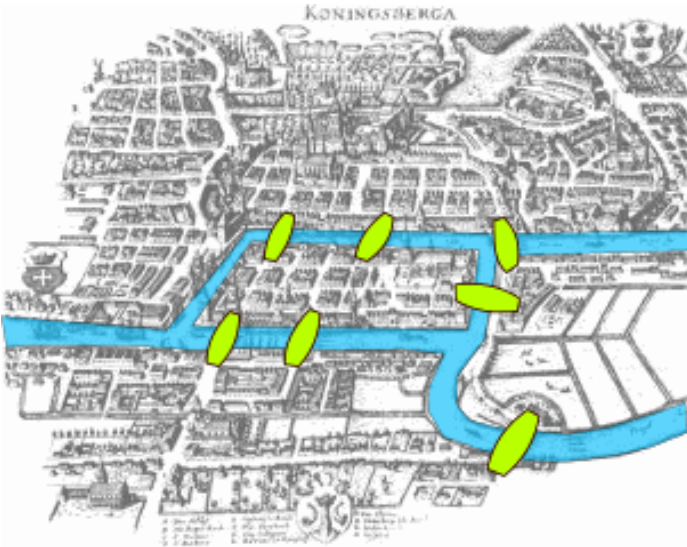


Traveling Salesperson Problem

Outline

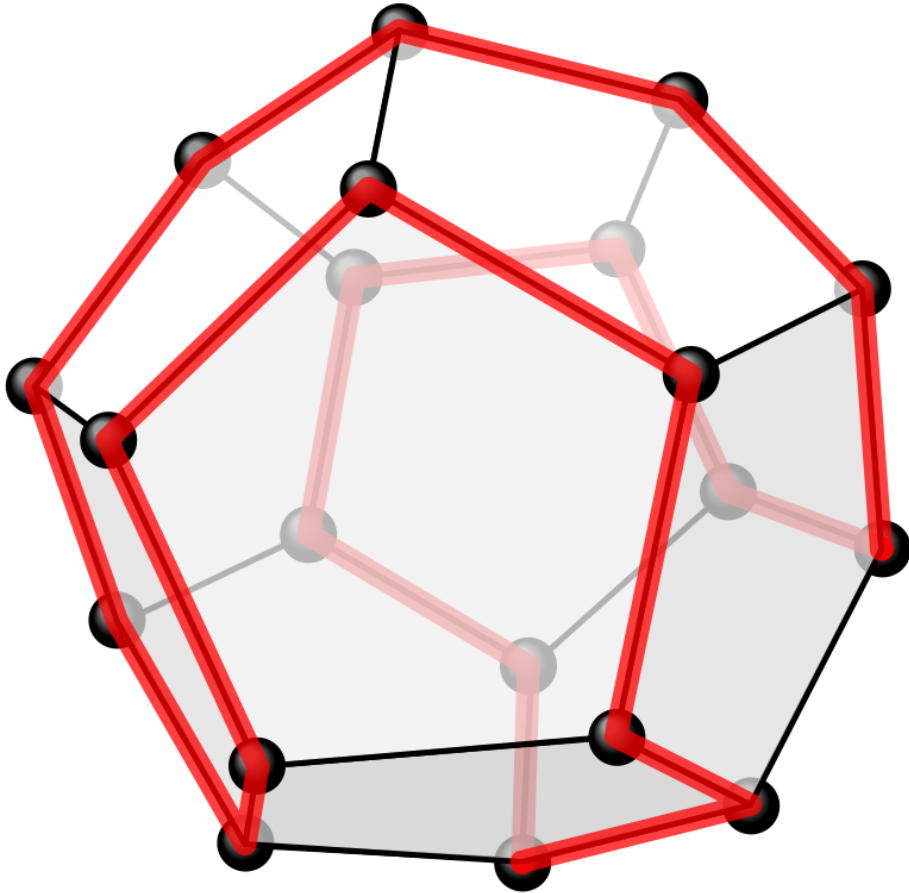
- Problem Background
- Notable Variants
- Popular Heuristics
- Christofides Algorithm

Graph Theory and Eulerian Cycles



- In 1736 Leonhard Euler laid the foundations of graph theory with the Königsberg Bridge problem
- He wanted to take a walk where he would pass over each bridge exactly once
- An Eulerian Cycle (circuit) is a path that goes over each edge once and returns to the origin
- Euler proved that in order to have such a circuit all nodes needed to have an even degree

TSP Origins



- William Rowan Hamilton invented the icosian game which involved finding a path that visits every vertex of a dodecahedron once
- A **Hamiltonian cycle** is the general term in graph theory for a cycle on an undirected graph that visits every node once and then return to the origin (starting node)

Integer Programming Formulation

$$x_{ij} = \begin{cases} 1 & \text{the path goes from city } i \text{ to city } j \\ 0 & \text{otherwise} \end{cases}$$

$$\min \sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}:$$

$$x_{ij} \in \{0, 1\}$$

$$i, j = 1, \dots, n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1$$

$$j = 1, \dots, n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1$$

$$i = 1, \dots, n;$$

$$\sum_{i \in Q} \sum_{j \neq i, j \in Q} x_{ij} \leq |Q| - 1 \quad \forall Q \subsetneq \{1, \dots, n\}, |Q| \geq 2$$

- IP solutions are slow compared to heuristic and meta-heuristic approaches
- There is a lot about the specific geometry of the problem that can be leveraged for a better solution

The Cutting-Plane Method

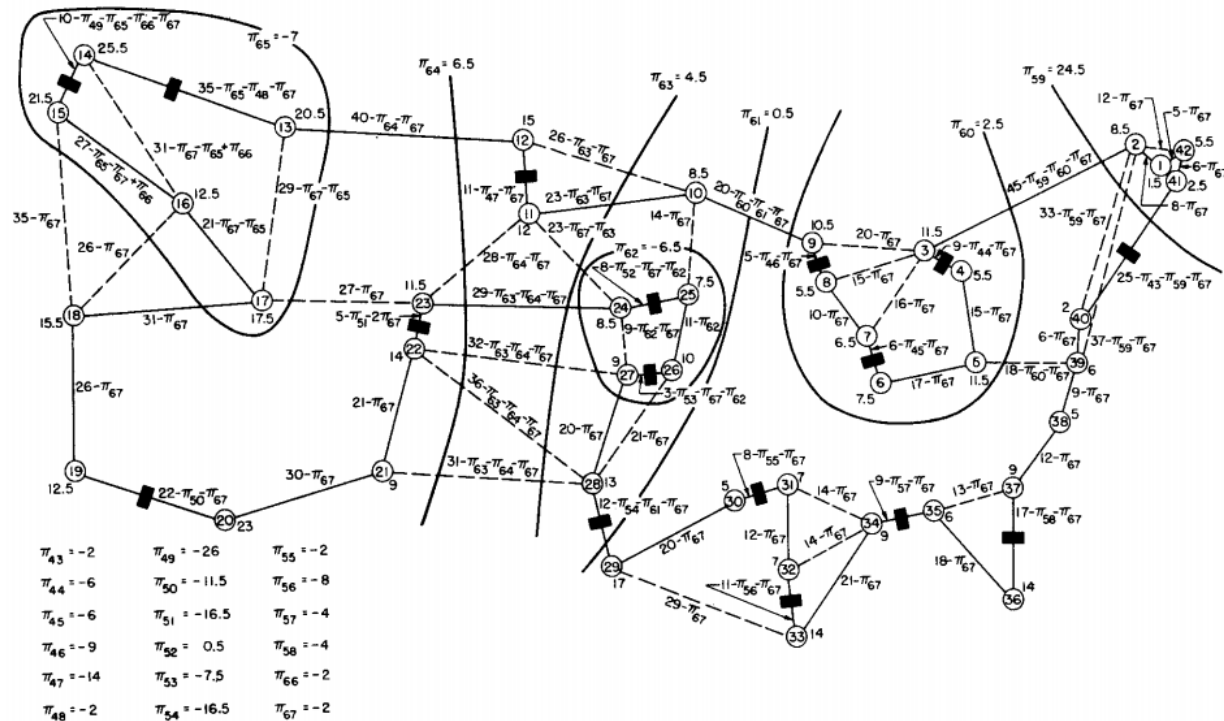


FIG. 17. Only the right-hand side of the equations satisfied by π_I are shown on the map; the left-hand side on line (I, J) is $\pi_I + \pi_J$. Dotted links (I, J) correspond to additional basic variables x_{IJ} .

