

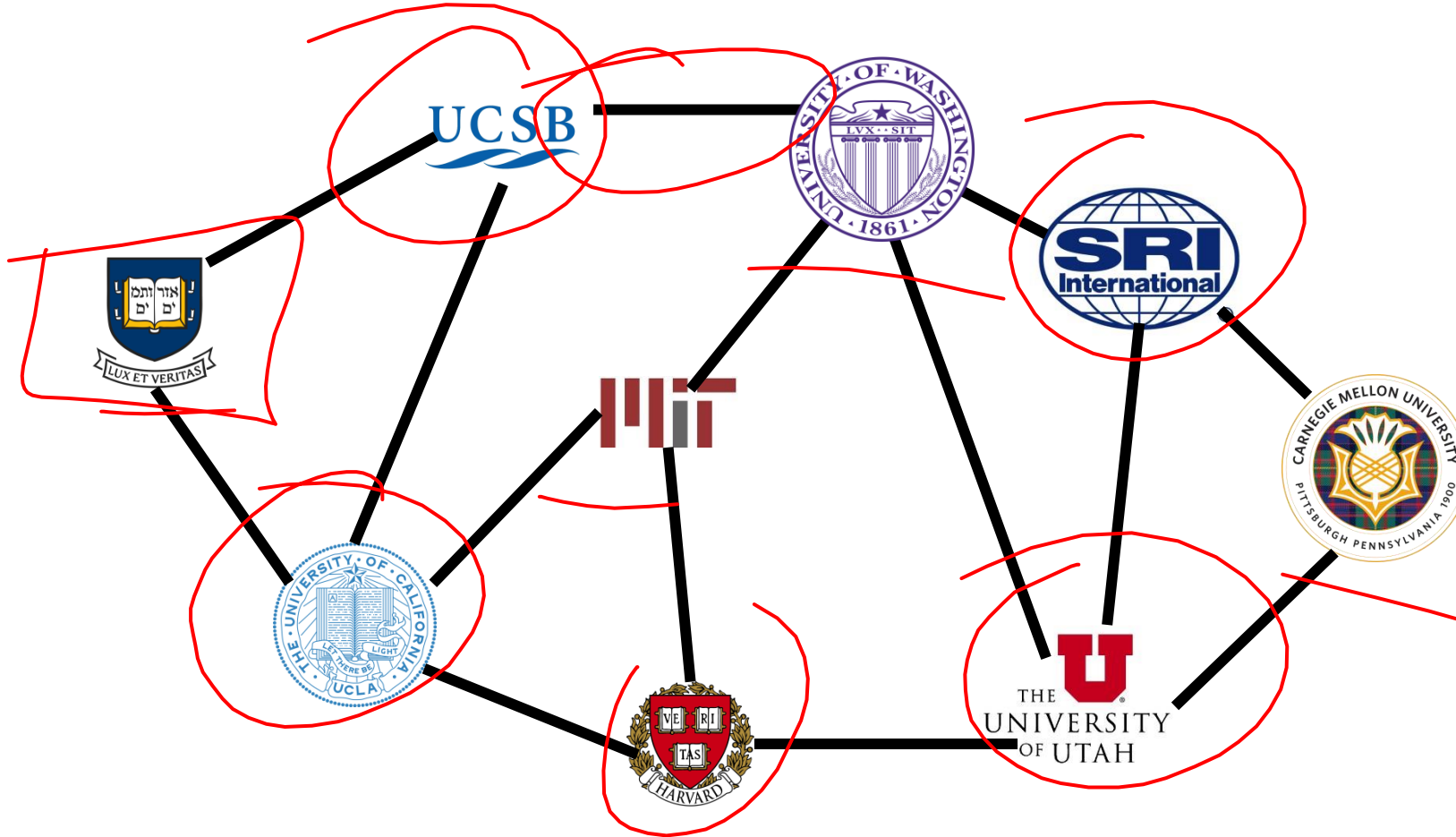
CSE 332 Summer 2024

Lecture 14: Graphs

Nathan Brunelle

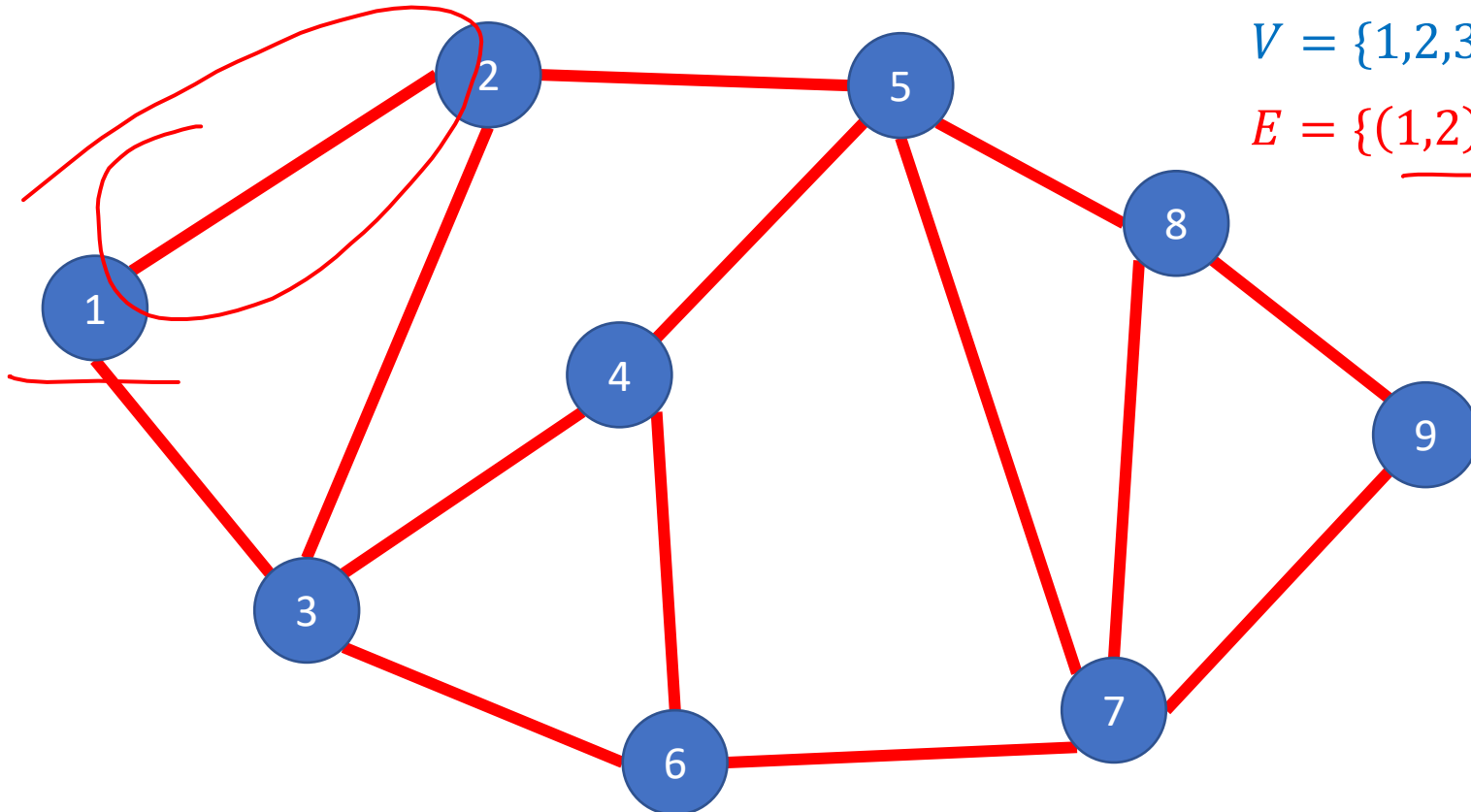
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ARPANET



Undirected Graphs

Definition: $G = (V, E)$
Vertices/Nodes
Edges

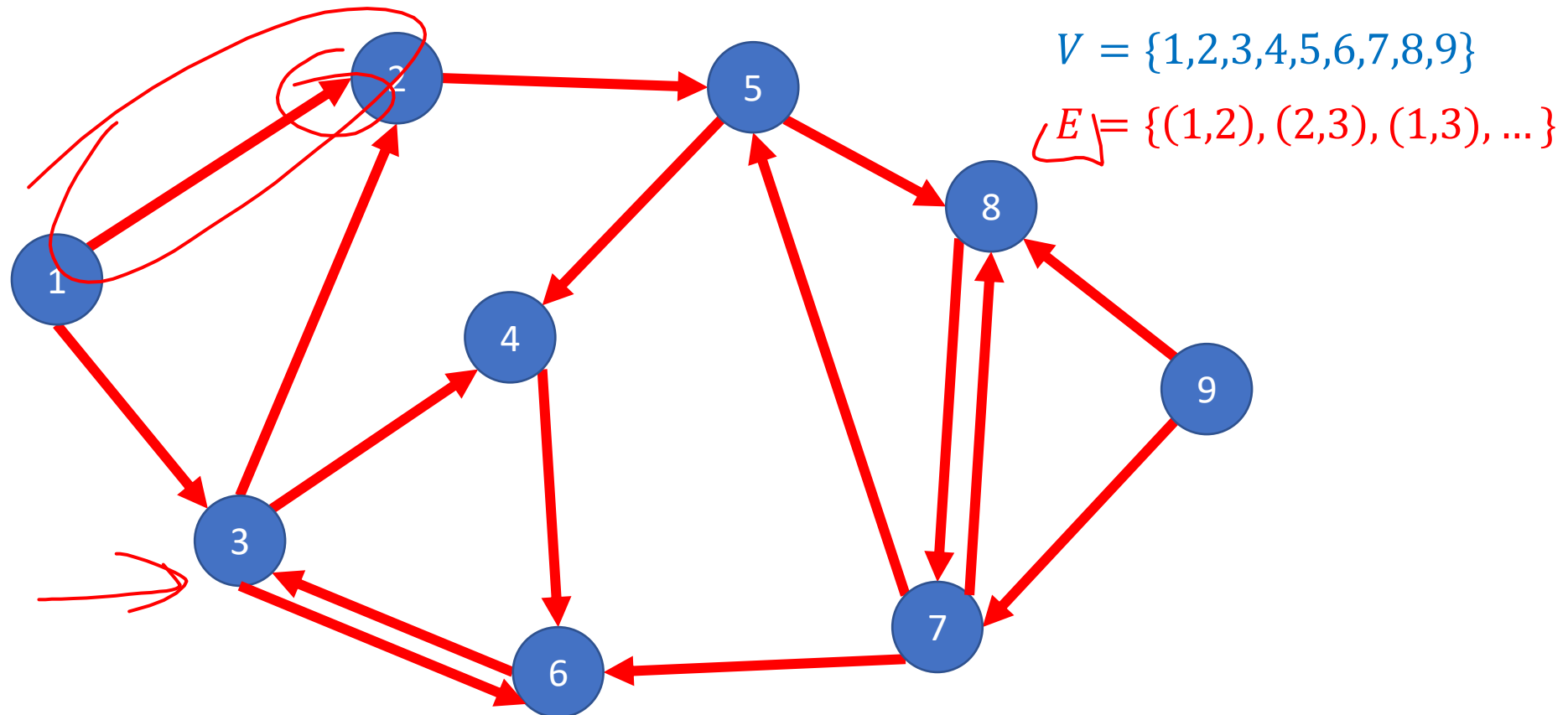


$V = \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$

$E = \{(1, 2), (2, 3), (1, 3), \dots\}$

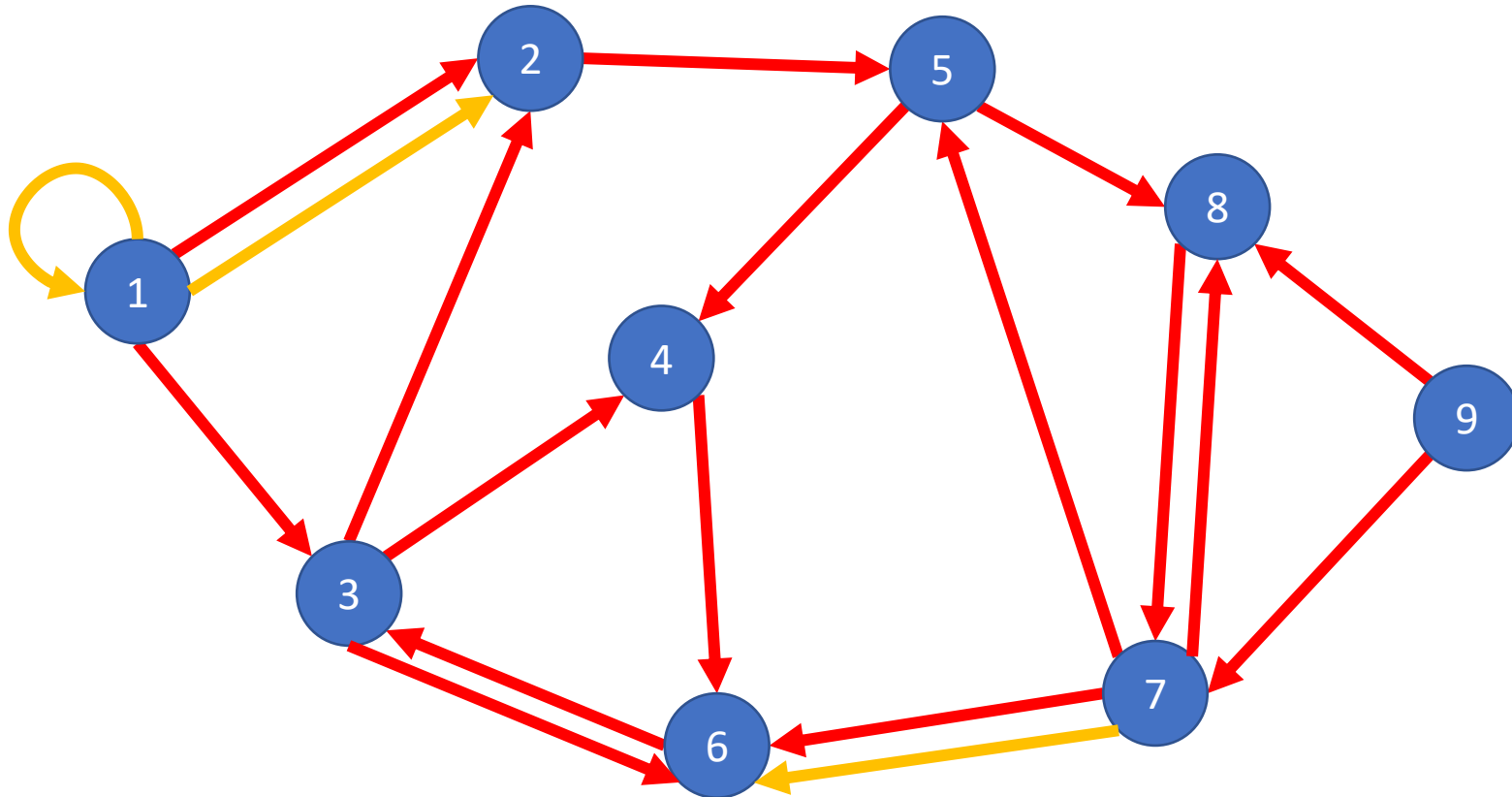
Directed Graphs

Definition: $G = (V, E)$
Vertices/Nodes
Edges



Self-Edges and Duplicate Edges

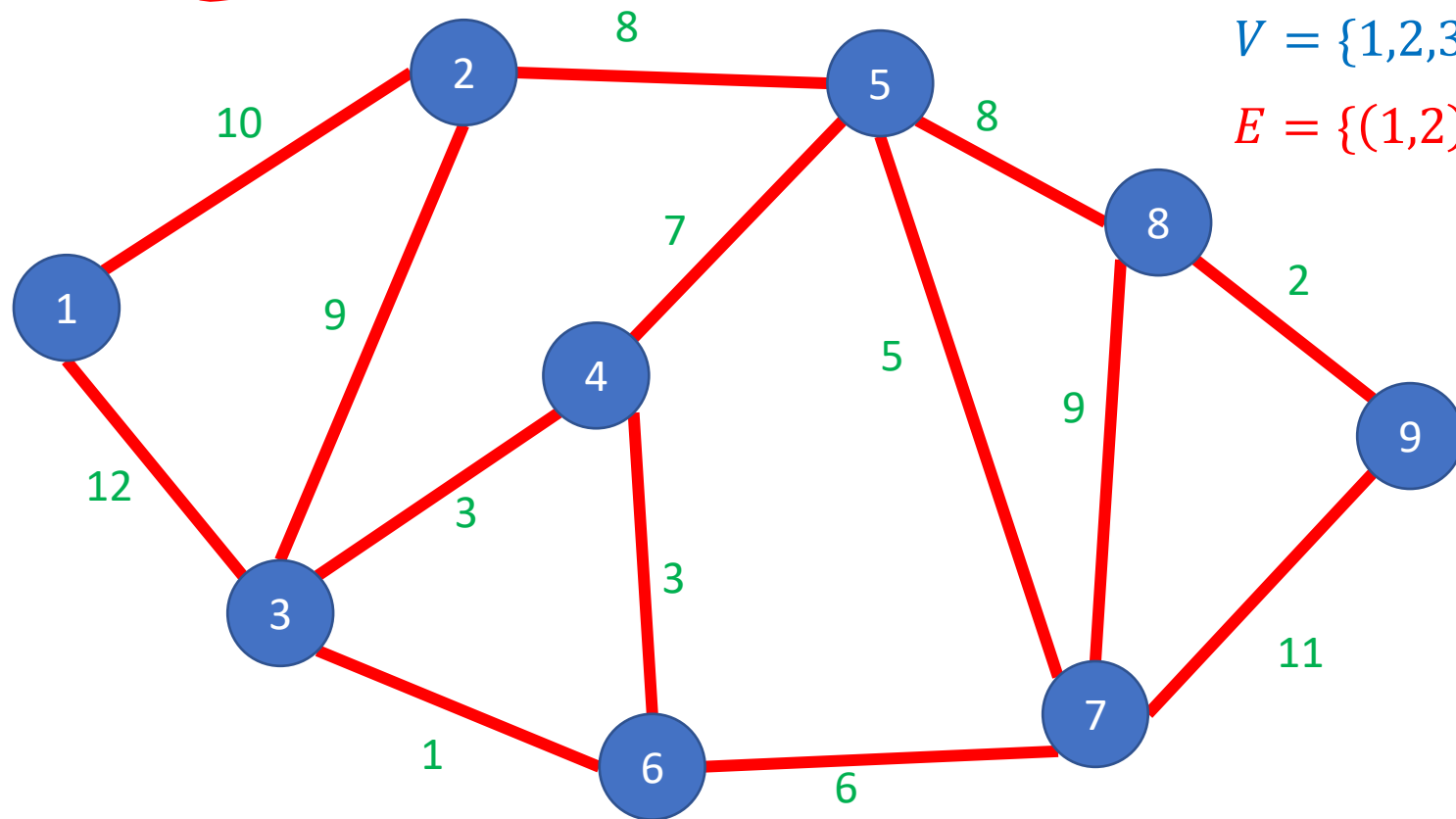
Some graphs may have duplicate edges (e.g. here we have the edge (1,2) twice).
Some may also have self-edges (e.g. here there is an edge from 1 to 1).
Graph with Neither self-edges nor duplicate edges are called simple graphs



