

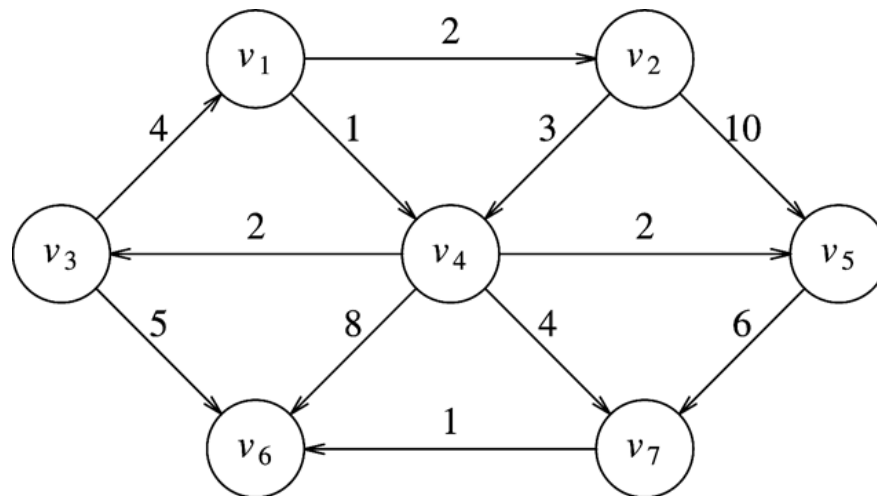
CS 10C: INTRO TO DATA STRUCTURES AND ALGORITHMS

Ryan Rusich

rusichr@cs.ucr.edu

Department of Computer Science
and Engineering
UC Riverside

Graphs



Graph

- A Graph $G = (V, E)$
 - A set V of vertices and a set E edges
- A graph is a way representing connections/relationships between pairs of objects in V
- Each edge is a pair (v, w) , where $v, w \in V$.
- Edges are either directed or undirected
 - Directed referred to as *ordered*
 - Directed graphs are called Digraphs
 - Directed graphs with no cycles (acyclic) are called DAGs
 - Undirected referred to as *unordered*

Graph

- An edge (v, v) is a loop
- Vertices can have incoming and outgoing edges
 - indegree – number of incoming edges of a vertex v
 - outdegree – number of outgoing edges of a vertex v
- A path in a graph is a sequence of vertices $w_1, w_2, w_3, \dots, w_N$ such that $(w_i, w_{i+1}) \in E$ for $1 \leq i < N$
 - A **simple path** is a path where all vertices are distinct except possibly first/last
 - **path length**: $N-1$ for N vertices
- Edges can have an associated cost called the **weight**.

Graphs - Definitions

- An **undirected graph** is **connected** if there is a path from every vertex to every other vertex
- A **directed graph** that is connected is called **strongly connected**.
- **Weakly connected** - If directed graph is not strongly connected, but removing direction makes graph connected
- **Complete graph** – edge between every pair of vertices.

Graph Representations

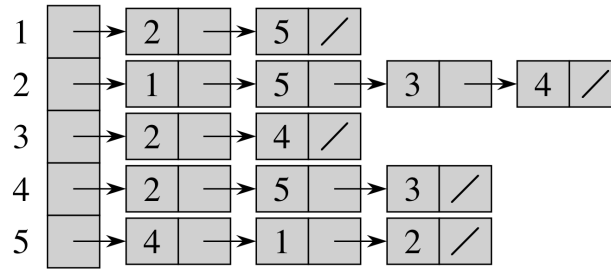
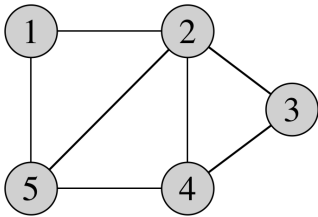
- Adjacency matrix – $|V|^2$ matrix, with **1** if there is an edge between two vertices, **0** otherwise.
 - Good for dense graphs.
 - Wasted space if not dense.
 - If graph is sparse, adjacency list is better.
- Adjacency list – for each vertex, store a list of vertices that share an edge
 - Example...

$$|E| = \Theta(|V|^2)$$

Adjacency list

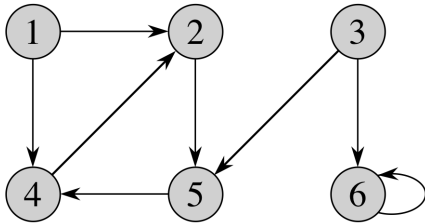
1	2, 4, 3
2	4, 5
3	6
4	6, 7, 3
5	4, 7
6	(empty)
7	6

Undirected Graph

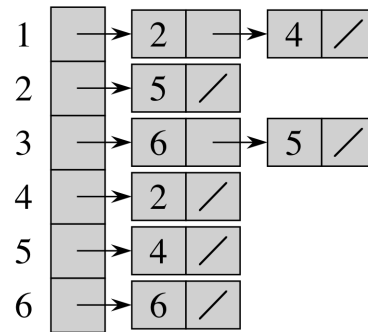


	1	2	3	4	5
1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0

Directed Graph



(a)

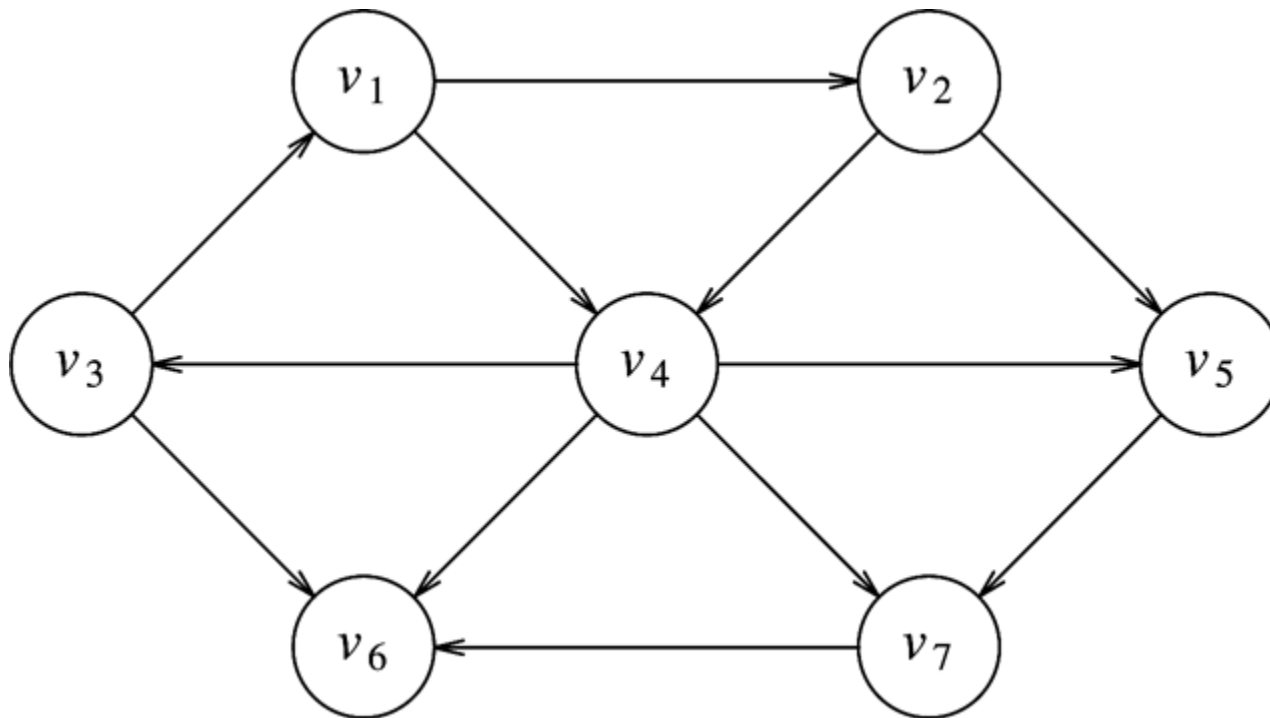


(b)

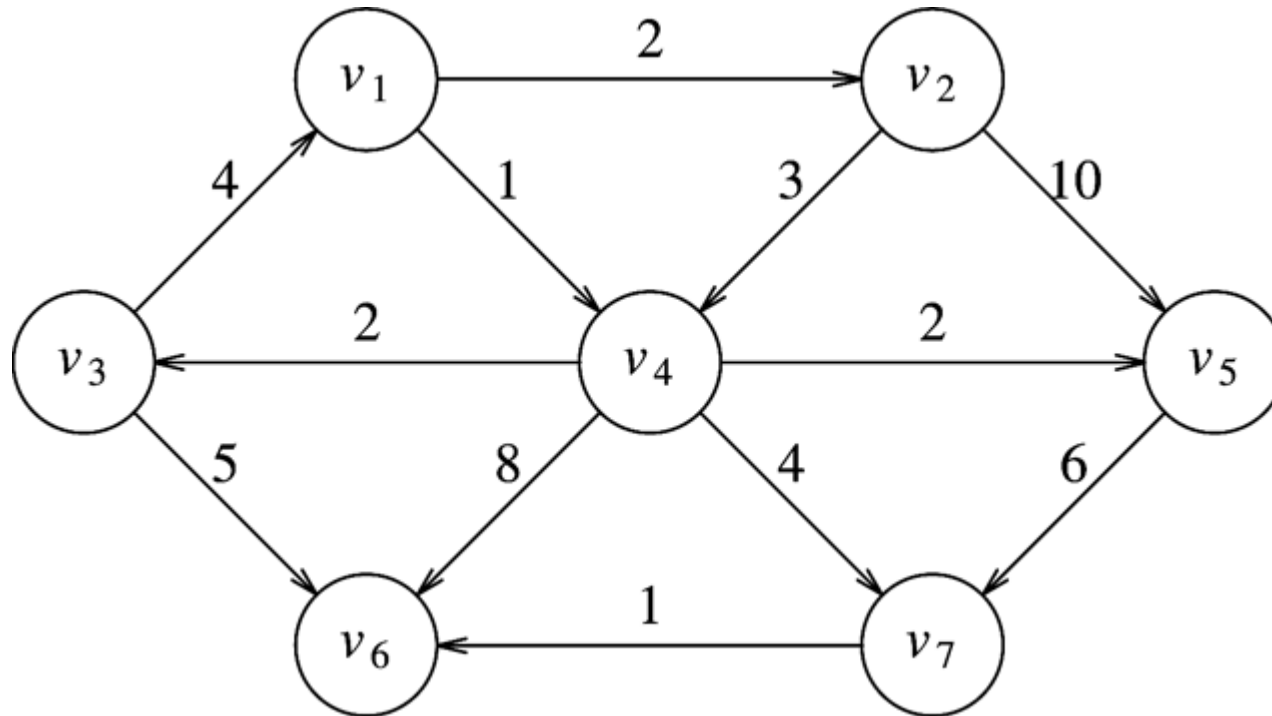
	1	2	3	4	5	6
1	0	1	0	1	0	0
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

(c)

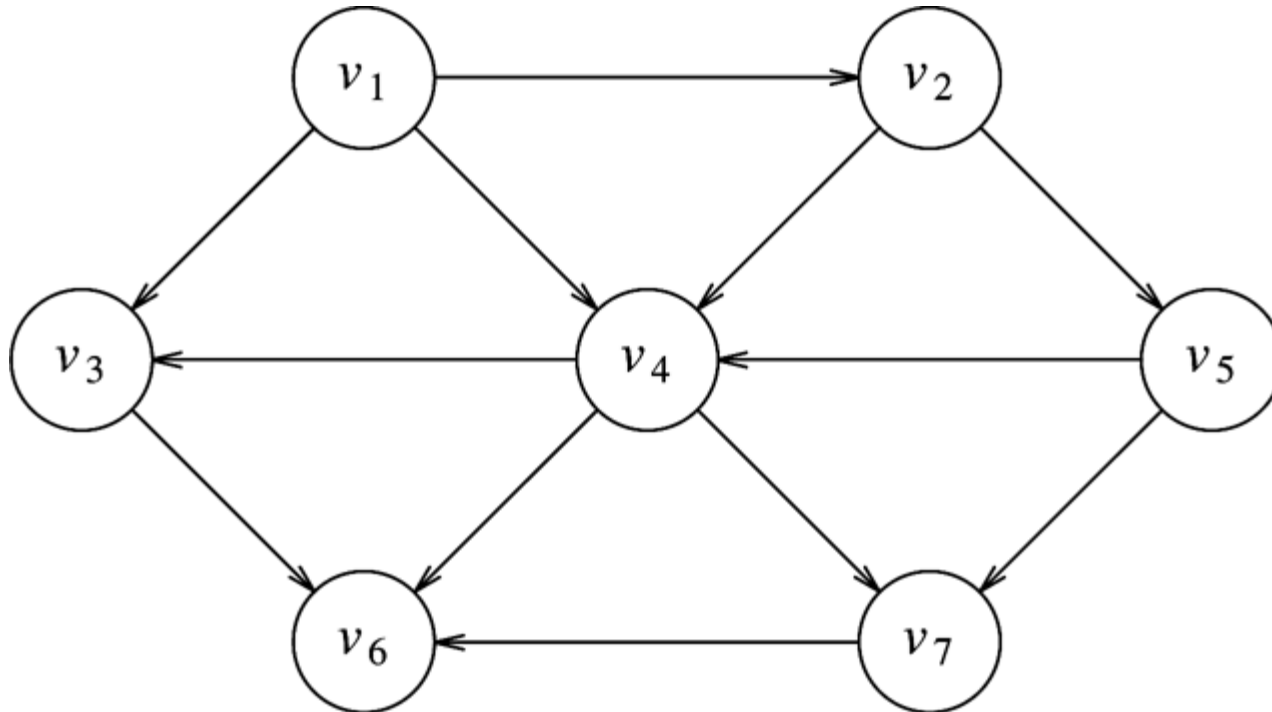
Directed Graph – Unweighted

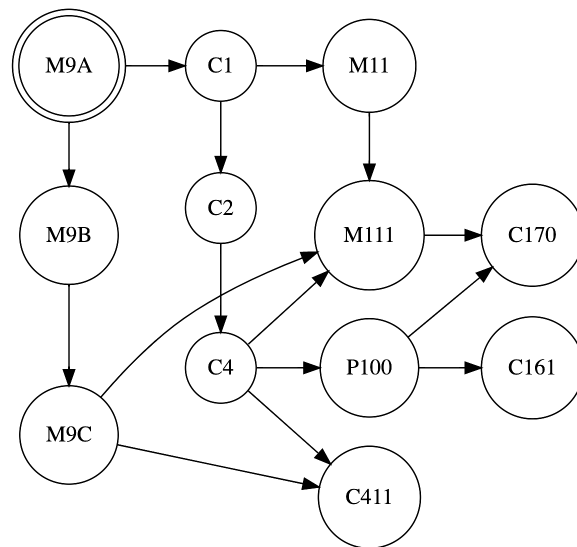


Directed Graph - Weighted



Directed Acyclic Graph

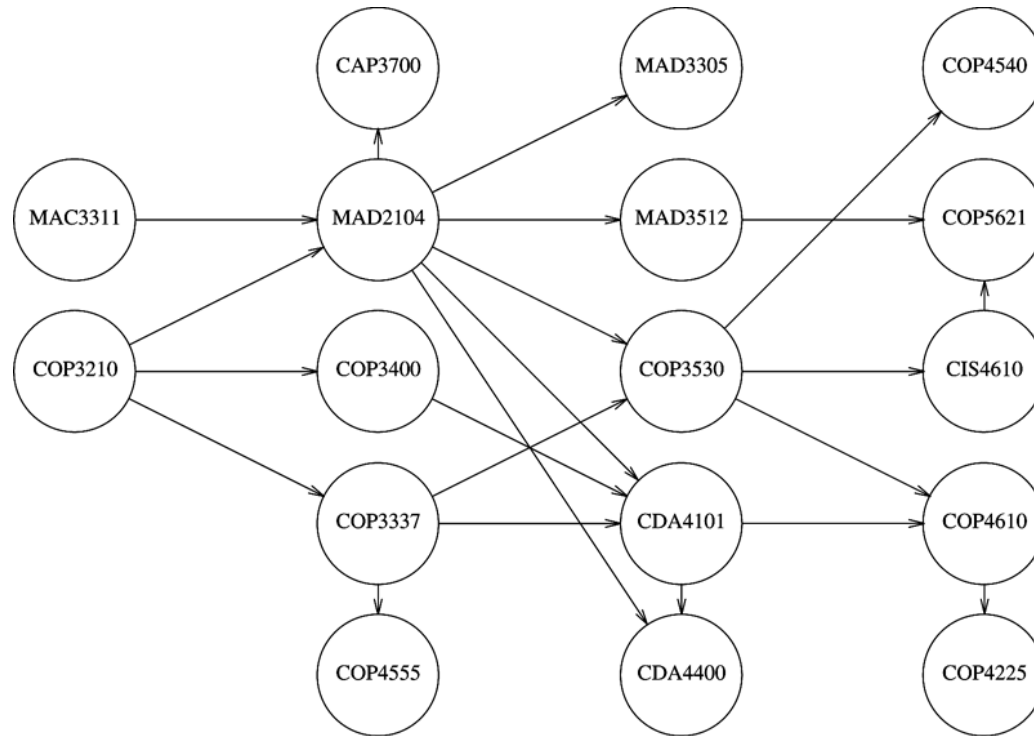




Topological Sort

- Linear ordering of vertices for a DAG (Directed Acyclic Graph).
- Every vertex comes before all vertices to which it has outgoing edges.
- Every DAG has at least 1 topological sort
 - 1 indicates a Hamiltonian path
 - 2 or more, no Hamiltonian path
 - Hamiltonian path – every vertex is visited exactly once.
- Topological Sort - used in scheduling jobs or tasks

Scheduling



Topological Sort

- **Simple algorithm:**

1. Find an initial vertex v with no incoming edges
2. Print v
3. Remove v from graph
4. Remove v 's outgoing edges from graph, updating effected vertices
5. Iterate (steps 1-4) over the remaining graph

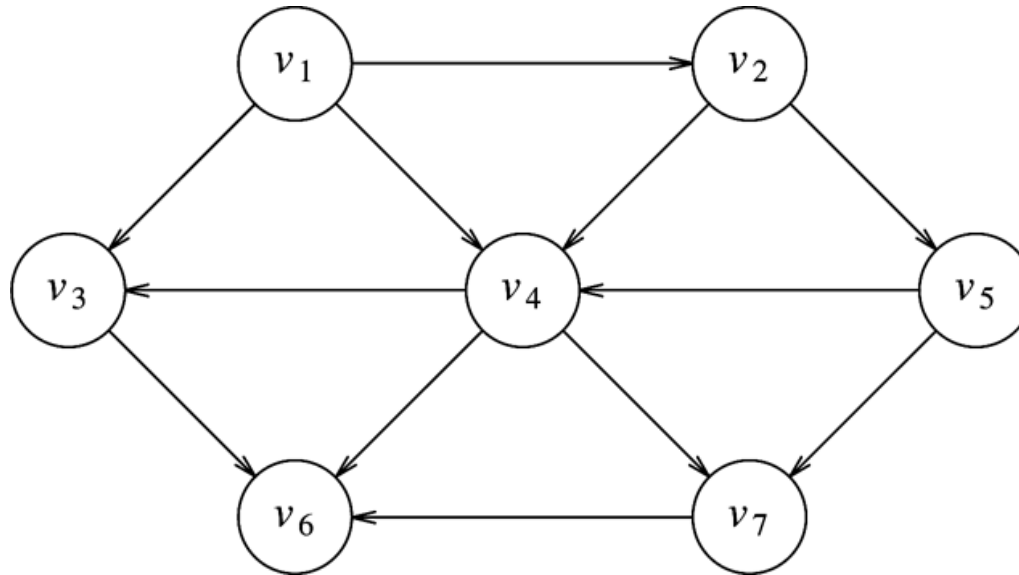
Running Time : $O(|V|^2)$

- Finding a vertex with in-degree zero, linear scan of V
- There are $|V|$ calls to do this

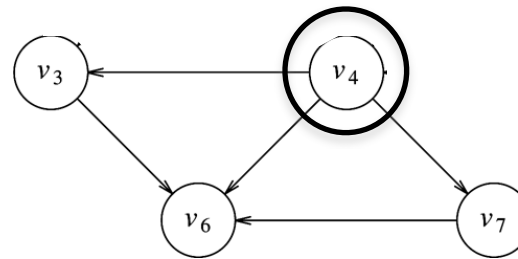
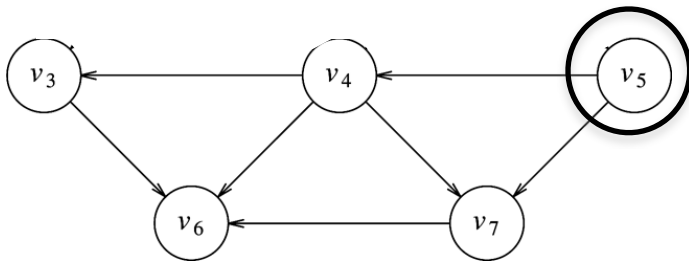
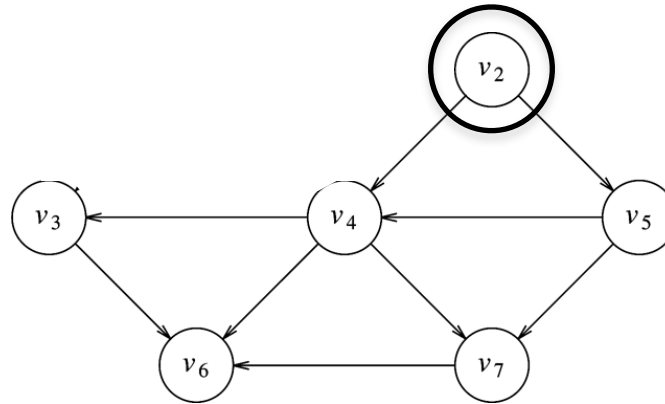
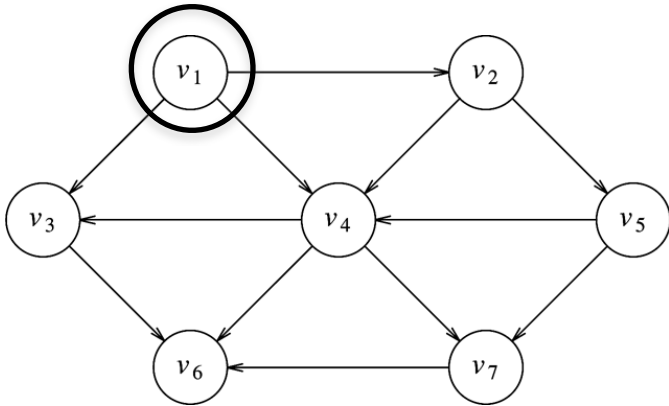
- **Speed up:**

- Use a Queue to store vertices with in-degree zero
- Only have to remove the front of Queue, skip linear scan of V
- **Running Time:** $O(|E| + |V|)$

Topological Sort

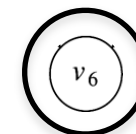
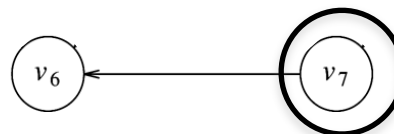
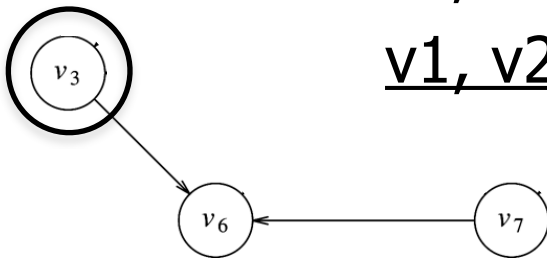


