

Heuristic Algorithms

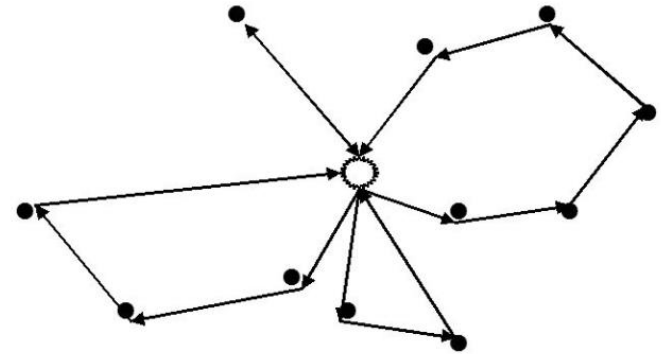
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Heuristic Algorithms

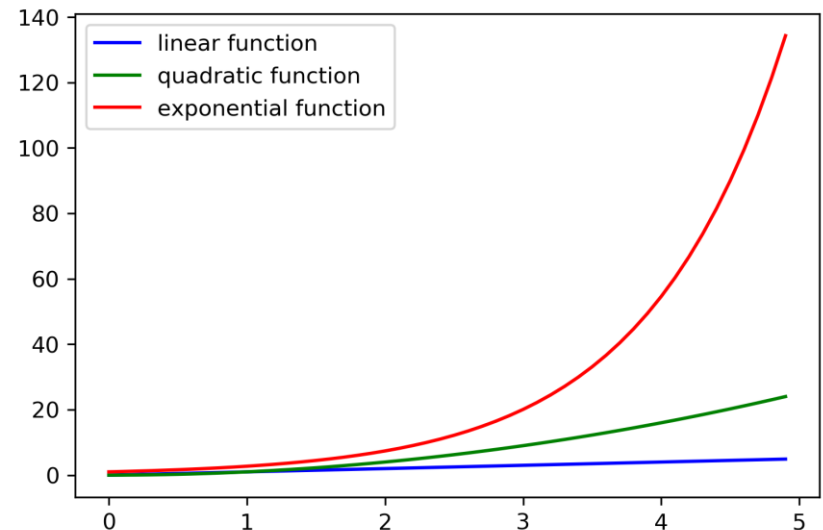
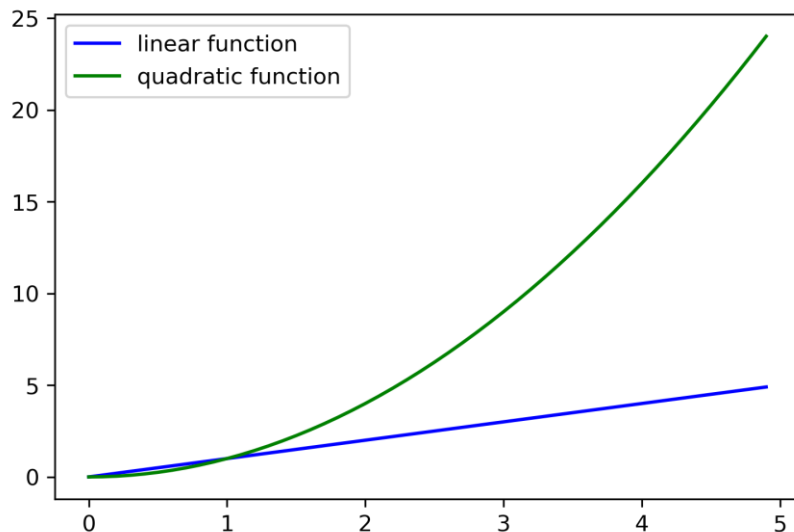
- Various **Combinatorial** Optimization Problems (COP) in industry
 - TSP
 - VRP
 - Scheduling
 - Packing
 - Location
 - Course timetabling
 - Portfolio optimization
 - ...





