

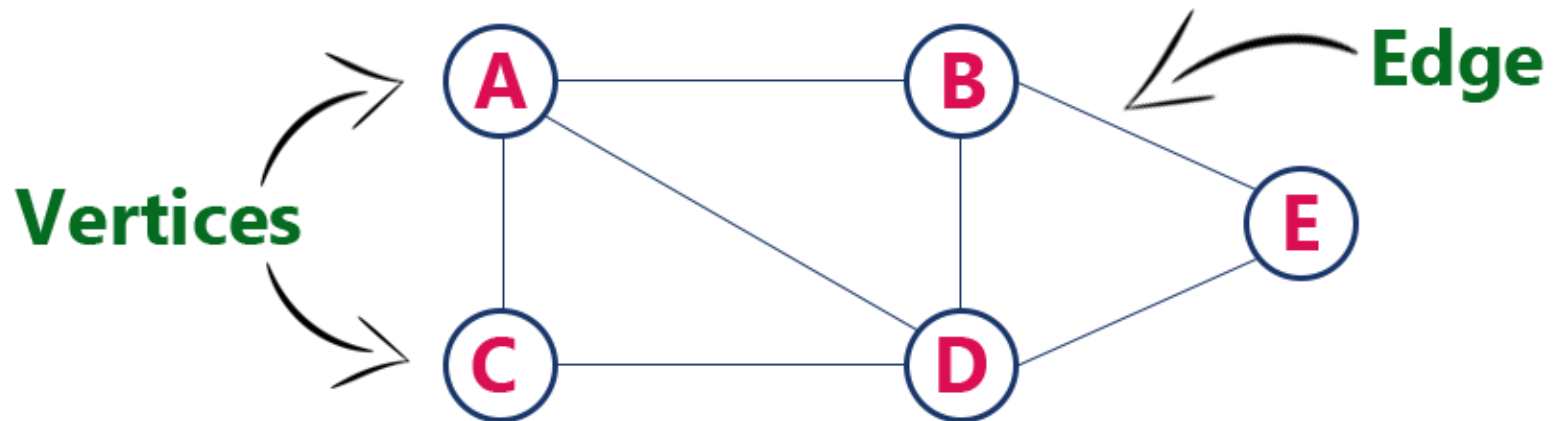
CS141: Intermediate Data Structures and Algorithms

Graphs

Amr Magdy

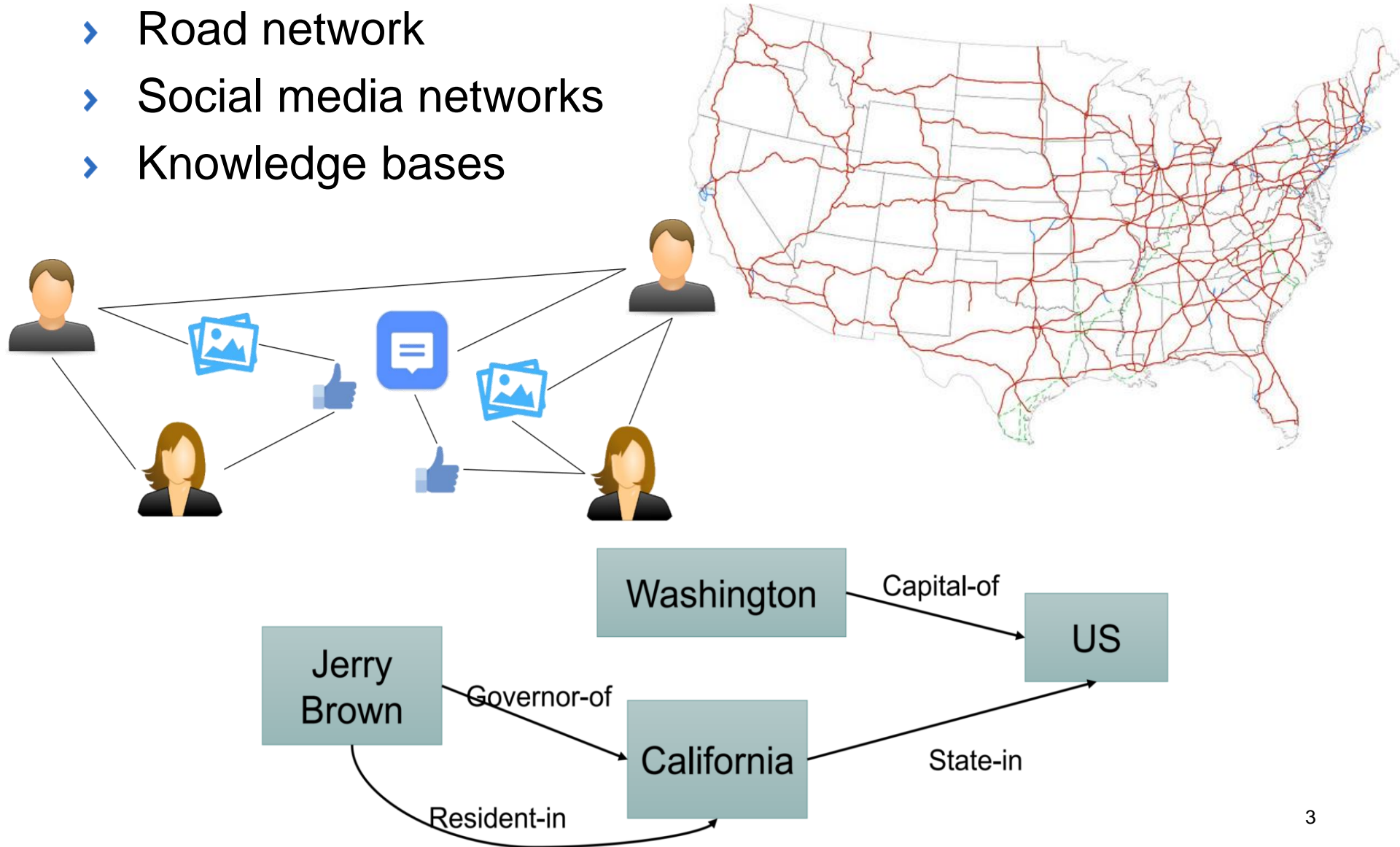
Graph Data Structure

- › A set of nodes (vertices) and edges connecting them



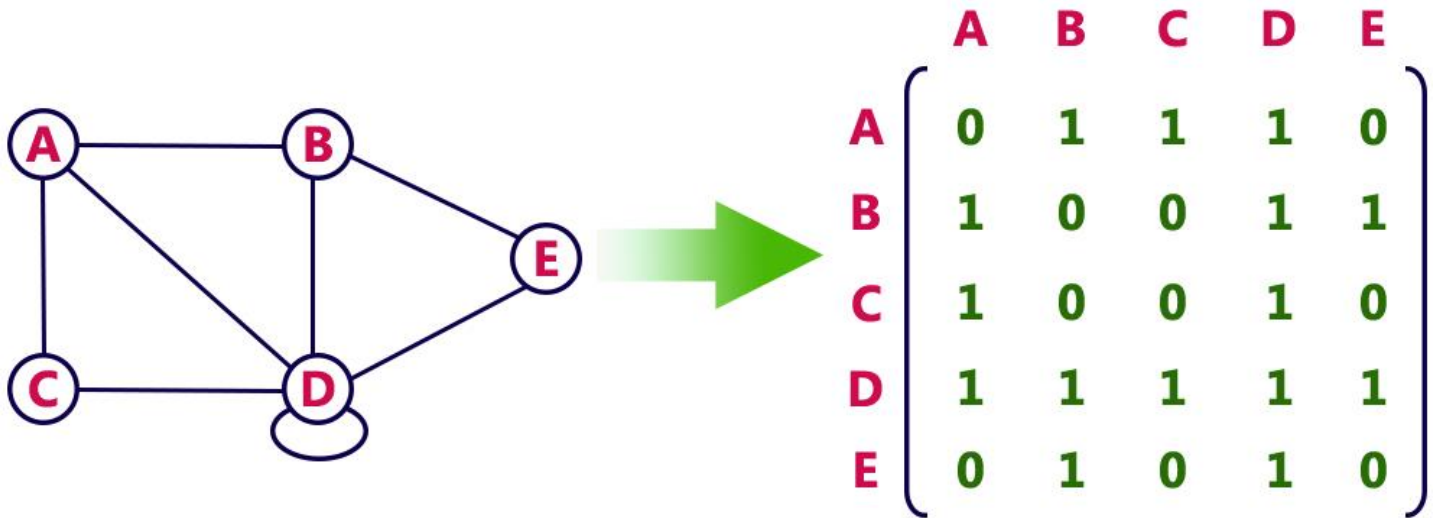
Graph Applications

- › Road network
- › Social media networks
- › Knowledge bases



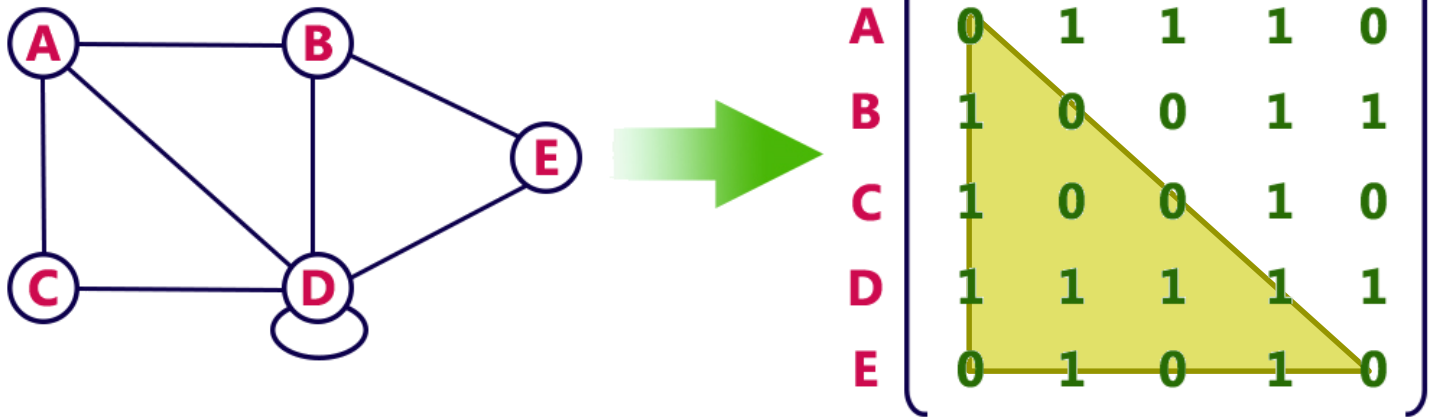
Graph Representations

- Adjacency matrix
 - Storage and access efficient when many edges exist



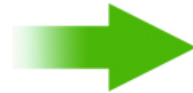
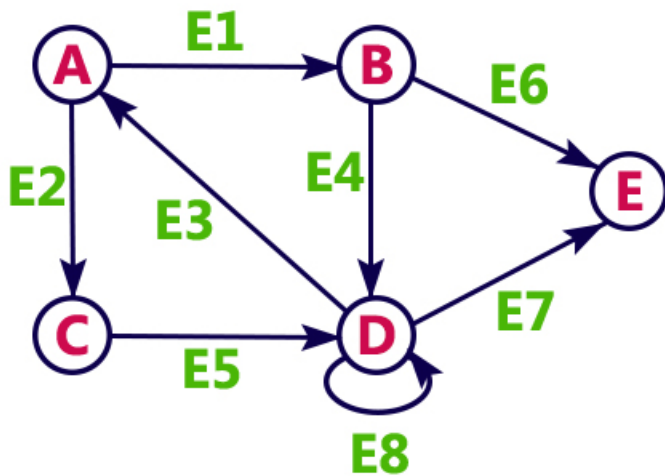
Graph Representations

- Adjacency matrix
 - Storage and access efficient when many edges exist



Graph Representations

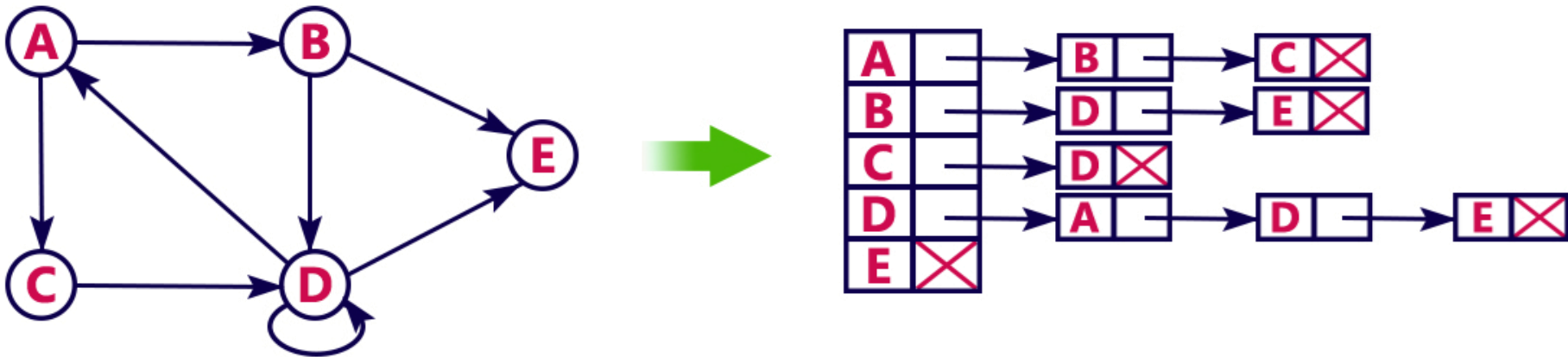
- › Incidence Matrix
 - › Expensive storage, not popular



	E1	E2	E3	E4	E5	E6	E7	E8
A	1	1	-1	0	0	0	0	0
B	-1	0	0	1	0	1	0	0
C	0	-1	0	0	1	0	0	0
D	0	0	1	-1	-1	0	1	1
E	0	0	0	0	0	-1	-1	0

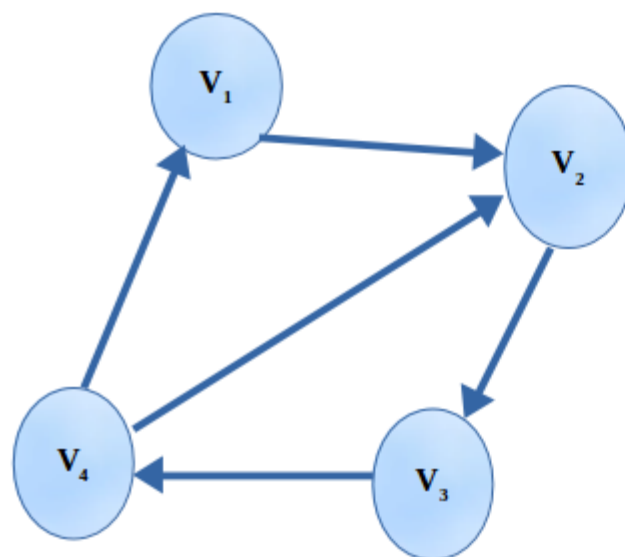
Graph Representations

- › Adjacency list
 - › Storage efficient when few edges exit (sparse graphs)
 - › Sequential access to edges (vs random access in matrix)

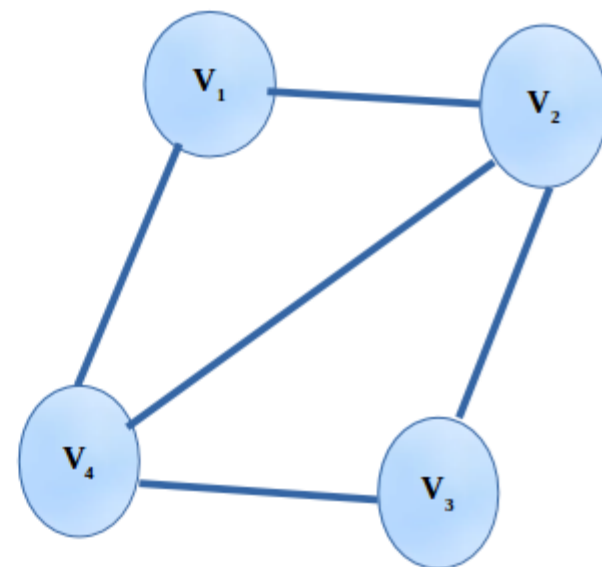


Types of Graphs

- › Directed and Undirected graphs
- › Weighted and Unweighted graphs
- › Connected graphs
- › Bipartite graphs
- › Acyclic graphs
- › Tree/Forest



