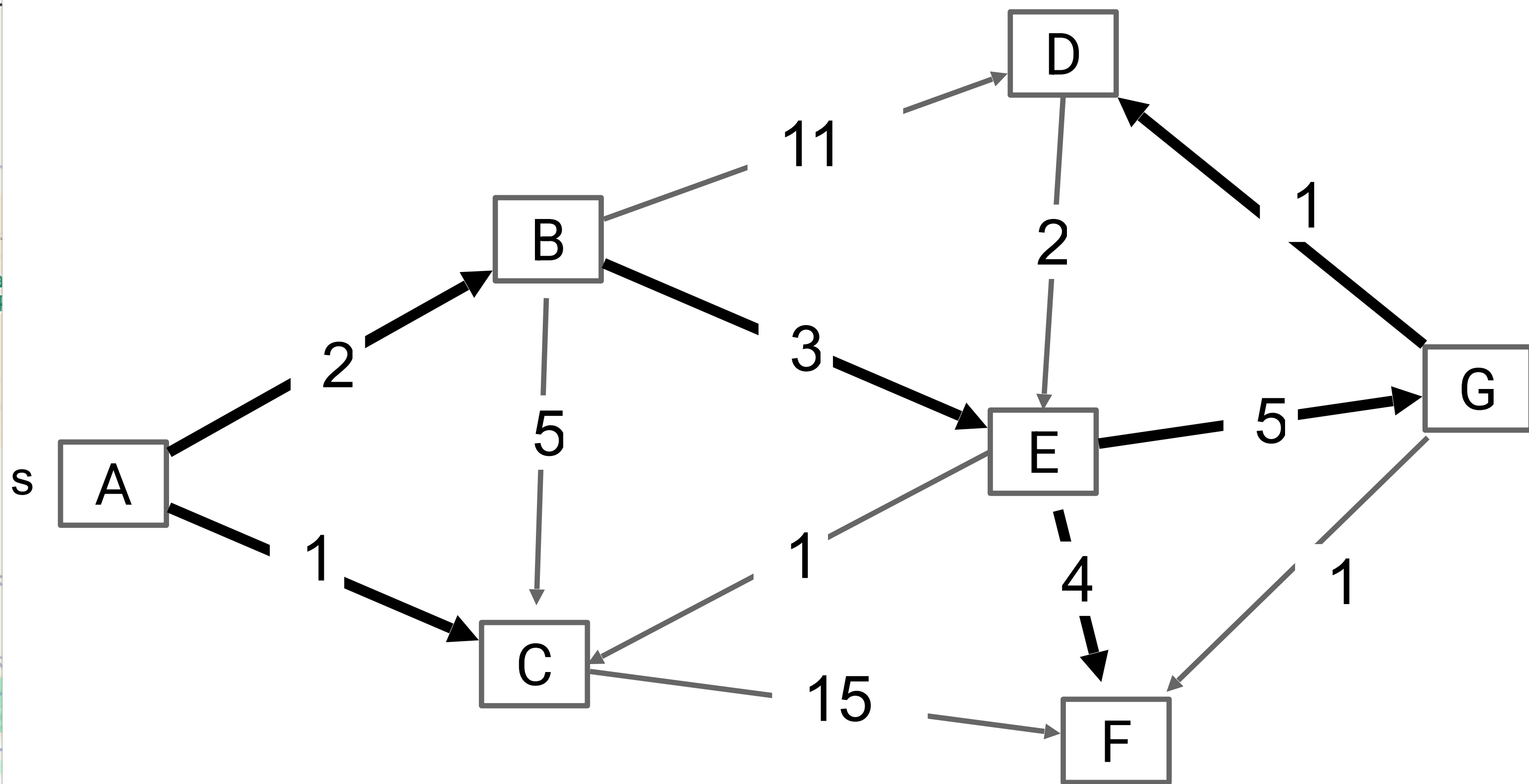
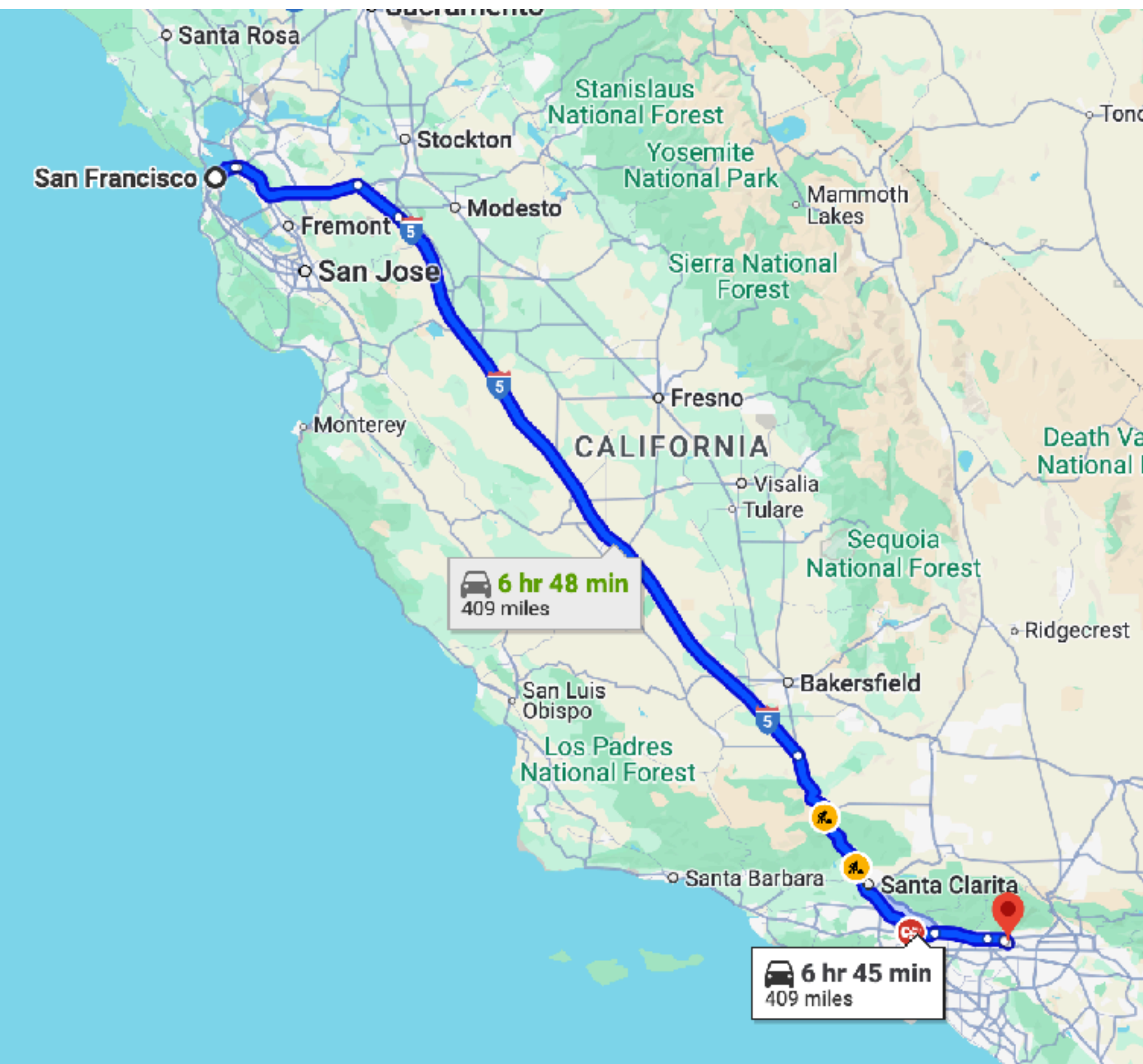


# CS62 Class 23: Shortest paths

Graphs



# Hashtable review

Insert the keys into both a separate chaining and open addressing (with linear probing) hashtable  $M = 5$ . Assume the hash function  $h(k) = k \% M$ . Assume no resizing.

1, 12, 22, 32, 42

- 1) What is the length of the longest chain?
- 2) What is the average number of probes per insertion?

# Hashtable review

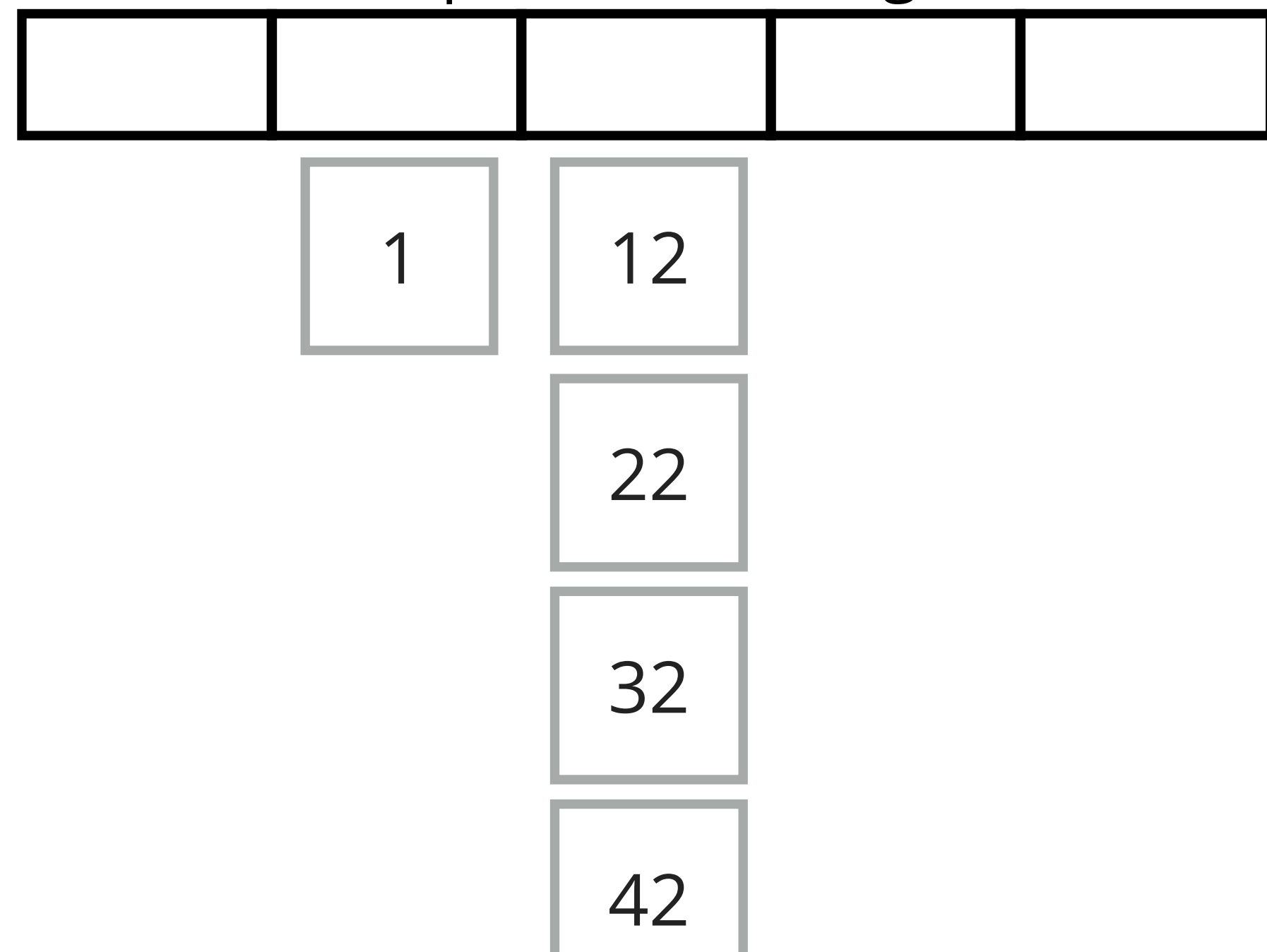
Insert the keys into both a separate chaining and open addressing (with linear probing) hashtable  $M = 5$ . Assume the hash function  $h(k) = k \% M$ . Assume no resizing.

1, 12, 22, 32, 42

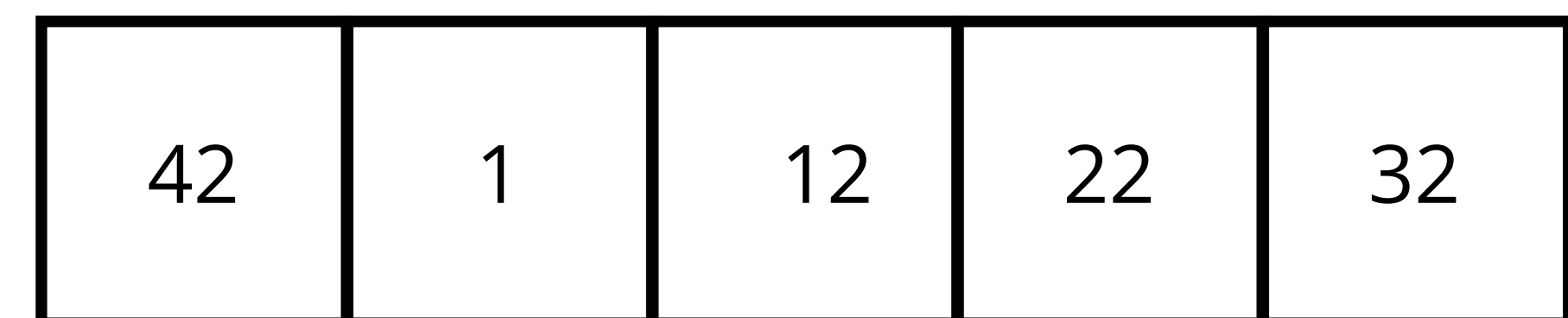
1) What is the length of the longest chain? - **4**

2) What is the average number of probes per insertion? - **2.2**

separate chaining



open addressing



Probes:  $1 (1) + 1 (12) + 2 (32) + 3 (42) + 4 (42) = (1+1+2+3+4)/5 = 2.2$

# Graph Problems

Last time, saw two ways to find paths in a graph, DFS & BFS.

Give an example of a graph that would make the space efficiency bad for DFS or BFS.

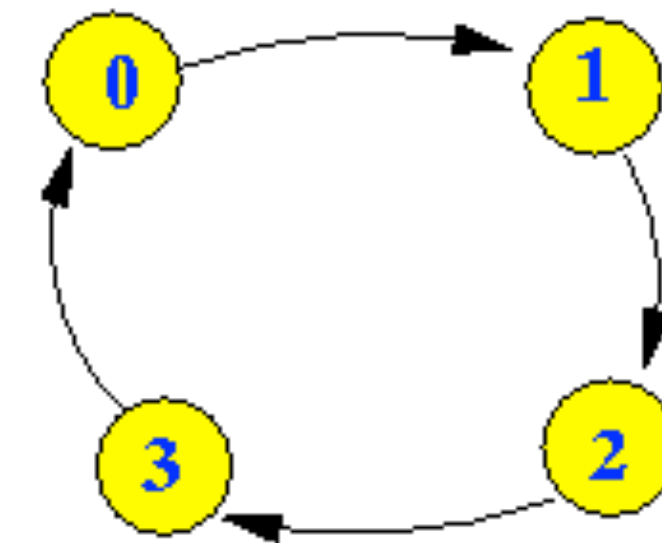
Problem	Problem Description	Solution	Efficiency (adj. list)
s-t paths	Find a path from s to every reachable vertex.	DFS	$O(V+E)$ time $\Theta(V)$ space
s-t shortest paths	Find a shortest path from s to every reachable vertex.	BFS	$O(V+E)$ time $\Theta(V)$ space

# BFS vs. DFS for space efficiency

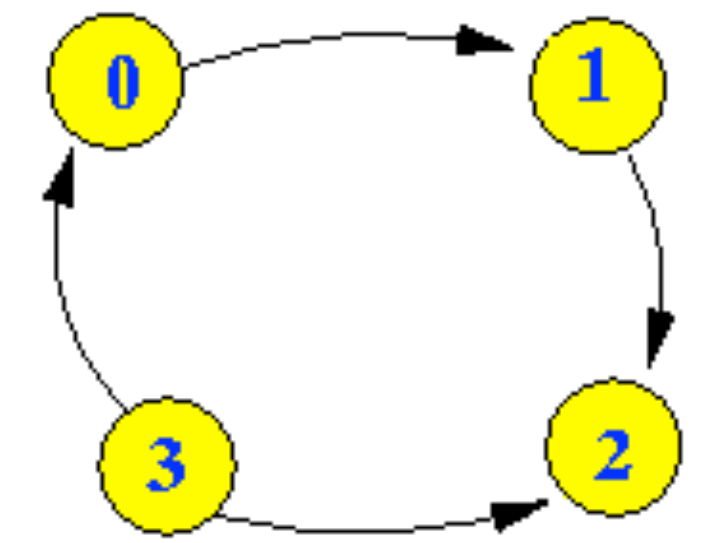
- DFS is worse for spindly graphs.
  - Call stack gets very deep.
  - Computer needs  $\Theta(V)$  memory to remember recursive calls.
- BFS is worse for absurdly “bushy” graphs.
  - Queue gets very large. In worst case, queue will require  $\Theta(V)$  memory.
  - Example: 1,000,000 vertices that are all connected. 999,999 will be enqueued at once.
- Note: In our implementations, we have to spend  $\Theta(V)$  memory anyway to track `distTo` and `edgeTo` arrays.
  - Can optimize by storing `distTo` and `edgeTo` in a map instead of an array.

# Strongly connected digraph algorithm

- A **strongly connected digraph** is a directed graph in which it is possible to reach any vertex starting from any other vertex by traversing edges.
- Pick a random starting vertex  $s$ .
- Run DFS/BFS starting at  $s$ .
  - If have not reached all vertices, return false.
- Reverse edges.
- Run DFS/BFS again on reversed graph.
  - If have not reached all vertices, return false.
  - Else return true.



*Strongly connected*



*Not strongly connected*

# Agenda

- Edge-weighted graphs
- Shortest paths
- Dijkstra's algorithm

**Edge-weighted graph**



# Edge-weighted graphs

- **Edge-weighted digraph**: a digraph where we associate weights/costs with each edge.
- **Shortest path from vertex  $s$  to vertex  $t$** : a directed path from  $s$  to  $t$  with the property that no other such path has a lower weight (total weight sum of edges it consists of).
- **Assumptions**:
  - Not all vertices need to be reachable.
  - Weights are positive.
  - Shortest paths are not necessarily unique but they are simple.

## edge-weighted digraph

4→5	0.35
5→4	0.35
4→7	0.37
5→7	0.28
7→5	0.28
5→1	0.32
0→4	0.38
0→2	0.26
7→3	0.39
1→3	0.29
2→7	0.34
6→2	0.40
3→6	0.52
6→0	0.58
6→4	0.93



## shortest path from 0 to 6

0→2	0.26
2→7	0.34
7→3	0.39
3→6	0.52

# Weighted directed edge API

- `public class DirectedEdge` only difference is we now have weights
  - `DirectedEdge(int v, int w, double weight)`
    - Constructs a weighted edge from v to w ( $v \rightarrow w$ ) with the provided weight.
  - `int from()`
    - Returns vertex source of this edge.
  - `int to()`
    - Returns vertex destination of this edge.
  - `double weight()`
    - Returns weight of this edge.
  - `String toString()`
    - Returns the string representation of this edge.

# Weighted directed edge in Java

```
public class DirectedEdge {
    private final int v;
    private final int w;
    private final double weight;

    public DirectedEdge(int v, int w, double weight) {
        this.v = v;
        this.w = w;
        this.weight = weight;
    }

    public int from() {
        return v;
    }

    public int to() {
        return w;
    }

    public double weight() {
        return weight;
    }
}
```

# Edge-weighted digraph API

- `public class EdgeWeightedDigraph`
  - `EdgeWeightedDigraph(int v)`
    - Constructs an edge-weighted digraph with  $v$  vertices.
  - `void addEdge(DirectedEdge e)`
    - Add weighted directed edge  $e$ .
  - `Iterable<DirectedEdge> adj(int v)`
    - Returns edges adjacent from  $v$ .
  - `int V()`
    - Returns number of vertices.
  - `int E()`
    - Returns number of edges.
  - `Iterable<DirectedEdge> edges()`
    - Returns all edges.

only difference is edges are `DirectedEdge` objects instead of integers

# Edge-weighted digraph adjacency list representation

- `public class EdgeWeightedDigraph`

- `EdgeWeightedDigraph(int v)`

- Constructs an edge-weighted digraph with  $V$  vertices.

- `void addEdge(DirectedEdge e)`

- Add weighted directed edge  $e$ .

- `Iterable<DirectedEdge> adj(int v)`

- Returns edges adjacent from  $v$ .

- `int V()`

- Returns number of vertices.

- `int E()`

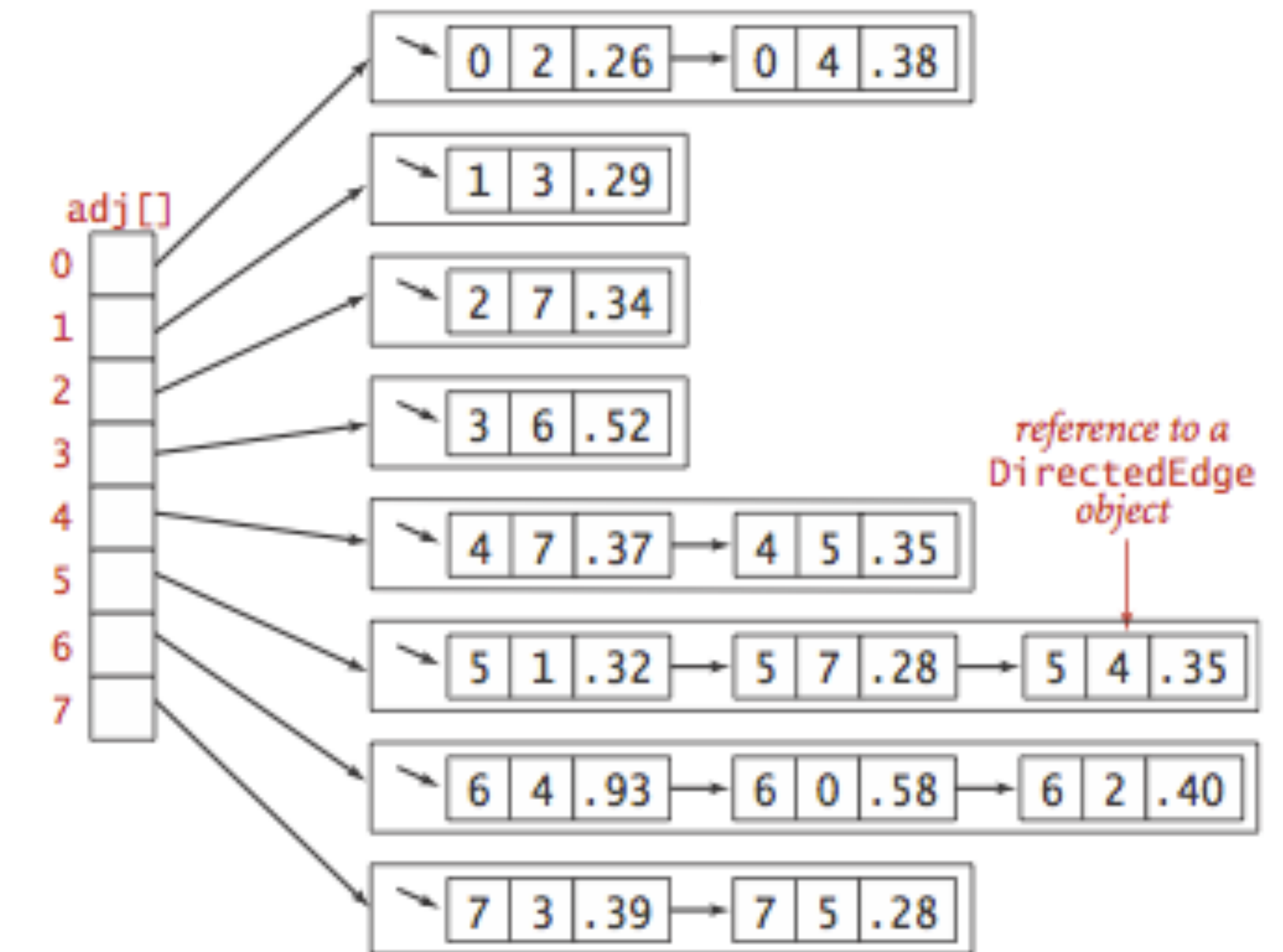
- Returns number of edges.

- `Iterable<DirectedEdge> edges()`

- Returns all edges.

*tinyEWD.txt*

$V \rightarrow 8$   
 $E \rightarrow 15$   
4 5 0.35  
5 4 0.35  
4 7 0.37  
5 7 0.28  
7 5 0.28  
5 1 0.32  
0 4 0.38  
0 2 0.26  
7 3 0.39  
1 3 0.29  
2 7 0.34  
6 2 0.40  
3 6 0.52  
6 0 0.58  
6 4 0.93



Edge-weighted digraph representation

# Edge-weighted digraph in Java

```
public class EdgeWeightedDigraph {
    private final int V;           // number of vertices in this digraph
    private int E;                 // number of edges in this digraph
    private SinglyLinkedList<DirectedEdge> adj[];
    // adj[v] = adjacency list for v

    public EdgeWeightedDigraph(int V) {
        this.V = V;
        this.E = 0;
        adj = new SinglyLinkedList<DirectedEdge>[V];
        for (int v = 0; v < V; v++)
            adj[v] = new SinglyLinkedList<DirectedEdge>();
    }

    public void addEdge(DirectedEdge e) {
        int v = e.from();
        int w = e.to();
        adj[v].add(e);           extract v & w with .from() and .to() getters
        E++;
    }

    public Iterable<DirectedEdge> adj(int v) {
        return adj[v];
    }
}
```

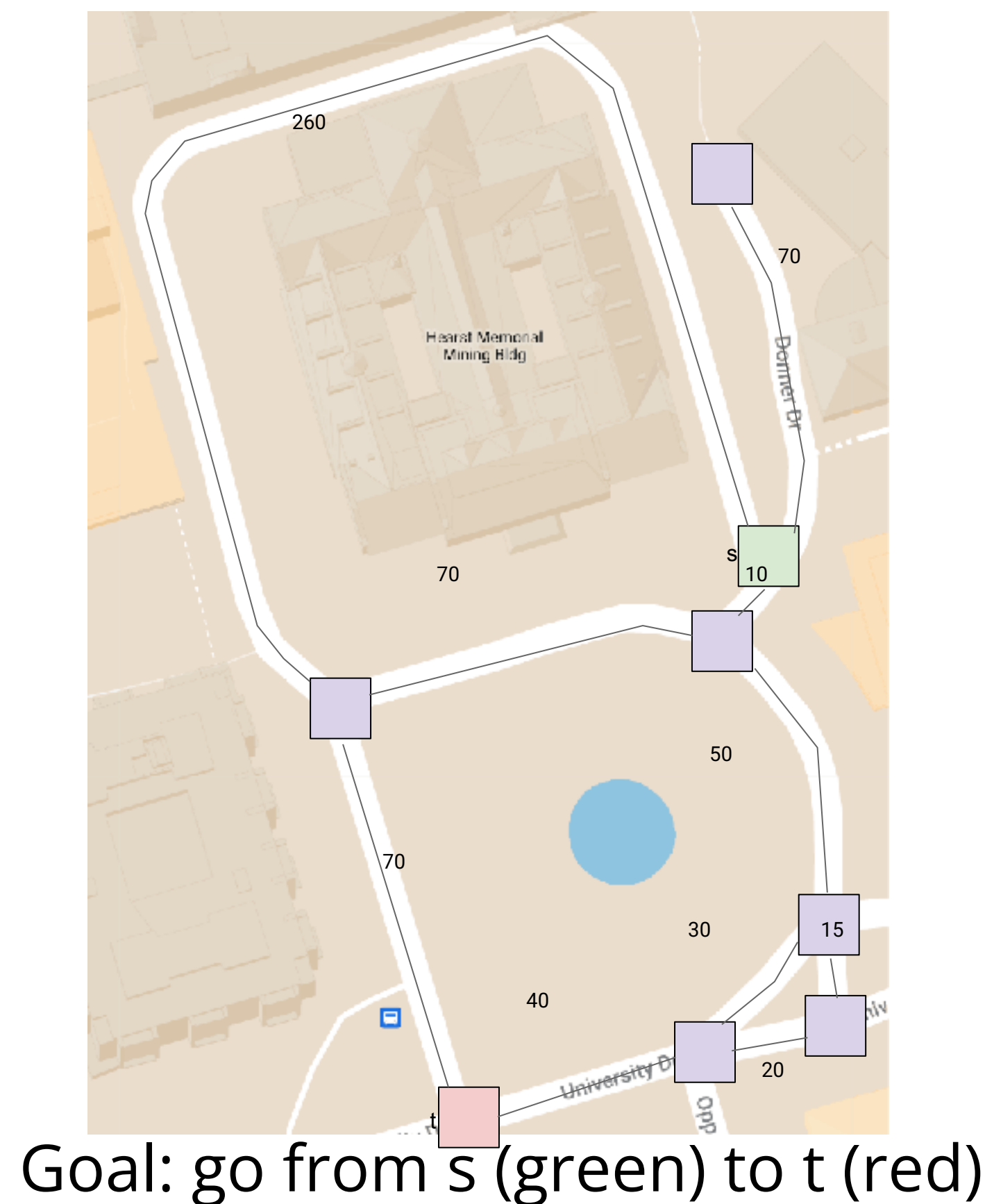
DirectedEdge instead of int

# Shortest paths

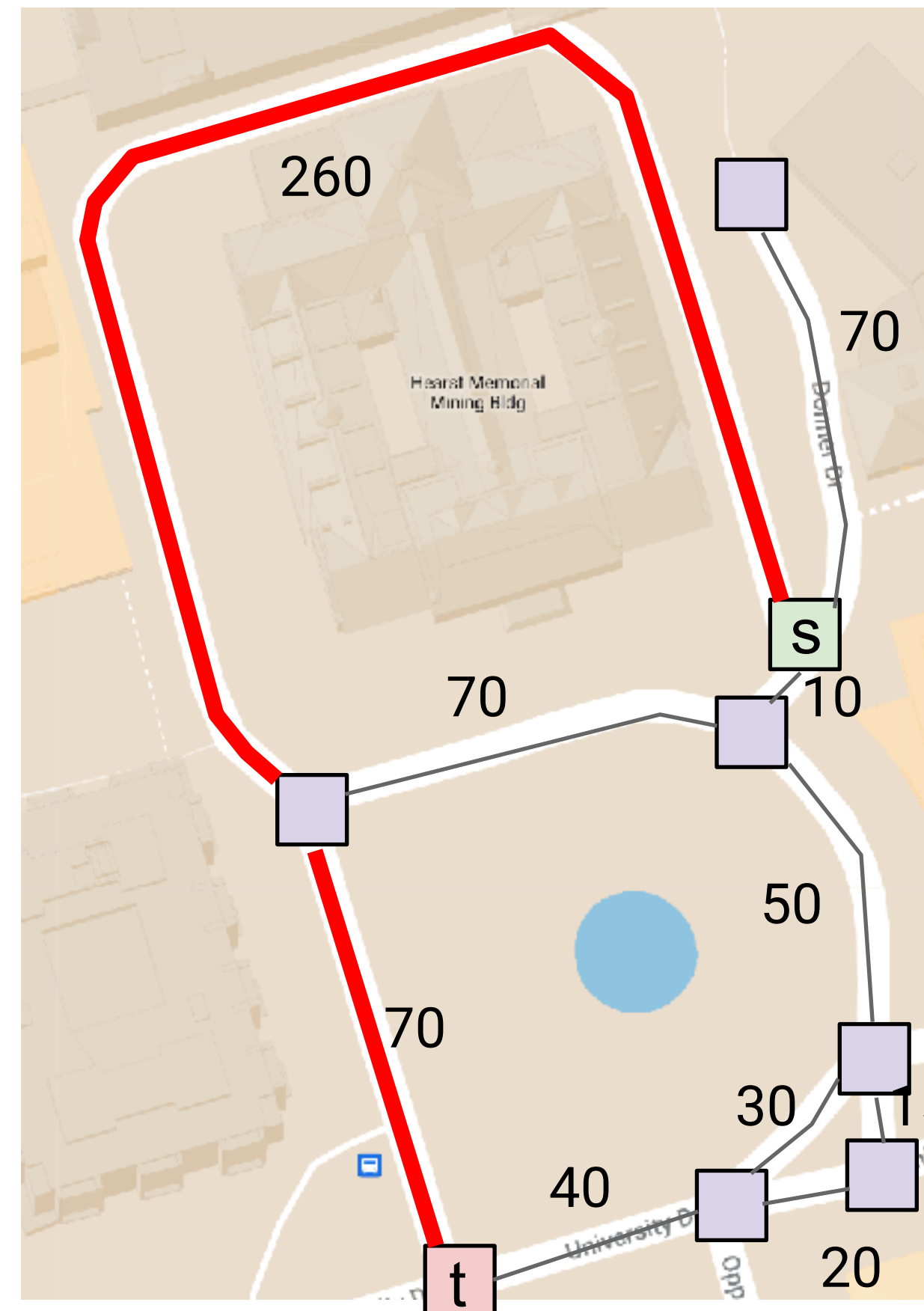
# BreadthFirstSearch for Google Maps

BFS would not be a good choice for a google maps style navigation application.

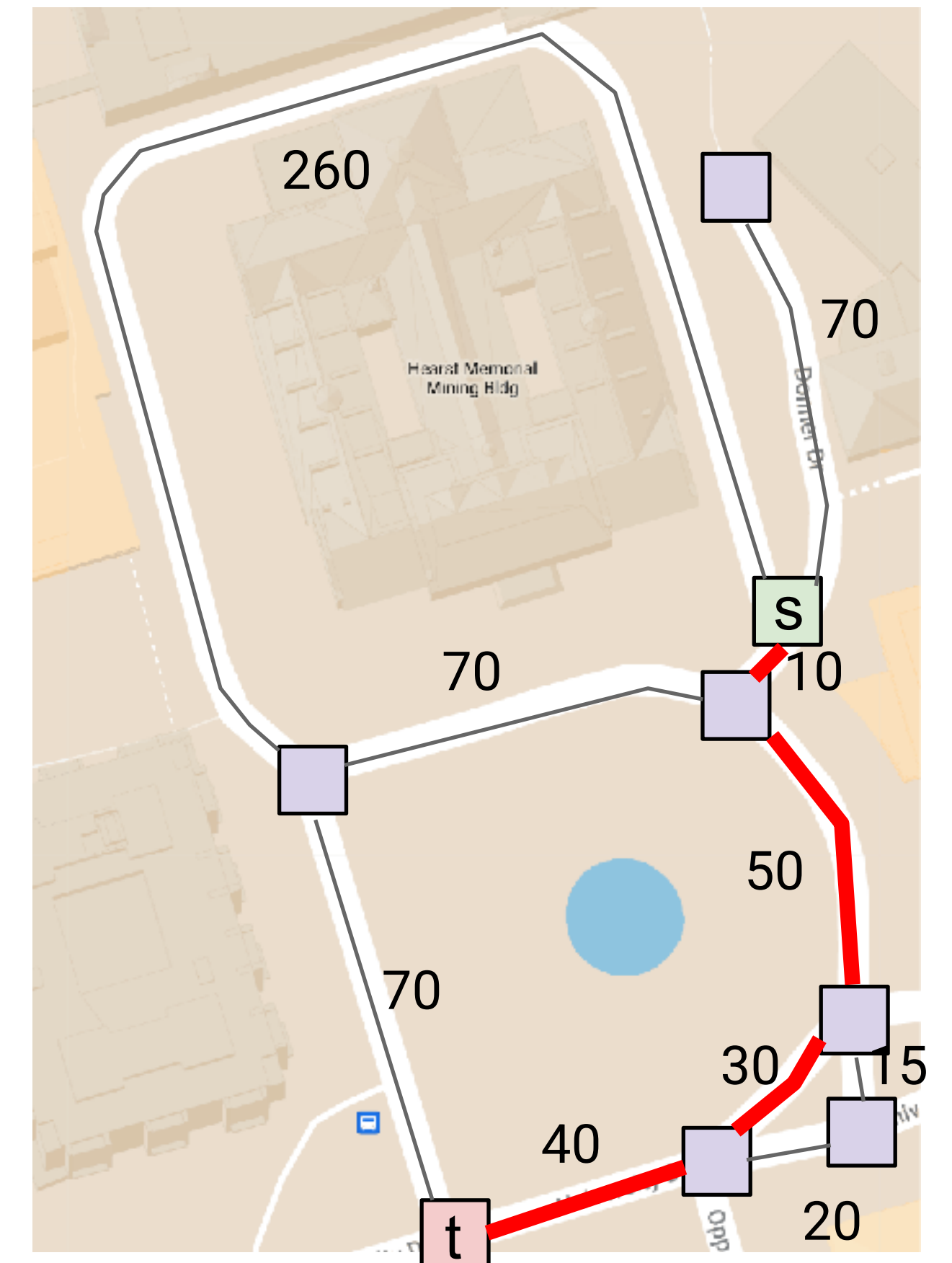
- The problem: BFS returns path with shortest number of edges, not necessarily the shortest path.
- That's why we need an edge-weighted graph.



BFS answer



Correct shortest path





# Shortest Path variants

- **Single source**: from one vertex  $s$  to every other vertex.
- **Single sink**: from every vertex to one vertex  $t$ .
- **Source-sink**: from one vertex  $s$  to another vertex  $t$ .
- **All pairs**: from every vertex to every other vertex.
- What version is there in Google Maps?

# Shortest Paths Assumptions

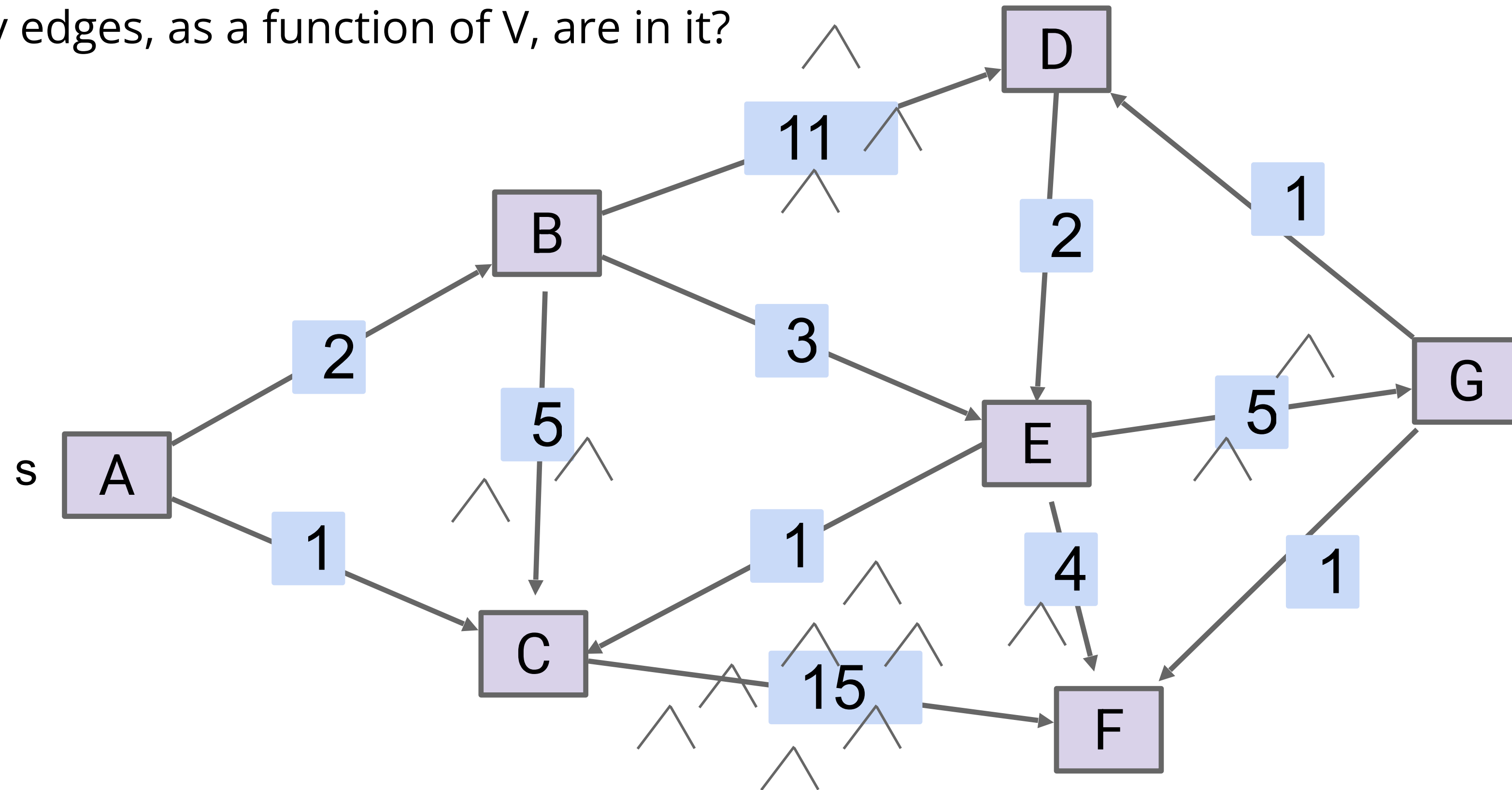
- Not all vertices need to be reachable.
  - We will assume so in this lecture.
- Weights are non-negative.
  - There are algorithms that can handle negative weights.
- Shortest paths are not necessarily unique but they are simple.

# Worksheet time!

Find the shortest paths from source vertex  $s$  to every other vertex. (Single source shortest path)

What data structure does your path look like?

How many edges, as a function of  $V$ , are in it?



What algorithm did you as a human come up with?

# Worksheet answers

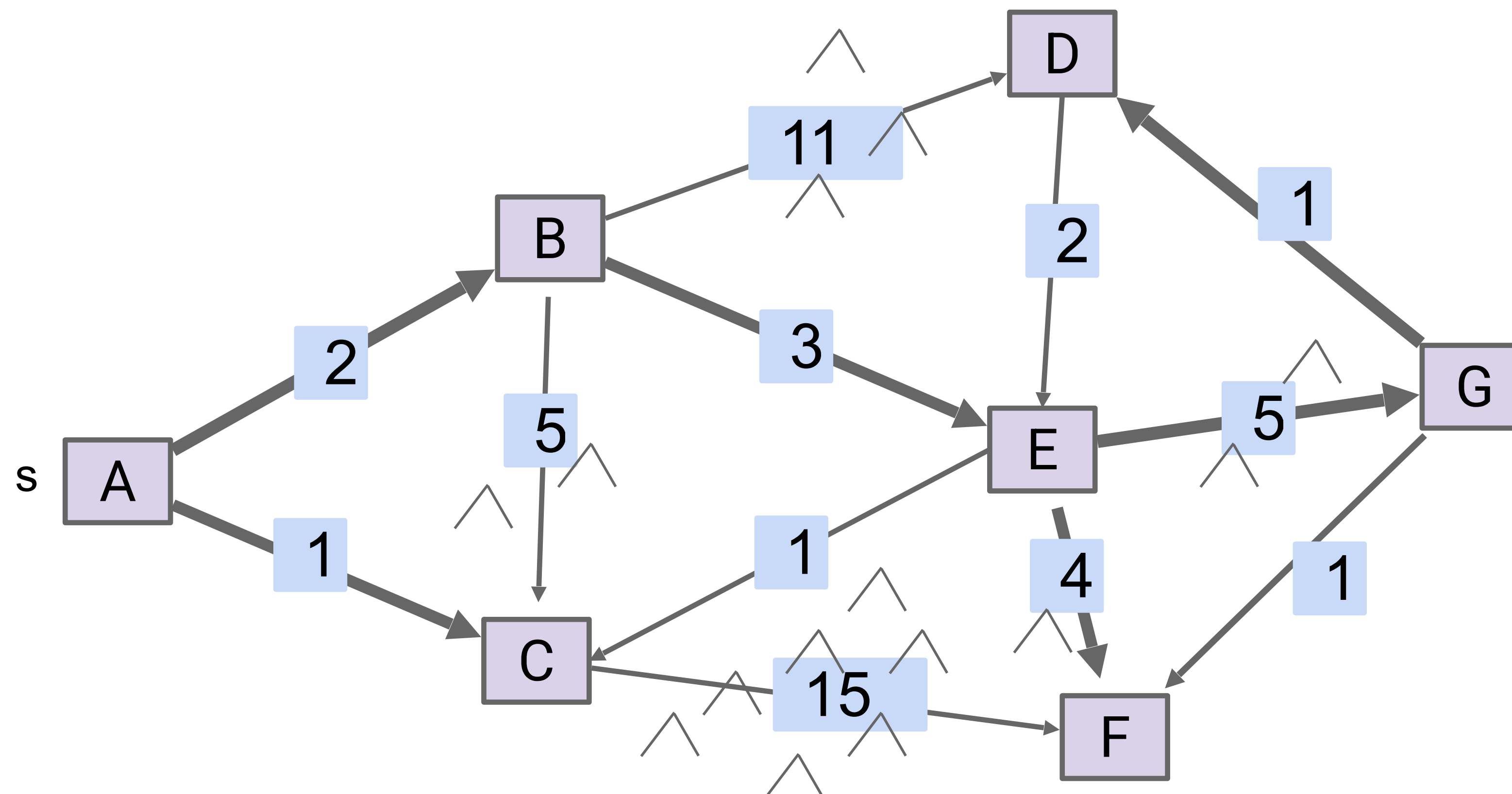
Find the shortest paths from source vertex  $s$  to every other vertex. (Single source shortest path)

What data structure does your path look like?

