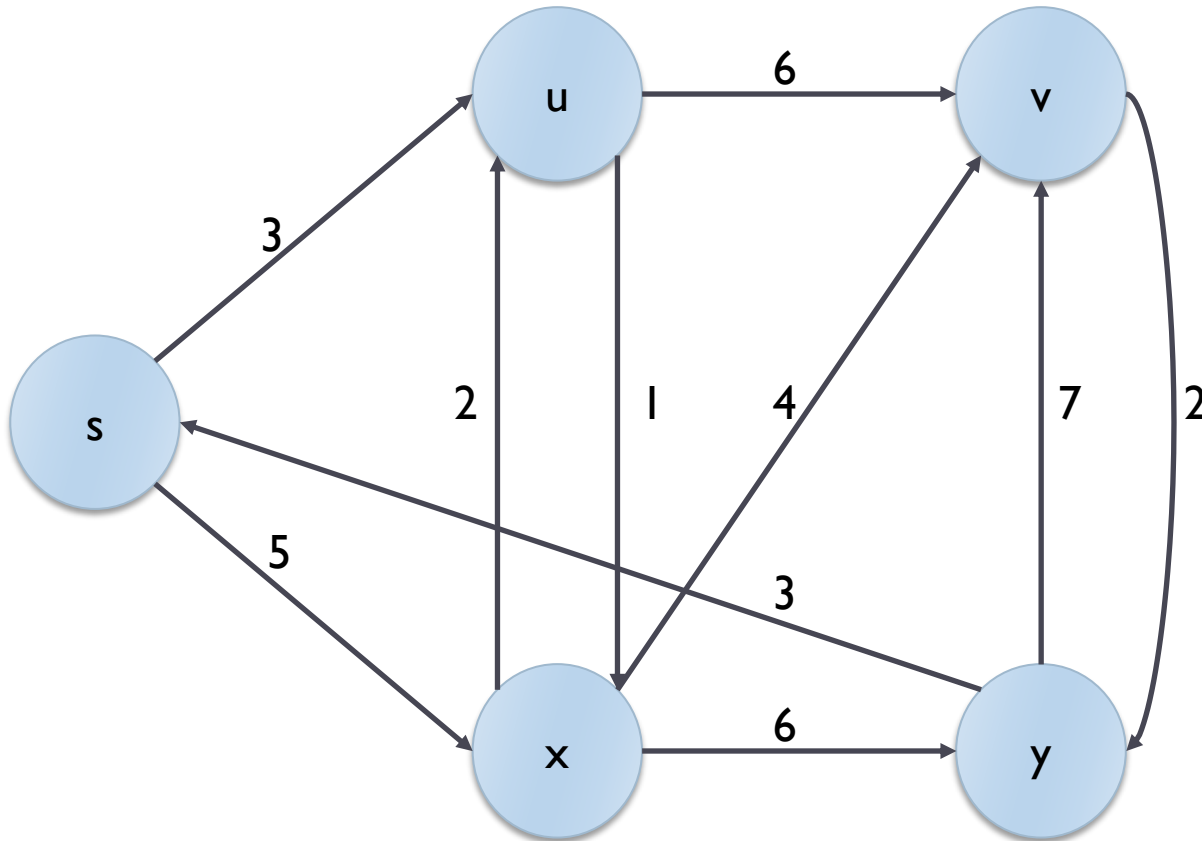




# Example

What is the shortest path between s and v ?



# Summary

---

- ▶ Definitions
- ▶ Floyd-Warshall algorithm
- ▶ Bellman-Ford-Moore algorithm
- ▶ Dijkstra algorithm



# Definition: weight of a path

---

- ▶ Consider a directed, weighted graph  $G=(V, E)$ , with weight function  $w: E \rightarrow \mathbb{R}$ 
  - ▶ This is the general case: undirected or un-weighted are automatically included
- ▶ The weight  $w(p)$  of a path  $p$  is the sum of the weights of the edges composing the path

$$w(p) = \sum_{(u,v) \in p} w(u, v)$$

# Definition: shortest path

---

- ▶ The shortest path between vertex  $u$  and vertex  $v$  is defined as the minimum-weight path between  $u$  and  $v$ , if the path exists.
- ▶ The weight of the shortest path is represented as  $\delta(u,v)$
- ▶ If  $v$  is not reachable from  $u$ , then (by definition)  $\delta(u,v)=\infty$

# Finding shortest paths

---

- ▶ **Single-source shortest path (SS-SP)**
  - ▶ Given  $u$  and  $v$ , find the shortest path between  $u$  and  $v$
  - ▶ Given  $u$ , find the shortest path between  $u$  and any other vertex
- ▶ **All-pairs shortest path (AP-SP)**
  - ▶ Given a graph, find the shortest path between any pair of vertices

# What to find?

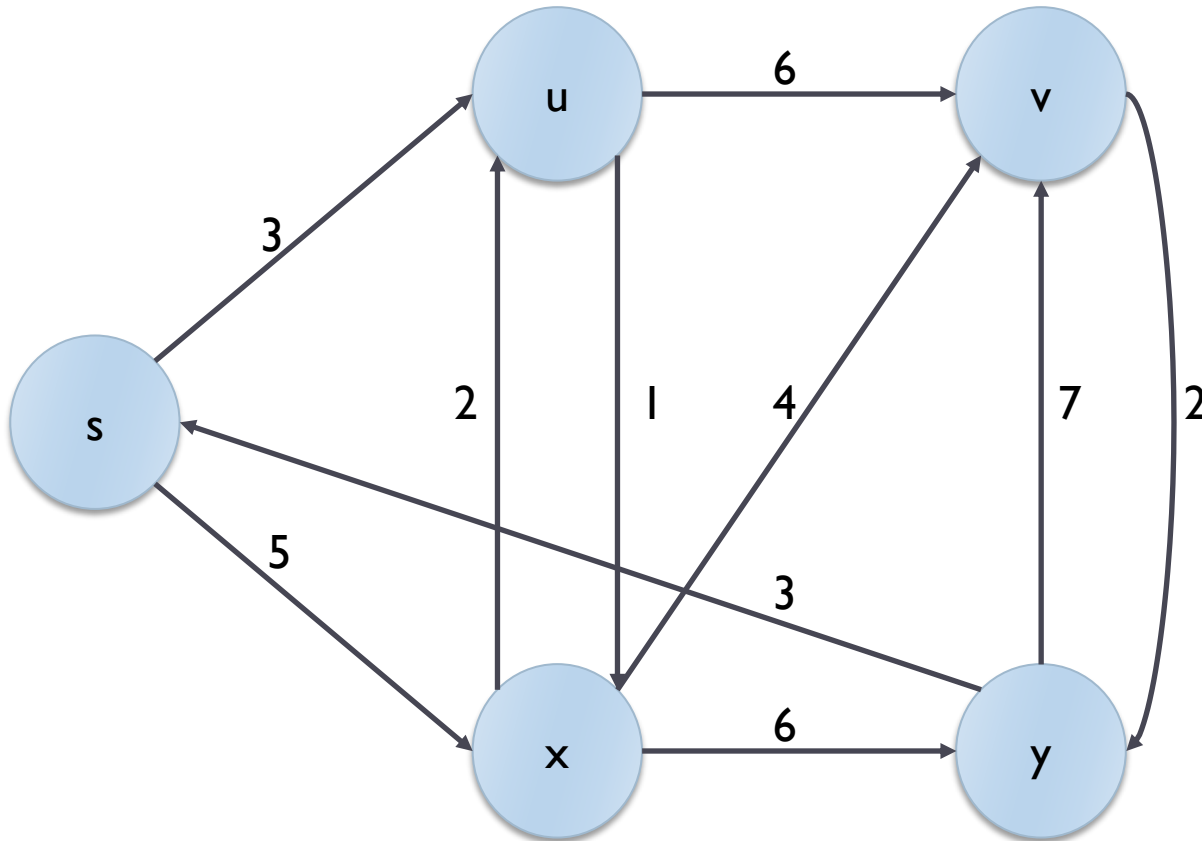
---

- ▶ Depending on the problem, you might want:
  - ▶ The **value** of the shortest path weight
    - ▶ Just a real number
  - ▶ The **actual path** having such minimum weight
    - ▶ For simple graphs, a sequence of vertices.
    - ▶ For multigraphs, a sequence of edges



# Example

What is the shortest path between s and v ?

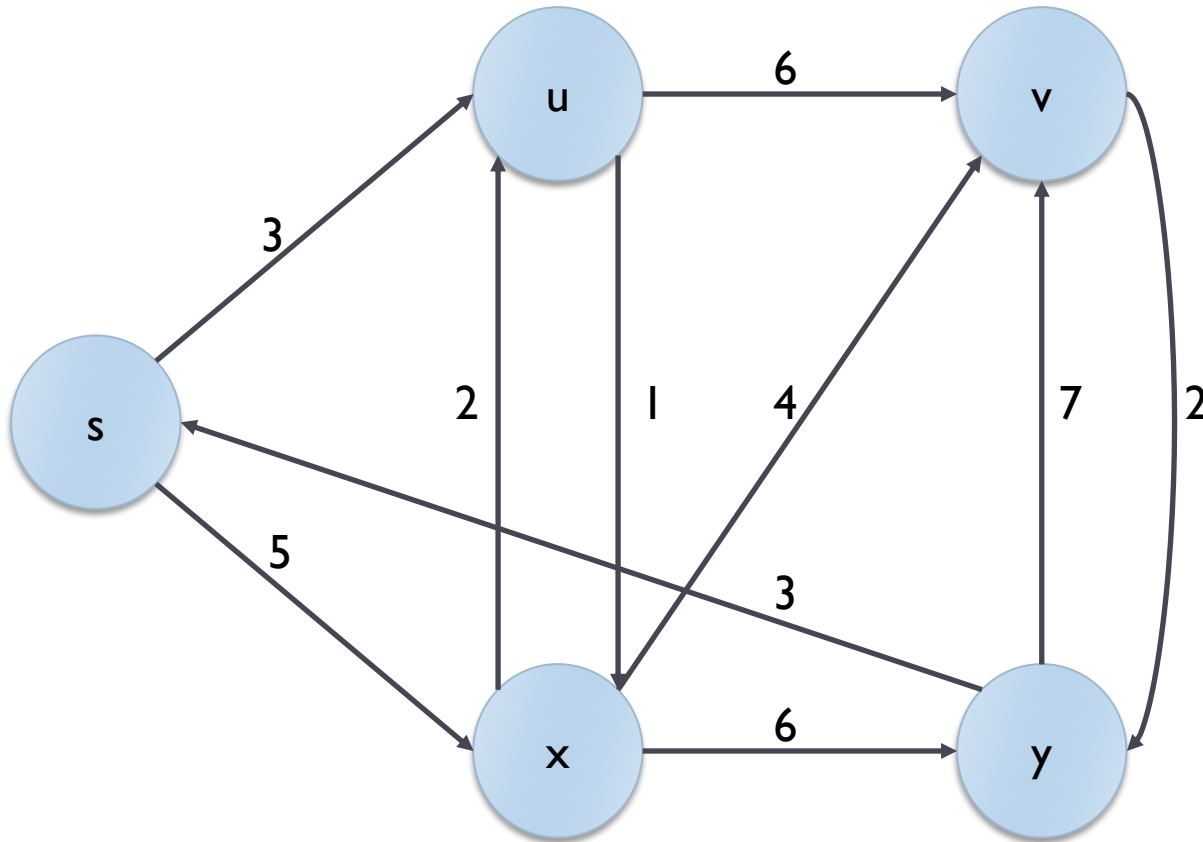


# Representing shortest paths

---

- ▶ To store all shortest paths from a single source  $u$ , we may add
  - ▶ For each vertex  $v$ , the **weight** of the shortest path  $\delta(u,v)$
  - ▶ For each vertex  $v$ , the “**preceding**” vertex  $\pi(v)$  that allows to reach  $v$  in the shortest path
    - ▶ For multigraphs, we need the preceding edge
- ▶ **Example:**
  - ▶ Source vertex:  $u$
  - ▶ For any vertex  $v$ :
    - ▶ `double v.weight ;`
    - ▶ `Vertex v.preceding ;`

