



# CS590/CPE590

**Shortest Path Algorithms**

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**Spring 2023**

# Shortest-Path

- Given a graph (directed or undirected)  $G = (V, E)$  with weight function  $w: E \rightarrow \mathbf{R}$  and a vertex  $s \in V$ , find for all vertices  $v \in V$  the minimum possible weight for path from  $s$  to  $v$ .

- The weight of path  $p = v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_k$  is

$$w(p) = \sum_{i=1}^{k-1} w(v_i, v_{i+1})$$

- Shortest path = a path of the minimum weight
- Algorithm will compute a **shortest-path tree**.

# Shortest-Path Problems

- Shortest-Path problems
  - **Single-Source (Single-Destination):** Find a shortest path from a given source (vertex  $s$ ) to each of the vertices.
  - **Single-Pair:** Given two vertices, find a shortest path between them. Solution to single-source problem solves this problem efficiently, too.
  - **All-Pairs:** Find shortest-paths for every pair of vertices. Dynamic programming algorithm.

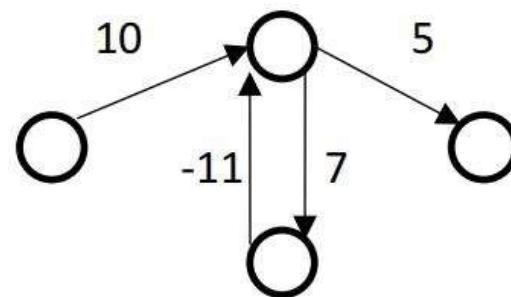
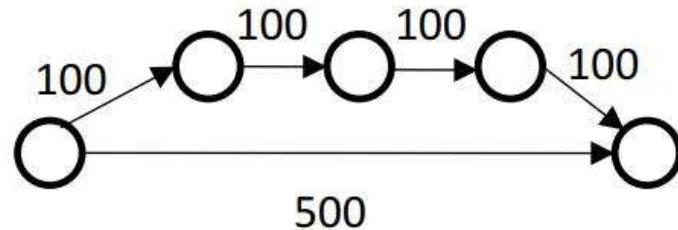
# Single-Source Shortest Path

- Given a graph (directed or undirected)  $G = (V, E)$  with weight function  $w: E \rightarrow \mathbf{R}$  and a vertex  $s \in V$ , find for all vertices  $v \in V$  the minimum possible weight for path from  $s$  to  $v$ .
- We will discuss two general case algorithms:
  - Dijkstra's Algorithm** (non-negative edge weights only)
  - Bellman-Ford Algorithm** (positive and negative edge weights)
- If all edge weights are equal (let's say 1), the problem is solved by BFS in  $\Theta(V + E)$  time. That's why BFS works for undirected graphs as well if the edges are of equal weight or unweighted graph.

# BFS no longer Works!

Why BFS won't work: Shortest path may not have the fewest edges

- Annoying when this happens with costs of flights

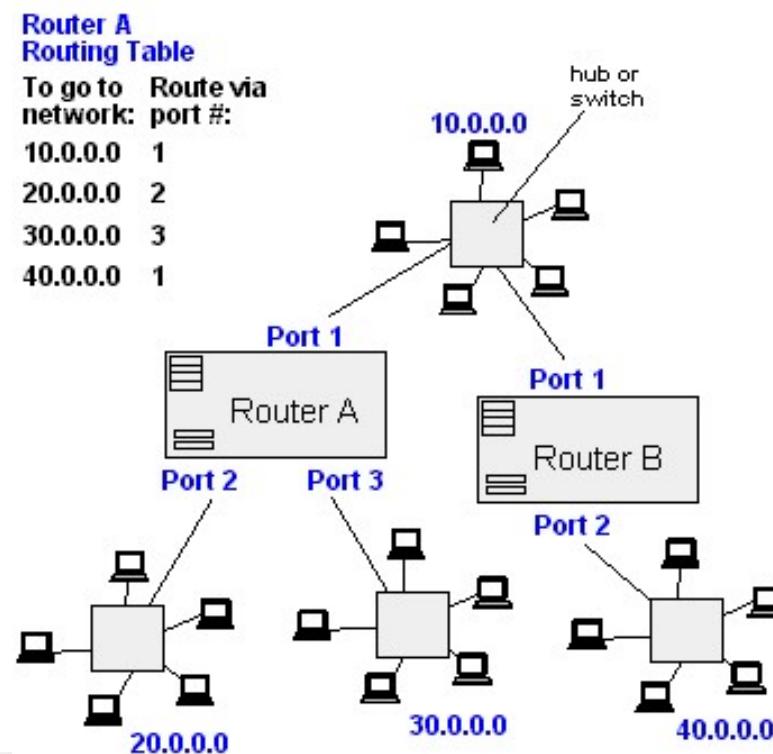


We will assume there are no negative weights

- Problem* is *ill-defined* if there are negative-cost cycles

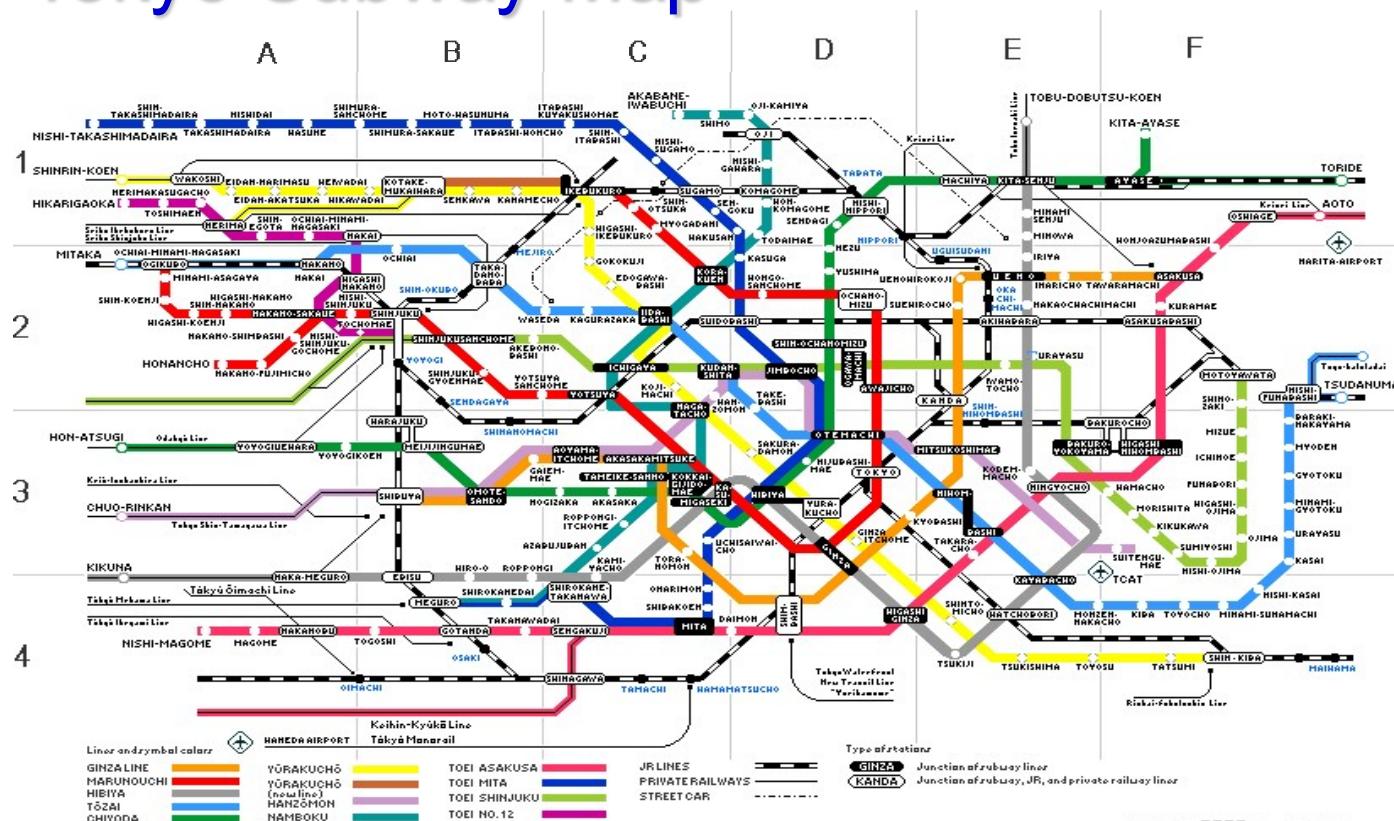
# Single-Source Shortest Path

- Traffic Information Systems are most prominent use
- Mapping (Map Quest, Google Maps)
- Routing Systems
- Flight Reservations



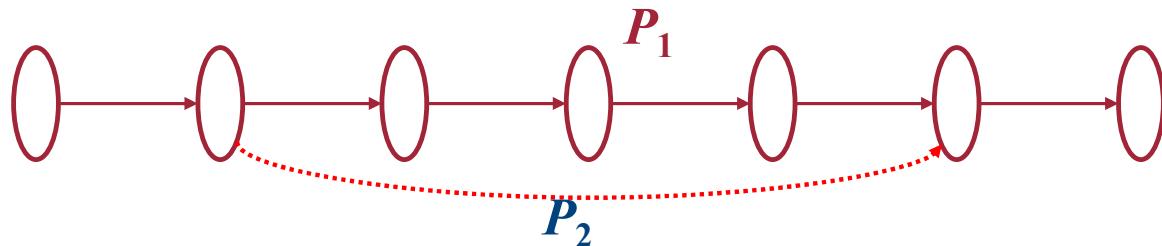
# Single-Source Shortest Path

## Tokyo Subway Map



# Shortest Path Properties

- The shortest path problem satisfies the ***optimal substructure property***:
  - Sub-paths of shortest paths are shortest paths.