A tree

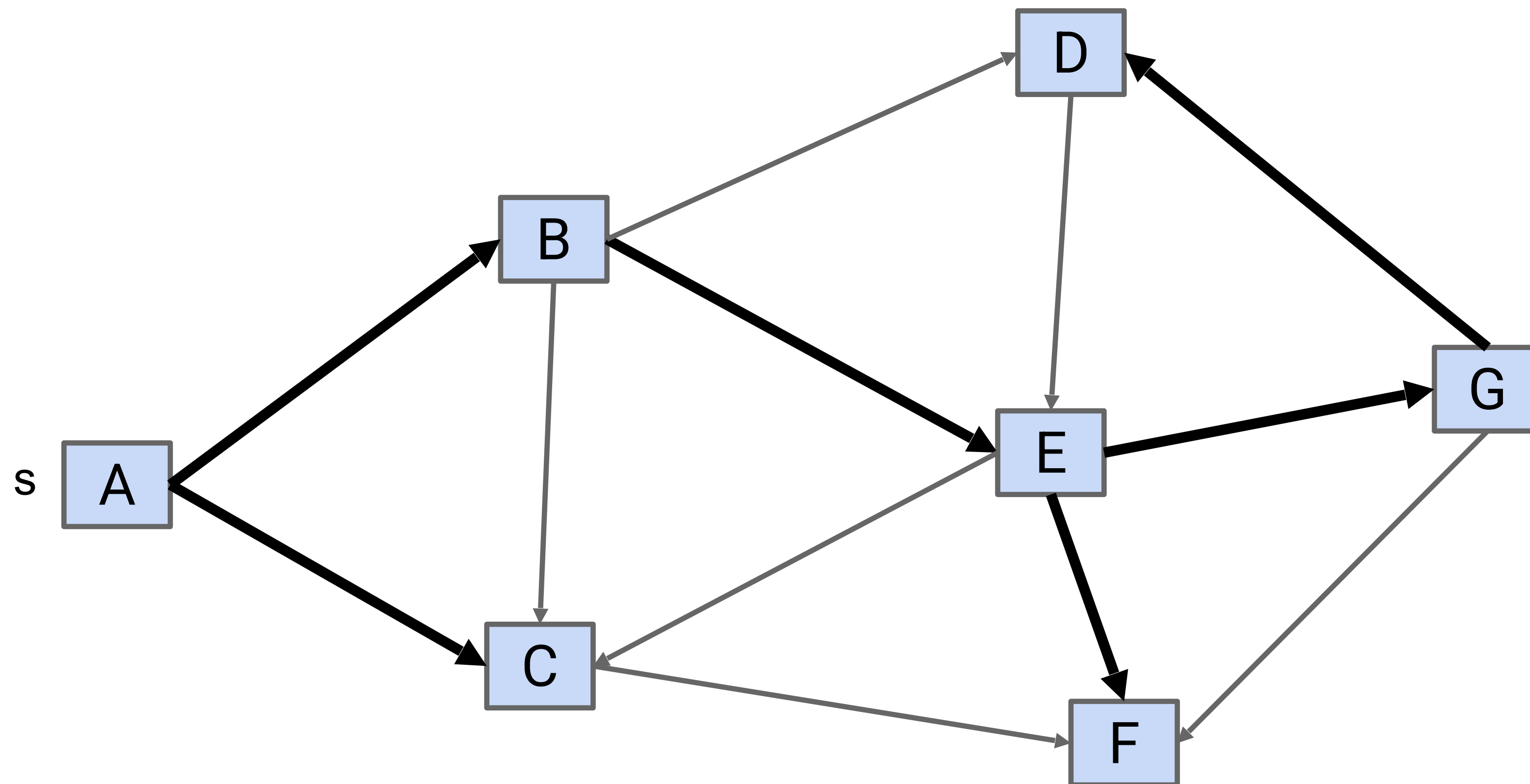
How many edges, as a function of  $V$ , are in it?

$E = V - 1$  (7 vertices, 6 edges)



# SPT Edge Count

If  $G$  is a connected edge-weighted graph with  $V$  vertices and  $E$  edges, there are exactly  $V-1$  edges in the **Shortest Paths Tree** (SPT) of  $G$ , assuming every vertex is reachable.



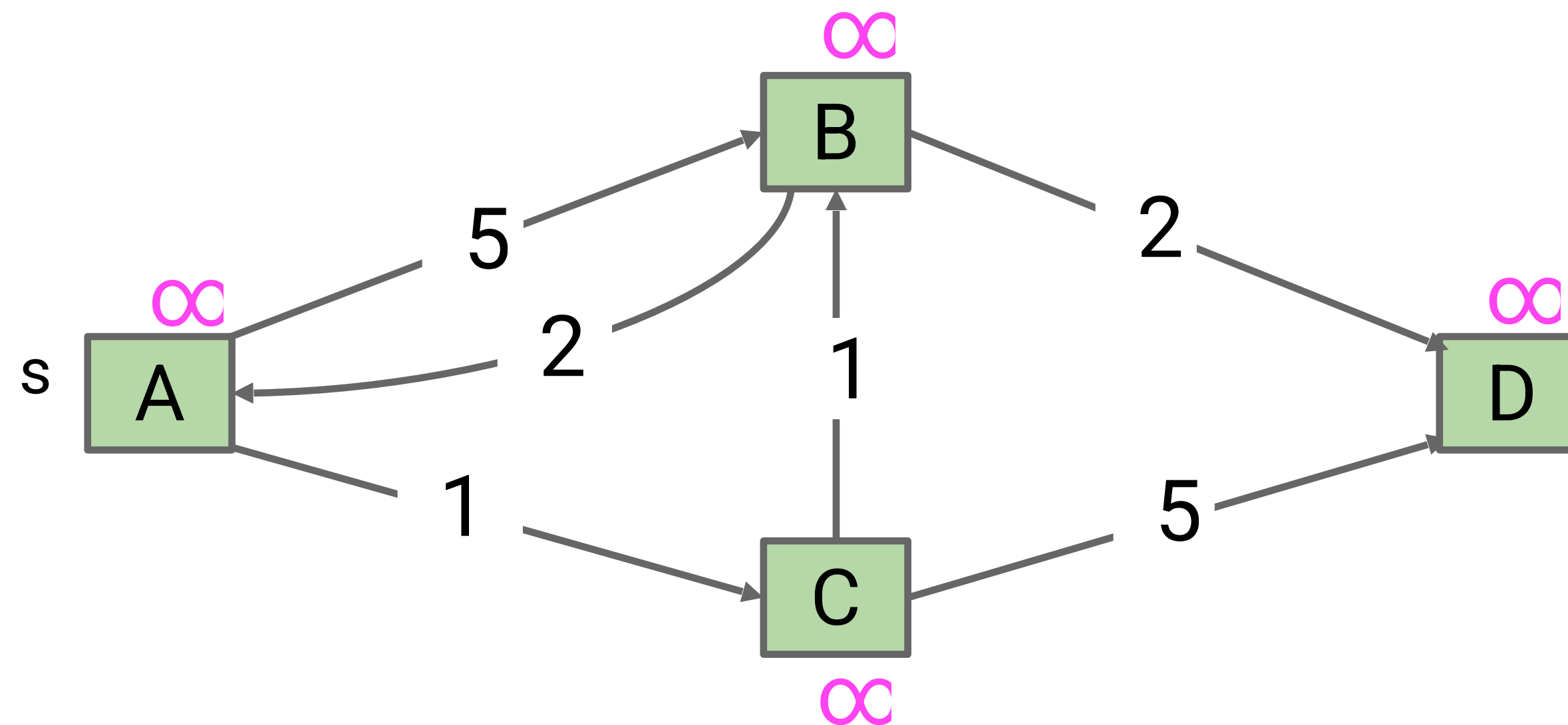
# Dijkstra's Algorithm (bad examples)

# Creating an Algorithm

Let's create an algorithm for finding the shortest paths.

Will start with a bad algorithm and then successively improve it.

- Algorithm begins in state below. All vertices unmarked. All distances infinite. No edges in the SPT.



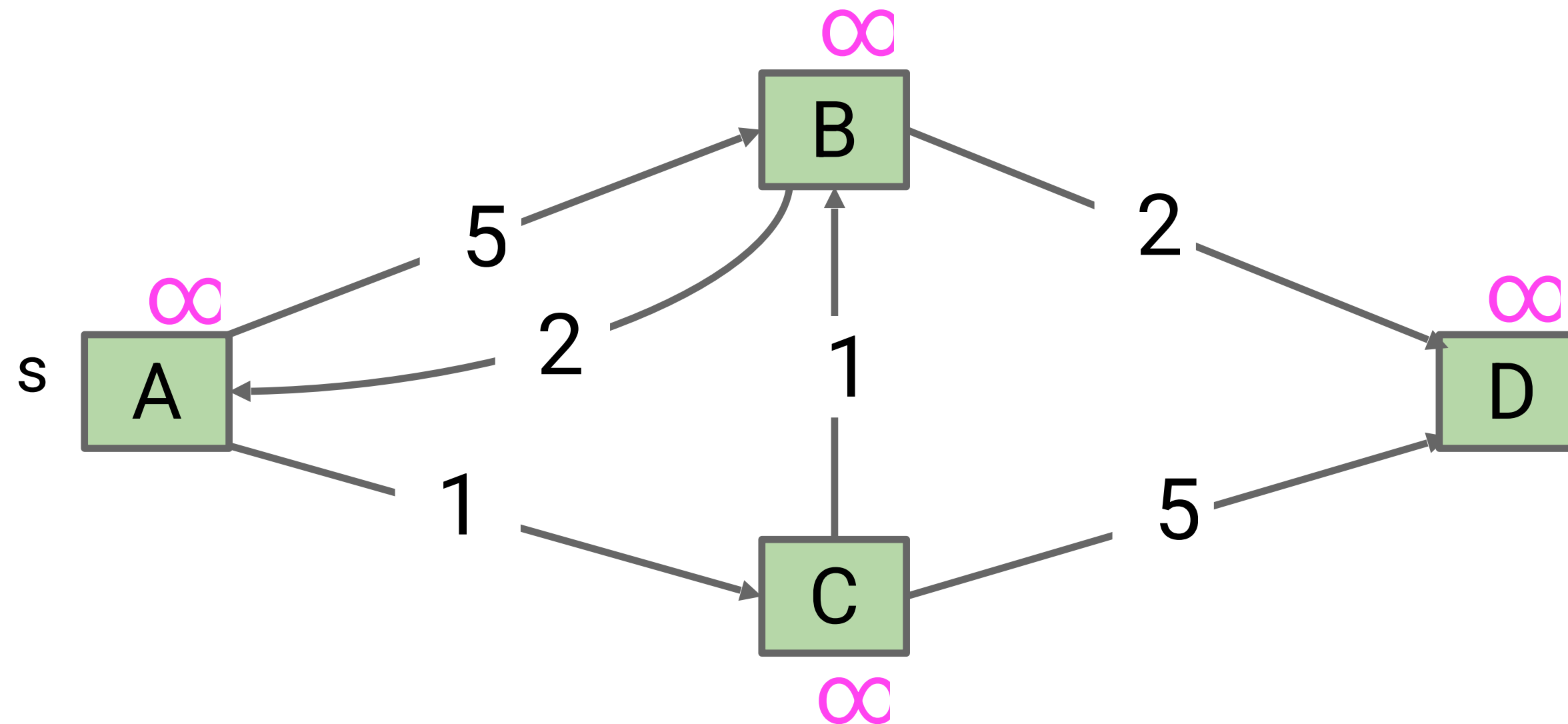
# Bad Algorithm #1 (Inspired by BFS)

Add the start (A) to the fringe.

While fringe is not empty:

Remove a vertex from the fringe and mark it.

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.





# Bad Algorithm #1 (Inspired by BFS)

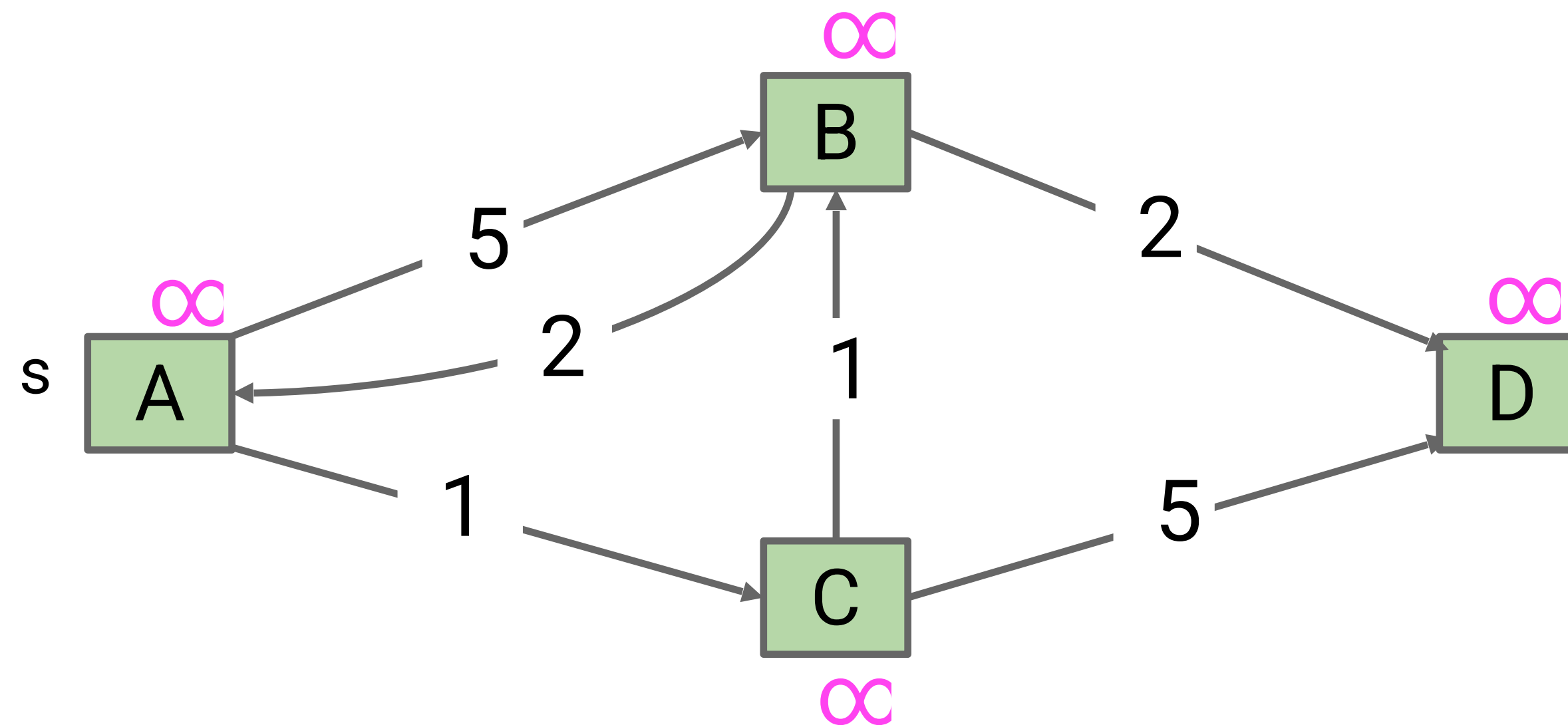
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While fringe is not empty:

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Fringe: [A]



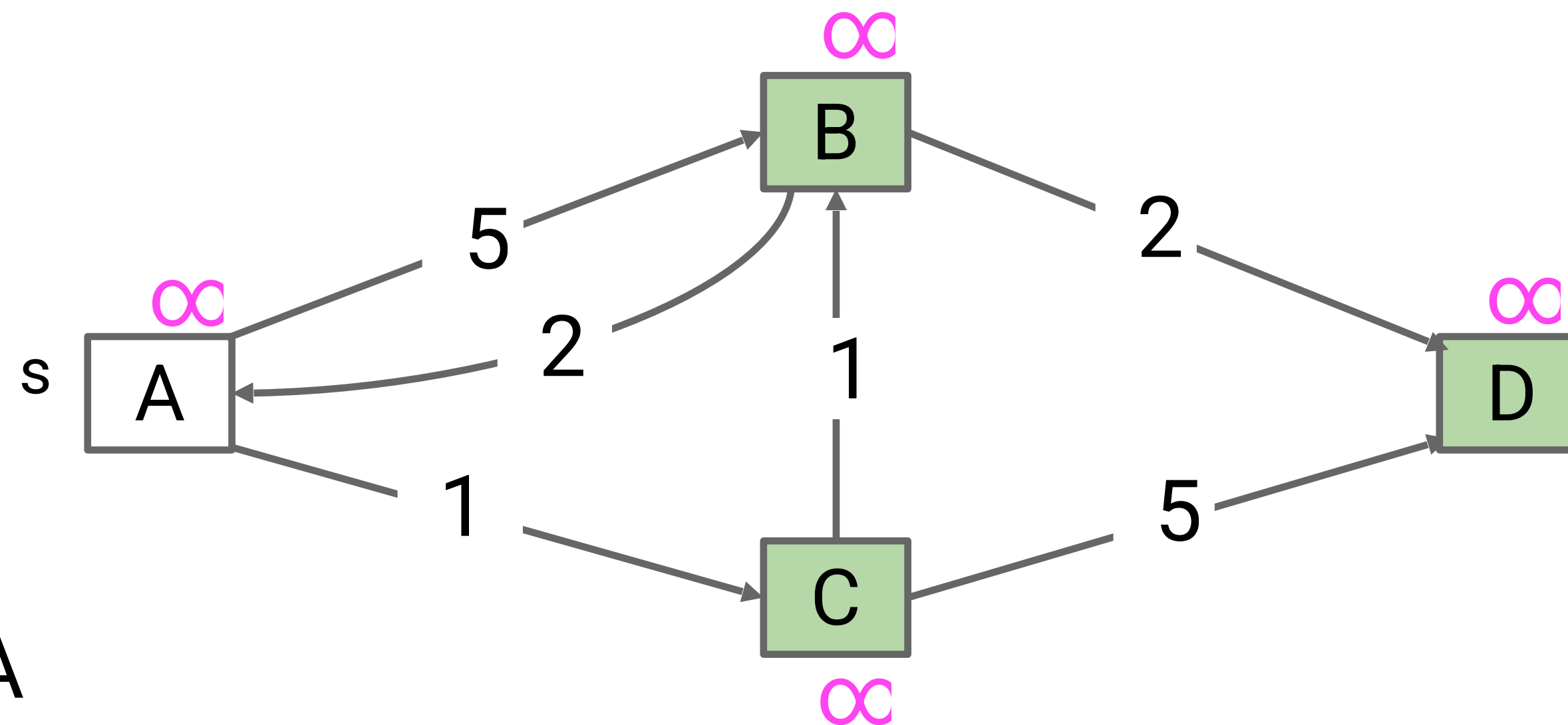
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Fringe: [A]

Removed vertex: A

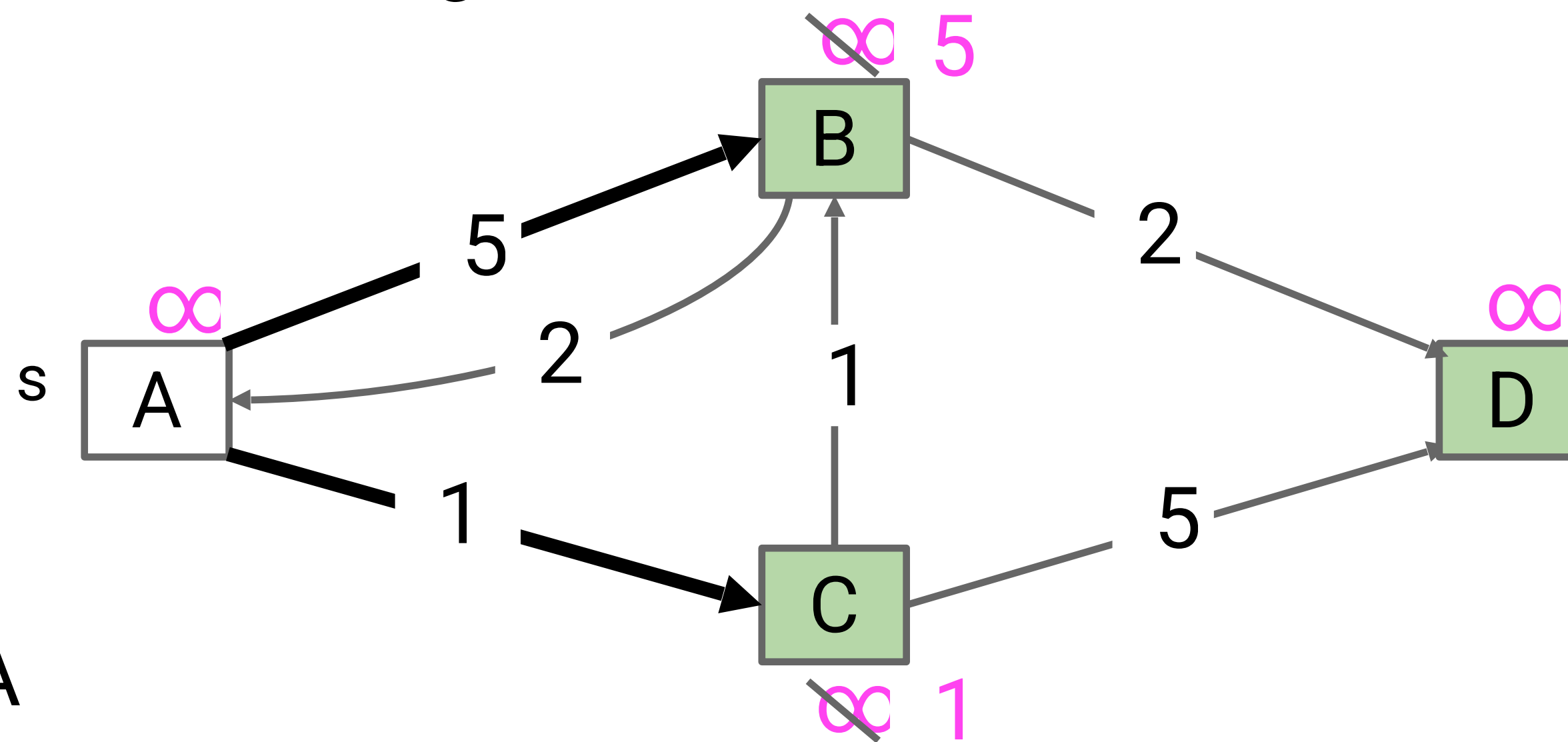
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Fringe: [A, B, C]

Removed vertex: A

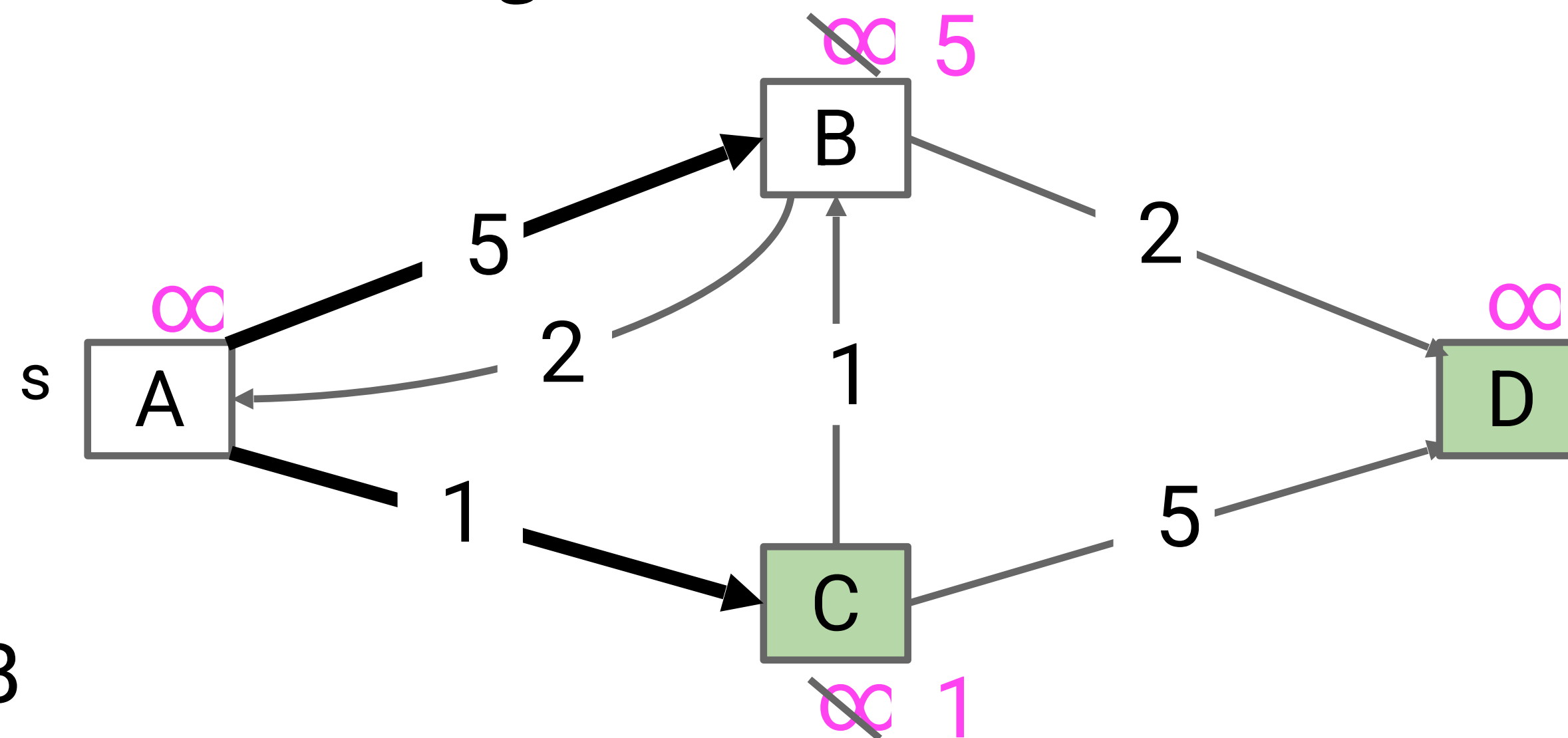
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While fringe is not empty:

**Remove a vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.



Fringe: [A, B, C]

Removed vertex: B

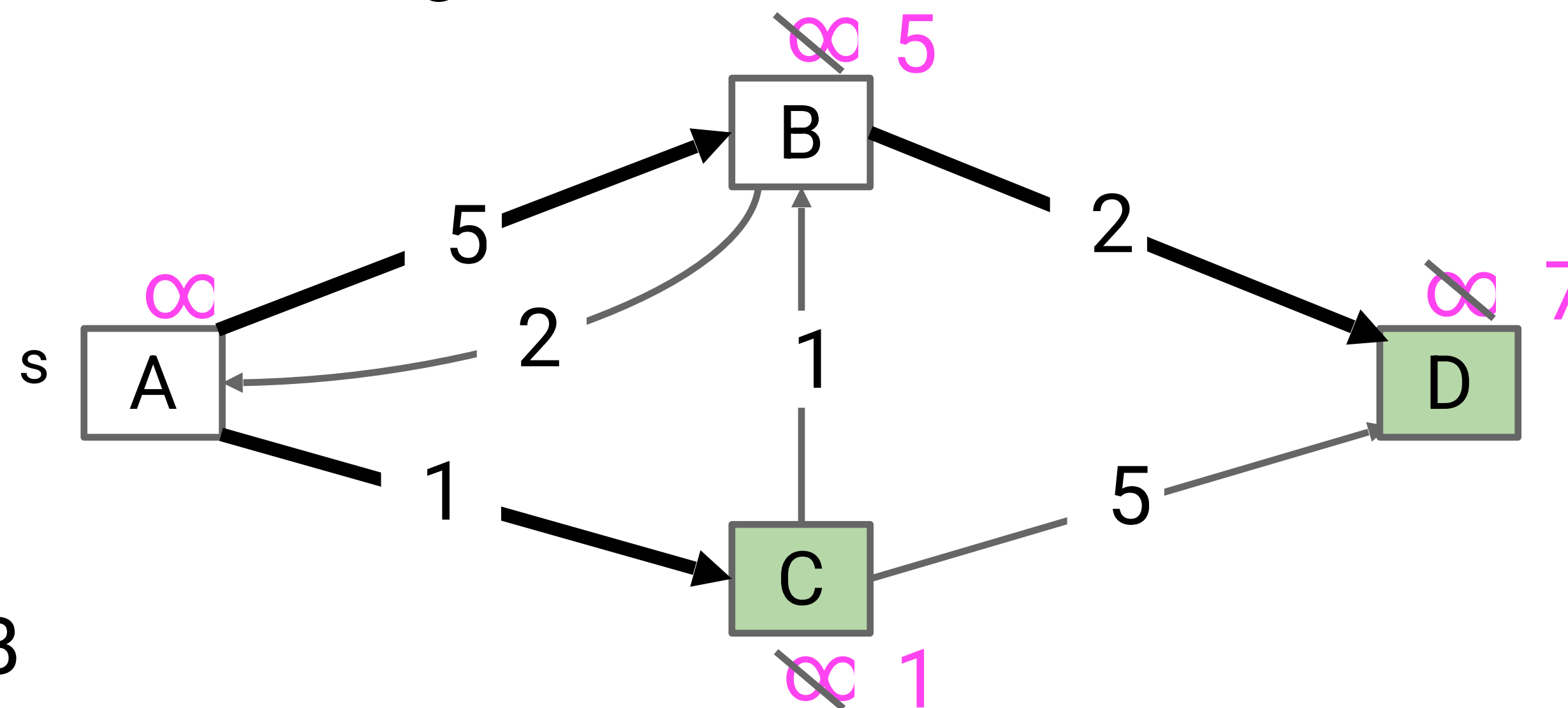
# Bad Algorithm #1 (Inspired by BFS)

Add the start (A) to the fringe.

While fringe is not empty:

Remove a vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.**



Fringe: [A, B, C, D]  
Removed vertex: B

The edge  $B \rightarrow A$  is not added to SPT, because A is already part of the SPT.

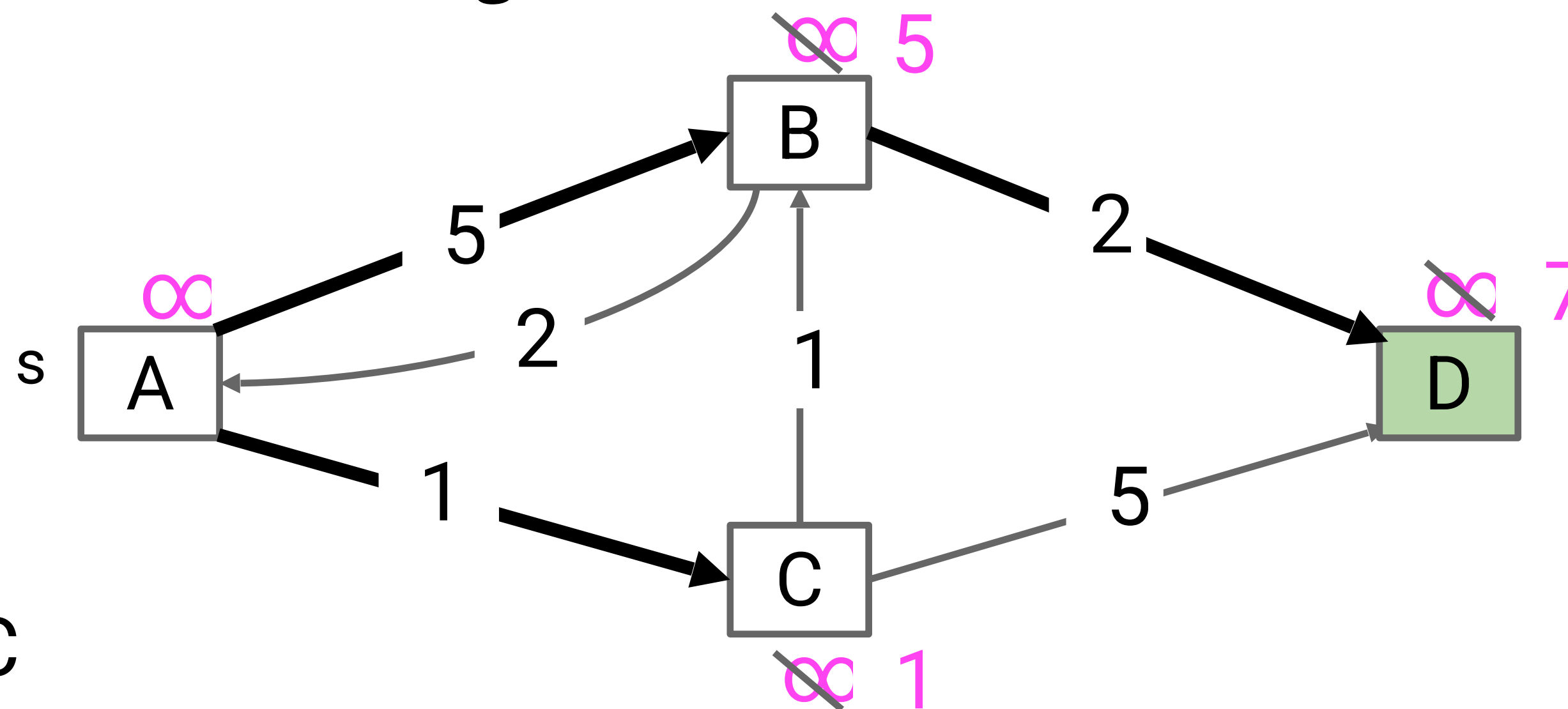
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Add the start (A) to the fringe.

While fringe is not empty:

**Remove a vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.



Fringe: [A, B, C, D]  
Removed vertex: C

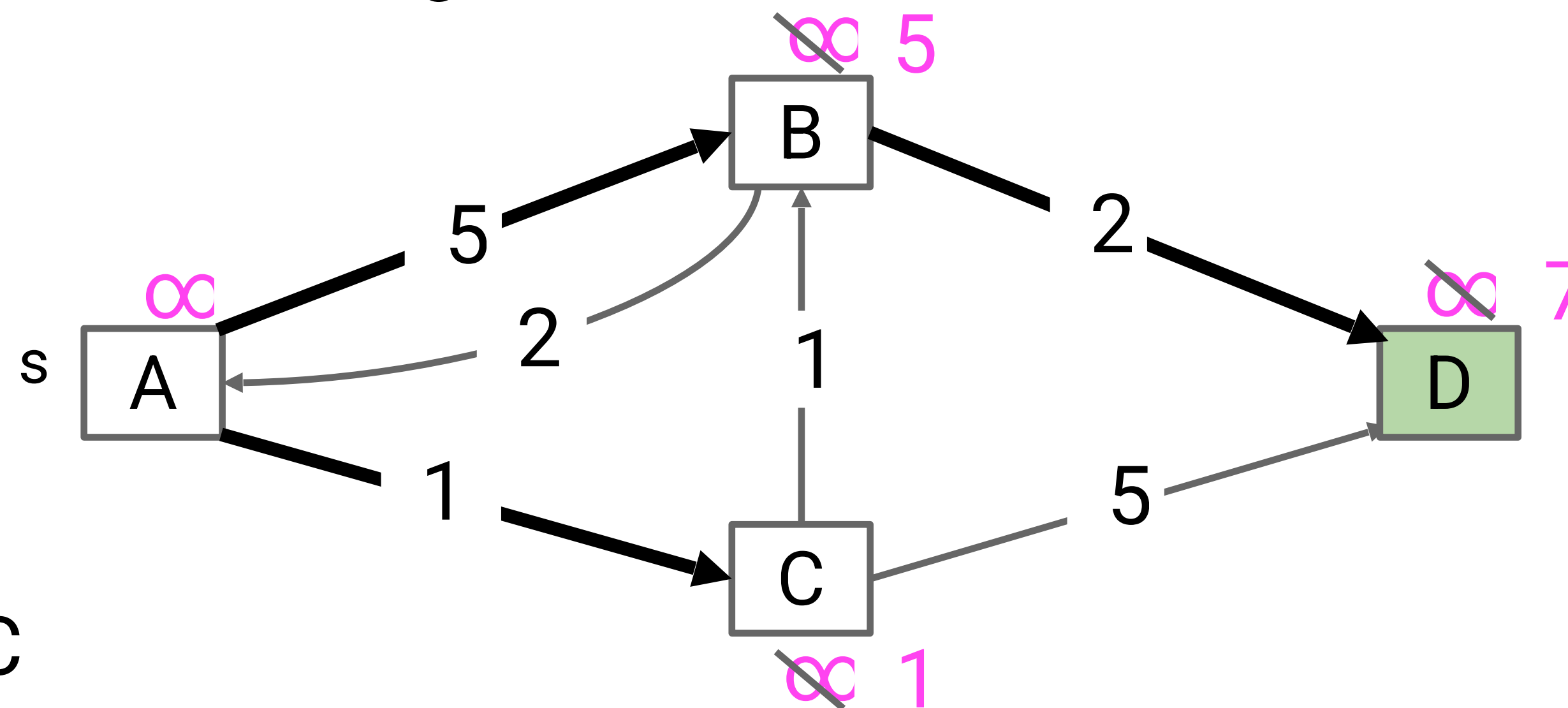
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Add the start (A) to the fringe.

While fringe is not empty:

Remove a vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.**



Fringe: [A, B, C, D]  
Removed vertex: C

Nothing happens.  
C  $\rightarrow$  B not added, B already in SPT.  
C  $\rightarrow$  D not added, D already in SPT.

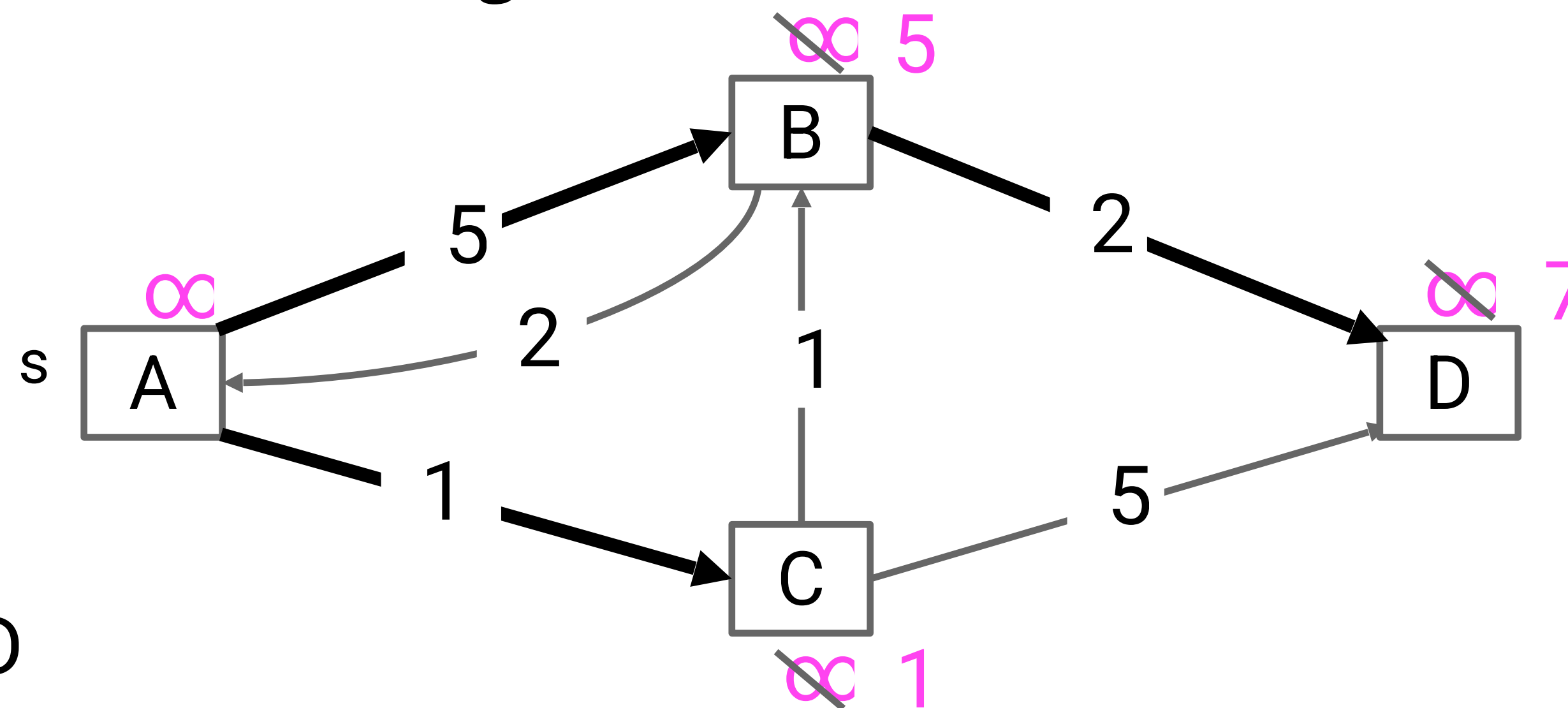
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Add the start (A) to the fringe.

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For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.



Fringe: [A, B, C, D]  
Removed vertex: D





# Bad Algorithm #1 (Inspired by BFS)

Add the start (A) to the fringe.

While fringe is not empty:

    Remove a vertex from the fringe and mark it.

    For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT,  
    add the edge, and add  $w$  to fringe.

Takeaways:

Algorithm #1 (BFS) visits:

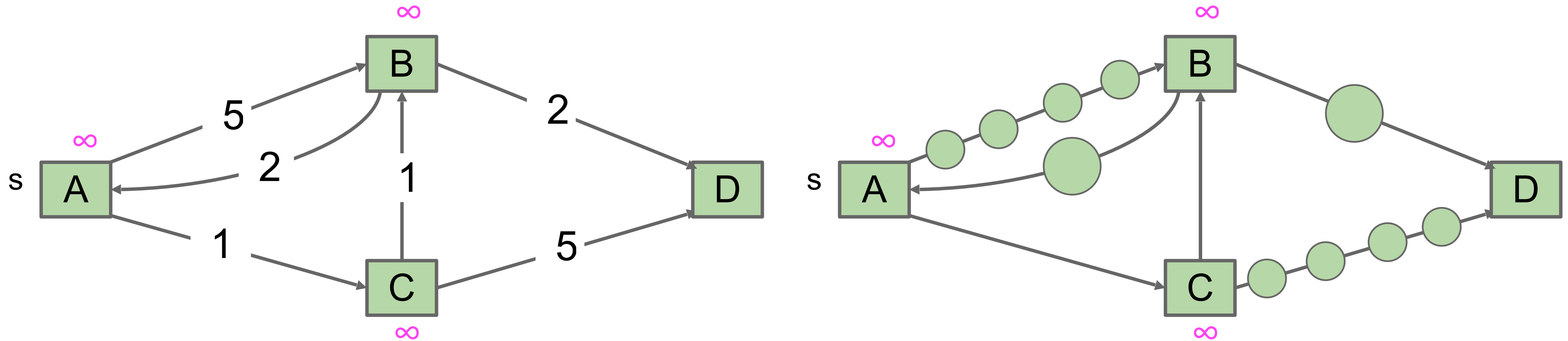
    every node 1 edge away,  
    then every node 2 edges away,  
    then every node 3 edges away, etc.

- This algorithm would work if all our edges were the same length.

# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

- When we hit one of our original nodes, add edge to the SPT.

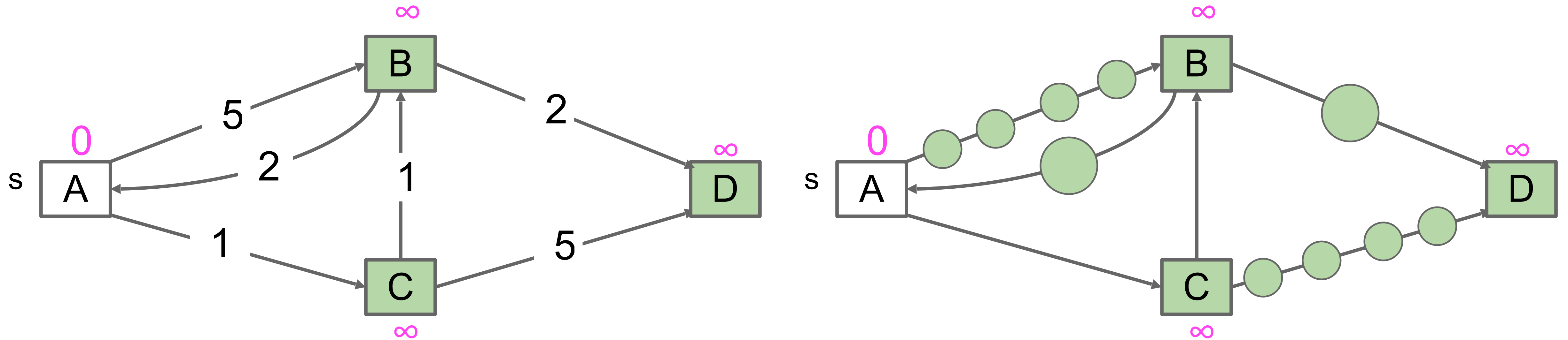


Order of visited nodes:

# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

- When we hit one of our original nodes, add edge to the SPT.

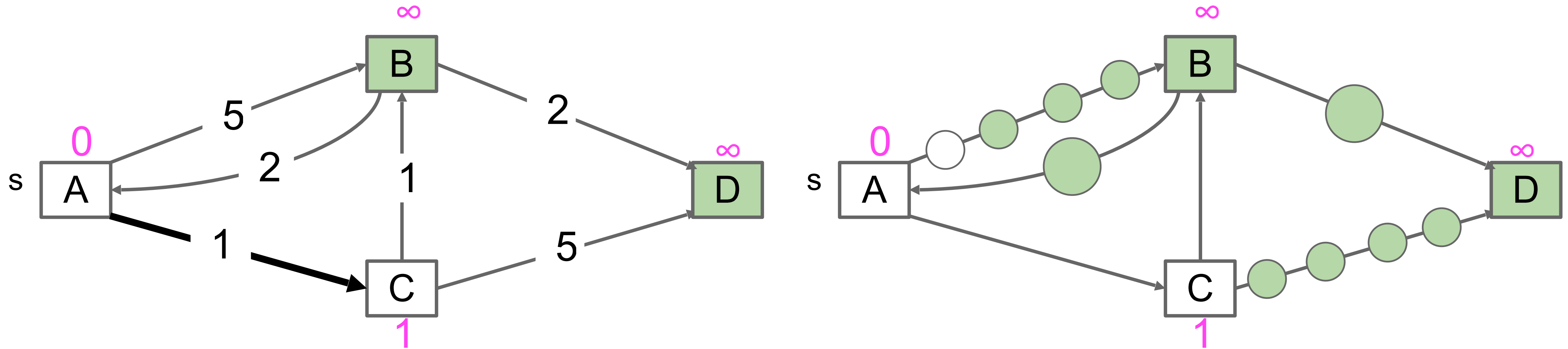


Order of visited nodes: A

# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

- When we hit one of our original nodes, add edge to the SPT.

