

Artificial Intelligence Methods (G52AIM)

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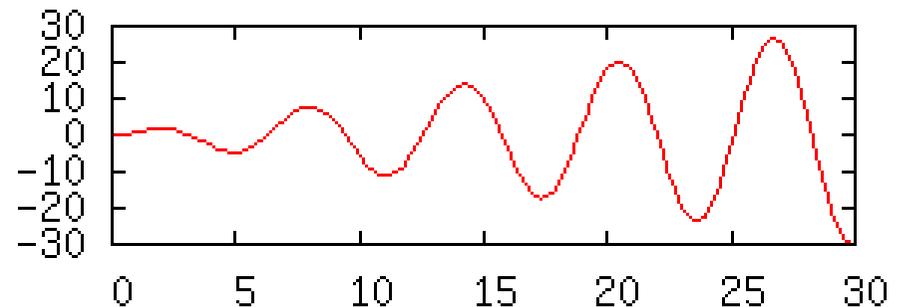
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Local Search

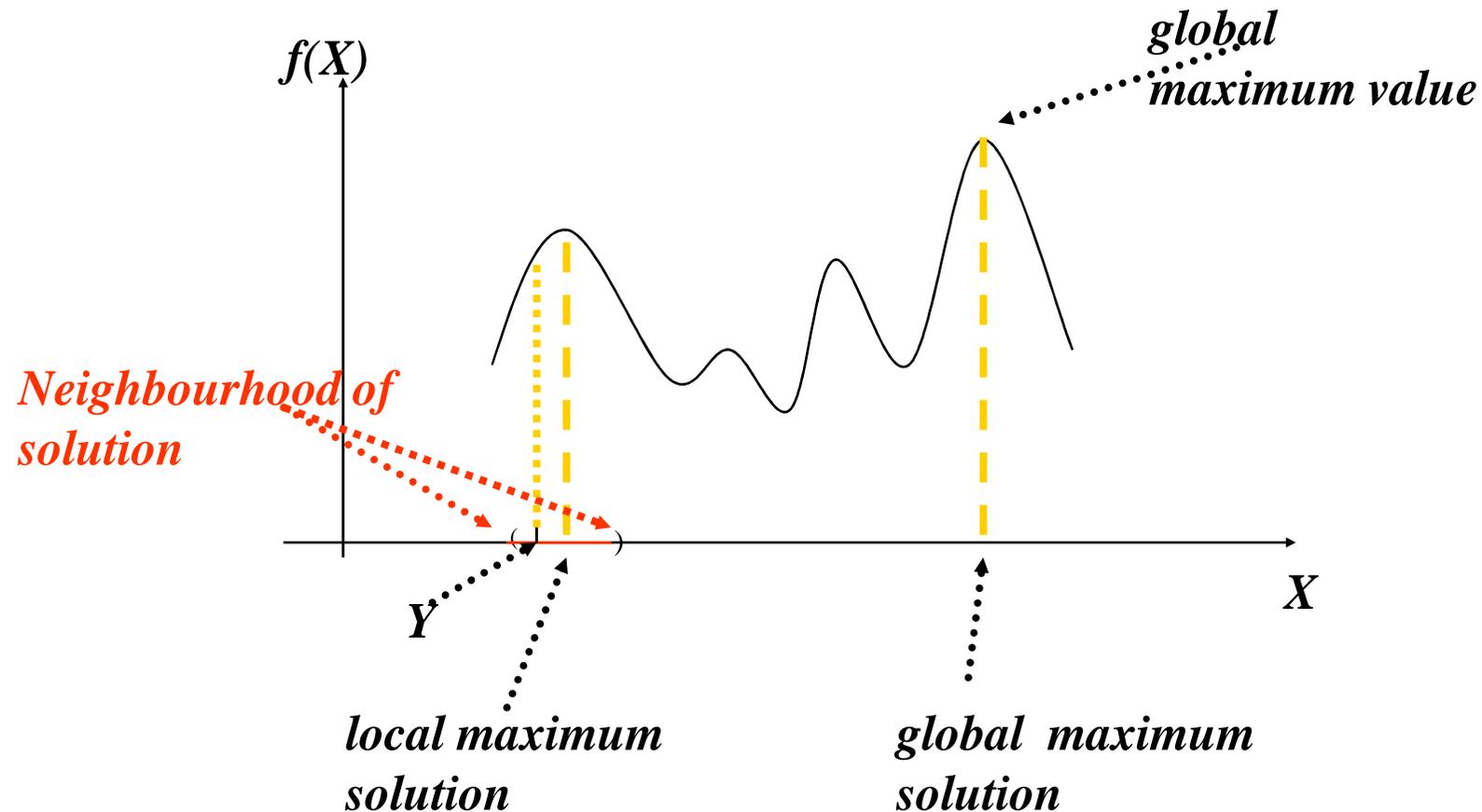
Optimisation Problems: Definition

- 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 $0 \leq x \leq 30$

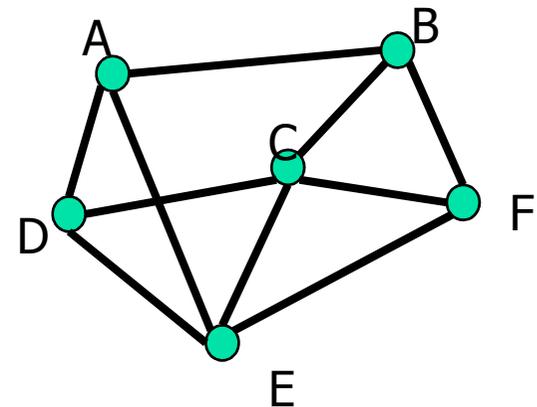
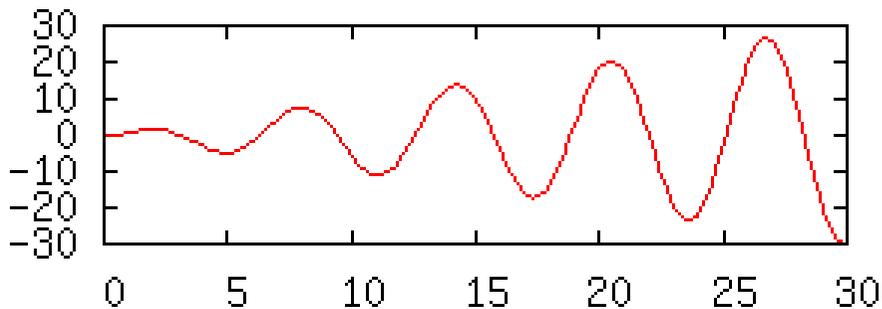


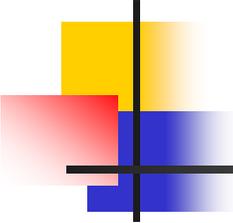
