

Reproducibility & replicability in MRI Gaëlle Leroux

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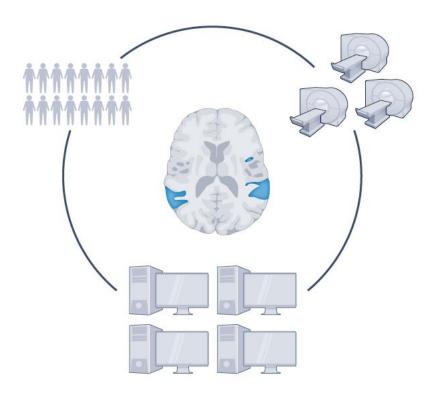
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Reproducibility & replicability in MRI







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ïD

Illustration from Vogt. Reproducibility in MRI. Nat Methods (2023)





Research data College

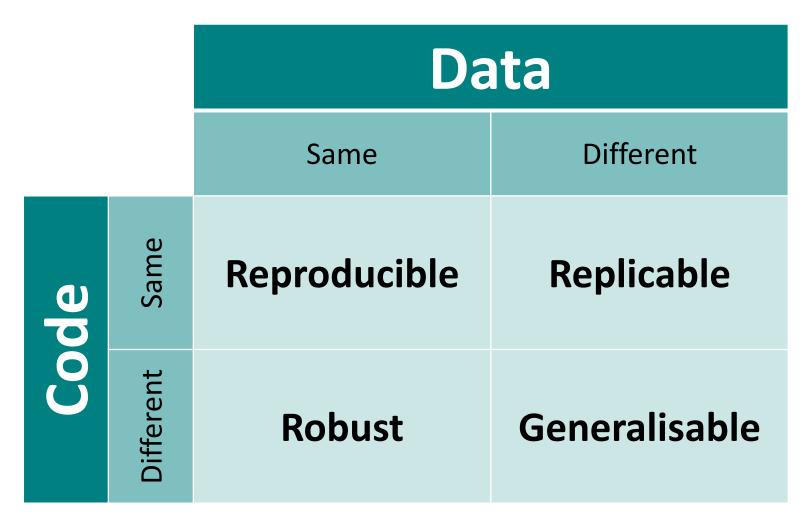
Glossary

Generalisable Replicable Data Reproducible Robust Different Same Same 2 Code Different 3 4

From Kirstie Whitaker 2017, doi: https://dx.doi.org/10.6084/m9.figshare.4244996



Glossary

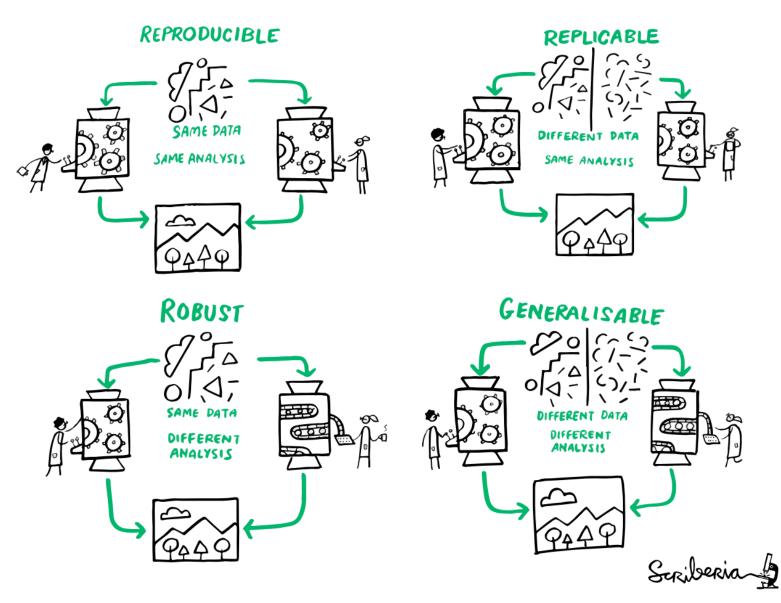


From Kirstie Whitaker 2017, doi: <u>https://dx.doi.org/10.6084/m9.figshare.4244996</u>

ED NSCo – 2023 MR courses – Reproducibility & replicability in MRI – Gaëlle Leroux – 2023/06



