Class Scheduling

Genetic Algorithm Problem for college timetable





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INFO 6205 Final Project Report Problem Description

When designing a timetable, we need to consider many factors at same time. In this project, we are going to solve schedule problem by using genetic algorithm. We will check different scenarios at first. Then we will build a class scheduler, which can be expanded to support more complex implementations.

As mentioned above, class scheduling is a complex case, which needs to be satisfied all constraints. Violation of any one factor, the schedule is flagged with failure. Constraints has two parts, hard constraints and soft constraints.

The hard constraints:

The classroom need to be big enough to hold the class



Classroom must contain any required equipment



Classroom can only hold one class at any given time



Professor can only be in one class at any given time



The soft constraints:

Room capacity should be suitable for the class size







Preferred classroom of the professor

Preferred class time of the professor

The hard constraints are the first priority, which need to be satisfied to produce a functional solution. Although the soft constraints are not necessary, the executor should also consider the cost. Like, if one classroom has 50 seats providing to ten-students class, it is not a good solution. Sometimes, multiple soft constraints may conflict, the algorithm should provide suitable solution for this case.

We will notice that class scheduling problem is little different from others algorithm problem. Our program will stop at the generation that are not violated with all hard constraints and are compatible with some soft constraints. It is because that our program is based the data we create. Once the schedule is suitable, the program will stop mutation and print result.

We could add more soft constraints in order to guarantee more accurate generation. More details will explain later.

Model Details Design

Scenario:

Each class will be assigned a timeslot, a professor, a room and a student group by the class scheduler. We can calculate the total number of classes that will need scheduling by summing the number of student groups multiplied by the number of modules each student group is enrolled in.

The schedule should consider following hard constraints and soft constraints.

hard constraints

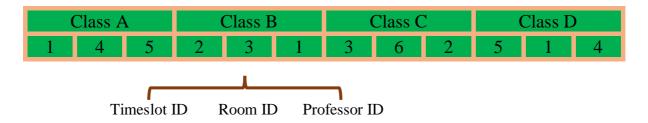
- Classroom should big enough to hold the class
- Classroom should be served one class and the class should be set in free room
- The professor can only teach one class at any given time.

soft constraints

- Preferred classroom of the professor
- Preferred class time of the professor

Chromosome Design

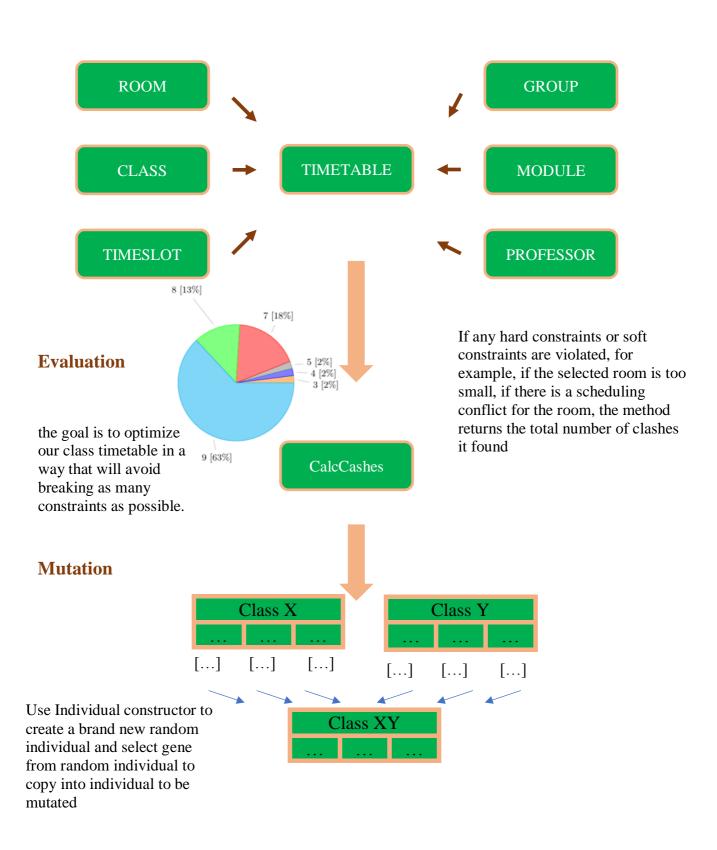
Chromosome is an important factor in the genetic algorithm where the design of chromosome will affect how the crossover implemented. We could assign several ID to each professor, classroom and timeslot. Form of gene are as below.



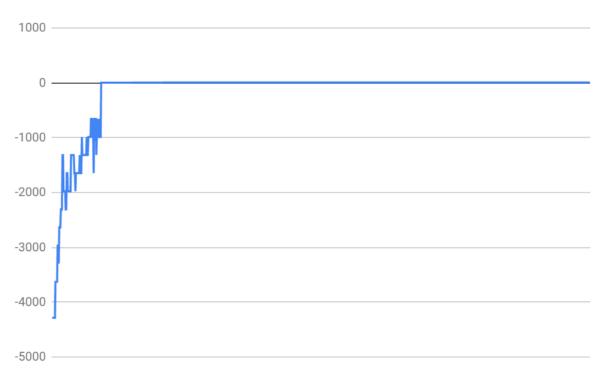
Each class has three components, 1 gene for time, 1 gene for professor, 1 gene for room.