Optimisation Problems: terminology



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.



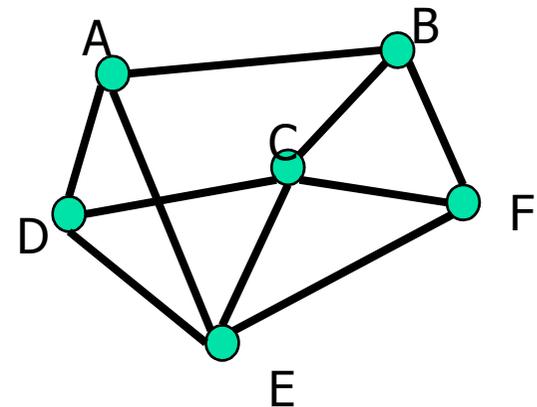


Methods of optimisation

- 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

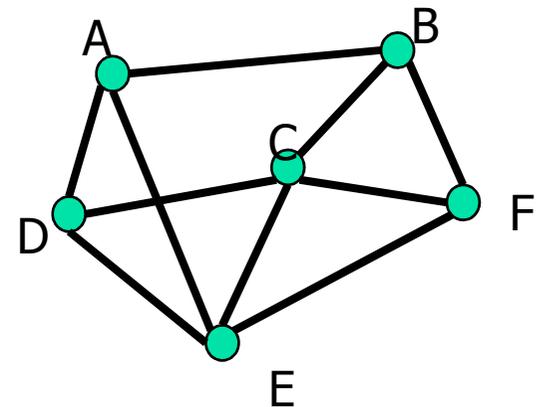
- Mathematical optimisation
 - +: Guarantee of optimality
 - - : Unable to solve larger instances of difficult problems due to large amount of computational time and memory needed;