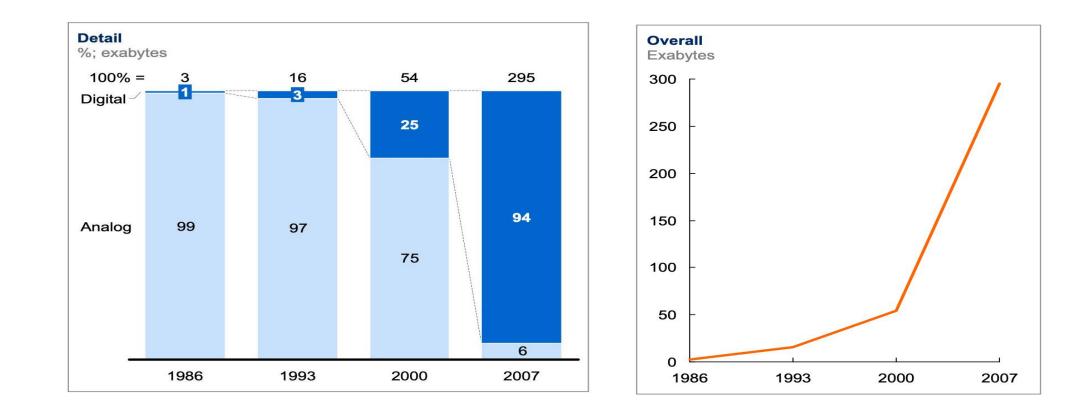
Glossary



The Turing Way project illustration by Scriberia. DOI: 10.5281/zenodo.3332807



1. An overview of the "reproducibility crisis": origins

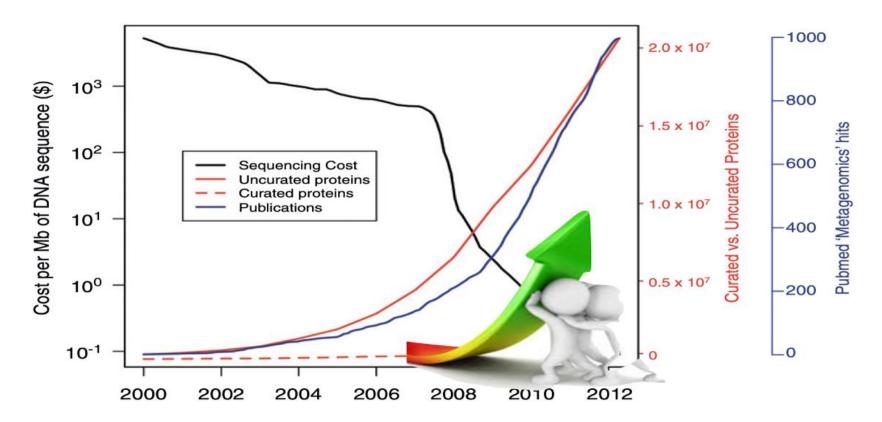




 (∞)

1. An overview of the "reproducibility crisis": origins

Human genome: in 1990 = 13 years & \$3 billion in 2015 = a few hours & \$1,000



→ The amount of data to be stored & to be analysed, explodes
 → Analysis performance dramatically falls

ED NSCo – 2023 MR courses – Reproducibility & replicability in MRI – Gaëlle Leroux – 2023/06



Source

1. An overview of the "reproducibility crisis": research tasks have moved

1. Design the experiment



- 2. Collect data
- 3. Analyze results

A change of paradigm

- 1. Massive data production / reuse
- 2. Organize (store, document, annotate)
- 3. Analyze (extract information)
- 4. Share information





1. An overview of the "reproducibility crisis": alarm in 2005

Open access, freely available online

Essay

Why Most Published Research Findings Are False

John P. A. Ioannidi

Summary

false. The probability that a research claim is true may depend on study power and same question, and, importantly, the ratio of true to no relationships among the greater flexibility in designs, definitions, designs and settings, it is more likely for often be simply accurate measures of the implications of these problems for the

factors that influence this problem and some corollaries thereof.

Modeling the Framework for False Positive Findings

Several methodologists have pointed out [9–11] that the high rate of nonreplication (lack of confirmation) of research discoveries is a consequence of the convenient, yet ill-founded strategy of claiming conclusive research findings solely on the basis of a single study assessed by formal statistical significance, typically for a *p*-value less than 0.05. Research is not most appropriately represented and summarized by *p*-values, but, unfortunately, there is a widespread notion that medical research articles

It can be proven that most claimed research findings are false.

should be interpreted based only on *p*-values. Research findings are defined here as any relationship reaching formal statistical significance, e.g., effective interventions, informative predictors, risk factors, or associations. "Negative" research is also very useful. is characteristic of the field and can vary a lot depending on whether the field targets highly likely relationships or searches for only one or a few true relationships among thousands and millions of hypotheses that may be postulated. Let us also consider, for computational simplicity, circumscribed fields where either there is only one true relationship (among many that can be hypothesized) or the power is similar to find any of the several existing true relationships. The pre-study probability of a relationship being true is R/(R+1). The probability of a study finding a true relationship reflects the power $1 - \beta$ (one minus the Type II error rate). The probability of claiming a relationship when none truly exists reflects the Type I error rate, α. Assuming that *c* relationships are being probed in the field, the expected values of the 2 × 2 table are given in Table 1. After a research finding has been claimed based on achieving formal statistical significance, the post-study probability that it is true is the positive predictive value, PPV. The PPV is also the complementary probability of what Wacholder et al. have called the false positive report probability [10]. According to the 2

74,946 7,388 Save Citation 3,001,940 9,576 View Share

Inadequate methodologies

A result is all the more likely to be false when:

- effect sizes are small
- there is a greater number and lesser preselection of tested relationships
- there is greater flexibility in designs, definitions, outcomes, and analytical modes
- there is greater financial and other interest and prejudice

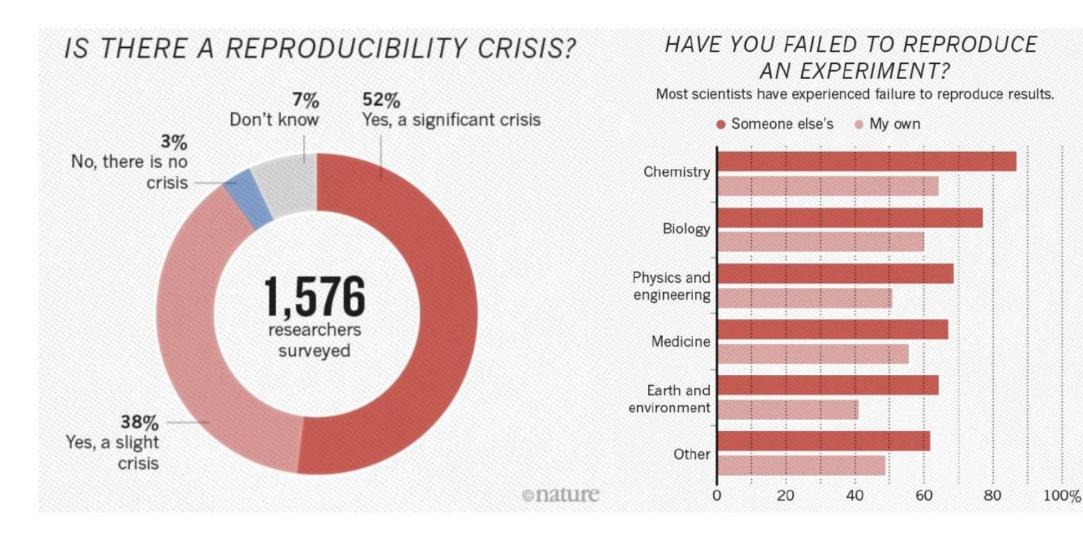
+ the propensity of journals to publish only positive results

