

# Graphs

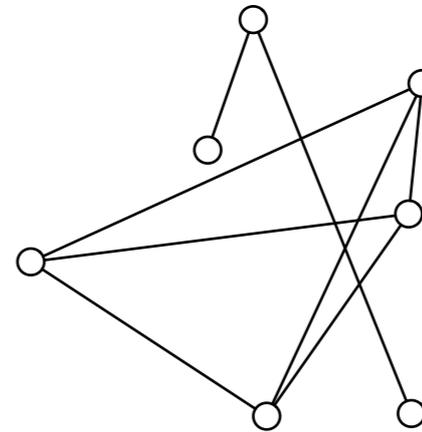
Thomas Schwarz, SJ

# Graph Definition

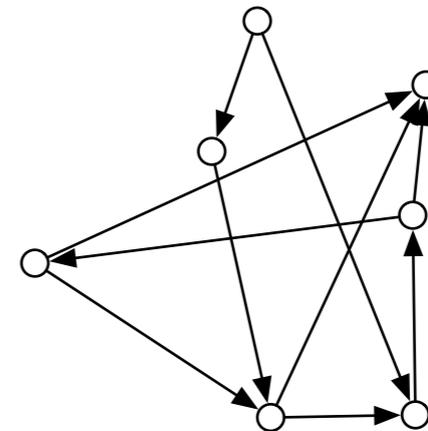
- A graph has a set of vertices  $V$  and a set of edges.
  - Directed edges are pairs  $(u, v)$  with  $u, v \in V$
  - Undirected edges are two-sets  $\{u, v\}$  with  $u, v \in V$
- A graph with directed edges is called a directed graph
- A graph with undirected edges is just called a graph

# Graph Definition

- Graphs are represented by:
  - drawing the vertices as small circles
  - drawing the edges as edges
- Directed edges are drawn as arrows



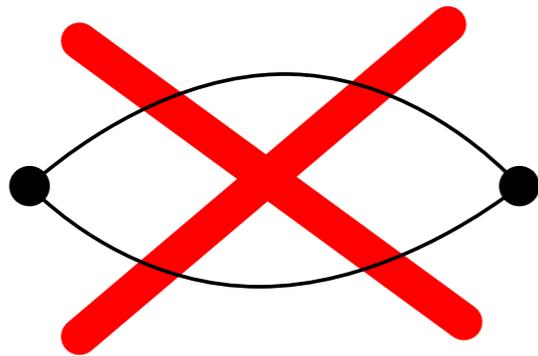
An undirected graph with 7 vertices and 7 edges



A directed graph

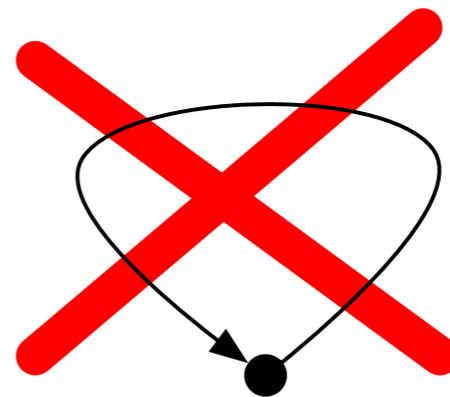
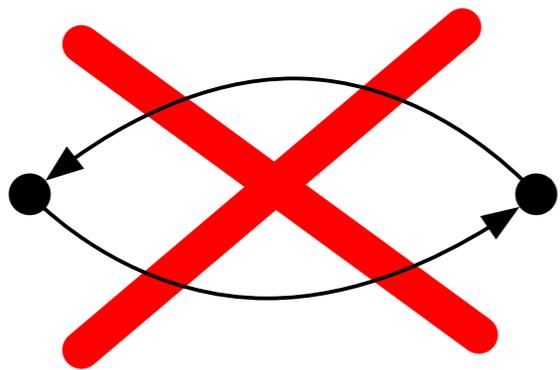
# Graph Definition

- Computer scientist sometimes differ from mathematicians in what is called a graph
  - In Mathematics, a(n undirected) graph can
    - Have only one edge at most between two vertices
    - Cannot have an edge to the same vertex



# Graph Definition

- Computer scientist sometimes differ from mathematicians in what is called a graph
  - In Mathematics, a directed graph can
    - Have only one edge at most between two vertices
    - Cannot have an edge to the same vertex



# Graph Definition

- Mathematicians call a graph that allows multiple edges between the same pair of vertices
  - a multigraph

# Graph Representations

- To understand graphs, we can use:
  - The visual representation
    - E.g. The neighbor graph
    - Take a political map



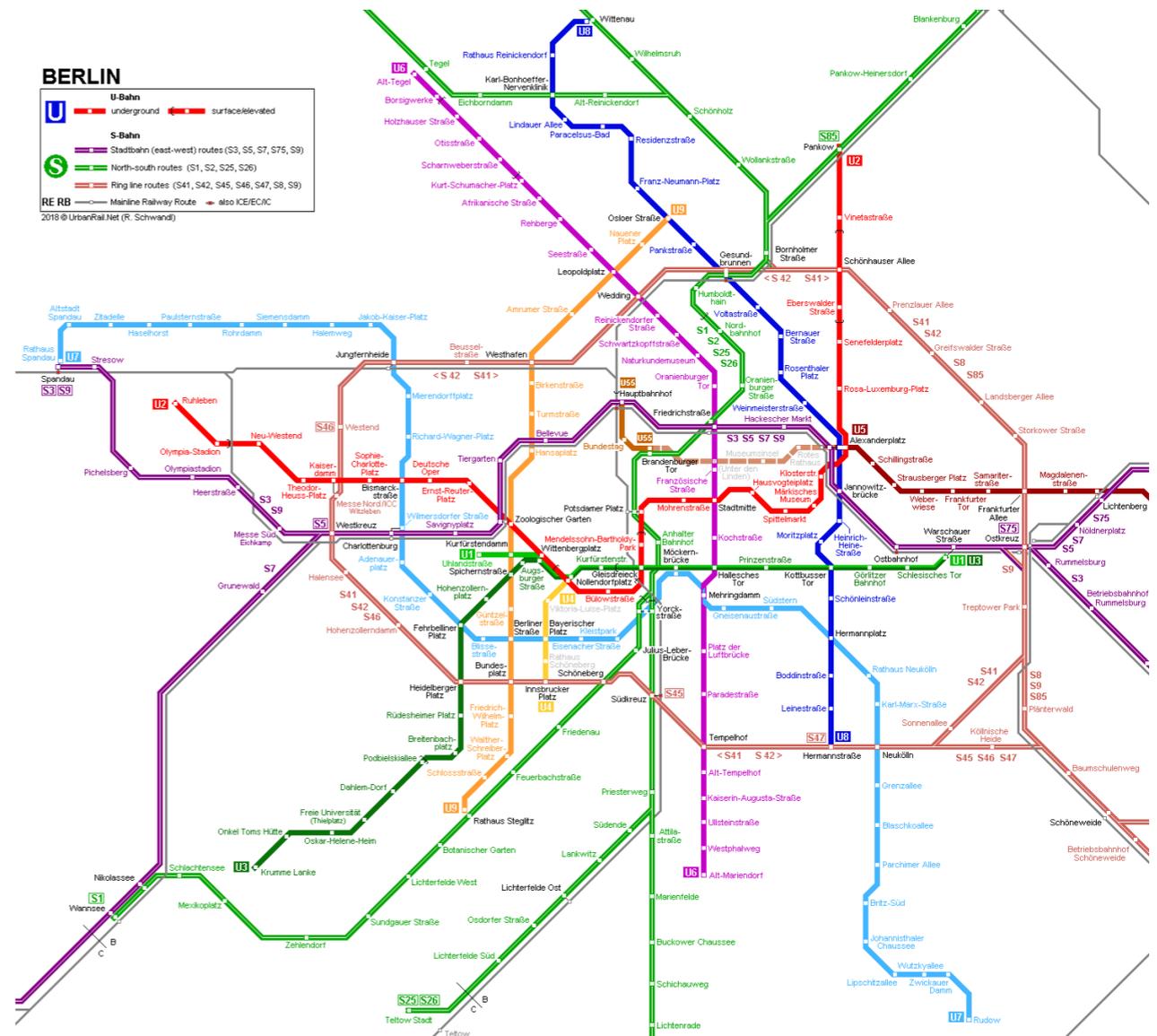
# Graph Representations

- Examples:
  - Place a vertex in every entity (state, not DDFF).
  - Connect vertices if the entities have a common border



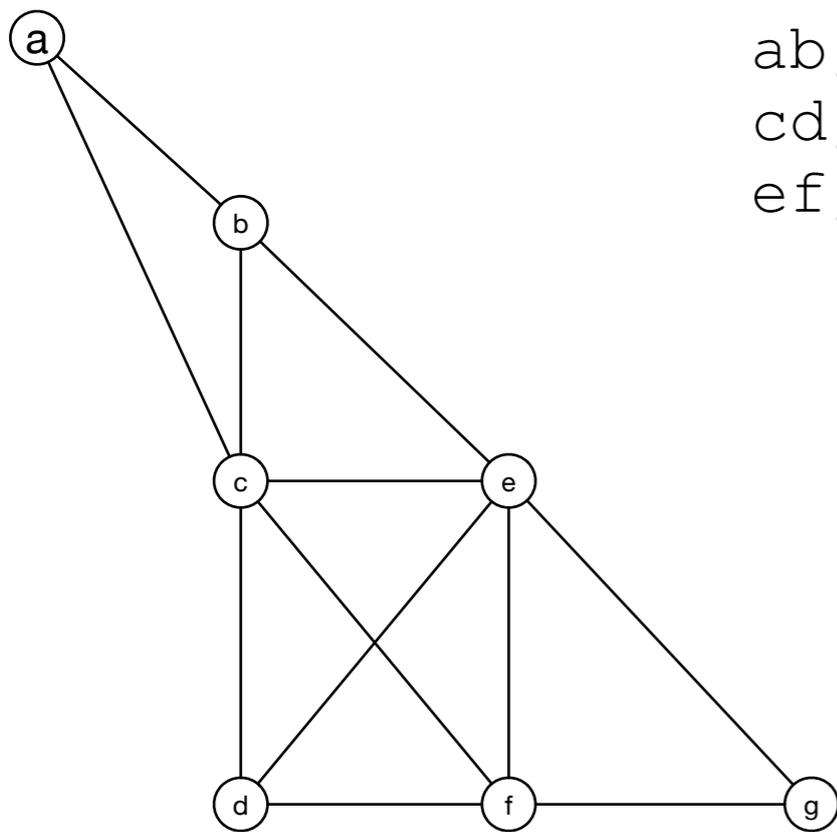
# Graph Representations

- Vertices are stations
- Edges represent a connection via underground or light rail
- This is multi-graph because several edges can connect a station

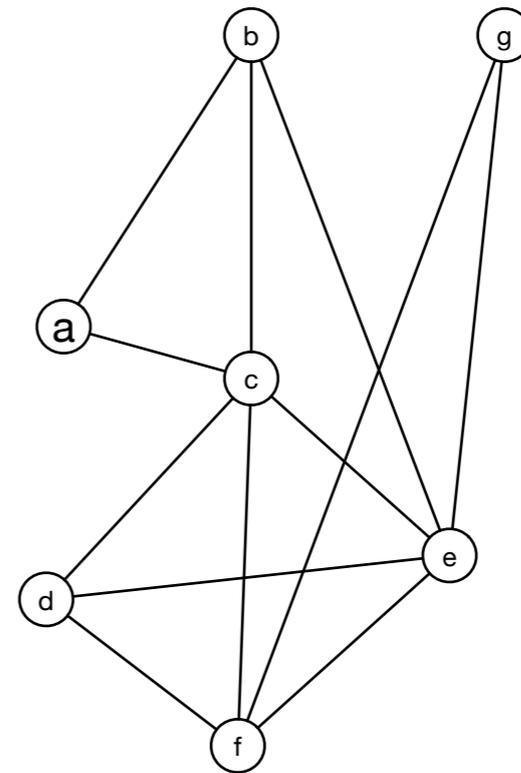


# Graph Definition

- Different visualizations can still give you the same graph, as you can see from the examples below

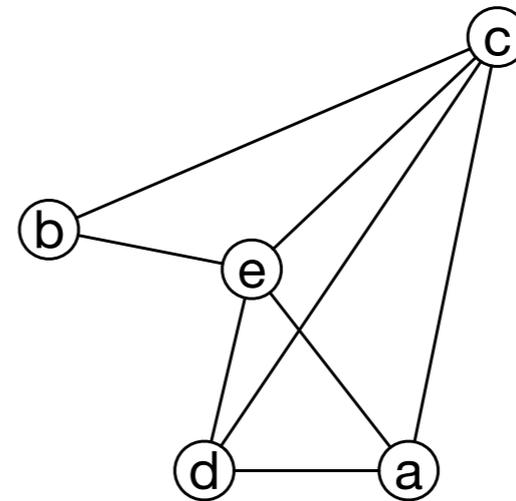
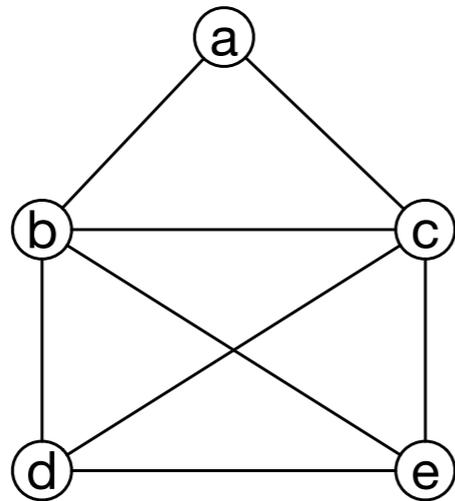


ab, ac, bc, ce  
cd, cf, de, df  
ef, eg, fg



# Graph Definition

- Two graphs are isomorphic, if there is a renaming of the vertices that converts one into the other and vice versa
- Mathematically, a renaming is a bijection



- These two do not look the same, but they are isomorphic:  $a \rightarrow b, b \rightarrow c, c \rightarrow e, d \rightarrow d, e \rightarrow a$

# Graph Definition

- Two graphs are isomorphic, if there is a renaming of the vertices that converts one into the other and vice versa

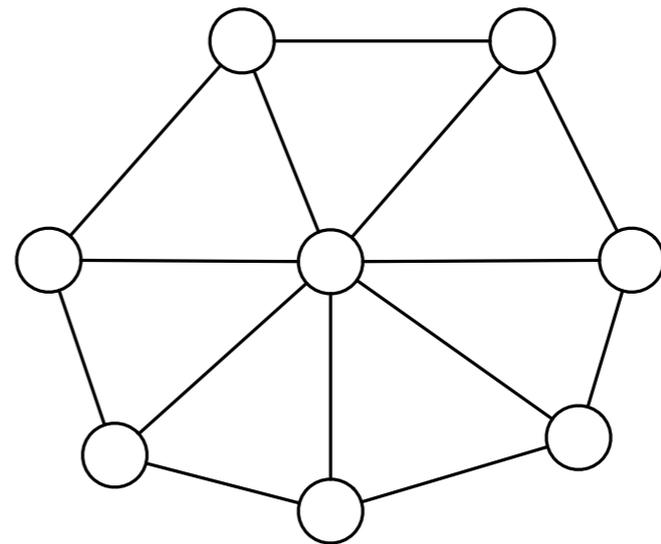
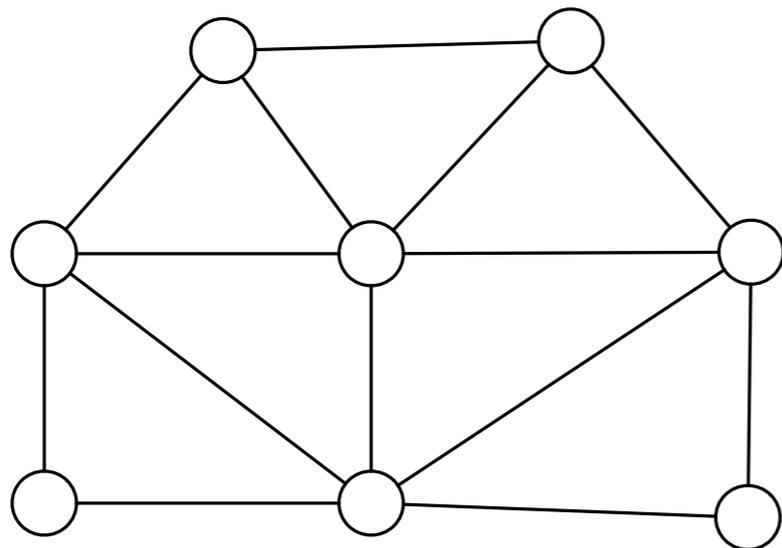
$G = (V, E)$  is isomorphic to  $G' = (V', E')$



$\exists f : V \rightarrow V'$  bijection :  $\forall v_1, v_2 \in V : (f(v_1), f(v_2)) \in E' \Leftrightarrow (v_1, v_2) \in E$

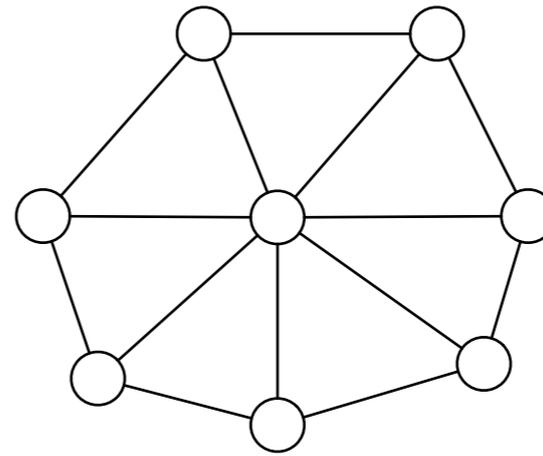
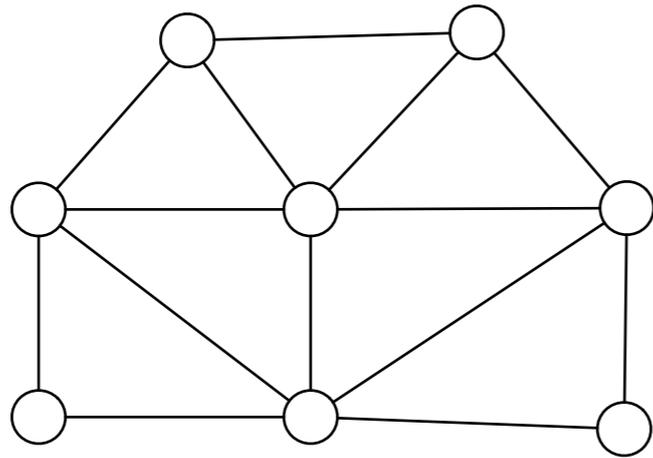
# Graph Definition

- Determining whether two graphs are isomorphic is a known, difficult question
- Some results are easy, e.g. vertices of the same *rank* (the number of edges adjacent to a vertex) need to be mapped to vertices of the same rank
- So, these two graphs **cannot** be isomorphic



# Graph Definition

- These two graphs *cannot* be isomorphic



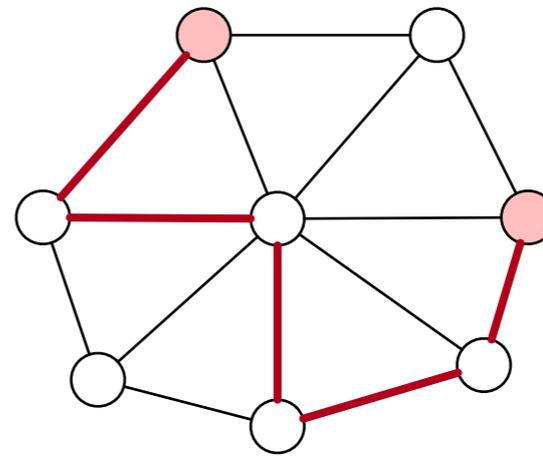
- The left graph has two vertices of degree 2
- The right graph has no vertices of degree 2
- But the number of vertices and edges is equal

# Graph Definitions

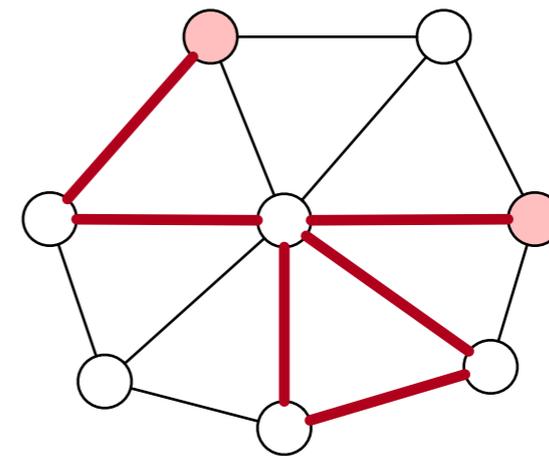
- There are a number of important properties of graphs
  - No need to learn them by heart, the ones used in CS will get repeated over and over again
    - A path between two vertices  $u, w \in V$  of a graph  $G = (V, E)$  is a list of vertices  $u = v_0, v_1, \dots, v_{n-1}, v_n = w$  such that there is an edge between all  $v_i$  and  $v_{i+1}$
    - Furthermore, no vertices can be repeated

# Graph Definitions

- Example for a path:
  - Has length 5 (number of edges)



- Example for a walk that is not a path
  - We visit the center vertex twice

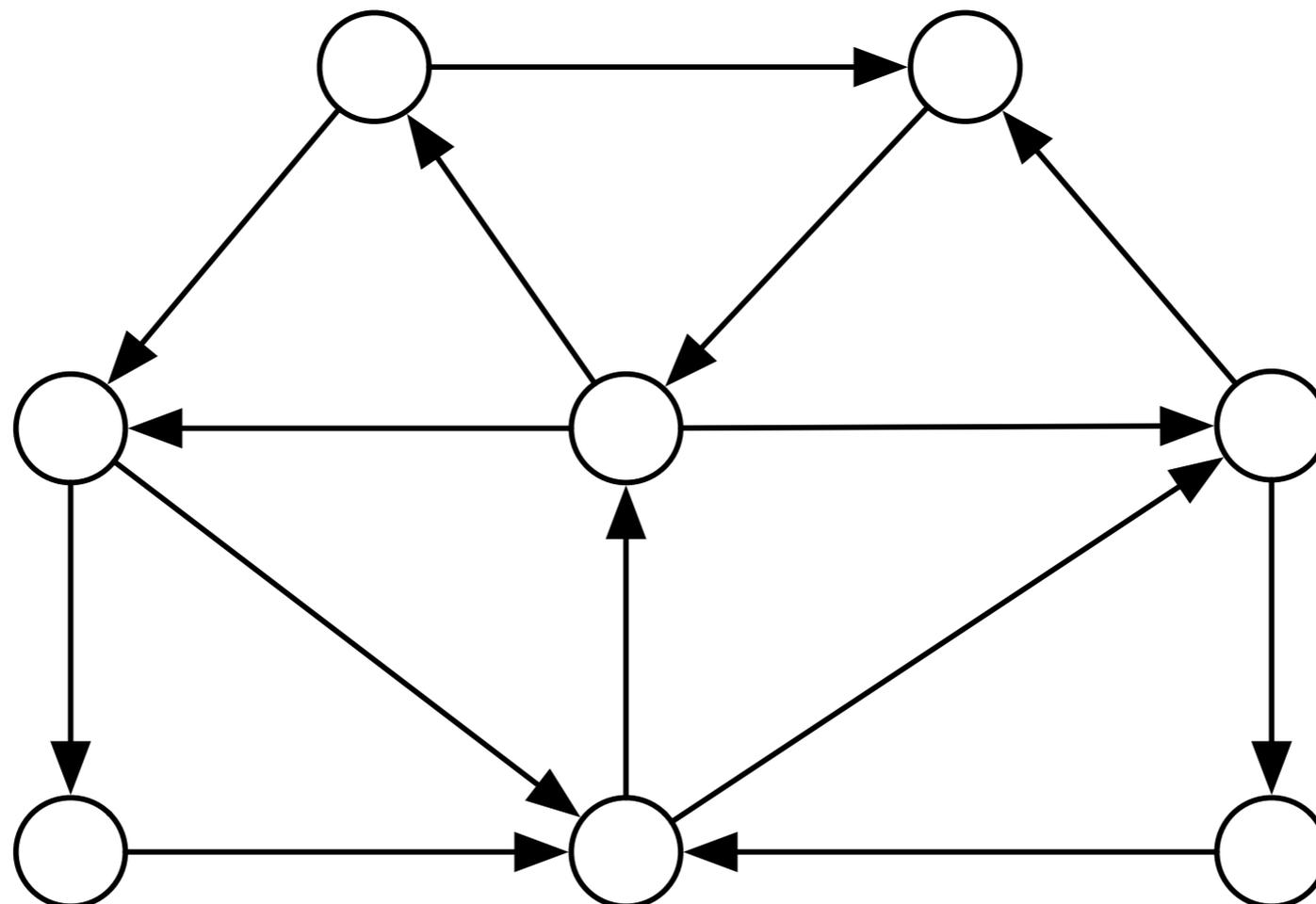


# Graph Definitions

- For directed graphs, the paths need to follow the arrow

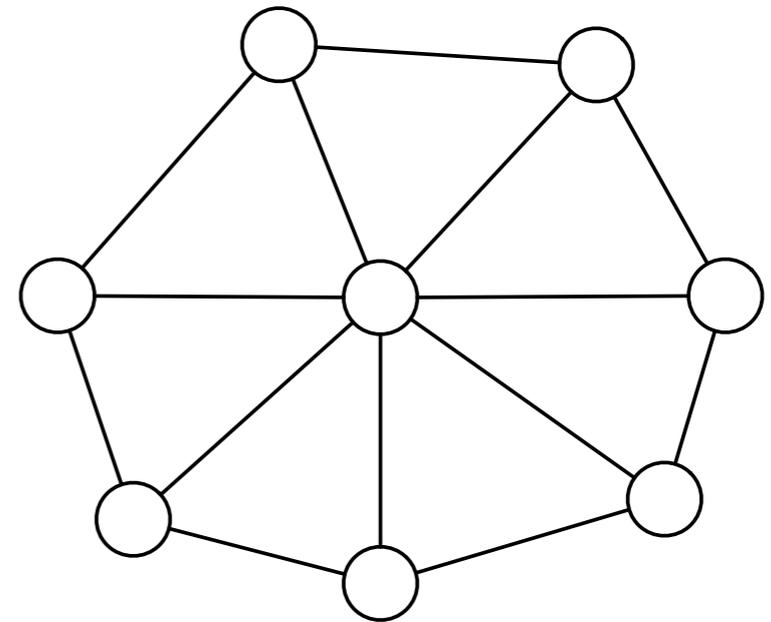
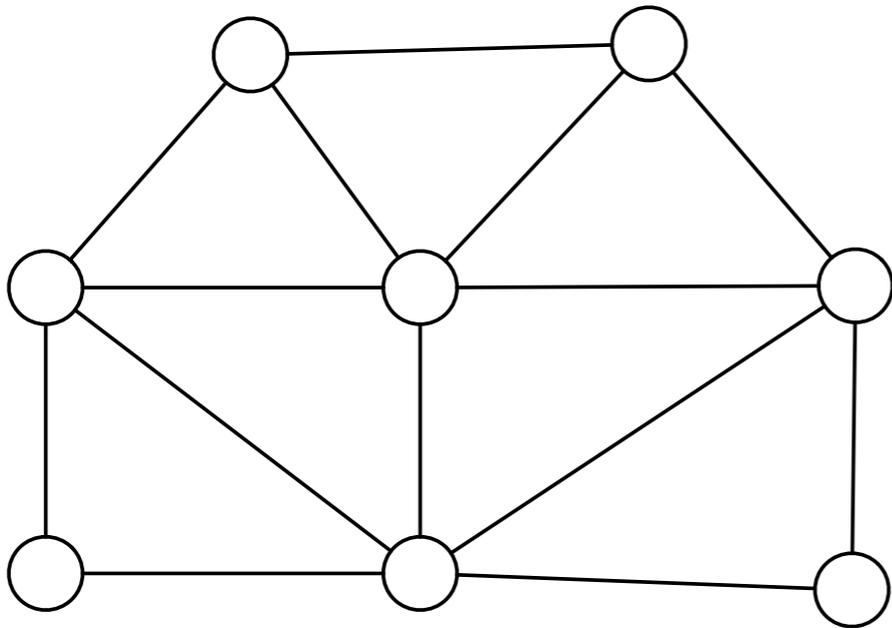
# Graph Definitions