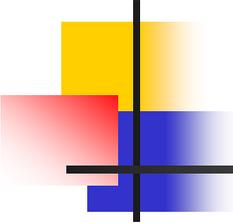


Methods of optimisation

- 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;



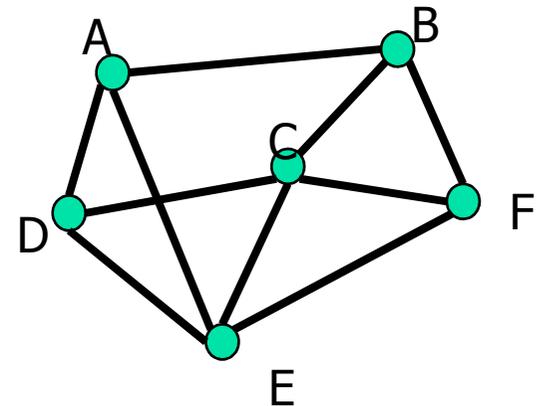


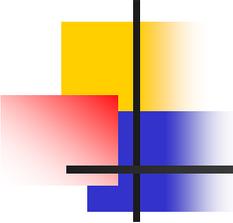
Methods of optimisation

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

Methods of optimisation (cont.)

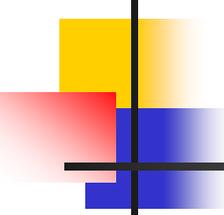
- Local Search algorithms
 - A neighbourhood search or so called local search method starts from some **initial (complete) solution** and moves to a better neighbouring solution until it arrives at a **local optimum**, one that does not have a better neighbour.
 - Examples: *k-opt* for TSP, etc;





Methods of optimisation (cont.)

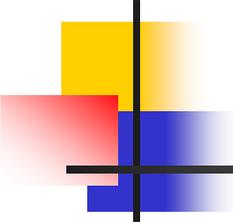
- Local Search algorithms
 - +: Ease of implementation;
 - +: Guarantee of local optimality usually in small computational time;
 - +: No need for exact model of the problem;
 - -: Poor quality of solution due to getting stuck in poor local optima;



Methods in G52AIM

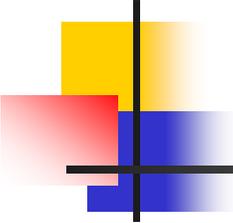
- Meta-heuristics

- 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;



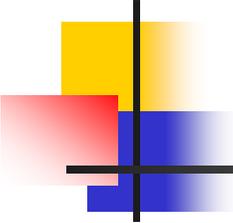
Methods in G52AIM

- 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

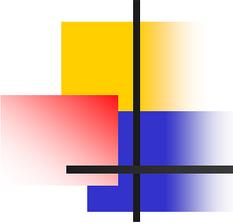
- Meta-heuristics
 - Population based approach
 - Genetic algorithms
 - “survival of the fittest”
 - Ant algorithms
 - “wander around a lot and leave a trail”
 - Genetic programming
 - Learn to program
 - Evolutionary algorithms, etc



Methods in G52AIM

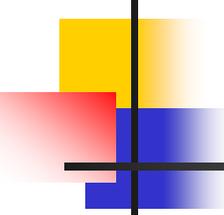
■ Meta-heuristics

- +: 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.