+ the "**publish or perish**" culture for researchers

Ioannidis (2005) Why most published research findings are false. *PLoS Med* 2(8): e124 DOI: 10.1371/journal.pmed.0020124



1. An overview of the "reproducibility crisis": awareness in **2016**

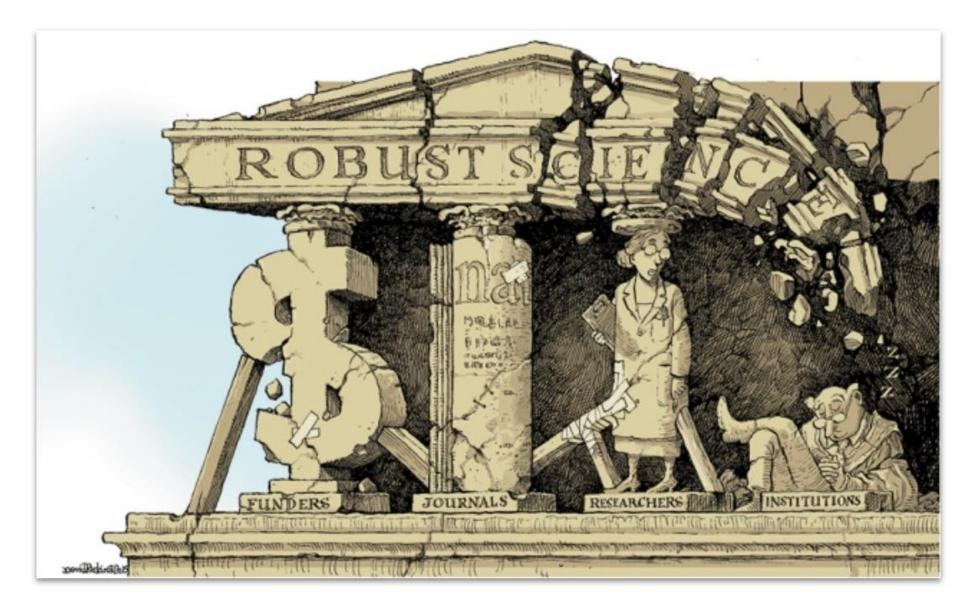


"1,500 scientists lift the lid on reproducibility". Nature. 533: 452-454 - 2016

4



1. An overview of the "reproducibility crisis": awareness in 2016

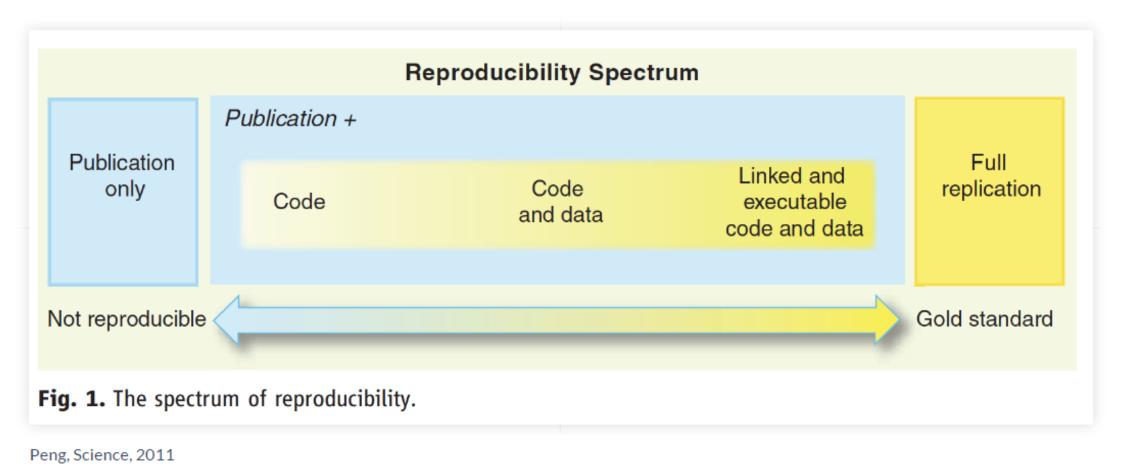




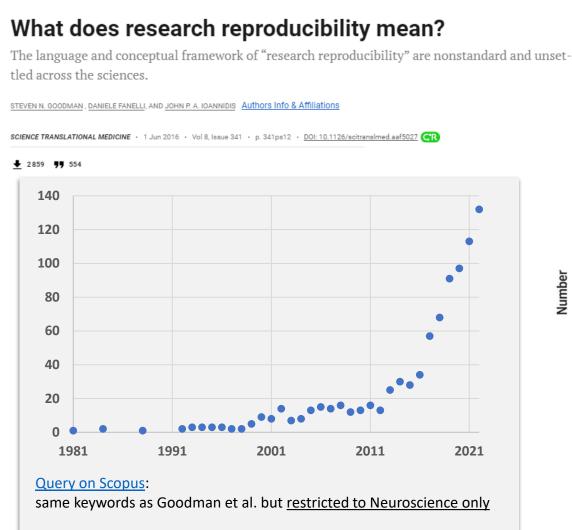


1. An overview of the "reproducibility crisis": not a binary phenomenon

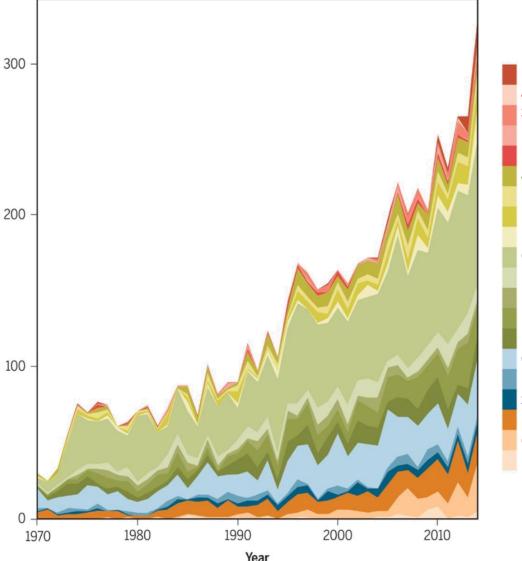
Reproducibility is not black an white, but rather a continuum







