



# Artificial Intelligence Methods (G52AIM)

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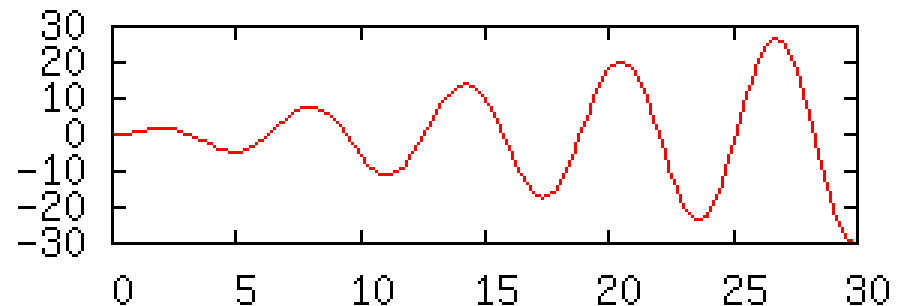
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***Local Search***

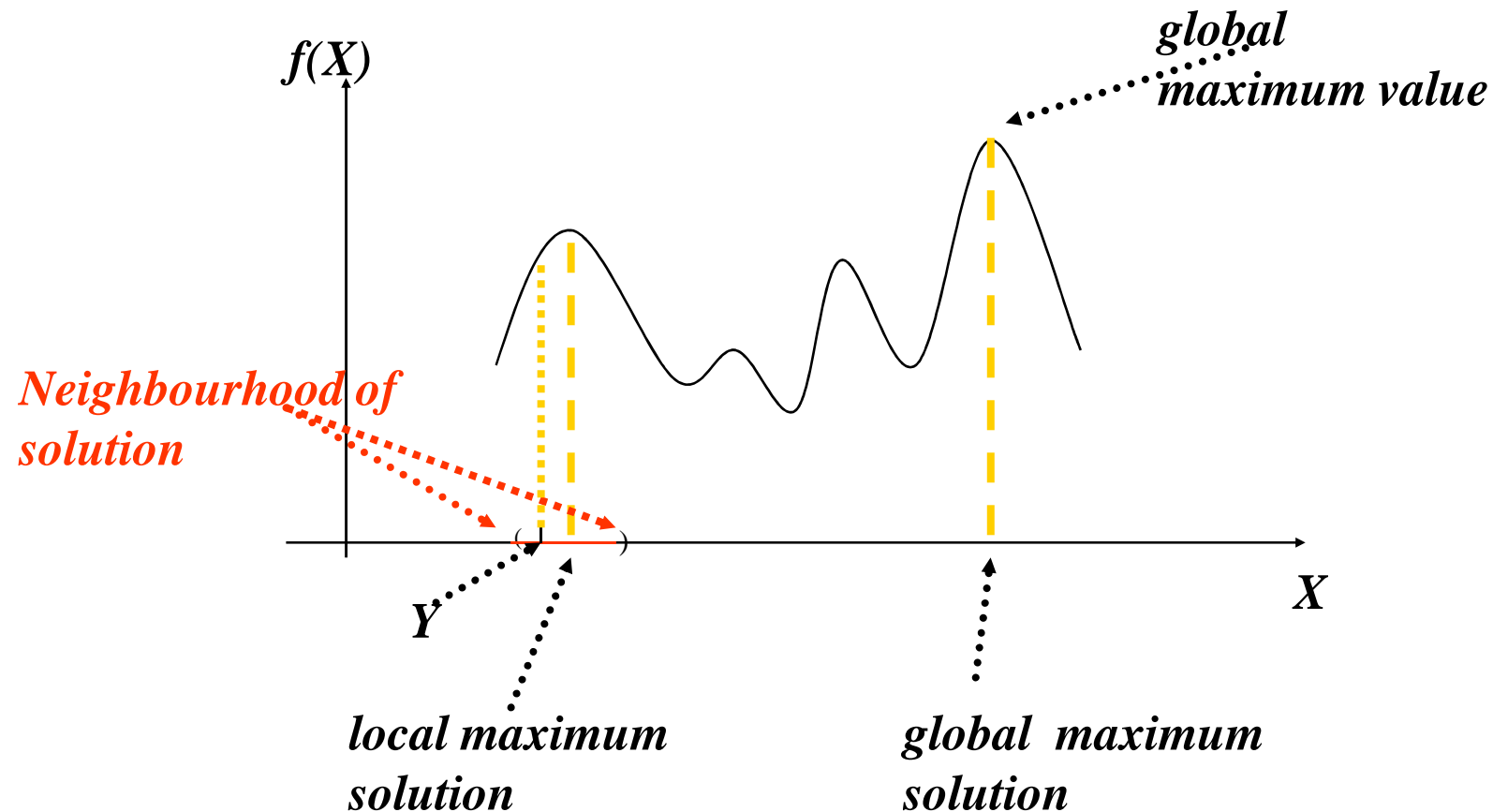
# Optimisation Problems: Definition

- Find values of a given set of decision variables:  $X=(x_1, x_2, \dots, x_n)$  which maximises (or minimises) the value of an objective function:  $x_0 = f(x_1, x_2, \dots, x_n)$ , subject to a set of constraints
- Any vector  $X$ , which satisfies the constraints is called a **feasible solution** and among them, the one which maximise (or minimise) the objective function is called the **optimal solution**

Maximise:  $x \sin(x)$   
subject to  $0 \leq x \leq 30$

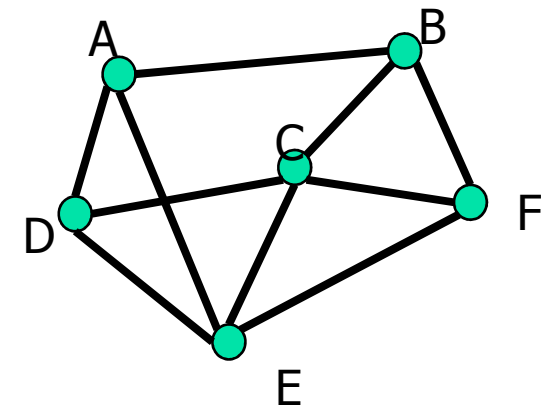