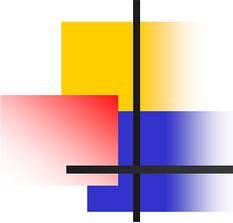
Problems in G52AIM

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



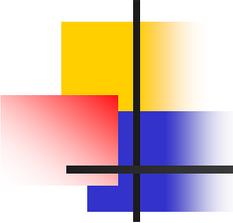
Local Search Methods

- 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”



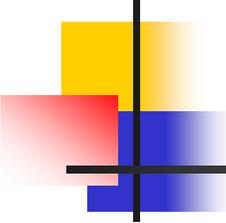
Local Search Methods

- 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

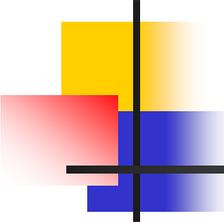
- Why local search?
 - Exponential growth of the solution space for most of the practical problems;
 - Ambiguity of the model of the problem for being solved with exact algorithms;
 - Ease of use of problem specific knowledge in design of algorithm than in design of classical optimisation methods for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



Local Search Methods

- Elements

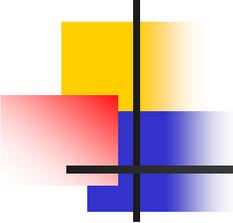
- Representation of the solution;
- Evaluation function;
- Neighbourhood function
 - To define solutions which can be considered close to a given solution.
 - For example: For optimisation of real-valued functions in elementary calculus, for a current solution x_0 , neighbourhood is defined as an interval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



Local Search Methods

- Elements

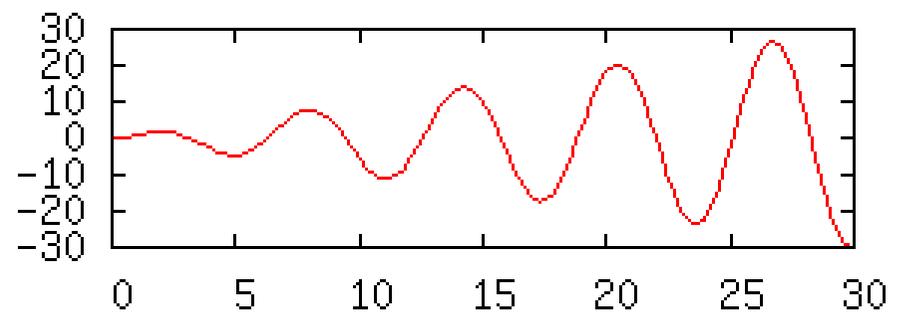
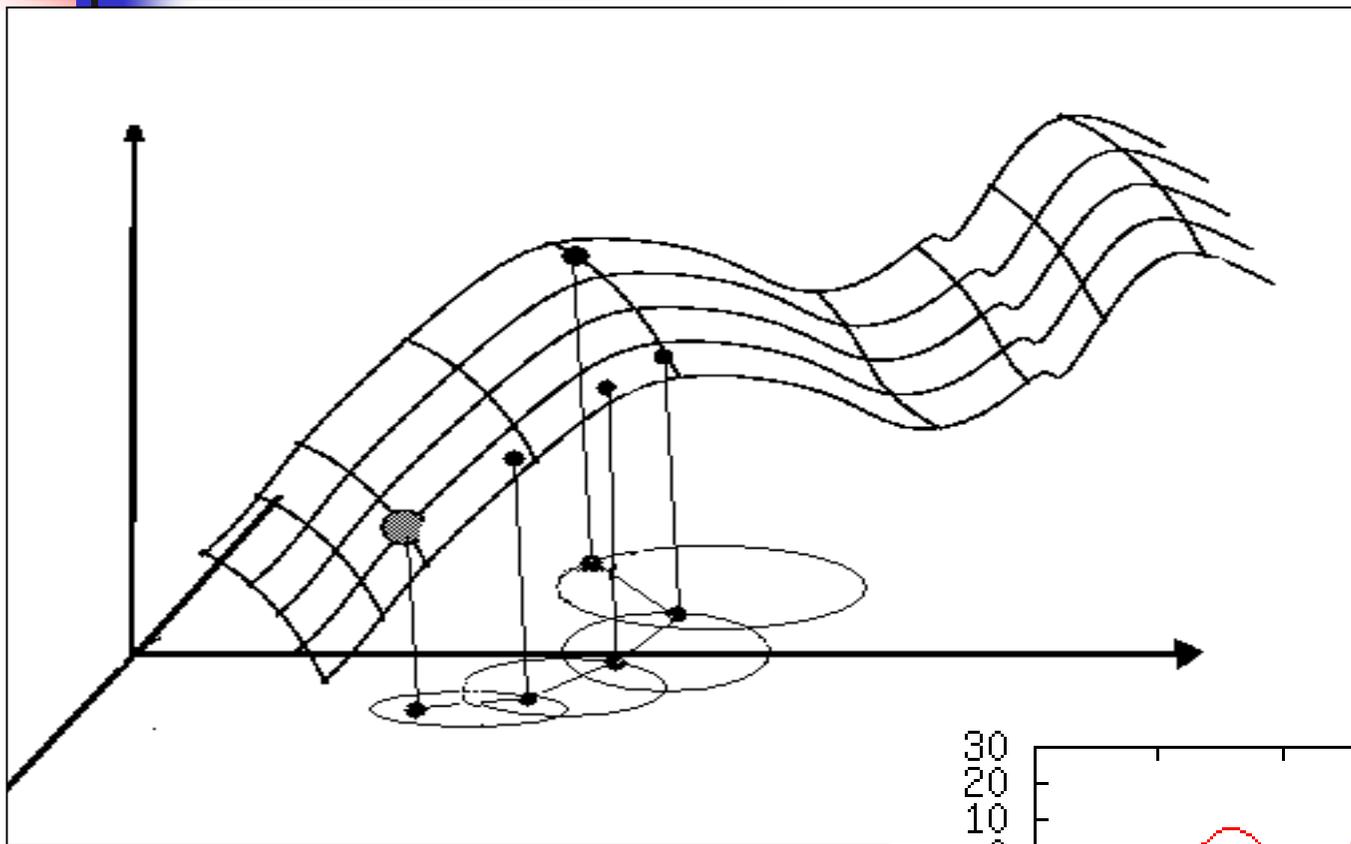
- Neighbourhood search strategy
 - Random and systematic search;
- Acceptance criterion
 - first improvement; best improvement; best of non-improving solutions; random criteria

