

# Minimum Spanning Tree

- Terminology and Properties
  - Prim's Algorithm
  - Kruskal's Algorithm
  - Baruvka's Algorithm
  - Traveling Salesperson Problem
- 
- A large, dark blue, curved shape that starts from the bottom left and extends diagonally upwards towards the right, covering the bottom half of the slide.

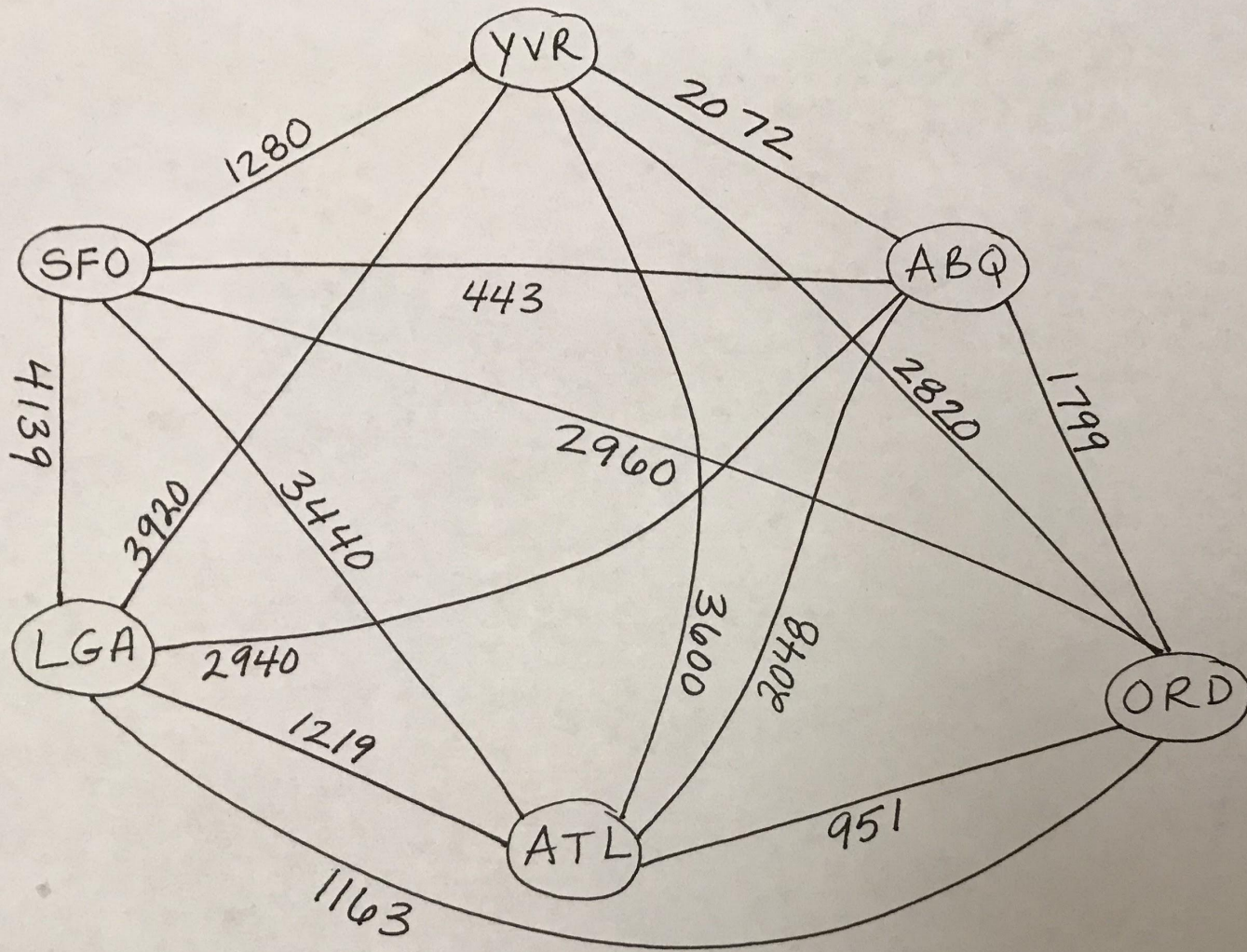
# Imagine...

You are employed at a brand new airline, and your first task is to determine which flights the airline should provide according to the following guidelines:

- There needs to be a way to get between each pair of cities/airports in the following list: San Francisco, CA (SFO); Vancouver, BC (YVR); Albuquerque, NM (ABQ); Chicago, IL (ORD); Atlanta, GA (ATL); New York, NY (LGA)
- The route between two cities does not have to be a direct flight.
- The total number of flights should be minimized.
- The total distance covered by all flights should be minimized.
- Assume that if a flight exists, it goes both ways (e.g. SFO→YVR and YVR→SFO)

How would you go about solving this problem?

	SFO	YVR	ABQ	ORD	ATL	LGA
SFO	0	1280	443	2960	3440	4139
YVR	1280	0	2072	2820	3600	3920
ABQ	443	2072	0	1799	2048	2940
ORD	2960	2820	1799	0	951	1163
ATL	3440	3600	2048	951	0	1219
LGA	4139	3920	2940	1163	1219	0



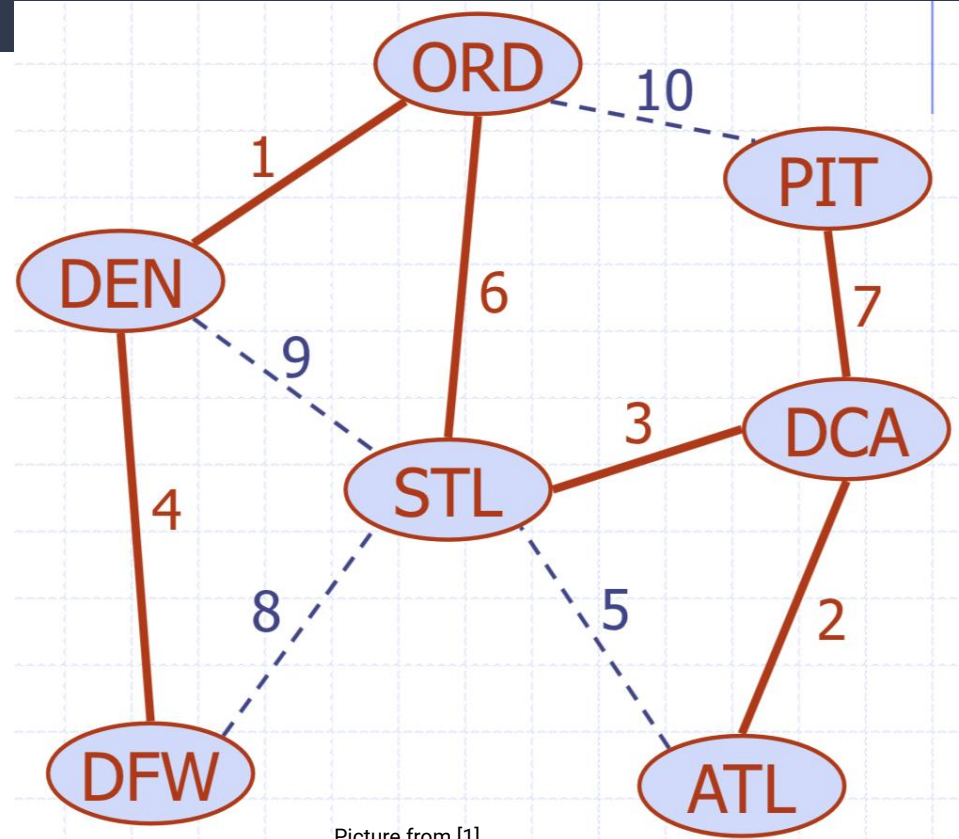
How does this problem differ from a shortest path problem?

In graph terms, how would you describe what we are looking for?

# What we really want is a minimum spanning tree (MST)...

A minimum spanning tree of a graph G...

- ...is a spanning subgraph (i.e. it contains all the vertices of G)
- ...is a tree (i.e. it has no cycles)
- ...is minimal weight-wise (i.e. the total weight of all the edges it uses is minimized so that no other spanning tree has a smaller total weight)
- ...does NOT guarantee the shortest path between two vertices!



