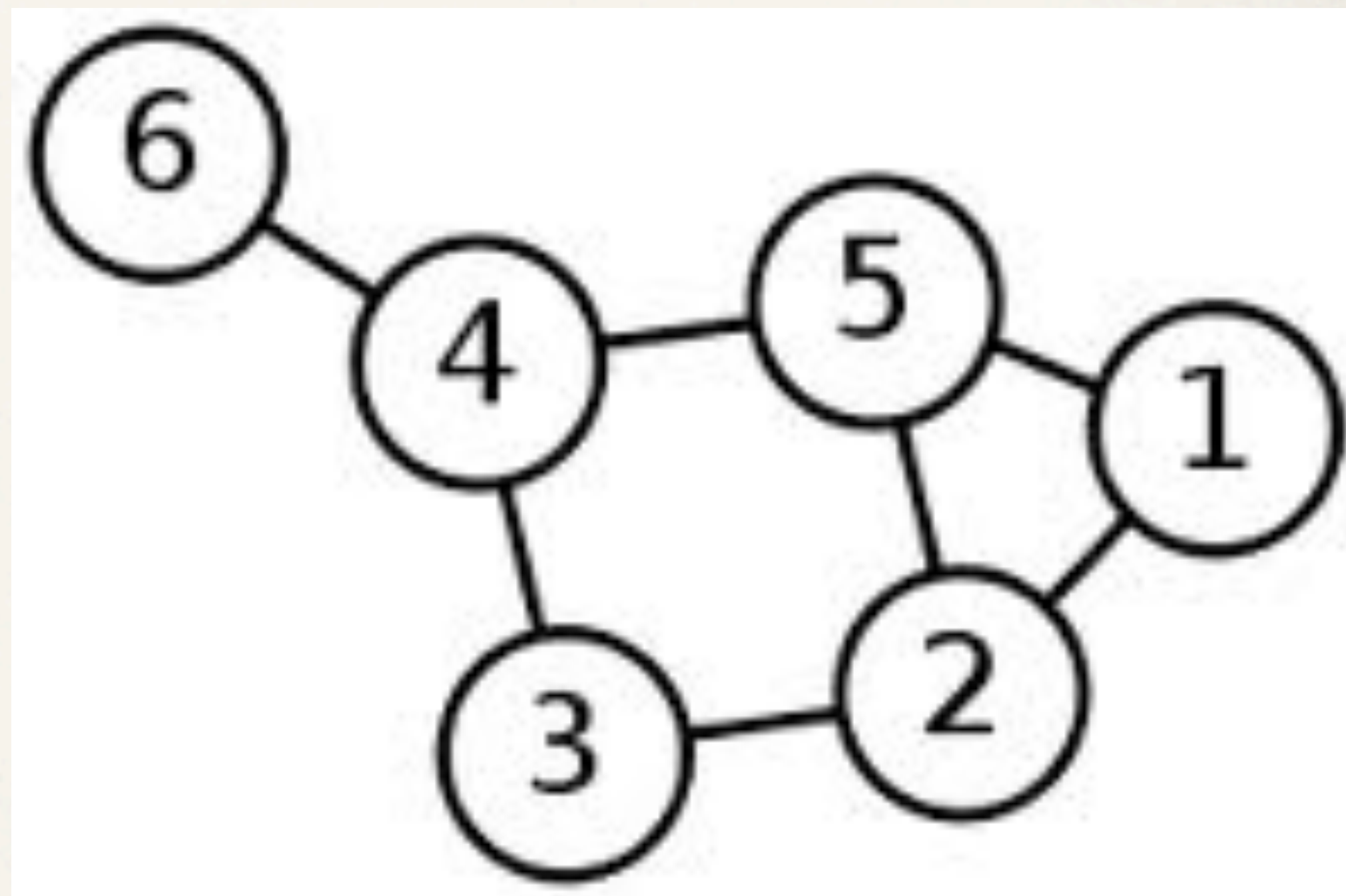


Graph Theory

Chapter 9

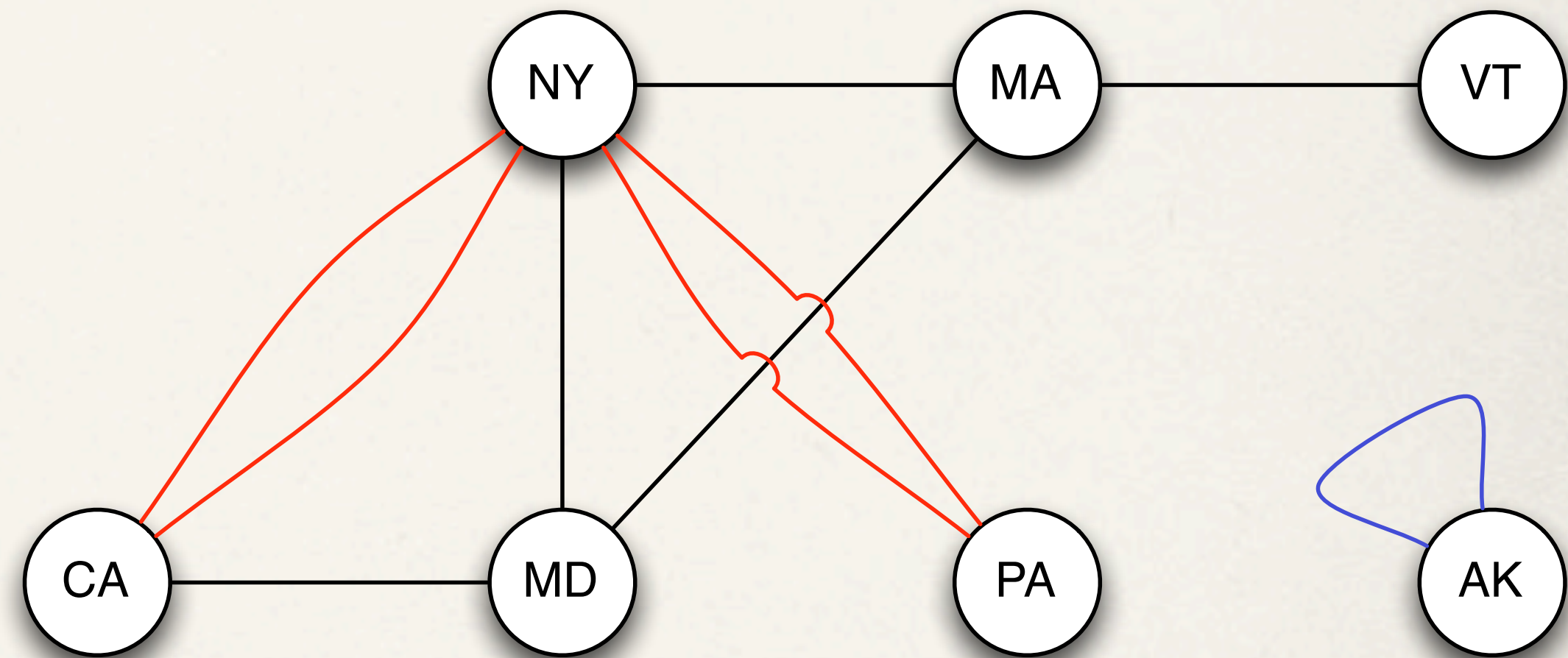
Graphs

- ❖ A **graph** consists of **nodes** and **edges**
 - ❖ The set of all nodes (or **vertices**) in a graph is V
 - ❖ The set of all edges in a graph is E
 - ❖ Each edge **connects** one or two vertices (called its' **endpoints**)
- ❖ A graph is formally represented (V, E)
- ❖ We will (usually) stick to **simple graphs**



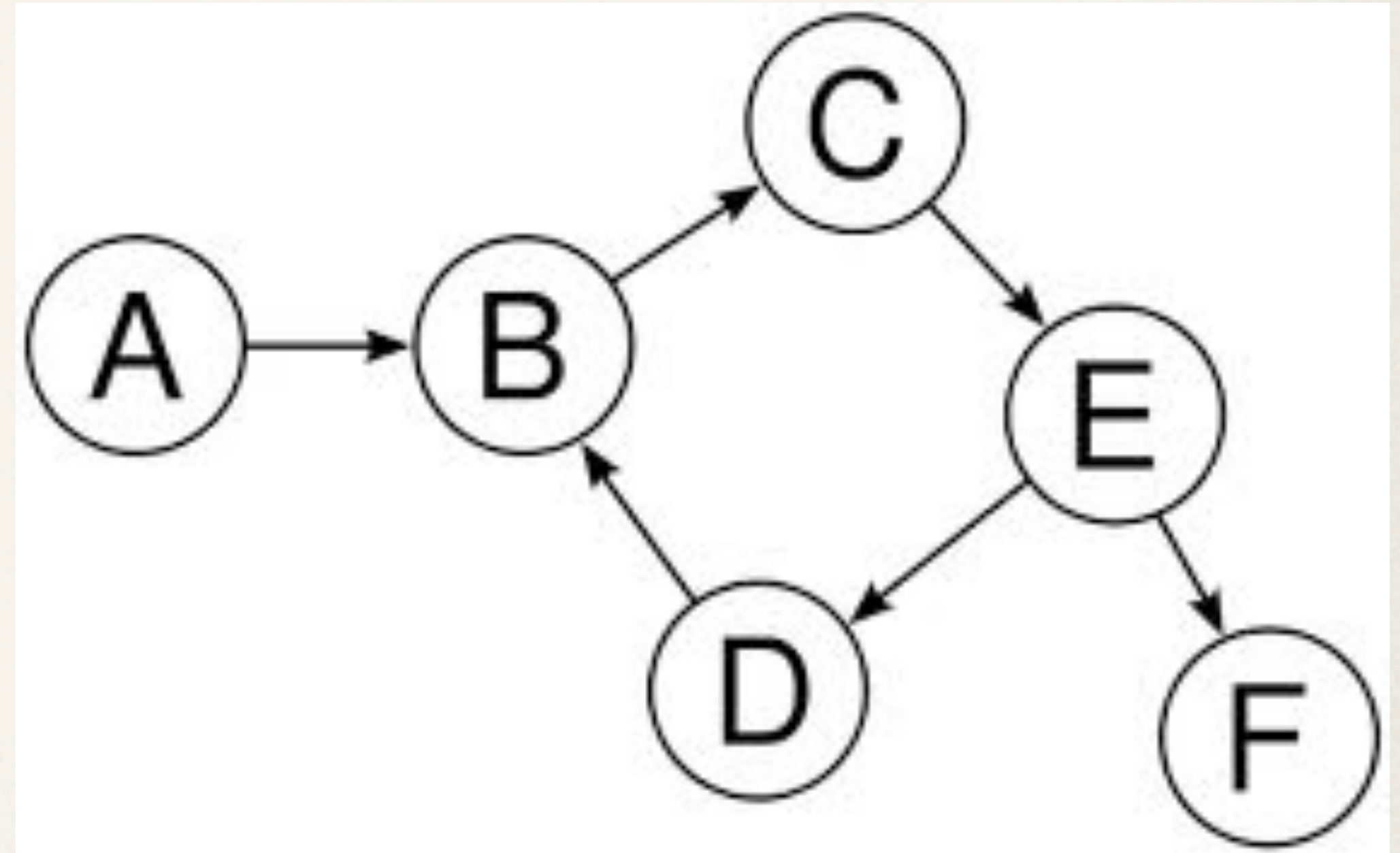
Simple Graphs

- ❖ **No loops**
 - ❖ Endpoints of an edge are distinct
- ❖ **No multiple edges**
 - ❖ between the same two vertices

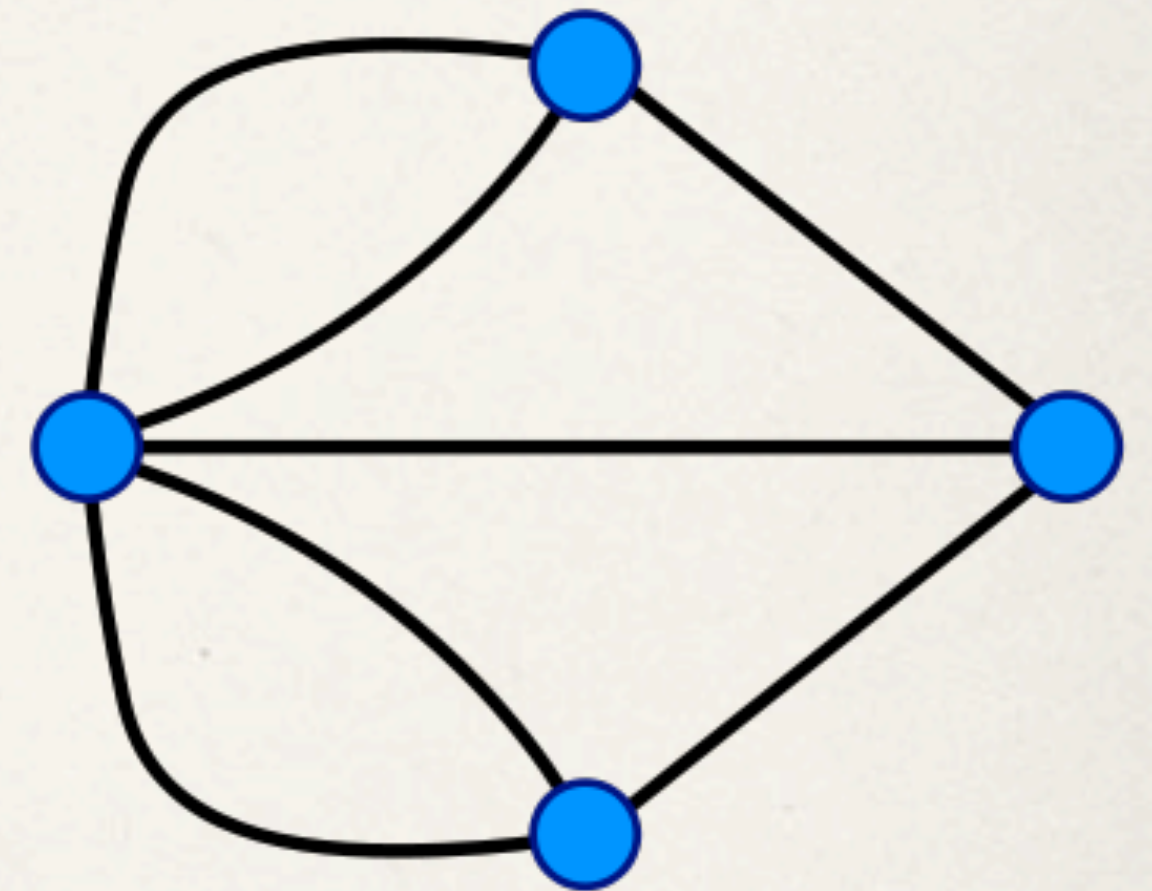
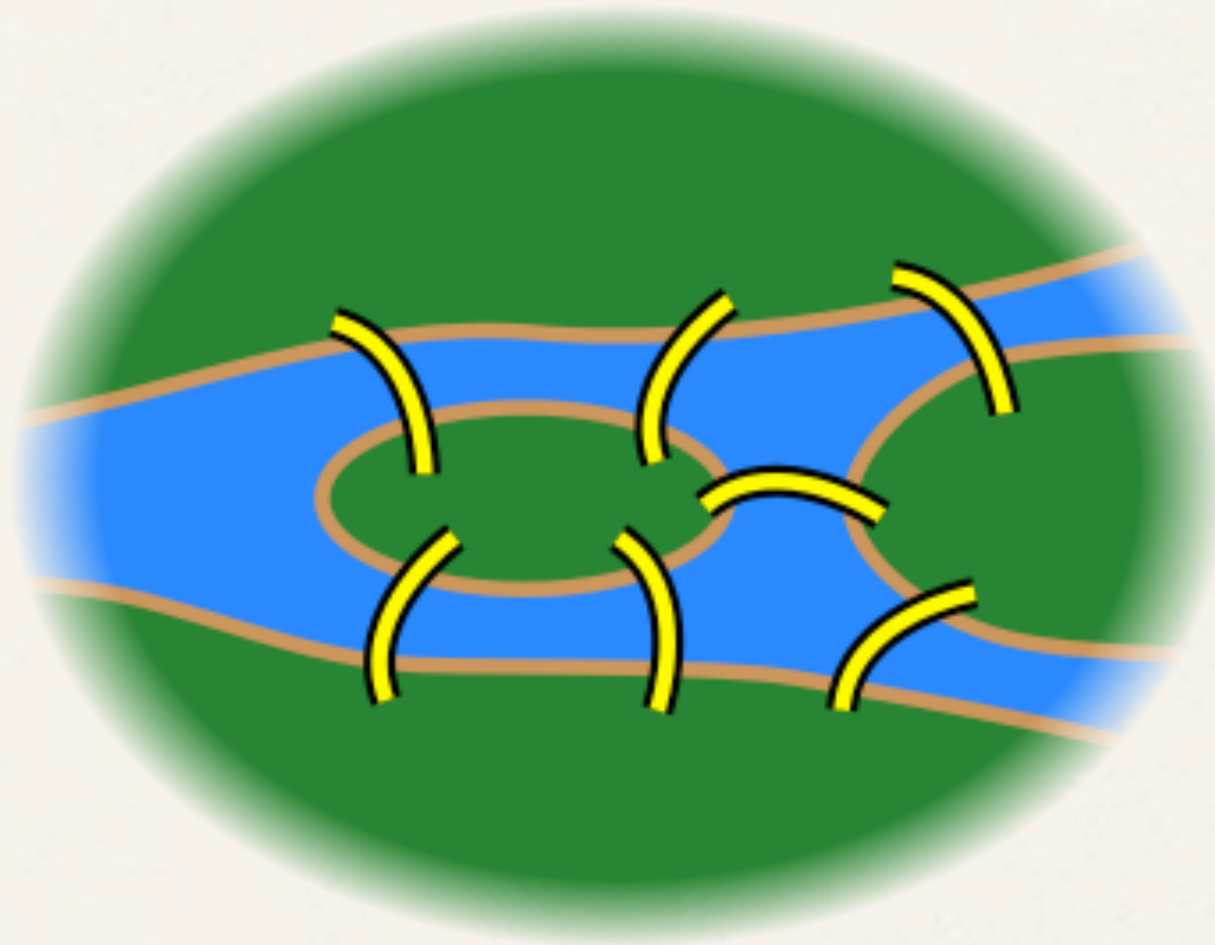
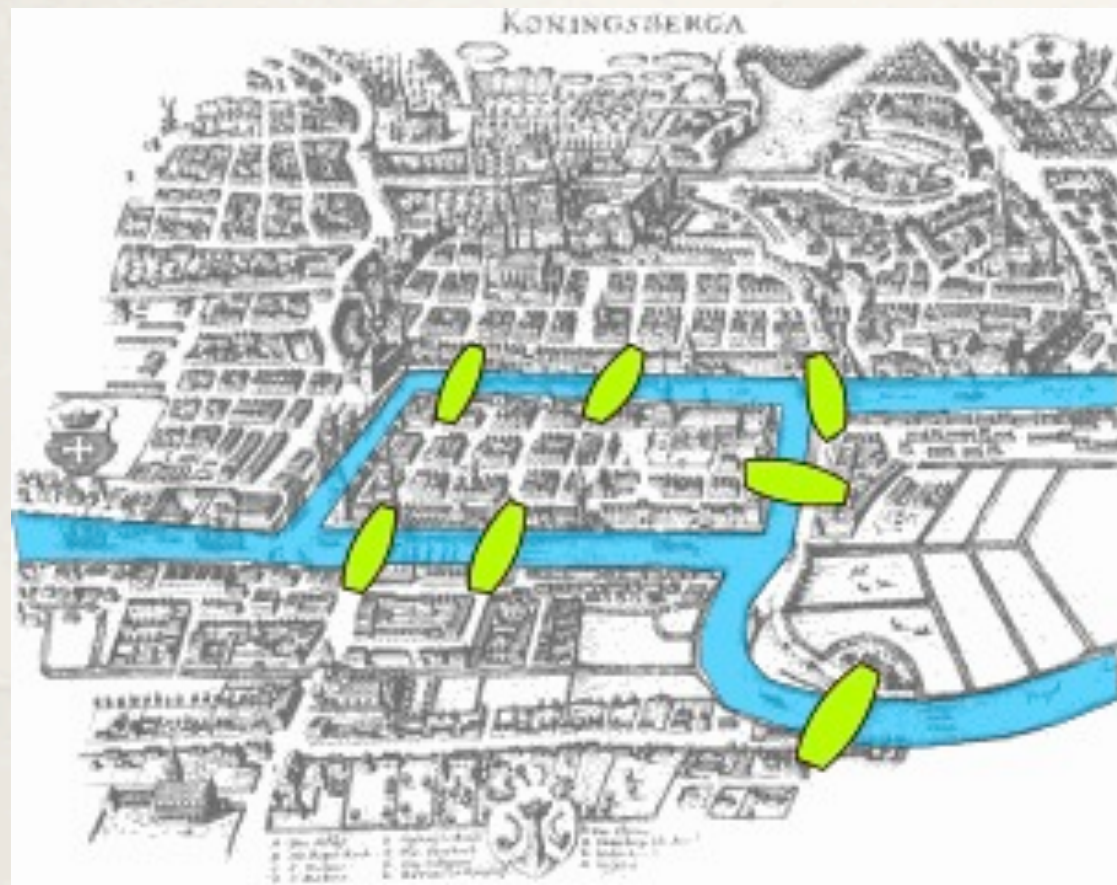


Directed Graphs

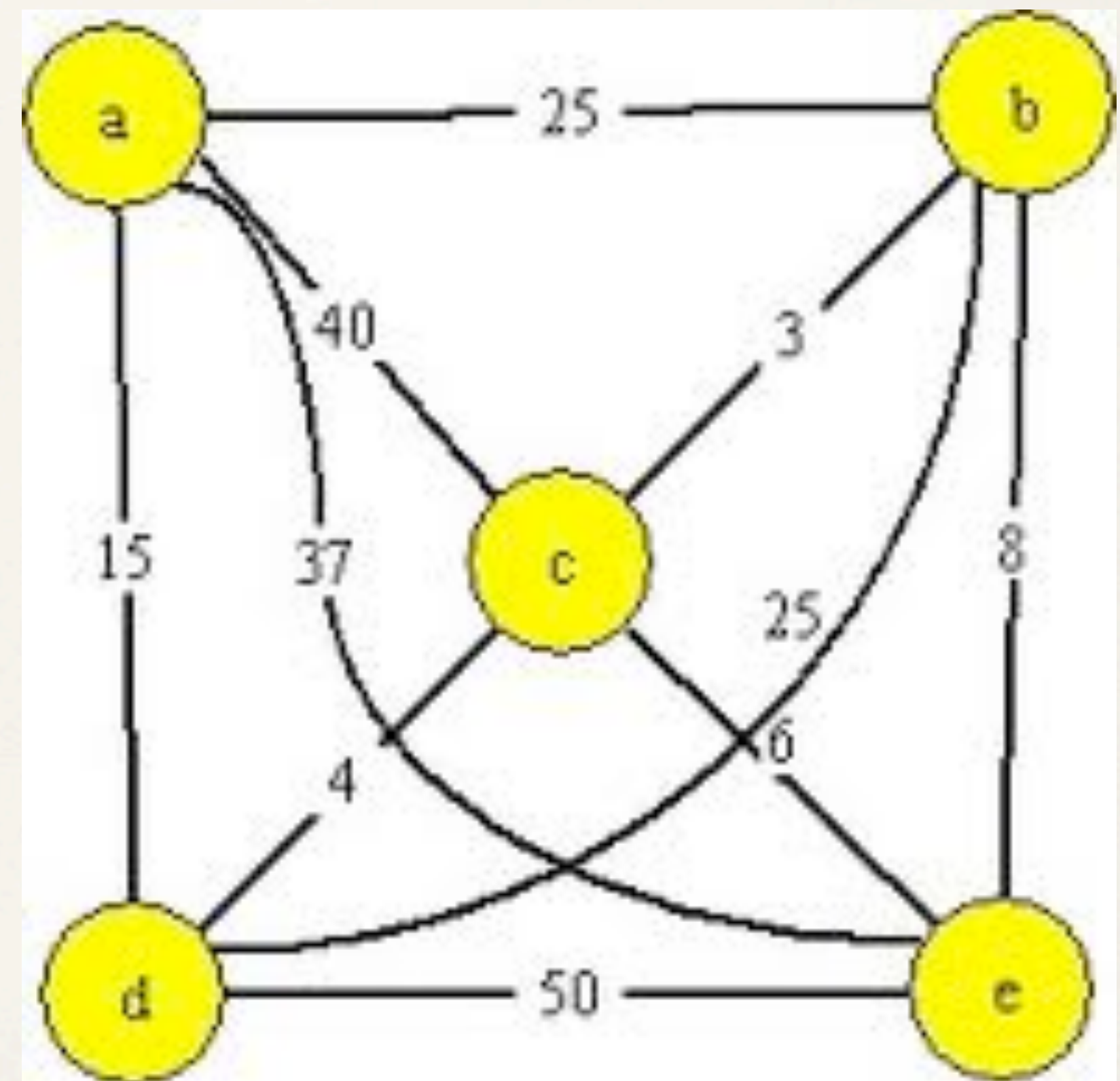
- ❖ A **directed edge**, or **arc** connects two vertices, but has a **start** and **end**
 - ❖ Formally an arc (u, v) starts at u and ends at v
- ❖ A graph with directed edges is a **directed graph** or **digraph**



Seven Bridges of Königsberg

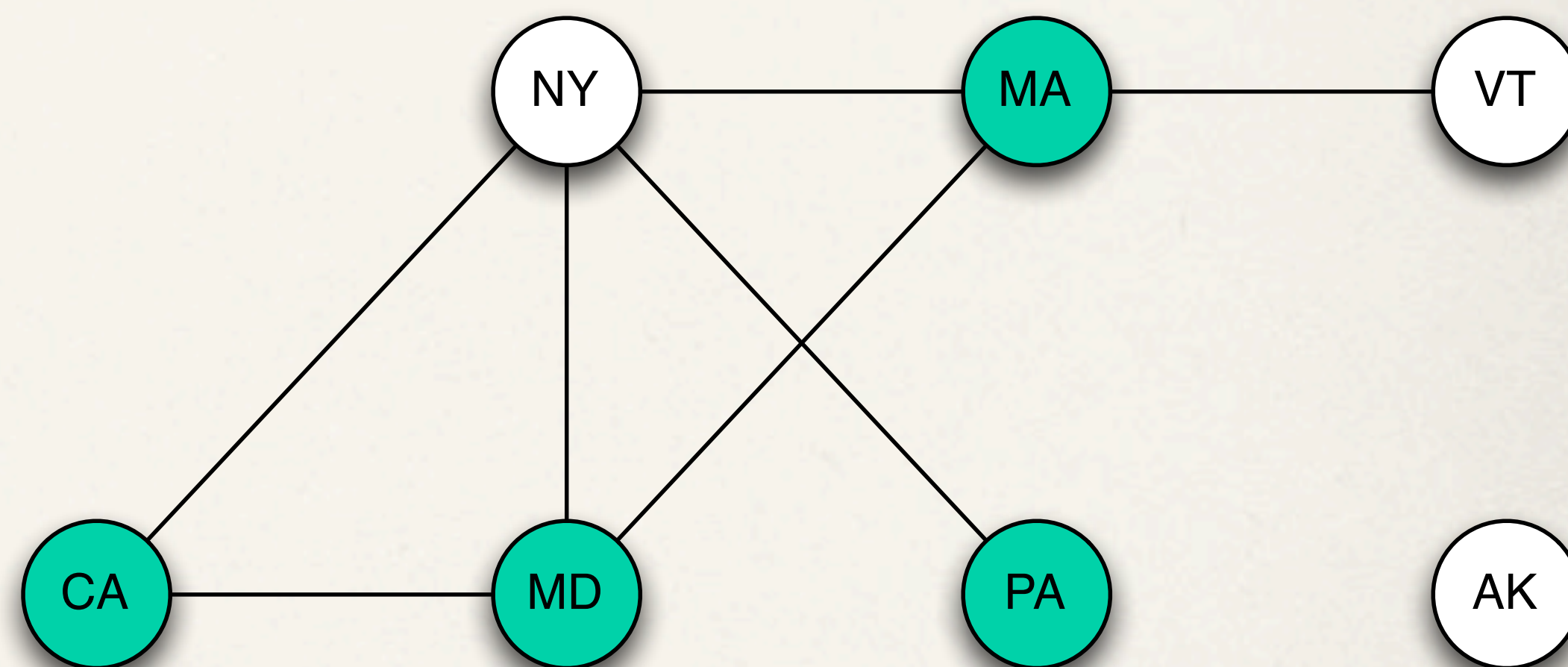


Traveling Salesman Problem



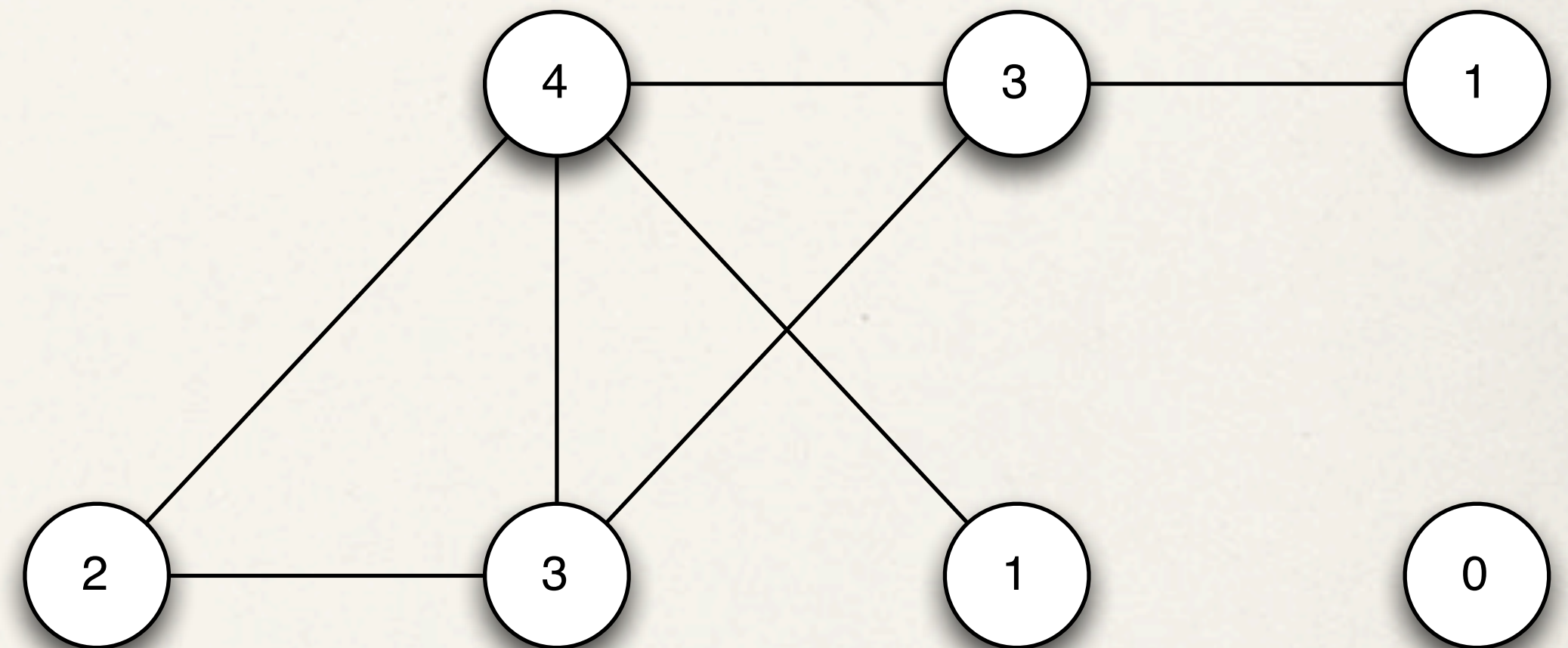
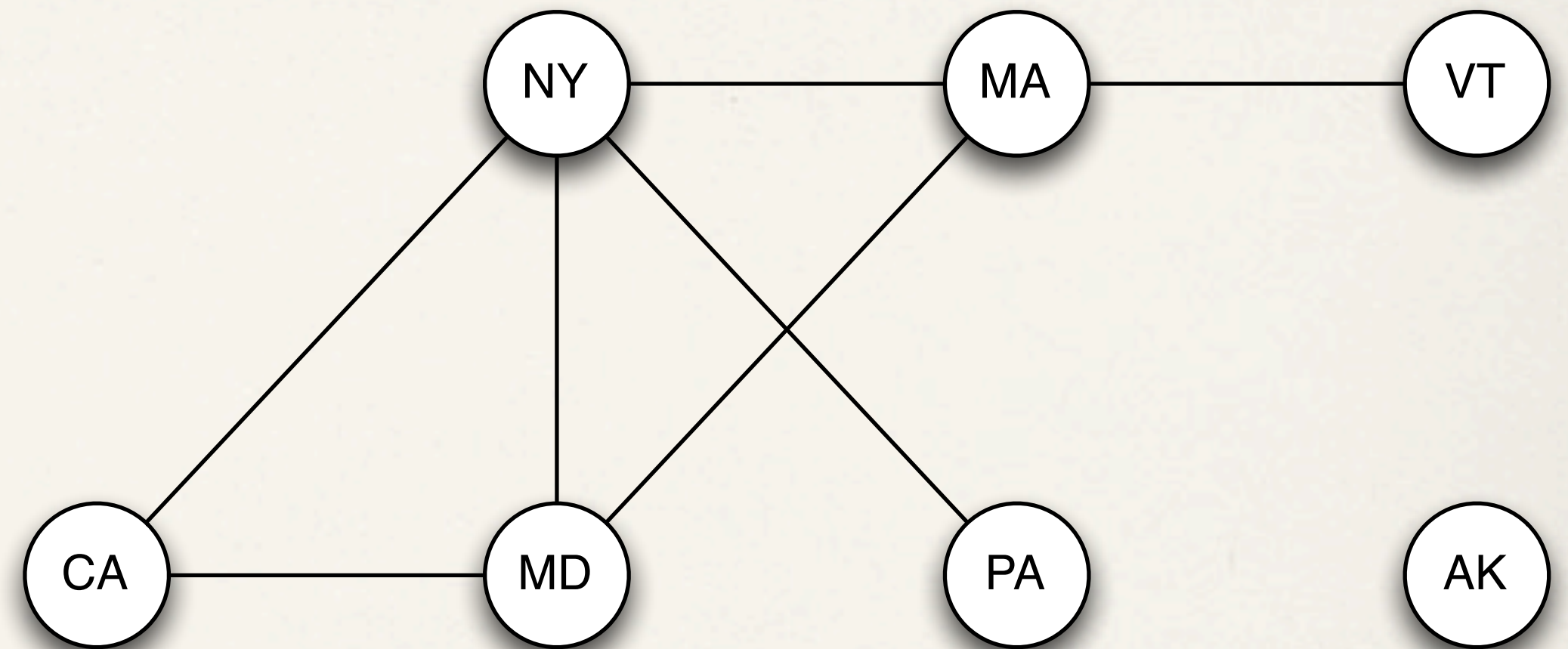
Terminology

