

## University Timetabling Problems

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### Literature

1. "Recent Research Directions in Automated Timetabling", Burke, E.K., Petrovic, S., accepted for publication in *European Journal of Operational Research - EJOR*, 2002.
2. "A Memetic Algorithm for University Exam Timetabling", Burke, E.K., Newall, J.P., Weare, R.F., 1996b. In: (Eds.) Burke, E., Ross, P. *The Practice and Theory of Automated Timetabling: Selected Papers from the 1st Int'l Conf. on the Practice and Theory of Automated Timetabling*, Napier University, August/September 1995, *Springer Lecture Notes in Computer Science Series*, Vol. 1153., pp. 241-250.

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### Highly recommended additional reading:

"A Survey of Practical Applications of Examination Timetabling Algorithms", Carter, M.W., *Operations Research* Vol.34, 1986, pages 193-202.

"A Multi-Stage Evolutionary Algorithm for the Timetable Problem, Burke, E.K., Newall, J.P., *IEEE Transactions on Evolutionary Computation* Vol.3 No.1, 1999, pages 63-74.

"Automated University Timetabling: The State of the Art", Burke, E., Kingston, J., Jackson, K., Weare, R., *The Computer Journal* Vol.40 No9, 1997, pages 565-571.

"Examination Timetabling: Algorithmic Strategies and Applications", Carter M.W., Laporte G., Lee, S.Y., *Journal of Operational Research Society* Vol. 74, 1996, pages 373-383

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### Statement of Timetabling Problems

**Timetabling problem:** assign a number of events into a limited number of time periods.

#### • **Course timetabling**

Assign each lecture to some period of the week in such a way that no student is required to take more than one lecture at a time

#### • **Examination timetabling**

Assign a number of exams into a limited number of time slots in such a way that no student is required to take more than one exam at a time

#### Graph representation

Vertices: courses / exams

Edges: if two courses/exams have at least one student in common

### Hard constraints:

- no resource can be demanded to be in more than one place at any time
- the total resources at any time must not exceed the resources available

### Soft constraints:

- time assignment
- time constraints between events
- spreading events out in time
- coherence
- resource assignment

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### Approaches to Timetabling problems

#### **1. Sequential Methods**

##### Graph Colouring Heuristics

- **Largest degree first** orders the events by their number of conflicts with other events.
- **Largest weighted degree** weights each conflict by the number of students involved in the conflict.
- **Saturation degree** selects an event which has the smallest number of valid periods available for scheduling.
- **Colour degree** prioritises those events that have the largest number of conflicts with the events that have already been scheduled.

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How to select a valid period for each event?

- earliest valid period
- "best" valid period in the context of objective functions

Example of a cost function for an exam timetabling:

$$cost(t) = 5000 * unscheduled(t) + 3 * sameDay(t) + overNight(t)$$

*unscheduled(t)* holds the number of exams not scheduled in a valid period

*sameDay(t)* present the number of conflicts where students have exams in adjacent periods on the same day

*overNight(t)* present the number of conflicts where students have exams in overnight adjacent periods

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☛ How to add random element in the search?

- Tournament selection

1. Generates a subset of events randomly
2. Selects the event that is ranked the first with respect to the heuristics.

- Bias selection

An event is randomly selected from the list of the first  $n$  events.

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## 2. Cluster Methods

1. The set of events is split into groups which satisfy hard constraints.
2. The groups are assigned to time periods to fulfil the soft constraints.

## 3. Constraint Based Approaches

A timetabling problem is modelled as a set of variables (i.e. events) to which values (i.e. resources such as rooms and time periods) have to be assigned to satisfy a number of constraints.

## 4. Meta-Heuristic Methods

Simulated annealing, Tabu search, Genetic algorithms

## 5. Multicriteria Approach

Criteria express the measures of the violations of the soft constraints.

## 6. Case-Based Reasoning in Timetabling

The previously solved timetabling problems are used in solving new timetabling problems.

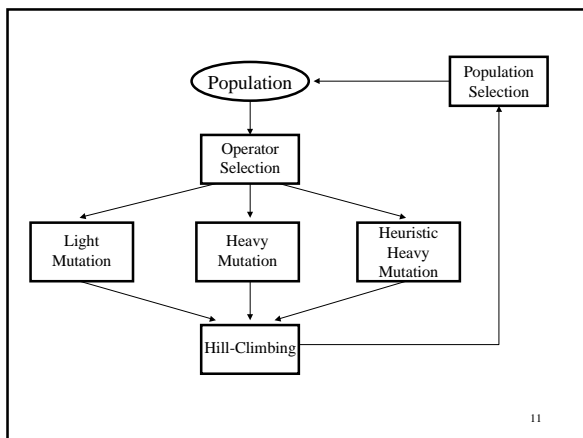
## Some Innovations in Meta-Heuristic Methods for Timetabling

Genetic algorithm + Local search + Heuristics



**Memetic algorithm**

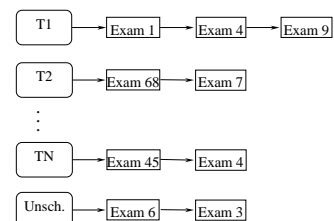
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## Representation of the solution

Member of the population



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### Evaluation function

$$cost(t) = 5000 * unscheduled(t) + 3 * sameDay(t) + overNight(t)$$

*unscheduled(t)* holds the number of exams not scheduled in a valid period

*sameDay(t)* presents the number of conflicts where students have exams in adjacent periods on the same day and, respectively.

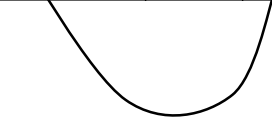
*overNight(t)* presents the number of conflicts where students have exams in overnight adjacent periods

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### Evolutionary Operators

#### • Light mutation operator

Period 1	Period 2	...	Period N-1	Period N
OR	Probability		Discr. Math.	Algebra
Statistics	...		Geometry	Data struct.
Programm.		...	AI	...
...				



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#### • Heavy mutation operator

Period 1	Period 2	...	Period N-1	Period N
OR	Probability		Discr. Math.	Algebra
Statistics	...		Geometry	Data struct.
Programm.		...	AI	...
...				

rescheduling of time periods

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#### • Heuristic Heavy Mutation

Period 1	Period 2	...	Period N-1	Period N
OR	Probability		Discr. Math.	Algebra
Statistics	...		Geometry	Data struct.
Programm.		...	AI	...
...				



heuristic rescheduling of events

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### Hill-climbing

- Direct the search toward the local optima.
- The hill-climbing operator takes exams sequentially from periods that are randomly ordered, and reschedules them in valid periods that cause the least penalties.

#### Selection phase

- Members of the population are ordered by the values of their evaluation functions.
- According to a certain probability it is decided whether to leave the solution for the next generation or not.

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### Decomposition

Period 1	Period 2	Period 3	...
...	...	...	...



Phase 1 Phase 2

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### **Heuristic Initialisation**

- Gives initial solutions of higher quality than randomly generated initial solutions.
- Diversity is desirable in evolutionary algorithms.
- Measures of diversity of the initial solutions are defined.

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### **Summary**

- Various operators that utilise heuristics are used to improve and accelerate the search process within the evolutionary method.
- Algorithms for timetabling might be more efficient if they consider subsets of events, rather than the whole set of events.
- Employment of heuristic initialisation strategies can provide significant benefits in terms of the quality and diversity of the members of the population.

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