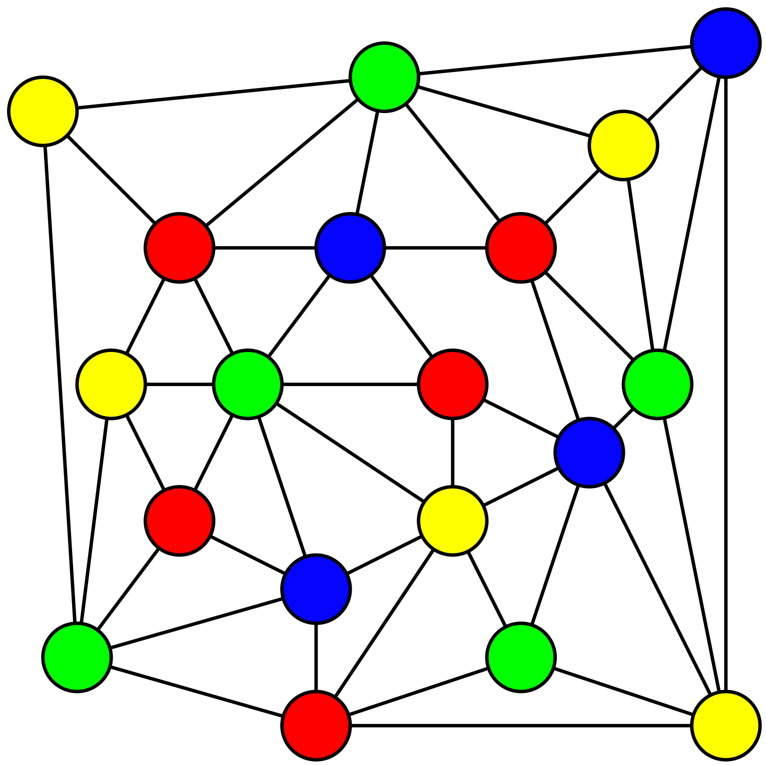
Further Mathematics and Algorithms

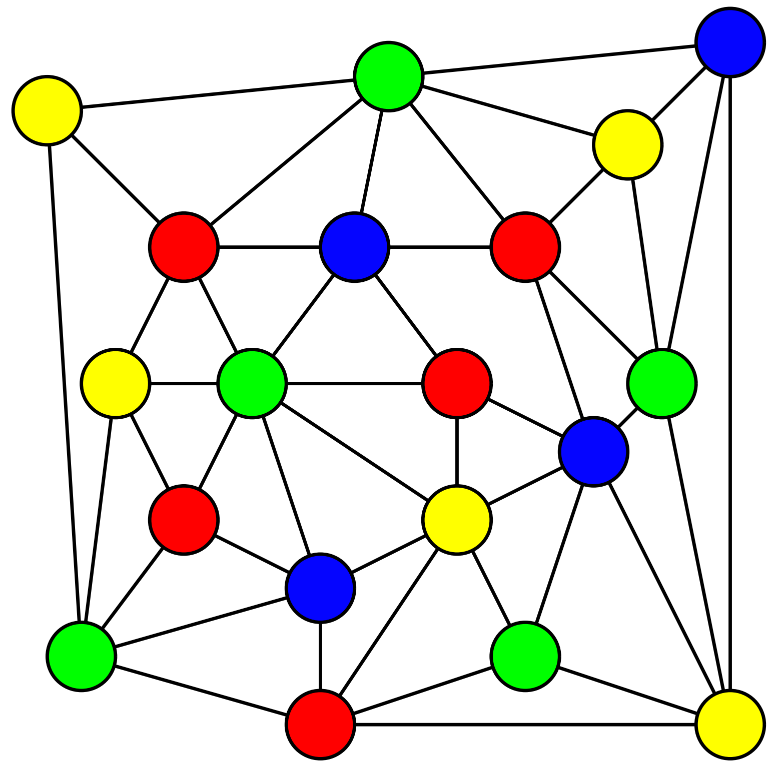
Lesson 17: *Think Graphically*



Graph theory, applications of graphs, graph problems

Outline

1. **Graph Theory**
2. Applications of Graphs
   - Geometric applications
   - Relational applications
3. Implementing Graphs
4. Graph Problems



Motivation

- Many different problems can be described in terms of graphs
- This often reveals the true nature of the problem
- It unifies many apparently different problems
- As much is known about graph problems it often provides a pointer to the solution

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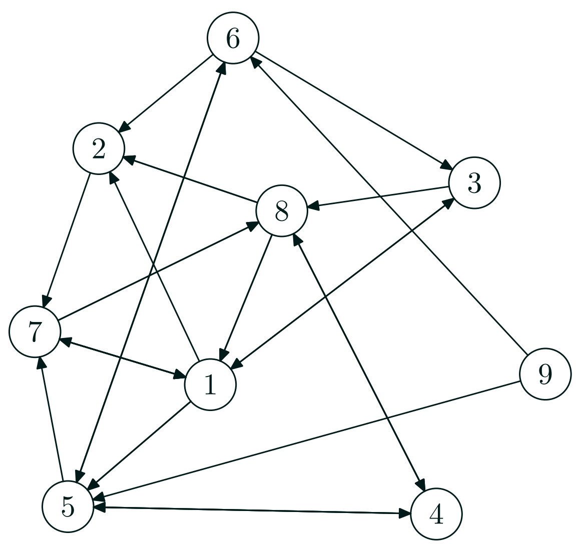
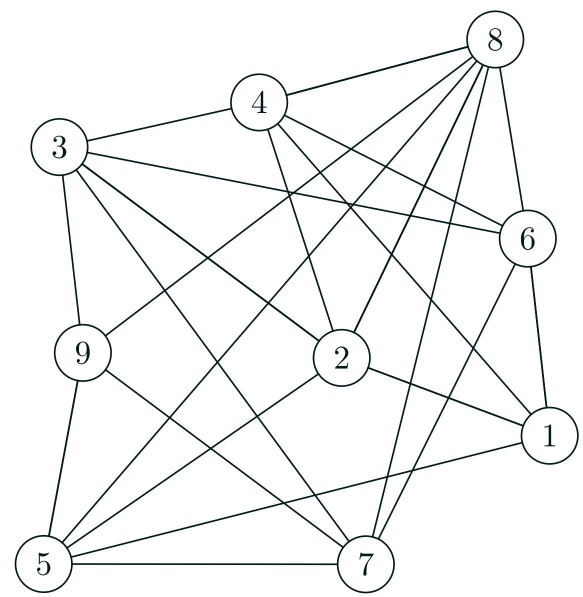
Definition of a Graph

- A graph, G, can be described by
  - ⋆ A set of vertices or nodes $\mathcal{V} = \{1, 2, 3 \ldots n\}$
  - ⋆ A set of edges $\mathcal{E} = \{(i, j)|\text{vertex } i \text{ is connected to vertex } j\}$
- The edges may be
  - ⋆ **directed**—sometimes called a **digraph**
  - ⋆ **undirected**





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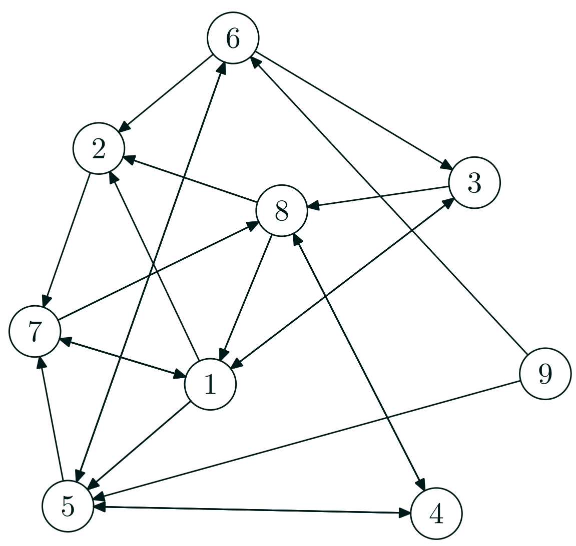

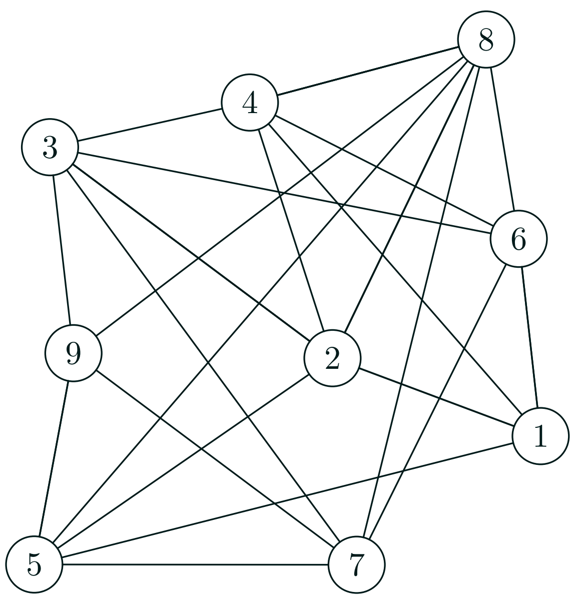

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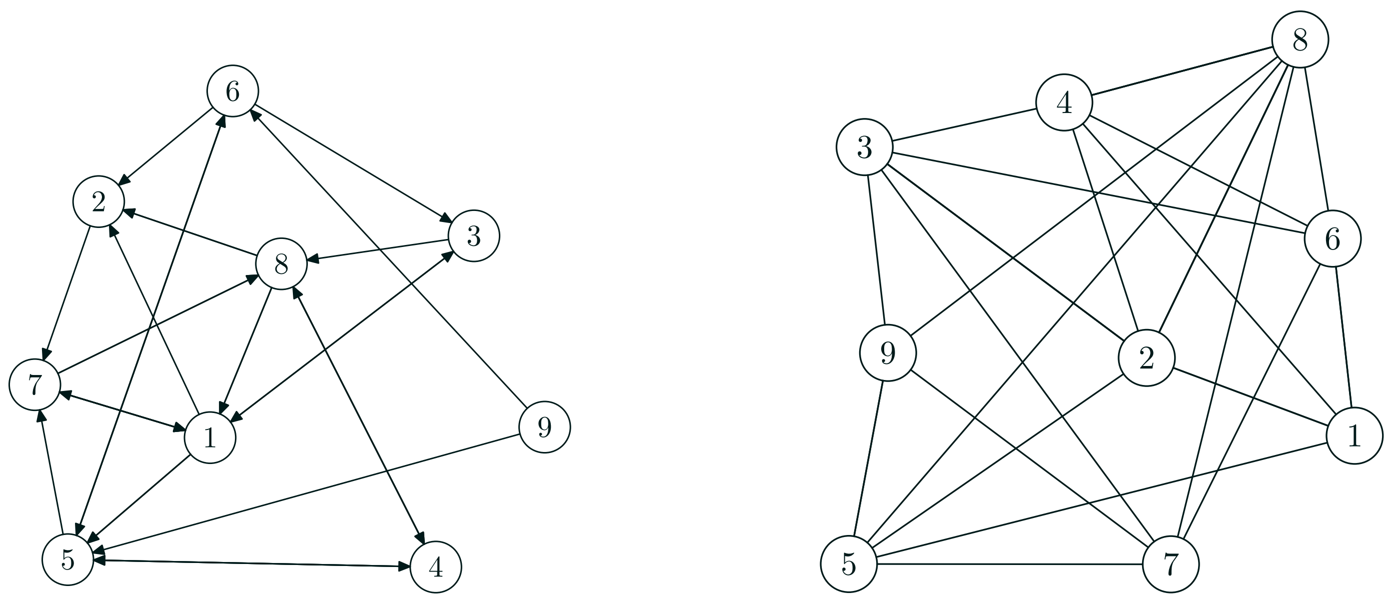

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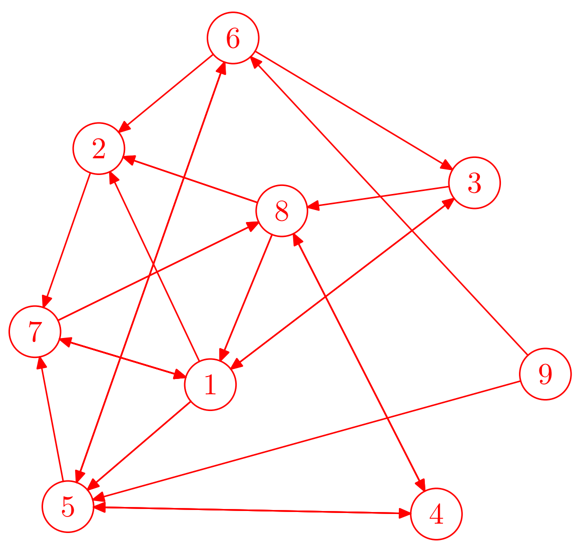

