

# Classical groups and Haar integration Teo Banica

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# **Classical groups and Haar integration**

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ABSTRACT. This is an introduction to the closed subgroups  $G \subset U_N$ , with all needed preliminaries included. We discuss the basic theory, and we perform some probability computations, in the finite case. In the general case, we explain how the Weingarten integration formula works, and we present some basic  $N \to \infty$  applications.

## Preface

Linear algebra is the source of many good things in this world. First of all, everything algebra, for sure. But also geometry and analysis, because any smooth function or manifold, taken locally, perturbes a certain linear transformation of  $\mathbb{R}^N$ . And finally probability too, remember indeed that Gauss integral needed for talking about normal laws, which can only be computed by using polar coordinates and their Jacobian.

The purpose of this book is to talk about linear algebra in a large sense, theory and applications, at a somewhat more advanced level than the beginner one, and by insisting on beautiful things. And with some graduate level mathematics, and quantum physics too, in mind. We will particularly insist on the groups of matrices, which are extremely useful for all sorts of mathematics and physics, and which are perhaps the most beautiful topic one could study, once the basics of linear algebra and matrices understood.

The book itself is accessible with 0 knowledge or almost, with page 1 of chapter 1 talking about vectors in  $\mathbb{R}^2$ . However, things will escalate quickly, and some prior knowledge of linear algebra, even in some rough form, is recommended. All in all, the book covers what can be taught during a 1-year upper division undergraduate course. Some other forms of applications are possible as well, and more on this later.

The first half of the book is concerned with linear algebra and its applications. Part I is a quick journey through basic linear algebra, from basic definitions and fun with  $2 \times 2$  matrices, up to the Spectral Theorem in its most general form, for the normal matrices  $A \in M_N(\mathbb{C})$ . Among the features of our presentation, the determinant will be introduced as it should, as a signed volume of a system of vectors. And also, we will discuss all sorts of useful matrix tricks, which are more advanced, and good to know.

As a continuation of this, Part II deals with various applications of linear algebra, to questions in analysis. After a quick look at differentiation and integration, which in several variables are intimately related to matrix theory, via the Jacobian, Hessian and so on, we will develop some useful probability theory, in relation with the normal and hyperspherical laws, by using spherical coordinates and their Jacobian. We will also discuss some other analytic topics, such as spectral theory, and special matrices.

#### PREFACE

The second half of the book is concerned with matrix groups. As already mentioned, this is perhaps the most beautiful topic one could study, once the basics of linear algebra understood. The subject is however huge, and Part III will be just a modest introduction to it. Our philosophy will be that of talking about all sorts of interesting closed subgroups  $G \subset U_N$ , finite and continuous alike, and by using very basic methods, coming from standard calculus, combinatorics and probability, for their study.

As a conclusion to this, the finite group case will appear to be reasonably understood, while the continuous case, not. Part IV will be dedicated to the study of the closed subgroups  $G \subset U_N$ , and more specifically the continuous ones, by using heavy machinery, as heavy as it gets. We will discuss here the basics of representation theory, then the existence of the Haar measure, and the Peter-Weyl theory, and then more advanced topics, such as Tannakian duality, Brauer theorems, and Weingarten calculus.

All in all, as already mentioned above, we will cover here what can be taught during a 1-year upper division undergraduate course. There are of course some other options too, as for instance using Parts I and III, with minimal input from Parts II and IV, for a 1-semester upper division undergraduate course, on linear algebra and group theory. And so on, with many other options being possible. Finally, it is our hope that mathematicians and physicists and engineers alike, teachers and students, will find this book useful.

At the level of things which are not done in this book, notable topics include the Jordan decomposition, which is the nightmare of everyone involved, teacher or student, and this remains between us, as well as some basic Lie algebra theory, which would have perfectly make sense to include, as a main topic for the last part, but that we preferred to replace by representation theory, and its relation with combinatorics and probability, which are somewhat more elementary, and fitting better with the rest of the book.

Finally, let us mention that this way of presenting things has its origins in some recent research work on the quantum groups, and more specifically on the so-called easy quantum groups. The idea there is that there is no much smoothness and geometry, with the main tools belonging to combinatorics and probability. Thus, as main philosophy, the present book, while dealing with classical topics, is written with a "quantum" touch.

Most of this book is based on lecture notes from various classes at Cergy, and I would like to thank my students. The final part goes into research topics, and I am grateful to Benoît Collins, Steve Curran and Roland Speicher, for our joint work on the subject. Many thanks go as well to my cats. There is so much to learn from them, too.

# Contents

Preface	3
Part I. Linear algebra	9
Chapter 1. Real matrices	11
1a. Linear maps	11
1b. Matrix calculus	19
1c. Diagonalization	24
1d. Scalar products	27
1e. Exercises	31
Chapter 2. The determinant	33
2a. Matrix inversion	33
2b. The determinant	35
2c. Basic properties	42
2d. Sarrus and beyond	47
2e. Exercises	55
Chapter 3. Complex matrices	57
3a. Complex numbers	57
3b. Linear maps	67
3c. Diagonalization	70
3d. The determinant	73
3e. Exercises	79
Chapter 4. Diagonalization	81
4a. Linear spaces	81
4b. Diagonalization	83
4c. Matrix tricks	87
4d. Spectral theorems	94
4e. Exercises	103

Part II. Basic analysis	105
Chapter 5. Standard calculus	107
5a. Derivatives	107
5b. Integration	117
5c. Normal laws	121
5d. Central limits	124
5e. Exercises	127
Chapter 6. Advanced calculus	129
6a. Volumes of spheres	129
6b. Spherical integrals	139
6c. Hyperspherical laws	145
6d. Complex spheres	146
6e. Exercises	151
Chapter 7. Spectral theory	153
7a. Hilbert spaces	153
7b. Linear operators	158
7c. Spectral theory	161
7d. Operator algebras	166
7e. Exercises	175
Chapter 8. Special matrices	177
8a. Fourier matrices	177
8b. Circulant matrices	183
8c. Bistochastic matrices	188
8d. Hadamard conjecture	193
8e. Exercises	199
Part III. Matrix groups	201
Chapter 9. Group theory	203
9a. Basic examples	203
9b. Dihedral groups	206
9c. Cayley embeddings	212
9d. Abelian groups	218
9e. Exercises	223

CONTENTS	7
Chapter 10. Rotation groups	225
10a. Rotation groups	225
10b. Euler-Rodrigues	228
10c. Symplectic groups	238
10d. Reflection groups	243
10e. Exercises	247
Chapter 11. Symmetric groups	249
11a. Character laws	249
11b. Poisson limits	255
11c. Truncated characters	261
11d. Further results	264
11e. Exercises	271
Chapter 12. Reflection groups	273
12a. Real reflections	273
12b. Bessel laws	275
12c. Complex reflections	277
12d. Wigner laws	286
12e. Exercises	295
Part IV. Haar integration	297
Chapter 13. Representations	299
13a. Basic theory	299
13b. Peter-Weyl theory	303
13c. Haar integration	309
13d. More Peter-Weyl	316
13e. Exercises	318
Chapter 14. Tannakian duality	321
14a. Tensor categories	321
14b. Abstract algebra	326
14c. The correspondence	330
14d. Brauer theorems	337
14e. Exercises	343
Chapter 15. Diagrams, easiness	345

### CONTENTS

15a. Easy groups	345
15b. Reflection groups	352
15c. Basic operations	355
15d. Classification results	363
15e. Exercises	367
Chapter 16. Weingarten calculus	369
16a. Weingarten formula	369
16b. Laws of characters	374
16c. Truncated characters	381
16d. Standard estimates	383
16e. Exercises	392
Bibliography	393
Index	397

# Part I

# Linear algebra

So close, no matter how far Couldn't be much more from the heart Forever trusting who we are And nothing else matters

### CHAPTER 1

## **Real matrices**

#### 1a. Linear maps

We are interested in what follows in symmetries, rotations, projections and other such basic transformations, in 2, 3 or even more dimensions. Such transformations appear a bit everywhere, in physics. To be more precise, each physical problem or equation has some "symmetries", and exploiting these symmetries is usually a very useful thing.

Let us start with 2 dimensions, and leave 3 and more dimensions for later. The maps that we are interested in are "affine", in the following sense:

DEFINITION 1.1. A map  $f : \mathbb{R}^2 \to \mathbb{R}^2$  is called

- (1) Affine, if it maps straight lines to straight lines,
- (2) Linear, if it maps lines passing through 0 to lines passing through 0,

with the convention that  $\{0\}$  itself is a "degenerate" line.

Here the last convention is there in order for some familiar maps, such as the projections, to be affine. Indeed, the projection on the horizontal axis sends the whole vertical axis to  $\{0\}$ , so if we want this projection to be affine,  $\{0\}$  must be a "line".

Here are some basic examples of symmetries, all being linear in the above sense:

**PROPOSITION 1.2.** The symmetries with respect to Ox and Oy are:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} x \\ -y \end{pmatrix} \quad , \quad \begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} -x \\ y \end{pmatrix}$$

The symmetries with respect to the x = y and x = -y diagonals are:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} y \\ x \end{pmatrix} \quad , \quad \begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} -y \\ -x \end{pmatrix}$$

All these maps are linear, in the above sense.

PROOF. The fact that all these maps are linear is clear, because they map lines passing through 0 to lines passing through 0. As for the explicit formulae in the statement, these are clear as well, by drawing pictures for each of the maps involved.  $\Box$ 

Here are now some basic examples of rotations, once again all being linear:

#### 1. REAL MATRICES

**PROPOSITION 1.3.** The rotations of angle  $0^{\circ}$  and of angle  $90^{\circ}$  are:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} x \\ y \end{pmatrix} \quad , \quad \begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} -y \\ x \end{pmatrix}$$

The rotations of angle  $180^{\circ}$  and of angle  $270^{\circ}$  are:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} -x \\ -y \end{pmatrix} \quad , \quad \begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} y \\ -x \end{pmatrix}$$

All these maps are linear, in the above sense.

PROOF. As before, these rotations are all linear, because they map lines passing through 0 to lines passing through 0. As for the explicit formulae in the statement, these are clear as well, by drawing pictures for each of the maps involved.  $\Box$ 

Here are as well some basic examples of projections, once again all being linear:

**PROPOSITION 1.4.** The projections on Ox and Oy are:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} x \\ 0 \end{pmatrix} \quad , \quad \begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} 0 \\ y \end{pmatrix}$$

The projections on the x = y and x = -y diagonals are:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \frac{1}{2} \begin{pmatrix} x+y \\ x+y \end{pmatrix} \quad , \quad \begin{pmatrix} x \\ y \end{pmatrix} \to \frac{1}{2} \begin{pmatrix} x-y \\ y-x \end{pmatrix}$$

All these maps are linear, in the above sense.

**PROOF.** As before, these projections are all linear. As for the formulae in the statement, these are clear as well, by drawing pictures for each of the maps involved.  $\Box$ 

Finally, we have as well translations, as follows:

**PROPOSITION 1.5.** The translations are exactly the maps of the form

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} x+p \\ y+q \end{pmatrix}$$

with  $p, q \in \mathbb{R}$ , and these maps are all affine, in the above sense.

PROOF. A translation  $f : \mathbb{R}^2 \to \mathbb{R}^2$  is clearly affine, because it maps lines to lines. Also, such a translation is uniquely determined by the following vector:

$$f\begin{pmatrix}0\\0\end{pmatrix} = \begin{pmatrix}p\\q\end{pmatrix}$$

To be more precise, f must be the map which takes a vector  $\binom{x}{y}$ , and adds this vector  $\binom{p}{a}$  to it. But this gives the formula in the statement.

#### 1A. LINEAR MAPS

Summarizing, we have many interesting examples of linear and affine maps. Let us develop now some general theory, for such maps. As a first result, the linear/affine maps  $f : \mathbb{R}^2 \to \mathbb{R}^2$  are fully described by 4/6 parameters, which appear as follows:

THEOREM 1.6. The linear maps  $f : \mathbb{R}^2 \to \mathbb{R}^2$  are precisely the maps of type

$$f\begin{pmatrix}x\\y\end{pmatrix} = \begin{pmatrix}ax+by\\cx+dy\end{pmatrix}$$

and the affine maps  $f : \mathbb{R}^2 \to \mathbb{R}^2$  are precisely the maps of type:

$$f\begin{pmatrix}x\\y\end{pmatrix} = \begin{pmatrix}ax+by\\cx+dy\end{pmatrix} + \begin{pmatrix}p\\q\end{pmatrix}$$

PROOF. It is routine to check, by drawing pictures, that the linear maps must map sums to sums, and scalar multiples to scalar multiples. Thus, a linear map  $f : \mathbb{R}^2 \to \mathbb{R}^2$ is uniquely determined by its values on the basic coordinate vectors, as follows:

$$f\begin{pmatrix}x\\y\end{pmatrix} = f\left(\begin{pmatrix}x\\0\end{pmatrix} + \begin{pmatrix}0\\y\end{pmatrix}\right)$$
$$= f\begin{pmatrix}x\\0\end{pmatrix} + f\begin{pmatrix}0\\y\end{pmatrix}$$
$$= f\left(x\begin{pmatrix}1\\0\end{pmatrix}\right) + f\left(y\begin{pmatrix}0\\1\end{pmatrix}\right)$$
$$= xf\begin{pmatrix}1\\0\end{pmatrix} + yf\begin{pmatrix}0\\1\end{pmatrix}$$

Thus, we obtain the formula in the statement, with  $a, b, c, d \in \mathbb{R}$  being given by:

$$f\begin{pmatrix}1\\0\end{pmatrix} = \begin{pmatrix}a\\c\end{pmatrix}$$
 ,  $f\begin{pmatrix}0\\1\end{pmatrix} = \begin{pmatrix}b\\d\end{pmatrix}$ 

In the affine case now, we have as extra piece of data a vector, as follows:

$$f\begin{pmatrix}0\\0\end{pmatrix} = \begin{pmatrix}p\\q\end{pmatrix}$$

Indeed, if  $f: \mathbb{R}^2 \to \mathbb{R}^2$  is affine, then the following map is linear:

$$f - \begin{pmatrix} p \\ q \end{pmatrix} : \mathbb{R}^2 \to \mathbb{R}^2$$

Thus, by using the formula in (1) we obtain the result.

In order to simplify now all this, which might be a bit complicated to memorize, the idea is to put our parameters a, b, c, d into a matrix, in the following way:

DEFINITION 1.7. A matrix  $A \in M_2(\mathbb{R})$  is an array as follows:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

These matrices act on the vectors in the following way,

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} ax + by \\ cx + dy \end{pmatrix}$$

the rule being "multiply the rows of the matrix by the vector".

The above multiplication formula might seem a bit complicated, at a first glance, but it is not. Here is an example for it, quickly worked out:

$$\begin{pmatrix} 1 & 2 \\ 5 & 6 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \cdot 3 + 2 \cdot 1 \\ 5 \cdot 3 + 6 \cdot 1 \end{pmatrix} = \begin{pmatrix} 5 \\ 21 \end{pmatrix}$$

As already mentioned, all this comes from our findings from Theorem 1.6. Indeed, with the above multiplication convention for matrices and vectors, we can turn Theorem 1.6 into something much simpler, and better-looking, as follows:

THEOREM 1.8. The linear maps  $f : \mathbb{R}^2 \to \mathbb{R}^2$  are precisely the maps of type

$$f(v) = Av$$

and the affine maps  $f: \mathbb{R}^2 \to \mathbb{R}^2$  are precisely the maps of type

$$f(v) = Av + w$$

with A being a  $2 \times 2$  matrix, and with  $v, w \in \mathbb{R}^2$  being vectors, written vertically.

**PROOF.** With the above conventions, the formulae in Theorem 1.6 read:

$$f\begin{pmatrix}x\\y\end{pmatrix} = \begin{pmatrix}a & b\\c & d\end{pmatrix}\begin{pmatrix}x\\y\end{pmatrix}$$
$$f\begin{pmatrix}x\\y\end{pmatrix} = \begin{pmatrix}a & b\\c & d\end{pmatrix}\begin{pmatrix}x\\y\end{pmatrix} + \begin{pmatrix}p\\q\end{pmatrix}$$

But these are exactly the formulae in the statement, with:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \quad , \quad v = \begin{pmatrix} x \\ y \end{pmatrix} \quad , \quad w = \begin{pmatrix} p \\ q \end{pmatrix}$$

Thus, we have proved our theorem.

Before going further, let us discuss some examples. First, we have:

14

**PROPOSITION 1.9.** The symmetries with respect to Ox and Oy are given by:

$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$
$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

The symmetries with respect to the x = y and x = -y diagonals are given by:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$
$$\begin{pmatrix} x \\ y \end{pmatrix} \to \begin{pmatrix} 0 & -1 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

PROOF. According to the formulae in Proposition 1.2, the transformations in the statement map an arbitrary vector  $\begin{pmatrix} x \\ y \end{pmatrix}$  to the following vectors:

$$\begin{pmatrix} x \\ -y \end{pmatrix} , \begin{pmatrix} -x \\ y \end{pmatrix} , \begin{pmatrix} y \\ x \end{pmatrix} , \begin{pmatrix} -y \\ -x \end{pmatrix}$$

But this gives the various formulae in the statement, by guessing in each case the matrix which does the job, in the obvious way.  $\hfill \Box$ 

Regarding now the basic rotations, we have here:

**PROPOSITION 1.10.** The rotations of angle 0° and of angle 90° are given by:

$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$
$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

The rotations of angle  $180^{\circ}$  and of angle  $270^{\circ}$  are given by:

$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$
$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

PROOF. As before, but according now to the formulae in Proposition 1.3, the transformations in the statement map an arbitrary vector  $\begin{pmatrix} x \\ y \end{pmatrix}$  to the following vectors:

$$\begin{pmatrix} x \\ y \end{pmatrix} , \begin{pmatrix} -y \\ x \end{pmatrix} , \begin{pmatrix} -x \\ -y \end{pmatrix} , \begin{pmatrix} y \\ -x \end{pmatrix}$$

But this gives the formulae in the statement, as before by guessing the matrix.  $\Box$ Finally, regarding the basic projections, we have here: **PROPOSITION 1.11.** The projections on Ox and Oy are given by:

$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$
$$\begin{pmatrix} x \\ y \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

The projections on the x = y and x = -y diagonals are given by:

$$\begin{pmatrix} x \\ y \end{pmatrix} \to \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$
$$\begin{pmatrix} x \\ y \end{pmatrix} \to \frac{1}{2} \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

**PROOF.** As before, but according now to Proposition 1.4, the transformations in the statement map an arbitrary vector  $\begin{pmatrix} x \\ y \end{pmatrix}$  to the following vectors:

$$\begin{pmatrix} x \\ 0 \end{pmatrix} , \begin{pmatrix} 0 \\ y \end{pmatrix} , \frac{1}{2} \begin{pmatrix} x+y \\ x+y \end{pmatrix} , \frac{1}{2} \begin{pmatrix} x-y \\ y-x \end{pmatrix}$$

But this gives the formulae in the statement, as before by guessing the matrix.  $\Box$ 

In addition to the above transformations, there are many other examples. We have for instance the null transformation, which is given by:

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Here is now a more bizarre map, which can still be understood, however, as being the map which "switches the coordinates, then kills the second one":

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ 0 \end{pmatrix}$$

Even more bizarrely now, here is a certain linear map, whose interpretation is more complicated, and is left to the reader:

$$\begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x+y \\ 0 \end{pmatrix}$$

And here is another linear map, which once again, being something geometric, in 2 dimensions, can definitely be understood, at least in theory:

$$\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x+y \\ y \end{pmatrix}$$

Summarizing, things can escalate quicky, but our linear map formalism is certainly something powerful, and everything can be a priori understood in terms of matrices.

Let us discuss now the computation of the arbitrary symmetries, rotations and projections. We begin with the rotations, whose formula is a must-know:

THEOREM 1.12. The rotation of angle  $t \in \mathbb{R}$  is given by the matrix

$$R_t = \begin{pmatrix} \cos t & -\sin t\\ \sin t & \cos t \end{pmatrix}$$

depending on  $t \in \mathbb{R}$  taken modulo  $2\pi$ .

**PROOF.** The rotation being linear, it must correspond to a certain matrix:

$$R_t = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

We can guess this matrix, via its action on the basic coordinate vectors  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$ . A quick picture shows that we must have:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} \cos t \\ \sin t \end{pmatrix}$$

Also, by paying attention to positives and negatives, we must have:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -\sin t \\ \cos t \end{pmatrix}$$

Guessing now the matrix is not complicated, because the first equation gives us the first column, and the second equation gives us the second column:

$$\begin{pmatrix} a \\ c \end{pmatrix} = \begin{pmatrix} \cos t \\ \sin t \end{pmatrix} \quad , \quad \begin{pmatrix} b \\ d \end{pmatrix} = \begin{pmatrix} -\sin t \\ \cos t \end{pmatrix}$$

Thus, we can just put together these two vectors, and we obtain our matrix.

Regarding now the symmetries, the formula here is as follows:

THEOREM 1.13. The symmetry with respect to the Ox axis rotated by an angle  $t/2 \in \mathbb{R}$  is given by the matrix

$$S_t = \begin{pmatrix} \cos t & \sin t \\ \sin t & -\cos t \end{pmatrix}$$

depending on  $t \in \mathbb{R}$  taken modulo  $2\pi$ .

**PROOF.** The symmetry being linear, it must correspond to a certain matrix:

$$S_t = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

As before, we can guess this matrix via its action on the basic coordinate vectors  $\begin{pmatrix} 1\\0 \end{pmatrix}$  and  $\begin{pmatrix} 0\\1 \end{pmatrix}$ . A quick picture shows that we must have:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} \cos t \\ \sin t \end{pmatrix}$$

#### 1. REAL MATRICES

Also, by paying attention to positives and negatives, we must have:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} \sin t \\ -\cos t \end{pmatrix}$$

Guessing now the matrix is not complicated, because we must have:

$$\begin{pmatrix} a \\ c \end{pmatrix} = \begin{pmatrix} \cos t \\ \sin t \end{pmatrix} \quad , \quad \begin{pmatrix} b \\ d \end{pmatrix} = \begin{pmatrix} \sin t \\ -\cos t \end{pmatrix}$$

Thus, we can just put together these two vectors, and we obtain our matrix.

Finally, regarding the projections, the formula here is as follows:

THEOREM 1.14. The projection on the Ox axis rotated by an angle  $t/2 \in \mathbb{R}$  is given by the matrix

$$P_t = \frac{1}{2} \begin{pmatrix} 1 + \cos t & \sin t \\ \sin t & 1 - \cos t \end{pmatrix}$$

depending on  $t \in \mathbb{R}$  taken modulo  $2\pi$ .

**PROOF.** We will need here some trigonometry, and more precisely the formulae for the duplication of the angles. Regarding the sine, the formula here is:

$$\sin(2t) = 2\sin t \cos t$$

Regarding the cosine, we have here 3 equivalent formulae, with the equivalence coming from  $\cos^2 + \sin^2 = 1$ , which are all useful, and worth memorizing, as follows:

$$cos(2t) = cos2 t - sin2 t$$
$$= 2 cos2 t - 1$$
$$= 1 - 2 sin2 t$$

Getting back now to our problem, a quick picture, using similarity of triangles, and then the above trigonometry formulae, show that we must have:

$$P_t \begin{pmatrix} 1\\0 \end{pmatrix} = \cos \frac{t}{2} \begin{pmatrix} \cos \frac{t}{2}\\\sin \frac{t}{2} \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1+\cos t\\\sin t \end{pmatrix}$$

Similarly, another quick picture plus trigonometry show that we must have:

$$P_t \begin{pmatrix} 0\\1 \end{pmatrix} = \sin \frac{t}{2} \begin{pmatrix} \cos \frac{t}{2}\\\sin \frac{t}{2} \end{pmatrix} = \frac{1}{2} \begin{pmatrix} \sin t\\1 - \cos t \end{pmatrix}$$

Now by putting together these two vectors, and we obtain our matrix.

We will be back to rotations, symmetries and projections later on, with a number of general formulae for them, generalizing the above ones, in several dimensions.

#### **1B. MATRIX CALCULUS**

#### 1b. Matrix calculus

In order to formulate now our second theorem, dealing with compositions of maps, let us make the following multiplication convention, between matrices and matrices:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} p & q \\ r & s \end{pmatrix} = \begin{pmatrix} ap+br & aq+bs \\ cp+dr & cq+ds \end{pmatrix}$$

This might look a bit complicated, but as before, in what was concerning the multiplication of matrices and vectors, the idea is very simple, namely "multiply the rows of the first matrix by the columns of the second matrix". With this convention, we have:

THEOREM 1.15. If we denote by  $f_A : \mathbb{R}^2 \to \mathbb{R}^2$  the linear map associated to a matrix A, given by the formula

$$f_A(v) = Av$$

then we have the following multiplication formula for such maps:

$$f_A f_B = f_{AB}$$

In other words, the composition of linear maps corresponds to the multiplication of the corresponding matrices.

**PROOF.** We want to prove that we have the following formula, valid for any two matrices  $A, B \in M_2(\mathbb{R})$ , and any vector  $v \in \mathbb{R}^2$ :

$$A(Bv) = (AB)\iota$$

For this purpose, let us write our matrices and vector as follows:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \quad , \quad B = \begin{pmatrix} p & q \\ r & s \end{pmatrix} \quad , \quad v = \begin{pmatrix} x \\ y \end{pmatrix}$$

The formula that we want to prove becomes:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{bmatrix} \begin{pmatrix} p & q \\ r & s \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{bmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} p & q \\ r & s \end{pmatrix} \begin{bmatrix} x \\ y \end{pmatrix}$$

But this is the same as saying that:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} px + qy \\ rx + sy \end{pmatrix} = \begin{pmatrix} ap + br & aq + bs \\ cp + dr & cq + ds \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

And this latter formula does hold indeed, because on both sides we get:

$$\begin{pmatrix} apx + aqy + brx + bsy \\ cpx + cqy + drx + dsy \end{pmatrix}$$

Thus, we have proved the result.

#### 1. REAL MATRICES

As a verification for the above result, let us compose two rotations. The computation here is as follows, yieding a rotation, as it should, and of the correct angle:

$$R_{s}R_{t} = \begin{pmatrix} \cos s & -\sin s \\ \sin s & \cos s \end{pmatrix} \begin{pmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{pmatrix}$$
$$= \begin{pmatrix} \cos s \cos t - \sin s \sin t & -\cos s \sin t - \sin t \cos s \\ \sin s \cos t + \cos s \sin t & -\sin s \sin t + \cos s \cos t \end{pmatrix}$$
$$= \begin{pmatrix} \cos(s+t) & -\sin(s+t) \\ \sin(s+t) & \cos(s+t) \end{pmatrix}$$
$$= R_{s+t}$$

We are ready now to pass to 3 dimensions. The idea is to select what we learned in 2 dimensions, nice results only, and generalize to 3 dimensions. We obtain:

THEOREM 1.16. Consider a map  $f : \mathbb{R}^3 \to \mathbb{R}^3$ .

(1) f is linear, sending lines through 0 to lines through 0, when it is of the form

f(v) = Av

with A being a certain  $3 \times 3$  matrix.

(2) f is affine, sending lines to lines, precisely when it is of the form

$$f(v) = Av + w$$

with A being a certain  $3 \times 3$  matrix, and  $w \in \mathbb{R}^3$  being a vector.

(3) In addition, we have a composition formula of type

$$f_A f_B = f_{AB}$$

similar to the 2D one.

**PROOF.** Here (1) and (2) can be proved exactly as in the 2D case, with the multiplication convention being as usual, "multiply the rows of the matrix by the vector":

$$\begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} ax + by + cz \\ dx + ey + fz \\ gx + hy + iz \end{pmatrix}$$

As for (3), once again the 2D idea applies, with the same product rule, "multiply the rows of the first matrix by the columns of the second matrix":

$$\begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} \begin{pmatrix} p & q & r \\ s & t & u \\ v & w & x \end{pmatrix} = \begin{pmatrix} ap+bs+cv & aq+bt+cw & ar+bu+cx \\ dp+es+fv & dq+et+fw & dr+eu+fx \\ gp+hs+iv & gq+ht+iw & gr+hu+ix \end{pmatrix}$$

Thus, we have proved our theorem. Of course, we are going a bit fast here, and some verifications are missing, but we will discuss all this in detail, in N dimensions.

#### 1B. MATRIX CALCULUS

We are now ready to discuss 4 and more dimensions. Before doing so, let us point out however that the maps of type  $f : \mathbb{R}^3 \to \mathbb{R}^2$ , or  $f : \mathbb{R} \to \mathbb{R}^2$ , and so on, are not covered by our results. Since there are many interesting such maps, say obtained by projecting and then rotating, and so on, we will be interested here in the maps  $f : \mathbb{R}^N \to \mathbb{R}^M$ .

A bit of thinking suggests that such maps should come from the  $M \times N$  matrices. Indeed, this is what happens at M = N = 2 and M = N = 3, of course. But this happens as well at M = 1, because a linear map  $f : \mathbb{R} \to \mathbb{R}^N$  can only be something of the form  $f(\lambda) = \lambda v$ , with  $v \in \mathbb{R}^N$ . But  $v \in \mathbb{R}^N$  means that v is a  $N \times 1$  matrix.

So, let us start with the product rule for such matrices, which is as follows:

DEFINITION 1.17. We can multiply the  $M \times N$  matrices with  $N \times K$  matrices,

$$\begin{pmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{M1} & \dots & a_{MN} \end{pmatrix} \begin{pmatrix} b_{11} & \dots & b_{1K} \\ \vdots & & \vdots \\ b_{N1} & \dots & b_{NK} \end{pmatrix}$$

the product being given by the following formula,

$$\begin{pmatrix} a_{11}b_{11} + \ldots + a_{1N}b_{N1} & \ldots & a_{11}b_{1K} + \ldots + a_{1N}b_{NK} \\ \vdots & & \vdots \\ a_{M1}b_{11} + \ldots + a_{MN}b_{N1} & \ldots & a_{M1}b_{1K} + \ldots + a_{MN}b_{NK} \end{pmatrix}$$

obtained via the usual rule "multiply rows by columns".

Observe that this formula generalizes all the multiplication rules that we have been using so far, between various types of matrices and vectors. Thus, in practice, we can simply forget all the previous multiplication rules, and simply memorize this one.

In case the above formula looks hard to memorize, here is an alternative formulation of it, which is simpler and more powerful, by using the standard algebraic notation for the matrices,  $A = (A_{ij})$ , along with some consequences coming from it:

**PROPOSITION 1.18.** The matrix multiplication is given by the following formula:

$$(AB)_{ij} = \sum_{k} A_{ik} B_{kj}$$

In addition, we have the following general formula, valid for any matrices A, B, C,

$$(AB)C = A(BC)$$

provided that the sizes of our matrices A, B, C fit.

PROOF. The first formula is just a shorthand for the formula in Definition 1.17, by following the rule there, namely "multiply the rows of A by the columns of B". As for the second formula, this follows from it. Indeed, we have:

$$((AB)C)_{ij} = \sum_{k} (AB)_{ik} C_{kj}$$
$$= \sum_{kl} A_{il} B_{lk} C_{kj}$$

On the other hand, we have as well:

$$A(BC))_{ij} = \sum_{l} A_{il}(BC)_{lj}$$
$$= \sum_{kl} A_{il} B_{lk} C_{kj}$$

Thus we have (AB)C = A(BC), and we have proved our result.

We can now talk about linear maps between spaces of arbitrary dimension, generalizing what we have been doing so far. The main result here is as follows:

THEOREM 1.19. Consider a map  $f : \mathbb{R}^N \to \mathbb{R}^M$ . (1) f is linear, sending lines through 0 to lines through 0, when it is of the form

(

f(v) = Av

with A being a certain  $M \times N$  matrix.

(2) f is affine, sending lines to lines, precisely when it is of the form

f(v) = Av + w

with A being a certain  $M \times N$  matrix, and  $w \in \mathbb{R}^M$  being a vector.

(3) In addition, we have  $f_A f_B = f_{AB}$ , whenever the sizes fit.

**PROOF.** We already know that this happens at M = N = 2, and at M = N = 3 as well. In general, the proof is similar, by doing some elementary computations.

As a first example here, we have the identity matrix, acting as the identity:

$$\begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix}$$

We have as well the null matrix, acting as the null map:

$$\begin{pmatrix} 0 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$$

22

Here is now an important result, providing us with many examples:

**PROPOSITION 1.20.** The diagonal matrices act as follows:

$$\begin{pmatrix} \lambda_1 & 0 \\ & \ddots & \\ 0 & & \lambda_N \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} \lambda_1 x_1 \\ \vdots \\ \lambda_N x_N \end{pmatrix}$$

**PROOF.** This is clear, indeed, from definitions.

As a more specialized example now, we have:

**PROPOSITION 1.21.** The flat matrix, which is as follows,

$$\mathbb{I}_N = \begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix}$$

acts via N times the projection on the all-one vector.

**PROOF.** The flat matrix acts in the following way:

$$\begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} x_1 + \dots + x_N \\ \vdots \\ x_1 + \dots + x_N \end{pmatrix}$$

Thus, in terms of the matrix  $P = \mathbb{I}_N / N$ , we have the following formula:

$$P\begin{pmatrix}x_1\\\vdots\\x_N\end{pmatrix} = \frac{x_1 + \ldots + x_N}{N} \begin{pmatrix}1\\\vdots\\1\end{pmatrix}$$

Since the linear map f(x) = Px satisfies  $f^2 = f$ , and since Im(f) consists of the scalar multiples of the all-one vector  $\xi \in \mathbb{R}^N$ , we conclude that f is a projection on  $\mathbb{R}\xi$ . Also, with the standard scalar product convention  $\langle x, y \rangle = \sum x_i y_i$ , we have:

$$\langle f(x) - x, \xi \rangle = \langle f(x), \xi \rangle - \langle x, \xi \rangle$$
  
=  $\frac{\sum x_i}{N} \times N - \sum x_i$   
= 0

Thus, our projection is an orthogonal projection, and we are done.

We will be back to the flat matrix later on, when systematically investigating scalar products, orthogonality, and orthogonal projections.

#### 1c. Diagonalization

Let us develop now some general theory for the square matrices. We will need the following standard result, regarding the changes of coordinates on  $\mathbb{R}^N$ :

THEOREM 1.22. For a system of vectors  $\{v_1, \ldots, v_N\} \subset \mathbb{R}^N$ , the following conditions are equivalent:

(1) The vectors  $v_i$  form a basis of  $\mathbb{R}^N$ , in the sense that each vector  $x \in \mathbb{R}^N$  can be written in a unique way as a linear combination of these vectors:

$$x = \sum \lambda_i v_i$$

(2) The following linear map associated to these vectors is bijective:

$$f: \mathbb{R}^N \to \mathbb{R}^N$$
$$\lambda \to \sum \lambda_i v_i$$

(3) The matrix formed by these vectors, regarded as usual as column vectors,

$$P = [v_1, \dots, v_N] \in M_N(\mathbb{R})$$

is invertible, with respect to the usual multiplication of the matrices.

PROOF. Here the equivalence (1)  $\iff$  (2) is clear from definitions, and the equivalence (2)  $\iff$  (3) is clear as well, because we have f(x) = Px.

Getting back now to the matrices, as an important definition, we have:

DEFINITION 1.23. Let  $A \in M_N(\mathbb{R})$  be a square matrix. We say that  $v \in \mathbb{R}^N$  is an eigenvector of A, with corresponding eigenvalue  $\lambda \in \mathbb{R}^N$ , when:

$$Av = \lambda v$$

Also, we say that A is diagonalizable when  $\mathbb{R}^N$  has a basis formed by eigenvectors of A.

We will see in a moment examples of eigenvectors and eigenvalues, and of diagonalizable matrices. However, even before seeing the examples, it is quite clear that these are key notions. Indeed, for a matrix  $A \in M_N(\mathbb{R})$ , being diagonalizable is the best thing that can happen, because in this case, once the basis changed, A becomes diagonal.

To be more precise here, we have the following result:

**PROPOSITION 1.24.** Assuming that  $A \in M_N(\mathbb{R})$  is diagonalizable, we have the formula

$$A = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_N \end{pmatrix}$$

with respect to the basis  $\{v_1, \ldots, v_N\}$  of  $\mathbb{R}^N$  consisting of eigenvectors of A.

#### 1C. DIAGONALIZATION

**PROOF.** This is clear from the definition of eigenvalues and eigenvectors, and from the formula of linear maps associated to diagonal matrices, from Proposition 1.20. 

Here is an equivalent form of the above result, which is often used in practice, when we prefer not to change the basis, and stay with the usual basis of  $\mathbb{R}^N$ :

THEOREM 1.25. Assuming that  $A \in M_N(\mathbb{R})$  is diagonalizable, with

$$v_1, \ldots, v_N \in \mathbb{R}^N$$
  
 $\lambda_1, \ldots, \lambda_N \in \mathbb{R}$ 

$$\lambda_1,\ldots,\lambda_N\in\mathbb{F}$$

as eigenvectors and corresponding eigenvalues, we have the formula

$$A = PDP^{-1}$$

with the matrices  $P, D \in M_N(\mathbb{R})$  being given by the formulae

$$P = [v_1, \ldots, v_N]$$

$$D = diag(\lambda_1, \dots, \lambda_N)$$

and respectively called passage matrix, and diagonal form of A.

**PROOF.** This can be viewed in two possible ways, as follows:

(1) As already mentioned, with respect to the basis  $v_1, \ldots, v_N \in \mathbb{R}^N$  formed by the eigenvectors, our matrix A is given by:

$$A = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_N \end{pmatrix}$$

But this corresponds precisely to the formula  $A = PDP^{-1}$  from the statement, with P and its inverse appearing there due to our change of basis.

(2) We can equally establish the formula in the statement by a direct computation. Indeed, we have  $Pe_i = v_i$ , where  $\{e_1, \ldots, e_N\}$  is the standard basis of  $\mathbb{R}^N$ , and so:

$$\begin{aligned} APe_i &= Av_i \\ &= \lambda_i v_i \end{aligned}$$

On the other hand, once again by using  $Pe_i = v_i$ , we have as well:

$$PDe_i = P\lambda_i e_i$$
$$= \lambda_i Pe_i$$
$$= \lambda_i v_i$$

Thus we have AP = PD, and so  $A = PDP^{-1}$ , as claimed.

Let us discuss now some basic examples, namely the rotations, symmetries and projections in 2 dimensions. The situation is very simple for the projections, as follows:

#### 1. REAL MATRICES

**PROPOSITION 1.26.** The projection on the Ox axis rotated by an angle  $t/2 \in \mathbb{R}$ ,

$$P_t = \frac{1}{2} \begin{pmatrix} 1 + \cos t & \sin t \\ \sin t & 1 - \cos t \end{pmatrix}$$

is diagonalizable, its diagonal form being as follows:

$$P_t \sim \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

**PROOF.** This is clear, because if we denote by L the line where our projection projects, we can pick any vector  $v \in L$ , and this will be an eigenvector with eigenvalue 1, and then pick any vector  $w \in L^{\perp}$ , and this will be an eigenvector with eigenvalue 0. Thus, even without computations, we are led to the conclusion in the statement.

The computation for the symmetries is similar, as follows:

**PROPOSITION 1.27.** The symmetry with respect to the Ox axis rotated by  $t/2 \in \mathbb{R}$ ,

$$S_t = \begin{pmatrix} \cos t & \sin t \\ \sin t & -\cos t \end{pmatrix}$$

is diagonalizable, its diagonal form being as follows:

$$S_t \sim \begin{pmatrix} 1 & 0\\ 0 & -1 \end{pmatrix}$$

**PROOF.** This is once again clear, because if we denote by L the line with respect to which our symmetry symmetrizes, we can pick any vector  $v \in L$ , and this will be an eigenvector with eigenvalue 1, and then pick any vector  $w \in L^{\perp}$ , and this will be an eigenvector with eigenvalue -1. Thus, we are led to the conclusion in the statement.  $\Box$ 

Regarding now the rotations, here the situation is different, as follows:

**PROPOSITION 1.28.** The rotation of angle  $t \in [0, 2\pi)$ , given by the formula

$$R_t = \begin{pmatrix} \cos t & -\sin t\\ \sin t & \cos t \end{pmatrix}$$

is diagonal at  $t = 0, \pi$ , and is not diagonalizable at  $t \neq 0, \pi$ .

**PROOF.** The first assertion is clear, because at  $t = 0, \pi$  the rotations are:

$$R_0 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \quad , \quad R_\pi = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}$$

As for the rotations of angle  $t \neq 0, \pi$ , these clearly cannot have eigenvectors.

Finally, here is one more example, or rather counterexample, which is the most important of them all, and definitely worth memorizing:

THEOREM 1.29. The following matrix is not diagonalizable,

$$J = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$

because it has only 1 eigenvector.

**PROOF.** The matrix in the statement acts as follows:

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ 0 \end{pmatrix}$$

Thus the eigenvector/eigenvalue equation  $Jv = \lambda v$  reads:

$$\begin{pmatrix} y \\ 0 \end{pmatrix} = \begin{pmatrix} \lambda x \\ \lambda y \end{pmatrix}$$

We have two cases here, depending on  $\lambda$ , as follows:

(1) For  $\lambda \neq 0$  we must have y = 0, coming from the second row, and so x = 0 as well, coming from the first row, so we have no nontrivial eigenvectors.

(2) As for the case  $\lambda = 0$ , here we must have y = 0, coming from the first row, and so the eigenvectors here are the vectors of the form  $\binom{x}{0}$ .

Thus, our matrix is not diagonalizable, as stated.

#### 1d. Scalar products

In order to discuss some interesting examples of matrices, and their diagonalization, in arbitrary dimensions, we will need the following standard fact:

**PROPOSITION 1.30.** Consider the scalar product on  $\mathbb{R}^N$ , given by:

$$\langle x, y \rangle = \sum_{i} x_{i} y_{i}$$

We have then the following formula, valid for any vectors x, y and any matrix A,

$$\langle Ax, y \rangle = \langle x, A^t y \rangle$$

with  $A^t$  being the transpose matrix.

**PROOF.** By linearity, it is enough to prove the above formula on the standard basis vectors  $e_1, \ldots, e_N$  of  $\mathbb{R}^N$ . Thus, we want to prove that for any i, j we have:

 $\langle Ae_j, e_i \rangle = \langle e_j, A^t e_i \rangle$ 

The scalar product being symmetric, this is the same as proving that:

$$< Ae_j, e_i > = < A^t e_i, e_j >$$

On the other hand, for any matrix M we have the following formula:

$$M_{ij} = \langle Me_j, e_i \rangle$$

Thus, the formula to be proved simply reads:

$$A_{ij} = (A^t)_{ji}$$

But this precisely the definition of  $A^t$ , and we are done.

With this, we can develop some theory. We first have:

**THEOREM** 1.31. The orthogonal projections are the matrices satisfying:

$$P^2 = P = P^t$$

These projections are diagonalizable, with eigenvalues 0, 1.

**PROOF.** It is obvious that a linear map f(x) = Px is a projection precisely when:

$$P^2 = P$$

In order now for this projection to be an orthogonal projection, the condition to be satisfied can be written and then processed as follows:

$$\langle Px - Py, Px - x \rangle = 0 \iff \langle x - y, P^t Px - P^t x \rangle = 0$$
  
 $\iff P^t Px - P^t x = 0$   
 $\iff P^t P - P^t = 0$ 

Thus we must have  $P^t = P^t P$ . Now observe that by transposing, we have as well:

$$P = (P^t P)^t$$
$$= P^t (P^t)^t$$
$$= P^t P$$

Thus we must have  $P = P^t$ , as claimed. Finally, regarding the diagonalization assertion, this is clear by taking a basis of Im(f), which consists of 1-eigenvectors, and then completing with 0-eigenvectors, which can be found inside the orthogonal of Im(f).  $\Box$ 

Here is now a key computation of such projections:

THEOREM 1.32. The rank 1 projections are given by the formula

$$P_x = \frac{1}{||x||^2} (x_i x_j)_{ij}$$

where the constant, namely

$$||x|| = \sqrt{\sum_i x_i^2}$$

is the length of the vector.

28

**PROOF.** Consider a vector  $y \in \mathbb{R}^N$ . Its projection on  $\mathbb{R}x$  must be a certain multiple of x, and we are led in this way to the following formula:

$$P_x y = \frac{\langle y, x \rangle}{\langle x, x \rangle} x$$
$$= \frac{1}{||x||^2} \langle y, x \rangle x$$

With this in hand, we can now compute the entries of  $P_x$ , as follows:

$$(P_x)_{ij} = \langle P_x e_j, e_i \rangle \\ = \frac{1}{||x||^2} \langle e_j, x \rangle \langle x, e_i \rangle \\ = \frac{x_j x_i}{||x||^2}$$

Thus, we are led to the formula in the statement.

As an application, we can recover a result that we already know, namely:

PROPOSITION 1.33. In 2 dimensions, the rank 1 projections, which are the projections on the Ox axis rotated by an angle  $t/2 \in [0, \pi)$ , are given by the following formula:

$$P_t = \frac{1}{2} \begin{pmatrix} 1 + \cos t & \sin t \\ \sin t & 1 - \cos t \end{pmatrix}$$

Together with the following two matrices, which are the rank 0 and 2 projections in  $\mathbb{R}^2$ ,

$$0 = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \quad , \quad 1 = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$$

these are all the projections in 2 dimensions.

**PROOF.** The first assertion follows from the general formula in Theorem 1.32, by plugging in the following vector, depending on a parameter  $s \in [0, \pi)$ :

$$x = \begin{pmatrix} \cos s \\ \sin s \end{pmatrix}$$

We obtain in this way the following matrix, which with t = 2s is the one in the statement, via some trigonometry:

$$P_{2s} = \begin{pmatrix} \cos^2 s & \cos s \sin s \\ \cos s \sin s & \sin^2 s \end{pmatrix}$$

As for the second assertion, this is clear from the first one, because outside rank 1 we can only have rank 0 or rank 2, corresponding to the matrices in the statement.  $\Box$ 

Here is another interesting application, this time in N dimensions:

**PROPOSITION 1.34.** The projection on the all-1 vector  $\xi \in \mathbb{R}^N$  is

$$P_{\xi} = \frac{1}{N} \begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix}$$

with the all-1 matrix on the right being called the flat matrix.

**PROOF.** As already pointed out in the proof of Proposition 1.21 above, the matrix in the statement acts in the following way:

$$P_{\xi}\begin{pmatrix}x_1\\\vdots\\x_N\end{pmatrix} = \frac{x_1 + \ldots + x_N}{N}\begin{pmatrix}1\\\vdots\\1\end{pmatrix}$$

Thus  $P_{\xi}$  is indeed a projection onto  $\mathbb{R}\xi$ , and the fact that this projection is indeed the orthogonal one follows either by a direct orthogonality computation, or by using the general formula in Theorem 1.32 above, by plugging in the all-1 vector  $\xi$ .

Let us discuss now, as a final topic of this chapter, the isometries of  $\mathbb{R}^N$ . We have here the following general result:

THEOREM 1.35. The linear maps  $f : \mathbb{R}^N \to \mathbb{R}^N$  which are isometries, in the sense that they preserve the distances, are those coming from the matrices satisfying:

 $U^t = U^{-1}$ 

These latter matrices are called orthogonal, and they form a set  $O_N \subset M_N(\mathbb{R})$  which is stable under taking compositions, and inverses.

**PROOF.** We have several things to be proved, the idea being as follows:

(1) We recall that we can pass from scalar products to distances, as follows:

$$||x|| = \sqrt{\langle x, x \rangle}$$

Conversely, we can compute the scalar products in terms of distances, by using the parallelogram identity, which is as follows:

$$\begin{split} ||x+y||^2 - ||x-y||^2 \\ &= ||x||^2 + ||y||^2 + 2 < x, y > -||x||^2 - ||y||^2 + 2 < x, y > \\ &= 4 < x, y > \end{split}$$

#### 1E. EXERCISES

Now given a matrix  $U \in M_N(\mathbb{R})$ , we have the following equivalences, with the first one coming from the above identities, and with the other ones being clear:

$$\begin{split} ||Ux|| &= ||x|| &\iff < Ux, Uy > = < x, y > \\ &\iff < x, U^t Uy > = < x, y > \\ &\iff U^t Uy = y \\ &\iff U^t U = 1 \\ &\iff U^t = U^{-1} \end{split}$$

(2) The second assertion is clear from the definition of the isometries, and can be established as well by using matrices, and the  $U^t = U^{-1}$  criterion.

As a basic illustration here, we have:

THEOREM 1.36. The rotations in the plane, given by the formula

$$R_t = \begin{pmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{pmatrix}$$

are isometries. The symmetries in the plane, given by formula

$$S_t = \begin{pmatrix} \cos t & \sin t \\ \sin t & -\cos t \end{pmatrix}$$

are isometries as well. These are all the isometries in 2 dimensions.

PROOF. We already know that  $R_t$  is the rotation of angle t. As for  $S_t$ , this is the symmetry with respect to the Ox axis rotated by  $t/2 \in \mathbb{R}$ . But this gives the result, since the isometries in 2 dimensions are either rotations, or symmetries.

As a conclusion, the set  $O_N$  from Theorem 1.35 is a quite fundamental object, with  $O_2$  already consisting of some interesting  $2 \times 2$  matrices, namely the matrices  $R_t, S_t$ . We will be back to  $O_N$ , which is a so-called group, and is actually one of the most important examples of groups, on several occasions, in what follows.

#### 1e. Exercises

The key thing in linear algebra is that of geometrically understanding the linear maps  $x \to Ax$  associated to the matrices  $A \in M_N(\mathbb{R})$ . Here is an exercise on this:

EXERCISE 1.37. Work out the geometric interpretation of the map f(x) = Ax, with

$$A \in M_2(\pm 1)$$

and then discuss as well the diagonalization of these matrices.

#### 1. REAL MATRICES

To be more precise, there are  $2^4 = 16$  matrices here, some of which were already discussed in the above. As a bonus exercise, you can try as well  $A \in M_2(0,1)$ , which is 16 more matrices. And for the black belt, try  $A \in M_2(-1,0,1)$ .

Here is now an interesting diagonalization exercise:

EXERCISE 1.38. Diagonalize explicitly the third flat matrix, namely

$$\mathbb{I}_3 = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

and then study as well the general case, that of the matrix  $\mathbb{I}_N$ .

Here we already know from the above that the diagonal form is D = (N, 0, ..., 0), and the problem is that of finding the passage matrix P, as to write the diagonalization formula  $\mathbb{I}_N = PDP^{-1}$ . The case to start with, as a warm-up for the exercise, is N = 2, where  $\mathbb{I}_2$  is twice the orthogonal projection on the x = y diagonal, which was already discussed in the above. Then, go with N = 3, and then with general  $N \in \mathbb{N}$ .

Here is now something that we used in the above, worth remembering in detail:

EXERCISE 1.39. Work out the trigonometry formulae

$$\sin(2t) = 2\sin t\cos t$$
$$\cos(2t) = 2\cos^2 t - 1$$

by using elementary methods, coming from plane geometry.

There are many ways of solving this exercise, and of course enjoy.

Here is another thing that we used in the above, worth clarifying:

EXERCISE 1.40. Prove that the isometries in 2 dimensions are either rotations, or symmetries, as to complete the proof of Theorem 1.36.

As before, there are many ways of dealing with this, all being nice geometry.

Finally, here is an exercise which is more research-style:

EXERCISE 1.41. Develop a theory of angles between the vectors  $x, y \in \mathbb{R}^N$ .

No indication here, the more complete the theory that you develop, the better.

#### CHAPTER 2

### The determinant

#### 2a. Matrix inversion

We have seen in the previous chapter that most of the interesting maps  $f : \mathbb{R}^N \to \mathbb{R}^N$ that we know, such as the rotations, symmetries and projections, are linear, and can be written in the following form, with  $A \in M_N(\mathbb{R})$  being a square matrix:

$$f(v) = Av$$

In this chapter we develop more general theory for the linear maps  $f : \mathbb{R}^N \to \mathbb{R}^N$ , written in the above form. To be more precise, we will investigate the following question, which is of particular importance: which linear maps f(v) = Av are invertible?

This is a quite tricky problem, so let us first recall what we know so far about it, from chapter 1 above. First, at the level of the basic examples we have:

**PROPOSITION** 2.1. The basic linear maps are as follows:

- (1) The rotation  $R_t$  is invertible, its inverse being  $R_{-t}$ .
- (2) Any symmetry S is invertible, with  $S^{-1} = S$ .
- (3) The projections  $P \neq 1$  are not invertible.

**PROOF.** All this is clear from definitions, as follows:

(1) This is obvious, because rotating by an angle  $t \in \mathbb{R}$ , and then rotating backwards by the same angle, amounts in doing nothing.

(2) Once again this is obvious, because a symmetry S is by definition a linear transformation satisfying  $S^2 = id$ , and so  $S^{-1} = S$ .

(3) This is clear too, because when our projection P is non-degenerate, in the sense that  $P \neq 1$ , we are certainly losing information when applying it.  $\Box$ 

Observe that the above results do not use matrix theory of any kind. In general, however, we have to use matrices, in order to get to results, and we have here:

THEOREM 2.2. A linear map  $f : \mathbb{R}^N \to \mathbb{R}^N$ , written as

$$f(v) = Av$$

is invertible when A is invertible, and in this case we have  $f^{-1}(v) = A^{-1}v$ .

#### 2. THE DETERMINANT

PROOF. This is something that we basically know, coming from the fact that, with the notation  $f_A(v) = Av$ , we have the following formula:

$$f_A f_B = f_{AB}$$

Thus, we are led to the conclusion in the statement.

In view of the above result, we are led to the question of inverting the square matrices  $A \in M_N(\mathbb{R})$ . In the simplest case, in 2 dimensions, the result is as follows:

THEOREM 2.3. We have the following inversion formula, for the  $2 \times 2$  matrices:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

When ad - bc = 0, the matrix is not invertible.

**PROOF.** We have two assertions to be proved, the idea being as follows:

(1) As a first observation, when ad - bc = 0 we must have, for some  $\lambda \in \mathbb{R}$ :

 $b = \lambda a$  ,  $d = \lambda c$ 

Thus our matrix must be of the following special type:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} a & \lambda a \\ a & \lambda c \end{pmatrix}$$

Thus in this case the columns are proportional, and so the linear map associated to the matrix is not invertible, and so the matrix itself is not invertible either.

(2) When  $ad - bc \neq 0$ , let us look for an inversion formula of the following type:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} * & * \\ * & * \end{pmatrix}$$

We must therefore solve the following equations:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} * & * \\ * & * \end{pmatrix} = \begin{pmatrix} ad - bc & 0 \\ 0 & ad - bc \end{pmatrix}$$

The obvious solution here is as follows:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix} = \begin{pmatrix} ad - bc & 0 \\ 0 & ad - bc \end{pmatrix}$$

Thus, we are led to the formula in the statement.

The  $3 \times 3$  matrices can be inverted as well, but this is something more complicated, requiring more theory. We will be back to this, later on.

#### 2B. THE DETERMINANT

#### 2b. The determinant

In order to deal now with the inversion problem in general, for the arbitrary matrices  $A \in M_N(\mathbb{R})$ , we will use the same method as the one above, at N = 2. Let us write indeed our matrix as follows, with  $v_1, \ldots, v_N \in \mathbb{R}^N$  being its column vectors:

 $A = [v_1, \ldots, v_N]$ 

We know from the general results from chapter 1 that, in order for A to be invertible, the vectors  $v_1, \ldots, v_N$  must be linearly independent. Thus, following the observations (1) from the above proof of Theorem 2.3, we are led into the question of understanding when a family of vectors  $v_1, \ldots, v_N \in \mathbb{R}^N$  are linearly independent.

In order to deal with this latter question, let us introduce the following notion:

DEFINITION 2.4. Associated to any vectors  $v_1, \ldots, v_N \in \mathbb{R}^N$  is the volume

$$\det^+(v_1 \dots v_N) = vol < v_1, \dots, v_N >$$

of the parallelepiped made by these vectors.

Here the volume is taken in the standard N-dimensional sense. At N = 1 this volume is a length, at N = 2 this volume is an area, at N = 3 this is the usual 3D volume, and so on. In general, the volume of a body  $X \subset \mathbb{R}^N$  is by definition the number  $vol(X) \in [0, \infty]$ of copies of the unit cube  $C \subset \mathbb{R}^N$  which are needed for filling X.

In order to compute this volume we can use various geometric techniques, and we will see soon that, in what regards the case that we are interested in, namely that of the parallelepipeds  $P \subset \mathbb{R}^N$ , we can basically compute here everything, just by using very basic geometric techniques, essentially based on the Thales theorem.

In relation with our inversion problem, we have the following statement:

**PROPOSITION 2.5.** The quantity  $det^+$  that we constructed, regarded as a function of the corresponding square matrices, formed by column vectors,

$$\det^+: M_N(\mathbb{R}) \to \mathbb{R}_+$$

has the property that a matrix  $A \in M_N(\mathbb{R})$  is invertible precisely when  $det^+(A) > 0$ .

**PROOF.** This follows from Theorem 2.2, and from the general results from chapter 1, which tell us that a matrix  $A \in M_N(\mathbb{R})$  is invertible precisely when its column vectors  $v_1, \ldots, v_N \in \mathbb{R}^N$  are linearly independent. But this latter condition is equivalent to the fact that we must have the following strict inequality:

$$vol < v_1, \ldots, v_N >> 0$$

Thus, we are led to the conclusion in the statement.

35
Summarizing, all this leads us into the explicit computation of det<sup>+</sup>. As a first observation, in 1 dimension we obtain the absolute value of the real numbers:

$$\det^+(a) = |a|$$

In 2 dimensions now, the computation is non-trivial, and we have the following result, making the link with our main result so far, namely Theorem 2.3 above:

THEOREM 2.6. In 2 dimensions we have the following formula,

$$\det^+ \begin{pmatrix} a & b \\ c & d \end{pmatrix} = |ad - bc|$$

with  $det^+: M_2(\mathbb{R}) \to \mathbb{R}_+$  being the function constructed above.

**PROOF.** We must show that the area of the parallelogram formed by  $\binom{a}{c}$ ,  $\binom{b}{d}$  equals the quantity |ad - bc|, which is naturally a difference of areas of two rectangles, either ad - bc, or bc - ad. But this can be done by drawing the parallelogram, along with the vertical lines at a, b, and the horizontal lines at c, d, and then proceeding as follows:

(1) We can either compute the lenghts of all segments appearing on the picture, by using the Thales theorem, and then the area of the parallelogram, by dividing it into triangles, and then using the usual A = bh/2 formula for the area of the triangles.

(2) Or we can solve the problem in "puzzle" style, without algebraic computations, by suitably chopping the parallelogram, along the vertical and horizontal lines, and then rearranging and modifying the pieces as to cover the difference of the two rectangles.

(3) There is as well a third proof, which is somewhat hybrid between the above two ones, consisting in deforming the parallelogram, via linear transformations which keep the area constant, into a rectangle, and then computing the area of this rectangle.

Let us explain the proof (3), which is the most appropriate for our present purposes, especially in what concerns some further generalizations, to be discussed later on. We can assume a, b, c, d > 0 for simplifying, the proof in general being similar. Moreover, by switching if needed the vectors  $\binom{a}{c}, \binom{b}{d}$ , we can assume that we have:

$$\frac{a}{c} > \frac{b}{d}$$

According to these conventions, the picture of our parallelogram is as follows:



Now let us slide the upper side to the left, until we reach the Oy axis. Our parallelogram, which has not changed its area in this process, becomes:



We can further modify this parallelogram, once again by not altering its area, by sliding the right side downwards, until we reach the Ox axis:



Let us compute now the area. In order to compute the quantity x, observe that we have two similar triangles, according to the following picture:



Thus, we are led to the following equation for the number x:

$$\frac{d-x}{b} = \frac{c}{a}$$

By solving this equation, we obtain the following value for x:

$$x = d - \frac{bc}{a}$$

Thus the area of our parallelogram, or rather of the above rectangle obtained from it, which has the same area as the original parallelogram, is given by:

$$ax = ad - bc$$

In linear algebra terms, this leads to the conclusion in the statement.  $\Box$ 

All this is very nice, and we obviously have a beginning of theory here. However, when looking carefully, we can see that our theory has a potential flaw, because:

- (1) In 1 dimension the number a, which is the simplest function of a itself, is certainly a better quantity than the number |a|.
- (2) In 2 dimensions the number ad bc, which is a degree 1 polynomial in a, b, c, d, is certainly a better quantity than the number |ad bc|.

So, let us upgrade right away our theory, by constructing a better function, which does the same job, namely checking if the vectors are proportional, of the following type:

$$\det: M_N(\mathbb{R}) \to \mathbb{R}$$
$$\det = \pm \det^+$$

In order to do this, we just need to find a tricky definition for the sign, as to obtain nice quantities such as a in 1 dimension, ad - bc in 2 dimensions, and so on.

This sign can be introduced as follows:

DEFINITION 2.7. A system of vectors  $v_1, \ldots, v_N \in \mathbb{R}^N$  is called:

- (1) Oriented, if one can continuously pass from the standard basis to it.
- (2) Unoriented, otherwise.

The associated sign is + in the oriented case, and - in the unoriented case.

As a first example, in 1 dimension the basis consists of the single vector e = 1, which can be continuously deformed into any vector a > 0. Thus, the sign is the usual one:

$$sgn(a) = \begin{cases} + & \text{if } a > 0 \\ - & \text{if } a < 0 \end{cases}$$

Thus, in connection with our original question, we are definitely on the good track, because when multiplying |a| by this sign we obtain a itself, as desired:

$$a = sgn(a)|a|$$

In 2 dimensions now, the explicit formula of the sign is as follows:

**PROPOSITION 2.8.** We have the following formula, valid for any 2 vectors in  $\mathbb{R}^2$ ,

$$sgn\left[\binom{a}{c}, \binom{b}{d}\right] = sgn(ad - bc)$$

with the sign function on the right being the usual one, in 1 dimension.

**PROOF.** According to our definitions, the sign of  $\binom{a}{c}$ ,  $\binom{b}{d}$  is as follows:

(1) The sign is + when these vectors come in this order with respect to the counterclockwise rotation.

(2) The sign is - otherwise, when these vectors come in this order with respect to the clockwise rotation.

Thus, when assuming a, b, c, d > 0 for simplifying, we are left with comparing the angles having the numbers c/a and d/b as tangents, and we obtain in this way:

$$sgn\left[\binom{a}{c},\binom{b}{d}\right] = \begin{cases} + & \text{if } \frac{c}{a} < \frac{d}{b} \\ - & \text{if } \frac{c}{a} > \frac{d}{b} \end{cases}$$

But this gives the formula in the statement. The proof in general is similar.

Once again, in connection with our original question, we are on the good track, because when multiplying |ad - bc| by this sign we obtain ad - bc itself, as desired:

$$ad - bc = sgn(ad - bc)|ad - bc|$$

Let us do as well a computation at N = 3. Things here are quite complicated, and we will discuss this later on. In the lack of something elementary, let us formulate:

**PROPOSITION 2.9.** Consider the standard basis of  $\mathbb{R}^3$ , namely:

$$e_1 = \begin{pmatrix} 1\\0\\0 \end{pmatrix} \qquad , \qquad e_2 = \begin{pmatrix} 0\\1\\0 \end{pmatrix} \qquad , \qquad e_3 = \begin{pmatrix} 0\\0\\1 \end{pmatrix}$$

We have then the following sign computations:

- (1)  $sgn(e_1, e_2, e_3) = +.$
- (2)  $sgn(e_1, e_3, e_2) = -.$
- (3)  $sgn(e_2, e_1, e_3) = -.$ (4)  $sgn(e_2, e_3, e_1) = +.$
- (4)  $sgn(e_2, e_3, e_1) = +$ . (5)  $sgn(e_3, e_1, e_2) = +$ .
- (6)  $sgn(e_3, e_2, e_1) = -.$

**PROOF.** In each case the problem is whether one can continuously pass from  $(e_1, e_2, e_3)$  to the basis in statement, and the computations can be done as follows:

(1) In three of the cases under investigation, namely (2,3,6), one of the vectors is unchanged, and the other two are switched. Thus, we are more or less in 2 dimensions, and since the switch here clearly corresponds to -, the sign in these cases is -.

(2) As for the remaining three cases, namely (1,4,5), here the sign can only be +, since things must be 50-50 between + and -, say by symmetry reasons. And this is indeed the case, because what we have here are rotations of the standard basis.

As already mentioned, we will be back to this later, with a general formula for the sign in 3 dimensions. This formula is quite complicated, the idea being that of making out of the  $3 \times 3 = 9$  entries of our vectors a certain quantity, somewhat in the spirit of the one in Proposition 2.8, and then taking the sign of this quantity.

At the level of the general results now, we have:

**PROPOSITION 2.10.** The orientation of a system of vectors changes as follows:

- (1) If we switch the sign of a vector, the associated sign switches.
- (2) If we permute two vectors, the associated sign switches as well.

**PROOF.** Both these assertions are clear from the definition of the sign, because the two operations in question change the orientation of the system of vectors.  $\Box$ 

With the above notion in hand, we can now formulate:

DEFINITION 2.11. The determinant of  $v_1, \ldots, v_N \in \mathbb{R}^N$  is the signed volume

$$\det(v_1 \dots v_N) = \pm vol < v_1, \dots, v_N >$$

of the parallelepiped made by these vectors.

In other words, we are upgrading here Definition 2.4, by adding a sign to the quantity det<sup>+</sup> constructed there, as to potentially reach to good additivity properties:

$$\det(v_1 \dots v_N) = \pm \det^+(v_1 \dots v_N)$$

In relation with our original inversion problem for the square matrices, this upgrade does not change what we have so far, and we have the following statement:

THEOREM 2.12. The quantity det that we constructed, regarded as a function of the corresponding square matrices, formed by column vectors,

$$\det: M_N(\mathbb{R}) \to \mathbb{R}$$

has the property that a matrix  $A \in M_N(\mathbb{R})$  is invertible precisely when  $det(A) \neq 0$ .

**PROOF.** We know from Proposition 2.5 that a matrix  $A \in M_N(\mathbb{R})$  is invertible precisely when  $\det^+(A) = |\det A|$  is strictly positive, and this gives the result.

In the matrix context, we will often use the symbol |. | instead of det:

$$|A| = \det A$$

Let us try now to compute the determinant. In 1 dimension we have of course the formula det(a) = a, because the absolute value fits, and so does the sign:

$$det(a) = sgn(a) \times |a|$$
$$= a$$

In 2 dimensions now, we have the following result:

THEOREM 2.13. In 2 dimensions we have the following formula,

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

with  $|.| = \det$  being the determinant function constructed above.

PROOF. According to our definition, to the computation in Theorem 2.6, and to sign formula from Proposition 2.8, the determinant of a  $2 \times 2$  matrix is given by:

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = sgn \left[ \begin{pmatrix} a \\ c \end{pmatrix}, \begin{pmatrix} b \\ d \end{pmatrix} \right] \times \det^+ \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$
$$= sgn \left[ \begin{pmatrix} a \\ c \end{pmatrix}, \begin{pmatrix} b \\ d \end{pmatrix} \right] \times |ad - bc|$$
$$= sgn(ad - bc) \times |ad - bc|$$
$$= ad - bc$$

Thus, we have obtained the formula in the statement.

## 2c. Basic properties

In order to discuss now arbitrary dimensions, we will need a number of theoretical results. Here is a first series of formulae, coming straight from the definitions:

THEOREM 2.14. The determinant has the following properties:

(1) When multiplying by scalars, the determinant gets multiplied as well:

 $\det(\lambda_1 v_1, \ldots, \lambda_N v_N) = \lambda_1 \ldots \lambda_N \det(v_1, \ldots, v_N)$ 

(2) When permuting two columns, the determinant changes the sign:

$$\det(\ldots, u, \ldots, v, \ldots) = -\det(\ldots, v, \ldots, u, \ldots)$$

(3) The determinant  $det(e_1, \ldots, e_N)$  of the standard basis of  $\mathbb{R}^N$  is 1.

**PROOF.** All this is clear from definitions, as follows:

(1) This follows from definitions, and from Proposition 2.10 (1).

(2) This follows as well from definitions, and from Proposition 2.10(2).

(3) This is clear from our definition of the determinant.

As an application of the above result, we have:

THEOREM 2.15. The determinant of a diagonal matrix is given by:

$$\begin{vmatrix} \lambda_1 \\ & \ddots \\ & & \lambda_N \end{vmatrix} = \lambda_1 \dots \lambda_N$$

That is, we obtain the product of diagonal entries, or of eigenvalues.

**PROOF.** The formula in the statement is clear by using the rules (1) and (3) in Theorem 2.14 above, which in matrix terms give:

$$\begin{vmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_N \end{vmatrix} = \begin{vmatrix} \lambda_1 \dots \lambda_N \end{vmatrix} \begin{vmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{vmatrix}$$
$$= \begin{vmatrix} \lambda_1 \dots \lambda_N \end{vmatrix}$$

As for the last assertion, this is rather a remark.

We will see in a moment that, more generally, the determinant of any diagonalizable matrix is the product of its eigenvalues.

In order to reach to a more advanced theory, let us adopt now the linear map point of view. In this setting, the definition of the determinant reformulates as follows:

### 2C. BASIC PROPERTIES

THEOREM 2.16. Given a linear map, written as f(v) = Av, its "inflation coefficient", obtained as the signed volume of the image of the unit cube, is given by:

$$I_f = \det A$$

More generally,  $I_f$  is the inflation ratio of any parallelepiped in  $\mathbb{R}^N$ , via the transformation f. In particular f is invertible precisely when det  $A \neq 0$ .

PROOF. The only non-trivial thing in all this is the fact that the inflation coefficient  $I_f$ , as defined above, is independent of the choice of the parallelepiped. But this is a generalization of the Thales theorem, which follows from the Thales theorem itself.  $\Box$ 

As a first application of the above linear map viewpoint, we have:

THEOREM 2.17. We have the following formula, valid for any matrices A, B:

$$\det(AB) = \det A \cdot \det B$$

In particular, we have det(AB) = det(BA).

**PROOF.** The decomposition formula in the statement follows by using the associated linear maps, which multiply as follows:

$$f_{AB} = f_A f_B$$

Indeed, when computing the determinant, by using the "inflation coefficient" viewpoint from Theorem 2.16, we obtain the same thing on both sides.

As for the formula det(AB) = det(BA), this is clear from the first formula.

Getting back now to explicit computations, we have the following key result:

**THEOREM** 2.18. The determinant of a diagonalizable matrix

$$A \sim \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_N \end{pmatrix}$$

is the product of its eigenvalues, det  $A = \lambda_1 \dots \lambda_N$ .

PROOF. We know that a diagonalizable matrix can be written in the form  $A = PDP^{-1}$ , with  $D = diag(\lambda_1, \ldots, \lambda_N)$ . Now by using Theorem 2.17, we obtain:

$$\det A = \det(PDP^{-1})$$
$$= \det(DP^{-1}P)$$
$$= \det D$$
$$= \lambda_1 \dots \lambda_N$$

Thus, we are led to the formula in the statement.

Here is another important result, which is very useful for diagonalization:

THEOREM 2.19. The eigenvalues of a matrix  $A \in M_N(\mathbb{R})$  are the roots of

 $P(x) = \det(A - x1_N)$ 

called characteristic polynomial of the matrix.

**PROOF.** We have the following computation, using the fact that a linear map is bijective precisely when the determinant of the associated matrix is nonzero:

$$\exists v, Av = \lambda v \iff \exists v, (A - \lambda 1_N)v = 0$$
$$\iff \det(A - \lambda 1_N) = 0$$

Thus, we are led to the conclusion in the statement.

Here are now some other computations, once again in arbitrary dimensions:

**PROPOSITION 2.20.** We have the following results:

- (1) The determinant of an orthogonal matrix must be  $\pm 1$ .
- (2) The determinant of a projection must be 0 or 1.

**PROOF.** These are elementary results, the idea being as follows:

(1) Here the determinant must be indeed  $\pm 1$ , because the orthogonal matrices map the unit cube to a copy of the unit cube.

(2) Here the determinant is in general 0, because the projections flatten the unit cube, unless we have the identity, where the determinant is 1.  $\Box$ 

We will be back with more explicit computations, later on.

In general now, at the theoretical level, we have the following key result:

THEOREM 2.21. The determinant has the additivity property

 $\det(\dots, u+v, \dots) = \det(\dots, u, \dots) + \det(\dots, v, \dots)$ 

valid for any choice of the vectors involved.

**PROOF.** This follows by doing some elementary geometry, in the spirit of the computations in the proof of Theorem 2.6 above, as follows:

(1) We can either use the Thales theorem, and then compute the volumes of all the parallelepipeds involved, by using basic algebraic formulae.

(2) Or we can solve the problem in "puzzle" style, the idea being to cut the big parallelepiped, and then recover the small ones, after some manipulations.

(3) We can do as well something hybrid, consisting in deforming the parallelepipeds involved, without changing their volumes, and then cutting and gluing.  $\Box$ 

As a basic application of the above result, we have:

44

### 2C. BASIC PROPERTIES

THEOREM 2.22. We have the following results:

- (1) The determinant of a diagonal matrix is the product of diagonal entries.
- (2) The same is true for the upper triangular matrices.
- (3) The same is true for the lower triangular matrices.

**PROOF.** All this can be deduced by using our various general formulae, as follows:

- (1) This is something that we already know, from Theorem 2.15 above.
- (2) This follows by using Theorem 2.14 and Theorem 2.21, then (1), as follows:

(3) This follows as well from Theorem 2.14 and Theorem 2.21, then (1), by proceeding this time from right to left, from the last column towards the first column.  $\Box$ 

We can see from the above that the rules in Theorem 2.14 and Theorem 2.21 are quite powerful, taken altogether. For future reference, let us record these rules:

THEOREM 2.23. The determinant has the following properties:

(1) When adding two columns, the determinants get added:

 $\det(\dots, u+v, \dots) = \det(\dots, u, \dots) + \det(\dots, v, \dots)$ 

(2) When multiplying columns by scalars, the determinant gets multiplied:

 $\det(\lambda_1 v_1, \ldots, \lambda_N v_N) = \lambda_1 \ldots \lambda_N \det(v_1, \ldots, v_N)$ 

(3) When permuting two columns, the determinant changes the sign:

 $\det(\ldots, u, \ldots, v, \ldots) = -\det(\ldots, v, \ldots, u, \ldots)$ 

(4) The determinant  $det(e_1, \ldots, e_N)$  of the standard basis of  $\mathbb{R}^N$  is 1.

**PROOF.** This is something that we already know, which follows by putting together the various formulae from Theorem 2.14 and Theorem 2.21 above.  $\Box$ 

As an important theoretical result now, which will ultimately lead to an algebraic reformulation of the whole determinant problematics, we have:

**THEOREM** 2.24. The determinant of square matrices is the unique map

 $\det: M_N(\mathbb{R}) \to \mathbb{R}$ 

satisfying the conditions in Theorem 2.23 above.

**PROOF.** This can be done in two steps, as follows:

(1) Our first claim is that any map det':  $M_N(\mathbb{R}) \to \mathbb{R}$  satisfying the conditions in Theorem 2.23 must coincide with det on the upper triangular matrices. But this is clear from the proof of Theorem 2.22, which only uses the rules in Theorem 2.23.

(2) Our second claim is that we have det' = det, on all matrices. But this can be proved by putting the matrix in upper triangular form, by using operations on the columns, in the spirit of the manipulations from the proof of Theorem 2.22 above. 

Here is now another important theoretical result:

THEOREM 2.25. The determinant is subject to the row expansion formula

$$\begin{vmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N1} & \dots & a_{NN} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & \dots & a_{2N} \\ \vdots & & \vdots \\ a_{N2} & \dots & a_{NN} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} & \dots & a_{2N} \\ \vdots & & \vdots \\ a_{N1} & a_{N3} & \dots & a_{NN} \end{vmatrix} + \dots + (-1)^{N+1} a_{1N} \begin{vmatrix} a_{21} & \dots & a_{2,N-1} \\ \vdots & & \vdots \\ a_{N1} & \dots & a_{N,N-1} \end{vmatrix}$$

and this method fully computes it, by recurrence.

**PROOF.** This follows from the fact that the formula in the statement produces a certain function det :  $M_N(\mathbb{R}) \to \mathbb{R}$ , which has the 4 properties in Theorem 2.23.  $\Box$ 

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We can expand as well over the columns, as follows:

THEOREM 2.26. The determinant is subject to the column expansion formula

$$\begin{vmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N1} & \dots & a_{NN} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & \dots & a_{2N} \\ \vdots & & \vdots \\ a_{N2} & \dots & a_{NN} \end{vmatrix} - a_{21} \begin{vmatrix} a_{12} & \dots & a_{1N} \\ a_{32} & \dots & a_{3N} \\ \vdots & & \vdots \\ a_{N2} & \dots & a_{NN} \end{vmatrix} + \dots + (-1)^{N+1} a_{N1} \begin{vmatrix} a_{12} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N-1,2} & \dots & a_{N-1,N} \end{vmatrix}$$

and this method fully computes it, by recurrence.

**PROOF.** This follows by using the same argument as for the rows.

We can now complement Theorem 2.23 with a similar result for the rows:

THEOREM 2.27. The determinant has the following properties:

(1) When adding two rows, the determinants get added:

$$\det \begin{pmatrix} \vdots \\ u+v \\ \vdots \end{pmatrix} = \det \begin{pmatrix} \vdots \\ u \\ \vdots \end{pmatrix} + \det \begin{pmatrix} \vdots \\ v \\ \vdots \end{pmatrix}$$

(2) When multiplying row by scalars, the determinant gets multiplied:

$$\det \begin{pmatrix} \lambda_1 v_1 \\ \vdots \\ \lambda_N v_N \end{pmatrix} = \lambda_1 \dots \lambda_N \det \begin{pmatrix} v_1 \\ \vdots \\ v_N \end{pmatrix}$$

(3) When permuting two rows, the determinant changes the sign.

PROOF. This follows indeed by using the using various formulae established above, and is best seen by using the column expansion formula from Theorem 2.26.  $\Box$ 

We can see from the above that the determinant is the subject to many interesting formulae, and that some of these formulae, when taken altogether, uniquely determine it. In all this, what is the most luminous is certainly the definition of the determinant, as a volume. As for the second most luminous of our statements, this is Theorem 2.24 above, which is something a bit abstract, but both beautiful and useful. So, as a final theoretical statement now, here is an alternative reformulation of Theorem 2.24:

**THEOREM** 2.28. The determinant of the systems of vectors

$$\det: \mathbb{R}^N \times \ldots \times \mathbb{R}^N \to \mathbb{R}$$

is multilinear, alternate and unital, and unique with these properties.

PROOF. This is a fancy reformulation of Theorem 2.24, with the various properties of det from the statement being those from Theorem 2.23.  $\Box$ 

As a conclusion to all this, we have now a full theory for the determinant, and we can freely use all the above results, definitions and theorems alike, and even start forgetting what is actually definition, and what is theorem.

# 2d. Sarrus and beyond

As a first application of the above methods, we can now prove:

 $\square$ 

THEOREM 2.29. The determinant of the  $3 \times 3$  matrices is given by

$$\begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = aei + bfg + cdh - ceg - bdi - afh$$

which can be memorized by using Sarrus' triangle method,

$$\det = \begin{pmatrix} * & * \\ & * \end{pmatrix} + \begin{pmatrix} * & * \\ & * \end{pmatrix} + \begin{pmatrix} * & * \\ & * \end{pmatrix} + \begin{pmatrix} * & * \\ & * \end{pmatrix} - \begin{pmatrix} & * \\ & * \end{pmatrix} + \begin{pmatrix} * & * \\ & * \end{pmatrix} + \begin{pmatrix} * & * \\ & * \end{pmatrix} + \begin{pmatrix} * & * \\ & * \end{pmatrix}$$

"triangles parallel to the diagonal, minus triangles parallel to the antidiagonal".

**PROOF.** Here is the computation, using Theorem 2.25 above:

$$\begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = a \begin{vmatrix} e & f \\ h & i \end{vmatrix} - b \begin{vmatrix} d & f \\ g & i \end{vmatrix} + c \begin{vmatrix} d & e \\ g & h \end{vmatrix}$$
$$= a(ei - fh) - b(di - fg) + c(dh - eg)$$
$$= aei - afh - bdi + bfg + cdh - ceg$$
$$= aei + bfg + cdh - ceg - bdi - afh$$

Thus, we obtain the formula in the statement.

As a first application, let us go back to the inversion problem for the  $3 \times 3$  matrices, that we left open in the above. We can now solve this problem, as follows:

THEOREM 2.30. The inverses of the  $3 \times 3$  matrices are given by

$$\begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix}^{-1} = \frac{1}{D} \begin{pmatrix} ei - fh & ch - bi & bf - ce \\ fg - di & ai - cg & cd - af \\ dh - eg & bg - ah & ae - bd \end{pmatrix}$$

with D being the determinant. When D = 0, the matrix is not invertible.

PROOF. We can use here the same method as for the  $2 \times 2$  matrices. To be more precise, in order for the matrix to be invertible, we must have:

 $D \neq 0$ 

The trick now is to look for solutions of the following problem:

$$\begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} \begin{pmatrix} * & * & * \\ * & * & * \\ * & * & * \end{pmatrix} = \begin{pmatrix} D & 0 & 0 \\ 0 & D & 0 \\ 0 & 0 & D \end{pmatrix}$$

We know from Theorem 2.29 above that the determinant is given by:

$$D = aei + bfg + cdh - ceg - bdi - afh$$

But this leads, via some obvious choices, to the following solution:

$$\begin{pmatrix} * & * & * \\ * & * & * \\ * & * & * \end{pmatrix} = \begin{pmatrix} ei - fh & ch - bi & bf - ce \\ fg - di & ai - cg & cd - af \\ dh - eg & bg - ah & ae - bd \end{pmatrix}$$

Thus, by rescaling, we obtain the formula in the statement.

In fact, we can now fully solve the inversion problem, as follows:

THEOREM 2.31. The inverse of a square matrix, having nonzero determinant,

$$A = \begin{pmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N1} & \dots & a_{NN} \end{pmatrix}$$

is given by the following formula,

$$A^{-1} = \frac{1}{\det A} \begin{pmatrix} \det A^{(11)} & -\det A^{(21)} & \det A^{(31)} & \dots \\ -\det A^{(12)} & \det A^{(22)} & -\det A^{(32)} & \dots \\ \det A^{(13)} & -\det A^{(23)} & \det A^{(33)} & \dots \\ \vdots & \vdots & \vdots & \end{pmatrix}$$

where  $A^{(ij)}$  is the matrix A, with the *i*-th row and *j*-th column removed.

PROOF. This follows indeed by using the row expansion formula from Theorem 2.25, which in terms of the matrix  $A^{-1}$  in the statement reads  $AA^{-1} = 1$ .

In practice, the above result leads to the following algorithm, which is quite easy to memorize, for computing the inverse:

- (1) Delete rows and columns, and compute the corresponding determinants.
- (2) Transpose, and add checkered signs.
- (3) Divide by the determinant.

Observe that this generalizes our previous computations at N = 2, 3. As an illustration, consider an arbitrary  $2 \times 2$  matrix, written as follows:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

By deleting rows and columns we obtain  $1 \times 1$  matrices, and so the matrix formed by the determinants det $(A^{(ij)})$  is as follows:

$$M = \begin{pmatrix} d & c \\ b & a \end{pmatrix}$$

Now by transposing, adding checkered signs and dividing by  $\det A$ , we obtain:

$$A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Similarly, at N = 3 what we obtain is the Sarrus formula, from Theorem 2.29.

As a new application now, let us record the following result, at N = 4:

THEOREM 2.32. The determinant of the  $4 \times 4$  matrices is given by

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$$\begin{vmatrix} a_1 & a_2 & a_3 & a_4 \\ b_1 & b_2 & b_3 & b_4 \\ c_1 & c_2 & c_3 & c_4 \\ d_1 & d_2 & d_3 & d_4 \end{vmatrix}$$

$$= a_1 b_2 c_3 d_4 - a_1 b_2 c_4 d_3 - a_1 b_3 c_2 d_4 + a_1 b_3 c_4 d_2 + a_1 b_4 c_2 d_3 - a_1 b_4 c_3 d_2$$

$$- a_2 b_1 c_3 d_4 + a_2 b_1 c_4 d_3 + a_2 b_3 c_1 d_4 - a_2 b_3 c_4 d_1 - a_2 b_4 c_1 d_3 + a_2 b_4 c_3 d_1$$

$$+ a_3 b_1 c_2 d_4 + a_3 b_1 c_4 d_2 - a_3 b_2 c_1 d_4 + a_3 b_2 c_4 d_1 + a_3 b_4 c_1 d_2 - a_3 b_4 c_2 d_1$$

$$- a_4 b_1 c_2 d_3 + a_4 b_1 c_3 d_2 - a_4 b_2 c_1 d_3 - a_4 b_2 c_3 d_1 - a_4 b_3 c_1 d_2 + a_4 b_3 c_2 d_1$$

and the formula of the inverse is as follows, involving 16 Sarrus determinants,

$$A^{-1} = \frac{1}{\det A} \begin{pmatrix} \det A^{(11)} & -\det A^{(21)} & \det A^{(31)} & -\det A^{(41)} \\ -\det A^{(12)} & \det A^{(22)} & -\det A^{(32)} & \det A^{(42)} \\ \det A^{(13)} & -\det A^{(23)} & \det A^{(33)} & -\det A^{(43)} \\ -\det A^{(14)} & \det A^{(24)} & -\det A^{(34)} & \det A^{(44)} \end{pmatrix}$$

where  $A^{(ij)}$  is the matrix A, with the *i*-th row and *j*-th column removed.

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**PROOF.** The formula for the determinant follows by developing over the first row, then by using the Sarrus formula, for each of the 4 smaller determinants which appear:

$\begin{vmatrix} a_1 \\ b_1 \\ c_1 \\ d_1 \end{vmatrix}$	$a_2 \\ b_2 \\ c_2 \\ d_2$	$egin{array}{c} a_3 \ b_3 \ c_3 \ d_3 \end{array}$	$egin{array}{c} a_4 \ b_4 \ c_4 \ d_4 \end{array}$	=	$a_1$	$\begin{vmatrix} b_2 \\ c_2 \\ d_2 \end{vmatrix}$	$b_3$ $c_3$ $d_3$	$\begin{vmatrix} b_4 \\ c_4 \\ d_4 \end{vmatrix}$	$-a_{2}$	$egin{array}{c} b_1 \ c_1 \ d_1 \end{array}$	$b_3$ $c_3$ $d_3$	$\begin{array}{c} b_4 \\ c_4 \\ d_4 \end{array}$
				+	$a_3$	$\begin{vmatrix} b_1 \\ c_1 \\ d_1 \end{vmatrix}$	$b_2 \\ c_2 \\ d_2$	$\begin{array}{c} b_4 \\ c_4 \\ d_4 \end{array}$	$-a_{4}$	$egin{array}{c} b_1 \ c_1 \ d_1 \end{array}$	$b_2 \\ c_2 \\ d_2$	$egin{array}{c} b_3 \ c_3 \ d_3 \end{array}$

As for the formula of the inverse, this is something that we already know.

Let us discuss now the general formula of the determinant, at arbitrary values  $N \in \mathbb{N}$  of the matrix size, generalizing those that we have at N = 2, 3. We will need:

DEFINITION 2.33. A permutation of  $\{1, \ldots, N\}$  is a bijection, as follows:

$$\sigma: \{1, \ldots, N\} \to \{1, \ldots, N\}$$

The set of such permutations is denoted  $S_N$ .

There are many possible notations for the permutations, the simplest one consisting in writing the numbers  $1, \ldots, N$ , and below them, their permuted versions:

$$\sigma = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 \\ 2 & 1 & 4 & 5 & 3 \end{pmatrix}$$

Another method, which is faster, and has many other advantages, is by denoting the permutations as diagrams, going by definition from top to bottom:



Here are some basic properties of the permutations:

**THEOREM** 2.34. The permutations have the following properties:

- (1) There are N! of them.
- (2) They are stable by composition, and inversion.

**PROOF.** In order to construct a permutation  $\sigma \in S_N$ , we have:

- N choices for the value of  $\sigma(N)$ .

-(N-1) choices for the value of  $\sigma(N-1)$ .

- (N-2) choices for the value of  $\sigma(N-2)$ .
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– and so on, up to 1 choice for the value of  $\sigma(1)$ .

Thus, we have N! choices, as claimed. As for the second assertion, this is clear.  $\Box$ 

We will need the following key result:

**THEOREM** 2.35. The permutations have a signature function

 $\varepsilon: S_N \to \{\pm 1\}$ 

which can be defined in the following equivalent ways:

- (1) As  $(-1)^c$ , where c is the number of inversions.
- (2) As  $(-1)^t$ , where t is the number of transpositions.
- (3) As  $(-1)^{\circ}$ , where o is the number of odd cycles.
- (4) As  $(-1)^x$ , where x is the number of crossings.
- (5) As the sign of the corresponding permuted basis of  $\mathbb{R}^N$ .

PROOF. This is something important, and quite subtle, to be systematically used in what follows. As a first observation, we can see right away a relation with the determinant, coming from (5) above. Thus, we already have some knowledge here, for instance coming from Proposition 2.9 above, which computes the signature of the permutations  $\sigma \in S_3$ .

In practice now, we have explain what the numbers c, t, o, x appearing in (1-4) above exactly are, then why they are well-defined modulo 2, then why they are equal to each other, and finally why the constructions (1-4) yield the same sign as (5).

Let us begin with the first two steps, namely precise definition of c, t, o, x, and fact that these numbers are well-defined modulo 2:

(1) The idea here is that given any two numbers i < j among  $1, \ldots, N$ , the permutation can either keep them in the same order,  $\sigma(i) < \sigma(j)$ , or invert them:

$$\sigma(j) > \sigma(i)$$

Now by making i < j vary over all pairs of numbers in  $1, \ldots, N$ , we can count the number of inversions, and call it c. This is an integer,  $c \in \mathbb{N}$ , which is well-defined.

(2) Here the idea, which is something quite intuitive, is that any permutation appears as a product of switches, also called transpositions:

 $i \leftrightarrow j$ 

The decomposition as a product of transpositions is not unique, but the number t of the needed transpositions is unique, when considered modulo 2. This follows for instance from the equivalence of (2) with (1,3,4,5), explained below.

(3) Here the point is that any permutation decomposes, in a unique way, as a product of cycles, which are by definition permutations of the following type:

$$i_1 \rightarrow i_2 \rightarrow i_3 \rightarrow \ldots \rightarrow i_k \rightarrow i_1$$

Some of these cycles have even length, and some others have odd length. By counting those having odd length, we obtain a well-defined number  $o \in \mathbb{N}$ .

(4) Here the method is that of drawing the permutation, as we usually do, and by avoiding triple crossings, and then counting the number of crossings. This number x depends on the way we draw the permutations, but modulo 2, we always get the same number. Indeed, this follows from the fact that we can continuously pass from a drawing to each other, and that when doing so, the number of crossings can only jump by  $\pm 2$ .

Summarizing, we have 4 different definitions for the signature of the permutations, which all make sense, constructed according to (1-4) above. Regarding now the fact that we always obtain the same number, this can be established as follows:

(1)=(2) This is clear, because any transposition inverts once, modulo 2.

(1)=(3) This is clear as well, because the odd cycles invert once, modulo 2.

### 2D. SARRUS AND BEYOND

(1)=(4) This comes from the fact that the crossings correspond to inversions.

(2)=(3) This follows by decomposing the cycles into transpositions.

(2)=(4) This comes from the fact that the crossings correspond to transpositions.

(3)=(4) This follows by drawing a product of cycles, and counting the crossings.

Finally, in what regards the equivalence of all these constructions with (5), here simplest is to use (2). Indeed, we already know that the sign of a system of vectors switches when interchaning two vectors, and so the equivalence between (2,5) is clear.

We can now formulate a key result, as follows:

THEOREM 2.36. We have the following formula for the determinant,

$$\det A = \sum_{\sigma \in S_N} \varepsilon(\sigma) A_{1\sigma(1)} \dots A_{N\sigma(N)}$$

with the signature function being the one introduced above.

PROOF. Indeed, when computing the determinant by using the standard rules, we are led into permutations, and into the formula in the statement.  $\Box$ 

As a basic example, in 2 dimensions we recover the usual formula:

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

In 3 dimensions, we recover the Sarrus formula:

$$\begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = aei + bfg + cdh - ceg - bdi - afh$$

Observe that the triangles in the Sarrus formula correspond to the permutations of  $\{1, 2, 3\}$ , and their signs correspond to the signatures of these permutations:

$$det = \begin{pmatrix} * & & \\ & * & \\ & & * \end{pmatrix} + \begin{pmatrix} & * & \\ & * & \\ * & & \end{pmatrix} + \begin{pmatrix} * & & \\ & * & \\ & & * \end{pmatrix} + \begin{pmatrix} & * & \\ & * & \\ & & * \end{pmatrix} + \begin{pmatrix} * & & \\ & * & \\ & & * \end{pmatrix} + \begin{pmatrix} * & & \\ & * & \\ & & * \end{pmatrix}$$

Also, in 4 dimensions, we recover the formula that we already know, as follows:

THEOREM 2.37. The determinant of the  $4 \times 4$  matrices is given by

$$\begin{vmatrix} a_1 & a_2 & a_3 & a_4 \\ b_1 & b_2 & b_3 & b_4 \\ c_1 & c_2 & c_3 & c_4 \\ d_1 & d_2 & d_3 & d_4 \end{vmatrix}$$
  
=  $a_1b_2c_3d_4 - a_1b_2c_4d_3 - a_1b_3c_2d_4 + a_1b_3c_4d_2 + a_1b_4c_2d_3 - a_1b_4c_3d_2$   
-  $a_2b_1c_3d_4 + a_2b_1c_4d_3 + a_2b_3c_1d_4 - a_2b_3c_4d_1 - a_2b_4c_1d_3 + a_2b_4c_3d_1$   
+  $a_3b_1c_2d_4 + a_3b_1c_4d_2 - a_3b_2c_1d_4 + a_3b_2c_4d_1 + a_3b_4c_1d_2 - a_3b_4c_2d_1$   
-  $a_4b_1c_2d_3 + a_4b_1c_3d_2 - a_4b_2c_1d_3 - a_4b_2c_3d_1 - a_4b_3c_1d_2 + a_4b_3c_2d_1$ 

with the generic term being of the following form, with  $\sigma \in S_4$ ,

 $\pm a_{\sigma(1)}b_{\sigma(2)}c_{\sigma(3)}d_{\sigma(4)}$ 

and with the sign being  $\varepsilon(\sigma)$ , computable by using Theorem 2.35.

PROOF. We can indeed recover this formula as well as a particular case of Theorem 2.36. To be more precise, the permutations in the statement are listed according to the lexicographic order, and the computation of the corresponding signatures is something quite elementary, by using the various rules from Theorem 2.35.  $\Box$ 

As another application, we have the following key result:

THEOREM 2.38. We have the formula

$$\det A = \det A^t$$

valid for any square matrix A.

PROOF. This follows from the formula in Theorem 2.36. Indeed, we have:

$$\det A^{t} = \sum_{\sigma \in S_{N}} \varepsilon(\sigma) (A^{t})_{1\sigma(1)} \dots (A^{t})_{N\sigma(N)}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma) A_{\sigma(1)1} \dots A_{\sigma(N)N}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma) A_{1\sigma^{-1}(1)} \dots A_{N\sigma^{-1}(N)}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma^{-1}) A_{1\sigma^{-1}(1)} \dots A_{N\sigma^{-1}(N)}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma) A_{1\sigma(1)} \dots A_{N\sigma(N)}$$

$$= \det A$$

Thus, we are led to the formula in the statement.

Good news, we have now in our bag all the needed techniques for computing the determinant.

Here is however a nice and important example of a determinant, whose computation uses some interesting new techniques, going beyond what has been said above:

THEOREM 2.39. We have the Vandermonde determinant formula

$$\begin{vmatrix} 1 & 1 & 1 & \dots & 1 \\ x_1 & x_2 & x_3 & \dots & x_N \\ x_1^2 & x_2^2 & x_3^2 & \dots & x_N \\ \vdots & \vdots & \vdots & & \vdots \\ \vdots & \vdots & \vdots & & \vdots \\ x_1^{N-1} & x_2^{N-1} & x_3^{N-1} & \dots & x_N^{N-1} \end{vmatrix} = \prod_{i>j} (x_i - x_j)$$

valid for any  $x_1, \ldots, x_N \in \mathbb{R}$ .

**PROOF.** By expanding over the columns, the determinant in question, say D, is a polynomial in the variables  $x_1, \ldots, x_N$ , having degree N - 1 in each variable.

Now since  $x_i = x_j$  makes D vanish, for any  $i \neq j$ , we conclude that we must have:

$$D = c \prod_{i>j} (x_i - x_j)$$

As for the constant  $c \in \mathbb{R}$ , this can be computed by recurrence, and we obtain:

c = 1

Thus, we are led to the formula in the statement.

Getting back now to the diagonalization problem, we know from Theorem 2.19 that the eigenvalues of a matrix  $A \in M_N(\mathbb{R})$  appear as roots of the characteristic polynomial:

$$P(x) = \det(A - x1_N)$$

Thus, we can compute these eigenvalues. The problem, however, is that certain matrices can have characteristic polynomials of type  $P(x) = x^2 + 1$ , and this suggests that these matrices might be diagonalizable over  $\mathbb{C}$ . Thus, before getting into diagonalization problems, we must upgrade our theory, and talk about complex matrices.

## 2e. Exercises

There has been a lot of exciting theory in this chapter, with some details sometimes missing, and our first exercises will be about this. First, we have:

EXERCISE 2.40. Fill in all the geometric details in the basic theory of the determinant, by using the same type of arguments as those in the proof of

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$$

which was fully proved in the above, namely geometric manipulations, and Thales.

To be more precise here, passed some issues with the sign and orientation, which are all elementary, the above  $2 \times 2$  determinant formula was subject of Theorem 2.6, coming with a full and honest proof. The problem is that of using the same arguments, namely basic geometry, as to have a full proof of Theorem 2.16 and Theorem 2.21 as well.

As a second theoretical exercise, we have:

EXERCISE 2.41. Prove with full details, based on the above, that the determinant of the systems of vectors

$$\det: \mathbb{R}^N \times \ldots \times \mathbb{R}^N \to \mathbb{R}$$

is multilinear, alternate and unital, and unique with these properties. Then try to prove as well this directly, without any reference to geometry.

To be more precise, in what regards the first question, this is something that we already discussed in the above, with only a few details missing, and the problem is that of recovering these details. As for the second question, this is something more tricky, and there are several possible approaches here, all being interesting and enjoyable.

As a third and last theoretical exercise, we have:

EXERCISE 2.42. Work out, with full details, the theory of the signature map

$$\varepsilon: S_N \to \{\pm 1\}$$

as outlined in Theorem 2.35 above and its proof.

As before, these are things that we already discussed, with a few details missing.

Finally, as a concrete, computational exercise, we have:

EXERCISE 2.43. Prove that for a matrix  $H \in M_N(\pm 1)$ , we have

$$|\det H| \leq N^{N/2}$$

and then find the maximizers of  $|\det H|$ , at small values of N.

Here the first question is theoretical, and its proof should not be difficult. As for the second question, which is quite tricky, the higher the  $N \in \mathbb{N}$  you get to, the better.

# CHAPTER 3

# Complex matrices

# 3a. Complex numbers

We have seen in the previous chapter that the study of the real matrices  $A \in M_N(\mathbb{R})$  suggests the use of the complex numbers. Indeed, even simple matrices like the  $2 \times 2$  ones can, at least in a formal sense, have complex eigenvalues.

In what follows we discuss the complex matrices  $A \in M_N(\mathbb{C})$ . We will see that the theory here is much more complete than in the real case. As an application, we will solve in this way the problems left open in the real case, concerning the diagonalization.

Let us begin with the complex numbers. There is a lot of magic here, and we will carefully explain this material. Their definition is as follows:

DEFINITION 3.1. The complex numbers are variables of the form

$$x = a + ib$$

which add in the obvious way, and multiply according to the following rule:

 $i^2 = -1$ 

In other words, we consider variables as above, without bothering for the moment with their precise meaning. Now consider two such complex numbers:

$$x = a + ib$$

$$y = c + id$$

The formula for the sum is then the obvious one, as follows:

$$x + y = (a + c) + i(b + d)$$

As for the formula of the product, by using the rule  $i^2 = -1$ , we obtain:

$$xy = (a+ib)(c+id)$$
  
=  $ac+iad+ibc+i^{2}bd$   
=  $ac+iad+ibc-bd$   
=  $(ac-bd)+i(ad+bc)$ 

### 3. COMPLEX MATRICES

Thus, the complex numbers as introduced above are well-defined. The multiplication formula is of course quite tricky, and hard to memorize, but we will see later some alternative ways, which are more conceptual, for performing the multiplication.

