

# Lecture 5.0

## Shortest Paths

Department of Computer Science  
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# Lecture Goals

- In this lecture we study **shortest-paths** problems. We begin by analyzing some basic properties of shortest paths and a generic algorithm for the problem.
- We introduce and analyze **Dijkstra's algorithm** for shortest-paths problems with nonnegative weights.
- We conclude with the **Bellman–Ford** algorithm for edge-weighted digraphs with no negative cycles.

# Shortest Paths in an Edge-weighted Digraph

Given an edge-weighted digraph, find the shortest path from  $s$  to  $t$ .

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 |

## Variants

### ❖ Which vertices?

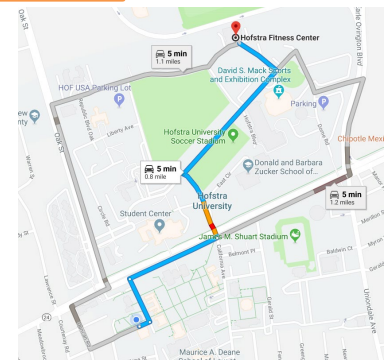
- Single source: from one vertex  $s$  to every other vertex.
- Source-sink: from one vertex  $s$  to another  $t$ .
- All pairs: between all pairs of vertices.

### ❖ Nonnegative weights?

### ❖ Cycles?

- Negative cycles.

Can we use BFS?



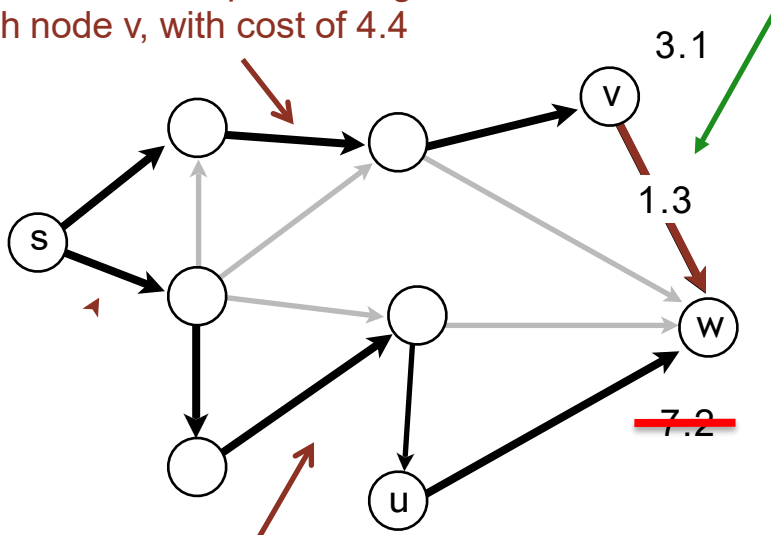
Simplifying assumption: Each vertex is reachable from  $s$ .

# Edge Relaxation

Relax edge  $e = v \rightarrow w$ . (basic of building SPT)

- $\text{distTo}[v]$  is length of shortest **known** path from  $s$  to  $v$ .
- $\text{distTo}[w]$  is length of shortest **known** path from  $s$  to  $w$ .
- $\text{prevNode}[w]$  is the previous node on shortest **known** path from  $s$  to  $w$ .
- If  $e = v \rightarrow w$  gives shorter path to  $w$  through  $v$ , update  $\text{distTo}[w]$  and  $\text{prevNode}[w]$ .
  - $\text{distTo}[w] = \min(\text{distTo}[w], \text{distTo}[v] + e.\text{weight}()); \text{prevNode}[w] = v$

After relaxing edge  $v \rightarrow w$ , the shortest path from  $s$  to  $w$  is updated to go through node  $v$ , with cost of 4.4



Previous shortest path from  $s$  to  $w$  goes through node  $u$ , with cost of 7.2

```
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();
        prevNode[w] = v;
    }
}
```

OLD  $\text{distTo}[w] = 7.2 > \text{distTo}[v] + e.\text{weight}() = 3.1 + 1.3 = 4.4$   
NEW  $\text{distTo}[w] \leftarrow \text{distTo}[v] + e.\text{weight}() = 4.4$ ,  
 $\text{prevNode}[w] = v$

# Generic Shortest-paths Algorithm

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## Generic algorithm (to compute SPT from $s$ )

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For each vertex  $v$ :  $\text{distTo}[v] = \infty$ .

For each vertex  $v$ :  $\text{prevNode}[v] = \text{null}$ .

$\text{distTo}[s] = 0$ .

Repeat until done:

- Relax any edge.

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**Proposition.** Generic algorithm computes SPT (if it exists) from  $s$ .

**Pf.**

- Throughout algorithm,  $\text{distTo}[v]$  is the length of a simple path from  $s$  to  $v$  (and  $\text{prevNode}[v]$  is its previous node on the path).
- Each successful relaxation decreases  $\text{distTo}[v]$  for some  $v$ .
- The entry  $\text{distTo}[v]$  can decrease at most a finite number of times.

**Efficient implementations.** How to choose which edge to relax?

- Ex 1. Dijkstra's algorithm. (**no negative weights**).
- Ex 2. Bellman–Ford algorithm. (**negative weights, no negative cycles**).



# Dijkstra's Algorithm

- Initialization:
  - Set the distance to the source node as 0 and to all other nodes as infinity.
  - Mark all nodes as unvisited and store them in a priority queue.
- Main Loop:
  - Visit the **unvisited node  $v$  with the shortest known distance** from the queue.
  - For each **unvisited neighbor node  $w$  of node  $v$** , calculate its tentative distance through the current node. **If this distance is smaller than the previously recorded distance, update it with edge relaxation for edge  $v-w$ .**
  - Mark the current node as visited once all its neighbors are processed.
- Termination:
  - The algorithm continues until all reachable nodes are visited, or until the shortest path to a specific destination is found.
- (Note: Dijkstra's algorithm works for both undirected and directed graphs. The only difference is the function for getting the neighbors of node  $v$ , which follows the edge arrow direction for directed graphs.)

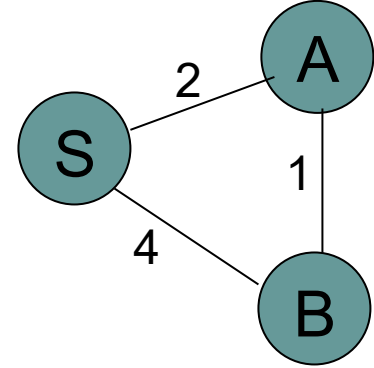
# Dijkstra's Algorithm: Correctness Proof

**Proposition.** Dijkstra's algorithm computes a SPT in any edge-weighted digraph with nonnegative weights.

**Pf.**

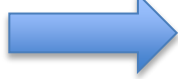
- Each edge  $e = v \rightarrow w$  is relaxed exactly once (when  $v$  is relaxed),
  - leaving  $\text{distTo}[w] \leq \text{distTo}[v] + e.\text{weight}()$ .
- Inequality holds until algorithm terminates because:
  - $\text{distTo}[w]$  cannot increase   $\text{distTo}[\ ]$  values are monotone decreasing
  - $\text{distTo}[v]$  will not change  we choose lowest  $\text{distTo}[\ ]$  value at each step (and edge weights are nonnegative)
- Thus, upon termination, shortest-paths optimality conditions hold.

Toy Example: Run Dijkstra's algorithm starting from source node S



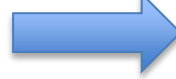
| N1 | SD       | PN |
|----|----------|----|
| S  | 0        |    |
| A  | $\infty$ |    |
| B  | $\infty$ |    |

Visit S



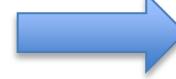
| N1 | SD | PN |
|----|----|----|
| S  | 0  |    |
| A  | 2  | S  |
| B  | 4  | S  |

Visit A

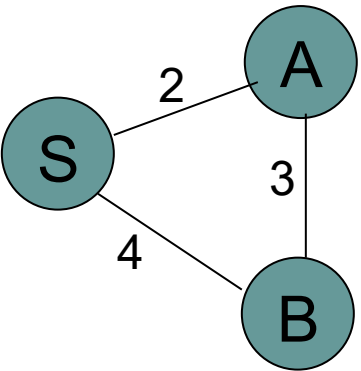


| N1 | SD | PN |
|----|----|----|
| S  | 0  |    |
| A  | 2  | S  |
| B  | 3  | A  |

Visit B

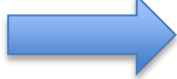


| N1 | SD | PN |
|----|----|----|
| S  | 0  |    |
| A  | 2  | S  |
| B  | 3  | A  |



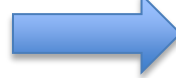
| N1 | SD       | PN |
|----|----------|----|
| S  | 0        |    |
| A  | $\infty$ |    |
| B  | $\infty$ |    |

Visit S



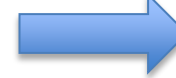
| N1 | SD | PN |
|----|----|----|
| S  | 0  |    |
| A  | 2  | S  |
| B  | 4  | S  |