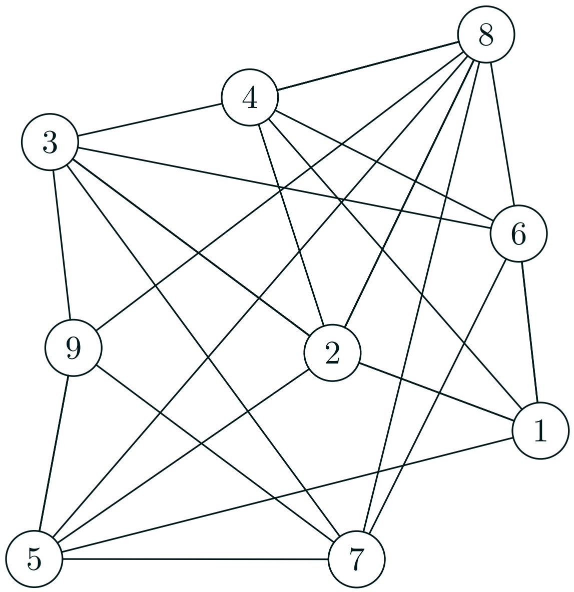

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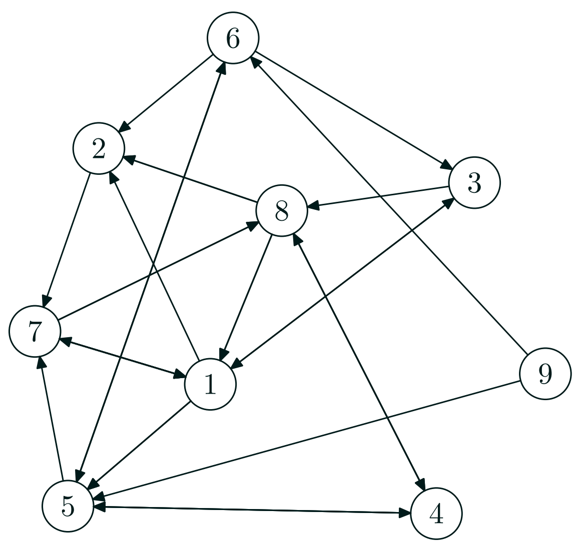
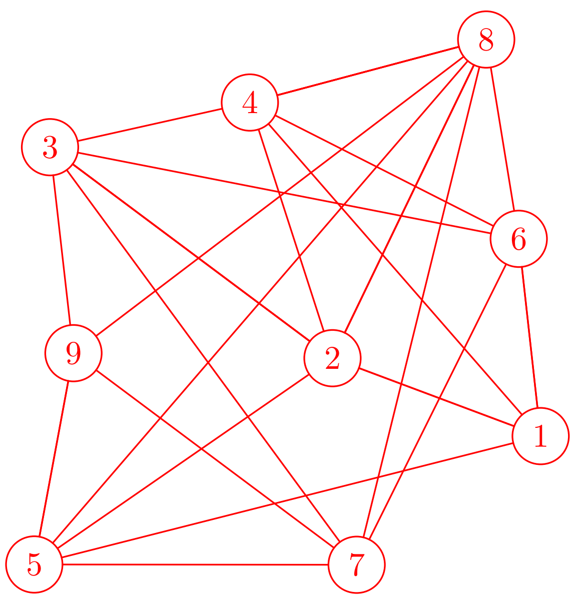
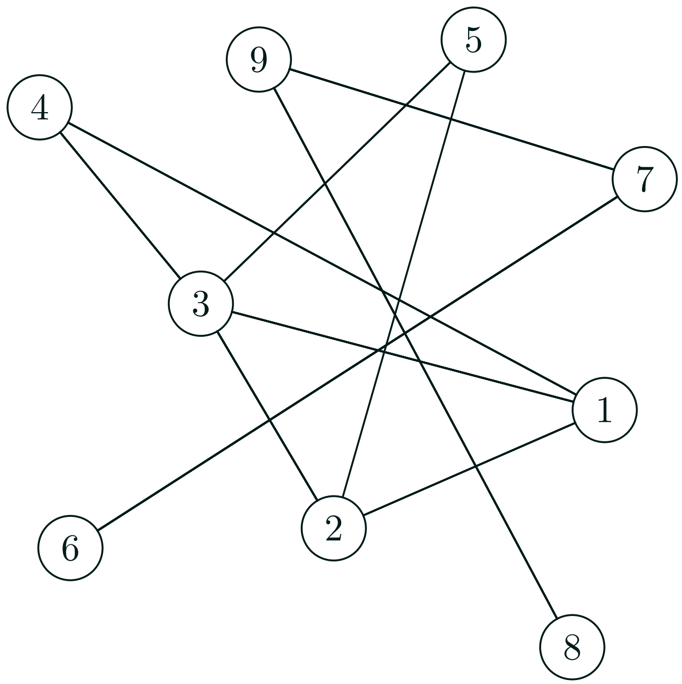




Connected and Unconnected Graphs

- A graph is **connected** if you can get from one node to any other along a series of edges

- Otherwise it is **disconnected**



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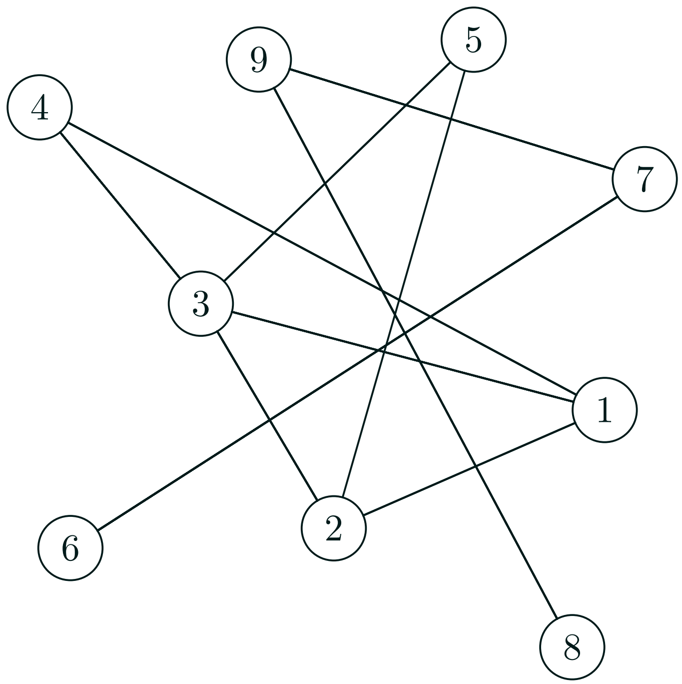

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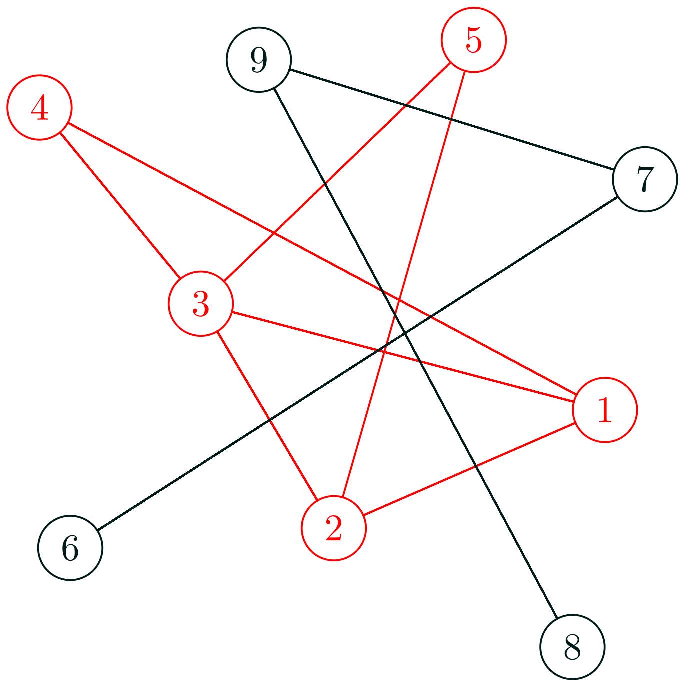

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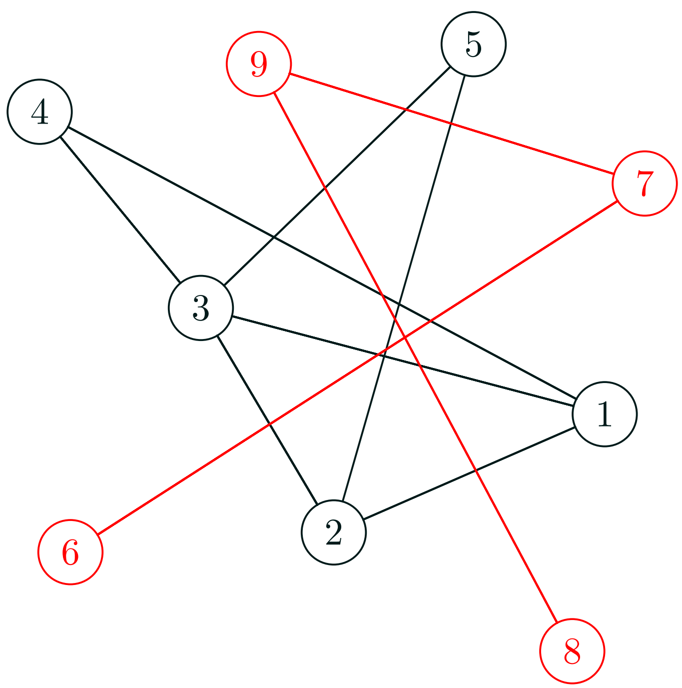

Trees

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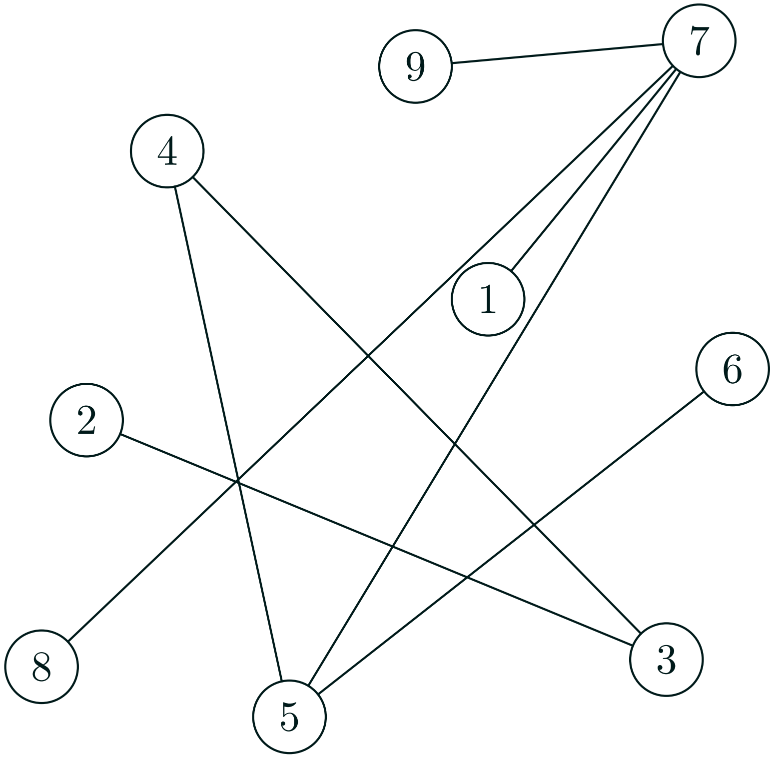
- A tree will have $n - 1$ edges



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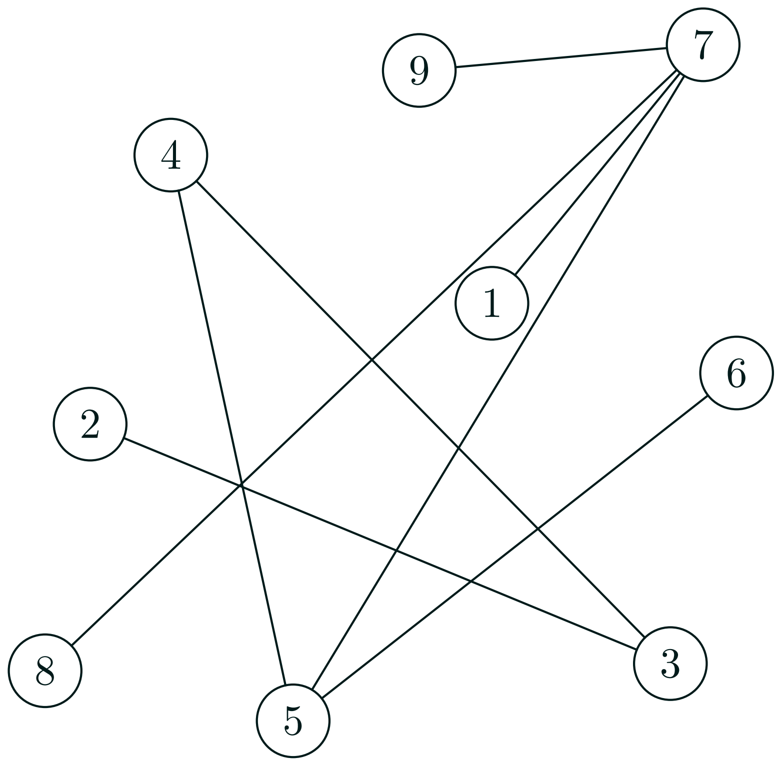
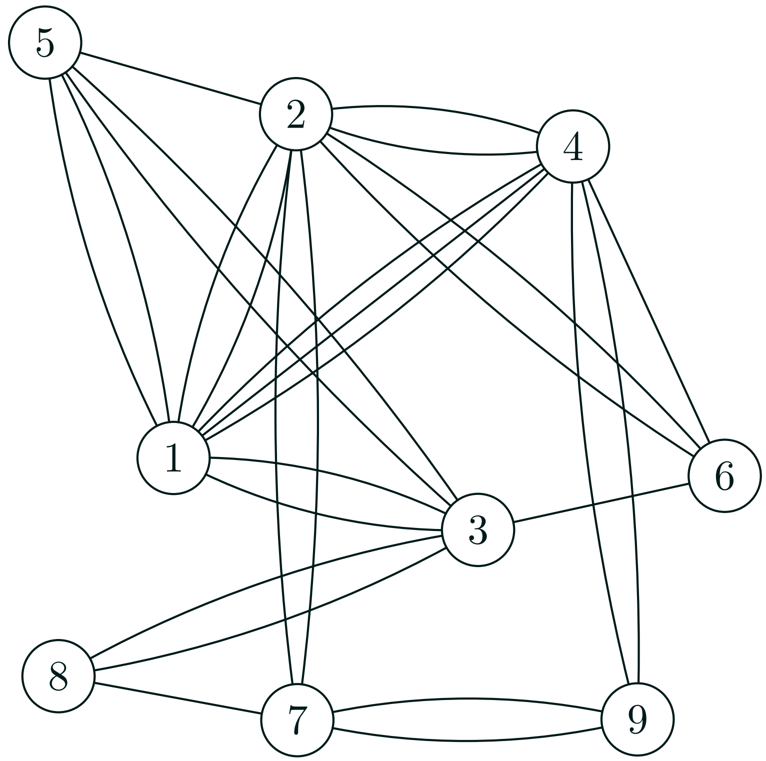