Weighted Graphs

Definition: $G = (V, E)$
Vertices/Nodes
Edges

$w(e)$ = weight of edge e



$V = \{1,2,3,4,5,6,7,8,9\}$

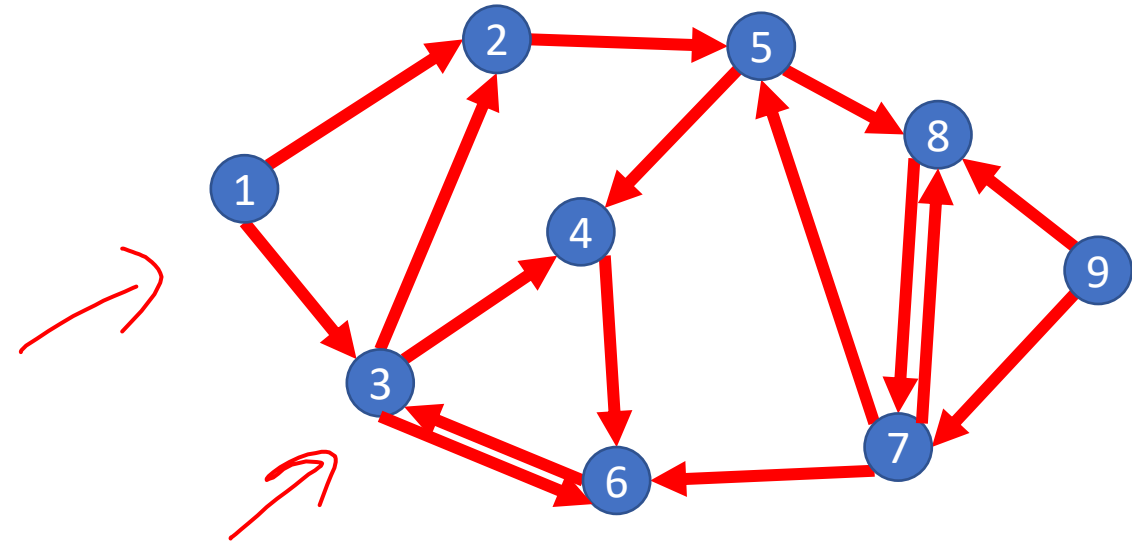
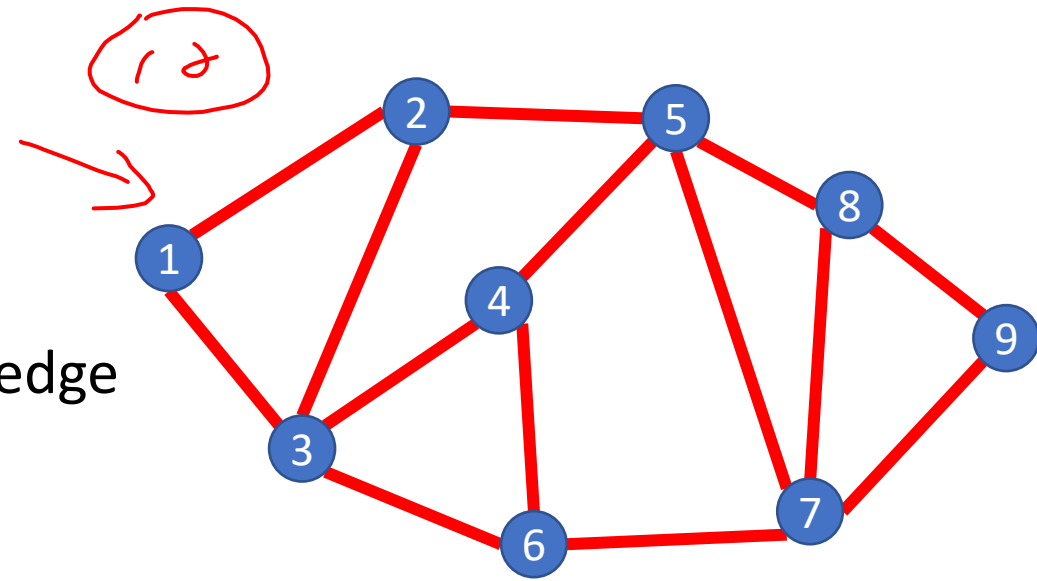
$E = \{(1,2), (2,3), (1,3), \dots\}$

Graph Applications

- For each application below, consider:
 - What are the nodes, what are the edges?
 - Is the graph directed?
 - Is the graph simple?
 - Is the graph weighted?
- Facebook friends
 - Nodes: Users
 - Undirected: friendship is required to be mutual
 - Simple
 - Depends
- Twitter/X followers
 - Nodes: Users
 - Directed
 - Simple
 - Depends (same as above)
- Java inheritance
 - Nodes: classes
 - Directed
 - Simple
 - unweighted
- Airline Routes
 - Nodes: airports
 - Directed
 - Non-simple
 - Weighted

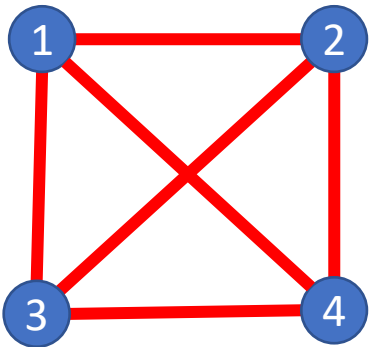
Some Graph Terms

- **Adjacent/Neighbors**
 - Nodes are adjacent/neighbors if they share an edge
- **Degree**
 - Number of edges “touching” a vertex
- **Indegree**
 - Number of incoming edges
- **Outdegree**
 - Number of outgoing edges

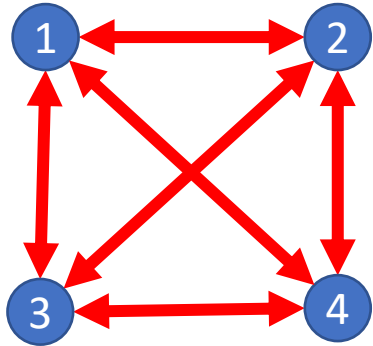


Definition: Complete Graph

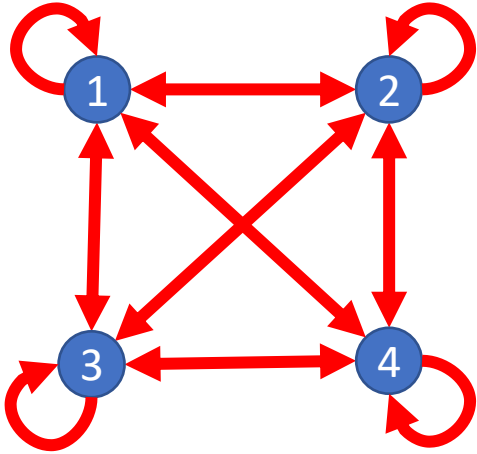
A Graph $G = (V, E)$ s.t. for any pair of nodes $v_1, v_2 \in V$ there is an edge from v_1 to v_2



Complete Undirected Graph



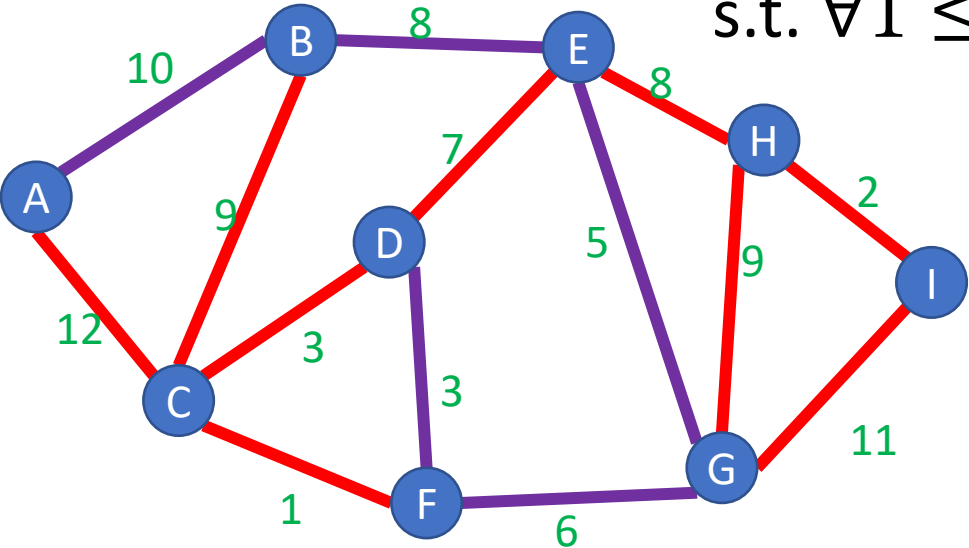
Complete Directed Graph



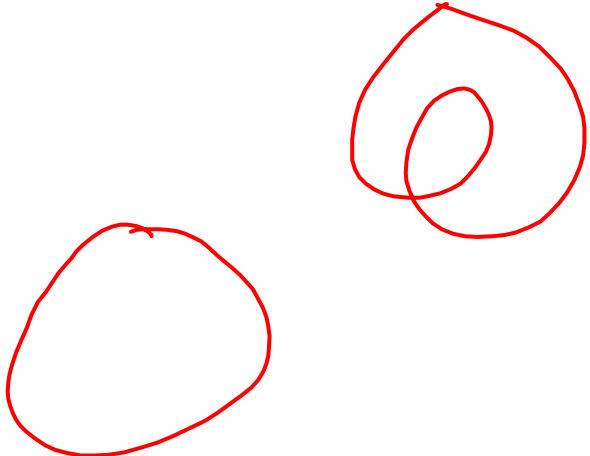
Complete Directed Non-simple Graph

Definition: Path

A sequence of nodes (v_1, v_2, \dots, v_k)
s.t. $\forall 1 \leq i \leq k - 1, (v_i, v_{i+1}) \in E$



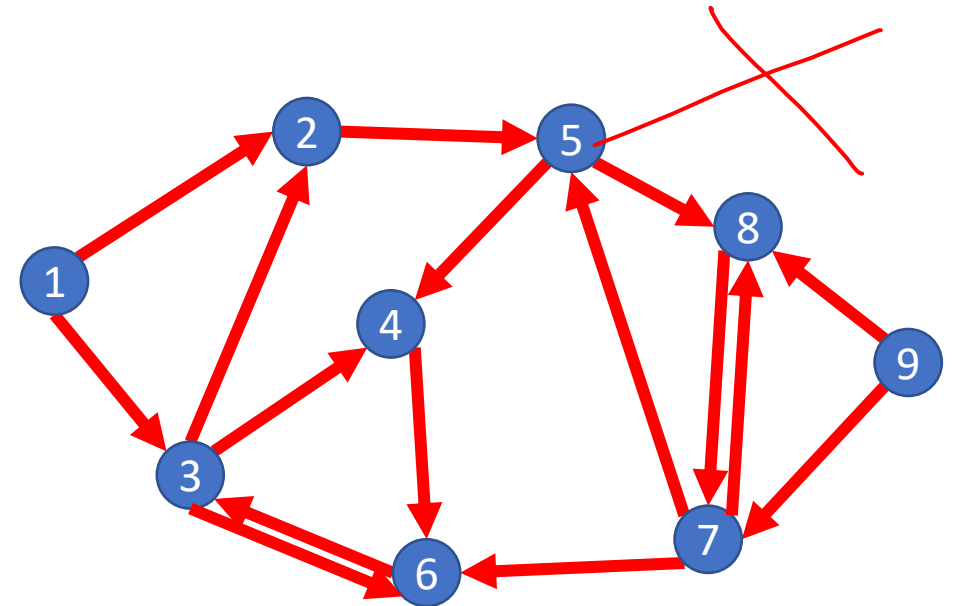
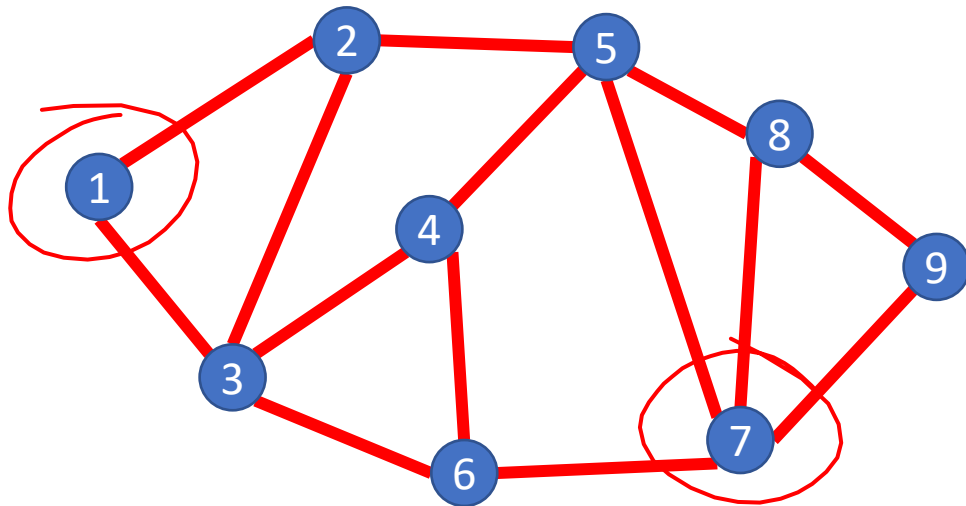
Simple Path:
A path in which each node appears at most once



Cycle:
A path which starts and ends in the same place

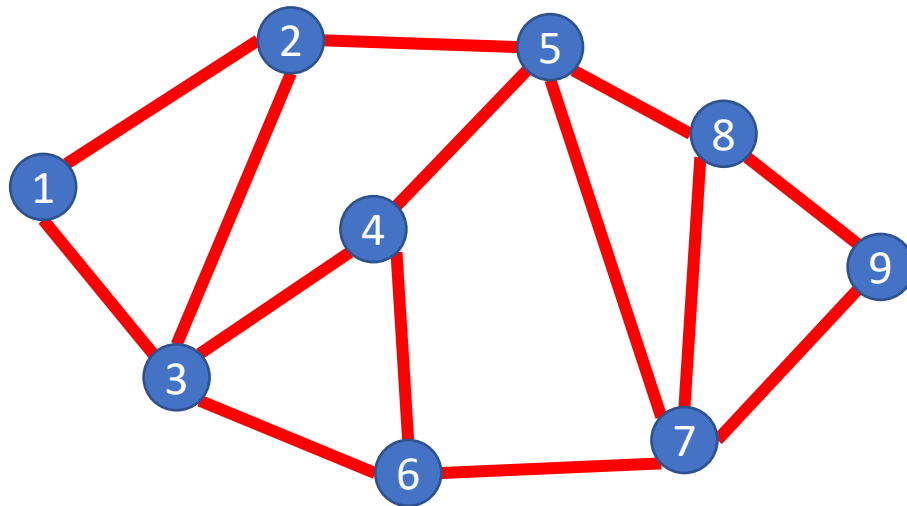
Definition: (Strongly) Connected Graph

A Graph $G = (V, E)$ s.t. for any pair of nodes $v_1, v_2 \in V$ there is a path from v_1 to v_2

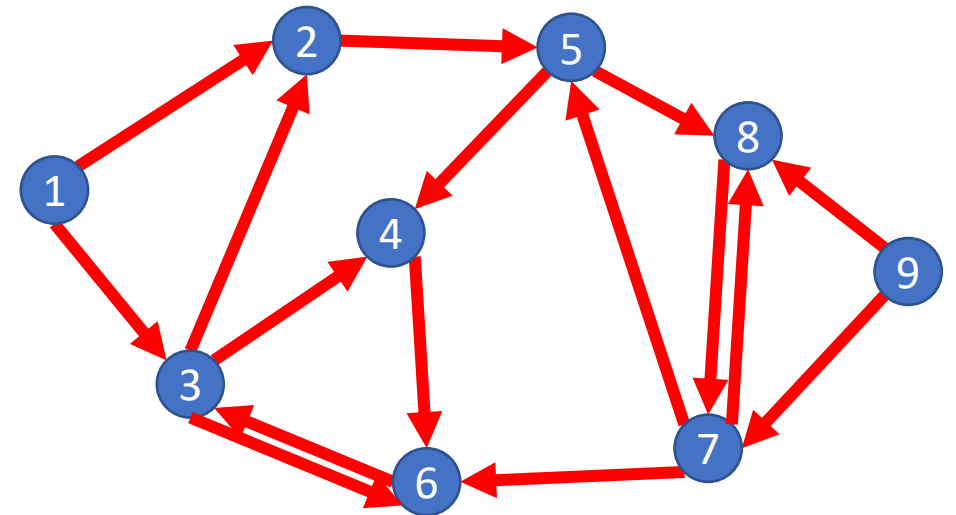


Definition: (Strongly) Connected Graph

A Graph $G = (V, E)$ s.t. for any pair of nodes $v_1, v_2 \in V$ there is a path from v_1 to v_2



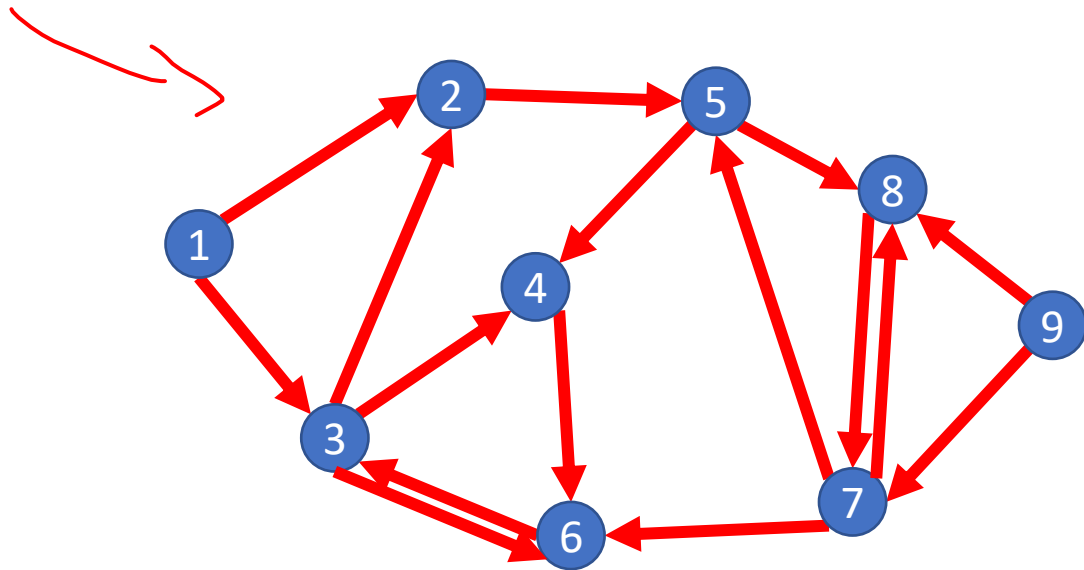
Connected



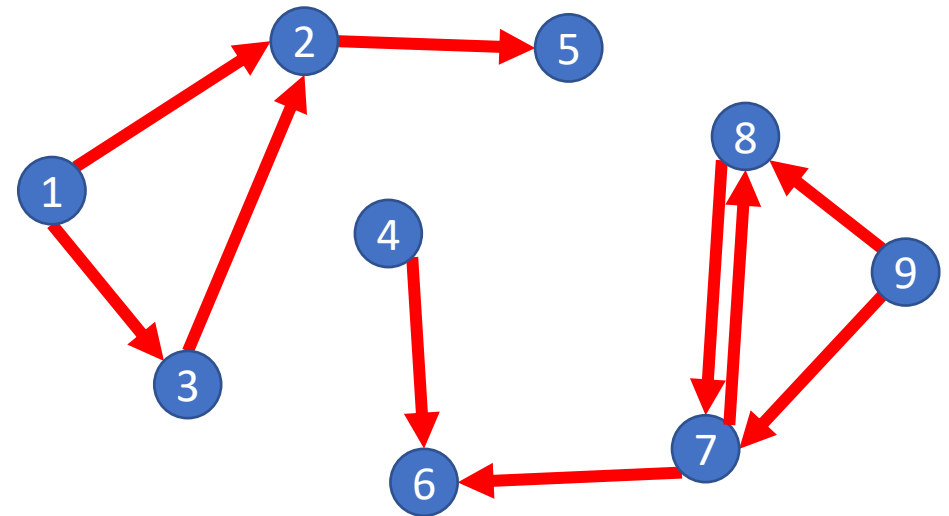
Not (strongly) Connected

Definition: Weakly Connected Graph

A Graph $G = (V, E)$ s.t. for any pair of nodes $v_1, v_2 \in V$ there is a path from v_1 to v_2 ignoring direction of edges



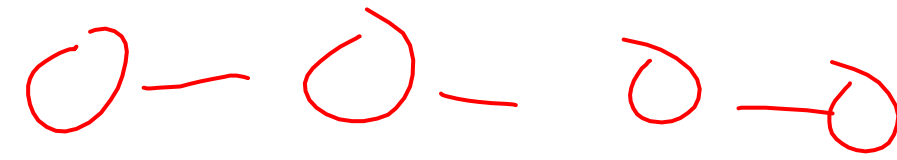
Weakly Connected



Not Weakly Connected

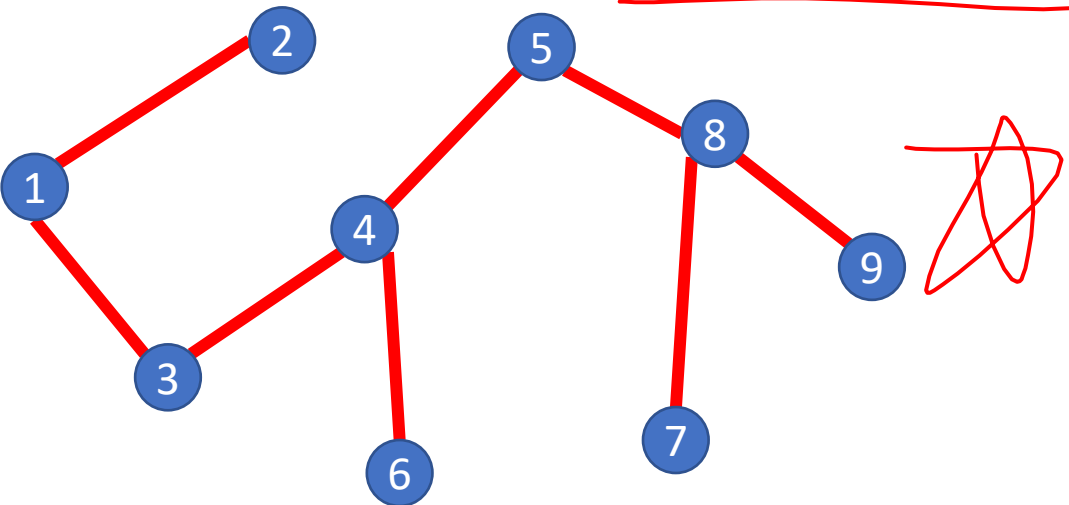
Graph Density, Data Structures, Efficiency