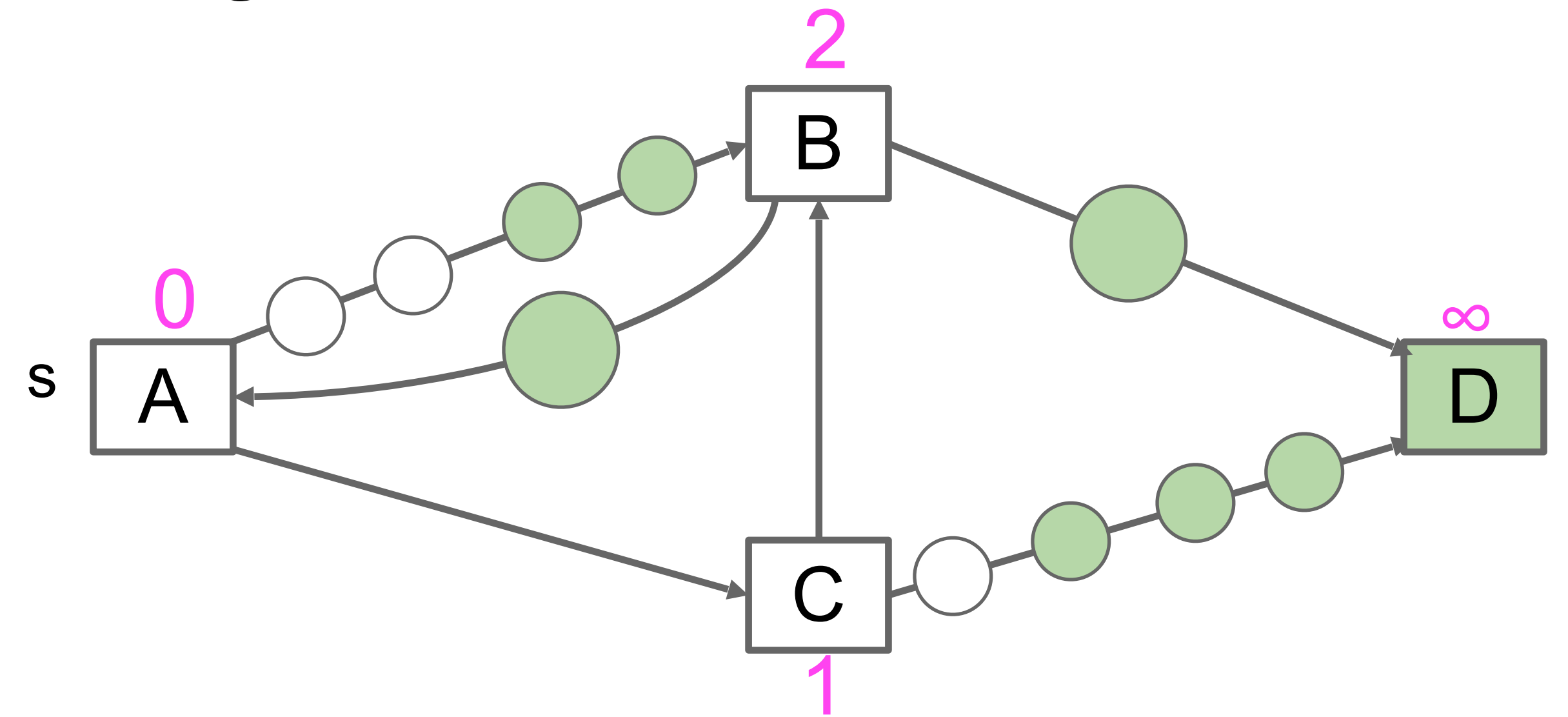
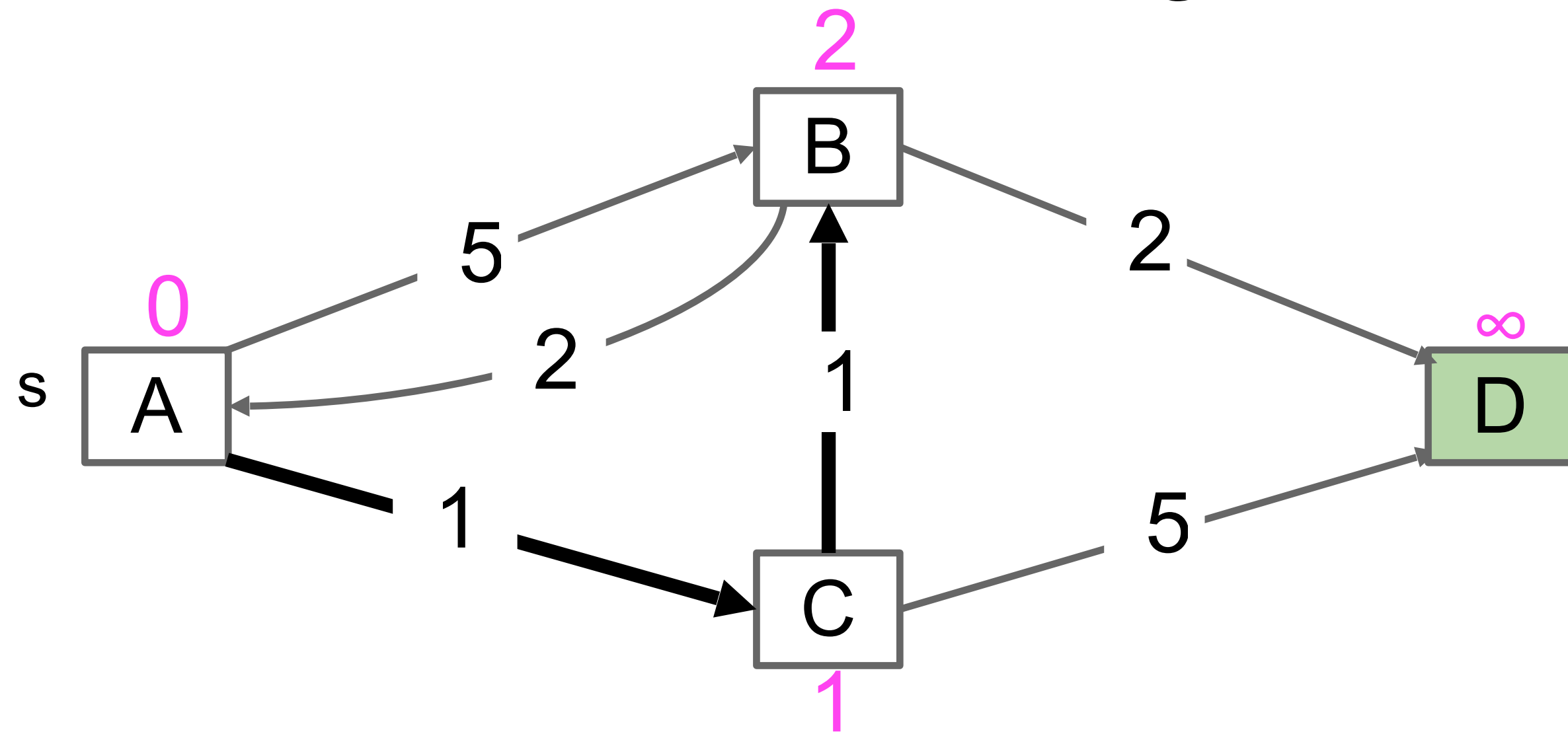


Order of visited nodes: AC

# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

- When we hit one of our original nodes, add edge to the SPT.

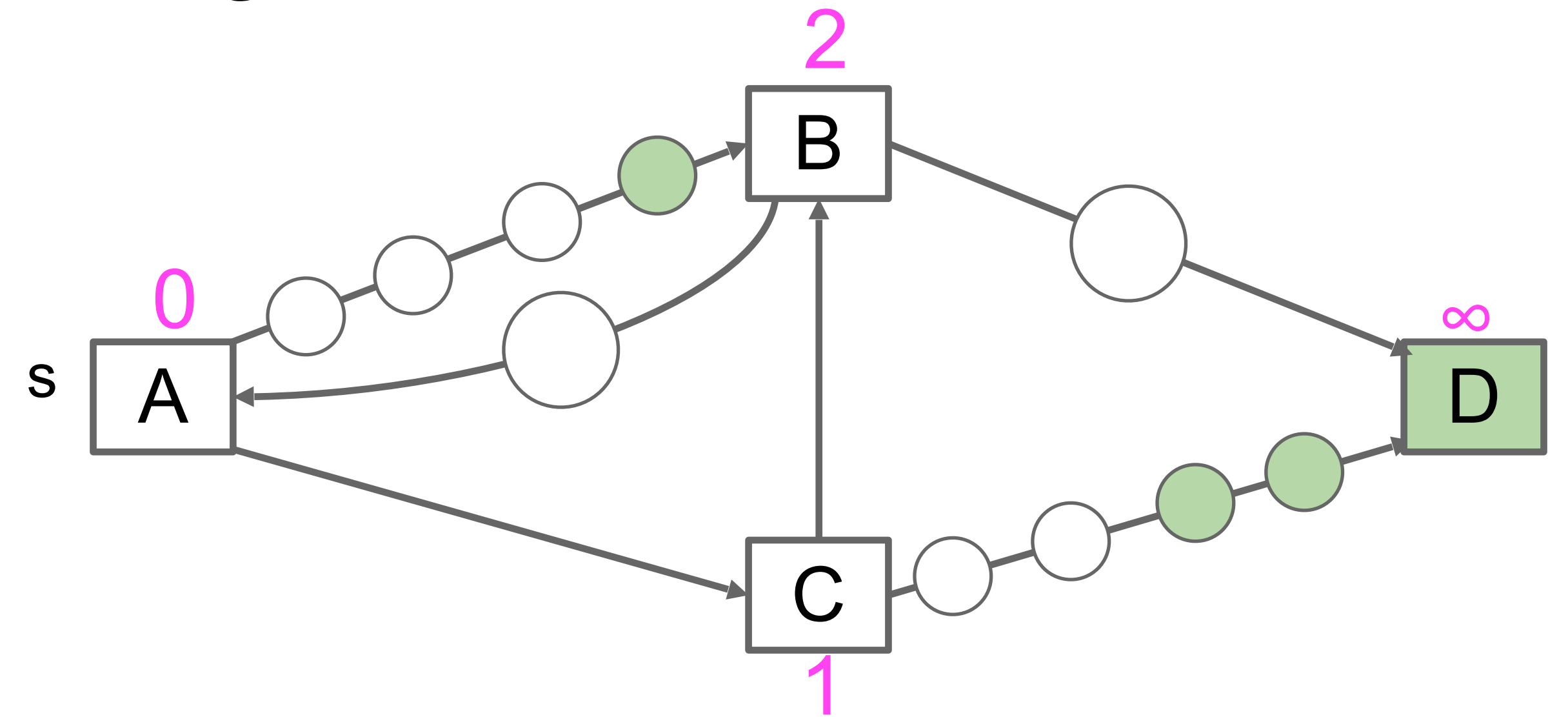
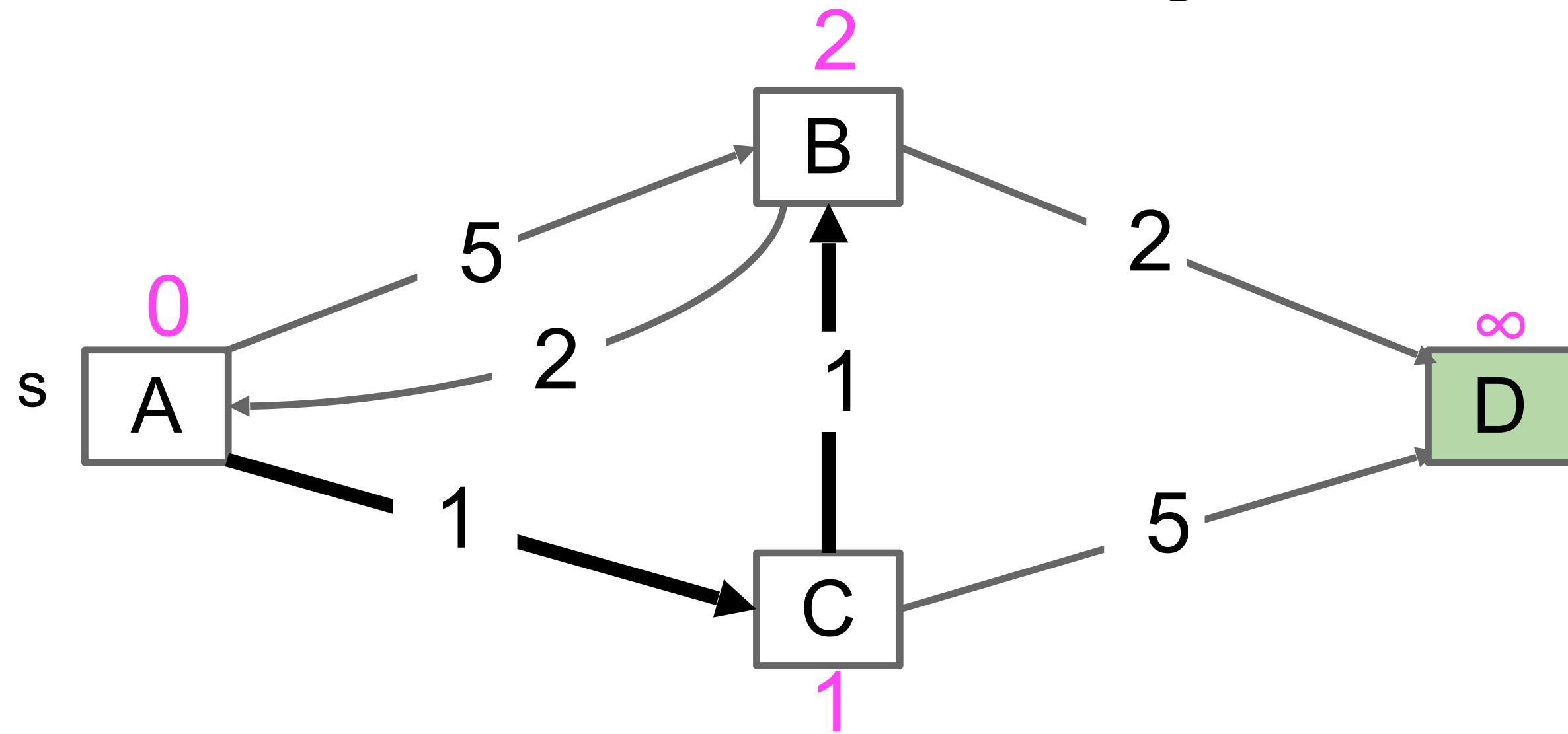


Order of visited nodes: ACB

# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

- When we hit one of our original nodes, add edge to the SPT.

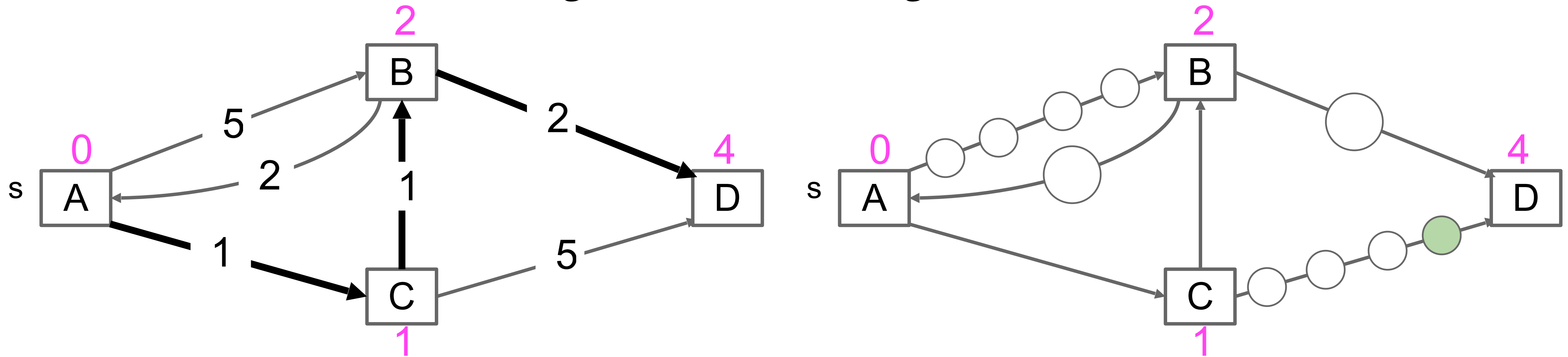


Order of visited nodes: ACB

# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

- When we hit one of our original nodes, add edge to the SPT.



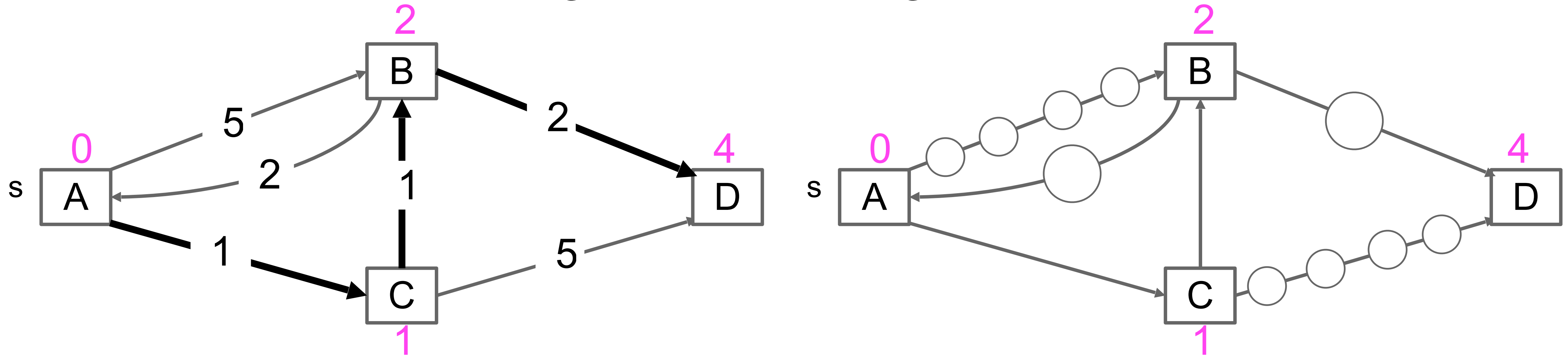
Order of visited nodes: ACBD



# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

- When we hit one of our original nodes, add edge to the SPT.



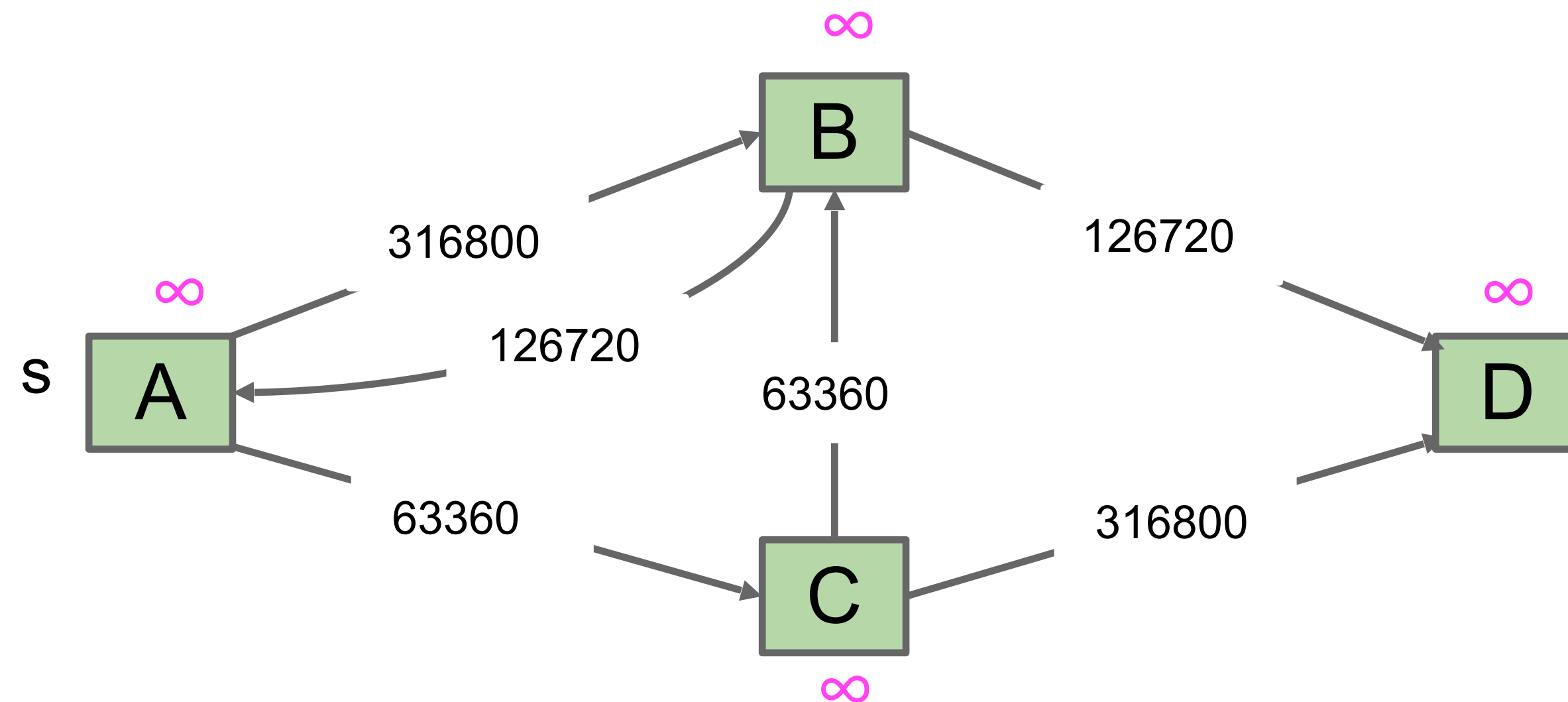
Order of visited nodes: ACBD

# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

Takeaways:

- It works, but can be really slow. For example, consider the graph below.
- What if we measured in inches instead of miles? Or had fractional weights?



# Bad Algorithm #2 (Dummy Nodes)

Bad algorithm #2: Create a new graph by adding a bunch of dummy nodes every unit along an edge, then run breadth-first search.

Takeaways:

Algorithm #1 (BFS) visits:

every node 1 edge away,  
then every node 2 edges away,  
then every node 3 edges away, etc.

Algorithm #2 (dummy nodes) visits:

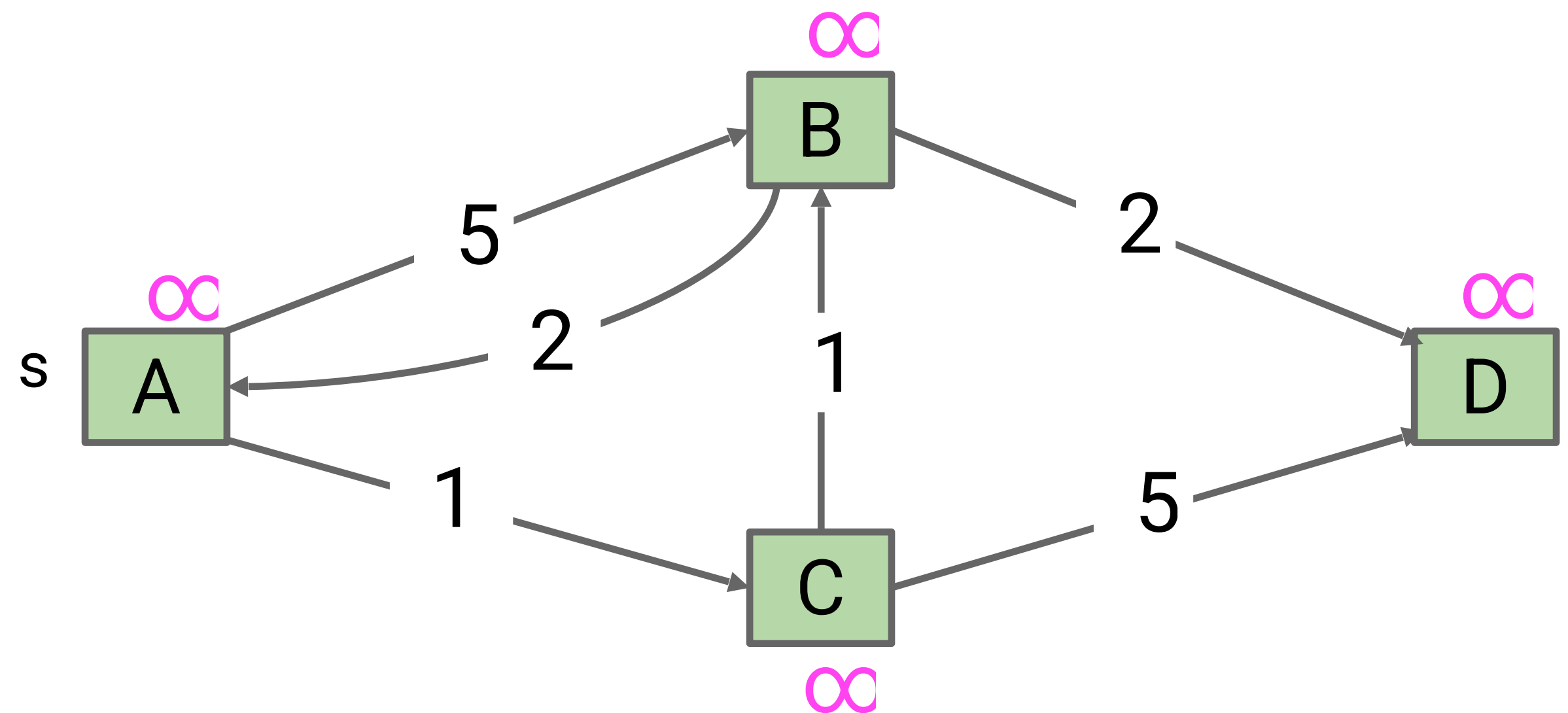
every node distance 1 away,  
then every node distance 2 away,  
then every node distance 3 away, etc.

- Algorithm #2 order is sometimes called **best-first** order.
- Let's try to visit the nodes in the same order as Algorithm #2 did, but without creating dummy nodes.

# Bad Algorithm #3 (Best-First Search)

Bad algorithm #3: Perform best-first search.

- Similar to BFS, but we remove the closest edge from the fringe each time.
- We can use a **priority queue** to track the closest edge.



# Bad Algorithm #3 (Best-First Search)

Add the start (A) to the fringe.

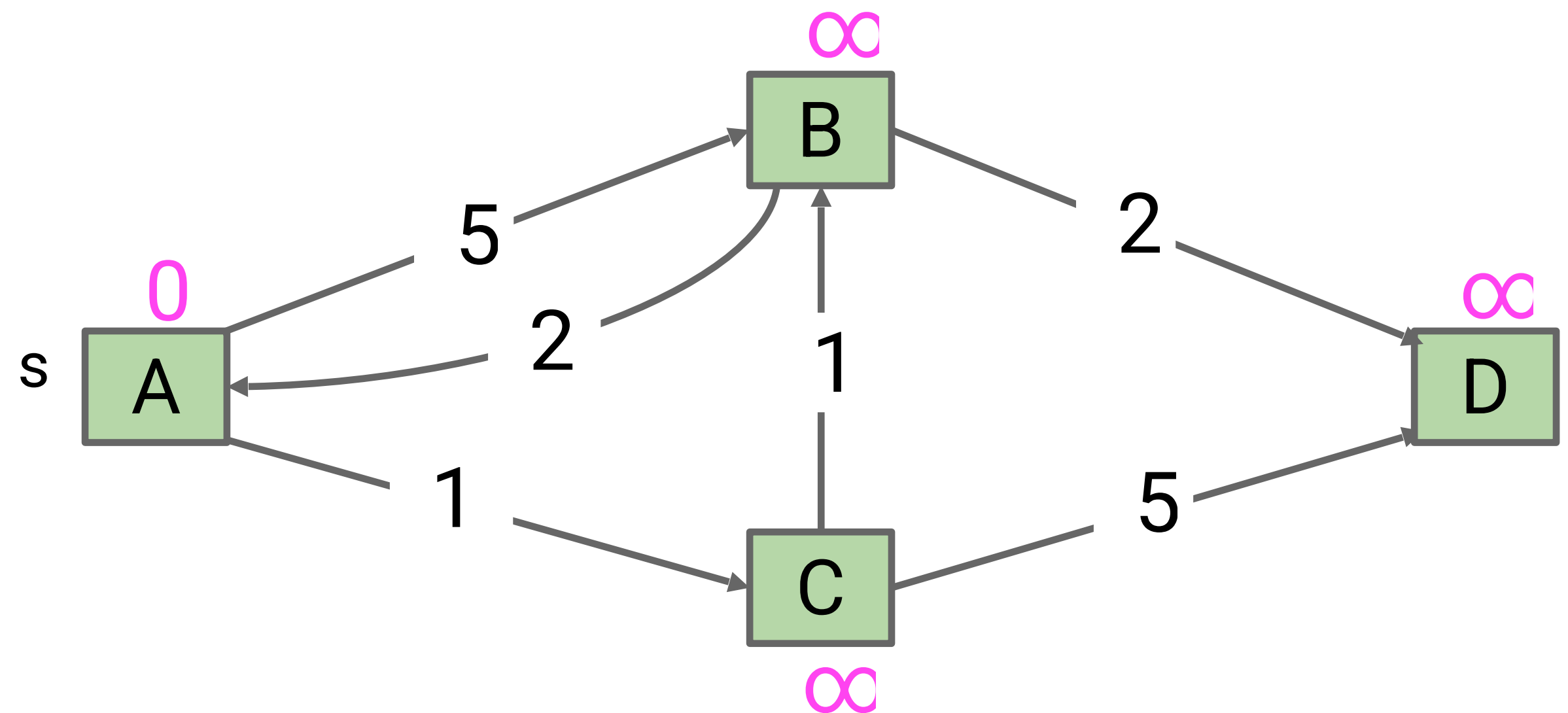
Only difference from Algorithm #1:  
We added the word "closest".

While fringe is not empty:

Remove the **closest** vertex from the fringe and mark it.

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.

Fringe: [A=0]



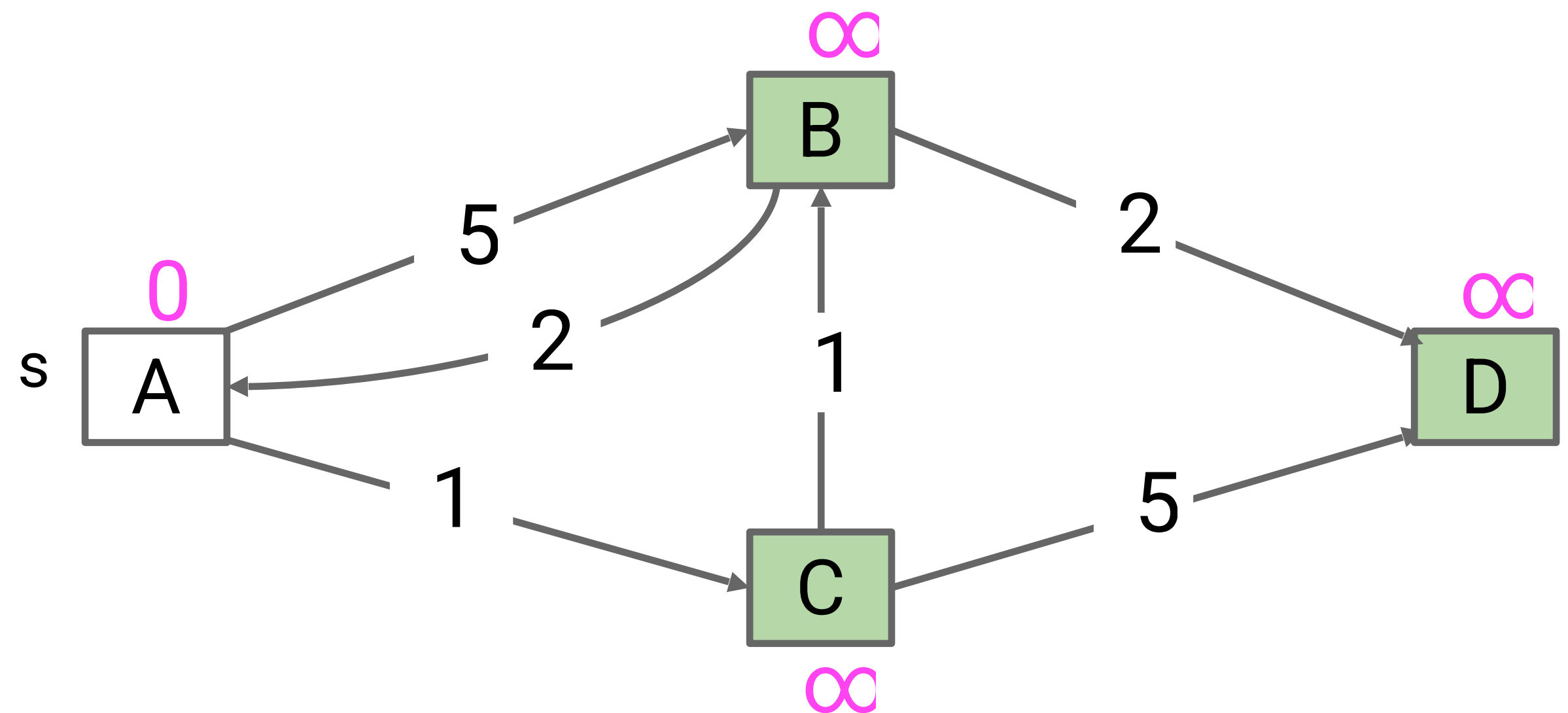
# Bad Algorithm #3 (Best-First Search)

Add the start (A) to the fringe.

While fringe is not empty:

**Remove the closest vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.



Fringe: [~~A=0~~]

Removed vertex: A

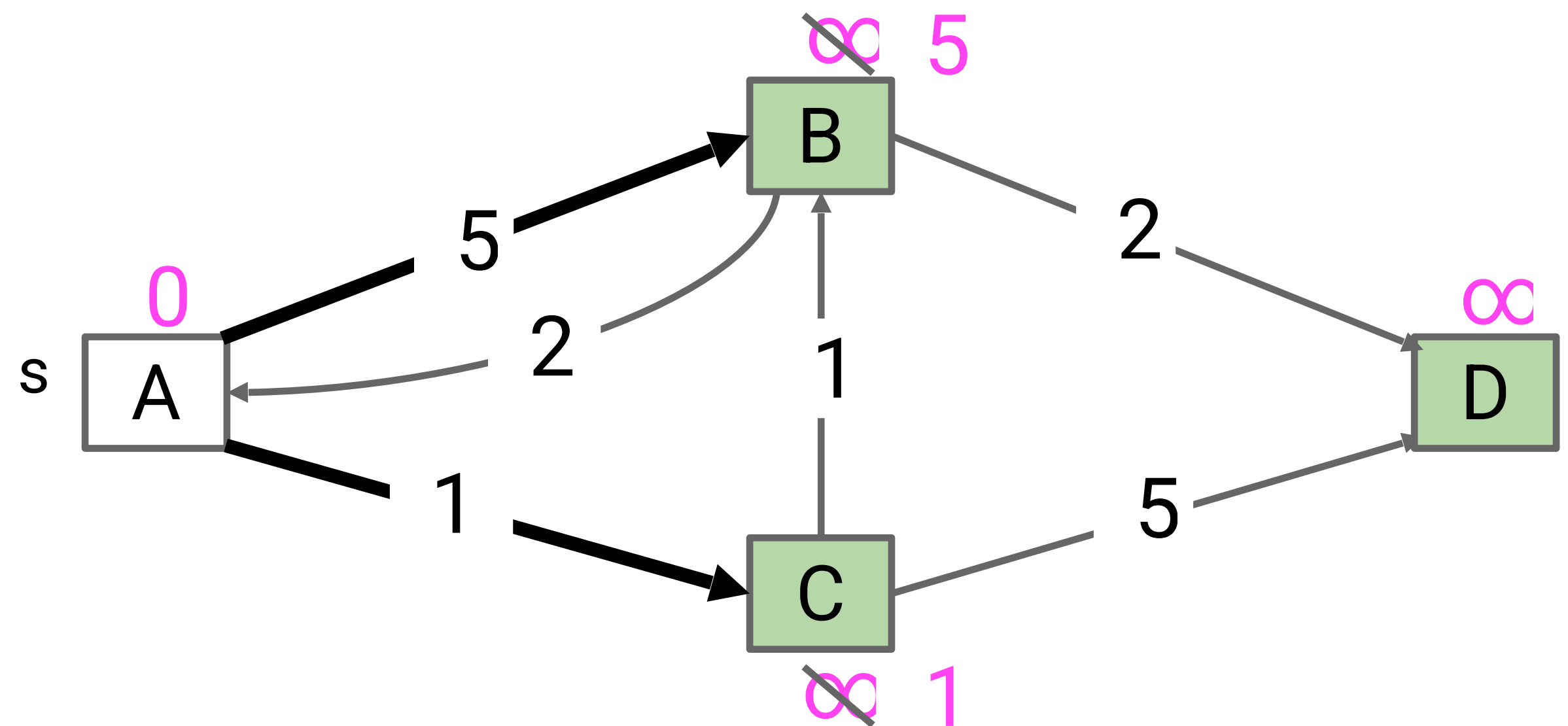
# Bad Algorithm #3 (Best-First Search)

Add the start (A) to the fringe.

While fringe is not empty:

Remove the closest vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.**



Fringe: [~~A=0~~, C=1, B=5]

Removed vertex: A





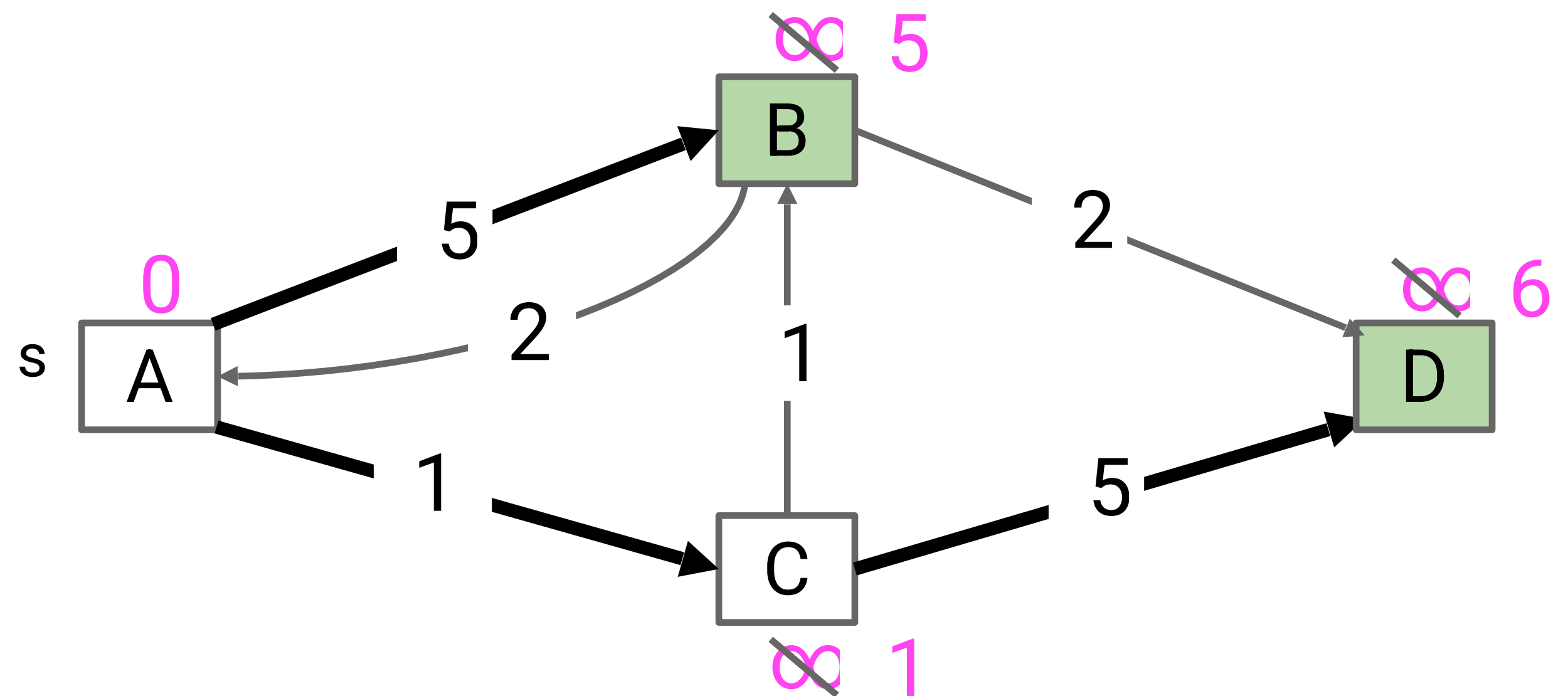
# Bad Algorithm #3 (Best-First Search)

Add the start (A) to the fringe.

While fringe is not empty:

Remove the closest vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.**



Fringe: [~~A=0~~, ~~C=1~~, B=5, D=6]

Removed vertex: C

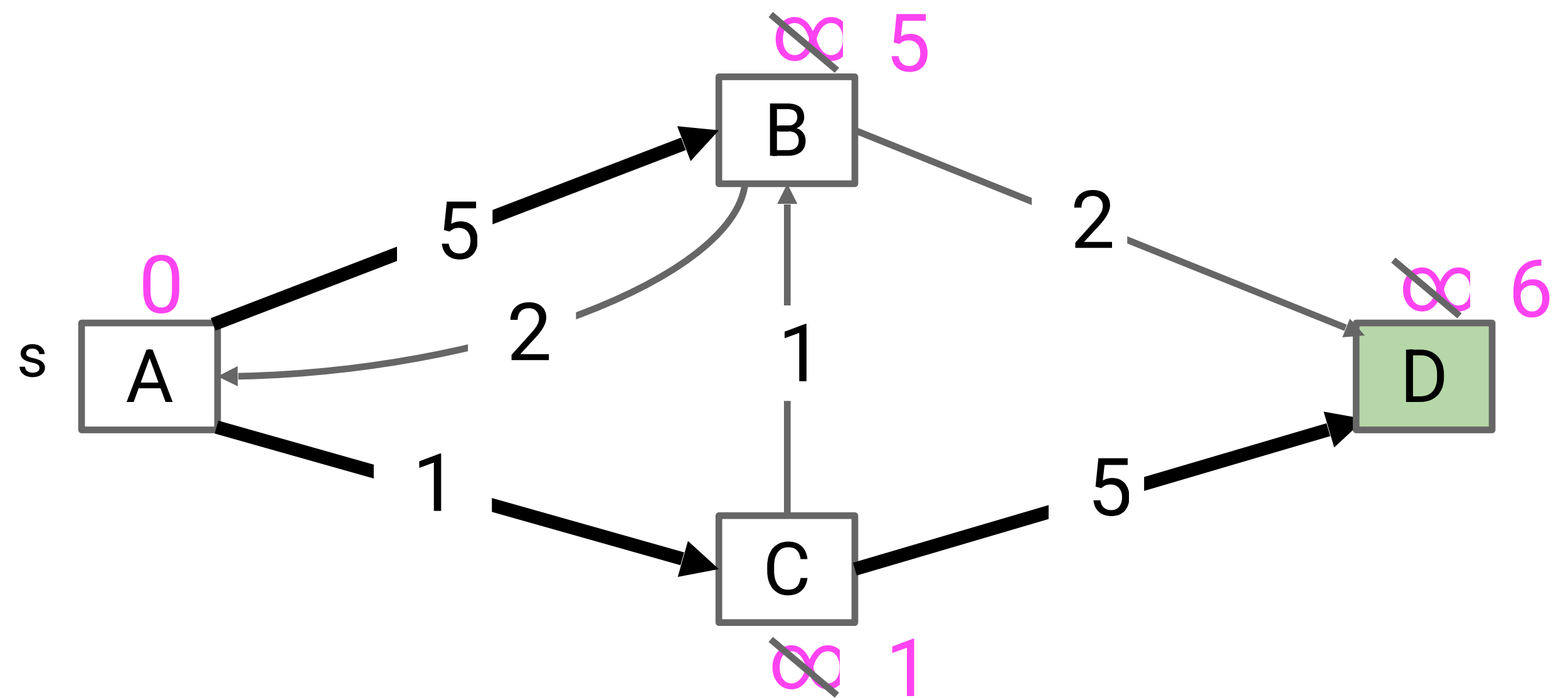
# Bad Algorithm #3 (Best-First Search)

Add the start (A) to the fringe.

While fringe is not empty:

**Remove the closest vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.



Fringe: [~~A=0~~, ~~C=1~~, ~~B=5~~, D=6]

Removed vertex: B

# Bad Algorithm #3 (Best-First Search)

Add the start (A) to the fringe.