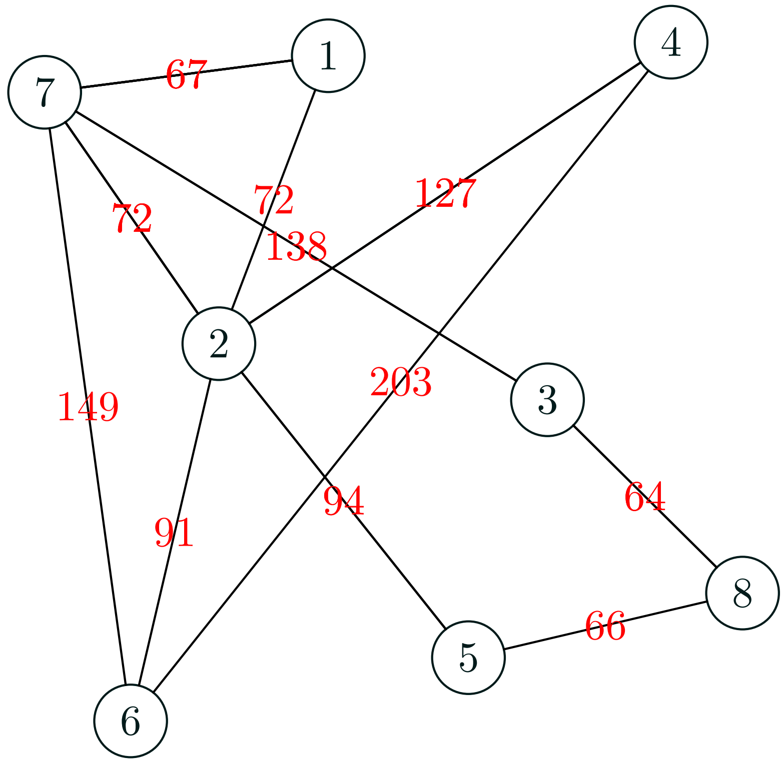
Multigraphs

- If the collection of edges is a *multiset* then we obtain a **multigraphs** where more than one edge is allowed between pairs of vertices



Weighted Graphs

- If we assign a number to an edge we obtain a **weighted graph**



Networks

- Sometimes we add more information to the graph
- E.g. attributes to the nodes or edges
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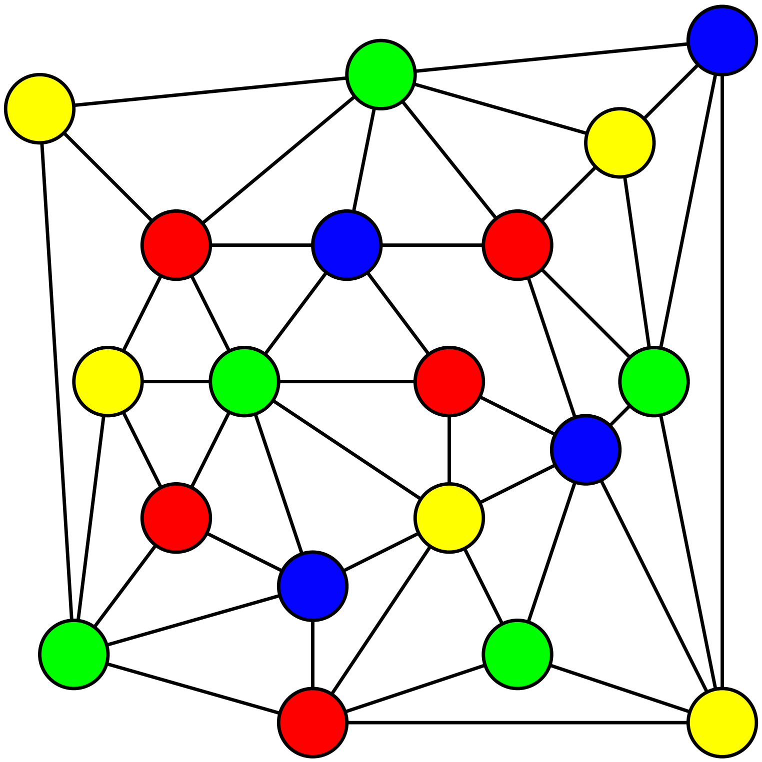
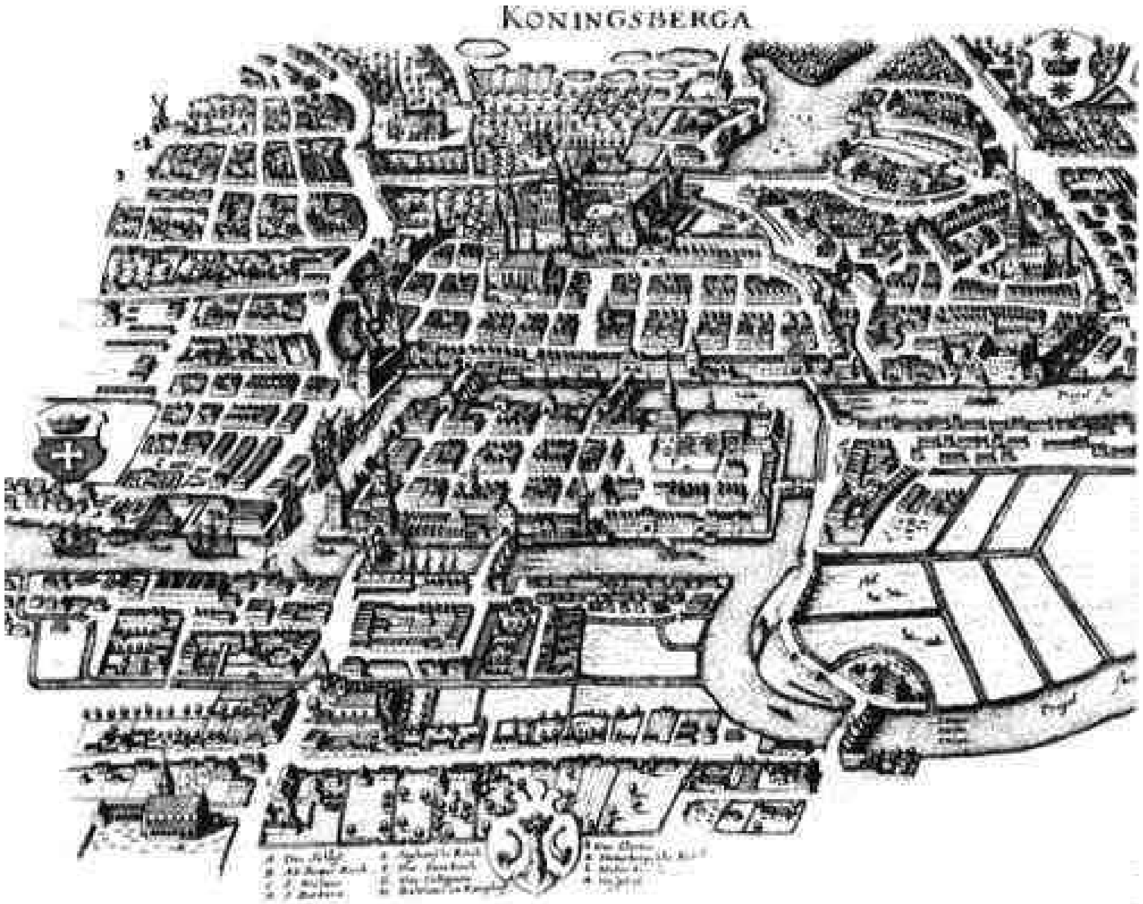


Bridges of Königsberg

Is there a tour around Königsberg going over every bridge once?



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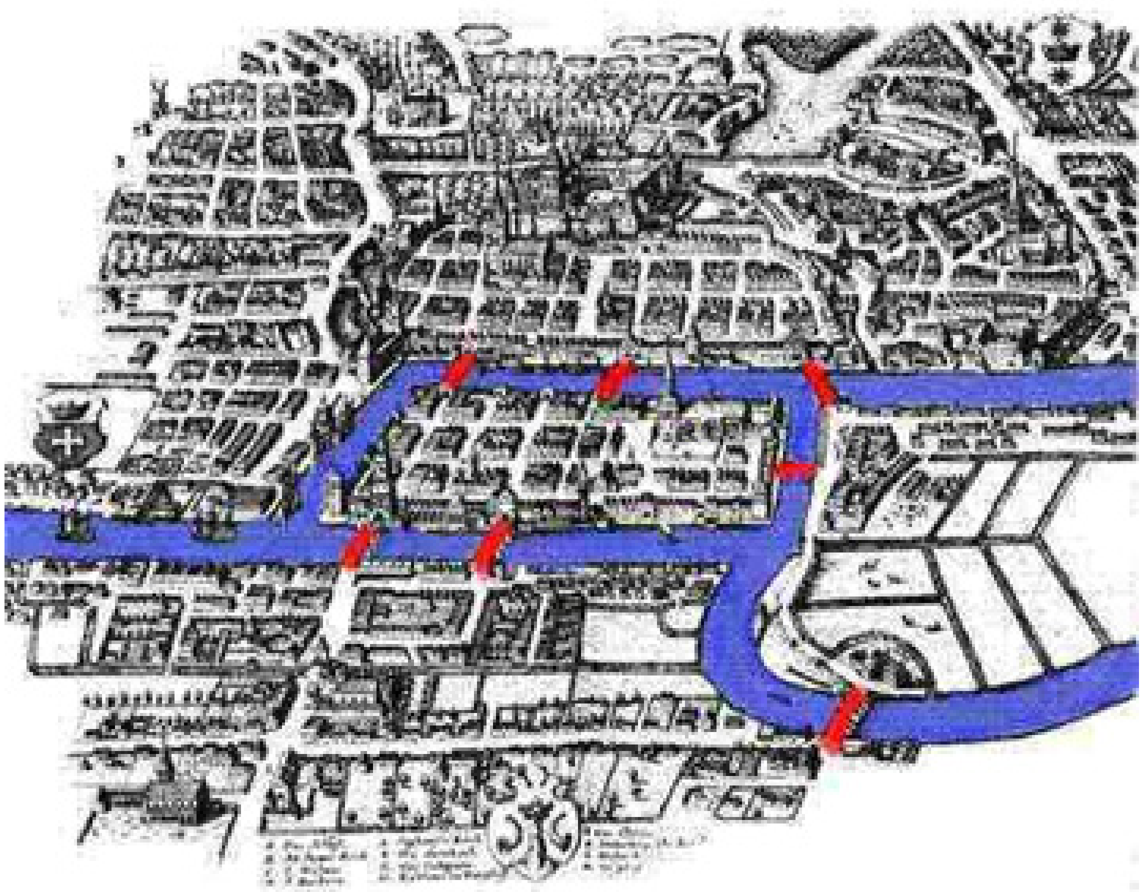

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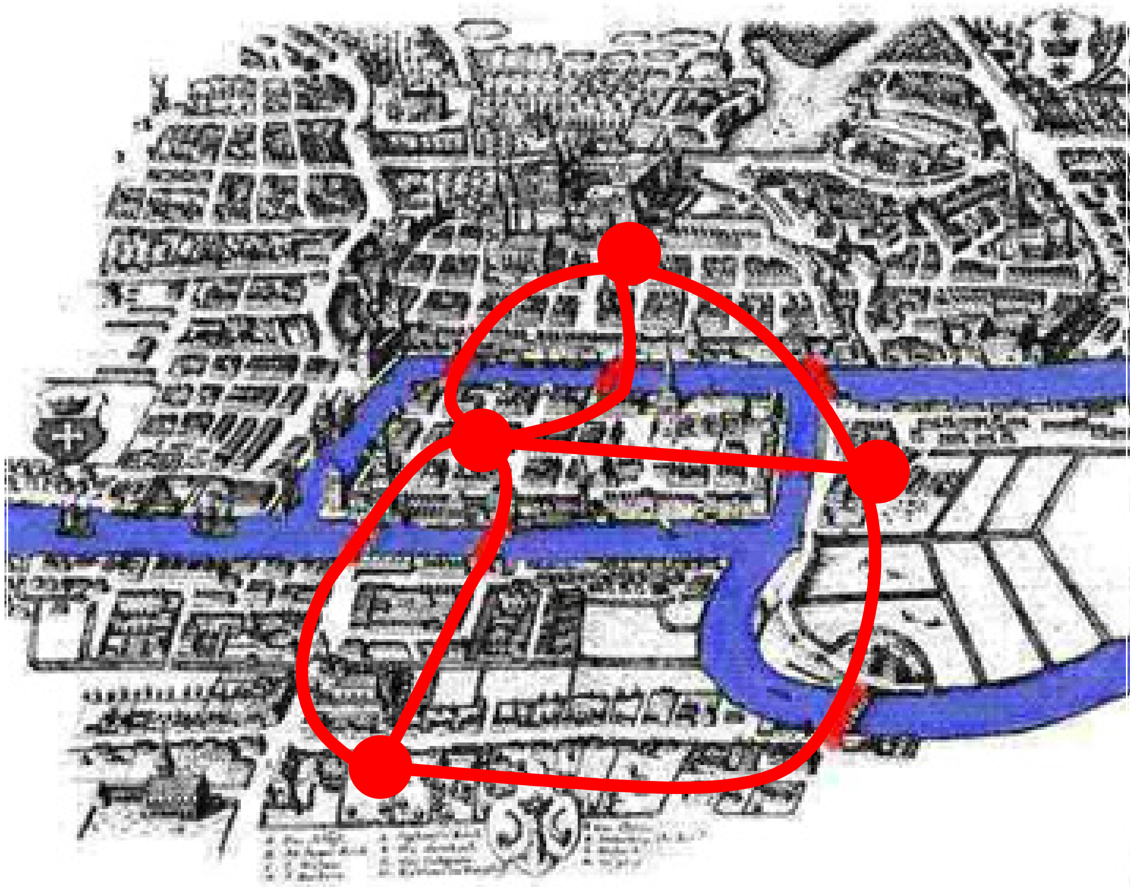

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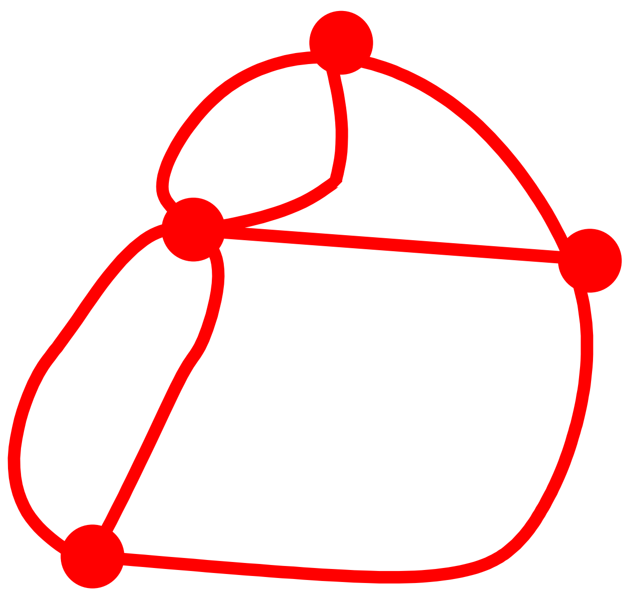
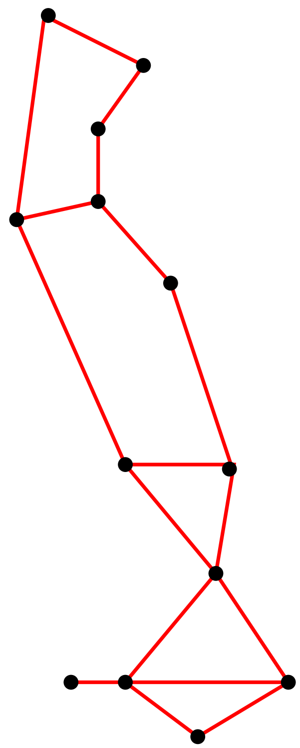


In 1736 Euler published a paper answering this question and founding graph theory

Representing Distances



- Consider some graph
- With weights representing the distance between nodes
- What is the shortest distance between S and I?