Picture from [1]

# Property 1: The Cycle Property

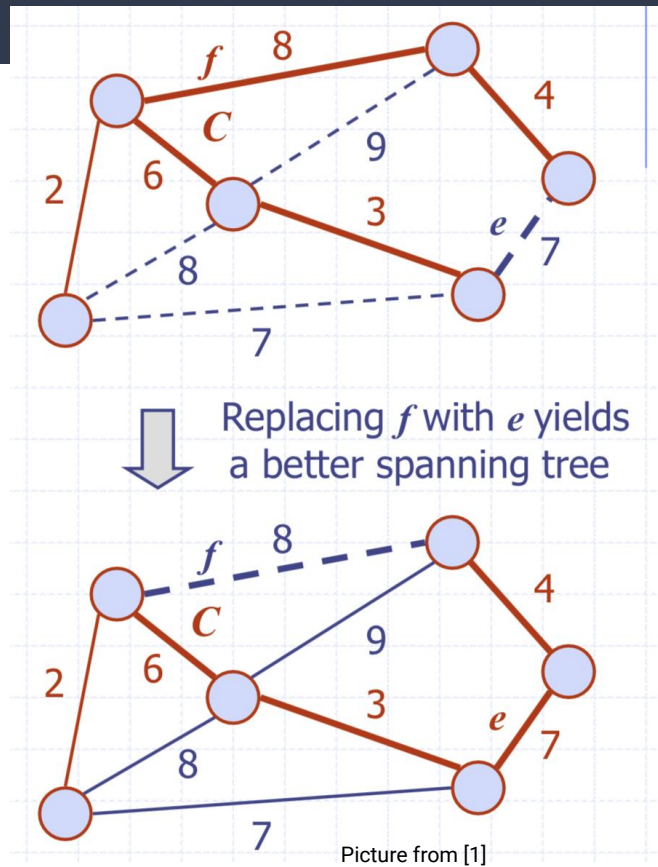
- Let **T** be a minimum spanning tree of a weighted graph **G**
- Let **e** be an edge of **G** that is not in **T**
- Let **C** be the cycle formed by adding **e** to **T**
- Claim: For every edge **f** of **C**:  
 $\text{weight}(f) \leq \text{weight}(e)$

**Proof by Contradiction: ???**

# Property 1: The Cycle Property

- Let  $T$  be a minimum spanning tree of a weighted graph  $G$
- Let  $e$  be an edge of  $G$  that is not in  $T$
- Let  $C$  be the cycle formed by adding  $e$  to  $T$
- Claim: For every edge  $f$  of  $C$ :  
 $\text{weight}(f) \leq \text{weight}(e)$

**Proof by Contradiction:** Given all the properties above, assume that for some edge  $f$  in  $C$ ,  $\text{weight}(f) > \text{weight}(e)$ . Then replacing  $f$  with  $e$  will produce a spanning tree  $T'$  such that the total weight of  $T'$  is smaller than the total weight of  $T$ . But that contradicts the definition of  $T$  as an MST of  $G$ .



# Property 2: The Partition Property

- Consider a partition of a graph  $\mathbf{G}$  into subsets  $\mathbf{U}$  and  $\mathbf{V}$
- Let  $\mathbf{e}$  be an edge of minimum weight across the partition (i.e.  $\mathbf{e}$  has one endpoint in  $\mathbf{U}$  and one endpoint in  $\mathbf{V}$ )
- Claim: There is a minimum spanning tree of  $\mathbf{G}$  that includes edge  $\mathbf{e}$

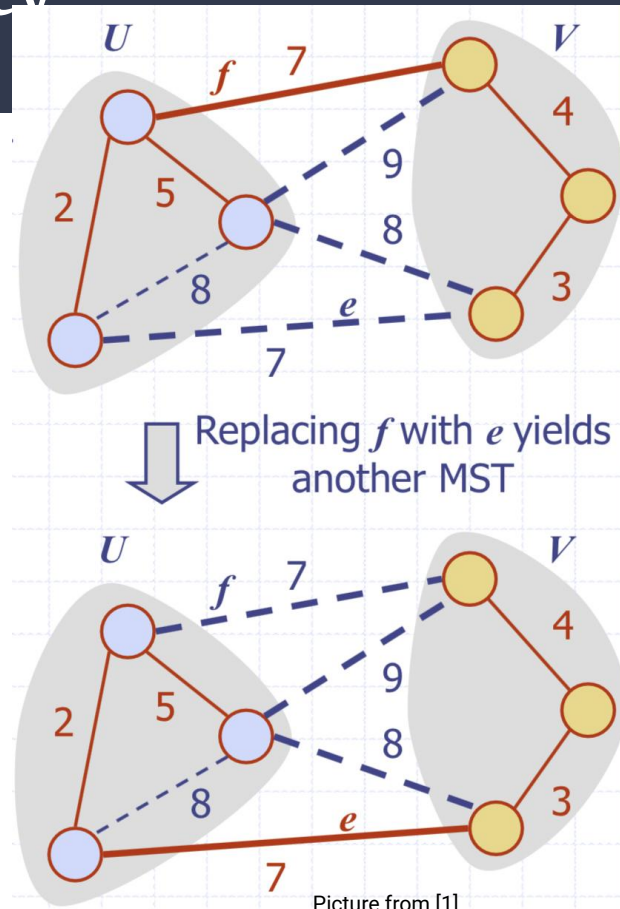
**Proof:** ???



# Property 2: The Partition Property

- Consider a partition of a graph  $G$  into subsets  $U$  and  $V$
- Let  $e$  be an edge of minimum weight across the partition (i.e.  $e$  has one endpoint in  $U$  and one endpoint in  $V$ )
- Claim: There is a minimum spanning tree of  $G$  that includes edge  $e$

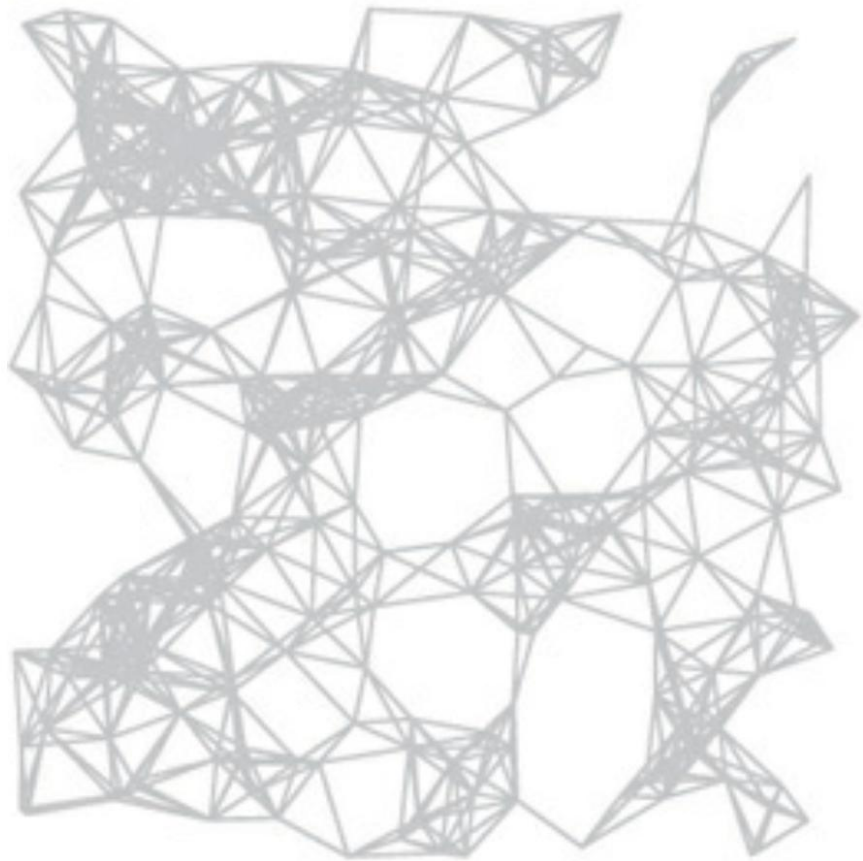
**Proof:** Let  $T$  be an MST of  $G$  and let the definitions above be true. If  $T$  does not contain  $e$ , consider the cycle  $C$  formed by  $e$  with  $T$  and let  $f$  be an edge of  $C$  across the partition. By the cycle property,  $\text{weight}(f) \leq \text{weight}(e)$ . Thus,  $\text{weight}(f) = \text{weight}(e)$ , and we obtain another MST by replacing  $f$  with  $e$ .



# Questions to think about...

1. Given an undirected, weighted graph  $\mathbf{G}$ , let  $\mathbf{T}$  be an MST of  $\mathbf{G}$ . Is  $\mathbf{T}$  unique?
2. Does an MST also give you the shortest path between a pair of vertices?
3. Does every weighted undirected graph have an MST?
4. Can you find an MST of a weighted undirected graph if there are negative weight edges?
5. Can you find an MST in a directed graph?
6. How would you find an MST of a weighted undirected graph?