- The maximum number of edges in a graph is $\Theta(|V|^2)$:
 - Undirected and simple: $\frac{|V|(|V|-1)}{2}$
 - Directed and simple: $|V|(|V|-1)$
 - Direct and non-simple (but no duplicates): $|V|^2$
- If the graph is connected, the minimum number of edges is $|V| - 1$
- If $|E| \in \Theta(|V|^2)$ we say the graph is **dense**
- If $|E| \in \Theta(|V|)$ we say the graph is **sparse**
- Because $|E|$ is not always near to $|V|^2$ we do not typically substitute $|V|^2$ for $|E|$ in running times, but leave it as a separate variable
 - However, $\log(|E|) \in \Theta(\log(|V|))$

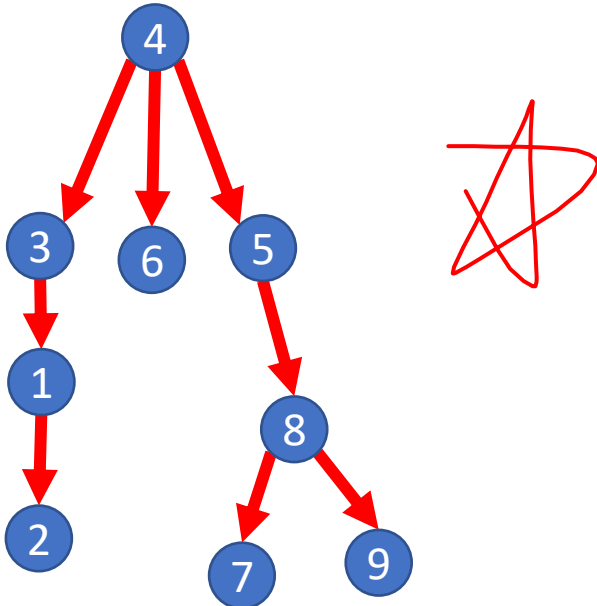


Definition: Tree

A Graph $G = (V, E)$ is a tree if it is undirect, connected, and has no cycles (i.e. is acyclic). Often one node is identified as the “root”







A Tree

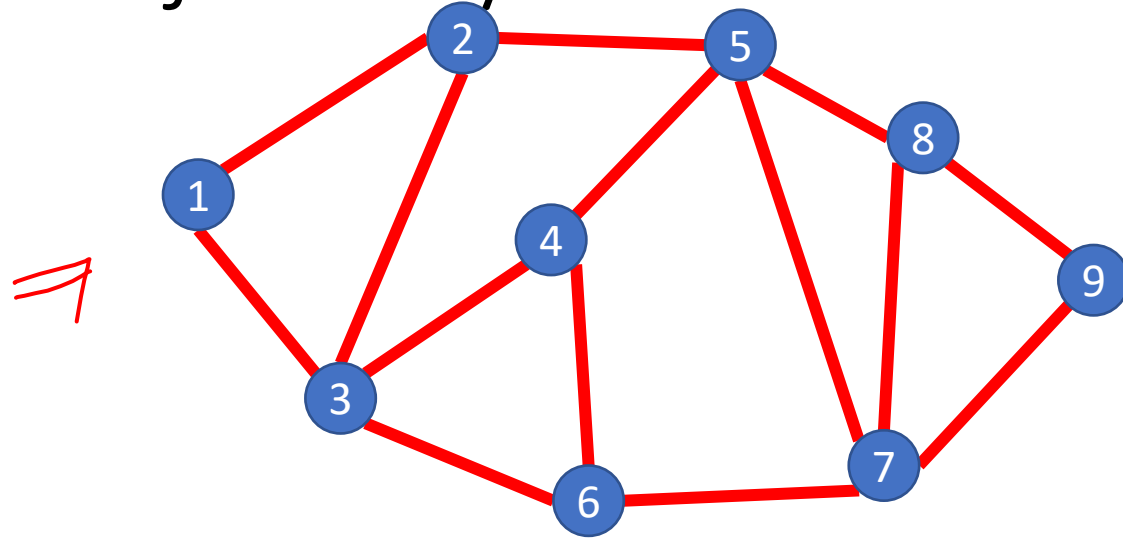


A Rooted Tree

Graph Operations

- To represent a Graph (i.e. build a data structure) we need:
 - Add Edge 
 - Remove Edge 
 - Check if Edge Exists 
 - Get Neighbors (incoming) 
 - Get Neighbors (outgoing)

Adjacency List



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

Time/Space Tradeoffs

Space to represent: $\Theta(n + m)$

Add Edge (v, w) : $\Theta(\deg(v))$

Remove Edge (v, w) : $\Theta(\deg(v))$

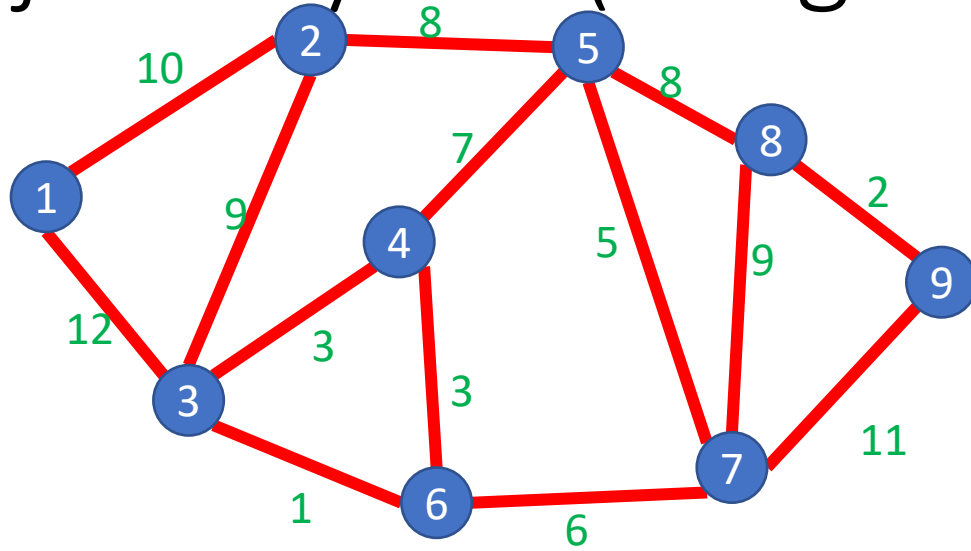
Check if Edge (v, w) Exists: $\Theta(\deg(v))$

Get Neighbors (incoming): $\Theta(n + m)$

Get Neighbors (outgoing): $\Theta(\deg(v))$

$|V| = n$
 $|E| = m$

Adjacency List (Weighted)



Time/Space Tradeoffs

Space to represent: $\Theta(n + m)$

Add Edge (v, w) : $\Theta(\deg(v))$

Remove Edge (v, w) : $\Theta(\deg(v))$

Check if Edge (v, w) Exists: $\Theta(\deg(v))$

Get Neighbors (incoming): $\Theta(n + m)$

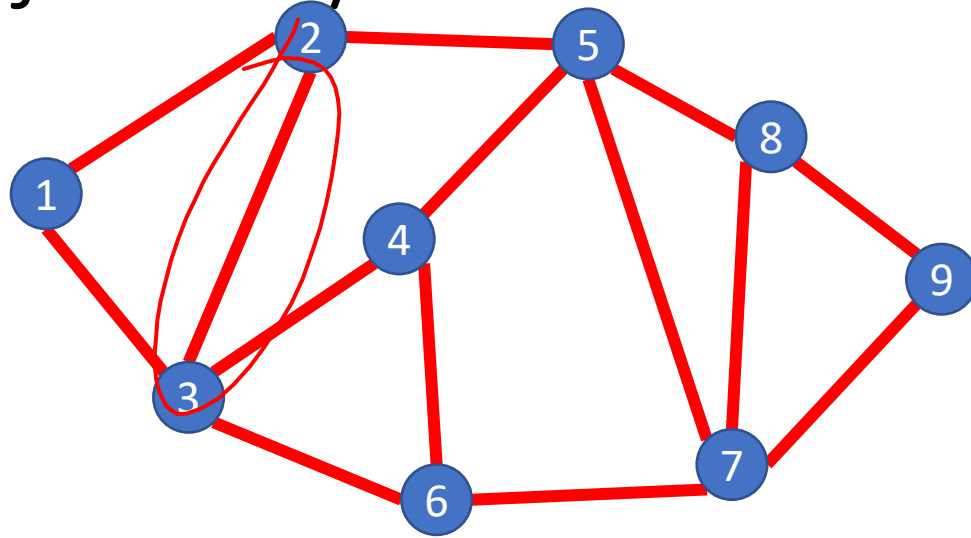
Get Neighbors (outgoing): $\Theta(\deg(v))$

$$|V| = n$$

$$|E| = m$$

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

Adjacency Matrix



1
→ 2
3
4

	A	B	C	D	E	F	G	H	I
A		1	1						
B	1		1		1				
C	1	1		1		1			
D			1		1	1			
E		1		1			1	1	
F			1	1			1		
G					1	1		1	1
H					1		1		1
I							1	1	

Time/Space Tradeoffs

Space to represent: $\Theta(n^2)$

Add Edge (v, w) : $\Theta(1)$

Remove Edge (v, w) : $\Theta(1)$

Check if Edge (v, w) Exists: $\Theta(1)$

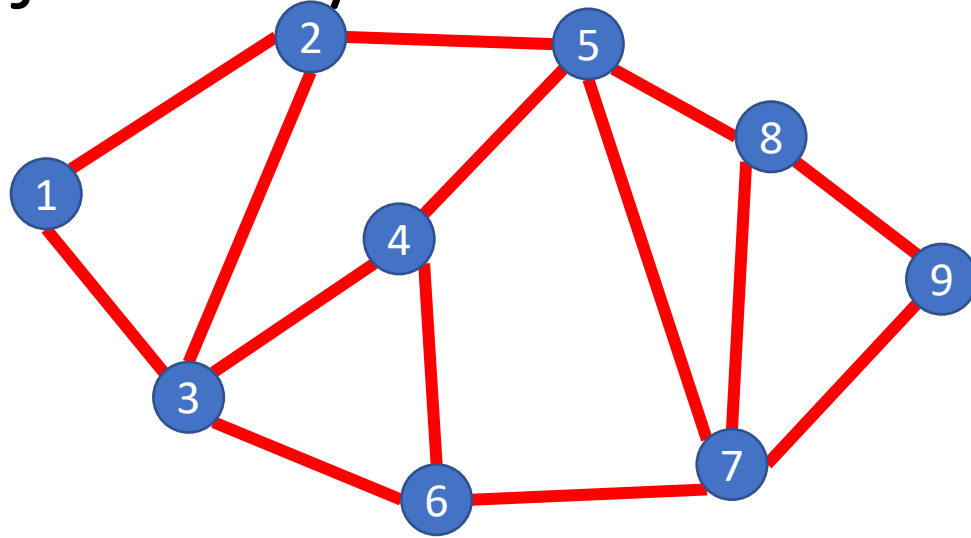
Get Neighbors (incoming): $\Theta(n)$

Get Neighbors (outgoing): $\Theta(n)$

$$|V| = n$$

$$|E| = m$$

Adjacency Matrix



Time/Space Tradeoffs

Space to represent: $\Theta(n^2)$

Add Edge (v, w) : $\Theta(1)$

Remove Edge (v, w) : $\Theta(1)$

Check if Edge (v, w) Exists: $\Theta(1)$

Get Neighbors (incoming): $\Theta(n)$

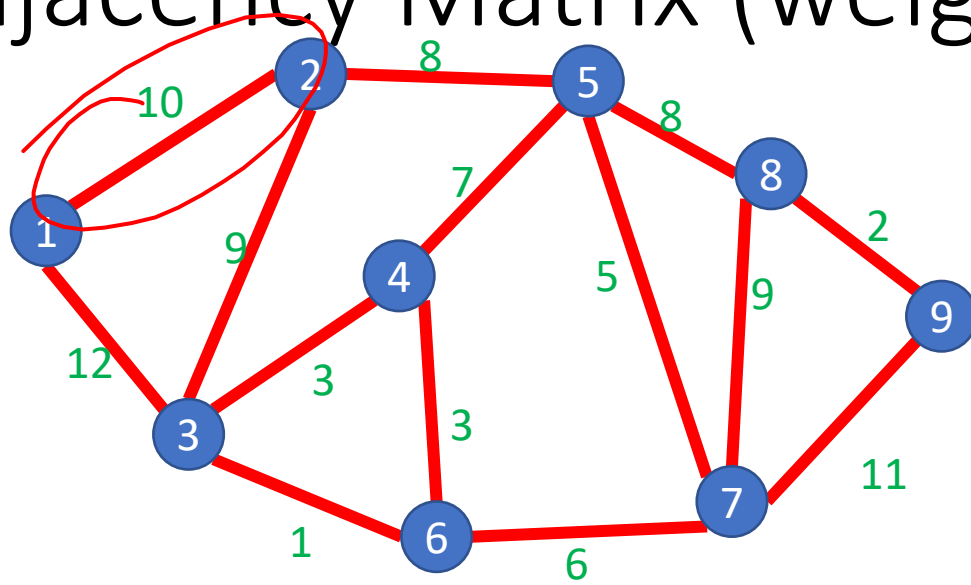
Get Neighbors (outgoing): $\Theta(n)$

$$|V| = n$$

$$|E| = m$$

	A	B	C	D	E	F	G	H	I
A		1	1						
B	1		1		1				
C	1	1		1		1			
D			1		1	1			
E		1		1			1	1	
F			1	1			1		
G					1	1		1	1
H					1		1		1
I							1	1	

Adjacency Matrix (weighted)



Time/Space Tradeoffs

Space to represent: $\Theta(n^2)$

Add Edge (v, w) : $\Theta(1)$

Remove Edge (v, w) : $\Theta(1)$

Check if Edge (v, w) Exists: $\Theta(1)$

Get Neighbors (incoming): $\Theta(n)$

Get Neighbors (outgoing): $\Theta(n)$

$$|V| = n$$

$$|E| = m$$

	A	B	C	D	E	F	G	H	I
A		1	1						
B	1		1		1				
C	1	1		1		1			
D			1		1	1			
E		1		1			1	1	
F			1	1			1		
G					1	1		1	1
H					1		1		1
I							1	1	

Comparison

- Adjacency List:

- Less memory when $|E| < |V|^2$
- Operations with running time linear in degree of source node
 - Add an edge
 - Remove an edge
 - Check for edge
 - Get neighbors



- Adjacency Matrix:

- Similar amount of memory when $|E| \approx |V|^2$
- Constant time operations:
 - Add an edge
 - Remove an edge
 - Check for an edge
- Operations running with linear time in $|V|$



- Get neighbors



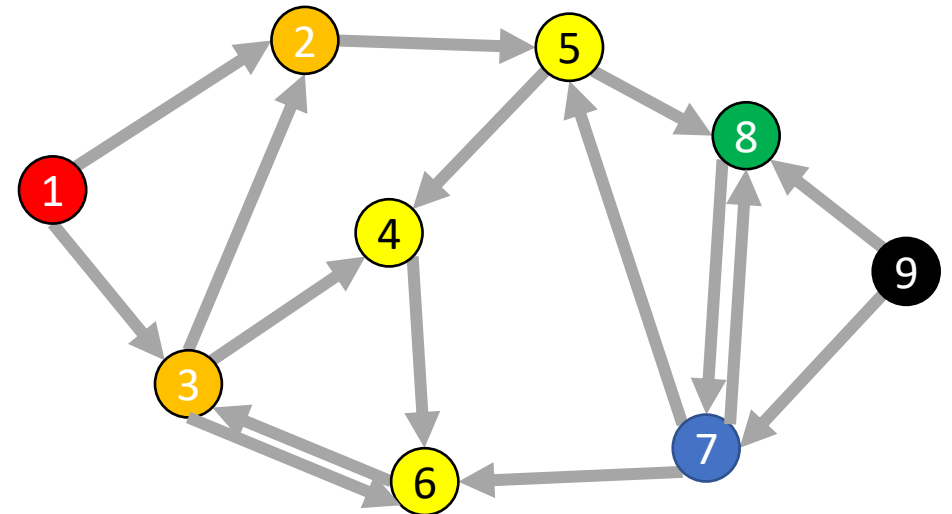
Adjacency List is more common in practice:

- Most graphs have $|E| \ll |V|^2$
 - Saves memory
 - Most nodes will have ~~small degree~~
- Getting neighbors is a common operation
- Adjacency Matrix may be better if the graph is “dense” or if its edges change a lot

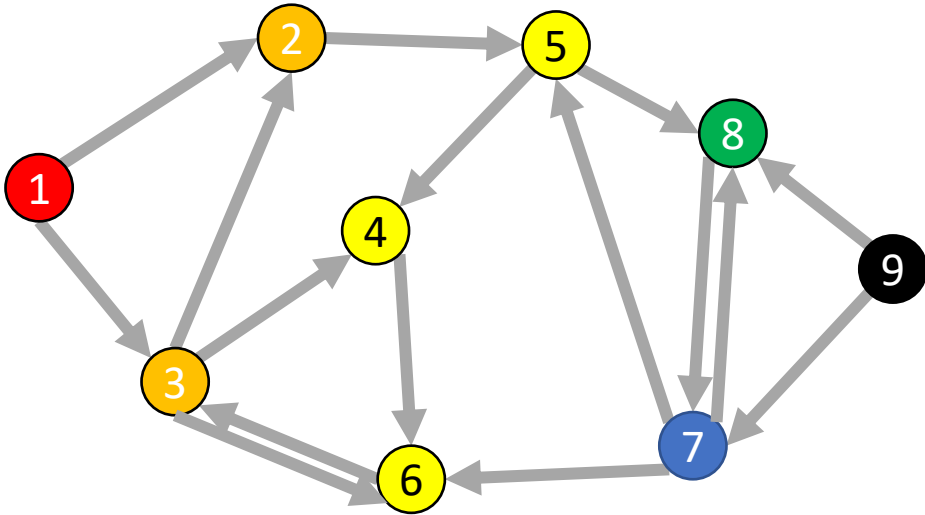


Breadth-First Search

- Input: a node s
- Behavior: Start with node s , visit all neighbors of s , then all neighbors of neighbors of s , ...
- Visits every node reachable from s in order of distance
- Output:
 - How long is the shortest path?
 - Is the graph connected?