The advantage of using the complex numbers comes from the fact that the equation  $x^2 = 1$  has now a solution, x = i. In fact, this equation has two solutions, namely:

$$x = \pm i$$

This is of course very good news. More generally, we have the following key result, regarding the arbitrary degree 2 equations, with real coefficients:

THEOREM 3.2. The complex solutions of  $ax^2 + bx + c = 0$  with  $a, b, c \in \mathbb{R}$  are

$$x_{1,2} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

with the square root of negative real numbers being defined as:

$$\sqrt{-m} = \pm i\sqrt{m}$$

**PROOF.** We can write our equation in the following way:

$$ax^{2} + bx + c = 0 \iff x^{2} + \frac{b}{a}x + \frac{c}{a} = 0$$
$$\iff \left(x + \frac{b}{2a}\right)^{2} - \frac{b^{2}}{4a^{2}} + \frac{c}{a} = 0$$
$$\iff \left(x + \frac{b}{2a}\right)^{2} = \frac{b^{2} - 4ac}{4a^{2}}$$
$$\iff x + \frac{b}{2a} = \pm \frac{\sqrt{b^{2} - 4ac}}{2a}$$

Thus, we are led to the conclusion in the statement.

We will see later that any degree 2 complex equation has solutions as well, and that more generally, any polynomial equation, real or complex, with arbitrary degree  $N \in \mathbb{N}$ , has solutions. These are all important results, having applications to linear algebra.

We can represent the complex numbers in the plane, in the following way:

**PROPOSITION 3.3.** The complex numbers, written as usual

$$x = a + ib$$

can be represented in the plane, according to the following identification:

$$x = \begin{pmatrix} a \\ b \end{pmatrix}$$

With this convention, the sum of complex numbers is the usual sum of vectors.

**PROOF.** Consider indeed two arbitrary complex numbers:

$$x = a + ib$$

$$y = c + id$$

Their sum is then by definition the following complex number:

$$x + y = (a + c) + i(b + d)$$

Now let us represent x, y in the plane, as in the statement:

$$x = \begin{pmatrix} a \\ b \end{pmatrix} \quad , \quad y = \begin{pmatrix} c \\ d \end{pmatrix}$$

In this picture, their sum is given by the following formula:

$$x + y = \begin{pmatrix} a + c \\ b + d \end{pmatrix}$$

But this is indeed the vector corresponding to x + y, so we are done.

Observe that in the above picture, the real numbers correspond to the numbers on the Ox axis. As for the purely imaginary numbers, these lie on the Oy axis, with:

$$i = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

All this is very nice, but it order to understand now the multiplication, we must do something more complicated, namely using polar coordinates. Let us start with:

DEFINITION 3.4. The complex numbers x = a + ib can be written in polar coordinates,

$$x = r(\cos t + i\sin t)$$

with the connecting formulae being

$$a = r\cos t$$
$$b = r\sin t$$

and in the other sense being

$$r = \sqrt{a^2 + b^2}$$
$$\tan t = b/a$$

and with r, t being called modulus, and argument.

There is a clear relation here with the vector notation from Proposition 3.3, because r is the length of the vector, and t is the angle made by the vector with the Ox axis.

As a basic example here, the number i takes the following form:

$$i = \cos\left(\frac{\pi}{2}\right) + i\sin\left(\frac{\pi}{2}\right)$$

### 3. COMPLEX MATRICES

The point now is that in polar coordinates, the multiplication formula for the complex numbers, which was so far something quite opaque, takes a very simple form:

THEOREM 3.5. Two complex numbers written in polar coordinates,

$$x = r(\cos s + i\sin s)$$

$$y = p(\cos t + i\sin t)$$

multiply according to the following formula:

$$xy = rp(\cos(s+t) + i\sin(s+t))$$

In other words, the moduli multiply, and the arguments sum up.

**PROOF.** This follows from the following well-known trigonometry formulae, which can be proved by doing some plane geometry:

 $\cos(s+t) = \cos s \cos t - \sin s \sin t$  $\sin(s+t) = \cos s \sin t + \sin s \cos t$ 

Indeed, we can assume that we have r = p = 1, by dividing everything by these numbers. Now with this assumption made, we have the following computation:

$$xy = (\cos s + i \sin s)(\cos t + i \sin t)$$
  
= (\cos s \cos t - \sin s \sin t) + i(\cos s \sin t + \sin s \cos t)  
= \cos(s + t) + i \sin(s + t)

Thus, we are led to the conclusion in the statement.

The above result, which is based on some non-trivial trigonometry, is quite powerful. As a basic application of it, we can now compute powers, as follows:

THEOREM 3.6. The powers of a complex number, written in polar form,

$$x = r(\cos t + i\sin t)$$

are given by the following formula, valid for any exponent  $k \in \mathbb{N}$ :

$$x^k = r^k (\cos kt + i\sin kt)$$

Moreover, this formula holds in fact for any  $k \in \mathbb{Z}$ , and even for any  $k \in \mathbb{Q}$ .

**PROOF.** Given a complex number x, written in polar form as above, and an exponent  $k \in \mathbb{N}$ , we have indeed the following computation, with k terms everywhere:

$$\begin{aligned} x^k &= x \dots x \\ &= r(\cos t + i \sin t) \dots r(\cos t + i \sin t) \\ &= r \dots r([\cos(t + \dots + t) + i \sin(t + \dots + t))) \\ &= r^k(\cos kt + i \sin kt) \end{aligned}$$

60

Thus, we are done with the case  $k \in \mathbb{N}$ . Regarding now the generalization to the case  $k \in \mathbb{Z}$ , it is enough here to do the verification for k = -1, where the formula is:

$$x^{-1} = r^{-1}(\cos(-t) + i\sin(-t))$$

But this number  $x^{-1}$  is indeed the inverse of x, because:

$$xx^{-1} = r(\cos t + i\sin t) \cdot r^{-1}(\cos(-t) + i\sin(-t))$$
  
=  $\cos(t - t) + i\sin(t - t)$   
=  $\cos 0 + i\sin 0$   
= 1

Finally, regarding the generalization to the case  $k \in \mathbb{Q}$ , it is enough to do the verification for exponents of type k = 1/n, with  $n \in \mathbb{N}$ . The claim here is that:

$$x^{1/n} = r^{1/n} \left[ \cos\left(\frac{t}{n}\right) + i\sin\left(\frac{t}{n}\right) \right]$$

In order to prove this, let us compute the *n*-th power of this number. We can use the power formula for the exponent  $n \in \mathbb{N}$ , that we already established, and we obtain:

$$(x^{1/n})^n = (r^{1/n})^n \left[ \cos\left(n \cdot \frac{t}{n}\right) + i \sin\left(n \cdot \frac{t}{n}\right) \right]$$
$$= r(\cos t + i \sin t)$$
$$= x$$

Thus, we have indeed a n-th root of x, and our proof is now complete.

We should mention that there is a bit of ambiguity in the above, in the case of the exponents  $k \in \mathbb{Q}$ , due to the fact that the square roots, and the higher roots as well, can take multiple values, in the complex number setting. We will be back to this.

Let us discuss now the final and most convenient writing of the complex numbers, which is a well-known variation on the polar writing, as follows:

$$x = re^{it}$$

In what follows we will not really need the true power of this formula, which is of analytic nature, due to occurrence of the number e. However, we would like to use the notation  $x = re^{it}$ , as everyone does, among others because it simplifies the writing.

The point with the above formula comes from the following deep result:

THEOREM 3.7. We have the following formula, valid for any  $t \in \mathbb{R}$ ,

$$e^{it} = \cos t + i\sin t$$

where e = 2.7182... is the usual constant from analysis.

### 3. COMPLEX MATRICES

PROOF. In order to prove such a result, we must first recall what e is, and what  $e^x$  is. One way of viewing things is that  $e^x$  is the unique function satisfying:

$$(e^x)' = e^x$$
$$e^0 = 1$$

Then, we can set  $e = e^1$ , and then prove that  $e^x$  equals indeed e to the power x. This is a bit abstract, but is convenient for our purposes. Indeed, the solution to the above derivative problem is easy to work out, as a series, by recurrence, the answer being:

$$e^x = \sum_k \frac{x^k}{k!}$$

Now let us plug x = it in this formula. We obtain the following formula:

$$e^{it} = \sum_{k} \frac{(it)^{k}}{k!}$$
  
=  $\sum_{k=2l} \frac{(it)^{k}}{k!} + \sum_{k=2l+1} \frac{(it)^{k}}{k!}$   
=  $\sum_{l} (-1)^{l} \frac{t^{2l}}{(2l)!} + i \sum_{l} (-1)^{l} \frac{t^{2l+1}}{(2l+1)!}$ 

Our claim now, which will complete the proof, is that we have:

$$\cos t = \sum_{l} (-1)^{l} \frac{t^{2l}}{(2l)!}$$
$$\sin t = \sum_{l} (-1)^{l} \frac{t^{2l+1}}{(2l+1)!}$$

In order to prove this claim, let us compute the Taylor series of cos and sin. By using the formulae for sums of angles, used in the proof of Theorem 3.5, we have:

Thus, we know how to differentiate sin and cos, once, then twice, then as many times as we want to, and with this we can compute the corresponding Taylor series, the answers being those given above. Now by putting everything together, we have:

$$e^{it} = \cos t + i\sin t$$

Thus, we are led to the conclusion in the statement.

All this was quite brief, but we will be back to it with full details in chapters 5-8 below, when doing analysis. For the moment, let us just enjoy all this. We first have:

THEOREM 3.8. We have the following formula,

$$e^{\pi i} = -1$$

and we have  $E = mc^2$  as well.

**PROOF.** This comes from Theorem 3.7, by setting  $t = \pi$ . Indeed, we obtain:

f

$$e^{\pi i} = \cos \pi + i \sin \pi$$
$$= -1 + i \cdot 0$$
$$= -1$$

As for the second formula, this is something quite deep as well.

Now back to our  $x = re^{it}$  objectives, with the above theory in hand we can indeed use from now on this notation, the complete statement being as follows:

THEOREM 3.9. The complex numbers x = a + ib can be written in polar coordinates,

$$x = re^{it}$$

with the connecting formulae being

$$a = r \cos t$$
$$b = r \sin t$$

and in the other sense being

$$r = \sqrt{a^2 + b^2}$$
$$\tan t = b/a$$

and with r, t being called modulus, and argument.

PROOF. This is a reformulation of Definition 3.4, by using the formula  $e^{it} = \cos t + i \sin t$  from Theorem 3.7, and multiplying everything by r.

We can now go back to the basics, namely the addition and multiplication of the complex numbers, and we have the following result:

THEOREM 3.10. In polar coordinates, the complex numbers multiply as

$$re^{is} \cdot pe^{it} = rp \, e^{i(s+t)}$$

with the arguments s, t being taken modulo  $2\pi$ .

PROOF. This is something that we already know, from Theorem 3.5 above, reformulated by using the notations from Theorem 3.9. Observe that this follows as well directly, from the fact that we have  $e^{a+b} = e^a e^b$ , that we know from analysis.

The above formula is obviously very powerful. However, in polar coordinates we do not have a simple formula for the sum. Thus, this formalism has its limitations.

We can now investigate more complicated operations, as follows:

### **3. COMPLEX MATRICES**

THEOREM 3.11. We have the following operations on the complex numbers, written in polar form, as above:

- (1) Inversion:  $(re^{it})^{-1} = r^{-1}e^{-it}$ .
- (2) Square roots:  $\sqrt{re^{it}} = \pm \sqrt{r}e^{it/2}$ . (3) Powers:  $(re^{it})^a = r^a e^{ita}$ .

**PROOF.** This is something that we already know, from Theorem 3.6, but we can now discuss all this, from a more conceptual viewpoint, the idea being as follows:

(1) We have indeed the following computation, using Theorem 3.10:

$$(re^{it})(r^{-1}e^{-it}) = rr^{-1} \cdot e^{i(t-t)}$$
  
= 1 \cdot 1  
= 1

(2) Once again by using Theorem 3.10, we have:

$$(\pm \sqrt{r}e^{it/2})^2 = (\sqrt{r})^2 e^{i(t/2+t/2)}$$
  
=  $re^{it}$ 

(3) Given an arbitrary number  $a \in \mathbb{R}$ , we can define, as stated:

$$(re^{it})^a = r^a e^{ita}$$

Due to Theorem 3.10, this operation  $x \to x^a$  is indeed the correct one.

With the above results in hand, and notably with the square root formula from Theorem 3.11 (2), we can now go back to the degree 2 equations, and we have:

THEOREM 3.12. The complex solutions of  $ax^2 + bx + c = 0$  with  $a, b, c \in \mathbb{C}$  are

$$x_{1,2} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

with the square root of complex numbers being defined as above.

**PROOF.** This is clear, the computations being the same as in the real case. To be more precise, our degree 2 equation can be written as follows:

$$\left(x+\frac{b}{2a}\right)^2 = \frac{b^2-4aa}{4a^2}$$

Now since we know from Theorem 3.11 (2) that any complex number has a square root, we are led to the conclusion in the statement.

More generally now, we can prove that any polynomial equation, of arbitrary degree  $N \in \mathbb{N}$ , has exactly N complex solutions, counted with multiplicities:

64

### 3A. COMPLEX NUMBERS

THEOREM 3.13. Any polynomial  $P \in \mathbb{C}[X]$  decomposes as

$$P = c(X - a_1) \dots (X - a_N)$$

with  $c \in \mathbb{C}$  and with  $a_1, \ldots, a_N \in \mathbb{C}$ .

PROOF. The problem is that of proving that our polynomial has at least one root, because afterwards we can proceed by recurrence. We prove this by contradiction. So, assume that P has no roots, and pick a number  $z \in \mathbb{C}$  where |P| attains its minimum:

$$|P(z)| = \min_{x \in \mathbb{C}} |P(x)| > 0$$

Since Q(t) = P(z+t) - P(z) is a polynomial which vanishes at t = 0, this polynomial must be of the form  $ct^k$  + higher terms, with  $c \neq 0$ , and with  $k \geq 1$  being an integer. We obtain from this that, with  $t \in \mathbb{C}$  small, we have the following estimate:

$$P(z+t) \simeq P(z) + ct^k$$

Now let us write t = rw, with r > 0 small, and with |w| = 1. Our estimate becomes:

$$P(z+rw) \simeq P(z) + cr^k w^k$$

Now recall that we have assumed  $P(z) \neq 0$ . We can therefore choose  $w \in \mathbb{T}$  such that  $cw^k$  points in the opposite direction to that of P(z), and we obtain in this way:

$$|P(z+rw)| \simeq |P(z)+cr^kw^k|$$
  
= |P(z)|(1-|c|r^k)

Now by choosing r > 0 small enough, as for the error in the first estimate to be small, and overcame by the negative quantity  $-|c|r^k$ , we obtain from this:

$$|P(z+rw)| < |P(z)|$$

But this contradicts our definition of  $z \in \mathbb{C}$ , as a point where |P| attains its minimum. Thus P has a root, and by recurrence it has N roots, as stated.

All this is very nice, and we will see applications in a moment. As a word of warning, however, we should say that the above result remains something quite theoretical. Indeed, the proof is by contradiction, and there is no way of recycling the material there into something explicit, which can be used for effectively computing the roots.

As a last topic regarding the complex numbers, we have:

THEOREM 3.14. The equation  $x^N = 1$  has N complex solutions, namely

$$\left\{ w^k \middle| k = 0, 1, \dots, N-1 \right\}$$
,  $w = e^{2\pi i/N}$ 

which are called roots of unity of order N.

### 3. COMPLEX MATRICES

**PROOF.** This follows from Theorem 3.10. Indeed, with  $x = re^{it}$  our equation reads:

$$r^N e^{itN} = 1$$

Thus r = 1, and  $t \in [0, 2\pi)$  must be a multiple of  $2\pi/N$ , as stated.

As an illustration here, the roots of unity of small order, along with some of their basic properties, which are very useful for computations, are as follows:

N = 1. Here the unique root of unity is 1.

<u>N=2</u>. Here we have two roots of unity, namely 1 and -1.

<u>N = 3</u>. Here we have 1, then  $w = e^{2\pi i/3}$ , and then  $w^2 = \bar{w} = e^{4\pi i/3}$ .

<u>N=4</u>. Here the roots of unity, read as usual counterclockwise, are 1, i, -1, -i.

<u>N = 5</u>. Here, with  $w = e^{2\pi i/5}$ , the roots of unity are  $1, w, w^2, w^3, w^4$ .

<u>N = 6</u>. Here a useful alternative writing is  $\{\pm 1, \pm w, \pm w^2\}$ , with  $w = e^{2\pi i/3}$ .

The roots of unity are very useful variables, and have many interesting properties. As a first application, we can now solve the ambiguity questions related to the extraction of N-th roots, from Theorem 3.6 and Theorem 3.11, the statement being as follows:

THEOREM 3.15. Any nonzero complex number, written as

 $x = re^{it}$ 

has exactly N roots of order N, which appear as

$$y = r^{1/N} e^{it/N}$$

# multiplied by the N roots of unity of order N.

**PROOF.** We must solve the equation  $z^N = x$ , over the complex numbers. Since the number y in the statement clearly satisfies  $y^N = x$ , our equation is equivalent to:

$$z^N = y^N$$

Now observe that we can write this equation as follows:

$$\left(\frac{z}{y}\right)^N = 1$$

We conclude that the solutions z appear by multiplying y by the solutions of  $t^N = 1$ , which are the N-th roots of unity, as claimed.

The roots of unity appear in connection with many other interesting questions, and there are many useful formulae relating them, which are good to know. Here is a basic such formula, to be used many times in what follows:

THEOREM 3.16. The roots of unity,  $\{w^k\}$  with  $w = e^{2\pi i/N}$ , have the property

$$\sum_{k=0}^{N-1} (w^k)^s = N\delta_{N|s}$$

for any exponent  $s \in \mathbb{N}$ , where on the right we have a Kronecker symbol.

PROOF. The numbers in the statement, when written more conveniently as  $(w^s)^k$  with  $k = 0, \ldots, N-1$ , form a certain regular polygon in the plane  $P_s$ . Thus, if we denote by  $C_s$  the barycenter of this polygon, we have the following formula:

$$\frac{1}{N}\sum_{k=0}^{N-1} w^{ks} = C_s$$

Now observe that in the case N/s our polygon  $P_s$  is non-degenerate, circling around the unit circle, and having center  $C_s = 0$ . As for the case N|s, here the polygon is degenerate, lying at 1, and having center  $C_s = 1$ . Thus, we have the following formula:

$$C_s = \delta_{N|s}$$

Thus, we obtain the formula in the statement.

This was for our basic presentation of the theory of complex numbers. We will be back to more things regarding the complex numbers, and the roots of unity, later on.

### 3b. Linear maps

In relation now with linear algebra, our first task is that of extending the results that we know, from the real case, to the complex case. We first have:

THEOREM 3.17. The linear maps  $f : \mathbb{C}^N \to \mathbb{C}^M$  are the maps of the form

$$f(x) = Ax$$

with A being a rectangular matrix,  $A \in M_{M \times N}(\mathbb{C})$ .

PROOF. This follows as in the real case. Indeed,  $f : \mathbb{C}^N \to \mathbb{C}^M$  must send a vector  $x \in \mathbb{C}^N$  to a certain vector  $f(x) \in \mathbb{C}^M$ , all whose components are linear combinations of the components of x. Thus, we can write, for certain complex numbers  $a_{ij} \in \mathbb{C}$ :

$$f\begin{pmatrix}x_1\\\vdots\\\vdots\\x_N\end{pmatrix} = \begin{pmatrix}a_{11}x_1 + \dots + a_{1N}x_N\\\vdots\\\vdots\\a_{M1}x_1 + \dots + a_{MN}x_N\end{pmatrix}$$

The point now is that the parameters  $a_{ij} \in \mathbb{C}$  can be regarded as being the entries of a rectangular matrix, as follows:

$$A = (a_{ij}) \in M_{M \times N}(\mathbb{C})$$

### 3. COMPLEX MATRICES

With the usual convention for the rectangular matrix multiplication, exactly as in the real case, the above formula is precisely the one in the statement, namely:

$$f(x) = Ax$$

Thus, we are led to the conclusion in the statement.

We have as well the following result:

THEOREM 3.18. A linear map  $f : \mathbb{C}^N \to \mathbb{C}^M$ , written as

f(v) = Av

is invertible precisely when A is invertible, and in this case we have:

$$f^{-1}(v) = A^{-1}v$$

**PROOF.** As in the real case, with the convention  $f_A(v) = Av$ , we have the following multiplication formula for such linear maps:

$$f_A f_B(v) = f_{AB}(v)$$

But this shows that  $f_A f_B = 1$  is equivalent to AB = 1, as desired.

With respect to the real case, some subtleties appear at the level of the scalar products, isometries and projections. The basic theory here is as follows:

THEOREM 3.19. Consider the usual scalar product on  $\mathbb{C}^N$ , namely:

$$\langle x, y \rangle = \sum_{i} x_i \bar{y}_i$$

(1) We have the following formula, where  $(A^*)_{ij} = \overline{A}_{ji}$  is the adjoint matrix:

 $\langle Ax, y \rangle = \langle x, A^*y \rangle$ 

(2) A linear map  $f : \mathbb{C}^N \to \mathbb{C}^N$ , written as f(x) = Ux with  $U \in M_N(\mathbb{C})$ , is an isometry precisely when U is unitary, in the sense that:

$$U^* = U^{-1}$$

(3) A linear map  $f : \mathbb{C}^N \to \mathbb{C}^N$ , written as f(x) = Px with  $P \in M_N(\mathbb{C})$ , is a porojection precisely when P is projection, in the sense that:

$$P = P^2 = P^*$$

(4) The formula for the rank 1 projections is as follows:

$$P_x = \frac{1}{||x||^2} (x_i \bar{x}_j)_{ij}$$

68

### **3B. LINEAR MAPS**

**PROOF.** We follow the study from the real case, from chapter 1 above, by performing modifications where needed:

(1) By linearity, it is enough to prove the formula in the statement on the standard basis vectors  $e_1, \ldots, e_N$  of  $\mathbb{C}^N$ . Thus, we want to prove that for any i, j we have:

$$\langle Ae_j, e_i \rangle = \langle e_j, A^*e_i \rangle$$

The scalar product being now antisymmetric, this is the same as proving that:

$$\langle Ae_j, e_i \rangle = \overline{\langle A^*e_i, e_j \rangle}$$

On the other hand, for any matrix M we have the following formula:

$$M_{ij} = \langle Me_j, e_i \rangle$$

Thus, the formula to be proved simply reads:

$$A_{ij} = \overline{(A^*)_{ji}}$$

But this is precisely the definition of  $A^*$ , and we are done.

(2) Let first recall that we can pass from scalar products to distances, as follows:

$$|x|| = \sqrt{\langle x, x \rangle}$$

Conversely, we can compute the scalar products in terms of distances, by using the complex polarization identity, which is as follows:

$$\begin{aligned} ||x+y||^2 - ||x-y||^2 + i||x+iy||^2 - i||x-iy||^2 \\ &= ||x||^2 + ||y||^2 - ||x||^2 - ||y||^2 + i||x||^2 + i||y||^2 - i||x||^2 - i||y||^2 \\ &+ 2Re(< x, y >) + 2Re(< x, y >) + 2iIm(< x, y >) + 2iIm(< x, y >) \\ &= 4 < x, y > \end{aligned}$$

Now given a matrix  $U \in M_N(\mathbb{C})$ , we have the following equivalences, with the first one coming from the above identities, and with the other ones being clear:

$$\begin{split} ||Ux|| &= ||x|| &\iff < Ux, Uy > = < x, y > \\ &\iff < x, U^*Uy > = < x, y > \\ &\iff U^*Uy = y \\ &\iff U^*U = 1 \\ &\iff U^* = U^{-1} \end{split}$$

(3) As in the real case, P is an abstract projection, not necessarily orthogonal, when  $P^2 = P$ . The point now is that this projection is orthogonal when:

$$\langle Px - Py, Px - x \rangle = 0 \iff \langle x - y, P^*Px - P^*x \rangle = 0$$
  
 $\iff P^*Px - P^*x = 0$   
 $\iff P^*P - P^* = 0$ 

## 3. COMPLEX MATRICES

Thus we must have  $P^* = P^*P$ . Now observe that by conjugating, we obtain:

$$P = (P^*P)^*$$
  
=  $P^*(P^*)^*$   
=  $P^*P$ 

Now by comparing with the original relation,  $P^* = P^*P$ , we conclude that  $P = P^*$ . Thus, we have shown that any orthogonal projection must satisfy, as claimed:

$$P^2 = P = P^*$$

Conversely, if this condition is satisfied,  $P^2 = P$  shows that P is a projection, and  $P = P^*$  shows via the above computation that P is indeed orthogonal.

(4) Once again in analogy with the real case, we have the following formula:

$$P_x y = \frac{\langle y, x \rangle}{\langle x, x \rangle} x$$
$$= \frac{1}{||x||^2} \langle y, x \rangle x$$

With this in hand, we can now compute the entries of  $P_x$ , as follows:

$$(P_x)_{ij} = \langle P_x e_j, e_i \rangle \\ = \frac{1}{||x||^2} \langle e_j, x \rangle \langle x, e_i \rangle \\ = \frac{\bar{x}_j x_i}{||x||^2}$$

Thus, we are led to the formula in the statement.

## **3c.** Diagonalization

In the present complex matrix setting, we can talk as well about eigenvalues and eigenvectors, exactly as in the real case, as follows:

DEFINITION 3.20. Let  $A \in M_N(\mathbb{C})$  be a square matrix. When  $Av = \lambda v$  we say that:

(1) v is an eigenvector of A.

(2)  $\lambda$  is an eigenvalue of A.

We say that A is diagonalizable when  $\mathbb{C}^N$  has a basis of eigenvectors of A.

When A is diagonalizable, in that basis of eigenvectors we can write:

$$A = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_N \end{pmatrix}$$

### **3C. DIAGONALIZATION**

In general, this means that we have a formula as follows, with D diagonal:

$$A = PDP^{-1}$$

Indeed, we can take P to be the matrix formed by the eigenvectors:

$$P = [v_1 \dots v_N]$$

We have many basic examples here, similar to those in the real case, and we refer here to the various elementary computations from chapters 1-2 above.

As a first interesting result now, regarding the real matrices, we have:

THEOREM 3.21. The eigenvalues of a real matrix  $A \in M_N(\mathbb{R})$  are the roots of the characteristic polynomial, given by:

$$P(x) = \det(A - x1_N)$$

In particular, any such matrix  $A \in M_N(\mathbb{R})$  has at least 1 complex eigenvalue.

**PROOF.** The first assertion is something that we already know from chapter 2 above, which follows from the following computation:

$$\exists v, Av = \lambda v \iff \exists v, (A - \lambda 1_N)v = 0$$
$$\iff \det(A - \lambda 1_N) = 0$$

As for the second assertion, this follows from the first assertion, and from Theorem 3.13 above, which shows in particular that P has at least 1 complex root.

We will see in a moment, once the determinant of the complex matrices constructed, that the above result holds in fact for all the complex matrices  $A \in M_N(\mathbb{C})$ .

It is possible to further build on these results, but this is quite long, and we will do this in detail in the next chapter. For the moment, let us keep in mind the conclusion that a real matrix  $A \in M_N(\mathbb{R})$  has substantially more chances of being diagonalizable over the complex numbers, than over the real numbers.

As an illustration for this principle, and as a first concrete result, which is of true complex nature, we can now diagonalize the rotations in the real plane, as follows:

THEOREM 3.22. The rotation of angle  $t \in \mathbb{R}$  in the real plane, namely

$$R_t = \begin{pmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{pmatrix}$$

can be diagonalized over the complex numbers, as follows:

$$R_t = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ i & -i \end{pmatrix} \begin{pmatrix} e^{-it} & 0 \\ 0 & e^{it} \end{pmatrix} \begin{pmatrix} 1 & -i \\ 1 & i \end{pmatrix}$$

Over the real numbers this is impossible, unless  $t = 0, \pi$ .
## 3. COMPLEX MATRICES

**PROOF.** The last assertion is something clear, that we already know, coming from the fact that at  $t \neq 0, \pi$  our rotation is a "true" rotation, having no eigenvectors in the plane. Regarding the first assertion, the point is that we have the following computation:

$$R_t \begin{pmatrix} 1 \\ i \end{pmatrix} = \begin{pmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{pmatrix} \begin{pmatrix} 1 \\ i \end{pmatrix}$$
$$= \begin{pmatrix} \cos t - i \sin t \\ \sin t + i \cos t \end{pmatrix}$$
$$= \begin{pmatrix} \cos t - i \sin t \\ i \cos t + \sin t \end{pmatrix}$$
$$= e^{-it} \begin{pmatrix} 1 \\ i \end{pmatrix}$$

We have as well a second eigenvector, as follows:

$$R_t \begin{pmatrix} 1 \\ -i \end{pmatrix} = \begin{pmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{pmatrix} \begin{pmatrix} 1 \\ -i \end{pmatrix}$$
$$= \begin{pmatrix} \cos t + i \sin t \\ \sin t - i \cos t \end{pmatrix}$$
$$= \begin{pmatrix} \cos t + i \sin t \\ -i \cos t + \sin t \end{pmatrix}$$
$$= e^{it} \begin{pmatrix} 1 \\ -i \end{pmatrix}$$

Thus our matrix  $R_t$  is diagonalizable over  $\mathbb{C}$ , with the diagonal form being:

$$R_t \sim \begin{pmatrix} e^{-it} & 0\\ 0 & e^{it} \end{pmatrix}$$

As for the passage matrix, obtained by putting together the eigenvectors, this is:

$$P = \begin{pmatrix} 1 & 1\\ i & -i \end{pmatrix}$$

In order to invert now P, we can use the standard inversion formula for the  $2 \times 2$  matrices, which is the same as the one in the real case, and which gives:

$$P^{-1} = \frac{1}{-2i} \begin{pmatrix} -i & -1 \\ -i & 1 \end{pmatrix}$$
$$= \frac{i}{2} \begin{pmatrix} -i & -1 \\ -i & 1 \end{pmatrix}$$
$$= \frac{1}{2} \begin{pmatrix} 1 & -i \\ 1 & i \end{pmatrix}$$

#### **3D. THE DETERMINANT**

Our diagonalization formula is therefore as follows:

$$R_t = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ i & -i \end{pmatrix} \begin{pmatrix} e^{-it} & 0 \\ 0 & e^{it} \end{pmatrix} \begin{pmatrix} 1 & -i \\ 1 & i \end{pmatrix}$$

Thus, we are led to the conclusion in the statement.

We will be back to the diagonalization question for rotations and for other isometries later, after developing the theory of the determinant.

## 3d. The determinant

Regarding now the determinant, for the complex matrices it is more convenient to follow an abstract approach, and this due to our lack of geometric intuition with the space  $\mathbb{C}^N$ , at  $N \geq 2$ , and with the volumes of the bodies there.

Technically speaking, the best is to start with the following definition:

DEFINITION 3.23. The determinant of a complex matrix  $A \in M_N(\mathbb{C})$  is given by

$$\det A = \sum_{\sigma \in S_N} \varepsilon(\sigma) A_{1\sigma(1)} \dots A_{N\sigma(N)}$$

with  $\varepsilon = \pm 1$  being the signature of the permutations.

Generally speaking, the theory of the determinant from the real case extends well, except of course for everything related to its interpretation, as a volume.

To be more precise, we first have the following result, summarizing the needed properties and formulae of the determinant of the complex matrices:

**THEOREM** 3.24. The determinant has the following properties:

(1) When adding two columns, the determinants get added:

 $\det(\ldots, u+v, \ldots) = \det(\ldots, u, \ldots) + \det(\ldots, v, \ldots)$ 

(2) When multiplying columns by scalars, the determinant gets multiplied:

 $\det(\lambda v_1,\ldots,\lambda_N v_N) = \lambda_1\ldots\lambda_N\det(v_1,\ldots,v_N)$ 

(3) When permuting two columns, the determinant changes the sign:

 $\det(\dots, v, \dots, w, \dots) = -\det(\dots, w, \dots, v, \dots)$ 

PROOF. This follows indeed by doing some elementary algebraic computations with permutations, which are similar to those in the real case, but now done backwards, based on the formula of the determinant from Definition 3.23 above.

We have as well a similar result for the rows, which completes the list of needed operations which are usually needed in practice, as follows:

## 3. COMPLEX MATRICES

THEOREM 3.25. The determinant has the following properties:

(1) When adding two rows, the determinants get added:

$$\det \begin{pmatrix} \vdots \\ u+v \\ \vdots \end{pmatrix} = \det \begin{pmatrix} \vdots \\ u \\ \vdots \end{pmatrix} + \det \begin{pmatrix} \vdots \\ v \\ \vdots \end{pmatrix}$$

(2) When multiplying rows by scalars, the determinant gets multiplied:

$$\det \begin{pmatrix} \lambda_1 v_1 \\ \vdots \\ \lambda_N v_N \end{pmatrix} = \lambda_1 \dots \lambda_N \det \begin{pmatrix} v_1 \\ \vdots \\ v_N \end{pmatrix}$$

(3) When permuting two rows, the determinant changes the sign.

**PROOF.** This follows once again by doing some algebraic computations with permutations, based on the formula of the determinant from Definition 3.23 above.  $\Box$ 

Next in line, we have the following result, which is very useful in practice:

THEOREM 3.26. The determinant is subject to the row expansion formula

and this method fully computes it, by recurrence.

**PROOF.** This follows indeed by doing some algebraic computations.

We can expand as well over the columns, as follows:

74

THEOREM 3.27. The determinant is subject to the column expansion formula

$$\begin{vmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N1} & \dots & a_{NN} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & \dots & a_{2N} \\ \vdots & & \vdots \\ a_{N2} & \dots & a_{NN} \end{vmatrix}$$
$$- a_{21} \begin{vmatrix} a_{12} & \dots & a_{1N} \\ a_{32} & \dots & a_{3N} \\ \vdots & & \vdots \\ a_{N2} & \dots & a_{NN} \end{vmatrix}$$
$$\vdots$$
$$+ (-1)^{N+1} a_{N1} \begin{vmatrix} a_{12} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N-1,2} & \dots & a_{N-1} \end{vmatrix}$$

and this method fully computes it, by recurrence.

**PROOF.** Once again, this follows by doing some algebraic computations.

Still in analogy with the real case, we have the following result:

THEOREM 3.28. The determinant of the systems of vectors

 $\det: \mathbb{C}^N \times \ldots \times \mathbb{C}^N \to \mathbb{C}$ 

is multilinear, alternate and unital, and unique with these properties.

PROOF. This is something that we know in the real case, and the proof in the complex case is similar, with the conditions in the statement corresponding to those in Theorem 3.24. It is possible to prove this result as well directly, by doing some abstract algebra.  $\Box$ 

Finally, once again at the general level, let us record the following result:

THEOREM 3.29. We have the following formulae,

$$\det \bar{A} = \overline{\det A}$$
$$\det A^t = \det A$$
$$\det A^* = \overline{\det A}$$

valid for any square matrix  $A \in M_N(\mathbb{C})$ .

**PROOF.** The first formula is clear from Definition 3.23, because when conjugating the entries of A, the determinant will get conjugated:

$$\det \bar{A} = \sum_{\sigma \in S_N} \varepsilon(\sigma) \overline{A_{1\sigma(1)} \dots A_{N\sigma(N)}}$$

N

## 3. COMPLEX MATRICES

The second formula follows as in the real case, as follows:

$$\det A^{t} = \sum_{\sigma \in S_{N}} \varepsilon(\sigma) (A^{t})_{1\sigma(1)} \dots (A^{t})_{N\sigma(N)}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma) A_{\sigma(1)1} \dots A_{\sigma(N)N}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma) A_{1\sigma^{-1}(1)} \dots A_{N\sigma^{-1}(N)}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma^{-1}) A_{1\sigma^{-1}(1)} \dots A_{N\sigma^{-1}(N)}$$

$$= \sum_{\sigma \in S_{N}} \varepsilon(\sigma) A_{1\sigma(1)} \dots A_{N\sigma(N)}$$

$$= \det A$$

As for the third formula, this follows from the first two formulae, by using:

$$\det A^* = \det A^t$$

Thus, we are led to the conclusions in the statement.

Summarizing, the theory from the real case extends well, and we have complex analogues of all results. As in the real case, as a main application of all this, we have:

THEOREM 3.30. The inverse of a square matrix, having nonzero determinant,

$$A = \begin{pmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N1} & \dots & a_{NN} \end{pmatrix}$$

is given by the following formula,

$$A^{-1} = \frac{1}{\det A} \begin{pmatrix} \det A^{(11)} & -\det A^{(21)} & \det A^{(31)} & \dots \\ -\det A^{(12)} & \det A^{(22)} & -\det A^{(32)} & \dots \\ \det A^{(13)} & -\det A^{(23)} & \det A^{(33)} & \dots \\ \vdots & \vdots & \vdots & \end{pmatrix}$$

where  $A^{(ij)}$  is the matrix A, with the *i*-th row and *j*-th column removed.

PROOF. This follows indeed by using the row expansion formula from Theorem 3.26, which in terms of the matrix  $A^{-1}$  in the statement reads  $AA^{-1} = 1$ .

The above result is very useful for solving systems of equations, and for many other questions, and we will discuss this later on.

As a final topic now, regarding the complex matrices, let us discuss some interesting examples of such matrices, which definitely do not exist in the real setting, and which are very useful, even in connection with real matrix questions. Let us start with:

DEFINITION 3.31. The Fourier matrix is as follows,

$$F_N = (w^{ij})_{ij}$$

with  $w = e^{2\pi i/N}$ , and with the convention that the indices are

$$i, j \in \{0, 1, \dots, N-1\}$$

and are usually taken modulo N.

Here the conventions regarding the indices are standard, and are there for various reasons, as for instance for having the first row and column consisting of 1 entries.

Observe that, in standard matrix form, and with the above conventions for the indices, starting at 0, the Fourier matrix is as follows:

$$F_N = \begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & w & w^2 & \dots & w^{N-1} \\ 1 & w^2 & w^4 & \dots & w^{2(N-1)} \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & w^{N-1} & w^{2(N-1)} & \dots & w^{(N-1)^2} \end{pmatrix}$$

Thus, what we have here is a Vandermonde matrix, in the sense of chapter 2 above, of very special type. Let us record as well the first few values of these matrices:

**PROPOSITION 3.32.** The second Fourier matrix is as follows:

$$F_2 = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

The third Fourier matrix is as follows, with  $w = e^{2\pi i/3}$ :

$$F_3 = \begin{pmatrix} 1 & 1 & 1 \\ 1 & w & w^2 \\ 1 & w^2 & w \end{pmatrix}$$

As for the fourth Fourier matrix, this is as follows:

$$F_4 = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{pmatrix}$$

**PROOF.** All these formulae are clear from definitions, with our usual convention for the indices of the Fourier matrices, from Definition 3.31 above.

#### 3. COMPLEX MATRICES

Our claim now is that the Fourier matrix can be used in order to solve a variety of linear algebra questions, via various powerful and unexpected tricks, a bit in a same way as the Fourier transform can be used in order to solve analysis questions.

Before discussing all this, however, let us analyze the Fourier matrix  $F_N$ , from a linear algebra perspective. We have the following result:

THEOREM 3.33. The Fourier matrix  $F_N$  has the following properties:

- (1) It is symmetric,  $F_N^t = F_N$ .
- (2) The matrix  $F_N/\sqrt{N}$  is unitary.
- (3) Its inverse is the matrix  $F_N^*/N$ .

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**PROOF.** This is a collection of elementary results, the idea being as follows:

(1) This is clear from definitions.

(2) The row vectors  $R_0, \ldots, R_{N-1}$  of the rescaled matrix  $F_N/\sqrt{N}$  have all length 1, and by using the barycenter formula in Theorem 3.16, we have, for any  $i \neq j$ :

$$\langle R_i, R_j \rangle = \frac{1}{N} \sum_k w^{ik} w^{-jk}$$
  
 $= \frac{1}{N} \sum_k (w^{i-j})^k$   
 $= 0$ 

Thus,  $R_0, \ldots, R_{N-1}$  are pairwise orthogonal, and so  $F_N/\sqrt{N}$  is unitary, as claimed.

(3) This follows from (1) and (2), because for a symmetric matrix, the adjoint is the conjugate, and in the unitary case, this is the inverse.  $\Box$ 

At a more advanced level,  $F_N$  is a Vandermonde matrix, in the sense of chapter 2 above, so we know how to compute its determinant. It is equally possible to compute eigenvalues, eigenvectors, and to explicitly diagonalize  $F_N$ , but all this gradually becomes fairly complicated. We will stop here, and recommend all this as an exercise.

Now back to our motivations, we were saying before that the Fourier matrix is to linear algebra what the Fourier transform is to analysis, namely advanced technology. In order to discuss now an illustrating application of the theory developed above, let us go back to our favorite example of a  $N \times N$  matrix, namely the flat matrix:

$$\mathbb{I}_N = \begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix}$$

#### **3E. EXERCISES**

This is a real matrix, and we know that we have  $\mathbb{I}_N = NP_N$ , with  $P_N$  being the projection on the all-1 vector  $\xi = (1)_i \in \mathbb{R}^N$ . Thus,  $\mathbb{I}_N$  diagonalizes over  $\mathbb{R}$ :

$$\mathbb{I}_N \sim \begin{pmatrix} N & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{pmatrix}$$

The problem, however, is that when looking for 0-eigenvectors, in order to have an explicit diagonalization formula, we must solve the following equation:

$$x_1 + \ldots + x_N = 0$$

And this is not an easy task, if our objective is that of finding a very nice and explicit basis for the space of solutions. To be more precise, if we want linearly independent vectors  $v_1, \ldots, v_{N-1} \in \mathbb{R}^N$ , each with components summing up to 0, and which are given by simple formulae, of type  $(v_i)_j$  = explicit function of i, j, we are in trouble.

Fortunately, complex numbers and Fourier analysis are there, and we have:

THEOREM 3.34. The flat matrix, namely

$$\mathbb{I}_N = \begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix}$$

has the following explicit diagonalization, over the complex numbers,

$$\mathbb{I}_N = \frac{1}{N} F_N Q F_N^*$$

with  $F_N = (w^{ij})_{ij}$  being the Fourier matrix, and with Q = diag(N, 0, ..., 0).

PROOF. Indeed, the 0-eigenvector problem discussed above can be solved explicitly over the complex numbers, with the solution  $(v_i)_j = w^{ij}$ , with  $w = e^{2\pi i/N}$ .

There are many other uses of  $F_N$ , along the same lines. We will be back to all this in chapter 8 below, with a complete discussion of the Fourier matrices, and of their generalizations, called complex Hadamard matrices.

#### **3e.** Exercises

As a first exercise, in relation with the complex numbers, we have:

EXERCISE 3.35. Try to use a complex numbers type idea in order to deal with the vectors of  $\mathbb{R}^3$ , then  $\mathbb{R}^4$ , and report on what you found.

#### 3. COMPLEX MATRICES

This is something quite tricky, and a piece of hint, do not worry if you find nothing interesting at N = 3. However, the N = 4 case is definitely worth some study.

As a second exercise, still in relation with the complex numbers, we have:

EXERCISE 3.36. Can you use complex numbers in order to explicitly find the roots of arbitrary degree 3 polynomials, a bit in the same way as in degree 2?

As an indication here, better ignore what comes out of an internet search on this question. Or rather read that stuff, then try applying it to some concrete degree 3 polynomial, see what you get, if you are satisfied with the answer or not.

Still in relation with the complex numbers, and their philosophy, we have:

EXERCISE 3.37. Write down a complete proof for the formula

 $e^{\pi i} = -1$ 

using any method of your choice.

This is something that we talked about in the above, with the proof however still missing a few details, regarding the basic properties of the function  $e^x$ . Thus, you can either recover all these details, or go with some other idea, of your choice.

As a first exercise now about the complex matrices, we have:

EXERCISE 3.38. Find a geometric interpretation of the formula

$$\begin{pmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ i & -i \end{pmatrix} \begin{pmatrix} e^{-it} & 0 \\ 0 & e^{it} \end{pmatrix} \begin{pmatrix} 1 & -i \\ 1 & i \end{pmatrix}$$

which diagonalizes the rotation of angle  $t \in \mathbb{R}$  in the real plane.

This is something quite tricky, and of course, enjoy.

As a second exercise about the complex matrices, we have:

EXERCISE 3.39. Develop a complete theory of diagonalization for the  $2 \times 2$  matrices, notably by deciding when exactly such a matrix is diagonalizable.

This is quite non-trivial, but all the needed ingredients are in the above.

As a final exercise now, in relation with the material at the end, we have:

EXERCISE 3.40. Work out all the details of the diagonalization formula

$$\mathbb{I}_N = \frac{1}{N} F_N Q F_N^*$$

with Q = diag(N, 0, ..., 0), and then try formulating a generalization of this.

Here the first question is standard, amounting in completing the proof that was given in the above. As for the second question, this is something more tricky.

# CHAPTER 4

# Diagonalization

## 4a. Linear spaces

In this chapter we discuss in detail the diagonalization question, with a number of advanced results, for the complex matrices  $A \in M_N(\mathbb{C})$ . Our techniques will apply in particular to the real case,  $A \in M_N(\mathbb{R})$ , and we will obtain a number of non-trivial results regarding the diagonalization of such matrices, over the complex numbers.

In order to obtain our results, the idea will be that of looking at the eigenvectors  $v \in \mathbb{C}^N$  corresponding to a given eigenvalue  $\lambda \in \mathbb{C}$ , taken altogether:

DEFINITION 4.1. Given a matrix  $A \in M_N(\mathbb{C})$ , its eigenspaces are given by:

$$E_{\lambda} = \left\{ v \in \mathbb{C}^{N} \middle| Av = \lambda v \right\}$$

That is,  $E_{\lambda}$  collects all the eigenvectors having a given eigenvalue  $\lambda \in \mathbb{C}$ .

As a first observation, these eingenspaces are stable under taking sums, and also under multiplying by scalars, and this due to the following trivial facts:

$$Av = \lambda v, Aw = \lambda w \implies A(v+w) = \lambda(v+w)$$
$$Av = \lambda v, \mu \in \mathbb{C} \implies A(\mu v) = \lambda(\mu v)$$

Abstractly speaking, these two properties tell us that the eigenspaces  $E_{\lambda}$  are linear subspaces of the vector space  $\mathbb{C}^{N}$ , in the following sense:

DEFINITION 4.2. A subset  $V \subset \mathbb{C}^N$  is called a linear space, or vector space, when:

$$v, w \in V \implies v + w \in V$$
  
 $\lambda \in \mathbb{C}, v \in V \implies \lambda v \in V$ 

That is, V must be stable under taking sums, and multiplying by scalars.

Before getting into the study of the eigenspaces of a given matrix  $A \in M_N(\mathbb{C})$ , let us develop some general theory for the arbitrary linear spaces  $V \subset \mathbb{C}^N$ .

As a first objective, we would like to talk about bases of such subspaces  $V \subset \mathbb{C}^N$ . We already know, from chapter 1 above, how to do this for  $V = \mathbb{C}^N$  itself. In order to review and improve the material there, let us start with the following key result:

**PROPOSITION 4.3.** For a matrix  $A \in M_N(\mathbb{C})$ , the following are equivalent:

- (1) A is right invertible:  $\exists B \in M_N(\mathbb{C}), AB = 1$ .
- (2) A is left invertible:  $\exists B \in M_N(\mathbb{C}), BA = 1.$
- (3) A is invertible:  $\exists B \in M_N(\mathbb{C}), AB = BA = 1.$
- (4) A has nonzero determinant: det  $A \neq 0$ .

**PROOF.** This follows indeed from the theory of the determinant, as developed in chapter 3 above, with all the conditions being equivalent to (4).  $\Box$ 

In terms of linear maps, we are led to the following statement:

**PROPOSITION 4.4.** For a linear map  $f : \mathbb{C}^N \to \mathbb{C}^N$ , the following are equivalent:

- (1) f is injective.
- (2) f is surjective.
- (3) f is bijective.
- (4) f(x) = Ax, with A invertible.

PROOF. This follows indeed from Proposition 4.3, by using the composition formula  $f_{AB} = f_A f_B$  for the linear maps  $\mathbb{C}^N \to \mathbb{C}^N$ , written in the form  $f_A(x) = Ax$ .

Getting now to the point where we wanted to get, we have:

THEOREM 4.5. For a system of vectors  $\{v_1, \ldots, v_N\} \subset \mathbb{C}^N$ , the following conditions are equivalent:

(1) The vectors  $v_i$  are linearly independent, in the sense that:

$$\sum \lambda_i v_i = 0 \implies \lambda_i = 0$$

(2) The vectors  $v_i$  span  $\mathbb{C}^N$ , in the sense that each  $x \in \mathbb{C}^N$  can be written as:

$$x = \sum \lambda_i v_i$$

(3) The vectors  $v_i$  form a basis of  $\mathbb{C}^N$ , in the sense that each vector  $x \in \mathbb{C}^N$  can be written in a unique way as:

$$x = \sum \lambda_i v_i$$

(4) The matrix formed by these vectors, regarded as usual as column vectors,

$$P = [v_1, \ldots, v_N] \in M_N(\mathbb{C})$$

is invertible, with respect to the usual multiplication of the matrices.

**PROOF.** Given a family of vectors  $\{v_1, \ldots, v_N\} \subset \mathbb{C}^N$  as in the statement, consider the following linear map, associated to these vectors:

$$f: \mathbb{C}^N \to \mathbb{C}^N$$
$$\lambda \to \sum \lambda_i v_i$$

The conditions (1,2,3) in the statement tell us that this map must be respectively injective, surjective, bijective. As for the condition (4) in the statement, this is related as well to this map, because the matrix of this map is the matrix P there:

$$f(\lambda) = P\lambda$$

With these observations in hand, the result follows from Proposition 4.4.

More generally now, we have the following result:

THEOREM 4.6. For a system of vectors  $\{v_1, \ldots, v_M\} \subset V$  belonging to a linear subspace  $V \subset \mathbb{C}^N$ , the following conditions are equivalent:

(1) The vectors  $v_i$  are linearly independent, in the sense that:

$$\sum \lambda_i v_i = 0 \implies \lambda_i = 0$$

(2) The vectors  $v_i$  span V, in the sense that each  $x \in V$  can be written as:

$$x = \sum \lambda_i v_i$$

(3) The vectors  $v_i$  form a basis of V, in the sense that each vector  $x \in V$  can be written in a unique way as:

$$x = \sum \lambda_i v_i$$

If these conditions are satisfied, we say that V has dimension M, and this dimension is independent on the choice of the basis.

PROOF. This is something that we know for  $V = \mathbb{C}^N$  itself, coming from Theorem 4.5 above, with the dimension discussion at the end being something elementary. As for the extension to the general case, involving an arbitrary linear space  $V \subset \mathbb{C}^N$ , this is something which is routine, by using a few recurrence arguments, where needed.

# 4b. Diagonalization

Let us begin with a reminder of the basic diagonalization theory, that we already know. The basic theory that we have so far can be summarized as follows:

THEOREM 4.7. Assuming that a matrix  $A \in M_N(\mathbb{C})$  is diagonalizable, in the sense that  $\mathbb{C}^N$  has a basis formed by eigenvectors of A, we have

$$A = PDP^{-1}$$

where  $P = [v_1 \dots v_N]$  is the square matrix formed by the eigenvectors of A, and  $D = diag(\lambda_1, \dots, \lambda_N)$  is the diagonal matrix formed by the corresponding eigenvalues.

PROOF. This is something that we already know, coming by changing the basis. We can prove this by direct computation as well, because we have  $Pe_i = v_i$ , and so the matrices A and  $PDP^{-1}$  follow to act in the same way on the basis vectors  $v_i$ :

$$PDP^{-1}v_i = PDe_i$$
$$= P\lambda_i e_i$$
$$= \lambda_i Pe_i$$
$$= \lambda_i v_i$$

Thus, the matrices A and  $PDP^{-1}$  coincide, as stated.

In general, in order to study the diagonalization problem, the idea is that the eigenvectors can be grouped into linear spaces, called eigenspaces, as follows:

THEOREM 4.8. Let  $A \in M_N(\mathbb{C})$ , and for any eigenvalue  $\lambda \in \mathbb{C}$  define the corresponding eigenspace as being the vector space formed by the corresponding eigenvectors:

$$E_{\lambda} = \left\{ v \in \mathbb{C}^{N} \middle| Av = \lambda v \right\}$$

These eigenspaces  $E_{\lambda}$  are then in a direct sum position, in the sense that given vectors  $v_1 \in E_{\lambda_1}, \ldots, v_k \in E_{\lambda_k}$  corresponding to different eigenvalues  $\lambda_1, \ldots, \lambda_k$ , we have:

$$\sum_{i} c_i v_i = 0 \implies c_i = 0$$

In particular, we have the following dimension inequality, with the sum being over all the eigenvalues  $\lambda \in \mathbb{C}$  of our matrix A,

$$\sum_{\lambda} \dim(E_{\lambda}) \le N$$

and our matrix is diagonalizable precisely when we have equality.

**PROOF.** We prove the first assertion by recurrence on  $k \in \mathbb{N}$ . Assume by contradiction that we have a formula as follows, with the scalars  $c_1, \ldots, c_k$  being not all zero:

$$c_1v_1 + \ldots + c_kv_k = 0$$

By dividing by one of these scalars, we can assume that our formula is:

$$v_k = c_1 v_1 + \ldots + c_{k-1} v_{k-1}$$

Now let us apply A to this vector. On the left we obtain:

$$Av_k = \lambda_k v_k$$
  
=  $\lambda_k c_1 v_1 + \ldots + \lambda_k c_{k-1} v_{k-1}$ 

84

On the right we obtain something different, as follows:

$$A(c_1v_1 + \ldots + c_{k-1}v_{k-1}) = c_1Av_1 + \ldots + c_{k-1}Av_{k-1}$$
  
=  $c_1\lambda_1v_1 + \ldots + c_{k-1}\lambda_{k-1}v_{k-1}$ 

We conclude from this that the following equality must hold:

$$\lambda_k c_1 v_1 + \ldots + \lambda_k c_{k-1} v_{k-1} = c_1 \lambda_1 v_1 + \ldots + c_{k-1} \lambda_{k-1} v_{k-1}$$

On the other hand, we know by recurrence that the vectors  $v_1, \ldots, v_{k-1}$  must be linearly independent. Thus, the coefficients must be equal, at right and at left:

$$\lambda_k c_1 = c_1 \lambda_1$$
$$\vdots$$

$$\lambda_k c_{k-1} = c_{k-1} \lambda_{k-1}$$

Now since at least one  $c_i$  must be nonzero, from  $\lambda_k c_i = c_i \lambda_i$  we obtain  $\lambda_k = \lambda_i$ , which is a contradiction. Thus our proof by recurrence of the first assertion is complete.

As for the second assertion, this follows from the first one.

In order to reach now to more advanced results, we can use the characteristic polynomial. Here is a result summarizing and improving our knowledge of the subject:

THEOREM 4.9. Given a matrix  $A \in M_N(\mathbb{C})$ , consider its characteristic polynomial:

$$P(x) = \det(A - x1_N)$$

The eigenvalues of A are then the roots of P. Also, we have the inequality

 $\dim(E_{\lambda}) \le m_{\lambda}$ 

where  $m_{\lambda}$  is the multiplicity of  $\lambda$ , as root of P.

**PROOF.** The first assertion follows from the following computation, using the fact that a linear map is bijective when the determinant of the associated matrix is nonzero:

$$\exists v, Av = \lambda v \iff \exists v, (A - \lambda 1_N)v = 0$$
$$\iff \det(A - \lambda 1_N) = 0$$

Regarding now the second assertion, given an eigenvalue  $\lambda$  of our matrix A, consider the dimension of the corresponding eigenspace:

$$d_{\lambda} = \dim(E_{\lambda})$$

By changing the basis of  $\mathbb{C}^N$ , as for the eigenspace  $E_{\lambda}$  to be spanned by the first  $d_{\lambda}$  basis elements, our matrix becomes as follows, with B being a certain smaller matrix:

$$A \sim \begin{pmatrix} \lambda 1_{d_{\lambda}} & 0\\ 0 & B \end{pmatrix}$$

We conclude that the characteristic polynomial of A is of the following form:

$$P_A = P_{\lambda 1_{d_\lambda}} P_B$$
$$= (\lambda - x)^{d_\lambda} P_B$$

Thus the multiplicity  $m_{\lambda}$  of our eigenvalue  $\lambda$ , as a root of P, satisfies:

 $m_{\lambda} \ge d_{\lambda}$ 

But this leads to the conclusion in the statement.

We can put together Theorem 4.8 and Theorem 4.9, and by using as well the fact that any complex polynomial of degree N has exactly N complex roots, when counted with multiplicities, that we know from chapter 3 above, we obtain the following result:

THEOREM 4.10. Given a matrix  $A \in M_N(\mathbb{C})$ , consider its characteristic polynomial  $P(X) = \det(A - X1_N)$ 

then factorize this polynomial, by computing the complex roots, with multiplicities,

$$P(X) = (-1)^N (X - \lambda_1)^{n_1} \dots (X - \lambda_k)^{n_k}$$

and finally compute the corresponding eigenspaces, for each eigenvalue found:

$$E_i = \left\{ v \in \mathbb{C}^N \middle| Av = \lambda_i v \right\}$$

The dimensions of these eigenspaces satisfy then the following inequalities,

$$\dim(E_i) \le n_i$$

and A is diagonalizable precisely when we have equality for any i.

**PROOF.** This follows indeed from Theorem 4.8 and Theorem 4.9, and from the abovementioned result regarding the complex roots of complex polynomials.  $\Box$ 

This was for the main result of linear algebra. There are countless applications of this, and generally speaking, advanced linear algebra consists in building on Theorem 4.10.

Let us record as well a useful algorithmic version of the above result:

THEOREM 4.11. The square matrices  $A \in M_N(\mathbb{C})$  can be diagonalized as follows:

- (1) Compute the characteristic polynomial.
- (2) Factorize the characteristic polynomial.
- (3) Compute the eigenvectors, for each eigenvalue found.
- (4) If there are no N eigenvectors, A is not diagonalizable.
- (5) Otherwise, A is diagonalizable,  $A = PDP^{-1}$ .

PROOF. This is an informal reformulation of Theorem 4.10 above, with (4) referring to the total number of linearly independent eigenvectors found in (3), and with  $A = PDP^{-1}$  in (5) being the usual diagonalization formula, with P, D being as before.

86

#### 4C. MATRIX TRICKS

As a remark here, in step (3) it is always better to start with the eigenvalues having big multiplicity. Indeed, a multiplicity 1 eigenvalue, for instance, can never lead to the end of the computation, via (4), simply because the eigenvectors always exist.

## 4c. Matrix tricks

At the level of basic examples of diagonalizable matrices, we first have:

THEOREM 4.12. For a matrix  $A \in M_N(\mathbb{C})$  the following conditions are equivalent,

- (1) The eigenvalues are different,  $\lambda_i \neq \lambda_j$ ,
- (2) The characteristic polynomial P has simple roots,
- (3) The characteristic polynomial satisfies (P, P') = 1,

and in this case, the matrix is diagonalizable.

**PROOF.** The equivalences in the statement are clear, the idea being as follows:

(1)  $\iff$  (2) This follows from Theorem 4.10.

(2)  $\iff$  (3) This is standard, the double roots of P being roots of P'.

Regarding now the last assertion, this follows from Theorem 4.10 as well. Indeed, the criterion there for diagonalization involving the sum of dimensions of eigenspaces is trivially satisfied when the eigenvalues are different, because:

$$\sum_{\lambda} \dim(E_{\lambda}) = \sum_{\lambda} 1 = N$$

Thus, we are led to the conclusion in the statement.

We will see in a moment that the matrices having distinct eigenvalues are "generic", from a certain analytic viewpoint, and so the above result basically says it all.

As another important comment, the assumptions of Theorem 4.12 can be effectively verified in practice, without the need for factorizing polynomials, the idea here being that of using the condition (3) there. In order to discuss this, let us start with:

THEOREM 4.13. Given two polynomials  $P, Q \in \mathbb{C}[X]$ , written as follows,

$$P = c(X - a_1) \dots (X - a_k)$$
$$Q = d(X - b_1) \dots (X - b_l)$$

the following quantity, which is called resultant of P, Q,

$$R(P,Q) = c^l d^k \prod_{ij} (a_i - b_j)$$

is a polynomial in the coefficients of P, Q, with integer coefficients, and we have

$$R(P,Q) = 0$$

precisely when P, Q have a common root.

**PROOF.** Given  $P, Q \in \mathbb{C}[X]$ , we can certainly construct the quantity R(P, Q) in the statement, and we have then R(P, Q) = 0 precisely when P, Q have a common root.

The whole point is that of proving that R(P, Q) is a polynomial in the coefficients of P, Q, with integer coefficients. But this can be checked as follows:

(1) We can expand the formula of R(P,Q), and in what regards  $a_1, \ldots, a_k$ , which are the roots of P, we obtain in this way certain symmetric functions in these variables, which will be therefore polynomials in the coefficients of P, with integer coefficients.

(2) We can then look what happens with respect to the remaining variables  $b_1, \ldots, b_l$ , which are the roots of Q. Once again what we have here are certain symmetric functions, and so polynomials in the coefficients of Q, with integer coefficients.

(3) Thus, we are led to the conclusion in the statement, that R(P,Q) is a polynomial in the coefficients of P, Q, with integer coefficients, and with the remark that the  $c^l d^k$ factor is there for these latter coefficients to be indeed integers, instead of rationals.  $\Box$ 

All this might seem a bit complicated, so as an illustration, let us work out an example. Consider the case of a polynomial of degree 2, and a polynomial of degree 1:

$$P = ax^2 + bx + c$$
$$Q = dx + e$$

In order to compute the resultant, let us factorize our polynomials:

$$P = a(x - p)(x - q)$$
$$Q = d(x - r)$$

The resultant can be then computed as follows, by using the method above:

$$R(P,Q) = ad^{2}(p-r)(q-r)$$
  
=  $ad^{2}(pq - (p+q)r + r^{2})$   
=  $cd^{2} + bd^{2}r + ad^{2}r^{2}$   
=  $cd^{2} - bde + ae^{2}$ 

Finally, observe that R(P,Q) = 0 corresponds indeed to the fact that P,Q have a common root. Indeed, the root of Q is r = -e/d, and we have:

$$P(r) = \frac{ae^2}{d^2} - \frac{be}{d} + c$$
$$= \frac{R(P,Q)}{d^2}$$

Thus P(r) = 0 precisely when R(P, Q) = 0, as predicted by Theorem 4.13.

Regarding now the explicit formula of R(P,Q), this is something quite complicated, and there are several methods for dealing with this problem.

#### 4C. MATRIX TRICKS

There is a slight similarity between Theorem 4.13 and the Vandermonde determinants discussed in chapter 2, and we have in fact the following formula for R(P,Q):

THEOREM 4.14. The resultant of two polynomials, written as

$$P = p_k X^k + \ldots + p_1 X + p_0$$
$$Q = q_l X^l + \ldots + q_1 X + q_0$$

appears as the determinant of an associated matrix, as follows,

$$R(P,Q) = \begin{vmatrix} p_k & q_l \\ \vdots & \ddots & \vdots & \ddots \\ p_0 & p_k & q_0 & q_k \\ & \ddots & \vdots & & \ddots & \vdots \\ & & p_0 & & q_0 \end{vmatrix}$$

with the matrix having size k + l, and having 0 coefficients at the blank spaces.

**PROOF.** This follows by doing some linear algebra computations, mixed with algebra, in the spirit of those from the proof of the Vandermonde determinant theorem.  $\Box$ 

In what follows we will not really need the above formula, so let us just check now that this formula works indeed. Consider our favorite polynomials, as before:

$$P = ax^2 + bx + c$$
$$Q = dx + e$$

According to the above result, the resultant should be then, as it should:

$$R(P,Q) = \begin{vmatrix} a & d & 0 \\ b & e & d \\ c & 0 & e \end{vmatrix}$$
$$= ae^{2} - bde + cd^{2}$$

Now back to our diagonalization questions, we want to compute R(P, P'), where P is the characteristic polynomial. So, we need one more piece of theory, as follows:

THEOREM 4.15. Given a polynomial  $P \in \mathbb{C}[X]$ , written as

$$P(X) = cX^N + dX^{N-1} + \dots$$

its discriminant, defined as being the following quantity,

$$\Delta(P) = \frac{(-1)^{\binom{N}{2}}}{c} R(P, P')$$

is a polynomial in the coefficients of P, with integer coefficients, and

$$\Delta(P) = 0$$

happens precisely when P has a double root.

**PROOF.** This follows from Theorem 4.13, applied with P = Q, with the division by c being indeed possible, under  $\mathbb{Z}$ , and with the sign being there for various reasons, including the compatibility with some well-known formulae, at small values of  $N \in \mathbb{N}$ .

Observe that, by using Theorem 4.14, we have an explicit formula for the discriminant, as the determinant of a certain associated matrix. There is a lot of theory here, and in order to terminate the discussion, let us see what happens in degree 2. Here we have:

$$P = aX^2 + bX + c$$
$$P' = 2aX + b$$

Thus, the resultant is given by the following formula:

$$R(P, P') = ab^{2} - b(2a)b + c(2a)^{2}$$
  
=  $4a^{2}c - ab^{2}$   
=  $-a(b^{2} - 4ac)$ 

With the normalizations in Theorem 4.15 made, we obtain, as we should:

$$\Delta(P) = b^2 - 4ac$$

Now back to our questions, we can upgrade Theorem 4.12, as follows:

THEOREM 4.16. For a matrix  $A \in M_N(\mathbb{C})$  the following conditions are equivalent,

- (1) The eigenvalues are different,  $\lambda_i \neq \lambda_j$ ,
- (2) The characteristic polynomial P has simple roots,
- (3) The discriminant of P is nonzero,  $\Delta(P) \neq 0$ ,

and in this case, the matrix is diagonalizable.

PROOF. This is indeed an upgrade of Theorem 4.12 above, by replacing the condition (3) there with the condition  $\Delta(P) \neq 0$ , which is something much better, because it is something algorithmic, ultimately requiring only computations of determinants.

As already mentioned before, one can prove that the matrices having distinct eigenvalues are "generic", and so the above result basically captures the whole situation. We have in fact the following collection of density results, all being very useful:

THEOREM 4.17. The following happen, inside  $M_N(\mathbb{C})$ :

- (1) The invertible matrices are dense.
- (2) The matrices having distinct eigenvalues are dense.
- (3) The diagonalizable matrices are dense.

**PROOF.** These are quite advanced linear algebra results, which can be proved as follows, with the technology that we have so far:

#### 4C. MATRIX TRICKS

(1) This is clear, intuitively speaking, because the invertible matrices are given by the condition det  $A \neq 0$ . Thus, the set formed by these matrices appears as the complement of the surface det A = 0, and so must be dense inside  $M_N(\mathbb{C})$ , as claimed.

(2) Here we can use a similar argument, this time by saying that the set formed by the matrices having distinct eigenvalues appears as the complement of the surface given by  $\Delta(P_A) = 0$ , and so must be dense inside  $M_N(\mathbb{C})$ , as claimed.

(3) This follows from (2), via the fact that the matrices having distinct eigenvalues are diagonalizable, that we know from Theorem 4.16 above. There are of course some other proofs as well, for instance by putting the matrix in Jordan form.  $\Box$ 

As an application of the above results, and of our methods in general, we can now establish a number of useful and interesting linear algebra results, as follows:

THEOREM 4.18. The following happen:

(1) We have  $P_{AB} = P_{BA}$ , for any two matrices  $A, B \in M_N(\mathbb{C})$ .

(2) AB, BA have the same eigenvalues, with the same multiplicities.

(3) If A has eigenvalues  $\lambda_1, \ldots, \lambda_N$ , then f(A) has eigenvalues  $f(\lambda_1), \ldots, f(\lambda_N)$ .

**PROOF.** These results can be deduced by using Theorem 4.17, as follows:

(1) It follows from definitions that the characteristic polynomial of a matrix is invariant under conjugation, in the sense that we have:

$$P_C = P_{ACA^{-1}}$$

Now observe that, when assuming that A is invertible, we have:

$$AB = A(BA)A^{-1}$$

Thus, we have the result when A is invertible. By using now Theorem 4.17 (1) above, we conclude that this formula holds for any matrix A, by continuity.

(2) This is a reformulation of (1) above, via the fact that P encodes the eigenvalues, with multiplicities, which is hard to prove with bare hands.

(3) This is something more informal, the idea being that this is clear for the diagonal matrices D, then for the diagonalizable matrices  $PDP^{-1}$ , and finally for all the matrices, by using Theorem 4.17 (3), provided that f has suitable regularity properties.

We will be back to the above results, with more details, at the end of the present chapter, and then in chapter 7 below, when doing spectral theory.

Getting away now from all this advanced mathematics, let us go back to the main problem raised by the diagonalization procedure, namely the computation of the roots of characteristic polynomials. As a first trivial observation, in degree 2 we have:

**PROPOSITION 4.19.** The roots of a degree 2 polynomial of the form

$$P = X^2 - aX + b$$

are precisely the numbers r, s satisfying r + s = a, rs = b.

**PROOF.** This is trivial, coming from P = (X - r)(X - s). In practice, this is very useful, and this is how calculus teachers are often faster than their students.

In the matrix setting now, the result is as follows:

**PROPOSITION 4.20.** The eigenvalues of a square matrix

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

are precisely the numbers r, s satisfying r + s = a + d, rs = ad - bc.

PROOF. Once again, this is something trivial, coming from Proposition 4.19. As before, this is a well-known calculus teacher trick.  $\Box$ 

Finally, we have the following result, which is useful as well:

**PROPOSITION** 4.21. Assume that we have a polynomial as follows, with integer coefficients, and with the leading term being 1:

$$P = X^{N} + a_{N-1}X^{N-1} + \ldots + a_{1}X + a_{0}$$

The integer roots of P must then divide the last coefficient  $a_0$ .

**PROOF.** This is clear, because any integer root  $c \in \mathbb{Z}$  of our polynomial must satisfy:

$$c^{N} + a_{N-1}c^{N-1} + \ldots + a_{1}c + a_{0} = 0$$

But modulo c, this equation simply reads  $a_0 = 0$ , as desired.

Getting now into more conceptual mathematics, the two formulae from Proposition 4.20 above are quite interesting, and can be generalized as follows:

THEOREM 4.22. The complex eigenvalues of a matrix  $A \in M_N(\mathbb{C})$ , counted with multiplicities, have the following properties:

- (1) Their sum is the trace.
- (2) Their product is the determinant.

**PROOF.** Consider indeed the characteristic polynomial P of the matrix:

$$P(X) = \det(A - X1_N)$$
  
=  $(-1)^N X^N + (-1)^{N-1} Tr(A) X^{N-1} + \dots + \det(A)$ 

We can factorize this polynomial, by using its N complex roots, and we obtain:

$$P(X) = (-1)^{N} (X - \lambda_{1}) \dots (X - \lambda_{N})$$
  
=  $(-1)^{N} X^{N} + (-1)^{N-1} \left(\sum_{i} \lambda_{i}\right) X^{N-1} + \dots + \prod_{i} \lambda_{i}$ 

Thus, we are led to the conclusion in the statement.

Regarding now the intermediate terms, we have here:

THEOREM 4.23. Assume that  $A \in M_N(\mathbb{C})$  has eigenvalues  $\lambda_1, \ldots, \lambda_N \in \mathbb{C}$ , counted with multiplicities. The basic symmetric functions of these eigenvalues, namely

$$c_k = \sum_{i_1 < \ldots < i_k} \lambda_{i_1} \ldots \lambda_{i_k}$$

are then given by the fact that the characteristic polynomial of the matrix is:

$$P(X) = (-1)^N \sum_{k=0}^N (-1)^k c_k X^k$$

Moreover, all symmetric functions of the eigenvalues, such as the sums of powers

 $d_s = \lambda_1^s + \ldots + \lambda_N^s$ 

appear as polynomials in these characteristic polynomial coefficients  $c_k$ .

**PROOF.** These results can be proved by doing some algebra, as follows:

(1) Consider indeed the characteristic polynomial P of the matrix, factorized by using its N complex roots, taken with multiplicities. By expanding, we obtain:

$$P(X) = (-1)^{N} (X - \lambda_{1}) \dots (X - \lambda_{N})$$
  
=  $(-1)^{N} X^{N} + (-1)^{N-1} \left( \sum_{i} \lambda_{i} \right) X^{N-1} + \dots + \prod_{i} \lambda_{i}$   
=  $(-1)^{N} X^{N} + (-1)^{N-1} c_{1} X^{N-1} + \dots + (-1)^{0} c_{N}$   
=  $(-1)^{N} \left( X^{N} - c_{1} X^{N-1} + \dots + (-1)^{N} c_{N} \right)$ 

With the convention  $c_0 = 1$ , we are led to the conclusion in the statement.

(2) This is something standard, coming by doing some abstract algebra. Working out the formulae for the sums of powers  $d_s = \sum_i \lambda_i^s$ , at small values of the exponent  $s \in \mathbb{N}$ , is an excellent exercise, which shows how to proceed in general, by recurrence.

# 4d. Spectral theorems

Let us go back now to the diagonalization question. Here is a key result:

THEOREM 4.24. Any matrix  $A \in M_N(\mathbb{C})$  which is self-adjoint,  $A = A^*$ , is diagonalizable, with the diagonalization being of the following type,

$$A = UDU^*$$

with  $U \in U_N$ , and with  $D \in M_N(\mathbb{R})$  diagonal. The converse holds too.

PROOF. As a first remark, the converse trivially holds, because if we take a matrix of the form  $A = UDU^*$ , with U unitary and D diagonal and real, then we have:

$$A^* = (UDU^*)^*$$
$$= UD^*U^*$$
$$= UDU^*$$
$$= A$$

In the other sense now, assume that A is self-adjoint,  $A = A^*$ . Our first claim is that the eigenvalues are real. Indeed, assuming  $Av = \lambda v$ , we have:

$$\lambda < v, v > = < \lambda v, v >$$

$$= < Av, v >$$

$$= < Av, v >$$

$$= < v, Av >$$

$$= < v, \lambda v >$$

$$= \overline{\lambda} < v, v >$$

Thus we obtain  $\lambda \in \mathbb{R}$ , as claimed. Our next claim now is that the eigenspaces corresponding to different eigenvalues are pairwise orthogonal. Assume indeed that:

$$Av = \lambda v$$
 ,  $Aw = \mu w$ 

We have then the following computation, using  $\lambda, \mu \in \mathbb{R}$ :

Thus  $\lambda \neq \mu$  implies  $v \perp w$ , as claimed. In order now to finish, it remains to prove that the eigenspaces span  $\mathbb{C}^N$ . For this purpose, we will use a recurrence method. Let us

pick an eigenvector,  $Av = \lambda v$ . Assuming  $v \perp w$ , we have:

$$\langle Aw, v \rangle = \langle w, Av \rangle$$
  
=  $\langle w, \lambda v \rangle$   
=  $\lambda \langle w, v \rangle$   
=  $0$ 

Thus, if v is an eigenvector, then the vector space  $v^{\perp}$  is invariant under A. In order to do the recurrence, it still remains to prove that the restriction of A to the vector space  $v^{\perp}$  is self-adjoint. But this comes from a general property of the self-adjoint matrices, that we will explain now. Our claim is that an arbitrary square matrix A is self-adjoint precisely when the following happens, for any vector v:

$$\langle Av, v \rangle \in \mathbb{R}$$

Indeed, the fact that the above scalar product is real is equivalent to:

$$\langle (A - A^*)v, v \rangle = 0$$

But this is equivalent to  $A = A^*$ , by using the complex polarization identity. Now back to our questions, it is clear from our self-adjointness criterion above that the restriction of A to any invariant subspace, and in particular to the subspace  $v^{\perp}$ , is self-adjoint. Thus, we can proceed by recurrence, and we obtain the result.  $\Box$ 

Let us record as well the real version of the above result:

THEOREM 4.25. Any matrix  $A \in M_N(\mathbb{R})$  which is symmetric, in the sense that

$$A = A^t$$

is diagonalizable, with the diagonalization being of the following type,

$$A = UDU^t$$

with  $U \in O_N$ , and with  $D \in M_N(\mathbb{R})$  diagonal. The converse holds too.

**PROOF.** As before, the converse trivially holds, because if we take a matrix of the form  $A = UDU^t$ , with U orthogonal and D diagonal and real, then we have:

$$A^{t} = (UDU^{t})^{t}$$
$$= UD^{t}U^{t}$$
$$= UDU^{t}$$
$$= A$$

In the other sense now, this follows from Theorem 4.24, and its proof.

As basic examples of self-adjoint matrices, we have the orthogonal projections. The diagonalization result regarding them is as follows:

PROPOSITION 4.26. The matrices  $P \in M_N(\mathbb{C})$  which are projections,  $P^2 = P = P^*$ , are precisely those which diagonalize as follows,

$$P = UDU^*$$

with  $U \in U_N$ , and with  $D \in M_N(0,1)$  being diagonal.

**PROOF.** We know that the equation for the projections is as follows:

$$P^2 = P = P^*$$

Thus the eigenvalues  $\lambda$  of our projection must be real, and we have as well the following condition, coming from  $P^2 = P$ :

$$\begin{array}{rcl} \lambda < v,v > &=& <\lambda v,v > \\ &=& < Pv,v > \\ &=& < P^2v,v > \\ &=& < Pv,Pv > \\ &=& <\lambda v,\lambda v > \\ &=& \lambda^2 < v,v > \end{array}$$

Thus we obtain  $\lambda \in \{0, 1\}$ , as claimed, and as a final conclusion here, the diagonalization of the self-adjoint matrices is as follows, with  $e_i \in \{0, 1\}$ :

$$P \sim \begin{pmatrix} e_1 & & \\ & \ddots & \\ & & e_N \end{pmatrix}$$

To be more precise, the number of 1 values is the dimension of the image of P, and the number of 0 values is the dimension of space of vectors sent to 0 by P.

In the real case, the result regarding the projections is as follows:

PROPOSITION 4.27. The matrices  $P \in M_N(\mathbb{R})$  which are projections,

$$P^2 = P = P^t$$

are precisely those which diagonalize as follows,

$$P = UDU^t$$

with  $U \in O_N$ , and with  $D \in M_N(0,1)$  being diagonal.

**PROOF.** This follows indeed from Proposition 4.26, and its proof.

An important class of self-adjoint matrices, that we will discuss now, which includes all the projections, are the positive matrices.

The general theory here is as follows:

THEOREM 4.28. For a matrix  $A \in M_N(\mathbb{C})$  the following conditions are equivalent, and if they are satisfied, we say that A is positive:

- (1)  $A = B^2$ , with  $B = B^*$ .
- (2)  $A = CC^*$ , for some  $C \in M_N(\mathbb{C})$ .
- (3)  $\langle Ax, x \rangle \geq 0$ , for any vector  $x \in \mathbb{C}^N$ .
- (4)  $A = A^*$ , and the eigenvalues are positive,  $\lambda_i \ge 0$ .
- (5)  $A = UDU^*$ , with  $U \in U_N$  and with  $D \in M_N(\mathbb{R}_+)$  diagonal.

**PROOF.** The idea is that the equivalences in the statement basically follow from some elementary computations, with only Theorem 4.24 needed, at some point:

- (1)  $\implies$  (2) This is clear, because we can take C = B.
- $(2) \implies (3)$  This follows from the following computation:

$$\langle Ax, x \rangle = \langle CC^*x, x \rangle$$
  
=  $\langle C^*x, C^*x \rangle$   
 $\geq 0$ 

(3)  $\implies$  (4) By using the fact that  $\langle Ax, x \rangle$  is real, we have:

$$\langle Ax, x \rangle = \langle x, A^*x \rangle$$
  
=  $\langle A^*x, x \rangle$ 

Thus we have  $A = A^*$ , and the remaining assertion, regarding the eigenvalues, follows from the following computation, assuming  $Ax = \lambda x$ :

$$\langle Ax, x \rangle = \langle \lambda x, x \rangle$$
  
=  $\lambda \langle x, x \rangle$   
 $\geq 0$ 

(4)  $\implies$  (5) This follows by using Theorem 4.24 above.

(5)  $\implies$  (1) Assuming  $A = UDU^*$  is as in the statement, with  $U \in U_N$ , and with  $D \in M_N(\mathbb{R}_+)$  being diagonal, we can set:

$$B = U\sqrt{D}U^*$$

Then B is self-adjoint, and its square is given by:

$$B^{2} = U\sqrt{D}U^{*} \cdot U\sqrt{D}U^{*}$$
$$= UDU^{*}$$
$$= A$$

Thus, we are led to the conclusion in the statement.

Let us record as well the following technical version of the above result:

THEOREM 4.29. For a matrix  $A \in M_N(\mathbb{C})$  the following conditions are equivalent, and if they are satisfied, we say that A is strictly positive:

(1)  $A = B^2$ , with  $B = B^*$ , invertible.

(2)  $A = CC^*$ , for some  $C \in M_N(\mathbb{C})$  invertible.

(3)  $\langle Ax, x \rangle > 0$ , for any nonzero vector  $x \in \mathbb{C}^N$ .

(4)  $A = A^*$ , and the eigenvalues are strictly positive,  $\lambda_i > 0$ .

(5)  $A = UDU^*$ , with  $U \in U_N$  and with  $D \in M_N(\mathbb{R}^*_{\perp})$  diagonal.

**PROOF.** This follows either from Theorem 4.28, by adding the various extra assumptions in the statement, or from the proof of Theorem 4.28, by modifying where needed.  $\Box$ 

The positive matrices are quite important, for a number of reasons. On one hand, these are the matrices  $A \in M_N(\mathbb{C})$  having a square root  $\sqrt{A} \in M_N(\mathbb{C})$ , as shown by our positivity condition (1). On the other hand, any matrix  $A \in M_N(\mathbb{C})$  produces the positive matrix  $A^*A \in M_N(\mathbb{C})$ , as shown by our positivity condition (2). We can combine these two observations, and we are led to the following construction, for any  $A \in M_N(\mathbb{C})$ :

$$A \to \sqrt{A^*A}$$

This is something quite interesting, because at N = 1 what we have here is the construction of the absolute value of the complex numbers,  $|z| = \sqrt{z\overline{z}}$ . This suggests using the notation  $|A| = \sqrt{A^*A}$ , and then looking for a decomposition result of type:

$$A = U|A|$$

We will be back to this type of decomposition later, called polar decomposition, at the end of the present chapter, after developing some more general theory.

Let us discuss now the case of the unitary matrices. We have here:

THEOREM 4.30. Any matrix  $U \in M_N(\mathbb{C})$  which is unitary,  $U^* = U^{-1}$ , is diagonalizable, with the eigenvalues on  $\mathbb{T}$ . More precisely we have

$$U = VDV^*$$

with  $V \in U_N$ , and with  $D \in M_N(\mathbb{T})$  diagonal. The converse holds too.

PROOF. As a first remark, the converse trivially holds, because given a matrix of type  $U = VDV^*$ , with  $V \in U_N$ , and with  $D \in M_N(\mathbb{T})$  being diagonal, we have:

$$U^* = (VDV^*)^*$$
  
=  $VD^*V^*$   
=  $VD^{-1}V^{-1}$   
=  $(V^*)^{-1}D^{-1}V^{-1}$   
=  $(VDV^*)^{-1}$   
=  $U^{-1}$ 

Let us prove now the first assertion, stating that the eigenvalues of a unitary matrix  $U \in U_N$  belong to  $\mathbb{T}$ . Indeed, assuming  $Uv = \lambda v$ , we have:

$$\langle v, v \rangle = \langle U^*Uv, v \rangle$$
  
=  $\langle Uv, Uv \rangle$   
=  $\langle \lambda v, \lambda v \rangle$   
=  $|\lambda|^2 \langle v, v \rangle$ 

Thus we obtain  $\lambda \in \mathbb{T}$ , as claimed. Our next claim now is that the eigenspaces corresponding to different eigenvalues are pairwise orthogonal. Assume indeed that:

$$Uv = \lambda v$$
 ,  $Uw = \mu w$ 

We have then the following computation, using  $U^* = U^{-1}$  and  $\lambda, \mu \in \mathbb{T}$ :

$$\lambda < v, w > = < \lambda v, w >$$

$$= < Uv, w >$$

$$= < Uv, w >$$

$$= < v, U^*w >$$

$$= < v, U^{-1}w >$$

$$= < v, \mu^{-1}w >$$

$$= \mu < v, w >$$

Thus  $\lambda \neq \mu$  implies  $v \perp w$ , as claimed. In order now to finish, it remains to prove that the eigenspaces span  $\mathbb{C}^N$ . For this purpose, we will use a recurrence method. Let us pick an eigenvector,  $Uv = \lambda v$ . Assuming  $v \perp w$ , we have:

$$\begin{array}{rcl} < Uw,v> &=& < w,U^*v> \\ &=& < w,U^{-1}v> \\ &=& < w,\lambda^{-1}v> \\ &=& \lambda < w,v> \\ &=& 0 \end{array}$$

Thus, if v is an eigenvector, then the vector space  $v^{\perp}$  is invariant under U. Now since U is an isometry, so is its restriction to this space  $v^{\perp}$ . Thus this restriction is a unitary, and so we can proceed by recurrence, and we obtain the result.

Let us record as well the real version of the above result, in a weak form:

THEOREM 4.31. Any matrix  $U \in M_N(\mathbb{R})$  which is orthogonal,  $U^t = U^{-1}$ , is diagonalizable, with the eigenvalues on  $\mathbb{T}$ . More precisely we have

$$U = VDV^*$$

with  $V \in U_N$ , and with  $D \in M_N(\mathbb{T})$  being diagonal.

**PROOF.** This follows indeed from Theorem 4.30 above.

99

Observe that the above result does not provide us with a complete characterization of the matrices  $U \in M_N(\mathbb{R})$  which are orthogonal. To be more precise, the question left is that of understanding when the matrices of type  $U = VDV^*$ , with  $V \in U_N$ , and with  $D \in M_N(\mathbb{T})$  being diagonal, are real, and this is something non-trivial.

Observe that the above generalizes our previous computation for the rotation matrix  $R_t$ , which has no real eigenvalues, but has two complex eigenvalues, of modulus 1.

Back to generalities, the self-adjoint matrices and the unitary matrices are particular cases of the general notion of a "normal matrix", and we have here:

THEOREM 4.32. Any matrix  $A \in M_N(\mathbb{C})$  which is normal,  $AA^* = A^*A$ , is diagonalizable, with the diagonalization being of the following type,

$$A = UDU^*$$

with  $U \in U_N$ , and with  $D \in M_N(\mathbb{C})$  diagonal. The converse holds too.

PROOF. As a first remark, the converse trivially holds, because if we take a matrix of the form  $A = UDU^*$ , with U unitary and D diagonal, then we have:

$$AA^* = UDU^* \cdot UD^*U^*$$
$$= UDD^*U^*$$
$$= UD^*DU^*$$
$$= UD^*U^* \cdot UDU^*$$
$$= A^*A$$

In the other sense now, this is something more technical. Our first claim is that a matrix A is normal precisely when the following happens, for any vector v:

$$||Av|| = ||A^*v||$$

Indeed, the above equality can be written as follows:

$$< AA^*v, v > = < A^*Av, v >$$

But this is equivalent to  $AA^* = A^*A$ , by using the polarization identity. Our claim now is that  $A, A^*$  have the same eigenvectors, with conjugate eigenvalues:

$$Av = \lambda v \implies A^*v = \overline{\lambda}v$$

Indeed, this follows from the following computation, and from the trivial fact that if A is normal, then so is any matrix of type  $A - \lambda 1_N$ :

$$||(A^* - \bar{\lambda} 1_N)v|| = ||(A - \lambda 1_N)^*v|| = ||(A - \lambda 1_N)v|| = 0$$

Let us prove now, by using this, that the eigenspaces of A are pairwise orthogonal. Assuming  $Av = \lambda v$  and  $Aw = \mu w$  with  $\lambda \neq \mu$ , we have:

$$\lambda < v, w > = < \lambda v, w >$$
  
=  $< Av, w >$   
=  $< v, A^*w >$   
=  $< v, \bar{\mu}w >$   
=  $\mu < v, w >$ 

Thus  $\lambda \neq \mu$  implies  $v \perp w$ , as claimed. In order to finish now the proof, it remains to prove that the eigenspaces of A span the whole  $\mathbb{C}^N$ . This is something that we have already seen for the self-adjoint matrices, and for the unitaries, and we will use here these results, in order to deal with the general normal case. As a first observation, given an arbitrary matrix A, the matrix  $AA^*$  is self-adjoint:

$$(AA^*)^* = AA^*$$

Thus, we can diagonalize this matrix  $AA^*$ , as follows, with the passage matrix being a unitary,  $V \in U_N$ , and with the diagonal form being real,  $E \in M_N(\mathbb{R})$ :

$$AA^* = VEV^*$$

Now observe that, for matrices of type  $A = UDU^*$ , which are those that we supposed to deal with, we have  $V = U, E = D\overline{D}$ . In particular, A and  $AA^*$  have the same eigenspaces. So, this will be our idea, proving that the eigenspaces of  $AA^*$  are eigenspaces of A. In order to do so, let us pick two eigenvectors v, w of the matrix  $AA^*$ , corresponding to different eigenvalues,  $\lambda \neq \mu$ . The eigenvalue equations are then as follows:

$$AA^*v = \lambda v$$
 ,  $AA^*w = \mu w$ 

We have the following computation, using the normality condition  $AA^* = A^*A$ , and the fact that the eigenvalues of  $AA^*$ , and in particular  $\mu$ , are real:

$$\lambda < Av, w > = < \lambda Av, w >$$

$$= < A\lambda v, w >$$

$$= < A\lambda v, w >$$

$$= < AAA^*v, w >$$

$$= < AA^*Av, w >$$

$$= < Av, AA^*w >$$

$$= < Av, \mu w >$$

$$= \mu < Av, w >$$

We conclude that we have  $\langle Av, w \rangle = 0$ . But this reformulates as follows:

$$\lambda \neq \mu \implies A(E_{\lambda}) \perp E_{\mu}$$

Now since the eigenspaces of  $AA^*$  are pairwise orthogonal, and span the whole  $\mathbb{C}^N$ , we deduce from this that these eigenspaces are invariant under A:

 $A(E_{\lambda}) \subset E_{\lambda}$ 

But with this result in hand, we can finish. Indeed, we can decompose the problem, and the matrix A itself, following these eigenspaces of  $AA^*$ , which in practice amounts in saying that we can assume that we only have 1 eigenspace. By rescaling, this is the same as assuming that we have  $AA^* = 1$ , and so we are now into the unitary case, that we know how to solve, as explained in Theorem 4.30 above.

Let us discuss now, as a final topic, one more important result, namely the polar decomposition. The idea will be that of writing a formula as follows:

A = U|A|

To be more precise, U will be here a partial isometry, a generalization of the notion of isometry, and the matrix |A| will be a kind of absolute value of A.

In order to discuss this, let us first discuss the absolute values. We have here:

THEOREM 4.33. Given a matrix  $A \in M_N(\mathbb{C})$ , we can construct a matrix |A| as follows, by using the fact that  $A^*A$  is diagonalizable, with positive eigenvalues:

$$|A| = \sqrt{A^*A}$$

This matrix |A| is then positive, and its square is  $|A|^2 = A$ . In the case N = 1, we obtain in this way the usual absolute value of the complex numbers.

PROOF. Consider indeed the matrix  $A^*A$ , which is normal. According to Theorem 4.32, we can diagonalize this matrix as follows, with  $U \in U_N$ , and with D diagonal:

$$A = UDU^{*}$$

Since we have  $A^*A \ge 0$ , it follows that we have  $D \ge 0$ , which means that the entries of D are real, and positive. Thus we can extract the square root  $\sqrt{D}$ , and then set:

$$\sqrt{A^*A} = U\sqrt{D}U^*$$

Now if we call this latter matrix |A|, we are led to the conclusions in the statement, namely  $|A| \ge 0$ , and  $|A|^2 = A$ . Finally, the last assertion is clear from definitions.

We can now formulate a first polar decomposition result, as follows:

THEOREM 4.34. Any invertible matrix  $A \in M_N(\mathbb{C})$  decomposes as

$$A = U|A|$$

with  $U \in U_N$ , and with  $|A| = \sqrt{A^*A}$  as above.

**PROOF.** This is routine, and follows by comparing the actions of A, |A| on the vectors  $v \in \mathbb{C}^N$ , and deducing from this the existence of a unitary  $U \in U_N$  as above.  $\Box$ 

#### 4E. EXERCISES

Observe that at N = 1 we obtain in this way the usual polar decomposition of the nonzero complex numbers. There are of course many other examples.

More generally now, we have the following result:

THEOREM 4.35. Any square matrix  $A \in M_N(\mathbb{C})$  decomposes as

A = U|A|

with U being a partial isometry, and with  $|A| = \sqrt{A^*A}$  as above.

**PROOF.** Once again, this follows by comparing the actions of A, |A| on the vectors  $v \in \mathbb{C}^N$ , and deducing from this the existence of a partial isometry U as above. Alternatively, we can get this from Theorem 4.34, applied on the complement of the 0-eigenvectors.  $\Box$ 

We will be back to this in chapter 7 below, when doing spectral theory.

## 4e. Exercises

Things have been quite dense in this chapter, which was our last one on linear algebra, and the main purpose of the exercises below will be that of filling some gaps in the above material, and also discussing some important things which must be known as well. As a first exercise, in relation with linear spaces, bases and dimensions, we have:

EXERCISE 4.36. Clarify the theory of bases and dimensions for the linear subspaces  $V \subset \mathbb{R}^N$ , notably by establishing the formula

$$\dim(\ker f) + \dim(Imf) = N$$

valid for any linear map  $f : \mathbb{C}^N \to \mathbb{C}^N$ , and then extend this into a theory of abstract linear spaces V, which are not necessarily subspaces of  $\mathbb{C}^N$ .

Here the first question is something quite standard, by using the material that we already have. As for the second question, things are a bit more tricky here, because once the abstract linear spaces V being defined, the only available tool is recurrence.

In relation now with the general diagonalization theorems, we have:

EXERCISE 4.37. Work out what happens to the main diagonalization theorem for the matrices  $A \in M_N(\mathbb{C})$ , in the cases  $A \in M_2(\mathbb{C})$ ,  $A \in M_N(\mathbb{R})$ , and  $A \in M_2(\mathbb{R})$ .

As before, this is a rather theoretical exercise, the point being that of carefully reviewing all the material above, in the 3 particular cases which are indicated.

In relation now with resultants and discriminants, there is a lot of theory here, only partly discussed in the above, and as an exercise on the subject, we have:

EXERCISE 4.38. Establish with full details the resultant formula

$$R(P,Q) = \begin{vmatrix} p_k & q_l \\ \vdots & \ddots & \vdots & \ddots \\ p_0 & p_k & q_0 & q_k \\ & \ddots & \vdots & & \ddots & \vdots \\ & & p_0 & & q_0 \end{vmatrix}$$

and discuss as well what happens in the context of the discriminant.

As before, there are many things that can be done or learned here.

In relation with the density tricks, and their applications, we have:

EXERCISE 4.39. Clarify which functions can be applied to which matrices, as to have results stating that the eigenvalues of f(A) are  $f(\lambda_1), \ldots, f(\lambda_N)$ .

This exercise is actually quite difficult. We will be back to this.

In relation with the spectral theorems, we have:

EXERCISE 4.40. Work out specialized spectral theorems for the orthogonal matrices  $U \in O_N$ , going beyond what has been said in the above.

To be more precise here, we have proved many spectral theorems in the above, but the case  $U \in O_N$ , where our statement here was something quite weak, coming without a converse, is obviously still in need of some discussion.

Finally, no linear algebra book would be complete without:

EXERCISE 4.41. Prove that any matrix can be put in Jordan form,

$$A \sim \begin{pmatrix} J_1 & & \\ & \ddots & \\ & & J_k \end{pmatrix}$$

with each of the blocks which appear, called Jordan blocks, being as follows,

$$J_i = \begin{pmatrix} \lambda_i & 1 & & \\ & \lambda_i & 1 & & \\ & & \ddots & \ddots & \\ & & & \lambda_i & 1 \\ & & & & & \lambda_i \end{pmatrix}$$

with the size being the multiplicity of  $\lambda_i$ .

This is something useful, somewhat the "nuclear option" in linear algebra.

# Part II

# **Basic** analysis

Everything dies, baby, that's a fact But maybe everything that dies some day comes back Put your makeup on, fix your hair up pretty And meet me tonight in Atlantic City

# CHAPTER 5

# Standard calculus

## 5a. Derivatives

We discuss in what follows some applications of the theory that we developed above, for the most to questions of analysis. The main idea here will be that the functions of several variables  $f : \mathbb{R}^N \to \mathbb{R}^M$  can be locally approximated by linear maps, in the same way as the functions  $f : \mathbb{R} \to \mathbb{R}$  can be locally approximated by using derivatives:

$$f(x+v) \simeq f(x) + Av$$

Let us begin, however, with some straightforward applications to algebra. First, we have the following fundamental result, which has countless concrete applications, and which is actually at the origin of the whole linear algebra theory:

THEOREM 5.1. Any linear system of equations

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \ldots + a_{1N}x_N &= v_1 \\ a_{21}x_1 + a_{22}x_2 + \ldots + a_{2N}x_N &= v_2 \\ \vdots \\ a_{N1}x_1 + a_{N2}x_2 + \ldots + a_{NN}x_N &= v_N \end{cases}$$

can be written in matrix form, as follows,

Ax = v

and when A is invertible, its solution is given by  $x = A^{-1}v$ .

**PROOF.** With linear algebra conventions, our system reads:

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \vdots & & & \vdots \\ a_{N1} & a_{N2} & \dots & a_{NN} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_N \end{pmatrix}$$

Thus, we are led to the conclusions in the statement.

There are many applications of the above result. In each case, we are led to the computation of an inverse matrix  $A^{-1}$ , and this is something that we know how to solve. Thus, we have here a first true application of the theory developed so far.
Along the same lines, we have as well the following result:

THEOREM 5.2. Any linear recurrence system

$$\begin{cases} x_{k+1} = a_{11}x_k + a_{12}y_k + \dots + a_{1N}z_k \\ y_{k+1} = a_{21}x_k + a_{22}y_k + \dots + a_{2N}z_k \\ \vdots \\ z_{k+1} = a_{N1}x_k + a_{N2}y_k + \dots + a_{NN}z_k \end{cases}$$

can be written in matrix form, as follows,

$$v_{k+1} = Av_k$$

and so the solution is given by  $v_k = A^k v_0$ .

**PROOF.** With linear algebra conventions, the recurrence reads:

$$\begin{pmatrix} x_{k+1} \\ y_{k+1} \\ \vdots \\ z_{k+1} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \vdots & & & \vdots \\ a_{N1} & a_{N2} & \dots & a_{NN} \end{pmatrix} \begin{pmatrix} x_k \\ y_k \\ \vdots \\ z_k \end{pmatrix}$$

Thus, we are led to the conclusions in the statement.

Let us discuss now applications to analysis. As already mentioned, the main idea will be that of approximating the functions  $f : \mathbb{R}^N \to \mathbb{R}^M$  by affine maps, as follows:

$$f(x+v) \simeq f(x) + Av$$

In order to discuss this, let us start with the simplest case, M = N = 1. Here we have the following result, which is the starting point for everything in analysis:

THEOREM 5.3. Any function of one variable  $f : \mathbb{R} \to \mathbb{R}$  is locally affine,

$$f(x+t) \simeq f(x) + f'(x)t$$

with  $f'(x) \in \mathbb{R}$  being the derivative of f at the point x, given by

$$f'(x) = \lim_{t \to 0} \frac{f(x+t) - f(x)}{t}$$

provided that this latter limit converges indeed.

**PROOF.** Assume indeed that the limit in the statement converges. By multiplying by t, we obtain that we have, once again in the  $t \to 0$  limit:

$$f(x+t) - f(x) \simeq f'(x)t$$

Thus, we are led to the conclusion in the statement.

As an illustration, the derivatives of the power functions are as follows:

108

#### 5A. DERIVATIVES

**PROPOSITION 5.4.** We have the differentiation formula

$$(x^p)' = px^{p-1}$$

valid for any exponent  $p \in \mathbb{R}$ .

