



# Shortest Path, Bellman-Ford algorithm, Dijkstra's algorithm, and Applications

by

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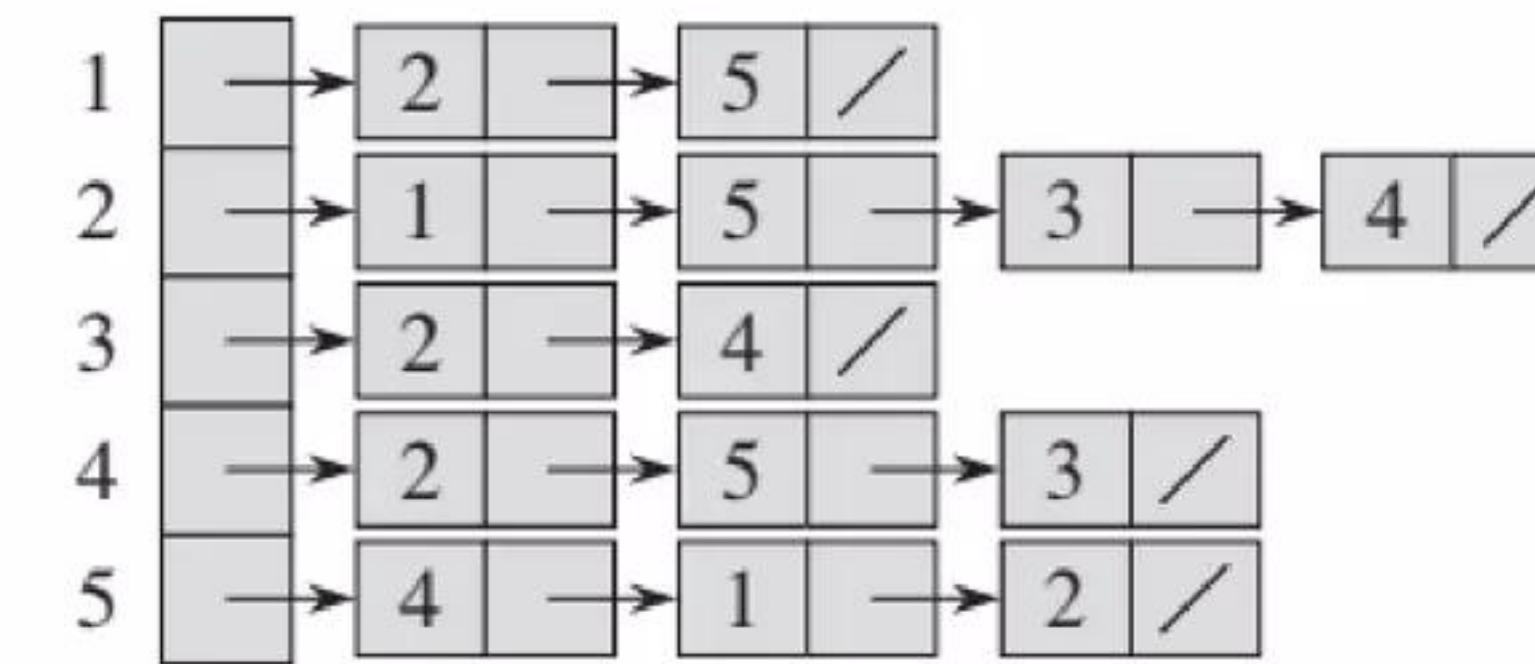
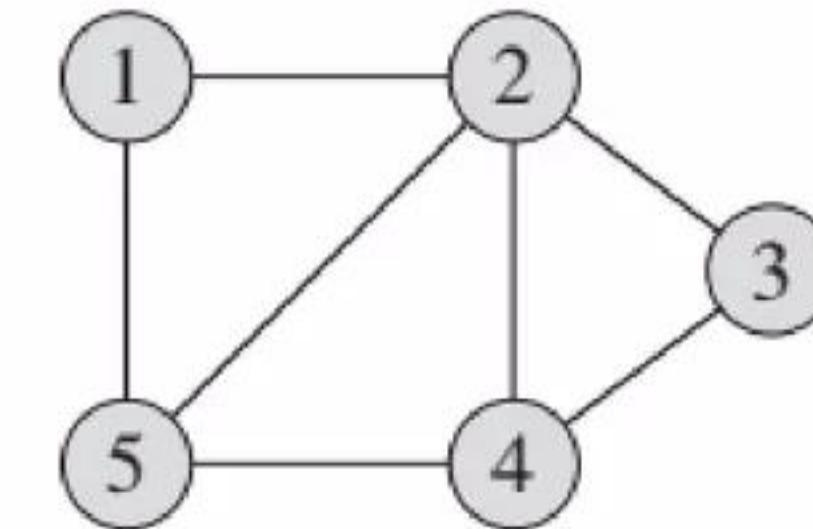


# Graph and Shortest Path

# Graph representations

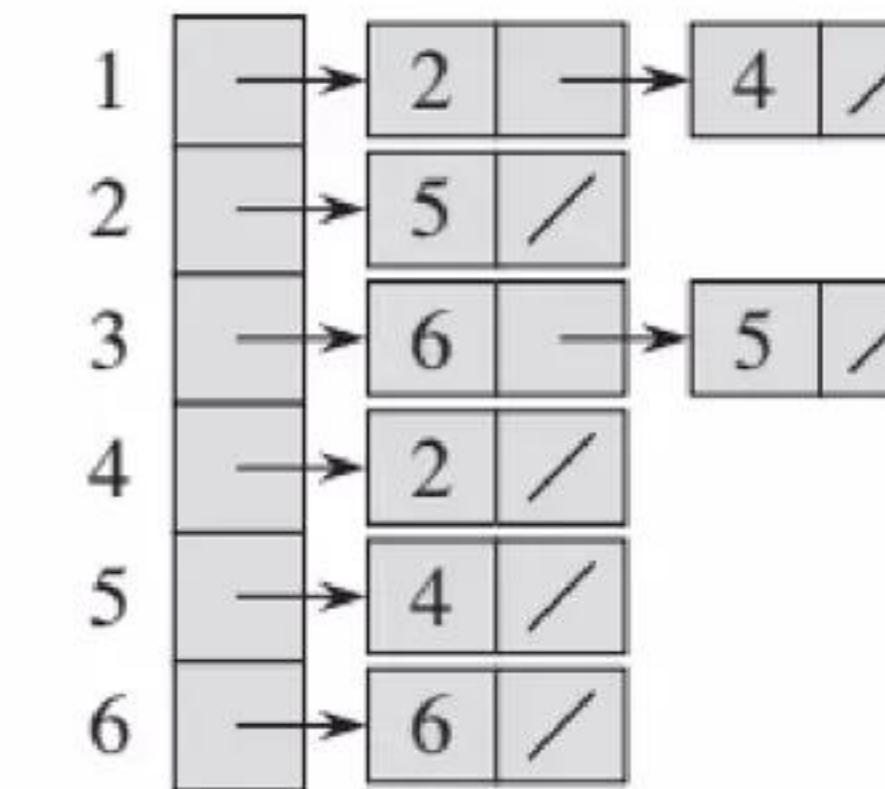
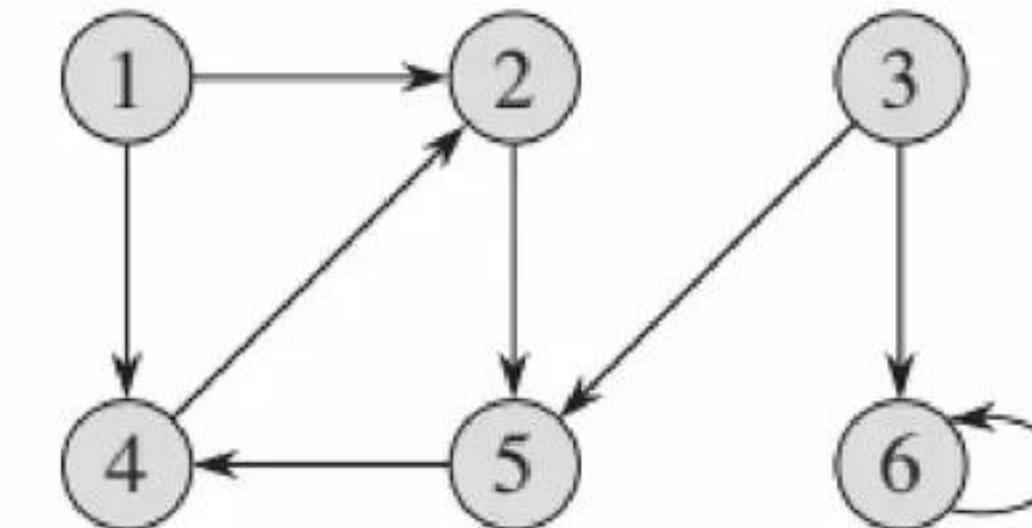
$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E, \\ 0 & \text{otherwise.} \end{cases}$$

- Adjacency-list and Adjacency-matrix representation of undirected graph G with 5 vertices and 7 edges.



	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

- Adjacency-list and Adjacency-matrix representations of a directed graph G with 6 vertices and 8 edges.



	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

# Shortest Path

- Given a road map of the India on which the distance between each pair of adjacent intersections is marked between route from Delhi to Mumbai, how can you find the shortest possible route?
- To examine an enormous number of possibilities, most of which are simply not worth considering!
- Model the road map as a graph vertices represent intersections, edges represent road segments between intersections, and edge weights represent road distances.
- For other examples weights can represent metrics other than distances, such as time, cost, penalties, loss, or any other quantity that accumulates.

# Shortest Path

- In a *shortest-paths problem*, given a weighted directed graph  $G = (V, E)$  with edges mapped to real-valued weights.
- The **weight**  $w(p)$  of path  $p = \langle v_0, v_1, \dots, v_k \rangle$  is the sum of the weights of its constituent edges:

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

We define the *shortest-path weight*  $\delta(u, v)$  from  $u$  to  $v$  by

$$\delta(u, v) = \begin{cases} \min\{w(p) : u \xrightarrow{p} v\} & \text{if there is a path from } u \text{ to } v, \\ \infty & \text{otherwise.} \end{cases}$$

A *shortest path* from vertex  $u$  to vertex  $v$  is then defined as any path  $p$  with weight  $w(p) = \delta(u, v)$ .

# Shortest Path Properties

- *Paths are directed.* A shortest path must respect the direction of its edges.
- *The weights are not necessarily distances.* Geometric intuition can be helpful, but the edge weights might represent time or cost.
- *Not all vertices need be reachable.* If  $t$  is not reachable from  $s$ , there is no path at all, and therefore there is no shortest path from  $s$  to  $t$ .
- *Negative weights introduce complications.*
- *Shortest paths are normally simple.*
- *Shortest paths are not necessarily unique.* There may be multiple paths of the lowest weight from one vertex to another; we are content to find any one of them.
- *Parallel edges and self-loops may be present.*

# Shortest Path: Variant

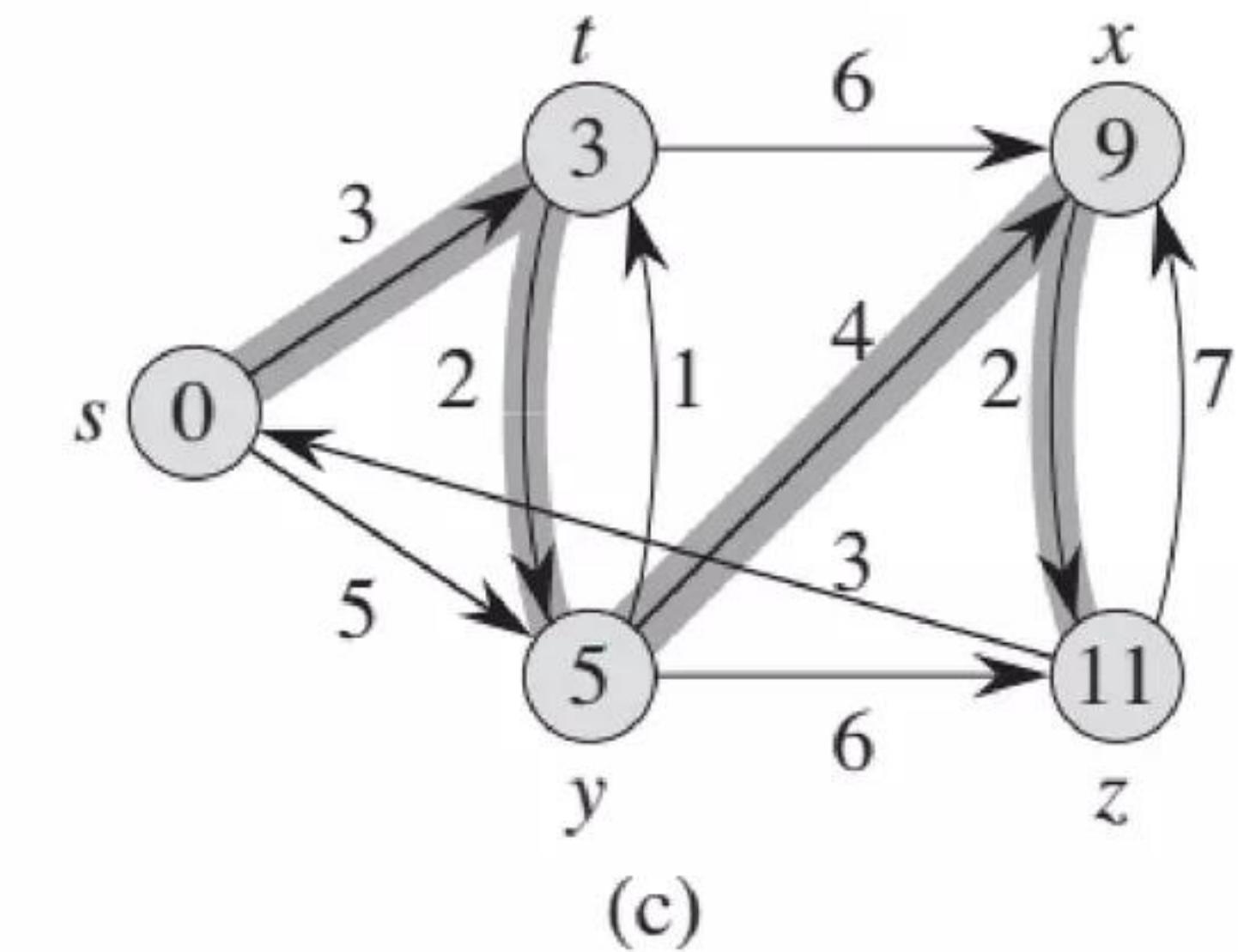
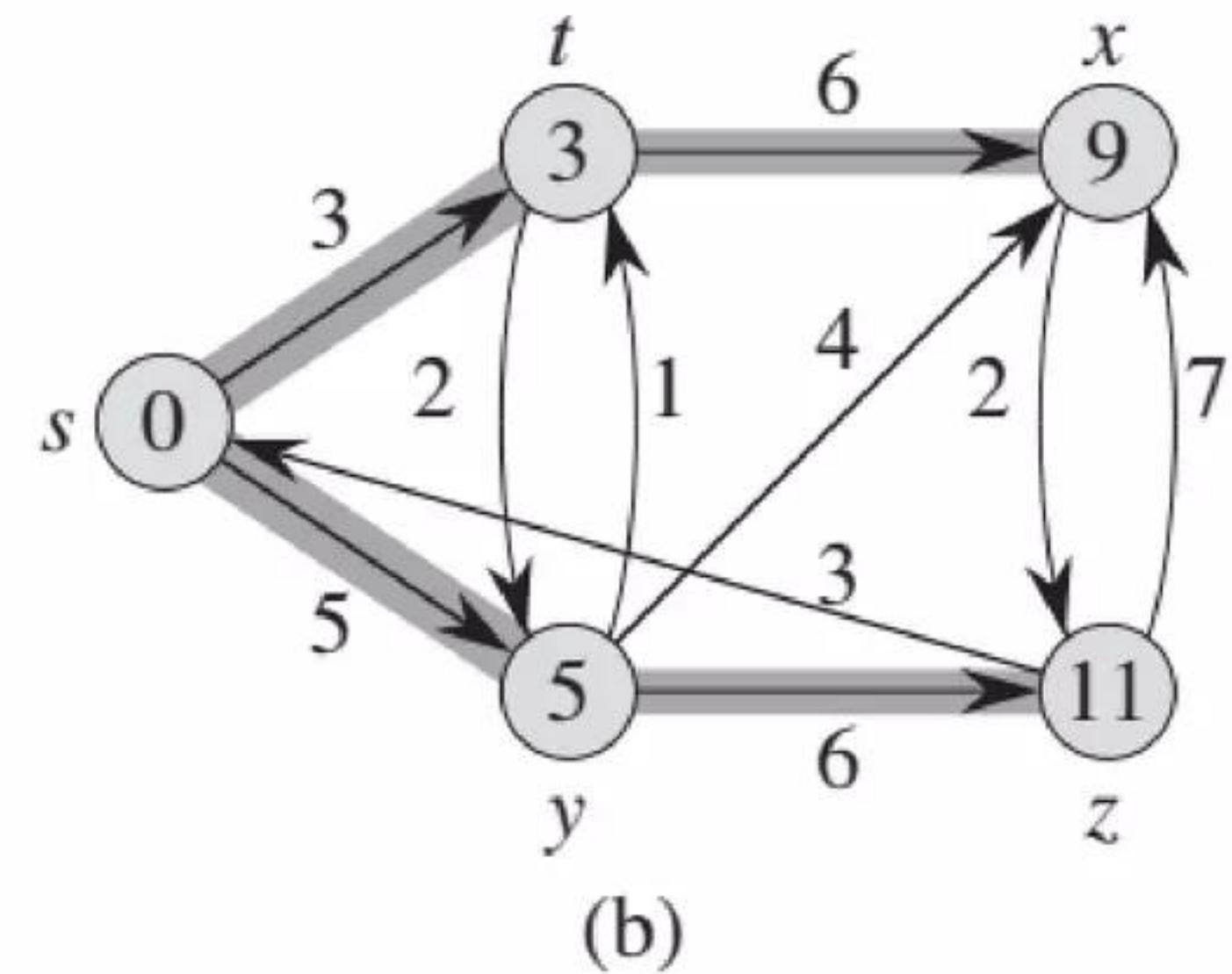
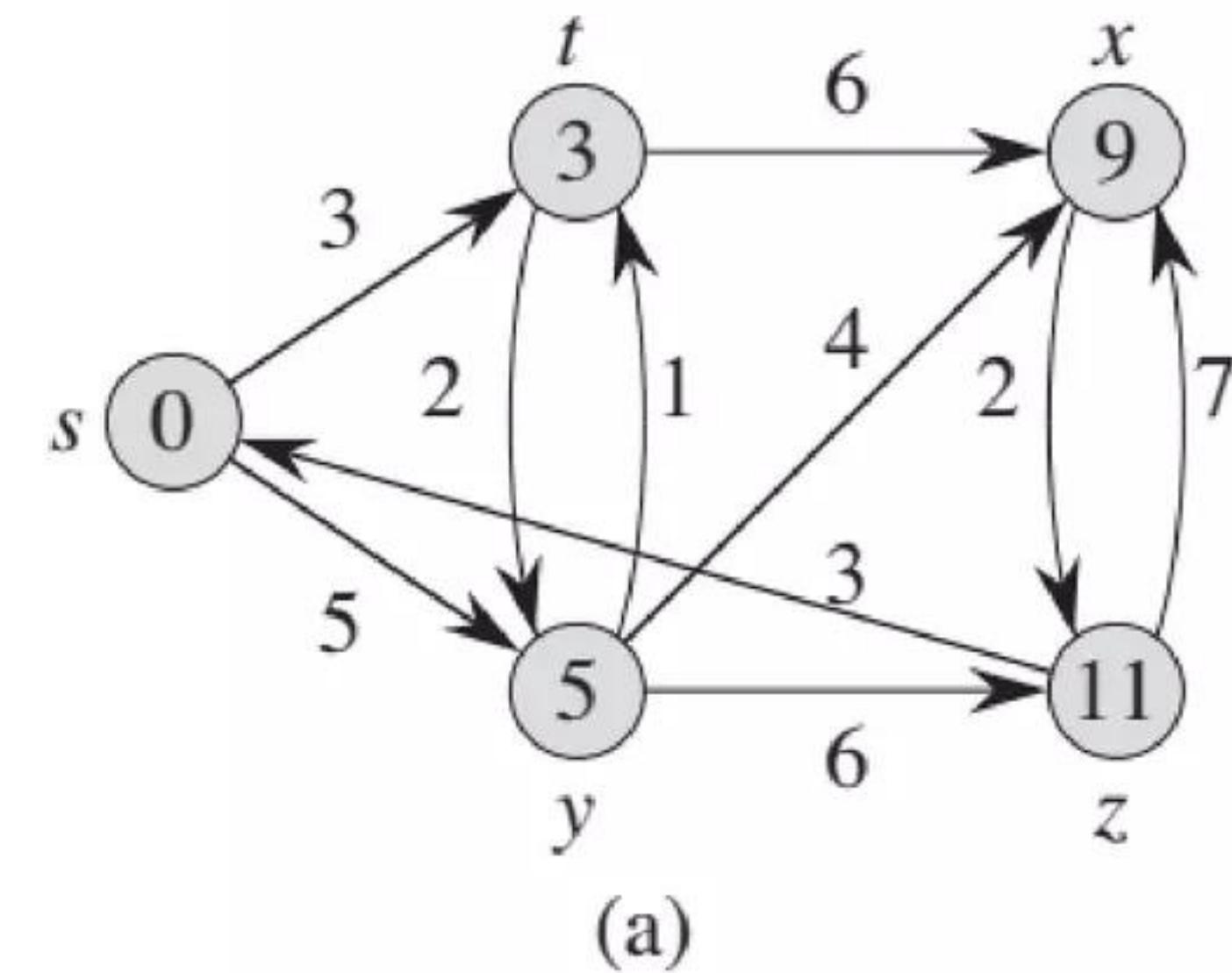
- **Single-destination shortest-paths problem:**
  - Find a shortest path to a given *destination* vertex  $t$  from each vertex  $v$ .
  - Reversing the direction of each edge → reduce this problem to a single-source problem.
- **Single-pair shortest-path problem:**
  - Find a shortest path from  $u$  to  $v$  for given vertices  $u$  and  $v$ , where source vertex is  $u$
- **All-pairs shortest-paths problem:**
  - Find a shortest path from  $u$  to  $v$  for every pair of vertices  $u$  and  $v$ .
  - Solve this problem by running a single source algorithm once from each vertex.

# Shortest Paths Tree

- A *shortest-paths tree* rooted at  $s$  is a directed subgraph  $G' = (V', E')$ , where  $V' \subseteq V$  and  $E' \subseteq E$ , such that
  1.  $V'$  is the set of vertices reachable from  $s$  in  $G$ ,
  2.  $G'$  forms a rooted tree with root  $s$ , and
  3. for all  $v \in V'$ , the unique simple path from  $s$  to  $v$  in  $G'$  is a shortest path from  $s$  to  $v$  in  $G$ .

# Shortest Paths Tree

- Shortest paths are not necessarily unique, and neither are shortest-paths trees.
- A weighted, directed graph and two shortest-paths trees with the same root.