- ❖ Two vertices u and v in an undirected graph G are **neighbors** if there exists an edge e in G which connects u , and v
 - ❖ Equivalently, u and v are said to be **adjacent** if there exists an edge which connects them
- ❖ The set of all neighbors of u (u 's **neighborhood**) is $N(u)$
- ❖ This is defined similarly for sets of vertices



Degree

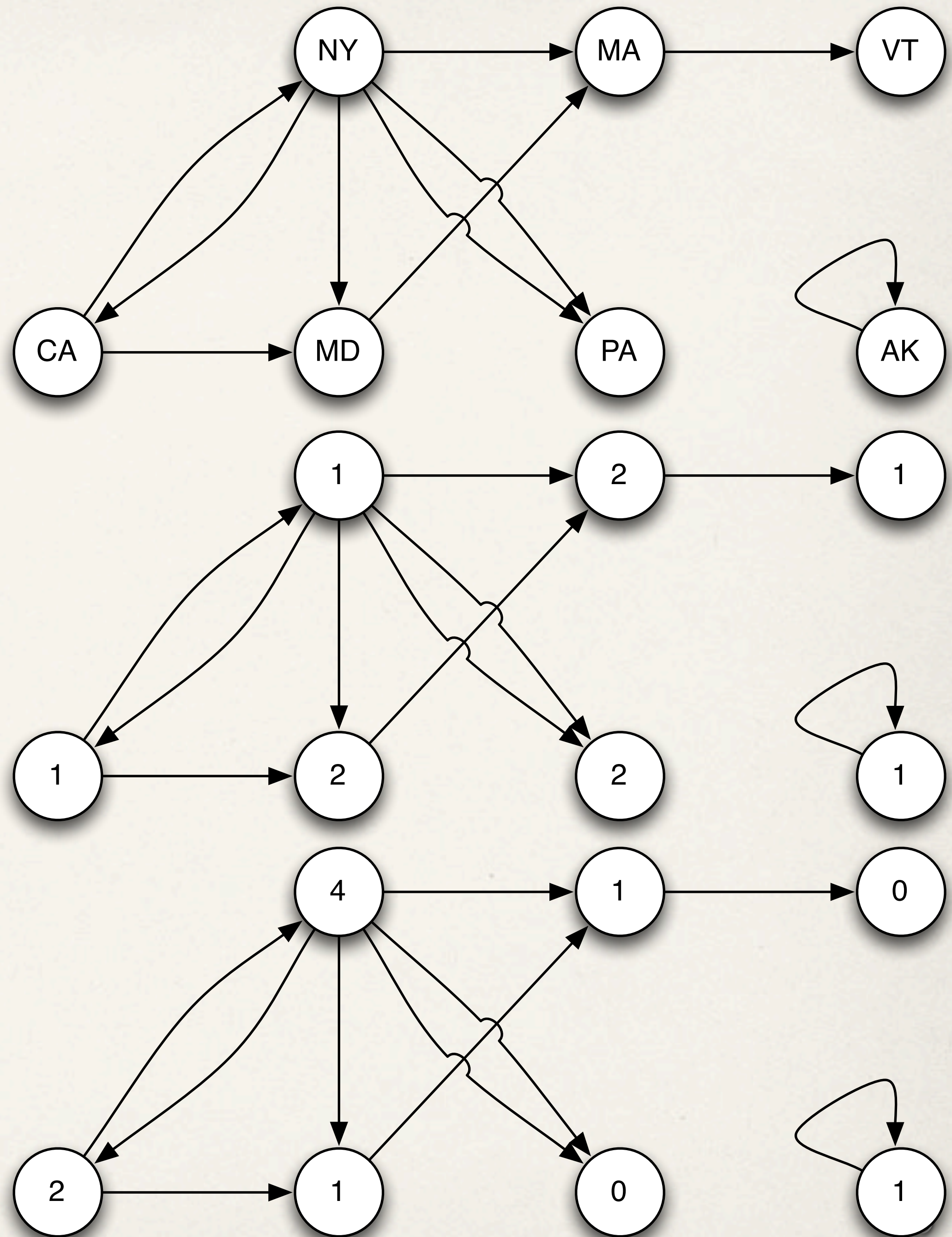
- ❖ If an edge e has vertex u as one of its' endpoints, e is **incident with u**
- ❖ The **degree** of a node u , $\text{deg}(u)$, is the count of the number of edges incident with u
- ❖ A node with degree 0 is **isolated**



In/Out -Degree

- ❖ In directed graphs, separate degrees
 - ❖ **In-Degree:** # of edges ending at u
 $\text{deg}^-(u)$
 - ❖ **Out-Degree:** # of edges starting at u
 $\text{deg}^+(u)$
- ❖ Vertices with in-degree 0 are **sources**
- ❖ Vertices with out-degree 0 are **sinks**

$$\sum_{v \in V} \text{deg}^-(v) = \sum_{v \in V} \text{deg}^+(v) = |E|$$



Representations

Graph ADT

- ❖ V = set of vertices
 E = set of edges
- ❖ Three operations
 - ❖ `getDegree(u)`
 - ❖ `getAdjacent(u)`
 - ❖ `isAdjacentTo(u, v)`

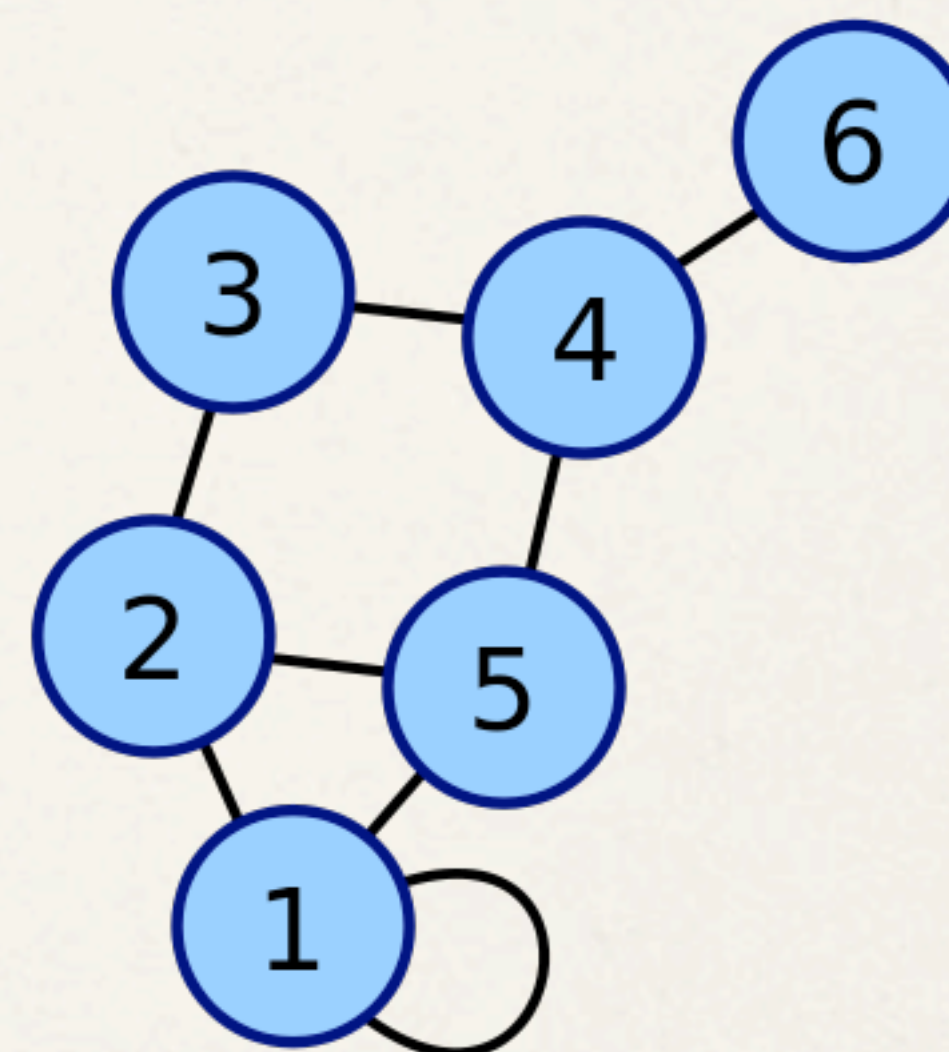
Graph Representations

- ❖ **Adjacency list**

- ❖ For each vertex, list the adjacent vertices
- ❖ Good for **sparse** graphs with few edges

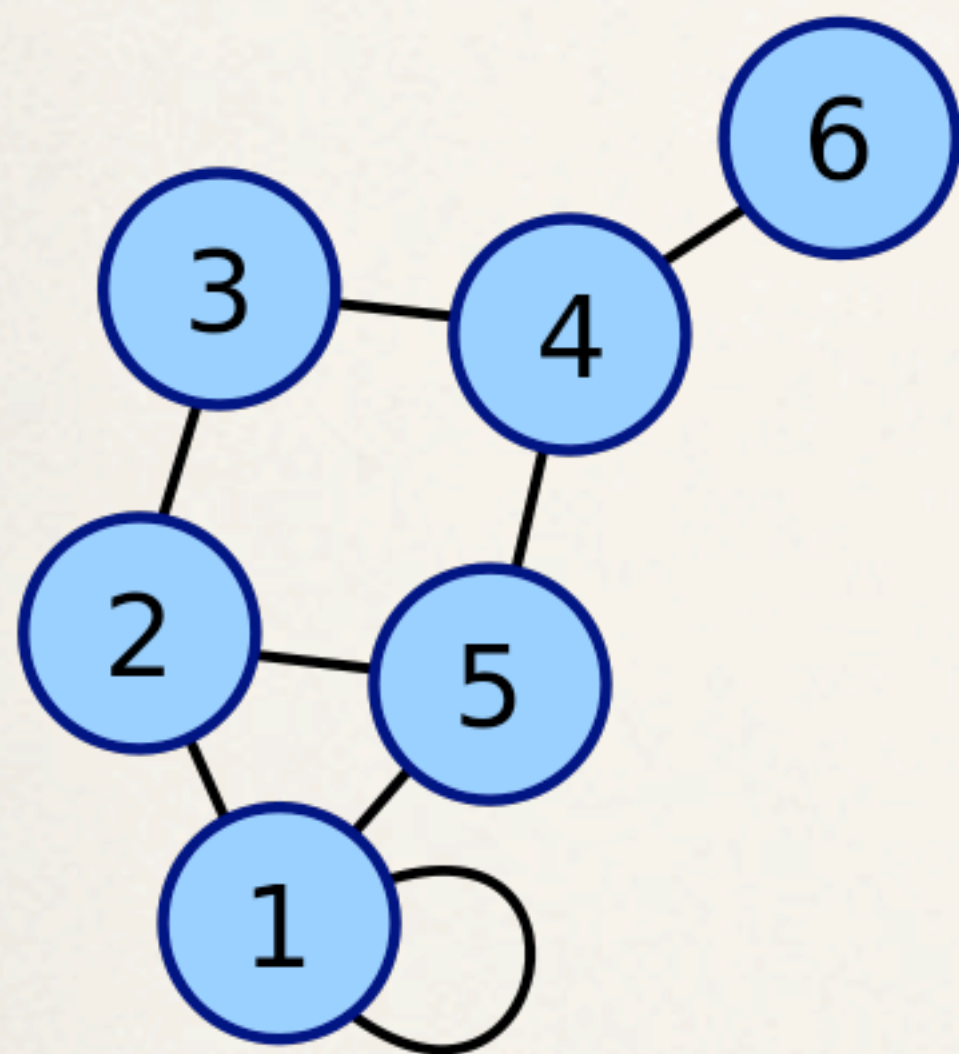
- ❖ **Adjacency matrix**

- ❖ 2D matrix of 1s and 0s
 - ❖ 1 iff there is an edge from i to j
- ❖ Good for **dense** graphs with many edges



Graph Representations

Graph



Adjacency List

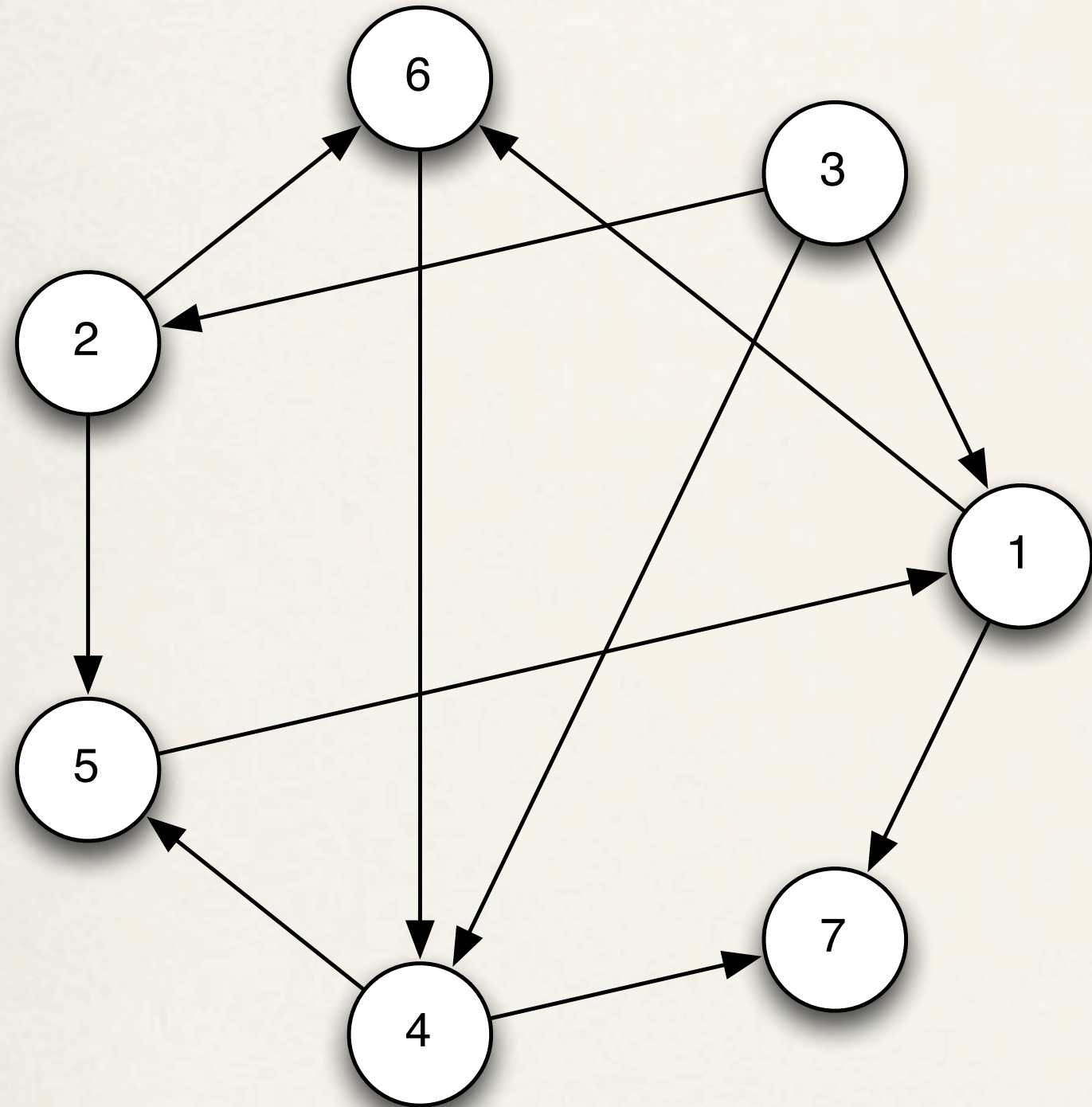
V	Adj.
1	1,2,5
2	1,3,5
3	2,4
4	3,5,6
5	1,2,4
6	4

Adjacency Matrix

V	1	2	3	4	5	6
1	1	1	0	0	1	0
2	1	0	1	0	1	0
3	0	1	0	1	0	0
4	0	0	1	0	1	1
5	1	1	0	1	0	0
6	0	0	0	1	0	0

Directed Graphs

Graph



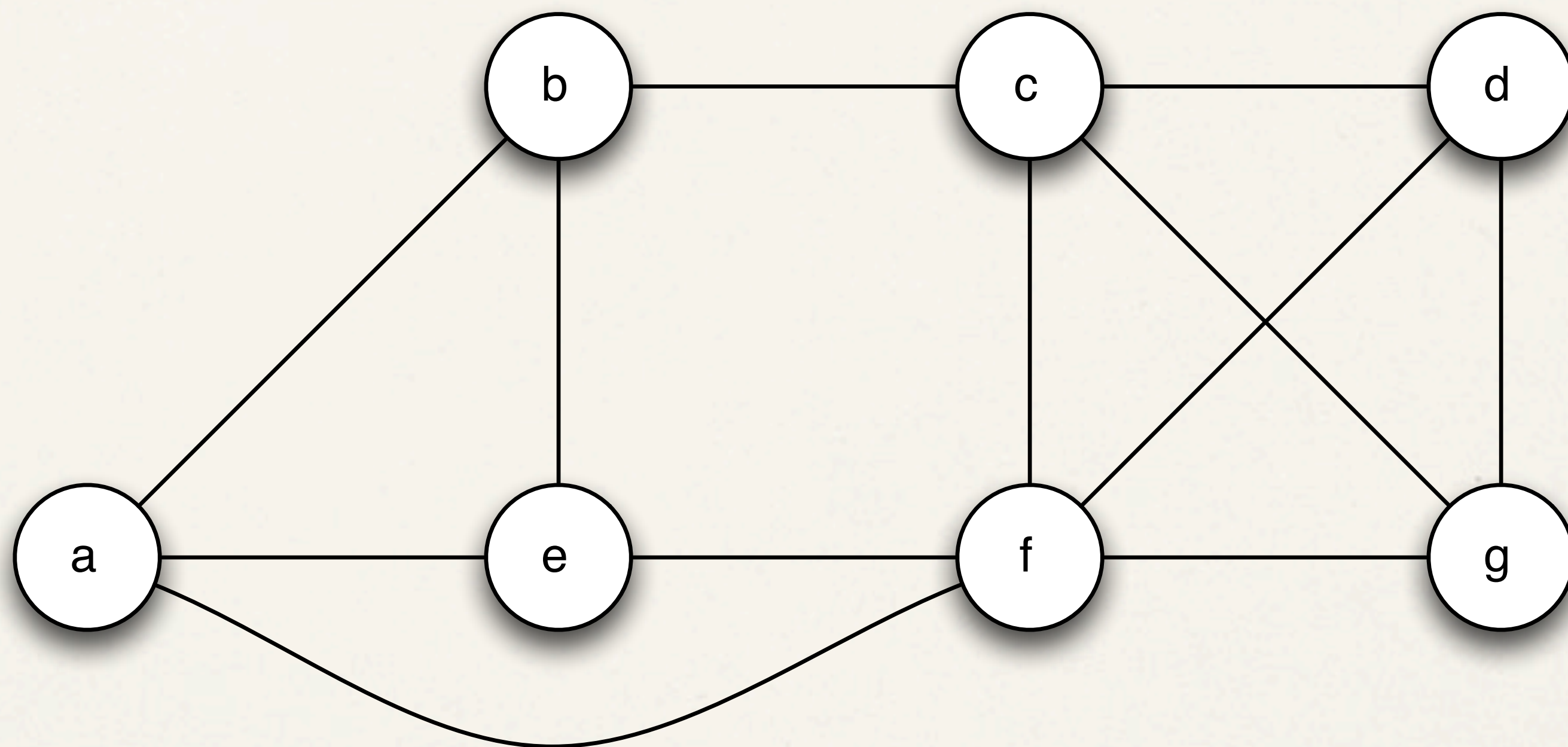
Adjacency List

V	Term
1	6,7
2	5,6
3	1,2,4
4	5,7
5	1
6	4
7	

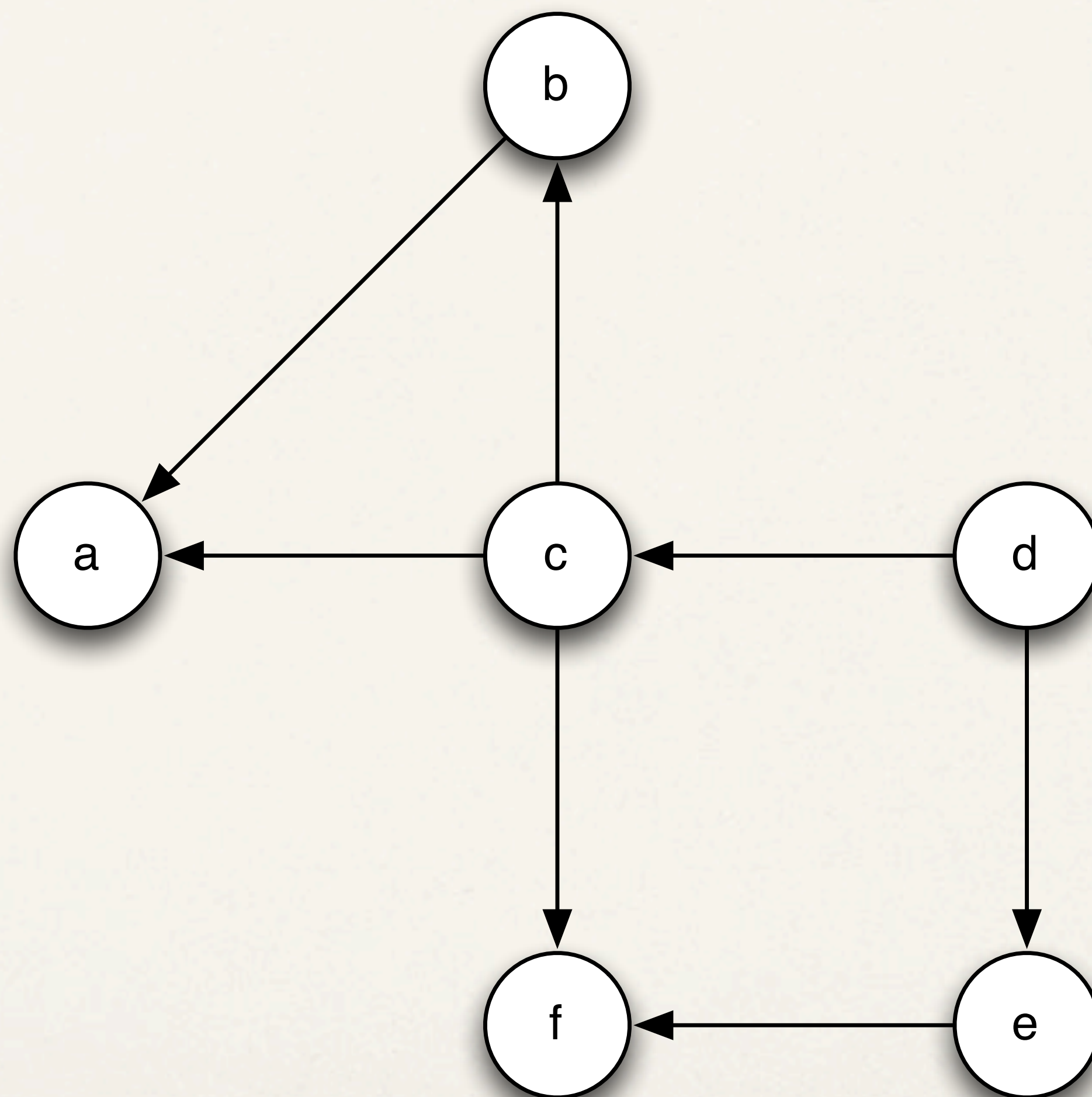
Adjacency Matrix

V	1	2	3	4	5	6	7
1	0	0	0	0	0	1	1
2	0	0	0	0	1	1	0
3	1	1	0	1	0	0	0
4	0	0	0	0	1	0	1
5	1	0	0	0	0	0	0
6	0	0	0	1	0	0	0
7	0	0	0	0	0	0	0

Practice



Practice

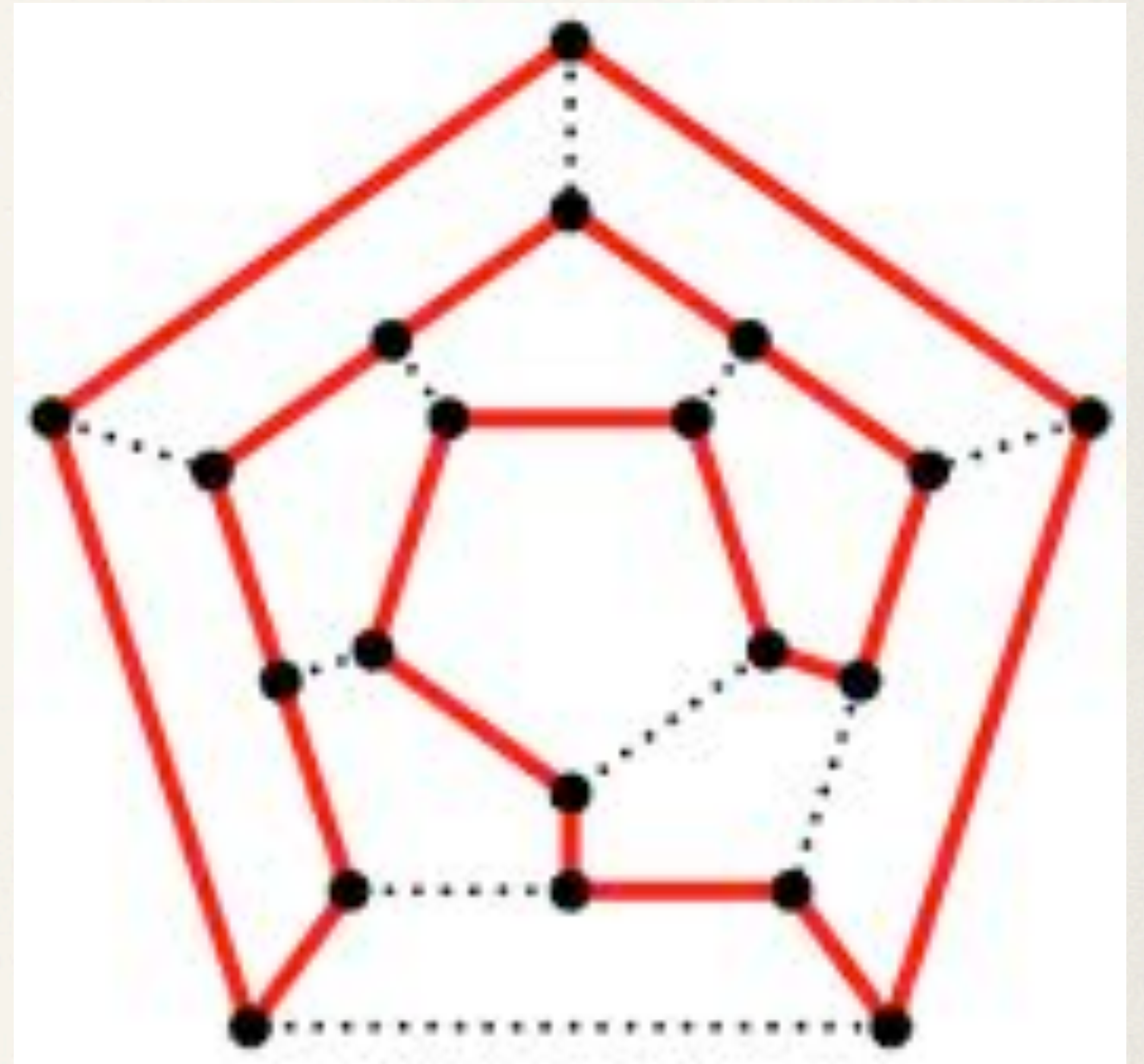


Performance of List vs Matrix

	Space	getDegree (u)	isAdjacentTo (u, v)	getAdjacent (u)
Adjacency List	$O(V+E)$	$O(D(u))$	$O(D(u))$	$O(D(u))$
Adjacency Matrix	$O(V^2)$	$O(V)$	$O(1)$	$O(V)$