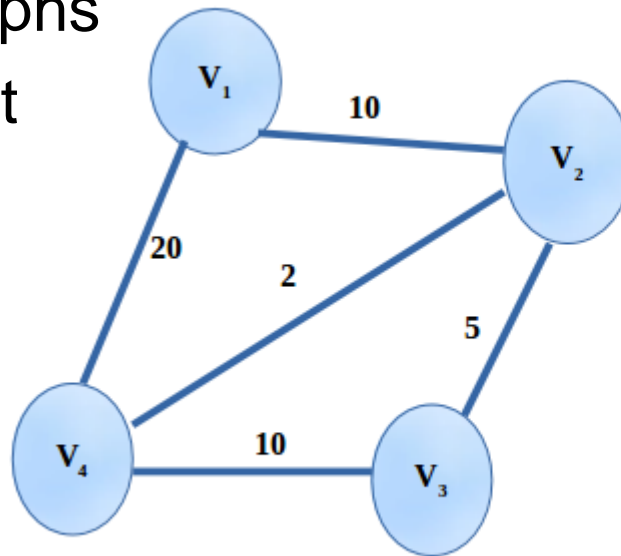
Directed Graph



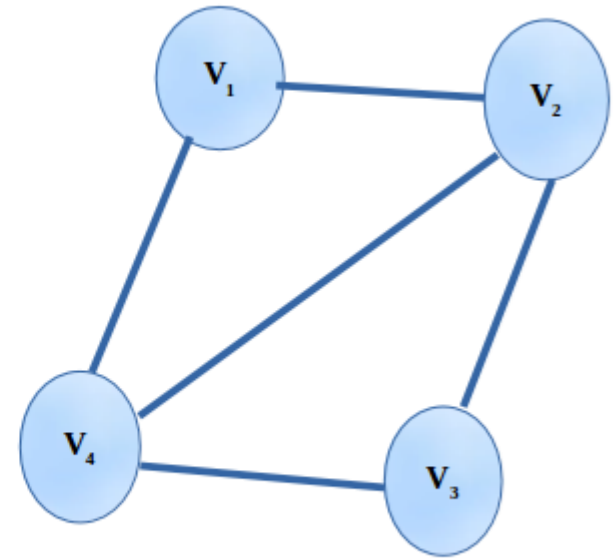
Undirected Graph

Types of Graphs

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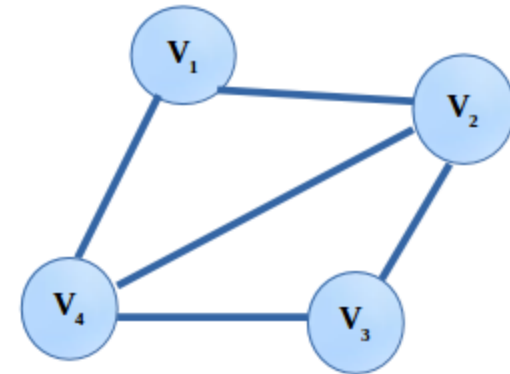
Weighted Graph



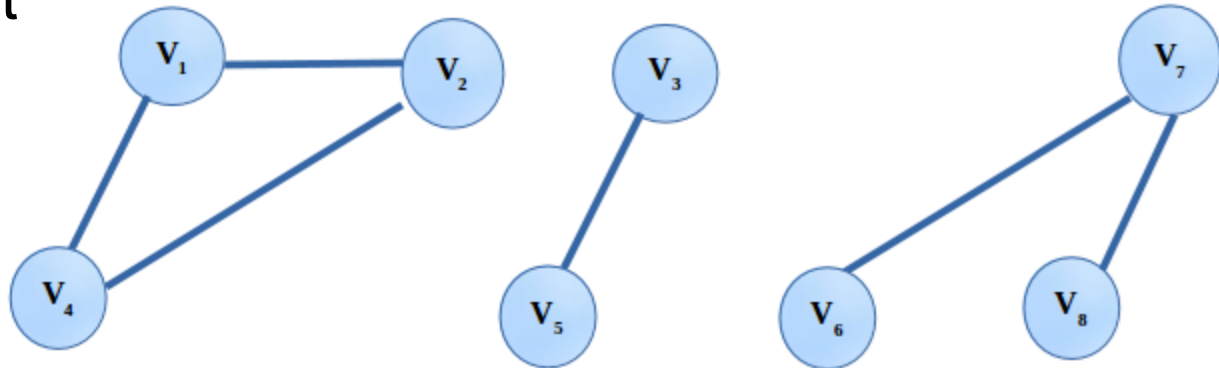
Unweighted Graph

Types of Graphs

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Fig(i):
Connected
Graph

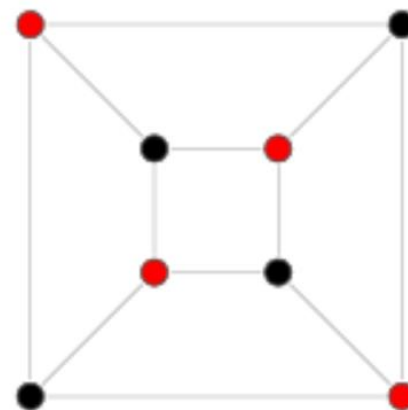
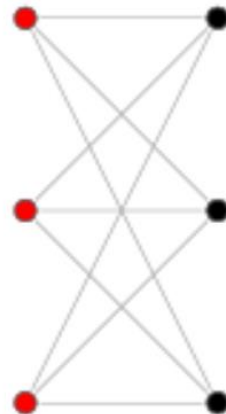
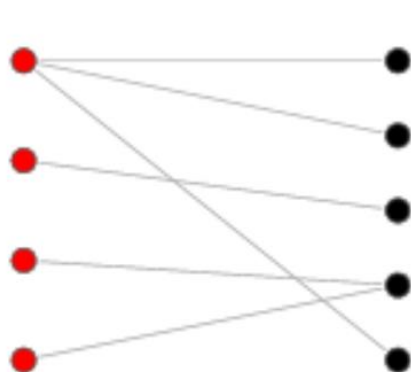


Fig(ii):
Unconnected Graph

There are three component of above unconnected graph

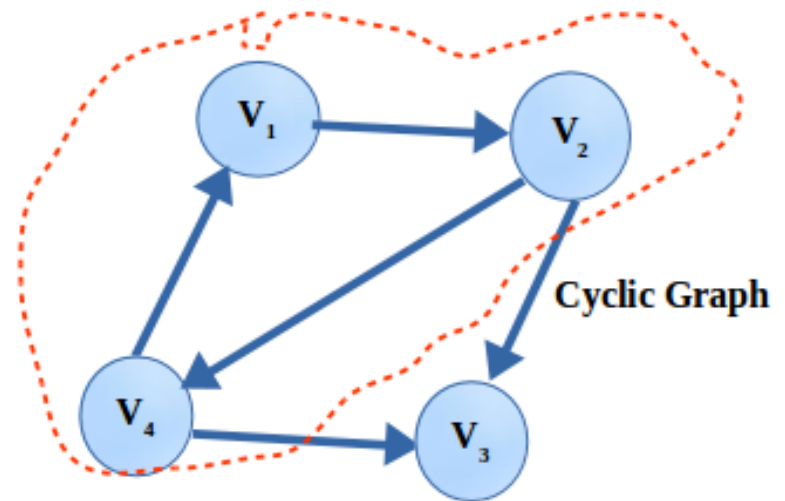
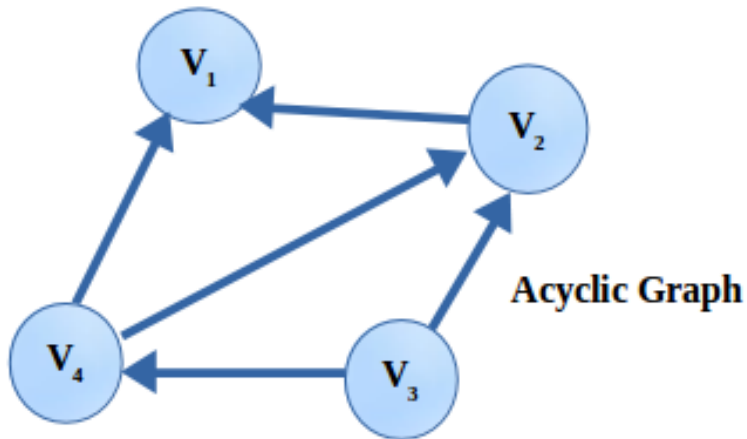
Types of Graphs

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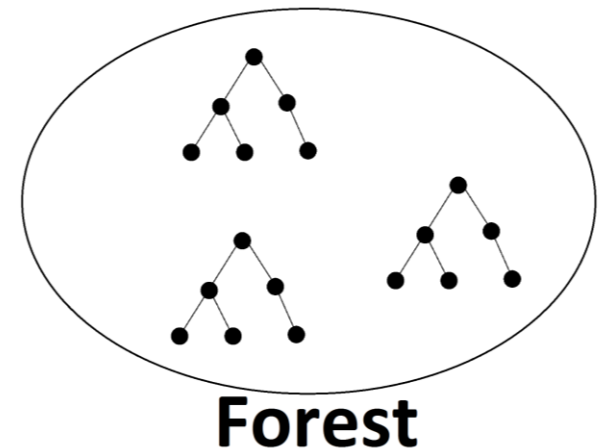
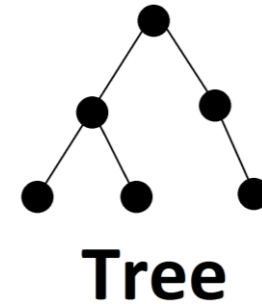
Types of Graphs

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Types of Graphs

- › Directed and Undirected graphs
- › Weighted and Unweighted graphs
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- › Acyclic graphs
- › Tree/Forest
 - › Tree: directed acyclic graph with max of one path between any two nodes
 - › Forest: set of disjoint trees

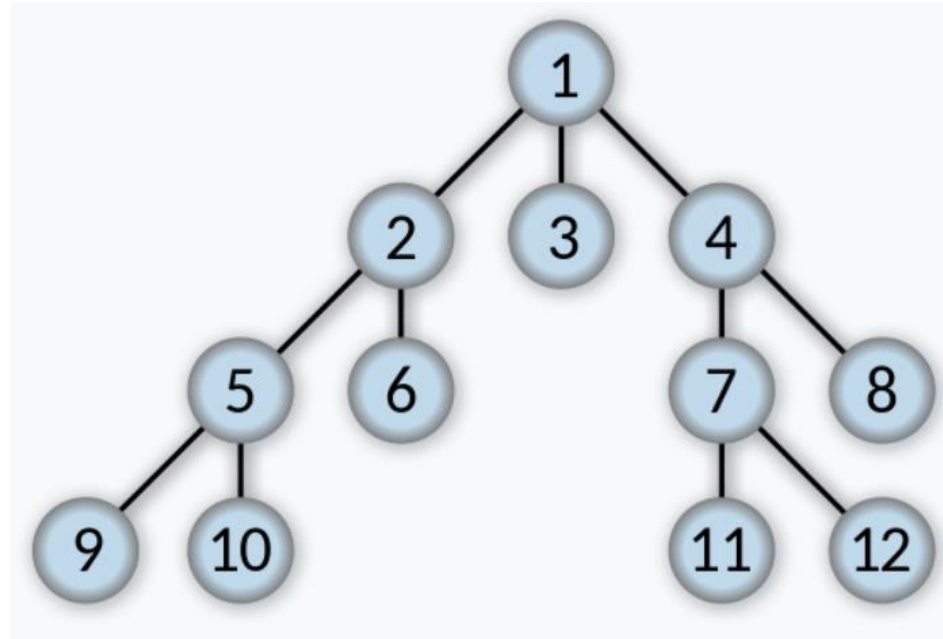


Basic Graph Algorithms

- › Graph traversal algorithms
 - › Bread-first Search (BFS)
 - › Depth-first Search (DFS)
- › Topological Sort
- › Graph Connectivity
- › Cycle Detection

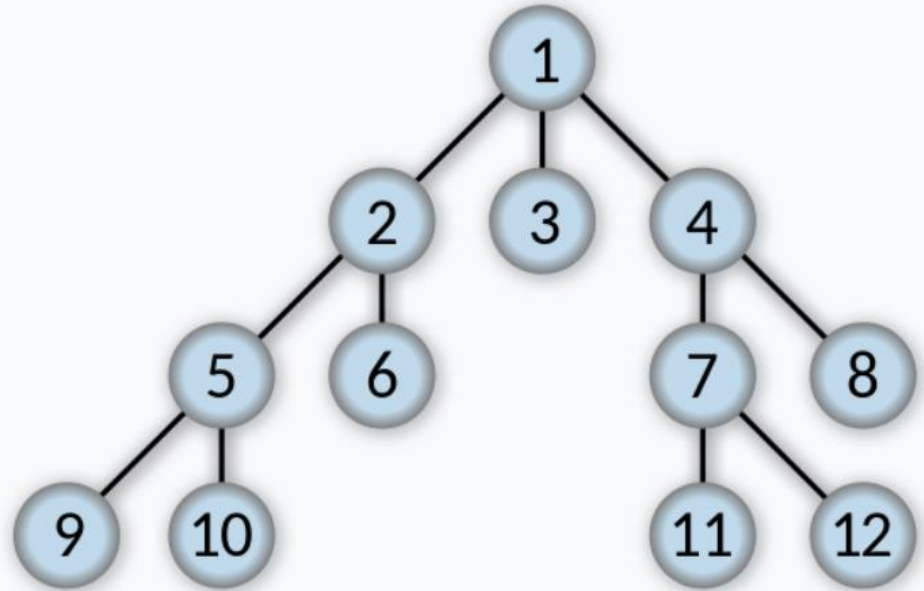
Breadth-first Search (BFS)

- How to traverse?



Breadth-first Search (BFS)

- › How to traverse?
- › Use a queue



Breadth-first Search (BFS)