# Shortest Path Algorithms

- Bellman-Ford algorithm
  - Negative weights are allowed
  - Negative cycles reachable from the source are not allowed.
- Dijkstra's algorithm
  - Negative weights are not allowed
- Operations common in both algorithms:
  - Initialization
  - Relaxation

# Shortest Path Algorithms Initialization

- For each vertex  $v \in V$ , we maintain an attribute  $v.d$ , which is an upper bound on the weight of a shortest path from source  $s$  to  $v$ .
- We call  $v.d$  a shortest-path estimate.
- We initialize the shortest-path estimates and predecessors by the following  $O(V)$  time procedure:

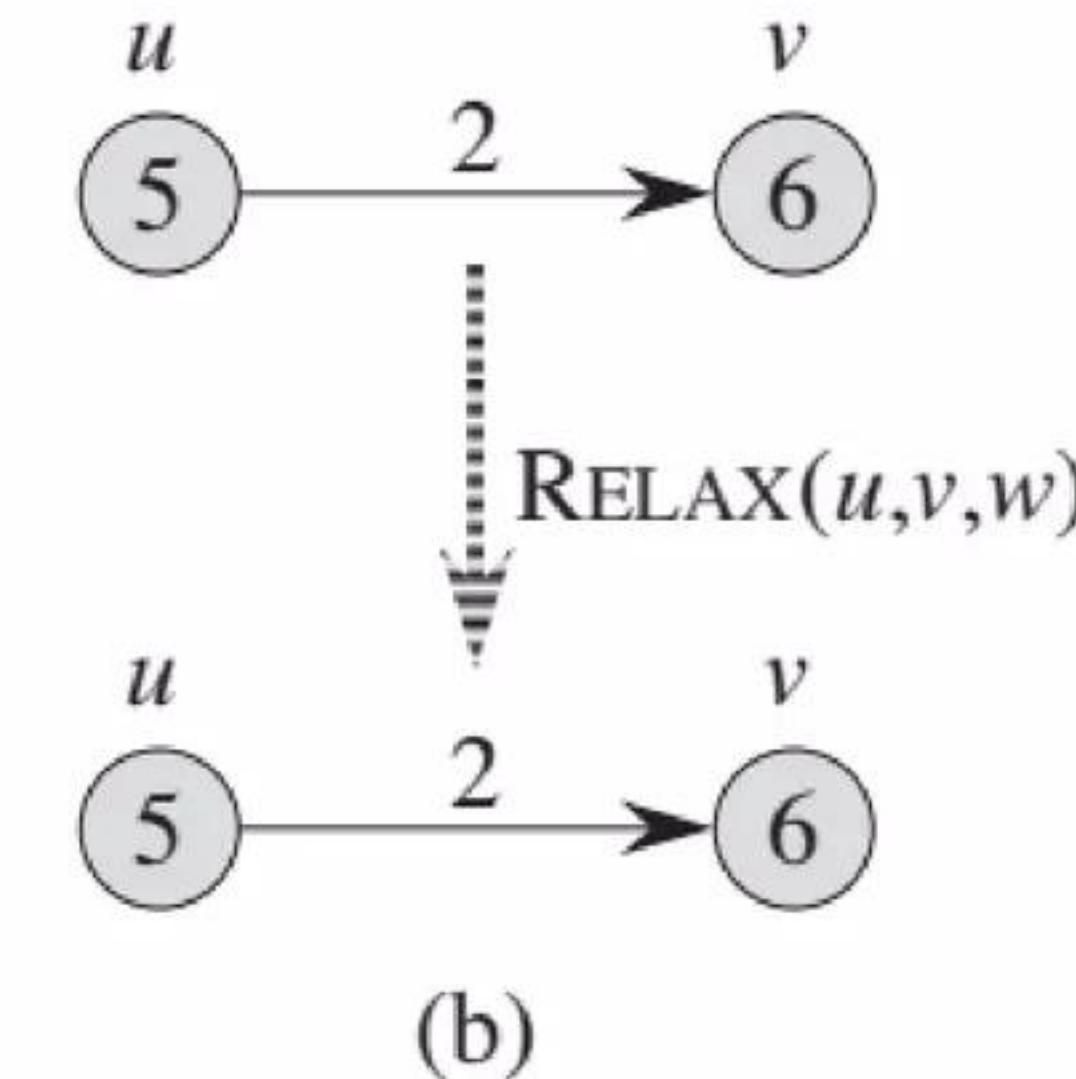
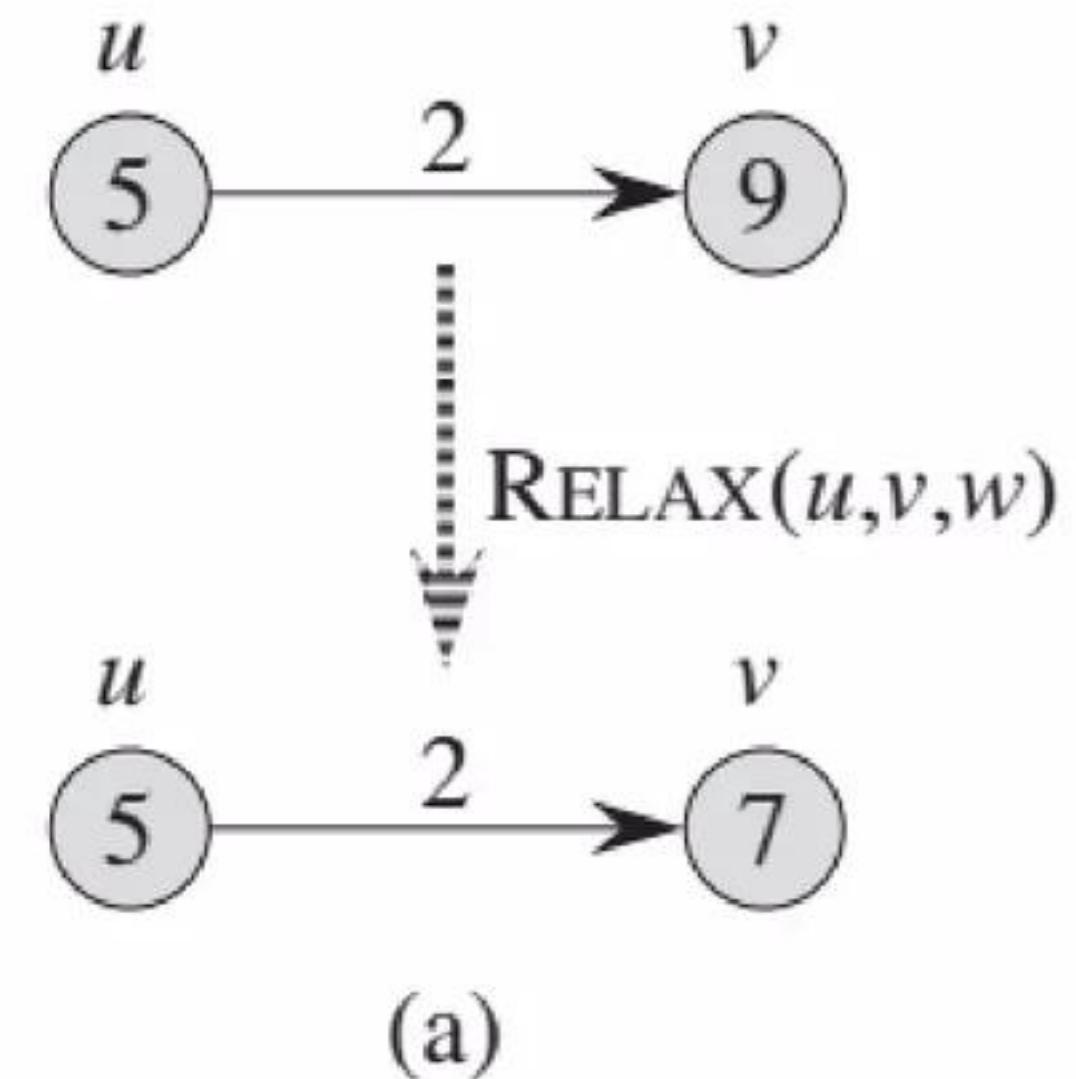
INITIALIZE-SINGLE-SOURCE( $G, s$ )

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

# Shortest Path Algorithms Relaxation

Relaxing an edge  $(u, v)$  with weight  $w(u, v) = 2$ . The shortest-path estimate of each vertex appears within the vertex.

- (a) Because  $v.d > u.d + w(u, v)$  prior to relaxation, the value of  $v.d$  decreases.
- (b) Here,  $v.d \leq u.d + w(u, v)$  before relaxing the edge, and so the relaxation step leaves  $v.d$  unchanged.



# Shortest Path Algorithms Relaxation

- Relaxation is the only means by which shortest path estimates and predecessors change.
- Algorithms differ in how many times they relax each edge and the order in which they relax edges.
- Dijkstra's algorithm and the shortest-paths algorithm for directed acyclic graphs relax each edge exactly once.
- The Bellman-Ford algorithm relaxes each edge  $|V| - 1$  times.

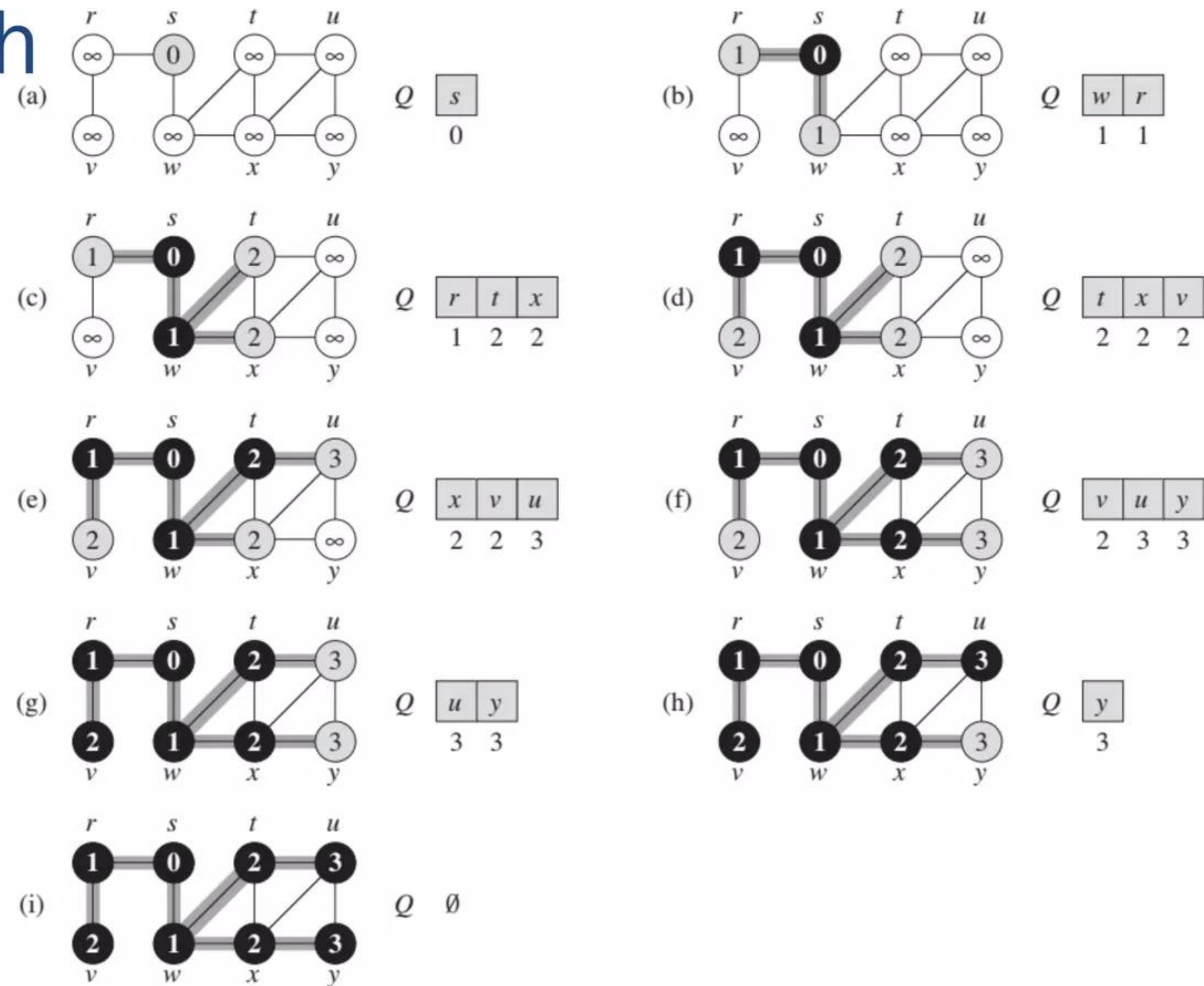
```
RELAX( $u, v, w$ )  
1  if  $v.d > u.d + w(u, v)$   
2       $v.d = u.d + w(u, v)$   
3       $v.\pi = u$ 
```

# Breadth First Search

BFS algorithm  
is a shortest-  
paths algorithm  
that works on  
unweighted  
graphs, that is,  
graphs in which  
each edge has  
unit weight.

```

BFS( $G, s$ )
1   for each vertex  $u \in G.V - \{s\}$ 
2      $u.color = \text{WHITE}$ 
3      $u.d = \infty$ 
4      $u.\pi = \text{NIL}$ 
5    $s.color = \text{GRAY}$ 
6    $s.d = 0$ 
7    $s.\pi = \text{NIL}$ 
8    $Q = \emptyset$ 
9   ENQUEUE( $Q, s$ )
10  while  $Q \neq \emptyset$ 
11     $u = \text{DEQUEUE}(Q)$ 
12    for each  $v \in G.Adj[u]$ 
13      if  $v.color == \text{WHITE}$ 
14         $v.color = \text{GRAY}$ 
15         $v.d = u.d + 1$ 
16         $v.\pi = u$ 
17        ENQUEUE( $Q, v$ )
18     $u.color = \text{BLACK}$ 
```



# Bellman-Ford Algorithm

# Bellman-Ford Algorithm

- Single-source shortest path problem
  - Computes  $d(s, v)$  and  $\pi.v$  for all  $v \in V$
- Allows negative edge weights - can detect negative cycles.
  - Returns TRUE if no negative-weight cycles are reachable from the source s
  - Returns FALSE otherwise  $\Rightarrow$  no solution exists

# Bellman-Ford Algorithm

- After initializing the  $d$  and  $\pi$  values of all vertices in line 1,
- The algorithm makes  $|V| - 1$  passes over the edges of the graph. Each pass is one iteration of the **for** loop of lines 2–4 and consists of relaxing each edge of the graph once.
- After making  $|V| - 1$  passes, lines 5–8 check for a negative-weight cycle and return the appropriate Boolean value.

