

1 Modern Constraint Programming Education: 2 Lessons for the Future

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7 — Abstract —

8 This paper details an outlook on modern constraint programming (CP) education through the
9 lens of a CP instructor. A general overview of current CP courses and instructional methods is
10 presented, with a focus on online and virtually-delivered courses. This is followed by a discussion of
11 the novel approach taken to introductory CP education for engineering students at large scale at
12 the Georgia Institute of Technology (Georgia Tech) in Atlanta, GA, USA. The paper summarizes
13 important takeaways from the Georgia Tech CP course and ends with a discussion on the future
14 of CP education. Some ideas for instructional methods, promotional methods, and organizational
15 changes are proposed to aid in the long-term growth of CP education.

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22 over the years. May the Force be with them.

23 **1 Introduction and Current CP Education**

24 Constraint programming (CP) is a methodology for solving combinatorial problems to find
25 feasible or optimal solutions by using constraints to reduce the set of values each variable
26 in a problem can potentially take. The field lies at the intersection of operations research
27 and computer science and drives numerous applications in the real world, such as hospital
28 scheduling, sports tournament bracketing, delivery vehicle routing, and evacuation planning.

29 In general, CP education at the university level tends to be fairly decentralized. Unlike
30 the abundance of machine learning courses, for example, that one can find at almost every
31 university today, existence of CP courses is largely predicated on having qualified and
32 motivated CP practitioners willing to design or teach such a course. For many years, quality
33 CP education was largely limited to graduate students who were fortunate enough to work
34 in departments that offered CP courses. Christine Solmon led an online artificial intelligence
35 (AI) course for graduate students in France in the early 2000s that contained 12 hours worth
36 of sessions on Gnu-Prolog and CP. The course focused on the basics of CP modeling and CP
37 solvers while working to solve puzzles like map coloring and “SEND + MORE = MONEY”
38 [12]. Helmut Simonis started, and still runs, an online course teaching ECLiPSE in order to
39 learn CP modeling and solving techniques. Today, one may self-study using the videos and
40 handouts on the course website [10, 11]. However, in the last decade, the advent of Massive
41 Open Online Courses (MOOCs) brought opportunities to democratize CP education in a
42 structured manner with lessons and assignments for members of the general public. Pascal
43 Van Hentenryck’s Discrete Optimization course on Coursera introduced some basics of CP
44 and Large Neighborhood Search (LNS) with over 18 hours of interactive material comprising



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45 Indiana Jones-themed videos, readings, and quizzes [13]. The course also featured the use of
46 an autograder system to grade the vast number of submissions from members of the general
47 public. Johannes Waldmann also developed an ‘autotool’ autograder framework along with
48 some exercises focused on understanding the fundamental algorithms behind CP solvers [14].
49 More recently, Jimmy Lee and Peter Stuckey co-developed three Coursera MOOCs on the
50 subject of “Modeling and Solving Discrete Optimization Problems” [1, 5]. The courses feature
51 a form of problem-based learning encapsulated in a coherent story plot following classic
52 Chinese novels. Lee and Stuckey presented learner statistics and feedback, and discussed
53 their experience with adopting the online materials in a smaller flipped classroom setting
54 as well. In 2022, Hoffmann et al. produced the first human-centered study addressing how
55 people approach constraint modeling and solving [3]. This information will be useful in
56 pedagogical design for future CP courses. Pierre Schaus, Laurent Michel, and Pascal Van
57 Hentenryck launched a CP MOOC in February 2023 on edX using the Mini-CP solver, a
58 lightweight, open-source CP solver designed for educational purposes [6]. The Mini-CP solver
59 comes with a series of more than 20 implementation projects to help students with the basics
60 of CP modeling and search heuristics.

61 The effects of the COVID-19 pandemic over the last few years have signaled a shift in
62 the way educational content is delivered and consumed. In order to continue growth as a
63 practice, the CP community must adapt to these new pedagogical styles. This paper draws
64 on the authors’ experiences from an online CP course at the Georgia Institute of Technology
65 (Georgia Tech) to detail lessons learned and novel strategies for introductory CP education.

66 **2 The Georgia Tech CP Course**

67 The Georgia Tech CP Course was started in Fall 2018 by Pascal Van Hentenryck. The course
68 was initially taught as an in-person course and focused on using the OPL programming
69 language inside the IBM ILOG CPLEX Optimization Studio to model CP problems. The
70 first third of the course focused on learning OPL and modeling puzzles with CP. The second
71 third of the course goes from modeling basic optimization toy problems to condensed versions
72 of real-world optimization problems. The last third of the course focused on CP applications
73 in scheduling, routing, and evacuation planning. The course was housed in the Industrial
74 and Systems Engineering department at Georgia Tech and was open to both advanced
75 undergraduate and graduate students within the department. The course primarily focused
76 on finite-domain CP as an approachable introduction within the College of Engineering. The
77 enrollment of this initial CP course offering was 26 students.

