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1 A preliminary Case Study on University Course 2 Timetabling using Constraint Programming and 3 Operations Research

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10 — Abstract —

11 This paper addresses a real-world problem in university timetable management. We begin by
12 describing the problem in detail, focusing on one particular teaching department, and then we
13 propose a possible solution using constraint programming and operations research techniques. Finally,
14 we discuss the advancements and trends in university timetable management over recent years.

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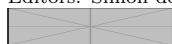
24 **1** Introduction

25 Timetabling problems are one of the most common and well-known classes of constrained
26 optimization problems. These problems concern all companies and organizations, including
27 universities, particularly with the management of their teaching schedules. Although there
28 are probably many scientific works on the subject, the practical use of constrained optimiza-
29 tion techniques for the management of university timetables remains non-existent in many
30 establishments.

31
32 This paper aims to present preliminary work on this topic that we are currently exploring
33 in real-world context. For the next academic year, we are trying to explore the possibilities
34 to improve the management of timetables within our teaching department. To this aim, we
35 need to study what are the specific needs to be taken into account in this department. Based
36 on this knowledge, we will propose a first solution in order to obtain a prototype for the
37 next school year. In the longer term, we will end by asking some questions about the most
38 effective techniques for solving this type of problem, and the different difficulties that may
39 be encountered.

40

¹ corresponding author



5:2 University Timetable Management: a Beginning Case Study

41 This paper is organized as follows. We start with a preliminary Section 2 on one more
42 restricted problematic which is the oral and written defenses of apprenticeship students.
43 Then, Section 3 presents the university timetabling problem and specific needs of our teaching
44 department. Section 4 presents the proposed solution for automatic timetable management.
45 Section 5 offers discussions on the theme of timetable management in order to open up
46 interesting perspectives of this preliminary work.

47 **2 Preliminary: Oral and Written Defense for Apprenticeship Students**

48 In this section, the case study is restricted to the oral and written defenses of apprenticeship
49 students only. This will provide a framework for addressing the issues related to the overall
50 scheduling challenges of the teaching department.

52 **2.1 Data**

53 There are between 30 and 50 students, between 10 and 15 teachers, and between 8 and 12
54 time slots.

56 Each student must be assigned a time slot for their oral defense (not for their written
57 defense which is independent of the time slot). Each student has a supervisor (among the
58 teachers) which is assigned to the oral defense and the written defense of the student. A
59 second teacher must be assigned for each oral defense and for each written defense, and it is
60 forbidden to have the same teach for both defenses.

62 Each time slot has a limited number of oral defenses (approximately 5) and corresponds
63 to a specific day and hour. Each teacher must participate in exactly a certain number of
64 defenses (usually twice the number of students they supervise). Some teachers have time
65 slots during which they are unavailable. Certain specific constraints exist, such as a pair of
66 teachers where at least one must be absent during certain time slots (but either one of the two).

68 The quality of the proposed schedule depends on several factors, including:

- 69 ■ The students involved have already had written and oral defenses in previous years. It is
70 important to minimize assigning a teacher to a student if this assignment has already
71 occurred before.
- 72 ■ It is necessary to group the teachers' participation as much as possible so that they make
73 the fewest possible trips to conduct the defenses (as some teachers also work at other
74 locations).

75 **2.2 A first prototype using Hexaly**

76 A first usable prototype has been implemented using the optimization solver Hexaly (formerly
77 called LocalSolver) [3]. The prototype can be broken down into three parts:

- 78 ■ An Excel file [10] containing all the data.
- 79 ■ A computer program which models and proposes a solution to the timetabling problem.
- 80 ■ An Excel file which contains the proposed timetable.

81 We describe informally below the mathematical model (which is mathematically repres-
82 ented in a format similar to that of a Weighted Constraint Satisfaction Problem) and we use

83 the optimization solver Hexaly (formerly called LocalSolver) [3] to solve a solution within a
 84 one-minut timeout.

85 **Decision variables:**

86 The decision variables are all boolean and are the following ones:

- 87 ■ For each student and for each time slot: Is the oral defense of this student assigned to
 88 this time slot?
- 89 ■ For each student and for each teacher: Is this teacher assigned to be the second teacher
 90 in the oral defense of this student?
- 91 ■ For each student and for each teacher: Is this teacher assigned to be the second teacher
 92 in the written defense of this student?