PROOF. We have to compute a certain limit, depending on an exponent  $p \in \mathbb{R}$ , and we can do this in three steps, as follows:

(1) Let us first discuss the case of integer exponents,  $p \in \mathbb{N}$ . Here we can use the binomial formula, which gives, with  $t \to 0$ :

$$(x+t)^{p} = \sum_{k=0}^{n} {p \choose k} x^{p-k} t^{k}$$
$$= x^{p} + p x^{p-1} t + \ldots + t^{p}$$
$$\simeq x^{p} + p x^{p-1} t$$

Thus, we have our locally affine approximation of  $x^p$ , and the derivative that we are looking for is the coefficient of t. We obtain  $(x^p)' = px^{p-1}$ , as desired.

(2) Let us discuss now the general case  $p \in \mathbb{Q}$ . We write p = m/n, with  $m \in \mathbb{N}$  and  $n \in \mathbb{Z}$ . In order to do the computation, we use the following formula:

$$a^{n} - b^{n} = (a - b)(a^{n-1} + a^{n-2}b + \dots + b^{n-1})$$

To be more precise, we will use this formula, written as follows:

$$a - b = \frac{a^n - b^n}{a^{n-1} + a^{n-2}b + \dots + b^{n-1}}$$

We set in this formula  $a = (x + t)^{m/n}$  and  $b = x^{m/n}$ . We obtain in this way, in the  $t \to 0$  limit, and by using the binomial formula for approximating on top:

 $| \rangle m$ 

$$(x+t)^{m/n} - x^{m/n} = \frac{(x+t)^m - x^m}{(x+t)^{m(n-1)/n} + \dots + x^{m(n-1)/n}}$$
  

$$\simeq \frac{(x+t)^m - x^m}{nx^{m(n-1)/n}}$$
  

$$\simeq \frac{mx^{m-1}t}{nx^{m(n-1)/n}}$$
  

$$= \frac{m}{n} \cdot x^{m-1-m+m/n} \cdot t$$
  

$$= \frac{m}{n} \cdot x^{m/n-1} \cdot t$$

Thus, we have our first order approximation of  $x^p$ , with p = m/n, and the derivative that we are looking for is the coefficient of t found above,  $(x^p)' = px^{p-1}$ .

(3) In the general case now, where  $p \in \mathbb{R}$  is real, the same formula holds, namely  $(x^p)' = px^{p-1}$ , by using what we found above, and a continuity argument.  $\Box$ 

There are many applications of the derivative, and we have for instance:

PROPOSITION 5.5. The local minima and maxima of a differentiable function  $f : \mathbb{R} \to \mathbb{R}$  appear at the points  $x \in \mathbb{R}$  where:

$$f'(x) = 0$$

However, the converse of this fact is not true in general.

**PROOF.** The first assertion is clear from the formula in Theorem 5.3, namely:

$$f(x+t) \simeq f(x) + f'(x)t$$

As for the converse, the simplest counterexample is  $f(x) = x^3$ , at x = 0.

At a more advanced level now, we have the following result:

THEOREM 5.6. Any function of one variable  $f : \mathbb{R} \to \mathbb{R}$  is locally quadratic,

$$f(x+t) \simeq f(x) + f'(x)t + \frac{f''(x)}{2}t^2$$

where f''(x) is the derivative of the function  $f' : \mathbb{R} \to \mathbb{R}$  at the point x.

**PROOF.** Assume indeed that f is twice differentiable at x, and let us try to construct an approximation of f around x by a quadratic function, as follows:

$$f(x+t) \simeq a + bt + ct^2$$

We must have a = f(x), and we also know from Theorem 5.3 that b = f'(x) is the correct choice for the coefficient of t. Thus, our approximation must be as follows:

$$f(x+t) \simeq f(x) + f'(x)t + ct^2$$

In order to find the correct choice for  $c \in \mathbb{R}$ , observe that the function  $t \to f(x+t)$  matches with  $t \to f(x) + f'(x)t + ct^2$  in what regards the value at t = 0, and also in what regards the value of the derivative at t = 0. Thus, the correct choice of  $c \in \mathbb{R}$  should be the one making match the second derivatives at t = 0, and this gives:

$$f''(x) = 2c$$

We are therefore led to the formula in the statement, namely:

$$f(x+t) \simeq f(x) + f'(x)t + \frac{f''(x)}{2}t^2$$

In order to prove now that this formula holds indeed, we will use L'Hôpital's rule, which states that the 0/0 type limits can be computed as follows:

$$\frac{f(x)}{g(x)} \simeq \frac{f'(x)}{g'(x)}$$

Observe that this formula holds indeed, as an application of Theorem 5.3. Now by using this, if we denote by  $\varphi(t) \simeq P(t)$  the formula to be proved, we have:

$$\frac{\varphi(t) - P(t)}{t^2} \simeq \frac{\varphi'(t) - P'(t)}{2t}$$
$$\simeq \frac{\varphi''(t) - P''(t)}{2}$$
$$= \frac{f''(x) - f''(x)}{2}$$
$$= 0$$

Thus, we are led to the conclusion in the statement.

The above result substantially improves Theorem 5.3, and there are many applications of this. We can improve for instance Proposition 5.5, as follows:

PROPOSITION 5.7. The local minima and maxima of a twice differentiable function  $f : \mathbb{R} \to \mathbb{R}$  appear at the points  $x \in \mathbb{R}$  where

$$f'(x) = 0$$

with the local minima corresponding to the case  $f'(x) \ge 0$ , and with the local maxima corresponding to the case  $f''(x) \le 0$ .

**PROOF.** The first assertion is something that we already know. As for the second assertion, we can use the formula in Theorem 5.6, which in the case f'(x) = 0 reads:

$$f(x+t) \simeq f(x) + \frac{f''(x)}{2}t^2$$

Indeed, assuming  $f''(x) \neq 0$ , it is clear that the condition f''(x) > 0 will produce a local minimum, and that the condition f''(x) < 0 will produce a local maximum.

We can further develop the above method, at order 3, at order 4, and so on, the ultimate result on the subject, called Taylor formula, being as follows:

THEOREM 5.8. Any function  $f : \mathbb{R} \to \mathbb{R}$  can be locally approximated as

$$f(x+t) = \sum_{k=0}^{\infty} \frac{f^{(k)}(x)}{k!} t^k$$

where  $f^{(k)}(x)$  are the higher derivatives of f at the point x.

**PROOF.** Consider the function to be approximated, namely:

$$\varphi(t) = f(x+t)$$

Let us try to best approximate this function at a given order  $n \in \mathbb{N}$ . We are therefore looking for a certain polynomial in t, of the following type:

$$P(t) = a_0 + a_1 t + \ldots + a_n t^n$$

The natural conditions to be imposed are those stating that P and  $\varphi$  should match at t = 0, at the level of the actual value, of the derivative, second derivative, and so on up the *n*-th derivative. Thus, we are led to the approximation in the statement:

$$f(x+t) \simeq \sum_{k=0}^{n} \frac{f^{(k)}(x)}{k!} t^{k}$$

In order to prove now that this approximation holds indeed, we use L'Hôpital's rule, applied several times, as in the proof of Theorem 5.6 above. To be more precise, if we denote by  $\varphi(t) \simeq P(t)$  the approximation to be proved, we have:

$$\frac{\varphi(t) - P(t)}{t^n} \simeq \frac{\varphi'(t) - P'(t)}{nt^{n-1}}$$
$$\simeq \frac{\varphi''(t) - P''(t)}{n(n-1)t^{n-2}}$$
$$\vdots$$
$$\approx \frac{\varphi^{(n)}(t) - P^{(n)}(t)}{n!}$$
$$= \frac{f^{(n)}(x) - f^{(n)}(x)}{n!}$$
$$= 0$$

Thus, we are led to the conclusion in the statement.

While fully proved in the above, the Taylor formula remains something quite mysterious. Here is a related interesting statement, inspired from the proof:

**PROPOSITION 5.9.** For a polynomial of degree n, the Taylor approximation

$$f(x+t) \simeq \sum_{k=0}^{n} \frac{f^{(k)}(x)}{k!} t^{k}$$

is an equality. The converse of this statement holds too.

PROOF. By linearity, it is enough to check the equality in question for the monomials  $f(x) = x^p$ , with  $p \le n$ . But here, the formula to be proved is as follows:

$$(x+t)^p \simeq \sum_{k=0}^p \frac{p(p-1)\dots(p-k+1)}{k!} x^{p-k} t^k$$

We recognize the binomial formula, so our result holds indeed. As for the converse, this is clear, because the Taylor approximation is a polynomial of degree n.

As an application of the Taylor formula, we can now improve the binomial formula, which was actually our main tool so far, in the following way:

112

THEOREM 5.10. We have the following generalized binomial formula, with  $p \in \mathbb{R}$ ,

$$(x+t)^p = \sum_{k=0}^{\infty} \binom{p}{k} x^{p-k} t^k$$

with the generalized binomial coefficients being given by the formula

$$\binom{p}{k} = \frac{p(p-1)\dots(p-k+1)}{k!}$$

valid for any |t| < |x|. With  $p \in \mathbb{N}$ , we recover the usual binomial formula.

PROOF. It is customary to divide everything by x, which is the same as assuming x = 1. The formula to be proved is then as follows, under the assumption |t| < 1:

$$(1+t)^p = \sum_{k=0}^{\infty} \binom{p}{k} t^k$$

Let us discuss now the validity of this formula, depending on  $p \in \mathbb{R}$ :

(1) Case  $p \in \mathbb{N}$ . According to our definition of the generalized binomial coefficients, we have  $\binom{p}{k} = 0$  for k > p, so the series is stationary, and the formula to be proved is:

$$(1+t)^p = \sum_{k=0}^p \binom{p}{k} t^k$$

But this is the usual binomial formula, which holds for any  $t \in \mathbb{R}$ .

(2) Case p = -1. Here we can use the following formula, valid for |t| < 1:

$$\frac{1}{1+t} = 1 - t + t^2 - t^3 + \dots$$

But this is exactly our generalized binomial formula at p = -1, because:

$$\binom{-1}{k} = \frac{(-1)(-2)\dots(-k)}{k!} = (-1)^k$$

(3) Case  $p \in -\mathbb{N}$ . This is a continuation of our study at p = -1, which will finish the study at  $p \in \mathbb{Z}$ . With p = -m, the generalized binomial coefficients are:

$$\begin{pmatrix} -m \\ k \end{pmatrix} = \frac{(-m)(-m-1)\dots(-m-k+1)}{k!}$$

$$= (-1)^k \frac{m(m+1)\dots(m+k-1)}{k!}$$

$$= (-1)^k \frac{(m+k-1)!}{(m-1)!k!}$$

$$= (-1)^k \binom{m+k-1}{m-1}$$

Thus, our generalized binomial formula at p = -m reads:

$$\frac{1}{(1+t)^m} = \sum_{k=0}^{\infty} (-1)^k \binom{m+k-1}{m-1} t^k$$

But this is something which holds indeed, with |t| < 1, one proof being by multiplying everything by  $(1+t)^m$ , expanded by using the usual binomial formula. Indeed, by working out the combinatorics, on the right side, we eventually obtain 1.

(4) General case,  $p \in \mathbb{R}$ . As we can see, things escalate quickly, so we will skip the next step,  $p \in \mathbb{Q}$ , and discuss directly the case  $p \in \mathbb{R}$ . Consider the following function:

$$f(x) = x^p$$

The derivatives at x = 1 are then given by the following formula:

$$f^{(k)}(1) = p(p-1)\dots(p-k+1)$$

Thus, the Taylor approximation at x = 1 is as follows:

$$f(1+t) = \sum_{k=0}^{\infty} \frac{p(p-1)\dots(p-k+1)}{k!} t^k$$

But this is exactly our generalized binomial formula, so we are done with the case where t is small. With a bit more care, we obtain that this holds for any |t| < 1.

We can see from the above the power of the Taylor formula. As an application now of our generalized binomial formula, we can extract square roots, as follows:

**PROPOSITION 5.11.** We have the following formula,

$$\sqrt{1+t} = 1 - 2\sum_{k=1}^{\infty} C_{k-1} \left(\frac{-t}{4}\right)^k$$

with  $C_k = \frac{1}{k+1} {\binom{2k}{k}}$  being the Catalan numbers. Also, we have

$$\frac{1}{\sqrt{1+t}} = \sum_{k=0}^{\infty} D_k \left(\frac{-t}{4}\right)^k$$

with  $D_k = \binom{2k}{k}$  being the central binomial coefficients.

**PROOF.** The above formulae both follow from Theorem 5.10, as follows:

(1) At p = 1/2, the generalized binomial coefficients are:

$$\binom{1/2}{k} = \frac{1/2(-1/2)\dots(3/2-k)}{k!}$$

$$= (-1)^{k-1} \frac{1\cdot 3\cdot 5\dots(2k-3)}{2^k k!}$$

$$= (-1)^{k-1} \frac{(2k-2)!}{2^{k-1}(k-1)!2^k k!}$$

$$= \frac{(-1)^{k-1}}{2^{2k-1}} \cdot \frac{1}{k} \binom{2k-2}{k-1}$$

$$= -2 \left(\frac{-1}{4}\right)^k C_{k-1}$$

(2) At p = -1/2, the generalized binomial coefficients are:

$$\begin{pmatrix} -1/2 \\ k \end{pmatrix} = \frac{-1/2(-3/2)\dots(1/2-k)}{k!}$$

$$= (-1)^k \frac{1 \cdot 3 \cdot 5 \dots (2k-1)}{2^k k!}$$

$$= (-1)^k \frac{(2k)!}{2^k k! 2^k k!}$$

$$= \frac{(-1)^k}{4^k} \binom{2k}{k}$$

$$= \left(\frac{-1}{4}\right)^k D_k$$

Thus, we obtain the formulae in the statement.

As another basic application of the Taylor series, we have:

THEOREM 5.12. We have the following formulae,

$$\sin t = \sum_{l} (-1)^{l} \frac{t^{2l+1}}{(2l+1)!} \quad , \quad \cos t = \sum_{l} (-1)^{l} \frac{t^{2l}}{(2l)!}$$

as well as the following formulae,

$$e^x = \sum_k \frac{x^k}{k!}$$
,  $\log(1+t) = \sum_k (-1)^{k+1} \frac{x^k}{k}$ 

as Taylor series, and in general as well, with |t| < 1 needed for log.

**PROOF.** There are several statements here, the proofs being as follows:

(1) Regarding sin and cos, we can use here the following well-known formulae:

$$\sin(x+t) = \sin x \cos t + \cos x \sin t$$

$$\cos(x+t) = \cos x \cos t - \sin x \sin t$$

With these formulae in hand we can appproximate both sin and cos, and we get:

$$(\sin t)' = \cos t$$
$$(\cos t)' = -\sin t$$

Thus, we can differentiate sin and cos as many times as we want to, so we can compute the corresponding Taylor series, and we obtain the formulae in the statement.

(2) Regarding exp and log, here the needed formulae, which lead to the formulae in the statement for the corresponding Taylor series, are as follows:

$$(e^{x})' = e^{x}$$
$$(\log x)' = x^{-1}$$
$$(x^{p})' = px^{p-1}$$

(3) Finally, the fact that the formulae in the statement extend beyond the small t setting, coming from Taylor series, is something standard too.

Let us discuss now what happens in several variables. At order 1, the situation is quite similar to the one in 1 variable, but this time involving matrices, as follows:

THEOREM 5.13. Any function of several variables can be locally approximated as

 $f(x+t) \simeq f(x) + At$ 

with A being the matrix of partial derivatives at x,

$$A = \left(\frac{\partial f_i}{\partial x_j}(x)\right)_{ij}$$

acting on the vectors  $t \in \mathbb{C}^N$  by usual multiplication.

**PROOF.** This is indeed standard, by using a recurrence method, with respect to the number of dimensions. To be more precise, consider the matrix in the statement:

$$A = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(x) & \dots & \frac{\partial f_1}{\partial x_N}(x) \\ \vdots & & \vdots \\ \frac{\partial f_N}{\partial x_1}(x) & \dots & \frac{\partial f_N}{\partial x_N}(x) \end{pmatrix}$$

#### **5B. INTEGRATION**

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The idea is then that around x, our function behaves exactly as  $t \to At$ :

$$f\begin{pmatrix}x_1+t_1\\\vdots\\x_N+t_N\end{pmatrix}\simeq f\begin{pmatrix}x_1\\\vdots\\x_N\end{pmatrix}+\begin{pmatrix}\frac{\partial f_1}{\partial x_1}(x)&\ldots&\frac{\partial f_1}{\partial x_N}(x)\\\vdots&\vdots\\\frac{\partial f_N}{\partial x_1}(x)&\ldots&\frac{\partial f_N}{\partial x_N}(x)\end{pmatrix}\begin{pmatrix}t_1\\\vdots\\t_N\end{pmatrix}$$

Thus, we are led to the conclusion in the statement.

Regarding the higher derivatives, the situation here is more complicated. Let us record, however, the following fundamental result, happening at order 2, and which does the job, the job in analysis being usually that of finding the minima or maxima:

**THEOREM** 5.14. Associated to any numeric function of several variables

$$f: \mathbb{R}^N \to \mathbb{R}$$

is its Hessian function, given by the following formula,

$$H(x) = \left(\frac{\partial^2 f}{\partial x_j \partial x_j}(x)\right)_{ij}$$

whose positivity properties are related to the local minima and maxima of f.

**PROOF.** This is quite standard, by using the same method as in the 1D case, namely building on Theorem 5.13, and approximating the function at order 2. 

## **5b.** Integration

The derivative, and the linear algebra methods, are very useful when integrating, and we will discuss this now. Let us first recall the basic theory here:

THEOREM 5.15. We have the Riemann integration formula,

$$\int_{a}^{b} f(x)dx = (b-a) \times \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^{N} f\left(a + \frac{b-a}{N} \cdot k\right)$$

which can serve as a formal definition for the integral.

PROOF. This is indeed standard, by drawing rectangles. We have indeed:

$$\int_{a}^{b} f(x)dx = (b-a) \times \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^{N} f(x_{i})$$

Now when the points  $x_1, \ldots, x_N \in [a, b]$  used in the above are chosen uniformly distributed, we obtain the formula in the statement. 

As a basic application, we can sum powers, with fixed results at small exponents, and estimates in general. We have indeed the following result:

THEOREM 5.16. The basic Riemann sums, namely  $1^k + 2^k + \ldots + N^k$ , depending on an exponent  $k \in \mathbb{N}$ , are as follows:

(1) At 
$$k = 1$$
 we have  $1 + 2 + \ldots + N = \frac{N(N+1)}{2}$ .  
(2) At  $k = 2$  we have  $1^2 + 2^2 + \ldots + N^2 = \frac{N(N+1)(2N+1)}{6}$ .  
(3) At  $k = 3$  we have  $1^3 + 2^3 + \ldots + N^3 = \left[\frac{N(N+1)}{2}\right]^2$ .  
(4) At  $k \in \mathbb{N}$  we have  $1^k + 2^k + \ldots + N^k \simeq \frac{N^{k+1}}{k+1}$ .

**PROOF.** This is something well-known, the idea being as follows:

(1) The average of the numbers  $1, 2, \ldots, N$  is given by:

$$\frac{1}{N}(1+2+\ldots+N) = \frac{N+1}{2}$$

Thus, we obtain the formula in the statement, without any computation.

(2) A well-known proof here is by drawing certain squares in the plane.

(3) A well-known proof here is by drawing certain cubes in the space.

(4) Here there is no trick, and we must use Riemann integration. We have:

$$\int_0^1 x^k dx = \frac{1}{k+1}$$

But this gives the estimate in the statement, by using a Riemann sum.

Getting back now to the basics, and to the definition of the integral, we have as well the following alternative approach:

THEOREM 5.17. We have the Monte Carlo integration formula,

$$\int_{a}^{b} f(x)dx = (b-a) \times \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^{N} f(x_{i})$$

with  $x_1, \ldots, x_N \in [a, b]$  being random.

**PROOF.** We recall from the construction of the integral that the idea is to use a formula as follows, with  $x_1, \ldots, x_N \in [a, b]$  being uniformly distributed:

$$\int_{a}^{b} f(x)dx = (b-a) \times \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^{N} f(x_{i})$$

But this works as well when the points  $x_1, \ldots, x_N \in [a, b]$  are randomly distributed, for somewhat obvious reasons, and this gives the result.

#### 5B. INTEGRATION

119

Observe that the Monte Carlo integration works better than Riemann integration, for instance when trying to improve the estimate, via  $N \rightarrow N + 1$ . Also, the Monte Carlo integration works better for functions having various symmetries. Indeed, the Riemann integration can get "fooled" by these symmetries, but not Monte Carlo.

Getting back now to linear algebra, we have the following key result:

THEOREM 5.18. We have the change of variable formula

$$\int_{a}^{b} f(x)dx = \int_{c}^{d} f(\varphi(t))\varphi'(t)dt$$

where  $c = \varphi^{-1}(a)$  and  $d = \varphi^{-1}(b)$ .

**PROOF.** This follows with f = F', from the following differentiation rule, whose proof is something elementary:

$$(F\varphi)'(t) = F'(\varphi(t))\varphi'(t)$$

Indeed, by integrating between c and d, we obtain the result.

In several variables now, the result is as follows:

THEOREM 5.19. Given a transformation in several variables,  $\varphi = (\varphi_1, \ldots, \varphi_N)$ , we have the following change of variable formula,

$$\int_{E} f(x)dx = \int_{\varphi^{-1}(E)} f(\varphi(t)) |J_{\varphi}(t)| dt$$

with the  $J_{\varphi}$  quantity, called Jacobian, being given by:

$$J_{\varphi}(t) = \det\left[\left(\frac{\partial \varphi_i}{\partial x_j}(x)\right)_{ij}\right]$$

PROOF. Observe first that this generalizes the change of variable formula in 1 dimension, from Theorem 5.18 above. In general, the proof is quite tricky, the idea being that of assuming that the change of variables is linear, and by using the definition of the determinant as a volume. The details and computations here are quite non-trivial.  $\Box$ 

In what regards the applications, these often come via:

THEOREM 5.20. We have polar coordinates in 2 dimensions,

$$\begin{cases} x = r\cos t \\ y = r\sin t \end{cases}$$

the corresponding Jacobian being J = r.

**PROOF.** This is elementary, the Jacobian being:

$$J = \begin{vmatrix} \frac{\partial r \cos t}{\partial r} & \frac{\partial r \cos t}{\partial t} \\ \frac{\partial r \sin t}{\partial r} & \frac{\partial r \sin t}{\partial t} \end{vmatrix}$$
$$= \begin{vmatrix} \cos t & -r \sin t \\ \sin t & r \cos t \end{vmatrix}$$
$$= r \cos^2 t + r \sin^2 t$$
$$= r$$

Thus, we have indeed the formula in the statement.

In 3 dimensions the formula is similar, as follows:

THEOREM 5.21. We have spherical coordinates in 3 dimensions,

$$\begin{cases} x = r \cos s \\ y = r \sin s \cos t \\ z = r \sin s \sin t \end{cases}$$

the corresponding Jacobian being  $J(r,s,t) = r^2 \sin s$ .

**PROOF.** The fact that we have indeed spherical coordinates is clear. Regarding now the Jacobian, this is given by the following formula:

$$J(r, s, t) = \begin{vmatrix} \cos s & -r\sin s & 0 \\ \sin s\cos t & r\cos s\cos t & -r\sin s\sin t \\ \sin s\sin t & r\cos s\sin t & r\sin s\cos t \end{vmatrix}$$
$$= r^{2}\sin s\sin t \begin{vmatrix} \cos s & -r\sin s \\ \sin s\sin t & r\cos s\sin t \end{vmatrix} + r\sin s\cos t \begin{vmatrix} \cos s & -r\sin s \\ \sin s\cos t & r\cos s\cos t \end{vmatrix}$$
$$= r\sin s\sin^{2} t \begin{vmatrix} \cos s & -r\sin s \\ \sin s & r\cos s \end{vmatrix} + r\sin s\cos^{2} t \begin{vmatrix} \cos s & -r\sin s \\ \sin s & r\cos s \end{vmatrix}$$
$$= r\sin s(\sin^{2} t + \cos^{2} t) \begin{vmatrix} \cos s & -r\sin s \\ \sin s & r\cos s \end{vmatrix}$$
$$= r\sin s \times 1 \times r$$
$$= r^{2}\sin s$$

Thus, we have indeed the formula in the statement.

Let us work out now the spherical coordinate formula in N dimensions. The formula here, which generalizes those at N = 2, 3, is as follows:

120

THEOREM 5.22. We have spherical coordinates in N dimensions,

$$\begin{cases} x_1 &= r \cos t_1 \\ x_2 &= r \sin t_1 \cos t_2 \\ \vdots \\ x_{N-1} &= r \sin t_1 \sin t_2 \dots \sin t_{N-2} \cos t_{N-1} \\ x_N &= r \sin t_1 \sin t_2 \dots \sin t_{N-2} \sin t_{N-1} \end{cases}$$

the corresponding Jacobian being given by the following formula:

$$J(r,t) = r^{N-1} \sin^{N-2} t_1 \sin^{N-3} t_2 \dots \sin^2 t_{N-3} \sin t_{N-2}$$

**PROOF.** As before, by developing the determinant over the last column, we have:

$$J_{N} = r \sin t_{1} \dots \sin t_{N-2} \sin t_{N-1} \times \sin t_{N-1} J_{N-1}$$
  
+  $r \sin t_{1} \dots \sin t_{N-2} \cos t_{N-1} \times \cos t_{N-1} J_{N-1}$   
=  $r \sin t_{1} \dots \sin t_{N-2} (\sin^{2} t_{N-1} + \cos^{2} t_{N-1}) J_{N-1}$   
=  $r \sin t_{1} \dots \sin t_{N-2} J_{N-1}$ 

Thus, we obtain the formula in the statement, by recurrence.

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# 5c. Normal laws

As a main application of all the above, we can compute the Gauss integral:

THEOREM 5.23. We have the following formula,

$$\int_{\mathbb{R}} e^{-x^2} dx = \sqrt{\pi}$$

called Gauss integral formula.

**PROOF.** Let I be the above integral. By using polar coordinates, we obtain:

$$I^{2} = \int_{\mathbb{R}} \int_{\mathbb{R}} e^{-x^{2}-y^{2}} dx dy$$
$$= \int_{0}^{2\pi} \int_{0}^{\infty} e^{-r^{2}} r dr dt$$
$$= 2\pi \int_{0}^{\infty} \left(-\frac{e^{-r^{2}}}{2}\right)' dr$$
$$= 2\pi \left[0 - \left(-\frac{1}{2}\right)\right]$$
$$= \pi$$

Thus, we are led to the formula in the statement.

As a main application of the Gauss formula, we can now formulate:

**DEFINITION** 5.24. The normal law of parameter 1 is the following measure:

$$g_1 = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

More generally, the normal law of parameter t > 0 is the following measure:

$$g_t = \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} dx$$

These are also called Gaussian distributions, with "g" standing for Gauss.

These laws are traditionally denoted  $\mathcal{N}(0,1)$  and  $\mathcal{N}(0,t)$ , but since we will be doing here all kinds of probability, namely real and complex, discrete and continuous, and so on, we will have to deal with lots of laws, and we will be using simplified notations for them. Let us mention as well that the normal laws traditionally have 2 parameters, the mean and the variance. Here we do not need the mean, all our theory using centered laws.

As a second remark, the above laws have indeed mass 1, as they should. This follows indeed from the Gauss formula, which gives, with  $x = \sqrt{2t} y$ :

$$\int_{\mathbb{R}} e^{-x^2/2t} dx = \int_{\mathbb{R}} e^{-y^2} \sqrt{2t} \, dy$$
$$= \sqrt{2t} \int_{\mathbb{R}} e^{-y^2} dy$$
$$= \sqrt{2t} \times \sqrt{\pi}$$
$$= \sqrt{2\pi t}$$

Generally speaking, the normal laws appear as bit everywhere, in real life. The reasons behind this phenomenon come from the Central Limit Theorem (CLT), that we will explain in a moment. At the level of basic facts now, we have:

THEOREM 5.25. The moments of the normal law are the numbers

$$M_k(q_t) = t^{k/2} \times k!!$$

where the double factorials are by definition given by

$$k!! = 1 \cdot 3 \cdot 5 \dots (k-1)$$

with the convention k!! = 0 when k is odd.

**PROOF.** We have the following computation:

$$M_{k} = \frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} x^{k} e^{-x^{2}/2t} dx$$
  
$$= \frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} (tx^{k-1}) \left(-e^{-x^{2}/2t}\right)' dx$$
  
$$= \frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} t(k-1)x^{k-2} e^{-x^{2}/2t} dx$$
  
$$= t(k-1) \times \frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} x^{k-2} e^{-x^{2}/2t} dx$$
  
$$= t(k-1)M_{k-2}$$

On the other hand, we have  $M_0 = 1$ ,  $M_1 = 0$ . Thus by recurrence, the even moments vanish, and the odd moments are given by the formula in the statement.

We can improve the above result, as follows:

THEOREM 5.26. The moments of the normal law are the numbers

$$M_k(g_t) = t^{k/2} |P_2(k)|$$

where  $P_2(k)$  is the set of pairings of  $\{1, \ldots, k\}$ .

**PROOF.** Let us count the pairings of  $\{1, \ldots, k\}$ . In order to have such a pairing, we must pair 1 with one of  $2, \ldots, k$ , and then use a pairing of the remaining k - 2 points. Thus, we have the following recurrence formula:

$$|P_2(k)| = (k-1)|P_2(k-2)|$$

Thus  $|P_2(k)| = k!!$ , and we are led to the conclusion in the statement.

We are done done yet, and here is one more improvement:

THEOREM 5.27. The moments of the normal law are the numbers

$$M_k(g_t) = \sum_{\pi \in P_2(k)} t^{|\pi|}$$

where  $P_2(k)$  is the set of pairings of  $\{1, \ldots, k\}$ , and |.| is the number of blocks.

**PROOF.** This follows indeed from Theorem 5.26 above, because the number of blocks of a pairing of  $\{1, \ldots, k\}$  is trivially k/2, independently of the pairing.

We will see later on that many other interesting probability distributions are subject to similar formulae regarding their moments, involving partitions, and a lot of interesting combinatorics. Discussing this will be in fact a main theme of the present book.

123

### 5d. Central limits

The fundamental result in probability is the Central Limit Theorem (CLT), and our next task will be that of explaining this. Let us start with:

DEFINITION 5.28. Let X be a probability space, that is to say, a space with a probability measure, and with the corresponding integration denoted  $\mathbb{E}$ , and called expectation.

(1) The random variables are the real functions as follows:

$$f \in L^{\infty}(X)$$

(2) The moments of such a variable are the following numbers:

$$M_k(f) = \mathbb{E}(f^k)$$

(3) The law of such a variable is the measure given by:

$$M_k(f) = \int_{\mathbb{R}} x^k d\mu_f(x)$$

Here the fact that  $\mu_f$  exists indeed is not trivial. By linearity, we would like to have a real probability measure making hold the following formula, for any  $P \in \mathbb{R}[X]$ :

$$\mathbb{E}(P(f)) = \int_{\mathbb{R}} P(x) d\mu_f(x)$$

By using a continuity argument, it is enough to have this formula for the characteristic functions  $\chi_I$  of the arbitrary measurable sets of real numbers  $I \subset \mathbb{R}$ :

$$\mathbb{E}(\chi_I(f)) = \int_{\mathbb{R}} \chi_I(x) d\mu_f(x)$$

Thus, we would like to have a measure  $\mu_f$  such that:

$$\mathbb{P}(f \in I) = \mu_f(I)$$

But this latter formula can serve as a definition for  $\mu_f$ , with the axioms of real probability measures being trivially satisfied, and so we are done.

Next in line, we need to talk about independence. Once again with the idea of doing things a bit abstractly, the definition here is as follows:

DEFINITION 5.29. Two variables  $f, g \in L^{\infty}(X)$  are called independent when

$$\mathbb{E}(f^k g^l) = \mathbb{E}(f^k) \cdot \mathbb{E}(g^l)$$

happens, for any  $k, l \in \mathbb{N}$ .

Once again, this definition hides some non-trivial things. Indeed, by linearity, we would like to have a formula as follows, valid for any polynomials  $P, Q \in \mathbb{R}[X]$ :

$$\mathbb{E}(P(f)Q(g)) = \mathbb{E}(P(f)) \cdot \mathbb{E}(Q(g))$$

By continuity, it is enough to have this formula for characteristic functions  $\chi_I, \chi_J$  of the arbitrary measurable sets of real numbers  $I, J \subset \mathbb{R}$ :

$$\mathbb{E}(\chi_I(f)\chi_J(g)) = \mathbb{E}(\chi_I(f)) \cdot \mathbb{E}(\chi_J(g))$$

Thus, we are led to the usual definition of independence, namely:

$$\mathbb{P}(f \in I, g \in J) = \mathbb{P}(f \in I) \cdot \mathbb{P}(g \in J)$$

All this might seem a bit abstract, but in practice, the idea is of course that f, g must be independent, in an intuitive, real-life sense.

Here is now our first result, regarding this notion of independence:

**PROPOSITION 5.30.** Assuming that  $f, g \in L^{\infty}(X)$  are independent, we have

 $\mu_{f+g} = \mu_f * \mu_g$ 

where \* is the convolution of real probability measures.

**PROOF.** We have the following computation, using the independence of f, g:

$$M_k(f+g) = \mathbb{E}((f+g)^k)$$
  
=  $\sum_l \binom{k}{l} \mathbb{E}(f^l g^{k-l})$   
=  $\sum_l \binom{k}{l} M_l(f) M_{k-l}(g)$ 

On the other hand, by using the Fubini theorem, we have as well:

$$\int_{\mathbb{R}} x^k d(\mu_f * \mu_g)(x) = \int_{\mathbb{R} \times \mathbb{R}} (x+y)^k d\mu_f(x) d\mu_g(y)$$
$$= \sum_l \binom{k}{l} \int_{\mathbb{R}} x^k d\mu_f(x) \int_{\mathbb{R}} y^l d\mu_g(y)$$
$$= \sum_l \binom{k}{l} M_l(f) M_{k-l}(g)$$

Thus the measures  $\mu_{f+g}$  and  $\mu_f * \mu_g$  have the same moments:

$$M_k(\mu_{f+g}) = M_k(\mu_f * \mu_g)$$

We conclude that these two measures coincide, as stated.

125

Here is now our second result, which is something more advanced, providing us with some efficient tools for the study of the independence:

Theorem 5.31. Assuming that  $f, g \in L^{\infty}(X)$  are independent, we have

$$F_{f+g} = F_f F_g$$

where  $F_f(x) = \mathbb{E}(e^{ixf})$  is the Fourier transform.

**PROOF.** We have the following computation, using Proposition 5.30 and Fubini:

$$F_{f+g}(x) = \int_{\mathbb{R}} e^{ixy} d\mu_{f+g}(y)$$
  
$$= \int_{\mathbb{R}} e^{ixy} d(\mu_f * \mu_g)(y)$$
  
$$= \int_{\mathbb{R} \times \mathbb{R}} e^{ix(y+z)} d\mu_f(y) d\mu_g(z)$$
  
$$= \int_{\mathbb{R}} e^{ixy} d\mu_f(y) \int_{\mathbb{R}} e^{ixz} d\mu_g(z)$$
  
$$= F_f(x) F_g(x)$$

Thus, we are led to the conclusion in the statement.

Getting now to the normal laws, we have the following key result:

THEOREM 5.32. We have the following formula, valid for any t > 0:

$$F_{q_t}(x) = e^{-tx^2/2}$$

In particular, the normal laws satisfy  $g_s * g_t = g_{s+t}$ , for any s, t > 0.

**PROOF.** The Fourier transform formula can be established as follows:

$$F_{g_t}(x) = \frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} e^{-y^2/2t + ixy} dy$$
  
=  $\frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} e^{-(y/\sqrt{2t} - \sqrt{t/2}ix)^2 - tx^2/2} dy$   
=  $\frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} e^{-z^2 - tx^2/2} \sqrt{2t} dz$   
=  $\frac{1}{\sqrt{\pi}} e^{-tx^2/2} \int_{\mathbb{R}} e^{-z^2} dz$   
=  $\frac{1}{\sqrt{\pi}} e^{-tx^2/2} \cdot \sqrt{\pi}$   
=  $e^{-tx^2/2}$ 

As for the last assertion, this follows from the fact that  $\log F_{g_t}$  is linear in t.

#### 5E. EXERCISES

We are now ready to state and prove the CLT, as follows:

THEOREM 5.33 (CLT). Given random variables  $f_1, f_2, f_3, \ldots \in L^{\infty}(X)$  which are *i.i.d.*, centered, and with variance t > 0, we have, with  $n \to \infty$ , in moments,

$$\frac{1}{\sqrt{n}}\sum_{i=1}^n f_i \sim g_t$$

where  $g_t$  is the Gaussian law of parameter t, having as density  $\frac{1}{\sqrt{2\pi t}}e^{-y^2/2t}dy$ .

**PROOF.** We use the Fourier transform, which is by definition given by:

$$F_f(x) = \mathbb{E}(e^{ixf})$$

In terms of moments, we have the following formula:

$$F_f(x) = \mathbb{E}\left(\sum_{k=0}^{\infty} \frac{(ixf)^k}{k!}\right)$$
$$= \sum_{k=0}^{\infty} \frac{(ix)^k \mathbb{E}(f^k)}{k!}$$
$$= \sum_{k=0}^{\infty} \frac{i^k M_k(f)}{k!} x^k$$

Thus, the Fourier transform of the variable in the statement is:

$$F(x) = \left[F_f\left(\frac{x}{\sqrt{n}}\right)\right]^n$$
$$= \left[1 - \frac{tx^2}{2n} + O(n^{-2})\right]^n$$
$$\simeq \left[1 - \frac{tx^2}{2n}\right]^n$$
$$\sim e^{-tx^2/2}$$

But this latter function being the Fourier transform of  $g_t$ , we obtain the result.  $\Box$ 

This was for our basic presentation of calculus, and of standard probability theory, which was of course quite informal. We will be back to this, on several occasions.

# 5e. Exercises

There has been a lot of material in this chapter, and we have several exercises here, with the aim of completing the above. In what regards the functions of one variable, and more specifically the second derivative, the standard exercise here is:

EXERCISE 5.34. Given a convex function  $f : \mathbb{R} \to \mathbb{R}$ , prove that we have the following Jensen inequality, for any  $x_1, \ldots, x_N \in \mathbb{R}$ , and any  $\lambda_1, \ldots, \lambda_N > 0$  summing up to 1,

 $f(\lambda_1 x_1 + \ldots + \lambda_N x_N) \le \lambda_1 f(x_1) + \ldots + \lambda_N x_N$ 

with equality when  $x_1 = \ldots = x_N$ . In particular, by taking the weights  $\lambda_i$  to be all equal, we obtain the following Jensen inequality, valid for any  $x_1, \ldots, x_N \in \mathbb{R}$ ,

$$f\left(\frac{x_1 + \ldots + x_N}{N}\right) \le \frac{f(x_1) + \ldots + f(x_N)}{N}$$

and once again with equality when  $x_1 = \ldots = x_N$ . Prove also that a similar statement holds for the concave functions, with all the inequalities being reversed.

Observe that there is no derivative appearing in the above, so as a preliminary exercise, you have to clarify first the relation between convexity and the second derivative.

As a second exercise on the subject, which is very classical as well, we have:

EXERCISE 5.35. Prove that for  $p \in (1, \infty)$  we have the following Hölder inequality

$$\left|\frac{x_1 + \ldots + x_N}{N}\right|^p \le \frac{|x_1|^p + \ldots + |x_N|^p}{N}$$

and that for  $p \in (0,1)$  we have the following reverse Hölder inequality

$$\left|\frac{x_1 + \ldots + x_N}{N}\right|^p \ge \frac{|x_1|^p + \ldots + |x_N|^p}{N}$$

with in both cases equality precisely when  $|x_1| = \ldots = |x_N|$ .

As a bonus exercise here, try as well, directly, the case p = 2.

In what regards now integration, and several variables, we have:

EXERCISE 5.36. Work out with some further details the theory of integration and differentiation in several variables, by using linear algebra techniques.

This is something that we discussed, and the problem is that of getting beyond that.

Finally, in relation with probability theory, we have:

EXERCISE 5.37. Work out the precise convergence conclusions in the CLT, going beyond the convergence in moments, which was established in the above.

As before with the previous exercise, this is something long, and theoretical.

# CHAPTER 6

# Advanced calculus

# 6a. Volumes of spheres

We discuss here more advanced questions, related to the computation of the volumes of the spheres, and to the integration over the spheres. Before anything, we first have:

THEOREM 6.1. Assuming that the length of the unit circle is

$$L = 2\pi$$

it follows that the area of the unit disk is:

 $A = \pi$ 

PROOF. This follows by drawing polygons, and taking the  $N \to \infty$  limit. To be more precise, let us cut the disk as a pizza, in N slices, and leave aside the rounded parts. The area to be eaten can be then computed as follows, where H is the height of the slices, S is the length of their sides, and P = NS is the total length of the sides:

$$A = N \times \frac{1}{2} \times H \times S$$
$$= \frac{1}{2} \times H \times P$$
$$\simeq \frac{1}{2} \times 1 \times L$$
$$= \frac{2\pi}{2}$$
$$= \pi$$

Thus, we are led to the conclusion in the statement.

In general, we need spherical coordinates, in order to deal with such questions. Let us recall from the previous chapter that we have:

#### 6. ADVANCED CALCULUS

THEOREM 6.2. We have spherical coordinates in N dimensions,

$$\begin{cases} x_1 = r \cos t_1 \\ x_2 = r \sin t_1 \cos t_2 \\ \vdots \\ x_{N-1} = r \sin t_1 \sin t_2 \dots \sin t_{N-2} \cos t_{N-1} \\ x_N = r \sin t_1 \sin t_2 \dots \sin t_{N-2} \sin t_{N-1} \end{cases}$$

the corresponding Jacobian being given by the following formula:

$$J(r,t) = r^{N-1} \sin^{N-2} t_1 \sin^{N-3} t_2 \dots \sin^2 t_{N-3} \sin t_{N-2}$$

**PROOF.** This follows by developing the determinant over the last column, and then by proceeding by recurrence. Indeed, by developing, we have:

$$J_{N} = r \sin t_{1} \dots \sin t_{N-2} \sin t_{N-1} \times \sin t_{N-1} J_{N-1} + r \sin t_{1} \dots \sin t_{N-2} \cos t_{N-1} \times \cos t_{N-1} J_{N-1} = r \sin t_{1} \dots \sin t_{N-2} (\sin^{2} t_{N-1} + \cos^{2} t_{N-1}) J_{N-1} = r \sin t_{1} \dots \sin t_{N-2} J_{N-1}$$

Thus, we obtain the formula in the statement, by recurrence.

With the above formula in hand, let us compute now the volumes of spheres. For this purpose, we must understand how the products of coordinates integrate over spheres.

Let us start with the case N = 2. Here the sphere is the unit circle  $\mathbb{T}$ , and with  $z = e^{it}$  the coordinates are  $\cos t$ ,  $\sin t$ . We first integrate arbitrary powers of these coordinates:

**PROPOSITION 6.3.** We have the following formulae,

$$\int_0^{\pi/2} \cos^p t \, dt = \int_0^{\pi/2} \sin^p t \, dt = \left(\frac{\pi}{2}\right)^{\varepsilon(p)} \frac{p!!}{(p+1)!!}$$

where  $\varepsilon(p) = 1$  if p is even, and  $\varepsilon(p) = 0$  if p is odd, and where

$$m!! = (m-1)(m-3)(m-5)\dots$$

with the product ending at 2 if m is odd, and ending at 1 if m is even.

**PROOF.** Let us first compute the integral on the left in the statement:

$$I_p = \int_0^{\pi/2} \cos^p t \, dt$$

We do this by partial integration. We have the following formula:

$$(\cos^{p} t \sin t)' = p \cos^{p-1} t(-\sin t) \sin t + \cos^{p} t \cos t$$
  
=  $p \cos^{p+1} t - p \cos^{p-1} t + \cos^{p+1} t$   
=  $(p+1) \cos^{p+1} t - p \cos^{p-1} t$ 

By integrating between 0 and  $\pi/2$ , we obtain the following formula:

$$(p+1)I_{p+1} = pI_{p-1}$$

Thus we can compute  $I_p$  by recurrence, and we obtain:

$$I_{p} = \frac{p-1}{p} I_{p-2}$$

$$= \frac{p-1}{p} \cdot \frac{p-3}{p-2} I_{p-4}$$

$$= \frac{p-1}{p} \cdot \frac{p-3}{p-2} \cdot \frac{p-5}{p-4} I_{p-6}$$

$$\vdots$$

$$= \frac{p!!}{(p+1)!!} I_{1-\varepsilon(p)}$$

On the other hand, at p = 0 we have the following formula:

$$I_0 = \int_0^{\pi/2} 1 \, dt = \frac{\pi}{2}$$

Also, at p = 1 we have the following formula:

$$I_1 = \int_0^{\pi/2} \cos t \, dt = 1$$

Thus, we obtain the result, by recurrence. As for the second formula, regarding  $\sin t$ , this follows from the first formula, with the following change of variables:

$$t = \frac{\pi}{2} - s$$

Thus, we have proved both formulae in the statement.

We can compute the volume of the sphere, as follows:

THEOREM 6.4. The volume of the unit sphere in  $\mathbb{R}^N$  is given by

$$V = \left(\frac{\pi}{2}\right)^{[N/2]} \frac{2^N}{(N+1)!!}$$

with the convention

$$N!! = (N-1)(N-3)(N-5)\dots$$

with the product ending at 2 if N is odd, and ending at 1 if N is even.

**PROOF.** Let us denote by  $B^+$  the positive part of the unit sphere, or rather unit ball B, obtained by cutting this unit ball in  $2^N$  parts. At the level of volumes, we have:

$$V = 2^N V^+$$

#### 6. ADVANCED CALCULUS

We have the following computation, using spherical coordinates, and Fubini:

$$V^{+} = \int_{B^{+}}^{V^{+}} 1$$

$$= \int_{0}^{1} \int_{0}^{\pi/2} \dots \int_{0}^{\pi/2} r^{N-1} \sin^{N-2} t_{1} \dots \sin t_{N-2} dr dt_{1} \dots dt_{N-1}$$

$$= \int_{0}^{1} r^{N-1} dr \int_{0}^{\pi/2} \sin^{N-2} t_{1} dt_{1} \dots \int_{0}^{\pi/2} \sin t_{N-2} dt_{N-2} \int_{0}^{\pi/2} 1 dt_{N-1}$$

$$= \frac{1}{N} \times \left(\frac{\pi}{2}\right)^{[N/2]} \times \frac{(N-2)!!}{(N-1)!!} \cdot \frac{(N-3)!!}{(N-2)!!} \dots \frac{2!!}{3!!} \cdot \frac{1!!}{2!!} \cdot 1$$

$$= \frac{1}{N} \times \left(\frac{\pi}{2}\right)^{[N/2]} \times \frac{1}{(N-1)!!}$$

Here we have used the following formula, for computing the exponent of  $\pi/2$ :

$$\varepsilon(0) + \varepsilon(1) + \varepsilon(2) + \dots + \varepsilon(N-2)$$
  
= 1 + 0 + 1 + \dots + \varepsilon(N-2)  
=  $\left[\frac{N-2}{2}\right] + 1$   
=  $\left[\frac{N}{2}\right]$ 

Thus, we obtain the formula in the statement.

There are many applications of the above formula, as we will see in what follows. As main particular cases of the above formula, we have:

THEOREM 6.5. The volumes of the low-dimensional spheres are as follows:

- (1) At N = 1, the length of the unit interval is V = 2.
- (2) At N = 2, the area of the unit disk is  $V = \pi$ .
- (3) At N = 3, the volume of the unit sphere is  $V = \frac{4\pi}{3}$
- (4) At N = 4, the volume of the corresponding unit sphere is  $V = \frac{\pi^2}{2}$ .

**PROOF.** Some of these results are well-known, but we can obtain all of them as particular cases of the general formula in Theorem 6.4, as follows:

- (1) At N = 1 we obtain  $V = 1 \cdot \frac{2}{1} = 2$ .
- (2) At N = 2 we obtain  $V = \frac{\pi}{2} \cdot \frac{4}{2} = \pi$ .

(3) At N = 3 we obtain  $V = \frac{\pi}{2} \cdot \frac{8}{3} = \frac{4\pi}{3}$ . (4) At N = 4 we obtain  $V = \frac{\pi^2}{4} \cdot \frac{16}{8} = \frac{\pi^2}{2}$ .

In order to obtain estimates for the volumes, in the large N limit, we can use:

THEOREM 6.6. We have the Stirling formula

$$N! \simeq \left(\frac{N}{e}\right)^N \sqrt{2\pi N}$$

valid in the  $N \to \infty$  limit.

**PROOF.** This is something quite tricky, the idea being to use the integral of the logarithm. Indeed, by using a Riemann sum, we have:

$$\int_{0}^{1} \log x \, dx \simeq \frac{1}{N} \sum_{k=1}^{N} \log\left(\frac{k}{N}\right)$$
$$= \frac{1}{N} \left(\sum_{k=1}^{N} \log k\right) - \log N$$
$$= \frac{\log N!}{N} - \log N$$

On the other hand, our integral can be explicitly computed, as follows:

$$\int_0^1 \log x \, dx = \left[ x \log x - x \right]_0^1$$
$$= -1 - 0$$
$$= -1$$

We are therefore led to the following formula:

$$\frac{\log N!}{N} \simeq \log N - 1$$

Now by exponentiating, this leads to the following estimate:

$$\left(N!\right)^{1/N} \simeq \frac{N}{e}$$

With a bit more care, we obtain, a bit in the same way:

$$N! \sim \left(\frac{N}{e}\right)^N$$

The point now is that, with even more care, by reviewing the Riemann sums used above, we can obtain the formula in the statement.  $\hfill \Box$ 

With the above formula in hand, we have many useful applications, such as:

133

 $\square$ 

### 6. ADVANCED CALCULUS

**PROPOSITION 6.7.** We have the following estimate for binomial coefficients,

$$\binom{N}{K} \simeq \left(\frac{1}{t^t (1-t)^{1-t}}\right)^N \frac{1}{\sqrt{2\pi t (1-t)N}}$$

in the  $K \simeq tN \rightarrow \infty$  limit, with  $t \in (0,1]$ . In particular we have

$$\binom{2N}{N} \simeq \frac{4^N}{\sqrt{\pi N}}$$

in the  $N \to \infty$  limit, for the central binomial coefficients.

**PROOF.** All this is very standard, by using the Stirling formula etablished above, for the various factorials which appear, the idea being as follows:

(1) This follows from the definition of the binomial coefficients, namely:

$$\begin{pmatrix} N \\ K \end{pmatrix} = \frac{N!}{K!(N-K)!}$$

$$\simeq \left(\frac{N}{e}\right)^N \sqrt{2\pi N} \left(\frac{e}{K}\right)^K \frac{1}{\sqrt{2\pi K}} \left(\frac{e}{N-K}\right)^{N-K} \frac{1}{\sqrt{2\pi (N-K)}}$$

$$= \frac{N^N}{K^K (N-K)^{N-K}} \sqrt{\frac{N}{2\pi K (N-K)}}$$

$$\simeq \frac{N^N}{(tN)^{tN} ((1-t)N)^{(1-t)N}} \sqrt{\frac{N}{2\pi t N (1-t)N}}$$

$$= \left(\frac{1}{t^t (1-t)^{1-t}}\right)^N \frac{1}{\sqrt{2\pi t (1-t)N}}$$

Thus, we are led to the conclusion in the statement.

(2) This estimate follows from a similar computation, as follows:

$$\begin{pmatrix} 2N\\N \end{pmatrix} = \frac{(2N)!}{N!N!} \\ \simeq \left(\frac{2N}{e}\right)^{2N} \sqrt{4\pi N} \left(\frac{e}{N}\right)^{2N} \frac{1}{2\pi N} \\ = \frac{4^N}{\sqrt{\pi N}}$$

Alternatively, we can take t = 1/2 in (1), then rescale. Indeed, we have:

$$\binom{N}{[N/2]} \simeq \left(\frac{1}{(\frac{1}{2})^{1/2}(\frac{1}{2})^{1/2}}\right)^N \frac{1}{\sqrt{2\pi \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot N}}$$
$$= 2^N \sqrt{\frac{2}{\pi N}}$$

Thus with the change  $N \to 2N$  we obtain the formula in the statement.

Summarizing, we have so far complete estimate for the factorials. Regarding now the double factorials, that we will need as well, the result here is as follows:

**PROPOSITION 6.8.** We have the following estimate for the double factorials,

$$N!! \simeq \left(\frac{N}{e}\right)^{N/2} C$$

with  $C = \sqrt{2}$  for N even, and  $C = \sqrt{\pi}$  for N odd. Alternatively, we have

$$(N+1)!! \simeq \left(\frac{N}{e}\right)^{N/2} D$$

with  $D = \sqrt{\pi N}$  for N even, and  $D = \sqrt{2N}$  for N odd.

**PROOF.** Once again this is standard, the idea being as follows:

(1) When N = 2K is even, we have the following computation:

$$N!! = (2K - 1)(2K - 3) \dots 1$$
  
=  $\frac{(2K)!}{2^{K}K!}$   
 $\simeq \frac{1}{2^{K}} \left(\frac{2K}{e}\right)^{2K} \sqrt{4\pi K} \left(\frac{e}{K}\right)^{K} \frac{1}{\sqrt{2\pi K}}$   
=  $\left(\frac{2K}{e}\right)^{K} \sqrt{2}$   
=  $\left(\frac{N}{e}\right)^{N/2} \sqrt{2}$ 

# 6. ADVANCED CALCULUS

(2) When N = 2K + 1 is odd, we have the following computation:

$$N!! = (2K)(2K - 2) \dots 2$$
  
=  $2^{K}K!$   
 $\simeq \left(\frac{2K}{e}\right)^{K}\sqrt{2\pi K}$   
=  $\left(\frac{2K+1}{e}\right)^{K+1/2}\sqrt{\frac{e}{2K+1}}\left(\frac{2K}{2K+1}\right)^{K}\sqrt{2\pi K}$   
 $\simeq \left(\frac{N}{e}\right)^{N/2}\sqrt{\frac{e}{2K}} \cdot \frac{1}{\sqrt{e}} \cdot \sqrt{2\pi K}$   
=  $\left(\frac{N}{e}\right)^{N/2}\sqrt{\pi}$ 

(3) Back to the case where N = 2K is even, by using (2) we obtain:

$$(N+1)!! \simeq \left(\frac{N+1}{e}\right)^{(N+1)/2} \sqrt{\pi}$$

$$= \left(\frac{N+1}{e}\right)^{N/2} \sqrt{\frac{N+1}{e}} \cdot \sqrt{\pi}$$

$$= \left(\frac{N}{e}\right)^{N/2} \left(\frac{N+1}{N}\right)^{N/2} \sqrt{\frac{N+1}{e}} \cdot \sqrt{\pi}$$

$$\simeq \left(\frac{N}{e}\right)^{N/2} \sqrt{e} \cdot \sqrt{\frac{N}{e}} \cdot \sqrt{\pi}$$

$$= \left(\frac{N}{e}\right)^{N/2} \sqrt{\pi N}$$

(4) Finally, back to the case where N = 2K + 1 is odd, by using (1) we obtain:

$$(N+1)!! \simeq \left(\frac{N+1}{e}\right)^{(N+1)/2} \sqrt{2}$$

$$= \left(\frac{N+1}{e}\right)^{N/2} \sqrt{\frac{N+1}{e}} \cdot \sqrt{2}$$

$$= \left(\frac{N}{e}\right)^{N/2} \left(\frac{N+1}{N}\right)^{N/2} \sqrt{\frac{N+1}{e}} \cdot \sqrt{2}$$

$$\simeq \left(\frac{N}{e}\right)^{N/2} \sqrt{e} \cdot \sqrt{\frac{N}{e}} \cdot \sqrt{2}$$

$$= \left(\frac{N}{e}\right)^{N/2} \sqrt{2N}$$

Thus, we have proved the estimates in the statement.

We can now estimate the volumes of the spheres, as follows:

THEOREM 6.9. The volume of the unit sphere in  $\mathbb{R}^N$  is given by

$$V \simeq \left(\frac{2\pi e}{N}\right)^{N/2} \frac{1}{\sqrt{\pi N}}$$

in the  $N \to \infty$  limit.

**PROOF.** We use the formula for V found in Theorem 6.4, namely:

$$V = \left(\frac{\pi}{2}\right)^{[N/2]} \frac{2^N}{(N+1)!!}$$

In the case where N is even, the estimate goes as follows:

$$V = \left(\frac{\pi}{2}\right)^{N/2} \frac{2^N}{(N+1)!!}$$
$$\simeq \left(\frac{\pi}{2}\right)^{N/2} 2^N \left(\frac{e}{N}\right)^{N/2} \frac{1}{\sqrt{\pi N}}$$
$$= \left(\frac{2\pi e}{N}\right)^{N/2} \frac{1}{\sqrt{\pi N}}$$

#### 6. ADVANCED CALCULUS

In the case where N is odd, the estimate goes as follows:

$$V = \left(\frac{\pi}{2}\right)^{(N-1)/2} \frac{2^N}{(N+1)!!}$$
$$\simeq \left(\frac{\pi}{2}\right)^{(N-1)/2} 2^N \left(\frac{e}{N}\right)^{N/2} \frac{1}{\sqrt{2N}}$$
$$= \sqrt{\frac{2}{\pi}} \left(\frac{2\pi e}{N}\right)^{N/2} \frac{1}{\sqrt{2N}}$$
$$= \left(\frac{2\pi e}{N}\right)^{N/2} \frac{1}{\sqrt{\pi N}}$$

Thus, we are led to the uniform formula in the statement.

Getting back now to our main result so far, Theorem 6.4 above, that result computes the volume of the sphere, or rather unit ball, to use the standard terminology, in  $\mathbb{R}^N$ .

We can compute in the same way the area of the sphere, the result being as follows:

THEOREM 6.10. The area of the unit sphere in  $\mathbb{R}^N$  is given by

$$A = \left(\frac{\pi}{2}\right)^{[N/2]} \frac{2^N}{(N-1)!!}$$

with the our usual convention for double factorials, namely:

$$N!! = (N-1)(N-3)(N-5)\dots$$

In particular, at N = 2, 3, 4 we obtain respectively the following areas,

$$A = 2\pi, 4\pi, 2\pi^2$$

and we have as well asymptotic estimates for A, in the  $N \to \infty$  limit.

**PROOF.** Regarding the first assertion, there is no need to compute again, because the formula in the statement can be deduced from Theorem 6.4, as follows:

(1) We can either use the slicing argument from the proof of Theorem 6.1, which shows that the area and volume of the sphere in  $\mathbb{R}^N$  are related by the following formula:

$$A = N \cdot V$$

Together with the formula in Theorem 6.4 for V, this gives the result.

(2) Or, we can start the computation in the same way as we started the proof of Theorem 6.4, the beginning of this computation being as follows, with the convention

138

that  $S^+$  denotes the part of the sphere where all the coordinates are positive:

$$vol(S^{+}) = \int_{S^{+}} 1$$
  
=  $\int_{0}^{\pi/2} \dots \int_{0}^{\pi/2} \sin^{N-2} t_{1} \dots \sin t_{N-2} dt_{1} \dots dt_{N-1}$ 

Now by comparing with the beginning of the proof of Theorem 6.4, the only thing that changes is the following quantity, which now dissapears:

$$\int_0^1 r^{N-1} \, dr = \frac{1}{N}$$

Thus, we have  $vol(S^+) = N \cdot vol(B^+)$ , and so:

$$vol(S) = N \cdot vol(B)$$

But this means  $A = N \cdot V$ , and together with the formula in Theorem 6.4 for V, this gives the result. As for the last assertion, this can be either worked out directly, or deduced from the results for volumes that we have so far, by multiplying by N.

# 6b. Spherical integrals

Let us discuss now the computation of the arbitrary integrals over the sphere. We will need a technical result extending Proposition 6.3 above, as follows:

THEOREM 6.11. We have the following formula,

$$\int_0^{\pi/2} \cos^p t \sin^q t \, dt = \left(\frac{\pi}{2}\right)^{\varepsilon(p)\varepsilon(q)} \frac{p!!q!!}{(p+q+1)!!}$$

where  $\varepsilon(p) = 1$  if p is even, and  $\varepsilon(p) = 0$  if p is odd, and where

$$m!! = (m-1)(m-3)(m-5)\dots$$

with the product ending at 2 if m is odd, and ending at 1 if m is even.

**PROOF.** Let  $I_{pq}$  be the integral in the statement. In order to do the partial integration, observe that we have:

$$(\cos^{p} t \sin^{q} t)' = p \cos^{p-1} t (-\sin t) \sin^{q} t + \cos^{p} t \cdot q \sin^{q-1} t \cos t = -p \cos^{p-1} t \sin^{q+1} t + q \cos^{p+1} t \sin^{q-1} t$$

By integrating between 0 and  $\pi/2$ , we obtain, for p, q > 0:

$$pI_{p-1,q+1} = qI_{p+1,q-1}$$

### 6. ADVANCED CALCULUS

Thus, we can compute  $I_{pq}$  by recurrence. When q is even we have:

$$I_{pq} = \frac{q-1}{p+1} I_{p+2,q-2}$$
  
=  $\frac{q-1}{p+1} \cdot \frac{q-3}{p+3} I_{p+4,q-4}$   
=  $\frac{q-1}{p+1} \cdot \frac{q-3}{p+3} \cdot \frac{q-5}{p+5} I_{p+6,q-6}$   
= :  
=  $\frac{p!!q!!}{(p+q)!!} I_{p+q}$ 

But the last term comes from Proposition 6.3, and we obtain the result:

$$I_{pq} = \frac{p!!q!!}{(p+q)!!} I_{p+q}$$
  
=  $\frac{p!!q!!}{(p+q)!!} \left(\frac{\pi}{2}\right)^{\varepsilon(p+q)} \frac{(p+q)!!}{(p+q+1)!!}$   
=  $\left(\frac{\pi}{2}\right)^{\varepsilon(p)\varepsilon(q)} \frac{p!!q!!}{(p+q+1)!!}$ 

Observe that this gives the result for p even as well, by symmetry. Indeed, we have  $I_{pq} = I_{qp}$ , by using the following change of variables:

$$t = \frac{\pi}{2} - s$$

In the remaining case now, where both p, q are odd, we can use once again the formula  $pI_{p-1,q+1} = qI_{p+1,q-1}$  established above, and the recurrence goes as follows:

$$I_{pq} = \frac{q-1}{p+1} I_{p+2,q-2}$$
  
=  $\frac{q-1}{p+1} \cdot \frac{q-3}{p+3} I_{p+4,q-4}$   
=  $\frac{q-1}{p+1} \cdot \frac{q-3}{p+3} \cdot \frac{q-5}{p+5} I_{p+6,q-6}$   
= :  
=  $\frac{p!!q!!}{(p+q-1)!!} I_{p+q-1,1}$ 

In order to compute the last term, observe that we have:

$$I_{p1} = \int_{0}^{\pi/2} \cos^{p} t \sin t \, dt$$
$$= -\frac{1}{p+1} \int_{0}^{\pi/2} (\cos^{p+1} t)' \, dt$$
$$= \frac{1}{p+1}$$

Thus, we can finish our computation in the case p, q odd, as follows:

$$I_{pq} = \frac{p!!q!!}{(p+q-1)!!} I_{p+q-1,1}$$
$$= \frac{p!!q!!}{(p+q-1)!!} \cdot \frac{1}{p+q}$$
$$= \frac{p!!q!!}{(p+q+1)!!}$$

Thus, we obtain the formula in the statement, the exponent of  $\pi/2$  appearing there being  $\varepsilon(p)\varepsilon(q) = 0 \cdot 0 = 0$  in the present case, and this finishes the proof.

We can now integrate over the spheres, as follows:

THEOREM 6.12. The polynomial integrals over the unit sphere  $S_{\mathbb{R}}^{N-1} \subset \mathbb{R}^N$ , with respect to the normalized, mass 1 measure, are given by the following formula,

$$\int_{S_{\mathbb{R}}^{N-1}} x_1^{k_1} \dots x_N^{k_N} \, dx = \frac{(N-1)!!k_1!!\dots k_N!!}{(N+\Sigma k_i - 1)!!}$$

valid when all exponents  $k_i$  are even. If an exponent is odd, the integral vanishes.

**PROOF.** Assume first that one of the exponents  $k_i$  is odd. We can make then the following change of variables, which shows that the integral in the statement vanishes:

$$x_i \to -x_i$$

Assume now that all the exponents  $k_i$  are even. As a first observation, the result holds indeed at N = 2, due to the formula from Theorem 6.11 above, which reads:

$$\int_0^{\pi/2} \cos^p t \sin^q t \, dt = \left(\frac{\pi}{2}\right)^{\varepsilon(p)\varepsilon(q)} \frac{p!!q!!}{(p+q+1)!!}$$
$$= \frac{p!!q!!}{(p+q+1)!!}$$

Indeed, this formula computes the integral in the statement over the first quadrant. But since the exponents  $p, q \in \mathbb{N}$  are assumed to be even, the integrals over the other quadrants are given by the same formula, so when averaging we obtain the result.

### 6. ADVANCED CALCULUS

In the general case now, where the dimension  $N \in \mathbb{N}$  is arbitrary, the integral in the statement can be written in spherical coordinates, as follows:

$$I = \frac{2^N}{A} \int_0^{\pi/2} \dots \int_0^{\pi/2} x_1^{k_1} \dots x_N^{k_N} J \, dt_1 \dots dt_{N-1}$$

Here A is the area of the sphere, J is the Jacobian, and the  $2^N$  factor comes from the restriction to the  $1/2^N$  part of the sphere where all the coordinates are positive. According to Theorem 6.10, the normalization constant in front of the integral is:

$$\frac{2^N}{A} = \left(\frac{2}{\pi}\right)^{[N/2]} (N-1)!!$$

As for the unnormalized integral, this is given by:

$$I' = \int_0^{\pi/2} \dots \int_0^{\pi/2} (\cos t_1)^{k_1} (\sin t_1 \cos t_2)^{k_2}$$
  

$$\vdots$$
  

$$(\sin t_1 \sin t_2 \dots \sin t_{N-2} \cos t_{N-1})^{k_{N-1}}$$
  

$$(\sin t_1 \sin t_2 \dots \sin t_{N-2} \sin t_{N-1})^{k_N}$$
  

$$\sin^{N-2} t_1 \sin^{N-3} t_2 \dots \sin^2 t_{N-3} \sin t_{N-2}$$
  

$$dt_1 \dots dt_{N-1}$$

By rearranging the terms, we obtain:

$$I' = \int_{0}^{\pi/2} \cos^{k_{1}} t_{1} \sin^{k_{2}+...+k_{N}+N-2} t_{1} dt_{1}$$
$$\int_{0}^{\pi/2} \cos^{k_{2}} t_{2} \sin^{k_{3}+...+k_{N}+N-3} t_{2} dt_{2}$$
$$\vdots$$
$$\int_{0}^{\pi/2} \cos^{k_{N-2}} t_{N-2} \sin^{k_{N-1}+k_{N}+1} t_{N-2} dt_{N-2}$$
$$\int_{0}^{\pi/2} \cos^{k_{N-1}} t_{N-1} \sin^{k_{N}} t_{N-1} dt_{N-1}$$

Now by using the above-mentioned formula at N = 2, this gives:

$$I' = \frac{k_1!!(k_2 + \ldots + k_N + N - 2)!!}{(k_1 + \ldots + k_N + N - 1)!!} \left(\frac{\pi}{2}\right)^{\varepsilon(N-2)}$$

$$\frac{k_2!!(k_3 + \ldots + k_N + N - 3)!!}{(k_2 + \ldots + k_N + N - 2)!!} \left(\frac{\pi}{2}\right)^{\varepsilon(N-3)}$$

$$\vdots$$

$$\frac{k_{N-2}!!(k_{N-1} + k_N + 1)!!}{(k_{N-2} + k_{N-1} + l_N + 2)!!} \left(\frac{\pi}{2}\right)^{\varepsilon(1)}$$

$$\frac{k_{N-1}!!k_N!!}{(k_{N-1} + k_N + 1)!!} \left(\frac{\pi}{2}\right)^{\varepsilon(0)}$$

Now let F be the part involving the double factorials, and P be the part involving the powers of  $\pi/2$ , so that  $I' = F \cdot P$ . Regarding F, by cancelling terms we have:

$$F = \frac{k_1 !! \dots k_N !!}{(\Sigma k_i + N - 1)!!}$$

As in what regards P, by summing the exponents, we obtain  $P = \left(\frac{\pi}{2}\right)^{[N/2]}$ . We can now put everything together, and we obtain:

$$I = \frac{2^{N}}{A} \times F \times P$$
  
=  $\left(\frac{2}{\pi}\right)^{[N/2]} (N-1)!! \times \frac{k_{1}!! \dots k_{N}!!}{(\Sigma k_{i} + N - 1)!!} \times \left(\frac{\pi}{2}\right)^{[N/2]}$   
=  $\frac{(N-1)!!k_{1}!! \dots k_{N}!!}{(\Sigma k_{i} + N - 1)!!}$ 

Thus, we are led to the conclusion in the statement.

We have the following useful generalization of the above formula:

THEOREM 6.13. We have the following integration formula over the sphere  $S_{\mathbb{R}}^{N-1} \subset \mathbb{R}^N$ , with respect to the normalized measure, valid for any exponents  $k_i \in \mathbb{N}$ ,

$$\int_{S_{\mathbb{R}}^{N-1}} |x_1^{k_1} \dots x_N^{k_N}| \, dx = \left(\frac{2}{\pi}\right)^{\Sigma(k_1,\dots,k_N)} \frac{(N-1)!!k_1!!\dots k_N!!}{(N+\Sigma k_i-1)!!}$$

with  $\Sigma = [odds/2]$  if N is odd and  $\Sigma = [(odds + 1)/2]$  if N is even, where "odds" denotes the number of odd numbers in the sequence  $k_1, \ldots, k_N$ .

143
## 6. ADVANCED CALCULUS

**PROOF.** As before, the formula holds indeed at N = 2, due to Theorem 6.11. In general, the integral in the statement can be written in spherical coordinates, as follows:

$$I = \frac{2^N}{A} \int_0^{\pi/2} \dots \int_0^{\pi/2} x_1^{k_1} \dots x_N^{k_N} J \, dt_1 \dots dt_{N-1}$$

Here A is the area of the sphere, J is the Jacobian, and the  $2^N$  factor comes from the restriction to the  $1/2^N$  part of the sphere where all the coordinates are positive. The normalization constant in front of the integral is, as before:

$$\frac{2^N}{A} = \left(\frac{2}{\pi}\right)^{[N/2]} (N-1)!!$$

As for the unnormalized integral, this can be written as before, as follows:

$$I' = \int_{0}^{\pi/2} \cos^{k_{1}} t_{1} \sin^{k_{2}+\ldots+k_{N}+N-2} t_{1} dt_{1}$$
$$\int_{0}^{\pi/2} \cos^{k_{2}} t_{2} \sin^{k_{3}+\ldots+k_{N}+N-3} t_{2} dt_{2}$$
$$\vdots$$
$$\int_{0}^{\pi/2} \cos^{k_{N-2}} t_{N-2} \sin^{k_{N-1}+k_{N}+1} t_{N-2} dt_{N-2}$$
$$\int_{0}^{\pi/2} \cos^{k_{N-1}} t_{N-1} \sin^{k_{N}} t_{N-1} dt_{N-1}$$

Now by using the formula at N = 2, we get:

$$I' = \frac{\pi}{2} \cdot \frac{k_1!!(k_2 + \ldots + k_N + N - 2)!!}{(k_1 + \ldots + k_N + N - 1)!!} \left(\frac{2}{\pi}\right)^{\delta(k_1, k_2 + \ldots + k_N + N - 2)} \\ \frac{\pi}{2} \cdot \frac{k_2!!(k_3 + \ldots + k_N + N - 3)!!}{(k_2 + \ldots + k_N + N - 2)!!} \left(\frac{2}{\pi}\right)^{\delta(k_2, k_3 + \ldots + k_N + N - 3)} \\ \vdots \\ \frac{\pi}{2} \cdot \frac{k_{N-2}!!(k_{N-1} + k_N + 1)!!}{(k_{N-2} + k_{N-1} + k_N + 2)!!} \left(\frac{2}{\pi}\right)^{\delta(k_{N-2}, k_{N-1} + k_N + 1)} \\ \frac{\pi}{2} \cdot \frac{k_{N-1}!!k_N!!}{(k_{N-1} + k_N + 1)!!} \left(\frac{2}{\pi}\right)^{\delta(k_{N-1}, k_N)}$$

In order to compute this quantity, let us denote by F the part involving the double factorials, and by P the part involving the powers of  $\pi/2$ , so that we have:

$$I' = F \cdot P$$

Regarding F, there are many cancellations there, and we end up with:

$$F = \frac{k_1!!\dots k_N!!}{(\Sigma k_i + N - 1)!!}$$

As in what regards P, the  $\delta$  exponents on the right sum up to the following number:

$$\Delta(k_1, \dots, k_N) = \sum_{i=1}^{N-1} \delta(k_i, k_{i+1} + \dots + k_N + N - i - 1)$$

In other words, with this notation, the above formula reads:

$$I' = \left(\frac{\pi}{2}\right)^{N-1} \frac{k_1!!k_2!!\dots k_N!!}{(k_1 + \dots + k_N + N - 1)!!} \left(\frac{2}{\pi}\right)^{\Delta(k_1,\dots,k_N)}$$
$$= \left(\frac{2}{\pi}\right)^{\Delta(k_1,\dots,k_N) - N + 1} \frac{k_1!!k_2!!\dots k_N!!}{(k_1 + \dots + k_N + N - 1)!!}$$
$$= \left(\frac{2}{\pi}\right)^{\Sigma(k_1,\dots,k_N) - [N/2]} \frac{k_1!!k_2!!\dots k_N!!}{(k_1 + \dots + k_N + N - 1)!!}$$

Here the formula relating  $\Delta$  to  $\Sigma$  follows from a number of simple observations, the first of which being the act that, due to obvious parity reasons, the sequence of  $\delta$  numbers appearing in the definition of  $\Delta$  cannot contain two consecutive zeroes. Together with  $I = (2^N/V)I'$ , this gives the formula in the statement.

Summarizing, we have complete results for the integration over the spheres, with the answers involving various multinomial type coefficients, defined in terms of factorials, or of double factorials. All these formulae are of course very useful, in practice.

# 6c. Hyperspherical laws

As a first application of all this, we have the following result:

THEOREM 6.14. The moments of the hyperspherical variables are

$$\int_{S_{\mathbb{R}}^{N-1}} x_i^k dx = \frac{(N-1)!!k!!}{(N+k-1)!!}$$

and the normalized hyperspherical variables

$$y_i = \frac{x_i}{\sqrt{N}}$$

become normal and independent with  $N \to \infty$ .

**PROOF.** We have two things to be proved, the idea being as follows:

#### 6. ADVANCED CALCULUS

(1) The formula in the statement follows from the general integration formula over the sphere, established above, which is as follows:

$$\int_{S^{N-1}} x_{i_1} \dots x_{i_k} \, dx = \frac{(N-1)!!l_1!! \dots l_N!!}{(N+\sum l_i-1)!!}$$

Indeed, with  $i_1 = \ldots = i_k = i$ , we obtain from this:

$$\int_{S^{N-1}} x_i^k dx = \frac{(N-1)!!k!!}{(N+k-1)!!}$$

Now observe that with  $N \to \infty$  we have the following estimate:

$$\int_{S^{N-1}} x_i^k dx = \frac{(N-1)!!}{(N+k-1)!!} \times k!!$$
  

$$\simeq N^{k/2} k!!$$
  

$$= N^{k/2} M_k(g_1)$$

Thus, the variables  $y_i = \frac{x_i}{\sqrt{N}}$  become normal with  $N \to \infty$ .

(2) As for the asymptotic independence result, this is standard as well, once again by using Theorem 6.12, for computing mixed moments, and taking the  $N \to \infty$  limit.  $\Box$ 

The above results are very interesting for a variety of reasons, and we will be back to them on several occasions, in what follows.

# 6d. Complex spheres

Let us discuss now the complex analogues of all the above. We must first introduce the complex analogues of the normal laws, and this can be done as follows:

DEFINITION 6.15. The complex Gaussian law of parameter t > 0 is

$$G_t = law\left(\frac{1}{\sqrt{2}}(a+ib)\right)$$

where a, b are independent, each following the law  $g_t$ .

The combinatorics here is a bit more complicated than in the real case, and we will be back to this in a moment, notably with a formula for the moments of such variables.

As in the real case, these measures form convolution semigroups:

**THEOREM 6.16.** The complex Gaussian laws have the property

$$G_s * G_t = G_{s+t}$$

for any s, t > 0, and so they form a convolution semigroup.

PROOF. This follows indeed from the real result, established above, by taking real and imaginary parts.  $\hfill \Box$ 

We have the following complex analogue of the CLT:

THEOREM 6.17 (CLT). Given complex random variables

$$f_1, f_2, f_3, \ldots \in L^{\infty}(X)$$

which are i.i.d., centered, and with variance t > 0, we have, with  $n \to \infty$ , in moments,

$$\frac{1}{\sqrt{n}}\sum_{i=1}^n f_i \sim G_t$$

where  $G_t$  is the complex Gaussian law of parameter t.

PROOF. This follows indeed from the real CLT, established above, by taking real and imaginary parts.  $\hfill \Box$ 

Regarding now the moments, things are a bit more complicated than before, because our variables are now complex instead of real. In order to deal with this issue, we will use "colored moments", which are the expectations of the "colored powers", with these latter powers being defined by the following formulae, and multiplicativity:

$$f^{\emptyset} = 1$$
 ,  $f^{\circ} = f$  ,  $f^{\bullet} = \bar{f}$ 

With these conventions made, the result is as follows:

THEOREM 6.18. The moments of the complex normal law are the numbers

$$M_k(G_t) = \sum_{\pi \in \mathcal{P}_2(k)} t^{|\pi|}$$

where  $\mathcal{P}_2(k)$  are the matching pairings of  $\{1, \ldots, k\}$ , and |.| is the number of blocks.

**PROOF.** This is something well-known, but quite technical, which can be done in several steps, as follows:

(1) We recall from the above that the moments of the real Gaussian law  $g_1$ , with respect to integer exponents  $k \in \mathbb{N}$ , are the following numbers:

$$m_k = |P_2(k)|$$

Numerically, we have the following formula, explained as well in section 5:

$$m_k = \begin{cases} k!! & (k \text{ even}) \\ 0 & (k \text{ odd}) \end{cases}$$

(2) We will show here that in what concerns the complex Gaussian law  $G_1$ , similar results hold. Numerically, we will prove that we have the following formula, where a

#### 6. ADVANCED CALCULUS

colored integer  $k = \circ \bullet \circ \circ \ldots$  is called uniform when it contains the same number of  $\circ$  and  $\bullet$ , and where  $|k| \in \mathbb{N}$  is the length of such a colored integer:

$$M_k = \begin{cases} (|k|/2)! & (k \text{ uniform}) \\ 0 & (k \text{ not uniform}) \end{cases}$$

Now since the matching partitions  $\pi \in \mathcal{P}_2(k)$  are counted by exactly the same numbers, and this for trivial reasons, we will obtain the formula in the statement, namely:

$$M_k = |\mathcal{P}_2(k)|$$

(3) This was for the plan. In practice now, we must compute the moments, with respect to colored integer exponents  $k = \circ \bullet \circ \circ \ldots$ , of the variable in the statement:

$$c = \frac{1}{\sqrt{2}}(a + ib)$$

As a first observation, in the case where such an exponent  $k = \circ \bullet \circ \circ \ldots$  is not uniform in  $\circ, \bullet$ , a rotation argument shows that the corresponding moment of c vanishes. To be more precise, the variable c' = wc can be shown to be complex Gaussian too, for any  $w \in \mathbb{C}$ , and from  $M_k(c) = M_k(c')$  we obtain  $M_k(c) = 0$ , in this case.

(4) In the uniform case now, where  $k = \circ \bullet \bullet \circ \ldots$  consists of p copies of  $\circ$  and p copies of  $\bullet$ , the corresponding moment can be computed as follows:

$$M_{k} = \int (c\bar{c})^{p}$$

$$= \frac{1}{2^{p}} \int (a^{2} + b^{2})^{p}$$

$$= \frac{1}{2^{p}} \sum_{s} {p \choose s} \int a^{2s} \int b^{2p-2s}$$

$$= \frac{1}{2^{p}} \sum_{s} {p \choose s} (2s)!! (2p - 2s)!!$$

$$= \frac{1}{2^{p}} \sum_{s} \frac{p!}{s!(p-s)!} \cdot \frac{(2s)!}{2^{s}s!} \cdot \frac{(2p - 2s)!}{2^{p-s}(p-s)!}$$

$$= \frac{p!}{4^{p}} \sum_{s} {2s \choose s} {2p - 2s \choose p-s}$$

(5) In order to finish now the computation, let us recall that we have the following formula, coming from the generalized binomial formula, or from the Taylor formula:

$$\frac{1}{\sqrt{1+t}} = \sum_{k=0}^{\infty} \binom{2k}{k} \left(\frac{-t}{4}\right)^k$$

By taking the square of this series, we obtain the following formula:

$$\frac{1}{1+t} = \sum_{ks} \binom{2k}{k} \binom{2s}{s} \left(\frac{-t}{4}\right)^{k+s}$$
$$= \sum_{p} \left(\frac{-t}{4}\right)^{p} \sum_{s} \binom{2s}{s} \binom{2p-2s}{p-s}$$

Now by looking at the coefficient of  $t^p$  on both sides, we conclude that the sum on the right equals  $4^p$ . Thus, we can finish the moment computation in (4), as follows:

$$M_p = \frac{p!}{4^p} \times 4^p = p!$$

(6) As a conclusion, if we denote by |k| the length of a colored integer  $k = \circ \bullet \circ \circ \ldots$ , the moments of the variable c in the statement are given by:

$$M_k = \begin{cases} (|k|/2)! & (k \text{ uniform}) \\ 0 & (k \text{ not uniform}) \end{cases}$$

On the other hand, the numbers  $|\mathcal{P}_2(k)|$  in the statement are given by exactly the same formula. Indeed, in order to have matching pairings of k, our exponent  $k = \circ \bullet \circ \circ \ldots$  must be uniform, consisting of p copies of  $\circ$  and p copies of  $\bullet$ , with:

$$p = \frac{|k|}{2}$$

But then the matching pairings of k correspond to the permutations of the  $\bullet$  symbols, as to be matched with  $\circ$  symbols, and so we have p! such matching pairings.

Thus, we have exactly the same formula as for the moments of the variable c, and we are led to the conclusion in the statement.

There are of course many other possible proofs for the above result, and some further theory as well, that can be developed for the complex normal variables. We will be back to this, on several occasions, in what follows.

In relation now with the spheres, we first have the following variation of the integration formula in Theorem 6.12, dealing this time with integrals over the complex sphere:

THEOREM 6.19. We have the following integration formula over the complex sphere  $S_{\mathbb{C}}^{N-1} \subset \mathbb{R}^N$ , with respect to the normalized measure,

$$\int_{S_{\mathbb{C}}^{N-1}} |z_1|^{2l_1} \dots |z_N|^{2l_N} dz = 4^{\sum l_i} \frac{(2N-1)! l_1! \dots l_n!}{(2N+\sum l_i-1)!}$$

valid for any exponents  $l_i \in \mathbb{N}$ . As for the other polynomial integrals in  $z_1, \ldots, z_N$  and their conjugates  $\bar{z}_1, \ldots, \bar{z}_N$ , these all vanish.

## 6. ADVANCED CALCULUS

PROOF. Consider an arbitrary polynomial integral over  $S_{\mathbb{C}}^{N-1}$ , written as follows:

$$I = \int_{S_{\mathbb{C}}^{N-1}} z_{i_1} \bar{z}_{i_2} \dots z_{i_{2l-1}} \bar{z}_{i_{2l}} \, dz$$

By using transformations of type  $p \to \lambda p$  with  $|\lambda| = 1$ , we see that I vanishes, unless each  $z_a$  appears as many times as  $\bar{z}_a$  does, and this gives the last assertion.

Assume now that we are in the non-vanishing case. Then the  $l_a$  copies of  $z_a$  and the  $l_a$  copies of  $\bar{z}_a$  produce by multiplication a factor  $|z_a|^{2l_a}$ , so we have:

$$I = \int_{S_{\mathbb{C}}^{N-1}} |z_1|^{2l_1} \dots |z_N|^{2l_N} \, dz$$

Now by using the standard identification  $S^{N-1}_{\mathbb{C}}\simeq S^{2N-1}_{\mathbb{R}},$  we obtain:

$$I = \int_{S_{\mathbb{R}}^{2N-1}} (x_1^2 + y_1^2)^{l_1} \dots (x_N^2 + y_N^2)^{l_N} d(x, y)$$
  
= 
$$\sum_{r_1 \dots r_N} {l_1 \choose r_1} \dots {l_N \choose r_N} \int_{S_{\mathbb{R}}^{2N-1}} x_1^{2l_1 - 2r_1} y_1^{2r_1} \dots x_N^{2l_N - 2r_N} y_N^{2r_N} d(x, y)$$

By using the formula in Theorem 6.12, we obtain:

$$= \sum_{r_1...r_N} {l_1 \choose r_1} \cdots {l_N \choose r_N} \frac{(2N-1)!!(2r_1)!! \cdots (2r_N)!!(2l_1-2r_1)!! \cdots (2l_N-2r_N)!!}{(2N+2\sum l_i-1)!!}$$
  
$$= \sum_{r_1...r_N} {l_1 \choose r_1} \cdots {l_N \choose r_N} \frac{(2N-1)!(2r_1)! \cdots (2r_N)!(2l_1-2r_1)! \cdots (2l_N-2r_N)!}{(2N+\sum l_i-1)!r_1! \cdots r_N!(l_1-r_1)! \cdots (l_N-r_N)!}$$

We can rewrite the sum on the right in the following way:

$$= \sum_{r_1...r_N} \frac{l_1! \dots l_N! (2N-1)! (2r_1)! \dots (2r_N)! (2l_1-2r_1)! \dots (2l_N-2r_N)!}{(2N+\sum l_i-1)! (r_1! \dots r_N! (l_1-r_1)! \dots (l_N-r_N)!)^2}$$
  

$$= \sum_{r_1} \binom{2r_1}{r_1} \binom{2l_1-2r_1}{l_1-r_1} \dots \sum_{r_N} \binom{2r_N}{r_N} \binom{2l_N-2r_N}{l_N-r_N} \frac{(2N-1)! l_1! \dots l_N!}{(2N+\sum l_i-1)!}$$
  

$$= 4^{l_1} \times \dots \times 4^{l_N} \times \frac{(2N-1)! l_1! \dots l_N!}{(2N+\sum l_i-1)!}$$

Thus, we obtain the formula in the statement.

Regarding now the hyperspherical variables, investigated in the above in the real case, we have similar results for the complex spheres, as follows:

#### 6E. EXERCISES

THEOREM 6.20. The rescaled coordinates on the complex sphere 
$$S_{\mathbb{C}}^{N-1}$$

$$w_i = \frac{z_i}{\sqrt{N}}$$

become complex Gaussian and independent with  $N \to \infty$ .

**PROOF.** This follows indeed by using Theorem 6.18 and Theorem 6.19.

## 6e. Exercises

There have been a lot of subtle computations in this chapter, and our exercises here will be about this. Let us start however with a theoretical exercise:

EXERCISE 6.21. Further meditate about the definition of  $\pi$ , notably by improving, if needed, our result about the length and area of the unit circle,

$$L = 2\pi \implies A = \pi$$

into a result as follows, regarding the length and area of a circle of radius R:

$$L = 2\pi R \implies A = \pi R^2$$

Also, reformulate the proof given in the above as to avoid trashing the rounded parts of the pizza, which gastronomically speaking, is an heresy.

Philosophically speaking, the question amounts in understanding what 1 is, geometrically speaking. There might be a discussion about pineapple toppings as well.

As a second exercise, on the same topic, even more theoretical, we have:

EXERCISE 6.22. Push to the limits the meditations about  $\pi$ , by reformulating things as follows, in case you find areas to be more familiar than lengths,

$$A = \pi \implies L = 2\pi$$

or even as follows, with V being the volume of unit cylinder, in case you find volumes to be more familiar than areas and lengths:

$$V = \pi \implies L = 2\pi$$

Also, in case you end up with the conclusion that you actually disagree with all the above, work out your own theory of  $\pi$ , say by starting with your favorite  $\emptyset$  set.

The point with the first question is that, in real life, as opposed to human-made mathematics as we know it, the volumes are far easier to compute than the areas and lengths, simply by immersing your 3D body into a square glass of water. Also, in the unfortunate case where you ended up disagreeing with all the above, it is probably because you have pushed your meditations too far. Have a beer, with that pizza, and relax.

As a first technical exercise now, in relation with higher dimensions, we have:

#### 6. ADVANCED CALCULUS

EXERCISE 6.23. Develop the theory of the gamma function, defined as

$$\Gamma(z) = \int_0^\infty x^{z-1} e^{-x} dx$$

notably by establishing the following formula, for any  $N \in \mathbb{N}$ ,

$$\Gamma(N) = (N-1)!$$

and then comment on the formulae for the volumes and areas of spheres.

To be more precise, the first question is that of establishing the well-known formula  $\Gamma(z+1) = z\Gamma(z)$ . The next step is that of computing  $\Gamma(z)$  for  $z \in \mathbb{N}/2$ , with the above formula in the case  $z \in \mathbb{N}$ . And then, the problem is that of deciding if all this can be useful in connection with the formulae for the volumes and areas of spheres. With the comment that the answer to this latter question is of course something quite subjective.

As a second technical exercise, which is actually quite difficult, we have:

EXERCISE 6.24. Clarify the occurrence of the  $\sqrt{2\pi N}$  term in the Stirling formula

$$N! \simeq \left(\frac{N}{e}\right)^N \sqrt{2\pi N}$$

and work out an estimate for the error term as well.

To be more precise here, in what regards the first question, this is something that we briefly discussed in the above, and the problem is that of working out the details. As for the second question, things are quite tricky here, and we warmly recommend a search of the literature on the subject. Good old-style mathematics, you will be amazed.

Finally, in relation with hyperspherical variables, we have:

EXERCISE 6.25. Prove, with full details, that the rescaled coordinates

$$y_i = \frac{x_i}{\sqrt{N}}$$

become independent with  $N \to \infty$ , for both the real and complex spheres.

To be more precise here, we have proved in the above that the rescaled coordinates become respectively real and complex Gaussian, in the  $N \to \infty$  limit, for the real and complex spheres. The problem now is that of using the same method, namely a straightforward application of our spherical integral formulae, in order to compute and then estimate the joint moments, as to prove the independence result as well.

# CHAPTER 7

# Spectral theory

# 7a. Hilbert spaces

We have seen so far the basics of linear algebra, concerning the relation between linear maps and matrices, the determinant, the diagonalization procedure, along with some applications. In this chapter we discuss what happens in infinite dimensions. Among our motivations is the fact that the spaces of infinite dimensions are of great importance in various branches of theoretical physics, such as quantum mechanics.

Generally speaking, linear algebra is not very interesting in infinite dimensions, due to a number of reasons, the idea being that the infinite dimensionality prevents us from doing many basic things, to the point that we cannot even have things started.

Skipping some details here, and getting directly to the solution to these questions, the idea will be that of using infinite dimensional vector spaces with some extra structure. And the ideal such extra structure is a scalar product:

DEFINITION 7.1. A scalar product on a complex vector space H is an operation

$$H\times H\to \mathbb{C}$$

denoted  $(x, y) \rightarrow < x, y >$ , satisfying the following conditions:

- (1)  $\langle x, y \rangle$  is linear in x, and antilinear in y.
- (2)  $\overline{\langle x, y \rangle} = \langle y, x \rangle$ , for any x, y.
- (3) < x, x >> 0, for any  $x \neq 0$ .

As a basic example here, we have the finite dimensional vector space  $H = \mathbb{C}^N$ , with its usual scalar product, which is as follows:

$$\langle x, y \rangle = \sum_{i} x_i \bar{y}_i$$

There are many other examples, and notably various spaces of  $L^2$  functions, which naturally appear in problems coming from physics. We will discuss them later on.

In order to study the scalar products, let us formulate the following definition:

DEFINITION 7.2. The norm of a vector  $x \in H$  is the following quantity:

$$||x|| = \sqrt{\langle x, x \rangle}$$

We also call this number length of x, or distance from x to the origin.

The terminology comes from what happens in  $\mathbb{C}^N$ , where the length of the vector, as defined by the above formula, coincides with the usual length.

In analogy with what happens in finite dimensions, we have two important results regarding the norms. First is the Cauchy-Schwarz inequality, as follows:

THEOREM 7.3. We have the Cauchy-Schwarz inequality

$$|\langle x, y \rangle| \le ||x|| \cdot ||y||$$

and the equality case holds precisely when x, y are proportional.

**PROOF.** Consider the following quantity, depending on a real variable  $t \in \mathbb{R}$ , and on a variable on the unit circle,  $w \in \mathbb{T}$ :

$$f(t) = ||twx + y||^2$$

By developing f, we see that this is a degree 2 polynomial in t:

$$\begin{aligned} f(t) &= \langle twx + y, twx + y \rangle \\ &= t^2 \langle x, x \rangle + tw \langle x, y \rangle + t\bar{w} \langle y, x \rangle + \langle y, y \rangle \\ &= t^2 ||x||^2 + 2t Re(w \langle x, y \rangle) + ||y||^2 \end{aligned}$$

Since f is obviously positive, its discriminant must be negative:

$$4Re(w < x, y >)^2 - 4||x||^2 \cdot ||y||^2 \le 0$$

But this is equivalent to the following condition:

$$|Re(w < x, y >)| \le ||x|| \cdot ||y||$$

Now the point is that we can arrange for the number  $w \in \mathbb{T}$  to be such that the quantity w < x, y > is real. Thus, we obtain the following inequality:

$$|\langle x, y \rangle| \le ||x|| \cdot ||y||$$

Finally, the study of the equality case is straightforward, by using the fact that the discriminant of f vanishes precisely when we have a root. But this leads to the conclusion in the statement, namely that the vectors x, y must be proportional.

As a second main result now, we have the Minkowski inequality:

THEOREM 7.4. We have the Minkowski inequality

$$||x + y|| \le ||x|| + ||y||$$

and the equality case holds precisely when x, y are proportional.

**PROOF.** This follows indeed from the Cauchy-Schwarz inequality, as follows:

$$\begin{split} ||x+y|| &\leq ||x|| + ||y|| \\ \iff \quad ||x+y||^2 \leq (||x|| + ||y||)^2 \\ \iff \quad ||x||^2 + ||y||^2 + 2Re < x, y > \leq ||x||^2 + ||y||^2 + 2||x|| \cdot ||y|| \\ \iff \quad Re < x, y > \leq ||x|| \cdot ||y|| \end{split}$$

As for the equality case, this is clear from Cauchy-Schwarz as well.

As a consequence of this, we have the following result:

THEOREM 7.5. The following function is a distance on H,

$$d(x,y) = ||x - y||$$

in the usual sense, that of the abstract metric spaces.

**PROOF.** This follows indeed from the Minkowski inequality, which corresponds to the triangle inequality, the other two axioms for a distance being trivially satisfied.  $\Box$ 

The above result is quite important, because it shows that we can do geometry in our present setting, a bit as in the finite dimensional case.

Finally, in connection with doing geometry, we have the following key technical result, which shows that everything can be recovered in terms of distances:

**PROPOSITION** 7.6. The scalar products can be recovered from distances, via the formula

$$4 < x, y > = ||x + y||^{2} - ||x - y||^{2} + i||x + iy||^{2} - i||x - iy||^{2}$$

called complex polarization identity.

**PROOF.** This is something that we have already met in finite dimensions. In arbitrary dimensions the proof is similar, as follows:

$$\begin{split} ||x+y||^2 - ||x-y||^2 + i||x+iy||^2 - i||x-iy||^2 \\ &= ||x||^2 + ||y||^2 - ||x||^2 - ||y||^2 + i||x||^2 + i||y||^2 - i||x||^2 - i||y||^2 \\ &+ 2Re(< x, y >) + 2Re(< x, y >) + 2iIm(< x, y >) + 2iIm(< x, y >) \\ &= 4 < x, y > \end{split}$$

Thus, we are led to the conclusion in the statement.

Let us discuss now some more advanced aspects. In order to do analysis on our spaces, we need the Cauchy sequences that we construct to converge. This is something which is automatic in finite dimensions, but in arbitrary dimensions, this can fail.

Thus, we must add an extra axiom, stating that H is complete with respect to the norm. It is convenient here to formulate a detailed new definition, as follows, which will be the starting point for our various considerations to follow:

DEFINITION 7.7. A Hilbert space is a complex vector space H given with a scalar product  $\langle x, y \rangle$ , satisfying the following conditions:

- (1) < x, y > is linear in x, and antilinear in y.
- (2)  $\overline{\langle x, y \rangle} = \langle y, x \rangle$ , for any x, y.
- (3) < x, x >> 0, for any  $x \neq 0$ .
- (4) *H* is complete with respect to the norm  $||x|| = \sqrt{\langle x, x \rangle}$ .

In other words, we have taken here Definition 7.1 above, and added the condition that H must be complete with respect to the norm  $||x|| = \sqrt{\langle x, x \rangle}$ , that we know indeed to be a norm, according to the Minkowski inequality proved above.

As a basic example, we have the space  $H = \mathbb{C}^N$ , with its usual scalar product:

$$\langle x, y \rangle = \sum_{i} x_i \bar{y}_i$$

More generally now, we have the following construction of Hilbert spaces:

**PROPOSITION 7.8.** The sequences of numbers  $x = (x_i)$  which are square-summable,

$$\sum_i |x_i|^2 < \infty$$

form a Hilbert space, denoted  $l^2(\mathbb{N})$ , with the following scalar product:

$$\langle x, y \rangle = \sum_{i} x_i \bar{y}_i$$

In fact, given any index set I, we can construct a Hilbert space  $l^2(I)$ , in this way.

**PROOF.** The fact that we have indeed a complex vector space with a scalar product is elementary, and the fact that this space is indeed complete is very standard too.  $\Box$ 

On the other hand, we can talk as well about spaces of functions, as follows:

**PROPOSITION 7.9.** Given an interval  $X \subset \mathbb{R}$ , the quantity

$$\langle f,g \rangle = \int_X f(x)\overline{g(x)}dx$$

is a scalar product, making  $H = L^2(X)$  a Hilbert space.

PROOF. Once again this is routine, coming this time from basic measure theory, with  $H = L^2(X)$  being the space of square-integrable functions  $f : X \to \mathbb{C}$ , with the convention that two such functions are identified when they coincide almost everywhere.

We can unify the above two constructions, as follows:

THEOREM 7.10. Given a measured space X, the quantity

$$\langle f,g \rangle = \int_X f(x)\overline{g(x)}dx$$

is a scalar product, making  $H = L^2(X)$  a Hilbert space.

**PROOF.** Here the first assertion is clear, and the fact that the Cauchy sequences converge is clear as well, by taking the pointwise limit, and using a standard argument.  $\Box$ 

Observe that with  $X = \{1, \ldots, N\}$  we obtain the space  $H = \mathbb{C}^N$ . Also, with  $X = \mathbb{N}$ , with the counting measure, we obtain the space  $H = l^2(\mathbb{N})$ . In fact, with an arbitrary set I, once again with the counting mesure, we obtain the space  $H = l^2(I)$ . Thus, the construction in Theorem 7.10 unifies all the Hilbert space constructions that we have.

Quite remarkably, the converse of this holds, in the sense that any Hilbert space must be of the form  $L^2(X)$ . This follows indeed from the following key result, which tells us that, in addition to this, we can always assume that X = I is a discrete space:

THEOREM 7.11. Let H be a Hilbert space.

- (1) Any algebraic basis of this space  $\{f_i\}_{i \in I}$  can be turned into an orthonormal basis  $\{e_i\}_{i \in I}$ , by using the Gram-Schmidt procedure.
- (2) Thus, H has an orthonormal basis, and so we have  $H \simeq l^2(I)$ , with I being the indexing set for this orthonormal basis.

PROOF. This is standard, by recurrence in finite dimensions, using Gram-Schmidt, as stated, and by recurrence as well in infinite, countable dimensions. As for the case of infinite, uncountable dimensions, here the result holds as well, with the proof using transfinite recurrence arguments, borrowed from logic.  $\Box$ 

We have the following definition, based on the above:

DEFINITION 7.12. A Hilbert space H is called separable when the following equivalent conditions are satisfied:

- (1) *H* has a countable algebraic basis  $\{f_i\}_{i \in \mathbb{N}}$ .
- (2) *H* has a countable orthonormal basis  $\{e_i\}_{i \in \mathbb{N}}$ .
- (3) We have  $H \simeq l^2(\mathbb{N})$ , isomorphism of Hilbert spaces.