graph



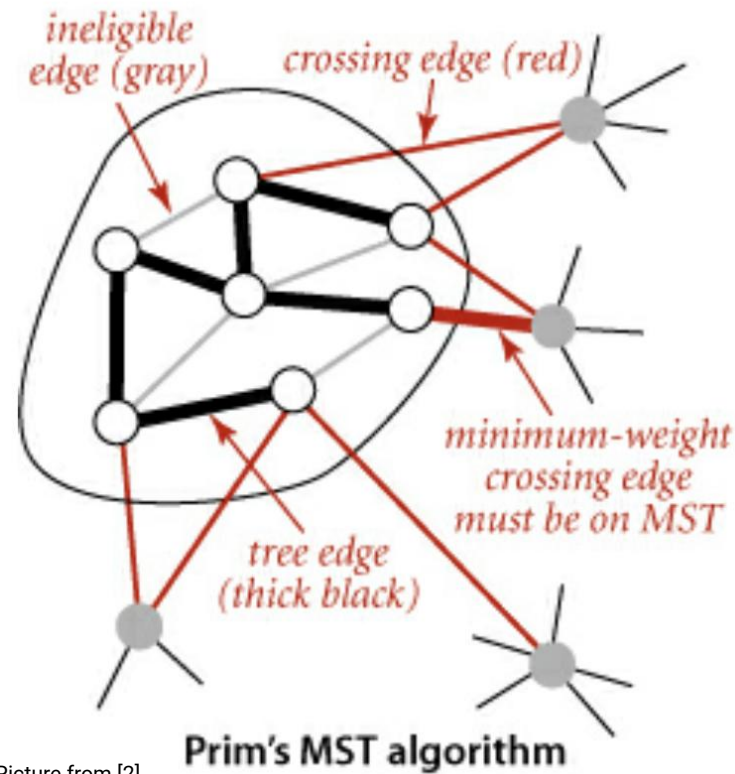
MST



A 250-vertex Euclidean graph (with 1,273 edges) and its MST

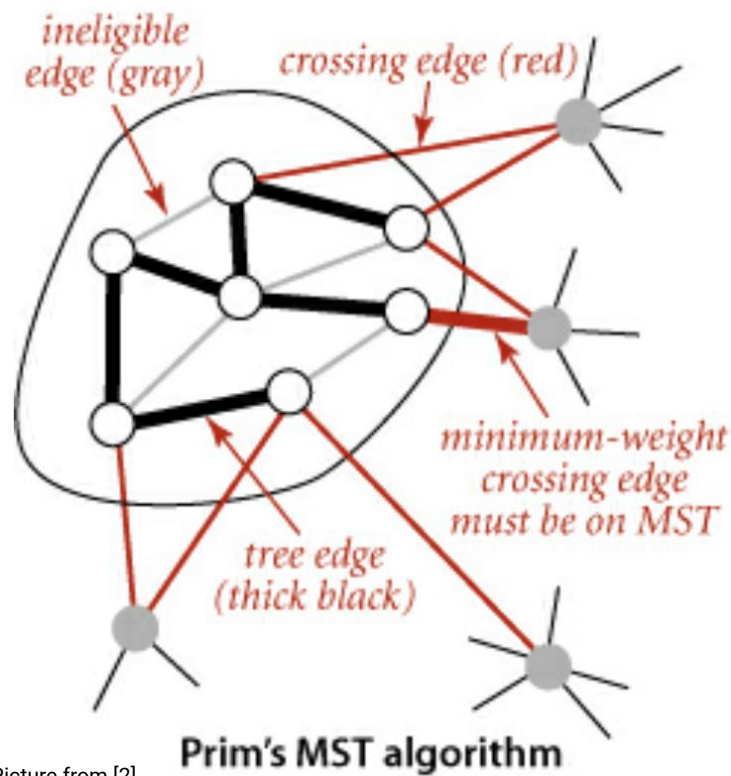
# Prim's Algorithm (aka Jarnik's Algorithm)

- Grow the tree one edge at a time
- Start with a single vertex (counts as a tree)
- Add  $|V| - 1$  edges to it by adding the minimum-weight edge that connects a new vertex to the tree (crossing the partition that is defined by the current tree vertices--called a *crossing edge*)
- NOTE: When you add an edge to the tree, you are also adding a vertex to the tree.
- Like Dijkstra's Shortest Path, this is a greedy algorithm.



# Prim's Algorithm (aka Jarnik's Algorithm)

**Proof of Correctness:** Follows directly from the Partition Property because we are choosing the minimum weight edge across the tree-defined partition.



# Implementation of Prim's Algorithm

- **marked[]**: an array of booleans to keep track of vertices on the tree
- **edgeTo[]**: an array to keep track of the lightest edge connecting a new vertex to the tree
- **distTo[]**: an array to keep track of the weight of the lightest edge connecting a new vertex to the tree
- **pq**: a minimum priority queue to keep track of eligible crossing edges (key is the weight of the edges)

### Algorithm *PrimsMST*( $G$ )

**Input:**  $G = (V, E)$ , a weighted, undirected graph

**Output:** A minimum-weight spanning tree of  $G$

$edgeTo[]$  := a  $|V|$ -sized array to store the edge connecting a vertex to the tree

$distTo[]$  := a  $|V|$ -sized array to keep track of the distance of the edge connecting a vertex to a tree

$marked[]$  := a  $|V|$ -sized array to keep track of which vertices have been visited

$pq$  := a heap-based min priority queue with weights as keys and vertices as values

//initialize structures

for all  $v \in V$ :

$distTo[v] = \infty$

$distTo[0] = 0$

$pq.insert(0, 0)$

//main loop

while ! $pq.isEmpty$ :

$v = pq.delMin()$

$marked[v] = T$

for each edge  $e = (v, u)$  adjacent to  $v$ :

if  $marked[u]$

continue

if  $e.weight() < distTo[u]$ :

$edgeTo[u] = e$

$distTo[u] = e.weight()$

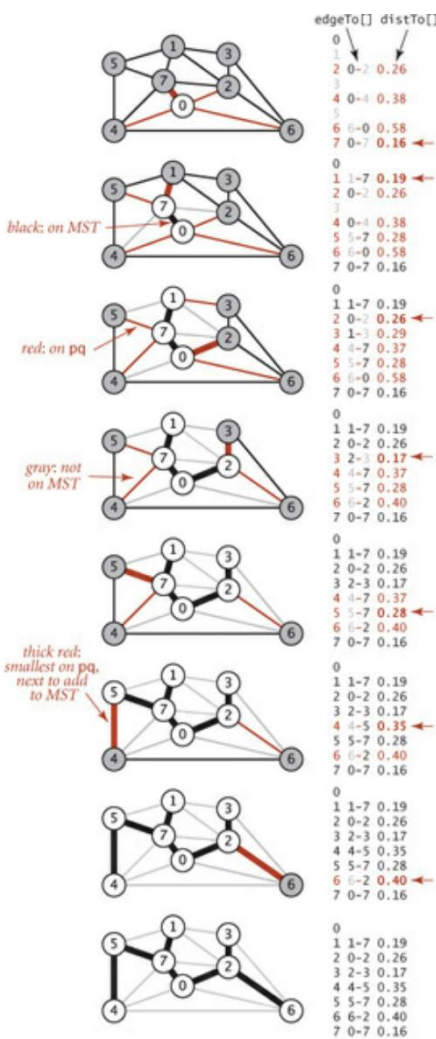
if  $pq.contains(u)$ :

$pq.changeKey(u, distTo[u])$

else:

$pq.insert(u, distTo[u])$





//main loop

while !pq.isEmpty:

$v = pq.delMin()$

$marked[v] = T$

for each edge  $e = (v, u)$  adjacent to  $v$ :

if  $marked[u]$

continue

if  $e.weight() < distTo[u]$ :

$edgeTo[u] = e$

$distTo[u] = e.weight()$

if  $pq.contains(u)$ :

$pq.changeKey(u, distTo[u])$

else:

$pq.insert(u, distTo[u])$



### Algorithm *PrimsMST*( $G$ )

**Input:**  $G = (V, E)$ , a weighted, undirected graph

**Output:** A minimum-weight spanning tree of  $G$

$edgeTo[]$  := a  $|V|$ -sized array to store the edge connecting a vertex to the tree

$distTo[]$  := a  $|V|$ -sized array to keep track of the distance of the edge connecting a vertex to a tree

$marked[]$  := a  $|V|$ -sized array to keep track of which vertices have been visited

$pq$  := a heap-based min priority queue with weights as keys and vertices as values

//initialize structures

for all  $v \in V$ :

$distTo[v] = \infty$

$distTo[0] = 0$

$pq.insert(0, 0)$

$O(|V|)$

//main loop

while ! $pq.isEmpty$ :

$v = pq.delMin()$

$marked[v] = T$

for each edge  $e = (v, u)$  adjacent to  $v$ :

if  $marked[u]$

continue

if  $e.weight() < distTo[u]$ :

$edgeTo[u] = e$

$distTo[u] = e.weight()$

if  $pq.contains(u)$ :

$pq.changeKey(u, distTo[u])$

else:

$pq.insert(u, distTo[u])$

```
//main loop
```

```
while !pq.isEmpty:
```

```
    v = pq.delMin()
```

```
    marked[v] = T
```

```
    for each edge e = (v, u) adjacent to v:
```

```
        if marked[u]
```

```
            continue
```

```
        if e.weight() < distTo[u]:
```

```
            edgeTo[u] = e
```

```
            distTo[u] = e.weight()
```

```
            if pq.contains(u):
```