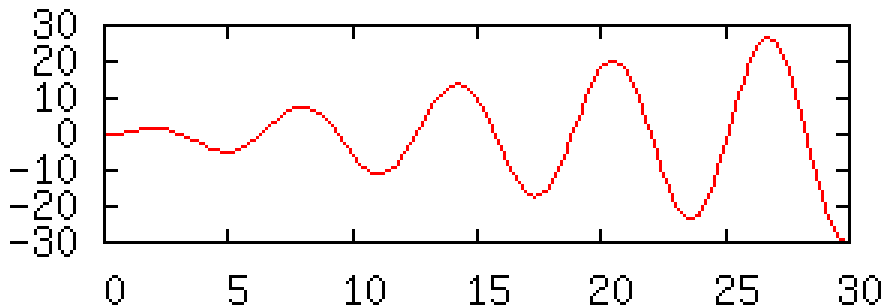


# Optimisation Problems: terminology



# Optimisation Problems: Difficulties

- For most of real world problems
  - An exact model cannot be built easily;
  - Number of feasible solutions grow exponentially with growth in the size of the problem.





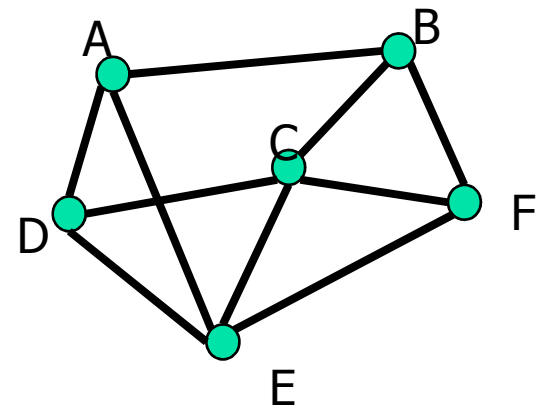
# Methods of optimisation

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- Mathematical optimisation
  - Based on Mathematical techniques to solve the optimisation problem exactly or approximately with guarantee for quality of the solution;
  - Examples:
    - Simplex method: by far the worlds most widely used optimization algorithm!
    - Lagrange multipliers, branch and bound, cutting planes, interior point methods, etc;

