



CS202 – Data Structures

LECTURE-20

Graphs – III

Graph Problems

Dr. Maryam Abdul Ghafoor

Assistant Professor

Department of Computer Science, SBASSE

Agenda

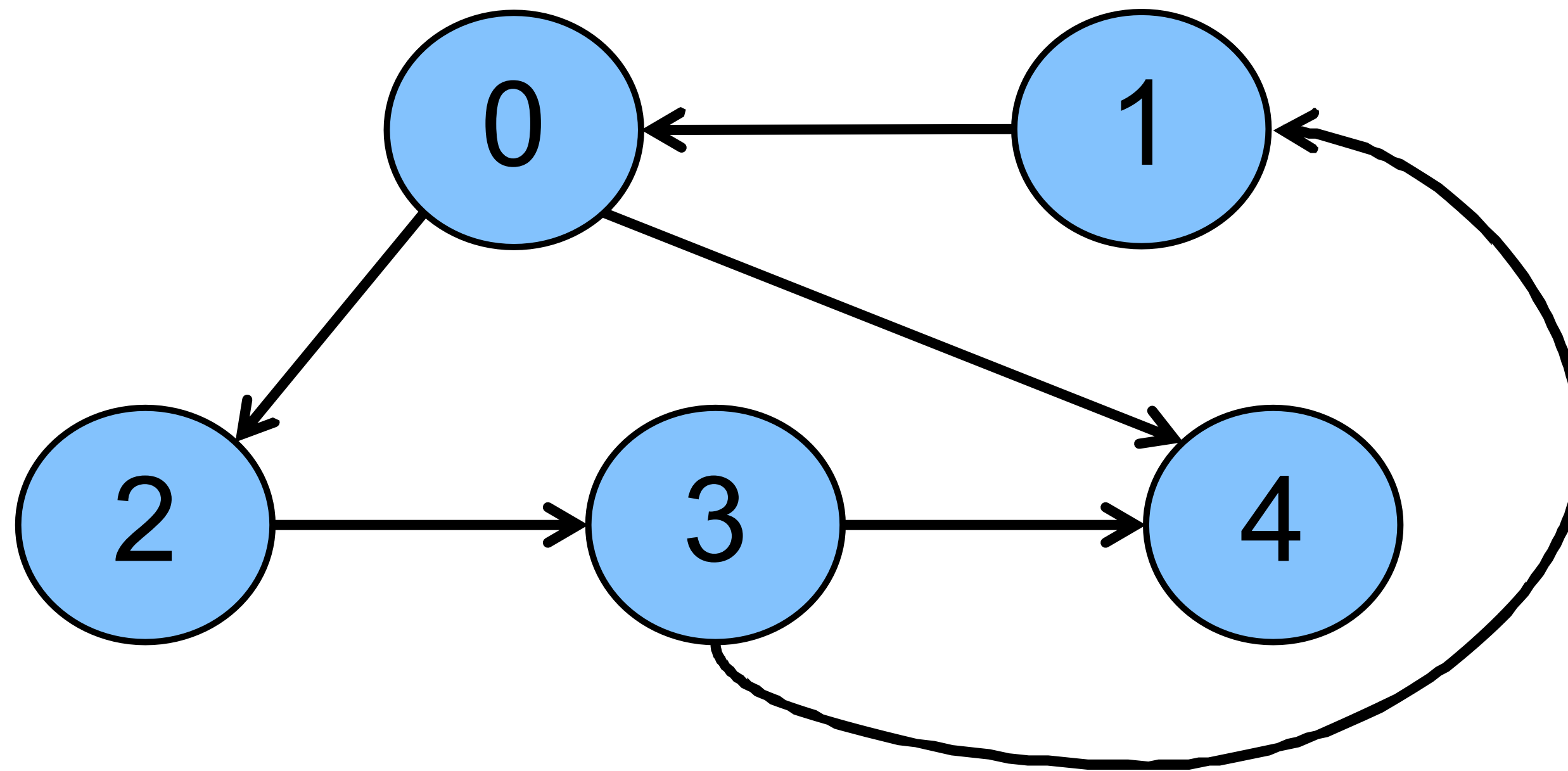
- Problem Solving using Graphs
 - Cycle detection
 - Path between two vertices
 - Shortest path algorithms

Problem: Cycle detection

“Does the graph G contain a cycle?”