78 With the COVID-19 pandemic forcing instruction to be delivered online in 2020, Van
79 Hentenryck adapted the method of course delivery considerably in order to maintain effective-
80 ness in a virtual format. This redevelopment modified the Georgia Tech CP course into the
81 form which it takes today. The size of the Georgia Tech CP course grew considerably after
82 this point. While the course was initially offered yearly in the Fall semesters, the frequency
83 was increased to both Fall and Spring semesters starting in Spring 2023. Also in Spring 2023,
84 Tejas Santanam joined the CP course as an instructor, having been a prior student and head
85 teaching assistant for the Georgia Tech CP course.

86 This is the only CP course offered at any level at Georgia Tech.

87 The key features of course logistics for the Georgia Tech CP course are detailed below.
88 In total, the course lasts for a full semester (approximately 15 weeks), with an expected 8 to
89 15 hours of workload each week.

90 Course Goals

91 The learning outcomes for students are as follows.

- 92 ■ Describe the fundamental properties of good constraint programming models and how
- 93 they differ from other methodologies.
- 94 ■ Be able to determine when/how to use constraint programming for practical applications
- 95 in areas such as scheduling, routing, and resource allocation.
- 96 ■ Achieve fluency in the modeling language OPL for constraint programming.
- 97 ■ Recognize when additional features (e.g., new constraints and dedicated search procedures)
- 98 are necessary to solve a problem and understand what this involves.

99 Topic Outline

100 The topics covered through the course are as follows. In some iterations of the course, the

101 advanced topics towards the end of the list were required only for graduate students.

- 102 ■ Basic Concepts
 - 103 ■ Getting started
 - 104 ■ Basic concepts I
 - 105 ■ Basic concepts II
 - 106 ■ OPL Primer
- 107 ■ Elements of Constraint Programming
 - 108 ■ Reified constraints
 - 109 ■ Optimization
 - 110 ■ Expressions
- 111 ■ Theoretical Foundation
 - 112 ■ Computational Model
- 113 ■ Global Constraints
 - 114 ■ The element constraint
 - 115 ■ The table constraint
 - 116 ■ Combinatorial Constraints
 - 117 ■ The pack constraint
 - 118 ■ The circuit constraint
 - 119 ■ The lex constraints
- 120 ■ Modeling in Constraint Programming
 - 121 ■ Symmetry breaking
 - 122 ■ Subexpression elimination
 - 123 ■ Redundant constraints I
 - 124 ■ Redundant constraints II
- 125 ■ Search in Constraint Programming
 - 126 ■ Search tree and Impact
 - 127 ■ Restart and nogoods
- 128 ■ Implementation of Constraint Programming
 - 129 ■ Packing
 - 130 ■ AllDifferent
 - 131 ■ NoOverlap
- 132 ■ Scheduling in Constraint Programming
 - 133 ■ Interval variables and noOverlap
 - 134 ■ The Sequence Constraints
 - 135 ■ Cumulative Constraints

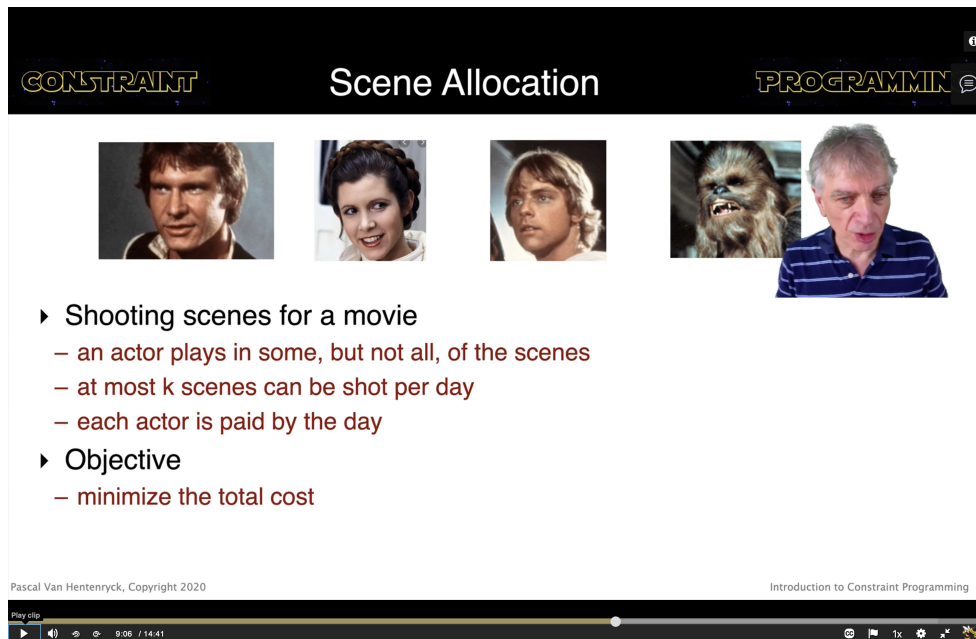
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- 136 ■ The House Problem II
- 137 ■ The House Problem III
- 138 ■ The Perfect Square Problem
- 139 ■ State Constraints
- 140 ■ The Trolley Application
- 141 ■ Optional Activities
- 142 ■ Standard Scheduling Problems
- 143 ■ Calendars
- 144 ■ Advanced Topics
 - 145 ■ Large neighborhood search
 - 146 ■ Scripting models
 - 147 ■ Routing
 - 148 ■ CP in Python
- 149 ■ Implementation in MiniCP
 - 150 ■ Semantics of CP
 - 151 ■ Operational Model of CP
 - 152 ■ Inference
 - 153 ■ Search
 - 154 ■ Advanced Inference
 - 155 ■ Advanced Search