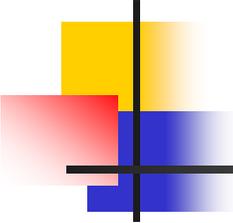


Hill Climbing - Algorithm

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





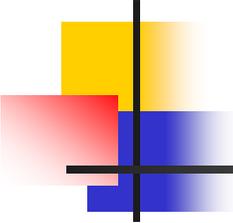
Hill Climbing – Exercise

- 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

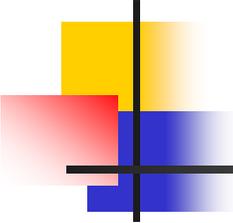
- Task

- Apply the hill climbing method to solve TSP



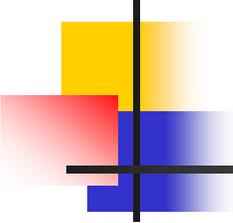
Hill Climbing – Exercise

- 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

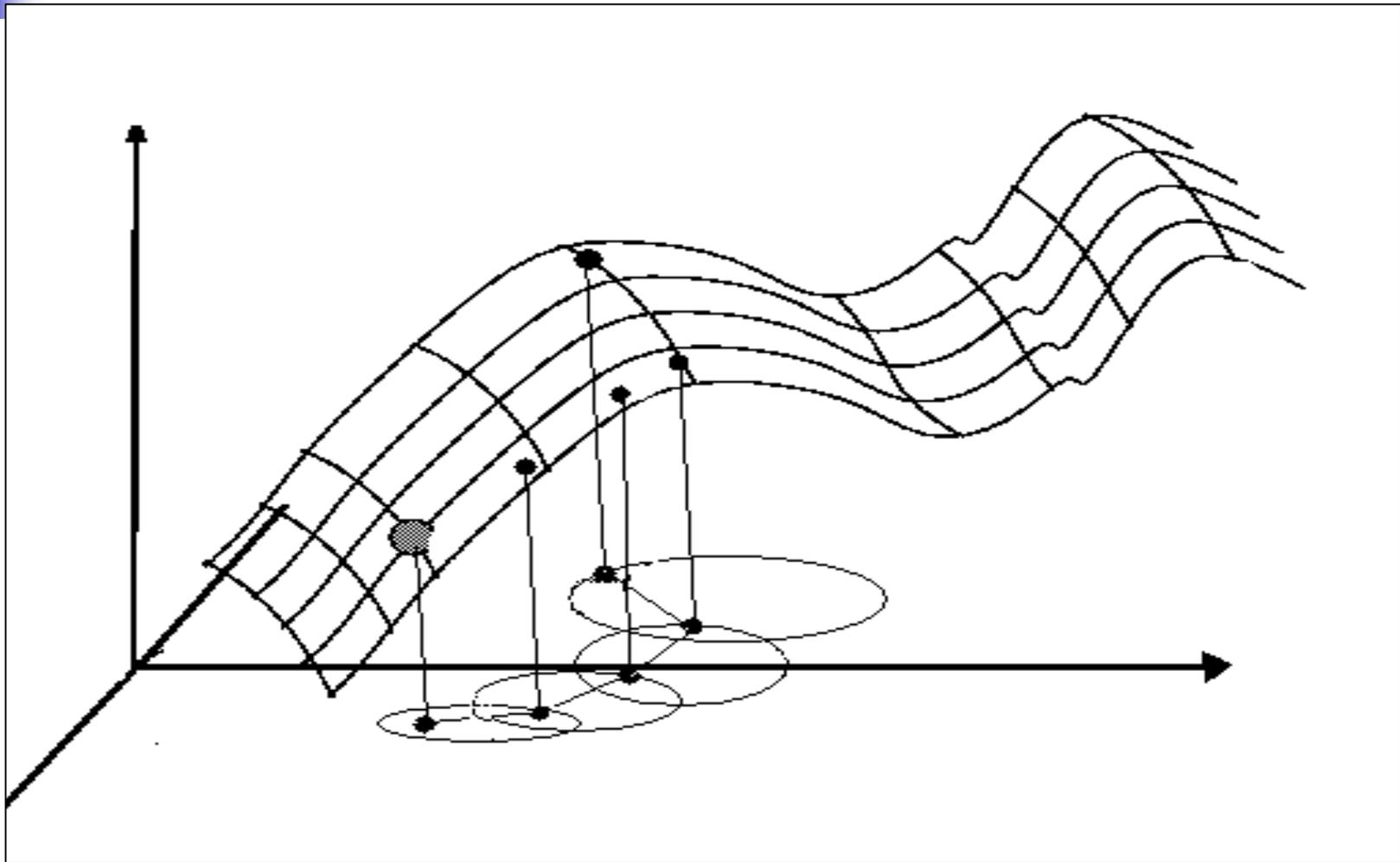
- How do we design the HC for TSP?
 - What is your evaluation function?
 - What is your neighbour?
 - What is your search strategy?
 - What is your acceptance criteria?