In what follows we will be mainly interested in the separable Hilbert spaces, where most of the questions coming from physics take place. In view of the above, the following philosophical question appears: why not simply talking about  $l^2(\mathbb{N})$ ?

In answer to this, we cannot really do so, because many of the separable spaces that we are interested in appear as spaces of functions, and such spaces do not necessarily have a very simple or explicit orthonormal basis, as shown by the following result:

PROPOSITION 7.13. The Hilbert space  $H = L^2[0, 1]$  is separable, having as orthonormal basis the orthonormalized version of the algebraic basis

$$f_n = x^n$$

with  $n \in \mathbb{N}$ , coming from the Weierstrass density theorem.

PROOF. The fact that the space  $H = L^2[0, 1]$  is indeed separable is clear from the Weierstrass theorem, which provides us with the algebraic basis  $f_n = x^n$ , which can be orthogonalized by using the Gram-Schmidt procedure, as explained in Theorem 7.11 above. Working out the details here is actually an excellent exercise.

As a conclusion to all this, we are interested in 1 space, namely the unique separable Hilbert space H, but due to various technical reasons, it is often better to forget that we have  $H = l^2(\mathbb{N})$ , and say instead that we have  $H = L^2(X)$ , with X being a separable measured space, or simply say that H is an abstract separable Hilbert space.

### 7b. Linear operators

Let us get now into the study of linear operators  $T: H \to H$ , which will eventually lead us into the correct infinite dimensional version of linear algebra. We first have:

THEOREM 7.14. Let H be an arbitrary Hilbert space, with orthonormal basis  $\{e_i\}_{i \in I}$ . The algebra of all linear operators from H to itself,

$$\mathcal{L}(H) = \left\{ T : H \to H \text{ linear} \right\}$$

embeds then into the algebra of the  $I \times I$  complex matrices,

$$M_I(\mathbb{C}) = \left\{ (M_{ij})_{i,j \in I} \middle| M_{ij} \in \mathbb{C} \right\}$$

with an operator T corresponding to the following matrix:

$$M_{ij} = \langle Te_j, e_i \rangle$$

In the case  $H = \mathbb{C}^N$  we obtain in this way the usual isomorphism  $\mathcal{L}(H) \simeq M_N(\mathbb{C})$ . In the separable case we obtain in this way a proper embedding  $\mathcal{L}(H) \subset M_\infty(\mathbb{C})$ .

**PROOF.** We have three assertions to be proved, the idea being as follows:

(1) The correspondence  $T \to M$  constructed in the statement is indeed linear, and its kernel is  $\{0\}$ , so we have indeed an embedding as follows, as claimed:

$$\mathcal{L}(H) \subset M_I(\mathbb{C})$$

(2) In finite dimensions we obtain an isomorphism, because any matrix  $M \in M_N(\mathbb{C})$  determines an operator  $T : \mathbb{C}^N \to \mathbb{C}^N$ , according to the formula  $\langle Te_j, e_i \rangle = M_{ij}$ .

(3) In infinite dimensions, however, we do not have an isomorphism. For instance on  $H = l^2(\mathbb{N})$  the following matrix does not define an operator:

$$M = \begin{pmatrix} 1 & 1 & \dots \\ 1 & 1 & \dots \\ \vdots & \vdots & \end{pmatrix}$$

Indeed,  $T(e_1)$  should be the all-1 vector, but this vector is not square-summable.  $\Box$ 

In connection with our previous comments, the above result is something quite theoretical, because for basic Hilbert spaces like  $L^2[0, 1]$ , which do not have a simple orthonormal basis, the embedding  $\mathcal{L}(H) \subset M_{\infty}(\mathbb{C})$  that we obtain is not something very useful.

In short, while the operators  $T: H \to H$  are basically some infinite matrices, it is better to think of these operators as being objects on their own.

In what follows we will be interested in the operators  $T: H \to H$  which are bounded. Regarding such operators, we have the following result:

THEOREM 7.15. Given a Hilbert space H, the linear operators  $T: H \to H$  which are bounded, in the sense that we have

$$||T|| = \sup_{||x|| \le 1} ||Tx|| < \infty$$

form a complex algebra with unit B(H), having the property

 $||ST|| \le ||S|| \cdot ||T||$ 

and which is complete with respect to the norm.

**PROOF.** The fact that we have indeed an algebra, satisfying the product condition in the statement, follows from the following estimates, which are all elementary:

$$||S + T|| \le ||S|| + ||T||$$
$$||\lambda T|| = |\lambda| \cdot ||T||$$
$$||ST|| \le ||S|| \cdot ||T||$$

Regarding now the last assertion, if  $\{T_n\} \subset B(H)$  is Cauchy then  $\{T_nx\}$  is Cauchy for any  $x \in H$ , so we can define the limit  $T = \lim_{n \to \infty} T_n$  by setting:

$$Tx = \lim_{n \to \infty} T_n x$$

Let us first check that the application  $x \to Tx$  is linear. We have:

$$T(x+y) = \lim_{n \to \infty} T_n(x+y)$$
  
= 
$$\lim_{n \to \infty} T_n(x) + T_n(y)$$
  
= 
$$\lim_{n \to \infty} T_n(x) + \lim_{n \to \infty} T_n(y)$$
  
= 
$$T(x) + T(y)$$

Similarly, we have as well the following computation:

$$T(\lambda x) = \lim_{n \to \infty} T_n(\lambda x)$$
$$= \lambda \lim_{n \to \infty} T_n(x)$$
$$= \lambda T(x)$$

Thus we have  $T \in \mathcal{L}(H)$ . It remains now to prove that we have  $T \in B(H)$ , and that we have  $T_n \to T$  in norm. For this purpose, observe that we have:

$$\begin{split} ||T_n - T_m|| &\leq \varepsilon \ , \ \forall n, m \geq N \\ \implies \quad ||T_n x - T_m x|| \leq \varepsilon \ , \ \forall ||x|| = 1 \ , \ \forall n, m \geq N \\ \implies \quad ||T_n x - T x|| \leq \varepsilon \ , \ \forall ||x|| = 1 \ , \ \forall n \geq N \\ \implies \quad ||T_N x - T x|| \leq \varepsilon \ , \ \forall ||x|| = 1 \\ \implies \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad ||T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| \leq \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x|| = \varepsilon \\ \Rightarrow \quad |T_N x - T x||$$

As a first consequence, we obtain  $T \in B(H)$ , because we have:

$$||T|| = ||T_N + (T - T_N)||$$
  

$$\leq ||T_N|| + ||T - T_N||$$
  

$$\leq ||T_N|| + \varepsilon$$
  

$$< \infty$$

As a second consequence, we obtain  $T_N \to T$  in norm, and we are done.

In relation with the construction from Theorem 7.14 above, we have:

**PROPOSITION** 7.16. We have embeddings of algebras as follows,

$$B(H) \subset \mathcal{L}(H) \subset M_I(\mathbb{C})$$

which are both proper, in the infinite dimensional case.

**PROOF.** According to Theorem 7.14, the algebra B(H) consists of the  $I \times I$  complex matrices which define indeed linear maps  $T: H \to H$ , and which satisfy as well a second boundedness condition, coming from the boundedness of the norm of T:

$$||T|| < \infty$$

In finite dimensions we have equalities everywhere, but in general this is not true, the standard example of a matrix not producing an operator being as follows:

$$M = \begin{pmatrix} 1 & 1 & \dots \\ 1 & 1 & \dots \\ \vdots & \vdots & \end{pmatrix}$$

As for the examples of linear operators which are not bounded, these are more complicated, coming from logic, and we will not need them in what follows.  $\Box$ 

As already mentioned after Theorem 7.14, all this is something quite theoretical, because for basic function spaces like  $L^2[0,1]$ , which do not have a simple orthonormal basis, the embedding  $B(H) \subset M_I(\mathbb{C})$  that we obtain is not very useful.

As a conclusion, while the bounded operators  $T: H \to H$  are basically some infinite matrices, it is better to think of these operators as being objects on their own.

# 7c. Spectral theory

We will be interested in what follows in B(H) and its closed subalgebras  $A \subset B(H)$ . It is convenient to formulate the following definition:

DEFINITION 7.17. A Banach algebra is a complex algebra with unit A, having a vector space norm ||.|| satisfying

$$|ab|| \le ||a|| \cdot ||b||$$

and which makes it a Banach space, in the sense that the Cauchy sequences converge.

As said above, the basic examples of Banach algebras, or at least the basic examples that we will be interested in here, are the operator algebra B(H), and its norm closed subalgebras  $A \subset B(H)$ , such as the algebras  $A = \langle T \rangle$  generated by a single operator  $T \in B(H)$ . There are many other examples, but more on this later.

Generally speaking, the elements  $a \in A$  of a Banach algebra can be thought of as being bounded operators on some Hilbert space, which is not present. With this idea in mind, we can emulate spectral theory in our setting, the starting point being:

DEFINITION 7.18. The spectrum of an element  $a \in A$  is the set

$$\sigma(a) = \left\{ \lambda \in \mathbb{C} \middle| a - \lambda \notin A^{-1} \right\}$$

where  $A^{-1} \subset A$  is the set of invertible elements.

As a basic example, the spectrum of a usual matrix  $M \in M_N(\mathbb{C})$  is the collection of its eigenvalues, taken of course without multiplicities. In the case of the trivial algebra  $A = \mathbb{C}$ , appearing at N = 1, the spectrum of an element is the element itself.

As a first, basic result regarding spectra, we have:

**PROPOSITION 7.19.** We have the following formula, valid for any  $a, b \in A$ :

 $\sigma(ab) \cup \{0\} = \sigma(ba) \cup \{0\}$ 

Also, there are examples where  $\sigma(ab) \neq \sigma(ba)$ .

**PROOF.** We first prove that we have the following implication:

$$1 \notin \sigma(ab) \implies 1 \notin \sigma(ba)$$

Assume indeed that 1 - ab is invertible, with inverse denoted c:

$$c = (1 - ab)^{-1}$$

We have then the following formulae, relating our variables a, b, c:

$$abc = cab = c - 1$$

By using these formulae, we obtain:

$$(1 + bca)(1 - ba) = 1 + bca - ba - bcaba$$
$$= 1 + bca - ba - bca + ba$$
$$= 1$$

A similar computation shows that we have:

$$(1 - ba)(1 + bca) = 1$$

Thus 1 - ba is invertible, with inverse 1 + bca, which proves our claim. Now by multiplying by scalars, we deduce from this that for any  $\lambda \in \mathbb{C} - \{0\}$  we have:

$$\lambda \notin \sigma(ab) \implies \lambda \notin \sigma(ba)$$

But this leads to the conclusion in the statement, namely:

$$\sigma(ab) \cup \{0\} = \sigma(ba) \cup \{0\}$$

Regarding now the last claim, we know from linear algebra that  $\sigma(ab) = \sigma(ba)$  holds for the usual matrices, for instace because of the above, and because ab is invertible if any only if ba is. However, this latter fact fails for general operators on Hilbert spaces. Indeed, we can take our operator a to be the shift on the space  $l^2(\mathbb{N})$ , given by:

$$S(e_i) = e_{i+1}$$

As for b, we can take the adjoint of S, which is the following operator:

$$S^{*}(e_{i}) = \begin{cases} e_{i-1} & \text{if } i > 0\\ 0 & \text{if } i = 0 \end{cases}$$

Let us compose now these two operators. In one sense, we have:

$$S^*S = 1 \implies 0 \notin \sigma(SS^*)$$

In the other sense, however, the situation is different, as follows:

$$SS^* = Proj(e_0^{\perp}) \implies 0 \in \sigma(SS^*)$$

Thus, the spectra do not match on 0, and we have our counterexample, as desired.  $\Box$ 

Let us discuss now a second basic result about spectra, which is very useful. Given an arbitrary Banach algebra element  $a \in A$ , and a rational function f = P/Q having poles outside  $\sigma(a)$ , we can construct the following element:

$$f(a) = P(a)Q(a)^{-1}$$

For simplicity, and due to the fact that the elements P(a), Q(a) commute, so that the order is irrelevant, we write this element as a usual fraction, as follows:

$$f(a) = \frac{P(a)}{Q(a)}$$

With this convention, we have the following result:

THEOREM 7.20. We have the "rational functional calculus" formula

$$\sigma(f(a)) = f(\sigma(a))$$

valid for any rational function  $f \in \mathbb{C}(X)$  having poles outside  $\sigma(a)$ .

**PROOF.** In order to prove this result, we can proceed in two steps, as follows:

(1) Assume first that we are in the polynomial function case,  $f \in \mathbb{C}[X]$ . We pick a scalar  $\lambda \in \mathbb{C}$ , and we write:

$$f(X) - \lambda = c(X - r_1) \dots (X - r_n)$$

We have then, as desired:

$$\lambda \notin \sigma(f(a)) \iff f(a) - \lambda \in A^{-1}$$
$$\iff c(a - r_1) \dots (a - r_n) \in A^{-1}$$
$$\iff a - r_1, \dots, a - r_n \in A^{-1}$$
$$\iff r_1, \dots, r_n \notin \sigma(a)$$
$$\iff \lambda \notin f(\sigma(a))$$

(2) Assume now that we are in the general rational function case,  $f \in \mathbb{C}(X)$ . We pick a scalar  $\lambda \in \mathbb{C}$ , we write f = P/Q, and we set:

$$F = P - \lambda Q$$

By using now (1), for this polynomial, we obtain:

$$\begin{split} \lambda \in \sigma(f(a)) & \iff F(a) \notin A^{-1} \\ & \iff 0 \in \sigma(F(a)) \\ & \iff 0 \in F(\sigma(a)) \\ & \iff \exists \mu \in \sigma(a), F(\mu) = 0 \\ & \iff \lambda \in f(\sigma(a)) \end{split}$$

Thus, we have obtained the formula in the statement.

Summarizing, we have so far a beginning of theory. Let us prove now something that we do not know yet, namely that the spectra are non-empty:

$$\sigma(a) \neq \emptyset$$

This is something that we know well for the usual matrices. However, a bit of thinking tells us that, even for the usual matrices, this is something rather advanced.

In the present Banach algebra setting, this is definitely something non-trivial. In order to establish this result, we will need a number of analytic preliminaries, as follows:

PROPOSITION 7.21. Let A be a Banach algebra.

(1) 
$$||a|| < 1 \implies (1-a)^{-1} = 1 + a + a^2 + \dots$$

- (2) The set  $A^{-1}$  is open.
- (3) The map  $a \to a^{-1}$  is differentiable.

**PROOF.** All these assertions are elementary, as follows:

(1) This follows as in the scalar case, the computation being as follows, provided that everything converges under the norm, which amounts in saying that ||a|| < 1:

$$(1-a)(1+a+a^{2}+...)$$
  
= 1-a+a-a^{2}+a^{2}-a^{3}+...  
= 1

(2) Assuming  $a \in A^{-1}$ , let us pick  $b \in A$  such that:

$$||a-b|| < \frac{1}{||a^{-1}||}$$

We have then the following estimate:

$$||1 - a^{-1}b|| = ||a^{-1}(a - b)||$$
  

$$\leq ||a^{-1}|| \cdot ||a - b||$$
  

$$< 1$$

Thus by (1) we obtain  $a^{-1}b \in A^{-1}$ , and so  $b \in A^{-1}$ , as desired.

164

(3) This follows as in the scalar case, where the derivative of  $f(t) = t^{-1}$  is:

$$f'(t) = -t^{-2}$$

To be more precise, in the present Banach algebra setting the derivative is no longer a number, but rather a linear transformation. But this linear transformation can be found by developing the function  $f(a) = a^{-1}$  at order 1, as follows:

$$(a+h)^{-1} = ((1+ha^{-1})a)^{-1}$$
  
=  $a^{-1}(1+ha^{-1})^{-1}$   
=  $a^{-1}(1-ha^{-1}+(ha^{-1})^2-\ldots)$   
 $\simeq a^{-1}(1-ha^{-1})$   
=  $a^{-1}-a^{-1}ha^{-1}$ 

We conclude that the derivative that we are looking for is:

$$f'(a)h = -a^{-1}ha^{-1}$$

Thus, we are led to the conclusion in the statement.

We can now formulate a key theorem about the Banach algebras, as follows:

THEOREM 7.22. The spectrum of a Banach algebra element  $\sigma(a) \subset \mathbb{C}$  is:

- (1) Compact.
- (2) Contained in the disc  $D_0(||a||)$ .
- (3) Non-empty.

**PROOF.** This can be proved by using the above results, as follows:

(1) In view of (2) below, it is enough to prove that  $\sigma(a)$  is closed. But this follows from the following computation, with  $|\varepsilon|$  being small:

$$\lambda \notin \sigma(a) \implies a - \lambda \in A^{-1}$$
$$\implies a - \lambda - \varepsilon \in A^{-1}$$
$$\implies \lambda + \varepsilon \notin \sigma(a)$$

(2) This follows from the following computation:

$$\begin{split} \lambda > ||a|| &\implies \left| \left| \frac{a}{\lambda} \right| \right| < 1 \\ &\implies 1 - \frac{a}{\lambda} \in A^{-1} \\ &\implies \lambda - a \in A^{-1} \\ &\implies \lambda \notin \sigma(a) \end{split}$$

165

(3) Assume by contradiction  $\sigma(a) = \emptyset$ . Given a linear form  $f \in A^*$ , consider the following map, which is well-defined, due to our assumption  $\sigma(a) = \emptyset$ :

$$\varphi: \mathbb{C} \to \mathbb{C} \quad , \quad \lambda \to f((a-\lambda)^{-1})$$

By using Proposition 7.21 this map is differentiable, and so is a power series:

$$\varphi(\lambda) = \sum_{k=0}^{\infty} c_k \lambda^k$$

On the other hand, we have the following estimate:

$$\begin{array}{rcl} \lambda \to \infty & \Longrightarrow & a - \lambda \to \infty \\ & \Longrightarrow & (a - \lambda)^{-1} \to 0 \\ & \Longrightarrow & \varphi(\lambda) \to 0 \end{array}$$

Thus  $\varphi = 0$ , and so  $(a - \lambda)^{-1} = 0$ . But this is a contradiction, as desired.

This was for the basic spectral theory in Banach algebras, which notably applies to the case A = B(H). It is possible to go beyond the above, for instance with a holomorphic function extension of the rational functional calculus formula  $\sigma(f(a)) = f(\sigma(a))$  from Theorem 7.20. Also, in the case of the algebras of operators, more can be said.

## 7d. Operator algebras

Let us get back now to the operator algebra B(H), from Theorem 7.15. The point is that this Banach algebra is of a very special type, due to the following fact:

THEOREM 7.23. The Banach algebra B(H) has an involution  $T \to T^*$ , given by

$$\langle Tx, y \rangle = \langle x, T^*y \rangle$$

which is antilinear, antimultiplicative, and is an isometry, in the sense that:

$$||T|| = ||T^*||$$

Moreover, the norm the involution are related as well by the following formula:

$$||TT^*|| = ||T||^2$$

**PROOF.** We have several things to be proved, the idea being as follows:

(1) As a preliminary fact, that we will need in what follows, our claim is that any linear form  $\varphi: H \to \mathbb{C}$  must be of the following type, for a certain vector  $z \in H$ :

$$\varphi(x) = \langle x, z \rangle$$

Indeed, this is something clear for any Hilbert space of type  $H = l^2(I)$ . But, by using a basis, any Hilbert space is of this form, and so we have proved our claim.

(2) The existence of the adjoint operator  $T^*$ , given by the formula in the statement, comes from the fact that the function  $\varphi(x) = \langle Tx, y \rangle$  being a linear map  $H \to \mathbb{C}$ , we must have a formula as follows, for a certain vector  $T^*y \in H$ :

$$\varphi(x) = \langle x, T^*y \rangle$$

Moreover, since this vector is unique,  $T^*$  is unique too, and we have as well:

$$(S+T)^* = S^* + T^*$$
$$(\lambda T)^* = \overline{\lambda} T^*$$
$$(ST)^* = T^* S^*$$
$$(T^*)^* = T$$

Observe also that we have indeed  $T^* \in B(H)$ , because:

$$||T|| = \sup_{||x||=1} \sup_{||y||=1} < Tx, y >$$
  
= 
$$\sup_{||y||=1} \sup_{||x||=1} < x, T^*y >$$
  
= 
$$||T^*||$$

Summarizing, we have proved everything, up to the last assertion.

(3) Regarding now the last assertion, observe that we have:

$$||TT^*|| \le ||T|| \cdot ||T^*|| = ||T||^2$$

On the other hand, we have as well the following estimate:

$$||T||^{2} = \sup_{||x||=1} | < Tx, Tx > |$$
  
= 
$$\sup_{||x||=1} | < x, T^{*}Tx > |$$
  
$$\leq ||T^{*}T||$$

By replacing  $T \to T^*$  we obtain from this that we have as well:

$$||T||^2 \le ||TT^*||$$

Thus, we have obtained the needed inequality, and we are done.

As a first observation, in the context of the construction  $T \to M$  from Theorem 7.14 above, the adjoint operation  $T \to T^*$  takes a very simple form, namely:

$$(M^*)_{ij} = M_{ji}$$

However, as already explained before, while the operators  $T: H \to H$  are basically some infinite matrices, it is better to think of them as being objects on their own.

The above result suggests the following definition:

DEFINITION 7.24. A  $C^*$ -algebra is a complex algebra with unit A, having:

- (1) A norm  $a \rightarrow ||a||$ , making it a Banach algebra.
- (2) An involution  $a \to a^*$ , which satisfies  $||aa^*|| = ||a||^2$ , for any  $a \in A$ .

At the level of the basic examples, we know from Theorem 7.23 that the full operator algebra B(H) is a  $C^*$ -algebra, in the above sense. More generally, any closed \*-subalgebra  $A \subset B(H)$  is a  $C^*$ -algebra. We will see later on that any  $C^*$ -algebra appears in fact in this way, as a closed \*-subalgebra  $A \subset B(H)$ , for a certain Hilbert space H.

For the moment, we are interested in developing the theory of  $C^*$ -algebras, without reference to operators, or Hilbert spaces. As a first observation, we have:

PROPOSITION 7.25. If X is an abstract compact space, the algebra C(X) of continuous functions  $f: X \to \mathbb{C}$  is a  $C^*$ -algebra, with structure as follows:

(1) The norm is the usual sup norm, given by:

$$||f|| = \sup_{x \in X} |f(x)|$$

(2) The involution is the usual involution, given by:

$$f^*(x) = \overline{f(x)}$$

This algebra is commutative, in the sense that fg = gf, for any f, g.

**PROOF.** Almost everything here is trivial. Observe that we have indeed:

$$|ff^*|| = \sup_{x \in X} |f(x)\overline{f(x)}|$$
$$= \sup_{x \in X} |f(x)|^2$$
$$= ||f||^2$$

Thus, the axioms are satisfied, and finally fg = gf is clear.

Our claim now is that any commutative  $C^*$ -algebra appears as above. This is something non-trivial, which requires a number of preliminaries. We will need:

DEFINITION 7.26. Given an element  $a \in A$ , its spectral radius

 $\rho(a) \in (0, ||a||)$ 

is the radius of the smallest disk centered at 0 containing  $\sigma(a)$ .

Here we have included a number of results that we already know, from Theorem 7.22, namely the fact that the spectrum is nonzero, and contained in the disk  $D_0(||a||)$ .

We have the following key result, extending our spectral theory knowledge, from the general Banach algebra setting, to the present  $C^*$ -algebra setting:

THEOREM 7.27. Let A be a  $C^*$ -algebra.

- (1) The spectrum of a unitary element  $(a^* = a^{-1})$  is on the unit circle.
- (2) The spectrum of a self-adjoint element  $(a = a^*)$  consists of real numbers.
- (3) The spectral radius of a normal element ( $aa^* = a^*a$ ) is equal to its norm.

**PROOF.** We use the various results established above, and notably the rational calculus formula from Theorem 7.20, and the various results from Theorem 7.22:

(1) Assuming  $a^* = a^{-1}$ , we have the following norm computations:

$$||a|| = \sqrt{||aa^*||} = \sqrt{1} = 1$$
$$||a^{-1}|| = ||a^*|| = ||a|| = 1$$

Now if we denote by D the unit disk, we obtain from this:

$$||a|| = 1 \implies \sigma(a) \subset D$$
$$||a^{-1}|| = 1 \implies \sigma(a^{-1}) \subset D$$

On the other hand, by using the rational function  $f(z) = z^{-1}$ , we have:

$$\sigma(a^{-1}) \subset D \implies \sigma(a) \subset D^{-1}$$

Now by putting everything together we obtain, as desired:

$$\sigma(a) \subset D \cap D^{-1} = \mathbb{T}$$

(2) This follows by using the result (1), just established above, and Theorem 7.20, with the following rational function, depending on  $t \in \mathbb{R}$ :

$$f(z) = \frac{z + it}{z - it}$$

Indeed, for t >> 0 the element f(a) is well-defined, and we have:

$$\left(\frac{a+it}{a-it}\right)^* = \frac{(a+it)^*}{(a-it)^*}$$
$$= \frac{a-it}{a+it}$$
$$= \left(\frac{a+it}{a-it}\right)^{-1}$$

Thus the element f(a) is a unitary, and by using (1) its spectrum is contained in  $\mathbb{T}$ . We conclude from this that we have:

$$f(\sigma(a)) = \sigma(f(a)) \subset \mathbb{T}$$

But this shows that we have, as desired:

$$\sigma(a) \subset f^{-1}(\mathbb{T}) = \mathbb{R}$$

(3) We already know that we have the inequality in one sense,  $\rho(a) \leq ||a||$ , and this for any  $a \in A$ . For the reverse inequality, when a is normal, we fix a number as follows:

$$\rho > \rho(a)$$

We have then the following computation, with the convention that the integration over the circle  $|z| = \rho$  is normalized, as for the integral of the 1 function to be 1:

$$\int_{|z|=\rho} \frac{z^n}{z-a} dz = \int_{|z|=\rho} \sum_{k=0}^{\infty} z^{n-k-1} a^k dz$$
$$= \sum_{k=0}^{\infty} \left( \int_{|z|=\rho} z^{n-k-1} dz \right) a^k$$
$$= \sum_{k=0}^{\infty} \delta_{n,k+1} a^k$$
$$= a^{n-1}$$

Here we have used the following formula, with  $m \in \mathbb{Z}$ , whose proof is elementary:

$$\int_{|z|=\rho} z^m \, dz = \delta_{m0}$$

By applying now the norm and taking n-th roots we obtain from the above formula, modulo some elementary manipulations, the following estimate:

$$\rho \ge \lim_{n \to \infty} ||a^n||^{1/n}$$

Now recall that  $\rho$  was by definiton an arbitrary number satisfying  $\rho > \rho(a)$ . Thus, we have obtained the following estimate, valid for any  $a \in A$ :

$$\rho(a) \ge \lim_{n \to \infty} ||a^n||^{1/n}$$

In order to finish, we must prove that when a is normal, this estimate implies the missing estimate, namely  $\rho(a) \ge ||a||$ . We can proceed in two steps, as follows:

Step 1. In the case  $a = a^*$  we have  $||a^n|| = ||a||^n$  for any exponent of the form  $n = 2^k$ , by using the  $C^*$ -algebra condition  $||aa^*|| = ||a||^2$ , and by taking *n*-th roots we get:

$$\rho(a) \ge ||a||$$

Thus, we are done with the self-adjoint case, with the result  $\rho(a) = ||a||$ .

<u>Step 2</u>. In the general normal case  $aa^* = a^*a$  we have  $a^n(a^n)^* = (aa^*)^n$ , and by using this, along with the result from Step 1, applied to  $aa^*$ , we obtain:

$$\rho(a) \geq \lim_{n \to \infty} ||a^n||^{1/n}$$

$$= \sqrt{\lim_{n \to \infty} ||a^n(a^n)^*||^{1/n}}$$

$$= \sqrt{\lim_{n \to \infty} ||(aa^*)^n||^{1/n}}$$

$$= \sqrt{\rho(aa^*)}$$

$$= \sqrt{||a||^2}$$

$$= ||a||$$

Thus, we are led to the conclusion in the statement.

As a first comment, the spectral radius formula  $\rho(a) = ||a||$  does not hold in general, the simplest counterexample being the following non-normal matrix:

$$M = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$

As another comment, we can combine the formula  $\rho(a) = ||a||$  for normal elements with the formula  $||aa^*|| = ||a||^2$ , and we are led to the following statement:

PROPOSITION 7.28. In a  $C^*$ -algebra, the norm is given by

$$||a|| = \sqrt{\sup\left\{\lambda \in \mathbb{C} \middle| aa^* - \lambda \notin A^{-1}\right\}}$$

and so is an algebraic quantity.

**PROOF.** We have the following computation, using the condition  $||aa^*|| = ||a||^2$ , then the spectral radius formula for  $aa^*$ , and finally the definition of the spectral radius:

$$\begin{aligned} ||a|| &= \sqrt{||aa^*||} \\ &= \sqrt{\rho(aa^*)} \\ &= \sqrt{\sup\left\{\lambda \in \mathbb{C} \middle| \lambda \in \sigma(aa^*)\right\}} \\ &= \sqrt{\sup\left\{\lambda \in \mathbb{C} \middle| aa^* - \lambda \notin A^{-1}\right\}} \end{aligned}$$

Thus, we are led to the conclusion in the statement.

The above result is quite interesting, because it raises the possibility of axiomatizing the  $C^*$ -algebras as being the Banach \*-algebras having the property that the formula in Proposition 7.28 defines a norm, which must satisfy the usual  $C^*$ -algebra conditions. However, this is something rather philosophical, and we will not follow this path.

 $\square$ 

We are now in position of proving a key result, namely:

THEOREM 7.29 (Gelfand). Any commutative  $C^*$ -algebra is the form

A = C(X)

with the compact space X, called spectrum of A, and denoted

X = Spec(A)

appearing as the space of Banach algebra characters  $\chi : A \to \mathbb{C}$ .

**PROOF.** This can be deduced from our spectral theory results, as follows:

(1) Given a commutative  $C^*$ -algebra A, we can define indeed X to be the set of characters  $\chi : A \to \mathbb{C}$ , with the topology making continuous all the evaluation maps:

$$ev_a: \chi \to \chi(a)$$

Then X is a compact space, and  $a \to ev_a$  is a morphism of algebras:

$$ev: A \to C(X)$$

(2) We first prove that ev is involutive. We use the following formula:

$$a = rac{a+a^*}{2} - i \cdot rac{i(a-a^*)}{2}$$

Thus it is enough to prove the following equality, for self-adjoint elements a:

$$ev_{a^*} = ev_a^*$$

But this is the same as proving that  $a = a^*$  implies that  $ev_a$  is a real function, which is in turn true, because  $ev_a(\chi) = \chi(a)$  is an element of  $\sigma(a)$ , contained in  $\mathbb{R}$ .

(3) Since A is commutative, each element is normal, so ev is isometric:

$$||ev_a|| = \rho(a) = ||a||$$

(4) It remains to prove that ev is surjective. But this follows from the Stone-Weierstrass theorem, because ev(A) is a closed subalgebra of C(X), which separates the points.  $\Box$ 

As a first consequence of the Gelfand theorem, we can extend the rational calculus formula from Theorem 7.20 above, in the case of the normal elements, as follows:

THEOREM 7.30. We have the "continuous functional calculus" formula

$$\sigma(f(a)) = f(\sigma(a))$$

valid for any normal element  $a \in A$ , and any continuous function  $f \in C(\sigma(a))$ .

**PROOF.** Since our element a is normal, the  $C^*$ -algebra  $\langle a \rangle$  that is generates is commutative, and the Gelfand theorem gives an identification as follows:

$$\langle a \rangle = C(X)$$

In order to compute X, observe that the map  $X \to \sigma(a)$  given by evaluation at a is bijective. Thus, we have an identification of compact spaces, as follows:

$$X = \sigma(a)$$

As a conclusion, the Gelfand theorem provides us with an identification as follows:

$$\langle a \rangle = C(\sigma(a))$$

Now given  $f \in C(\sigma(a))$ , we can define indeed an element  $f(a) \in A$ , with  $f \to f(a)$  being a morphism of  $C^*$ -algebras, and we have  $\sigma(f(a)) = f(\sigma(a))$ , as claimed.

The above result adds to a series of similar statements, namely Theorem 7.20, dealing with rational calculus, and the known holomorphic calculus in Banach algebras, briefly mentioned after Theorem 7.22 above. The story is not over here, because in certain special  $C^*$ -algebras, such as the matrix algebras  $M_N(\mathbb{C})$ , or more generally the so-called von Neumann algebras, we can apply if we want arbitrary measurable functions to the normal elements, and we have  $\sigma(f(a)) = f(\sigma(a))$  as well. We will not get into this.

As another important remark, the above result, or rather the formula  $\langle a \rangle = C(\sigma(a))$ from its proof, when applied to the normal operators  $T \in B(H)$ , is more of less the spectral theorem for such operators. Once again, we will not get into this.

As a last topic, let us discuss now the GNS representation theorem, providing us with embeddings  $A \subset B(H)$ . We will need some more spectral theory, as follows:

**PROPOSITION 7.31.** For a normal element  $a \in A$ , the following are equivalent:

- (1) a is positive, in the sense that  $\sigma(a) \subset [0, \infty)$ .
- (2)  $a = b^2$ , for some  $b \in A$  satisfying  $b = b^*$ .
- (3)  $a = cc^*$ , for some  $c \in A$ .

**PROOF.** This is something very standard, as follows:

(1)  $\implies$  (2) Since a is normal, we can use Theorem 7.30, and set  $b = \sqrt{a}$ .

(2)  $\implies$  (3) This is trivial, because we can set c = b.

(3)  $\implies$  (1) We proceed by contradiction. By multiplying c by a suitable element of  $\langle cc^* \rangle$ , we are led to the existence of an element  $d \neq 0$  satisfying  $-dd^* \geq 0$ . By writing now d = x + iy with  $x = x^*, y = y^*$  we have:

$$dd^* + d^*d = 2(x^2 + y^2) \ge 0$$

Thus  $d^*d \ge 0$ . But this contradicts the elementary fact that  $\sigma(dd^*), \sigma(d^*d)$  must coincide outside  $\{0\}$ , that we know from Proposition 7.19 above.

Here is now the GNS representation theorem for the  $C^*$ -algebras, due to Gelfand, Naimark and Segal, along with the idea of the proof:

THEOREM 7.32 (GNS theorem). Let A be a  $C^*$ -algebra.

(1) A appears as a closed \*-subalgebra  $A \subset B(H)$ , for some Hilbert space H.

(2) When A is separable (usually the case), H can be chosen to be separable.

(3) When A is finite dimensional, H can be chosen to be finite dimensional.

**PROOF.** Let us first discuss the commutative case, A = C(X). Our claim here is that if we pick a probability measure on X, we have an embedding as follows:

$$C(X) \subset B(L^2(X))$$

$$f \to (g \to fg)$$

Indeed, given a function  $f \in C(X)$ , consider the operator  $T_f(g) = fg$ , acting on  $H = L^2(X)$ . Observe that  $T_f$  is indeed well-defined, and bounded as well, because:

$$|fg||_2 = \sqrt{\int_X |f(x)|^2 |g(x)|^2 dx}$$
  
  $\leq ||f||_{\infty} ||g||_2$ 

The application  $f \to T_f$  being linear, involutive, continuous, and injective as well, we obtain in this way a  $C^*$ -algebra embedding  $C(X) \subset B(H)$ , as claimed.

In general, we can use a similar idea, with the algebraic aspects being fine, and with the positivity issues being taken care of by Proposition 7.31.

Indeed, assuming that a linear form  $\varphi : A \to \mathbb{C}$  has some suitable positivity properties, making it analogous to the integration functionals  $\int_X : A \to \mathbb{C}$  from the commutative case, we can define a scalar product on A, by the following formula:

$$\langle a, b \rangle = \varphi(ab^*)$$

By completing we obtain a Hilbert space H, and we have an embedding as follows:

$$A \subset B(H)$$
$$a \to (b \to ab)$$

Thus we obtain the assertion (1), and a careful examination of the construction  $A \rightarrow H$ , outlined above, shows that the assertions (2,3) are in fact proved as well.

There are of course many other things that can be said about the bounded operators and the operator algebras, but for our purposes here, the above material, and especially the Gelfand theorem, will be basically all that we will need, in what follows.

#### 7E. EXERCISES

### 7e. Exercises

The present chapter was an introduction to linear algebra in infinite dimensions, and most of our exercises here will be about straightforward continuations of all this. As a first exercise, dealing with the general theory of Hilbert spaces, we have:

EXERCISE 7.33. Find an explicit orthonormal basis of the separable Hilbert space

 $H = L^2[0, 1]$ 

by applying the Gram-Schmidt procedure to the polynomials  $f_n = x^n$ , with  $n \in \mathbb{N}$ .

This is something both fundamental and scary, and the answer can be found by doing an internet search with the keyword "orthogonal polynomials", then carefully reading what comes out of that, and adapting it if needed to the  $H = L^2[0, 1]$  situation.

Here is another instructive exercise, about explicit operators  $T \in B(H)$ :

EXERCISE 7.34. Develop a theory of projections, isometries and symmetries inside B(H), notably by examining the validity of the formula

$$\lim_{n \to \infty} (PQ)^n = P \land Q$$

when talking about projections, and also by taking into account the fact that

 $UU^* = 1 \iff U^*U = 1$ 

does not necessarily hold in infinite dimensions, when talking about isometries.

To be more precise, there are countless things to be done here. Regarding the projections, their definition should be  $P^2 = P = P^*$ , but there are a few things to be clarified in relation with the fact that an arbitrary subspace  $K \subset H$  is not necessarily closed. Once all this understood, the limiting formula in the statement is a good object of study. As in what regards the isometries, and related operators such as the symmetries, here things are tricky as well, and the hint here would be to understand first when the operators  $UU^*, U^*U$  are projections, and what is the precise relation between these projections.

Going ahead with spectra, here is a key exercise:

EXERCISE 7.35. Prove that for the usual matrices  $A, B \in M_N(\mathbb{C})$  we have

$$\sigma^+(AB) = \sigma^+(BA)$$

where  $\sigma^+$  denotes the set of eigenvalues, taken with multiplicities.

As a remark, we have seen in the above that  $\sigma(AB) = \sigma(BA)$  holds outside {0}, and the equality on {0} holds as well, because AB is invertible if and only if BA is invertible. However, in what regards the eigenvalues taken with multiplicities, things are more tricky here, and the answer should be somewhere inside your linear algebra knowledge.

Here is now an exercise in relation with the notion of positivity:

EXERCISE 7.36. Prove that an operator  $T \in B(H)$  satisfies the condition

 $\langle Tx, x \rangle \geq 0$ 

for any  $x \in H$  precisely when it is positive in our sense,  $\sigma(T) \subset [0, \infty)$ .

In one sense this is normally something quite clear, and in the other sense this needs some tricks with vectors and scalar products, such as the polarization identity. Working out first the case of the usual matrices,  $M \in M_N(\mathbb{C})$ , with not much advanced linear algebra involved, is actually a good preliminary exercise.

At a more theoretical level now, also in connection with operators, we have:

EXERCISE 7.37. Clarify, with examples and counterexamples, the relation between the eigenvalues of an operator  $T \in B(H)$ , and its spectrum  $\sigma(T) \subset \mathbb{C}$ .

Here, as usual, the counterexamples could only come from the shift operator S, on the space  $H = l^2(\mathbb{N})$ . As a bonus exercise here, try computing the spectrum of S.

Here is now a key exercise, once again about operator theory:

EXERCISE 7.38. Assuming that an operator  $T \in B(H)$  is normal,  $TT^* = T^*T$ , apply the Gelfand theorem to the  $C^*$ -algebra that it generates

$$< T > \subset B(H)$$

in order to deduce a diagonalization theorem for T.

This is actually something quite heavy, so better start with an internet search.

Finally, also in relation with the Gelfand theorem, we have:

EXERCISE 7.39. Develop a theory of noncommutative geometry, by formally writing any  $C^*$ -algebra, not necessarily commutative, as

$$A = C(X)$$

with X being a "compact quantum space", and report on what you found.

This is of course a very broad question, and almost anything would be welcome here, depending on your geometric taste and knowledge, plus of course on the ability to generalize what you know, to the case of the quantum spaces as defined above. We will be actually back to such questions and to  $C^*$ -algebras later on, when doing group theory.

# CHAPTER 8

# Special matrices

### 8a. Fourier matrices

In this chapter we go back to finite dimensions, and we discuss a number of more specialized questions. To be more precise, we will be interested in various classes of "special matrices", and in the tools for dealing with them. We will see that all this leads us into Fourier analysis, and more specifically into "discrete Fourier analysis".

As a first and central example of a special matrix, we have the flat matrix:

DEFINITION 8.1. The flat matrix  $\mathbb{I}_N$  is the all-one  $N \times N$  matrix:

$$\mathbb{I}_N = \begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix}$$

Equivalently,  $\mathbb{I}_N/N$  is the orthogonal projection on the all-one vector  $\xi \in \mathbb{C}^N$ .

Observe that  $\mathbb{I}_N$  has a lot of interesting properties, such as being circulant, and bistochastic. The idea will be that many techniques that can be applied to  $\mathbb{I}_N$ , with quite trivial results, apply to such special classes of matrices, with non-trivial consequences.

A first interesting question regarding  $\mathbb{I}_N$  concerns its diagonalization. Since  $\mathbb{I}_N$  is a multiple of a rank 1 projection, we have right away the following result:

**PROPOSITION 8.2.** The flat matrix  $\mathbb{I}_N$  diagonalizes as follows,

$$\begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix} = P \begin{pmatrix} N & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{pmatrix} P^{-1}$$

where  $P \in M_N(\mathbb{C})$  can be any matrix formed by the all one-vector  $\xi$ , followed by N-1linearly independent solutions  $x \in \mathbb{C}^N$  of the following equation:

$$x_1 + \ldots + x_N = 0$$

PROOF. This follows indeed from our linear algebra knowledge from chapters 1-4 above, by using the fact that  $\mathbb{I}_N/N$  is the orthogonal projection onto  $\mathbb{C}\xi$ .

#### 8. SPECIAL MATRICES

In practice now, the problem which is left is that of finding an explicit matrix  $P \in M_N(\mathbb{C})$ , as above. To be more precise, there are plently of solutions here, some of them being even real,  $P \in M_N(\mathbb{R})$ , and the problem is that of finding a "nice" such solution, say having the property that  $P_{ij}$  appears as an explicit function of i, j.

Long story short, we are led to the question of solving, in a somewhat canonical and elegant way, the following equation, over the real or the complex numbers:

$$x_1 + \ldots + x_N = 0$$

And this is more tricky than it seems. To be more precise, there is no hope of doing this over the real numbers. As in what regards the complex numbers, there is a ray of light here coming from the roots of unity, and more specifically, from:

**PROPOSITION 8.3.** We have the formula

$$\frac{1}{N} \sum_{k=0}^{N-1} z^k = \begin{cases} 0 & \text{if } z \neq 1\\ 1 & \text{if } z = 1 \end{cases}$$

valid for any N-th root of unity z.

PROOF. This is something that we know from chapter 3 above, coming from the fact that the average in the statement computes the barycenter of the polygon formed by the numbers  $1, z, z^2, \ldots, z^{N-1}$ , in the complex plane. Indeed, since this polygon is regular and centered at 0, the barycenter in question is obviously 0, and this unless we are in the case z = 1, where the polygon degenerates, and the barycenter is obviously 1.

Summarizing, we have a beginning of an answer to our question, with the idea being that of using a matrix  $P \in M_N(\mathbb{C})$  formed by the N-th roots of unity.

To be more precise, this leads us into the Fourier matrix, which is as follows:

DEFINITION 8.4. The Fourier matrix  $F_N$  is the following matrix, with  $w = e^{2\pi i/N}$ :

$$F_N = \begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & w & w^2 & \dots & w^{N-1} \\ 1 & w^2 & w^4 & \dots & w^{2(N-1)} \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & w^{N-1} & w^{2(N-1)} & \dots & w^{(N-1)^2} \end{pmatrix}$$

That is,  $F_N = (w^{ij})_{ij}$ , with indices  $i, j \in \{0, 1, \dots, N-1\}$ , taken modulo N.

Before getting further, observe that this matrix  $F_N$  is "special" as well, but in a different sense, its main properties being the fact that it is a Vandermonde matrix, and a rescaled unitary. We will axiomatize later on the matrices of this type.

Getting back now to the diagonalization problem for the flat matrix  $\mathbb{I}_N$ , this can be solved by using the Fourier matrix  $F_N$ , in the following elegant way:

THEOREM 8.5. The flat matrix  $\mathbb{I}_N$  diagonalizes as follows,

$$\begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix} = \frac{1}{N} F_N \begin{pmatrix} N & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{pmatrix} F_N^*$$

with  $F_N = (w^{ij})_{ij}$  being the Fourier matrix.

PROOF. According to Proposition 8.2, we are left with finding the 0-eigenvectors of  $\mathbb{I}_N$ , which amounts in solving the following equation:

$$x_0 + \ldots + x_{N-1} = 0$$

For this purpose, we use the root of unity  $w = e^{2\pi i/N}$ , and more specifically, the following standard formula, coming from Proposition 8.3:

$$\sum_{i=0}^{N-1} w^{ij} = N\delta_{j0}$$

This formula shows that for j = 1, ..., N-1, the vector  $v_j = (w^{ij})_i$  is a 0-eigenvector. Moreover, these vectors are pairwise orthogonal, because we have:

$$\langle v_j, v_k \rangle = \sum_i w^{ij-ik}$$
  
 $= N\delta_{jk}$ 

Thus, we have our basis  $\{v_1, \ldots, v_{N-1}\}$  of 0-eigenvectors, and since the N-eigenvector is  $\xi = v_0$ , the passage matrix P that we are looking is given by:

$$P = \begin{bmatrix} v_0 & v_1 & \dots & v_{N-1} \end{bmatrix}$$

But this is precisely the Fourier matrix,  $P = F_N$ . In order to finish now, observe that the above computation of  $\langle v_i, v_j \rangle$  shows that  $F_N/\sqrt{N}$  is unitary, and so:

$$F_N^{-1} = \frac{1}{N} F_N^*$$

Thus, we are led to the diagonalization formula in the statement.

Generally speaking, the above result will be the template for what we will be doing here. On one hand we will have special matrices to be studied, of  $\mathbb{I}_N$  type, and on the other hand we will have special matrices that can be used as tools, of  $F_N$  type.

Let us begin with a discussion of the "tools". The Fourier matrix  $F_N$  certainly has many interesting properties, and there are many possible generalizations of it, but what
really stands out, for instance when looking at the above proof, is the fact that  $F_N$  has entries on the unit circle, and its columns are pairwise orthogonal.

This discussion leads to the following definition:

**DEFINITION 8.6.** A complex Hadamard matrix is a square matrix of type

 $H \in M_N(\mathbb{T})$ 

where  $\mathbb{T}$  is the unit circle, satisfying the following equivalent conditions:

- (1) The rows are pairwise orthogonal.
- (2) The columns are pairwise orthogonal.
- (3) The rescaled matrix  $H/\sqrt{N}$  is unitary.
- (4) The rescaled matrix  $H^t/\sqrt{N}$  is unitary.

Here the fact that the above conditions are equivalent comes from basic linear algebra, and more specifically from the fact that a matrix  $U \in M_N(\mathbb{C})$  is a unitary precisely when the rows, or the columns, have norm 1, and are pairwise orthogonal.

We already know, from the proof of Theorem 8.5, that the Fourier matrix  $F_N$  is a complex Hadamard matrix. Here is the list of such matrices, at small values of N:

**PROPOSITION 8.7.** The first Fourier matrix is:

$$F_1 = (1)$$

The second Fourier matrix is as follows:

$$F_2 = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

The third Fourier matrix is as follows, with  $w = e^{2\pi i/3}$ :

$$F_3 = \begin{pmatrix} 1 & 1 & 1 \\ 1 & w & w^2 \\ 1 & w^2 & w \end{pmatrix}$$

As for the fourth Fourier matrix, this is as follows:

$$F_4 = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{pmatrix}$$

PROOF. All these formulae are clear from definitions, with our usual convention for the indices of the Fourier matrices, from Definition 8.4 above.  $\hfill \Box$ 

There are many other examples of complex Hadamard matrices, and the basic theory of such matrices can be summarized as follows:

### 8A. FOURIER MATRICES

THEOREM 8.8. The class of the  $N \times N$  complex Hadamard matrices has the following properties:

- (1) It contains the Fourier matrix  $F_N$ .
- (2) It is stable under taking tensor products.
- (3) It is stable under taking transposes, conjugates and adjoints.
- (4) It is stable under permuting rows, or permuting columns.
- (5) It is stable under multiplying rows or columns by numbers in  $\mathbb{T}$ .

**PROOF.** All this is elementary, the idea being as follows:

(1) This is something that we already know, from the proof of Theorem 8.5 above, with the orthogonality coming from the following standard formula:

$$\sum_{i} w^{ij} = N\delta_{j0}$$

(2) Assume that  $H \in M_M(\mathbb{T})$  and  $K \in M_N(\mathbb{T})$  are Hadamard matrices, and consider their tensor product, which in double index notation is as follows:

$$(H \otimes K)_{ia,jb} = H_{ij}K_{ab}$$

We have then  $H \otimes K \in M_{MN}(\mathbb{T})$ , and the rows  $R_{ia}$  of this matrix are pairwise orthogonal, as shown by the following computation:

$$\langle R_{ia}, R_{kc} \rangle = \sum_{jb} H_{ij} K_{ab} \cdot \bar{H}_{kj} \bar{K}_{cb}$$
$$= \sum_{j} H_{ij} \bar{H}_{kj} \sum_{b} K_{ab} \bar{K}_{cb}$$
$$= MN \delta_{ik} \delta_{ac}$$

(3) We know that the set formed by the  $N \times N$  complex Hadamard matrices appears as follows, with the intersection being taken inside  $M_N(\mathbb{C})$ :

$$X_N = M_N(\mathbb{T}) \cap \sqrt{NU_N}$$

The set  $M_N(\mathbb{T})$  is stable under the operations in the statement. As for the set  $\sqrt{NU_N}$ , here we can use the well-known fact that if a matrix is unitary,  $U \in U_N$ , then so is its complex conjugate  $\overline{U} = (\overline{U}_{ij})$ , the inversion formulae being as follows:

$$U^* = U^{-1}$$
$$U^t = \bar{U}^{-1}$$

Thus the unitary group  $U_N$  is stable under the following operations:

$$U \to U^t$$
$$U \to \bar{U}$$
$$U \to U^*$$

It follows that the above set  $X_N$  is stable as well under these operations, as desired.

(4) This is clear from definitions, because permuting rows or permuting columns leaves invariant both the sets  $M_N(\mathbb{T})$  and  $\sqrt{N}U_N$ .

(5) This is once again clear from definitions, because multiplying rows or columns by numbers in  $\mathbb{T}$  leaves invariant both the sets  $M_N(\mathbb{T})$  and  $\sqrt{N}U_N$ .

In the above result (1,2) are really important, and (3,4,5) are rather technical remarks. As a consequence, coming from (1,2), let us formulate:

THEOREM 8.9. The following matrices, called generalized Fourier matrices,

$$F_{N_1,\ldots,N_k} = F_{N_1} \otimes \ldots \otimes F_{N_k}$$

are Hadamard, for any choice of  $N_1, \ldots, N_k$ . In particular the following matrices,

$$W_N = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}^{\otimes l}$$

having size  $N = 2^k$ , and called Walsh matrices, are all Hadamard.

**PROOF.** This is a consequence of Theorem 8.8, as follows:

(1) The first assertion follows from (1,2) in Theorem 8.8.

(2) The second assertion is a consequence of (1), by taking  $N_1 = \ldots = N_k = 2$ . Indeed, the generalized Fourier matrix that we obtain is:

$$F_{2,\dots,2} = F_2^{\otimes k} = \begin{pmatrix} 1 & 1\\ 1 & -1 \end{pmatrix}^{\otimes k}$$

Thus, we are led to the conclusion in the statement.

As an illustration for the above result, the second Walsh matrix, which is an Hadamard matrix having real entries, as is the case with all the Walsh matrices, is as follows:

The generalized Fourier matrices, and in particular the Walsh matrices, have many applications, to questions in coding, radio transmissions, quantum physics, and many other questions. We will be back to them later, on several occasions.

182

### 8b. Circulant matrices

With the above theory in hand, let us go back now to the general linear algebra considerations from the beginning of this chapter.

We have seen that  $F_N$  diagonalizes in an elegant way the flat matrix  $\mathbb{I}_N$ , and the idea in what follows will be that of  $F_N$ , or other real or complex Hadamard matrices, can be used in order to deal with other matrices, of  $\mathbb{I}_N$  type.

A first feature of the flat matrix  $\mathbb{I}_N$  is that it is circulant, in the following sense:

DEFINITION 8.10. A real or complex matrix M is called circulant if

$$M_{ij} = \xi_{j-i}$$

for a certain vector  $\xi$ , with the indices taken modulo N.

The circulant matrices are beautiful mathematical objects, which appear of course in many serious problems as well. As an example, at N = 4, we must have:

$$M = \begin{pmatrix} a & b & c & d \\ d & a & b & c \\ c & d & a & b \\ b & c & d & a \end{pmatrix}$$

While certainly gently looking, these matrices can be quite diabolic, when it comes to diagonalization, and other computations with them. For instance, when M is real, the computations with M are usually very complicated over the real numbers.

Fortunately the complex numbers and the Fourier matrices are there, and we have:

THEOREM 8.11. For a matrix  $M \in M_N(\mathbb{C})$ , the following are equivalent:

(1) M is circulant, in the sense that we have, for a certain vector  $\xi \in \mathbb{C}^N$ :

$$M_{ij} = \xi_{j-i}$$

(2) M is Fourier-diagonal, in the sense that, for a certain diagonal matrix Q:

$$M = F_N Q F_N^*$$

In addition, if these conditions hold, then  $\xi, Q$  are related by the formula

$$\xi = F_N^* q$$

where  $q \in \mathbb{C}^N$  is the column vector formed by the diagonal entries of Q.

**PROOF.** This follows from some basic computations with roots of unity, as follows:

(1)  $\implies$  (2) Assuming  $M_{ij} = \xi_{j-i}$ , the matrix  $Q = F_N^* M F_N$  is indeed diagonal, as shown by the following computation:

$$Q_{ij} = \sum_{kl} w^{-ik} M_{kl} w^{lj}$$
  
$$= \sum_{kl} w^{jl-ik} \xi_{l-k}$$
  
$$= \sum_{kr} w^{j(k+r)-ik} \xi_{r}$$
  
$$= \sum_{r} w^{jr} \xi_{r} \sum_{k} w^{(j-i)k}$$
  
$$= N \delta_{ij} \sum_{r} w^{jr} \xi_{r}$$

(2)  $\implies$  (1) Assuming  $Q = diag(q_1, \ldots, q_N)$ , the matrix  $M = F_N Q F_N^*$  is indeed circulant, as shown by the following computation:

$$M_{ij} = \sum_{k} w^{ik} Q_{kk} w^{-jk}$$
$$= \sum_{k} w^{(i-j)k} q_{k}$$

To be more precise, in this formula the last term depends only on j - i, and so shows that we have  $M_{ij} = \xi_{j-i}$ , with  $\xi$  being the following vector:

$$\xi_i = \sum_k w^{-ik} q_k$$
$$= (F_N^* q)_i$$

Thus, we are led to the conclusions in the statement.

As a basic illustration for the above result, for the circulant matrix  $M = \mathbb{I}_N$  we recover in this way the diagonalization result from Theorem 8.5, namely:

$$\begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix} = \frac{1}{N} F_N \begin{pmatrix} N & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{pmatrix} F_N^*$$

In general, the above result is something quite powerful, and very useful, and suggests doing everything in Fourier, when dealing with circulant matrices. Quite remarkably, this can be effectively done, because the Fourier transform behaves well with respect to the various other basic properties that our matrix might have, as follows:

THEOREM 8.12. The various basic sets of  $N \times N$  circulant matrices are as follows, with the convention that associated to any  $q \in \mathbb{C}^N$  is the matrix  $Q = diag(q_1, \ldots, q_N)$ :

(1) The set of all circulant matrices is:

$$M_N(\mathbb{C})^{circ} = \left\{ F_N Q F_N^* \middle| q \in \mathbb{C}^N \right\}$$

(2) The set of all circulant unitary matrices is:

$$U_N^{circ} = \left\{ \frac{1}{N} F_N Q F_N^* \middle| q \in \mathbb{T}^N \right\}$$

(3) The set of all circulant orthogonal matrices is:

$$O_N^{circ} = \left\{ \frac{1}{N} F_N Q F_N^* \middle| q \in \mathbb{T}^N, \bar{q}_i = q_{-i}, \forall i \right\}$$

In addition, in this picture, the first row vector of  $F_N Q F_N^*$  is always given by:

$$\xi = F_N^* q$$

**PROOF.** All this follows from Theorem 8.11, as follows:

(1) This assertion, along with the last one, is Theorem 8.11 itself.

(2) This is clear from (1), and from the fact that the rescaled matrix  $F_N/\sqrt{N}$  is unitary, because the eigenvalues of a unitary matrix must be on the unit circle  $\mathbb{T}$ .

(3) This follows from (2), because the matrix is real when  $\xi_i = \bar{\xi}_i$ , and in Fourier transform,  $\xi = F_N^* q$ , this corresponds to the condition  $\bar{q}_i = q_{-i}$ .

There are many other things that can be said about the circulant matrices, along these lines. Importantly, all this can be generalized to the setting of the matrices which are  $(N_1, \ldots, N_k)$  patterned, in a certain technical sense, and the matrix which does the job here is the corresponding generalized Fourier matrix, namely:

$$F_{N_1,\ldots,N_k} = F_{N_1} \otimes \ldots \otimes F_{N_k}$$

However, all this would lead us into the science of understanding patterns and their symmetries, which is called finite group theory, and that we have not developed yet. We will be back to this later, towards the end of the present book.

As a last topic regarding the circulant matrices, which is somehow one level above the above considerations, let us discuss the circulant Hadamard matrices. It is convenient to use the following notion, coming from the various results in Theorem 8.8 above:

DEFINITION 8.13. Two complex Hadamard matrices are called equivalent, and we write  $H \sim K$ , when it is possible to pass from H to K via the following operations:

- (1) Permuting the rows, or permuting the columns.
- (2) Multiplying the rows or columns by numbers in  $\mathbb{T}$ .

Here we have not taken into account all the results in Theorem 8.8 when formulating the above definition, because the operations  $H \to H^t, \bar{H}, H^*$  are far more subtle than those in (1,2) above, and can complicate things, if included in the equivalence.

The point now is that, up to equivalence, we can put the Fourier matrix  $F_N$  in circulant form. At small values of N, this can be done as follows:

**PROPOSITION 8.14.** The following are circulant and symmetric Hadamard matrices,

$$F_{2}' = \begin{pmatrix} i & 1 \\ 1 & i \end{pmatrix} , \quad F_{3}' = \begin{pmatrix} w & 1 & 1 \\ 1 & w & 1 \\ 1 & 1 & w \end{pmatrix}$$
$$F_{4}'' = \begin{pmatrix} -1 & \nu & 1 & \nu \\ \nu & -1 & \nu & 1 \\ 1 & \nu & -1 & \nu \\ \nu & 1 & \nu & -1 \end{pmatrix}$$

where  $w = e^{2\pi i/3}$ ,  $\nu = e^{\pi i/4}$ , equivalent to the Fourier matrices  $F_2, F_3, F_4$ .

PROOF. The orthogonality between rows being clear, we have here complex Hadamard matrices. The fact that we have an equivalence  $F_2 \sim F'_2$  follows from:

$$\begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \sim \begin{pmatrix} i & i \\ 1 & -1 \end{pmatrix} \sim \begin{pmatrix} i & 1 \\ 1 & i \end{pmatrix}$$

At N = 3 now, the equivalence  $F_3 \sim F_3'$  can be constructed as follows:

$$\begin{pmatrix} 1 & 1 & 1 \\ 1 & w & w^2 \\ 1 & w^2 & w \end{pmatrix} \sim \begin{pmatrix} 1 & 1 & w \\ 1 & w & 1 \\ w & 1 & 1 \end{pmatrix} \sim \begin{pmatrix} w & 1 & 1 \\ 1 & w & 1 \\ 1 & 1 & w \end{pmatrix}$$

As for the case N = 4, here the equivalence  $F_4 \sim F''_4$  can be constructed as follows, where we use the logarithmic notation  $[k]_s = e^{2\pi k i/s}$ , with respect to s = 8:

0	0	0	0		0	1	4	1		4	1	0	1	
0	2	4	6		1	4	1	0		1	4	1	0	
0	4	0	4	$\sim$	4	1	0	1		0	1	4	1	
0	6	4	2	8	1	0	1	4	8	1	0	1	4	8

Thus, the Fourier matrices  $F_2, F_3, F_4$  can be put indeed in circulant form.

We will explain later the reasons for denoting the above matrix  $F''_4$ , instead of  $F'_4$ , the idea being that  $F'_4$  will be a matrix belonging to a certain series.

In order to discuss now the general case, we will use a technical method for dealing with the circulant matrices, namely Björck's cyclic root formalism, which is as follows:

THEOREM 8.15. Assume that a matrix  $H \in M_N(\mathbb{T})$  is circulant,  $H_{ij} = \gamma_{j-i}$ . Then H is is a complex Hadamard matrix if and only if the vector

$$z = (z_0, z_1, \ldots, z_{N-1})$$

given by  $z_i = \gamma_i / \gamma_{i-1}$  satisfies the following equations:

 $z_0 + z_1 + \ldots + z_{N-1} = 0$   $z_0 z_1 + z_1 z_2 + \ldots + z_{N-1} z_0 = 0$   $\ldots$   $z_0 z_1 \ldots z_{N-2} + \ldots + z_{N-1} z_0 \ldots z_{N-3} = 0$  $z_0 z_1 \ldots z_{N-1} = 1$ 

If so is the case, we say that  $z = (z_0, \ldots, z_{N-1})$  is a cyclic N-root.

PROOF. This follows from a direct computation, the idea being that, with  $H_{ij} = \gamma_{j-i}$  as above, the orthogonality conditions between the rows are best written in terms of the variables  $z_i = \gamma_i / \gamma_{i-1}$ , and correspond to the equations in the statement.

Now back to the Fourier matrices, we have the following result:

THEOREM 8.16. Given  $N \in \mathbb{N}$ , construct the following complex numbers:

$$\nu = e^{\pi i/N}$$
 ,  $q = \nu^{N-1}$  ,  $w = \nu^2$ 

We have then a cyclic N-root as follows,

$$(q, qw, qw^2, \ldots, qw^{N-1})$$

and the corresponding complex Hadamard matrix  $F'_N$  is circulant and symmetric, and equivalent to the Fourier matrix  $F_N$ .

**PROOF.** Given two numbers  $q, w \in \mathbb{T}$ , let us find out when  $(q, qw, qw^2, \ldots, qw^{N-1})$  is a cyclic root. We have two conditions to be verified, as follows:

(1) In order for the = 0 equations in Theorem 8.15 to be satisfied, the value of q is irrelevant, and w must be a primitive N-root of unity.

(2) As for the = 1 equation in Theorem 8.15, this states that we must have:

$$q^N w^{\frac{N(N-1)}{2}} = 1$$

Thus, we must have  $q^N = (-1)^{N-1}$ , so with the values of  $q, w \in \mathbb{T}$  in the statement, we have a cyclic N-root. Now construct  $H_{ij} = \gamma_{j-i}$  as in Theorem 8.15. We have:

$$\gamma_k = \gamma_{-k} \iff q^{k+1} w^{\frac{k(k+1)}{2}} = q^{-k+1} w^{\frac{k(k-1)}{2}}$$
$$\iff q^{2k} w^k = 1$$
$$\iff q^2 = w^{-1}$$

But this latter condition holds indeed, because we have:

$$q^2 = \nu^{2N-2} = \nu^{-2} = w^{-1}$$

We conclude that our circulant matrix H is symmetric as well, as claimed. It remains to construct an equivalence  $H \sim F_N$ . In order to do this, observe that, due to our conventions  $q = \nu^{N-1}, w = \nu^2$ , the first row vector of H is given by:

$$\begin{aligned} \gamma_k &= q^{k+1} w^{\frac{k(k+1)}{2}} \\ &= \nu^{(N-1)(k+1)} \nu^{k(k+1)} \\ &= \nu^{(N+k-1)(k+1)} \end{aligned}$$

Thus, the entries of H are given by the following formula:

$$H_{-i,j} = H_{0,i+j}$$
  
=  $\nu^{(N+i+j-1)(i+j+1)}$   
=  $\nu^{i^2+j^2+2ij+Ni+Nj+N-1}$   
=  $\nu^{N-1} \cdot \nu^{i^2+Ni} \cdot \nu^{j^2+Nj} \cdot \nu^{2ij}$ 

With this formula in hand, we can now finish. Indeed, the matrix  $H = (H_{ij})$  is equivalent to the following matrix:

$$H' = (H_{-i,i})$$

Now regarding this latter matrix H', observe that in the above formula, the factors  $\nu^{N-1}$ ,  $\nu^{i^2+Ni}$ ,  $\nu^{j^2+Nj}$  correspond respectively to a global multiplication by a scalar, and to row and column multiplications by scalars. Thus H' is equivalent to the matrix H'' obtained from it by deleting these factors. But this latter matrix, given by  $H''_{ij} = \nu^{2ij}$  with  $\nu = e^{\pi i/N}$ , is precisely the Fourier matrix  $F_N$ , and we are done.

As an illustration, at N = 2, 3 we obtain the old matrices  $F'_2, F'_3$ . As for the case N = 4, here we obtain the following matrix, with  $\nu = e^{\pi i/4}$ :

$$F_4' = \begin{pmatrix} \nu^3 & 1 & \nu^7 & 1\\ 1 & \nu^3 & 1 & \nu^7\\ \nu^7 & 1 & \nu^3 & 1\\ 1 & \nu^7 & 1 & \nu^3 \end{pmatrix}$$

This matrix is equivalent to the matrix  $F_4''$  from Proposition 8.14, with the equivalence  $F_4' \sim F_4''$  being obtained by multiplying everything by the number  $\nu = e^{\pi i/4}$ .

### 8c. Bistochastic matrices

Getting back now to the main idea behind what we are doing, namely building on the relation between  $\mathbb{I}_N$  and  $F_N$ , let us study now the class of bistochastic matrices:

DEFINITION 8.17. A square matrix  $M \in M_N(\mathbb{C})$  is called bistochastic if each row and each column sum up to the same number:



If this happens only for the rows, or only for the columns, the matrix is called rowstochastic, respectively column-stochastic.

As the name indicates, these matrices are useful in statistics, with the case of the matrices having entries in [0, 1], which sum up to  $\lambda = 1$ , being the important one.

As a basic example of a bistochastic matrix, we have of course the flat matrix  $\mathbb{I}_N$ . In fact, the various above notions of stochasticity are closely related to  $\mathbb{I}_N$ , or rather to the all-one vector  $\xi$  that the matrix  $\mathbb{I}_N/N$  projects on, in the following way:

PROPOSITION 8.18. Let  $M \in M_N(\mathbb{C})$  be a square matrix.

- (1) M is row stochastic, with sums  $\lambda$ , when  $M\xi = \lambda\xi$ .
- (2) M is column stochastic, with sums  $\lambda$ , when  $M^t \xi = \lambda \xi$ .
- (3) M is bistochastic, with sums  $\lambda$ , when  $M\xi = M^t\xi = \lambda\xi$ .

**PROOF.** All these assertions are clear from definitions, because when multiplying a matrix by  $\xi$ , we obtain the vector formed by the row sums.

As an observation here, we can reformulate if we want the above statement in a purely matrix-theoretic form, by using the flat matrix  $\mathbb{I}_N$ , as follows:

**PROPOSITION 8.19.** Let  $M \in M_N(\mathbb{C})$  be a square matrix.

- (1) *M* is row stochastic, with sums  $\lambda$ , when  $M\mathbb{I}_N = \lambda \mathbb{I}_N$ .
- (2) M is column stochastic, with sums  $\lambda$ , when  $\mathbb{I}_N M = \lambda \mathbb{I}_N$ .
- (3) *M* is bistochastic, with sums  $\lambda$ , when  $M\mathbb{I}_N = \mathbb{I}_N M = \lambda \mathbb{I}_N$ .

**PROOF.** This follows from Proposition 8.18, and from the fact that both the rows and the columns of the flat matrix  $\mathbb{I}_N$  are copies of the all-one vector  $\xi$ .

In what follows we will be mainly interested in the unitary bistochastic matrices, which are quite interesting objects. These do not exactly cover the flat matrix  $\mathbb{I}_N$ , but cover instead the following related matrix, which appears in many linear algebra questions:

$$K_N = \frac{1}{N} \begin{pmatrix} 2 - N & 2 \\ & \ddots & \\ 2 & 2 - N \end{pmatrix}$$

As a first result, regarding such matrices, we have the following statement:

THEOREM 8.20. For a unitary matrix  $U \in U_N$ , the following conditions are equivalent:

- (1) H is bistochastic, with sums  $\lambda$ .
- (2) *H* is row stochastic, with sums  $\lambda$ , and  $|\lambda| = 1$ .
- (3) *H* is column stochastic, with sums  $\lambda$ , and  $|\lambda| = 1$ .

**PROOF.** By using a symmetry argument we just need to prove (1)  $\iff$  (2), and both the implications are elementary, as follows:

(1)  $\implies$  (2) If we denote by  $U_1, \ldots, U_N \in \mathbb{C}^N$  the rows of U, we have indeed:

$$1 = \langle U_1, U_1 \rangle$$
$$= \sum_i \langle U_1, U_i \rangle$$
$$= \sum_i \sum_j U_{1j} \overline{U}_{ij}$$
$$= \sum_j U_{1j} \sum_i \overline{U}_{ij}$$
$$= \sum_j U_{1j} \cdot \overline{\lambda}$$
$$= \lambda \cdot \overline{\lambda}$$
$$= |\lambda|^2$$

(2)  $\implies$  (1) Consider the all-one vector  $\xi = (1)_i \in \mathbb{C}^N$ . The fact that U is rowstochastic with sums  $\lambda$  reads:

$$\sum_{j} U_{ij} = \lambda, \forall i \quad \Longleftrightarrow \quad \sum_{j} U_{ij} \xi_j = \lambda \xi_i, \forall i$$
$$\iff \quad U\xi = \lambda \xi$$

Also, the fact that U is column-stochastic with sums  $\lambda$  reads:

$$\sum_{i} U_{ij} = \lambda, \forall j \iff \sum_{j} U_{ij}\xi_i = \lambda\xi_j, \forall j$$
$$\iff U^t\xi = \lambda\xi$$

We must prove that the first condition implies the second one, provided that the row sum  $\lambda$  satisfies  $|\lambda| = 1$ . But this follows from the following computation:

$$U\xi = \lambda \xi \implies U^* U\xi = \lambda U^* \xi$$
$$\implies \xi = \lambda U^* \xi$$
$$\implies \xi = \bar{\lambda} U^t \xi$$
$$\implies U^t \xi = \lambda \xi$$

Thus, we have proved both the implications, and we are done.

The unitary bistochastic matrices are stable under a number of operations, and in particular under taking products, and we have the following result, making the link with the Fourier matrix, and with various discrete Fourier analysis questions:

THEOREM 8.21. The real and complex bistochastic groups, which are the sets

$$B_N \subset O_N$$
$$C_N \subset U_N$$

consisting of matrices which are bistochastic, are isomorphic to  $O_{N-1}$ ,  $U_{N-1}$ .

**PROOF.** Let us pick a unitary matrix  $F \in U_N$  satisfying the following condition, where  $e_0, \ldots, e_{N-1}$  is the standard basis of  $\mathbb{C}^N$ , and where  $\xi$  is the all-one vector:

$$Fe_0 = \frac{1}{\sqrt{N}}\xi$$

Observe that such matrices  $F \in U_N$  exist indeed, the basic example being the normalized Fourier matrix, which with  $w = e^{2\pi i/N}$  is as follows:

$$F = \frac{1}{\sqrt{N}} (w^{ij})$$

We have then, by using the above property of F:

$$u\xi = \xi \iff uFe_0 = Fe_0$$
$$\iff F^*uFe_0 = e_0$$
$$\iff F^*uF = diag(1, w)$$

Thus we have isomorphisms as in the statement, given by:

$$w_{ij} \to (F^* u F)_{ij}$$

But this gives both the assertions.

We will be back to the real and complex bistochastic groups  $B_N \subset O_N$  and  $C_N \subset U_N$ later on in this book, when systematically doing group theory.

Finally, in relation with the Hadamard matrices, we already know that the Fourier matrices  $F_N$  can be put in circulant form, and so in bistochastic form. However, in what regards more general Hadamard matrices, things are quite tricky here.

As a first remark, the first Walsh matrix  $W_2$  can be put, and actually looks better in complex bistochastic form, modulo the standard equivalence relation:

$$\begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \sim \begin{pmatrix} i & i \\ 1 & -1 \end{pmatrix} \sim \begin{pmatrix} i & 1 \\ 1 & i \end{pmatrix}$$

The second Walsh matrix  $W_4 = W_2 \otimes W_2$  can be put as well, and also looks as well better, in complex bistochastic form:

By using the above formulae, we are led to the following statement:

**PROPOSITION 8.22.** All the Walsh matrices,  $W_N = W_2^{\otimes n}$  with  $N = 2^n$ , can be put in bistochastic form, up to the standard equivalence relation, as follows:

(1) The matrices  $W_N$  with  $N = 4^n$  admit a real bistochastic form, namely:

$$W_N \sim \begin{pmatrix} -1 & 1 & 1 & 1 \\ 1 & -1 & 1 & 1 \\ 1 & 1 & -1 & 1 \\ 1 & 1 & 1 & -1 \end{pmatrix}^{\otimes}$$

(2) The matrices  $W_N$  with  $N = 2 \times 4^n$  admit a complex bistochastic form, namely:

$$W_N \sim \begin{pmatrix} i & 1 \\ 1 & i \end{pmatrix} \otimes \begin{pmatrix} -1 & 1 & 1 & 1 \\ 1 & -1 & 1 & 1 \\ 1 & 1 & -1 & 1 \\ 1 & 1 & 1 & -1 \end{pmatrix}^{\otimes i}$$

**PROOF.** This follows indeed from the above discussion.

All this is quite interesting, because it suggests that even when studying the real Hadamard matrices,  $H \in M_N(\pm 1)$ , which is normally a combinatorial business, going to complex numbers, by putting such matrices in complex bistochastic form,  $H' \in M_N(\mathbb{T})$ , might be a good idea. We will be back to this, with further comments, later on.

Regarding now the question of putting the general Hadamard matrices, real or complex, in complex bistochastic form, things here are tricky. We first have:

192

THEOREM 8.23. The class of the bistochastic complex Hadamard matrices has the following properties:

- (1) It contains the circulant symmetric forms  $F'_N$  of the Fourier matrices  $F_N$ .
- (2) It is stable under permuting rows and columns.
- (3) It is stable under taking tensor products.

In particular, any generalized Fourier matrix  $F_{N_1,\ldots,N_k} = F_{N_1} \otimes \ldots \otimes F_{N_k}$  can be put in bistochastic and symmetric form, up to the equivalence relation.

**PROOF.** We have several things to be proved, the idea being as follows:

(1) We know from the above that any Fourier matrix  $F_N$  has a circulant and symmetric form  $F'_N$ . But since circulant implies bistochastic, this gives the result.

(2) The claim regarding permuting rows and columns is clear.

(3) Assuming that H, K are bistochastic, with sums  $\lambda, \mu$ , we have:

$$\sum_{ia} (H \otimes K)_{ia,jb} = \sum_{ia} H_{ij} K_{ab}$$
$$= \sum_{i} H_{ij} \sum_{a} K_{ab}$$
$$= \lambda \mu$$

We have as well the following computation:

$$\sum_{jb} (H \otimes K)_{ia,jb} = \sum_{jb} H_{ij} K_{ab}$$
$$= \sum_{j} H_{ij} \sum_{b} K_{ab}$$
$$= \lambda \mu$$

Thus, the matrix  $H \otimes K$  is bistochastic as well.

(4) As for the last assertion, this follows from (1,2,3).

In general now, putting an arbitrary complex Hadamard matrix in bistochastic form can be theoretically done, according to a general theorem of Idel-Wolf. The proof of this latter theorem is however based on a quite advanced, and non-explicit argument, coming from symplectic geometry, and there are many interesting open questions here.

# 8d. Hadamard conjecture

As a final topic for this chapter, let us discuss the real Hadamard matrices. The definition here, going back to 19th century work of Sylvester, is as follows:

DEFINITION 8.24. A real Hadamard matrix is a square binary matrix,

 $H \in M_N(\pm 1)$ 

whose rows are pairwise orthogonal, with respect to the scalar product on  $\mathbb{R}^N$ .

Observe that we do not really need real numbers in order to talk about the Hadamard matrices, because the orthogonality condition tells us that, when comparing two rows, the number of matchings should equal the number of mismatchings.

As a first result regarding such matrices, we have:

**PROPOSITION 8.25.** For a square matrix  $H \in M_N(\pm 1)$ , the following are equivalent:

- (1) The rows of H are pairwise orthogonal, and so H is Hadamard.
- (2) The columns of H are pairwise orthogonal, and so  $H^t$  is Hadamard.
- (3) The rescaled matrix  $U = H/\sqrt{N}$  is orthogonal,  $U \in O_N$ .

PROOF. This is something that we already know for the complex Hadamard matrices, with the orthogonal group  $O_N$  being replaced by the unitary group  $U_N$ . In the real case the proof is similar, with everything coming from definitions, and linear algebra.

As an abstract consequence of the above result, let us record:

THEOREM 8.26. The set of the  $N \times N$  Hadamard matrices is

$$Y_N = M_N(\pm 1) \cap \sqrt{NO_N}$$

where  $O_N$  is the orthogonal group, the intersection being taken inside  $M_N(\mathbb{R})$ .

PROOF. This follows from Proposition 8.25, which tells us that an arbitrary matrix  $H \in M_N(\pm 1)$  belongs to  $Y_N$  if and only if it belongs to  $\sqrt{NO_N}$ .

As a conclusion here, the set  $Y_N$  that we are interested in appears as a kind of set of "special rational points" of the real algebraic manifold  $\sqrt{NO_N}$ . Thus, we are doing some kind of algebraic geometry here, of precise type to be determined.

As before in the complex matrix case, it is convenient to introduce:

DEFINITION 8.27. Two real Hadamard matrices are called equivalent, and we write  $H \sim K$ , when it is possible to pass from H to K via the following operations:

- (1) Permuting the rows, or the columns.
- (2) Multiplying the rows or columns by -1.

Observe that we do not include the transposition operation  $H \to H^t$  in our list of allowed operations. This is because Proposition 8.25 above, while looking quite elementary, rests however on a deep linear algebra fact, namely that the transpose of an orthogonal matrix is orthogonal as well, and this can produce complications later on.

Let us do now some classification work. Here is the result at N = 4:

**PROPOSITION 8.28.** There is only one Hadamard matrix at N = 4, namely

$$W_4 = W_2 \otimes W_2$$

up to the standard equivalence relation for such matrices.

**PROOF.** Consider an Hadamard matrix  $H \in M_4(\pm 1)$ , assumed to be dephased:

$$H = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & a & b & c \\ 1 & d & e & f \\ 1 & g & h & i \end{pmatrix}$$

By orthogonality of the first 2 rows we must have  $\{a, b, c\} = \{-1, -1, 1\}$ , and so by permuting the last 3 columns, we can further assume that our matrix is as follows:

$$H = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & m & n & o \\ 1 & p & q & r \end{pmatrix}$$

By orthogonality of the first 2 columns we must have  $\{m, p\} = \{-1, 1\}$ , and so by permuting the last 2 rows, we can further assume that our matrix is as follows:

$$H = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & x & y \\ 1 & -1 & z & t \end{pmatrix}$$

But this gives the result, because from the orthogonality of the rows we obtain x = y = -1. Indeed, with these values of x, y, from the orthogonality of the columns we obtain z = -1, t = 1. Thus, up to equivalence we have  $H = W_4$ , as claimed.

The case N = 5 is excluded, because the orthogonality condition forces  $N \in 2\mathbb{N}$ . The point now is that the case N = 6 is excluded as well, because we have:

**PROPOSITION 8.29.** The size of an Hadamard matrix must be

$$N \in \{2\} \cup 4\mathbb{N}$$

and we have examples at any values  $N = 2^n$ .

**PROOF.** By permuting the rows and columns or by multiplying them by -1, as to rearrange the first 3 rows, we can always assume that our matrix looks as follows:

$$H = \begin{pmatrix} 1 \dots 1 & 1 \dots 1 & 1 \dots 1 & 1 \dots 1 & 1 \dots 1 \\ 1 \dots 1 & 1 & \dots 1 & -1 \dots -1 & -1 \dots -1 \\ 1 \dots 1 & -1 \dots -1 & 1 \dots \dots 1 & -1 \dots -1 \\ \vdots \dots \vdots \dots \vdots & \vdots \dots \vdots & \vdots & \vdots & \vdots \\ x & y & z & t \end{pmatrix}$$

The orthogonality conditions between the first 3 rows give the following system:

$$(1 \perp 2)$$
 :  $x + y = z + t$   
 $(1 \perp 3)$  :  $x + z = y + t$   
 $(2 \perp 3)$  :  $x + t = y + z$ 

The solution of this system being x = y = z = t, we must have 4|N, as claimed.  $\Box$ 

The above result, and various other findings, suggest the following conjecture:

CONJECTURE 8.30 (Hadamard Conjecture (HC)). There is at least one Hadamard matrix

$$H \in M_N(\pm 1)$$

for any integer  $N \in 4\mathbb{N}$ .

This conjecture, going back to the 19th century, is one of the most beautiful statements in combinatorics, linear algebra, and mathematics in general. Quite remarkably, the numeric verification so far goes up to the number of the beast:

$$\mathfrak{N} = 666$$

Our purpose now will be that of gathering some evidence for this conjecture. At N = 8 we have the Walsh matrix  $W_8$ . Thus, the next existence problem comes at N = 12. And here, we can use the following key construction, due to Paley:

THEOREM 8.31. Let  $q = p^r$  be an odd prime power, define

$$\chi: \mathbb{F}_q \to \{-1, 0, 1\}$$

by  $\chi(0) = 0$ ,  $\chi(a) = 1$  if  $a = b^2$  for some  $b \neq 0$ , and  $\chi(a) = -1$  otherwise, and finally set  $Q_{ab} = \chi(a - b)$ 

We have then constructions of Hadamard matrices, as follows:

(1) Paley 1: if q = 3(4) we have a matrix of size N = q + 1, as follows:

$$P_N^1 = 1 + \begin{pmatrix} 0 & 1 & \dots & 1 \\ -1 & & & \\ \vdots & & Q & \\ -1 & & & \end{pmatrix}$$

(2) Paley 2: if q = 1(4) we have a matrix of size N = 2q + 2, as follows:

$$P_N^2 = \begin{pmatrix} 0 & 1 & \dots & 1 \\ 1 & & & \\ \vdots & Q & \\ 1 & & & \end{pmatrix} \quad : \quad 0 \to \begin{pmatrix} 1 & -1 \\ -1 & -1 \end{pmatrix} \quad , \quad \pm 1 \to \pm \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

These matrices are skew-symmetric  $(H + H^t = 2)$ , respectively symmetric  $(H = H^t)$ .

PROOF. In order to simplify the presentation, we will denote by 1 all the identity matrices, of any size, and by  $\mathbb{I}$  all the rectangular all-one matrices, of any size as well.

It is elementary to check that the matrix  $Q_{ab} = \chi(a-b)$  has the following properties:

$$QQ^t = q1 - \mathbb{I}$$
$$Q\mathbb{I} = \mathbb{I}Q = 0$$

In addition, we have the following formulae, which are elementary as well, coming from the fact that -1 is a square in  $\mathbb{F}_q$  precisely when q = 1(4):

$$q = 1(4) \implies Q = Q^{t}$$
$$q = 3(4) \implies Q = -Q^{t}$$

With these observations in hand, the proof goes as follows:

(1) With our conventions for the symbols 1 and  $\mathbb{I}$ , the matrix in the statement is:

$$P_N^1 = \begin{pmatrix} 1 & \mathbb{I} \\ -\mathbb{I} & 1+Q \end{pmatrix}$$

With this formula in hand, the Hadamard matrix condition follows from:

$$P_N^1(P_N^1)^t = \begin{pmatrix} 1 & \mathbb{I} \\ -\mathbb{I} & 1+Q \end{pmatrix} \begin{pmatrix} 1 & -\mathbb{I} \\ \mathbb{I} & 1-Q \end{pmatrix}$$
$$= \begin{pmatrix} N & 0 \\ 0 & \mathbb{I}+1-Q^2 \end{pmatrix}$$
$$= \begin{pmatrix} N & 0 \\ 0 & N \end{pmatrix}$$

(2) If we denote by G, F the matrices in the statement, which replace respectively the 0, 1 entries, then we have the following formula for our matrix:

$$P_N^2 = \begin{pmatrix} 0 & \mathbb{I} \\ \mathbb{I} & Q \end{pmatrix} \otimes F + 1 \otimes G$$

With this formula in hand, the Hadamard matrix condition follows from:

$$(P_N^2)^2 = \begin{pmatrix} 0 & \mathbb{I} \\ \mathbb{I} & Q \end{pmatrix}^2 \otimes F^2 + \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \otimes G^2 + \begin{pmatrix} 0 & \mathbb{I} \\ \mathbb{I} & Q \end{pmatrix} \otimes (FG + GF)$$
$$= \begin{pmatrix} q & 0 \\ 0 & q \end{pmatrix} \otimes 2 + \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \otimes 2 + \begin{pmatrix} 0 & \mathbb{I} \\ \mathbb{I} & Q \end{pmatrix} \otimes 0$$
$$= \begin{pmatrix} N & 0 \\ 0 & N \end{pmatrix}$$

Finally, the last assertion is clear, from the above formulae relating  $Q, Q^t$ . The above constructions allow us to get well beyond the Walsh matrix level:

THEOREM 8.32. The HC is verified at least up to N = 88, as follows:

- (1) At N = 4, 8, 16, 32, 64 we have Walsh matrices.
- (2) At N = 12, 20, 24, 28, 44, 48, 60, 68, 72, 80, 84, 88 we have Paley 1 matrices.
- (3) At N = 36, 52, 76 we have Paley 2 matrices.
- (4) At N = 40,56 we have Paley 1 matrices tensored with  $W_2$ .

**PROOF.** First of all, the numbers in (1-4) are indeed all the multiples of 4, up to 88. As for the various assertions, the proof here goes as follows:

(1) This is clear.

(2) Since N-1 takes the values q = 11, 19, 23, 27, 43, 47, 59, 67, 71, 79, 83, 87, all prime powers, we can indeed apply the Paley 1 construction, in all these cases.

(3) Since N = 4(8) here, and N/2 - 1 takes the values q = 17, 25, 37, all prime powers, we can indeed apply the Paley 2 construction, in these cases.

(4) At N = 40 we have indeed  $P_{20}^1 \otimes W_2$ , and at N = 56 we have  $P_{28}^1 \otimes W_2$ .

As a continuation of all this, at N = 92 we have  $92 - 1 = 7 \times 13$ , so the Paley 1 construction does not work, and 92/2 = 46, so the Paley 2 construction, or tensoring with  $W_2$ , does not work either. However, the problem can be solved by using a computer.

Things get even worse at higher values of N, where more and more complicated constructions are needed. The whole subject is quite technical, and, as already mentioned, human knowledge here stops so far at  $\mathfrak{N} = 666$ .

Another well-known open question concerns the circulant case. Given a binary vector  $\gamma \in (\pm 1)^N$ , one can ask whether the matrix  $H \in M_N(\pm 1)$  defined by  $H_{ij} = \gamma_{j-i}$  is Hadamard or not. Here is a solution to the problem:

$$K_4 = \begin{pmatrix} -1 & 1 & 1 & 1\\ 1 & -1 & 1 & 1\\ 1 & 1 & -1 & 1\\ 1 & 1 & 1 & -1 \end{pmatrix}$$

More generally, any vector  $\gamma \in (\pm 1)^4$  satisfying  $\sum \gamma_i = \pm 1$  is a solution to the problem. The following conjecture, from the 50s, states that there are no other solutions:

CONJECTURE 8.33 (Circulant Hadamard Conjecture (CHC)). The only Hadamard matrices which are circulant are

$$K_4 = \begin{pmatrix} -1 & 1 & 1 & 1\\ 1 & -1 & 1 & 1\\ 1 & 1 & -1 & 1\\ 1 & 1 & 1 & -1 \end{pmatrix}$$

and its conjugates, regardless of the value of  $N \in \mathbb{N}$ .

### 8E. EXERCISES

The fact that such a simple-looking problem is still open might seem quite surprising. Indeed, if we denote by  $S \subset \{1, \ldots, N\}$  the set of positions of the -1 entries of  $\gamma$ , the Hadamard matrix condition is simply, for any  $k \neq 0$ , taken modulo N:

$$|S \cap (S+k)| = |S| - N/4$$

Thus, the above conjecture simply states that at  $N \neq 4$ , such a set S cannot exist. This is a well-known problem in combinatorics, raised by Ryser a long time ago.

Summarizing, we have many interesting questions in the real case. The situation is quite different from the one in complex case, where at any  $N \in \mathbb{N}$  we have the Fourier matrix  $F_N$ , which makes the HC problematics disappear. Since  $F_N$  can be put in circulant form, the CHC disappears as well. There are however many interesting questions in the complex case, for the most in relation with questions in quantum physics.

### 8e. Exercises

There have been many questions, ranging from elementary to quite difficult, left open in this chapter, and we will leave things like that. Our exercises here will focus on something new, or rather not really discussed in the above, namely the systematic development of the theory of complex Hadamard matrices. First, we have:

EXERCISE 8.34. Prove that the Fourier matrices  $F_2, F_3$ , which are given by

$$F_2 = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \quad , \quad F_3 = \begin{pmatrix} 1 & 1 & 1 \\ 1 & w & w^2 \\ 1 & w^2 & w \end{pmatrix}$$

with  $w = e^{2\pi i/3}$  are the only Hadamard matrices at N = 2, 3, up to equivalence.

Here the question about  $F_2$  is quite trivial, but the question about  $F_3$  is interesting, requiring studying the equation a + b + c = 0 in the plane, with  $a, b, c \in \mathbb{T}$ . Enjoy.

In order to deal now with the N = 4 case, we will need:

EXERCISE 8.35. If  $H \in M_M(\mathbb{T})$  and  $K \in M_N(\mathbb{T})$  are complex Hadamard matrices, prove that so is the matrix

$$H \otimes_Q K \in M_{MN}(\mathbb{T})$$

given by the following formula, with  $Q \in M_{M \times N}(\mathbb{T})$ ,

$$(H \otimes_Q K)_{ia,jb} = Q_{ib}H_{ij}K_{ab}$$

called Dită deformation of  $H \otimes K$ , with parameter Q.

Normally this is just a quick, standard verification. More difficult, however, is the question of explicitly writing down the matrices that can be constructed in this way, because this requires things like struggling with double indices. Coming soon.

Regarding now the classification at N = 4, we have:

EXERCISE 8.36. Prove that the only complex Hadamard matrices at N = 4 are, up to the standard equivalence relation, the matrices

$$F_4^q = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & q & -1 & -q \\ 1 & -q & -1 & q \end{pmatrix}$$

with  $q \in \mathbb{T}$ , which appear as Diță deformations of  $W_4 = F_2 \otimes F_2$ .

Here the first question is quite standard, in the spirit of the computations at N = 3, mentioned before. As for the second question, good luck here with the double indices.

In order to discuss the case N = 5, we first have:

EXERCISE 8.37. Given an Hadamard matrix  $H \in M_5(\mathbb{T})$ , chosen dephased,

$$H = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & a & x & * & * \\ 1 & y & b & * & * \\ 1 & * & * & * & * \\ 1 & * & * & * & * \end{pmatrix}$$

prove that the numbers a, b, x, y must satisfy the following equation:

$$(x-y)(x-ab)(y-ab) = 0$$

This is something quite tricky, called Haagerup lemma.

And here is now the exercise at N = 5:

EXERCISE 8.38. Prove that the only Hadamard matrix at N = 5 is the Fourier matrix,

$$F_5 = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & w & w^2 & w^3 & w^4 \\ 1 & w^2 & w^4 & w & w^3 \\ 1 & w^3 & w & w^4 & w^2 \\ 1 & w^4 & w^3 & w^2 & w \end{pmatrix}$$

with  $w = e^{2\pi i/5}$ , up to the standard equivalence relation for such matrices.

As a hint here, try first to apply the Haagerup lemma, all across the matrix. Finally, as a bonus exercise, take a look at the case N = 6, and report on what you found.

# Part III

# Matrix groups

Castles out of fairy tales Timbers shivered where once there sailed The lovesick men who caught her eye And no one knew but Lorelei

# CHAPTER 9

# Group theory

# 9a. Basic examples

We have seen so far the basics of linear algebra, with the conclusion that the theory is very useful, and quickly becomes non-trivial. We have seen as well some abstract applications, to questions in analysis, and combinatorics, and with some results on the infinite dimensional case as well. All this is of course very useful in physics.

In this second half of this book we discuss a related topic, which is of key interest, namely the matrix groups. The theory here is once again very useful in connection with various questions in physics, the general idea being that any physical system S has a group of symmetries G(S), whose study can lead to concrete conclusions about S.

Let us begin with some abstract aspects. An abstract group is something very simple, namely a set, with a composition operation, which must satisfy what we should expect from a "multiplication". The precise definition of the groups is as follows:

DEFINITION 9.1. A group is a set G endowed with a multiplication operation

 $(g,h) \to gh$ 

which must satisfy the following conditions:

- (1) Associativity: we have, (gh)k = g(hk), for any  $g, h, k \in G$ .
- (2) Unit: there is an element  $1 \in G$  such that g1 = 1g = g, for any  $g \in G$ .
- (3) Inverses: for any  $g \in G$  there is  $g^{-1} \in G$  such that  $gg^{-1} = g^{-1}g = 1$ .

The multiplication law is not necessarily commutative. In the case where it is, in the sense that gh = hg, for any  $g, h \in G$ , we call G abelian, and we usually denote its multiplication, unit and inverse operation as follows:

$$(g,h) \to g+h$$
$$0 \in G$$
$$g \to -g$$

However, this is not a general rule, and rather the converse is true, in the sense that if a group is denoted as above, this means that the group must be abelian.

### 9. GROUP THEORY

At the level of the examples, we have for instance the symmetric group  $S_N$ . There are many other examples of groups, with typically the basic sets of numbers being abelian groups, and the basic of matrices being non-abelian groups. Once again, this is of course not a general rule. Here are some basic examples and counterexamples:

**PROPOSITION** 9.2. We have the following groups, and non-groups:

(1)  $(\mathbb{Z}, +)$  is a group.

- (2)  $(\mathbb{Q}, +)$ ,  $(\mathbb{R}, +)$ ,  $(\mathbb{C}, +)$  are groups as well.
- (3)  $(\mathbb{N}, +)$  is not a group.
- (4)  $(\mathbb{Q}^*, \cdot)$  is a group.
- (5)  $(\mathbb{R}^*, \cdot)$ ,  $(\mathbb{C}^*, \cdot)$  are groups as well.
- (6)  $(\mathbb{N}^*, \cdot), (\mathbb{Z}^*, \cdot)$  are not groups.

**PROOF.** All this is clear from the definition of the groups, as follows:

(1) The group axioms are indeed satisfied for  $\mathbb{Z}$ , with the sum g + h being the usual sum, 0 being the usual 0, and -g being the usual -g.

(2) Once again, the axioms are satisfied for  $\mathbb{Q}, \mathbb{R}, \mathbb{C}$ , with the remark that for  $\mathbb{Q}$  we are using here the fact that the sum of two rational numbers is rational, coming from:

$$\frac{a}{b} + \frac{c}{d} = \frac{ad + bc}{bd}$$

(3) In  $\mathbb{N}$  we do not have inverses, so we do not have a group:

$$-1 \notin \mathbb{N}$$

(4) The group axioms are indeed satisfied for  $\mathbb{Q}^*$ , with the product gh being the usual product, 1 being the usual 1, and  $g^{-1}$  being the usual  $g^{-1}$ . Observe that we must remove indeed the element  $0 \in \mathbb{Q}$ , because in a group, any element must be invertible.

(5) Once again, the axioms are satisfied for  $\mathbb{R}^*, \mathbb{C}^*$ , with the remark that for  $\mathbb{C}$  we are using here the fact that the nonzero complex numbers can be inverted, coming from:

$$z\bar{z} = |z|^2$$

(6) Here in  $\mathbb{N}^*, \mathbb{Z}^*$  we do not have inverses, so we do not have groups, as claimed.  $\Box$ 

There are many interesting groups coming from linear algebra, as follows:

THEOREM 9.3. We have the following groups:

(1)  $(\mathbb{R}^{N}, +)$  and  $(\mathbb{C}^{N}, +)$ .

(2)  $(M_N(\mathbb{R}), +)$  and  $(M_N(\mathbb{C}), +)$ .

- (3)  $(GL_N(\mathbb{R}), \cdot)$  and  $(GL_N(\mathbb{C}), \cdot)$ , the invertible matrices.
- (4)  $(SL_N(\mathbb{R}), \cdot)$  and  $(SL_N(\mathbb{C}), \cdot)$ , with S standing for "special", meaning det = 1.
- (5)  $(O_N, \cdot)$  and  $(U_N, \cdot)$ , the orthogonal and unitary matrices.
- (6)  $(SO_N, \cdot)$  and  $(SU_N, \cdot)$ , with S standing as above for det = 1.

### 9A. BASIC EXAMPLES

**PROOF.** All this is clear from definitions, and from our linear algebra knowledge, from chapters 1-4 above:

(1) The axioms are indeed clearly satisfied for  $\mathbb{R}^N$ ,  $\mathbb{C}^N$ , with the sum being the usual sum of vectors, -v being the usual -v, and the null vector 0 being the unit.

(2) Once again, the axioms are clearly satisfied for  $M_N(\mathbb{R}), M_N(\mathbb{C})$ , with the sum being the usual sum of matrices, -M being the usual -M, and the null matrix 0 being the unit. Observe that what we have here is in fact a particular case of (1), because any  $N \times N$  matrix can be regarded as a  $N^2 \times 1$  vector, and so at the group level we have:

$$(M_N(\mathbb{R}), +) \simeq (\mathbb{R}^{N^2}, +)$$
$$(M_N(\mathbb{C}), +) \simeq (\mathbb{C}^{N^2}, +)$$

(3) Regarding now  $GL_N(\mathbb{R}), GL_N(\mathbb{C})$ , these are groups because the product of invertible matrices is invertible, according to the following formula:

$$(AB)^{-1} = B^{-1}A^{-1}$$

Observe that at N = 1 we obtain the groups  $(\mathbb{R}^*, \cdot), (\mathbb{C}^*, \cdot)$ . At  $N \geq 2$  the groups  $GL_N(\mathbb{R}), GL_N(\mathbb{C})$  are not abelian, because we do not have AB = BA in general.

(4) The sets  $SL_N(\mathbb{R})$ ,  $SL_N(\mathbb{C})$  formed by the real and complex matrices of determinant 1 are subgroups of the groups in (3), because of the following formula, which shows that the matrices satisfying det A = 1 are stable under multiplication:

$$\det(AB) = \det(A)\det(B)$$

(5) Regarding now  $O_N, U_N$ , here the group property is clear too from definitions, and is best seen by using the associated linear maps, because the composition of two isometries is an isometry. Equivalently, assuming  $U^* = U^{-1}$  and  $V^* = V^{-1}$ , we have:

$$(UV)^* = V^*U^*$$
  
=  $V^{-1}U^{-1}$   
=  $(UV)^{-1}$ 

(6) The sets of matrices  $SO_N$ ,  $SU_N$  in the statement are obtained by intersecting the groups in (4) and (5), and so they are groups indeed:

$$SO_N = O_N \cap SL_N(\mathbb{R})$$
  
 $SU_N = U_N \cap SL_N(\mathbb{C})$ 

Thus, all the sets in the statement are indeed groups, as claimed.

We can see from the above that the notion of group is something extremely wide. In fact, in basically any mathematical question, there is a group somewhere around.

At the level of generality of Definition 9.1, there is nothing much that can be said. Let us record, however, as an example of what can be done with an arbitrary group:

9. GROUP THEORY

THEOREM 9.4. Given a group  $(G, \cdot)$ , we have the formula

 $(g^{-1})^{-1} = g$ 

valid for any element  $g \in G$ .