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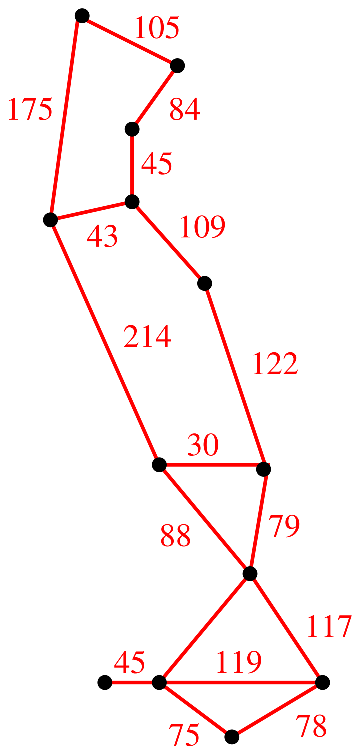

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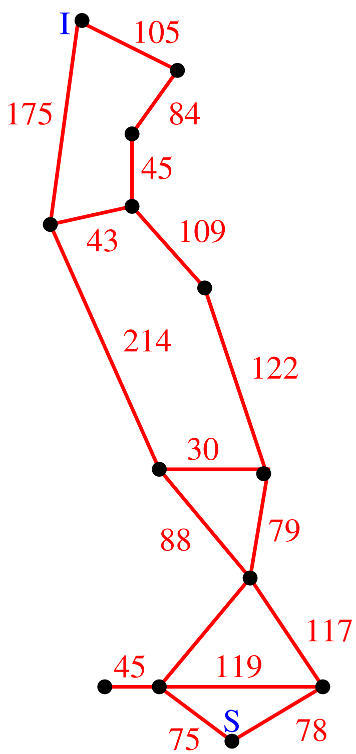

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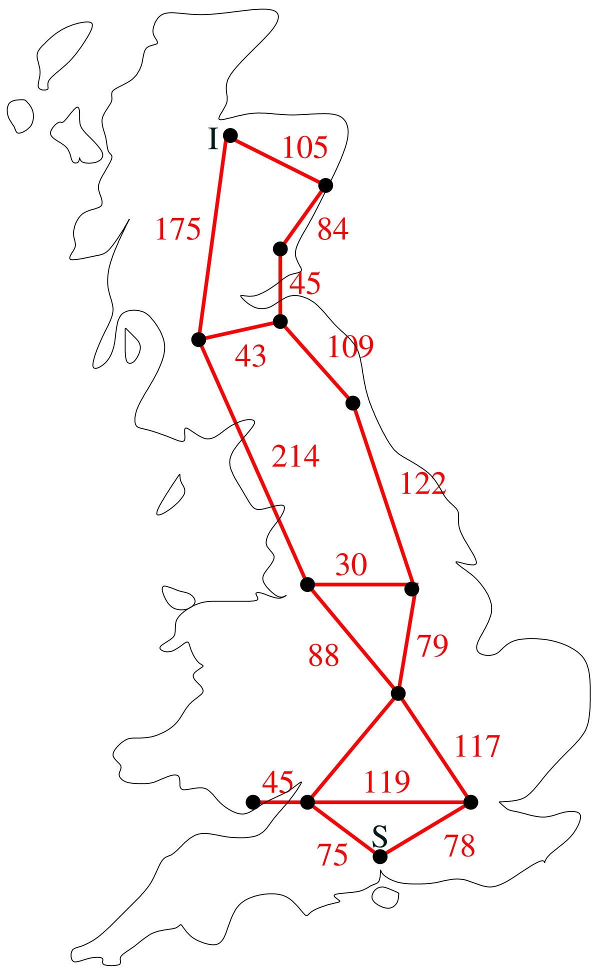

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Other Applications

- We could take the weights to represent the time taken to travel between nodes
- In a computer network the weights might represent the bandwidth
- In a representation of a transport system the weights might represent the carrying capacity of the traffic on a road
- Graphs can be used to represent other kinds of relationships
- E.g. We could create a digraph of links between web pages

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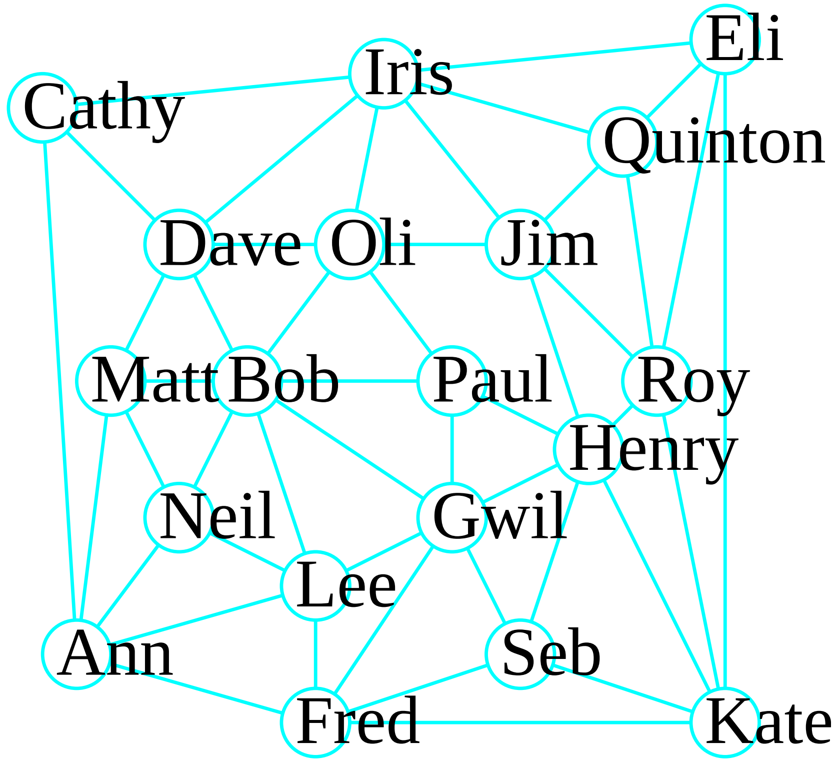
Christmas Card Problem

- I have four types of Christmas cards
- Some of my friends know each other



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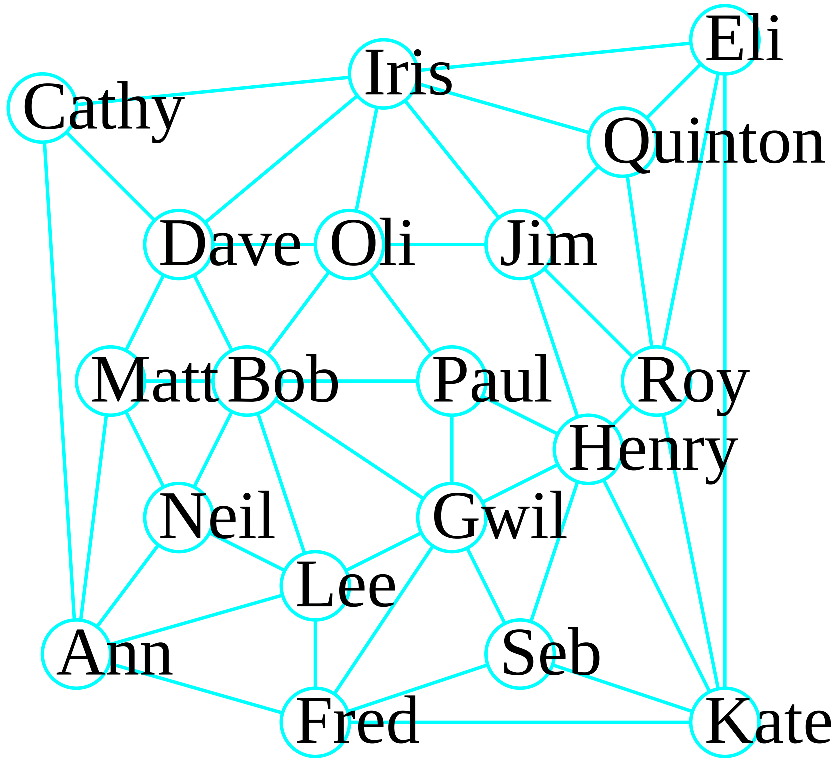

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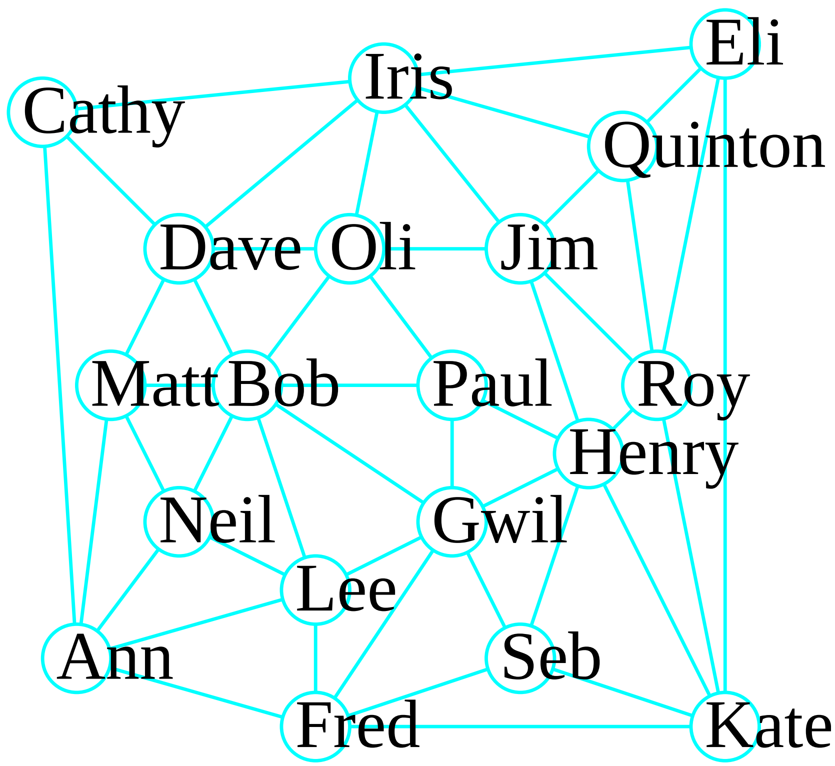

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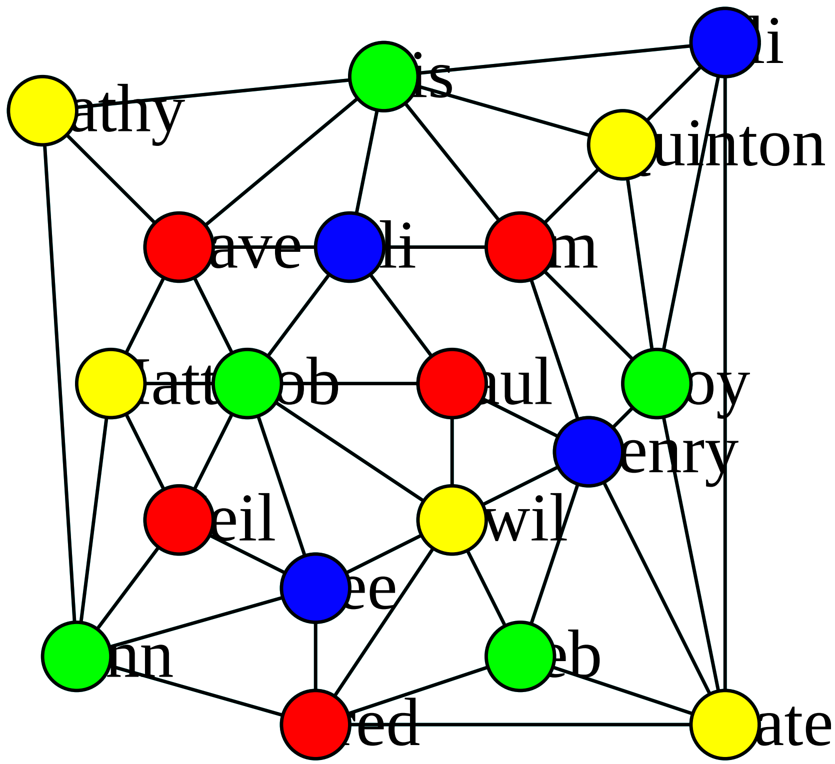

