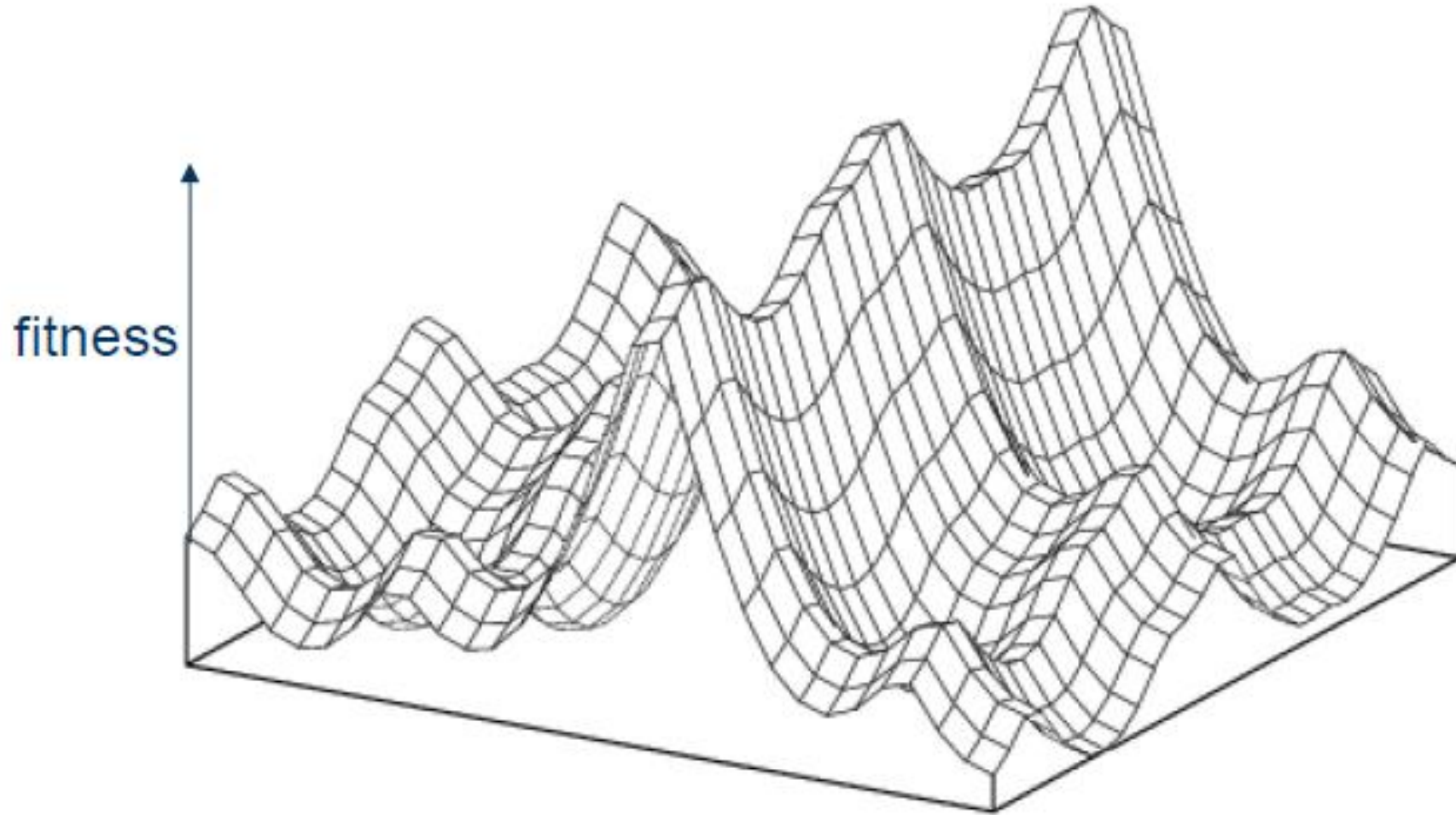




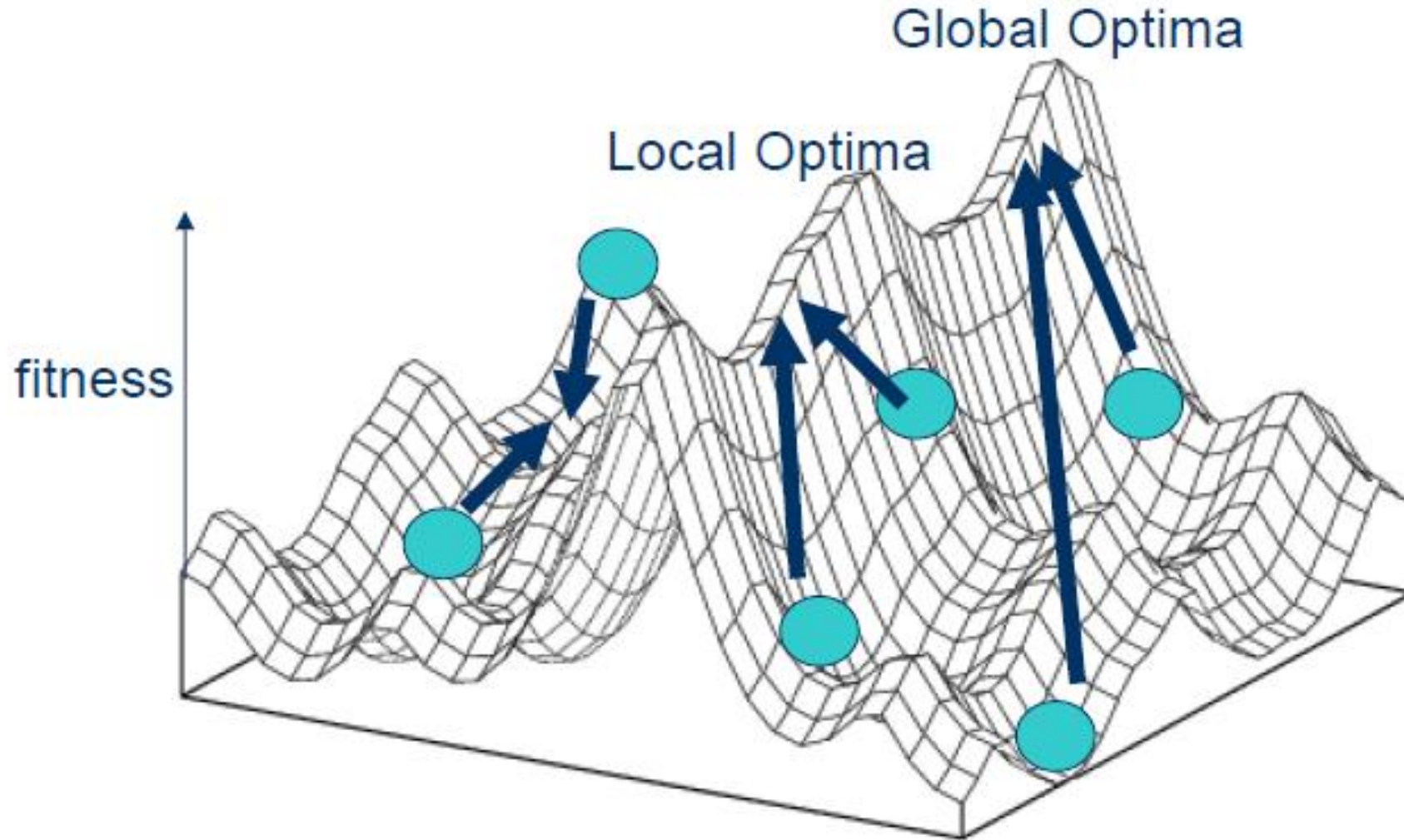
Metaheuristic Optimization and Local Search