Visit A



| N1 | SD | PN |
|----|----|----|
| S  | 0  |    |
| A  | 2  | S  |
| B  | 4  | S  |

Visit B



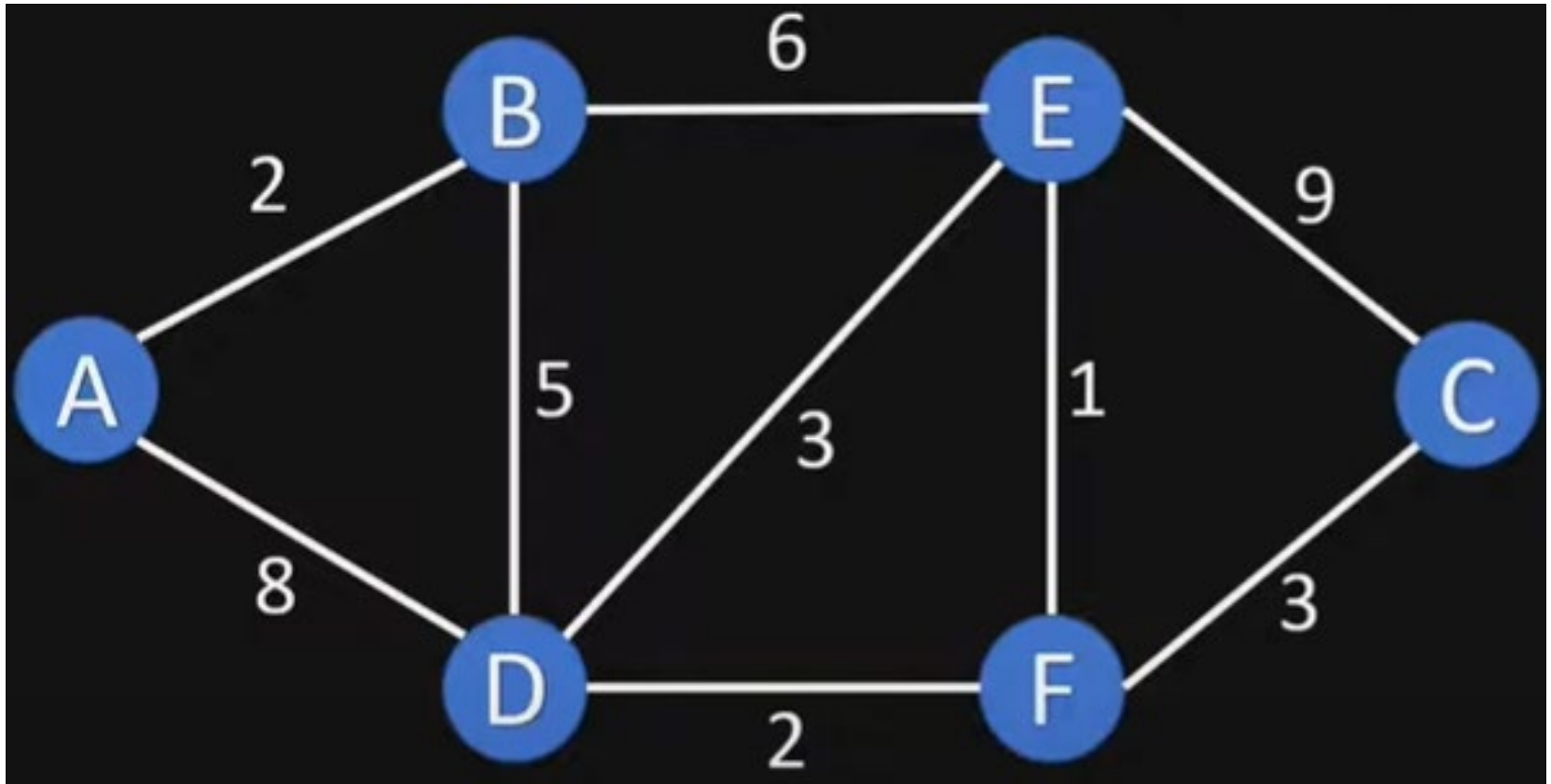
| N1 | SD | PN |
|----|----|----|
| S  | 0  |    |
| A  | 2  | S  |
| B  | 4  | S  |



# Video Tutorials

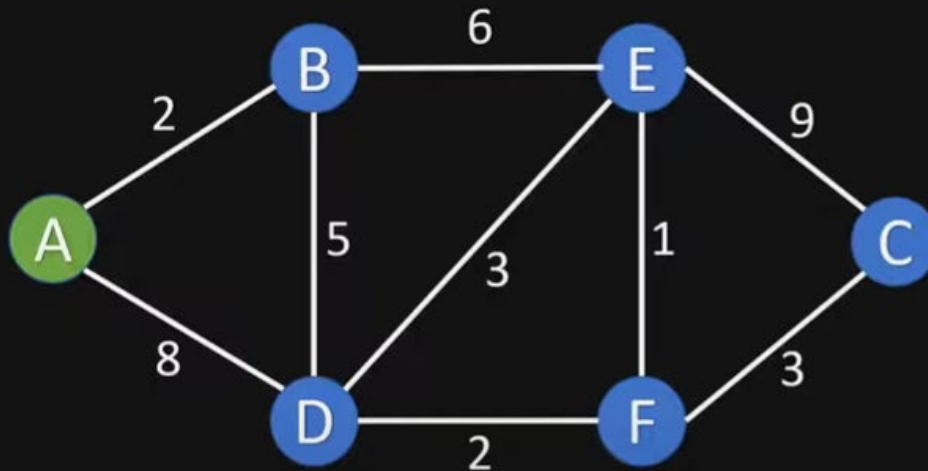
- Dijkstras Shortest Path Algorithm Explained | With Example | Graph Theory
  - <https://www.youtube.com/watch?v=bZkzH5x0SKU>
  - The following lecture slides are based on this video
- Dijkstra's algorithm in 3 minutes
  - [https://www.youtube.com/watch?v=\\_lHSawdgXpI](https://www.youtube.com/watch?v=_lHSawdgXpI)
- Bellman-Ford in 4 minutes — Theory
  - <https://www.youtube.com/watch?v=9PHkk0UavIM>
- Bellman-Ford in 5 minutes — Step by step example
  - <https://www.youtube.com/watch?v=obWXjtg0L64>

# Example Graph



# Initialize

2. Assign to all nodes a tentative distance value



Visited Nodes: []

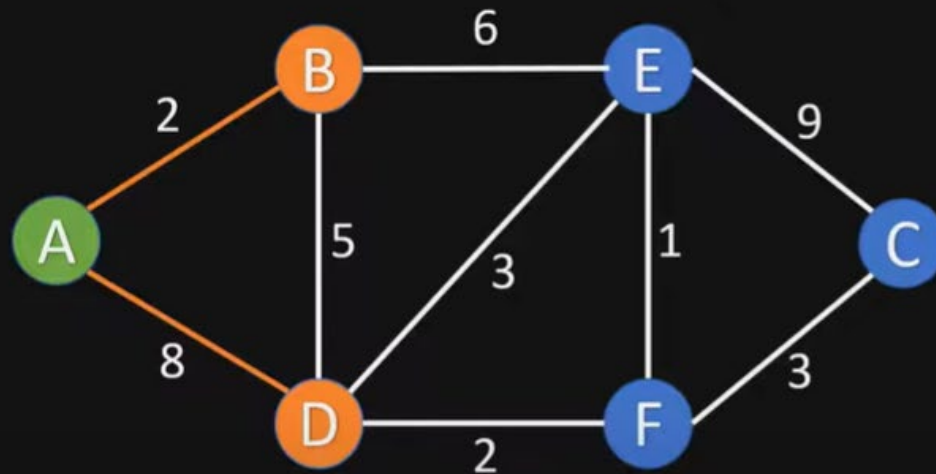
Unvisited Nodes: [A, B, C, D, E, F]

| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | $\infty$          |               |
| C    | $\infty$          |               |
| D    | $\infty$          |               |
| E    | $\infty$          |               |
| F    | $\infty$          |               |

# Visit Node A

3. For the current node calculate the distance to all unvisited neighbours

3.1. Update shortest distance, if new distance is shorter than old distance



Visited Nodes: []

Unvisited Nodes: [A, B, C, D, E, F]

| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | 2                 | A             |
| C    | $\infty$          |               |
| D    | 8                 | A             |
| E    | $\infty$          |               |
| F    | $\infty$          |               |

OLD  $\text{distTo}[B] = \infty > \text{distTo}[A] + e[A][B].\text{weight}() = 0 + 2 = 2$

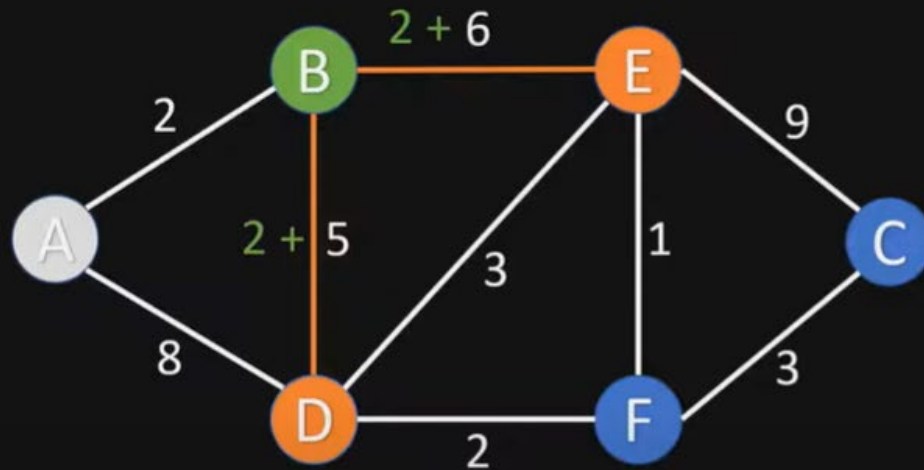
NEW  $\text{distTo}[B] \leftarrow \text{distTo}[A] + e[A][B].\text{weight}() = 2$ ,  $\text{prevNode}[B] = A$

OLD  $\text{distTo}[D] = \infty > \text{distTo}[A] + e[A][D].\text{weight}() = 0 + 8 = 8$

NEW  $\text{distTo}[D] \leftarrow \text{distTo}[A] + e[A][D].\text{weight}() = 8$ ,  $\text{prevNode}[D] = A$

# Visit Node B

3. For the current node calculate the distance to all unvisited neighbours  
3.1. Update shortest distance, if new distance is shorter than old distance



| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | 2                 | A             |
| C    | $\infty$          |               |
| D    | 7                 | B             |
| E    | 8                 | B             |
| F    | $\infty$          |               |

OLD  $\text{distTo}[D] = 8 > \text{distTo}[B] + e[B][D].\text{weight}() = 2 + 5 = 7$

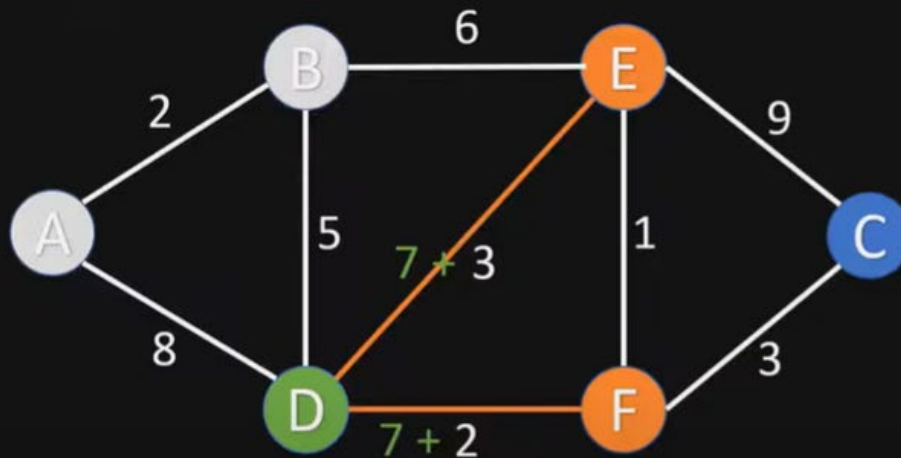
NEW  $\text{distTo}[D] \leftarrow \text{distTo}[B] + e[B][D].\text{weight}() = 7$ ,  $\text{prevNode}[D] = B$

OLD  $\text{distTo}[E] = \infty > \text{distTo}[B] + e[B][E].\text{weight}() = 2 + 6 = 8$