Conclusion: "any scientific field, require a clear specification of the kind of reproducibility being discussed (methods, results, or inferences)"

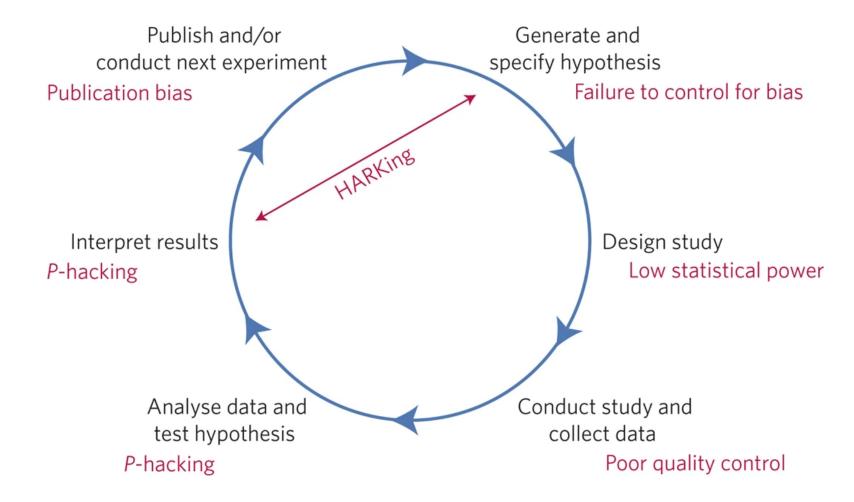


Multidisciplinary Arts & Humanities Social Sciences Psychology & Psychiatry Economics & Business Agricultural sciences **Environment & Ecology** Plant & Animal sciences Neurobiology & Behavior **Clinical Medicine** Pharmacology & Toxicology Microbiology Molecular Biology **Biology & Biochemistry** Chemistry Physics Space science Engineering **Computer Sciences Mathematics**

Keywords (at least 1 of the following expressions): research reproducibility, reproducibility of research, reproducibility of results, results reproducibility, reproducibility of study, study reproducibility, reproducible research, reproducible finding, or reproducible result

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Legend

Blue: An idealized version of the hypothetico-deductive model of the scientific method Pink: threats to this model

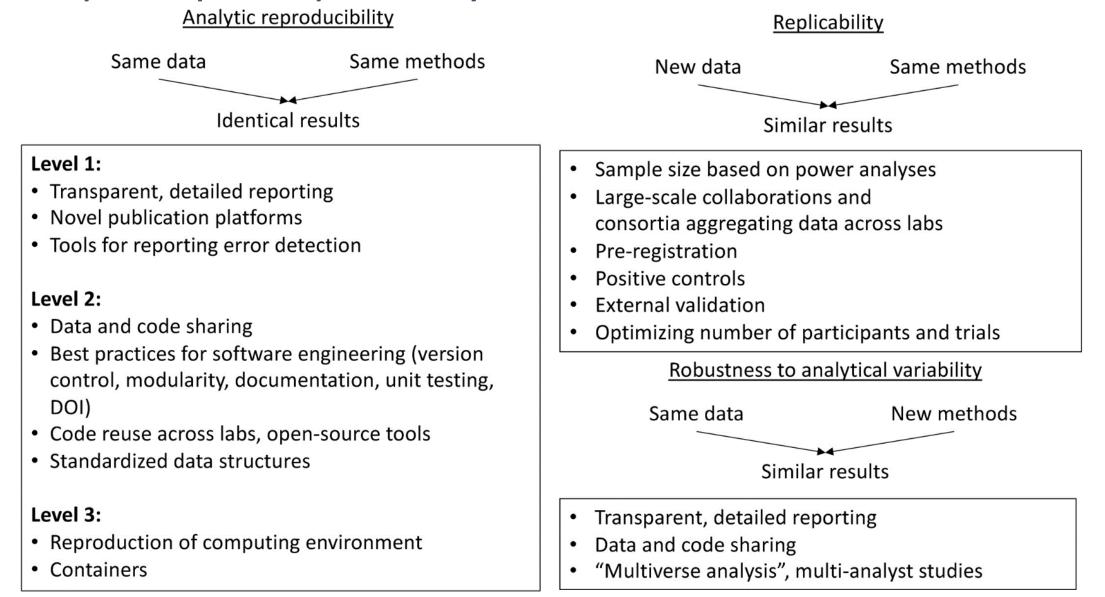
Munafò et al. A manifesto for reproducible science. Nat Hum Behav 1, 0021 (2017). https://doi-org.insb.bib.cnrs.fr/10.1038/s41562-016-0021



Theme	Proposal	Examples of initiatives/potential solutions (extent of current adoption)	Stakeholder(s)
Methods	Protecting against cognitive biases	All of the initiatives listed below (* to ****) Blinding (**)	J, F
	Improving methodological training	Rigorous training in statistics and research methods for future researchers (*) Rigorous continuing education in statistics and methods for researchers (*)	I, F
	Independent methodological support	Involvement of methodologists in research (**) Independent oversight (*)	F
	Collaboration and team science	Multi-site studies/distributed data collection (*) Team-science consortia (*)	I, F
Reporting and dissemination	Promoting study pre-registration	Registered Reports (*) Open Science Framework (*)	J, F
	Improving the quality of reporting	Use of reporting checklists (**) Protocol checklists (*)	J
	Protecting against conflicts of interest	Disclosure of conflicts of interest (***) Exclusion/containment of financial and non-financial conflicts of interest (*)	J
Reproducibility	Encouraging transparency and open science	Open data, materials, software and so on (* to **) Pre-registration (**** for clinical trials, * for other studies)	J, F, R
Evaluation	Diversifying peer review	Preprints (* in biomedical/behavioural sciences, **** in physical sciences) Pre- and post-publication peer review, for example, Publons, PubMed Commons (*)	ſ
Incentives	Rewarding open and reproducible practices	Badges (*) Registered Reports (*) Transparency and Openness Promotion guidelines (*) Funding replication studies (*) Open science practices in hiring and promotion (*)	J, I, F