**PROOF.** This is clear from the definition of the inverses. Assume indeed that:

$$gg^{-1} = g^{-1}g = 1$$

We can write this formula in the following way:

$$g^{-1}g = gg^{-1} = 1$$

But this shows that g is the inverse of  $g^{-1}$ , as claimed.

As a side comment here, the above result, while being something trivial, has led to a lot of controversy among professional mathematicians, in recent times. The point indeed is that, for the needs of quantum mechanics, the notion of group must be replaced with something more general, called "quantum group", and the hot question is whether a quantum group must satisfy or not the above formula  $(g^{-1})^{-1} = g$ , taken in an appropriate sense. If you ever see on campus mathematicians or theoretical physicists fighting each other, covered in blood, they are probably discussing this issue.

# 9b. Dihedral groups

In order to have now some theory going, we obviously have to impose some conditions on the groups that we consider, and this because in general, the theory is too wide. With this idea in mind, let us work out some examples, in the finite group case.

The simplest possible finite group is the cyclic group  $\mathbb{Z}_N$ . There are many ways of picturing  $\mathbb{Z}_N$ , both additive and multiplicative, which are all useful, as follows:

DEFINITION 9.5. The cyclic group  $\mathbb{Z}_N$  is defined as follows:

- (1) As the additive group of remainders modulo N.
- (2) As the multiplicative group of the N-th roots of unity.

The two definitions are equivalent, because if we set  $w = e^{2\pi i/N}$ , then any remainder modulo N defines a N-th root of unity, according to the following formula:

 $k \to w^k$ 

We obtain in this way all the N-roots of unity, and so our correspondence is bijective. Moreover, our correspondence transforms the sum of remainders modulo N into the multiplication of the N-th roots of unity, due to the following formula:

$$w^k w^l = w^{k+l}$$

Thus, the groups defined in (1,2) above are isomorphic, via  $k \to w^k$ , and we agree to denote by  $\mathbb{Z}_N$  the corresponding group. Observe that this group  $\mathbb{Z}_N$  is abelian.

206

### 9B. DIHEDRAL GROUPS

Another interesting example of a finite group, which is more advanced, and which is non-abelian this time, is the dihedral group  $D_N$ , which appears as follows:

DEFINITION 9.6. The dihedral group  $D_N$  is the symmetry group of the regular polygon having N vertices, which is called regular N-gon.

Here are some basic examples of regular N-gons, at small values of the parameter  $N \in \mathbb{N}$ , and of their symmetry groups:

<u>N=2</u>. Here the N-gon is just a segment, and its symmetries are the identity *id* and the obvious symmetry  $\tau$ . Thus  $D_2 = \{id, \tau\}$ , and in group theory terms,  $D_2 = \mathbb{Z}_2$ .

<u>N=3</u>. Here the N-gon is an equilateral triangle, and the symmetries are the 3!=6 possible permutations of the vertices. Thus we have  $D_3 = S_3$ .

<u>N = 4</u>. Here the N-gon is a square, and as symmetries we have 4 rotations, of angles 0°, 90°, 180°, 270°, as well as 4 symmetries, with respect to the 4 symmetry axes, which are the 2 diagonals, and the 2 segments joining the midpoints of opposite sides.

<u>N = 5</u>. Here the N-gon is a regular pentagon, and as symmetries we have 5 rotations, of angles 0°, 72°, 144°, 216°, 288°, as well as 5 symmetries, with respect to the 5 symmetry axes, which join the vertices to the midpoints of the opposite sides.

<u>N = 6</u>. Here the *N*-gon is a regular hexagon, and we have 6 rotations, of angles 0°, 60°, 120°, 180°, 240°, 300°, and 6 symmetries, with respect to the 6 symmetry axes, which are the 3 diagonals, and the 3 segments joining the midpoints of opposite sides.

We can see from the above that the various dihedral groups  $D_N$  have many common features, and that there are some differences as well, for instance in what concerns the structure of the N symmetries, depending on whether  $N \in \mathbb{N}$  is even or odd.

In general, we have the following result, coming from the above observations:

**PROPOSITION 9.7.** The dihedral group  $D_N$  has 2N elements, as follows:

(1) We have N rotations  $R_1, \ldots, R_N$ , with  $R_k$  being the rotation of angle  $2k\pi/N$ . When labelling the vertices of the N-gon  $1, \ldots, N$ , the rotation formula is:

$$R_k: i \to k+i$$

(2) We have N symmetries  $S_1, \ldots, S_N$ , with  $S_k$  being the symmetry with respect to the Ox axis rotated by  $k\pi/N$ . The symmetry formula is:

$$S_k: i \to k-i$$

**PROOF.** This is clear, indeed. To be more precise,  $D_N$  consists of:

(1) The N rotations, of angles  $2k\pi/N$  with k = 1, ..., N. But these are exactly the rotations  $R_1, ..., R_N$  from the statement.

### 9. GROUP THEORY

(2) The N symmetries with respect to the N possible symmetry axes, which are the N medians of the N-gon when N is odd, and are the N/2 diagonals plus the N/2 lines connecting the midpoints of opposite edges, when N is even. But these are exactly the symmetries  $S_1, \ldots, S_N$  from the statement.

With the above description of  $D_N$  in hand, we can forget if we want about geometry and the regular N-gon, and talk about  $D_N$  abstractly, as follows:

THEOREM 9.8. The dihedral group  $D_N$  is the group having 2N elements,  $R_1, \ldots, R_N$ and  $S_1, \ldots, S_N$ , called rotations and symmetries, which multiply as follows,

$$R_k R_l = R_{k+l}$$
$$R_k S_l = S_{k+l}$$
$$S_k R_l = S_{k-l}$$
$$S_k S_l = R_{k-l}$$

with all the indices being taken modulo N.

**PROOF.** With notations from Proposition 9.7 above, the various compositions between rotations and symmetries can be computed as follows:

$$R_k R_l : i \to l + i \to k + l + i$$

$$R_k S_l : i \to l - i \to k + l - i$$

$$S_k R_l : i \to l + i \to k - l - i$$

$$S_k S_l : i \to l - i \to k - l + i$$

But these are exactly the formulae for  $R_{k+l}, S_{k+l}, S_{k-l}, R_{k-l}$ , as stated. Now since a group is uniquely determined by its multiplication rules, this gives the result.

Observe that  $D_N$  has the same cardinality as  $E_N = \mathbb{Z}_N \times \mathbb{Z}_2$ . We obviously don't have  $D_N \simeq E_N$ , because  $D_N$  is not abelian, while  $E_N$  is. So, our next goal will be that of proving that  $D_N$  appears by "twisting"  $E_N$ . In order to do this, let us start with:

**PROPOSITION** 9.9. The group  $E_N = \mathbb{Z}_N \times \mathbb{Z}_2$  is the group having 2N elements,  $r_1, \ldots, r_N$  and  $s_1, \ldots, s_N$ , which multiply according to the following rules,

 $r_k r_l = r_{k+l}$  $r_k s_l = s_{k+l}$  $s_k r_l = s_{k+l}$  $s_k s_l = r_{k+l}$ 

with all the indices being taken modulo N.

PROOF. With the notation  $\mathbb{Z}_2 = \{1, \tau\}$ , the elements of the product group  $E_N = \mathbb{Z}_N \times \mathbb{Z}_2$  can be labelled  $r_1, \ldots, r_N$  and  $s_1, \ldots, s_N$ , as follows:

$$r_k = (k, 1)$$
$$s_k = (k, \tau)$$

These elements multiply then according to the formulae in the statement. Now since a group is uniquely determined by its multiplication rules, this gives the result.  $\Box$ 

Let us compare now Theorem 9.8 and Proposition 9.9. In order to formally obtain  $D_N$  from  $E_N$ , we must twist some of the multiplication rules of  $E_N$ , namely:

$$s_k r_l = s_{k+l} \to s_{k-l}$$
$$s_k s_l = r_{k+l} \to r_{k-l}$$

Informally, this amounts in following the rule " $\tau$  switches the sign of what comes afterwards", and we are led in this way to the following definition:

DEFINITION 9.10. Given two groups A, G, with an action  $A \curvearrowright G$ , the crossed product

 $P = G \rtimes A$ 

is the set  $G \times A$ , with multiplication as follows:

$$(g,a)(h,b) = (gh^a,ab)$$

It is routine to check that P is indeed a group. Observe that when the action is trivial,  $h^a = h$  for any  $a \in A$  and  $h \in H$ , we obtain the usual product  $G \times A$ .

Now by getting back to the dihedral group  $D_N$ , we can improve Theorem 9.8 above, into a final result on the subject, as follows:

THEOREM 9.11. We have a crossed product decomposition as follows,

$$D_N = \mathbb{Z}_N \rtimes \mathbb{Z}_2$$

with  $\mathbb{Z}_2 = \{1, \tau\}$  acting on  $\mathbb{Z}_N$  via switching signs,  $k^{\tau} = -k$ .

**PROOF.** We have an action  $\mathbb{Z}_2 \curvearrowright \mathbb{Z}_N$  given by the formula in the statement, namely  $k^{\tau} = -k$ , so we can consider the corresponding crossed product group:

$$P_N = \mathbb{Z}_N \rtimes \mathbb{Z}_2$$

In order to understand the structure of  $P_N$ , we follow Proposition 9.9. The elements of  $P_N$  can indeed be labelled  $\rho_1, \ldots, \rho_N$  and  $\sigma_1, \ldots, \sigma_N$ , as follows:

$$\rho_k = (k, 1)$$
$$\sigma_k = (k, \tau)$$

### 9. GROUP THEORY

Now when computing the products of such elements, we basically obtain the formulae in Proposition 9.9, perturbed as in Definition 9.10. To be more precise, we have:

$$\rho_k \rho_l = \rho_{k+l}$$
$$\rho_k \sigma_l = \sigma_{k+l}$$
$$\sigma_k \rho_l = \sigma_{k+l}$$
$$\sigma_k \sigma_l = \rho_{k+l}$$

But these are exactly the multiplication formulae for  $D_N$ , from Theorem 9.8. Thus, we have an isomorphism  $D_N \simeq P_N$  given by  $R_k \to \rho_k$  and  $S_k \to \sigma_k$ , as desired.

As a third basic example of a finite group, we have the symmetric group  $S_N$ . This is a group that we already met, when talking about the determinant:

THEOREM 9.12. The permutations of  $\{1, \ldots, N\}$  form a group, denoted  $S_N$ , and called symmetric group. This group has N! elements. The signature map

$$\varepsilon: S_N \to \mathbb{Z}_2$$

can be regarded as being a group morphism, with values in  $\mathbb{Z}_2 = \{\pm 1\}$ .

PROOF. These are things that we already know, from chapter 2 above. To be more precise, the group property is clear, and the formula  $|S_N| = N!$  is clear as well.

To be more precise, in order to construct an element  $\sigma \in S_N$ , we have:

- N choices for the value of  $\sigma(N)$ . - (N-1) choices for the value of  $\sigma(N-1)$ . - (N-2) choices for the value of  $\sigma(N-2)$ . : :

– and so on, up to 1 choice for the value of  $\sigma(1)$ .

Summing up, the number of choices for a permutation  $\sigma \in S_N$  is, as desired:

$$N(N-1)(N-2)\dots 1 = N!$$

Regarding now the last assertion, recall the following formula for the signatures of the permutations, that we know from chapter 2 above:

$$\varepsilon(\sigma\tau) = \varepsilon(\sigma)\varepsilon(\tau)$$

But this tells us precisely that  $\varepsilon$  is a group morphism, as stated.

We will be back to  $S_N$ , on several occasions.

At an even more advanced level now, we have the hyperoctahedral group  $H_N$ , which appears as follows:

### 9B. DIHEDRAL GROUPS

211

DEFINITION 9.13. The hyperoctahedral group  $H_N$  is the group of symmetries of the unit cube in  $\mathbb{R}^N$ , which is called N-cube.

The hyperoctahedral group is a quite interesting group, whose definition, as a symmetry group, reminds that of the dihedral group  $D_N$ . So, let us start our study in the same way as we did for  $D_N$ , with a discussion at small values of  $N \in \mathbb{N}$ :

<u>N = 1</u>. Here the 1-cube is the segment, whose symmetries are the identity *id* and the flip  $\tau$ . Thus, we obtain the group with 2 elements, which is a very familiar object:

$$H_1 = D_2 = S_2 = \mathbb{Z}_2$$

<u>N=2</u>. Here the 2-cube is the square, and so the corresponding symmetry group is the dihedral group  $D_4$ , which is a group that we know well:

$$H_2 = D_4 = \mathbb{Z}_4 \rtimes \mathbb{Z}_2$$

N = 3. Here the 3-cube is the usual cube, and the situation is considerably more complicated, because this usual cube has no less than 48 symmetries. Identifying and counting these symmetries is actually an excellent exercise.

All this looks quite complicated, but fortunately we can count  $H_N$ , at N = 3, and at higher N as well, by using some tricks, the result being as follows:

THEOREM 9.14. We have the cardinality formula

$$|H_N| = 2^N N!$$

coming from the fact that  $H_N$  is the symmetry group of the coordinate axes of  $\mathbb{R}^N$ .

**PROOF.** Consider the standard cube in  $\mathbb{R}^N$ , centered at 0, and having as vertices the points having coordinates  $\pm 1$ . With this picture in hand, it is clear that the symmetries of the cube coincide with the symmetries of the N coordinate axes of  $\mathbb{R}^N$ .

In order to count now these latter symmetries, a bit as we did for the dihedral group, observe first that we have N! permutations of these N coordinate axes.

But each of these permutations of the coordinate axes  $\sigma \in S_N$  can be further "decorated" by a sign vector  $e \in \{\pm 1\}^N$ , consisting of the possible  $\pm 1$  flips which can be applied to each coordinate axis, at the arrival. Thus, we have:

$$|H_N| = |S_N| \cdot |\mathbb{Z}_2^N|$$
$$= N! \cdot 2^N$$

Thus, we are led to the conclusions in the statement.

As in the dihedral group case, it is possible to go beyond this, with a crossed product decomposition, of quite special type, called wreath product decomposition:

9. GROUP THEORY

THEOREM 9.15. We have a wreath product decomposition as follows,

$$H_N = \mathbb{Z}_2 \wr S_N$$

which means by definition that we have a crossed product decomposition

$$H_N = \mathbb{Z}_2^N \rtimes S_N$$

with the permutations  $\sigma \in S_N$  acting on the elements  $e \in \mathbb{Z}_2^N$  as follows:

$$\sigma(e_1,\ldots,e_k) = (e_{\sigma(1)},\ldots,e_{\sigma(k)})$$

**PROOF.** As explained in the proof of Theorem 9.14 above, the elements of  $H_N$  can be identified with the pairs  $g = (e, \sigma)$  consisting of a permutation  $\sigma \in S_N$ , and a sign vector  $e \in \mathbb{Z}_2^N$ , so that at the level of the cardinalities, we have:

$$|H_N| = |\mathbb{Z}_2^N \times S_N|$$

To be more precise, given an element  $g \in H_N$ , the element  $\sigma \in S_N$  is the corresponding permutation of the N coordinate axes, regarded as unoriented lines in  $\mathbb{R}^N$ , and  $e \in \mathbb{Z}_2^N$  is the vector collecting the possible flips of these coordinate axes, at the arrival.

Now observe that the product formula for two such pairs  $g = (e, \sigma)$  is as follows, with the permutations  $\sigma \in S_N$  acting on the elements  $f \in \mathbb{Z}_2^N$  as in the statement:

$$(e,\sigma)(f,\tau) = (ef^{\sigma},\sigma\tau)$$

Thus, we are precisely in the framework of Definition 9.10 above, and we conclude that we have a crossed product decomposition, as follows:

$$H_N = \mathbb{Z}_2^N \rtimes S_N$$

Thus, we are led to the conclusion in the statement, with the formula  $H_N = \mathbb{Z}_2 \wr S_N$ being just a shorthand for the decomposition  $H_N = \mathbb{Z}_2^N \rtimes S_N$  that we found.  $\Box$ 

Summarizing, we have so far many interesting examples of finite groups, and as a sequence of main examples, we have the following groups:

$$\mathbb{Z}_N \subset D_N \subset S_N \subset H_N$$

We will be back to these fundamental finite groups later on, on several occasions, with further results on them, both of algebraic and of analytic type.

# 9c. Cayley embeddings

At the level of the general theory now, we have the following fundamental result regarding the finite groups, due to Cayley:

THEOREM 9.16. Given a finite group G, we have an embedding as follows,

$$G \subset S_N$$
 ,  $g \to (h \to gh)$ 

with N = |G|. Thus, any finite group is a permutation group.

### 9C. CAYLEY EMBEDDINGS

**PROOF.** Given a group element  $g \in G$ , we can associate to it the following map:

$$\sigma_g: G \to G \quad , \quad h \to gh$$

Since gh = gh' implies h = h', this map is bijective, and so is a permutation of G. Thus, with N = |G|, we can view this map as a usual permutation,  $\sigma_G \in S_N$ .

Summarizing, we have constructed so far a map as follows:

$$G \to S_N$$
 ,  $g \to \sigma_g$ 

Our first claim is that this is a group morphism. Indeed, this follows from:

$$\sigma_g \sigma_h(k) = \sigma_g(hk)$$
  
=  $ghk$   
=  $\sigma_{gh}(k)$ 

It remains to prove that this group morphism is injective. But this follows from:

$$g \neq h \implies \sigma_g(1) \neq \sigma_h(1)$$
$$\implies \sigma_g \neq \sigma_h$$

Thus, we are led to the conclusion in the statement.

Observe that in the above statement the embedding  $G \subset S_N$  that we constructed depends on a particular writing  $G = \{g_1, \ldots, g_N\}$ , which is needed in order to identify the permutations of G with the elements of the symmetric group  $S_N$ .

As another remark now, of rather warning, for the basic examples of groups that we know, the Cayley theorem provides us with embeddings as follows:

$$\mathbb{Z}_N \subset S_N$$
$$D_N \subset S_{2N}$$
$$S_N \subset S_{N!}$$
$$H_N \subset S_{2^N N!}$$

Here the first embedding is the good one, the second one is not the best possible one, but can be useful, and the third and fourth embeddings are useless.

Thus, as a conclusion, the Cayley theorem remains something quite theoretical. We will be back to this later on, with a systematic study of the "representation" problem.

Getting back now to our main series of finite groups,  $\mathbb{Z}_N \subset D_N \subset S_N \subset H_N$ , these are of course permutation groups, according to the above. However, and perhaps even more interestingly, these are as well subgroups of the orthogonal group  $O_N$ :

$$\mathbb{Z}_N \subset D_N \subset S_N \subset H_N \subset O_N$$

### 9. GROUP THEORY

Indeed, it is enough to check here that we have an embedding  $H_N \subset O_N$ , and this is clear from the definition of  $H_N$ , because any transformation of the unit cube in  $\mathbb{R}^N$  must extend into an isometry of the whole  $\mathbb{R}^N$ , in the obvious way.

In view of this, it makes sense to look at the finite subgroups  $G \subset O_N$ . And this is what we will do, but with two modifications, namely: (1) although we do not have examples yet, it is better to look at the general subgroups  $G \subset U_N$ , and (2) it is better to replace the assumption that G is finite by the assumption that G is compact, and this in order to cover some interesting continuous groups, such as  $O_N, U_N, SO_N, SU_N$ .

Thus, we are led to the study of the closed subgroups  $G \subset U_N$ . Let us start our discussion here with the following simple fact:

**PROPOSITION 9.17.** The closed subgroups  $G \subset U_N$  are precisely the closed sets of matrices  $G \subset U_N$  satisfying the following conditions:

- (1)  $U, V \in G \implies UV \in G$ .
- $(2) \ 1 \in G.$
- $(3) \ U \in G \implies U^{-1} \in G.$

**PROOF.** This is clear from definitions, the only point with this statement being the fact that a subset  $G \subset U_N$  can be a group or not, as indicated above.

It is possible to get beyond this, first with a result stating that any closed subgroup  $G \subset U_N$  is a smooth manifold, and so is a compact Lie group, having a Lie algebra and so on. Moreover, and importantly, a converse of this fact is known to hold, in the sense that any compact Lie group appears as a closed subgroup  $G \subset U_N$  of a unitary group.

However, all this is quite advanced, and we will not need it, in what follows.

Let us go back now to the finite groups. We first have the following key result:

THEOREM 9.18. We have a group embedding as follows, obtained by regarding  $S_N$  as the permutation group of the N coordinate axes of  $\mathbb{R}^N$ ,

 $S_N \subset O_N$ 

which makes  $\sigma \in S_N$  correspond to the matrix having 1 on row i and column  $\sigma(i)$ , for any i, and having 0 entries elsewhere.

**PROOF.** The first assertion is clear, because the permutations of the N coordinate axes of  $\mathbb{R}^N$  are isometries. Regarding now the explicit formula, we have by definition:

$$\sigma(e_j) = e_{\sigma(j)}$$

Thus, the permutation matrix corresponding to  $\sigma$  is given by:

$$\sigma_{ij} = \begin{cases} 1 & \text{if } \sigma(j) = i \\ 0 & \text{otherwise} \end{cases}$$

Thus, we are led to the formula in the statement.

We can combine the above result with the Cayley theorem, and we obtain the following result, which is something very nice, having theoretical importance:

THEOREM 9.19. Given a finite group G, we have an embedding as follows,

$$G \subset O_N \quad , \quad g \to (e_h \to e_{gh})$$

with N = |G|. Thus, any finite group is an orthogonal matrix group.

**PROOF.** The Cayley theorem gives an embedding as follows:

$$G \subset S_N$$
 ,  $g \to (h \to gh)$ 

On the other hand, Theorem 9.18 provides us with an embedding as follows:

$$S_N \subset O_N \quad , \quad \sigma \to (e_i \to e_{\sigma(i)})$$

Thus, we are led to the conclusion in the statement.

The same remarks as for the Cayley theorem apply. First, the embedding  $G \subset O_N$  that we constructed depends on a particular writing  $G = \{g_1, \ldots, g_N\}$ . And also, for the basic examples of groups that we know, the embeddings that we obtain are as follows:

$$\mathbb{Z}_N \subset O_N$$
$$D_N \subset O_{2N}$$
$$S_N \subset O_{N!}$$
$$H_N \subset O_{2^N N!}$$

As before, here the first embedding is the good one, the second one is not the best possible one, but can be useful, and the third and fourth embeddings are useless.

Summarizing, in order to advance, it is better to forget about the Cayley theorem, and build on Theorem 9.18 instead. In relation with the "basic" groups, we have:

THEOREM 9.20. We have the following finite groups of matrices:

- (1)  $\mathbb{Z}_N$ , the cyclic permutation matrices.
- (2)  $D_N$ , the dihedral matrices.
- (3)  $S_N$ , the permutation matrices.
- (4)  $H_N$ , the signed permutation matrices.

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## 9. GROUP THEORY

**PROOF.** This is clear, indeed:

(1) The cyclic permutation matrices are as follows, and form a group:

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(2) The dihedral matrices form a group, indeed.

(3) These are just the permutation matrices.

(4) These are just the signed permutation matrices.

The above groups are all groups of orthogonal matrices. When looking into general unitary matrices, we led to the following interesting class of groups:

DEFINITION 9.21. The complex reflection group  $H_N^s \subset U_N$ , depending on parameters

$$N \in \mathbb{N}$$
 ,  $s \in \mathbb{N} \cup \{\infty\}$ 

is the group of permutation-type matrices with s-th roots of unity as entries,

$$H_N^s = M_N(\mathbb{Z}_s \cup \{0\}) \cap U_N$$

with the convention  $\mathbb{Z}_{\infty} = \mathbb{T}$ , at  $s = \infty$ .

Observe that at s = 1, 2 we obtain the following groups:

$$H_N^1 = S_N$$
$$H_N^2 = H_N$$

Another important particular case is  $s = \infty$ , where we obtain a group which is actually not finite, but is still compact, denoted as follows:

$$K_N \subset U_N$$

In general, in analogy with what we know about  $S_N, H_N$ , we first have:

**PROPOSITION 9.22.** The number of elements of  $H_N^s$  with  $s \in \mathbb{N}$  is:

$$|H_N^s| = s^N N!$$

At  $s = \infty$ , the group  $K_N = H_N^{\infty}$  that we obtain is infinite.

PROOF. This is indeed clear from our definition of  $H_N^s$ , as a matrix group as above, because there are N! choices for a permutation-type matrix, and then  $s^N$  choices for the corresponding s-roots of unity, which must decorate the N nonzero entries.

216

Once again in analogy with what we already know at s = 1, 2, we have as well the following decomposition result:

THEOREM 9.23. We have a wreath product decomposition as follows,

$$H_N^s = \mathbb{Z}_s \wr S_N$$

which means by definition that we have a crossed product decomposition

$$H_N^s = \mathbb{Z}_s^N \rtimes S_N$$

with the permutations  $\sigma \in S_N$  acting on the elements  $e \in \mathbb{Z}_s^N$  as follows:

$$\sigma(e_1,\ldots,e_k) = (e_{\sigma(1)},\ldots,e_{\sigma(k)})$$

PROOF. As explained in the proof of Proposition 9.22, the elements of  $H_N^s$  can be identified with the pairs  $g = (e, \sigma)$  consisting of a permutation  $\sigma \in S_N$ , and a decorating vector  $e \in \mathbb{Z}_s^N$ , so that at the level of the cardinalities, we have:

$$|H_N| = |\mathbb{Z}_s^N \times S_N|$$

Now observe that the product formula for two such pairs  $g = (e, \sigma)$  is as follows, with the permutations  $\sigma \in S_N$  acting on the elements  $f \in \mathbb{Z}_s^N$  as in the statement:

$$(e,\sigma)(f,\tau) = (ef^{\sigma},\sigma\tau)$$

Thus, we are precisely in the framework of Definition 9.10 above, and we conclude that we have a crossed product decomposition, as follows:

$$H_N^s = \mathbb{Z}_s^N \rtimes S_N$$

Thus, we are led to the conclusion in the statement, with the formula  $H_N^s = \mathbb{Z}_s \wr S_N$  being just a shorthand for the decomposition  $H_N^s = \mathbb{Z}_s^N \rtimes S_N$  that we found.

Summarizing, and by focusing now only on the cases  $s = 1, 2, \infty$ , which are the most important, we have extended our series of basic unitary groups, as follows:

$$\mathbb{Z}_N \subset D_N \subset S_N \subset H_N \subset K_N$$

In addition to this, we have the groups  $H_N^s$  with  $s \in \{3, 4, \ldots, \}$ . However, these will not fit well into the above series of inclusions, because we only have:

$$s|t \implies H_N^s \subset H_N^t$$

Thus, we can only extend our series of inclusions as follows:

$$\mathbb{Z}_N \subset D_N \subset S_N \subset H_N \subset H_N^4 \subset H_N^8 \subset \ldots \subset K_N$$

We will be back later to the groups  $H_N^s$ , and to the other reflection groups, with some more theory, and with some generalizations as well.

## 9. GROUP THEORY

## 9d. Abelian groups

We have seen so far that the basic examples of groups, even taken finite, lead us into linear algebra, and more specifically, into the study of the groups of unitary matrices:

 $G \subset U_N$ 

This is indeed a good idea, and we will systematically do this in this book, starting from the next chapter. Before getting into this, however, let us go back to the definition of the abstract groups, from the beginning of this chapter, and make a last attempt of developing some useful general theory there, without relation to linear algebra.

Basic common sense suggests looking into the case of the finite abelian groups, which can only be far less complicated than the arbitrary finite groups. In order to develop some theory for such groups, let us start with a general definition, as follows:

DEFINITION 9.24. A representation of a finite group G is a group morphism

$$u: G \to U_N$$

into a unitary group. The character of such a representation is the function

 $\chi: G \to \mathbb{C}$  $q \to Tr(u_q)$ 

where Tr is the usual, unnormalized trace of the  $N \times N$  matrices.

We have already seen some examples of such representations, and we will be back to them, and to some general representation theory as well, gradually, in what follows.

For our purposes here we only need the case N = 1. Here the representations coincide with their characters, and are simply called characters of the group:

DEFINITION 9.25. The characters of a finite group G are the group morphisms as follows, from G to the unit circle in the complex plane:

 $\chi:G\to \mathbb{T}$ 

That is, the characters of G are its 1-dimensional representations, or, equivalently, are the characters of these 1-dimensional representations.

A group might have or not characters, and this is a subtle question, that we will investigate later. As a first observation, the characters from a group:

**PROPOSITION** 9.26. Given a finite group G, its characters

$$\chi: G \to \mathbb{T}$$

form a finite abelian group, denoted  $\widehat{G}$ , and called dual of G.

**PROOF.** There are several things to be proved, the idea being as follows:

(1) Our first claim is that the set  $\widehat{G}$  formed by the characters  $\chi : G \to \mathbb{T}$  is a group, with the pointwise multiplication of characters, namely:

$$(\chi\rho)(g) = \chi(g)\rho(g)$$

Indeed, if  $\chi, \rho$  are characters, so is  $\chi\rho$ , and so the multiplication is well-defined on  $\widehat{G}$ . Regarding the unit, this is the trivial character, constructed as follows:

$$1: G \to \mathbb{T}$$

$$g \rightarrow 1$$

Finally, we have inverses, with the inverse of  $\chi: G \to \mathbb{T}$  being its conjugate:

$$\bar{\chi}: G \to \mathbb{T}$$
$$g \to \overline{\chi(g)}$$

(2) Our next claim is that  $\widehat{G}$  is finite. Indeed, given a group element  $g \in G$ , we can talk about its order, which is smallest integer  $k \in \mathbb{N}$  such that  $g^k = 1$ . Now assuming that we have a character  $\chi : G \to \mathbb{T}$ , we have the following formula:

$$\chi(g)^k = 1$$

Thus  $\chi(g)$  must be one of the k-th roots of unity, and in particular there are finitely many choices for  $\chi(g)$ . Thus, there are finitely many choices for  $\chi$ , as desired.

(3) Finally, the fact that  $\widehat{G}$  is abelian follows from definitions, because the pointwise multiplication of functions, and in particular of characters, is commutative.

As a basic illustration for the above construction, we have:

THEOREM 9.27. The dual of a cyclic group is the group itself:

$$\widehat{\mathbb{Z}}_N = \mathbb{Z}_N$$

More generally, any finite direct product of cyclic groups

$$G = \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$$

is self-dual, in the sense that we have  $\widehat{G} = G$ .

**PROOF.** We have two assertions to be proved, the idea being as follows:

(1) A character  $\chi : \mathbb{Z}_N \to \mathbb{T}$  is uniquely determined by its value  $z = \chi(g)$  on the standard generator  $g \in \mathbb{Z}_N$ . But this value must satisfy:

$$z^{N} = 1$$

## 9. GROUP THEORY

Thus we must have  $z \in \mathbb{Z}_N$ , with the cyclic group  $\mathbb{Z}_N$  being regarded this time as being the group of N-th roots of unity. Now conversely, any N-th root of unity  $z \in \mathbb{Z}_N$ defines a character  $\chi : \mathbb{Z}_N \to \mathbb{T}$ , by setting, for any  $r \in \mathbb{N}$ :

$$\chi(g^r) = z^r$$

Thus we have an identification  $\widehat{\mathbb{Z}}_N = \mathbb{Z}_N$ , as claimed.

(2) Regarding now the case of a product of cyclic groups  $G = \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$ , the analysis here is similar, with a character  $\chi : G \to \mathbb{T}$  being uniquely determined by its values  $\chi(g_1), \ldots, \chi(g_k)$  on the standard generators of  $\mathbb{Z}_{N_1}, \ldots, \mathbb{Z}_{N_k}$ . But these values must belong respectively to  $\mathbb{Z}_{N_1}, \ldots, \mathbb{Z}_{N_k}$ , regarded now as groups of corresponding roots of unity in the complex plane, and this gives an identification  $\widehat{G} = G$ , as claimed.  $\Box$ 

We can get some further insight into duality by using the spectral theory methods developed in chapter 7 above, and we have the following result:

THEOREM 9.28. Any finite abelian group G has the following property:

$$G = \widehat{\widehat{G}}$$

Moreover, we have an isomorphism of commutative  $C^*$ -algebras as follows,

$$\mathbb{C}[G] \simeq C(\widehat{G})$$

and via this isomorphism, the functional defined by linearity and the formula

$$\int_{\widehat{G}} g = \delta_{g1}$$

for any  $g \in G$ , is the integration with respect to the normalized counting measure on  $\widehat{G}$ .

**PROOF.** We have several assertions here, the idea being as follows:

(1) Given a finite abelian group G, consider indeed the group algebra  $\mathbb{C}[G]$ , having as elements the formal combinations of elements of G. This is a commutative  $C^*$ -algebra, and so by the Gelfand theorem we obtain an isomorphism as follows:

$$\mathbb{C}[G] = C(X)$$

To be more precise, X is the space of the following Banach algebra characters:

$$\chi:\mathbb{C}[G]\to\mathbb{C}$$

The point now is that by delinearizing, such a Banach algebra character must come from a usual group character of G, as follows:

$$\chi:G\to\mathbb{T}$$

Thus  $X = \hat{G}$ , and we are led to the isomorphism in the statement, namely:

$$\mathbb{C}[G] \simeq C(G)$$

(2) Consider now the following canonical group morphism:

$$\begin{split} G \to \widehat{G} \\ g \to (\chi \to \chi(g)) \end{split}$$

By using the various identifications coming from (1), it follows that this group morphism is injective, and so bijective, and so we obtain  $G = \hat{G}$ , as claimed.

(3) Finally, the last assertion, regarding the integration, is clear as well.

By doing some further algebra, we can in fact say more, as follows:

THEOREM 9.29. The finite abelian groups are the following groups,

$$G = \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$$

and these groups are self-dual.

**PROOF.** This is something more tricky, the idea being as follows:

(1) For any prime number  $p \in \mathbb{N}$ , let us define indeed  $G_p \subset G$  to be the subset of elements having as order a power of p. It is then routine to check that  $G_p$  is a subgroup, and that we have a direct product decomposition as follows:

$$G = \prod_{p} G_{p}$$

(2) Thus, we are left with proving that each component  $G_p$  decomposes as a product of cyclic groups, having as orders powers of p, as follows:

$$G_p = \mathbb{Z}_{p^{r_1}} \times \ldots \times \mathbb{Z}_{p^{r_s}}$$

But this is something that can be checked by recurrence on  $|G_p|$ , via some routine computations, and so we are led to the conclusion in the statement.

As an application of the above, let us go back to the generalized Fourier matrices, from chapter 8 above. We have here the following result:

THEOREM 9.30. Given a finite abelian group G, with dual group  $\widehat{G} = \{\chi : G \to \mathbb{T}\}$ , consider the corresponding Fourier coupling, namely:

$$\mathcal{F}_G: G \times \widehat{G} \to \mathbb{T}$$
$$(i, \chi) \to \chi(i)$$

- (1) Via the standard isomorphism  $G \simeq \widehat{G}$ , this Fourier coupling can be regarded as a square matrix,  $F_G \in M_G(\mathbb{T})$ , which is a complex Hadamard matrix.
- (2) In the case of the cyclic group  $G = \mathbb{Z}_N$  we obtain in this way, via the standard identification  $\mathbb{Z}_N = \{1, \ldots, N\}$ , the Fourier matrix  $F_N$ .
- (3) In general, when using a decomposition  $G = \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$ , the corresponding Fourier matrix is given by  $F_G = F_{N_1} \otimes \ldots \otimes F_{N_k}$ .

## 9. GROUP THEORY

**PROOF.** This follows indeed by using the above finite abelian group theory:

(1) With the identification  $G \simeq \widehat{G}$  made our matrix is given by  $(F_G)_{i\chi} = \chi(i)$ , and the scalar products between the rows are computed as follows:

$$\langle R_i, R_j \rangle = \sum_{\chi} \chi(i) \overline{\chi(j)}$$
  
 $= \sum_{\chi} \chi(i-j)$   
 $= |G| \cdot \delta_{ij}$ 

Thus, we obtain indeed a complex Hadamard matrix.

(2) This follows from the well-known and elementary fact that, via the identifications  $\mathbb{Z}_N = \widehat{\mathbb{Z}_N} = \{1, \ldots, N\}$ , the Fourier coupling here is as follows, with  $w = e^{2\pi i/N}$ :

$$(i,j) \to w^{ij}$$

(3) We use here the following well-known formula, for the duals of products:

$$\widehat{H \times K} = \widehat{H} \times \widehat{K}$$

At the level of the corresponding Fourier couplings, we obtain from this:

$$F_{H\times K} = F_H \otimes F_K$$

Now by decomposing G into cyclic groups, as in the statement, and by using (2) for the cyclic components, we obtain the formula in the statement.  $\Box$ 

As a nice application of the above result, we have:

**PROPOSITION 9.31.** The Walsh matrix,  $W_N$  with  $N = 2^n$ , which is given by

$$W_N = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}^{\otimes n}$$

is the Fourier matrix of the finite abelian group  $K_N = \mathbb{Z}_2^n$ .

**PROOF.** We know that the first Walsh matrix is a Fourier matrix:

$$W_2 = F_2 = F_{K_2}$$

Now by taking tensor powers we obtain from this that we have, for any  $N = 2^n$ :

$$W_N = W_2^{\otimes n} = F_{K_2}^{\otimes n} = F_{K_2}^n = F_{K_N}$$

Thus, we are led to the conclusion in the statement.

Summarizing, we have now a better understanding of the generalized Fourier matrices, and of the complex Hadamard matrices in general, and also a new and fresh point of view on the various discrete Fourier analysis considerations from chapter 8 above.

### 9E. EXERCISES

## 9e. Exercises

There are many things that can be said about groups, especially in the matrix case,  $G \subset U_N$ , and we will discuss this later in this book. Our exercises here will rather focus on the abstract groups, as in the end of the present chapter, and we first have:

EXERCISE 9.32. Develop a full theory of the order of group elements,

 $ord(g) \in \mathbb{N} \cup \{\infty\}$ 

with examples and results, and prove in particular that we have the formula

 $ord(G) \mid |G|$ 

valid for any element  $g \in G$  of a finite group G.

Here we are a bit vague, but there are just so many things that can be done. As an example here, try to understand when |G| can be a power of a prime.

Along the same lines, in relation with what was discussed above, we have:

EXERCISE 9.33. Prove with full details that any finite abelian group decomposes as

$$G = \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$$

either by using abstract algebra, or spectral theory and matrix theory.

To be more precise, the abstract algebra approach was already outlined in the above, and the problem is that of clarifying the 2 steps mentioned there, by working out the previous exercise first, as a useful preliminary. It is possible as well to prove the above result by building on our spectral theory study, but this is more tricky.

Let us get now into a generalization of all the above, going beyond the finite group setting. In analogy with what we know about finite groups, let us start with:

EXERCISE 9.34. Given a locally compact abelian group G, prove that its group characters, which must be by definition continuous,

$$\chi:G\to\mathbb{T}$$

form a locally compact abelian group, denoted  $\widehat{G}$ , and called dual of G.

Here locally compact means that any group element  $g \in G$  has a neighborhood which is compact, a bit in analogy with what happens for the real numbers  $r \in \mathbb{R}$ .

At the level of the main examples now, we have:

EXERCISE 9.35. Prove that the integers are dual to the unit circle, and vice versa:

$$\widehat{\mathbb{Z}}=\mathbb{T}$$
 ,  $\widehat{\mathbb{T}}=\mathbb{Z}$ 

Also, prove that the group of real numbers is self-dual,  $\widehat{\mathbb{R}} = \mathbb{R}$ .

## 9. GROUP THEORY

To be more precise, we already know from the above that we have  $\widehat{\mathbb{Z}}_N = \mathbb{Z}_N$ , for any  $N \in \mathbb{N}$ , and the first question, regarding  $\mathbb{Z}$  and  $\mathbb{T}$ , is a kind of " $N = \infty$ " version of this. As for the second question, regarding  $\mathbb{R}$ , this is related to all this as well.

Getting now into spectral theory methods, we have here:

EXERCISE 9.36. Prove that any abelian group G has the following property:

$$G = \widehat{\widehat{G}}$$

Also, prove that we have an isomorphism of commutative  $C^*$ -algebras as follows,

$$C^*(G) \simeq C(\widehat{G})$$

with the integration over  $\widehat{G}$  corresponding to the functional  $\int_{\widehat{G}} g = \delta_{g_1}$ .

This is something that we already know from the above, in the case where G is finite, and the problem is that of carefully working out the extension to the compact setting, notably by clarifying what the algebra  $C^*(G)$  should exactly mean. As a bonus exercise, try as well to state and prove a similar result in the general, locally compact setting.

At a more advanced level now, we have:

EXERCISE 9.37. Prove that the finitely generated abelian groups are

 $G = \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$ 

with the convention  $\mathbb{Z}_{\infty} = \mathbb{Z}$ , and that the compact matrix abelian groups are

 $H = \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$ 

with this time the convention  $\mathbb{Z}_{\infty} = \mathbb{T}$ . Also, prove that  $G = \widehat{H}$  and  $H = \widehat{G}$ .

This exercise, generalizing everything that we know, or almost, is actually something quite tricky, requiring a good knowledge of both algebra and analysis.

Finally, on the same theme, but more philosophically now, we have:

EXERCISE 9.38. Clarify the relation between the dualities

$$\widehat{\mathbb{Z}}_N = \mathbb{Z}_N$$
$$\widehat{\mathbb{Z}} = \mathbb{T} \quad , \quad \widehat{\mathbb{T}} = \mathbb{Z}$$
$$\widehat{\mathbb{R}} = \mathbb{R}$$

and the various types of Fourier transforms available.

To be more precise here, the problem is that of understanding why the above 3 dualities correspond to the main 3 types of known Fourier transforms, namely discrete Fourier transforms, usual Fourier series, and usual Fourier transforms. With the remark that this is something that we already know, for the first duality.

# CHAPTER 10

# **Rotation groups**

# 10a. Rotation groups

We have seen in the previous chapter that there are many interesting examples of finite groups G, which usually appear as finite groups of unitary matrices,  $G \subset U_N$ , and that we have many interesting questions regarding these groups.

In this chapter we discuss similar questions, in the continuous case. We will be mainly interested here in the unitary group  $U_N$  itself, in its real version, which is the orthogonal group  $O_N$ , and in various technical versions of these basic groups  $O_N, U_N$ .

Let us start with some reminders, regarding  $O_N, U_N$ . These are the groups of rotations of  $\mathbb{R}^N$ ,  $\mathbb{C}^N$ , regarded as square matrices, the precise result being as follows:

THEOREM 10.1. We have the following results:

(1) The rotations of  $\mathbb{R}^N$  form the orthogonal group  $O_N$ , which is given by:

$$O_N = \left\{ U \in M_N(\mathbb{R}) \middle| U^t = U^{-1} \right\}$$

(2) The rotations of  $\mathbb{C}^N$  form the unitary group  $U_N$ , which is given by:

$$U_N = \left\{ U \in M_N(\mathbb{C}) \middle| U^* = U^{-1} \right\}$$

In addition, we can restrict the attention to the rotations of the corresponding spheres.

**PROOF.** This is something that we already know, the idea being as follows:

(1) We have seen in chapter 1 that a linear map  $T : \mathbb{R}^N \to \mathbb{R}^N$ , written as T(x) = Ux with  $U \in M_N(\mathbb{R})$ , is a rotation, in the sense that it preserves the distances and the angles, precisely when the associated matrix U is orthogonal, in the following sense:

$$U^t = U^{-1}$$

Thus, we obtain the result. As for the last assertion, this is clear as well, because an isometry of  $\mathbb{R}^N$  is the same as an isometry of the unit sphere  $S_{\mathbb{R}}^{N-1} \subset \mathbb{R}^N$ .

(2) We have seen in chapter 3 that a linear map  $T : \mathbb{C}^N \to \mathbb{C}^N$ , written as T(x) = Ux with  $U \in M_N(\mathbb{C})$ , is a rotation, in the sense that it preserves the distances and the scalar

products, precisely when the associated matrix U is unitary, in the following sense:

$$U^* = U^{-1}$$

Thus, we obtain the result. As for the last assertion, this is clear as well, because an isometry of  $\mathbb{C}^N$  is the same as an isometry of the unit sphere  $S_{\mathbb{C}}^{N-1} \subset \mathbb{C}^N$ .

Our next examples of continuous groups, which are very important as well, are the subgroups  $SO_N \subset O_N$  and  $SU_N \subset U_N$ , constructed by using the determinant.

In order to introduce these groups, let us start with the following simple fact:

**PROPOSITION 10.2.** We have the following results:

- (1) For an orthogonal matrix  $U \in O_N$  we have det  $U \in \{\pm 1\}$ .
- (2) For a unitary matrix  $U \in U_N$  we have  $\det U \in \mathbb{T}$ .

**PROOF.** This is clear from the equations defining  $O_N, U_N$ , as follows:

(1) We have indeed the following implications:

$$U \in O_N \implies U^t = U^{-1}$$
$$\implies \det U^t = \det U^{-1}$$
$$\implies \det U = (\det U)^{-1}$$
$$\implies \det U \in \{\pm 1\}$$

(2) We have indeed the following implications:

$$U \in U_N \implies U^* = U^{-1}$$
$$\implies \det U^* = \det U^{-1}$$
$$\implies \det U = (\det U)^{-1}$$
$$\implies \det U \in \mathbb{T}$$

Here we have used the fact that  $\overline{z} = z^{-1}$  means  $z\overline{z} = 1$ , and so  $z \in \mathbb{T}$ .

We can now introduce the subgroups  $SO_N \subset O_N$  and  $SU_N \subset U_N$ , as being the subgroups consisting of the rotations which preserve the orientation, as follows:

THEOREM 10.3. The following are groups of matrices,

$$SO_N = \left\{ U \in O_N \middle| \det U = 1 \right\}$$
$$SU_N = \left\{ U \in U_N \middle| \det U = 1 \right\}$$

consisting of the rotations which preserve the orientation.

**PROOF.** The fact that we have indeed groups follows from the properties of the determinant, of from the property of preserving the orientation, which is clear as well.  $\Box$ 

Summarizing, we have constructed so far 4 continuous groups of matrices, consisting of various rotations, with inclusions between them, as follows:



Observe that this is an intersection diagram, in the sense that:

$$SO_N = SU_N \cap O_N$$

As an illustration, let us work out what happens at N = 1, 2. At N = 1 the situation is quite trivial, and we obtain very simple groups, as follows:

**PROPOSITION 10.4.** The basic continuous groups at N = 1, namely



are the following groups of complex numbers,



or, equivalently, are the following cyclic groups,



with the convention that  $\mathbb{Z}_s$  is the group of s-th roots of unity.

**PROOF.** This is clear from definitions, because for a  $1 \times 1$  matrix the unitarity condition reads  $\overline{U} = U^{-1}$ , and so  $U \in \mathbb{T}$ , and this gives all the results.

At N = 2 now, let us first discuss the real case. The result here is as follows:

THEOREM 10.5. We have the following results:

(1)  $SO_2$  is the group of usual rotations in the plane, which are given by:

$$R_t = \begin{pmatrix} \cos t & -\sin t\\ \sin t & \cos t \end{pmatrix}$$

(2)  $O_2$  consists in addition of the usual symmetries in the plane, given by:

$$S_t = \begin{pmatrix} \cos t & \sin t \\ \sin t & -\cos t \end{pmatrix}$$

(3) Abstractly speaking, we have isomorphisms as follows:

$$SO_2 \simeq \mathbb{T}$$
 ,  $O_2 = \mathbb{T} \rtimes \mathbb{Z}_2$ 

(4) When discretizing all this, by replacing the 2-dimensional unit sphere  $\mathbb{T}$  by the regular N-gon, the latter isomorphism discretizes as  $D_N = \mathbb{Z}_N \rtimes \mathbb{Z}_2$ .

**PROOF.** This follows from some elementary computations, as follows:

(1) The first assertion is clear, because only the rotations of the plane in the usual sense preserve the orientation. As for the formula of  $R_t$ , this is something that we already know, from section 1 above, obtained by computing  $R_t \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $R_t \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ .

(2) The first assertion is clear, because rotations left aside, we are left with the symmetries of the plane, in the usual sense. As for formula of  $S_t$ , this is something that we basically know too, obtained by computing  $S_t \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $S_t \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ .

(3) The first assertion is clear, because the angles  $t \in \mathbb{R}$ , taken as usual modulo  $2\pi$ , form the group  $\mathbb{T}$ . As for the second assertion, the proof here is similar to the proof of the crossed product decomposition  $D_N = \mathbb{Z}_N \rtimes \mathbb{Z}_2$  for the dihedral groups.

(4) This is something more speculative, the idea here being that the isomorphism  $O_2 = \mathbb{T} \rtimes \mathbb{Z}_2$  appears from  $D_N = \mathbb{Z}_N \rtimes \mathbb{Z}_2$  by taking the  $N \to \infty$  limit.  $\Box$ 

# 10b. Euler-Rodrigues

In general, the structure of  $O_N$  and  $SO_N$ , and the relation between them, is far more complicated than what happens at N = 1, 2. We will be back to this later.

For the moment, let us keep working out what happens at N = 2, this time with a study in the complex case. We first have here the following result:

THEOREM 10.6. We have the following formula,

$$SU_2 = \left\{ \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix} \mid |a|^2 + |b|^2 = 1 \right\}$$

which makes  $SU_2$  isomorphic to the unit sphere  $S^1_{\mathbb{C}} \subset \mathbb{C}^2$ .

**PROOF.** Consider an arbitrary  $2 \times 2$  matrix, written as follows:

$$U = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

Assuming  $\det U = 1$ , the inverse is then given by:

$$U^{-1} = \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

On the other hand, assuming  $U \in U_2$ , the inverse must be the adjoint:

$$U^{-1} = \begin{pmatrix} \bar{a} & \bar{c} \\ \bar{b} & \bar{d} \end{pmatrix}$$

We are therefore led to the following equations:

$$d = \bar{a}$$
$$c = -\bar{b}$$

Thus our matrix must be of the following form:

$$U = \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix}$$

Since the determinant is 1, we must have, as stated:

$$|a|^2 + |b|^2 = 1$$

Thus, we are done with one direction. As for the converse, this is clear, the matrices in the statement being unitaries, and of determinant 1, and so being elements of  $SU_2$ .

Regarding the last assertion, recall that the unit sphere  $S^1_{\mathbb{C}} \subset \mathbb{C}^2$  is given by:

$$S_{\mathbb{C}}^{1} = \left\{ (a, b) \mid |a|^{2} + |b|^{2} = 1 \right\}$$

Thus, we have an isomorphism of compact spaces, as follows:

$$SU_2 \simeq S^1_{\mathbb{C}}$$
$$\begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix} \to (a, b)$$

We have therefore proved our theorem.

As a comment here, we will see later that the isomorphism  $SU_2 \simeq S_{\mathbb{C}}^1$  constructed above goes beyond the compact space level, for instance with a result stating that the uniform measure on  $S_{\mathbb{C}}^1$  corresponds in this way to the uniform measure on  $SU_2$ , which is by definition the unique probability measure which is invariant under translations by the various group elements  $g \in SU_2$ . We will be back to this later, with full details.

Regarding now the unitary group  $U_2$ , the result here is as follows:

THEOREM 10.7. We have the following formula,

$$U_2 = \left\{ d \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix} \mid |a|^2 + |b|^2 = 1, |d| = 1 \right\}$$

which makes  $U_2$  be a quotient compact space, as follows,

 $S^1_{\mathbb{C}} \times \mathbb{T} \to U_2$ 

but with this parametrization being no longer bijective.

**PROOF.** In one sense, this is clear, because we have:

$$|d| = 1 \implies dSU_2 \subset U_2$$

In the other sense, let  $U \in U_2$ . We have then:

$$|\det(U)|^2 = \det(U)\overline{\det(U)}$$
  
=  $\det(U)\det(U^*)$   
=  $\det(UU^*)$   
=  $\det(1)$   
= 1

Consider now the following complex number, up to a binary choice:

$$d = \sqrt{\det U}$$

We know from Proposition 10.2 that we have |d| = 1. Thus the rescaled matrix V = U/d is unitary,  $V \in U_2$ . As for the determinant of this matrix, this is given by:

$$det(V) = det(d^{-1}U)$$
  
=  $d^{-2} det(U)$   
=  $det(U)^{-1} det(U)$   
= 1

Thus we have  $V \in SU_2$ , and so we can write, with  $|a|^2 + |b|^2 = 1$ :

$$V = \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix}$$

Thus the matrix U = dV appears as in the statement. Finally, observe that the result that we have just proved provides us with a quotient map as follows:

$$S^1_{\mathbb{C}} \times \mathbb{T} \to U_2 \quad , \quad ((a,b),d) \to d \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix}$$

However, the parametrization is no longer bijective, because when we globally switch signs, the element ((-a, -b), -d) produces the same element of  $U_2$ .

In general, the structure of  $U_N$  and  $SU_N$ , and the relation between them, is far more complicated than what happens at N = 1, 2. We will be back to this later.

Let us record now a few more results regarding  $SU_2, U_2$ , which are key groups in mathematics and physics. First, we have the following reformulation of Theorem 10.6:

THEOREM 10.8. We have the formula

$$SU_{2} = \left\{ \begin{pmatrix} x + iy & z + it \\ -z + it & x - iy \end{pmatrix} \mid x^{2} + y^{2} + z^{2} + t^{2} = 1 \right\}$$

which makes  $SU_2$  isomorphic to the unit real sphere  $S^3_{\mathbb{R}} \subset \mathbb{R}^3$ .

PROOF. We recall from Theorem 10.6 above that we have:

$$SU_2 = \left\{ \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix} \mid |a|^2 + |b|^2 = 1 \right\}$$

Now let us write our parameters  $a, b \in \mathbb{C}$ , which belong to the complex unit sphere  $S^1_{\mathbb{C}} \subset \mathbb{C}^2$ , in terms of their real and imaginary parts, as follows:

$$a = x + iy$$
$$b = z + it$$

In terms of these new parameters  $x, y, z, t \in \mathbb{R}$ , our formula for a generic matrix  $U \in SU_2$ , that we established before, reads:

$$U = \begin{pmatrix} x + iy & z + it \\ -z + it & x - iy \end{pmatrix}$$

As for the condition to be satisfied by the parameters  $x, y, z, t \in \mathbb{R}$ , this comes the condition  $|a|^2 + |b|^2 = 1$  to be satisfied by  $a, b \in \mathbb{C}$ , which reads:

$$x^2 + y^2 + z^2 + t^2 = 1$$

Thus, we are led to the conclusion in the statement. Regarding now the last assertion, recall that the unit sphere  $S^3_{\mathbb{R}} \subset \mathbb{R}^4$  is given by:

$$S_{\mathbb{R}}^{3} = \left\{ (x, y, z, t) \mid x^{2} + y^{2} + z^{2} + t^{2} = 1 \right\}$$

Thus, we have an isomorphism of compact spaces, as follows:

$$SU_2 \simeq S^3_{\mathbb{R}}$$
  
+  $iy \quad z + it$ )

$$\begin{pmatrix} x+iy & z+it \\ -z+it & x-iy \end{pmatrix} \to (x,y,z,t)$$

We have therefore proved our theorem.

231

As before, we will see later that the isomorphism  $SU_2 \simeq S^3_{\mathbb{R}}$  constructed above goes beyond the compact space level, for instance with a result stating that the uniform measure on  $S^3_{\mathbb{R}}$  corresponds in this way to the uniform measure on  $SU_2$ , which is by definition the unique probability measure which is invariant under translations by the various group elements  $g \in SU_2$ . We will be back to this later, with full details.

As another comment, philosophical this time, the above parametrization of  $SU_2$  is something very nice, because the parameters (x, y, z, t) range now over the sphere of space-time. Thus, we are probably doing some physics here.

Regarding now the group  $U_2$ , we have here a similar result, as follows:

THEOREM 10.9. We have the following formula,

$$U_2 = \left\{ (p+iq) \begin{pmatrix} x+iy & z+it \\ -z+it & x-iy \end{pmatrix} \mid x^2+y^2+z^2+t^2 = 1, \ p^2+q^2 = 1 \right\}$$

which makes  $U_2$  be a quotient compact space, as follows,

$$S^3_{\mathbb{R}} \times S^1_{\mathbb{R}} \to U_2$$

but with this parametrization being no longer bijective.

**PROOF.** We recall from Theorem 10.7 above that we have:

$$U_2 = \left\{ d \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix} \mid |a|^2 + |b|^2 = 1, \ |d| = 1 \right\}$$

Now let us write our parameters  $a, b \in \mathbb{C}$ , which belong to the complex unit sphere  $S^1_{\mathbb{C}} \subset \mathbb{C}^2$ , and  $d \in \mathbb{T}$ , in terms of their real and imaginary parts, as follows:

$$a = x + iy$$
$$b = z + it$$
$$d = p + iq$$

In terms of these new parameters  $x, y, z, t, p, q \in \mathbb{R}$ , our formula for a generic matrix  $U \in SU_2$ , that we established before, reads:

$$U = (p + iq) \begin{pmatrix} x + iy & z + it \\ -z + it & x - iy \end{pmatrix}$$

As for the condition to be satisfied by the parameters  $x, y, z, t, p, q \in \mathbb{R}$ , this comes the conditions  $|a|^2 + |b|^2 = 1$  and |d| = 1 to be satisfied by  $a, b, d \in \mathbb{C}$ , which read:

$$x^2 + y^2 + z^2 + t^2 = 1$$
$$p^2 + q^2 = 1$$

Thus, we are led to the conclusion in the statement. Regarding now the last assertion, recall that the unit spheres  $S^3_{\mathbb{R}} \subset \mathbb{R}^4$  and  $S^1_{\mathbb{R}} \subset \mathbb{R}^2$  are given by:

$$S^{3}_{\mathbb{R}} = \left\{ (x, y, z, t) \mid x^{2} + y^{2} + z^{2} + t^{2} = 1 \right\}$$
$$S^{1}_{\mathbb{R}} = \left\{ (p, q) \mid p^{2} + q^{2} = 1 \right\}$$

Thus, we have quotient map of compact spaces, as follows:

$$S^{3}_{\mathbb{R}} \times S^{1}_{\mathbb{R}} \to U_{2}$$
$$((x, y, z, t), (p, q)) \to (p + iq) \begin{pmatrix} x + iy & z + it \\ -z + it & x - iy \end{pmatrix}$$

However, the parametrization is no longer bijective, because when we globally switch signs, the element ((-x, -y, -z, -t), (-p, -q)) produces the same element of  $U_2$ .

Here is now another reformulation of our main result so far, regarding  $SU_2$ , obtained by further building on the parametrization from Theorem 10.8:

THEOREM 10.10. We have the following formula,

$$SU_{2} = \left\{ xc_{1} + yc_{2} + zc_{3} + tc_{4} \mid x^{2} + y^{2} + z^{2} + t^{2} = 1 \right\}$$

where  $c_1, c_2, c_3, c_4$  are the Pauli matrices, given by:

$$c_1 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} , \quad c_2 = \begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix}$$
$$c_3 = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} , \quad c_4 = \begin{pmatrix} 0 & i \\ i & 0 \end{pmatrix}$$

**PROOF.** We recall from Theorem 10.8 above that the group  $SU_2$  can be parametrized by the real sphere  $S^3_{\mathbb{R}} \subset \mathbb{R}^4$ , in the following way:

$$SU_{2} = \left\{ \begin{pmatrix} x + iy & z + it \\ -z + it & x - iy \end{pmatrix} \mid x^{2} + y^{2} + z^{2} + t^{2} = 1 \right\}$$

But this gives the formula in the statement, with the Pauli matrices  $c_1, c_2, c_3, c_4$  being the coefficients of x, y, z, t, in this parametrization.

The above result is often the most convenient one, when dealing with  $SU_2$ .

This is because the Pauli matrices have a number of remarkable properties, which are very useful when doing computations. These properties can be summarized as follows:

THEOREM 10.11. The Pauli matrices multiply according to the following formulae,

 $c_2^2 = c_3^2 = c_4^2 = -1$   $c_2c_3 = -c_3c_2 = c_4$   $c_3c_4 = -c_4c_3 = c_2$  $c_4c_2 = -c_2c_4 = c_3$ 

they conjugate according to the following rules,

$$c_1^* = c_1, \ c_2^* = -c_2, \ c_3^* = -c_3, \ c_4^* = -c_4$$

and they form an orthonormal basis of  $M_2(\mathbb{C})$ , with respect to the scalar product

 $\langle a, b \rangle = tr(ab^*)$ 

with  $tr: M_2(\mathbb{C}) \to \mathbb{C}$  being the normalized trace of  $2 \times 2$  matrices, tr = Tr/2.

PROOF. The first two assertions, regarding the multiplication and conjugation rules for the Pauli matrices, follow from some elementary computations. As for the last assertion, this follows by using these rules. Indeed, the fact that the Pauli matrices are pairwise orthogonal follows from computations of the following type, for  $i \neq j$ :

$$< c_i, c_j >= tr(c_i c_j^*) = tr(\pm c_i c_j) = tr(\pm c_k) = 0$$

As for the fact that the Pauli matrices have norm 1, this follows from:

$$< c_i, c_i > = tr(c_i c_i^*) = tr(\pm c_i^2) = tr(c_1) = 1$$

Thus, we are led to the conclusion in the statement.

Regarding now the basic unitary groups in 3 or more dimensions, the situation here becomes fairly complicated. It is possible however to explicitly compute the rotation groups  $SO_3$  and  $O_3$ , and explaining this result, due to Euler-Rodrigues, which is something non-trivial and very useful, for all sorts of practical purposes, will be our next goal.

The proof of the Euler-Rodrigues formula is something quite tricky. Let us start with the following construction, whose usefulness will become clear later on:

**PROPOSITION 10.12.** The adjoint action  $SU_2 \curvearrowright M_2(\mathbb{C})$ , given by

$$T_U(M) = UMU^*$$

leaves invariant the following real vector subspace of  $M_2(\mathbb{C})$ ,

$$\mathbb{R}^4 = span(c_1, c_2, c_3, c_4)$$

and so we obtain in this way a group morphism as follows:

$$SU_2 \to GL_4(\mathbb{R})$$

Moreover, we obtain in fact in this way a group morphism  $SU_2 \rightarrow O_4$ .

#### 10B. EULER-RODRIGUES

PROOF. Everything here is clear from the multiplication formulae for the Pauli matrices, from Theorem 10.11 above. In fact, all this will come as well as a consequence of the following result, where the morphism  $SU_2 \rightarrow O_4$  is computed explicitly.

The point now is that when computing the morphism constructed in Proposition 10.12, we are led to something quite interesting, namely the group  $SO_3$ , as follows:

PROPOSITION 10.13. With respect to the standard basis  $c_1, c_2, c_3, c_4$  of the vector space  $\mathbb{R}^4 = span(c_1, c_2, c_3, c_4)$ , the morphism  $T: SU_2 \to GL_4(\mathbb{R})$  is given by:

$$T_U = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & x^2 + y^2 - z^2 - t^2 & 2(yz - xt) & 2(xz + yt) \\ 0 & 2(xt + yz) & x^2 + z^2 - y^2 - t^2 & 2(zt - xy) \\ 0 & 2(yt - xz) & 2(xy + zt) & x^2 + t^2 - y^2 - z^2 \end{pmatrix}$$

Thus, when looking at T as a group morphism  $SU_2 \rightarrow O_4$ , what we have in fact is a group morphism  $SU_2 \rightarrow O_3$ , and even  $SU_2 \rightarrow SO_3$ .

**PROOF.** With notations from Proposition 10.12 and its proof, let us first look at the action  $L: SU_2 \sim \mathbb{R}^4$  by left multiplication, which is by definition given by:

$$L_U(M) = UM$$

In order to compute the matrix of this action, let us write, as usual:

$$U = xc_1 + yc_2 + zc_3 + tc_4$$
$$M = ac_1 + bc_2 + cc_3 + dc_4$$

By using the multiplication formulae in Theorem 10.11 above, we obtain:

$$UM = (xc_1 + yc_2 + zc_3 + tc_4)(ac_1 + bc_2 + cc_3 + dc_4)$$
  
=  $(xa - yb - zc - td)c_1$   
+  $(xb + ya + zd - tc)c_2$   
+  $(xc - yd + za + tb)c_3$   
+  $(xd + yc - zb + ta)c_4$ 

We conclude that the matrix of the left action considered above is:

$$L_U = \begin{pmatrix} x & -y & -z & -t \\ y & x & -t & z \\ z & t & x & -y \\ t & -z & y & x \end{pmatrix}$$

Similarly, let us look now at the action  $R : SU_2 \curvearrowright \mathbb{R}^4$  by right multiplication, which is by definition given by the following formula:

$$R_U(M) = MU^*$$

In order to compute the matrix of this action, let us write, as before:

$$U = xc_1 + yc_2 + zc_3 + tc_4$$
$$M = ac_1 + bc_2 + cc_3 + dc_4$$

By using the multiplication formulae in Theorem 10.11 above, we obtain:

$$MU^* = (ac_1 + bc_2 + cc_3 + dc_4)(xc_1 - yc_2 - zc_3 - tc_4)$$
  
=  $(ax + by + cz + dt)c_1$   
+  $(-ay + bx - ct + dz)c_2$   
+  $(-az + bt + cx - dy)c_3$   
+  $(-at - bz + cy + dx)c_4$ 

We conclude that the matrix of the right action considered above is:

$$R_U = \begin{pmatrix} x & y & z & t \\ -y & x & -t & z \\ -z & t & x & -y \\ -t & -z & y & x \end{pmatrix}$$

Now by composing, the matrix of the adjoint matrix in the statement is:

$$\begin{aligned} T_U &= R_U L_U \\ &= \begin{pmatrix} x & y & z & t \\ -y & x & -t & z \\ -z & t & x & -y \\ -t & -z & y & x \end{pmatrix} \begin{pmatrix} x & -y & -z & -t \\ y & x & -t & z \\ z & t & x & -y \\ t & -z & y & x \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & x^2 + y^2 - z^2 - t^2 & 2(yz - xt) & 2(xz + yt) \\ 0 & 2(xt + yz) & x^2 + z^2 - y^2 - t^2 & 2(zt - xy) \\ 0 & 2(yt - xz) & 2(xy + zt) & x^2 + t^2 - y^2 - z^2 \end{pmatrix} \end{aligned}$$

Thus, we have indeed the formula in the statement. As for the remaining assertions, these are all clear either from this formula, or from standard abstract considerations.  $\Box$ 

We can now formulate the Euler-Rodrigues result, as follows:

THEOREM 10.14. We have a double cover map, obtained via the adjoint representation,

$$SU_2 \rightarrow SO_3$$

and this map produces the Euler-Rodrigues formula

$$U = \begin{pmatrix} x^2 + y^2 - z^2 - t^2 & 2(yz - xt) & 2(xz + yt) \\ 2(xt + yz) & x^2 + z^2 - y^2 - t^2 & 2(zt - xy) \\ 2(yt - xz) & 2(xy + zt) & x^2 + t^2 - y^2 - z^2 \end{pmatrix}$$

for the generic elements of  $SO_3$ .

## 10B. EULER-RODRIGUES

237

PROOF. This follows indeed from the formula in Proposition 10.13, with the fact that the morphism  $SU_2 \rightarrow SO_3$  constructed there is indeed surjective coming for instance from the fact that any rotation  $U \in SO_3$  has an axis, and so, after some computations, can be written in the form in the statement, and finally with the fact that we have indeed a double cover map being something elementary, obtained by computing the kernel.

We will see later on a more conceptual explanation for the above formula, or rather some more details in what regards the proof, and the main ideas that are used there, when doing systematic representation theory, in chapter 13 below.

Regarding now  $O_3$ , the extension from  $SO_3$  is very simple, as follows:

THEOREM 10.15. We have the Euler-Rodrigues formula

$$U = \pm \begin{pmatrix} x^2 + y^2 - z^2 - t^2 & 2(yz - xt) & 2(xz + yt) \\ 2(xt + yz) & x^2 + z^2 - y^2 - t^2 & 2(zt - xy) \\ 2(yt - xz) & 2(xy + zt) & x^2 + t^2 - y^2 - z^2 \end{pmatrix}$$

for the generic elements of  $O_3$ .

PROOF. This follows from Theorem 10.14, because the determinant of a matrix  $U \in O_3$  must satisfy det  $U = \pm 1$ , and in the case det U = -1, we have:

$$det(-U) = (-1)^3 det U$$
$$= -det U$$
$$= 1$$

Thus, assuming det U = -1, we can therefore rescale U into an element  $-U \in SO_3$ , and this leads to the conclusion in the statement.

With all these small N examples worked out, let us discuss now the general theory, at arbitrary values of  $N \in \mathbb{N}$ . In the real case, we have the following result:

**PROPOSITION** 10.16. We have a decomposition as follows, with  $SO_N^{-1}$  consisting by definition of the orthogonal matrices having determinant -1:

$$O_N = SO_N \cup SO_N^{-1}$$

When N is odd the set  $SO_N^{-1}$  is simply given by:

$$SO_N^{-1} = -SO_N$$

When N is even the situation is more complicated, and requires complex numbers.

PROOF. The first assertion is clear from definitions, because the determinant of an orthogonal matrix must be  $\pm 1$ . The second assertion is clear too, and we have seen this already at N = 3, in the proof of Theorem 10.15 above. As for the last assertion, regarding the case where N is even, here we can use the same trick as in the N odd case, but by dividing by *i* instead of -1, which leads us into  $U_N$ . We will be back to this.

In the complex case now, the result is simpler, as follows:

**PROPOSITION** 10.17. We have a decomposition as follows, with  $SU_N^d$  consisting by definition of the unitary matrices having determinant  $d \in \mathbb{T}$ :

$$O_N = \bigcup_{d \in \mathbb{T}} SU_N^d$$

Moreover, the components are  $SU_N^d = f \cdot SU_N$ , where  $f \in \mathbb{T}$  is such that  $f^N = d$ .

**PROOF.** This is clear from definitions, and from the fact that the determinant of a unitary matrix belongs to  $\mathbb{T}$ , by extracting a suitable square root of the determinant.  $\Box$ 

It is possible to use Proposition 10.17 in order to say more about what happens in the real case, in the context of Proposition 10.16, but we will not get into this.

We will basically stop here with our study of  $O_N, U_N$ , and of their versions  $SO_N, SU_N$ . As a last result on the subject, however, let us record:

THEOREM 10.18. We have subgroups of  $O_N, U_N$  constructed via the condition

 $(\det U)^d = 1$ 

with  $d \in \mathbb{N} \cup \{\infty\}$ , which generalize both  $O_N, U_N$  and  $SO_N, SU_N$ .

**PROOF.** This is indeed from definitions, and from the multiplicativity property of the determinant. We will be back to these groups, which are quite specialized, later on.  $\Box$ 

## 10c. Symplectic groups

At a more specialized level as well, but not very far from basic, we have the groups  $B_N$  and  $C_N$ , consisting of the orthogonal and unitary bistochastic matrices.

Let us start with a study of the bistochastic matrices, in the general, non-unitary case. These matrices are defined as follows:

DEFINITION 10.19. A square matrix  $M \in M_N(\mathbb{C})$  is called bistochastic if each row and each column sum up to the same number:

If this happens only for the rows, or only for the columns, the matrix is called rowstochastic, respectively column-stochastic.

As the name indicates, these matrices are useful in statistics, with the case of the matrices having entries in [0, 1], which sum up to  $\lambda = 1$ , being the important one.

As a basic example of a bistochastic matrix, we have of course the flat matrix  $\mathbb{I}_N$ . In fact, the various above notions of stochasticity are closely related to  $\mathbb{I}_N$ , or rather to the all-one vector  $\xi$  that the matrix  $\mathbb{I}_N/N$  projects on, in the following way:

PROPOSITION 10.20. Let  $M \in M_N(\mathbb{C})$  be a square matrix.

- (1) M is row stochastic, with sums  $\lambda$ , when  $M\xi = \lambda\xi$ .
- (2) M is column stochastic, with sums  $\lambda$ , when  $M^t \xi = \lambda \xi$ .
- (3) M is bistochastic, with sums  $\lambda$ , when  $M\xi = M^t\xi = \lambda\xi$ .

PROOF. All these assertions are clear from definitions, because when multiplying a matrix by  $\xi$ , we obtain the vector formed by the row sums.

As an observation here, we can reformulate if we want the above statement in a purely matrix-theoretic form, by using the flat matrix  $\mathbb{I}_N$ , as follows:

**PROPOSITION 10.21.** Let  $M \in M_N(\mathbb{C})$  be a square matrix.

- (1) *M* is row stochastic, with sums  $\lambda$ , when  $M\mathbb{I}_N = \lambda \mathbb{I}_N$ .
- (2) M is column stochastic, with sums  $\lambda$ , when  $\mathbb{I}_N M = \lambda \mathbb{I}_N$ .
- (3) *M* is bistochastic, with sums  $\lambda$ , when  $M\mathbb{I}_N = \mathbb{I}_N M = \lambda \mathbb{I}_N$ .

**PROOF.** This follows from Proposition 10.20, and from the fact that both the rows and the columns of the flat matrix  $\mathbb{I}_N$  are copies of the all-one vector  $\xi$ .

In what follows we will be mainly interested in the unitary bistochastic matrices, which are quite interesting objects. These do not exactly cover the flat matrix  $\mathbb{I}_N$ , but cover instead the following related matrix, which appears in many linear algebra questions:

$$K_N = \frac{1}{N} \begin{pmatrix} 2-N & 2\\ & \ddots & \\ 2 & 2-N \end{pmatrix}$$

As a first result, regarding such matrices, we have the following statement:

THEOREM 10.22. For a unitary matrix  $U \in U_N$ , the following conditions are equivalent:

- (1) H is bistochastic, with sums  $\lambda$ .
- (2) *H* is row stochastic, with sums  $\lambda$ , and  $|\lambda| = 1$ .
- (3) *H* is column stochastic, with sums  $\lambda$ , and  $|\lambda| = 1$ .

**PROOF.** By using a symmetry argument we just need to prove (1)  $\iff$  (2), and both the implications are elementary, as follows:

(1)  $\implies$  (2) If we denote by  $U_1, \ldots, U_N \in \mathbb{C}^N$  the rows of U, we have indeed:

1

$$= \langle U_1, U_1 \rangle$$

$$= \sum_i \langle U_1, U_i \rangle$$

$$= \sum_i \sum_j U_{1j} \overline{U}_{ij}$$

$$= \sum_j U_{1j} \sum_i \overline{U}_{ij}$$

$$= \sum_j U_{1j} \cdot \overline{\lambda}$$

$$= \lambda \cdot \overline{\lambda}$$

$$= |\lambda|^2$$

(2)  $\implies$  (1) Consider the all-one vector  $\xi = (1)_i \in \mathbb{C}^N$ . The fact that U is rowstochastic with sums  $\lambda$  reads:

$$\sum_{j} U_{ij} = \lambda, \forall i \quad \Longleftrightarrow \quad \sum_{j} U_{ij} \xi_j = \lambda \xi_i, \forall i$$
$$\iff \quad U\xi = \lambda \xi$$

Also, the fact that U is column-stochastic with sums  $\lambda$  reads:

$$\sum_{i} U_{ij} = \lambda, \forall j \iff \sum_{j} U_{ij}\xi_i = \lambda\xi_j, \forall j$$
$$\iff U^t\xi = \lambda\xi$$

We must prove that the first condition implies the second one, provided that the row sum  $\lambda$  satisfies  $|\lambda| = 1$ . But this follows from the following computation:

$$U\xi = \lambda \xi \implies U^*U\xi = \lambda U^*\xi$$
$$\implies \xi = \lambda U^*\xi$$
$$\implies \xi = \bar{\lambda} U^t\xi$$
$$\implies U^t\xi = \lambda\xi$$

Thus, we have proved both the implications, and we are done.

The unitary bistochastic matrices are stable under a number of operations, and in particular under taking products. Thus, these matrices form a group.

We have in fact the following result, making the link with the Fourier matrix, and with various discrete Fourier analysis questions:

THEOREM 10.23. The real and complex bistochastic groups, which are the sets

$$B_N \subset O_N$$
$$C_N \subset U_N$$

consisting of matrices which are bistochastic, are isomorphic to  $O_{N-1}$ ,  $U_{N-1}$ .

PROOF. Let us pick a unitary matrix  $F \in U_N$  satisfying the following condition, where  $e_0, \ldots, e_{N-1}$  is the standard basis of  $\mathbb{C}^N$ , and where  $\xi$  is the all-one vector:

$$Fe_0 = \frac{1}{\sqrt{N}}\xi$$

Observe that such matrices  $F \in U_N$  exist indeed, the basic example being the normalized Fourier matrix, which with  $w = e^{2\pi i/N}$  is as follows:

$$F = \frac{1}{\sqrt{N}} (w^{ij})$$

We have then, by using the above property of F:

$$u\xi = \xi \iff uFe_0 = Fe_0$$
$$\iff F^*uFe_0 = e_0$$
$$\iff F^*uF = diag(1, w)$$

Thus we have isomorphisms as in the statement, given by:

$$w_{ij} \to (F^* u F)_{ij}$$

But this gives both the assertions.

Finally, as yet another basic example, we have the symplectic group  $Sp_N$ , consisting of the symplectic matrices. This is something more technical. Let us begin with:

DEFINITION 10.24. The "super-space"  $\overline{\mathbb{C}}^N$  is the usual space  $\mathbb{C}^N$ , with its standard basis  $\{e_1, \ldots, e_N\}$ , with a chosen sign

 $\varepsilon = \pm 1$ 

and a chosen involution on the set of indices:

 $i \to \overline{i}$ 

The "super-identity" matrix is  $J_{ij} = \delta_{i\bar{j}}$  for  $i \leq j$  and  $J_{ij} = \varepsilon \delta_{i\bar{j}}$  for  $i \geq j$ .

Up to a permutation of the indices, we have a decomposition N = 2p + q, such that the involution is, in standard permutation notation:

$$(12) \dots (2p-1, 2p)(2p+1) \dots (q)$$

Thus, up to a base change, the super-identity is as follows, where N = 2p + q and  $\varepsilon = \pm 1$ , with the  $1_q$  block at right disappearing if  $\varepsilon = -1$ :

$$J = \begin{pmatrix} 0 & 1 & & & \\ \varepsilon 1 & 0_{(0)} & & & \\ & & \ddots & & \\ & & 0 & 1 & & \\ & & & \varepsilon 1 & 0_{(p)} & & \\ & & & & & 1_{(1)} & \\ & & & & & \ddots & \\ & & & & & & 1_{(q)} \end{pmatrix}$$

In the case  $\varepsilon = 1$ , the super-identity is the following matrix:

$$J_{+}(p,q) = \begin{pmatrix} 0 & 1 & & & & \\ 1 & 0_{(1)} & & & & \\ & & \ddots & & & \\ & & 0 & 1 & & \\ & & & 1 & 0_{(p)} & & \\ & & & & & 1_{(1)} & \\ & & & & & & \ddots & \\ & & & & & & & 1_{(q)} \end{pmatrix}$$

In the case  $\varepsilon = -1$  now, the diagonal terms vanish, and the super-identity is:

$$J_{-}(p,0) = \begin{pmatrix} 0 & 1 & & & \\ -1 & 0_{(1)} & & & \\ & & \ddots & & \\ & & & 0 & 1 \\ & & & -1 & 0_{(p)} \end{pmatrix}$$

With the above notions in hand, we have the following result:

THEOREM 10.25. The super-orthogonal group, which is by definition

$$\bar{O}_N = \left\{ U \in U_N \middle| U = J\bar{U}J^{-1} \right\}$$

with J being the super-identity matrix, is as follows:

(1) At  $\varepsilon = 1$  we have  $\bar{O}_N = O_N$ . (2) At  $\varepsilon = -1$  we have  $\bar{O}_N = Sp_N$ .

**PROOF.** These results are both elementary, as follows:

(1) At  $\varepsilon = -1$  this follows from definitions.

(2) At  $\varepsilon = 1$  now, consider the root of unity  $\rho = e^{\pi i/4}$ , and let:

$$\Gamma = \frac{1}{\sqrt{2}} \begin{pmatrix} \rho & \rho^7 \\ \rho^3 & \rho^5 \end{pmatrix}$$

Then  $\Gamma$  is unitary and we have:

$$\Gamma \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \Gamma^t = 1$$

Thus the following matrix is unitary as well, and satisfies  $CJC^t = 1$ :

$$C = \begin{pmatrix} \Gamma^{(1)} & & & \\ & \ddots & & \\ & & \Gamma^{(p)} & \\ & & & 1_q \end{pmatrix}$$

Thus in terms of  $V = CUC^*$  the relations  $U = J\overline{U}J^{-1} =$  unitary simply read:

$$V = \overline{V} = \text{unitary}$$

Thus we obtain an isomorphism  $\overline{O}_N = O_N$  as in the statement.

Regarding now  $Sp_N$ , we have the following result:

THEOREM 10.26. The symplectic group  $Sp_N \subset U_N$ , which is by definition

$$Sp_N = \left\{ U \in U_N \middle| U = J\bar{U}J^{-1} \right\}$$

consists of the  $SU_2$  patterned matrices,

$$U = \begin{pmatrix} a & b & \dots \\ -\bar{b} & \bar{a} & \\ \vdots & \ddots \end{pmatrix}$$

which are unitary,  $U \in U_N$ . In particular, we have  $Sp_2 = SU_2$ .

PROOF. This follows indeed from definitions, because the condition  $U = J\bar{U}J^{-1}$  corresponds precisely to the fact that U must be a  $SU_2$ -patterned matrix.

We will be back later to the symplectic groups, with more results about them.

## 10d. Reflection groups

We have now a quite good understanding of the main examples of continuous groups of unitary matrices  $G \subset U_N$ , especially at small values of N. Let us go back now to the finite groups from the previous chapter, and make a link with the material there.

Let us first recall from there that we have:

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DEFINITION 10.27. The complex reflection group  $H_N^s \subset U_N$ , depending on parameters

$$N \in \mathbb{N}$$
 ,  $s \in \mathbb{N} \cup \{\infty\}$ 

are the groups of permutation-type matrices with s-th roots of unity as entries,

$$H_N^s = M_N(\mathbb{Z}_s \cup \{0\}) \cap U_N$$

with the convention  $\mathbb{Z}_{\infty} = \mathbb{T}$ , at  $s = \infty$ .

Observe that at s = 1, 2 we obtain the symmetric and hyperoctahedral groups:

$$H_N^1 = S_N$$
$$H_N^2 = H_N$$

Another important particular case is  $s = \infty$ , where we obtain a group which is actually not finite, denoted as follows:

$$H_N^\infty = K_N$$

This group is new, and its basic theory is as follows:

THEOREM 10.28. The full complex reflection group, denoted

$$K_N \subset U_N$$

and formed of the permutation-type matrices with numbers in  $\mathbb{T}$  as nonzero entries,

 $K_N = M_N(\mathbb{T} \cup \{0\}) \cap U_N$ 

has a wreath product decomposition, as follows,

$$K_N = \mathbb{T} \wr S_N$$

with  $S_N$  acting on  $\mathbb{T}^N$  in the standard way, by permuting the factors.

**PROOF.** This is something that we already know from the above, at the  $s = \infty$  particular case of the results established for the complex reflection groups  $H_N^s$ .

By using the above full complex reflection group  $K_N$ , we can talk in fact about the reflection subgroup of any compact group  $G \subset U_N$ , as follows:

DEFINITION 10.29. Given  $G \subset U_N$ , we define its reflection subgroup to be

$$K = G \cap K_N^+$$

with the intersection taken inside  $U_N$ .

There are many basic examples here, that can be worked out.

The central object in all the above is the symmetric group  $S_N$ , and in connection with the considerations from the first part of this chapter, where the central notion was the determinant of the orthogonal or unitary matrices, we have the following result:

THEOREM 10.30. For the symmetric group, regarded as group of permutation matrices,

 $S_N \subset O_N$ 

the determinant is the signature. The subgroup  $A_N \subset S_N$  given by

$$A_N = S_N \cap SO_N$$

and called alternating group, consists of the even permutations.

**PROOF.** In this statement the first assertion is clear from the definition of the determinant, and of the permutation matrices, and all the rest is standard.  $\Box$ 

We can further complicate the above construction, in the following way:

THEOREM 10.31. We have subgroups of the basic complex reflection groups,

$$H_N^{sd} \subset H_N^s$$

constructed via the following condition, with  $d \in \mathbb{N} \cup \{\infty\}$ ,

 $(\det U)^d = 1$ 

which generalize all the complex reflection groups that we have so far.

**PROOF.** We have several things to be proved, the idea being as follows:

(1) The first assertion is clear from definitions, and from the multiplicativity property of the determinant.

(2) As for the second assertion, this is rather a remark, coming from the fact that the alternating group  $A_N$ , which is the ony finite group so far not fitting into the series  $\{H_N^s\}$ , is indeed of this type, obtained from  $H_N^1 = S_N$  by using d = 1.

We should mention here that, by a well-known and deep result in group theory, the complex reflection groups consist of the series  $\{H_N^{sd}\}$  constructed above, and of a number of exceptional groups, which can be fully classified.

To be more precise, we have the following result:

THEOREM 10.32. The irreducible complex reflection groups are

$$H_N^{sd} = \left\{ U \in H_N^s \middle| (\det U)^d = 1 \right\}$$

along with 34 exceptional examples.

**PROOF.** This is something quite advanced, and we refer here to the paper of Shephard and Todd [67], and to the subsequent literature on the subject.  $\Box$ 

Let us discuss now a number of more specialized questions. Consider the following diagram, formed by the main rotation and reflection groups:



We know from the above that this is an intersection and generation diagram. Now assume that we have an intermediate compact group, as follows:

$$H_N \subset G_N \subset U_N$$

The point is that we can think of our group  $G_N$  as living inside the above square, and so project it on the edges, as to obtain information about it. Let us start with:

DEFINITION 10.33. Associated to any closed subgroup  $G_N \subset U_N$  are its discrete and real versions, given by

$$G_N^d = G_N \cap K_N$$
$$G_N^r = G_N \cap O_N$$

as well as its smooth and unitary versions, given by

$$G_N^s = \langle G_N, O_N \rangle$$
$$G_N^u = \langle G_N, K_N \rangle$$

where <,> is the topological generation operation.

Assuming now that we have an intermediate compact group  $H_N \subset G_N \subset U_N$ , as above, we are led in this way to the following notion:

DEFINITION 10.34. A compact group  $H_N \subset G_N \subset U_N$  is called oriented if



is an intersection and generation diagram.

## 10E. EXERCISES

Most of our examples of compact groups  $G_N \subset U_N$ , finite or continuous, usually come in series, depending uniformly on  $N \in \mathbb{N}$ . This suggests the following definition:

DEFINITION 10.35. A family of compact groups  $G = (G_N)$ , with

 $S_N \subset G_N \subset U_N$ 

is called uniform if it satisfies the following equivalent conditions:

(1) We have  $G_{N-1} = G_N \cap U_{N-1}$ , via the embedding  $U_{N-1} \subset U_N$  given by:

$$u \to \begin{pmatrix} u & 0 \\ 0 & 1 \end{pmatrix}$$

(2) We have  $G_{N-1} = G_N \cap U_{N-1}$ , via the N possible diagonal embeddings

$$U_{N-1} \subset U_N$$

obtained as above, by inserting 1 somewhere on the diagonal.

Here the fact that the above two conditions are indeed equivalent comes from our assumption  $S_N \subset G_N$ , which makes everything to be permutation invariant.

We have many examples of uniform oriented groups, and normally some classification results are possible. To be more precise, the uniform oriented groups should be the obvious examples of such groups, that we already know from the above.

We will be back to this later on, in chapters 13-16 below, under a number of supplementary assumptions on the finite compact groups that we consider, which will allow us to derive, in a quite elementary way, a number of classification results.

# 10e. Exercises

There has been a lot of theory in this chapter, and this is just the tip of the iceberg, on what can be said about the continuous groups. As a first exercise, we have:

EXERCISE 10.36. Work out all the details of the Euler-Rodrigues formula, by using Pauli matrices and the quotient map  $SU_2 \rightarrow SO_3$ .

This is something that was already discussed in the above, and the problem now is that of filling out all the details.

Along the same lines, we have the following exercise, which is complementary to the above one, and very instructive as well:

EXERCISE 10.37. Work out all the details of the Euler-Rodrigues formula, by using this time the fact that any rotation in  $\mathbb{R}^3$  has a rotation axis.

Here the problem, once the rotation axis found, is that of drawing the picture, identifying the relevant angles, and then doing the mathematics in terms of these angles.

In relation now with various versions of  $SO_N, SU_N$ , we first have:

EXERCISE 10.38. Work out the theory of the subgroups of  $O_N, U_N$  constructed via

 $(\det U)^d = 1$ 

with  $d \in \mathbb{N} \cup \{\infty\}$ , which generalize both  $O_N, U_N$  and  $SO_N, SU_N$ .

There are many things that can be done here, and the more, the better.

In relation now with the bistochastic groups and matrices, we have:

EXERCISE 10.39. Verify that all basic examples of unitary matrices can be put in bistochastic form, and then find and write a brief account of the Idel-Wolf theorem, based on symplectic geometry, stating that this is true in general.

This is of course something quite difficult, but everything here is very instructive.

Along the same lines, but in relation now with symplectic groups, we have:

EXERCISE 10.40. Look up the literature, and find the relevance of the symplectic groups, and of symplectic geometry in general, to questions in mechanics, and then write down a brief account of that.

As before with the previous exercise, many things that can be learned and done here, and the more, the better.

Regarding now the reflection groups, we have:

EXERCISE 10.41. Find and then write down a brief account of the Shephard-Todd theorem, stating that the irreducible complex reflection groups are

$$H_N^{sd} = \left\{ U \in H_N^s \middle| (\det U)^d = 1 \right\}$$

along with a number of exceptional examples, more precisely 34 of them.

As before with the previous exercises, the more, the better.

Finally, we have the following exercise, the most important of them all:

EXERCISE 10.42. Learn some basic Lie algebra theory, and write down a brief account of that, in relation with the groups studied in the present chapter.

This is a good and instructive exercise, leading you into many interesting things, which must be learned too, in one form or another. By the way do not forget to come back to this book afterwards, because there are many things to be learned from here as well, starting with the results from the next chapter, about the symmetric group  $S_N$ .

# CHAPTER 11

# Symmetric groups

# 11a. Character laws

We develop in what follows some general theory for the compact subgroups  $G \subset U_N$ , usually taken finite, with our main example being the symmetric group  $S_N \subset O_N$ . Let us start with a definition that we have already met in chapter 9 above, namely:

**DEFINITION 11.1.** A representation of a finite group G is a group morphism

 $u: G \to U_N$ 

into a unitary group. The character of such a representation is the function

 $\chi: G \to \mathbb{C}$  $g \to Tr(u_g)$ 

where Tr is the usual, unnormalized trace of the  $N \times N$  matrices.

As explained in chapter 9, the simplest case of all this, namely N = 1, is of particular interest. Here the representations coincide with their characters, and are by definition the group morphisms as follows, called characters of the group:

 $\chi:G\to \mathbb{T}$ 

These characters from an abelian group  $\widehat{G}$ , and when G itself is abelian, the correspondence  $G \to \widehat{G}$  is a duality, in the sense that it maps  $\widehat{G} \to G$  as well. Moreover, a more detailed study shows that we have in fact an isomorphism  $G \simeq \widehat{G}$ , with this being something quite subtle, related at the same time to the structure theorem for the finite abelian groups,  $G \simeq \mathbb{Z}_{N_1} \times \ldots \times \mathbb{Z}_{N_k}$ , and to the Fourier transforms over such groups.

In what follows we will be interested in the general case,  $N \in \mathbb{N}$ . It is technically convenient to assume that the representation  $\pi : G \to U_N$  is faithful, by replacing if necessary G with its image. Thus, we are led to the following definition:

DEFINITION 11.2. The main character of a compact group  $G \subset U_N$  is the map

$$\chi: G \to \mathbb{C}$$

$$g \to Tr(g)$$

which associates to the group elements, viewed as unitary matrices, their trace.

## 11. SYMMETRIC GROUPS

We will see in a moment some motivations for the study of these characters. From a naive viewpoint, which is ours at the present stage, we want to do some linear algebra with our group elements  $g \in U_N$ , and we have several choices here, as follows:

- (1) A first idea would be to look at the determinant, det  $g \in \mathbb{T}$ . However, this is usually not a very interesting quantity, for instance because  $g \in O_N$  implies det  $g = \pm 1$ . Also, for groups like  $SO_N, SU_N$ , this determinant is by definition 1.
- (2) A second idea would be to try to compute eigenvalues and eigenvectors for the group elements  $g \in G$ , and then solve diagonalization questions for these elements. However, all this is quite complicated, so this idea is not good either.
- (3) Thus, we are left with looking at the trace,  $Tr(g) \in \mathbb{C}$ . We will see soon that this is a very reasonable choice, with the mathematics being at the same time non-trivial, doable, and also interesting, for a number of reasons.

Before starting our study, let us briefly discuss as well some more complicated reasons, leading to the study of characters. The idea here is that a given finite or compact group G can have several representations  $\pi : G \to U_N$ , and these representations can be studied via their characters  $\chi_{\pi} : G \to \mathbb{C}$ , with a well-known and deep theorem basically stating that  $\pi$  can be recovered from its character  $\chi_{\pi}$ . We will be back to this later.

As a basic result regarding the characters, we have:

THEOREM 11.3. Given a compact group  $G \subset U_N$ , its main character  $\chi : G \to \mathbb{C}$  is a central function, in the sense that it satisfies the following condition:

 $\chi(gh) = \chi(hg)$ 

Equivalently,  $\chi$  is constant on the conjugacy classes of G.

**PROOF.** This is clear from the fact that the trace of matrices satisfies:

$$Tr(AB) = Tr(BA)$$

Thus, we are led to the conclusion in the statement.

As before, there is some interesting mathematics behind all this. We will prove later, when doing representation theory, that any central function  $f: G \to \mathbb{C}$  appears as a linear combination of characters  $\chi_{\pi}: G \to \mathbb{C}$  of representations  $\pi: G \to U_N$ .

In order to work out now some examples, let us get back now to our main examples of finite groups, constructed in chapter 9 above, namely:

$$\mathbb{Z}_N \subset D_N \subset S_N \subset H_N$$

We will do in what follows some character computations for these groups, which are all quite elementary, or at least not requiring very advanced theory.

## 11A. CHARACTER LAWS

Let us start with the following result, which covers  $\mathbb{Z}_N \subset D_N \subset S_N$ , or rather tells us what is to be done with these groups, in relation with their main characters:

PROPOSITION 11.4. For the symmetric group, regarded as group of permutation matrices,  $S_N \subset O_N$ , the main character counts the number of fixed points:

$$\chi(g) = \#\left\{i \in \{1, \dots, N\} \middle| \sigma(i) = i\right\}$$

The same goes for any  $G \subset S_N$ , regarded as a matrix group via  $G \subset S_N \subset O_N$ .

PROOF. This is indeed clear from definitions, because the diagonal entries of the permutation matrices correspond to the fixed points of the permutation.  $\Box$ 

Summarizing, we are left with counting fixed points. For the simplest possible group, namely the cyclic group  $\mathbb{Z}_N \subset S_N$ , the computation is as follows:

**PROPOSITION 11.5.** The character of  $\mathbb{Z}_N \subset O_N$  is given by:

$$\chi(g) = \begin{cases} 0 & \text{if } g \neq 1\\ N & \text{if } g = 1 \end{cases}$$

**PROOF.** This is clear from definitions, because the cyclic permutation matrices have 0 on the diagonal, and so 0 as trace, unless the matrix is the identity, having trace N.

Let us record as well a probabilistic version of the above result. In probabilistic terms, the result states that the corresponding distribution is a Bernoulli law:

$$law(\chi) = \left(1 - \frac{1}{N}\right)\delta_0 + \frac{1}{N}\delta_N$$

For the dihedral group now, which is the next one in our hierarchy, the computation is more interesting, and the final answer is not uniform in N, as follows:

**PROPOSITION 11.6.** For the dihedral group  $D_N \subset S_N$  we have:

$$law(\chi) = \begin{cases} \left(\frac{3}{4} - \frac{1}{2N}\right)\delta_0 + \frac{1}{4}\delta_2 + \frac{1}{2N}\delta_N & (N \ even) \\ \\ \left(\frac{1}{2} - \frac{1}{2N}\right)\delta_0 + \frac{1}{2}\delta_1 + \frac{1}{2N}\delta_N & (N \ odd) \end{cases}$$

**PROOF.** The dihedral group  $D_N$  consists indeed of:

(1) N symmetries, having each 1 fixed point when N is odd, and having 0 or 2 fixed points, distributed 50 - 50, when N is even.

(2) N rotations, each having 0 fixed points, except for the identity, which is technically a rotation too, and which has N fixed points.

Thus, we are led to the formulae in the statement.

251
Regarding now the symmetric group  $S_N$  itself, the permutations having no fixed points at all are called derangements, and the first question which appears, which is a classical question in combinatorics, is that of counting these derangements.

The result here, which is something remarkable, is as follows:

THEOREM 11.7. The probability for a random permutation  $\sigma \in S_N$  to be a derangement is given by the following formula:

$$P = 1 - \frac{1}{1!} + \frac{1}{2!} - \ldots + (-1)^{N-1} \frac{1}{(N-1)!} + (-1)^N \frac{1}{N!}$$

Thus we have the following asymptotic formula, in the  $N \to \infty$  limit

$$P \simeq \frac{1}{e}$$

where e = 2.7182... is the usual constant from analysis.

**PROOF.** This is something very classical, which is best viewed by using the inclusionexclusion principle. Consider indeed the following sets:

$$S_N^i = \left\{ \sigma \in S_N \middle| \sigma(i) = i \right\}$$

The set of permutations having no fixed points is then:

$$X_N = \left(\bigcup_i S_N^i\right)^c$$

In order to compute now the cardinality  $|X_N|$ , consider as well the following sets, depending on indices  $i_1 < \ldots < i_k$ , obtained by taking intersections:

$$S_N^{i_1\dots i_k} = S_N^{i_1}\bigcap\dots\bigcap S_N^{i_k}$$

Observe that we have the following formula:

$$S_N^{i_1\dots i_k} = \left\{ \sigma \in S_N \middle| \sigma(i_1) = i_1, \dots, \sigma(i_k) = i_k \right\}$$

The inclusion-exclusion principle tells us that we have:

$$|X_N| = |S_N| - \sum_i |S_N^i| + \sum_{i < j} |S_N^i \cap S_N^j| - \dots + (-1)^N \sum_{i_1 < \dots < i_N} |S_N^{i_1} \cup \dots \cup S_N^{i_N}|$$
  
=  $|S_N| - \sum_i |S_N^i| + \sum_{i < j} |S_N^{i_j}| - \dots + (-1)^N \sum_{i_1 < \dots < i_N} |S_N^{i_1 \dots i_N}|$ 

Thus, the probability that we are interested in is given by:

$$P = \frac{1}{N!} \left( |S_N| - \sum_i |S_N^i| + \sum_{i < j} |S_N^{ij}| - \dots + (-1)^N \sum_{i_1 < \dots < i_N} |S_N^{i_1 \dots i_N}| \right)$$
  

$$= \frac{1}{N!} \sum_{k=0}^N (-1)^k \sum_{i_1 < \dots < i_k} |S_N^{i_1 \dots i_k}|$$
  

$$= \frac{1}{N!} \sum_{k=0}^N (-1)^k \sum_{i_1 < \dots < i_k} (N-k)!$$
  

$$= \frac{1}{N!} \sum_{k=0}^N (-1)^k \binom{N}{k} (N-k)!$$
  

$$= \sum_{k=0}^N \frac{(-1)^k}{k!}$$
  

$$= 1 - \frac{1}{1!} + \frac{1}{2!} - \dots + (-1)^{N-1} \frac{1}{(N-1)!} + (-1)^N \frac{1}{N!}$$

Since on the right we have the expansion of  $\frac{1}{e}$ , we obtain:

$$P \simeq \frac{1}{e}$$

Thus, we are led to the conclusion in the statement.

The above result is something quite remarkable, and there are many versions and generalizations of it. We will discuss this gradually, in what follows.

In terms of characters, the above result reformulates as follows:

THEOREM 11.8. For the symmetric group, the probability for main character

$$\chi:S_N\to\mathbb{N}$$

to vanish is given by the following formula:

$$\mathbb{P}(\chi = 0) = 1 - \frac{1}{1!} + \frac{1}{2!} - \ldots + (-1)^{N-1} \frac{1}{(N-1)!} + (-1)^N \frac{1}{N!}$$

Thus we have the following asymptotic formula, in the  $N \to \infty$  limit

$$\mathbb{P}(\chi = 0) \simeq \frac{1}{e}$$

where e = 2.7182... is the usual constant from analysis.

PROOF. This follows indeed by combining Proposition 11.4, which tells us that  $\chi$  counts the number of fixed points, and Theorem 11.7 above.