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1/13/2026

Combinatorial Problems: Fitness Landscape

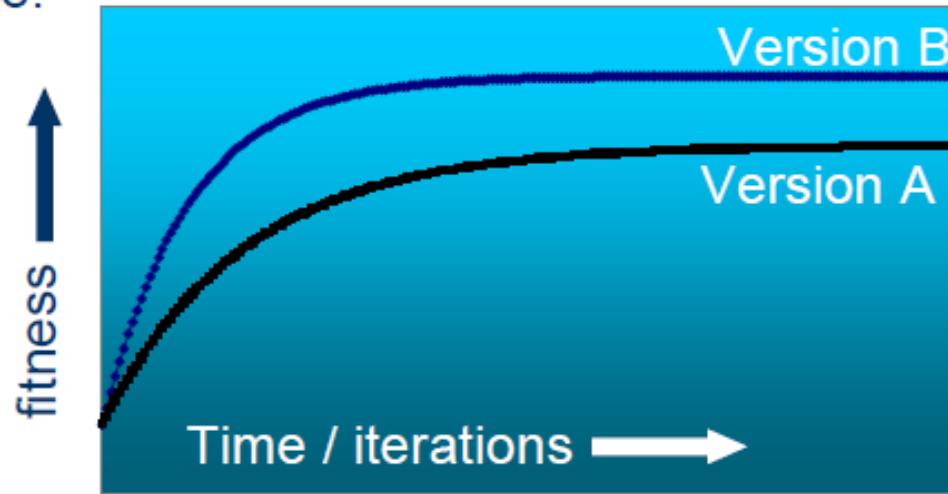


Combinatorial Problems: Fitness Landscape



Typical Behavior of an Evolutionary Algorithm

- As the population evolves, the quality of the solutions in the population tends to increase.



- Typically, the performance of the EA will be affected by choice of:
 - Parameter Settings (Population size, mutation rate, etc.)
 - Types of operators, population policies used etc.Unfortunately, this is somewhat of a “black art”

Ant Colony Optimization

- Another important metaheuristic

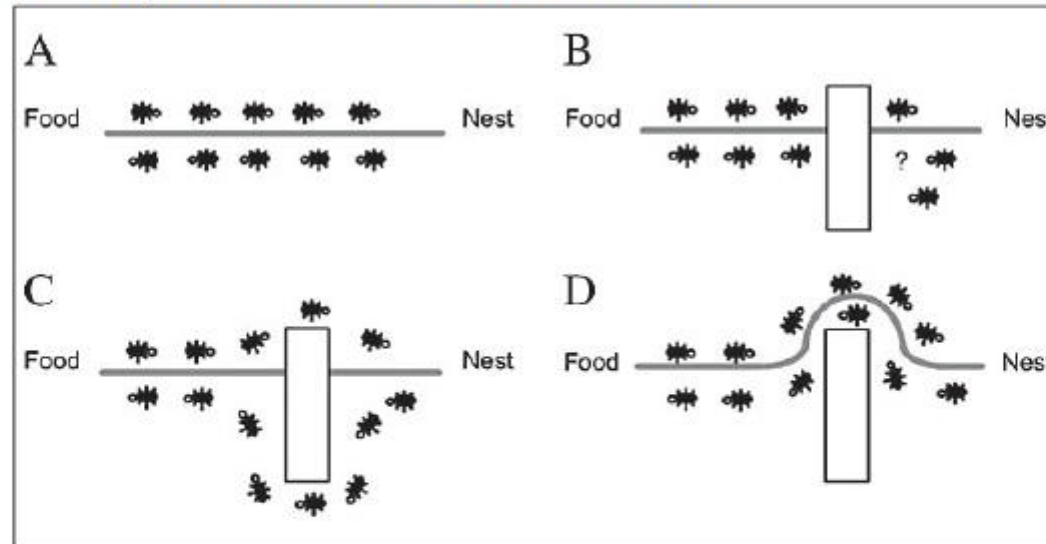


Figure 2. A. Ants in a pheromone trail between nest and food; B. an obstacle interrupts the trail; C. ants find two paths to go around the obstacle; D. a new pheromone trail is formed along the shorter path.