156 **Lecture Videos**

157 The material in the course is presented in high-quality videos. In total there are about 90
158 videos comprising around 30 hours of material. Van Hentenryck recorded a video for each
159 topic listed above with greenscreen backgrounds, animations, and more. The videos generally
160 have Van Hentenryck's head superimposed on slides or images, which make for clearer and
161 higher quality videos compared to recordings of traditional classroom proceedings. The
162 videos are relatively short and digestible, with lengths ranging from 10 to 30 minutes each.
163 The students are able to play, pause, and rewind these videos, as well as change the playback
164 speed. A flipped classroom style approach is utilized where the lecture videos are posted in
165 advance of the scheduled class sessions. Lecture videos are complemented with interactive
166 online sessions that happen during the scheduled course time. An example of one of the
167 lecture videos can be seen below in Figure 1. The PDF files of the slides used in each lecture
168 video are also provided to the students.

■ **Figure 1** A look at one of the Star Wars-themed lecture videos



169 Interactive Sessions

170 The interactive sessions take place on Zoom and focus on review of the material from the
 171 most recent set of lecture videos. The interactive sessions include time for Q&A and go over
 172 the course assignments. They also include one-on-one sessions in Zoom breakout rooms with
 173 the instructors and the teaching assistants. These breakout room sessions allow students to
 174 get customized face-to-face help. The sessions are frequently used for help with conceptual
 175 understanding and debugging code for the assignments. The interactive sessions take place
 176 two to three times a week for about an hour each day. Teaching assistants also offer office
 177 hours on an ad hoc basis to help those unable to join a particular interactive session.

178 Discussion Forums

179 The Georgia Tech CP Course also has an attached discussion forum where students can
 180 ask conceptual questions publicly and assignment code debugging questions in a private
 181 post to course instructors and teaching assistants. Students may answer the questions of
 182 their peers or leave follow-up comments and questions to others' questions. Students are
 183 guaranteed responses within 24 hours; however, the mean instructor (including teaching
 184 assistants) response time is 47 minutes. The response time is often even faster than that
 185 during the workday.

186 Assignments

187 The course contains many assignments that are due on a roughly weekly basis. Each
 188 assignment relates to applying the topics covered in the lecture videos and interactive sessions
 189 from the previous week. Each assignment can be solved with the knowledge students have
 190 learned to that point, as well as with the existing functionalities of OPL and the OPL
 191 IDE. Every assignment involves an application-focused modeling problem that requires a

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192 model coded in OPL as a solution. The assignments start simple with problems like map
193 coloring before ramping up to more difficult problems like the Capacitated Vehicle Routing
194 Problem with Time Windows (CVRPTW) and flood evacuation planning. Students start
195 with assignments from the very first week to cultivate their constraint programming mindset
196 and familiarize themselves with the OPL language. This is in line with the philosophy
197 espoused by Patrick Prosser in his CP 2014 talk, where he discussed getting students to solve
198 problems as soon as possible [8]. The focus in the assignments is on modeling and writing
199 appropriate constraints. Students in the course use the solver from IBM ILOG CP Optimizer
200 out of the box, rather than writing custom search procedures. There are a few assignments
201 towards the end of the course that involve CP in Python (due to students' familiarity with
202 Python) and Mini-CP (primarily for graduates students; a way to explore writing search
203 procedures). Some of the introductory assignments are reused year to year, though most
204 of the assignments receive changes ranging from minor tweaks all the way to brand-new
205 exercises. Some of the assignments require data file inputs which are provided to the students.
206 For many of the assignments in the latter half of the course, multiple data files are provided
207 for testing as students devise their formulations. However, students are asked to make their
208 formulation general in nature, as the provided dataset to the students is not necessarily the
209 one used for verifying and grading their model.

210 Theme

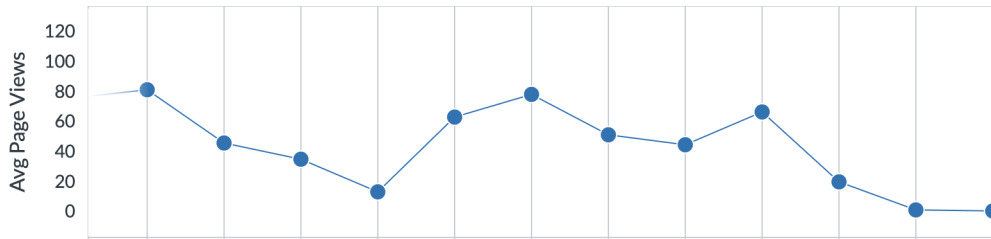
211 Similar to the adoption of an Indiana Jones theme in Van Hentenryck's earlier MOOC [13] or
212 the classic Chinese novel themes in Lee and Stuckey's MOOCs [1, 5], the Georgia Tech CP
213 course uses a Star Wars theme. All of the lecture videos involve Star Wars-themed examples,
214 graphics, and even costumes! At the time of writing, a viral video of Van Hentenryck teaching
215 while dressed as Yoda has garnered over 1.5 million views on TikTok. The assignments also
216 all have problem descriptions pertaining to happenings in the Star Wars universe. All course
217 instructors are referred to as Jedi Masters (Van Hentenryck is Yoda, while Santanam is
218 Obi-Wan Kenobi), teaching assistants are referred to as Jedi Assistants, and students are
219 referred to as Padawans. At first glance, the theme may seem gimmicky, but it actually
220 serves to motivate the students and keep them engaged with some light-heartedness. There
221 is no requirement of prior Star Wars knowledge, nor does such knowledge provide a leg
222 up in the course. The use of the Star Wars theme, specifically, is allowed for educational
223 use internally. However, it would infringe on Disney's copyright if the lecture videos and
224 assignments were to be public-facing.