# Example



$\pi$

Vertex	Previous
s	NULL
u	s
x	u
v	x
y	v

$\delta$

Vertex	Weight
s	0
u	3
x	4
v	8
y	10

# Lemma

---

- ▶ The “previous” vertex in an intermediate node of a minimum path does **not** depend on the **final** destination
- ▶ **Example:**
  - ▶ Let  $p_1$  = shortest path between  $u$  and  $v_1$
  - ▶ Let  $p_2$  = shortest path between  $u$  and  $v_2$
  - ▶ Consider a vertex  $w \in p_1 \cap p_2$
  - ▶ The value of  $\pi(w)$  may be chosen in a single way and still guarantee that both  $p_1$  and  $p_2$  are shortest

# Shortest path graph

---

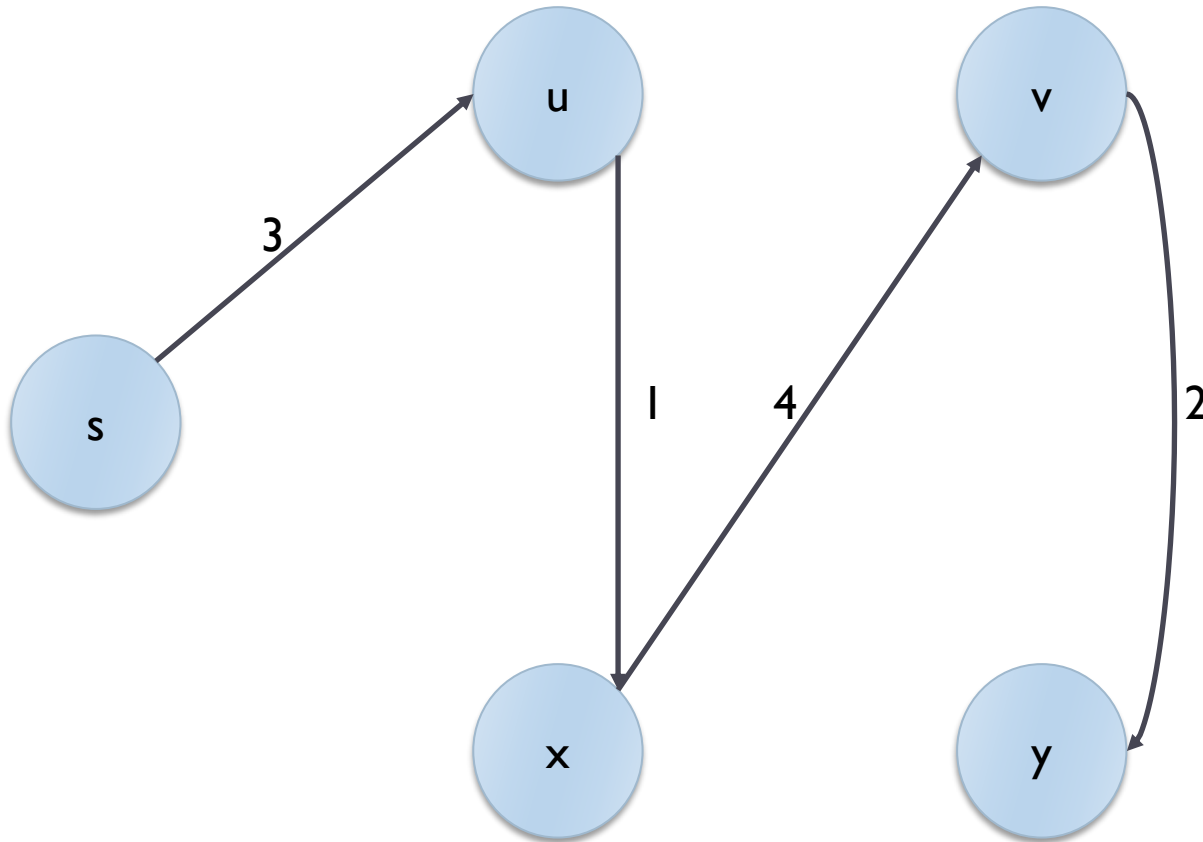
- ▶ Consider a source node  $u$
- ▶ Compute all shortest paths from  $u$
- ▶ Consider the relation  $E\pi = \{ (v.\text{preceding}, v) \}$
- ▶  $E\pi \subseteq E$
- ▶  $V\pi = \{ v \in V : v \text{ reachable from } u \}$
- ▶  $G\pi = G(V\pi, E\pi)$  is a subgraph of  $G(V, E)$
- ▶  $G\pi$ : the predecessor-subgraph

# Shortest path tree

---

- ▶  $G_\pi$  is a tree (due to the Lemma) rooted in  $u$
- ▶ In  $G_\pi$ , the (unique) paths starting from  $u$  are always shortest paths
- ▶  $G_\pi$  is not unique, but all possible  $G_\pi$  are equivalent (same weight for every shortest path)

# Example



$\pi$

Vertex	Previous
s	NULL
u	s
x	u
v	x
y	v

$\delta$

Vertex	Weight
s	0
u	3
x	4
v	8
y	10

# Special case

---

- ▶ If  $G$  is an un-weighted graph, then the shortest paths may be computed just with a breadth-first visit



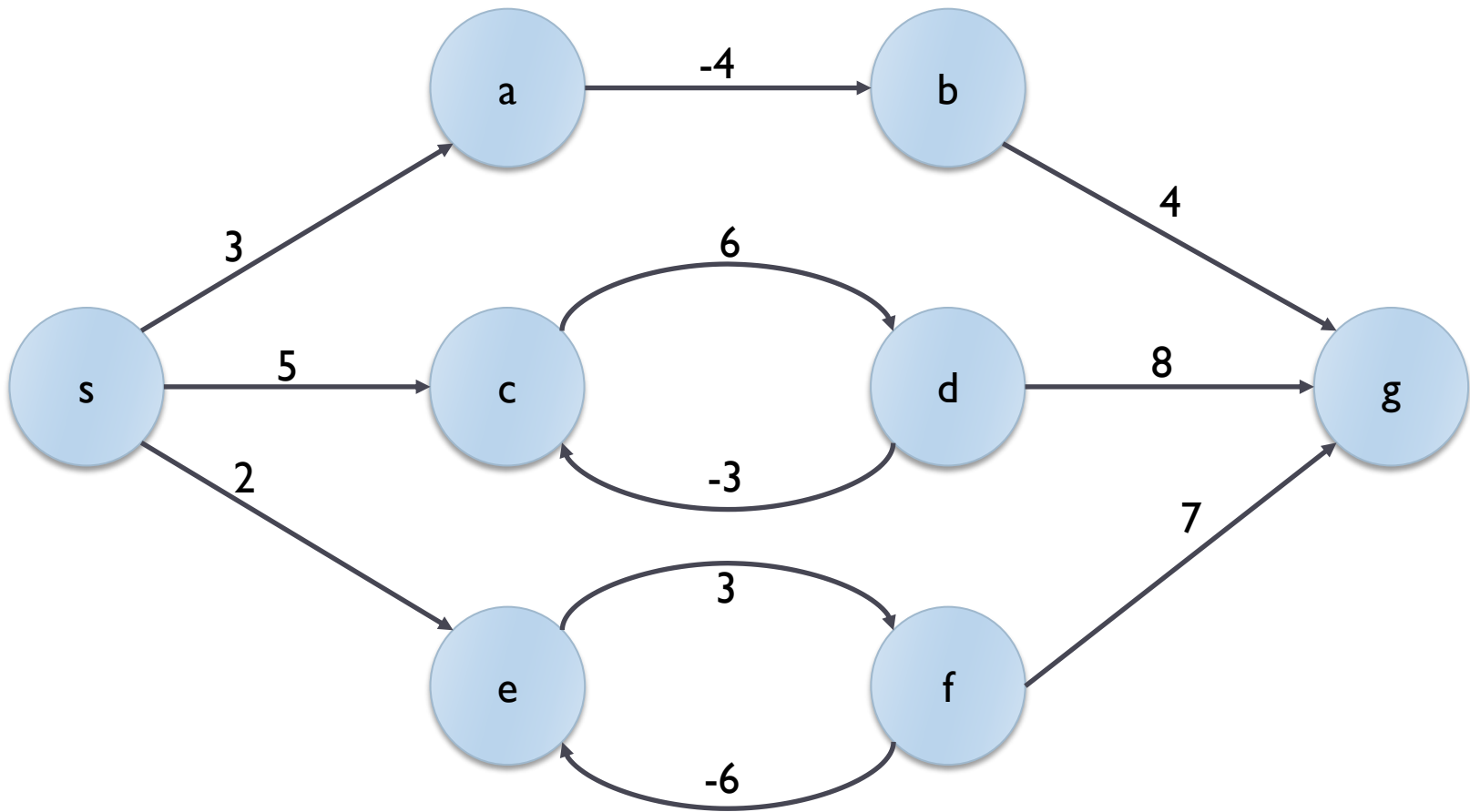
# Negative-weight cycles

---

- ▶ Minimum paths cannot be defined if there are negative-weight cycles in the graph
- ▶ In this case, the minimum path does not exist, because you may always decrease the path weight by going once more through the loop.
- ▶ Conventionally, in these case we say that the path weight is  $-\infty$ .