NEW  $\text{distTo}[E] \leftarrow \text{distTo}[B] + e[B][E].\text{weight}() = 8$ ,  $\text{prevNode}[E] = B$

# Visit Node D

3. For the current node calculate the distance to all unvisited neighbours  
3.1. Update shortest distance, if new distance is shorter than old distance



| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | 2                 | A             |
| C    | $\infty$          |               |
| D    | 7                 | B             |
| E    | 8                 | B             |
| F    | 9                 | D             |

OLD  $\text{distTo}[E] = 8 < \text{distTo}[D] + e[D][E].\text{weight}() = 7 + 3 = 10$

No update,  $\text{distTo}[E]$  stays 8,  $\text{prevNode}[E]$  stays B

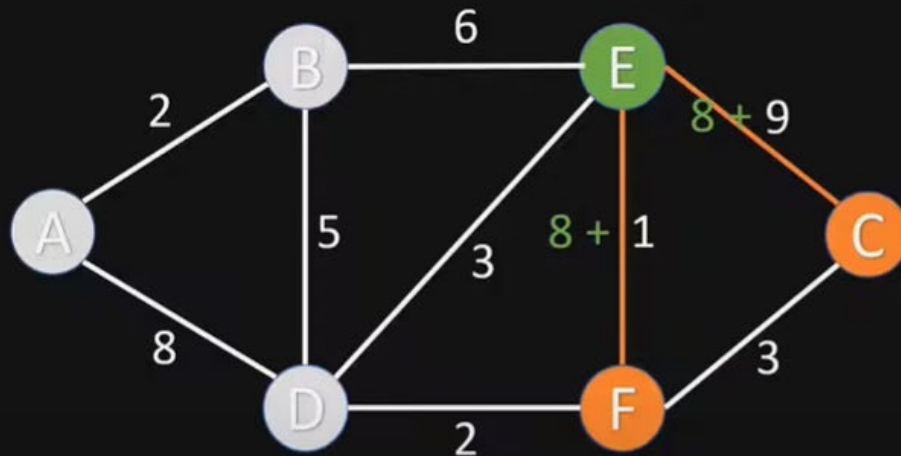
OLD  $\text{distTo}[F] = \infty > \text{distTo}[D] + e[D][F].\text{weight}() = 7 + 2 = 9$

NEW  $\text{distTo}[F] \leftarrow \text{distTo}[D] + e[D][F].\text{weight}() = 9$ ,  $\text{prevNode}[F] = D$

# Visit Node E

3. For the current node calculate the distance to all unvisited neighbours

3.1. Update shortest distance, if new distance is shorter than old distance



Visited Nodes: [A, B, D] Unvisited Nodes: [C, E, F]

| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | 2                 | A             |
| C    | 17                | E             |
| D    | 7                 | B             |
| E    | 8                 | B             |
| F    | 9                 | D             |

OLD  $\text{distTo}[C] = \infty > \text{distTo}[E] + e[E][C].\text{weight}() = 8 + 9 = 17$

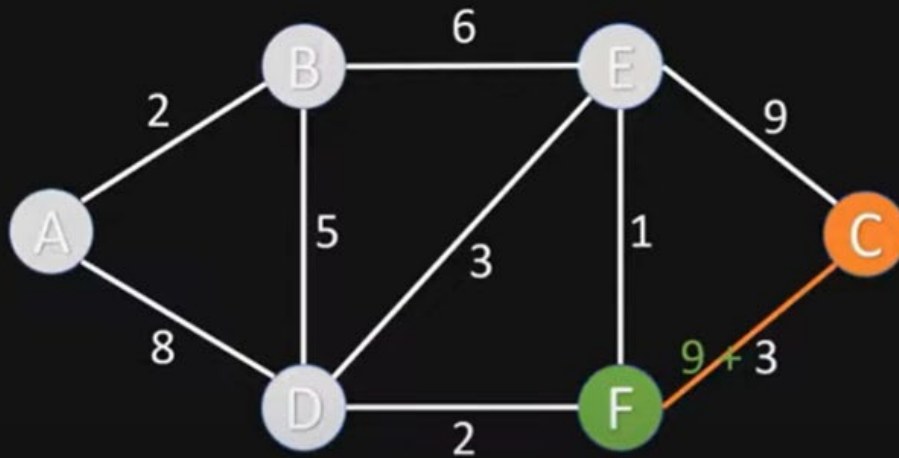
NEW  $\text{distTo}[C] \leftarrow \text{distTo}[E] + e[E][C].\text{weight}() = 17$ ,  $\text{prevNode}[C] = E$

OLD  $\text{distTo}[F] = 9 = \text{distTo}[E] + e[E][F].\text{weight}() = 8 + 1 = 9$

No update,  $\text{distTo}[F]$  stays 9,  $\text{prevNode}[F] = D$  (You can also update  $\text{prevNode}[F] = E$ .)

# Visit Node F

3. For the current node calculate the distance to all unvisited neighbours  
3.1. Update shortest distance, if new distance is shorter than old distance



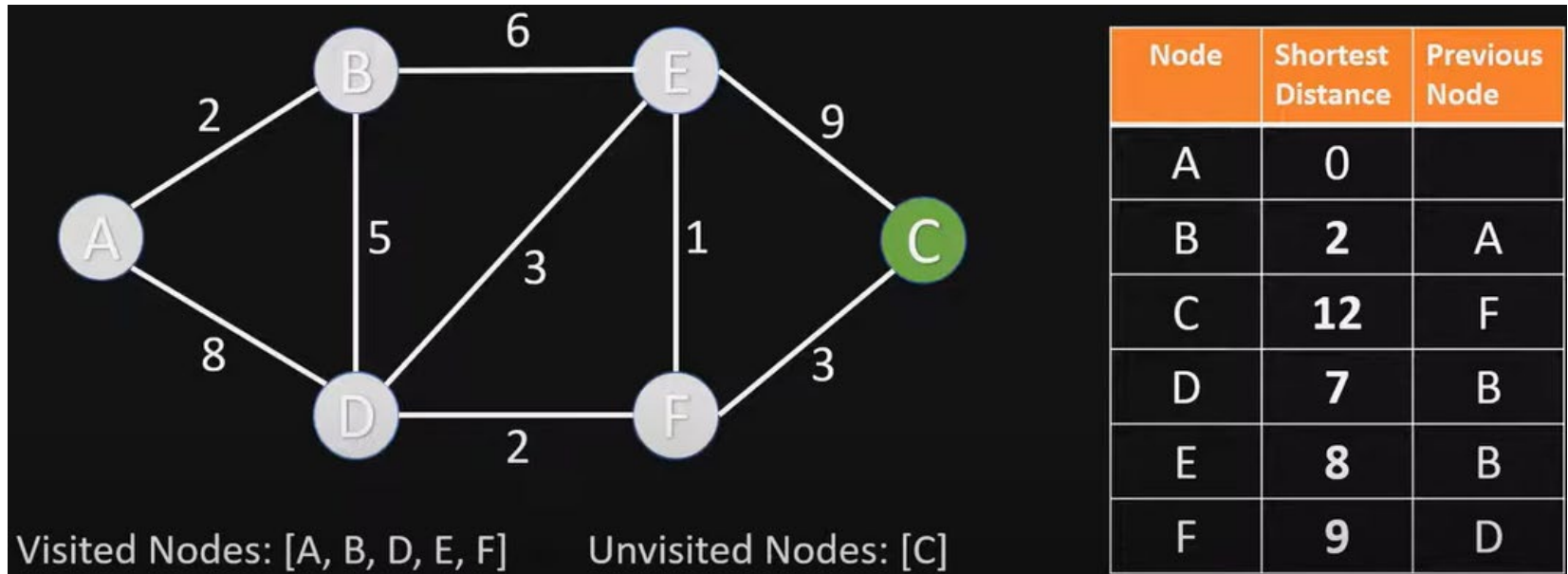
Visited Nodes: [A, B, D, E]    Unvisited Nodes: [C, F]

| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | 2                 | A             |
| C    | 12                | F             |
| D    | 7                 | B             |
| E    | 8                 | B             |
| F    | 9                 | D             |

OLD  $\text{distTo}[C] = 17 > \text{distTo}[F] + e[F][C].\text{weight}() = 9 + 3 = 12$   
NEW  $\text{distTo}[C] \leftarrow \text{distTo}[F] + e[F][C].\text{weight}() = 12$ ,  $\text{prevNode}[C] = F$



# Visit Node C

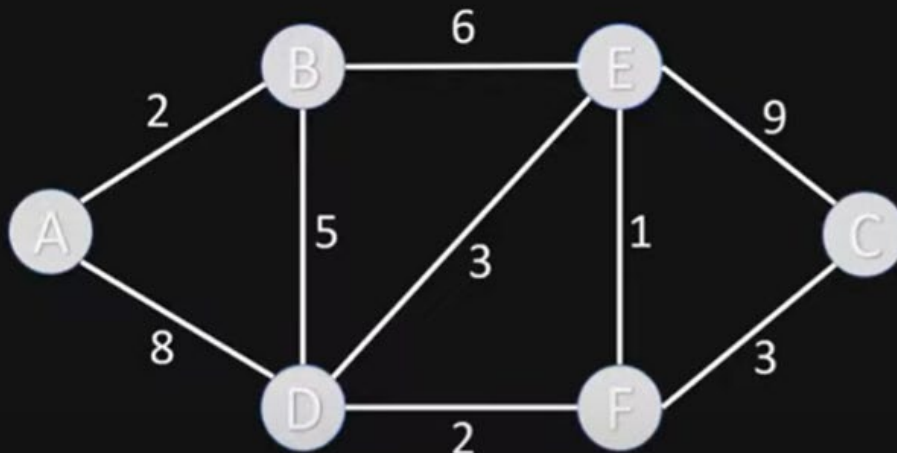


Nothing changes, since C has no unvisited neighbor nodes

# End of Algorithm

- Table contains the shortest distance to each node N from the source node A, and its previous node in the shortest path

4. Mark current node as visited



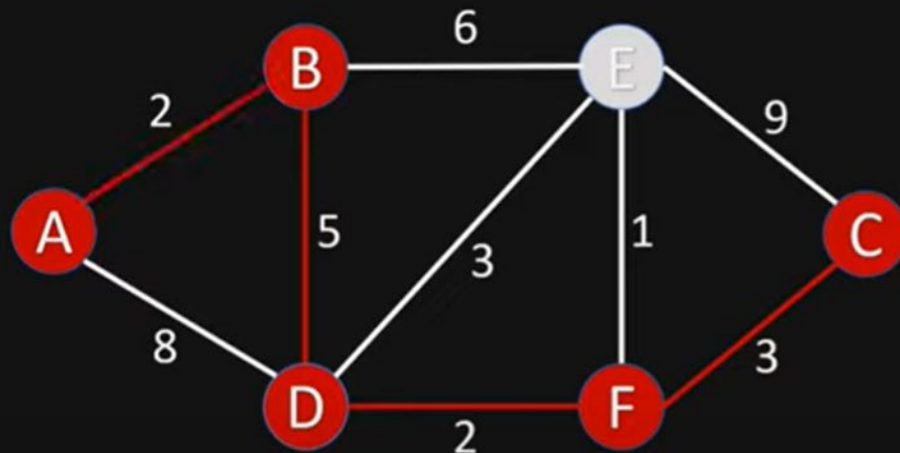
Visited Nodes: [A, B, D, E, F, C] Unvisited Nodes: []

| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | 2                 | A             |
| C    | 12                | F             |
| D    | 7                 | B             |
| E    | 8                 | B             |
| F    | 9                 | D             |