- Like evolutionary algorithms, ACO is applicable to a wide range of problems

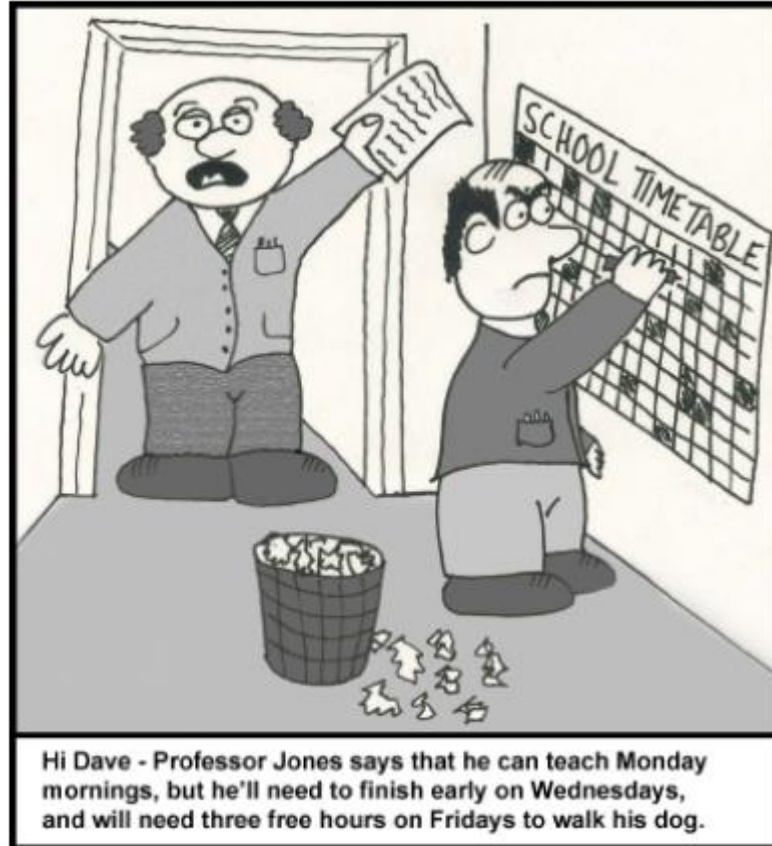
History of Metaheuristics

- 1965: first evolution strategy (Hans-Paul Schwefel)
- 1975: first genetic algorithm(s) (Turing -1954, Alex Fraser)
- 1983: simulated annealing (Pincus 1970 & many others independently)
- 1986: tabu search (Fred Glover)
 - Takes potential soln & checks immediate neighbors (differ by minor details)
 - Worsening moves accepted, “prohibitions” prevent revisiting old soln’s
 - If soln w/in short period of time or violates rule, marked as “tabu”
- 1991: ant colony optimization (Pierre-Paul Grasse)
- 1997: variable neighborhood search (Mladenovic & Hasen)
 - Descent to local optimum, then perturb to get out of valleys
- 2000+: parallel and distributed computing in metaheuristics

Questions in Metaheuristic Research

- Research questions include:
 - What quality of solutions can we expect from our algorithm?
 - How fast is the algorithm?
 - How do the solutions / run-times compare to other methods?
 - How robust/reliable is the algorithm?
 - Is the algorithm more reliable with certain types of problem instances (e.g. those of a certain size)?
- Such questions are usually answered empirically