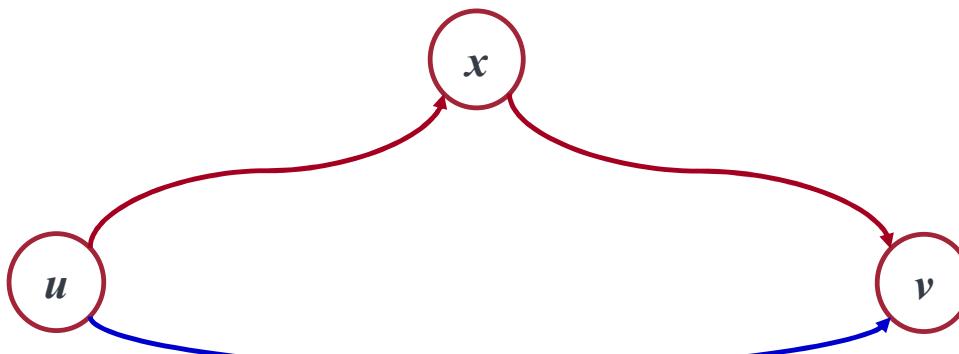


- **Proof:** suppose some sub-path  $P_1$  is not a shortest path
  - There must then exist a shorter sub-path  $P_2$
  - Could substitute the sub-path  $P_1$  by the shorter path  $P_2$
  - But then overall path is not the shortest path. **Contradiction**

# Shortest Path Properties

- Define  $\delta(u, v)$  to be the weight of the shortest path from  $u$  to  $v$
- Shortest paths satisfy the *triangle inequality*:  
$$\delta(u, v) \leq \delta(u, x) + \delta(x, v)$$

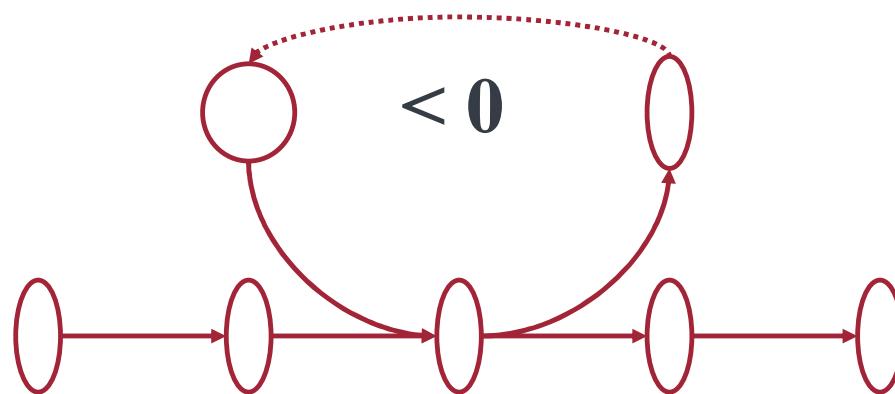
- **Proof :**



*This path is no longer than any other path*

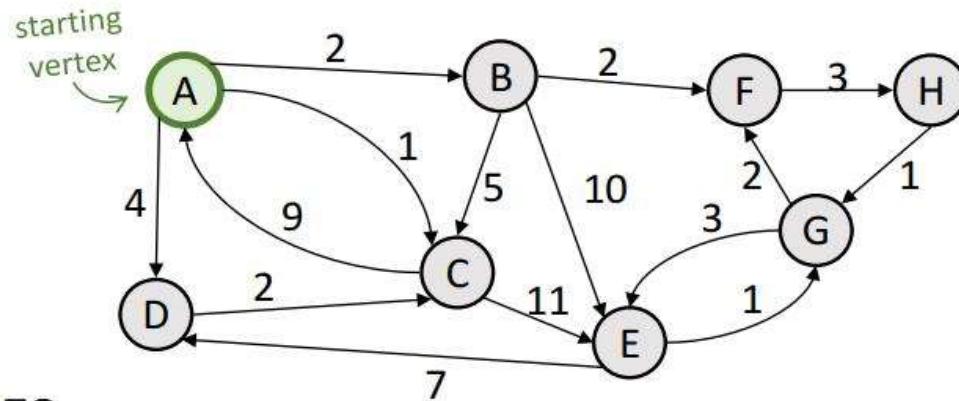
# Shortest Path Properties

- In graphs with negative weight cycles, some shortest paths will not exist (*Why?*):



# Algorithm: General Idea

**Goal:** From one starting vertex, what are the shortest paths to each of the other vertices (for a weighted graph)?

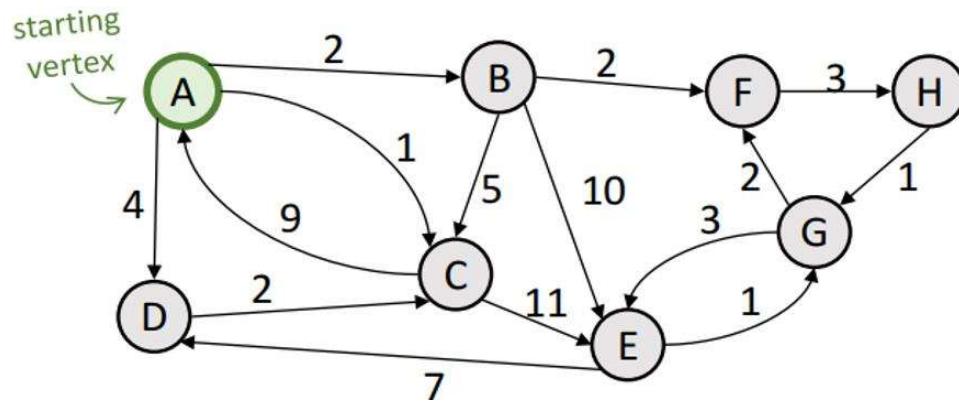


**Idea:** Similar to BFS

- Repeatedly increase a “set of vertices with known shortest distances”
- Any vertex not in this set will have a “best distance so far”
- Each vertex has a “cost” to represent these shortest/best distances
- Update costs (i.e. “best distances so far”) as we add vertices to set

# Single-Source Shortest Path Problem

**Single-Source Shortest Path Problem** - The problem of finding shortest paths from a source vertex  $v$  to all other vertices in the graph.



# Dijkstra's algorithm

Dijkstra's algorithm - is a solution to the single-source shortest path problem in graph theory.

-- Named after its inventor Edsger Dijkstra (1930-2002)

Works on both directed and undirected graphs. However, all edges must have **nonnegative weights**.

**Approach:** Greedy

**Input:** Weighted graph  $G=\{E,V\}$  and source vertex  $v \in V$ , such that all edge weights are **nonnegative**

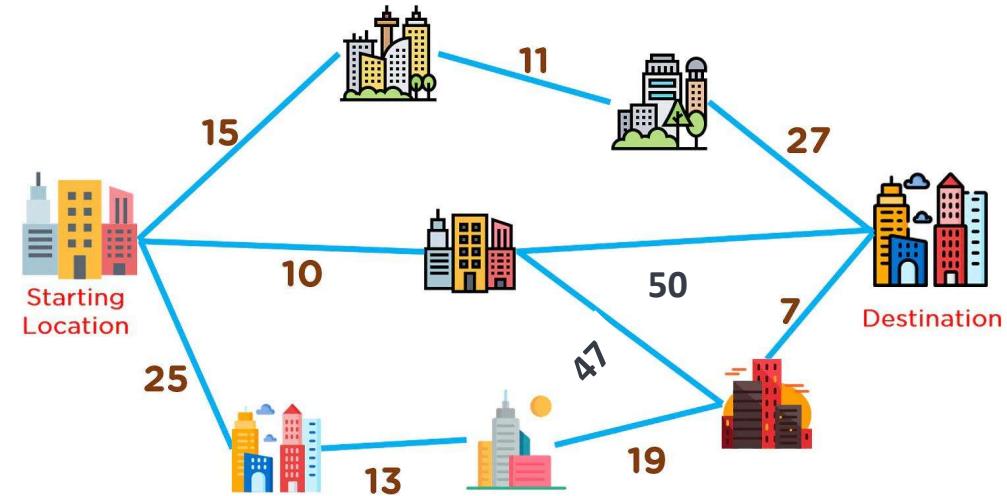
**Output:** Lengths of shortest paths (or the shortest paths themselves) from a given source vertex  $v \in V$  to all other vertices

# Greedy Algorithm

- Dijkstra's Algorithm is one example of ...
  - A greedy algorithm:
    - Make a locally optimal choice at each stage to (hopefully) find a global optimum
    - Settle on the best-looking option at each repeated step
    - For some problems, greedy algorithms cannot find best answer!

# Greedy Algorithms: Principles

- A greedy algorithm works in phases
- At each phase:
  - You take the **best you can get right now**, without regard for future consequences.
  - You hope that by choosing a local optimum at each step, you will end up at a global optimum.



# Dijkstra's algorithm - Pseudocode

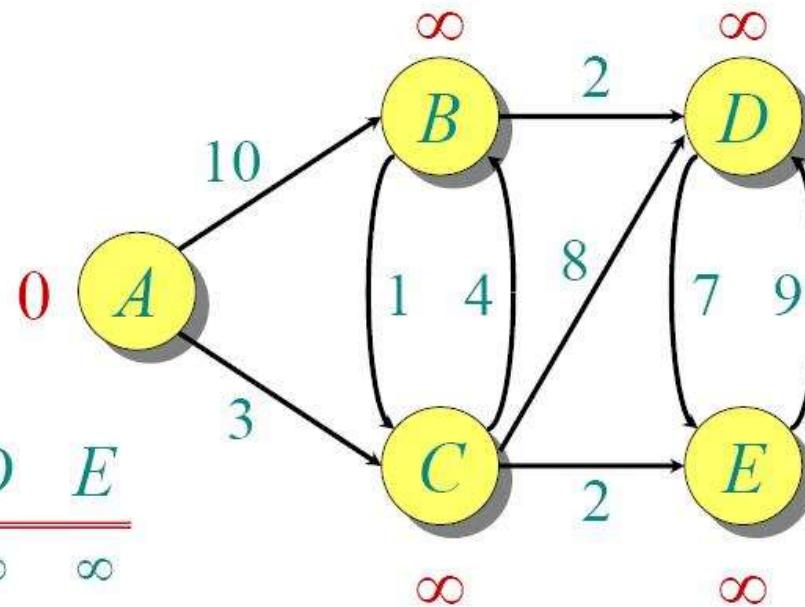
```
dist[s] ← 0                                (distance to source vertex is zero)
for all  $v \in V - \{s\}$ 
    do dist[v] ←  $\infty$                       (set all other distances to infinity)
S ←  $\emptyset$                                 ( $S$ , the set of visited vertices is initially empty)
Q ← V                                     ( $Q$ , the queue initially contains all vertices)
while  $Q \neq \emptyset$                          (while the queue is not empty)
    do  $u \leftarrow \min.\text{distance}(Q, \text{dist})$  (select the element of  $Q$  with the min. distance)
        S ← S ∪ {u}                          (add  $u$  to list of visited vertices)
        for all  $v \in \text{Adj.}[u]$ 
            do if dist[v] > dist[u] + w(u, v)   (if new shortest path found)
                   then d[v] ← d[u] + w(u, v)      (set new value of shortest path)
return dist
```

# Dijkstra Example

Initialize:

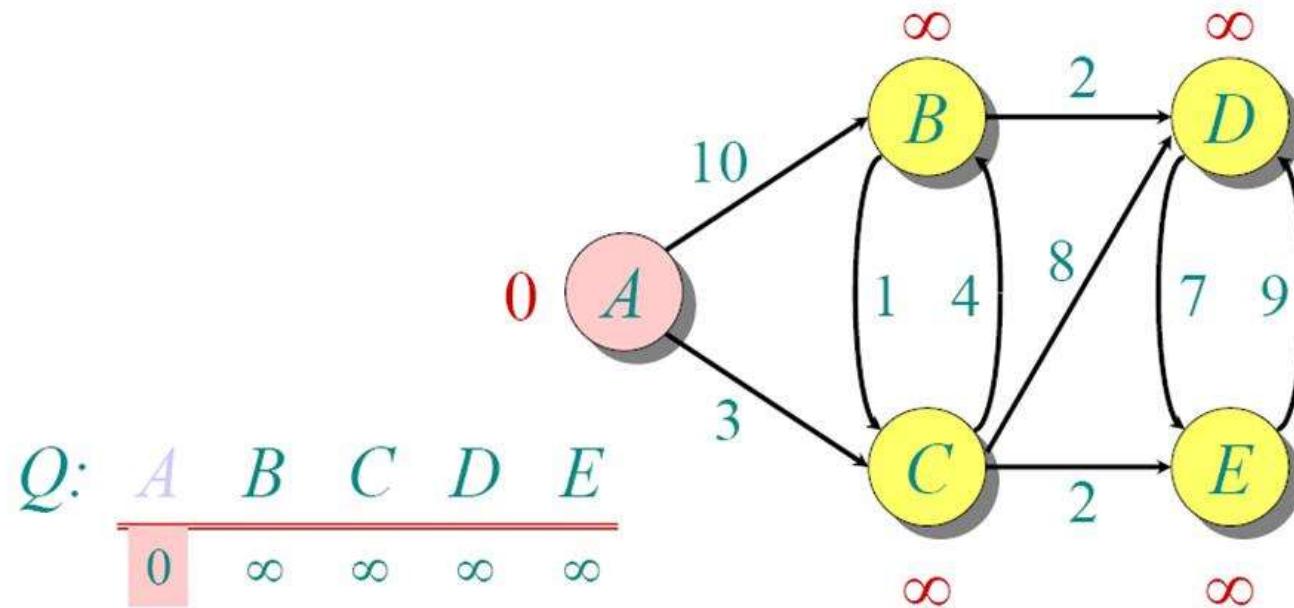
Source: A

$$Q: \begin{array}{ccccc} A & B & C & D & E \\ \hline 0 & \infty & \infty & \infty & \infty \end{array}$$