Let us discuss now, more generally, what happens when counting permutations having exactly k fixed points. The result here, extending Theorem 11.7, is as follows:

THEOREM 11.9. The probability for a random permutation  $\sigma \in S_N$  to have exactly k fixed points, with  $k \in \mathbb{N}$ , is given by the following formula:

$$P = \frac{1}{k!} \left( 1 - \frac{1}{1!} + \frac{1}{2!} - \ldots + (-1)^{N-1} \frac{1}{(N-1)!} + (-1)^N \frac{1}{N!} \right)$$

Thus we have the following asymptotic formula, in the  $N \to \infty$  limit

$$P \simeq \frac{1}{ek!}$$

where e = 2.7182... is the usual constant from analysis.

PROOF. We already know, from Theorem 11.7, that this formula holds at k = 0. In the general case now, we have to count the permutations  $\sigma \in S_N$  having exactly k points. Since having such a permutation amounts in choosing k points among  $1, \ldots, N$ , and then permuting the N - k points left, without fixed points allowed, we have:

$$\# \left\{ \sigma \in S_N \middle| \chi(\sigma) = k \right\} = \binom{N}{k} \# \left\{ \sigma \in S_{N-k} \middle| \chi(\sigma) = 0 \right\} \\
= \frac{N!}{k!(N-k)!} \# \left\{ \sigma \in S_{N-k} \middle| \chi(\sigma) = 0 \right\} \\
= N! \times \frac{1}{k!} \times \frac{\# \left\{ \sigma \in S_{N-k} \middle| \chi(\sigma) = 0 \right\}}{(N-k)!}$$

Now by dividing everything by N!, we obtain from this the following formula:

$$\frac{\#\left\{\sigma \in S_N \middle| \chi(\sigma) = k\right\}}{N!} = \frac{1}{k!} \times \frac{\#\left\{\sigma \in S_{N-k} \middle| \chi(\sigma) = 0\right\}}{(N-k)!}$$

By using now the computation at k = 0, that we already have, from Theorem 11.7 above, it follows that with  $N \to \infty$  we have the following estimate:

$$P(\chi = k) \simeq \frac{1}{k!} \cdot P(\chi = 0)$$
$$\simeq \frac{1}{k!} \cdot \frac{1}{e}$$

Thus, we are led to the conclusion in the statement.

As before, in regards with derengements, we can reformulate what we found in terms of the main character, and we obtain in this way the following statement:

THEOREM 11.10. For the symmetric group, the distribution of the main character

$$\chi: S_N \to \mathbb{N}$$

is given by the following formula:

$$\mathbb{P}(\chi = k) = \frac{1}{k1} \left( 1 - \frac{1}{1!} + \frac{1}{2!} - \dots + (-1)^{N-1} \frac{1}{(N-1)!} + (-1)^N \frac{1}{N!} \right)$$

Thus we have the following asymptotic formula, in the  $N \to \infty$  limit,

$$\mathbb{P}(\chi = k) \simeq \frac{1}{ek!}$$

where e = 2.7182... is the usual constant from analysis.

PROOF. This follows indeed by combining Proposition 11.4, which tells us that  $\chi$  counts the number of fixed points, and Theorem 11.9 above.

# 11b. Poisson limits

In order to interpret the above results, we will need some probability theory, coming as the "discrete" counterpart of the theory developed in chapter 5. We first have:

DEFINITION 11.11. The Poisson law of parameter 1 is the following measure,

$$p_1 = \frac{1}{e} \sum_k \frac{\delta_k}{k!}$$

and the Poisson law of parameter t > 0 is the following measure,

$$p_t = e^{-t} \sum_k \frac{t^k}{k!} \,\delta_k$$

with the letter "p" standing for Poisson.

Observe that these laws have indeed mass 1, as they should, and this due to the following well-known formula, which is the foundational formula of calculus:

$$e^t = \sum_k \frac{t^k}{k!}$$

We will see in the moment why these measures appear a bit everywhere, in discrete contexts, the reasons behind this coming from the Poisson Limit Theorem (PLT).

Let us first develop some general theory. We first have:

THEOREM 11.12. We have the following formula, for any s, t > 0,

$$p_s * p_t = p_{s+t}$$

so the Poisson laws form a convolution semigroup.

PROOF. The convolution of Dirac masses is given by:

$$\delta_k * \delta_l = \delta_{k+l}$$

By using this formula and the binomial formula, we obtain:

$$p_s * p_t = e^{-s} \sum_k \frac{s^k}{k!} \delta_k * e^{-t} \sum_l \frac{t^l}{l!} \delta_l$$

$$= e^{-s-t} \sum_{kl} \frac{s^k t^l}{k! l!} \delta_{k+l}$$

$$= e^{-s-t} \sum_n \delta_n \sum_{k+l=n} \frac{s^k t^l}{k! l!}$$

$$= e^{-s-t} \sum_n \frac{\delta_n}{n!} \sum_{k+l=n} \frac{n!}{k! l!} s^k t^l$$

$$= e^{-s-t} \sum_n \frac{(s+t)^n}{n!} \delta_n$$

$$= p_{s+t}$$

Thus, we are led to the conclusion in the statement.

We have as well the following result:

THEOREM 11.13. The Poisson laws appear as exponentials

$$p_t = \sum_k \frac{t^k (\delta_1 - \delta_0)^{*k}}{k!}$$

with respect to the convolution of measures \*.

**PROOF.** By using the binomial formula, the measure at right is:

$$\mu = \sum_{k} \frac{t^{k}}{k!} \sum_{p+q=k} (-1)^{q} \frac{k!}{p!q!} \delta_{p}$$

$$= \sum_{k} t^{k} \sum_{p+q=k} (-1)^{q} \frac{\delta_{p}}{p!q!}$$

$$= \sum_{p} \frac{t^{p} \delta_{p}}{p!} \sum_{q} \frac{(-1)^{q}}{q!}$$

$$= \frac{1}{e} \sum_{p} \frac{t^{p} \delta_{p}}{p!}$$

$$= p_{t}$$

Thus, we are led to the conclusion in the statement.

The Fourier transform computation is as follows:

Theorem 11.14. The Fourier transform of  $p_t$  is given by

$$F_{p_t}(x) = \exp\left((e^{ix} - 1)t\right)$$

for any t > 0.

**PROOF.** We have by definition:

$$F_f(x) = \mathbb{E}(e^{ixf})$$

We therefore obtain the following formula:

$$F_{p_t}(x) = e^{-t} \sum_k \frac{t^k}{k!} F_{\delta_k}(x)$$
$$= e^{-t} \sum_k \frac{t^k}{k!} e^{ikx}$$
$$= e^{-t} \sum_k \frac{(e^{ix}t)^k}{k!}$$
$$= \exp(-t) \exp(e^{ix}t)$$
$$= \exp\left((e^{ix} - 1)t\right)$$

Thus, we obtain the formula in the statement.

Observe that we obtain in this way another proof for the convolution semigroup property of the Poisson laws, that we established above.

We can now establish the Poisson Limit Theorem (PLT), as follows:

THEOREM 11.15. We have the following convergence, in moments,

$$\left(\left(1-\frac{t}{n}\right)\delta_0 + \frac{t}{n}\delta_1\right)^{*n} \to p_t$$

for any t > 0.

**PROOF.** Let us denote by  $\mu_n$  the measure under the convolution sign:

$$\mu_n = \left(1 - \frac{t}{n}\right)\delta_0 + \frac{t}{n}\delta_1$$

257

We have the following computation:

$$F_{\delta_r}(x) = e^{irx} \implies F_{\mu_n}(x) = \left(1 - \frac{t}{n}\right) + \frac{t}{n}e^{ix}$$
$$\implies F_{\mu_n^{*n}}(x) = \left(\left(1 - \frac{t}{n}\right) + \frac{t}{n}e^{ix}\right)^n$$
$$\implies F_{\mu_n^{*n}}(x) = \left(1 + \frac{(e^{ix} - 1)t}{n}\right)^n$$
$$\implies F(x) = \exp\left((e^{ix} - 1)t\right)$$

Thus, we obtain the Fourier transform of  $p_t$ , as desired.

At the level of moments now, things are quite subtle, and we first have:

THEOREM 11.16. The moments of  $p_1$  are the Bell numbers,

$$M_k(p_1) = |P(k)|$$

where P(k) is the set of partitions of  $\{1, \ldots, k\}$ .

**PROOF.** The moments of  $p_1$  are given by the following formula:

$$M_k = \frac{1}{e} \sum_s \frac{s^k}{s!}$$

We have the following recurrence formula for these moments:

$$M_{k+1} = \frac{1}{e} \sum_{s} \frac{(s+1)^{k+1}}{(s+1)!}$$
$$= \frac{1}{e} \sum_{s} \frac{(s+1)^{k}}{s!}$$
$$= \frac{1}{e} \sum_{s} \frac{s^{k}}{s!} \left(1 + \frac{1}{s}\right)^{k}$$
$$= \frac{1}{e} \sum_{s} \frac{s^{k}}{s!} \sum_{r} \binom{k}{r} s^{-r}$$
$$= \sum_{r} \binom{k}{r} \cdot \frac{1}{e} \sum_{s} \frac{s^{k-r}}{s!}$$
$$= \sum_{r} \binom{k}{r} M_{k-r}$$

Let us try now to find a recurrence for the Bell numbers:

$$B_k = |P(k)|$$

258

A partition of  $\{1, \ldots, k+1\}$  appears by choosing r neighbors for 1, among the k numbers available, and then partitioning the k - r elements left. Thus, we have:

$$B_{k+1} = \sum_{r} \binom{k}{r} B_{k-r}$$

Thus, the numbers  $M_k$  satisfy the same recurrence as the numbers  $B_k$ .

Regarding now the initial values, for the moments of  $p_1$ , these are:

$$M_0 = 1$$
 ,  $M_1 = 1$ 

Indeed, the formula  $M_0 = 1$  is clear, and the formula  $M_1 = 1$  follows from:

$$M_1 = \frac{1}{e} \sum_s \frac{s}{s!}$$
$$= \frac{1}{e} \sum_s \frac{1}{(s-1)!}$$
$$= \frac{1}{e} \times e$$
$$= 1$$

Now by using the above recurrence we obtain from this:

$$M_2 = \sum_{r} {\binom{1}{r}} M_{k-r}$$
$$= 1+1$$
$$= 2$$

Thus, we can say that the initial values for the moments are:

$$M_1 = 1$$
 ,  $M_2 = 2$ 

As for the Bell numbers, here the initial values are:

$$B_1 = 1 \quad , \quad B_2 = 2$$

Thus the initial values coincide, and so these numbers are equal, as stated.

More generally, we have the following result:

THEOREM 11.17. The moments of  $p_t$  are given by

$$M_k(p_t) = \sum_{\pi \in P(k)} t^{|\pi|}$$

where |.| is the number of blocks.

PROOF. Observe first that the formula in the statement generalizes the one in Theorem 11.16 above, because at t = 1 we obtain, as we should:

$$M_k(p_1) = \sum_{\pi \in P(k)} 1^{|\pi|}$$
$$= |P(k)|$$
$$= B_k$$

In general now, the moments of  $p_t$  with t > 0 are given by:

$$M_k = e^{-t} \sum_s \frac{t^s s^k}{s!}$$

We have the following recurrence formula for the moments of  $p_t$ :

$$M_{k+1} = e^{-t} \sum_{s} \frac{t^{s+1}(s+1)^{k+1}}{(s+1)!}$$
  
=  $e^{-t} \sum_{s} \frac{t^{s+1}(s+1)^{k}}{s!}$   
=  $e^{-t} \sum_{s} \frac{t^{s+1}s^{k}}{s!} \left(1 + \frac{1}{s}\right)^{k}$   
=  $e^{-t} \sum_{s} \frac{t^{s+1}s^{k}}{s!} \sum_{r} {\binom{k}{r}} s^{-r}$   
=  $\sum_{r} {\binom{k}{r}} \cdot e^{-t} \sum_{s} \frac{t^{s+1}s^{k-r}}{s!}$   
=  $t \sum_{r} {\binom{k}{r}} M_{k-r}$ 

As for the initial values of these moments, these are as follows:

$$M_1 = t \quad , \quad M_2 = t + t^2$$

On the other hand, consider the numbers in the statement, namely:

$$S_k = \sum_{\pi \in P(k)} t^{|\pi|}$$

Since a partition of  $\{1, \ldots, k+1\}$  appears by choosing r neighbors for 1, among the k numbers available, and then partitioning the k - r elements left, we have:

$$S_{k+1} = t \sum_{r} \binom{k}{r} S_{k-r}$$

As for the initial values of these numbers, these are:

$$S_1 = t \quad , \quad S_2 = t + t^2$$

Thus the initial values coincide, so these numbers are the moments, as stated.  $\Box$ 

Observe the analogy with the moment formulae for  $g_t$  and  $G_t$ , discussed before. To be more precise, the moments for the main laws come from partitions, as follows:

$$p_t \to P$$
$$g_t \to P_2$$
$$G_t \to \mathcal{P}_2$$

We will be back later with some more conceptual explanations for these results.

# 11c. Truncated characters

With the above probabilistic preliminaries done, let us get back now to finite groups, and compute laws of characters. As a first piece of good news, our main result so far, namely Theorem 11.10, reformulates into something very simple, as follows:

THEOREM 11.18. For the symmetric group  $S_N \subset O_N$  we have

$$\chi \sim p_1$$

in the  $N \to \infty$  limit.

PROOF. This is indeed a reformulation of Theorem 11.10 above, which tells us that with  $N \to \infty$  we have the following estimate:

$$\mathbb{P}(\chi = k) \simeq \frac{1}{ek!}$$

But, according to our definition of the Poisson laws, this tells us precisely that the asymptotic law of the main character  $\chi$  is Poisson (1), as stated.

An interesting question now is that of recovering all the Poisson laws  $p_t$ , by using group theory. In order to do this, let us formulate the following definition:

DEFINITION 11.19. Given a closed subgroup  $G \subset U_N$ , the function

$$\chi: G \to \mathbb{C}$$
$$\chi_t(g) = \sum_{i=1}^{[tN]} g_{ii}$$

is called main truncated character of G, of parameter  $t \in (0, 1]$ .

As before with the plain characters, there is some general theory behind this definition, and we will discuss this later on, systematically, in chapters 13-16 below.

Getting back now to the symmetric groups, we first have the following result:

PROPOSITION 11.20. Consider the symmetric group  $S_N$ , regarded as the permutation group of the N coordinate axes of  $\mathbb{R}^N$ . This picture provides us with an embedding

$$S_N \subset O_N$$

the coordinate functions for the permutations being as follows:

$$g_{ij} = \chi \left( \sigma \in S_N \middle| \sigma(j) = i \right)$$

In this picture, the truncated characters count the number of partial fixed points

$$\chi_t(\sigma) = \#\left\{i \in \{1, \dots, [tN]\} \middle| \sigma(i) = i\right\}$$

with respect to the truncation parameter  $t \in (0, 1]$ .

PROOF. All this is clear from definitions, with the formula for the coordinates being clear from the definition of the embedding  $S_N \subset O_N$ , and with the character formulae following from it, by summing over i = j. To be more precise, we have:

$$\chi_t(\sigma) = \sum_{i=1}^{[tN]} \sigma_{ii}$$
$$= \sum_{i=1}^{[tN]} \delta_{\sigma(i)i}$$
$$= \# \left\{ i \in \{1, \dots, [tN]\} \middle| \sigma(i) = i \right\}$$

Thus, we are led to the conclusions in the statement.

Regarding now the asymptotic laws of the truncated characters, the result here, generalizing everything that we have so far, is as follows:

THEOREM 11.21. For the symmetric group  $S_N \subset O_N$  we have

$$\chi_t \sim p_t$$

in the  $N \to \infty$  limit, for any  $t \in (0, 1]$ .

PROOF. We already know from Theorem 11.18 that the result holds at t = 1. In general, the proof is similar, the idea being as follows:

(1) Consider indeed the following sets, as in the proof of Theorem 11.18, or rather as in the proof of Theorem 11.7, leading to Theorem 11.18:

$$S_N^i = \left\{ \sigma \in S_N \middle| \sigma(i) = i \right\}$$

The set of permutations having no fixed points among  $1, \ldots, [tN]$  is then:

$$X_N = \left(\bigcup_{i \le [tN]} S_N^i\right)^c$$

In order to compute now the cardinality  $|X_N|$ , consider as well the following sets, depending on indices  $i_1 < \ldots < i_k$ , obtained by taking intersections:

$$S_N^{i_1\dots i_k} = S_N^{i_1} \bigcap \dots \bigcap S_N^{i_k}$$

As before in the proof of Theorem 11.18, we obtain by inclusion-exclusion that:

$$P(\chi_t = 0) = \frac{1}{N!} \sum_{k=0}^{[tN]} (-1)^k \sum_{i_1 < \dots < i_k < [tN]} |S_N^{i_1 \dots i_k}|$$
  
$$= \frac{1}{N!} \sum_{k=0}^{[tN]} (-1)^k \sum_{i_1 < \dots < i_k < [tN]} (N-k)!$$
  
$$= \frac{1}{N!} \sum_{k=0}^{[tN]} (-1)^k {\binom{[tN]}{k}} (N-k)!$$
  
$$= \sum_{k=0}^{[tN]} \frac{(-1)^k}{k!} \cdot \frac{[tN]!(N-k)!}{N!([tN]-k)!}$$

With  $N \to \infty$ , we obtain from this the following estimate:

$$P(\chi_t = 0) \simeq \sum_{k=0}^{[tN]} \frac{(-1)^k}{k!} \cdot t^k$$
$$= \sum_{k=0}^{[tN]} \frac{(-t)^k}{k!}$$
$$\simeq e^{-t}$$

(2) More generally now, by counting the permutations  $\sigma \in S_N$  having exactly k fixed points among  $1, \ldots, [tN]$ , as in the proof of Theorem 11.9, our claim is that we get:

$$P(\chi_t = k) \simeq \frac{t^k}{k!e^t}$$

We already know from (1) that this formula holds at k = 0. In the general case now, we have to count the permutations  $\sigma \in S_N$  having exactly k fixed points among  $1, \ldots, [tN]$ . Since having such a permutation amounts in choosing k points among  $1, \ldots, [tN]$ , and

then permuting the N - k points left, without fixed points among  $1, \ldots, [tN]$  allowed, we obtain the following formula, where  $s \in (0, 1]$  is such that [s(N - k)] = [tN] - k:

$$\# \left\{ \sigma \in S_N \middle| \chi_t(\sigma) = k \right\} = \binom{[tN]}{k} \# \left\{ \sigma \in S_{N-k} \middle| \chi_s(\sigma) = 0 \right\} \\
= \frac{[tN]!}{k!([tN] - k)!} \# \left\{ \sigma \in S_{N-k} \middle| \chi_s(\sigma) = 0 \right\} \\
= \frac{1}{k!} \times \frac{[tN]!(N-k)!}{([tN] - k)!} \times \frac{\# \left\{ \sigma \in S_{N-k} \middle| \chi_s(\sigma) = 0 \right\}}{(N-k)!}$$

Now by dividing everything by N!, we obtain from this the following formula:

$$\frac{\#\left\{\sigma \in S_N \middle| \chi_t(\sigma) = k\right\}}{N!} = \frac{1}{k!} \times \frac{[tN]!(N-k)!}{N!([tN]-k)!} \times \frac{\#\left\{\sigma \in S_{N-k} \middle| \chi_s(\sigma) = 0\right\}}{(N-k)!}$$

By using now the computation at k = 0, that we already have, from (1) above, it follows that with  $N \to \infty$  we have the following estimate:

$$P(\chi_t = k) \simeq \frac{1}{k!} \times \frac{[tN]!(N-k)!}{N!([tN]-k)!} \cdot P(\chi_s = 0)$$
$$\simeq \frac{t^k}{k!} \cdot P(\chi_s = 0)$$
$$\simeq \frac{t^k}{k!} \cdot \frac{1}{e^s}$$

Now recall that the parameter  $s \in (0, 1]$  was chosen in the above such that:

$$[s(N-k)] = [tN] - k$$

Thus in the  $N \to \infty$  limit we have s = t, and so we obtain, as claimed:

$$P(\chi_t = k) \simeq \frac{t^k}{k!} \cdot \frac{1}{e^t}$$

It follows that we obtain in the limit a Poisson law of parameter t, as stated.

# 11d. Further results

All the above is quite interesting, and is at the core of the theory that we want to develop, so let us present now a new proof for the above character results.

The point indeed is that we can approch the problems as well directly, by integrating over  $S_N$ , and in order to do so, we can use the following result:

THEOREM 11.22. Consider the symmetric group  $S_N$ , with its standard coordinates:

$$g_{ij} = \chi \left( \sigma \in S_N \middle| \sigma(j) = i \right)$$

The products of these coordinates span the algebra  $C(S_N)$ , and the arbitrary integrals over  $S_N$  are given, modulo linearity, by the formula

$$\int_{S_N} g_{i_1 j_1} \dots g_{i_k j_k} = \begin{cases} \frac{(N-|\ker i|)!}{N!} & \text{if } \ker i = \ker j\\ 0 & \text{otherwise} \end{cases}$$

where ker *i* denotes as usual the partition of  $\{1, \ldots, k\}$  whose blocks collect the equal indices of *i*, and where |.| denotes the number of blocks.

PROOF. The first assertion follows from the Stone-Weierstrass theorem, because the standard coordinates  $g_{ij}$  separate the points of  $S_N$ , and so the algebra  $\langle g_{ij} \rangle$  that they generate must be equal to the whole function algebra  $C(S_N)$ :

$$\langle g_{ij} \rangle = C(S_N)$$

Regarding now the second assertion, according to the definition of  $g_{ij}$ , the integrals in the statement are given by:

$$\int_{S_N} g_{i_1 j_1} \dots g_{i_k j_k} = \frac{1}{N!} \# \left\{ \sigma \in S_N \middle| \sigma(j_1) = i_1, \dots, \sigma(j_k) = i_k \right\}$$

Now observe that the existence of  $\sigma \in S_N$  as above requires:

$$i_m = i_n \iff j_m = j_n$$

Thus, the above integral vanishes when:

$$\ker i \neq \ker j$$

Regarding now the case ker  $i = \ker j$ , if we denote by  $b \in \{1, \ldots, k\}$  the number of blocks of this partition ker  $i = \ker j$ , we have N - b points to be sent bijectively to N - b points, and so (N - b)! solutions, and the integral is  $\frac{(N-b)!}{N!}$ , as claimed.

As an illustration for the above formula, we can recover the computation of the asymptotic laws of the truncated characters  $\chi_t$ . We have indeed:

THEOREM 11.23. For the symmetric group  $S_N \subset O_N$ , regarded as a compact group of matrices,  $S_N \subset O_N$ , via the standard permutation matrices, the truncated character

$$\chi_t(g) = \sum_{i=1}^{[tN]} g_{ii}$$

counts the number of fixed points among  $\{1, \ldots, [tN]\}$ , and its law with respect to the counting measure becomes, with  $N \to \infty$ , a Poisson law of parameter t.

**PROOF.** The first assertion comes from the following formula:

$$g_{ij} = \chi\left(\sigma \middle| \sigma(j) = i\right)$$

Regarding now the second assertion, we can use here the integration formula in Theorem 11.22 above. With  $S_{kb}$  being the Stirling numbers, counting the partitions of  $\{1, \ldots, k\}$  having exactly b blocks, we have the following formula:

$$\int_{S_N} \chi_t^k = \sum_{i_1 \dots i_k=1}^{[tN]} \int_{S_N} g_{i_1 i_1} \dots g_{i_k i_k}$$
$$= \sum_{\pi \in P(k)} \frac{[tN]!}{([tN] - |\pi|!)} \cdot \frac{(N - |\pi|!)}{N!}$$
$$= \sum_{b=1}^{[tN]} \frac{[tN]!}{([tN] - b)!} \cdot \frac{(N - b)!}{N!} \cdot S_{kb}$$

In particular with  $N \to \infty$  we obtain the following formula:

$$\lim_{N \to \infty} \int_{S_N} \chi_t^k = \sum_{b=1}^k S_{kb} t^b$$

But this is the k-th moment of the Poisson law  $p_t$ , and so we are done.

Summarizing, we have a good understanding of our main result so far, involving the characters of the symmetric group  $S_N$  and the Poisson laws of parameter  $t \in (0, 1]$ , by using 2 different methods. We will see in a moment a third proof as well, and we will be actually back to this in chapters 13-16 too, with a fourth method as well.

As another result now regarding  $S_N$ , here is a useful related formula:

THEOREM 11.24. We have the law formula

$$law(g_{11} + \ldots + g_{ss}) = \frac{s!}{N!} \sum_{p=0}^{s} \frac{(N-p)!}{(s-p)!} \cdot \frac{(\delta_1 - \delta_0)^{*p}}{p!}$$

where  $g_{ij}$  are the standard coordinates of  $S_N \subset O_N$ .

**PROOF.** We have the following moment formula, where  $m_f$  is the number of permutations of  $\{1, \ldots, N\}$  having exactly f fixed points in the set  $\{1, \ldots, s\}$ :

$$\int_{S_N} (u_{11} + \ldots + u_{ss})^k = \frac{1}{N!} \sum_{f=0}^s m_f f^k$$

Thus the law in the statement, say  $\nu_{sN}$ , is the following average of Dirac masses:

$$\nu_{sN} = \frac{1}{N!} \sum_{f=0}^{s} m_f \,\delta_f$$

Now observe that the permutations contributing to  $m_f$  are obtained by choosing f points in the set  $\{1, \ldots, s\}$ , then by permuting the remaining N - f points in  $\{1, \ldots, n\}$  in such a way that there is no fixed point in  $\{1, \ldots, s\}$ .

But these latter permutations are counted as follows: we start with all permutations, we substract those having one fixed point, we add those having two fixed points, and so on. We obtain in this way the following formula:

$$\nu_{sN} = \frac{1}{N!} \sum_{f=0}^{s} {\binom{s}{f}} \left( \sum_{k=0}^{s-f} (-1)^{k} {\binom{s-f}{k}} (N-f-k)! \right) \delta_{f}$$
  
$$= \sum_{f=0}^{s} \sum_{k=0}^{s-f} (-1)^{k} \frac{1}{N!} \cdot \frac{s!}{f!(s-f)!} \cdot \frac{(s-f)!(N-f-k)!}{k!(s-f-k)!} \delta_{f}$$
  
$$= \frac{s!}{N!} \sum_{f=0}^{s} \sum_{k=0}^{s-f} \frac{(-1)^{k}(N-f-k)!}{f!k!(s-f-k)!} \delta_{f}$$

We can proceed as follows, by using the new index p = f + k:

$$\nu_{sN} = \frac{s!}{N!} \sum_{p=0}^{s} \sum_{k=0}^{p} \frac{(-1)^{k} (N-p)!}{(p-k)! k! (s-p)!} \,\delta_{p-k}$$
$$= \frac{s!}{N!} \sum_{p=0}^{s} \frac{(N-p)!}{(s-p)! p!} \sum_{k=0}^{p} (-1)^{k} \binom{p}{k} \,\delta_{p-k}$$
$$= \frac{s!}{N!} \sum_{p=0}^{s} \frac{(N-p)!}{(s-p)!} \cdot \frac{(\delta_{1}-\delta_{0})^{*p}}{p!}$$

Here \* is convolution of real measures, and the assertion follows.

Observe that the above formula is finer than most of our previous formulae, which were asymptotic, because it is valid at any  $N \in \mathbb{N}$ .

We can use the above formula, as follows:

THEOREM 11.25. Let  $g_{ij}$  be the standard coordinates of  $C(S_N)$ . (1)  $u_{11} + \ldots + u_{ss}$  with s = o(N) is a projection of trace s/N.

(2)  $u_{11} + \ldots + u_{ss}$  with s = tN + o(N) is Poisson of parameter t.

**PROOF.** We use the formula in Theorem 11.24 above.

(1) With s fixed and  $N \to \infty$  we have the following estimate:

$$\begin{aligned}
& = \sum_{p=0}^{s} \frac{(N-p)!}{N!} \cdot \frac{s!}{(s-p)!} \cdot \frac{(\delta_1 - \delta_0)^{*p}}{p!} \\
& = \delta_0 + \frac{s}{N} \left(\delta_1 - \delta_0\right) + O(N^{-2})
\end{aligned}$$

But the law on the right is that of a projection of trace s/N, as desired.

(2) We have a law formula of the following type:

$$law(u_{11} + \ldots + u_{ss}) = \sum_{p=0}^{s} c_p \cdot \frac{(\delta_1 - \delta_0)^{*p}}{p!}$$

The coefficients  $c_p$  can be estimated by using the Stirling formula, as follows:

$$c_p = \frac{(tN)!}{N!} \cdot \frac{(N-p)!}{(tN-p)!}$$

$$\simeq \frac{(tN)^{tN}}{N^N} \cdot \frac{(N-p)^{N-p}}{(tN-p)^{tN-p}}$$

$$= \left(\frac{tN}{tN-p}\right)^{tN-p} \left(\frac{N-p}{N}\right)^{N-p} \left(\frac{tN}{N}\right)^p$$

The last expression can be estimated by using the definition of exponentials, and we obtain the following estimate:

$$c_p \simeq e^p e^{-p} t^p$$
$$= t^p$$

We can now compute the Fourier transform with respect to a variable y:

$$\mathcal{F}\left(\operatorname{law}(u_{11} + \ldots + u_{ss})\right) \simeq \sum_{p=0}^{s} t^{p} \cdot \frac{(e^{y} - 1)^{p}}{p!}$$
$$= e^{t(e^{y} - 1)}$$

But this is the Fourier transform of the Poisson law  $p_t$ , as explained in Theorem 11.14 above, and this gives the second assertion.

Let us discuss now, as an instructive variation of the above, the computation for the alternating group  $A_N \subset S_N$ . We will see that with  $N \to \infty$  nothing changes, and with this being part of a more general phenomenon, regarding more general types of reflection groups and subgroups, that we will further discuss in the next chapter.

Let us start with some algebraic considerations. We first have:

PROPOSITION 11.26. For the symmetric group, regarded as group of permutations of the N coordinate axes of  $\mathbb{R}^N$ , and so as group of permutation matrices,

$$S_N \subset O_N$$

the determinant is the signature. The subgroup  $A_N \subset S_N$  given by

$$A_N = S_N \cap SO_N$$

and called alternating group, consists of the even permutations.

PROOF. In this statement the first assertion is clear from the definition of the determinant, and of the permutation matrices, and all the rest is standard.  $\Box$ 

Regarding now character computations, the best here is to use an analogue of Theorem 11.22 above. To be more precise, we have here the following result:

THEOREM 11.27. Consider the alternating group  $A_N$ , regarded as group of permutation matrices, with its standard coordinates:

$$g_{ij} = \chi \left( \sigma \in A_N \middle| \sigma(j) = i \right)$$

The products of these coordinates span the algebra  $C(A_N)$ , and the arbitrary integrals over  $A_N$  are given, modulo linearity, by the formula

$$\int_{A_N} g_{i_1 j_1} \dots g_{i_k j_k} \simeq \begin{cases} \frac{(N - |\ker i|)!}{N!} & \text{if } \ker i = \ker j \\ 0 & \text{otherwise} \end{cases}$$

with  $N \to \infty$ , where ker *i* denotes as usual the partition of  $\{1, \ldots, k\}$  whose blocks collect the equal indices of *i*, and where |.| denotes the number of blocks.

**PROOF.** The first assertion follows from the Stone-Weierstrass theorem, because the standard coordinates  $g_{ij}$  separate the points of  $A_N$ , and so the algebra  $\langle g_{ij} \rangle$  that they generate must be equal to the whole function algebra  $C(A_N)$ :

$$\langle g_{ij} \rangle = C(A_N)$$

Regarding now the second assertion, according to the definition of the standard coordinates  $g_{ij}$ , the integrals in the statement are given by:

$$\int_{A_N} g_{i_1 j_1} \dots g_{i_k j_k} = \frac{1}{N!/2} \# \left\{ \sigma \in A_N \left| \sigma(j_1) = i_1, \dots, \sigma(j_k) = i_k \right\} \right\}$$

Now observe that the existence of  $\sigma \in A_N$  as above requires:

 $i_m = i_n \iff j_m = j_n$ 

Thus, the above integral vanishes when:

 $\ker i \neq \ker j$ 

Regarding now the case ker  $i = \ker j$ , if we denote by  $b \in \{1, \ldots, k\}$  the number of blocks of this partition ker  $i = \ker j$ , we have N - b points to be sent bijectively to N - b points. But when assuming N >> 0, and more specifically N > k, half of these bijections will be alternating, and so we have (N - b)!/2 solutions. Thus, the integral is:

$$\int_{A_N} g_{i_1 j_1} \dots g_{i_k j_k} = \frac{1}{N!/2} \# \left\{ \sigma \in A_N \middle| \sigma(j_1) = i_1, \dots, \sigma(j_k) = i_k \right\}$$
$$= \frac{(N-b)!/2}{N!/2}$$
$$= \frac{(N-b)!}{N!}$$

Thus, we are led to the conclusion in the statement.

As an illustration for the above formula, we can recover the computation of the asymptotic laws of the truncated characters  $\chi_t$ . We have indeed:

THEOREM 11.28. For the alternating group  $A_N \subset O_N$ , regarded as a compact group of matrices,  $A_N \subset O_N$ , via the standard permutation matrices, the truncated character

$$\chi_t(g) = \sum_{i=1}^{[tN]} g_{ii}$$

counts the number of fixed points among  $\{1, \ldots, [tN]\}$ , and its law with respect to the counting measure becomes, with  $N \to \infty$ , a Poisson law of parameter t.

**PROOF.** The first assertion comes from the following formula:

$$g_{ij} = \chi\left(\sigma \middle| \sigma(j) = i\right)$$

Regarding now the second assertion, we can use here the integration formula in Theorem 11.27 above. With  $S_{kb}$  being the Stirling numbers, counting the partitions of  $\{1, \ldots, k\}$  having exactly b blocks, we have the following formula:

$$\int_{A_N} \chi_t^k = \sum_{i_1...i_k=1}^{[tN]} \int_{A_N} g_{i_1 i_1} \dots g_{i_k i_k}$$
$$\simeq \sum_{\pi \in P(k)} \frac{[tN]!}{([tN] - |\pi|!)} \cdot \frac{(N - |\pi|!)}{N!}$$
$$= \sum_{b=1}^{[tN]} \frac{[tN]!}{([tN] - b)!} \cdot \frac{(N - b)!}{N!} \cdot S_{kb}$$

270

#### 11E. EXERCISES

In particular with  $N \to \infty$  we obtain the following formula:

$$\lim_{N \to \infty} \int_{S_N} \chi_t^k = \sum_{b=1}^k S_{kb} t^b$$

But this is the k-th moment of the Poisson law  $p_t$ , and so we are done.

As a conclusion to this, when passing from the symmetric group  $S_N$  to its subgroup  $A_N \subset S_N$ , in what concerns character computations, with  $N \to \infty$  nothing changes. This is actually part of a more general phenomenon, regarding more general types of reflection groups and subgroups, that we will further discuss in the next chapter.

As a conclusion to all this, we have seen that the truncated characters  $\chi_t$  of the symmetric group  $S_N$  have the Poisson laws  $p_t$  as limiting distributions, with  $N \to \infty$ . Moreover, we have seen several proofs for this fundamental fact, using inclusion-exclusion, direct integration, and convolution exponentials and Fourier transforms as well.

We will keep building on all this in the next chapter, by stating and proving similar results for more general reflection groups  $G \subset U_N$ . Also, we will be back to the symmetric group  $S_N$  and to the Poisson laws in chapters 13-16, with a fourth proof for our results, using representation theory, and a property of  $S_N$  called easiness. More on this later.

# 11e. Exercises

There are many interesting exercises in connection with the above. First, in relation with derangements and fixed points, we have:

EXERCISE 11.29. Compute the number of derangements in  $S_4$ , by explicitly listing them, and then comment on the estimate of

$$e = 2.7182...$$

that you obtain in this way.

Here the first question is of course elementary, but the problem is that of finding out what the best notation for permutations is, in order to solve this problem quickly. As for the second question, that you can investigate at higher N too, based on the various formulae established in this chapter, this is something quite instructive too.

No discussion about e would be complete without a similar discussion about  $\pi$ , and we have here the following well-known and beautiful exercise:

EXERCISE 11.30. Show that the probability for a length 1 needle to intersect, when thrown, a 1-spaced grid is  $2/\pi$ , and then comment on the estimate on

$$\pi = 3.1415...$$

that you obtain in this way.

271

Here the first question is quite tricky, because there are several possible ways of modelling the problem, but only one of them gives the correct, real-life answer. As for the second question, this is a good introduction to applied mathematics too.

In relation now with the Poisson laws, we have:

EXERCISE 11.31. Interpret the abstract PLT formula established above, namely

$$\left(\left(1-\frac{t}{n}\right)\delta_0+\frac{t}{n}\delta_1\right)^{*n}\to p_t$$

as a Poisson Limit Theorem, with full probabilistic details.

Here the problem is to understand why the above formula, that we already know, makes the Poisson law appears everywhere, in real life.

Also in relation with the Poisson laws, we have the following question, regarding their moments, which are the Bell numbers:

EXERCISE 11.32. Find some formulae for the Bell numbers  $B_k$ , or rather for their generating series, or suitable transforms of that series, and the more the better.

There is a lot of interesting mathematics here, and after solving the exercise, you can check the internet, and complete your knowledge with more things.

In relation now with character computations, we have:

EXERCISE 11.33. Show that the truncated characters of  $S_N$ , suitably moved over the diagonal, as to not overlap, become independent with  $N \to \infty$ .

Here the formulation is of course a bit loose, but this is intentional, and finding the precise formulation is part of the exercise. As for the proof, this can only come by using the various integration formulae over  $S_N$  established in the above.

Finally, in relation with the alternating group, we have:

EXERCISE 11.34. Find some alternative proofs for the fact, that we already know, that the truncated charcters for  $A_N \subset O_N$  become Poisson, with  $N \to \infty$ .

This is a bit technical, the problem being that of picking the best alternative proof for  $S_N$ , from the above, and then extending it to  $A_N$ . As a bonus exercise, you can work out as well independence aspects for  $A_N$ , in the spirit of the previous exercise.

# CHAPTER 12

# **Reflection groups**

# 12a. Real reflections

We have seen in the previous chapter that some interesting computations, in relation with the law of the main character, happen for the symmetric group  $S_N$ , in the  $N \to \infty$ limit. Moreover, we have seen as well that the same kind of conclusions, with the law of the main character becoming Poisson (1), and the laws of the truncated characters becoming Poisson (t), in the  $N \to \infty$  limit, happen for the alternating subgroup  $A_N \subset S_N$ .

All this suggests systematically looking at the general series of complex reflection groups  $H_N^{sd}$ . However, things are quite technical here, and we will do this slowly. First we will study the hyperoctahedral group  $H_N$ . Then we will develop some more general probabilistic theory, in relation with the Poisson laws and their versions. Then we will discuss the groups  $H_N^s$ , generalizing both  $S_N$  and  $H_N$ . And finally, we will comment on the groups  $H_N^{sd}$ , and we will do some new computations for continuous groups as well.

Summarizing, a lot of things to be done. Let us start by discussing the hyperoctahedral group  $H_N$ . We first recall, from chapter 9 above, that we have:

THEOREM 12.1. Consider the hyperoctahedral group  $H_N$ , which appears as the symmetry group of the N-cube, or the symmetry group of the N coordinate axes of  $\mathbb{R}^N$ :

$$S_N \subset H_N \subset O_N$$

In matrix terms,  $H_N$  consists of the permutation-type matrices having  $\pm 1$  as nonzero entries, and we have a wreath product decomposition as follows:

$$H_N = \mathbb{Z}_2 \wr S_N$$

In this picture, the main character counts the signed number of fixed points, among the coordinate axes, and its truncations count the truncations of such numbers.

**PROOF.** This is something that was already discussed before, the idea being that the first assertions are clear, and that the wreath product decomposition in the statement corresponds by definition to a crossed product decomposition, as follows:

$$H_N = \mathbb{Z}_2^N \rtimes S_N$$

As for the assertions regarding the main character and its truncations, once again these are clear, exactly as in the case of the symmetric group  $S_N$ .

#### 12. REFLECTION GROUPS

Regarding now the character laws, we can compute them by using the same method as for the symmetric group  $S_N$ , namely inclusion-exclusion, and we have:

THEOREM 12.2. For the hyperoctahedral group  $H_N \subset O_N$ , the law of the variable

$$\chi_t = \sum_{i=1}^{[tN]} g_{ii}$$

is in the  $N \to \infty$  limit the measure

$$b_t = e^{-t} \sum_{k=-\infty}^{\infty} \delta_k \sum_{p=0}^{\infty} \frac{(t/2)^{|k|+2p}}{(|k|+p)!p!}$$

where  $\delta_k$  is the Dirac mass at  $k \in \mathbb{Z}$ .

**PROOF.** We regard  $H_N$  as being the symmetry group of the graph  $I_N = \{I^1, \ldots, I^N\}$  formed by N segments. The diagonal coefficients are given by:

$$u_{ii}(g) = \begin{cases} 0 & \text{if } g \text{ moves } I^i \\ 1 & \text{if } g \text{ fixes } I^i \\ -1 & \text{if } g \text{ returns } I^i \end{cases}$$

We denote by  $\uparrow g, \downarrow g$  the number of segments among  $\{I^1, \ldots, I^s\}$  which are fixed, respectively returned by an element  $g \in H_N$ . With this notation, we have:

$$u_{11} + \ldots + u_{ss} = \uparrow g - \downarrow g$$

We denote by  $P_N$  probabilities computed over the group  $H_N$ . The density of the law of  $u_{11} + \ldots + u_{ss}$  at a point  $k \ge 0$  is given by the following formula:

$$D(k) = P_N(\uparrow g - \downarrow g = k)$$
  
= 
$$\sum_{p=0}^{\infty} P_N(\uparrow g = k + p, \downarrow g = p)$$

Assume first that we have t = 1. We use the fact that the probability of  $\sigma \in S_N$  to have no fixed points is asymptotically:

$$P_0 = 1/e$$

Thus the probability of  $\sigma \in S_N$  to have m fixed points is asymptotically:

$$P_m = 1/(em!)$$

In terms of probabilities over  $H_N$ , we get:

$$\lim_{N \to \infty} D(k) = \lim_{N \to \infty} \sum_{p=0}^{\infty} (1/2)^{k+2p} \binom{k+2p}{k+p} P_N(\uparrow g + \downarrow g = k+2p)$$
$$= \sum_{p=0}^{\infty} (1/2)^{k+2p} \binom{k+2p}{k+p} \frac{1}{e(k+2p)!}$$
$$= \frac{1}{e} \sum_{p=0}^{\infty} \frac{(1/2)^{k+2p}}{(k+p)!p!}$$

The general case  $0 < t \leq 1$  follows by performing some modifications in the above computation. The asymptotic density is computed as follows:

$$\lim_{N \to \infty} D(k) = \lim_{N \to \infty} \sum_{p=0}^{\infty} (1/2)^{k+2p} \binom{k+2p}{k+p} P_N(\uparrow g + \downarrow g = k+2p)$$
$$= \sum_{p=0}^{\infty} (1/2)^{k+2p} \binom{k+2p}{k+p} \frac{t^{k+2p}}{e^t(k+2p)!}$$
$$= e^{-t} \sum_{p=0}^{\infty} \frac{(t/2)^{k+2p}}{(k+p)!p!}$$

Together with D(-k) = D(k), this gives the formula in the statement.

# 12b. Bessel laws

The above result is quite interesting, because the densities there are the Bessel functions of the first kind. Due to this fact, the limiting measures are called Bessel laws:

DEFINITION 12.3. The Bessel law of parameter t > 0 is the measure

$$b_t = e^{-t} \sum_{k=-\infty}^{\infty} \delta_k f_k(t/2)$$

with the density being the Bessel function of the first kind:

$$f_k(t) = \sum_{p=0}^{\infty} \frac{t^{|k|+2p}}{(|k|+p)!p!}$$

Let us study now these Bessel laws. We first have the following result:

THEOREM 12.4. The Bessel laws  $b_t$  have the property

$$b_s * b_t = b_{s+t}$$

so they form a truncated one-parameter semigroup with respect to convolution.

# 12. REFLECTION GROUPS

PROOF. We use the formula in Definition 12.3, namely:

$$b_t = e^{-t} \sum_{k=-\infty}^{\infty} \delta_k f_k(t/2)$$

The Fourier transform of this measure is given by:

$$Fb_t(y) = e^{-t} \sum_{k=-\infty}^{\infty} e^{ky} f_k(t/2)$$

We compute now the derivative with respect to t:

$$Fb_t(y)' = -Fb_t(y) + \frac{e^{-t}}{2} \sum_{k=-\infty}^{\infty} e^{ky} f'_k(t/2)$$

On the other hand, the derivative of  $f_k$  with  $k \ge 1$  is given by:

$$\begin{split} f_k'(t) &= \sum_{p=0}^{\infty} \frac{(k+2p)t^{k+2p-1}}{(k+p)!p!} \\ &= \sum_{p=0}^{\infty} \frac{(k+p)t^{k+2p-1}}{(k+p)!p!} + \sum_{p=0}^{\infty} \frac{pt^{k+2p-1}}{(k+p)!p!} \\ &= \sum_{p=0}^{\infty} \frac{t^{k+2p-1}}{(k+p-1)!p!} + \sum_{p=1}^{\infty} \frac{t^{k+2p-1}}{(k+p)!(p-1)!} \\ &= \sum_{p=0}^{\infty} \frac{t^{(k-1)+2p}}{((k-1)+p)!p!} + \sum_{p=1}^{\infty} \frac{t^{(k+1)+2(p-1)}}{((k+1)+(p-1))!(p-1)!} \\ &= f_{k-1}(t) + f_{k+1}(t) \end{split}$$

This computation works in fact for any k, so we get:

$$Fb_{t}(y)' = -Fb_{t}(y) + \frac{e^{-t}}{2} \sum_{k=-\infty}^{\infty} e^{ky} (f_{k-1}(t/2) + f_{k+1}(t/2))$$
  
$$= -Fb_{t}(y) + \frac{e^{-t}}{2} \sum_{k=-\infty}^{\infty} e^{(k+1)y} f_{k}(t/2) + e^{(k-1)y} f_{k}(t/2)$$
  
$$= -Fb_{t}(y) + \frac{e^{y} + e^{-y}}{2} Fb_{t}(y)$$
  
$$= \left(\frac{e^{y} + e^{-y}}{2} - 1\right) Fb_{t}(y)$$

Thus the log of the Fourier transform is linear in t, and we get the assertion.

#### 12C. COMPLEX REFLECTIONS

# 12c. Complex reflections

In order to further discuss all this, we will need a number of probabilistic preliminaries. So, let us pause now from our reflection group study, and do some theoretical probability theory, as a continuation of the material from chapter 11 above.

We recall that, conceptually speaking, the Poisson laws are the laws appearing via the Poisson Limit Theorem (PLT). In order to generalize this construction, as to cover for instance for Bessel laws that we found in connection with the hyperoctahedral group  $H_N$ , we have the following notion, extending the Poisson limit theory from chapter 11:

DEFINITION 12.5. Associated to any compactly supported positive measure  $\nu$  on  $\mathbb{R}$  is the probability measure

$$p_{\nu} = \lim_{n \to \infty} \left( \left( 1 - \frac{c}{n} \right) \delta_0 + \frac{1}{n} \nu \right)^{*n}$$

where  $c = mass(\nu)$ , called compound Poisson law.

In other words, what we are doing here is to generalize the construction in the Poisson Limit Theorem, by allowing the only parameter there, which was the positive real number t > 0, to be replaced by a certain probability measure  $\nu$ , or arbitrary mass c > 0.

In what follows we will be interested in the case where  $\nu$  is discrete, as is for instance the case for  $\nu = t\delta_1$  with t > 0, which produces the Poisson laws.

To be more precise, we will be mainly interested in the case where  $\nu$  is a multiple of the uniform measure on the *s*-th roots of unity. More on this later.

The following result allows one to detect compound Poisson laws:

**PROPOSITION 12.6.** For a discrete measure, written as

$$\nu = \sum_{i=1}^{s} c_i \delta_{z_i}$$

with  $c_i > 0$  and  $z_i \in \mathbb{R}$ , we have

$$F_{p_{\nu}}(y) = \exp\left(\sum_{i=1}^{s} c_i(e^{iyz_i} - 1)\right)$$

where F denotes the Fourier transform.

**PROOF.** Let  $\mu_n$  be the measure appearing in Definition 12.5, under the convolution signs, namely:

$$\mu_n = \left(1 - \frac{c}{n}\right)\delta_0 + \frac{1}{n}\nu$$

We have the following computation:

$$F_{\mu_n}(y) = \left(1 - \frac{c}{n}\right) + \frac{1}{n} \sum_{i=1}^s c_i e^{iyz_i}$$
$$\implies F_{\mu_n^{*n}}(y) = \left(\left(1 - \frac{c}{n}\right) + \frac{1}{n} \sum_{i=1}^s c_i e^{iyz_i}\right)^n$$
$$\implies F_{p_\nu}(y) = \exp\left(\sum_{i=1}^s c_i (e^{iyz_i} - 1)\right)$$

Thus, we have obtained the formula in the statement.

We have as well the following result, providing an alternative to Definition 12.5:

THEOREM 12.7. For a discrete measure, written as

$$\nu = \sum_{i=1}^{s} c_i \delta_{z_i}$$

with  $c_i > 0$  and  $z_i \in \mathbb{R}$ , we have

$$p_{\nu} = \operatorname{law}\left(\sum_{i=1}^{s} z_{i}\alpha_{i}\right)$$

where the variables  $\alpha_i$  are Poisson  $(c_i)$ , independent.

**PROOF.** Let  $\alpha$  be the sum of Poisson variables in the statement:

$$\alpha = \sum_{i=1}^{s} z_i \alpha_i$$

By using some well-known Fourier transform formulae, we have:

$$F_{\alpha_i}(y) = \exp(c_i(e^{iy} - 1)) \implies F_{z_i\alpha_i}(y) = \exp(c_i(e^{iyz_i} - 1))$$
$$\implies F_{\alpha}(y) = \exp\left(\sum_{i=1}^s c_i(e^{iyz_i} - 1)\right)$$

Thus we have indeed the same formula as in Proposition 12.6.

Getting back now to the Bessel laws, we have:

THEOREM 12.8. The Bessel laws  $b_t$  are compound Poisson laws, given by

$$b_t = p_{t\varepsilon}$$

where  $\varepsilon = \frac{1}{2}(\delta_{-1} + \delta_1)$  is the uniform measure on  $\mathbb{Z}_2$ .

**PROOF.** This follows indeed by comparing the formula of the Fourier transform of  $b_t$ , from the proof of Theorem 12.2 above, with the formula in Proposition 12.7.

278

Our next task will be that of generalizing the results that we have for  $S_N, H_N$ . For this purpose, let us consider the following family of groups:

DEFINITION 12.9. The complex reflection group  $H_N^s \subset U_N$ , depending on parameters

 $N \in \mathbb{N}$  ,  $s \in \mathbb{N} \cup \{\infty\}$ 

are the groups of permutation-type matrices with s-th roots of unity as entries,

 $H_N^s = M_N(\mathbb{Z}_s \cup \{0\}) \cap U_N$ 

with the convention  $\mathbb{Z}_{\infty} = \mathbb{T}$ , at  $s = \infty$ .

Observe that at s = 1, 2 we obtain the symmetric and hyperoctahedral groups:

$$H_N^1 = S_N$$
$$H_N^2 = H_N$$

Another important particular case is  $s = \infty$ , where we obtain a group which is actually not finite, denoted as follows:

$$H_N^\infty = K_N$$

In order to do now the character computations for  $H_N^s$ , in general, we need a number of further probabilistic preliminaries. Let us start with the following definition:

DEFINITION 12.10. The Bessel law of level  $s \in \mathbb{N} \cup \{\infty\}$  and parameter t > 0 is

$$b_t^s = p_{t\varepsilon_s}$$

with  $\varepsilon_s$  being the uniform measure on the s-th roots of unity.

Observe that at s = 1, 2 we obtain the Poisson and real Bessel laws:

$$b_t^1 = p_t$$
$$b_t^2 = b_t$$

Another important particular case is  $s = \infty$ , where we obtain a measure which is actually not discrete, denoted as follows:

$$b_t^{\infty} = B_t$$

As a basic result on these laws, generalizing those about  $p_t, b_t$ , we have:

THEOREM 12.11. The generalized Bessel laws  $b_t^s$  have the property

$$b_t^s * b_{t'}^s = b_{t+t'}^s$$

so they form a truncated one-parameter semigroup with respect to convolution.

**PROOF.** This follows indeed from the Fourier transform formulae from Proposition 12.6, because the log of these Fourier transforms are linear in t.

Regarding now the moments, the result here is as follows:

#### 12. REFLECTION GROUPS

THEOREM 12.12. The moments of the Bessel law  $b_t^s$  are the numbers

 $M_k = |P^s(k)|$ 

where  $P^{s}(k)$  is the set of partitions of  $\{1, \ldots, k\}$  satisfying

$$\#\circ = \# \bullet (s)$$

as a weighted sum, in each block.

PROOF. We already know that the formula in the statement holds indeed at s = 1, where  $b_t^1 = p_t$  is the Poisson law of parameter t > 0, and where  $P^1 = P$  is the set of all partitions. At s = 2 we have  $P^2 = P_{even}$ , and the result is elementary as well, and already proved in the above. In general, this follows by doing some combinatorics.

We can go back now to the reflection groups, and we have:

THEOREM 12.13. For the complex reflection group

$$H_N^s = \mathbb{Z}_s \wr S_N$$

we have, with  $N \to \infty$ , the estimate

$$\chi_t \sim b_t^s$$

where  $b_t^s = p_{t\varepsilon_s}$ , with  $\varepsilon_s$  being the uniform measure on the s-th roots of unity.

**PROOF.** We denote by  $\rho$  the uniform measure on the *s*-roots of unity. The best is to proceed in two steps, as follows:

(1) We work out first the case t = 1. Since the limit probability for a random permutation to have exactly k fixed points is  $e^{-1}/k!$ , we get:

$$\lim_{N \to \infty} \text{law}(\chi_1) = e^{-1} \sum_{k=0}^{\infty} \frac{1}{k!} \, \rho^{*k}$$

On the other hand, we get from the definition of the Bessel law  $b_1^s$ :

$$b_1^s = \lim_{N \to \infty} \left( \left( 1 - \frac{1}{N} \right) \delta_0 + \frac{1}{N} \rho \right)^{*N}$$
$$= \lim_{N \to \infty} \sum_{k=0}^N \binom{N}{k} \left( 1 - \frac{1}{N} \right)^{N-k} \frac{1}{N^k} \rho^{*k}$$
$$= e^{-1} \sum_{k=0}^\infty \frac{1}{k!} \rho^{*k}$$

But this gives the assertion for t = 1.

(2) Now in the case t > 0 arbitrary, we can use the same method, by performing the following modifications to the above computation:

$$\lim_{N \to \infty} \operatorname{law}(\chi_t) = e^{-t} \sum_{k=0}^{\infty} \frac{t^k}{k!} \rho^{*k}$$
$$= \lim_{N \to \infty} \left( \left( 1 - \frac{1}{N} \right) \delta_0 + \frac{1}{N} \rho \right)^{*[tN]}$$
$$= b_t^s$$

Thus, we are led to the conclusion in the statement.

Here is now some more theory for the Bessel laws. First, it is convenient to introduce as well "modified" versions of the Bessel laws, as follows:

DEFINITION 12.14. The Bessel and modified Bessel laws are given by

$$b_t^s = \log\left(\sum_{k=1}^s w^k a_k\right)$$
$$\tilde{b}_t^s = \log\left(\sum_{k=1}^s w^k a_k\right)^s$$

where  $a_1, \ldots, a_s$  are independent random variables, each of them following the Poisson law of parameter t/s, and  $w = e^{2\pi i/s}$ .

As a first remark, at s = 1 we get the Poisson law of parameter t:

$$b_t^1 = \tilde{b}_t^1 = e^{-t} \sum_{r=0}^\infty \frac{t^r}{r!} \,\delta_r$$

Consider now the level s exponential function, given by:

$$\exp_s z = \sum_{k=0}^{\infty} \frac{z^{sk}}{(sk)!}$$

We have the following formula, in terms of  $w = e^{2\pi i/s}$ :

$$\exp_s z = \frac{1}{s} \sum_{k=1}^s \exp(w^k z)$$

Observe that we have the following formulae:

$$exp_1 = exp$$
$$exp_2 = \cosh$$

We have the following result, regarding both the plain and modified Bessel laws, which is a more explicit version of Proposition 12.6 above, for the Bessel laws:

## 12. REFLECTION GROUPS

THEOREM 12.15. The Fourier transform of  $b_t^s$  is given by

$$\log F_t^s(z) = t \left( \exp_s z - 1 \right)$$

so in particular the measures  $b_t^s$  are additive with respect to t.

PROOF. Consider, as in Definition 12.14, the following variable:

$$a = \sum_{k=1}^{s} w^k a_k$$

We have then the following computation:

$$\log F_a(z) = \sum_{k=1}^s \log F_{a_k}(w^k z)$$
$$= \sum_{k=1}^s \frac{t}{s} \left( \exp(w^k z) - 1 \right)$$

This gives the following formula:

$$\log F_a(z) = t \left( \left( \frac{1}{s} \sum_{k=1}^s \exp(w^k z) \right) - 1 \right)$$
$$= t \left( \exp_s(z) - 1 \right)$$

Now since  $b_t^s$  is the law of a, this gives the formula in the statement.

Let us study now the densities of  $b_t^s, \tilde{b}_t^s$ . We have here the following result: THEOREM 12.16. We have the formulae

$$b_t^s = e^{-t} \sum_{p_1=0}^{\infty} \dots \sum_{p_s=0}^{\infty} \frac{1}{p_1! \dots p_s!} \left(\frac{t}{s}\right)^{p_1+\dots+p_s} \delta\left(\sum_{k=1}^s w^k p_k\right)$$
$$\tilde{b}_t^s = e^{-t} \sum_{p_1=0}^{\infty} \dots \sum_{p_s=0}^{\infty} \frac{1}{p_1! \dots p_s!} \left(\frac{t}{s}\right)^{p_1+\dots+p_s} \delta\left(\sum_{k=1}^s w^k p_k\right)^s$$

where  $w = e^{2\pi i/s}$ , and the  $\delta$  symbol is a Dirac mass.

**PROOF.** It is enough to prove the formula for  $b_t^s$ . For this purpose, we compute the Fourier transform of the measure on the right. This is given by:

$$F(z) = e^{-t} \sum_{p_1=0}^{\infty} \dots \sum_{p_s=0}^{\infty} \frac{1}{p_1! \dots p_s!} \left(\frac{t}{s}\right)^{p_1+\dots+p_s} F\delta\left(\sum_{k=1}^s w^k p_k\right)(z)$$
  
=  $e^{-t} \sum_{p_1=0}^{\infty} \dots \sum_{p_s=0}^{\infty} \frac{1}{p_1! \dots p_s!} \left(\frac{t}{s}\right)^{p_1+\dots+p_s} \exp\left(\sum_{k=1}^s w^k p_k z\right)$   
=  $e^{-t} \sum_{r=0}^{\infty} \left(\frac{t}{s}\right)^r \sum_{\Sigma p_i=r} \frac{\exp\left(\sum_{k=1}^s w^k p_k z\right)}{p_1! \dots p_s!}$ 

We multiply by  $e^t$ , and we compute the derivative with respect to t:

$$(e^{t}F(z))' = \sum_{r=1}^{\infty} \frac{r}{s} \left(\frac{t}{s}\right)^{r-1} \sum_{\Sigma p_{i}=r} \frac{\exp\left(\sum_{k=1}^{s} w^{k} p_{k} z\right)}{p_{1}! \dots p_{s}!}$$
  
$$= \frac{1}{s} \sum_{r=1}^{\infty} \left(\frac{t}{s}\right)^{r-1} \sum_{\Sigma p_{i}=r} \left(\sum_{l=1}^{s} p_{l}\right) \frac{\exp\left(\sum_{k=1}^{s} w^{k} p_{k} z\right)}{p_{1}! \dots p_{s}!}$$
  
$$= \frac{1}{s} \sum_{r=1}^{\infty} \left(\frac{t}{s}\right)^{r-1} \sum_{\Sigma p_{i}=r} \sum_{l=1}^{s} \frac{\exp\left(\sum_{k=1}^{s} w^{k} p_{k} z\right)}{p_{1}! \dots p_{l-1}! (p_{l}-1)! p_{l+1}! \dots p_{s}!}$$

By using the variable u = r - 1, we get:

$$(e^t F(z))' = \frac{1}{s} \sum_{u=0}^{\infty} \left(\frac{t}{s}\right)^u \sum_{\Sigma q_i=u} \sum_{l=1}^s \frac{\exp\left(w^l z + \sum_{k=1}^s w^k q_k z\right)}{q_1! \dots q_s!}$$
$$= \left(\frac{1}{s} \sum_{l=1}^s \exp(w^l z)\right) \left(\sum_{u=0}^{\infty} \left(\frac{t}{s}\right)^u \sum_{\Sigma q_i=u} \frac{\exp\left(\sum_{k=1}^s w^k q_k z\right)}{q_1! \dots q_s!}\right)$$
$$= (\exp_s z)(e^t F(z))$$

On the other hand, consider the following function:

$$\Phi(t) = \exp(t \exp_s z)$$

This function satisfies as well the equation found above, namely:

$$\Phi'(t) = (\exp_s z)\Phi(t)$$

Thus, we have the following equality of functions:

$$e^t F(z) = \Phi(t)$$

But this gives the following formula:

$$\log F = \log(e^{-t} \exp(t \exp_s z))$$
$$= \log(\exp(t(\exp_s z - 1)))$$
$$= t(\exp_s z - 1)$$

Thus, we obtain the formulae in the statement.

Let us discuss now the  $s = \infty$  particular case of all the above, which will be of interest in what follows, along with the particular cases s = 1, 2.

First, we have the following result, at the algebraic level:

THEOREM 12.17. The full complex reflection group, denoted

$$K_N \subset U_N$$

and formed of the permutation-type matrices with numbers in  $\mathbb{T}$  as nonzero entries,

$$K_N = M_N(\mathbb{T} \cup \{0\}) \cap U_N$$

has a wreath product decomposition, as follows,

 $K_N = \mathbb{T} \wr S_N$ 

with  $S_N$  acting on  $\mathbb{T}^N$  in the standard way, by permuting the factors.

PROOF. This is something that we already know from the above, at the  $s = \infty$  particular case of the results established for the complex reflection groups  $H_N^s$ .

At the probabilistic level, we have the following result:

THEOREM 12.18. For the full reflection group  $K_N \subset U_N$ , the law of the variable

$$\chi_t = \sum_{i=1}^{[tN]} g_{ii}$$

is in the  $N \to \infty$  limit the purely complex Bessel law,

$$B_t = p_{t\varepsilon}$$

where  $\varepsilon$  is the uniform measure on  $\mathbb{T}$ .

PROOF. Once again, this is something that we already know from the above, at the  $s = \infty$  particular case of the results established for the complex reflection groups  $H_N^s$ .  $\Box$ 

There are many other interesting probability formulae, regarding the real and complex Bessel laws  $b_t$  and  $B_t$ , and we refer here to the literature on the subject.

284

Let us end this presentation with some philosophical considerations, in connection with abstract probability theory. We have 4 main limiting results in probability, namely discrete and continuous, and real and complex, which are as follows:



We have seen that the limiting laws in these main limiting theorems are the real and complex Gaussian and Bessel laws, which are as follows:



Moreover, we have seen that at the level of the moments, these come from certain collections of partitions, as follows:



Finally, we have seen that there are some Lie groups behind all this, namely the basic real and complex rotation and reflection groups, as follows:



All this is quite interesting, and we will be back to this, with more explanations, in chapters 13-16 below, by using some advanced representation theory tools.

Finally, in order for our discussion to be complete, we still have to talk about the general series of complex reflection groups  $H_N^{sd}$ . Here the determinant does not really contribute, in the  $N \to \infty$  limit, and we obtain the same asymptotic laws as for the

#### 12. REFLECTION GROUPS

groups  $H_N^s$ . This is something that we have already seen in chapter 11 above, in relation with the alternating group  $A_N \subset S_N$ , and the proof in general is similar.

As before with other more advanced questions, we will leave all this for further discussion in chapters 13-16 below, after developing more powerful tools for dealing with such questions, and more specifically, advanced representation theory tools.

## 12d. Wigner laws

In the continuous group case now, as a continuation of the above investigations, an interesting input comes from the computations from chapter 6 above.

In order to discuss all this, let us first recall some useful formulae from chapter 6. One of the key results there, which is very useful in practice, was as follows:

**PROPOSITION 12.19.** We have the following formula,

$$\int_0^{\pi/2} \cos^p t \sin^q t \, dt = \left(\frac{\pi}{2}\right)^{\varepsilon(p)\varepsilon(q)} \frac{p!!q!!}{(p+q+1)!!}$$

where  $\varepsilon(p) = 1$  if p is even, and  $\varepsilon(p) = 0$  if p is odd, and where

$$m!! = (m-1)(m-3)(m-5)\dots$$

with the product ending at 2 if m is odd, and ending at 1 if m is even.

**PROOF.** This is something that follows by partial integration, and then a double recurrence on  $p, q \in \mathbb{N}$ , worked out in chapter 6 above.

More generally now, we can in fact compute the polynomial integrals over the unit sphere in arbitrary dimensions, the result here being as follows:

THEOREM 12.20. The polynomial integrals over the unit sphere  $S_{\mathbb{R}}^{N-1} \subset \mathbb{R}^N$ , with respect to the normalized, mass 1 measure, are given by the following formula,

$$\int_{S_{\mathbb{R}}^{N-1}} x_1^{k_1} \dots x_N^{k_N} \, dx = \frac{(N-1)!!k_1!!\dots k_N!!}{(N+\Sigma k_i - 1)!!}$$

valid when all exponents  $k_i$  are even. If an exponent is odd, the integral vanishes.

**PROOF.** As before this is something that we know from chapter 6, the idea being that the N = 2 case is solved by Proposition 12.19 above, and that the general case,  $N \in \mathbb{N}$ , follows from this, by using spherical coordinates and the Fubini theorem.

As an application of the above formula, we can investigate the hyperspherical laws and their asymptotics, and we have the following result:

**THEOREM** 12.21. The moments of the hyperspherical variables are

$$\int_{S^{N-1}_{\mathbb{R}}} x_i^k dx = \frac{(N-1)!!k!!}{(N+k-1)!!}$$

and the normalized hyperspherical variables

$$y_i = \frac{x_i}{\sqrt{N}}$$

become normal and independent with  $N \to \infty$ .

**PROOF.** We have two things to be proved, the idea being as follows:

(1) The formula in the statement follows from the general integration formula over the sphere, established above, which is as follows:

$$\int_{S_{\mathbb{R}}^{N-1}} x_{i_1} \dots x_{i_k} \, dx = \frac{(N-1)!! l_1!! \dots l_N!!}{(N+\sum l_i-1)!!}$$

Indeed, with  $i_1 = \ldots = i_k = i$ , we obtain from this:

$$\int_{S_{\mathbb{R}}^{N-1}} x_i^k dx = \frac{(N-1)!!k!!}{(N+k-1)!!}$$

Now observe that with  $N \to \infty$  we have the following estimate:

$$\int_{S_{\mathbb{R}}^{N-1}} x_i^k dx = \frac{(N-1)!!}{(N+k-1)!!} \times k!!$$
  
$$\simeq N^{k/2} k!!$$
  
$$= N^{k/2} M_k(g_1)$$

Thus, the variables  $y_i = \frac{x_i}{\sqrt{N}}$  become normal with  $N \to \infty$ .

(2) As for the asymptotic independence result, this is standard as well, once again by using Theorem 12.20, for computing mixed moments, and taking the  $N \to \infty$  limit.  $\Box$ 

We can talk as well about rotations, as follows:

THEOREM 12.22. We have the integration formula

$$\int_{O_N} U_{ij}^k dU = \frac{(N-1)!!k!!}{(N+k-1)!!}$$

and the normalized coordinates on  $O_N$ , constructed as follows,

$$V_{ij} = \frac{U_{ij}}{\sqrt{N}}$$

become normal and independent with  $N \to \infty$ .
#### 12. REFLECTION GROUPS

PROOF. We use the well-known fact that we have an embedding as follows, for any i, which makes correspond the respective integration functionals:

$$C(S_{\mathbb{R}}^{N-1}) \subset C(O_N)$$
$$x_i \to U_{1i}$$

With this identification made, the result follows from Theorem 12.21.

We have similar results in the unitary case. First, we have:

THEOREM 12.23. We have the following integration formula over the complex sphere  $S_{\mathbb{C}}^{N-1} \subset \mathbb{R}^N$ , with respect to the normalized measure,

$$\int_{S_{\mathbb{C}}^{N-1}} |z_1|^{2l_1} \dots |z_N|^{2l_N} dz = 4^{\sum l_i} \frac{(2N-1)! l_1! \dots l_n!}{(2N+\sum l_i-1)!}$$

valid for any exponents  $l_i \in \mathbb{N}$ . As for the other polynomial integrals in  $z_1, \ldots, z_N$  and their conjugates  $\overline{z}_1, \ldots, \overline{z}_N$ , these all vanish.

**PROOF.** As before, this is something that we know from chapter 6 above, and which can be proved either directly, or by using the formula in Theorem 12.20 above.  $\Box$ 

We can talk about complex hyperspherical laws, and we have:

THEOREM 12.24. The rescaled coordinates on the complex sphere  $S_{\mathbb{C}}^{N-1}$ ,

$$w_i = \frac{z_i}{\sqrt{N}}$$

become complex Gaussian and independent with  $N \to \infty$ .

**PROOF.** This follows indeed by using Theorem 12.22 and Theorem 12.23.

In relation now with Lie groups, the result that we obtain is as follows:

THEOREM 12.25. For the unitary group  $U_N$ , the normalized coordinates

$$V_{ij} = \frac{U_{ij}}{\sqrt{N}}$$

become complex Gaussian and independent with  $N \to \infty$ .

PROOF. We use the well-known fact that we have an embedding as follows, for any i, which makes correspond the respective integration functionals:

$$C(S_{\mathbb{C}}^{N-1}) \subset C(U_N)$$
$$x_i \to U_{1i}$$

With this identification made, the result follows from Theorem 12.21.

Observe now that the above results can be reformulated in terms of the truncated characters introduced in chapter 11. Let us recall indeed from there that we have:

DEFINITION 12.26. Given a closed subgroup  $G \subset U_N$ , the function

$$\chi: G \to \mathbb{C}$$

$$\chi_t(g) = \sum_{i=1}^{k} g_{ii}$$

is called main truncated character of G, of parameter  $t \in (0, 1]$ .

We refer to chapter 11 above for more on this notion. Also, we will be back to all this in chapters 13-16 below, with more details about all this, and motivations.

In connection with the present considerations, the point is that with the above notion in hand, our above results reformulate as follows:

THEOREM 12.27. For the orthogonal and unitary groups  $O_N, U_N$ , the rescalings

$$\chi = \frac{\chi_{1/N}}{\sqrt{N}}$$

become respectively real and complex Gaussian, in the  $N \to \infty$  limit.

**PROOF.** According to our conventions, given a closed subgroup  $G \subset U_N$ , the main character truncated at t = 1/N is simply the first coordinate:

$$\chi_{1/N}(g) = g_{11}$$

With this remark made, the conclusions from the statement follow from the computations performed above, for the laws of coordinates on  $O_N, U_N$ .

It is possible to get beyond such results, by using advanced representation theory methods, with full results about all the truncated characters, and in particular about the main characters. We will be back to this in chapters 13-16 below.

Also, it is possible to compute as well the laws of individual coordinates for some of the remaining continuous groups. To be more precise, it is possible to work out results for the bistochastic groups  $B_N, C_N$ , as well as for the symplectic groups  $Sp_N$ , the computations being not very complicated. However, we prefer here to defer the discussion to chapters 13-16 below, after developing some appropriate tools, for dealing with such questions.

As a last topic now, let us discuss the case where N is fixed. Things are quite complicated here, and as a main goal, we would like to find the law of the main character for our favorite rotation groups, namely  $SU_2$  and  $SO_3$ . In order to do so, we will need some combinatorial preliminaries. We first have the following well-known result:

THEOREM 12.28. The Catalan numbers, which are by definition given by

 $C_k = |NC_2(2k)|$ 

satisfy the following recurrence formula,

$$C_{k+1} = \sum_{a+b=k} C_a C_b$$

and their generating series, given by

$$f(z) = \sum_{k \ge 0} C_k z^k$$

satisfies the following degree 2 equation,

$$zf^2 - f + 1 = 0$$

and we have the following explicit formula for these numbers:

$$C_k = \frac{1}{k+1} \binom{2k}{k}$$

**PROOF.** We must count the noncrossing pairings of  $\{1, \ldots, 2k\}$ . But such a pairing appears by pairing 1 to an odd number, 2a + 1, and then inserting a noncrossing pairing of  $\{2, \ldots, 2a\}$ , and a noncrossing pairing of  $\{2a + 2, \ldots, 2l\}$ . We conclude from this that we have the following recurrence formula for the Catalan numbers:

$$C_k = \sum_{a+b=k-1} C_a C_b$$

In terms of the generating series, the above recurrence gives:

$$zf^{2} = \sum_{a,b\geq 0} C_{a}C_{b}z^{a+b+1}$$
$$= \sum_{k\geq 1} \sum_{a+b=k-1} C_{a}C_{b}z^{k}$$
$$= \sum_{k\geq 1} C_{k}z^{k}$$
$$= f - 1$$

Thus f satisfies the following degree 2 equation:

$$zf^2 - f + 1 = 0$$

By choosing the solution which is bounded at z = 0, we obtain:

$$f(z) = \frac{1 - \sqrt{1 - 4z}}{2z}$$

By using now the Taylor formula for  $\sqrt{x}$ , we obtain the following formula:

$$f(z) = \sum_{k \ge 0} \frac{1}{k+1} \binom{2k}{k} z^k$$

It follows that the Catalan numbers are given by the formula the statement.

The Catalan numbers are central objects in probability as well, and we have the following key result here, complementing the formulae from Theorem 12.28:

THEOREM 12.29. The normalized Wigner semicircle law, which is by definition

$$\gamma_1 = \frac{1}{2\pi}\sqrt{4 - x^2}dx$$

has the Catalan numbers as even moments. As for the odd moments, these all vanish.

**PROOF.** The even moments of the Wigner law can be computed with the change of variable  $x = 2 \cos t$ , and we are led to the following formula:

$$M_{2k} = \frac{1}{\pi} \int_{0}^{2} \sqrt{4 - x^{2}} x^{2k} dx$$
  

$$= \frac{1}{\pi} \int_{0}^{\pi/2} \sqrt{4 - 4\cos^{2}t} (2\cos t)^{2k} 2\sin t dt$$
  

$$= \frac{4^{k+1}}{\pi} \int_{0}^{\pi/2} \cos^{2k}t \sin^{2}t dt$$
  

$$= \frac{4^{k+1}}{\pi} \cdot \frac{\pi}{2} \cdot \frac{(2k)!!2!!}{(2k+3)!!}$$
  

$$= 2 \cdot 4^{k} \cdot \frac{(2k)!/2^{k}k!}{2^{k+1}(k+1)!}$$
  

$$= C_{k}$$

As for the odd moments, these all vanish, because the density of  $\gamma_1$  is an even function. Thus, we are led to the conclusion in the statement.

We can now formulate our result regarding  $SU_2$ , as follows:

THEOREM 12.30. The main character of  $SU_2$ , given by

$$\chi \begin{pmatrix} a & b \\ -\bar{b} & \bar{a} \end{pmatrix} = 2Re(a)$$

follows a Wigner semicircle law  $\gamma_1$ .

**PROOF.** This follows by identifying  $SU_2$  with the sphere  $S^3_{\mathbb{R}} \subset \mathbb{R}^4$ , and the uniform measure on  $SU_2$  with the uniform measure on this sphere.

#### 12. REFLECTION GROUPS

Indeed, in real notation for the standard parametrization of  $SU_2$ , from chapter 9 above, we have the following formula, for the main character of  $SU_2$ :

$$\chi \begin{pmatrix} x+iy & z+it \\ -z+it & x-iy \end{pmatrix} = 2x$$

We are therefore left with computing the law of the following variable:

$$x \in C(S^3_{\mathbb{R}})$$

But for this purpose, we can use the moment method. Indeed, by using the trigonometric integral formulae from chapter 6 above, we obtain:

$$\int_{S_{\mathbb{R}}^{3}} x^{2k} = \frac{3!!(2k)!!}{(2k+3)!!}$$

$$= 2 \cdot \frac{3 \cdot 5 \cdot 7 \dots (2k-1)}{2 \cdot 4 \cdot 6 \dots (2k+2)}$$

$$= 2 \cdot \frac{(2k)!}{2^{k} k! 2^{k+1} (k+1)!}$$

$$= \frac{1}{4^{k}} \cdot \frac{1}{k+1} \binom{2k}{k}$$

$$= \frac{C_{k}}{4^{k}}$$

Thus the variable  $2x \in C(S^3_{\mathbb{R}})$  has the Catalan numbers as even moments, and so by Theorem 12.28 its distribution is the Wigner semicircle law  $\gamma_1$ , as claimed.

In order to do the computation for  $SO_3$ , we will need some more probabilistic preliminaries, which are standard random matrix theory material. Let us start with:

**PROPOSITION** 12.31. We have a bijection between noncrossing partitions and pairings

$$NC(k) \simeq NC_2(2k)$$

constructed as follows:

- (1) The application  $NC(k) \rightarrow NC_2(2k)$  is the "fattening" one, obtained by doubling all the legs, and doubling all the strings as well.
- (2) Its inverse  $NC_2(2k) \rightarrow NC(k)$  is the "shrinking" application, obtained by collapsing pairs of consecutive neighbors.

**PROOF.** The fact that the two operations in the statement are indeed inverse to each other is clear, by computing the corresponding two compositions, with the remark that the construction of the fattening operation requires the partitions to be noncrossing.  $\Box$ 

As a consequence of the above result, we have a new look on the Catalan numbers, which is more adapted to our present  $SO_3$  considerations, as follows:

**PROPOSITION 12.32.** The Catalan numbers  $C_k = |NC_2(2k)|$  appear as well as

$$C_k = |NC(k)|$$

where NC(k) is the set of all noncrossing partitions of  $\{1, \ldots, k\}$ .

**PROOF.** This follows indeed from Proposition 12.31 above.

Let us formulate now the following definition:

DEFINITION 12.33. The standard Marchenko-Pastur law  $\pi_1$  is given by:

$$f \sim \gamma_1 \implies f^2 \sim \pi_1$$

That is,  $\pi_1$  is the law of the square of a variable following the semicircle law  $\gamma_1$ .

Here the fact that  $\pi_1$  is indeed well-defined comes from the fact that a measure is uniquely determined by its moments. More explicitly now, we have:

**PROPOSITION 12.34.** The density of the Marchenko-Pastur law is

$$\pi_1 = \frac{1}{2\pi} \sqrt{4x^{-1} - 1} \, dx$$

and the moments of this measure are the Catalan numbers.

PROOF. There are several proofs here, either by using Definition 12.33, or by Stieltjes inversion, of just by cheating. Whis this latter method, the point is that the moments of the law in the statement can be computed with the following change of variables:

$$x = 4\cos^2 t$$

To be more precise, the moments of the law in the statement can be computed with the change of variable  $x = 4 \cos^2 t$ , and we are led to the following formula:

$$M_{k} = \frac{1}{2\pi} \int_{0}^{4} \sqrt{4x^{-1} - 1} x^{k} dx$$

$$= \frac{1}{2\pi} \int_{0}^{\pi/2} \frac{\sin t}{\cos t} \cdot (4\cos^{2} t)^{k} \cdot 2\cos t \sin t \, dt$$

$$= \frac{4^{k+1}}{\pi} \int_{0}^{\pi/2} \cos^{2k} t \sin^{2} t \, dt$$

$$= \frac{4^{k+1}}{\pi} \cdot \frac{\pi}{2} \cdot \frac{(2k)!!2!!}{(2k+3)!!}$$

$$= 2 \cdot 4^{k} \cdot \frac{(2k)!/2^{k}k!}{2^{k+1}(k+1)!}$$

$$= C_{k}$$

Thus, we are led to the conclusion in the statement.

#### 12. REFLECTION GROUPS

We can do now the character computation for  $SO_3$ , as follows:

THEOREM 12.35. The main character of  $SO_3$ , modified by adding 1 to it, given in standard Euler-Rodrigues coordinates by

$$\chi = 3x^2 - y^2 - z^2 - t^2$$

follows a squared semicircle law, or Marchenko-Pastur law  $\pi_1$ .

PROOF. This follows by using the canonical quotient map  $SU_2 \rightarrow SO_3$ , and the result for  $SU_2$  from Theorem 12.30. To be more precise, let us recall from chapter 9 above that the elements of  $SU_2$  can be parametrized as follows:

$$U = \begin{pmatrix} x + iy & z + it \\ -z + it & x - iy \end{pmatrix}$$

As for the elements of  $SO_3$ , these can be parametrized as follows:

$$V = \begin{pmatrix} x^2 + y^2 - z^2 - t^2 & 2(yz - xt) & 2(xz + yt) \\ 2(xt + yz) & x^2 + z^2 - y^2 - t^2 & 2(zt - xy) \\ 2(yt - xz) & 2(xy + zt) & x^2 + t^2 - y^2 - z^2 \end{pmatrix}$$

The point now is that, by using these formulae, in the context of Theorem 12.30, the main character of  $SO_3$  is then given by:

$$\chi = 4Re(a)^2$$

Now recall from the proof of Theorem 12.30 above that we have:

$$2Re(a) \sim \gamma_1$$

On the other hand, a quick comparison between the moment formulae for the Wigner and Marchenko-Pastur laws, which are very similar, shows that we have:

$$f \sim \gamma_1 \implies f^2 \sim \pi_1$$

Thus, with f = 2Re(a), we obtain the result in the statement.

As an interesting question now, appearing from the above, and which is quite philosophical, we have the problem of understanding how the Wigner and Marchenko-Pastur laws  $\gamma_t$ ,  $\pi_t$  fit in regards with the main limiting laws from classical probability.

The answer here is quite tricky, the idea being that, with a suitable formalism for freeness,  $\gamma_t, \pi_t$  can be thought of as being "free analogues" of the Gaussian and Poisson laws  $g_t, p_t$ . This is something quite subtle, requiring some further knowledge, and we will be back to this in chapters 13-16 below, when doing representation theory.

#### 12E. EXERCISES

#### 12e. Exercises

There has been a lot of technical material in this chapter, and technical as well will be most of our exercises. First, we have:

EXERCISE 12.36. Work out an alternative proof for the main result regarding the truncated characters of the hyperoctahedral group  $H_N$ , namely

$$\chi_t \sim e^{-t} \sum_{k=-\infty}^{\infty} \delta_k \sum_{p=0}^{\infty} \frac{(t/2)^{|k|+2p}}{(|k|+p)!p!}$$

with  $N \to \infty$ , by working our first an explicit formula for the polynomials integrals over  $H_N$ , and then using that for computing the laws of truncated characters.

The idea here is of course that of adapting the computation that we have for the symmetric group  $S_N$ , from the previous chapter.

As a second question now, technical as well, we have:

EXERCISE 12.37. Work out all the details for the moment formula for the Bessel laws,

$$M_k = |P^s(k)|$$

where  $P^{s}(k)$  are the partitions satisfying the formula

$$\#\circ = \# \bullet (s)$$

as a weighted sum, in each block.

This is something that we briefly discussed in the above, and the problem is now that of working out all the details, first as  $s = 1, 2, \infty$ , and then in general.

In the same spirit, we have the following exercise:

EXERCISE 12.38. Work out all the details for the truncated character formula for  $H_N^s$ ,

 $\chi_t \sim b_t^s$ 

where  $b_t^s = p_{t\varepsilon_s}$ , with  $\varepsilon_s$  being the uniform measure on the s-th roots of unity.

As before, this is something that we briefly discussed in the above, and the problem is now that of working out all the details, first as  $s = 1, 2, \infty$ , and then in general.

As a more complicated exercise now, fully closing the discussion for the complex reflection groups, we have:

EXERCISE 12.39. Show that the passage from  $H_N^s$  to  $H_N^{sd}$  does not change the asymptotic laws of the truncated characters.

#### 12. REFLECTION GROUPS

This is something that we discussed in the previous chapter, in a particular case, namely for the passage from the symmetric group  $S_N$  to the alternating group  $A_N$ . The problem is that of having this done in general, by using a similar method.

In relation now with the continuous groups, we first have:

EXERCISE 12.40. Compute the asymptotic laws of characters and coordinates for the bistochastic groups  $B_N$  and  $C_N$ , as well as for the symplectic group  $Sp_N \subset U_N$ .

These computations are all quite standard, the idea being that, by using the various group theory considerations from chapter 10 above, the computation for  $B_N$  is quite similar to that from  $O_{N-1}$ , the computation for  $C_N$  is quite similar to that from  $U_{N-1}$ , and the computation for  $S_{p_N}$  is quite similar to that from  $O_{N-1}$ .

As another exercise in the continuous case, now at fixed  $N \in \mathbb{N}$ , we have:

EXERCISE 12.41. Compute the character laws for the groups  $O_1$ ,  $SO_1$ , then for the groups  $U_1$ ,  $SU_1$ , and then for the groups  $O_2$ ,  $SO_2$ .

As before with the previous exercise, the computations here are quite standard. In fact, the more difficult questions of this type concern the next groups in the above series, namely  $SU_2$  and  $SO_3$ , which were discussed in the above.

In relation now with the Wigner and Marchenko-Pastur laws, we have:

EXERCISE 12.42. Work out all the combinatorics and calculus details in relation with the Wigner and Marchenko-Pastur laws, and their moments, the Catalan numbers.

This is a very instructive exercise, with lots of nice combinatorics involved. Most of this combinatorics was actually already discussed in the above.

Finally, as a more advanced exercise, of research flavor, we have:

EXERCISE 12.43. Look up the Wigner and Marchenko-Pastur laws, given by

$$\gamma_1 = \frac{1}{2\pi} \sqrt{4 - x^2} dx$$
$$\pi_1 = \frac{1}{2\pi} \sqrt{4x^{-1} - 1} dx$$

and write down a brief account of what you found, and understood.

Here the solution is not unique, the Wigner and Marchenko-Pastur laws appearing in random matrix theory, or free probability, or both. All nice mathematics, enjoy.

# Part IV

Haar integration

And the band plays Waltzing Matilda And the old men still answer the call But year after year, their numbers get fewer Someday, no one will march there at all

## CHAPTER 13

# Representations

#### 13a. Basic theory

We have seen so far that some algebraic and probabilistic theory for the finite subgroups  $G \subset U_N$ , ranging from elementary to quite advanced, can be developed. We have seen as well a few computations for the continuous compact subgroups  $G \subset U_N$ .

In what follows we develop some systematic theory for the arbitrary closed subgroups  $G \subset U_N$ , covering both the finite and the infinite case. The main examples that we have in mind, and the questions that we would like to solve for them, are as follows:

- (1) The orthogonal and unitary groups  $O_N, U_N$ . Here we would like to have an integration formula, and results about character laws, in the  $N \to \infty$  limit.
- (2) Various versions of  $O_N, U_N$ , such as the bistochastic groups  $B_N, C_N$ , or the symplectic groups  $Sp_N$ , with similar questions to be solved.
- (3) The reflection groups  $H_N^{sd} \subset U_N$ , with results about characters extending, or at least putting in a more conceptual framework, what we already have.

There is a lot of theory to be developed, and we will do this gradually. To be more precise, in this chapter and in the next one we will work out algebraic aspects, and then in the chapter afterwards and in the last one we will use these algebraic techniques, in order to work out probabilistic results, and in particular to answer the above questions.

As before, the main notion that we will be interested in is that of a representation:

DEFINITION 13.1. A representation of a compact group G is a continuous group morphism, which can be faithful or not, into a unitary group:

$$u: G \to U_N$$

The character of such a representation is the function  $\chi: G \to \mathbb{C}$  given by

$$g \to Tr(u_q)$$

where Tr is the usual trace of the  $N \times N$  matrices,  $Tr(M) = \sum_{i} M_{ii}$ .

As a basic example here, for any compact group we always have available the trivial representation, which is by definition as follows:

$$u: G \to U_1 \quad , \quad g \to (1)$$

In the same spirit, we have as well the null representation, which is as follows:

$$u: G \to U_1 \quad , \quad g \to (0)$$

At the level of non-trivial examples now, most of the compact groups that we met so far, finite or continuous, naturally appear as closed subgroups  $G \subset U_N$ . In this case, the embedding  $G \subset U_N$  is of course a representation, called fundamental representation:

$$u: G \subset U_N \quad , \quad g \to g$$

In this situation, there are many other representations of G, which are equally interesting. For instance, we can define the representation conjugate to u, as being:

$$\bar{u}: G \subset U_N \quad , \quad g \to \bar{g}$$

In order to clarify all this, and see which representations are available, let us first discuss the various operations on the representations. The result here is as follows:

**PROPOSITION 13.2.** The representations of a given compact group G are subject to the following operations:

(1) Making sums. Given a N-dimensional representation u and a M-dimensional representation v, their sum is the N + M-dimensional representation:

$$u + v = diag(u, v)$$

(2) Making products. Given a N-dimensional representation u and a M-dimensional representation v, their tensor product is the NM-dimensional representation:

$$(u \otimes v)_{ia,jb} = u_{ij}v_{ab}$$

(3) Taking conjugates. Given a N-dimensional representation u, its conjugate is following the N-dimensional representation:

$$(\bar{u})_{ij} = \bar{u}_{ij}$$

(4) Spinning by unitaries. Given a N-dimensional representation u, and a unitary  $V \in U_N$ , we can spin u by this unitary V, as follows:

$$u \to V u V^*$$

**PROOF.** The fact that the operations in the statement are indeed well-defined, among maps from G to unitary groups, can be checked as follows:

(1) This follows from the trivial fact that if  $g \in U_N$  and  $h \in U_M$  are two unitaries, then their diagonal sum is a unitary too, as follows:

$$\begin{pmatrix} g & 0\\ 0 & h \end{pmatrix} \in U_{N+M}$$

(2) This follows from the fact, which is standard too, that if  $g \in U_N$  and  $h \in U_M$  are two unitaries, then their tensor product is a unitary too,  $g \otimes h \in U_{NM}$ . Given unitaries g, h, define indeed a matrix  $g \otimes h$  by the formula in the statement, namely:

$$(g \otimes h)_{ia,jb} = g_{ij}h_{ab}$$

This matrix is then a unitary too, due to the following computation:

$$\begin{split} [(g \otimes h)(g \otimes h)^*]_{ia,jb} &= \sum_{kc} (g \otimes h)_{ia,kc} ((g \otimes h)^*)_{kc,jb} \\ &= \sum_{kc} (g \otimes h)_{ia,kc} \overline{(g \otimes h)_{jb,kc}} \\ &= \sum_{kc} g_{ik} h_{ac} \overline{g}_{jk} \overline{h}_{bc} \\ &= \sum_{k} g_{ik} \overline{g}_{jk} \sum_{c} h_{ac} \overline{h}_{bc} \\ &= \delta_{ij} \delta_{ab} \\ &= \delta_{ia,jb} \end{split}$$

(3) This simply follows from the fact that if  $g \in U_N$  is unitary, then so is its complex conjugate,  $\bar{g} \in U_N$ , and this due to the following formula, obtained by conjugating:

$$g^* = g^{-1} \implies g^t = \bar{g}^{-1}$$

(4) This is clear as well, because if  $g \in U_N$  is unitary, and  $V \in U_N$  is another unitary, then we can spin g by this unitary, and we obtain a unitary as follows:

$$VgV^* \in U_N$$

Thus, our operations are well-defined, and this leads to the above conclusions.  $\Box$ 

As a consequence of the above result, assuming that we have a non-trivial representation  $u: G \to U_N$ , we can construct a whole family of representations of G, by using the above operations. We will be back to this in a moment, with some theory and details.

In relation now with characters, we have the following result:

**PROPOSITION 13.3.** We have the following formulae, regarding characters

$$\chi_{u+v} = \chi_u + \chi_v$$
$$\chi_{u\otimes v} = \chi_u \chi_v$$
$$\chi_{\bar{u}} = \bar{\chi}_u$$
$$\chi_{VuV^*} = \chi_u$$

in relation with the basic operations for the representations.

**PROOF.** All these assertions are elementary, by using the following well-known trace formulae, valid for any two square matrices g, h, and any unitary V:

$$Tr(diag(g,h)) = Tr(g) + Tr(h)$$
$$Tr(g \otimes h) = Tr(g)Tr(h)$$
$$Tr(\overline{g}) = \overline{Tr(g)}$$
$$Tr(VgV^*) = Tr(g)$$

To be more precise, the first formula is clear from definitions. Regarding now the second formula, the computation here is immediate too, as follows:

$$Tr(g \otimes h) = \sum_{ia} (g \otimes h)_{ia,ia}$$
$$= \sum_{ia} g_{ii} h_{aa}$$
$$= \sum_{i} g_{ii} \sum_{a} h_{aa}$$
$$= Tr(g)Tr(h)$$

Regarding nor the third formula, this is clear from definitions, by conjugating. Finally, regarding the fourth formula, this can be established as follows:

$$Tr(VgV^*) = Tr(V \cdot gV^*)$$
  
=  $Tr(gV^* \cdot V)$   
=  $Tr(g)$ 

Thus, we are led to the conclusions in the statement.

Assume now that we are given a closed subgroup  $G \subset U_N$ . By using the above operations, we can construct a whole family of representations of G, as follows:

DEFINITION 13.4. Given a closed subgroup  $G \subset U_N$ , its Peter-Weyl representations are the tensor products between the fundamental representation and its conjugate:

$$u: G \subset U_N$$
$$\bar{u}: G \subset U_N$$

We denote these tensor products  $u^{\otimes k}$ , with  $k = \circ \bullet \circ \circ \ldots$  being a colored integer, with the colored tensor powers being defined according to the rules

0.

$$u^{\otimes \circ} = u$$
$$u^{\otimes \bullet} = \bar{u}$$
$$u^{\otimes kl} = u^{\otimes k} \otimes u^{\otimes l}$$

and with the convention that  $u^{\otimes \emptyset}$  is the trivial representation  $1: G \to U_1$ .

302

Here are a few examples of such Peter-Weyl representations, namely those coming from the colored integers of length 2, to be often used in what follows:

$$\begin{split} u^{\otimes \circ \circ} &= u \otimes u \quad , \quad u^{\otimes \circ \bullet} = u \otimes \bar{u} \\ u^{\otimes \bullet \circ} &= \bar{u} \otimes u \quad , \quad u^{\otimes \bullet \bullet} = \bar{u} \otimes \bar{u} \end{split}$$

In relation now with characters, we have the following result:

**THEOREM** 13.5. The characters of the Peter-Weyl representations are given by

$$\chi_{u^{\otimes k}} = (\chi_u)^k$$

with the colored powers of a variable  $\chi$  being by definition given by

$$\chi^{\circ} = \chi$$
$$\chi^{\bullet} = \bar{\chi}$$
$$\chi^{kl} = \chi^k \chi^l$$

and with the convention that  $\chi^{\emptyset}$  equals by definition 1.

PROOF. This follows indeed from the additivity, multiplicativity and conjugation formulae established in Proposition 13.3 above, via the conventions in Definition 13.4.  $\Box$ 

Getting back now to our motivations, we can see the interest in the above constructions. Indeed, the joint moments of the main character  $\chi = \chi_u$  and its adjoint  $\bar{\chi} = \chi_{\bar{u}}$ are simply the expectations of the characters of various Peter-Weyl representations:

$$\int_G \chi^k = \int_G \chi_{u^{\otimes k}}$$

Summarizing, given  $G \subset U_N$ , we would like to understand its Peter-Weyl representations, and compute the expectations of the characters of these representations.

#### 13b. Peter-Weyl theory

In order to advance, we must develop some general theory. Let us start with:

DEFINITION 13.6. Given a compact group G, and two of its representations,

$$u: G \to U_N$$

$$v: G \to U_M$$

we define the linear space of intertwiners between these representations as being

$$Hom(u,v) = \left\{ T \in M_{M \times N}(\mathbb{C}) \middle| Tu(g) = v(g)T, \forall g \in G \right\}$$

and we use the following conventions:

- (1) We use the notations Fix(u) = Hom(1, u), and End(u) = Hom(u, u).
- (2) We write  $u \sim v$  when Hom(u, v) contains an invertible element.
- (3) We say that u is irreducible, and write  $u \in Irr(G)$ , when  $End(u) = \mathbb{C}1$ .

The terminology here is standard, with Hom and End standing for "homomorphisms" and "endomorphisms" between the representations in question, and with Fix standing for "fixed points". It is useful to think of the representations of G as being the "objects" of some kind of abstract picture of G, of rather combinatorial nature, and of the intertwiners between these representations as being the "arrows" between these objects.

Here are a few basic results, regarding the above intertwiner spaces:

**PROPOSITION 13.7.** We have the following results:

(1) The intertwiners are stable under composition:

$$T \in Hom(u, v)$$
,  $S \in Hom(v, w) \implies ST \in Hom(u, w)$ 

(2) The intertwiners are stable under taking tensor products:

$$S \in Hom(u, v)$$
,  $T \in Hom(w, t) \implies S \otimes T \in Hom(u \otimes w, v \otimes t)$ 

(3) The intertwiners are stable under taking adjoints:

$$T \in Hom(u, v) \implies T^* \in Hom(v, u)$$

In abstract terms, we say that the Hom spaces form a tensor \*-category.

**PROOF.** All the formulae in the statement are clear from definitions, or rather follow from some elementary computations, based on the main definition, namely:

$$Hom(u,v) = \left\{ T \in M_{M \times N}(\mathbb{C}) \middle| Tu(g) = v(g)T, \forall g \in G \right\}$$

As for the last assertion, this is something coming from (1,2,3). We will be back to tensor categories later on, in chapter 14 below, with more details on all this.

As a main consequence of the above result, we have:

**PROPOSITION 13.8.** Given a representation of a compact group,

$$u: G \to U_N$$

the corresponding linear space of self-intertwiners

$$End(u) \subset M_N(\mathbb{C})$$

is a \*-algebra, with respect to the usual involution of the matrices.

PROOF. By definition, End(u) is a linear subspace of  $M_N(\mathbb{C})$ . We know from Proposition 13.7 (1) that this subspace End(u) is a subalgebra of  $M_N(\mathbb{C})$ , and then we know as well from Proposition 13.7 (3) that this subalgebra is stable under the involution \*. Thus, what we have here is a \*-subalgebra of  $M_N(\mathbb{C})$ , as claimed.  $\Box$ 

The above result is quite interesting, because it gets us into linear algebra. To be more precise, associated to any group representation  $u : G \to U_N$ , which is something quite abstract, is now a quite familiar object, namely the algebra  $End(u) \subset M_N(\mathbb{C})$ .

In order to exploit this fact, we will need a basic, well-known result, complementing the basic operator algebra theory developed in chapter 7 above, as follows:

THEOREM 13.9. Let  $A \subset M_N(\mathbb{C})$  be a \*-algebra.

(1) The unit decomposes as follows, with  $p_i \in A$  being central minimal projections:

 $1 = p_1 + \ldots + p_k$ 

(2) Each of the following linear spaces is a non-unital \*-subalgebra of A:

$$A_i = p_i A p_i$$

(3) We have a non-unital \*-algebra sum decomposition, as follows:

 $A = A_1 \oplus \ldots \oplus A_k$ 

(4) We have unital \*-algebra isomorphisms as follows, with  $N_i = rank(p_i)$ :

 $A_i \simeq M_{N_i}(\mathbb{C})$ 

(5) Thus, we have a \*-algebra isomorphism as follows:

$$A \simeq M_{N_1}(\mathbb{C}) \oplus \ldots \oplus M_{N_k}(\mathbb{C})$$

PROOF. Consider indeed an arbitrary \*-algebra of the  $N \times N$  matrices,  $A \subset M_N(\mathbb{C})$ . Let us first look at the center of this algebra, which given by:

$$Z(A) = A \cap A'$$

It is elementary to prove that this center, as an algebra, is of the following form:

$$Z(A) \simeq \mathbb{C}^k$$

Consider now the standard basis  $e_1, \ldots, e_k \in \mathbb{C}^k$ , and let  $p_1, \ldots, p_k \in Z(A)$  be the images of these vectors via the above identification. In other words, these elements  $p_1, \ldots, p_k \in A$  are central minimal projections, summing up to 1:

$$p_1 + \ldots + p_k = 1$$

The idea is then that this partition of the unity will eventually lead to the block decomposition of A, as in the statement. We prove this in 4 steps, as follows:

Step 1. We first construct the matrix blocks, our claim here being that each of the following linear subspaces of A are non-unital \*-subalgebras of A:

$$A_i = p_i A p_i$$

But this is clear, with the fact that each  $A_i$  is closed under the various non-unital \*-subalgebra operations coming from the projection equations  $p_i^2 = p_i = p_i^*$ .

