Introducing a Novel parameter in Generation of Course Timetable with Genetic Algorithm

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Motivation

Genetic Algorithm produces result which is randomly selected from the solution space. This solution is not appropriate most of the times, even though it is fit.

The Happiness parameter is introduced for the faculty and the students, to provide comfort with the timetable.

Introduction
The Timetable Problem

Definition: The task of assigning events, such as lectures, exams to a limited set of timeslots and rooms, within the limits of specified constraints.

It is an NP-Hard Problem.

Introduction
Type of University Timetable Problem

- Course Timetable Problem
  - One event per room per timeslot
  - Fixed number of timeslots

- Exam Timetable Problem:
  - Multiple events per room at one time
  - Variable timeslots

Introduction
Constraints

- Hard Constraints
  - All should be satisfied for the solution to be feasible
  - Higher Priority

- Soft Constraints
  - Maximum of the soft constraints should be obeyed
  - Lower priority than Hard Constraints

Related Work

- Bambrick (1997), presents Genetic Algorithm for timetable problem along with repair strategy. All types of feasible and non-feasible solution exists together in the solution space with fitness count of 1.0.
- Abdullah and Turabieh (2008), local search along with genetic Algorithm is used for optimization. An additional computational cost is incurred using local search and it is performed on final solution which limits the final solution
- Yang and Hat (2011), used an additional data structure for storing information about events with zero penalty value

Problem Description
The University Course Timetable

- The course timetable of Department of Computer Engineering, G.B. Pant University of Agriculture and Technology, Pantnagar was used for generating the timetable.

- A course is represented as Course Number, followed by Lectures, Tutorials and Laboratories such as Operating System with course number TCT324 having total 3 credits is represented as TCT324(2-1-1). Tutorials are not included in the total number of credit hours.

Problem Description
Constraints

The Hard Constraints are:

1. A classroom must not be booked twice at one time.
2. Each event must be allotted exactly once.
3. Students must not have multiple classes simultaneously.
4. The classroom must have enough capacity to hold the allotted group of students.
5. Instructors must not be booked twice for one particular timeslot.
6. An instructor must not be allotted a lecture, laboratory or tutorial when he/she is unavailable. E.g. an instructor might have prior commitments.

The soft constraints have been exempted from consideration since they vary greatly with respect to departments and universities.

Problem Description
Sample Data

S. No.	Abbrv.	Course Name	Credits	Year
1	DS	Data Structure	3(3-1-0)	II
2	DSS	Discrete Structure	3(1-0-0)	II
3	FLAT	Finite Lang. and Automata Theory	3(3-1-0)	II
4	FCCS	Fund. of Computer Comm. Systems	4(3-1-1)	II
5	PL	Programming Language	3(3-1-0)	II
6	SYP	System Programming	4(3-0-1)	III
7	DBMS	Database Management System	4(3-0-1)	III
8	ORS	Operational Research	2(2-1-0)	III
9	MIS	Management Information System	3(2-0-1)	III
10	CA	Computer Architecture	3(2-1-1)	IV
11	DMW	Data Mining and Warehousing	3(2-1-1)	IV
12	CGA	Computer Graphics and Animations	4(3-0-1)	IV

Genetic Algorithm

- Derived from process of evolution in nature as stated by Charles Darwin
- Each individual is represented as Chromosome
- A group of Chromosomes form a population.
- These chromosomes undergo evolution process to produce new offsprings (individuals).

Genetic Algorithm Evolutionary Process

- Initially, chromosomes are randomly generated equal to the population size
- The chromosomes with maximum fitness are used for breeding
- The chromosomes produced after breeding are named as Generation
- The produced offspring are evaluated for fitness value
- The individual with maximum fitness value is added into the population and the one with least fitness value is discarded from the population, thus keeping constant population size.
- This process is repeated depending upon the requirement.