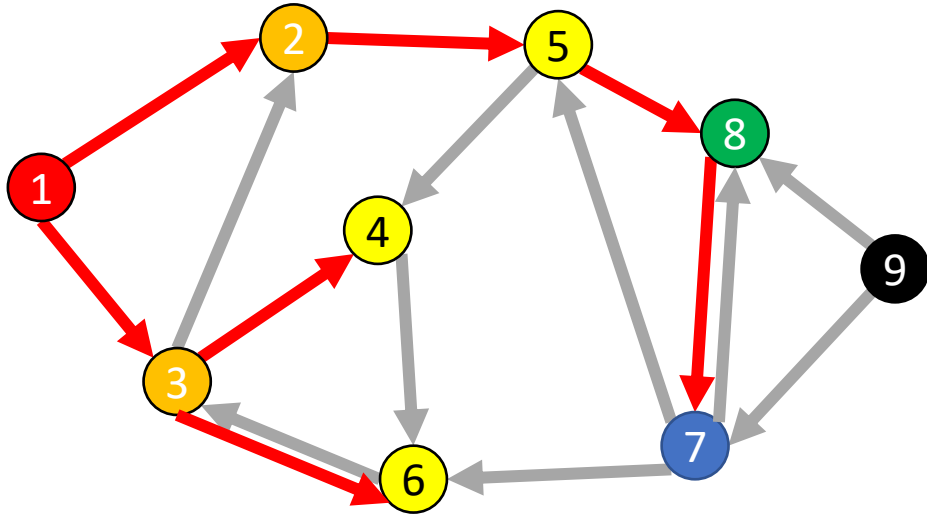
BFS



Running time: $\Theta(|V| + |E|)$

```
void bfs(graph, s){
    found = new Queue();
    found.enqueue(s);
    mark s as "visited";
    While (!found.isEmpty()){
        current = found.dequeue();
        for (v : neighbors(current)){
            if (! v marked "visited"){
                mark v as "visited";
                found.enqueue(v);
            }
        }
    }
}
```


Shortest Path (unweighted)



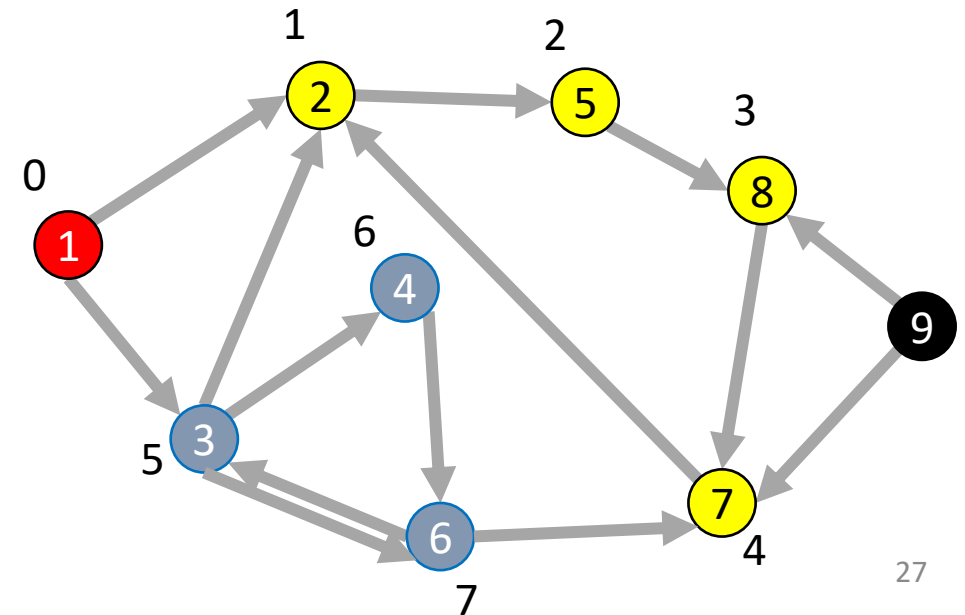
Idea: when it's seen, remember its "layer" depth!

```
int shortestPath(graph, s, t){
    found = new Queue();
    layer = 0;
    found.enqueue(s);
    mark s as "visited";
    While (!found.isEmpty()){
        current = found.dequeue();
        layer = depth of current;
        for (v : neighbors(current)){
            if (! v marked "visited"){
                mark v as "visited";
                depth of v = layer + 1;
                found.enqueue(v);
            }
        }
    }
    return depth of t;
}
```

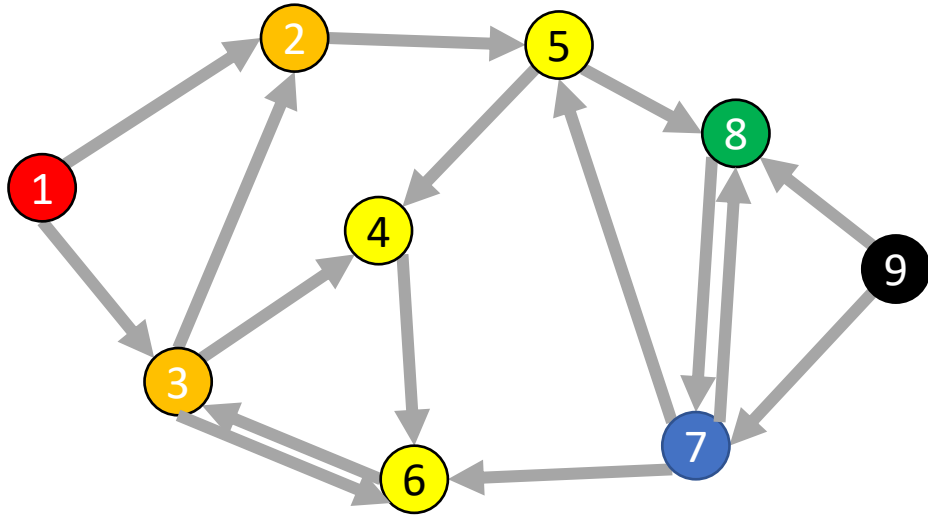
Depth-First Search

Depth-First Search

- Input: a node s
- Behavior: Start with node s , visit one neighbor of s , then all nodes reachable from that neighbor of s , then another neighbor of s ,...
 - Before moving on to the second neighbor of s , visit everything reachable from the first neighbor of s
- Output:
 - Does the graph have a cycle?
 - A **topological sort** of the graph.



DFS (non-recursive)

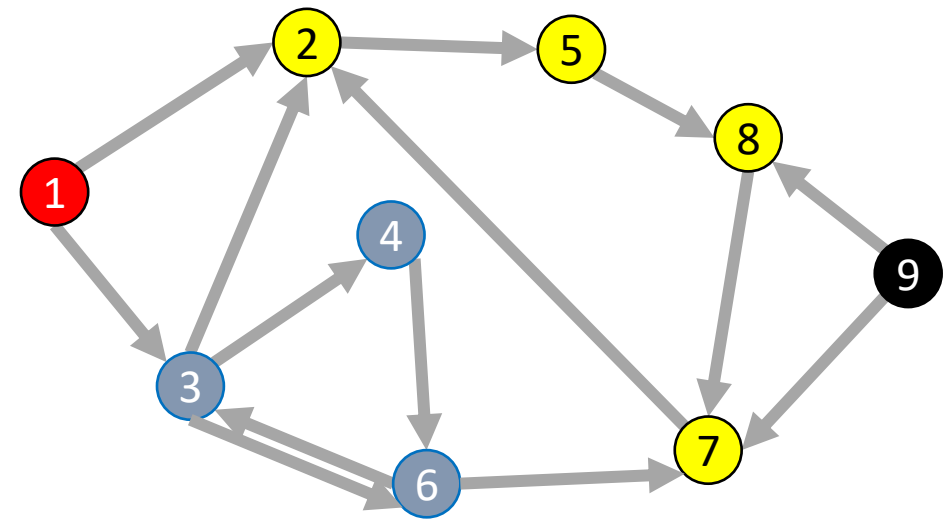


Running time: $\Theta(|V| + |E|)$

```
void dfs(graph, s){
    found = new Stack();
    found.pop(s);
    mark s as "visited";
    While (!found.isEmpty()){
        current = found.pop();
        for (v : neighbors(current)){
            if (! v marked "visited"){
                mark v as "visited";
                found.push(v);
            }
        }
    }
}
```

DFS Recursively (more common)

```
void dfs(graph, curr){  
    mark curr as "visited";  
    for (v : neighbors(current)){  
        if (! v marked "visited"){  
            dfs(graph, v);  
        }  
    }  
    mark curr as "done";  
}
```



Using DFS

- Consider the “visited times” and “done times”
- Edges can be categorized:

- Tree Edge

- (a, b) was followed when pushing
- (a, b) when b was unvisited when we were at a

- Back Edge

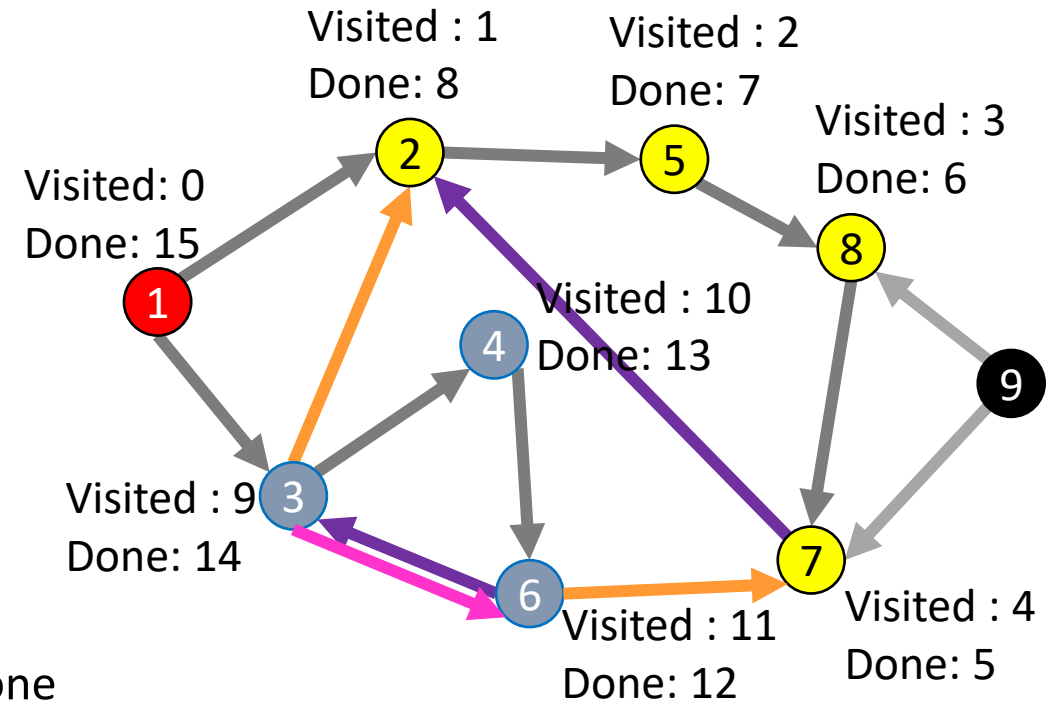
- (a, b) goes to an “ancestor”
- a and b visited but not done when we saw (a, b)
- $t_{visited}(b) < t_{visited}(a) < t_{done}(a) < t_{done}(b)$

- Forward Edge

- (a, b) goes to a “descendent”
- b was visited and done between when a was visited and done
- $t_{visited}(a) < t_{visited}(b) < t_{done}(b) < t_{done}(a)$

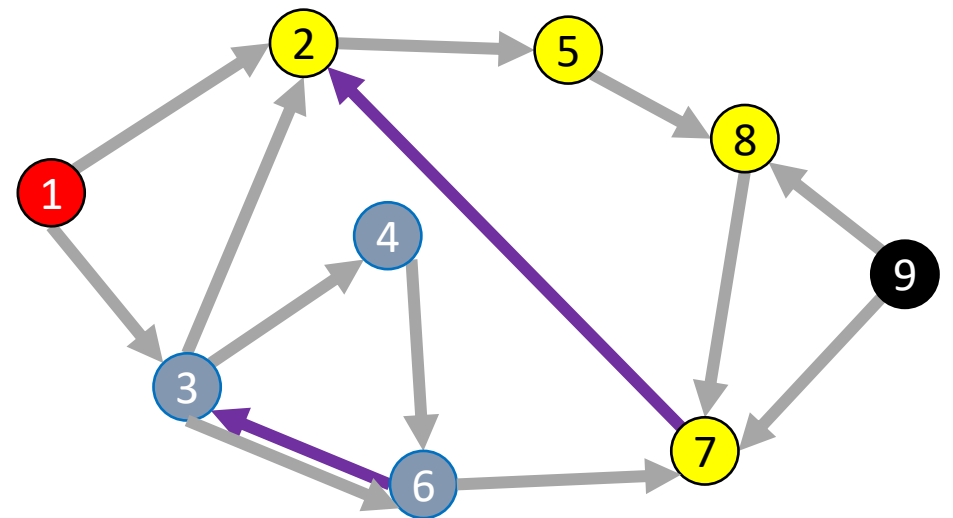
- Cross Edge

- (a, b) goes to a node that doesn't connect to a
- b was seen and done before a was ever visited
- $t_{done}(b) < t_{visited}(a)$



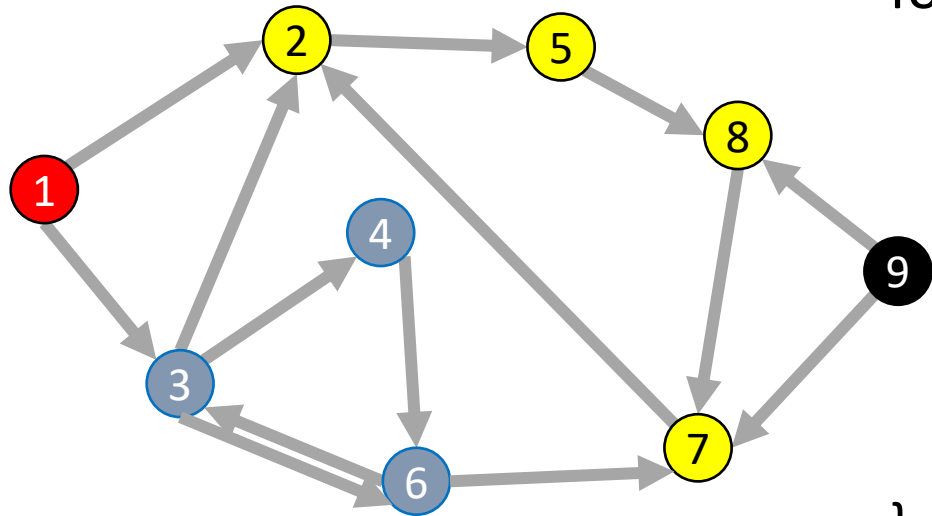
Back Edges

- Behavior of DFS:
 - “Visit everything reachable from the current node before going back”
- Back Edge:
 - The current node’s neighbor is an “in progress” node
 - Since that other node is “in progress”, the current node is reachable from it
 - The back edge is a path to that other node
 - **Cycle!**



Cycle Detection

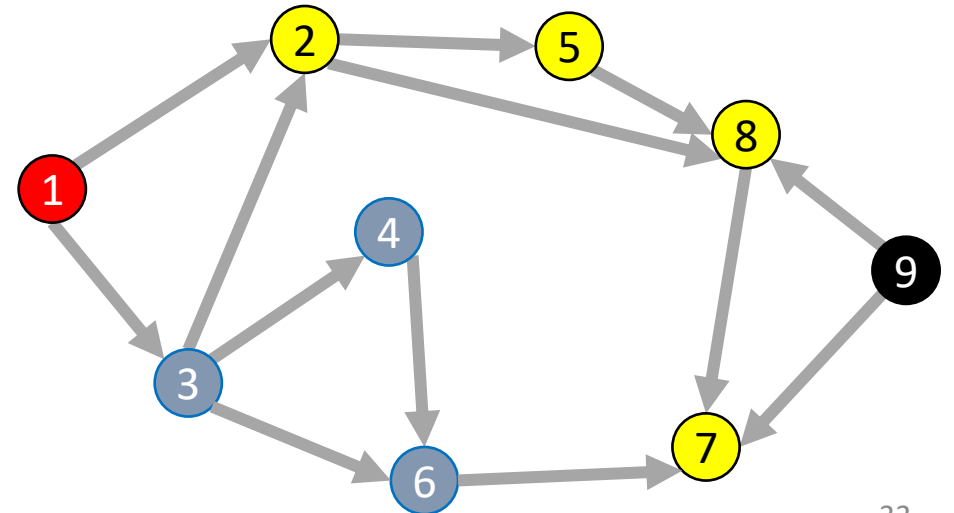
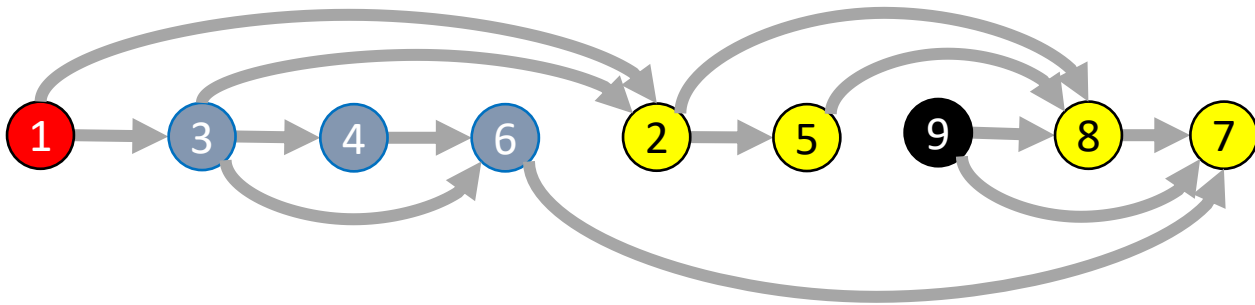
Idea: Look for a back edge!



```
boolean hasCycle(graph, curr){
  mark curr as "visited";
  cycleFound = false;
  for (v : neighbors(current)){
    if (v marked "visited" && ! v marked "done"){
      cycleFound=true;
    }
    if (! v marked "visited" && !cycleFound){
      cycleFound = hasCycle(graph, v);
    }
  }
  mark curr as "done";
  return cycleFound;
}
```


Topological Sort

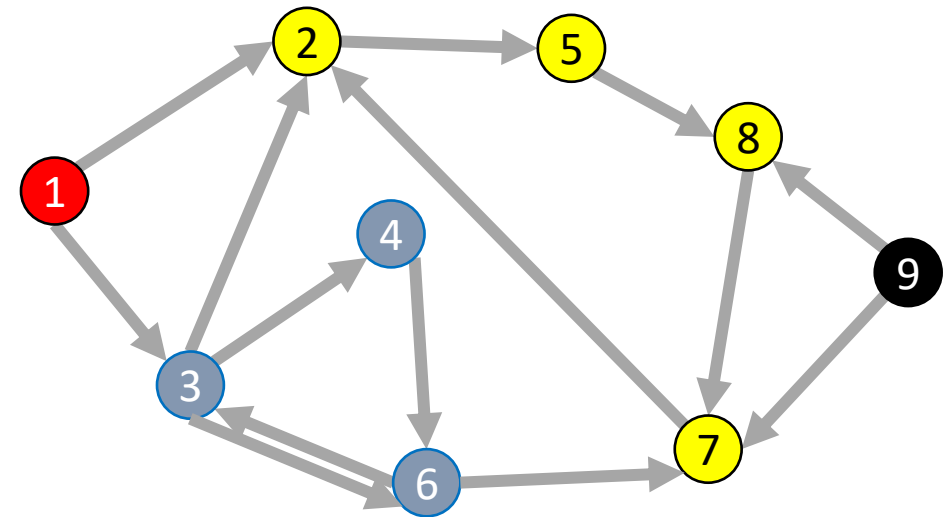
- A Topological Sort of a **directed acyclic graph** $G = (V, E)$ is a permutation of V such that if $(u, v) \in E$ then u is before v in the permutation



DFS Recursively

```
void dfs(graph, curr){  
    mark curr as "visited";  
    for (v : neighbors(current)){  
        if (! v marked "visited"){  
            dfs(graph, v);  
        }  
    }  
    mark curr as "done";  
}
```

Idea: List in reverse order by "done" time



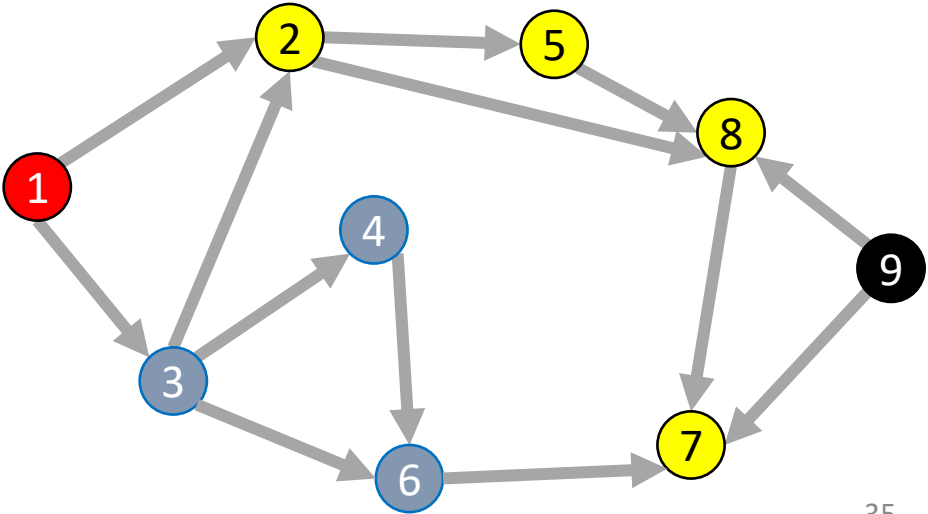
DFS: Topological sort

```
List topSort(graph){  
    List<Nodes> done = new List<>();  
    for (Node v : graph.vertices){  
        if (!v.visited){  
            finishTime(graph, v, finished);  
        }  
    }  
    done.reverse();  
    return done;  
}
```

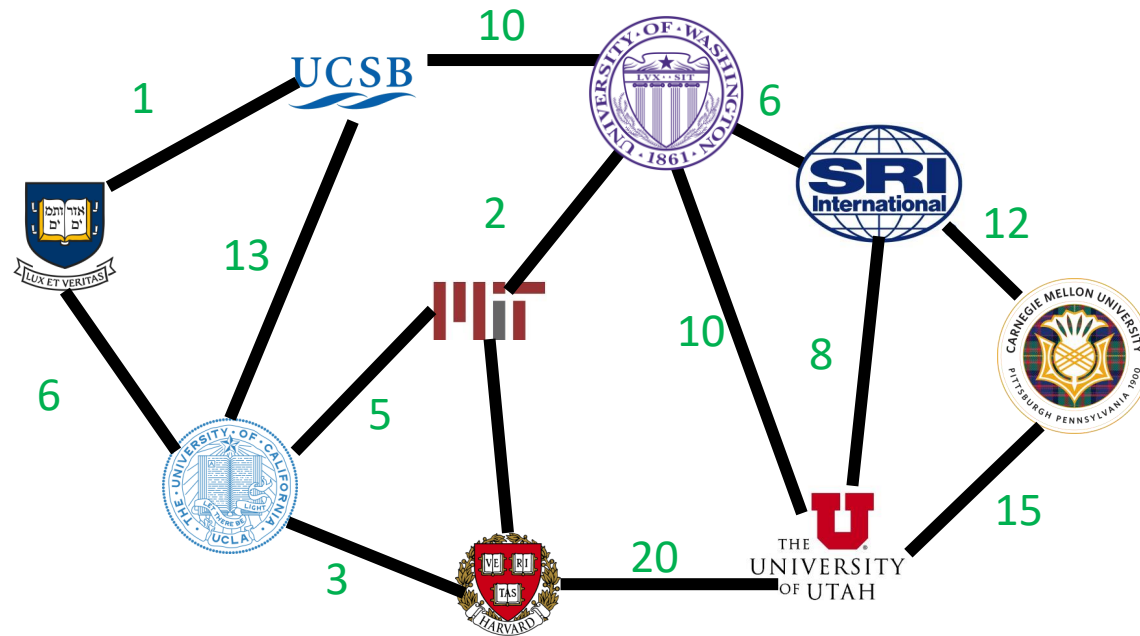
Idea: List in reverse order by “done” time



```
void finishTime(graph, curr, finished){  
    curr.visited = true;  
    for (Node v : curr.neighbors){  
        if (!v.visited){  
            finishTime(graph, v, finished);  
        }  
    }  
    done.add(curr)  
}
```