- A directed graph (digraph) is strongly connected if there is a path from every vertex to every other vertex



# Graph Definitions

- An undirected graph is connected if there is a path from every vertex to every other vertex
  - This is not a connected graph



- But it consists of two connected components

# Graph Definitions

- Interesting question
  - Is the friends graph on facebook connected
    - The "friend" relation is mutual, so all users are vertices and there is an edge if two users are in a friends-relation
  - Probably not, because we signed up my mom on facebook and she did not like it, so she is no longer friends with anyone
  - But how about "active users"
    - Could there be a republican and a democratic facebook
      - No, but maybe there are isolated groups

# Euler Tours

- An Euler tour is a closed tour that traverses each edge of the graph only once.
  - Graphs with an Euler tour are called Eulerian
- Theorem: An undirected, connected graph is Eulerian if each vertex has even degree.
  - Recall: Degree is the number of edges of the vertex



# Euler Tours

- Solved by Euler
  - Translate into a multi-graph (multiple edges allowed)

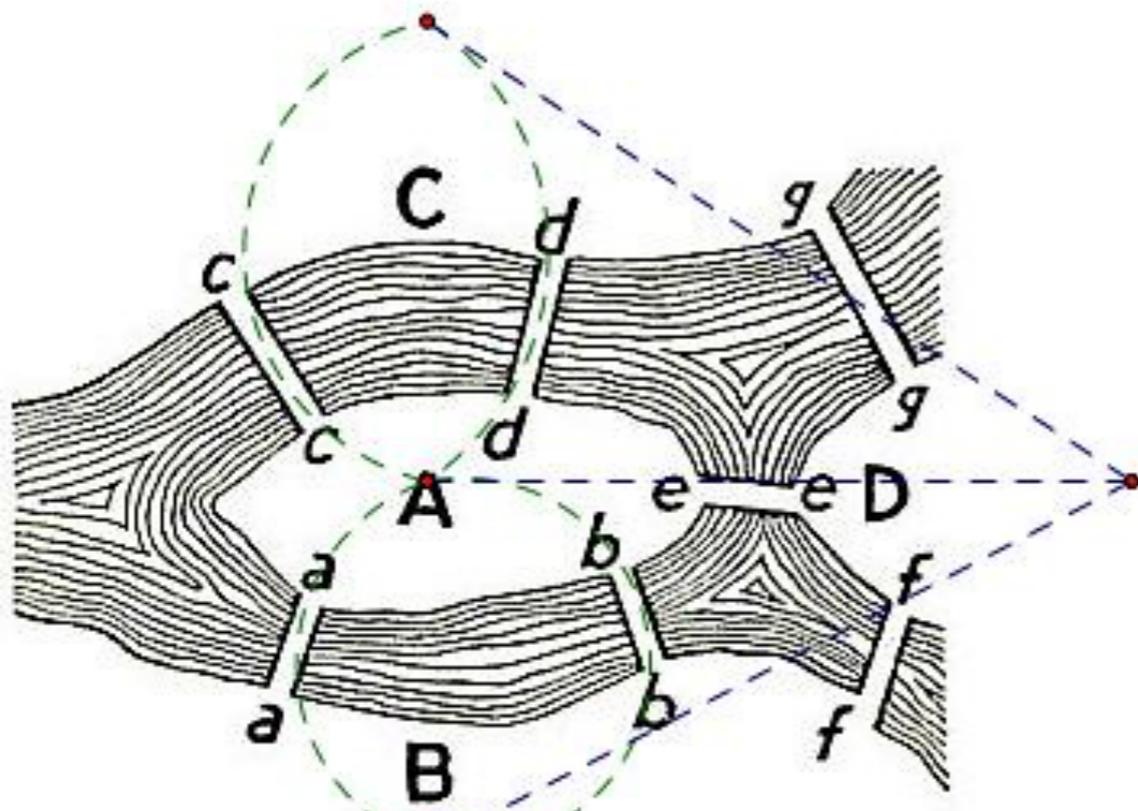
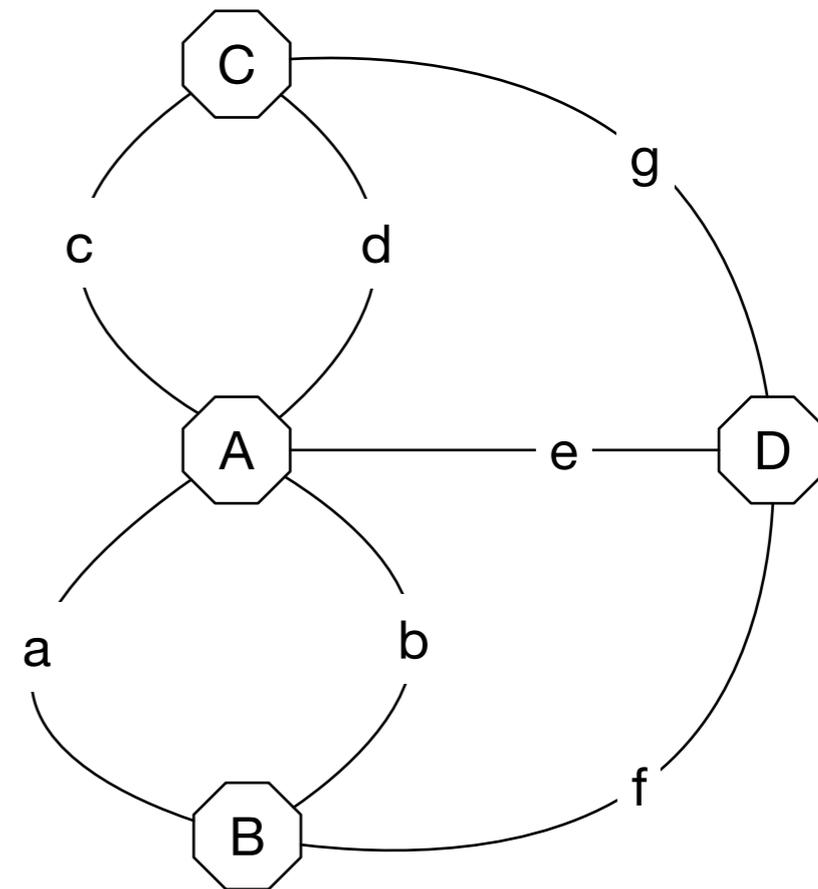
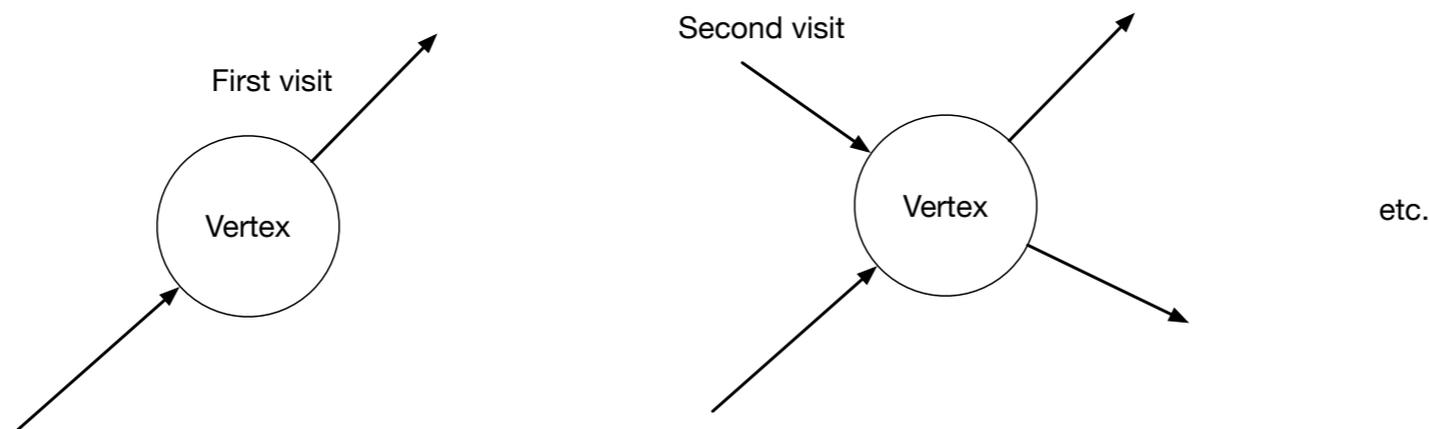


FIGURE 98. *Geographic Map:  
The Königsberg Bridges.*



# Euler Tours

- Actually, all edges have odd degree, so such a tour is not possible
- To show that the theorem is correct:
  - Euler tour exists implies all vertex degrees are even
    - Because an Euler tour visits all edges and every time it visits an edge, it needs to come and to go.



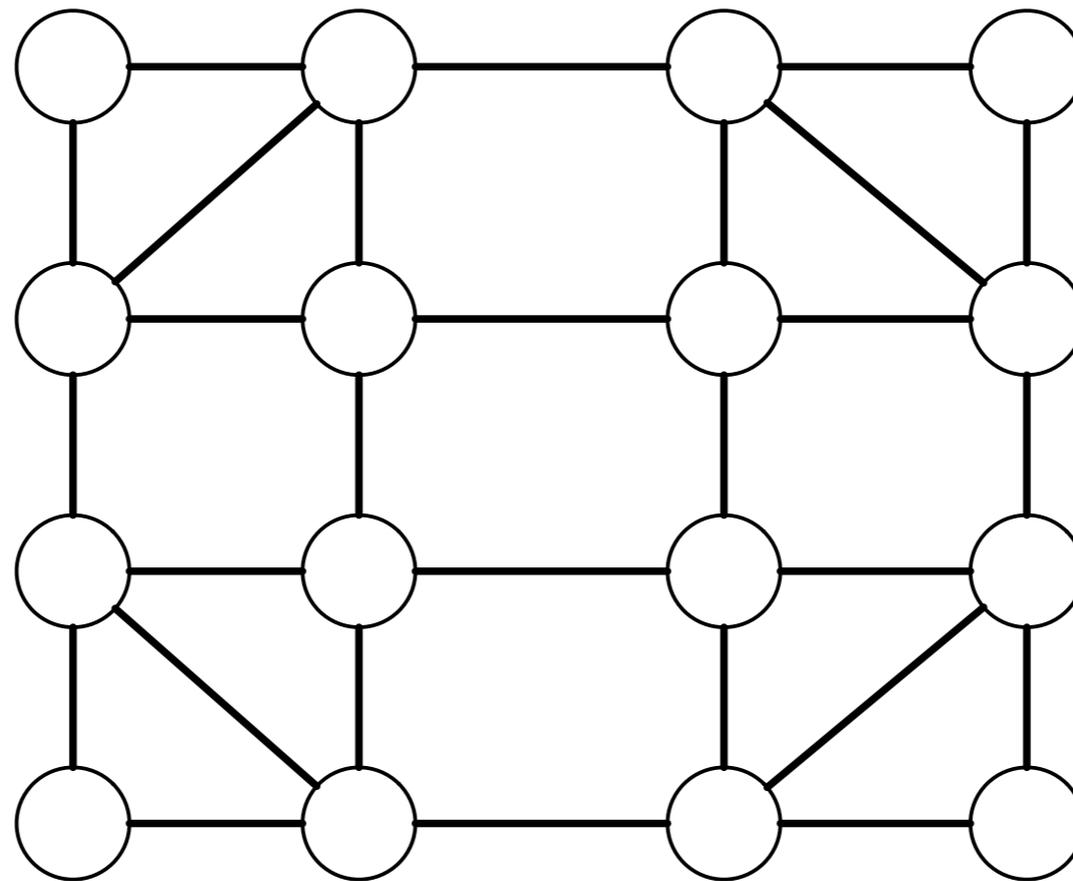


# Euler Tours

- Fleury's algorithm:
  - Start at a node and walk anywhere, marking the edge
  - Leave the node that you arrived at
  - Continue until you can no longer find an unused edge
    - At this point, you are back in the starting vertex
  - If any of the vertices visited has a unused edges, start with that edge until you are back at that edge.
  - Splice the new circuit into the old one

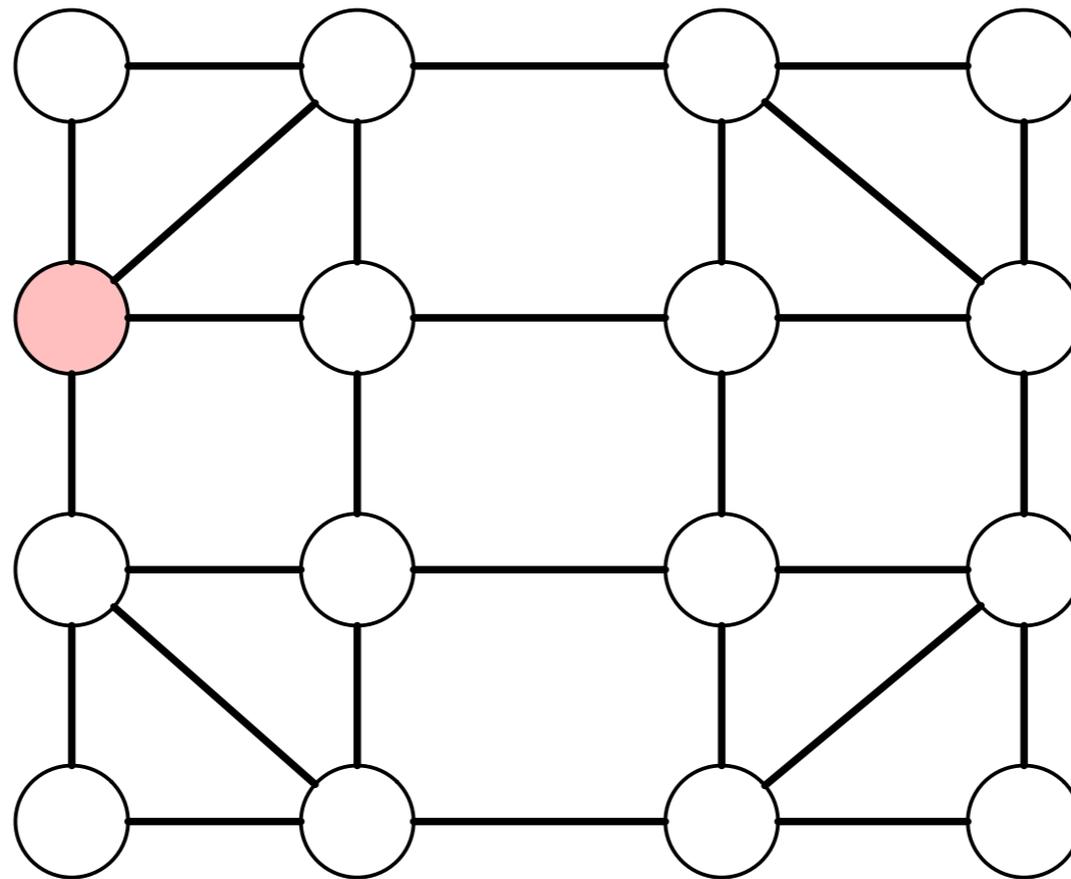
# Euler Tours

- Example



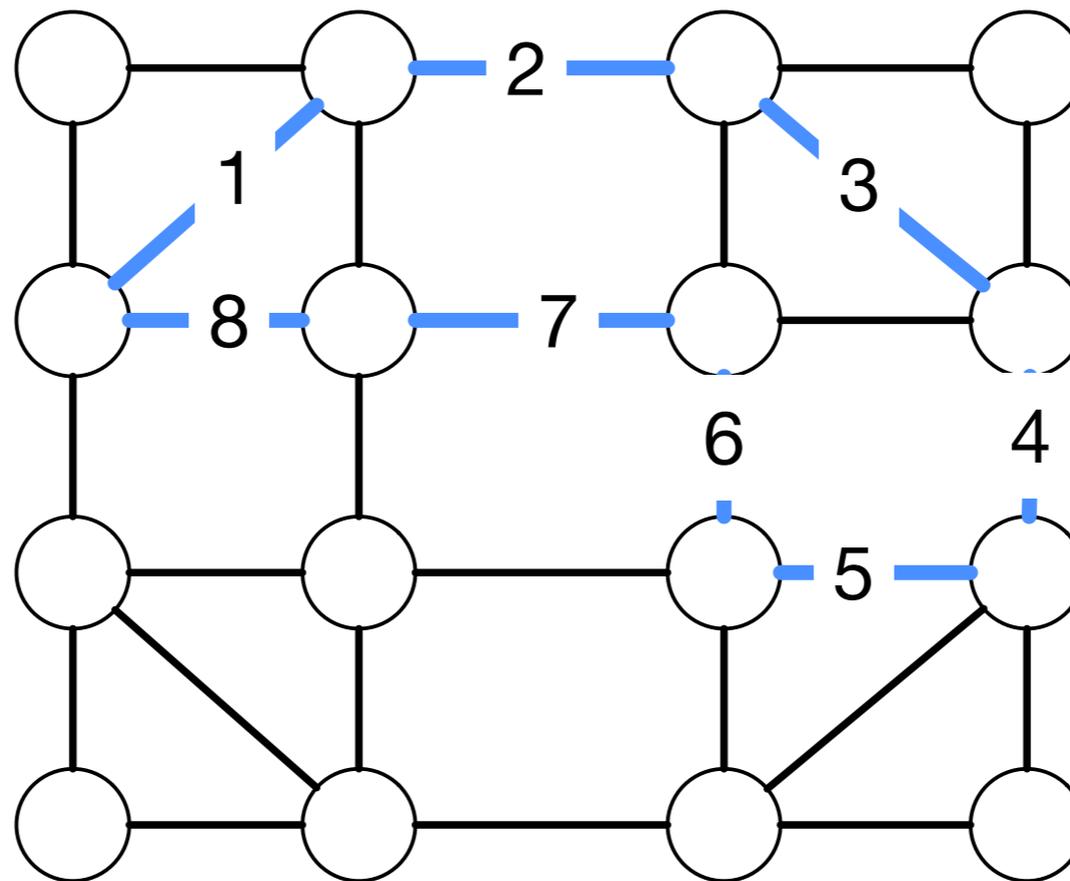
# Euler Tours

- Start at a random vertex



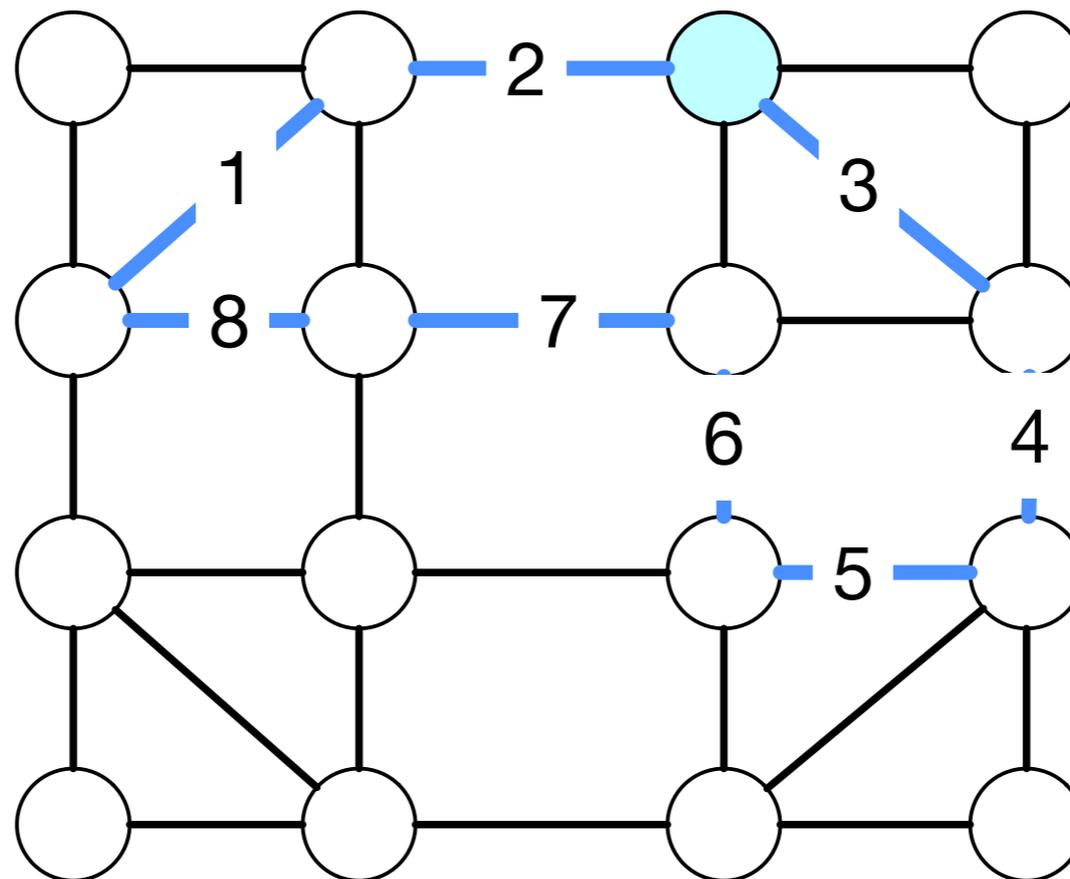
# Euler Tours

- Make a tour



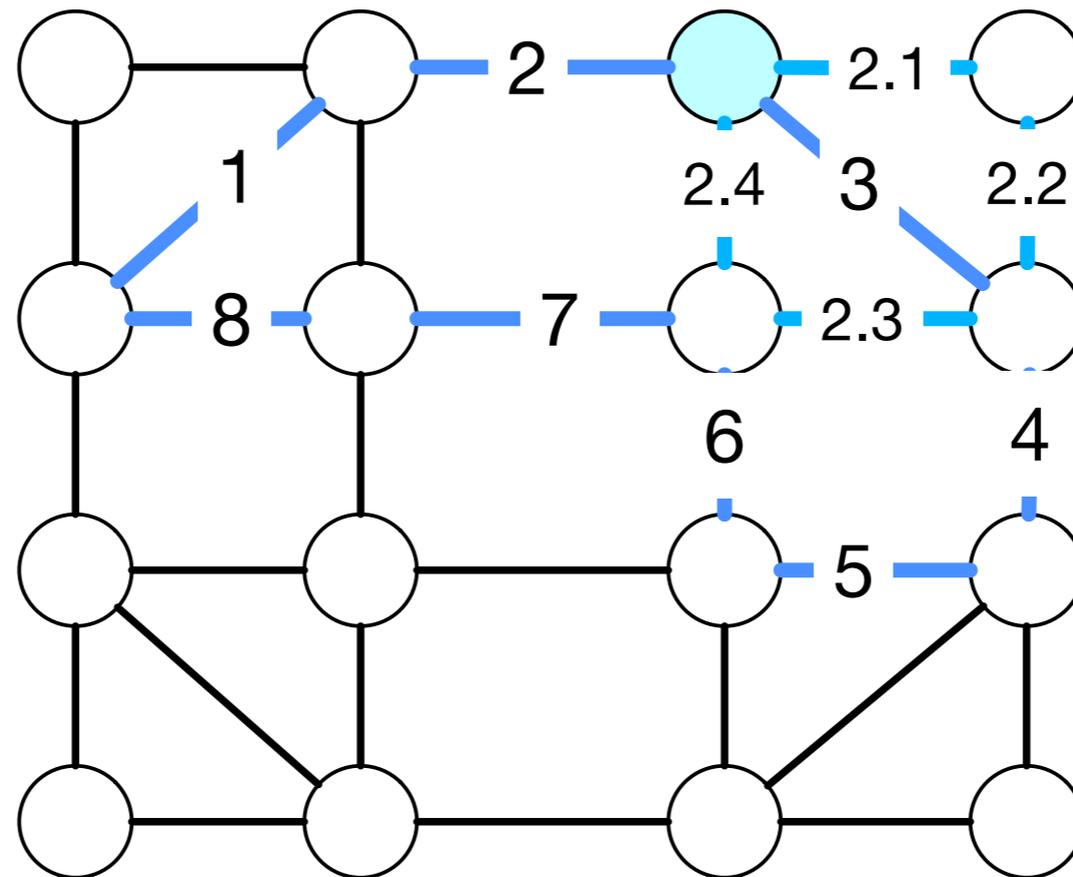
# Euler Tours

- Check for vertices with unused edges and pick a random one



# Euler Tours

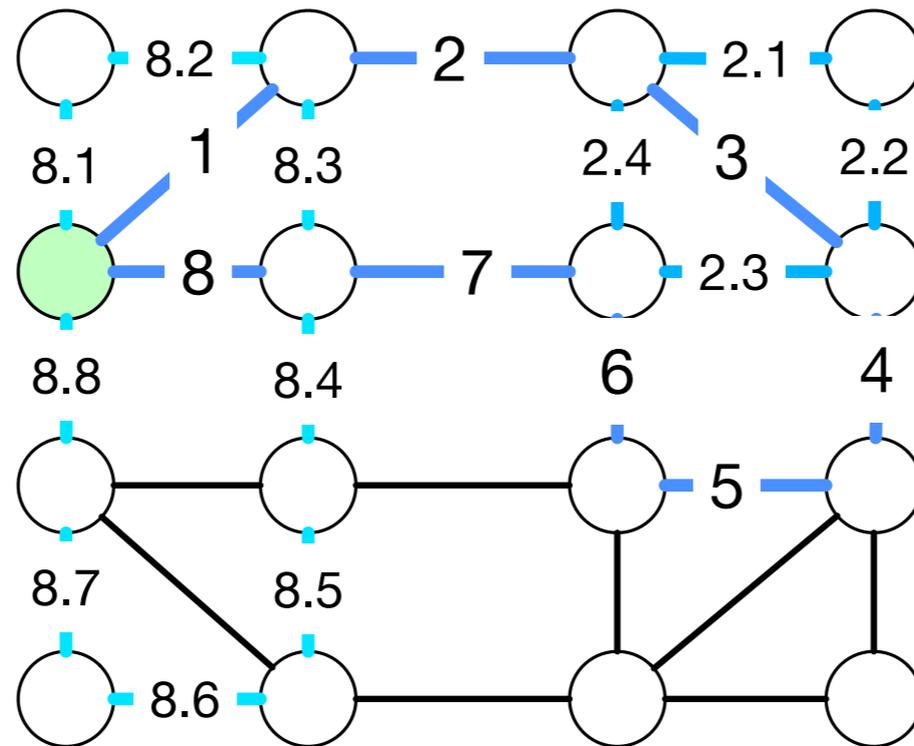
- Start out creating a random circuit of unused edges





# Euler Tours

- Start a new part of the circuit

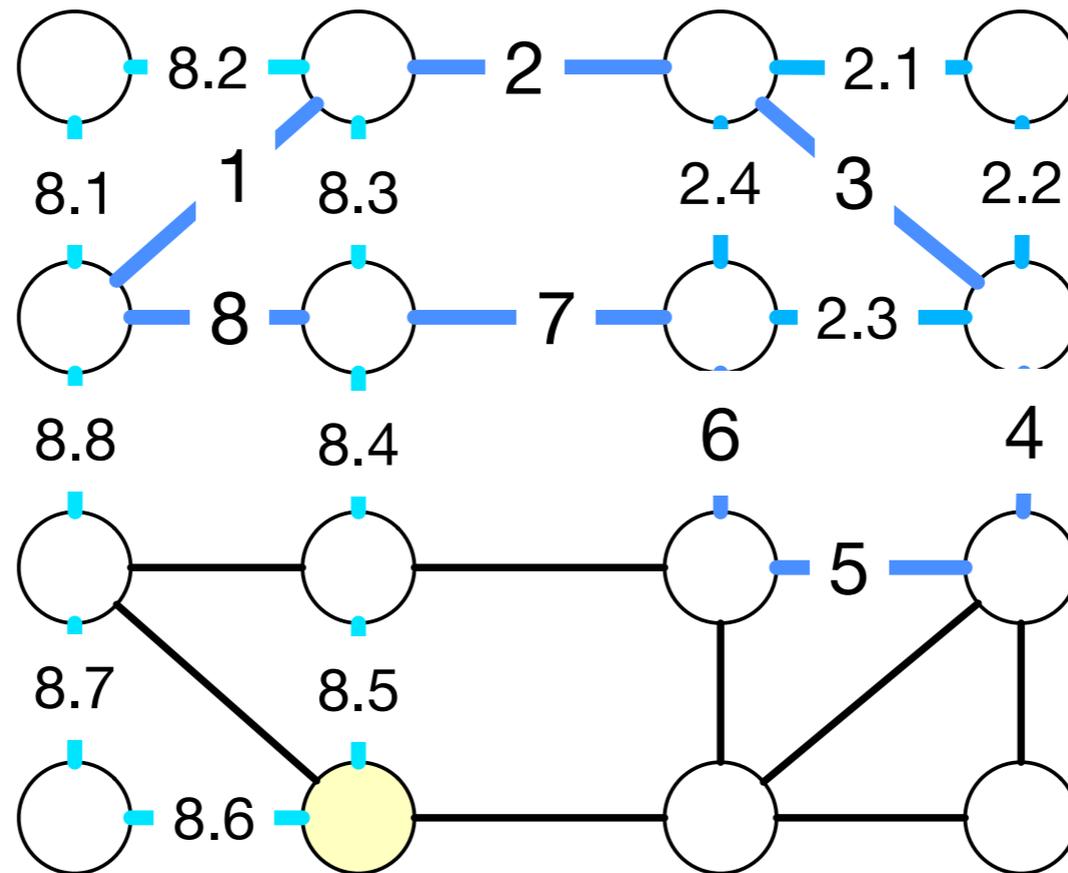


- Circuit so far: 1, 2, 2.1, 2.2, 2.3, 2.4, 3, 4, 5, 6, 7, 8, 8.1, 8.2, 8.3, 8.4, 8.5, 8.6, 8.7, 8.8

# Euler Tours

- In the new circuit, there are still vertices without all edges used.