Genetic Algorithm Operators

Crossover

Parent 1: 1010100001111 0001100110000110

Parent 2: 1010111011101 1000011101011010

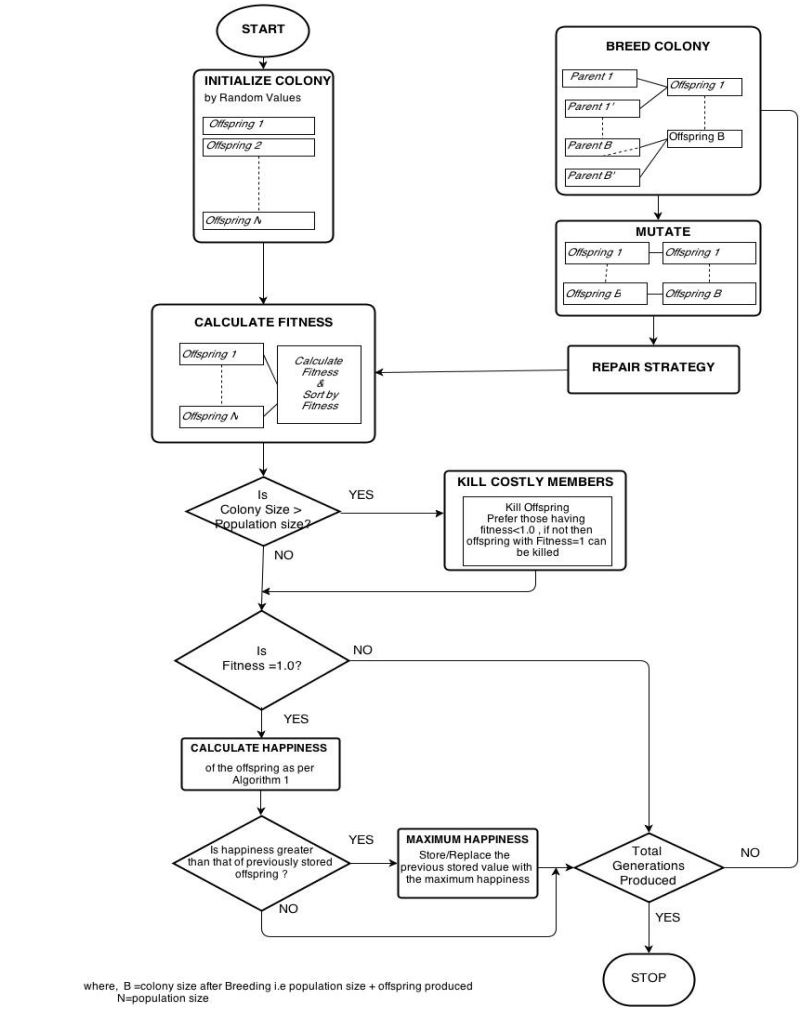
Offspring: 1010100001111 1000011101011010

Mutation

Before Mutation: 1010111011011100101 01

After Mutation: 1010111001011100101 01

Proposed Methodology



Proposed Methodology
The Happiness Parameter

- The Happiness parameter calculated happiness of the entire table
- The Happiness value is calculated after implementation of Genetic Algorithm

The Happiness Parameter:

It is the sum of free timeslots an instructor has after which no lectures are scheduled for that particular day, for all days of the week. A penalty is added for lectures at odd times like 08:00-09:00 (early morning) and/or 14:00-15:00 (immediately after lunch)

Happiness Parameter Formula

$$Happiness = \sum_{i=0}^{n} \{(Total\ Time\ slots(i.e.8) - Position\ of\ last\ lecture\ for\ i\ day) - p\}$$

Where,

i = Number of Working days {Mon, Tue, Wed, Thu, Fri, Sat}, and

$$\boldsymbol{Penalty\ ,p =} \begin{cases} \boldsymbol{-1,\ If\ a\ lecture\ is\ scheduled\ at\ 08{:}00 - 09{:}00\ OR\ 14{:}00 - 15{:}00} \\ \boldsymbol{-2,\ If\ a\ lecture\ is\ scheduled\ at\ 08{:}00 - 09{:}00\ AND\ 14{:}00 - 15{:}00} \end{cases}$$