# Methods of optimisation

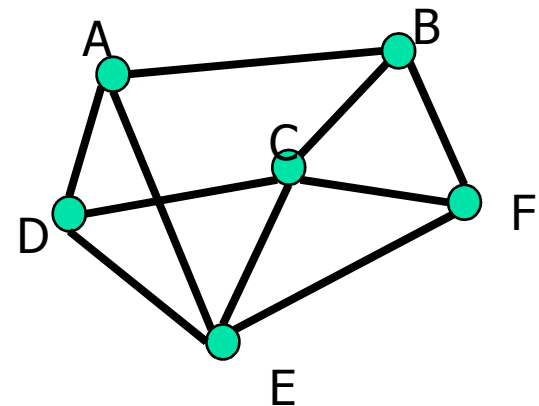
- Mathematical optimisation
  - +: Guarantee of optimality
  - - : Unable to solve larger instances of difficult problems due to large amount of computational time and memory needed;



# Methods of optimisation

- Constructive Heuristics

- Using simple minded greedy functions to evaluate different options (choices) to build a reasonable solution iteratively (one element at a time);
- Examples: Dijkstra method, Big M, Two phase method, Density constructive methods for clustering problems, etc;





# Methods of optimisation

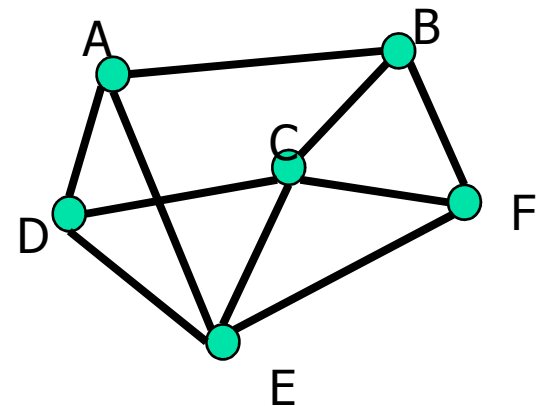
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- Constructive Heuristics
  - +: Ease of implementation;
  - -: Poor quality of solution;
  - -: Problem specific.



# Methods of optimisation (cont.)

- Local Search algorithms
  - A neighbourhood search or so called local search method starts from some **initial (complete) solution** and moves to a better neighbouring solution until it arrives at a **local optimum**, one that does not have a better neighbour.
  - Examples: *k-opt* for TSP, etc;





# Methods of optimisation (cont.)

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- Local Search algorithms
  - +: Ease of implementation;
  - +: Guarantee of local optimality usually in small computational time;
  - +: No need for exact model of the problem;
  - -: Poor quality of solution due to getting stuck in poor local optima;



# Methods in G52AIM

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- Meta-heuristics

- These algorithms guide an underlying heuristic / local search to escape from being trapped in a local optima and to explore better areas of the solution space;



# Methods in G52AIM

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- Meta-heuristics
  - Local search based approaches
    - Hill climbing
      - “Run uphill/downhill and hope you find the top/bottom”
    - Simulated annealing
      - “Shake it up a lot and then slowly let it settle”
    - Tabu search
      - “Don’t look under the same lamp-post twice”
    - GRASP, VNS, etc



# Methods in G52AIM

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- Meta-heuristics
  - Population based approach
    - Genetic algorithms
      - “survival of the fittest”
    - Ant algorithms
      - “wander around a lot and leave a trail”
    - Genetic programming
      - Learn to program
    - Evolutionary algorithms, etc



# Methods in G52AIM

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- Meta-heuristics

- +: Able to cope with inaccuracies of data and model, large sizes of the problem and real-time problem solving;
- +: Including mechanisms to escape from local optima of their embedded local search algorithms,
- +: Ease of implementation;
- +: No need for exact model of the problem;
- -: Usually no guarantee of optimality.