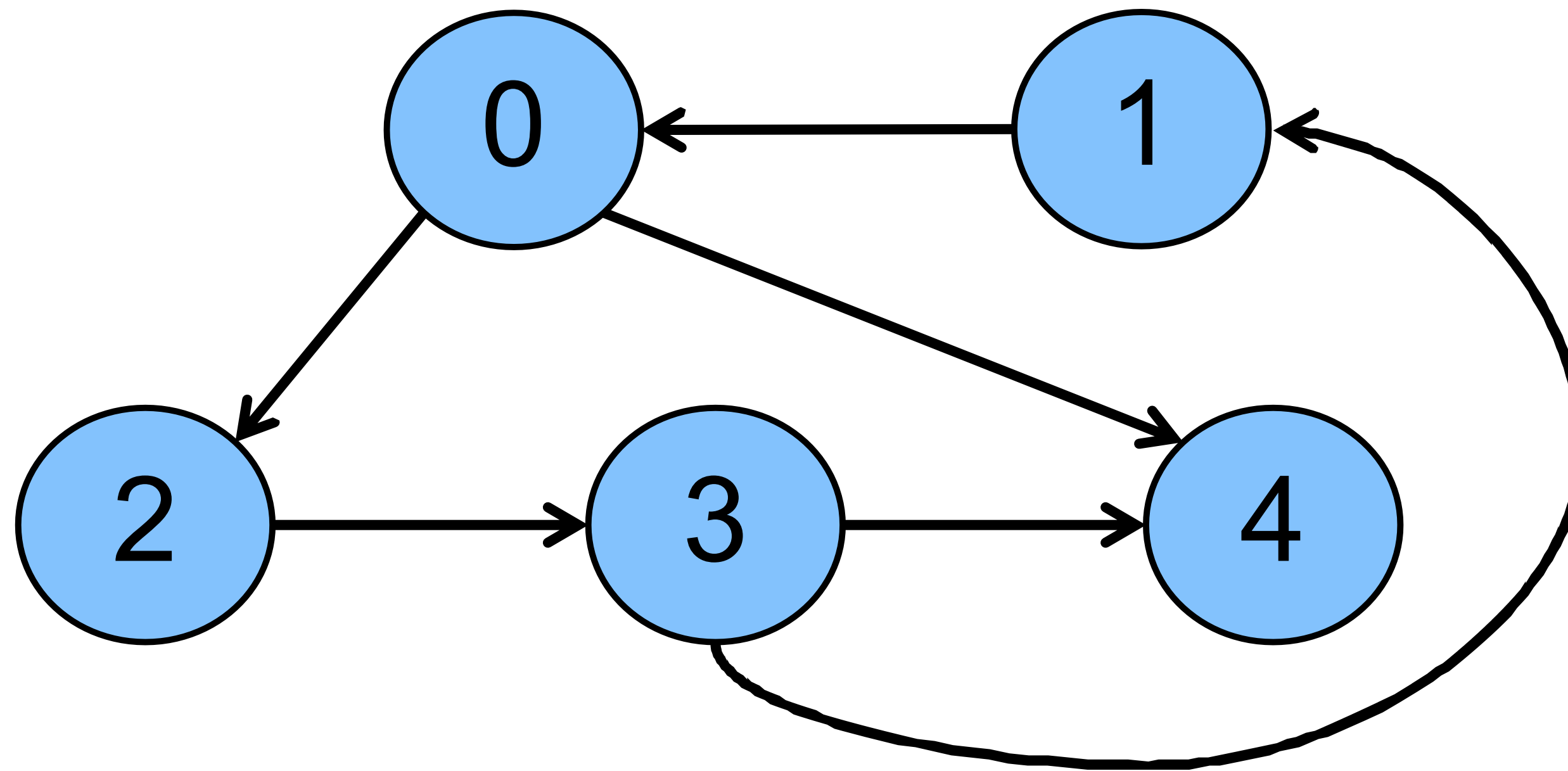
Cycle Detection

Cycle:
 $\{0, 2, 3, 1, 0\}$



Cycle Detection

Cycle:
 $\{0, 2, 3, 1, 0\}$

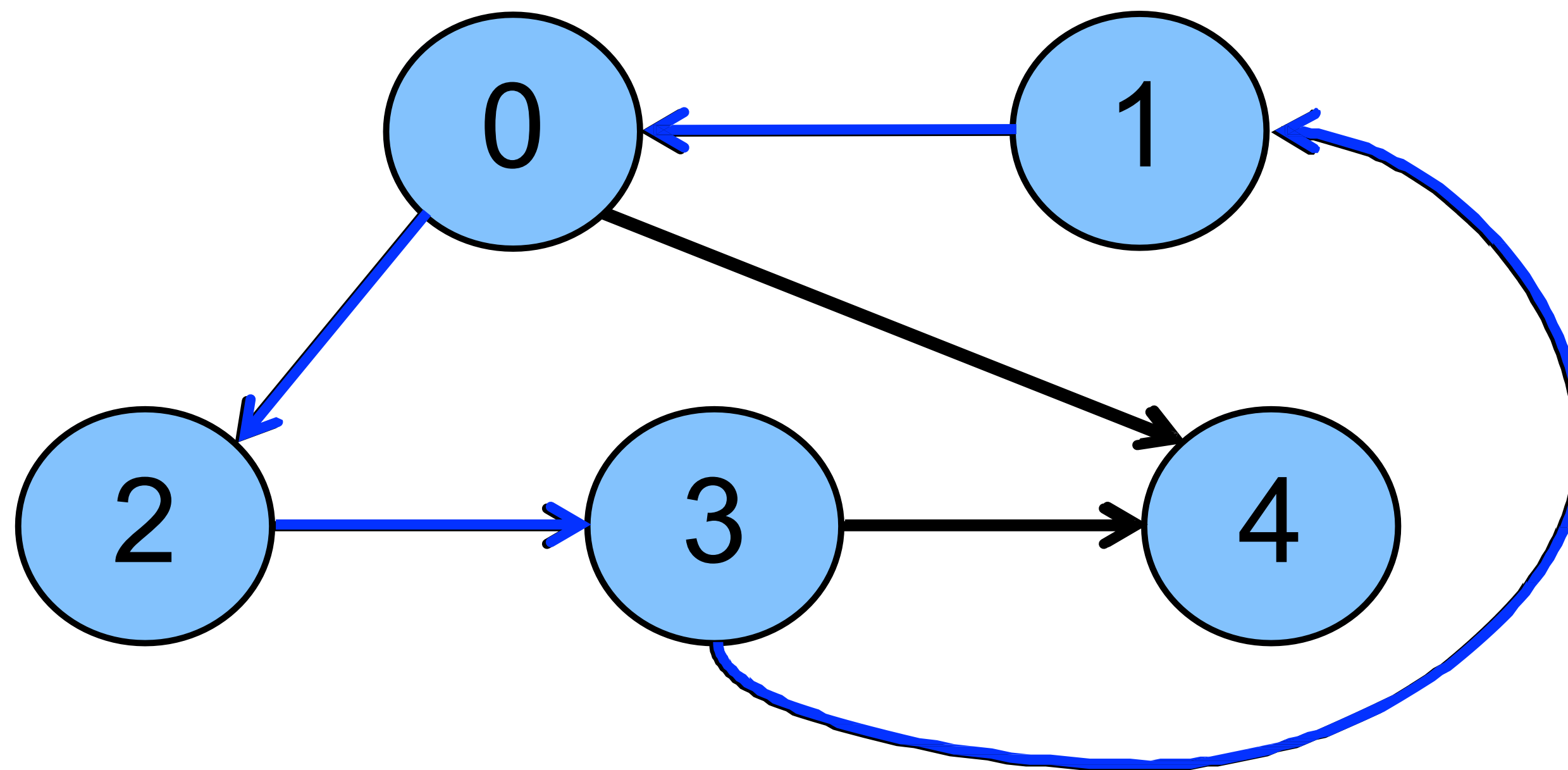


Approach 1: Run DFS, if you encounter a vertex that is already visited then return “there is a cycle”

Cycle Detection

- $\text{dfs}(0) = \{0, 2, 3, 1, 0\}$

Cycle:
 $\{0, 2, 3, 1, 0\}$

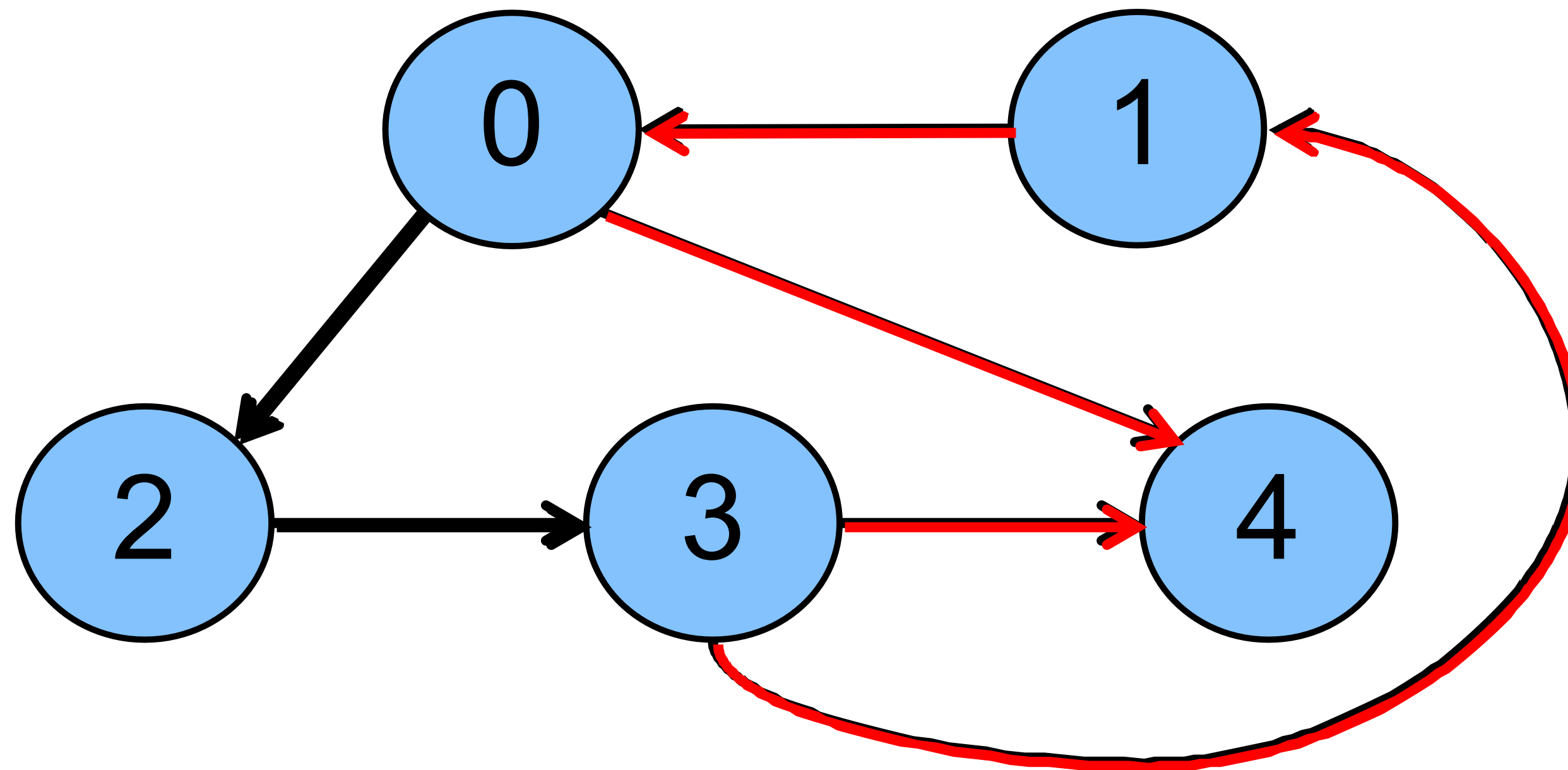


Approach 1: Run DFS, if you encounter a vertex that is already visited then return “there is a cycle”

Cycle Detection

- $\text{dfs}(0) = \{0, 2, 3, 1, 0\}$
- $\text{dfs}(3) = \{3, 4, 1, 0, 4^*\}$ NOT a cycle!

Cycle:
 $\{0, 2, 3, 1, 0\}$

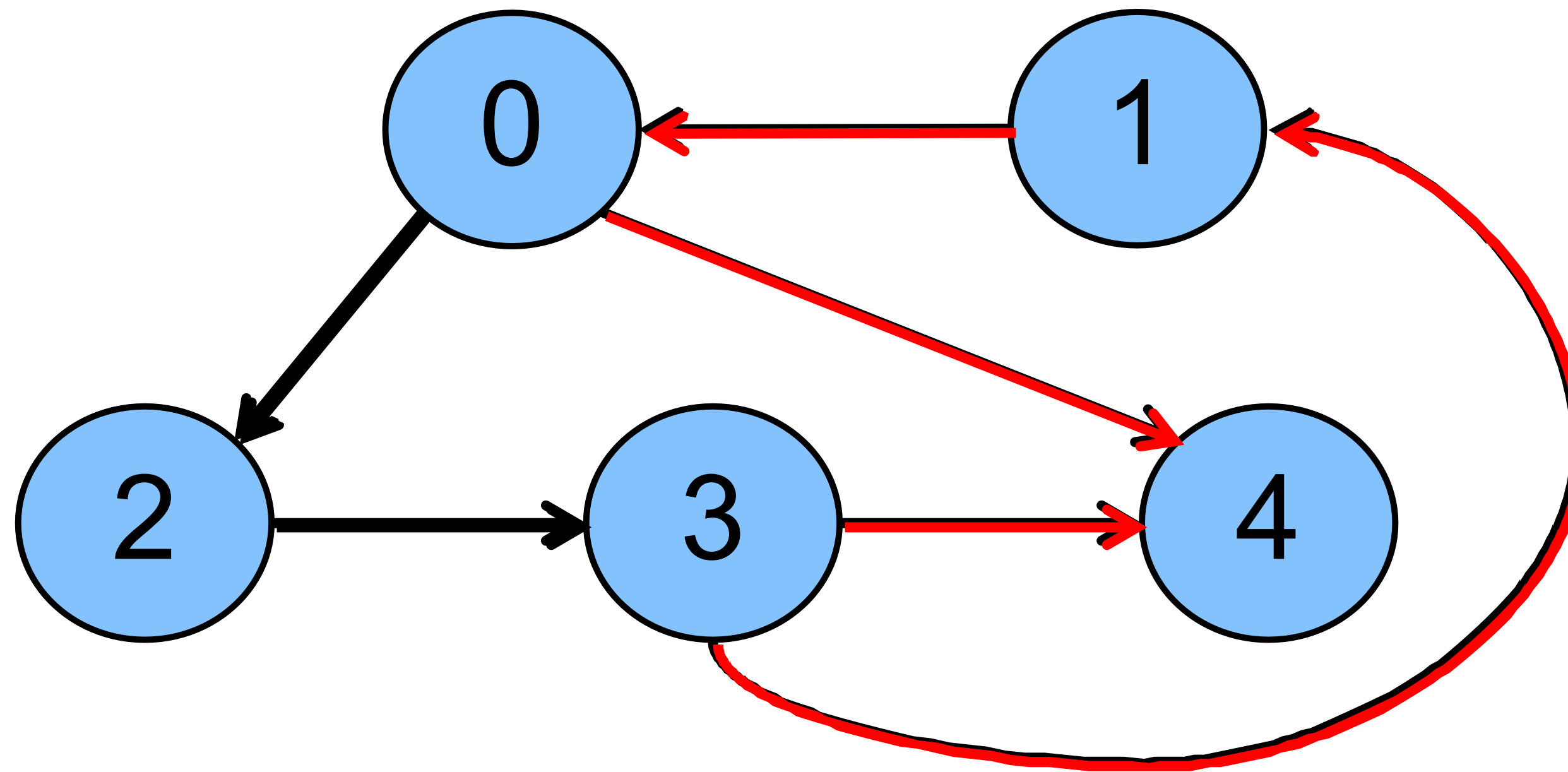


Approach 1: Run DFS, if you encounter a vertex that is already visited then return “there is a cycle”

Cycle Detection

- $\text{dfs}(0) = \{0, 2, 3, 1, 0\}$
- $\text{dfs}(3) = \{3, 4, 1, 0, 4^*\}$ NOT a cycle!