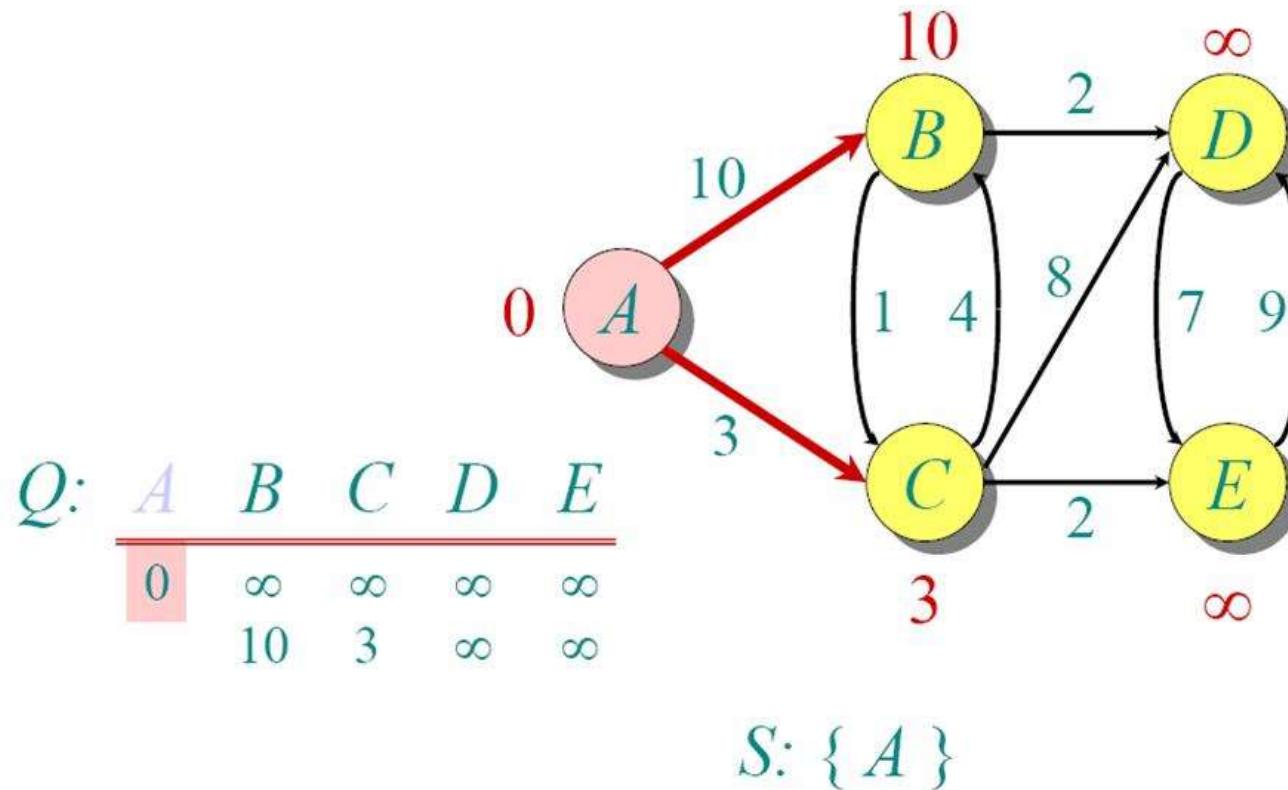


$$S: \{\}$$

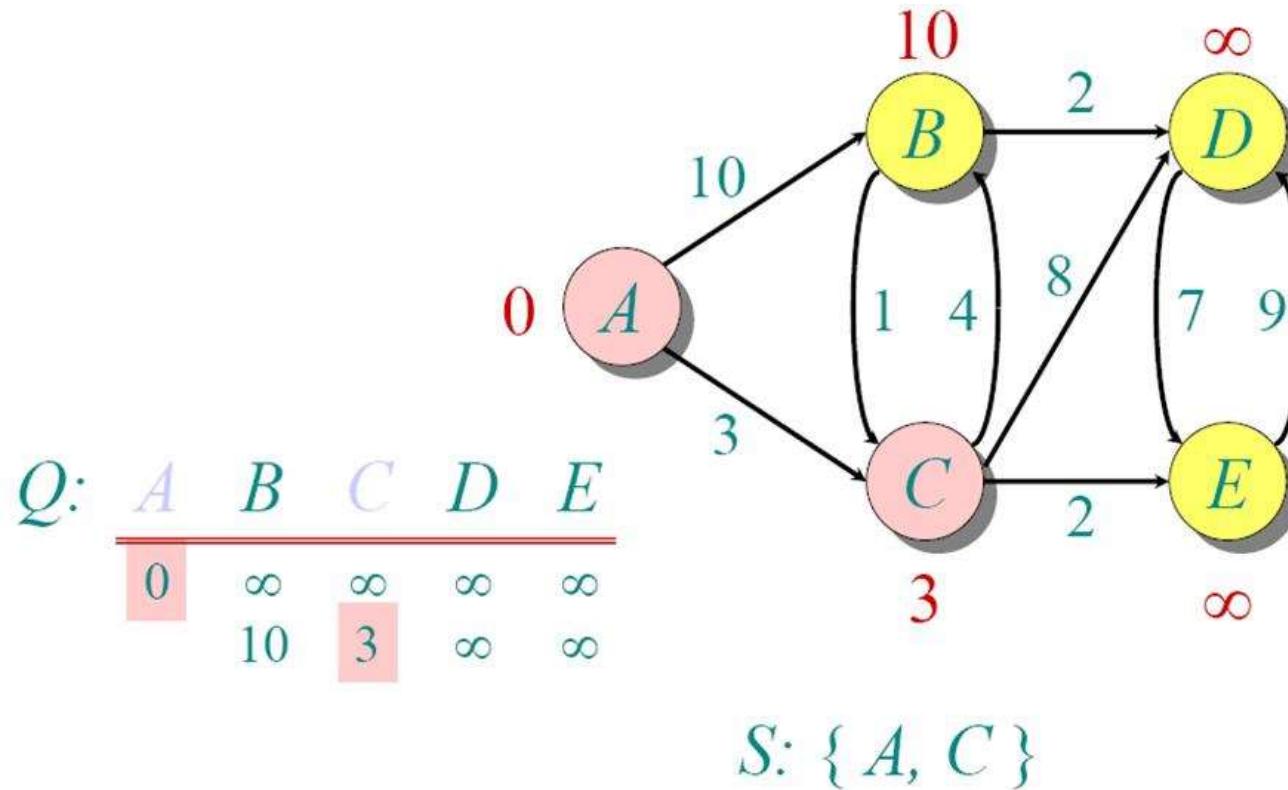
# Dijkstra Example



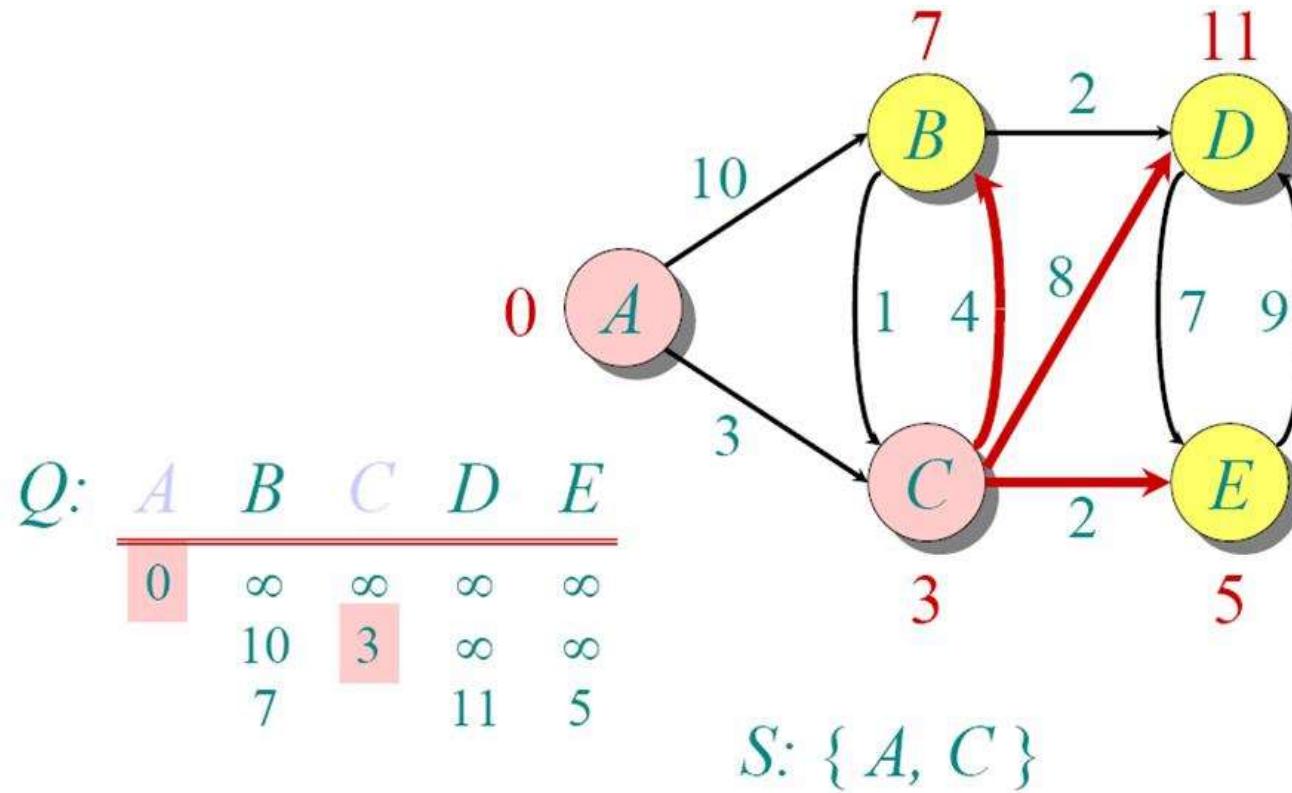
# Dijkstra Example



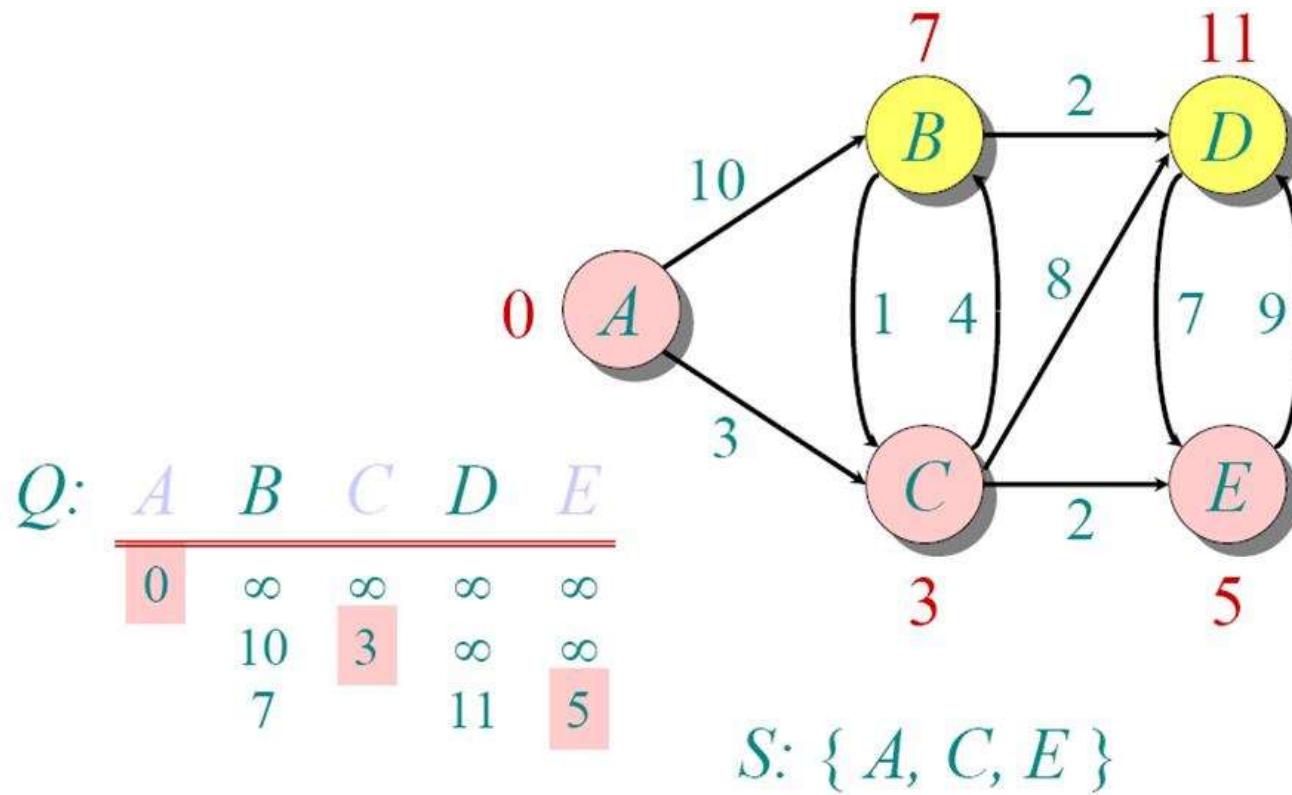
# Dijkstra Example



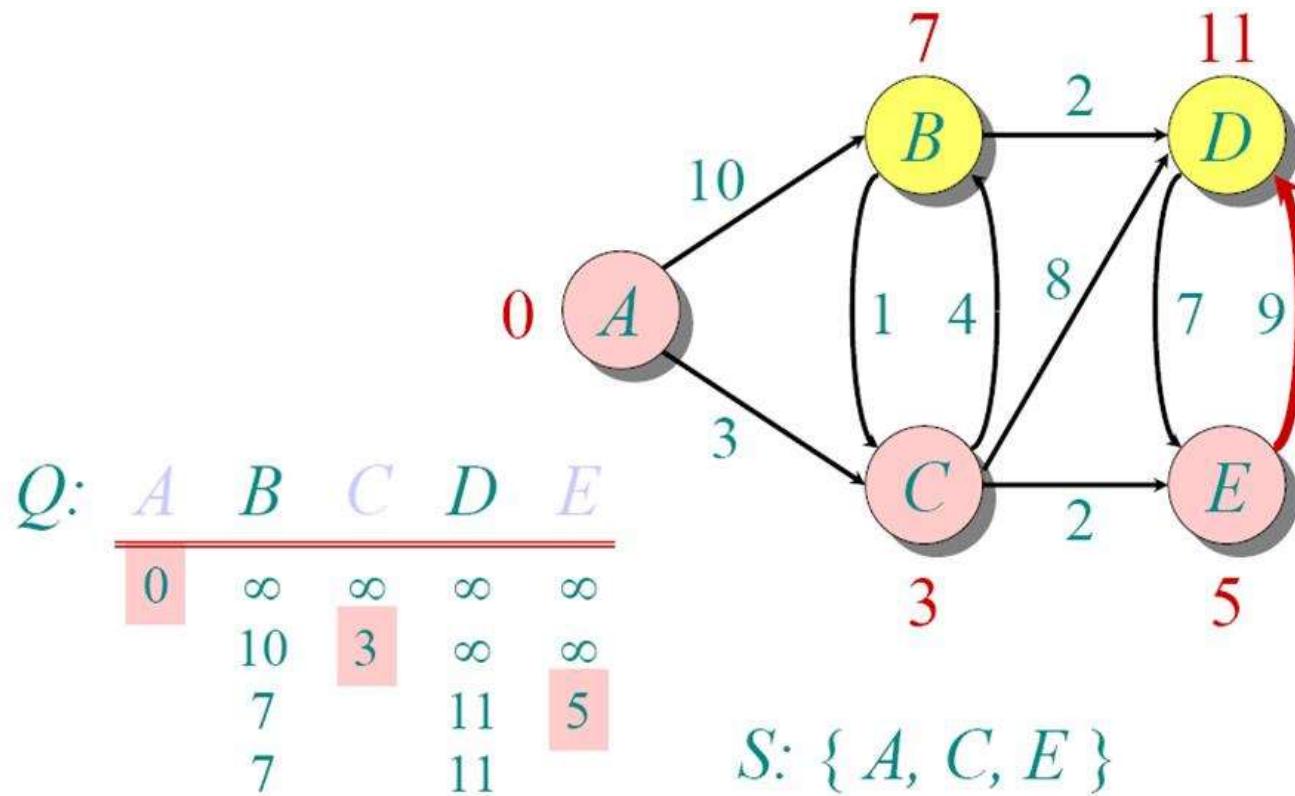
# Dijkstra Example



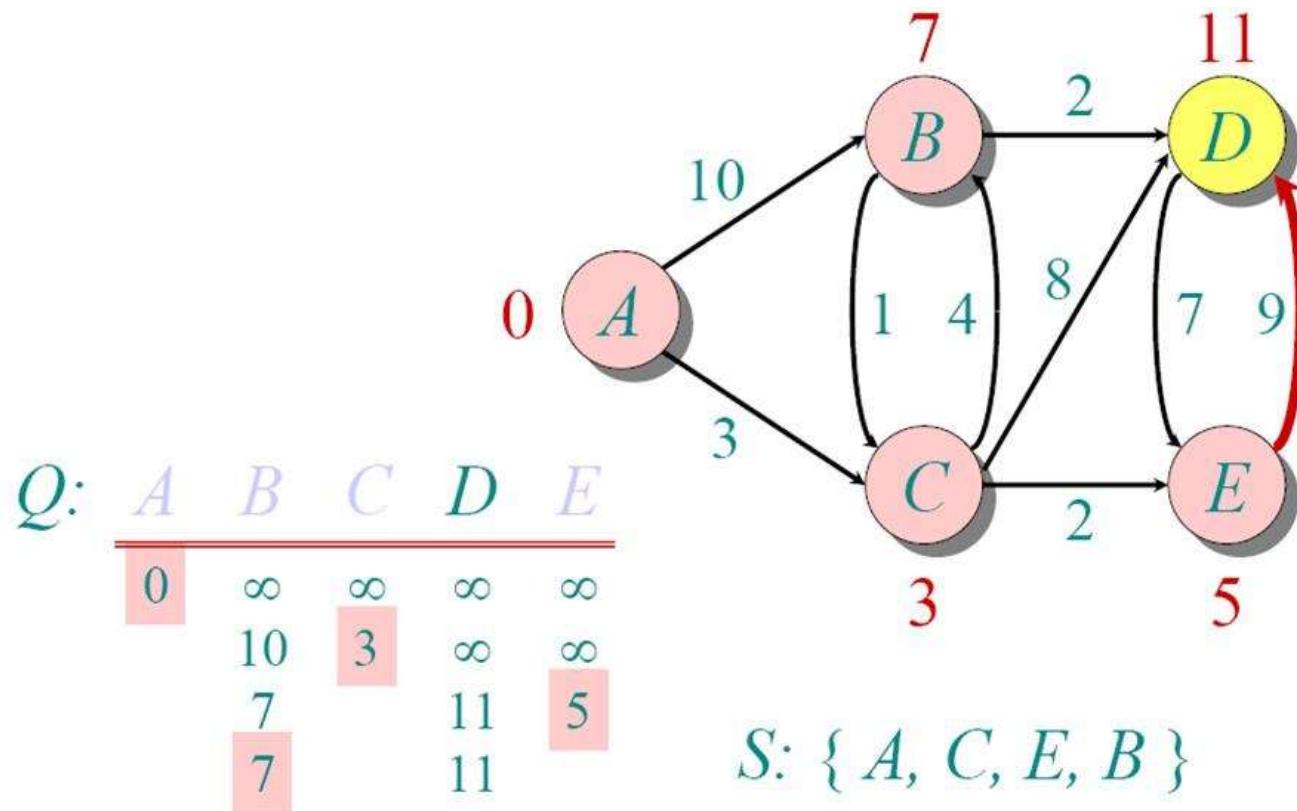
# Dijkstra Example



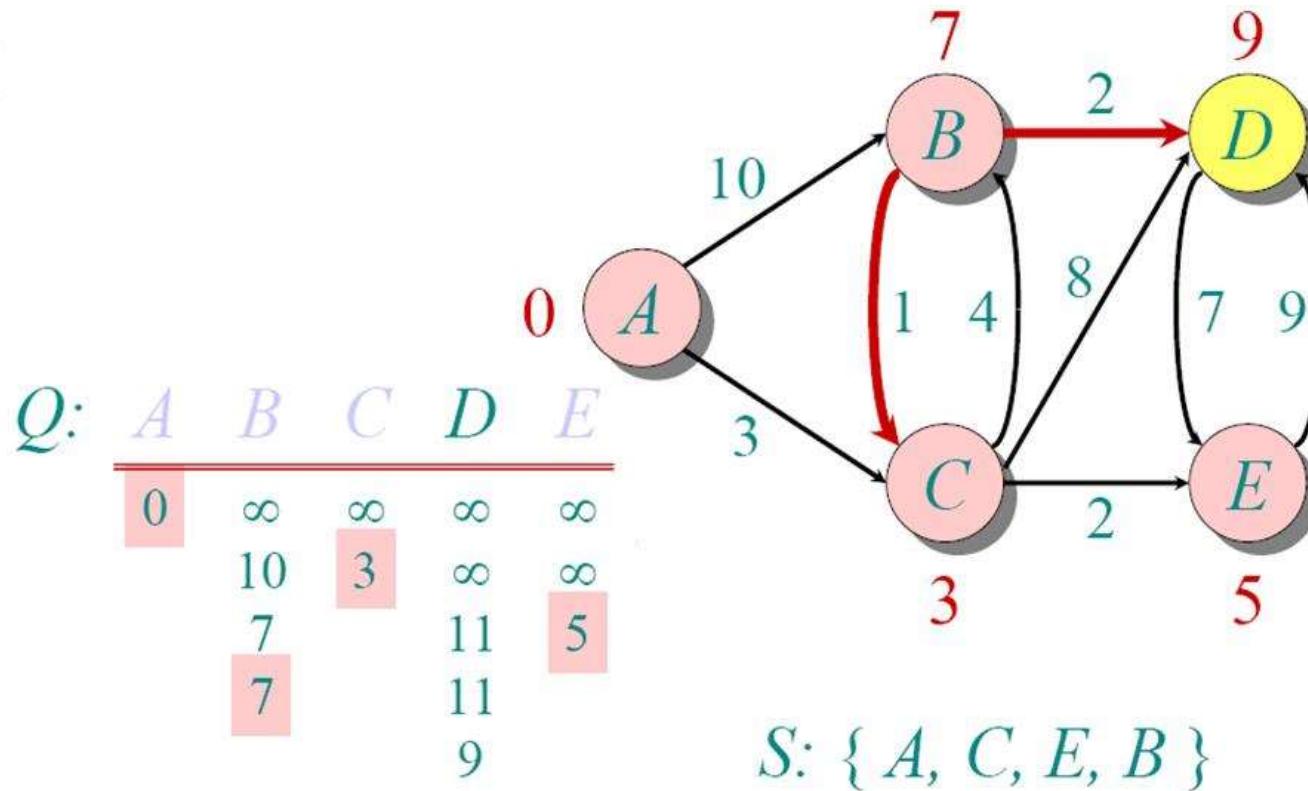
# Dijkstra Example



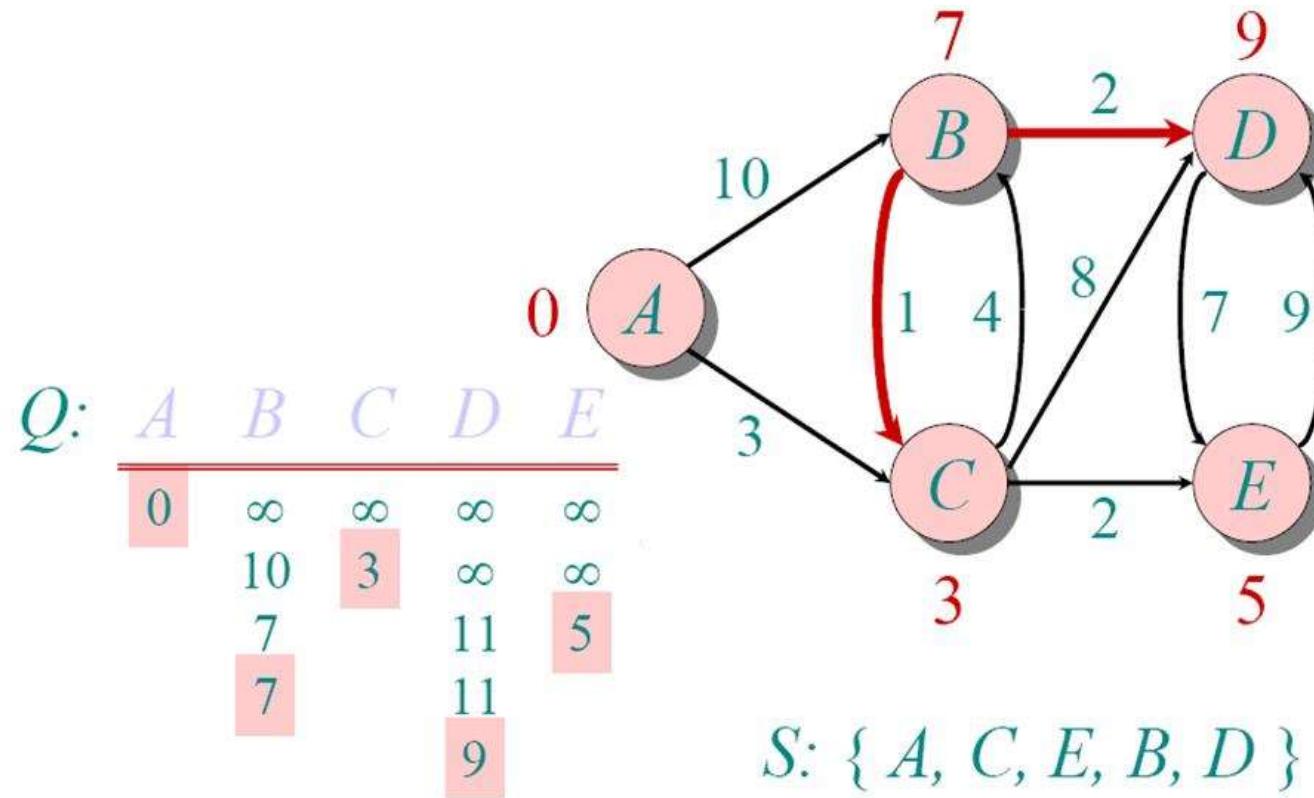
# Dijkstra Example



# Dijkstra Example



# Dijkstra Example



# Running Time: Using List/Array

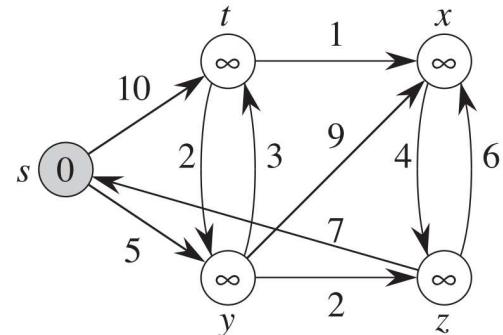
The simplest implementation of the Dijkstra's algorithm stores vertices in an ordinary linked list or array

- Good for dense graphs (many edges)
- $|V|$  vertices and  $|E|$  edges
- Initialization  $O(|V|)$
- While loop  $O(|V|)$ 
  - Find and remove min distance vertices  $O(|V|)$
  - Potentially  $|E|$  updates
    - Update costs  $O(1)$
- Reconstruct path  $O(|E|)$

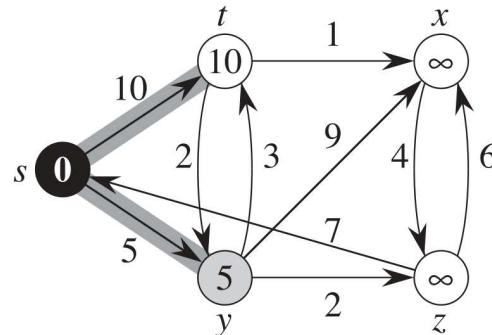
Total time  $O(|V^2| + |E|) = O(|V^2|)$

- Can be improved to  $O(|V| \log |V| + |E| \log |V|)$  using priority queue & heap data structure