# Problems in G52AIM

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- Problem domain: optimiation
  - Quality of solutions: given by evaluation function (objective function, fitness, etc)
  - Aim: minimize or maximize this objective



# Local Search Methods

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- Initially focus on algorithms that might be classed as “iterative improvement”
- Take a candidate complete ‘solution’ and then try to fix or repair it
- Simplest version of this is “local search”





# Local Search Methods

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- A **neighbourhood function** is usually defined by using the concept of a move, which changes one or more attributes of a given solution to generate another solution.
- Definition
  - A solution  $x$  is called a local optimum with respect to the neighbourhood function  $N$ , if  $f(x) < f(y)$  for every  $y$  in  $N(x)$ .



# Local Search Methods

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- Why local search?
  - Exponential growth of the solution space for most of the practical problems;
  - Ambiguity of the model of the problem for being solved with exact algorithms;
  - Ease of use of problem specific knowledge in design of algorithm than in design of classical optimisation methods for a specific problem.



# Local Search Methods

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- Elements

- Representation of the solution;
- Evaluation function;
- Neighbourhood function
  - To define solutions which can be considered close to a given solution.
  - For example: For optimisation of real-valued functions in elementary calculus, for a current solution  $x_0$ , neighbourhood is defined as an interval  $(x_0-r, x_0+r)$ .



# Local Search Methods

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- Elements

- Neighbourhood search strategy
  - Random and systematic search;
- Acceptance criterion
  - first improvement; best improvement; best of non-improving solutions; random criteria