# Getting the Shortest Path from A to C

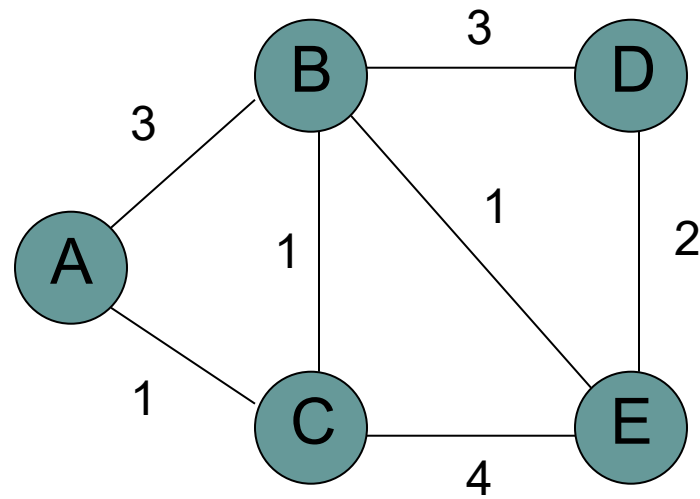
- C's previous node is F; F's previous node is D; D's previous node is B; B's previous node is A
- Shortest Path from A to C is ABDFC

Get shortest path from A to C

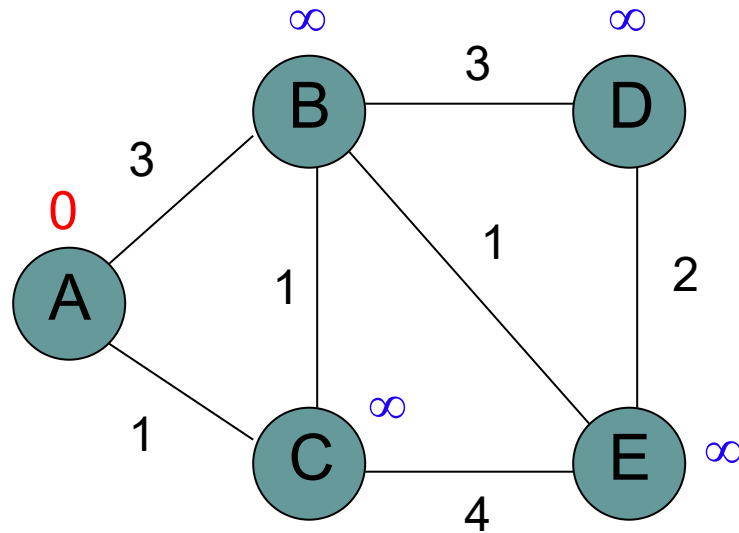


| Node | Shortest Distance | Previous Node |
|------|-------------------|---------------|
| A    | 0                 |               |
| B    | 2                 | A             |
| C    | 12                | F             |
| D    | 7                 | B             |
| E    | 8                 | B             |
| F    | 9                 | D             |

# Dijkstra's Algorithm Example 2

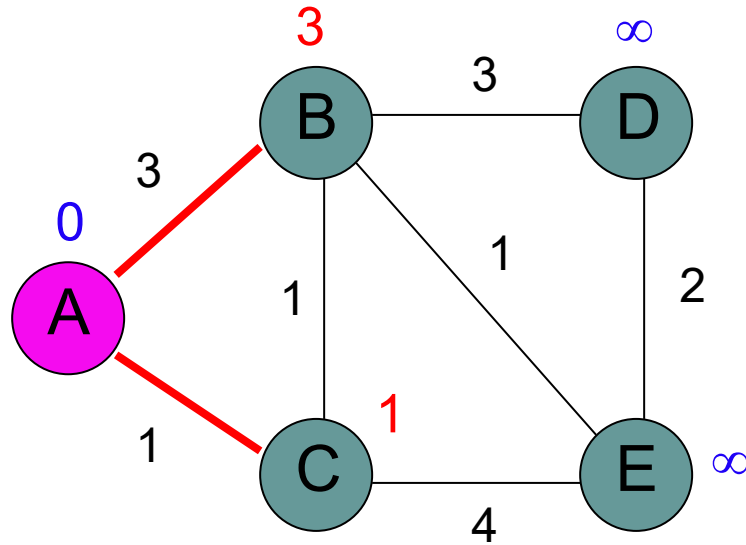


# Initialize



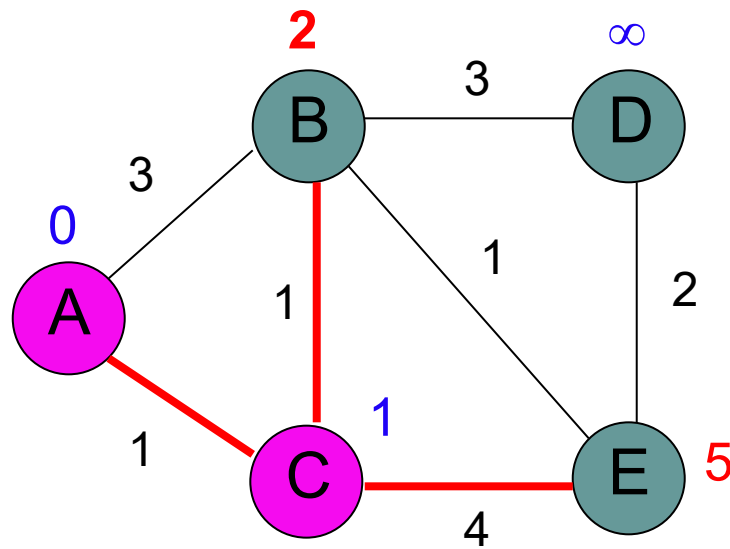
| N | SD       | PN |
|---|----------|----|
| A | 0        |    |
| B | $\infty$ |    |
| C | $\infty$ |    |
| D | $\infty$ |    |
| E | $\infty$ |    |

# Visit Node A



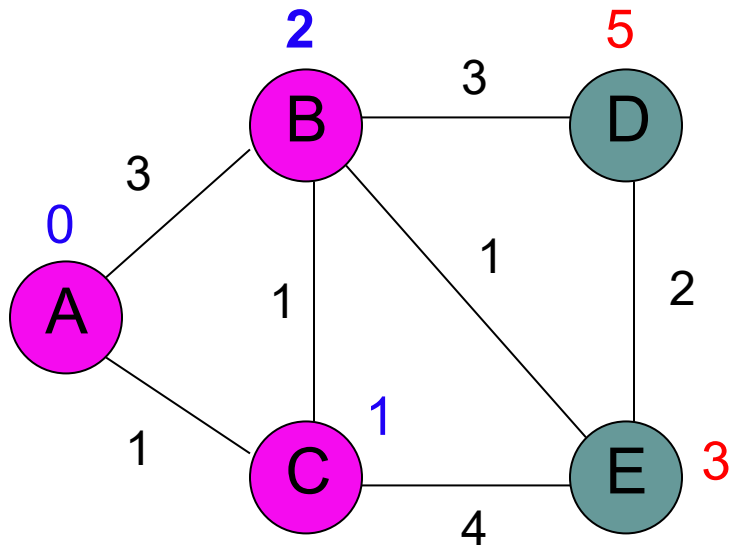
| N | SD       | PN |
|---|----------|----|
| A | 0        |    |
| B | 3        | A  |
| C | 1        | A  |
| D | $\infty$ |    |
| E | $\infty$ |    |