$v_1, v_2, v_5, v_4, v_3, v_7, v_6$

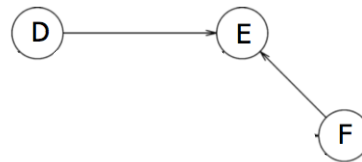
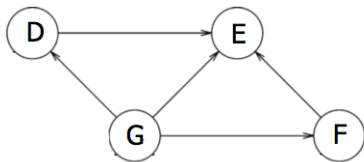
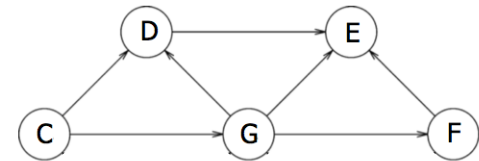
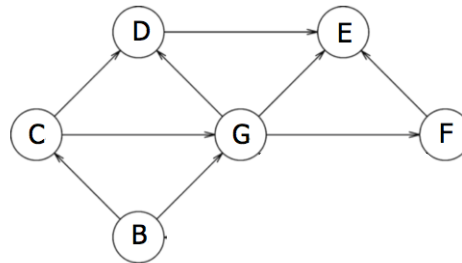
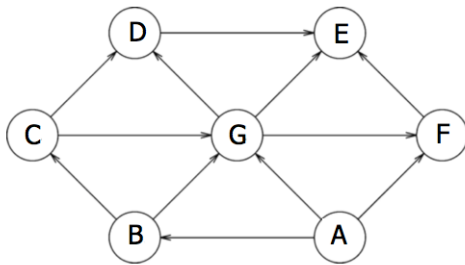


v1, v2, v5, v4, v3, v7, v6

v1, v2, v5, v4, v7, v3, v6

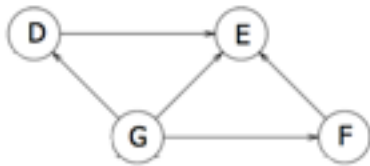
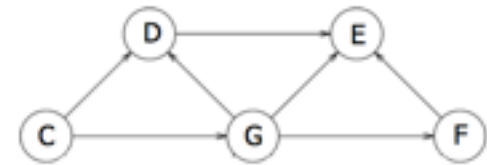
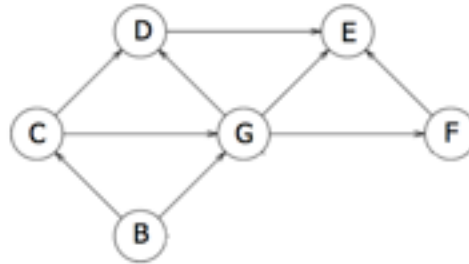
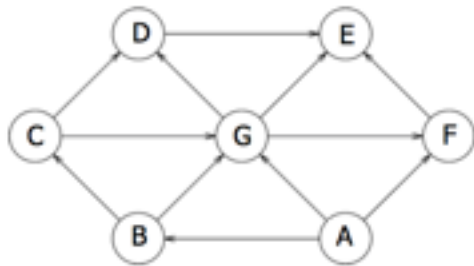


Topological Sort



A, B, C, G, D, E, F

Topological Sort

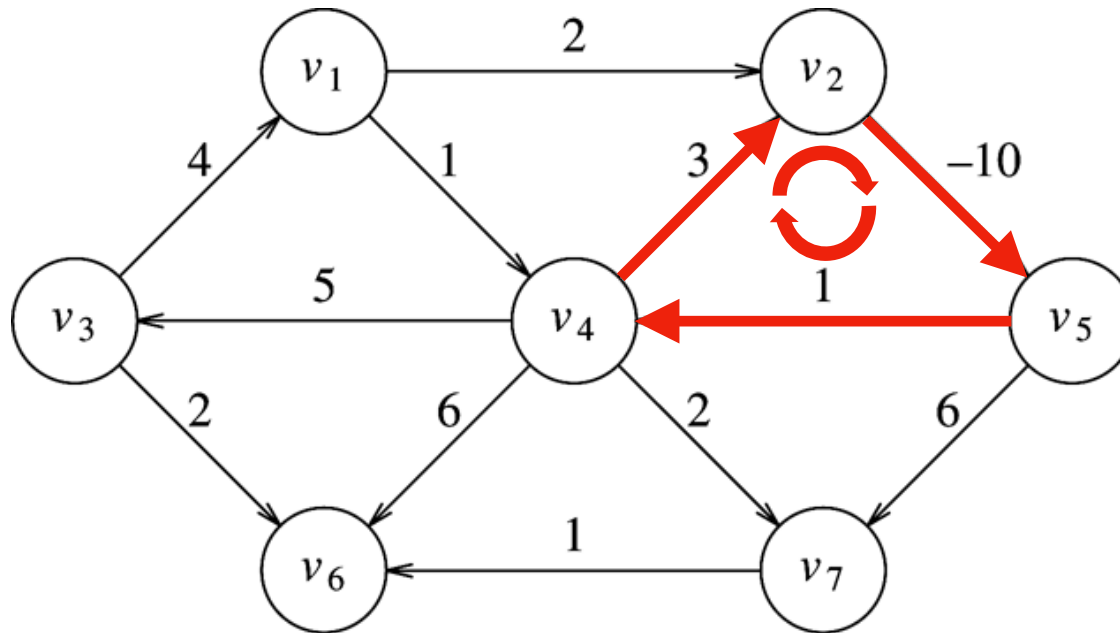


A, B, C, G, D, E, F

Single-Source Shortest Path

- **Problem:** find the path between two vertices that minimizes the number of edges or minimizes the sum of the weights of the edges.
 - Unweighted Graphs – number of edges
 - Weighted Graphs – sum of the costs
- **BFS** – solves for unweighted graphs
- **Dijkstra's** – solves for directed graph with positive weights
- **Bellman Ford** – solves for graphs that can include negative weights

Negative Cost Cycle



- Shortest Path is **NOT** defined.
- Reason: because you could complete cycle repeatedly and lower the cost.
- This is true for Dijkstra's and Bellman-Ford.

Single Pair Shortest Path Initialization Step

for each $u \in V - \{s\}$

$u.d = \infty$

$s.d = 0$

$Q = \emptyset$

ENQUEUE(Q, s)

- Set all distances to infinite, since a path from source not yet found.
- Set distance to source equal to zero (no cost to go from a node to itself).

BFS(V, E, s)

for each $u \in V - \{s\}$

$u.d = \infty$

$s.d = 0$

$Q = \emptyset$

ENQUEUE(Q, s)

while $Q \neq \emptyset$

$u = \text{DEQUEUE}(Q)$

for each $v \in G.\text{Adj}[u]$

if $v.d == \infty$

$v.d = u.d + 1$

ENQUEUE(Q, v)

(u, v)

if $v.d == \text{infinity}$

then update $v.d$ *to* $u.d + 1$

(u, v)

if $v.d == \text{infinity}$

then update $v.d$ *to* $0 + 1$

(u, v)

if $v.d == \text{infinity}$

then update $v.d$ *to* $1 + 1$

Dijkstra's - RELAX

$\text{RELAX}(u, v, w)$

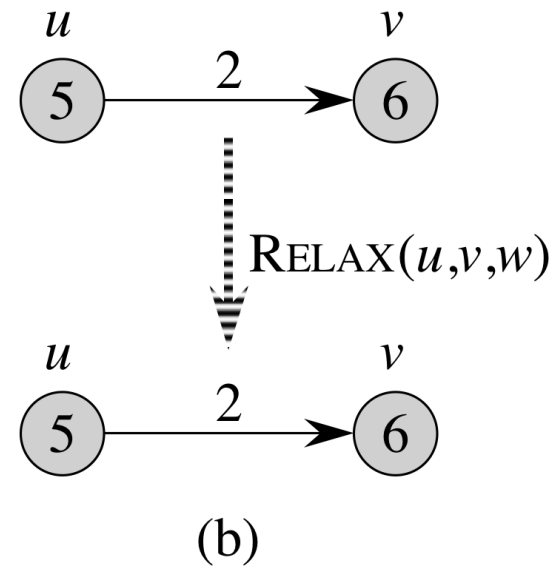
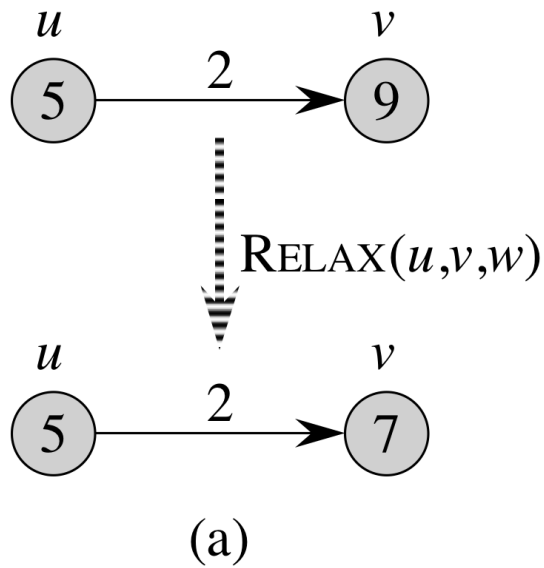
if $v.d > u.d + w(u, v)$

$v.d = u.d + w(u, v)$

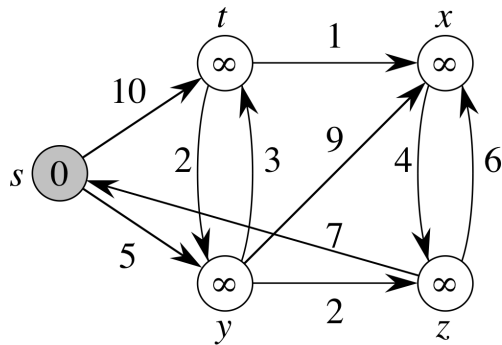
$v.\pi = u$

- Single-pair shortest path update distance step for Dijkstra's is similar to BFS.
- Main difference is that the weights of the edges, e.g. $w(u, v)$, are positive values.
- Edge weights can be larger than 1.
- Dijkstra's - edge weights cannot be negative.

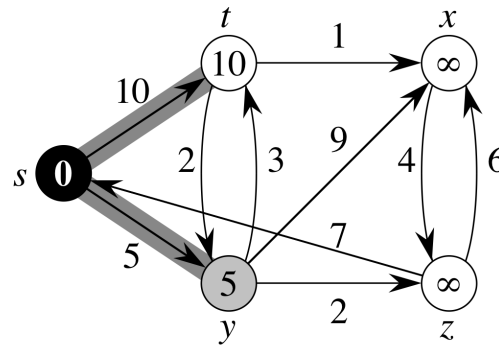
Dijkstra's - RELAX



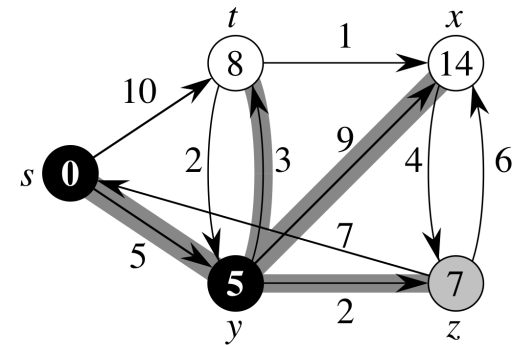
Dijkstra's - Example



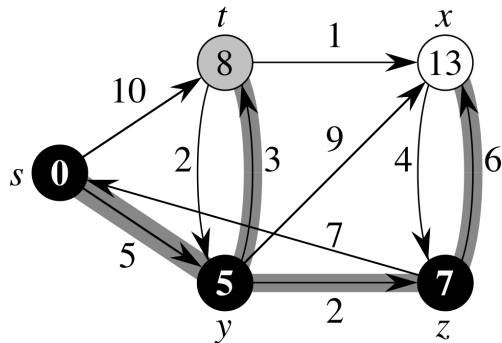
(a)



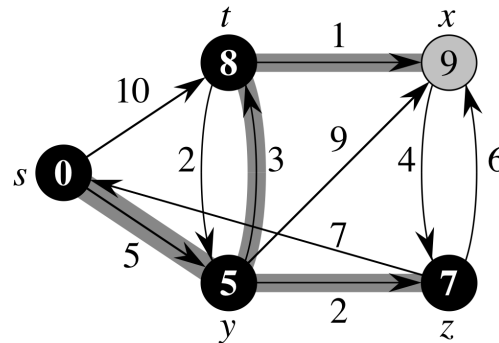
(b)



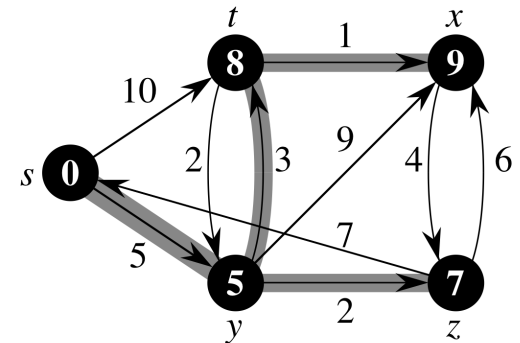
(c)



(d)

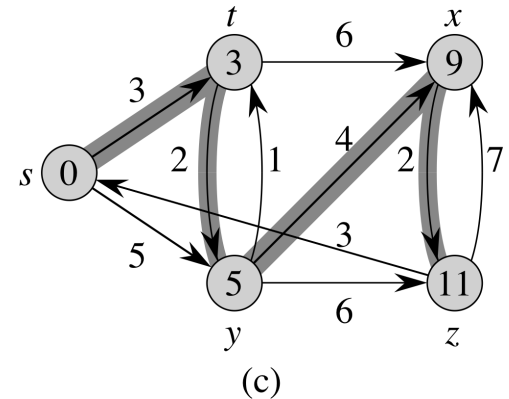
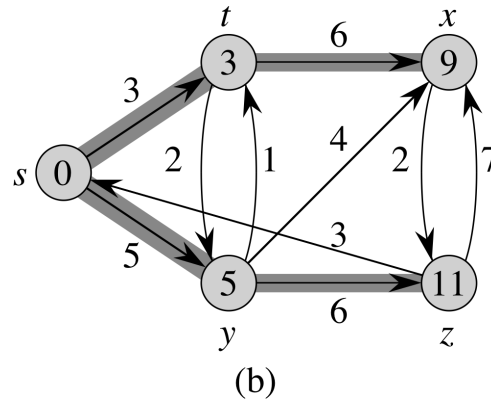
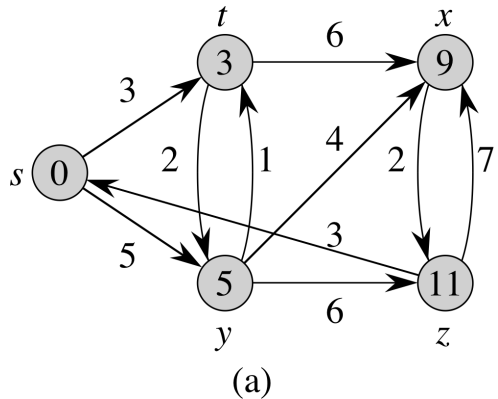


(e)



(f)

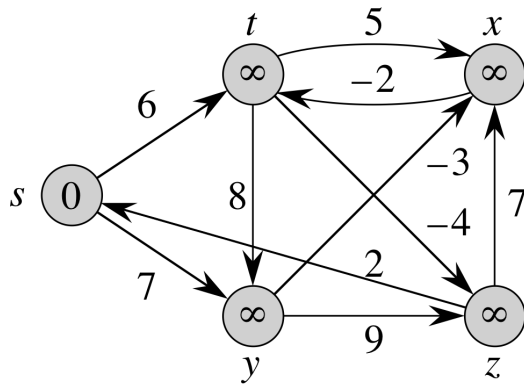
Non-Unique Shortest Path



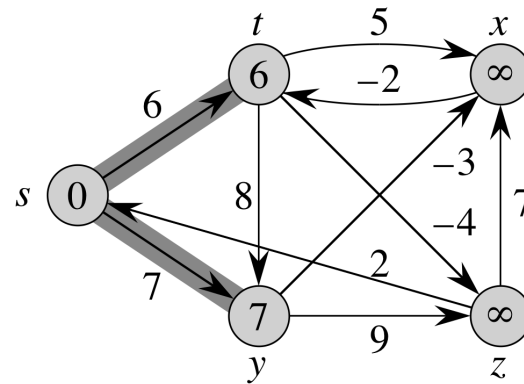
Bellman-Ford - CLRS

```
BELLMAN-FORD( $G, w, s$ )  
  INIT-SINGLE-SOURCE( $G, s$ )  
  for  $i = 1$  to  $|G.V| - 1$   
    for each edge  $(u, v) \in G.E$   
      RELAX( $u, v, w$ )  
  for each edge  $(u, v) \in G.E$   
    if  $v.d > u.d + w(u, v)$   
      return FALSE  
  return TRUE
```

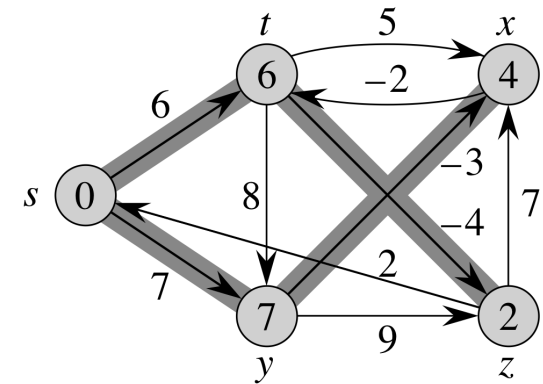
Bellman Ford - Example



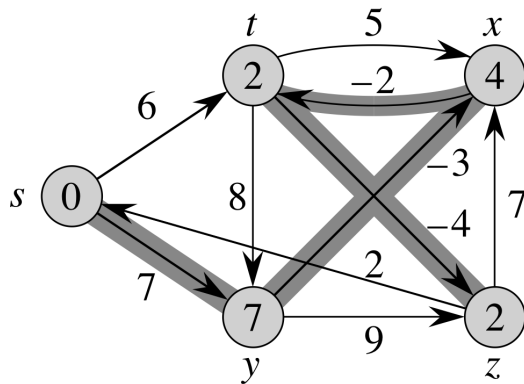
(a)



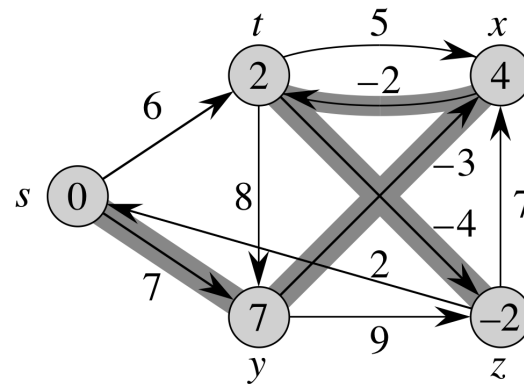
(b)



(c)

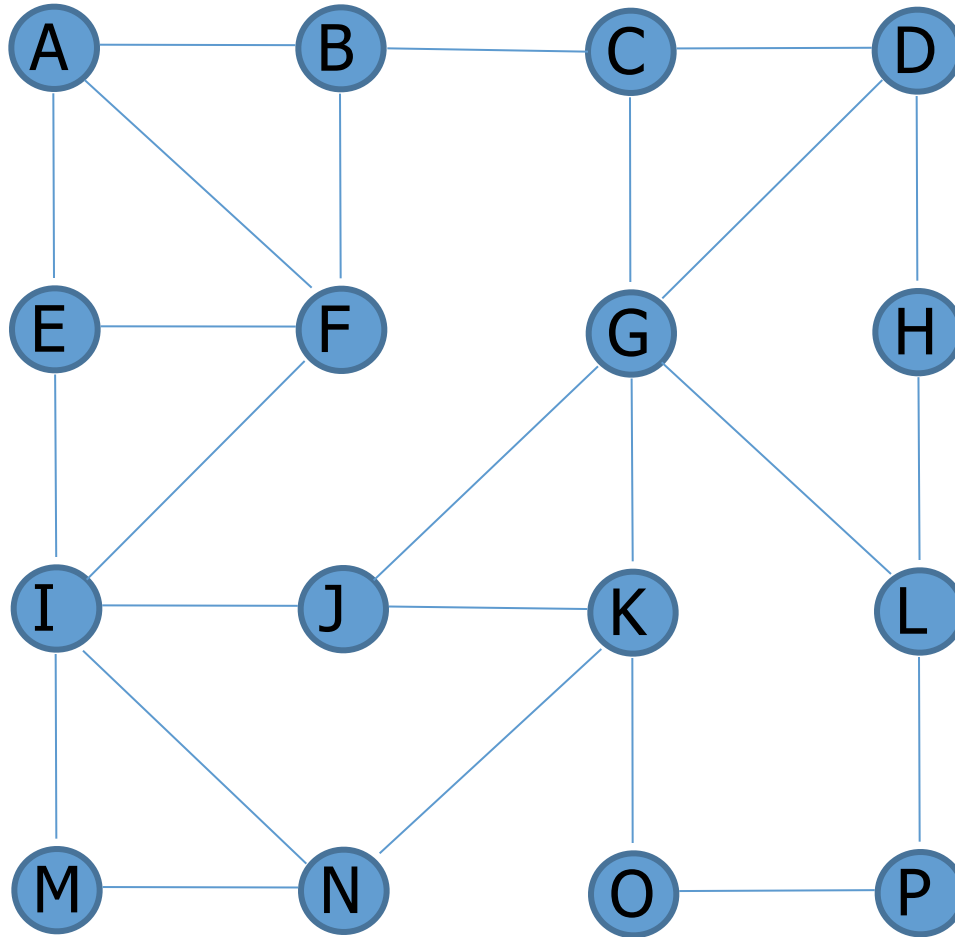


(d)



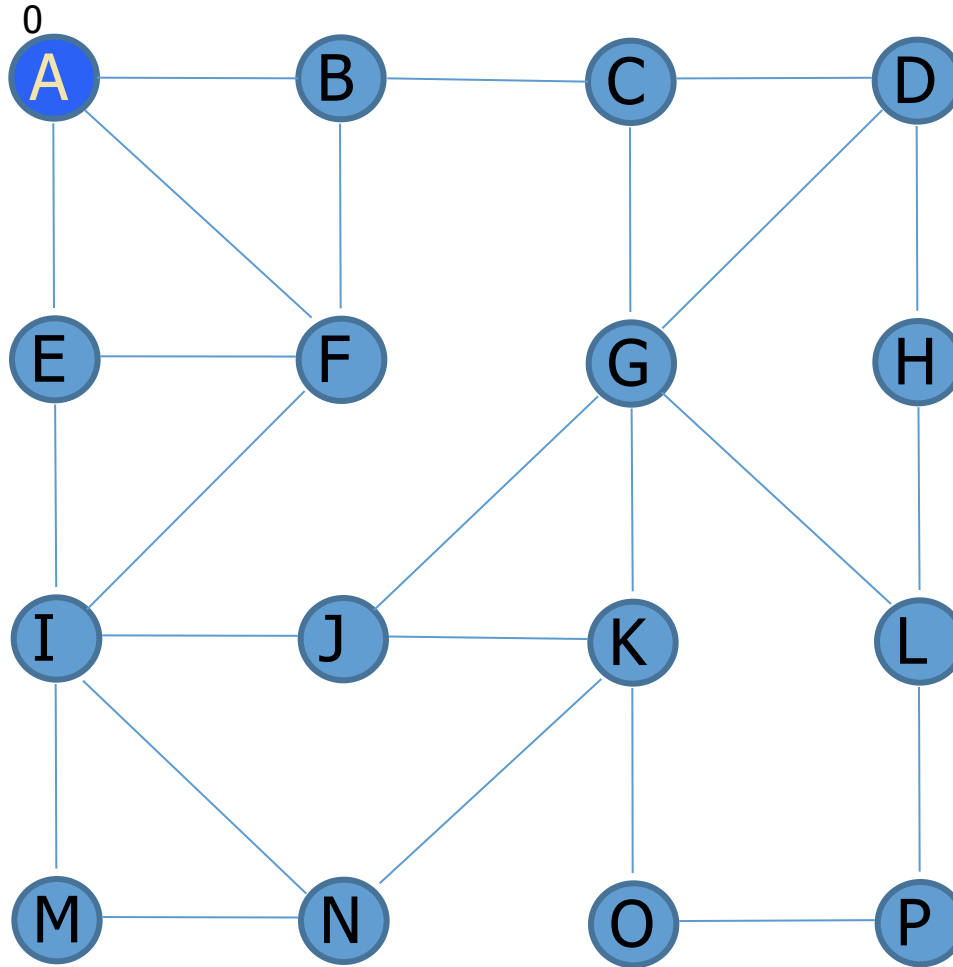
(e)

Breadth First Search (BFS)