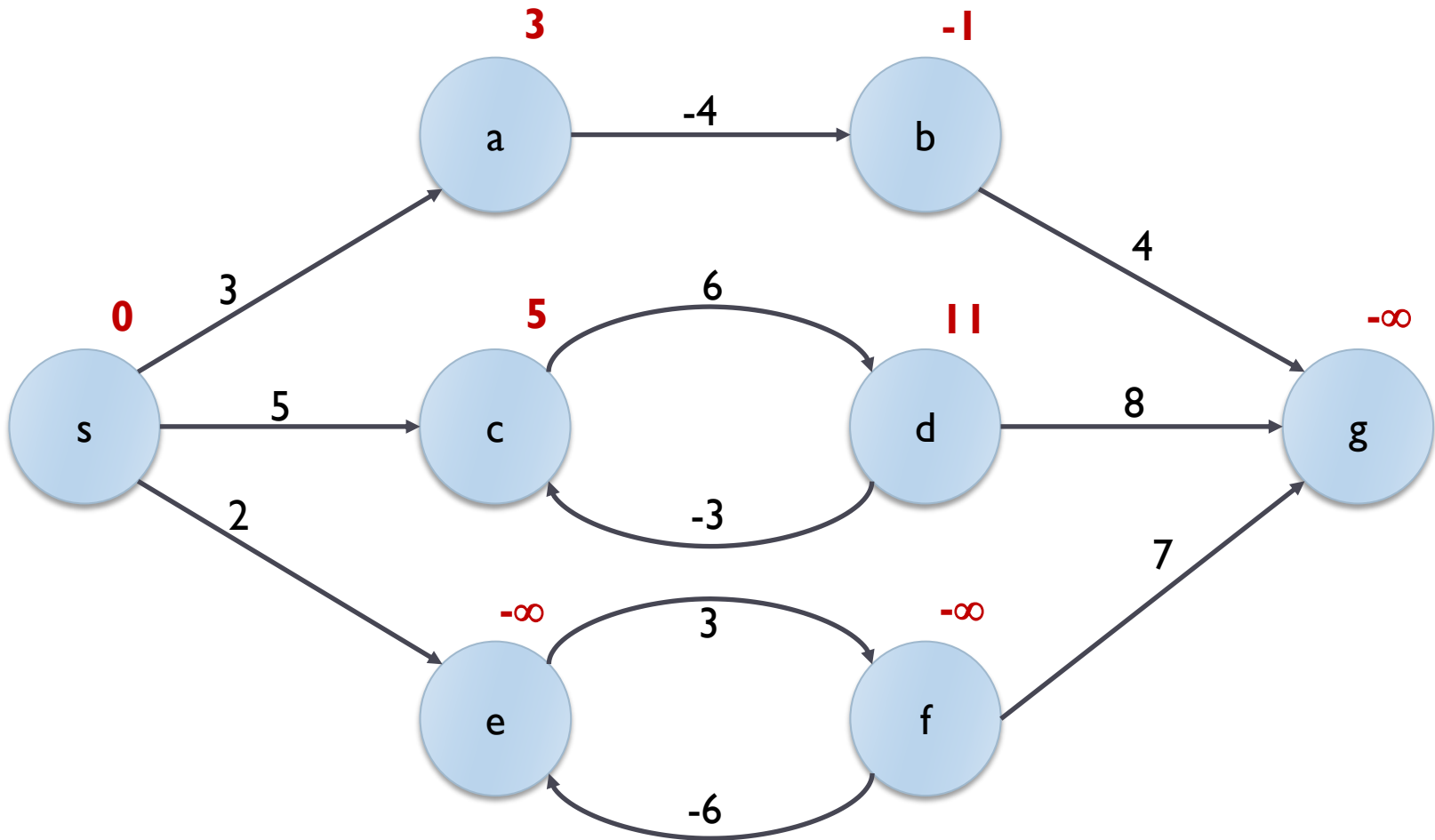
# Example

---



# Example

Minimum-weight paths from source vertex s



# Lemma

---

- ▶ Consider an ordered weighted graph  $G=(V,E)$ , with weight function  $w: E \rightarrow \mathbb{R}$ .
- ▶ Let  $p = \langle v_1, v_2, \dots, v_k \rangle$  a shortest path from vertex  $v_1$  to vertex  $v_k$ .
- ▶ For all  $i, j$  such that  $1 \leq i \leq j \leq k$ , let  $p_{ij} = \langle v_i, v_{i+1}, \dots, v_j \rangle$  be the sub-path of  $p$ , from vertex  $v_i$  to vertex  $v_j$ .
- ▶ Therefore,  $p_{ij}$  is a shortest path from  $v_i$  to  $v_j$ .

# Corollary

---

- ▶ Let  $p$  be a shortest path from  $s$  to  $v$
- ▶ Consider the vertex  $u$ , such that  $(u,v)$  is the last edge in the shortest path
- ▶ We may decompose  $p$  (from  $s$  to  $v$ ) into:
  - ▶ A sub-path from  $s$  to  $u$
  - ▶ The final edge  $(u,v)$
- ▶ Therefore
  - ▶  $\delta(s,v) = \delta(s,u) + w(u,v)$

# Lemma

---

- ▶ If we arbitrarily chose the vertex  $u'$ , then for all edges  $(u',v) \in E$  we may say that
  - ▶  $\delta(s,v) \leq \delta(s,u') + w(u',v)$

# Relaxation

---

- ▶ Most shortest-path algorithms are based on the relaxation technique
- ▶ It consists of
  - ▶ Vector  $d[u]$  represents  $\delta(s,u)$
  - ▶ Keeping track of an updated estimate  $d[u]$  of the shortest path towards each node  $u$
  - ▶ Relaxing (i.e., updating)  $d[v]$  (and therefore the predecessor  $\pi[v]$ ) whenever we discover that node  $v$  is more conveniently reached by traversing edge  $(u,v)$

# Initial state

---

► **Initialize-Single-Source( $G(V,E), s$ )**

1. **for** all vertices  $v \in V$
2. **do**
  1.  $d[v] \leftarrow \infty$
  2.  $\pi[v] \leftarrow \text{NIL}$
3.  $d[s] \leftarrow 0$

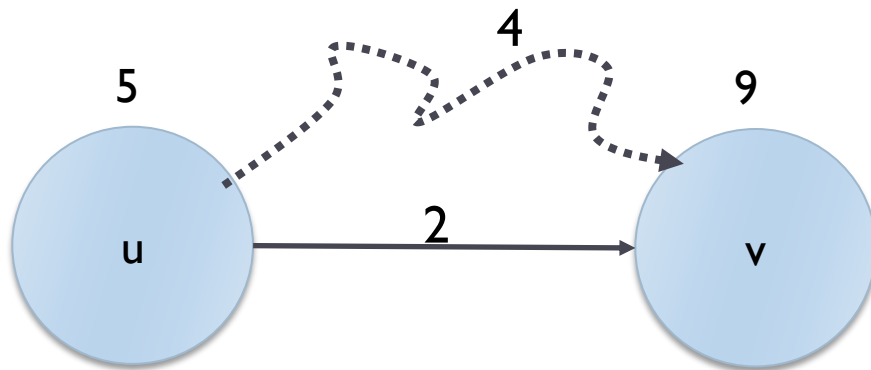


# Relaxation

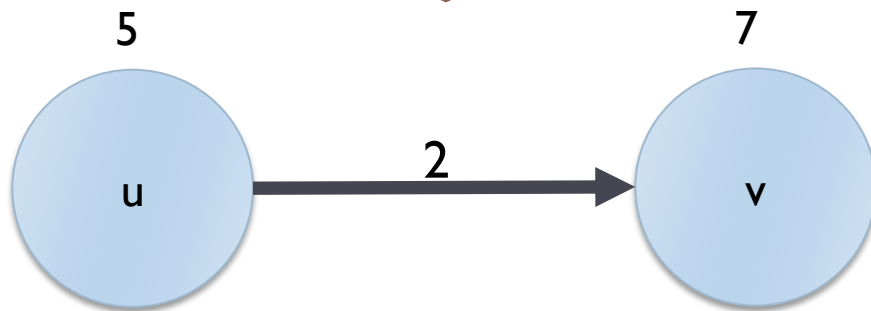
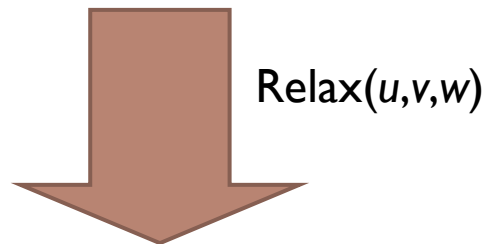
---

- ▶ We consider an edge  $(u,v)$  with weight  $w$
- ▶  $\text{Relax}(u, v, w)$ 
  1. **if**  $d[v] > d[u] + w(u,v)$
  2. **then**
    1.  $d[v] \leftarrow d[u] + w(u,v)$
    2.  $\pi[v] \leftarrow u$

# Example 1

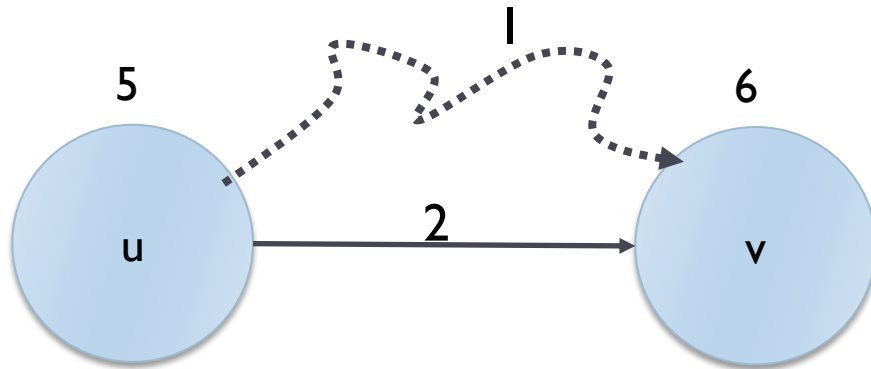


Before:  
Shortest known path to  $v$  weights 9, does not contain  $(u,v)$

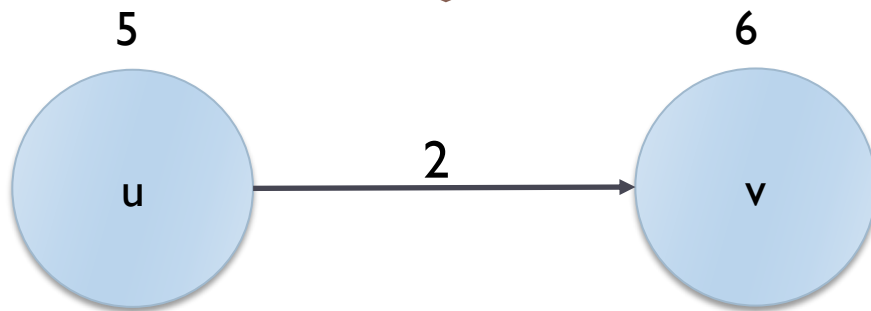
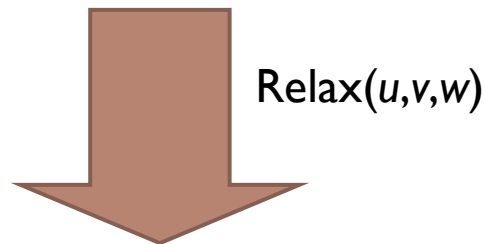


After:  
Shortest path to  $v$  weights 7, the path includes  $(u,v)$

## Example 2



Before:  
Shortest path to  $v$   
weights 6, does not  
contain  $(u,v)$



After:  
No relaxation possible,  
shortest path unchanged

# Lemma

---

- ▶ Consider an ordered weighted graph  $G=(V, E)$ , with weight function  $w: E \rightarrow \mathbb{R}$ .
- ▶ Let  $(u,v)$  be an edge in  $G$ .
- ▶ After relaxation of  $(u,v)$  we may write that:
  - ▶  $d[v] \leq d[u] + w(u,v)$

# Lemma

---

- ▶ Consider an ordered weighted graph  $G=(V, E)$ , with weight function  $w: E \rightarrow \mathbb{R}$  and source vertex  $s \in V$ . Assume that  $G$  has no negative-weight cycles reachable from  $s$ .
- ▶ Therefore
  - ▶ After calling `Initialize-Single-Source(G,s)`, the predecessor subgraph  $G_\pi$  is a rooted tree, with  $s$  as the root.
  - ▶ Any relaxation we may apply to the graph does not invalidate this property.