While fringe is not empty:

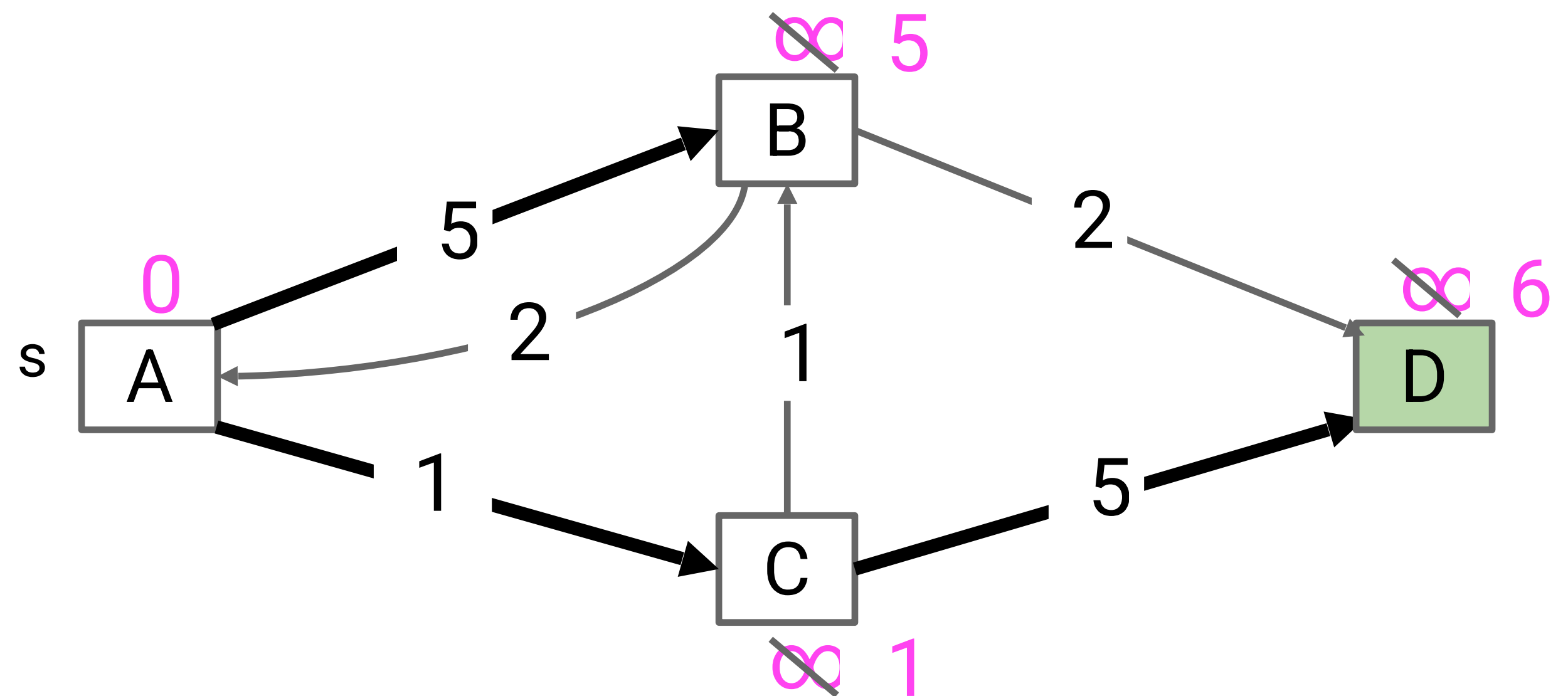
Remove the closest vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.**

The only outgoing edge is  $B \rightarrow D$ .  
D is already part of the SPT, so do nothing.

Fringe: [~~A=0~~, ~~C=1~~, ~~B=5~~, D=6]

Removed vertex: B



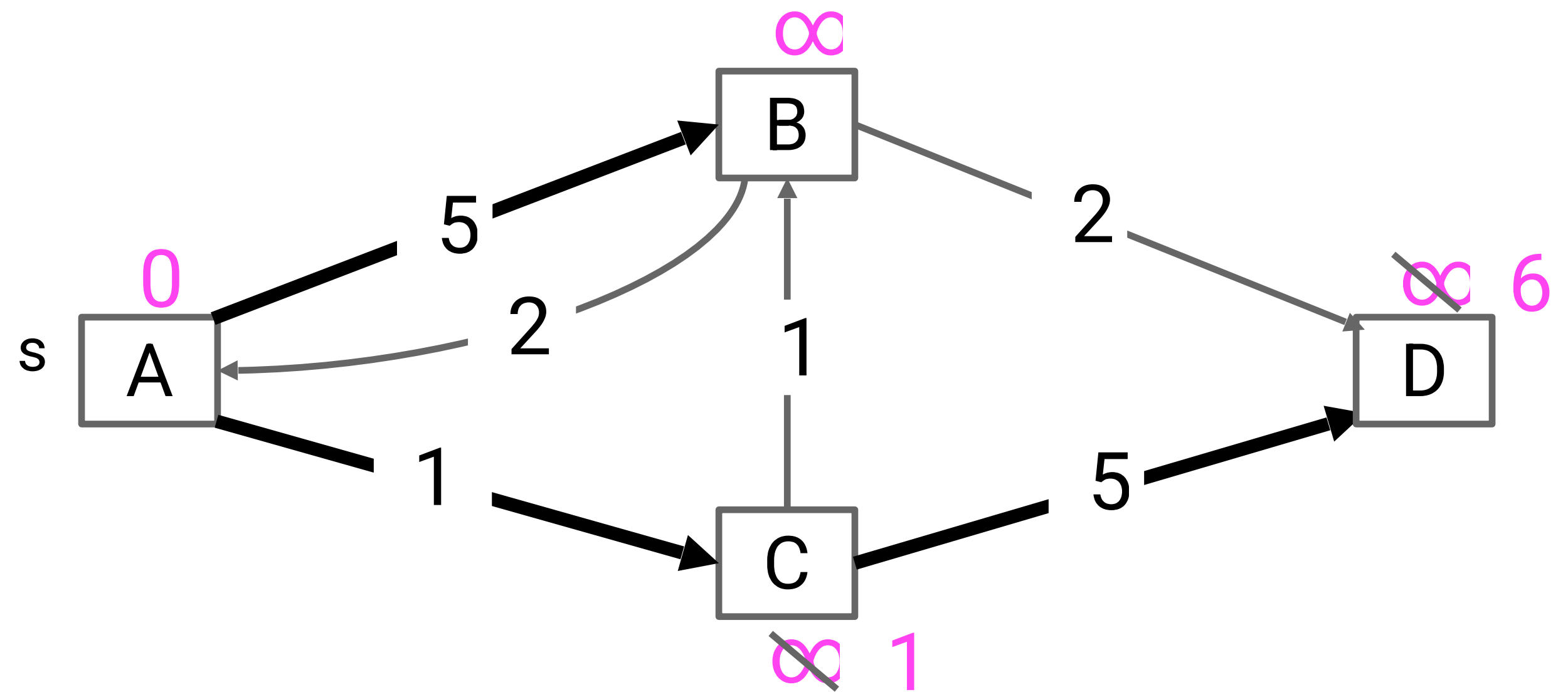
# Bad Algorithm #3 (Best-First Search)

Add the start (A) to the fringe.

While fringe is not empty:

**Remove the closest vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, and add  $w$  to fringe.



Fringe: [~~A=0~~, ~~C=1~~, ~~B=5~~, ~~D=6~~]

Removed vertex: D



# Bad Algorithm #3 (Best-First Search)

Bad algorithm #3: Perform best-first search.

- Similar to BFS, but we remove the closest edge from the fringe each time.
- We can use a priority queue to track the closest edge.

Takeaways:

- Pro: We visited the nodes in best-first order (same order as in Algorithm #2), without creating dummy nodes.
- Con: We got the wrong answer. Why?
- Let's revisit the step where things went wrong.

# Bad Algorithm #3 (Best-First Search)

For each outgoing edge  $v \rightarrow w$ : if  $w$  is not already part of SPT, add the edge, mark  $w$ , and add  $w$  to fringe.

$C \rightarrow B$  edge: B was in the SPT (via  $A \rightarrow B$ ), so we did nothing.

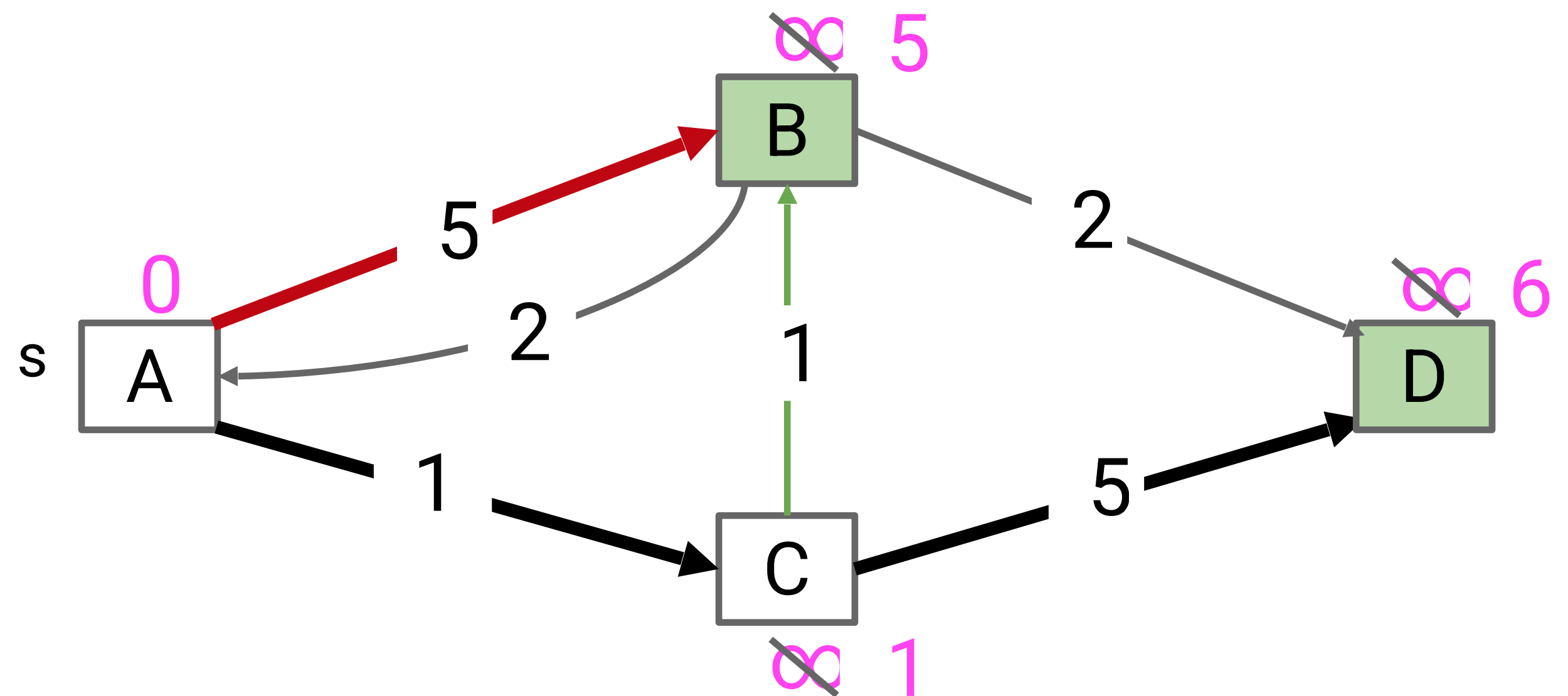
What should we have done here?

- We should have added edge  $C \rightarrow B$ , and thrown out the old edge ( $A \rightarrow B$ ) to B. Why?
- The distance to B via  $C \rightarrow B$  is 2.

This is better than the currently best known distance to B (5, via  $A \rightarrow B$ ).

Fringe: [~~A=0~~, ~~C=1~~, B=5, D=6]

Removed vertex: C

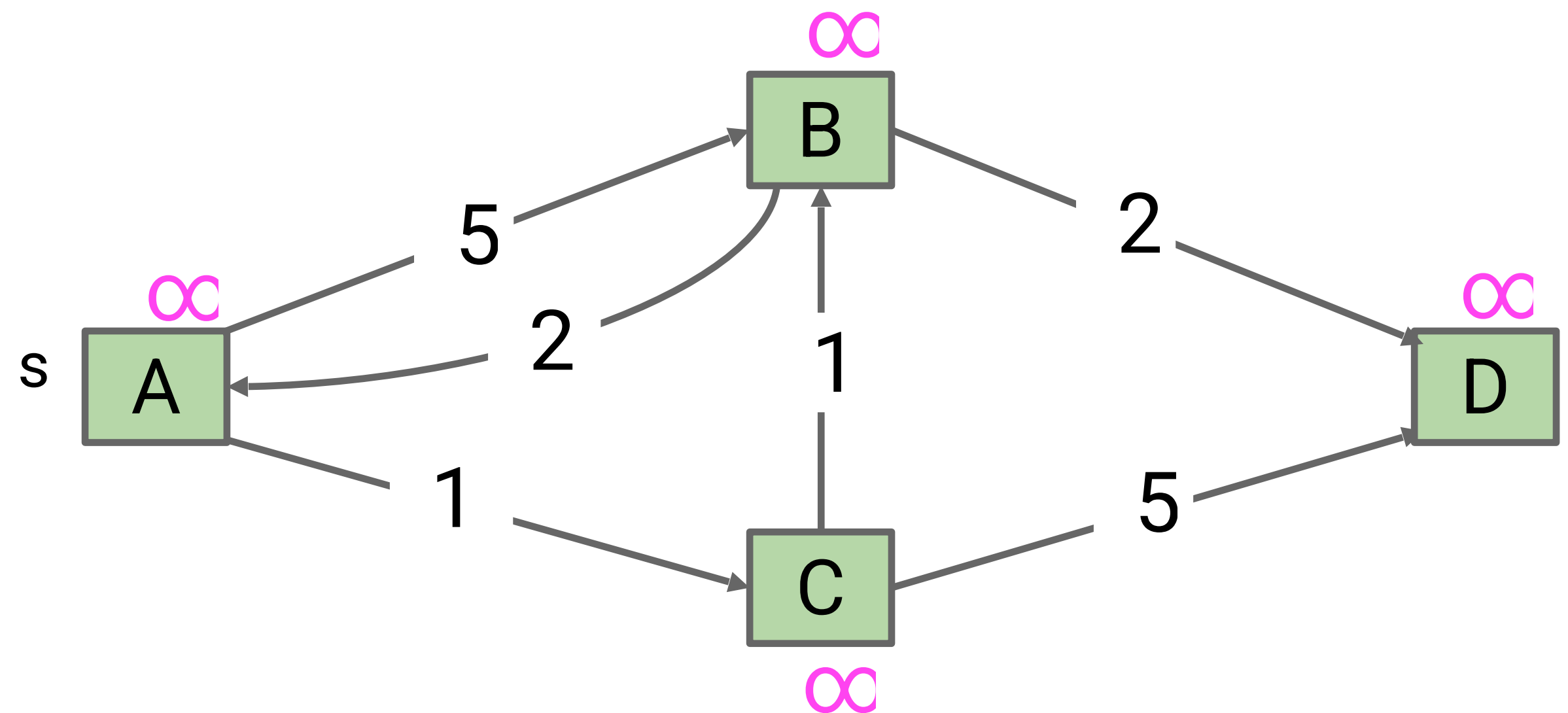


# Finding a Shortest Paths Tree Algorithmically

Dijkstra's Algorithm:

- So far, we've added an edge  $v \rightarrow w$  if  $w$  is not already part of the SPT.
- Instead, we should add an edge if that edge yields better distance.
- Use the priority queue to track best known distances.

We'll call this process "edge relaxation".





# Finding a Shortest Paths Tree Algorithmically

**Add all vertices to the fringe.**

While fringe is not empty:

Remove the closest vertex from the fringe and mark it.

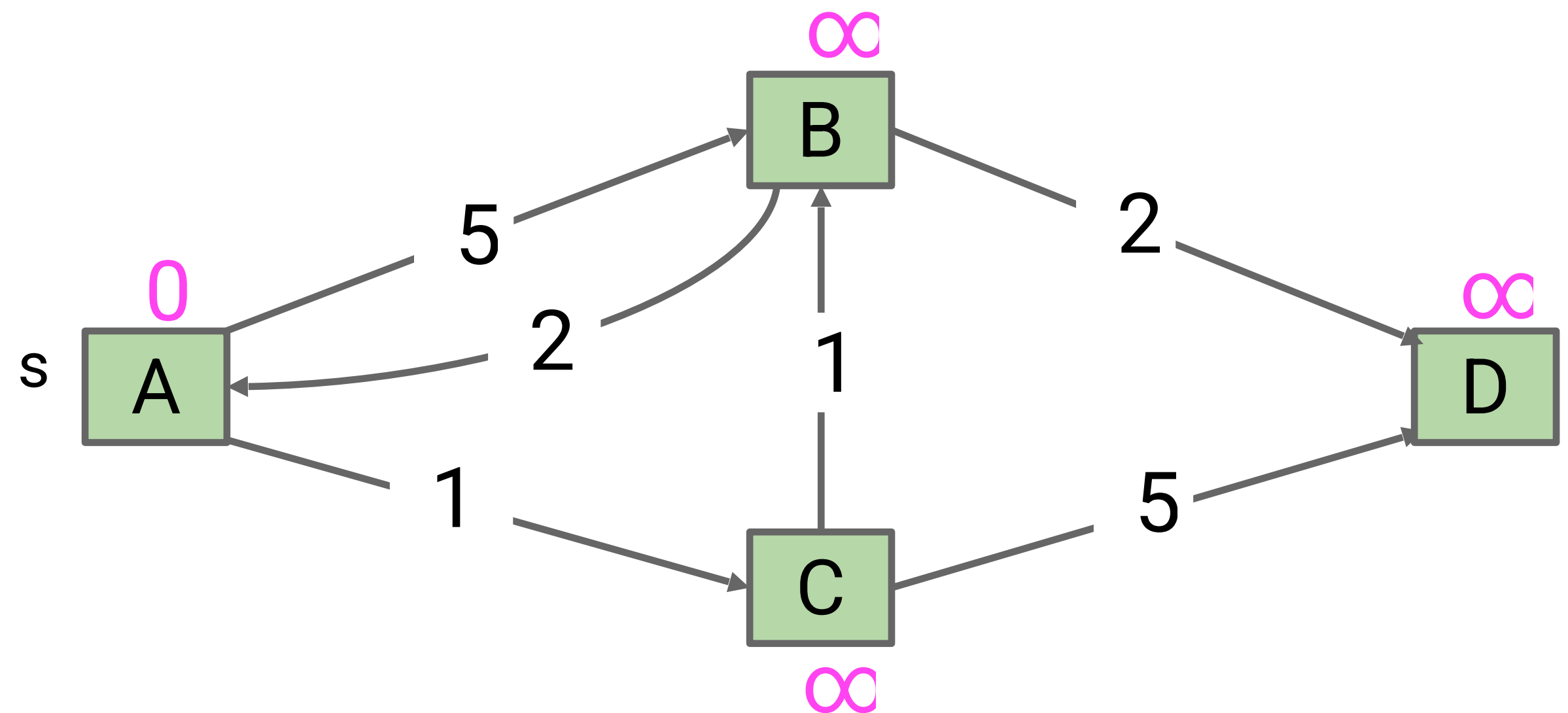
For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.

Key difference from Algorithm #3:  
The condition for adding an edge.  
(This used to say "if  $w$  not in SPT").

Extra bookkeeping: Instead of adding to the fringe as we go, we'll add all vertices to start.

This lets us track the best known distance to each vertex.

Fringe:  $[A=0, B=\infty, C=\infty, D=\infty]$



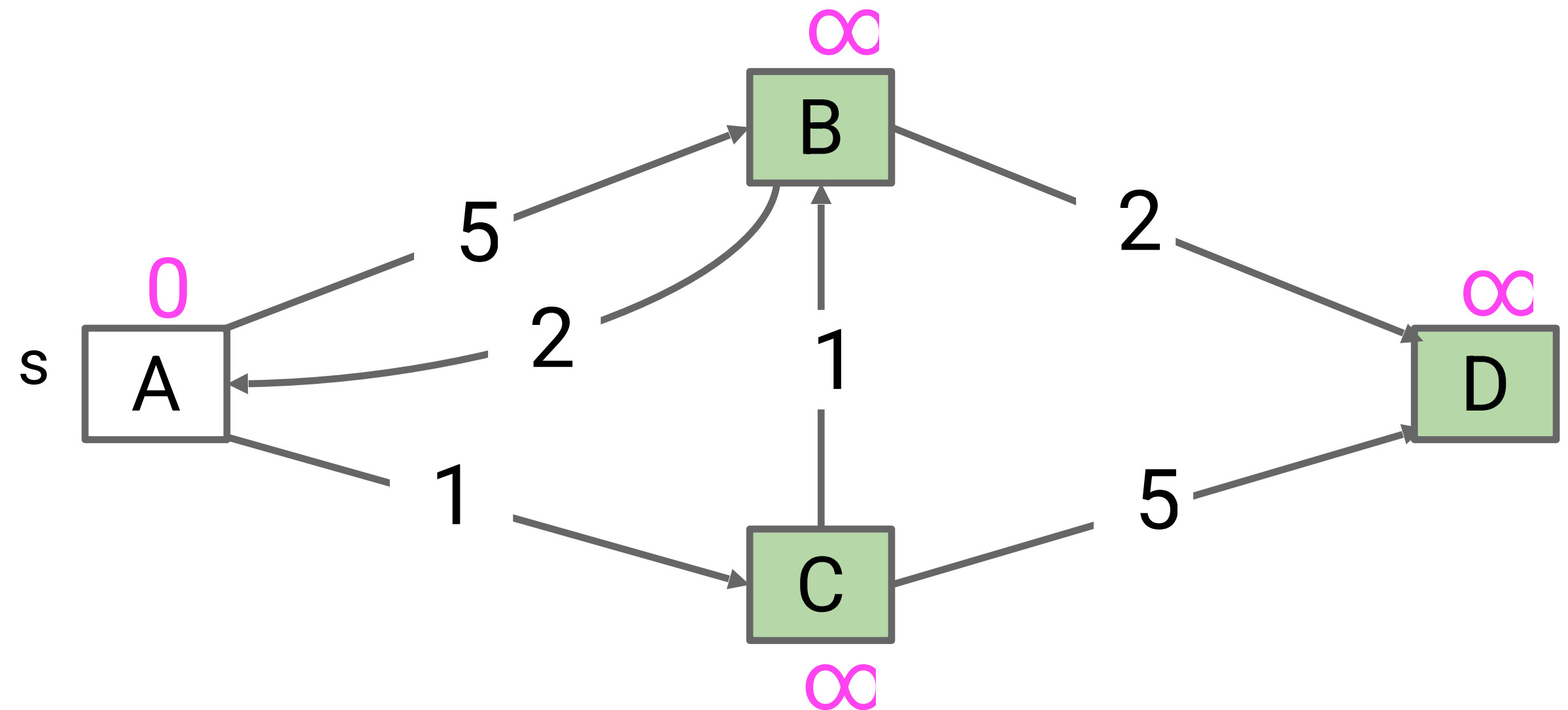
# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

**Remove the closest vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.



Fringe: [~~A=0~~, B= $\infty$ , C= $\infty$ , D= $\infty$ ]

Removed vertex: A

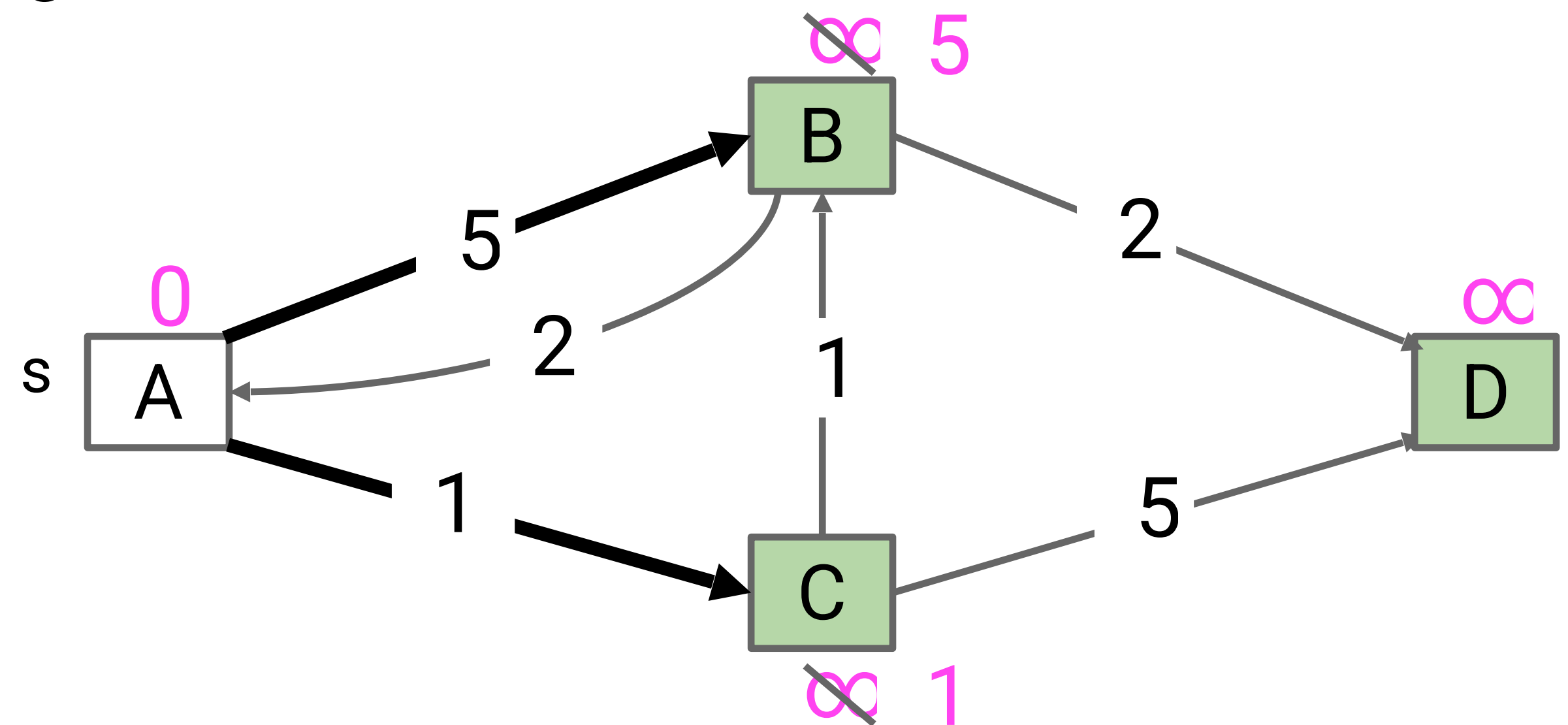
# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

Remove the closest vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.**



Fringe: [~~A=0~~, C=1, B=5, D=∞]

Removed vertex: A

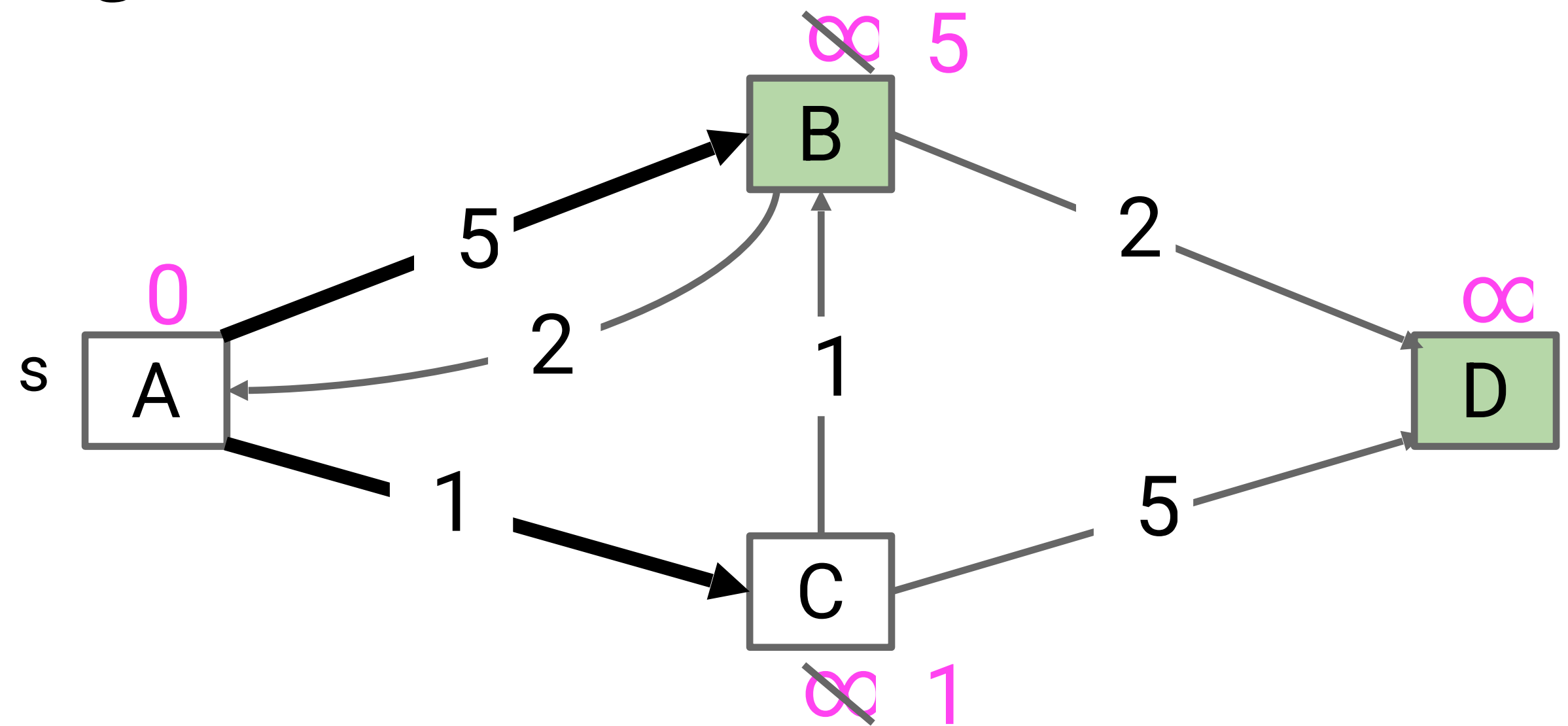
# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

**Remove the closest vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.



Fringe: [~~A=0~~, ~~C=1~~, B=5, D=∞]

Removed vertex: C

# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

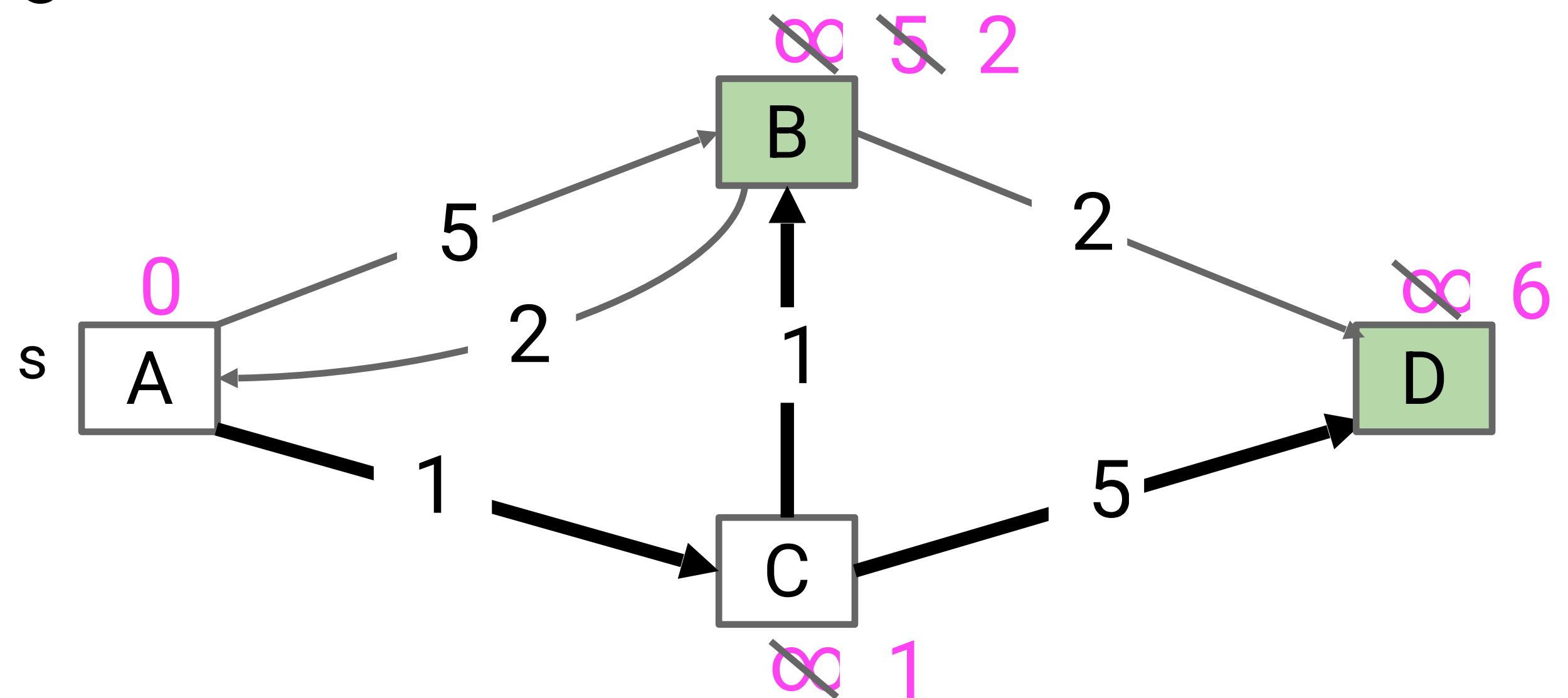
Remove the closest vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.**

Improvement: We used  $C \rightarrow B$  because the distance via  $C \rightarrow B$  (2) is better than the distance via  $A \rightarrow B$  (5). This also means we throw out the old edge ( $A \rightarrow B$ ) to B.

Fringe: [~~A=0~~, ~~C=1~~, B=2, D=6]

Removed vertex: C



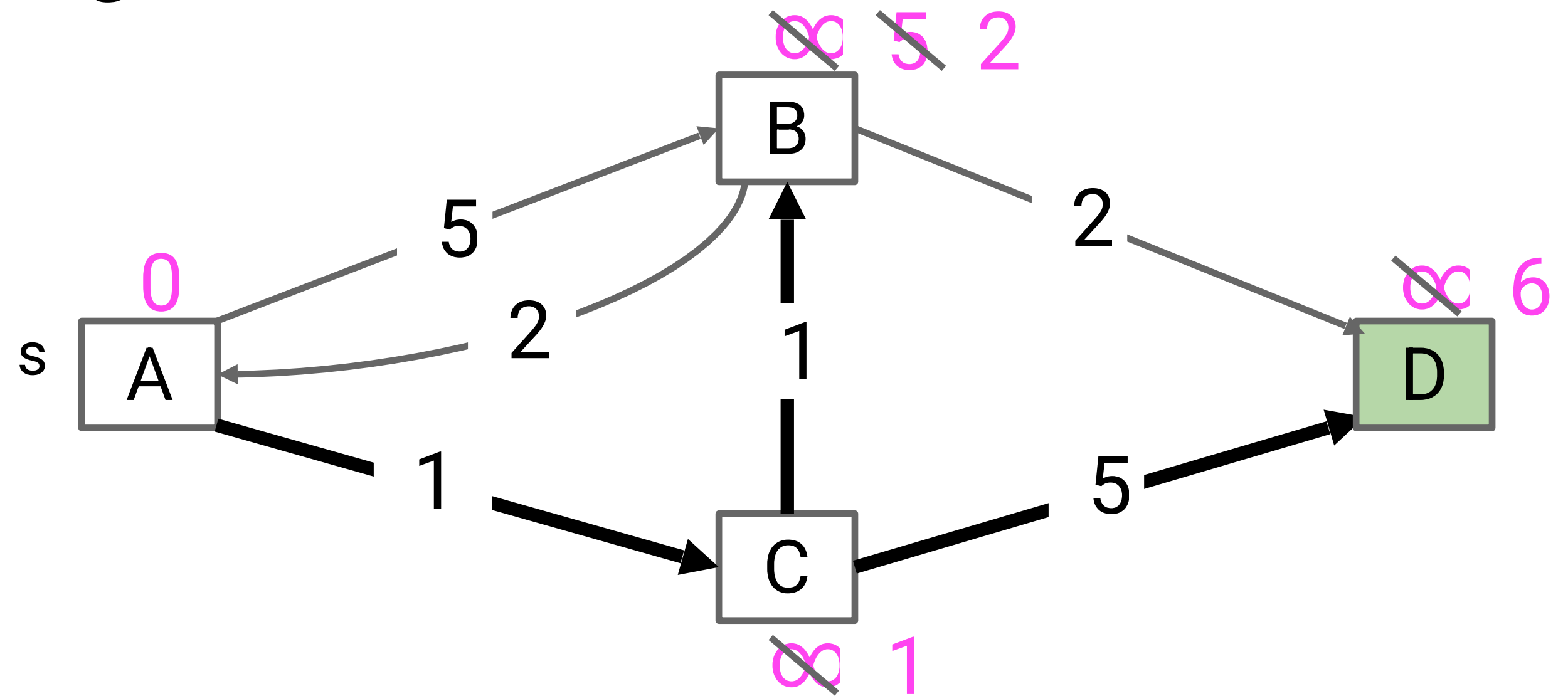
# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

**Remove the closest vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.



Fringe: [~~A=0~~, ~~C=1~~, B=2, D=6]

Removed vertex: B

# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

Remove the closest vertex from the fringe and mark it.

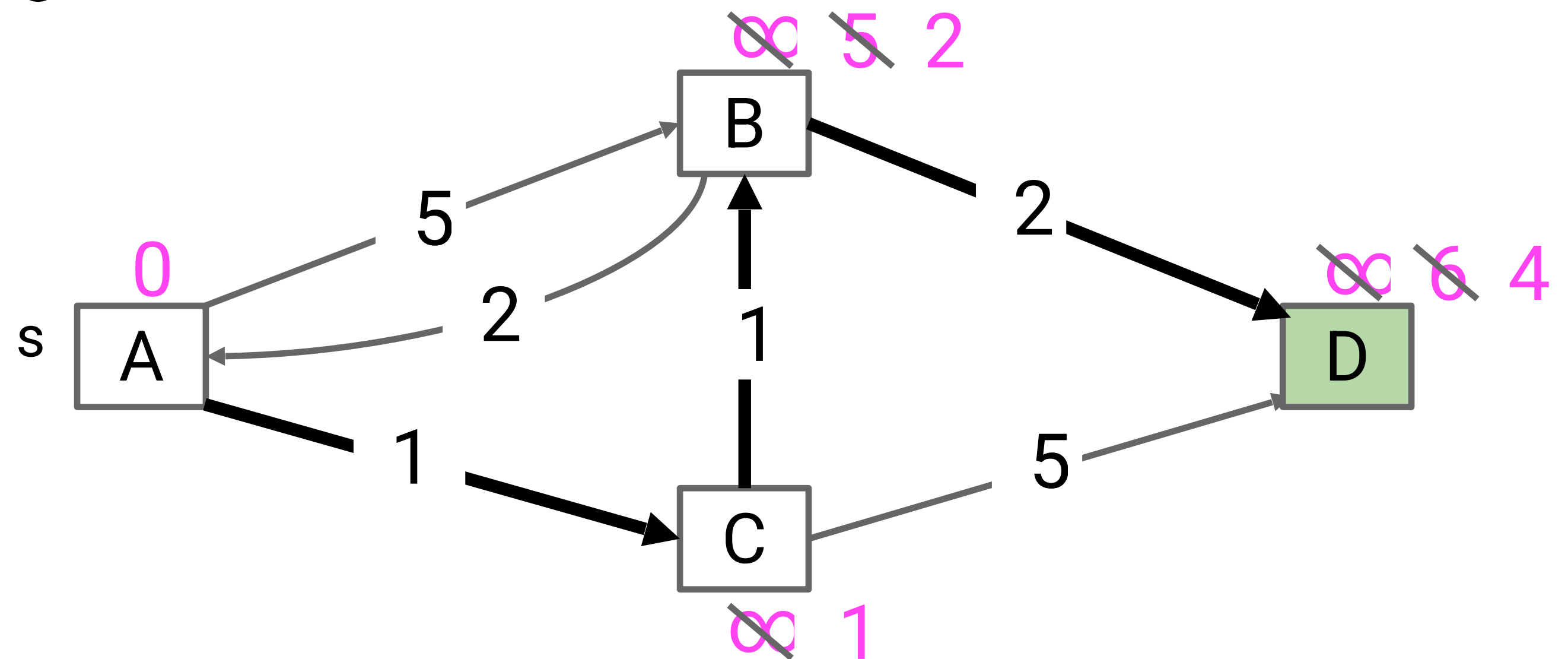
**For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.**

$B \rightarrow A$  (total=4) is not better than the best known way to A (0).

$B \rightarrow D$  (total=4) is better than the best known way to D (6, via  $C \rightarrow D$ ).  
So, we'll update the path to D.

Fringe: [A=0, C=1, B=2, D=4]

Removed vertex: B



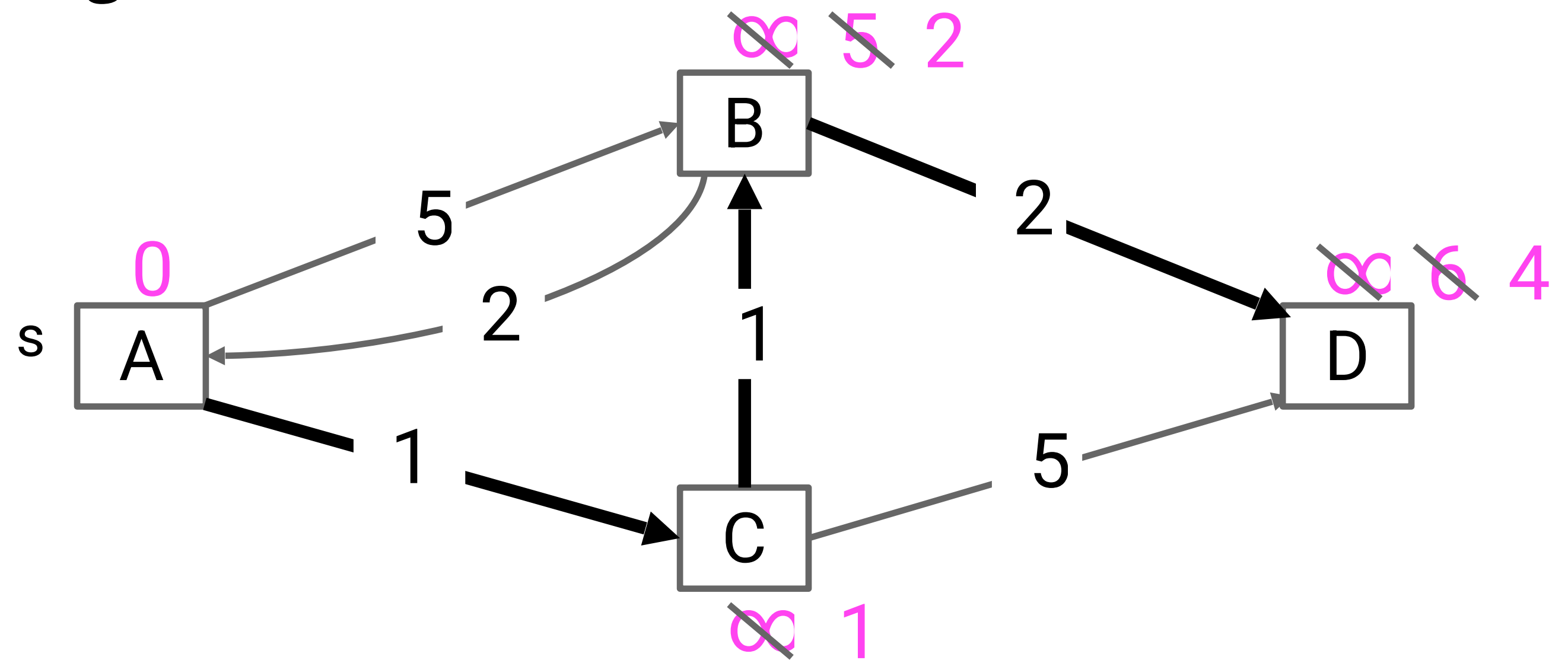
# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

**Remove the closest vertex from the fringe and mark it.**

For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.



Fringe: [~~A=0~~, ~~C=1~~, B=2, D=4]

Removed vertex: D



# Finding a Shortest Paths Tree Algorithmically

Add all vertices to the fringe.

While fringe is not empty:

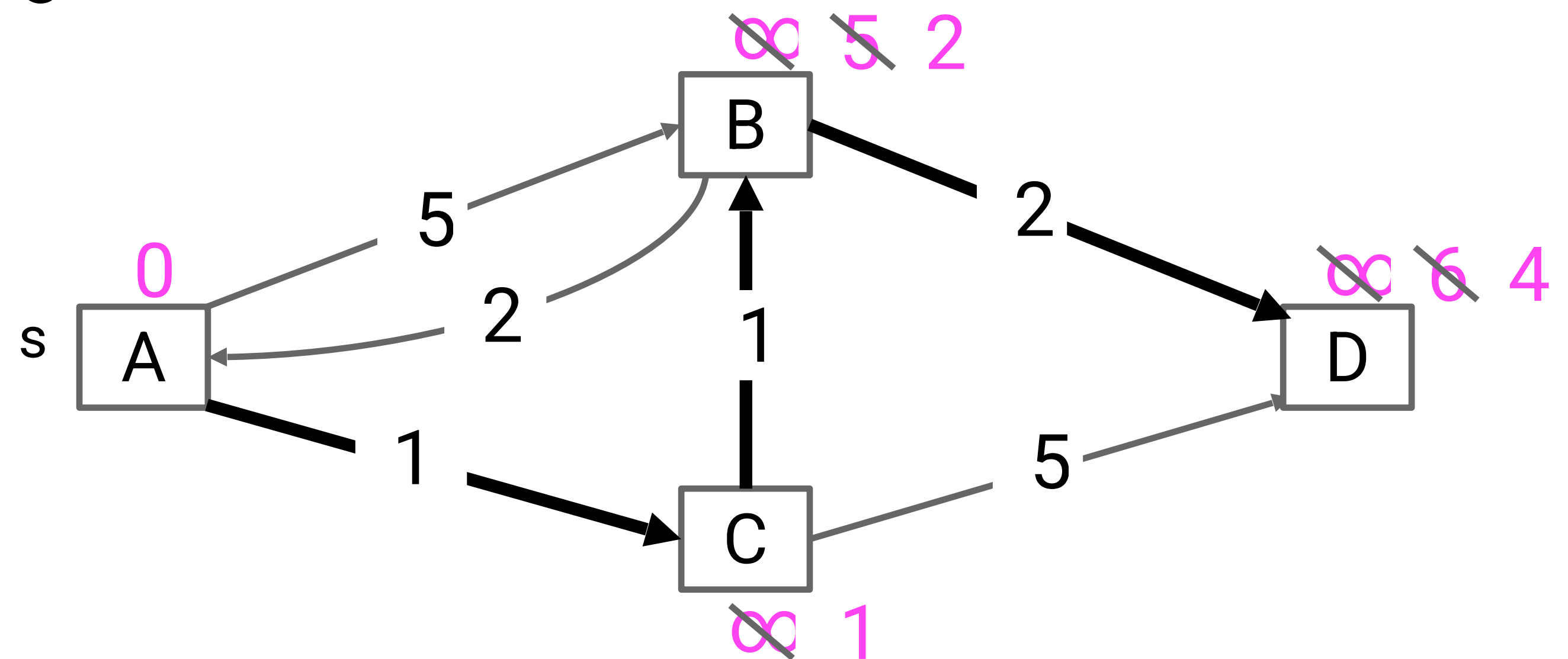
Remove the closest vertex from the fringe and mark it.