# Visit Node C



| N | SD | PN |
|---|----|----|
| A | 0  |    |
| B | 2  | C  |
| C | 1  | A  |
| D | ∞  |    |
| E | 5  | C  |

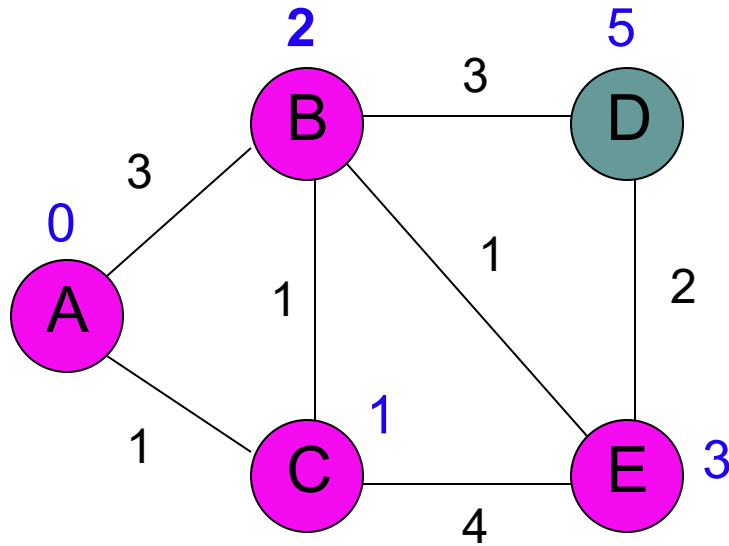
# Visit Node B



| N | SD | PN |
|---|----|----|
| A | 0  |    |
| B | 2  | C  |
| C | 1  | A  |
| D | 5  | B  |
| E | 3  | B  |



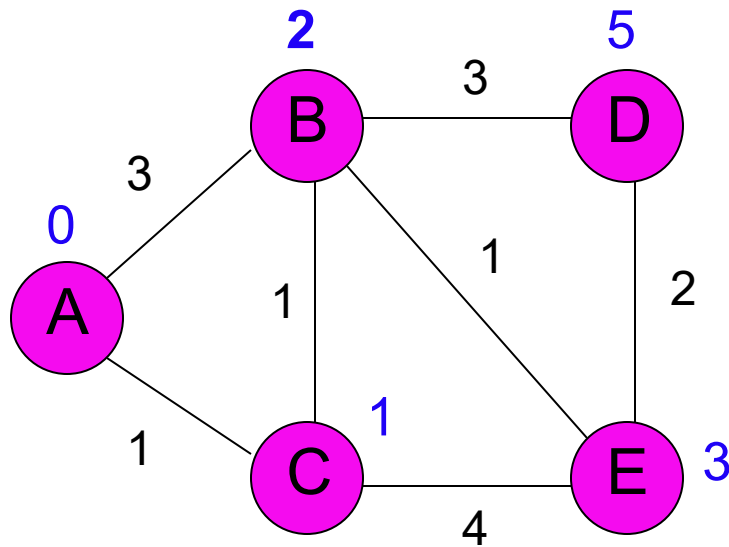
# Visit Node E



| N | SD | PN |
|---|----|----|
| A | 0  |    |
| B | 2  | C  |
| C | 1  | A  |
| D | 5  | B  |
| E | 3  | B  |

Nothing changes

# Visit Node D



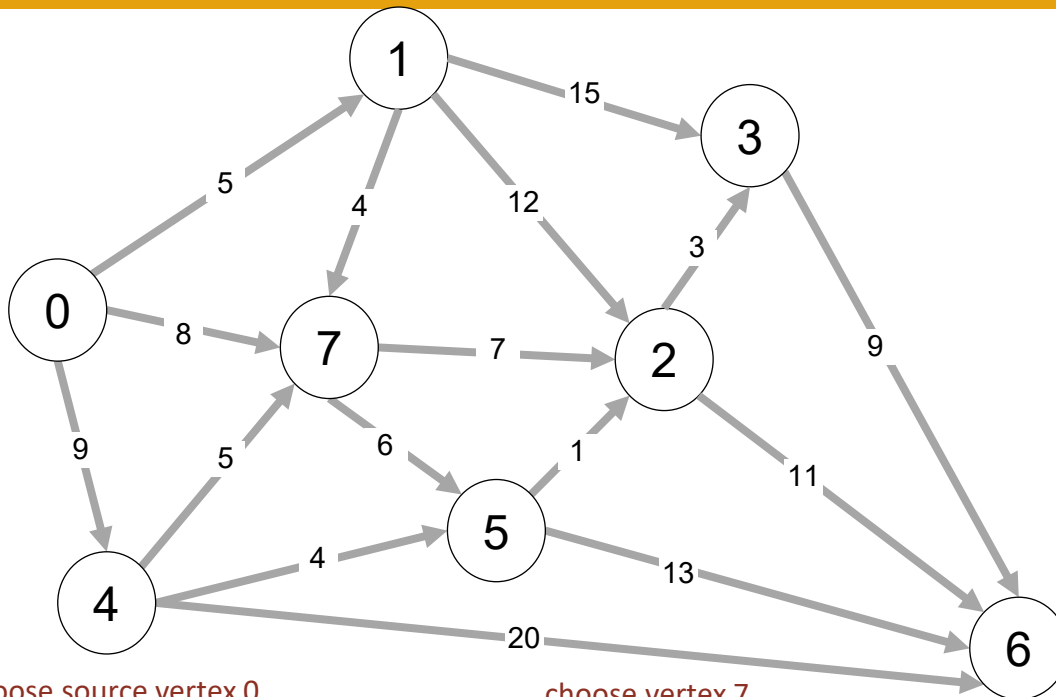
| N | SD | PN |
|---|----|----|
| A | 0  |    |
| B | 2  | C  |
| C | 1  | A  |
| D | 5  | B  |
| E | 3  | B  |

Nothing changes

# Dijkstra's Algorithm Example 3

- 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 pointing from that vertex.

choose vertex 5  
 relax all edges adjacent from 5  
 choose vertex 2  
 relax all edges adjacent from 2  
 choose vertex 3  
 relax all edges adjacent from 3  
 choose vertex 6  
 relax all edges adjacent from 6



choose source vertex 0  
 relax all edges adjacent from 0  
 choose vertex 1  
 relax all edges adjacent from 1

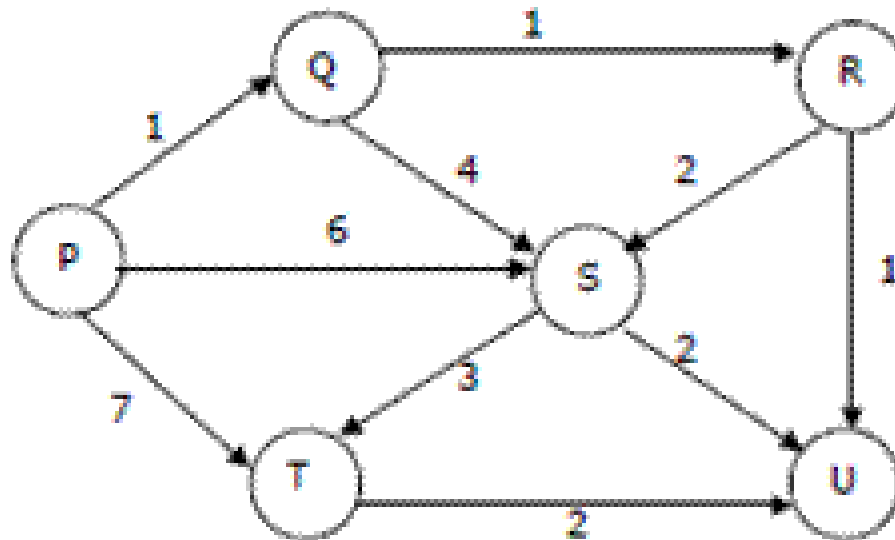
choose vertex 7  
 relax all edges adjacent from 7  
 choose vertex 4  
 relax all edges adjacent from 4

| v distTo[ ] |              |               |                  |
|-------------|--------------|---------------|------------------|
| → 0         | <del>∞</del> | 0             |                  |
| → 1         | <del>∞</del> | 5             |                  |
| → 2         | <del>∞</del> | <del>17</del> | <del>15</del> 14 |
| → 3         | <del>∞</del> | <del>20</del> | 17               |
| → 4         | <del>∞</del> | 9             |                  |
| → 5         | <del>∞</del> | <del>14</del> | 13               |
| → 6         | <del>∞</del> | <del>29</del> | <del>26</del> 25 |
| → 7         | <del>∞</del> | 8             |                  |

| v edgeTo[ ] |              |              |                |
|-------------|--------------|--------------|----------------|
| 0           | -            |              |                |
| 1           | <del>-</del> | 0            |                |
| 2           | <del>-</del> | <del>1</del> | <del>7</del> 5 |
| 3           | <del>-</del> | <del>1</del> | 2              |
| 4           | <del>-</del> | 0            |                |
| 5           | <del>-</del> | <del>7</del> | 4              |
| 6           | <del>-</del> | <del>4</del> | <del>5</del> 2 |
| 7           | <del>-</del> | 0            |                |

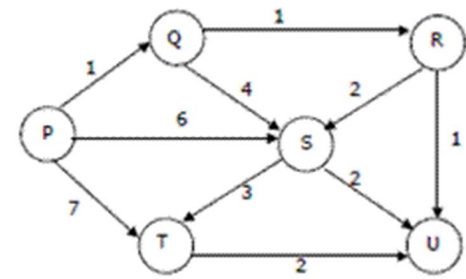
# Dijkstra's Algorithm Example 4

- Suppose we run Dijkstra's single source shortest-path algorithm on the following edge weighted directed graph with vertex P as the source. In what order do the nodes get included into the set of vertices for which the shortest path distances are finalized?
- ANS: P, Q, R, U, S, T



# SD: Shortest Distance

## PN: Previous Node



| N | SD       | PN |
|---|----------|----|
| P | 0        |    |
| Q | $\infty$ |    |
| R | $\infty$ |    |
| S | $\infty$ |    |
| T | $\infty$ |    |
| U | $\infty$ |    |

Visit P  
→

| N | SD       | PN |
|---|----------|----|
| P | 0        |    |
| Q | 1        | P  |
| R | $\infty$ |    |
| S | 6        | P  |
| T | 7        | P  |
| U | $\infty$ |    |

Visit Q  
→

| N | SD       | PN |
|---|----------|----|
| P | 0        |    |
| Q | 1        | P  |
| R | 2        | Q  |
| S | 5        | Q  |
| T | 7        | P  |
| U | $\infty$ |    |

Visit R  
→

| N | SD | PN |
|---|----|----|
| P | 0  |    |
| Q | 1  | P  |
| R | 2  | Q  |
| S | 4  | Q  |
| T | 7  | P  |
| U | 3  | R  |

← Visit U (nothing changes)

| N | SD | PN |
|---|----|----|
| P | 0  |    |
| Q | 1  | P  |
| R | 2  | Q  |
| S | 4  | Q  |
| T | 7  | P  |
| U | 3  | R  |

Visit S  
(nothing changes)  
→

| N | SD | PN |
|---|----|----|
| P | 0  |    |
| Q | 1  | P  |
| R | 2  | Q  |
| S | 4  | Q  |
| T | 7  | P  |
| U | 3  | R  |

Visit T  
(nothing changes)  
→

| N | SD | PN |
|---|----|----|
| P | 0  |    |
| Q | 1  | P  |
| R | 2  | Q  |
| S | 4  | Q  |
| T | 7  | P  |
| U | 3  | R  |

Finished  
→

| N | SD | PN |
|---|----|----|
| P | 0  |    |
| Q | 1  | P  |
| R | 2  | Q  |
| S | 4  | Q  |
| T | 7  | P  |
| U | 3  | R  |

# Bellman-Ford Algorithm

- Initialize distance array `distTo[]` for each vertex `v` as **`distTo[v] = ∞`**, and **`distTo[s] = 0`** to source vertex `s`.
- Relax all **edges**  $|V|-1$  times.

```
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;
    }
}
```

## Recall:

### Generic algorithm (to compute SPT from `s`)

For each vertex `v`: `distTo[v] = ∞`.

For each vertex `v`: `edgeTo[v] = null`.

`distTo[s] = 0`.

Repeat until done:

- Relax any edge.

### Bellman-Ford algorithm

For each vertex `v`: `distTo[v] = ∞`.

For each vertex `v`: `edgeTo[v] = null`.

`distTo[s] = 0`.

Repeat  $|V| - 1$  times:

- Relax each edge.

# Bellman-Ford Algorithm Proof of Correctness

- Relaxing edges  $|V|-1$  times in the Bellman-Ford Algorithm guarantees that the algorithm has explored all possible paths of length up to  $|V|-1$ , which is the maximum possible length of a shortest path in a graph with  $|V|$  vertices. This allows the algorithm to correctly calculate the shortest paths from the source vertex to all other vertices, given that there are no negative-weight cycles.