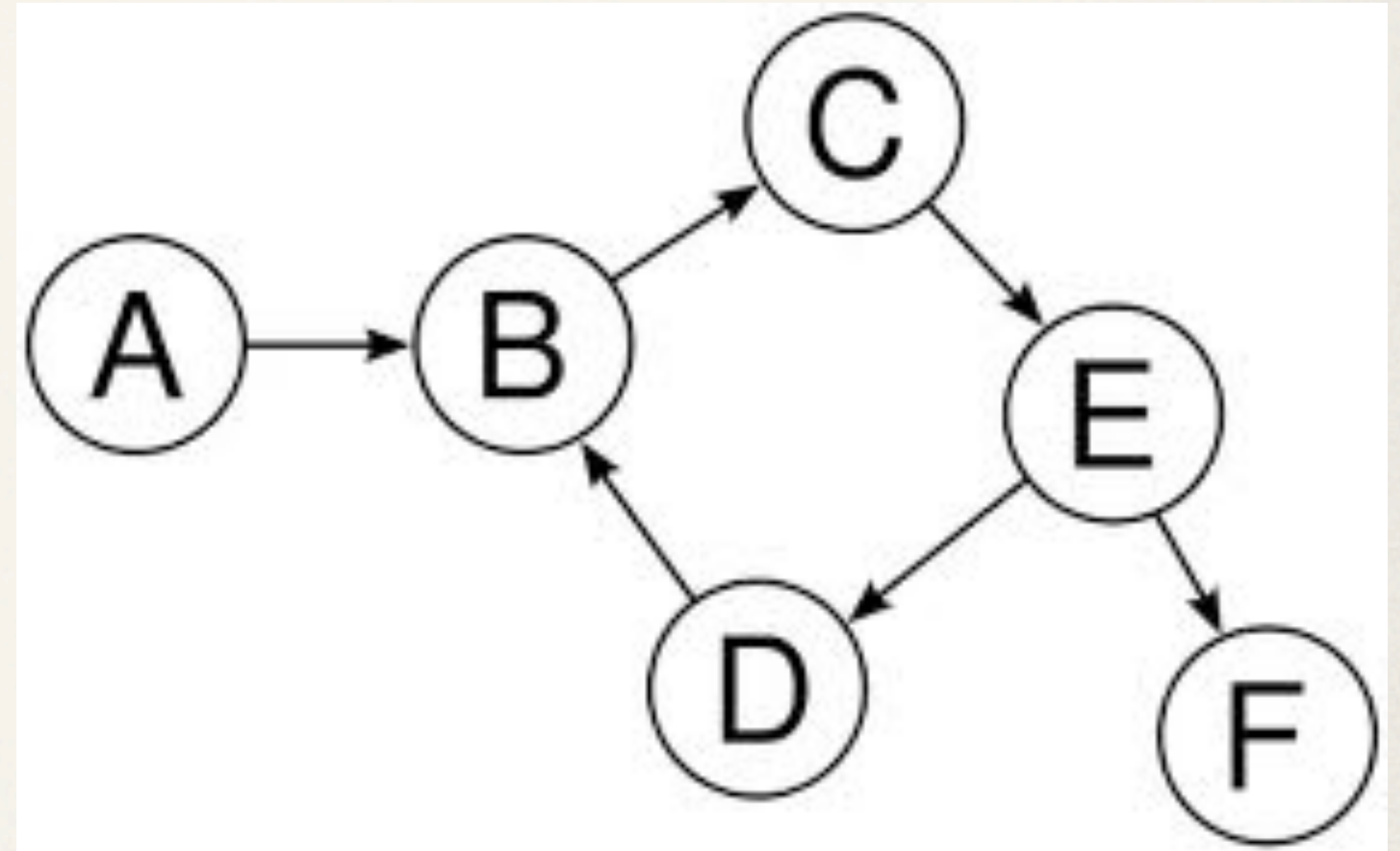
Paths

- ❖ A **path** (sometimes a **walk**) is a series of edges which starts at a vertex, travels from vertex to vertex along edges and ends at some vertex
 - ❖ Formally a path p is a set of edges s.t.
 $p.start = e_0.start$
 $p.end = e_n.end$
 $e_n.end = e_{n+1}.start$
- ❖ If $p.start = p.end$, then p is a **circuit (cycle)**
- ❖ If no edge is repeated, the path is **simple**

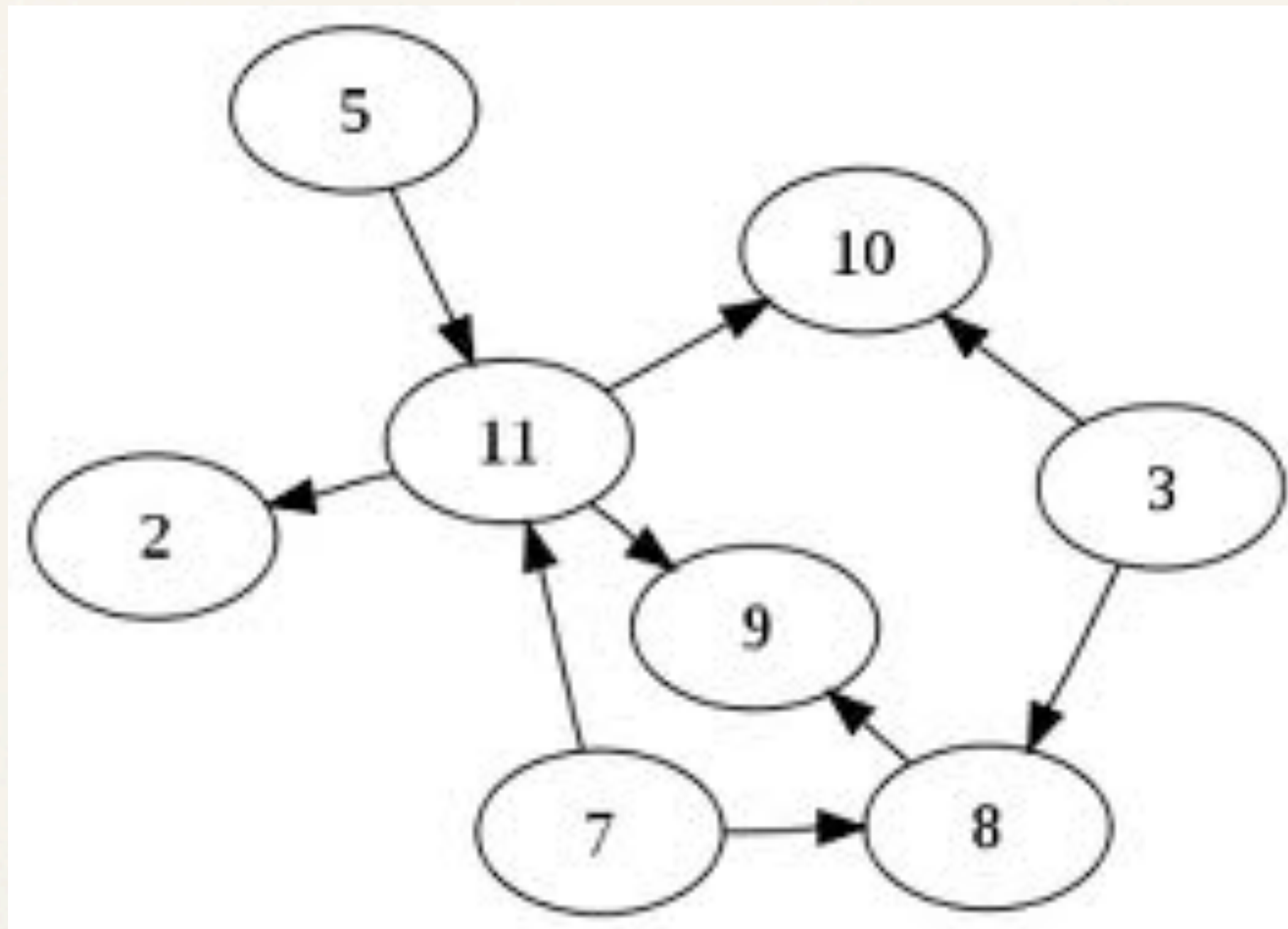


Cyclic vs Acyclic

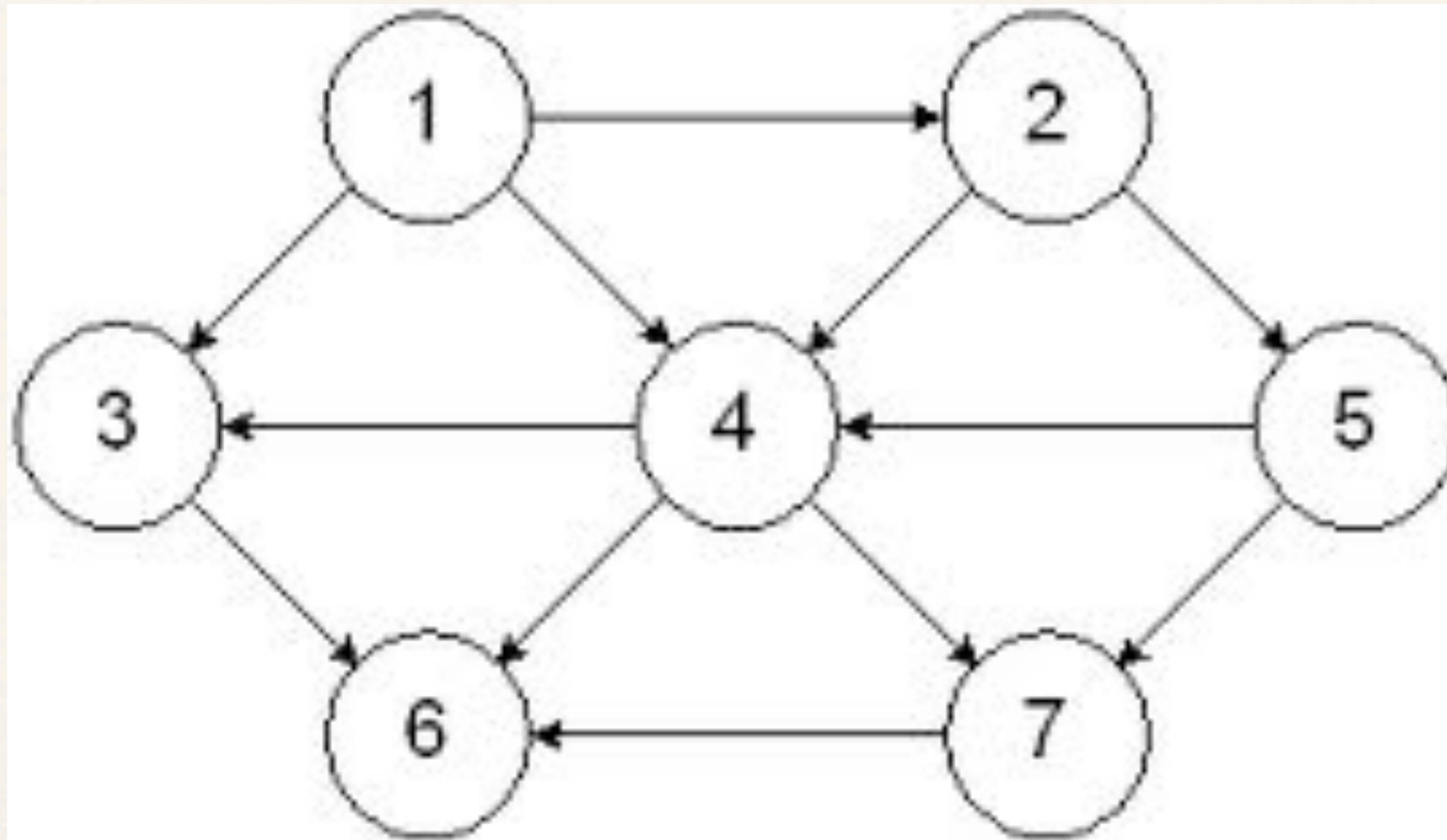
- ❖ Directed graphs can be **cyclic** or **acyclic**
 - ❖ Cyclic: contains a cycle
 - ❖ Acyclic: does not contain any cycles



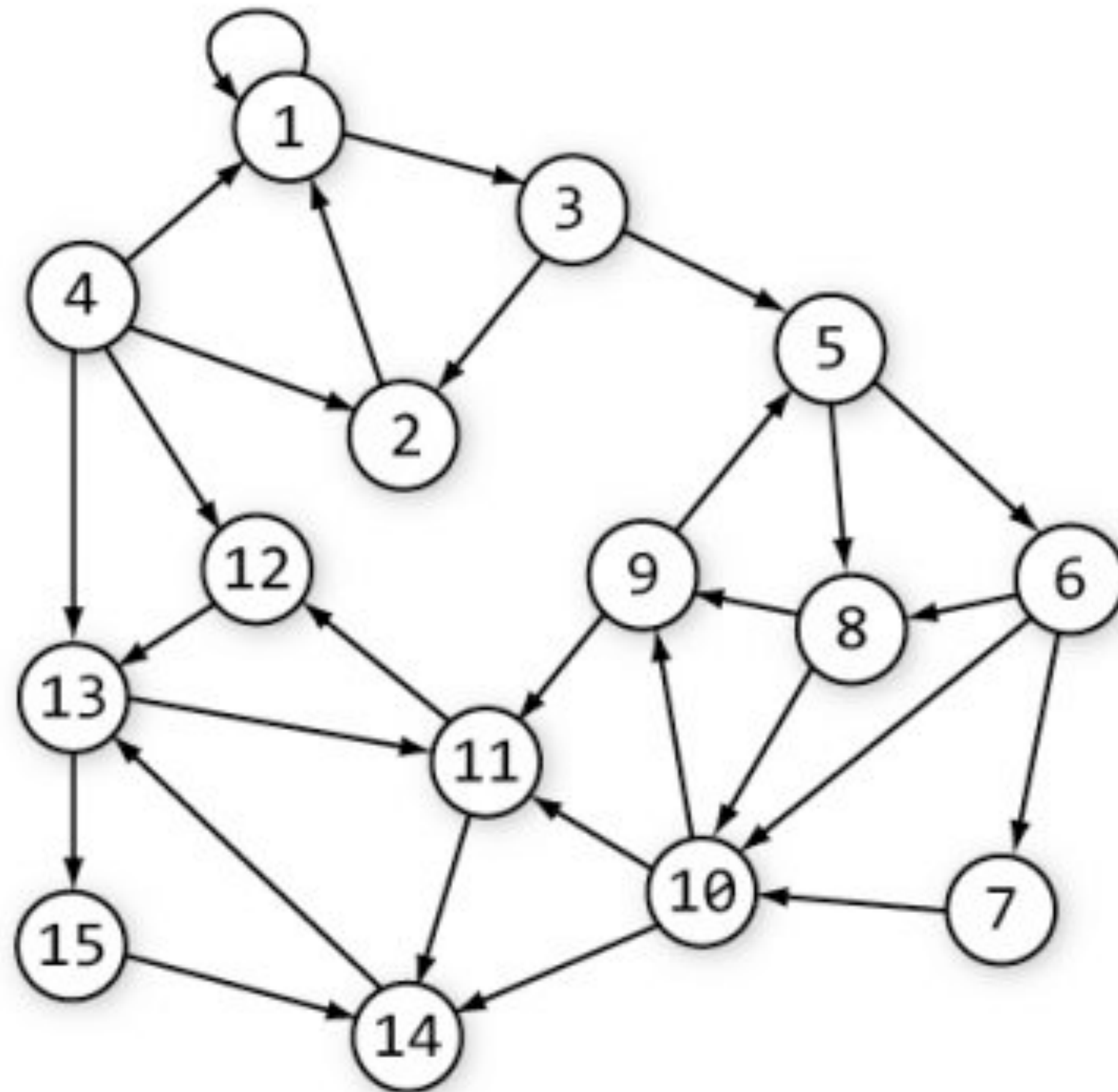
Cycle Practice



Cycle Practice



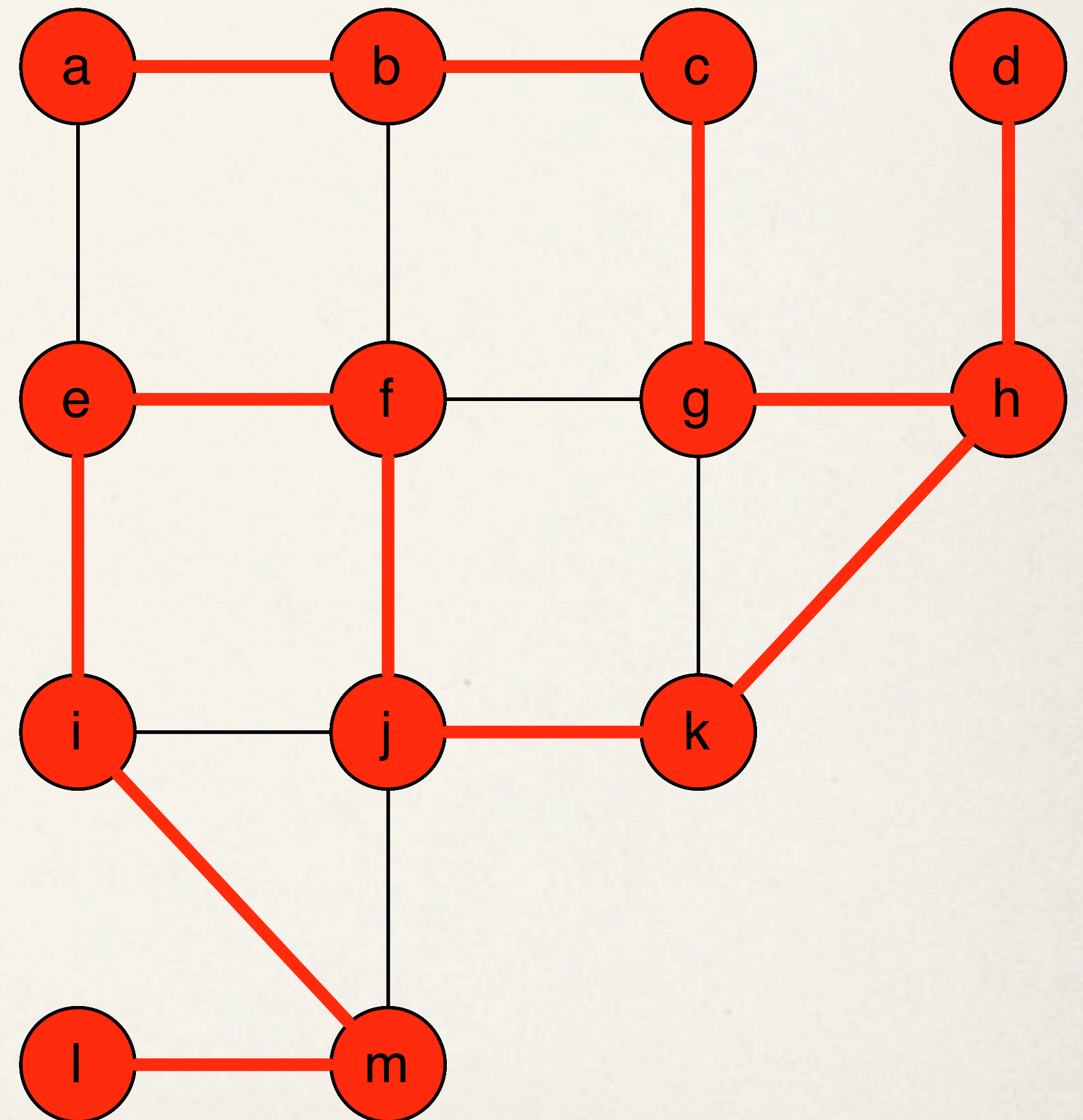
How many cycles are there?



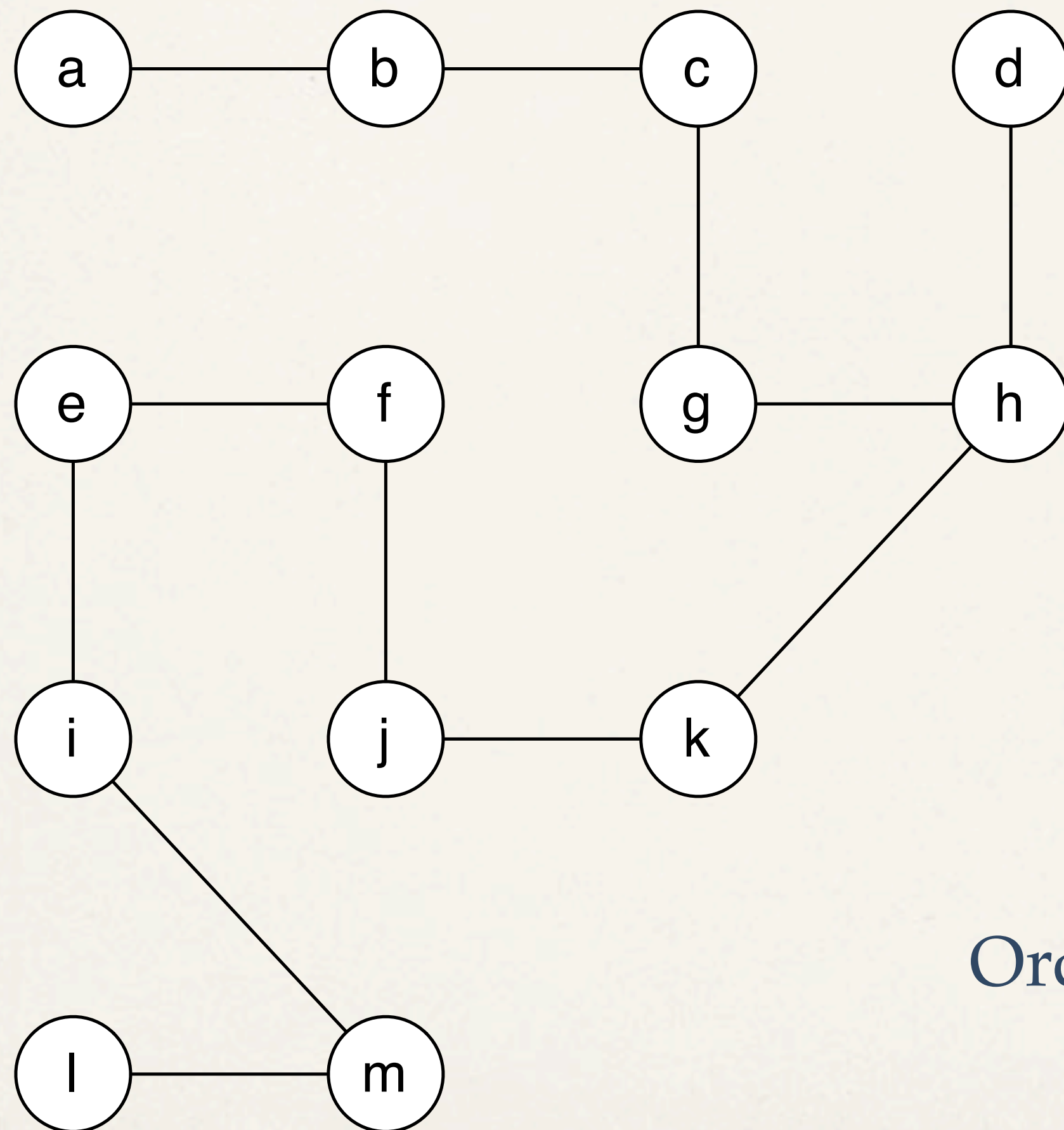
Graph Search

Depth-First Search (DFS)

- ❖ **Visit** some start vertex
- ❖ **Follow** an edge to a vertex which hasn't been explored
 - ❖ **Visit** that vertex
 - ❖ **Follow** an edge from that vertex to another unexplored vertex
 - ❖ If there are no edges to choose from, **backtrack** to the previous vertex



DFS



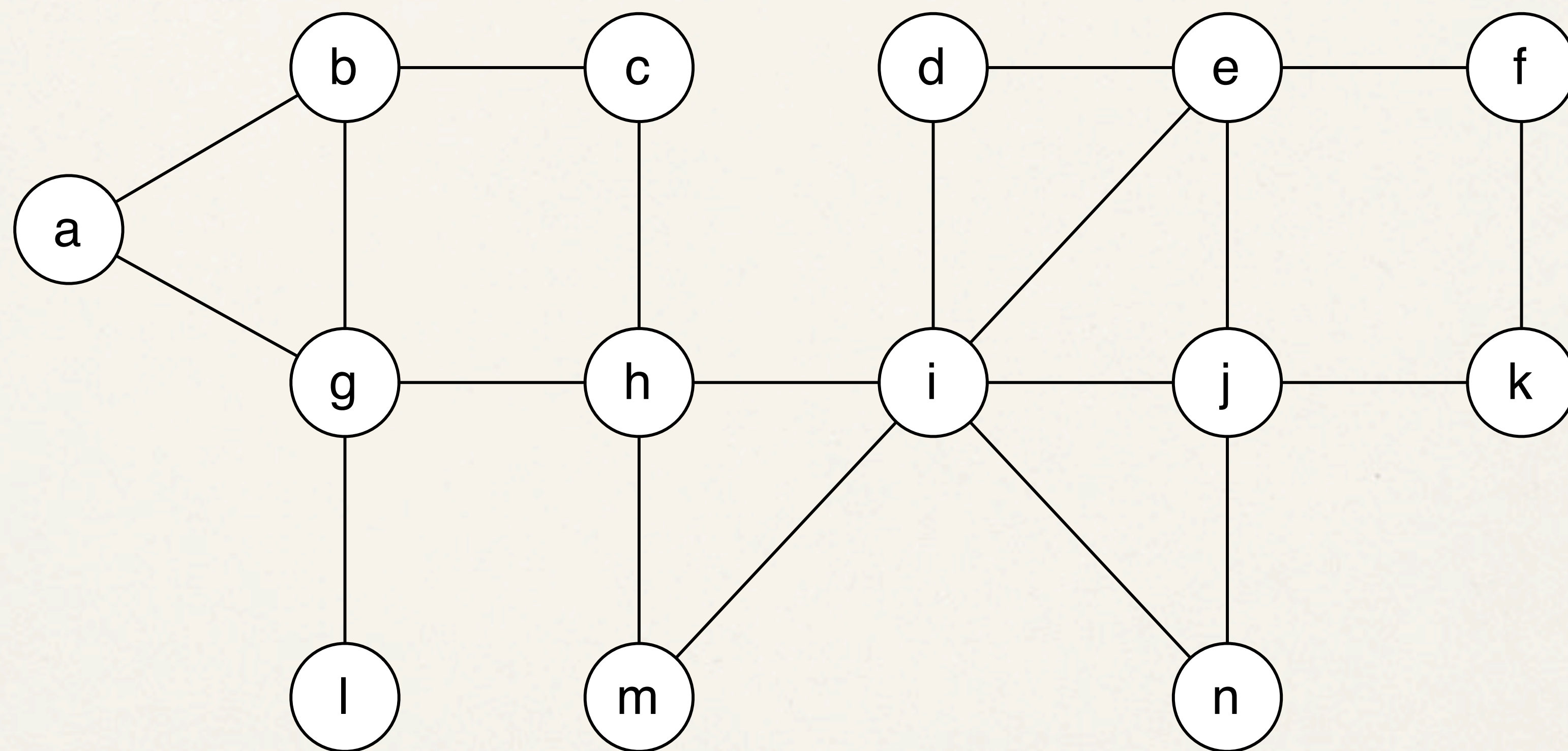
Order visited: a, b, c, g, h, d, k, j, f, e, i, m, l

A Recursive DFS Algorithm

procedure *DFS*(G : connected graph with vertices v_1, v_2, \dots, v_n)
 $T :=$ tree consisting only of the vertex v_1
visit(v_1)

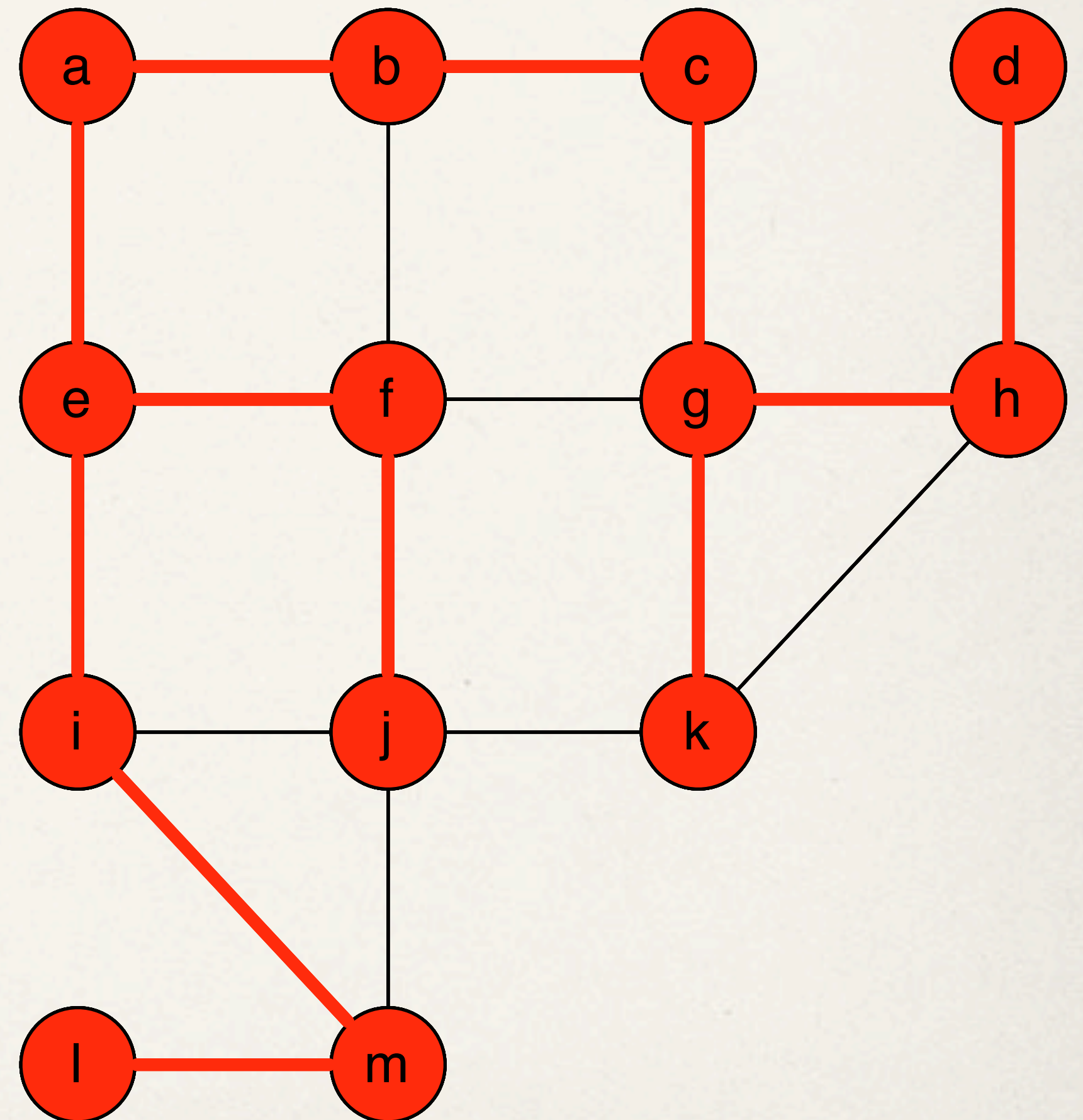
procedure *visit*(v : vertex of G)
for each vertex w adjacent to v and not yet in T
 add vertex w and edge $\{v, w\}$ to T
 visit(w)

Practice

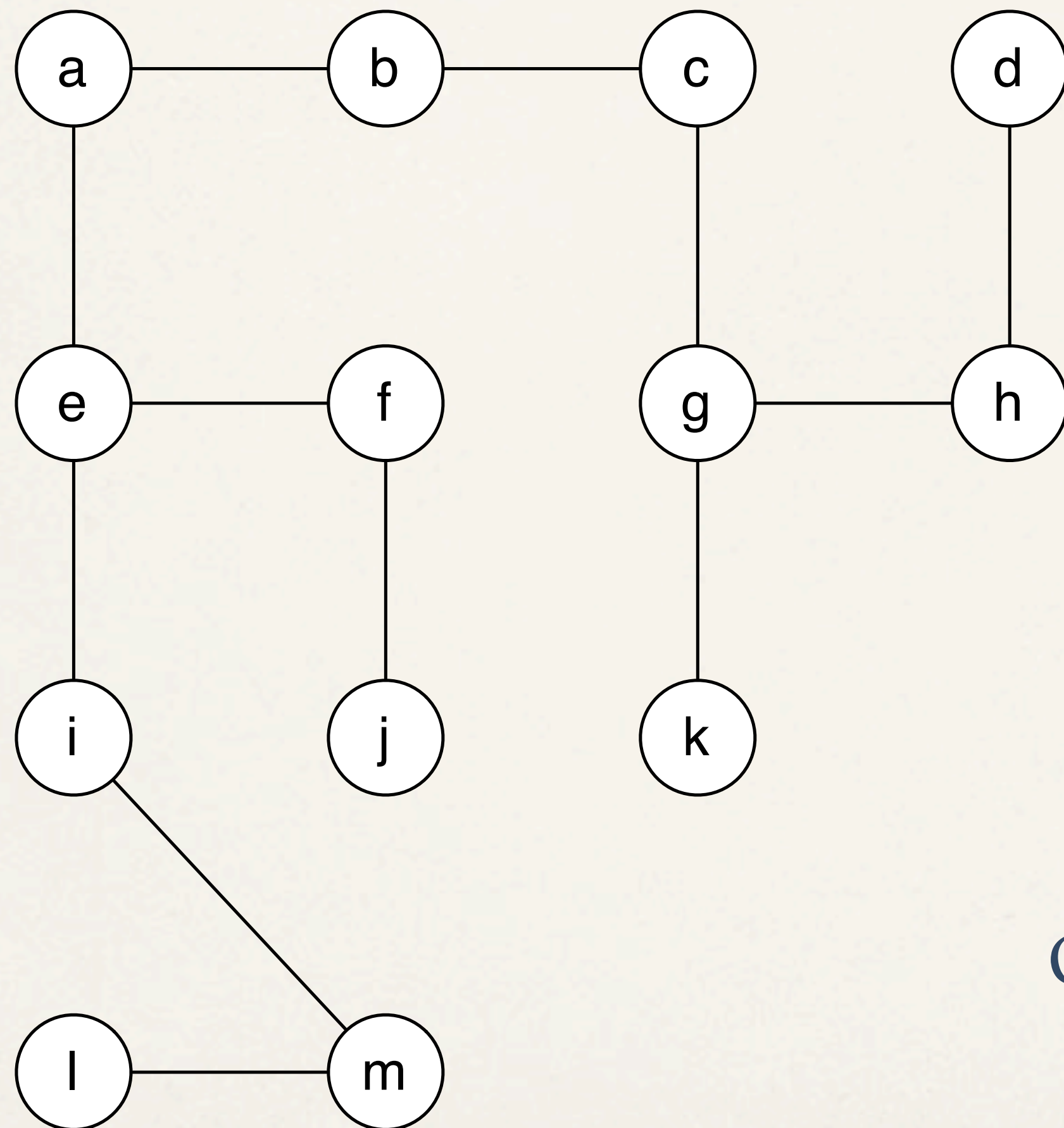


Breadth-First Search

- ❖ **Visit** some start vertex
- ❖ **Visit** all neighbors of the start vertex
- ❖ **Visit** all those neighbors' neighbors
- ❖ Repeat until all vertices visited