Single-Source Shortest Path



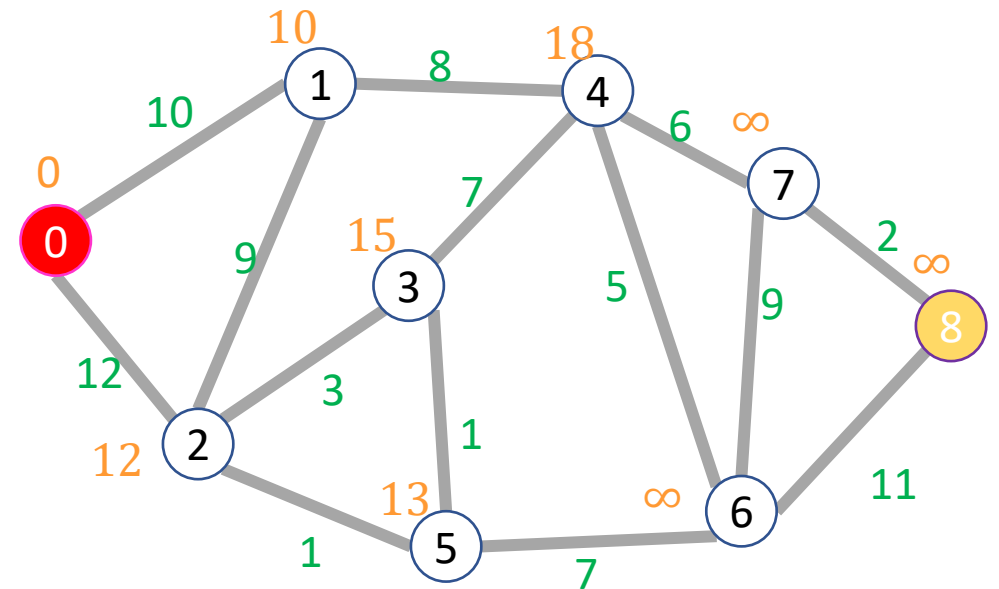
Find the quickest way to get from UVA to each of these other places

Given a graph $G = (V, E)$ and a start node $s \in V$, for each $v \in V$ find the least-weight path from $s \rightarrow v$ (call this weight $\delta(s, v)$)

(assumption: all edge weights are positive)

Dijkstra's Algorithm

- Input: graph with **no negative edge weights**, start node s , end node t
- Behavior: Start with node s , repeatedly go to the incomplete node “nearest” to s , stop when
- Output:
 - Distance from start to end
 - Distance from start to every node



Dijkstra's Algorithm

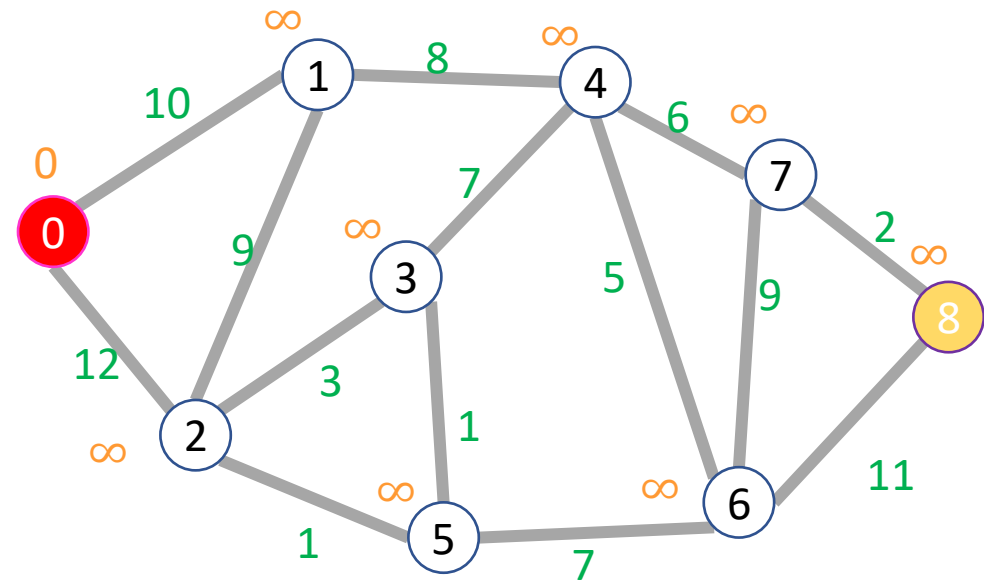
Start: 0

End: 8

Node	Done?
0	F
1	F
2	F
3	F
4	F
5	F
6	F
7	F
8	F

Node	Distance
0	0
1	∞
2	∞
3	∞
4	∞
5	∞
6	∞
7	∞
8	∞

Idea: When a node is the closest undiscovered thing to the start, we have found its shortest path



Dijkstra's Algorithm

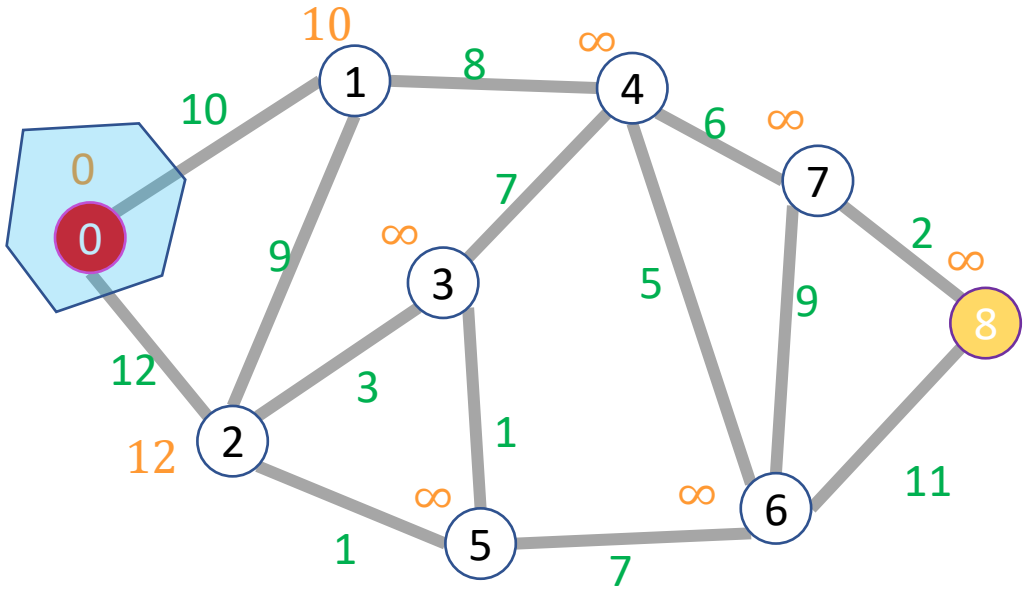
Start: 0

End: 8

Idea: When a node is the closest undiscovered thing to the start, we have found its shortest path

Node	Done?
0	T
1	F
2	F
3	F
4	F
5	F
6	F
7	F
8	F

Node	Distance
0	0
1	10
2	12
3	∞
4	∞
5	∞
6	∞
7	∞
8	∞



Dijkstra's Algorithm

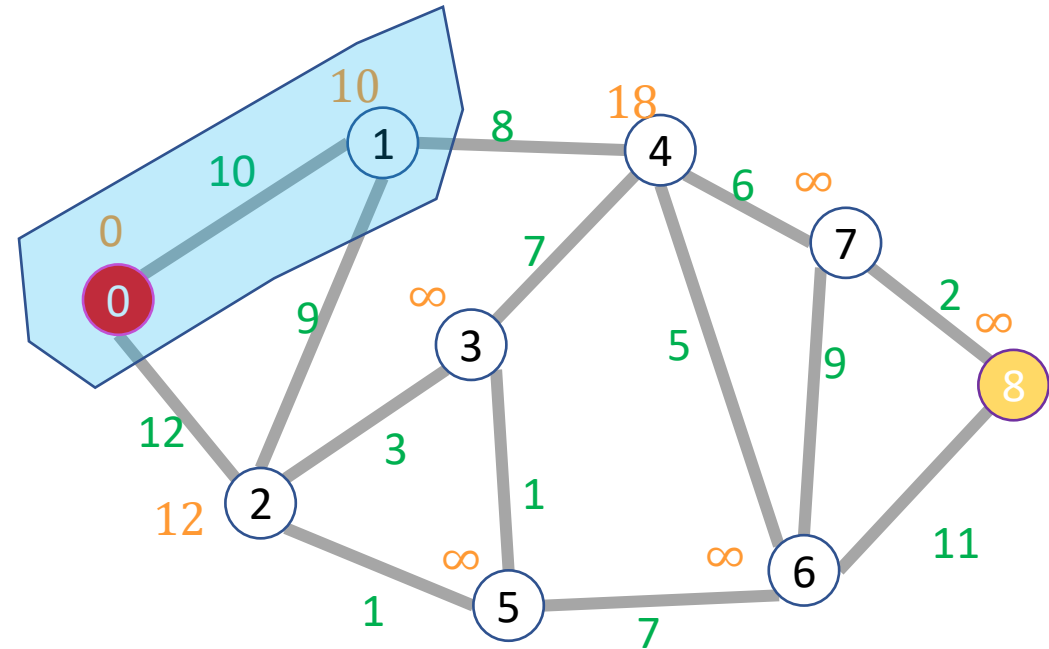
Start: 0

End: 8

Idea: When a node is the closest undiscovered thing to the start, we have found its shortest path

Node	Done?
0	T
1	T
2	F
3	F
4	F
5	F
6	F
7	F
8	F

Node	Distance
0	0
1	10
2	12
3	∞
4	18
5	∞
6	∞
7	∞
8	∞



Dijkstra's Algorithm

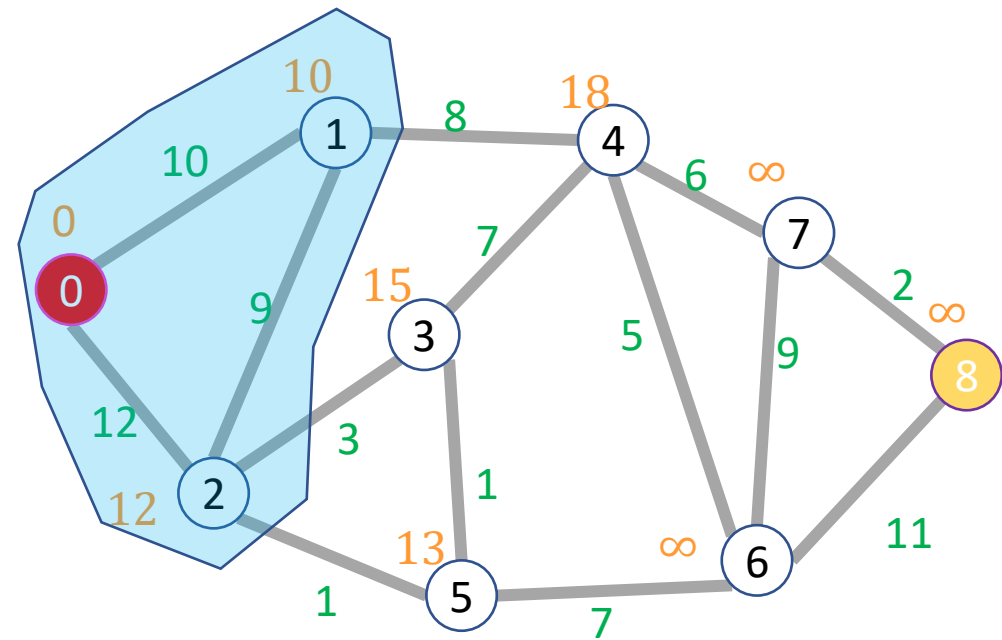
Start: 0

End: 8

Idea: When a node is the closest undiscovered thing to the start, we have found its shortest path

Node	Done?
0	T
1	T
2	T
3	F
4	F
5	F
6	F
7	F
8	F

Node	Distance
0	0
1	10
2	12
3	15
4	18
5	13
6	∞
7	∞
8	∞



Dijkstra's Algorithm

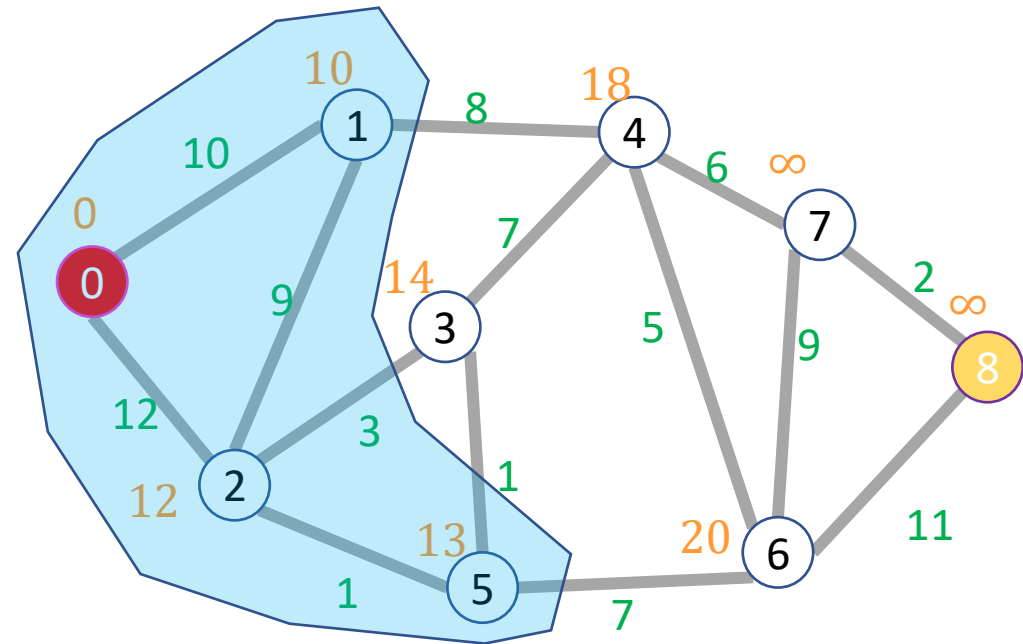
Start: 0

End: 8

Idea: When a node is the closest undiscovered thing to the start, we have found its shortest path

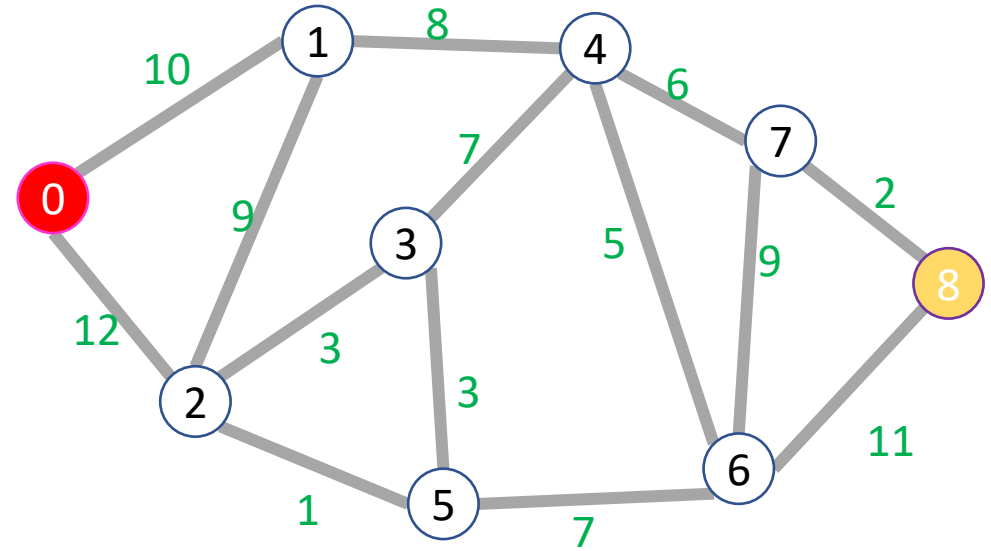
Node	Done?
0	T
1	T
2	T
3	F
4	F
5	T
6	F
7	F
8	F

Node	Distance
0	0
1	10
2	12
3	14
4	18
5	13
6	∞
7	20
8	∞



Dijkstra's Algorithm

```
int dijkstras(graph, start, end){
    distances = [ $\infty$ ,  $\infty$ ,  $\infty$ ,...]; // one index per node
    done = [False,False,False,...]; // one index per node
    PQ = new minheap();
    PQ.insert(0, start); // priority=0, value=start
    distances[start] = 0;
    while (!PQ.isEmpty){
        current = PQ.deleteMin();
        done[current] = true;
        for (neighbor : current.neighbors){
            if (!done[neighbor]){
                new_dist = distances[current]+weight(current,neighbor);
                if(distances[neighbor] ==  $\infty$ ){
                    distances[neighbor] = new_dist;
                    PQ.insert(new_dist, neighbor);
                }
                if (new_dist < distances[neighbor]){
                    distances[neighbor] = new_dist;
                    PQ.decreaseKey(new_dist,neighbor); }
            }
        }
    }
    return distances[end]
}
```

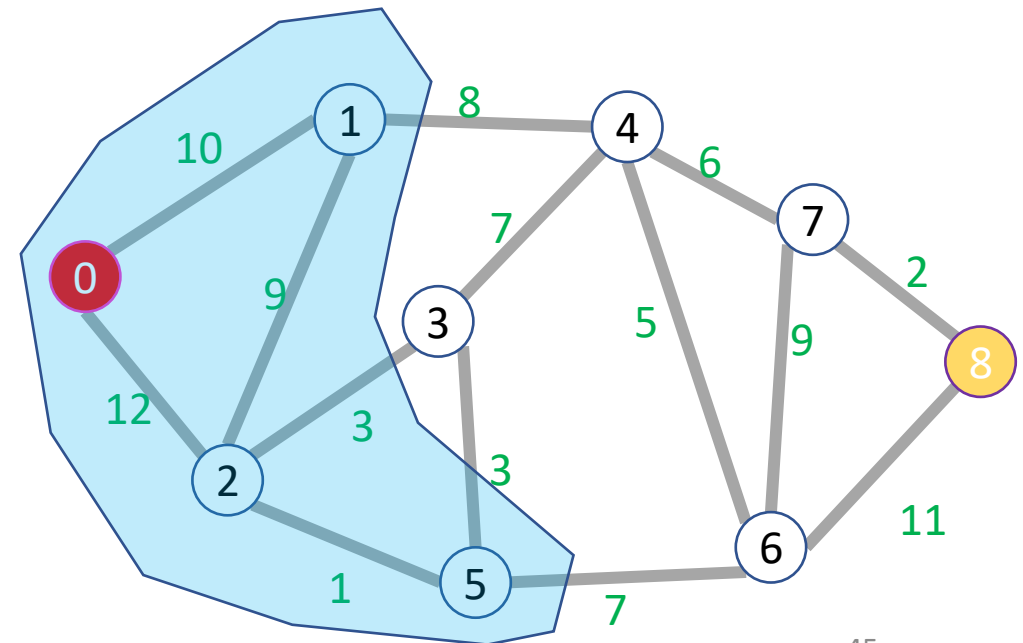


Dijkstra's Algorithm: Running Time

- How many total priority queue operations are necessary?
 - How many times is each node added to the priority queue?
 - How many times might a node's priority be changed?
- What's the running time of each priority queue operation?
- Overall running time:
 - $\Theta(|E| \log |V|)$

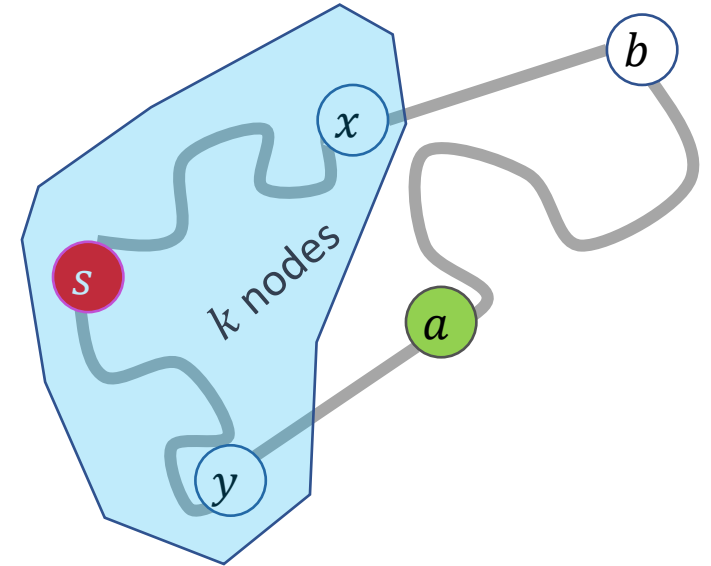
Dijkstra's Algorithm: Correctness

- Claim: when a node is removed from the priority queue, we have found its shortest path
- Induction over number of completed nodes
- Base Case:
- Inductive Step:



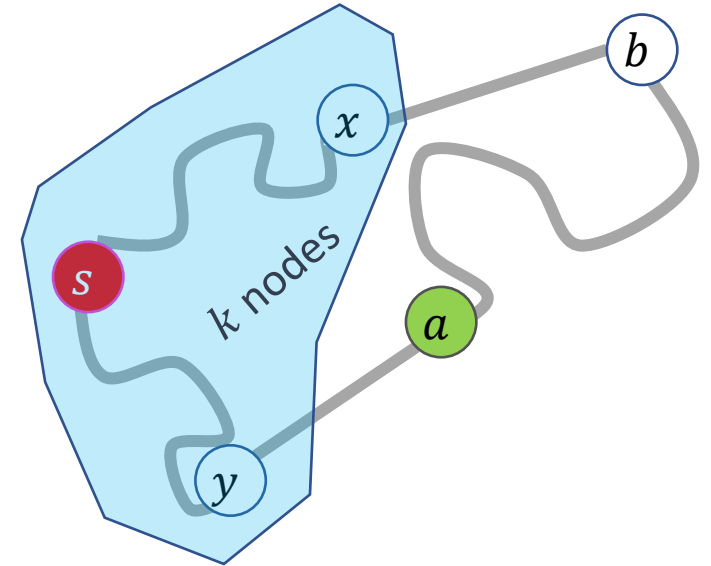
Dijkstra's Algorithm: Correctness

- Claim: when a node is removed from the priority queue, its distance is that of the shortest path
- Induction over number of completed nodes
- Base Case: Only the start node removed
 - It is indeed 0 away from itself
- Inductive Step:
 - If we have correctly found shortest paths for the first k nodes, then when we remove node $k + 1$ we have found its shortest path



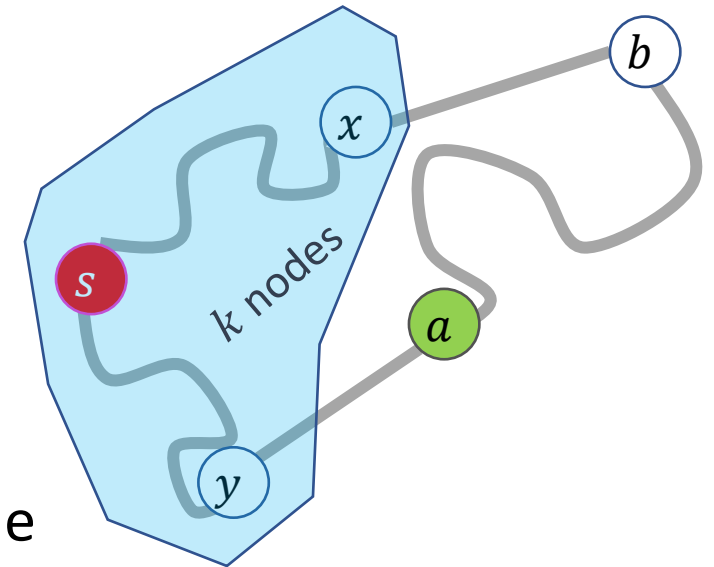
Dijkstra's Algorithm: Correctness

- Suppose a is the next node removed from the queue. What do we know about a ?



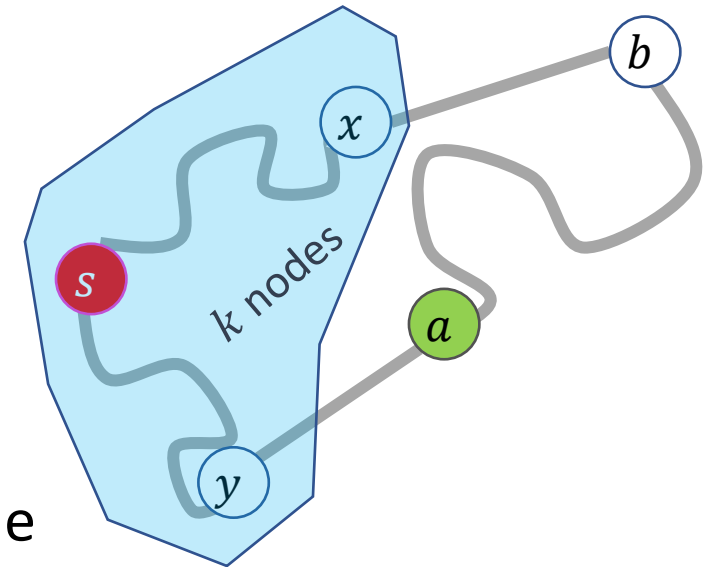
Dijkstra's Algorithm: Correctness

- Suppose a is the next node removed from the queue.
 - No other node incomplete node has a shorter path discovered so far
- Claim: no undiscovered path to a could be shorter
 - Consider any other incomplete node b that is 1 edge away from a complete node
 - a is the closest node that is one away from a complete node
 - Thus no path that includes b can be a shorter path to a
 - Therefore the shortest path to a must use only complete nodes, and therefore we have found it already!



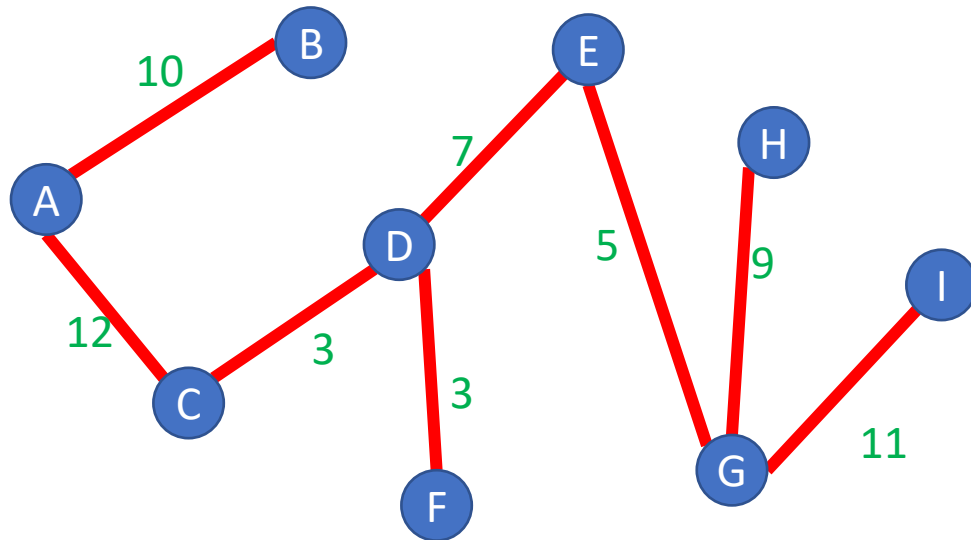
Dijkstra's Algorithm: Correctness

- Suppose a is the next node removed from the queue.
 - No other node incomplete node has a shorter path discovered so far
- Claim: no undiscovered path to a could be shorter
 - Consider any other incomplete node b that is 1 edge away from a complete node
 - a is the closest node that is one away from a complete node
 - **No path from b to a can have negative weight**
 - Thus no path that includes b can be a shorter path to a
 - Therefore the shortest path to a must use only complete nodes, and therefore we have found it already!



Definition: Tree

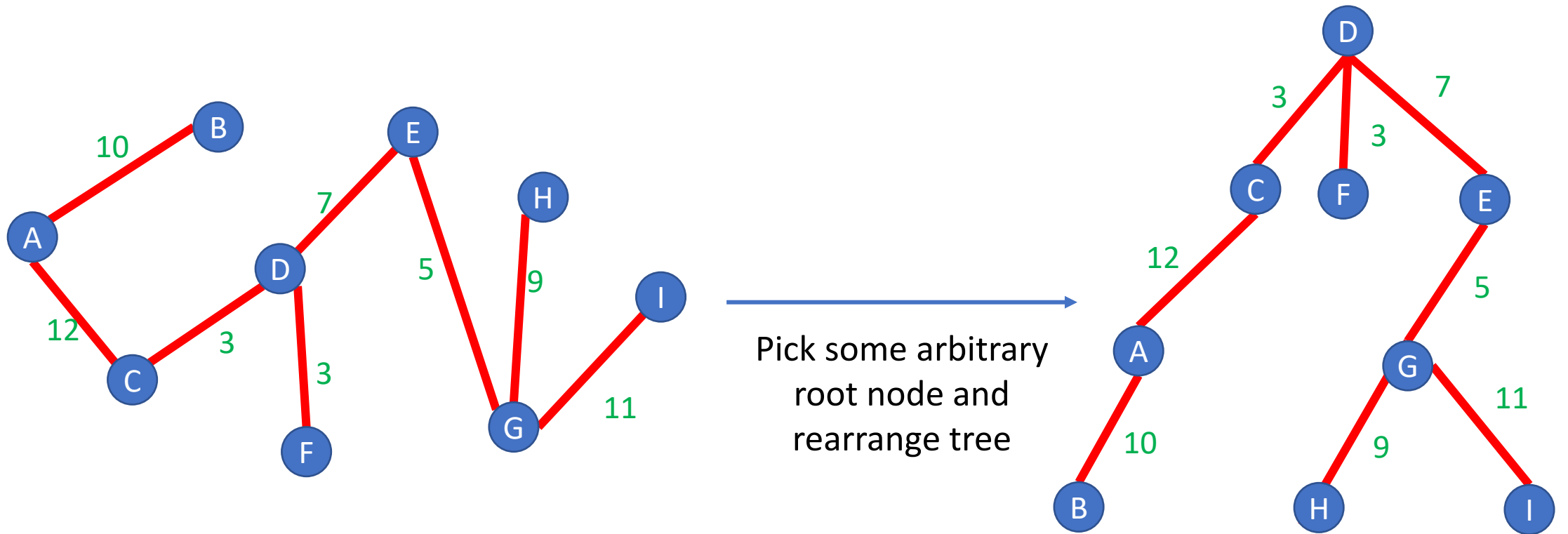
A connected graph with no cycles



Note: A tree does not need a root, but they often do!

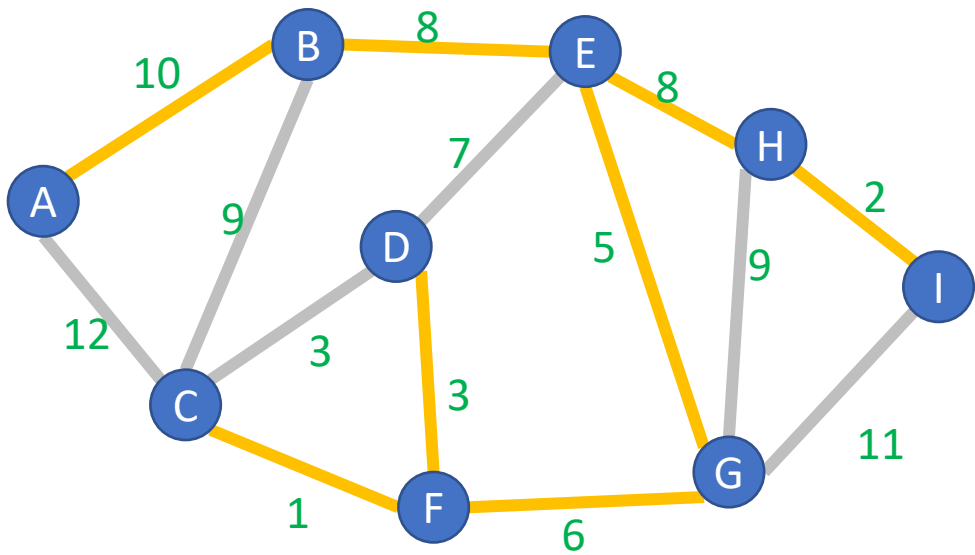
Definition: Tree

A connected graph with no cycles



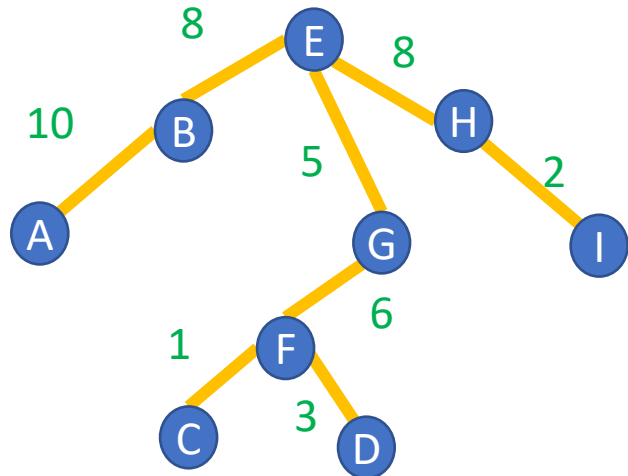
Definition: Spanning Tree

A Tree $T = (V_T, E_T)$ which connects (“spans”) all the nodes in a graph $G = (V, E)$



How many edges does T have?
 $V - 1$

→
Pick some arbitrary root node and rearrange tree

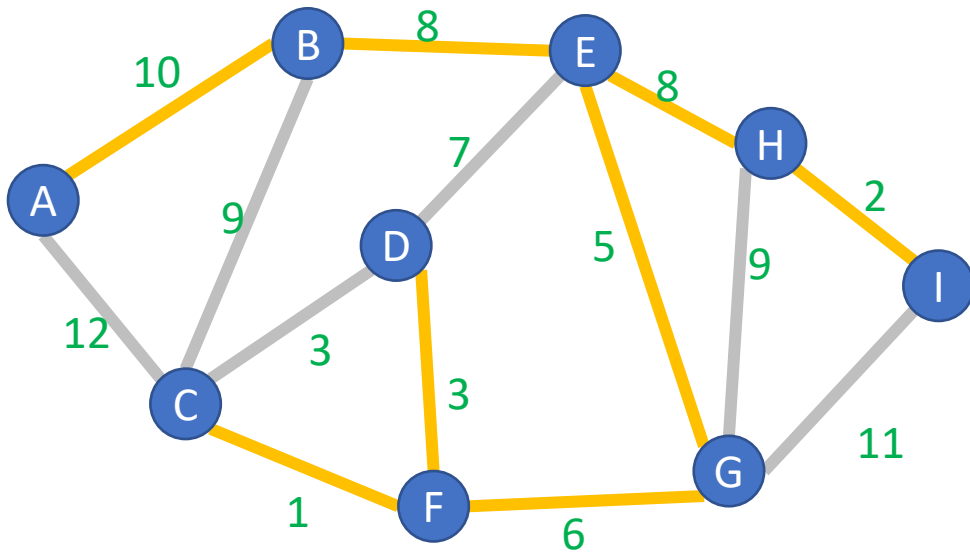


Any set of $V-1$ edges in the graph that doesn't have any cycles is guaranteed to be a spanning tree!

Any set of $V-1$ edges that connects all the nodes in the graph is guaranteed to be a spanning tree!

Definition: Minimum Spanning Tree

A Tree $T = (V_T, E_T)$ which connects (“spans”) all the nodes in a graph $G = (V, E)$, that has minimal **cost**

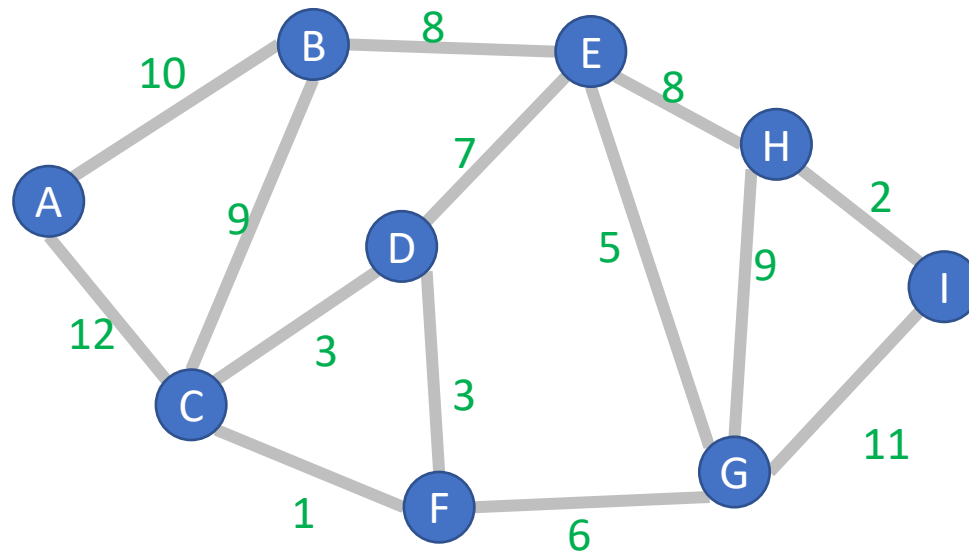


$$Cost(T) = \sum_{e \in E_T} w(e)$$

Kruskal's Algorithm

Start with an empty tree A

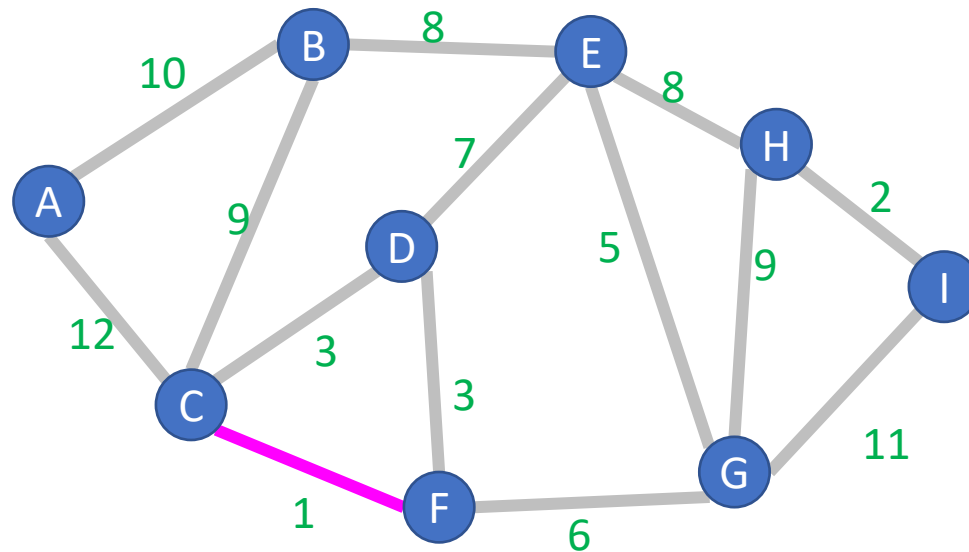
Add to A the lowest-weight edge that does not create a cycle



Kruskal's Algorithm

Start with an empty tree A

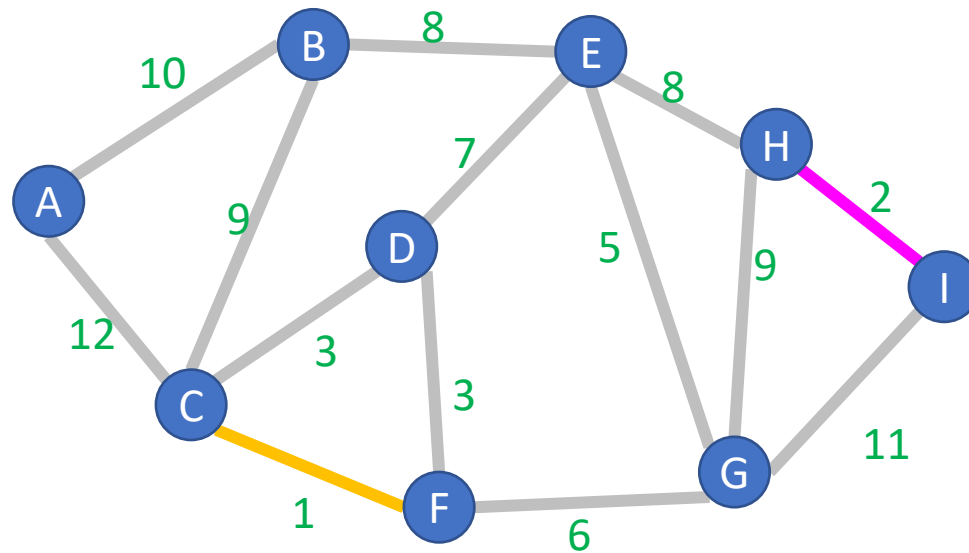
Add to A the lowest-weight edge that does not create a cycle