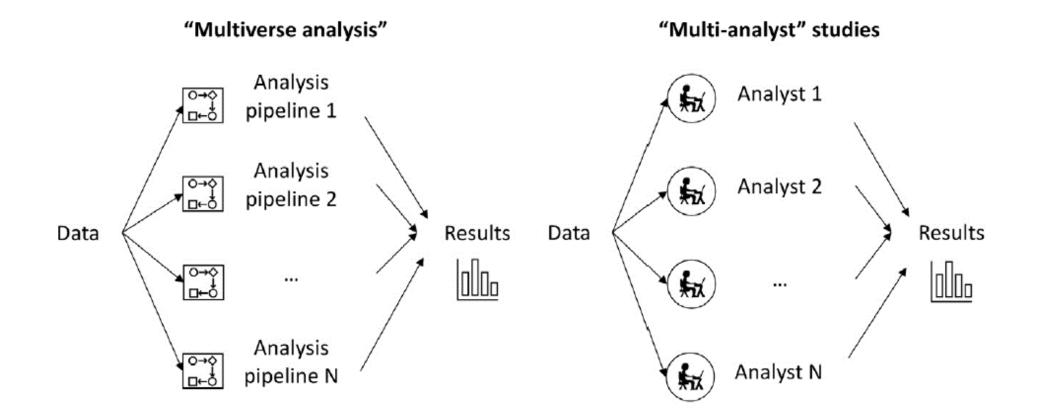
 Table1 | A manifesto for reproducible science.

Estimated extent of current adoption: *, <5%; ***, 5-30%; ****, 30-60%; ****, >60%. Abbreviations for key stakeholders: J, journals/publishers; F, funders; I, institutions; R, regulators.



Botvinik-Nezer & Wager. Reproducibility in Neuroimaging Analysis: Challenges and Solutions. *Biological Psychiatry: Cognitive Neuroscience and Neuroimaging* (2022) https://doi.org/10.1016/j.bpsc.2022.12.006





Botvinik-Nezer & Wager. Reproducibility in Neuroimaging Analysis: Challenges and Solutions. *Biological Psychiatry: Cognitive Neuroscience and Neuroimaging* (2022) <u>https://doi.org/10.1016/j.bpsc.2022.12.006</u>

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Organizational principles:

- 1. Put each project in its own project directory
- 2. Keep together what needs to be kept together
- 3. Keep the source data
- 4. Make file names human-readable
- 5. Make file names **machine-readable**
- 6. Make file names easy to sort
- 7. Back up your data



XNAT @ CRNL (see wiki for details)



Bienvenue sur le XNAT du CRNL dédié au stockage des données source de tous les protocoles du CRNL !

TOUS les types de données peuvent être déposés.

Les pré-traitements, les données générées et les analyses statistiques doivent être réalisés dans les espaces disques des équipes, comme d'habitude.

Une question sur l'identification ? > merci d'ouvrir un ticket info (https://extranet.crnl.fr/ticket)

Une question sur les projets ? > merci d'envoyer un email à gaelle.leroux@cnrs.fr

Welcome to the CRNL XNAT dedicated to the storage of the source data of all CRNL protocols!

ALL data types can be uploaded.

The pre-processing, derived data and statistical analyses are performed on the team disk space, as usual.

Any question about xnat login? > please, open a ticket to the IT service at the CRNL (https://extranet.crnl.fr/ticket).

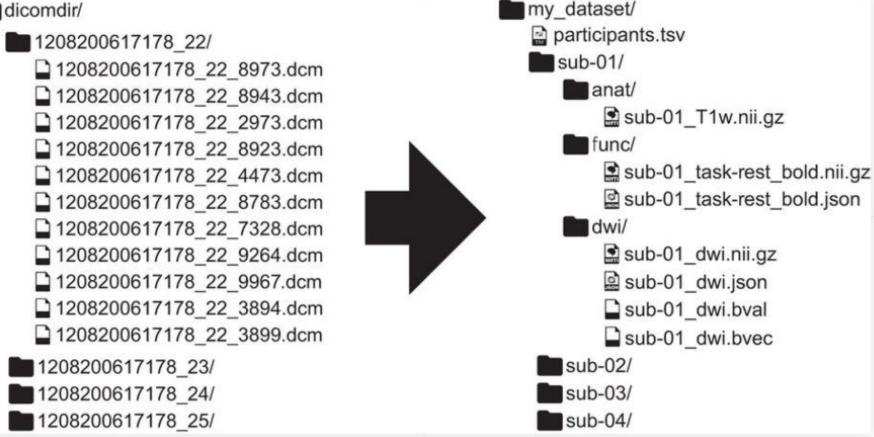
Any question about the projects? > please, send an email to gaelle.leroux@cnrs.fr

Idap1 •	
USER	
PASSWORD	
Forgot login or password?	





dicomdir/



Source & further reading: Gorgolewski et al. Sci Data, 2016, <u>http://bids.neuroimaging.io</u>

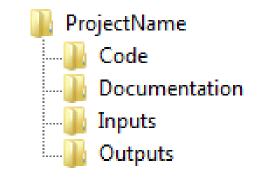


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- 7. Back up your data

Documentation principles:

- 8. Create an **overview** of your project
- 9. Make software **dependencies** explicit
- 10. Describe variables in a **codebook**
- 11. **Document** your code and its use
- 12. Give **meaningful names** to functions and variables
- 13. Keep track of how each result was produced





 Pour le code : nom explicite, indication du workflow, (version)