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A Real World Problem

- A food company used different colour bags for each of it products
- To save money they reduced the stock of bags to 25
- They wanted to know what items to put in what bags so that as few customers as possible would have items with the same colour bags
- This can again be reduced to a graph colouring problem
  - ⋆ Each node represents an item
  - ⋆ The edges were weighted by the number of customers that took both items
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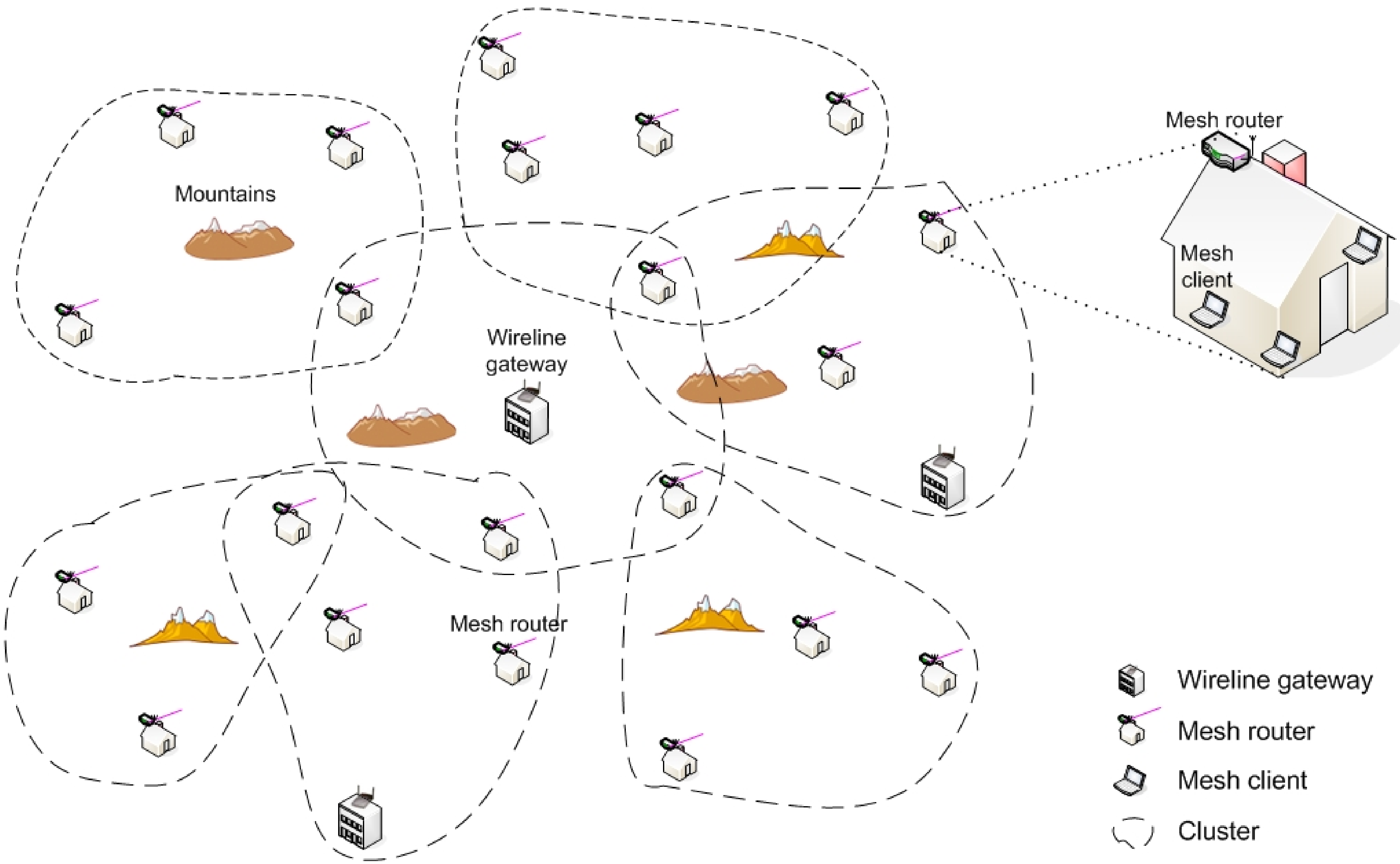
Frequency Assignment Problem



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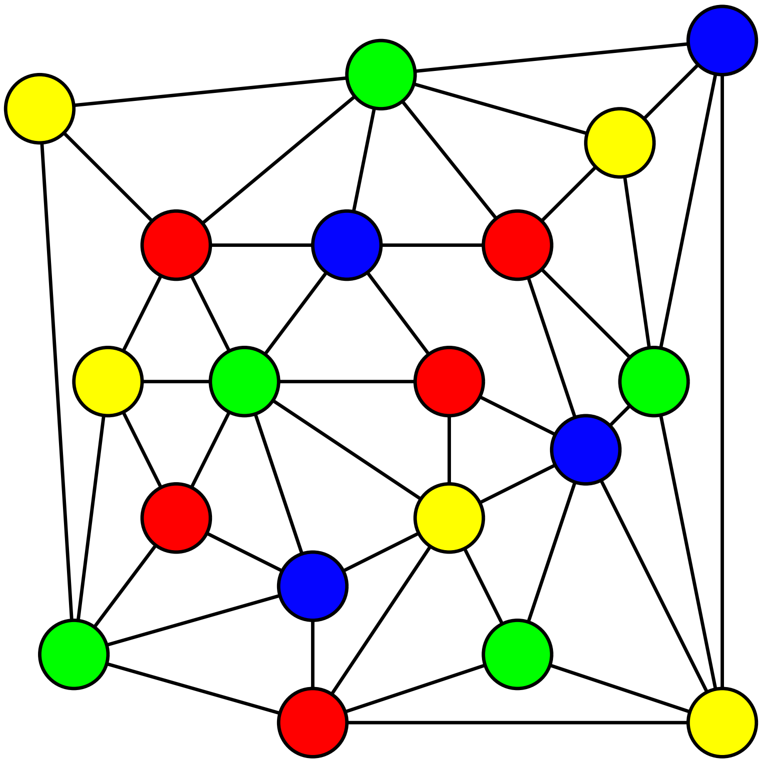

Representations

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- The best representation depends on the graph
- Some books describe a $Graph\ ADT$—graphs are too varied for this to be very useful
- An important issue in representing a graph is how to store the edge information

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Adjacency Matrices

- One representation of a graph $G = (\mathcal{V}, \mathcal{E})$ is in term of an $n \times n$ **adjacency matrix** **A** with elements

$$A_{ij} = \begin{cases} 1 & \text{if } (i,j) \in \mathcal{E} \\ 0 & \text{if } (i,j) \notin \mathcal{E} \end{cases}$$

  where $n = |\mathcal{V}|$

- For undirected graphs **A** is a symmetric matrix, i.e. $\mathbf{A} = \mathbf{A}^\mathsf{T}$

- For weighted graphs we often store the **connectivity matrix** or **cost-adjacency matrix**, **C**, where

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- But in **sparse** graphs where the number of edges is $\Theta(n)$ the adjacency matrix has a very large number of zeros
- A more efficient representation is in terms of the adjacency list where the set of outgoing edges is stored for each node
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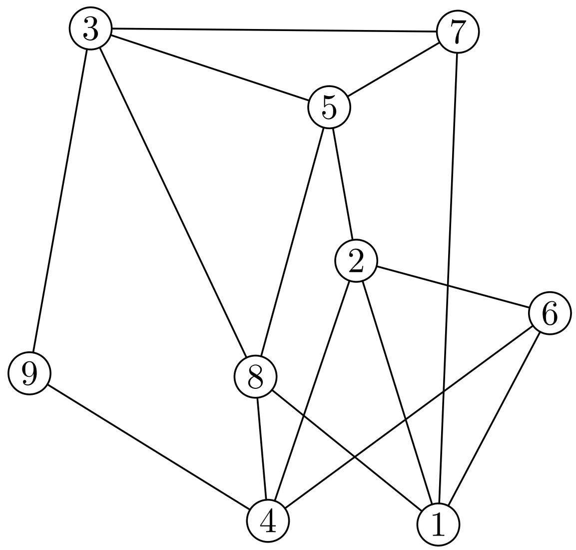
- For **dense** graphs where the number of edges is $\Theta(n^2)$ the adjacency matrix is often a useful representation
- But in **sparse** graphs where the number of edges is $\Theta(n)$ the adjacency matrix has a very large number of zeros
- A more efficient representation is in terms of the adjacency list where the set of outgoing edges is stored for each node
- In some applications it is useful to store both the adjacency matrix and the adjacency list

Representing Undirected Graphs

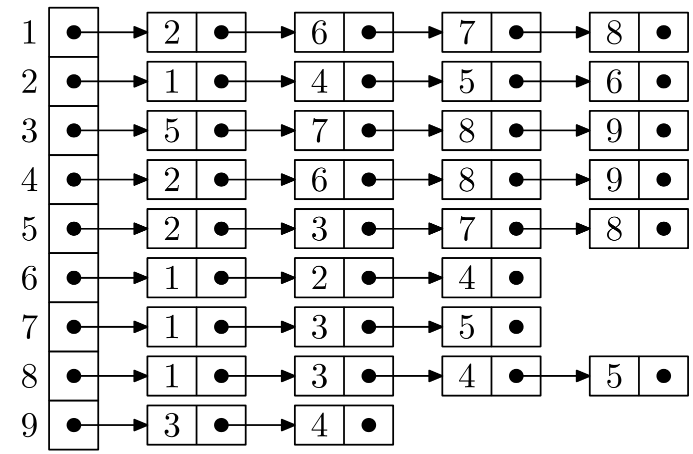


Graph

from

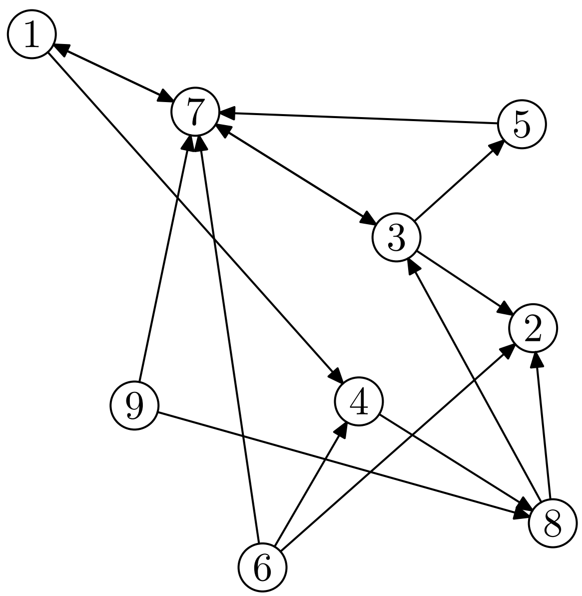
		1	2	3	4	5	6	7	8	9
	1	0	1	0	0	0	1	1	1	0
	2	1	0	0	1	1	1	0	0	0
	3	0	0	0	0	1	0	1	1	1
	4	0	1	0	0	0	1	0	1	1
to	5	0	1	1	0	0	0	1	1	0
	6	1	1	0	1	0	0	0	0	0
	7	1	0	1	0	1	0	0	0	0
	8	1	0	1	1	1	0	0	0	0
	9	0	0	1	1	0	0	0	0	0

Adjacency Matrix



Adjacency List

Representing Digraphs



Graph

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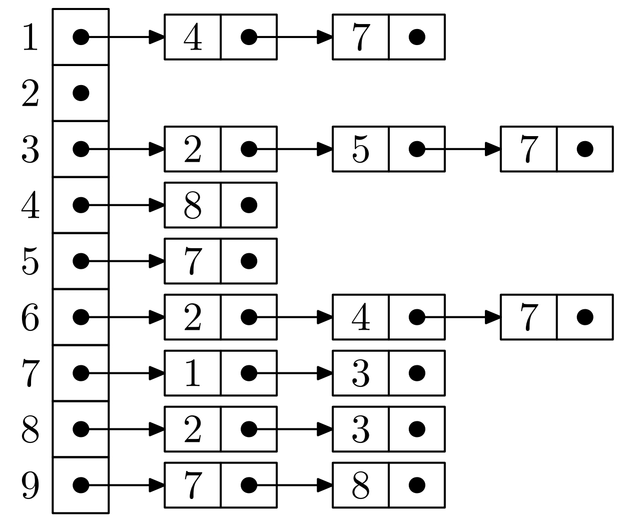
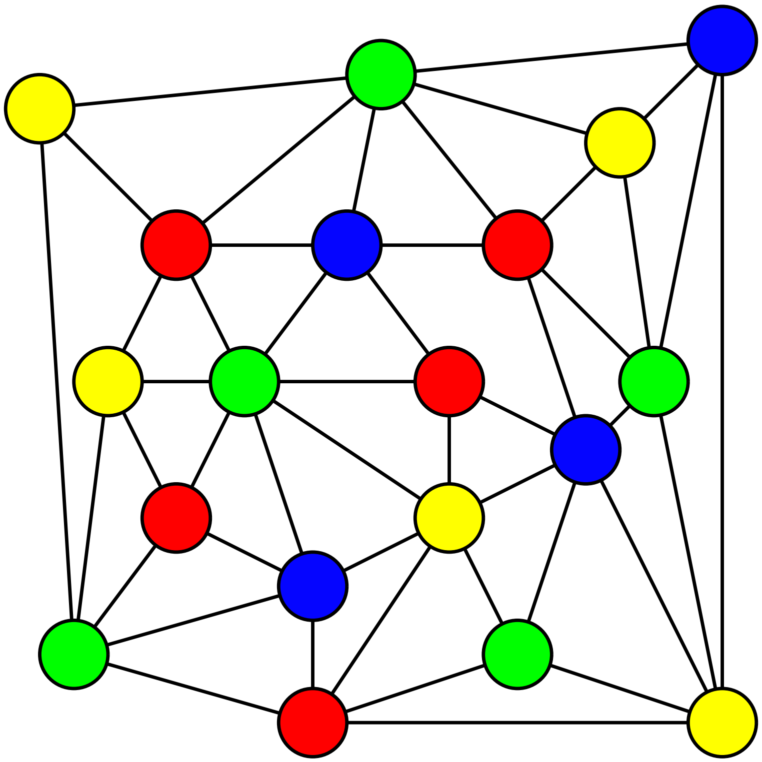
Adjancency Matrix



Adjancency List

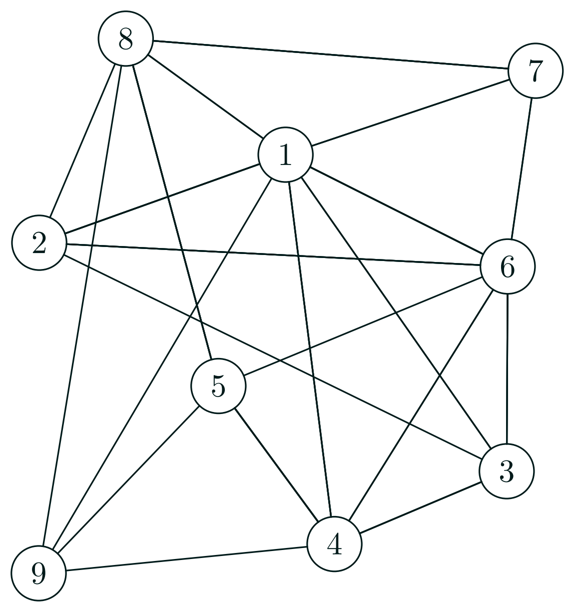
Outline

1. Graph Theory
2. Applications of Graphs
   - Geometric applications
   - Relational applications
3. Implementing Graphs
4. **Graph Problems**



Hamilton Cycle

- The Euler path problem is to find a path through a multigraph that passes through every edge once—easy to solve
- The Hamilton cycle problem is to find a cycle that goes through each vertex exactly once