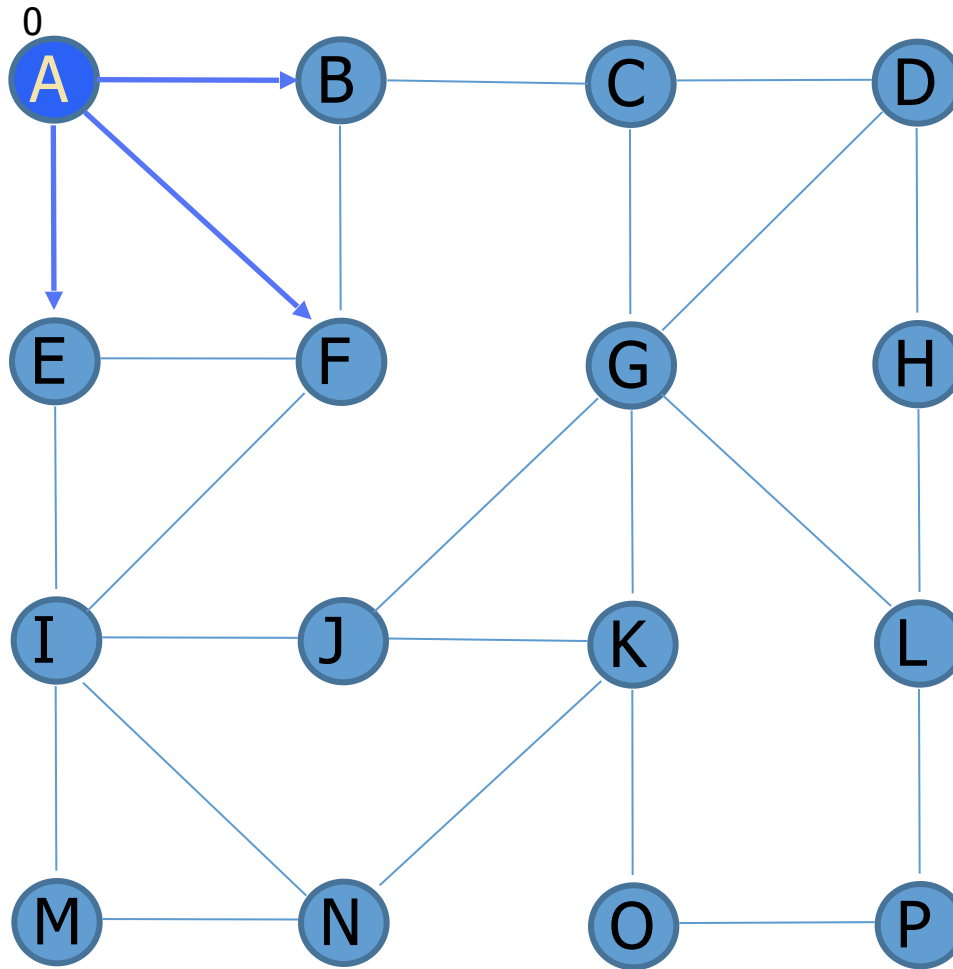
BFS

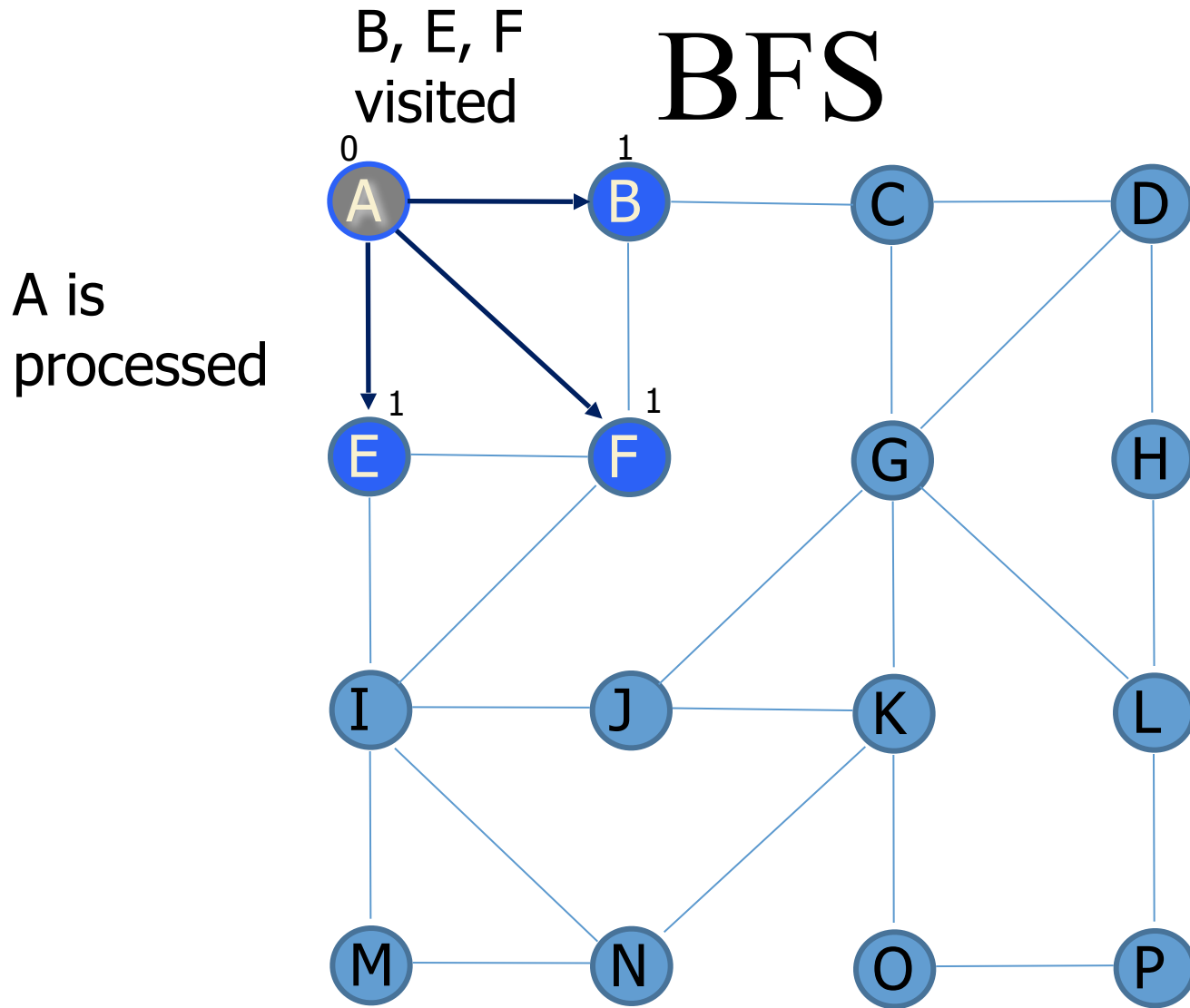
A visited



BFS

Explore
A's
neighbors

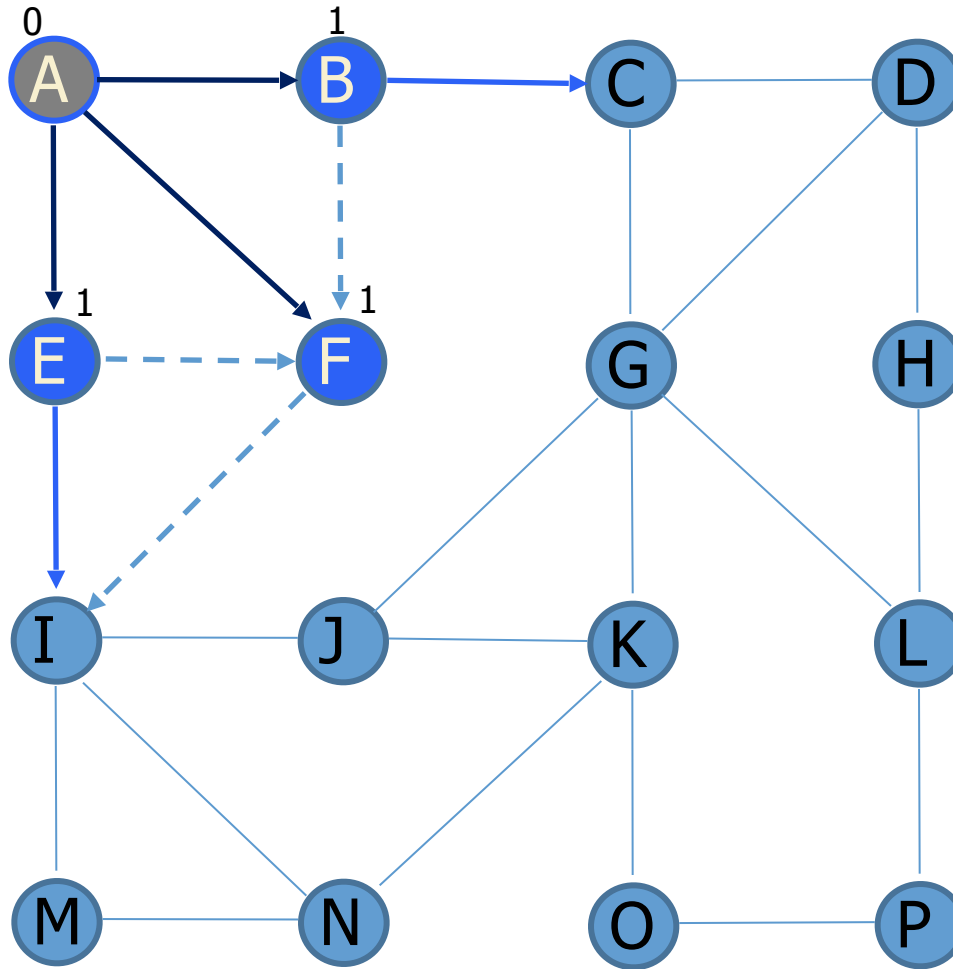




Explore B's
neighbors

BFS

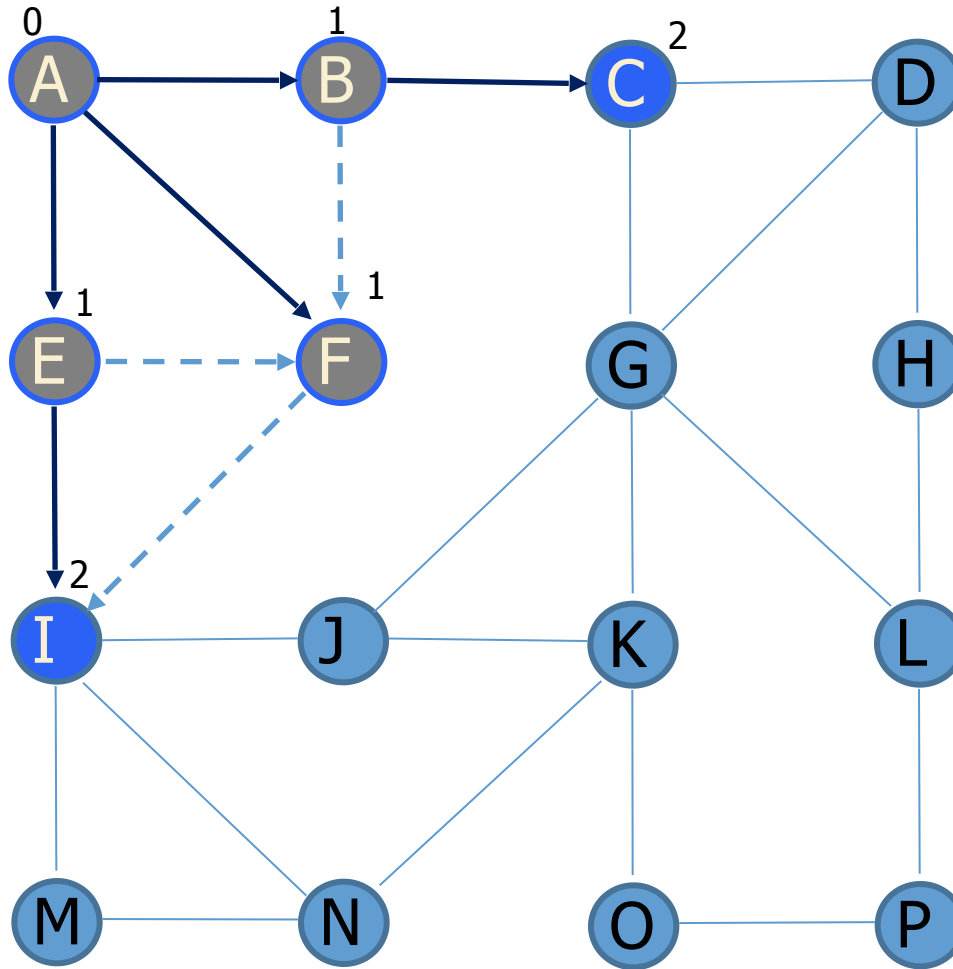
Explore E
then F's
neighbors



BFS

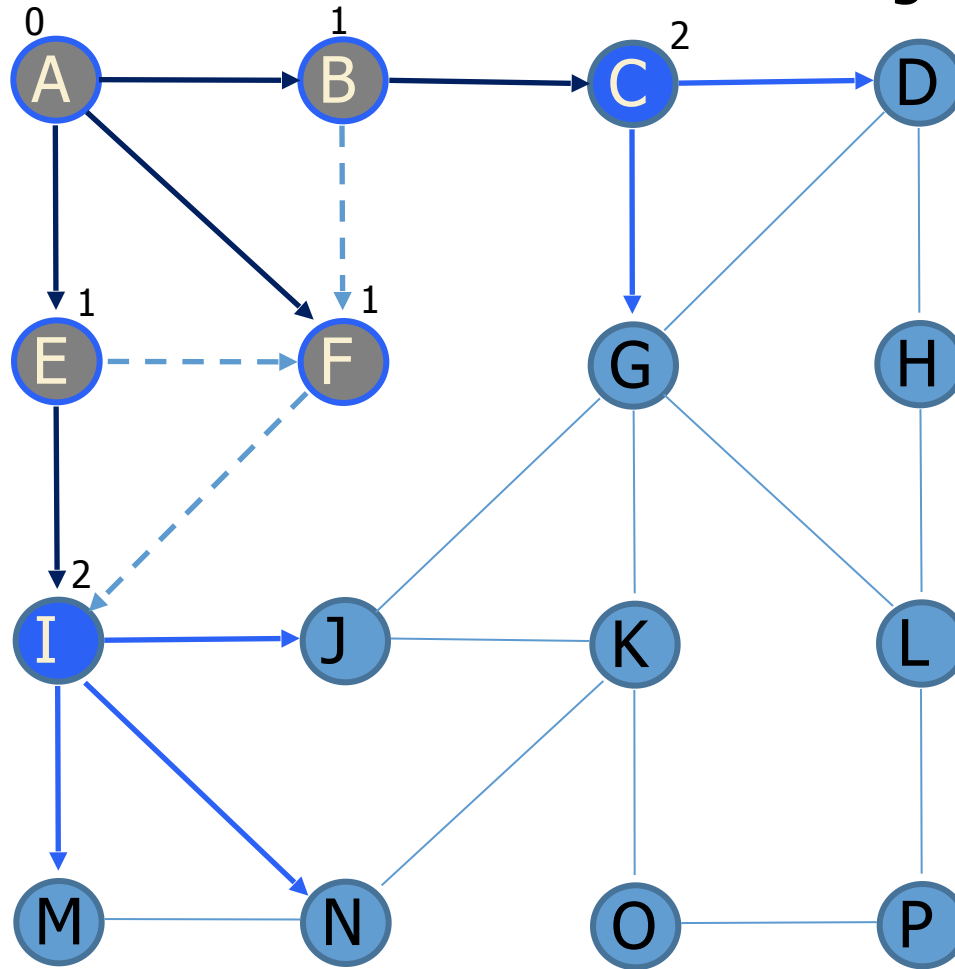
C, I
visited

B, E, F
are
processed



BFS

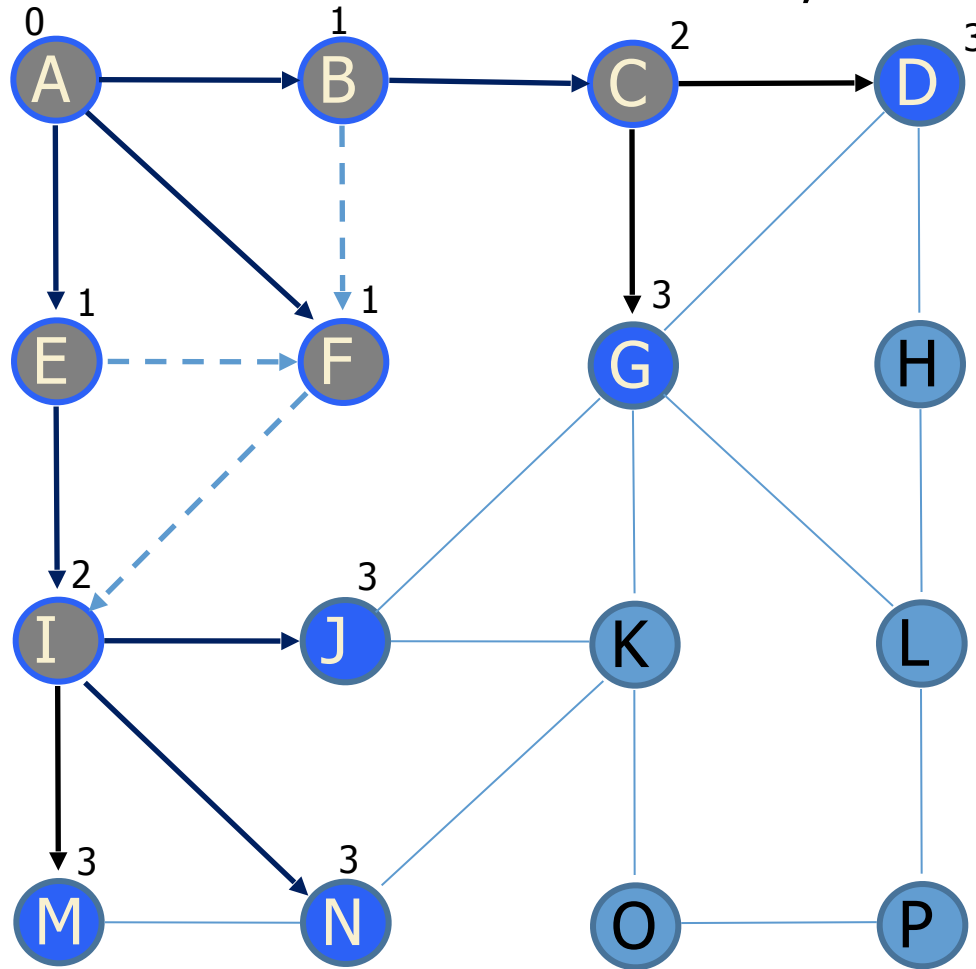
Explore C's
neighbors



Explore I's
neighbors

BFS

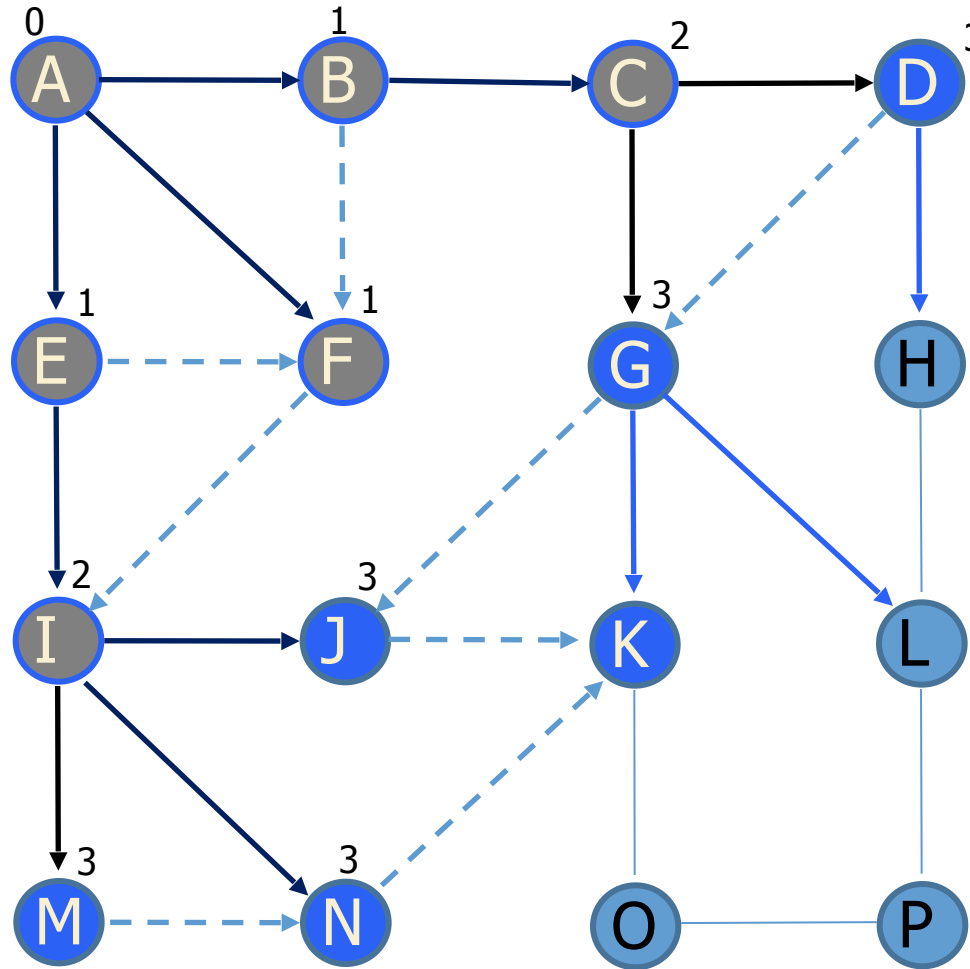
C processed
D,E visited



I processed
J,M,N
visited

BFS

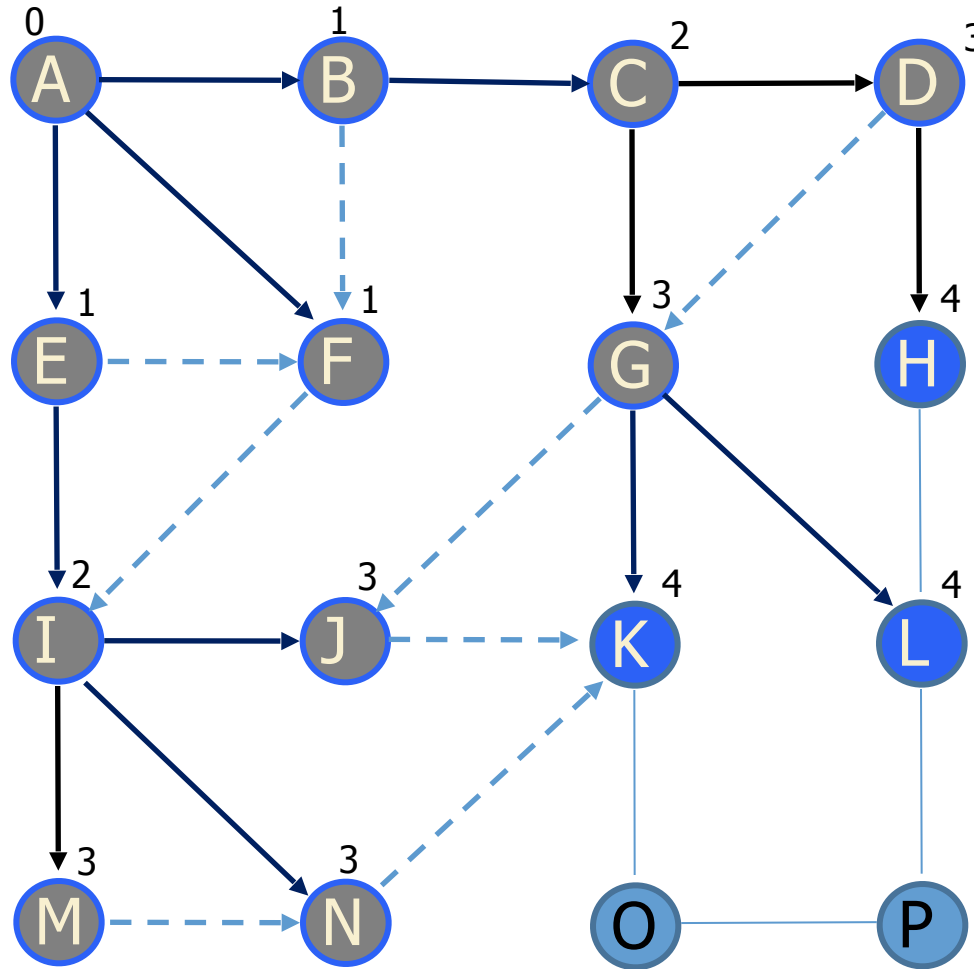
Explore D's, G's
neighbors



Explore
J's, M's, N's