225 Student Patterns of Interaction

226 It is also important to shed light on the manner in which students engage with course material.
227 The graph below in Figure 2 shows the average page views per student per week over the
228 course of a semester. Page views includes actions like viewing a lecture video, reading lecture
229 slides, or going to the interactive session page. The interaction pattern follows a cyclical
230 up and down flow. Students spend more time with the material when it is new, and then
231 focus on the applications as they get more used to the material. Then, when new material is
232 introduced, the page views jump up again. Apart from the drop near the beginning of the
233 semester associated with a school holiday week, interactions until Georgia Tech's final exam
234 period remain consistently above 30 page views per student per week. Page views tend to
235 be higher in the second half of the course when there is a focus on scheduling and routing
236 applications in CP. In general over the years, the most watched lecture videos have been

237 the videos on global constraints, sequence variables, reification, and routing. We can likely
 238 assume that these videos reflect the topics the students found most challenging or needed
 239 the most review on. Global constraints and reification are covered in the first half of the
 240 course, while scheduling are routing are covered in the second half. Thus, the Georgia Tech
 241 CP course does well at distributing these challenging topics.

■ **Figure 2** Average weekly student interaction with the course material over the semester



242 As far as patterns of interaction within the interactive session, there have been three groups
 243 of students that have been generally observed in the course. One group of students attend
 244 every interactive session no matter what. The second group of students attend sporadically,
 245 which is usually only when they need help and at times close to assignment deadline. The
 246 last group of students function independently and don't come to any interactive sessions.
 247 The first group of students are generally strong performers in the course and understand
 248 the concepts well. The second group of students tend to be grade motivated and sometimes
 249 perform well, but do not usually build a deeper understanding of CP. The third group is a
 250 unique case where some students establish a good enough understanding from the videos
 251 and don't need the sessions while other students in the group are falling behind and struggle
 252 in the course. It is important to look at the assignments of students in this third group
 253 to identify if they need extra support or not. In addition to the grade received on the
 254 assignment, it is also helpful to look at the way constraints are written to see if the students
 255 has a certain maturity in their modeling. After all, there is a large difference between a
 256 concise and effective constraint and a hardcoded constraint specific to the dataset, even if
 257 they produce the same result.

258 **Reception**

259 To date, there have been six completed CP courses at Georgia Tech between Fall 2018 and
 260 Spring 2023. With each iteration of the course, the popularity of the course within the
 261 Industrial and Systems Engineering department has grown. Table 1 shows the number of
 262 students who took each iteration of the Georgia Tech CP course. The Spring 2023 course
 263 was intentionally set to a lower capacity as it was the first time the course was offered
 264 in back-to-back semesters. At the time of writing, the upcoming Fall 2023 course has 96
 265 students registered with a further 33 on the waitlist. Enrollment in the Georgia Tech CP
 266 course has had full registration with spillover onto waitlists in every semester since Fall 2019.

■ **Table 1** Enrollment for past iterations of the Georgia Tech CP course

Semester	Fall 2018	Fall 2019	Fall 2020	Fall 2021	Fall 2022	Spring 2023
Enrollment	26	41	94	100	183	30

267 Despite the growth in enrollment, the quality of the Georgia Tech CP course has been

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268 maintained. Surveys were given to all students in the last five iterations of the course. Some
269 questions with average ratings on a 1-5 scale, with 1 being the lowest possible rating and 5
270 being the highest possible rating, are shown in Table 2 below. The course has maintained
271 consistently high ratings since its inception.