Heuristic Algorithms

- Bad news: COP is generally **NP-hard**
 - Can be solved to optimality
 - Solution time increases **exponentially**
 - May not be practical
 - E.g.: solve VRP with **1000+** nodes within **5** minutes





Heuristic Algorithms

- Good news: compromised solution
 - Solve for suboptimal solution **quickly**
 - How? Heuristic algorithms!
- **Heuristic**: a procedure that determines **near-optimal** solutions to an optimization problem.
 - Sometimes can find optimal solution, but cannot prove its optimality
 - **Widely used in industry!**



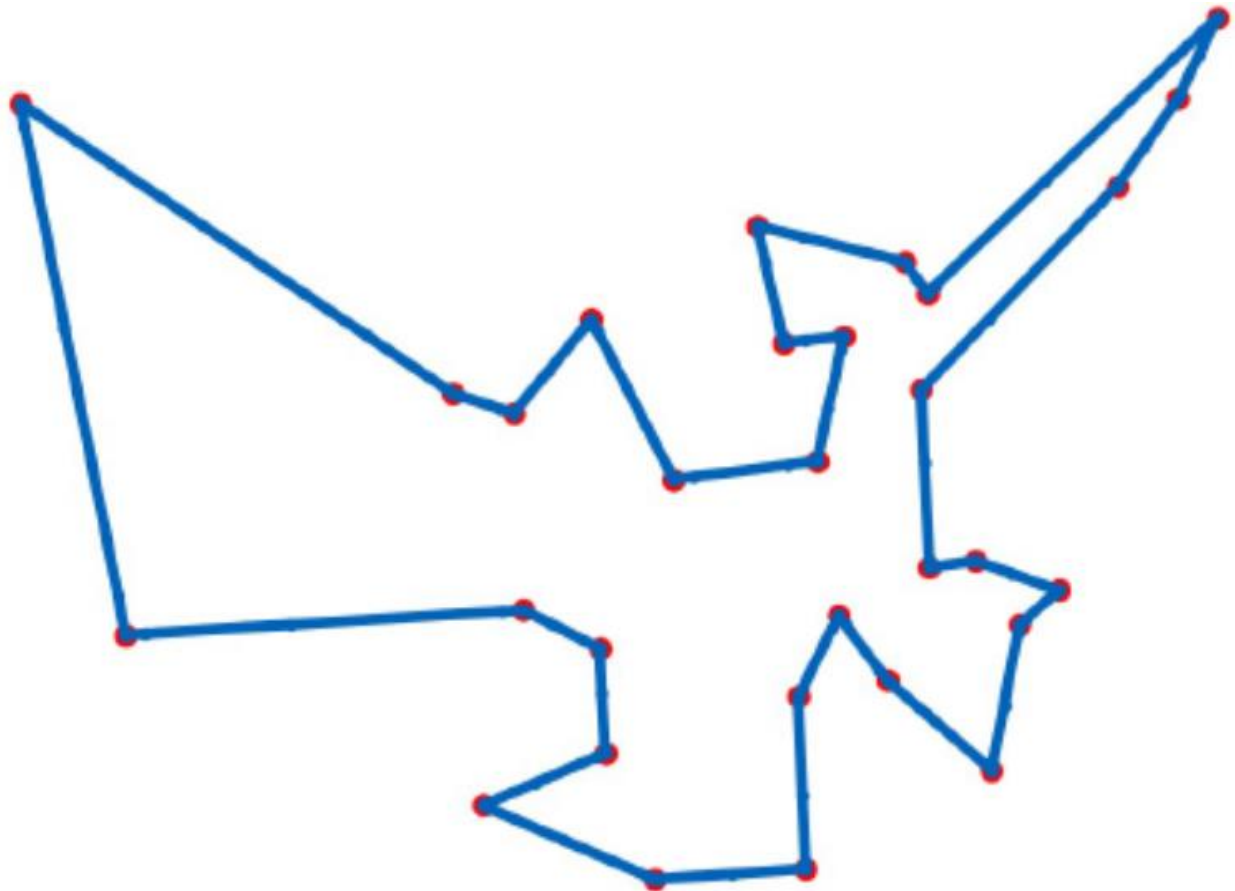
Competition

- JD
 - <https://jdata.jd.com/html/detail.html?id=5>
- Huawei
 - https://www.noahlab.com.hk/logistics-ranking/#/competition_details
 - <https://competition.huaweicloud.com/information/1000041601/circumstance>