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
```

# Bellman-Ford Algorithm

- Runs in time  $O(VE)$ ,
- Initialization takes  $O(V)$  time,
- Each  $|V| - 1$  passes over the edges takes  $O(V)$  time
  - For loop takes  $O(E)$  time
    - Relax will take  $O(VE)$  times
- Running time:
  - $O(V+VE+E) = O(VE)$

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

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

# Bellman-Ford Algorithm example

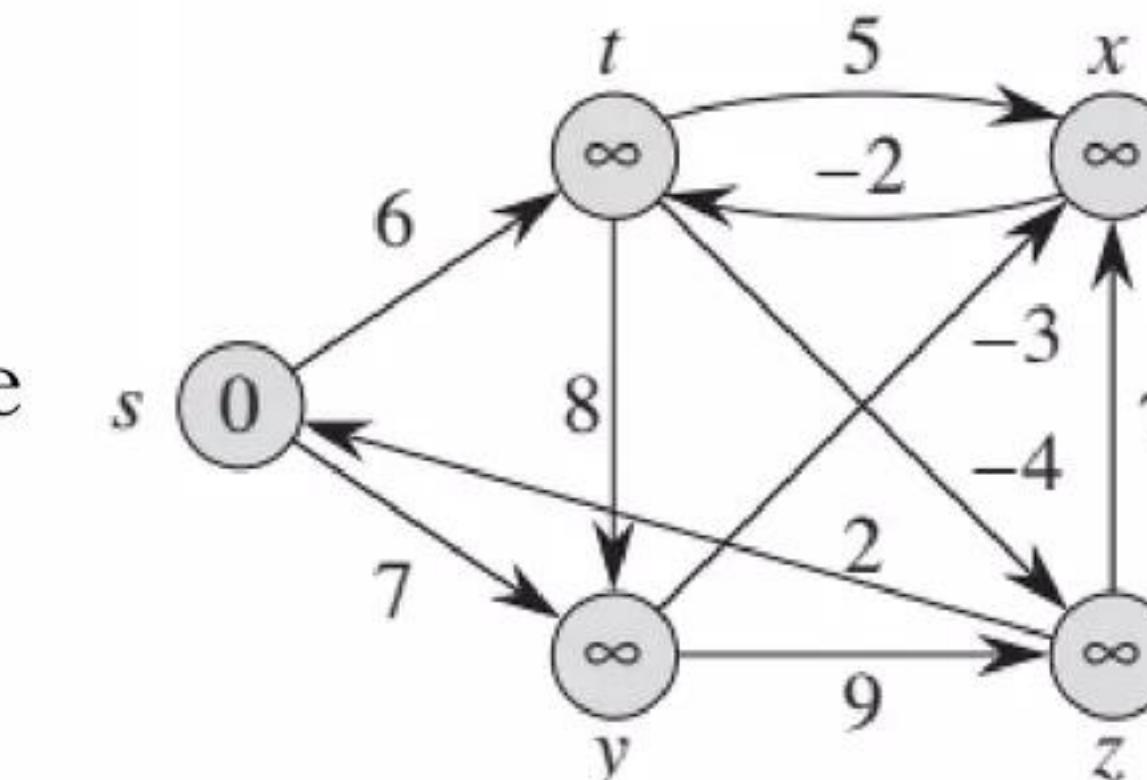
- Source is vertex  $s$
- shaded edges indicate predecessor values: if edge  $(u, v)$  is shaded, then
  - $\pi_v = u$
- Each pass relaxes the edges in the order  $(t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)$

(a) Initialization

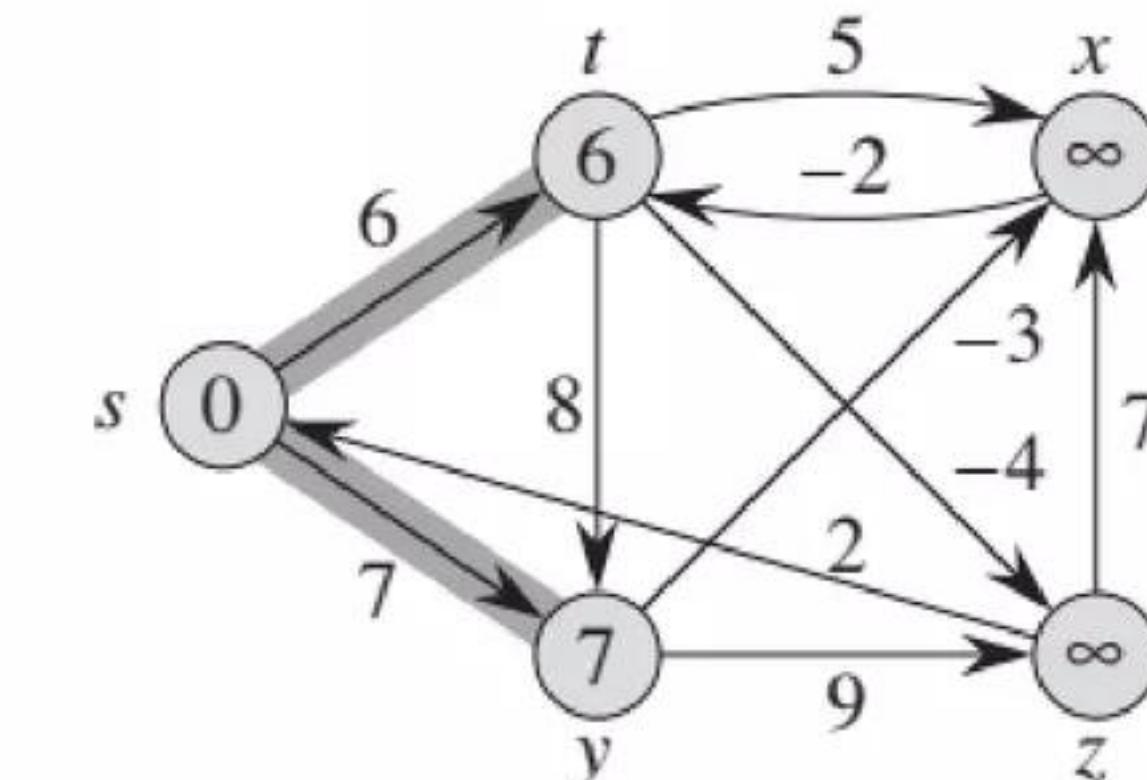
(b)–(e) Each successive pass  $|V| - 1 = 5 - 1 = 4$  over edges

(e) The  $d$  and  $\pi$  values are the final values

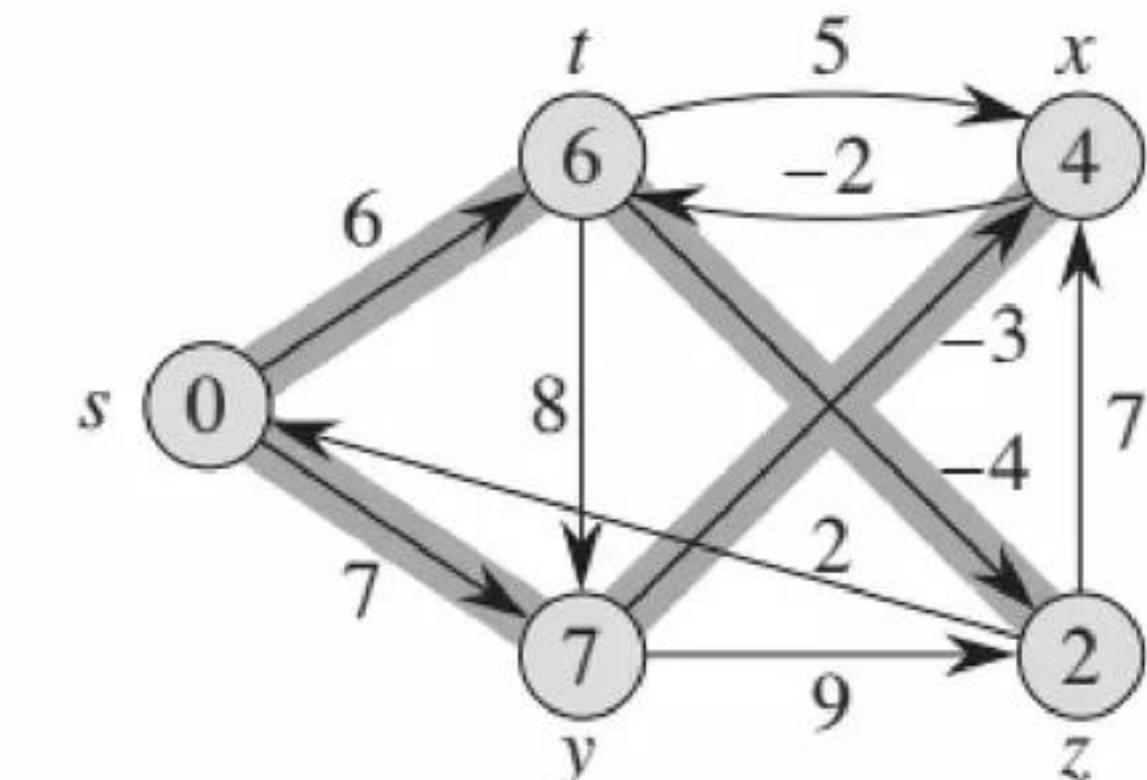
The Bellman-Ford algorithm returns TRUE



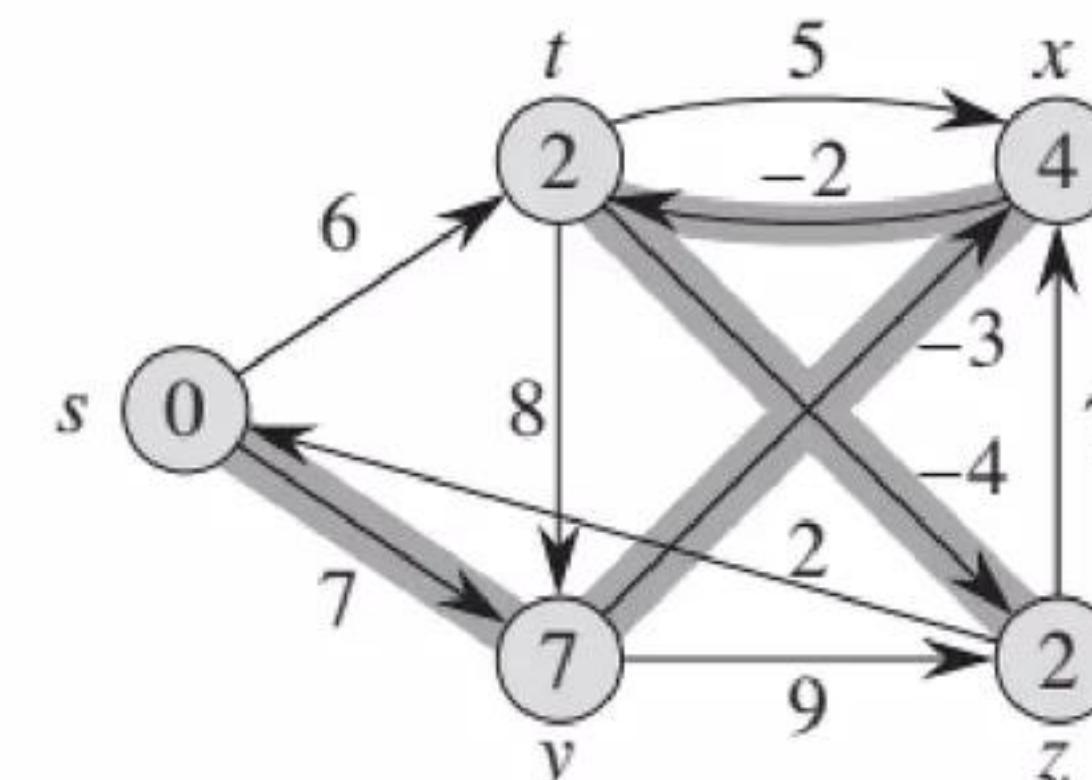
(a)



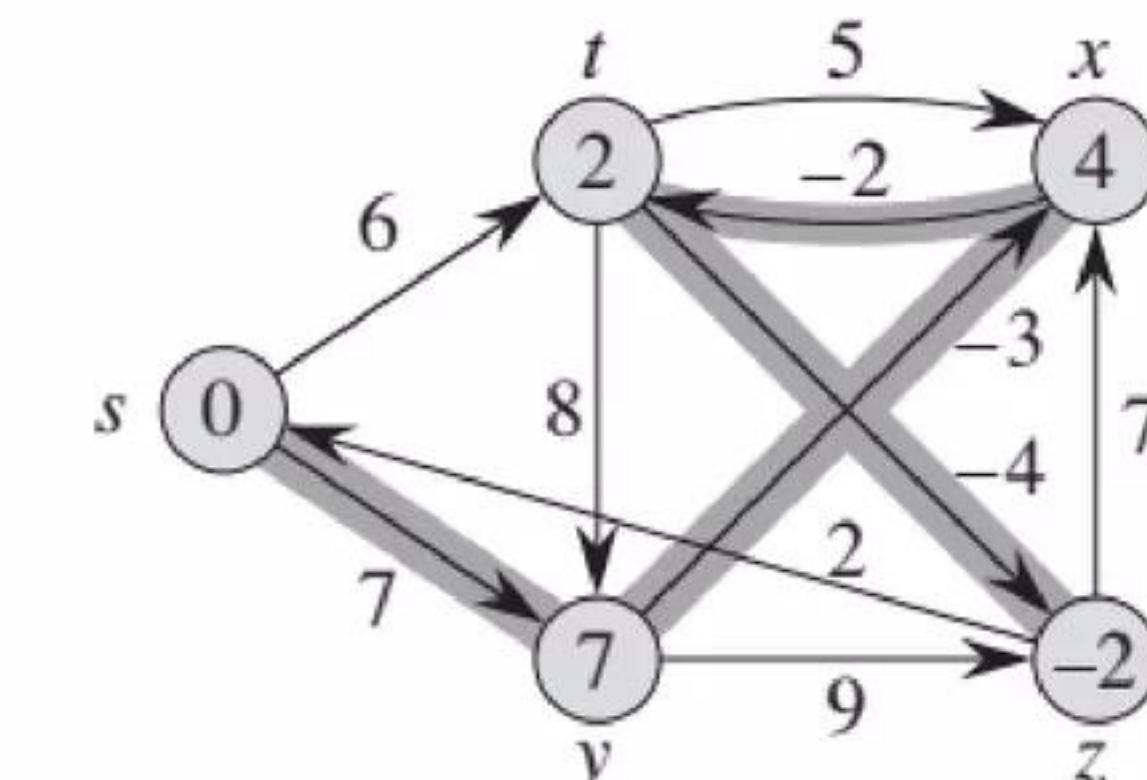
(b)



(c)



(d)

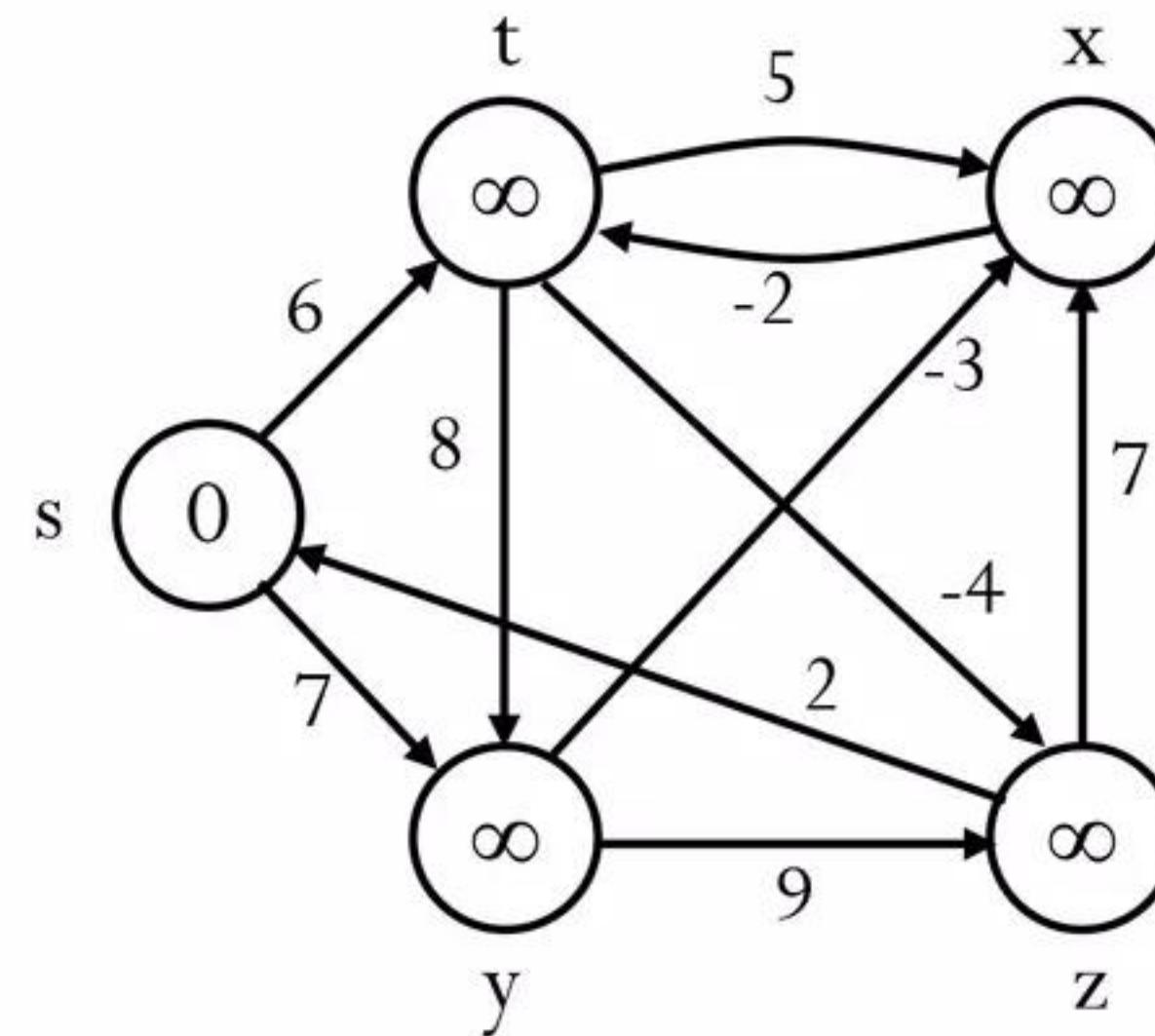


(e)

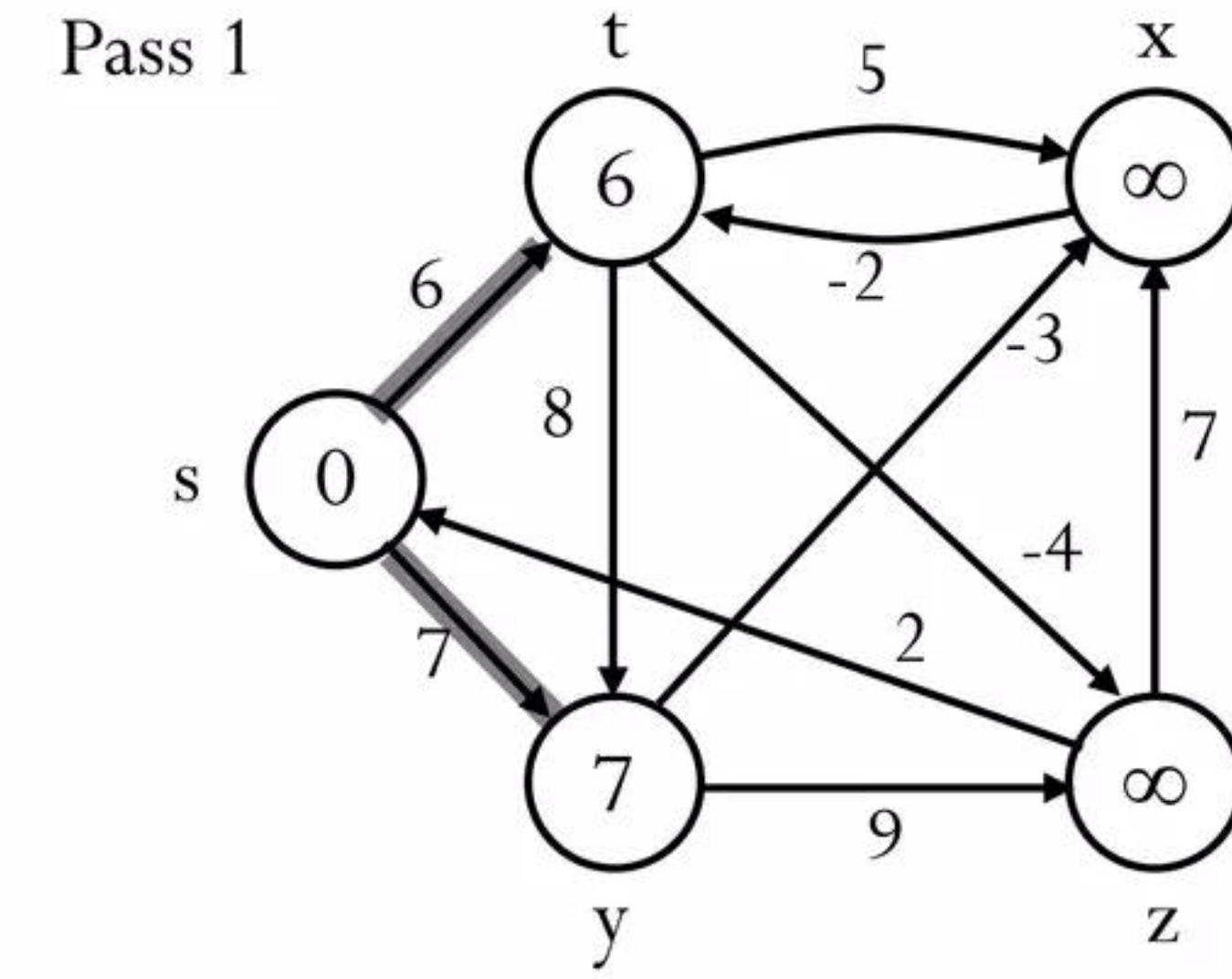
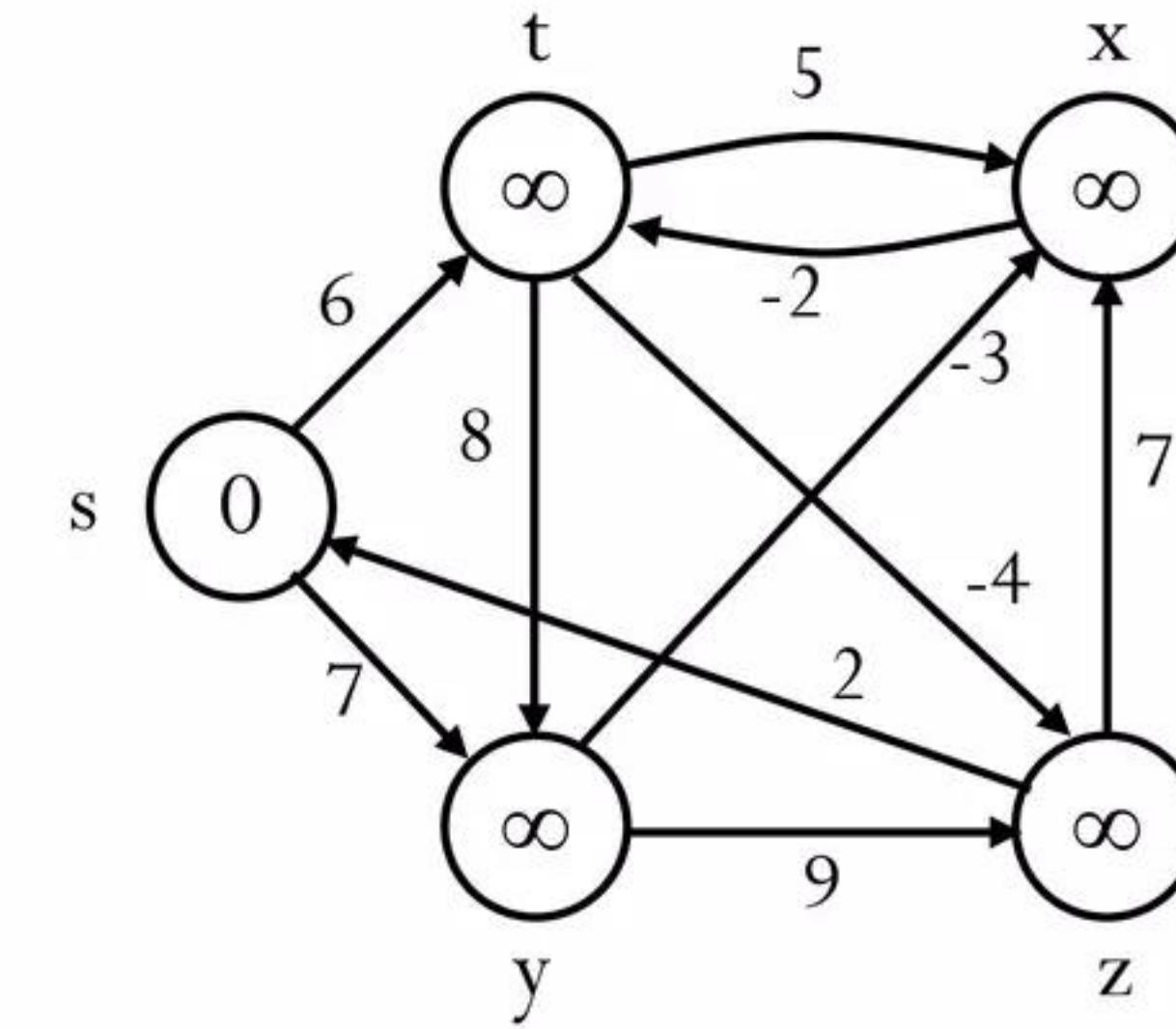
# Bellman-Ford Algorithm

- Each edge is relaxed  $|V| - 1$  times by making  $|V| - 1$  passes over the whole edge set.
- To make sure that each edge is relaxed exactly  $|V - 1|$  times, it puts the edges in an unordered list and goes over the list  $|V - 1|$  times.

$(t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)$



# BELLMAN-FORD( $V$ , $E$ , $w$ , $s$ )

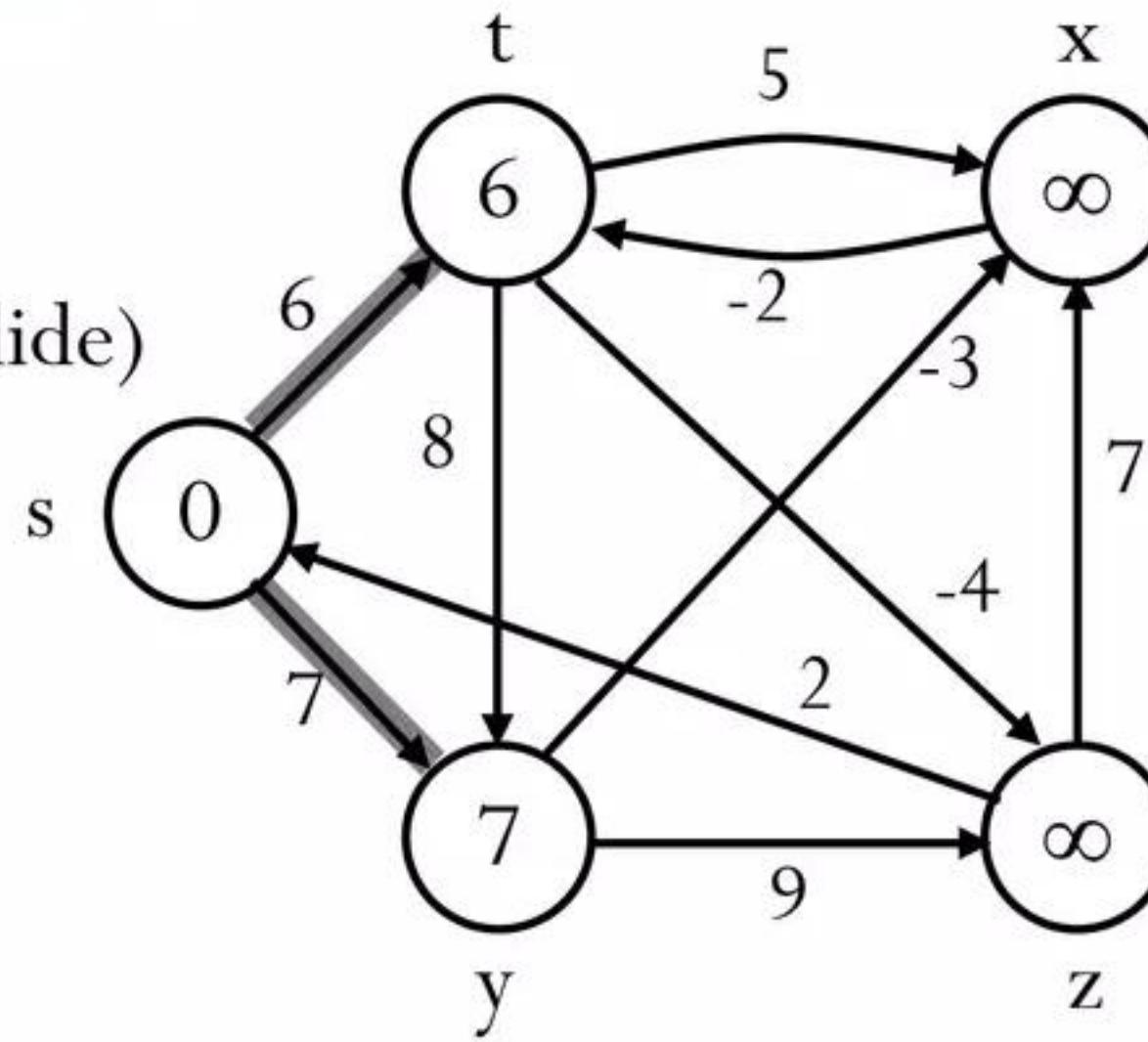


$E: (t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)$

# Example

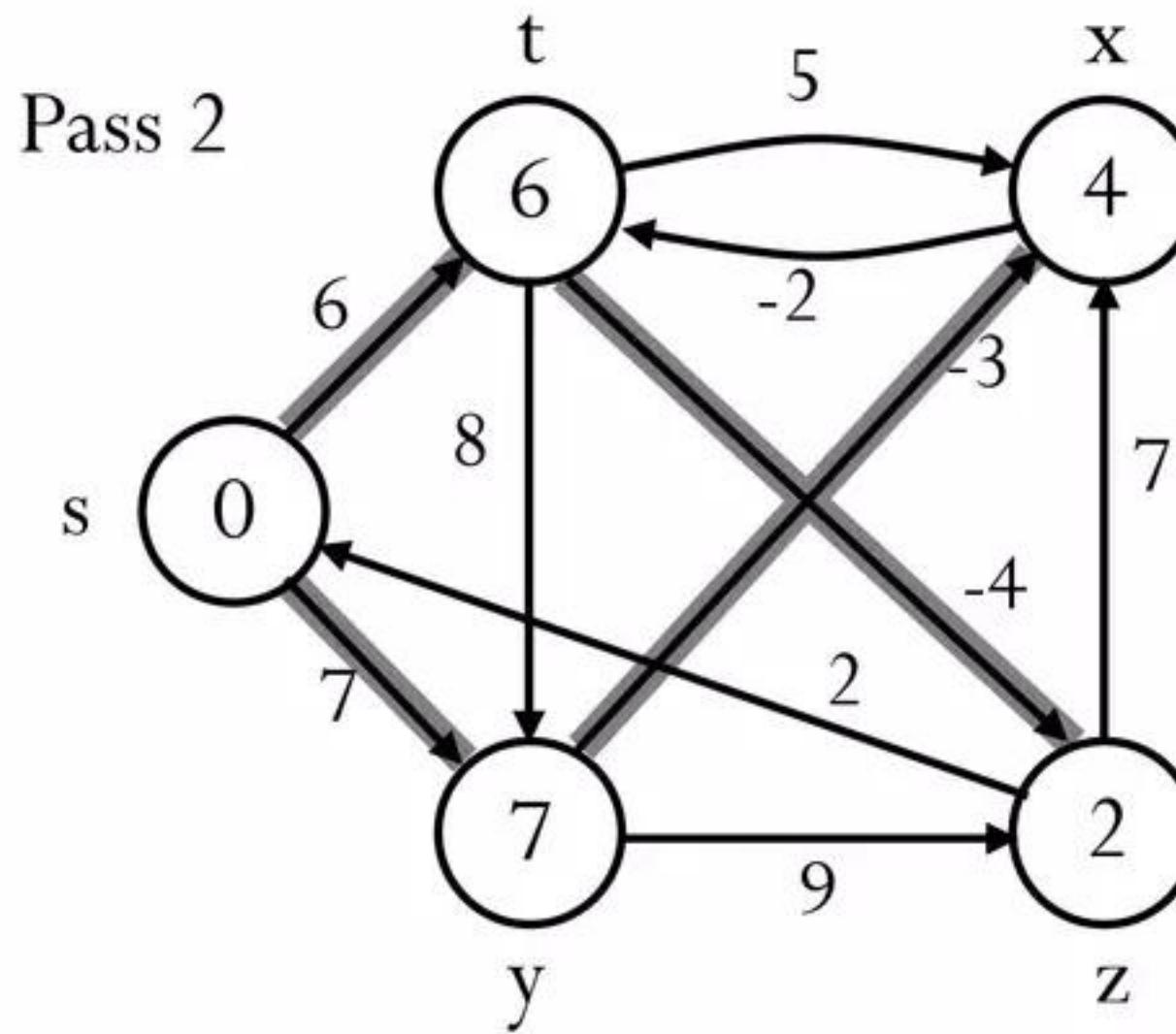
Pass 1

(from previous slide)

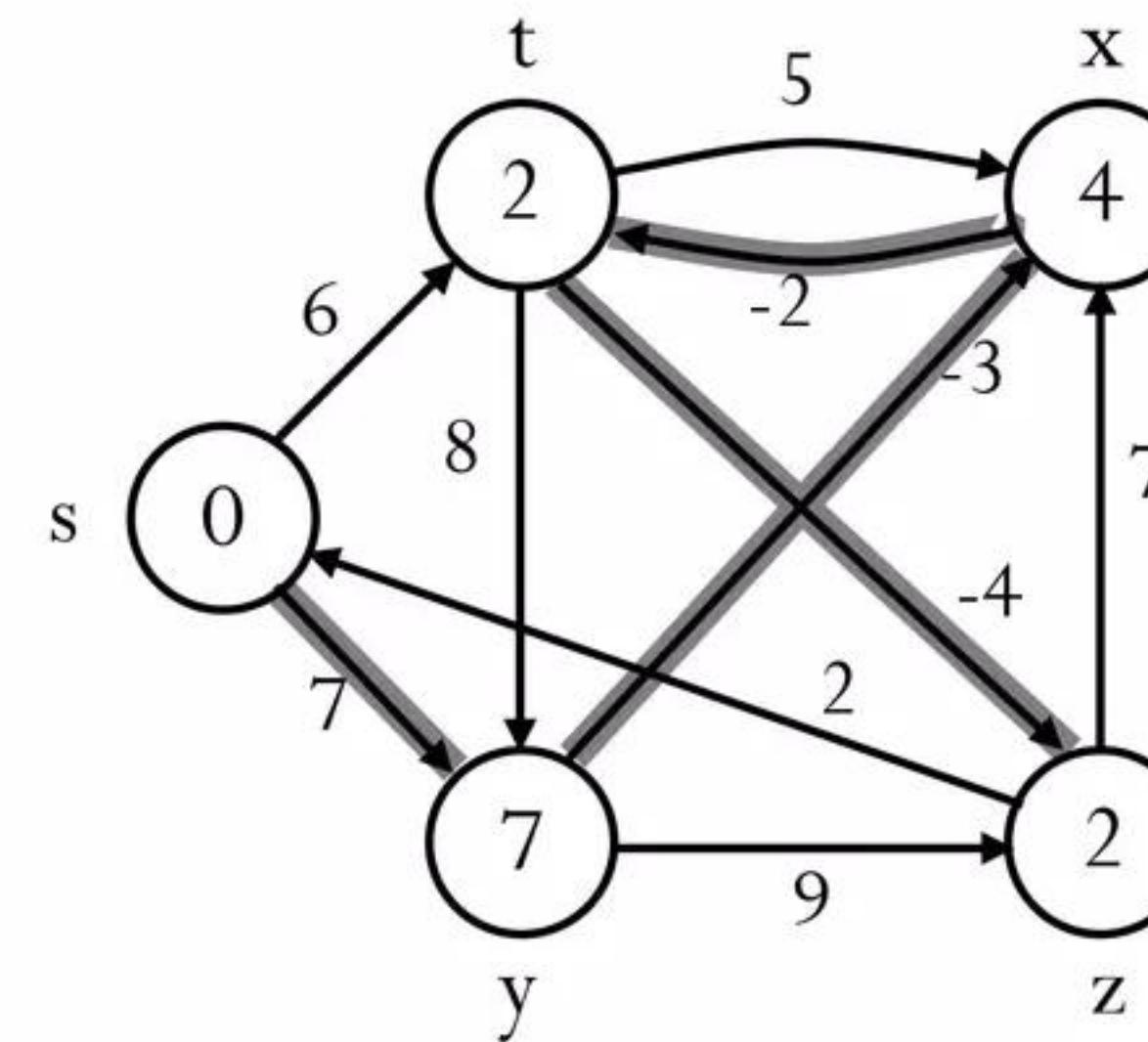


$(t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)$

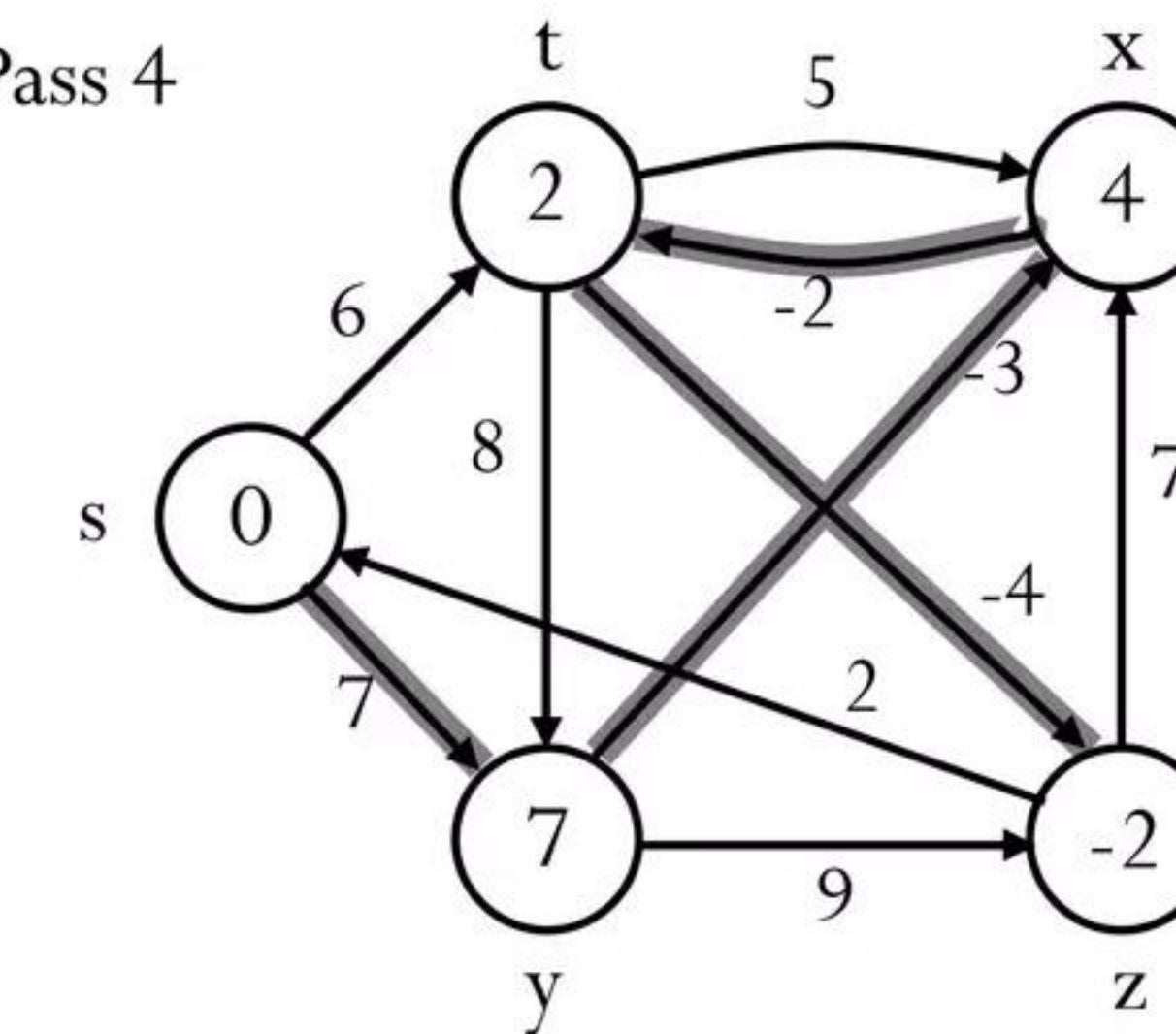
Pass 2



Pass 3

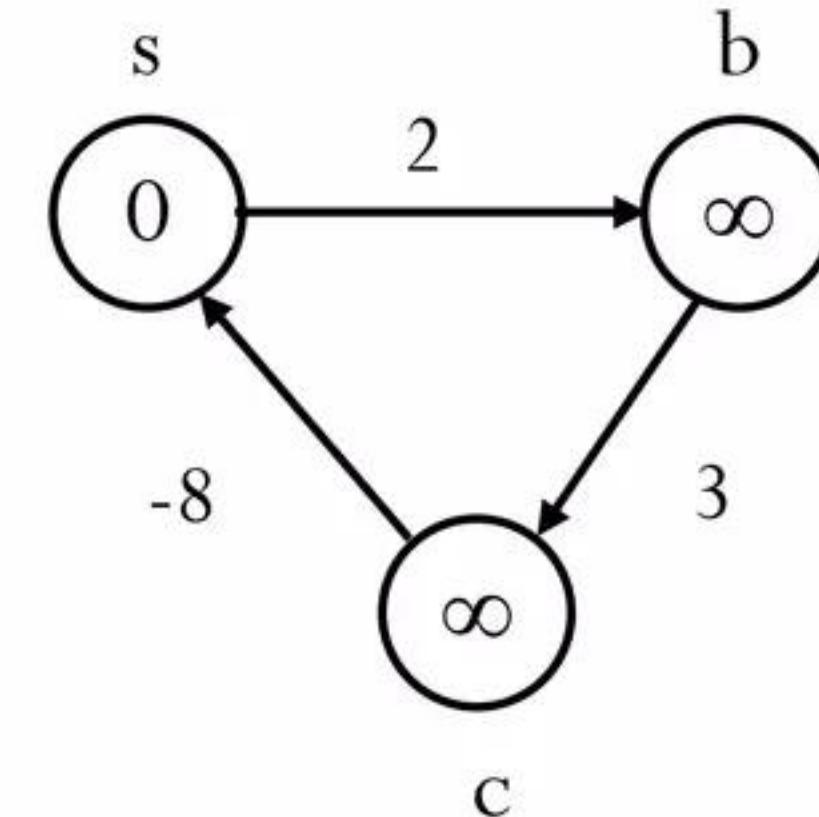
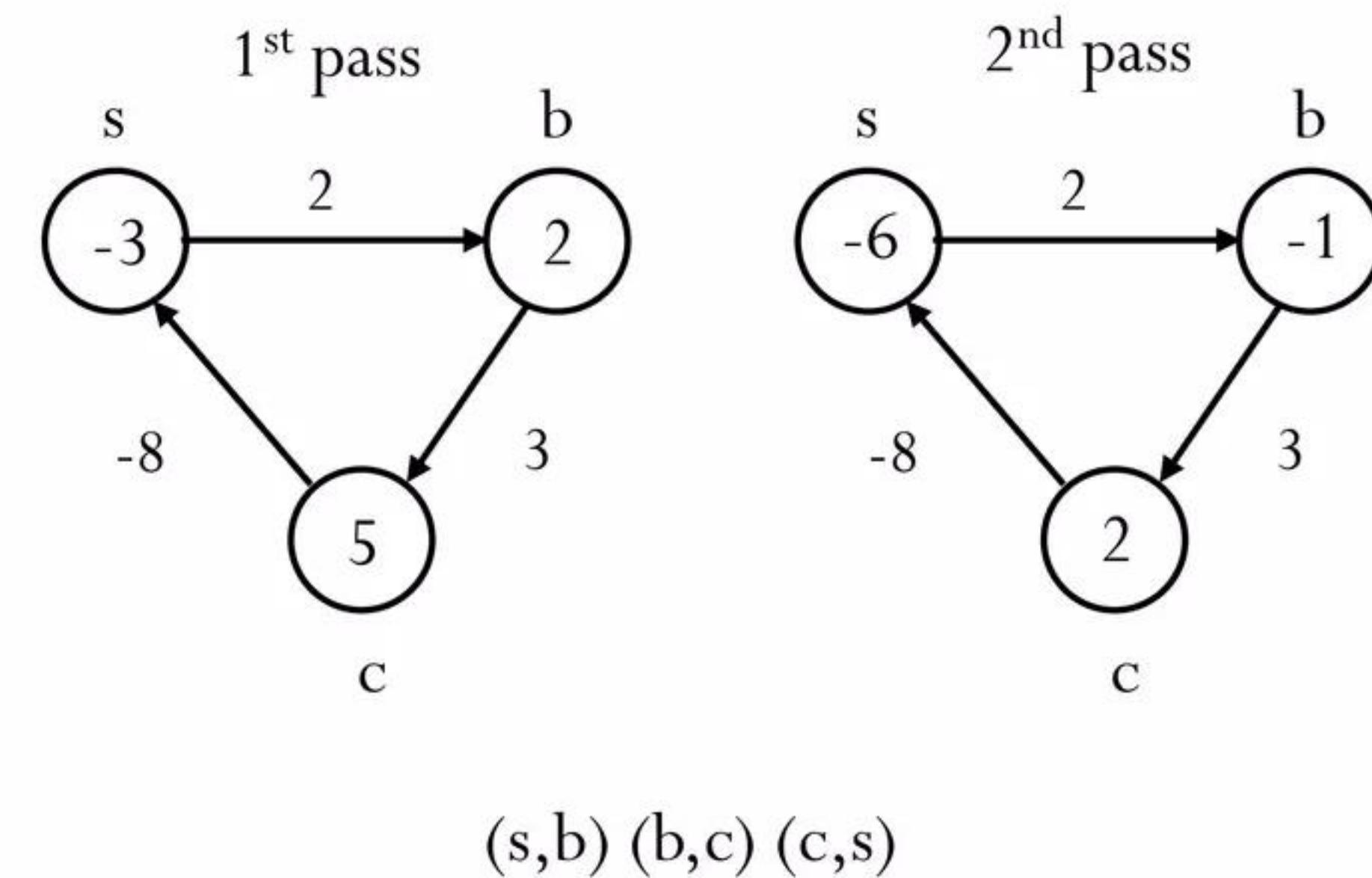


Pass 4



# Detecting Negative Cycles

- (perform extra test after  $V-1$  iterations)  
**for** each edge  $(u, v) \in E$   
    **do if**  $d[v] > d[u] + w(u, v)$   
    **then return** FALSE  
**return** TRUE



Look at edge  $(s, b)$ :

$$d[b] = -1$$

$$d[s] + w(s, b) = -4$$

$$\Rightarrow d[b] > d[s] + w(s, b)$$