Cycle:
 $\{0, 2, 3, 1, 0\}$

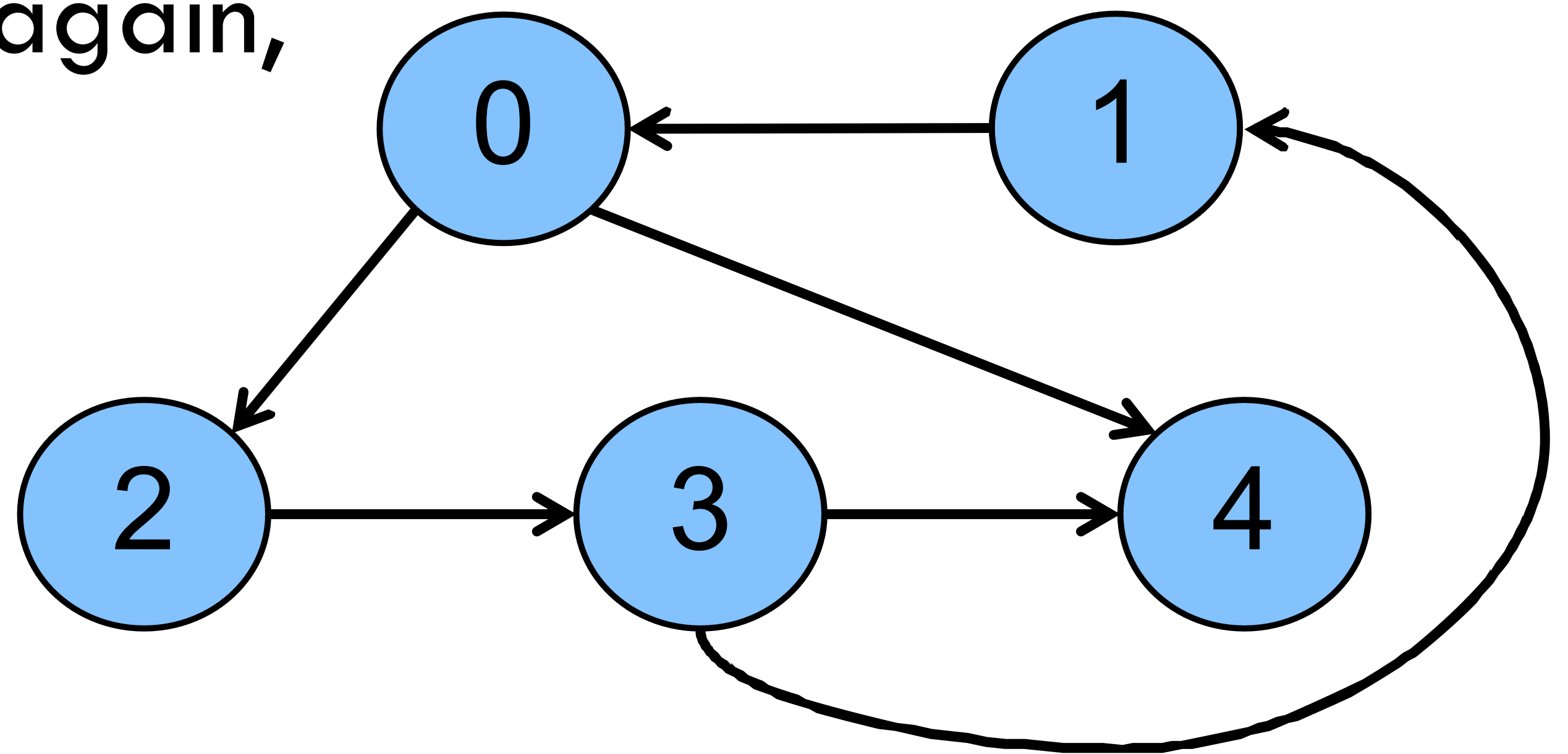


Approach 1: Run DFS, if you encounter a vertex that is already visited then return “there is a cycle”

Cycle Detection – Observations

- **Case-1:** when we visited vertex 0 again, dfs(0) was still active!
- **Case-2:** when we visited 4 again, dfs(4) was already done

Cycle:
 $\{0, 2, 3, 1, 0\}$



Approach 2: Keep track of when a vertex is “inprogress”
Use a 3-state field to mark progress: (**unvisited**, **inprogress**, **done**)

Cycle Detection – Observations

- **Approach 2:** Keep track of when a vertex is “inprogress”
- Use a 3-state field to mark progress: (unvisited, inprogress, done)
 1. Initially, all nodes are unvisited
 2. When a node is first visited, we mark it as “inprogress”
 3. Once all successor nodes are visited, we mark it as done
 4. There is a cyclic path reachable from vertex i iff some node's successor is found to be marked “inProgress” during $\text{dfs}(i)$

Time Complexity

AL: $O(|V| + |E|)$

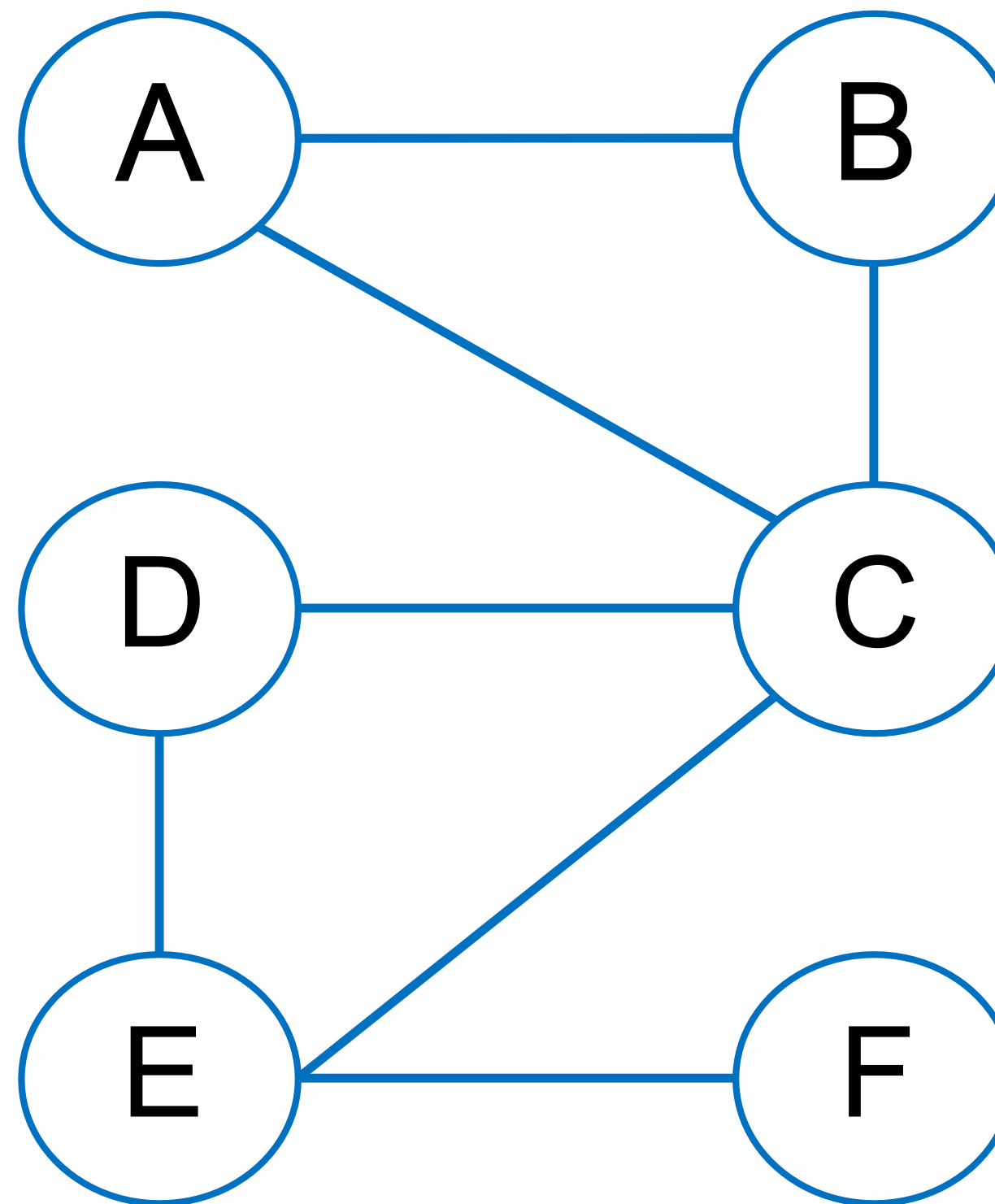
AM: $O(|V|^2)$

Let's try to solve some problems

- **Problem (Path detection):** “Is there a path from vertex s to vertex t ?”
- **Solution:**
 - Run DFS start from vertex s
 - If vertex k is visited \rightarrow there is a path from j to k
 - Time complexity?