TSP

- Problem: finding a shortest circle visiting each node exactly once.
- E.g.:





TSP

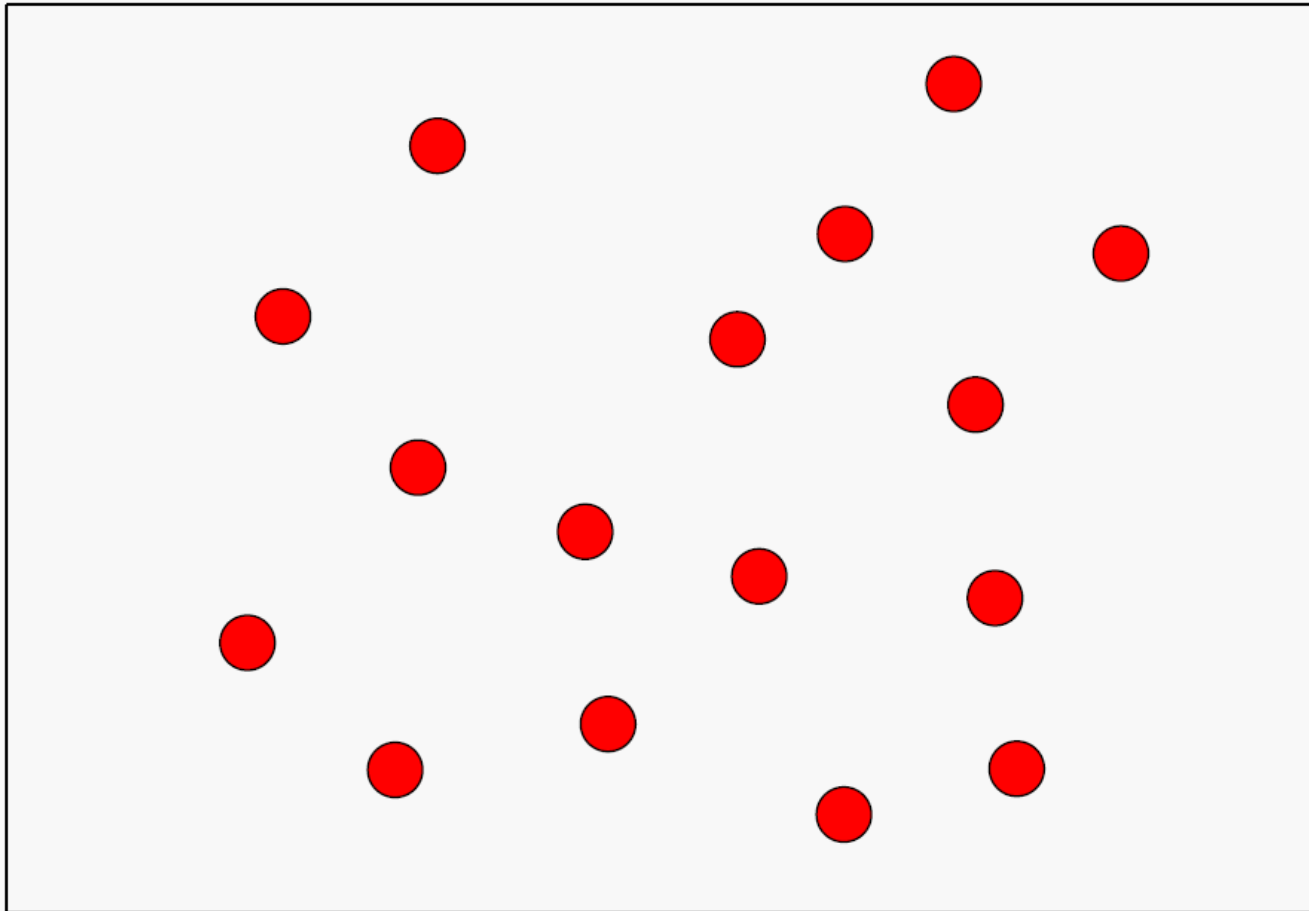
- Mona Lisa Instance
 - 100,000 nodes
 - Best known solution by heuristics:
 - Dist = 5,757,191
 - Best known lower bound
 - Dist = 5,757,084
 - Gap = 0.0019%
 - \$1,000 for finding better
 - <http://www.math.uwaterloo.ca/tsp/>