Initialization

First of all, we have to initialize some information including class, classroom, professor, timeslot, group and module. Forms of each Java class are as below.



Result and Conclusion



(the first generation gets so bad fitness, which lead to the best generation that gets 8.83 nearly close to zero)

It can be predicted that the first generation could perform really bad because they have such a small amount of good genes generated by list of random number. Under such circumstance, the functional schedule could not make a right class timetable.

With generation increasing, the fitness gets better. The generation stops changing when the generation is 468. Unlike the others genetic algorithm problem, most solutions in this scenario are invalid and we stop only when we find the first valid solution or run out of time.

The difference between the two problems is as follows: in other genetic algorithm problem, it's easy to create a valid but in the class scheduler, creating a valid solution is the difficult part.

When considering only three hard constraints, it is easy to stop at nearly 200 generation. This is because the database we create is too small that means there is no very complex situations. It is enough for program to find valid results.

In order get more accurate generations, we introduce soft constraints, adding soft constraints to the class scheduler changes the problem significantly.

To guarantee a valid from the GA algorithm, we need to make sure that hard constraints have significant impact on the fitness. In other words, if the individual is not valid, the sum of points gain from match the soft constraints should less than the penalties of hard constraints which make sure that the final fitness less than 0. So, penalties of hard constraints fellow the change of population size. If population size scales up, the penalties goes up which still larger than the sum of rewards gained from matching the soft constraints.

Bonus: parallel processing

We implement parallel processing in calculating fitness. The fitness function is often going to be the bottleneck of genetic algorithm. Using parallel processing can calculate the fitness of numerous individuals simultaneously, which makes a huge difference when there are often hundreds of individuals to evaluate per population.

We use IntStream () to achieve parallel processing in fitness function. To ensure the objects in thread will not affect objects in another thread, we clone the timetable object before calculating the individuals' fitness.

Range from 0 to the size of population, parallel() returns an equivalent stream that is parallel. In the forEach(), we give the action-calculate fitness to it.

Evidence of Running

```
G98 Best fitness: -656.68
G99 Best fitness: -656.68
G109 Best fitness: -326.68
G101 Best fitness: -326.68
G101 Best fitness: -326.68
G102 Best fitness: -326.68
G103 Best fitness: -326.68
G104 Best fitness: -326.68
G105 Best fitness: -326.97
G107 Best fitness: -326.97
G107 Best fitness: -326.97
G108 Best fitness: 1.87
G109 Best fitness: 1.87
G110 Best fitness: 3.9
G111 Best fitness: 3.9
G112 Best fitness: 3.9
G113 Best fitness: 3.9
G114 Best fitness: 3.9
G115 Best fitness: 3.9
G116 Best fitness: 3.9
G117 Best fitness: 3.9
G118 Best fitness: 3.9
G119 Best fitness: 3.9
G110 Best fitness: 3.9
G110 Best fitness: 3.9
G111 Best fitness: 3.9
G112 Best fitness: 3.9
G112 Best fitness: 3.9
G120 Best fitness: 3.9
G121 Best fitness: 3.9
G121 Best fitness: 3.9
G122 Best fitness: 3.9
G123 Best fitness: 3.9
G124 Best fitness: 3.9
G125 Best fitness: 3.9
G126 Best fitness: 3.9
G127 Best fitness: 3.9
G128 Best fitness: 3.9
G129 Best fitness: 3.9
G130 Best fitness: 3.9
G131 Best fitness: 3.9
G132 Best fitness: 3.9
G133 Best fitness: 3.9
G134 Best fitness: 3.9
G135 Best fitness: 3.9
G137 Best fitness: 3.9
G138 Best fitness: 3.9
G139 Best fitness: 3.9
G139 Best fitness: 9.12
G999 Best fitness: 9.12
                Solution found in 1001 generations
Final solution fitness: 9.12
Classes: 912
Class 1:
Module: Computer Science
Group: 1
Room: A1
Professor: Mr A Thompson
Time: Thu 9:00 - 11:00
                Module: Maths
Group: 1
Room: B1
Professor: Mrs E Mitchell
Time: Fri 9:00 - 11:00
                        Class 3:
Module: Physics
Group: 1
Room: A1
```

Unit Tests

TestGeneExpression

TestEval

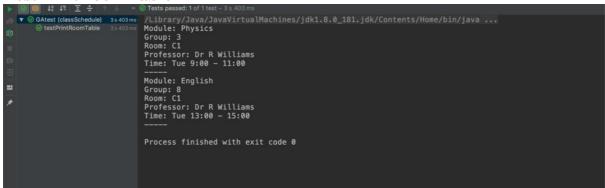
TestMutate

TestCrossover

TestPrintProfessorTable

TestPrintGroupTable

TestPrintRoomTable



TestPrintModuleTable

```
Tests passed: 1 of 1 test - 2 s 800 ms

| Contest class Schedule| 2 s 800 ms
| Contest class Schedule| 2 s 800 ms
| Contest PrintModule Table| 2 s 800 ms
| Contest P
```