93 **(Hard) Constraints:**

94 The constraints are the following ones:

- 95 ■ Each student must be assigned to exactly one time slot, one teacher for the oral defense
 96 (plus their supervisor) and one teacher for the written defense (plus their supervisor).
- 97 ■ Each time slot must contain at most a certain number of oral defenses.
- 98 ■ Each teacher can only be assigned to one oral defense per time slot.
- 99 ■ Each teacher cannot be assigned to the oral defense and the written defense of the same
 100 student.
- 101 ■ Each teacher must be assigned to the number of oral defenses given in the data.
- 102 ■ Each teacher cannot be assigned on time slot on which he is absent.
- 103 ■ ... (additional requests must be taken into account)

104 **Objective Functions / Soft Constraints:**

105 The objective function contains the following parts:

- 106 ■ For each teacher and for each student he/she have already evaluated in a previous defense,
 107 a penalty is applied if a teacher is assigned to this student.
- 108 ■ For each teacher, a penalty is applied for each day the teacher is assigned to at least one
 109 oral defense.
- 110 ■ ... (additional demands must be taken into account)

	A	B	C	D	E	F	G	H	I	J	K	L	M
1		Student	Supervisor	Oral T.	Written T.			Student	Supervisor	Oral T.	Written T.		Stu
2													
3	Thursday 8h-10h	Student5	Teacher9	Teacher6	Teacher1			Student11	Teacher12	Teacher5	Teacher2		Stu
4													
5	Thursday 10h-12h	Student3	Teacher7	Teacher12	Teacher6			Student22	Teacher1	Teacher6	Teacher7		Stu
6													
7	Thursday 14h-16h	Student7	Teacher2	Teacher3	Teacher6			Student8	Teacher5	Teacher6	Teacher12		Stu
8													
9	Thursday 16h-18h	Student2	Teacher6	Teacher2	Teacher3			Student15	Teacher3	Teacher9	Teacher11		Stu
10													
11	Friday 8h-10h	Student10	Teacher6	Teacher8	Teacher3			Student18	Teacher9	Teacher11	Teacher8		Stu
12													
13	Friday 10h-12h	Student6	Teacher3	Teacher9	Teacher7			Student19	Teacher7	Teacher6	Teacher12		Stu
14													
15	Friday 14h-16h	Student4	Teacher11	Teacher8	Teacher5			Student13	Teacher6	Teacher9	Teacher8		Stu
16													
17	Friday 16h-18h	Student1	Teacher8	Teacher9	Teacher7			Student14	Teacher6	Teacher4	Teacher5		Stu

■ **Figure 1** An example of generated timetabling solution for the Oral and Written Defense for Apprenticeship Students

111 **2.3 Results and Post-Optimization**

112 An example of result given by the prototype is given in Figure 1. However, it is common to
 113 have to modify the schedule several times depending on new constraints (for example teacher
 114 unavailability). An additional part of the objective function has consequently be added,
 115 which is minimizing the number of modifications of the solution if an update is needed. An
 116 example of a such update is given in Figure 2, after Teacher 11 was known absent on Friday.

	A	B	C	D	E	F	G	H	I	J	K	L	M
1		Student	Supervisor	Oral T.	Written T.			Student	Supervisor	Oral T.	Written T.		Stud
2													
3	Thursday 8h-10h	Student5	Teacher9	Teacher6	Teacher3			Student11	Teacher12	Teacher5	Teacher6		Stude
4													
5	Thursday 10h-12h	Student3	Teacher7	Teacher12	Teacher8			Student26	Teacher11	Teacher6	Teacher7		Stude
6													
7	Thursday 14h-16h	Student4	Teacher11	Teacher12	Teacher8			Student7	Teacher2	Teacher3	Teacher6		Stude
8													
9	Thursday 16h-18h	Student2	Teacher6	Teacher2	Teacher3			Student15	Teacher3	Teacher9	Teacher5		Stude
10													
11	Friday 8h-10h	Student10	Teacher6	Teacher8	Teacher9			Student18	Teacher9	Teacher1	Teacher11		Stude
12													
13	Friday 10h-12h	Student6	Teacher3	Teacher1	Teacher9			Student19	Teacher7	Teacher6	Teacher2		Stude
14													
15	Friday 14h-16h	Student13	Teacher6	Teacher8	Teacher9			Student21	Teacher1	Teacher7	Teacher12		Stude
16													
17	Friday 16h-18h	Student1	Teacher8	Teacher9	Teacher4			Student14	Teacher6	Teacher4	Teacher7		Stude

■ **Figure 2** An example of generated timetabling solution for the Oral and Written Defense for Apprenticeship Students after the modification of a first version

117 **3 Case study and presentation of the problem**

118 In this section, we focus on the study of the global needs of one teaching department for one
 119 semester. Currently, timetabling is managed by a teacher whose are of expertise does not
 120 include mathematics or computers science. This colleague must manage two complementary
 121 aspects of timetable management:

- 122 ■ The forecast part: this involves planning the schedules in advance for future weeks
- 123 ■ The adaptive part: it involves modifying the forecast schedule so that it adapts to daily
 124 unforeseen events that affect the current schedule.