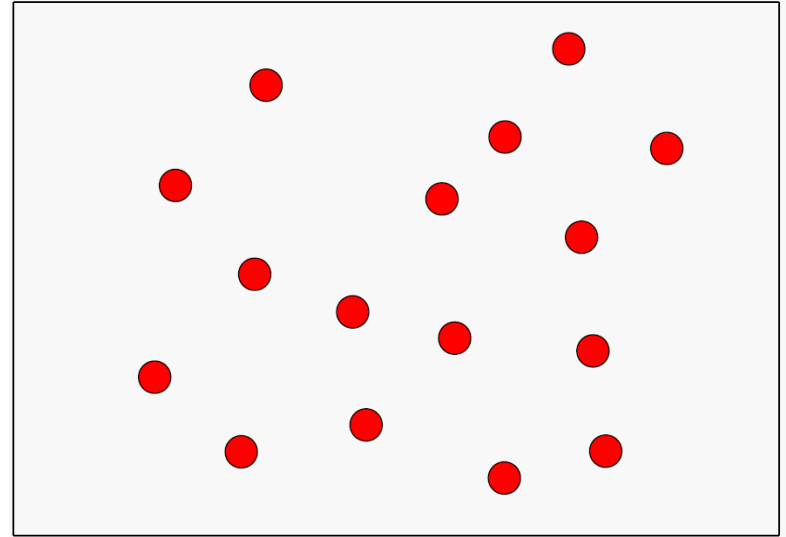
TSP

- How to solve it? Brain storming...





TSP



- Idea 1: Enumeration
- # feasible routes?
 - $(\text{\#nodes} - 1)!$
- E.g.:
 - 101 nodes; # routes = $100! \approx 10^{158}$
 - Super powerful computer
 - # CPUs is $10^{80} \approx$ # atoms in the universe
 - 5.39×10^{-44} seconds (Planck time) per operation
 - $\sim 5.39 \times 10^{34}$ seconds to enumerate
 - Compare to 4.33×10^{17} s. \approx age of universe



TSP

- Idea 2: Greedy Algorithm
 - Iteratively select the **shortest** arc



TSP

- Idea 3: Nearest neighbor
 - Iteratively select the **nearest** unvisited node



TSP

- Idea 4: Cheapest insertion
 - Iteratively insert a node to a subtour in a **shortest** possible way.



TSP

- Idea 5: Random insertion
 - Iteratively insert a random node to current subtour



TSP

- **Guess:** which performs better?
 - 1. Nearest neighbor
 - 2. Cheapest insertion



TSP

- Idea 6: 2-opt
 - Swap two edges



TSP

- Idea 7: 3-opt
 - Swap three edges



TSP

- Summary
- 1. Construction heuristics
 - Nearest neighbor
 - Cheapest insertion
 -
- 2. Improvement heuristics (local search)
 - 2-opt
 - 3-opt
 -



TSP

- Key idea:

construct, then improve

- Common wisdom:

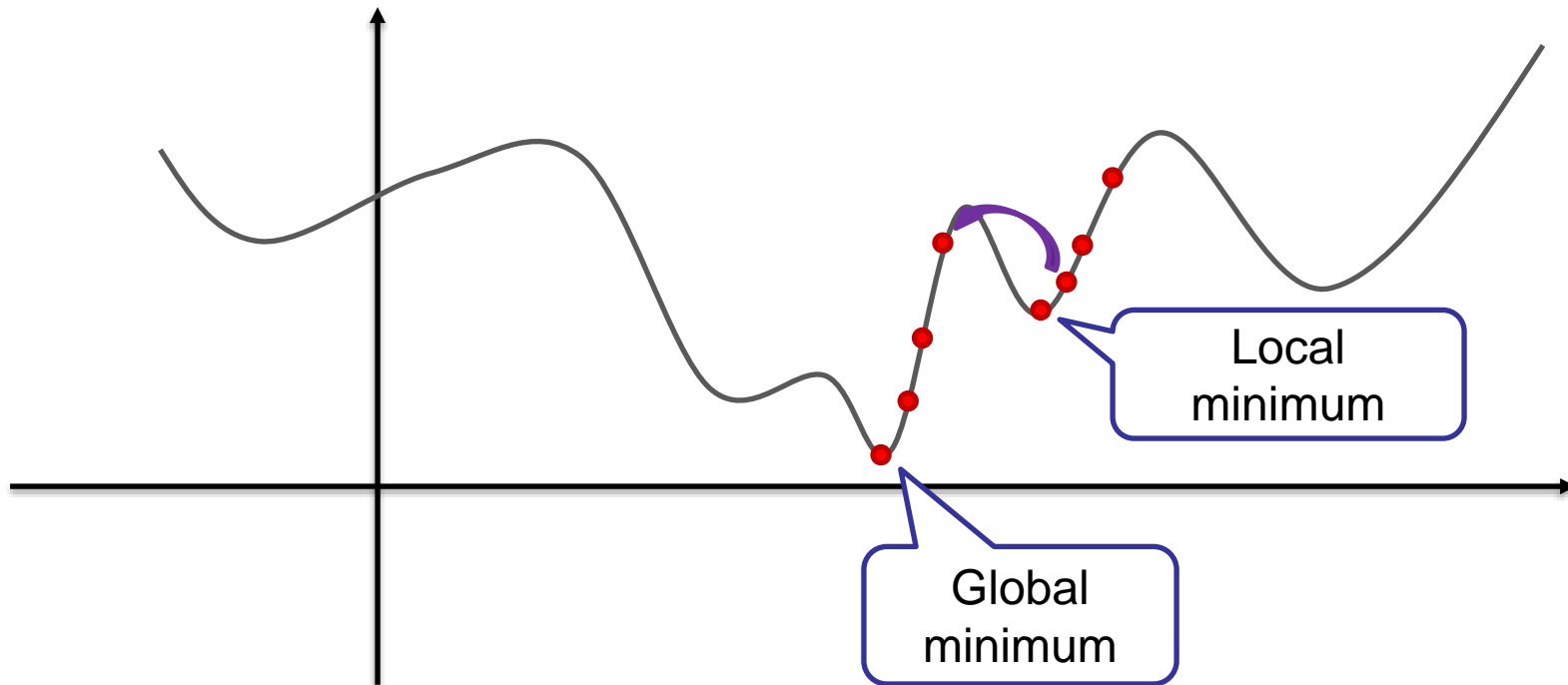
**Solution quality depends largely on
improvement heuristics!**

- Improvement
 - Solution representation (node sequence)
 - Definition of neighbor (2-opt)
 - Search strategy (optimal neighbor)



TSP

- Caution: **Local optimal**



How to escape from local minimum?