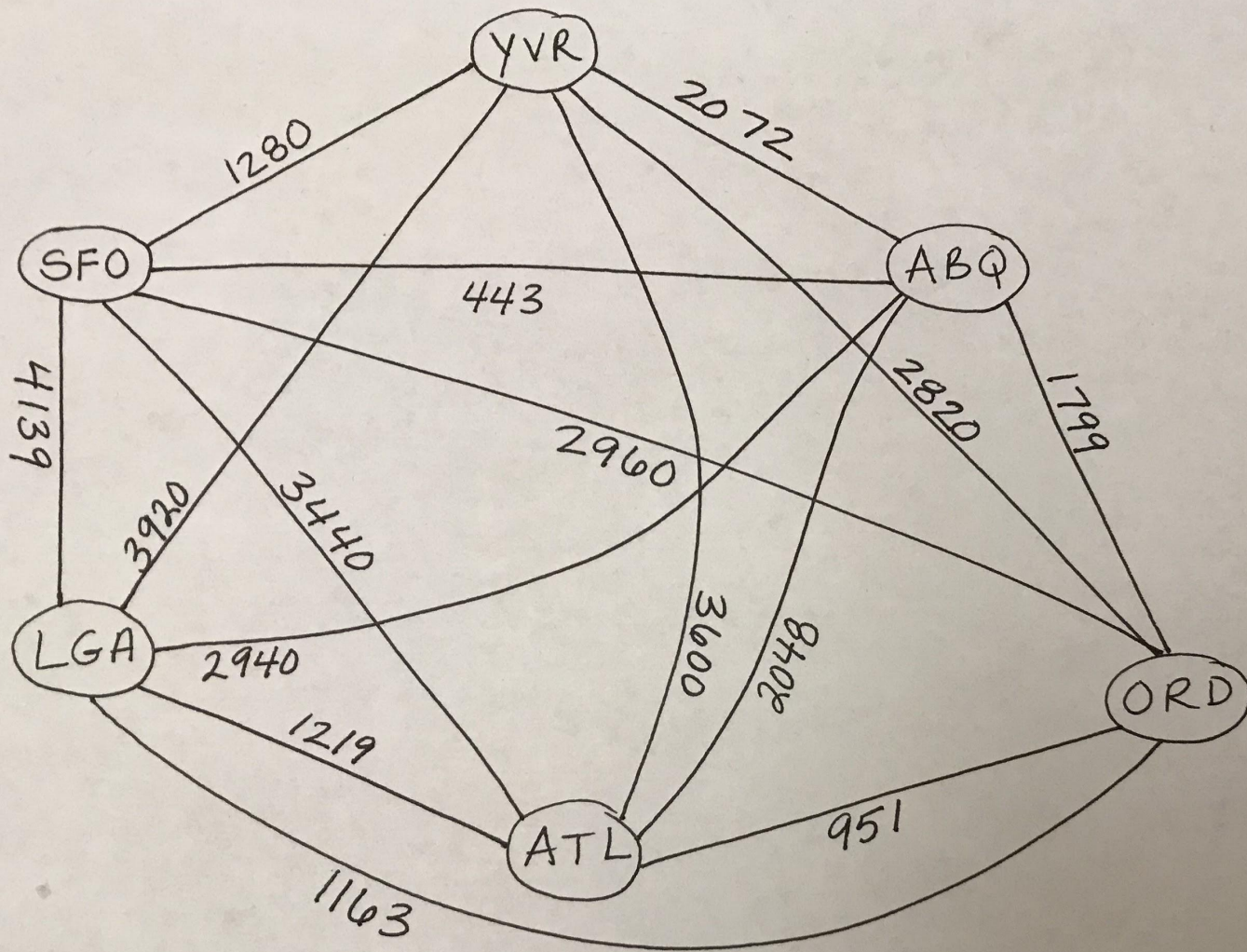
```
                pq.changeKey(u, distTo[u])
```

```
            else:
```

```
                pq.insert(u, distTo[u])
```

Assuming adjacency list representation, total time is  $O((V + E)\log V) = O(E\log V)$  for a connected graph

- delMin in a heap-based PQ:  $O(\log V)$
- checking if PQ contains a vertex and changing a key is  $O(\log V)$  as long as there is an auxiliary data structure keeping track of positions in the queue
- key of any vertex  $v$  is updated at most  $\deg(v)$  times and sum of all degrees is  $O(E)$



# Imagine...

You are employed at a company that has offices in several different cities. Your first task is to work with a phone company to ensure that there is a connection between every pair of offices according to the following guidelines:

- The connection between two offices does not have to be direct.
- The company wants to minimize the total cost of all the connections.

What kind of a problem is this?

# Imagine...

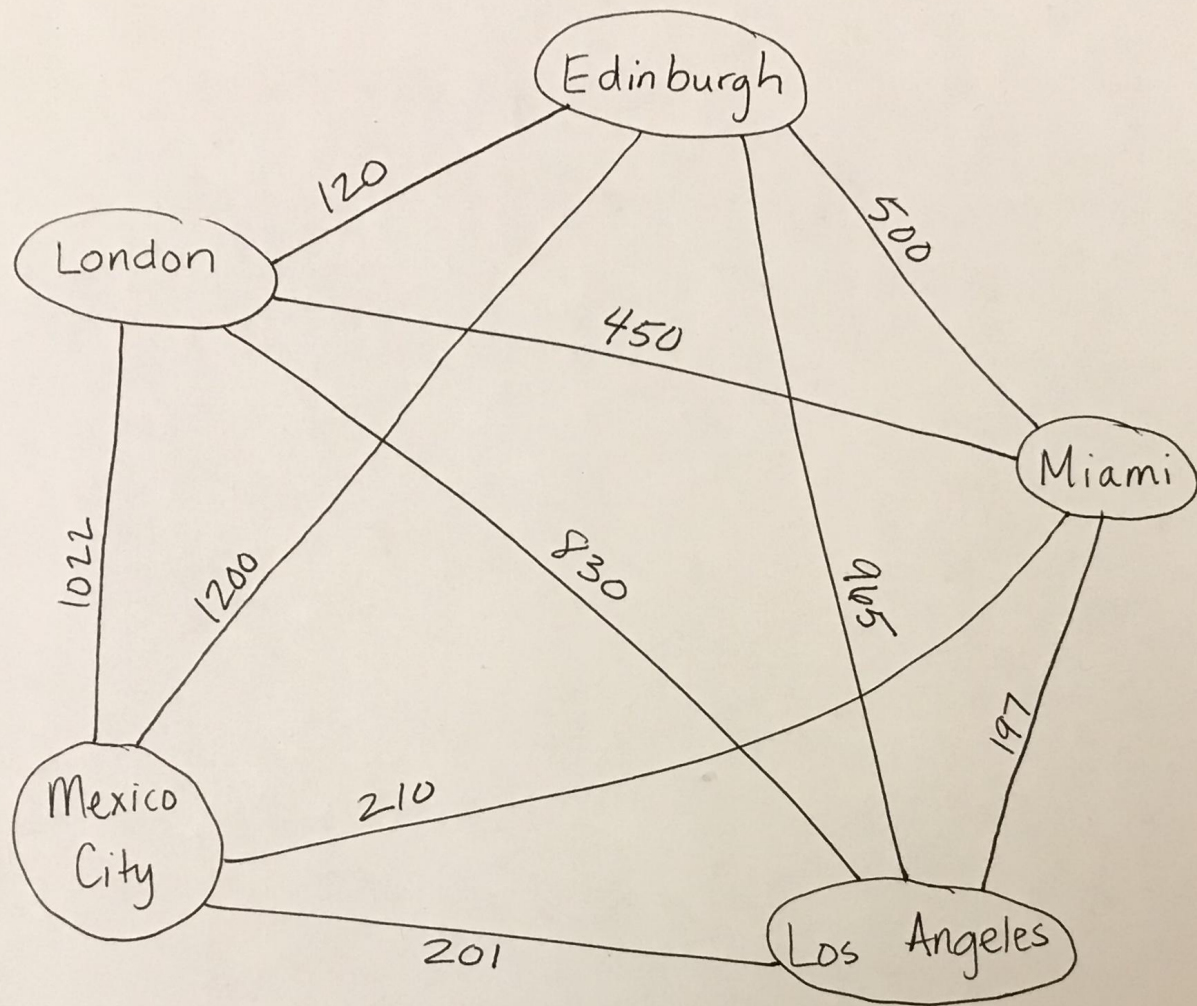
You are employed at a company that has offices in several different cities. Your first task is to work with a phone company to ensure that there is a connection between every pair of offices according to the following specifications:

- The phone company charges different rates to connect pairs of cities, specified on the following slide.
- The connection between two offices does not have to be direct.
- The company wants to minimize the total cost of all the connections.

What kind of a problem is this?