**For each outgoing edge  $v \rightarrow w$ : if the edge gives a better distance to  $w$ , add the edge, and update  $w$  in the fringe.**

No outgoing edges from D, so do nothing.

Fringe: [~~A=0~~, ~~C=1~~, B=2, ~~D=4~~]

Removed vertex: D

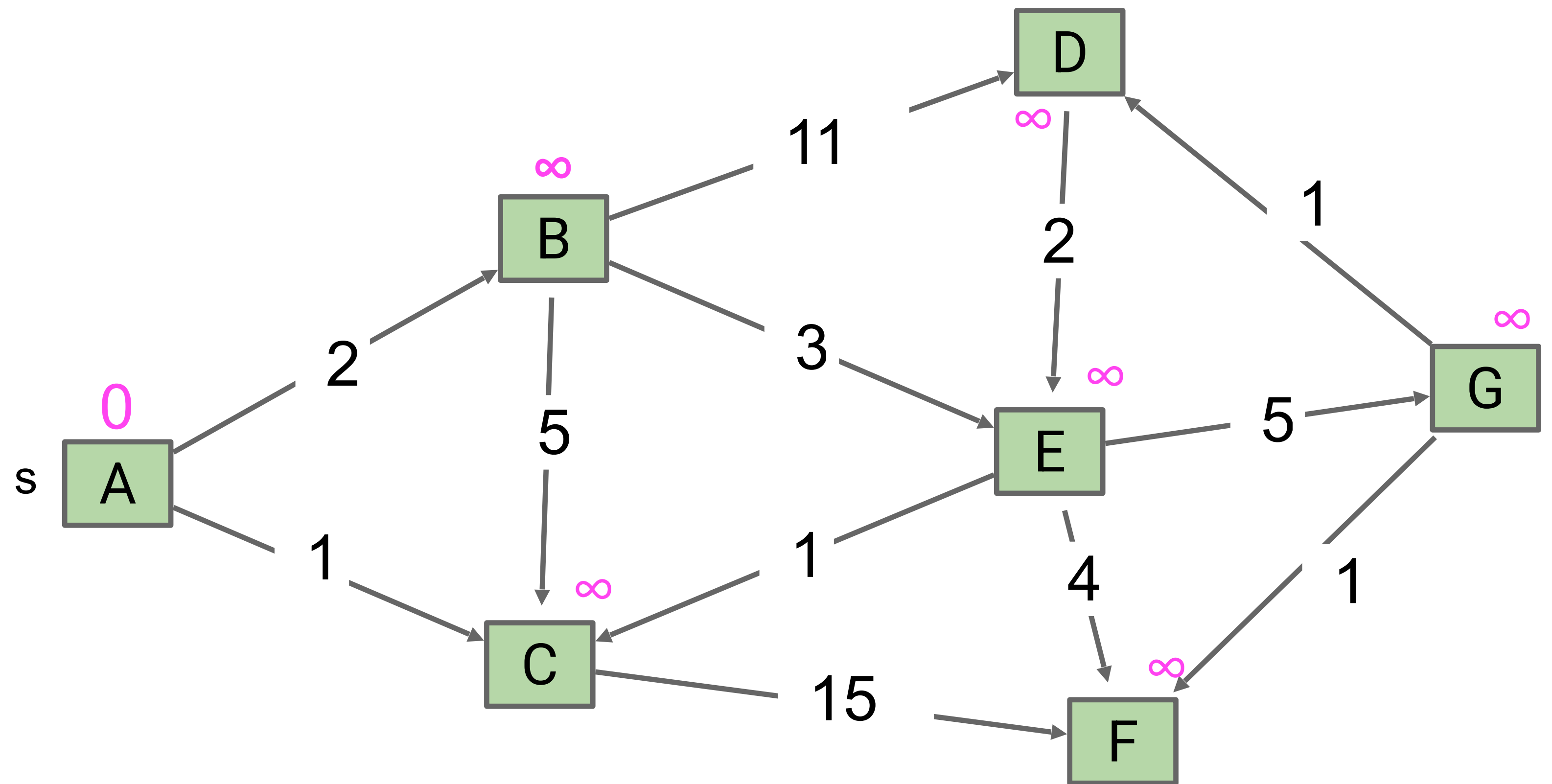


# Dijkstra's Algorithm

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

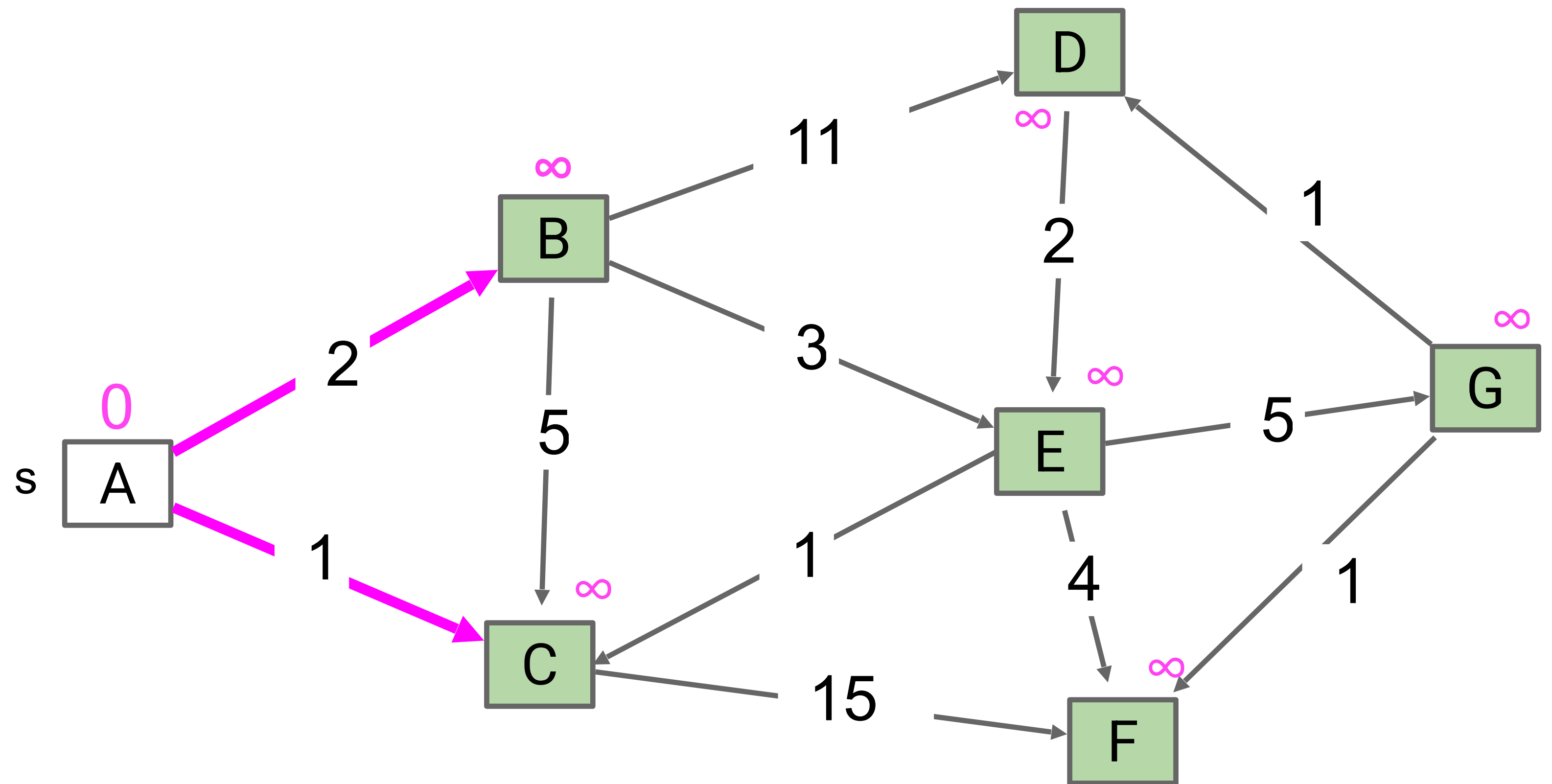


# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	$\infty$	-
C	$\infty$	-
D	$\infty$	-
E	$\infty$	-
F	$\infty$	-
G	$\infty$	-



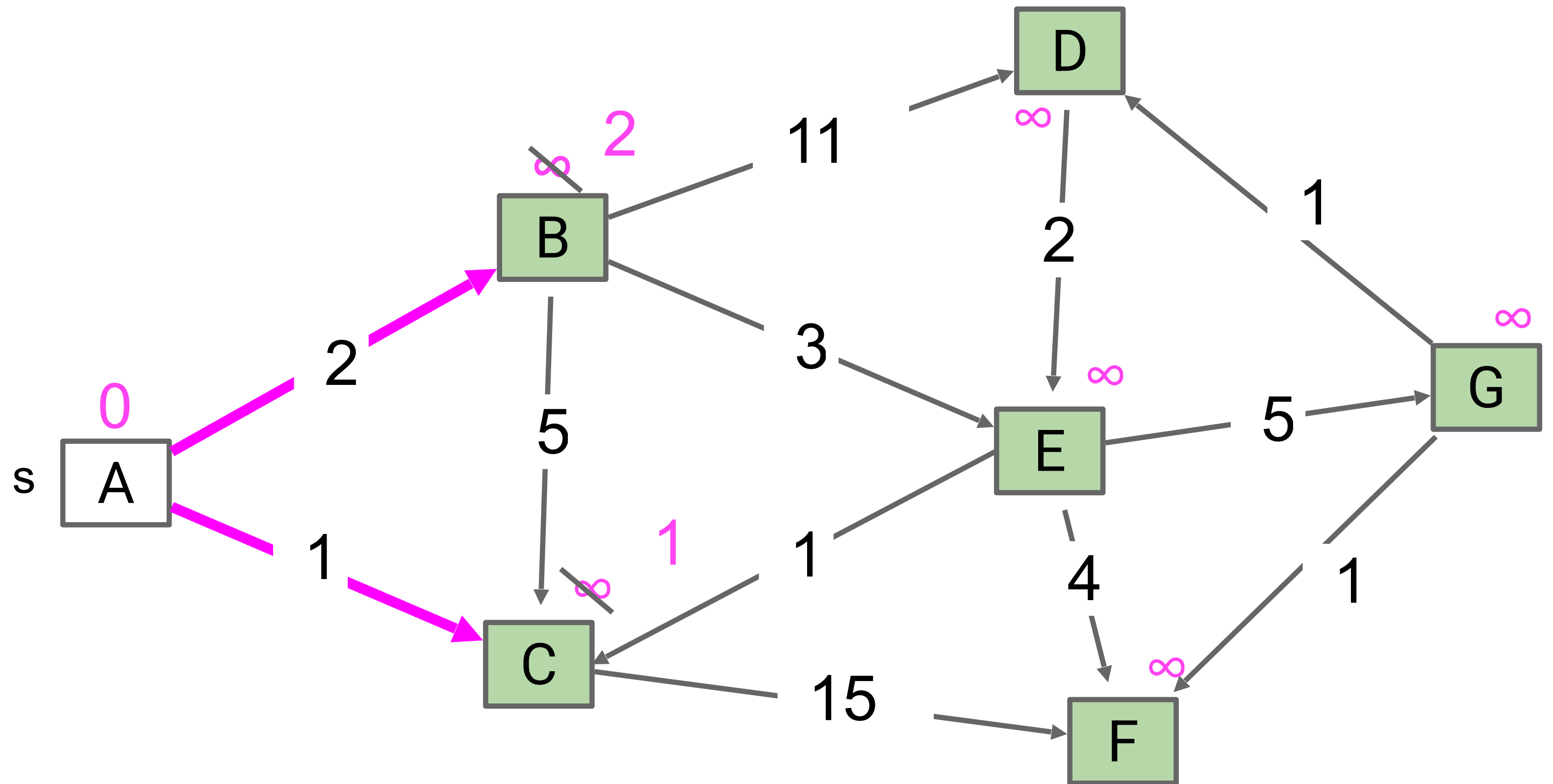
Fringe:  $[(B: \infty), (C: \infty), (D: \infty), (E: \infty), (F: \infty), (G: \infty)]$

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	$\infty$	-
E	$\infty$	-
F	$\infty$	-
G	$\infty$	-



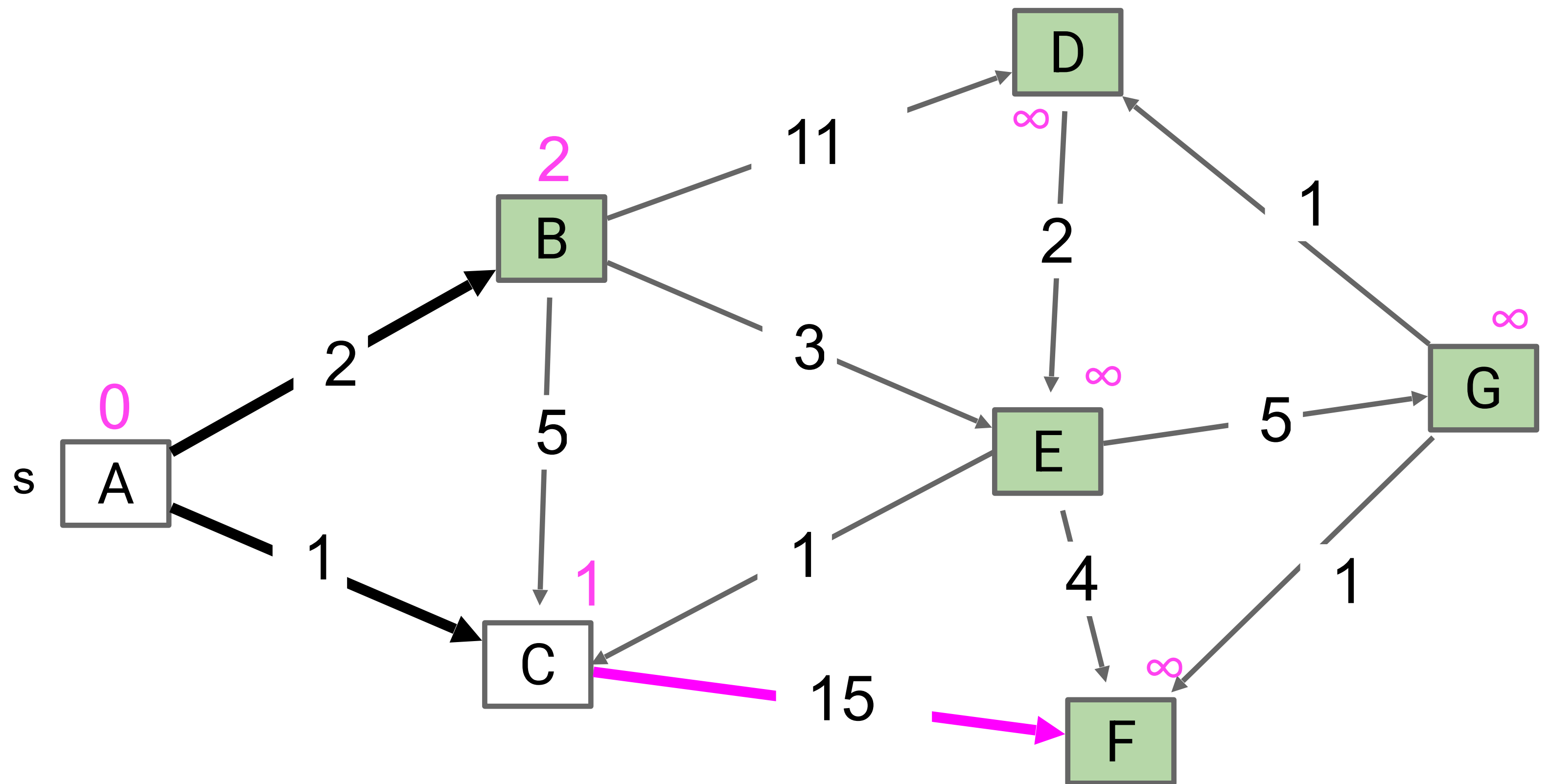
Fringe: [(C: 1), (B: 2)], (D:  $\infty$ ), (E:  $\infty$ ), (F:  $\infty$ ), (G:  $\infty$ )

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	$\infty$	-
E	$\infty$	-
F	$\infty$	-
G	$\infty$	-



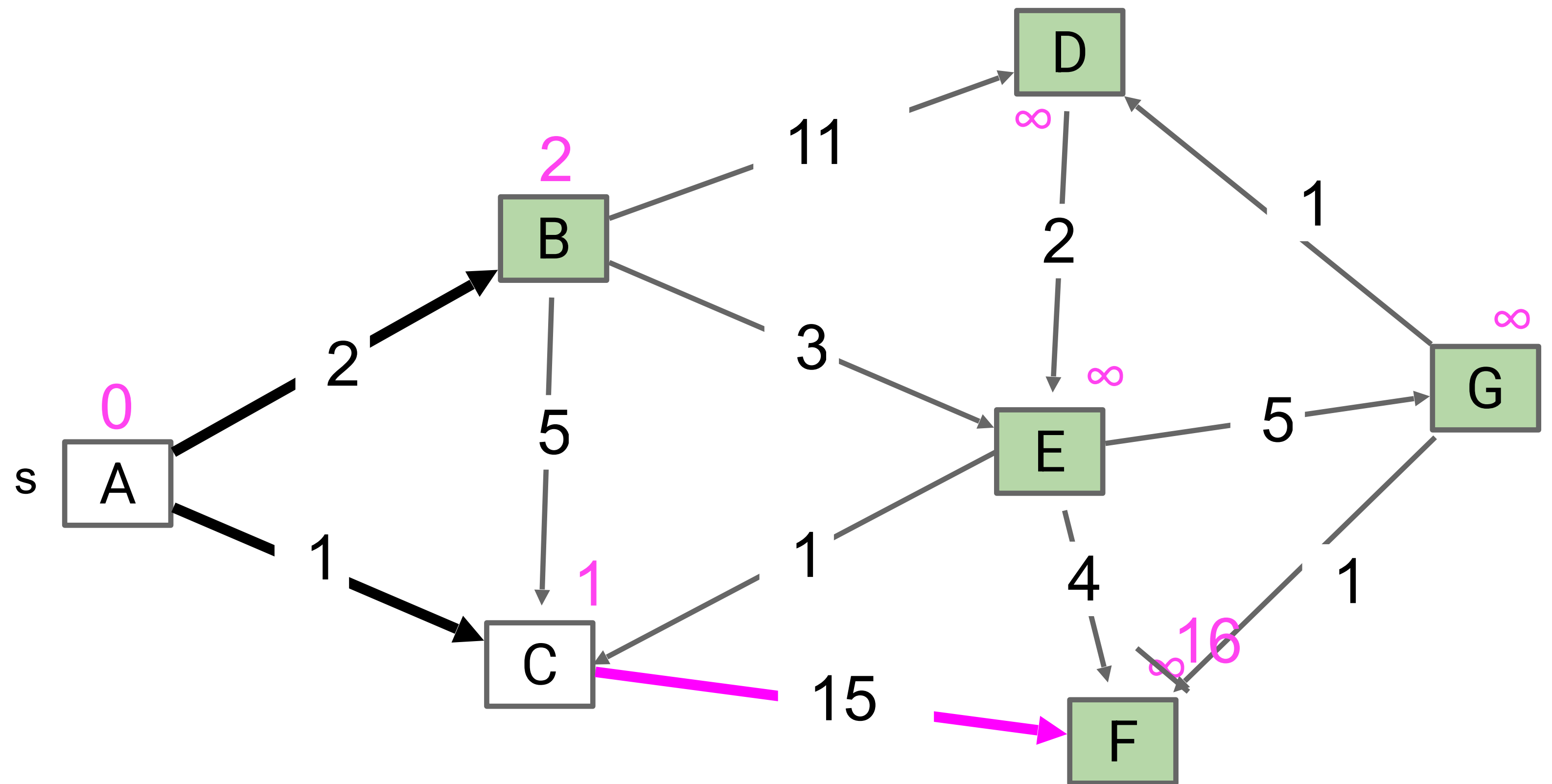
Fringe:  $[(B: 2), (D: \infty), (E: \infty), (F: \infty), (G: \infty)]$

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	$\infty$	-
E	$\infty$	-
F	16	C
G	$\infty$	-



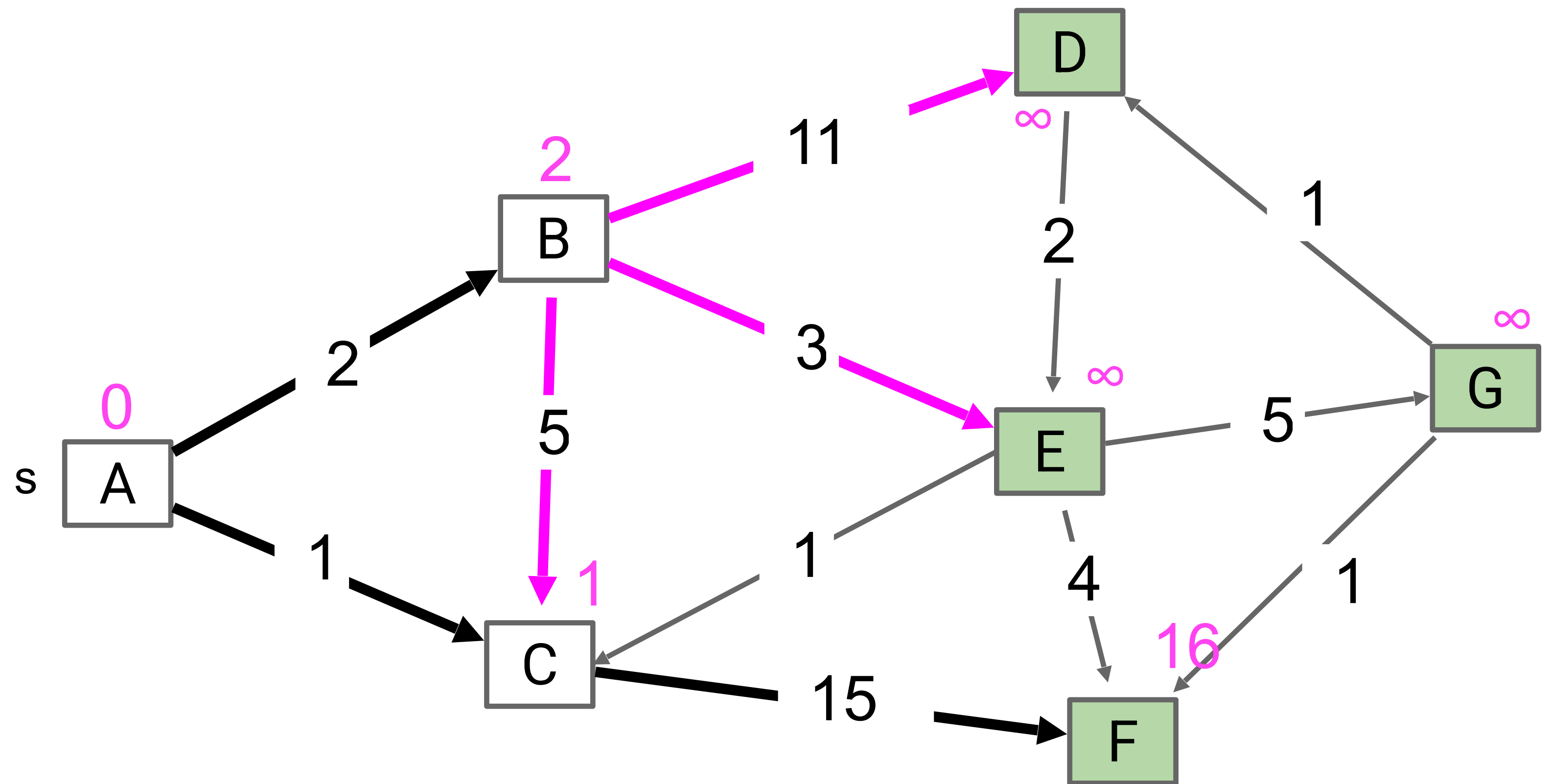
Fringe: [ (B: 2), (F: 16), (D:  $\infty$ ), (E:  $\infty$ ), (G:  $\infty$ ) ]

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	$\infty$	-
E	$\infty$	-
F	16	C
G	$\infty$	-



Fringe:  $[(F: 16), (D: \infty), (E: \infty), (G: \infty)]$

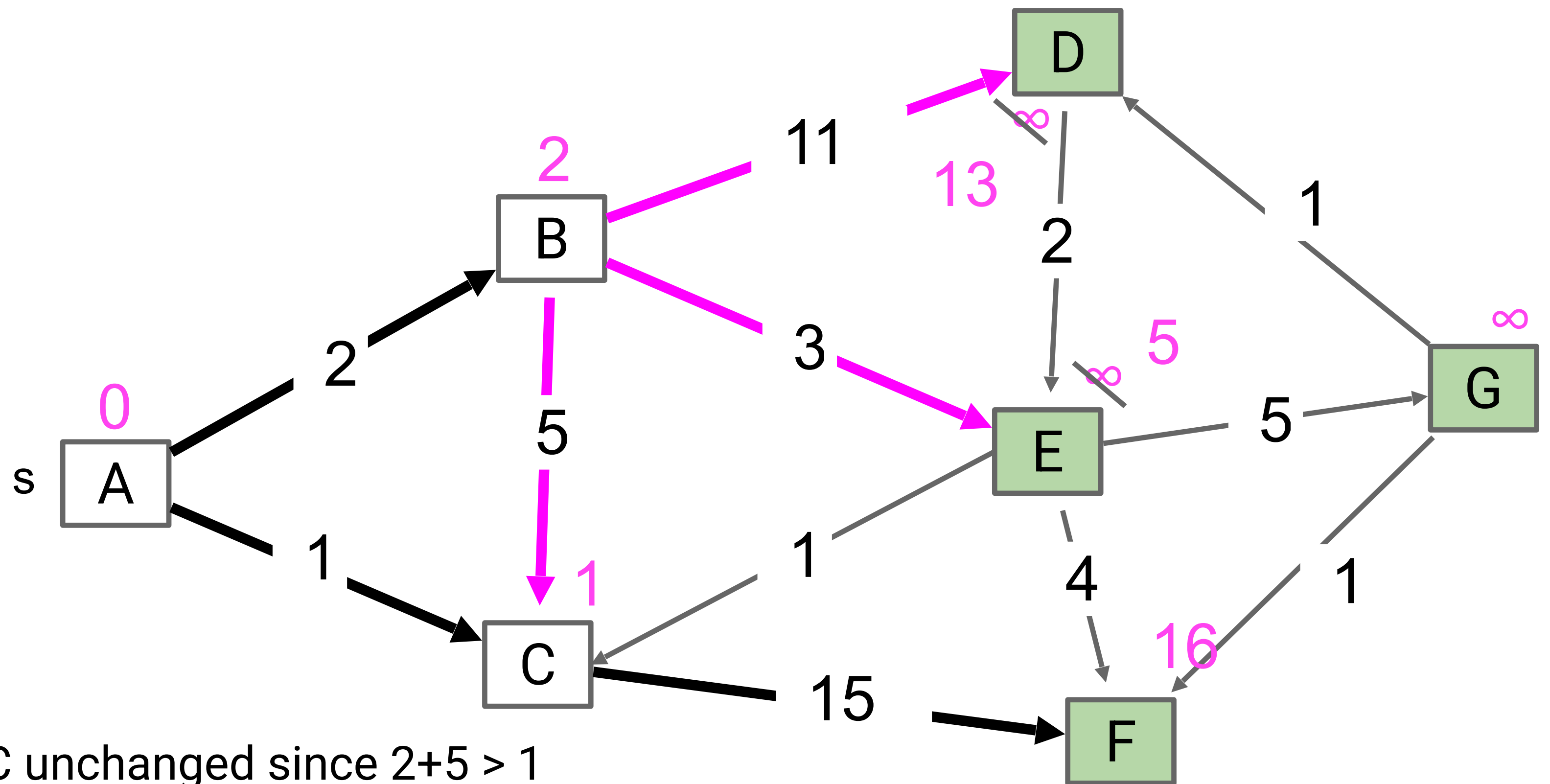


# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	13	B
E	5	B
F	16	C
G	$\infty$	-



Vertex C unchanged since  $2+5 > 1$

Fringe: [ (E: 5), (D: 13) ], (F: 16), (G:  $\infty$ ) ]

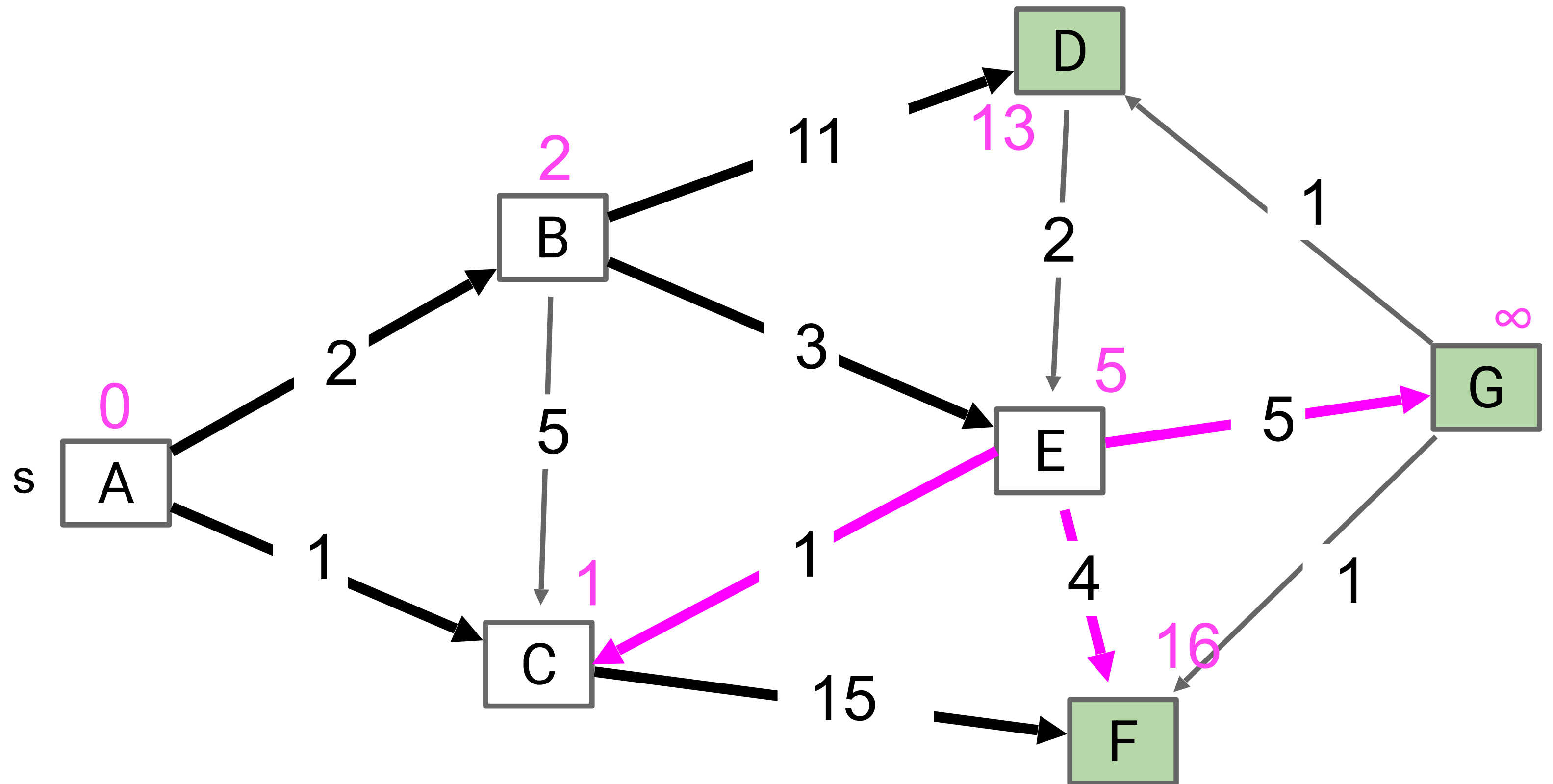
Which vertex is removed next?

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	13	B
E	5	B
F	16	C
G	$\infty$	-



Fringe: [ (D: 13), (F: 16), (G:  $\infty$ ) ]

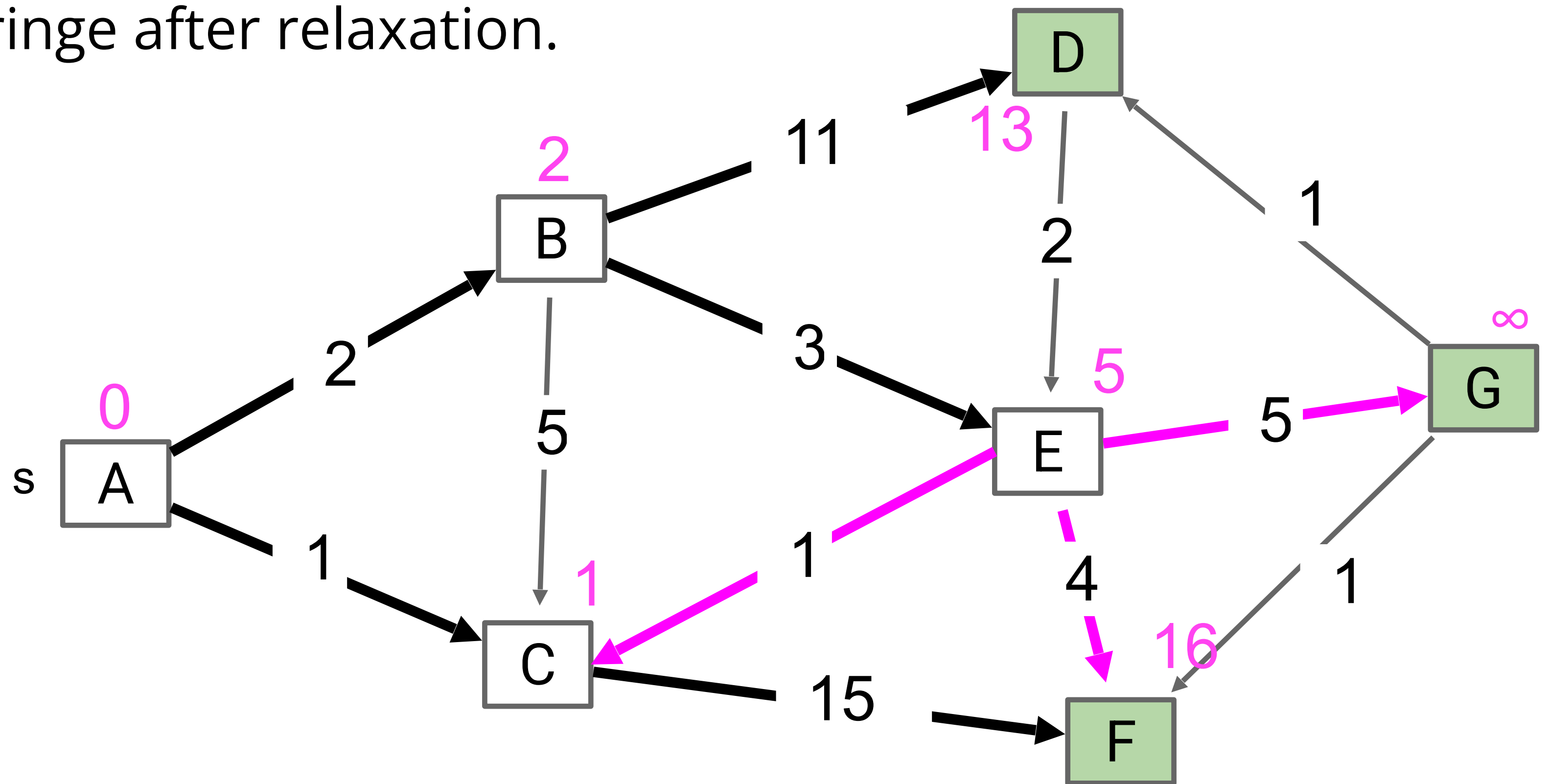
# Worksheet time!

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

- Show `distTo`, `edgeTo`, and fringe after relaxation.

Node	<code>distTo</code>	<code>edgeTo</code>
A	0	-
B	2	A
C	1	A
D	13	B
E	5	B
F	16	C
G	$\infty$	-



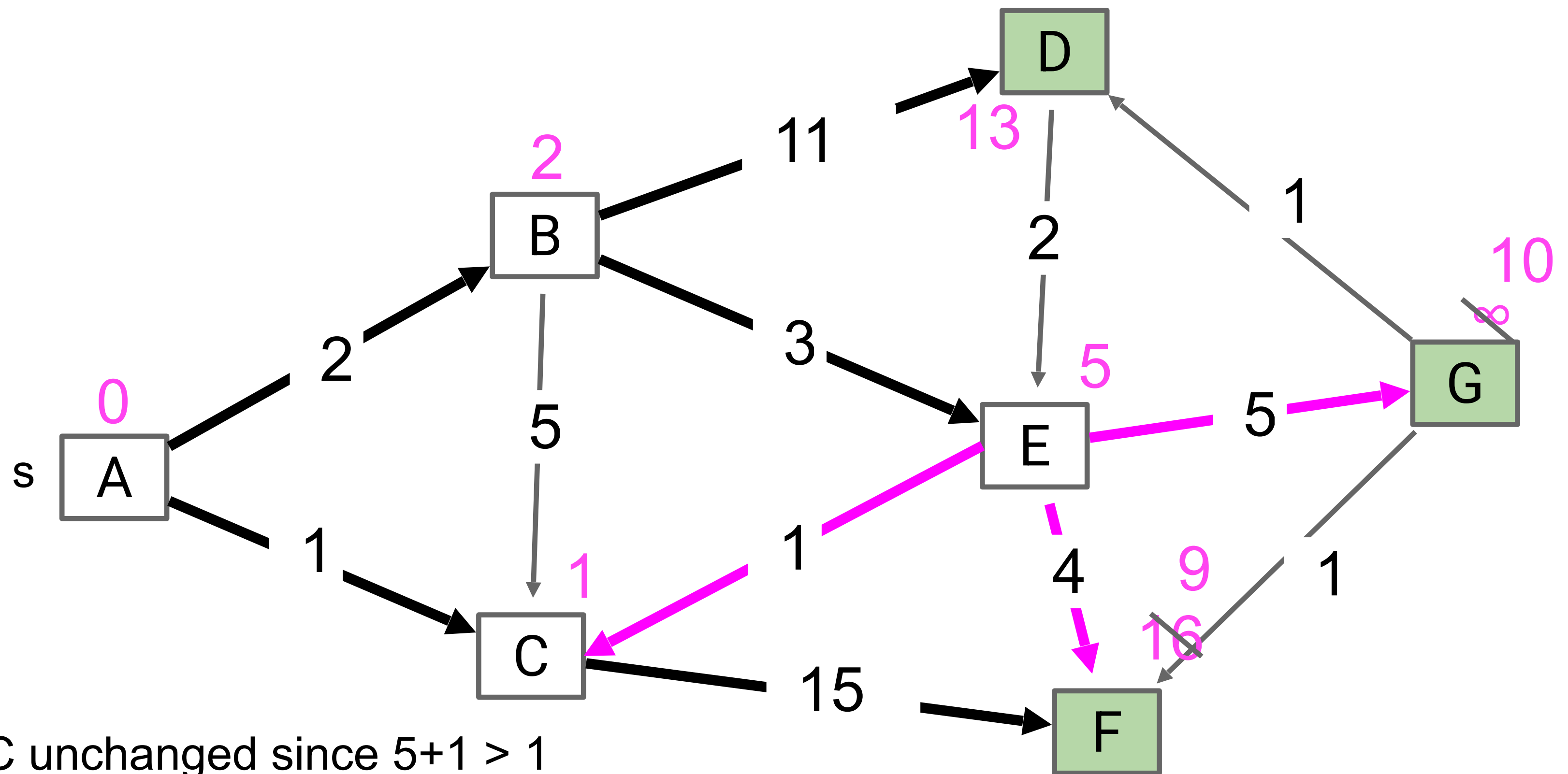
Fringe:  $[(D: 13), (F: 16), (G: \infty)]$

# Worksheet answers

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	13	B
E	5	B
F	9	E
G	10	E



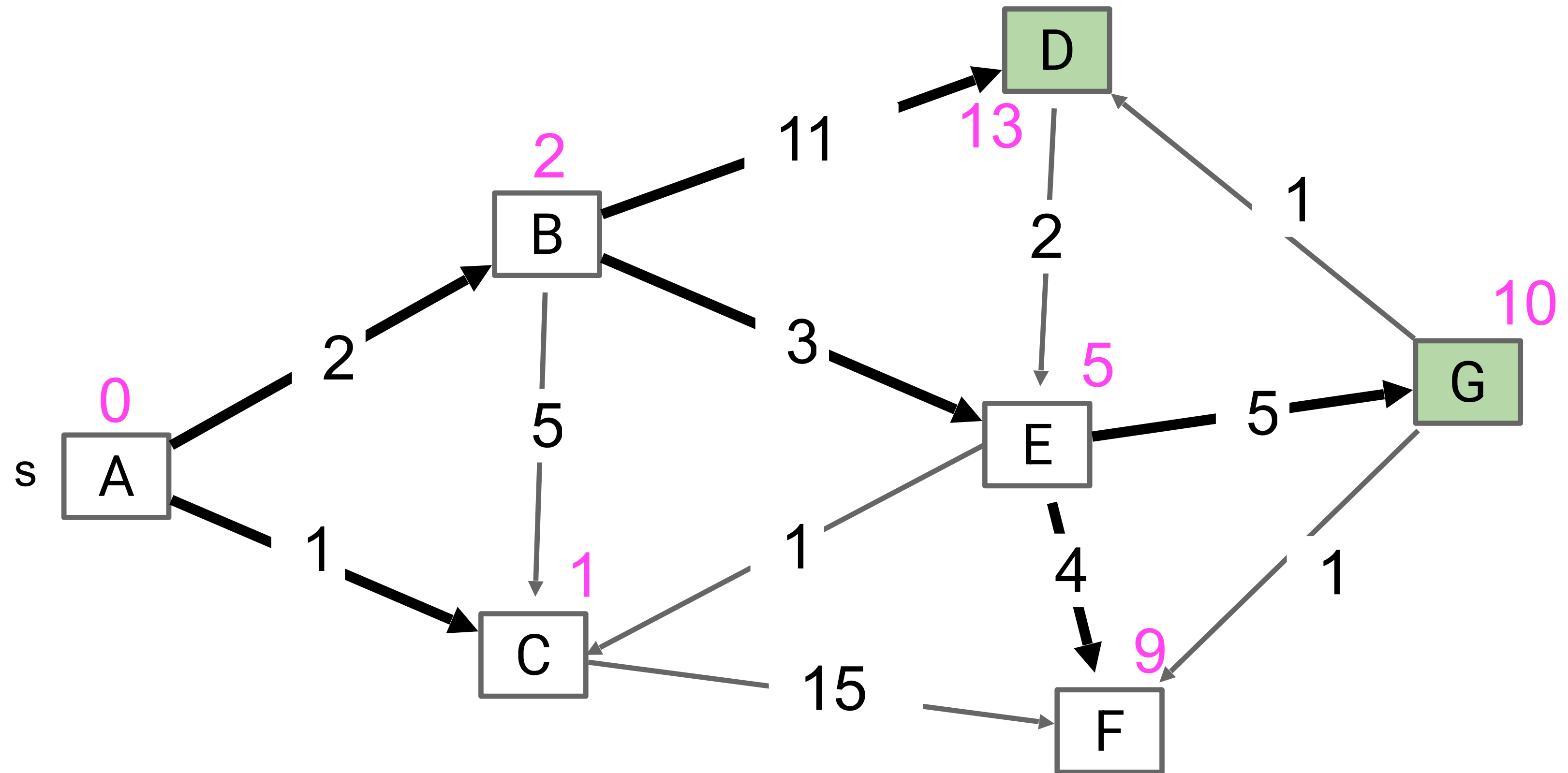
Fringe: [(F: 9), (G: 10), (D: 13)]