Simulated Annealing

- Main idea: introduce **randomness**
 - Permits worse movement with some probability
 - Probability decreases with time
 - tends to stability
 - analogy with **physical annealing**





Simulated Annealing

- How to define the probability
 - law of thermodynamics
 - Probability is

$$p = e^{\frac{\Delta E}{kT}}$$

- ΔE : difference of objective values
- k : a positive constant
- T : current temperature
- Hence, $p \in (0,1)$ when $\Delta E < 0$



Simulated Annealing

- Analogy
 - Tour distance \Leftrightarrow energy level
 - Feasible solution \Leftrightarrow system state
 - Neighbore solution \Leftrightarrow change of state
 - Control parameter \Leftrightarrow temperature
 - Final solution \Leftrightarrow status when solid





Simulated Annealing

Select an initial solution $\omega \in \Omega$

Select the temperature change counter $k = 0$

Select a temperature cooling schedule, t_k

Select an initial temperature $T = t_0 \geq 0$

Select a repetition schedule, M_k , that defines the number of iterations executed at each temperature, t_k

Repeat

Set repetition counter $m = 0$

Repeat

Generate a solution $\omega' \in N(\omega)$

Calculate $\Delta_{\omega, \omega'} = f(\omega') - f(\omega)$

If $\Delta_{\omega, \omega'} \leq 0$, then $\omega \leftarrow \omega'$

If $\Delta_{\omega, \omega'} > 0$, then $\omega \leftarrow \omega'$ with probability $\exp(-\Delta_{\omega, \omega'} / t_k)$

$m \leftarrow m + 1$

Until $m = M_k$

$k \leftarrow k + 1$

Until stopping criterion is met

“Metropolis Criterion”

Pseudo-code: A.G. Nikolaev and S. H. Jacobson, Simulated annealing. in Handbook of Metaheuristics, 2nd Ed.