Kruskal's Algorithm

Start with an empty tree A

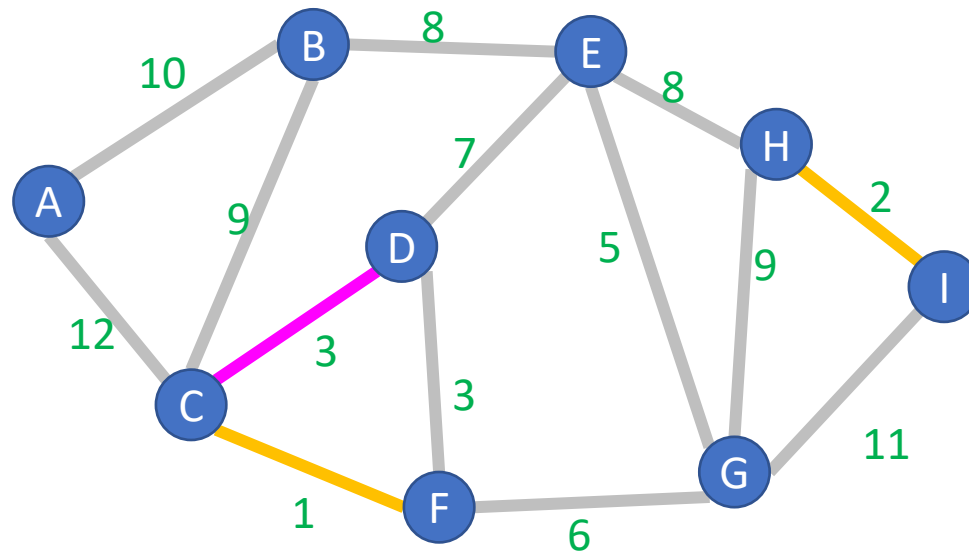
Add to A the lowest-weight edge that does not create a cycle



Kruskal's Algorithm

Start with an empty tree A

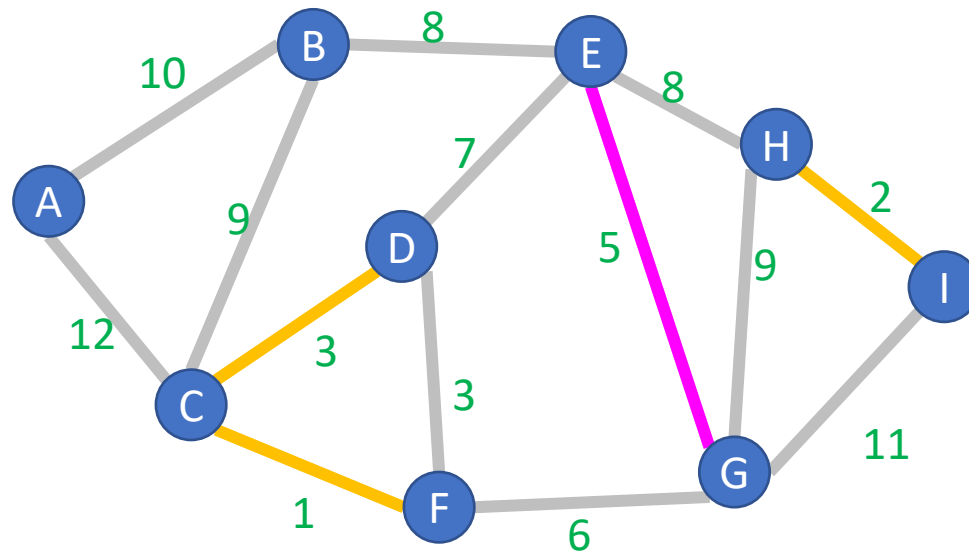
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Kruskal's Algorithm

Start with an empty tree A

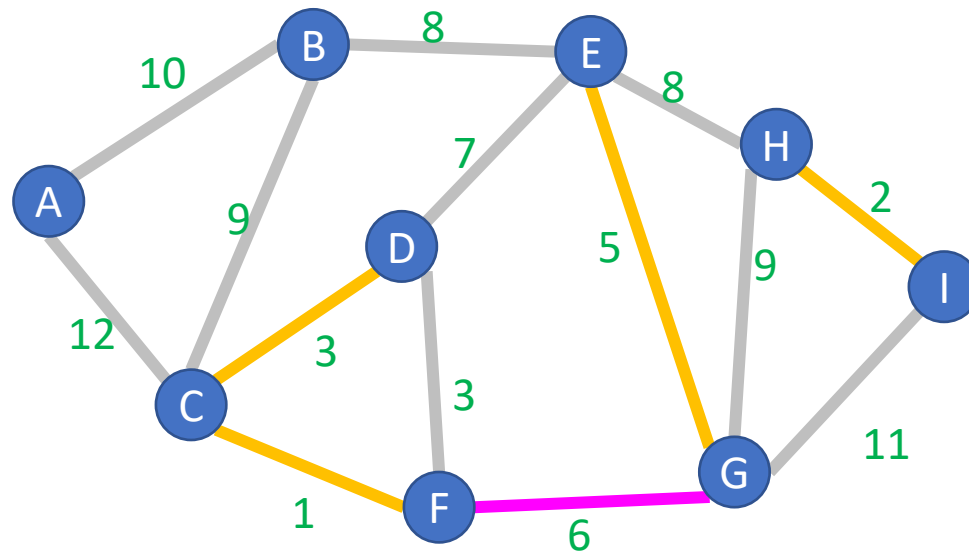
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Kruskal's Algorithm

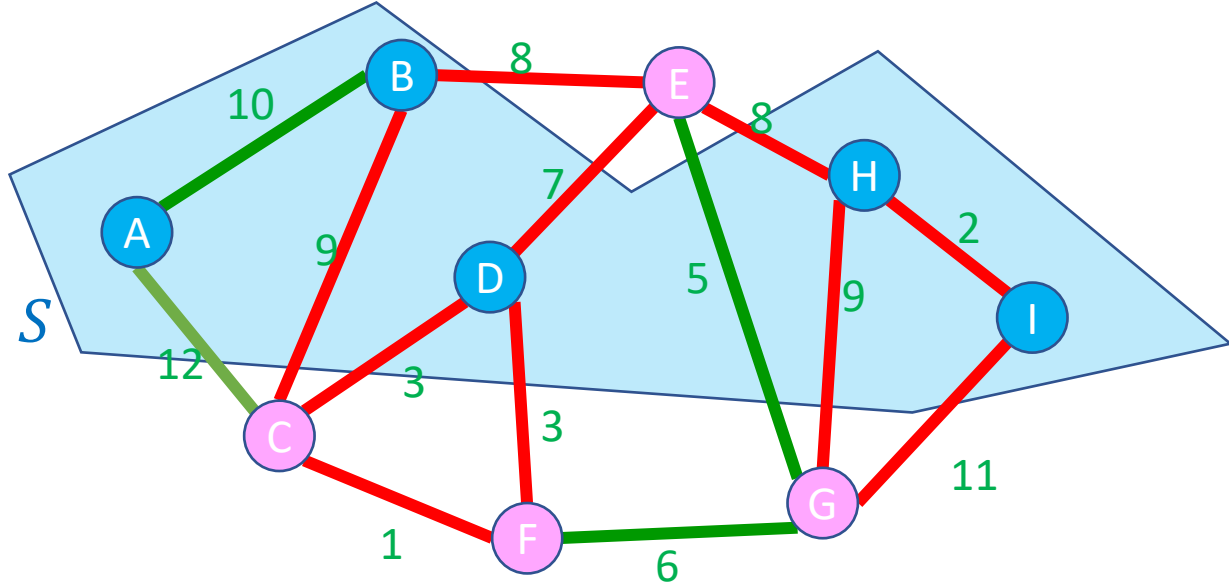
Start with an empty tree A

Add to A the lowest-weight edge that does not create a cycle



Definition: Cut

A Cut of graph $G = (V, E)$ is a partition of the nodes into two sets, S and $V - S$



Edge $(v_1, v_2) \in E$ crosses a cut if $v_1 \in S$ and $v_2 \in V - S$ (or opposite), e.g. (A, C)

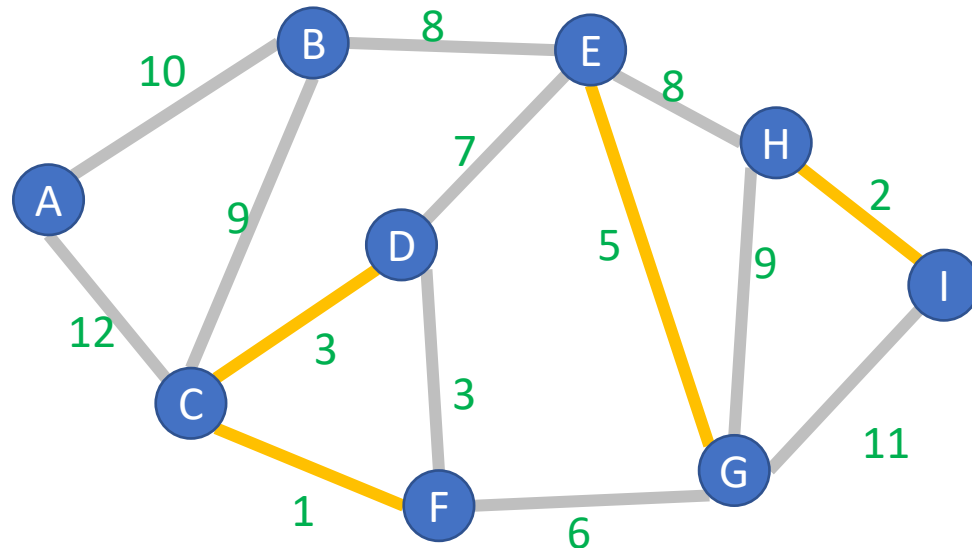
A set of edges R Respects a cut if no edges cross the cut
e.g. $R = \{(A, B), (E, G), (F, G)\}$

Cut Theorem

If a set of edges A is a subset of a minimum spanning tree T , let $(S, V - S)$ be any cut which A respects. Let e be the least-weight edge which crosses $(S, V - S)$. $A \cup \{e\}$ is also a subset of a minimum spanning tree.

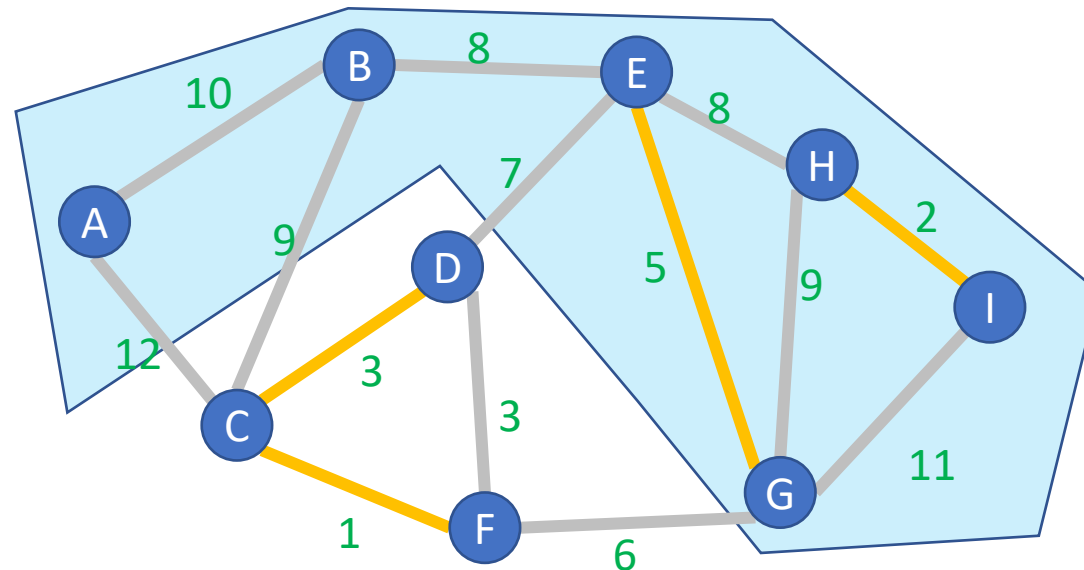
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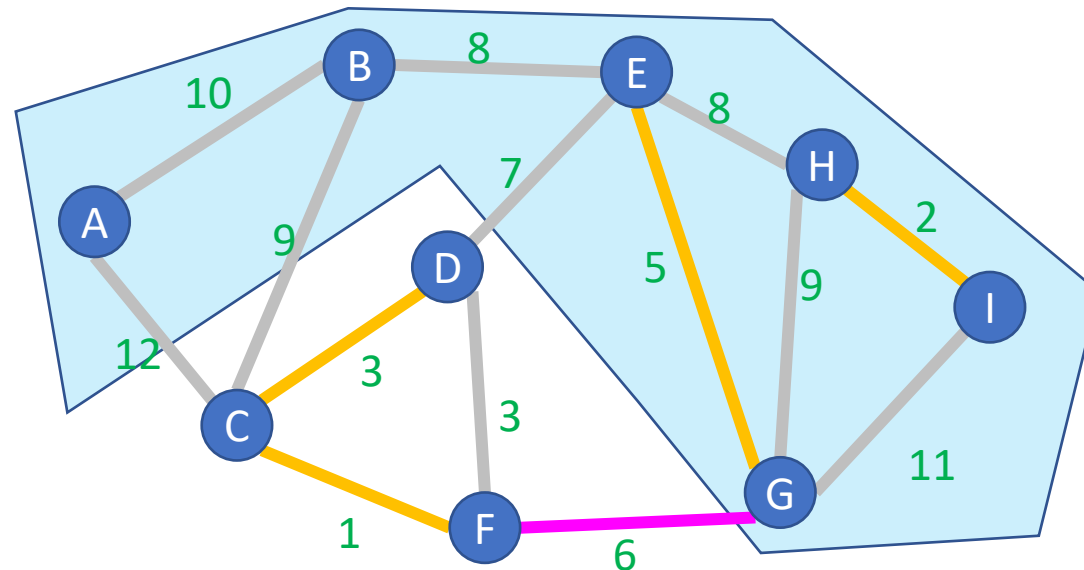
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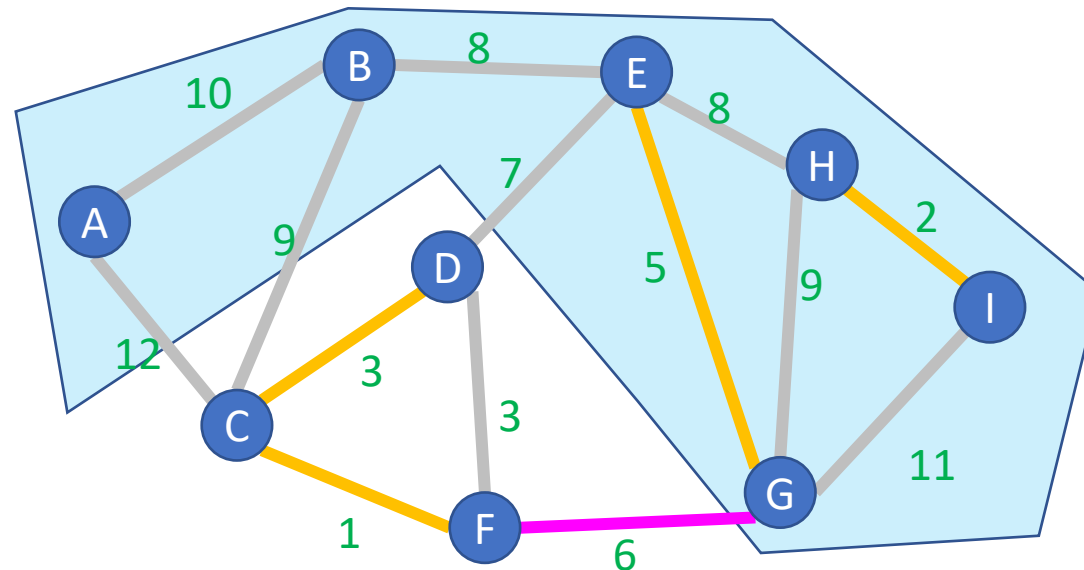
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Cut Theorem

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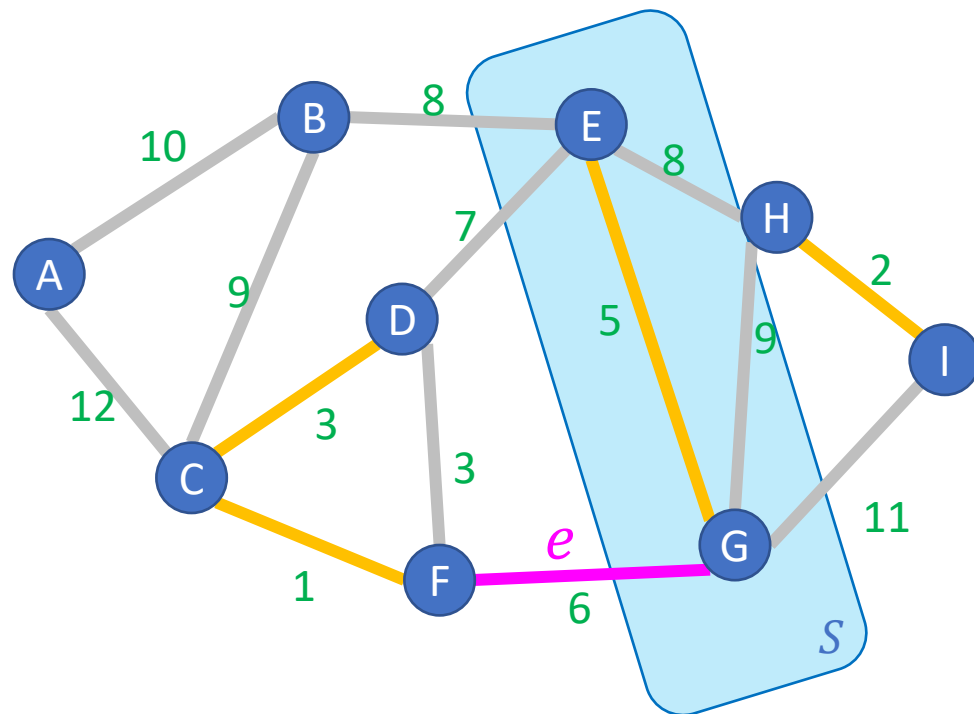


Proof of Kruskal's Algorithm

Start with an empty tree A

Repeat $V - 1$ times:

Add the min-weight edge that doesn't cause a cycle



Proof: Suppose we have some arbitrary set of edges A that Kruskal's has already selected to include in the MST. $e = (F, G)$ is the edge Kruskal's selects to add next

We know that there cannot exist a path from F to G using only edges in A because e does not cause a cycle

We can cut the graph therefore into 2 disjoint sets:

- nodes reachable from G using edges in A
- All other nodes

e is the minimum cost edge that crosses this cut, so by the Cut Theorem, Kruskal's is optimal!