Another Application: University Timetabling



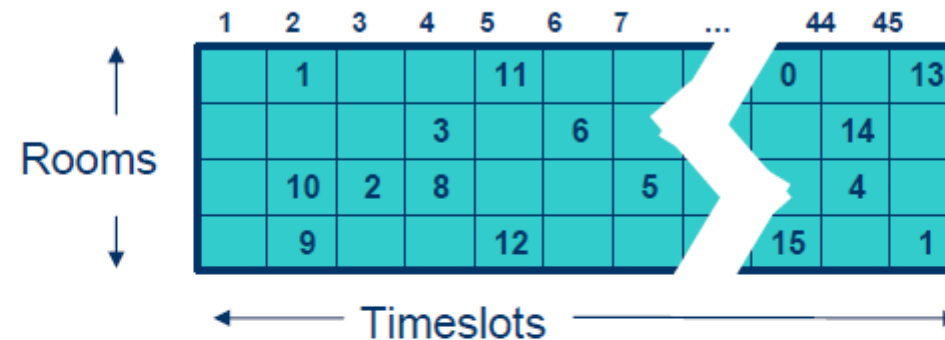
- A problem common to all universities
- Assign “events” to timeslots and rooms while obeying various constraints
- Typically constraints for this problem are idiosyncratic (every university is different)
- Research in the field typically disconnected

International Timetabling Competition (ITC)

- www.cs.qub.ac.uk/itc2007/
- Run between August '07 and January '08
- **Idea:** Design the best algorithm for a number of benchmark problem instances.
- Three competition tracks – exam timetabling, curriculum-based timetabling, and post enrolment-based timetabling
- Any type of algorithm was permitted, including commercial software.
- A strict run time limit imposed (approx 5 min. depending on machine and platform)
- Performance judged on solution quality *at the time limit*
- Algorithms were ranked against one another, and performance was verified on the organisers computers

Timetabling Competition: *ITC2007*

- Example: Post enrollment-based course timetabling (track 2)



- Assign each event to a room and timeslot such that:
 - No student or room is double-booked
 - Precedence constraints are obeyed
 - All events occur in *suitable* rooms
- Soft Constraints are also considered, such as:
 - Students should not have to sit three lectures in a row
 - Students should not have a lecture in the 5pm timeslot
 - Students should not have just one lecture in a day

Timetabling Competition: *ITC2007*

- Over 40 entrants from across the globe
- All finalists in each track used metaheuristic-based approaches

Results of Track 2

Rank	Entrants	Affiliation
(1)	Hadrien Cambazard, Emmanue Hebrard, Barry O'Sullivan, and Alexandre Papadopoulos	Cork Constraint Computation Centre, Ireland
(2)	Mitsunori Atsuta, Koji Nonobe, and Toshihide Ibaraki	Kwansei-Gakuin University, Japan
(3)	Marco Chiarandini, Chris Fawcett, and Holger Hoos	University of Southern Denmark
(4)	Clemens Nothegger, Alfred Mayer, Andreas Chwatal, and Gunther Raidl	Vienna University of Technology, Austria
(5)	Tomas Müller	Purdue University, USA

The Metaheuristic “Toolbox”

- Metaheuristics an effective tool in our armoury against intractable problems
- General algorithmic frameworks applicable to a wide range of problem types
- However:
 - There is no “one-size fits all” policy, different approaches *seem* to work well with different problems
 - Development times are often high
 - Theoretical studies are difficult. Algorithm design is often considered an art, and analysis is usually empirical
 - Difficult to state bounds on solution quality

Conclusions

- Metaheuristics an effective tool in our armoury against intractable problems.
 - General algorithmic frameworks applicable to a wide range of problem types.
 - However:
 - There is no “one-size-fits-all” policy, different approaches seem to work well with different problems.
 - Development times are often high
 - Theoretical studies are difficult. Algorithm design is often considered an art, and analysis is usually empirical
 - Difficult to state bounds on solution quality
- There is NO
free lunch...
(NFLT)!!!!!!**

Questions?