SOLUTION OF A LARGE-SCALE TRAVELING-SALESMAN PROBLEM*

G. DANTZIG, R. FULKERSON, AND S. JOHNSON

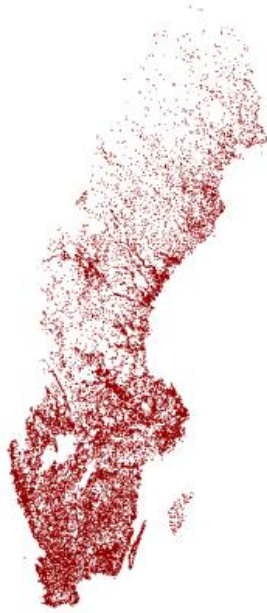
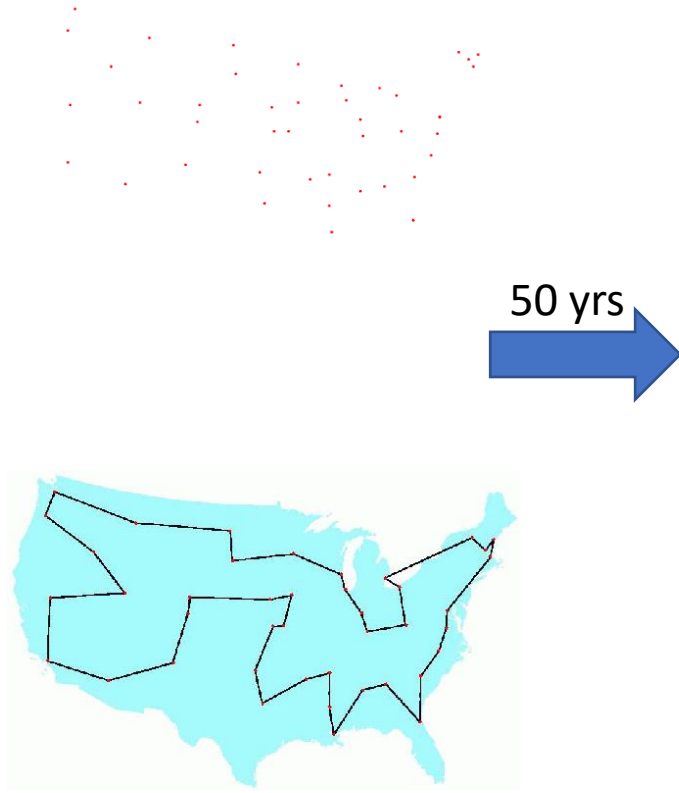
The Rand Corporation, Santa Monica, California

(Received August 9, 1954)

It is shown that a certain tour of 49 cities, one in each of the 48 states and Washington, D. C., has the shortest road distance.

THE TRAVELING-SALESMAN PROBLEM might be described as follows: Find the shortest route (tour) for a salesman starting from a given city, visiting each of a specified group of cities, and then returning to the original point of departure. More generally, given an n by n symmetric matrix $D = (d_{IJ})$, where d_{IJ} represents the 'distance' from I to J , arrange the points in a cyclic order in such a way that the sum of the d_{IJ} between consecutive points is minimal. Since there are only a finite number of possibilities (at most $\frac{1}{2}(n-1)!$) to consider, the problem is to devise a method of picking out the optimal arrangement which is reasonably efficient for fairly large values of n . Although algorithms have been devised for problems of similar nature, e.g., the optimal assignment problem,^{3,7,8} little is known about the traveling-salesman problem. We do not claim that this note alters the situation very much; what we shall do is outline a way of approaching the problem that sometimes, at least, enables one to find an optimal path and prove it so. In particular, it will be shown that a certain arrangement of 49 cities, one in each of the 48 states and Washington, D. C., is best, the d_{IJ} used representing road distances as taken from an atlas.

TSP Origins



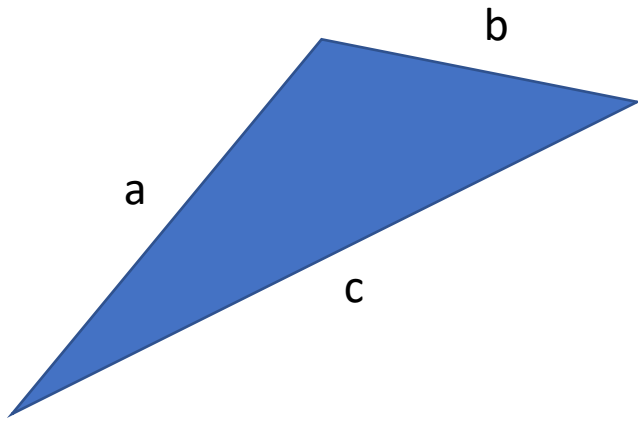
- The general form of the TSP and its mathematical treatment was first worked on in the 1930s
- In 1954 Dantzig (of LP fame) presented a solution for an instance with 49 cities
(6,206,957,796,268,036,335,431,144,523,686,687,519,260,743,177,338,880,000,000,000 possible tours)
- In 2004 a paper with a solution for 24,978 cities was published

TSP Today



- Today there exist excellent opensource solvers that can solve small problems to optimality and do well on much larger ones
[<http://www.math.uwaterloo.ca/tsp/concorde.html>]
- People continue to solve larger and larger problems using super computers taking decades worth of compute time on a single processor

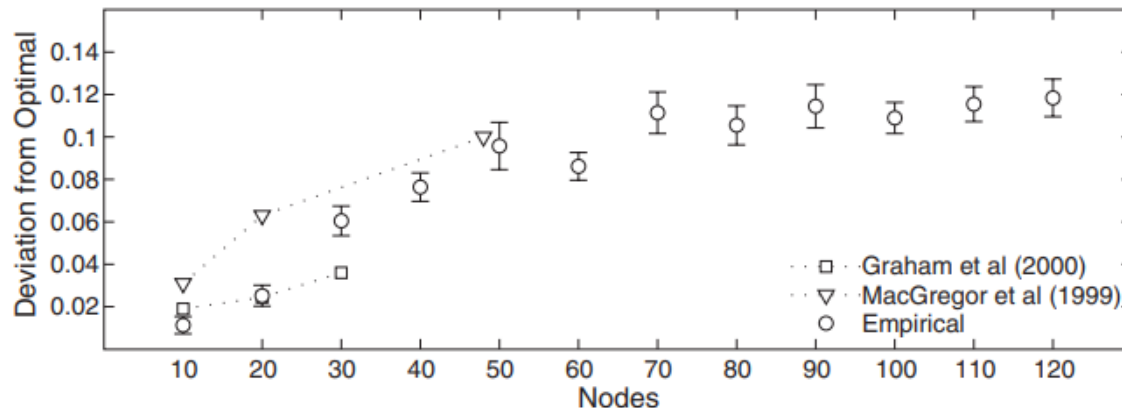
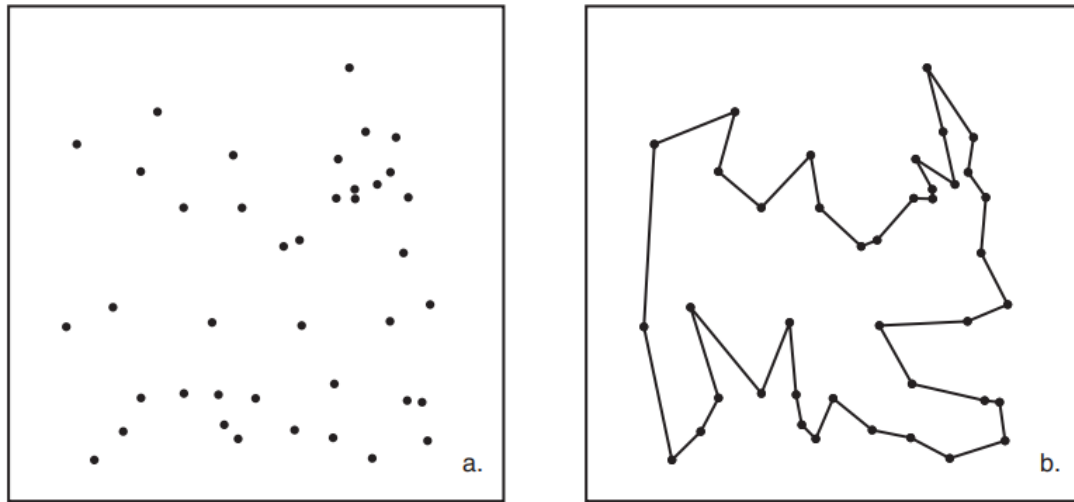
On a Graph vs On a Plane