# Lemma

---

- ▶ Given the previous definitions.
- ▶ Apply any possible sequence of relaxation operations
- ▶ Therefore, for each vertex  $v$ 
  - ▶  $d[v] \geq \delta(s,v)$
- ▶ Additionally, if  $d[v] = \delta(s,v)$ , then the value of  $d[v]$  will not change anymore due to relaxation operations.

# Shortest path algorithms

---

- ▶ Various algorithms
- ▶ Differ according to one-source or all-sources requirement
- ▶ Adopt repeated relaxation operations
- ▶ Vary in the order of relaxation operations they perform
- ▶ May be applicable (or not) to graph with negative edges (but no negative cycles)

# Implementations

## Package `org.jgrapht.alg.shortestpath`

### Class Summary

Class	Description
<code>AllDirectedPaths&lt;V,E&gt;</code>	A Dijkstra-like algorithm to find all paths between two sets of nodes in a directed graph, with options to search only simple paths and to limit the path length.
<code>ALTAdmissibleHeuristic&lt;V,E&gt;</code>	An admissible heuristic for the A* algorithm using a set of landmarks and the triangle inequality.
<code>AShortestPath&lt;V,E&gt;</code>	A* shortest path.
<code>BellmanFordShortestPath&lt;V,E&gt;</code>	The Bellman-Ford algorithm.
<code>BhandariKDisjointShortestPaths&lt;V,E&gt;</code>	An implementation of Bhandari algorithm for finding $K$ edge- <i>disjoint</i> shortest paths.
<code>BidirectionalDijkstraShortestPath&lt;V,E&gt;</code>	A bidirectional version of Dijkstra's algorithm.
<code>DijkstraShortestPath&lt;V,E&gt;</code>	An implementation of <b>Dijkstra's shortest path algorithm</b> using a Fibonacci heap.
<code>FloydWarshallShortestPaths&lt;V,E&gt;</code>	The Floyd-Warshall algorithm.
<code>GraphMeasurer&lt;V,E&gt;</code>	Algorithm class which computes a number of distance related metrics.
<code>JohnsonShortestPaths&lt;V,E&gt;</code>	Johnson's all pairs shortest paths algorithm.
<code>KShortestSimplePaths&lt;V,E&gt;</code>	The algorithm determines the k shortest simple paths in increasing order of weight.
<code>ListMultiObjectiveSingleSourcePathsImpl&lt;V,E&gt;</code>	An implementation of <b>MultiObjectiveShortestPathAlgorithm.MultiObjectiveSingleSourcePaths</b> which stores one list of paths per vertex.
<code>ListSingleSourcePathsImpl&lt;V,E&gt;</code>	An implementation of <b>ShortestPathAlgorithm.SingleSourcePaths</b> which stores one path per vertex.
<code>MartinShortestPath&lt;V,E&gt;</code>	Martin's algorithm for the multi-objective shortest paths problem.
<code>SuurballeKDisjointShortestPaths&lt;V,E&gt;</code>	An implementation of Suurballe algorithm for finding $K$ edge- <i>disjoint</i> shortest paths.
<code>TreeMeasurer&lt;V,E&gt;</code>	Algorithm class which computes a number of distance related metrics for trees.
<code>TreeSingleSourcePathsImpl&lt;V,E&gt;</code>	An implementation of <b>ShortestPathAlgorithm.SingleSourcePaths</b> which uses linear space.





# Floyd-Warshall algorithm

- ▶ Computes the all-source shortest path (AP-SP)
- ▶  $dist[i][j]$  is an  $n$ -by- $n$  matrix that contains the length of a shortest path from  $v_i$  to  $v_j$ .
- ▶ if  $dist[u][v]$  is  $\infty$ , there is no path from  $u$  to  $v$
- ▶  $pred[s][j]$  is used to reconstruct an actual shortest path: stores the predecessor vertex for reaching  $v_j$  starting from source  $v_s$

FLOYD-WARSHALL			 Weighted Directed Graph	Overflow
Best	Average	Worst		
$O(V^3)$	$O(V^3)$	$O(V^3)$	Dynamic Programming	2D Array

**allPairsShortestPath (G)**

1. **foreach**  $u \in V$  **do**
2.     **foreach**  $v \in V$  **do**
3.          $dist[u][v] = \infty$
4.          $pred[u][v] = -1$
5.          $dist[u][u] = 0$
6.     **foreach** neighbor  $v$  of  $u$  **do**
7.          $dist[u][v] = \text{weight of edge } (u,v)$
8.          $pred[u][v] = u$

Initialize  $dist[][]$  matrix with existing edges

	0	1	2	3	4
0	0	2	$\infty$	$\infty$	4
1	$\infty$	0	3	$\infty$	$\infty$
2	$\infty$	$\infty$	0	5	1
3	8	$\infty$	$\infty$	0	$\infty$
4	$\infty$	$\infty$	$\infty$	7	0

**dist[u][v]**

For each vertex  $t \in V$ , reduce paths between each pair of  $(u,v)$  vertices through  $t$  when possible

9. **foreach**  $t \in V$  **do**
10.     **foreach**  $u \in V$  **do**
11.         **foreach**  $v \in V$  **do**
12.              $newLen = dist[u][t] + dist[t][v]$
13.             **if** ( $newLen < dist[u][v]$ ) **then**
14.                  $dist[u][v] = newLen$
15.                  $pred[u][v] = pred[t][v]$

**end**

$t=1$

	0	1	2	3	4
0	0	2	$\infty$	$\infty$	4
1	$\infty$	0	3	$\infty$	$\infty$
2	$\infty$	$\infty$	0	5	1
3	8	10	0	0	12
4	$\infty$	$\infty$	$\infty$	7	0

$t=2$

	0	1	2	3	4
0	0	2	5	10	4
1	$\infty$	0	3	8	4
2	$\infty$	$\infty$	0	5	1
3	8	10	13	0	12
4	$\infty$	$\infty$	$\infty$	7	0

$t=3$


	0	1	2	3	4
0	0	2	5	10	4
1	16	0	3	8	4
2	13	15	0	5	1
3	8	10	13	0	12
4	15	17	20	7	0

This is the final result since processing vertex 4 has no impact

# Floyd-Warshall: initialization

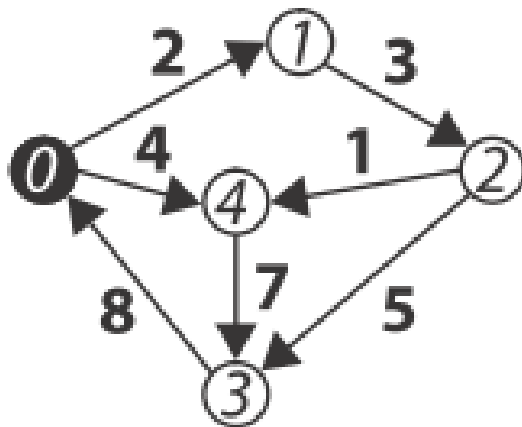
---

## **allPairsShortestPath** (G)

1. **foreach**  $u \in V$  **do**
2.     **foreach**  $v \in V$  **do** 
3.          $\text{dist}[u][v] = \infty$
4.          $\text{pred}[u][v] = -1$
5.      $\text{dist}[u][u] = 0$
6.     **foreach** neighbor  $v$  of  $u$  **do**
7.          $\text{dist}[u][v] = \text{weight of edge } (u,v)$
8.          $\text{pred}[u][v] = u$

# Example, after initialization

---




	0	1	2	3	4
0	0	2	$\infty$	$\infty$	4
1	$\infty$	0	3	$\infty$	$\infty$
2	$\infty$	$\infty$	0	5	1
3	8	$\infty$	$\infty$	0	$\infty$
4	$\infty$	$\infty$	$\infty$	7	0

**dist[u][v]**

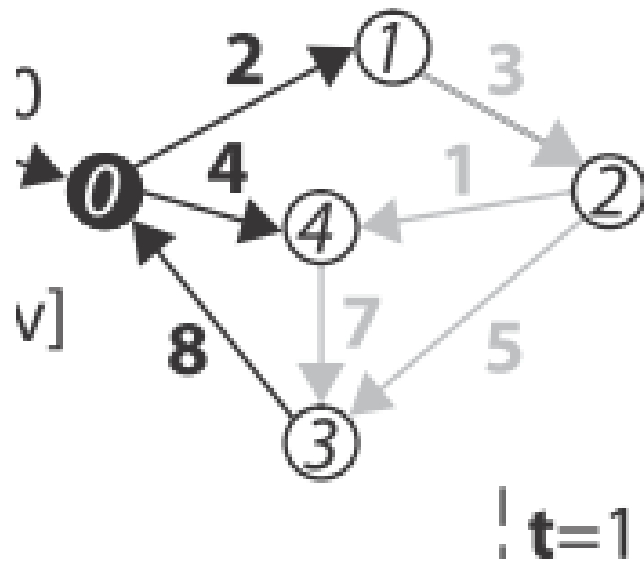
# Floyd-Warshall: relaxation

---

```
9.  foreach  $t \in V$  do
10.   foreach  $u \in V$  do
11.    foreach  $v \in V$  do
12.     newLen = dist[u][t] + dist[t][v]
13.     if (newLen < dist[u][v]) then
14.       dist[u][v] = newLen
15.       pred[u][v] = pred[t][v]
```

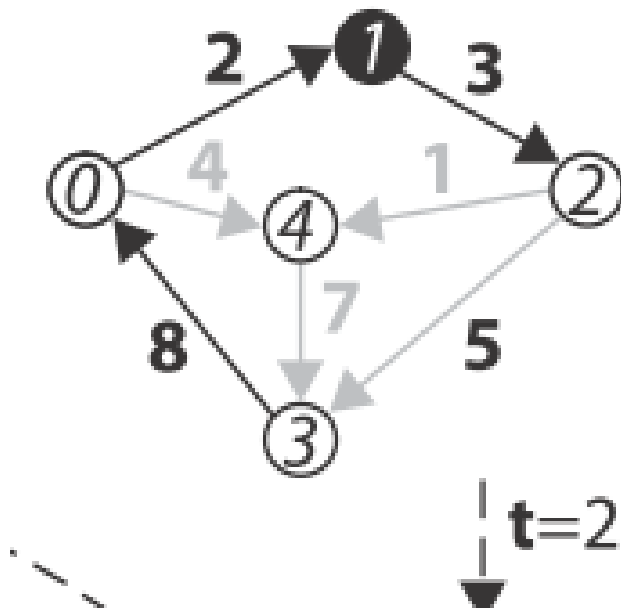


# Example, after step $t=0$



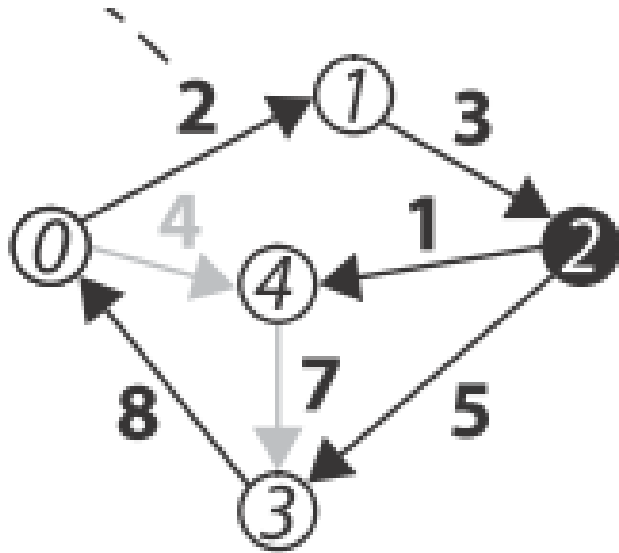
	0	1	2	3	4
0	0	2	$\infty$	$\infty$	4
1	$\infty$	0	3	$\infty$	$\infty$
2	$\infty$	$\infty$	0	5	1
3	8	10	$\infty$	0	12
4	$\infty$	$\infty$	$\infty$	7	0