- › How to traverse?
- › Use a queue
- › Start at a vertex s

Mark s as visited

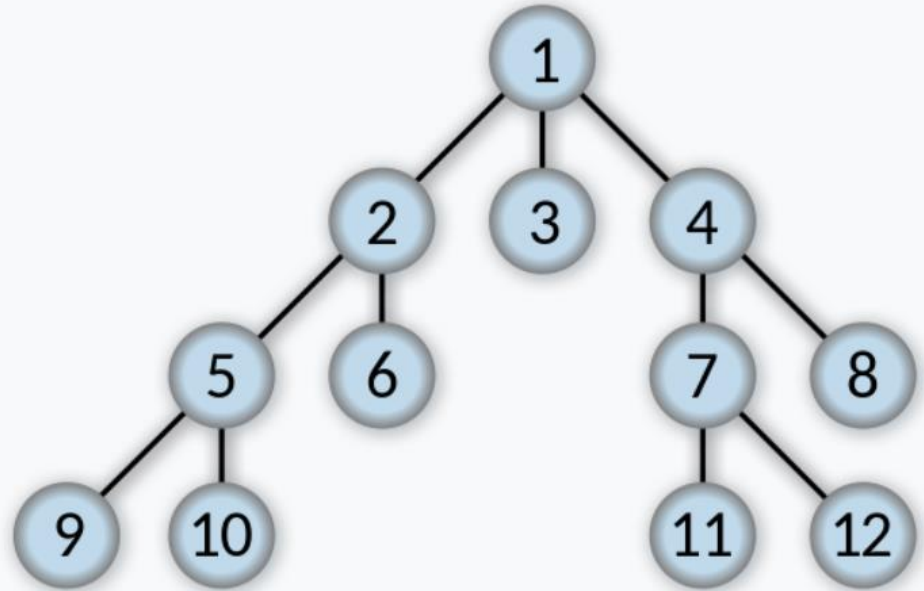
Enqueue neighbors of s

while Q not empty

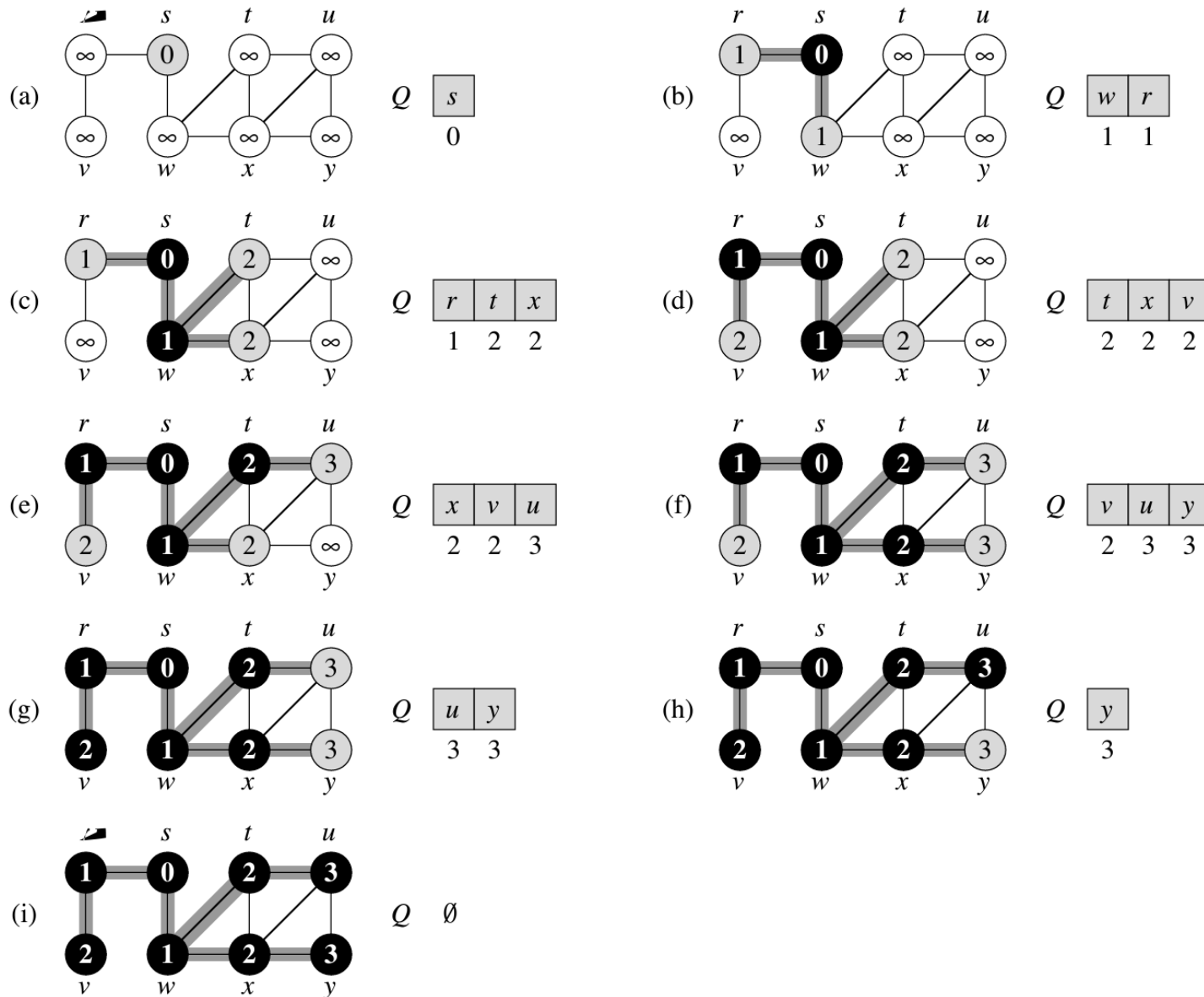
 Dequeue vertex u

 Mark u as visited

 Enqueue unvisited neighbors of u

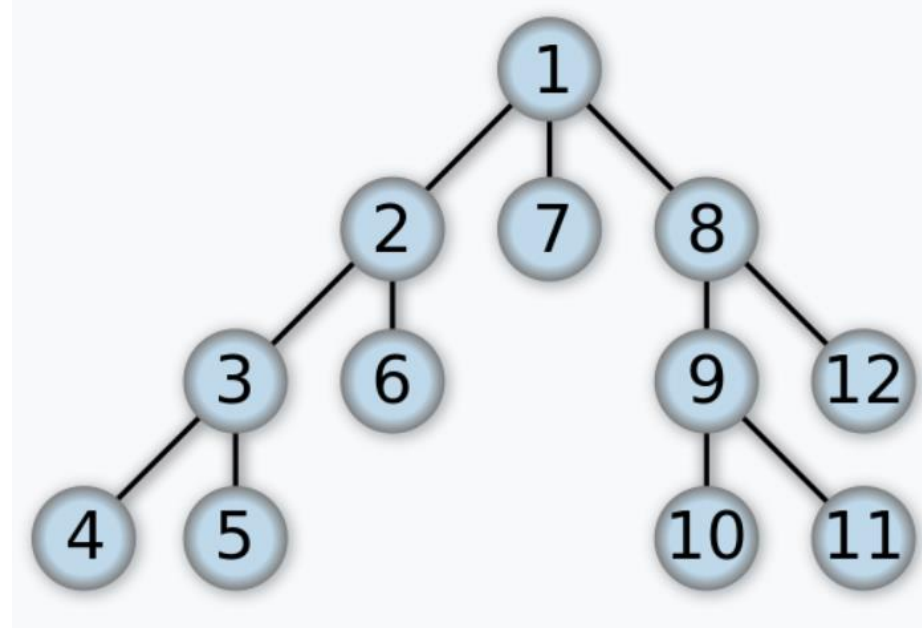


Breadth-first Search (BFS)



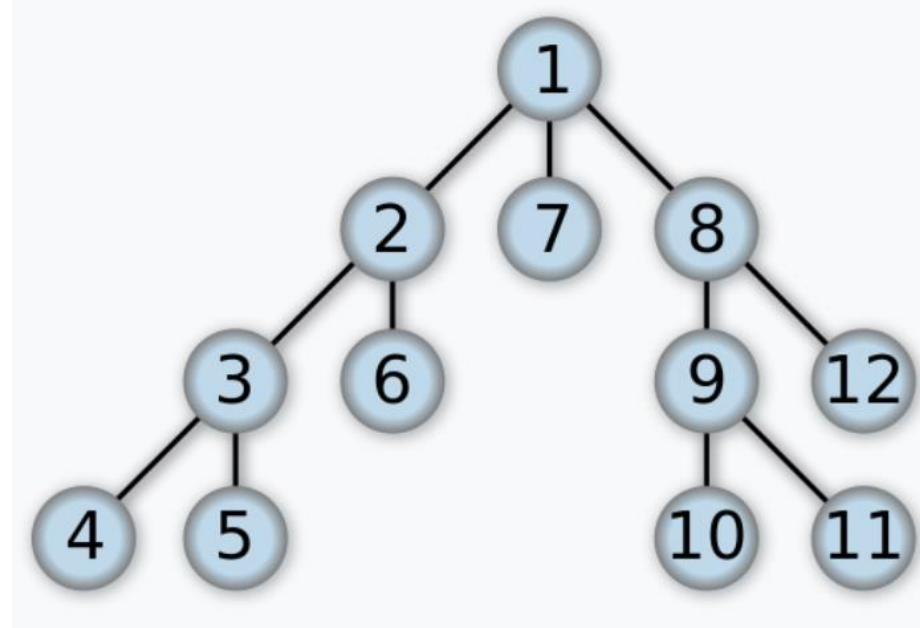
Depth-first Search (DFS)

- How to traverse?



Depth-first Search (DFS)

- › How to traverse?
- › Use a stack



Depth-first Search (DFS)

› How to traverse?

› Use a stack

› Start at a vertex s

Mark s as visited

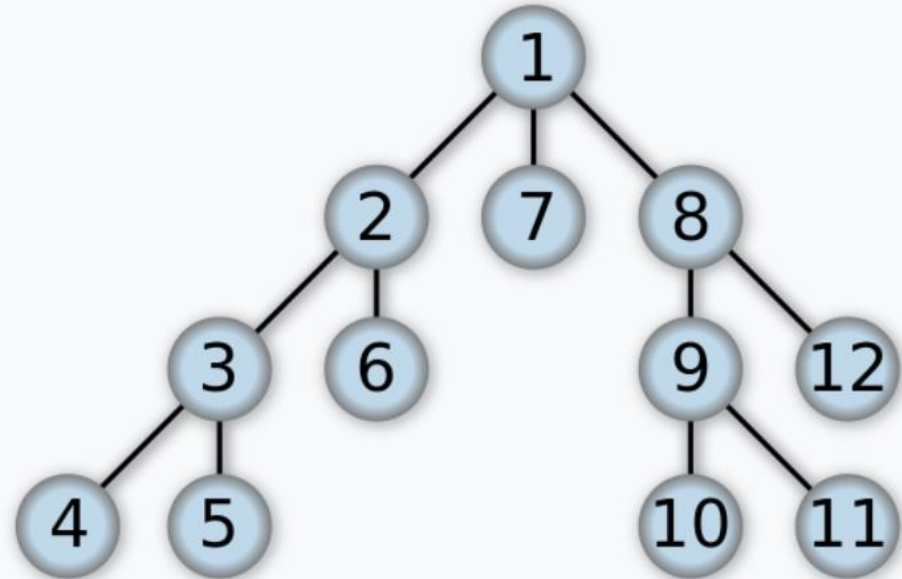
Push neighbors of s

while Stack not empty

 Pop vertex u

 Mark u as visited

 Push unvisited neighbors of u



Complexity of Graph Traversal

- For $G = (V, E)$, V set of vertices, E set of edges
- BFS
 - Time: $O(|V| + |E|)$
 - Space: $O(|V|)$ (plus graph representation)
- DFS
 - $O(|V| + |E|)$
 - Space: $O(|V|)$ (plus graph representation)

Graph Connectivity



- › Checking if graph is connected:

Graph Connectivity

- › Checking if graph is connected:

IsConnected(G)

{

 DFS(G)

 if any vertex not visited

 return false

 else

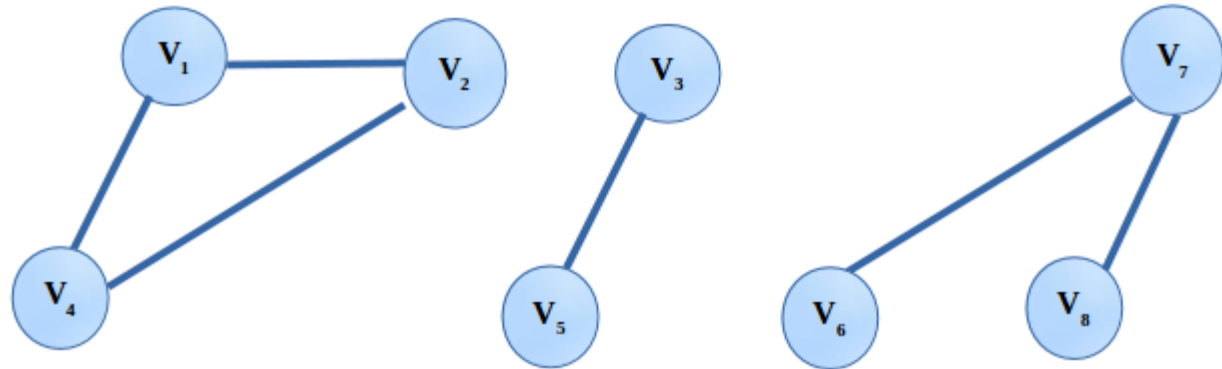
 return true

}

Time Complexity: $O(|V|+|E|)$

Graph Connected Components

- › Getting the graph connected components



Fig(ii):
Unconnected Graph

**There are three component of above
unconnected graph**

Graph Connected Components

- › Getting the graph connected components
- › Mark all nodes as unvisited

visitCycle = 1

while(there exists unvisited node n)

{

- Start DFS(G) at n, mark visited node with *visitCycle*

- Output all nodes with current *visitCycle* as one connected component

- *visitCycle* = *visitCycle*+1

}

Time Complexity: $O(|V|+|E|)$

Cycle Detection



- › Does a connected graph G contain a cycle?
(non-trivial cycle)

Cycle Detection

- › Does a connected graph G contain a cycle?
(non-trivial cycle)
- › General idea: if DFS procedure tries to revisit a visited node, then there is a cycle

Cycle Detection



- › Does a graph G contain a cycle? (non-trivial cycle)

IsAcyclic(G) {

 Start at unvisited vertex s

 Mark “ s ” as visited

 Push neighbors u of s in stack $\langle \text{node}:u, \text{parent}:s \rangle$

 while stack not empty

 Pop vertex u

 Mark u as visited

 if u has a visited neighbor v

 & v is non-parent for u

 return true

 Push unvisited neighbors v of u $\langle \text{node}:v, \text{parent}:u \rangle$

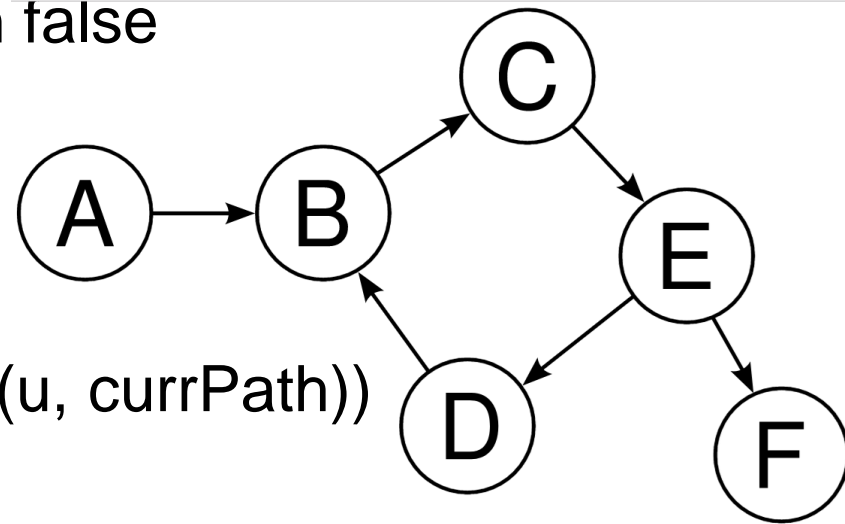
return false

Cycle Detection

- › Does a connected graph G contain a cycle?
(non-trivial cycle)
- › General idea: if DFS procedure tries to revisit a visited node, then there is a cycle
- › Why checking if v non-parent for u ?
 - › To eliminate trivial cycles, a cycle that involve only two nodes

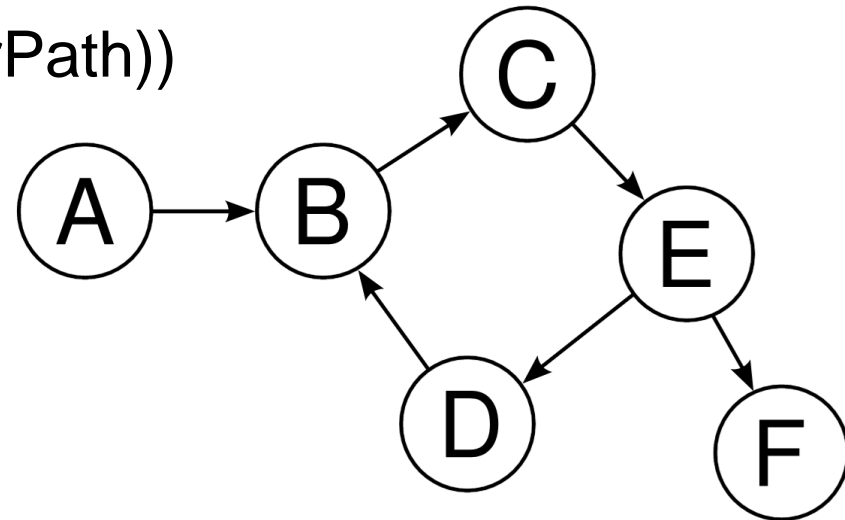
Cycle Detection in Directed Graphs

```
IsAcyclicDirected(node s, currPath) {  
    if s in currPath      return true  
    if s is visited       return false  
    Mark s as visited  
    Add s to currPath  
    for each neighbor u of s  
        if(IsAcyclicDirected(u, currPath))  
            return true  
    remove s from currPath  
    return false  
}
```



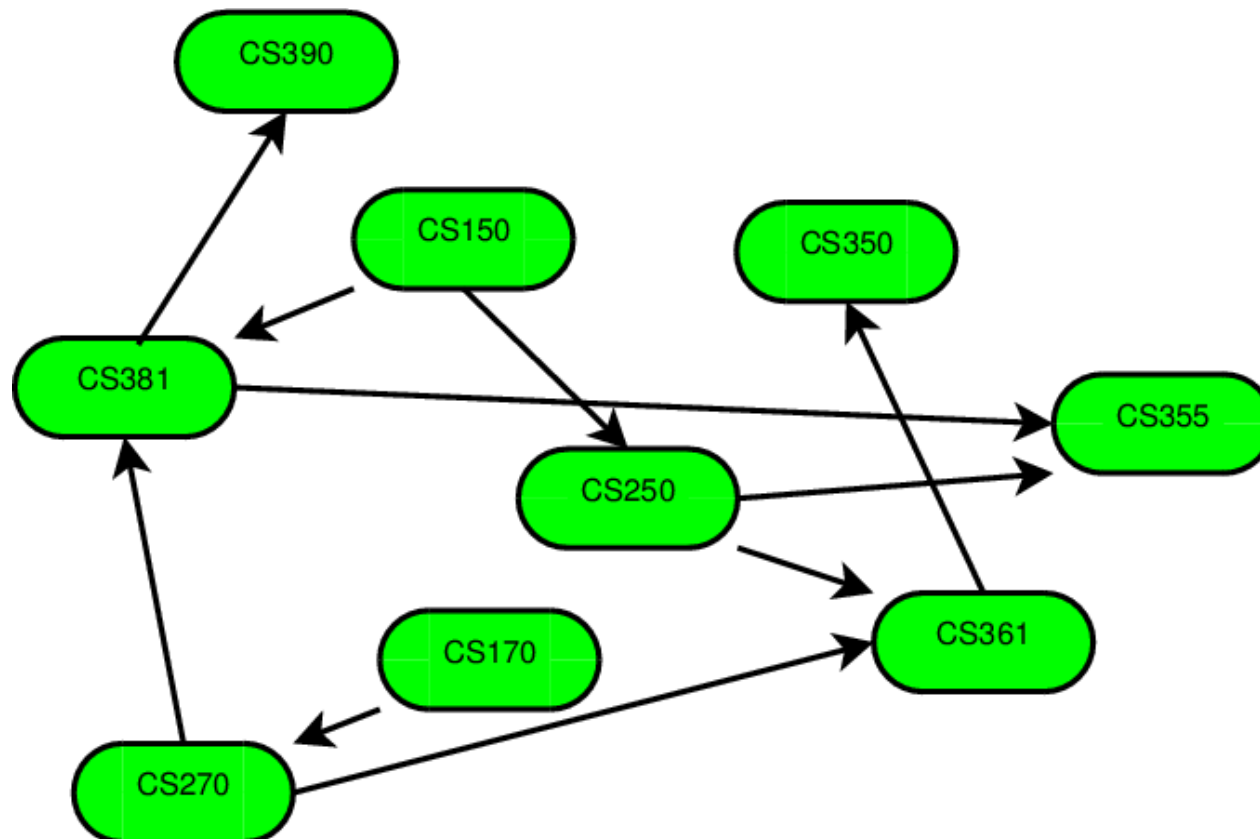
Cycle Detection in Directed Graphs

```
while(there is unvisited node s)
{
    currPath = {}
    if(IsAcyclicDirected(s, currPath))
        return true
}
return false
```



Topological Sort

- Determine a linear order for vertices of a directed acyclic graph (DAG)
 - Mostly dependency/precedence graphs
 - If edge (u,v) exists, then u appears before v in the order