Simulated Annealing

- Illustration

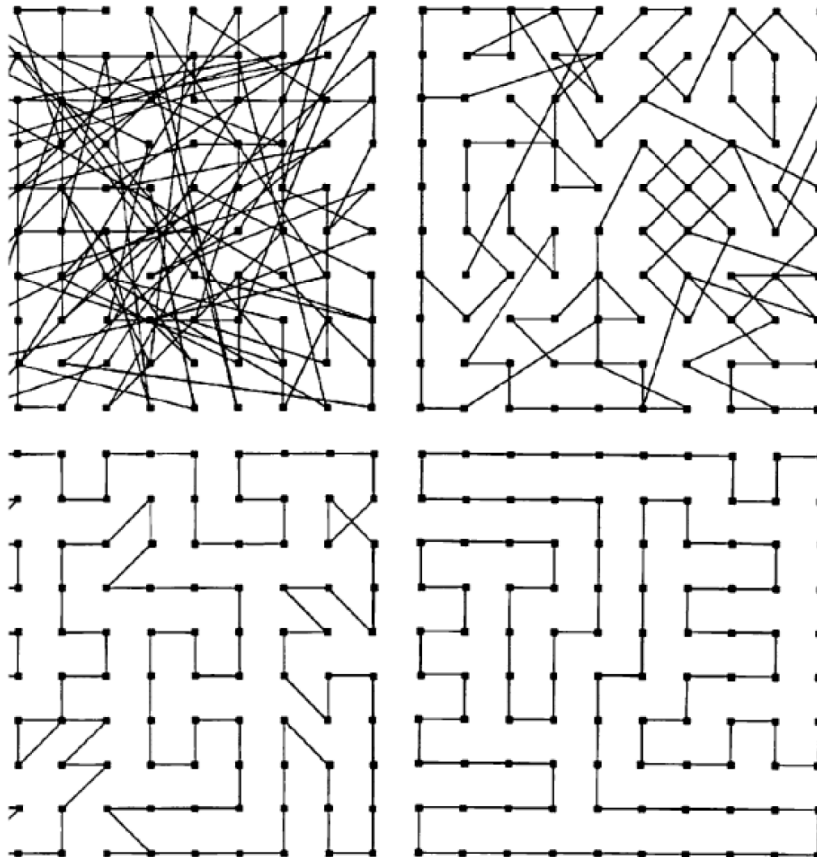
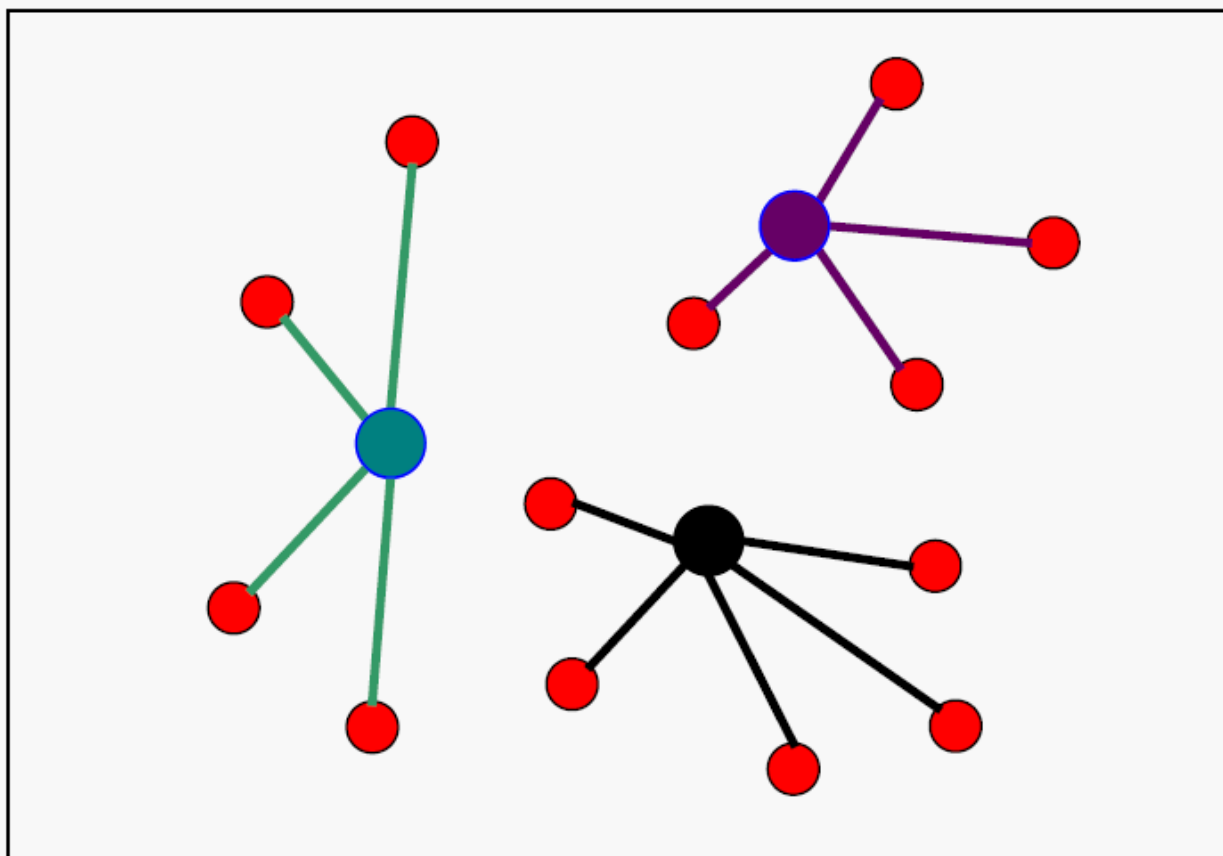


Image: E. K. Burke and G. Kendall, Search Methodologies, Introductory Tutorials in Optimization and Decision Support Techniques. Springer, 2014.



Facility Location

- Problem: Determine the facilities to open and assign customer to facilities to minimize the fixed and variable costs





Facility Location

- Question: Design a heuristic algorithm