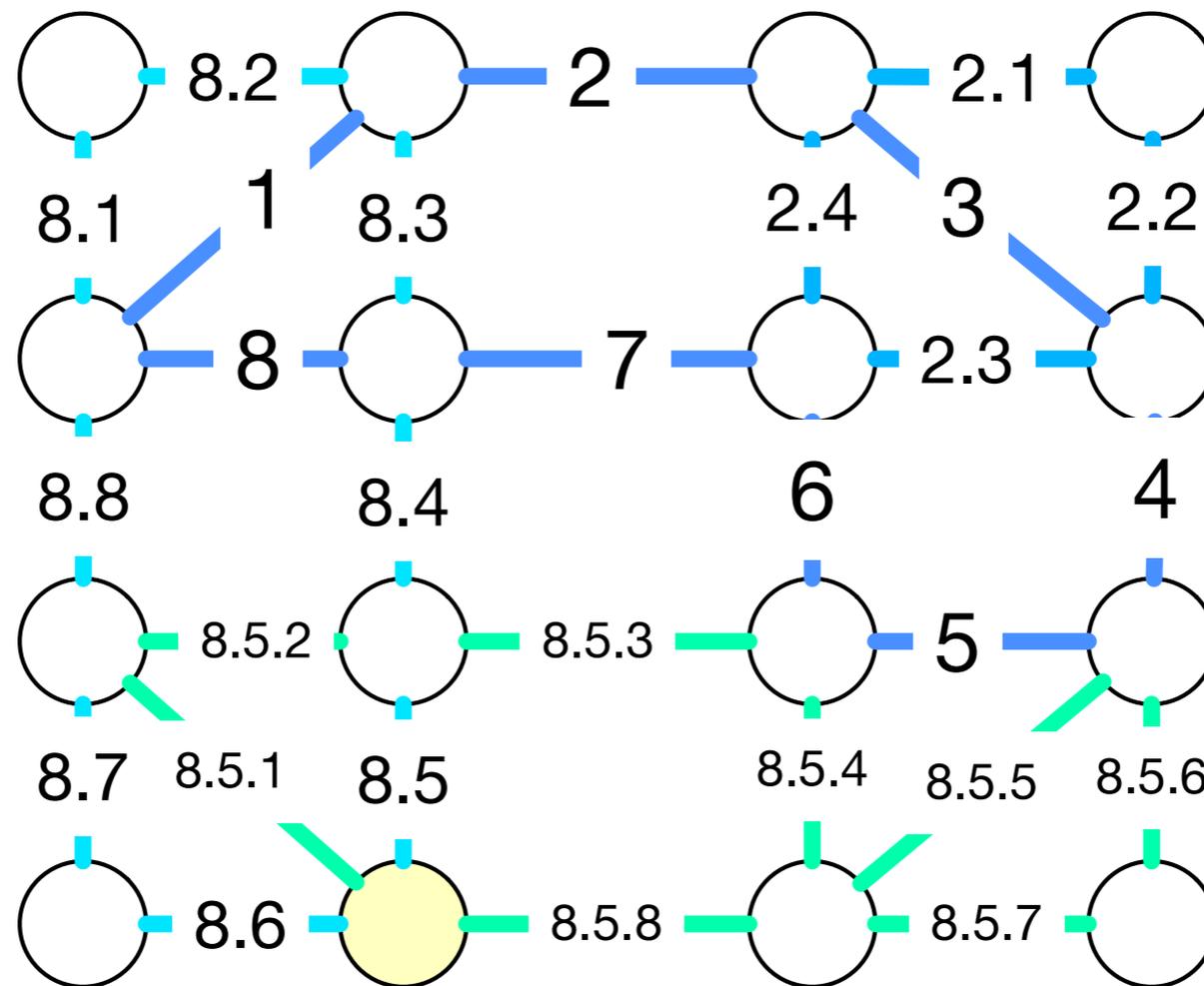
- Pick one



- Circuit so far: 1, 2, 2.1, 2.2, 2.3, 2.4, 3, 4, 5, 6, 7, 8, 8.1, 8.2, 8.3, 8.4, 8.5, 8.6, 8.7, 8.8

# Euler Tours

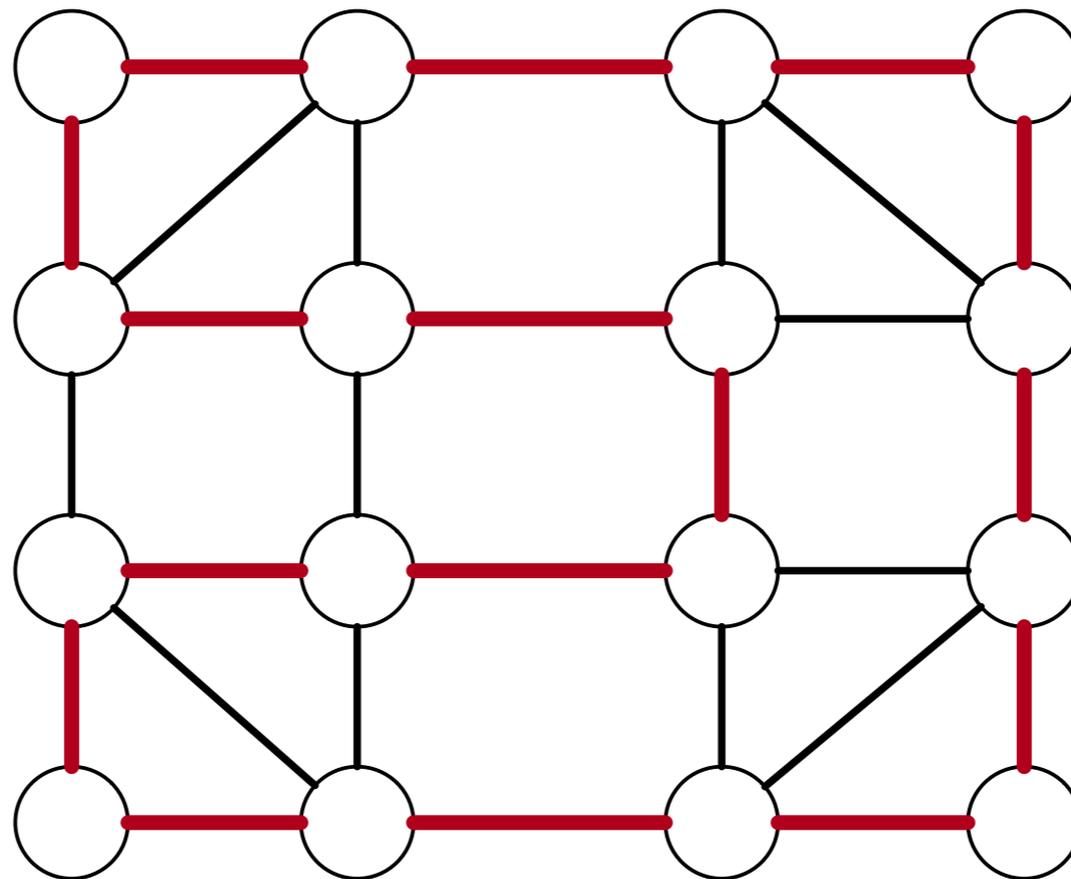
- And after this, we are done



- Circuit is: 1, 2, 2.1, 2.2, 2.3, 2.4, 3, 4, 5, 6, 7, 8, 8.1, 8.2, 8.3, 8.4, 8.5, 8.5.1, 8.5.2, 8.5.3, 8.5.4, 8.5.5, 8.5.6, 8.5.7, 8.5.8, 8.6, 8.7, 8.8

# Hamiltonian Circuit

- Similar question: Is there a circuit that goes through all vertices



# Hamiltonian Circuit

- Turns out to be very difficult
  - Can be shown to not be decidable with a polynomial time algorithm

# Graph Definitions

- Distance in a graph:
  - Length of the shortest path between two vertices

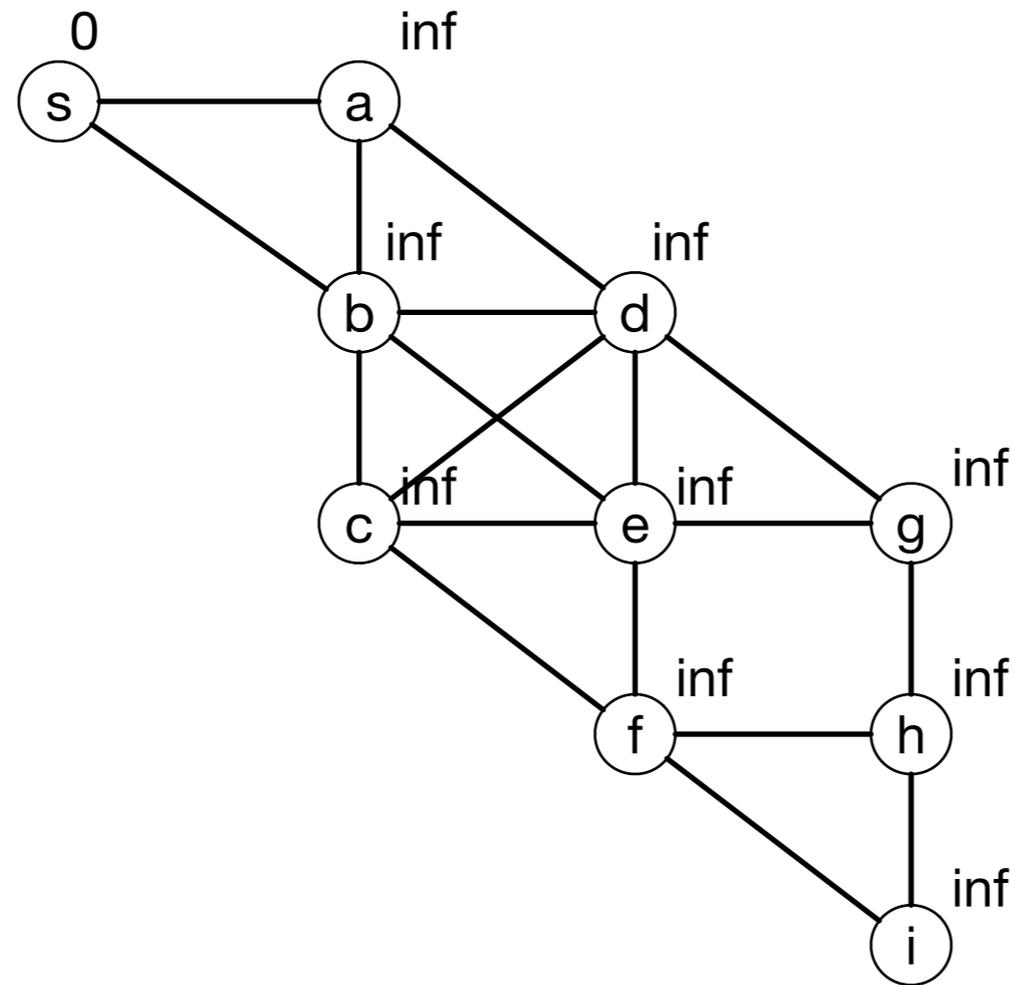
$$\delta(u, w) = \min\{n \mid \exists v_0 = u, v_1, \dots, v_n = w \text{ such that } (v_i, v_{i+1}) \in E \forall i \in \{0, \dots, n-1\}\}$$

# Dijkstra's Algorithm

- Want to determine the distance between a vertex  $s$  and all other vertices in an undirected graph
  - Dynamic programming algorithm
    - Add intermediate vertices one by one
    - Start: Every vertex not  $s$  gets distance infinity
      - $s$  gets distance 0
    - Put all vertices into a priority heap ordered by distance
      - We can quickly extract a vertex with minimum distance

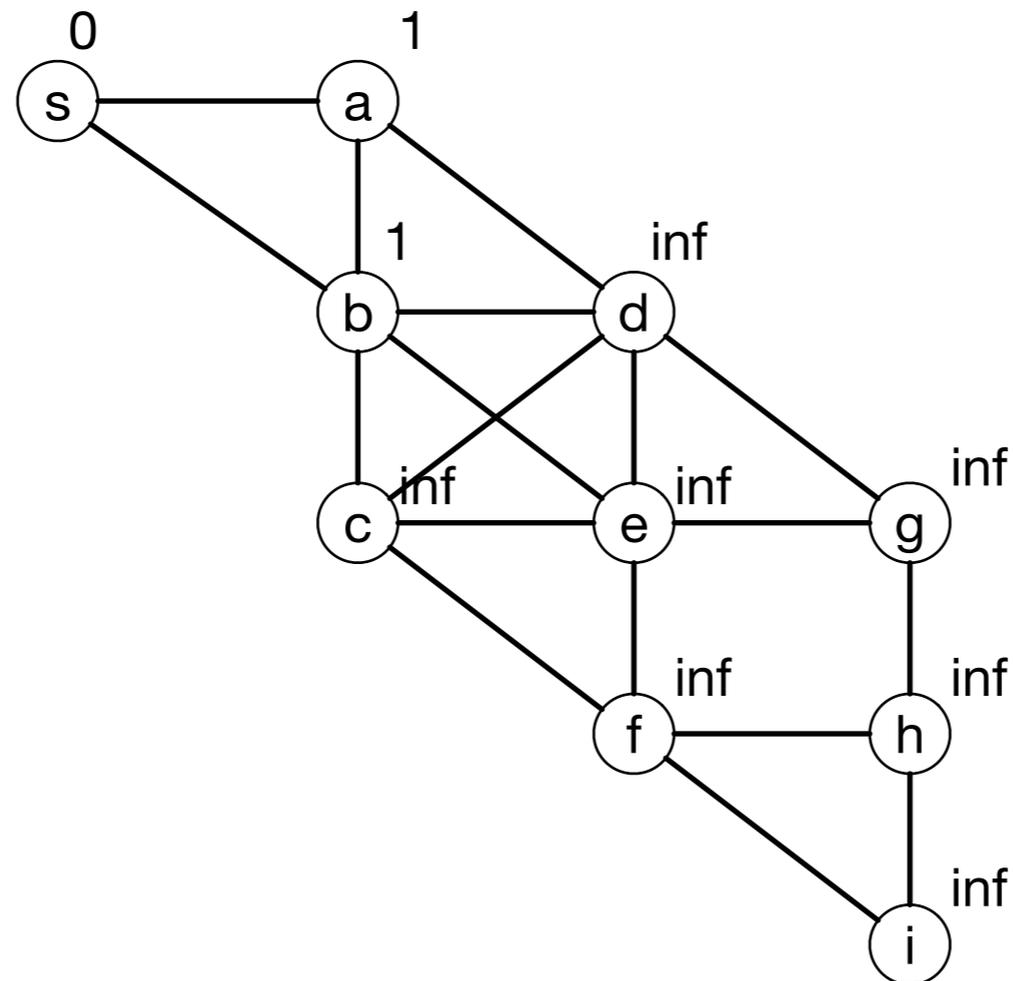
# Dijkstra's Algorithm

- Example:



# Dijkstra's Algorithm

- Update  $s$ :
  - Give all neighbors of  $s$  distance 1

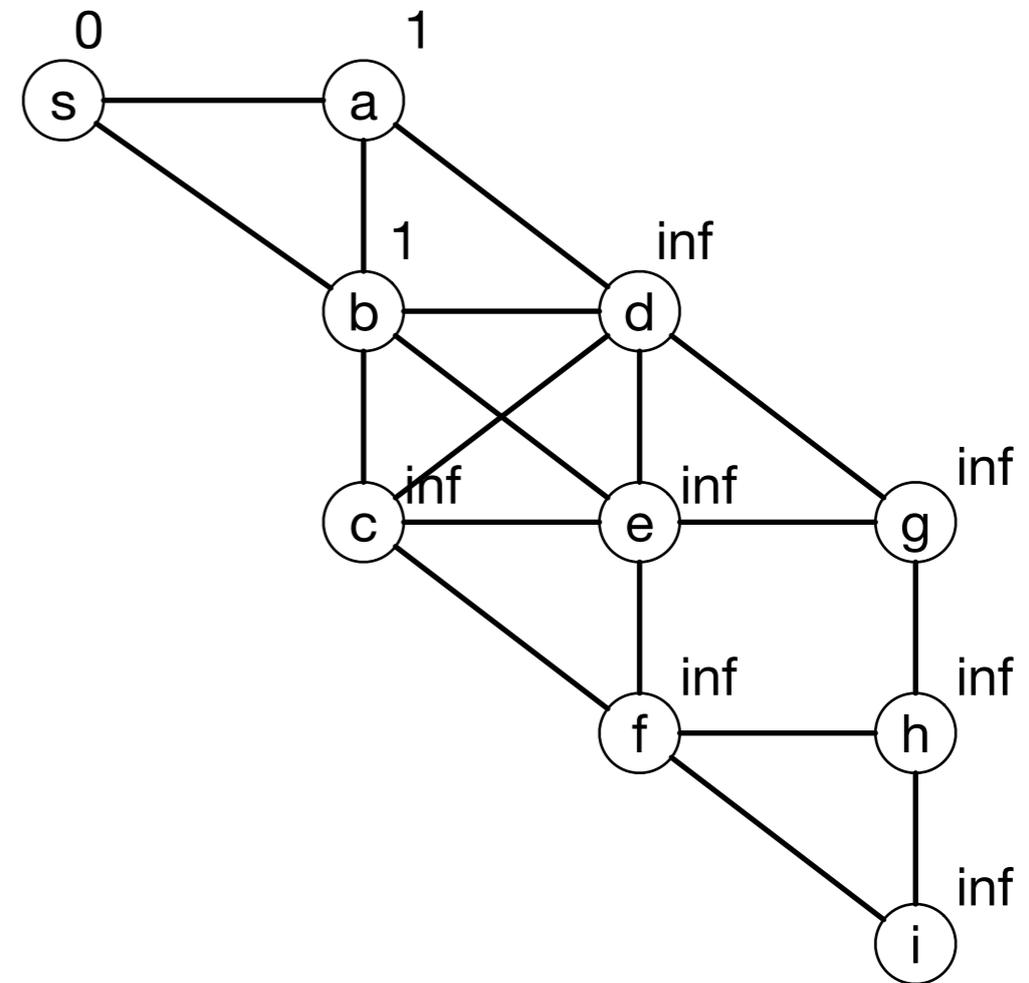


# Dijkstra's Algorithm

- The heap gives us one of  $\{a, b\}$  as a minimum distance node.
  - Pick  $a$ .
  - Update all its neighbors by giving them an updated distance
    - Minimum of current value
    - Value of  $a$  plus 1
  - $a$  is connected to  $b$ ,  $c$ , and  $s$

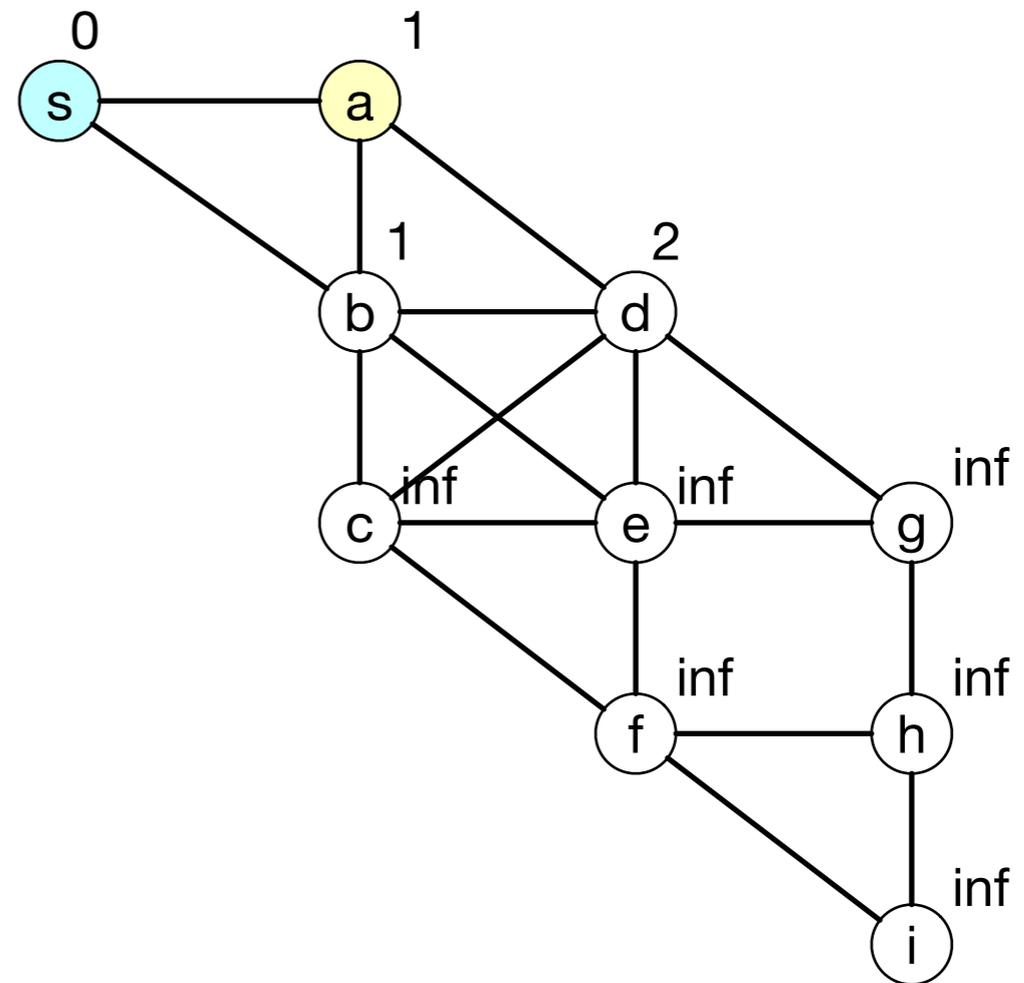
# Dijkstra's Algorithm

- b gets  $\min(1, 1+1)$
- s gets  $\min(0, 1+1)$
- d gets  $\min(\text{inf}, 1+1)$



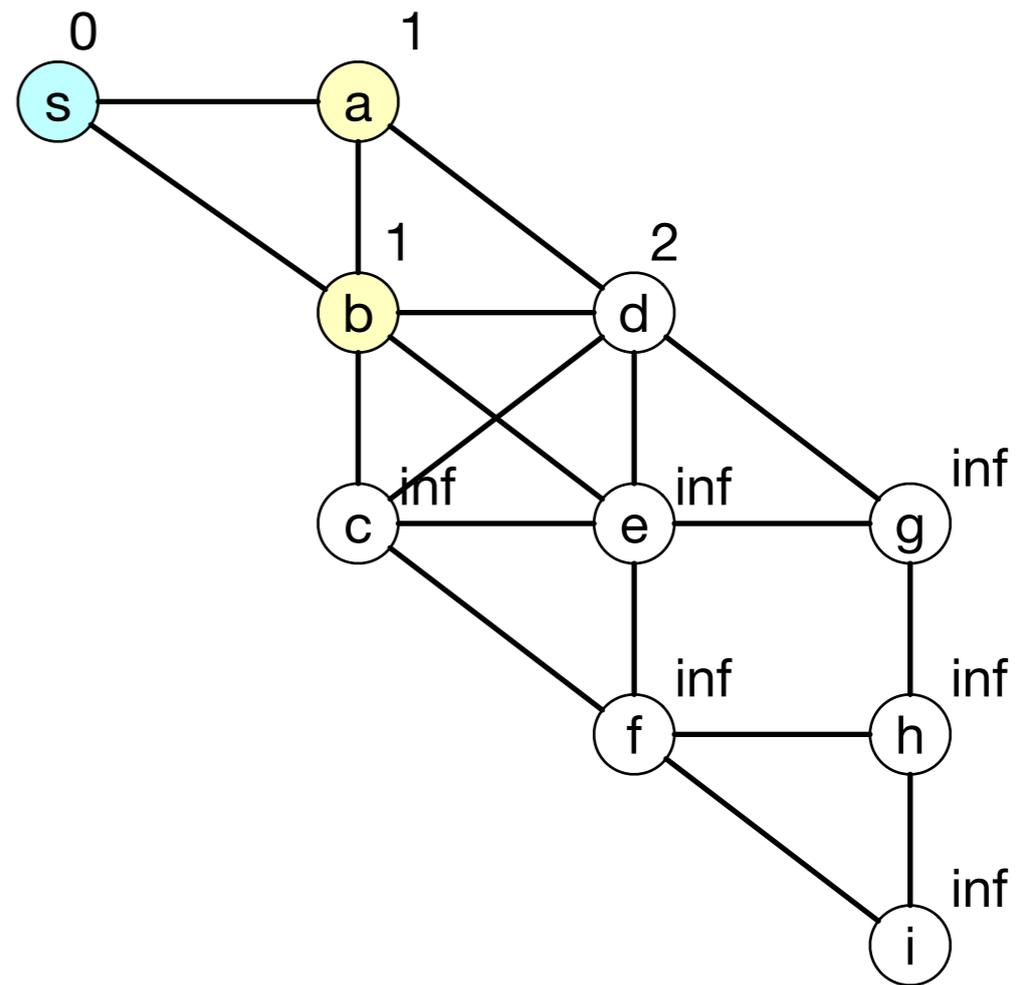
# Dijkstra's Algorithm

- b gets  $\min(1, 1+1)$
  - s gets  $\min(0, 1+1)$
  - d gets  $\min(\text{inf}, 1+1)$
- 
- After update, mark a as used by removing it from the priority queue