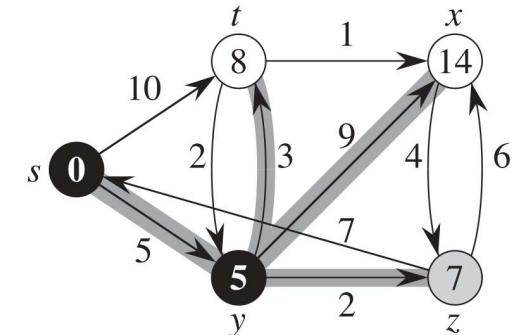
# Dijkstra's Algorithm: Another Example



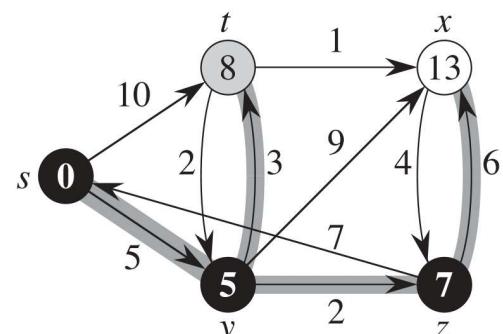
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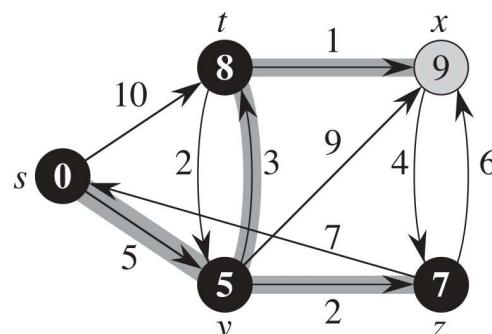
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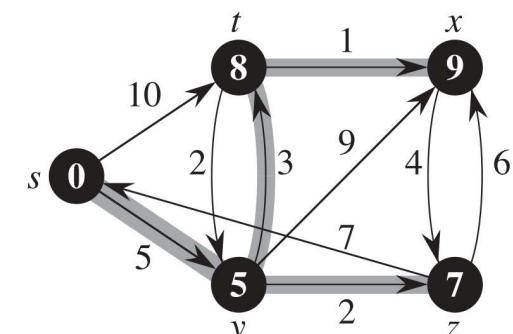
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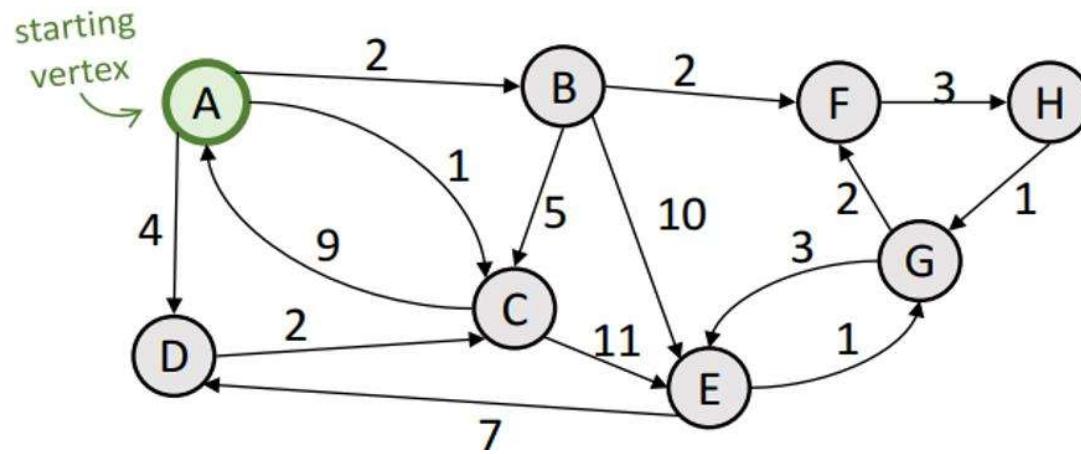
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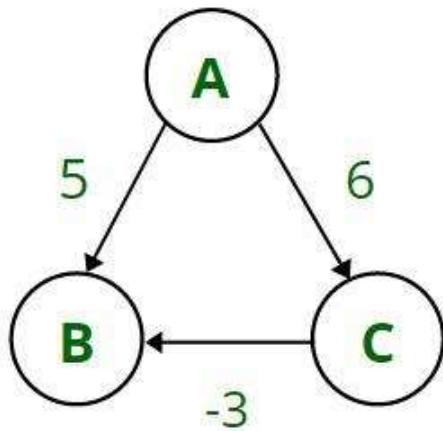
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# Exercise for Practice

- Try Dijkstra's Algo on the following graph:



# When Dijkstra's Algo Fails?



# Bellman-Ford Algorithm

BELLMAN-FORD( $G, w, s$ )

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
2  for  $i = 1$  to  $|G.V| - 1$ 
3      for each edge  $(u, v) \in G.E$ 
4          RELAX( $u, v, w$ )
5  for each edge  $(u, v) \in G.E$ 
6      if  $v.d > u.d + w(u, v)$ 
7          return FALSE
8  return TRUE
```

INITIALIZE-SINGLE-SOURCE( $G, s$ )

```
1  for each vertex  $v \in G.V$ 
2       $v.d = \infty$ 
3       $v.\pi = \text{NIL}$ 
4       $s.d = 0$ 
```

*Initialize d[], which  
will converge to  
shortest-path value  $\delta$*

*Relaxation:  
Make  $|V|-1$  passes,  
relaxing each edge*

*Test for solution  
Under what condition  
do we get a solution?*

RELAX( $u, v, w$ )

```
1  if  $v.d > u.d + w(u, v)$ 
2       $v.d = u.d + w(u, v)$ 
3       $v.\pi = u$ 
```

# Bellman-Ford Algorithm

BELLMAN-FORD( $G, w, s$ )

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1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
2  for  $i = 1$  to  $|G.V| - 1$ 
3    for each edge  $(u, v) \in G.E$ 
4      RELAX( $u, v, w$ )
5  for each edge  $(u, v) \in G.E$ 
6    if  $v.d > u.d + w(u, v)$ 
7      return FALSE
8  return TRUE
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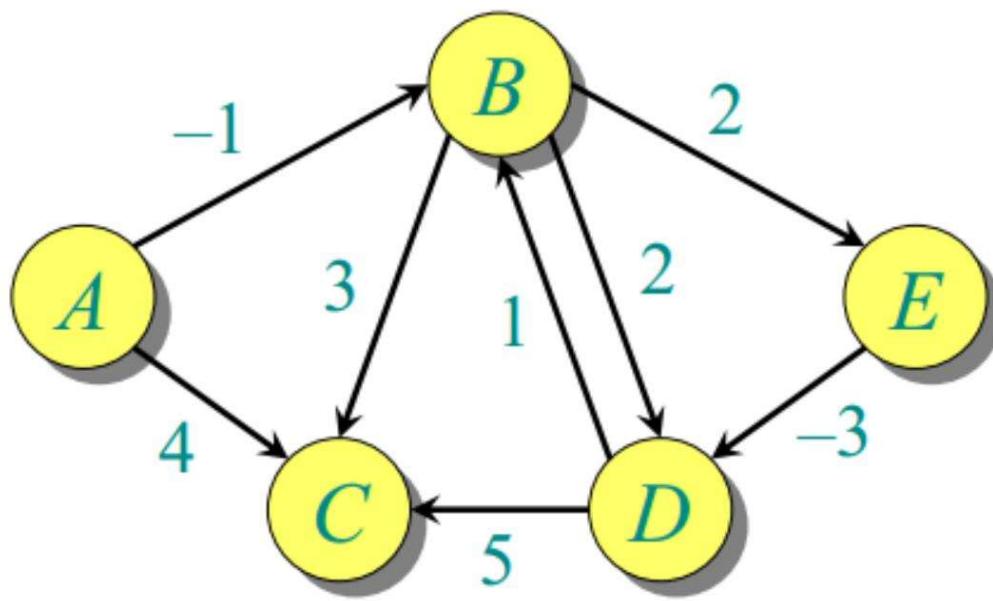
*Q: What will be the running time?*

*A:  $O(VE)$*

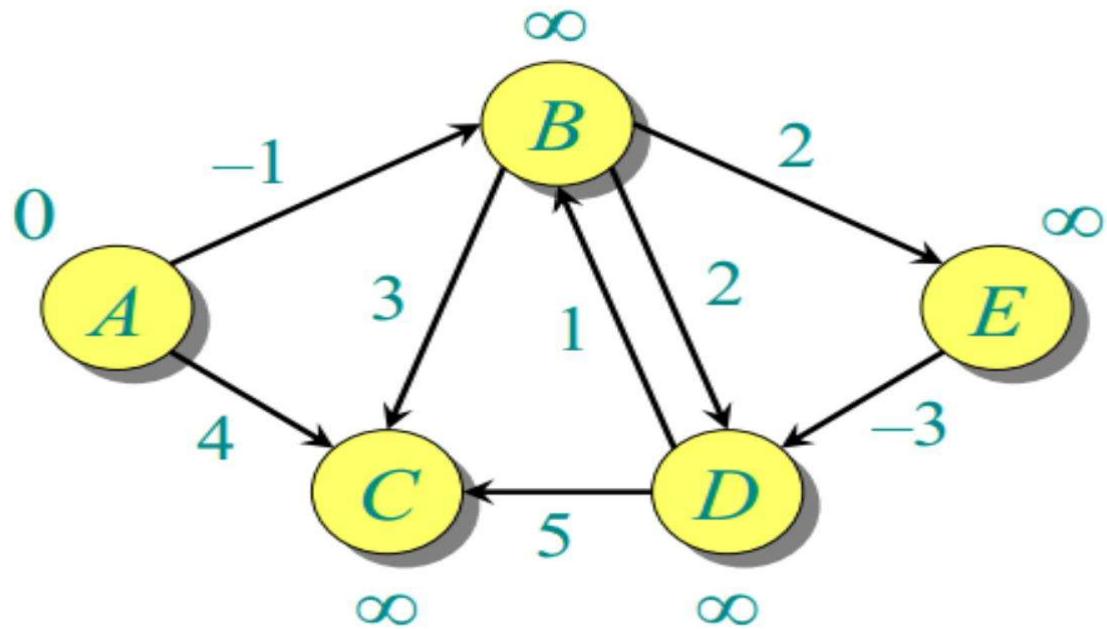
RELAX( $u, v, w$ )

```
1  if  $v.d > u.d + w(u, v)$ 
2     $v.d = u.d + w(u, v)$ 
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```

