

#### Institute Timetable Automation

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## Agenda



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- 2. Project Challenges
- 3. Synopsis of Prior Works
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### About the Project



This project aims to develop an automated solution where faculty and academic section can enter the data about the courses offered in a semester, the list of faculties offering courses, the slot system, and the details of available rooms.

It should also provide a workflow that allows the academic section to create an optimal timetable (room-slot-course allocation) by considering the various constraints.

### Project Challenges



- 1. Slot system allocation
- 2. Room allocation
- 3. Minimal clashes
- 4. Common courses allocation
- 5. Multiple campuses
- 6. Credit-based system compatibility

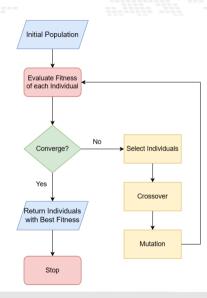
### Synopsis of Prior Works



- Our first attempt was to create a slot system as well along with the timetable to minimize the clashes
- For this we opted for a combination of brute force and greedy approaches and successfully implemented this approach
- We also constructed a basic web interface for quicker and faster visualization of the results of our algorithm
- ▶ But based on several reviews, it was concluded that having a fixed slot system is more convenient for both the students and the faculty
- ▶ Also, our approach was a complete brute force approach, so we need some better algorithms to solve all the problem constraints in an efficient manner

# Genetic Algorithm Algorithm Flowchart





# Genetic Algorithm (Contd.) Using Genetic Algorithm in Timetabling



#### Randomized Approach: Perform each step randomly with some associated probabilities

- Individual Representation: List of classes also called a Schedule
- Class: Contains a course and randomly allocated slots and rooms
- ▶ Fitness Function: Inversely proportional to the number of conflicts
- Selection Procedure: Using tournament selection with some elite schedules
- ► Crossover: Randomly mixing the classes from two schedules to create a new schedule
- ► Mutation: Randomly pick classes for an existing schedule from a newly created schedule

# Genetic Algorithm (Contd.) Using Genetic Algorithm in Timetabling

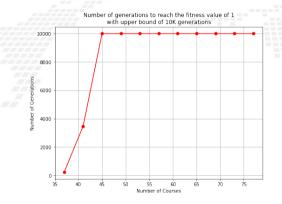


**Greedy and Maximal Approach**: Perform each step greedily to always have a maximal allocation of the list of classes

- Individual Representation: List of maximally allocated classes and unallocated classes, also called a Schedule
- ▶ Class: Contains a course and greedily allocated slots and rooms
- ► Fitness Function: Ratio of the number of allocated courses and the total number of courses
- ▶ Selection Procedure: Using tournament selection with some elite schedules
- ► Crossover: Take different permutations of the sequence of classes from the two parents, create their maximal allocation and pick the one with the maximum fitness
- ▶ Mutation: Unallocating an allocated class and trying to allocate an unallocated class such that the allocation is always maximal

#### Observations





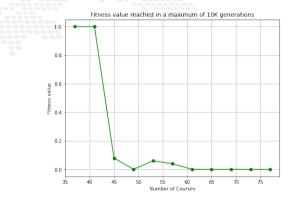


Figure: Results for 10K Generations

# Simulated Annealing Hill Climbing



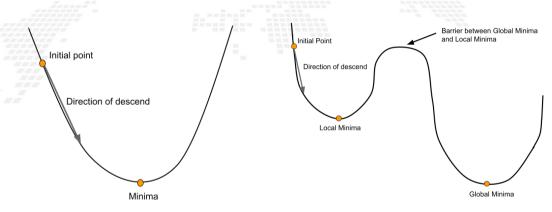
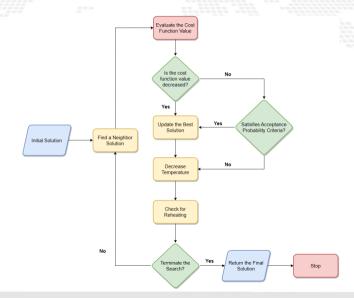


Figure: Successful and Unsuccessful Hill Climbing

#### Algorithm Flowchart





Using SA in Timetabling



- Cost Function The cost function is the sum of all the conflicts. Hard
   constraints are multiplied with a larger multiplier as compared to soft constraints
- Neighbor Function Swapping Neighbourhood one of the most used neighborhood searching algorithms in our implementation
   Other neighbor functions include simple searching and a mix of simple searching and swapping neighborhood algorithm

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Using SA in Timetabling

► Temperature Function - The temperature function which we choose in our implementation is

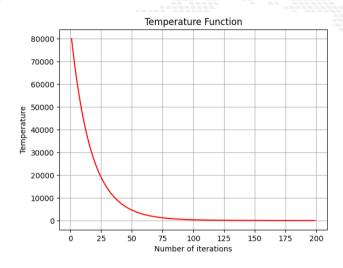
$$t = t \cdot (1 - \frac{\ln t - \ln FT}{NMOVE})$$

Acceptance Probability - The acceptance probability we used is widely used in most implementations of simulated annealing -

$$P(\textit{oldState}, \textit{ newState}, \textit{ temp}) = egin{cases} 1, & \text{if } \textit{newState} < \textit{oldState} \\ e^{\frac{\textit{oldState} - \textit{newState}}{\textit{temp}}}, & \text{otherwise} \end{cases}$$

#### Temperature Function





### Greedy Initialization



- Course Selection Heuristic Choose a course with the smallest value of  $\frac{apd_i(M)}{\sqrt{nl_i(\widetilde{M})}}$   $apd_i(\widetilde{M})$  Number of available slots in the timetable  $\widetilde{M}$  for course  $c_i$   $nl_i(\widetilde{M})$  Number of lectures of course  $c_i$  yet to be assigned in  $\widetilde{M}$
- ▶ Slot-Room Selection Heuristic After choosing a course we pick a slot-room pair for it. The pair with the smallest value of the below function is chosen -

$$k1 \cdot uac_{i,j}(\widetilde{M}) + k1 \cdot \gamma + k2 \cdot \Delta h$$

#### Results



No.of courses	Hard Conflicts	Hard Conflicts	Soft Conflicts
No.or courses	Genetic Algorithm	Simulated Annealing	Simulated Annealing
37 (random data)	0	0	0
45 (random data)	13	0	0
57 (random data)	43	0	2
69 (random data)	90	1	3
97 (random data)	141	3	16
94 (real data)	222	11	62

Table: Comparison between Genetic Algorithm and Simulated Annealing used on timetable initialized by Genetic Algorithm

# Results (Contd.)



No.of courses	Hard Conflicts	Soft Conflicts	Hard Conflicts	Soft Conflicts
	Greedy Algo.	Greedy Algo.	SA	SA
37 (random data)	0	0	0	0
45 (random data)	0	0	0	0
57 (random data)	0	0	0	0
69 (random data)	0	0	0	0
97 (random data)	6	33	0	39
94 (real data)	18	34	8	33

Table: Comparison between Greedy Algorithm and Simulated Annealing used on timetable initialized by Greedy Algorithm



# Demo Time



# Thank You!