Problem: Shortest Path

- How can we find the **shortest path** between two vertices in a graph

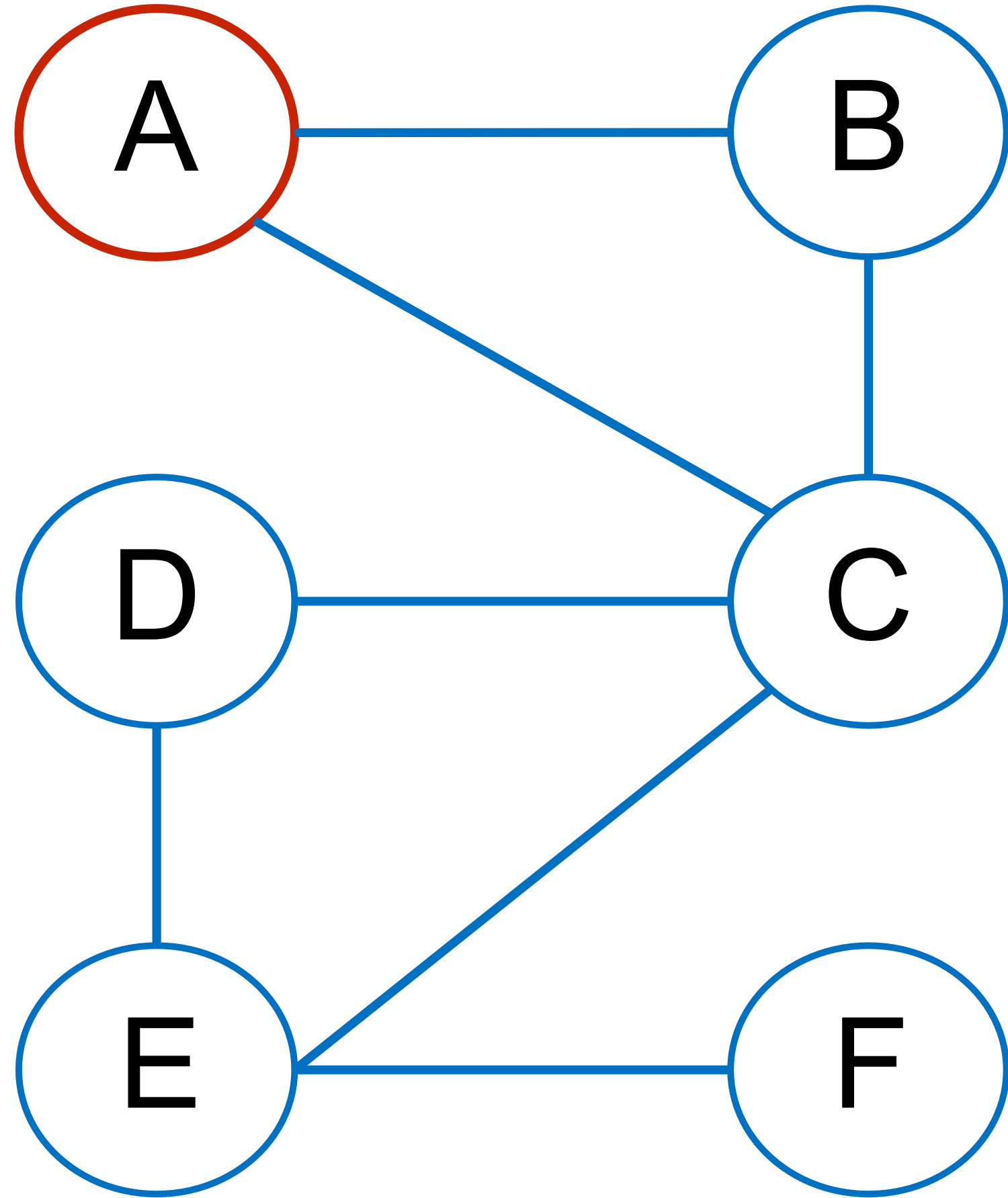


Problem: Shortest Path

- How can we find the **shortest path** between two vertices using BFS in an unweighted graph
- Solution:
 - (a) Add a **distance field** to each node
 - (b) When **bfs** is called on a vertex v , **set v 's distance to zero**
 - (c) When a **vertex w is to be enqueued**, set its distance to the distance of the **current vertex + 1**