# Bellman-Ford Algorithm: Example



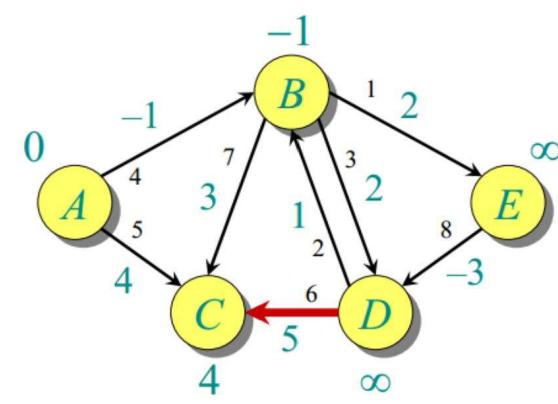
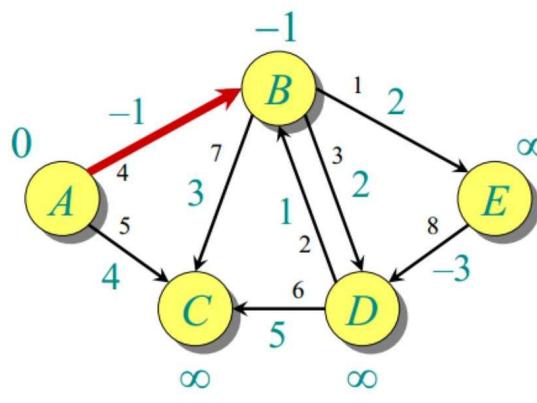
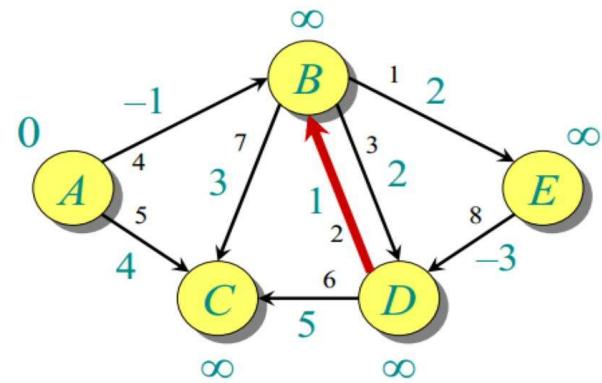
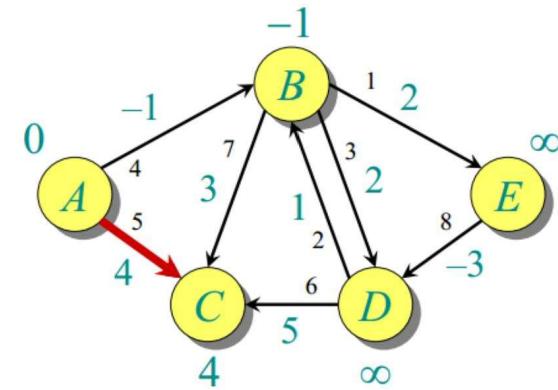
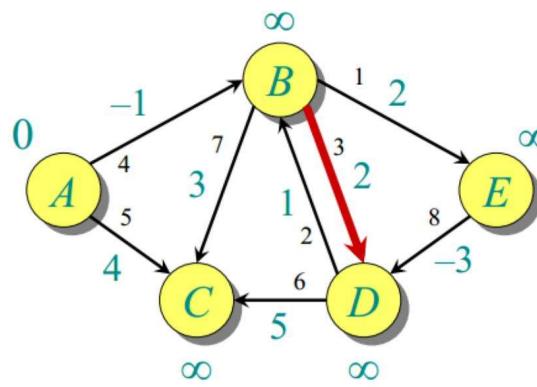
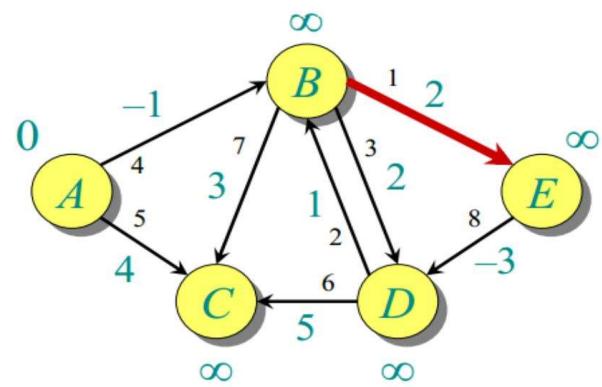
# Bellman-Ford Algorithm: Example



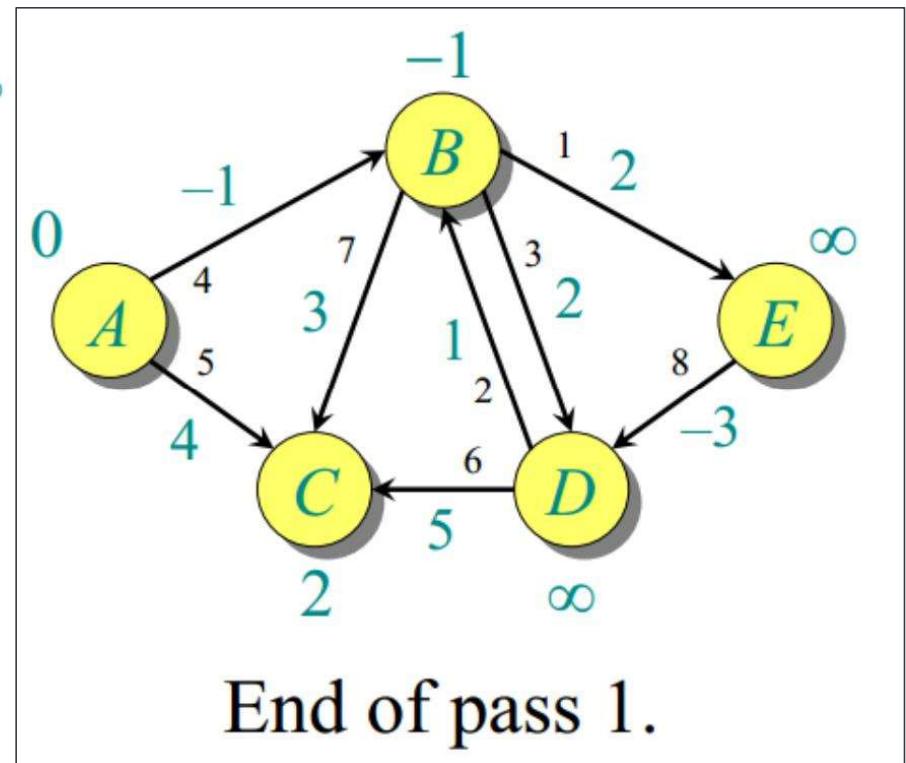
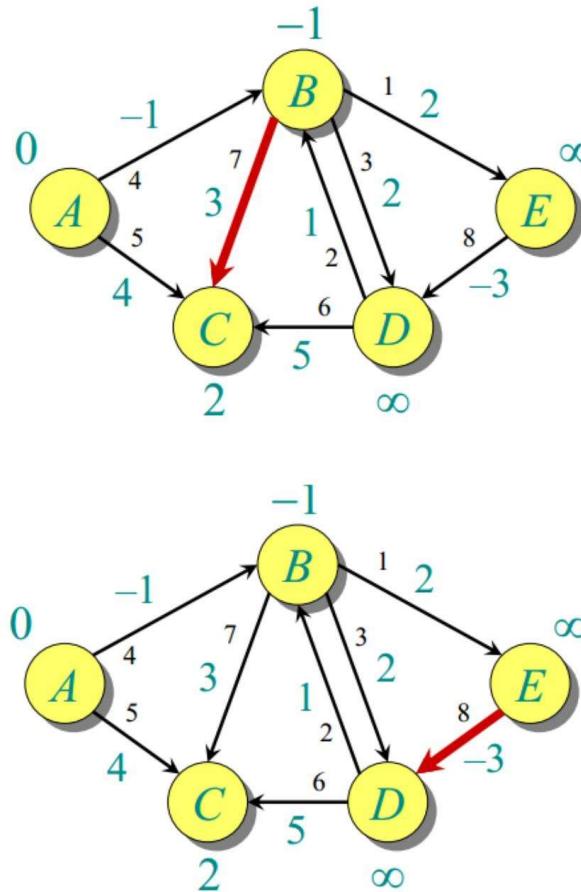
Initialization.