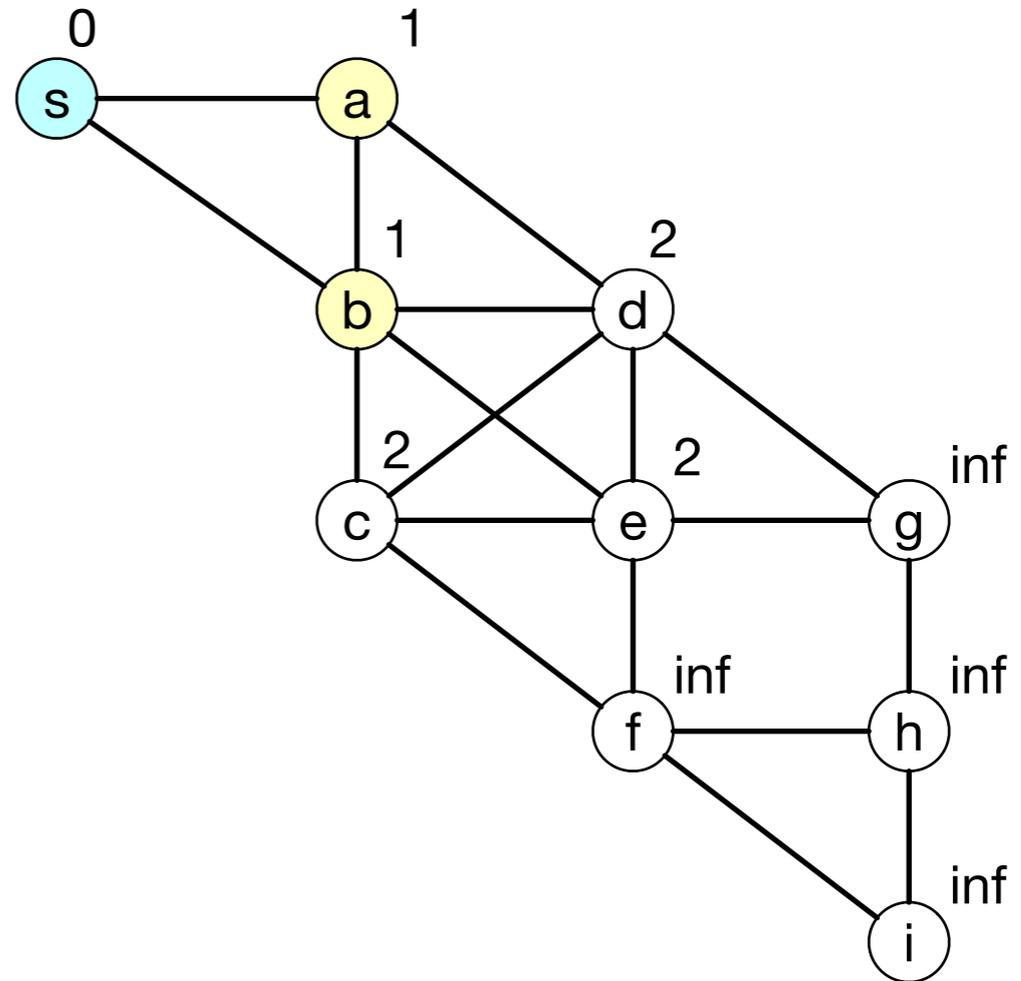
# Dijkstra's Algorithm

- Pick the node with minimum distance that is not marked
- Which would be b
- Update its neighbors



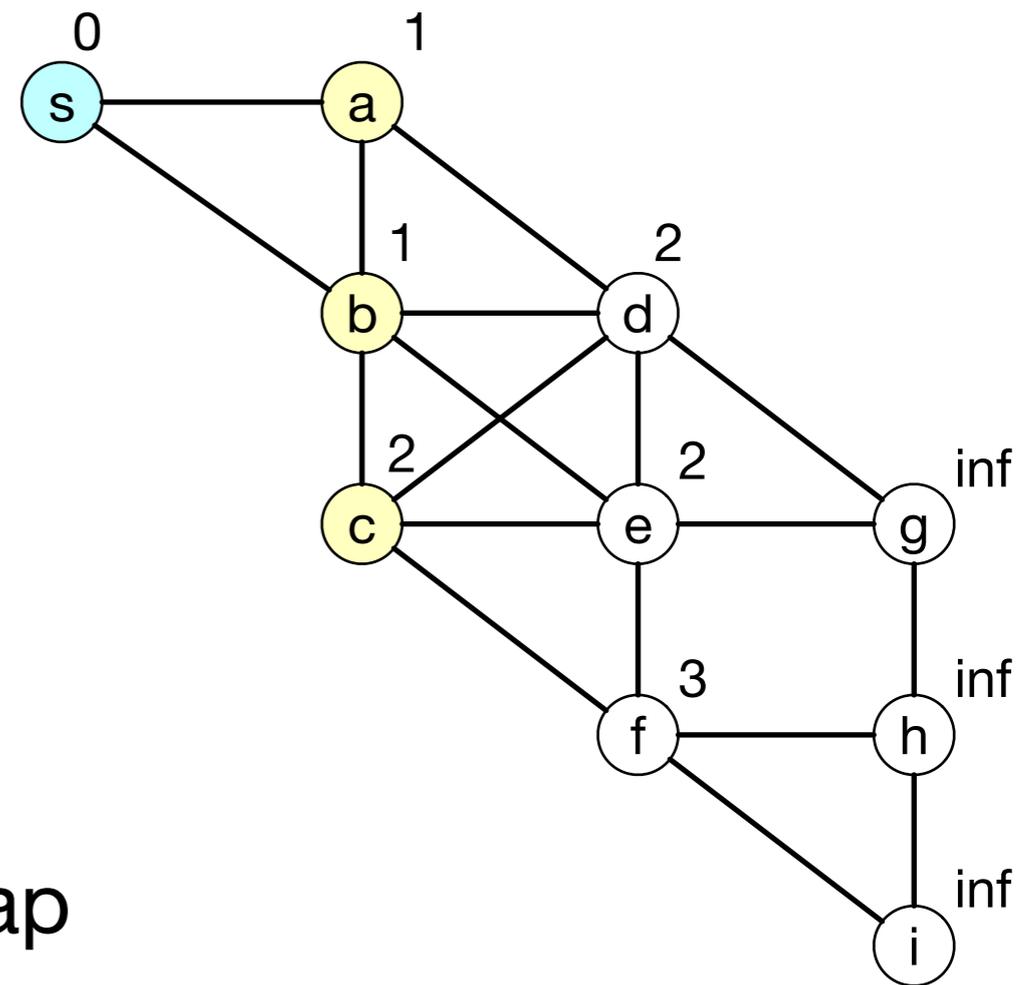
# Dijkstra's Algorithm

- d gets  $\min(2, 1+1)$
- c gets  $\min(\text{inf}, 1+1)$
- e gets  $\min(\text{inf}, 1+1)$
- s gets  $\min(0, 1+1)$



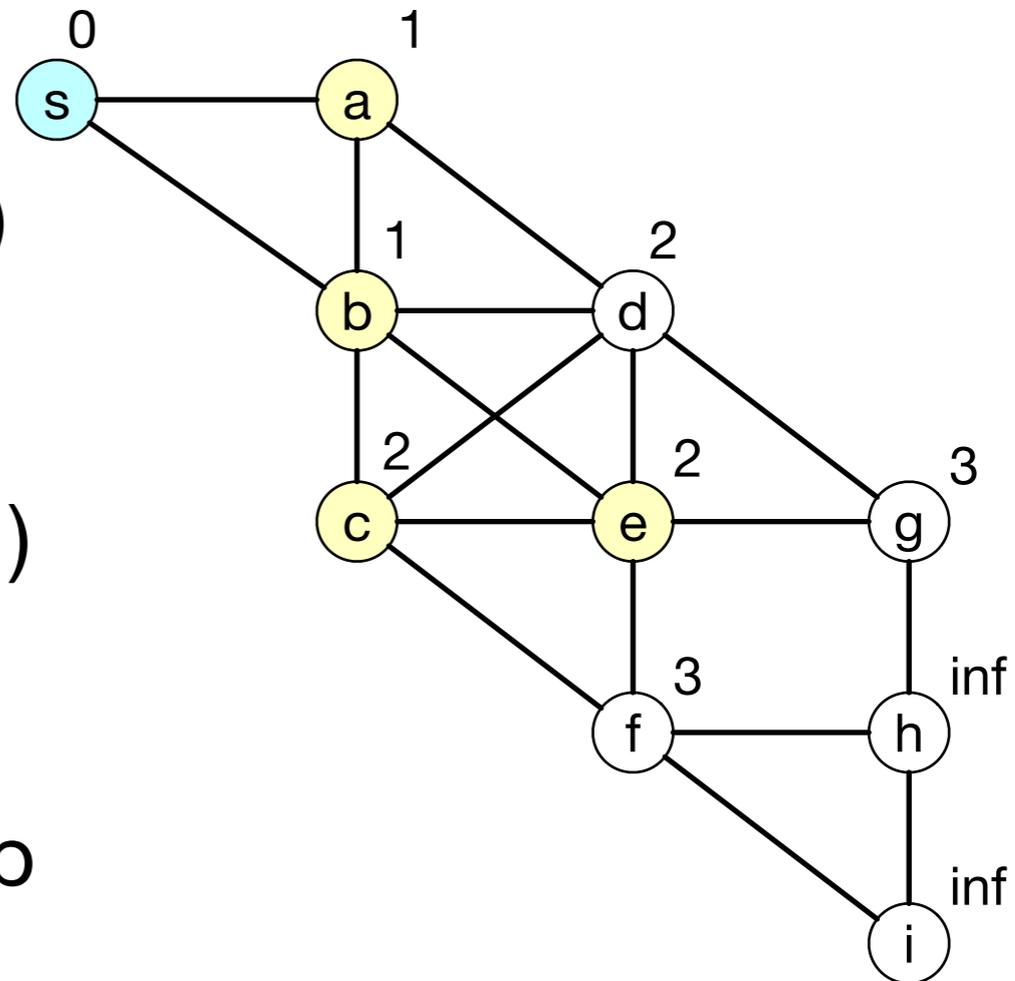
# Dijkstra's Algorithm

- Select one of the vertices with minimum distance:
  - Either c, d, or e
  - Pick c
    - b gets  $\min(1, 2+1)$
    - d gets  $\min(2, 2+1)$
    - e gets  $\min(2, 2+1)$
    - f gets  $\min(\text{inf}, 2+1)$
  - Remove c from the priority heap



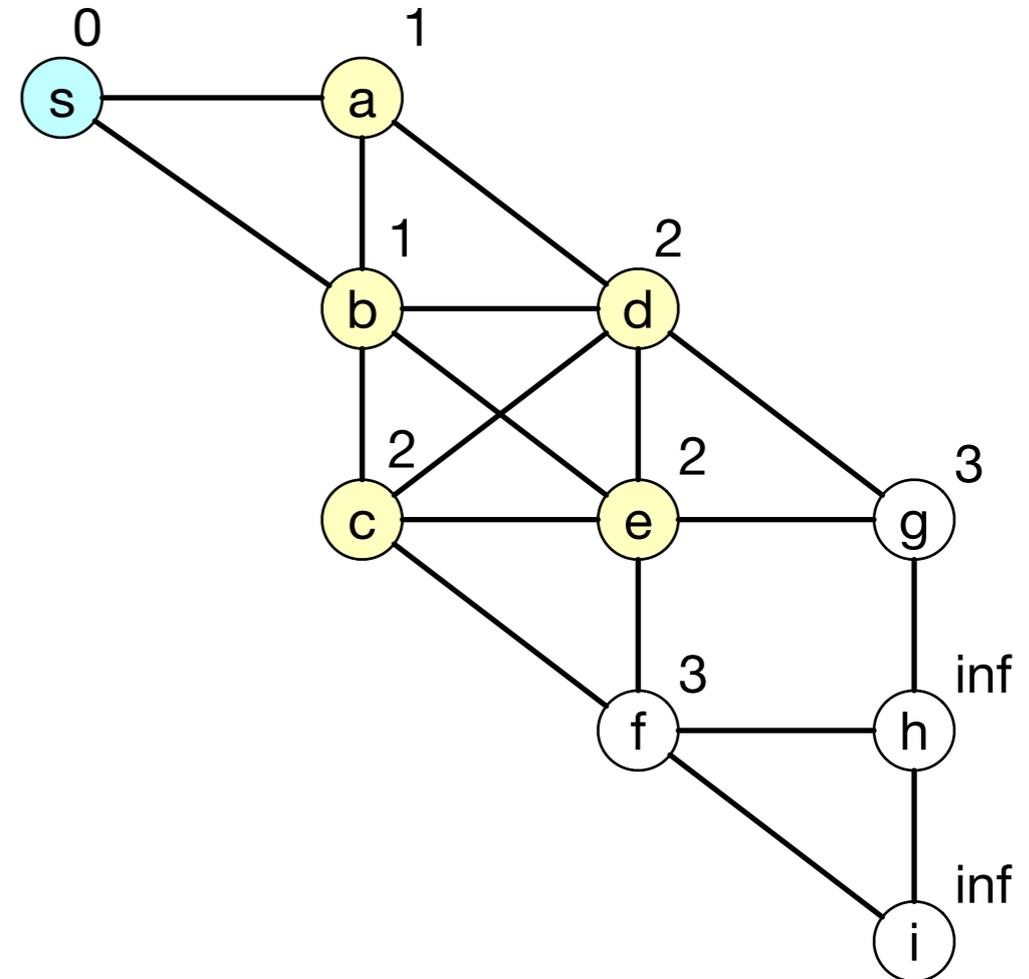
# Dijkstra's Algorithm

- Select e
  - Update b with  $\min(1, 2+1)$
  - Update c with  $\min(2, 2+1)$
  - Update d with  $\min(2, 2+1)$
  - Update f with  $\min(3, 2+1)$
- Remove e from priority heap



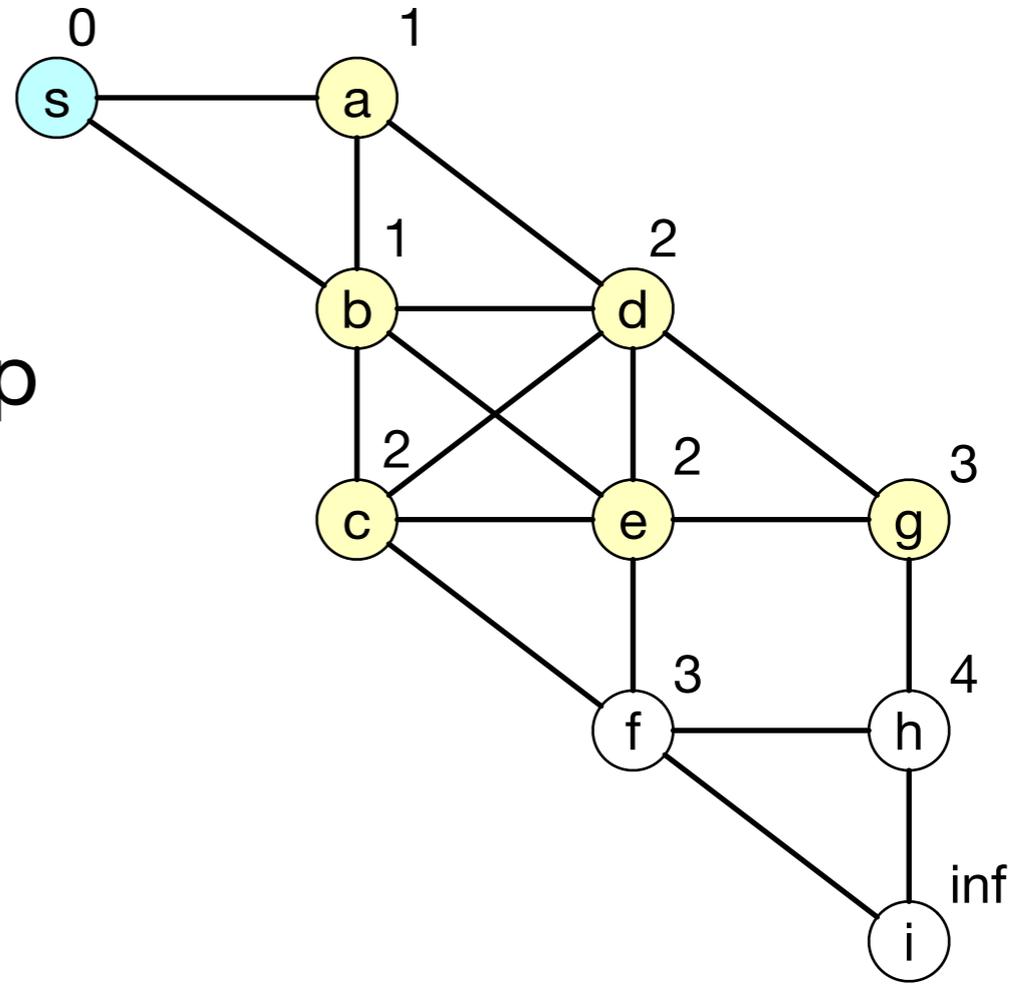
# Dijkstra's Algorithm

- Select d
  - Updates have no effect
- Remove d from heap



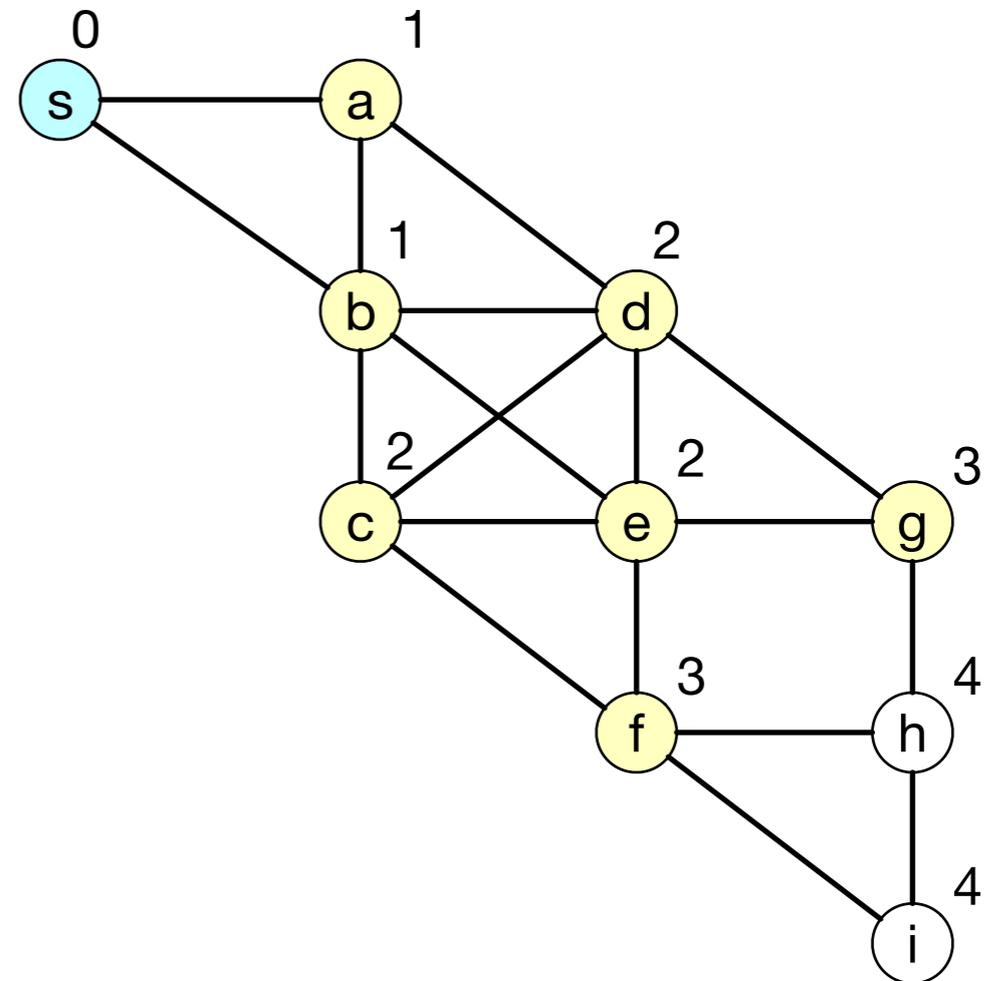
# Dijkstra's Algorithm

- Select g
  - Only change is h gets 4
- Remove g from priority heap



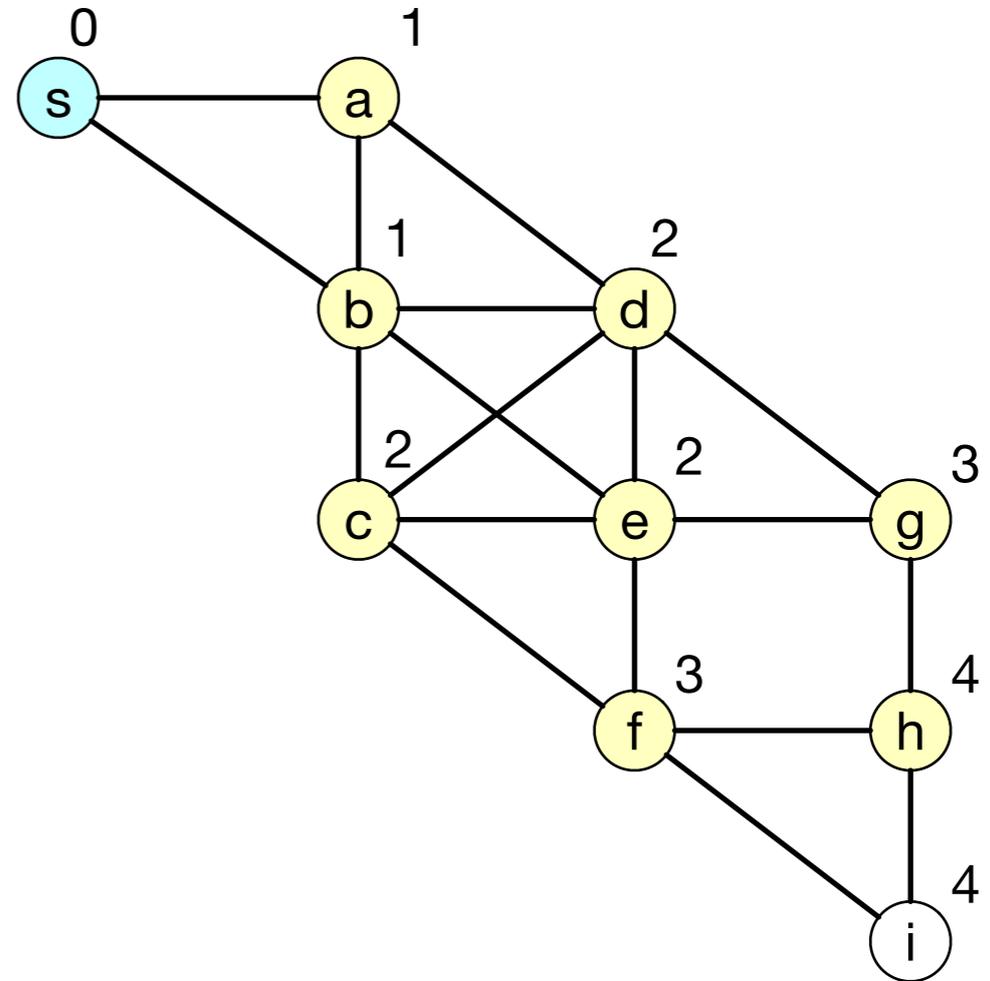
# Dijkstra's Algorithm

- Need to select f
  - Update only changes h



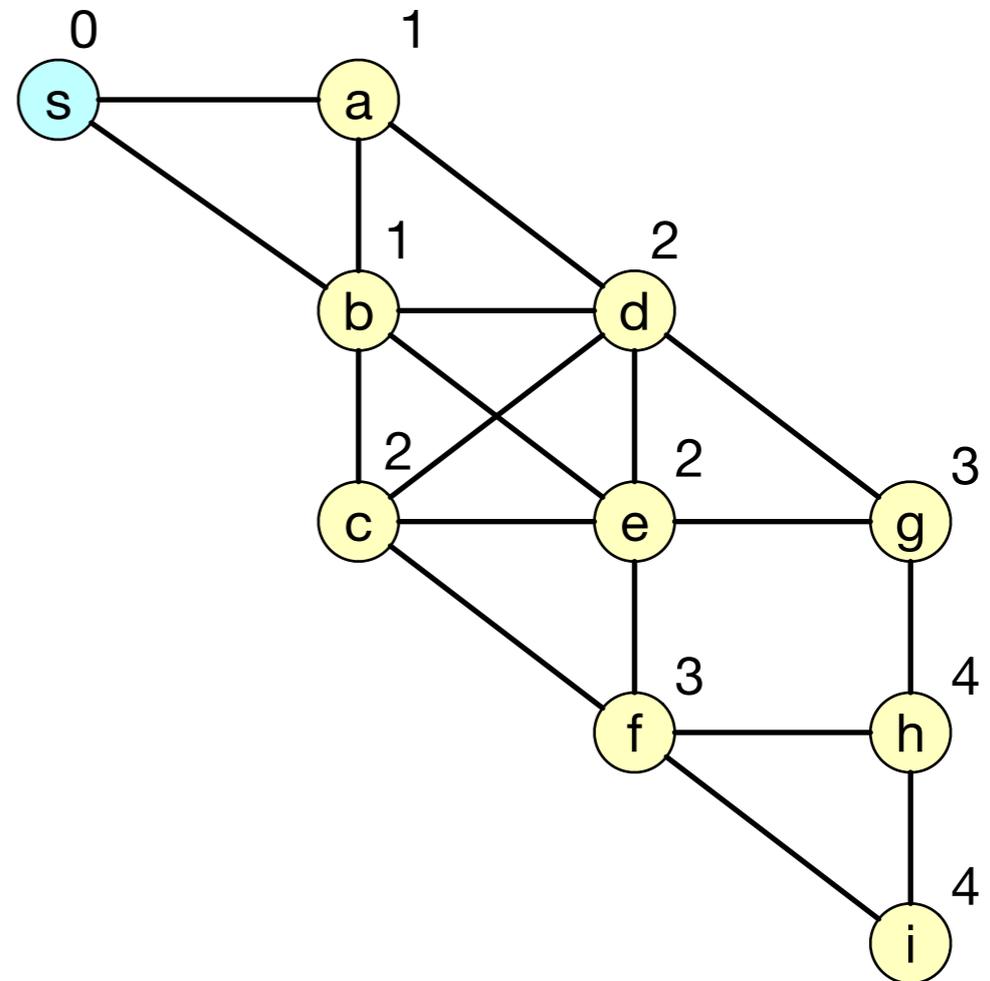
# Dijkstra's Algorithm

- Need to select h
  - Does not change any value



# Dijkstra's Algorithm

- Need to select i as the only node left
- But that does not change any values

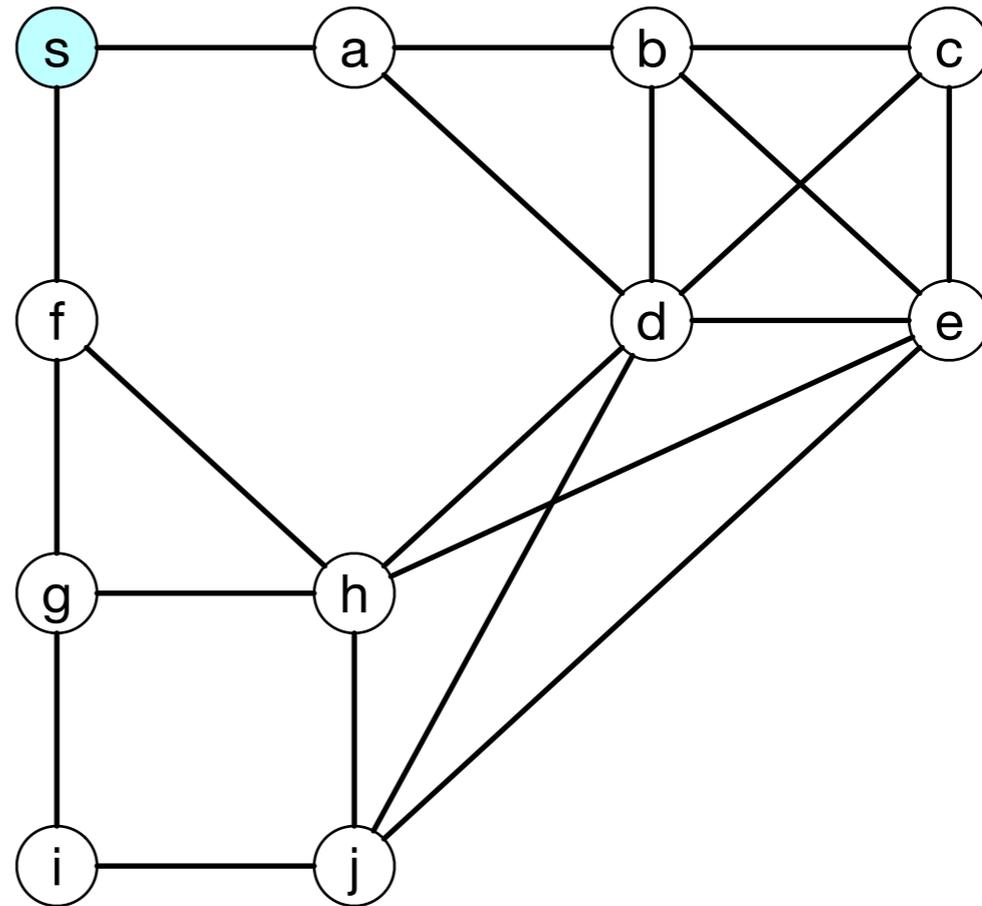


# Dijkstra's Algorithm

- Dijkstra's algorithm can be generalized to weighted graphs

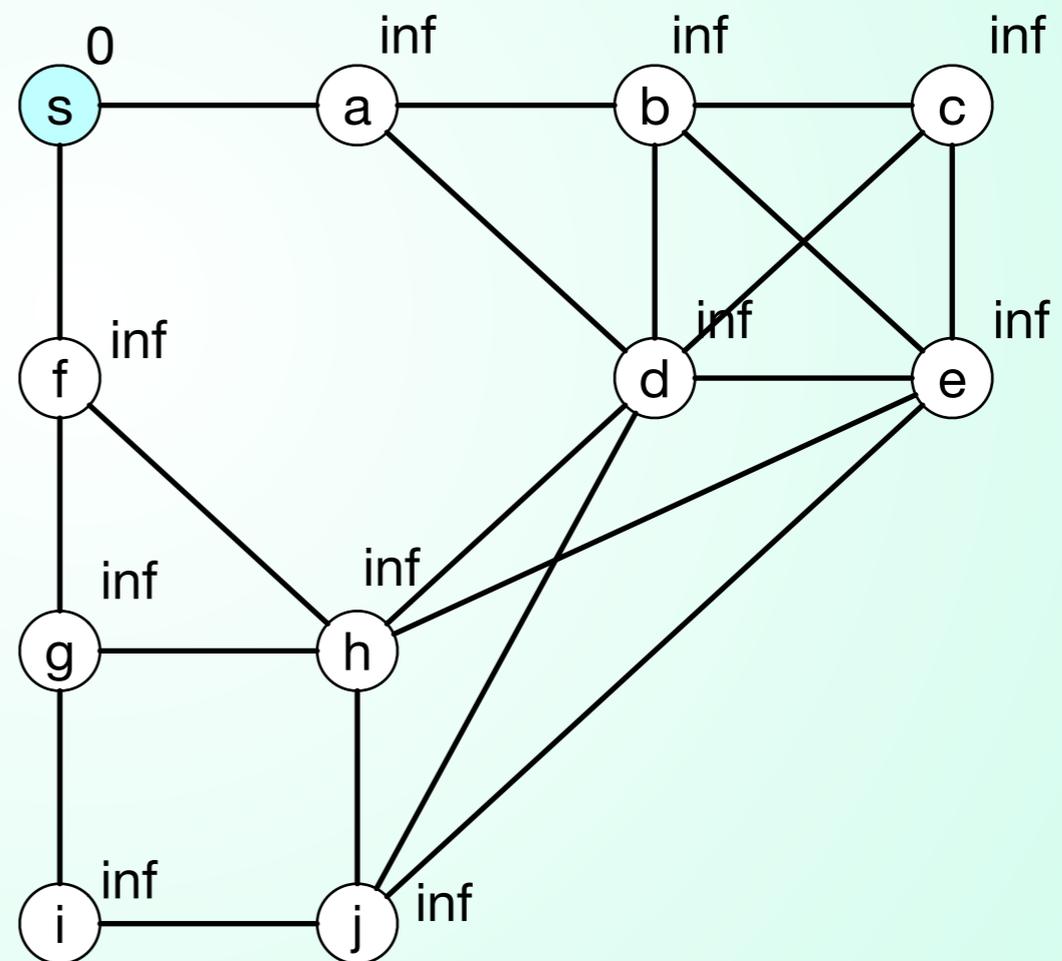
# Dijkstra's algorithm

- Your turn
- Rule:
  - Of course you choose smallest distance first, but you break ties in order of the alphabet, e.g. select a over f