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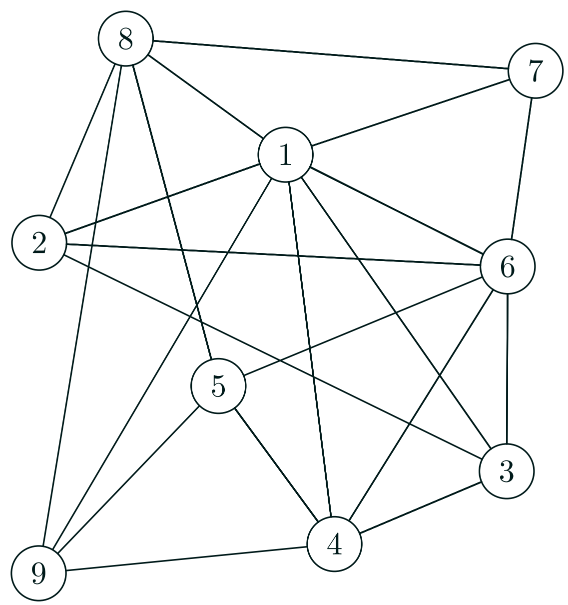

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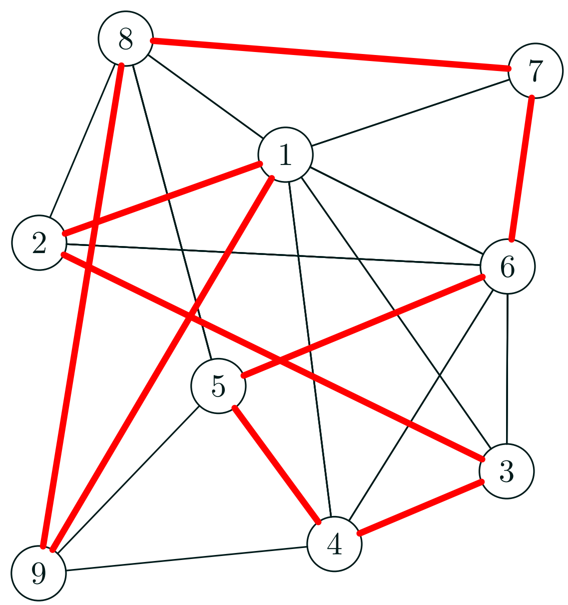

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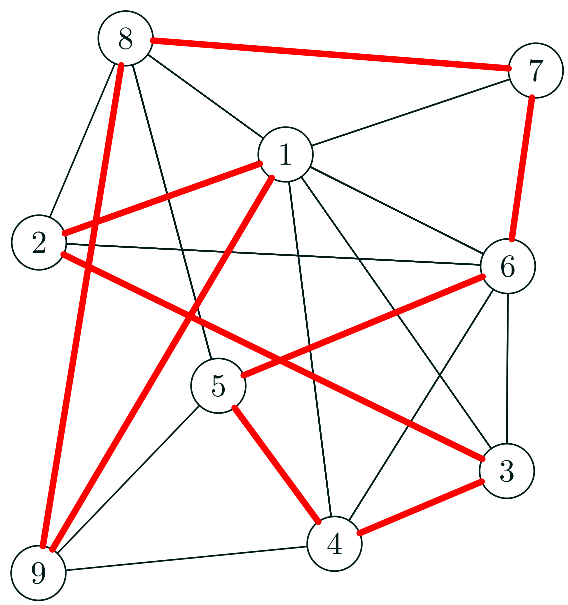

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Shortest Path and TSP

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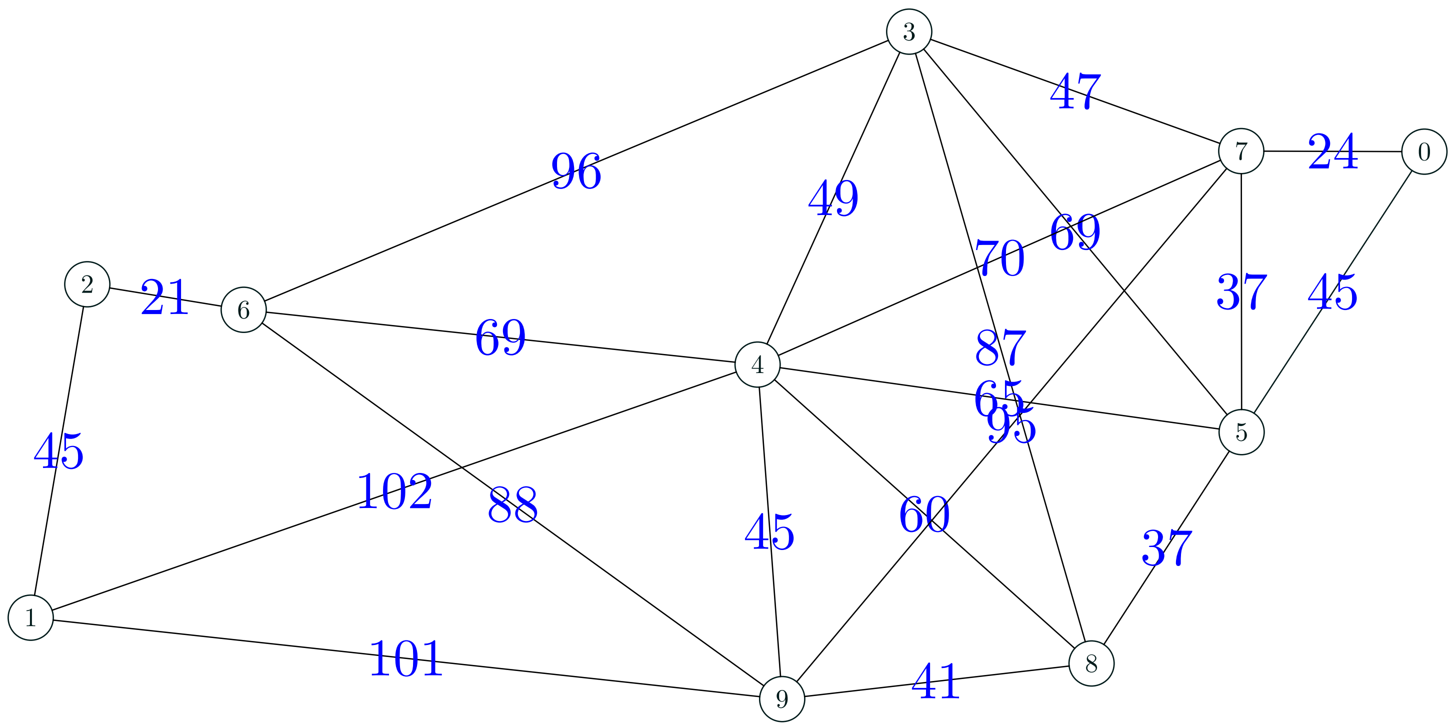
Minimum Spanning Tree

- Suppose we want to construct pylons connecting a number of cities using the least amount of cable



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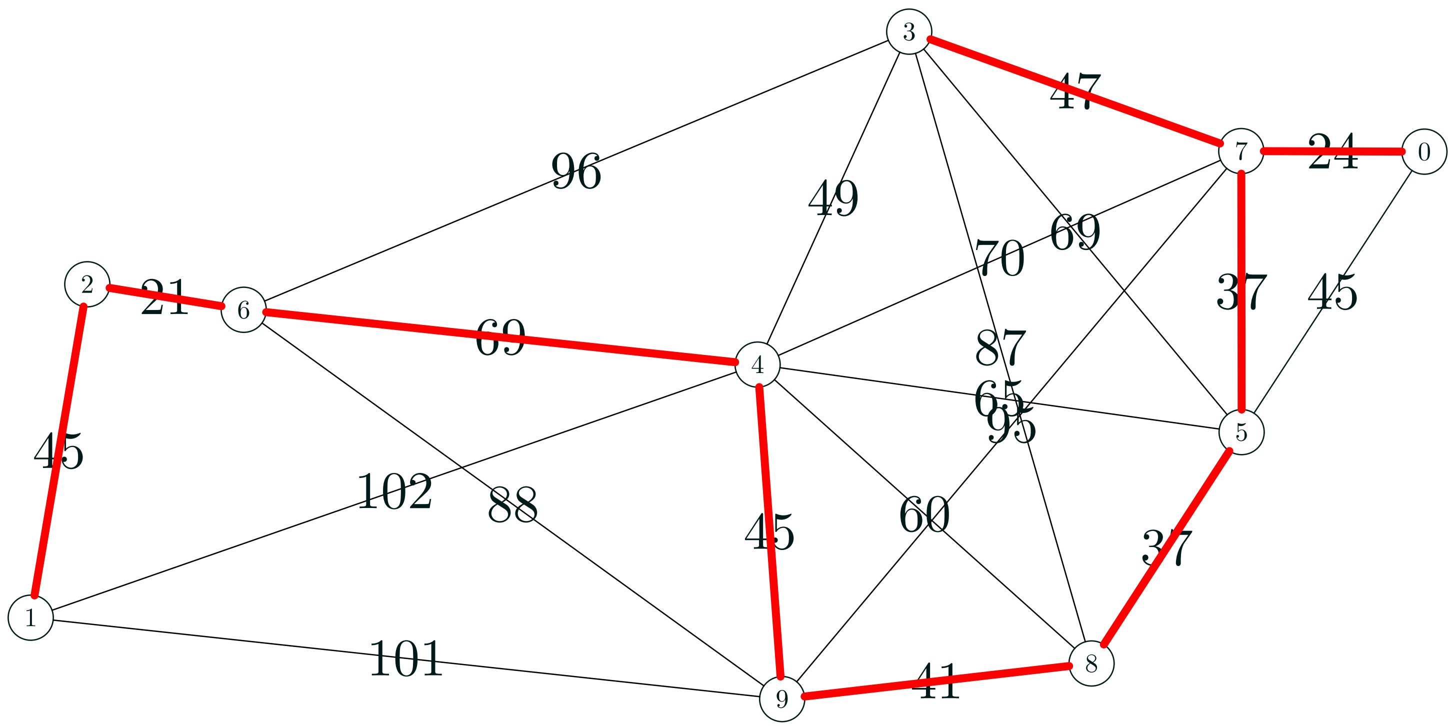

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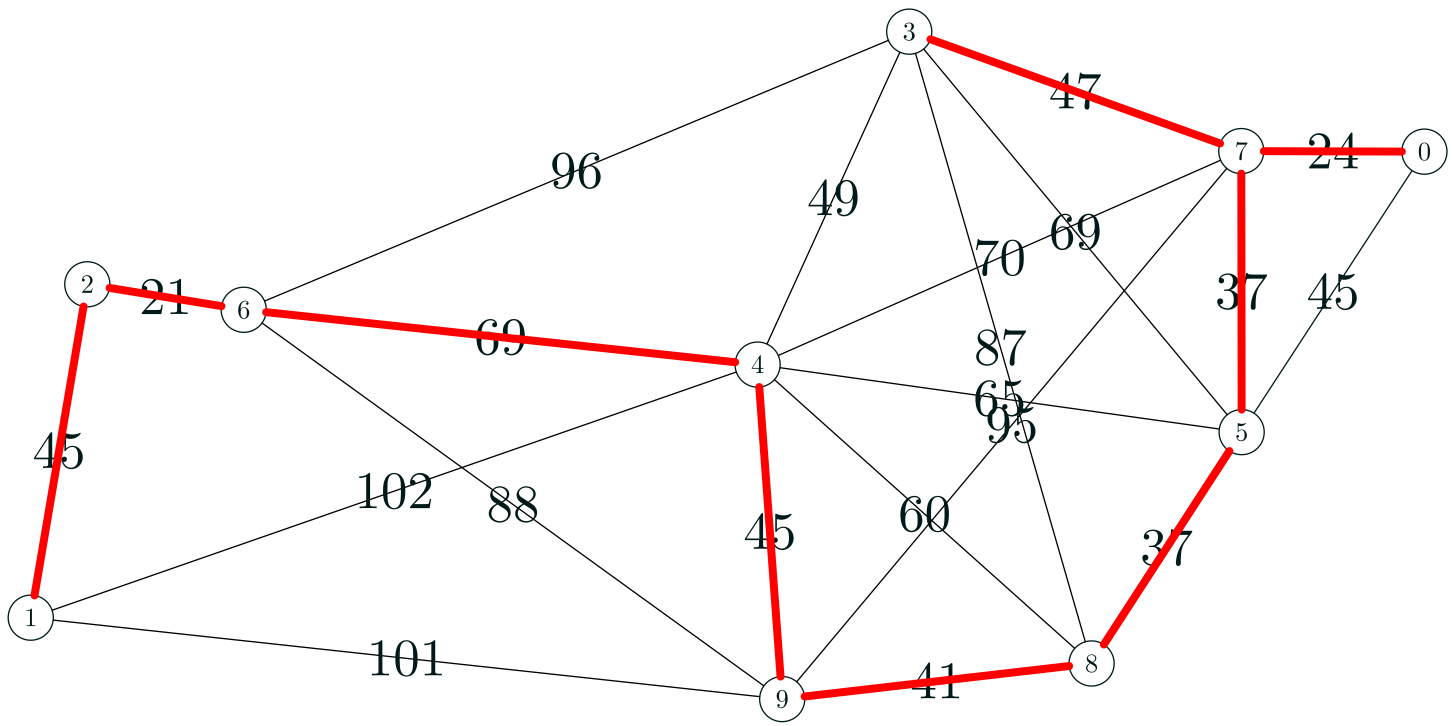

