

Graphs

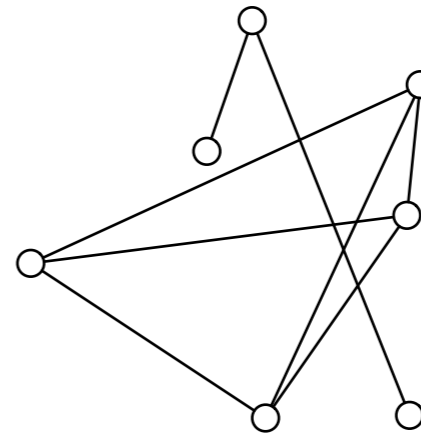
Thomas Schwarz, SJ

Graph Definition

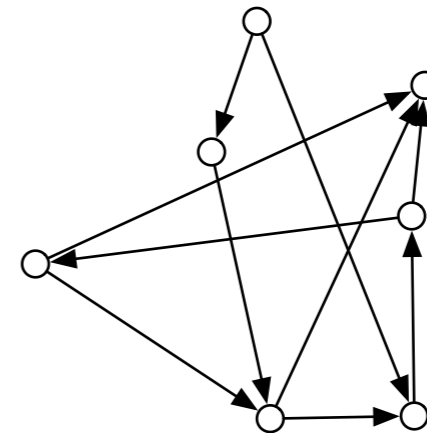
- A graph has a set of vertices V and a set of edges.
 - Directed edges are pairs (u, v) with $u, v \in V$
 - Undirected edges are two-sets $\{u, v\}$ with $u, v \in V$
- A graph with directed edges is called a directed graph
- A graph with undirected edges is just called a graph

Graph Definition

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



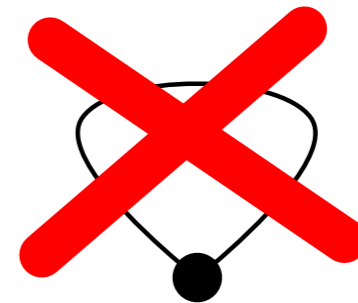
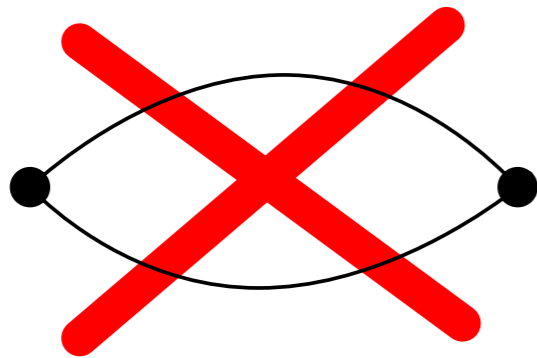
An undirected graph with 7 vertices and 7 edges



A directed graph

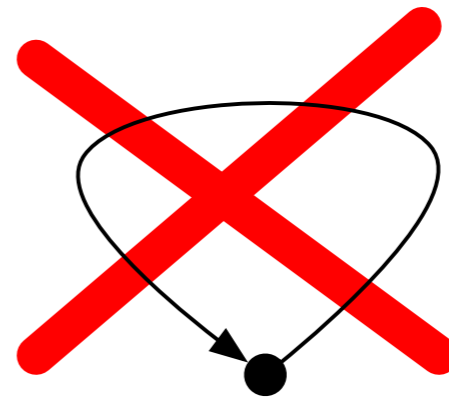
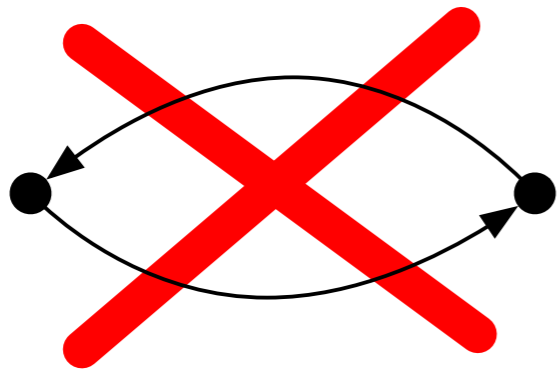
Graph Definition

- 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



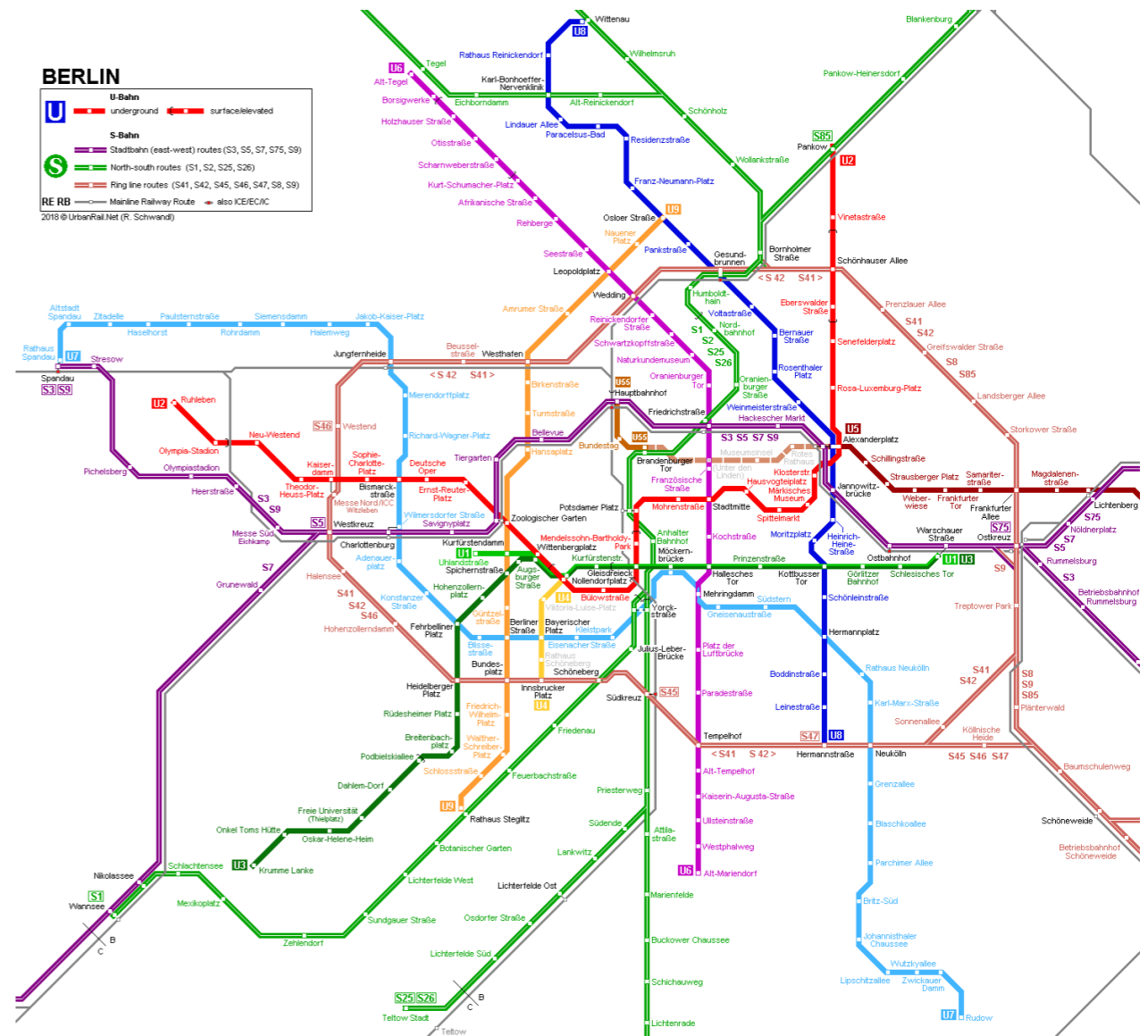
Graph Representations

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



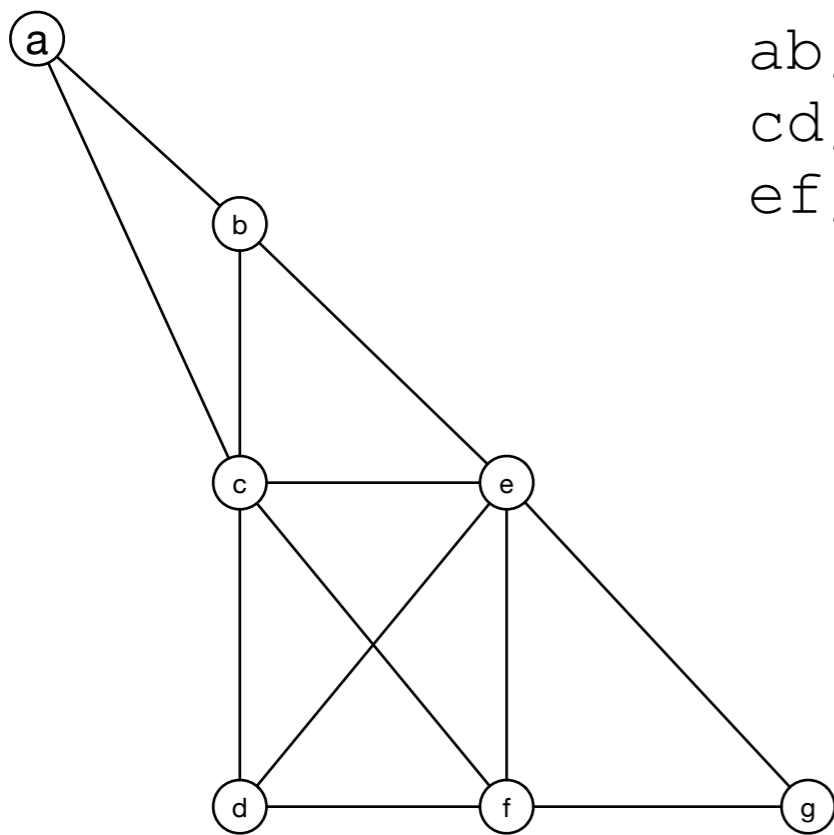
Graph Representations

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

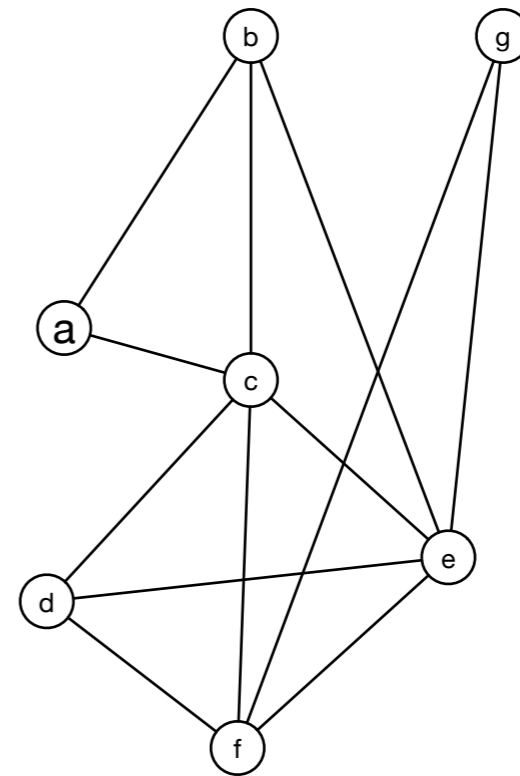


Graph Definition

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

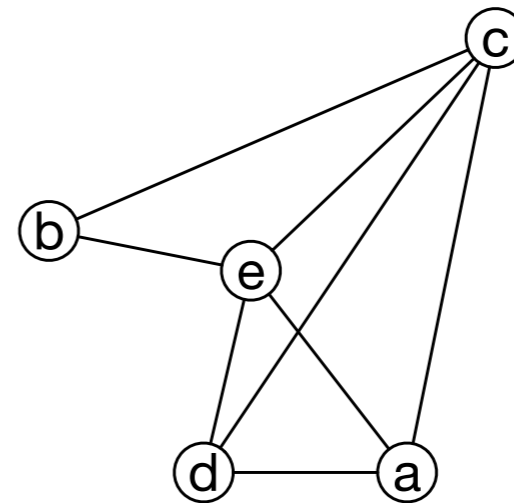
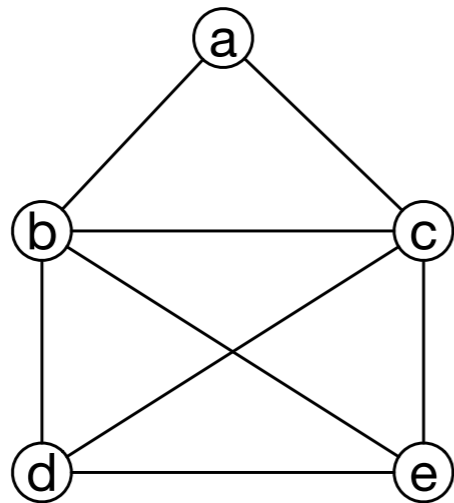


ab, ac, bc, ce
cd, cf, de, df
ef, eg, fg



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

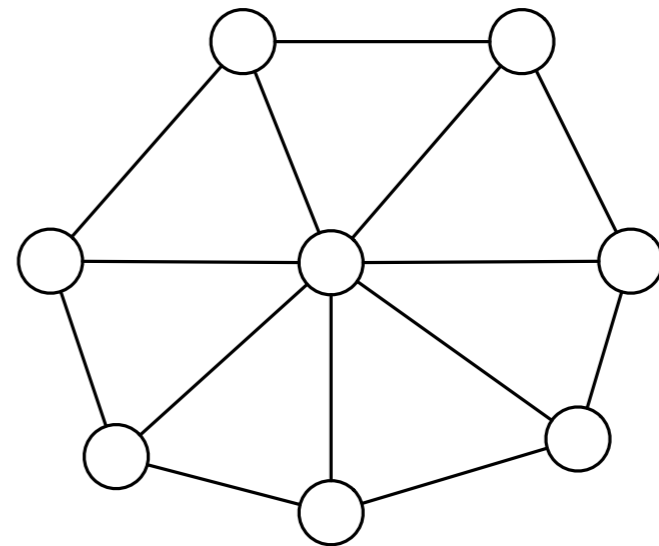
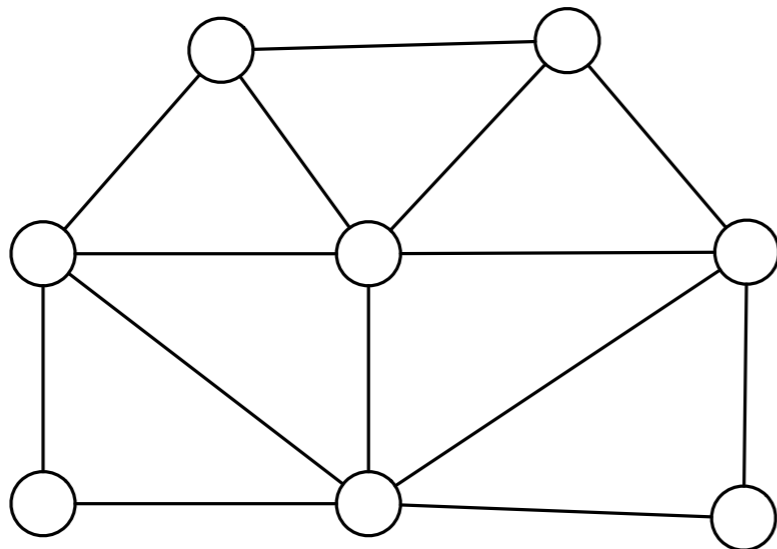
$G = (V, E)$ is isomorphic to $G' = (V', E')$



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

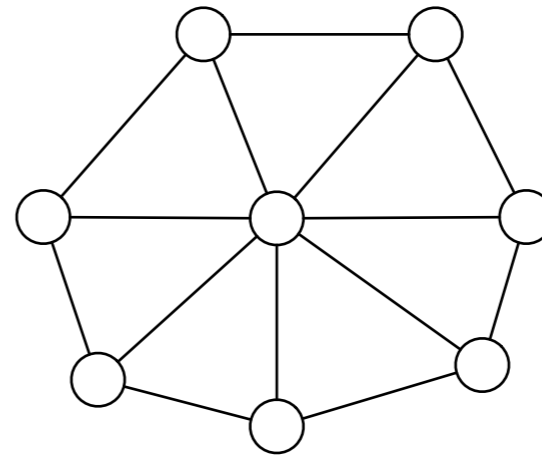
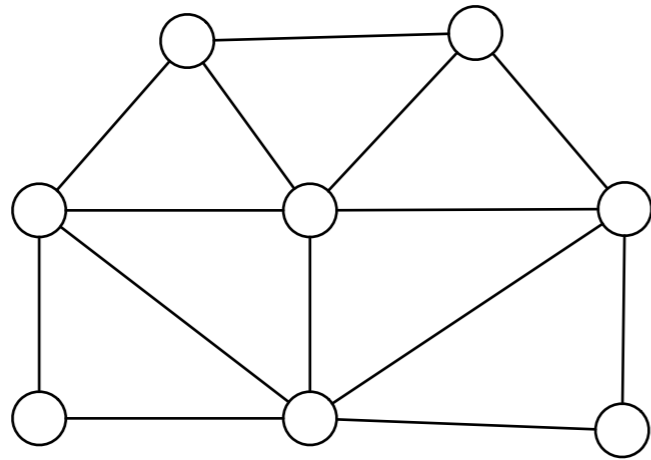
Graph Definition

- 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

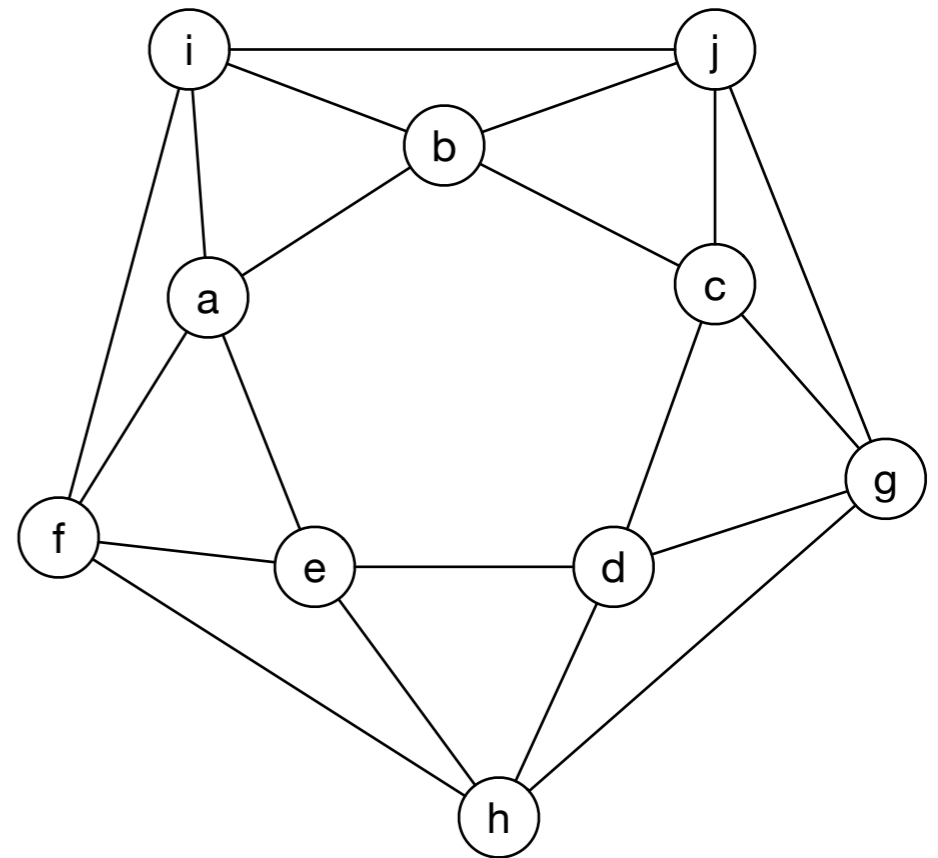
Graph-Definition

- Vertex degree: Number of edges incident to a vertex
- Vertex degree vector: Made up of vertex degrees of all vertices
- *Handshake Lemma*: The sum of vertex degrees is twice the number of edges
 - Proof: Each edge contributes twice to the sum of the vertex degrees
- Number of odd vertices (vertex degree is odd) is even