**A minimum spanning tree problem!**

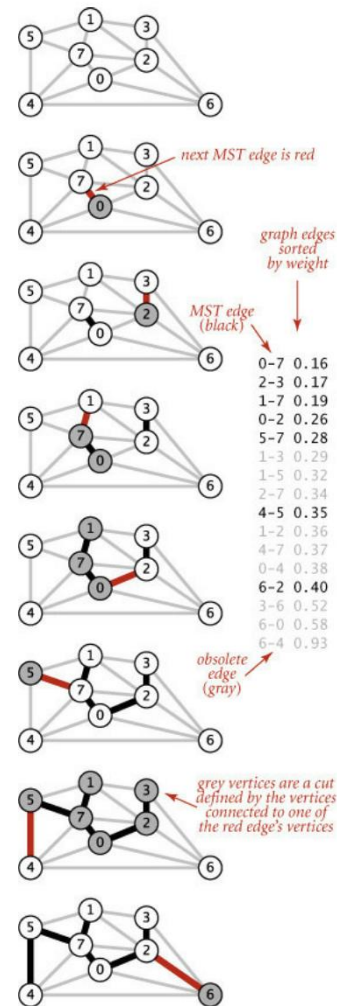
	London	Edinburgh	Miami	Mexico City	Los Angeles
London	0	\$120	\$450	\$1022	\$830
Edinburgh	\$120	0	\$500	\$1200	\$965
Miami	\$450	\$500	0	\$210	\$197
Mexico City	\$1022	\$1200	\$210	0	\$201
Los Angeles	\$830	\$965	\$197	\$201	0





# Kruskal's Algorithm

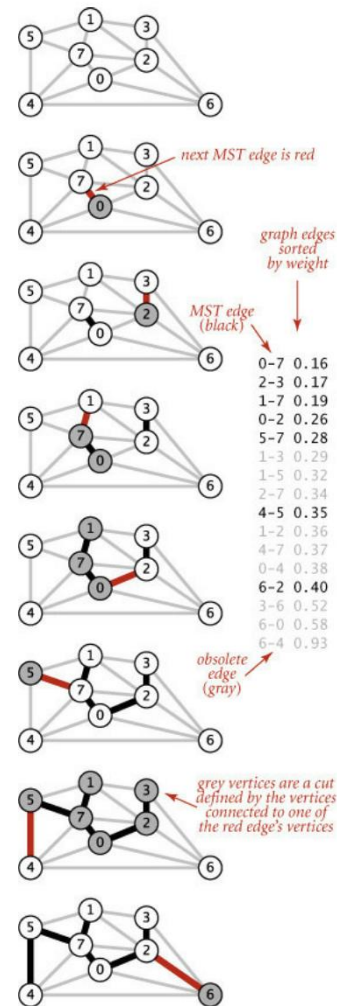
- Process edges, adding edges to the tree smallest-weight first **as long as the new edge does not form a cycle.**
- The result is that the algorithm creates a forest that eventually merges into a single tree.
- What would be some of the challenges in implementing this?
- What data structures would be useful?
- How would you represent the graph?





# Kruskal's Algorithm

- Can be implemented using disjoint sets for each separate component
- Start with  $|V|$  disjoint sets, each containing a single vertex
- Combine sets (union) by adding edges, reducing the number of separate components until there is only one



Trace of Kruskal's algorithm

### Algorithm *KruskalsMST*( $G$ )

**Input:**  $G = (V, E)$ , a connected, weighted, undirected graph

**Output:** A minimum-weight spanning tree of  $G$

$pq :=$  a minimum priority queue of edges where keys are weights

**for each** edge  $e = (u, v) \in E$ :

$pq.insert(e, e.weight())$

$A = \emptyset$

**for each**  $v \in V$ :

$makeSet(v)$

**while**  $|A| < |V| - 1$

$e = (u, v) = pq.delMin()$

**if**  $findSet(u) \neq findSet(v)$ :

$A = A \cup (u, v)$

$union(u, v)$

**return**  $A$

## Algorithm *KruskalsMST*( $G$ )

**Input:**  $G = (V, E)$ , a connected, weighted, undirected graph

**Output:** A minimum-weight spanning tree of  $G$

$pq :=$  a minimum priority queue of edges where keys are weights

**for each** edge  $e = (u, v) \in E$ :

$pq.insert(e, e.weight())$

$A = \emptyset$

**for each**  $v \in V$ :

$makeSet(v)$

**while**  $|A| < |V| - 1$

$e = (u, v) = pq.delMin()$

**if**  $findSet(u) \neq findSet(v)$ :

$A = A \cup (u, v)$

$union(u, v)$

**return**  $A$

- **$O(E \log E)$**  for doing  **$E$**  inserts into the  **$E$** -sized PQ
- Alternatively, we could just sort the edges in  **$O(E \log E)$**  time

## Algorithm *KruskalsMST*( $G$ )

**Input:**  $G = (V, E)$ , a connected, weighted, undirected graph

**Output:** A minimum-weight spanning tree of  $G$

$pq :=$  a minimum priority queue of edges where keys are weights

**for each** edge  $e = (u, v) \in E$ :

$pq.insert(e, e.weight())$

$A = \emptyset$

**for each**  $v \in V$ :

$makeSet(v)$

**while**  $|A| < |V| - 1$

$e = (u, v) = pq.delMin()$

**if**  $findSet(u) \neq findSet(v)$ :

$A = A \cup (u, v)$

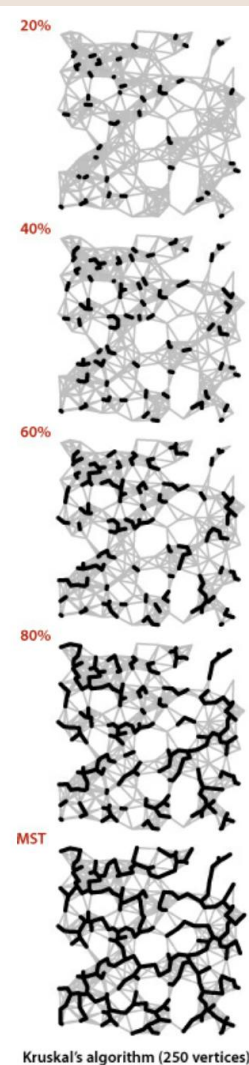
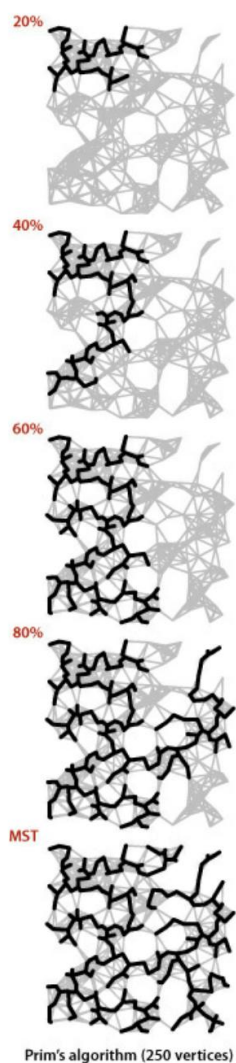
$union(u, v)$

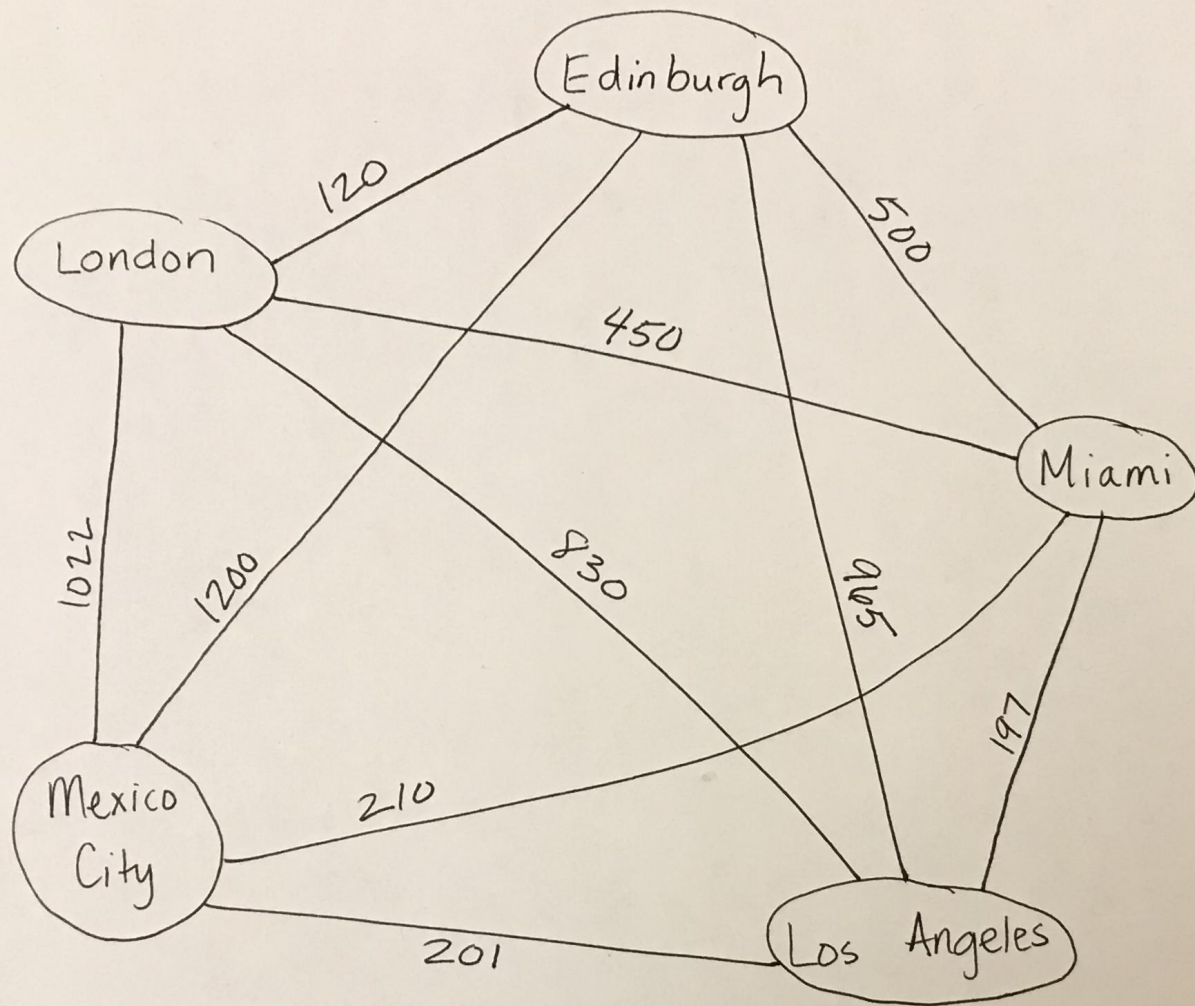
**return**  $A$

- $makeSet$  is an  $O(1)$  operation, so the total time is  **$O(V)$**
- The total time required for  **$E$  union** and *findSet* operations is  **$O(E \log V)$**
- **Total time:  $O(E \log E + E \log V) = O(E \log E)$**