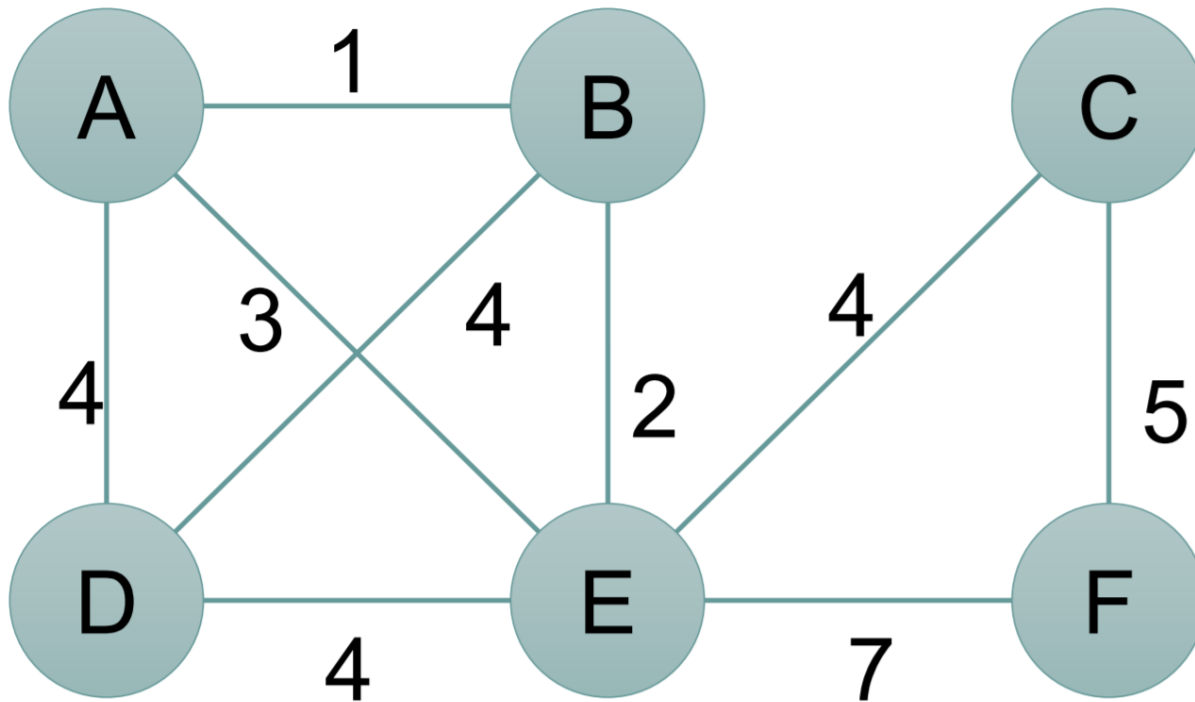
Topological Sort

```
L ← Empty list
S ← Set of all nodes with no incoming edge
while S is non-empty do
    remove a node n from S
    add n to end of L
    for each node m with an edge e from n to m do
        remove edge e from the graph
        if m has no other incoming edges then
            insert m into S
return L (a topologically sorted order)
```

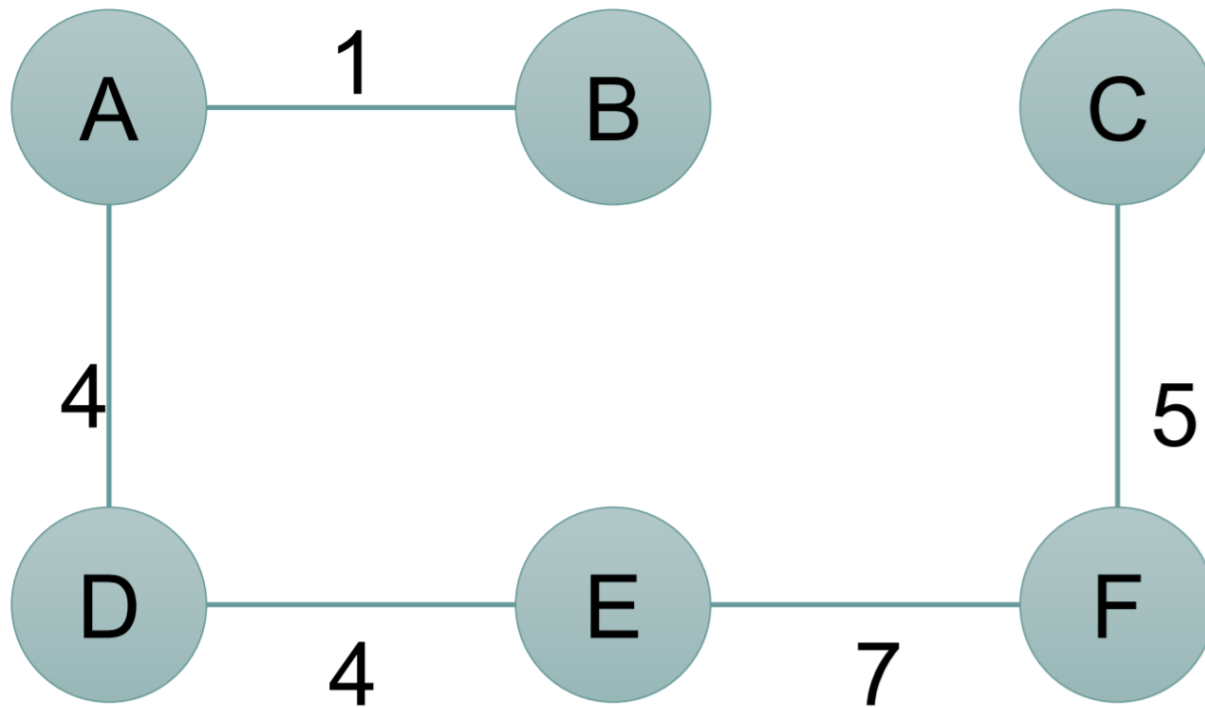
Spanning Tree

- ▶ Given a connected graph $G=(V,E)$, a spanning tree $T \subseteq E$ is a set of edges that “spans” (i.e., connects) all vertices in V .
- ▶ A **Minimum Spanning Tree (MST)**: a spanning tree with minimum total weight on edges of T
- ▶ Application:
 - ▶ The wiring problem in hardware circuit design

Spanning Tree: Example

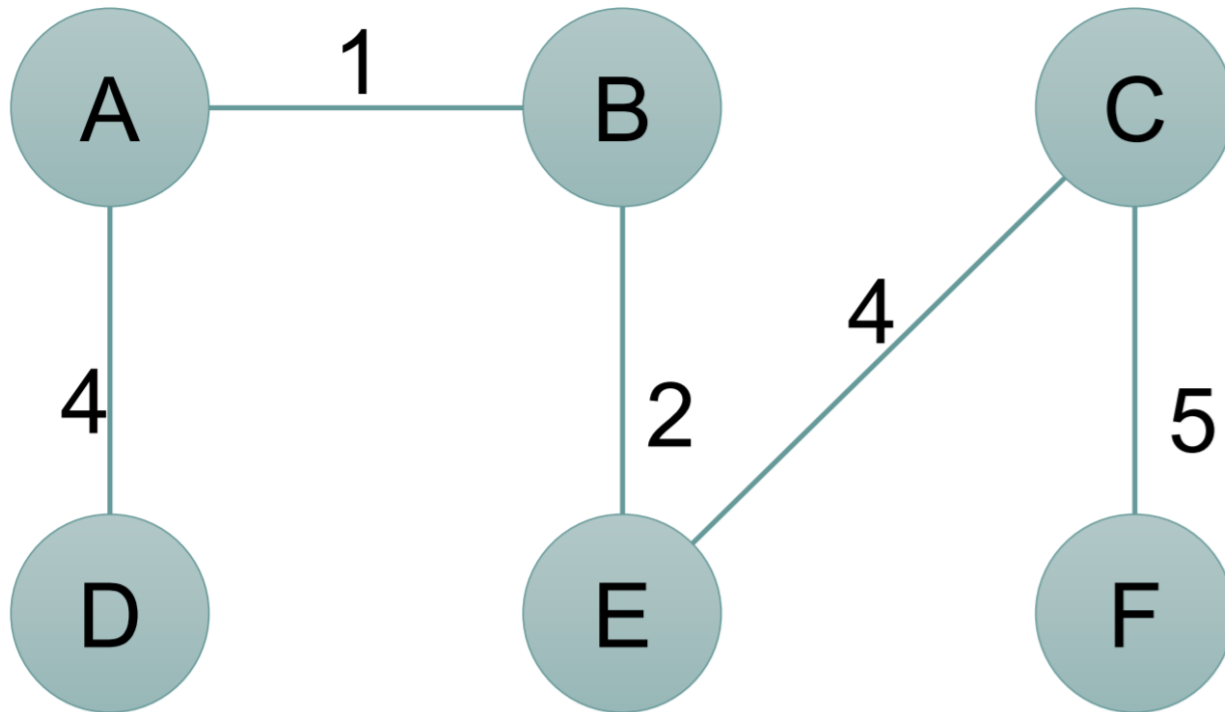


Spanning Tree: Not MST



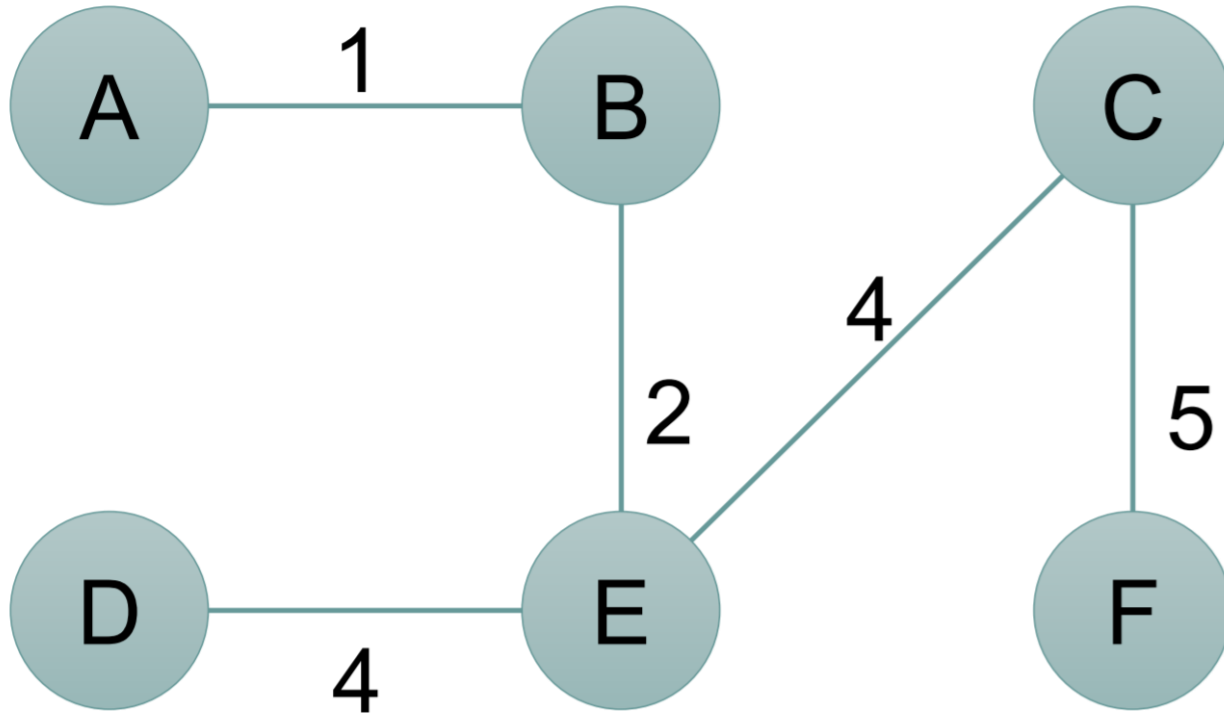
Total weight = 21

Spanning Tree: MST



Total weight = 16

Spanning Tree: Another MST

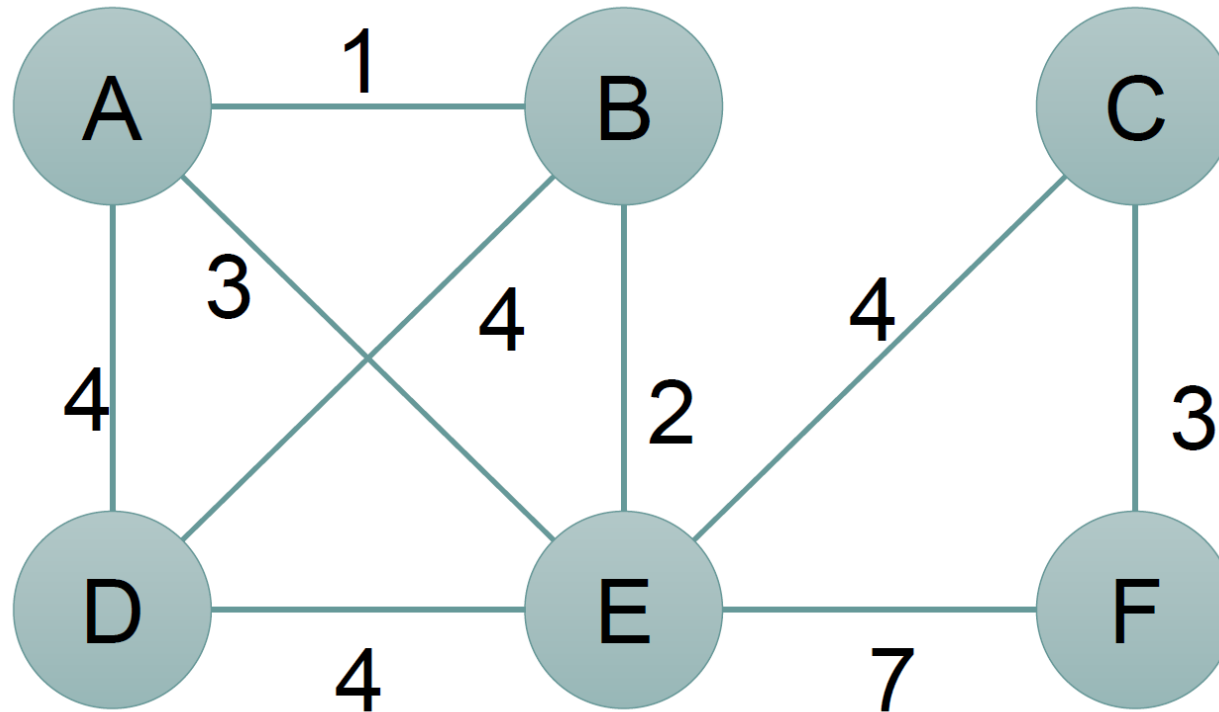


Total weight = 16

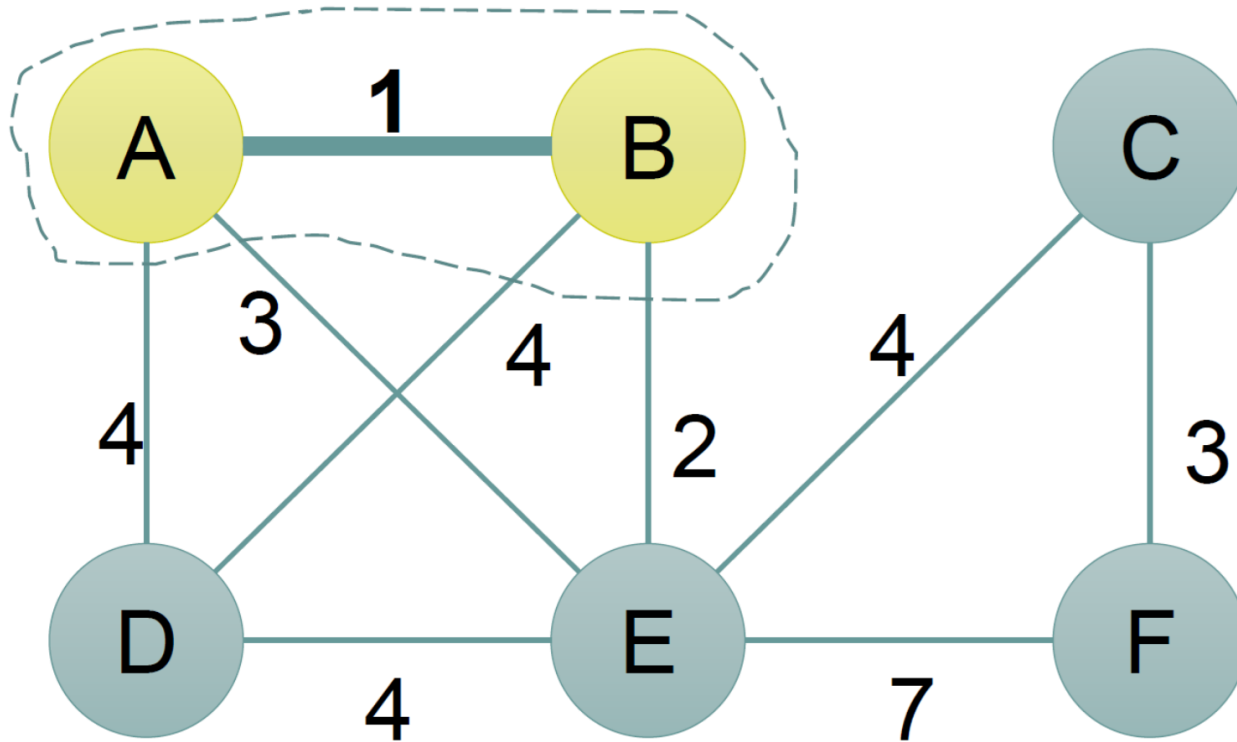
Finding MST: Kruskal's algorithm

- › Sort all the edges by weight
- › Scan the edges by weight from lowest to highest
- › If an edge introduces a cycle, drop it
- › If an edge does not introduce a cycle, pick it
- › Terminate when $n-1$ edges are picked
(n : number of vertices)