neighbors

BFS

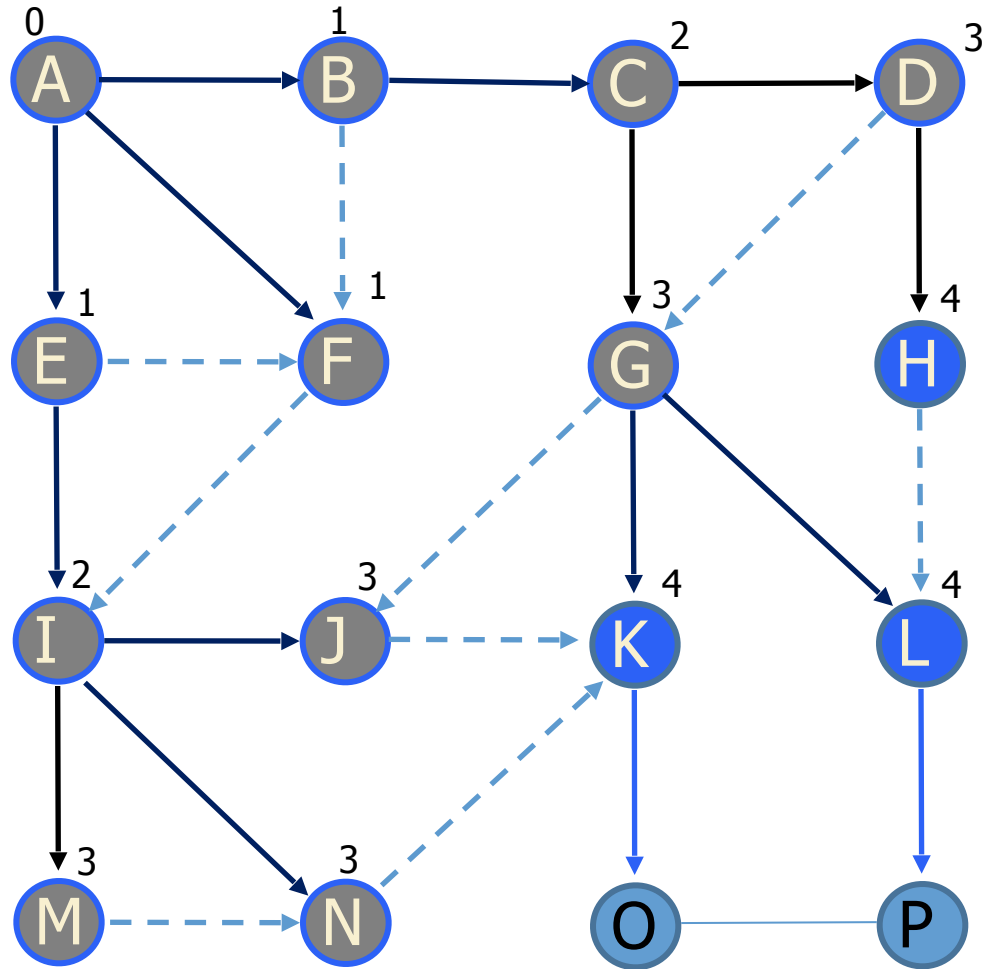
D, G
processed



H, K, L
visited

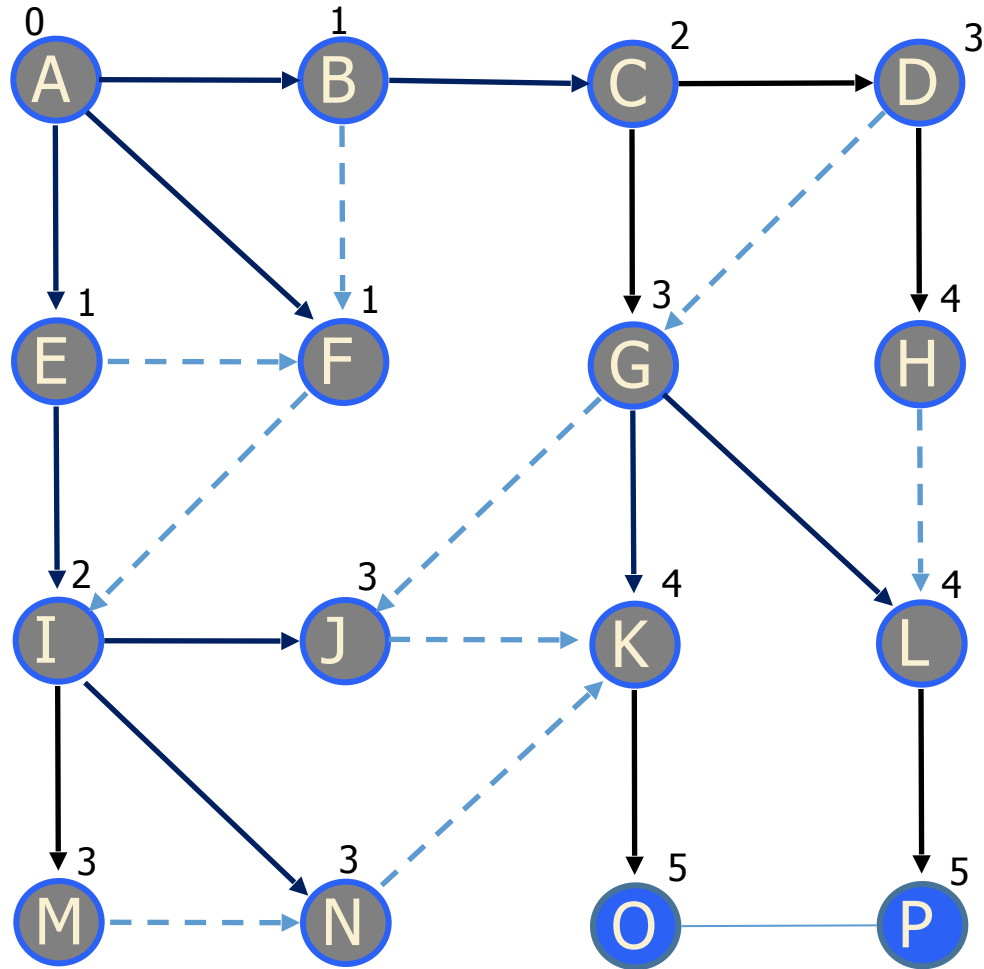
J, M, N
processed

BFS

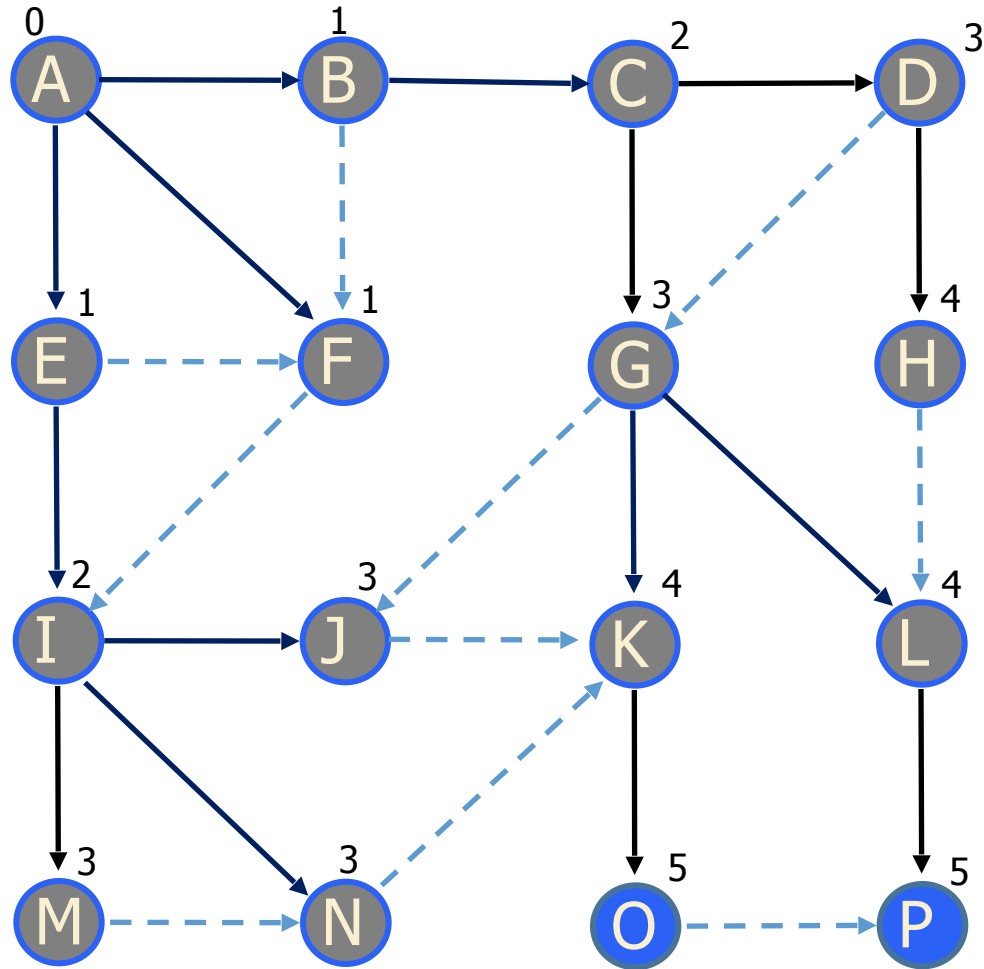


Explore
H, K, L
neighbors

BFS

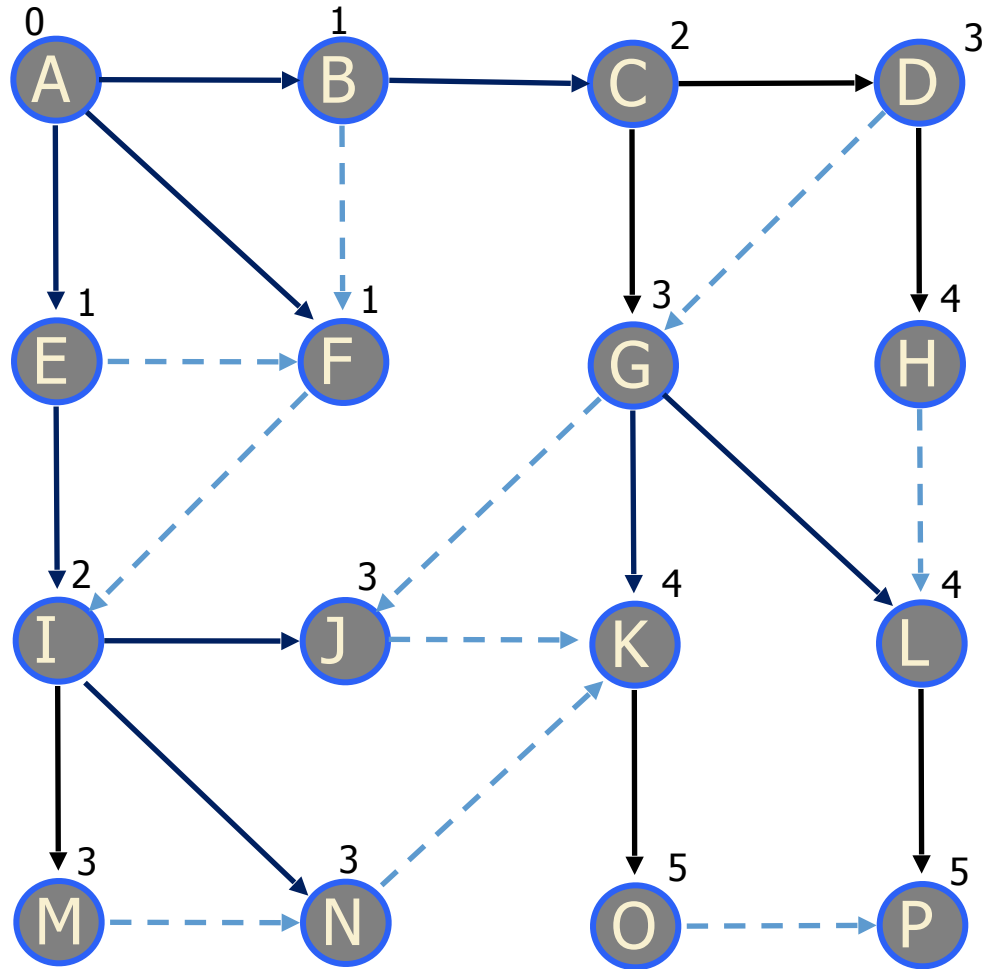


BFS



Explore O's
neighbor

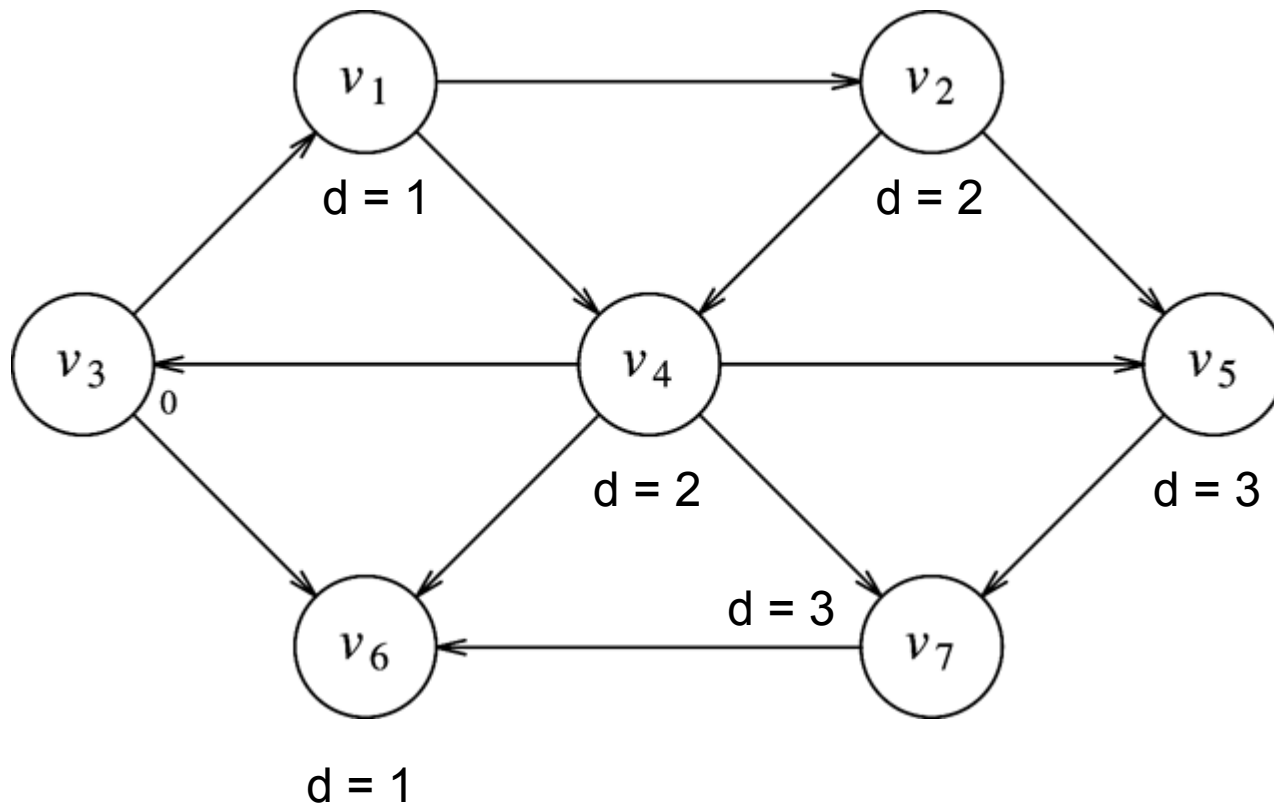
BFS



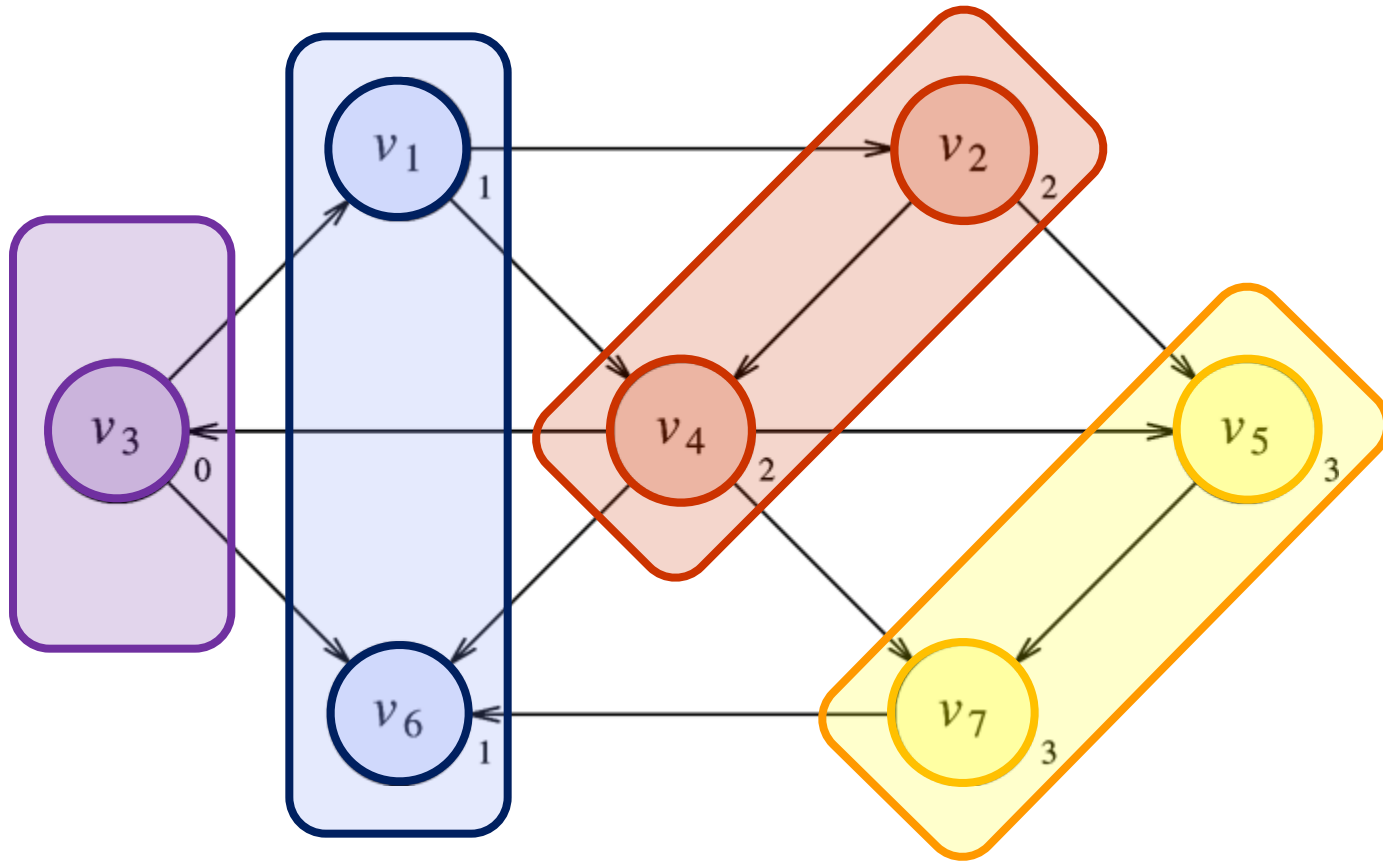
Done

$$\text{Running Time} = O(|E| + |V|)$$

BFS – In Class Activity

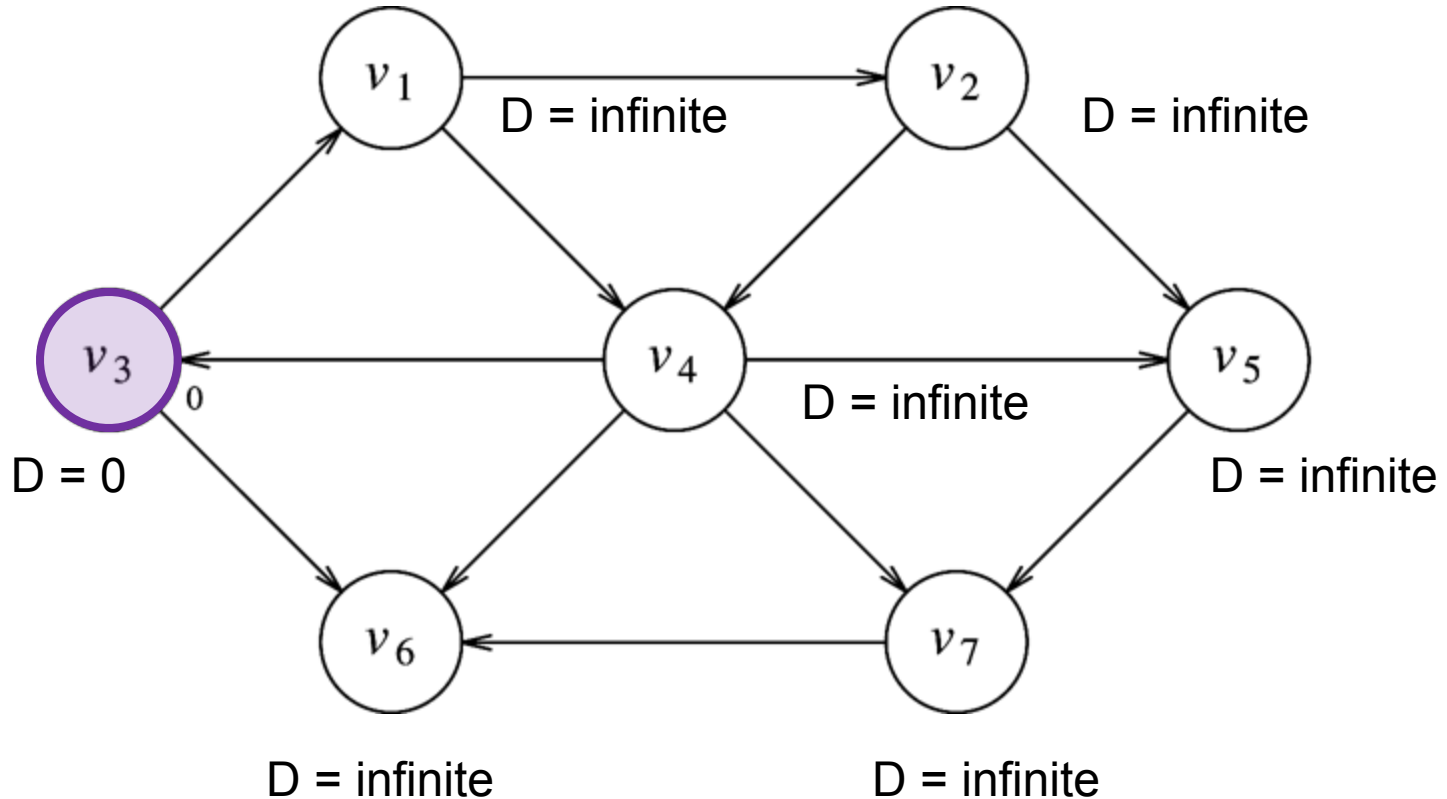


BFS – In Class Activity

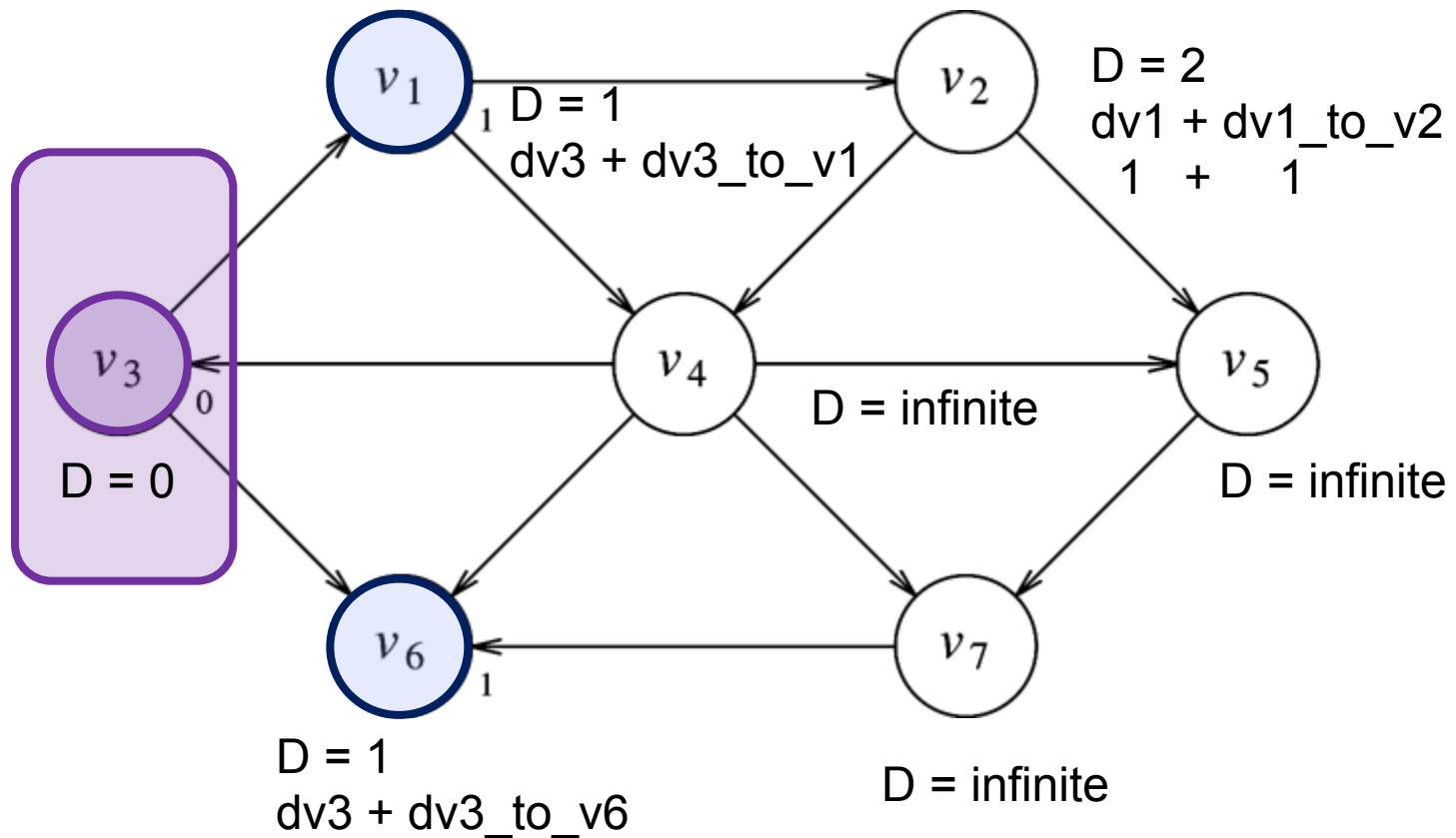


Running Time = $O(|E| + |V|)$, every V is explored, all E in worst case.