■ **Table 2** Survey ratings for past iterations of the Georgia Tech CP course

Semester	Fall 2019	Fall 2020	Fall 2021	Fall 2022	Spring 2023
Number of Respondents	15	82	93	153	26
Amount Learned	4.9	4.8	4.5	4.4	4.6
Instructor stimulates interest	4.9	4.95	4.8	4.6	4.8
Instructor effectiveness	5	4.97	4.9	4.7	4.7
Course effectiveness	4.9	4.92	4.8	4.3	4.6

272 Some written feedback from the surveys can be seen below:

- 273 ■ “The structure was well organized to incrementally build our familiarity with the material.
274 The flipped classroom format was also effective, in being able to review the recordings
275 several times, as well as being able to revisit the recordings later in the semester. The in
276 class sessions were useful to solidify our understanding.”
- 277 ■ “The lecture videos were high quality and engaging. The professor themed the entire course
278 and the amount of effort he put into making it work in a virtual format is commendable!
279 I loved that I could watch and re-watch the lecture videos as many times as I needed to.
280 The interactive sessions were a must have. Getting to interact with the professor after
281 having watched the lecture videos and read through the assignments really helped and he
282 generally gave great pointers and advice. The breakout rooms were must haves as well
283 as often the questions I wanted to ask would not have been appropriate to ask to the
284 class. Finally, I thought the assignments were incredible benchmarks for whether or not
285 we fully grasped and could synthesize many concepts from the course.”
- 286 ■ “I really enjoyed a lot of the assignments, and the theme of the course was perfect. I
287 am a huge star wars fan (currently getting through all of the animated shows which I
288 highly recommend), and the fun assignments made me want to complete them and I
289 always looked forward to the story of the assignments. The interactive sessions were also
290 super helpful and an amazing idea. I always got great help there especially in the 1-to-1
291 breakout rooms.”
- 292 ■ “It is hard to say because there were so many things that I loved about the way this class
293 was taught. 1. I would say the interactive sessions (at a convenient (5 PM) after my
294 full-time job) and the Ed Discussion were some of the best features for getting help; I
295 felt like I was able to get the face-to-face and virtual support I needed way more than in
296 other classes I’ve taken. The professor and TAs went out of their way to make themselves
297 available in this class, way more than in other classes. In other classes, I have had to seek
298 out help more, and it is harder as a Distance Learning student to ask questions without
299 the ability to do so in real-time. But this class does it! 2. As a close second, having our
300 grades be based on assignments is the way I like to learn – learning by applying and
301 doing (rather than solely being tested on theory).”

302 In addition to the favorable enrollment and positive survey results, the Georgia Tech CP
303 course has received further accolades. The course received the Teaching Excellence Award
304 for Online Teaching and the Student Recognition of Excellence in Teaching: Class of 1934

305 Award for the Fall 2020 iteration of the course. Furthermore, the course received the Student
306 Recognition of Excellence in Teaching: CIOS Honor Roll for the Fall 2020, Fall 2021, and
307 Fall 2022 iterations of the course, an award reserved for teaching excellence with large class
308 sizes.

309 **3 Lessons Learned**

310 Over the many iterations of the Georgia Tech CP course, both the authors have gained
311 numerous insights into the learning process. Van Hentenryck has been instructing the course
312 since its inception, while Santanam has been involved in every iteration of the course for
313 the last four years as either a student, teaching assistant, or instructor. Some key areas of
314 emphasis are expounded upon below.

315 **Teaching to Undergraduate Students**

316 One of the unique parts of the Georgia Tech CP course is the large presence of undergraduate
317 students each year. In every iteration of the course, undergraduate students have comprised at
318 least 75% of the class. Given the audience, this necessitates a different approach to CP than
319 what one might generally see at most universities. Given their age and relative inexperience
320 in the field, it is much more important to cultivate an interest in CP (or optimization in
321 general) in each student. The Bachelor of Science in Industrial Engineering degree at Georgia
322 Tech has specializations in analytics, statistics, economics, operations research, supply chain,
323 and a general studies track as well. It is very likely that the majority of students in the class
324 have other backgrounds and interest. Thus, many who take the CP course are doing so to
325 learn something new, rather than trying to build upon a longstanding interest. It is for this
326 reason that teaching in an engaging way is important. Presenting the real-world applications
327 of CP and allowing undergraduate students to get hands-on with modeling early in the course
328 is vital to enable a student to realize their interest or disinterest in the subject. A focus
329 on real-world applications is also helpful for an undergraduate audience due to the large
330 percentage of undergraduate students that go straight to work in industry upon completing
331 their degree. For them, the focus is on developing the modeling skills and the problem-solving
332 mindset over the theoretical underpinnings one might expect a graduate student to care
333 more about. A last distinctive adaptation needed when teaching to undergraduate audiences
334 is the role of coding in a CP course. At an undergraduate level, many students lack the
335 coding maturity that allows one to take a new programming language and efficiently write
336 the necessary syntax to execute on an ideated model. Learning a new programming language
337 in OPL is a difficult task for many, and most undergraduate students lack the full ability
338 to teach themselves new syntax solely by reading documentation. Undergraduate students
339 require and continually ask for a good deal of coding examples and demonstrations, as well
340 as help debugging and interpreting documentation. Thus, when teaching to undergraduate
341 students, it is important to make the coding parts of a course as accessible as possible. One
342 must decidedly think of coding exercises as a means to enable the building of CP-related
343 intuition in undergraduate students and lower the barriers to entry there.