# Example, after step $t=1$



	0	1	2	3	4
0	0	2	5	$\infty$	4
1	$\infty$	0	3	$\infty$	$\infty$
2	$\infty$	$\infty$	0	5	1
3	8	10	13	0	12
4	$\infty$	$\infty$	$\infty$	7	0

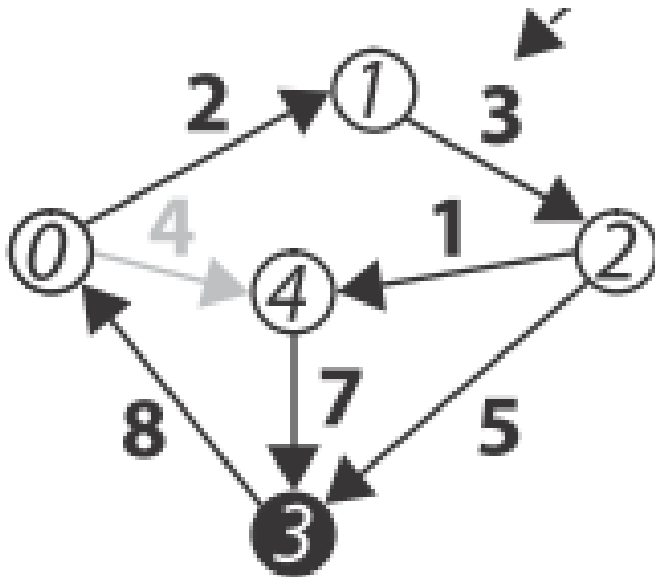
# Example, after step $t=2$



	0	1	2	3	4
0	0	2	5	10	4
1	$\infty$	0	3	8	4
2	$\infty$	$\infty$	0	5	1
3	8	10	13	0	12
4	$\infty$	$\infty$	$\infty$	7	0



# Example, after step $t=3$



	0	1	2	3	4
0	0	2	5	10	4
1	16	0	3	8	4
2	13	15	0	5	1
3	8	10	13	0	12
4	15	17	20	7	0

# Complexity

---

- ▶ The Floyd-Warshall is basically executing 3 nested loops, each iterating over all vertices in the graph
- ▶ Complexity:  $O(V^3)$

# Implementation

OVERVIEW PACKAGE **CLASS** USE TREE DEPRECATED INDEX HELP

PREV CLASS NEXT CLASS FRAMES NO FRAMES

SUMMARY: NESTED | FIELD | CONSTR | METHOD DETAIL: FIELD | CONSTR | METHOD

org.jgrapht.alg.shortestpath

## Class FloydWarshallShortestPaths<V,E>

java.lang.Object  
org.jgrapht.alg.shortestpath.FloydWarshallShortestPaths<V,E>

### Type Parameters:

V - the graph vertex type

E - the graph edge type

### All Implemented Interfaces:

ShortestPathAlgorithm<V,E>

```
public class FloydWarshallShortestPaths<V,E>  
extends Object
```

The Floyd-Warshall algorithm.

The Floyd-Warshall algorithm finds all shortest paths (all  $n^2$  of them) in  $O(n^3)$  time. Note that during construction time, no computations are performed! All computations are performed the first time one of the member methods of this class is invoked. The results are stored, so all subsequent calls to the same method are computationally efficient.

### Author:

Tom Larkworthy, Soren Davidsen (soren@tanasha.net), Joris Kinable, Dimitrios Michail

### Nested Class Summary

Nested classes/interfaces inherited from interface org.jgrapht.alg.interfaces.ShortestPathAlgorithm

ShortestPathAlgorithm.SingleSourcePaths<V,E>



# Bellman-Ford-Moore Algorithm

---

- ▶ Solution to the single-source shortest path (SS-SP) problem in graph theory
- ▶ Based on relaxation (for every vertex, relax all possible edges)
- ▶ Does not work in presence of negative cycles
  - ▶ but it is able to detect the problem
- ▶  $O(V \cdot E)$

# Bellman-Ford-Moore Algorithm

---

```
dist[s] ← 0 (distance to source vertex is zero)
for all  $v \in V - \{s\}$ 
  do  $\text{dist}[v] \leftarrow \infty$  (set all other distances to infinity)
for  $i \leftarrow 0$  to  $|V|$ 
  for all  $(u, v) \in E$ 
    do if  $\text{dist}[v] > \text{dist}[u] + w(u, v)$  (if new shortest path found)
      then  $d[v] \leftarrow d[u] + w(u, v)$  (set new value of shortest path)
        (if desired, add traceback code)

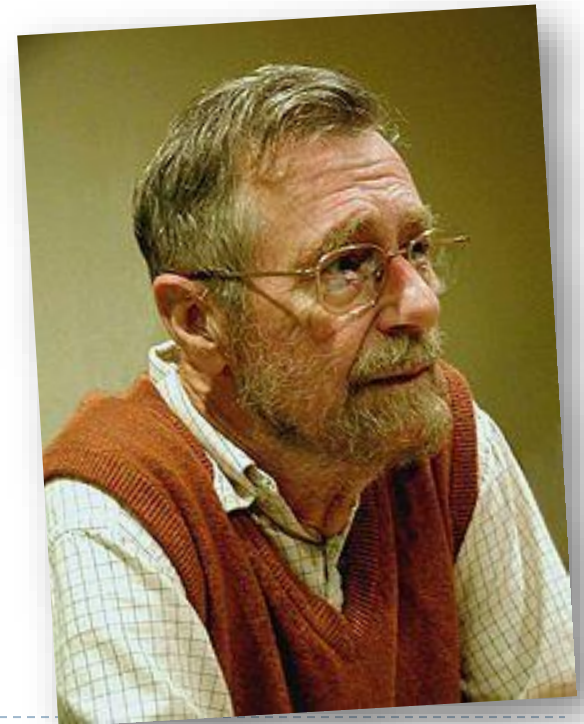
for all  $(u, v) \in E$  (sanity check)
  do if  $\text{dist}[v] > \text{dist}[u] + w(u, v)$ 
    then PANIC!
```



# Dijkstra's algorithm

---

- ▶ Solution to the single-source shortest path (SS-SP) problem in graph theory
- ▶ Works on both directed and undirected graphs
- ▶ All edges must have nonnegative weights
  - ▶ the algorithm would miserably fail
- ▶ Greedy
  - ... but guarantees the optimum!