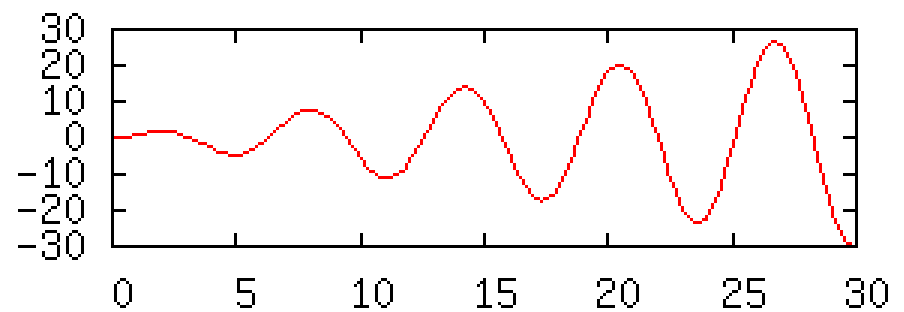
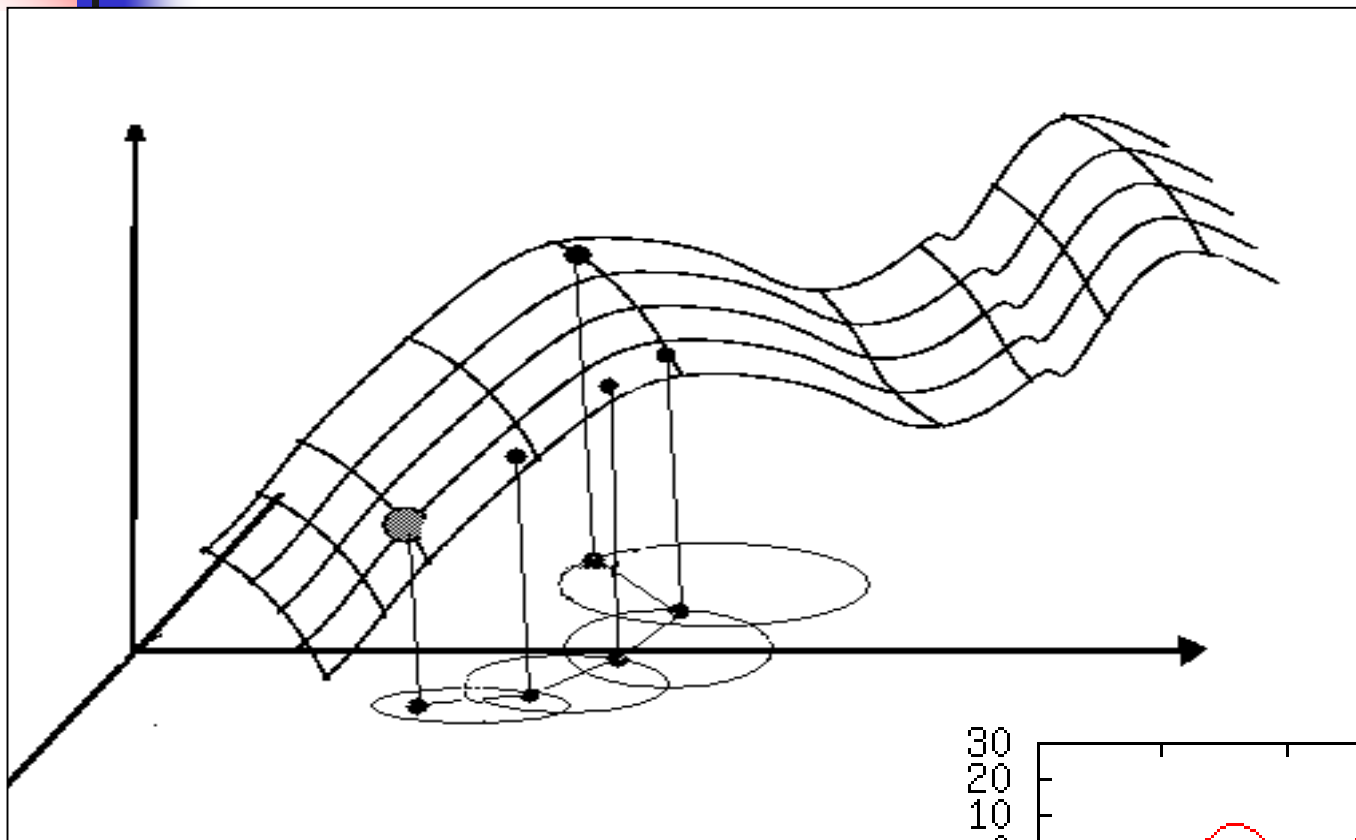


# Hill Climbing - Algorithm

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1. Pick a random point in the search space
2. Consider all the neighbours of the current state
3. Choose the neighbour with the best quality and move to that state
4. Repeat 2 thru 4 until all the neighbouring states are of lower quality
5. Return the current state as the solution state

# Hill Climbing - Examples





# Hill Climbing – Exercise

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- Problem

- TSP: A travelling salesman is visiting  $n$  cities
- Constraint: He can visit each city only once, and back to the starting city
- Objective: Find the shortest tour

- Task

- Apply the hill climbing method to solve TSP



# Hill Climbing – Exercise

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- How do we define the problem as a search problem?
  - What is a “point” in the **search space**?
  - How to represent a solution?
  - What is the search space?





# Hill Climbing – Exercise

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- How do we design the HC for TSP?
  - What is your evaluation function?
  - What is your neighbour?
  - What is your search strategy?
  - What is your acceptance criteria?