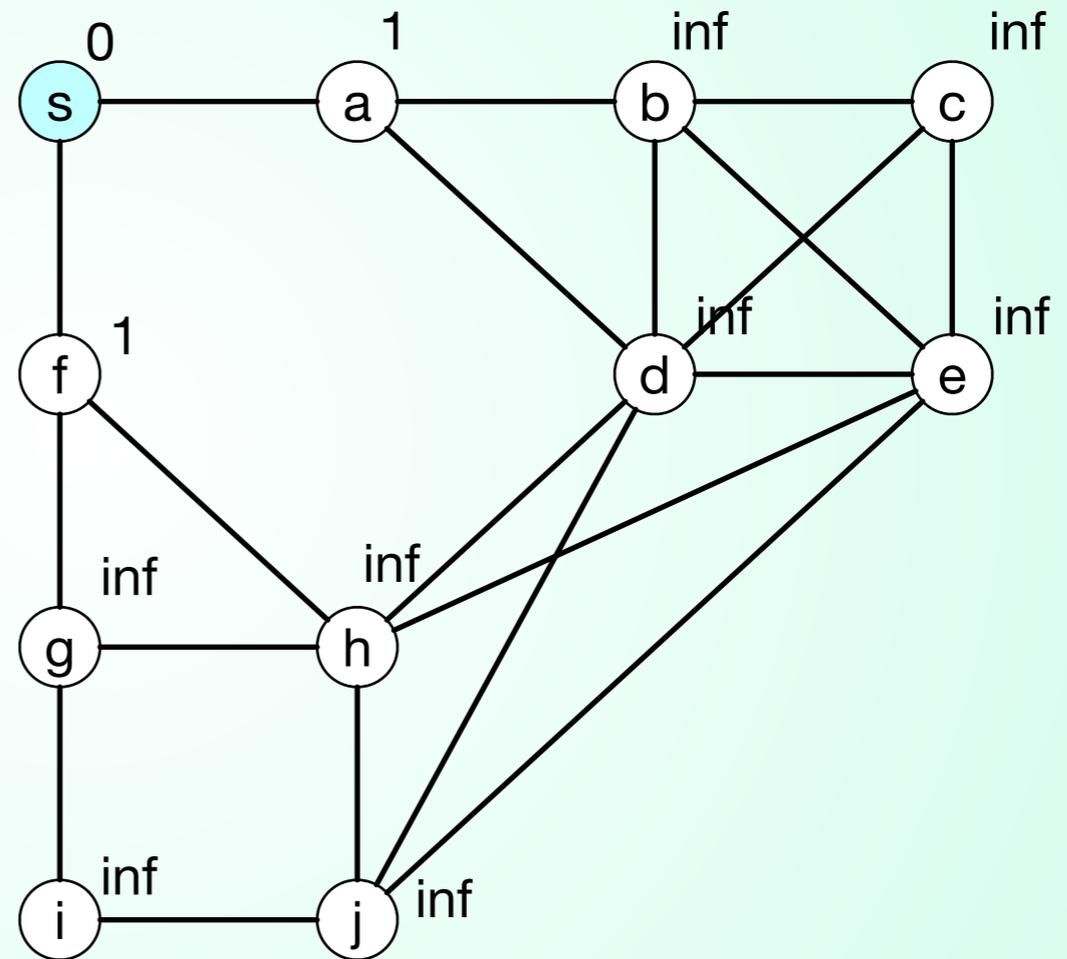
# Dijkstra's algorithm

- Select s



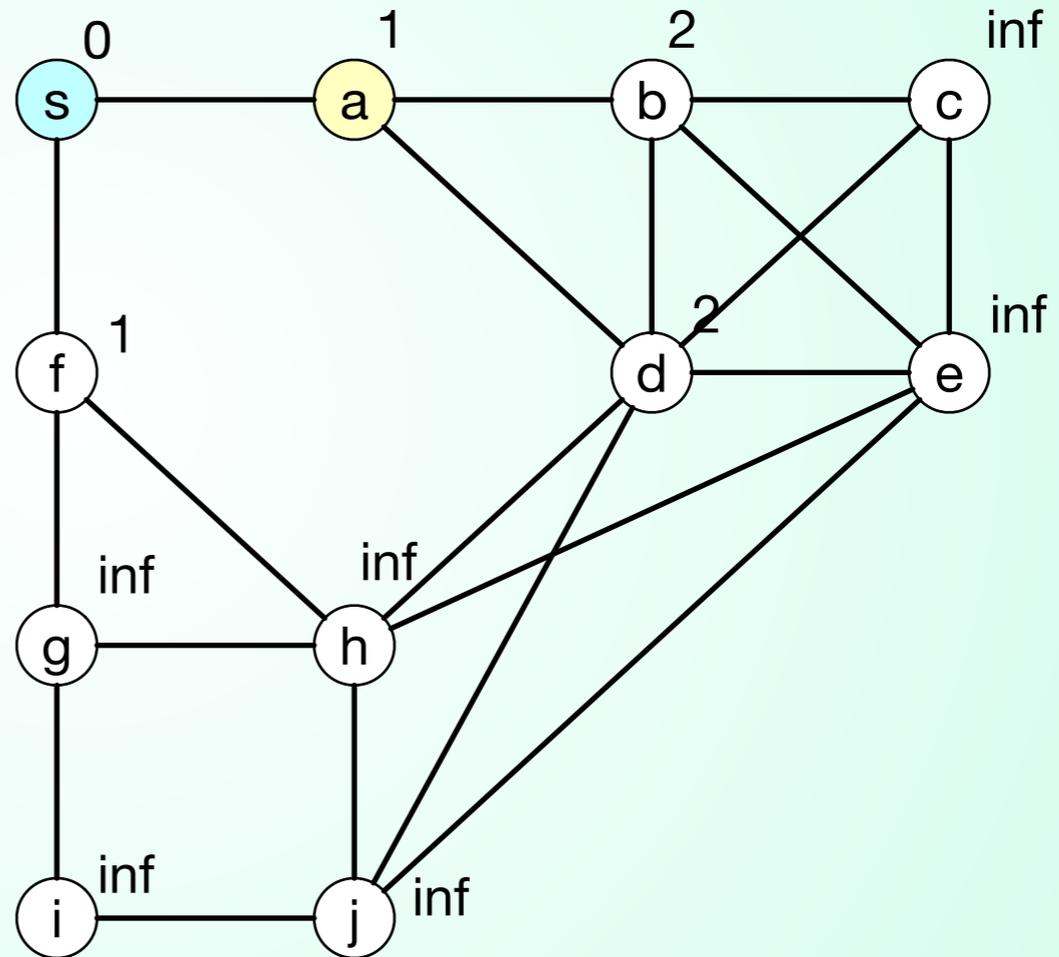
# Dijkstra's algorithm

- Update a and f



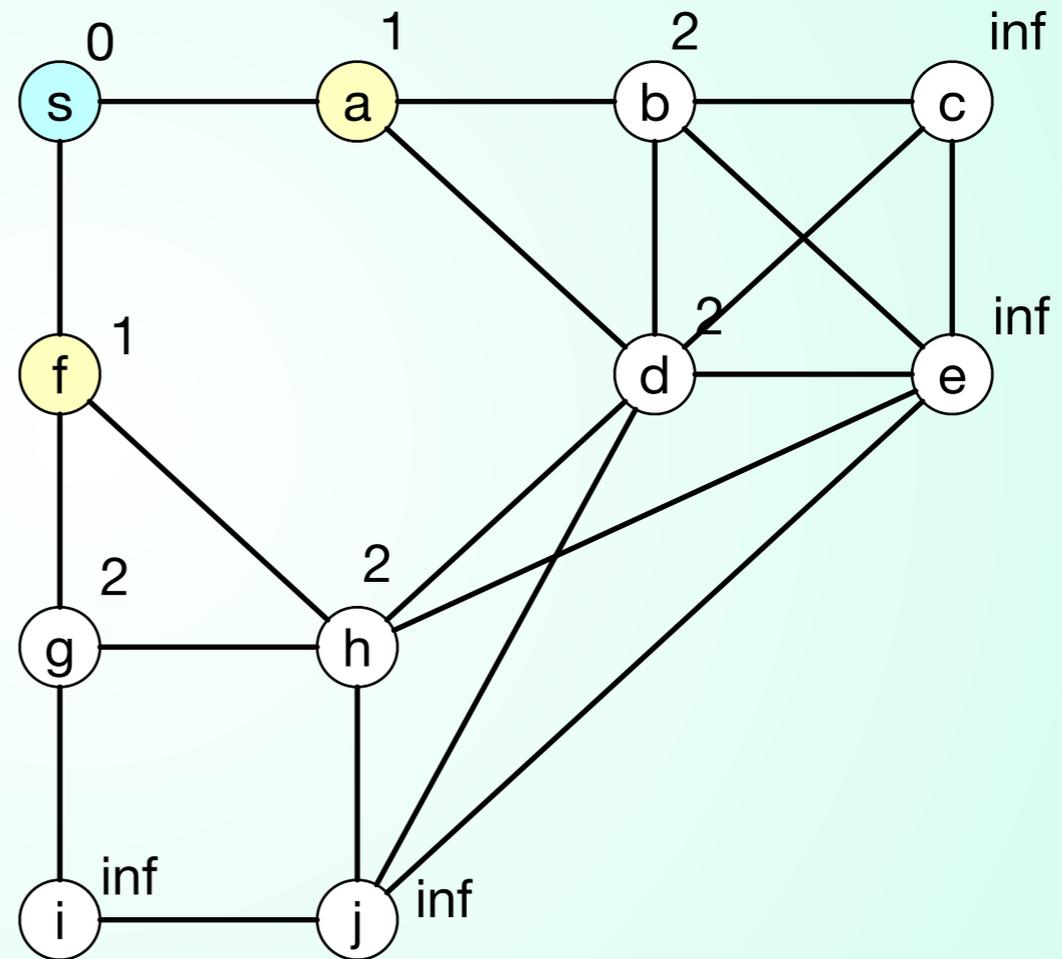
# Dijkstra's algorithm

- Select a
  - Update b and d
  - s stays the same



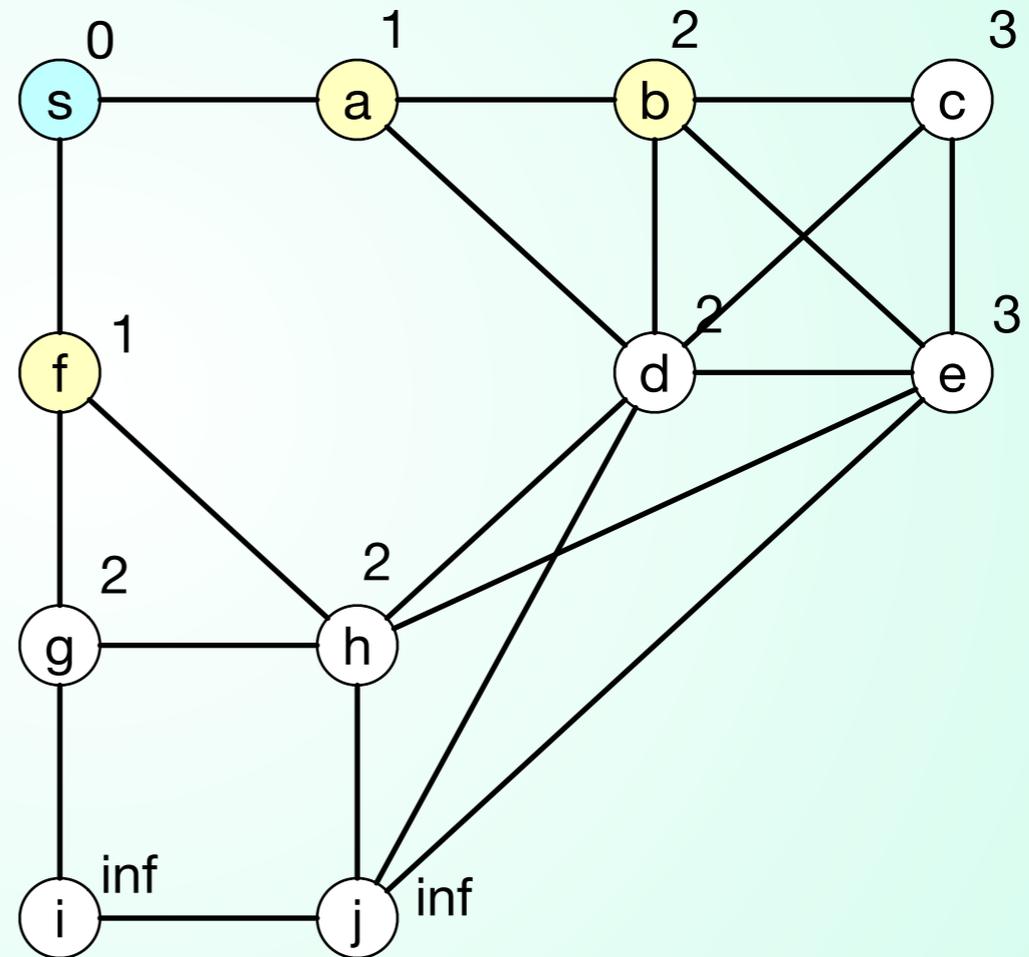
# Dijkstra's algorithm

- Select f (no choice here)



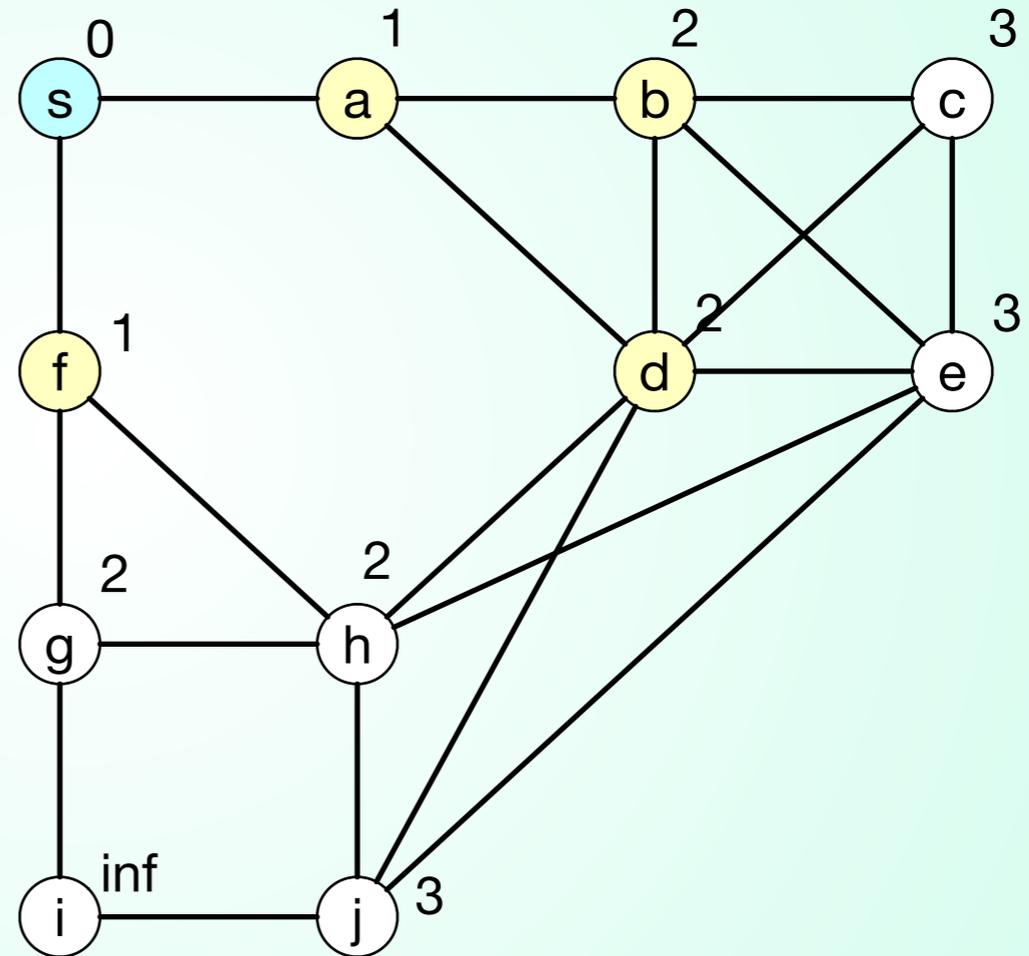
# Dijkstra's algorithm

- Select b



# Dijkstra's algorithm

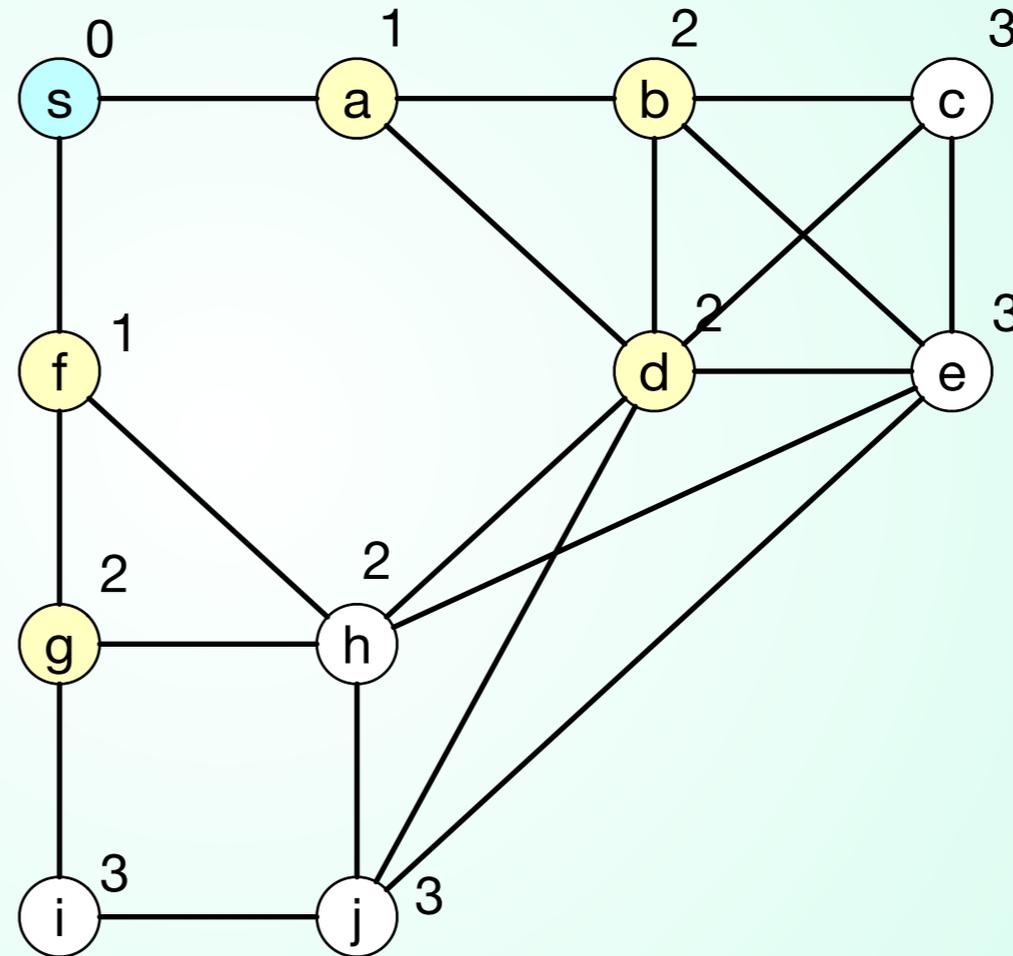
- Select d



# Dijkstra's algorithm

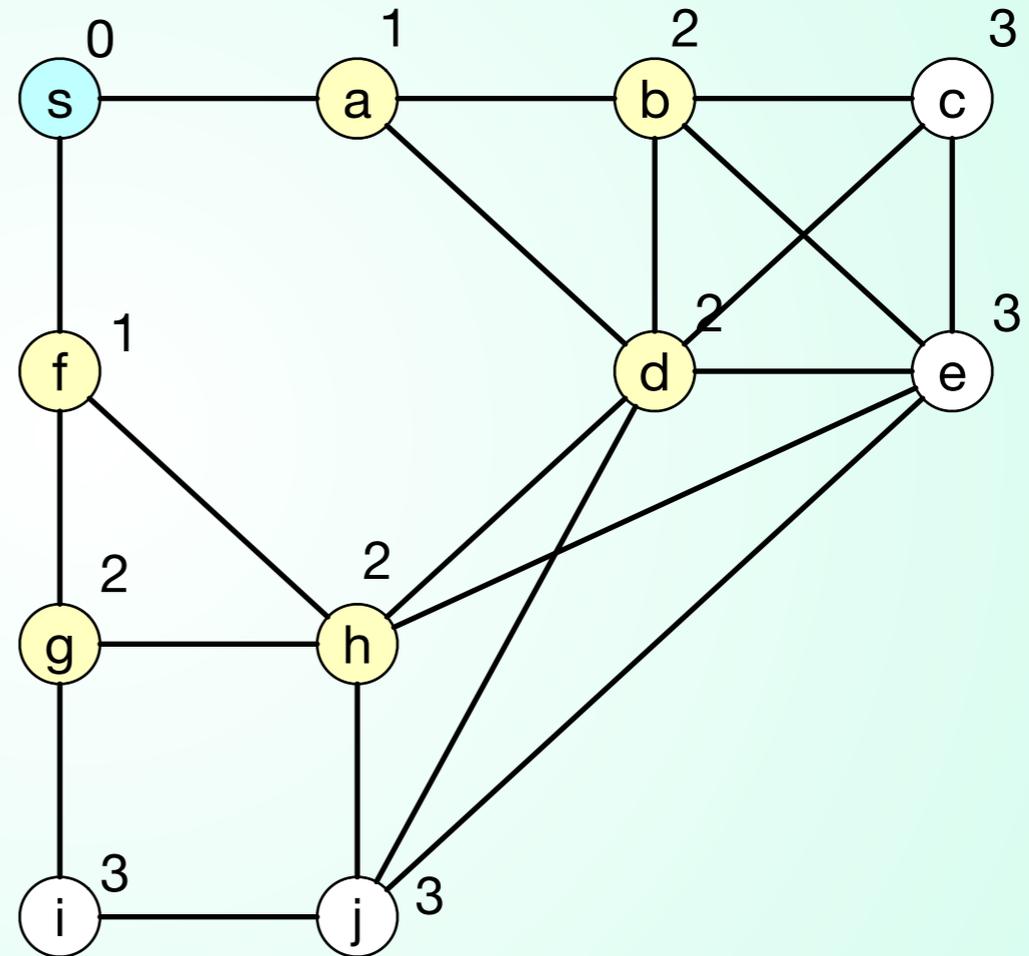
# Dijkstra's algorithm

- Select g



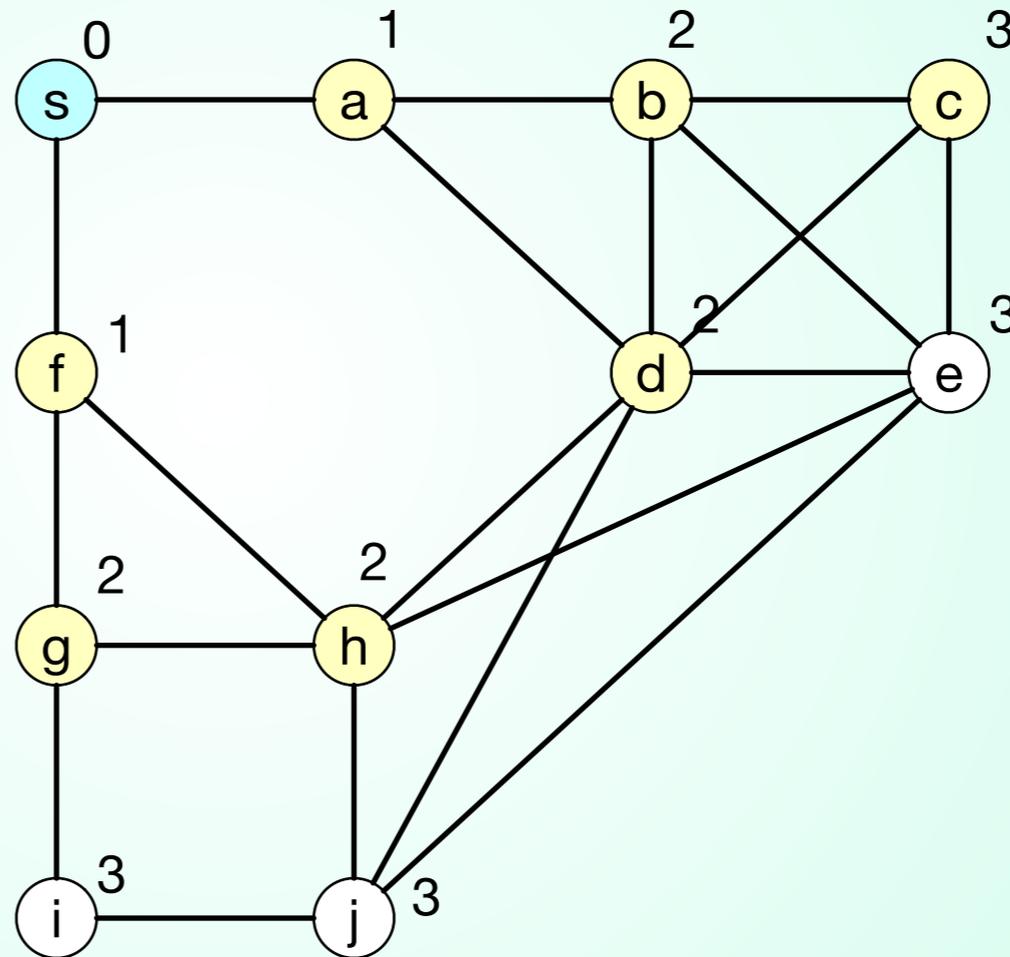
# Dijkstra's algorithm

- Select h



# Dijkstra's algorithm

- Select c
  - We might as well stop here
  - All updated values will be 4 or more, and every node has already a 3