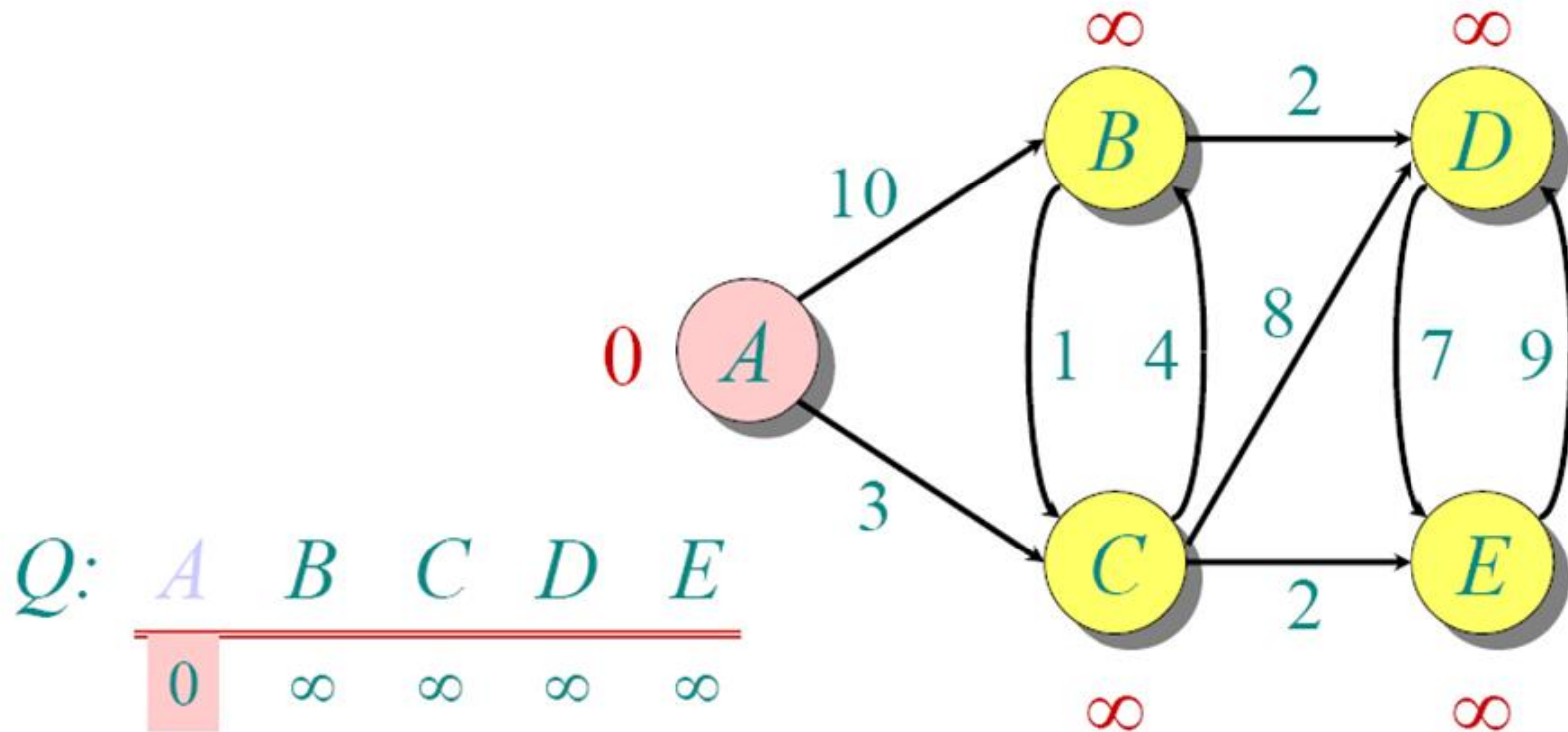
# Dijkstra's algorithm

---

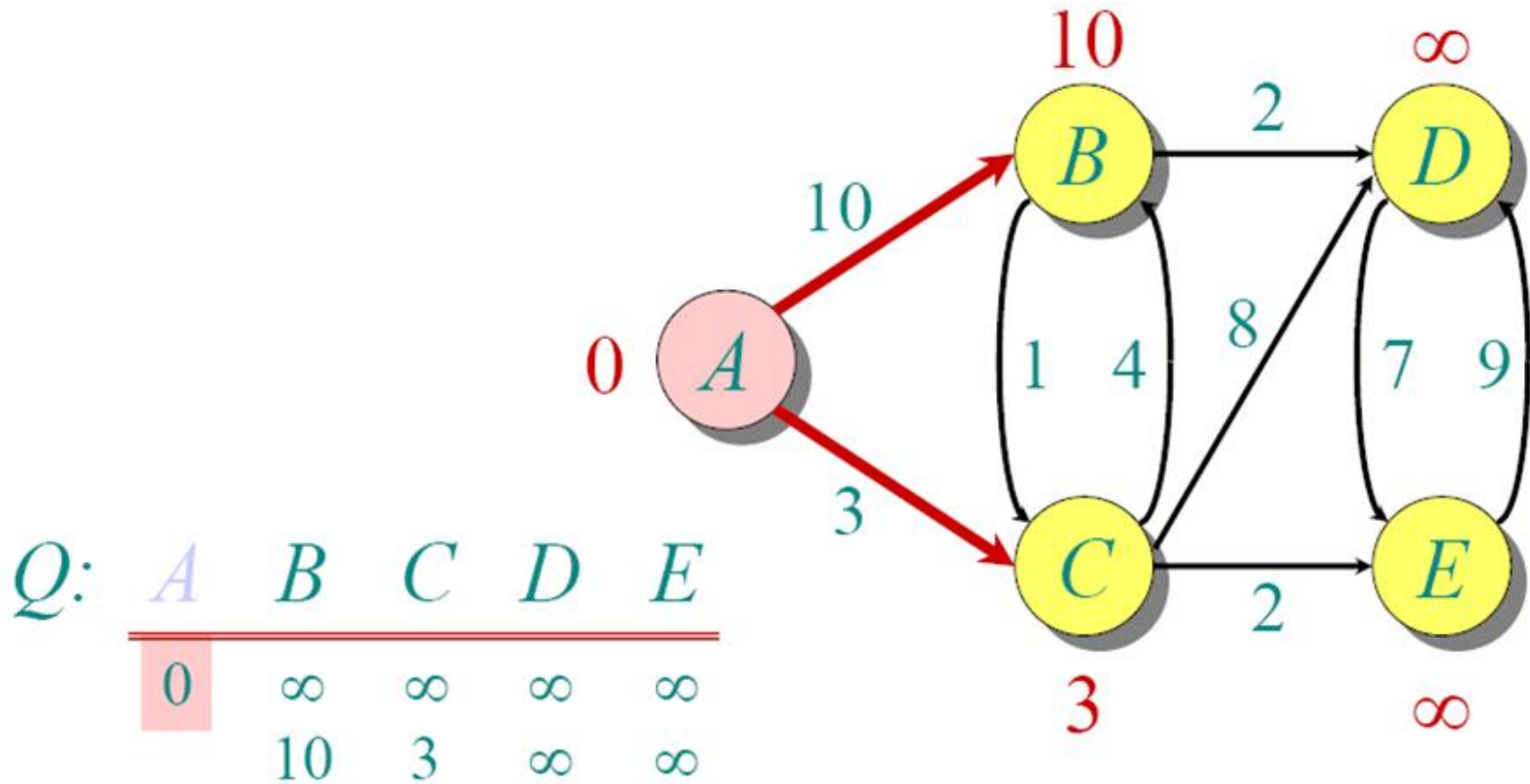
$\text{dist}[s] \leftarrow 0$  (distance to source vertex is zero)  
for all  $v \in V - \{s\}$   
do  $\text{dist}[v] \leftarrow \infty$  (set all other distances to infinity)  
 $S \leftarrow \emptyset$  ( $S$ , the set of visited vertices is initially empty)  
 $Q \leftarrow V$  ( $Q$ , the queue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