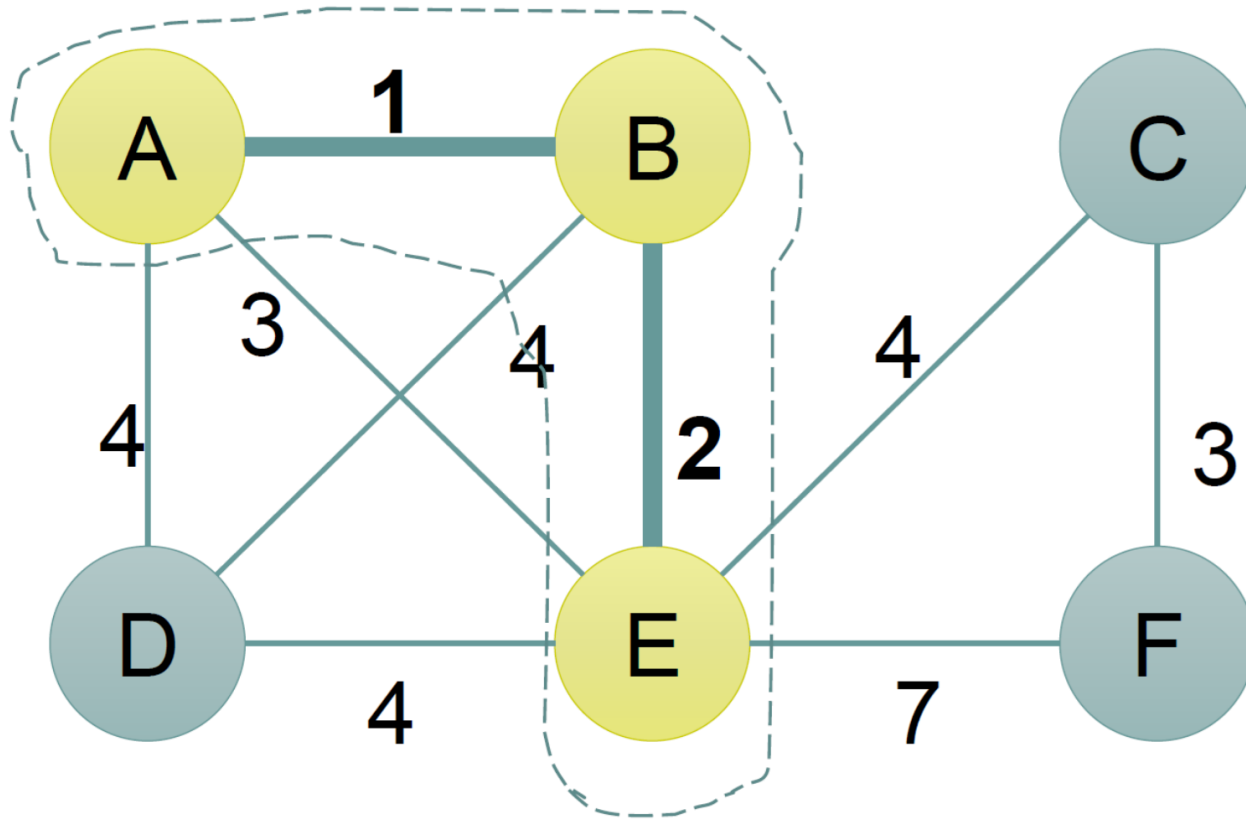
Finding MST: Kruskal's algorithm



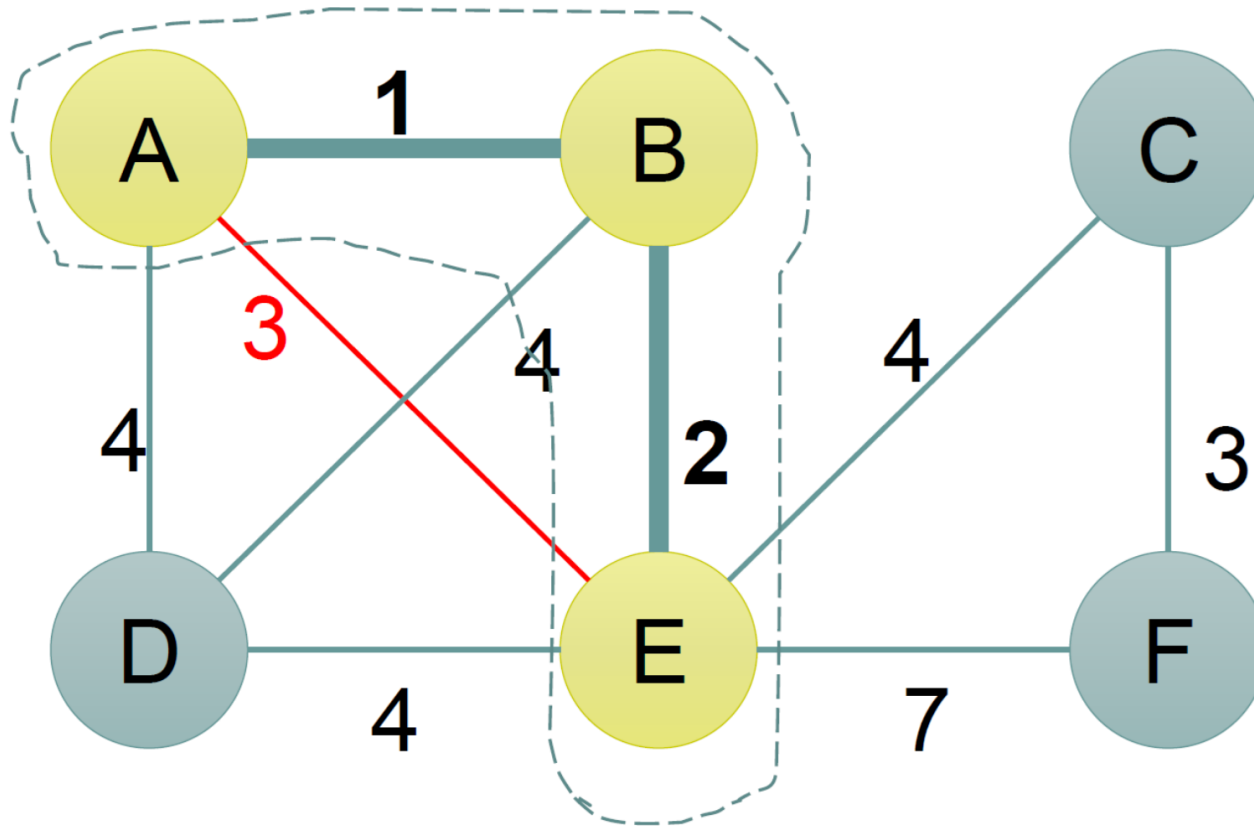
Finding MST: Kruskal's algorithm



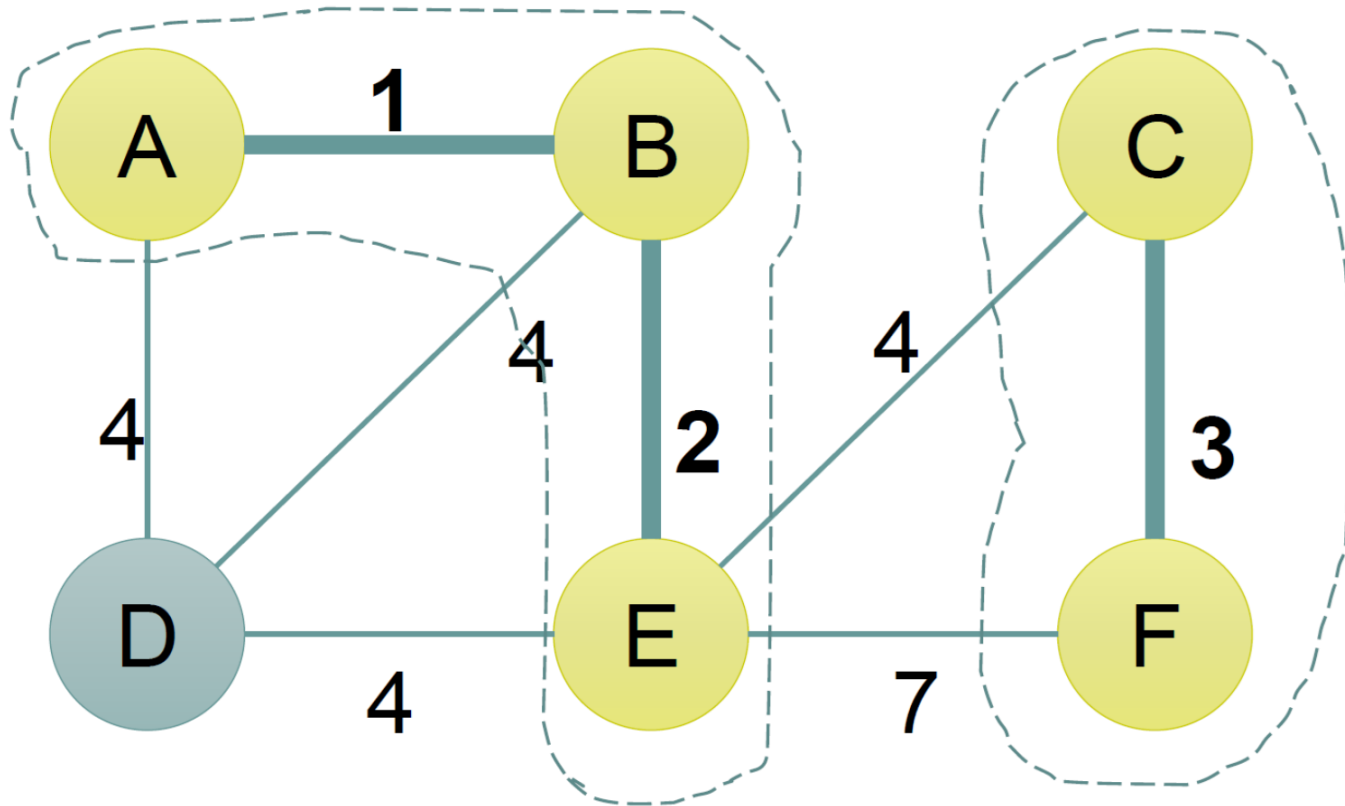
Finding MST: Kruskal's algorithm



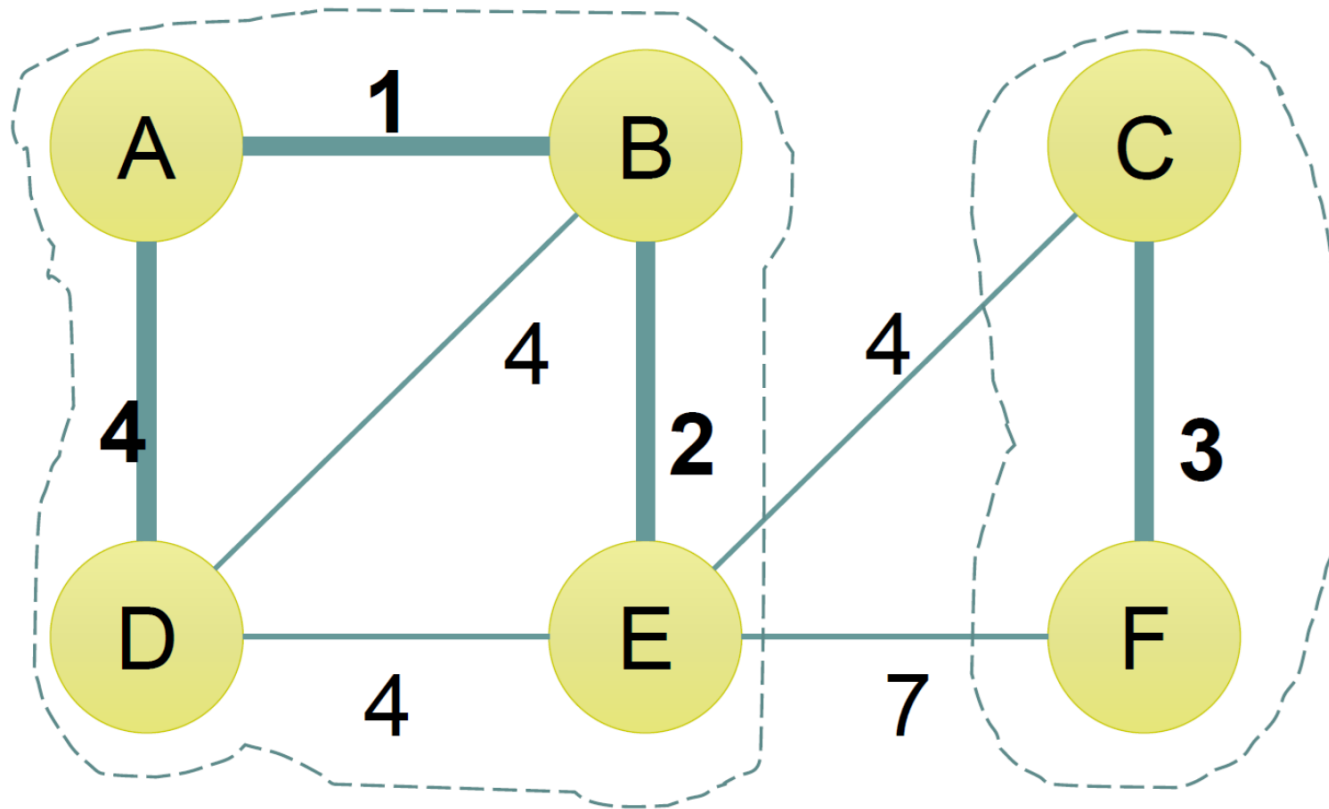
Finding MST: Kruskal's algorithm



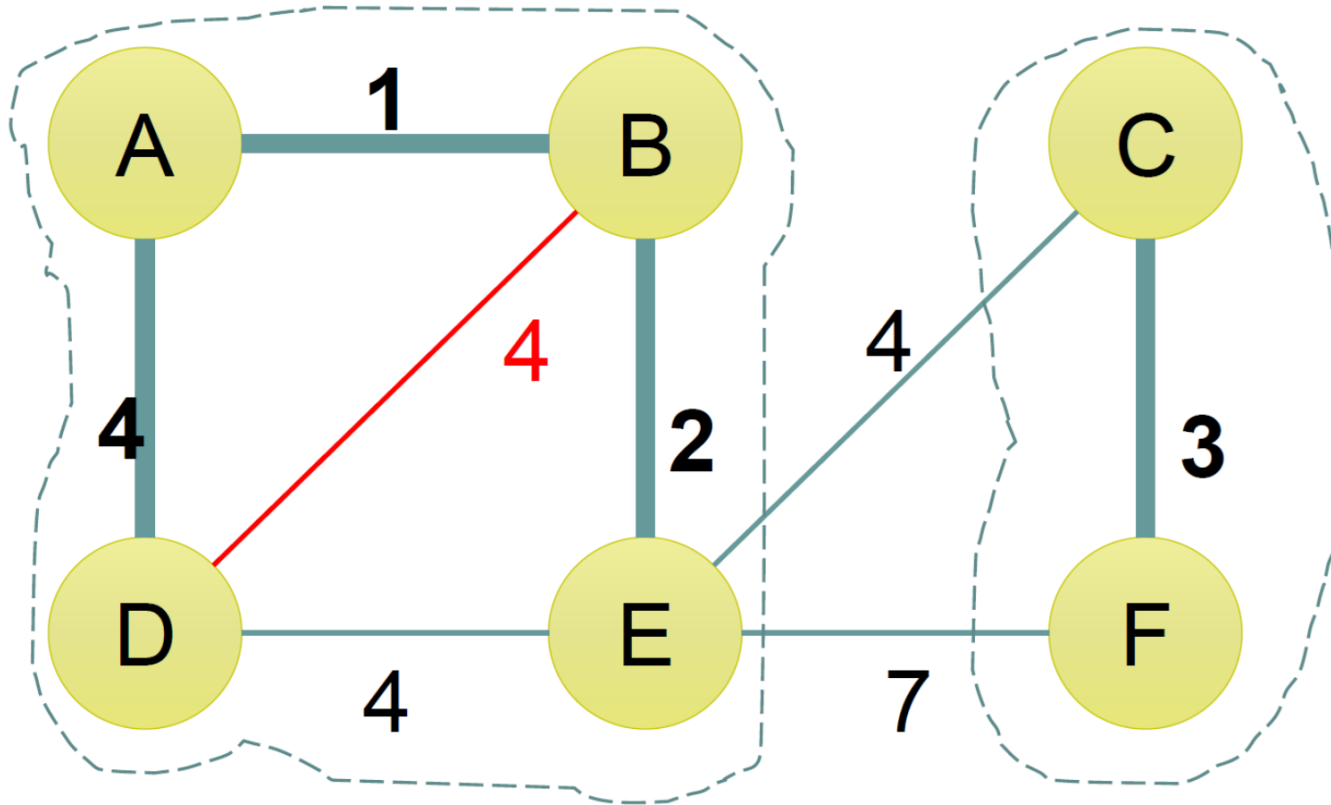
Finding MST: Kruskal's algorithm



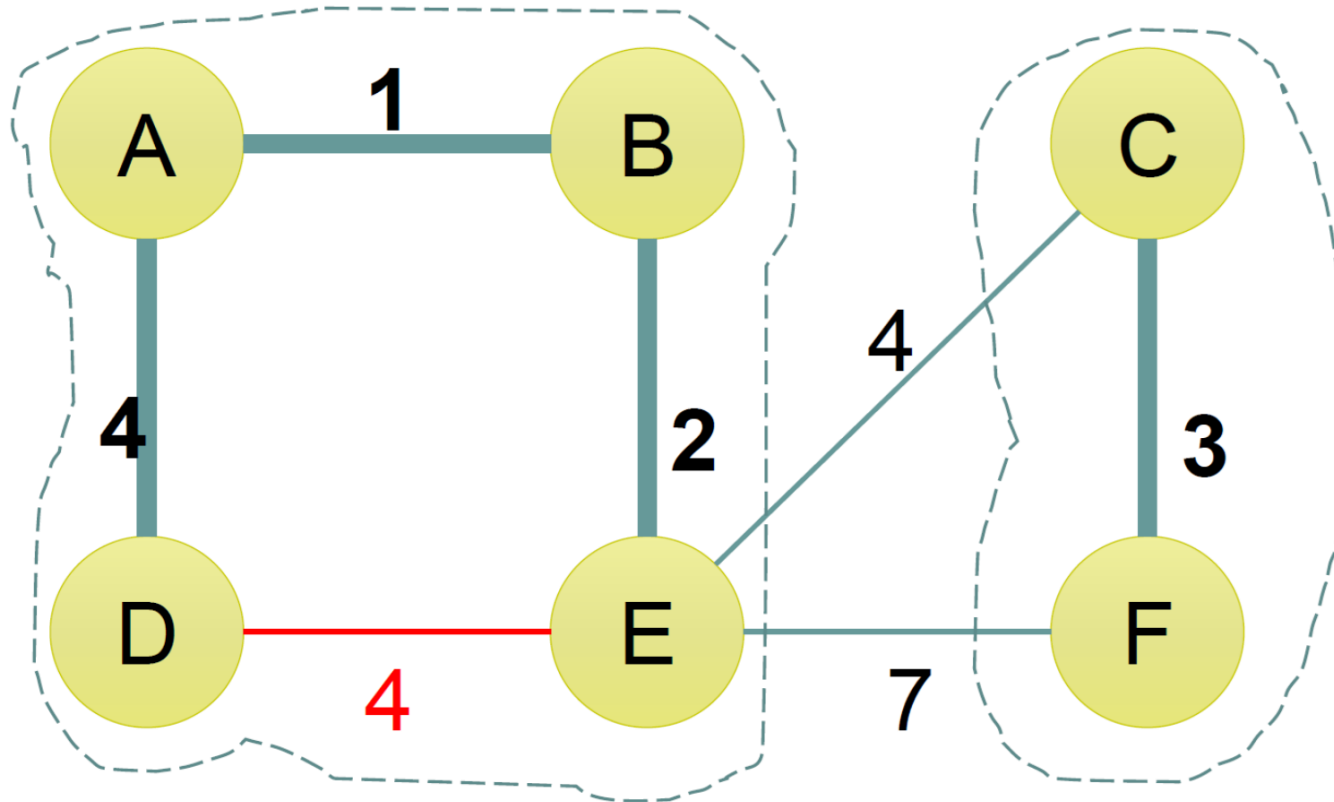
Finding MST: Kruskal's algorithm



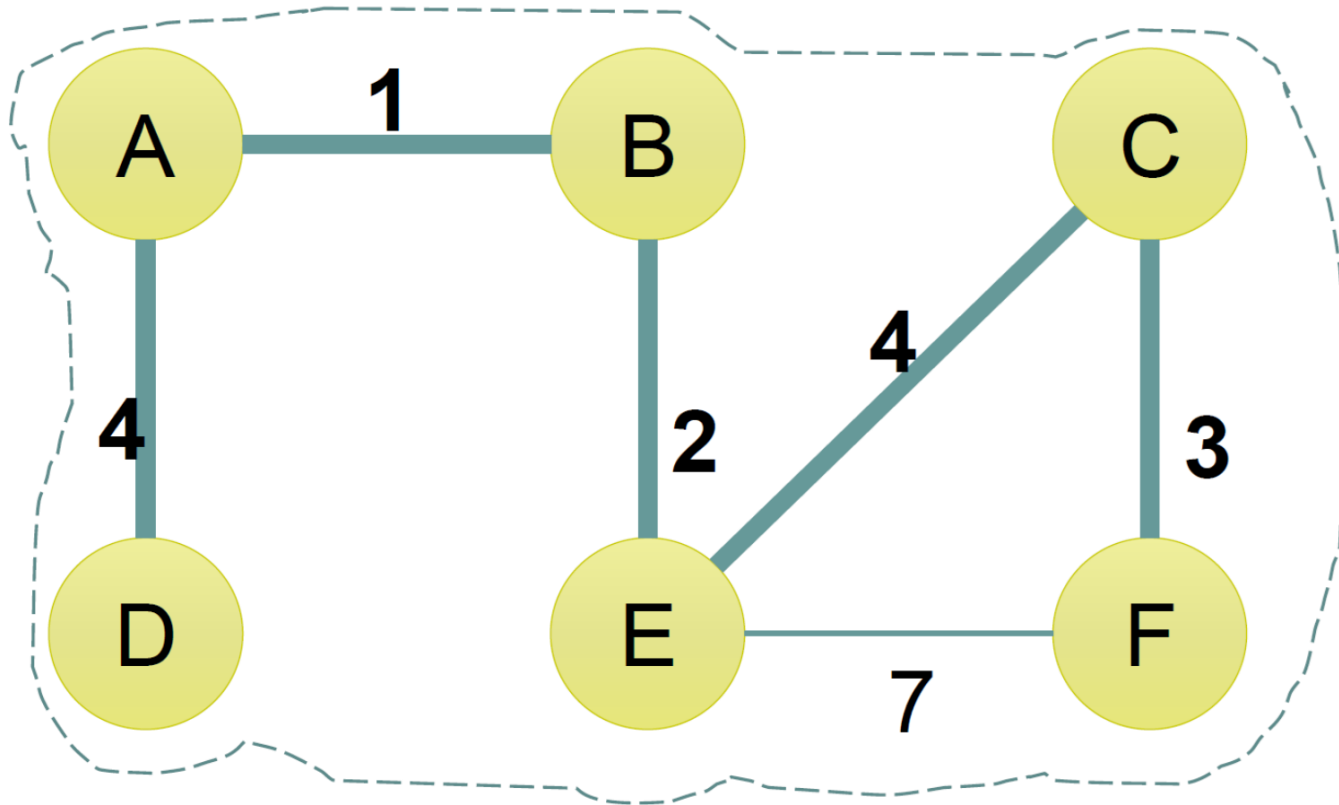
Finding MST: Kruskal's algorithm



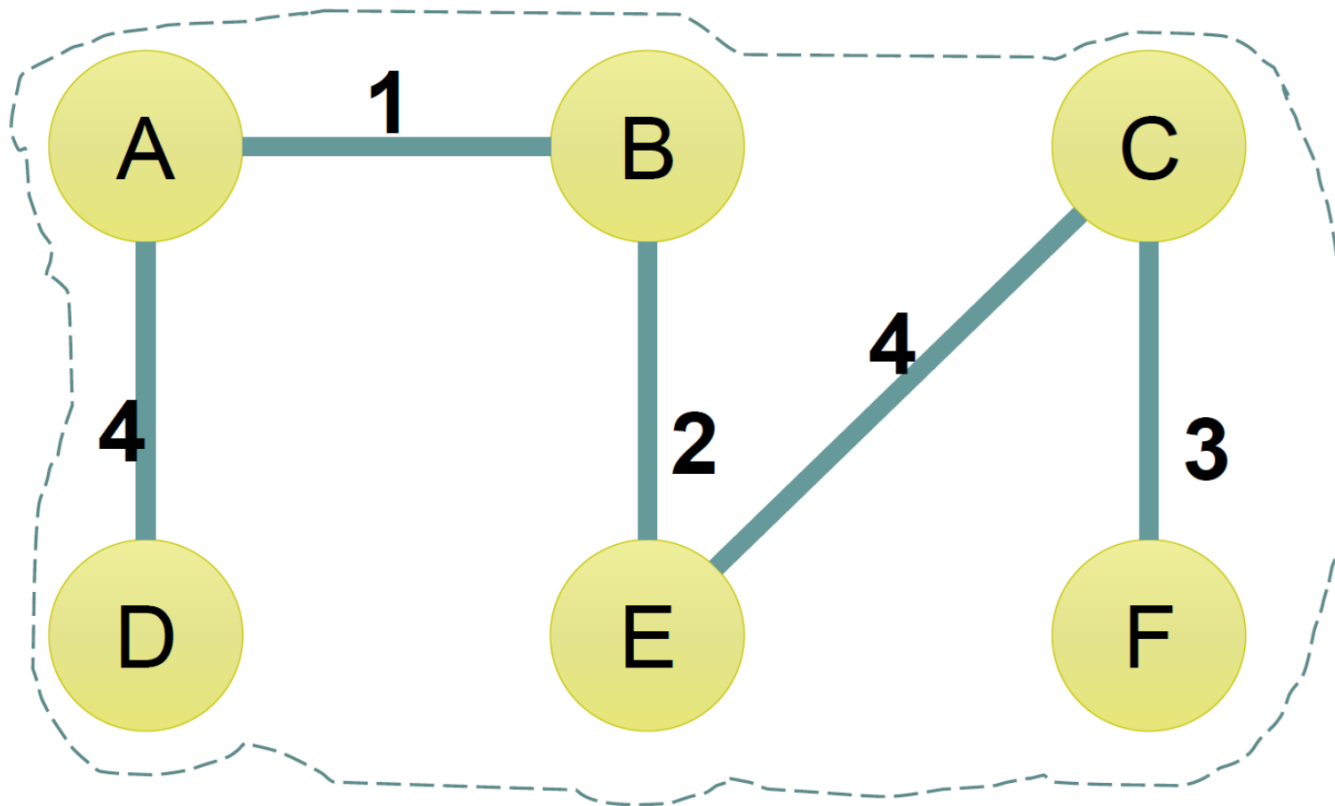
Finding MST: Kruskal's algorithm



Finding MST: Kruskal's algorithm



Finding MST: Kruskal's algorithm



Finding MST

- Kruskal's algorithm: greedy
 - Greedy choice: least weighted edge first
 - Complexity: $O(E \log E)$ – sorting edges by weight