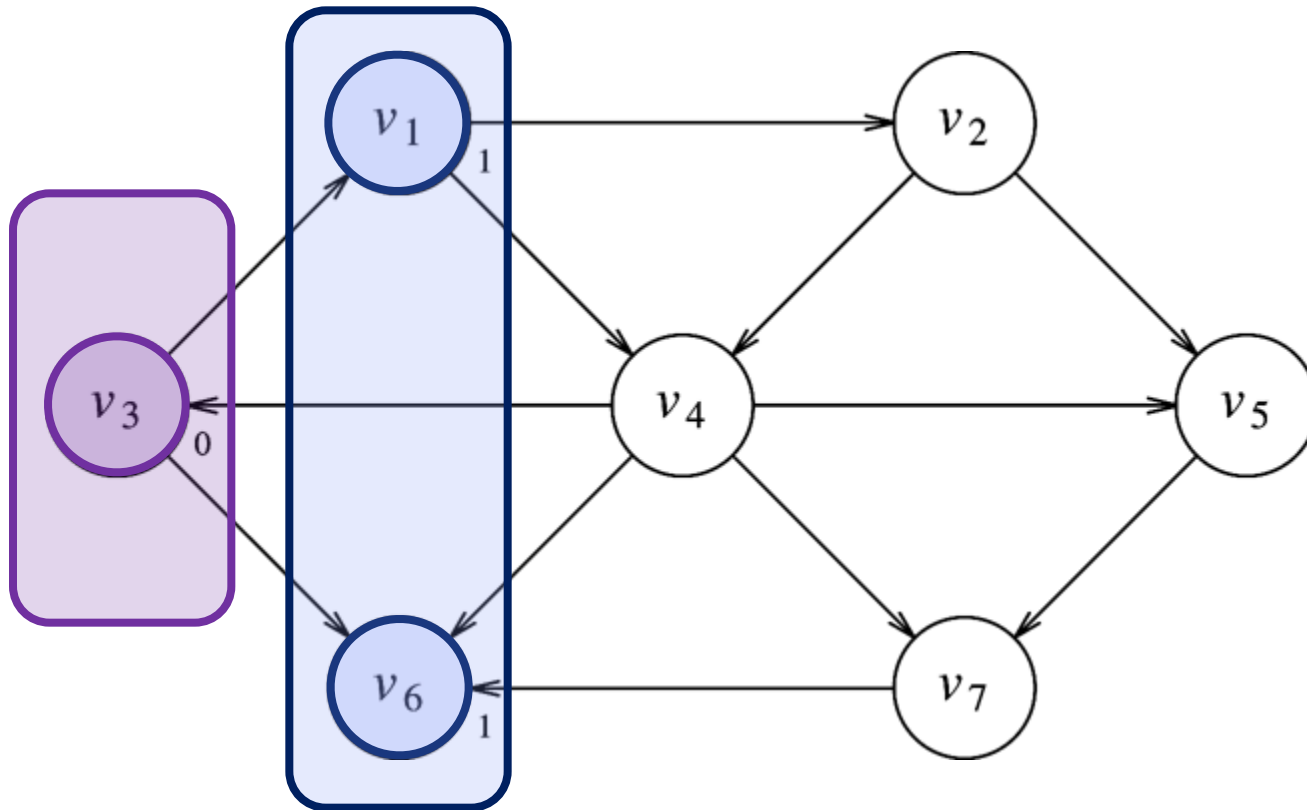
BFS – In Class Activity



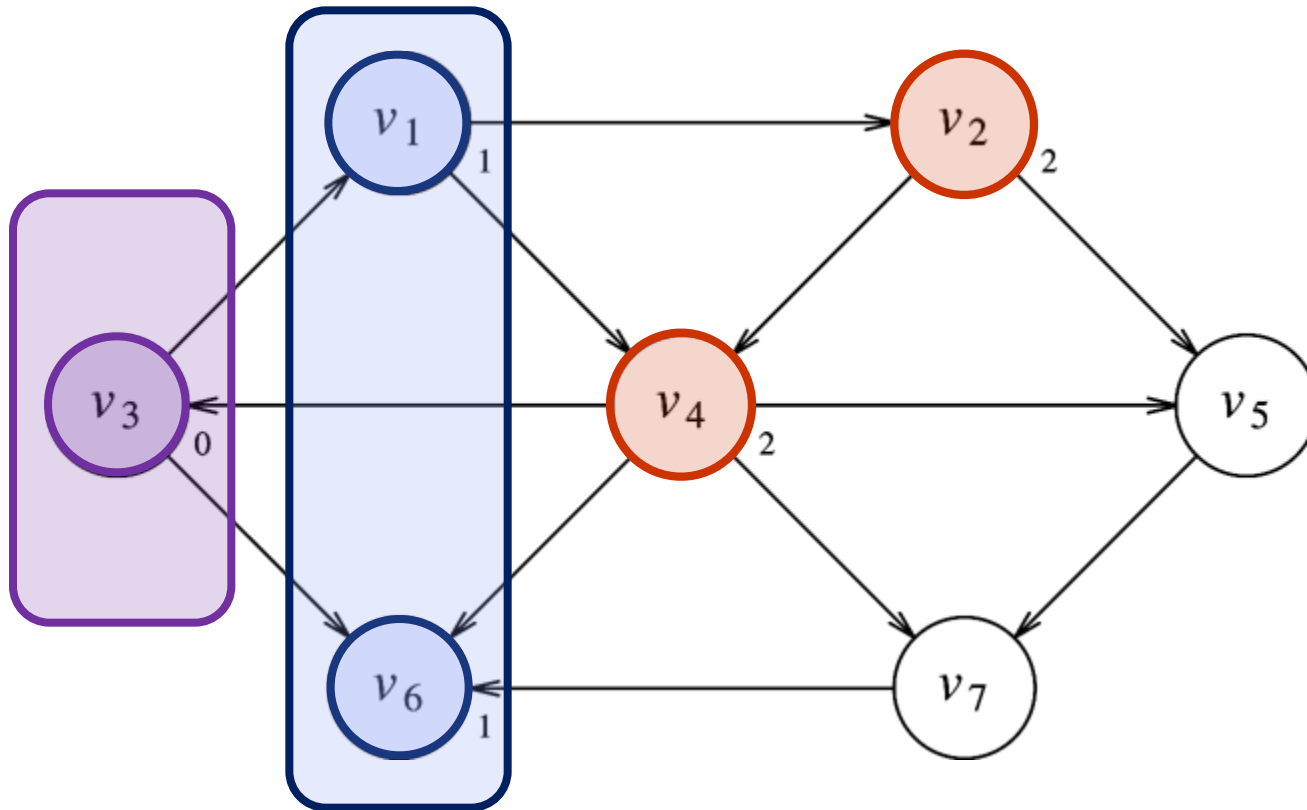
BFS – In Class Activity



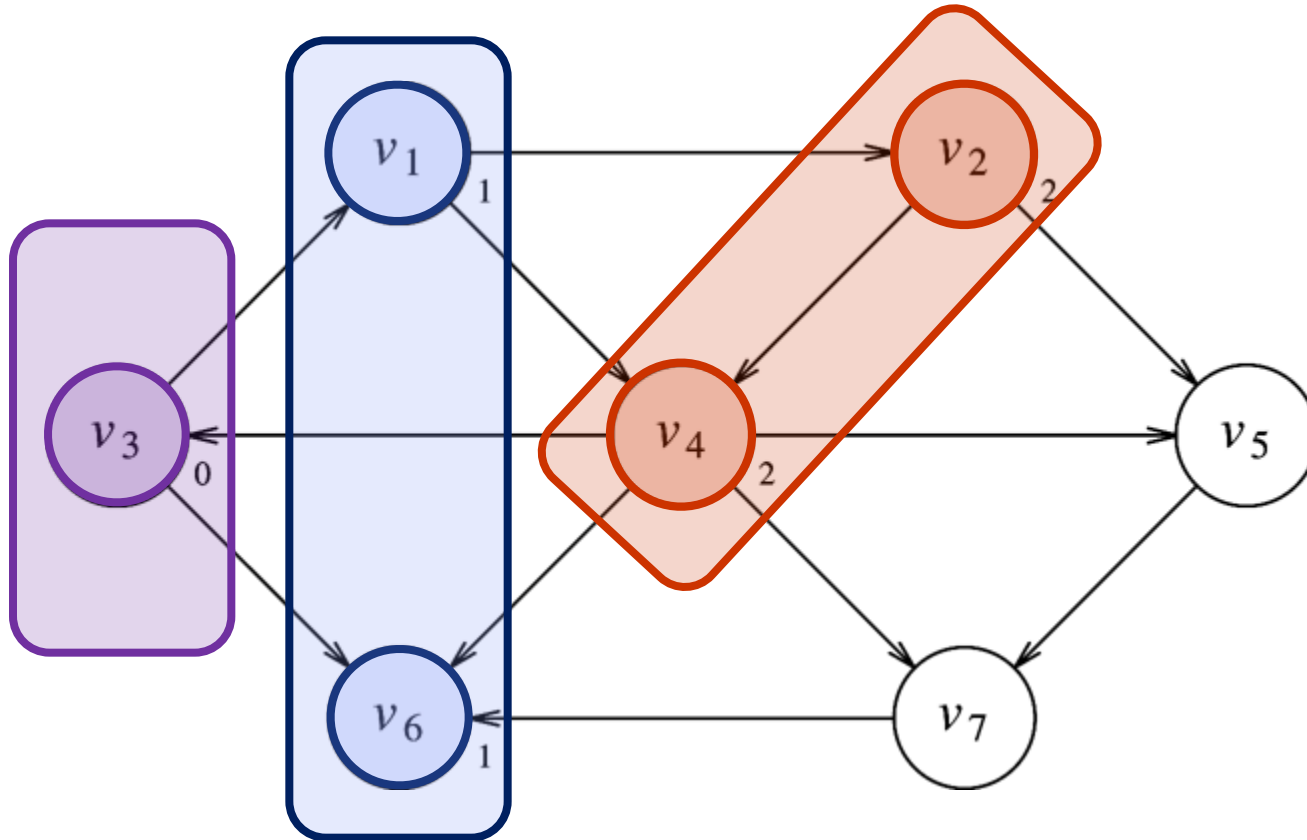
BFS – In Class Activity



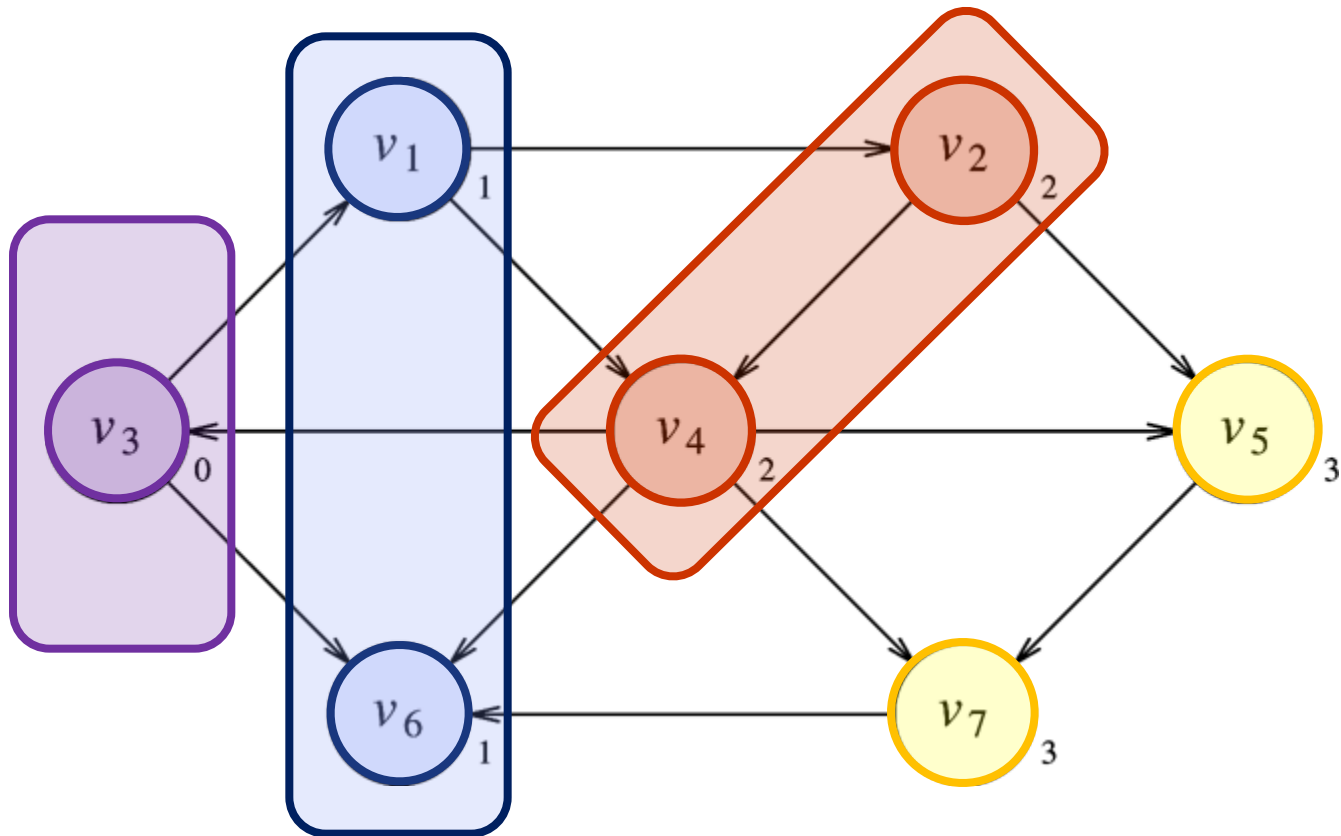
BFS – In Class Activity



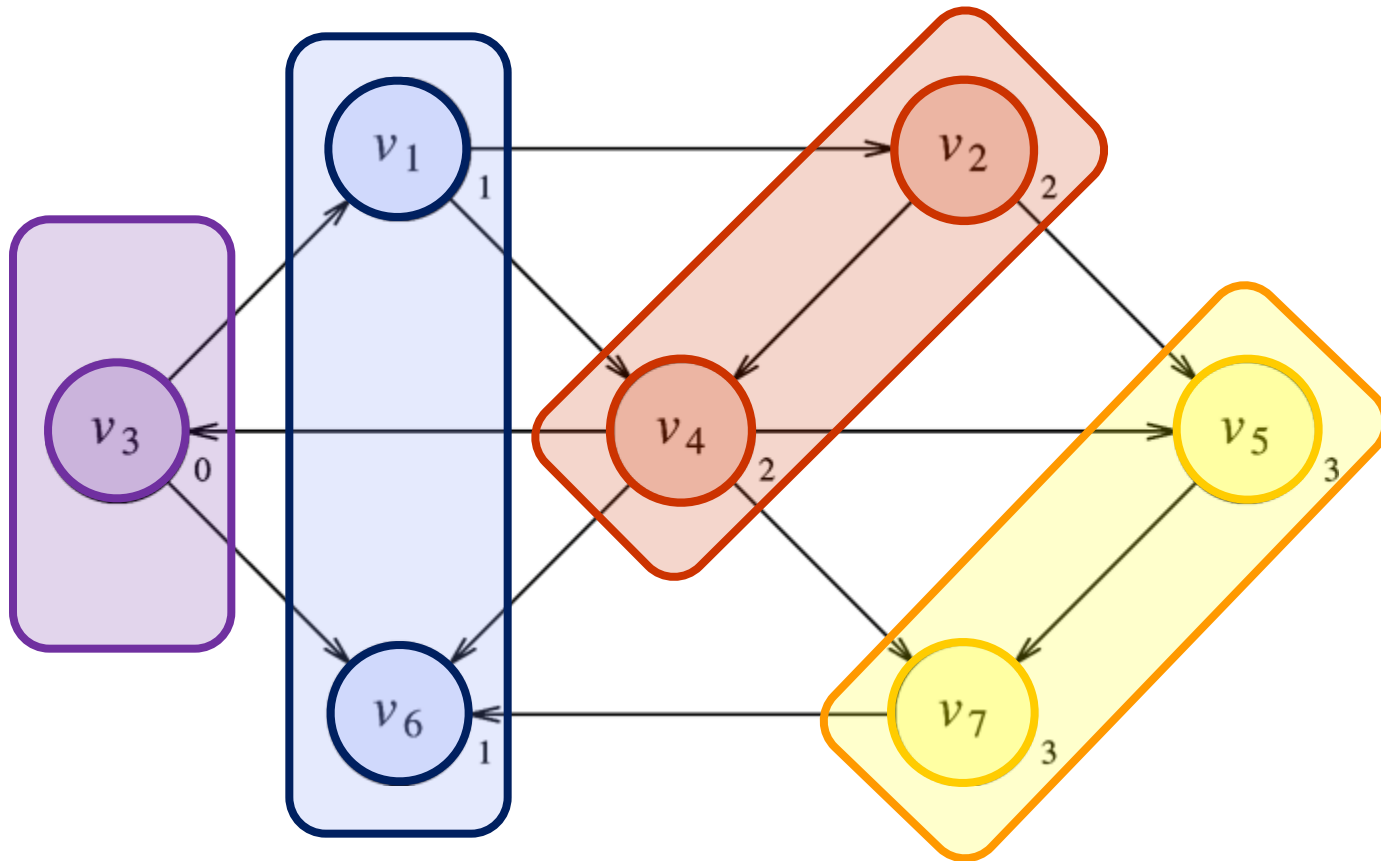
BFS – In Class Activity



BFS – In Class Activity



BFS – In Class Activity



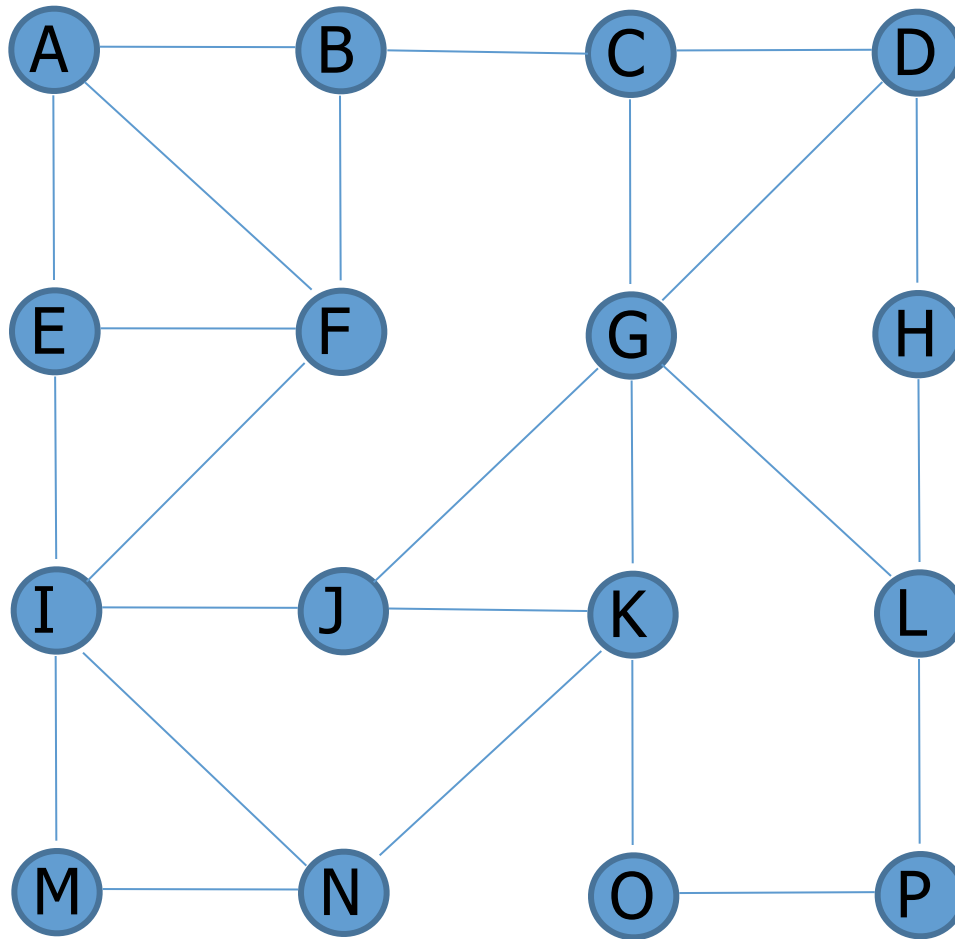
Running Time = $O(|E| + |V|)$

DFS - Depth First Search

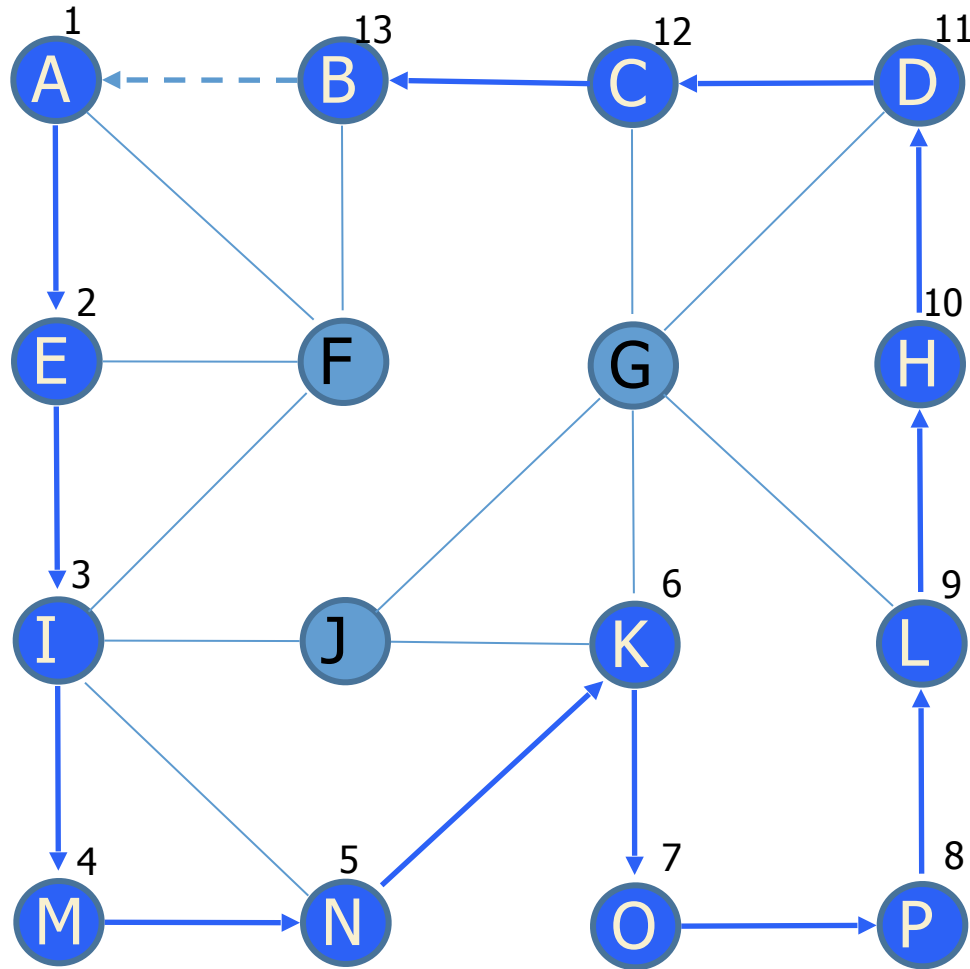
Generalization of a preorder traversal.

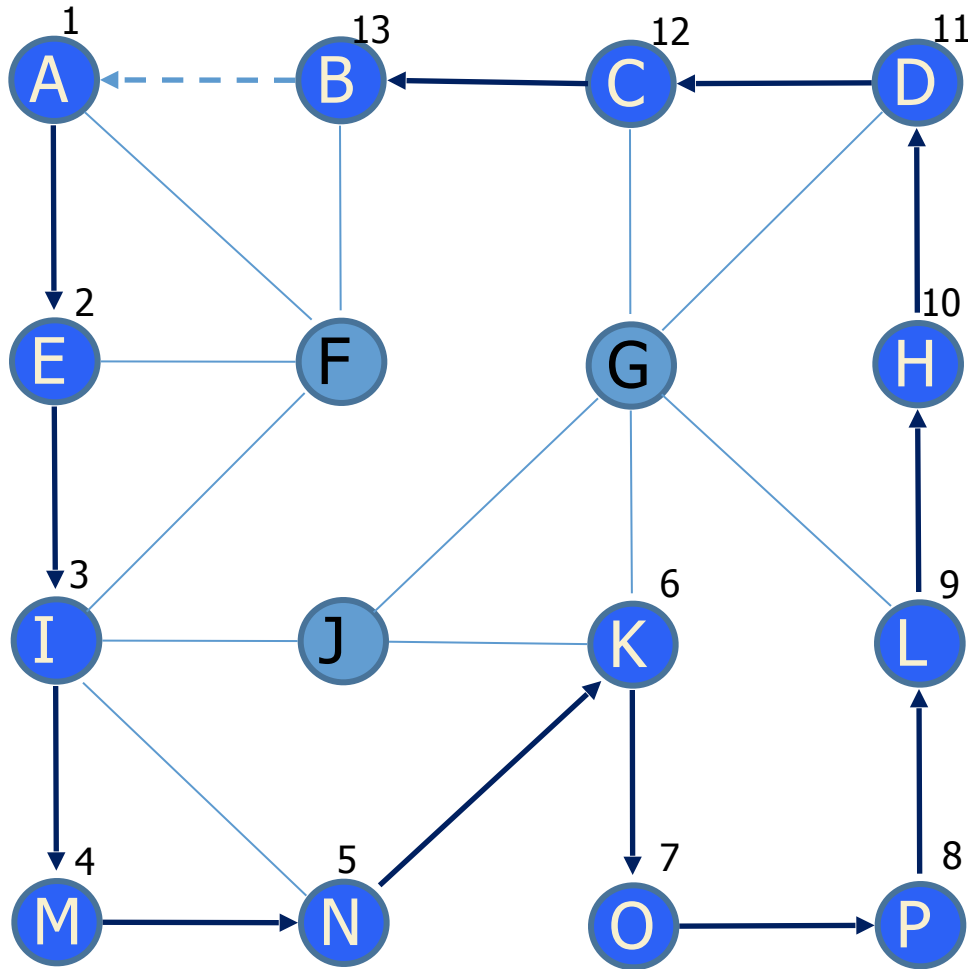
1. Start at an arbitrary vertex s .
2. Traverse (visit) as many descendants of s as possible without visiting a previously visited vertex.
3. Once a previously visited vertex is encountered, back out until a vertex v_i with unexplored edges is found.
4. Depth first search on v_i .
5. Repeat (steps 3-4) until all vertices visited.