> Usual prog1.do prog2.do test.R final.R

Better

01 preparing data.do 02 stat desc.do 03 model1.R 03 model2.R



Source









GitLab @ CRNL (see wiki for details)

or https://gitlab.in2p3.fr/





Variable Name	Description	Туре	Values or Characteristics
patientid	Patient ID	Numeric	Integers
sex	Patient Gender	Numeric	1=Female, 2=Male
proceduredate	Date of Procedure	Date, M/D/YYYY	
dob	Date of Birth	Date, M/D/YYYY	
patientrace	Patient Race	Numeric	O=Not Latino, 1=Latino,
tx	Treatment Group	Numeric	1-Tx 1, 2=Tx 2, 3=Tx 3
dischargeppt	PPt at Discharge	Numeric	2 decimal places
notes	Notes	Character	
location	Location	Numeric	1=local, 2=regional, 3=distant
patientdied	Patient is Dead	Numeric	O=No, 1=Yes
deathdate	Date of Death	Date, M/D/YYYY	
clinicaloutcome	Clinical Outcome	Character	
outcomecomments	Outcome Comments	Character	

Source: http://domstat.med.ucla.edu/pages/codebook2

Further reading: Codebook cookbook - A guide to writing a good codebook for data analysis projects in medicine, Creating a codebook - Document, Discover, and Interoperate alliance

Tools: codebook generator for R: memisc





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Principles of automation :

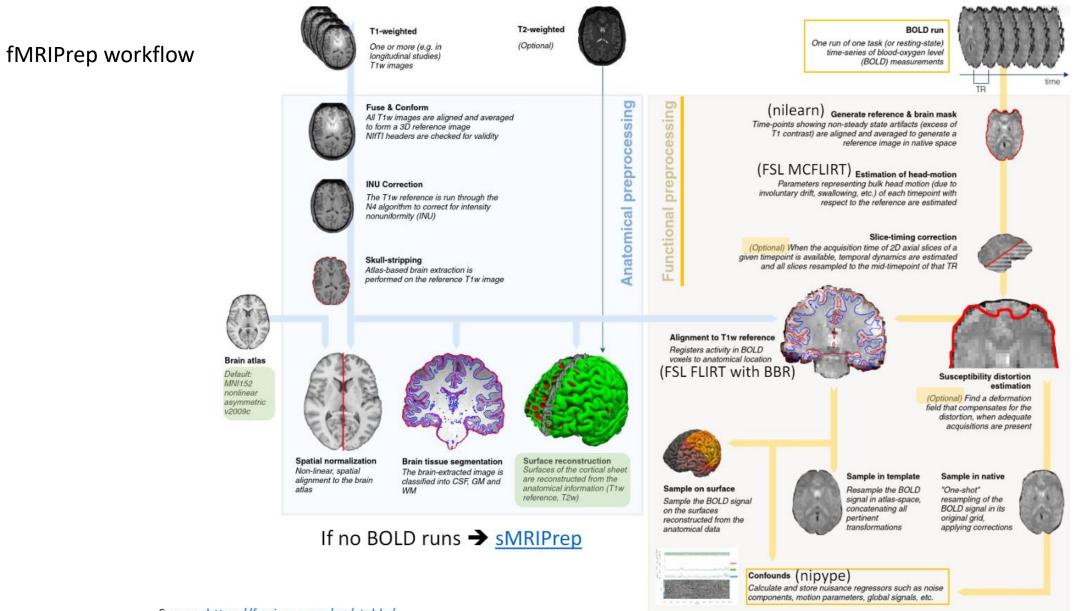
- 14. Avoid manual data manipulation steps
- 15. Modularize code rather than copy and paste
- 16. Reuse code rather than rewrite it
- 17. Don't comment or uncomment code sections to control program behavior

18. Add assertions to programs to check their operation

- 19. Save the random number generator seed
- 20. Record all intermediate results

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Source: <u>https://fmriprep.org/en/stable/</u>



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- 19. Save the random number generator seed
- 20. Record all intermediate results

Distribution principles:

- 21. Store data in **open formats**
- 22. Process data with **open software**
- 23. Eliminate hard-coded paths in your code
- 24. Link text statements to underlying results
- 25. Submit code and data to a **DOI-issuing repository** and reference it in your article
- 26. Make the **license** of code and data explicit



File format	Recommendation	Avoid
Document texte	PDF, TXT, ODT	MS Word, RTF
Feuille de calcul	ODS, CSV	MS Excel, PDF, OOXML
Base de données	SQL, SIARD, DB tables (.CSV)	MS Access, dBase (.dbf), HDF5
Données statistiques	SPSS Portable, STATA, XML, CSV, TXT	SAS et R
Images	JPEG, TIFF, PNG	DICOM
Audio	BWF, MXF, Matroska (.mka), FLAC, OPUS	<u>WAVE</u> , <u>MP3</u> , <u>AAC</u> , <u>AIFF</u> , <u>OGG</u>
Video	MXF, MKV	MPEG-4, MPEG-2, AVI, QuickTime (.mov, .qt)
Information géographique	GML, MIF/MID	ESRI Shapefiles, MapInfo, KML
lmages géoréférencées Raster	GeoTIFF (.tif, .tiff) ASCII GRID (.asc, .txt)	TIFF World File ESRI GRID

<u>https://facile.cines.fr/</u> service de validation des formats

In practice, you can often work with a popular closed format and convert it to an open format. But you need to check whether the conversion alters the information, and take compensatory measures if necessary. Example: XLSX -> CSV conversion loses formatting.



(cc)

eCOBIDAS checklist

experimental design reporting, 2. image acquisition reporting, 3. preprocessing reporting,
 statistical modeling, 5. results reporting, 6. data sharing, 7. reproducibility

For each of 7 areas of a study, a tabular listing of over 100 items to help plan, execute, report, and share research in the most transparent fashion is provided.