# Prim's vs. Kruskal's

- Space:  $|V|$  vs.  $|E|$
- Time:  $|E|\log|V|$  vs.  $|E|\log|E|$





Goal: Find an MST of the graph using Kruskal's Algorithm...



# Baruvka's Algorithm

- ◆ Like Kruskal's Algorithm, Baruvka's algorithm grows many "clouds" at once.

**Algorithm** *BaruvkaMST*( $G$ )

$T \leftarrow V$  {just the vertices of  $G$ }

**while**  $T$  has fewer than  $n-1$  edges **do**

**for each** connected component  $C$  in  $T$  **do**

    Let edge  $e$  be the smallest-weight edge from  $C$  to another component in  $T$ .

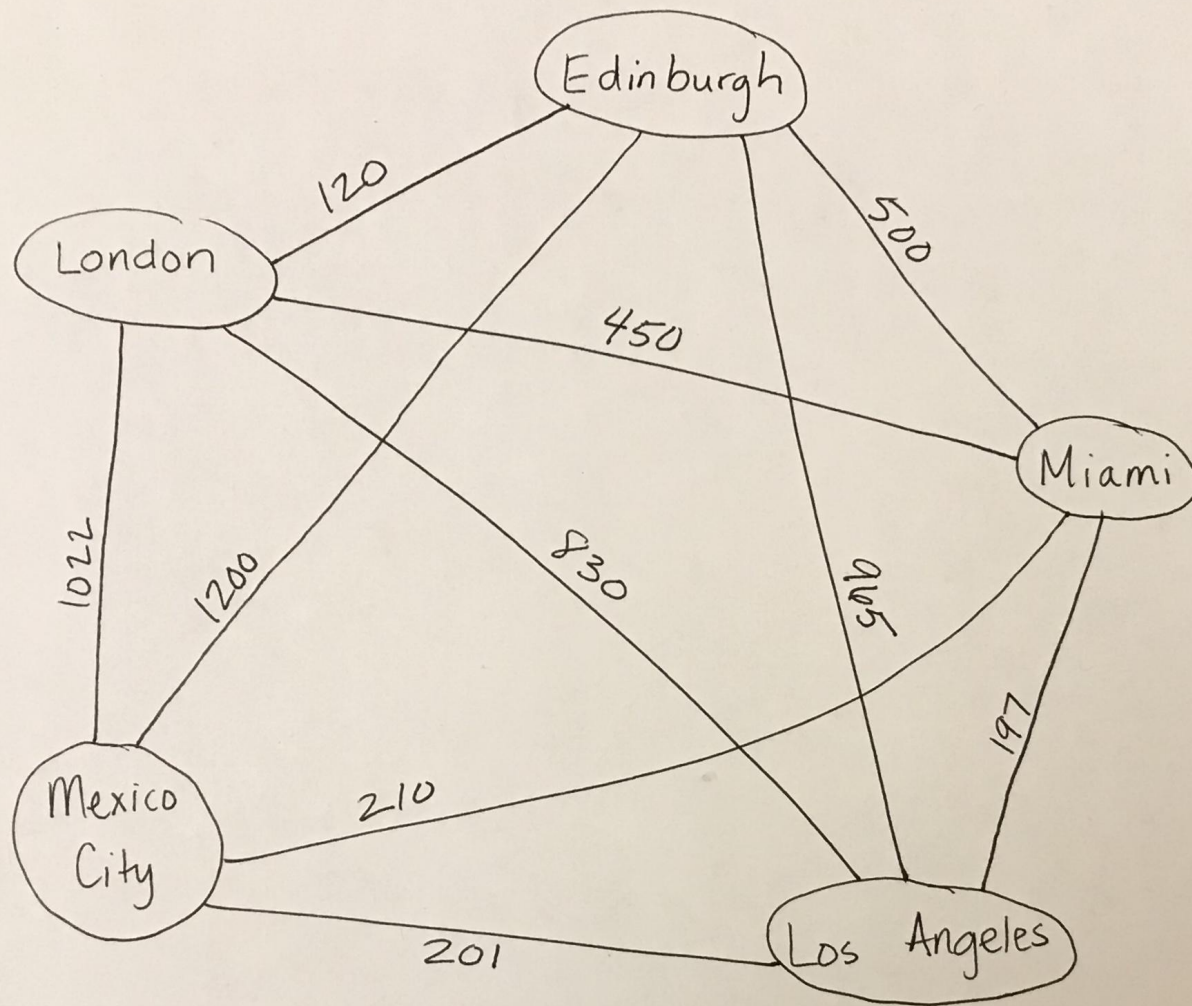
**if**  $e$  is not already in  $T$  **then**

      Add edge  $e$  to  $T$

**return**  $T$

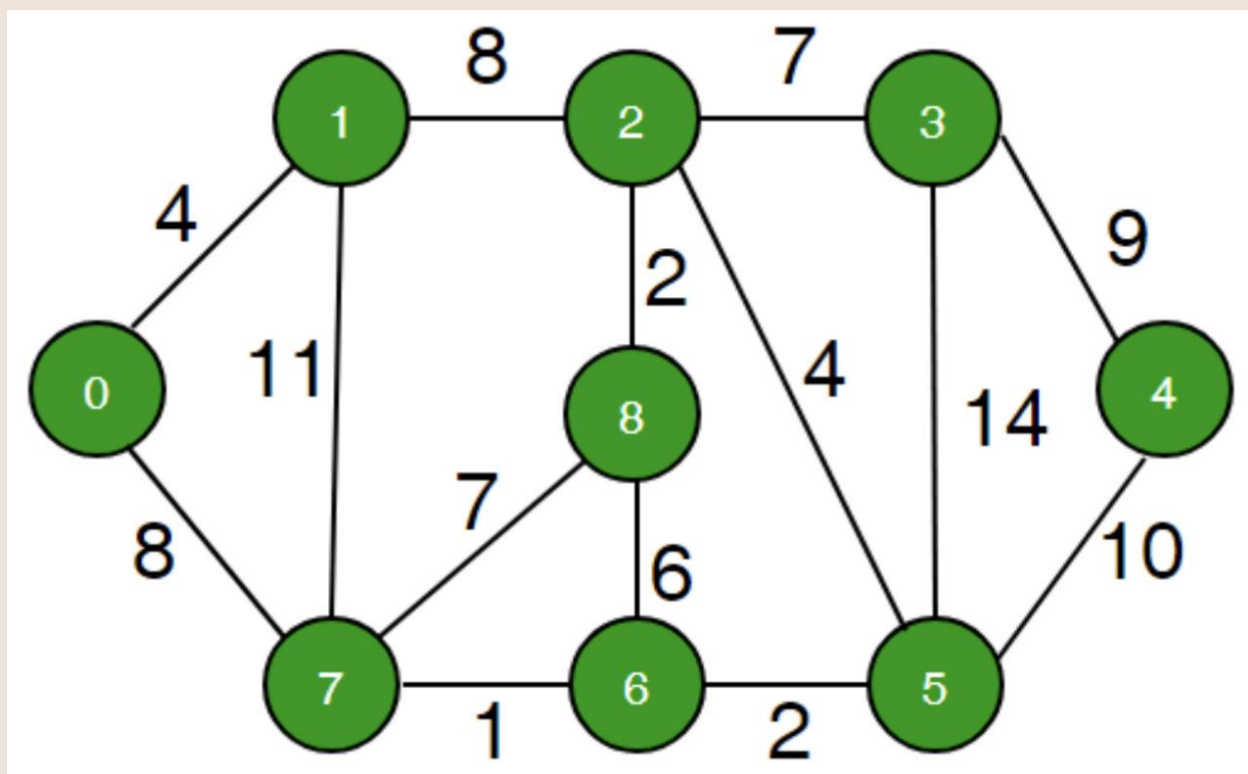
- ◆ Each iteration of the while-loop halves the number of connected components in  $T$ .
  - The running time is  $O(m \log n)$ .

