

An Evolutionary Approach to Timetabling

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Introduction

We constructed an easy to use study scheduling tool to assist students in managing their homework. It is based around genetic algorithm techniques, which are a subclass of evolutionary algorithm.

Background

Evolutionary algorithms are different from raw implementations of solutions, and are a subset of optimization problems that derive strategies from real world speciation and evolution. As opposed to building globally-optimal strategies or combining locally-optimal ones, evolutionary algorithms adjust and adapt their current solutions over many generations to best fit some pre-defined desired goal. The process is intended to follow evolutionary theory, where a species adapts its genepool over time to best match environmental conditions and maximize likelihood of survival.

Genetic algorithms phrase solution-finding in terms of those natural adaptive strategies, like gene mutation and natural selection. The solutions, which are all progressive generations on some defined common ancestor, are ranked based on their position in some “fitness landscape.” In the nature, the fitness function is most often the likelihood to survive and procreate. In the case of homebrew algorithms that fitness could be anything, like “how far a robot moves before falling.”

These differ from typical gradient descent or astar path finding algorithms (See (Patel 2020b)) in that we are not evaluating a gradient in order to determine a direction to move towards, but simply trusting that evolutionary processes will result in a fitness improvement.

In *similarity* to path finding algorithms, this process is done through heuristic methods which find approximations of a solution (See (Patel 2020a) and (Gour 2018) for more.)

We first thought to use a path finding heuristic setup. We used Amit Patel's fabulous pathfinding website (Patel 2020b) for help building our foundational knowledge on heuristic-based path finding. They have another page on the site, found at (Patel 2020a) which we used to learn more about heuristics in-depth in order to figure out how we may be able to adapt it to our own algorithm.

We eventually found the technique of genetic algorithms promising, so we dove into that. We used established crossover and mutation techniques (Miller and Goldberg 1995) found through research to construct a model for our system and evolve it.

Problem Space

We chose to study the details and implementation surrounding homework scheduling. We implemented a simple genetic homework scheduling system which assists students in optimizing when to study. It can be difficult to balance NINJA hours, class due dates, and lengthy assignments, but using a heuristic-based schedule in the context of time, we will try and give suggestions for how to improve the studying experience. Parameters to our weighting system could include due dates for specific homework, time entry from the users, when class is happening, ninja-hours optimization, and desired sleeping hours.

We see parallels to the knapsack and other pathfinding problems, and also in concepts used in CPU scheduling. Timetabling is also a very defined homework space in regards to university class scheduling methods, where improving the efficiency and optimality of solutions is important and increasingly complex.

Design Process

We decided that segmenting our problem into discrete chunks would be the first step to solving it. Each day is divided into half-hour chunks which is the level of fidelity we are solving the problem at. We figured that this would be sufficient due to most classes (at Olin) fitting into roughly half-hour increments. It would also simplify mutating and crossing solutions in very defined list of temporal possibilities.

Most implementations of genetic algorithms utilize list-based structures to assist in simplifying mutation and cross-ing functions. It also makes it easy to manipulate data structures when manipulating entire populations of information. Due to these considerations, we process the most basic time representation in the form of a single hour slot being filled one or zero.

Examination & Results

We ran the program through a set of real-life course and homework sets and also through randomly generated requirements to assess the real world suitability of the system along with suitable stress testing of it.

	S	U	M	T	W	R	F
0	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
1	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
2	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
3	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
4	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
5	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
6	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
7	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
8	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
9	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
10	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
11	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
12	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
13	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
14	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
15	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep	Sleep
16							
17							
18			DesNat	ISIM		DesNat	ISIM
19			DesNat	ISIM		DesNat	ISIM
20			DesNat	ISIM		DesNat	ISIM
21			DesNat	ISIM		DesNat	ISIM
22			DesNat	ISIM		DesNat	ISIM
23	Meals	Meals	Meals	Meals	Meals	Meals	Meals
24	Meals	Meals	Meals	Meals	Meals	Meals	Meals
25	Meals	Meals	Meals	Meals	Meals	Meals	Meals
26							
27		Elias	ModSim	SoftDes	Elias	ModSim	SoftDes
28		Elias	ModSim	SoftDes	Elias	ModSim	SoftDes
29		Elias	ModSim	SoftDes	Elias	ModSim	SoftDes
30		Elias	ModSim	SoftDes	Toolbox #3	ModSim	SoftDes
31		Elias			Toolbox #3		
32		Elias	Dieter		Elias	Dieter	
33			Do your report			Dieter	
34			Do your report			Dieter	
35			Dieter			Dieter	
36			Dieter			Dieter	
37							
38							
39			Build the hopper				
40			Build the hopper				
41			Build the hopper				
42			Build the hopper				
43			Build the hopper				
44			Build the hopper		Get lab working		
45					Get lab working		
46	NightSleep	NightSleep	NightSleep	NightSleep	NightSleep	NightSleep	NightSleep
47	NightSleep	NightSleep	NightSleep	NightSleep	NightSleep	NightSleep	NightSleep

Figure 1: Example result from the scheduling program

Conclusions

We successfully implemented a homework scheduler using evolutionary techniques, and tested it across a number of input sizes

References

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