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	13	B
E	5	B
F	9	E
G	10	E



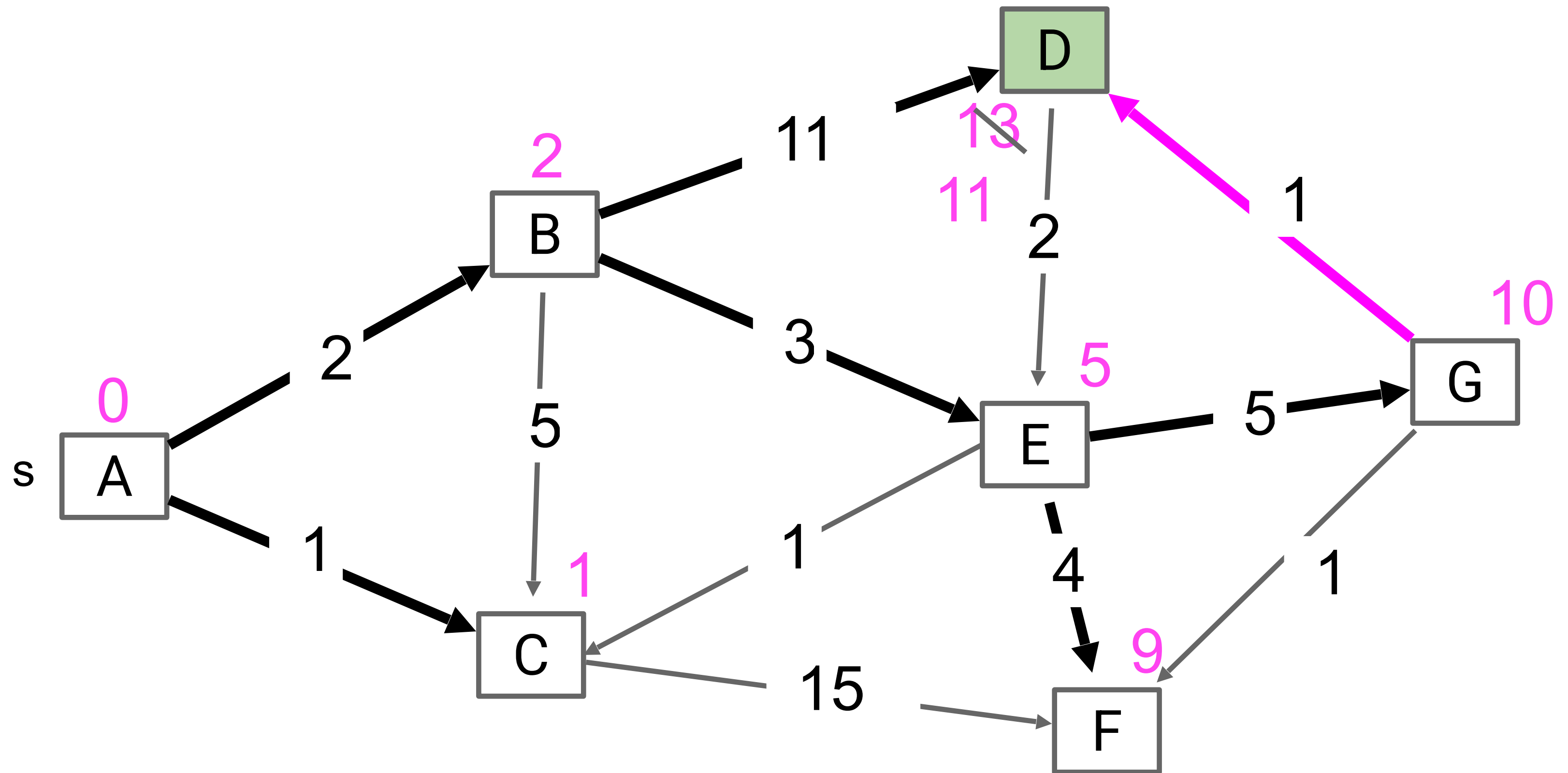
Fringe:  $[(G: 10), (D: 13)]$

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	11	G
E	5	B
F	9	E
G	10	E



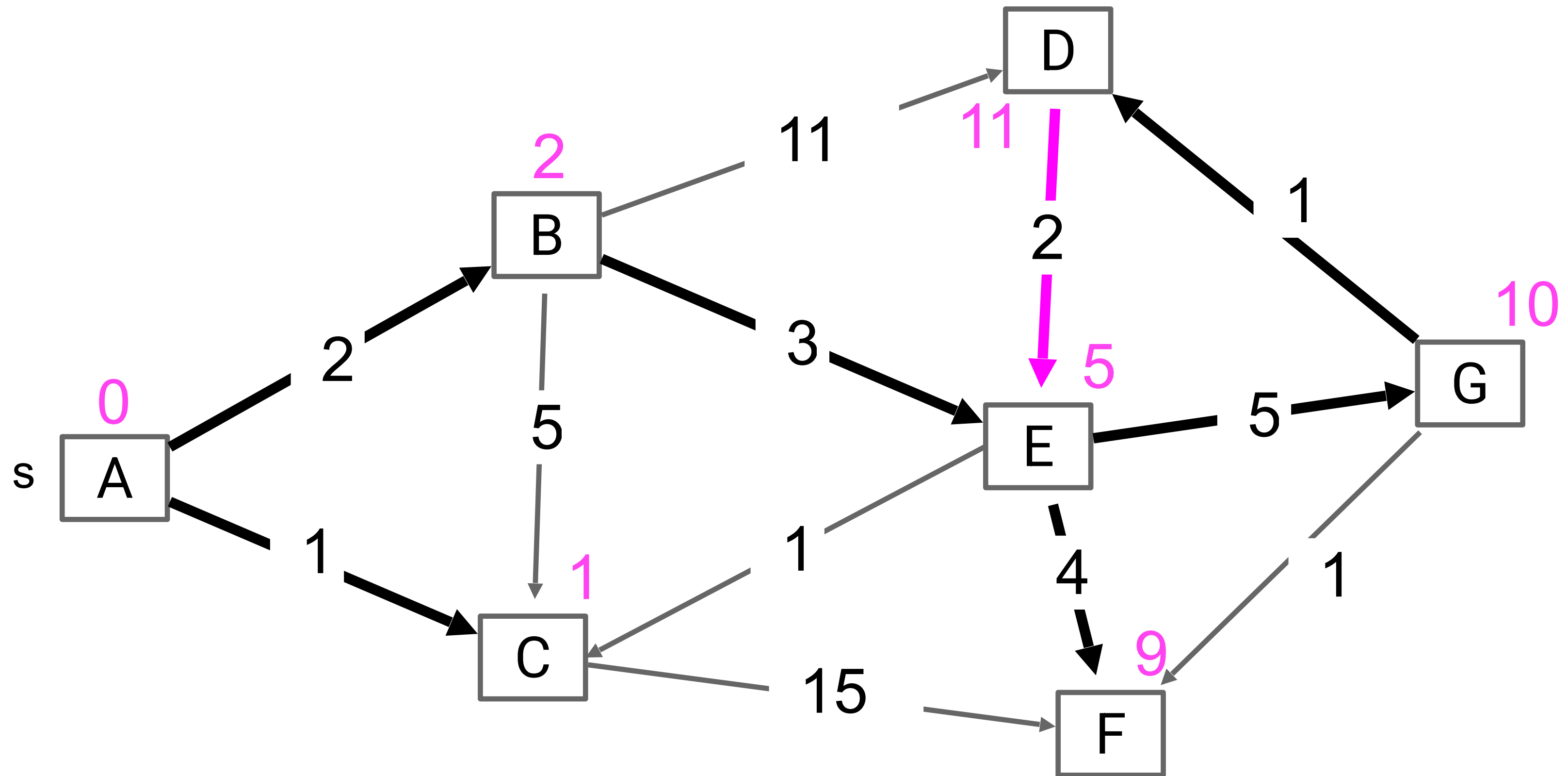
Fringe: (D: 11)

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	11	G
E	5	B
F	9	E
G	10	E



Vertex E unchanged since  $11 + 2 > 5$

Fringe: [ ]

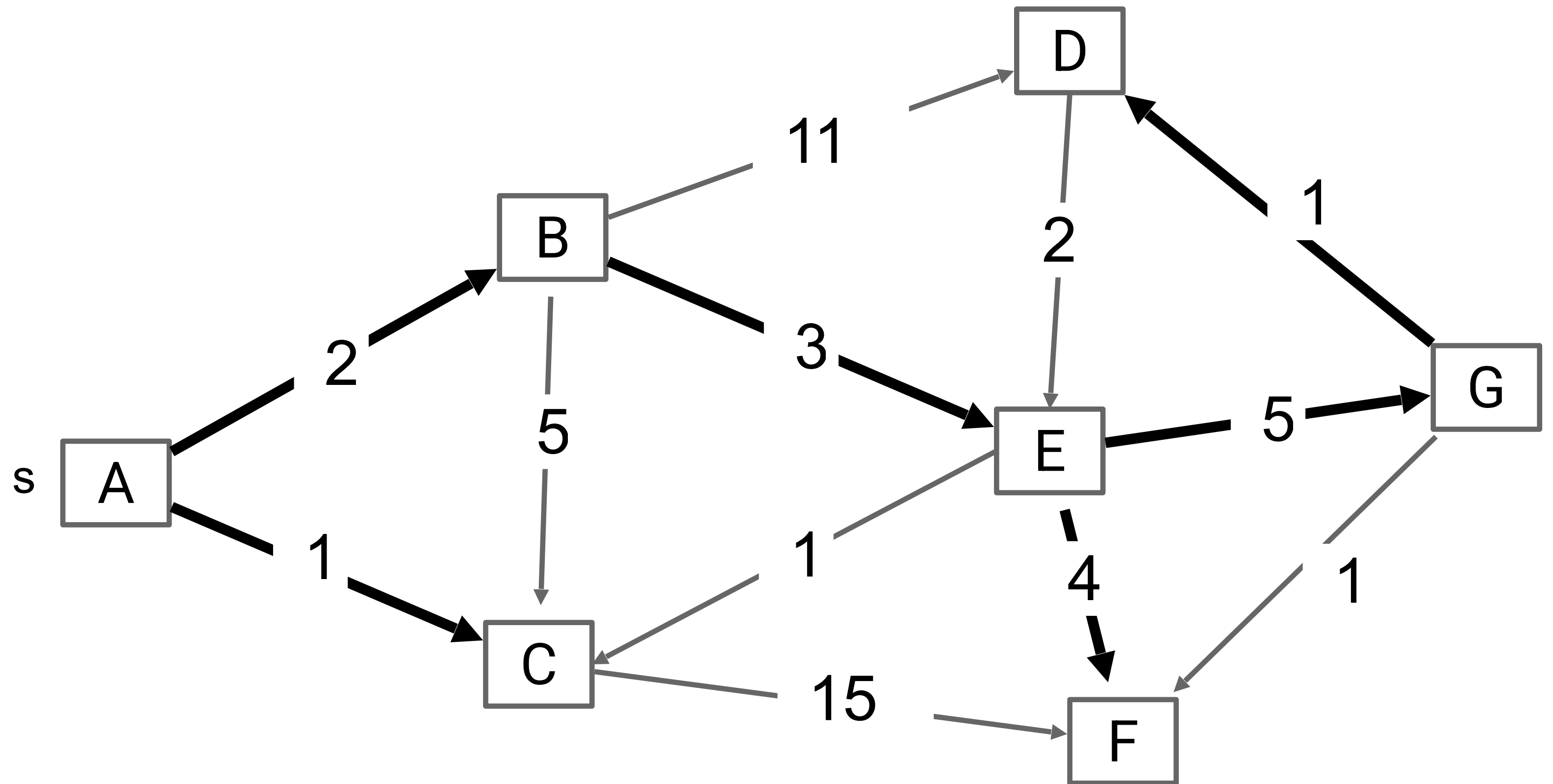
Note: If non-negative weights, **impossible for any inactive vertex** (white, not on fringe) **to be improved!**

# Dijkstra's Demo

Insert all vertices into fringe PQ (priority queue), storing vertices in order of distance from source.

Repeat: Remove (closest) vertex  $v$  from PQ, and relax all edges pointing from  $v$ .

Node	distTo	edgeTo
A	0	-
B	2	A
C	1	A
D	11	G
E	5	B
F	9	E
G	10	E

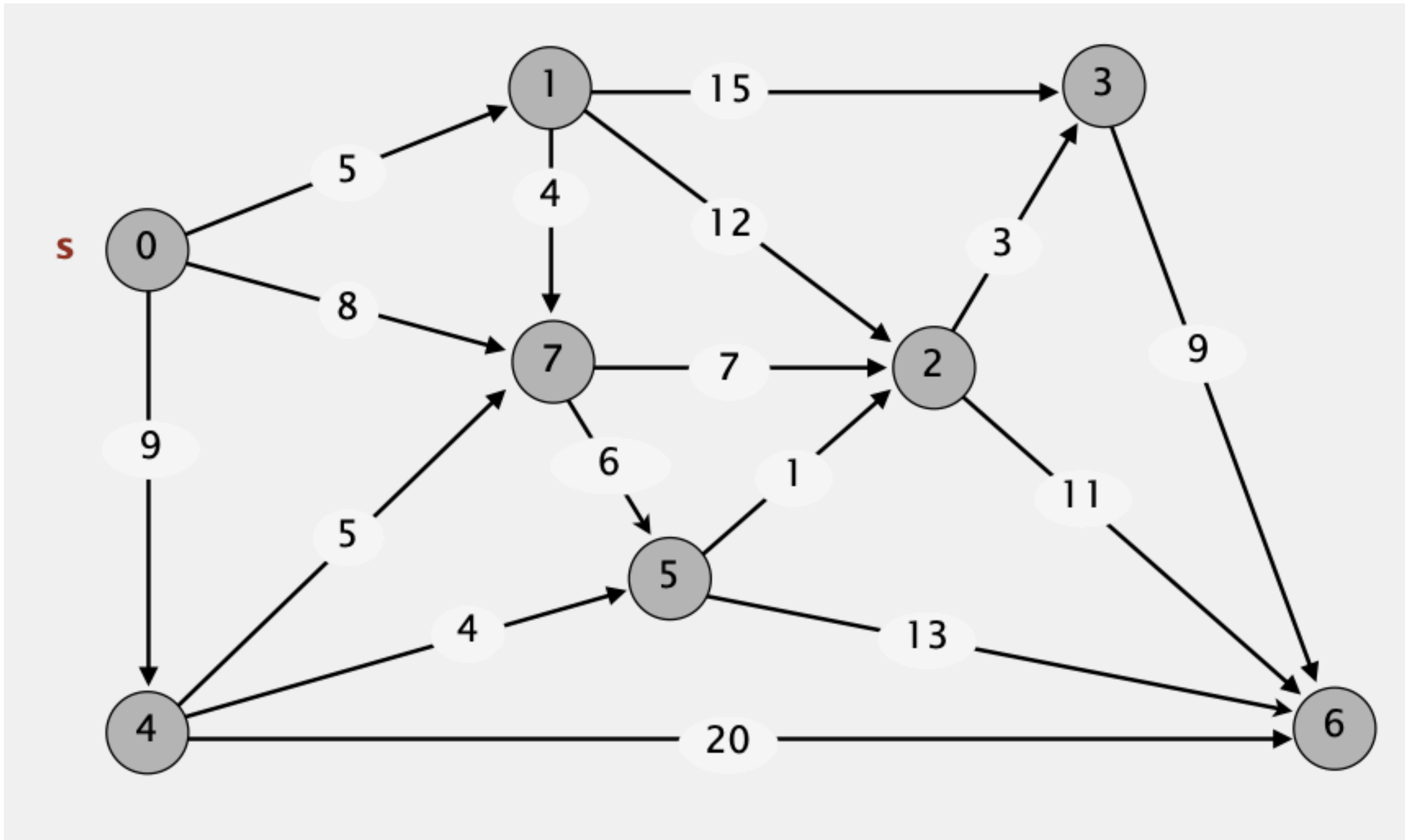


Fringe: [ ]



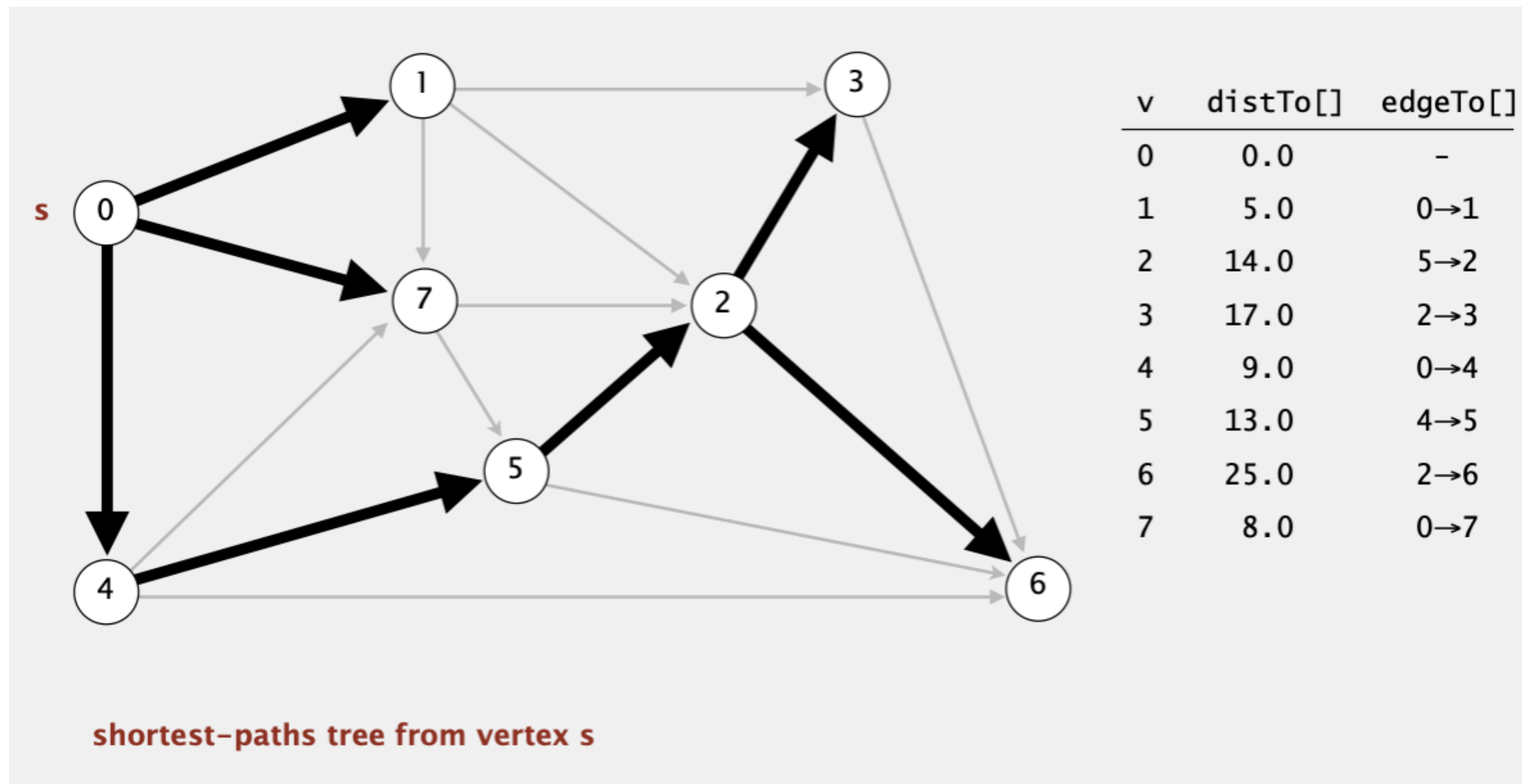
# Worksheet time!

- Run Dijkstra's algorithm to generate the shortest path tree from s below.



# Worksheet answers!

- Run Dijkstra's algorithm to generate the shortest path tree from  $s$  below.
- For a full walkthrough, see the slides in the appendix



# Dijkstra's Implementation

# Dijkstra's Algorithm Pseudocode

## Dijkstra's:

- `PQ.add(source, 0)`
- For other vertices  $v$ , `PQ.add(v, infinity)`
- While PQ is not empty:
  - $p = \text{PQ.removeSmallest}()$
  - Relax all edges from  $p$

## Relaxing an edge $p \rightarrow q$ with weight $w$ :

- If  $\text{distTo}[p] + w < \text{distTo}[q]$ :
  - $\text{distTo}[q] = \text{distTo}[p] + w$
  - $\text{edgeTo}[q] = p$
  - `PQ.changePriority(q, distTo[q])`

## Key invariants:

- $\text{edgeTo}[v]$  is the best known predecessor of  $v$ .
- $\text{distTo}[v]$  is the best known total distance from source to  $v$ .
- PQ contains all unvisited vertices in order of  $\text{distTo}$ .

## Important properties:

- Always visits vertices in order of total distance from source.
- Relaxation always fails on edges to already visited vertices.

# Framework for shortest-paths algorithm

```
public class DijkstraSP {
    private double[] distTo;           // distTo[v] = distance of shortest s->v path
    private DirectedEdge[] edgeTo;    // edgeTo[v] = last edge on shortest s->v path
    private IndexMinPQ<Double> pq;    // priority queue of vertices

    public DijkstraSP(EdgeWeightedDigraph G, int s) {
        distTo = new double[G.V()];
        edgeTo = new DirectedEdge[G.V()];

        for (int v = 0; v < G.V(); v++)
            distTo[v] = Double.POSITIVE_INFINITY;
        distTo[s] = 0.0;

        // relax vertices in order of distance from s
        pq = new IndexMinPQ<Double>(G.V());
        pq.insert(s, distTo[s]);
        while (!pq.isEmpty()) {
            int v = pq.delMin();
            for (DirectedEdge e : G.adj(v))
                relax(e);
        }

        // relax edge e and update pq if changed
        private void relax(DirectedEdge e) {
            int v = e.from(), w = e.to();
            if (distTo[w] > distTo[v] + e.weight()) {
                distTo[w] = distTo[v] + e.weight();
                edgeTo[w] = e;
                if (pq.contains(w)) pq.decreaseKey(w, distTo[w]);
                else pq.insert(w, distTo[w]);
            }
        }
    }
}
```

# Dijkstra's Algorithm Runtime

Priority Queue operation count, assuming min-binary heap based PQ:

- add: max  $V$  times, each costing  $O(\log V)$  time.
- removeSmallest: max  $V$  times, each costing  $O(\log V)$  time.
- changePriority: max  $E$  times, each costing  $O(\log V)$  time.

Overall runtime:  $O(V \cdot \log(V) + V \cdot \log(V) + E \cdot \log V)$ .

- Assuming  $E > V$ , this is just  **$O(E \log V)$**  for a connected graph.

	# Operations	Cost per operation	Total cost
PQ add	$V$	$O(\log V)$	$O(V \log V)$
PQ removeSmallest	$V$	$O(\log V)$	$O(V \log V)$
PQ changePriority	$E$	$O(\log V)$	$O(E \log V)$

# Lecture 23 wrap-up

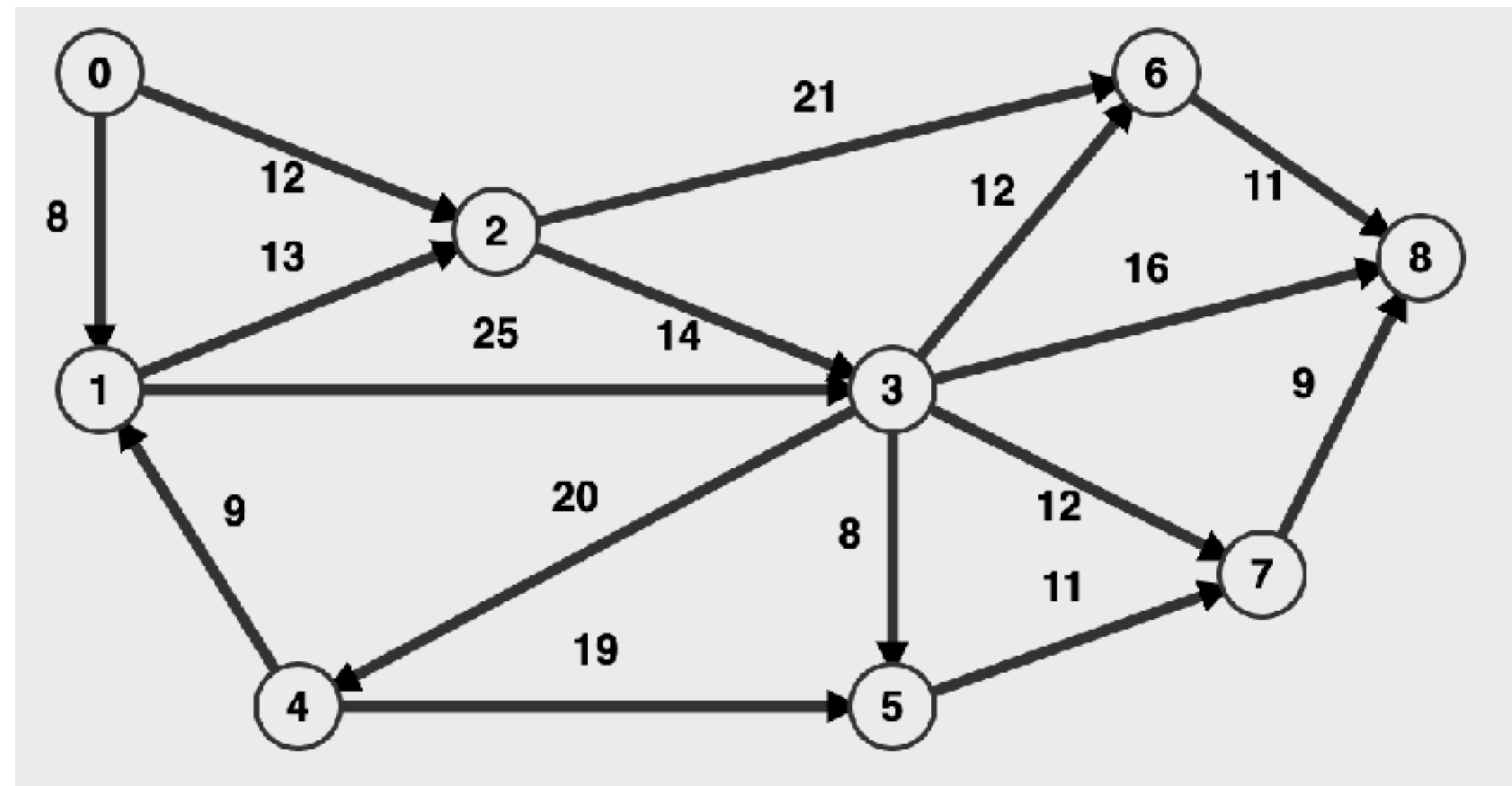
- HW9: Transplant Manager due 11:59pm
- Last(!) HW (10, text generator), quiz, & graded lab this week
- Next week: Zoom class
  - Tues: Course evals 2:45-3pm, 3-4pm Careers panel (Google, CS PhD, video game company)
  - Thurs: Final project pt 1 check-ins (more in lab tomorrow!), sign up for 10 min slot

## Resources

- Recommended Textbook: Chapter 4.4 (Pages 638-676)
- Website: <https://algs4.cs.princeton.edu/44sp/>
- Visualization: <https://visualgo.net/en/sssp>
- Practice problems behind this slide

# Problem 1

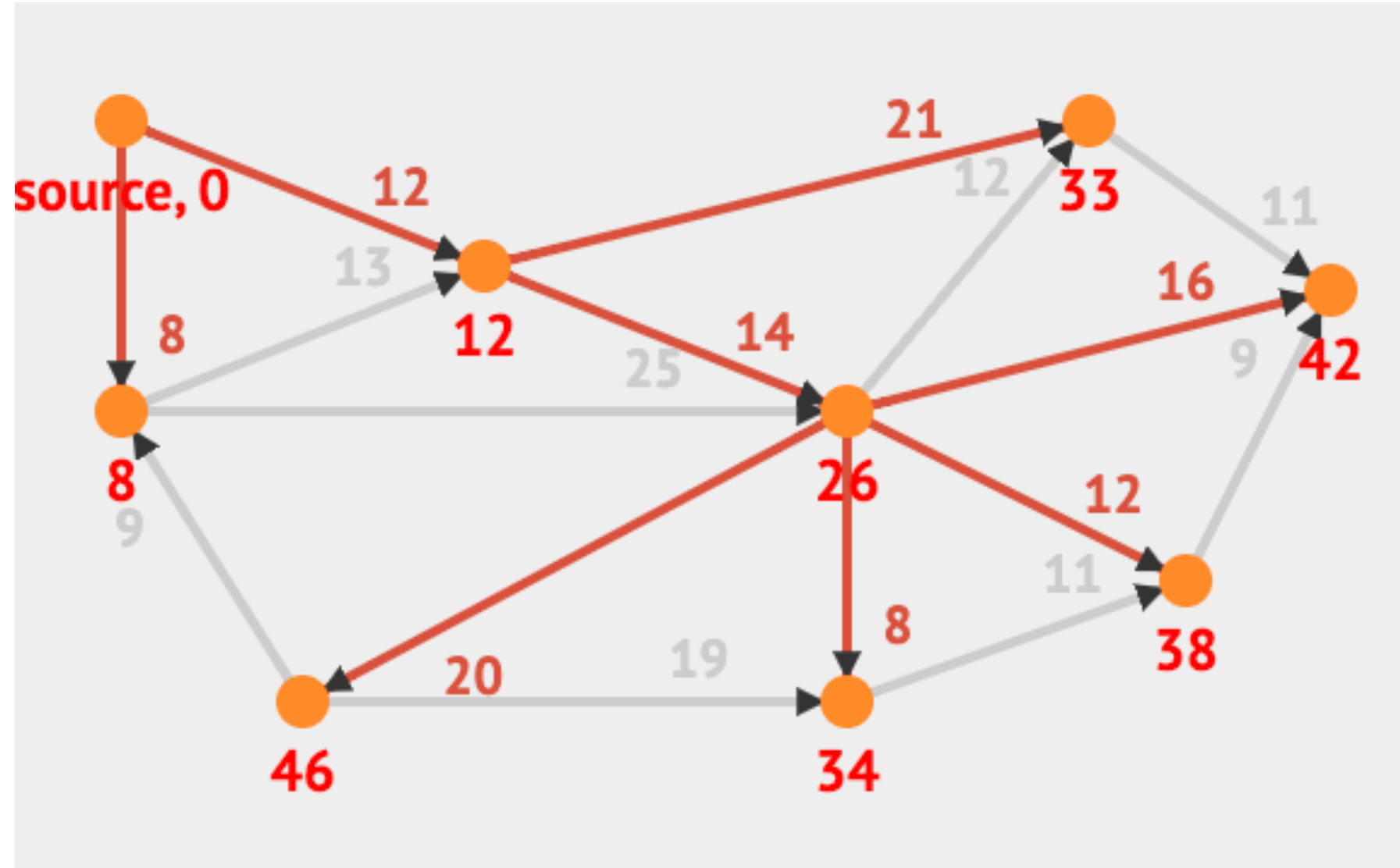
- Run Dijkstra's algorithm on the following graph with 0 being the starting vertex.





# Answer 1

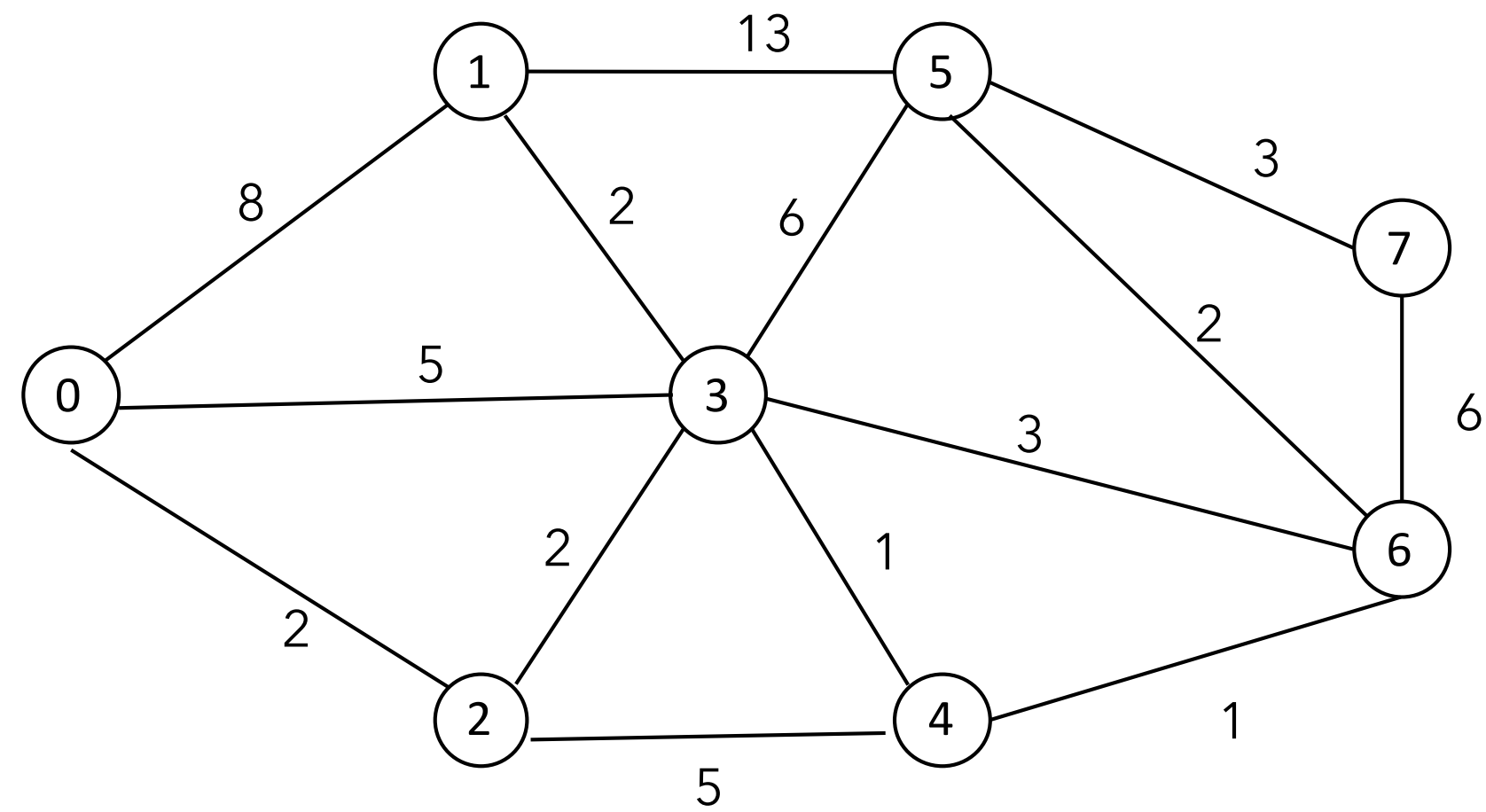
- Run Dijkstra's algorithm on the following graph with 0 being the starting vertex.



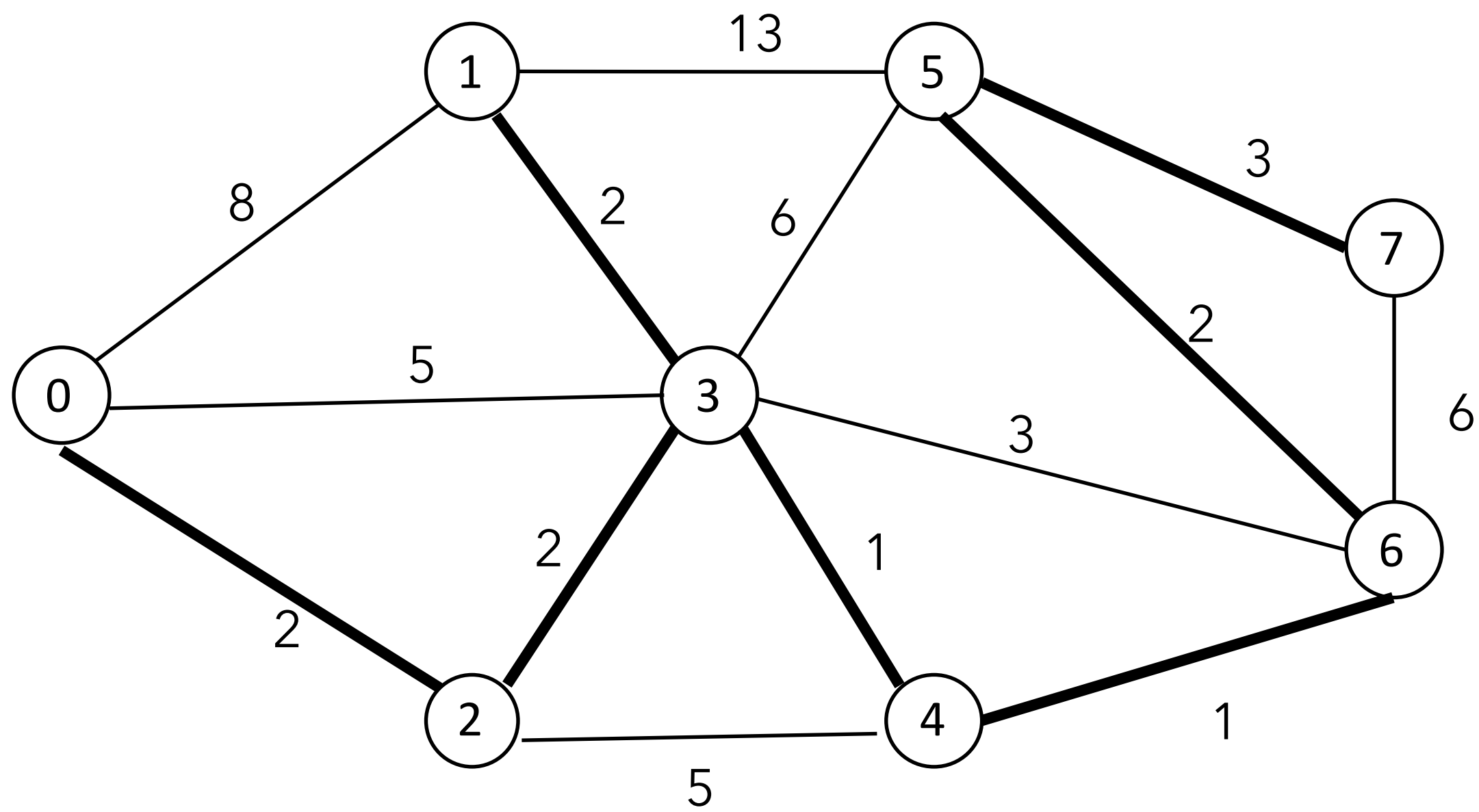
v	distTo[]	edgeTo[]
0	0	-
1	8	0->1
2	12	0->2
3	26	2->3
4	46	3->4
5	34	3->5
6	33	3->6
7	38	3->7
8	42	3->8

# Problem 2

Run Dijkstra's algorithm on the following graph with 0 being the starting vertex.



# Answer 2



v	distTo[]	edgeTo[]
0	0	-
1	6	3->1
2	2	0->2
3	4	2->3
4	5	3->4
5	8	6->5
6	6	4->6
7	11	5->7

# Problem 3

Dijkstra's algorithm is guaranteed to be optimal so long as there are no negative edges. Sketch a proof by induction proving why.

- Hint: The proof relies on the property that relaxation always fails on edges to visited (white) vertices.

# Answer 3

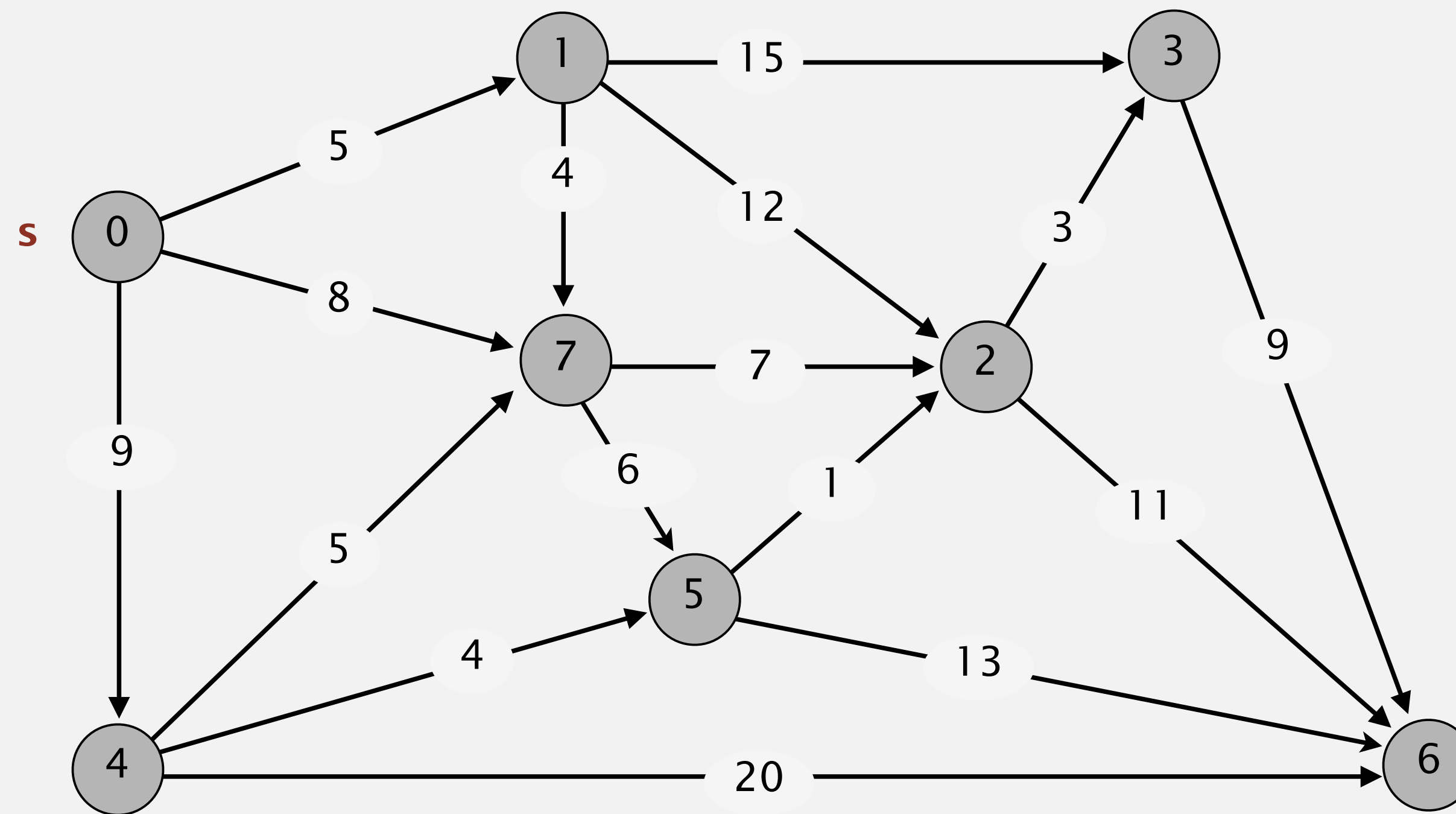
Proof sketch: Assume all edges have non-negative weights.

- At start,  $\text{distTo}[\text{source}] = 0$ , which is optimal.
- After relaxing all edges from source, let vertex  $v_1$  be the vertex with minimum weight, i.e. that is closest to the source. Claim:  $\text{distTo}[v_1]$  is optimal, and thus future relaxations will fail. Why?
  - $\text{distTo}[p] \geq \text{distTo}[v_1]$  for all  $p$ , therefore
  - $\text{distTo}[p] + w \geq \text{distTo}[v_1]$
- Can use induction to prove that this holds for all vertices after dequeuing.

# **Worksheet #3 full walkthrough**

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $\text{distTo}[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.