BFS



Order visited: a, b, e, c, f, i, g, j, m, h, k, l, d

A (non-recursive) BFS Algorithm

procedure *BFS* (*G*: connected graph with vertices v_1, v_2, \dots, v_n)

$T :=$ tree consisting only of vertex v_1

$L :=$ empty list

put v_1 in the list L of unprocessed vertices

while L is not empty

 remove the first vertex, v , from L

for each neighbor w of v

if w is not in L and not in T **then**

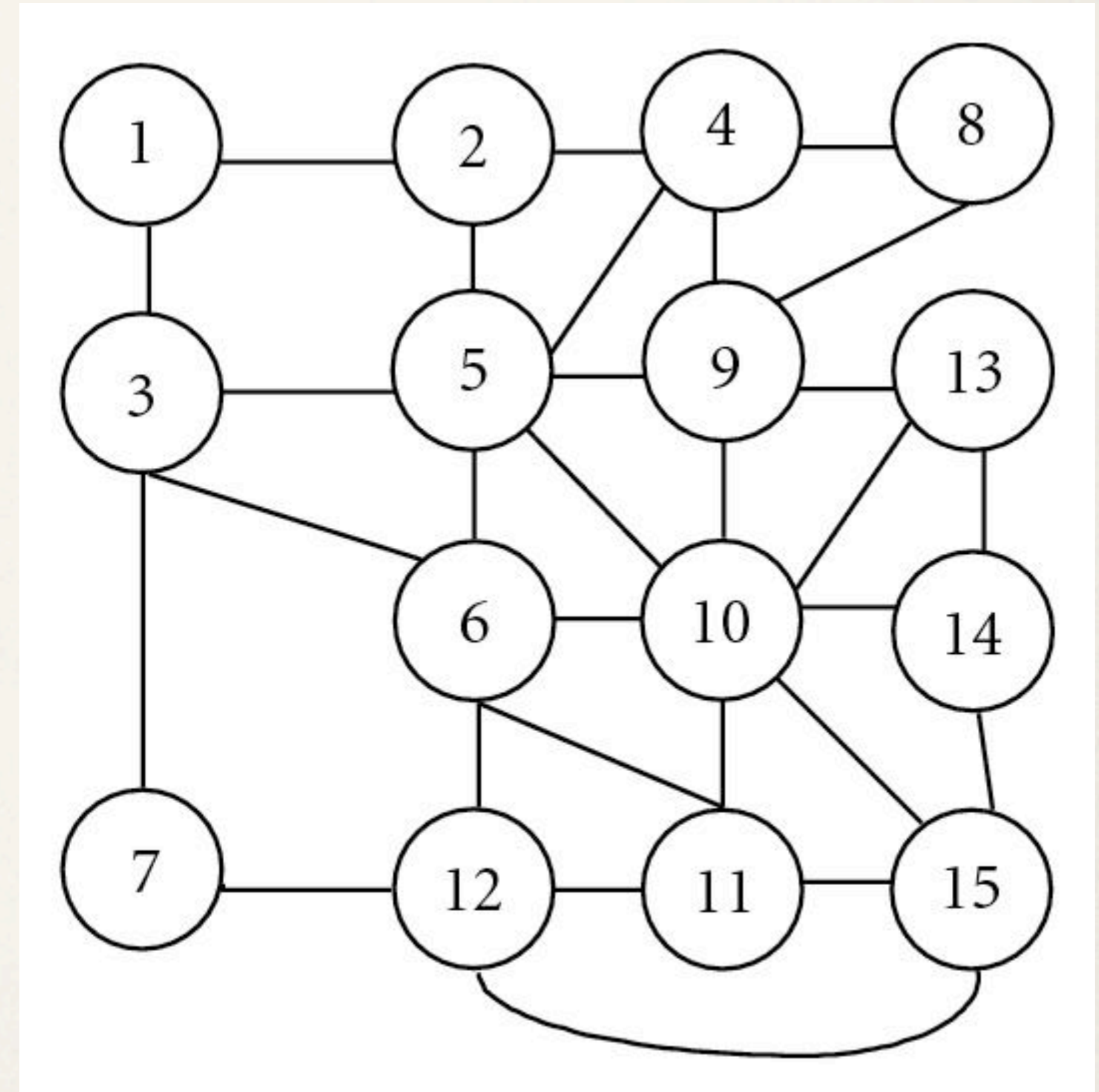
 add w to the end of the list L

 add w and edge $\{v, w\}$ to T

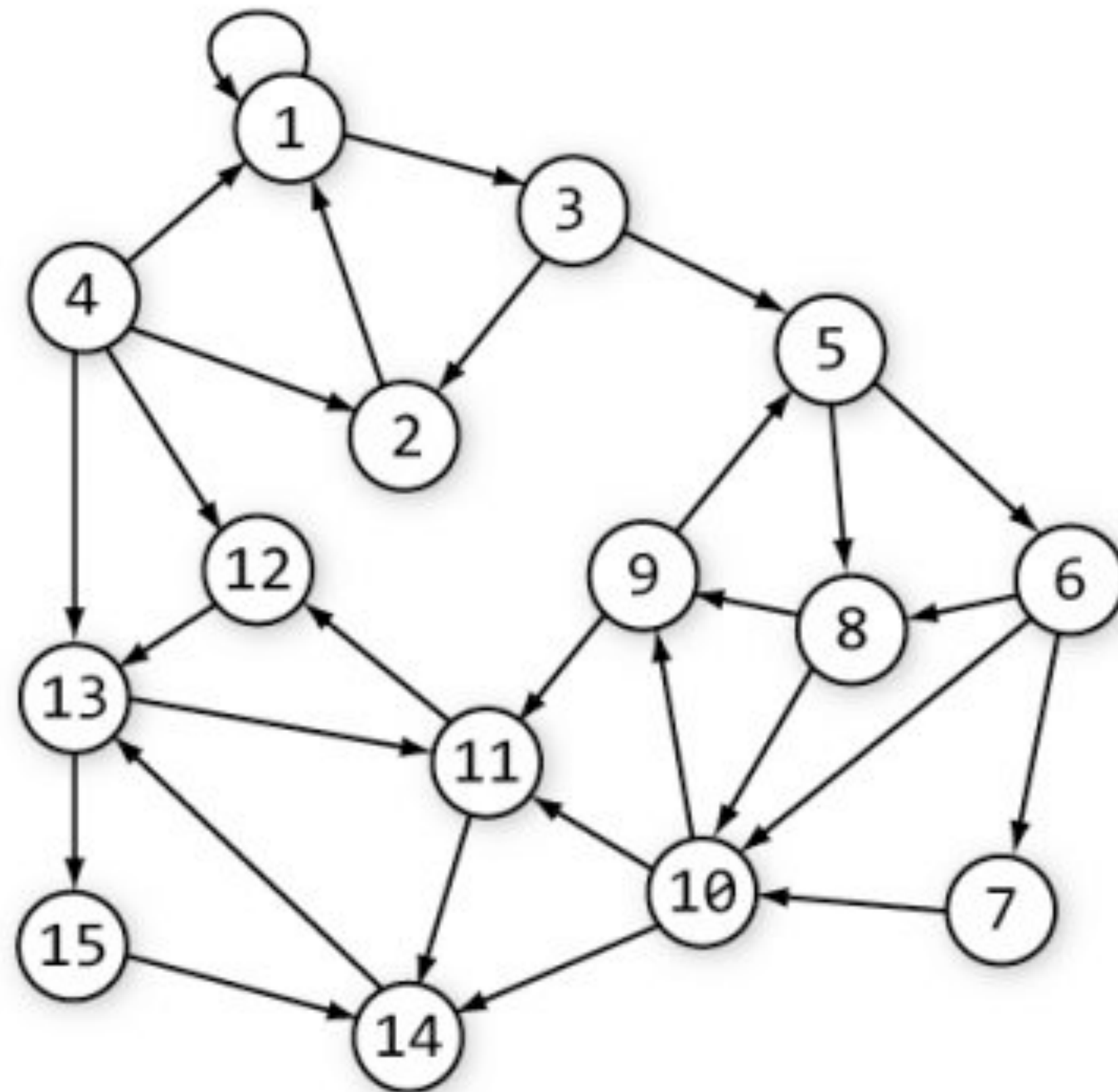
Shortest-Path Problems

Shortest Path

- ❖ The **length** of a path is the number of edges in it
- ❖ Many problems try to find the **shortest path** between two vertices
 - ❖ Flights with the fewest stopovers
 - ❖ Driving directions with few instructions
- ❖ *Can* use DFS or BFS

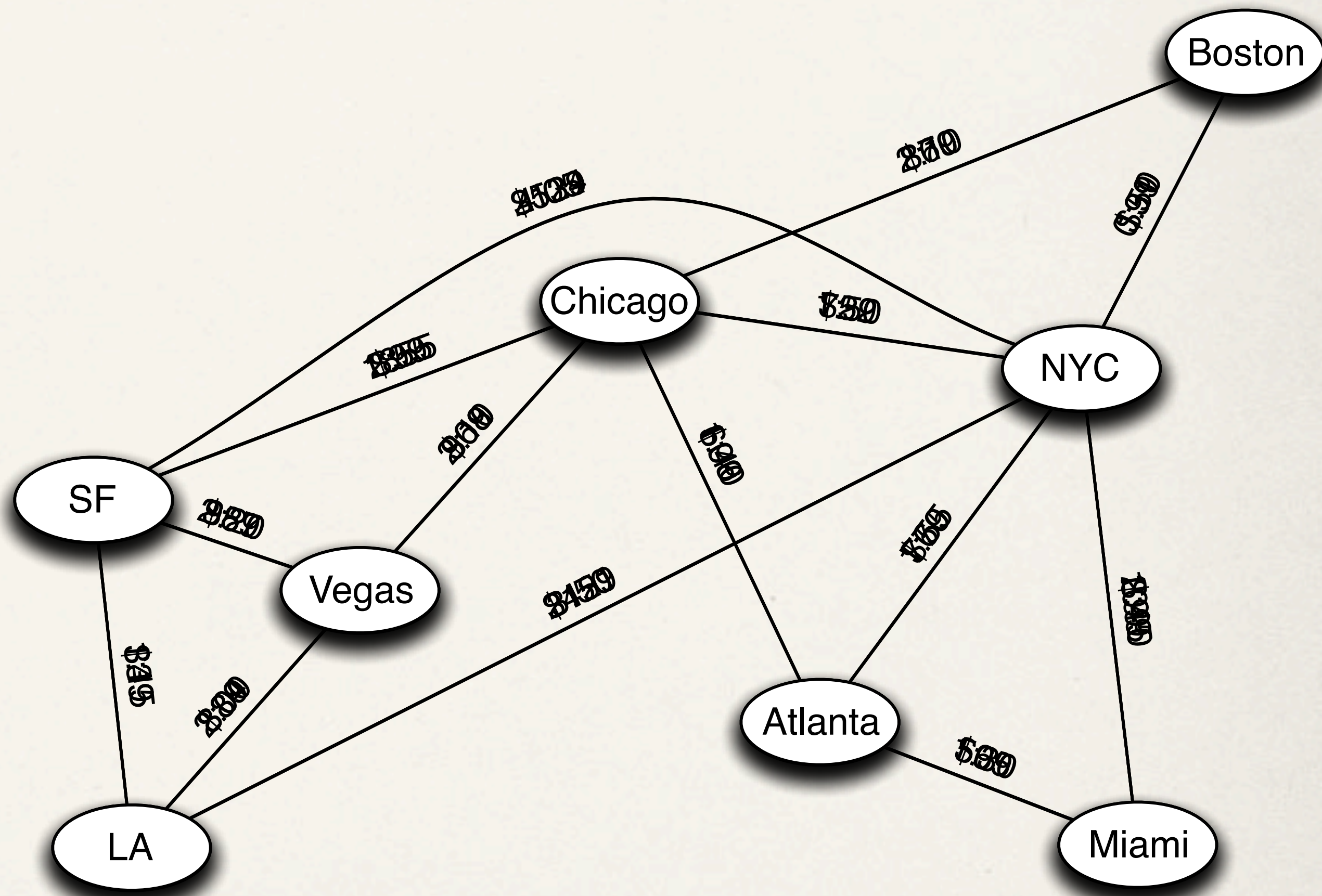


Practice



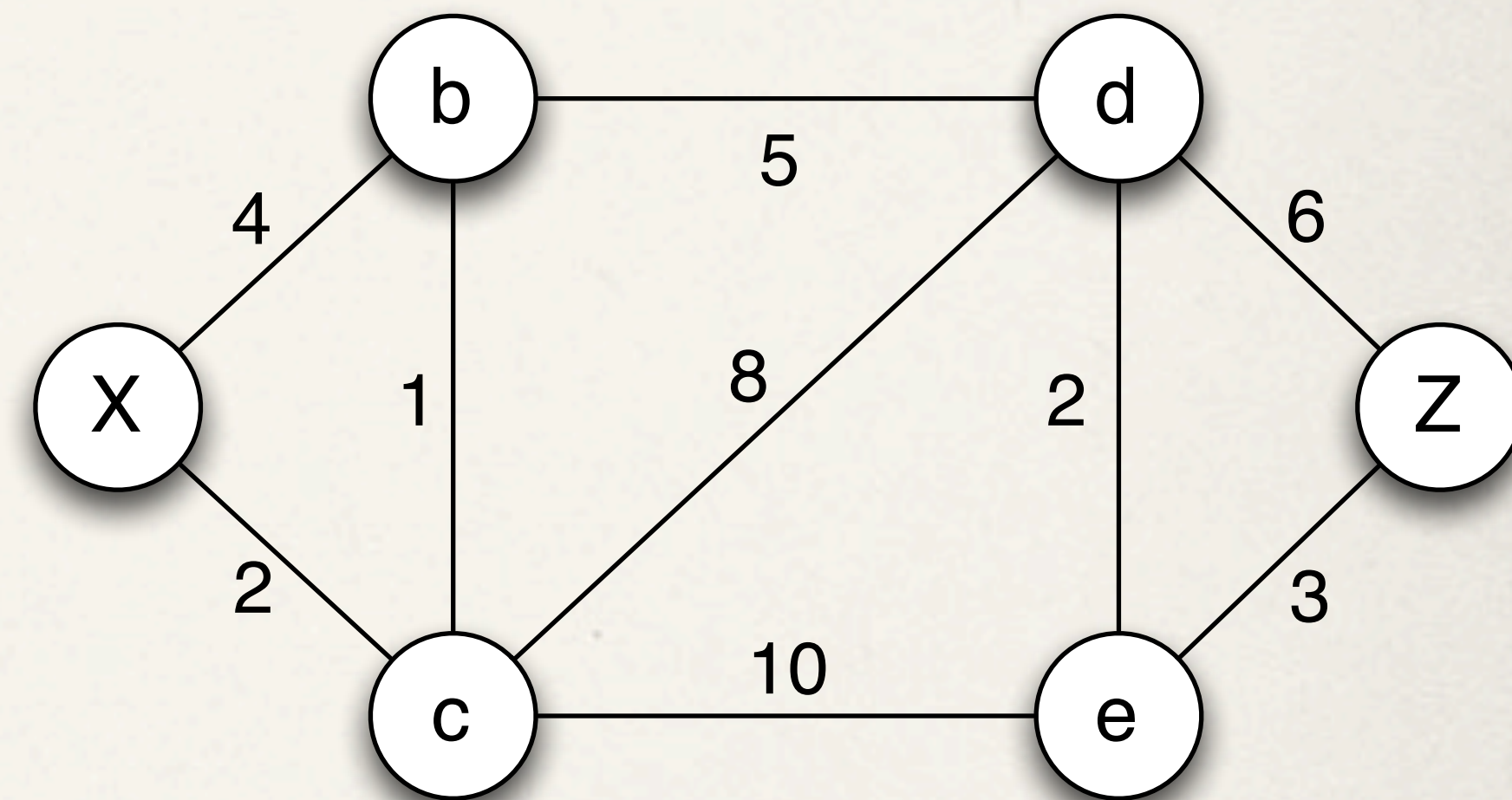
Weighted Graphs

- * Add a **weight** to every edge
 - * weights are normally “costs”
- * Length of a path is the sum of the weights of each edge

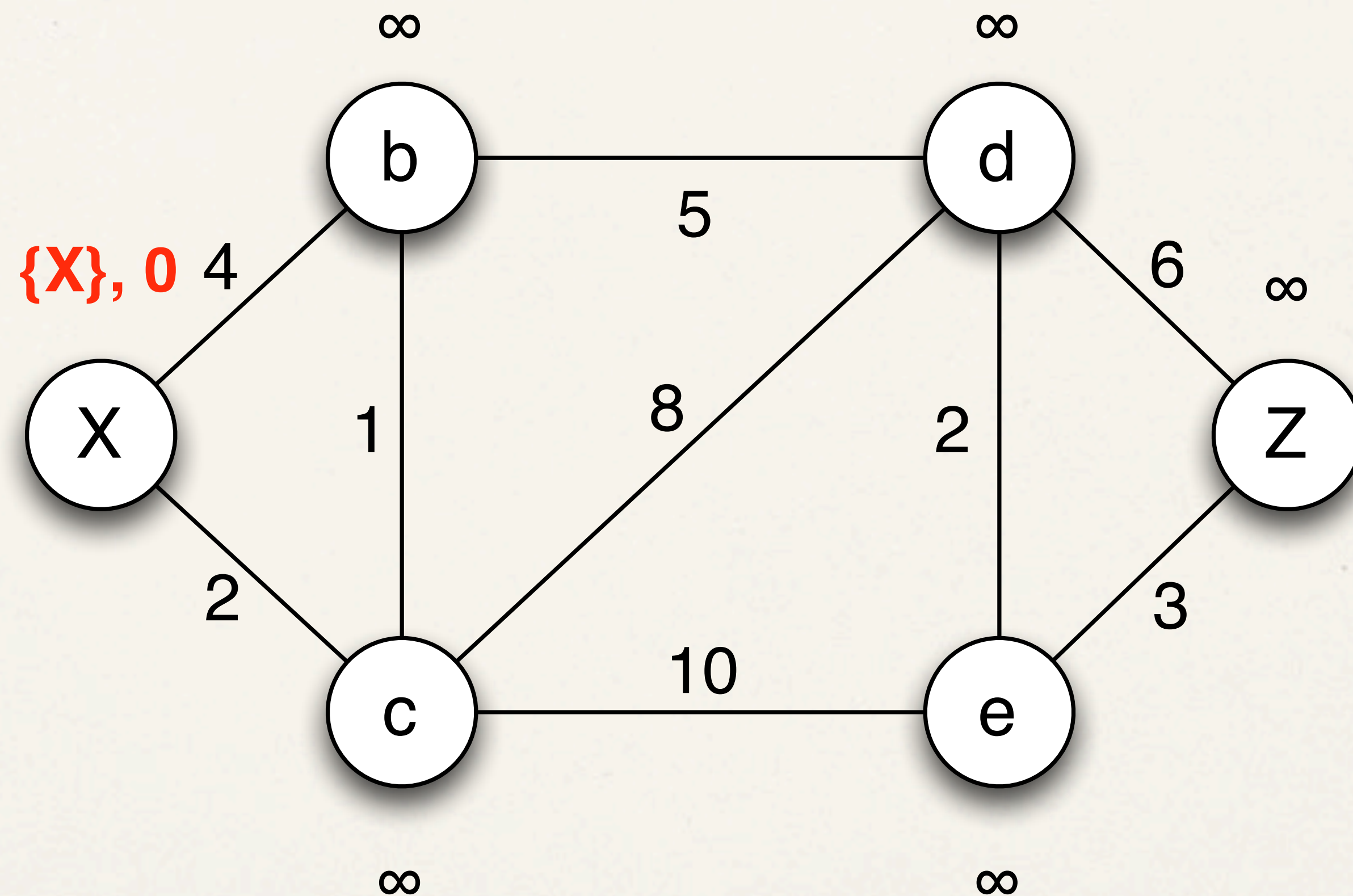


Dijkstra's Algorithm

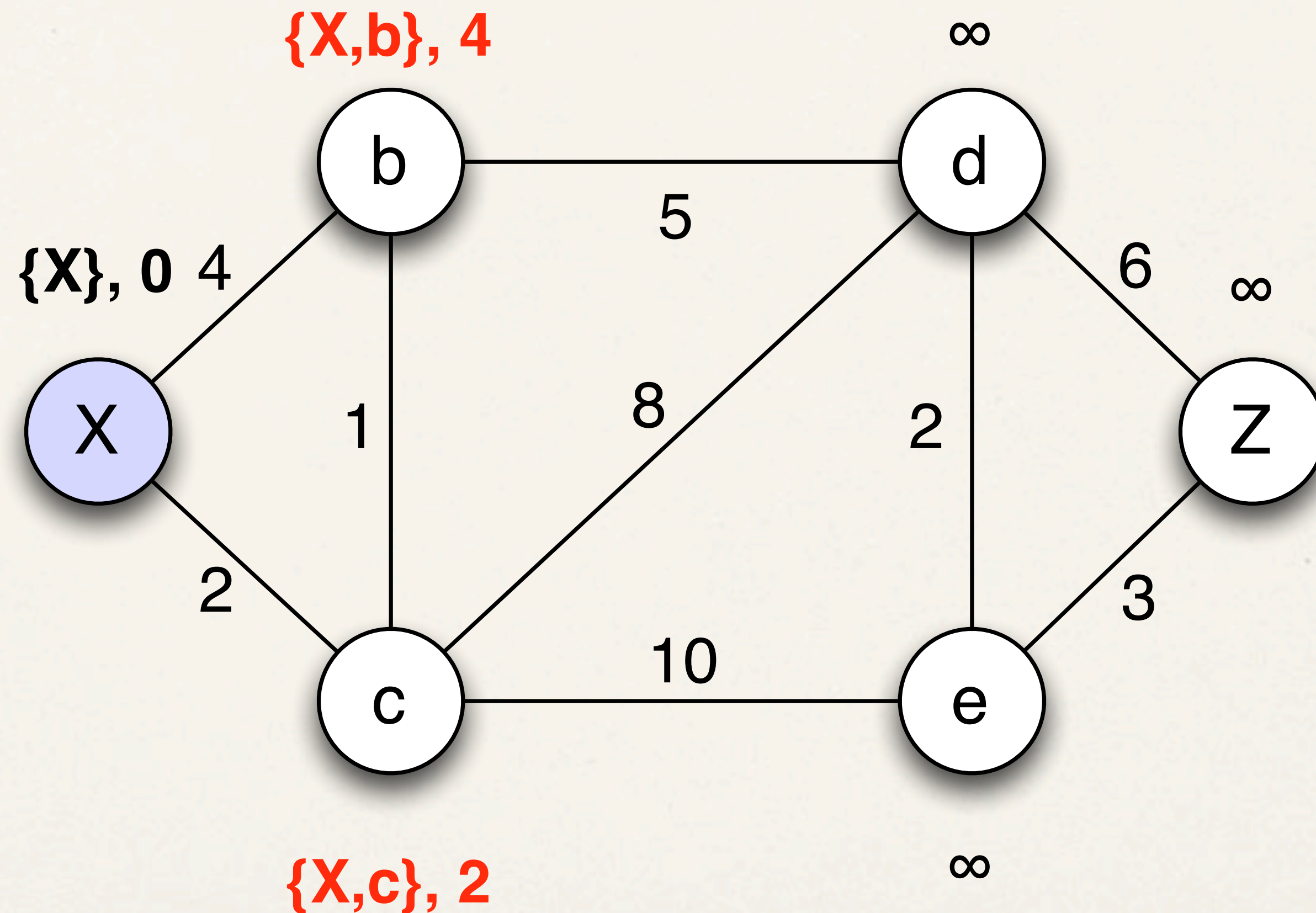
- ❖ Finds shortest path between X and Z in $O(n^2)$ time
- ❖ **Greedy algorithm:**
 - ❖ Start with $C = \{X\}$
 - ❖ Search the neighbors of C to find next closest node to X not in C
 - ❖ Add it to C
 - ❖ Repeat until Z is the next closest node to X