125 The objective of this current work is to lighten and help reduce the important management
 126 load of the timetabling task, especially in the forecast aspect.

127 **3.1 Students**

128 There are between 80 and 160 students each year studying in the teaching department.

129

130 A student belongs to:

- 131 ■ One year of teaching: first year, second year, third year or special year (the special year
 132 allows you to follow the first two years of teaching in a single year).
- 133 ■ One group of Supervised Work (each year contains between one and two groups of
 134 supervised work)
- 135 ■ A group of Practical Work (each year contains between one and four groups of practical
 136 work)
- 137 ■ A language group: each year contains between one and two groups for Spanish and each
 138 of the other languages, German, Italian and French, has one group containing all the
 139 students who participate whatever their year.

140 ■ Each student is either an Apprentice or an Intern. Apprentices alternate between weeks
 141 at university and weeks in which they work in a company. Interns remain at the university
 142 most of the year but have specific period reserved for an internship to be conducted
 143 within a company. First-year students are all interns. In second and third years, students
 144 choose between being an apprentice or being an intern.

145 We deduce from this information a set of student profiles P , where each student profile
 146 $p \in P$ is the non-empty intersection of a Year, a Supervised Work group, a Practical Work
 147 group, a Language group and of professional status (Apprentice or Intern). Each student
 148 profile can have unavailability and requests, like avoiding to have gaps between sessions on
 149 the same day (except for lunch).

150

151 3.2 Teaching Staff

152 One of the particularities of the training is the diversity of the instructors who provide the
 153 course sessions. In addition to teachers and teacher-researchers, many sessions are provided
 154 by external speakers, most often professionals from companies related to the areas of training.

155

156 Each instructor has constraints and requests, particularly linked to their status. For
 157 example:

- 158 ■ External speakers must travel to the training site as little as possible, which means
 159 grouping their course sessions together as much as possible.
- 160 ■ Teacher-researchers have recurring days dedicated to research.

161 Teachers and teacher-researchers assigned to the department site can have unavailabilities.
 162 Certain special cases must also be treated, including for example:

- 163 ■ Individual preferences: some teachers prefer to start early in the morning while others
 164 prefer to finish late in the evening.
- 165 ■ Family or personal constraints: For instance, we have a couple of teacher-researcher
 166 parents, at least one of whom must be free when it is time to drop the children off at
 167 school.
- 168 ■ The teaching load must be smoothed as much as possible over the year, which can be
 169 difficult because of apprentice students.

170 3.3 Classrooms

171 We can estimate at first glance that the classrooms are completely classic: amphitheater,
 172 Supervised Working Rooms and Practical Working Rooms. However, the use of the rooms
 173 differs depending on its specific characteristics. For example, one room is specifically designed
 174 for English Practical Work. This will be developed in the subsection on course sessions.

175 3.4 Calendar

176 We are currently interested in the autumn semester, which extends from approximately the
 177 beginning of September to the end of January (with some nuances depending on the years of
 178 students). Each student profile (year, apprentice or intern, etc.) has its own calendar, with
 179 its weeks of presence and absence.

180

181 Each week $w \in W$ has the following information:

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- 182 ■ The number of this week in the calendar
- 183 ■ For each year, the maximum number of working hours of this year in the week.

184 3.5 Teachings and Sessions

185 The autumn semester contains approximately 100 distinct courses, each course was composed
186 of a certain number of sessions. Each session belongs to a course, with the exception of a
187 few specific sessions, for example the back-to-school session. Depending on the number of
188 groups, we obtain approximately 1500 sessions to be placed over the semester.

189
190 Each session $s \in S$ has the following information:

- 191 ■ An identifier.
- 192 ■ The name of the overlying course. For simplification purposes, we consider a “Miscel-
193 laneous” course containing isolated sessions.
- 194 ■ A duration, in minutes. Most sessions last between 30 and 240 minutes.
- 195 ■ The list of student profiles who must follow this session.
- 196 ■ The list of speakers who must lead this session. Some sessions, such as exams, do not
197 have an assigned speaker but have a list of teachers who may be assigned to that session.
- 198 ■ A list of rooms that can be used for this session, and the number of rooms required.
- 199 ■ Information specific to the session, for example if it is an exam.