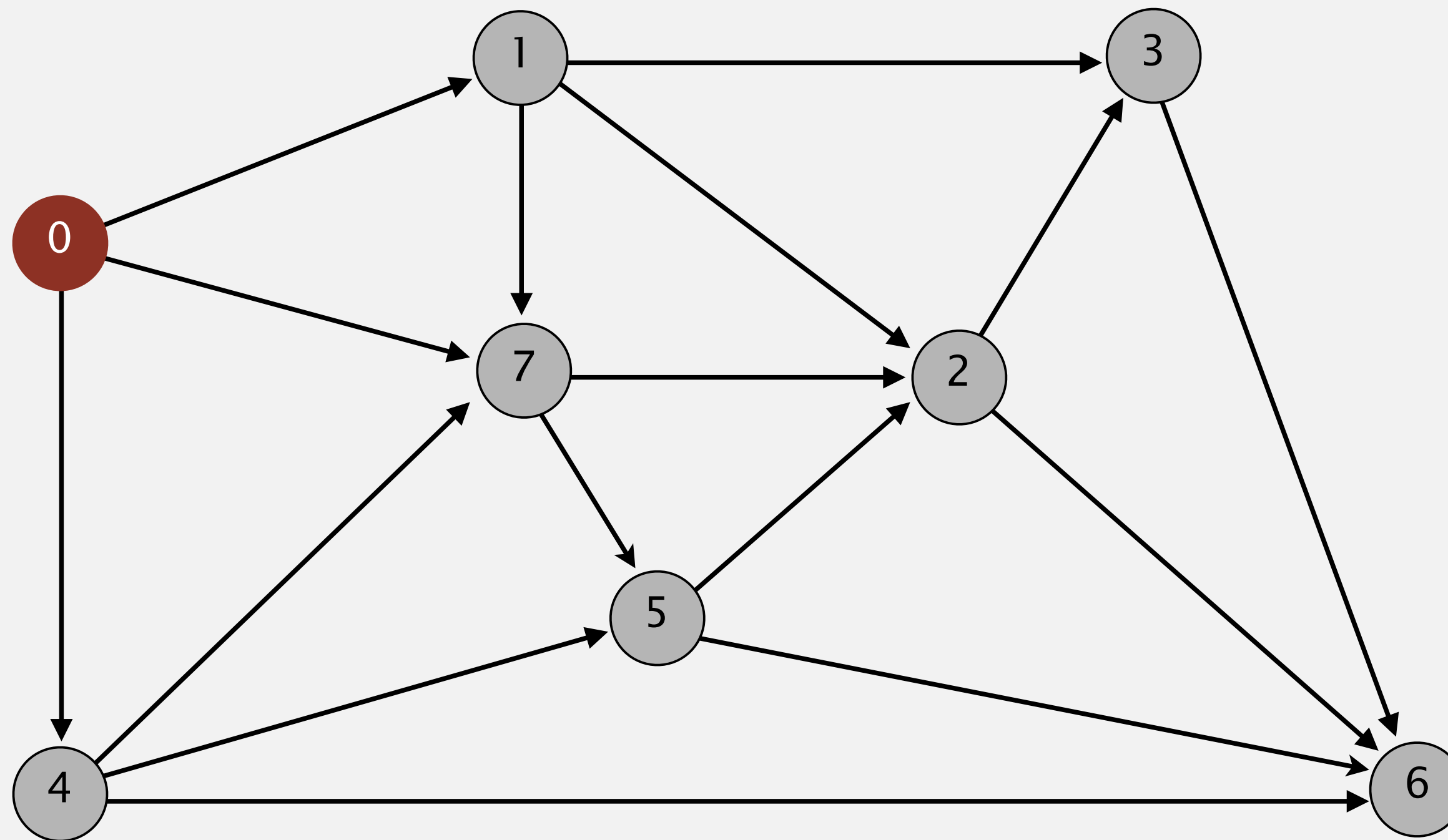


an edge-weighted digraph

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



choose source vertex 0

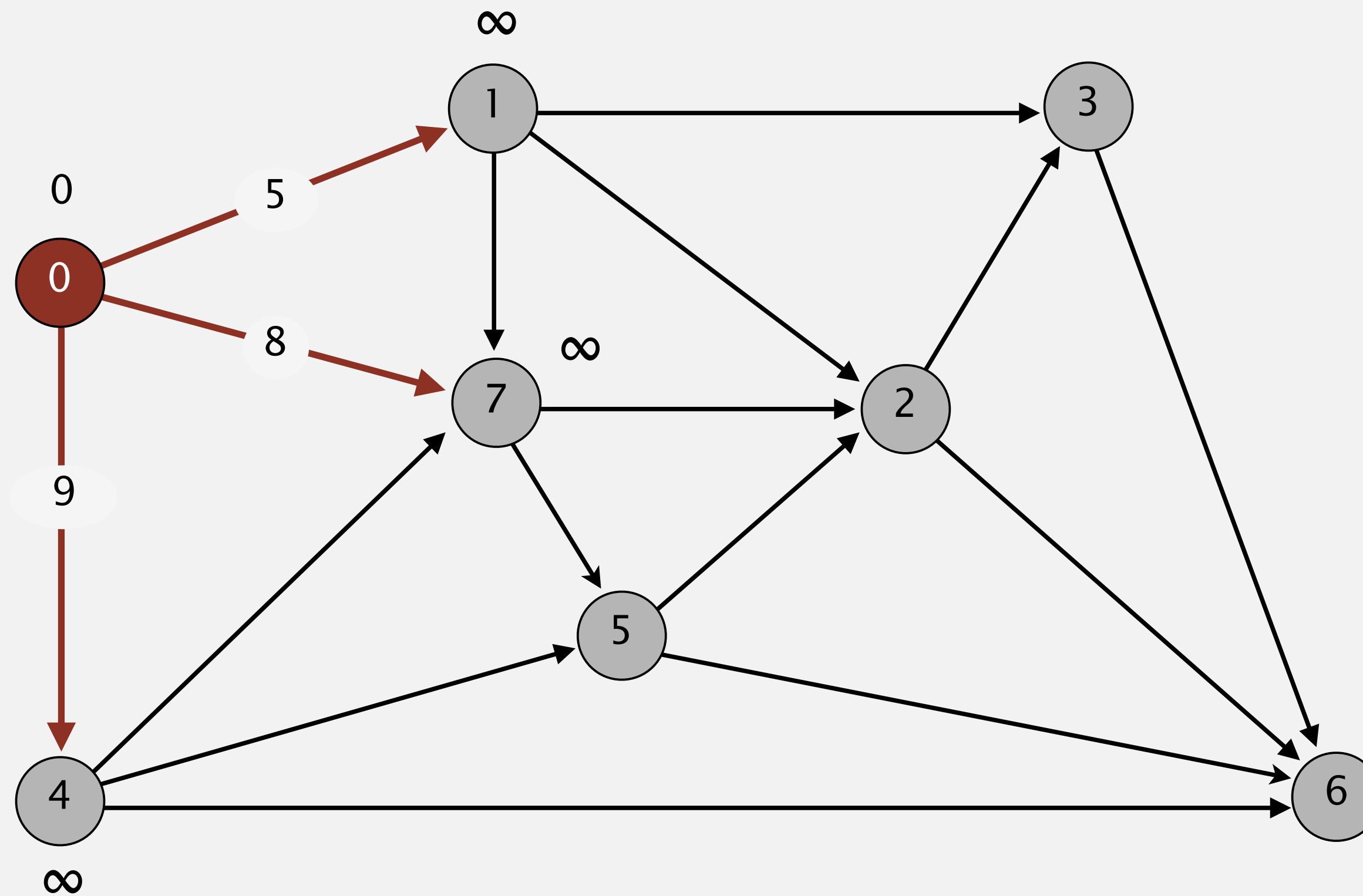
v	distTo[]	edgeTo[]
→ 0	0.0	-
1		
2		
3		
4		
5		
6		
7		

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0



# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



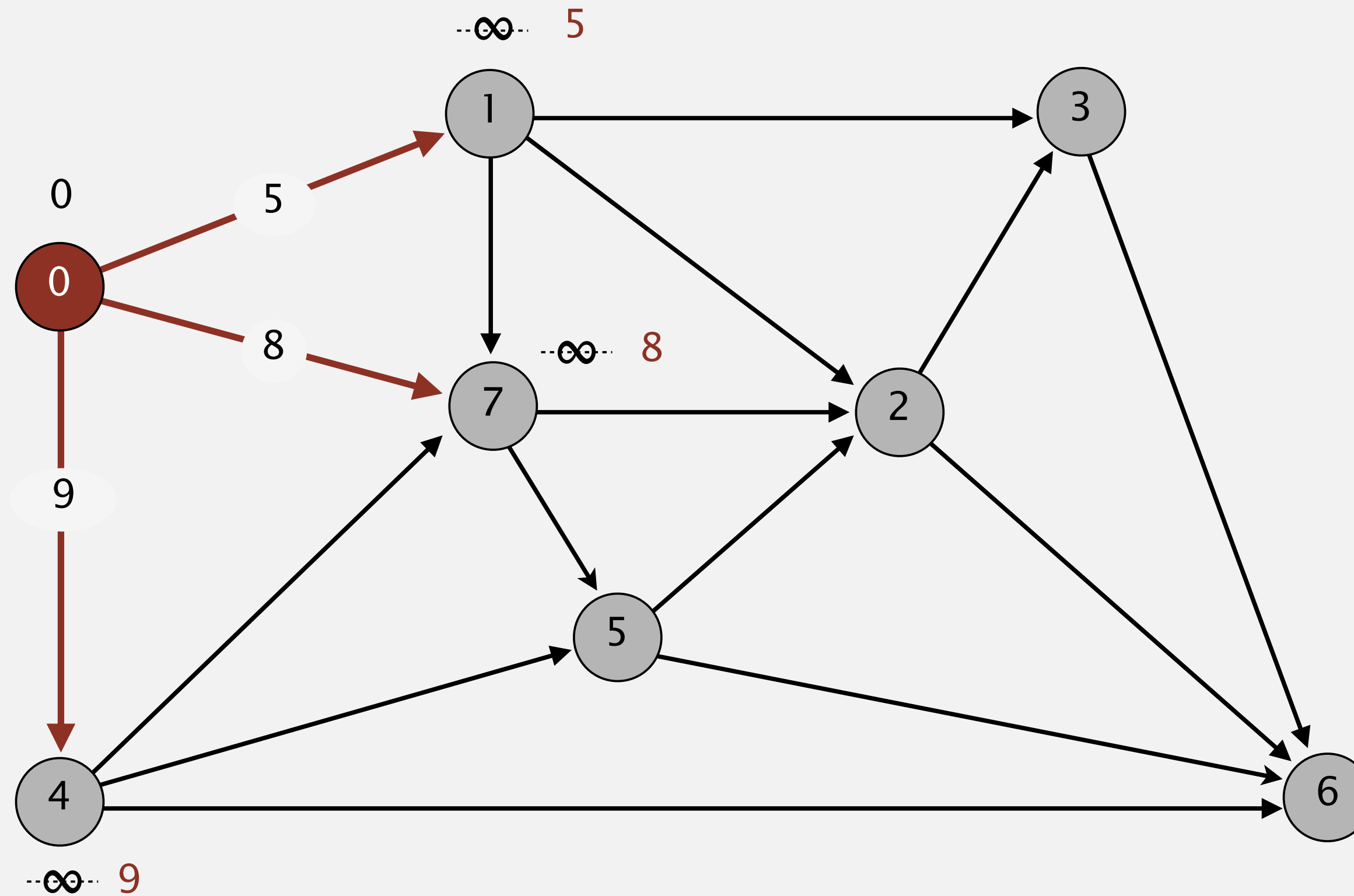
relax all edges adjacent from 0

v	distTo[]	edgeTo[]
→ 0	0.0	-
1		
2		
3		
4		
5		
6		
7		

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



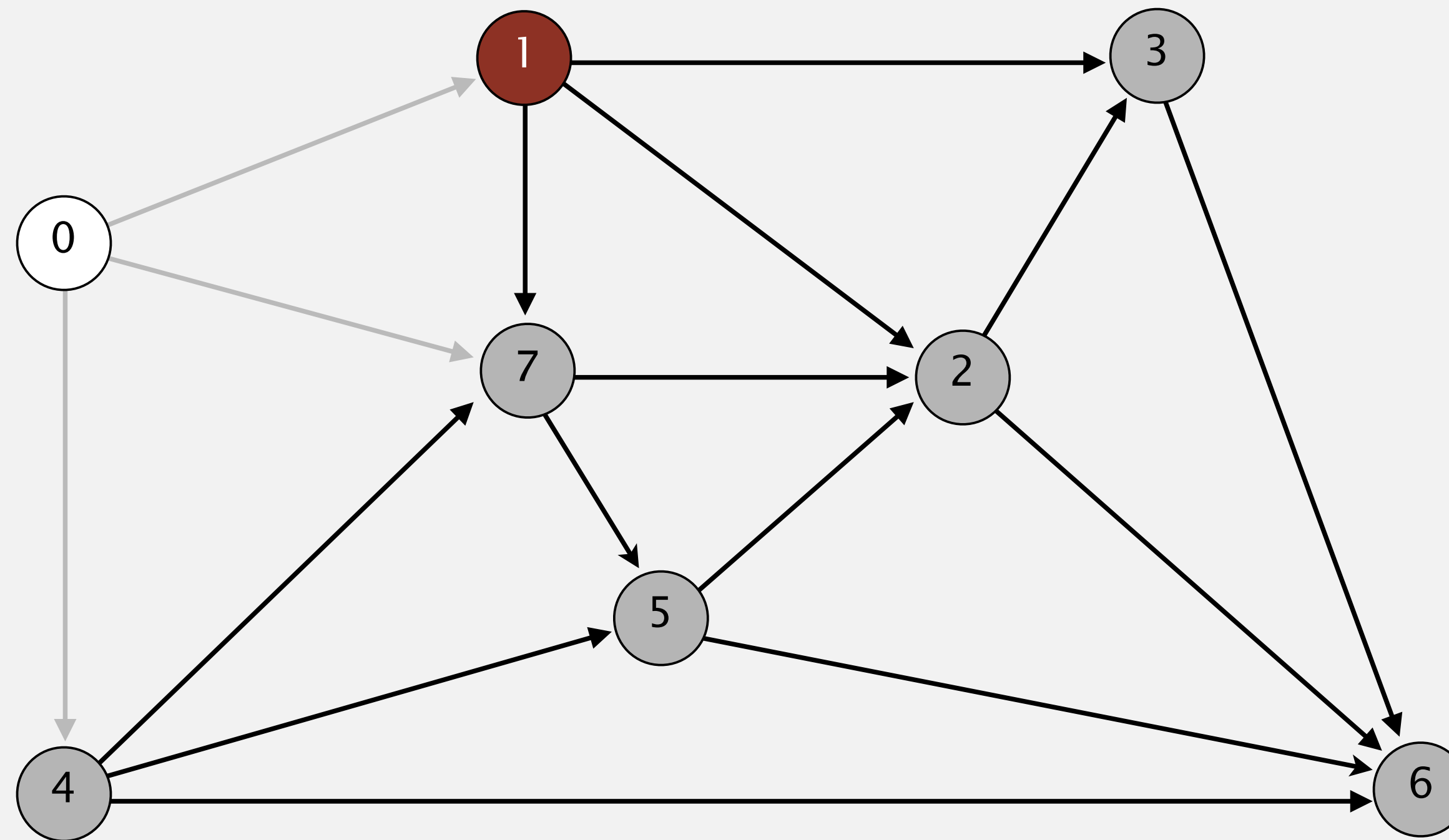
relax all edges adjacent from 0

v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	∞	
3	∞	
4	9.0	0→4
5	∞	
6	∞	
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



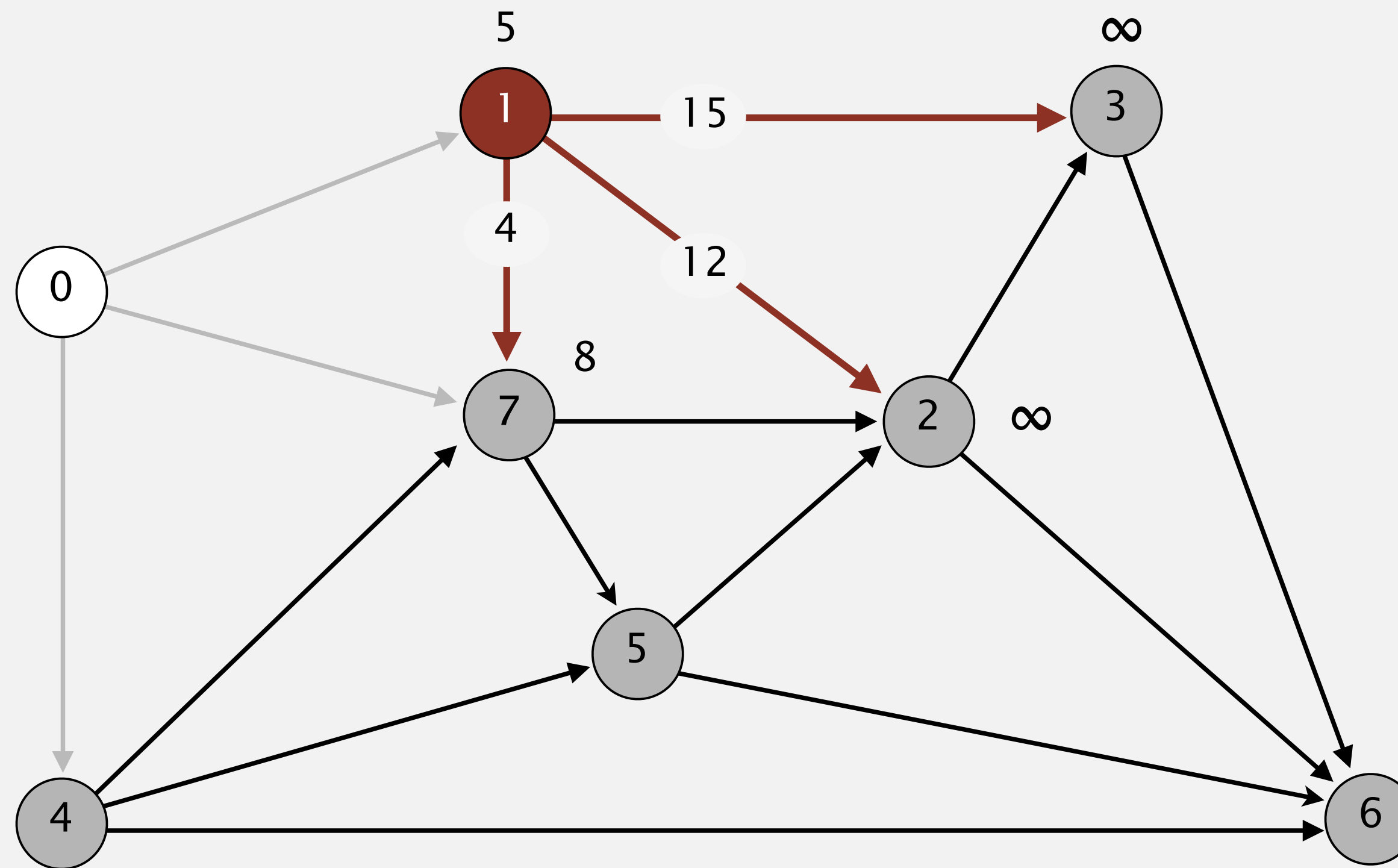
choose vertex 1

v	distTo[]	edgeTo[]
0	0.0	-
→ 1	5.0	0→1
2		
3		
4	9.0	0→4
5		
6		
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



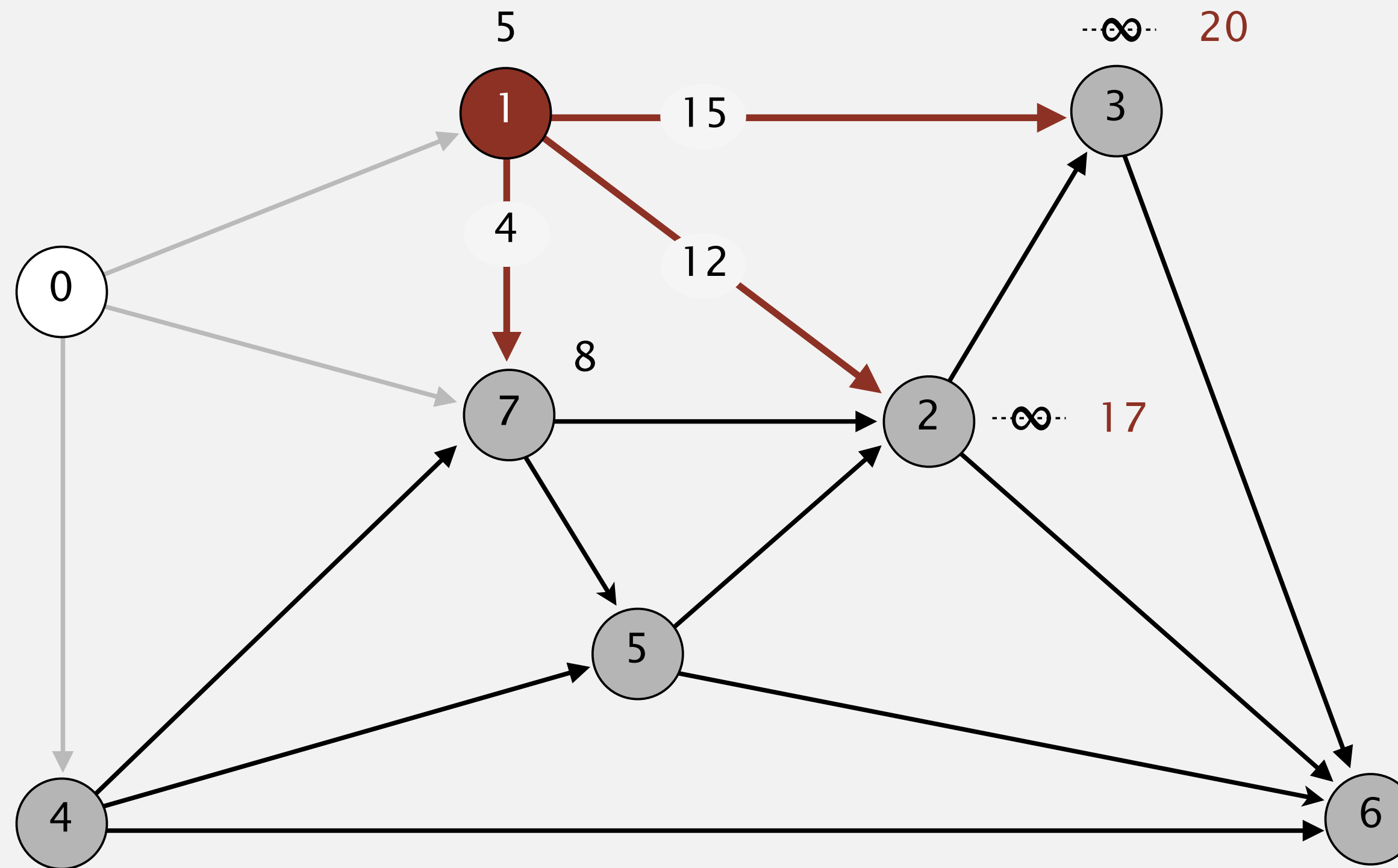
v	distTo[]	edgeTo[]
0	0.0	-
→ 1	5.0	0→1
2		
3		
4	9.0	0→4
5		
6		
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

relax all edges adjacent from 1

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



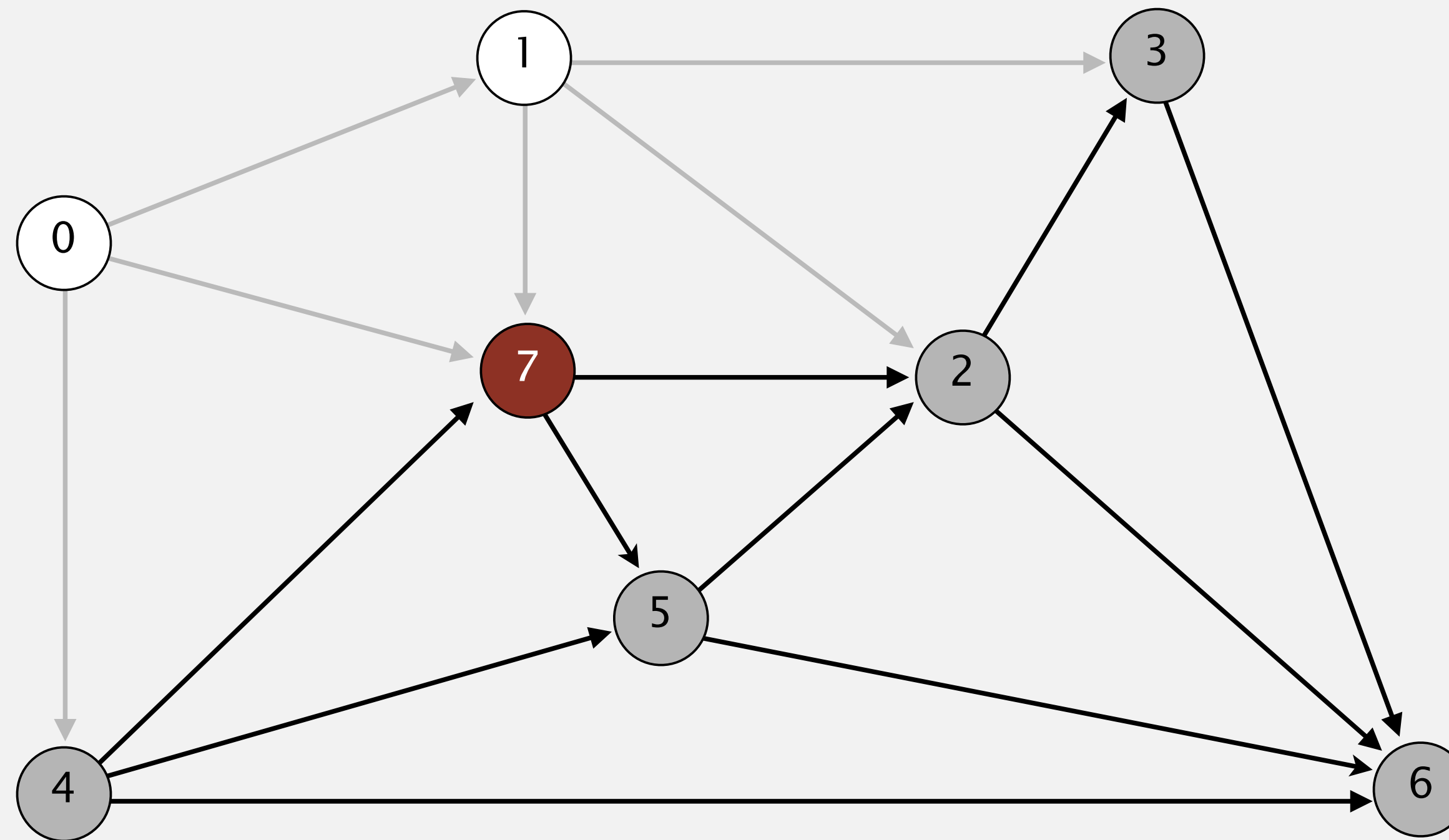
v	distTo[]	edgeTo[]
0	0.0	-
→ 1	5.0	0→1
2	17.0	1→2
3	20.0	1→3
4	9.0	0→4
5		
6		
7	8.0 ✓	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

relax all edges adjacent from 1

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



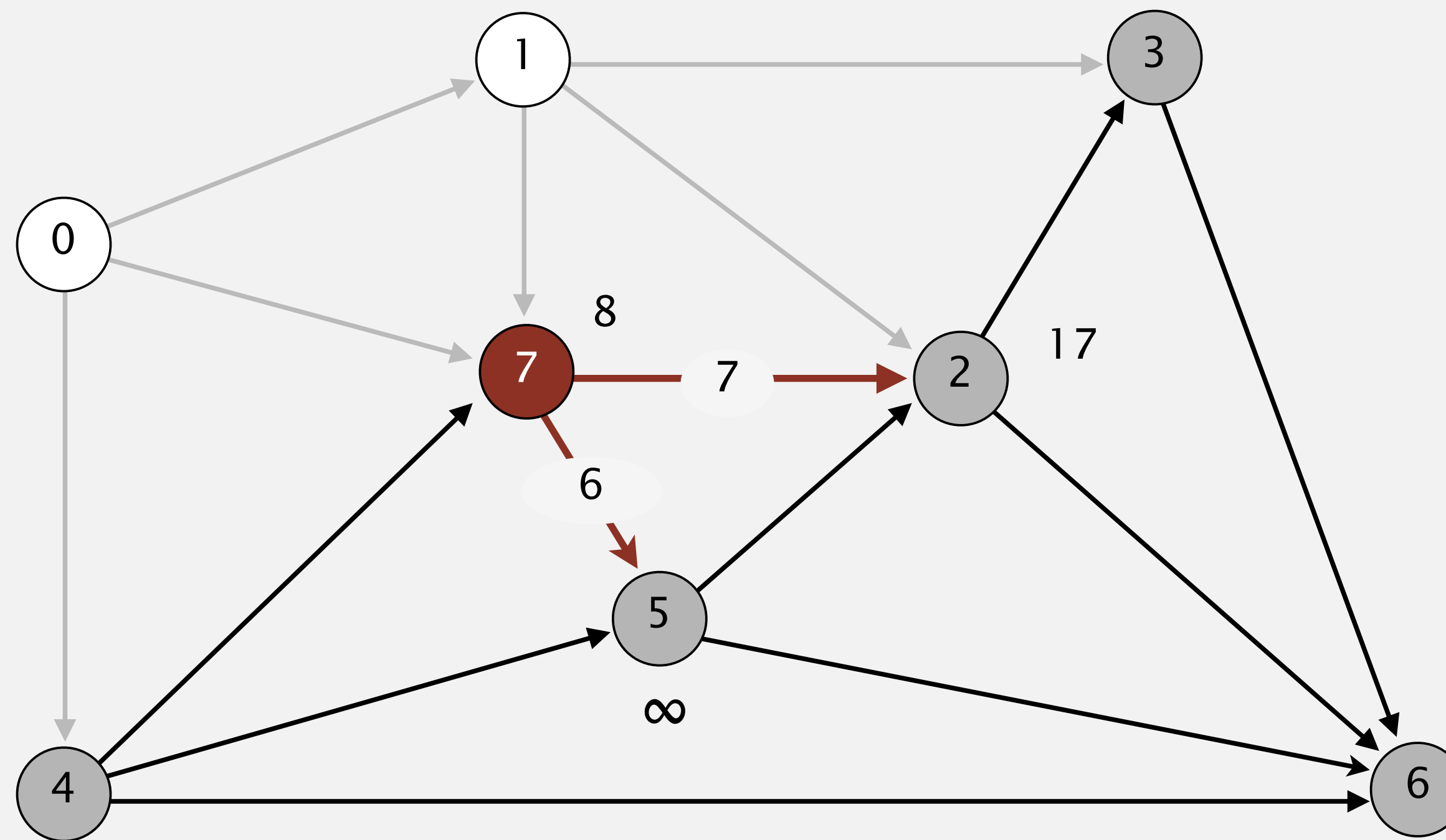
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	17.0	1→2
3	20.0	1→3
4	9.0	0→4
5		
6		
7	8.0	0→7

choose vertex 7

0→1 5.0  
 0→4 9.0  
 0→7 8.0  
 1→2 12.0  
 1→3 15.0  
 1→7 4.0  
 2→3 3.0  
 2→6 11.0  
 3→6 9.0  
 4→5 4.0  
 4→6 20.0  
 4→7 5.0  
 5→2 1.0  
 5→6 13.0  
 7→5 6.0  
 7→2 7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



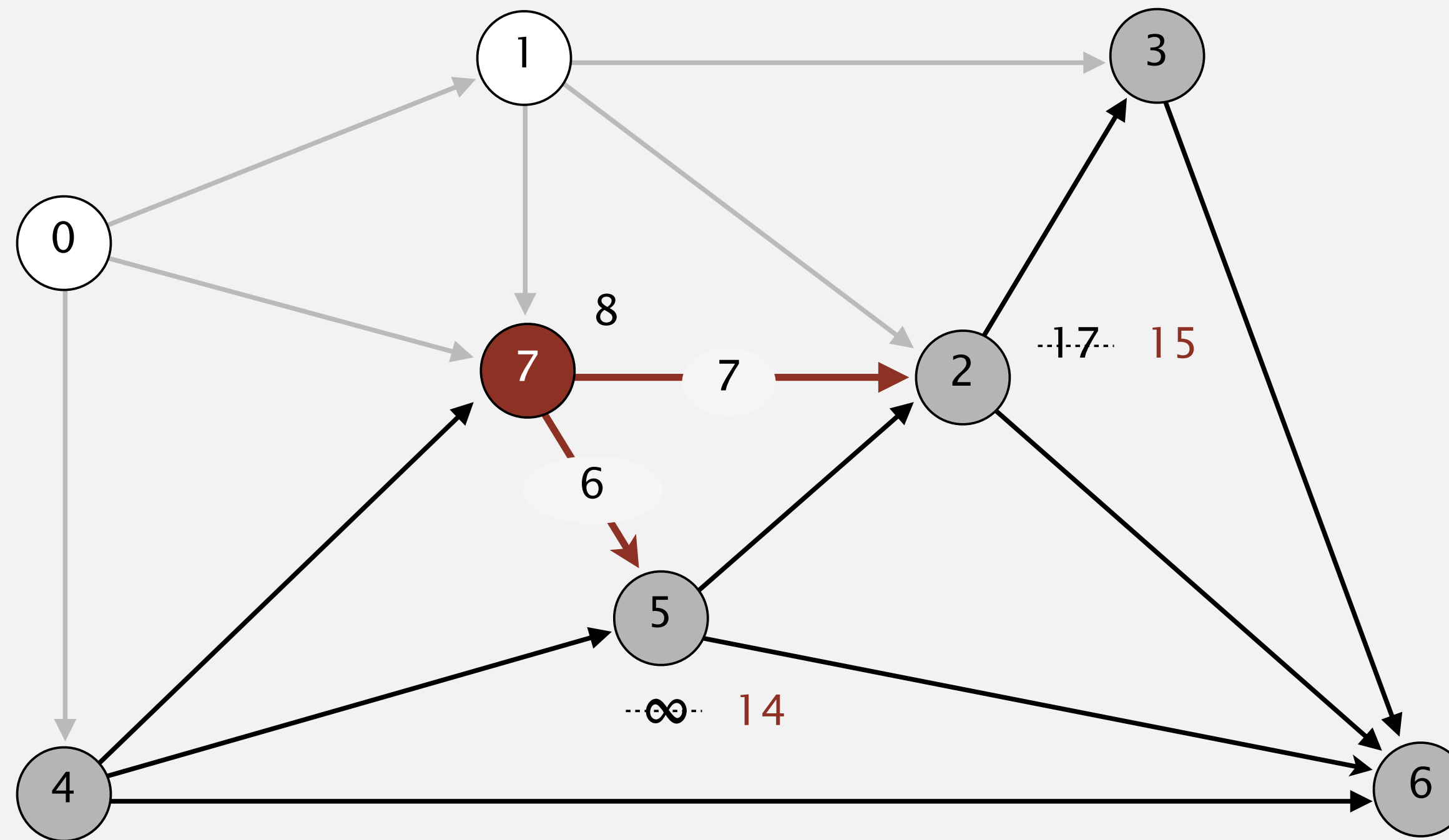
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	17.0	1→2
3	20.0	1→3
4	9.0	0→4
5		
6		
7	8.0	0→7

0→1 5.0  
 0→4 9.0  
 0→7 8.0  
 1→2 12.0  
 1→3 15.0  
 1→7 4.0  
 2→3 3.0  
 2→6 11.0  
 3→6 9.0  
 4→5 4.0  
 4→6 20.0  
 4→7 5.0  
 5→2 1.0  
 5→6 13.0  
 7→5 6.0  
 7→2 7.0

relax all edges adjacent from 7

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



relax all edges adjacent from 7

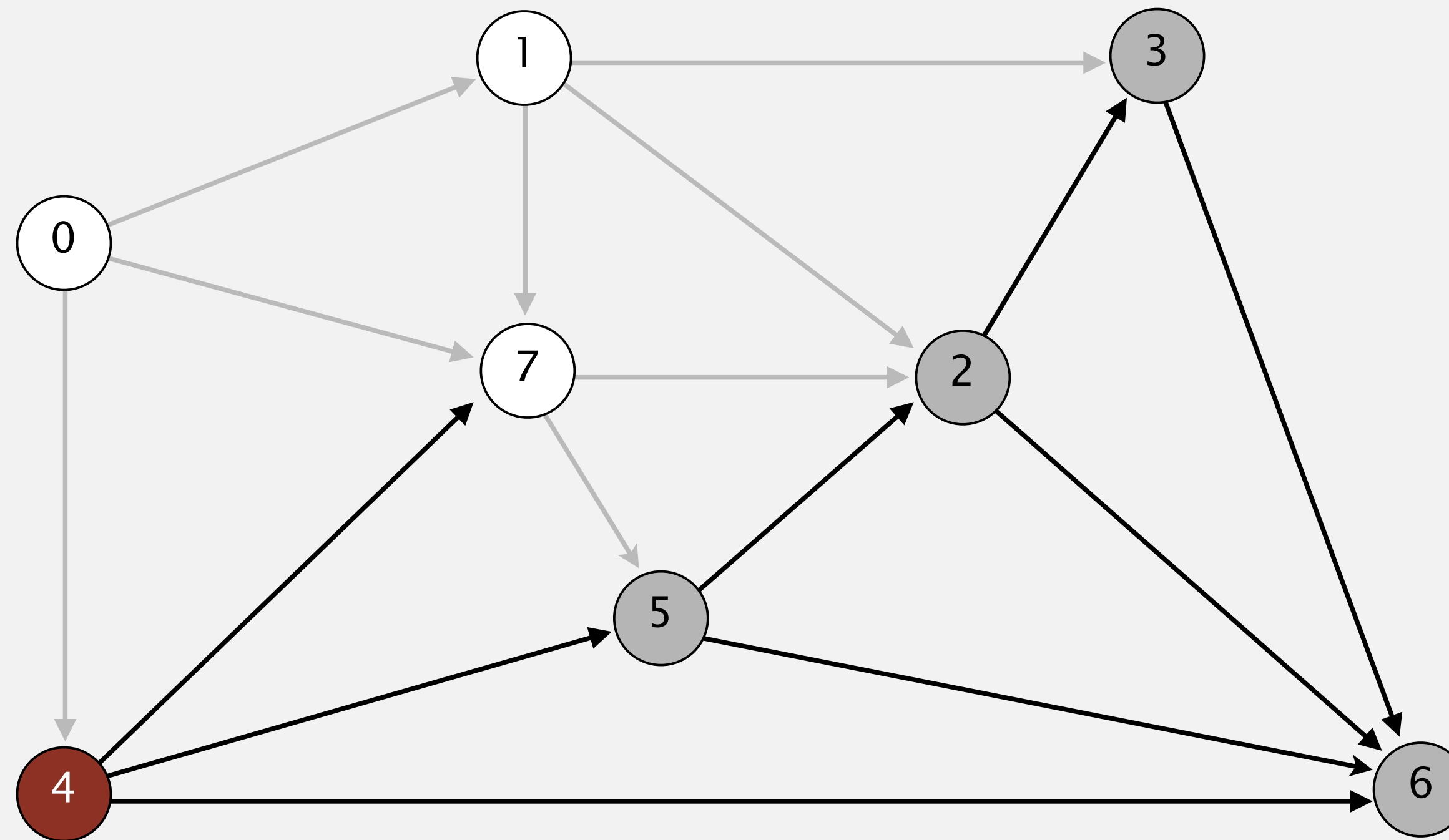
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	15.0	7→2
3	20.0	1→3
4	9.0	0→4
5	14.0	7→5
6		
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0



# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



select vertex 4

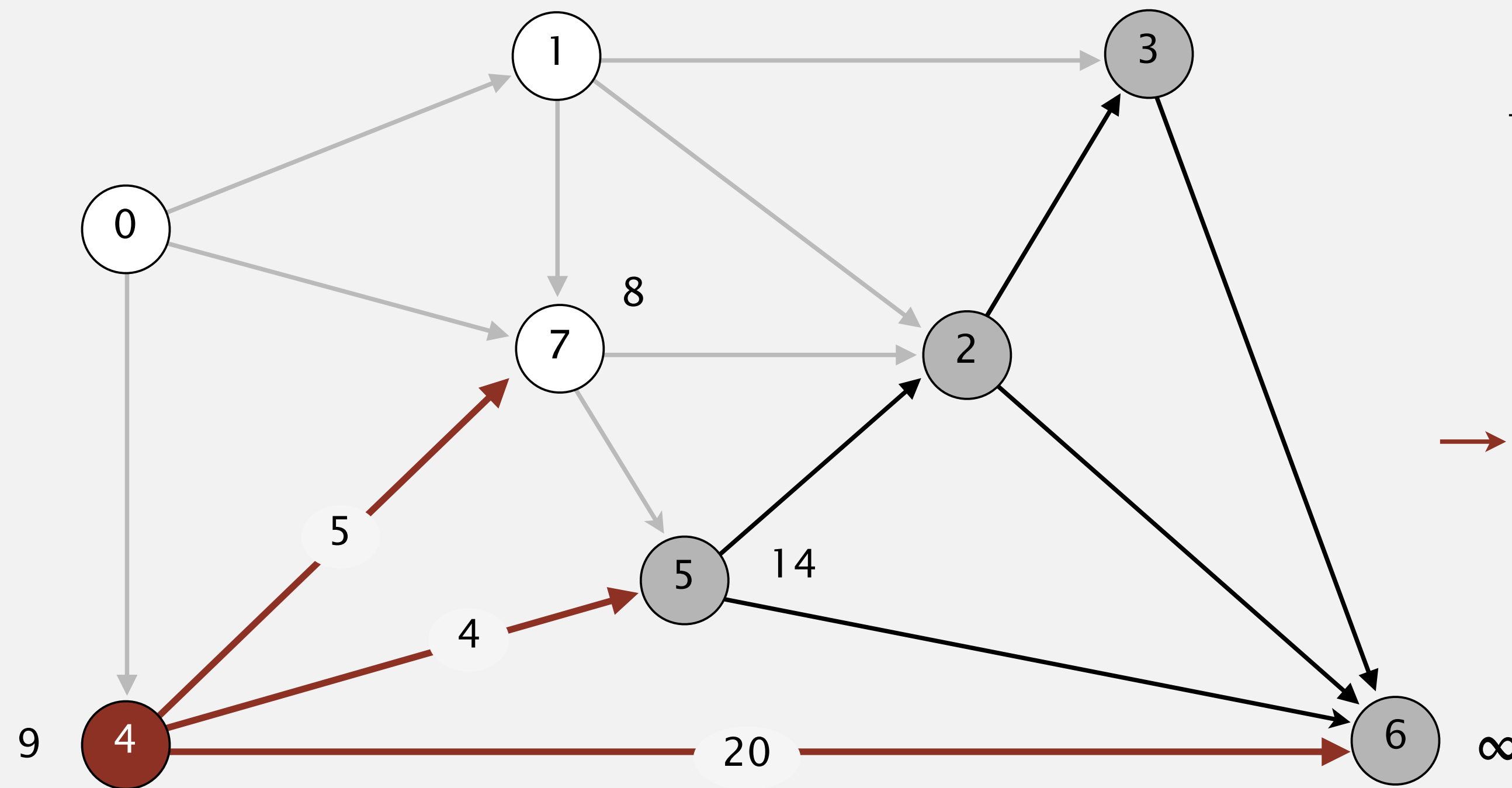
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	15.0	7→2
3	20.0	1→3
→ 4	9.0	0→4
5	14.0	7→5
6		
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.

0→1 5.0  
 0→4 9.0  
 0→7 8.0  
 1→2 12.0  
 1→3 15.0  
 1→7 4.0  
 2→3 3.0  
 2→6 11.0  
 3→6 9.0  
 4→5 4.0  
 4→6 20.0  
 4→7 5.0  
 5→2 1.0  
 5→6 13.0  
 7→5 6.0  
 7→2 7.0



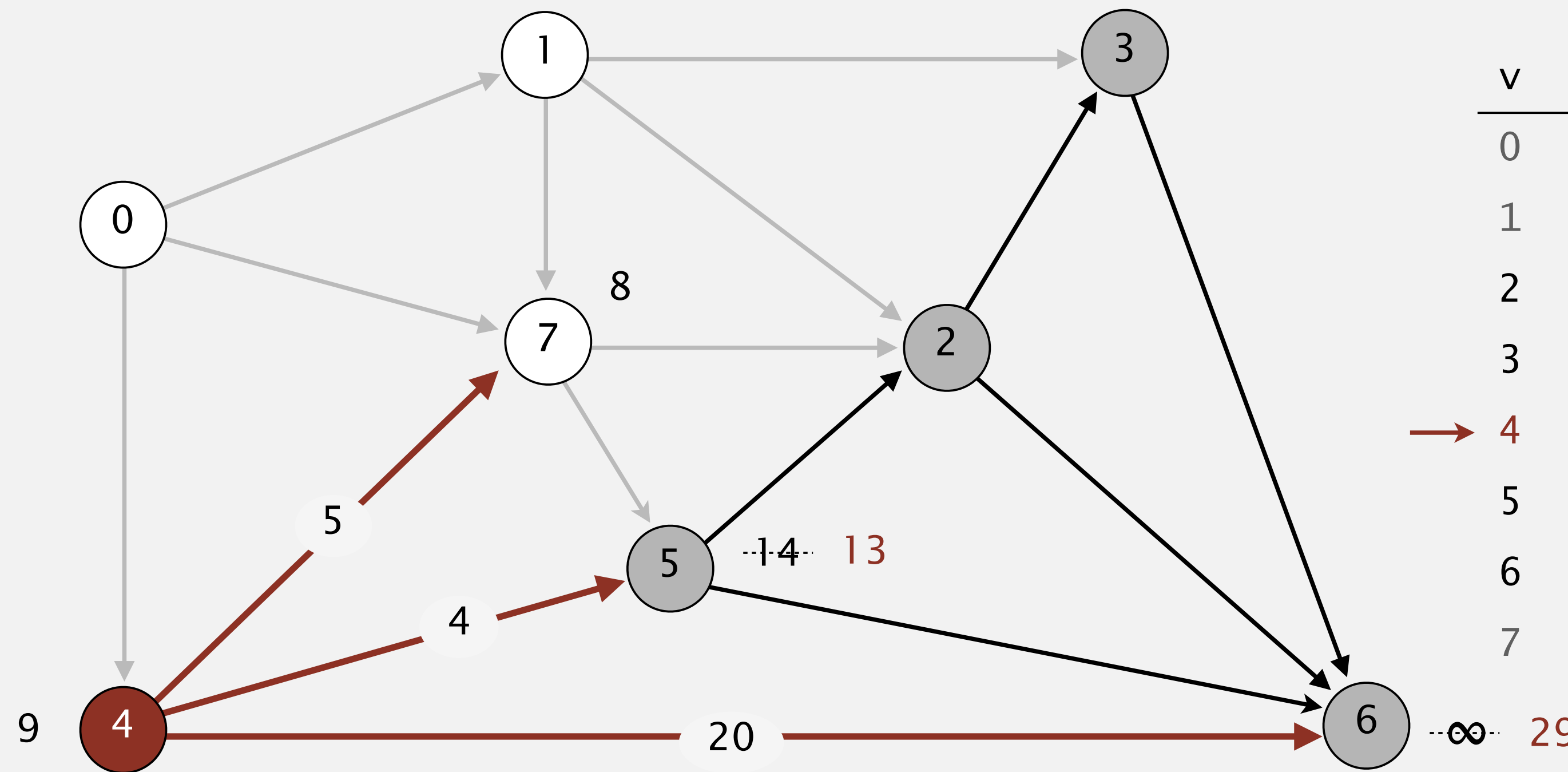
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	15.0	7→2
3	20.0	1→3
→ 4	9.0	0→4
5	14.0	7→5
6		
7	8.0	0→7

relax all edges adjacent from 4

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.