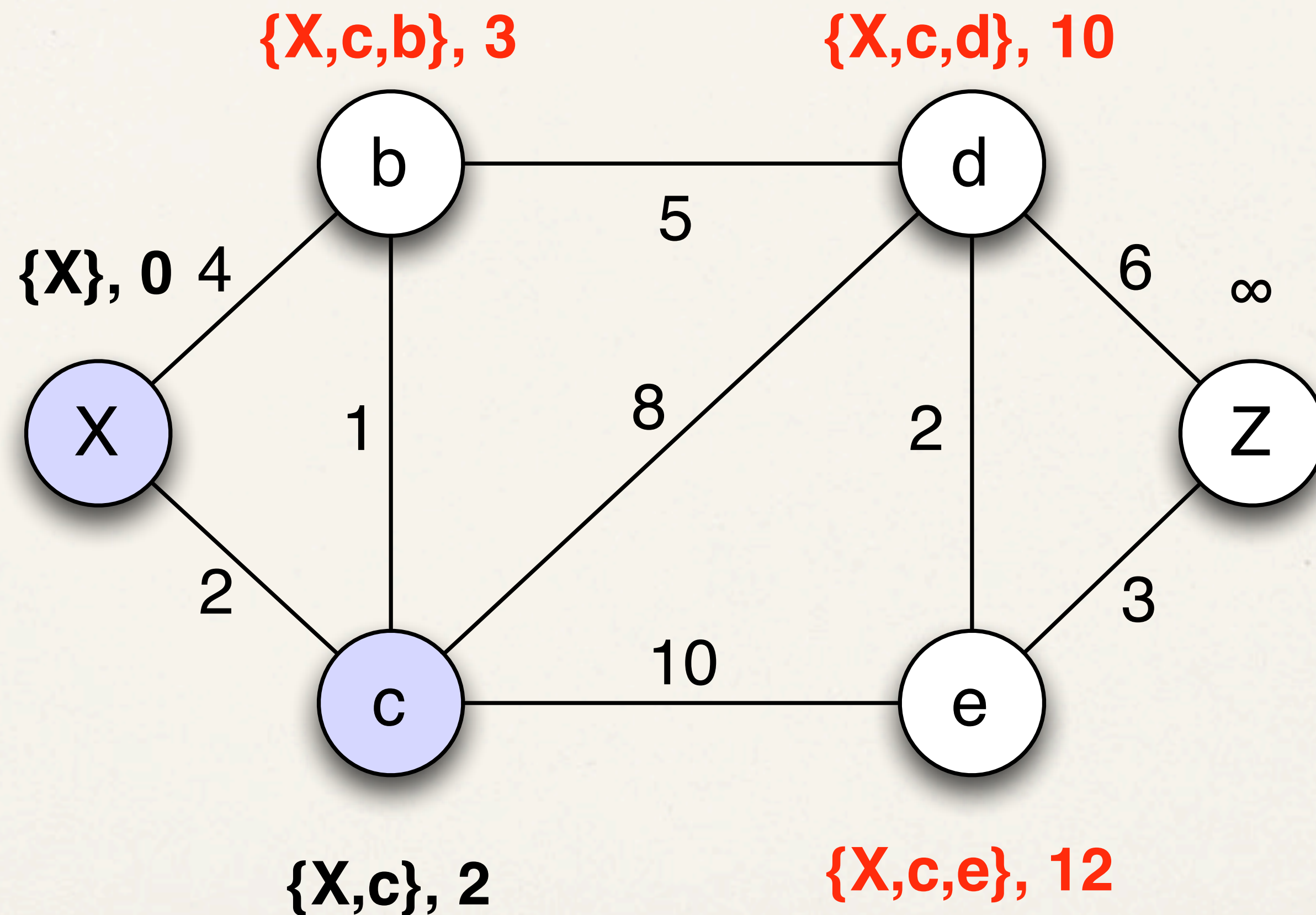
Dijkstra's Algorithm



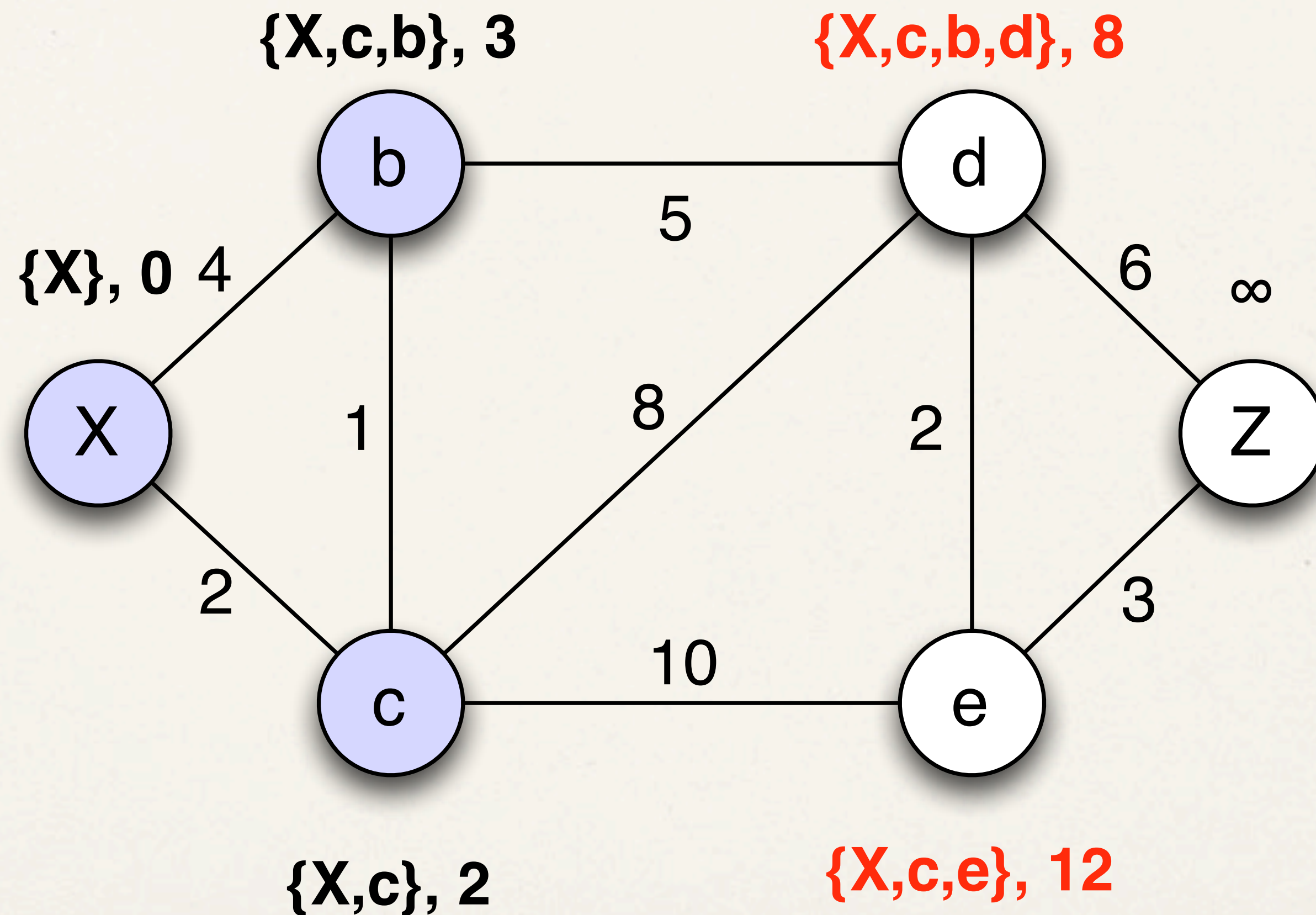
Dijkstra's Algorithm



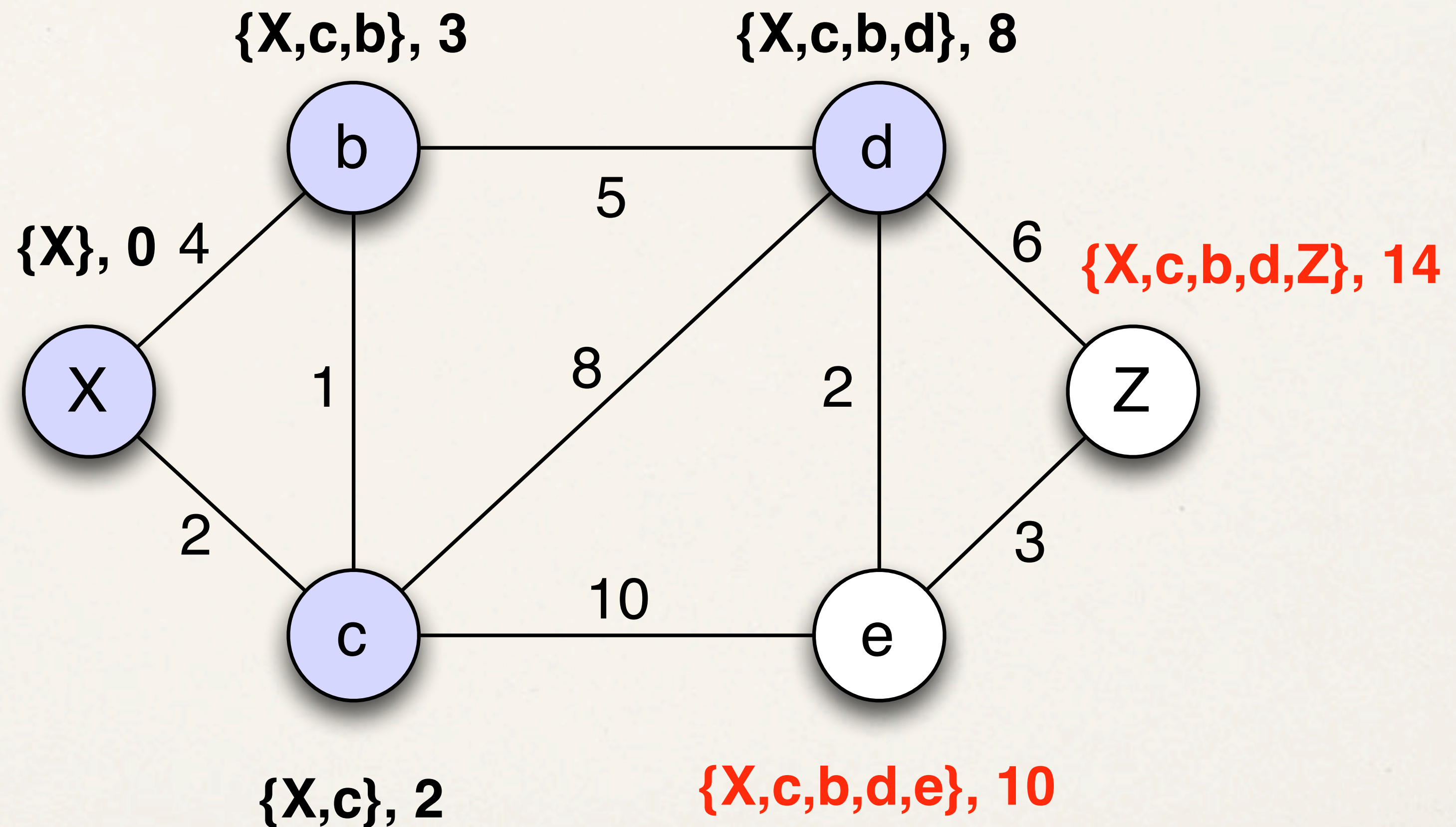
Dijkstra's Algorithm



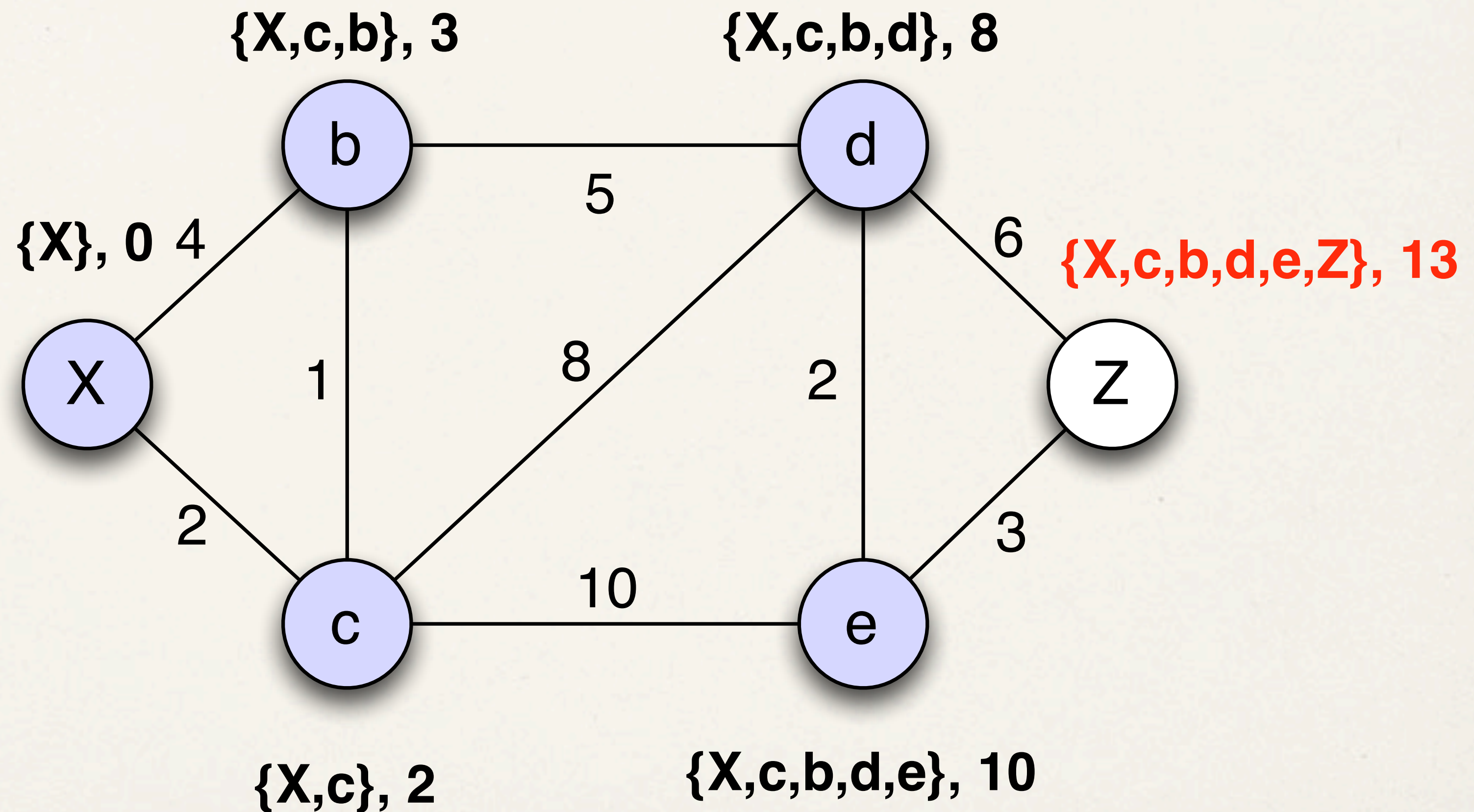
Dijkstra's Algorithm



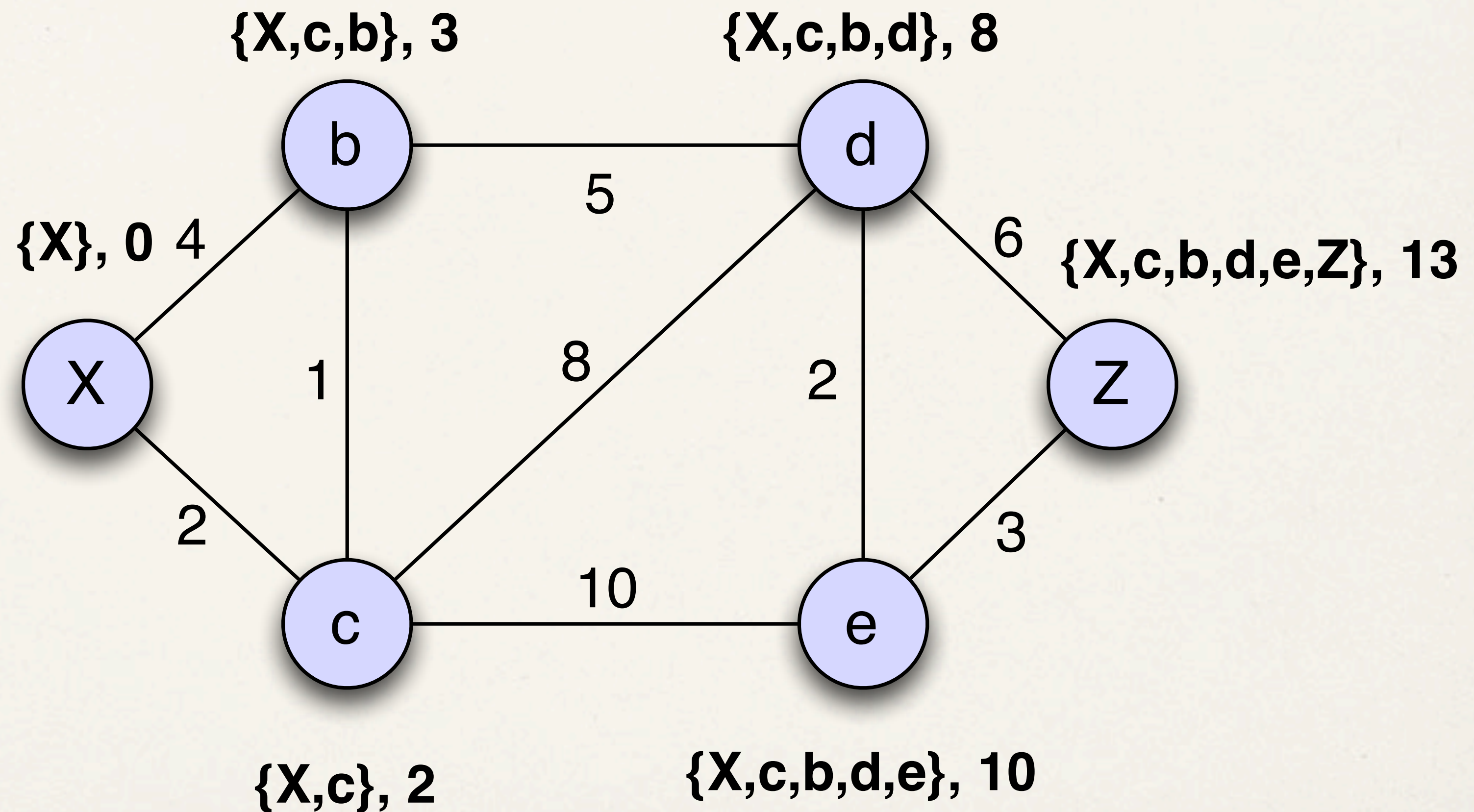
Dijkstra's Algorithm



Dijkstra's Algorithm

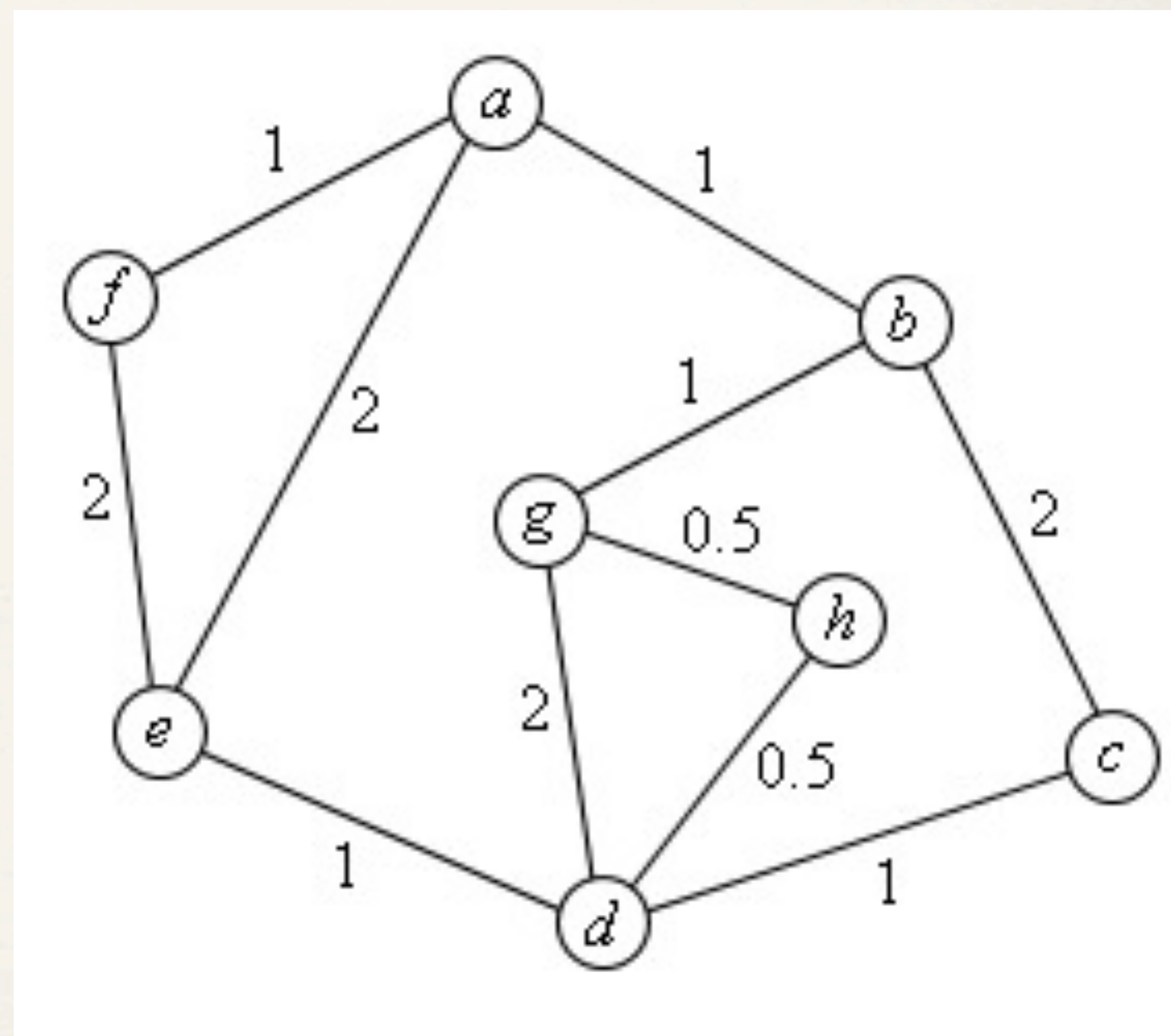


Dijkstra's Algorithm



Practice

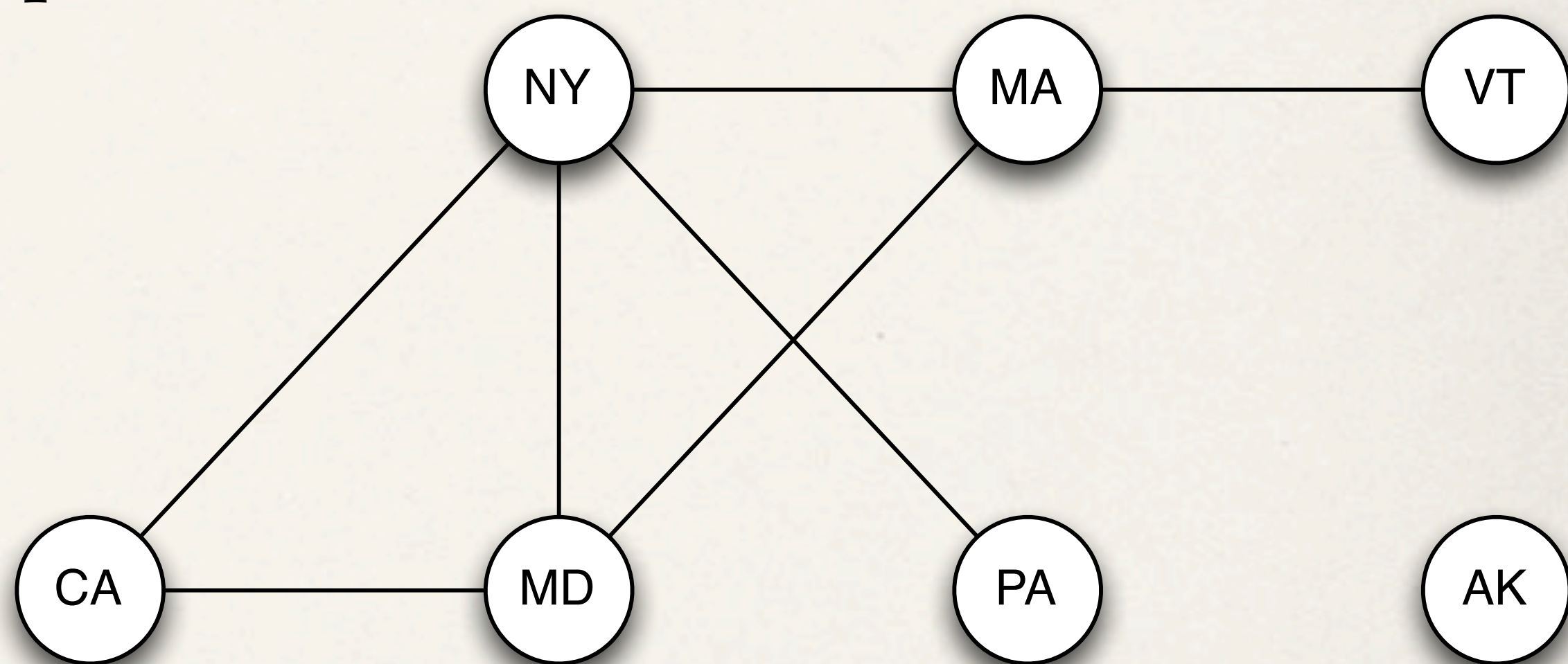
- ❖ Find the shortest path from g to f using Dijkstra's Algorithm



Connectivity

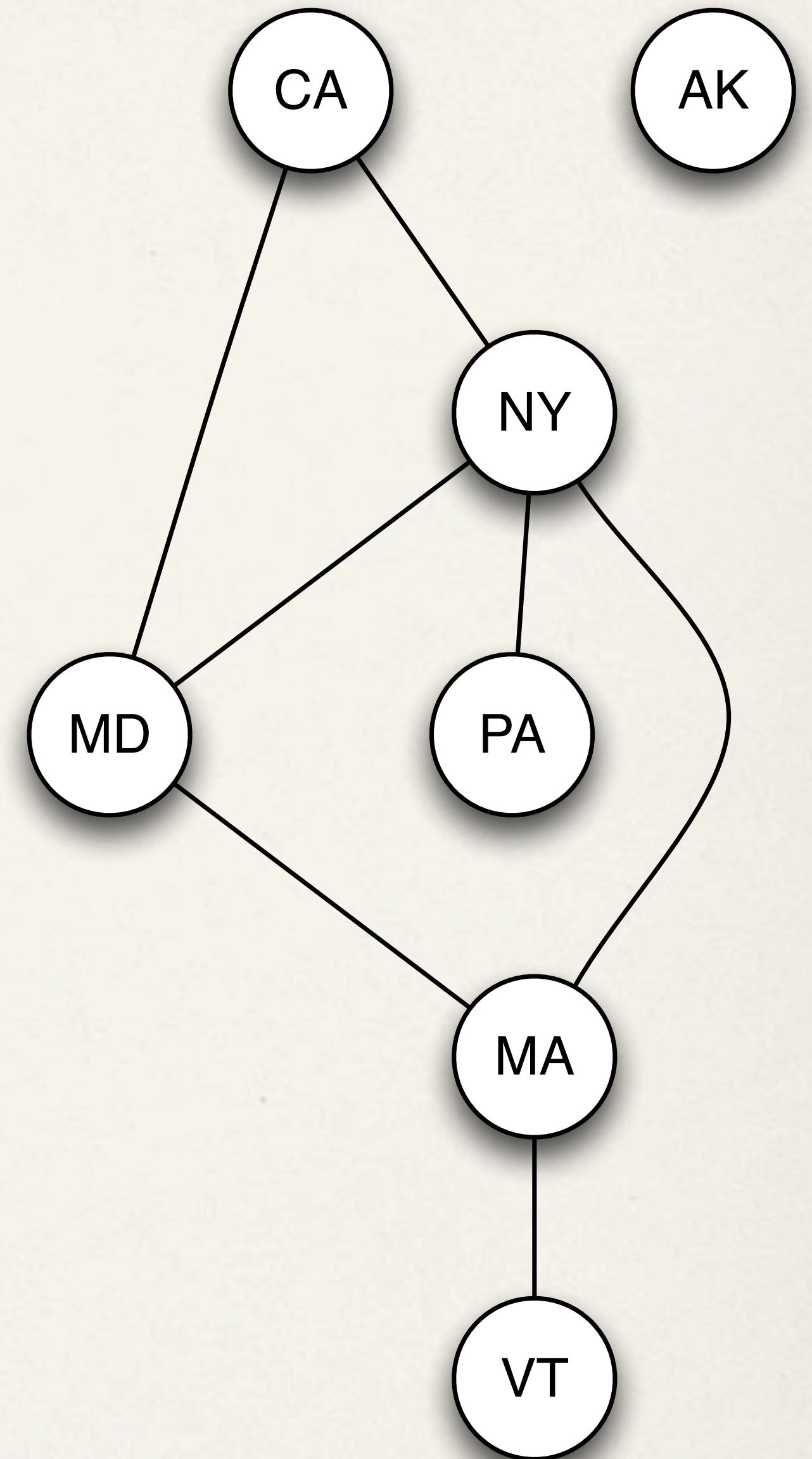
Network Reliability

- ❖ What would happen if my router in NY went offline? If CA got knocked out?
- ❖ I often want there to always be a path available between all the nodes in my graph

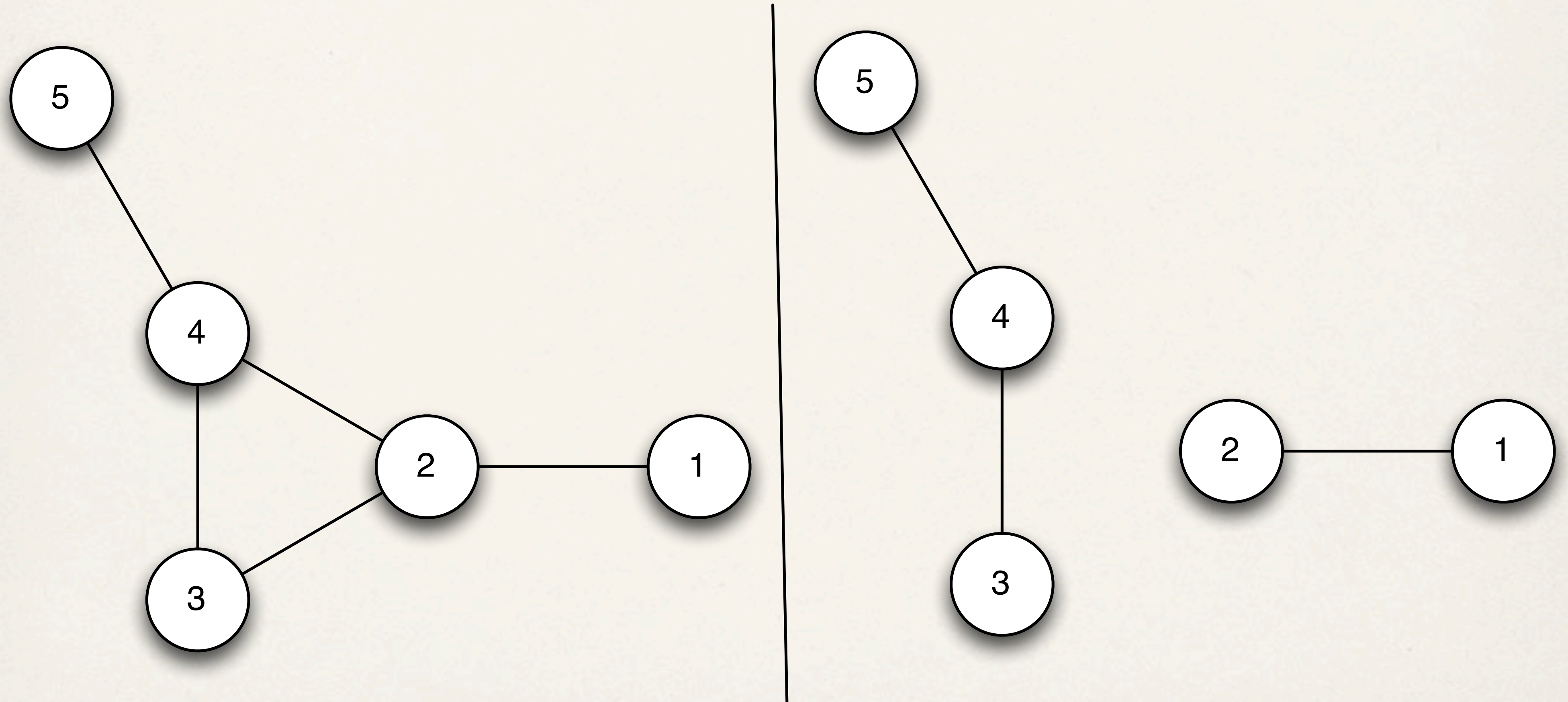


Connectedness

- ❖ Two nodes are **connected** if there exists a path between them
 - ❖ Otherwise they are **disconnected**
- ❖ If every pair of nodes in a graph is connected, then the whole graph is **connected**
 - ❖ Otherwise it is **disconnected**
- ❖ If a graph is connected, then there exists a simple path between every pair of vertices in the graph