DFS - Depth First Search

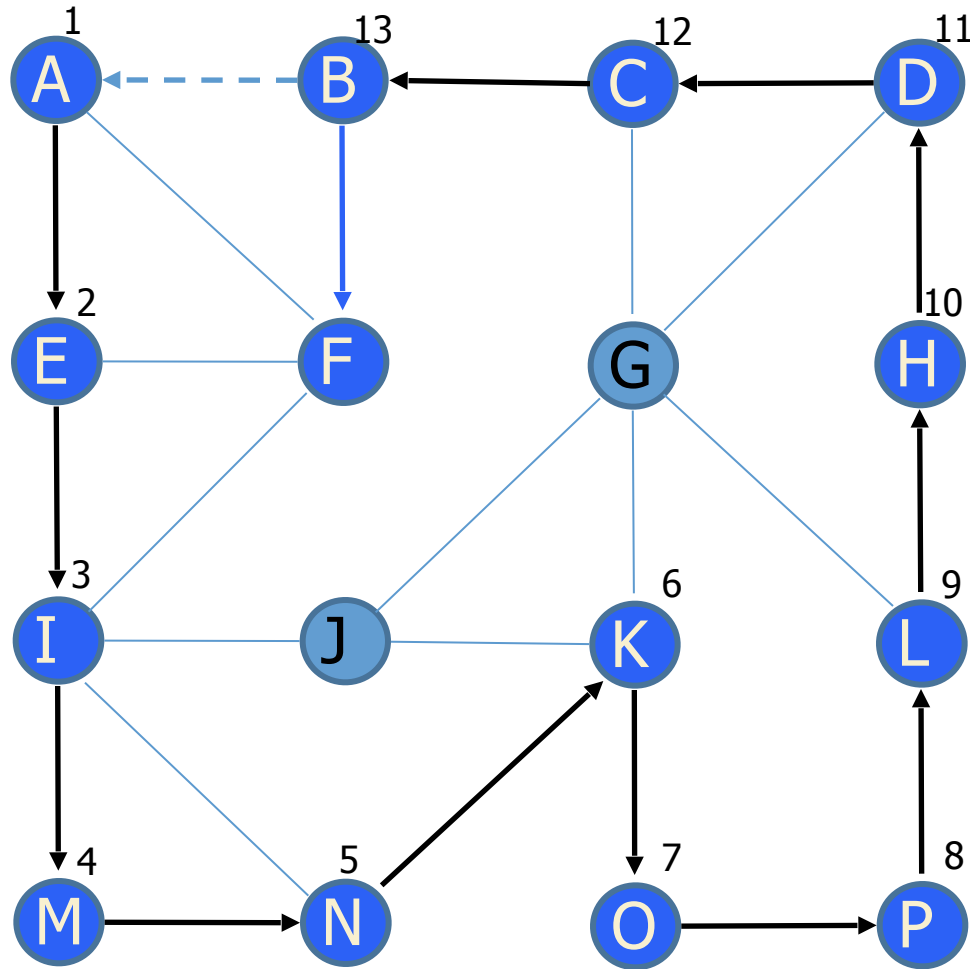


DFS - Depth First Search

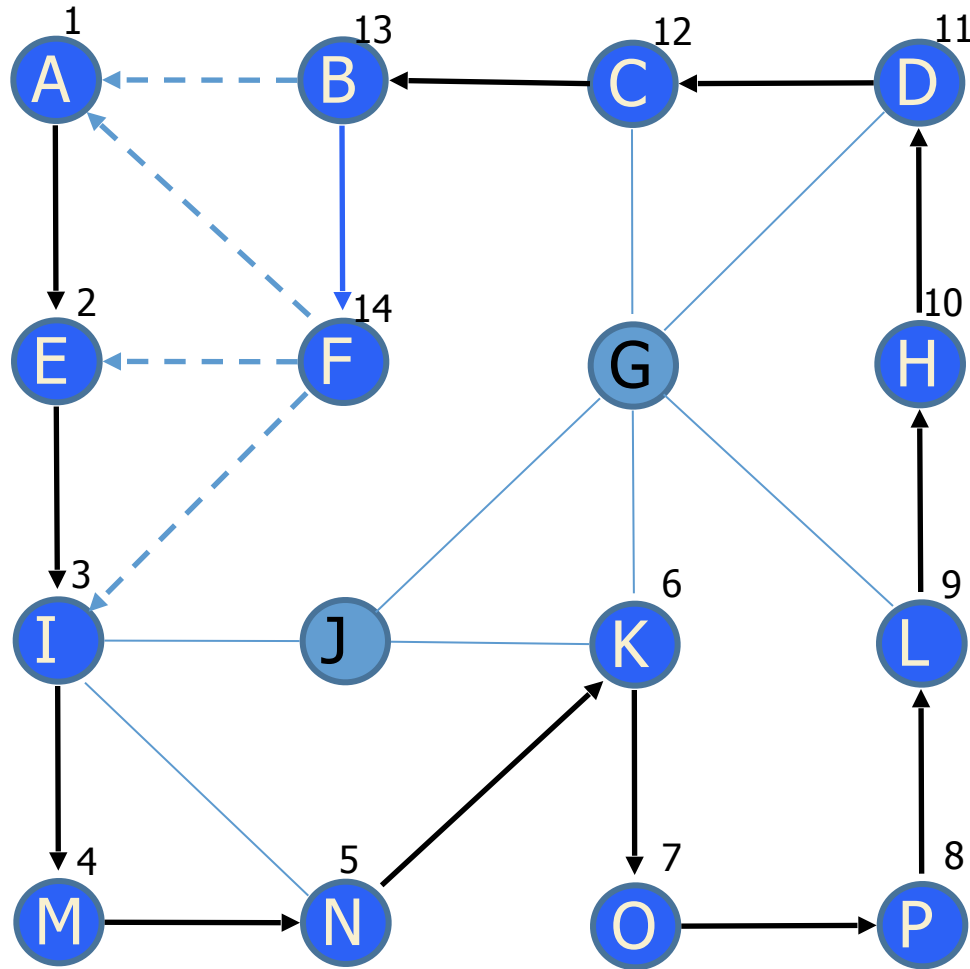




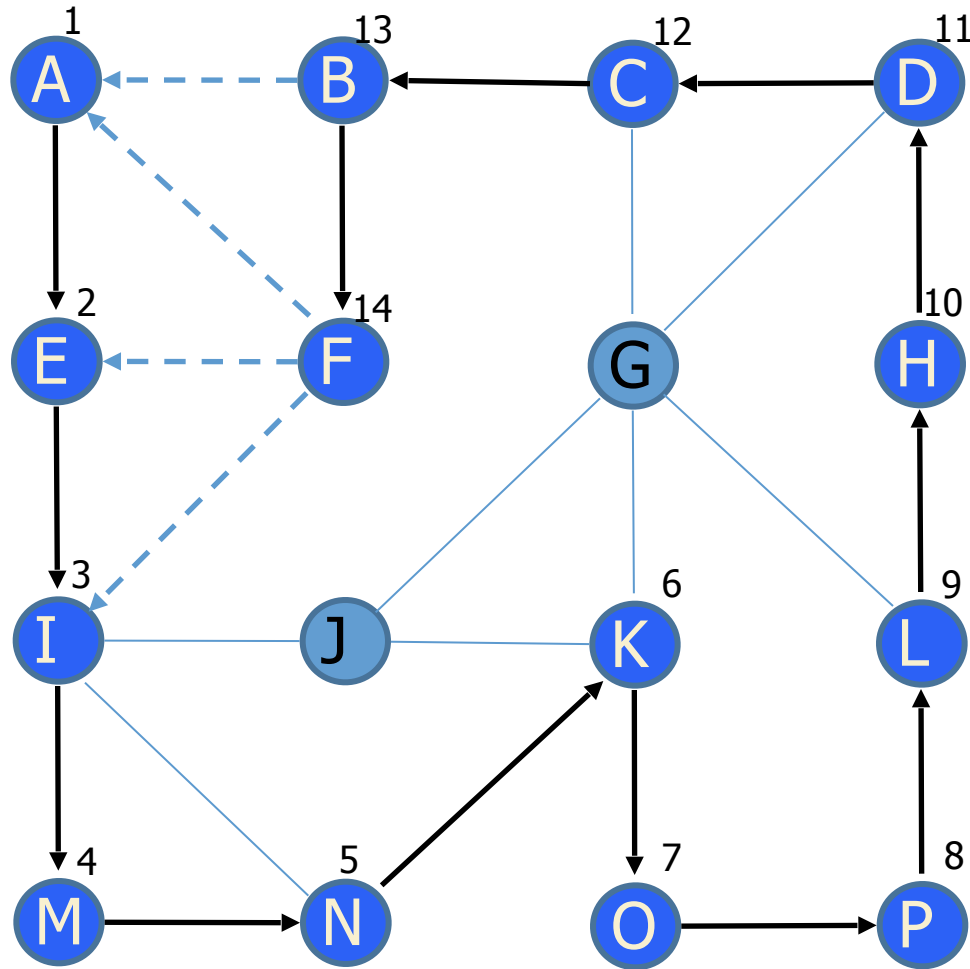
DFS - Depth First Search



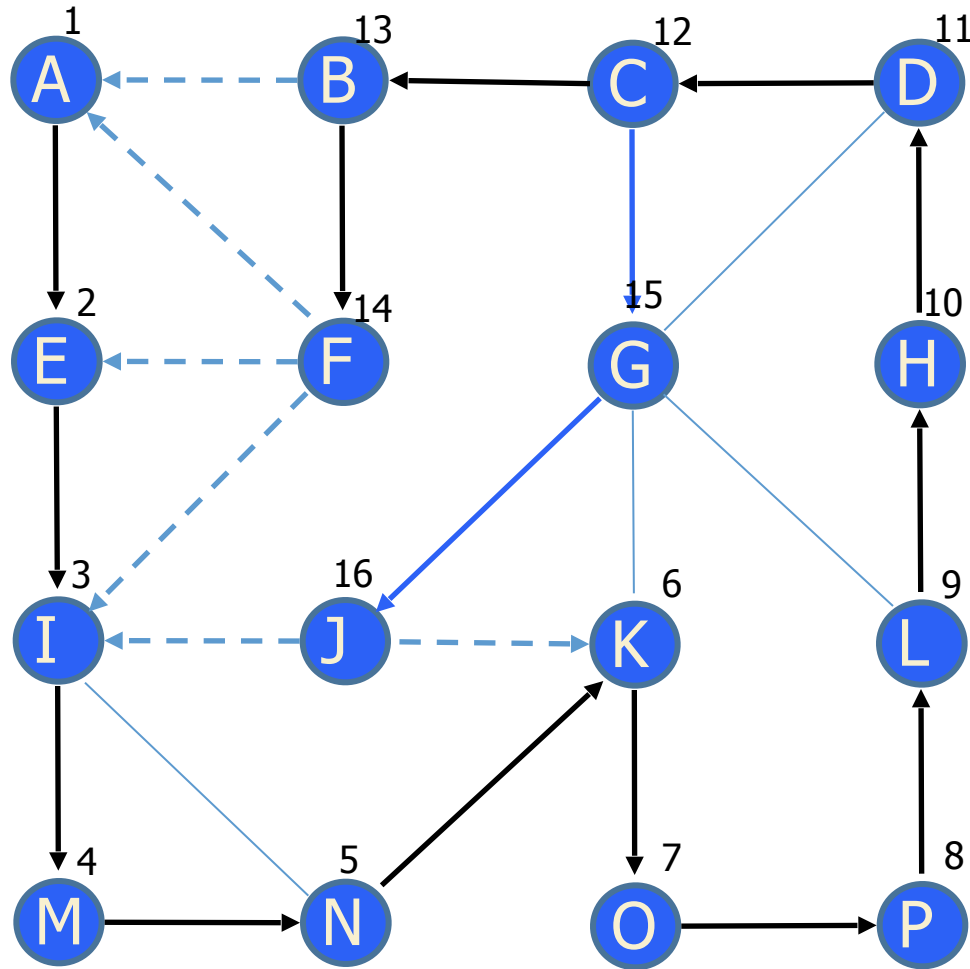
DFS - Depth First Search



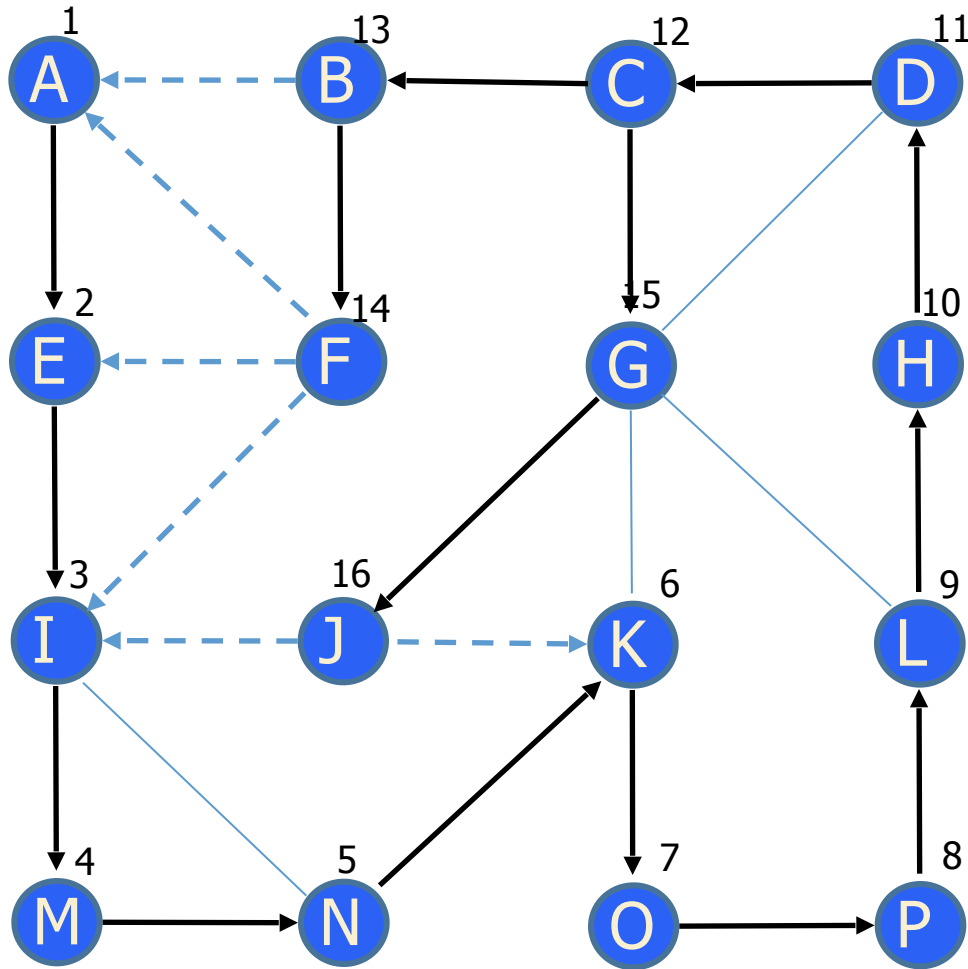
DFS - Depth First Search



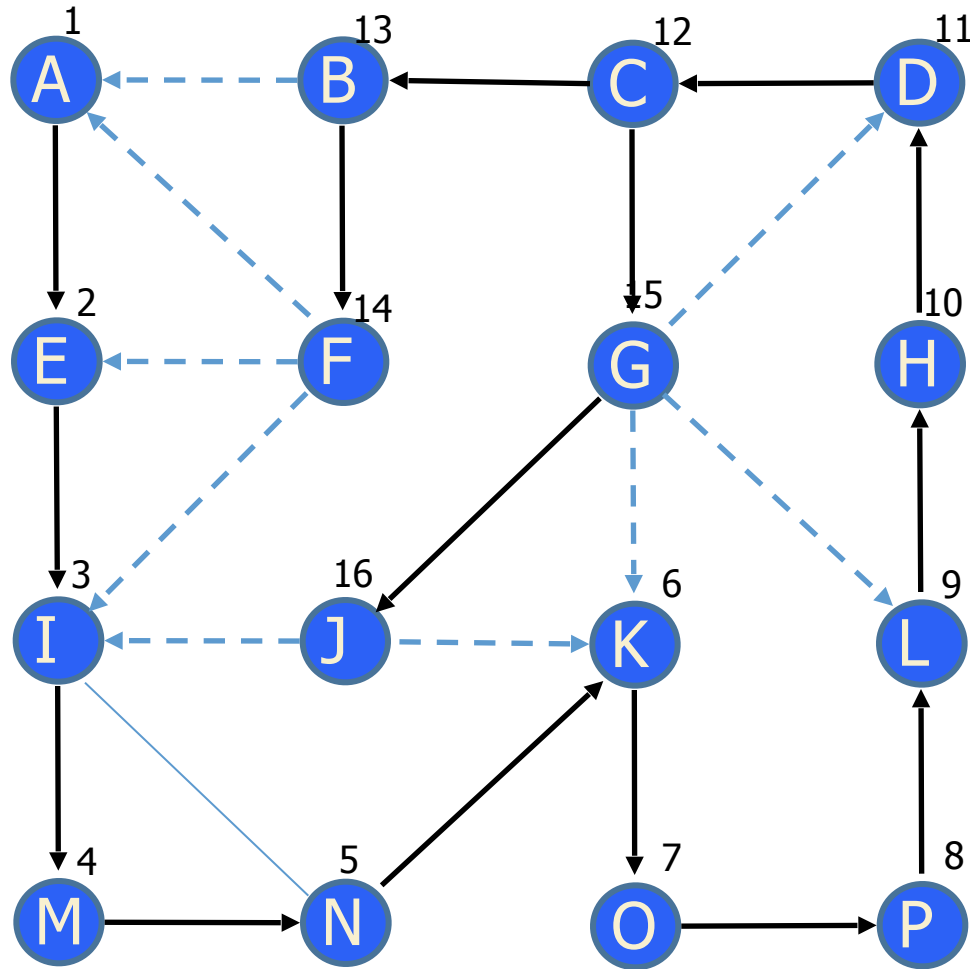
DFS - Depth First Search



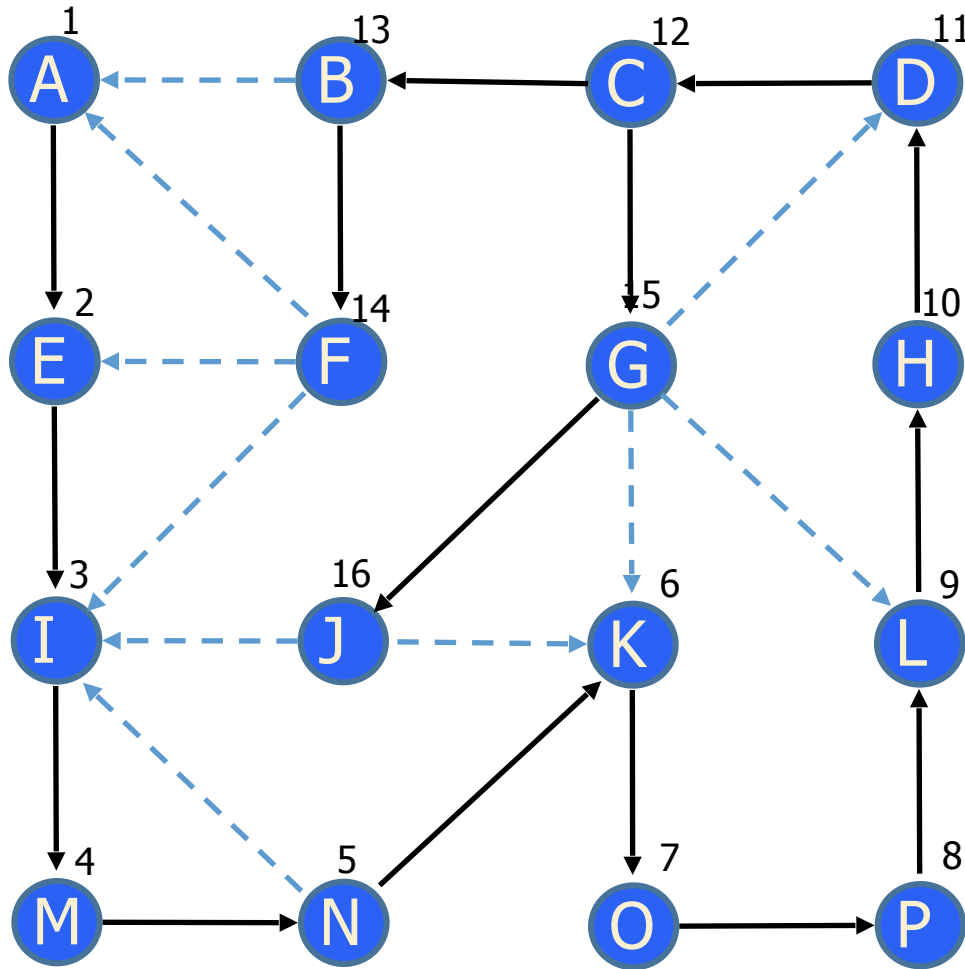
DFS - Depth First Search



DFS - Depth First Search

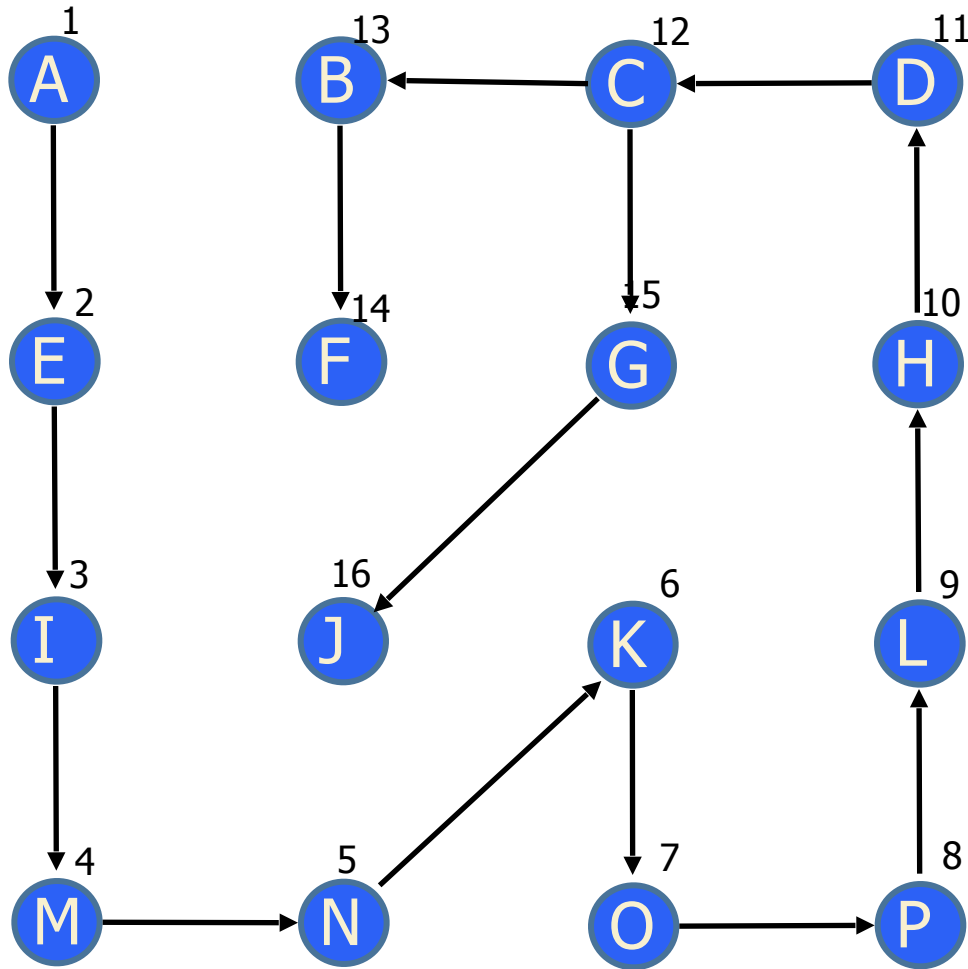


DFS - Depth First Search

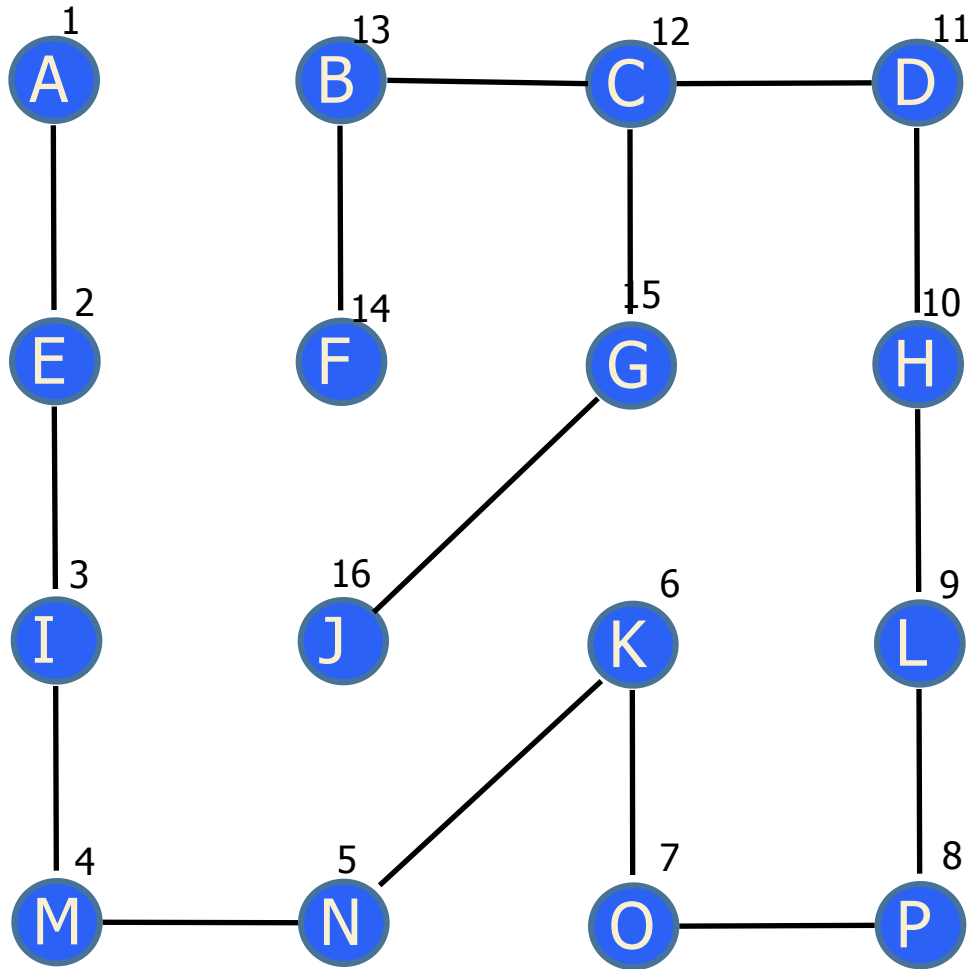


Running Time = $O(|E| + |V|)$

Spanning Tree



Spanning Tree



MST – Minimum Spanning Tree

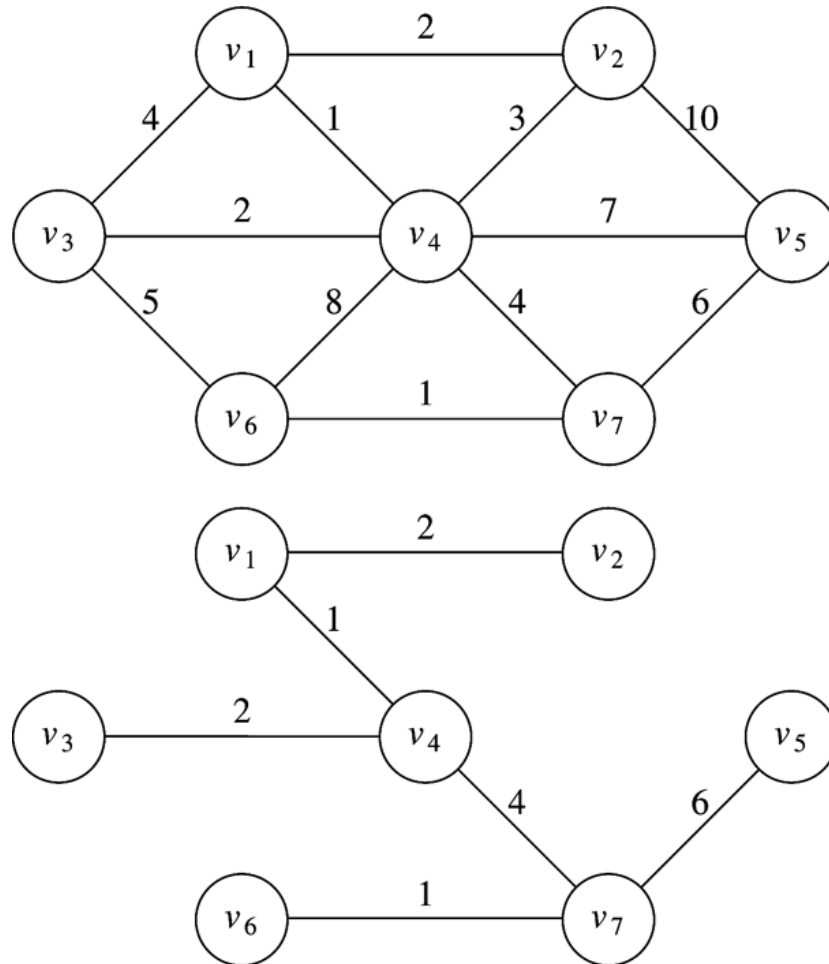
- **Spanning Tree** – Given an undirected, connected graph G , a spanning tree T is a connected acyclic subgraph (tree) that contains all vertices of the graph.
- **Minimum Spanning Tree** – the spanning tree of the smallest weight, where the weight is the sum of the weights on all T 's edges.
- **MST Problem** – Find the minimum spanning tree for a given weighted connected graph.
- Not always a unique MST. There can be more than 1 MST with a different set of edges, perhaps some shared, in 1 or more MSTs.

MST

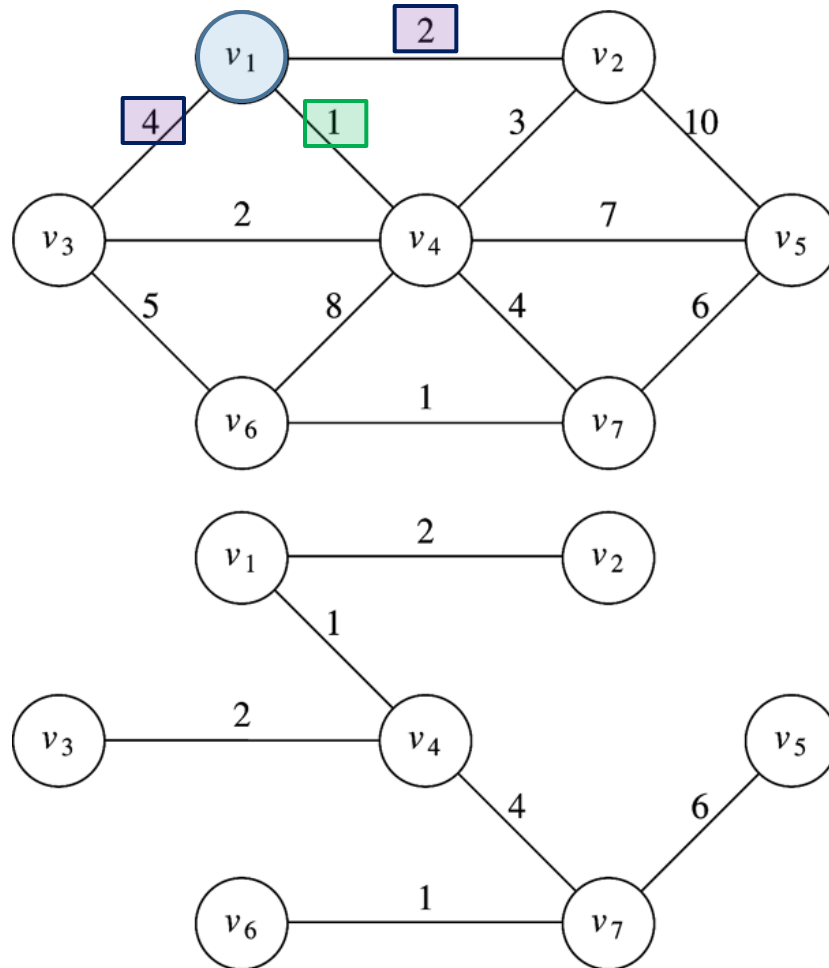
- useful for minimizing the amount of wiring needed for:
 - phone lines
 - cable lines
- used in wireless sensor networks
- used in network routing algorithms