Happiness Parameter Algorithm

Input: Timetable
Output: Happiness parameter value

```
var day = {Mon, Tue, Wed, Thu, Fri, Sat}
var timeslot = {08:00-09:00, 09:00-10:00, 10:00-11:00, 11:00-12:00, 12:00-13:00, 13:00-14:00,
                14:00-15:00, 15:00-16:00, 16:00-17:00}
var Happiness = 0;
var free_timeslots = 0;
var last_allocated_timeslot = 0;

for days = 0 to 5 do
        Calculate last_allocated_timeslot;
        Happiness = Happiness + free_timeslots;

        for timeslot = 0 to 8 do
                if Lecture allocated at timeslot = 08:00-09:00 OR 14:00-15:00
                        then
                        Happiness = Happiness - 1;
                end
                else If Lecture allocated at timeslot = 08:00-09:00 AND 14:00-15:00
                        then
                        Happiness = Happiness - 2;
                end
        end
return Happiness;
end
```

Result and Discussion
Specification for GA

S. No.	Parameters	Quantity
1	Number of Generations	20,000
2	Population Size	50
3	Number of Working Days	6
4	Number of Timeslots each day	8

The simulation was performed using Intel Core i3 processor 2.4Ghz on Windows 7 Operating System.

Result and Discussion

S. No.	G_0	$G(H_{neg})$	H_{max}	$G(H_{max})$	H_{min}	$G(H_{min})$	H_{avg}
1	14288	13566	5	2	-12	19	-4.608133
2	14317	13488	6	4	-13	6	-4.650268
3	14228	13300	6	1	-14	1	-4.556508
4	14190	13476	6	8	-12	5	-4.834179
5	14254	13514	6	11	-13	4	-4.687877
6	14427	13537	6	10	-13	17	-4.635198
7	14300	13192	7	2	-13	2	-4.419161
8	14422	13461	8	1	-12	6	-4.498613
9	14290	13298	5	9	-12	14	-4.348775
10	14158	12842	6	15	-13	3	-4.323138

Where,

G_0 = Generations with Fitness = 1.0,

$G(H_{max})$ = Generations with Maximum Happiness,

$G(H_{min})$ = Generations with Minimum Happiness

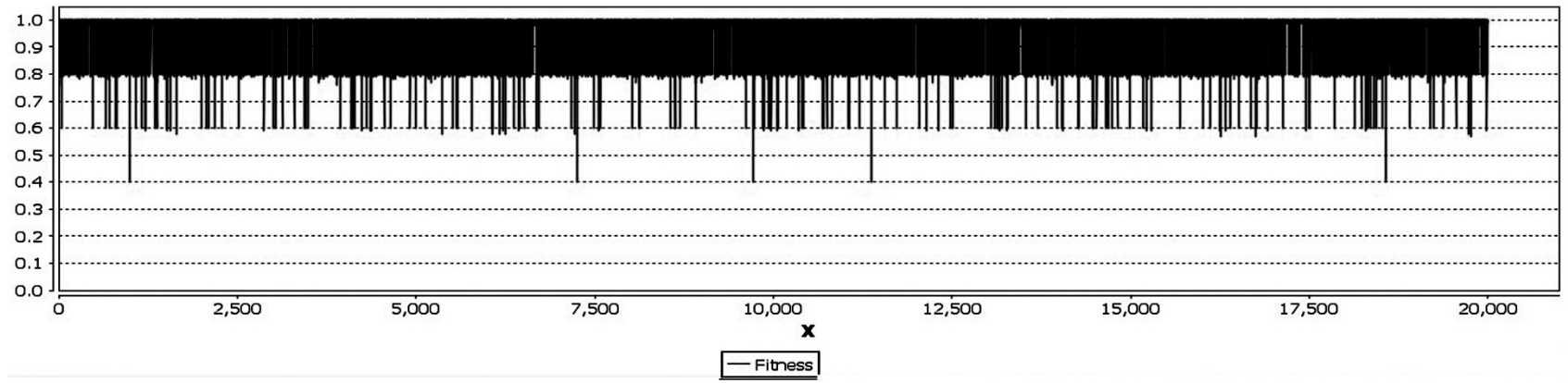
$G(H_{neg})$ = Generations with negative Happiness,