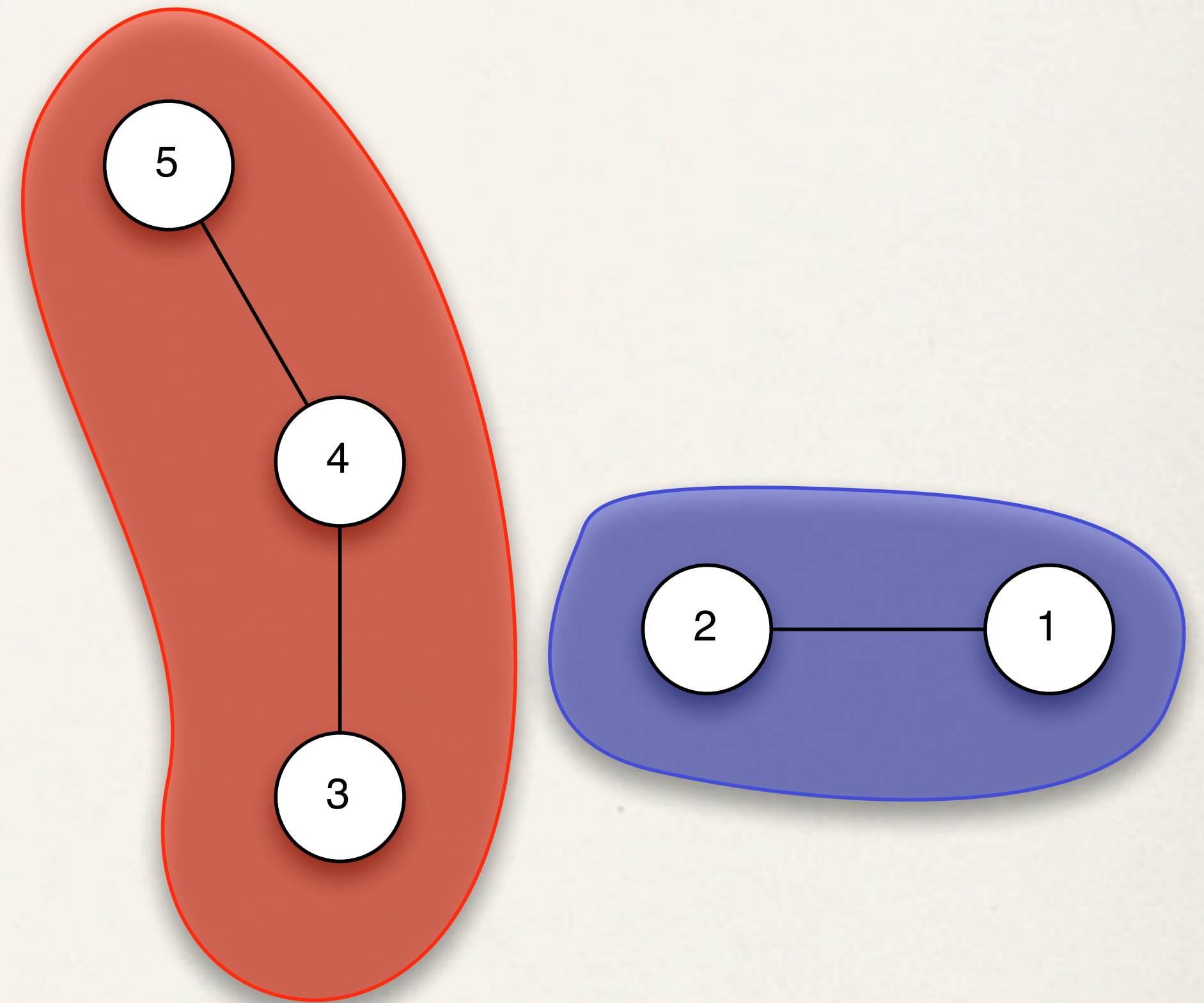


Connected vs Disconnected



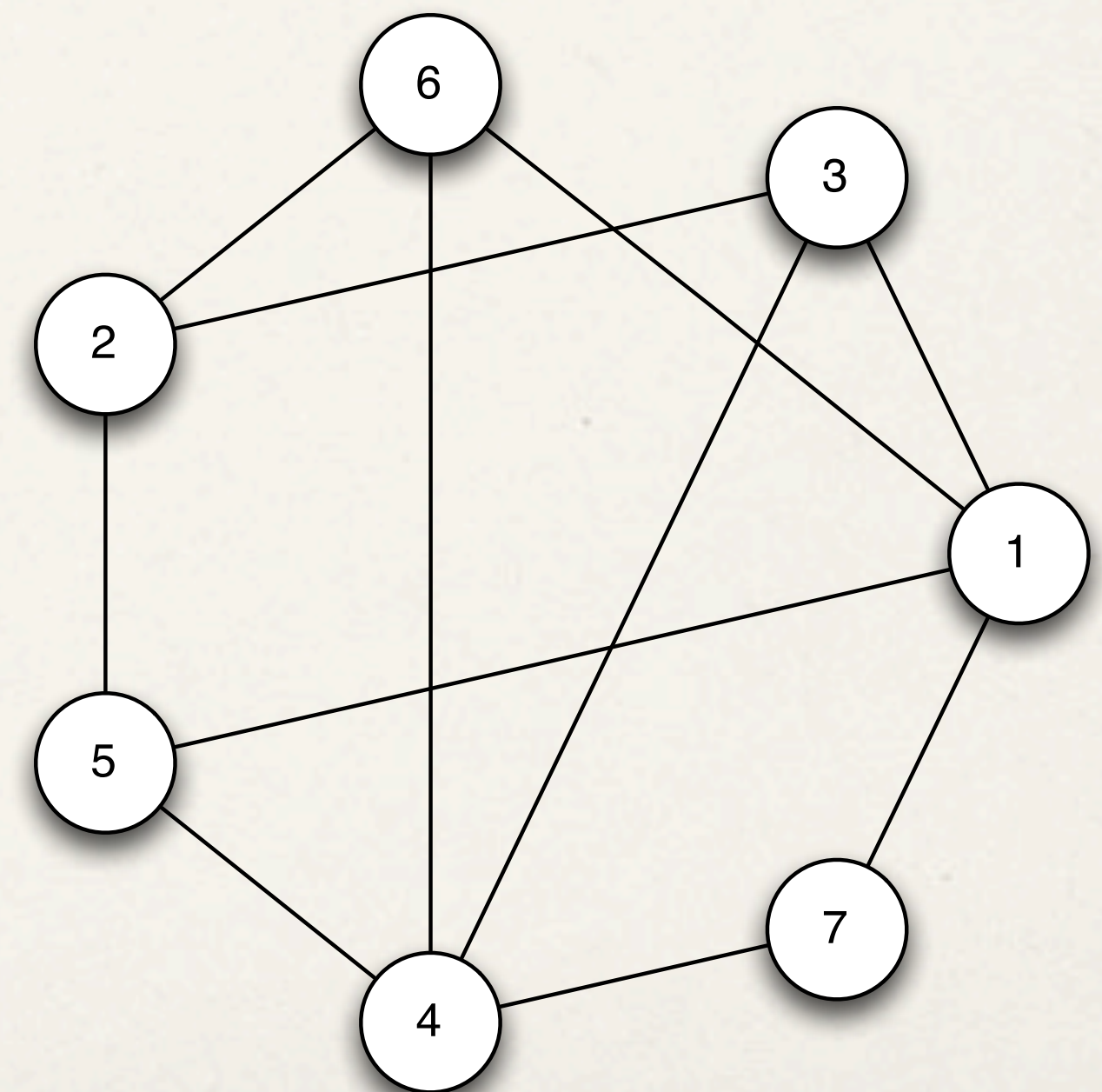
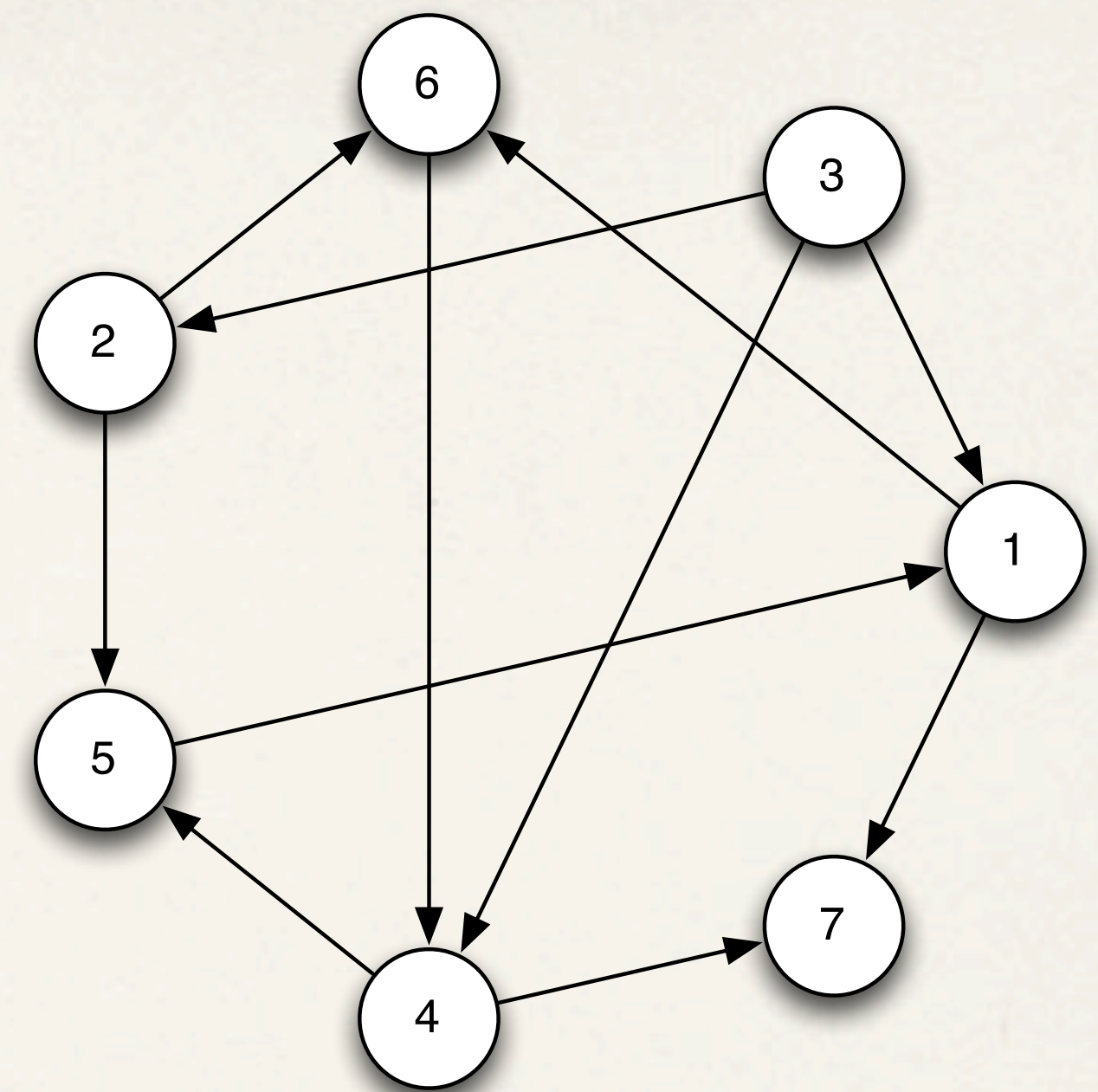
Connected Components

- ❖ A part of a graph which is disconnected from all other parts is called a **connected component**
- ❖ Formally, a connected subgraph of G which is not a proper subgraph of any other connected subgraph of G is a **connected component**
- ❖ A connected graph has 1 connected component
- ❖ A disconnected graph has 2+



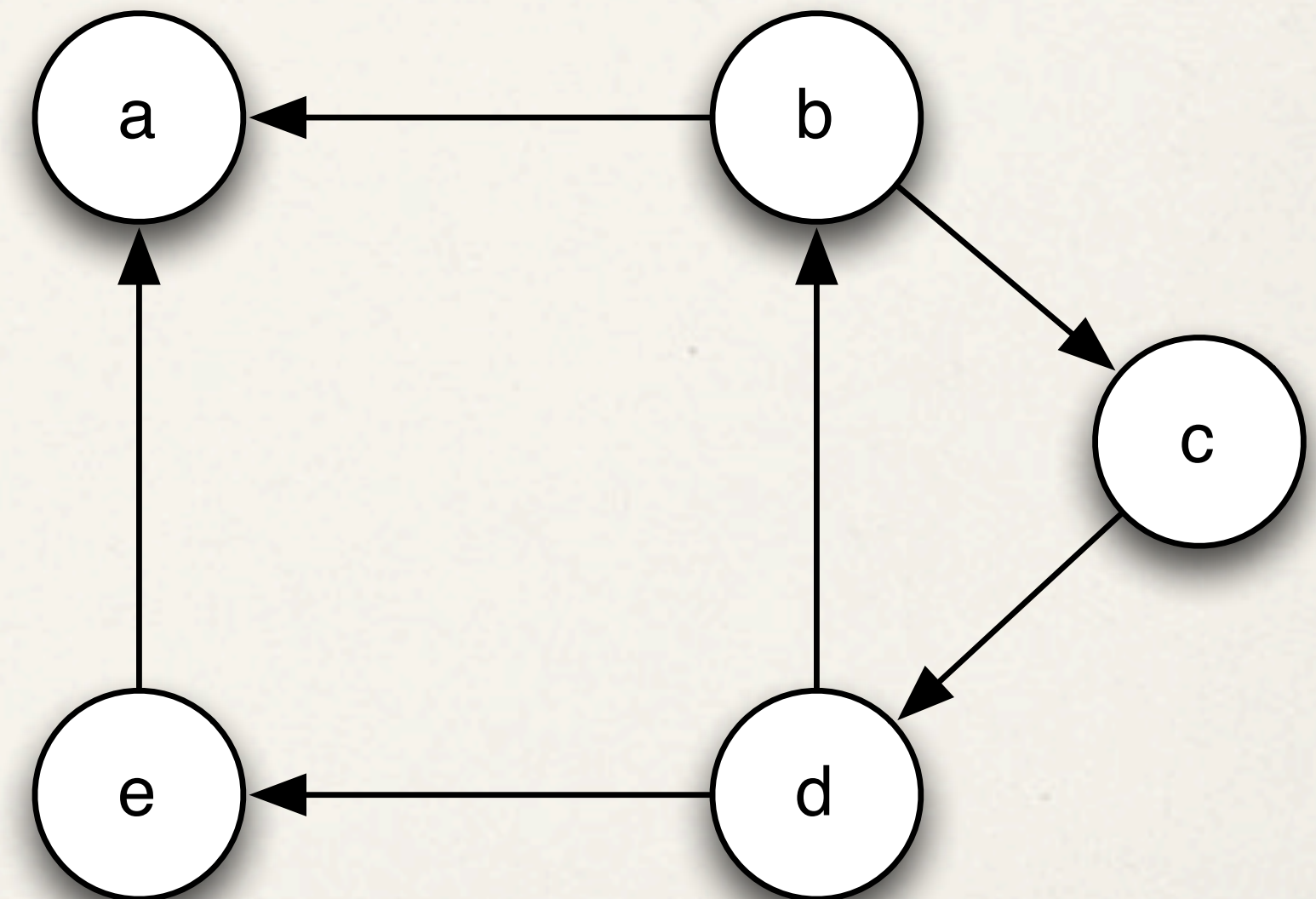
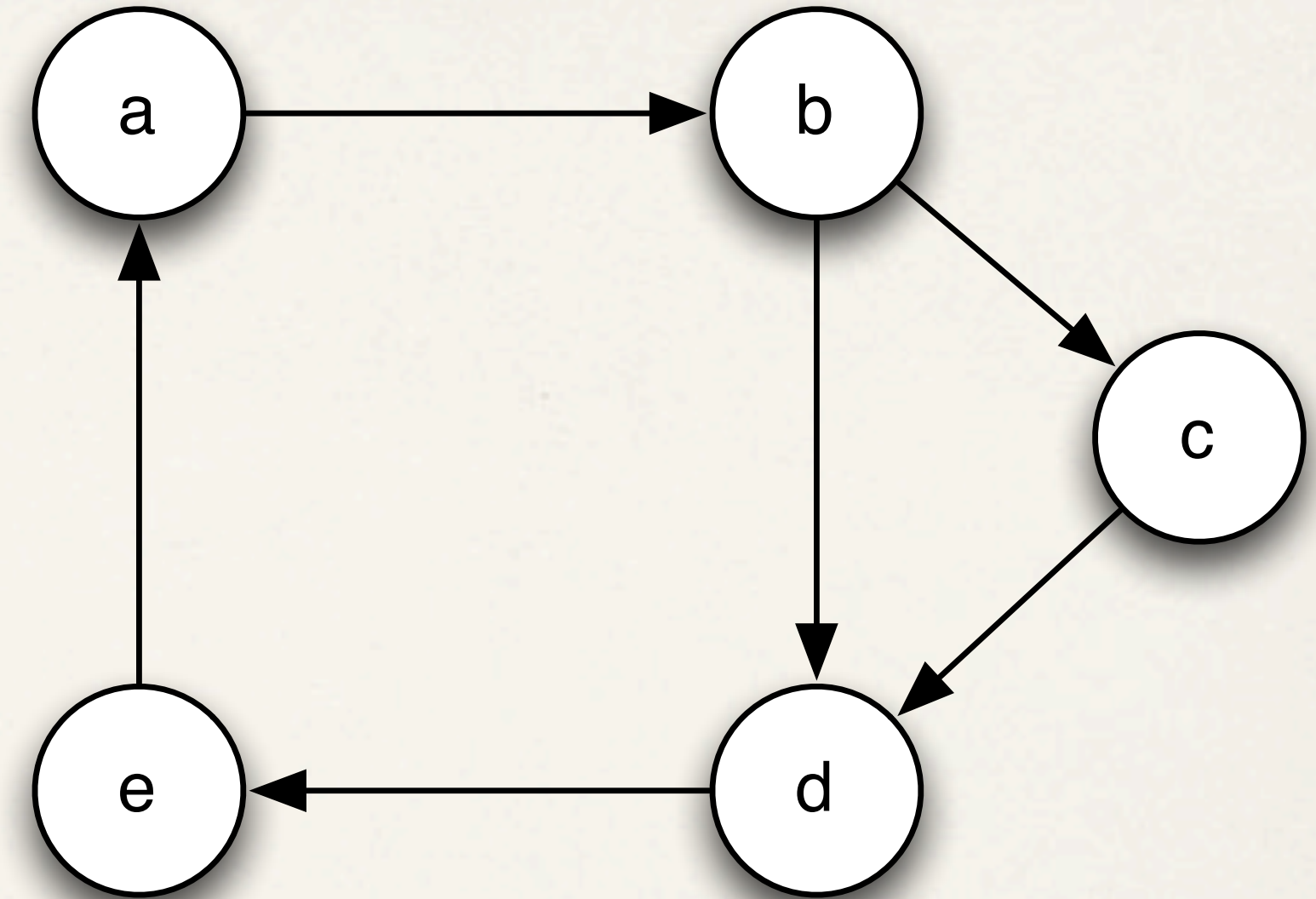
Directed Graphs

- ❖ The **underlying undirected graph** of a digraph is the same graph, minus directions
- ❖ A digraph is **weakly connected** if the underlying undirected graph is connected
 - ❖ i.e. if the digraph is “in one piece”

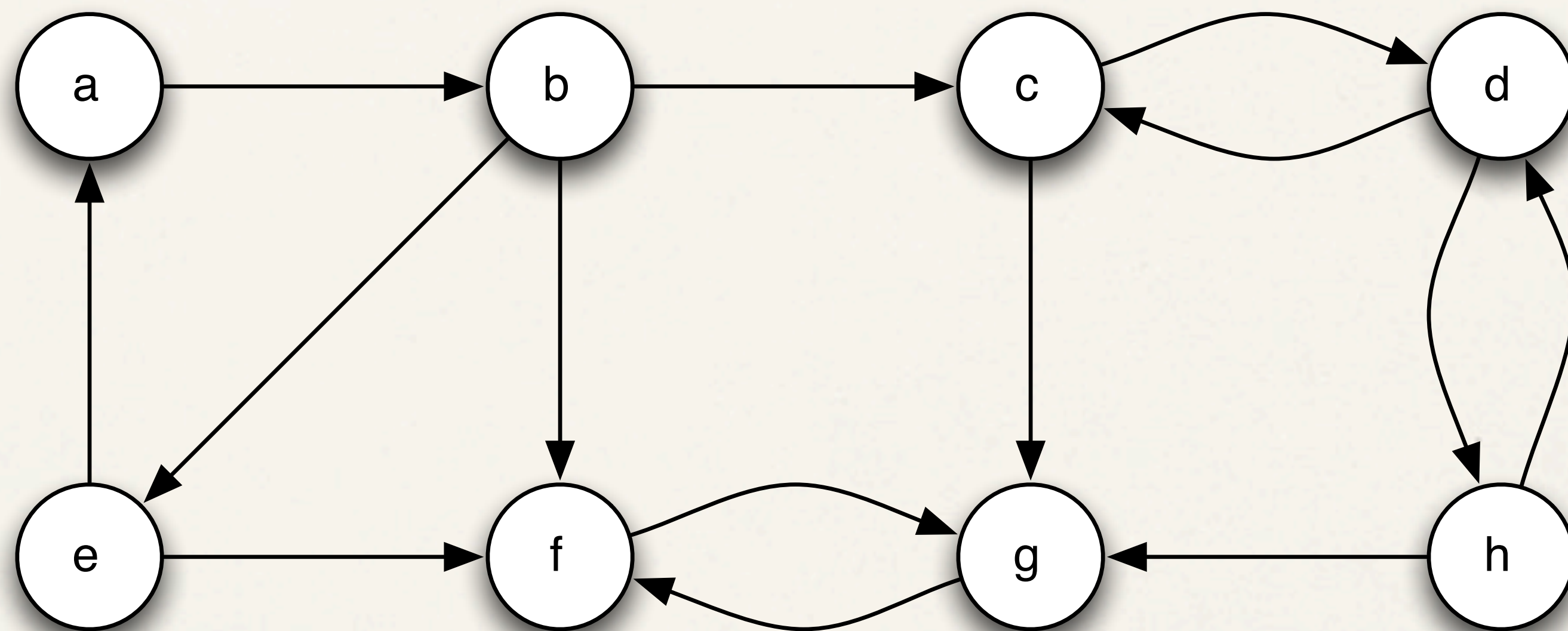


Strongly Connected

- ❖ A digraph G is **strongly connected** if for every pair of vertices $a, b \in G$, there exists a path from a to b and from b to a
 - ❖ Note: this includes the pair a, a



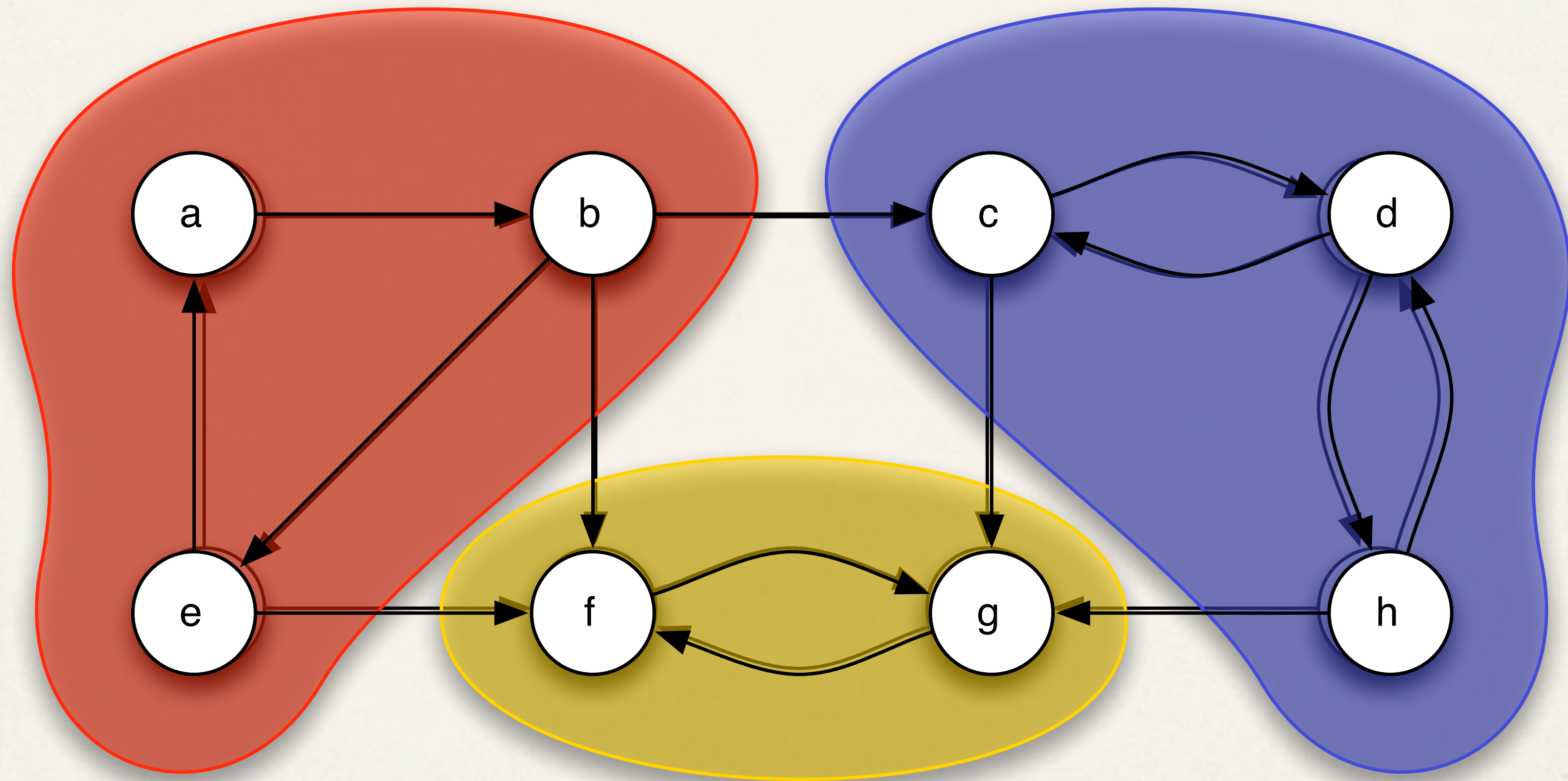
Practice



Strongly Connected Components

- ❖ Analogous to connected components
 - ❖ Maximal subgraphs of G which are strongly connected
i.e. they are not contained within any other such strongly connected subgraphs
- ❖ We can **compress** a digraph into a directed acyclic graph by reducing it to a graph of its strongly connected components
 - ❖ Treat all nodes in a SCC as interchangeable
 - ❖ Model how to hop from one SCC to another

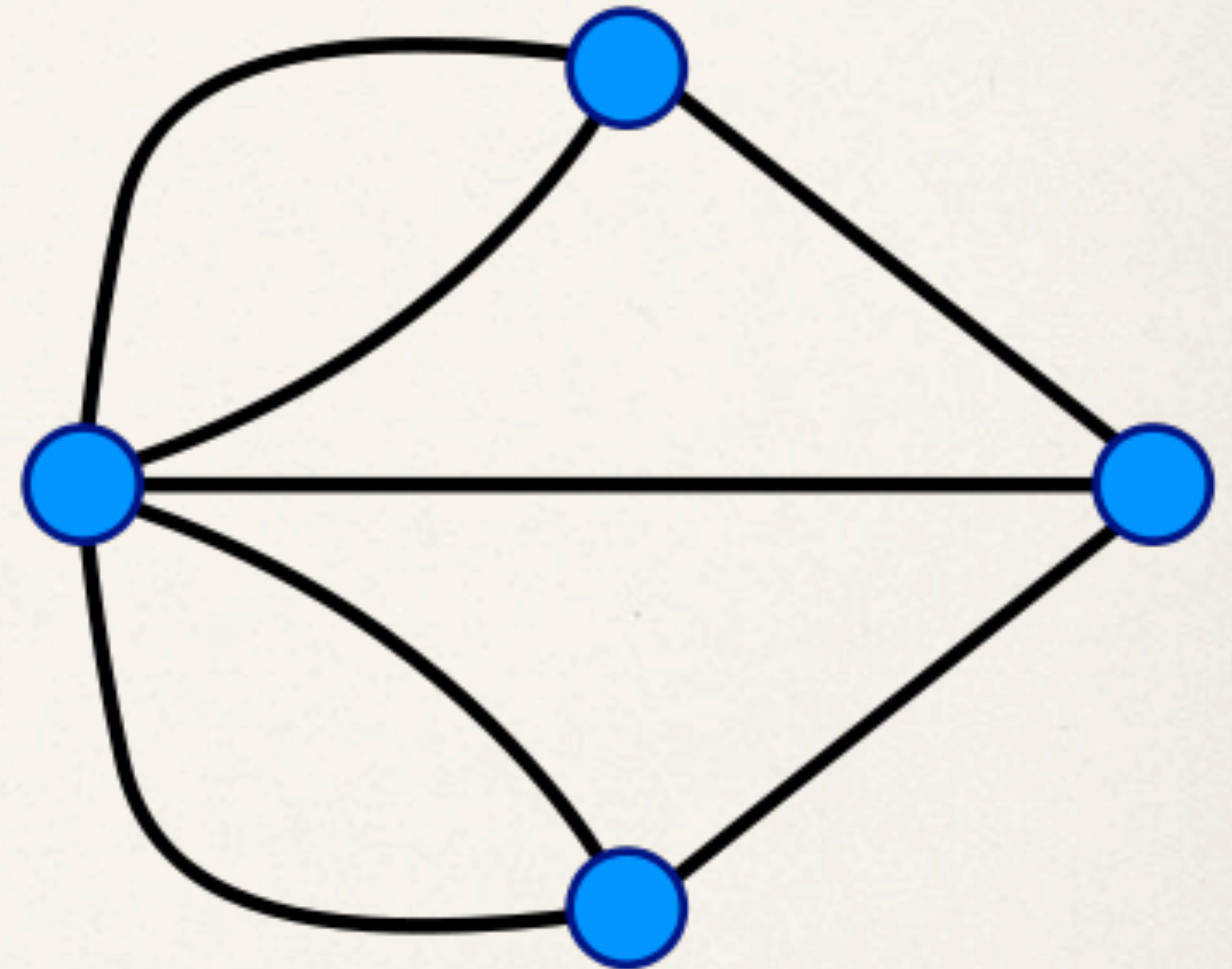
Strongly Connected Components



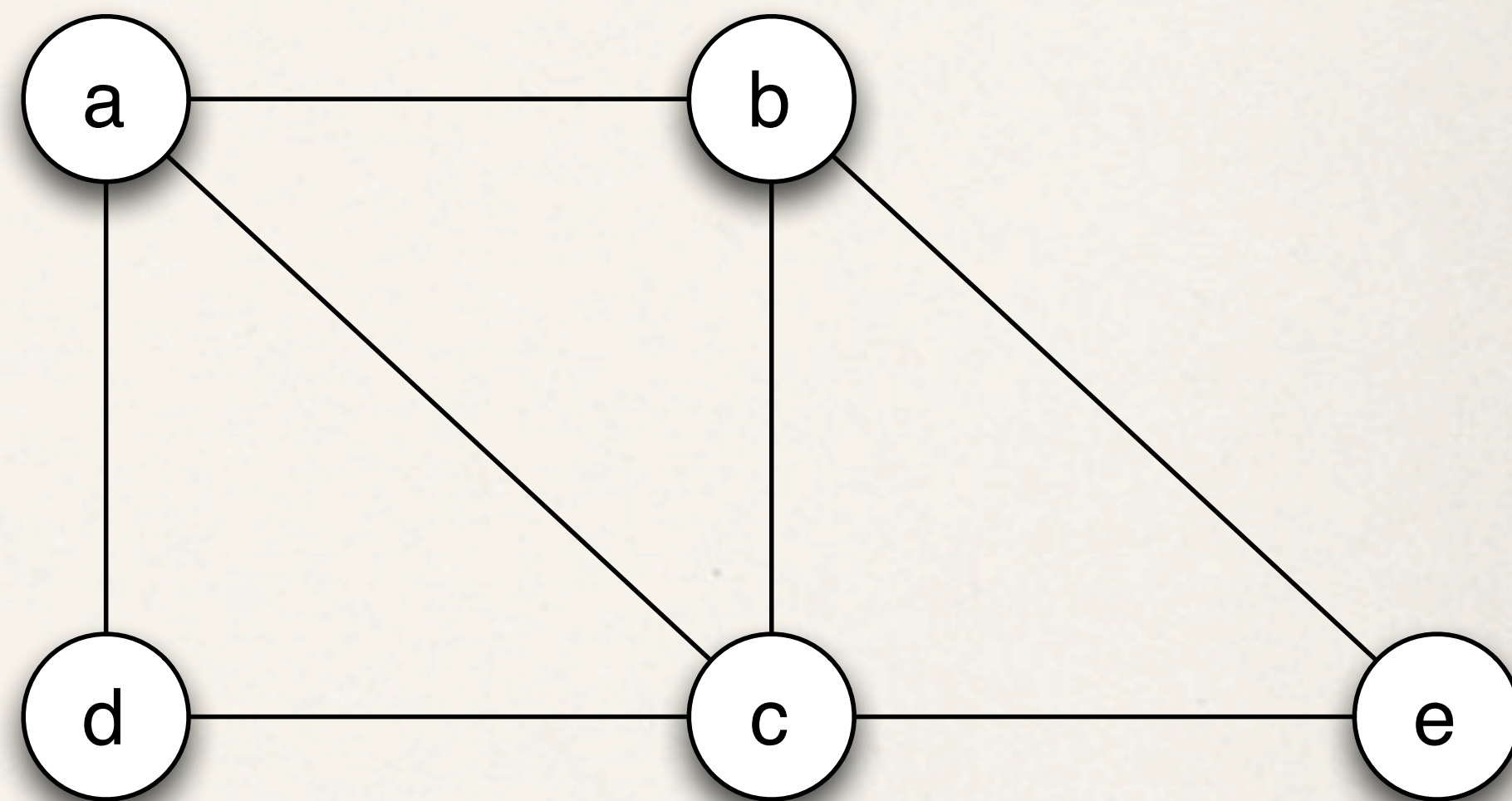
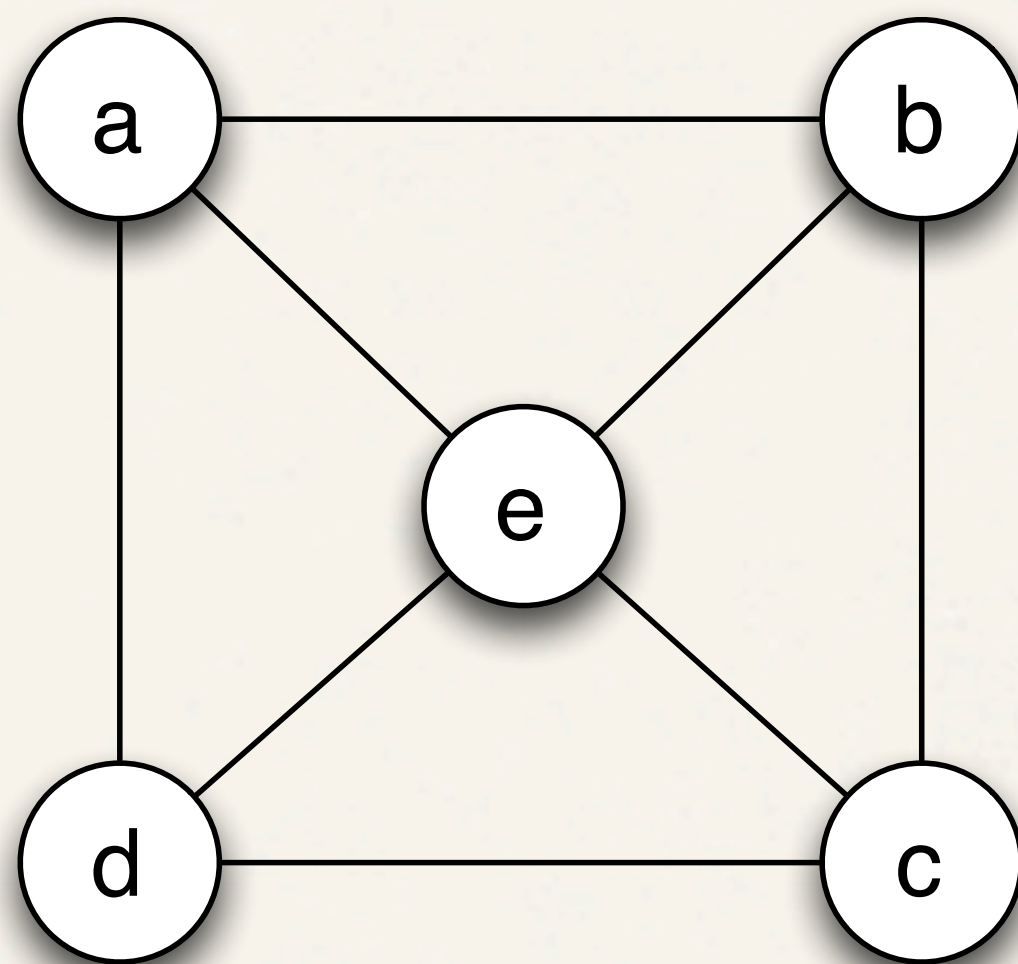
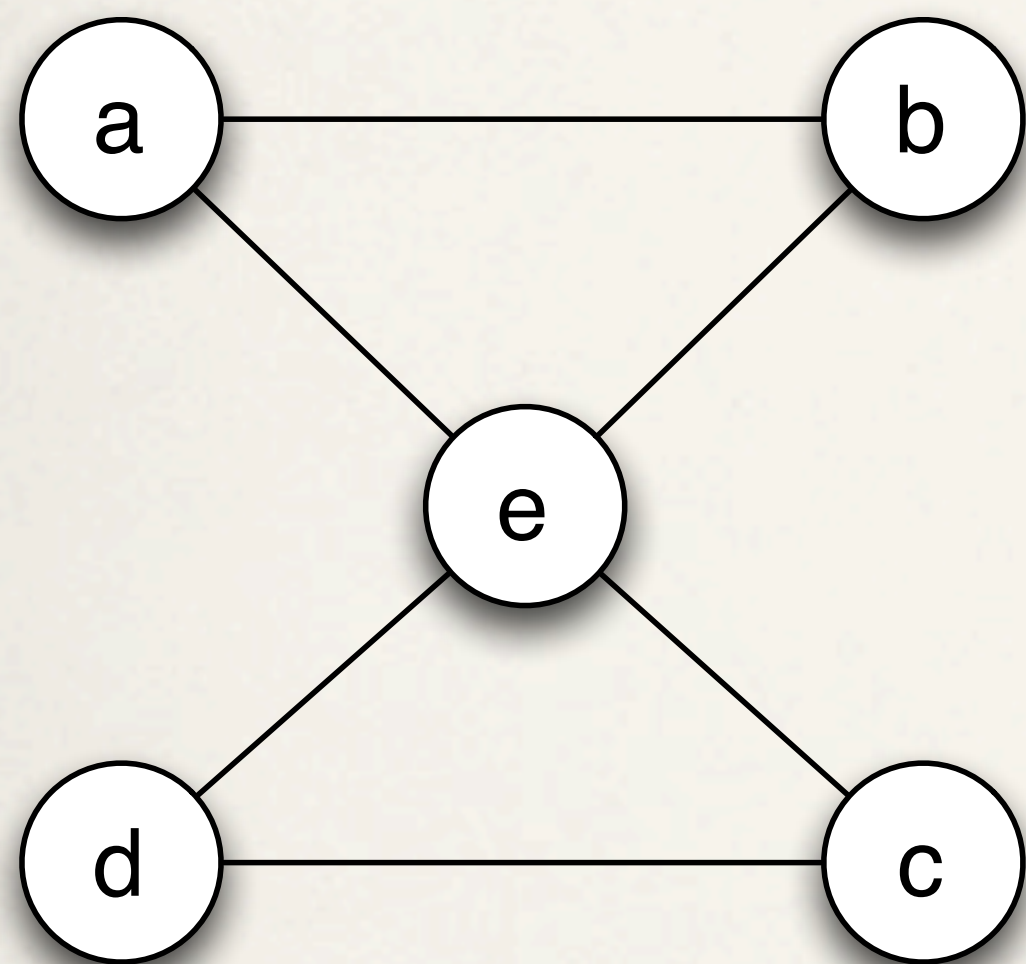
Euler and Hamilton Paths/Cycles