# Cycles

**Can shortest paths contain cycles?**

- Negative-weight cycles: NO
  - Shortest path is not well defined, because each iteration result in reduced shortest path
- Positive-weight cycles: NO
  - Path is a tree, property of tree that tree does not have cycle
  - By removing the cycle, we can get a shorter path, like we did in minimum spanning tree
- Zero-weight cycles
  - No reason to use them
  - Can remove them to obtain a path with same weight

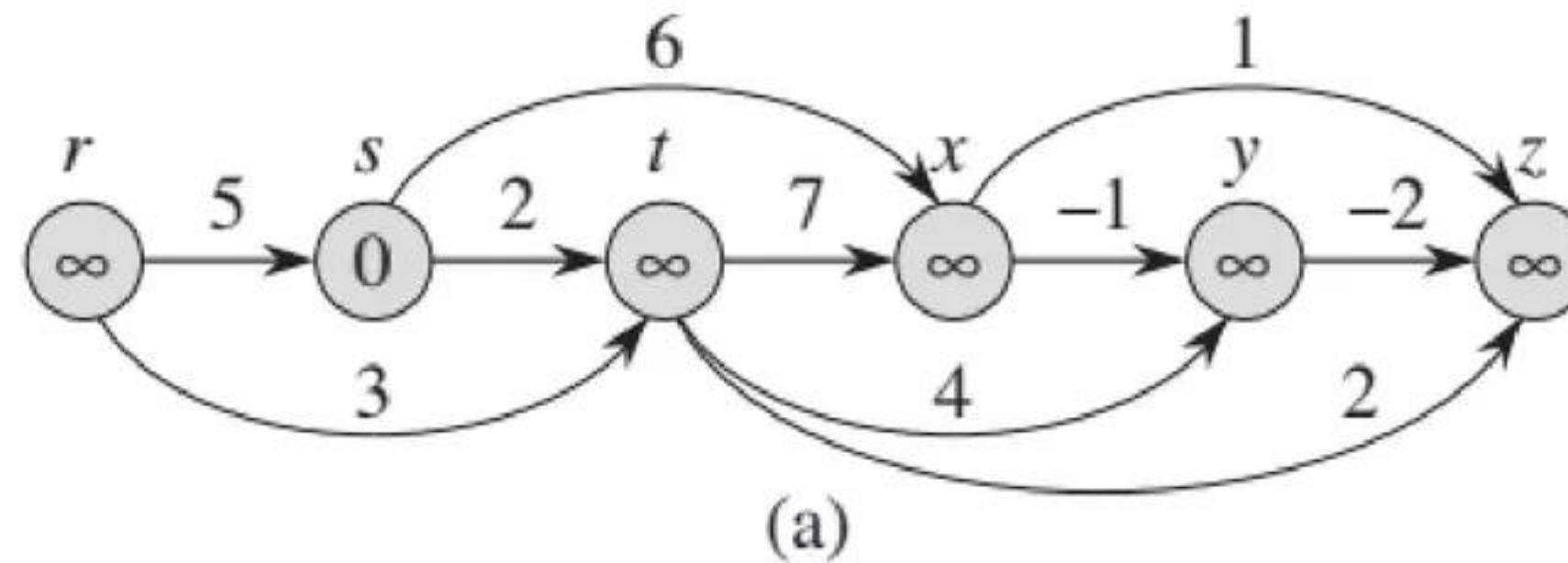
# Shortest paths in Directed Acyclic Graphs (DAG)

- Single-source Shortest paths in DAG
- Topological sort of line 1 takes  $O(V + E)$  time
- INITIALIZE line 2 takes  $O(V)$  time
- **for** loop of lines 3–5 makes one iteration per vertex
- **for** loop of lines 4–5 relaxes each edge exactly once.

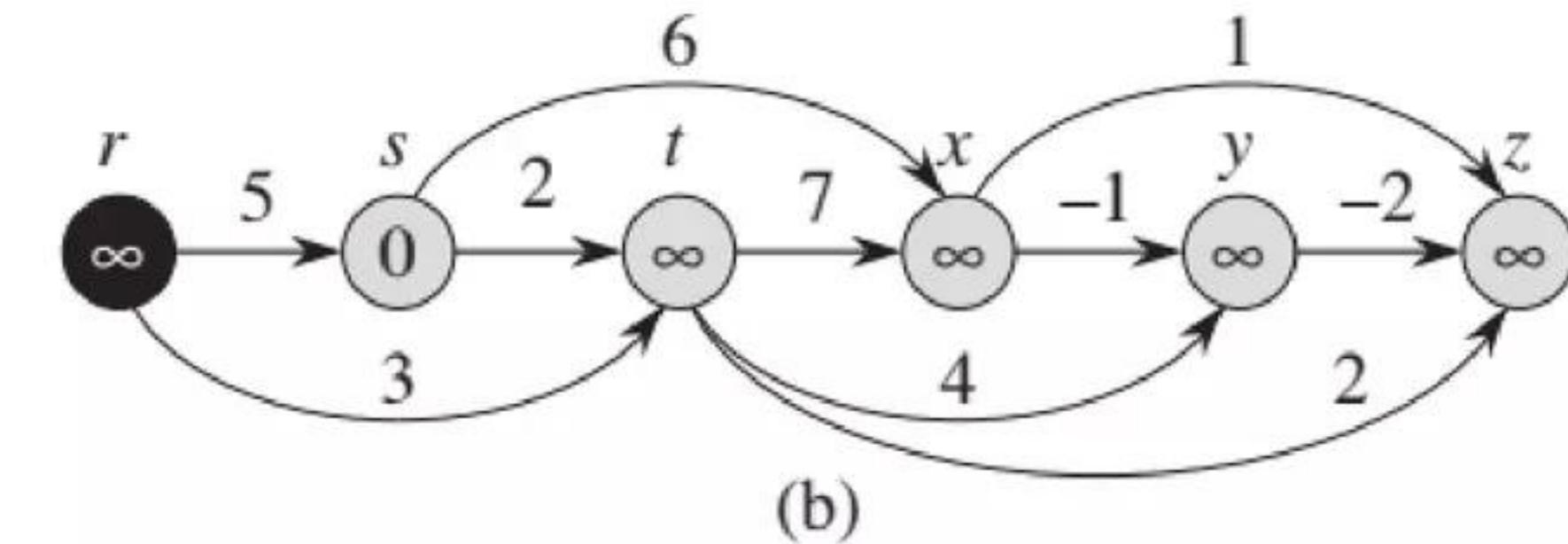
DAG-SHORTEST-PATHS( $G, w, s$ )

- 1 topologically sort the vertices of  $G$   $\leftarrow O(V+E)$
- 2 INITIALIZE-SINGLE-SOURCE( $G, s$ )  $\leftarrow O(V)$
- 3 **for** each vertex  $u$ , taken in topologically sorted order  $\leftarrow O(V)$
- 4     **for** each vertex  $v \in G.Adj[u]$   $\leftarrow O(E)$
- 5         RELAX( $u, v, w$ )  $\leftarrow O(E)$

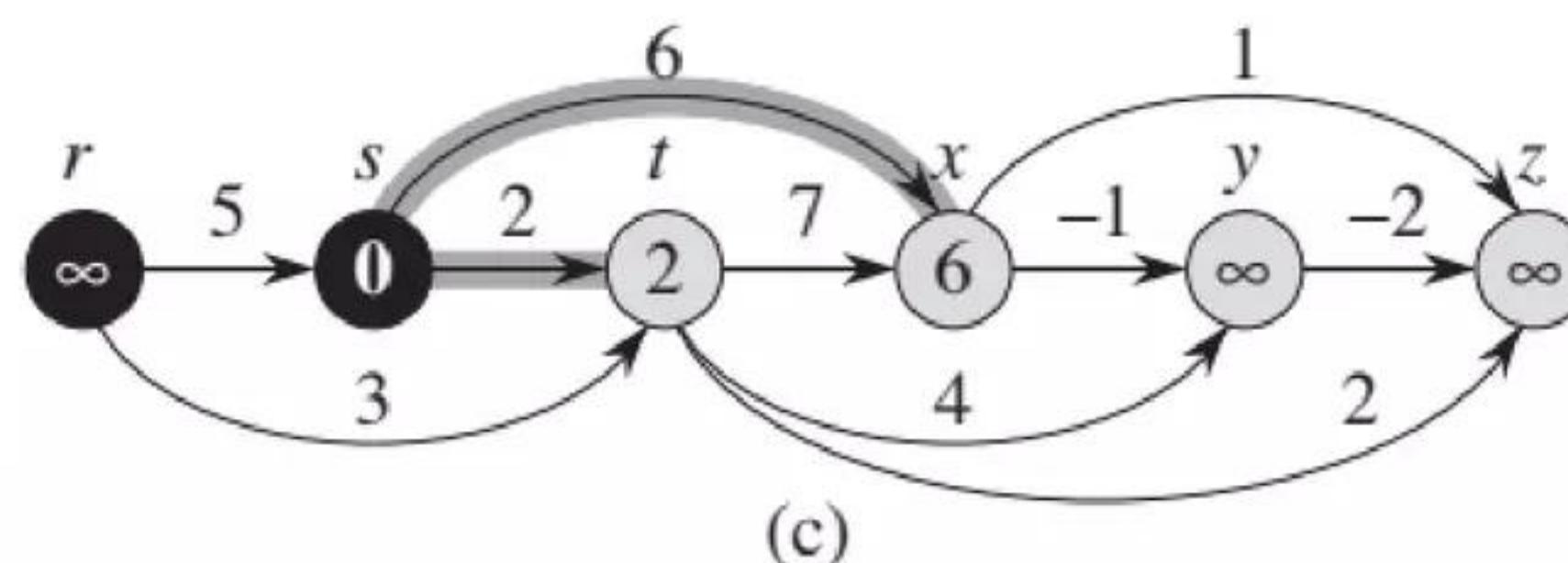
# Shortest paths in Directed Acyclic Graphs (DAG)



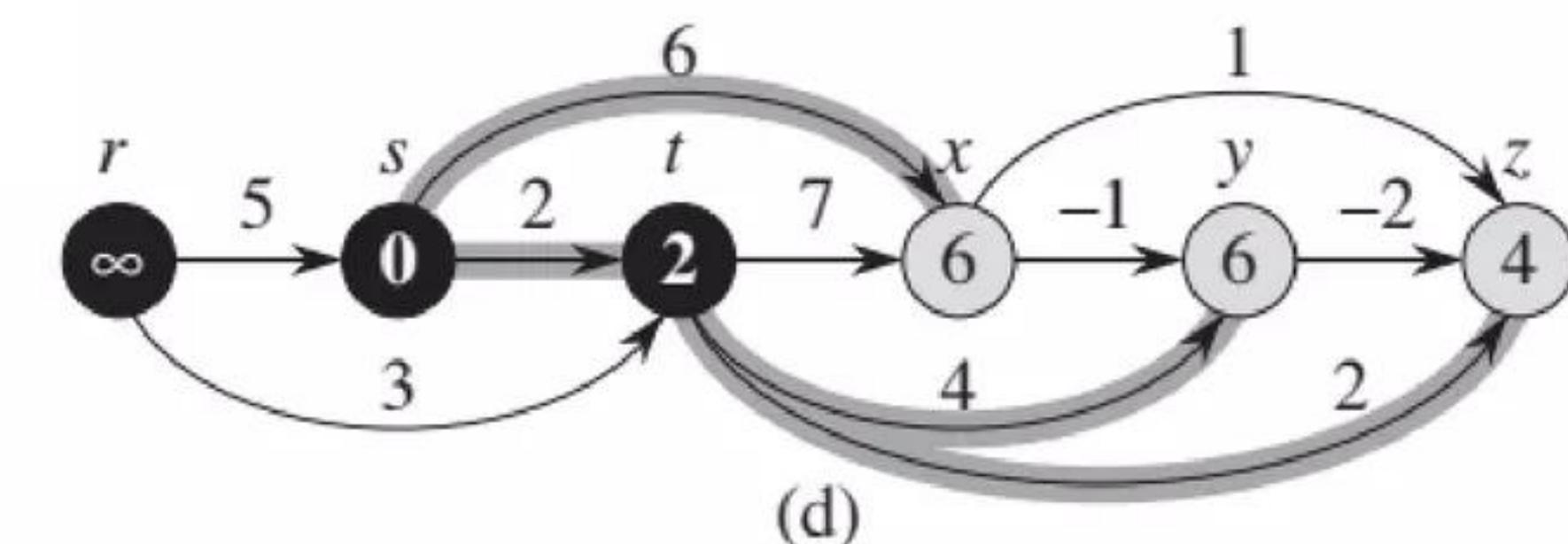
(a)



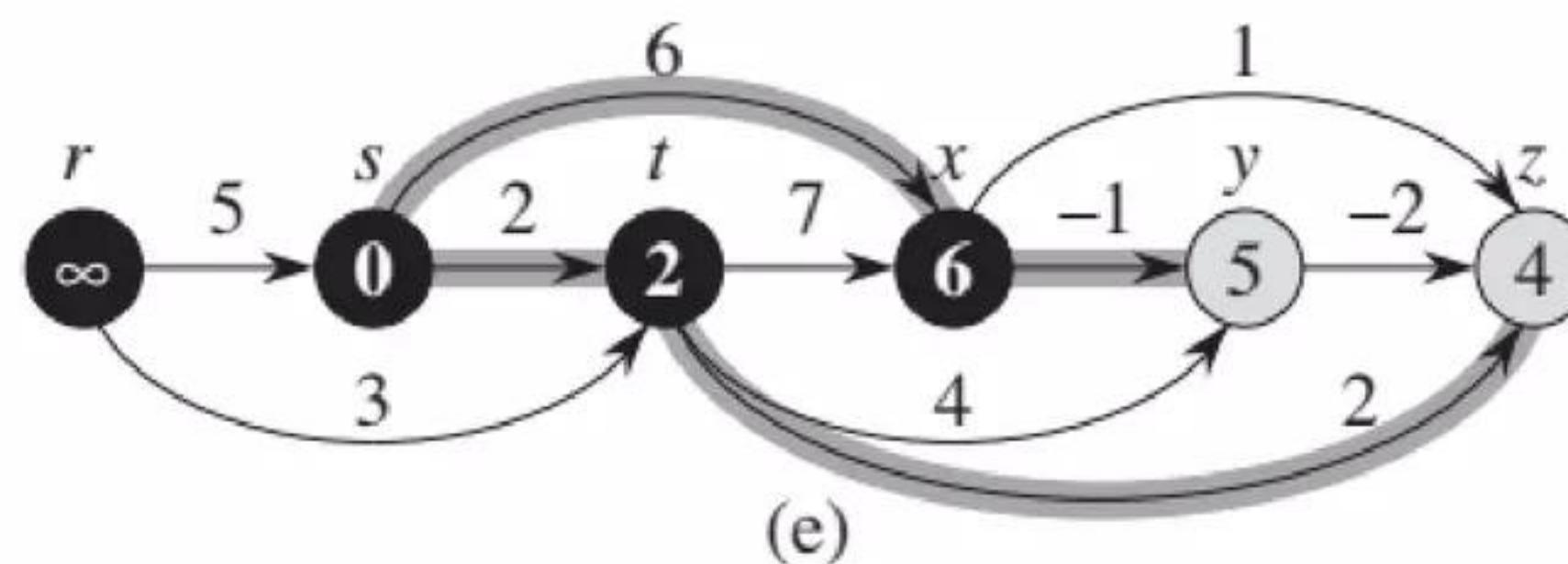
(b)



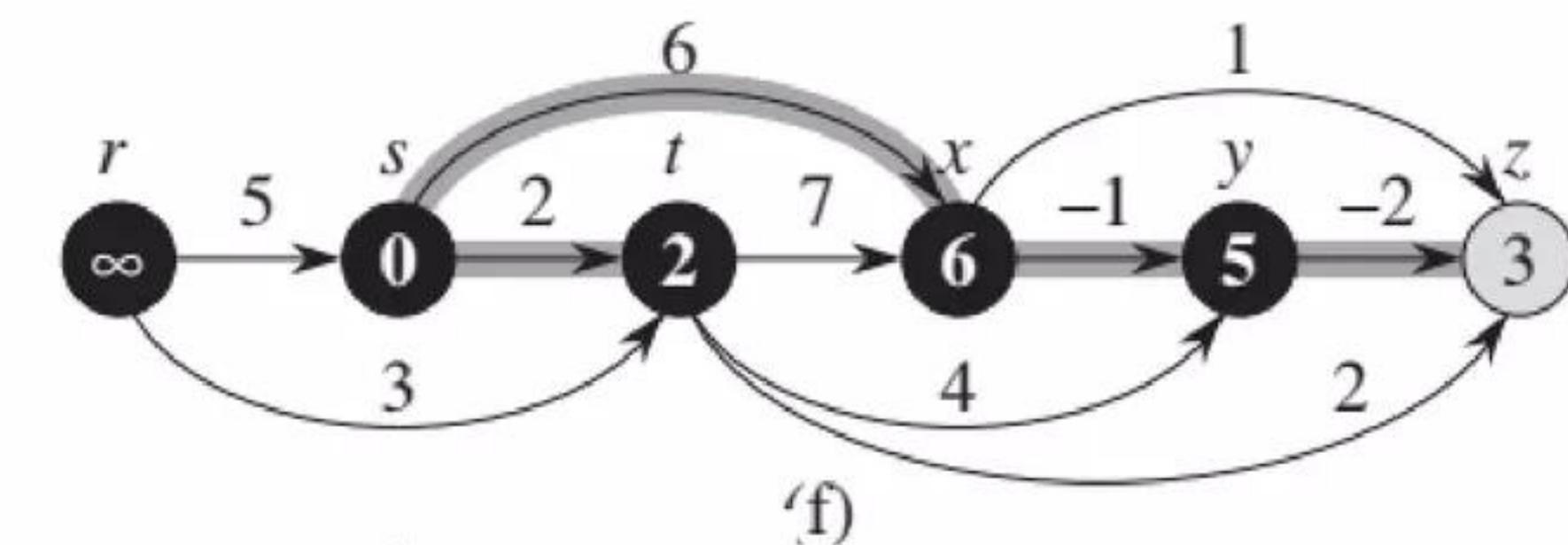
(c)



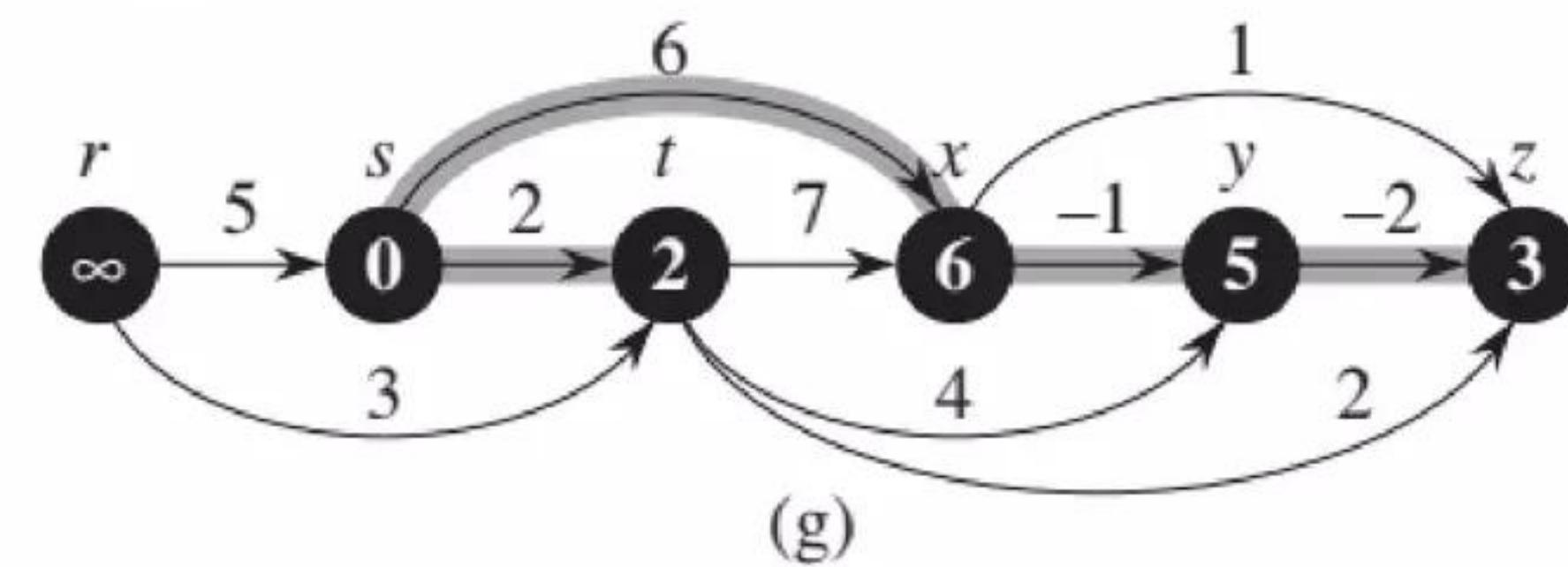
(d)



(e)



(f)

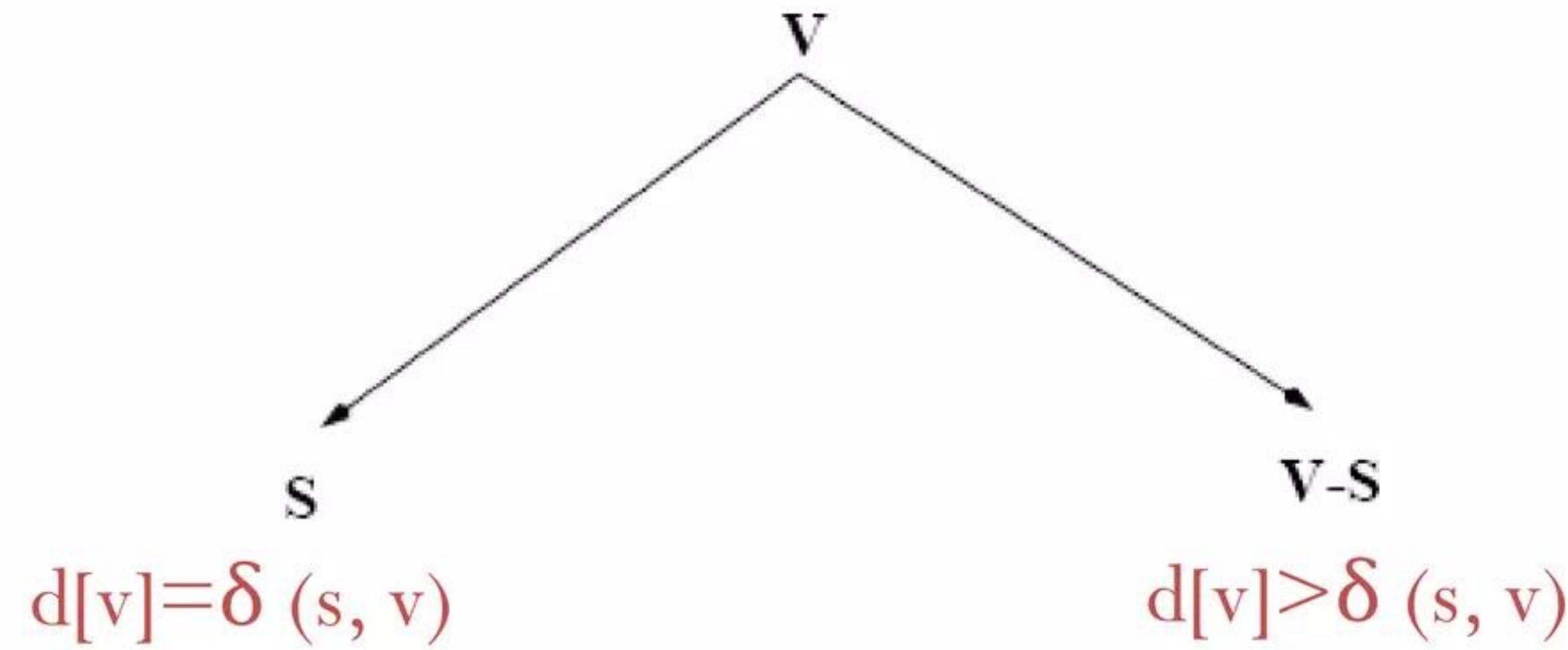


(g)

# Dijkstra's Algorithm

# Dijkstra's Algorithm

- Single-source shortest path problem:
  - No negative-weight edges:  $w(u, v) > 0, \forall (u, v) \in E$
- Each edge is relaxed **only once!**
- Maintains two sets of vertices:
- Similar to Prim's algorithm, which finds Minimum Spanning Tree (MST)



# Dijkstra's Algorithm

- Line 1 initializes the  $d$  and  $\pi$  values in the usual way,
- Line 2 initializes the set  $S$  to the empty set.
- Line 3 initializes the min-priority queue  $Q$  to contain all the vertices in  $V$ ;

DIJKSTRA( $G, w, s$ )

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
2   $S = \emptyset$ 
3   $Q = G.V$ 
4  while  $Q \neq \emptyset$ 
5       $u = \text{EXTRACT-MIN}(Q)$ 
6       $S = S \cup \{u\}$ 
7      for each vertex  $v \in G.\text{Adj}[u]$ 
8          RELAX( $u, v, w$ )
```