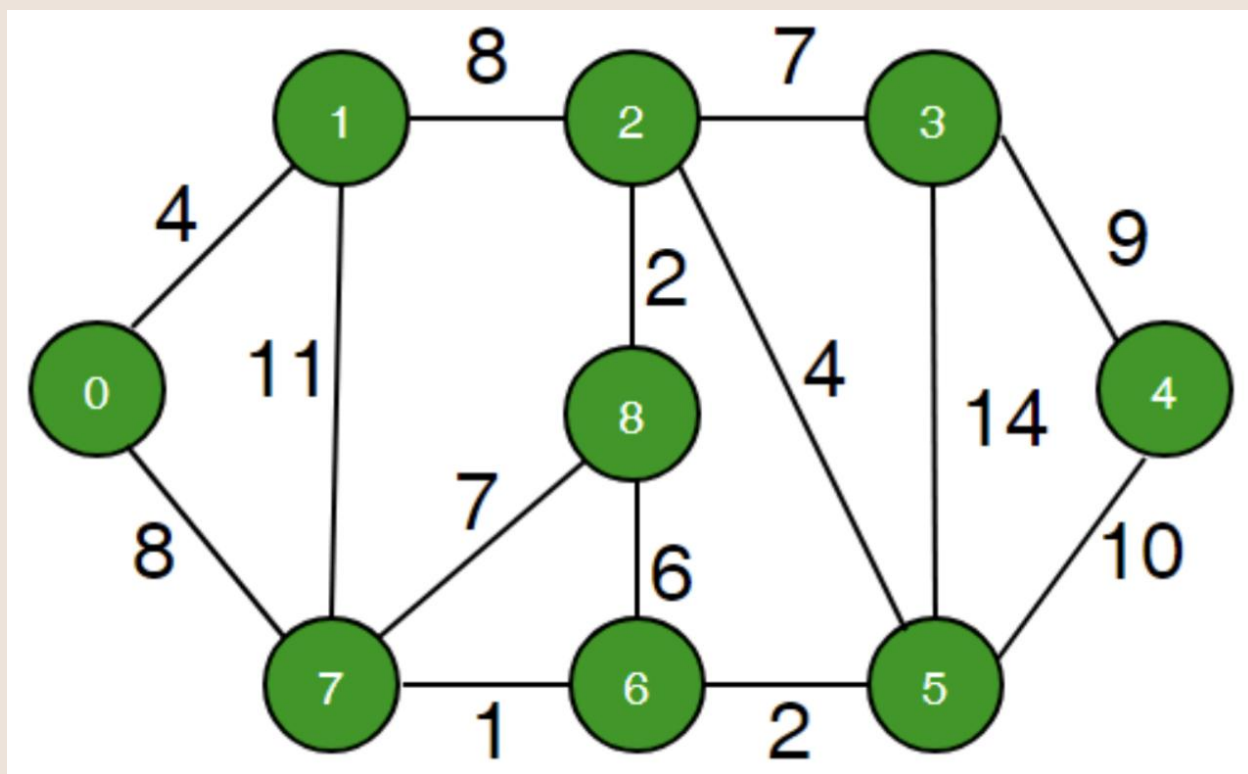
Goal: Find an MST of the graph using Baruvka's Algorithm...

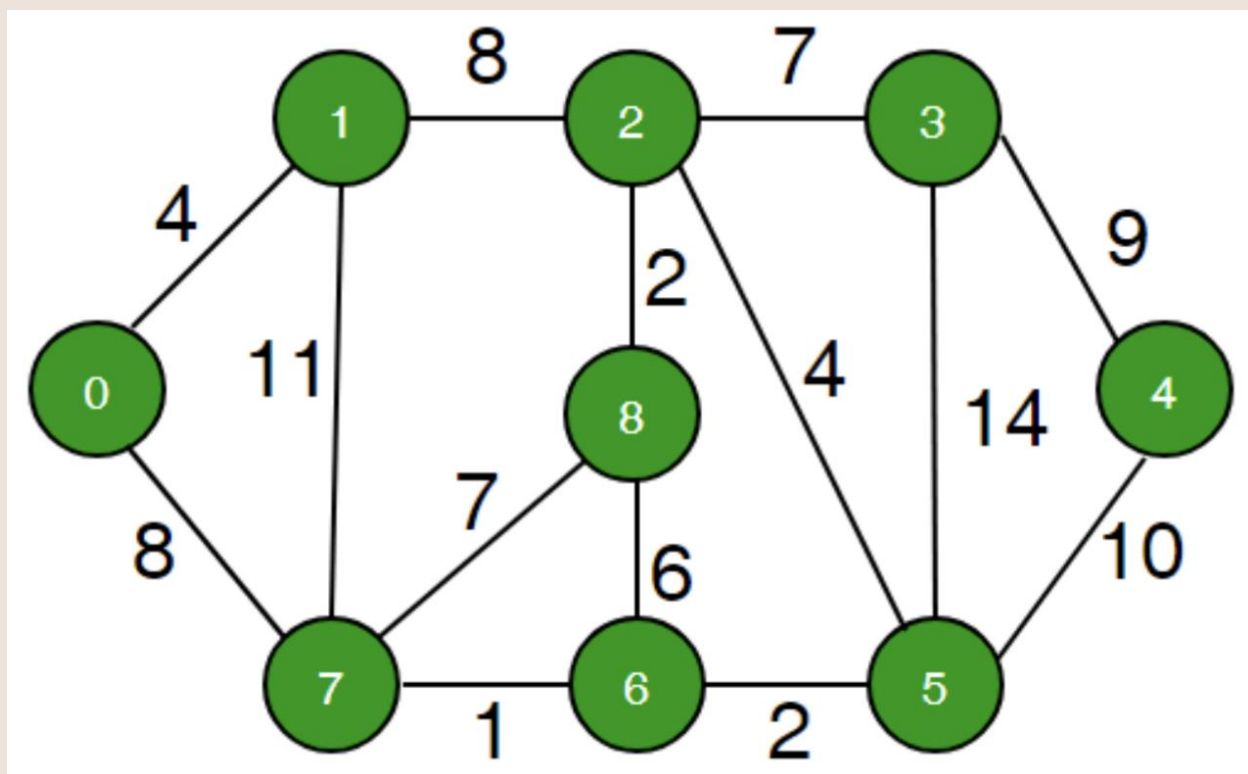




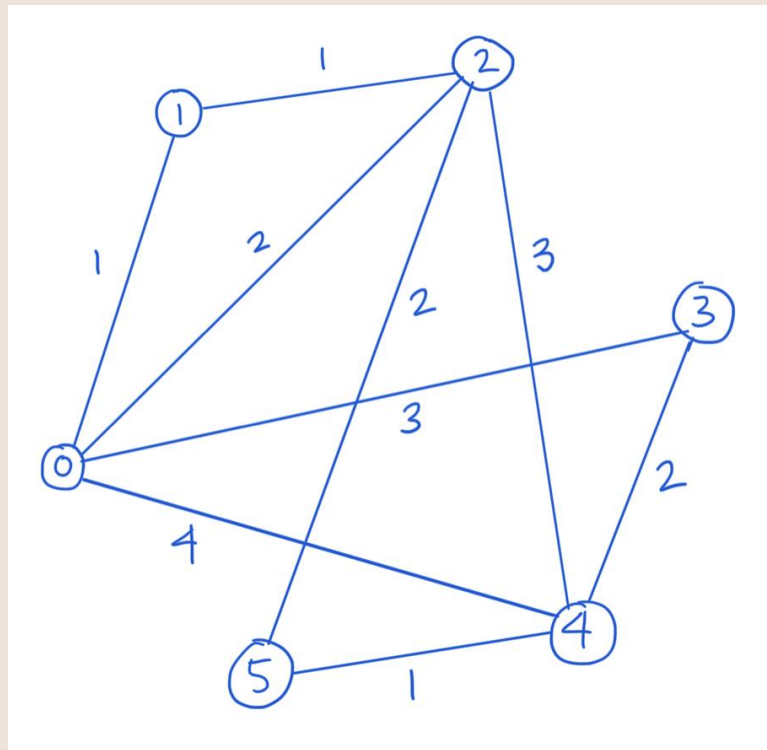
# More Examples







How many different MSTs does the graph below have?