$$a + b \geq c$$

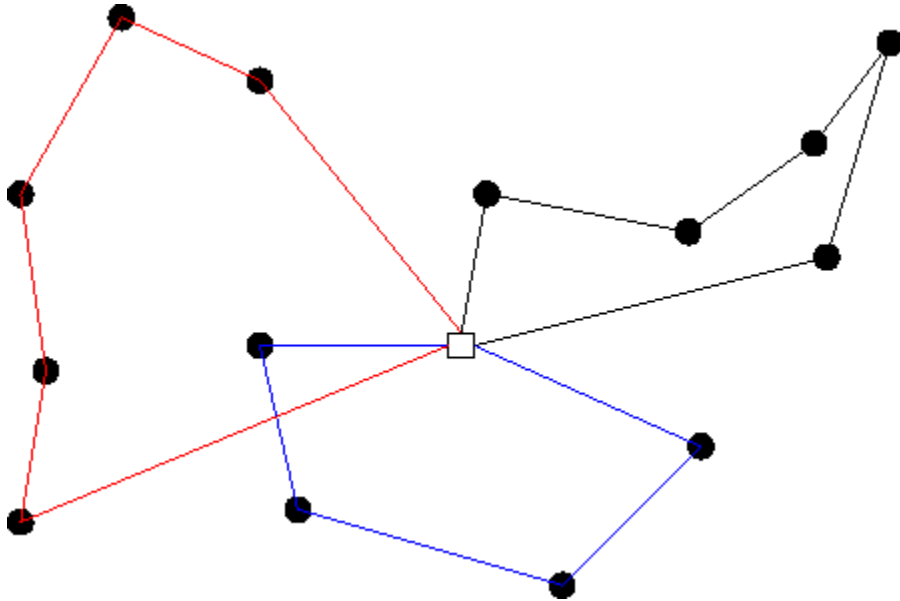
- There is an important distinction between cases that obey a triangle inequality and those that do not
- Performance of some algorithms can be arbitrarily bad on graphs that do not adhere to a triangle inequality

Human Solutions



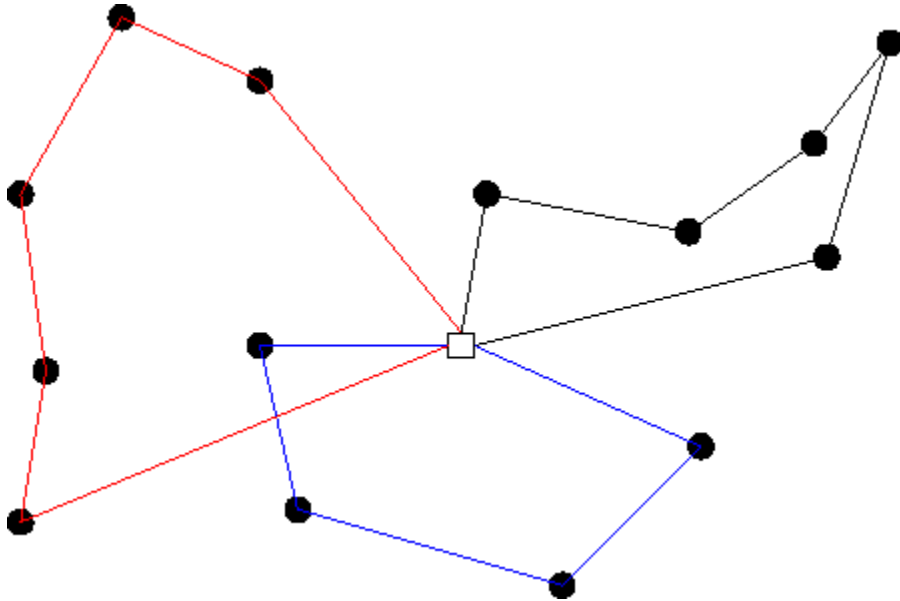
- Human's perform fairly well on the Euclidian TSP finding near optimal solutions for problems with 10-20 nodes
- For problems with more than 70 nodes the error is around 10%
- <https://citeseerx.ist.psu.edu/viewdoc/download;jsessionid=069B20F8000AD6C3D4100E5868F9B32F?doi=10.1.1.360.9763&rep=rep1&type=pdf>

TSP as a special case of VRP



- The Vehicle Routing Problem is a more generalized version of the TSP that involves minimizing the distance of multiple vehicles traveling between the vertices
- The TSP is just a VRP with a single truck
- VRP can be broken into a clustering (which nodes are visited by the same vehicle) and TSP problem (what order are those nodes visited)

TSP as a special case of VRP



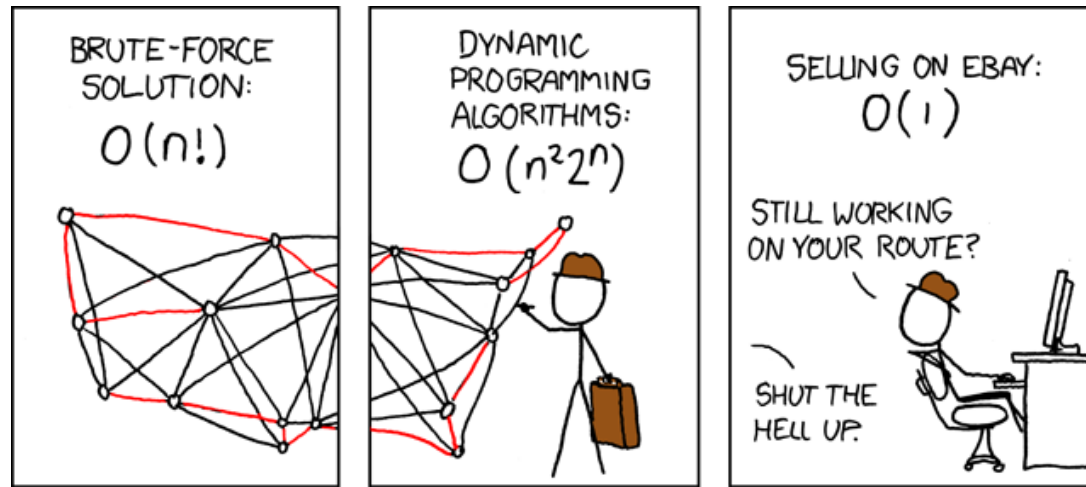
- Transportation accounts for about 10% of the cost of most items
- Small algorithmic improvements can lead to big changes as UPS, USPS, FedEx, etc control thousands of vehicles that use millions of gallons of gasoline
- US uses 390 million gallons of gas per day, a 0.01% improvement in routing equates to over 128,000 tons less CO₂ in a year

Many, many VRP Variants



- Multi-Depot
- Capacity constraints
- Route-length constraints
- Time-windows
- Uncertainty in nodes serviced
- Mixed Fleet
- ...

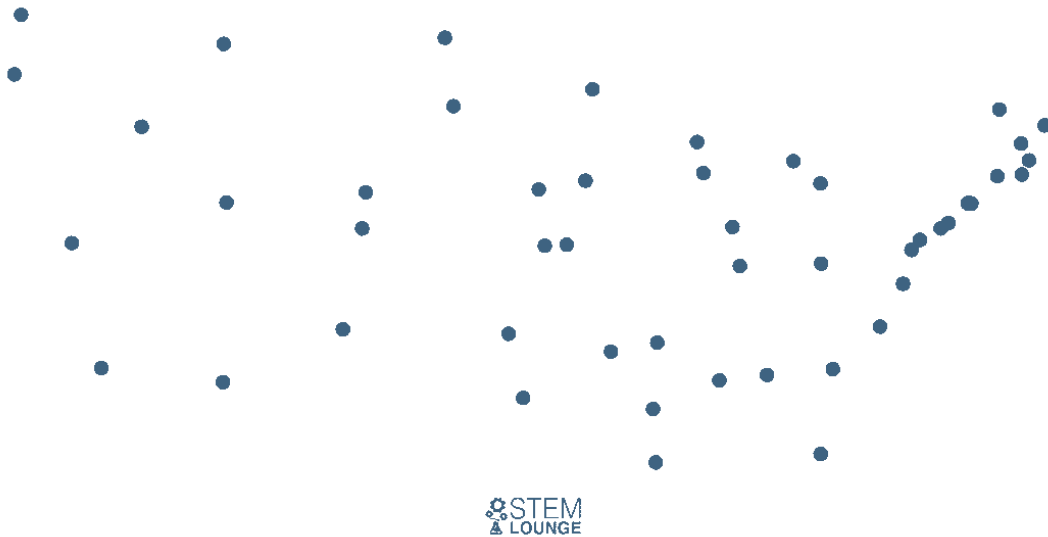
Complexity



- The general problem of finding the optimal solution is NP-hard
- The decision problem of, does there exist a route shorter than length x , is NP-complete