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
```

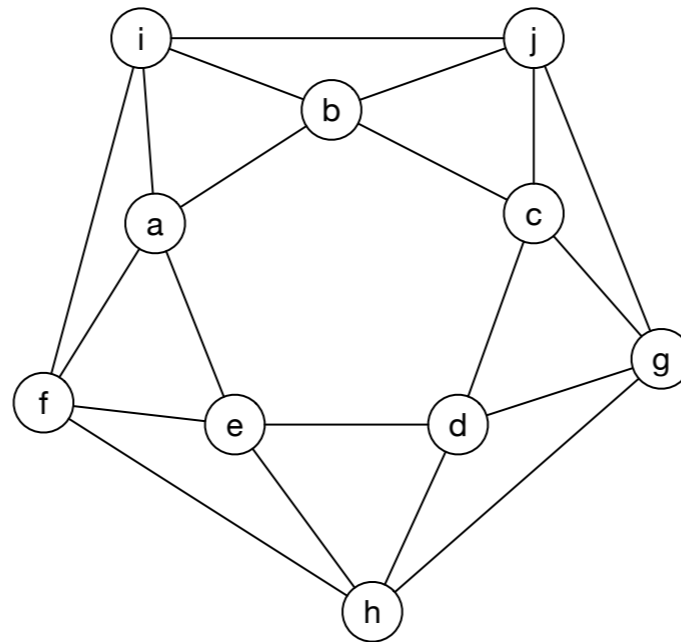


Graph Representations

- 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	i
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
i	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

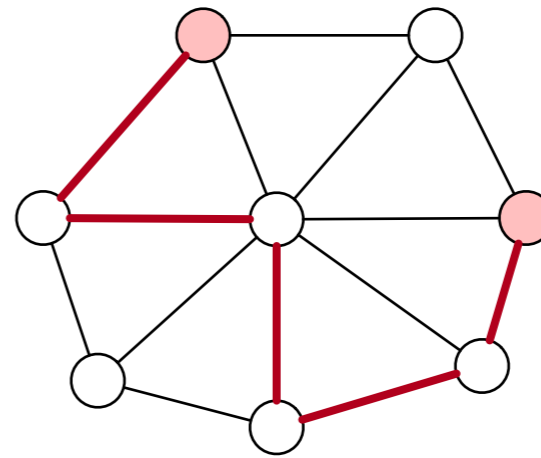
- Graph $G = (V, E)$ with vertices V and edges E
 - Graph algorithms usually need to look at each edge at least once
 - there are some idiosyncratic exceptions
 - They usually run in time at least $\Theta(|V|^2)$
 - However, many important graphs are sparse:
 - No edge between most pairs of vertices

Graph Definitions

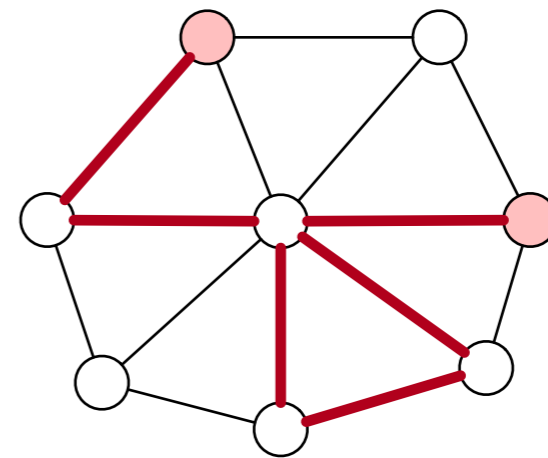
- There are a number of important properties of graphs
 - No need to learn them by heart, the ones used in CS will get repeated over and over again
 - A path between two vertices $u, w \in V$ of a graph $G = (V, E)$ is a list of vertices $u = v_0, v_1, \dots, v_{n-1}, v_n = w$ such that there is an edge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

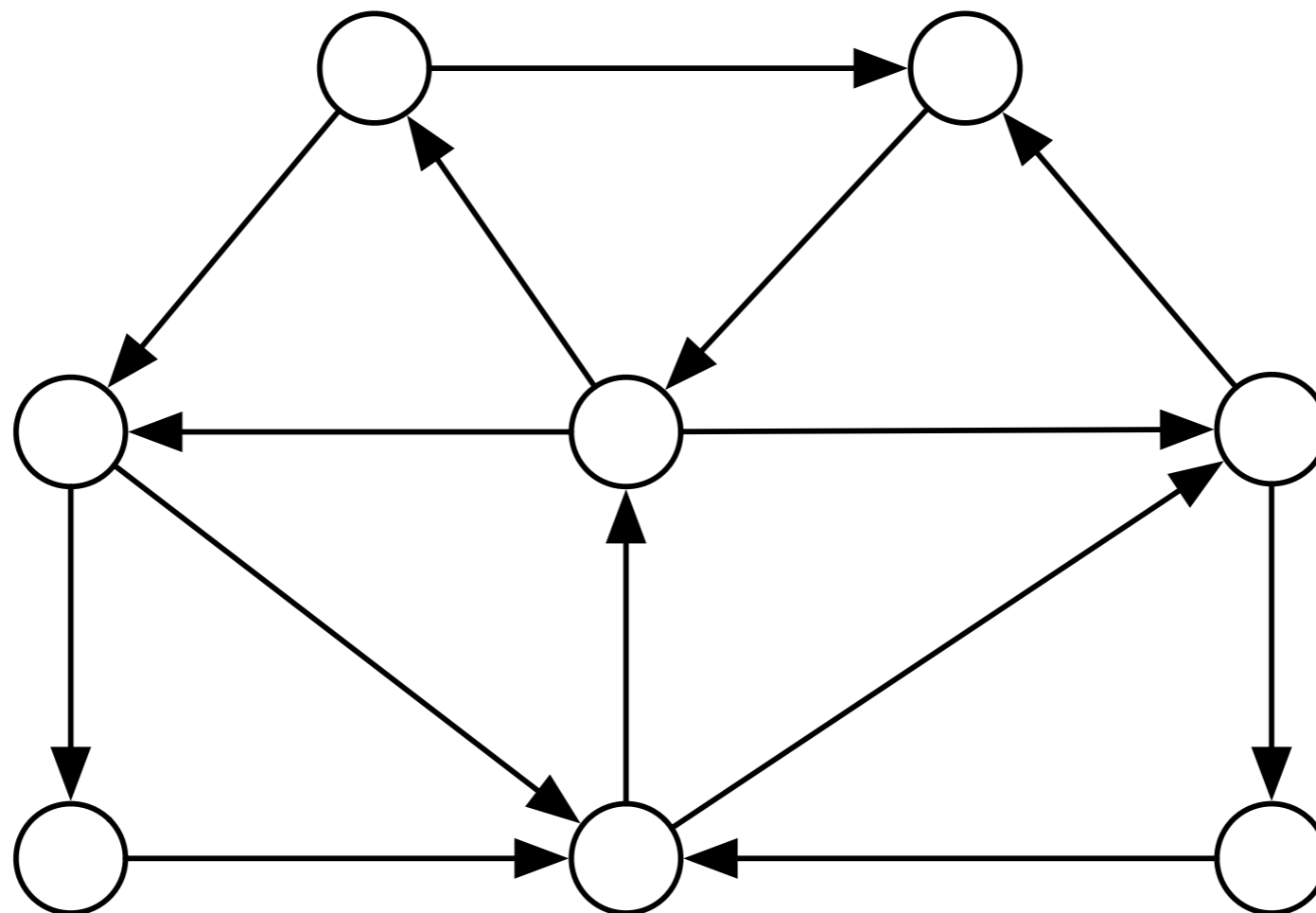


Graph Definitions

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

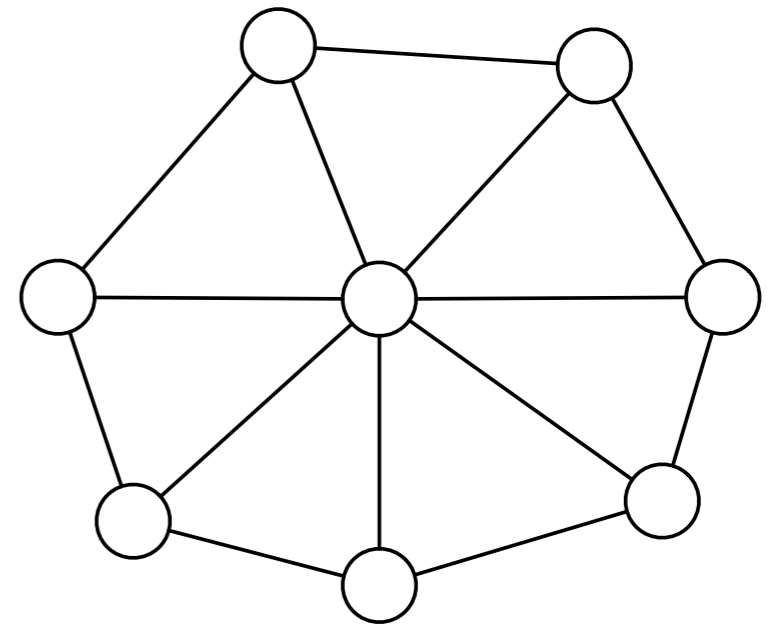
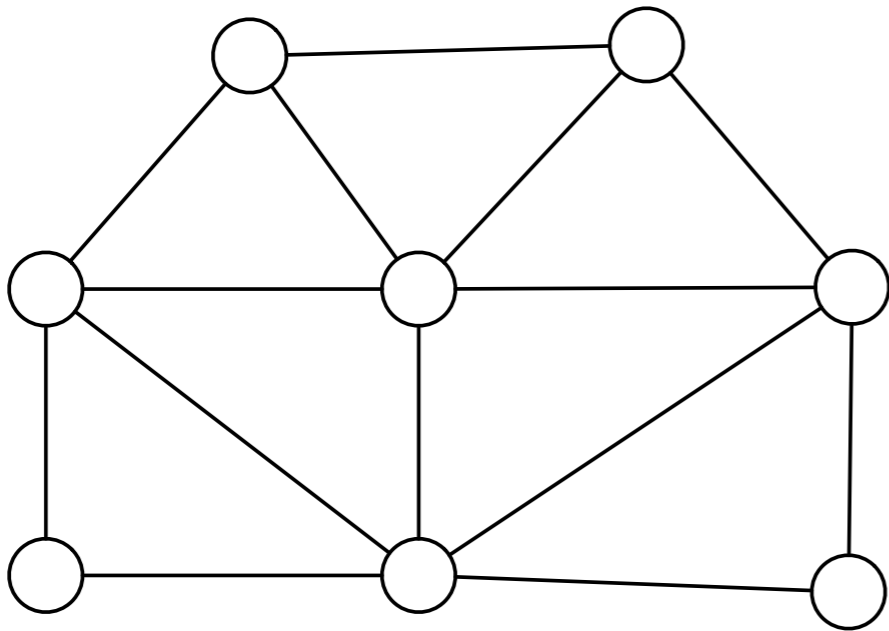