344 **Teaching to Engineering Students**

345 Another novel aspect of the Georgia Tech CP course is that it is housed within the Industrial
346 and Systems Engineering Department inside the College of Engineering. The majority of
347 CP courses around the world are taught by computer science faculty to computer science

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348 students, as opposed to the setup at Georgia Tech. Teaching CP to engineering students
349 also requires careful consideration in pedagogical choices. Since the students hail from an
350 engineering background, they tend to be very focused on the impact that CP-driven processes
351 have on real-world systems. Most students seem to be more motivated by those outcomes
352 CP can drive rather than by the inner workings of CP itself. For that reason, the Georgia
353 Tech CP course has videos on real-world case studies. For example, Van Hentenryck presents
354 two videos detailing applications of CP for integrated container terminal operations and
355 scheduling at ports. The coding aspect of CP is also somewhere where engineering students
356 may not be as advanced as their computer science peers. At Georgia Tech, students only
357 have exposure to two semesters of Python before taking Constraint Programming. This can
358 make it difficult for some students in learning a new programming language, and may lead
359 to difficulties in debugging or expressing complex constraints.

360 Modeling-Focused Teaching

361 One of the main ways that the Georgia Tech CP course makes the subject more accessible
362 at an introductory level is through a strong focus on writing CP models. The emphasis on
363 modeling helps students understand how to breakdown a problem from a CP perspective
364 and express it via a set of constraints. Since the students (particularly undergraduates in
365 engineering) are motivated by the applications of CP, the use of black-box solvers serves
366 to expedite the process as a whole and allows a student to go from problem to results and
367 output much quicker. Largely limiting the instruction and assessment to modeling enables
368 instructional staff to reclaim time that can be used to add depth or breadth in applications
369 of CP. By way of the modeling emphasis, students also learn about elegance in model
370 formulation and are encouraged to write constraints multiple different ways. Students end
371 up generating a wide variety of creative formulations. Details and concepts behind search
372 procedures and CP solvers are covered in the course, but the emphasis on the assignments
373 largely shifts away from that.

374 Autograders

375 In order to teach the Georgia Tech CP course to hundreds of students every year, an
376 autograder system had to be implemented for efficient grading. Manually having instructors
377 and teaching assistants run each model and verify each solution would be inefficient. For the
378 Georgia Tech CP course, a master Python script takes all the model files submitted by the
379 students and runs them a few at a time. The model files produce an output in a format that
380 is pre-specified in the assignment instructions. Students are encouraged to model however
381 they want with whatever variables they want. Normally, a wide range of formulations are
382 seen. The only ask is to convert their output into the pre-specified output which usually
383 comprises a trivial post-processing step. An OPL script is then used to verify if the students'
384 solutions are feasible and/or optimal. The results on each model's correctness is written
385 to a CSV file that instructional staff can use for grading. Students who did not pass the
386 autograder have their models manually checked for partial credit and are given feedback on
387 where errors were committed. Without this kind of autograder system, countless additional
388 hours would have to be spent by course instructional staff every semester. The lack of an
389 autograder system would mean having to set a lower capacity limit on the course. Thus, such
390 a system enables the spreading of CP education to larger numbers of students and reduces
391 the marginal cost of time and effort required for each additional students. In general, any
392 course, whether in-person or online, or MOOC or university-based, would be recommend

393 to have an autograder. Thanks to the fact that we can verify the solutions to NP-complete
394 problems in polynomial time, the autograder ensures fairness and consistency in grading,
395 doing so in a timely manner.

396 **Engaging Large Audiences**

397 Ensuring that every students stays engaged, interested, and has their educational needs met
398 is one of the most challenging things to do as an instructor. Especially when delivering CP
399 MOOCs or large CP classes over 100 students, connecting with each and every students
400 may seem like a daunting tasks. There are, however, a few strategies that one can employ.
401 The first method is to make the learning environment and process as fun as possible. In the
402 Georgia Tech CP course, this is achieved through immersion in the Star Wars theme. All
403 videos and assignments are tinged with Star Wars lore, characters, locations, and more. The
404 illusion is kept up in both verbal and written communication from the instructional staff,
405 who only use the Star Wars names for themselves. Within the interactive sessions on Zoom,
406 instructional staff also have their Zoom backgrounds set to Star Wars-related imagery. For
407 many students, this can bring a lightheartedness to the class by relating a new, daunting
408 subject with something familiar. Other students may also find the journey of “becoming
409 a Jedi” motivating and place themselves within the story as they strive to complete their
410 assignments as part of their mission. In addition to a fun theme, engaging large audiences
411 required multiple potential points of engagement with students. In the Georgia Tech CP
412 course, students can access the content via lecture videos 24/7, while also having interactive
413 sessions multiple times a week. These interactive sessions allow for tighter knit review, in
414 addition to one-on-one interactions with each student in attendance. For students with
415 schedule or time zone issues (especially in the case of online courses and MOOCs), interaction
416 via a discussion or Q&A forum is yet another option. Lastly, ad hoc touchpoints with any
417 member of the instructional staff makes learning and assignment help available for every
418 student. The last method for large audience engagement is about accommodation of different
419 learning styles. For those who learn by watching, videos are a great solution. Those who
420 prefer to read can review the posted PDF lecture slide handouts. Students who learn “by
421 doing” can code along with demos and gain numerous opportunities for practice through
422 the weekly hands-on assignments. Students who learn through one-on-one discussion find
423 the breakout rooms in the interactive sessions useful, while those students who do not need
424 any help on particular assignments are not required to engage with any of the resources; the
425 materials and touchpoints for help are there as and when each student needs them.