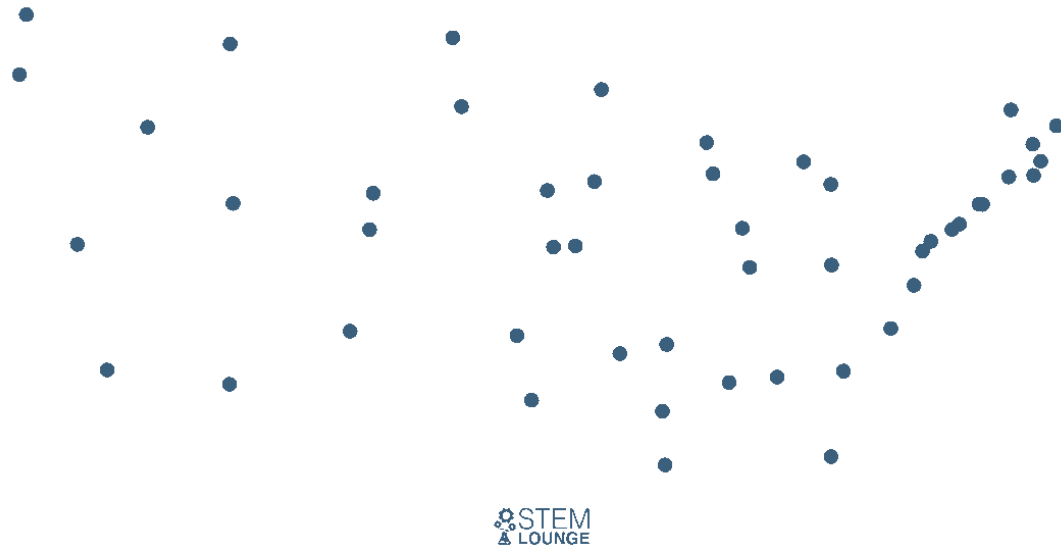
- <https://xkcd.com/399/>

Greedy Algorithm



- Add the edge x_{ij} , with the minimum length c_{ij} to the solution
- Remove any edges from the candidate set that would create a cycle or have more than two edges attached to a node
- Repeat until no edges left, connect two nodes with only one edge

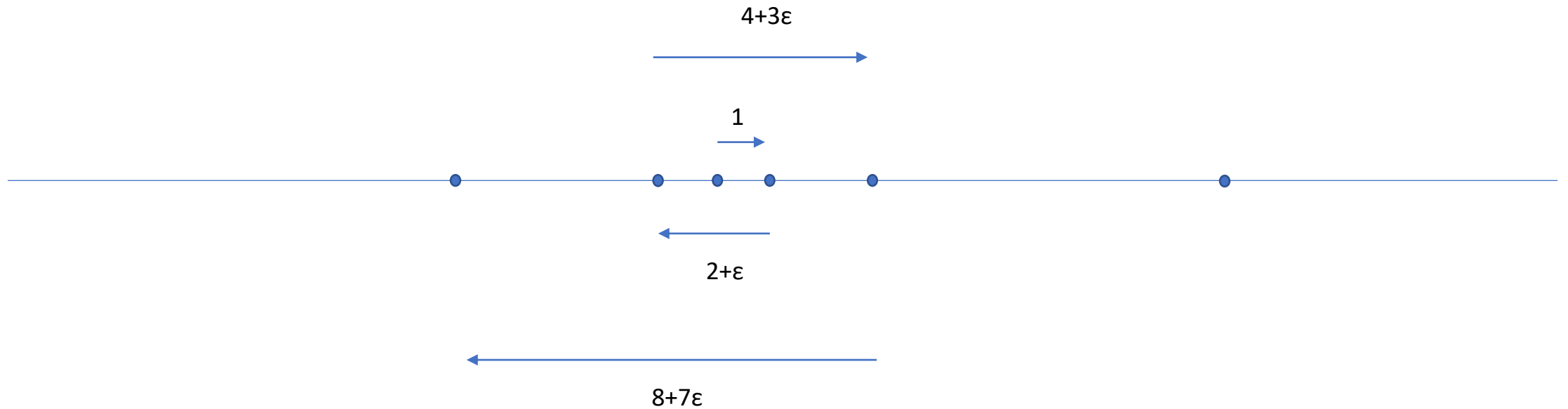
Nearest Neighbor



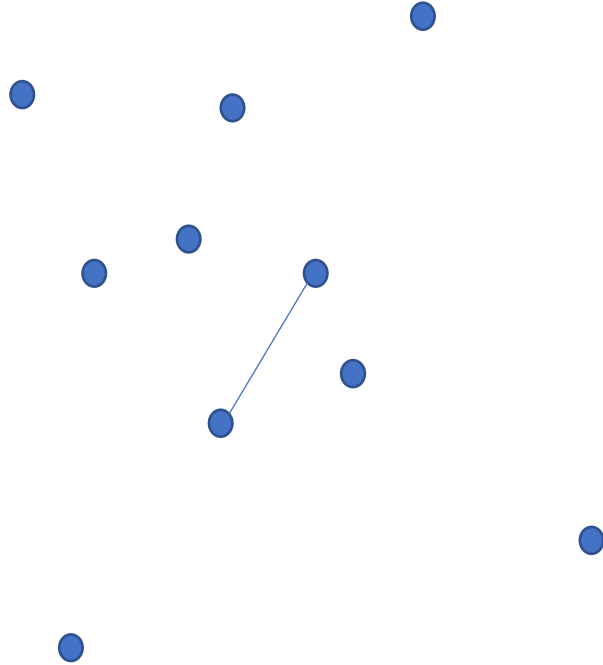
- Step 1: Choose a starting node
- Step 2: Find the next node is the closest node not in the path
- Step 3: Once all nodes have been visited, return to starting node

Worst Case Analysis

Given a bad starting node and the worst spacing of nodes possible, you can generate a worst-case scenario that shows how bad things can go