Graph Definitions

- 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)

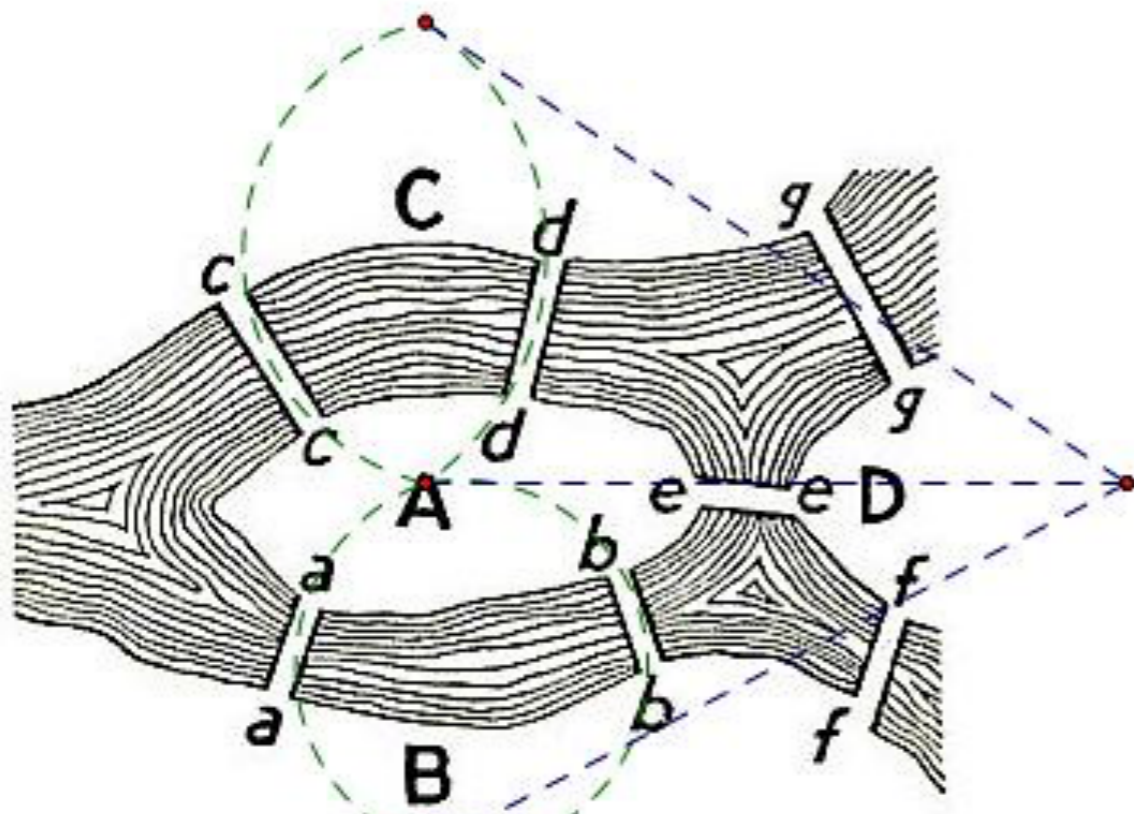
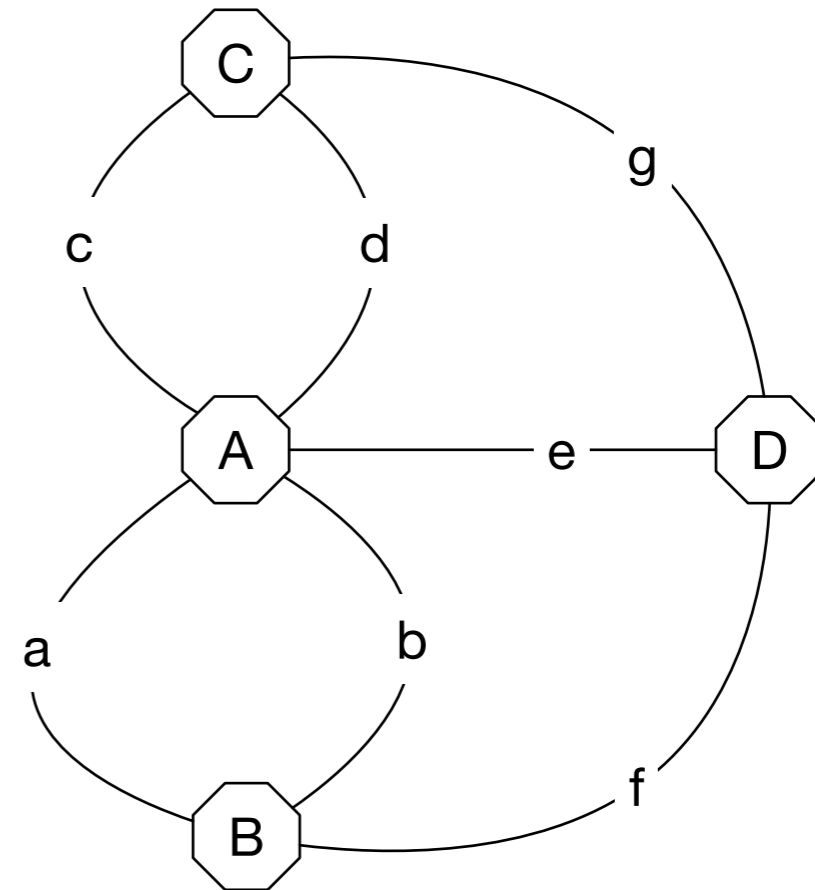
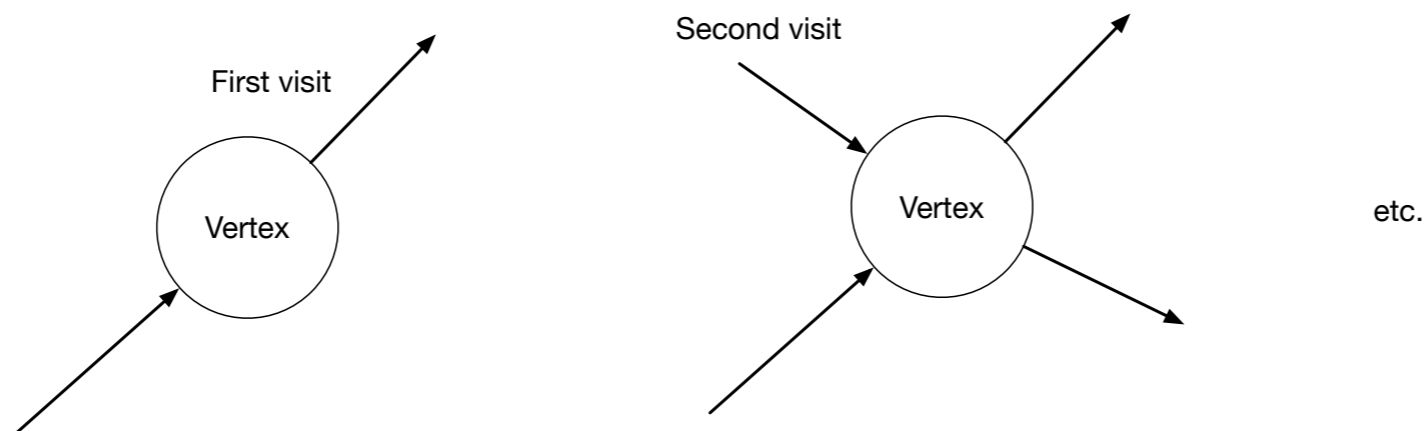


FIGURE 98. *Geographic Map:
The Königsberg Bridges.*



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.

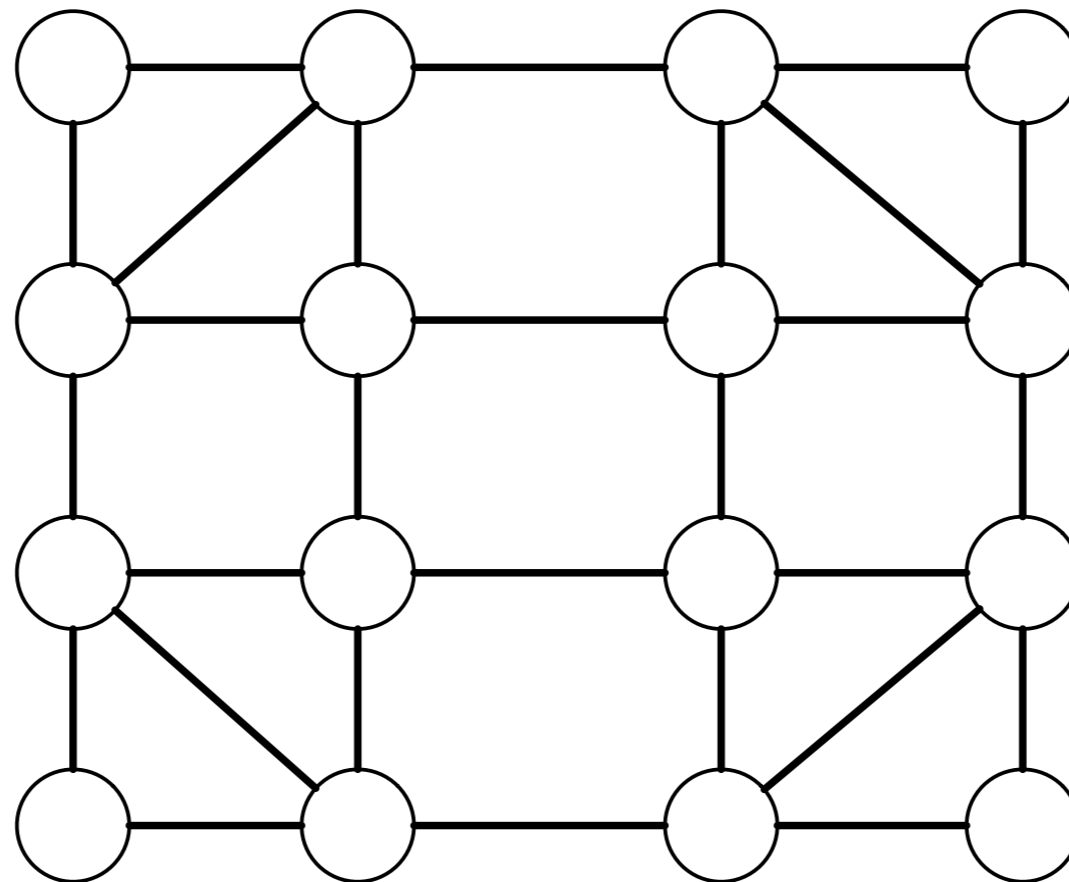


Euler Tours

- 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

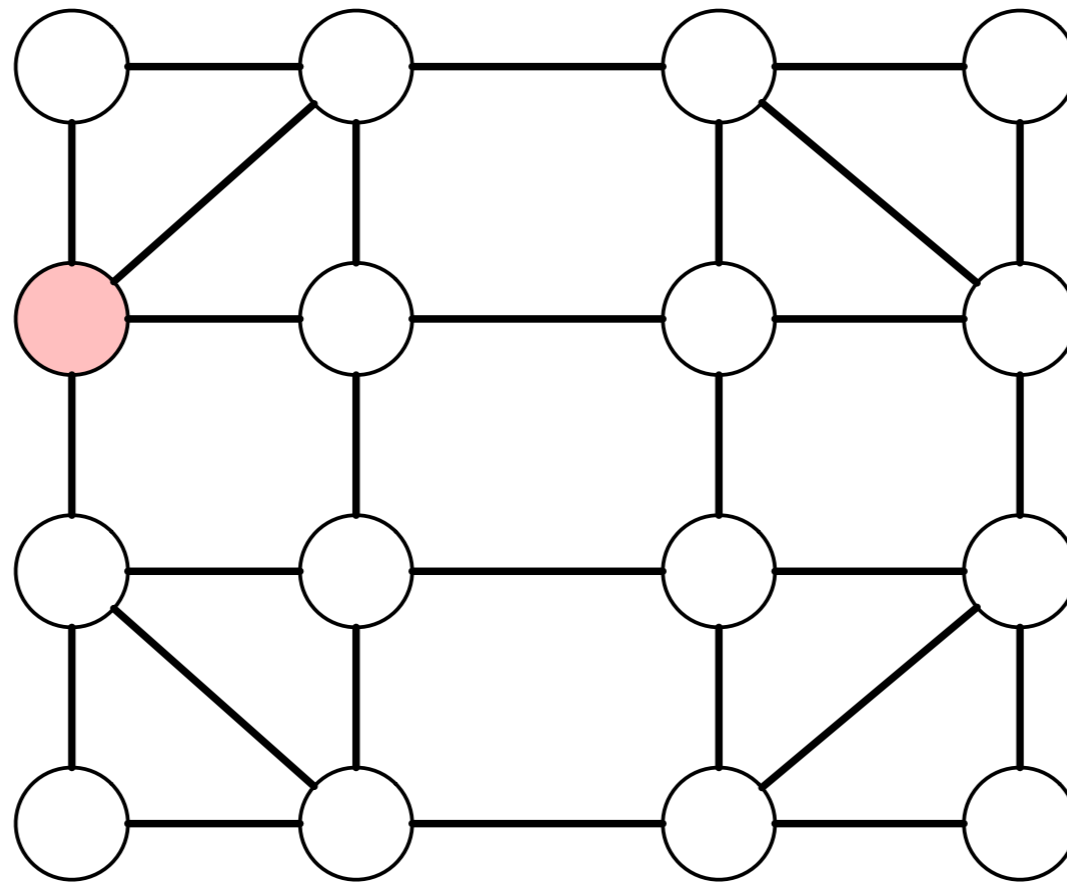
Euler Tours

- Example



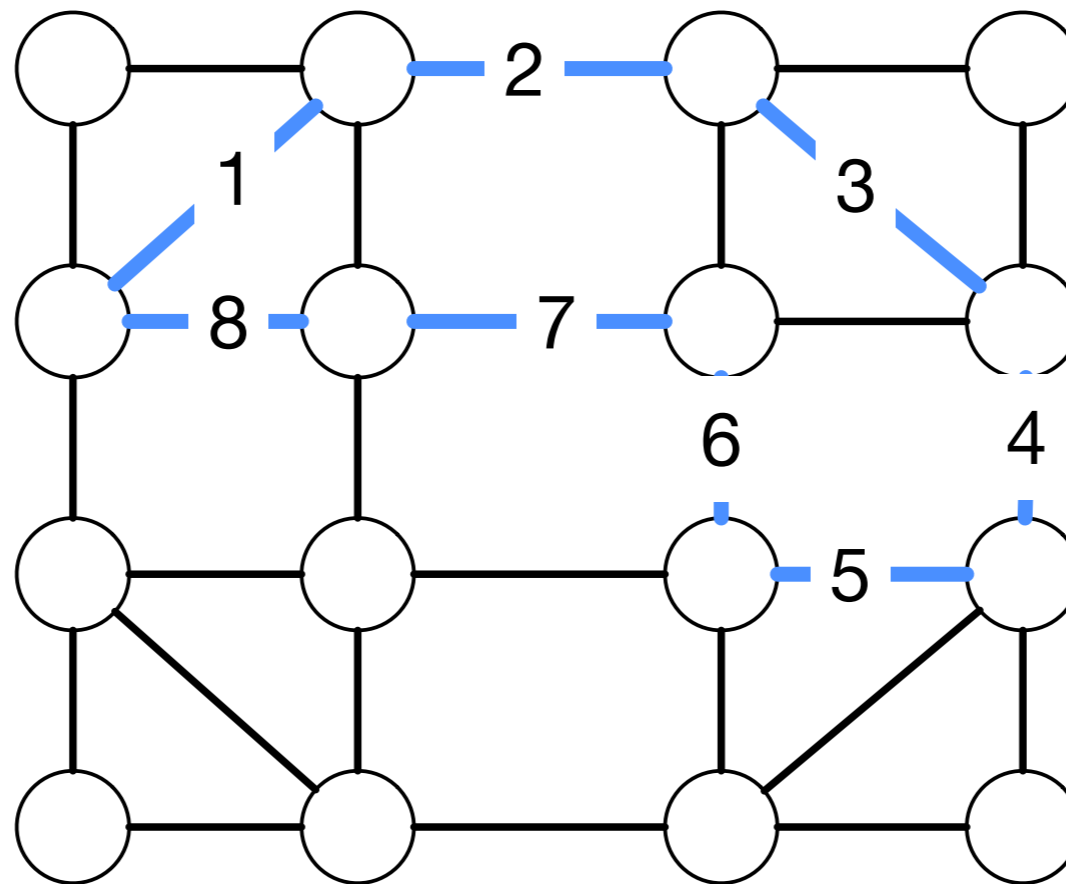
Euler Tours

- Start at a random vertex



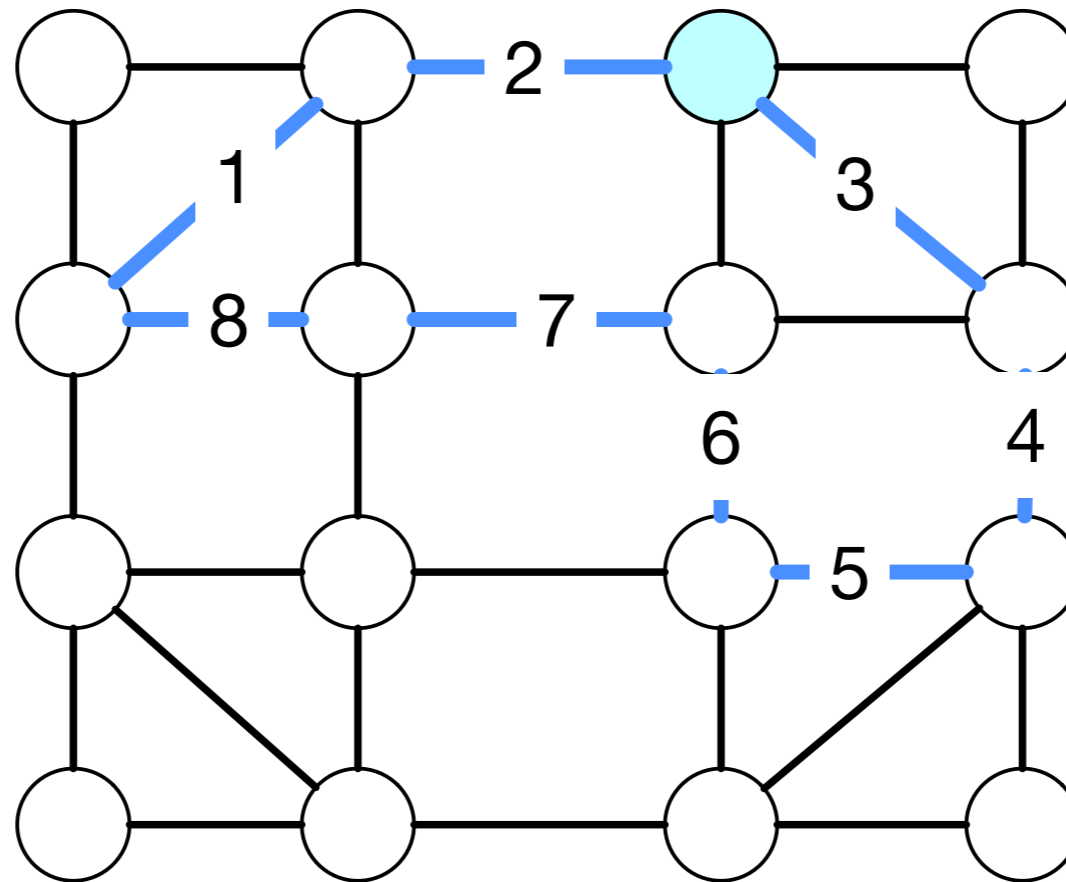
Euler Tours

- Make a tour



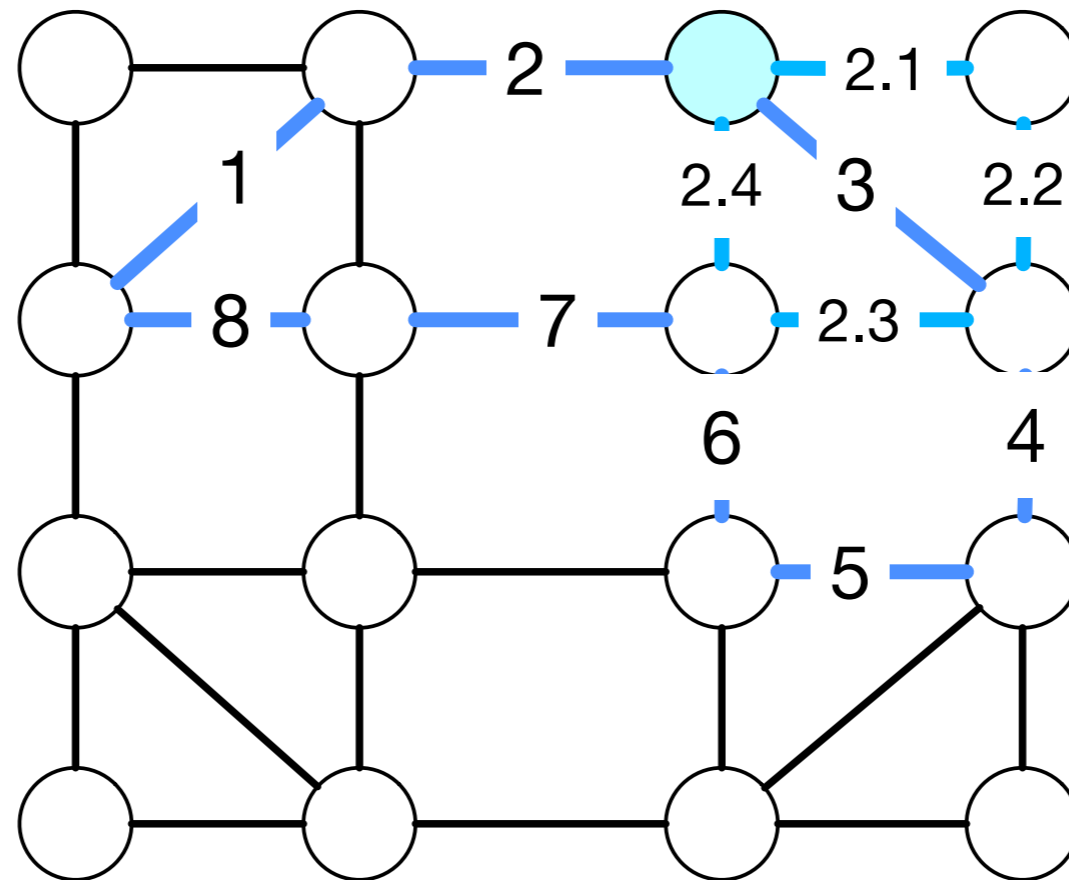
Euler Tours

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



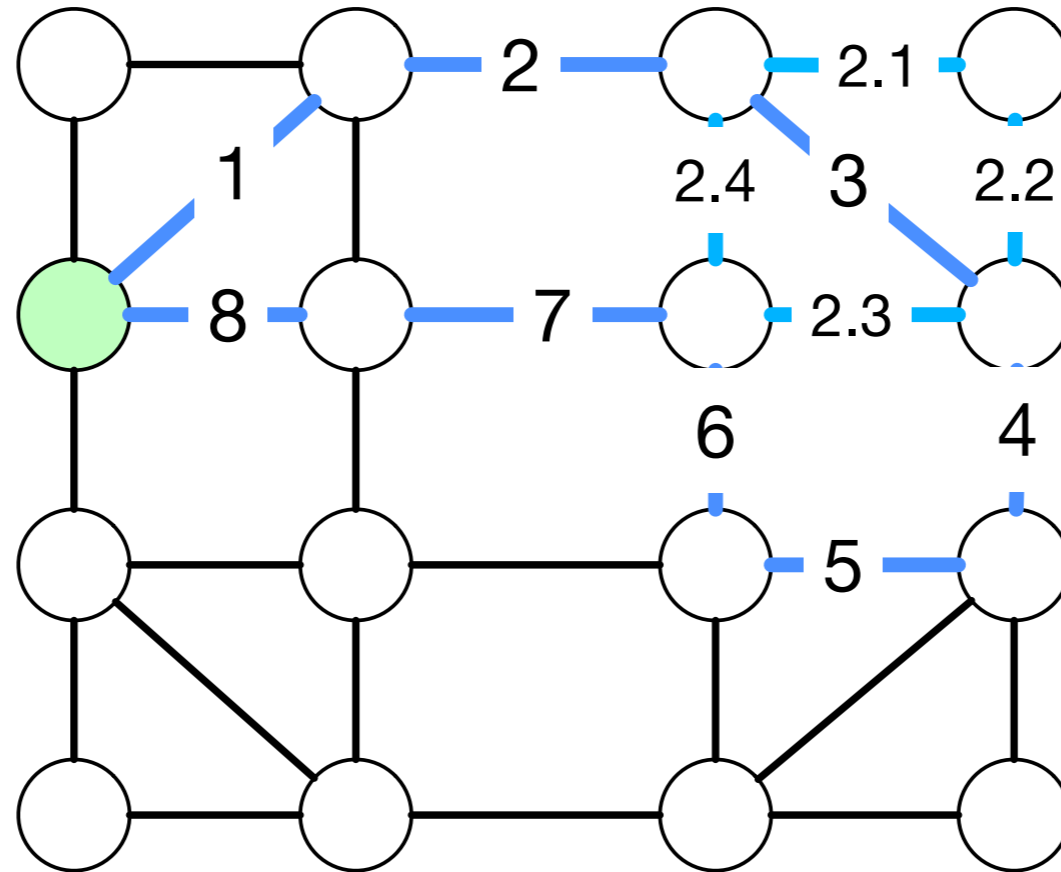
Euler Tours

- Start out creating a random circuit of unused edges



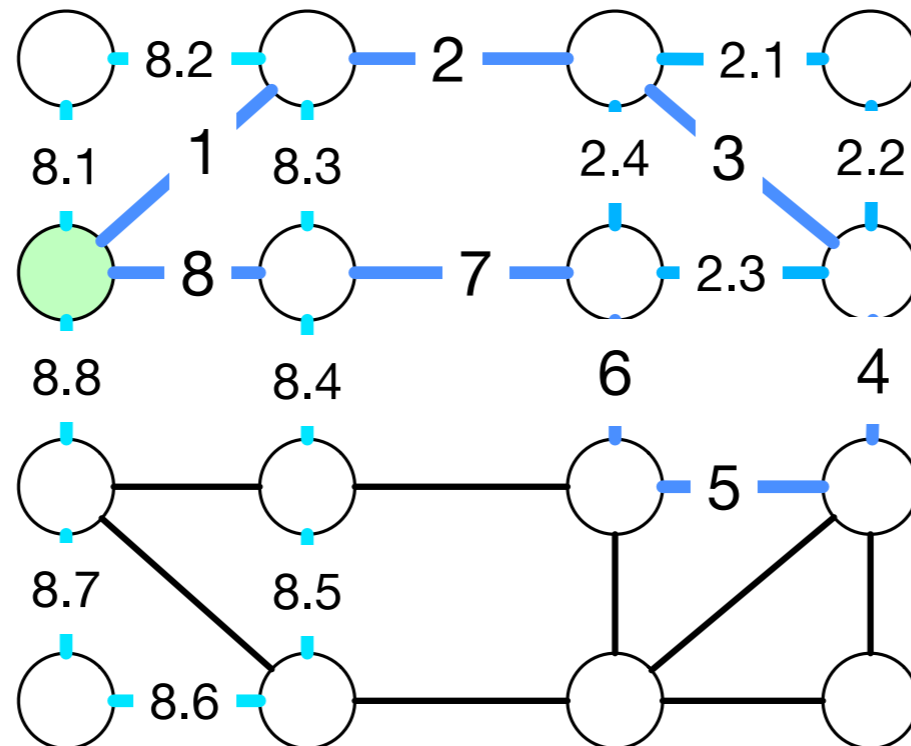
Euler Tours

- Pick another vertex with unused edges



Euler Tours

- Start a new part of the circuit

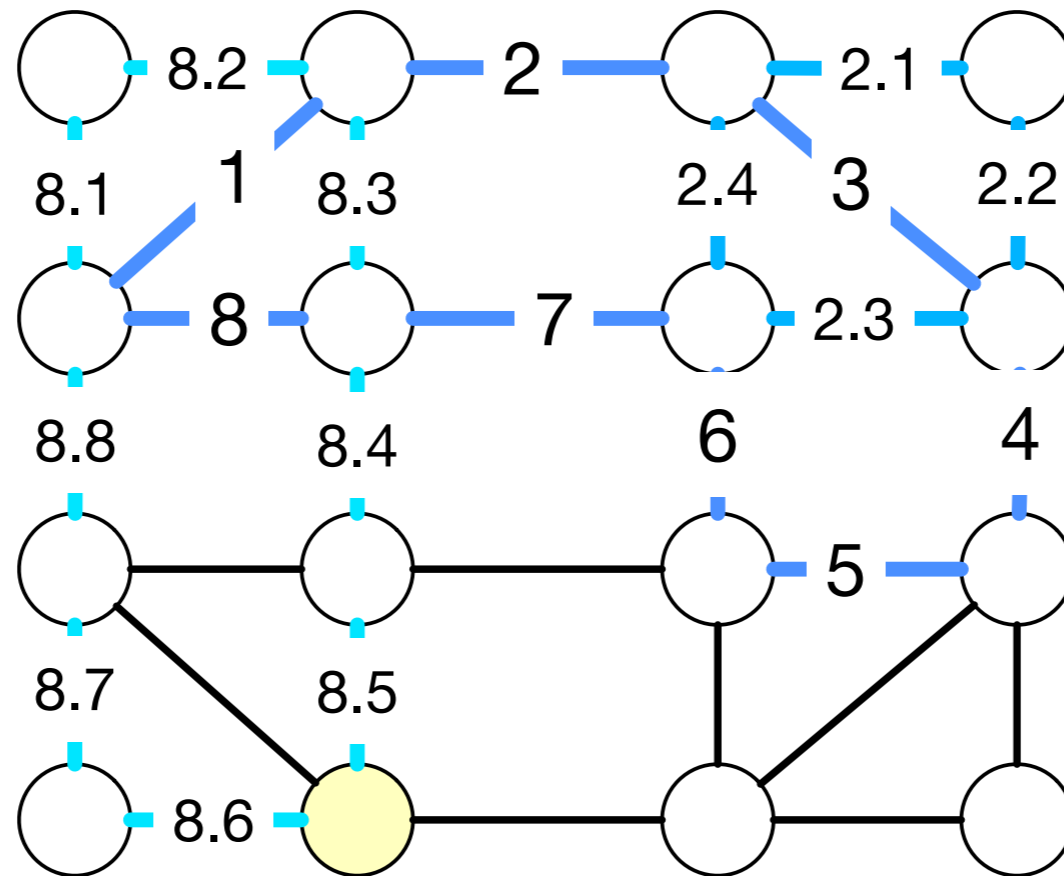


- Circuit so far: 1, 2, 2.1, 2.2, 2.3, 2.4, 3, 4, 5, 6, 7, 8, 8.1, 8.2, 8.3, 8.4, 8.5, 8.6, 8.7, 8.8

Euler Tours

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

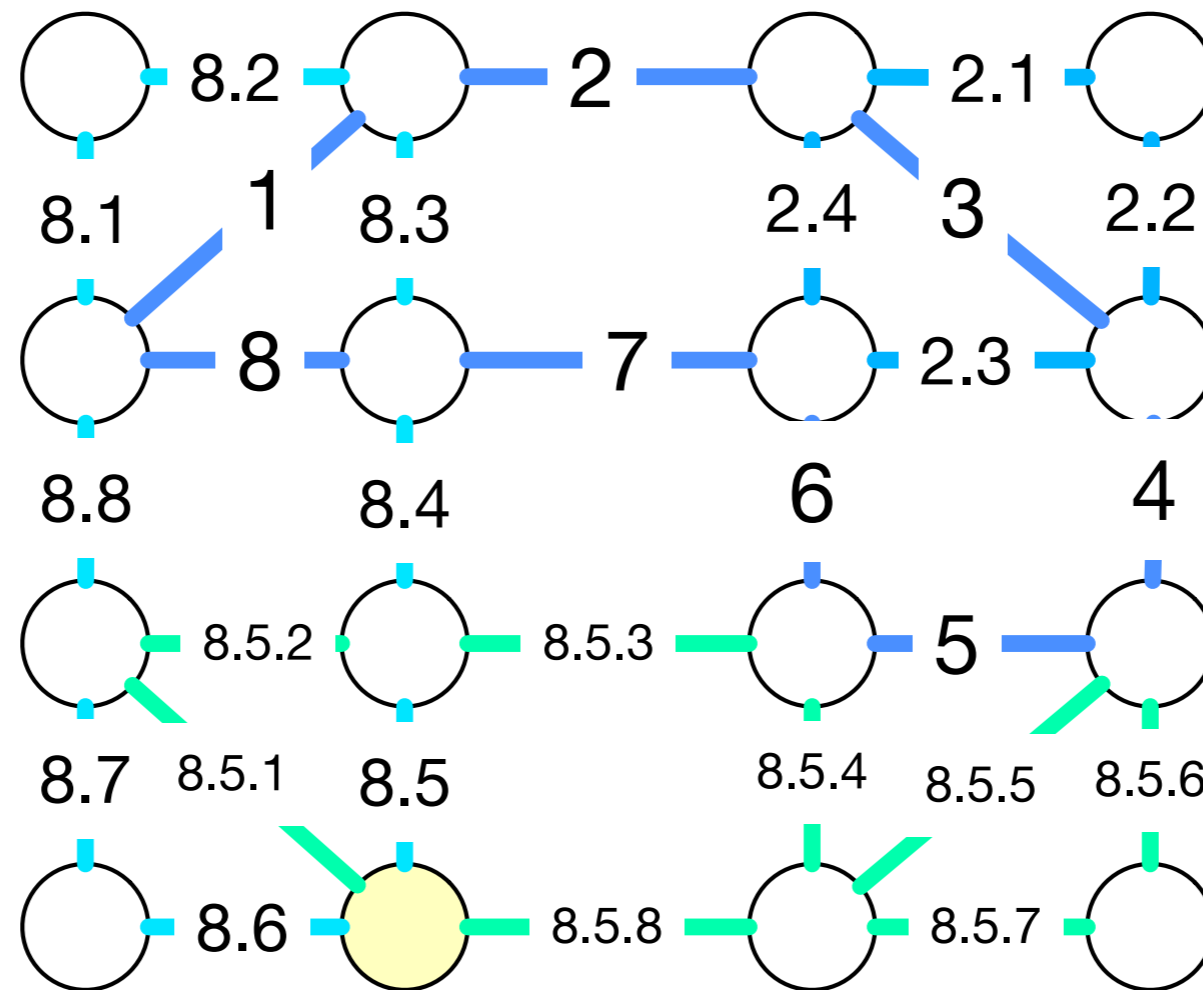