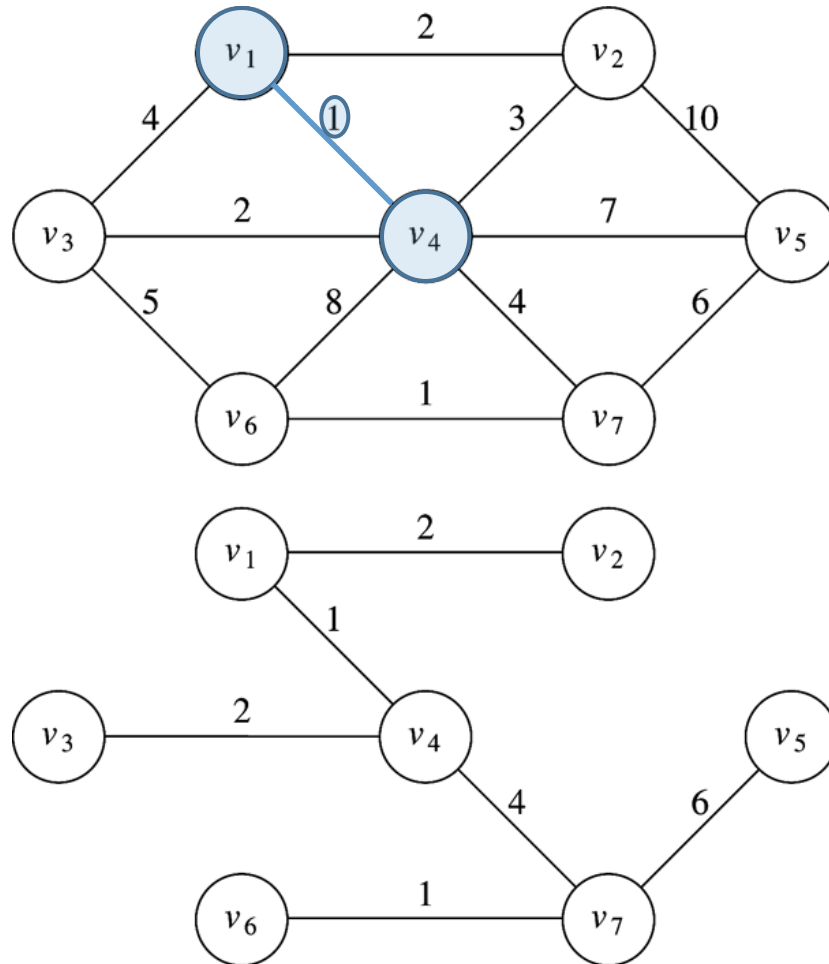
Prim's Algorithm



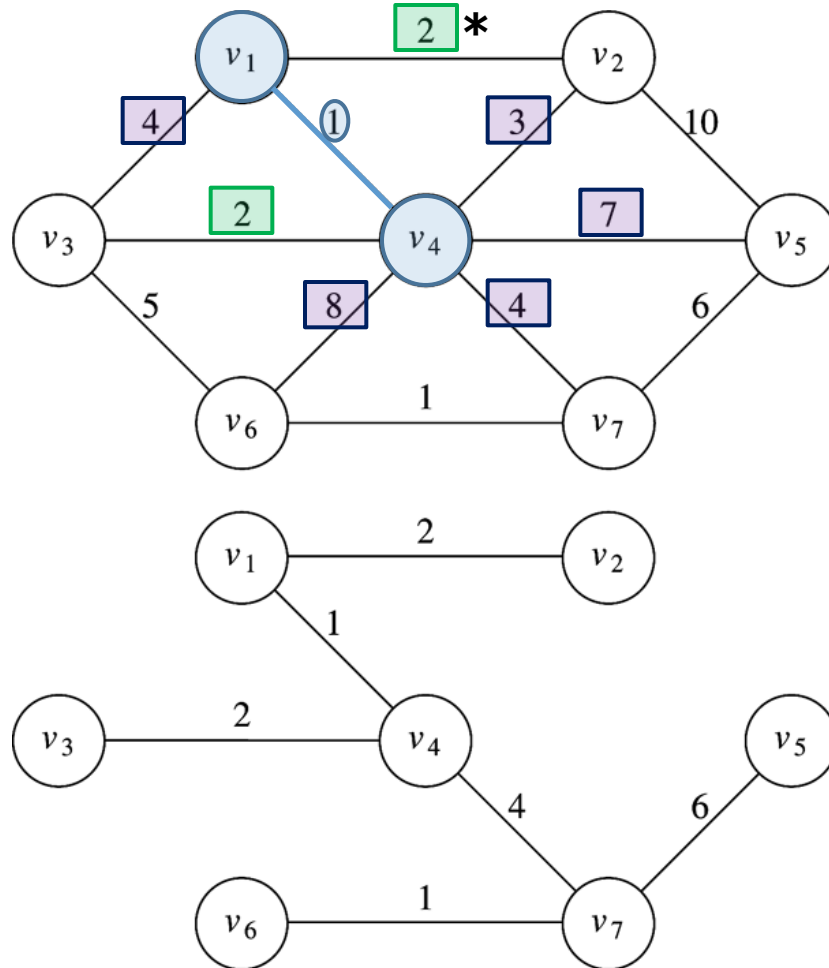
Prim's Algorithm



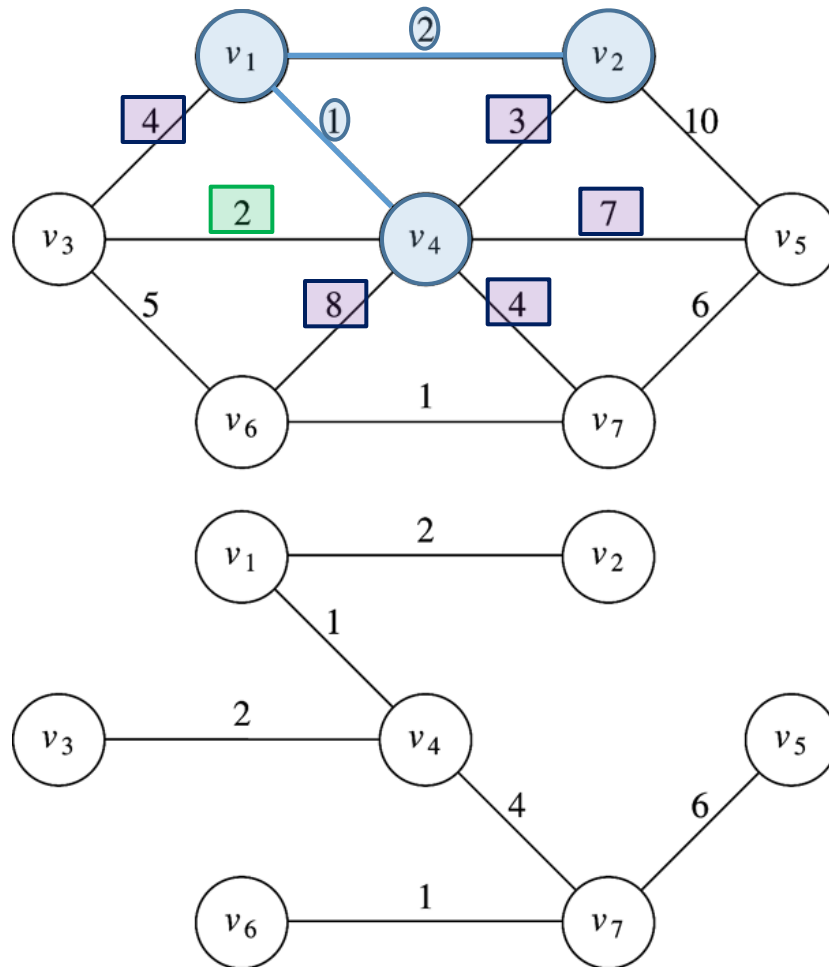
Prim's Algorithm



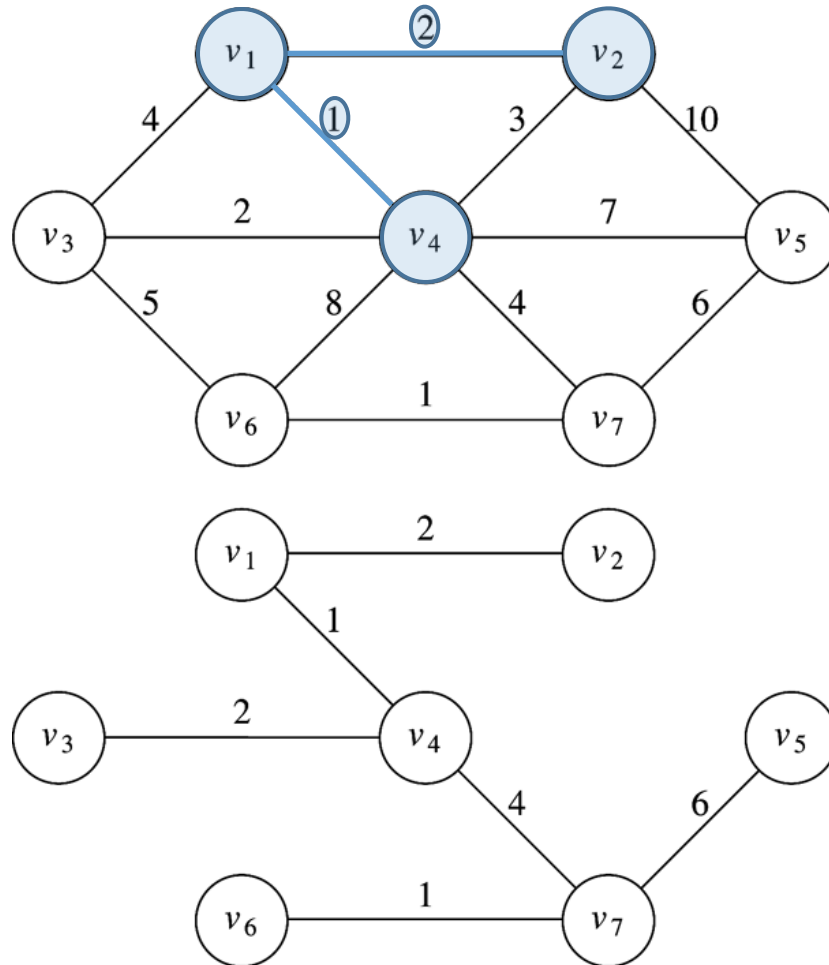
Prim's Algorithm



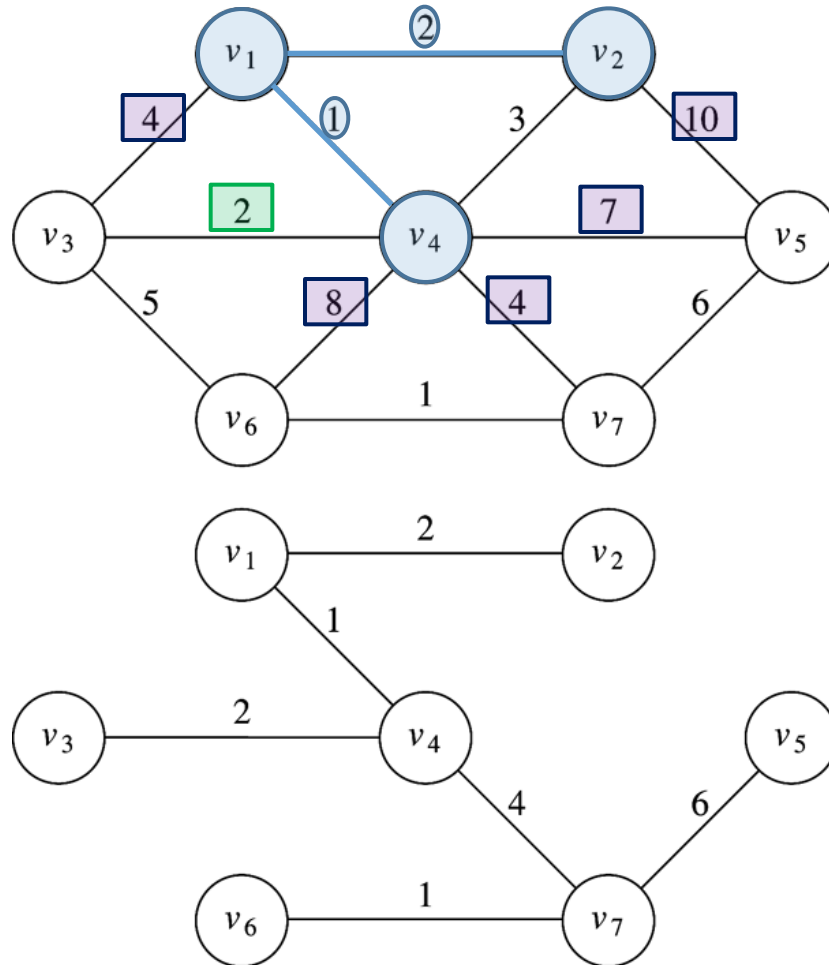
Prim's Algorithm



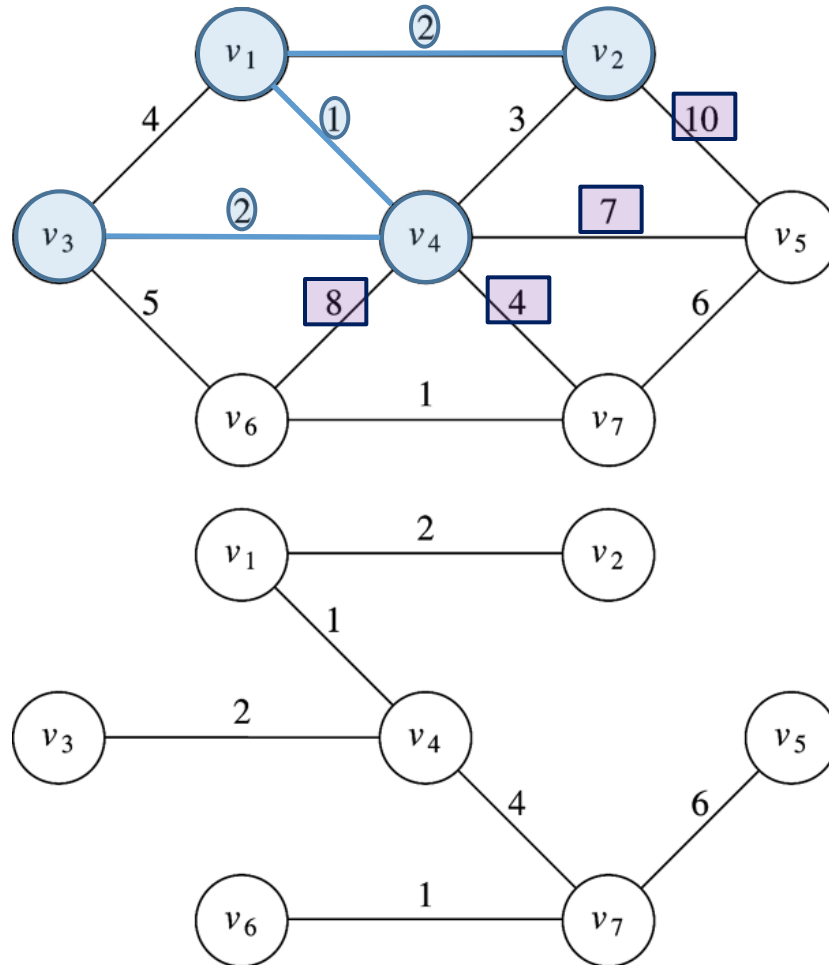
Prim's Algorithm



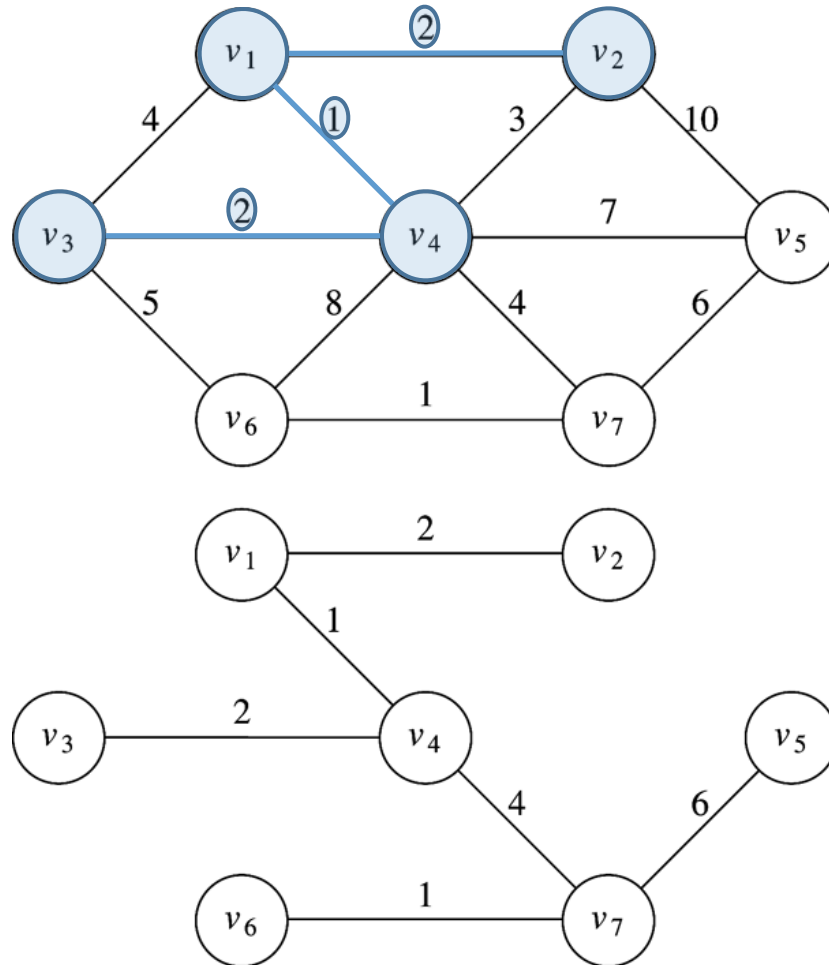
Prim's Algorithm



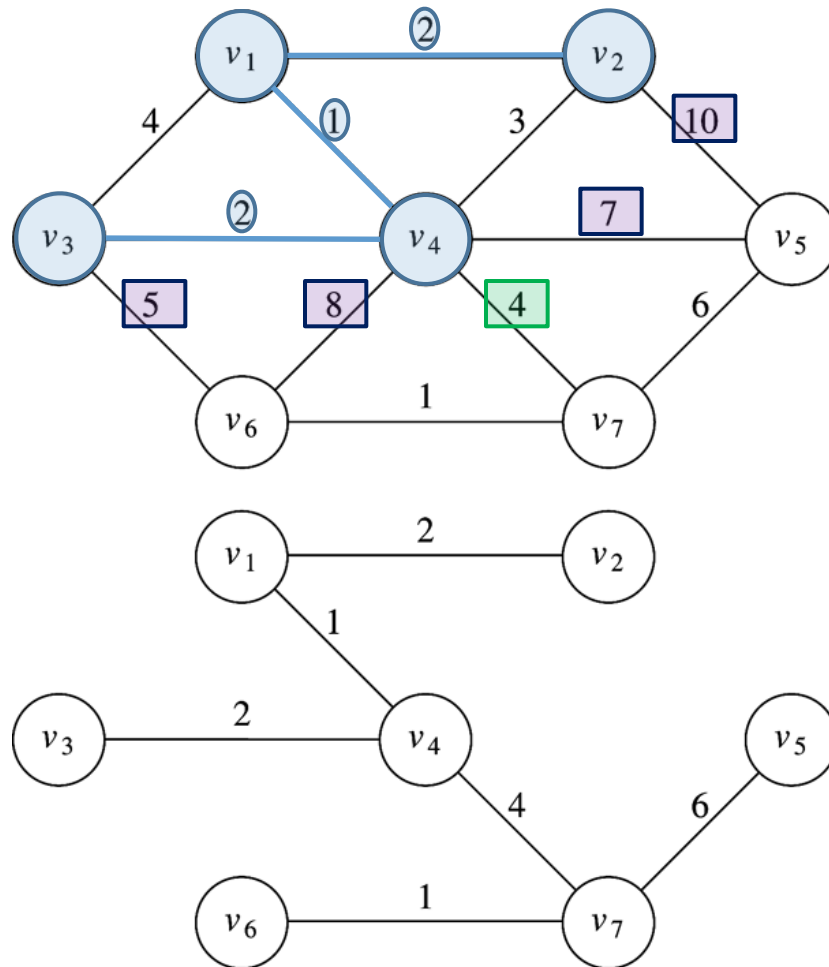
Prim's Algorithm



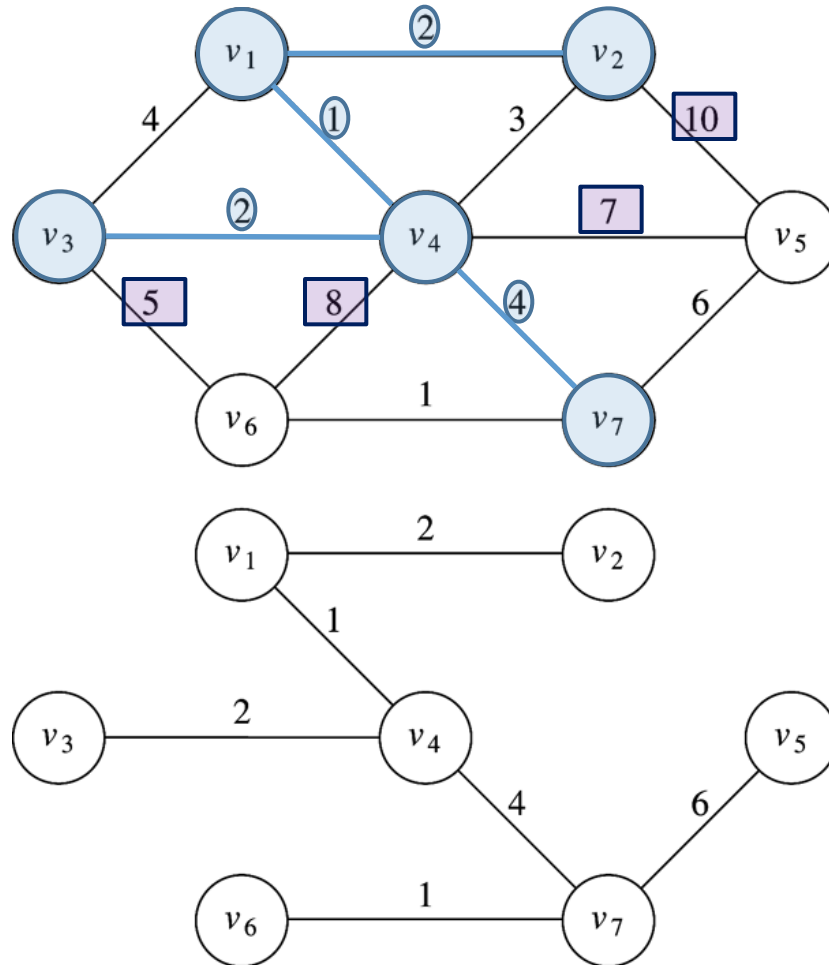
Prim's Algorithm



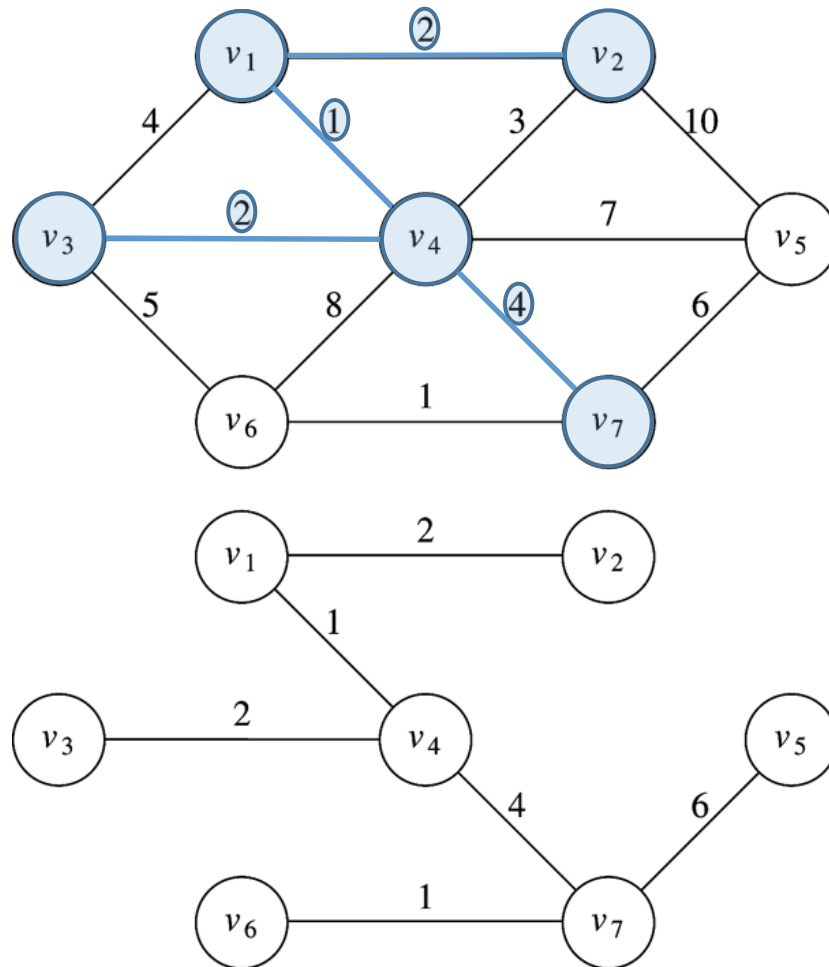