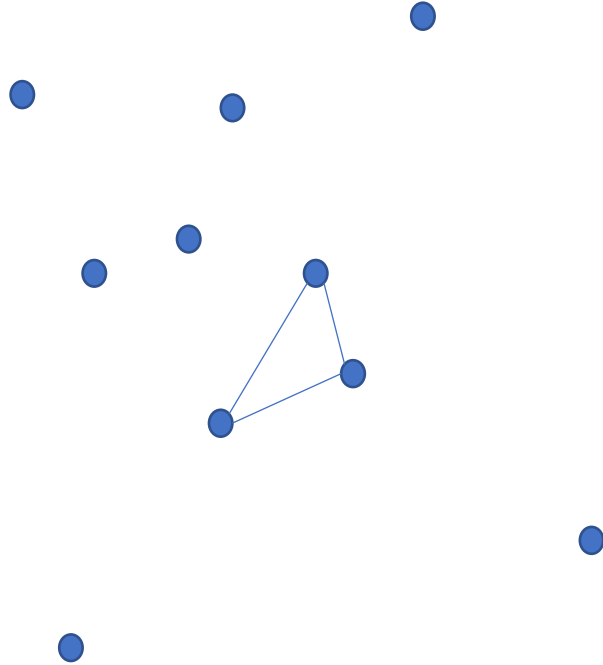


Nearest Insertion



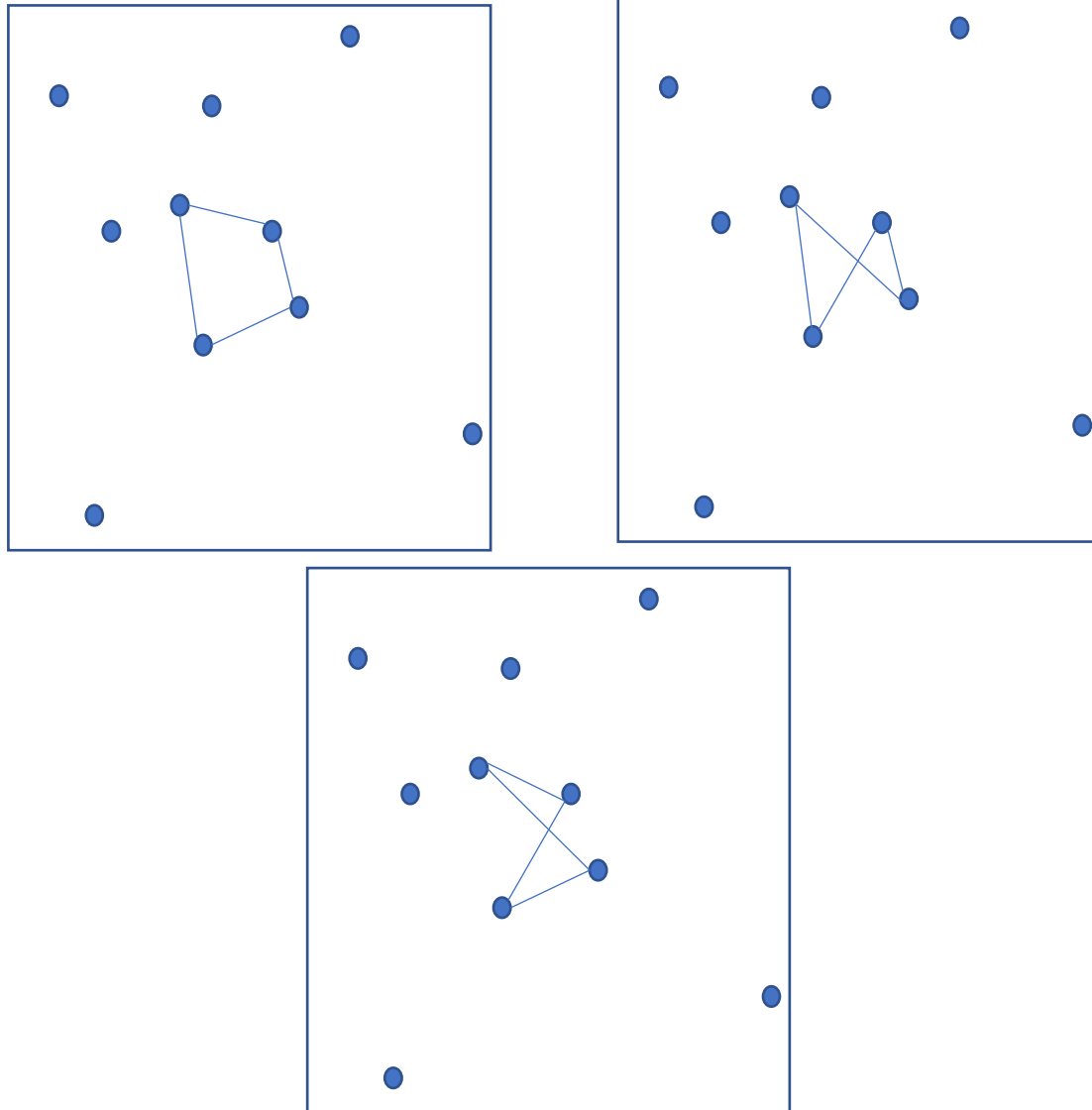
- Step 1: Choose 2 nodes to create a cycle
- Step 2: Add the closest node to the cycle in the way that minimizes the additional length,
$$c_{ik} + c_{kj} - c_{ij}$$
- Step 3: Repeat Step 2 until all nodes are added

Nearest Insertion



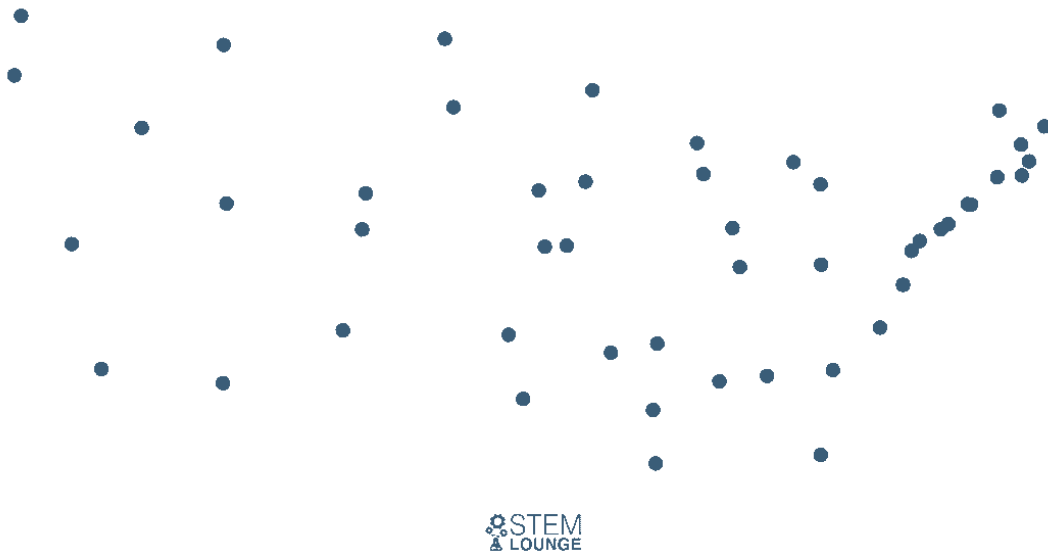
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Nearest Insertion



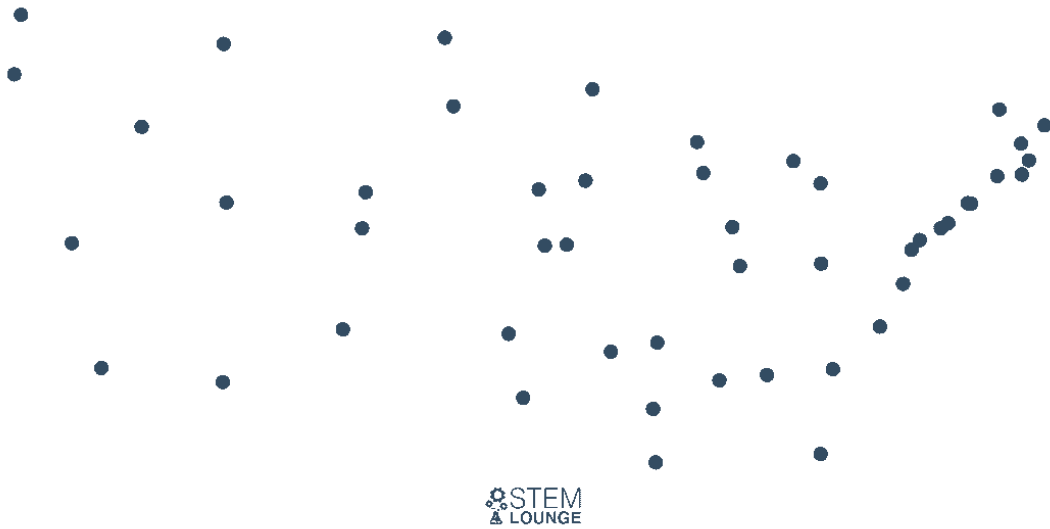
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Nearest Insertion



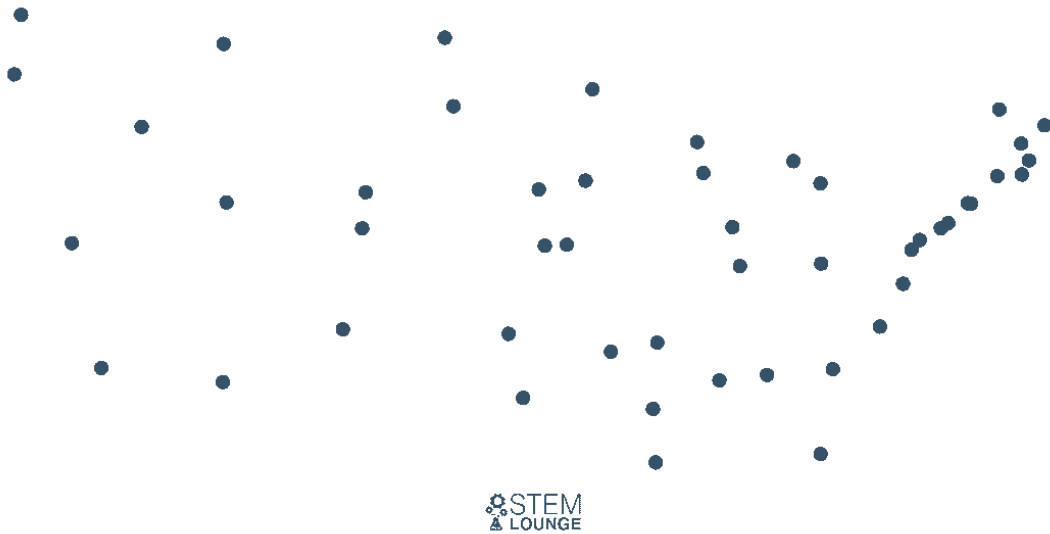
- Step 1: Choose 2 nodes to create a cycle
- Step 2: Add the closest node to the cycle in the way that minimizes the additional length,
$$c_{ik} + c_{kj} - c_{ij}$$
- Step 3: Repeat Step 2 until all nodes are added