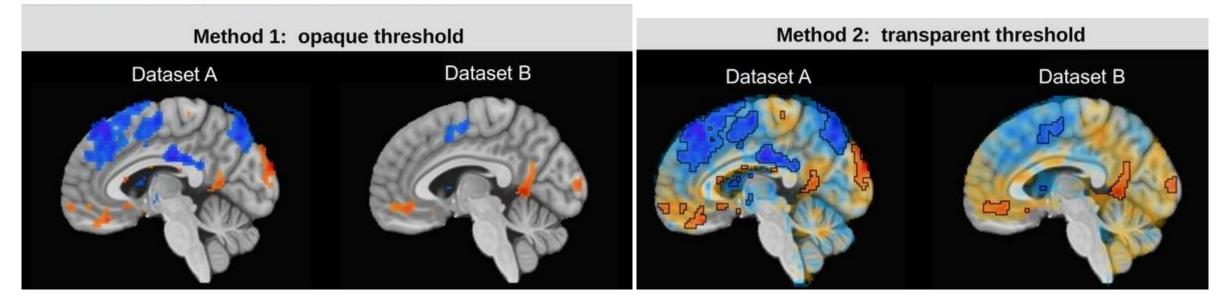
The short term goal of this project is to make the COBIDAS report (and other best practices for methods reporting) easier to use: we want to create a website with a clickable checklist that, at the end, automatically generates most of the method section of a (f)MRI / (i)EEG / MEG / PET paper.

https://remi-gau.github.io/eCobidas/



- Most studies do not present all results of their analysis, hiding subthreshold ones.
- Hiding results negatively affects the interpretation and understanding of the study.
- Neuroimagers should present all results of their study, highlighting key Q: Which method for viewing FMRI data makes it easier to

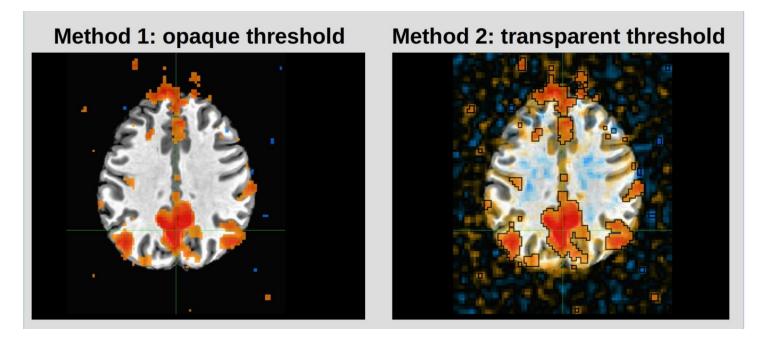
understand the results, and to interpret how they might be similar and/or different? How would your description of the "reproducibility" of the results change when moving from the 1st to the 2nd?



Thresholds hide interesting effects!

Taylor & al. Highlight results, don't hide them: Enhance interpretation, reduce biases and improve reproducibility. *NeuroImage* (2023)



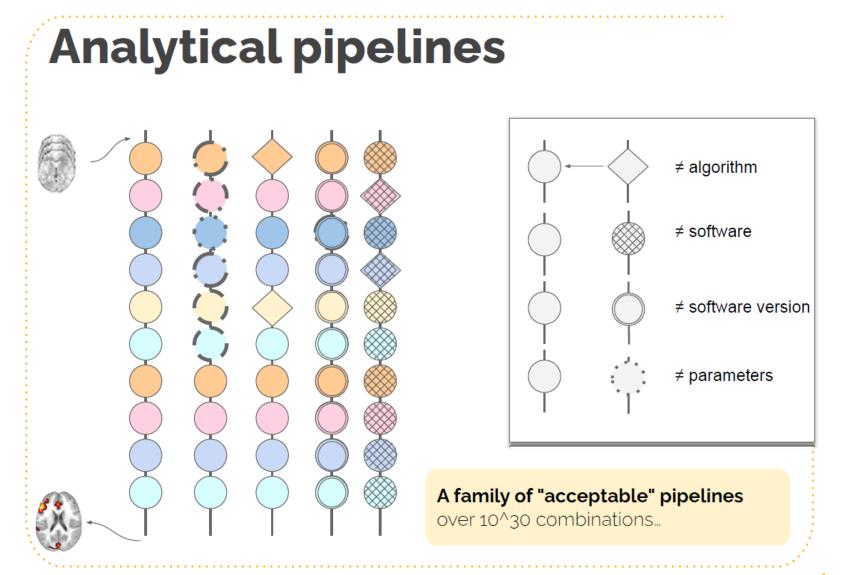


& other times, thresholds hide artifacts and badness.

"We need to see each. Showing more results helps with quality control (QC) evaluation. Become more confident in (or appropriately worried about!) your data." <u>Paul Taylor</u>

Taylor & al. Highlight results, don't hide them: Enhance interpretation, reduce biases and improve reproducibility. *NeuroImage* (2023)

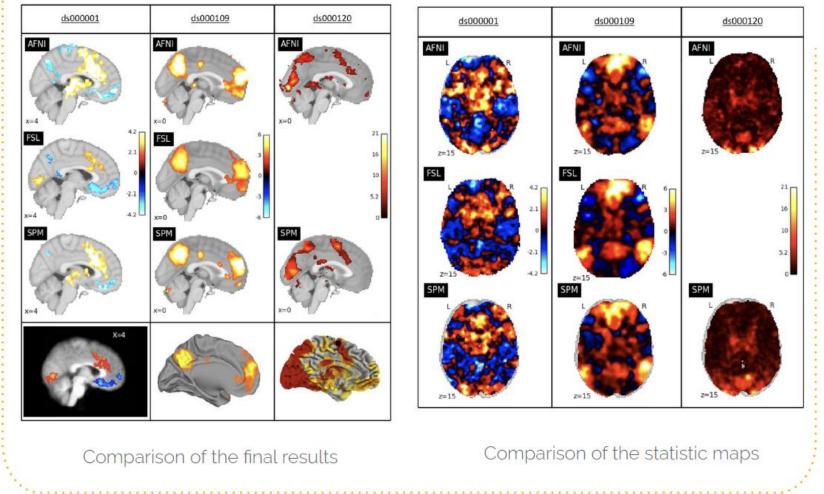




Maumet. Different analysis pipelines, different results... How to deal with analytical variability? 20th IEEE International Symposium on Biomedical Imaging (ISBI 2023), Cartagena (Colombia). ffhal-04073419f