Step 2. We prove now that the above algebras  $A_i \subset A$  are in a direct sum position, in the sense that we have a non-unital \*-algebra sum decomposition, as follows:

$$A = A_1 \oplus \ldots \oplus A_k$$

As with any direct sum question, we have two things to be proved here. First, by using the formula  $p_1 + \ldots + p_k = 1$  and the projection equations  $p_i^2 = p_i = p_i^*$ , we conclude that we have the needed generation property, namely:

$$A_1 + \ldots + A_k = A$$

As for the fact that the sum is indeed direct, this follows as well from the formula  $p_1 + \ldots + p_k = 1$ , and from the projection equations  $p_i^2 = p_i = p_i^*$ .

<u>Step 3.</u> Our claim now, which will finish the proof, is that each of the \*-subalgebras  $A_i = p_i A p_i$  constructed above is a full matrix algebra. To be more precise here, with  $r_i = rank(p_i)$ , our claim is that we have isomorphisms, as follows:

$$A_i \simeq M_{r_i}(\mathbb{C})$$

In order to prove this claim, recall that the projections  $p_i \in A$  were chosen central and minimal. Thus, the center of each of the algebras  $A_i$  reduces to the scalars:

$$Z(A_i) = \mathbb{C}$$

But this shows, either via a direct computation, or via the bicommutant theorem, that the each of the algebras  $A_i$  is a full matrix algebra, as claimed.

<u>Step 4</u>. We can now obtain the result, by putting together what we have. Indeed, by using the results from Step 2 and Step 3, we obtain an isomorphism as follows:

$$A \simeq M_{r_1}(\mathbb{C}) \oplus \ldots \oplus M_{r_k}(\mathbb{C})$$

Moreover, a careful look at the isomorphisms established in Step 3 shows that at the global level, of the algebra A itself, the above isomorphism simply comes by twisting the following standard multimatrix embedding, discussed in the beginning of the proof, (1) above, by a certain unitary matrix  $U \in U_N$ :

$$M_{r_1}(\mathbb{C}) \oplus \ldots \oplus M_{r_k}(\mathbb{C}) \subset M_N(\mathbb{C})$$

Now by putting everything together, we obtain the result.

Observe that, in terms of the basic spectral theory notions developed in chapter 7 above, the above result basically tells us that the finite dimensional  $C^*$ -algebras are exactly the direct sums of matrix algebras. We will use this several times, in what follows.

We can now formulate our first Peter-Weyl type theorem, as follows:

THEOREM 13.10 (PW1). Let  $u : G \to U_N$  be a group representation, consider the algebra A = End(u), and write its unit as above, as follows:

$$1 = p_1 + \ldots + p_k$$

We have then a decomoposition result for u, as follows,

$$u = u_1 + \ldots + u_k$$

with each  $u_i$  being an irreducible representation, obtained by restricting u to  $Im(p_i)$ .

**PROOF.** This follows from Proposition 13.8 and Theorem 13.9, as follows:

(1) We first associate to our representation  $u: G \to U_N$  the corresponding action map on  $\mathbb{C}^N$ . If a linear subspace  $V \subset \mathbb{C}^N$  is invariant, the restriction of the action map to Vis an action map too, which must come from a subrepresentation  $v \subset u$ .

(2) Consider now a projection  $p \in End(u)$ . From pu = up we obtain that the linear space V = Im(p) is invariant under u, and so this space must come from a subrepresentation  $v \subset u$ . It is routine to check that the operation  $p \to v$  maps subprojections to subrepresentations, and minimal projections to irreducible representations.

(3) With these preliminaries in hand, let us decompose the algebra End(u) as in Theorem 13.9, by using the decomposition  $1 = p_1 + \ldots + p_k$  into minimal projections. If we denote by  $u_i \subset u$  the subrepresentation coming from the vector space  $V_i = Im(p_i)$ , then we obtain in this way a decomposition  $u = u_1 + \ldots + u_k$ , as in the statement.  $\Box$ 

In order to formulate our second Peter-Weyl theorem, we need to talk about coefficients, and smoothness. Things here are quite tricky, and best is to proceed as follows:

DEFINITION 13.11. Given a closed subgroup  $G \subset U_N$ , and a unitary representation  $v: G \to U_M$ , the space of coefficients of this representation is:

$$C_v = \left\{ f \circ v \middle| f \in M_M(\mathbb{C})^* \right\}$$

In other words, by delinearizing,  $C_{\nu} \subset C(G)$  is the following linear space:

$$C_v = span \left[ g \to v(g)_{ij} \right]$$

We say that v is smooth if its matrix coefficients  $g \to v(g)_{ij}$  appear as polynomials in the standard matrix coordinates  $g \to g_{ij}$ , and their conjugates  $g \to \overline{g}_{ij}$ .

As a basic example of coefficient we have, besides the matrix coefficients  $g \to v(g)_{ij}$ , the character, which appears as the diagonal sum of these coefficients:

$$\chi_v(g) = \sum_i v(g)_{ii}$$

Regarding the notion of smoothness, things are quite tricky here, the idea being that any closed subgroup  $G \subset U_N$  can be shown to be a Lie group, and that, with this result in

hand, a representation  $v: G \to U_M$  is smooth precisely when the condition on coefficients from the above definition is satisfied. All this is quite technical, and we will not get into it. We will simply use Definition 13.11 as such, and further comment on this later on.

Here is now our second Peter-Weyl theorem, complementing Theorem 13.10:

THEOREM 13.12 (PW2). Given a closed subgroup  $G \subset_u U_N$ , any of its irreducible smooth representations

$$v: G \to U_M$$

appears inside a tensor product of the fundamental representation u and its adjoint  $\bar{u}$ .

**PROOF.** In order to prove the result, we will use the following three elementary facts, regarding the spaces of coefficients introduced above:

(1) The construction  $v \to C_v$  is functorial, in the sense that it maps subrepresentations into linear subspaces. This is indeed something which is routine to check.

(2) Our smoothness assumption on  $v : G \to U_M$ , as formulated in Definition 13.11, means that we have an inclusion of linear spaces as follows:

$$C_v \subset \langle g_{ij} \rangle$$

(3) By definition of the Peter-Weyl representations, as arbitrary tensor products between the fundamental representation u and its conjugate  $\bar{u}$ , we have:

$$< g_{ij} > = \sum_k C_{u^{\otimes k}}$$

Now by putting together the observations (2,3) we conclude that we must have an inclusion as follows, for certain exponents  $k_1, \ldots, k_p$ :

$$C_v \subset C_{u^{\otimes k_1} \oplus \ldots \oplus \pi^{\otimes k_p}}$$

By using now the functoriality result from (1), we deduce from this that we have an inclusion of representations, as follows:

$$v \subset u^{\otimes k_1} \oplus \ldots \oplus u^{\otimes k_k}$$

Together with Theorem 13.10, this leads to the conclusion in the statement.  $\Box$ 

As a conclusion to what we have so far, the problem to be solved is that of splitting the Peter-Weyl representations into sums of irreducible representations. We will be back to this, and to other aspects of the Peter-Weyl theory, after a short break.

#### 13C. HAAR INTEGRATION

#### 13c. Haar integration

In order to further advance, and complete the Peter-Weyl theory, we need to talk about integration over G. In the finite group case the situation is trivial, as follows:

**PROPOSITION 13.13.** Any finite group G has a unique probability measure which is invariant under left and right translations,

$$\mu(E) = \mu(gE) = \mu(Eg)$$

and this measure is the normalized counting measure on G, given by  $\mu(E) = \frac{|E|}{|G|}$ .

**PROOF.** The uniformity condition in the statement gives, with the choice  $E = \{h\}$ :

$$\mu\{h\} = \mu\{gh\} = \mu\{hg\}$$

Thus  $\mu$  must be the usual counting measure, normalized as to have mass 1.

In the continuous group case now, the simplest examples, to be studied first, are the compact abelian groups. Here things are standard again, as follows:

THEOREM 13.14. Given a compact abelian group G, with dual group denoted  $\Gamma = \widehat{G}$ , we have an isomorphism of commutative algebras

$$C(G) \simeq C^*(\Gamma)$$

and via this isomorphism, the functional defined by linearity and the following formula,

$$\int_G g = \delta_{g1}$$

for any  $g \in \Gamma$ , is the integration with respect to the unique uniform measure on G.

**PROOF.** This is something that we basically know, from chapters 7 and 9 above, coming as a consequence of the general results regarding the abelian groups and the commutative algebras developed there. To be more precise, and skipping some details here, the conclusions in the statement can be deduced as follows:

(1) We can either apply the Gelfand theorem, from chapter 7 above, to the group algebra  $C^*(\Gamma)$ , which is commutative, and this gives all the results.

(2) Or, we can use decomposition results for the compact abelian groups from chapter 9 above, and by reducing things to summands, once again we obtain the results.  $\Box$ 

Summarizing, we have results in the finite case, and in the compact abelian case. With the remark that the proof in the compact abelian case was quite brief, but this result, coming as an illustration for more general things to follow, is not crucial for us.

Let us discuss now the construction of the uniform probability measure in general. This is something quite technical, the idea being that the uniform measure  $\mu$  over G can be constructed by starting with an arbitrary probability measure  $\nu$ , and setting:

$$\mu = \lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} \nu^{*k}$$

Thus, our next task will be that of proving this result. It is convenient, for this purpose, to work with the integration functionals with respect to the various measures on G, instead of the measures themselves. Let us begin with the following key result:

**PROPOSITION 13.15.** Given a unital positive linear form  $\varphi : C(G) \to \mathbb{C}$ , the limit

$$\int_{\varphi} f = \lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} \varphi^{*k}(f)$$

exists, and for a coefficient of a representation  $f = (\tau \otimes id)v$  we have

$$\int_{\varphi} f = \tau(P)$$

where P is the orthogonal projection onto the 1-eigenspace of  $(id \otimes \varphi)v$ .

**PROOF.** By linearity it is enough to prove the first assertion for elements of the following type, where v is a Peter-Weyl representation, and  $\tau$  is a linear form:

$$a = (\tau \otimes id)v$$

Thus we are led into the second assertion, and more precisely we can have the whole result proved if we can establish the following formula, with  $a = (\tau \otimes id)v$ :

$$\lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} \varphi^{*k}(a) = \tau(P)$$

In order to prove this latter formula, observe that we have:

$$\varphi^{*k}(a) = (\tau \otimes \varphi^{*k})v = \tau((id \otimes \varphi^{*k})v)$$

Let us set  $M = (id \otimes \varphi)v$ . In terms of this matrix, we have:

$$((id \otimes \varphi^{*k})v)_{i_0i_{k+1}} = \sum_{i_1\dots i_k} M_{i_0i_1}\dots M_{i_ki_{k+1}}$$
$$= (M^k)_{i_0i_{k+1}}$$

Thus we have the following formula, for any  $k \in \mathbb{N}$ :

$$(id \otimes \varphi^{*k})v = M^k$$

It follows that our Cesàro limit is given by:

$$\lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} \varphi^{*k}(a) = \lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} \tau(M^{k})$$
$$= \tau \left( \lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} M^{k} \right)$$

Now since v is unitary we have ||v|| = 1, and we obtain from this:

$$||M|| \le 1$$

Thus the Cesàro limit on the right in the previous formula converges, and equals the orthogonal projection onto the 1-eigenspace of M:

$$\lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} M^k = P$$

Thus our initial Cesàro limit converges as well, to  $\tau(P)$ , as desired.

The point now is that when the linear form  $\varphi \in C(G)^*$  from the above result is chosen to be faithful, we obtain the following finer result:

**PROPOSITION 13.16.** Given a faithful unital linear form  $\varphi \in C(G)^*$ , the limit

$$\int_{\varphi} a = \lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} \varphi^{*k}(a)$$

exists, and is independent of  $\varphi$ , given on coefficients of representations by

$$\left(id \otimes \int_{\varphi}\right)v = P$$

where P is the orthogonal projection onto the following space:

$$Fix(v) = \left\{ \xi \in \mathbb{C}^n \middle| v\xi = \xi \right\}$$

PROOF. In view of Proposition 13.15, it remains to prove that when  $\varphi$  is faithful, the 1-eigenspace of the matrix  $M = (id \otimes \varphi)v$  equals the space Fix(v).

" $\supset$ " This is clear, and for any  $\varphi$ , because we have:

$$v\xi = \xi \implies M\xi = \xi$$

" $\subset$ " Here we must prove that, when  $\varphi$  is faithful, we have:

$$M\xi = \xi \implies v\xi = \xi$$

For this purpose, assume that we have  $M\xi = \xi$ , and consider the following element:

$$a = \sum_{i} \left( \sum_{j} v_{ij} \xi_j - \xi_i \right) \left( \sum_{k} v_{ik} \xi_k - \xi_i \right)^*$$

We must prove that we have a = 0. Since v is unitary, we have:

$$a = \sum_{i} \left( \sum_{j} \left( v_{ij}\xi_{j} - \frac{1}{N}\xi_{i} \right) \right) \left( \sum_{k} \left( v_{ik}^{*}\bar{\xi}_{k} - \frac{1}{N}\bar{\xi}_{i} \right) \right)$$
  
$$= \sum_{ijk} v_{ij}v_{ik}^{*}\xi_{j}\bar{\xi}_{k} - \frac{1}{N}v_{ij}\xi_{j}\bar{\xi}_{i} - \frac{1}{N}v_{ik}^{*}\xi_{i}\bar{\xi}_{k} + \frac{1}{N^{2}}\xi_{i}\bar{\xi}_{i}$$
  
$$= \sum_{j} |\xi_{j}|^{2} - \sum_{ij} v_{ij}\xi_{j}\bar{\xi}_{i} - \sum_{ik} v_{ik}^{*}\xi_{i}\bar{\xi}_{k} + \sum_{i} |\xi_{i}|^{2}$$
  
$$= ||\xi||^{2} - \langle v\xi, \xi \rangle - \langle v\xi, \xi \rangle + ||\xi||^{2}$$
  
$$= 2(||\xi||^{2} - Re(\langle v\xi, \xi \rangle))$$

By using now our assumption  $M\xi = \xi$ , we obtain from this:

$$\begin{aligned} \varphi(a) &= 2\varphi(||\xi||^2 - Re(\langle v\xi, \xi \rangle)) \\ &= 2(||\xi||^2 - Re(\langle M\xi, \xi \rangle)) \\ &= 2(||\xi||^2 - ||\xi||^2) \\ &= 0 \end{aligned}$$

Now since  $\varphi$  is faithful, this gives a = 0, and so  $v\xi = \xi$ , as claimed.

We can now formulate a main result, as follows:

THEOREM 13.17. Any compact group G has a unique Haar integration, which can be constructed by starting with any faithful positive unital state  $\varphi \in C(G)^*$ , and setting:

$$\int_G = \lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^n \varphi^{*k}$$

Moreover, for any representation v we have the formula

$$\left(id \otimes \int_G\right)v = P$$

where P is the orthogonal projection onto the following linear space:

$$Fix(v) = \left\{ \xi \in \mathbb{C}^n \middle| v\xi = \xi \right\}$$

312

**PROOF.** We can prove this from what we have, in several steps, as follows:

(1) Let us first go back to the general context of Proposition 13.15 above. Since convolving one more time with  $\varphi$  will not change the Cesàro limit appearing there, the functional  $\int_{\varphi} \in C(G)^*$  constructed there has the following invariance property:

$$\int_{\varphi} \ast \varphi = \varphi \ast \int_{\varphi} = \int_{\varphi}$$

In the case where  $\varphi$  is assumed to be faithful, as in Proposition 13.16 above, our claim is that we have the following formula, valid this time for any  $\psi \in C(G)^*$ :

$$\int_{\varphi} *\psi = \psi * \int_{\varphi} = \psi(1) \int_{\varphi}$$

It is enough to prove this formula on a coefficient of a corepresentation:

$$a = (\tau \otimes id)v$$

In order to do so, consider the following two matrices:

$$P = \left(id \otimes \int_{\varphi}\right)v \quad , \quad Q = (id \otimes \psi)v$$

We have then the following computation:

$$\left(\int_{\varphi} *\psi\right) a = \left(\tau \otimes \int_{\varphi} \otimes\psi\right) (v_{12}v_{13}) \\ = \tau(PQ)$$

Similarly, we have the following computation:

$$\left(\psi * \int_{\varphi}\right) a = \left(\tau \otimes \psi \otimes \int_{\varphi}\right) (v_{12}v_{13}) \\ = \tau(QP)$$

Finally, regarding the term on the right, this is given by:

$$\psi(1)\int_{\varphi}a=\psi(1)\tau(P)$$

Thus, our claim is equivalent to the following equality:

$$PQ = QP = \psi(1)P$$

But this follows from the fact, coming from Proposition 13.16, that  $P = (id \otimes \int_{\varphi})v$  equals the orthogonal projection onto Fix(v). Thus, we have proved our claim.

(2) In order to finish now, it is convenient to introduce the following abstract operation, on the continuous functions  $a, b : C(G) \to \mathbb{C}$  on our group:

$$\Delta(a\otimes b)(g\otimes h) = a(g)b(h)$$

With this convention, the formula that we established above can be written as:

$$\psi\left(\int_{\varphi} \otimes id\right)\Delta = \psi\left(id \otimes \int_{\varphi}\right)\Delta$$
$$= \psi\int_{\varphi} (.)1$$

This formula being true for any  $\psi \in C(G)^*$ , we can simply delete  $\psi$ . We conclude that the following invariance formula holds indeed, with  $\int_G = \int_{\varphi}$ :

$$\left(\int_{G} \otimes id\right) \Delta = \left(id \otimes \int_{G}\right) \Delta = \int_{G} (.)1$$

But this is exactly the left and right invariance formula we were looking for.

(3) Finally, in order to prove the uniqueness assertion, assuming that we have two invariant integrals  $\int_G$ ,  $\int'_G$ , we have, according to the above invariance formula:

$$\left(\int_{G} \otimes \int_{G}'\right) \Delta = \left(\int_{G}' \otimes \int_{G}\right) \Delta$$
$$= \int_{G} (.)1$$
$$= \int_{G}' (.)1$$

Thus we have  $\int_G = \int'_G$ , and this finishes the proof.

Summarizing, we can now integrate over G. As a first application, we have:

THEOREM 13.18. Given a compact group G, we have the following formula, valid for any unitary group representation  $v: G \to U_M$ :

$$\int_G \chi_v = \dim(Fix(v))$$

In particular, in the unitary matrix group case,  $G \subset_u U_N$ , the moments of the main character  $\chi = \chi_u$  are given by the following formula:

$$\int_G \chi^k = \dim(Fix(u^{\otimes k}))$$

Thus, knowing the law of the main character  $\chi = \chi_u$  is the same as knowing the number of fixed points of the Peter-Weyl representations  $u^{\otimes k}$ .

**PROOF.** We have three statements here, the idea being as follows:

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(1) Given a unitary representation  $v : G \to U_M$  as in the statement, its character  $\chi_v$  is a coefficient, so we can use the integration formula for coefficients in Theorem 13.17. If we denote by P the projection onto Fix(v), this formula gives, as desired:

$$\int_{G} \chi_{v} = Tr(P)$$
  
= dim(Im(P))  
= dim(Fix(v))

(2) This follows from (1), applied to the Peter-Weyl representations, as follows:

$$\int_{G} \chi^{k} = \int_{G} \chi^{k}_{u}$$
$$= \int_{G} \chi_{u^{\otimes k}}$$
$$= \dim(Fix(u^{\otimes k}))$$

(3) This follows from (2), and from the standard fact, which follows from definitions, that a probability measure is uniquely determined by its moments.  $\Box$ 

As a key remark now, in principle the integration formula for coefficients in Theorem 13.17, which is something quite powerful, allows the computation for the truncated characters too, because these truncated characters are coefficients as well.

To be more precise, all the probabilistic questions about G, regarding characters, or truncated characters, or more complicated variables, require a good knowledge of the integration over G, and more precisely, of the various polynomial integrals over G:

DEFINITION 13.19. Given a closed subgroup  $G \subset U_N$ , the quantities

$$I_k = \int_G g_{i_1 j_1}^{e_1} \dots g_{i_k j_k}^{e_k} \, dg$$

depending on a colored integer  $k = e_1 \dots e_k$ , are called polynomial integrals over G.

As a first observation, the knowledge of these integrals is the same as the full knowledge of the integration functional over G. Indeed, since the coordinate functions  $g \to g_{ij}$ separate the points of G, we can apply the Stone-Weierstrass theorem, and we obtain:

$$C(G) = < g_{ij} >$$

Thus, by linearity, the computation of any functional  $f : C(G) \to \mathbb{C}$ , and in particular of the integration functional, reduces to the computation of this functional on the polynomials of the coordinate functions  $g \to g_{ij}$  and their conjugates  $g \to \bar{g}_{ij}$ .

By using the Peter-Weyl theory, everything reduces to algebra, due to:

THEOREM 13.20. The Haar integration over a closed subgroup  $G \subset_u U_N$  is given on the dense subalgebra of smooth functions by the Weingarten type formula

$$\int_G g_{i_1 j_1}^{e_1} \dots g_{i_k j_k}^{e_k} dg = \sum_{\pi, \sigma \in D_k} \delta_{\pi}(i) \delta_{\sigma}(j) W_k(\pi, \sigma)$$

valid for any colored integer  $k = e_1 \dots e_k$  and any multi-indices i, j, where  $D_k$  is a linear basis of  $Fix(u^{\otimes k})$ , the associated generalized Kronecker symbols are given by

 $\delta_{\pi}(i) = <\pi, e_{i_1} \otimes \ldots \otimes e_{i_k} >$ 

and  $W_k = G_k^{-1}$  is the inverse of the Gram matrix,  $G_k(\pi, \sigma) = <\pi, \sigma >$ .

**PROOF.** We know from Peter-Weyl theory that the integrals in the statement form altogether the orthogonal projection  $P^k$  onto the following space:

$$Fix(u^{\otimes k}) = span(D_k)$$

Consider now the following linear map, with  $D_k = \{\xi_k\}$  being as in the statement:

$$E(x) = \sum_{\pi \in D_k} \langle x, \xi_\pi \rangle \xi_\pi$$

By a standard linear algebra computation, it follows that we have P = WE, where W is the inverse of the restriction of E to the following space:

$$K = span\left(T_{\pi} \middle| \pi \in D_k\right)$$

But this restriction is precisely the linear map given by the matrix  $G_k$ , and so W itself is the linear map given by the matrix  $W_k$ , and this gives the result.

We will be back to this in chapter 16 below, with some concrete applications.

#### 13d. More Peter-Weyl

In order to further develop now the Peter-Weyl theory, which is something very useful, we will need the following result, which is of independent interest:

**PROPOSITION 13.21.** We have a Frobenius type isomorphism

$$Hom(v,w) \simeq Fix(v \otimes \bar{w})$$

valid for any two representations v, w.

**PROOF.** According to the definitions, we have the following equivalences:

$$\begin{array}{lll} T \in Hom(v,w) & \Longleftrightarrow & Tv = wT \\ & \Longleftrightarrow & \sum_{j} T_{aj}v_{ji} = \sum_{b} w_{ab}T_{bi}, \forall a, i \end{array}$$

On the other hand, we have as well the following equivalences:

$$T \in Fix(v \otimes \bar{w}) \iff (v \otimes \bar{w})T = \xi$$
$$\iff \sum_{jb} v_{ij} w_{ab}^* T_{bj} = T_{ai} \forall a, i$$

With these formulae in hand, both inclusions follow from the unitarity of v, w.  $\Box$ 

We can now formulate our third Peter-Weyl theorem, as follows:

THEOREM 13.22 (PW3). The norm dense \*-subalgebra

$$\mathcal{C}(G) \subset C(G)$$

generated by the coefficients of the fundamental representation decomposes as a direct sum

$$\mathcal{C}(G) = \bigoplus_{v \in Irr(G)} M_{\dim(v)}(\mathbb{C})$$

with this being an isomorphism of \*-coalgebras, and with the summands being pairwise orthogonal with respect to the scalar product given by

$$\langle a,b \rangle = \int_{G} ab^{*}$$

where  $\int_G$  is the Haar integration over G.

**PROOF.** By combining the previous two Peter-Weyl results, we deduce that we have a linear space decomposition as follows:

$$\mathcal{C}(G) = \sum_{v \in Irr(G)} C_v$$
$$= \sum_{v \in Irr(G)} M_{\dim(v)}(\mathbb{C})$$

Thus, in order to conclude, it is enough to prove that for any two irreducible corepresentations  $v, w \in Irr(A)$ , the corresponding spaces of coefficients are orthogonal:

$$v \not\sim w \implies C_v \perp C_w$$

But this follows from Theorem 13.17, via Proposition 13.21. Let us set indeed:

$$P_{ia,jb} = \int_G v_{ij} w_{ab}^*$$

Then P is the orthogonal projection onto the following vector space:

$$Fix(v \otimes \bar{w}) \simeq Hom(v, w) = \{0\}$$

Thus we have P = 0, and this gives the result.

Finally, we have the following result, completing the Peter-Weyl theory:

THEOREM 13.23 (PW4). The characters of the irreducible representations of G belong to the algebra

$$\mathcal{C}(G)_{central} = \left\{ a \in \mathcal{C}(G) \middle| \Sigma \Delta(a) = \Delta(a) \right\}$$

of smooth central functions on G, and form an orthonormal basis of it.

PROOF. We have several things to be proved, the idea being as follows:

(1) Observe first that  $\mathcal{C}(G)_{central}$  is indeed an algebra, which contains all the characters. Conversely, consider an element  $a \in \mathcal{C}(G)$ , written as follows:

$$a = \sum_{v \in Irr(G)} a_v$$

The condition  $a \in \mathcal{C}(G)_{central}$  states then that for any  $v \in Irr(G)$ , we must have:

$$a_v \in \mathcal{C}(G)_{central}$$

But this means precisely that the coefficient  $a_v$  must be a scalar multiple of  $\chi_v$ , and so the characters form a basis of  $\mathcal{C}(G)_{central}$ , as stated.

(2) The fact that we have an orthogonal basis follows from Theorem 13.22.

(3) As for the fact that the characters have norm 1, this follows from:

$$\int_{G} \chi_{v} \chi_{v}^{*} = \sum_{ij} \int_{G} v_{ii} v_{jj}^{*}$$
$$= \sum_{i} \frac{1}{N}$$
$$= 1$$

Here we have used the fact, coming from Theorem 13.22, that the integrals  $\int_G v_{ij} v_{kl}^*$  form the orthogonal projection onto the following vector space:

$$Fix(v \otimes \bar{v}) \simeq End(v) = \mathbb{C}1$$

Thus, the proof of our theorem is now complete.

As a basic illustration for this, in the abelian group case the irreducible representations are all 1-dimensional, and form the dual group. There are also many things that can be said in the finite group case, in relation with central functions, and conjugacy classes.

#### 13e. Exercises

There has been a lot of theory on this chapter, which was often quite abstract, and as basic exercise on the subject, which is quite elementary, we have:

318

#### 13E. EXERCISES

EXERCISE 13.24. Work out the details for the Peter-Weyl decomposition formula

$$\mathcal{C}(G) = \bigoplus_{v \in Irr(G)} M_{\dim(v)}(\mathbb{C})$$

in an explicit way, for some finite non-abelian groups, of your choice.

There are of course many possible choices for G. As a bonus exercise here, you can try do the same for the subalgebra of central functions  $\mathcal{C}(G)_{central}$  too.

Going ahead now, one question which still remains unanswered is that of construcing the dual of a compact non-abelian group. There is a solution to this problem, based on the above, and we propose this now, in the form of a series of exercises. Let us first discuss the finite group case, which is simpler. We have the following exercise, to start with:

EXERCISE 13.25. Given a finite group G, setting A = C(G), prove that the maps

$$\Delta: A \to A \otimes A \quad , \quad \varepsilon: A \to \mathbb{C} \quad , \quad S: A \to A$$

which are transpose to the multiplication  $m : G \times G \to G$ , unit  $u : \{.\} \to G$  and inverse map  $i : G \to G$ , are subject to the following conditions

$$(\varepsilon \otimes id)\Delta = (id \otimes \varepsilon)\Delta = id$$

$$m(S \otimes id)\Delta = m(id \otimes S)\Delta = \varepsilon(.)1$$

in usual tensor product notation, along with the extra condition  $S^2 = id$ .

This does not look difficult, with the conditions in the statement reminding the usual group axioms, satisfied by m, u, i. Up to you to prove this now, with full details.

Along the same lines, we have the following exercise:

EXERCISE 13.26. Given a finite group H, setting  $A = C^*(H)$ , prove that the maps

 $\Delta: A \to A \otimes A \quad , \quad \varepsilon: A \to \mathbb{C} \quad , \quad S: A \to A^{opp}$ 

given by the formulae  $\Delta(g) = g \otimes g$ ,  $\varepsilon(g) = 1$ ,  $S(g) = g^{-1}$  and linearity, are subject to the same conditions as above, namely

$$(\varepsilon \otimes id)\Delta = (id \otimes \varepsilon)\Delta = id$$

$$m(S\otimes id)\Delta=m(id\otimes S)\Delta=\varepsilon(.)1$$

in usual tensor product notation, along with the extra condition  $S^2 = id$ .

As before with the previous exercise, this does not look very difficult, with most likely only some elementary algebraic computations involved.

The point now is that, once you solved the previous two exercises, you can solve:

EXERCISE 13.27. Let us call finite Hopf algebra a finite dimensional  $C^*$ -algebra, with maps as follows, called comultiplication, counit and antipode,

 $\Delta: A \to A \otimes A \quad, \quad \varepsilon: A \to \mathbb{C} \quad, \quad S: A \to A^{opp}$ 

satisfying the following conditions, which are those found above,

$$(\varepsilon \otimes id)\Delta = (id \otimes \varepsilon)\Delta = id$$

$$m(S \otimes id)\Delta = m(id \otimes S)\Delta = \varepsilon(.)1$$

along with  $S^2 = id$ . Prove that if G, H are finite abelian groups, dual to each other via Pontrjagin duality, then we have an identification of Hopf algebras as follows,

$$C(G) = C^*(H)$$

and based on this, go ahead and formally write any finite Hopf algebra as

$$A = C(G) = C^*(H)$$

and call G, H finite quantum groups, dual to each other.

Here the thing to be done, namely to establish the identification in the statement, looks like something quite routine, related to many things that we already know. As for the last part, there is nothing to be done here, just enjoying that definition.

Here is one more exercise on all this, further clarifying the above:

EXERCISE 13.28. Given a finite dimensional Hopf algebra A, prove that its dual  $A^*$  is a Hopf algebra too, with structural maps as follows:

$$\Delta^{t} : A^{*} \otimes A^{*} \to A^{*}$$

$$\varepsilon^{t} : \mathbb{C} \to A^{*}$$

$$m^{t} : A^{*} \to A^{*} \otimes A^{*}$$

$$u^{t} : A^{*} \to \mathbb{C}$$

$$S^{t} : A^{*} \to A^{*}$$

Also, check that A is commutative if and only if  $A^*$  is cocommutative, and also discuss what happens in the cases A = C(G) and  $A = C^*(H)$ , with G, H being finite groups.

This exercise is actually to be best worked out at the same time with the previous one. In this way you'll fully learn what the finite quantum groups are.

At a more advanced level now, and fully solving the original duality question, we have:

EXERCISE 13.29. Develop a theory of compact and discrete quantum groups, generalizing at the same time the theory of usual compact and discrete groups, and the Ponytrjagin duality for them, in the abelian case, and the theory of finite quantum groups, and the abstract duality for them, developed in the above series of exercises.

This latter exercise is probably quite difficult, better look it up.

## CHAPTER 14

# Tannakian duality

#### 14a. Tensor categories

We have seen that the representations of a closed subgroup  $G \subset U_N$  are subject to a number of non-trivial results, collectively known as Peter-Weyl theory. To be more precise, the main ideas of the Peter-Weyl theory were as follows:

- (1) The representations of G split as sums of irreducibles, and the irreducibles can be found by splitting the tensor powers  $u^{\otimes k}$  between the fundamental representation  $u: G \subset U_N$  and its adjoint  $\bar{u}: G \subset U_N$ , called Peter-Weyl representations.
- (2) The main problem is therefore that of splitting the Peter-Weyl representations  $u^{\otimes k}$  into irreducibles. Technically speaking, this leads to the question of explicitly computing the corresponding fixed point spaces  $Fix(u^{\otimes k})$ .
- (3) From a probabilistic perspective, in connection with characters and truncated characters, which require the explicit knowledge of  $\int_G$ , we are led into the same fundamental question, namely the computation of the spaces  $Fix(u^{\otimes k})$ .

Summarizing, no matter on what we want to do with our given closed subgroup  $G \subset_u U_N$ , algebra or analysis, we must compute the spaces  $Fix(u^{\otimes k})$ . There are of course some other motivations as well for the computation of these spaces, more abstract, coming from advanced algebra and quantum algebra, and more on this later.

As a first idea, it is technically convenient to slightly enlarge the class of spaces to be computed, by talking about Tannakian categories, as follows:

DEFINITION 14.1. The Tannakian category associated to a closed subgroup  $G \subset_u U_N$ is the collection C = (C(k, l)) of vector spaces

$$C(k,l) = Hom(u^{\otimes k}, u^{\otimes l})$$

where the representations  $u^{\otimes k}$  with  $k = \circ \bullet \circ \circ \ldots$  colored integer, defined by

$$u^{\otimes \emptyset} = 1$$
$$u^{\otimes \circ} = u$$
$$u^{\otimes \bullet} = \bar{u}$$

and multiplicativity,  $u^{\otimes kl} = u^{\otimes k} \otimes u^{\otimes l}$ , are the Peter-Weyl representations.

#### 14. TANNAKIAN DUALITY

Here are a few examples of such representations, namely those coming from the colored integers of length 2, to be often used in what follows:

$$u^{\otimes \circ \circ} = u \otimes u$$
$$u^{\otimes \circ \circ} = u \otimes \bar{u}$$
$$u^{\otimes \circ \circ} = \bar{u} \otimes u$$
$$u^{\otimes \circ \circ} = \bar{u} \otimes \bar{u}$$

As a first observation, the knowledge of the Tannakian category is more or less the same thing as the knowledge of the fixed point spaces, which appear as:

$$Fix(u^{\otimes k}) = C(0,k)$$

Indeed, these latter spaces fully determine all the spaces C(k, l), because of the Frobenius isomorphisms, which for the Peter-Weyl representations read:

$$C(k,l) = Hom(u^{\otimes k}, u^{\otimes l})$$
  

$$\simeq Hom(1, \bar{u}^{\otimes k} \otimes u^{\otimes l})$$
  

$$= Hom(1, u^{\otimes \bar{k}l})$$
  

$$= Fix(u^{\otimes \bar{k}l})$$

In practice, it is much better to study the spaces C(k, l) instead of the spaces  $Fix(u^{\otimes k})$ , because, as already pointed out in the previous chapter, C = (C(k, l)) is a tensor category. Thus, we have far more structure in this setting, that we can exploit.

In order to get started now, let us make a summary of what we have so far, regarding these spaces C(k, l), coming from the general theory developed in chapter 13 above.

In order to formulate our result, let us start with:

DEFINITION 14.2. Let H be a finite dimensional Hilbert space. A tensor category over H is a collection C = (C(k, l)) of linear spaces

$$C(k,l) \subset \mathcal{L}(H^{\otimes k}, H^{\otimes l})$$

satisfying the following conditions:

- (1)  $S, T \in C$  implies  $S \otimes T \in C$ .
- (2) If  $S, T \in C$  are composable, then  $ST \in C$ .
- (3)  $T \in C$  implies  $T^* \in C$ .
- (4) Each C(k, k) contains the identity operator.
- (5)  $C(\emptyset, k)$  with  $k = \circ \bullet, \bullet \circ$  contain the operator  $R: 1 \to \sum_i e_i \otimes e_i$ .
- (6) C(kl, lk) with  $k, l = 0, \bullet$  contain the flip operator  $\Sigma : a \otimes b \to b \otimes a$ .

Here the tensor powers  $H^{\otimes k}$ , which are Hilbert spaces depending on a colored integer  $k = \circ \bullet \bullet \circ \ldots$ , are defined by the following formulae, and multiplicativity:

$$H^{\otimes \emptyset} = \mathbb{C}$$
$$H^{\otimes \bullet} = H$$
$$H^{\otimes \bullet} = \bar{H} \simeq H$$

With these conventions, we have the following result, summarizing our knowledge on the subject, coming from the results from the previous chapter:

THEOREM 14.3. For a closed subgroup  $G \subset_u U_N$ , the associated Tannakian category

$$C(k,l) = Hom(u^{\otimes k}, u^{\otimes l})$$

is a tensor category over the Hilbert space  $H = \mathbb{C}^N$ .

PROOF. We know that the fundamental representation u acts on the Hilbert space  $H = \mathbb{C}^N$ , and that its conjugate  $\bar{u}$  acts on the Hilbert space  $\bar{H} = \mathbb{C}^N$ . Now by multiplicativity we conclude that any Peter-Weyl representation  $u^{\otimes k}$  acts on the Hilbert space  $H^{\otimes k}$ , and so that we have embeddings as in Definition 14.2, as follows:

$$C(k,l) \subset \mathcal{L}(H^{\otimes k}, H^{\otimes l})$$

Regarding now the fact that the axioms (1-6) in Definition 14.2 are indeed satisfied, this is something that we basically already know, as follows:

(1) This follows from definitions, and was explained in chapter 13 above.

- (2) Once again, this is clear, and was explained in chapter 13.
- (3) This result is clear too, and was explained in chapter 13.
- (4) This is something trivial, coming from definitions.

(5) This follows from the fact that each element  $g \in G$  is a unitary, which can be reformulated as follows, with  $R: a \to \sum_i e_i \otimes e_i$  being the map in Definition 14.2:

$$R \in Hom(1, g \otimes \overline{g})$$
$$R \in Hom(1, \overline{g} \otimes g)$$

Indeed, given an arbitrary matrix  $g \in M_N(\mathbb{C})$ , we have the following computation:

$$(g \otimes \bar{g})(R(1) \otimes 1) = \sum_{ijk} e_i \otimes e_k \otimes g_{ij}g_{kj}^*$$
$$= \sum_{ik} e_i \otimes e_k \otimes (gg^*)_{ik}$$

We conclude from this that we have the following equivalence:

$$R \in Hom(1, g \otimes \overline{g}) \iff gg^* = 1$$
Consider now the adjoint operator  $R^* : \mathbb{C}^N \otimes \mathbb{C}^N \to \mathbb{C}$ , which is given by:

$$R^*(e_i \otimes e_j) = \delta_{ij}$$

Once again with  $g \in M_N(\mathbb{C})$  being arbitrary, we have the following computation:

$$(R^* \otimes id)(g \otimes \bar{g})(e_j \otimes e_l \otimes 1) = \sum_i g_{ij}g^*_{il}$$
  
=  $(g^t \bar{g})_{jl}$ 

We conclude from this that we have the following equivalence:

$$R^* \in Hom(g \otimes \bar{g}, 1) \iff g^t \bar{g} = 1$$

Similarly, or simply by replacing g in the above two conclusions with its conjugate matrix  $\overline{g}$ , we have as well the following two equivalences:

$$R \in Hom(1, \bar{g} \otimes g) \iff \bar{g}g^t = 1$$
$$R^* \in Hom(\bar{g} \otimes g, 1) \iff g^*g = 1$$

Thus, the various intertwining conditions in Definition 14.2 (5) are all equivalent either to  $g^* = g^{-1}$ , or to  $g^t = \bar{g}^{-1}$ . But these two conditions are both equivalent to the fact that g is unitary, and so all the conditions in Definition 14.2 (5) are satisfied, as desired.

(6) This is something trivial, coming from the fact that the matrix coefficients  $g \to g_{ij}$ and their complex conjugates  $g \to \bar{g}_{ij}$  commute with each other.

With the above in hand, our purpose now will be that of showing that any closed subgroup  $G \subset U_N$  is uniquely determined by its Tannakian category C = (C(k, l)).

This result, known as Tannakian duality, or rather as Tannakian duality in "soft form", is something quite deep, and extremely useful. Indeed, the idea is that what we would have here is a "delinearization" of G, allowing us to do combinatorics, and to ultimately reach to very concrete and powerful results, regarding G itself.

The construction  $C \to G$  is in fact something very simple, as follows:

**THEOREM** 14.4. Given an abstract tensor category

$$C = (C(k, l))$$

over a finite dimensional Hilbert space  $H \simeq \mathbb{C}^N$ , the following construction,

$$G = \left\{ g \in U_N \middle| Tg^{\otimes k} = g^{\otimes l}T , \ \forall k, l, \forall T \in C(k, l) \right\}$$

produces a closed subgroup  $G \subset U_N$ .

PROOF. This is something elementary, with the fact that the closed subset  $G \subset U_N$  constructed in the statement being stable under the multiplication, unit and inversion operation for the unitary matrices  $g \in U_N$  being clear from definitions.

Summarizing, we have so far precise axioms for the tensor categories C = (C(k, l)), given in Definition 14.2 above, as well as two theorems, 14.3 and 14.4, which are quite elementary, and provide us with correspondences as follows:

$$G \to C$$
 ,  $C \to G$ 

We will show in what follows that these correspondences are inverse to each other, and so that, as a final conclusion, we have a one-to-one correspondence, as follows:

$$G \leftrightarrow C$$

In order to get started, we have the following result:

THEOREM 14.5. If we denote the correspondences in Theorem 14.3 and 14.4, between closed subgroups  $G \subset U_N$  and tensor categories C = (C(k, l)) over  $H = \mathbb{C}^N$ , as

 $G \to C_G$  ,  $C \to G_C$ 

then we have embeddings as follows, for any G and C respectively,

$$G \subset G_{C_G}$$
$$C \subset C_{G_G}$$

and proving that these correspondences are inverse to each other amounts in proving

$$C_{G_C} \subset C$$

for any tensor category C = (C(k, l)) over the space  $H = \mathbb{C}^N$ .

**PROOF.** This is something trivial, with the embeddings  $G \subset G_{C_G}$  and  $C \subset C_{G_C}$  being both clear from definitions, and with the last assertion coming from this.

Summarizing, after all these preliminaries, and discussions about motivations and methods and everything, we have now a concrete question to be solved, namely proving that for any Tannakian category C, we have an inclusion as follows:

$$C_{G_C} \subset C$$

The bad news, however, is that proving that this inclusion holds is something nontrivial, requiring a lot of work, and some clever ideas as well. To be more precise, we will first have to do some straightforward algebra, in order to further reformulate what we want to prove, in a more convenient way. Then, we will have to go into something more delicate, namely understanding how the correspondence  $C \rightarrow C_G$  really works. And after all this done, we will be able to establish Tannakian duality, in its formulation above.

## 14b. Abstract algebra

In order to establish Tannakian duality, we will need some abstract constructions. Following [52], let us start with the following elementary fact:

**PROPOSITION 14.6.** Given a tensor category C = C((k, l)) over a Hilbert space H,

$$E_C^{(s)} = \bigoplus_{|k|,|l| \le s} C(k,l)$$
$$\subset \bigoplus_{|k|,|l| \le s} B(H^{\otimes k}, H^{\otimes l})$$
$$= B\left(\bigoplus_{|k| \le s} H^{\otimes k}\right)$$

is a finite dimensional \*-subalgebra. Also,

$$E_C = \bigoplus_{k,l} C(k,l)$$
$$\subset \bigoplus_{k,l} B(H^{\otimes k}, H^{\otimes l})$$
$$\subset B\left(\bigoplus_k H^{\otimes k}\right)$$

is a closed \*-subalgebra.

PROOF. This is clear indeed from the categorical axioms from Definition 14.2 above, which, since satisfied, prove that the various linear spaces in the statement are stable under both the multiplication operation, and under taking the adjoints.  $\Box$ 

Now back to our reconstruction question, we want to prove  $C = C_{G_C}$ , which is the same as proving  $E_C = E_{C_{G_C}}$ . Equivalently, we want to prove that, for any  $s \in \mathbb{N}$ :

$$E_C^{(s)} = E_{C_{G_C}}^{(s)}$$

We will use a standard commutant trick, as follows:

THEOREM 14.7. For any \*-algebra  $A \subset M_n(\mathbb{C})$  we have

$$A = A''$$

where prime denotes the commutant, constructed as follows:

$$X' = \left\{ T \in M_n(\mathbb{C}) \middle| Tx = xT, \forall x \in X \right\}$$

PROOF. This is a particular case of von Neumann's bicommutant theorem, which follows from the explicit description of A given in chapter 13. To be more precice, we know from chapter 13 above that our algebra  $A \subset M_n(\mathbb{C})$  decomposes a follows:

$$A = M_{r_1}(\mathbb{C}) \oplus \ldots \oplus M_{r_k}(\mathbb{C})$$

The center of each matrix algebra being reduced to the scalars, the commutant of this algebra is then as follows, with each copy of  $\mathbb{C}$  corresponding to a matrix block:

$$A' = \mathbb{C} \oplus \ldots \oplus \mathbb{C}$$

Now when taking once again the commutant, things are trivial, and we obtain in this way A itself, and this leads to the conclusion in the statement.

As a side remark here, as already mentioned above, the about result is a particular case of a famous result in functional analysis, called von Neumann's bicommutant theorem. This theorem states that given a Hilbert space H, not necessarily finite dimensional, and a \*-algebra of bounded operators  $A \subset B(H)$ , this algebra is weakly closed, in the sense that it is closed with respect to the topology making continuous the evaluation maps  $T \to Tx$  with  $x \in H$ , if and only if this algebra is equal to its algebraic bicommutant, A = A''. For more on all this, we recommend any good operator algebra book.

Getting back to work now, we recall from the above that, in order to establish Tannakian duality, we want to prove that we have, for any Tannakian category C:

$$C = C_{G_C}$$

By using the above notions, and the bicommutant theorem, we have:

**PROPOSITION 14.8.** Given a Tannakian category C, the following are equivalent:

(1) 
$$C = C_{G_C}$$
.  
(2)  $E_C = E_{C_{G_C}}$ .  
(3)  $E_C^{(s)} = E_{C_{G_C}}^{(s)}$ , for any  $s \in \mathbb{N}$ .  
(4)  $E_C^{(s)'} = E_{C_{G_C}}^{(s)'}$ , for any  $s \in \mathbb{N}$ .

In addition, the inclusions  $\subset$ ,  $\subset$ ,  $\subset$ ,  $\supset$  are automatically satisfied.

**PROOF.** This follows from the above results, as follows:

(1)  $\iff$  (2) This is clear from definitions.

(2)  $\iff$  (3) This is clear from definitions as well.

(3)  $\iff$  (4) This comes from the bicommutant theorem. As for the last assertion, we have indeed  $C \subset C_{G_C}$  from Theorem 14.5, and this shows that we have as well:

$$E_C \subset E_{C_{G_C}}$$

We therefore obtain by truncating  $E_C^{(s)} \subset E_{C_{G_C}}^{(s)}$ , and by taking the commutants, this gives  $E_C^{(s)} \supset E_{C_{G_C}}^{(s)}$ . Thus, we are led to the conclusion in the statement.

Summarizing, we would like to prove that we have  $E_C^{(s)'} \subset E_{C_{G_C}}^{(s)'}$ . Let us first study the commutant on the right. As a first observation, we have:

**PROPOSITION 14.9.** We have the following equality

$$E_{C_G}^{(s)} = End\left(\bigoplus_{|k| \le s} u^{\otimes k}\right)$$

between subalgebras of the following algebra:

$$B\left(\bigoplus_{|k|\leq s}H^{\otimes k}\right)$$

**PROOF.** The category  $C_G$  is by definition given by:

$$C_G(k,l) = Hom(u^{\otimes k}, u^{\otimes l})$$

Thus, the corresponding algebra  $E_{C_G}^{(s)}$  appears as follows:

$$E_{C_G}^{(s)} = \bigoplus_{|k|,|l| \le s} Hom(u^{\otimes k}, u^{\otimes l})$$
$$\subset \bigoplus_{|k|,|l| \le s} B(H^{\otimes k}, H^{\otimes l})$$
$$= B\left(\bigoplus_{|k| \le s} H^{\otimes k}\right)$$

On the other hand, the algebra of intertwiners of  $\bigoplus_{|k| \le s} u^{\otimes k}$  is given by:

$$End\left(\bigoplus_{|k|\leq s} u^{\otimes k}\right) = \bigoplus_{|k|,|l|\leq s} Hom(u^{\otimes k}, u^{\otimes l})$$
$$\subset \bigoplus_{|k|,|l|\leq s} B(H^{\otimes k}, H^{\otimes l})$$
$$= B\left(\bigoplus_{|k|\leq s} H^{\otimes k}\right)$$

Thus we have indeed the same algebra, and we are done.

We have to compute the commutant of the above algebra. For this purpose, we can use the following general result, valid for any representation of a compact group:

**PROPOSITION** 14.10. Given a unitary group representation

$$v: G \to U_n$$

we have an algebra representation as follows,

$$\pi_v : C(G)^* \to M_n(\mathbb{C})$$
$$\varphi \to (\varphi(v_{ij}))_{ij}$$

whose image is given by:

$$Im(\pi_v) = End(v)'$$

**PROOF.** The first assertion is clear, with the multiplicativity claim for  $\pi_v$  coming from the following computation, where  $\Delta : C(G) \to C(G) \otimes C(G)$  is the comultiplication:

$$(\pi_v(\varphi * \psi))_{ij} = (\varphi \otimes \psi)\Delta(v_{ij})$$
  
=  $\sum_k \varphi(v_{ik})\psi(v_{kj})$   
=  $\sum_k (\pi_v(\varphi))_{ik}(\pi_v(\psi))_{kj}$   
=  $(\pi_v(\varphi)\pi_v(\psi))_{ij}$ 

Let us establish now the equality in the statement, namely:

$$Im(\pi_v) = End(v)'$$

Let us first prove the inclusion  $\subset$ . Given  $\varphi \in C(G)^*$  and  $T \in End(v)$ , we have:

$$[\pi_{v}(\varphi), T] = 0 \iff \sum_{k} \varphi(v_{ik}) T_{kj} = \sum_{k} T_{ik} \varphi(v_{kj}), \forall i, j$$
$$\iff \varphi\left(\sum_{k} v_{ik} T_{kj}\right) = \varphi\left(\sum_{k} T_{ik} v_{kj}\right), \forall i, j$$
$$\iff \varphi((vT)_{ij}) = \varphi((Tv)_{ij}), \forall i, j$$

But this latter formula is true, because  $T \in End(v)$  means that we have:

$$vT = Tv$$

As for the converse inclusion  $\supset$ , the proof is quite similar. Indeed, by using the bicommutant theorem, this is the same as proving that we have:

$$Im(\pi_v)' \subset End(v)$$

But, by using the above equivalences, we have the following computation:

$$T \in Im(\pi_v)' \iff [\pi_v(\varphi), T] = 0, \forall \varphi$$
$$\iff \varphi((vT)_{ij}) = \varphi((Tv)_{ij}), \forall \varphi, i, j$$
$$\iff vT = Tv$$

Thus, we have obtained the desired inclusion, and we are done.

By combining the above results, we obtain the following technical statement:

THEOREM 14.11. We have the following equality

$$E_{C_G}^{(s)'} = Im(\pi_v)$$

as subalgebras of the following algebra,

$$B\left(\bigoplus_{|k|\leq s}H^{\otimes k}\right)$$

where the representation v is the following direct sum,

$$v = \bigoplus_{|k| \le s} u^{\otimes k}$$

and where the algebra representation  $\pi_v : C(G)^* \to M_n(\mathbb{C})$  is given by:

$$\varphi \to (\varphi(v_{ij}))_{ij}$$

PROOF. This follows indeed by combining the above results, and more precisely by combining Proposition 14.9 and Proposition 14.10.  $\hfill \Box$ 

We recall that we want to prove that we have  $E_C^{(s)'} \subset E_{C_{G_C}}^{(s)'}$ , for any  $s \in \mathbb{N}$ . For this purpose, we must first refine Theorem 14.11, in the case  $G = G_C$ .

## 14c. The correspondence

Generally speaking, in order to prove anything about  $G_C$ , we are in need of an explicit model for this group. In order to construct such a model, let  $\langle u_{ij} \rangle$  be the free \*-algebra over dim $(H)^2$  variables, with comultiplication and counit as follows:

$$\Delta(u_{ij}) = \sum_{k} u_{ik} \otimes u_{kj}$$
$$\varepsilon(u_{ij}) = \delta_{ij}$$

Following [52], we can model this \*-bialgebra, in the following way:

**PROPOSITION** 14.12. Consider the following pair of dual vector spaces,

$$F = \bigoplus_{k} B(H^{\otimes k})$$
$$F^* = \bigoplus_{k} B(H^{\otimes k})^*$$

and let  $f_{ij}, f_{ij}^* \in F^*$  be the standard generators of  $B(H)^*, B(\bar{H})^*$ .

(1)  $F^*$  is a \*-algebra, with multiplication  $\otimes$  and involution as follows:

$$f_{ij} \leftrightarrow f_{ij}^*$$

(2)  $F^*$  is a \*-bialgebra, with \*-bialgebra operations as follows:

$$\Delta(f_{ij}) = \sum_{k} f_{ik} \otimes f_{kj}$$

$$\varepsilon(f_{ij}) = \delta_{ij}$$

(3) We have a \*-bialgebra isomorphism  $\langle u_{ij} \rangle \simeq F^*$ , given by  $u_{ij} \to f_{ij}$ .

**PROOF.** Since  $F^*$  is spanned by the various tensor products between the variables  $f_{ij}, f_{ij}^*$ , we have a vector space isomorphism as follows:

$$< u_{ij} > \simeq F^*$$
  
 $u_{ij} \to f_{ij}$   
 $u_{ij}^* \to f_{ij}^*$ 

The corresponding \*-bialgebra structure induced on the vector space  $F^*$  is then the one in the statement, and this gives the result.

Now back to our group  $G_C$ , we have the following modelling result for it:

**PROPOSITION 14.13.** The smooth part of the algebra  $A_C = C(G_C)$  is given by

$$\mathcal{A}_C \simeq F^*/J$$

where  $J \subset F^*$  is the ideal coming from the following relations,

$$\sum_{p_1,\dots,p_k} T_{i_1\dots i_l,p_1\dots p_k} f_{p_1 j_1} \otimes \dots \otimes f_{p_k j_k}$$
$$= \sum_{q_1,\dots,q_l} T_{q_1\dots q_l,j_1\dots j_k} f_{i_1 q_1} \otimes \dots \otimes f_{i_l q_l} \quad , \quad \forall i,j$$

one for each pair of colored integers k, l, and each  $T \in C(k, l)$ .

**PROOF.** As a first observation,  $A_C$  appears as enveloping  $C^*$ -algebra of the following universal \*-algebra, where  $u = (u_{ij})$  is regarded as a formal corepresentation:

$$\mathcal{A}_{C} = \left\langle (u_{ij})_{i,j=1,\dots,N} \middle| T \in Hom(u^{\otimes k}, u^{\otimes l}), \forall k, l, \forall T \in C(k,l) \right\rangle$$

With this observation in hand, the conclusion is that we have a formula as follows, where I is the ideal coming from the relations  $T \in Hom(u^{\otimes k}, u^{\otimes l})$ , with  $T \in C(k, l)$ :

$$\mathcal{A}_C = < u_{ij} > /I$$

Now if we denote by  $J \subset F^*$  the image of the ideal I via the \*-algebra isomorphism  $\langle u_{ij} \rangle \simeq F^*$  from Proposition 14.15, we obtain an identification as follows:

$$\mathcal{A}_C \simeq F^*/J$$

In order to compute J, let us go back to I.

With standard multi-index notations, and by assuming now that  $k, l \in \mathbb{N}$  are usual integers, for simplifying the presentation, the general case being similar, a relation of type  $T \in Hom(u^{\otimes k}, u^{\otimes l})$  inside  $\langle u_{ij} \rangle$  is equivalent to the following conditions:

$$\sum_{p_1,\dots,p_k} T_{i_1\dots i_l,p_1\dots p_k} u_{p_1 j_1} \dots u_{p_k j_k}$$
  
=  $\sum_{q_1,\dots,q_l} T_{q_1\dots q_l,j_1\dots j_k} u_{i_1 q_1} \dots u_{i_l q_l}$ ,  $\forall i$ ,

j

Now by recalling that the isomorphism of \*-algebras  $\langle u_{ij} \rangle \rightarrow F^*$  is given by  $u_{ij} \rightarrow f_{ij}$ , and that the multiplication operation of  $F^*$  corresponds to the tensor product operation  $\otimes$ , we conclude that  $J \subset F^*$  is the ideal from the statement.

With the above result in hand, let us go back to Theorem 14.11. We have:

**PROPOSITION 14.14.** The linear space  $\mathcal{A}_C^*$  is given by the formula

$$\mathcal{A}_{C}^{*} = \left\{ a \in F \middle| Ta_{k} = a_{l}T, \forall T \in C(k, l) \right\}$$

and the representation

$$\pi_v: \mathcal{A}_C^* \to B\left(\bigoplus_{|k| \le s} H^{\otimes k}\right)$$

appears diagonally, by truncating:

$$\pi_v: a \to (a_k)_{kk}$$

**PROOF.** We know from Proposition 14.13 above that we have an identification of \*-bialgebras as follows:

$$\mathcal{A}_C \simeq F^*/J$$

But this gives a quotient map as follows:

$$F^* \to \mathcal{A}_C$$

At the dual level now, we obtain an inclusion as follows:

$$\mathcal{A}_C^* \subset F$$

To be more precise, we have the following equality:

$$\mathcal{A}_{C}^{*} = \left\{ a \in F \middle| f(a) = 0, \forall f \in J \right\}$$

Now since  $J = \langle f_T \rangle$ , where  $f_T$  are the relations in Proposition 14.13, we obtain:

$$\mathcal{A}_{C}^{*} = \left\{ a \in F \middle| f_{T}(a) = 0, \forall T \in C \right\}$$

Given  $T \in C(k, l)$ , for an arbitrary element  $a = (a_k)$ , we have:

$$f_T(a) = 0$$

$$\iff \sum_{p_1,\dots,p_k} T_{i_1\dots i_l,p_1\dots p_k}(a_k)_{p_1\dots p_k,j_1\dots j_k} = \sum_{q_1,\dots,q_l} T_{q_1\dots q_l,j_1\dots j_k}(a_l)_{i_1\dots i_l,q_1\dots q_l}, \forall i,j$$

$$\iff (Ta_k)_{i_1\dots i_l,j_1\dots j_k} = (a_l T)_{i_1\dots i_l,j_1\dots j_k}, \forall i,j$$

$$\iff Ta_k = a_l T$$

Thus, the dual space  $\mathcal{A}_C^*$  is given by the formula in the statement. It remains to compute the representation  $\pi_v$ , which appears as follows:

$$\pi_v: \mathcal{A}_C^* \to B\left(\bigoplus_{|k| \le s} H^{\otimes k}\right)$$

With  $a = (a_k)$ , we have the following computation:

$$\pi_{v}(a)_{i_{1}...i_{k},j_{1}...j_{k}} = a(v_{i_{1}...i_{k},j_{1}...j_{k}})$$
  
=  $(f_{i_{1}j_{1}} \otimes ... \otimes f_{i_{k}j_{k}})(a)$   
=  $(a_{k})_{i_{1}...i_{k},j_{1}...j_{k}}$ 

Thus, our representation  $\pi_v$  appears diagonally, by truncating, as claimed. In order to further advance, consider the following vector spaces:

$$F_{s} = \bigoplus_{|k| \le s} B\left(H^{\otimes k}\right)$$
$$F_{s}^{*} = \bigoplus_{|k| \le s} B\left(H^{\otimes k}\right)^{*}$$

We denote by  $a \to a_s$  the truncation operation  $F \to F_s$ .

With these conventions, we have the following result:

**PROPOSITION 14.15.** The following hold:

(1)  $E_C^{(s)'} \subset F_s.$ (2)  $E_C' \subset F.$ (3)  $\mathcal{A}_C^* = E_C'.$ (4)  $Im(\pi_v) = (E_C')_s.$ 

**PROOF.** These results basically follow from what we have, as follows:

(1) We have an inclusion as follows, as a diagonal subalgebra:

$$F_s \subset B\left(\bigoplus_{|k|\leq s} H^{\otimes k}\right)$$

The commutant of this algebra is:

$$F'_{s} = \left\{ b \in F_{s} \middle| b = (b_{k}), b_{k} \in \mathbb{C}, \forall k \right\}$$

On the other hand, we know from the identity axiom for C that this algebra is contained inside  $E_C^{(s)}$ :

$$F'_s \subset E_C^{(s)}$$

Thus, our result follows from the bicommutant theorem, as follows:

$$F'_s \subset E_C^{(s)} \implies F_s \supset E_C^{(s)'}$$

- (2) This follows from (1), by taking inductive limits.
- (3) With the present notations, the formula of  $\mathcal{A}_C^*$  from Proposition 14.14 reads:

$$\mathcal{A}_C^* = F \cap E_C'$$

Now since by (2) we have  $E'_C \subset F$ , we obtain from this  $\mathcal{A}^*_C = E'_C$ .

(4) This follows from (3), and from the formula of  $\pi_{\nu}$  in Proposition 14.14.

Following [52], we can now state and prove our main result, as follows:

THEOREM 14.16. The Tannakian duality constructions

$$C \to G_C$$
$$G \to C_G$$

are inverse to each other.

**PROOF.** According to our various results above, we have to prove that, for any Tannakian category C, and any  $s \in \mathbb{N}$ , we have an inclusion as follows:

$$E_C^{(s)'} \subset (E_C')_s$$

By taking duals, this is the same as proving that we have:

$$\left\{ f \in F_s^* \middle| f_{|(E'_C)_s} = 0 \right\} \subset \left\{ f \in F_s^* \middle| f_{|E_C^{(s)'}} = 0 \right\}$$

In order to do so, we use the following formula, from Proposition 14.15:

$$\mathcal{A}_C^* = E_C'$$

We know that we have an identification as follows:

$$\mathcal{A}_C = F^*/J$$

We conclude that the ideal J is given by:

$$J = \left\{ f \in F^* \middle| f_{|E'_C} = 0 \right\}$$

Our claim is that we have the following formula, for any  $s \in \mathbb{N}$ :

$$J \cap F_s^* = \left\{ f \in F_s^* \middle| f_{|E_C^{(s)'}} = 0 \right\}$$

Indeed, let us denote by  $X_s$  the spaces on the right. The categorical axioms for C show that these spaces are increasing, that their union  $X = \bigcup_s X_s$  is an ideal, and that:

$$X_s = X \cap F_s^*$$

We must prove that we have J = X, and this can be done as follows:

" $\subset$ " This follows from the following fact, for any  $T \in C(k, l)$  with  $|k|, |l| \leq s$ :

$$(f_T)_{|\{T\}'} = 0 \implies (f_T)_{|E_C^{(s)'}} = 0$$
  
 $\implies f_T \in X_s$ 

" $\supset$ " This follows from our description of J, because from  $E_C^{(s)} \subset E_C$  we obtain:

$$f_{|E_C^{(s)'}} = 0 \implies f_{|E_C'} = 0$$

Summarizing, we have proved our claim. On the other hand, we have:

$$J \cap F_s^* = \left\{ f \in F^* \middle| f_{|E'_C} = 0 \right\} \cap F_s^*$$
$$= \left\{ f \in F_s^* \middle| f_{|E'_C} = 0 \right\}$$
$$= \left\{ f \in F_s^* \middle| f_{|(E'_C)_s} = 0 \right\}$$

Thus, our claim is exactly the inclusion that we wanted to prove, and we are done.  $\Box$ 

Summarizing, we have proved Tannakian duality. There are many other versions of this duality, following the original work of Tannaka and Krein, and the more recent work of Deligne and Doplicher-Roberts, and the work of Jones as well.

In what follows we will use Theorem 14.16 as such, with this formulation being well adpated to our motivations here, and to the examples that we have in mind.

For the discussion to be complete, however, let us briefly indicate the story of Tannakian duality, and of its various versions. The situation here is basically as follows:

- (1) The original version of the duality, due to Tannaka and Krein, states that any compact group G can be recovered from the knowledge of its category of representations  $\mathcal{R}_G$ , viewed as subcategory of the category  $\mathcal{H}$  of the finite dimensional Hilbert spaces, with each  $v \in \mathcal{R}_G$  corresponding to its Hilbert space  $H_v \in \mathcal{H}$ .
- (2) Regarding the proof of this fact, this is something established long ago by Tannaka and Krein. From a more modern perspective, coming from the work of Grothendieck, Saavedra and others, the group G simply appears as the group of endomorphisms of the embedding functor  $\mathcal{R}_G \subset \mathcal{H}$ . And that's all.
- (3) With this understood, and even labelled as "trivial", comes now a non-trivial result, due independently to Doplicher-Roberts and Deligne, stating that the compact group G can be recovered from the sole knowledge of the category  $\mathcal{R}_G$ . This is obviously something more advanced, and the proof is quite tricky.
- (4) Getting back to Earth now, as already explained in the above, for concrete applications it is technically convenient to replace the full category  $\mathcal{R}_G$  by its subcategory  $\mathcal{R}_G^\circ$  consisting of the Peter-Weyl representations. And here, we have the analogue of the original result of Tannaka-Krein, explained in the above.
- (5) For more specialized questions, it is convenient to further shrink the Peter-Weyl category  $\mathcal{R}_G^{\circ}$ , consisting of the spaces  $C(k, l) = Hom(u^{\otimes k}, u^{\otimes l})$ , to its diagonal algebra  $P_G = \Delta \mathcal{R}_G$ , consisting of the \*-algebras  $C(k, k) = End(u^{\otimes k})$ . This is the idea behind Jones' planar algebras, with  $P_G$  being such an algebra.
- (6) And this is not the end of the story, because one can still try to have Doplicher-Roberts and Deligne type results in the Peter-Weyl category setting, or in the planar algebra setting. We refer here to the modern quantum algebra literature, where Tannakian duality, in all its forms, is something highly valued.

Summarizing, many things to be learned, as a continuation of what was explained in the above. Importantly, Tannakian duality of all sorts is just a way of "linearizing" the group, and for linearizing there is as well a second method, namely considering the associated Lie algebra. So, the above long story is in fact only half of the story.

Again, we refer here to the modern quantum algebra literature.

## 14d. Brauer theorems

As a basic illustration for the Tannakian correspondence found above, we will work out the Brauer theorems for  $O_N, U_N$ . These theorems provide a very explicit description of the categories associated to these groups, as certain algebras formed by diagrams.

Let us start with the following definition:

DEFINITION 14.17. Given a pairing  $\pi \in P_2(k,l)$  and an integer  $N \in \mathbb{N}$ , we can construct a linear map between tensor powers of  $\mathbb{C}^N$ ,

$$T_{\pi}: (\mathbb{C}^N)^{\otimes k} \to (\mathbb{C}^N)^{\otimes k}$$

by the following formula, with  $e_1, \ldots, e_N$  being the standard basis of  $\mathbb{C}^N$ ,

$$T_{\pi}(e_{i_1} \otimes \ldots \otimes e_{i_k}) = \sum_{j_1 \dots j_l} \delta_{\pi} \begin{pmatrix} i_1 & \cdots & i_k \\ j_1 & \cdots & j_l \end{pmatrix} e_{j_1} \otimes \ldots \otimes e_{j_l}$$

and with the coefficients on the right being Kronecker type symbols,

$$\delta_{\pi} \begin{pmatrix} i_1 & \dots & i_k \\ j_1 & \dots & j_l \end{pmatrix} \in \{0, 1\}$$

whose values depend on whether the indices fit or not.

To be more precise here, in order to compute the Kronecker type symbols  $\delta_{\pi}({}^{i}_{j})$ , we put the multi-indices i, j on the legs of  $\pi$ , in the obvious way. In the case where any strings of  $\pi$  joins a pair of equal indices of i, j, we set  $\delta_{\pi}({}^{i}_{j}) = 1$ . Otherwise, we set  $\delta_{\pi}({}^{i}_{j}) = 0$ .

Here are a few basic examples of such linear maps:

**PROPOSITION 14.18.** The correspondence  $\pi \to T_{\pi}$  has the following properties:

(1)  $T_{\cap} = (1 \rightarrow \sum_{i} e_i \otimes e_i).$ (2)  $T_{\cup} = (e_i \otimes e_j \rightarrow \delta_{ij}).$ (3)  $T_{||\dots||} = id.$ (4)  $T_{\chi} = (a \otimes b \rightarrow b \otimes a).$ 

**PROOF.** We can assume if we want that all the upper and lower legs of  $\pi$  are colored  $\circ$ . With this assumption made, the proof goes as follows:

(1) We have  $\cap \in P_2(\emptyset, \infty)$ , so the corresponding linear map is as follows:

$$T_{\cap}:\mathbb{C}\to\mathbb{C}^N\otimes\mathbb{C}^N$$

The formula of this map is as follows, as claimed:

$$T_{\cap}(1) = \sum_{ij} \delta_{\cap}(i \ j) e_i \otimes e_j$$
$$= \sum_{ij} \delta_{ij} e_i \otimes e_j$$
$$= \sum_i e_i \otimes e_i$$

(2) Here we have  $\cup \in P_2(\circ\circ, \emptyset)$ , so the corresponding linear map is as follows:

$$T_{\cap}:\mathbb{C}^N\otimes\mathbb{C}^N\to\mathbb{C}$$

The formula of this linear form is then as follows:

$$T_{\cap}(e_i \otimes e_j) = \delta_{\cap}(i \ j) \\ = \delta_{ij}$$

(3) Consider indeed the "identity" pairing  $|| \dots || \in P_2(k, k)$ , with  $k = \circ \circ \dots \circ \circ$ . The corresponding linear map is then the identity, because we have:

$$T_{||\dots||}(e_{i_{1}} \otimes \dots \otimes e_{i_{k}})$$

$$= \sum_{j_{1}\dots j_{k}} \delta_{||\dots||} \begin{pmatrix} i_{1} & \dots & i_{k} \\ j_{1} & \dots & j_{k} \end{pmatrix} e_{j_{1}} \otimes \dots \otimes e_{j_{k}}$$

$$= \sum_{j_{1}\dots j_{k}} \delta_{i_{1}j_{1}} \dots \delta_{i_{k}j_{k}} e_{j_{1}} \otimes \dots \otimes e_{j_{k}}$$

$$= e_{i_{1}} \otimes \dots \otimes e_{i_{k}}$$

(4) For the basic crossing  $X \in P_2(\circ\circ, \circ\circ)$ , the corresponding linear map is as follows:

$$T_{\chi}:\mathbb{C}^N\otimes\mathbb{C}^N\to\mathbb{C}^N\otimes\mathbb{C}^N$$

This map can be computed as follows:

$$T_{\chi}(e_i \otimes e_j) = \sum_{kl} \delta_{\chi} \begin{pmatrix} i & j \\ k & l \end{pmatrix} e_k \otimes e_l$$
$$= \sum_{kl} \delta_{il} \delta_{jk} e_k \otimes e_l$$
$$= e_j \otimes e_i$$

Thus we obtain the flip operator  $\Sigma(a \otimes b) = b \otimes a$ , as claimed.

The relation with the Tannakian categories, as axiomatized in the above, comes from the following key result, regarding the operation  $\pi \to T_{\pi}$  from Definition 14.17:

THEOREM 14.19. The assignment constructed above

$$\pi \to T_{\pi}$$

is categorical, in the sense that we have the formulae

$$T_{\pi} \otimes T_{\sigma} = T_{[\pi\sigma]}$$
$$T_{\pi}T_{\sigma} = N^{c(\pi,\sigma)}T_{[\pi]}$$
$$T_{\pi}^{*} = T_{\pi^{*}}$$

where  $c(\pi, \sigma)$  are certain integers, coming from the erased components in the middle.

**PROOF.** This follows from some routine computations, as follows:

(1) The concatenation axiom follows from the following computation:

$$(T_{\pi} \otimes T_{\sigma})(e_{i_{1}} \otimes \ldots \otimes e_{i_{p}} \otimes e_{k_{1}} \otimes \ldots \otimes e_{k_{r}})$$

$$= \sum_{j_{1} \ldots j_{q}} \sum_{l_{1} \ldots l_{s}} \delta_{\pi} \begin{pmatrix} i_{1} & \ldots & i_{p} \\ j_{1} & \ldots & j_{q} \end{pmatrix} \delta_{\sigma} \begin{pmatrix} k_{1} & \ldots & k_{r} \\ l_{1} & \ldots & l_{s} \end{pmatrix} e_{j_{1}} \otimes \ldots \otimes e_{j_{q}} \otimes e_{l_{1}} \otimes \ldots \otimes e_{l_{s}}$$

$$= \sum_{j_{1} \ldots j_{q}} \sum_{l_{1} \ldots l_{s}} \delta_{[\pi\sigma]} \begin{pmatrix} i_{1} & \ldots & i_{p} & k_{1} & \ldots & k_{r} \\ j_{1} & \ldots & j_{q} & l_{1} & \ldots & l_{s} \end{pmatrix} e_{j_{1}} \otimes \ldots \otimes e_{j_{q}} \otimes e_{l_{1}} \otimes \ldots \otimes e_{l_{s}}$$

$$= T_{[\pi\sigma]}(e_{i_{1}} \otimes \ldots \otimes e_{i_{p}} \otimes e_{k_{1}} \otimes \ldots \otimes e_{k_{r}})$$

(2) The composition axiom follows from the following computation:  $\overline{T}$ 

$$T_{\pi}T_{\sigma}(e_{i_{1}}\otimes\ldots\otimes e_{i_{p}})$$

$$= \sum_{j_{1}\ldots j_{q}}\delta_{\sigma}\begin{pmatrix}i_{1}&\ldots&i_{p}\\j_{1}&\ldots&j_{q}\end{pmatrix}\sum_{k_{1}\ldots k_{r}}\delta_{\pi}\begin{pmatrix}j_{1}&\ldots&j_{q}\\k_{1}&\ldots&k_{r}\end{pmatrix}e_{k_{1}}\otimes\ldots\otimes e_{k_{r}}$$

$$= \sum_{k_{1}\ldots k_{r}}N^{c(\pi,\sigma)}\delta_{[\frac{\sigma}{\pi}]}\begin{pmatrix}i_{1}&\ldots&i_{p}\\k_{1}&\ldots&k_{r}\end{pmatrix}e_{k_{1}}\otimes\ldots\otimes e_{k_{r}}$$

$$= N^{c(\pi,\sigma)}T_{[\frac{\sigma}{\pi}]}(e_{i_{1}}\otimes\ldots\otimes e_{i_{p}})$$

(3) Finally, the involution axiom follows from the following computation:

$$T_{\pi}^{*}(e_{j_{1}} \otimes \ldots \otimes e_{j_{q}})$$

$$= \sum_{i_{1}\ldots i_{p}} < T_{\pi}^{*}(e_{j_{1}} \otimes \ldots \otimes e_{j_{q}}), e_{i_{1}} \otimes \ldots \otimes e_{i_{p}} > e_{i_{1}} \otimes \ldots \otimes e_{i_{p}}$$

$$= \sum_{i_{1}\ldots i_{p}} \delta_{\pi} \begin{pmatrix} i_{1} & \ldots & i_{p} \\ j_{1} & \ldots & j_{q} \end{pmatrix} e_{i_{1}} \otimes \ldots \otimes e_{i_{p}}$$

$$= T_{\pi^{*}}(e_{j_{1}} \otimes \ldots \otimes e_{j_{q}})$$

Summarizing, our correspondence is indeed categorical.

The above result suggests the following general definition, from [11]:

DEFINITION 14.20. Let  $P_2(k, l)$  be the set of pairings between an upper colored integer k, and a lower colored integer l. A collection of subsets

$$D = \bigsqcup_{k,l} D(k,l)$$

with  $D(k,l) \subset P_2(k,l)$  is called a category of pairings when it has the following properties:

- (1) Stability under the horizontal concatenation,  $(\pi, \sigma) \rightarrow [\pi\sigma]$ .
- (2) Stability under vertical concatenation  $(\pi, \sigma) \to [\sigma]$ , with matching middle symbols.
- (3) Stability under the upside-down turning \*, with switching of colors,  $\circ \leftrightarrow \bullet$ .
- (4) Each set P(k,k) contains the identity partition  $|| \dots ||$ .
- (5) The sets  $P(\emptyset, \bullet \bullet)$  and  $P(\emptyset, \bullet \circ)$  both contain the semicircle  $\cap$ .
- (6) The sets  $P(k, \bar{k})$  with |k| = 2 contain the crossing partition  $\chi$ .

Observe the similarity with the axioms for Tannakian categories, from the beginning of this chapter. We will see in a moment that this similarity can be turned into something very precise, with the categories of pairings producing Tannakian categories.

As basic examples of such categories of pairings, that we have already met in the above, we have the categories  $P_2, \mathcal{P}_2$  of pairings, and of matching pairings.

In relation with the compact groups, we have the following result:

THEOREM 14.21. Each category of pairings, in the above sense,

$$D = (D(k, l))$$

produces a family of compact groups  $G = (G_N)$ , one for each  $N \in \mathbb{N}$ , via the formula

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

and the Tannakian duality correspondence.

**PROOF.** Given an integer  $N \in \mathbb{N}$ , consider the correspondence  $\pi \to T_{\pi}$  constructed in Definition 14.17, and then the collection of linear spaces in the statement, namely:

$$C_{kl} = span\left(T_{\pi} \middle| \pi \in D(k,l)\right)$$

According to Theorem 14.19, and to our axioms for the categories of partitions, from Definition 14.20, this collection of spaces  $C = (C_{kl})$  satisfies the axioms for the Tannakian categories, from the beginning of this chapter. Thus the Tannakian duality result there applies, and provides us with a closed subgroup  $G_N \subset U_N$  such that:

$$C_{kl} = Hom(u^{\otimes k}, u^{\otimes l})$$

Thus, we are led to the conclusion in the statement.

#### 14D. BRAUER THEOREMS

The compact groups constructed above are called "easy", and this due to the fact that, from the point of view of Tannakian duality, these groups are indeed the "easiest" possible ones. We will be back to them, with a systematic study, in chapter 15 below.

As a preliminary remark, however, which is of theoretical nature, and that we should keep all the time in mind in dealing with such things, the category of pairings producing an easy group is not unique, and this for instance because at N = 1 all the possible categories of pairings produce the same easy group, namely the trivial group  $G = \{1\}$ . Thus, there are some subtleties going on here. More on this later.

Getting back now to concrete things, the point now is that with the above ingredients in hand, and as a first application of Tannakian duality, we can establish a useful result, namely the Brauer theorem for the unitary group  $U_N$ . The statement is a follows:

THEOREM 14.22. For the unitary group  $U_N$  we have

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in \mathcal{P}_{2}(k, l)\right)$$

where  $\mathcal{P}_2$  denotes as usual the category of all matching pairings.

**PROOF.** Consider the spaces on the right in the statement, namely:

$$C_{kl} = span\left(T_{\pi} \middle| \pi \in \mathcal{P}_2(k,l)\right)$$

According to Theorem 14.19 these spaces form a tensor category. Thus, according to Tannakian duality, these spaces must come from a certain closed subgroup, as follows:

$$G \subset U_N$$

To be more precise, if we denote by v the fundamental representation of G, then:

$$C_{kl} = Hom(v^{\otimes k}, v^{\otimes l})$$

We must prove that we have  $G = U_N$ . For this purpose, let us recall that the unitary group  $U_N$  is defined via the following relations:

$$u^* = u^{-1}$$
$$u^t = \bar{u}^{-1}$$

But these relations tell us precisely that the following two operators must be in the associated Tannakian category C:

$$\begin{array}{rcl} T_{\pi} & , & \pi = \begin{tabular}{c} & & \\ & & \\ T_{\pi} & , & \pi = \begin{tabular}{c} & & \\ & \bullet & \\ & \bullet & \\ \end{array}$$

Thus the associated Tannakian category is  $C = span(T_{\pi} | \pi \in D)$ , with:

$$D = < \stackrel{\cap}{_{\circ\circ}}, \stackrel{\cap}{_{\circ\circ}} > \\ = \mathcal{P}_2$$

Thus, we are led to the conclusion in the statement.

Regarding now the orthogonal group  $O_N$ , we have here a Brauer theorem as well, the statement being quite similar, as follows:

THEOREM 14.23. For the orthogonal group  $O_N$  we have

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in P_2(k, l)\right)$$

where  $P_2$  denotes as usual the category of all pairings.

**PROOF.** Consider the spaces on the right in the statement, namely:

$$C_{kl} = span\left(T_{\pi} \middle| \pi \in P_2(k,l)\right)$$

According to Theorem 14.19 these spaces form a tensor category. Thus, according to Tannakian duality, these spaces must come from a certain closed subgroup, as follows:

$$G \subset U_N$$

To be more precise, if we denote by v the fundamental representation of G, then:

$$C_{kl} = Hom(v^{\otimes k}, v^{\otimes l})$$

We must prove that we have  $G = O_N$ . For this purpose, let us recall that the orthogonal group  $O_N \subset U_N$  is defined by imposing the following relations:

$$u_{ij} = \bar{u}_{ij}$$

But these relations tell us precisely that the following two operators must be in the associated Tannakian category C:

$$T_{\pi} \quad , \quad \pi = \clubsuit$$
$$T_{\pi} \quad , \quad \pi = \clubsuit$$

Thus the associated Tannakian category is  $C = span(T_{\pi} | \pi \in D)$ , with:

$$D = \langle \mathcal{P}_2, \overset{\circ}{\bullet}, \overset{\circ}{\bullet} \rangle = P_2$$

Thus, we are led to the conclusion in the statement.

We will see later, in chapter 16 below, applications of the above results, to integration problems over  $O_N, U_N$ , by using the Peter-Weyl methods from chapter 13.

Also, we will see in the next chapter a unification and generalization of the above results and their proofs, which are quite similar.

### 14E. EXERCISES

### 14e. Exercises

Generally speaking, the best complement to the material presented in this chapter is more reading on Tannakian duality, in its various versions, which are all useful.

With the technology presented above, however, we can work out a few interesting particular cases of the Tannakian duality, and this will be the purpose of the first few exercises that we have here. Let us start with something quite elementary:

EXERCISE 14.24. Work out the Tannakian duality for the closed subgroups

 $G \subset O_N$ 

first as a consequence of the general results that we have, regarding the closed subgroups

 $G \subset U_N$ 

and then independently, by pointing out the simplifications that appear in the real case.

Regarding the first question, this is normally something quite quick, obtained by adding the assumption  $u = \bar{u}$  to the Tannakian statement that we have, and then working out the details. Regarding the second question, the idea here is basically that the colored exponents  $k, l = \circ \bullet \bullet \circ \ldots$  will become in this way usual exponents,  $k, l \in \mathbb{N}$ , and this brings a number of simplifications in the proof, which are to be found.

Here is a related question, which is a bit more complicated:

EXERCISE 14.25. Work out the Tannakian duality for the closed subgroups

 $G \subset U_N$ 

whose fundamental representation is self-adjoint, up to equivalence,

 $u\sim \bar{u}$ 

first as a consequence of the results that we have, and then independently.

Here are there are several possible paths, either by proceeding a bit as for the previous exercise, but with the condition  $u = \bar{u}$  there replaced by the more general condition  $u \sim \bar{u}$ , or by using what was done in the previous exercise, and generalizing, from  $u = \bar{u}$  to  $u \sim \bar{u}$ . In any case, regardless of the method which is chosen, the problem is that understanding what the condition  $u \sim \bar{u}$  really means, categorially speaking.

Moving ahead now, in relation with diagrams and Brauer theorems, we have:

EXERCISE 14.26. Check the Brauer theorems for  $O_N, U_N$ , which are both of type

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

for small values of the global length parameter,  $k + l \in \{1, 2, 3\}$ .

The idea here is to prove these results that we already know directly, by double inclusion, with the inclusion in one sense being normally something quite elementary, and with the inclusion in the other sense being probably somehing quite tricky.

As a second question now regarding the Brauer theorems, we have:

EXERCISE 14.27. Write down Brauer theorems for the bistochastic groups  $B_N, C_N$ , by identifying first the partition which produces them, as subgroups of  $O_N, U_N$ .

This is actually something that will be discussed later on in this book, but without too much details, so the answer "done in the book" will not do.

As already mentioned after the proof of Tannakian duality, there are plenty of other versions of Tannakian duality, which are all worth learning. First, we have:

EXERCISE 14.28. Look up the original version of Tannakian duality, stating that G can be recovered from the knowledge of its full category of representations  $\mathcal{R}_G$ , viewed as subcategory of the category  $\mathcal{H}$  of the finite dimensional Hilbert spaces, with each  $\pi \in \mathcal{R}_G$  corresponding to its Hilbert space  $H_{\pi} \in \mathcal{H}$ , and write down a brief account of this.

As already mentioned in the above, the idea is that the group G appears as the group of endomorphisms of the embedding functor  $\mathcal{R}_G \subset \mathcal{H}$ . Time to understand this.

As another exercise now, of the same type, but more advanced, we have:

EXERCISE 14.29. Look up the Doplicher-Roberts and Deligne theorems, stating that the compact group G can be in fact recovered from the sole knowledge of the category  $\mathcal{R}_G$ , with no need for the embedding into  $\mathcal{H}$ , and write down a brief account of this.

This is obviously something more advanced, and the proof is quite tricky. Try however to undertand the main ideas behid the proof, which are very instructive.

Getting back now to the category  $\mathcal{R}_G^{\circ} \subset \mathcal{R}_G$  consisting of Peter-Weyl representations, that we used in the above, a natural idea is that of further shrinking this category to its diagonal algebra  $P_G = \Delta \mathcal{R}_G$ , consisting of the \*-algebras  $C(k, k) = End(u^{\otimes k})$ . This is the idea behind Jones' planar algebras, with  $P_G$  being such an algebra, and we have:

EXERCISE 14.30. Given a closed subgroup  $G \subset_u U_N$ , understand and then briefly explain, in a short piece of writing, why the \*-algebras

$$C(k,k) = End(u^{\otimes k})$$

form a planar algebra in the sense of Jones, and then comment as well on the various formulations of Tannakian duality, in the planar algebra setting.

This is actually quite difficult. And as a final, bonus exercise, try learning as well some Lie algebras, and their relation with the above, and write down what you learned.

## CHAPTER 15

# Diagrams, easiness

## 15a. Easy groups

We have seen in the previous chapter that the Tannakian duals of the groups  $O_N, U_N$ are very simple objects. To be more precise, the Brauer theorem for these two groups states that we have equalities as follows, with  $D = P_2, \mathcal{P}_2$  respectively:

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

Our goal here is that of axiomatizing and studying the closed subgroups  $G \subset U_N$  which are of this type, that we will call "easy". Our results will be as follows:

(1) At the level of the examples, we will see that besides  $O_N, U_N$ , we have the bistochastic groups  $B_N, C_N$ , the symmetric group  $S_N$ , the hyperoctahedral group  $H_N$ , and more generally the series of complex reflection groups  $H_N^s$ .

(2) Also at the level of the basic examples, some key groups such as the symplectic group  $Sp_N$  are not easy, but we will show here that these are covered by a suitable "super-easiness" version of the easiness, as defined above.

(3) At the level of the general theory, we will develop some algebraic theory in this chapter, for the most in relation with various product operations, the idea being that in the easy case, everything eventually reduces to computations with partitions.

(4) Also at the level of the general theory, we will develop as well some analytic theory, in the next chapter, based on the same idea, namely that in the easy case, everything eventually reduces to some elementary computations with partitions.

In order to get started, let us formulate the following key definition, extending to the case of arbitrary partitions what we already know about pairings:

DEFINITION 15.1. Given a partition  $\pi \in P(k, l)$  and an integer  $N \in \mathbb{N}$ , we define  $T_{\pi} : (\mathbb{C}^N)^{\otimes k} \to (\mathbb{C}^N)^{\otimes l}$ 

by the following formula, with  $e_1, \ldots, e_N$  being the standard basis of  $\mathbb{C}^N$ ,

$$T_{\pi}(e_{i_1} \otimes \ldots \otimes e_{i_k}) = \sum_{j_1 \dots j_l} \delta_{\pi} \begin{pmatrix} i_1 & \cdots & i_k \\ j_1 & \cdots & j_l \end{pmatrix} e_{j_1} \otimes \ldots \otimes e_{j_l}$$

and with the coefficients on the right being Kronecker type symbols.

### 15. DIAGRAMS, EASINESS

To be more precise here, in order to compute the Kronecker type symbols  $\delta_{\pi}(_{j}^{i})$ , we put the multi-indices i, j on the legs of  $\pi$ , in the obvious way. In case all the blocks of  $\pi$  contain equal indices of i, j, we set  $\delta_{\pi}(_{j}^{i}) = 1$ . Otherwise, we set  $\delta_{\pi}(_{j}^{i}) = 0$ .

With the above notion in hand, we can now formulate the following key definition, from [11], motivated by the Brauer theorems for  $O_N, U_N$ , as indicated before:

DEFINITION 15.2. A closed subgroup  $G \subset U_N$  is called easy when

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

for any two colored integers  $k, l = \circ \bullet \circ \bullet \ldots$ , for certain sets of partitions

 $D(k,l) \subset P(k,l)$ 

where  $\pi \to T_{\pi}$  is the standard implementation of the partitions, as linear maps.

In other words, we call a group G easy when its Tannakian category appears in the simplest possible way: from linear maps associated to partitions. The terminology is quite natural, because Tannakian duality is basically our only serious tool.

As basic examples, the orthogonal and unitary groups  $O_N, U_N$  are both easy, coming respectively from the following collections of sets of partitions:

$$P_{2} = \bigsqcup_{k,l} P_{2}(k,l)$$
$$\mathcal{P}_{2} = \bigsqcup_{k,l} \mathcal{P}_{2}(k,l)$$

In the general case now, as a theoretical remark, in the context of Definition 15.2 above, consider the following collection of sets of partitions:

$$D = \bigsqcup_{k,l} D(k,l)$$

This collection of sets D obviously determines G, but the converse is not true. Indeed, at N = 1 for instance, both the choices  $D = P_2, \mathcal{P}_2$  produce the same easy group, namely  $G = \{1\}$ . We will be back to this issue on several occasions, with results about it.

In order to advance, our first goal will be that of establishing a duality between easy groups and certain special classes of collections of sets as above, namely:

$$D = \bigsqcup_{k,l} D(k,l)$$

Let us begin with a general definition, from [11], as follows:

### 15A. EASY GROUPS

DEFINITION 15.3. Let P(k, l) be the set of partitions between an upper colored integer k, and a lower colored integer l. A collection of subsets

$$D = \bigsqcup_{k,l} D(k,l)$$

with  $D(k,l) \subset P(k,l)$  is called a category of partitions when it has the following properties:

- (1) Stability under the horizontal concatenation,  $(\pi, \sigma) \rightarrow [\pi\sigma]$ .
- (2) Stability under vertical concatenation  $(\pi, \sigma) \to [\sigma]$ , with matching middle symbols.
- (3) Stability under the upside-down turning \*, with switching of colors,  $\circ \leftrightarrow \bullet$ .
- (4) Each set P(k,k) contains the identity partition  $\| \dots \|$ .
- (5) The sets  $P(\emptyset, \circ \bullet)$  and  $P(\emptyset, \bullet \circ)$  both contain the semicircle  $\cap$ .
- (6) The sets P(k,k) with |k| = 2 contain the crossing partition  $\chi$ .

As before, this is something that we already met in chapter 14, but for the pairings only. Observe the similarity with the axioms for Tannakian categories, from chapter 13. We will see in a moment that this similarity can be turned into something very precise, the idea being that such a category produces a family of easy quantum groups  $(G_N)_{N \in \mathbb{N}}$ , one for each  $N \in \mathbb{N}$ , via the formula in Definition 15.1, and Tannakian duality.

As basic examples of such categories of partitions, that we have already met in chapter 14 above, in connection with the representation theory of the groups  $O_N, U_N$ , we have the categories  $P_2, \mathcal{P}_2$  of pairings, and of matching pairings.

Further basic examples include the categories  $P, P_{even}$  of all partitions, and of all partitions whose blocks have even size. We will see later that these latter categories are related to the symmetric and hyperoctahedral groups  $S_N, H_N$ .

The relation with the Tannakian categories comes from the following result, which once again extends something that we already know about the pairings:

THEOREM 15.4. The assignment  $\pi \to T_{\pi}$  is categorical, in the sense that we have

$$T_{\pi} \otimes T_{\sigma} = T_{[\pi\sigma]}$$
$$T_{\pi}T_{\sigma} = N^{c(\pi,\sigma)}T_{[\pi]}$$
$$T_{\pi}^{*} = T_{\pi^{*}}$$

where  $c(\pi, \sigma)$  are certain integers, coming from the erased components in the middle.

**PROOF.** This is something that we already know for the pairings, from chapter 14 above, and the proof in general is similar, the computations being as follows:

## 15. DIAGRAMS, EASINESS

(1) The concatenation axiom follows from the following computation:

$$(T_{\pi} \otimes T_{\sigma})(e_{i_{1}} \otimes \ldots \otimes e_{i_{p}} \otimes e_{k_{1}} \otimes \ldots \otimes e_{k_{r}})$$

$$= \sum_{j_{1} \ldots j_{q}} \sum_{l_{1} \ldots l_{s}} \delta_{\pi} \begin{pmatrix} i_{1} & \ldots & i_{p} \\ j_{1} & \ldots & j_{q} \end{pmatrix} \delta_{\sigma} \begin{pmatrix} k_{1} & \ldots & k_{r} \\ l_{1} & \ldots & l_{s} \end{pmatrix} e_{j_{1}} \otimes \ldots \otimes e_{j_{q}} \otimes e_{l_{1}} \otimes \ldots \otimes e_{l_{s}}$$

$$= \sum_{j_{1} \ldots j_{q}} \sum_{l_{1} \ldots l_{s}} \delta_{[\pi\sigma]} \begin{pmatrix} i_{1} & \ldots & i_{p} & k_{1} & \ldots & k_{r} \\ j_{1} & \ldots & j_{q} & l_{1} & \ldots & l_{s} \end{pmatrix} e_{j_{1}} \otimes \ldots \otimes e_{j_{q}} \otimes e_{l_{1}} \otimes \ldots \otimes e_{l_{s}}$$

$$= T_{[\pi\sigma]}(e_{i_{1}} \otimes \ldots \otimes e_{i_{p}} \otimes e_{k_{1}} \otimes \ldots \otimes e_{k_{r}})$$

(2) The composition axiom follows from the following computation:

$$T_{\pi}T_{\sigma}(e_{i_{1}}\otimes\ldots\otimes e_{i_{p}})$$

$$=\sum_{j_{1}\ldots j_{q}}\delta_{\sigma}\begin{pmatrix}i_{1}&\ldots&i_{p}\\j_{1}&\ldots&j_{q}\end{pmatrix}\sum_{k_{1}\ldots k_{r}}\delta_{\pi}\begin{pmatrix}j_{1}&\ldots&j_{q}\\k_{1}&\ldots&k_{r}\end{pmatrix}e_{k_{1}}\otimes\ldots\otimes e_{k_{r}}$$

$$=\sum_{k_{1}\ldots k_{r}}N^{c(\pi,\sigma)}\delta_{[\frac{\sigma}{\pi}]}\begin{pmatrix}i_{1}&\ldots&i_{p}\\k_{1}&\ldots&k_{r}\end{pmatrix}e_{k_{1}}\otimes\ldots\otimes e_{k_{r}}$$

$$=N^{c(\pi,\sigma)}T_{[\frac{\sigma}{\pi}]}(e_{i_{1}}\otimes\ldots\otimes e_{i_{p}})$$

(3) Finally, the involution axiom follows from the following computation:

$$T_{\pi}^{*}(e_{j_{1}} \otimes \ldots \otimes e_{j_{q}})$$

$$= \sum_{i_{1} \ldots i_{p}} < T_{\pi}^{*}(e_{j_{1}} \otimes \ldots \otimes e_{j_{q}}), e_{i_{1}} \otimes \ldots \otimes e_{i_{p}} > e_{i_{1}} \otimes \ldots \otimes e_{i_{p}}$$

$$= \sum_{i_{1} \ldots i_{p}} \delta_{\pi} \begin{pmatrix} i_{1} & \cdots & i_{p} \\ j_{1} & \cdots & j_{q} \end{pmatrix} e_{i_{1}} \otimes \ldots \otimes e_{i_{p}}$$

$$= T_{\pi^{*}}(e_{j_{1}} \otimes \ldots \otimes e_{j_{q}})$$

Summarizing, our correspondence is indeed categorical.

It is time now to put everyting together. All the above was pure combinatorics, and in relation with the compact groups, we have the following result:

THEOREM 15.5. Each category of partitions D = (D(k, l)) produces a family of compact groups  $G = (G_N)$ , one for each  $N \in \mathbb{N}$ , via the formula

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

and the Tannakian duality correspondence.

### 15A. EASY GROUPS

**PROOF.** Given an integer  $N \in \mathbb{N}$ , consider the correspondence  $\pi \to T_{\pi}$  constructed in Definition 15.1, and then the collection of linear spaces in the statement, namely:

$$C_{kl} = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

According to Theorem 15.4, and to our axioms for the categories of partitions, from Definition 15.3, this collection of spaces  $C = (C_{kl})$  satisfies the axioms for the Tannakian categories, from chapter 14 above. Thus the Tannakian duality result there applies, and provides us with a closed subgroup  $G_N \subset U_N$  such that:

$$C_{kl} = Hom(u^{\otimes k}, u^{\otimes l})$$

Thus, we are led to the conclusion in the statement.

In relation with the easiness property, we can now formulate a key result, which can serve as an alternative definition for the easy groups, as follows:

THEOREM 15.6. A closed subgroup  $G \subset U_N$  is easy precisely when

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

for any colored integers k, l, for a certain category of partitions  $D \subset P$ .

**PROOF.** This basically follows from Theorem 15.5, as follows:

(1) In one sense, we know from Theorem 15.5 that any category of partitions  $D \subset P$  produces a family of closed groups  $G \subset U_N$ , one for each  $N \in \mathbb{N}$ , according to Tannakian duality and to the Hom space formula there, namely:

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

But these groups  $G \subset U_N$  are indeed easy, in the sense of Definition 15.2.

(2) In the other sense now, assume that  $G \subset U_N$  is easy, in the sense of Definition 15.2, coming via the above Hom space formula, from a collection of sets as follows:

$$D = \bigsqcup_{k,l} D(k,l)$$

Consider now the category of partitions generated by this family:

$$\widetilde{D} = \langle D \rangle$$

It follows then, via another application of Tannakian duality, that  $G \subset U_N$  can be viewed as well as coming from  $\widetilde{D}$ , and this gives the result.

As already mentioned above, Theorem 15.6 can be regarded as an alternative definition for easiness, with the assumption that  $D \subset P$  must be a category of partitions being added. In what follows we will rather use this new definition, which is more precise.

### 15. DIAGRAMS, EASINESS

Generally speaking, the same comments as before apply. First, G is easy when its Tannakian category appears in the simplest possible way: from a category of partitions. The terminology is quite natural, because Tannakian duality is our only serious tool.

Also, the category of partitions D is not unique, for instance because at N = 1 all the categories of partitions produce the same easy group, namely  $G = \{1\}$ . We will be back to this issue on several occasions, with various results about it.

We will see in what follows that many interesting examples of compact quantum groups are easy. Moreover, most of the known series of "basic" compact quantum groups,  $G = (G_N)$  with  $N \in \mathbb{N}$ , can be in principle made fit into some suitable extensions of the easy quantum group formalism. We will discuss this too, in what follows.

The notion of easiness goes back to the results of Brauer regarding the orthogonal group  $O_N$ , and the unitary group  $U_N$ , which reformulate as follows:

THEOREM 15.7. We have the following results:

- (1) The unitary group  $U_N$  is easy, coming from the category  $\mathcal{P}_2$ .
- (2) The orthogonal group  $O_N$  is easy as well, coming from the category  $P_2$ .