200 Sessions can be affected by specific constraints, including but not restricted to:

- 201 ■ Precedence constraint: in many cases, one session must completed before starting another
202 session
- 203 ■ Proximity constraint: the time gap between two sessions must be within a certain interval
204 of time
- 205 ■ Already-planned-session: for some sessions, we already know when the session will be
206 allocated.

207 4 Towards a first prototype

208 The initial objective is to obtain a usable prototype, which will then gradually evolve over
209 the years. The prototype will be broken down into three parts:

- 210 ■ An Excel file [10] containing all the problem data.
- 211 ■ A computer program which models and proposes a solution to the timetabling problem.
- 212 ■ An Excel file which contains the proposed timetable.

213 To solve the University Timetabling Problem, for this first prototype we propose to
214 build a mathematical model of this real-world problem (in a format similar to that of a
215 Weighted Constraint Satisfaction Problem) and to find a solution using the optimization
216 solver Hexaly (formerly called LocalSolver) [3]. Given the large number of sessions to be
217 placed, the modeling and resolution of the problem was divided into independent parts
218 summarized in the Algorithm 1:

- 219 ■ **Planning:** For each session, we decide (only) on which week this session will take place
- 220 ■ **Scheduling:** For each week and for each session planned for this week, we decide on
221 which day, at what time and in which room(s) this session will take place.

■ **Algorithm 1** University timetabling algorithm

Require: University Timetabling Data

Ensure: For each session, the week, day, time and room(s) affected to this session

- 1: Solve the Planning Problem
 - 2: **for all** Week in the Calendar **do**
 - 3: Solve the Scheduling Problem for every sessions affected to this week
 - 4: **end for**
-

222 Below is a brief overview of the proposed mathematical models.

223 4.1 Planification

224 The first part of the proposed algorithm is to assign, for each session, the week on which the
225 session will be executed.

226 **Decision variables:**

227 Decision variables are the following ones:

- 228 ■ For each session and for each week, is this session executed on this week?

229 **(Hard) Constraints:**

230 Constraints are the following ones:

- 231
- 232 1. Each session must be executed on exactly one week.
 - 233
 - 234 2. For each week and each student profile, the number of hours affected to this profile in
235 this week cannot exceed a maximum amount.
 - 236 3. Similarly, for each week and each instructor, the number of hours affected to this instructor
237 in this week cannot exceed a maximum amount.
 - 238
 - 239 4. For each week and each classroom, the number of hours affected to this classroom in this
240 week cannot exceed a maximum amount.
 - 241 ■ As we have not yet chosen the classroom of each session at this stage, we temporarily
242 consider that if a session needs one classroom among two possibilities for two hours, it
243 is the same as asking for both classrooms for one hour. This simplification will have to
244 be removed in later work.
 - 245
 - 246 5. We must respect every specific constraint requested about the sessions, the instructors,
247 the students, ...

248 **Objective Functions / Soft Constraints:**

249 The current objective function enables to compress the timetable as much as possible, i.e.
250 minimize the weighted sum by week ID of session assignments.

251

252 **4.2 Scheduling**

253 After every session is allocated to a week, we iteratively compute on every week, and on
254 every session affected to this week, the day, the time and the classroom of this session. If
255 some sessions cannot be successfully scheduled, they are rescheduled for the next available
256 week. The time is discretized on time periods which lasts 30 minutes.

257 **Decision variables:**

258 We define the following decision variables:

- 259 ■ For each session affected to this week, is this session successfully scheduled?
- 260 ■ For each session affected to this week and for each day, is this session scheduled on this
261 day?
- 262 ■ For each session affected to this week and for each time period, does this session start at
263 this time period?
- 264 ■ For each session affected to this week and for each classroom, is this session using this
265 classroom?

266 **(Hard) Constraints:**

267 Constraints are the following ones:

268

- 269 1. Each successfully scheduled session must be affected to exactly one day/time.
270
- 271 2. Each student profile can participate in at most one session at a time.
272
- 273 3. Similarly, each instructor can participate in at most one session at a time.
274
- 275 4. Similarly, each classroom can be used by at most one session at a time.
276
- 277 5. We must respect every specific constraint about the sessions, the instructors, the students,
278 ...