- Pick one



- 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

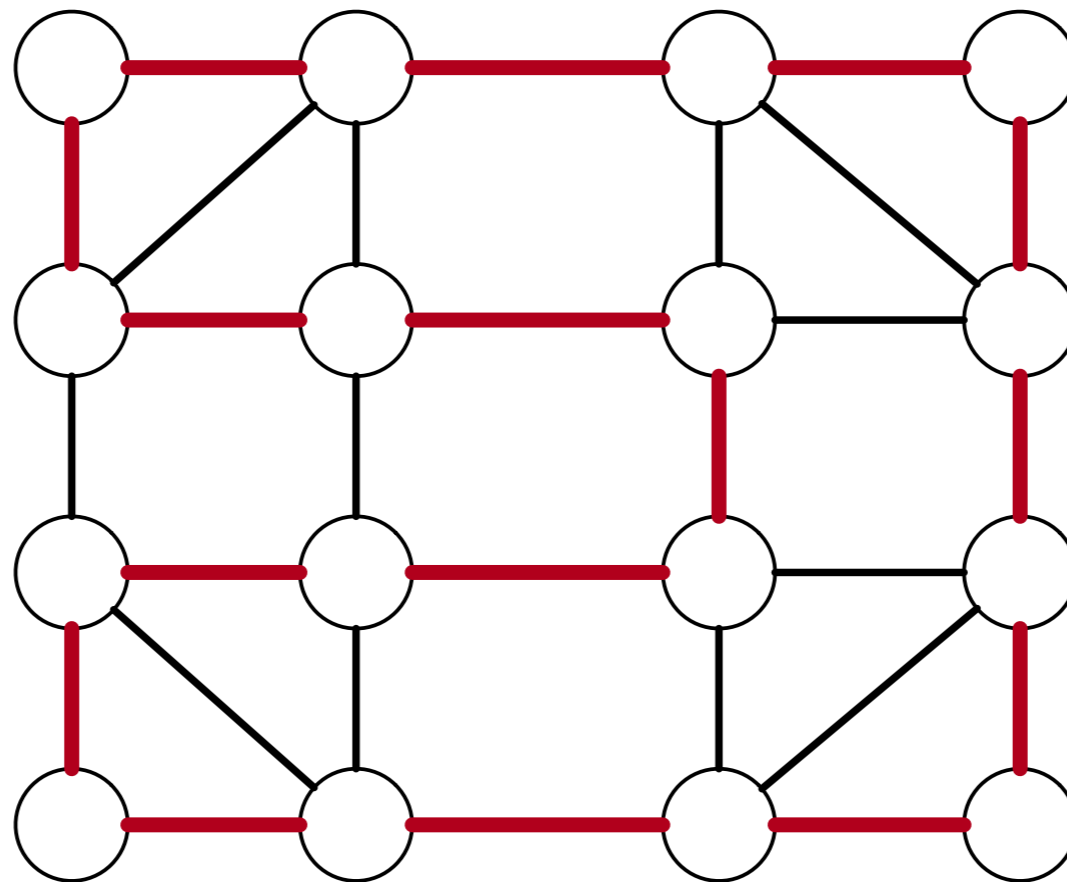
- And after this, we are done



- 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

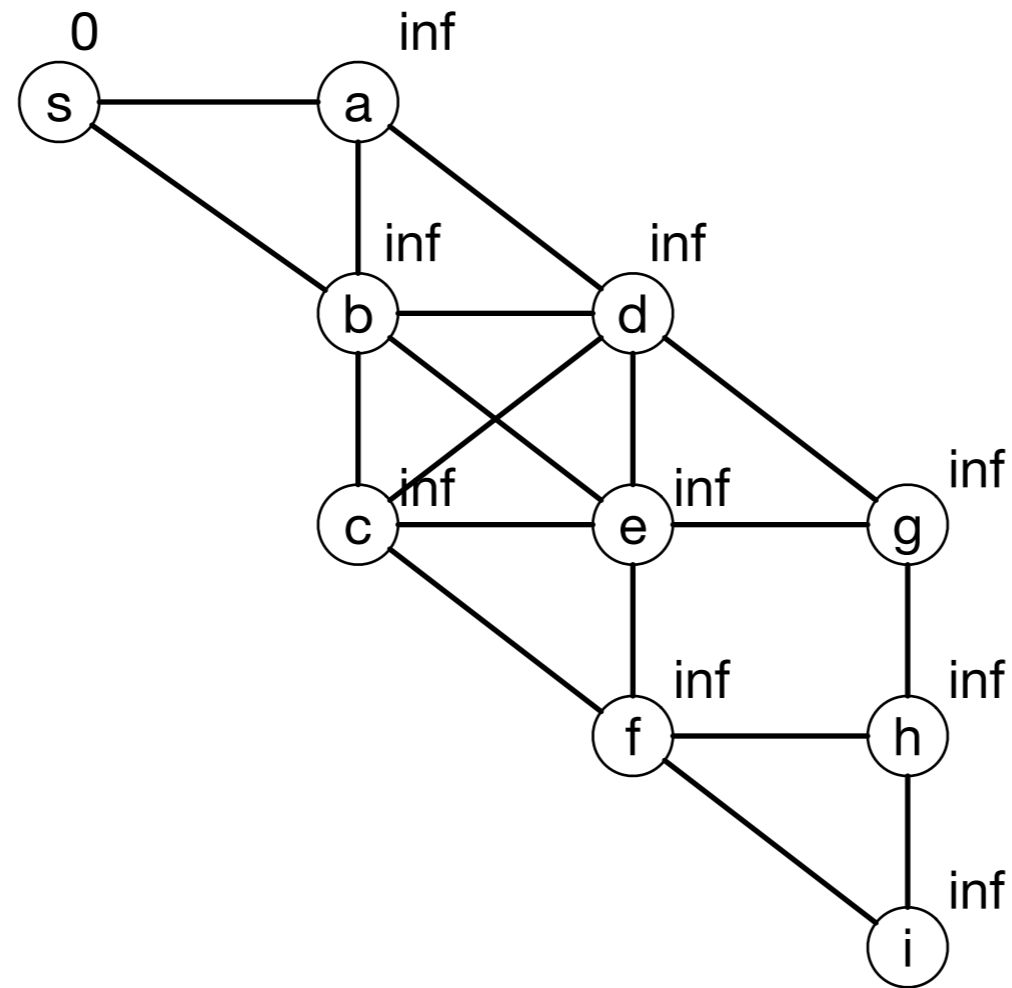
$$\delta(u, w) = \min\{n \mid \exists v_0 = u, v_1, \dots, v_n = w \text{ such that } (v_i, v_{i+1}) \in E \forall i \in \{0, \dots, n-1\}\}$$

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

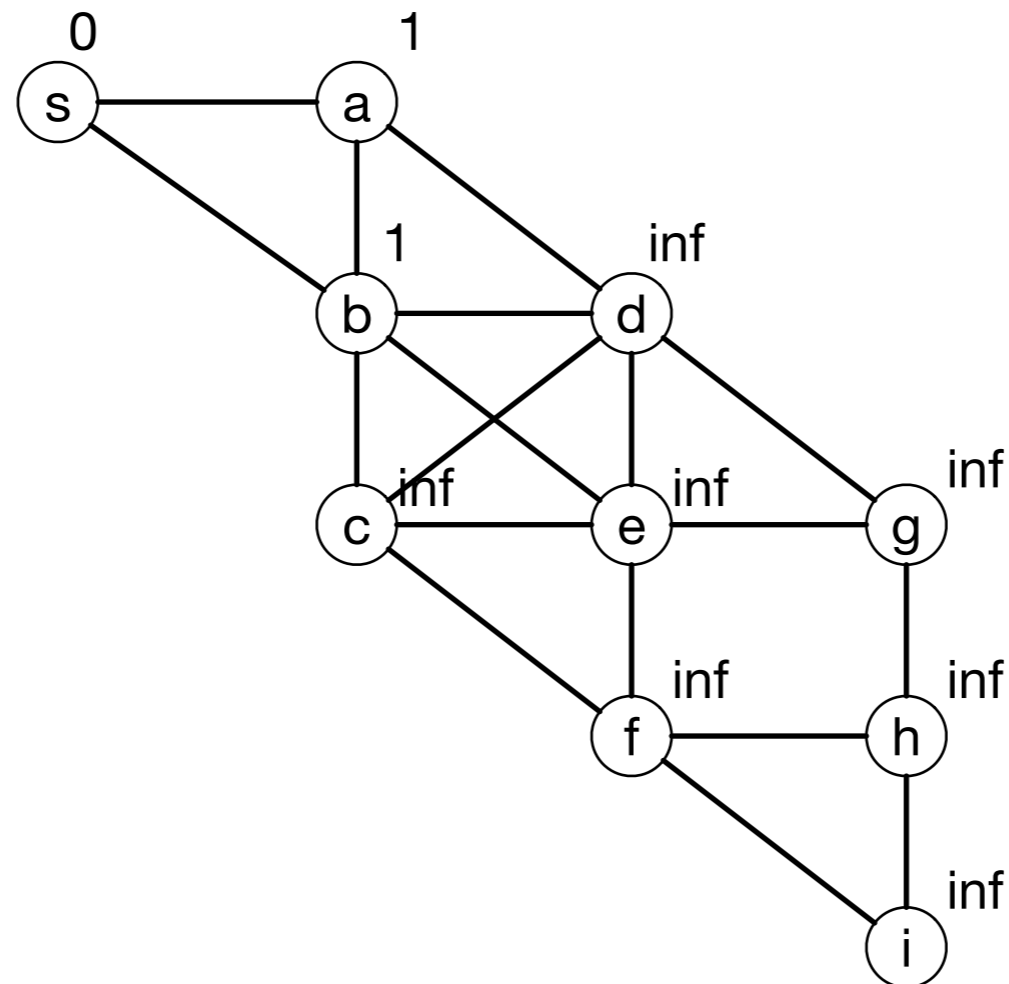
Dijkstra's Algorithm

- Example:



Dijkstra's Algorithm

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

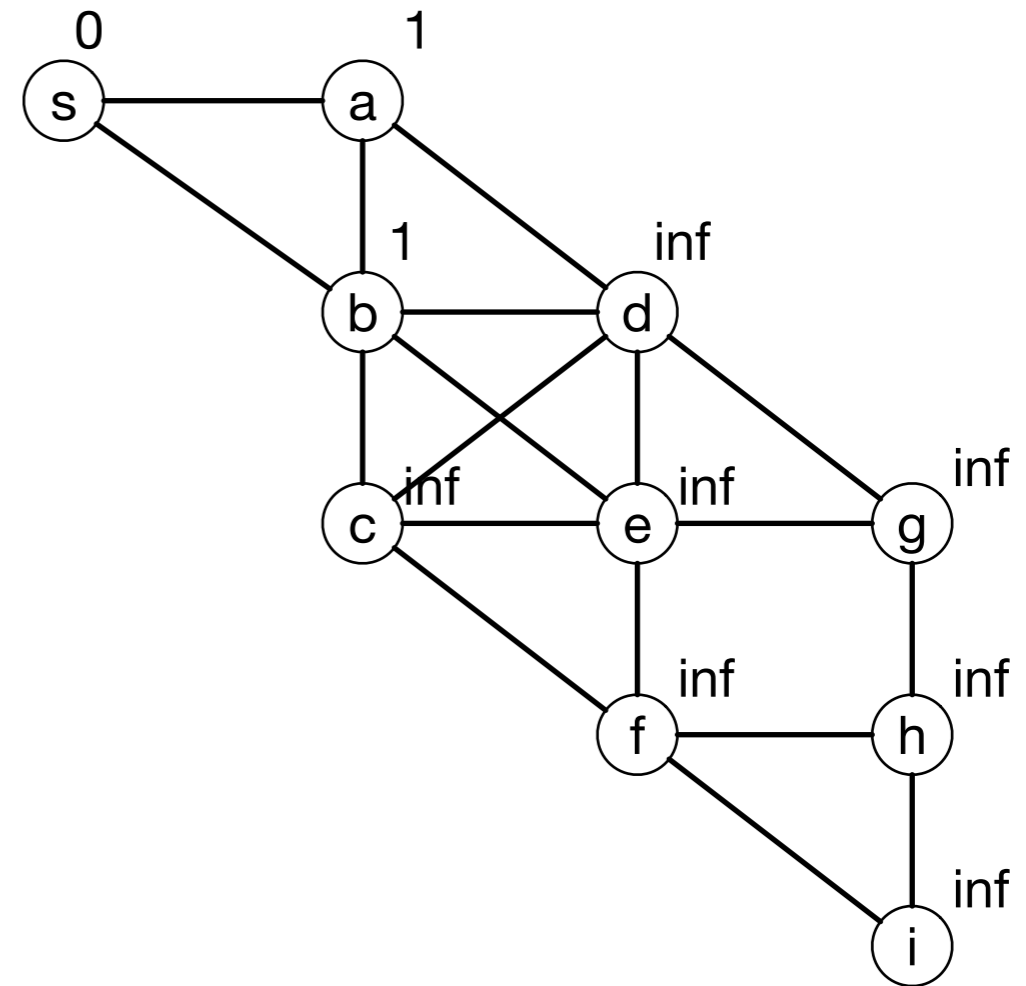


Dijkstra's Algorithm

- 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

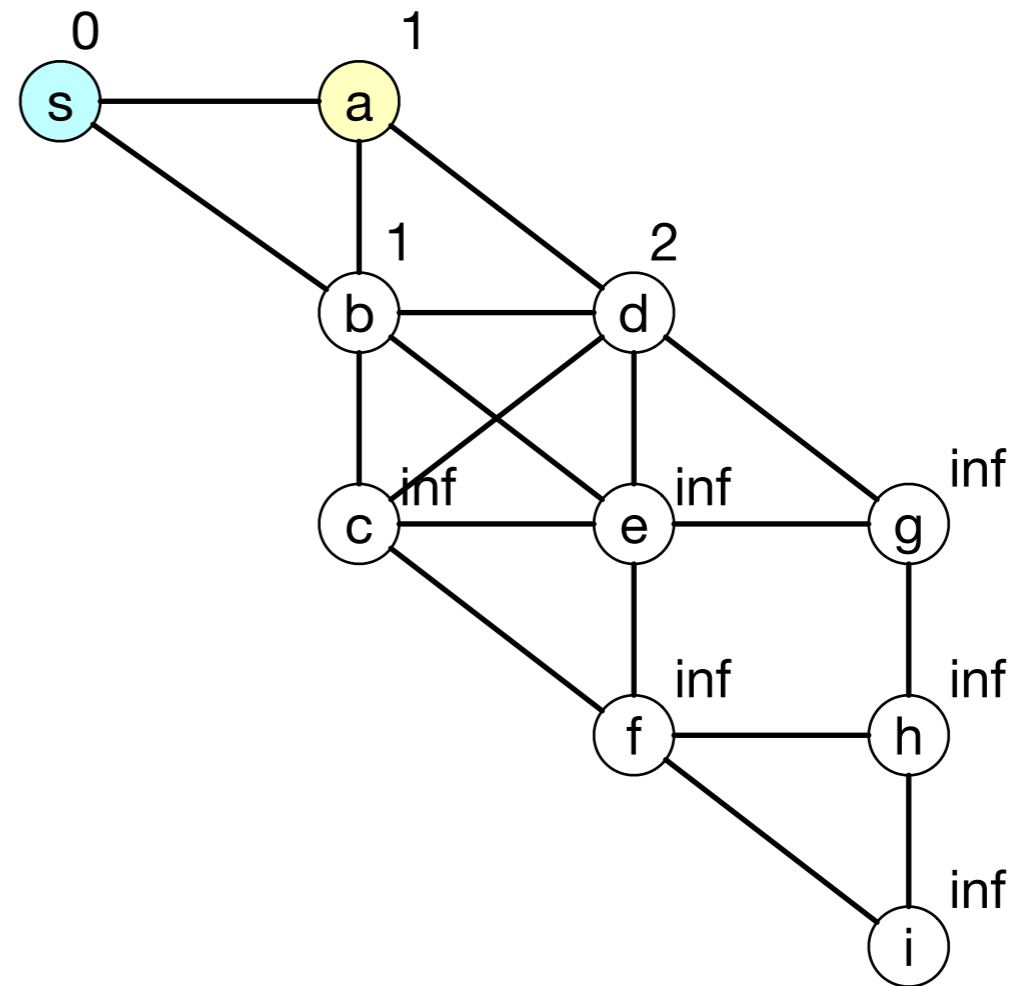
Dijkstra's Algorithm

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



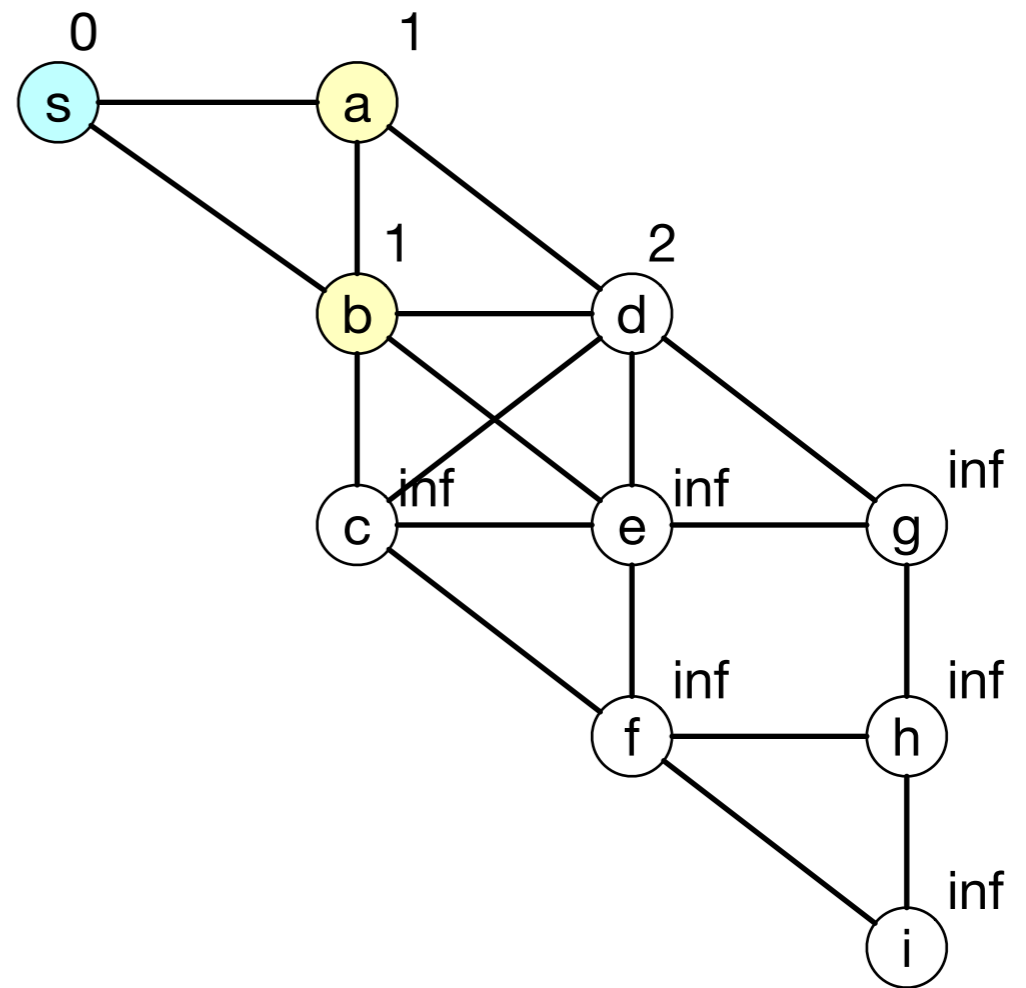
Dijkstra's Algorithm

- 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



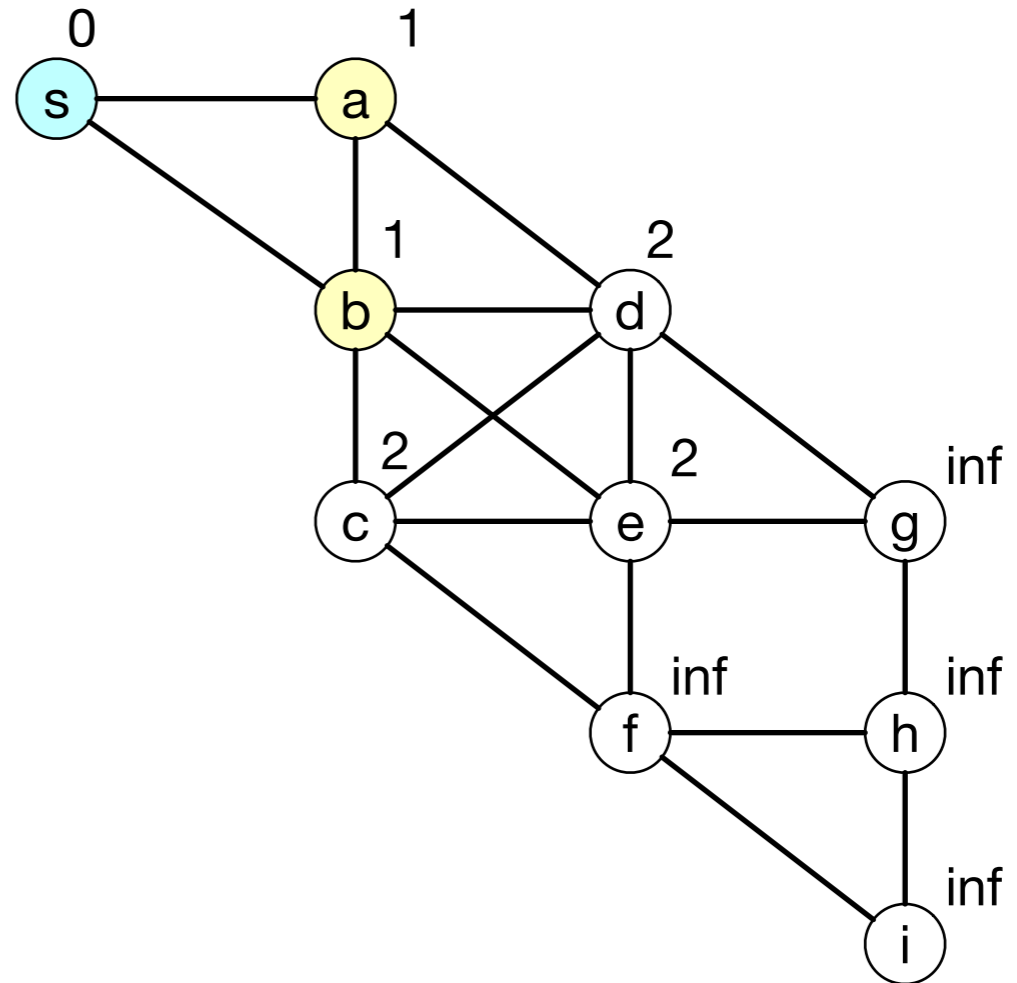
Dijkstra's Algorithm

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



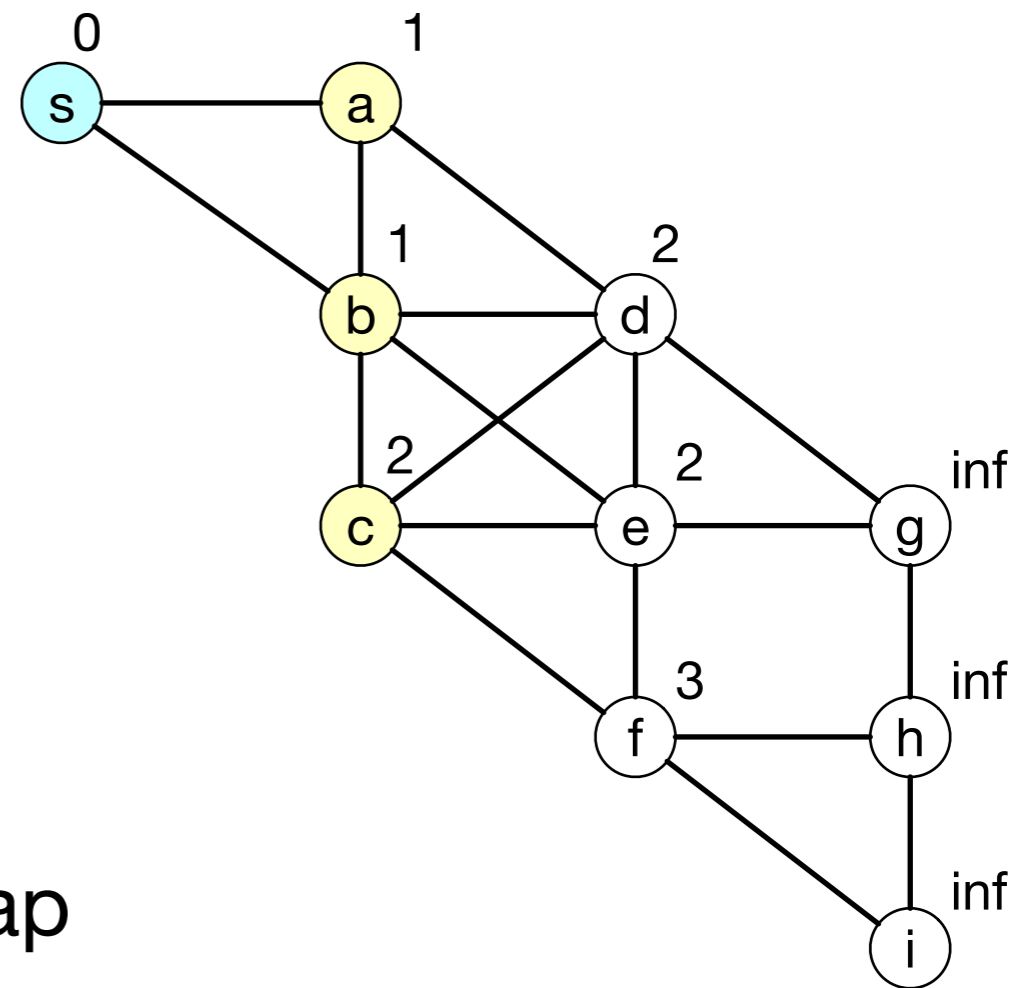
Dijkstra's Algorithm

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



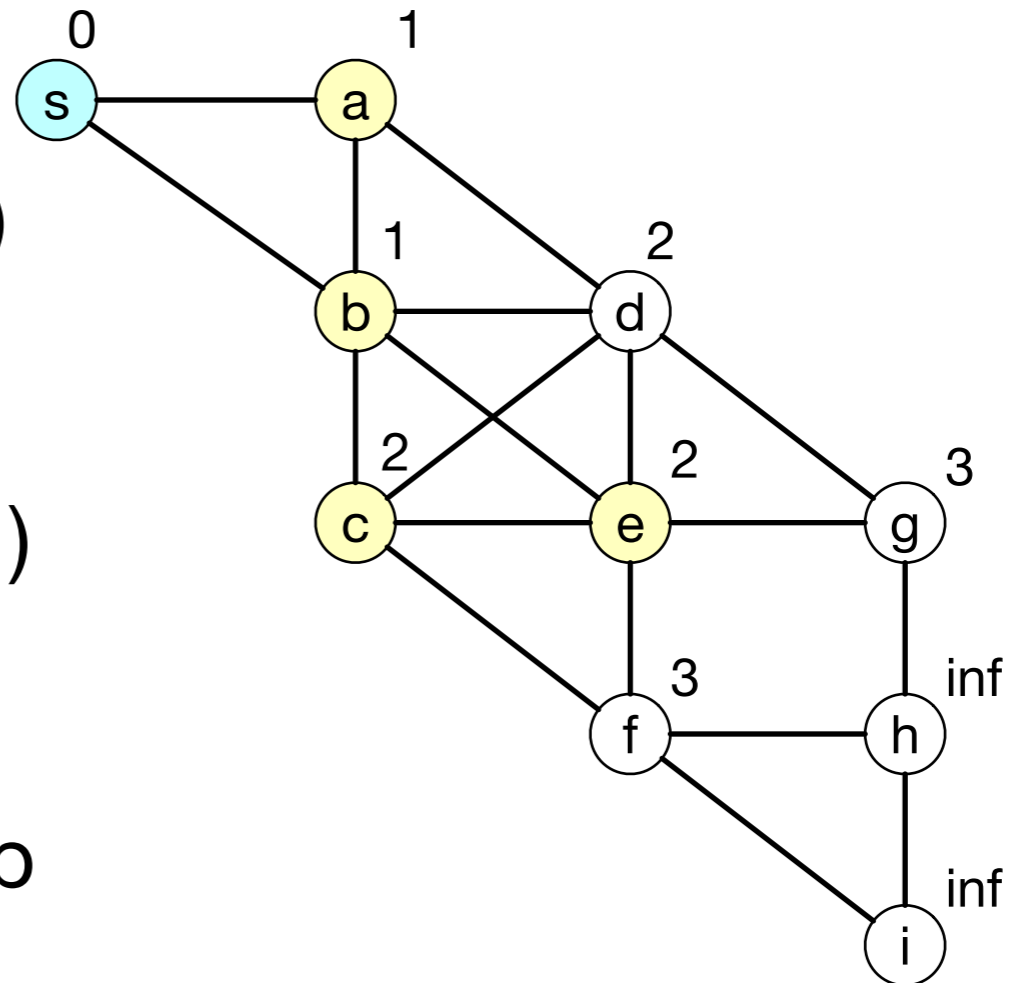
Dijkstra's Algorithm