426 **Teaching via Distance Learning**

427 Delivering a successful online course requires a reliable set of technologies to produce a
428 quality student experience. It can never be overstated how important good audio and video
429 equipment is in creating high-quality videos. Students need to be able to clearly see and hear
430 instruction in order to properly internalize it. Additional work in animation and graphics
431 is helpful but not required; clear content is more important than theatrics. Beyond lecture
432 video production, tools for interactive video sessions and breakout rooms are also vital for
433 seamless transition between small group and one-on-one environments. The Georgia Tech CP
434 course has used both Zoom and BlueJeans effectively in the past. A proper discussion forum
435 that handles question and answer posts between students and instructional staff is a useful
436 tool for communication, debugging, and conceptual help offline outside of interactive sessions.
437 The Georgia Tech CP course has used both Ed Discussion and Piazza. In a large online

438 environment where students are not all located in the same place, empowering students to
439 help other students in the discussion forums foments greater understanding of the material
440 between students. This can happen holistically, but it may behoove a CP instructor to offer
441 extra credit or some similar incentive for answering the questions of peers. For overall course
442 structure, housing course materials, announcements, and links in a learning management site
443 like Canvas, Sakai, or Google Classroom can also be beneficial organizationally. Lastly, it
444 is important to check in with students that one may see really struggling or falling behind.
445 Without seeing the students physically in the classroom, it can be easy for students to fall
446 through the cracks due to the isolation. Instructional staff may also not be fully aware of
447 the extent of a student's personal situation. For any students that are falling behind, it is
448 generally good practice to reach out and inquire on ways to accommodate or assist them.

449 **4 The Future of CP Education**

450 The lessons presented above offer some thoughts and guidance for constructing a successful
451 CP course in modern times. However, looking ahead to the future, there will need to be
452 further adaptations and changes within the CP community as a whole to ensure the long-term
453 growth of CP in both the education and research spheres.

454 **Promotion of CP**

455 One of the main obstacles facing CP education going forward has to do with promotion.
456 The fact of the matter is, CP is a form of artificial intelligence (AI). However, this fact is
457 grossly understated in educational spheres. The machine learning (ML) community has so
458 tightly hitched their wagon to AI that many members of the general (and scientific) public
459 often conflate the two and use the terms AI and ML interchangeably. This is not to say
460 that CP must suddenly match ML in terms of media coverage and hype, but the way the
461 CP community positions the subject can have a great impact on the desire of new students
462 wanting to take an introductory course in CP. If one were to describe CP as AI that solves
463 Sudoku in milliseconds and powers delivery systems within massive supply chains, this line
464 of messaging could be more appealing to young students today. The challenge of CP relative
465 to fields like ML is also something that could be spoken about more when promoting to
466 young students in computer science and engineering. ML at an introductory level is largely
467 a problem of organizing data appropriately and putting it into a black-box function that
468 provides a trained model. On the other hand, even at the most introductory stages, CP
469 involves writing out a uniquely customized and specific model for each problem and going a
470 step beyond data organization in order to solve problems. The additional skill needed here is
471 something that could prove appealing to students and motivate them in their CP journey.

472 **Introductory CP Resources**

473 Another positive development with regards to CP education would be greater availability of
474 introductory resources, especially at the undergraduate level. To use the ML community as
475 a point of comparison yet again, undergraduate learners can easily sink their teeth into *An*
476 *Introduction to Statistical Learning* [4], which is also fairly standardized and used at many
477 universities across the world. The mathematical programming community also has fairly
478 standard undergraduate-level introductory texts like *Optimization in Operations Research* [9]
479 and *Operations Research: Applications and Algorithms* [15]. It would be beneficial to the CP
480 community to have such approachable texts with relative standardization in adoption across

481 universities. The focus of this kind of text should be strongly in the applied realm. From
482 experience, undergraduate students are often loathe to dive into textbooks with numerous
483 theorems and proofs. Rather, they would prefer a book with numerous worked examples
484 and code samples showing toy examples of real-world applications. A book that does this
485 well in the ML field is *Hands-on machine learning with Scikit-Learn, Keras, and TensorFlow*
486 [2]. Of course, a textbook is not the only solution to this problem, but a standardized
487 collection of introductory-level information along with a guide or outline that instructors can
488 follow would be extremely beneficial, be it a textbook or an organized website. However, in
489 presenting application-driven textbooks and other resources, the CP community must also
490 standardize more on the set of tools used to introduce coding to students. In the ML world,
491 most students first use Python with libraries like Scikit-learn [7]. In the CP world, there are
492 multiple different languages used across many introductory courses. Rather than potentially
493 overwhelming students with the choice between OR-Tools, MiniZinc, CPLEX, Mini-CP, and
494 more, it would be better for the community as a whole to focus in on specific languages and
495 solvers to use when educating newcomers to the subject. It may seem unlikely after 40 years
496 of CP that the community would suddenly align on a set of tools, but adjustments to at
497 least focus on a smaller subset would be a step in the right direction.