0→1 5.0  
 0→4 9.0  
 0→7 8.0  
 1→2 12.0  
 1→3 15.0  
 1→7 4.0  
 2→3 3.0  
 2→6 11.0  
 3→6 9.0  
 4→5 4.0  
 4→6 20.0  
 4→7 5.0  
 5→2 1.0  
 5→6 13.0  
 7→5 6.0  
 7→2 7.0

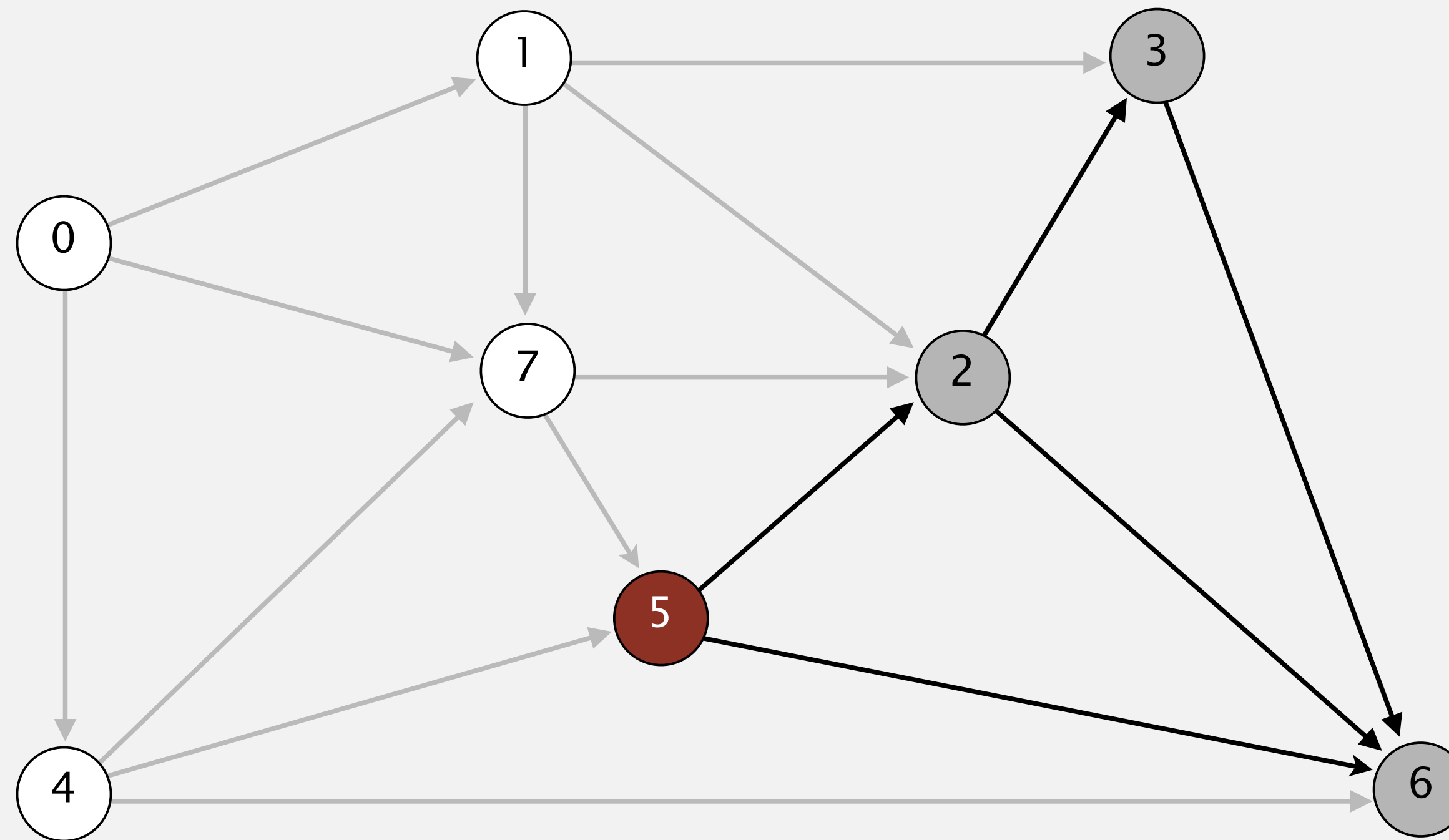


v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	15.0	7→2
3	20.0	1→3
→ 4	9.0	0→4
5	13.0	4→5
6	29.0	4→6
7	8.0 ✓	0→7

relax all edges adjacent from 4

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



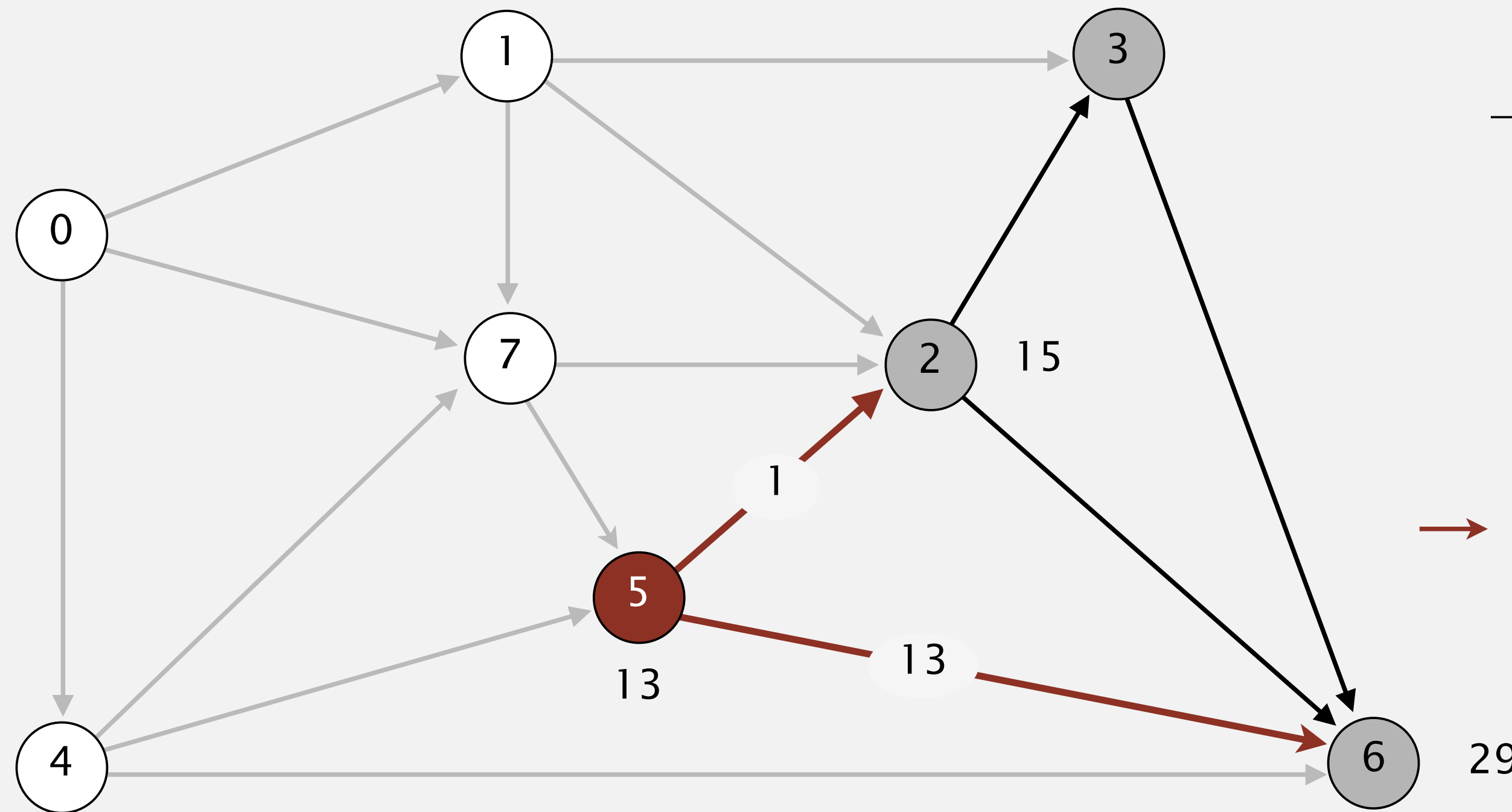
**select vertex 5**

v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	15.0	7→2
3	20.0	1→3
4	9.0	0→4
→ 5	13.0	4→5
6	29.0	4→6
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



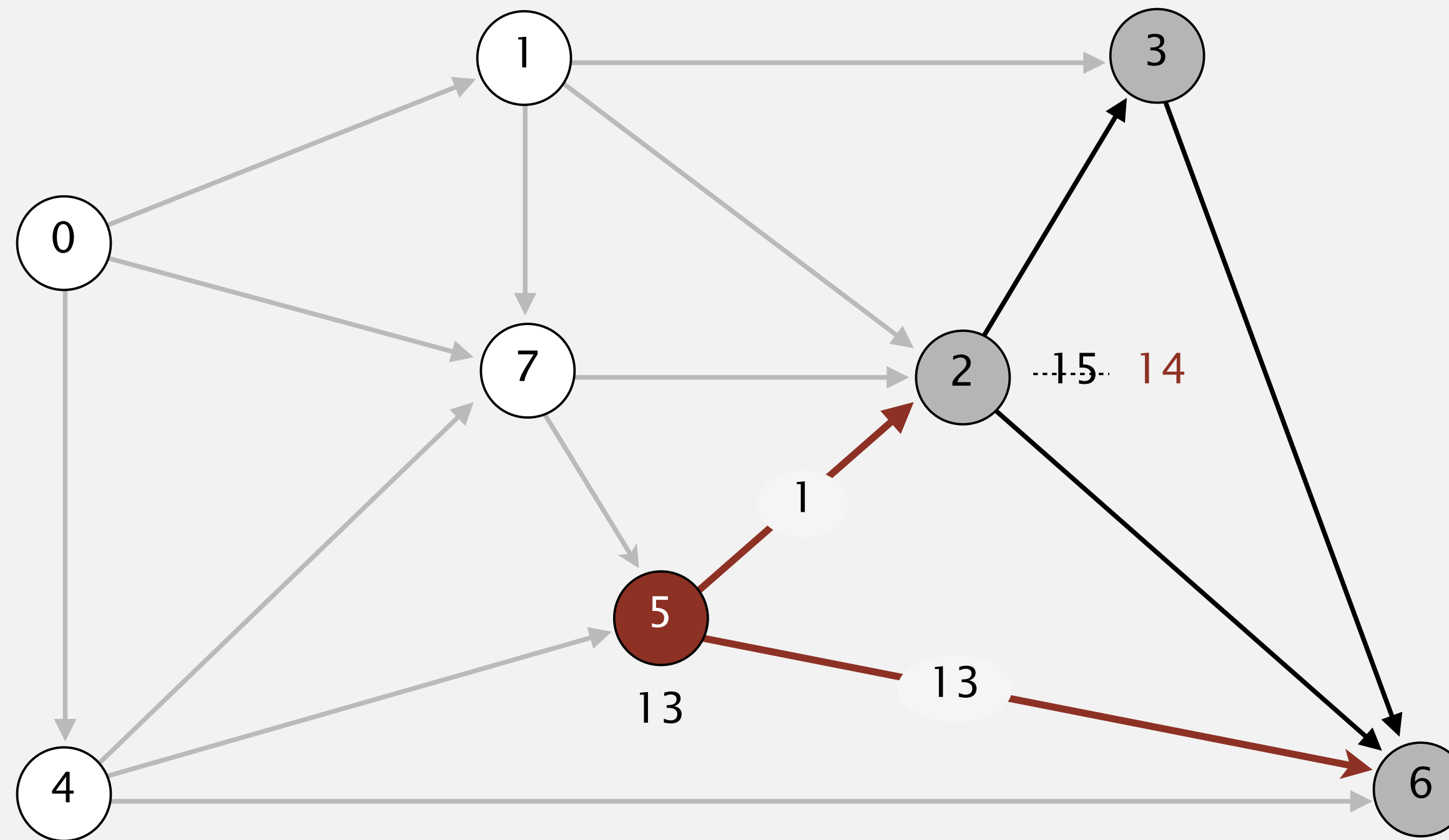
relax all edges adjacent from 5

v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	15.0	7→2
3	20.0	1→3
4	9.0	0→4
→ 5	13.0	4→5
6	29.0	4→6
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



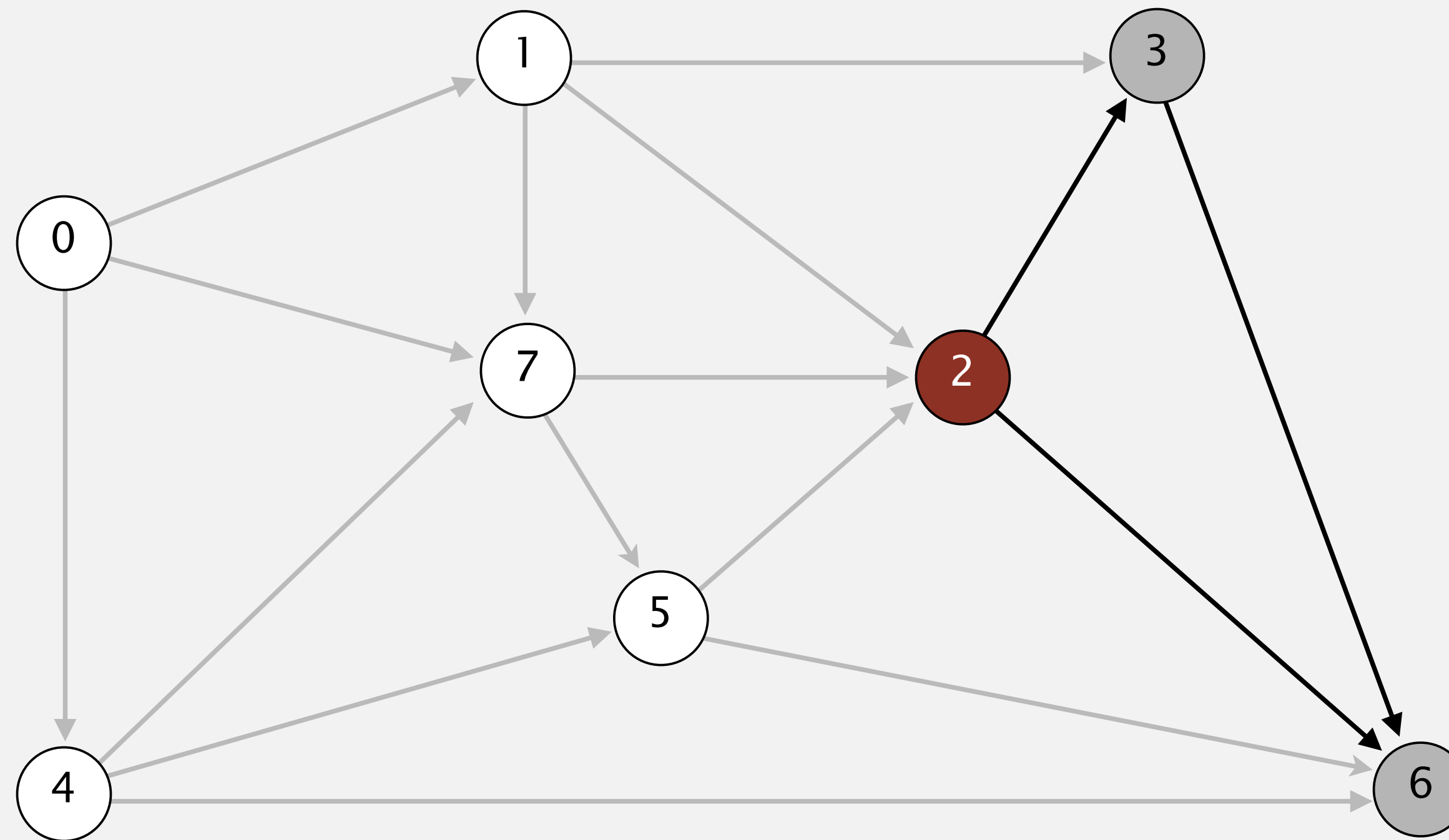
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	20.0	1→3
4	9.0	0→4
5	13.0	4→5
6	26.0	5→6
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

relax all edges adjacent from 5

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



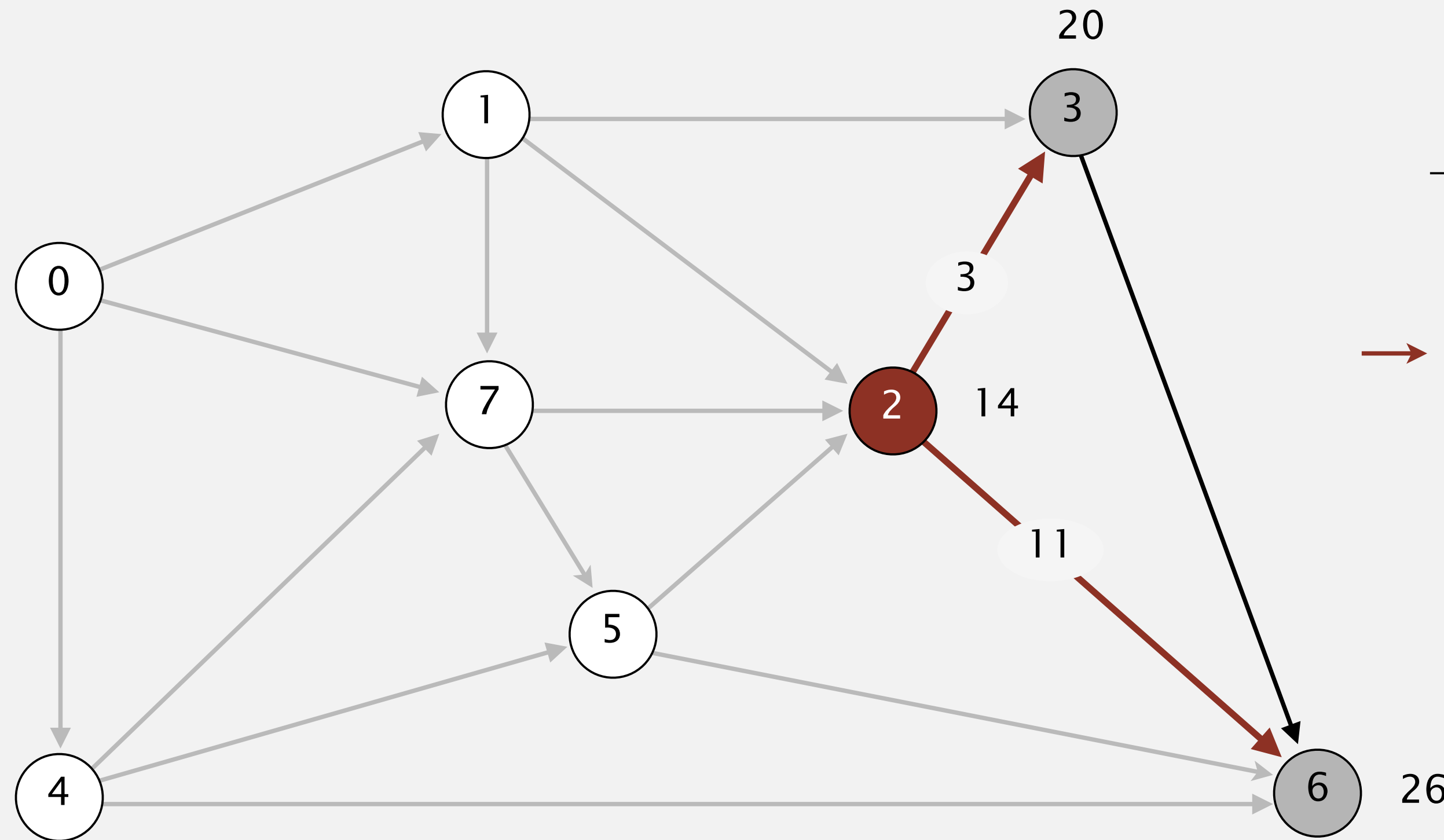
**select vertex 2**

v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
→ 2	14.0	5→2
3	20.0	1→3
4	9.0	0→4
5	13.0	4→5
6	26.0	5→6
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
→ 2	14.0	5→2
3	20.0	1→3
4	9.0	0→4
5	13.0	4→5
6	26.0	5→6
7	8.0	0→7

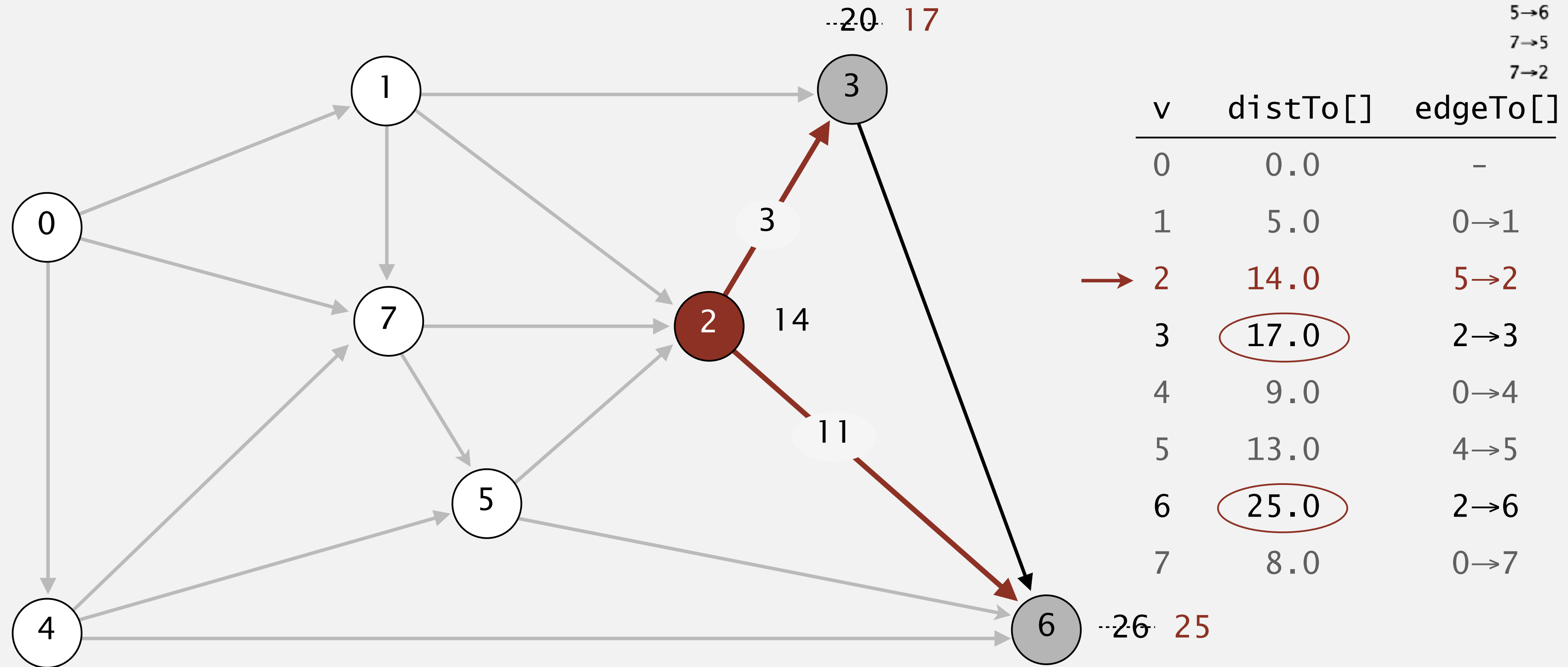
0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

relax all edges adjacent from 2



# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.

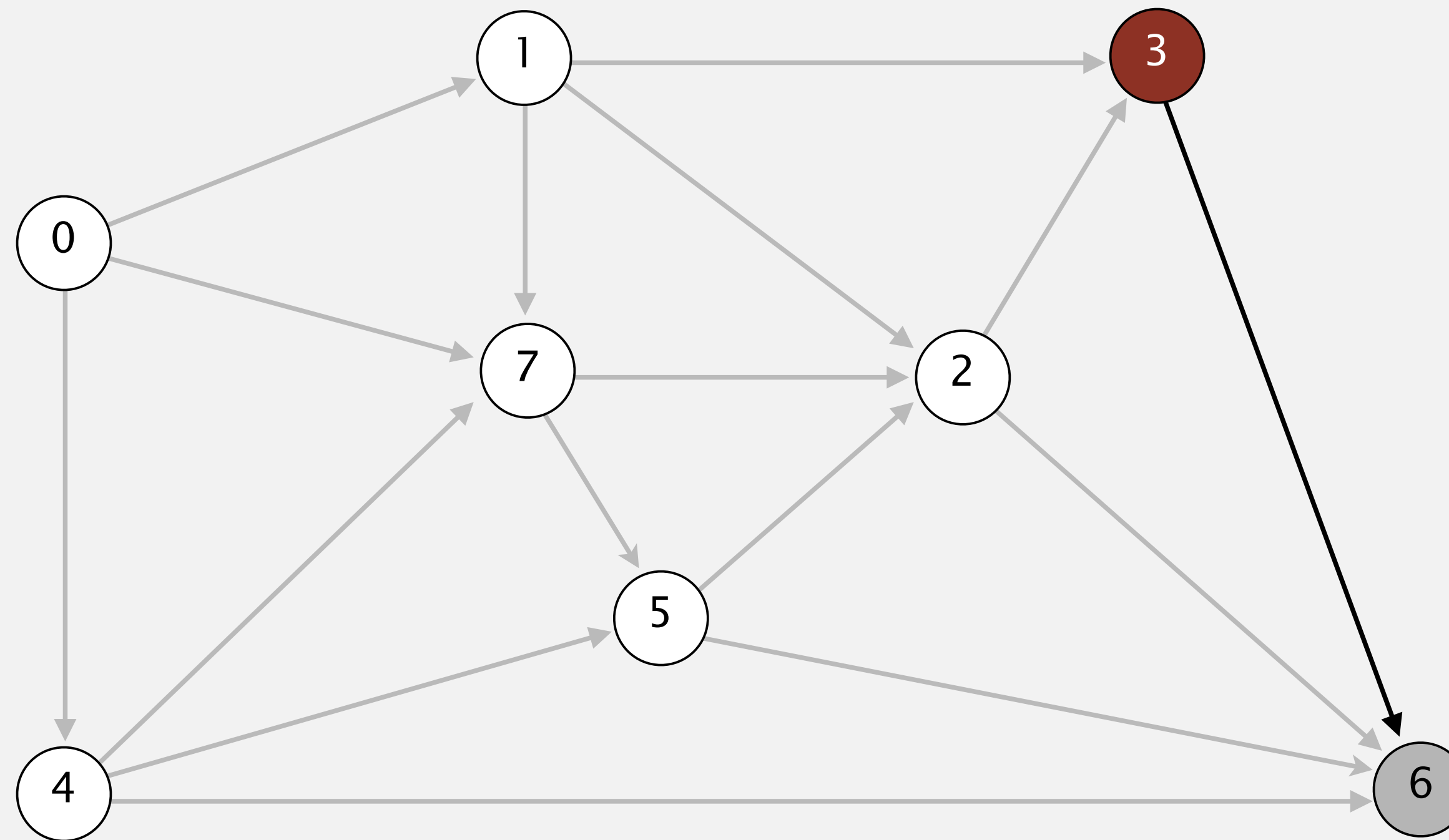


relax all edges adjacent from 2

0→1 5.0  
 0→4 9.0  
 0→7 8.0  
 1→2 12.0  
 1→3 15.0  
 1→7 4.0  
 2→3 3.0  
 2→6 11.0  
 3→6 9.0  
 4→5 4.0  
 4→6 20.0  
 4→7 5.0  
 5→2 1.0  
 5→6 13.0  
 7→5 6.0  
 7→2 7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



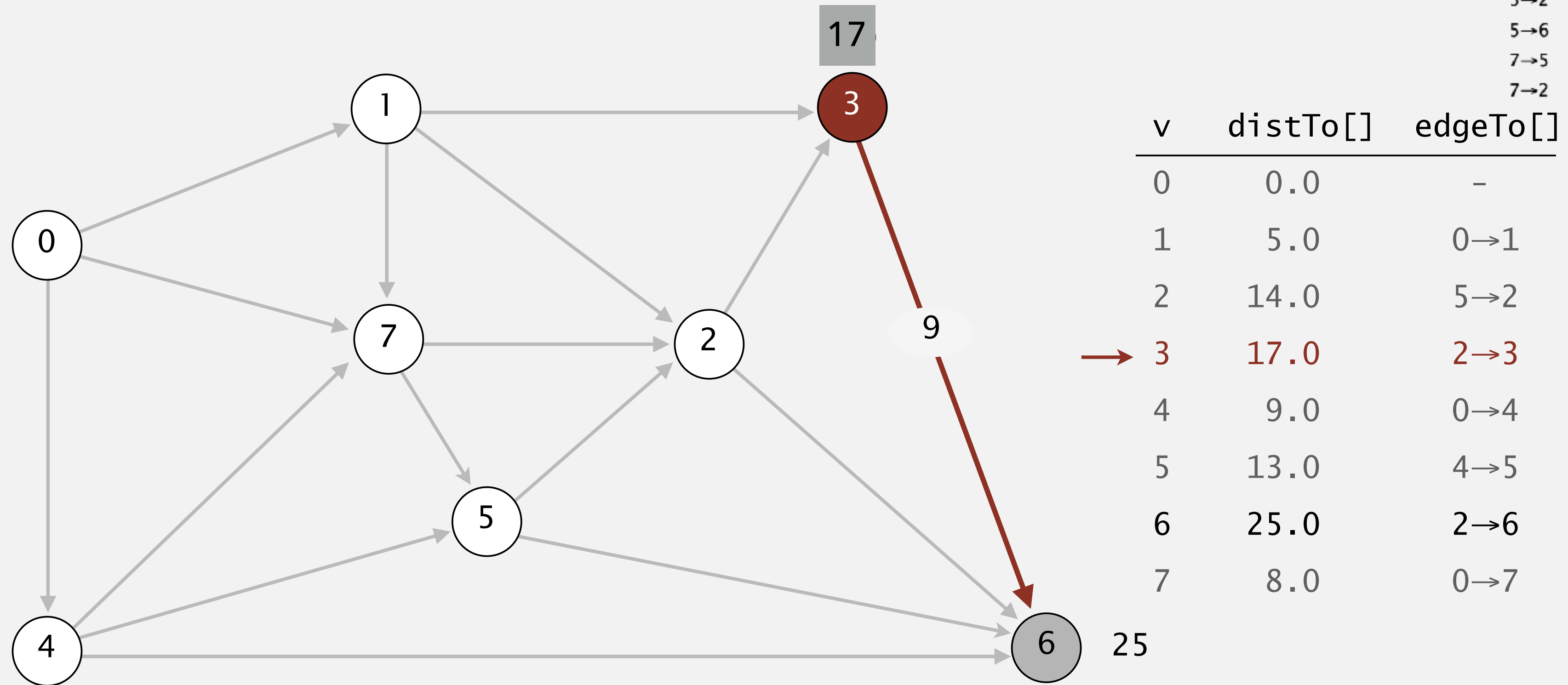
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
→ 3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0	2→6
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

**select vertex 3**

# Dijkstra's algorithm demo

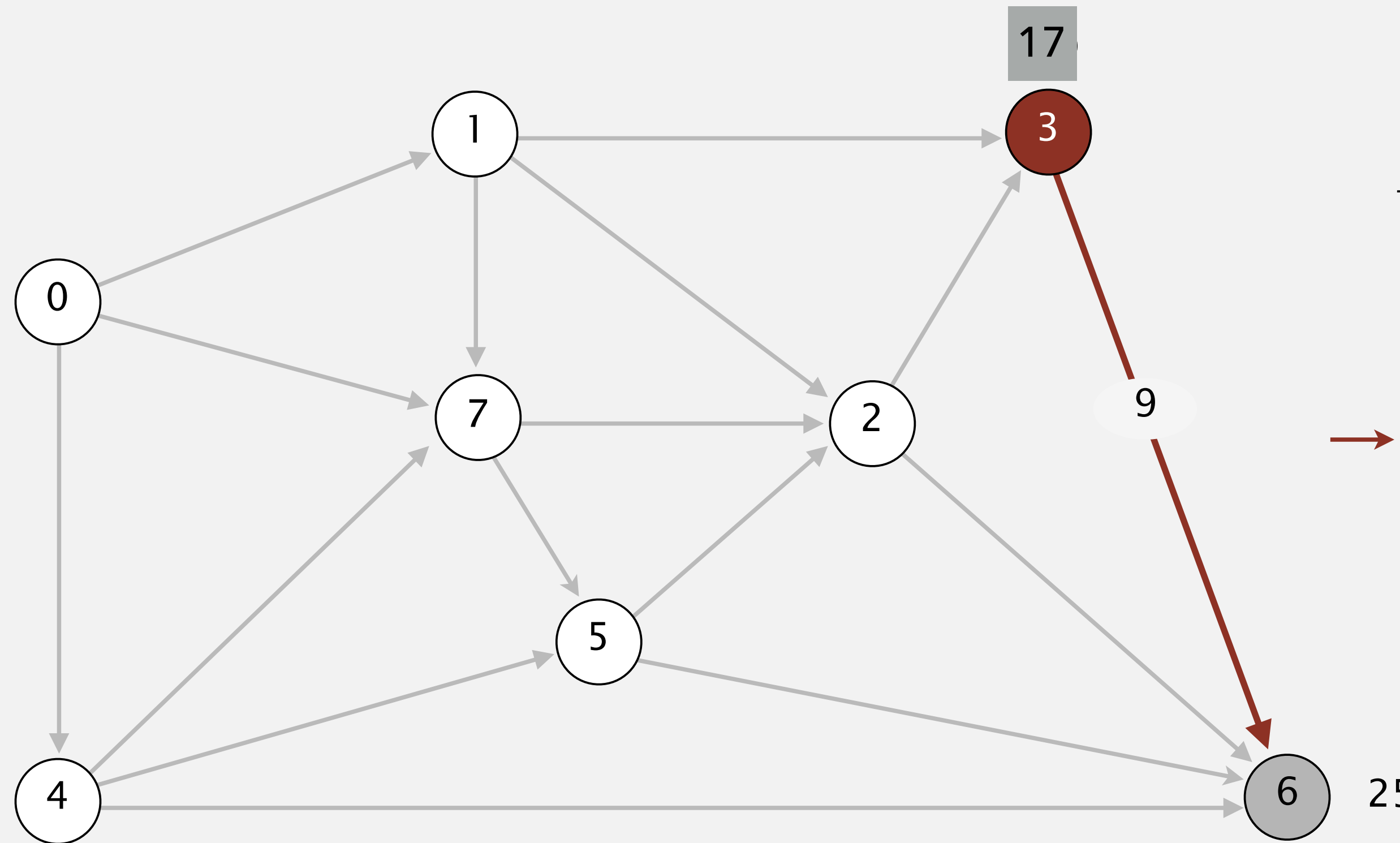
- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



relax all edges adjacent from 3

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



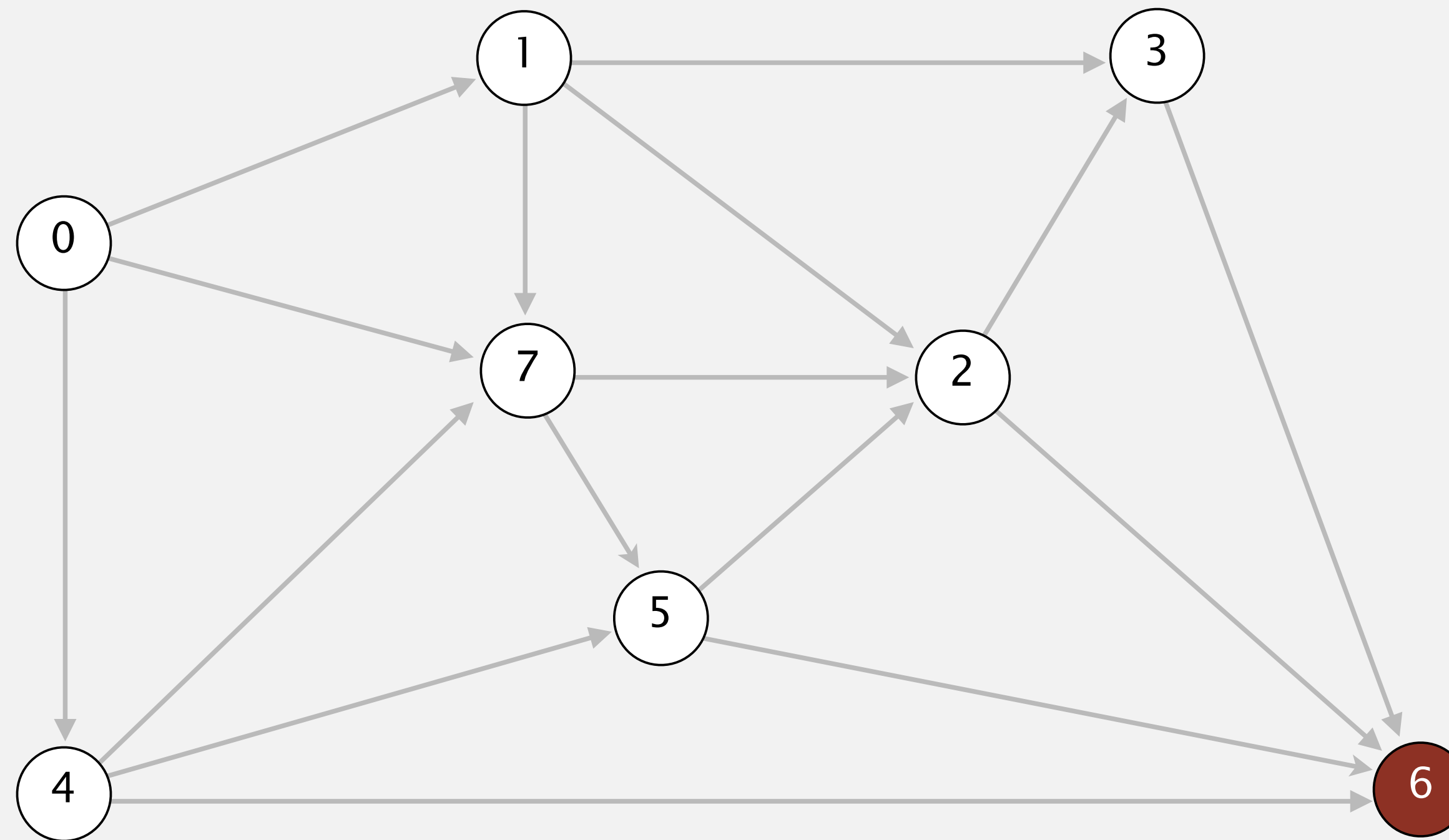
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0 ✓	2→6
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

relax all edges adjacent from 3

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



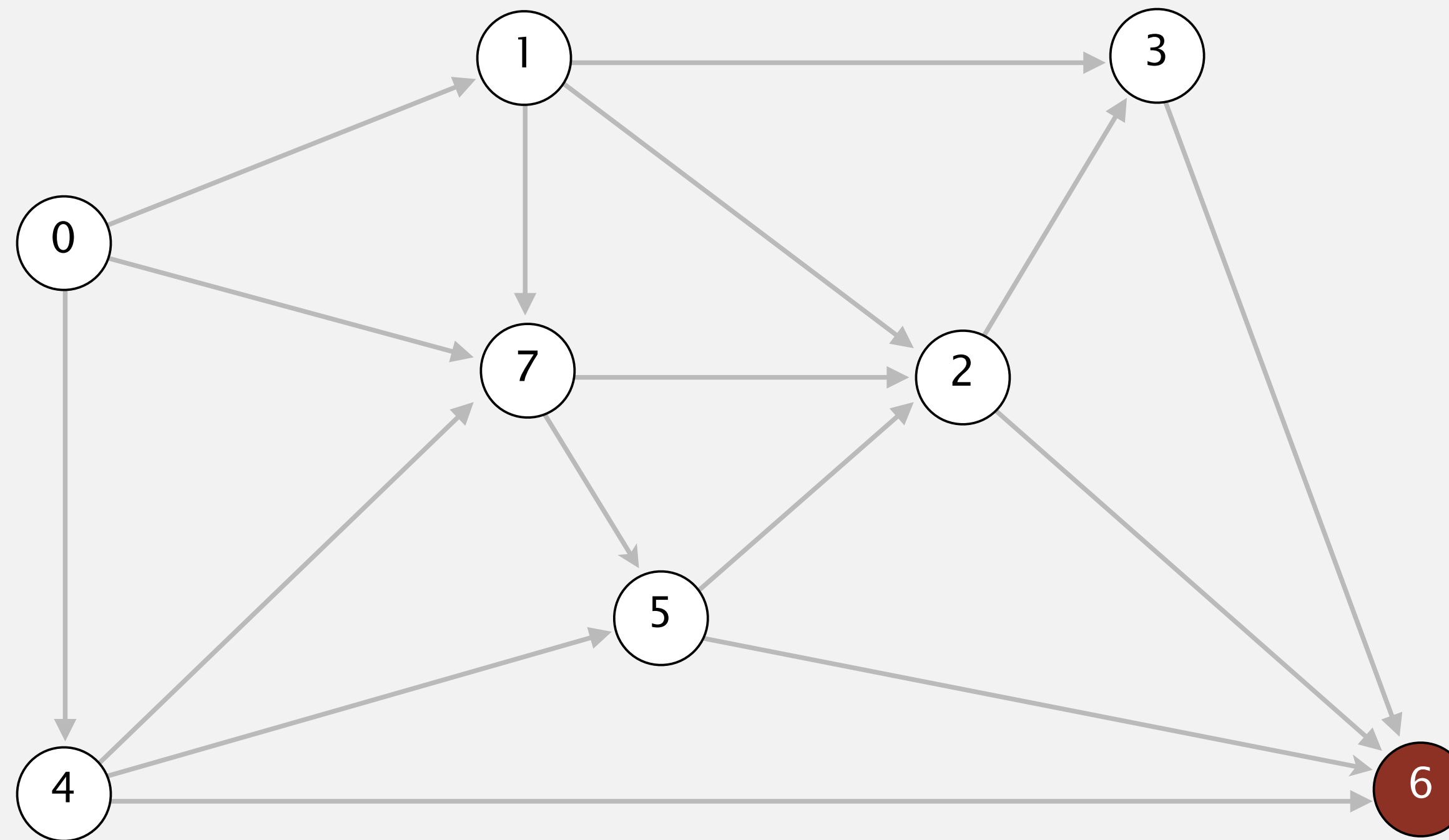
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
→ 6	25.0	2→6
7	8.0	0→7

**select vertex 6**

0→1 5.0  
 0→4 9.0  
 0→7 8.0  
 1→2 12.0  
 1→3 15.0  
 1→7 4.0  
 2→3 3.0  
 2→6 11.0  
 3→6 9.0  
 4→5 4.0  
 4→6 20.0  
 4→7 5.0  
 5→2 1.0  
 5→6 13.0  
 7→5 6.0  
 7→2 7.0

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $\text{distTo}[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



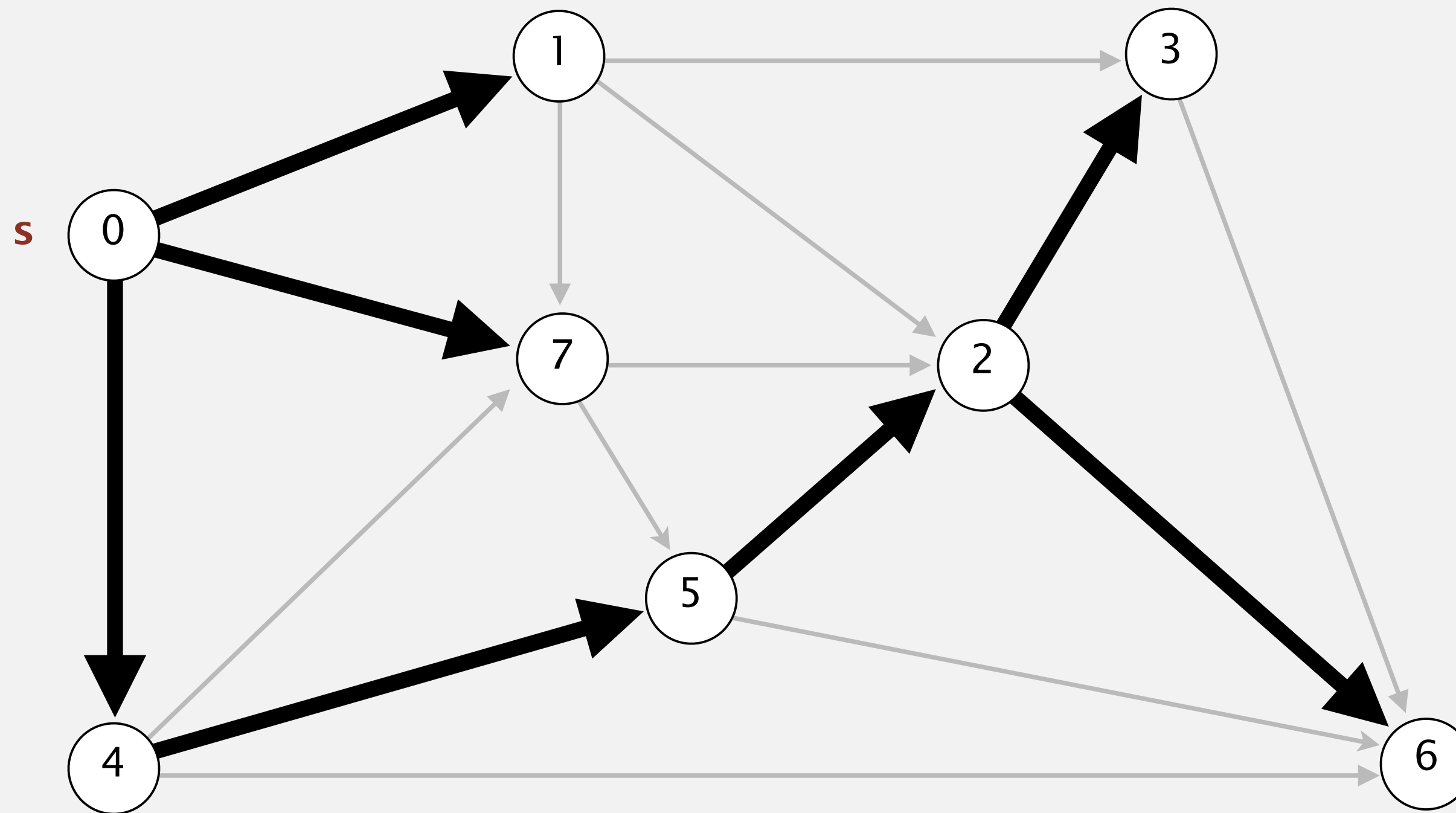
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
→ 6	25.0	2→6
7	8.0	0→7

- 0→1 5.0
- 0→4 9.0
- 0→7 8.0
- 1→2 12.0
- 1→3 15.0
- 1→7 4.0
- 2→3 3.0
- 2→6 11.0
- 3→6 9.0
- 4→5 4.0
- 4→6 20.0
- 4→7 5.0
- 5→2 1.0
- 5→6 13.0
- 7→5 6.0
- 7→2 7.0

relax all edges adjacent from 6

# Dijkstra's algorithm demo

- Consider vertices in increasing order of distance from  $s$  (non-tree vertex with the lowest  $distTo[]$  value).
- Add vertex to tree and relax all edges adjacent from that vertex.



shortest-paths tree from vertex  $s$

$v$	$distTo[]$	$edgeTo[]$
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0	2→6
7	8.0	0→7

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0