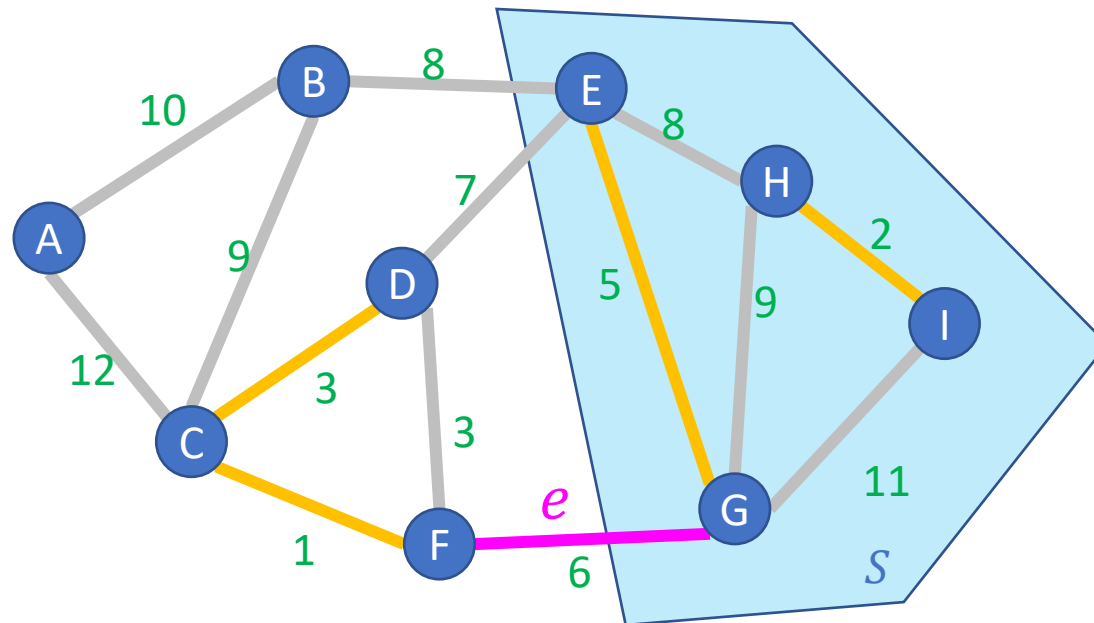
Kruskal's Algorithm Runtime

Start with an empty tree A

Repeat $V - 1$ times:

Add the min-weight edge that doesn't cause a cycle

Keep edges in a Disjoint-set data structure (very fancy)
 $O(E \log V)$



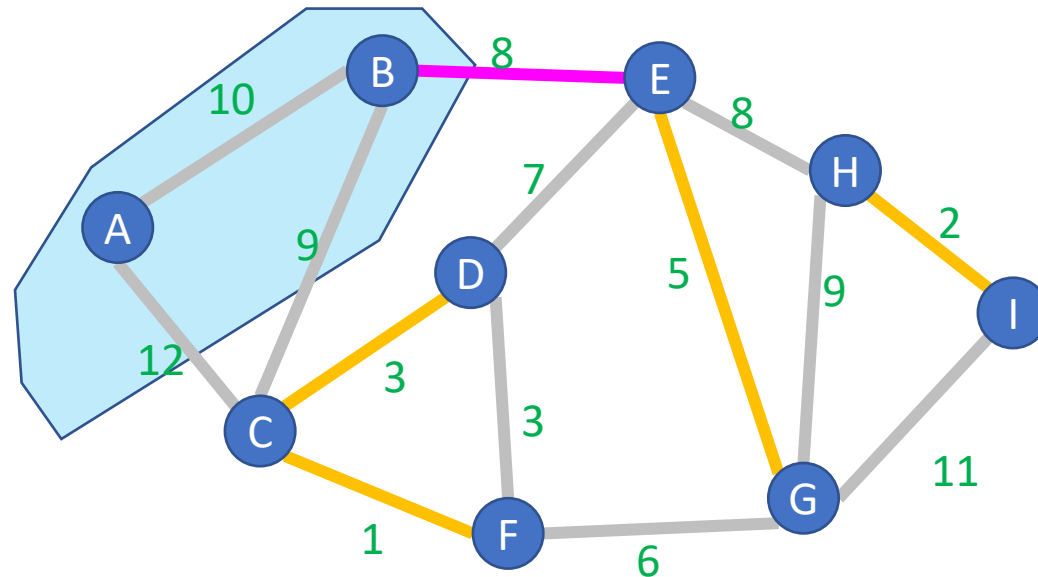
General MST Algorithm

Start with an empty tree A

Repeat $V - 1$ times:

Pick a cut $(S, V - S)$ which A respects (typically implicitly)

Add the **min-weight edge which crosses $(S, V - S)$**



Prim's Algorithm

Start with an empty tree A

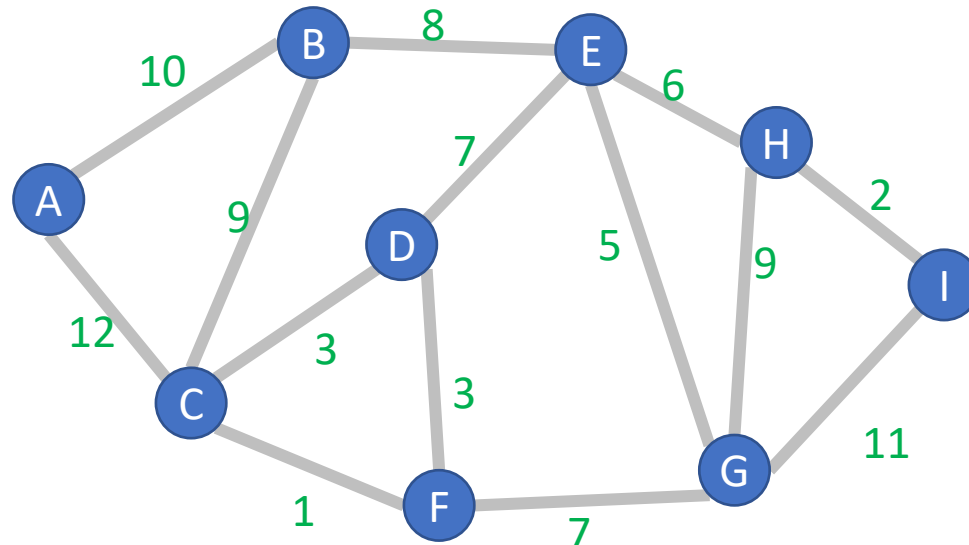
Repeat $V - 1$ times:

Pick a cut $(S, V - S)$ which A respects

Add the min-weight edge which crosses $(S, V - S)$

S is all endpoint of edges in A

e is the min-weight edge that grows the tree



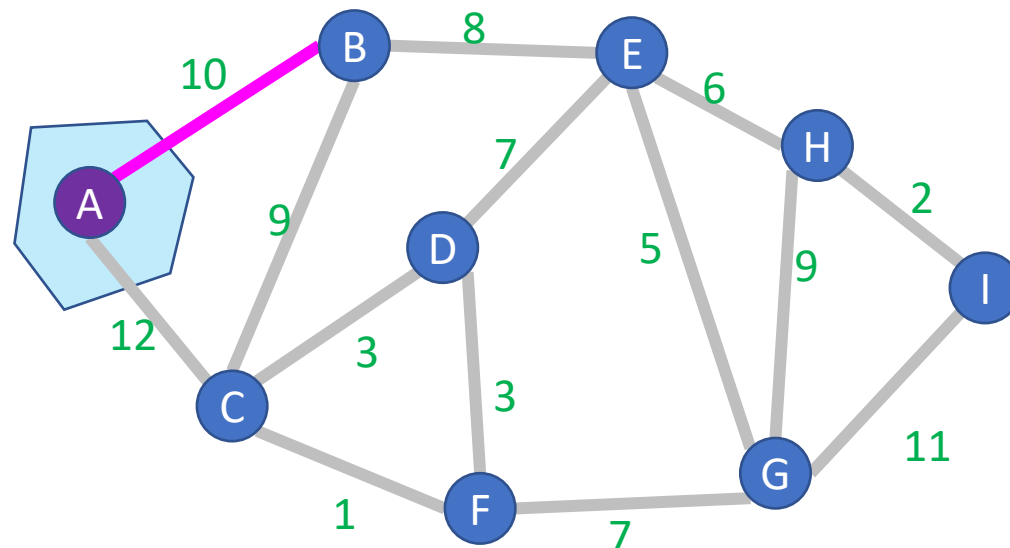
Prim's Algorithm

Start with an empty tree A

Pick a **start node**

Repeat $V - 1$ times:

Add **the min-weight edge** which connects to node
in A with a node not in A



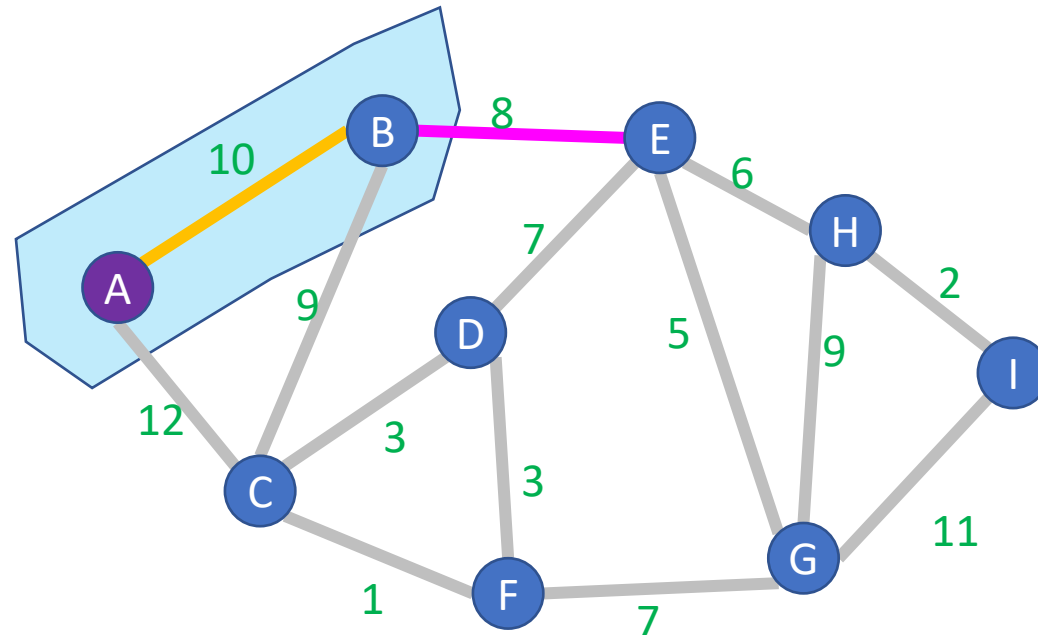
Prim's Algorithm

Start with an empty tree A

Pick a **start node**

Repeat $V - 1$ times:

Add **the min-weight edge** which connects to node
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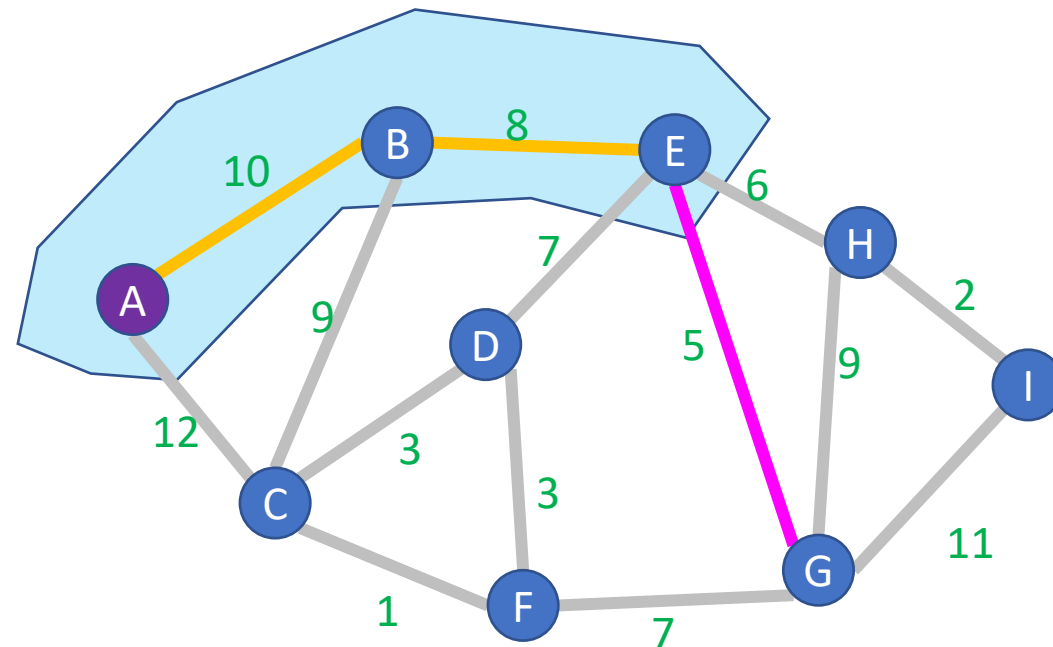
Prim's Algorithm

Start with an empty tree A

Pick a **start node**

Repeat $V - 1$ times:

Add **the min-weight edge** which connects to node
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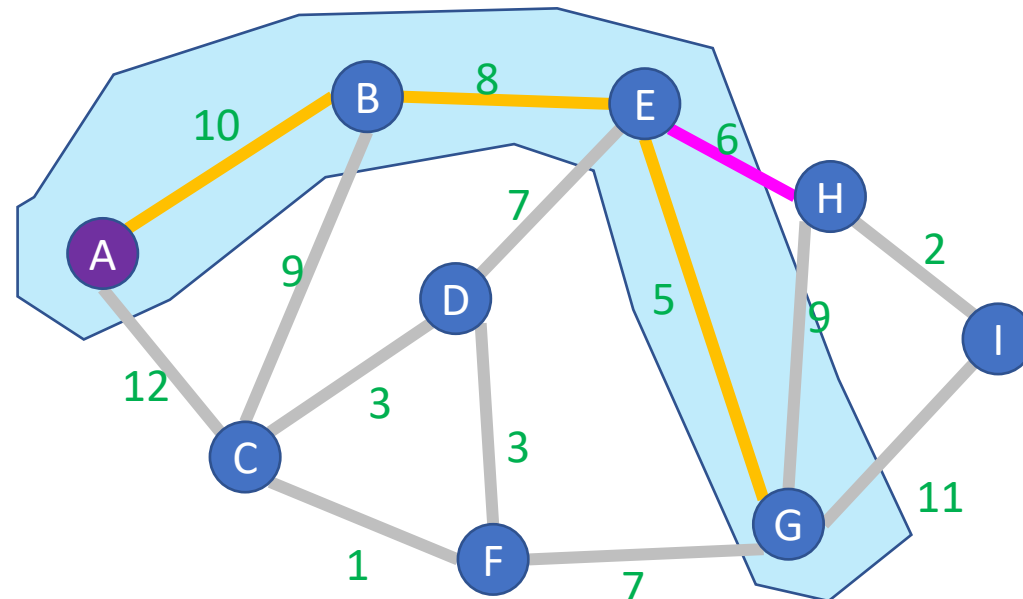
Prim's Algorithm

Start with an empty tree A

Pick a **start node**

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Add **the min-weight edge** which connects to node
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Prim's Algorithm

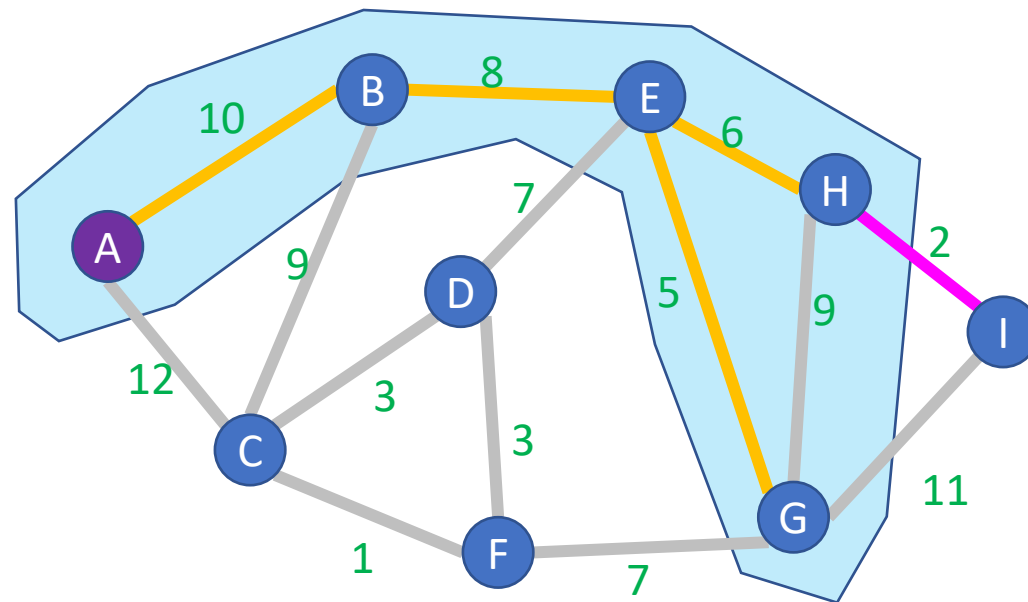
Start with an empty tree A

Pick a **start node**

Repeat $V - 1$ times:

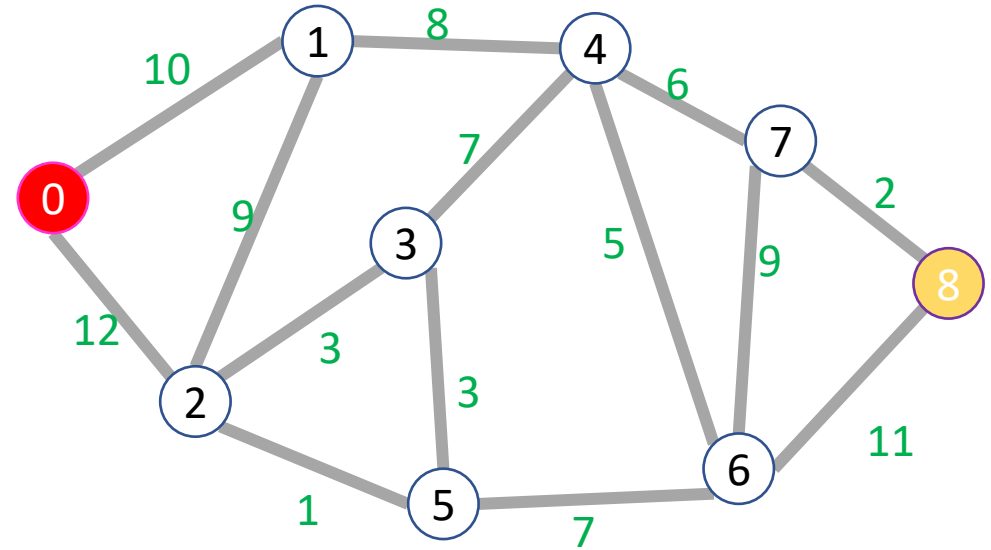
Add **the min-weight edge** which connects to node
in A with a node not in A

Keep edges in a Heap
 $O(E \log V)$



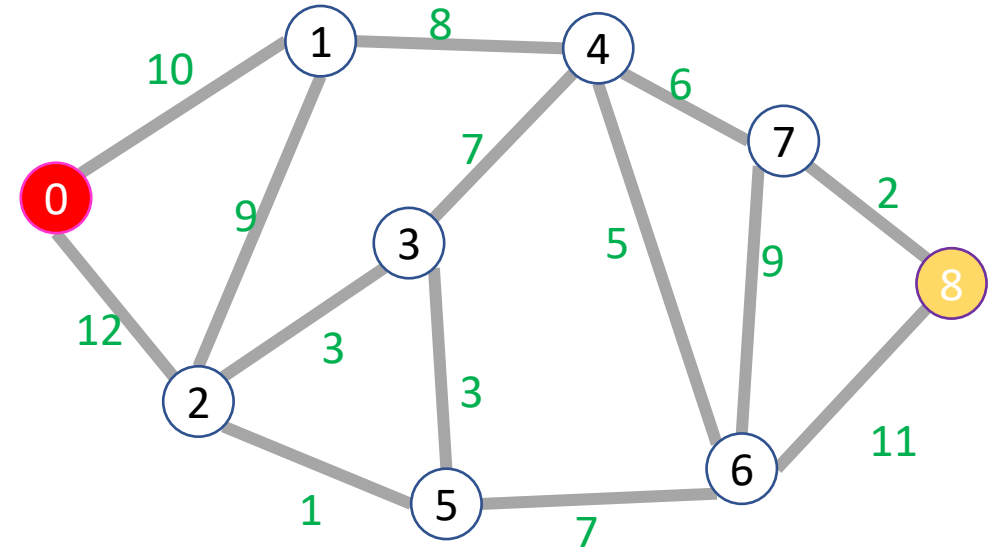
Dijkstra's Algorithm

```
int dijkstras(graph, start, end){
    distances = [ $\infty$ ,  $\infty$ ,  $\infty$ ,...]; // one index per node
    done = [False,False,False,...]; // one index per node
    PQ = new minheap();
    PQ.insert(0, start); // priority=0, value=start
    distances[start] = 0;
    while (!PQ.isEmpty){
        current = PQ.deleteMin();
        done[current] = true;
        for (neighbor : current.neighbors){
            if (!done[neighbor]){
                new_dist = distances[current]+weight(current,neighbor);
                if(distances[neighbor] ==  $\infty$ ){
                    distances[neighbor] = new_dist;
                    PQ.insert(new_dist, neighbor);
                }
                if (new_dist < distances[neighbor]){
                    distances[neighbor] = new_dist;
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            }
        }
    }
    return distances[end]
}
```



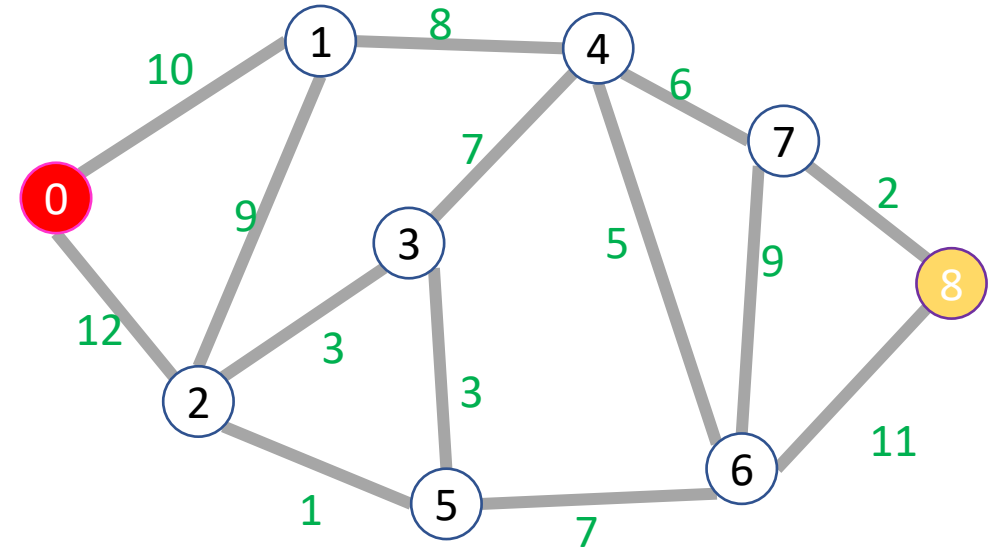
Prim's Algorithm

```
int primss(graph, start, end){
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```



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                }
                if (new_dist < distances[neighbor]){
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                    PQ.decreaseKey(new_dist,neighbor); }
            }
        }
    }
    return distances[end]
}
```



Prim's Algorithm

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            }
        }
    }
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```