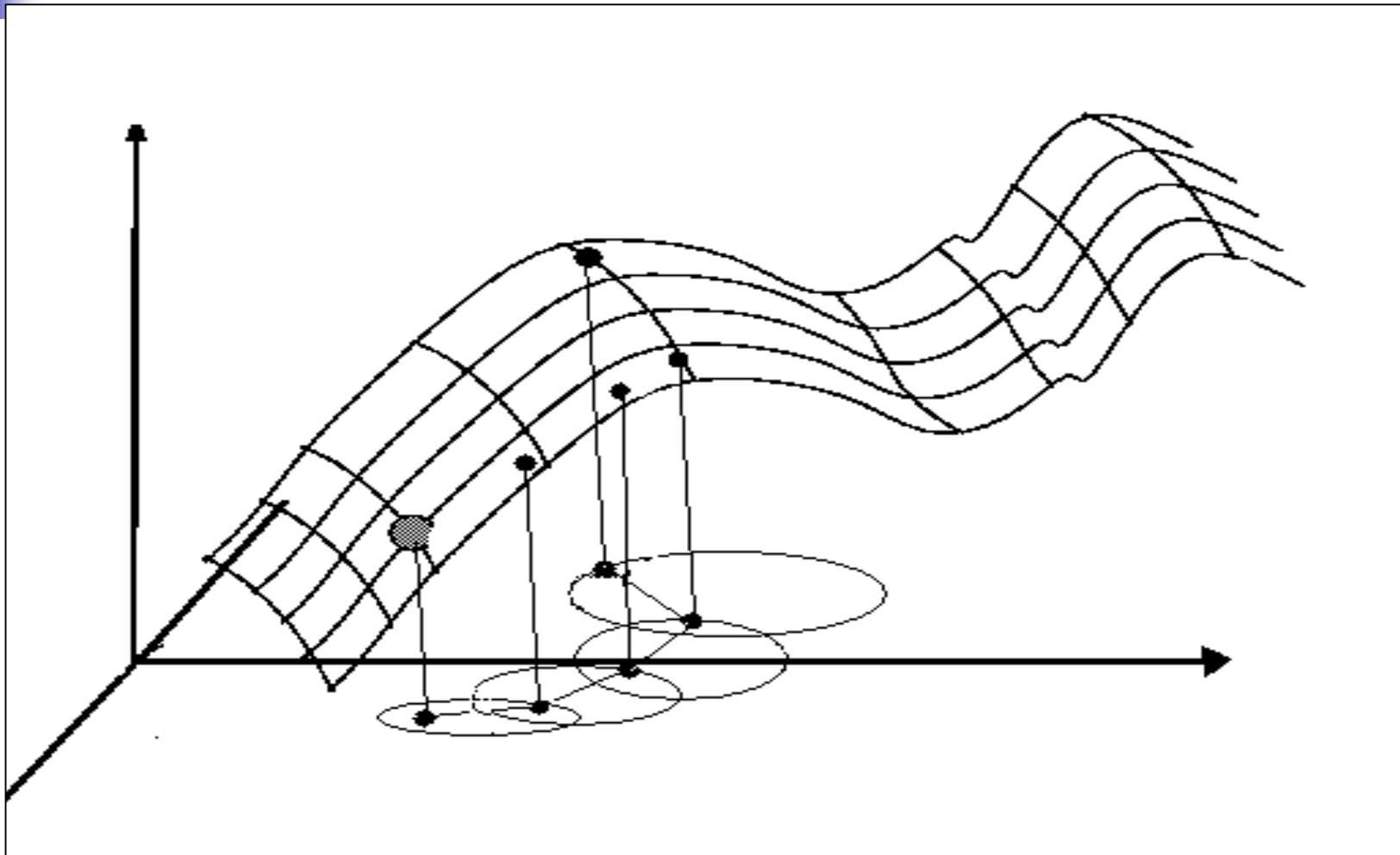


# Hill Climbing – Exercise

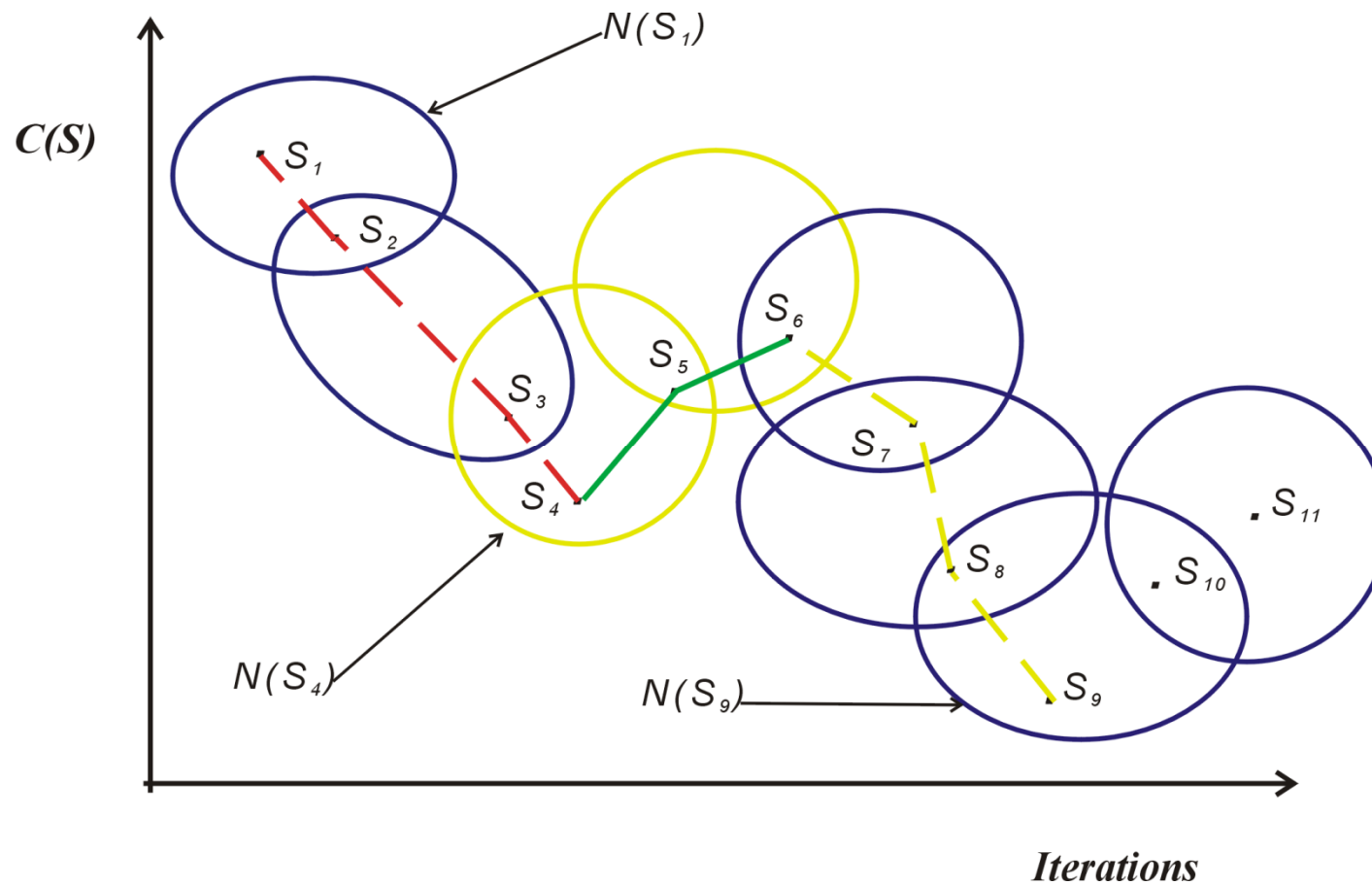
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- What's the problem in HC?
  - First improvement
    - Within the neighbourhood, if a point  $x'$  is an improvement then immediately terminate and jump straight to it
  - Best Improvement
    - Search the entire neighbourhood, and find the  $x'$  that gives the largest improvement
    - What to do in tie-breaks?

# Hill Climbing – local optima



# How can bad local optima be avoided?





# Summary

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- Optimisation problems
  - Definition
  - Methods
- Local search algorithms
  - Elements
  - Hill climbing



# Learning objectives

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- Terminology
  - Local vs. global optima
  - Neighborhood
  - Feasibility
- Local search algorithms
  - Concepts
  - Elements
  - Basic hill climbing