# Bellman-Ford Algorithm with Negative Cycle Detection

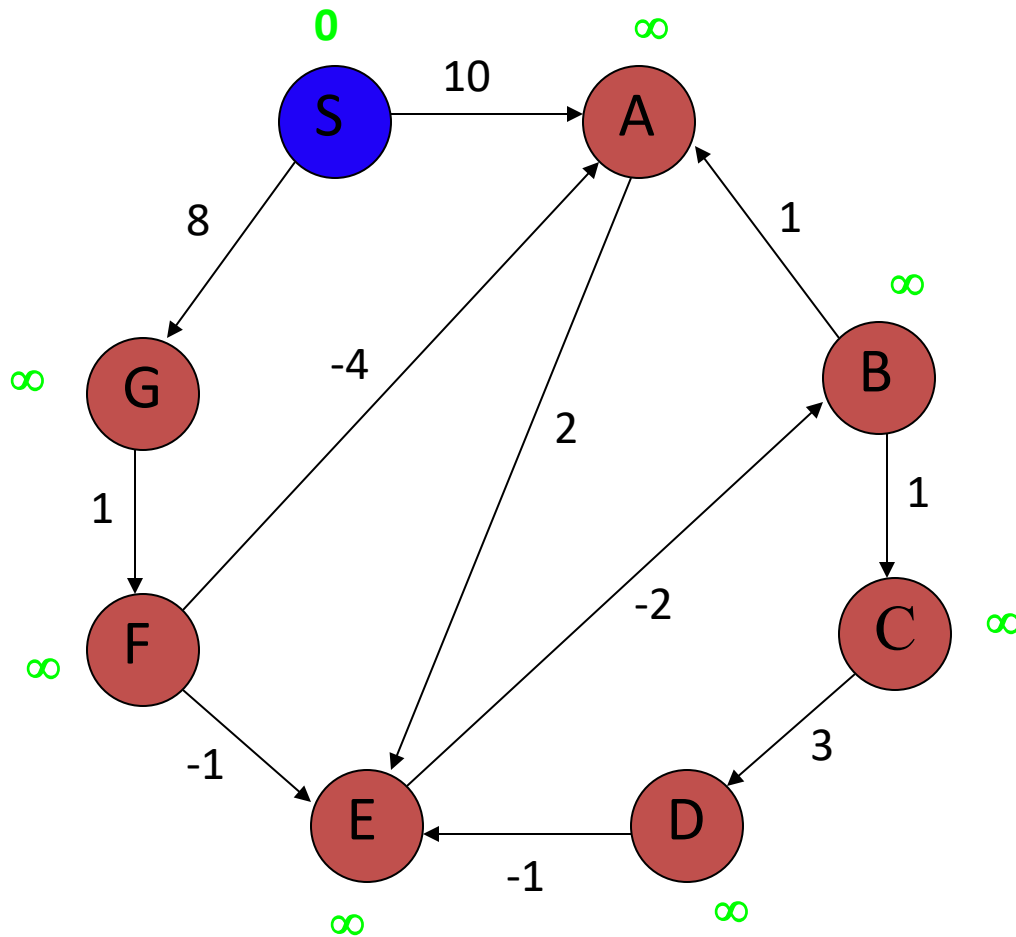
- Initialize distance array `distTo[]` for each vertex `v` as **`distTo[v] = ∞`**, and **`distTo[s] = 0`** to source vertex `s`.
- Relax all **edges** `|V|-1` times.
- Relax all the edges one more time i.e. the **N-th** time:
  - Case 1 (Negative cycle exists): if any edge can be further relaxed, i.e., for any **edge e**, if **`distTo[w] > distTo[v] + e.weight()`**
  - Case 2 (No Negative cycle) : case 1 fails for all the edges.



# Time Complexity of Bellman-Ford Algorithm

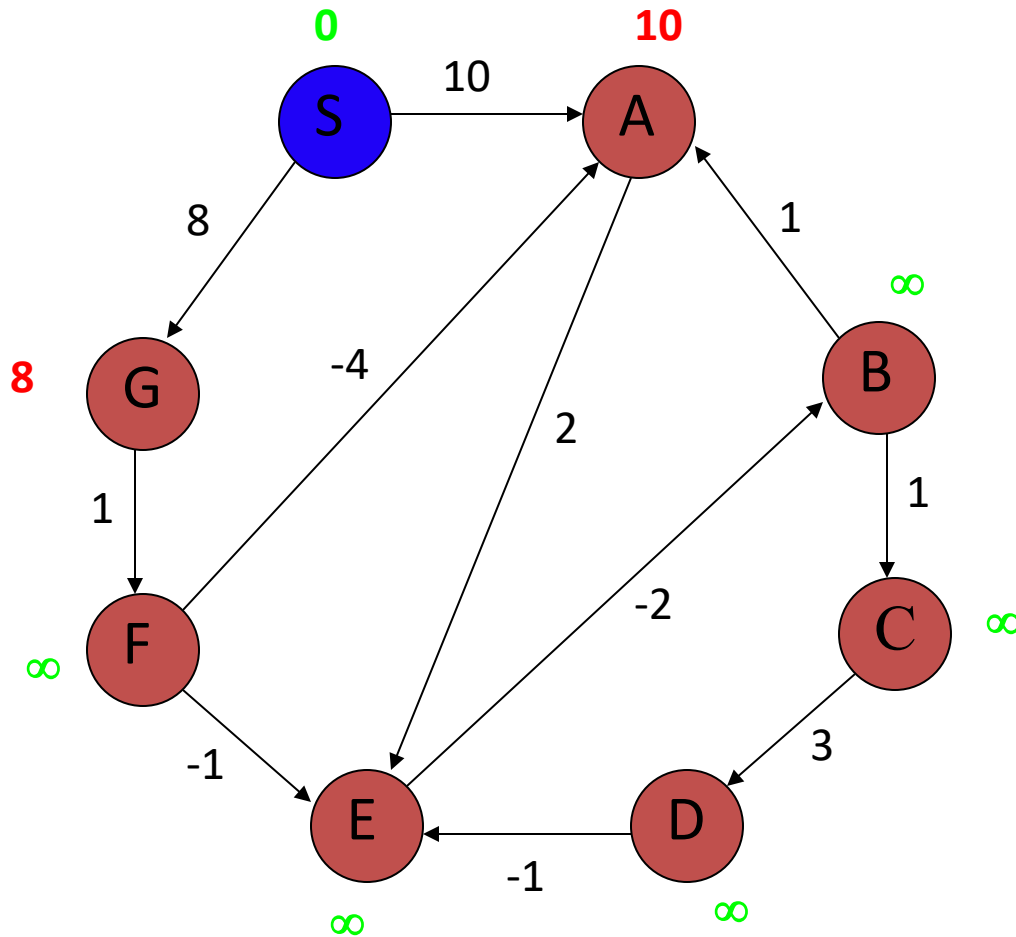
- Time complexity for connected graph:
- Best Case:  $O(|E|)$ , when distance array after 1st and 2nd relaxation are same, we can simply stop further processing after one iteration
- Average Case:  $O(|V|*|E|)$
- Worst Case:  $O(|V|*|E|)$ 
  - If the graph is complete, the value of  $E$  becomes  $O(|V|^2)$ . So overall time complexity becomes  $O(|V|^3)$