# Graph Representations

- For computational purposes, we can use:

- List of vertices and list of edges as pairs

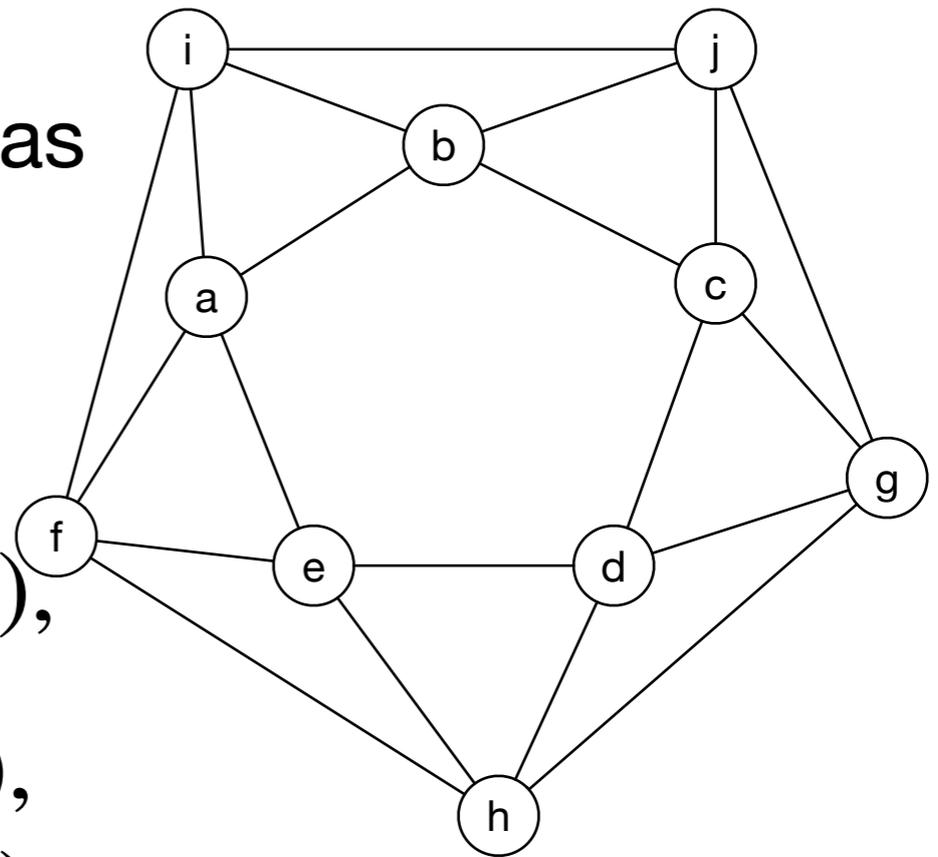
$$V = \{a, b, c, d, e, f, g, h, i, j\}$$

$$E = \{(a, b), (a, e), (a, f), (a, i), (b, c),$$

$$(b, i), (b, j), (c, d), (c, g), (c, j), (d, e),$$

$$(d, h), (d, g), (e, f), (e, h), (f, h), (f, i),$$

$$(g, h), (g, j), (i, j)\}$$



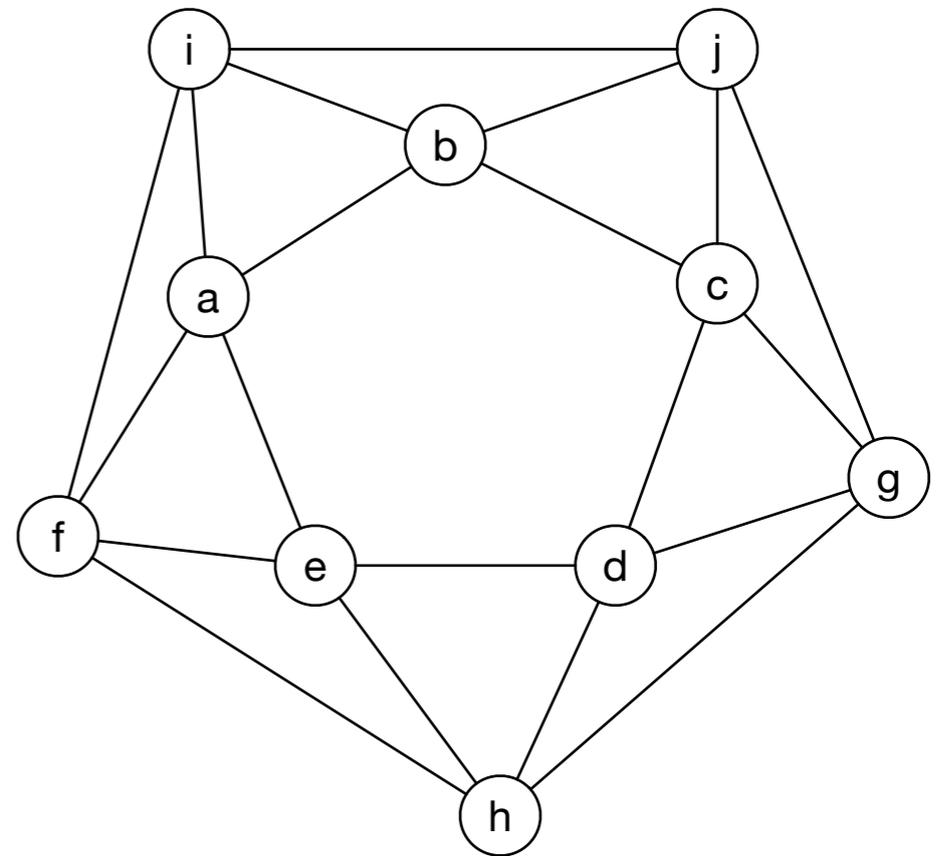
# Dijkstra's Algorithm

- Need to maintain a priority heap
  - Otherwise
    - Look at every node
    - And every edge twice

# Graph Representations

- An adjacency list
  - For every vertex the list of vertices to which there is an edge

- - a: b, e, f, i
  - b: a, c, i, j
  - c: b, d, g, j
  - d: c, e, g, h
  - e: a, f, d, h
  - f: a, e, h, i
  - g: c, d, h, j
  - h: d, e, f, g
  - i: a, b, f, j
  - j: b, c, g, i

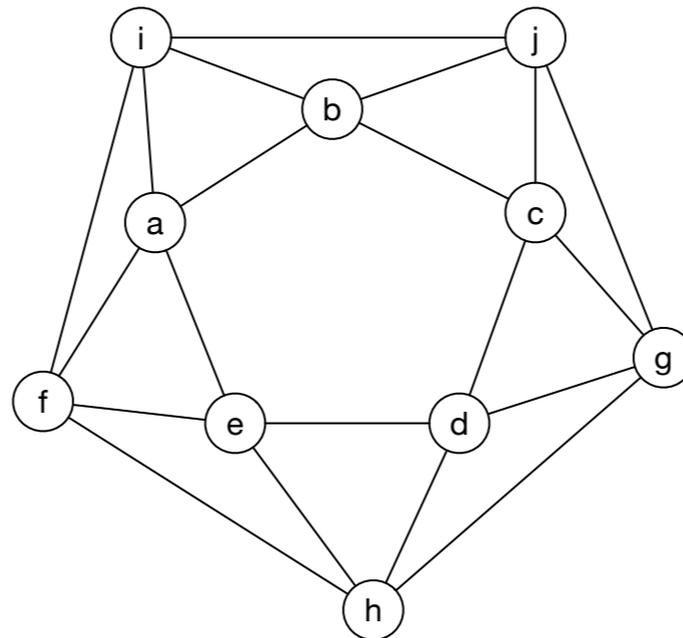


# Graph Representations

- An adjacency matrix
  - square matrix conceptually labeled with vertices

- coefficient

$$a_{i,j} = \begin{cases} 1 & \text{edge between } v_i \text{ and } v_j \\ 0 & \text{otherwise} \end{cases}$$



	a	b	c	d	e	f	g	h	i	j
a	0	1	0	0	1	1	0	0	1	0
b	1	0	1	0	0	0	0	0	1	1
c	0	1	0	1	0	0	1	0	0	1
d	0	0	1	0	1	0	1	1	0	0
e	1	0	0	1	0	1	0	1	0	0
f	1	0	0	0	1	0	0	1	1	0
g	0	0	1	1	0	0	0	1	0	1
h	0	0	0	1	1	1	1	0	0	0
i	1	1	0	0	0	1	0	0	0	1
j	0	1	1	0	0	0	1	0	1	0

# Number of Vertices and Edges

- Graph  $G = (V, E)$  with vertices  $V$  and edges  $E$ 
  - Whether directed or undirected, graph can have as many edges as there are pairs of vertices
  - The latter is  $\binom{|V|}{2} = \frac{|V|(|V| - 1)}{2}$
  - Number of edges is at most  $O(|V|^2)$

# Number of Vertices and Edges

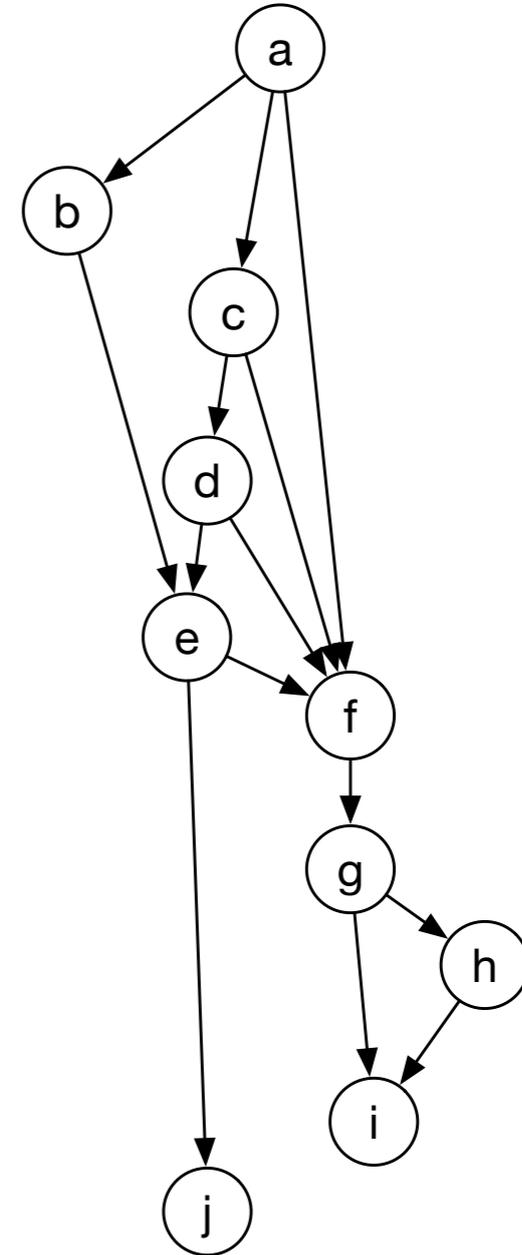
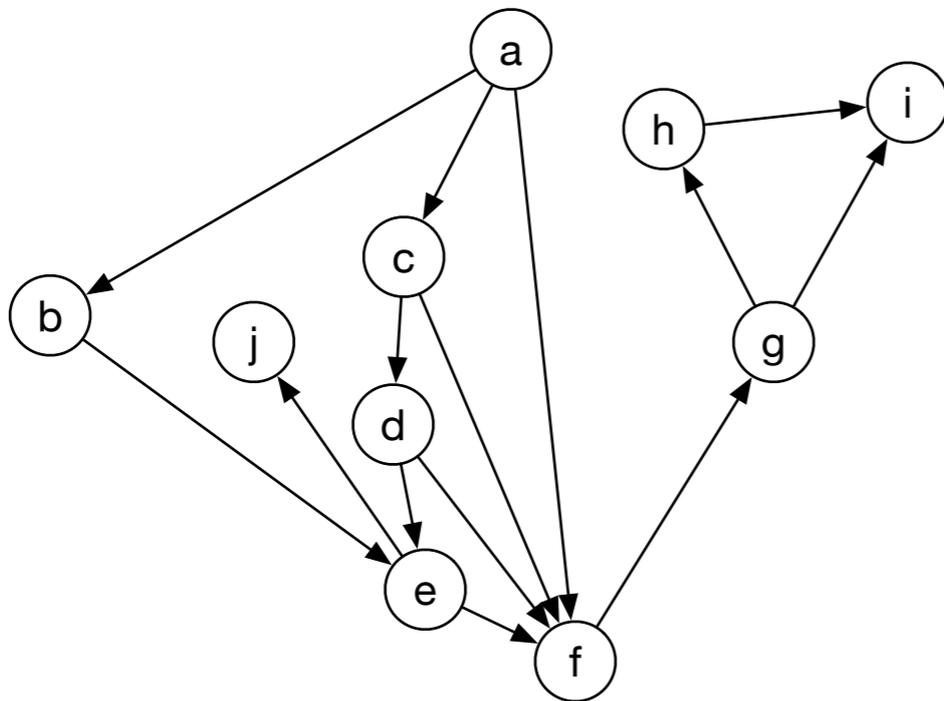
- Graph  $G = (V, E)$  with vertices  $V$  and edges  $E$ 
  - Graph algorithms usually need to look at each edge at least once
    - there are some idiosyncratic exceptions
    - They usually run in time at least  $\Theta(|V|^2)$
  - However, many important graphs are sparse:
    - No edge between most pairs of vertices

# Topological Sort

- We can use a directed graph in order to represent a precedence relation
  - Topological sort:
    - Given a directed graph:
      - Order all vertices in an order such that an edge always goes from a preceding to a succeeding vertex
      - Or show that this is impossible because there is a cycle

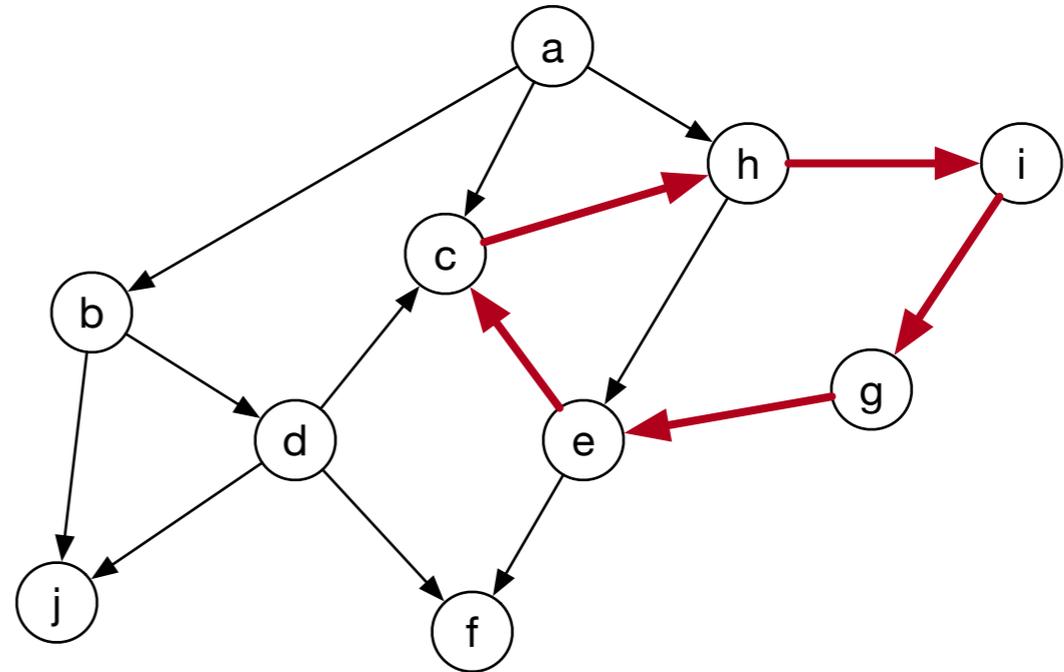
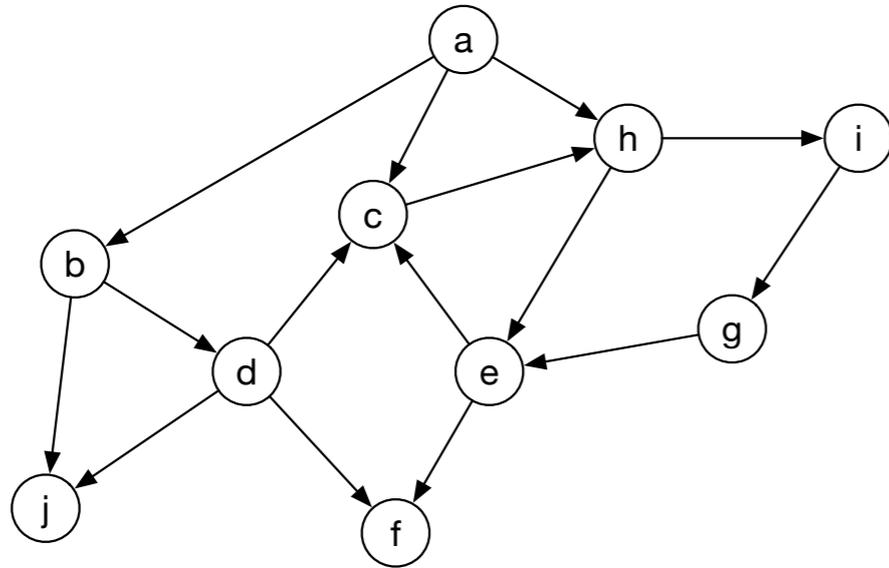
# Topological Sort

- Example 1:
  - Can arrange all vertices such that arrows only go down
  - Sort is a,b,c,d,e,f,g,h,i,j



# Topological Sort

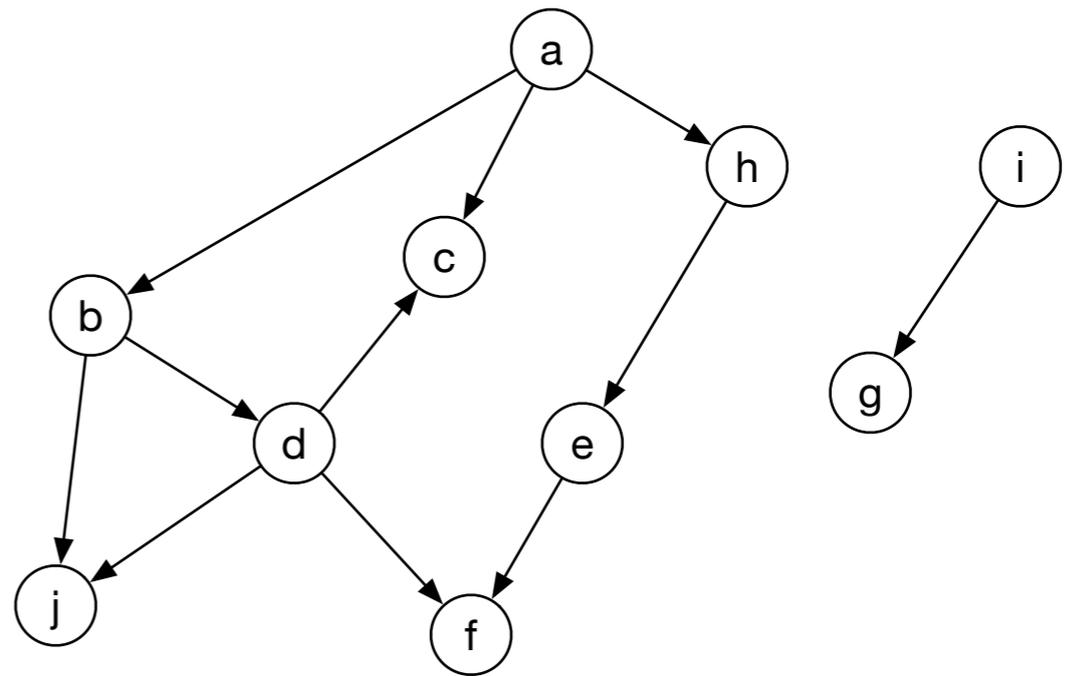
- Example:
  - There is a cycle, a topological sort is not possible



# Topological Sort

- A simple algorithm:
  - Go to the adjacency list

```
a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

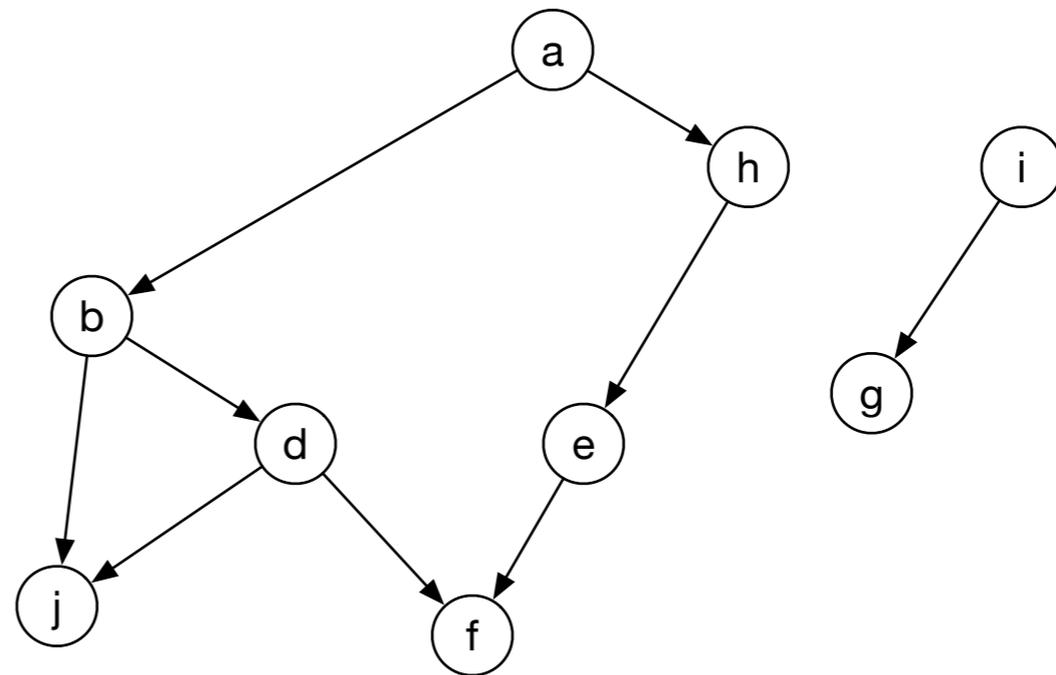


- Find a vertex with empty list, add it to a list, and remove it from the graph

# Topological Sort

- A simple algorithm

```
a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

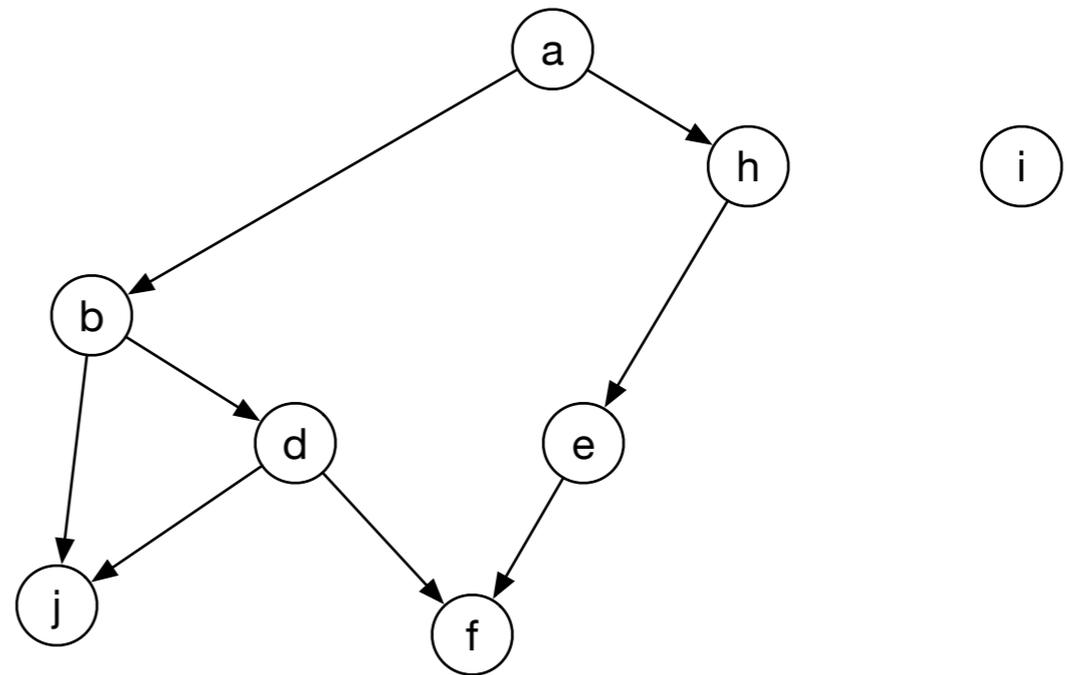


- List contains {c}

# Topological Sort

- A simple algorithm

a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:

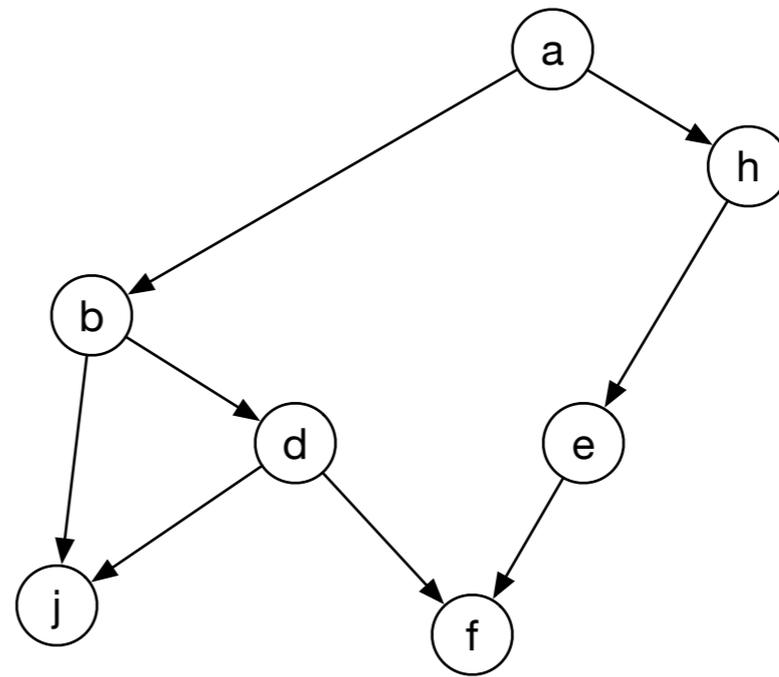


- Remove g and add it to the list {c, g}

# Topological Sort

- A simple algorithm

a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:

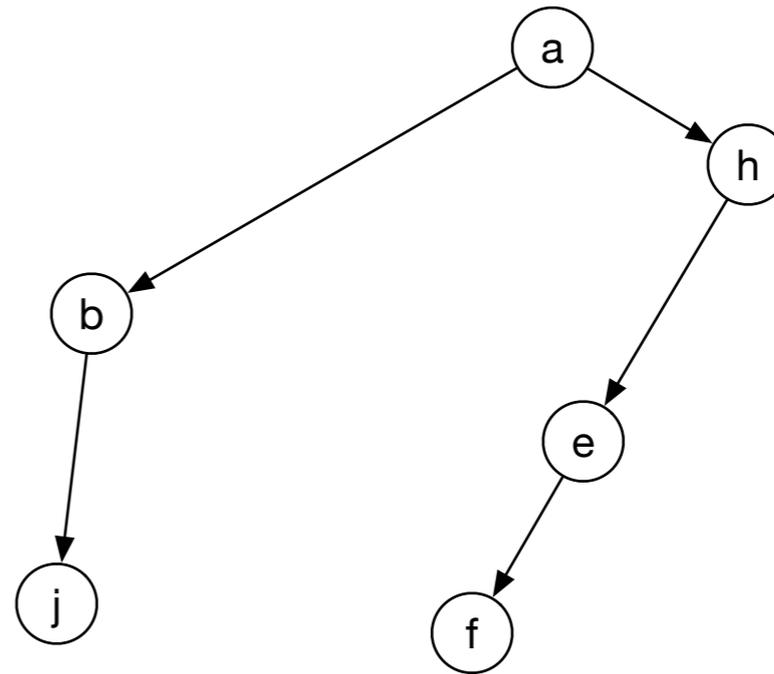


- Remove i and add it to the list  $\{c, g, i\}$

# Topological Sort

- A simple algorithm

```
a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

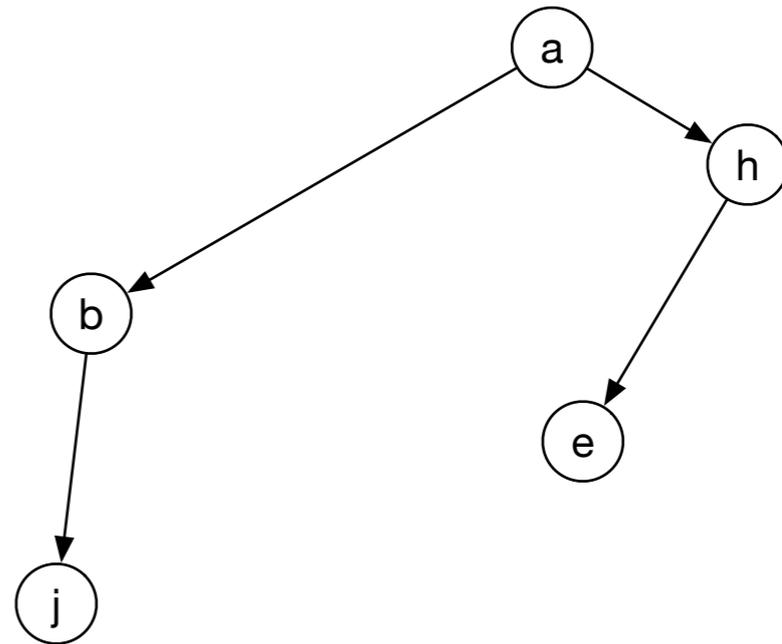


- Remove d and add it to the list  $\{c, g, i, d\}$

# Topological Sort

- A simple algorithm

a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:

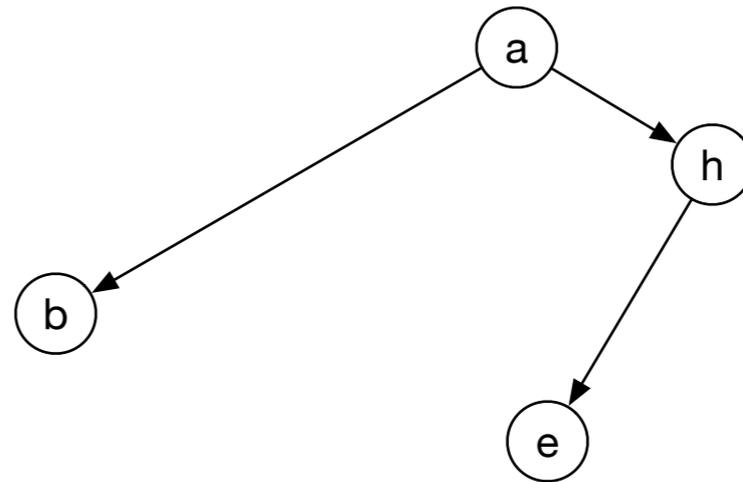


- Remove f and add it to the list  $\{c, g, i, d, f\}$

# Topological Sort

- A simple algorithm

a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:

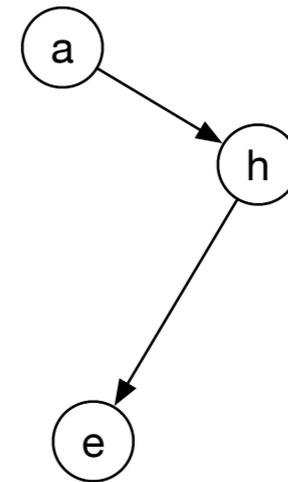


- Remove j and add it to the list  $\{c, g, i, d, f, j\}$

# Topological Sort

- A simple algorithm

```
a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

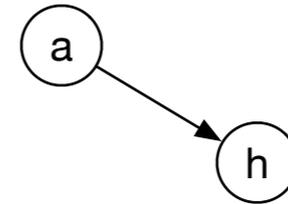


- Remove b and add it to the list  $\{c, g, i, d, f, j, b\}$

# Topological Sort

- A simple algorithm

```
a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

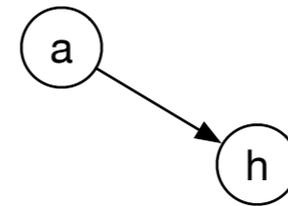


- Remove e and add it to the list  $\{c, g, i, d, f, j, b, e\}$

# Topological Sort

- A simple algorithm

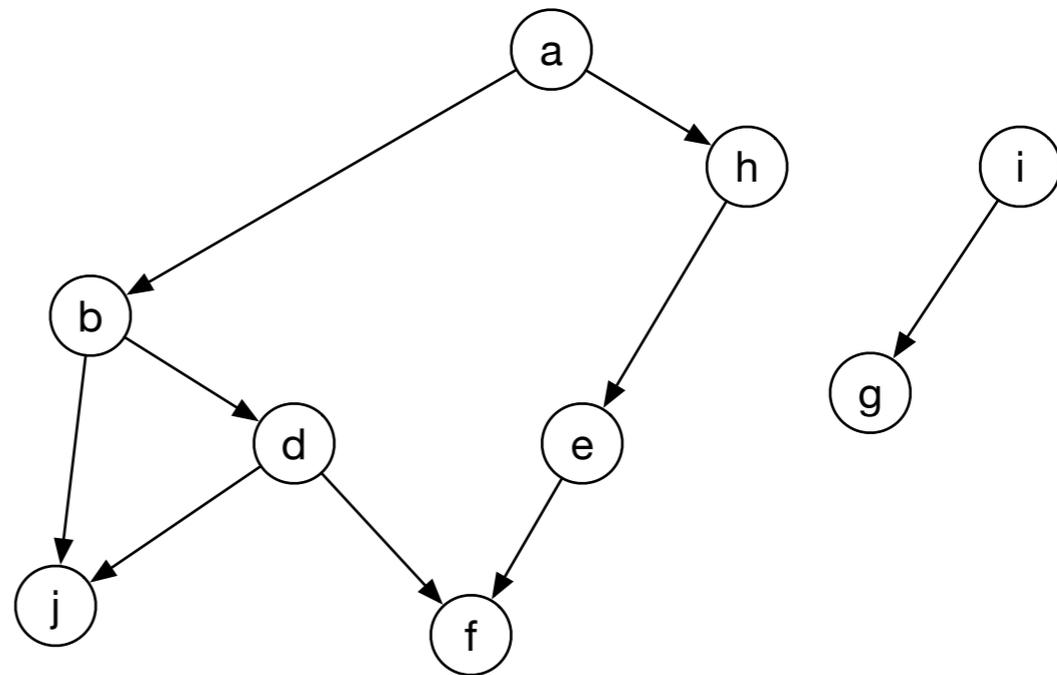
```
a: b, c, h  
b: d, j  
c:  
d: c  
e: f  
f:  
g:  
h: e  
i: g  
j:
```



- Remove a and add it to the list  $\{c, g, i, d, f, j, b, e, h, a\}$

# Topological Sort

- The reverse list is the topological sort:
  - $\{a, h, e, b, j, f, d, i, g, c\}$



# Topological Sort

- In this version, we have
  - To determine the length of the adjacency list
  - After selecting a vertex, delete that vertex from all the adjacency lists
- The latter means scanning all adjacency lists repeatedly
- This is inefficient

# Topological Sort

- Question: How can we do this better?

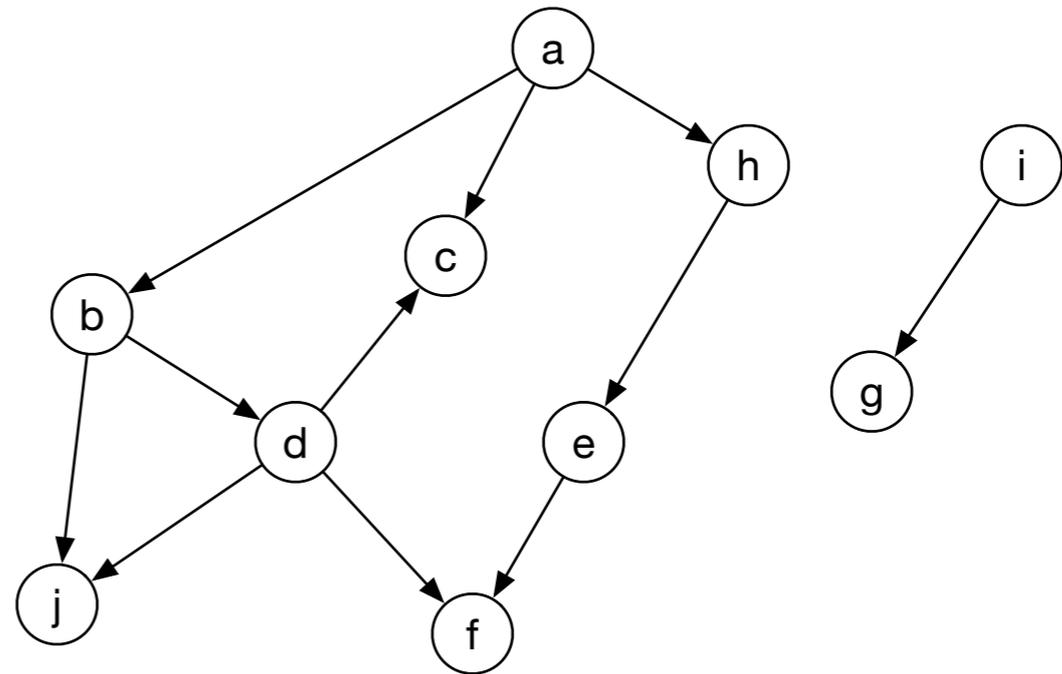
# Topological Sort

- Instead of optimizing the search for vertices, we can optimize the selection of the vertex for removal
- Better algorithm:
  - Find the in-degree for all vertices
    - That is the number of edges going into a vertex
  - While there are vertices with in-degree 0
    - Remove the vertex
    - Update the in-degrees

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:

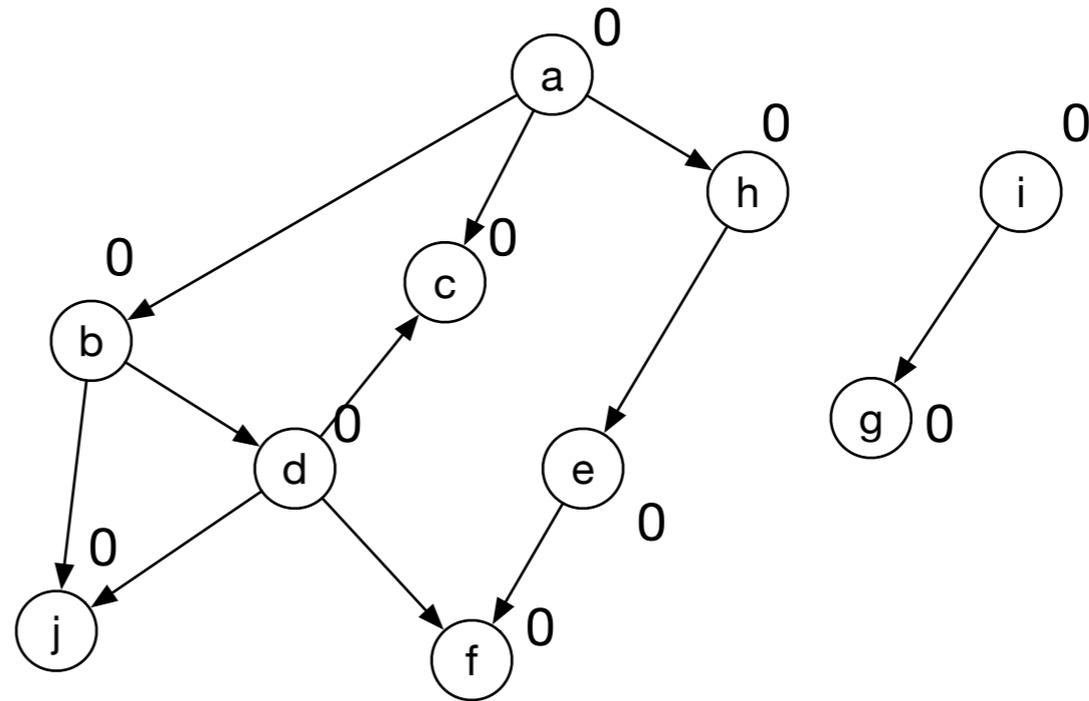


- Initialize in-degree 0 for all vertices

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:



- Initialize in-degree 0 for all vertices

# Topological Sort

- Example:

a: **b, c, h**

b: d, j

c:

d: c, j

e: f

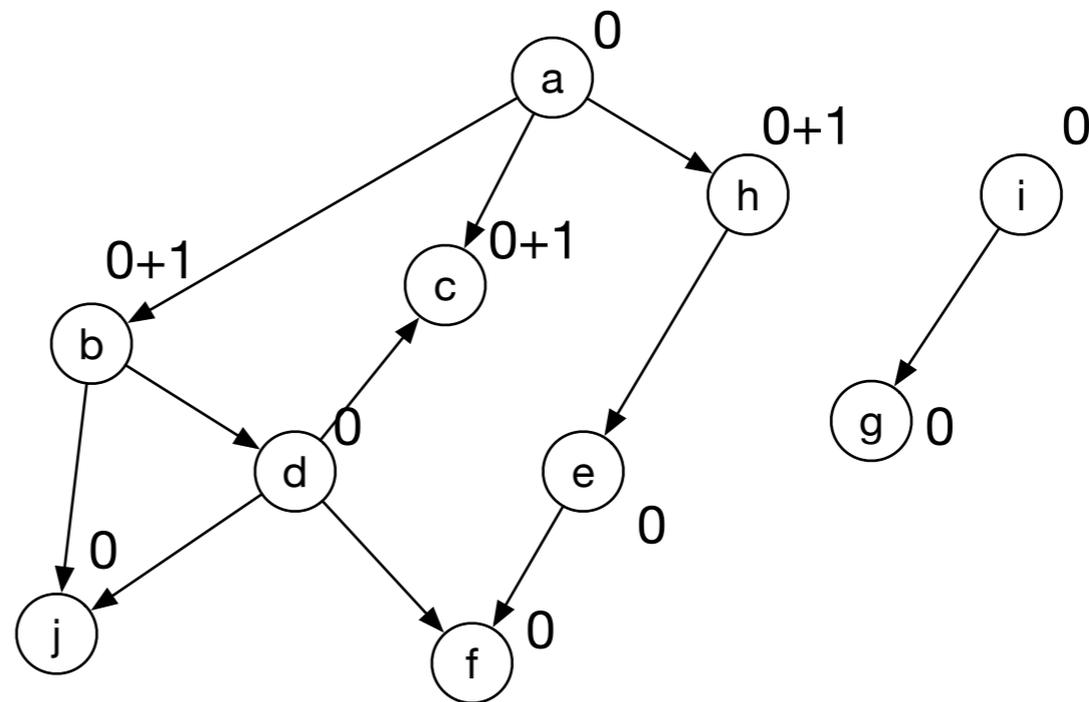
f:

g:

h: e

i: g

j:



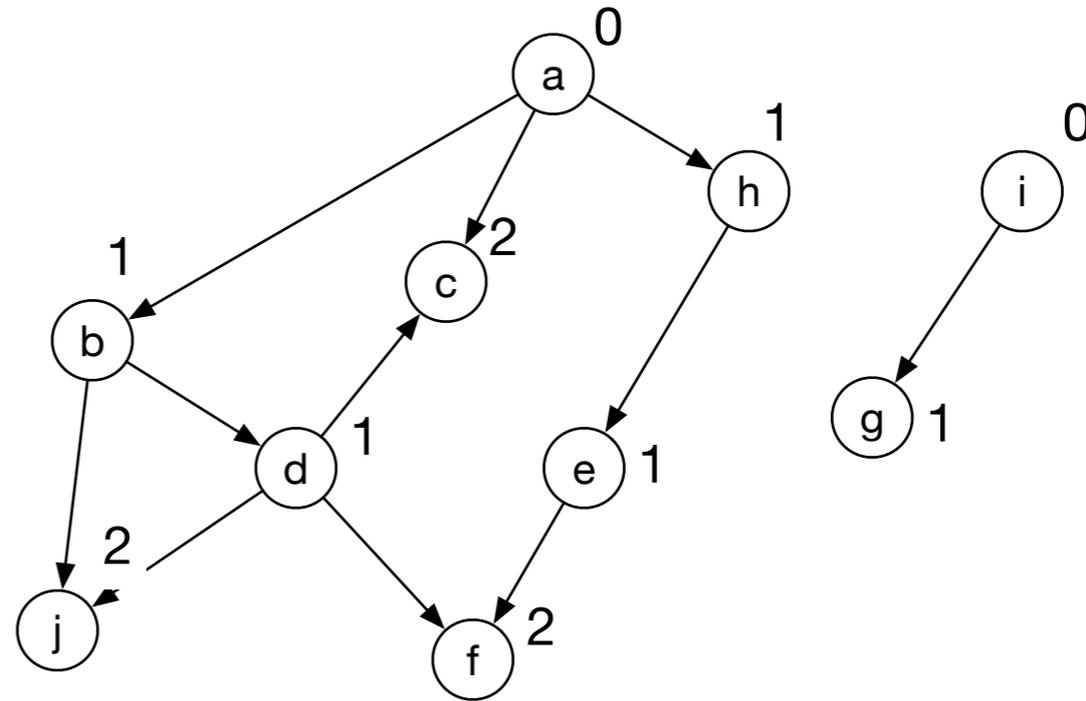
- Go through the adjacency list.

- For each vertex in an adjacency list, add 1 to the in-degree
- For a, we change three in-degrees

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:



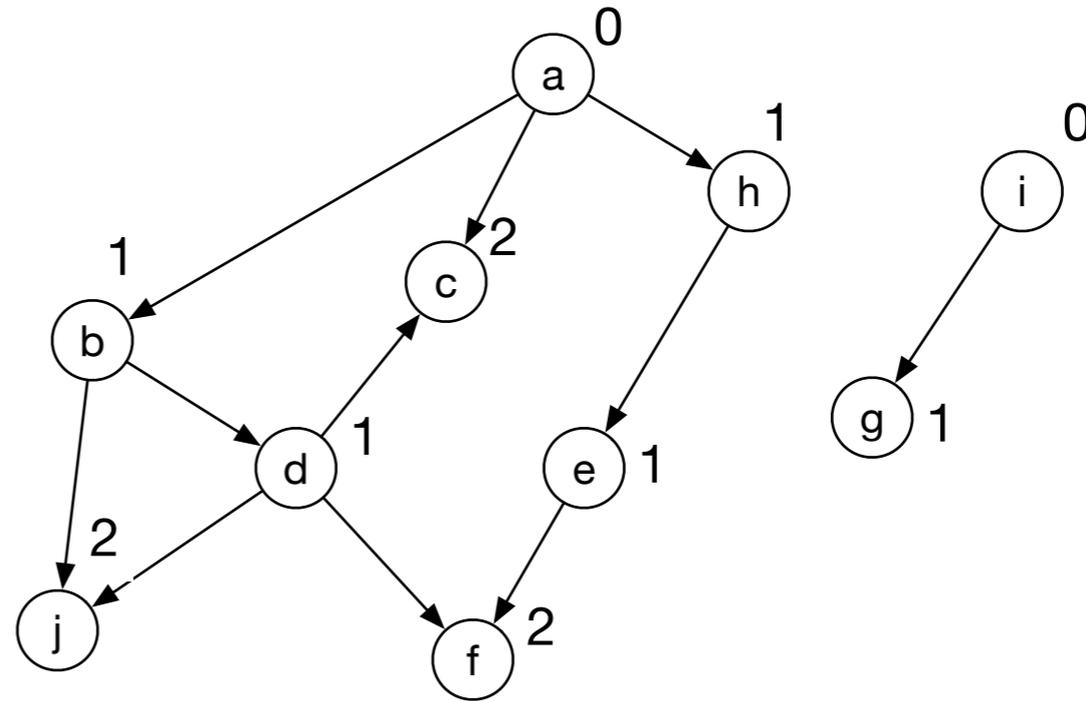
- Go through the adjacency list.

- After processing all adjacency lists, we have the correct in-degrees

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:



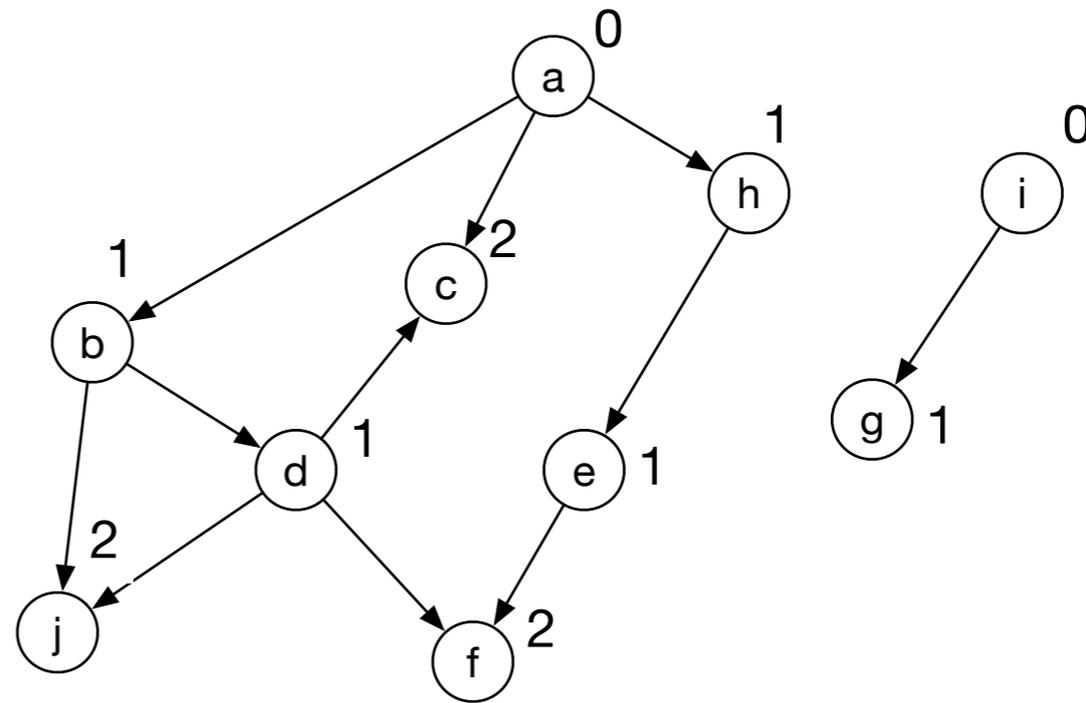
- Now we start the removal phase

- We need to find a vertex with in-degree 0
- How can we make this more efficient?

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:



- Now we start the removal phase

- We need to find a vertex with in-degree 0
- Could place the vertices in a heap

# Topological Sort

- Example:

a: b, c, h

b: d, j

c:

d: c, j

e: f

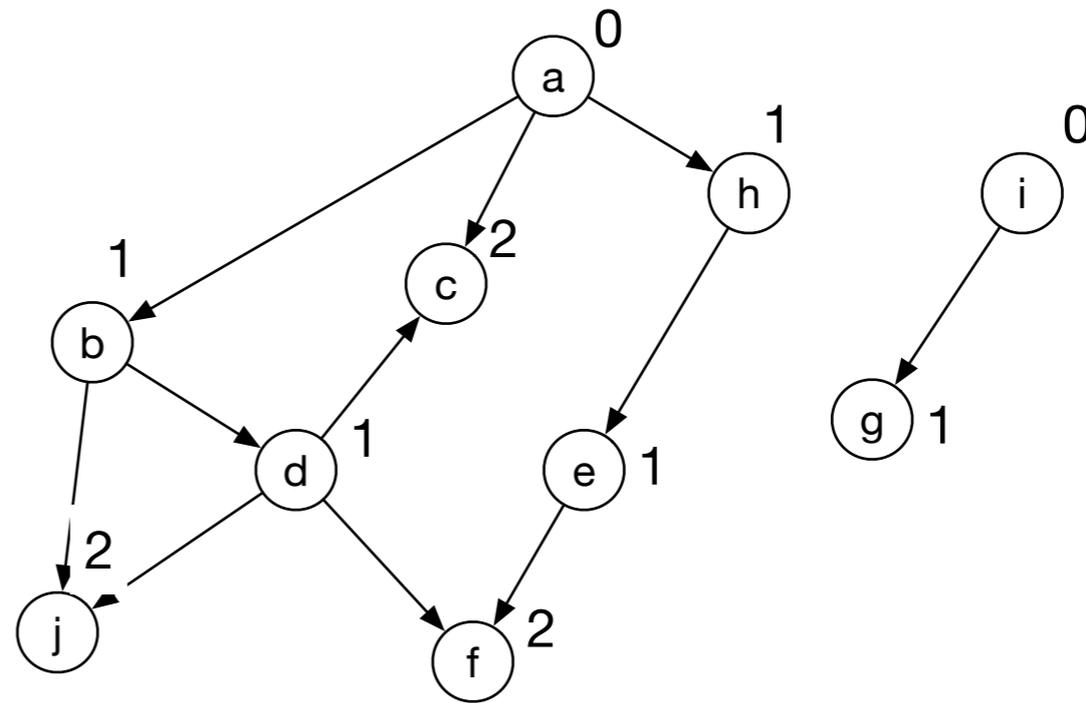
f:

g:

h: e

i: g

j:



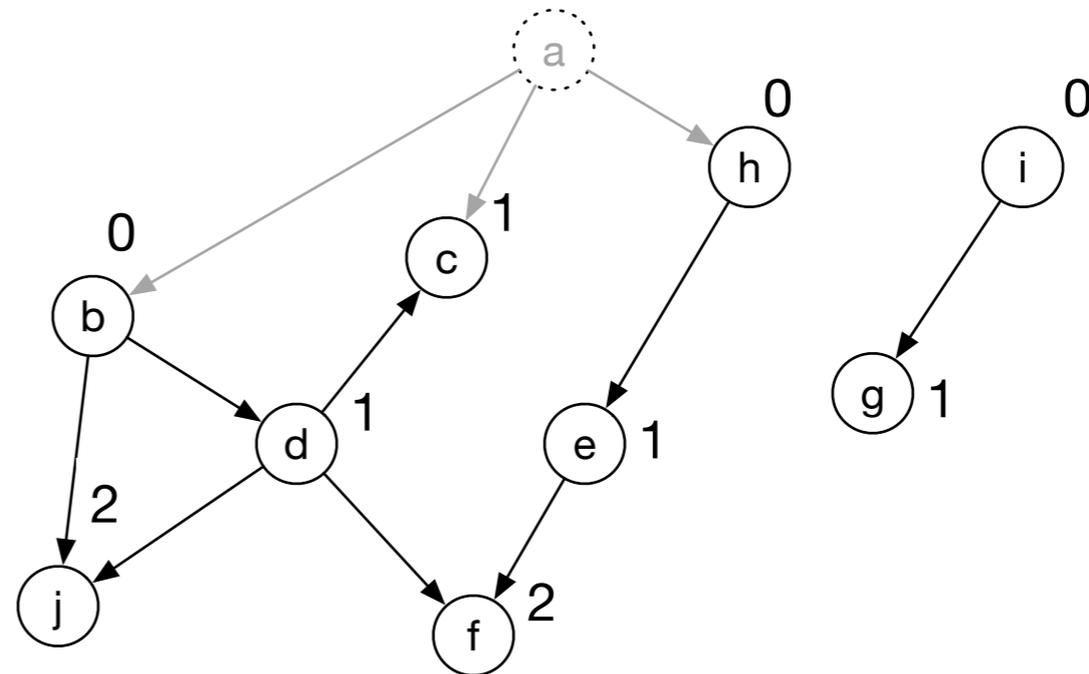
- We select a for the removal

- We go through its adjacency list and reset the in-degrees of the nodes there

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:

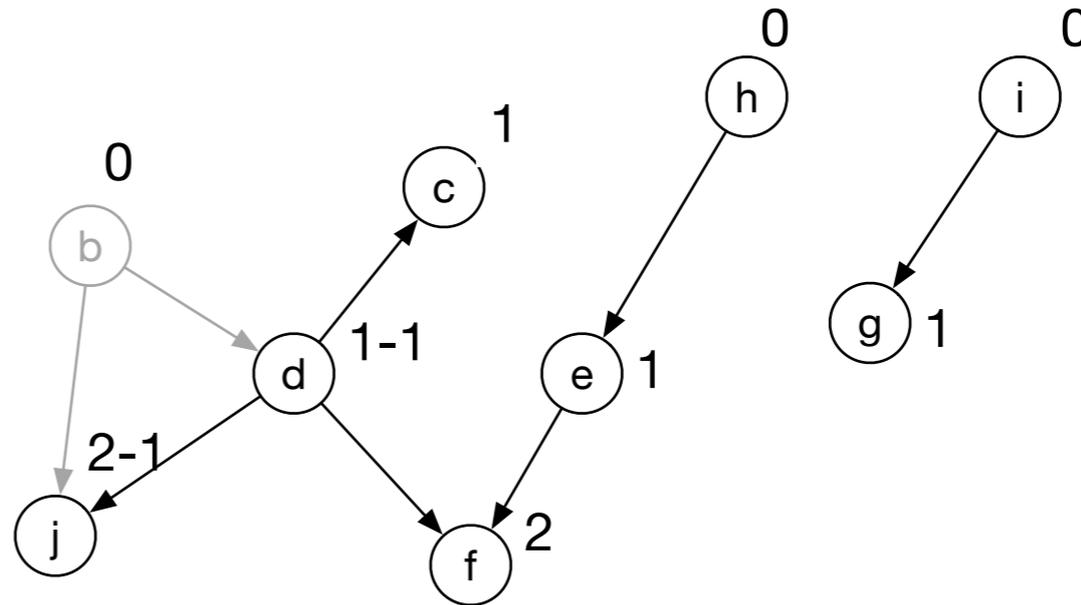


- We select a for the removal:  $\{a\}$ 
  - We go through its adjacency list and reset the in-degrees of the nodes there

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:



- We update our heap and select one of the 0-in-degree vertices:

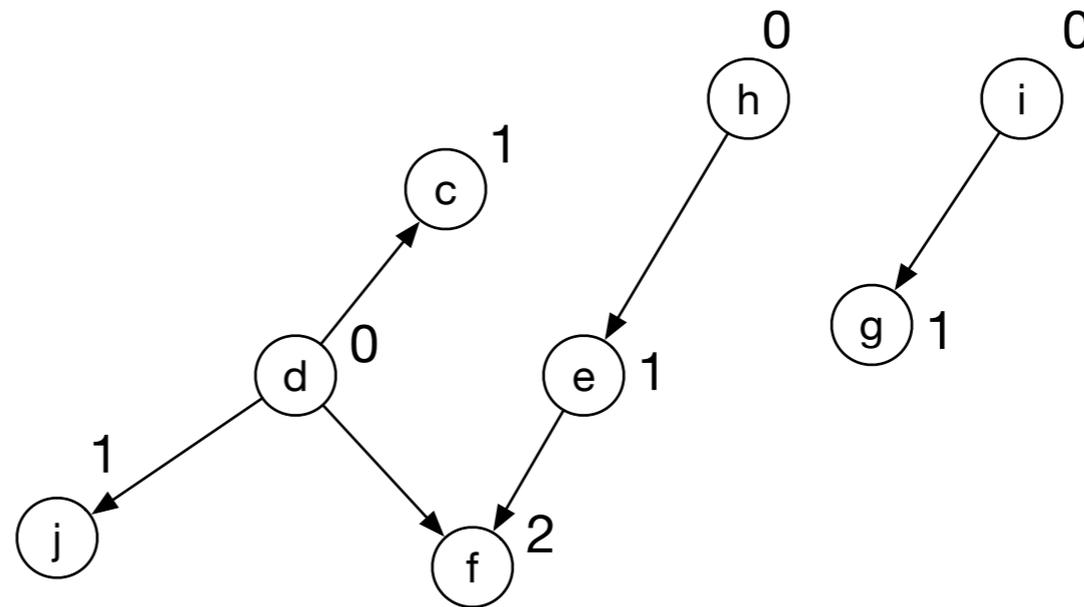
- b: {a, b}

- and update the in-degrees of d and j

# Topological Sort

- Example:

```
a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:
```



- We update our heap and select one of the 0-in-degree vertices:

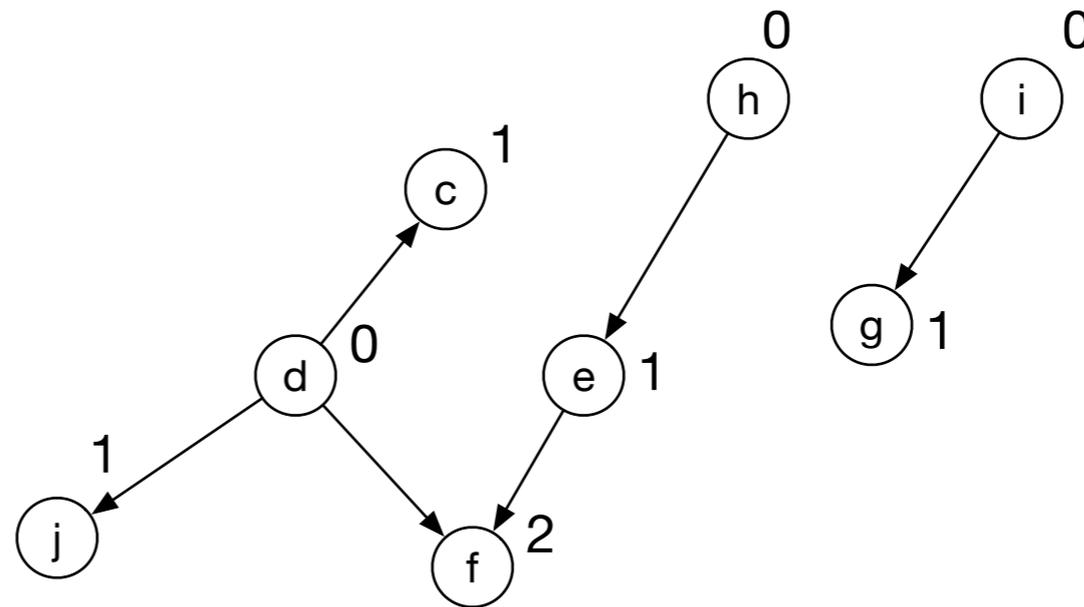
- $b: \{a, b\}$

- and update the in-degrees of d and j

# Topological Sort

- Example:

```
a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

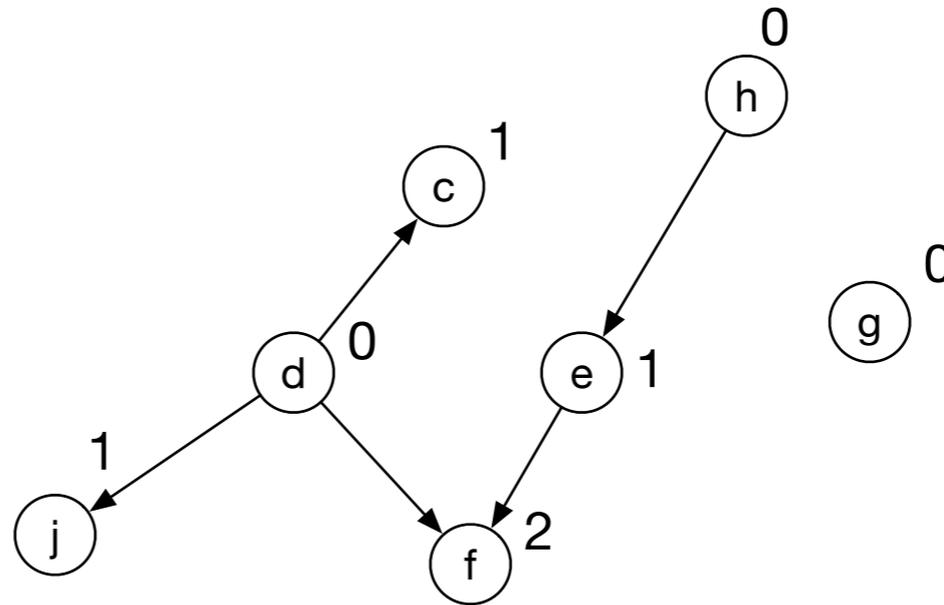


- We now randomly pick one of the vertices with degree 0, let's pick i
- Deleting it means just decrementing the in-degree of g

# Topological Sort

- Example:

a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:

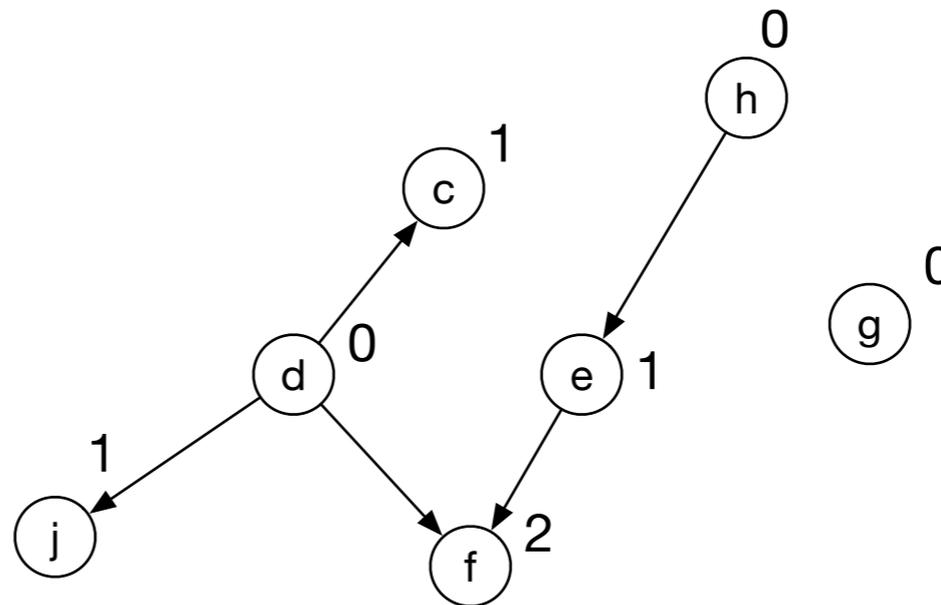


- We add g to our list  $\{a, b, i, g\}$

# Topological Sort

- Example:

```
a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

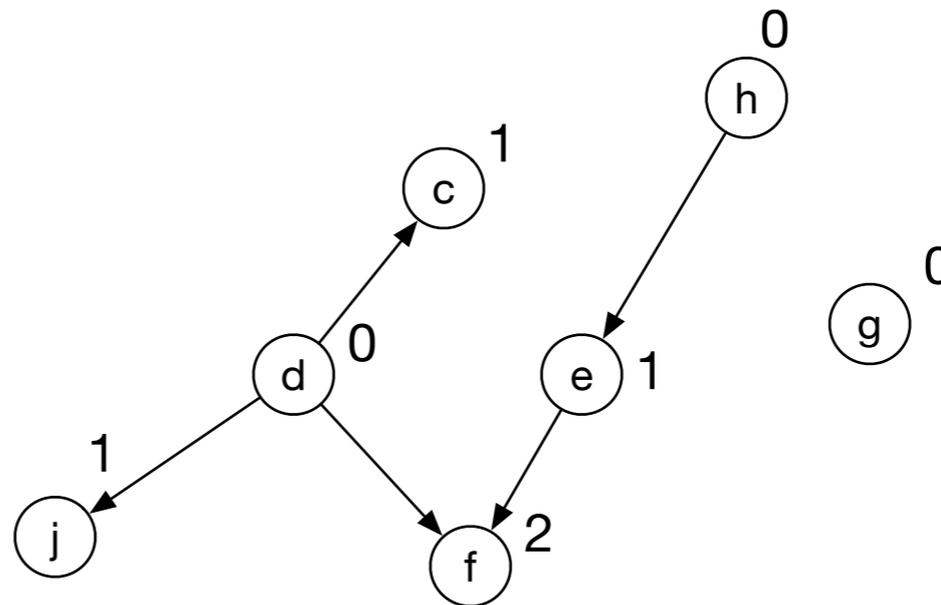


- There are three nodes with in-degree 0, let's pick h

# Topological Sort

- Example:

```
a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:
```



- There are three nodes with in-degree 0, let's pick h

# Topological Sort

- Example:

a: b, c, h

b: d, j

c:

d: c, j

e: f

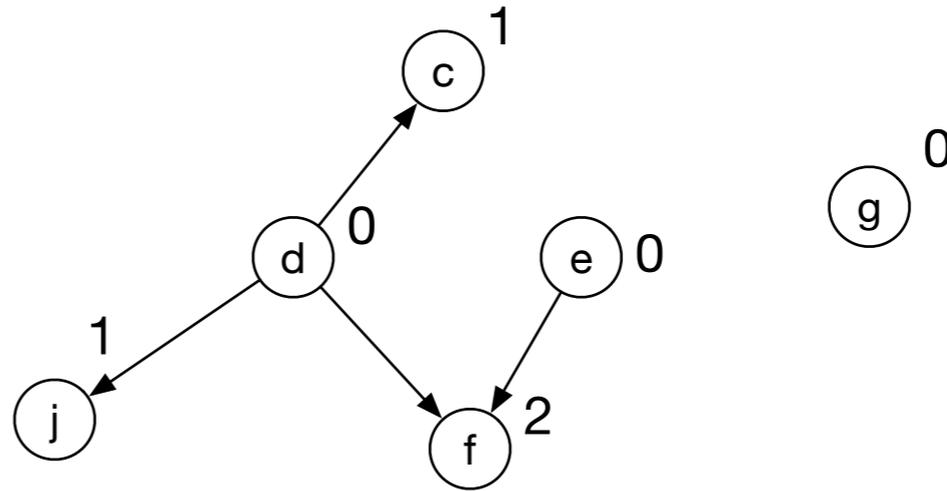
f:

g:

h: e

i: g

j:



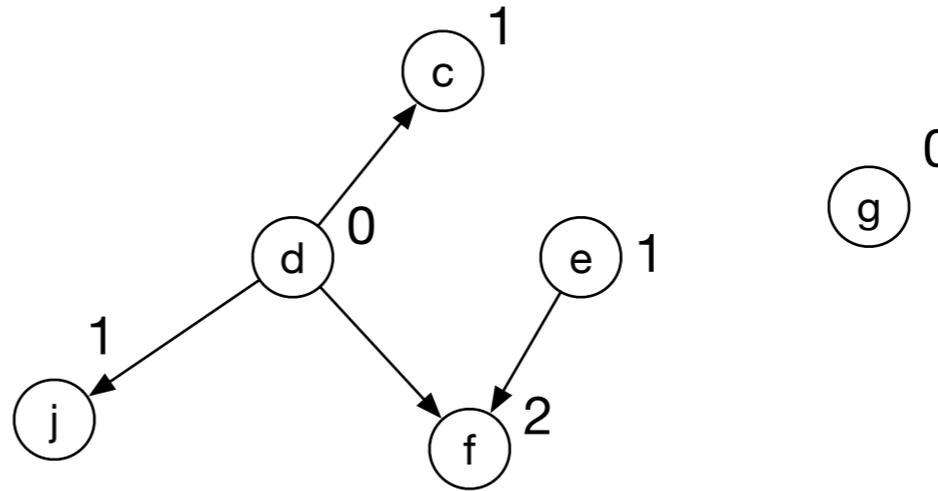
- Need to update in-degree of e

- $\{a, b, i, g, h\}$

# Topological Sort

- Example:

```
a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:
```



- There are two nodes with in-degree 0, let's pick d

# Topological Sort

- Example:

a: b, c, h

b: d, j

c:

d: c, j

e: f

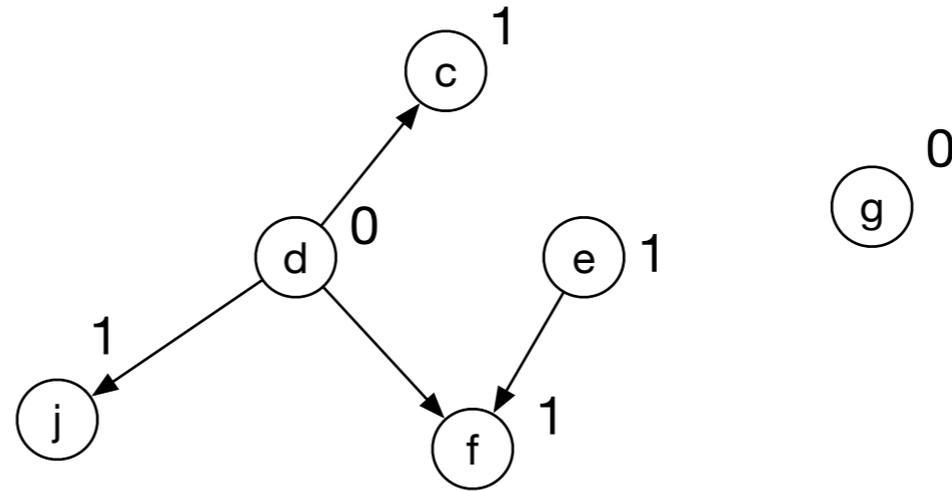
f:

g:

h: e

i: g

j:

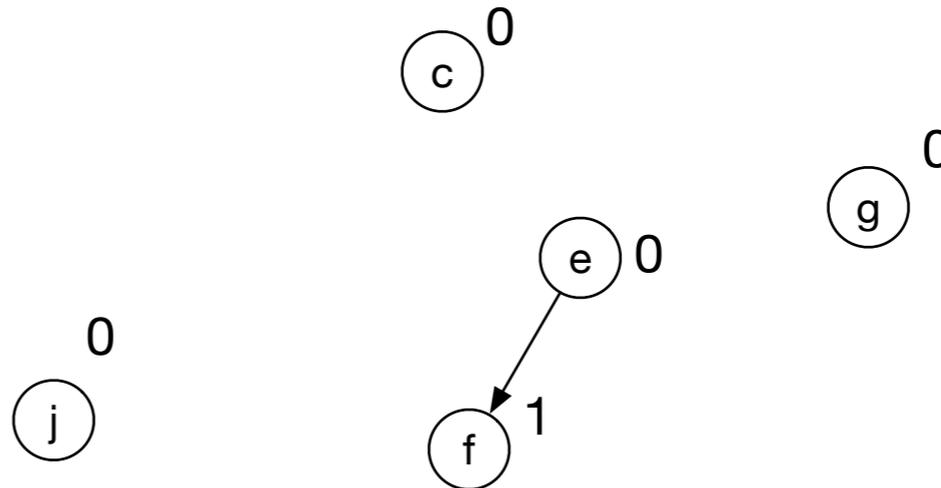


- $\{a, b, i, g, h, d\}$

# Topological Sort

- Example:

```
a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:
```

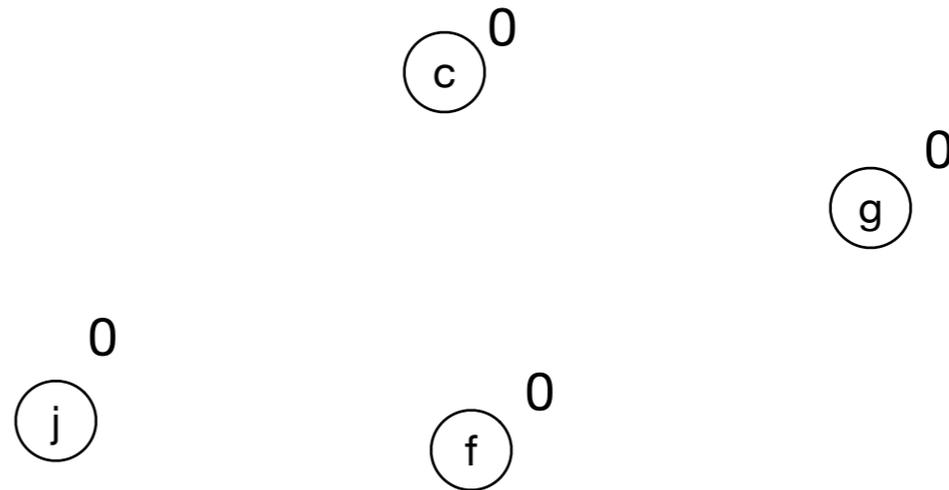


- $\{a, b, i, g, h, d\}$
- Can pick among four nodes: e

# Topological Sort

- Example:

```
a: b, c, h  
b: d, j  
c:  
d: c, j  
e: f  
f:  
g:  
h: e  
i: g  
j:
```



- $\{a, b, i, g, h, d, e\}$
- Can pick among four nodes in any order

# Topological Sort

- Example:

a: b, c, h

b: d, j

c:

d: c, j

e: f

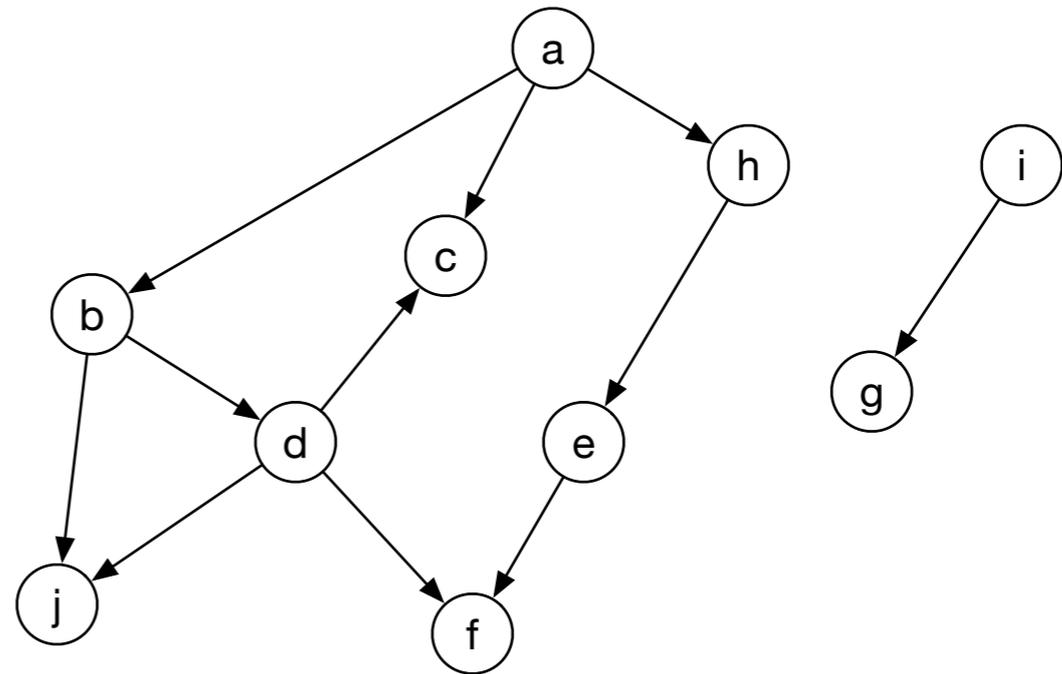
f:

g:

h: e

i: g

j:



- $\{a, b, i, g, h, d, e, h, j, f\}$

- Can pick among four nodes in any order

# Topological Sort

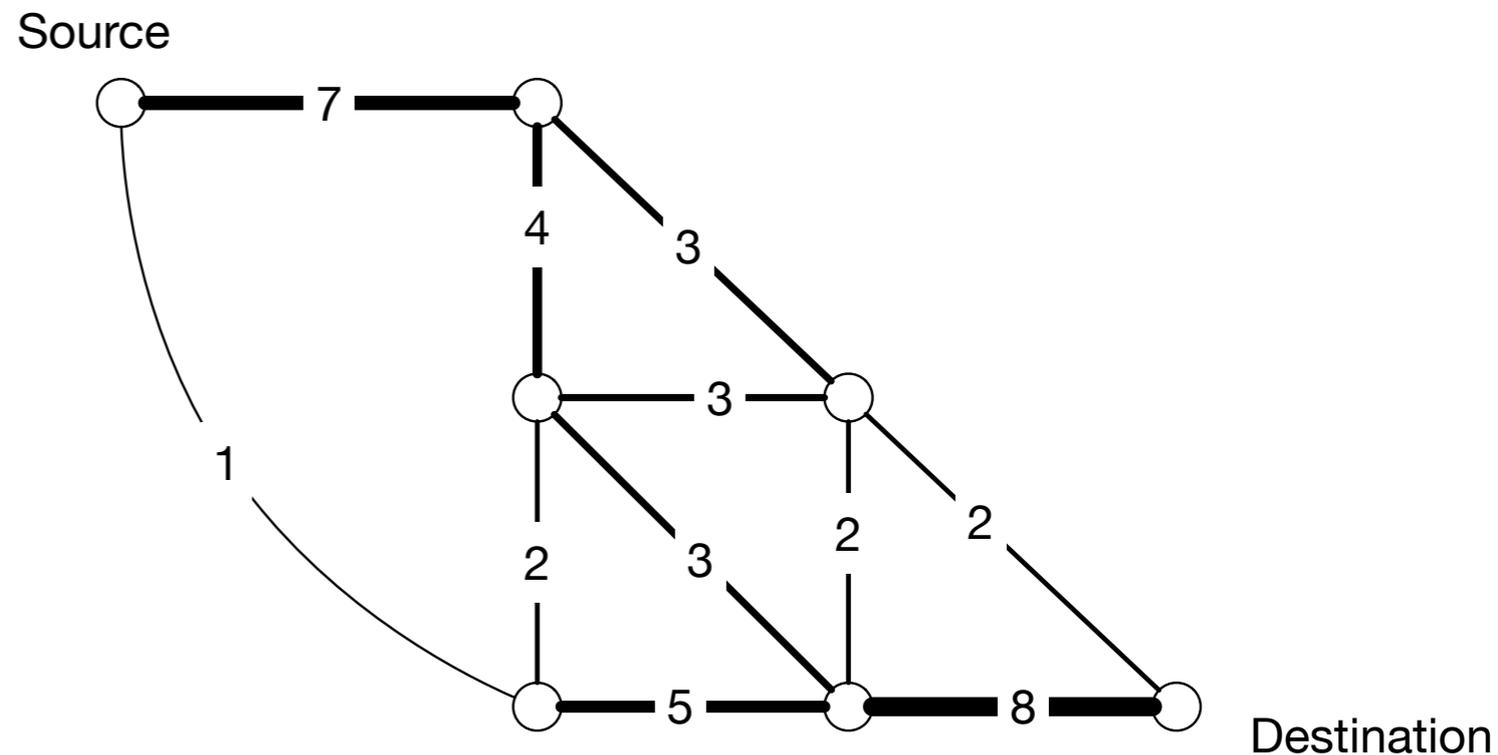
- Analysis for topological sort on  $G = (V, E)$ 
  - Need to establish in-degrees:
    - Process all elements in an adjacency list
      - Correspond to edges
      - work  $\sim |E|$
  - For each vertex:
    - find the vertex as a vertex of minimum in-degree
    - update in-degrees by going through the adjacency list
    - Latter work is  $\sim |E|$  because we process each adjacency list entry once
    - Delete the adjacency list
    - Work is  $\sim V$

# Topological Sort

- This algorithm is *almost*  $O(|E|)$  but for finding the minimum in-degree
- We will see a better algorithm shortly

# Weighted Graphs

- Graphs with edge weights
  - Often, graphs in CS have edge weights
    - Example: edge weight indicates the size of a pipeline
      - such as network connection, capacity of roads, etc.



How much can you pump from source to destination if the pipes have the indicated capacities (Flow Problem)

# Weighted Graphs

- Graphs with edge weights
  - Weights can indicate distance
    - What is the shortest distance from source to destination