**PROOF.** This is something that we already know from chapter 14 above, based on Tannakian duality, the idea being as follows:

(1) The group  $U_N$  being defined via the relations  $u^* = u^{-1}$ ,  $u^t = \bar{u}^{-1}$ , the associated Tannakian category is  $C = span(T_{\pi} | \pi \in D)$ , with:

$$D = < \cap_{\circ \bullet} \cap_{\circ \circ} \cap_{\circ \circ} > = \mathcal{P}_2$$

(2) The group  $O_N \subset U_N$  being defined by imposing the relations  $u_{ij} = \bar{u}_{ij}$ , the associated Tannakian category is  $C = span(T_{\pi} | \pi \in D)$ , with:

$$D = \langle \mathcal{P}_2, \overset{\circ}{\downarrow}, \overset{\circ}{\downarrow} \rangle = P_2$$

Thus, we are led to the conclusion in the statement.

As already mentioned in the beginning of this chapter, there are many other examples of easy groups, and we will gradually explore this, in what follows.

To start with, we have the following result, dealing with the orthogonal and unitary bistochastic groups  $B_N, C_N$ , that we discussed in chapter 10 above:

THEOREM 15.8. We have the following results:

- (1) The unitary bistochastic group  $C_N$  is easy, coming from the category  $\mathcal{P}_{12}$  of matching singletons and pairings.
- (2) The orthogonal bistochastic group  $B_N$  is easy, coming from the category  $P_{12}$  of singletons and pairings.

**PROOF.** The proof here is similar to the proof of Theorem 15.7. To be more precise, we can use the results there, and the proof goes as follows:

(1) The group  $C_N \subset U_N$  is defined by imposing the following relations, with  $\xi$  being the all-one vector, which correspond to the bistochasticity condition:

$$\begin{split} u\xi &= \xi \\ \bar{u}\xi &= \xi \end{split}$$

But these relations tell us precisely that the following two operators must be in the associated Tannakian category C:

$$T_{\pi} \quad , \quad \pi = \downarrow$$
$$T_{\pi} \quad , \quad \pi = \downarrow$$

Thus the associated Tannakian category is  $C = span(T_{\pi} | \pi \in D)$ , with:

$$D = \langle \mathcal{P}_2, \phi, \phi \rangle \geq \mathcal{P}_{12}$$

Thus, we are led to the conclusion in the statement.

(2) In order to deal now with the real bistochastic group  $B_N$ , we can either use a similar argument, or simply use the following intersection formula:

$$B_N = C_N \cap O_N$$

Indeed, at the categorical level, this intersection formula tells us that the associated Tannakian category is given by  $C = span(T_{\pi} | \pi \in D)$ , with:

$$D = \langle \mathcal{P}_{12}, \mathcal{P}_2 \rangle = \mathcal{P}_{12}$$

Thus, we are led to the conclusion in the statement.

As a comment here, we have used in the above the fact, which is something quite trivial, that the category of partitions associated to an intersection of easy quantum groups is the intersection of the corresponding categories of partitions. We will be back to this, and to some other product operations as well, with similar results, later on.

We can put now the results that we have together, as follows:

THEOREM 15.9. The basic unitary and bistochastic groups,



are all easy, coming from the various categories of singletons and pairings.

### 15. DIAGRAMS, EASINESS

**PROOF.** We know from the above results that the groups in the statement are indeed easy, the corresponding diagram of categories of partitions being as follows:



Thus, we are led to the conclusion in the statement.

Summarizing, what we have so far is a general notion of "easiness", coming from the Brauer theorems for  $O_N, U_N$ , and their straightforward extensions to  $B_N, C_N$ .

# 15b. Reflection groups

In view of the above, the notion of easiness is a quite interesting one, deserving a full, systematic investigation. As a first natural question that we would like to solve, we would like to compute the easy group associated to the category of all partitions P itself.

And here, no surprise, we are led to the most basic, but non-trivial, classical group that we know, namely the symmetric group  $S_N$ . To be more precise, we have the following Brauer type theorem for  $S_N$ , which answers our question formulated above:

THEOREM 15.10. The symmetric group  $S_N$ , regarded as group of unitary matrices,

 $S_N \subset O_N \subset U_N$ 

via the permutation matrices, is easy, coming from the category of all partitions P.

**PROOF.** Consider indeed the group  $S_N$ , regarded as a group of unitary matrices, with each permutation  $\sigma \in S_N$  corresponding to the associated permutation matrix:

$$\sigma(e_i) = e_{\sigma(i)}$$

Consider as well the easy group  $G \subset O_N$  coming from the category of all partitions P. Since P is generated by the one-block partition  $\mu \in P(2, 1)$ , we have:

$$C(G) = C(O_N) \Big/ \Big\langle T_{\mu} \in Hom(u^{\otimes 2}, u) \Big\rangle$$

The linear map associated to  $\mu$  is given by the following formula:

 $T_{\mu}(e_i \otimes e_j) = \delta_{ij} e_i$ 

Thus, we have the following formulae, using indices:

$$u = (u_{ij})_{ij}$$
$$u^{\otimes 2} = (u_{ij}u_{kl})_{ik,jl}$$
$$T_{\mu} = (\delta_{ijk})_{i,jk}$$

352

We therefore obtain the following formula:

$$\begin{aligned} (T_{\mu}u^{\otimes 2})_{i,jk} &= \sum_{lm} (T_{\mu})_{i,lm} (u^{\otimes 2})_{lm,jk} \\ &= u_{ij} u_{ik} \end{aligned}$$

We have as well the following formula:

$$(uT_{\mu})_{i,jk} = \sum_{l} u_{il}(T_{\mu})_{l,jk}$$
$$= \delta_{jk}u_{ij}$$

Thus, the relation defining  $G \subset O_N$  reformulates as follows:

$$T_{\mu} \in Hom(u^{\otimes 2}, u) \iff u_{ij}u_{ik} = \delta_{jk}u_{ij}, \forall i, j, k$$

In other words, the elements  $u_{ij}$  must be projections, and these projections must be pairwise orthogonal on the rows of  $u = (u_{ij})$ . We conclude that  $G \subset O_N$  is the subgroup of matrices  $g \in O_N$  having the following property:

$$g_{ij} \in \{0, 1\}$$

Thus we have  $G = S_N$ , as desired, and this gives the result.

As a continuation of all this, and following the hierarchy of finite groups explained in chapters 9-12 above, with the original motivations there being acutually probabilistic, but with these motivations leading to the same hierarchy that we need here, let us discuss now the hyperoctahedral group  $H_N$ . The result here is quite similar, as follows:

THEOREM 15.11. The hyperoctahedral group  $H_N$ , regarded as a group of matrices,

$$S_N \subset H_N \subset O_N$$

is easy, coming from the category of partitions with even blocks  $P_{even}$ .

**PROOF.** This follows as usual from Tannakian duality. To be more precise, consider the following one-block partition:

$$\pi \in P(4)$$

It is routine to check that this partition implements the relations producing the subgroup  $H_N \subset S_N$ , and so this group  $H_N$  is easy, coming from:

$$D = <\pi>$$

On the other hand, the category on the right can be computed by drawing pictures, and we obtain the following formula:

$$<\pi>=P_{even}$$

Thus, we are led to the conclusion in the statement.

In fact, we have the following general result, regarding now the series of complex reflection groups  $H_N^s$ , which covers both the groups  $S_N, H_N$ :

THEOREM 15.12. The complex reflection group

$$H_N^s = \mathbb{Z}_s \wr S_N$$

is easy, the corresponding category  $P^s$  consisting of the partitions satisfying

$$\#\circ = \# \bullet (s)$$

in each block. In particular, we have the following results:

- (1)  $S_N$  is easy, coming from the category P.
- (2)  $H_N = \mathbb{Z}_2 \wr S_N$  is easy, coming from the category  $P_{even}$ .
- (3)  $K_N = \mathbb{T} \wr S_N$  is easy, coming from the category  $\mathcal{P}_{even}$ .

PROOF. This is something that we already know at s = 1, 2, from Theorems 15.10 and 15.11 above. In general, the proof is similar, based on Tannakian duality.

To be more precise, in what regards the main assertion, the idea here is that the oneblock partition  $\pi \in P(s)$ , which generates the category of partitions  $P^s$  in the statement, implements the relations producing the subgroup  $H_N^s \subset S_N$ .

As for the last assertions, these follow from the following observations:

(1) At s = 1 we know that we have  $H_N^1 = S_N$ . Regarding now the corresponding category, here the condition  $\# \circ = \# \bullet (1)$  is automatic, and so  $P^1 = P$ .

(2) At s = 2 we know that we have  $H_N^2 = H_N$ . Regarding now the corresponding category, here the condition  $\# \circ = \# \bullet (2)$  reformulates as follows:

$$\# \circ + \# \bullet = 0(2)$$

Thus each block must have even size, and we obtain, as claimed,  $P^2 = P_{even}$ .

(3) At  $s = \infty$  we know that we have  $H_N^{\infty} = K_N$ . Regarding now the corresponding category, here the condition  $\# \circ = \# \bullet (\infty)$  reads:

$$\#\circ = \#\bullet$$

But this is the condition defining  $\mathcal{P}_{even}$ , and so  $P^{\infty} = \mathcal{P}_{even}$ , as claimed.

Summarizing, we have many examples. In fact, our list of easy groups has currently become quite big, and here is a selection of the main results that we have so far:

THEOREM 15.13. We have a diagram of compact groups as follows,



where  $H_N = \mathbb{Z}_2 \wr S_N$  and  $K_N = \mathbb{T} \wr S_N$ , and all these groups are easy.

354

**PROOF.** This follows from the above results. To be more precise, we know that the above groups are all easy, the corresponding categories of partitions being as follows:



Thus, we are led to the conclusion in the statement.

Summarizing, most of the finite and compact groups that we investigated so far in this book are covered by the easy group formalism. One exception is the symplectic group  $Sp_N$ , but we will see later that this group is covered as well, by a certain extension of the easy group formalism. We will see later as well analytic applications of all this.

## 15c. Basic operations

We develop now some general abstract theory for the easy groups. Let us first discuss some basic composition operations. We will be mainly interested in:

DEFINITION 15.14. The closed subgroups of  $U_N$  are subject to intersection and generation operations, constructed as follows:

- (1) Intersection:  $H \cap K$  is the usual intersection of H, K.
- (2) Generation:  $\langle H, K \rangle$  is the closed subgroup generated by H, K.

Alternatively, we can define these operations at the function algebra level, by performing operations on the associated ideals, as follows:

**PROPOSITION** 15.15. Assuming that we have presentation results as follows

$$C(H) = C(U_N)/I$$
$$C(K) = C(U_N)/J$$

the groups  $H \cap K$  and  $\langle H, K \rangle$  are given by the following formulae:

$$C(H \cap K) = C(U_N) / \langle I, J \rangle$$
$$C(\langle H, K \rangle) = C(U_N) / (I \cap J)$$

**PROOF.** This is indeed clear from the definition of the operations  $\cap$  and  $\langle , \rangle$ , and from the Stone-Weierstrass theorem.

In what follows we will need Tannakian formulations of the above two operations.

The result here, coming from the duality established in chaper 14, is as follows:

### 15. DIAGRAMS, EASINESS

THEOREM 15.16. The intersection and generation operations  $\cap$  and  $\langle , \rangle$  can be constructed via the Tannakian correspondence  $G \to C_G$ , as follows:

- (1) Intersection: defined via  $C_{G \cap H} = \langle C_G, C_H \rangle$ .
- (2) Generation: defined via  $C_{\langle G,H \rangle} = C_G \cap C_H$ .

**PROOF.** This follows indeed from the definition of our operations, in its functional analytic formulation, as explained above, and from the Tannakian duality result for the compact Lie groups, in its soft form, from chapter 14 above.

To be more precise, it follows from the Tannakian duality result worked out in chapter 14 that given a closed subgroup  $G \subset U_N$ , with fundamental representation v, the corresponding algebra of functions C(G) has the following presentation:

$$C(G) = C(U_N) \Big/ \left\langle T \in Hom(u^{\otimes k}, u^{\otimes l}) \Big| \forall k, \forall l, \forall T \in Hom(v^{\otimes k}, v^{\otimes l}) \right\rangle$$

In other words, given a closed subgroup  $G \subset U_N$ , we have a presentation of the following type, with  $I_G$  being the ideal coming from the Tannakian category of G:

$$C(G) = C(U_N)/I_G$$

But this leads to the conclusion in the statement.

In relation now with our easiness questions, we first have the following result:

**PROPOSITION 15.17.** Assuming that H, K are easy, then so is  $H \cap K$ , and we have

$$D_{H \cap K} = < D_H, D_K >$$

at the level of the corresponding categories of partitions.

**PROOF.** We have indeed the following computation:

$$C_{H \cap K} = \langle C_H, C_K \rangle$$
  
=  $\langle span(D_H), span(D_K) \rangle$   
=  $span(\langle D_H, D_K \rangle)$ 

Thus, by Tannakian duality we obtain the result.

Regarding now the generation operation, the situation is more complicated, and we have here the following statement:

**PROPOSITION 15.18.** Assuming that H, K are easy, we have an inclusion

$$\langle H, K \rangle \subset \{H, K\}$$

coming from an inclusion of Tannakian categories as follows,

$$C_H \cap C_K \supset span(D_H \cap D_K)$$

where  $\{H, K\}$  is the easy group having as category of partitions  $D_H \cap D_K$ .

356

**PROOF.** This follows from the definition and properties of the generation operation, and from the following computation:

$$C_{\langle H,K\rangle} = C_H \cap C_K$$
  
=  $span(D_H) \cap span(D_K)$   
 $\supset span(D_H \cap D_K)$ 

Indeed, by Tannakian duality we obtain from this all the assertions.

It is not clear if the inclusions in Proposition 15.18 are isomorphisms or not, and this even under a supplementary N >> 0 assumption. Technically speaking, the problem comes from the fact that the operation  $\pi \to T_{\pi}$  does not produce linearly independent maps, and so all that we are doing is sensitive to the value of  $N \in \mathbb{N}$ . The subject here is quite technical, to be further developed in chapter 16 below, with probabilistic motivations in mind, without however solving the present algebraic questions.

Summarizing, we have some problems here, and we must proceed as follows:

THEOREM 15.19. The intersection and easy generation operations  $\cap$  and  $\{,\}$  can be constructed via the Tannakian correspondence  $G \to D_G$ , as follows:

- (1) Intersection: defined via  $D_{G \cap H} = \langle D_G, D_H \rangle$ .
- (2) Easy generation: defined via  $D_{\{G,H\}} = D_G \cap D_H$ .

**PROOF.** Here the situation is as follows:

(1) This is a true and honest result, coming from Proposition 15.17.

(2) This is more of an empty statement, coming from Proposition 15.18.

As already mentioned, there is some interesting mathematics still to be worked out, in relation with all this. We will be back to this later.

With the above notions in hand, we can formulate a nice result, which improves our main result so far, namely Theorem 15.13 above, as follows:

THEOREM 15.20. The basic unitary and reflection groups, namely



are all easy, and they form an intersection and easy generation diagram, in the sense that the above square diagram satisfies  $U_N = \{K_N, O_N\}$ , and  $H_N = K_N \cap O_N$ .

### 15. DIAGRAMS, EASINESS

**PROOF.** We know from Theorem 15.13 that the groups in the statement are easy, the corresponding categories of partitions being as follows:



Now observe that this latter diagram is an intersection and generation diagram. By using Theorem 15.19, this reformulates into the fact that the diagram of quantum groups is an intersection and easy generation diagram, as claimed.  $\Box$ 

It is possible to further improve the above result, by proving that the diagram there is actually a plain generation diagram. However, this is something more technical.

Let us also mention that it is possible to develop some more general theory, at this level. Given a closed subgroup  $G \subset U_N$ , we can talk about its "easy envelope", which is the smallest easy quantum group  $\tilde{G}$  containing G, which is easy.

This easy envelope appears by definition as an intermediate closed subgroup, as follows:

$$G \subset \widetilde{G} \subset U_N$$

With this notion in hand, Proposition 15.18 can be refined into a result stating that given two easy groups H, K, we have inclusions as follows:

$$< H, K > \subset < H, K > \subset \{H, K\}$$

Howver, it is not clear when these inclusions, individually taken, are isomorphisms.

To be more precise now, in order to get started, we have:

DEFINITION 15.21. A closed subgroup  $G \subset U_N$  is called homogeneous when

$$S_N \subset G \subset U_N$$

with  $S_N \subset U_N$  being the standard embedding, via permutation matrices.

We will be interested in such groups. At the Tannakian level, we have:

THEOREM 15.22. The homogeneous groups  $S_N \subset G \subset U_N$  are in one-to-one correspondence with the intermediate tensor categories

$$span\left(T_{\pi} \middle| \pi \in \mathcal{P}_{2}\right) \subset C \subset span\left(T_{\pi} \middle| \pi \in P\right)$$

where P is the category of all partitions,  $\mathcal{P}_2$  is the category of the matching pairings, and  $\pi \to T_{\pi}$  is the standard implementation of partitions, as linear maps.

PROOF. This follows from Tannakian duality, and from the Brauer type results for  $S_N, U_N$ . To be more precise, we know from Tannakian duality that each closed subgroup  $G \subset U_N$  can be reconstructed from its Tannakian category C = (C(k, l)), as follows:

$$C(G) = C(U_N) \Big/ \left\langle T \in Hom(u^{\otimes k}, u^{\otimes l}) \Big| \forall k, l, \forall T \in C(k, l) \right\rangle$$

Thus we have a one-to-one correspondence  $G \leftrightarrow C$ , given by Tannakian duality, and since the endpoints  $G = S_N, U_N$  are both easy, corresponding to the categories  $C = span(T_{\pi}|\pi \in D)$  with  $D = P, \mathcal{P}_2$ , this gives the result.

Our purpose now will be that of using the Tannakian result in Theorem 15.22, in order to introduce and study a combinatorial notion of "easiness level", for the arbitrary intermediate groups  $S_N \subset G \subset U_N$ . Let us begin with the following simple fact:

PROPOSITION 15.23. Given a homogeneous group  $S_N \subset G \subset U_N$ , with associated Tannakian category C = (C(k, l)), the sets

$$D^{1}(k,l) = \left\{ \pi \in P(k,l) \middle| T_{\pi} \in C(k,l) \right\}$$

form a category of partitions, in the sense of Definition 15.3.

**PROOF.** We use the basic categorical properties of the correspondence  $\pi \to T_{\pi}$ , that we established above, namely:

$$T_{[\pi\sigma]} = T_{\pi} \otimes T_{\sigma}$$
$$T_{[\pi]} \sim T_{\pi}T_{\sigma}$$
$$T_{\pi^*} = T_{\pi}^*$$

Together with the fact that C is a tensor category, we deduce from these formulae that we have the following implication:

$$\pi, \sigma \in D^{1} \implies T_{\pi}, T_{\sigma} \in C$$
$$\implies T_{\pi} \otimes T_{\sigma} \in C$$
$$\implies T_{[\pi\sigma]} \in C$$
$$\implies [\pi\sigma] \in D^{1}$$

We have as well the following implication:

$$\pi, \sigma \in D^{1} \implies T_{\pi}, T_{\sigma} \in C$$
$$\implies T_{\pi}T_{\sigma} \in C$$
$$\implies T_{[\frac{\sigma}{\pi}]} \in C$$
$$\implies [\frac{\sigma}{\pi}] \in D^{1}$$
Finally, we have as well the following implication:

$$\begin{aligned} \mathbf{T} \in D^1 & \Longrightarrow & T_{\pi} \in C \\ & \Longrightarrow & T_{\pi}^* \in C \\ & \Longrightarrow & T_{\pi^*} \in C \\ & \Longrightarrow & \pi^* \in D^1 \end{aligned}$$

Thus  $D^1$  is indeed a category of partitions, as claimed.

We can further refine the above observation, in the following way:

PROPOSITION 15.24. Given a compact group  $S_N \subset G \subset U_N$ , construct  $D^1 \subset P$  as above, and let  $S_N \subset G^1 \subset U_N$  be the easy group associated to  $D^1$ . Then:

- (1) We have  $G \subset G^1$ , as subgroups of  $U_N$ .
- (2)  $G^1$  is the smallest easy group containing G.
- (3) G is easy precisely when  $G \subset G^1$  is an isomorphism.

**PROOF.** All this is elementary, the proofs being as follows:

(1) We know that the Tannakian category of  $G^1$  is given by:

$$C_{kl}^1 = span\left(T_{\pi} \middle| \pi \in D^1(k,l)\right)$$

Thus we have  $C^1 \subset C$ , and so  $G \subset G^1$ , as subgroups of  $U_N$ .

(2) Assuming that we have  $G \subset G'$ , with G' easy, coming from a Tannakian category C' = span(D'), we must have  $C' \subset C$ , and so  $D' \subset D^1$ . Thus,  $G^1 \subset G'$ , as desired.

(3) This is a trivial consequence of (2).

Summarizing, we have now a notion of "easy envelope", as follows:

DEFINITION 15.25. The easy envelope of a homogeneous group  $S_N \subset G \subset U_N$  is the easy group  $S_N \subset G^1 \subset U_N$  associated to the category of partitions

$$D^{1}(k,l) = \left\{ \pi \in P(k,l) \middle| T_{\pi} \in C(k,l) \right\}$$

where C = (C(k, l)) is the Tannakian category of G.

At the level of the examples, most of the known homogeneous groups  $S_N \subset G \subset U_N$  are in fact easy. However, there are many non-easy examples as well, and we will compute the easy envelopes in several cases of interest, later on.

As a technical observation now, we can in fact generalize the above construction to any closed subgroup  $G \subset U_N$ , and we have the following result:

PROPOSITION 15.26. Given a closed subgroup  $G \subset U_N$ , construct  $D^1 \subset P$  as above, and let  $S_N \subset G^1 \subset U_N$  be the easy quantum group associated to  $D^1$ . We have then

$$G^1 = (\langle G, S_N \rangle)^1$$

where  $\langle G, S_N \rangle \subset U_N$  is the smallest closed subgroup containing  $G, S_N$ .

PROOF. It is well-known, and elementary to show, using Tannakian duality, that the smallest subgroup  $\langle G, S_N \rangle \subset U_N$  from the statement exists indeed, and can be obtained by intersecting the Tannakian categories of  $G, S_N$ :

$$C_{\langle G, S_N \rangle} = C_G \cap C_{S_N}$$

We conclude from this that for any  $\pi \in P(k, l)$  we have:

$$T_{\pi} \in C_{\langle G, S_N \rangle}(k, l) \iff T_{\pi} \in C_G(k, l)$$

It follows that the  $D^1$  categories for the quantum groups  $\langle G, S_N \rangle$  and G coincide, and so the easy envelopes  $(\langle G, S_N \rangle)^1$  and  $G^1$  coincide as well, as stated.

In order now to fine-tune all this, by using an arbitrary parameter  $p \in \mathbb{N}$ , which can be thought of as being an "easiness level", we can proceed as follows:

DEFINITION 15.27. Given a compact group  $S_N \subset G \subset U_N$ , and an integer  $p \in \mathbb{N}$ , we construct the family of linear spaces

$$E^{p}(k,l) = \left\{ \alpha_{1}T_{\pi_{1}} + \ldots + \alpha_{p}T_{\pi_{p}} \in C(k,l) \middle| \alpha_{i} \in \mathbb{C}, \pi_{i} \in P(k,l) \right\}$$

and we denote by  $C^p$  the smallest tensor category containing  $E^p = (E^p(k, l))$ , and by  $S_N \subset G^p \subset U_N$  the compact group corresponding to this category  $C^p$ .

As a first observation, at p = 1 we have  $C^1 = E^1 = span(D^1)$ , where  $D^1$  is the category of partitions constructed in Proposition 15.24. Thus the quantum group  $G^1$  constructed above coincides with the "easy envelope" of G, from Definition 15.25 above.

In the general case,  $p \in \mathbb{N}$ , the family  $E^p = (E^p(k, l))$  constructed above is not necessarily a tensor category, but we can of course consider the tensor category  $C^p$  generated by it, as indicated. Finally, in the above definition we have used of course the Tannakian duality results, in order to perform the operation  $C^p \to G^p$ .

In practice, the construction in Definition 15.27 is often something quite complicated, and it is convenient to use the following observation:

**PROPOSITION 15.28.** The category  $C^p$  constructed above is generated by the spaces

$$E^{p}(l) = \left\{ \alpha_{1}T_{\pi_{1}} + \ldots + \alpha_{p}T_{\pi_{p}} \in C(l) \middle| \alpha_{i} \in \mathbb{C}, \pi_{i} \in P(l) \right\}$$

where C(l) = C(0, l), P(l) = P(0, l), with l ranging over the colored integers.

## 15. DIAGRAMS, EASINESS

**PROOF.** We use the well-known fact that given a closed subgroup  $G \subset U_N$ , we have a Frobenius type isomorphism, as follows:

$$Hom(u^{\otimes k}, u^{\otimes l}) \simeq Fix(u^{\otimes \bar{k}l})$$

If we apply this to the group  $G^p$  from Definition 15.27, we obtain an isomorphism as follows:

$$C(k,l) \simeq C(\bar{k}l)$$

On the other hand, we have as well an isomorphism  $P(k, l) \simeq P(\bar{k}l)$ , obtained by performing a counterclockwise rotation to the partitions  $\pi \in P(k, l)$ . According to the above definition of the spaces  $E^p(k, l)$ , this induces an isomorphism as follows:

$$E^p(k,l) \simeq E^p(\bar{k}l)$$

We deduce from this that for any partitions  $\pi_1, \ldots, \pi_p \in C(k, l)$ , having rotated versions  $\rho_1, \ldots, \rho_p \in C(\bar{k}l)$ , and for any scalars  $\alpha_1, \ldots, \alpha_p \in \mathbb{C}$ , we have:

$$\alpha_1 T_{\pi_1} + \ldots + \alpha_p T_{\pi_p} \in C(k,l) \iff \alpha_1 T_{\rho_1} + \ldots + \alpha_p T_{\rho_p} \in C(kl)$$

But this gives the conclusion in the statement, and we are done.

The main properties of the construction  $G \to G^p$  can be summarized as follows:

THEOREM 15.29. Given a compact group  $S_N \subset G \subset U_N$ , the compact groups  $G^p$  constructed above form a decreasing family, whose intersection is G:

$$G = \bigcap_{p \in \mathbb{N}} G^p$$

Moreover, G is easy when this decreasing limit is stationary,  $G = G^1$ .

**PROOF.** By definition of  $E^{p}(k, l)$ , and by using Proposition 15.28, these linear spaces form an increasing filtration of C(k, l). The same remains true when completing into tensor categories, and so we have an increasing filtration, as follows:

$$C = \bigcup_{p \in \mathbb{N}} C^p$$

At the compact group level now, we obtain the decreasing intersection in the statement. Finally, the last assertion is clear from Proposition 15.28.  $\hfill \Box$ 

As a main consequence of the above results, we can now formulate:

DEFINITION 15.30. We say that a compact group  $S_N \subset G \subset U_N^+$  is easy at order pwhen  $G = G^p$ . When p is chosen minimal, we also say that G has easiness level p.

Observe that the order 1 notion corresponds to the usual easiness. In general, all this is quite abstract, but there are several explicit examples, that can be worked out.

362

## 15D. CLASSIFICATION RESULTS

# 15d. Classification results

Let us go back now to easiness, and discuss some classification results. In order to cut from the complexity, we must impose an extra axiom, and we will use here:

THEOREM 15.31. For an easy group  $G = (G_N)$ , coming from a category of partitions  $D \subset P$ , the following conditions are equivalent:

- (1)  $G_{N-1} = G_N \cap U_{N-1}$ , via the embedding  $U_{N-1} \subset U_N$  given by  $u \to diag(u, 1)$ .
- (2)  $G_{N-1} = G_N \cap U_{N-1}$ , via the N possible diagonal embeddings  $U_{N-1} \subset U_N$ .
- (3) D is stable under the operation which consists in removing blocks.

If these conditions are satisfied, we say that  $G = (G_N)$  is uniform.

PROOF. We use here the general theory explained above.

(1)  $\iff$  (2) This is something standard, coming from the inclusion  $S_N \subset G_N$ , which makes everything  $S_N$ -invariant. The result follows as well from the proof of (1)  $\iff$  (3) below, which can be converted into a proof of (2)  $\iff$  (3), in the obvious way.

(1)  $\iff$  (3) Given a subgroup  $K \subset U_{N-1}$ , with fundamental representation u, consider the  $N \times N$  matrix v = diag(u, 1). Our claim is that for any  $\pi \in P(k)$  we have:

$$\xi_{\pi} \in Fix(v^{\otimes k}) \iff \xi_{\pi'} \in Fix(v^{\otimes k'}), \, \forall \pi' \in P(k'), \pi' \subset \pi$$

In order to prove this, we must study the condition on the left. We have:

$$\xi_{\pi} \in Fix(v^{\otimes k})$$

$$\iff (v^{\otimes k}\xi_{\pi})_{i_{1}\dots i_{k}} = (\xi_{\pi})_{i_{1}\dots i_{k}}, \forall i$$

$$\iff \sum_{j} (v^{\otimes k})_{i_{1}\dots i_{k}, j_{1}\dots j_{k}} (\xi_{\pi})_{j_{1}\dots j_{k}} = (\xi_{\pi})_{i_{1}\dots i_{k}}, \forall i$$

$$\iff \sum_{j} \delta_{\pi}(j_{1},\dots,j_{k})v_{i_{1}j_{1}}\dots v_{i_{k}j_{k}} = \delta_{\pi}(i_{1},\dots,i_{k}), \forall i$$

Now let us recall that our representation has the special form v = diag(u, 1). We conclude from this that for any index  $a \in \{1, \ldots, k\}$ , we must have:

$$i_a = N \implies j_a = N$$

With this observation in hand, if we denote by i', j' the multi-indices obtained from i, j obtained by erasing all the above  $i_a = j_a = N$  values, and by  $k' \leq k$  the common length of these new multi-indices, our condition becomes:

$$\sum_{j'} \delta_{\pi}(j_1, \dots, j_k)(v^{\otimes k'})_{i'j'} = \delta_{\pi}(i_1, \dots, i_k), \forall i$$

Here the index j is by definition obtained from j' by filling with N values. In order to finish now, we have two cases, depending on i, as follows:

# 15. DIAGRAMS, EASINESS

<u>Case 1</u>. Assume that the index set  $\{a|i_a = N\}$  corresponds to a certain subpartition  $\pi' \subset \pi$ . In this case, the N values will not matter, and our formula becomes:

$$\sum_{j'} \delta_{\pi}(j'_1, \dots, j'_{k'})(v^{\otimes k'})_{i'j'} = \delta_{\pi}(i'_1, \dots, i'_{k'})$$

<u>Case 2</u>. Assume now the opposite, namely that the set  $\{a|i_a = N\}$  does not correspond to a subpartition  $\pi' \subset \pi$ . In this case the indices mix, and our formula reads:

0 = 0

Thus, we are led to  $\xi_{\pi'} \in Fix(v^{\otimes k'})$ , for any subpartition  $\pi' \subset \pi$ , as claimed.

Now with this claim in hand, the result follows from Tannakian duality.

We can now formulate a first classification result, as follows:

THEOREM 15.32. The uniform orthogonal easy groups are as follows,



and this diagram is an intersection and easy generation diagram.

**PROOF.** We know that the quantum groups in the statement are indeed easy and uniform, the corresponding categories of partitions being as follows:



Since this latter diagram is an intersection and generation diagram, we conclude that we have an intersection and easy generation diagram of quantum groups, as stated.

Regarding now the classification, consider an easy group, as follows:

$$S_N \subset G_N \subset O_N$$

This group must come from a category of partitions, as follows:

$$P_2 \subset D \subset P$$

Now if we assume  $G = (G_N)$  to be uniform, this category D is uniquely determined by the subset  $L \subset \mathbb{N}$  consisting of the sizes of the blocks of the partitions in D.

Our claim is that the admissible sets are as follows:

- (1)  $L = \{2\}$ , producing  $O_N$ .
- (2)  $L = \{1, 2\}$ , producing  $B_N$ .
- (3)  $L = \{2, 4, 6, \ldots\}$ , producing  $H_N$ .
- (4)  $L = \{1, 2, 3, ...\},$  producing  $S_N$ .

Indeed, in one sense, this follows from our easiness results for  $O_N, B_N, H_N, S_N$ .

In the other sense now, assume that  $L \subset \mathbb{N}$  is such that the set  $P_L$  consisting of partitions whose sizes of the blocks belong to L is a category of partitions. We know from the axioms of the categories of partitions that the semicircle  $\cap$  must be in the category, so we have  $2 \in L$ . We claim that the following conditions must be satisfied as well:

$$k, l \in L, k > l \implies k - l \in L$$
  
 $k \in L, k \ge 2 \implies 2k - 2 \in L$ 

Indeed, we will prove that both conditions follow from the axioms of the categories of partitions. Let us denote by  $b_k \in P(0, k)$  the one-block partition:

$$b_k = \begin{cases} \sqcap & \ldots & \sqcap \\ 1 \ 2 & \ldots & k \end{cases}$$

For k > l, we can write  $b_{k-l}$  in the following way:

$$b_{k-l} = \begin{cases} \square & \dots & \dots & \dots & \square \\ 1 & 2 & \dots & l & l+1 & \dots & k \\ \square & \dots & \square & & | & \dots & | \\ & & & 1 & \dots & k-l \end{cases}$$

In other words, we have the following formula:

$$b_{k-l} = (b_l^* \otimes |^{\otimes k-l})b_k$$

Since all the terms of this composition are in  $P_L$ , we have  $b_{k-l} \in P_L$ , and this proves our first claim. As for the second claim, this can be proved in a similar way, by capping two adjacent k-blocks with a 2-block, in the middle.

With these conditions in hand, we can conclude in the following way:

<u>Case 1</u>. Assume  $1 \in L$ . By using the first condition with l = 1 we get:

$$k \in L \implies k-1 \in L$$

# 15. DIAGRAMS, EASINESS

This condition shows that we must have  $L = \{1, 2, ..., m\}$ , for a certain number  $m \in \{1, 2, ..., \infty\}$ . On the other hand, by using the second condition we get:

$$m \in L \implies 2m - 2 \in L$$
$$\implies 2m - 2 \leq m$$
$$\implies m \in \{1, 2, \infty\}$$

The case m = 1 being excluded by the condition  $2 \in L$ , we reach to one of the two sets producing the groups  $S_N, B_N$ .

<u>Case 2</u>. Assume  $1 \notin L$ . By using the first condition with l = 2 we get:

$$k \in L \implies k-2 \in L$$

This condition shows that we must have  $L = \{2, 4, ..., 2p\}$ , for a certain number  $p \in \{1, 2, ..., \infty\}$ . On the other hand, by using the second condition we get:

$$2p \in L \implies 4p - 2 \in L$$
$$\implies 4p - 2 \leq 2p$$
$$\implies p \in \{1, \infty\}$$

Thus L must be one of the two sets producing  $O_N, H_N$ , and we are done.

In the unitary case, the classification is quite similar. In particular we have:

THEOREM 15.33. The uniform, purely unitary easy groups are as follows,



and this diagram is an intersection and easy generation diagram.

**PROOF.** We know that the quantum groups in the statement are indeed easy and uniform, the corresponding categories of partitions being as follows:



Since this latter diagram is an intersection and generation diagram, we conclude that we have an intersection and easy generation diagram of quantum groups, as stated.

# 15E. EXERCISES

As for the uniqueness result, the proof here is similar to the proof from the real case, from Theorem 15.32 above, by examining the possible sizes of the blocks of the partitions in the category, and doing some direct combinatorics.

Alternatively, with a good definition for the notion "pure unitarity", expressing the fact that the group in question must appear as the complexification of a real easy group, the present result can be deduced from Theorem 15.32 above.  $\Box$ 

When lifting the uniformity assumption in all the above, things become more complicated, and the final classification results become more technical, due to the presence of various copies of  $\mathbb{Z}_2$ , that can be added, as for keeping the easiness property true.

To be more precise, in the real case, as explained in [11], we have exactly 6 solutions, which are as follows, with the convention  $G'_N = G_N \times \mathbb{Z}_2$ :



In the unitary case the situation is more complicated, and we refer here to [79].

Finally, the easy quantum group formalism can be extended into a "super-easy" group formalism, covering as well the symplectic group  $Sp_N$ . We refer here to [20].

# 15e. Exercises

In relation with the notion of easy envelope, which was introduced in the above, we have some explicit computations to be done, as follows:

EXERCISE 15.34. Compute the easy envelope of the general complex reflection groups

$$H_N^{sd} = \left\{ U \in H_N^s \middle| (\det U)^d = 1 \right\}$$

and of other closed subgroups of  $U_N$  that you know, that are not easy.

This is something which does not look very difficult, and you have the choice here, either by using combinatorics, or the universality property of the easy envelope.

Here is now one key exercise, in relation with the extensions of easiness:

EXERCISE 15.35. Work out the super-easiness property of the symplectic group

$$Sp_N \subset U_N$$

defined for  $N \in \mathbb{N}$  even, then try as well the groups  $SU_2$  and  $SO_3$ .

# 15. DIAGRAMS, EASINESS

This is actually a quite difficult exercise, and for  $Sp_N$  you can of course look it up, if you don't find the answer. As for  $SU_2$  and  $SO_3$ , you're into advanced mathematics here, and good luck with this. Also, as a bonus exercise, which brings you into hot research topics, you can learn as well about the exceptional Lie groups, and then try to understand what is the suitable extension of the easiness property, covering them.

Finally, we have several exercises in connection with classification problems, which were only briefly discussed in the above. First, in the real case, we have:

EXERCISE 15.36. Prove that when lifting the uniformity assumption, the groups



with the convention  $G'_N = G_N \times \mathbb{Z}_2$ , are the only easy real groups.

This is something quite standard, briefly discussed in the above, the idea being that of adapting the proof of the classification from the real uniform case.

In the unitary case now, we have a more modest exercise, as follows:

EXERCISE 15.37. Prove that the uniform, purely unitary easy groups are



with a suitable definition for the notion of pure unitarity.

As before, this is something quite standard, briefly discussed in the above, the idea being that of adapting the proof of the classification from the real uniform case.

Finally, as a bonus exercise, try classifying all the easy groups  $G \subset U_N$ , with no extra assumption at all. And of course, in case you're lost somewhere, don't hesitate to look it up, and then to write down a brief account of what you learned.

# CHAPTER 16

# Weingarten calculus

# 16a. Weingarten formula

We discuss in what follows applications of the theory developed in chapters 13-15, to the computation of the laws of characters, and truncated characters, as to solve the various questions left open in chapters 9-12, in the continuous group case.

Generally speaking, all these questions require a good knowledge of the integration over G, and more precisely, of the various polynomial integrals over G:

DEFINITION 16.1. Given a closed subgroup  $G \subset U_N$ , the quantities

$$I_k = \int_G g_{i_1 j_1}^{e_1} \dots g_{i_k j_k}^{e_k} \, dg$$

depending on a colored integer  $k = e_1 \dots e_k$ , are called polynomial integrals over G.

As a first observation, the knowledge of these integrals is the same as the full knowledge of the integration functional over G. Indeed, since the coordinate functions  $g \to g_{ij}$ separate the points of G, we can apply the Stone-Weierstrass theorem, and we obtain:

$$C(G) = \langle g_{ij} \rangle$$

Thus, by linearity, the computation of any functional  $f : C(G) \to \mathbb{C}$ , and in particular of the integration functional, reduces to the computation of this functional on the polynomials of the coordinate functions  $g \to g_{ij}$  and their conjugates  $g \to \bar{g}_{ij}$ .

By using the Peter-Weyl theory, everything reduces to algebra, due to:

THEOREM 16.2. The Haar integration over a closed subgroup  $G \subset U_N$  is given on the dense subalgebra of smooth functions by the Weingarten type formula

$$\int_G g_{i_1 j_1}^{e_1} \dots g_{i_k j_k}^{e_k} dg = \sum_{\pi, \sigma \in D_k} \delta_{\pi}(i) \delta_{\sigma}(j) W_k(\pi, \sigma)$$

valid for any colored integer  $k = e_1 \dots e_k$  and any multi-indices i, j, where  $D_k$  is a linear basis of  $Fix(u^{\otimes k})$ , the associated generalized Kronecker symbols are given by

$$\delta_{\pi}(i) = <\pi, e_{i_1} \otimes \ldots \otimes e_{i_k} >$$

and  $W_k = G_k^{-1}$  is the inverse of the Gram matrix,  $G_k(\pi, \sigma) = <\pi, \sigma >$ .

**PROOF.** We know from Peter-Weyl theory that the integrals in the statement form altogether the orthogonal projection  $P^k$  onto the following space:

$$Fix(u^{\otimes k}) = span(D_k)$$

Consider now the following linear map, with  $D_k = \{\xi_k\}$  being as in the statement:

$$E(x) = \sum_{\pi \in D_k} \langle x, \xi_\pi \rangle \xi_\pi$$

By a standard linear algebra computation, it follows that we have P = WE, where W is the inverse of the restriction of E to the following space:

$$K = span\left(T_{\pi} \middle| \pi \in D_k\right)$$

But this restriction is the linear map given by the matrix  $G_k$ , and so W is the linear map given by the inverse matrix  $W_k = G_k^{-1}$ , and this gives the result.

In the easy case, we have the following result:

THEOREM 16.3. For an easy group  $G \subset U_N$ , coming from a category of partitions

$$D = (D(k, l))$$

we have the Weingarten integration formula

$$\int_{G} u_{i_1 j_1}^{e_1} \dots u_{i_k j_k}^{e_k} = \sum_{\pi, \sigma \in D(k)} \delta_{\pi}(i) \delta_{\sigma}(j) W_{kN}(\pi, \sigma)$$

for any multi-indices i, j and any exponent  $k = e_1 \dots e_k$ , where  $D(k) = D(\emptyset, k)$ , the  $\delta$  numbers are the usual Kronecker type symbols, and  $W_{kN} = G_{kN}^{-1}$ , with

$$G_{kN}(\pi,\sigma) = N^{|\pi \vee \sigma|}$$

where |.| is the number of blocks.

**PROOF.** We use the abstract Weingarten formula, from Theorem 16.2 above. According to our easiness conventions, the Kronecker symbols are then given by:

$$\delta_{\xi_{\pi}}(i) = \langle \xi_{\pi}, e_{i_1} \otimes \ldots \otimes e_{i_k} \rangle$$

$$= \left\langle \sum_j \delta_{\pi}(j_1, \ldots, j_k) e_{j_1} \otimes \ldots \otimes e_{j_k}, e_{i_1} \otimes \ldots \otimes e_{i_k} \right\rangle$$

$$= \left\langle \delta_{\pi}(i_1, \ldots, i_k) \right\rangle$$

The Gram matrix being as well the correct one, we obtain the result.

Generally speaking, the above result is something quite powerful, because the main computation there, that of the inverse matrix  $W_{kN} = G_{kN}^1$ , can be run on an ordinary laptop, after implementing the formula of the Gram matrix, namely  $G_{kN}(\pi, \sigma) = N^{|\pi \vee \sigma|}$ , which is something quite easy to do. Thus, you can prove theorems about integrals over easy groups just by smoking cigars, and letting your computer do the work.

As a basic theoretical illustration, however, let us discuss the computation of the Weingarten function for the symmetric group  $S_N$ , which is something elementary.

For this purpose, we can use the explicit integration formula over  $S_N$ , which actually shows that the Weingarten formula is not needed in this case, which is as follows:

THEOREM 16.4. Consider the symmetric group  $S_N$ , with its standard coordinates coming from the standard embedding  $S_N \subset O_N$ , given by:

$$g_{ij} = \chi \left( \sigma \in S_N \middle| \sigma(j) = i \right)$$

The products of these coordinates span the algebra  $C(S_N)$ , and the arbitrary integrals over  $S_N$  are given, modulo linearity, by the formula

$$\int_{S_N} g_{i_1 j_1} \dots g_{i_k j_k} = \begin{cases} \frac{(N-|\ker i|)!}{N!} & \text{if } \ker i = \ker j\\ 0 & \text{otherwise} \end{cases}$$

where ker *i* denotes as usual the partition of  $\{1, ..., k\}$  whose blocks collect the equal indices of *i*, and where |.| denotes the number of blocks.

PROOF. The first assertion follows from the Stone-Weierstrass theorem, because the standard coordinates  $g_{ij}$  separate the points of  $S_N$ , and so the algebra  $\langle g_{ij} \rangle$  that they generate must be equal to the whole function algebra  $C(S_N)$ :

$$\langle g_{ij} \rangle = C(S_N)$$

Regarding now the second assertion, according to the definition of  $g_{ij}$ , the integrals in the statement are given by:

$$\int_{S_N} g_{i_1 j_1} \dots g_{i_k j_k} = \frac{1}{N!} \# \left\{ \sigma \in S_N \middle| \sigma(j_1) = i_1, \dots, \sigma(j_k) = i_k \right\}$$

Now observe that the existence of  $\sigma \in S_N$  as above requires:

 $i_m = i_n \iff j_m = j_n$ 

Thus, the above integral vanishes when:

$$\ker i \neq \ker j$$

Regarding now the case ker  $i = \ker j$ , if we denote by  $b \in \{1, \ldots, k\}$  the number of blocks of this partition ker  $i = \ker j$ , we have N - b points to be sent bijectively to N - b points, and so (N - b)! solutions, and the integral is  $\frac{(N-b)!}{N!}$ , as claimed.

The above result shows that the integration over  $S_N$  is something quite trivial, and so the computation of the Weingarten function should be something quite trivial too.

In practice now, in order to compute the Weingarten function for  $S_N$ , by using the above result, we will need some combinatorics, and more specifically the Möbius inversion formula. Let us begin with some standard definitions, as follows:

DEFINITION 16.5. Let P(k) be the set of partitions of  $\{1, \ldots, k\}$ , and let  $\pi, \sigma \in P(k)$ .

- (1) We write  $\pi \leq \sigma$  if each block of  $\pi$  is contained in a block of  $\sigma$ .
- (2) We let  $\pi \lor \sigma \in P(k)$  be the partition obtained by superposing  $\pi, \sigma$ .

As an illustration here, at k = 2 we have  $P(2) = \{||, \square\}$ , and we have:

$$|| \leq \Box$$

Also, at k = 3 we have  $P(3) = \{|||, \Box|, \Box, \Box, \Box, \Box\}$ , and the order relation is as follows:  $||| \le \Box|, \Box, \Box \le \Box \Box$ 

Observe also that we have the following inequalities:

$$\pi, \sigma \le \pi \lor \sigma$$

In fact, the partition  $\pi \lor \sigma$  is the smallest possible one with this property. Due to this fact, this partition  $\pi \lor \sigma$  is called supremum of  $\pi, \sigma$ .

We can now introduce the Möbius function, as follows:

DEFINITION 16.6. The Möbius function of any lattice, and so of P, is given by

$$\mu(\pi, \sigma) = \begin{cases} 1 & \text{if } \pi = \sigma \\ -\sum_{\pi \le \tau < \sigma} \mu(\pi, \tau) & \text{if } \pi < \sigma \\ 0 & \text{if } \pi \nleq \sigma \end{cases}$$

with the construction being performed by recurrence.

As an illustration here, let us go back to the set of 2-point partitions,  $P(2) = \{||, \square\}$ . We have by definition:

$$\mu(||,||) = \mu(\sqcap,\sqcap) = 1$$

Also, we know that we have  $|| < \Box$ , with no intermediate partition in between, and so the above recurrence procedure gives:

$$\mu(||, \Box) = -\mu(||, ||) = -1$$

Finally, we have  $\sqcap \not\leq \mid \mid$ , and so  $\mu(\sqcap, \mid \mid) = 0$ . Thus, as a conclusion, the Möbius matrix  $M_{\pi\sigma} = \mu(\pi, \sigma)$  of the lattice  $P(2) = \{\mid \mid, \sqcap\}$  is as follows:

$$M = \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix}$$

The interest in the Möbius function comes from the Möbius inversion formula:

$$f(\sigma) = \sum_{\pi \le \sigma} g(\pi) \implies g(\sigma) = \sum_{\pi \le \sigma} \mu(\pi, \sigma) f(\pi)$$

In linear algebra terms, the statement and proof of this formula are as follows:

THEOREM 16.7. The inverse of the adjacency matrix of P, given by

$$A_{\pi\sigma} = \begin{cases} 1 & \text{if } \pi \leq \sigma \\ 0 & \text{if } \pi \nleq \sigma \end{cases}$$

is the Möbius matrix of P, given by  $M_{\pi\sigma} = \mu(\pi, \sigma)$ .

**PROOF.** This is well-known, coming for instance from the fact that A is upper triangular. Indeed, when inverting, we are led into the recurrence from Definition 16.6.

As a first illustration, for P(2) the formula  $M = A^{-1}$  appears as follows:

$$\begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}^{-1}$$

Also, for  $P(3) = \{ |||, \Box|, \Box, \Box, \Box \}$  the formula  $M = A^{-1}$  reads:

$$\begin{pmatrix} 1 & -1 & -1 & -1 & 2 \\ 0 & 1 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}^{-1}$$

In general, the formula  $M = A^{-1}$  is quite similar.

With the above results in hand, we can now compute the Weingarten function of  $S_N$ , and also find a precise estimate for it, as follows:

THEOREM 16.8. For  $S_N$  the Weingarten function is given by

$$W_{kN}(\pi,\sigma) = \sum_{\tau \le \pi \land \sigma} \mu(\tau,\pi) \mu(\tau,\sigma) \frac{(N-|\tau|)}{N!}$$

and satisfies the following estimate,

$$W_{kN}(\pi,\sigma) = N^{-|\pi\wedge\sigma|}(\mu(\pi\wedge\sigma,\pi)\mu(\pi\wedge\sigma,\sigma) + O(N^{-1}))$$

with  $\mu$  being the Möbius function of P(k).

**PROOF.** The first assertion follows from the Weingarten formula, namely:

$$\int_{S_N} u_{i_1 j_1} \dots u_{i_k j_k} = \sum_{\pi, \sigma \in P(k)} \delta_{\pi}(i) \delta_{\sigma}(j) W_{kN}(\pi, \sigma)$$

Indeed, in this formula the integrals on the left are known, from the explicit integration formula over  $S_N$  that we established above, namely:

$$\int_{S_N} g_{i_1 j_1} \dots g_{i_k j_k} = \begin{cases} \frac{(N-|\ker i|)!}{N!} & \text{if } \ker i = \ker j\\ 0 & \text{otherwise} \end{cases}$$

But this allows the computation of the right term, via the Möbius inversion formula, explained above. As for the second assertion, this follows from the first one.  $\Box$ 

As an illustration, let us record the formulae at k = 2, 3. At k = 2, with indices  $||, \Box$ , and with the convention that  $\approx$  means componentwise dominant term, we have:

$$W_{2N} \approx \begin{pmatrix} N^{-2} & -N^{-2} \\ -N^{-2} & N^{-1} \end{pmatrix}$$

At k = 3 now, with indices  $|||, |\Box, \Box|, \Box, \Box \Box$ , and same meaning for  $\approx$ , we have:

$$W_{3N} \approx \begin{pmatrix} N^{-3} & -N^{-3} & -N^{-3} & -N^{-3} & 2N^{-3} \\ -N^{-3} & N^{-2} & N^{-3} & N^{-3} & -N^{-2} \\ -N^{-3} & N^{-3} & N^{-2} & N^{-3} & -N^{-2} \\ -N^{-3} & N^{-3} & N^{-3} & N^{-2} & -N^{-2} \\ 2N^{-3} & -N^{-2} & -N^{-2} & -N^{-2} & N^{-1} \end{pmatrix}$$

We will be back to all this later, with results about the orthogonal group  $O_N$  and about some other easy groups as well, where the Weingarten function is in general not explicitly computable, but where some useful estimates are still possible.

# 16b. Laws of characters

As an application of the above, let us discuss now the computation of the asymptotic laws of characters. In the easy group case, we do not need in fact the full power of the Weingarten formula, because we have the following result:

THEOREM 16.9. Assuming that  $G \subset U_N$  is an easy group, coming from a category of partitions

$$D = (D(k, l))$$

The moments of the main character are given by

$$\int_{G} \chi^{k} = \dim \left( span \left( \xi_{\pi} \big| \pi \in D(k) \right) \right)$$

where  $D(k) = D(\emptyset, k)$ , and where for  $\pi \in D(k)$  we use the notation  $\xi_{\pi} = T_{\pi}$ .

**PROOF.** We recall that for an easy group  $G \subset U_N$ , coming from a category of partitions D = (D(k, l)), we have by definition equalities as follows:

$$Hom(u^{\otimes k}, u^{\otimes l}) = span\left(T_{\pi} \middle| \pi \in D(k, l)\right)$$

By interchanging  $k \leftrightarrow l$  in this formula, and then setting  $l = \emptyset$ , we obtain:

$$Fix(u^{\otimes k}) = span\left(\xi_{\pi} \middle| \pi \in D(k)\right)$$

Now since by the Peter-Weyl theory integrating a character amounts in counting the fixed points, we are led to the conclusion in the statement.  $\hfill \Box$ 

In order to investigate the linear independence questions for the vectors  $\xi_{\pi}$ , we will use the Gram matrix of these vectors. We have the following result, to start with:

PROPOSITION 16.10. The Gram matrix  $G_{kN}(\pi, \sigma) = \langle \xi_{\pi}, \xi_{\sigma} \rangle$  is given by

$$G_{kN}(\pi,\sigma) = N^{|\pi \vee \sigma|}$$

where |.| is the number of blocks.

**PROOF.** According to the formula of the vectors  $\xi_{\pi}$ , we have:

$$<\xi_{\pi},\xi_{\sigma}> = \sum_{i_{1}...i_{k}} \delta_{\pi}(i_{1},...,i_{k})\delta_{\sigma}(i_{1},...,i_{k})$$
$$= \sum_{i_{1}...i_{k}} \delta_{\pi\vee\sigma}(i_{1},...,i_{k})$$
$$= N^{|\pi\vee\sigma|}$$

Thus, we have obtained the formula in the statement.

Next in line, we have the following key result:

**PROPOSITION 16.11.** The Gram matrix is given by  $G_{kN} = AL$ , where

$$L(\pi, \sigma) = \begin{cases} N(N-1)\dots(N-|\pi|+1) & \text{if } \sigma \le \pi\\ 0 & \text{otherwise} \end{cases}$$

and where  $A = M^{-1}$  is the adjacency matrix of P(k).

**PROOF.** We have indeed the following computation:

$$N^{|\pi \vee \sigma|} = \# \left\{ i_1, \dots, i_k \in \{1, \dots, N\} \middle| \ker i \ge \pi \vee \sigma \right\}$$
$$= \sum_{\tau \ge \pi \vee \sigma} \# \left\{ i_1, \dots, i_k \in \{1, \dots, N\} \middle| \ker i = \tau \right\}$$
$$= \sum_{\tau \ge \pi \vee \sigma} N(N-1) \dots (N-|\tau|+1)$$

According to Proposition 16.10 and to the definition of A, L, this formula reads:

$$(G_{kN})_{\pi\sigma} = \sum_{\tau \ge \pi} L_{\tau\sigma}$$
$$= \sum_{\tau} A_{\pi\tau} L_{\tau\sigma}$$
$$= (AL)_{\pi\sigma}$$

Thus, we obtain in this way the formula in the statement.

As an illustration for the above result, at k = 2 we have  $P(2) = \{||, \square\}$ , and the above formula  $G_{kN} = AL$  appears as follows:

$$\begin{pmatrix} N^2 & N \\ N & N \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} N^2 - N & 0 \\ N & N \end{pmatrix}$$

At k = 3 now, we have  $P(3) = \{|||, \Box|, \Box, \Box, \Box\}$ , and the Gram matrix is:

$$G_3 = \begin{pmatrix} N^3 & N^2 & N^2 & N^2 & N \\ N^2 & N^2 & N & N & N \\ N^2 & N & N^2 & N & N \\ N^2 & N & N & N^2 & N \\ N & N & N & N & N \end{pmatrix}$$

Regarding  $L_3$ , this can be computed by writing down the matrix  $E_3(\pi, \sigma) = \delta_{\sigma \leq \pi} |\pi|$ , and then replacing each entry by the corresponding polynomial in N. We reach to the conclusion that the product  $A_3L_3$  is as follows, producing the above matrix  $G_3$ :

$$A_{3}L_{3} = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} N^{3} - 3N^{2} + 2N & 0 & 0 & 0 & 0 \\ N^{2} - N & N^{2} - N & 0 & 0 & 0 \\ N^{2} - N & 0 & N^{2} - N & 0 & 0 \\ N^{2} - N & 0 & 0 & N^{2} - N & 0 \\ N & N & N & N & N \end{pmatrix}$$

In general, the formula  $G_k = A_k L_k$  appears a bit in the same way, with  $A_k$  being binary and upper triangular, and with  $L_k$  depending on N, and being lower triangular.

With the above result in hand, we can now investigate the linear independence properties of the vectors  $\xi_{\pi}$ . To be more precise, we have the following result:

THEOREM 16.12. The determinant of the Gram matrix  $G_{kN}$  is given by

$$\det(G_{kN}) = \prod_{\pi \in P(k)} \frac{N!}{(N - |\pi|)!}$$

and in particular, for  $N \ge k$ , the vectors  $\{\xi_{\pi} | \pi \in P(k)\}$  are linearly independent.

**PROOF.** According to the formula in Proposition 16.11 above, we have:

 $\det(G_{kN}) = \det(A) \det(L)$ 

Now if we order P(k) as above, with respect to the number of blocks, and then lexicographically, we see that A is upper triangular, and that L is lower triangular.

Thus  $\det(A)$  can be computed simply by making the product on the diagonal, and we obtain 1. As for  $\det(L)$ , this can computed as well by making the product on the diagonal, and we obtain the number in the statement, with the technical remark that in the case N < k the convention is that we obtain a vanishing determinant.

Now back to the laws of characters, we can formulate:

THEOREM 16.13. For an easy group  $G = (G_N)$ , coming from a category of partitions D = (D(k, l)), the asymptotic moments of the main character are given by

$$\lim_{N \to \infty} \int_{G_N} \chi^k = \# D(k)$$

where  $D(k) = D(\emptyset, k)$ , with the limiting sequence on the left consisting of certain integers, and being stationary at least starting from the k-th term.

PROOF. This follows indeed from the general formula from Theorem 16.9, by using the linear independence result from Theorem 16.12.  $\hfill \Box$ 

Our next purpose will be that of understanding what happens for the basic classes of easy quantum groups. In order to deal with the orthogonal case, we will need:

**PROPOSITION 16.14.** We have the formula

$$\#P_2(2k) = (2k)!!$$

with the following convention:

$$(2k)!! = 1.3.5...(2k-3)(2k-1)$$

**PROOF.** We have to count the pairings of  $\{1, \ldots, 2k\}$ . But, in order to construct such a pairing, we have 2k - 1 choices for the pair of the number 1, then 2k - 3 choices for the pair of the next number left, and so on. Thus, we obtain (2k)!!, as claimed.

In order to deal with the unitary case, we will need:

**PROPOSITION 16.15.** The moments of the complex normal law are the numbers

$$M_k(G_t) = \sum_{\pi \in \mathcal{P}_2(k)} t^{|\pi|}$$

where  $\mathcal{P}_2(k)$  are the matching pairings of  $\{1, \ldots, k\}$ , and |.| is the number of blocks.

**PROOF.** This is something well-known, which can be done in several steps, as follows:

(1) We recall from the above that the moments of the real Gaussian law  $g_1$ , with respect to integer exponents  $k \in \mathbb{N}$ , are the following numbers:

$$m_k = |P_2(k)|$$

Numerically, we have the following formula, explained as well in the above:

$$m_k = \begin{cases} k!! & (k \text{ even}) \\ 0 & (k \text{ odd}) \end{cases}$$

(2) We will show now that in what concerns the complex Gaussian law  $G_1$ , similar results hold. Numerically, we will prove that we have the following formula, where a colored integer  $k = \circ \bullet \circ \circ \ldots$  is called uniform when it contains the same number of  $\circ$  and  $\bullet$ , and where  $|k| \in \mathbb{N}$  is the length of such a colored integer:

$$M_k = \begin{cases} (|k|/2)! & (k \text{ uniform}) \\ 0 & (k \text{ not uniform}) \end{cases}$$

Now since the matching partitions  $\pi \in \mathcal{P}_2(k)$  are counted by exactly the same numbers, and this for trivial reasons, we will obtain the formula in the statement, namely:

$$M_k = |\mathcal{P}_2(k)|$$

(3) This was for the plan. In practice now, we must compute the moments, with respect to colored integer exponents  $k = \circ \bullet \circ \circ \ldots$ , of the variable in the statement:

$$c = \frac{1}{\sqrt{2}}(a+ib)$$

As a first observation, in the case where such an exponent  $k = \circ \bullet \bullet \circ \ldots$  is not uniform in  $\circ, \bullet$ , a rotation argument shows that the corresponding moment of c vanishes. To be more precise, the variable c' = wc can be shown to be complex Gaussian too, for any  $w \in \mathbb{C}$ , and from  $M_k(c) = M_k(c')$  we obtain  $M_k(c) = 0$ , in this case.

(4) In the uniform case now, where  $k = \circ \bullet \bullet \circ \ldots$  consists of p copies of  $\circ$  and p copies of  $\bullet$ , the corresponding moment can be computed as follows:

$$M_{k} = \int (c\bar{c})^{p}$$

$$= \frac{1}{2^{p}} \int (a^{2} + b^{2})^{p}$$

$$= \frac{1}{2^{p}} \sum_{s} {p \choose s} \int a^{2s} \int b^{2p-2s}$$

$$= \frac{1}{2^{p}} \sum_{s} {p \choose s} (2s)!! (2p - 2s)!!$$

$$= \frac{1}{2^{p}} \sum_{s} \frac{p!}{s!(p-s)!} \cdot \frac{(2s)!}{2^{s}s!} \cdot \frac{(2p - 2s)!}{2^{p-s}(p-s)!}$$

$$= \frac{p!}{4^{p}} \sum_{s} {2s \choose s} {2p - 2s \choose p-s}$$

(5) In order to finish now the computation, let us recall that we have the following formula, coming from the generalized binomial formula, or from the Taylor formula:

$$\frac{1}{\sqrt{1+t}} = \sum_{k=0}^{\infty} \binom{2k}{k} \left(\frac{-t}{4}\right)^k$$

By taking the square of this series, we obtain the following formula:

$$\frac{1}{1+t} = \sum_{ks} \binom{2k}{k} \binom{2s}{s} \left(\frac{-t}{4}\right)^{k+s}$$
$$= \sum_{p} \left(\frac{-t}{4}\right)^{p} \sum_{s} \binom{2s}{s} \binom{2p-2s}{p-s}$$

Now by looking at the coefficient of  $t^p$  on both sides, we conclude that the sum on the right equals  $4^p$ . Thus, we can finish the moment computation in (4), as follows:

$$M_p = \frac{p!}{4^p} \times 4^p = p!$$

(6) As a conclusion, if we denote by |k| the length of a colored integer  $k = \circ \bullet \circ \circ \ldots$ , the moments of the variable c in the statement are given by:

$$M_k = \begin{cases} (|k|/2)! & (k \text{ uniform}) \\ 0 & (k \text{ not uniform}) \end{cases}$$

On the other hand, the numbers  $|\mathcal{P}_2(k)|$  are given by exactly the same formula. Indeed, in order to have matching pairings of k, our exponent  $k = \circ \bullet \circ \circ \ldots$  must be uniform, consisting of p copies of  $\circ$  and p copies of  $\bullet$ , with p = |k|/2. But then the matching pairings of k correspond to the permutations of the  $\bullet$  symbols, as to be matched with  $\circ$ symbols, and so we have p! such matching pairings. Thus, we have the same formula as for the moments of c, and we are led to the conclusion in the statement. 

With these preliminaries in hand, we can now state and prove:

THEOREM 16.16. In the  $N \to \infty$  limit, the law of the main character

$$\chi_u = \sum_{i=1}^N u_{ii}$$

for the orthogonal and unitary groups is as follows:

- (1) For  $O_N$  we obtain a Gaussian law,  $\frac{1}{\sqrt{2\pi}}e^{-x^2/2}dx$ . (2) For  $U_N$  we obtain a complex Gaussian law  $G_1$ .

**PROOF.** These results follow from the general formula from Theorem 16.13 above, by using the knowledge of the associated categories of partitions, then the counting formulae from Proposition 16.14 and Proposition 16.15, and finally by doing some calculus:

(1) For  $O_N$  the associated category of partitions is  $P_2$ , so the asymptotic moments of the main character are as follows, with the convention k!! = 0 when k is odd:

$$M_k = \# P_2(k) = k!!$$

Thus, we obtain the real Gaussian law, as stated.

(2) This follows from some combinatorics. To be more precise, the asymptotic moments of the main character, with respect to the colored integers, are as follows:

$$M_k = \# \mathcal{P}_2(k)$$

Thus, we obtain the complex Gaussian law, as stated.

More generally now, we have the following result:

THEOREM 16.17. With  $N \to \infty$ , the laws of main character is as follows:

- (1) For  $O_N$  we obtain the Gaussian law  $g_1$ .
- (2) For  $U_N$  we obtain the complex Gaussian law  $G_1$ .
- (3) For  $S_N$  we obtain the Poisson law  $p_1$ .
- (4) For  $H_N$  we obtain the Bessel law  $b_1$ .
- (5) For  $H_N^s$  we obtain the generalized Bessel law  $b_1^s$ .
- (6) For  $K_N$  we obtain the complex Bessel law  $B_1$ .

Also, for  $B_N, C_N$  and for  $Sp_N$  we obtain modified normal laws.

PROOF. We already know the results for  $O_N$  and for  $U_N$ , from Theorem 16.16 above. In general, the proof is similar, by counting the partitions in the associated category of partitions, and then doing some calculus, based on the various moment results for the laws in the statement, coming from the general theory developed in the above.

There are of course many other applications of the above technology.

# 16c. Truncated characters

In order to fully solve the questions left open in chapters 9-12 above, we have to discuss now the more advanced question of computing the laws of truncated characters.

First, we have the following formula, in the general easy group setting:

**PROPOSITION 16.18.** The moments of truncated characters are given by the formula

$$\int_G (g_{11} + \ldots + g_{ss})^k = Tr(W_{kN}G_{ks})$$

where  $G_{kN}$  and  $W_{kN} = G_{kN}^{-1}$  are the associated Gram and Weingarten matrices.

**PROOF.** We have indeed the following computation:

$$\int_{G} (g_{11} + \ldots + g_{ss})^{k} = \sum_{i_{1}=1}^{s} \ldots \sum_{i_{k}=1}^{s} \int_{G} g_{i_{1}i_{1}} \ldots g_{i_{k}i_{k}}$$
$$= \sum_{\pi,\sigma \in D(k)} W_{kN}(\pi,\sigma) \sum_{i_{1}=1}^{s} \ldots \sum_{i_{k}=1}^{s} \delta_{\pi}(i) \delta_{\sigma}(i)$$
$$= \sum_{\pi,\sigma \in D(k)} W_{kN}(\pi,\sigma) G_{ks}(\sigma,\pi)$$
$$= Tr(W_{kN}G_{ks})$$

Thus, we have obtained the formula in the statement.

In order to process now the above formula, and reach to concrete results, we can impose the uniformity condition from chapter 15, originally used there for classification purposes. Let us recall indeed from there that we have:

DEFINITION 16.19. An easy group  $G = (G_N)$ , coming from a category of partitions  $D \subset P$ , is called uniform if it satisfies the following equivalent conditions:

- (1)  $G_{N-1} = G_N \cap U_{N-1}$ , via the embedding  $U_{N-1} \subset U_N$  given by  $u \to diag(u, 1)$ .
- (2)  $G_{N-1} = G_N \cap U_{N-1}$ , via the N possible diagonal embeddings  $U_{N-1} \subset U_N$ .
- (3) D is stable under the operation which consists in removing blocks.

Here the equivalence between the above three conditions is something standard, obtained by doing some combinatorics, and this was discussed in chapter 15 above.

In what follows we will be mostly interested in the condition (3) above, which makes the link with our computations for truncated characters, and simplifies them. To be more precise, by imposing the uniformity condition we obtain:

THEOREM 16.20. For a uniform easy group  $G = (G_N)$ , we have the formula

$$\lim_{N \to \infty} \int_{G_N} \chi_t^k = \sum_{\pi \in D(k)} t^{|\pi|}$$

with  $D \subset P$  being the associated category of partitions.

**PROOF.** We use the general moment formula from Proposition 16.18, namely:

$$\int_G (g_{11} + \ldots + g_{ss})^k = Tr(W_{kN}G_{ks})$$

By setting s = [tN], with t > 0 being a given parameter, this formula becomes:

$$\int_{G_N} \chi_t^k = Tr(W_{kN}G_{k[tN]})$$

The point now is that in the uniform case the Gram and Weingarten matrices are asymptotically diagonal, and this leads to the formula in the statement.  $\Box$ 

We can now improve our character results, as follows:

THEOREM 16.21. With  $N \to \infty$ , the laws of truncated characters are as follows:

- (1) For  $O_N$  we obtain the Gaussian law  $g_t$ .
- (2) For  $U_N$  we obtain the complex Gaussian law  $G_t$ .
- (3) For  $S_N$  we obtain the Poisson law  $p_t$ .
- (4) For  $H_N$  we obtain the Bessel law  $b_t$ .
- (5) For  $H_N^s$  we obtain the generalized Bessel law  $b_t^s$ .
- (6) For  $K_N$  we obtain the complex Bessel law  $B_t$ .

Also, for  $B_N, C_N$  and for  $Sp_N$  we obtain modified normal laws.

**PROOF.** We use the formula in Theorem 16.20 above, namely:

$$\lim_{N \to \infty} \int_{G_N} \chi_t^k = \sum_{\pi \in D(k)} t^{|\pi|}$$

By doing now some combinatorics, this gives the results.

There are of course many other applications of the above technology.

## 16D. STANDARD ESTIMATES

# 16d. Standard estimates

We have seen in the above that the Weingarten calculus is something very efficient in dealing with various probability questions over the easy groups  $G \subset U_N$ .

We discuss now, as a continuation of this, a number of more advanced aspects of the Weingarten function combinatorics. We will be mostly interested in the case  $G = O_N$ .

To be more precise, we will be interested in the computation of the polynomial integrals over  $O_N$ . These are best introduced in a "rectangular way", as follows:

DEFINITION 16.22. Associated to any matrix  $a \in M_{p \times q}(\mathbb{N})$  is the integral

$$I(a) = \int_{O_N} \prod_{i=1}^{p} \prod_{j=1}^{q} u_{ij}^{a_{ij}} \, du$$

with respect to the Haar measure of  $O_N$ , where  $N \ge p, q$ .

We can of course complete our matrix with 0 values, as to always deal with square matrices,  $a \in M_N(\mathbb{N})$ . However, the parameters p, q are very useful, because they measure the "complexity" of the problem, as shown for instance by the result below.

Let  $x!! = (x-1)(x-3)(x-5)\dots$ , with the product ending at 1 or 2. We have:

THEOREM 16.23. At p = 1 we have the formula

$$I(a_1 \quad \dots \quad a_q) = \varepsilon \cdot \frac{(n-1)!!a_1!!\dots a_q!!}{(n+\Sigma a_i-1)!!}$$

where  $\varepsilon = 1$  if all  $a_i$  are even, and  $\varepsilon = 0$  otherwise.

**PROOF.** This follows from the fact that the first slice of  $O_N$  is isomorphic to the real sphere  $S_{\mathbb{R}}^{N-1}$ . Indeed, this gives the following formula:

$$I(a_1 \ldots a_q) = \int_{S^{N-1}_{\mathbb{R}}} x_1^{a_1} \ldots x_q^{a_q} dx$$

This latter integral can be computed by using polar coordinates, and we obtain the formula in the statement. See e.g. [5].

Another instructive computation, as well of trigonometric nature, is the one at N = 2. We have here the following result, which completely solves the problem in this case:

THEOREM 16.24. At N = 2 we have the formula

$$I\begin{pmatrix} a & b \\ c & d \end{pmatrix} = \varepsilon \cdot \frac{(a+d)!!(b+c)!!}{(a+b+c+d+1)!!}$$

where  $\varepsilon = 1$  if a, b, c, d are even,  $\varepsilon = -1$  is a, b, c, d are odd, and  $\varepsilon = 0$  if not.

PROOF. When computing the integral over  $O_2$ , we can restrict the integration to  $SO_2 = \mathbb{T}$ , then further restrict the integration to the first quadrant. We get:

$$I\begin{pmatrix}a&b\\c&d\end{pmatrix} = \varepsilon \cdot \frac{2}{\pi} \int_0^{\pi/2} (\cos t)^{a+d} (\sin t)^{b+c} dt$$

This gives the formula in the statement.

The above computations might tend to suggest that I(a) always decomposes as a product of factorials. This is far from being true, but in the 2 × 2 case it is known that I(a) decomposes as a quite reasonable sum of products of factorials.

Let us discuss now the representation theory approach to the computation of I(a). The Weingarten formula reformulates as follows:

THEOREM 16.25. We have the Weingarten formula

$$I(a) = \sum_{\pi,\sigma} \delta_{\pi}(a_l) \delta_{\sigma}(a_r) W_{kN}(\pi,\sigma)$$

where  $k = \sum a_{ij}/2$ , and where the multi-indices  $a_l/a_r$  are defined as follows:

- (1) Start with  $a \in M_{p \times q}(\mathbb{N})$ , and replace each *ij*-entry by  $a_{ij}$  copies of i/j.
- (2) Read this matrix in the usual way, as to get the multi-indices  $a_l/a_r$ .

**PROOF.** This is simply a reformulation of the Weingarten formula. Indeed, according to our definitions, the integral in the statement is given by:

$$I(a) = \int_{O_n} \underbrace{u_{11} \dots u_{11}}_{a_{11}} \underbrace{u_{12} \dots u_{12}}_{a_{12}} \dots \underbrace{u_{pq} \dots u_{pq}}_{a_{pq}} du$$

Thus what we have here is an integral as in the Weingarten formula, the multi-indices being as follows:

$$a_{l} = (\underbrace{1 \dots 1}_{a_{11}} \underbrace{1 \dots 1}_{a_{12}} \dots \underbrace{p \dots p}_{a_{pq}})$$
$$a_{r} = (\underbrace{1 \dots 1}_{a_{11}} \underbrace{2 \dots 2}_{a_{12}} \dots \underbrace{q \dots q}_{a_{pq}})$$

The result follows now from the Weingarten formula.

We are now in position of deriving a first general corollary from our study. This extends the vanishing results appearing before:

PROPOSITION 16.26. We have I(a) = 0, unless the matrix a is "admissible", in the sense that all p + q sums on its rows and columns are even numbers.

384

PROOF. Observe first that the left multi-index associated to a consists of  $k_1 = \sum a_{1j}$  copies of 1,  $k_2 = \sum a_{2j}$  copies of 2, and so on, up to  $k_p = \sum a_{pj}$  copies of p. In the case where one of these numbers is odd we have  $\delta_{\pi}(a) = 0$  for any  $\pi$ , and this gives I(a) = 0.

A similar argument with the right multi-index associated to a shows that the sums on the columns of a must be even as well, and we are done.

A natural question now is whether the converse of Proposition 16.26 holds, and if so, the question of computing the sign of I(a) appears as well. These are both quite subtle questions, and we begin our investigations with a  $N \to \infty$  study. We have:

THEOREM 16.27. The Weingarten matrix is asymptotically diagonal, in the sense that:

$$W_{kN}(\pi,\sigma) = N^{-k}(\delta_{\pi\sigma} + O(N^{-1}))$$

Moreover, the  $O(N^{-1})$  remainder is asymptotically smaller that  $(2k/e)^k N^{-1}$ .

**PROOF.** It is convenient, for the purposes of this proof, to drop the indices k, N. We know that the Gram matrix is given by  $G(\pi, \sigma) = N^{loops(\pi, \sigma)}$ , so we have:

$$G(\pi, \sigma) = \begin{cases} N^k & \text{for } \pi = \sigma \\ N, N^2, \dots, N^{k-1} & \text{for } \pi \neq \sigma \end{cases}$$

Thus the Gram matrix is of the following form, with  $||H||_{\infty} \leq N^{-1}$ :

$$G = N^k (I + H)$$

Now recall that for any  $K \times K$  complex matrix X, we have the following lineup of standard inequalities, which are all elementary:

$$||X||_{\infty} \leq ||X||$$
  
$$\leq ||X||_{2}$$
  
$$\leq K||X||_{\infty}$$

In the case of our matrix H, the size is K = (2k)!!, so we have:

$$||H|| < KN^{-1}$$

We can use now the following formula:

$$(I+H)^{-1} = I - H + H^2 - H^3 + \dots$$

We conclude from this that we have:

$$||I - (I + H)^{-1}|| \le ||H||/(1 - ||H||)$$

Thus we have the following estimate:

$$||I - N^{k}W||_{\infty} = ||I - (1 + H)^{-1}||_{\infty}$$
  

$$\leq ||I - (1 + H)^{-1}||$$
  

$$\leq ||H||/(1 - ||H||)$$
  

$$\leq KN^{-1}/(1 - KN^{-1})$$
  

$$= K/(N - K)$$

Together with the standard estimate  $K \approx (2k/e)^k$ , this gives the result.

We have the following result:

THEOREM 16.28. We have the estimate

$$I(a) = N^{-k} \left( \prod_{i=1}^{p} \prod_{j=1}^{q} a_{ij} !! + O(N^{-1}) \right)$$

when all  $a_{ij}$  are even, and  $I(a) = O(N^{-k-1})$  otherwise.

**PROOF.** By using the above results, we obtain the following estimate:

$$I(a) = \sum_{\pi,\sigma} \delta_{\pi}(a_l) \delta_{\sigma}(a_r) W_{kN}(\pi,\sigma)$$
  
=  $n^{-k} \sum_{\pi,\sigma} \delta_{\pi}(a_l) \delta_{\sigma}(a_r) (\delta_{\pi\sigma} + O(N^{-1}))$   
=  $N^{-k} \left( \# \{\pi | \delta_{\pi}(a_l) = \delta_{\pi}(a_r) = 1 \} + O(N^{-1}) \right)$ 

In order to count the partitions appearing in the set on the right, let us go back to the multi-indices  $a_l, a_r$  described above. It is convenient to view both these multi-indices in a rectangular way, as follows:

$$a_{l} = \begin{pmatrix} \underbrace{1 \dots 1}_{a_{11}} & \dots & \underbrace{1 \dots 1}_{a_{1q}} \\ \dots & \dots & \dots \\ \underbrace{p \dots p}_{a_{p1}} & \dots & \underbrace{p \dots p}_{a_{pq}} \end{pmatrix}$$
$$a_{r} = \begin{pmatrix} \underbrace{1 \dots 1}_{a_{11}} & \dots & \underbrace{q \dots q}_{a_{1q}} \\ \dots & \dots & \dots \\ \underbrace{1 \dots 1}_{a_{p1}} & \dots & \underbrace{p \dots p}_{a_{pq}} \end{pmatrix}$$

In other words, the multi-indices  $a_l/a_r$  are now simply obtained from the matrix a by "dropping" from each entry  $a_{ij}$  a sequence of  $a_{ij}$  numbers, all equal to i/j.

## 16D. STANDARD ESTIMATES

These two multi-indices, now in matrix form, have total length  $2k = \sum a_{ij}$ . We agree to view as well any pairing of  $\{1, \ldots, 2k\}$  in matrix form, by following the same convention.

With this picture, the pairings  $\pi$  which contribute are simply those interconnecting sequences of indices "dropped" from the same  $a_{ij}$ , and this gives the following results:

(1) In the case where one of the entries  $a_{ij}$  is odd, there is no pairing that can contribute to the leading term under consideration, so we have  $I(a) = O(N^{-k-1})$ , and we are done.

(2) In the case where all the entries  $a_{ij}$  are even, the pairings that contribute to the leading term are those connecting points inside the pq "dropped" sets, i.e. are made out of a pairing of  $a_{11}$  points, a pairing of  $a_{12}$  points, and so on, up to a pairing of  $a_{pq}$  points. Now since an x-point set has x!! pairings, this gives the formula in the statement.

In order to further advance, let us formulate a key definition, as follows:

DEFINITION 16.29. The Brauer space  $D_k$  is defined as follows:

- (1) The points are the Brauer diagrams, i.e. the pairings of  $\{1, 2, \ldots, 2k\}$ .
- (2) The distance function is given by  $d(\pi, \sigma) = k loops(\pi, \sigma)$ .

It is indeed well-known, and elementary to check, that d satisfies the usual axioms for a distance function. This actually comes from some general categorical properties of the Brauer diagrams, which are valid in much more general situations. See [4], [22].

The Brauer space, which will play an important role in what follows, is by definition a metric space having (2k)!! = 1.3.5...(2k-1) points. An interesting question is to find a "geometric" realization of this space. This will be discussed later on.

The series expansion of the Weingarten function in terms of paths on the Brauer space was originally found by Collins in [15] in the unitary case, then by Collins and Śniady [20] in the orthogonal case. We present below a slightly modified statement, along with a complete proof, by using a somewhat lighter formalism:

THEOREM 16.30. The Weingarten function  $W_{kN}$  has a series expansion in  $N^{-1}$ ,

$$W_{kN}(\pi,\sigma) = N^{-k-d(\pi,\sigma)} \sum_{g=0}^{\infty} K_g(\pi,\sigma) N^{-g}$$

where the objects on the right are defined as follows:

- (1) A path from  $\pi$  to  $\sigma$  is a sequence  $p = [\pi = \tau_0 \neq \tau_1 \neq \ldots \neq \tau_r = \sigma]$ .
- (2) The signature of such a path is + when r is even, and when r is odd.
- (3) The geodesicity defect of such a path is  $g(p) = \sum_{i=1}^{r} d(\tau_{i-1}, \tau_i) d(\pi, \sigma)$ .
- (4)  $K_g$  counts the signed paths from  $\pi$  to  $\sigma$ , with geodesicity defect g.

**PROOF.** Let us go back to the proof of our main estimate so far, established in the above. We can write the Gram matrix in the following way:

$$G_{kn} = N^{-k}(I+H)$$

In terms of the Brauer space distance, the formula of H is simply:

$$H(\pi, \sigma) = \begin{cases} 0 & \text{for } \pi = \sigma \\ N^{-d(\pi, \sigma)} & \text{for } \pi \neq \sigma \end{cases}$$

Consider now the set  $P_r(\pi, \sigma)$  of r-paths between  $\pi$  and  $\sigma$ . According to the usual rule of matrix multiplication, the powers of H are given by:

$$H^{r}(\pi,\sigma) = \sum_{p \in P_{r}(\pi,\sigma)} H(\tau_{0},\tau_{1}) \dots H(\tau_{r-1},\tau_{r})$$
$$= \sum_{p \in P_{r}(\pi,\sigma)} N^{-d(\pi,\sigma)-g(p)}$$

We can use now the following standard inversion formula:

$$(1+H)^{-1} = 1 - H + H^2 - H^3 + \dots$$

By using this formula, we obtain:

$$W_{kN}(\pi,\sigma) = N^{-k} \sum_{r=0}^{\infty} (-1)^r H^r(\pi,\sigma)$$
  
=  $N^{-k-d(\pi,\sigma)} \sum_{r=0}^{\infty} \sum_{p \in P_r(\pi,\sigma)} (-1)^r N^{-g(p)}$ 

Now by rearranging the various terms of the double sum according to their geodesicity defect g = g(p), this gives the following formula:

$$W_{kN}(\pi,\sigma) = N^{-k-d(\pi,\sigma)} \sum_{g=0}^{\infty} K_g(\pi,\sigma) N^{-g}$$

Thus, we have obtained the formula in the statement.

In order to discuss now the I(a) reformulation of the above result, it is convenient to use the total length of a path, defined as follows:

$$d(p) = \sum_{i=1}^{r} d(\tau_{i-1}, \tau_i)$$

Observe that, in terms of this quantity, we have the following formula:

$$d(p) = d(\pi, \sigma) + g(p)$$

With these conventions, we have the following result:

THEOREM 16.31. The integral I(a) has a series expansion in  $N^{-1}$  of the form

$$I(a) = N^{-k} \sum_{d=0}^{\infty} H_d(a) N^{-d}$$

where the coefficient on the right can be interpreted as follows:

- (1) Starting from  $a \in M_{p \times q}(\mathbb{N})$ , construct the multi-indices  $a_l, a_r$  as usual.
- (2) Call a path "a-admissible" if its endpoints satisfy  $\delta_{\pi}(a_l) = 1$  and  $\delta_{\sigma}(a_r) = 1$ .
- (3) Then  $H_d(a)$  counts all a-admissible signed paths in  $D_k$ , of total length d.

**PROOF.** We combine first the above results:

$$I(a) = \sum_{\pi,\sigma} \delta_{\pi}(a_l) \delta_{\sigma}(a_r) W_{kN}(\pi,\sigma)$$
$$= N^{-k} \sum_{\pi,\sigma} \delta_{\pi}(a_l) \delta_{\sigma}(a_r) \sum_{g=0}^{\infty} K_g(\pi,\sigma) N^{-d(\pi,\sigma)-g}$$

Let us denote by  $H_d(\pi, \sigma)$  the number of signed paths between  $\pi$  and  $\sigma$ , of total length d. In terms of the new variable  $d = d(\pi, \sigma) + g$ , the above expression becomes:

$$I(a) = N^{-k} \sum_{\pi,\sigma} \delta_{\pi}(a_l) \delta_{\sigma}(a_r) \sum_{d=0}^{\infty} H_d(\pi,\sigma) N^{-d}$$
$$= N^{-k} \sum_{d=0}^{\infty} \left( \sum_{\pi,\sigma} \delta_{\pi}(a_l) \delta_{\sigma}(a_r) H_d(\pi,\sigma) \right) N^{-d}$$

We recognize in the middle the quantity  $H_d(a)$ , and this gives the result.

We derive now some concrete consequences from the abstract results in the previous section. First, let us recall the following result, due to Collins and Sniady [20]:

THEOREM 16.32. We have the estimate

$$W_{kN}(\pi,\sigma) = N^{-k-d(\pi,\sigma)}(\mu(\pi,\sigma) + O(N^{-1}))$$

where  $\mu$  is the Möbius function.

**PROOF.** We know from the above that we have the following estimate:

$$W_{kN}(\pi,\sigma) = N^{-k-d(\pi,\sigma)}(K_0(\pi,\sigma) + O(N^{-1}))$$

Now since one of the possible definitions of the Möbius function is that this counts the signed geodesic paths, we have  $K_0 = \mu$ , and we are done.

Let us go back now to our integrals I(a). We have the following result:

THEOREM 16.33. We have the estimate

$$I(a) = N^{-k-e(a)}(\mu(a) + O(N^{-1}))$$

where the objects on the right are as follows:

- (1)  $e(a) = \min\{d(\pi, \sigma) | \pi, \sigma \in D_k, \delta_\pi(a_l) = \delta_\sigma(a_r) = 1\}.$
- (2)  $\mu(a)$  counts all a-admissible signed paths in  $D_k$ , of total length e(a).

**PROOF.** We know that we have an estimate of the following type:

$$I(a) = N^{-k-e}(H_e(a) + O(N^{-1}))$$

Here, according to the various notations above,  $e \in \mathbb{N}$  is the smallest total length of an *a*-admissible path, and  $H_e(a)$  counts all signed *a*-admissible paths of total length *e*. Now since the smallest total length of such a path is of course attained when the path is just a segment, we have e = e(a) and  $H_e(a) = \mu(a)$ , and we are done.

Summarizing, we have now a full description of the asymptotic behavior of I(a).

A remarkable advance on this subject, using more advanced technology, comes from the paper of Collins and Matsumoto [16]. In what follows we will briefly explain their formula, and work out the corresponding consequences regarding the integrals I(a).

The formula of Collins and Matsumoto [16], also established by Zinn-Justin in [100], is something quite technical, as follows:

THEOREM 16.34. We have the formula

$$W_{kn}(\pi,\sigma) = \frac{\sum_{\lambda \vdash k, \, l(\lambda) \le k} \chi^{2\lambda}(1_k) w^{\lambda}(\pi^{-1}\sigma)}{(2k)!! \prod_{(i,j) \in \lambda} (n+2j-i-1)}$$

where the various objects on the right are as follows:

- (1) The sum is over all partitions of  $\{1, \ldots, 2k\}$  of length  $l(\lambda) \leq k$ .
- (2)  $w^{\lambda}$  is the corresponding zonal spherical function of  $(S_{2k}, H_k)$ .
- (3)  $\chi^{2\lambda}$  is the character of  $S_{2k}$  associated to  $2\lambda = (2\lambda_1, 2\lambda_2, \ldots)$ .
- (4) The product is over all squares of the Young diagram of  $\lambda$ .

**PROOF.** This is something quite technical, and we refer to [16], [100].

It is of course possible to deduce from this a new a formula for the integrals I(a), just by putting together the various formulae that we have. However, there are probably several non-trivial simplifications that might appear when doing the sum over  $\pi, \sigma$ , and it is actually not very clear how the final statement about I(a) should look like.

Instead, let us just record the following consequence:

# 16D. STANDARD ESTIMATES

THEOREM 16.35. The possible poles of I(a) can be at the numbers

$$-(k-1), -(k-2), \ldots, 2k-1, 2k$$

where the number  $k \in \mathbb{N}$  associated to the admissible matrix  $a \in M_{p \times q}(\mathbb{N})$  is given by:

$$k = \sum a_{ij}/2$$

**PROOF.** We know from the above that the possible poles of I(a) can only come from those of the Weingarten function. On the other hand, Theorem 16.34 tells us that these latter poles are located at the numbers of the form -2j + i + 1, with (i, j) ranging over all possible squares of all possible Young diagrams, and this gives the result.

As explained in [16], the main feature of the zonal function approach is its remarkable efficiency in respect with numeric computations. Normally the above results can be of help in producing more "data" for the exact computation of I(a).

As a last topic of this chapter and book, let us discuss Gram determinants.

In what regards the symmetric group  $S_N$ , we have the following result, that we already know, from the above:

THEOREM 16.36. The determinant of the Gram matrix of  $S_N$  is given by

$$\det(G_{kN}) = \prod_{\pi \in P(k)} \frac{N!}{(N - |\pi|)!}$$

with the convention that in the case N < k we obtain 0.

**PROOF.** This is something that we know, the idea being that  $G_{kN}$  naturally decomposes as a product of an upper triangular and lower triangular matrix.

Let us discuss now the case of the orthogonal group  $O_N$ . Here the combinatorics is that of the Young diagrams. We denote by |.| the number of boxes, and we use quantity  $f^{\lambda}$ , which gives the number of standard Young tableaux of shape  $\lambda$ .

With these conventions, the result is then as follows:

THEOREM 16.37. The determinant of the Gram matrix of  $O_N$  is given by

$$\det(G_{kN}) = \prod_{|\lambda|=k/2} f_N(\lambda)^{f^{2\lambda}}$$

where the quantities on the right are  $f_N(\lambda) = \prod_{(i,j) \in \lambda} (N+2j-i-1)$ .

**PROOF.** This follows from the results of Zinn-Justin [100]. Indeed, it is known from there that the Gram matrix is diagonalizable, as follows:

$$G_{kN} = \sum_{|\lambda|=k/2} f_N(\lambda) P_{2\lambda}$$

Here  $1 = \sum P_{2\lambda}$  is the standard partition of unity associated to the Young diagrams having k/2 boxes, and the coefficients  $f_N(\lambda)$  are those in the statement. Now since we have  $Tr(P_{2\lambda}) = f^{2\lambda}$ , this gives the result. See [6], [100].

We refer to [6], [36] for a further discussion on these topics.

# 16e. Exercises

Congratulations for having read this book, and there are no exercises for this final chapter. But you can try instead to read the various articles referenced below.

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396

## Index

abelian group, 203 absolute value, 102 adjoint matrix, 68 adjoint operator, 166 affine map, 11, 13, 14, 20, 22 algebra of characters, 317 algebraic basis, 157 all-one matrix, 23, 29, 177 all-one vector, 23, 29, 177, 189 alternating group, 244 area of sphere, 138 argument of complex number, 59 associativity, 203 asymptotic character, 380 Banach algebra, 161 barycenter, 66 basis, 24, 82, 83 Bell numbers, 258 Bessel function, 275 Bessel law, 275, 278 bicommutant theorem, 326 bijective linear map, 24, 82 binomial coefficients, 133 binomial formula, 112 bistochastic group, 191, 240, 350 bistochastic Hadamard matrix, 192 bistochastic matrix, 188, 192, 238 bounded operator, 159, 160 Brauer space, 387 Brauer theorem, 341, 342, 350, 352, 353 Catalan numbers, 114, 289, 291 category of partitions, 340, 346

category of partitions, 340, 346 Cauchy-Schwarz inequality, 154 Cayley embedding, 212, 215

central binomial coefficients, 114, 133 central function, 250, 317 Central Limit Theorem, 127 change of basis, 24, 25 change of variable, 119 character, 218, 249, 299, 301 characteristic polynomial, 43, 71, 85, 86, 93 CHC, 198 checkered signs, 49 Circulant Hadamard conjecture, 198 circulant matrix, 183, 186 closed subgroup, 214 CLT, 127 colored integer, 302 colored powers, 303 column expansion, 46, 74 column-stochastic matrix, 188 common roots, 87 complete algebra, 159 complex Bessel laws, 279 complex CLT, 147 complex Gaussian law, 146 complex normal law, 146 complex numbers, 57, 58 complex reflection group, 216, 245, 279, 353 complex roots, 64composition of linear maps, 19, 20, 22 compound Poisson law, 277 compound Poisson Limit theorem, 278 conjugacy classes, 250 conjugate representation, 300 continuous functional calculus, 172 convolution, 125, 146, 255, 256, 275 CPLT, 278 crossed product, 209, 211

398

## INDEX

crossed product decomposition, 209 crossings, 51 cyclic group, 206, 219, 251 cyclic root, 186 degree 2 equation, 64density, 90 derangement, 252 derivative, 108 determinant, 40, 73 determinant formula, 53, 73 determinant of products, 43 diagonal form, 25 diagonal matrix, 23 diagonalizable matrix, 24, 70, 83 diagonalization, 25, 86 dihedral group, 207, 209, 251 dimension, 83 dimension inequality, 84 discrete Fourier transform, 183, 184, 191, 240 discriminant, 89 distance, 155 distance preservation, 30 double cover map, 236 double factorial, 131 double factorials, 122, 130, 135, 141 double root, 89 dual group, 218, 220 easiness level, 361 easy envelope, 360 easy generation, 357 easy group, 346, 349, 350 eigenspaces, 81, 84, 86 eigenvalue, 24, 43, 70 eigenvector, 24, 70 eigenvector basis, 24 End space, 303 equivalent Hadamard matrices, 185 Euler-Rodrigues formula, 236  $\exp$  and  $\log$ , 115fattening partitions, 292 finite abelian group, 220, 221 finite dimensional algebra, 305 finite group, 215 Fix space, 303 fixed points, 251, 252

flat matrix, 23, 29, 79, 177, 179 formal exponential, 256 Fourier matrix, 77, 79, 178-180, 221 Fourier transform, 126, 220, 221, 257, 277 Fourier-diagonal matrix, 183 Frobenius isomorphism, 316 functions of matrices, 91 Gauss formula, 121 Gaussian law, 122 Gelfand theorem, 172 general linear group, 204 generalized Bessel laws, 279 generalized Fourier matrix, 182, 221 GNS theorem, 174 Gram matrix, 315, 370 Gram-Schmidt, 157 group, 203group of characters, 218 groups of matrices, 204 groups of numbers, 204 Haar integration, 312 Haar measure, 312 Hadamard conjecture, 196 Hadamard equivalence, 194 Hadamard matrix, 180, 182, 193 Hassian matrix, 117 HC, 196, 197 Hilbert space, 156 Hom space, 303 homogeneous group, 358 hypercube, 210 hyperoctahedral group, 210, 211, 217, 273, 353 hyperspherical law, 145, 150, 286 identity matrix, 22 independence, 124–126 infinite matrix, 158, 160 injective linear map, 82 inverse of inverse, 205 inversion formula, 34 invertible matrix, 24, 33, 35, 41, 68, 81, 82 isometry, 30, 68

Jacobian, 119, 129

Kronecker symbols, 337, 345

## INDEX

law. 124 left invertibility, 81 length of vector, 28linear dependence, 82 linear equations, 107 linear independence, 82, 83, 376 linear map, 11, 13, 14, 20, 22, 67 linear operator, 158, 160 linear recurrence, 108 linear space, 81 local maximum, 110, 111, 117 local minimum, 110, 111, 117 lower triangular matrix, 44 Möbius function, 372 main character, 249, 251, 253, 274, 314, 374 maps associated to partitions, 337, 345 Marchenko-Pastur law, 293 matching pairings, 147 matrices with distinct eigenvalues, 90 matrix, 13 matrix inversion, 48, 49, 76, 107 matrix multiplication, 13, 19, 21 matrix powers, 108metric space, 155 Minkovski inequality, 154 modified Bessel laws, 281 modulus, 102modulus of complex number, 59 moment, 124 moments, 122, 123, 147, 314, 374 Monte Carlo integration, 118 multilinear form, 47, 75 multiplication, 203 multiplication of complex numbers, 63 noncrossing pairings, 289, 292 noncrossing partitions, 292 norm of operator, 159 norm of vector, 153

normal law, 122, 126, 145, 380 normal matrix, 100 normal operator, 168, 172 null matrix, 22 number of inversions, 51 number of the beast, 196

odd cycles, 51

operator algebra, 167, 171, 174, 305 oriented group, 246 oriented system of vectors, 38 orthogonal group, 31, 204, 225, 350 orthogonal matrix, 30, 99 orthogonal polynomials, 157 orthogonal projection, 23, 28, 68 orthonormal basis, 157

pairings, 123 Paley matrices, 196, 197 parallelogram identity, 30 partial derivatives, 116 partial isometry, 103 passage matrix, 25 Pauli matrices, 233 permutation, 50 permutation group, 210, 215 permutation matrix, 214 Peter-Weyl, 306, 308, 317 Peter-Weyl representations, 302, 321 PLT, 257 Poisson law, 255, 257, 262 Poisson limit theorem, 257 polar coordinates, 59, 63, 119, 121 polar decomposition, 102, 103 polar writing, 61 polarization identity, 69, 155 polynomial integrals, 264, 315, 369 positive matrix, 96 positive operator, 173 powers of complex number, 60probability space, 124 product of cyclic groups, 219 product of eigenvalues, 43, 92 product of matrices, 21 product of representations, 300 products of matrices, 91 projection, 12, 15, 18, 25, 68, 95 projections, 29

random permutation, 252 random variable, 124 rank 1 projection, 28, 68 rational calculus, 163 rational function, 163 rectangular matrix, 21, 67 regular polygon, 207 399

INDEX

representation, 218, 249, 299 resultant, 87, 89 Riemann integration, 117 Riemann sum, 117, 133 right invertibility, 81 roots of polynomial, 43 roots of polynomials, 64 roots of unity, 65, 66, 206 rotation, 11, 15, 17, 26, 31, 71, 236 row expansion, 46, 74 row-stochastic matrix, 188 Sarrus formula, 47, 53 scalar product, 23, 27, 68, 153, 156 self-adjoint matrix, 94 self-adjoint operator, 168 self-dual group, 219 semicircle law, 291 separable Hilbert space, 157 Shephard-Todd, 245 shift, 161 shrinking partitions, 292 sign of system of vectors, 38 signature, 39, 51, 210, 244 signed volume, 40 simple roots, 87 single roots, 89 size of Hadamard matrix, 195 smooth representation, 307 space of coefficients, 307 space-time sphere, 232 special linear group, 204 special orthogonal group, 204, 226 special unitary group, 204, 226 spectral radius, 168 spectral theorem, 94, 95, 98–100 spectrum, 161, 165 spherical coordinates, 120, 129 spherical integral, 141, 143, 149 spinned representation, 300 square root, 114, 173 square-summable, 156 Stirling formula, 133 strictly positive matrix, 97 sum of eigenvalues, 92 sum of representations, 300 sums of roots, 178

super-identity, 241 super-orthogonal group, 242 super-space, 241 surjective linear map, 82 symmetric functions, 93 symmetric group, 210, 251, 262, 352 symmetric matrix, 95 symmetry, 11, 14, 17, 26, 31 symplectic group, 242

Tannakian category, 321 Tannakian duality, 334 Taylor formula, 110–112, 115 Taylor series, 62 tensor category, 304, 322 topological generation, 246 translation, 12 transpose matrix, 27, 54 transpositions, 51 trigonometric functions, 115 trigonometric integral, 130, 139 truncated character, 261

uniform group, 247, 363, 381 unitary group, 204, 225, 350 unitary matrix, 68, 98 unitary operator, 168 unoriented system of vectors, 38 upper triangular matrix, 44

Vandermonde formula, 55 vector space, 81 volume of parallelepiped, 35 volume of sphere, 131, 132, 137

Walsh matrix, 182, 192, 222 Weingarten formula, 315, 369, 370, 384 Weingarten matrix, 315, 370 Wigner law, 291 wreath product, 211, 217

400