- 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



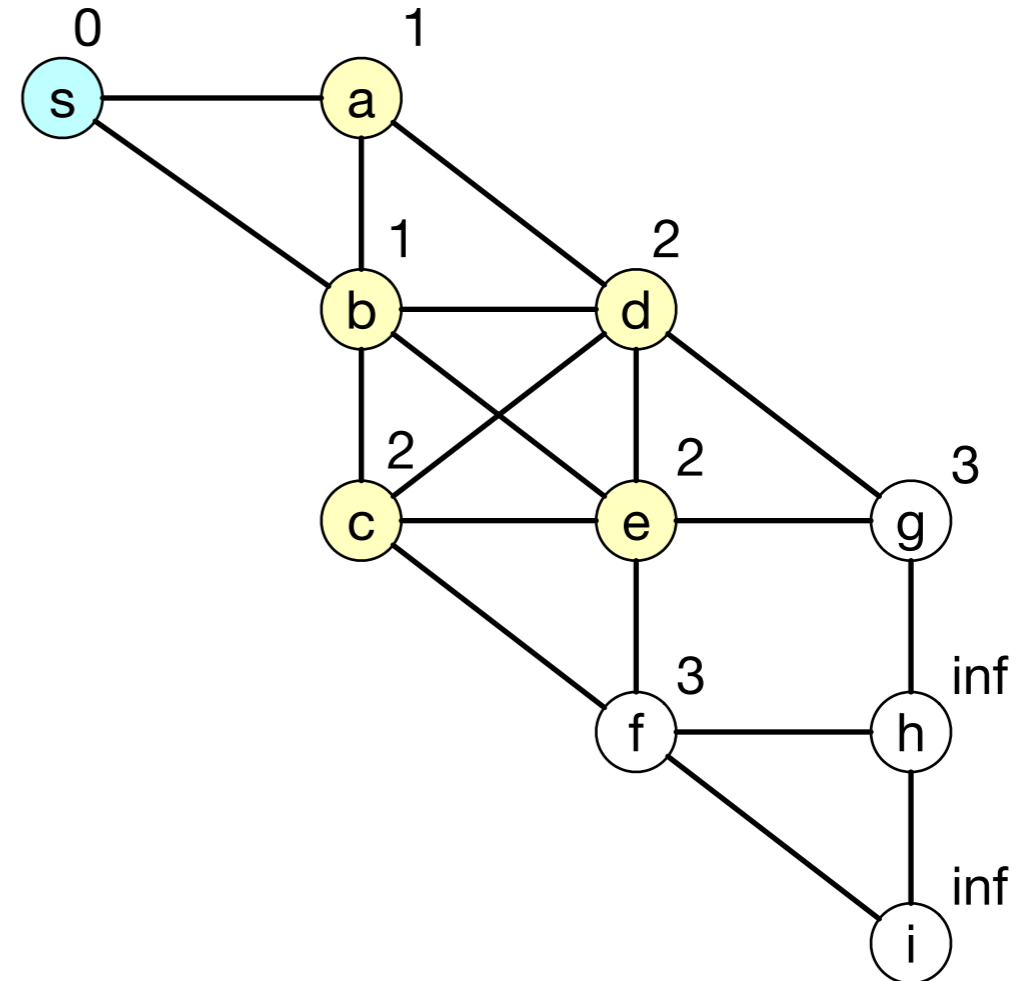
Dijkstra's Algorithm

- 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



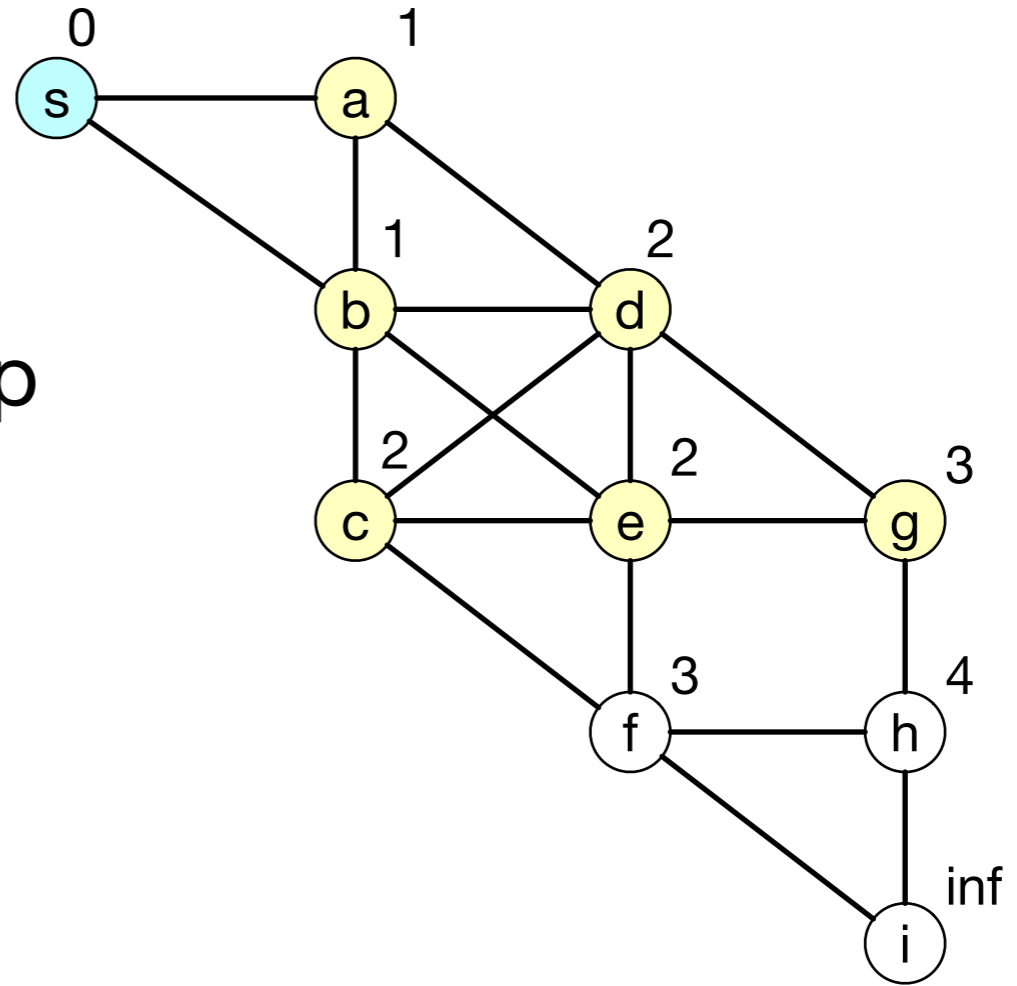
Dijkstra's Algorithm

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



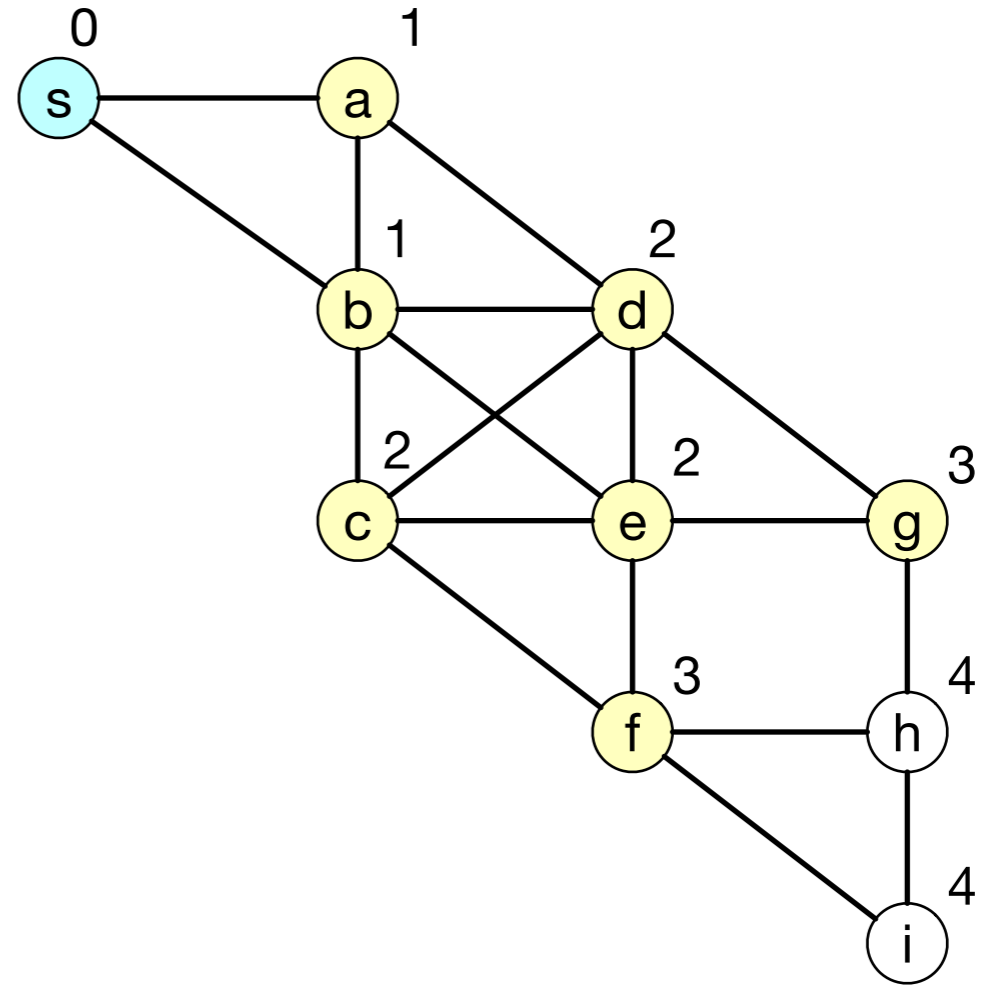
Dijkstra's Algorithm

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



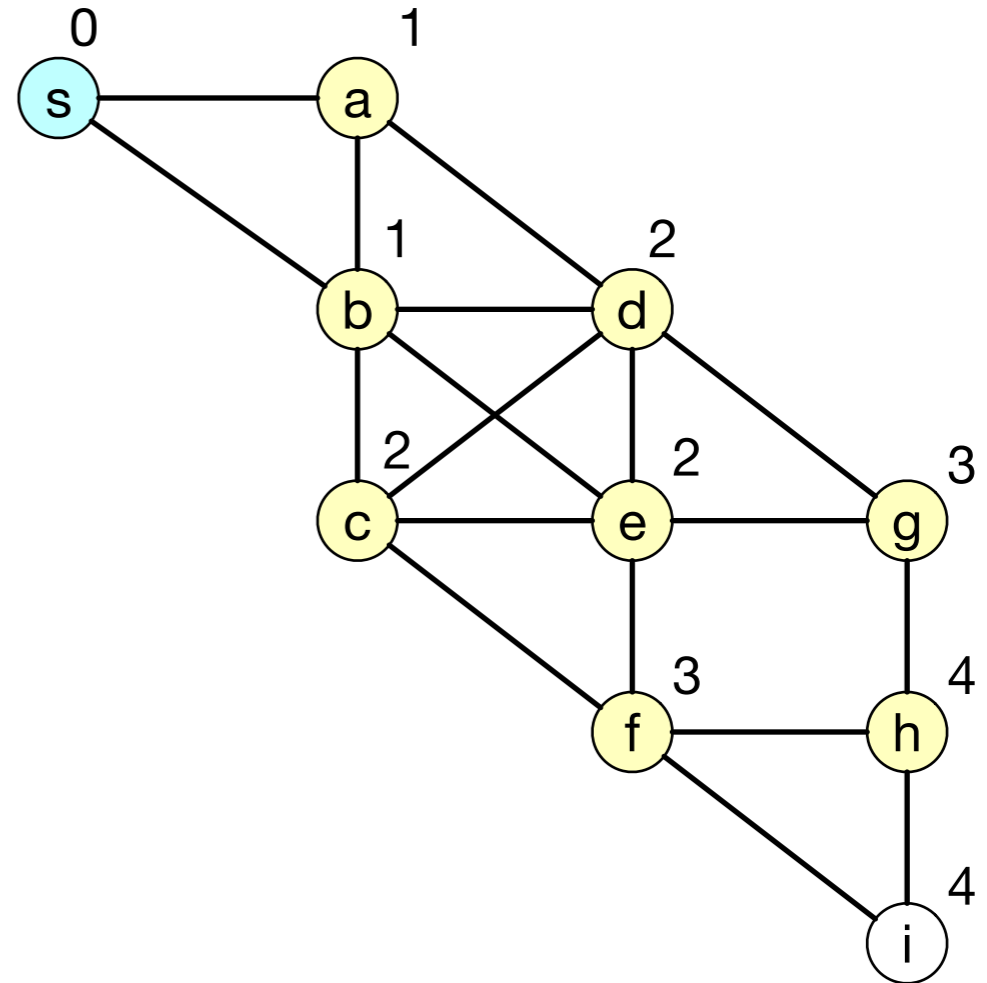
Dijkstra's Algorithm

- Need to select f
 - Update only changes i



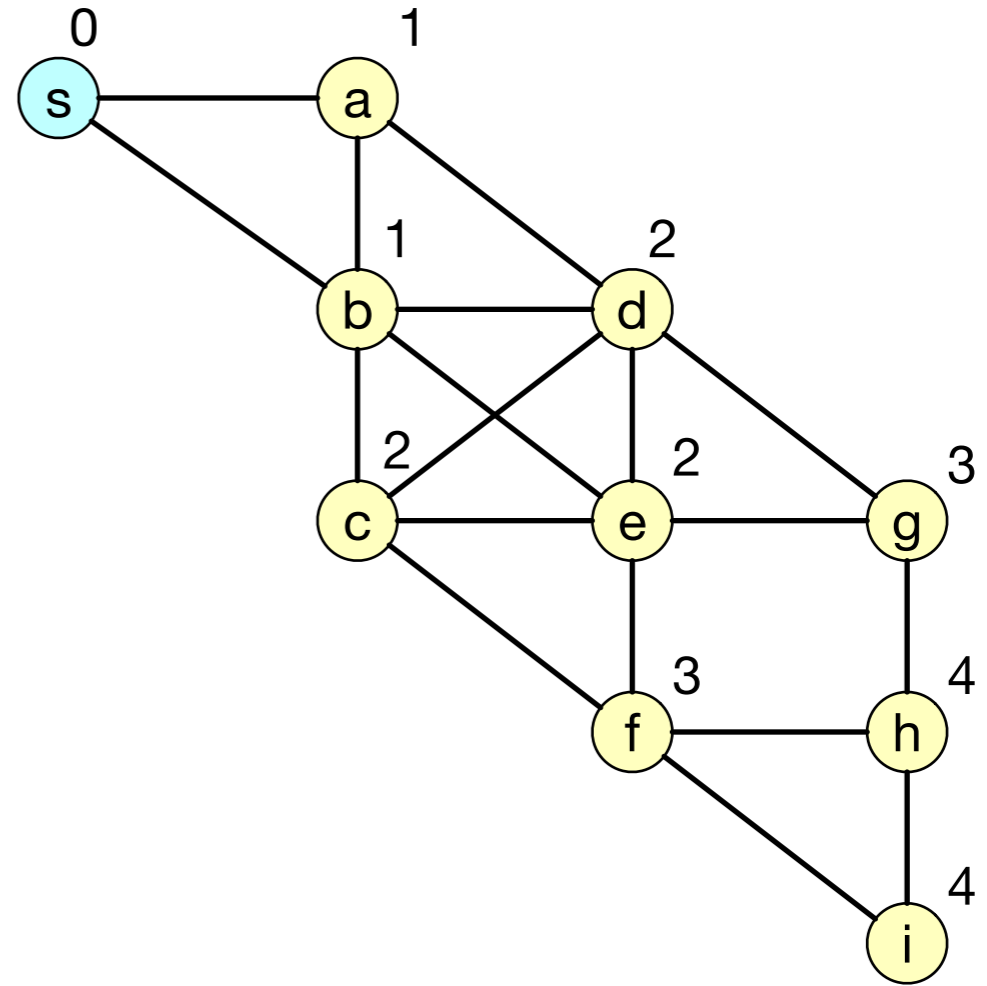
Dijkstra's Algorithm

- 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

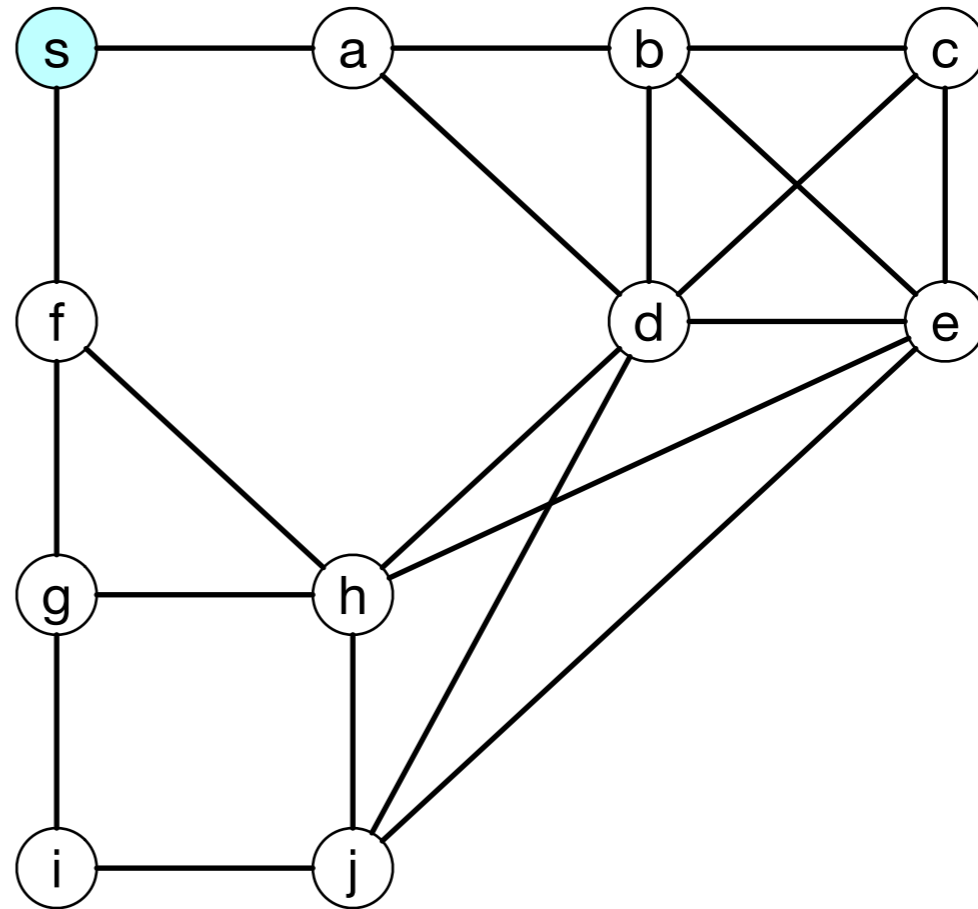


Dijkstra's Algorithm

- Dijkstra's algorithm can be generalized to weighted graphs

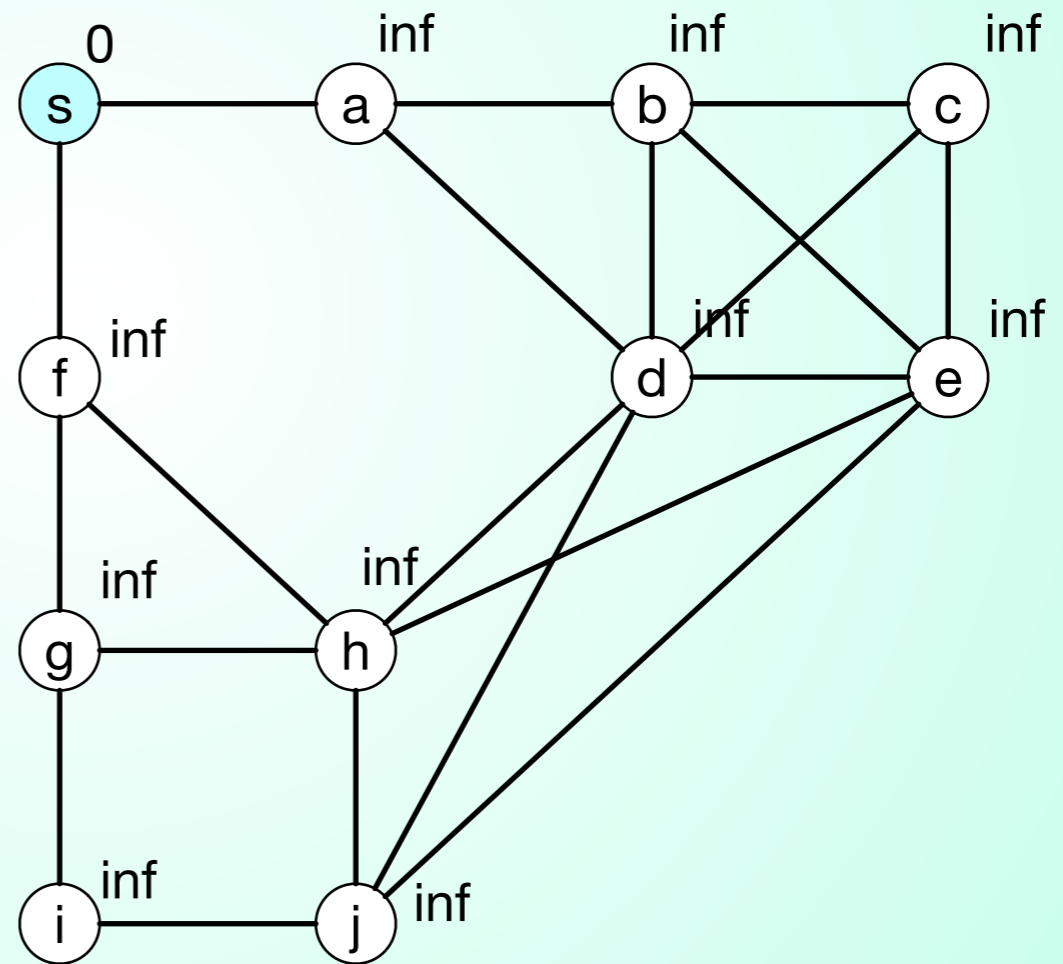
Dijkstra's algorithm

- 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



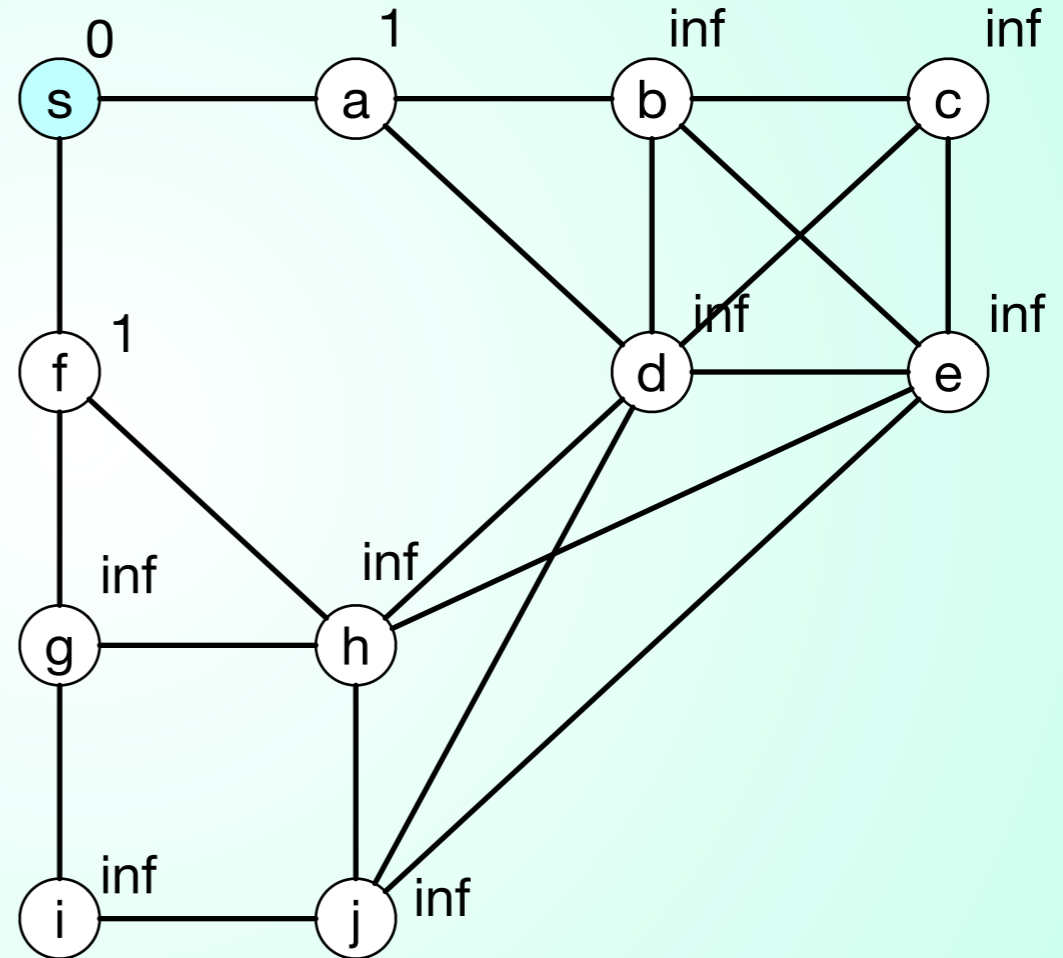
Dijkstra's algorithm

- Select s



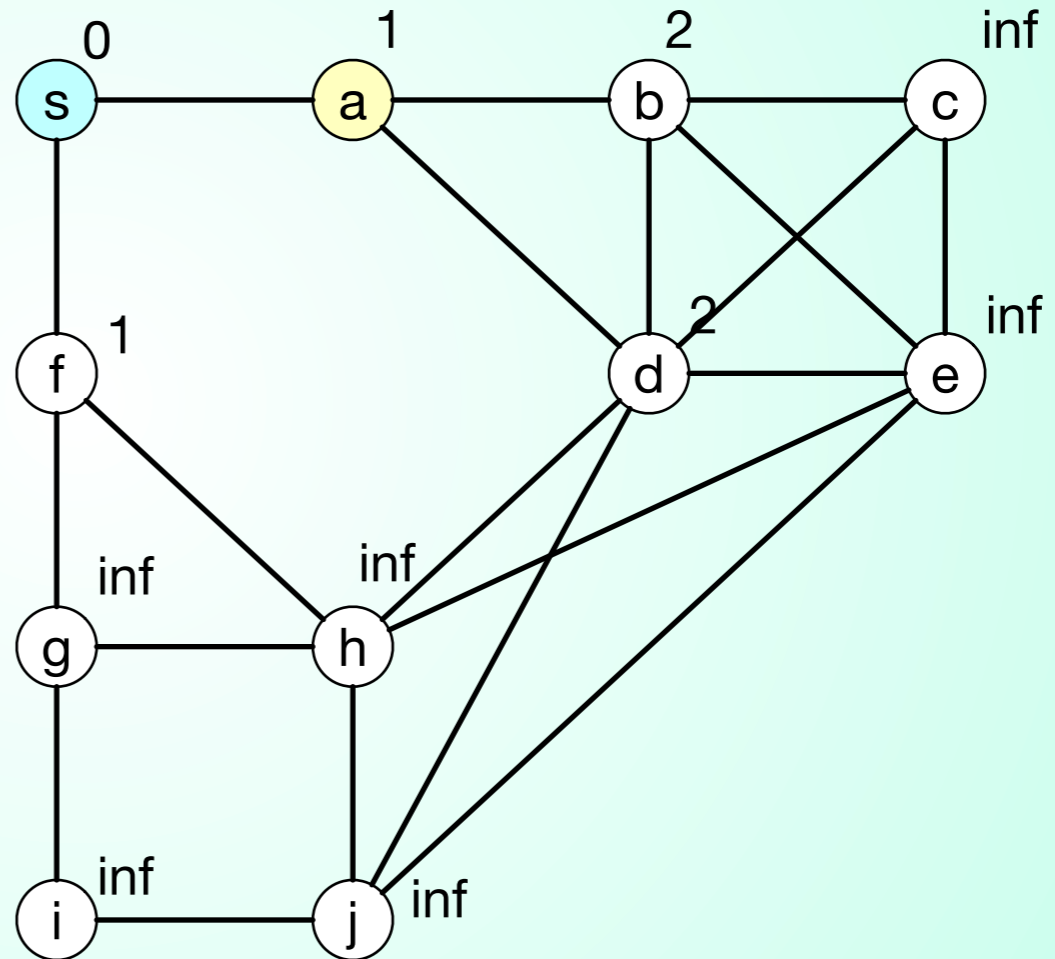
Dijkstra's algorithm

- Update a and f



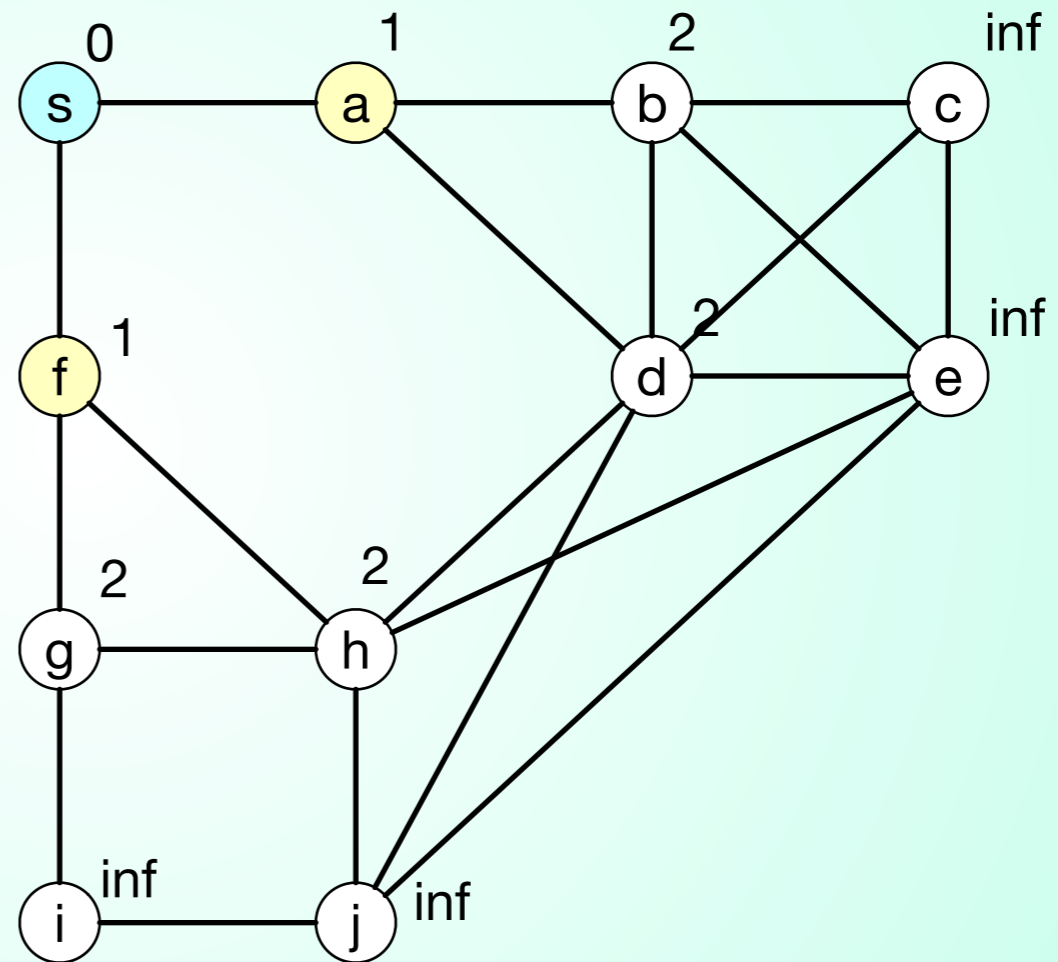
Dijkstra's algorithm

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



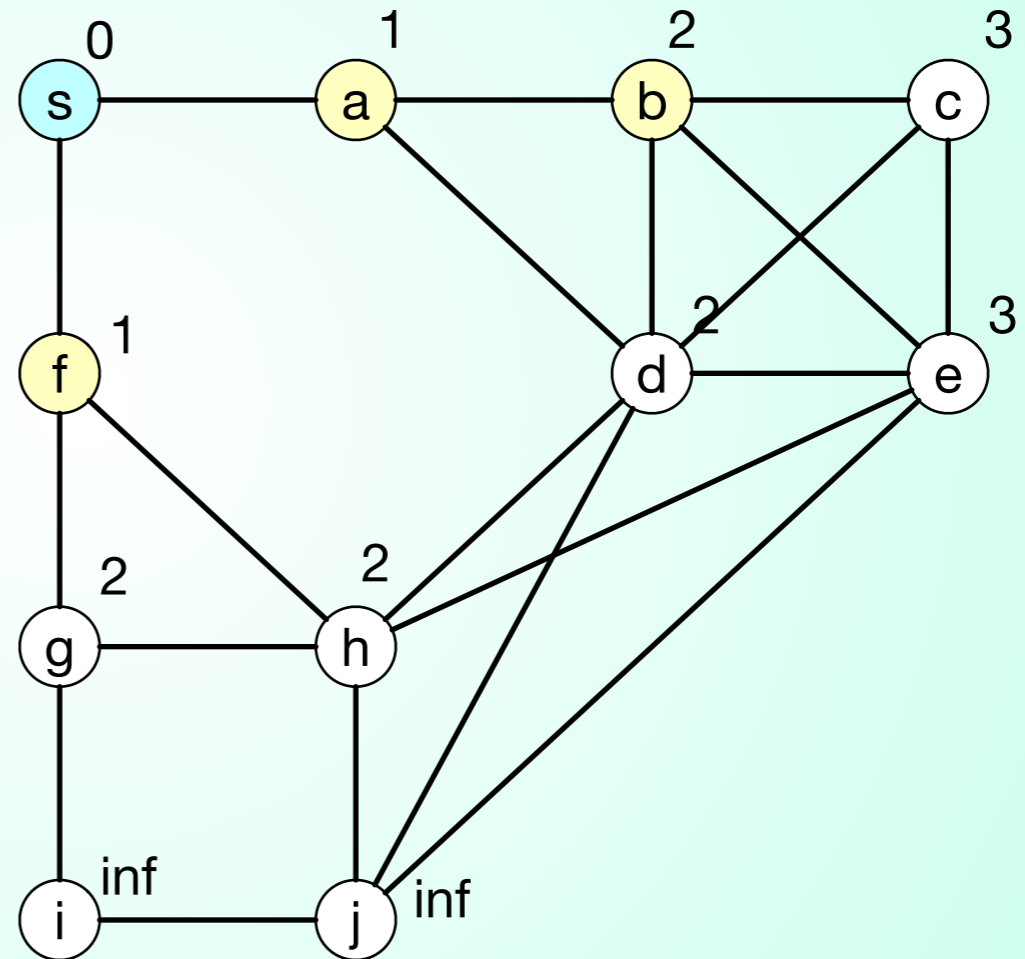
Dijkstra's algorithm