# Bellman-Ford Algorithm Example 1



Iteration: 0

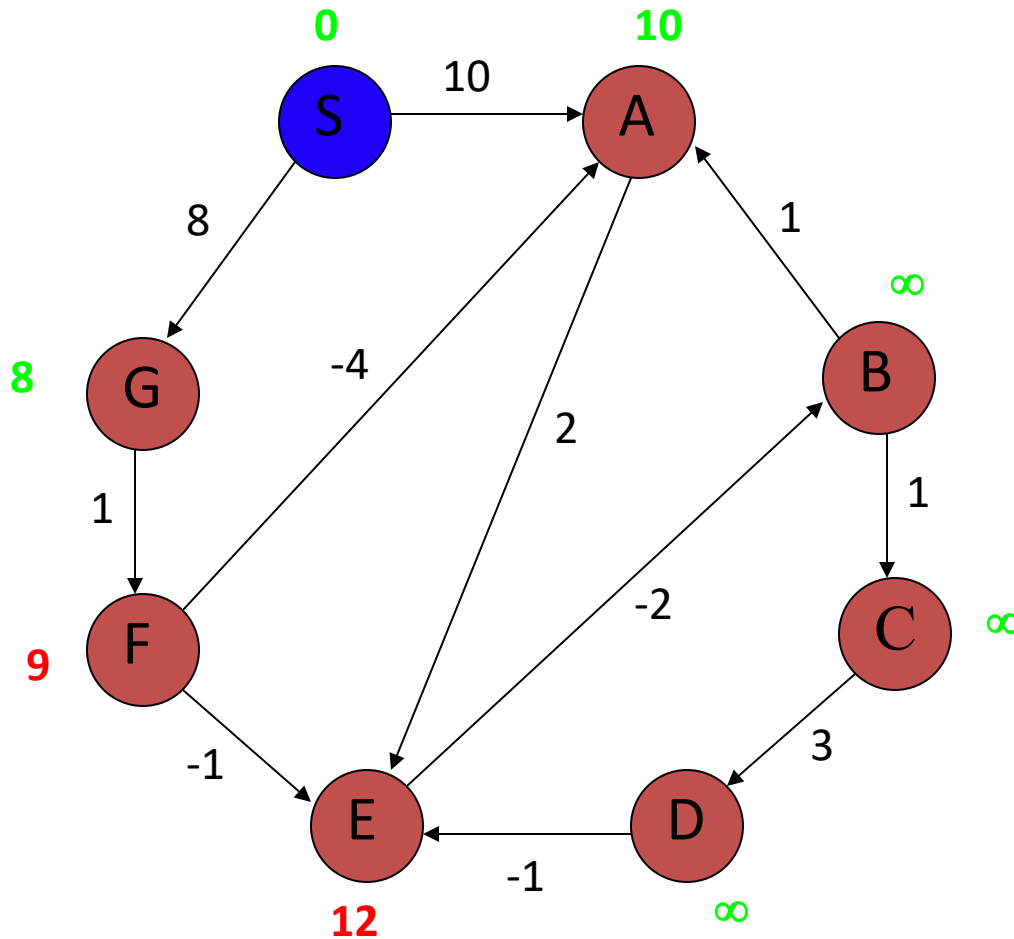
# Bellman-Ford Algorithm Example 1



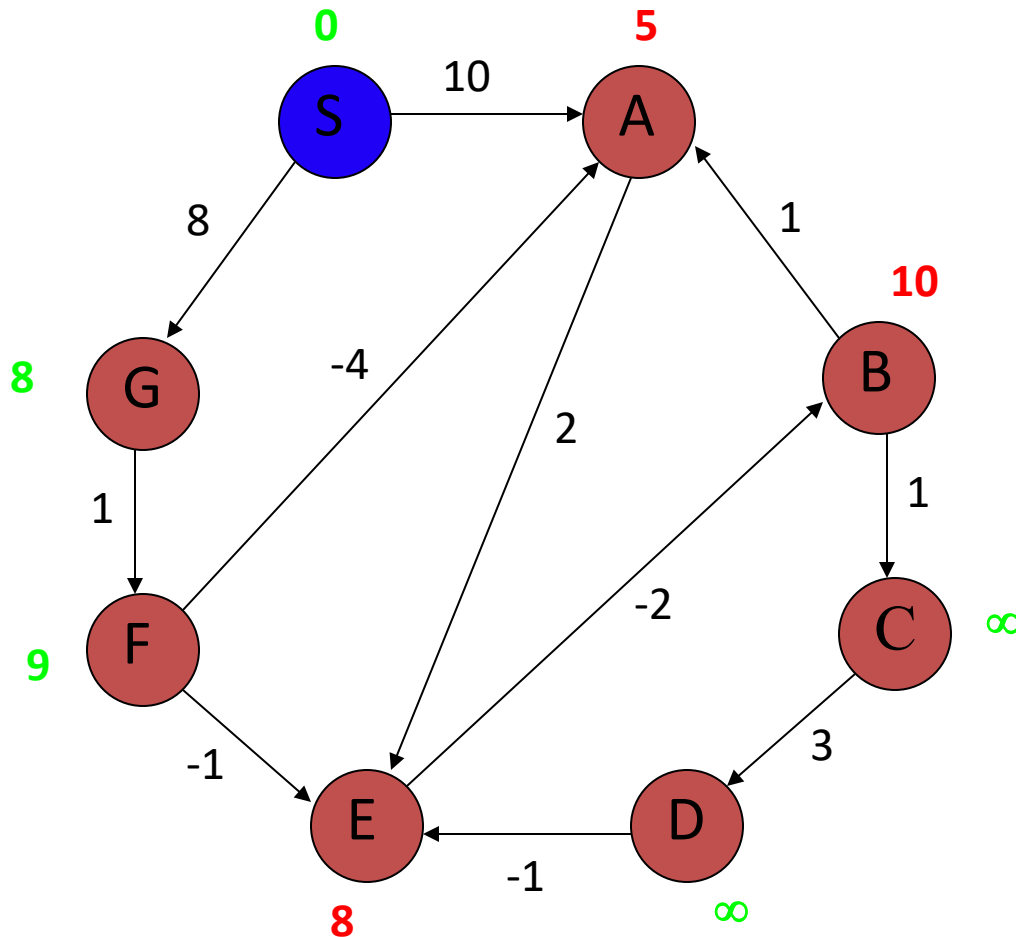
Iteration: 1

# Bellman-Ford Algorithm Example 1

Iteration: 2



# Bellman-Ford Algorithm Example 1

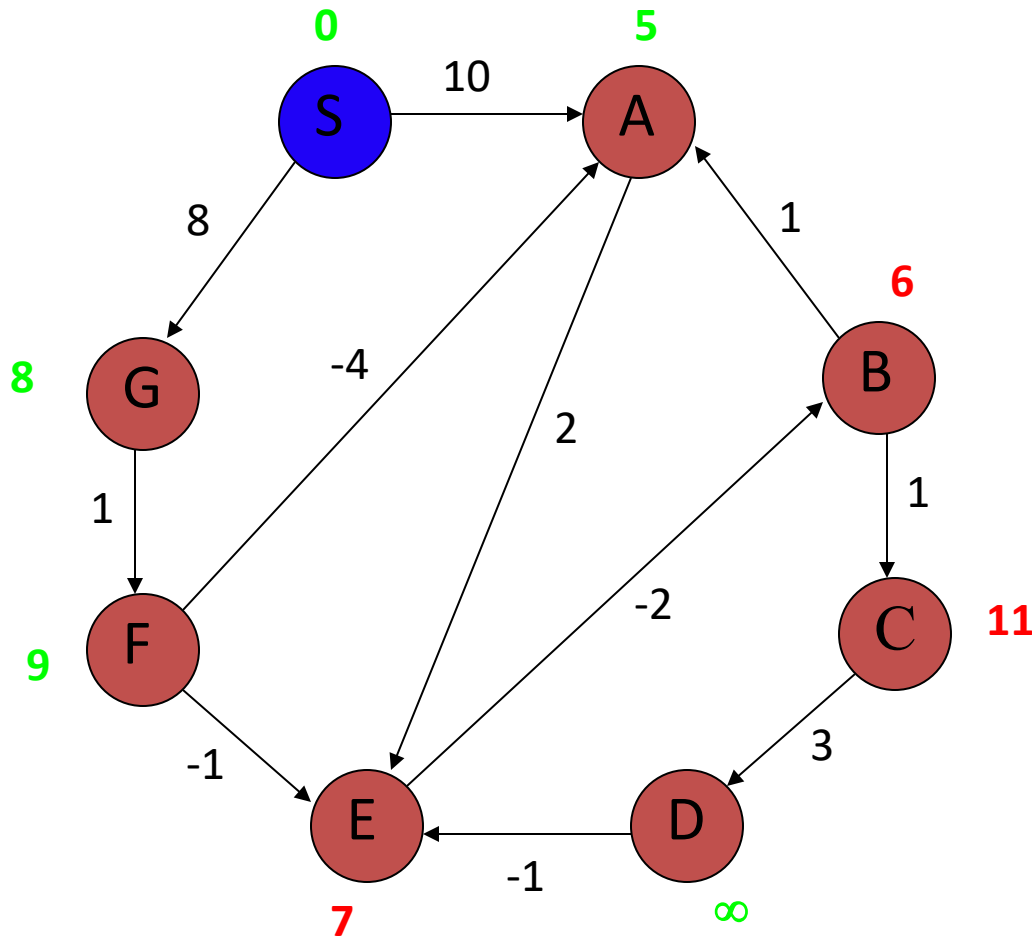


Iteration: 3

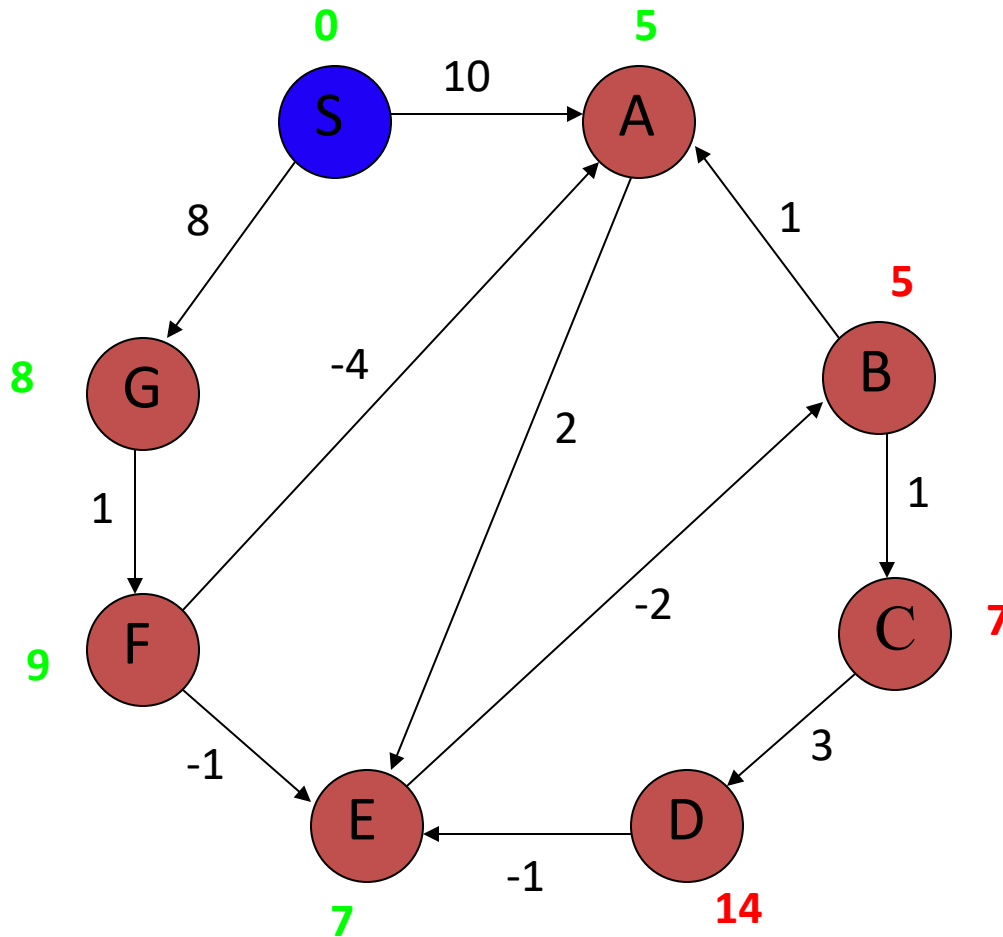
A has the correct  
distance and path

# Bellman-Ford Algorithm Example 1

Iteration: 4



# Bellman-Ford Algorithm Example 1

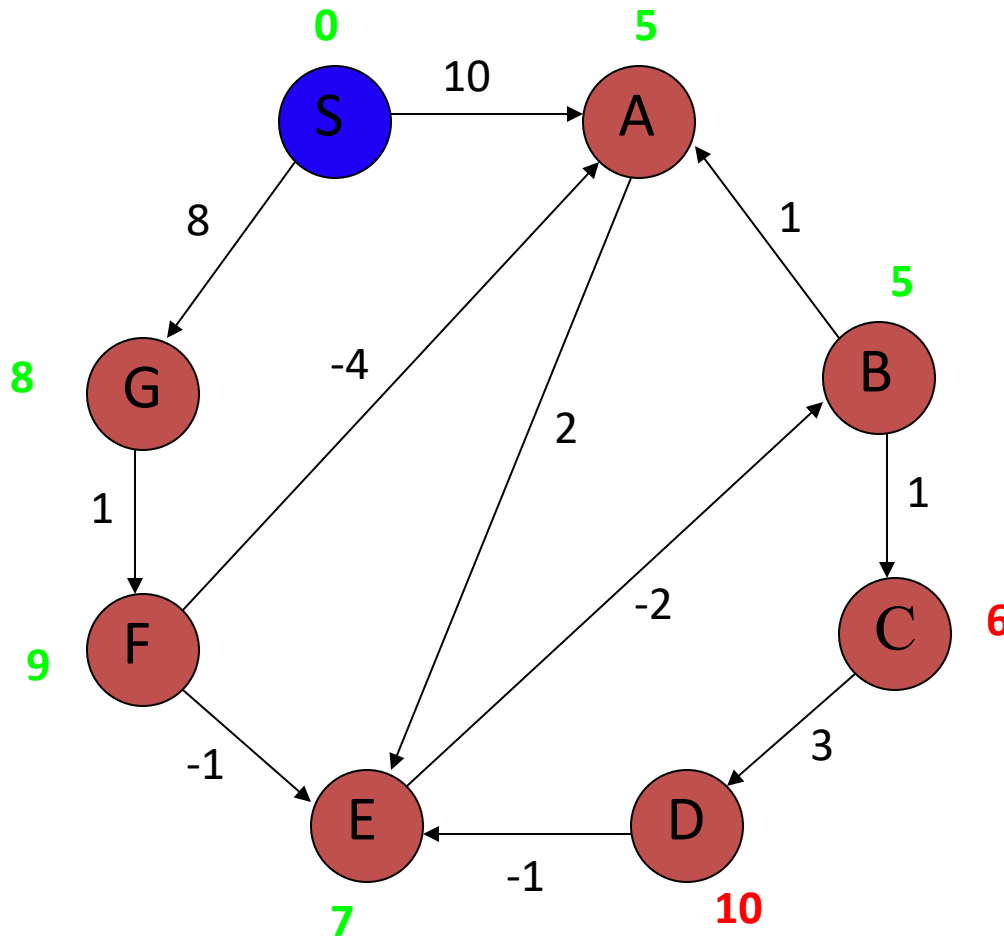


Iteration: 5

B has the correct  
distance and path

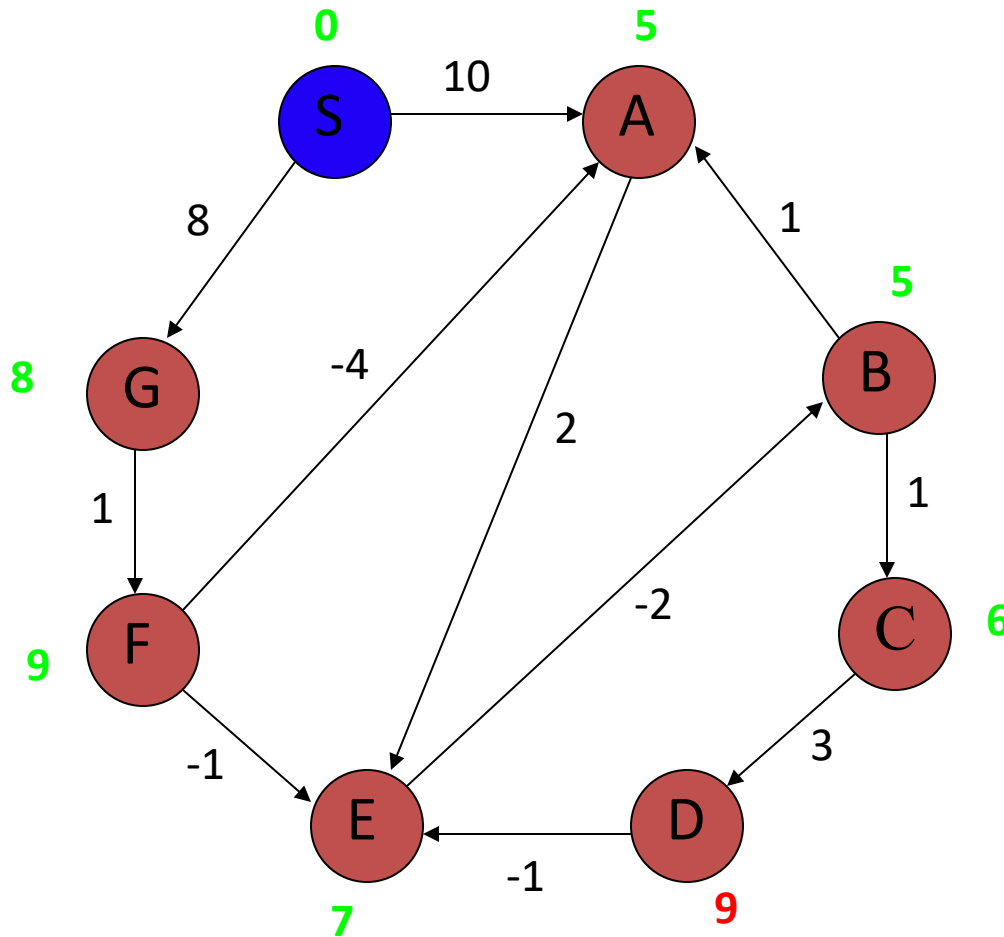
# Bellman-Ford Algorithm Example 1

Iteration: 6





# Bellman-Ford Algorithm Example 1

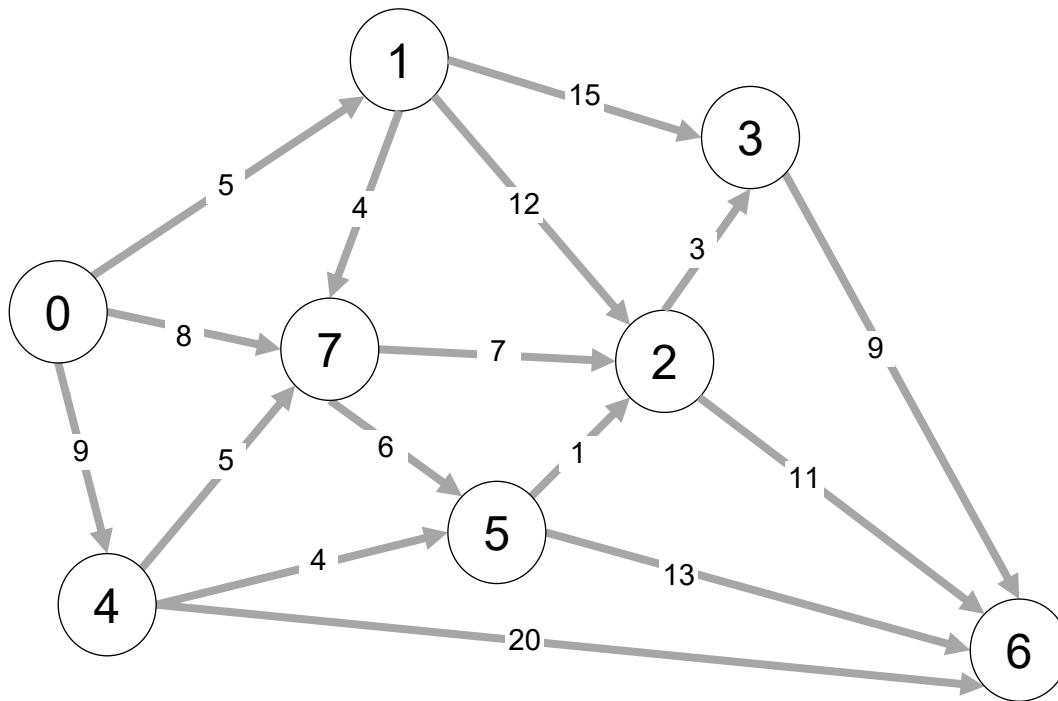


Iteration: 7

D (and all other nodes) have the correct distance and path

# Bellman-Ford Algorithm Example 2

Repeat  $V - 1$  times: relax all  $E$  edges.



| v | distTo[]     |               |                  |
|---|--------------|---------------|------------------|
| 0 | <del>∞</del> | 0             |                  |
| 1 | <del>∞</del> | 5             |                  |
| 2 | <del>∞</del> | <del>17</del> | 14               |
| 3 | <del>∞</del> | <del>20</del> | 17               |
| 4 | <del>∞</del> | 9             |                  |
| 5 | <del>∞</del> | 13            |                  |
| 6 | <del>∞</del> | <del>28</del> | <del>26</del> 25 |
| 7 | <del>∞</del> | 8             |                  |

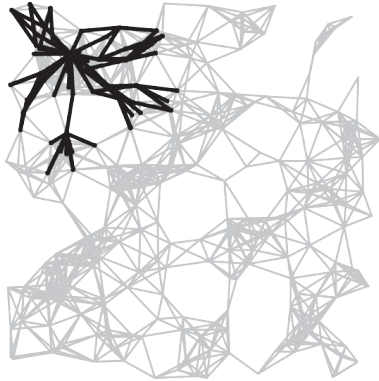
| v | edgeTo[]     |              |                |
|---|--------------|--------------|----------------|
| 0 | -            |              |                |
| 1 | <del>-</del> | 0            |                |
| 2 | <del>-</del> | <del>1</del> | 5              |
| 3 | <del>-</del> | <del>1</del> | 2              |