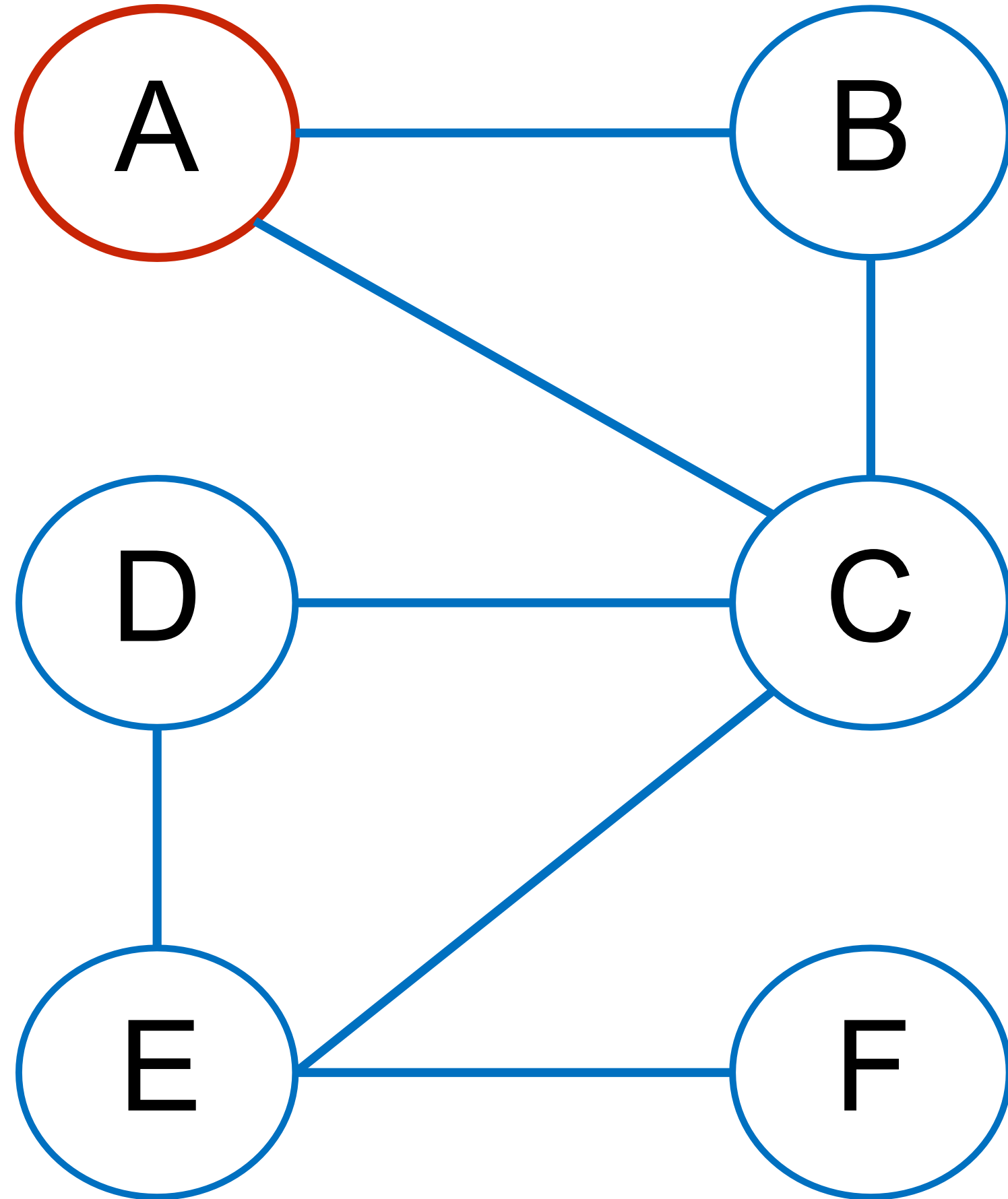
BFS

dist=0, parent=∅



BFS

dist=0, parent=∅

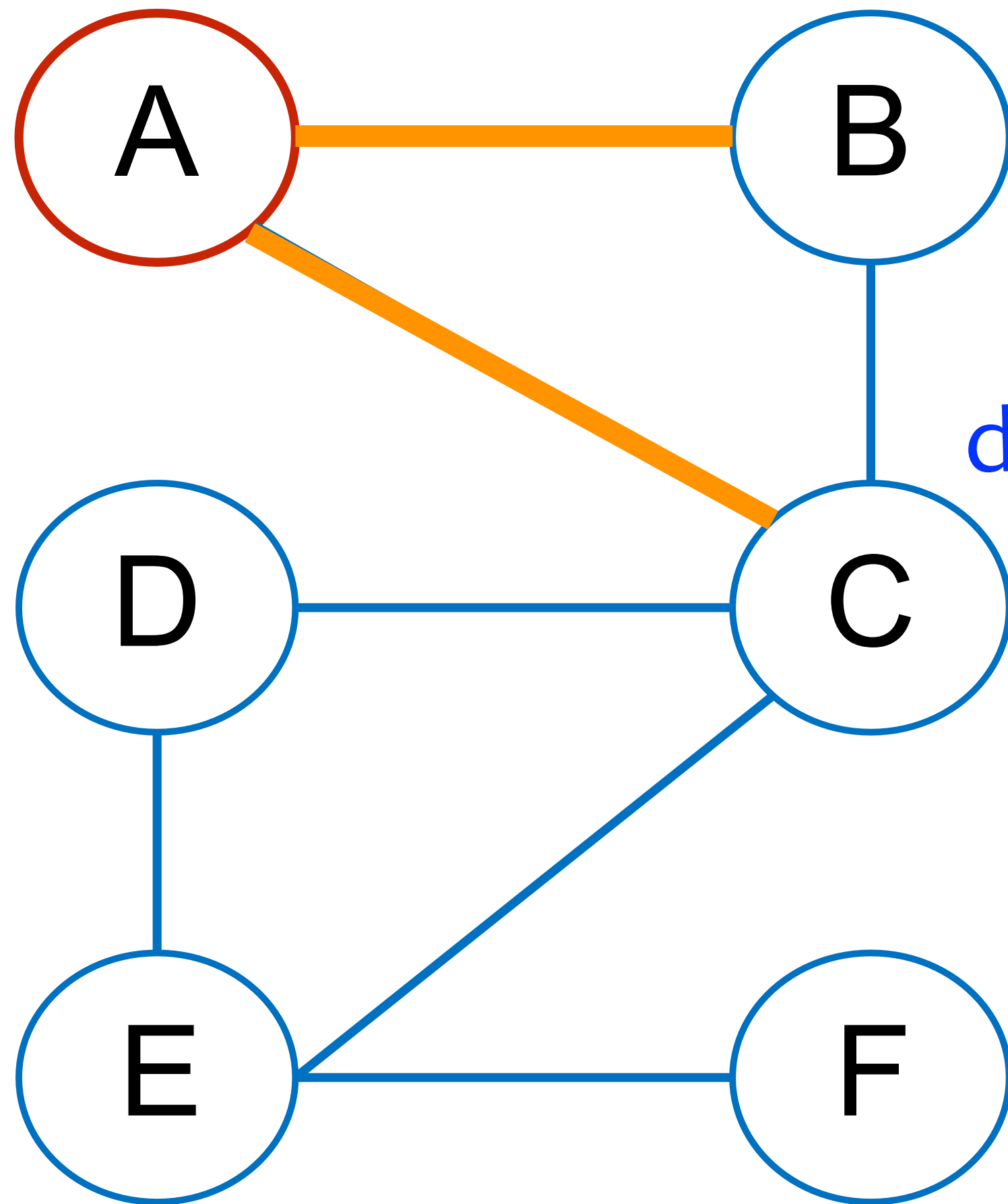


A

BFS

dist=0, parent=∅

dist=1, p=A

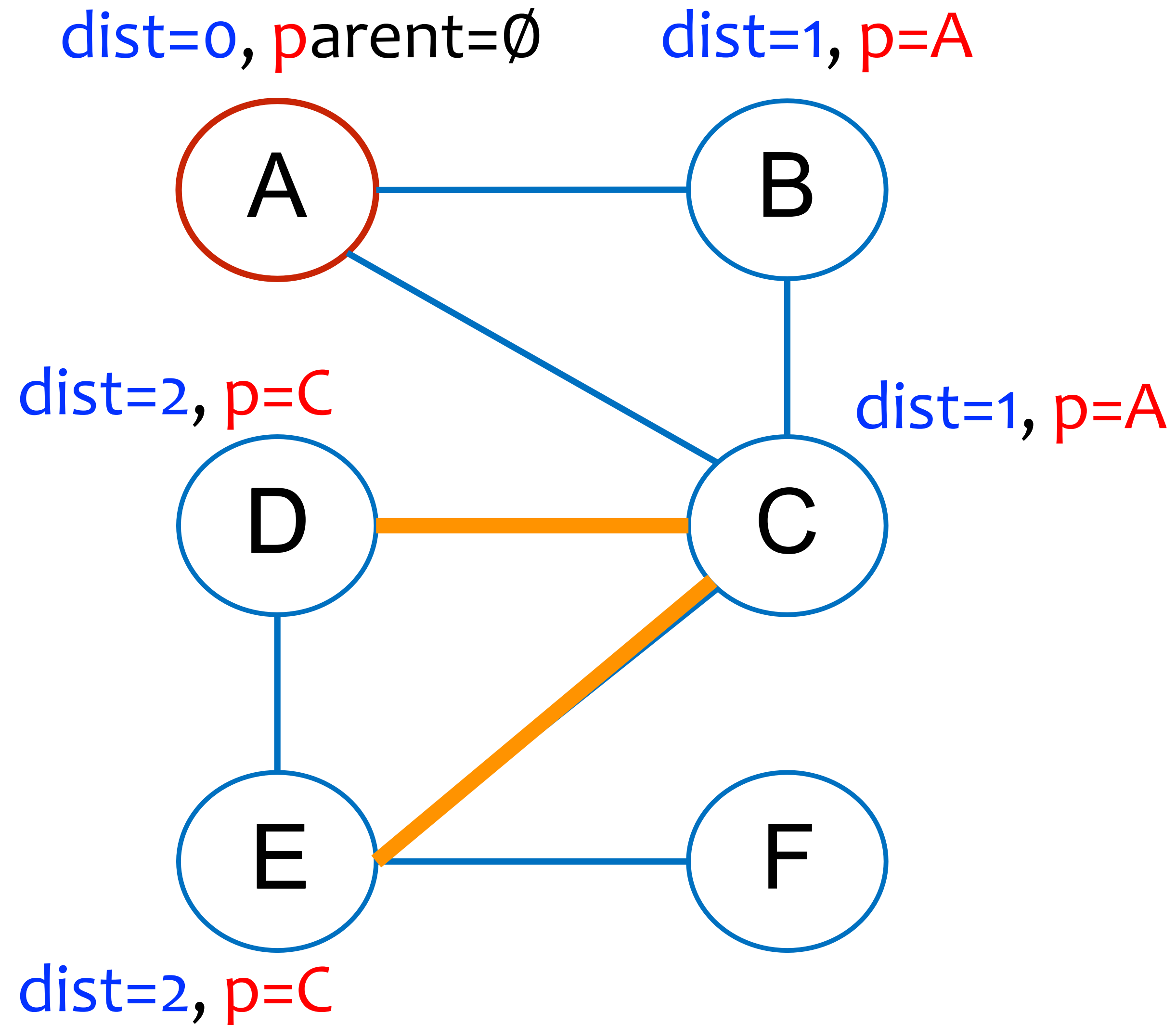


dist=1, p=A

A

B C

BFS

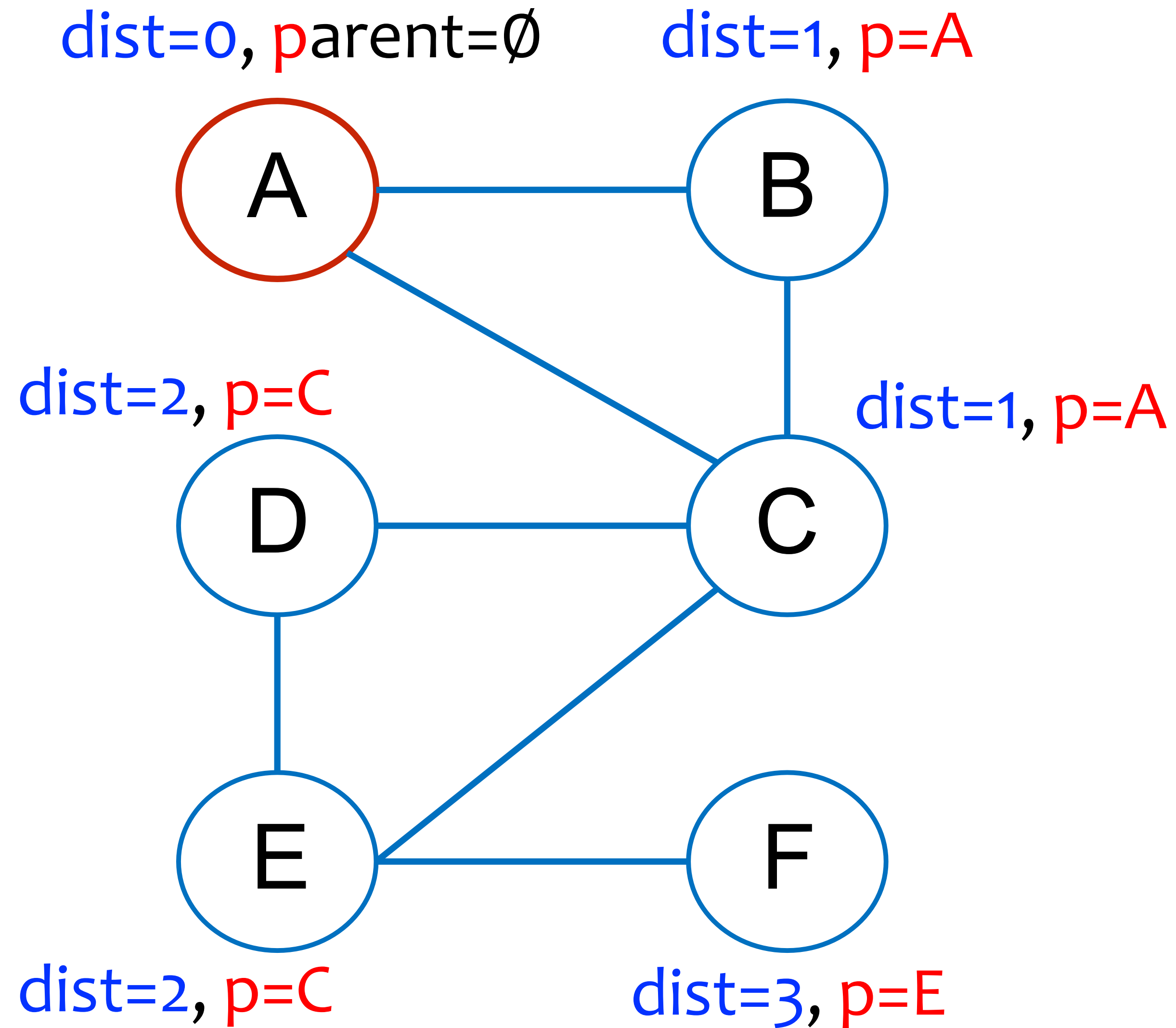


A

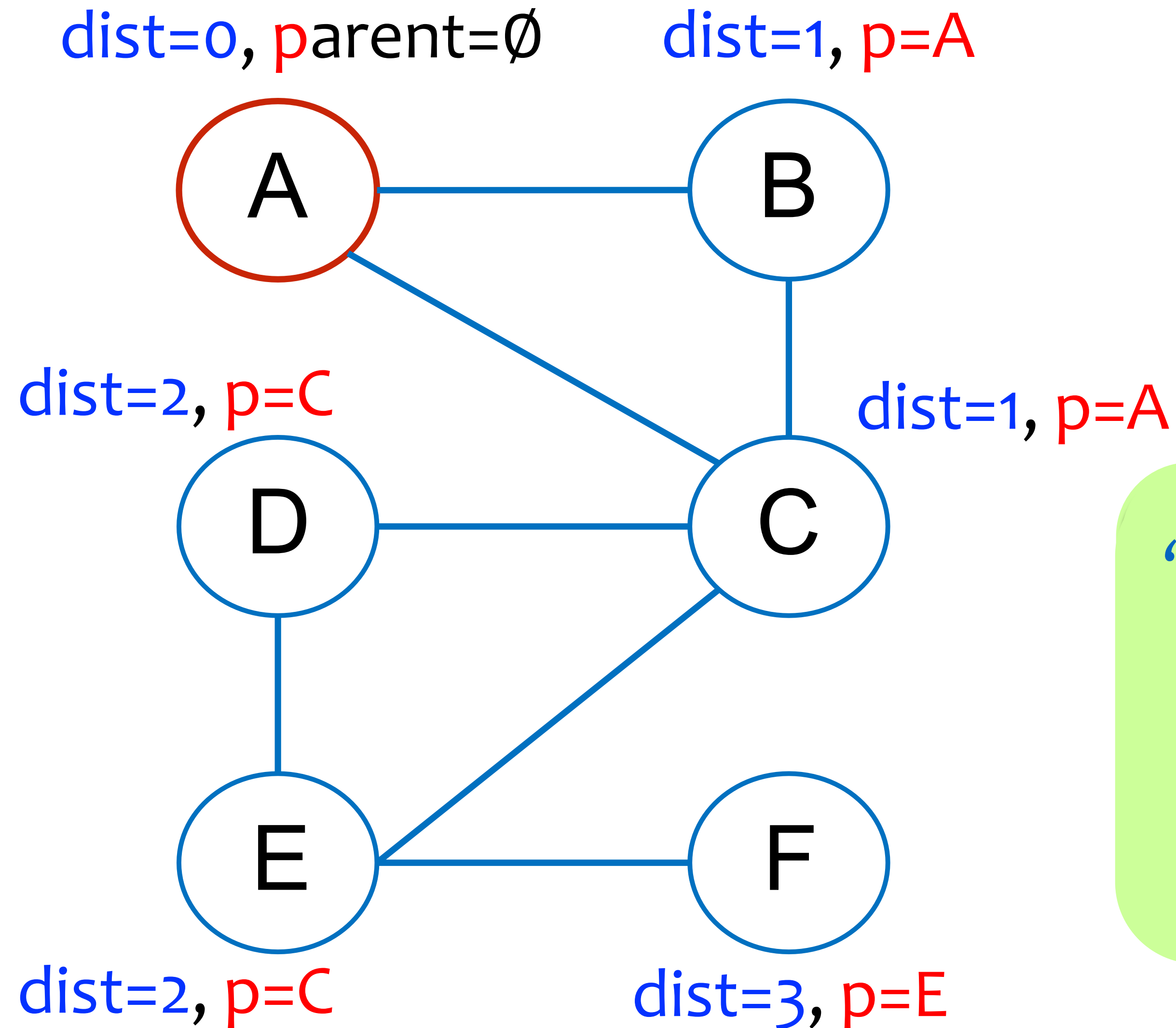
B C

E D B

BFS



BFS



A

B C

E D B

“Follow the parent pointer to find the shortest path from the destination vertex to the source vertex”

A C E F

Weighted Graphs

- Each edge is labelled with a **numerical weight**
 - **Distances** between nodes
 - **Cost** of moving from one node to another
 - **Signal strength** between two wireless nodes

How do we represent weighted graphs?