# Bellman-Ford Algorithm: Example

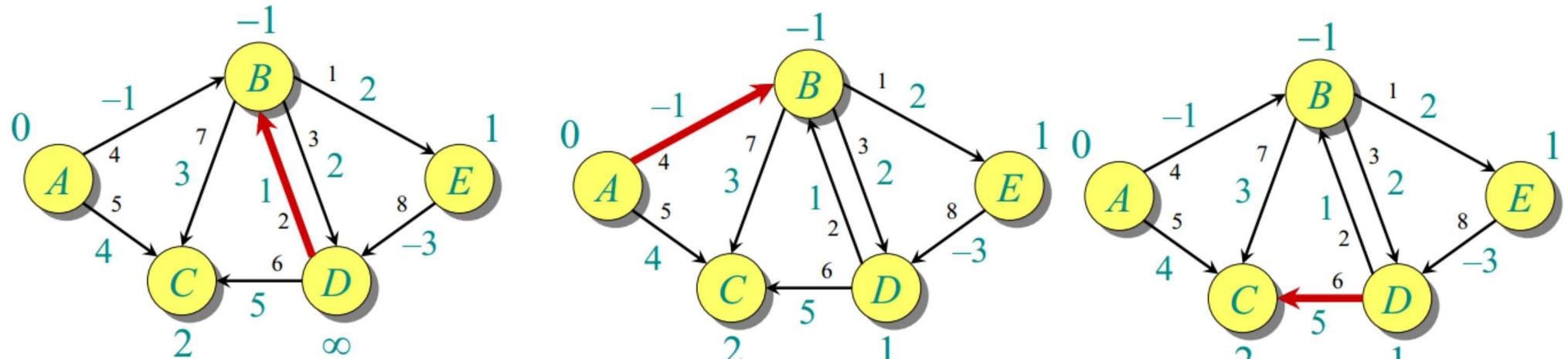
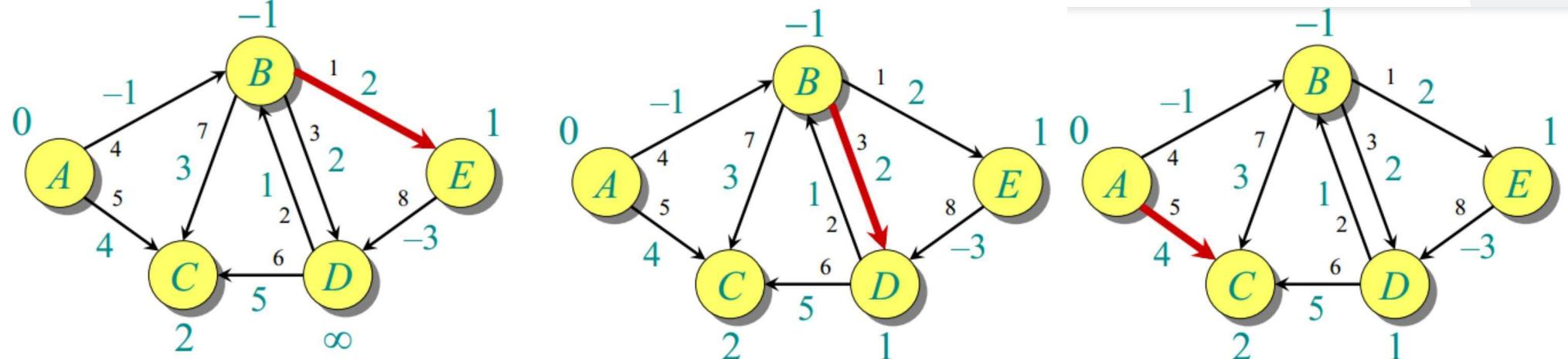
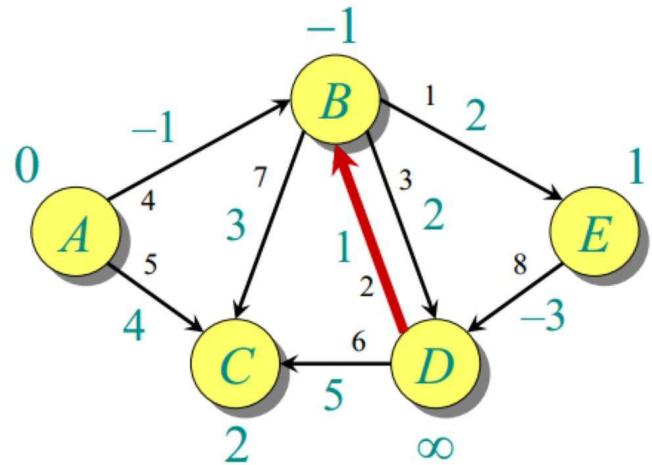
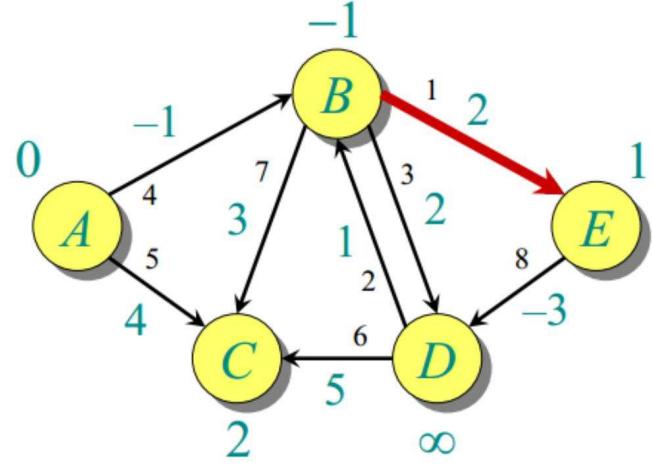
Order of edge relaxation.

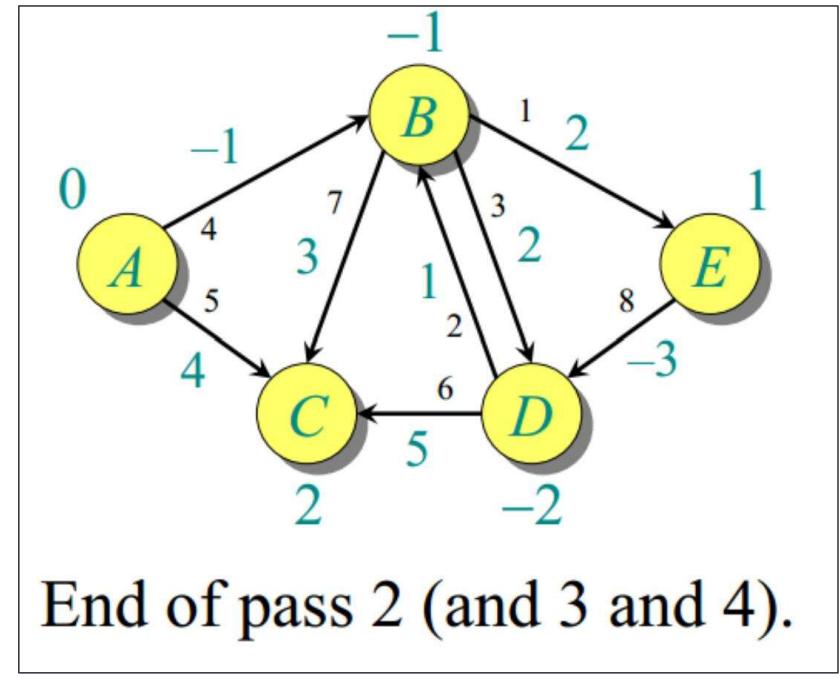
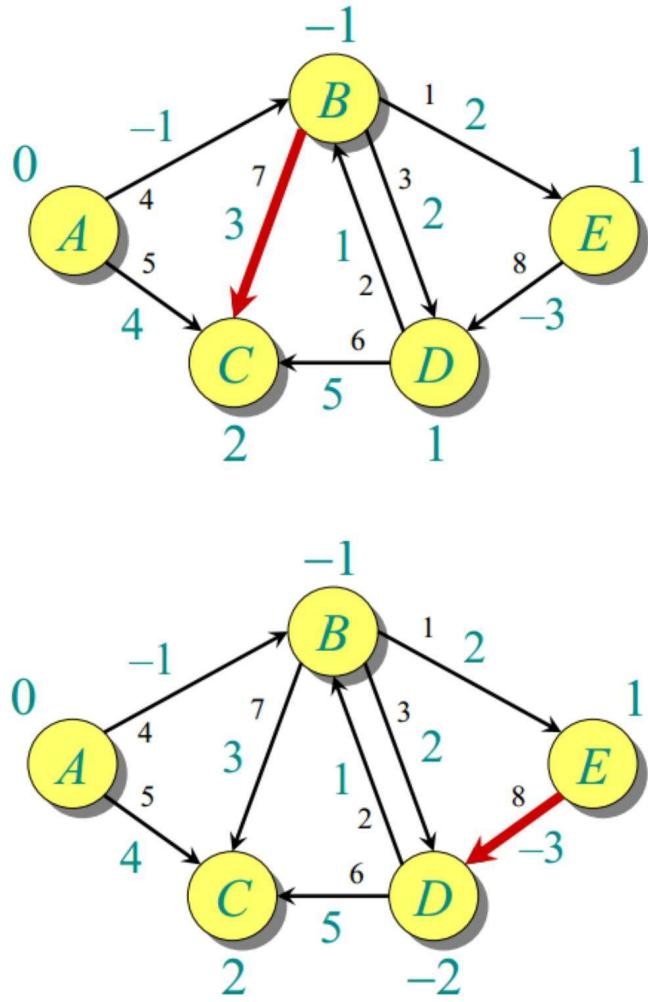


# Bellman-Ford Algorithm: Example

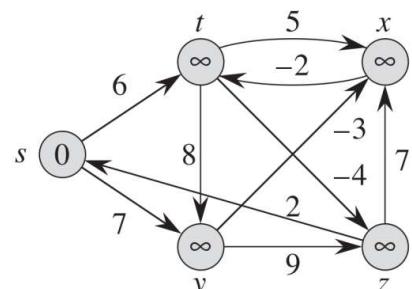


## Bellman-Ford Algorithm: Example

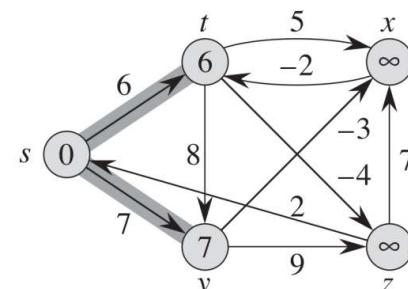




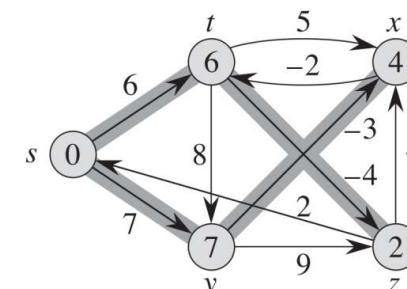
## Bellman-Ford Algorithm: Another Example



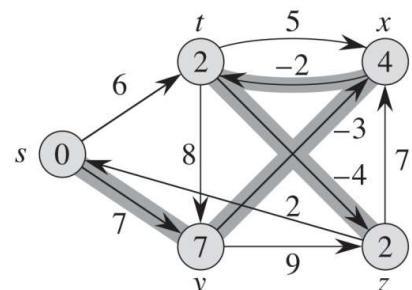
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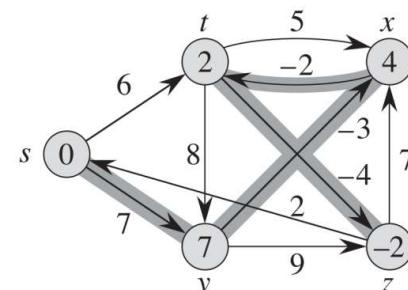
(b)



(c)



(d)



(e)

# Bellman-Ford Algorithm

- Prove: after  $|V|-1$  passes, all  $d$  values correct
  - Consider shortest path from  $s$  to  $v$ :  
 $s \rightarrow v_1 \rightarrow v_2 \rightarrow v_3 \rightarrow v_4 \rightarrow v$ 
    - Initially,  $d[s] = 0$  is correct, and doesn't change
    - After 1 pass through edges,  $d[v_1]$  is correct and doesn't change
    - After 2 passes,  $d[v_2]$  is correct and doesn't change
    - ...
    - Terminates in  $|V| - 1$  passes: (*Why?*)

---

Contents of this presentation are also based on  
Chapter-24, **Introduction to Algorithms** (3rd edition) Cormen, Leiserson, Rivest, & Stein

### Additional Reference

<https://www.cs.utexas.edu/~tandy/barrera>

MIT 6.046J/18.401J Introduction to Algorithms (Lecture 18 by Prof. Erik Demaine)



# THANK YOU

Stevens Institute of Technology