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Graph Partitioning

- The simplest version of this problem is to cut a graph into two equal halves so that you minimise the number of edges you cut
- If the edges are weighted then you want to minimise the sum of edges that are cut
- If the vertices are weighted you want to balance the sum of vertex weights in the two partitions
- An example of this problem is in dividing up a problem to run on a parallel computer
  - ⋆ Nodes are subtasks (weights on nodes are run times)
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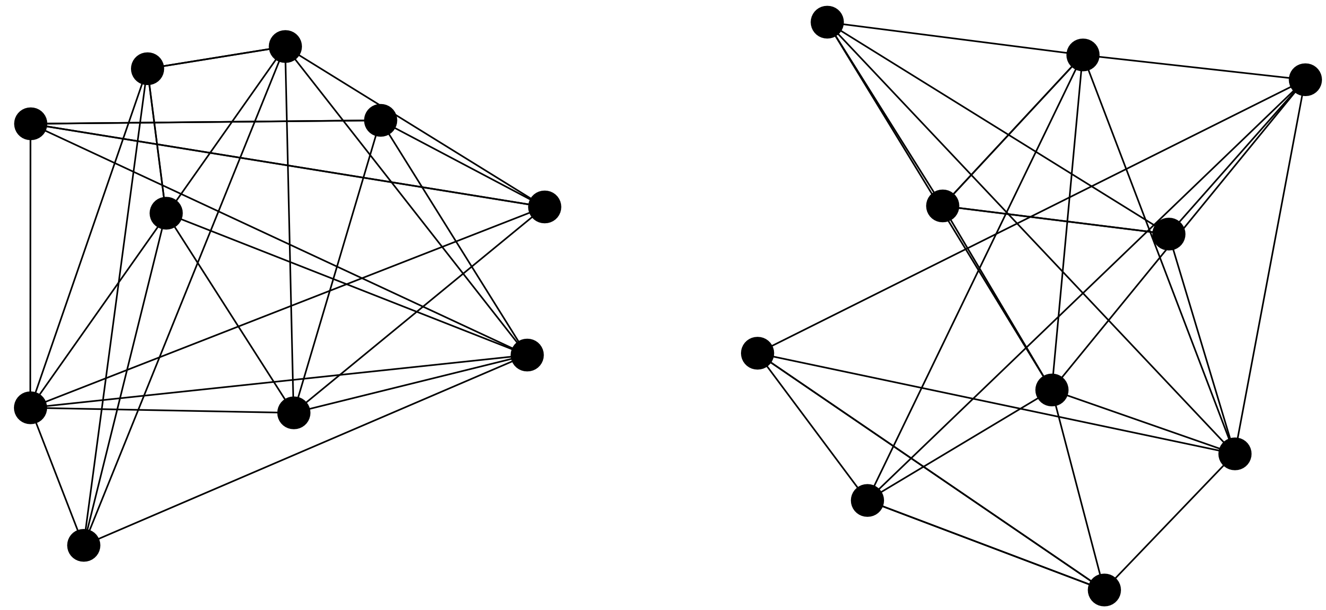

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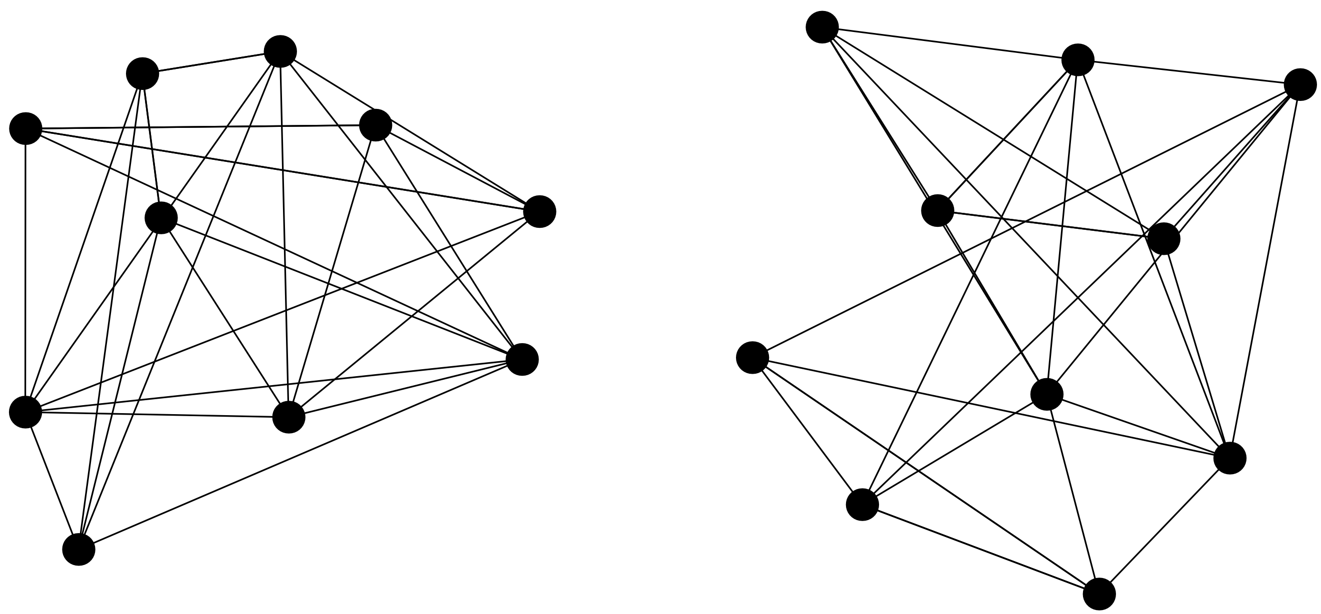

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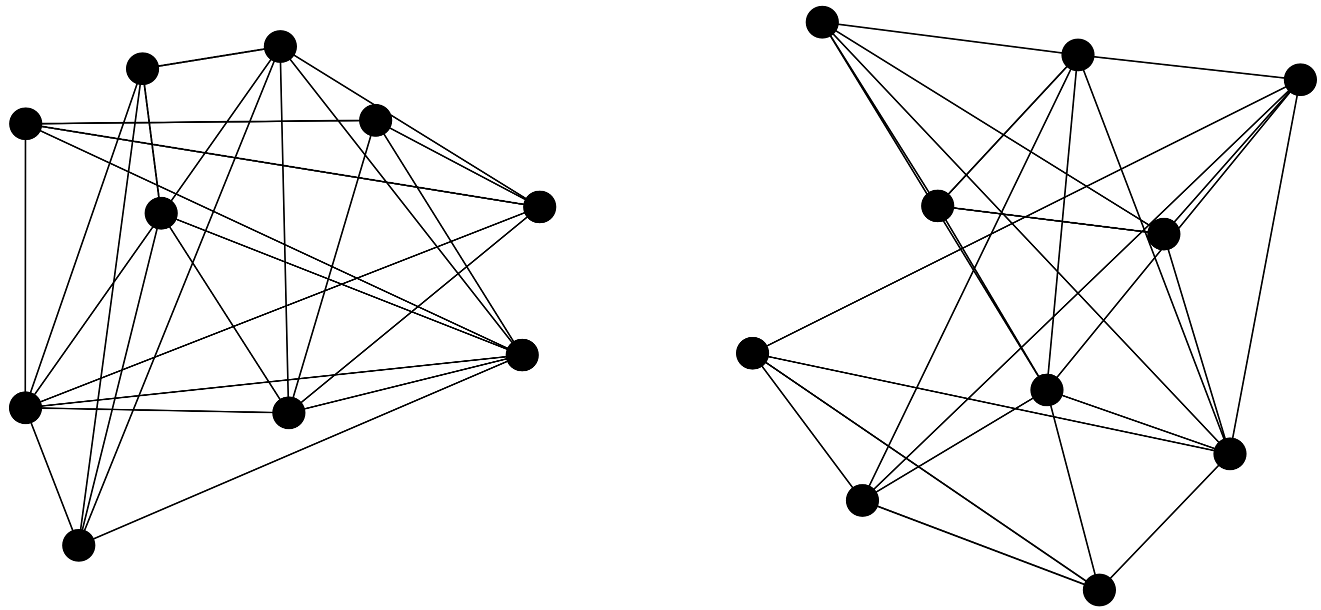

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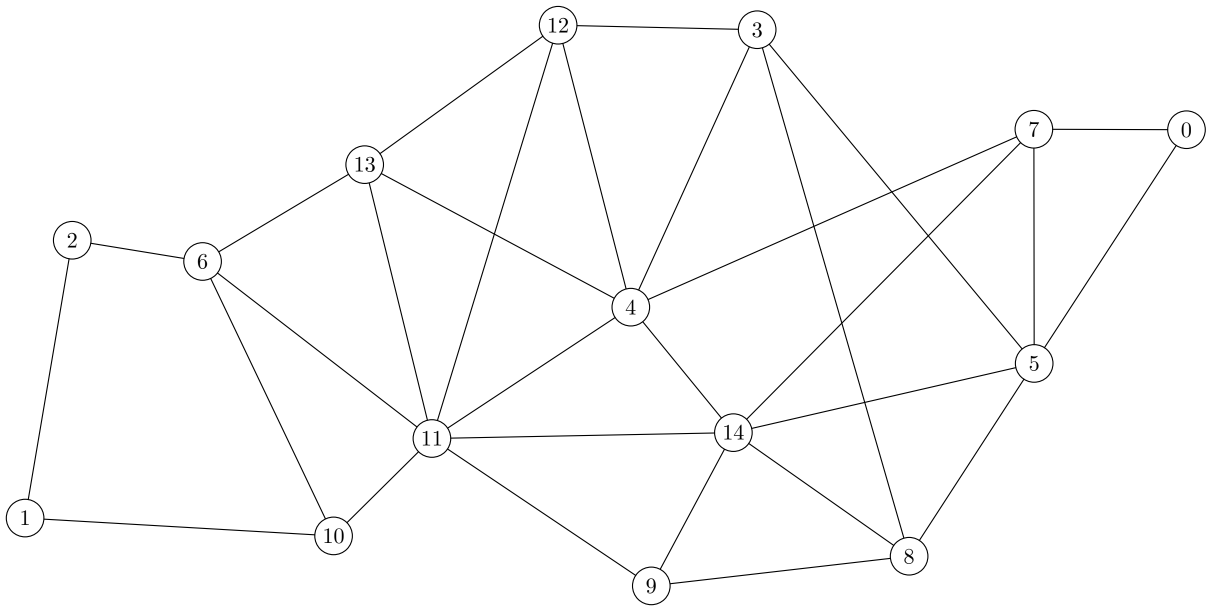
Vertex Cover

- How many guards do you need to cover all the corridors in a museum



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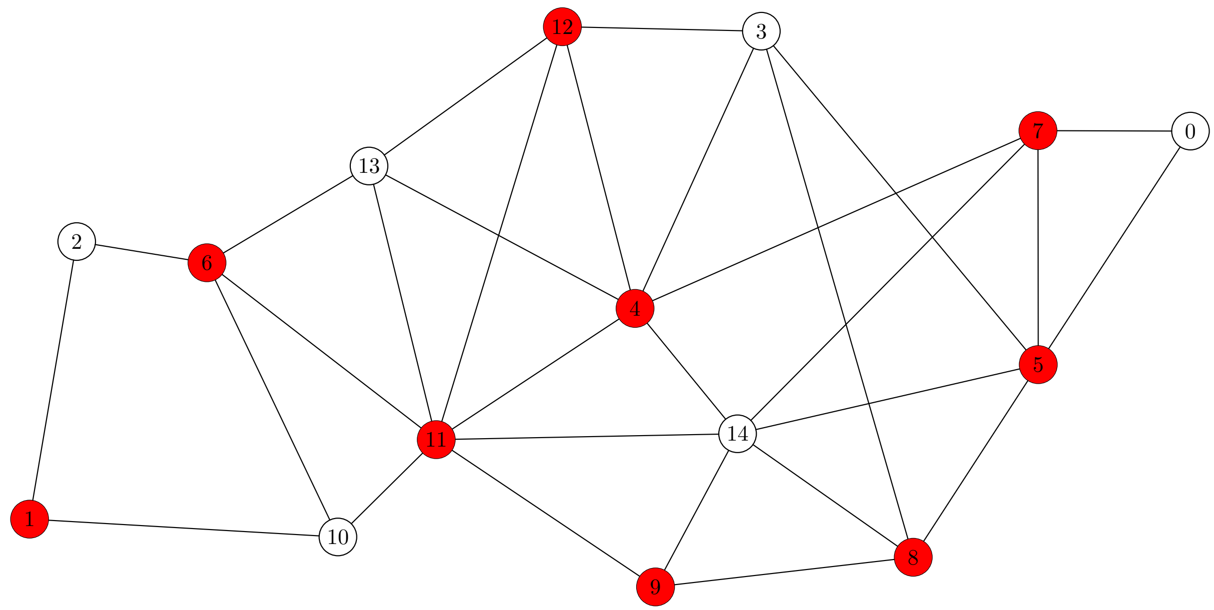

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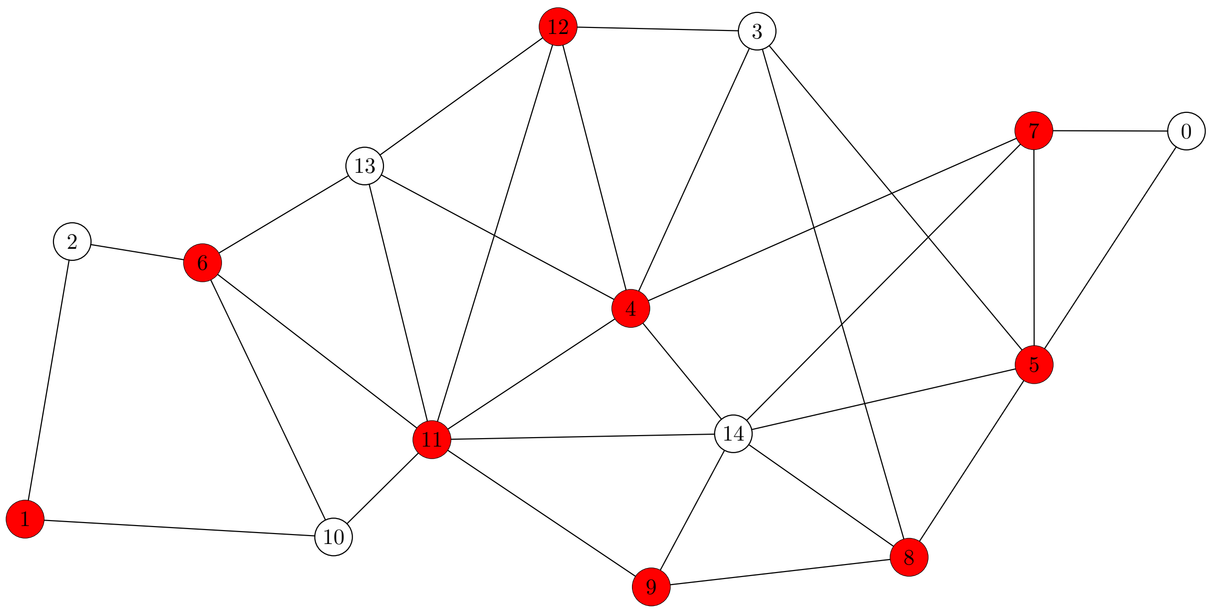