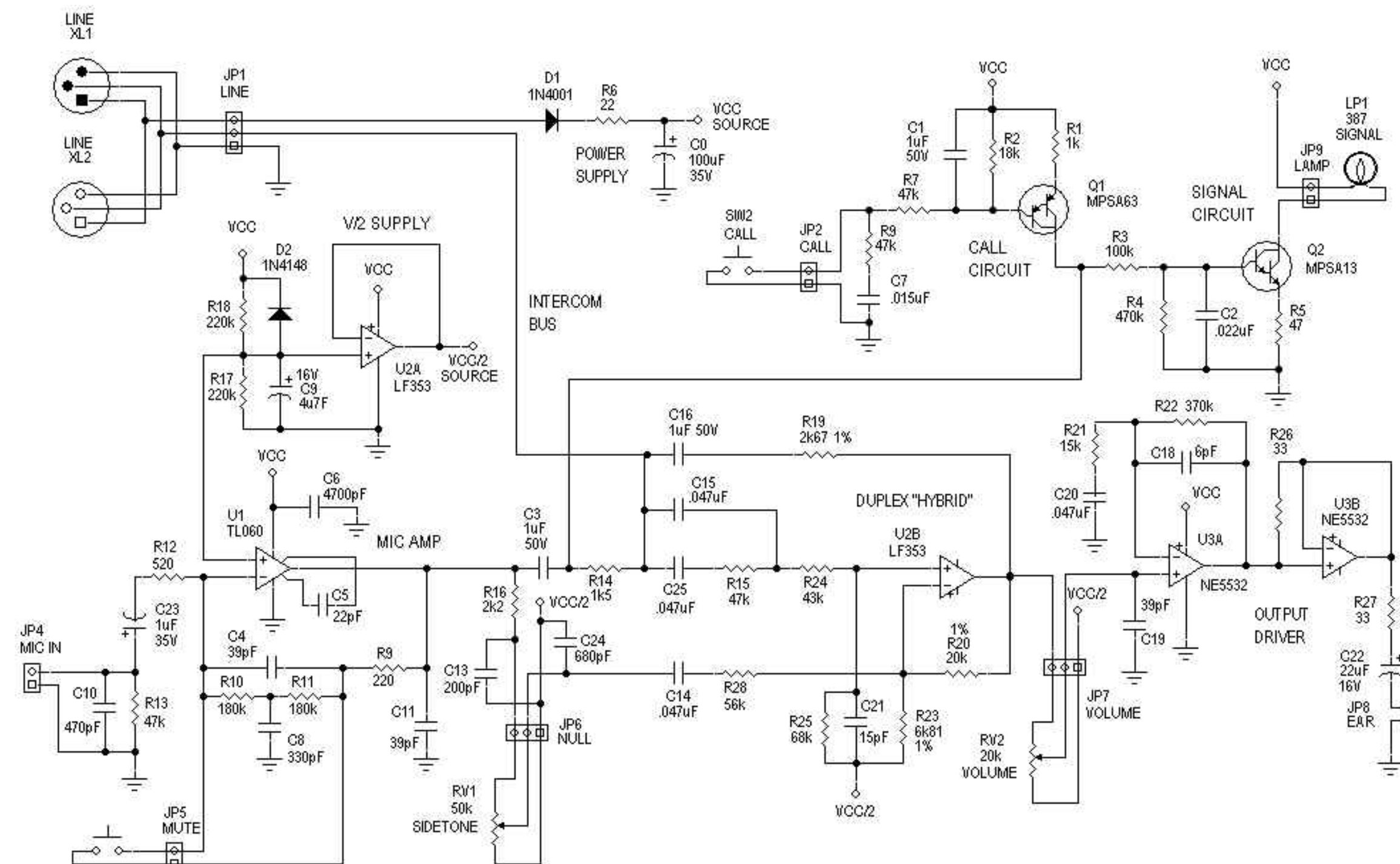
Weighted Graphs - Representation

- Adjacency Matrix
 - 2D array of ints/double/...
 - But... how do we represent the absence of an edge?
 - 2D array of structs having one field for edge connectivity and another for edge weights
- Adjacency List
 - Each list node contains a weight variable

| | 0 | 1 | 2 | 3 |
|---|---|----|---|----|
| 0 | 1 | 3 | 9 | 4 |
| 1 | 0 | 4 | 7 | 4 |
| 2 | 2 | -5 | 4 | -5 |
| 3 | 1 | 4 | 2 | -9 |

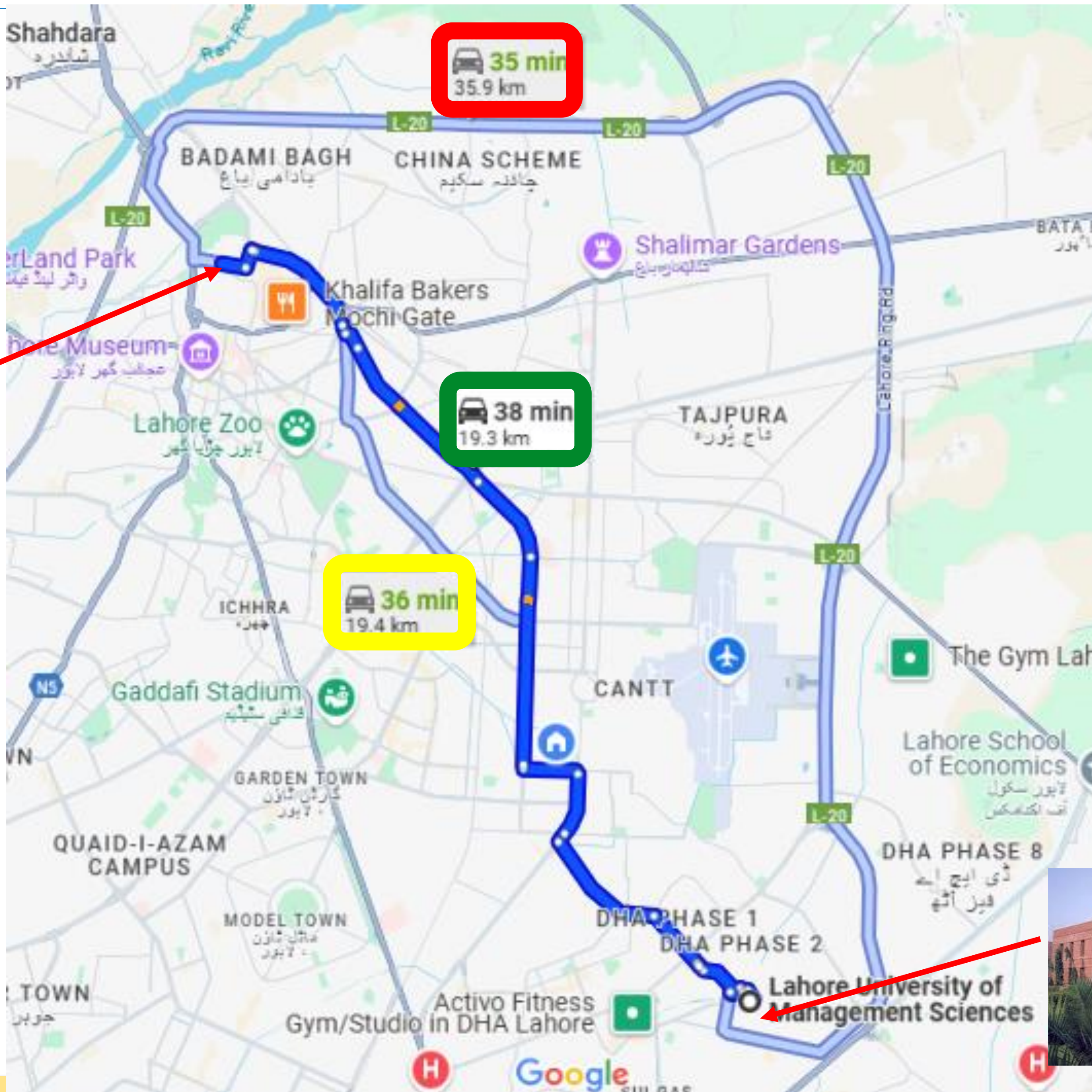
Circuit Design

- Signal **propagation delay** depends on **path length** between components
- Shortest path ensures **faster response times** in critical circuits
- Used in timing analysis to **optimize logic gate placement**