- Select f (no choice here)



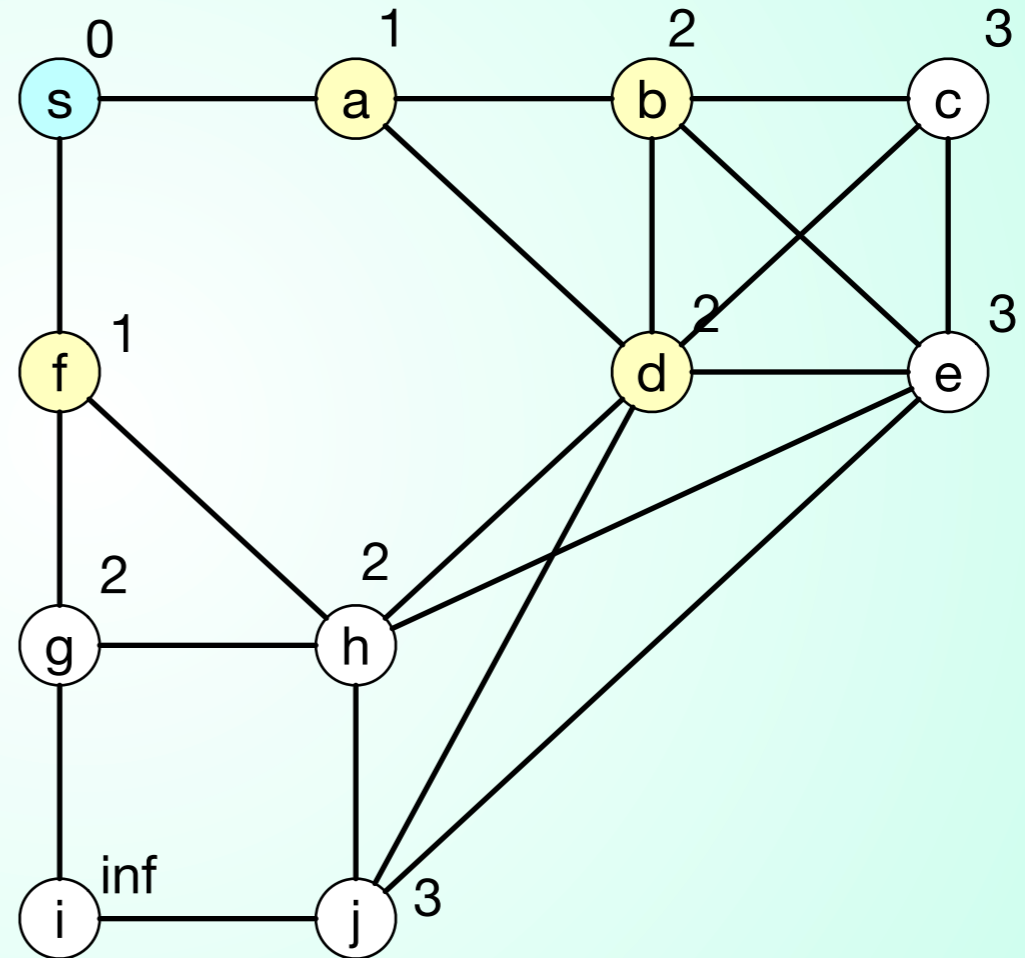
Dijkstra's algorithm

- Select b



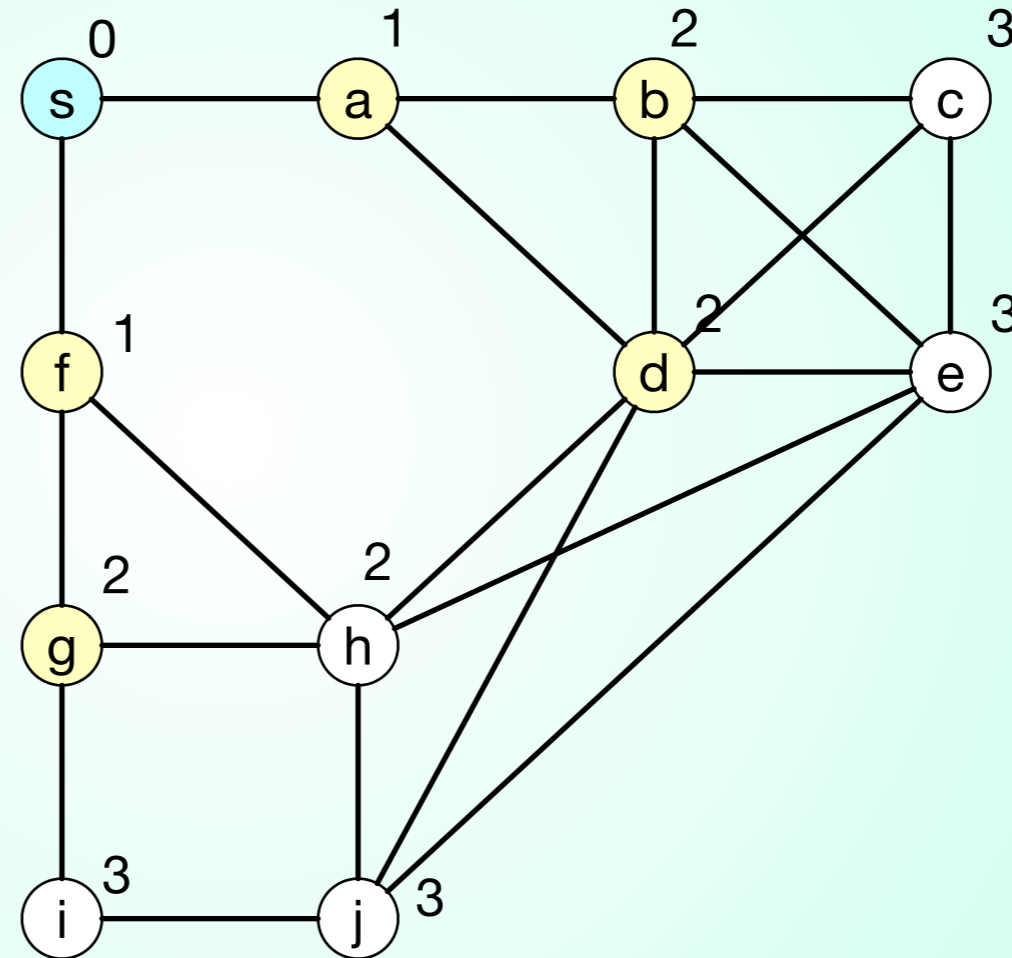
Dijkstra's algorithm

- Select d



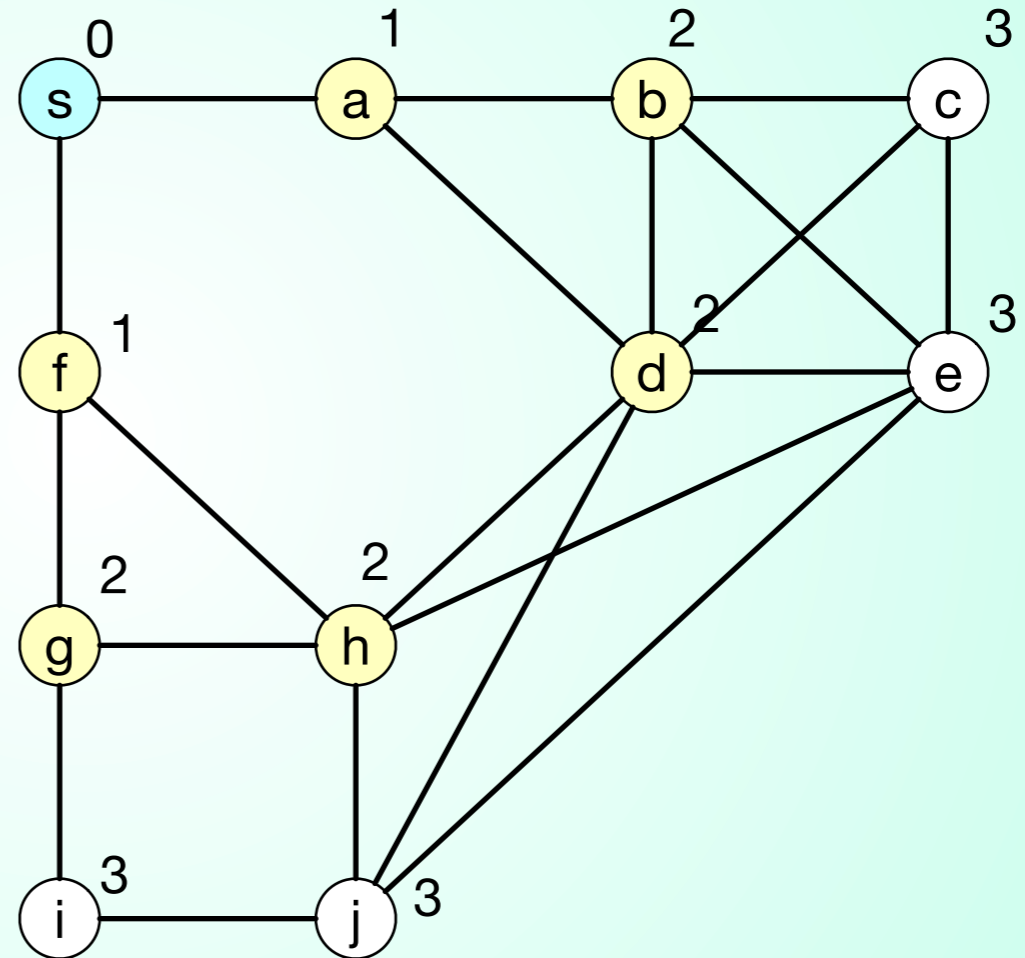
Dijkstra's algorithm

- Select g



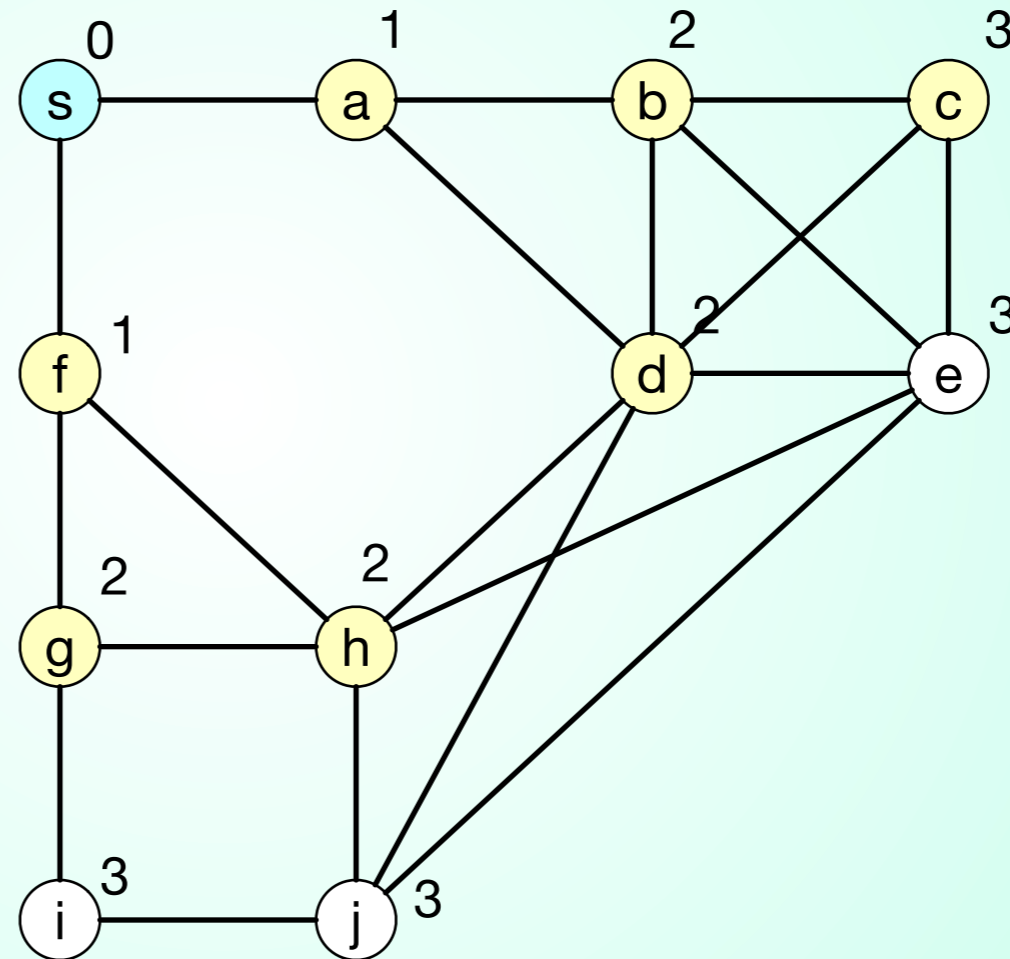
Dijkstra's algorithm

- 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

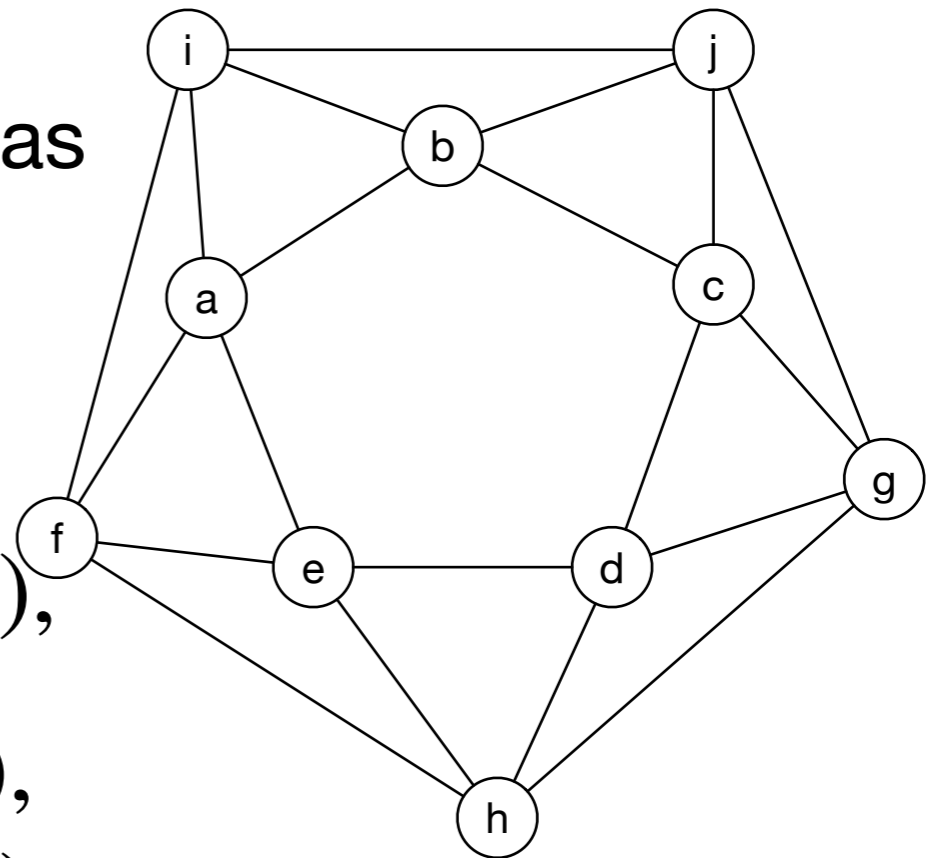
$$V = \{a, b, c, d, e, f, g, h, i, j\}$$

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

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

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

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



Dijkstra's Algorithm

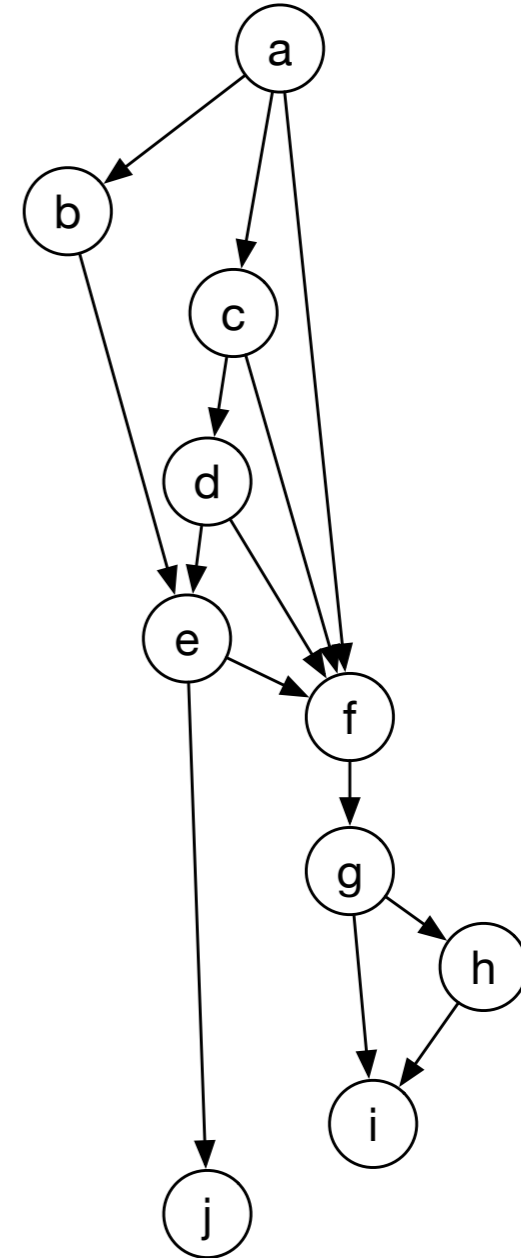
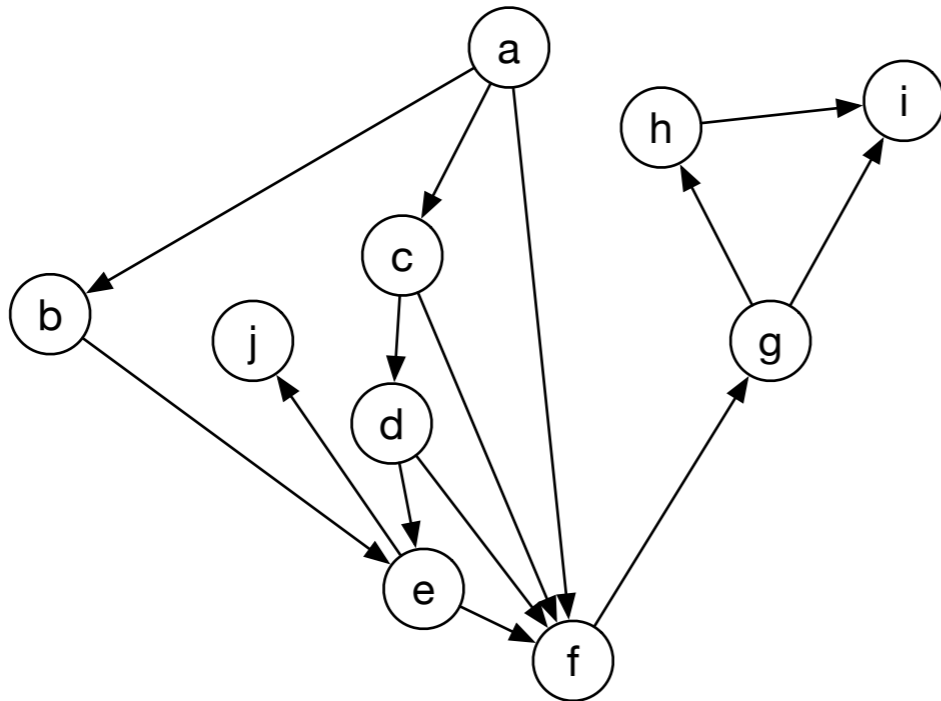
- Need to maintain a priority heap
 - Otherwise
 - Look at every node
 - And every edge twice

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

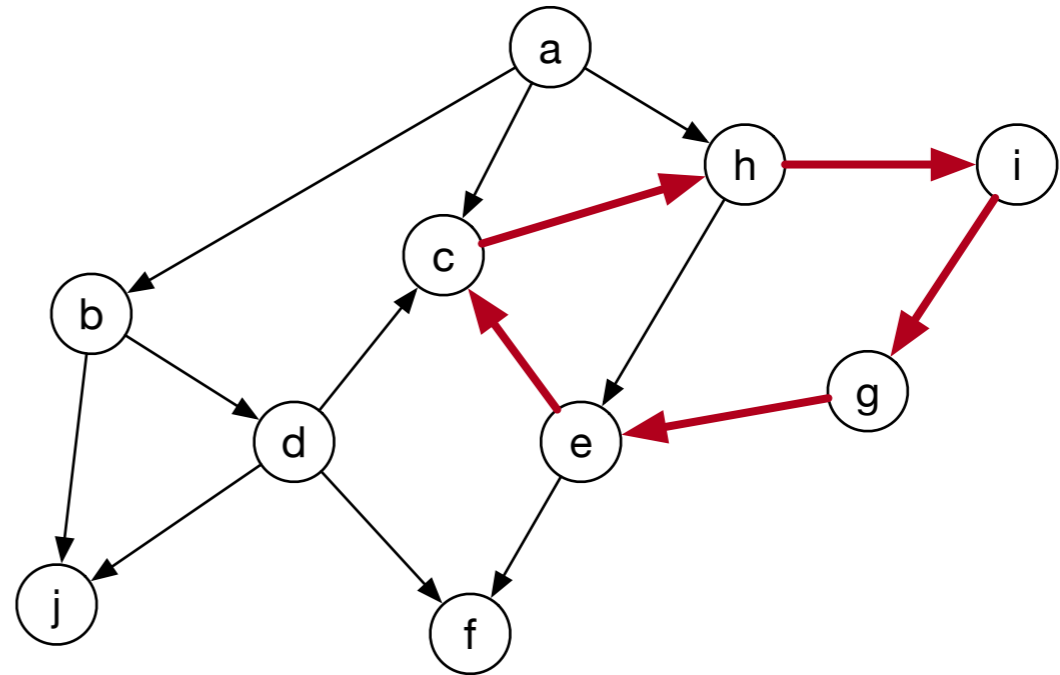
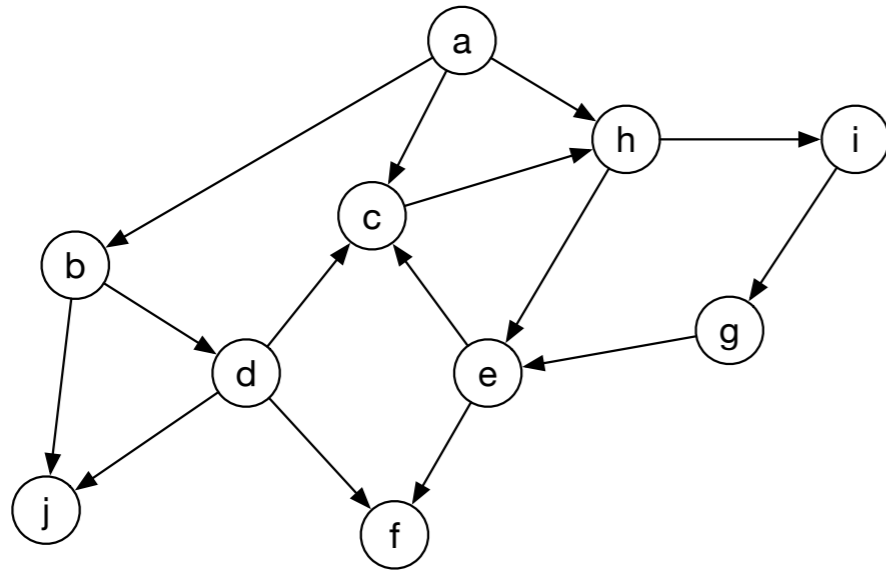
Topological Sort

- 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

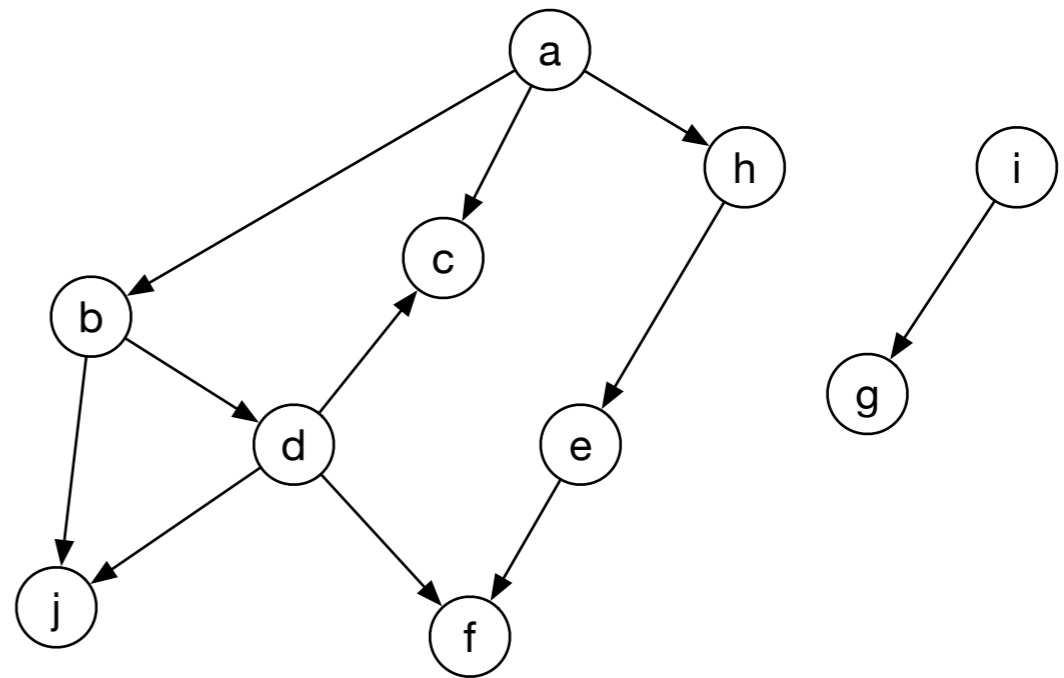
- Example:
 - There is a cycle, a topological sort is not possible



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:
```