# Dijkstra's Algorithm

- While loop of lines 4–8, line 5 extracts a vertex  $u$  from  $Q$  and
- Line 6 adds it to set  $S$ , thereby maintaining the invariant. Vertex  $u$ , therefore, has the smallest shortest-path estimate of any vertex in  $V - S$ .
- Then, lines 7–8 relax each edge, thus updating the estimate  $v.d$  and the predecessor  $v.\pi$

DIJKSTRA( $G, w, s$ )

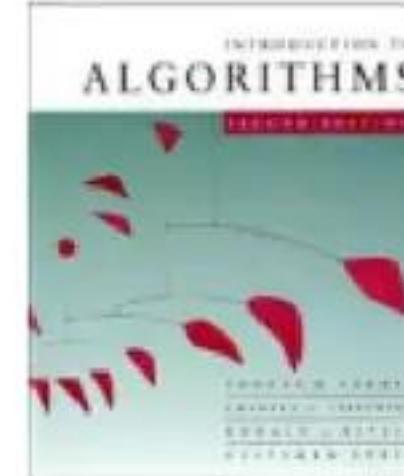
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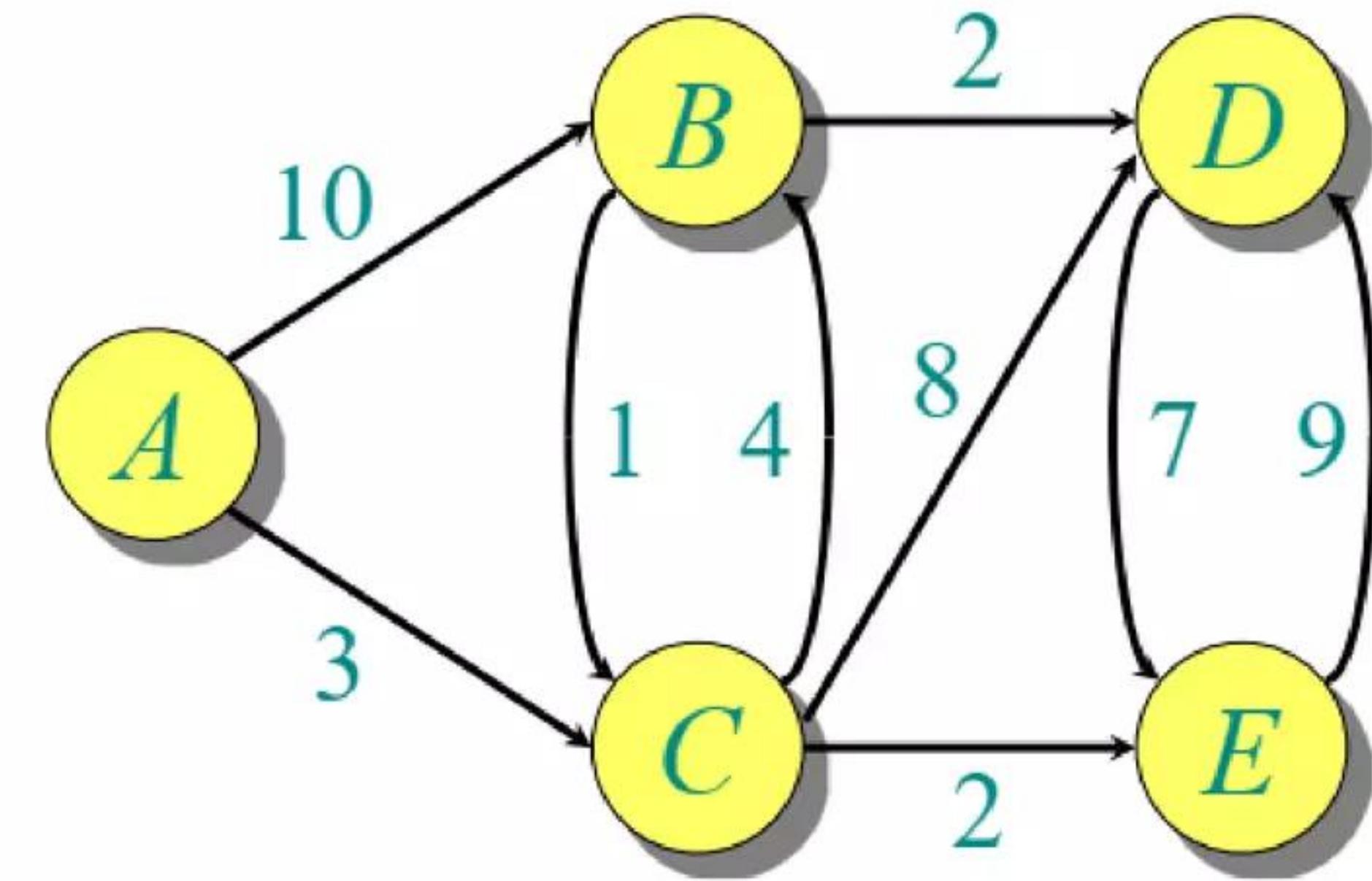
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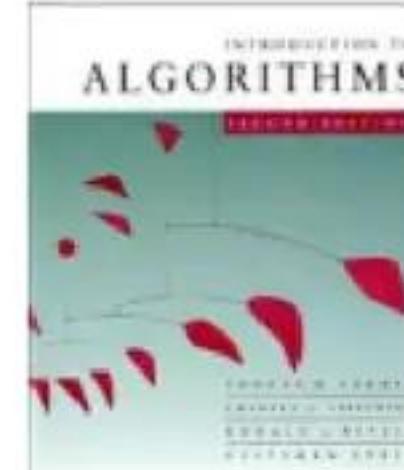
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```