# Dijkstra Animated Example

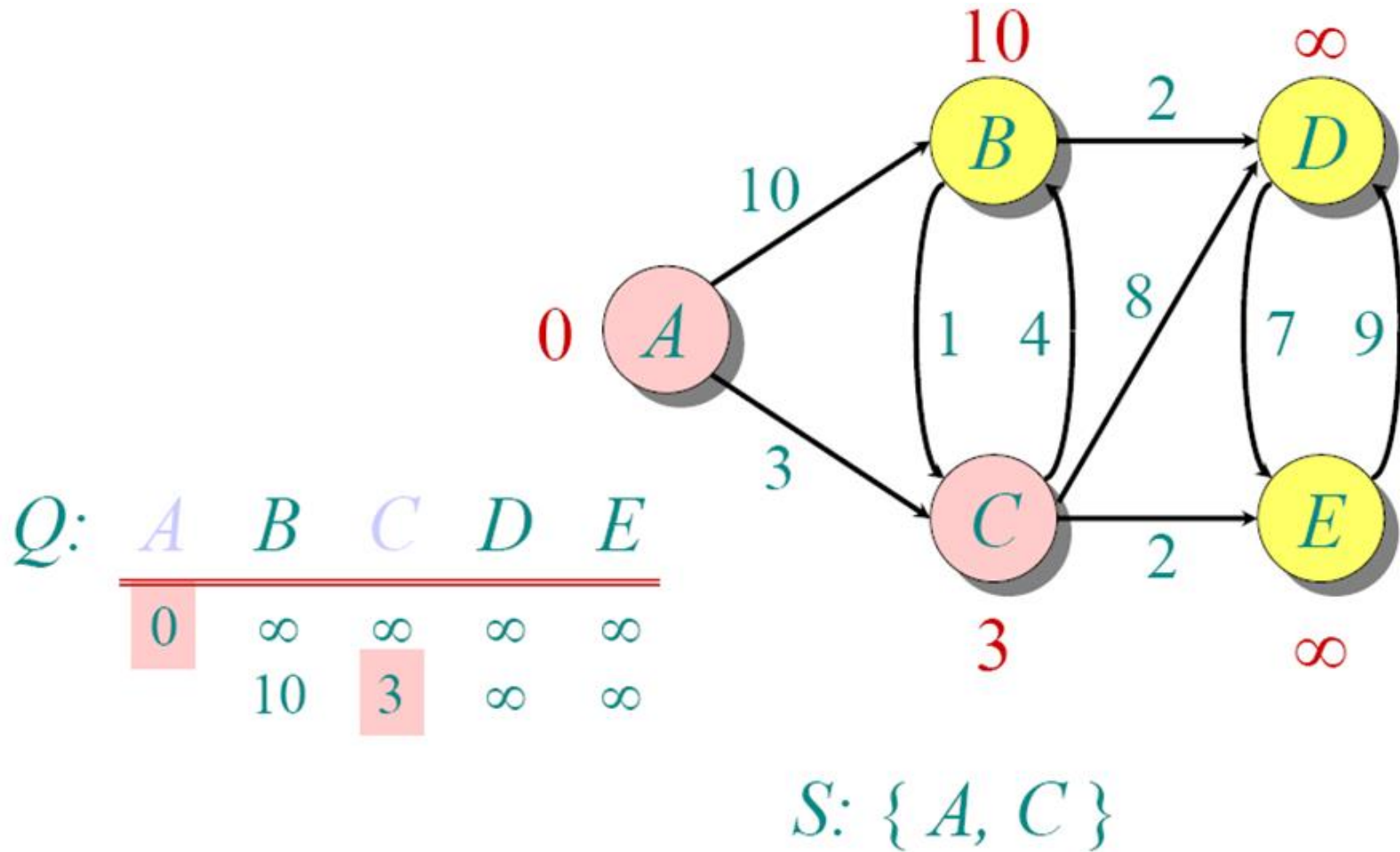


# Dijkstra Animated Example

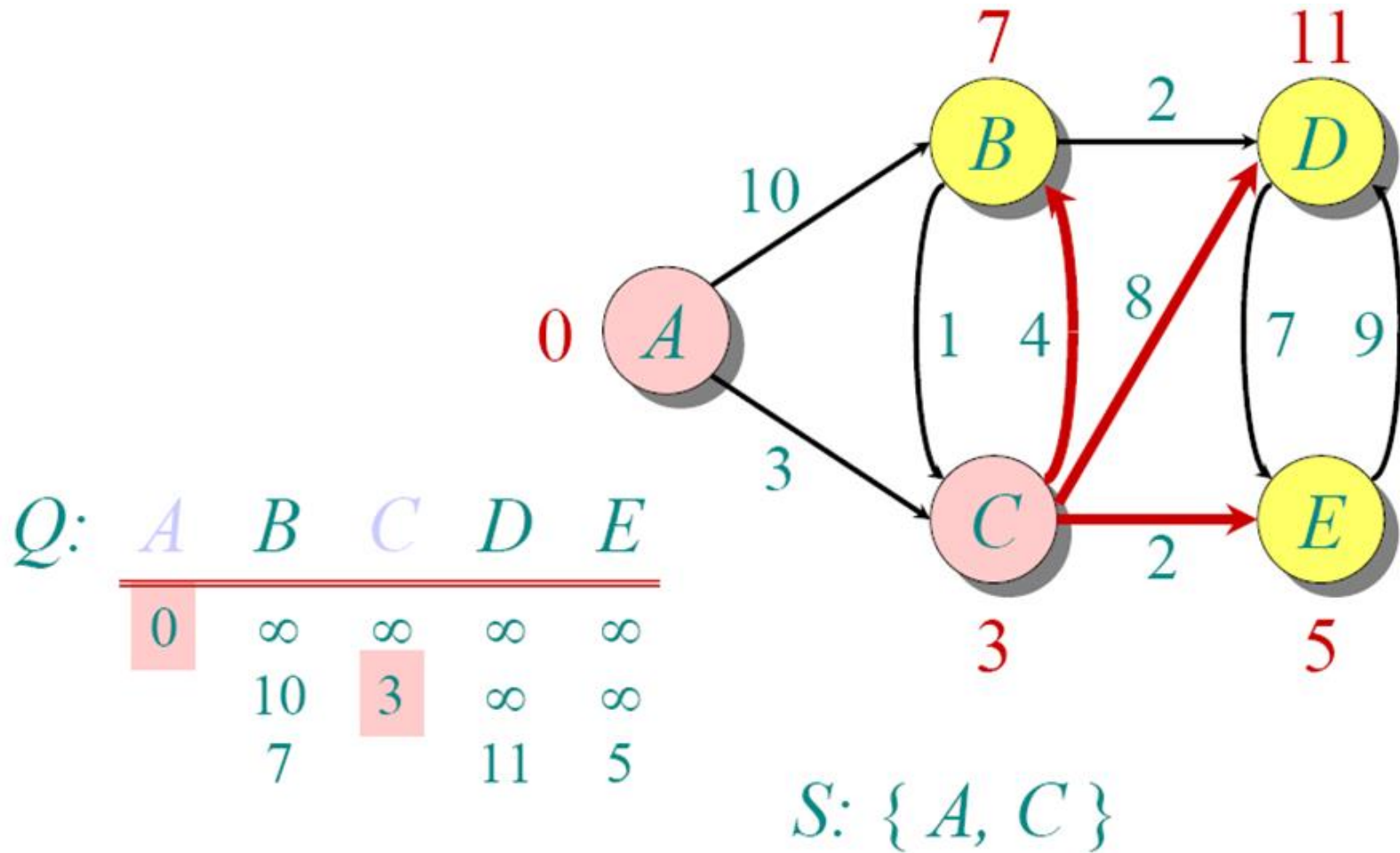


S: { A }

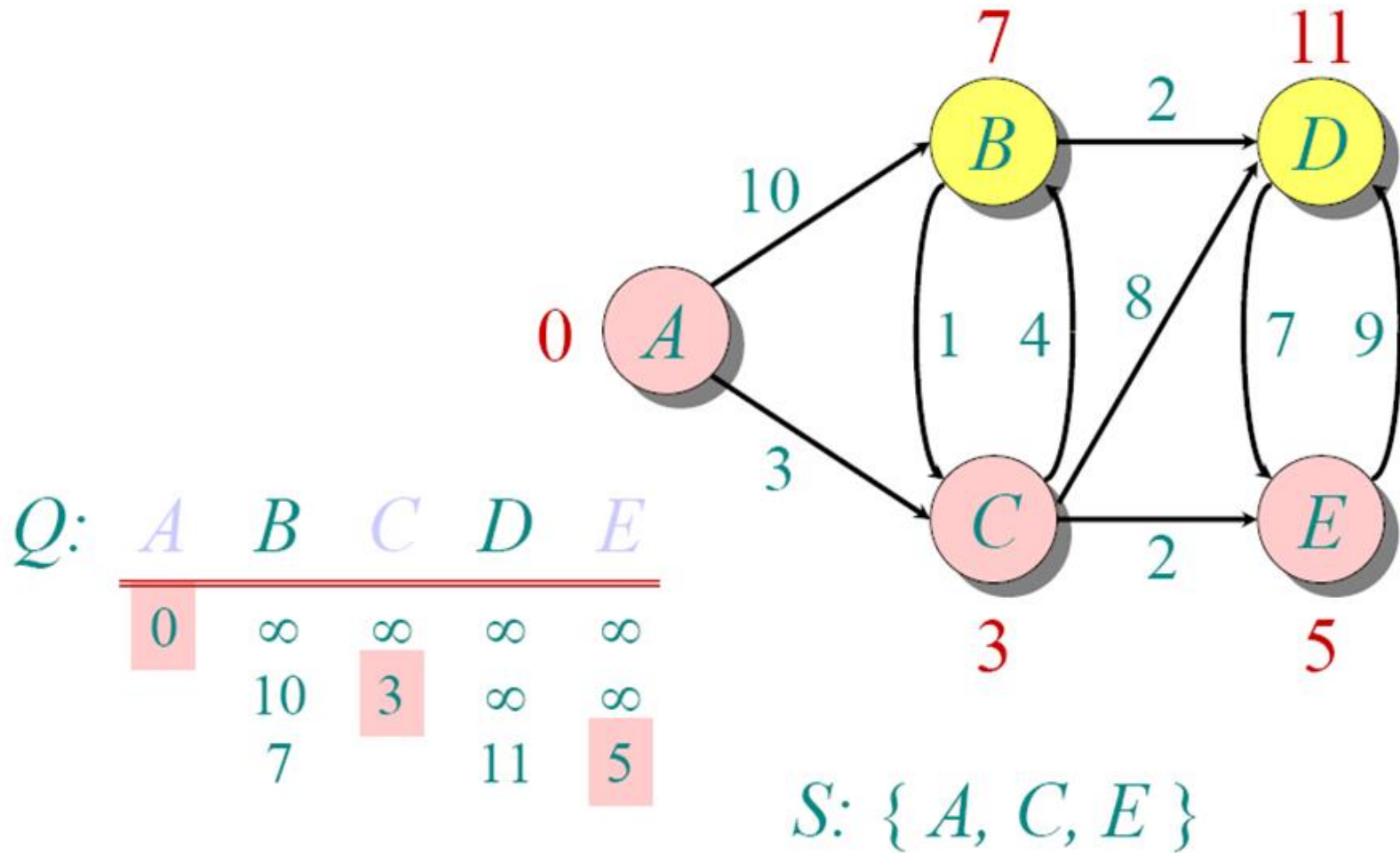
# Dijkstra Animated Example



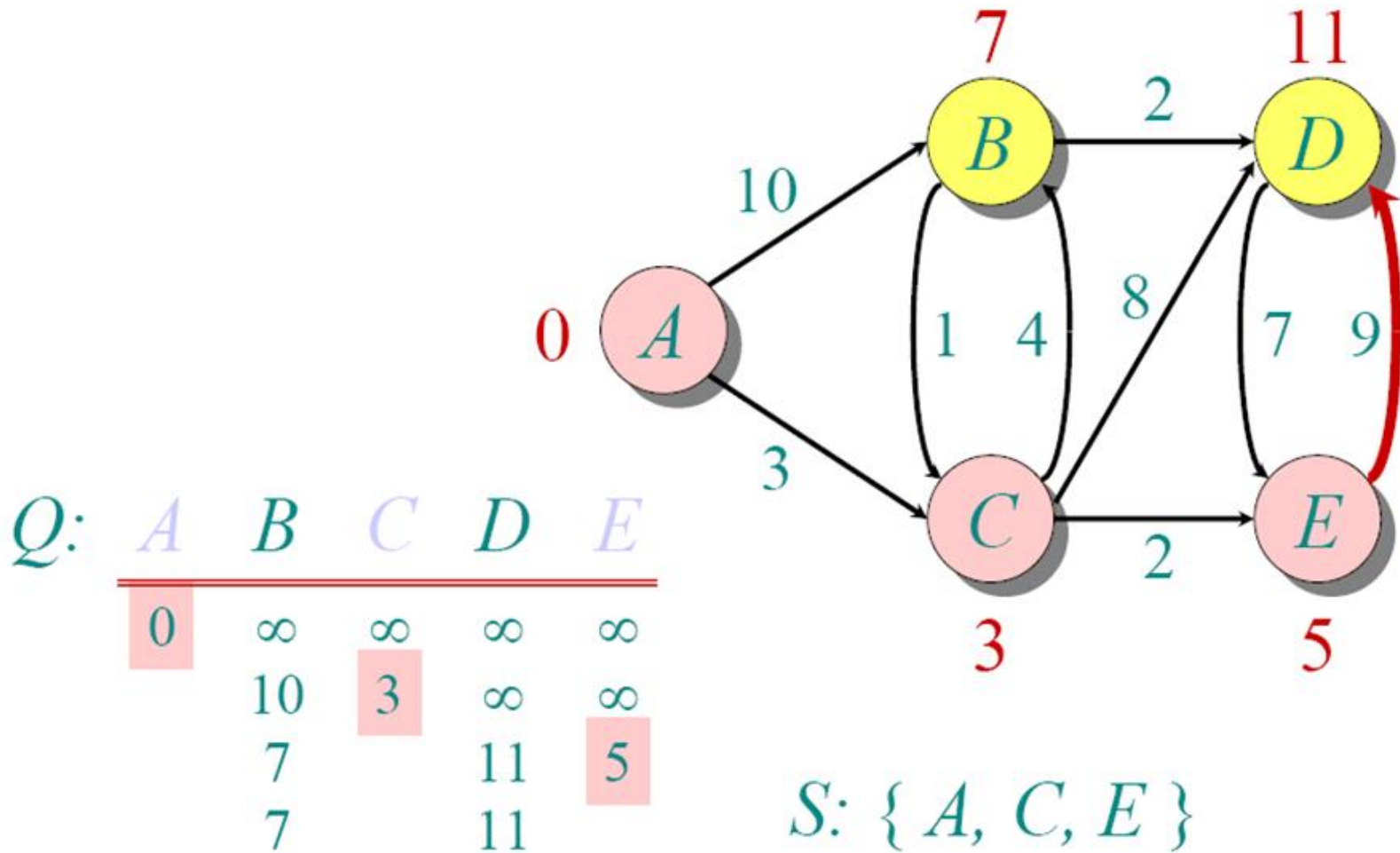
# Dijkstra Animated Example



# Dijkstra Animated Example

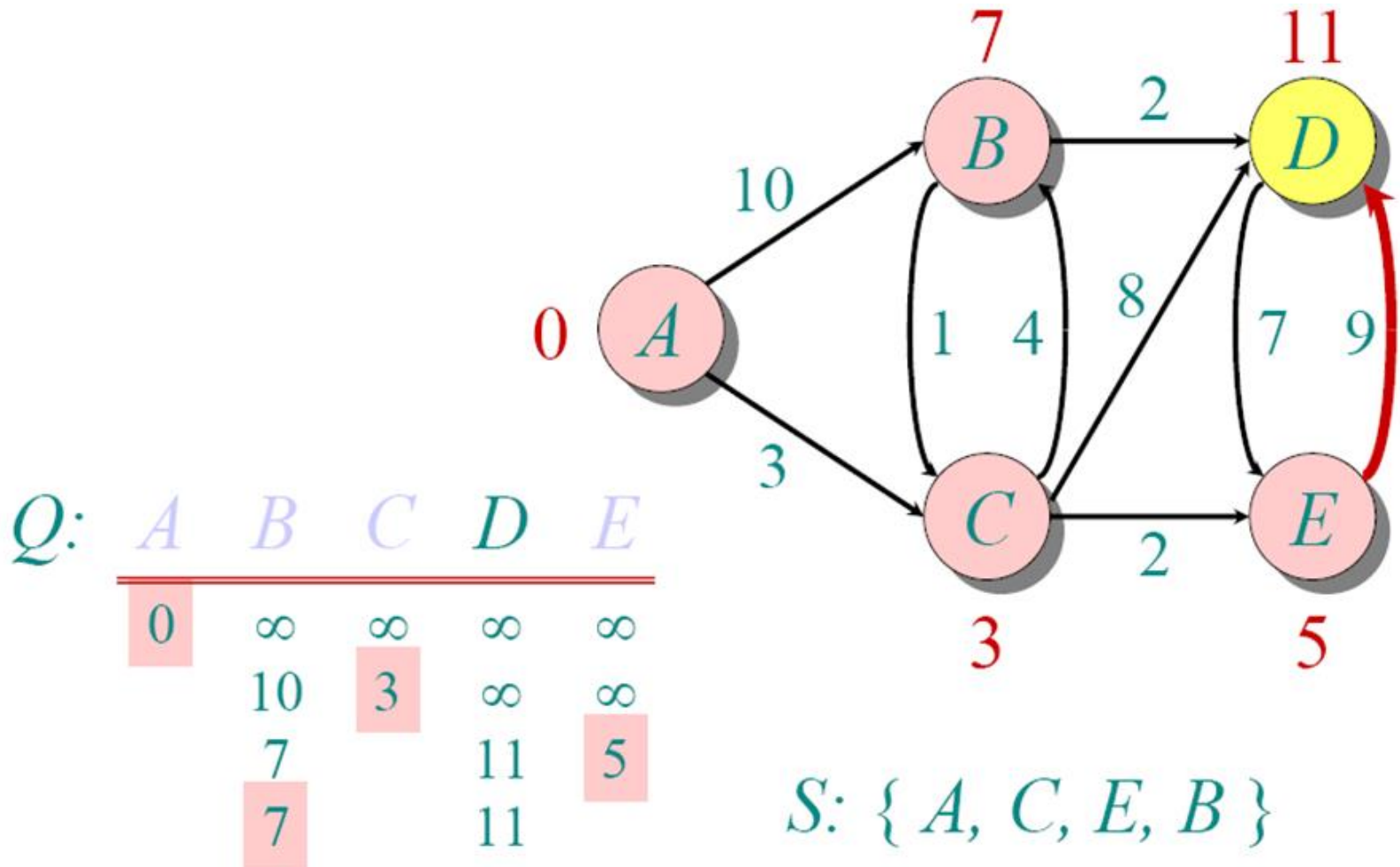


# Dijkstra Animated Example

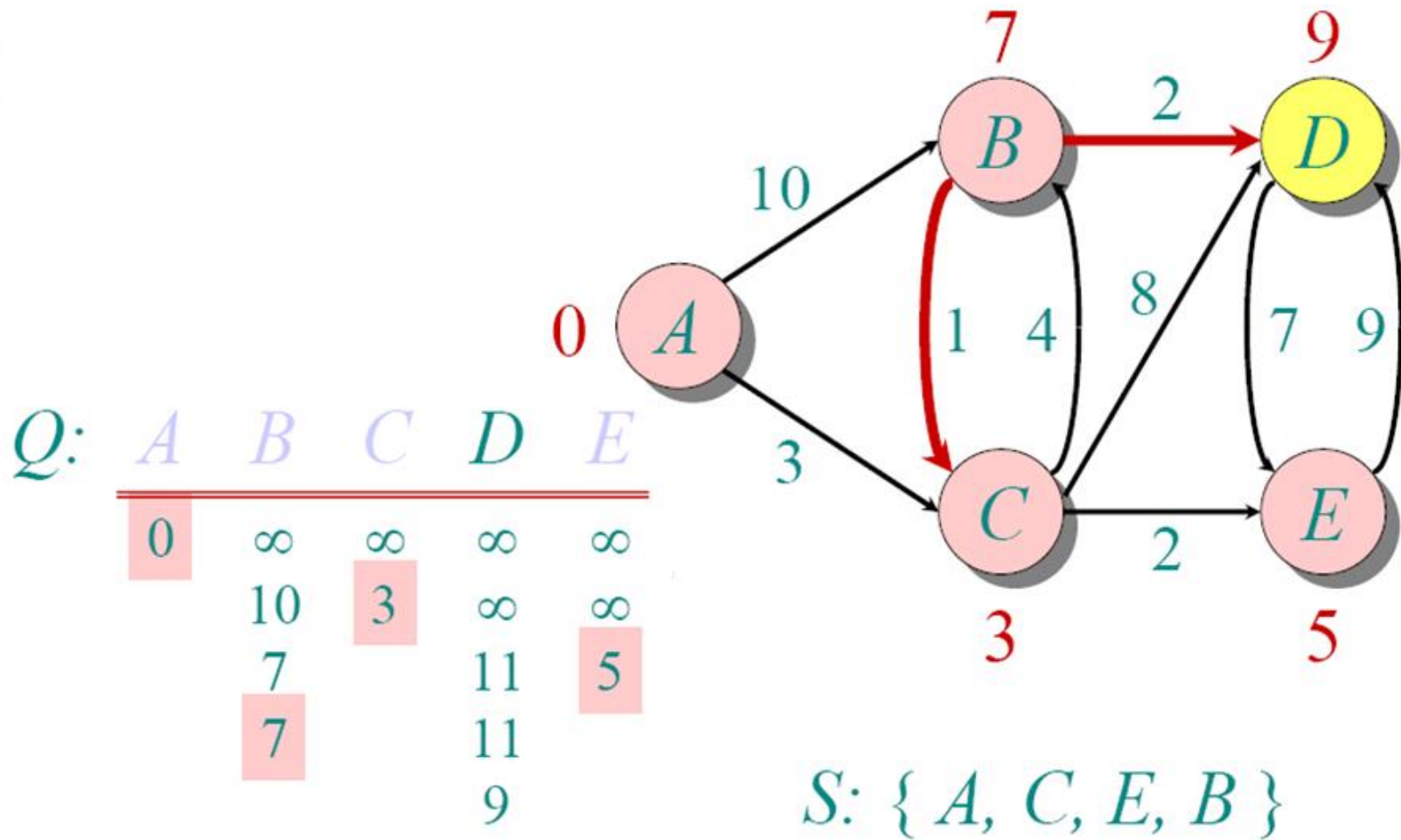




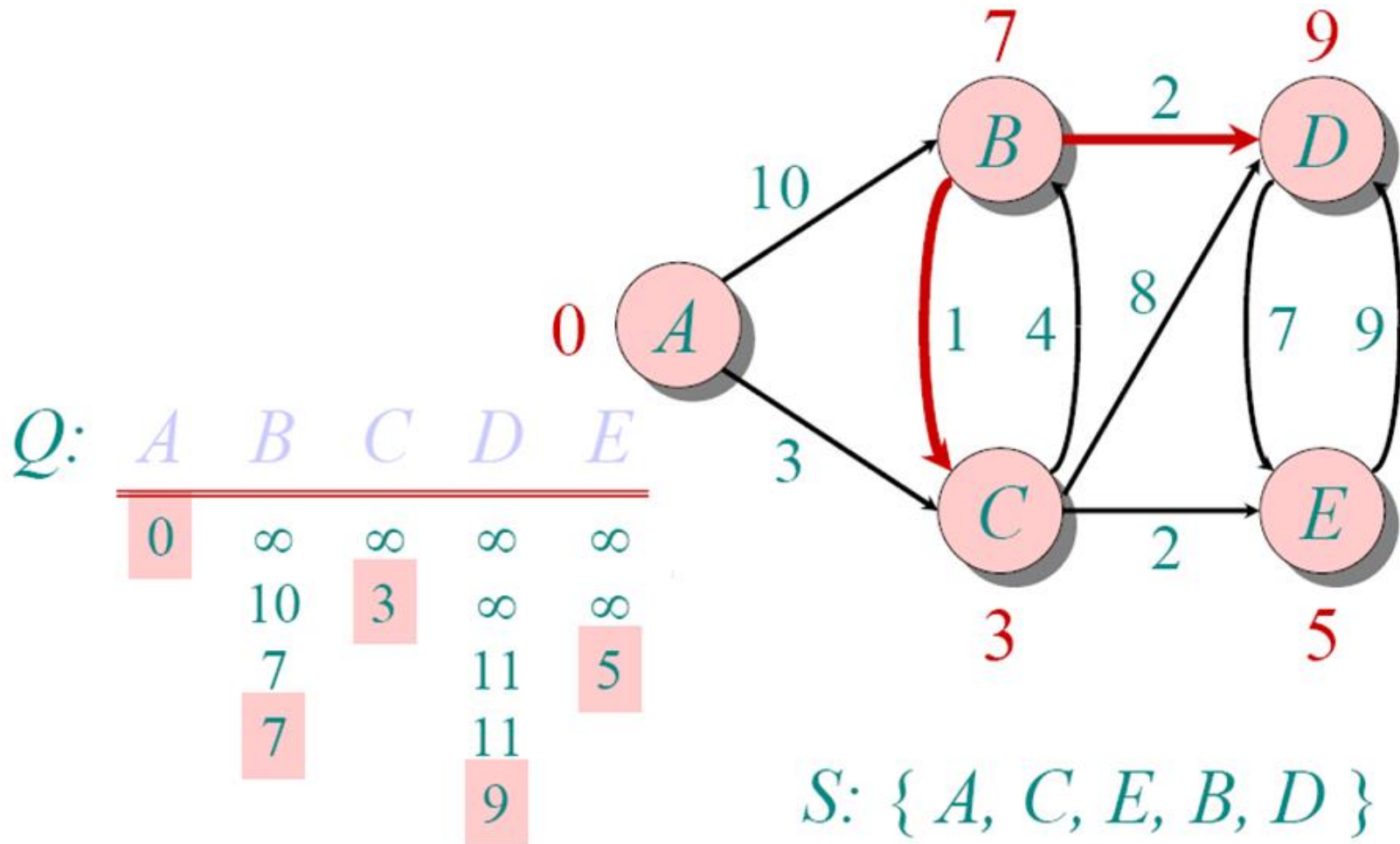
# Dijkstra Animated Example



# Dijkstra Animated Example



# Dijkstra Animated Example



# Dijkstra efficiency

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- ▶ The simplest implementation is:

$$O(E + V^2)$$

- ▶ But it can be implemented more efficiently:

$$O(E + V \cdot \log V)$$



Floyd–Warshall:  $O(V^3)$   
Bellman-Ford-Moore :  $O(V \cdot E)$

# Implementation

---

OVERVIEW PACKAGE **CLASS** USE TREE DEPRECATED INDEX HELP

PREV CLASS NEXT CLASS FRAMES NO FRAMES

SUMMARY: NESTED | FIELD | CONSTR | METHOD DETAIL: FIELD | CONSTR | METHOD

org.jgrapht.alg.shortestpath

## Class `DijkstraShortestPath<V,E>`

java.lang.Object  
org.jgrapht.alg.shortestpath.DijkstraShortestPath<V,E>

### Type Parameters:

V - the graph vertex type

E - the graph edge type

### All Implemented Interfaces:

ShortestPathAlgorithm<V,E>

---

```
public final class DijkstraShortestPath<V,E>  
    extends Object