Driving Directions

- Google Maps



Routing Data

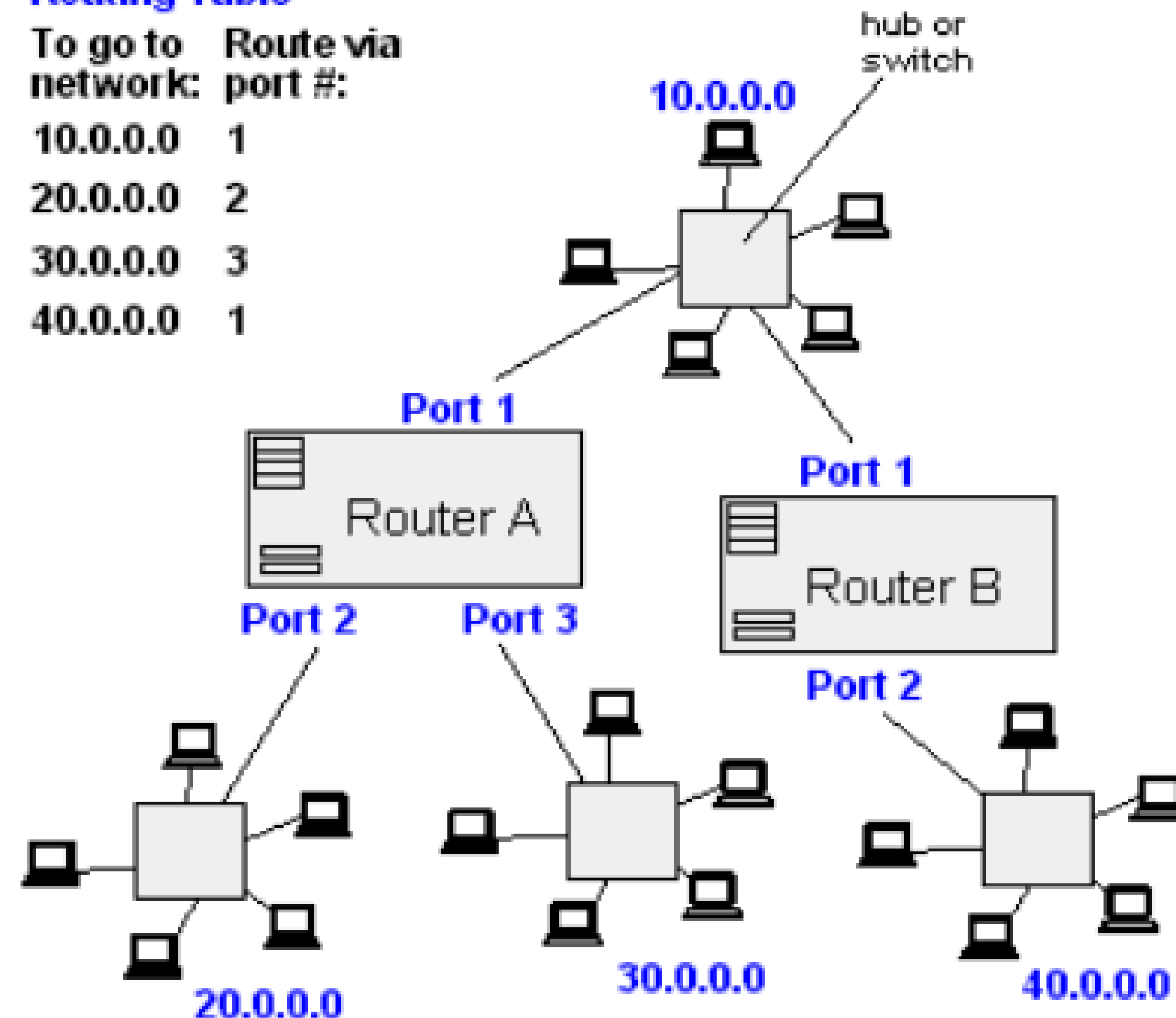
Routers use **graph algorithms** to find the **fastest path** for data packets

Edges represent links with **weights** like **latency** or **congestion**

From Computer Desktop Encyclopedia
© 1998 The Computer Language Co. Inc.

Router A
Routing Table

| To go to network: | Route via port #: |
|-------------------|-------------------|
| 10.0.0.0 | 1 |
| 20.0.0.0 | 2 |
| 30.0.0.0 | 3 |
| 40.0.0.0 | 1 |

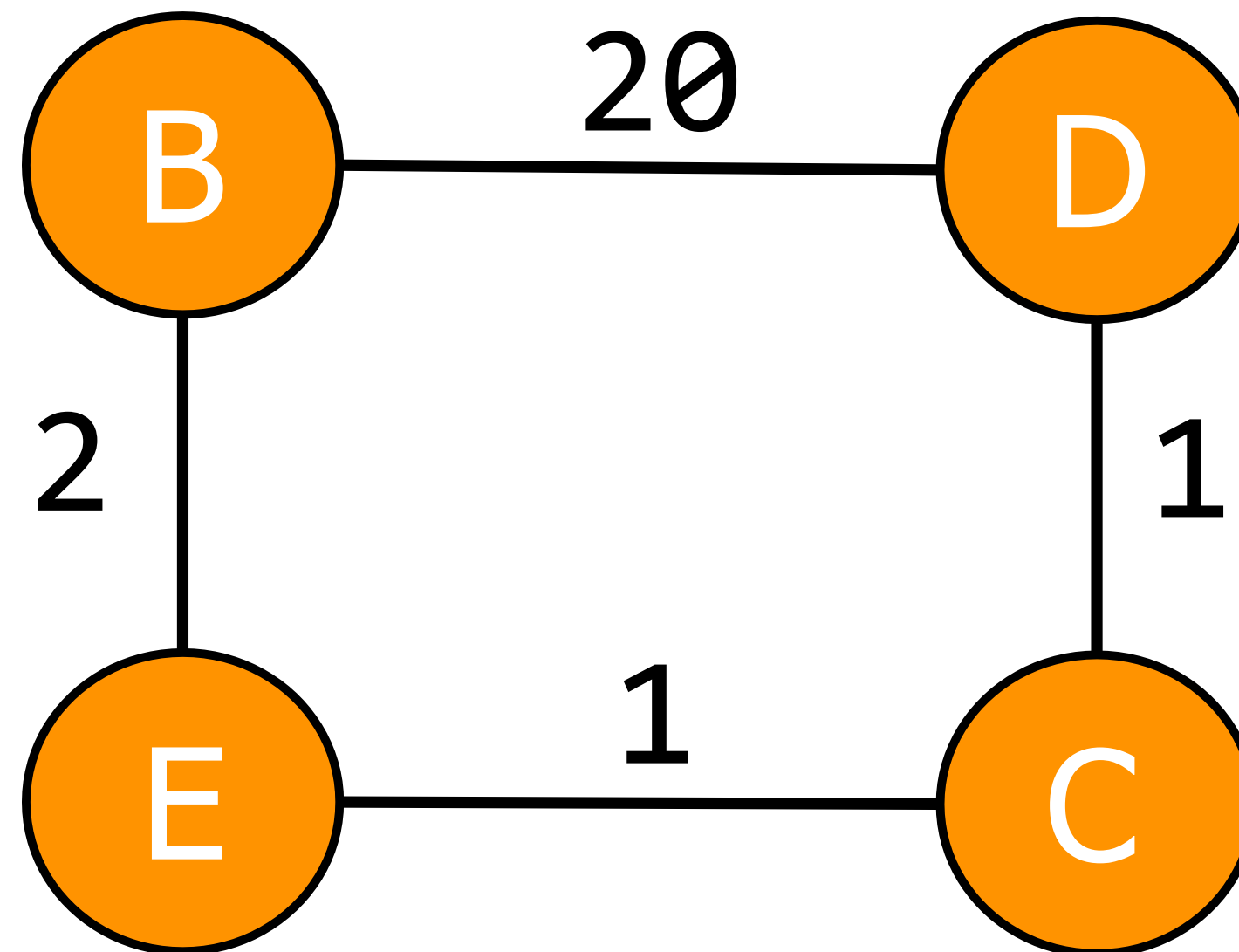


Shortest Path Problem

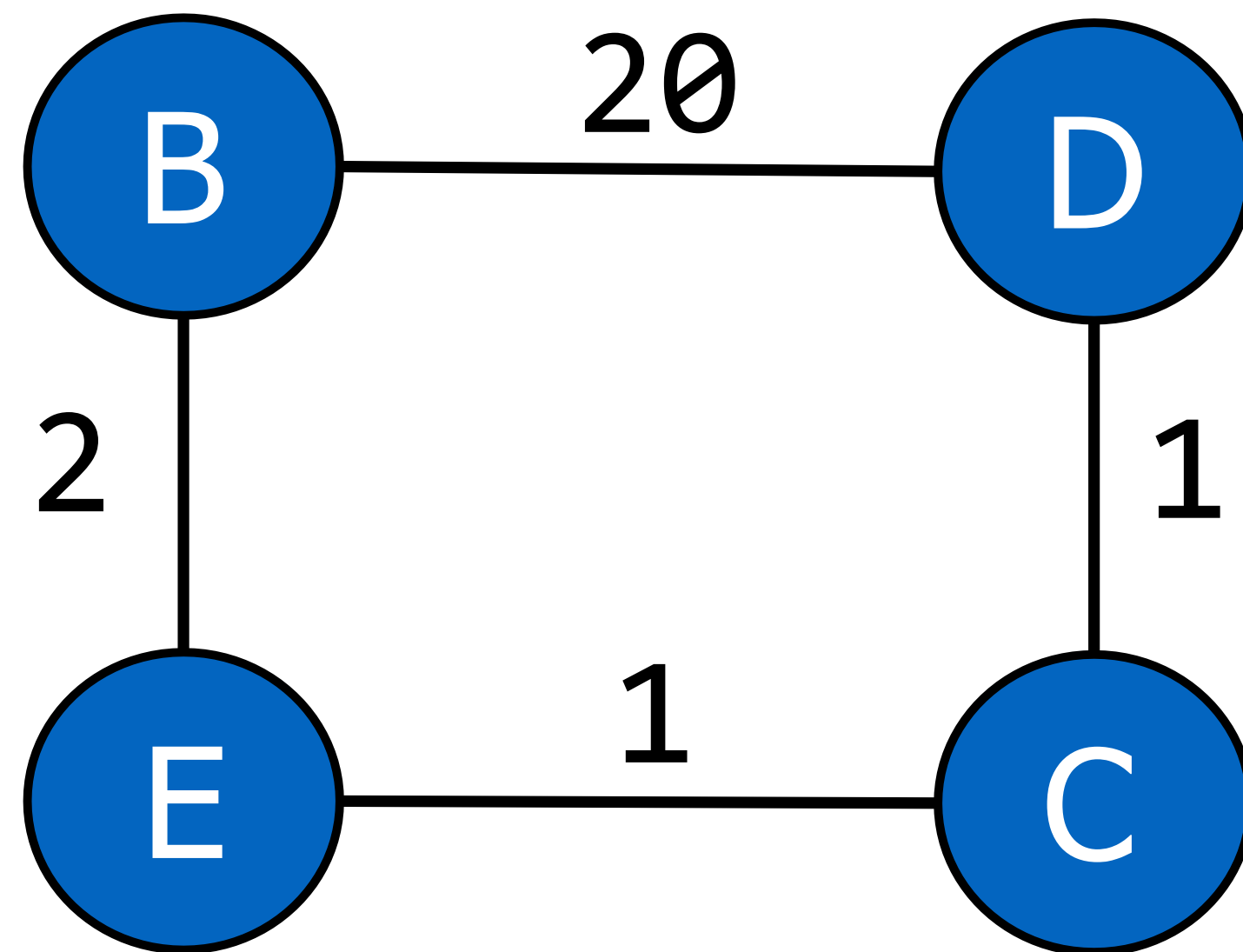
Problem: Given a graph $G = \{V, E\}$ representing different locations, where each edge has an associated cost (weight)