Prim's Algorithm



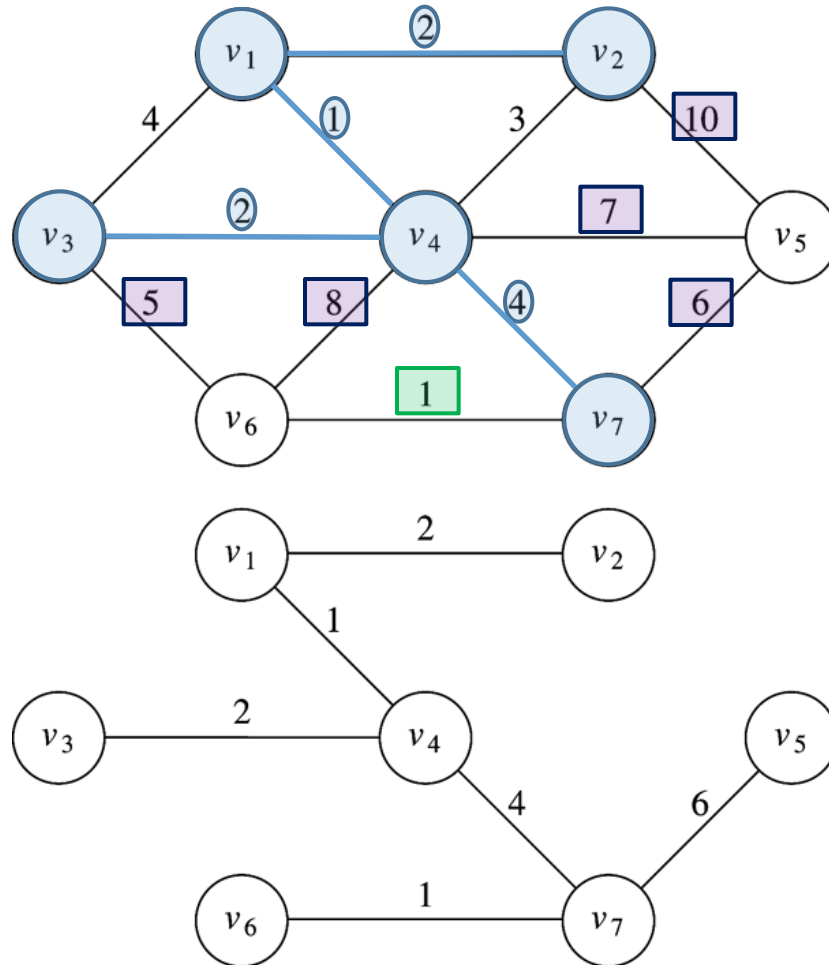
Prim's Algorithm



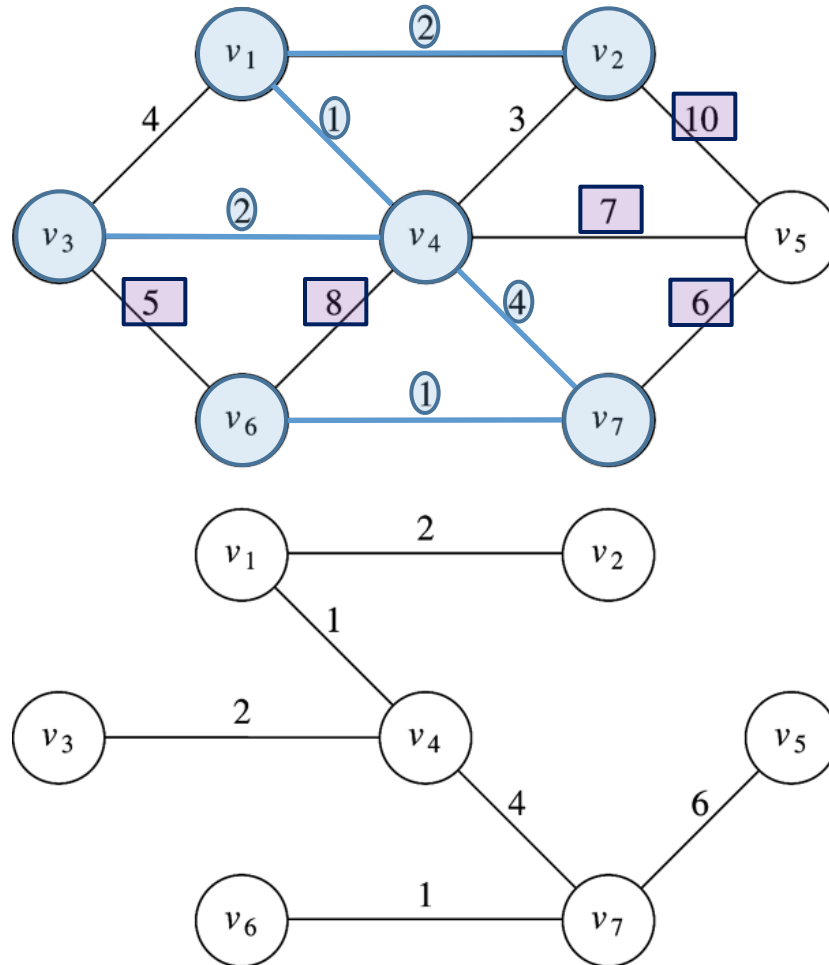
Prim's Algorithm



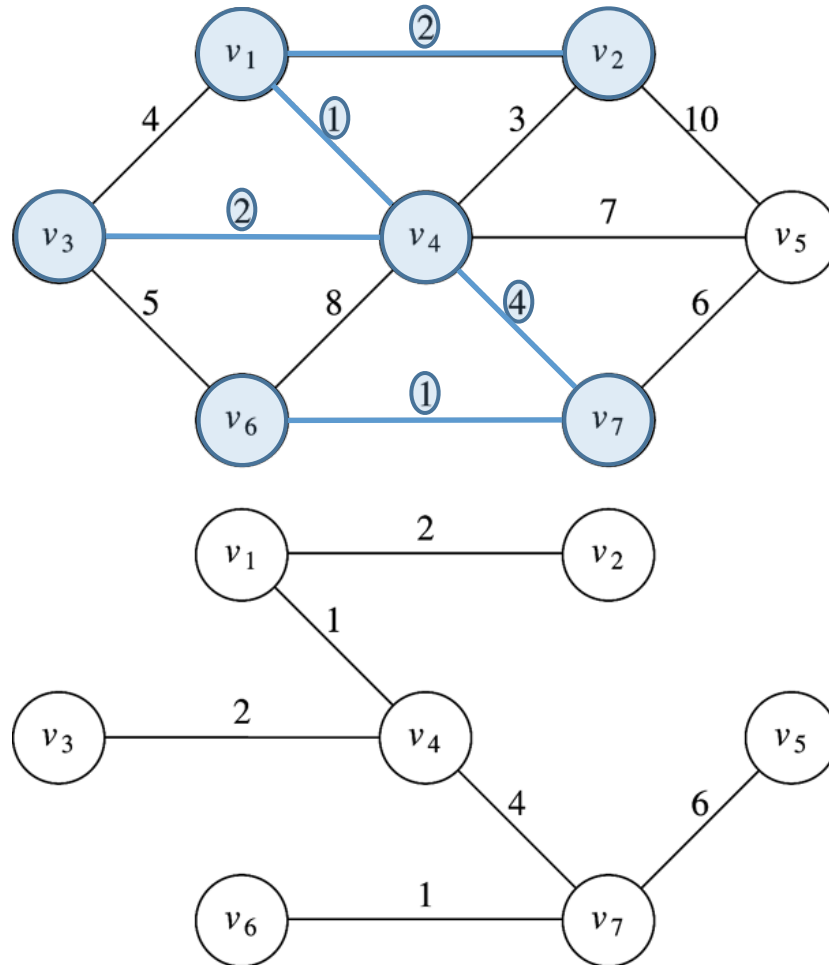
Prim's Algorithm



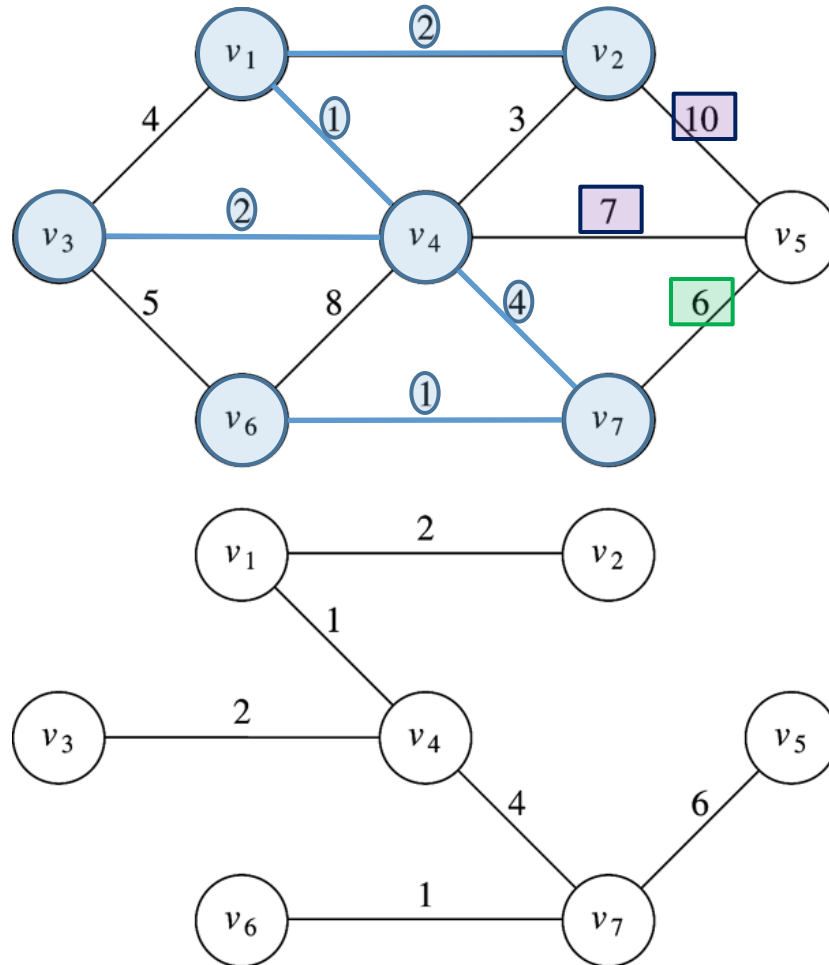
Prim's Algorithm



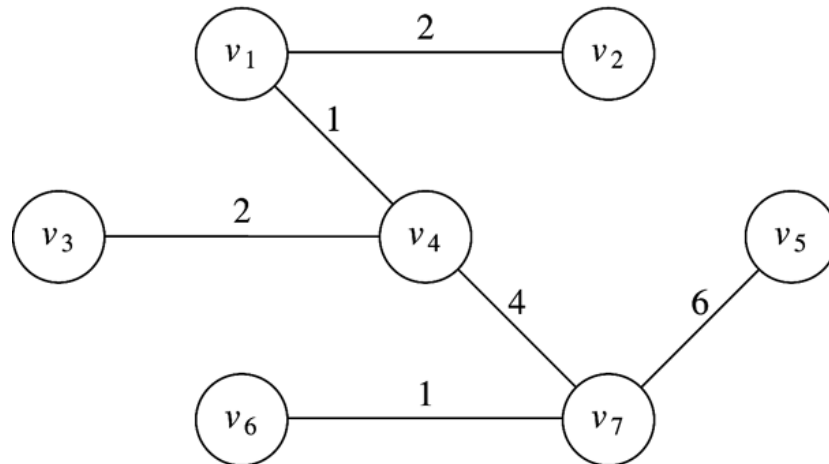
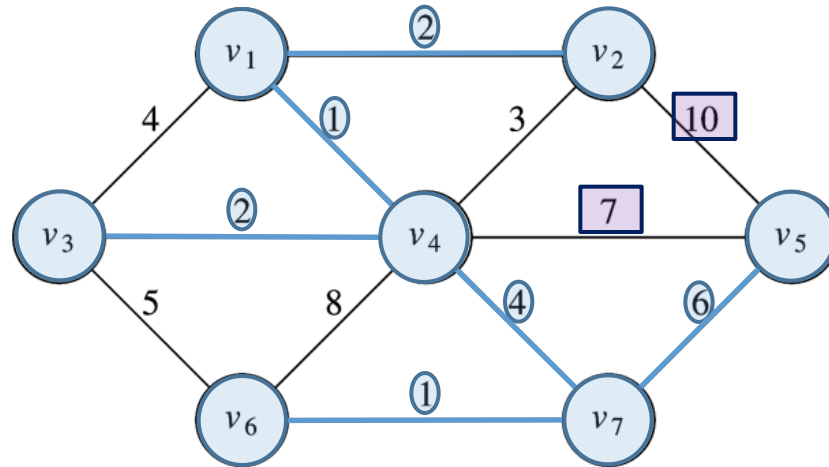
Prim's Algorithm



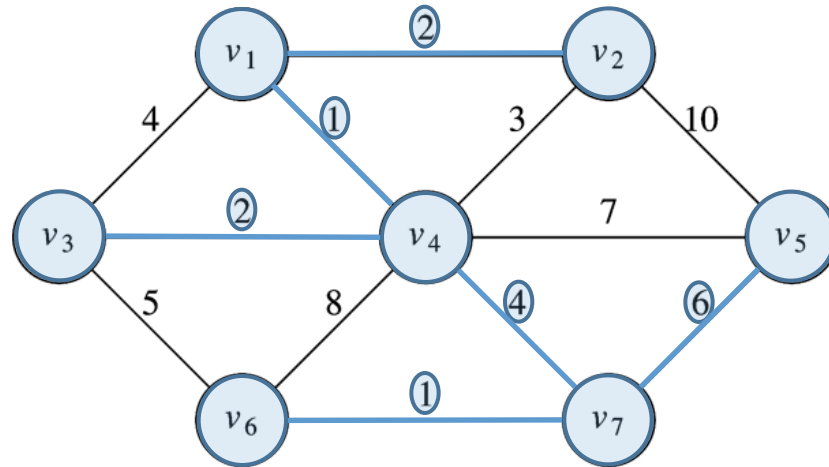
Prim's Algorithm



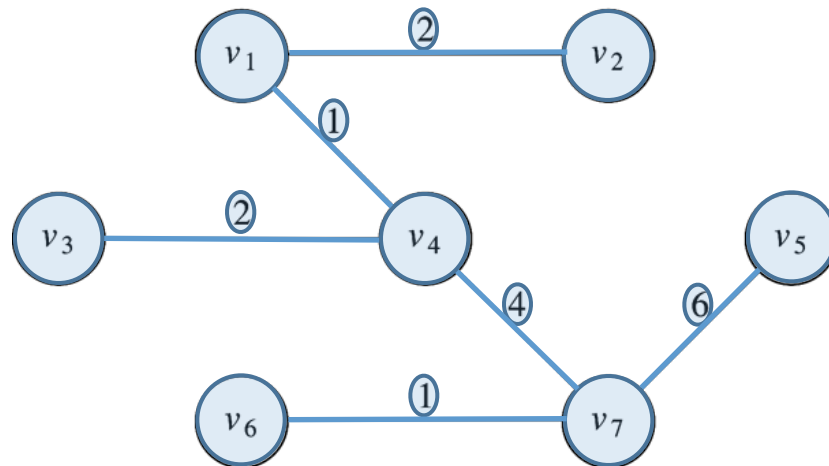
Prim's Algorithm



Prim's Algorithm



- $O(|V|^2)$
Linear scan to find min
- $O(|E|\log|V|)$
Binary heaps



Directed Graph - Weighted