# A little extra

An NP-Complete Problem

# Imagine...

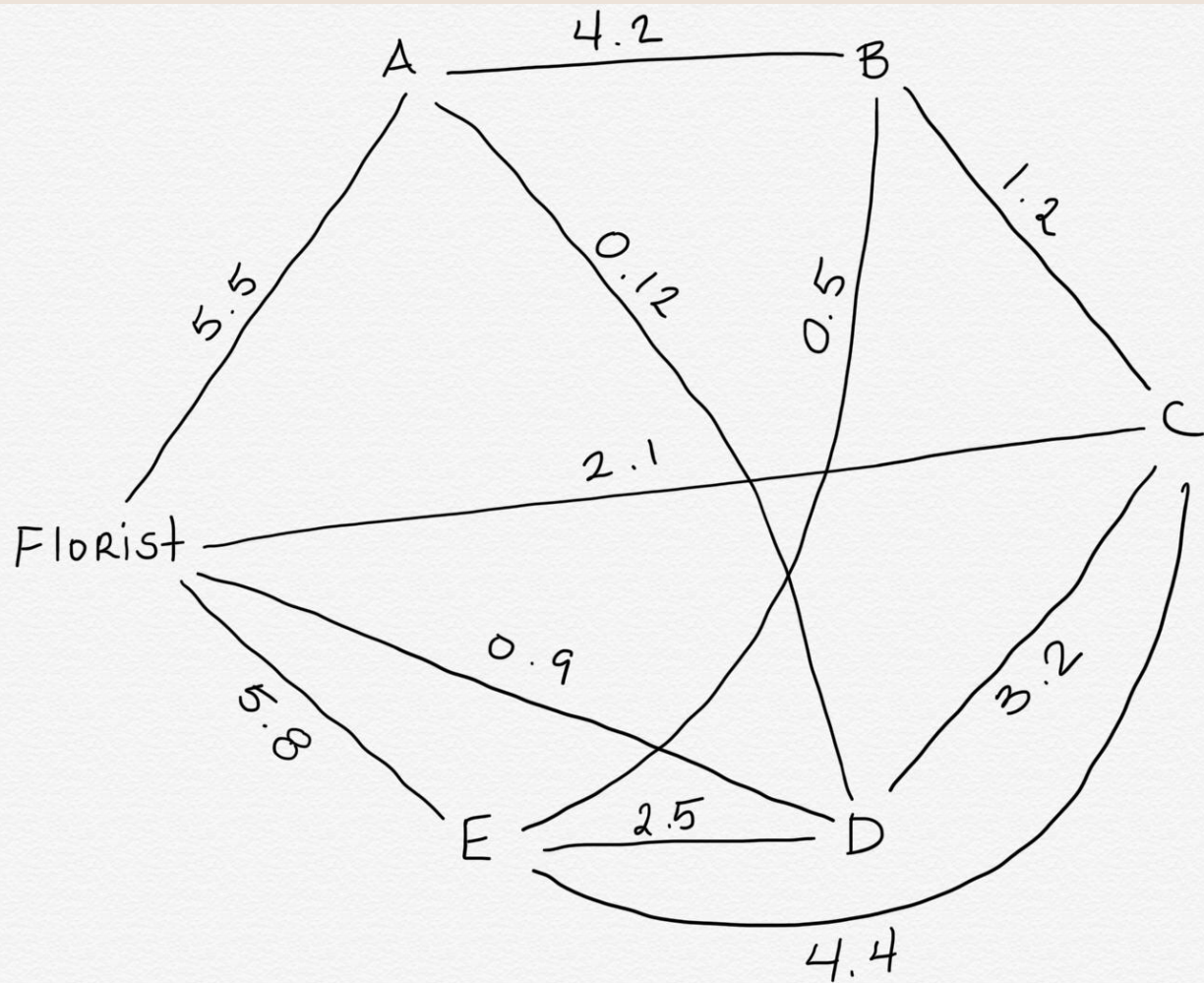
You are employed at a florist shop as a driver during their busy Valentine's Day season. Your first task is to take a list of delivery addresses and plan a route according to the following specifications:

- You want to start and end at the florist shop.
- You want to deliver ALL the flowers
- You want to minimize the distance you have to go.
- You don't want to visit any stop more than once.
- You don't want to drive the same road more than once.

How would you go about solving this problem?

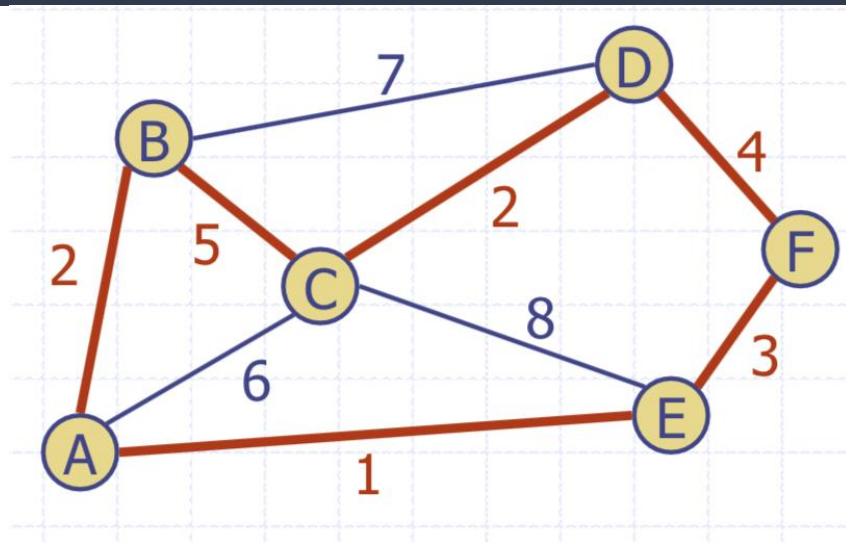
	Florist	A	B	C	D	E
Florist	0	5.5	NONE	2.1	0.9	5.8
A	5.5	0	4.2	NONE	0.12	NONE
B	NONE	4.2	0	1.2	NONE	0.5
C	2.1	NONE	1.2	0	3.2	4.4
D	0.9	0.12	NONE	3.2	0	2.5
E	5.8	NONE	0.5	4.4	2.5	0

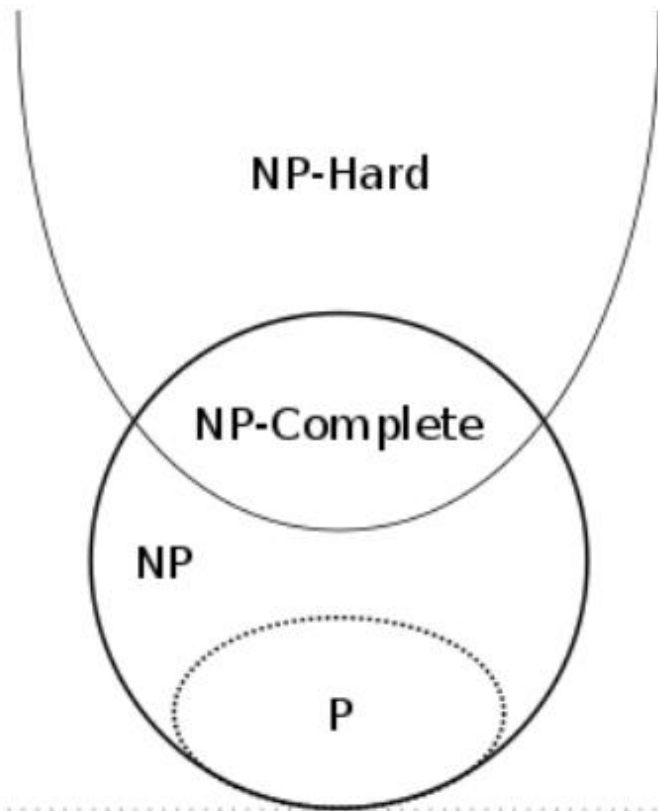




# The Traveling Salesperson Problem

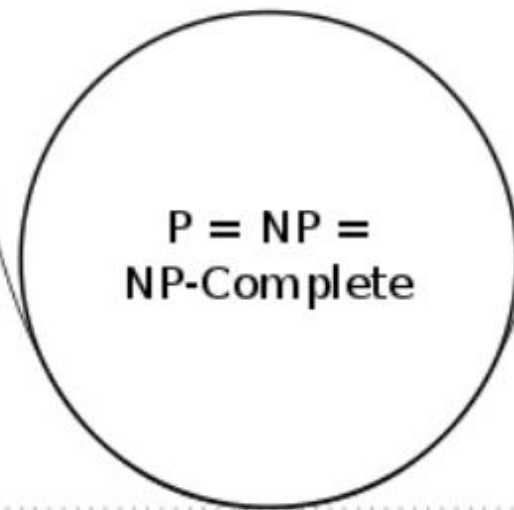
- A **tour** of a graph is a spanning cycle (goes through all the vertices)
- A TSP tour of a weighted graph is a tour that is simple (i.e. repeats no vertices or edges) and has minimum weight
- Determining if a TSP tour shorter than a given length  $L$  **exists** in a graph is NP-Complete (i.e. solution can be verified in polynomial time but not discovered in polynomial time—at least not yet)
- Finding a TSP tour is NP-hard (i.e. finding a minimum-weight tour)





$P \neq NP$

Complexity



$P = NP$

# Other NP-Complete Problems

- Boolean Satisfiability Problem (SAT)
- Knapsack Problem
- Hamiltonian Path Problem
- Subgraph Isomorphism Problem
- Subset sum problem
- Clique Problem
- Vertex Cover Problem
- Independent set problem
- Dominating set problem
- Graph Coloring problem

# References

- [1] Goodrich and Tamassia
- [2] Sedgewick and Wayne
- [3] [en.wikipedia.org](https://en.wikipedia.org)