Euler Paths and Circuits

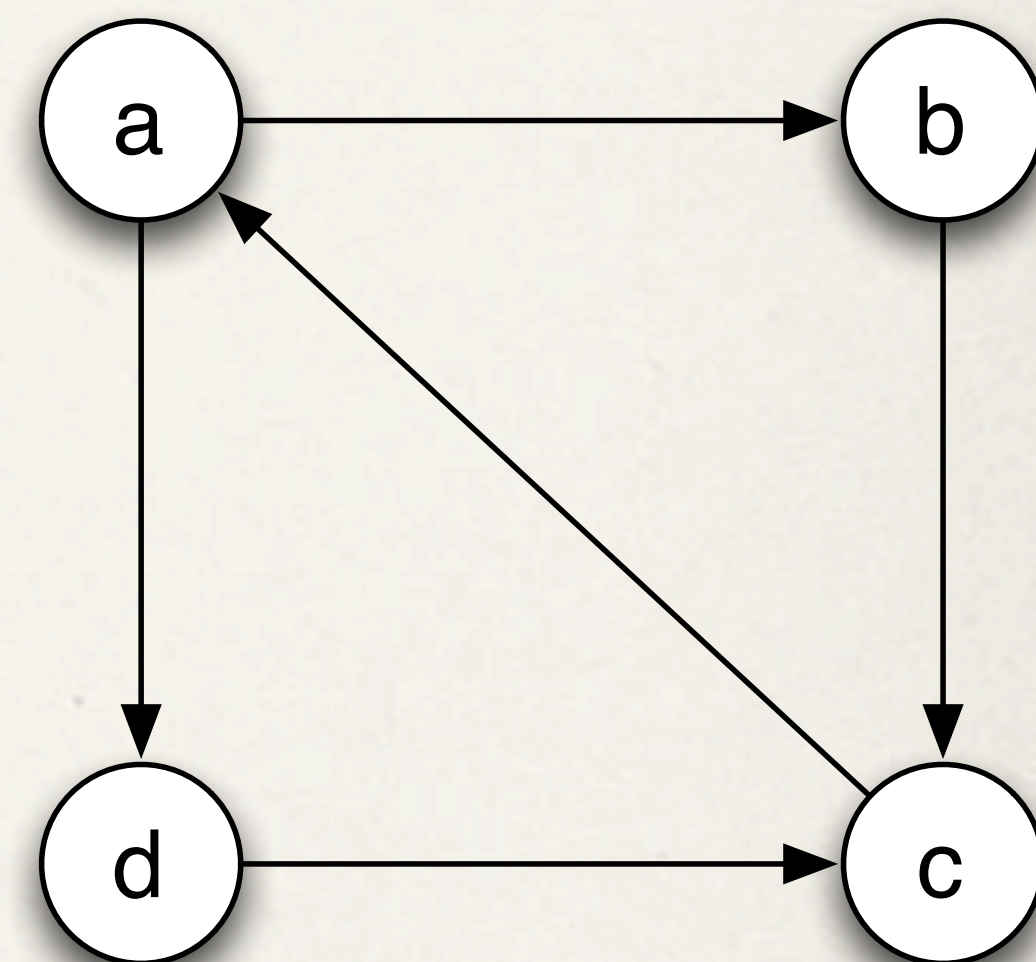
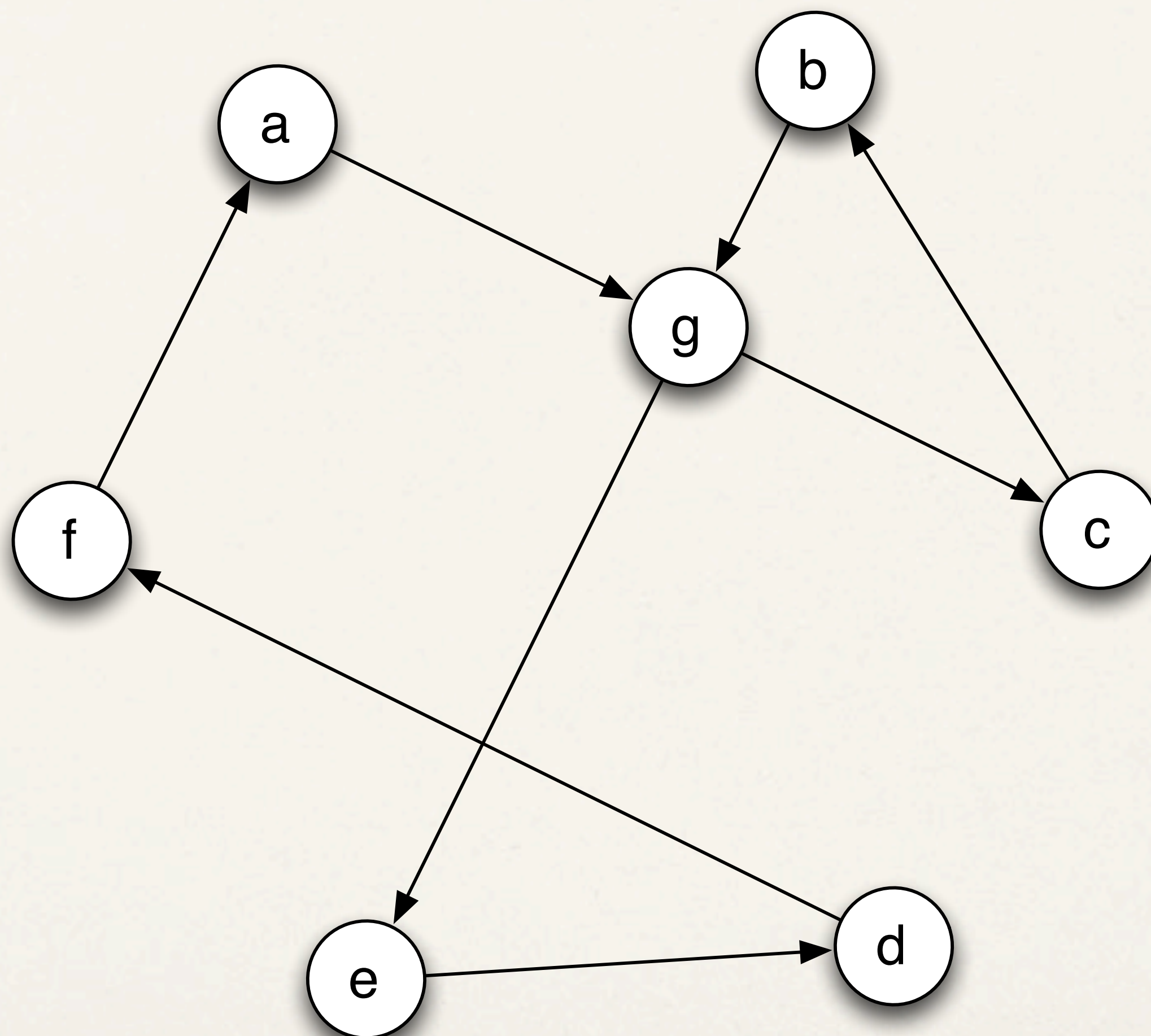
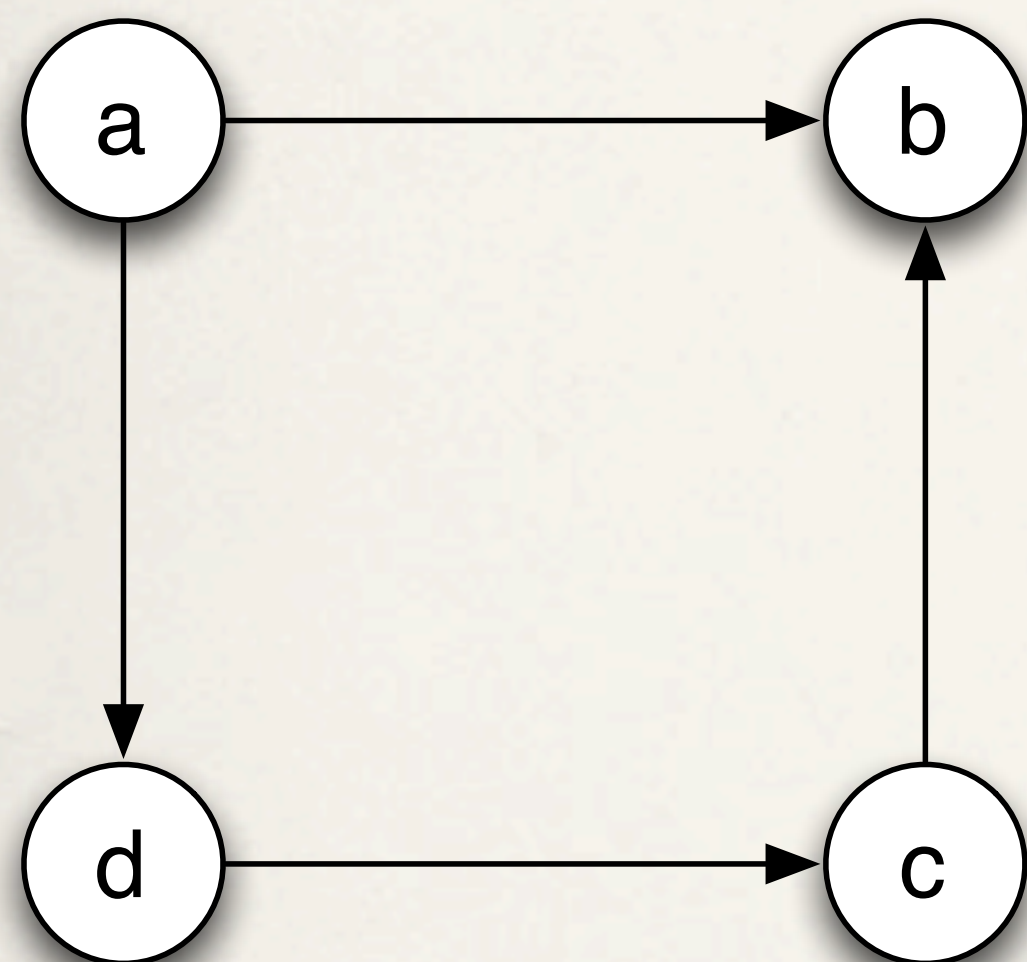
- ❖ **Euler Path:** path in G which uses every edge exactly once
 - ❖ can visit a node more than once
- ❖ **Euler Circuit:** Euler path which is also a circuit



Practice



Practice

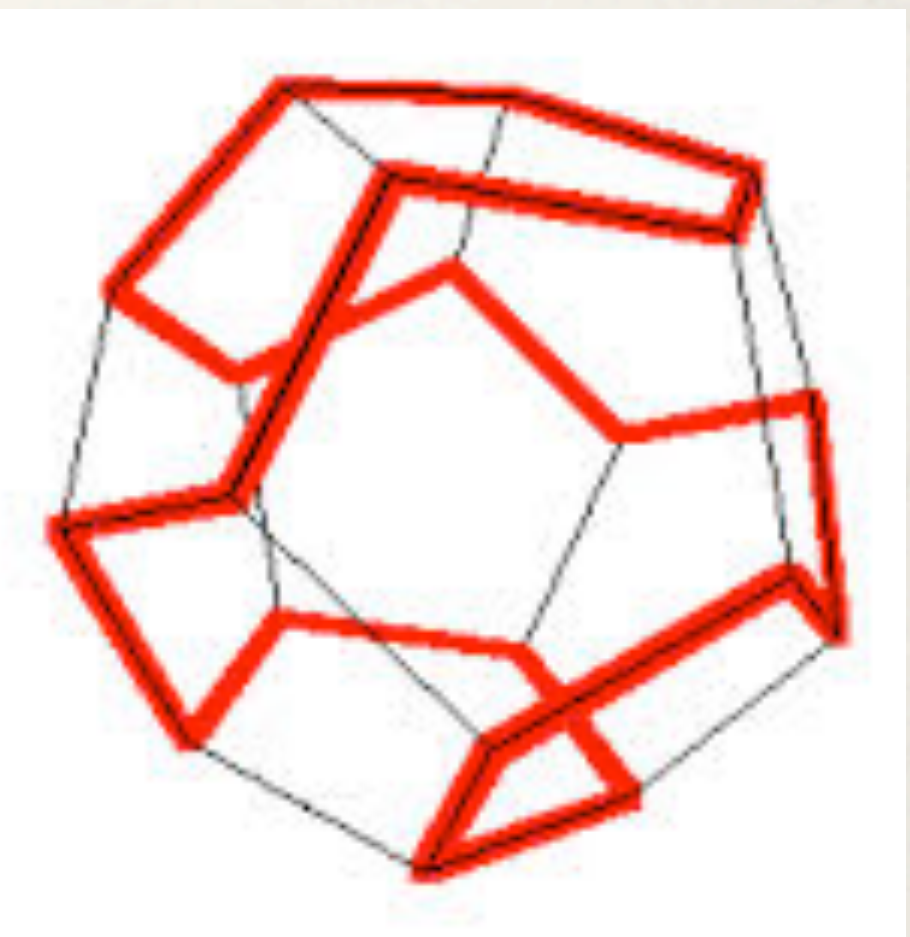
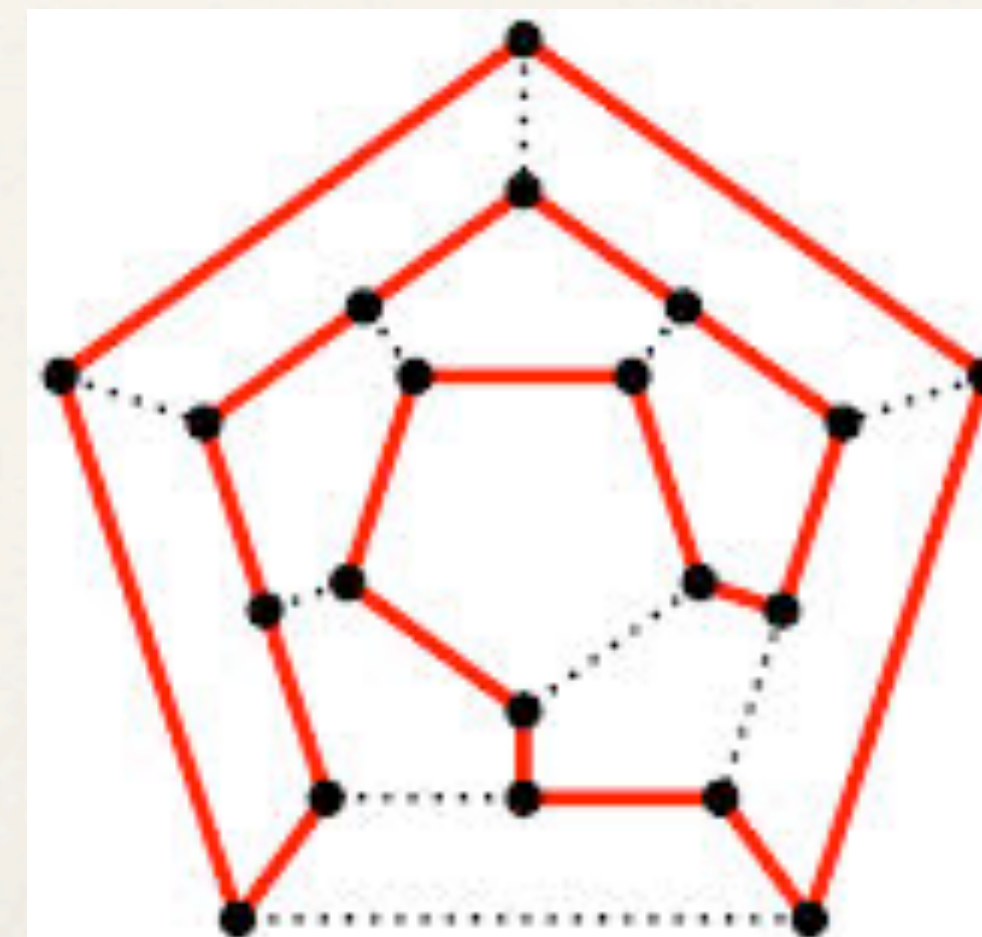
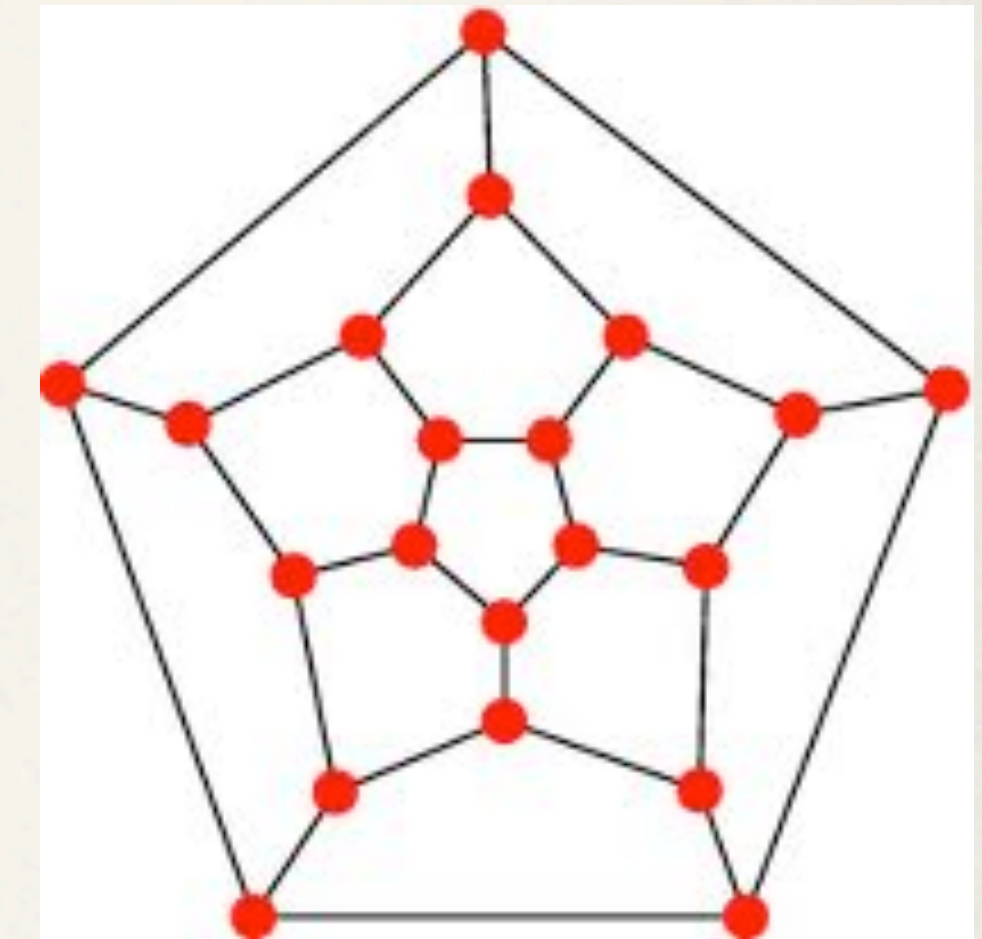
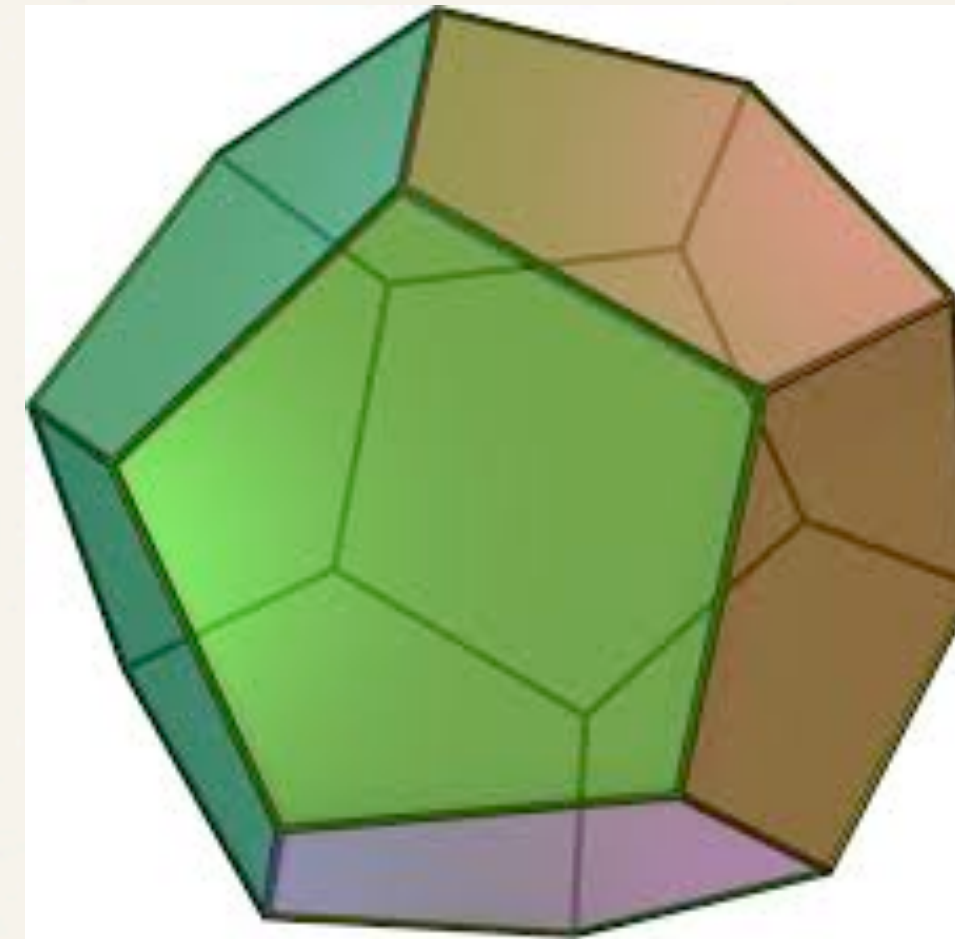


Hamilton Paths and Circuits

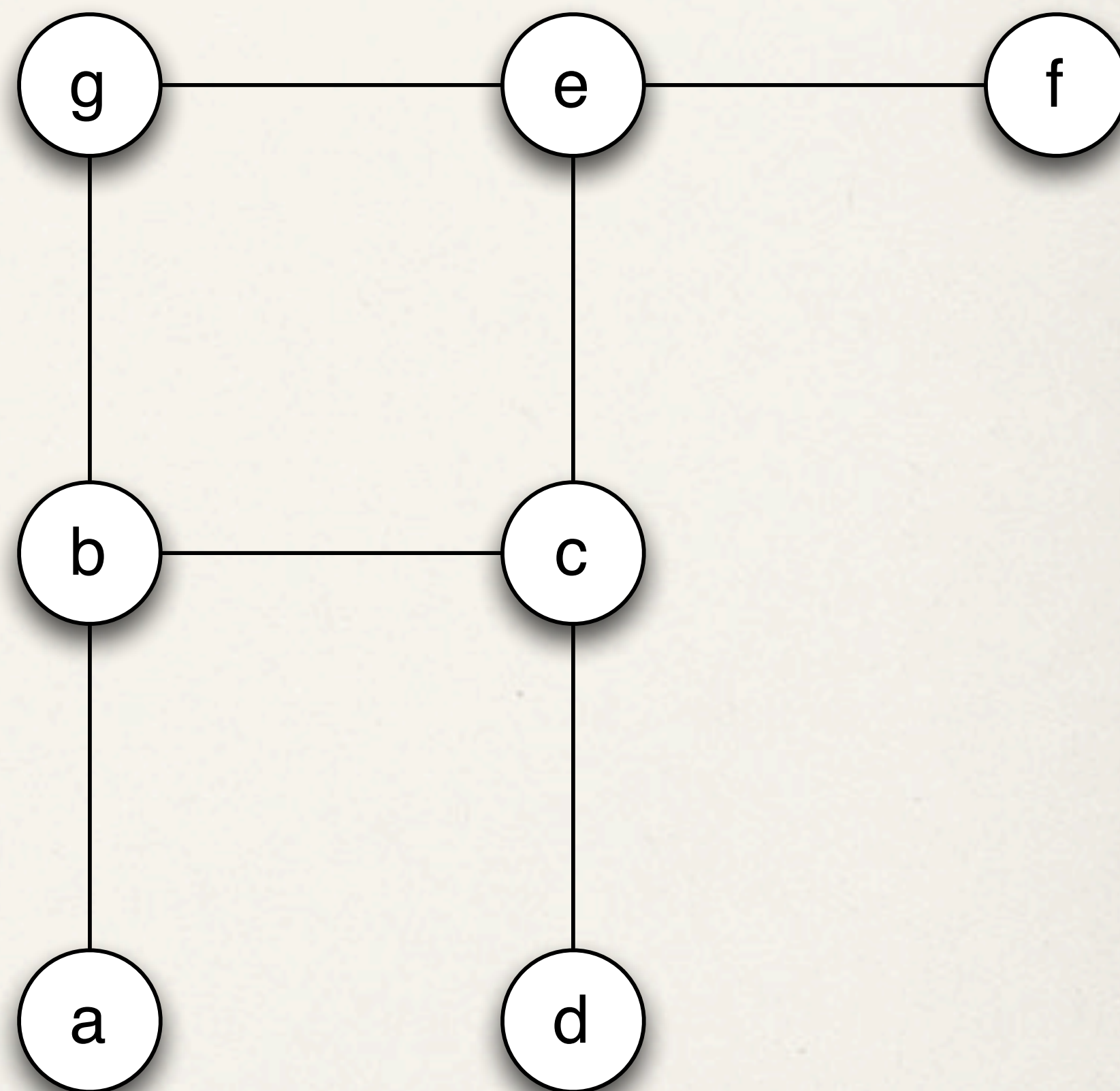
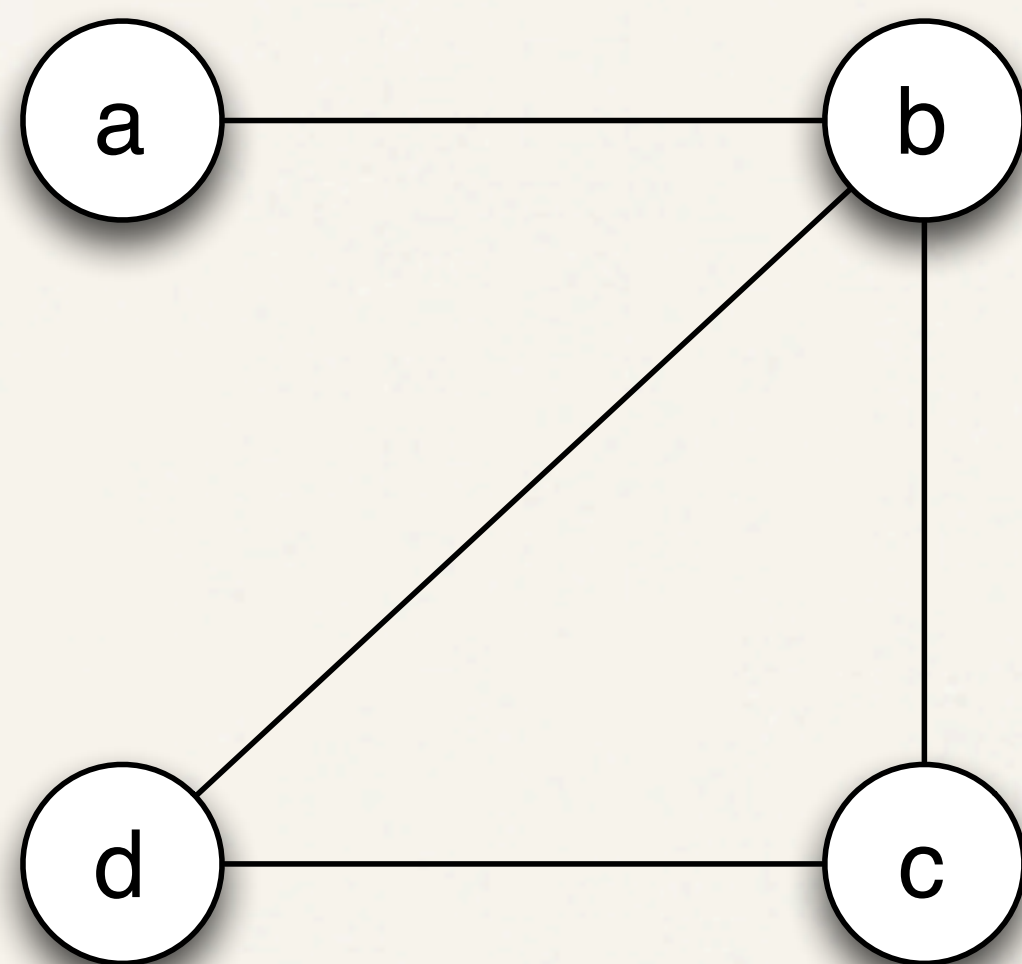
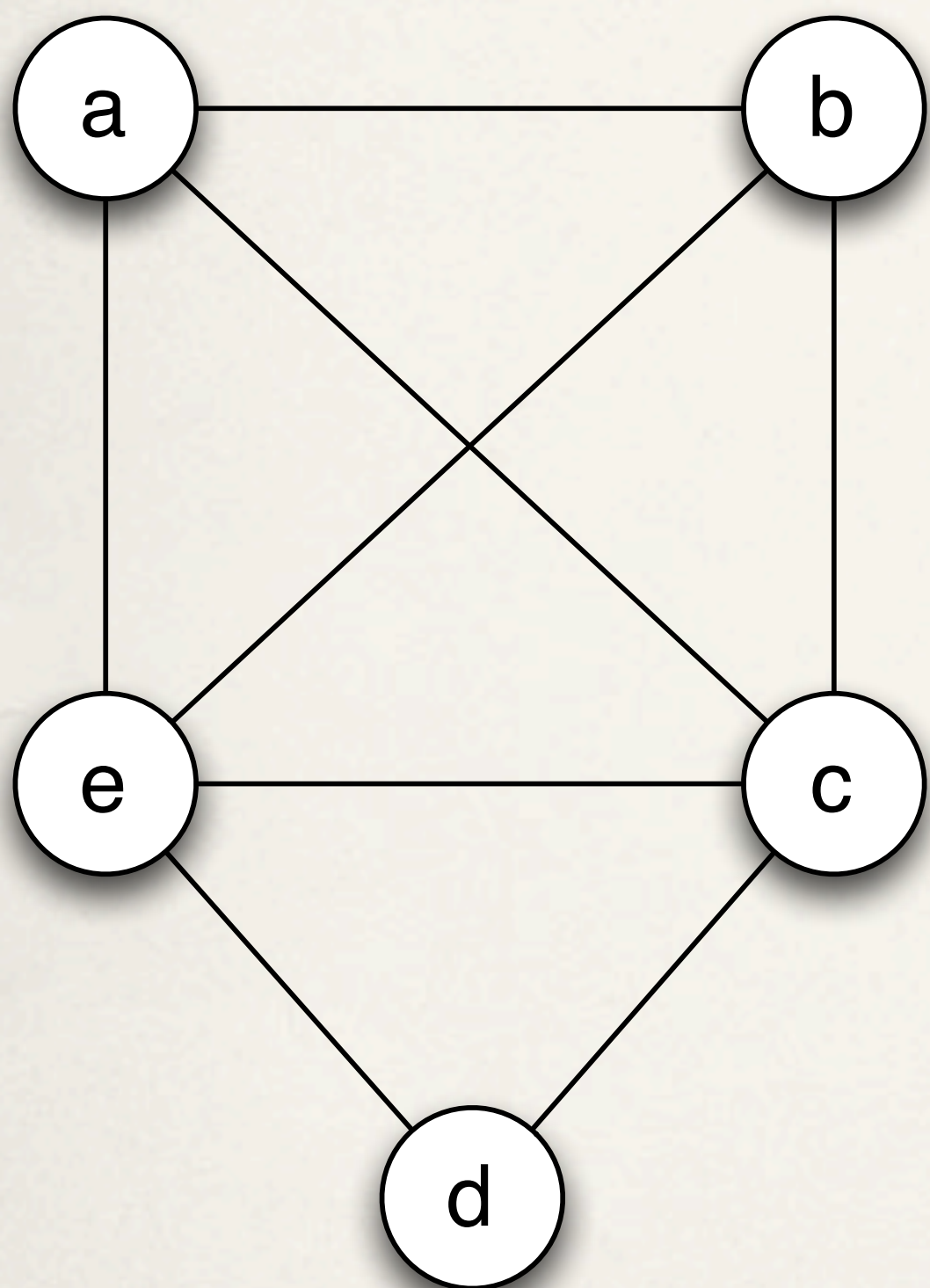
- ❖ **Hamilton Path:** path visiting every node exactly once