Farthest Insertion



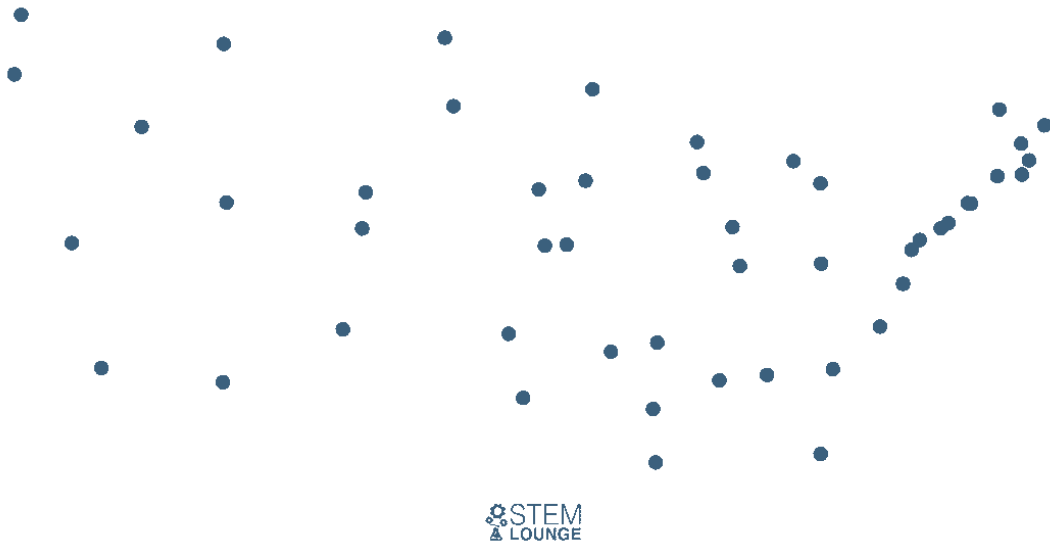
- Step 1: Choose a node to start
- Step 2: Add the farthest node to the cycle in the way that minimizes the additional length,
$$c_{ik} + c_{kj} - c_{ij}$$
- Step 3: Repeat Step 2 until all nodes are added

Random Insertion



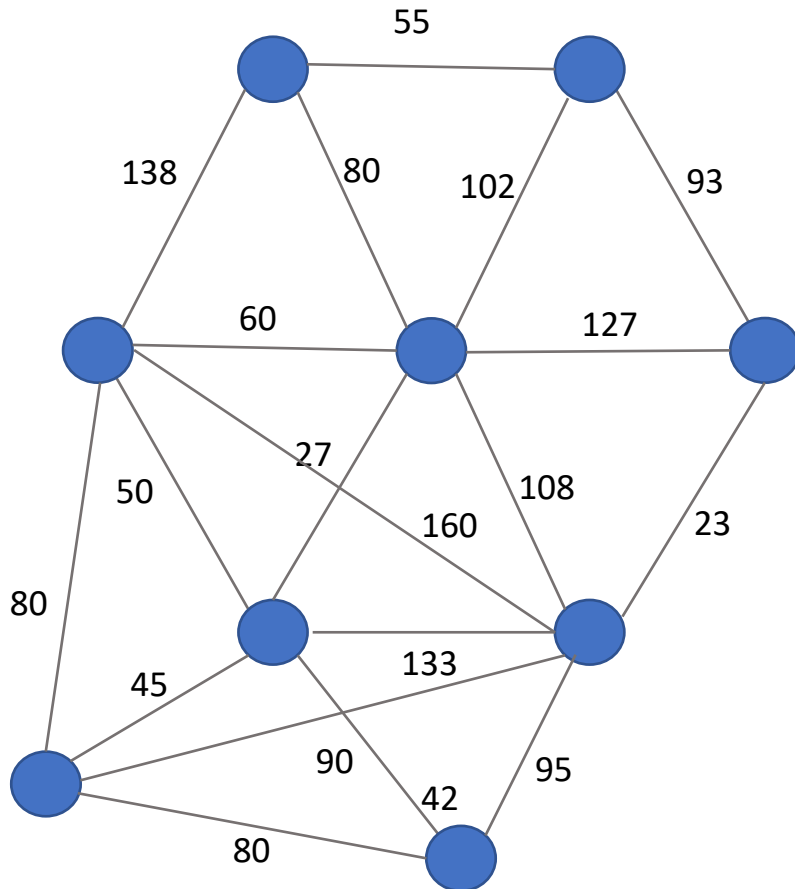
- Step 1: Choose a node to start
- Step 2: Add a random node to the cycle in the way that minimizes the additional length,
$$c_{ik} + c_{kj} - c_{ij}$$
- Step 3: Repeat Step 2 until all nodes are added

Cheapest Insertion



- Step 1: Start with a subgraph of one node, i
- Step 2: Find node such that c_{ik} is minimized and add to tour
- Step 3: Find (i,j) in subtour and k not, such that $c_{ik} + c_{kj} - c_{ij}$ is minimized
- Step 4: Repeat until finished

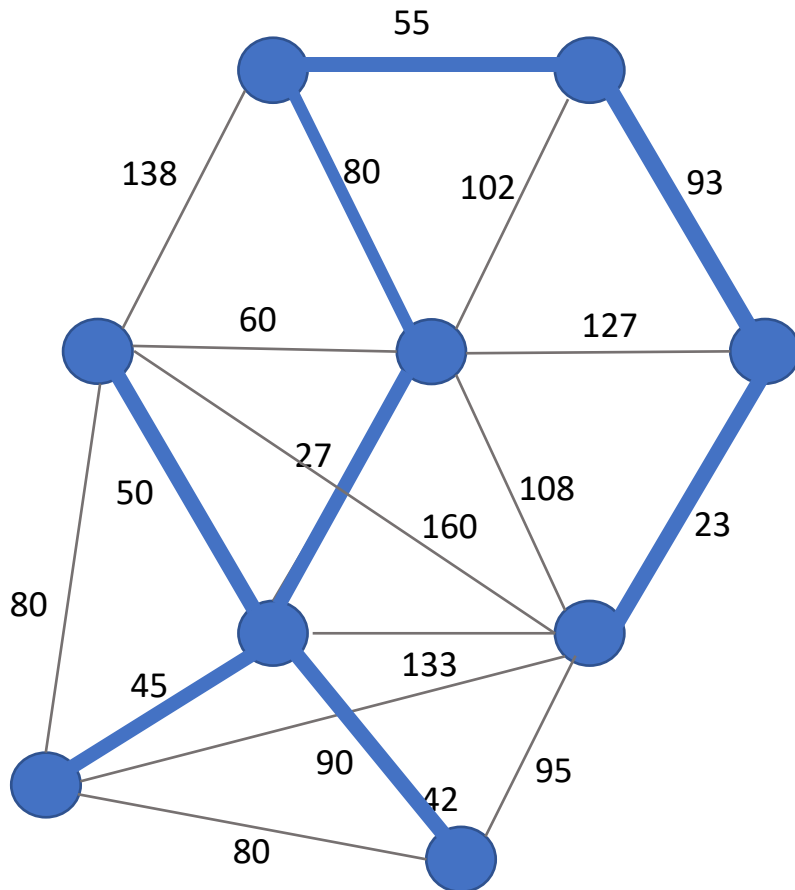
Christofides' Algorithm



- Step 1: Find minimal spanning tree
- Step 2: Identify all odd degree nodes and optimally match adding links to the graph
- Step 3: Draw an Eulerian circuit on the resulting graph, removing repeated vertices to result in a Hamiltonian circuit