# Example of Dijkstra's algorithm

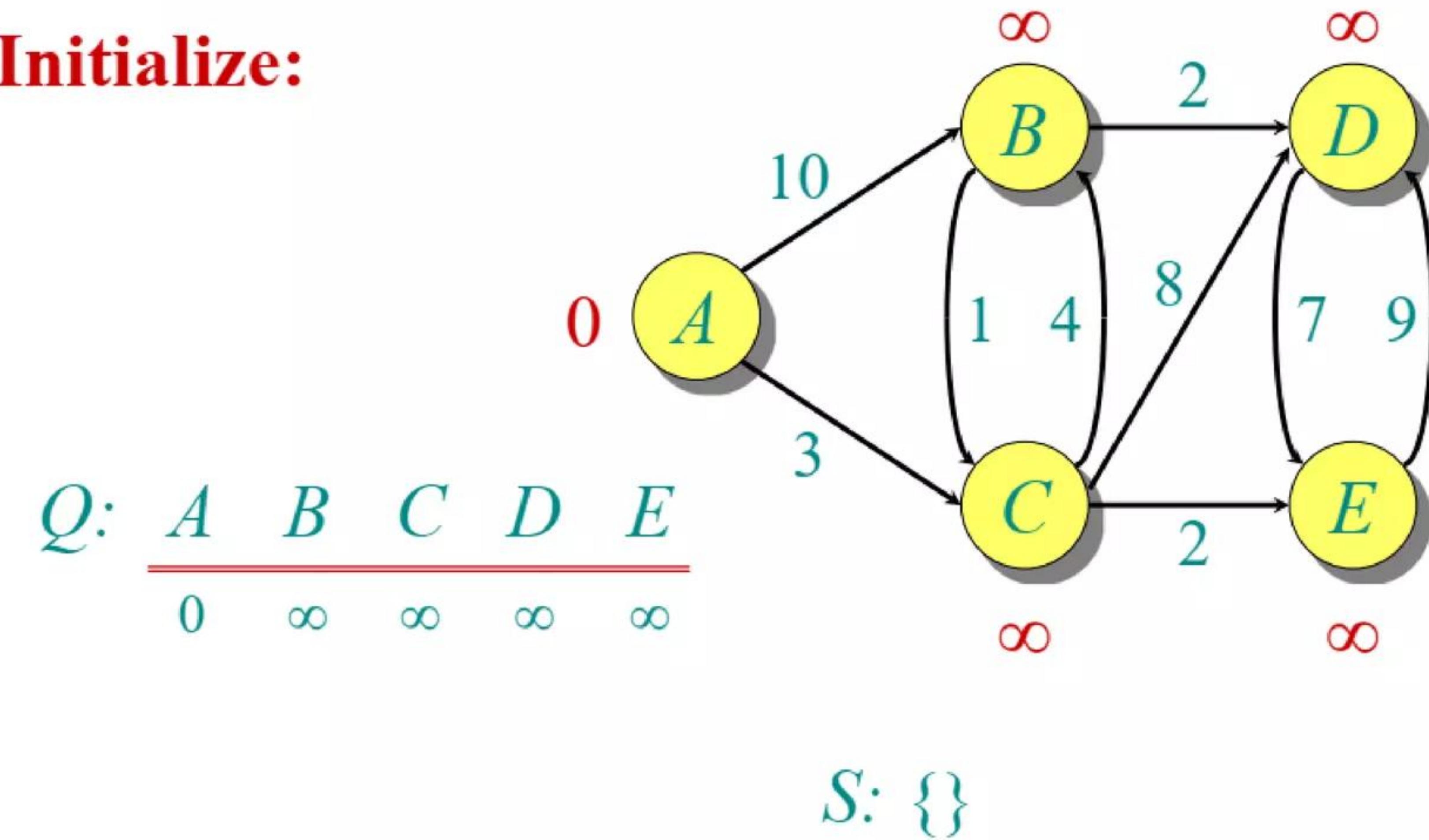
**Graph with  
nonnegative  
edge weights:**

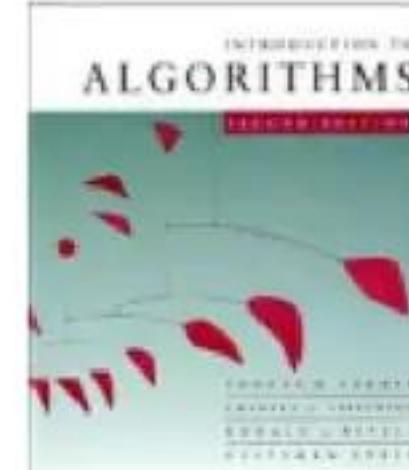




# Example of Dijkstra's algorithm

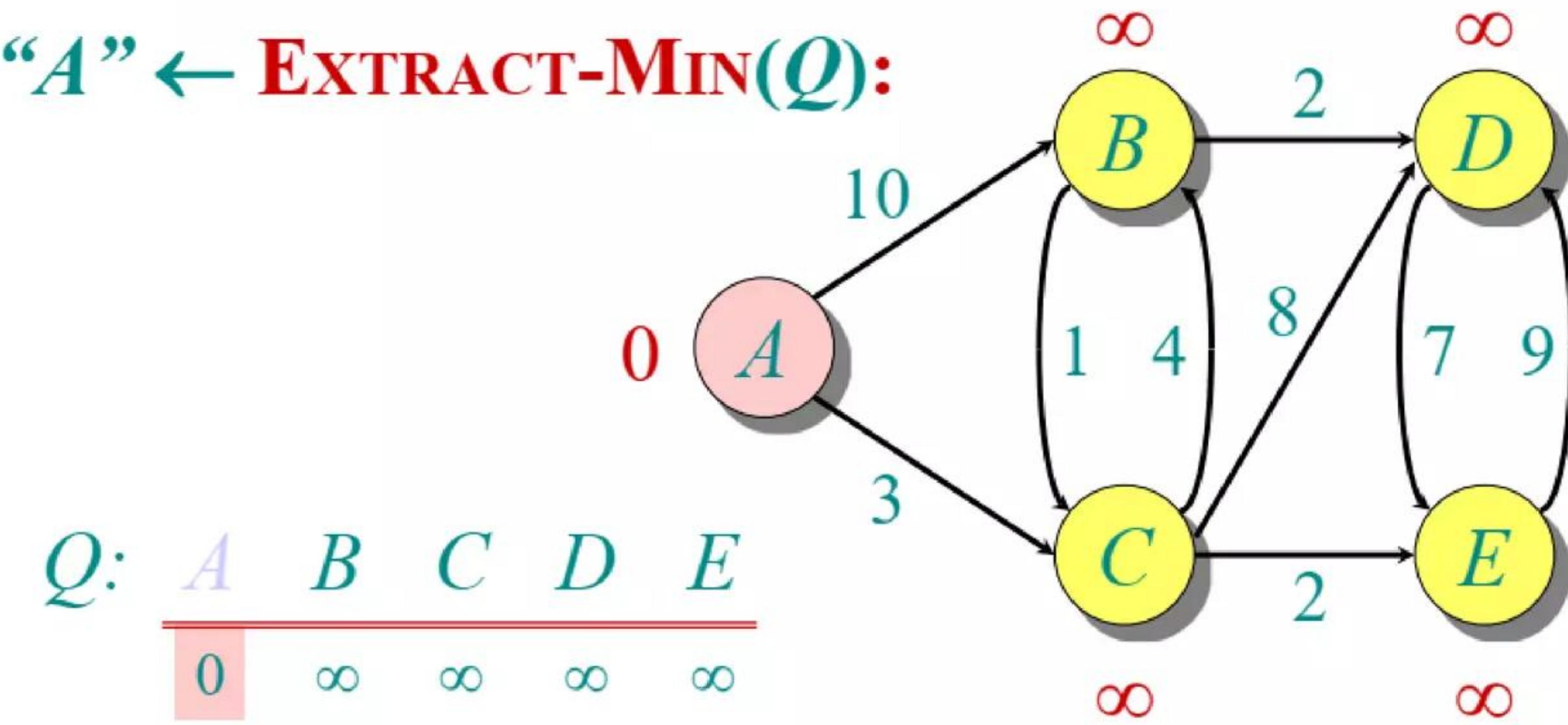
Initialize:



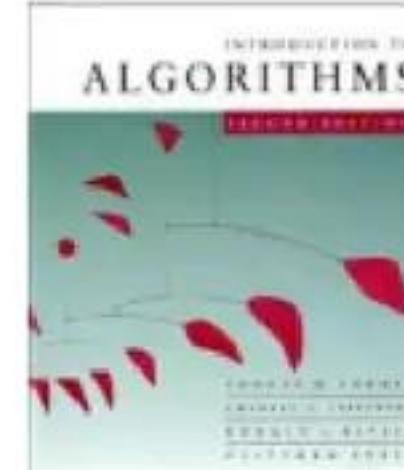


# Example of Dijkstra's algorithm

“ $A$ ”  $\leftarrow \text{EXTRACT-MIN}(Q)$ :

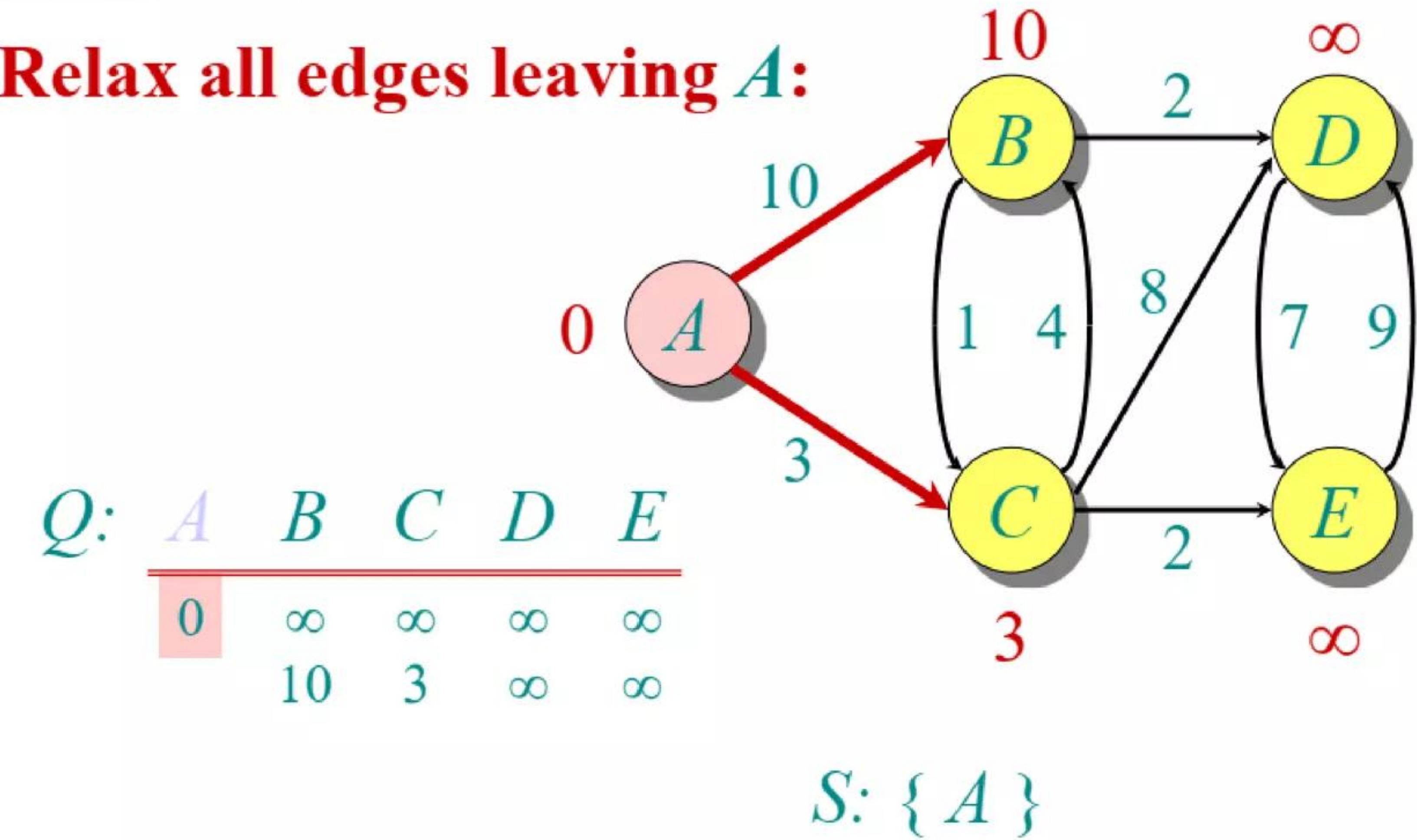


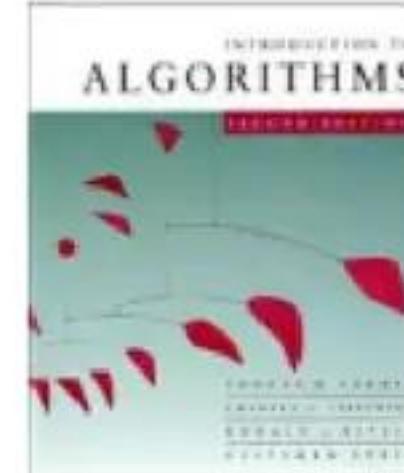
$S: \{ A \}$



# Example of Dijkstra's algorithm

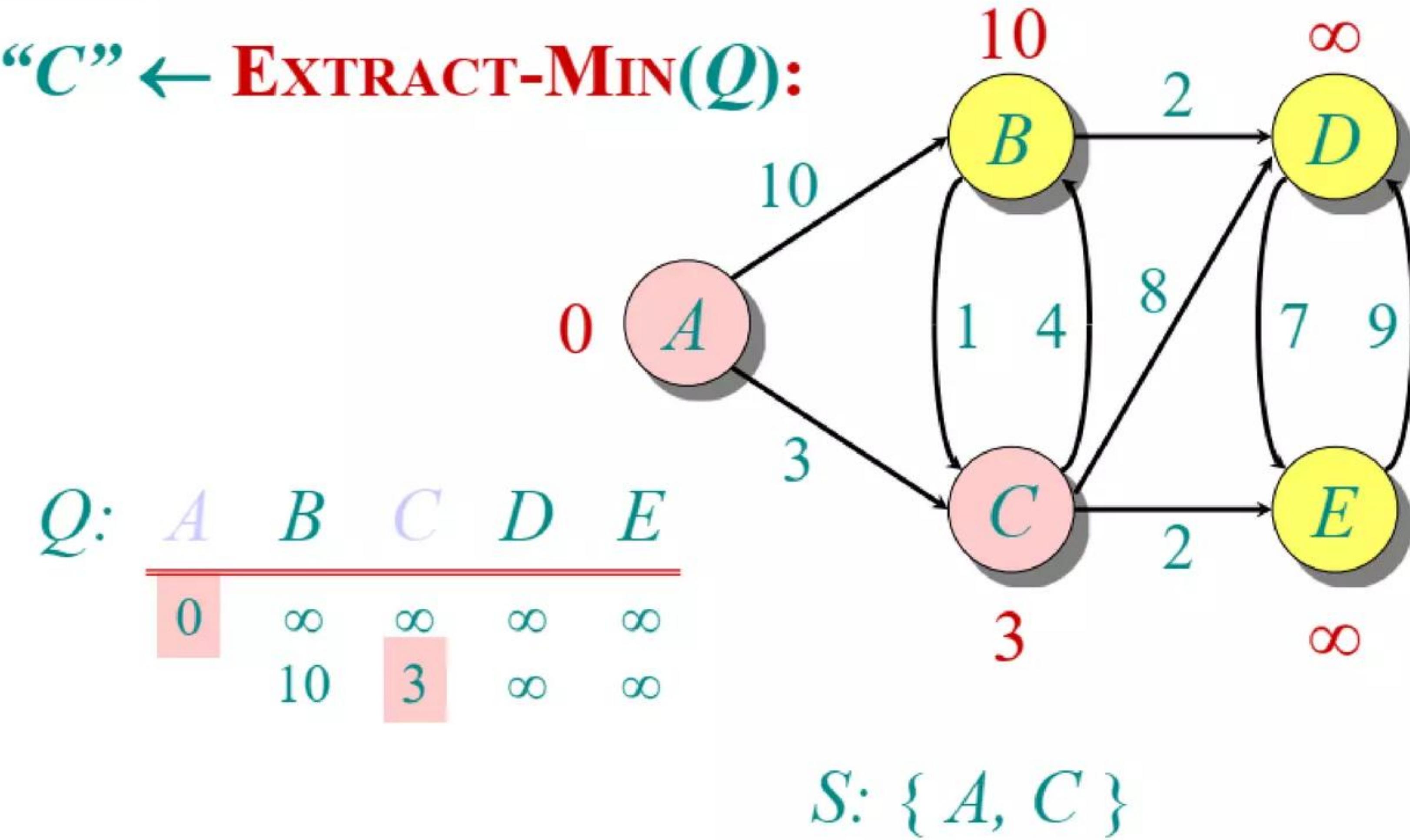
Relax all edges leaving  $A$ :

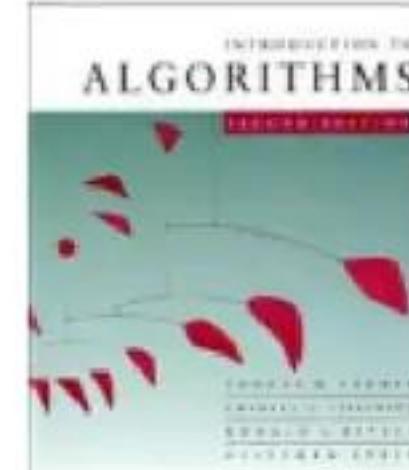




# Example of Dijkstra's algorithm

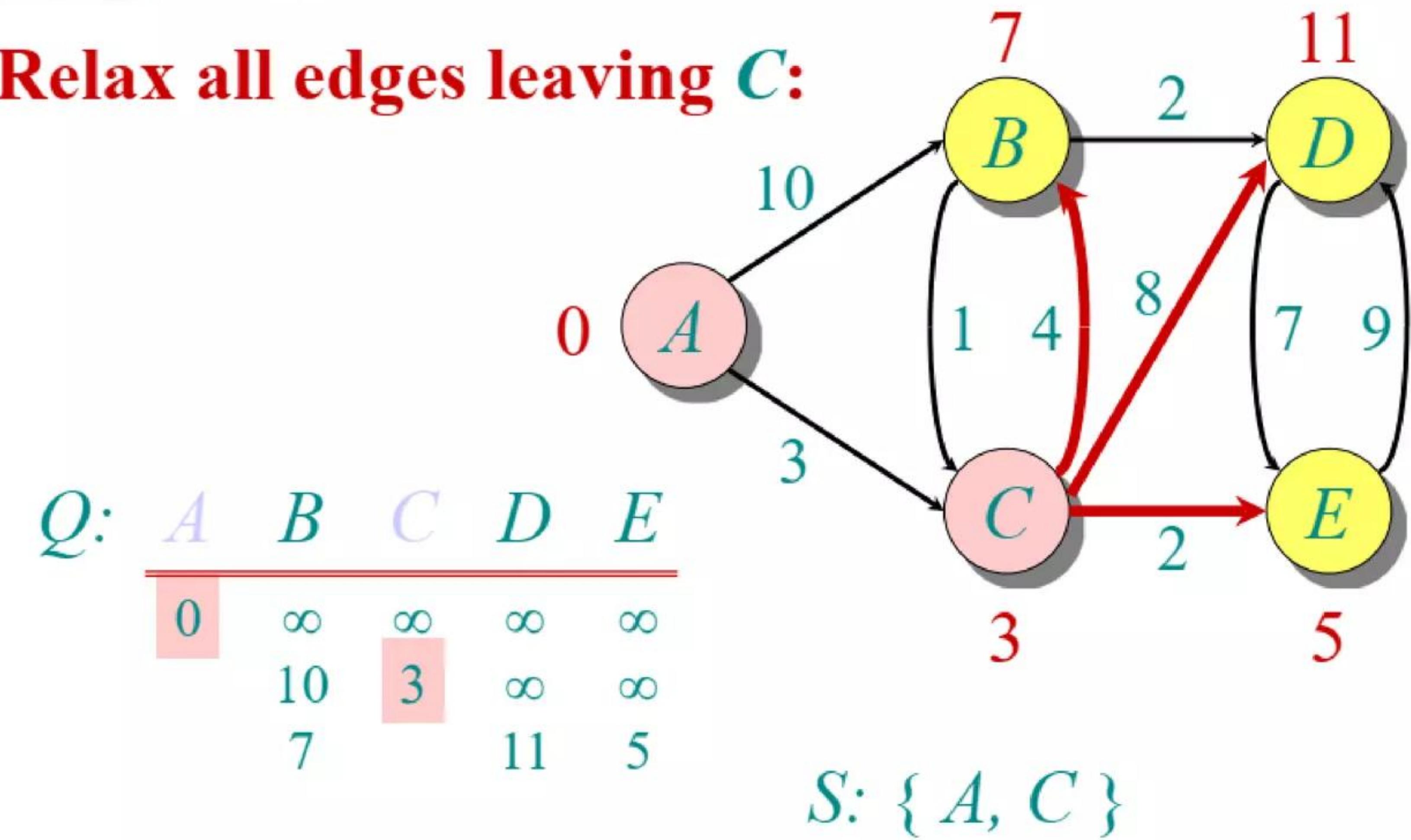
“C”  $\leftarrow \text{EXTRACT-MIN}(Q)$ :