- ❖ each edge can only be used once

- ❖ **Hamilton Circuit:** hamilton path which is also a circuit



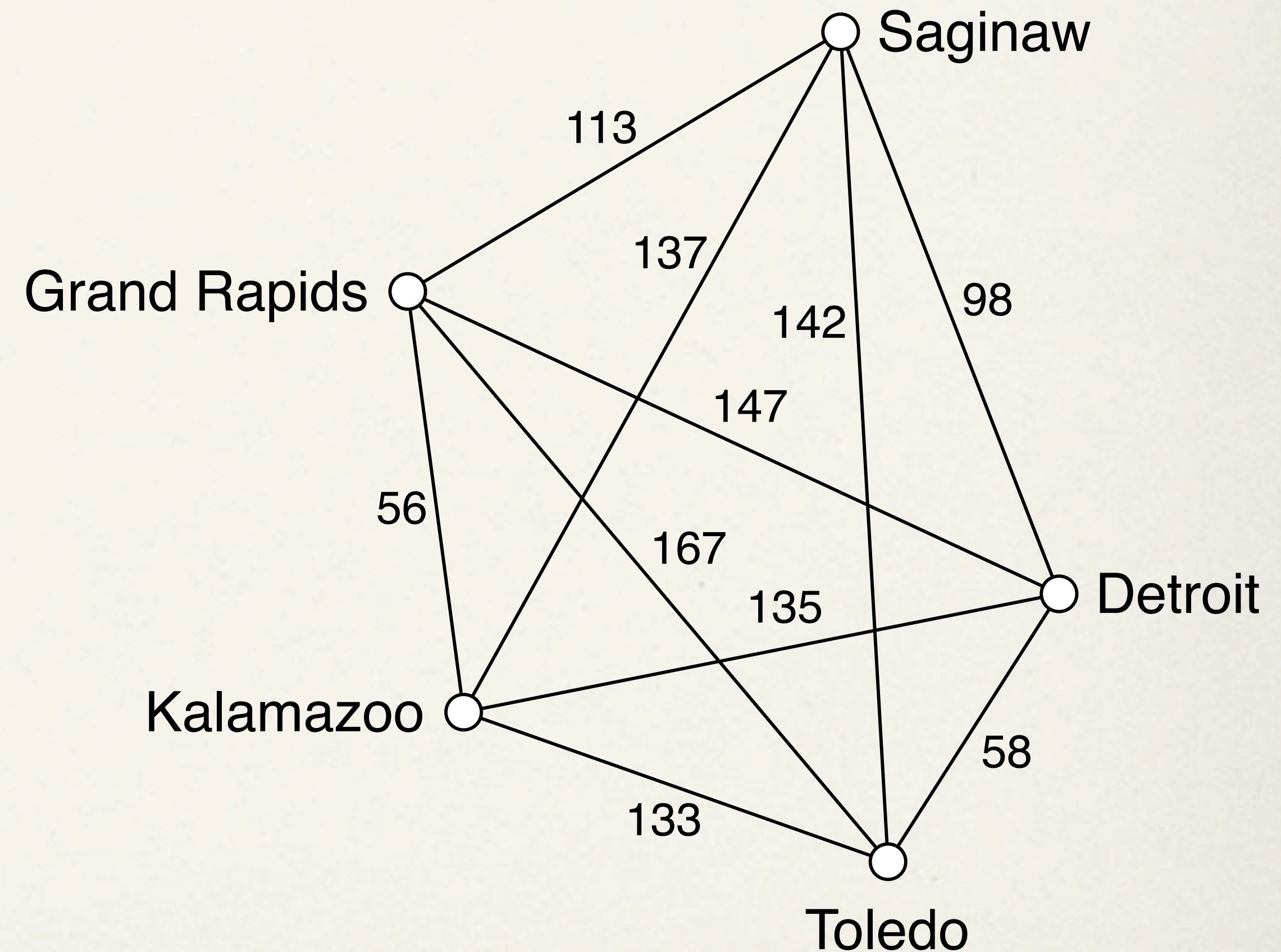
Practice



Traveling Salesperson Problem

- ❖ **Traveling Salesperson Problem (TSP)**
or **Traveling Salesman Problem**

- ❖ Find the shortest hamilton cycle
- ❖ i.e. visiting every node once and returning to the start



Traveling Salesperson Problem

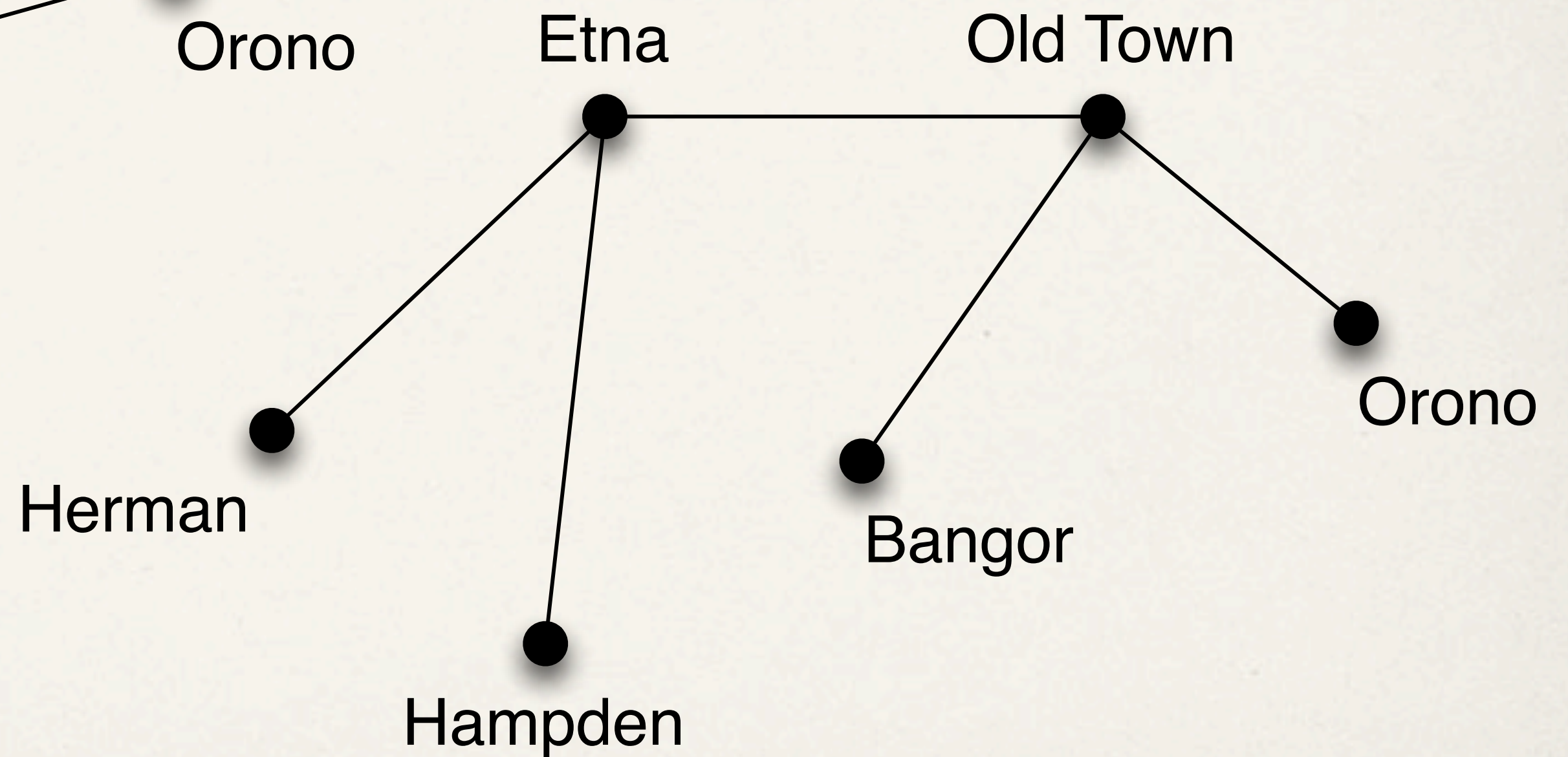
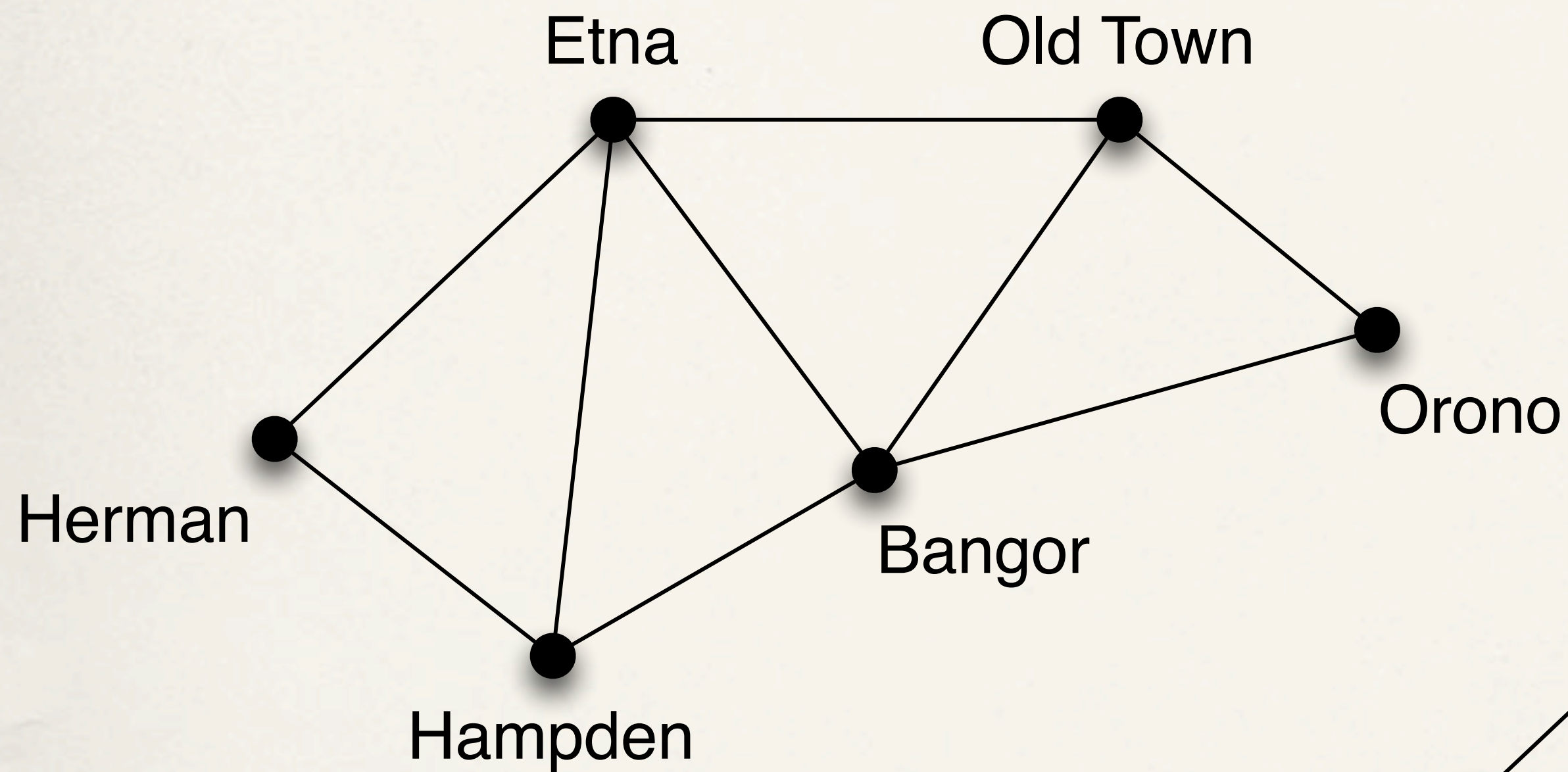
- ❖ This is an **NP-hard problem**
 - ❖ There are $(n - 1)! / 2$ Hamilton cycles
 - ❖ $O(n!)$ to compute exhaustively
 - ❖ FedEx pays lots and lots of money for improvements on this problem
 - ❖ Often okay to find an **approximate solution**

Route	Total
D - T - GR - S - K - D	610
D - T - GR - K - S - D	516
D - T - K - S - GR - D	588
D - T - K - GR - S - D	458
D - T - S - K - GR - D	540
D - T - S - GR - K - D	504
D - S - T - GR - K - D	598
D - S - T - K - GR - D	576
D - S - K - GR - T - D	682
D - S - GR - T - K - D	646
D - GR - S - T - K - D	670
D - GR - T - S - K - D	728

Spanning Trees

Section 9.5

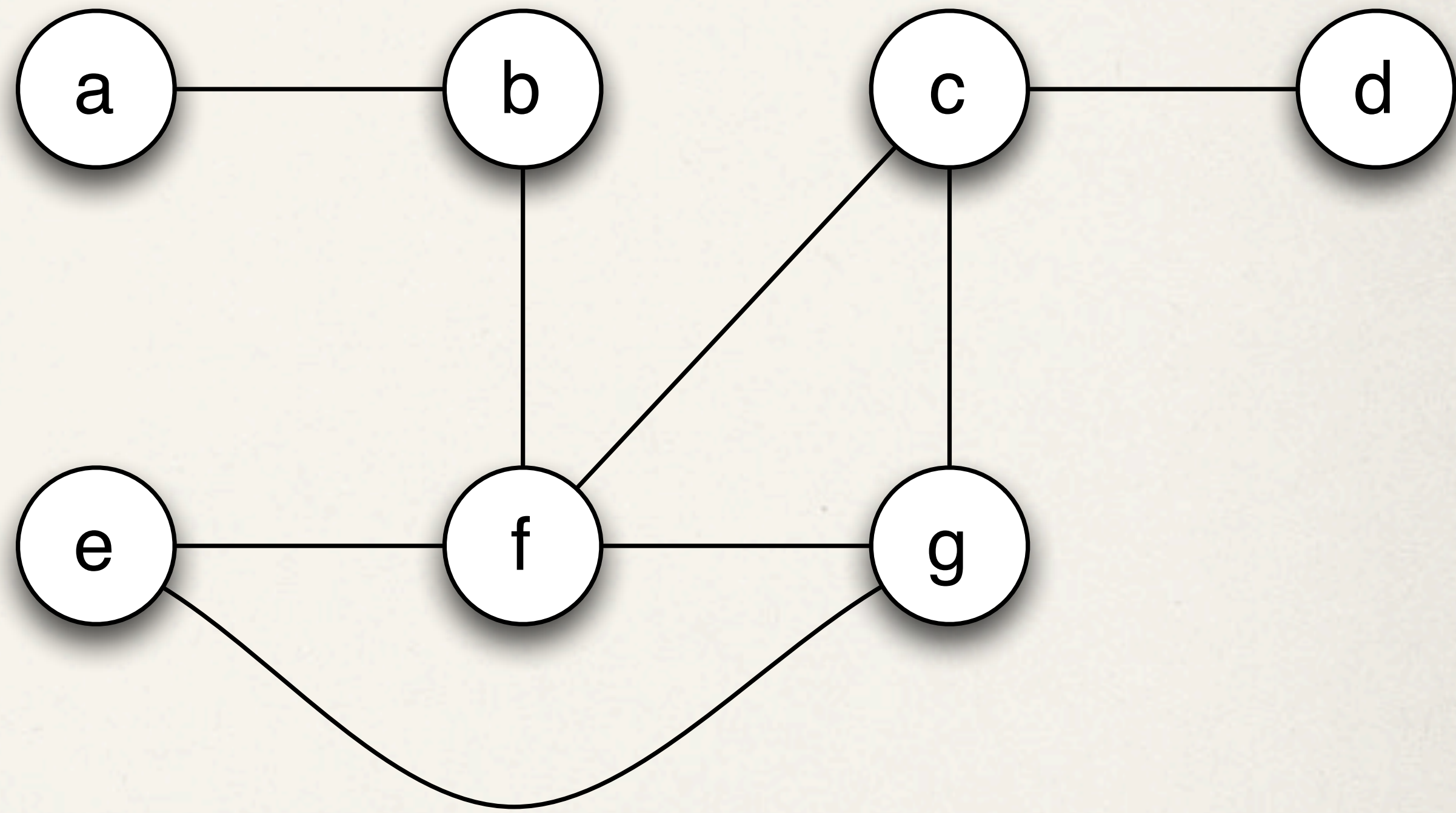
Spanning Tree



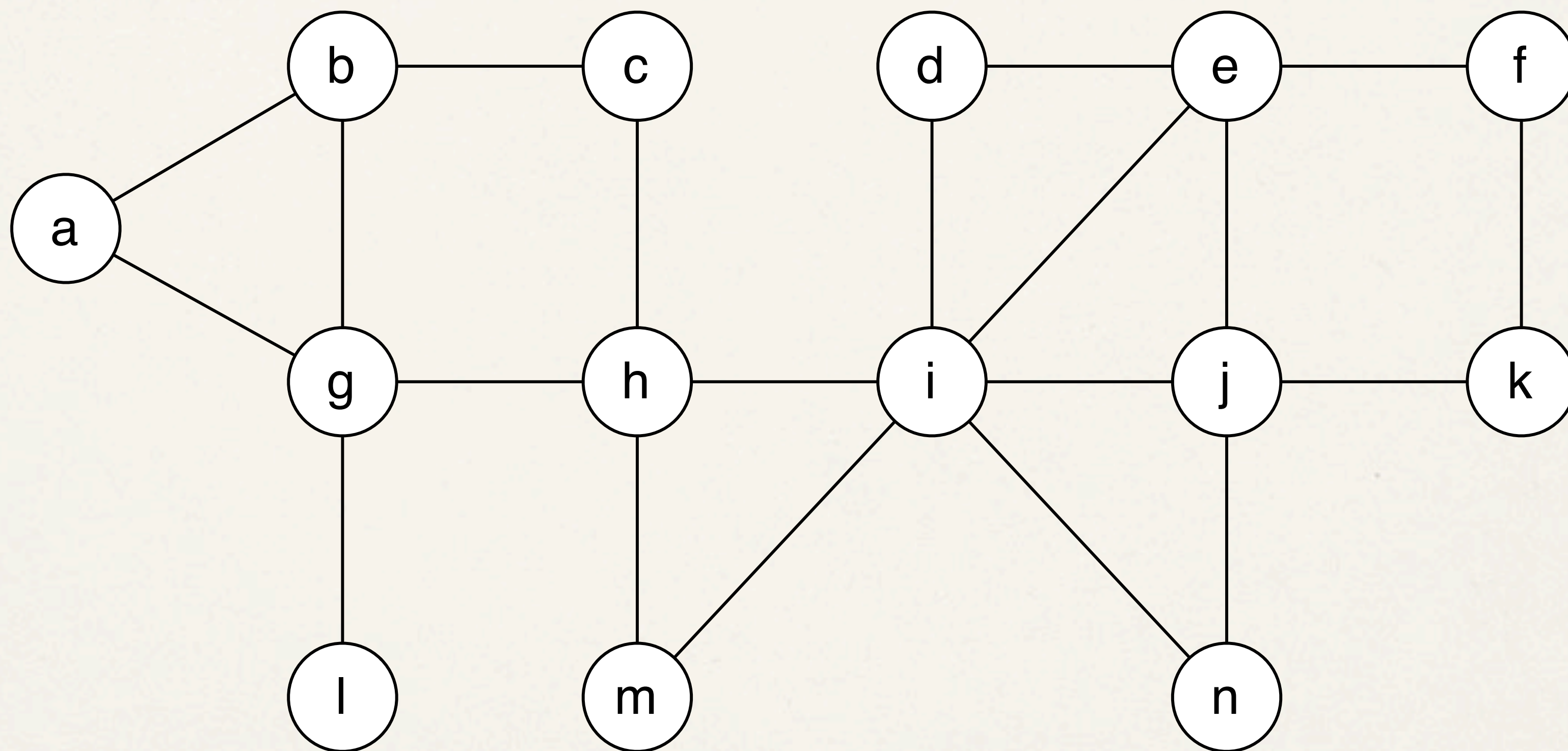
- ❖ **Spanning Tree**
- ❖ Tree w/ every vertex of G

Finding Spanning Trees

- ❖ Trees are acyclic, graphs aren't
- ❖ Remove edges to break up cycles
 - ❖ Keep graph connected

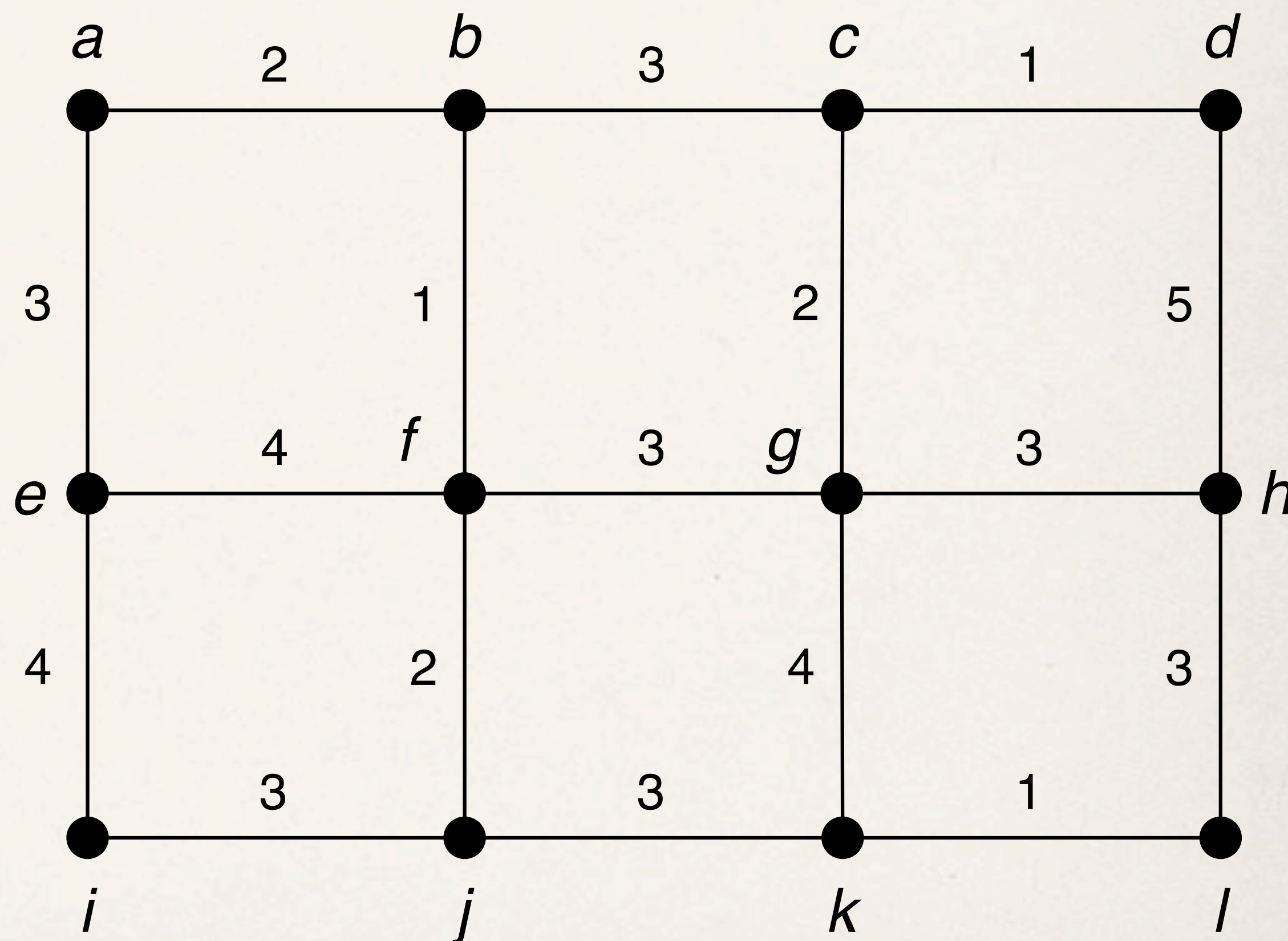


Practice



Minimum Spanning Trees

- ❖ Weighted graphs
- ❖ Find spanning tree with the smallest possible sum of edge weights
- ❖ Applications:
 - ❖ Making connected graphs



Prim's Algorithm

procedure *Prim* (G)

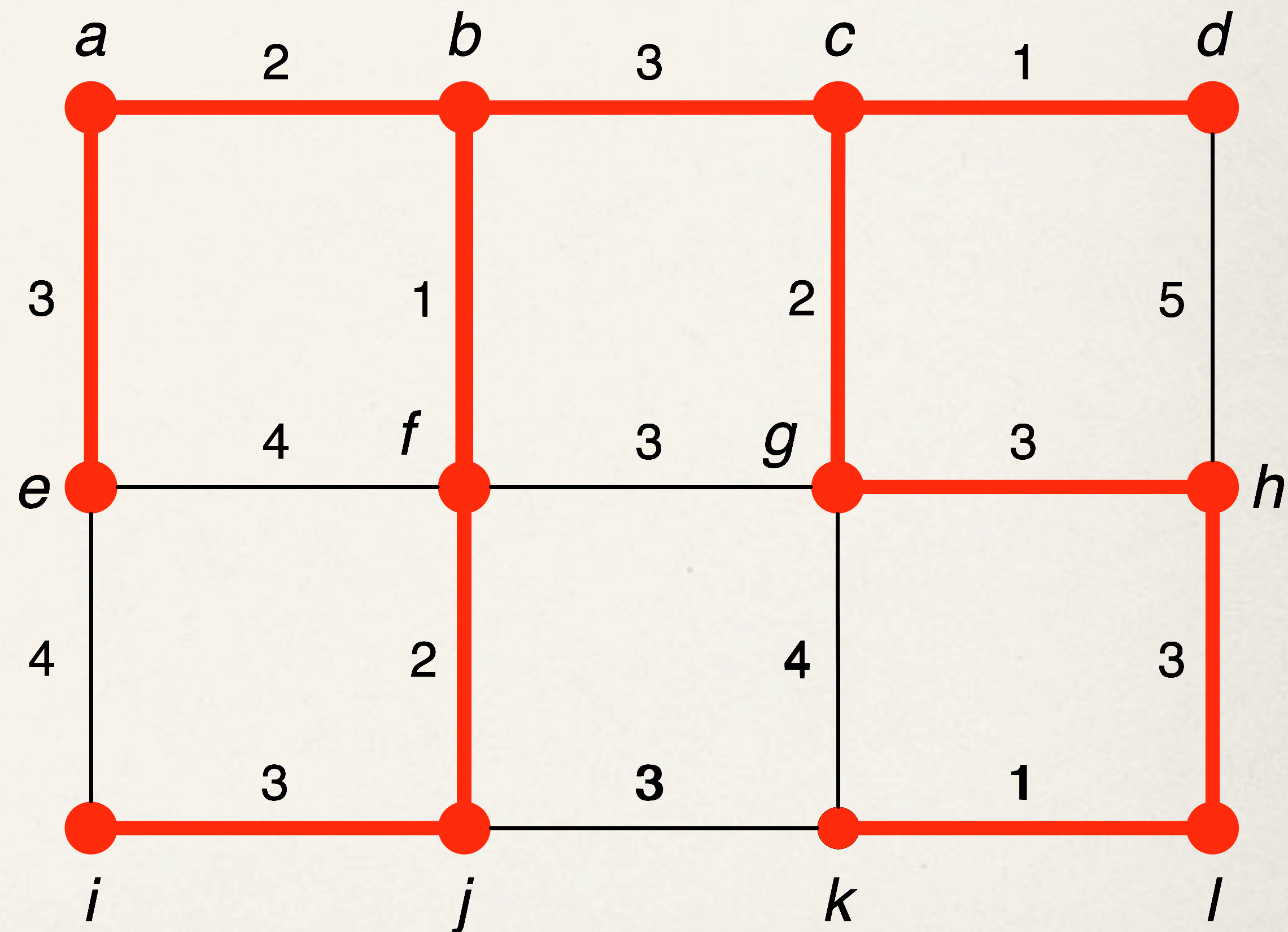
$T :=$ a minimum weight edge

for $i := 1$ **to** $n - 2$

$e :=$ an edge of minimum weight,
incident to a vertex in T ,
not forming a cycle in T

$T := T$ with e added

return T



Kruskal's Algorithm

procedure *Kruskal* (G)

$T :=$ an empty graph

for $i := 1$ **to** $n - 1$

$e :=$ any edge of minimum weight in G ,
not forming a cycle if added to T

$T := T$ with e added

return T