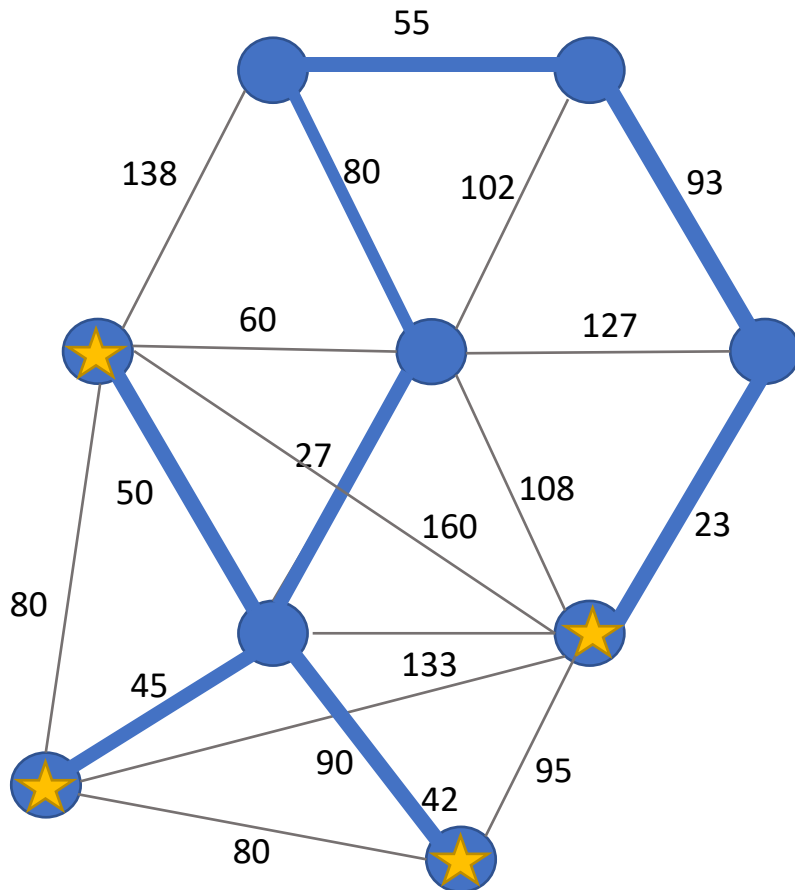
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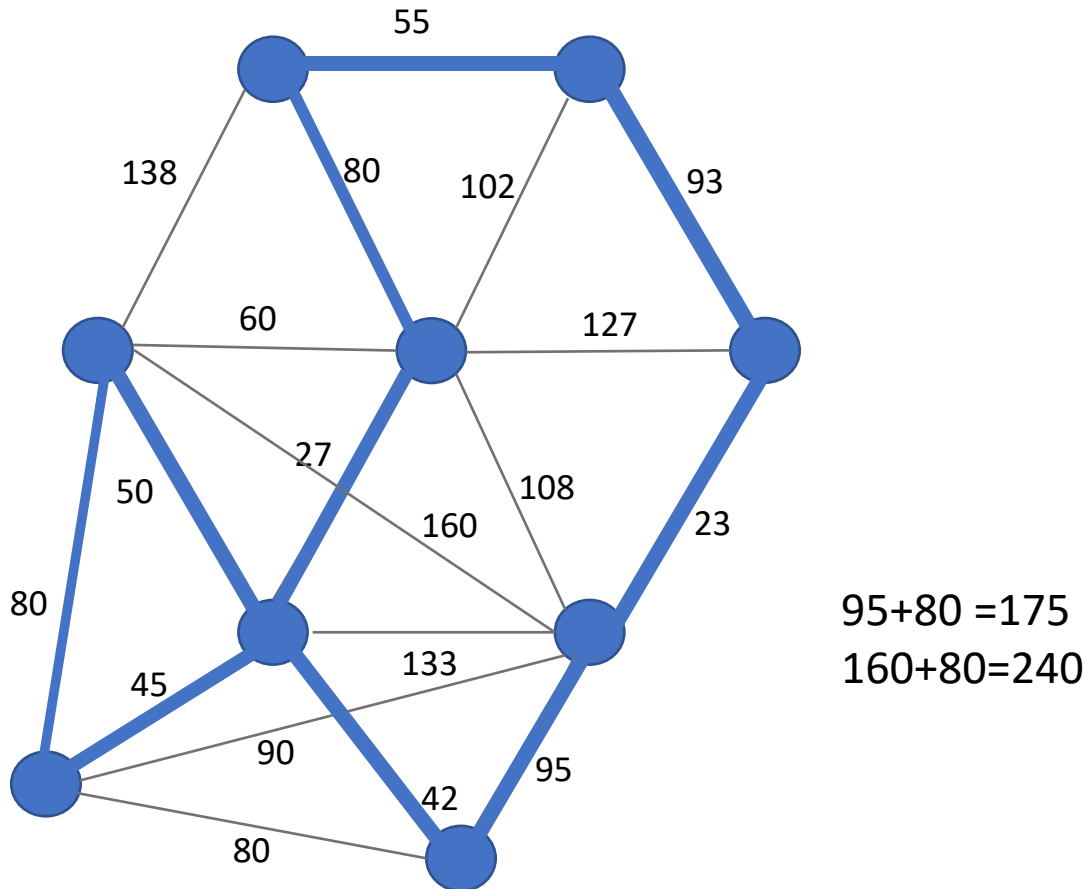
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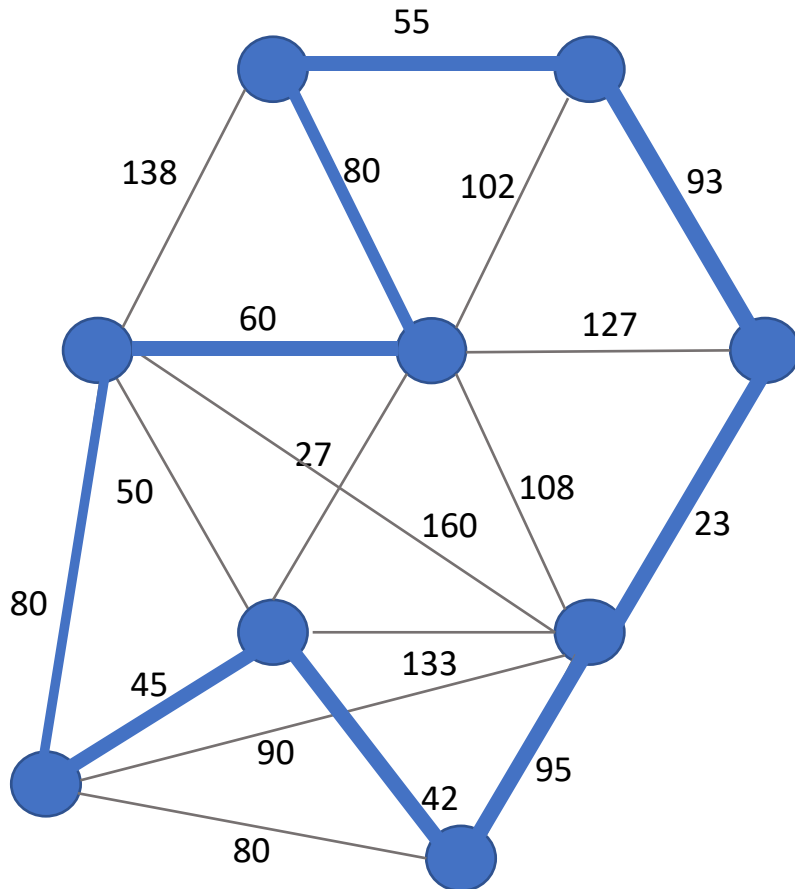


Christofides' Algorithm

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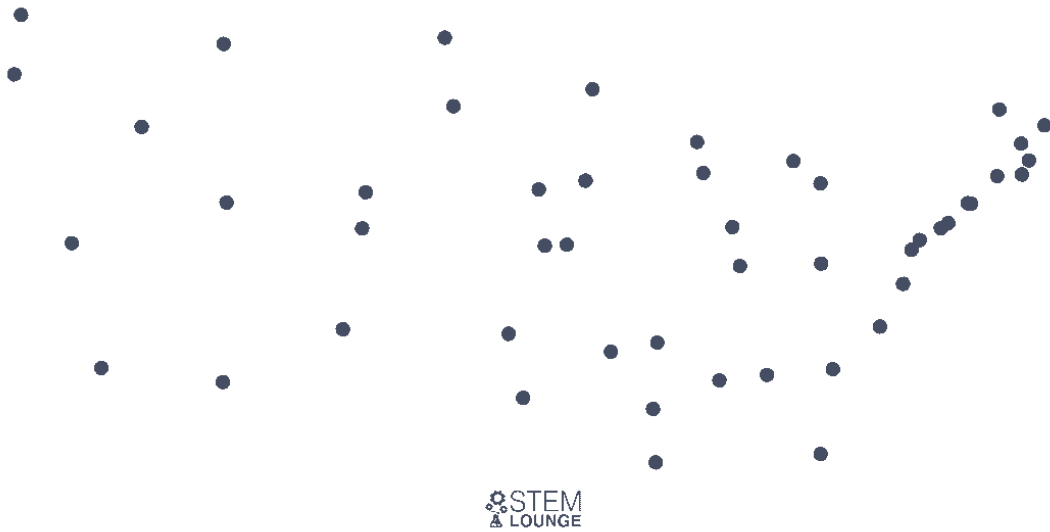
Christofides' Algorithm



$$\begin{aligned}80 - 45 - 42 &= -7 \\60 - 50 - 27 &= -17\end{aligned}$$

- Step 1: Find minimal spanning tree
- Step 2: Identify all odd degree nodes and optimally match adding links to the graph
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Christofides' Algorithm