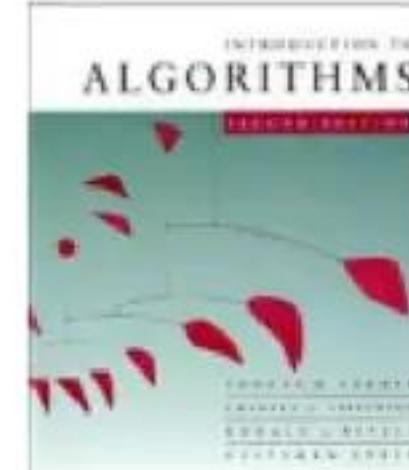




# Example of Dijkstra's algorithm

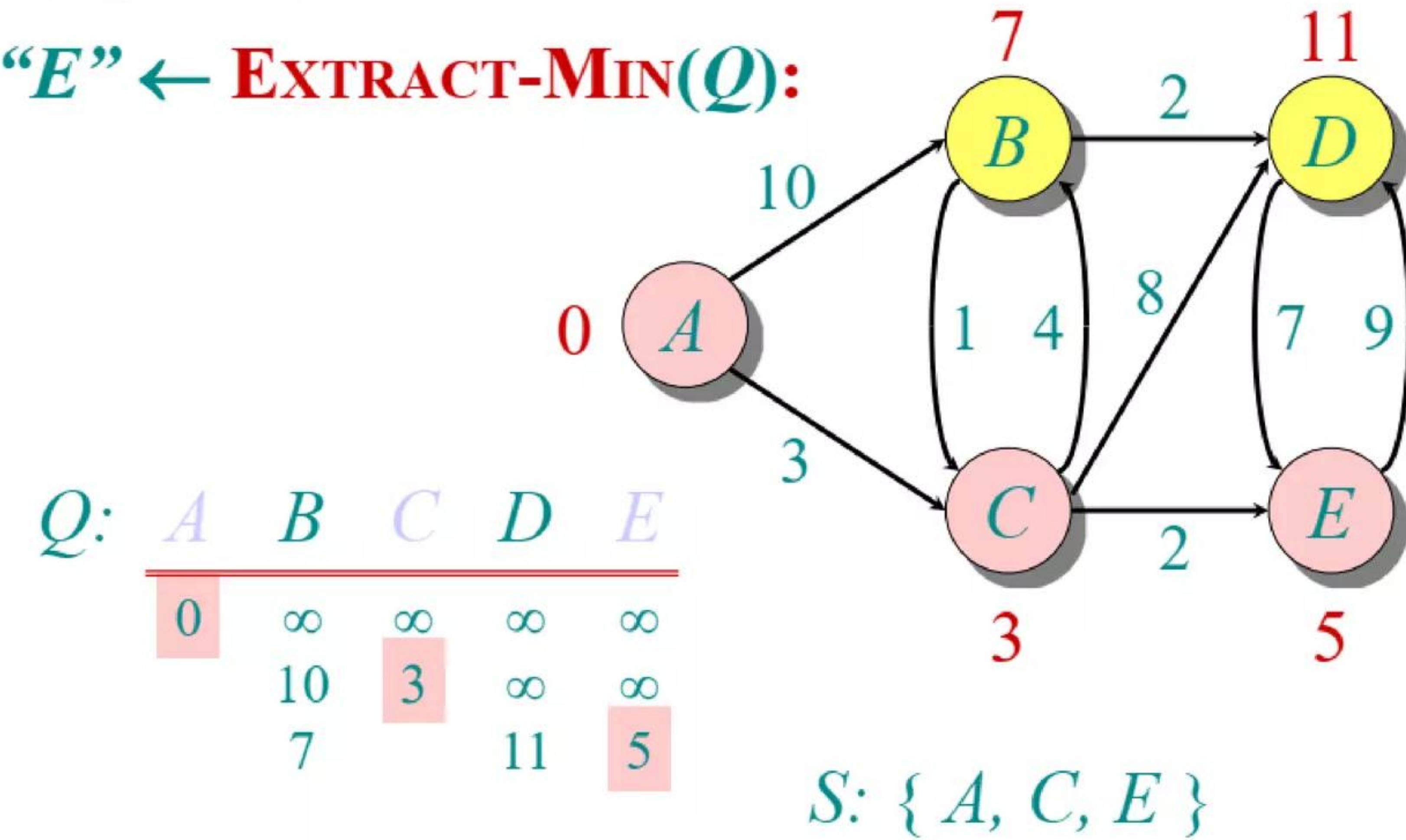
Relax all edges leaving  $C$ :

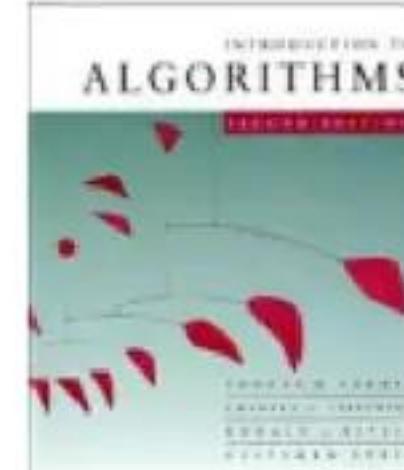




# Example of Dijkstra's algorithm

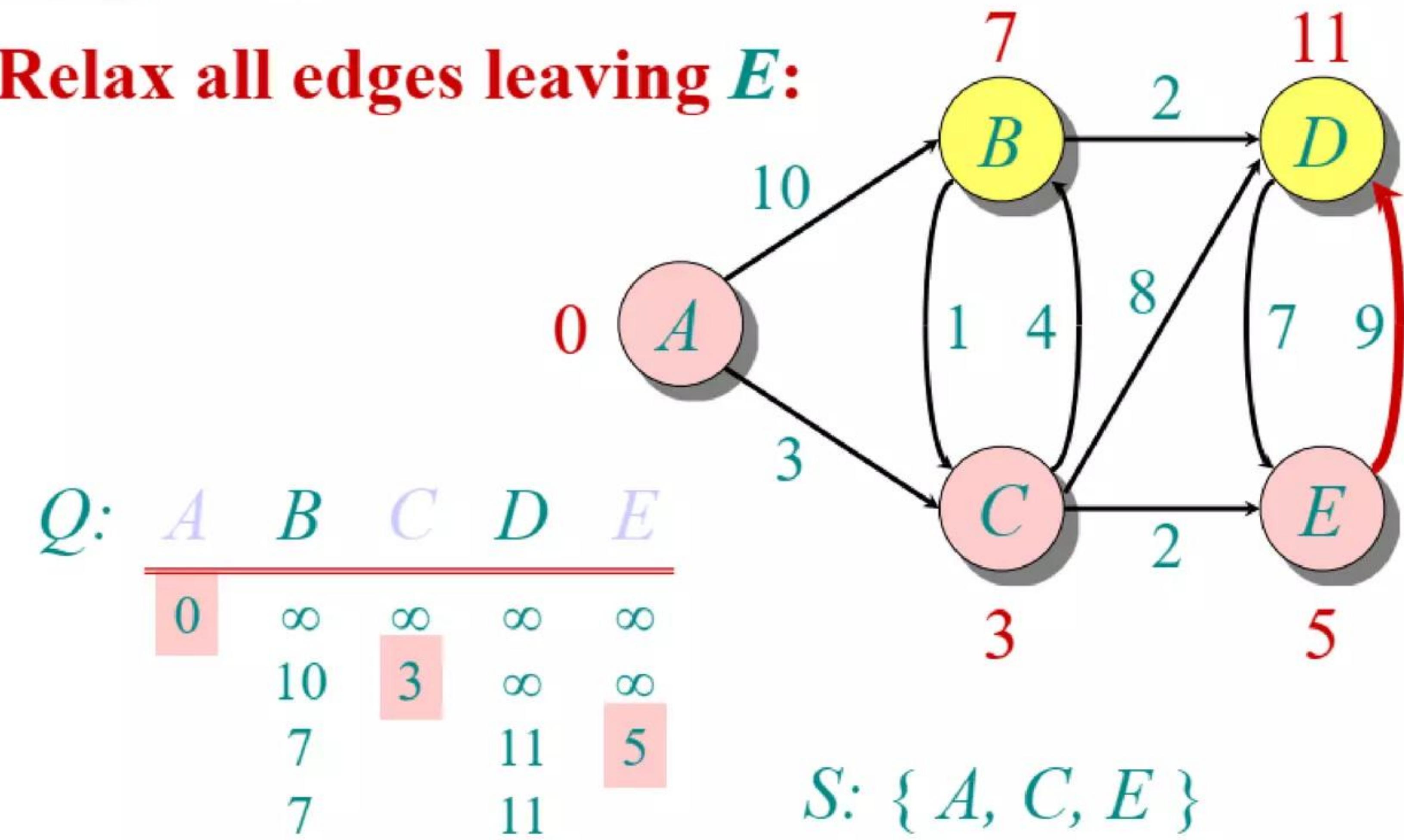
“ $E$ ”  $\leftarrow \text{EXTRACT-MIN}(Q)$ :

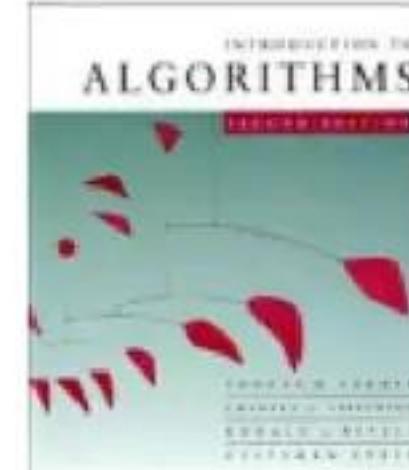




# Example of Dijkstra's algorithm

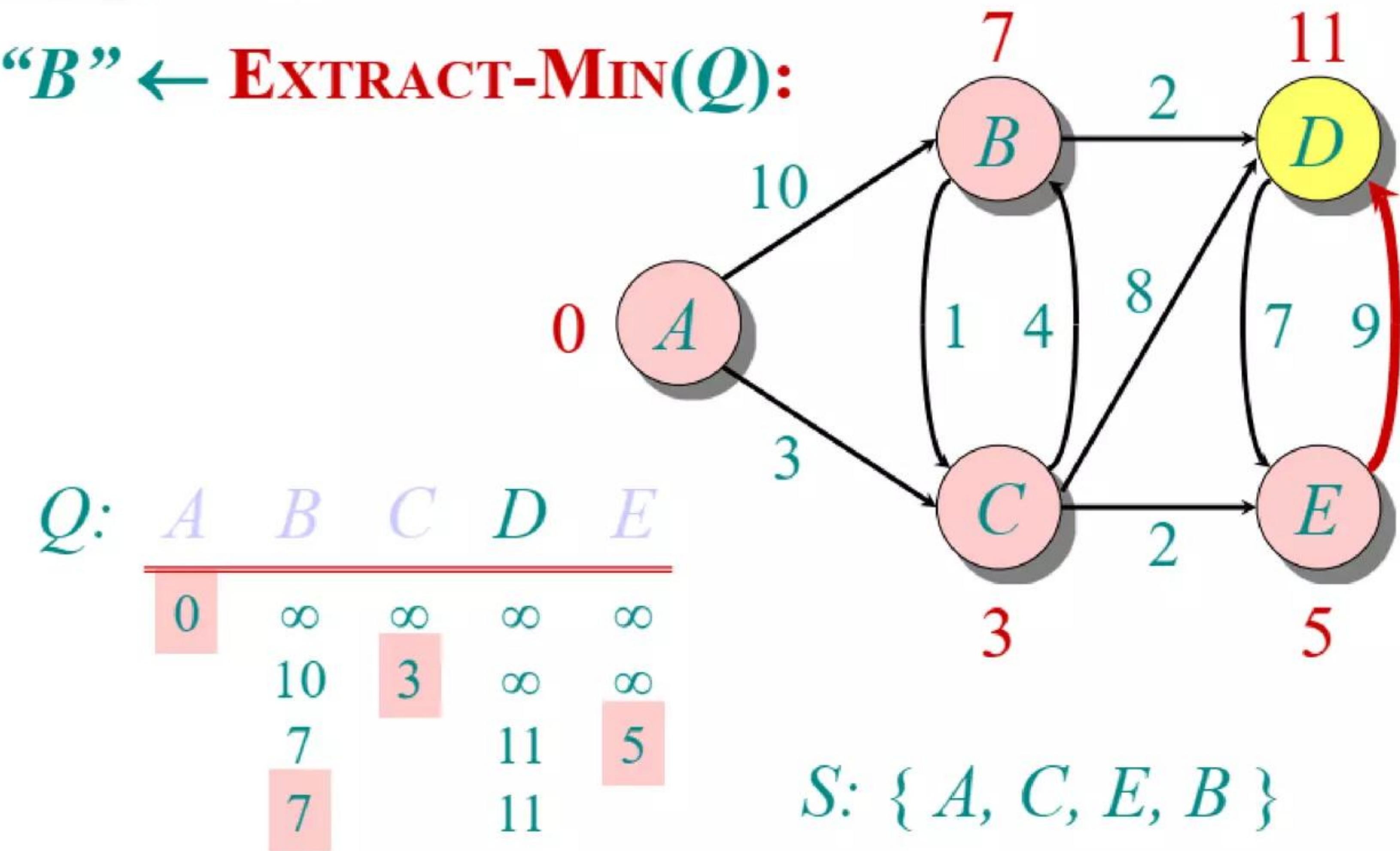
Relax all edges leaving  $E$ :

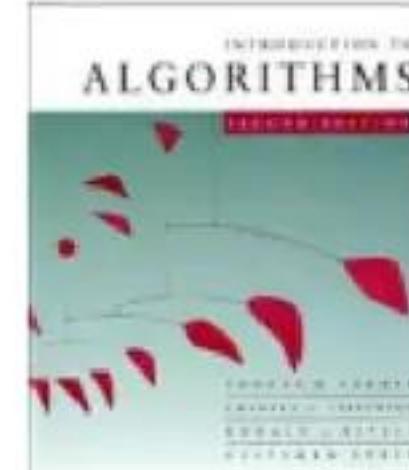




# Example of Dijkstra's algorithm

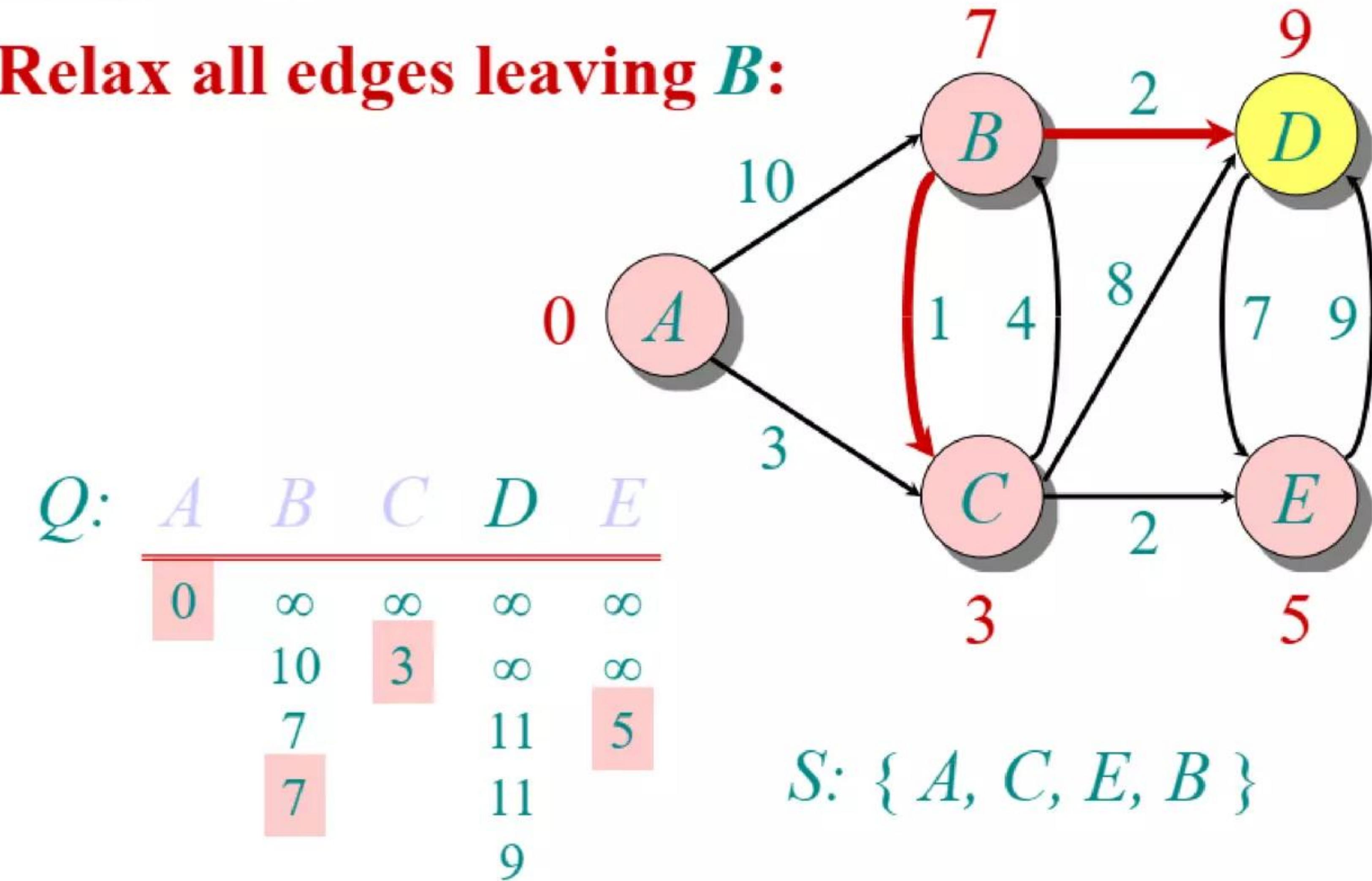
“B”  $\leftarrow \text{EXTRACT-MIN}(Q)$ :

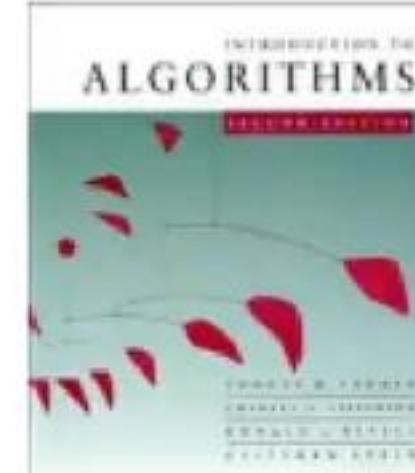




# Example of Dijkstra's algorithm

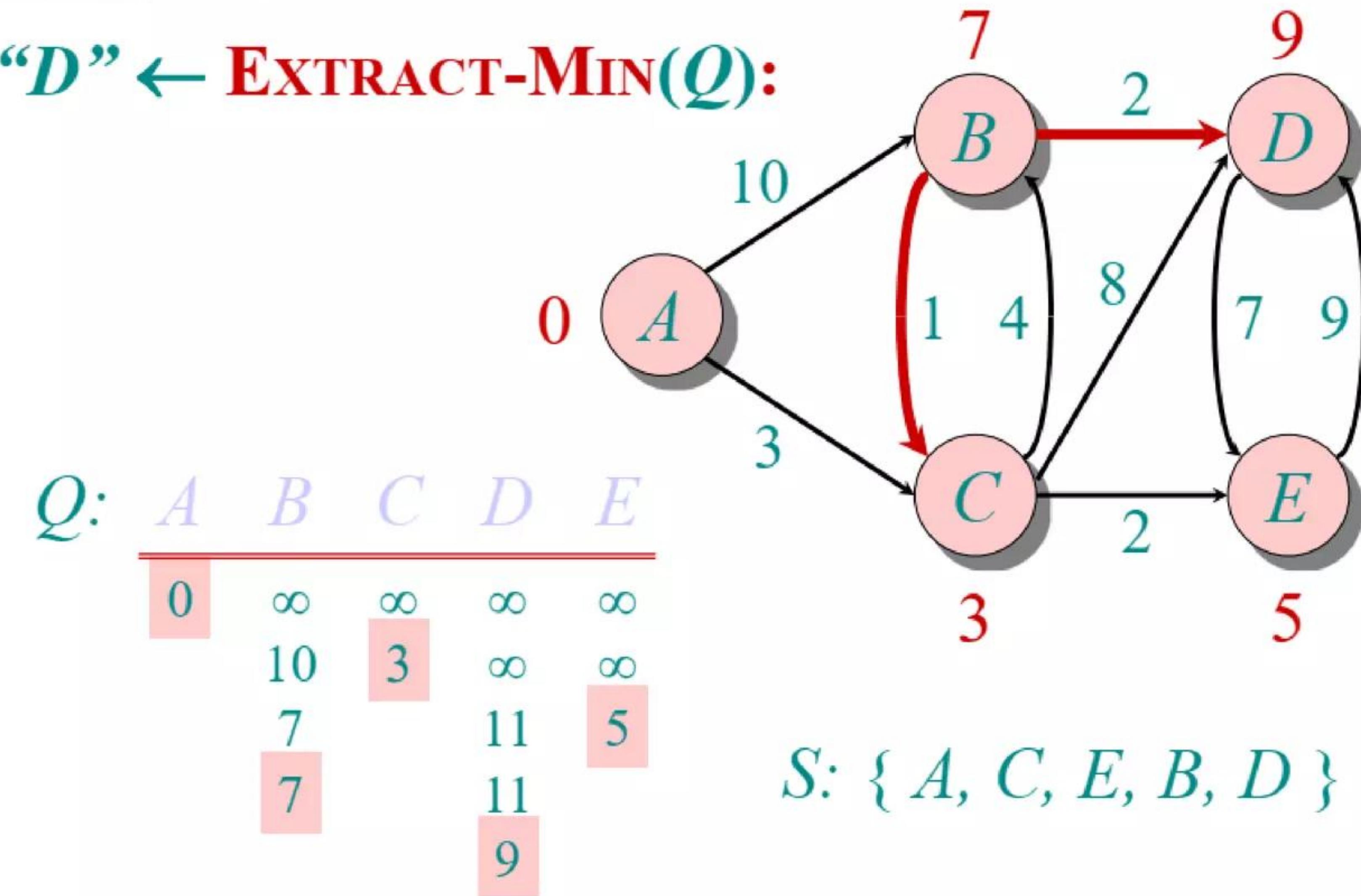
Relax all edges leaving  $B$ :





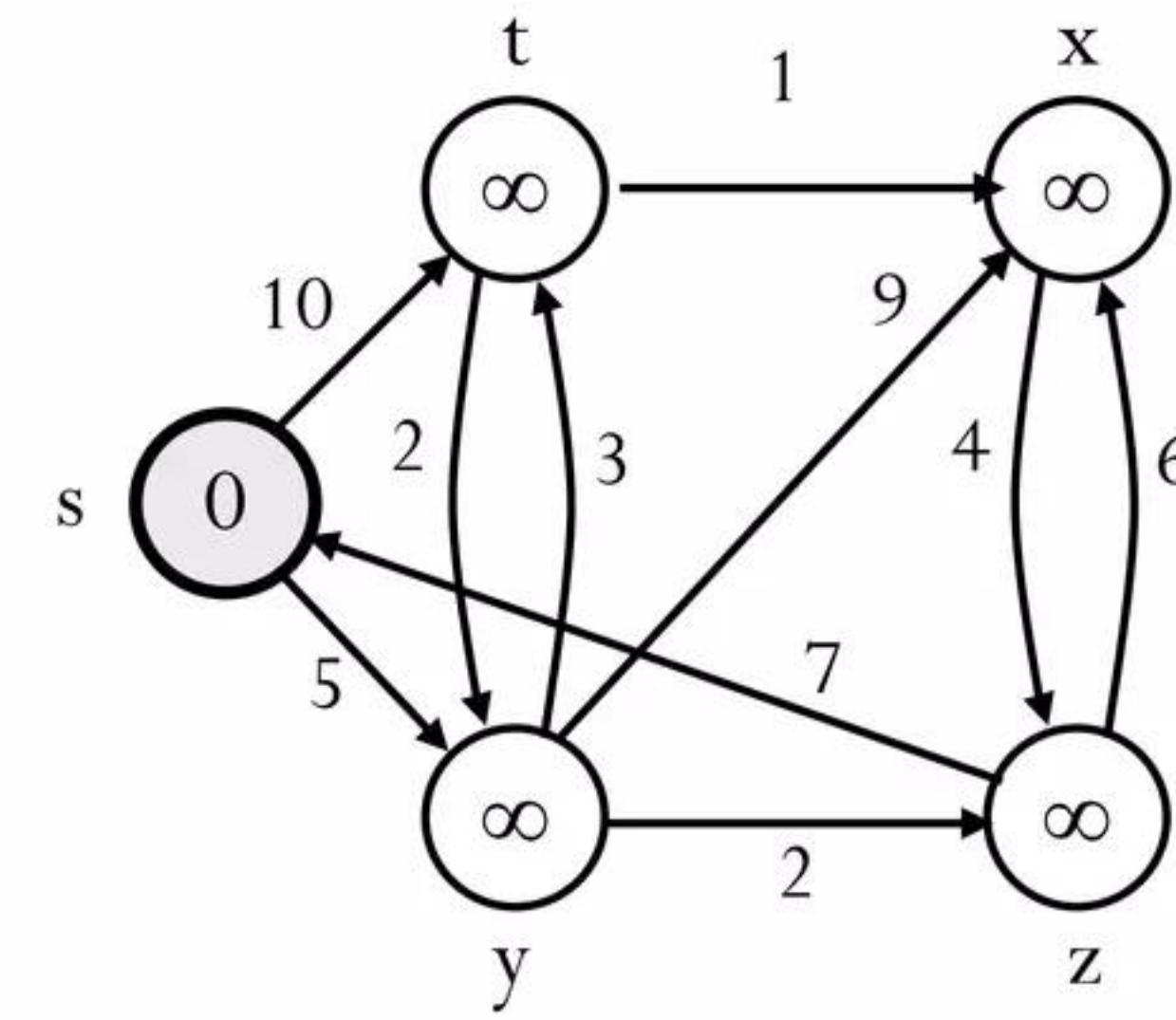
# Example of Dijkstra's algorithm

“ $D$ ”  $\leftarrow \text{EXTRACT-MIN}(Q)$ :