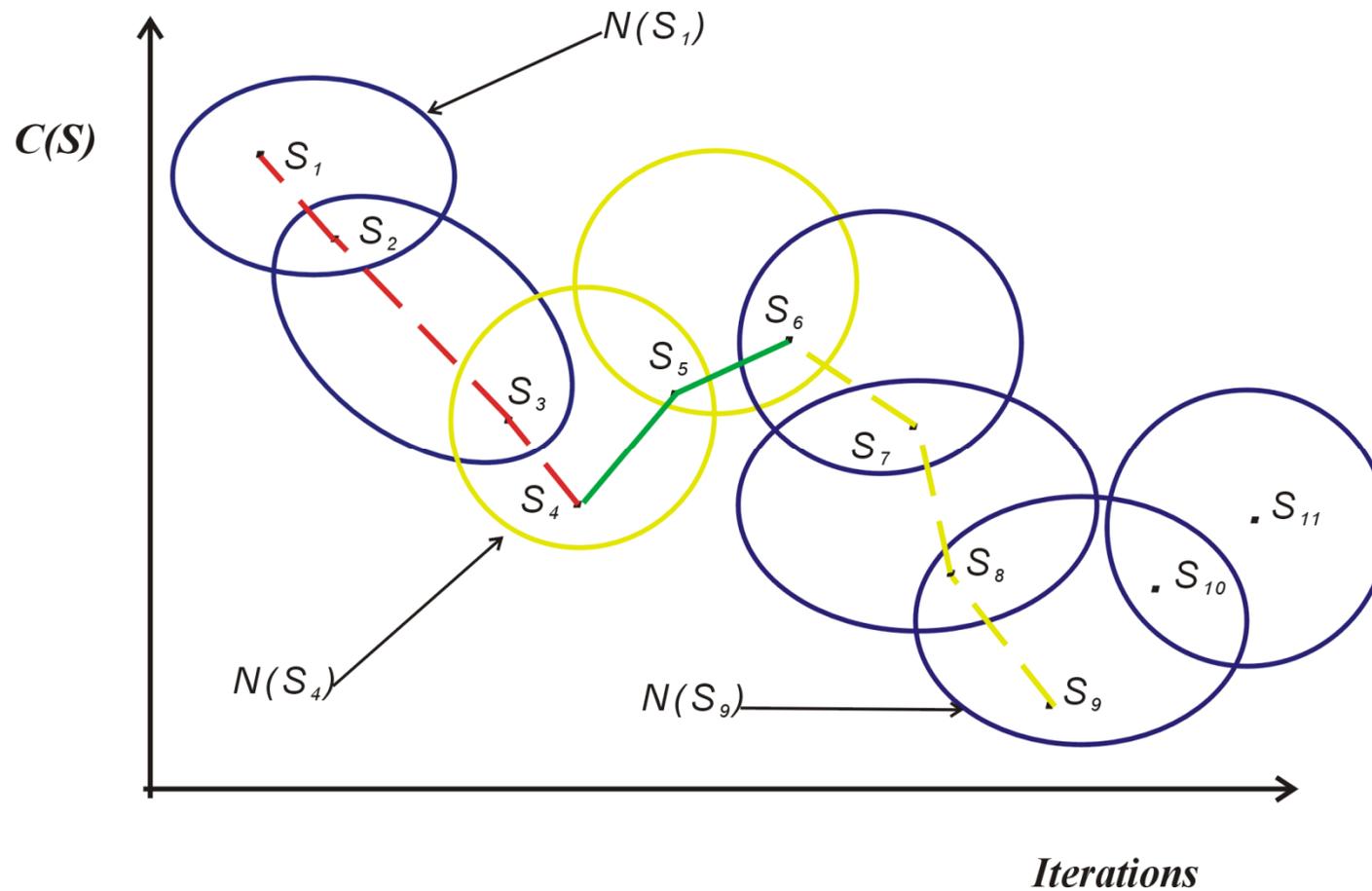
Hill Climbing – Exercise

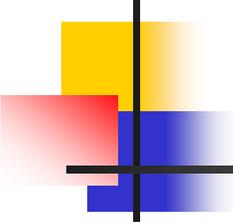
- What's the problem in HC?
 - First improvement
 - Within the neighbourhood, if a point x' is an improvement then immediately terminate and jump straight to it
 - Best Improvement
 - Search the entire neighbourhood, and find the x' that gives the largest improvement
 - What to do in tie-breaks?

Hill Climbing – local optima



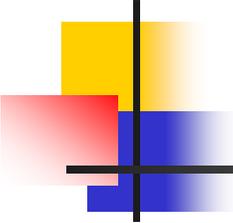
How can bad local optima be avoided?





Summary

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



Learning objectives

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