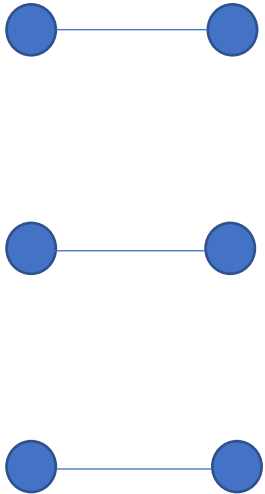
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Christofides' Algorithm

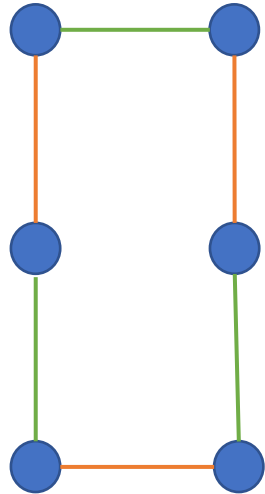
- Theorem: The Length of the Christofides' Solution, $l(\phi_c)$, divided by the optimal solution, $l(\phi_*)$, is strictly less than 1.5
- Let T be the minimal spanning tree of graph G with length $l(T)$
- $l(T) < l(\phi_*)$
- There are an even number of odd nodes in the spanning tree, and let $l(M_0)$ be the minimum matching of the odd degree nodes

Christofides' Algorithm

M_0



ϕ_0



Note that ϕ_0 contains two sets of connections between the nodes of M_0 and since M_0 is the minimal connection we know $l(M_0) \leq l(\phi_0)/2$

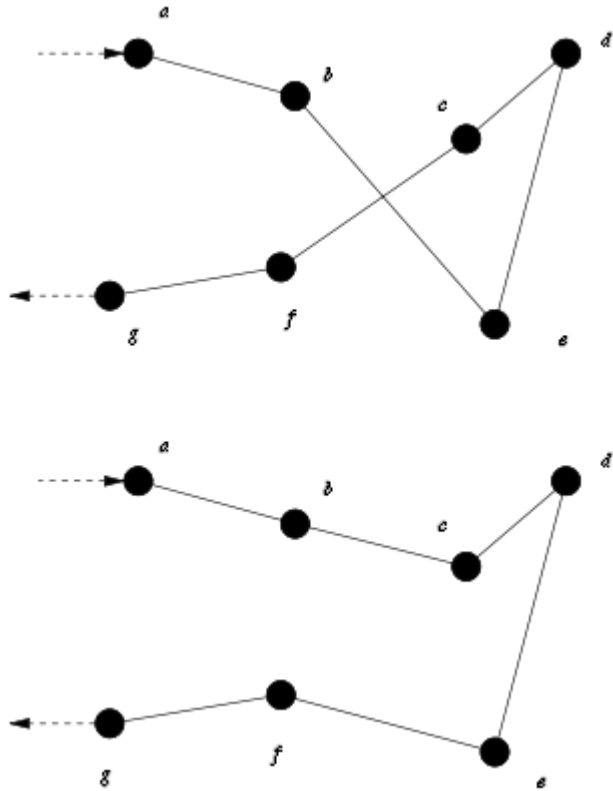
- $l(M_0) \leq l(\phi_0)/2$ (where ϕ_0 is the optimal TSP tour over the nodes in M_0)
- And since M_0 is a subset of points we know that $l(M_0) \leq l(\phi_0)/2 \leq l(\phi_*)/2$
- Thus worst case the optimal solution

$$l(\phi_c) \leq l(T) + l(M_0) \leq l(\phi_*) + l(\phi_*)/2$$

Christofides' Algorithm

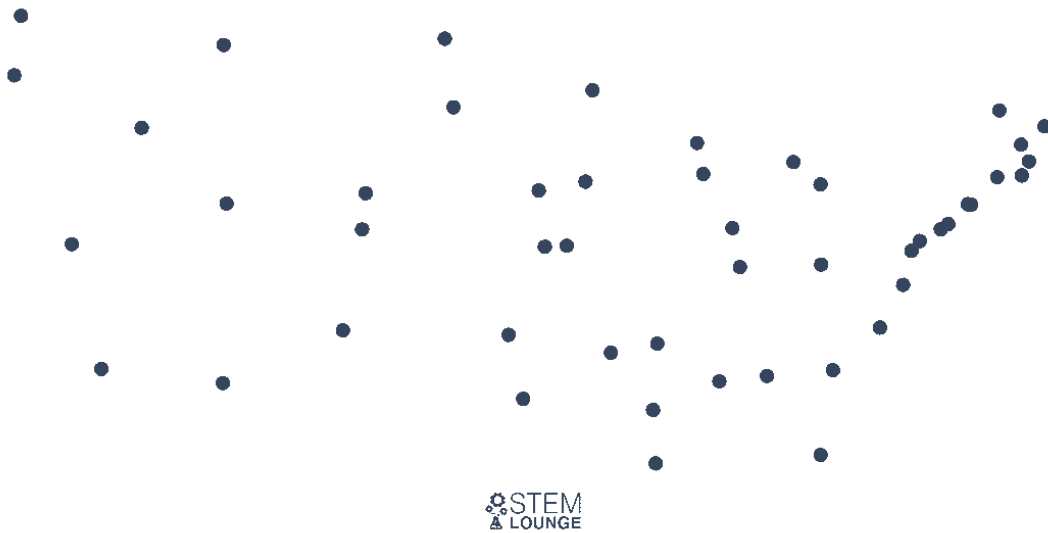
- [illegible]

2-opt, 3-opt, ...



- Is a local search algorithm for improving a solution to the TSP
- Takes the old route order:
 - [a,b,e,d,c,f,g] and cuts a section
 - [a,b, (e,d,c), f,g] reversing it
 - [a,b,c,d,e,f,g] and adding it back
- Only improvements are kept

2-opt, 3-opt, ...



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Complexity of Various Heuristics

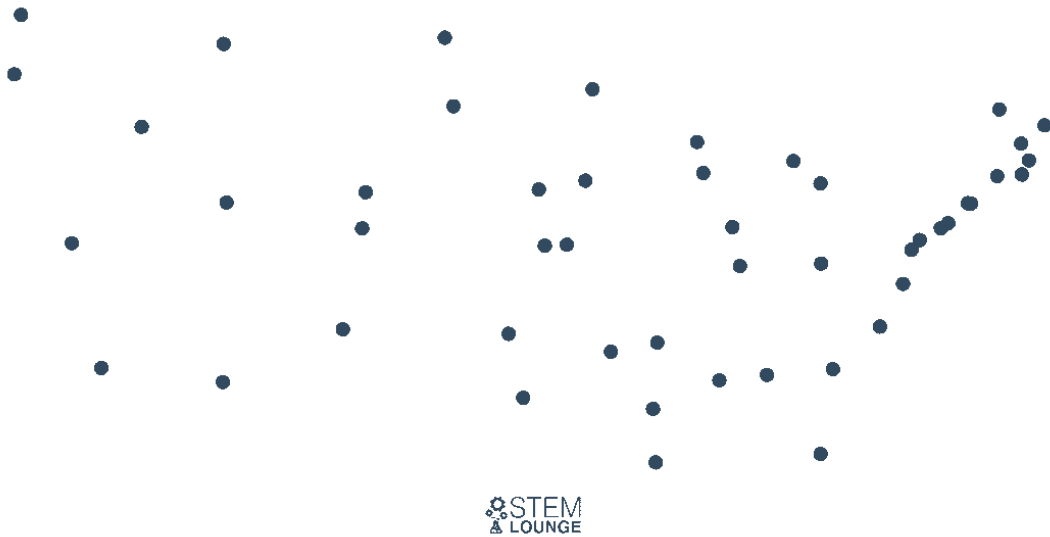
Heuristic	Complexity
Nearest Neighbor	$O(n^2)$
Nearest Insertion	$O(n^2)$
Cheapest Insertion	$O(n^2 \log_2 n)$
Christofides'	$O(n^4)$
Random Insertion	$O(n^2)$
Farthest Insertion	$O(n^2)$
2-opt	$O(n^2)$
3-opt	$O(n^3)$

Complexity \neq Better Performance

Heuristic	Avg % above Optimality
Nearest Neighbor	24.2
Nearest Insertion	20.0
Cheapest Insertion	16.8
Christofides'	19.5
Random Insertion	11.1
Farthest Insertion	9.9
2-opt	8.3
3-opt	3.8

- An article comparing the performance of 30 Euclidian TSPs from the literature with optimal solutions (ranging in size from 105 to 2392)

Iterated Lin-Kernighan



- A generalized form of the k-Opt procedure Lin-Kernighan is the most computationally complex heuristic in general use
- By iterating the process, it tends to result in solutions $<1\%$ from optimal (0.6% on average)