Software Comparison Project



Maumet. Different analysis pipelines, different results... How to deal with analytical variability? 20th IEEE International Symposium on Biomedical Imaging (ISBI 2023), Cartagena (Colombia). ffhal-04073419f

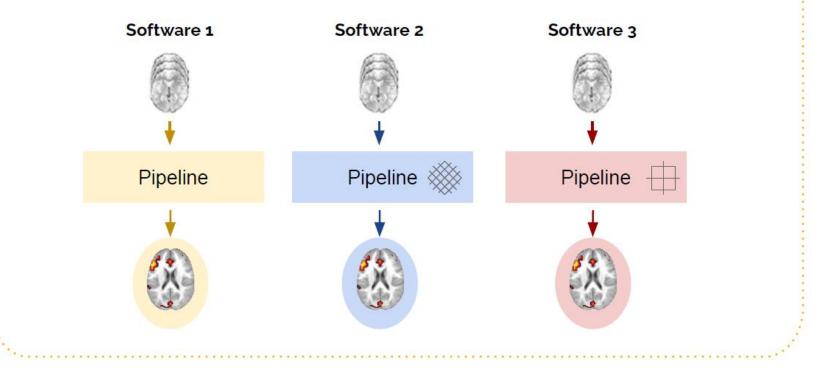


Variability across software

- Reproduced 3 published functional MRI studies
 - o ds001, ds109, ds120 (from Openneuro)
- Using 3 different software packages
 - AFNI, FSL, SPM



Alex Bowring Tom Nichols



Maumet. Different analysis pipelines, different results... How to deal with analytical variability? 20th IEEE International Symposium on Biomedical Imaging (ISBI 2023), Cartagena (Colombia). ffhal-04073419f



2. General tips to help with reproducibility: effect of analytical choices on fMRI findings

The Neuroimaging Analysis Replication and Prediction Study

What happens when 70 independent groups of researchers analyze the same fMRI dataset?





Conclusions of the study:

The strongest factors in explaining the differences between results:

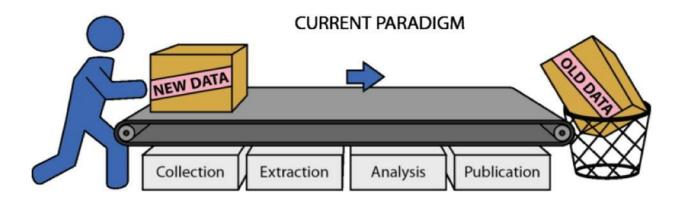
- **1. Spatial smoothness** : higher FWHM lead to a higher likelihood of significant outcomes
- 2. FSL vs other softs: FSL has higher likelihood of significant results across all hypotheses
- 3. Multiple correction method
- 4. No significant effect of standardized vs customized preprocessing pipelines

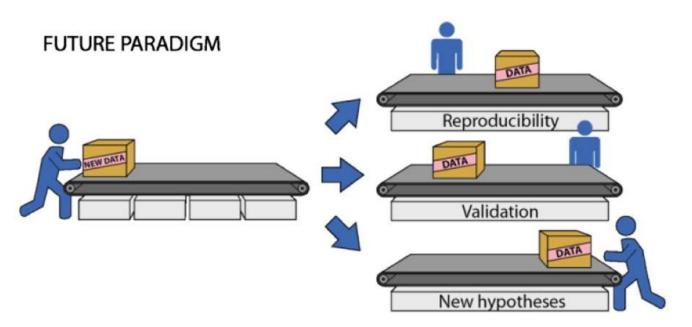
Recommendations to:

- **1. Pre-register** the study, hypotheses, methods
- 2. Share the pipelinee, data and code because the Mat & Methods section is never enough
- 3. Share un- and thresholded maps to allow image-based meta-analyses (e.g. on <u>NeuroVault</u>)
- 4. Use multiple pipelines to analyse the data ("multiverse analysis"), several teams
- 5. Report them and **find the consensus using a meta-analysis**, not the "best" one.
- 6. Use **pipeline-optimization tools** to reduce analysis flexibility (e.g. <u>FitLins</u> & <u>LaConte</u> article)
- 7. Use **sensitivity analyses** to evaluate the effects of pipeline decisions

Botvinik-Nezer *et al.* Variability in the analysis of a single neuroimaging dataset by many teams. *Nature* **582**, 84–88 (2020). <u>https://doi-org.insb.bib.cnrs.fr/10.1038/s41586-020-2314-9</u>









"My hope is that others will be able to reproduce some of our core findings and extend them in interesting new directions." -Pr. James R. Booth

https://lab.vanderbilt.edu/boothlab/datasets/

Lambin et al. Radiother. Oncol. 2013 doi: 10.1016/j.radonc.2013.07.007



Resources

Open science in the doctoral school https://nsco.universite-lyon.fr/science-ouverte/

French network on reproductible science: <u>https://www.recherche-reproductible.fr/</u> UK Reproductibility Network <u>https://www.ukrn.org/</u> <u>OSIRIS</u>: OPEN SCIENCE TO INCREASE REPRODUCIBILITY IN SCIENCE



"Scientific integrity" Club(CRNL): <u>https://project.crnl.fr/IntegriteScientifique/</u> NeuroImaging club meetings: <u>https://osf.io/sxkgq/</u>

Additional tools to the ones cited in the slides:

Electronic labbooks <u>https://datacc.elab.one/login.php</u> & <u>https://cle.inserm.fr/users/sign_in</u> Research Data Plan <u>https://dmp.opidor.fr/</u> Support to data management & sharing <u>https://doranum.fr/</u>