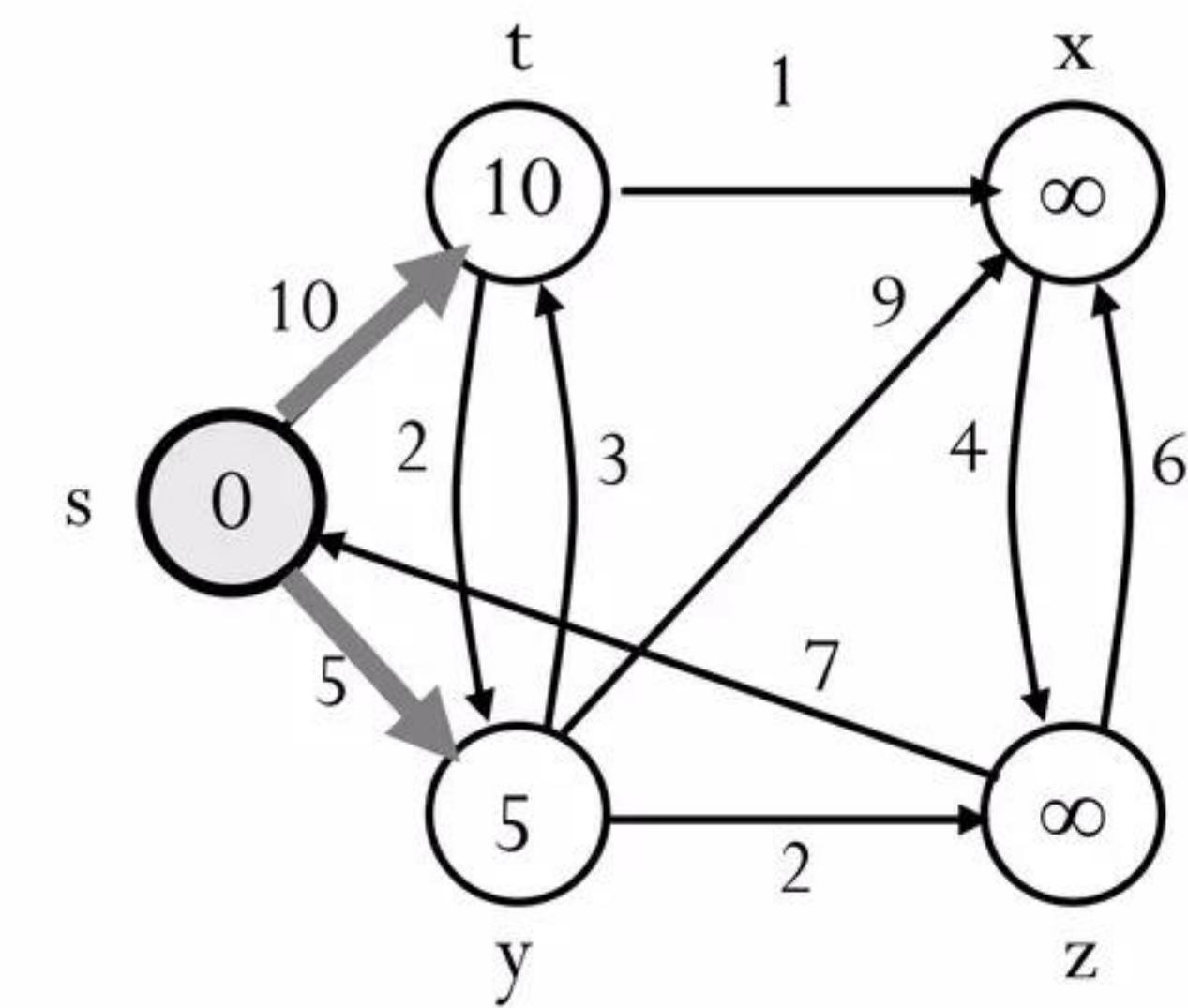


# Dijkstra ( $G, w, s$ )

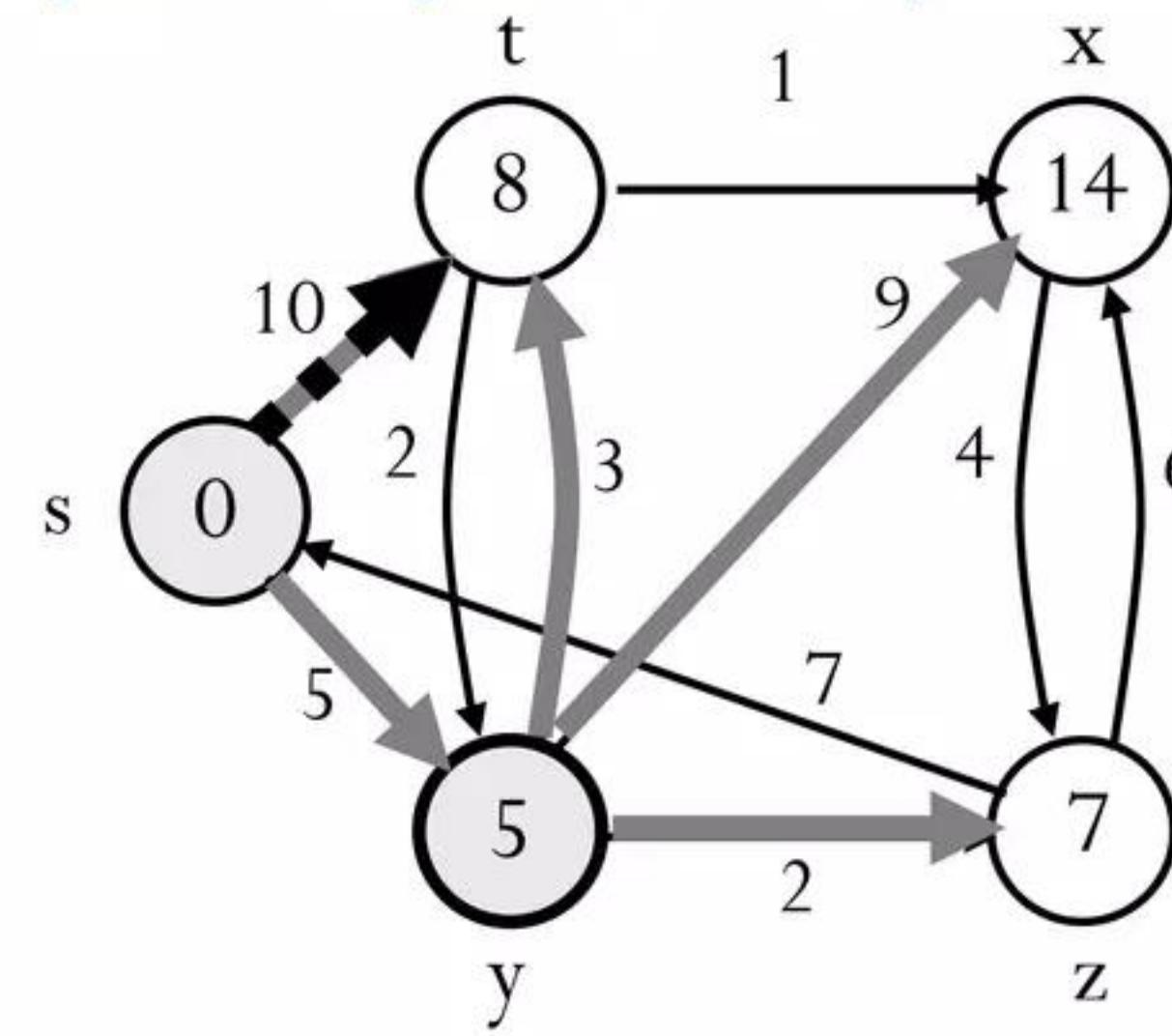
$S = \langle \rangle$   $Q = \langle s, t, x, z, y \rangle$



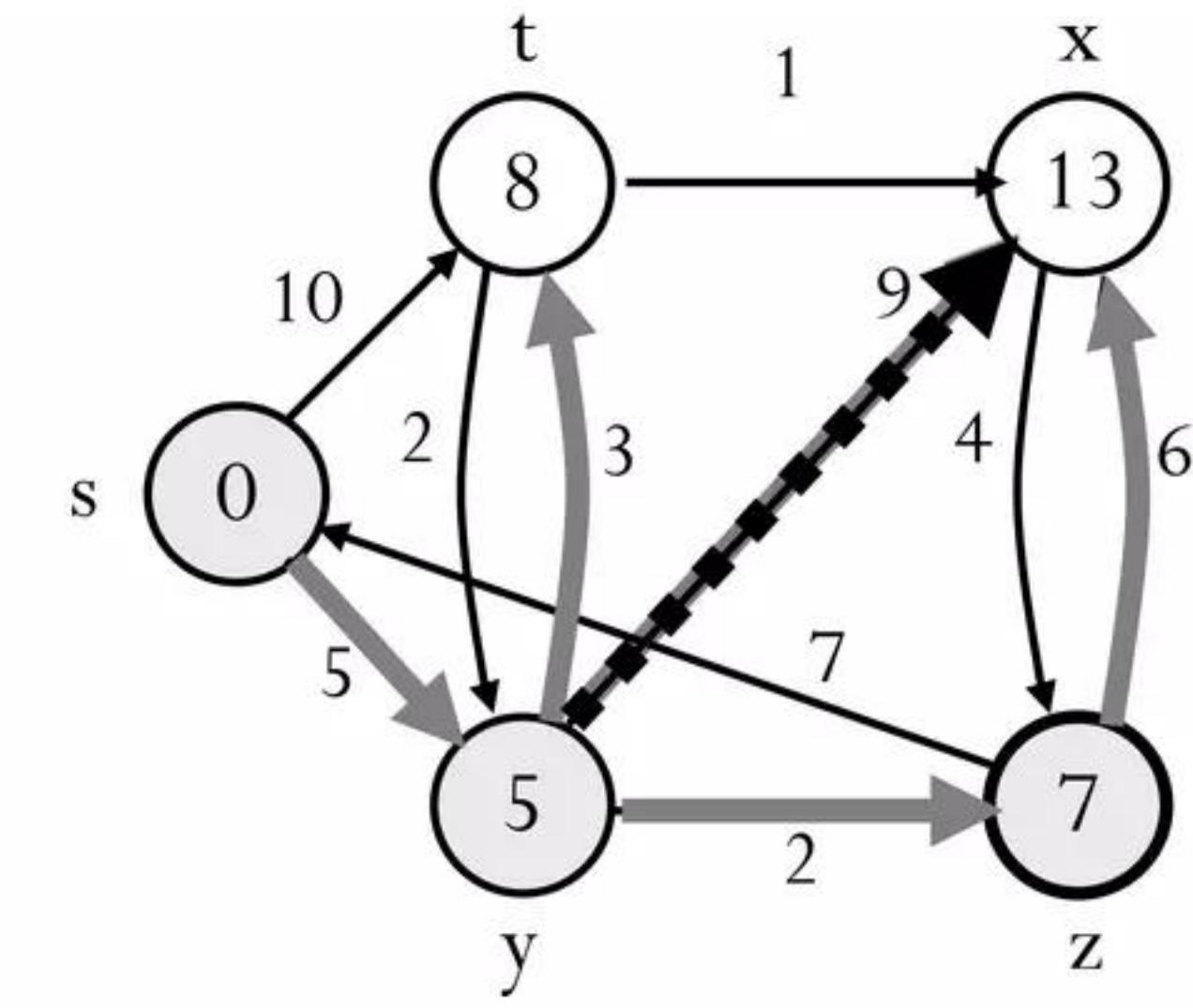
$S = \langle s \rangle$   $Q = \langle y, t, x, z \rangle$



## Example (cont.)



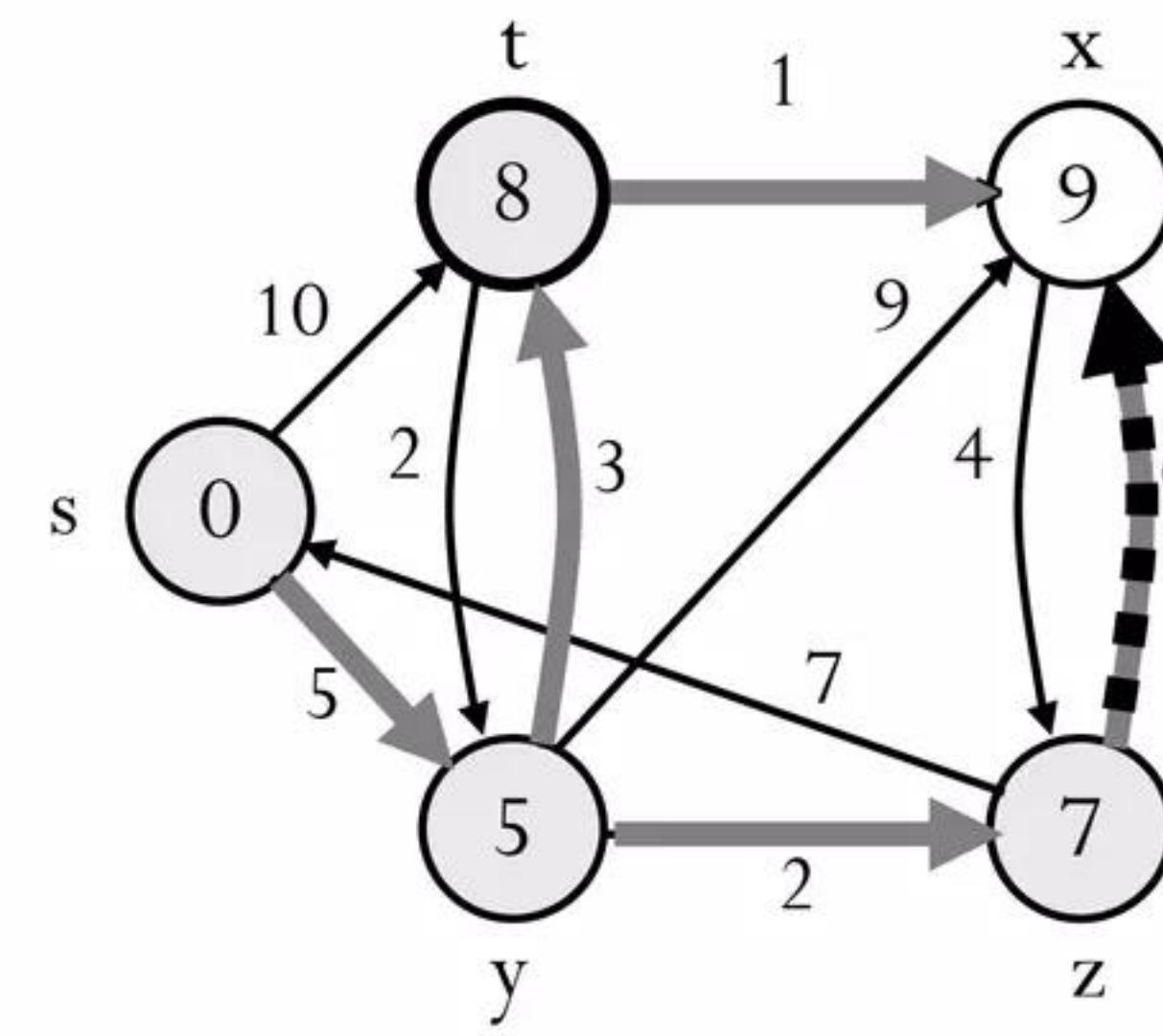
$$S = \langle s, y \rangle \quad Q = \langle z, t, x \rangle$$



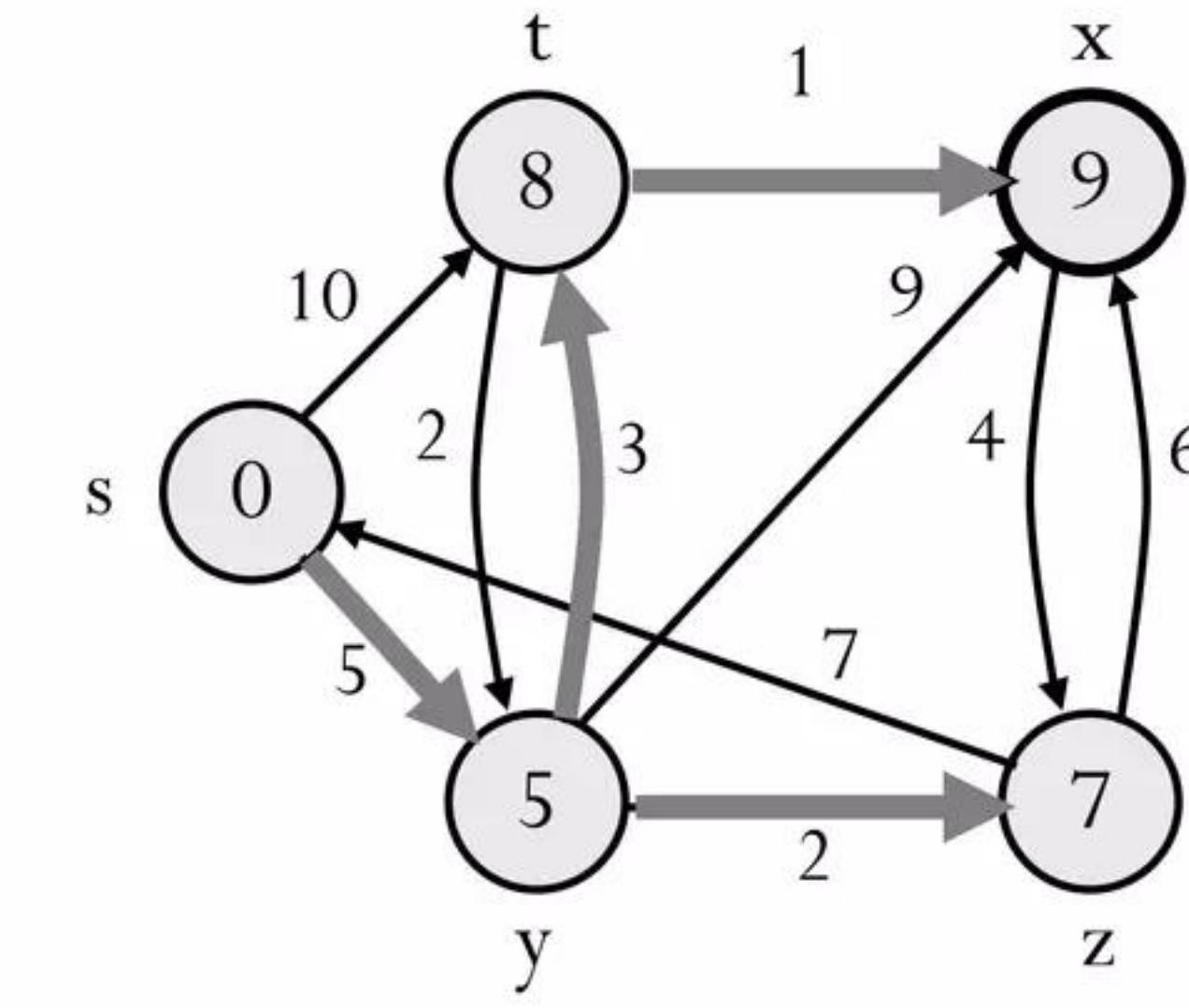
$$S = \langle s, y, z \rangle \quad Q = \langle t, x \rangle$$

## Example (cont.)

$S = \langle s, y, z, t \rangle$   $Q = \langle x \rangle$



$S = \langle s, y, z, t, x \rangle$   $Q = \langle \rangle$



# Dijkstra's Algorithm

- Running time:  $O(V \lg V + E \lg V) = O(E \lg V)$

DIJKSTRA( $G, w, s$ )

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )  $\leftarrow \Theta(V)$ 
2   $S = \emptyset$   $\xleftarrow{\quad} O(V)$  build min-heap
3   $Q = G.V$ 
4  while  $Q \neq \emptyset$   $\xleftarrow{\quad} O(V)$  times
5     $u = \text{EXTRACT-MIN}(Q) \leftarrow O(\lg V)$ 
6     $S = S \cup \{u\} \xleftarrow{\quad} O(V \lg V)$ 
7    for each vertex  $v \in G.Adj[u]$   $\xleftarrow{\quad} O(E)$  times
8      RELAX( $u, v, w$ )  $\xleftarrow{\quad} O(E \lg V)$ 
```

# Dijkstra's Algorithm

PQ implementation	insert	delete-min	decrease-key	total
<b>unordered array</b>	1	$V$	1	$V^2$
<b>binary heap</b>	$\log V$	$\log V$	$\log V$	$E \log V$
<b>d-way heap</b>	$\log_d V$	$d \log_d V$	$\log_d V$	$E \log_{E/V} V$
<b>Fibonacci heap</b>	$1^\dagger$	$\log V^\dagger$	$1^\dagger$	$E + V \log V$

$\dagger$  amortized

# Single Source Shortest-Paths Implementation

algorithm	restriction	typical case	worst case	extra space
<b>topological sort</b>	no directed cycles	$E + V$	$E + V$	$V$
<b>Dijkstra (binary heap)</b>	no negative weights	$E \log V$	$E \log V$	$V$
<b>Bellman-Ford</b>	no negative cycles	$E V$	$E V$	$V$
<b>Bellman-Ford (queue-based)</b>		$E + V$	$E V$	$V$

Remark 1. Directed cycles make the problem harder.

Remark 2. Negative weights make the problem harder.

Remark 3. Negative cycles makes the problem intractable.