

An Evolutionary Approach to Timetabling

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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 gene pool 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.”

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.

Conclusions

References

An overview of the A* algorithm: <https://theory.stanford.edu/~amitp/GameProgramming/AStarComparison.html>. This will help build our foundational knowledge on heuristic-based path finding.

A heuristic-based approach to optimal path finding:<https://theory.stanford.edu/~amitp/GameProgramming/Heuristics.html>. We need to know about heuristics in-depth in order to figure out how we may be able to adapt it to our own algorithm.

A comprehensive overview of heuristic functions:<https://medium.com/@rinu.gour123/heuristic-search-in-artificial-intelligence-python-3087ecfece4d> Algorithms like A* do not guarantee that they'll find the globally best solution, but do guarantee that they'll find the best solution for the given heuristic functions. Therefore, resources like these will be critical in learning more about how to select better and more optimal functions.