| 4 | <del>-</del> | 0            |                |
| 5 | <del>-</del> | 4            |                |
| 6 | <del>-</del> | <del>2</del> | <del>5</del> 2 |
| 7 | <del>-</del> | 0            |                |

pass 1 pass 2 pass 3 (no further changes) pass 4-7 (no further changes)

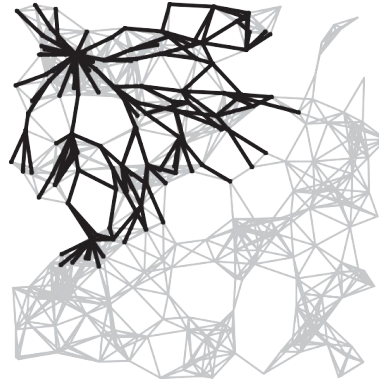
0→1 0→4 0→7 1→2 1→3 1→7 2→3 2→6 3→6 4→5 4→6 4→7 5→2 5→6 7→2 7→5

# Bellman-Ford Algorithm Visualization

passes 4



7



10



13



SPT



# Quiz

- Given a graph where all edges have positive weights, the shortest paths produced by Dijkstra and Bellman Ford algorithm may be different but path weight would always be same.
- ANS: True
- Dijkstra and Bellman-Ford both work fine for a graph with all positive weights, but they are different algorithms and may pick different edges for shortest paths.

# Quiz

- Let  $G$  be a directed graph whose vertex set is the set of numbers from 1 to 100. There is an edge from a vertex  $i$  to a vertex  $j$  if either  $j = i + 1$  or  $j = 3i$ . The minimum number of edges in a path in  $G$  from vertex 1 to vertex 100 is
- A. 4 B. 7 C. 23 D. 99
- ANS: 7
- The task is to find minimum number of edges in a path in  $G$  from vertex 1 to vertex 100 such that we can move to either  $i+1$  or  $3i$  from a vertex  $i$ .
- Since the task is to minimize number of edges, we would prefer to follow  $3*i$ . Let us follow multiple of 3.  $1 \Rightarrow 3 \Rightarrow 9 \Rightarrow 27 \Rightarrow 81$ , now we can't follow multiple of 3 anymore. So we will have to follow  $i+1$ . This solution gives a long path.
- What if we begin from end, and we reduce by 1 if the value is not multiple of 3, else we divide by 3.  $100 \Rightarrow 99 \Rightarrow 33 \Rightarrow 11 \Rightarrow 10 \Rightarrow 9 \Rightarrow 3 \Rightarrow 1$
- So we need total 7 edges.