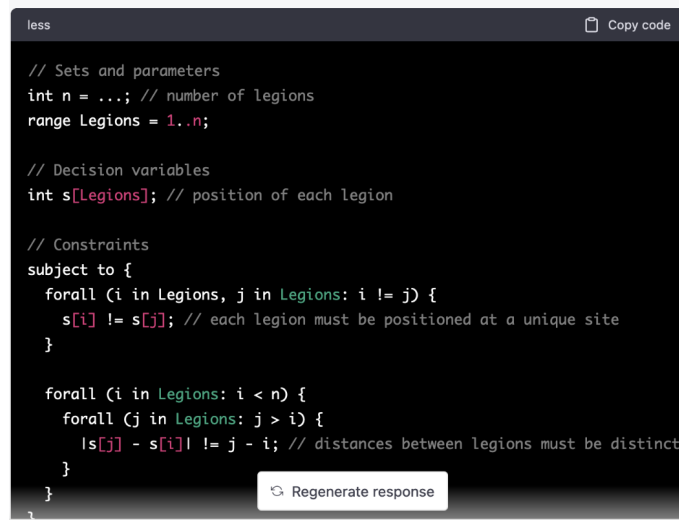
498 **Availability of CP Courses**

499 Further proliferation of CP courses at both the university level and MOOCs is vital for the
500 survival and growth of CP as a practice. Comparing again to ML, almost every university
501 with a computer science, math, or engineering department these days has at least one course
502 in ML. Similarly, there are innumerable websites with online courses for ML topics at all levels.
503 The CP community should place special focus on outreach to schools without CP courses
504 or faculty members with a CP background. Furthermore, continued promotion of existing
505 MOOCs along with the development of new MOOCs will help perpetuate a virtuous cycle.
506 Without new students and the continued development of those starting their CP journeys,
507 new researchers in CP cannot be trained and cannot go on to spread their ideas around the
508 world. The fact of the matter is that CP education and CP research go hand in hand. One
509 cannot survive without the other. Thus, a focus on teaching CP will ultimately derive benefits
510 in the research sphere as well. Additionally, the updating of centralized lists of CP courses like
511 the one found on Pearltrees (<http://www.pearltrees.com/constraints/courses/id39842792>)
512 would be helpful to students in finding educational resources, as well as helpful to instructors
513 to gain inspiration on ways to improve their own CP courses.

514 **The Impact of Large Language Models**

515 Large Language Models (LLMs) represent a fundamental challenge / threat for every project-
516 based course. Constraint programming is no exception. ChatGPT was run on the assignments
517 for the Georgia Tech CP course. The results were stunning. On the early assignments,
518 ChatGPT essentially produced the correct models. As the assignments got more involved,
519 ChatGPT output models that were close to the solutions, but typically had syntax or type
520 errors. However, these models certainly give students a strong basis to start from. Figure
521 3 shows a model for an assignment that abstracts a real problem. This assignment is the
522 first in the second part of the course, where the projects get more realistic. ChatGPT, at
523 this point, struggles for the third part of the class, which is heavily based on scheduling and
524 routing.

525 Assignments are changed every year in the Georgia Tech CP class, but they build around



```

less
Copy code

// Sets and parameters
int n = ...; // number of legions
range Legions = 1..n;

// Decision variables
int s[Legions]; // position of each legion

// Constraints
subject to {
  forall (i in Legions, j in Legions: i != j) {
    s[i] != s[j]; // each legion must be positioned at a unique site
  }

  forall (i in Legions: i < n) {
    forall (j in Legions: j > i) {
      |s[j] - s[i]| != j - i; // distances between legions must be distinct
    }
  }
}
Regenerate response

```

■ **Figure 3** A Model Produced by ChatGPT for an Assignment in Resource Allocation.

526 the same core problems. If students have access to a model from previous years (which they
 527 are forbidden to do), it is conceivable that LLMs will fill the gap, putting increasing burden
 528 on instructors to fundamentally change assignments each year and create assignments that
 529 are radically different. Or, perhaps, in a world where LLMs will become a fundamental
 530 tool, it becomes important to rethink entirely how modeling and problem-solving courses are
 531 taught.

532 **5 Conclusion**

533 The Georgia Tech CP course detailed in this paper demonstrates some novel ways to teach
 534 an introductory online CP course. The lecture videos and interactive sessions provide a
 535 fun way to mix large-scale instruction with focused efforts in small groups and one-on-one
 536 settings. The focus on modeling and solving applied problems helps hone the problem solving
 537 intuition of students while shoring up their coding skills.

538 The paper also discusses the adaptations needed when working with students from
 539 backgrounds like undergraduate engineering. Further observations are made on the use of
 540 technology in autograding and distance learning. Some thoughts from the authors on future
 541 directions of CP education (especially at an introductory level) are also discussed along with
 542 actionable recommendations for their implementation.

543 Ultimately, the experiences and thoughts put forward in this paper only comprise ob-
 544 servations in a small segment of the CP community. It would be most beneficial to have
 545 discussions with CP educators around the world to learn more about other ways of effectively
 546 teaching the subject.

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