H_{max} = Maximum Happiness

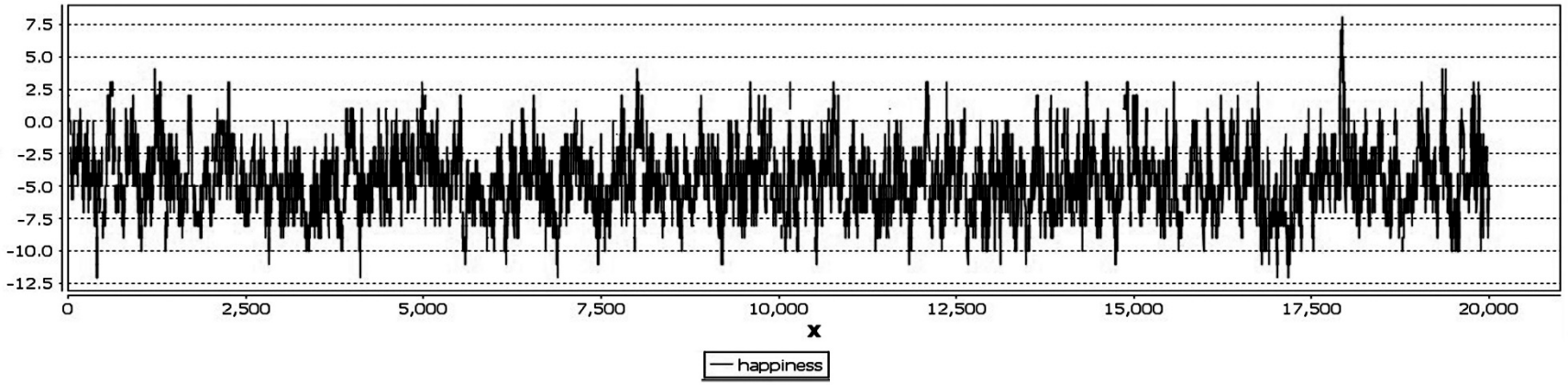
H_{min} = Minimum Happiness

H_{avg} = Average Happiness

Result and Discussion



Simulation Graph with Fitness = 1.0



Simulation Graph with Maximum Happiness = 8.0

Result and Discussion

$$Probability\ of\ selecting\ solution\ with\ negative\ Happiness = \frac{Total\ number\ of\ Generations\ with\ negative\ Happiness}{Total\ Number\ of\ Generations\ having\ Fitness}$$

S. No.	G_0	$G(H_{neg})$	Probability of selecting offspring with H_{neg}
1	14288	13566	0.94946
2	14317	13488	0.93855
3	14228	13300	0.93477
4	14190	13476	0.94968
5	14254	13514	0.94808
6	14427	13537	0.93831
7	14300	13192	0.92251
8	14422	13461	0.93336
9	14290	13298	0.93058
10	14158	12842	0.90704

Result and Discussion

- The Tab. 4 represents the probability of selection of a timetable with negative happiness for all the iterations done in Tab. 3.

- The values in Tab. 4 clearly depicts that there are more than 90% (approx.) chances of selection of non-appropriate timetable without using Happiness parameter.

Conclusion

- Adding Happiness parameter with the Genetic Algorithm narrows the solution space with only appropriately feasible solutions only.

- Thus, using this happiness parameter the non appropriate solutions can be discarded from consideration.

Future Work

- Happiness for Individual Instructors
- Versatility of Happiness Parameter
- Using Happiness parameter on Parallel systems

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Thanks