279 **Objective Functions / Soft Constraints:**

280 Two objective functions are identified in this situation. The first objective function is to
281 minimize the number of delayed session (i.e. maximize the number of successfully scheduled
282 sessions). The second one is to satisfy user demands as much as possible (students and
283 instructors).

Week S36										
Monday	8h	8h30	9h	9h30	10h	10h30	11h	11h30	12h	12h30
	13h	13h30	14h	14h30	15h	15h30	16h	16h30	17h	17h30
			Welcome session				Prof. Project (SY) TD1			
Tuesday	8h	8h30	9h	9h30	10h	10h30	11h	11h30	12h	12h30
	Inventory Management (SY) TD1				Excel (SY) TD1					
	13h	13h30	14h	14h30	15h	15h30	16h	16h30	17h	17h30
			Inventory Management (SY) TD2			Info. System TD1				
Wednesday	8h	8h30	9h	9h30	10h	10h30	11h	11h30	12h	12h30
	Info. System TP1				English TD1					
	13h	13h30	14h	14h30	15h	15h30	16h	16h30	17h	17h30
			Probabilities TD 1				Communication TD1			
Thursday	8h	8h30	9h	9h30	10h	10h30	11h	11h30	12h	12h30
	Excel (SY) TD2				Excel (SY) TD3					
	13h	13h30	14h	14h30	15h	15h30	16h	16h30	17h	17h30
			Probabilities TD 2			Excel (SY) TD4				
Friday	8h	8h30	9h	9h30	10h	10h30	11h	11h30	12h	12h30
					Excel (SY) TD5					
	13h	13h30	14h	14h30	15h	15h30	16h	16h30	17h	17h30
			Portfolio TD1			Excel TP1				

■ **Figure 3** An example of generated timetabling solution for the Special Year Students

5 Discussions

5.1 Towards a reliable and efficient prototype for our case study

We present in Figure 3 an example of what we are able to produce. This example shows, for one student profile, their planning for the first week of the calendar (we translated the information on the Figure in English compared to the real names). Despite this encouraging results and our abilities to satisfy every constraints of the problem, a lot of work still has to do to respect the wishes of users and the results remains not good enough. In particular, the overall timetable contains many empty slots (students have classes before the slot and after the slot, but not during the slot), empty slots caused by some bad assignments taken by the planification part of the process. Moreover, the identification of the specific needs of the teaching department, the collect of useful data, the creation of the mathematical model and the verification of the correctness of the implementation are currently being carried out.

Once this first step has been completed, it would be interesting to compare this first prototype with other techniques. The state of the art on this theme remains to be carried out but we can already cite some potential directions, like Meta-heuristics [1, 13, 14, 15], Linear Programming [6, 11, 12], Constraint Satisfaction Problem [16], and Boolean Satisfiability Problem [7, 8] among others [2, 4, 5, 9].

5.2 Multiplicity of actors, an obstacle to practical application

From our meager experience in a few French universities, the impression remains that very often, timetables were drawn up almost entirely by hand. One of the obstacles can be the multiplicity of the actors. Indeed, there are different actors involved in the university timetable management:

- There is at least one person who manages the schedules. This is often a person who has experience in management and human contact, sometimes a member of the administrative staff. It is rarely a person with skills in decision-making processes like constraint

310 programming or operational research.

311 ■ There is at least one person, or an external company, that manages the tool that displays
312 timetables. Similarly, this person rarely has skills in decision-making processes like
313 constraint programming or operational research. His/Her priority is to secure the process
314 of displaying and accessing timetables, and consequently the automation of their creation
315 often remains in contradiction.

316 ■ Finally, the people with knowledge in constraint programming or operational research
317 are often not involved in the timetabling management process.

318 How to bring together these people to ensure a better utilization of constraint programming
319 and operations research in order to improve decision-making in university timetabling? This
320 seems to be a relevant question.

321 5.3 How to generalize a such work to several teaching departments?

322 Another important part of our work will also aim to tackle the generalization of timetabling
323 management for several teaching departments. Automating this task across multiple teaching
324 departments presents significant challenges due to the unique and diverse needs of each
325 department. While one department may have to build a schedule and repeat it on several
326 weeks, others can aim to create a different schedule every week. The variability in course
327 structures means that a one-size-fits-all solution is very difficult to put in practice. Developing
328 such a system necessitates a deep understanding of the distinct operational nuances of each
329 department, making the task complex and resource-intensive.

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