 - Edge-cycle detection: $O(1)$ using hashing of $O(V)$ space

- Prim's algorithm: greedy
 - Complexity: $O(E + V \log V)$ – using Fibonacci heap data structure

Shortest Paths in Graphs



- Given graph $G=(V,E)$, find shortest paths from a given node *source* to all nodes in V . (Single-source All Destinations)

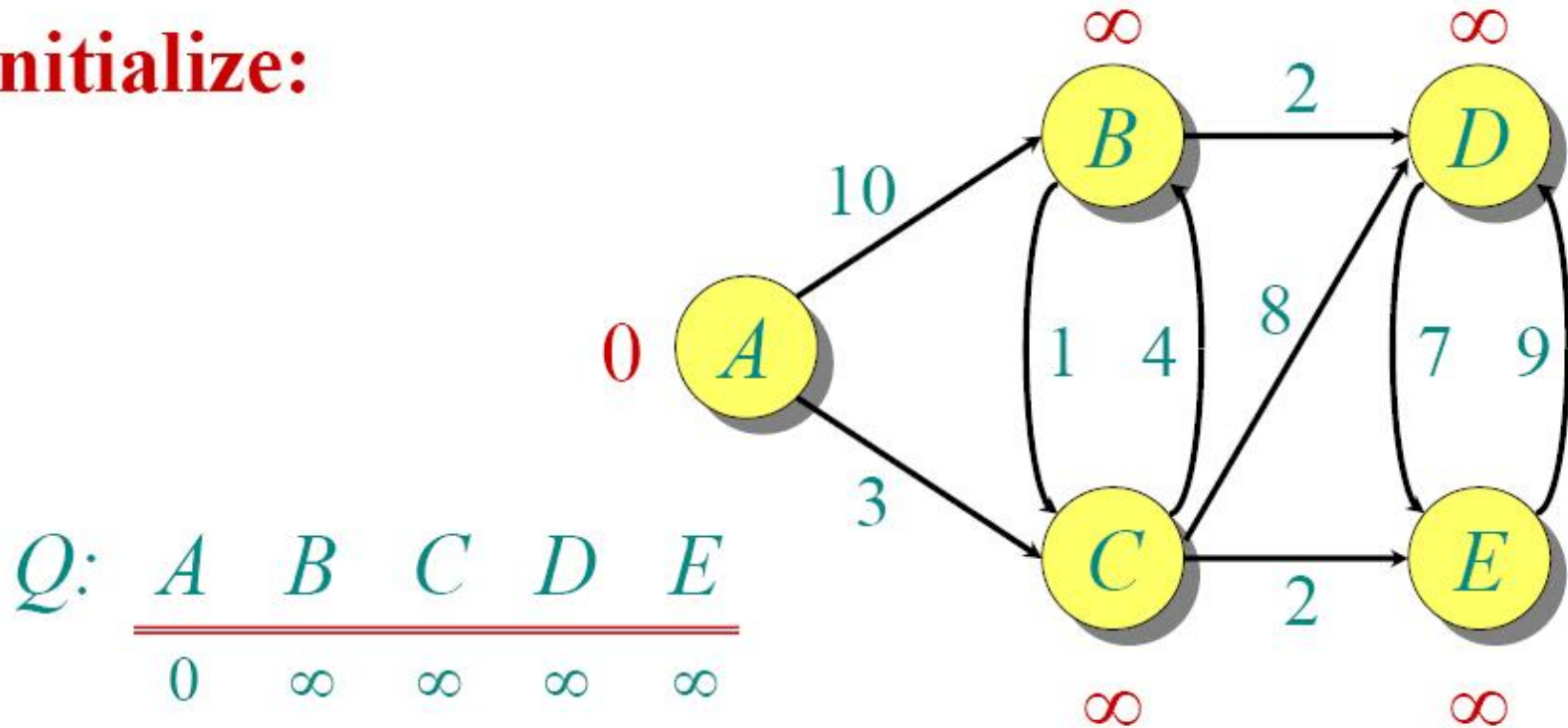
Shortest Paths in Graphs



- Given graph $G=(V,E)$, find shortest paths from a given node *source* to all nodes in V . (**Single-source All Destinations**)
- If negative weight cycle exist from $s \rightarrow t$, shortest is undefined
 - Can always reduce the cost by navigating the negative cycle
- If graph with all +ve weights \rightarrow Dijkstra's algorithm
- If graph with some -ve weights \rightarrow Bellman-Ford's algorithm

Dijkstra's Algorithm

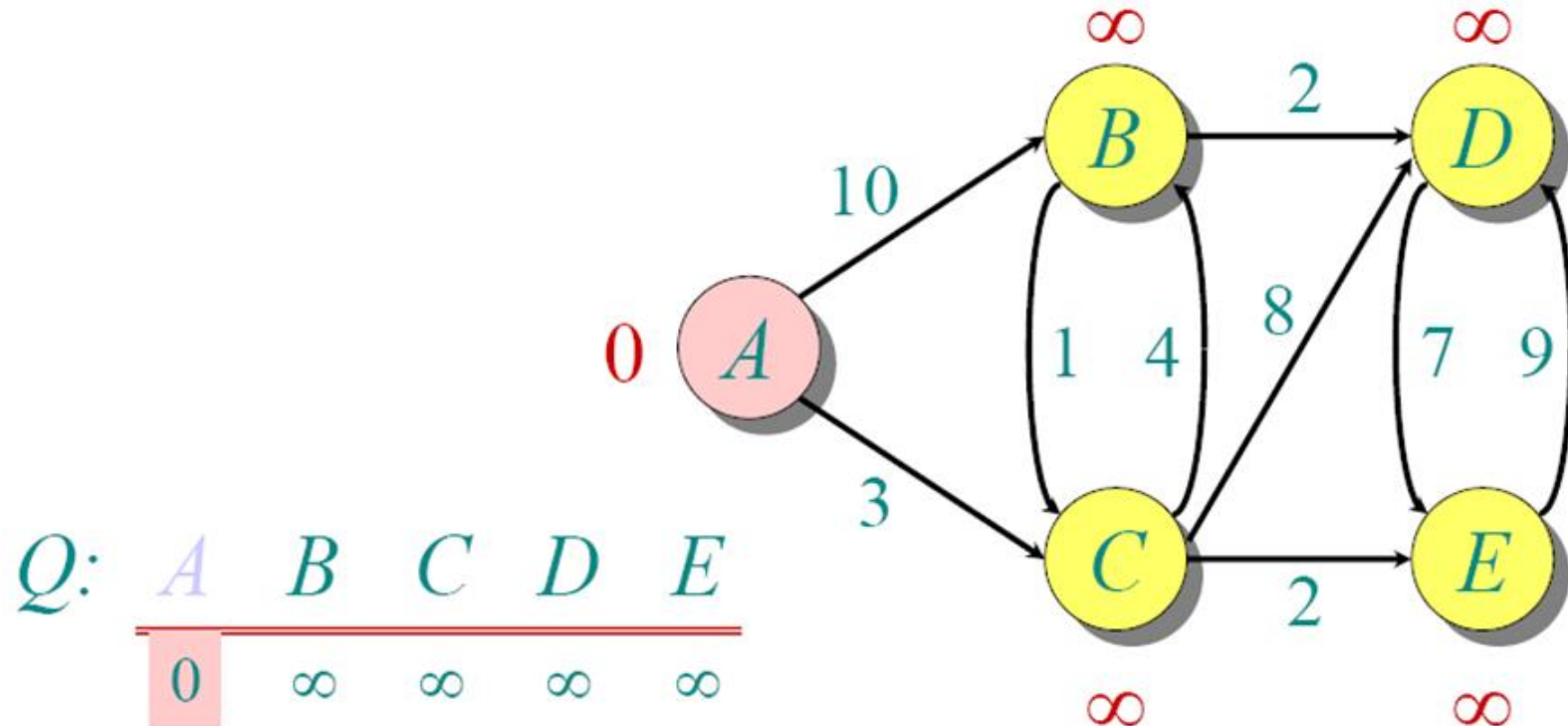
Initialize:



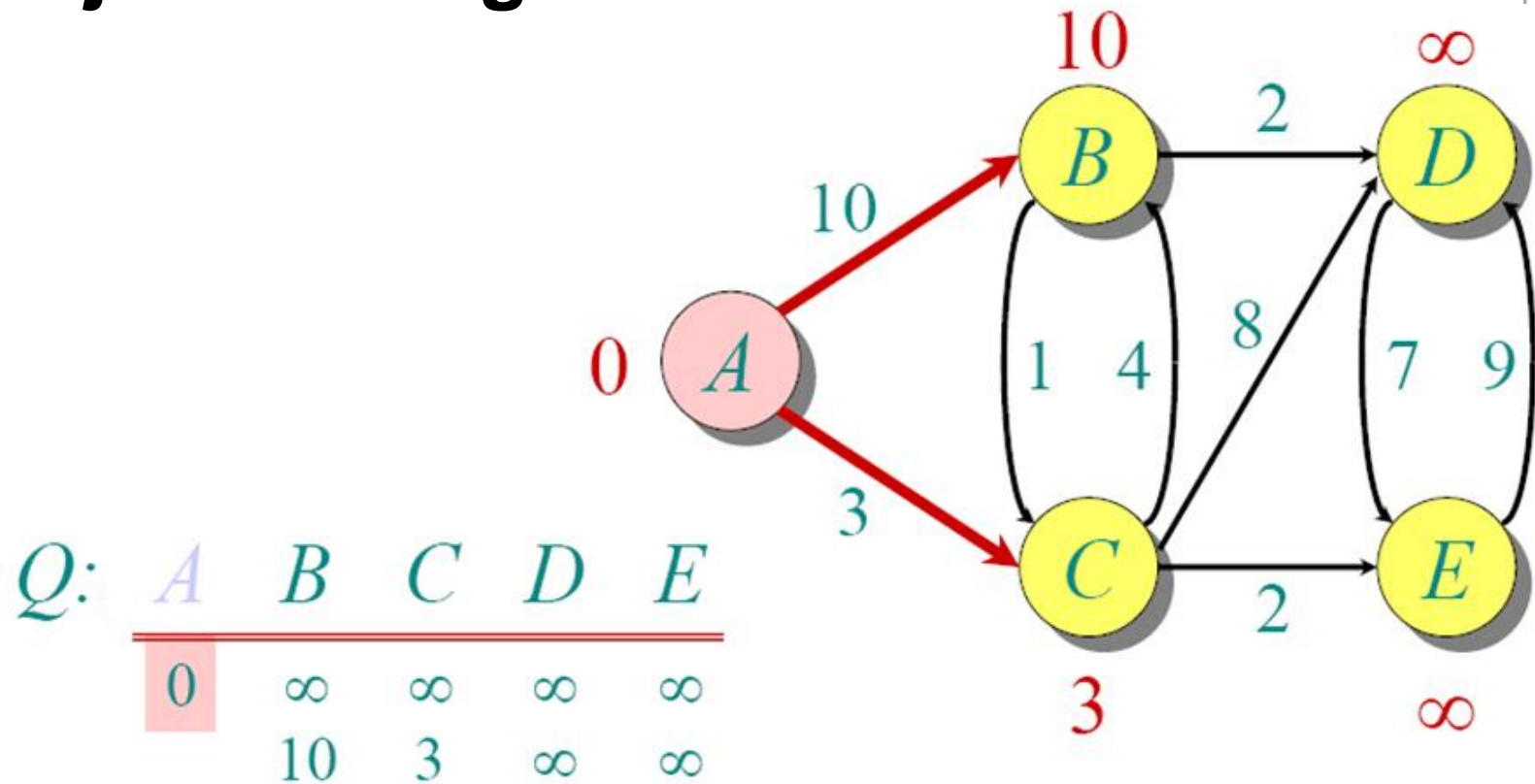
$S: \{\}$

$Prev: \{A, U, U, U, U\}$

Dijkstra's Algorithm



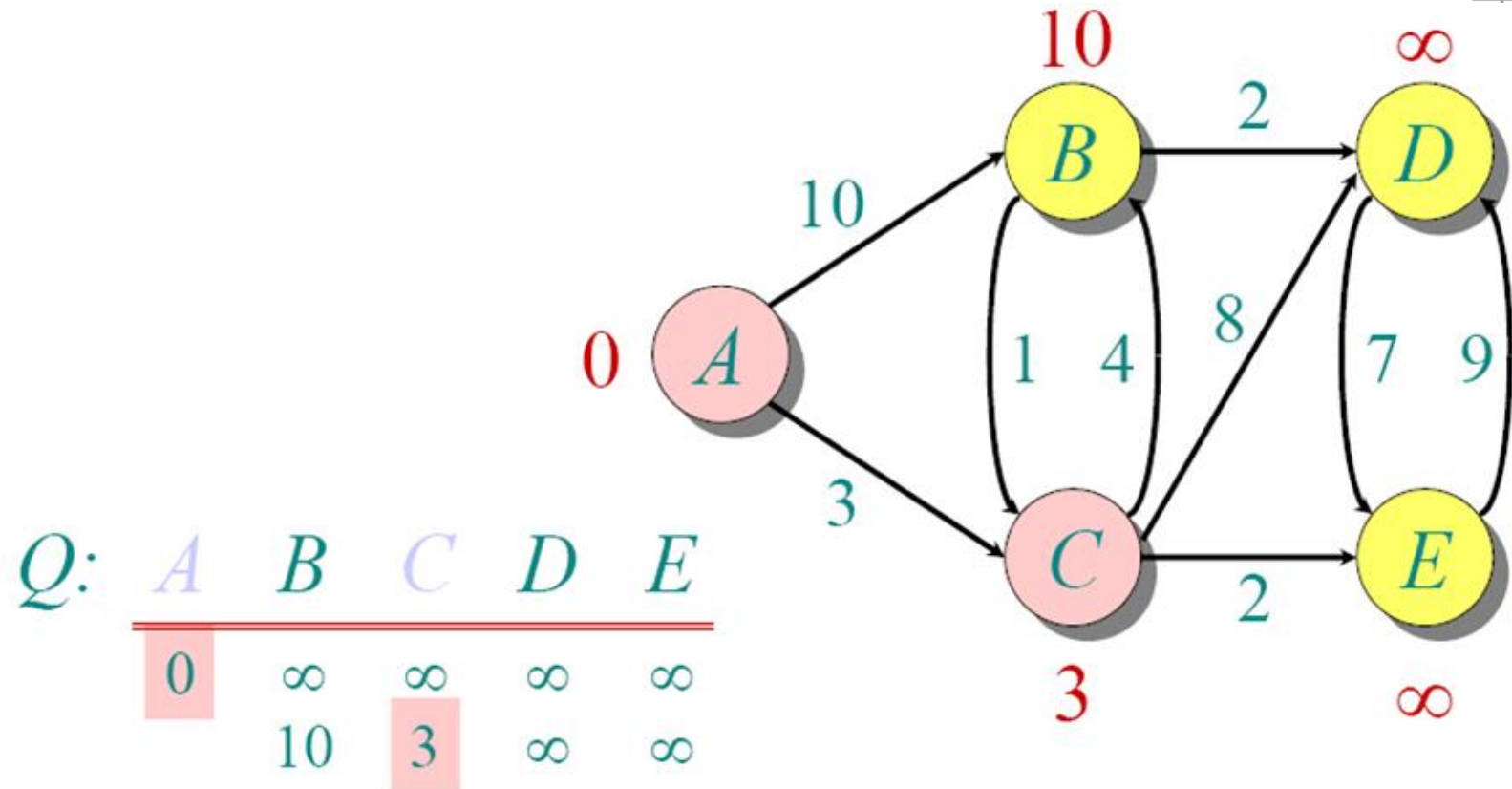
Dijkstra's Algorithm



$S: \{A\}$

$Prev: \{A, A, A, U, U\}$

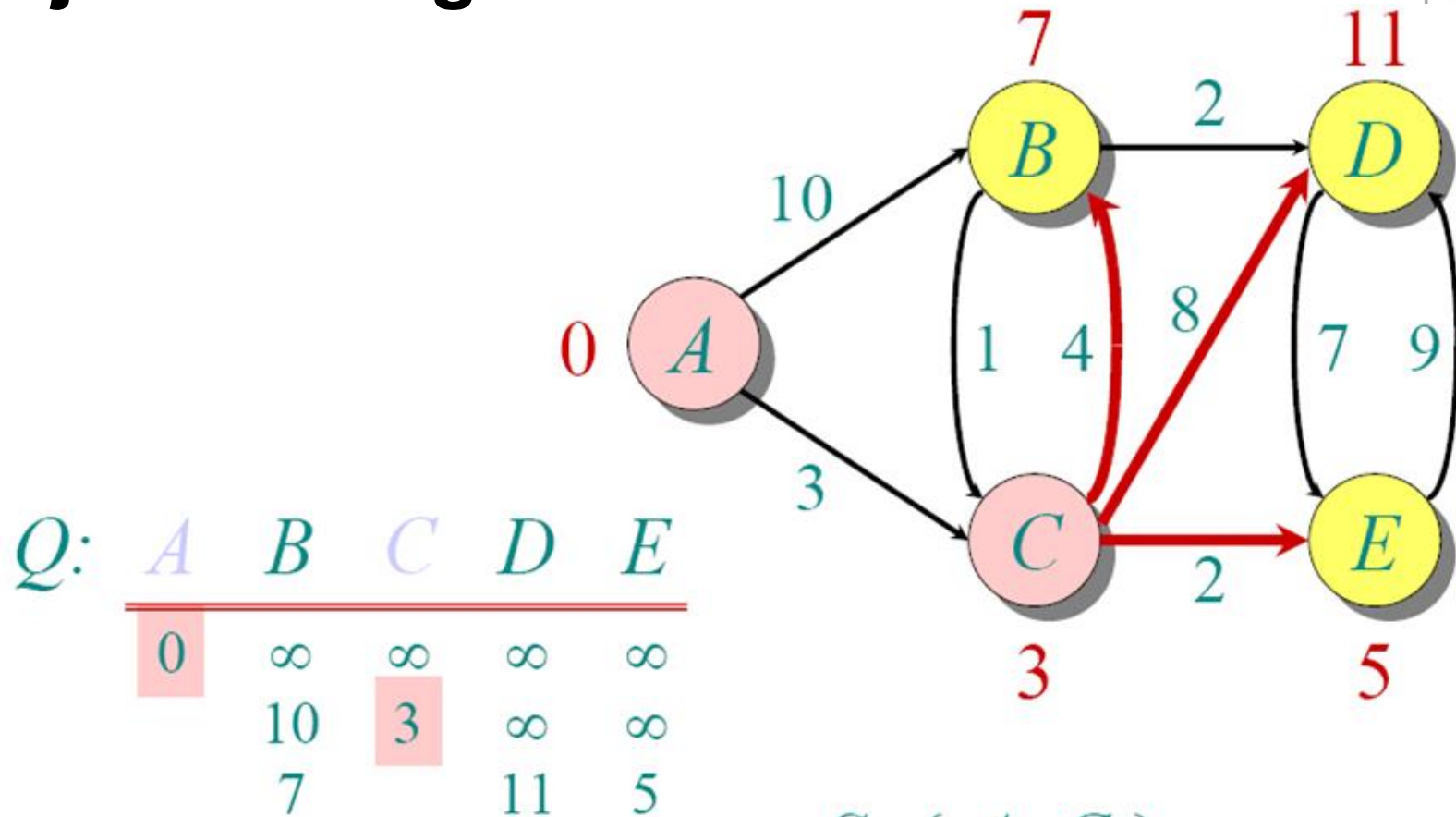
Dijkstra's Algorithm



$S: \{A, C\}$

$Prev: \{A, A, A, U, U\}$

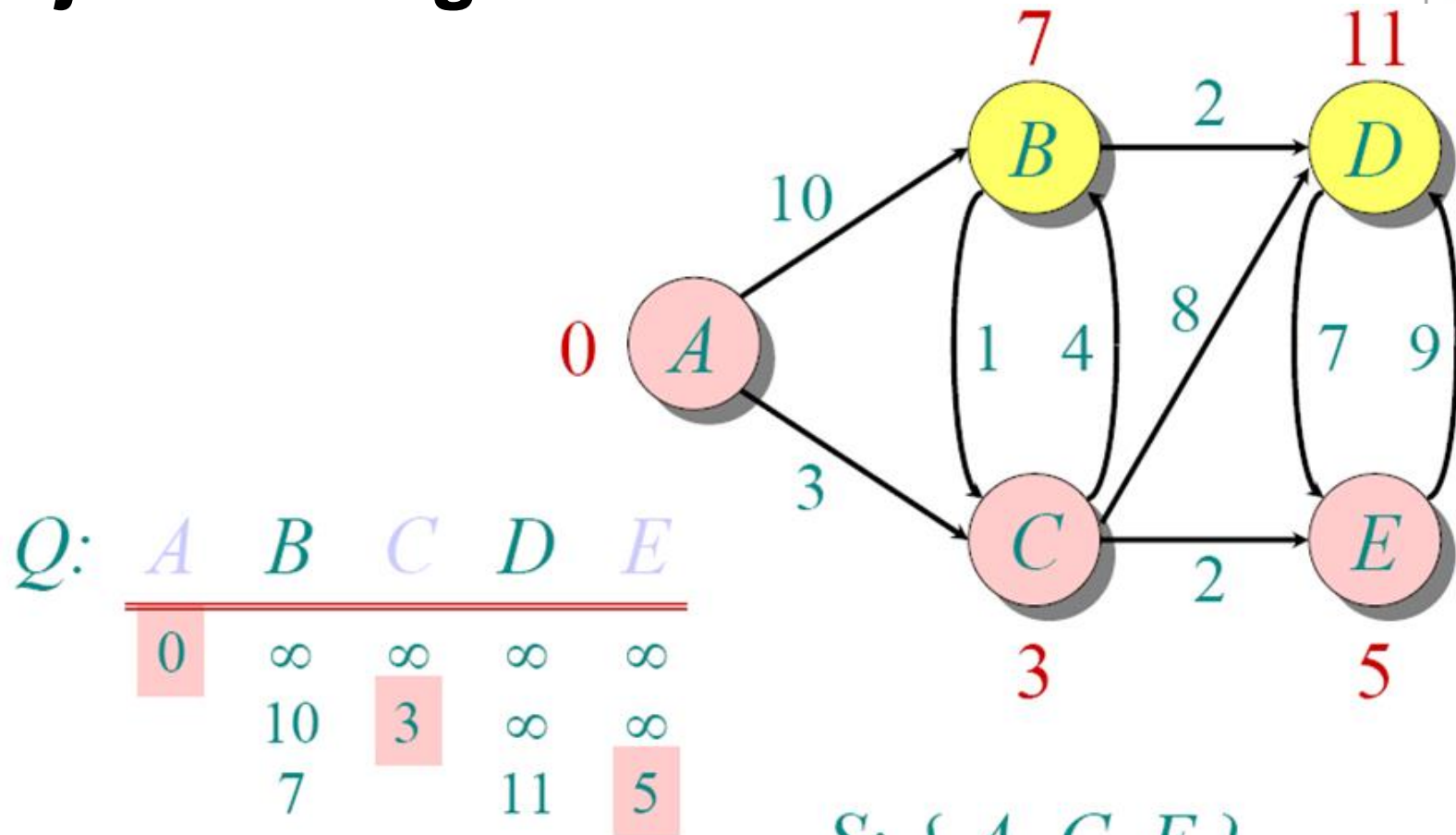
Dijkstra's Algorithm



$S: \{A, C\}$

$Prev: \{A, C, A, C, C\}$

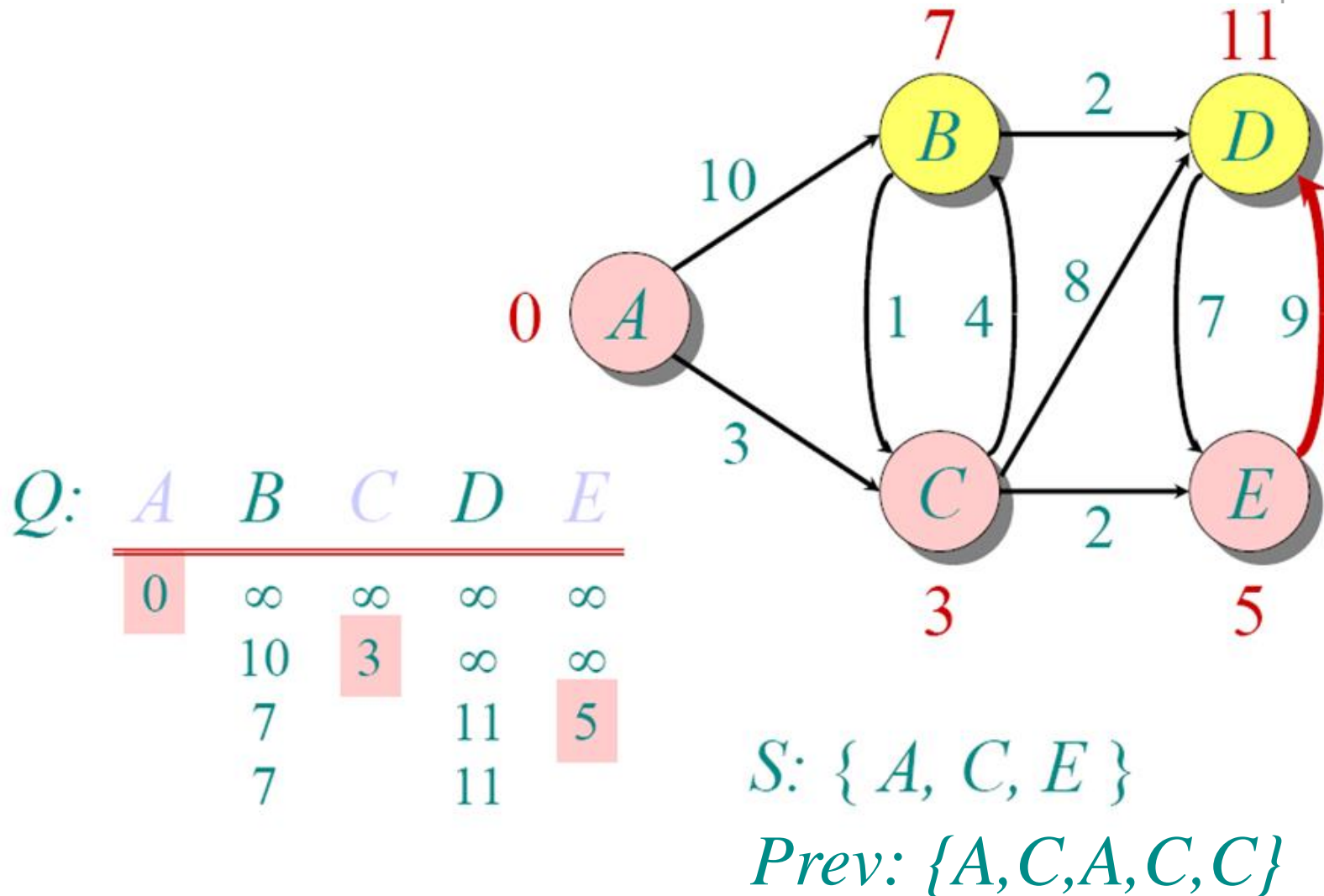
Dijkstra's Algorithm



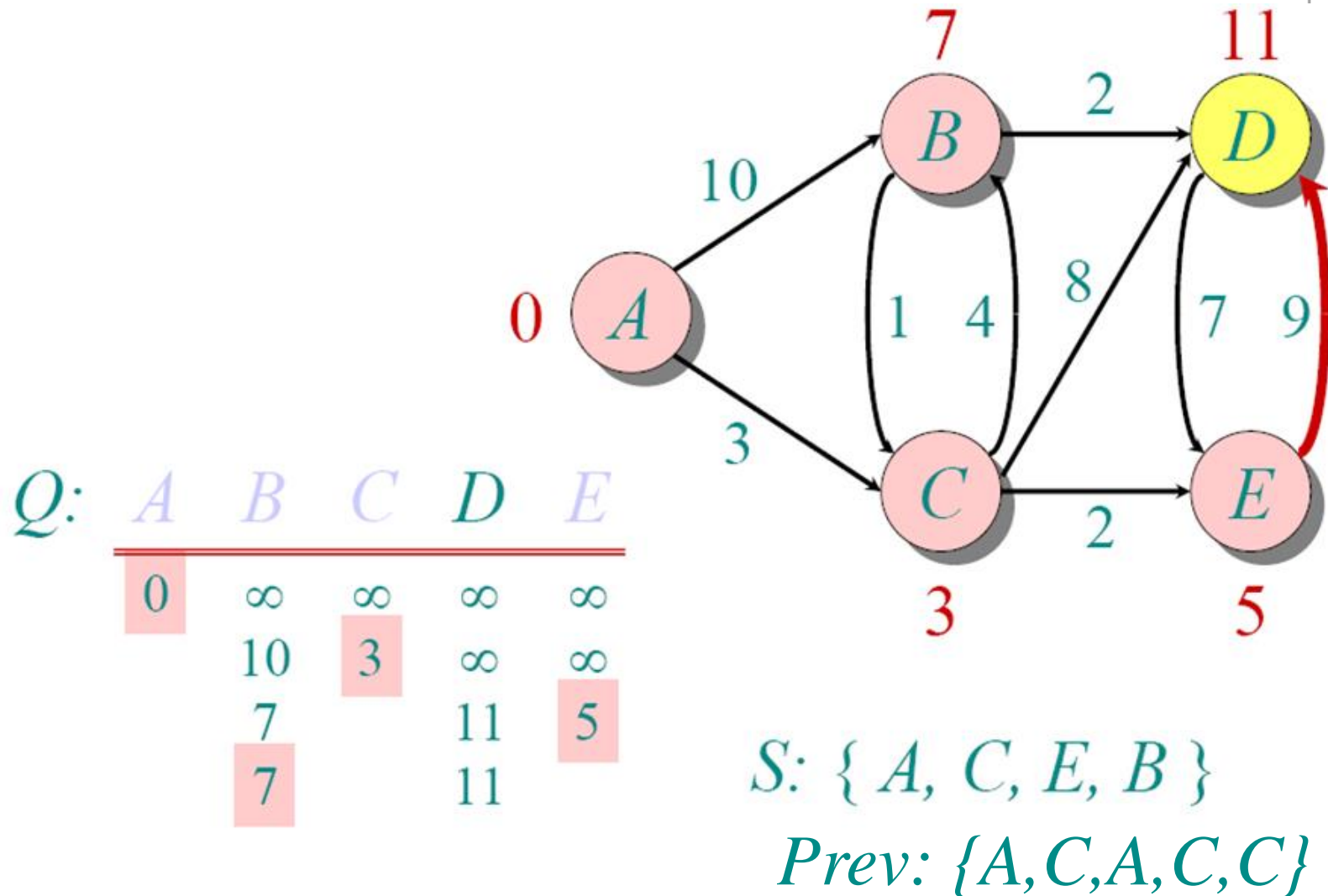
$S: \{A, C, E\}$

$Prev: \{A, C, A, C, C\}$

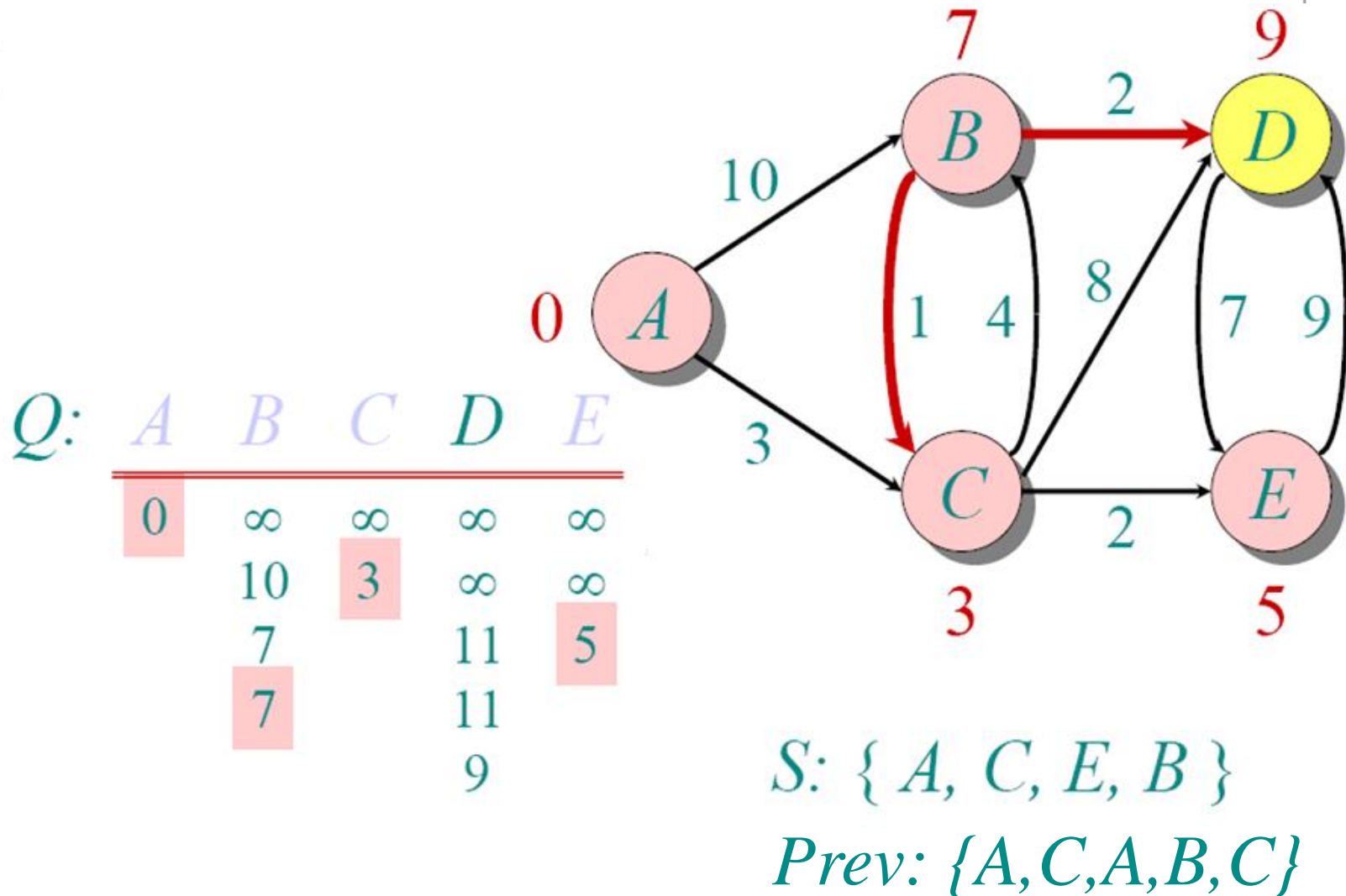
Dijkstra's Algorithm



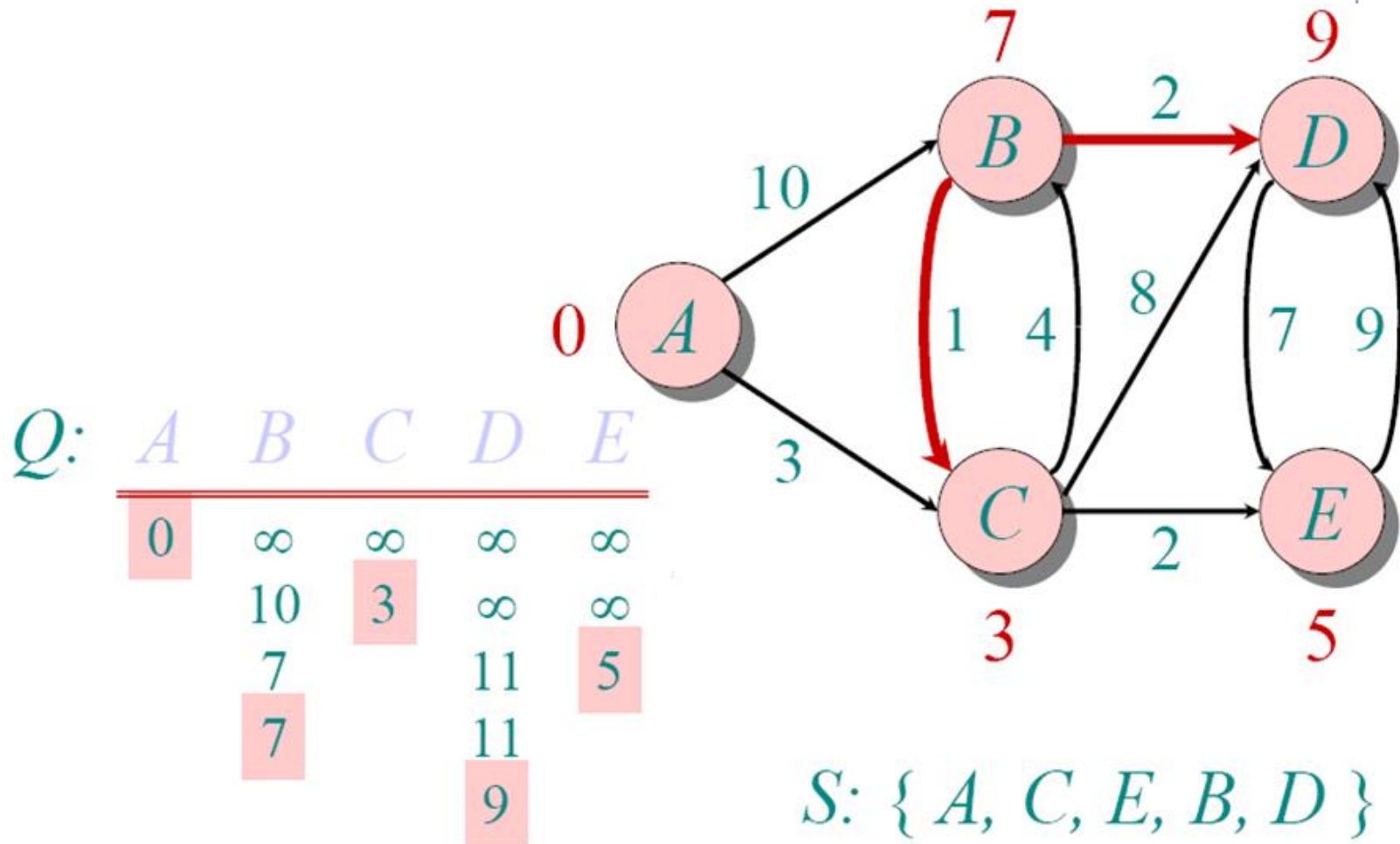
Dijkstra's Algorithm



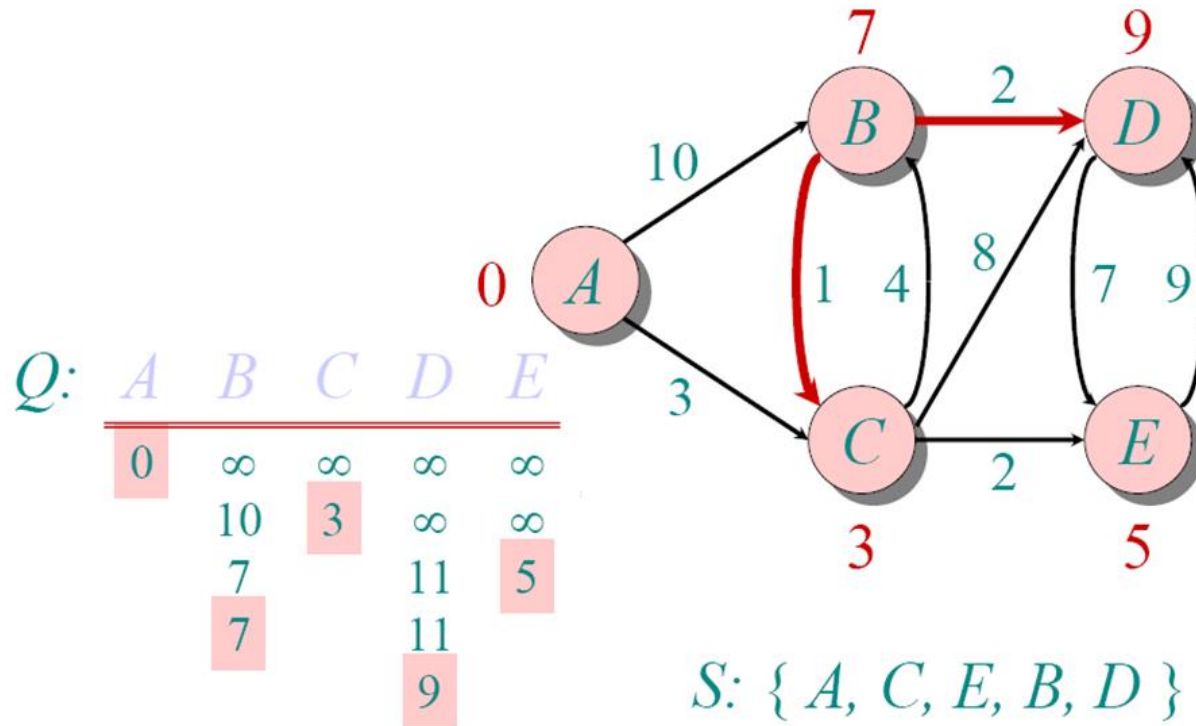
Dijkstra's Algorithm



Dijkstra's Algorithm



Dijkstra's Algorithm



A: A \rightarrow A

B: A \rightarrow C \rightarrow B

C: A \rightarrow C

D: A \rightarrow C \rightarrow B \rightarrow D

E: A \rightarrow C \rightarrow E

Dijkstra's Algorithm

```
1 function Dijkstra(Graph, source):
2
3     create vertex set Q
4
5     for each vertex v in Graph:    //Initialization
6         Dist[v] ← INFINITY        //Unknown distance from source to v
7         Prev[v] ← UNDEFINED       //Previous node in path from source to v
8         add v to Q                //All nodes initially unvisited (in Q)
9
10    Dist[source] ← 0               // Distance from source to source = 0
11    Prev[source] ← source
12    while Q is not empty:
13        u ← vertex in Q with min Dist[u] //Node with the least distance
14                                           // will be selected first
15        remove u from Q
16
17        for each neighbor v of u in Q:    //v is still in Q.
18            tmp ← Dist[u]+edge_length(u, v) //trying u as "source->u->v"
19            if tmp < Dist[v]:             //A shorter path to v has been found
20                Dist[v] ← tmp
21                Prev[v] ← u
22
23    return Dist[], S[]
```

Book Readings & Credits

- Book Readings:
 - Ch. 22, 23.2, 24.3
- Credits:
 - Figures:
 - Wikipedia
 - btechsmartclass.com
 - <https://www.codingeek.com/data-structure/graph-introductions-explanations-and-applications/>
 - Prof. Ahmed Eldawy notes
 - Laksman Veeravagu and Luis Barrera