```

An implementation of Dijkstra's shortest path algorithm using a Fibonacci heap.

### Author:

John V. Sichi

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# Shortest Paths wrap-up

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Algorithm	Problem	Efficiency	Limitation
Floyd-Warshall	AP	$O(V^3)$	No negative cycles
Bellman-Ford	SS	$O(V \cdot E)$	No negative cycles
Repeated Bellman-Ford	AP	$O(V^2 \cdot E)$	No negative cycles
Dijkstra	SS	$O(E + V \cdot \log V)$	No negative edges
Repeated Dijkstra	AP	$O(V \cdot E + V^2 \cdot \log V)$	No negative edges
Breadth-First visit	SS	$O(V + E)$	Unweighted graph



# JGraphT



```
public class FloydWarshallShortestPaths<V,E>
public class BellmanFordShortestPath<V,E>
public class DijkstraShortestPath<V,E>
```

```
// APSP
List<GraphPath<V,E>>  getShortestPaths (V v)
GraphPath<V,E>      getShortestPath (V a, V b)

// SSSP
GraphPath<V,E>      getPath ()
```

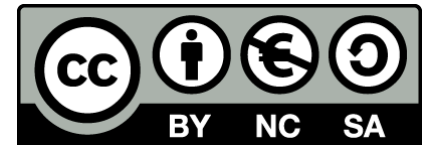
# Resources





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- ▶ Algorithms in a Nutshell, G. Heineman, G. Pollice, S. Selkow, O'Reilly, ISBN 978-0-596-51624-6, Chapter 6  
<http://shop.oreilly.com/product/9780596516246.do>
- ▶ [http://en.wikipedia.org/wiki/Floyd%E2%80%93Warshall\\_algorithm](http://en.wikipedia.org/wiki/Floyd%E2%80%93Warshall_algorithm)



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