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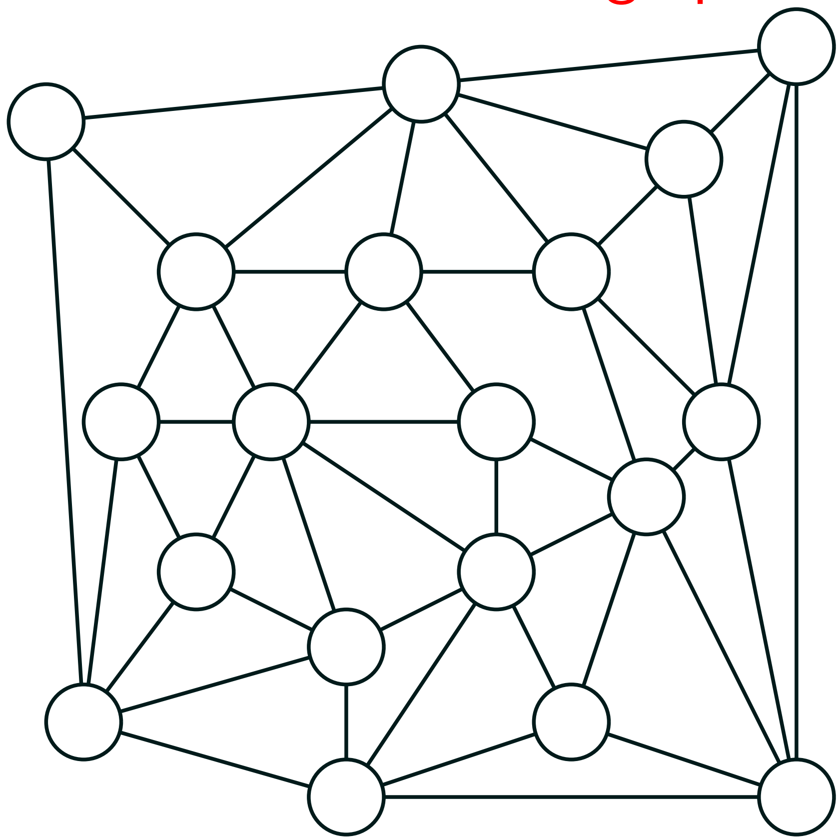

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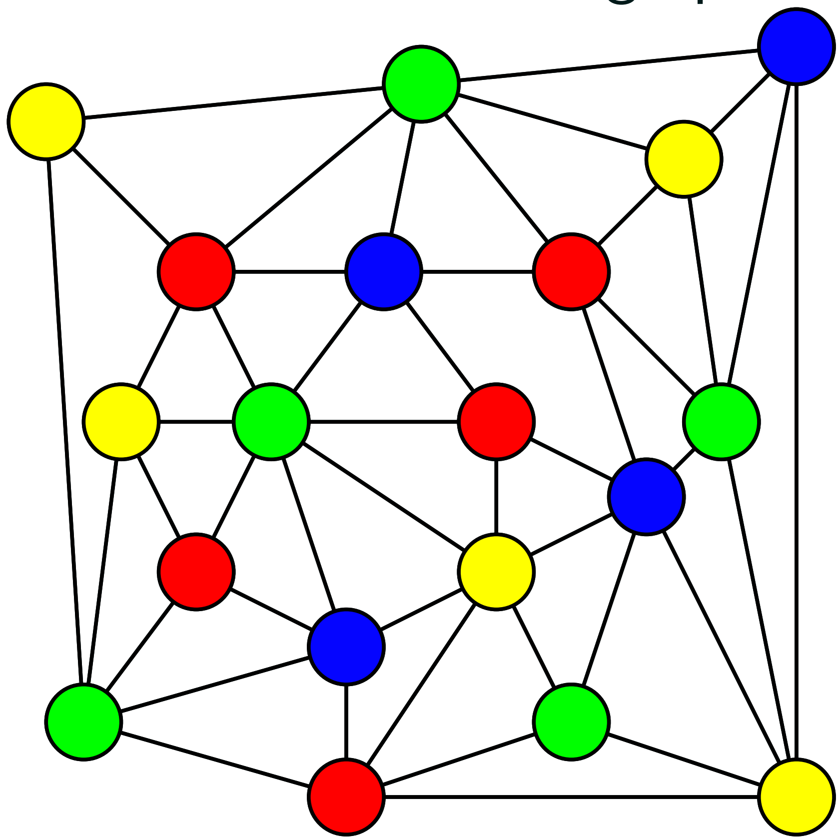

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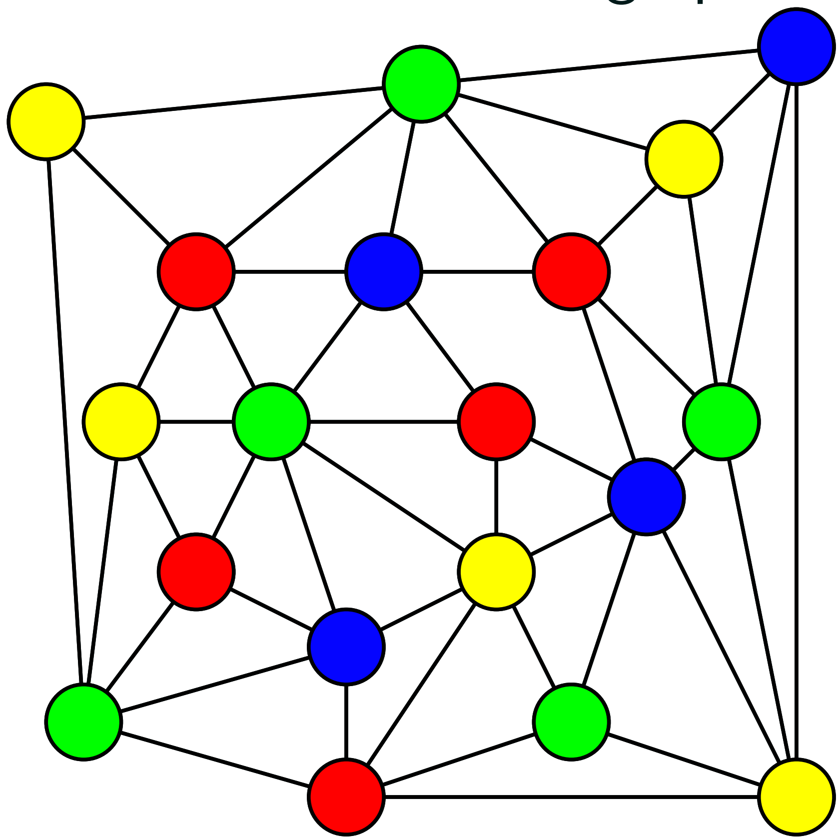

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Other Graph Problems

- These are only a sample of the many famous graph problems
- Others include
  - ⋆ Max-clique (hard)
  - ⋆ Maximal independent set (hard)
  - ⋆ Maximal flow problem (easy)
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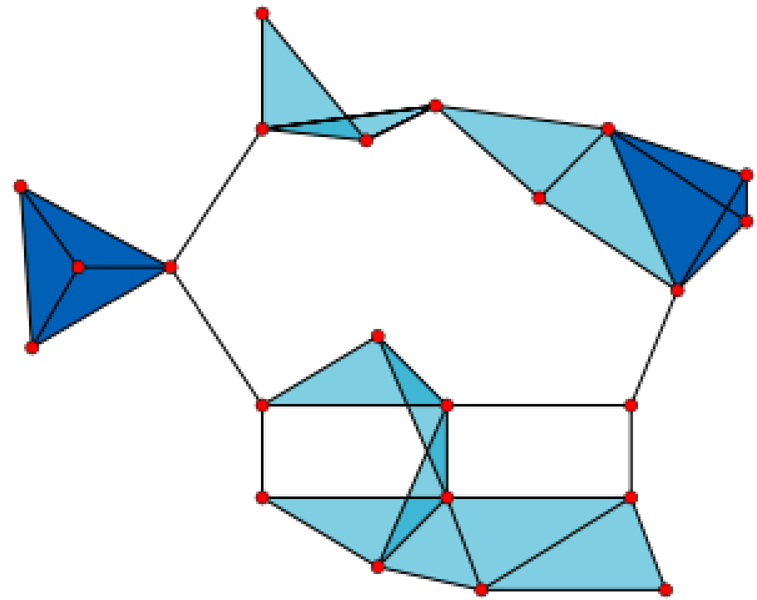

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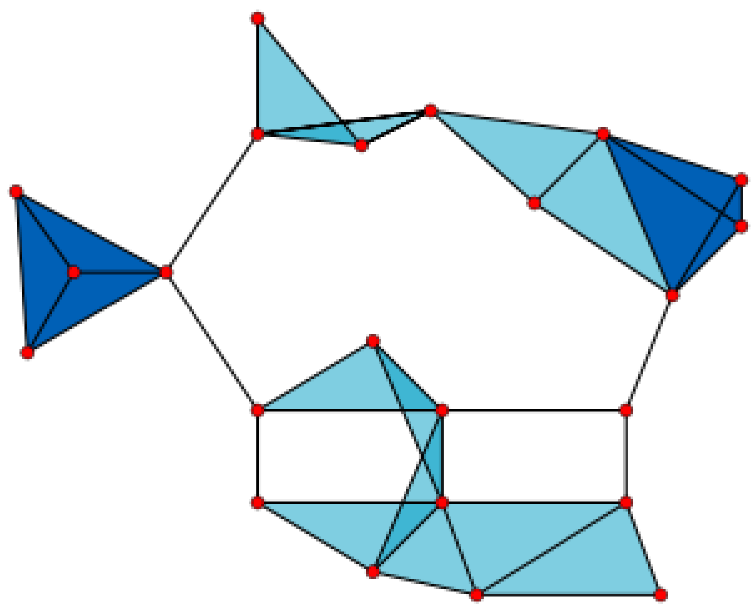

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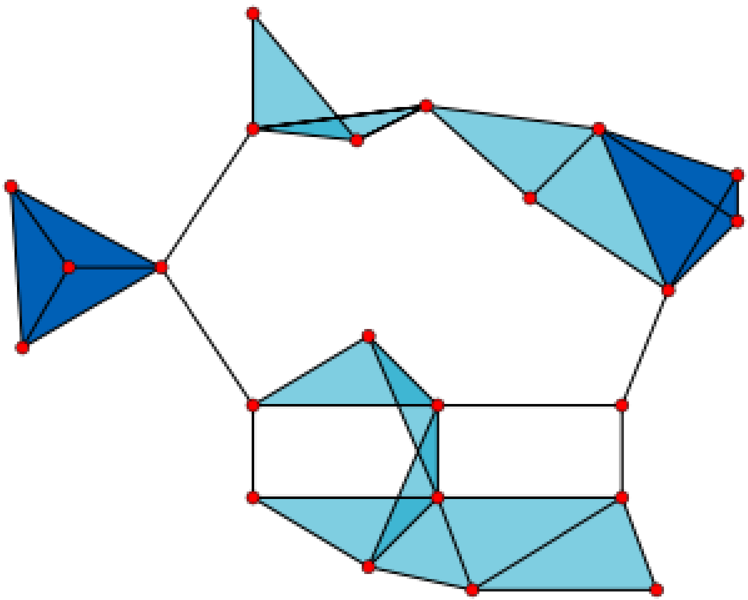

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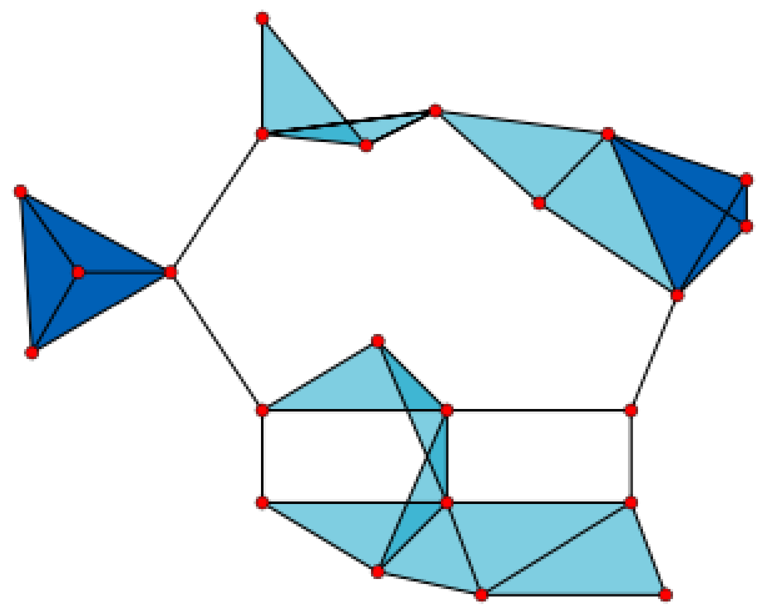

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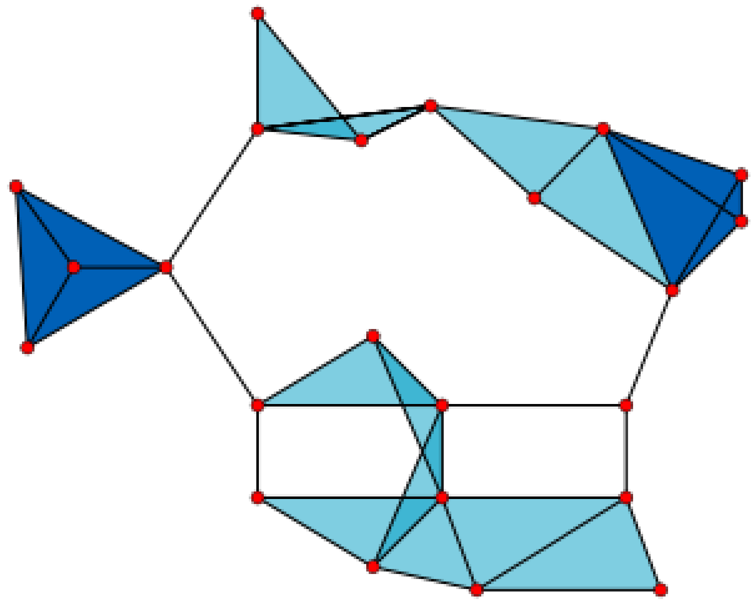

Lessons

- Graphs are an important method for abstracting problems
- They appear in a huge number of disparate fields
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- There are many problems which are believed to be hard—i.e. there aren't any efficient algorithms
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