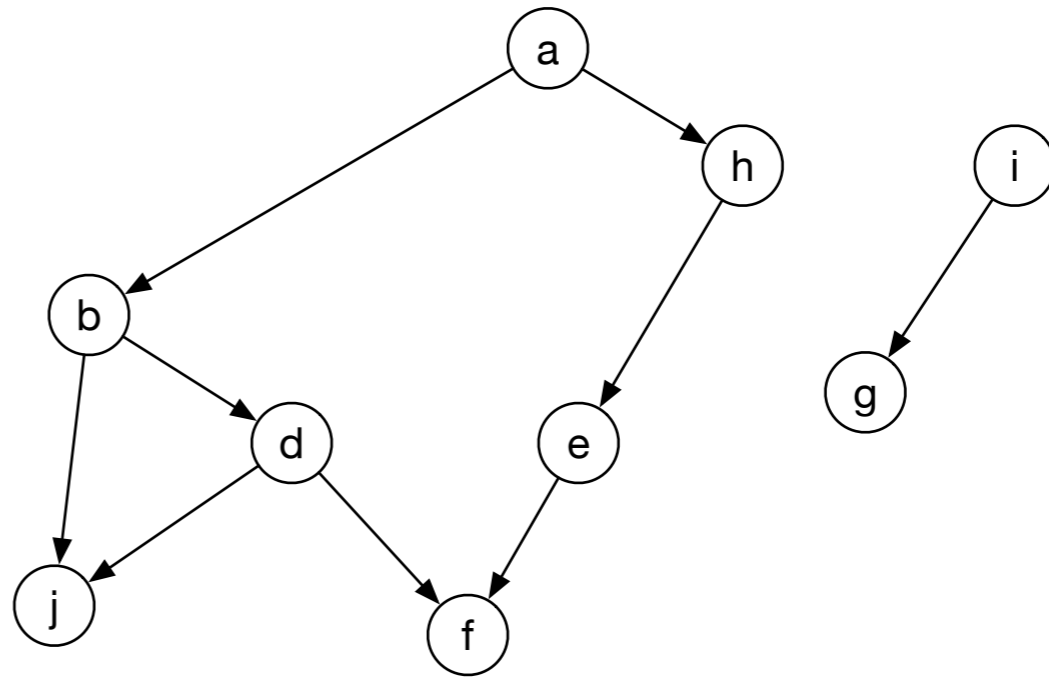


- Find a vertex with empty list, add it to a list, and remove it from the graph

Topological Sort

- A simple algorithm

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

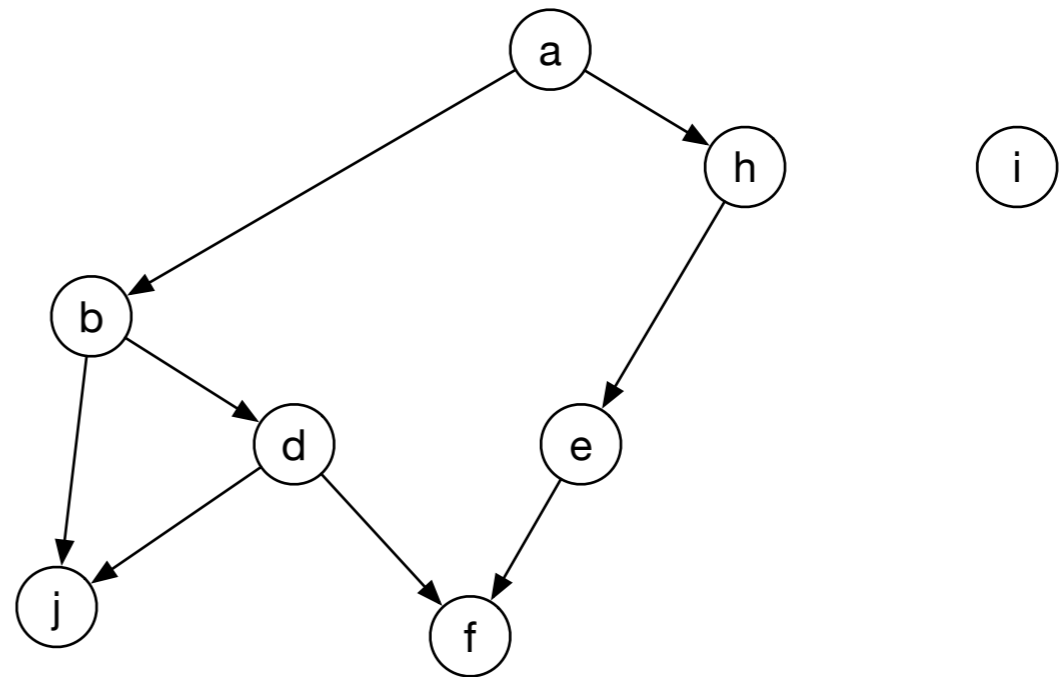


- List contains {c}

Topological Sort

- A simple algorithm

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

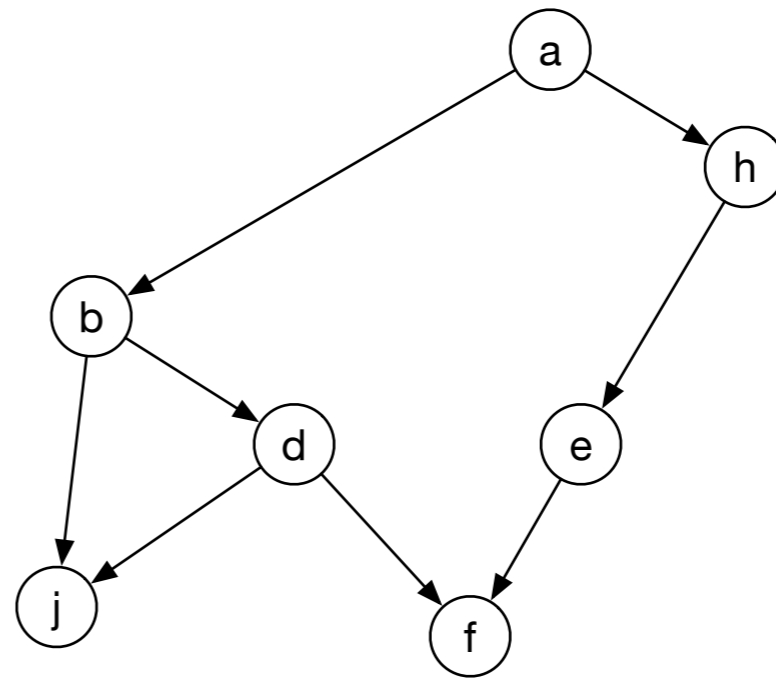


- Remove g and add it to the list {c, g}

Topological Sort

- A simple algorithm

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

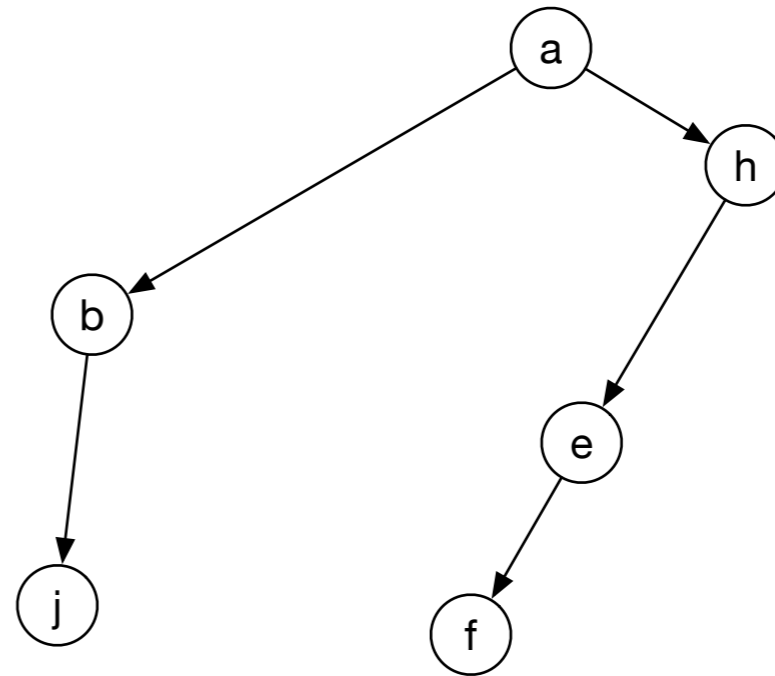


- Remove i and add it to the list $\{c, g, i\}$

Topological Sort

- A simple algorithm

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

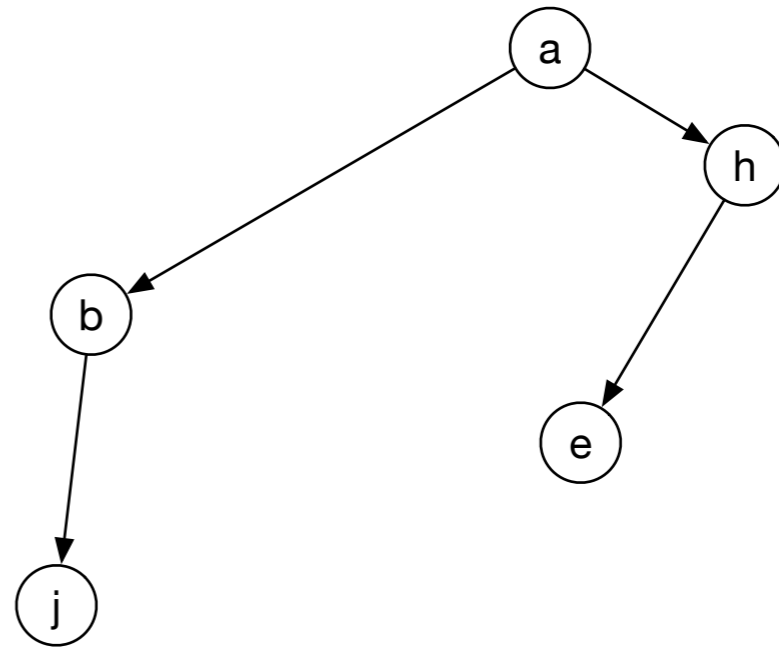


- Remove d and add it to the list $\{c, g, i, d\}$

Topological Sort

- A simple algorithm

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

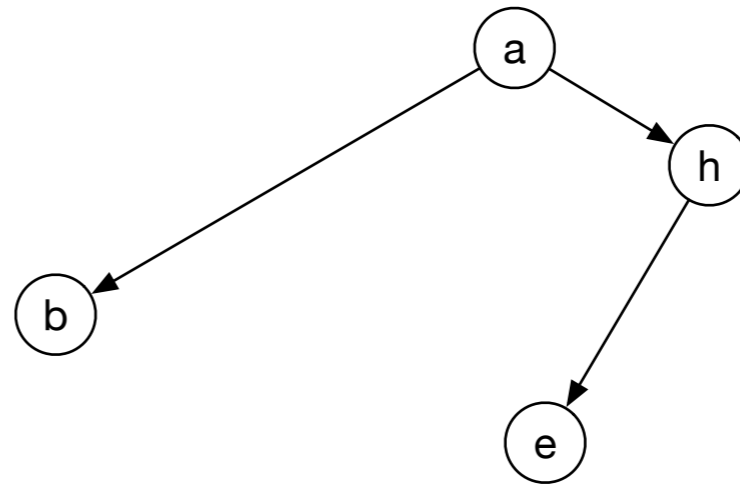


- Remove f and add it to the list $\{c, g, i, d, f\}$

Topological Sort

- A simple algorithm

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

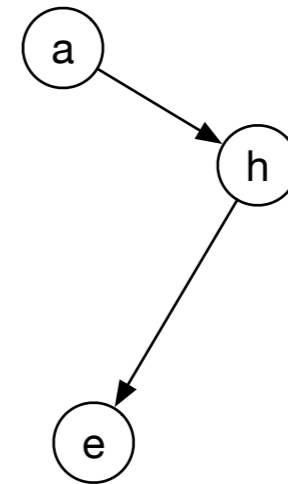


- Remove j and add it to the list $\{c, g, i, d, f, j\}$

Topological Sort

- A simple algorithm

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

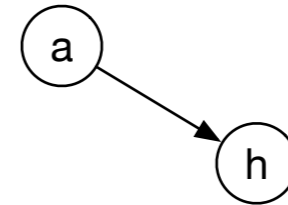


- Remove b and add it to the list $\{c, g, i, d, f, j, b\}$

Topological Sort

- A simple algorithm

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

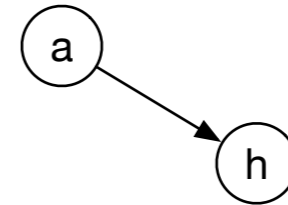


- Remove e and add it to the list $\{c, g, i, d, f, j, b, e\}$

Topological Sort

- A simple algorithm

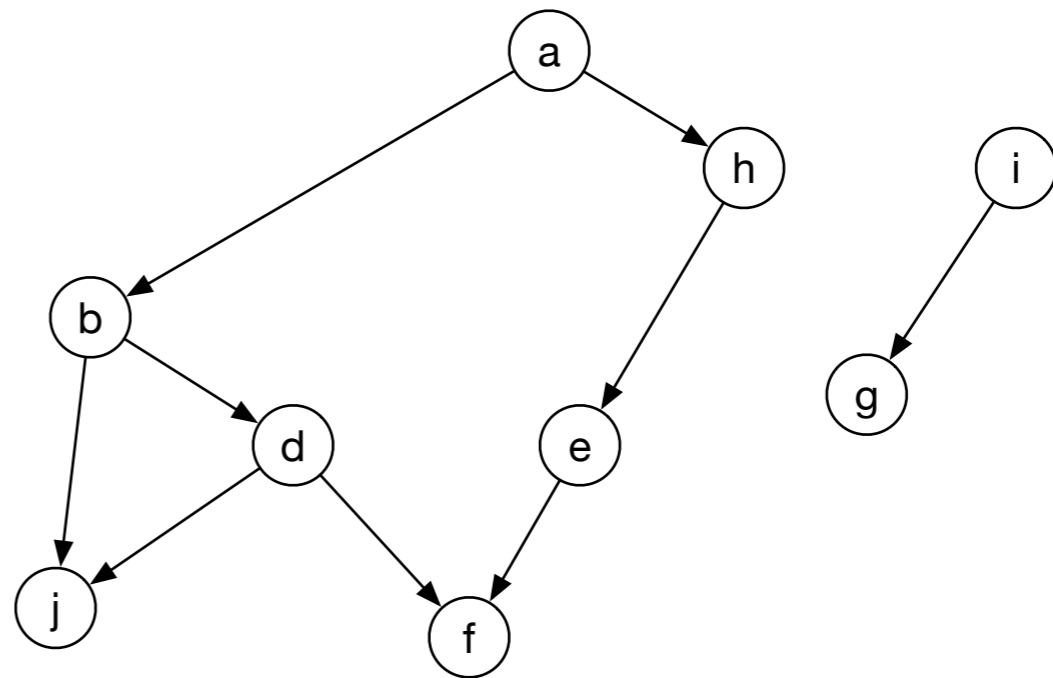
```
a: b, c, h
b: d, j
c:
d: c
e: f
f:
g:
h: e
i: g
j:
```



- 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?

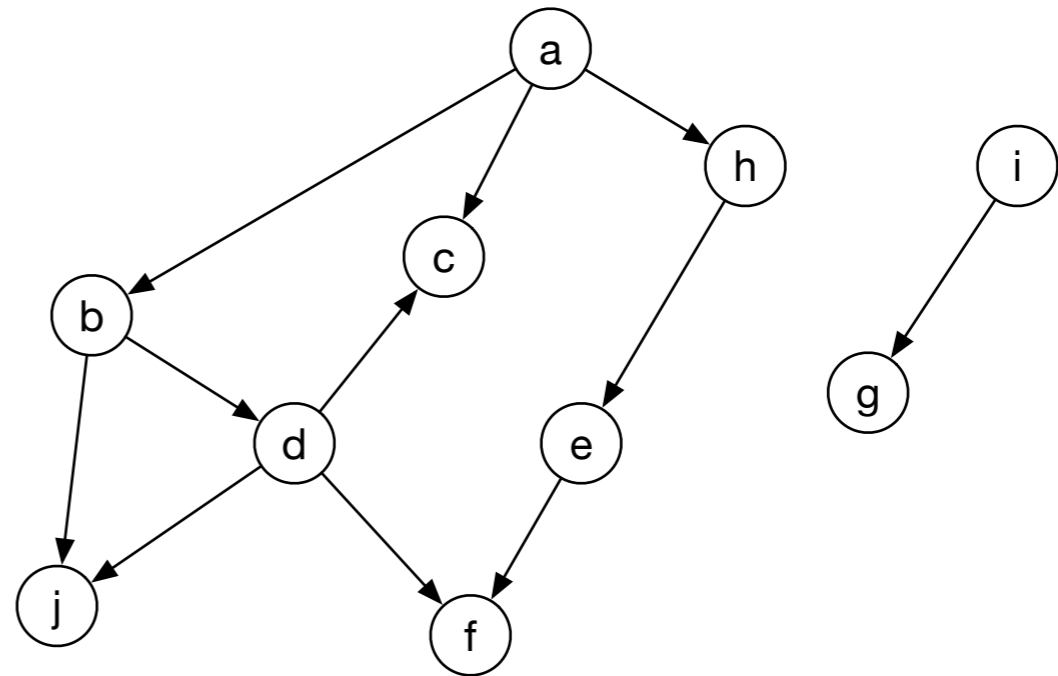
Topological Sort

- 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:

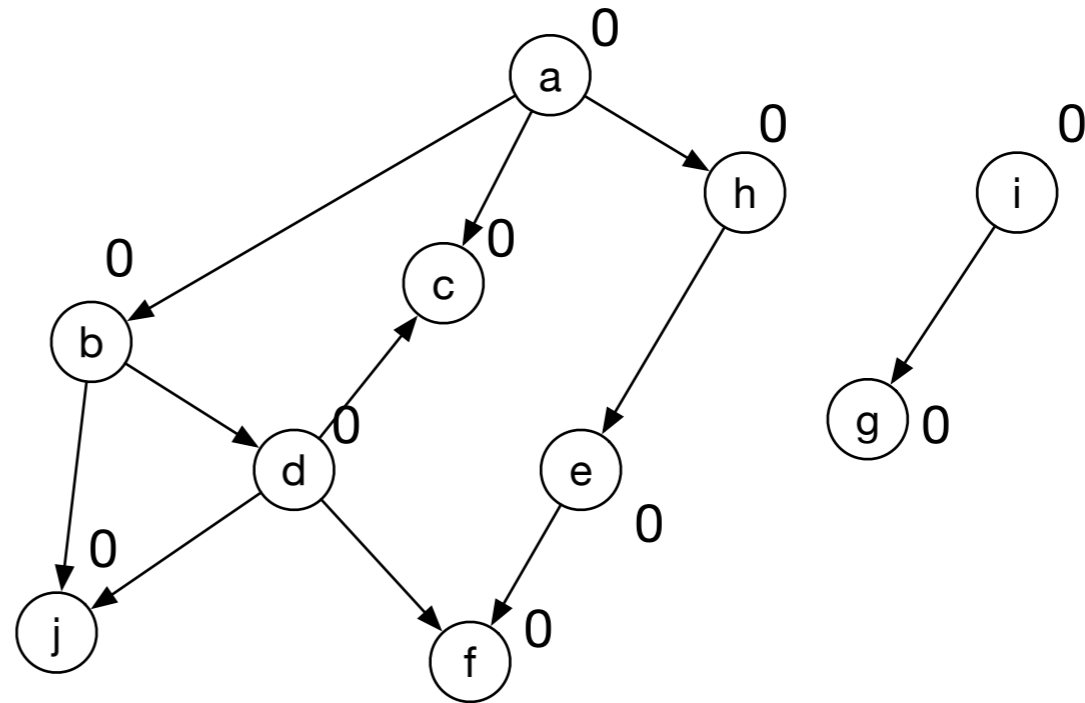


- Initialize in-degree 0 for all vertices

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

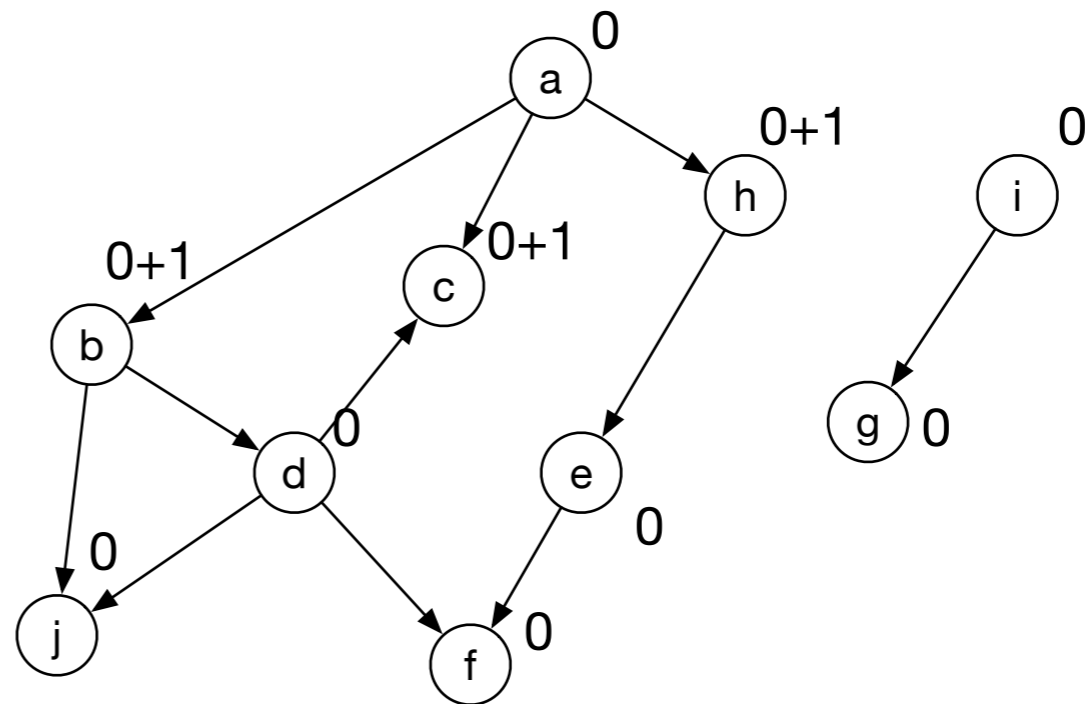
f:

g:

h: e

i: g

j:



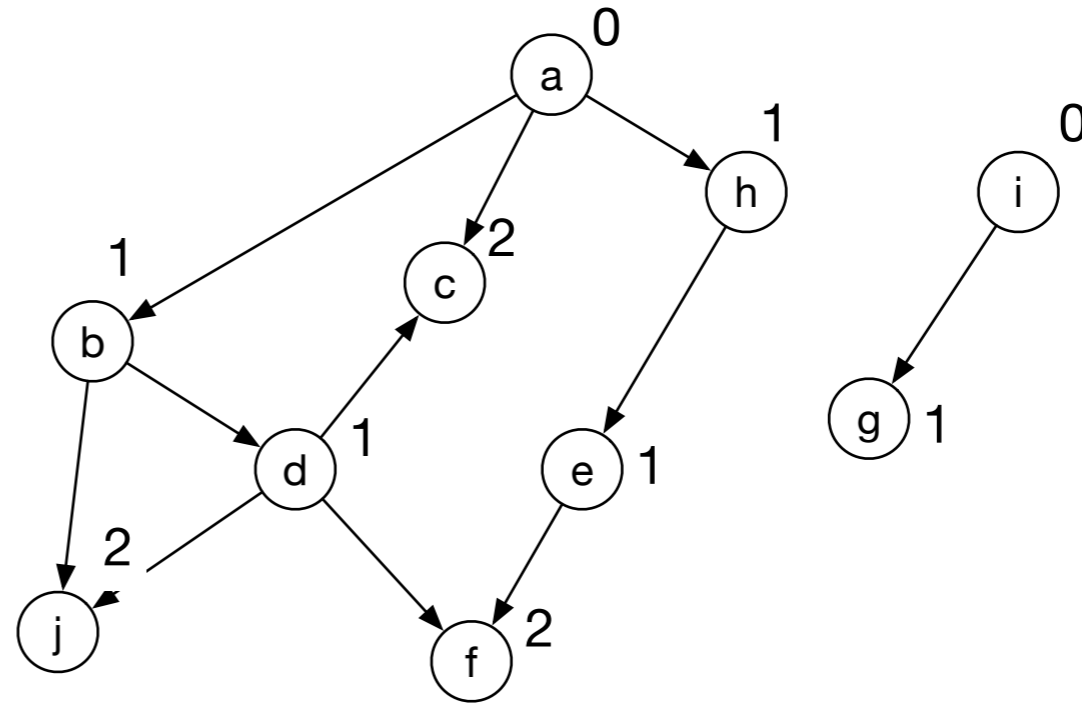
- Go through the adjacency list.

- 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:



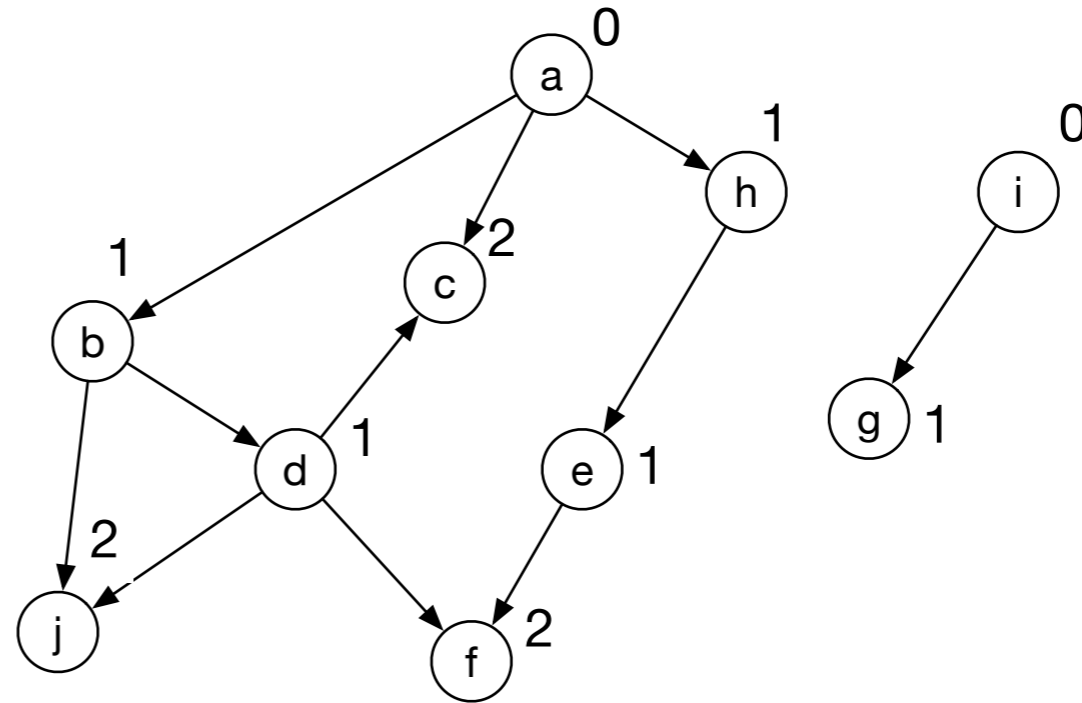
- Go through the adjacency list.

- 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:



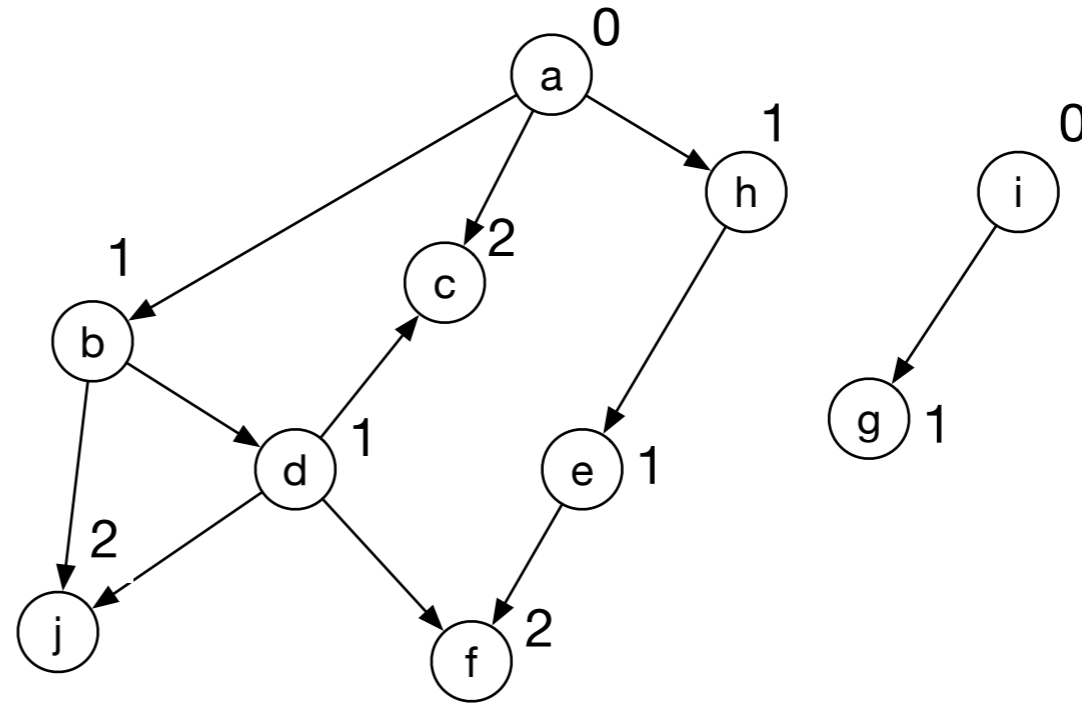
- Now we start the removal phase

- 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:



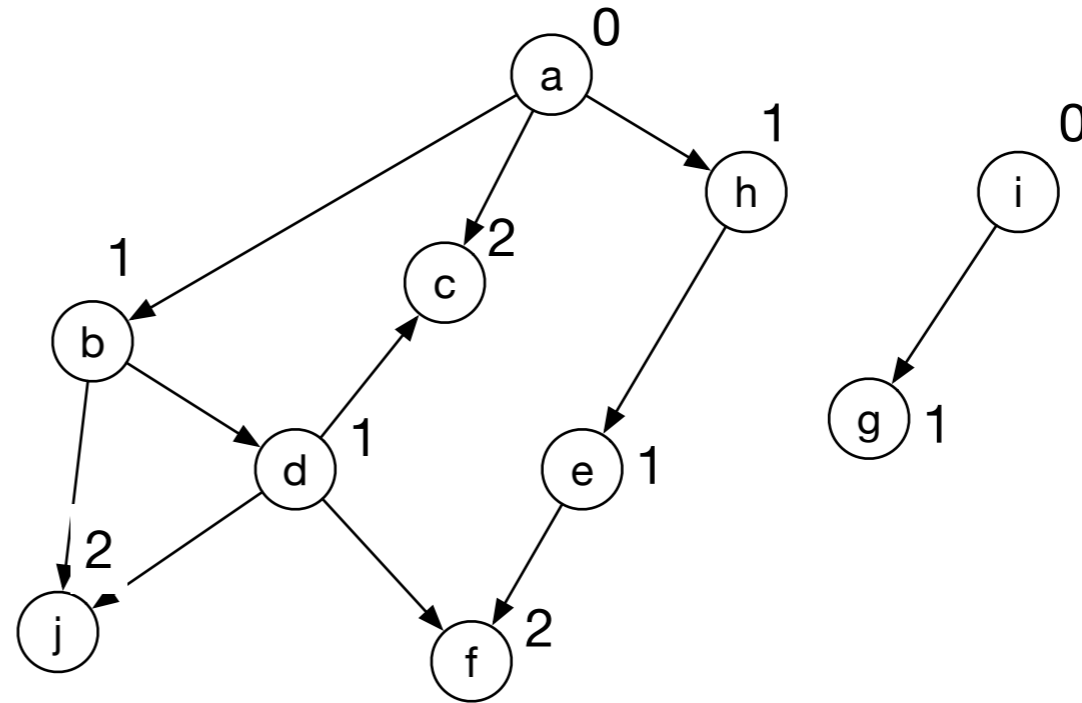
- Now we start the removal phase

- 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:



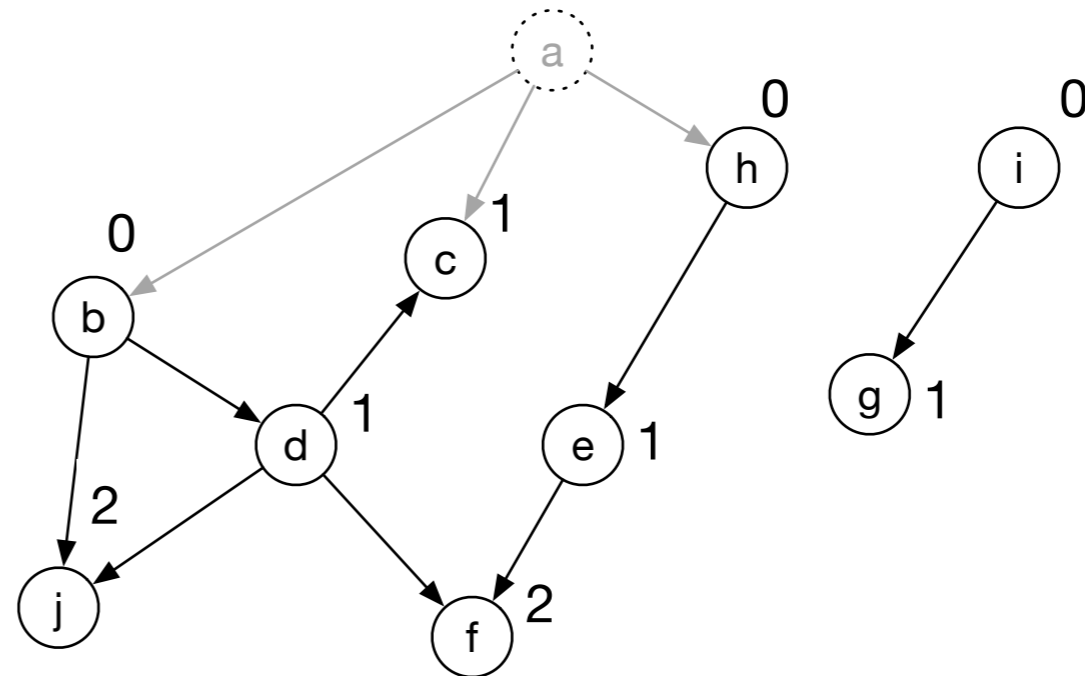
- We select a for the removal

- 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:

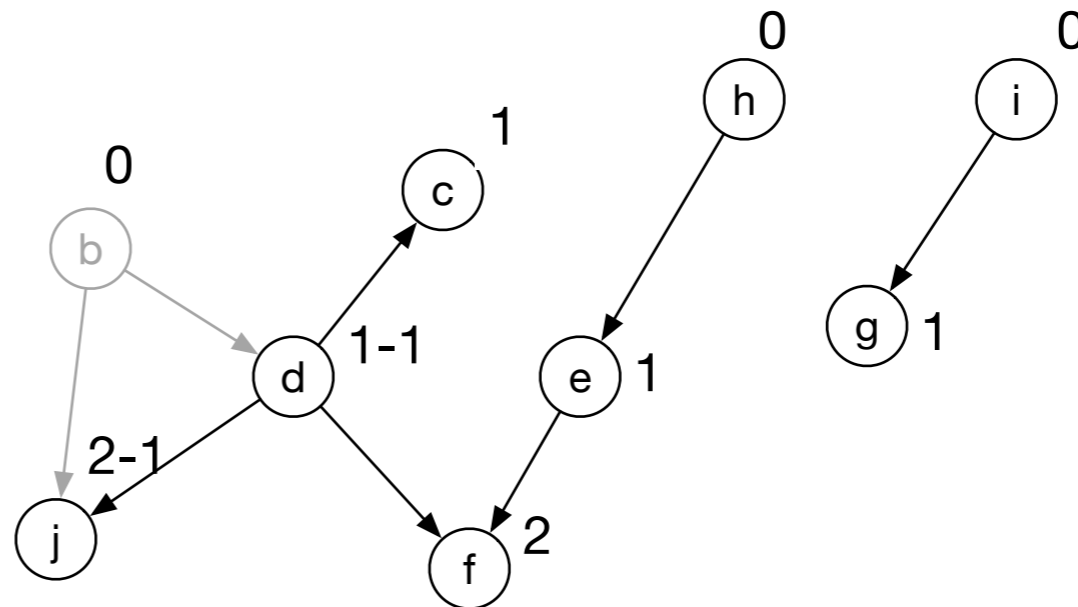


- We select a for the removal: $\{a\}$
 - We go through its adjacency list and reset the in-degrees of the nodes there

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:

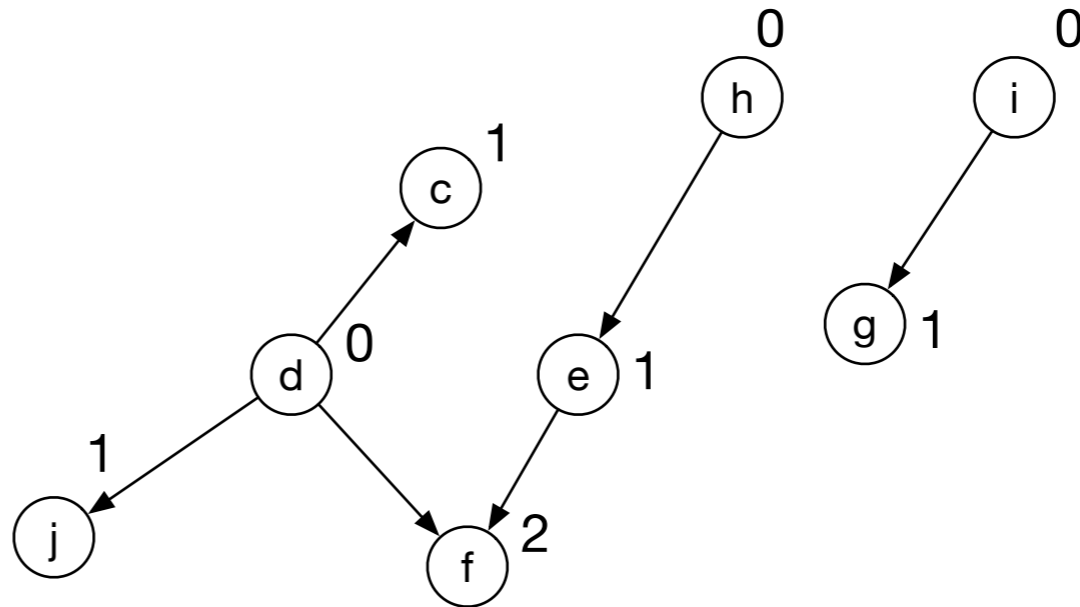
- b: {a, b}

- 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:

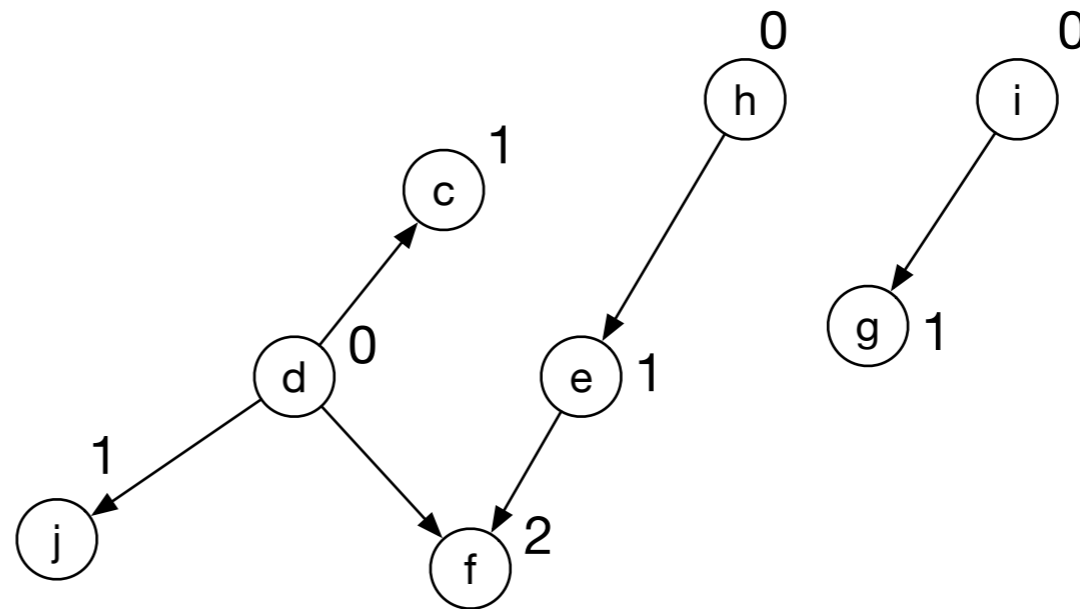
- $b: \{a, b\}$

- 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:
```

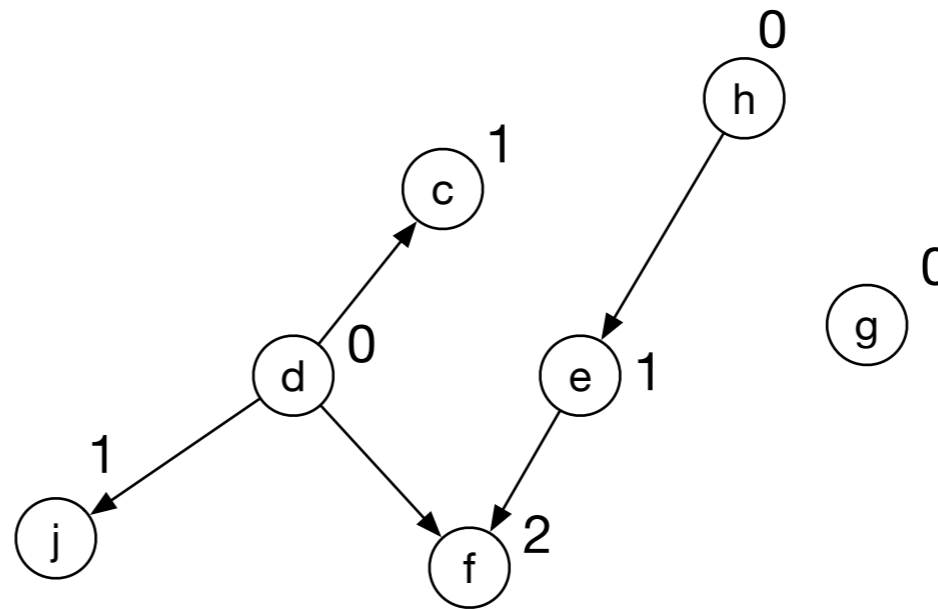


- We now randomly pick one of the vertices with degree 0, let's pick i
- Deleting it means just decrementing the in-degree of g

Topological Sort

- Example:

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

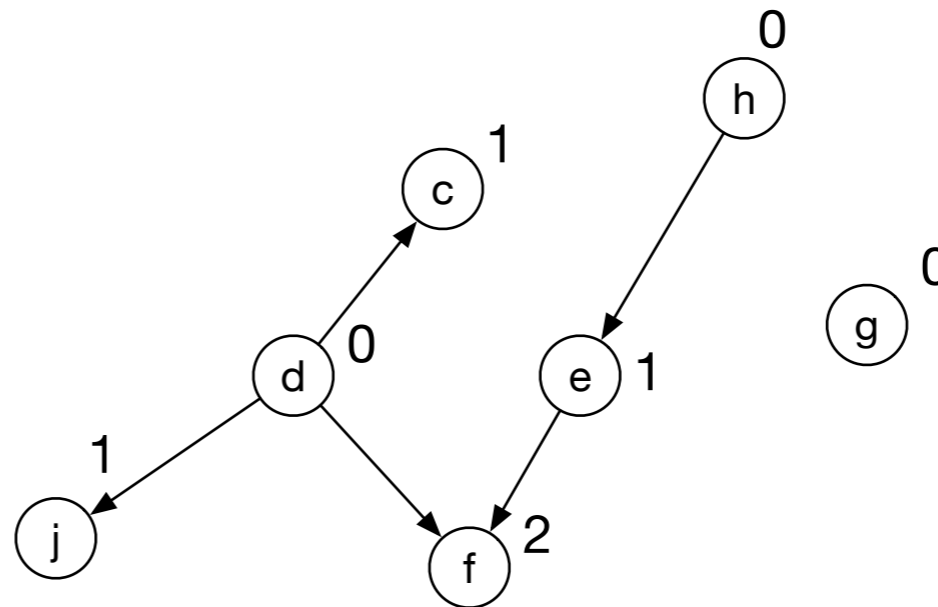


- We add g to our list $\{a, b, i, g\}$

Topological Sort

- Example:

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

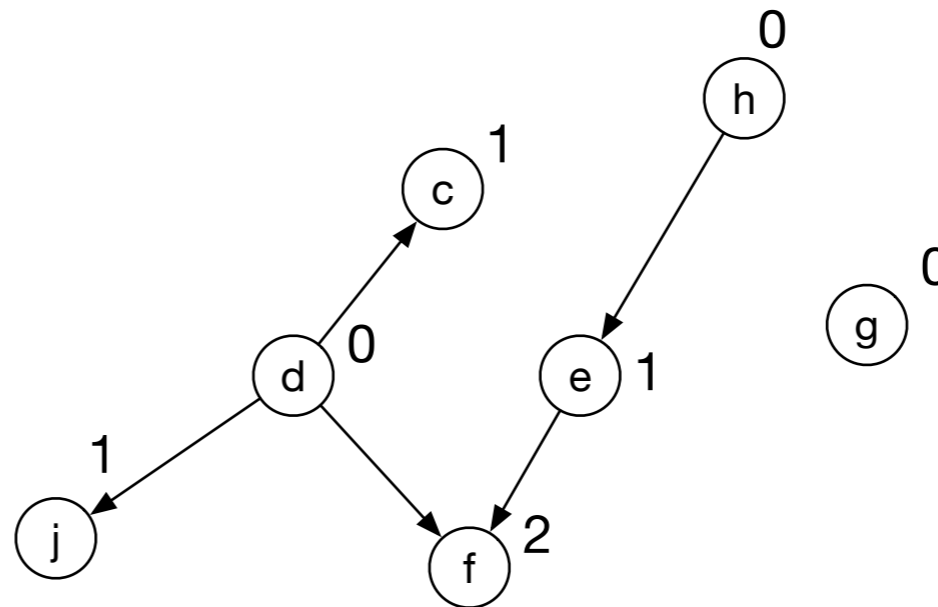


- There are three nodes with in-degree 0, let's pick h

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

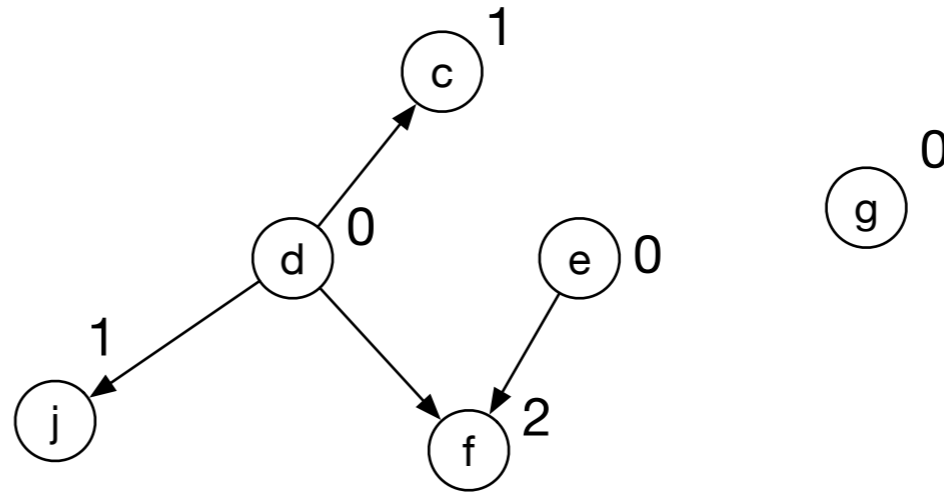
f:

g:

h: e

i: g

j:



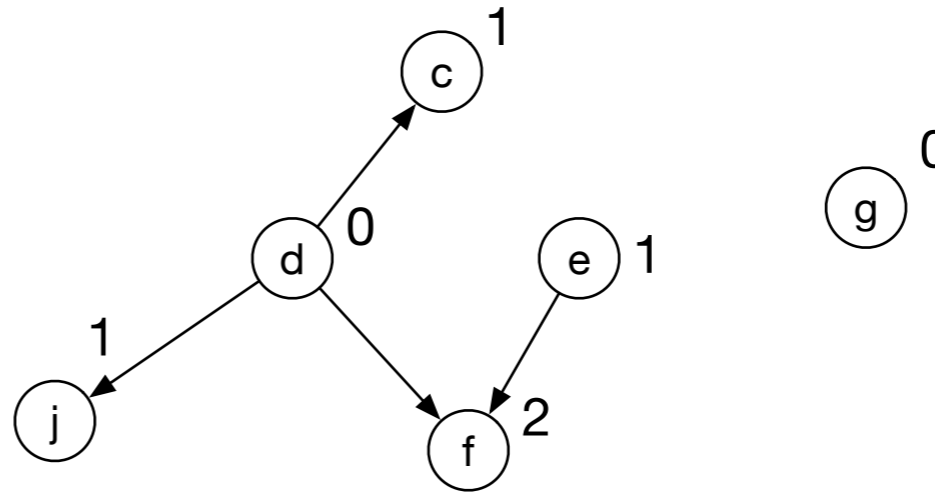
- Need to update in-degree of e

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

Topological Sort

- Example:

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

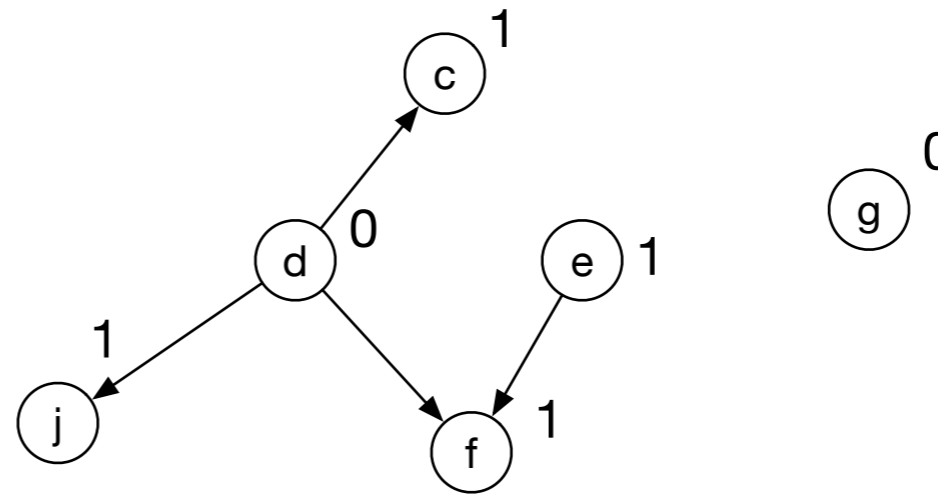


- There are two nodes with in-degree 0, let's pick d

Topological Sort

- Example:

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

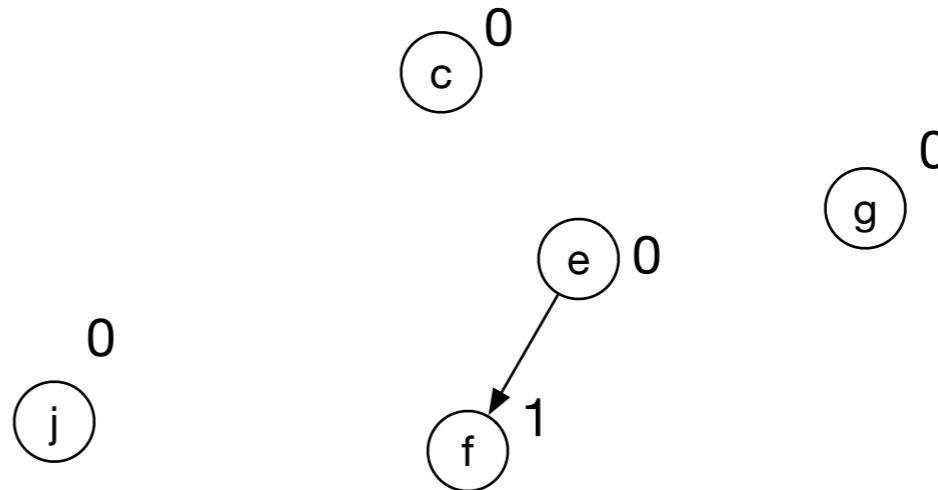


- $\{a, b, i, g, h, d\}$

Topological Sort

- Example:

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

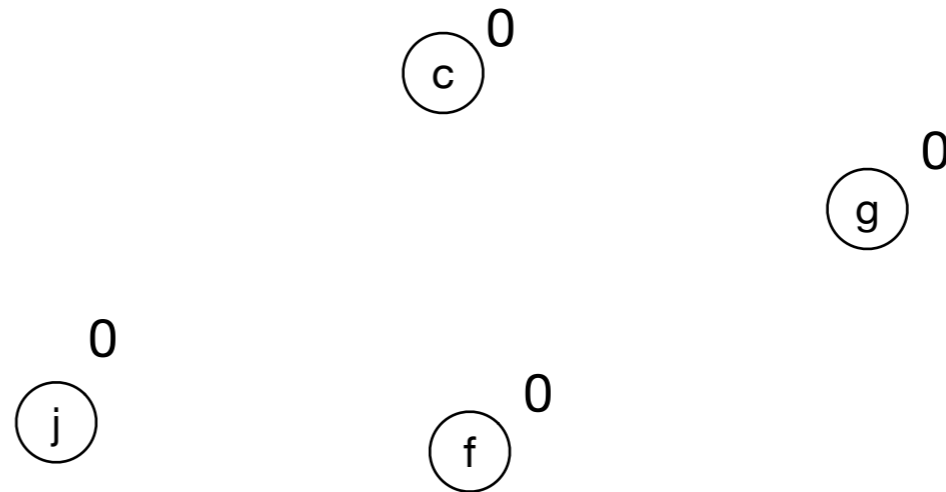


- $\{a, b, i, g, h, d\}$
- Can pick among four nodes: e

Topological Sort

- Example:

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

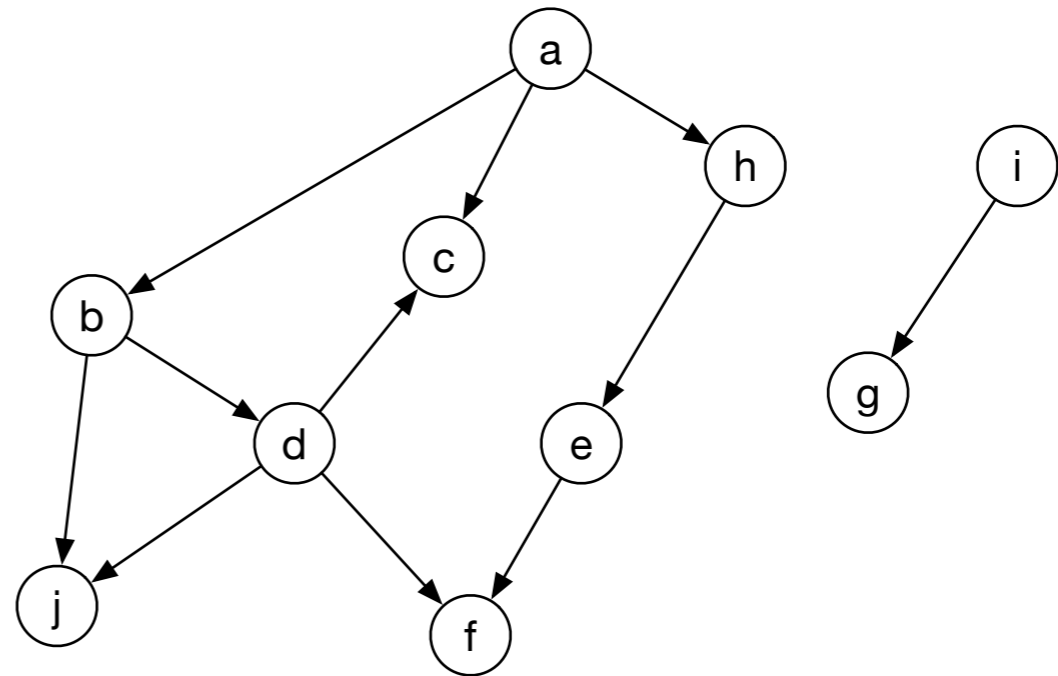


- $\{a, b, i, g, h, d, e\}$
- Can pick among four nodes in any order

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

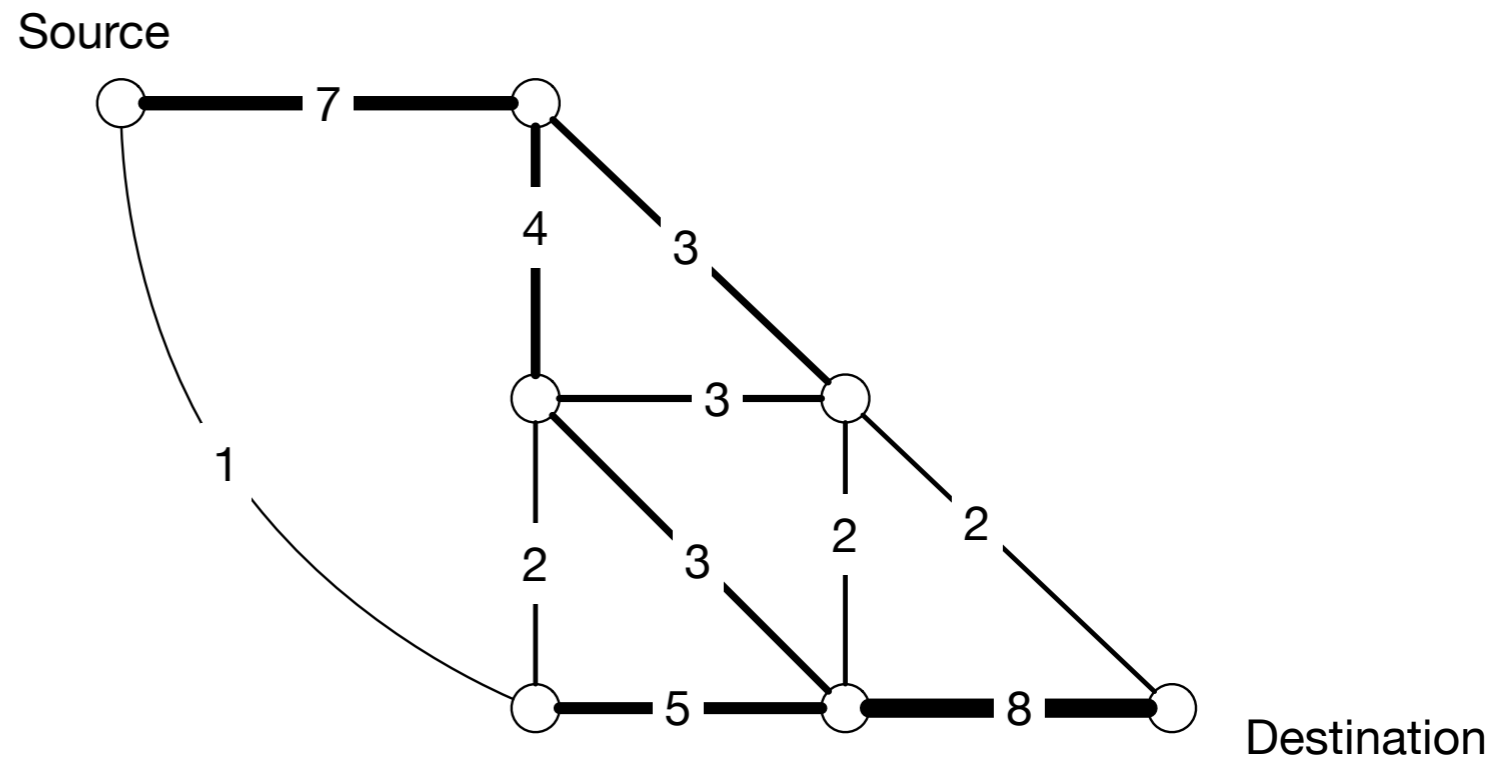
- Analysis for topological sort on $G = (V, E)$
 - Need to establish in-degrees:
 - Process all elements in an adjacency list
 - Correspond to edges
 - work $\sim |E|$
 - For each vertex:
 - find the vertex as a vertex of minimum in-degree
 - update in-degrees by going through the adjacency list
 - Latter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