Assumption: The costs are non-negative

Goal: Find the least-cost path from node B to a node D

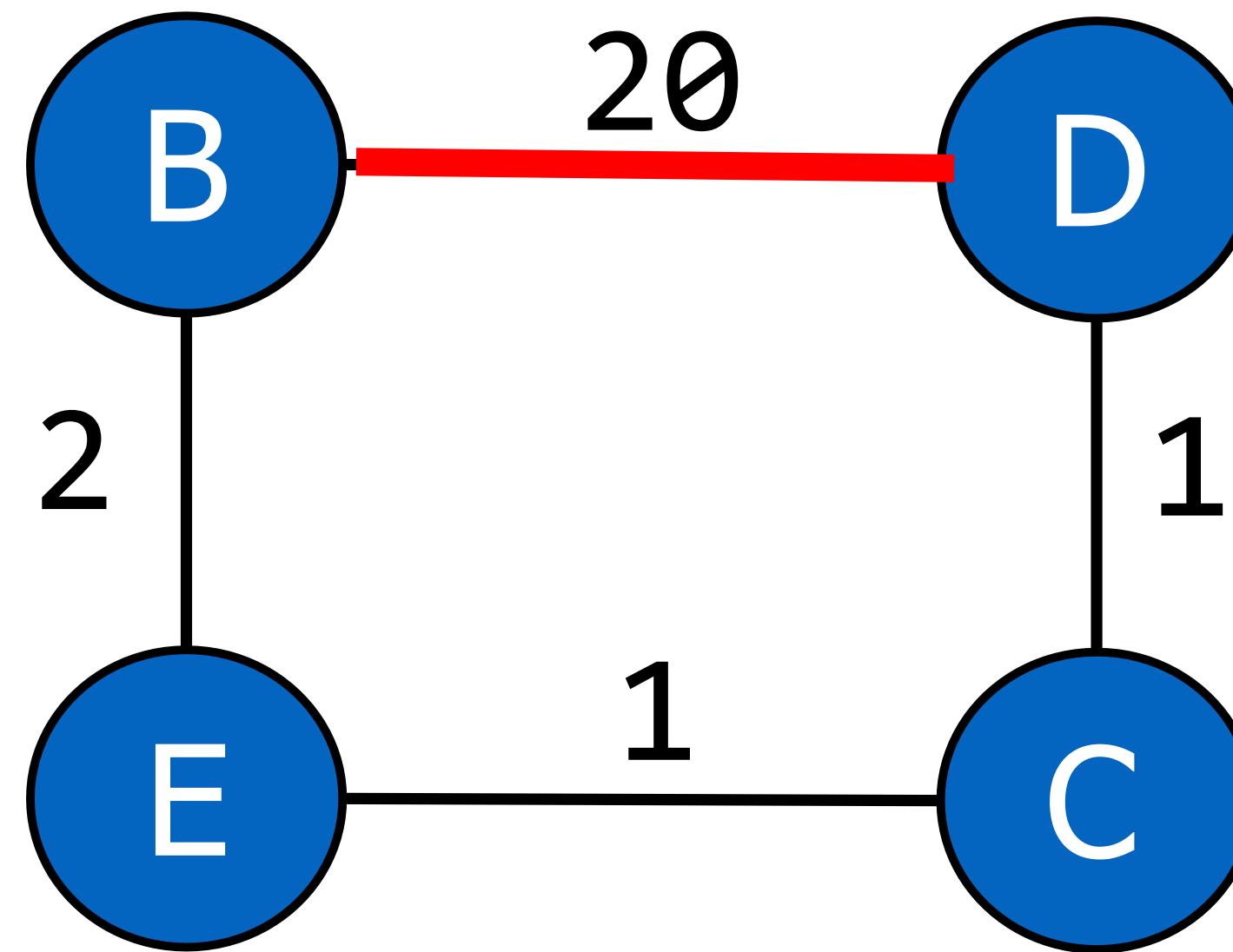


BFS for Shortest Path



Which Path BFS will find?

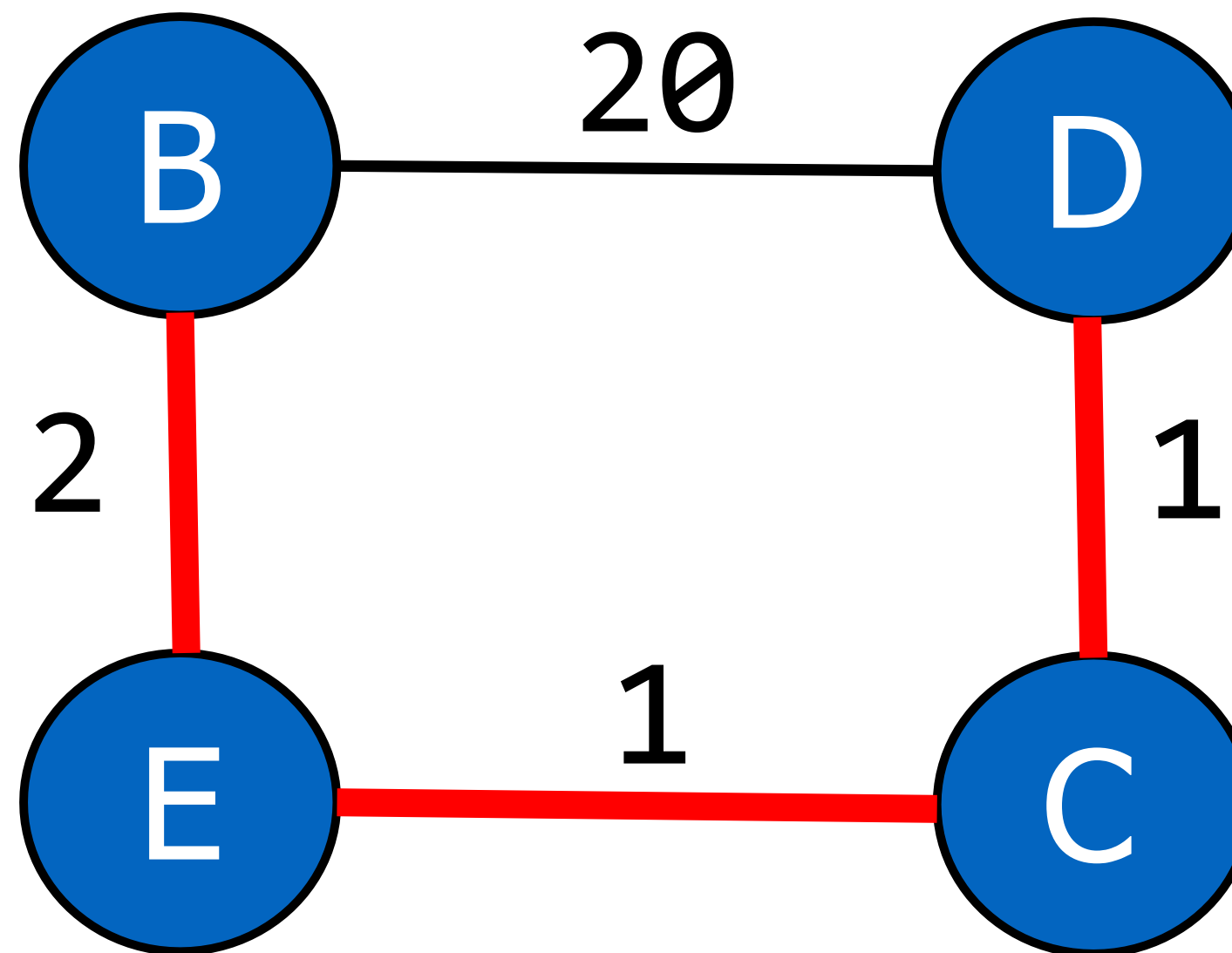
BFS for Shortest Path



BFS finds the **shortest path** based on the “**least number of edges**” but that path may have **greater cost**!

BFS for Shortest Path on Weighted Graph

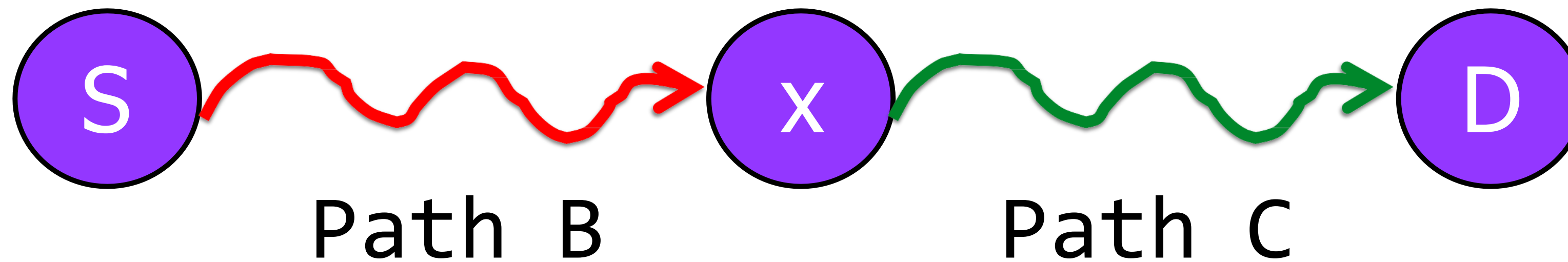
Always pick the **next vertex** with the **least cost**!



The **greedy choice** may **not be optimal**, need to keep track of other possible paths as they may turn out to be short!

Optimal Substructure Property

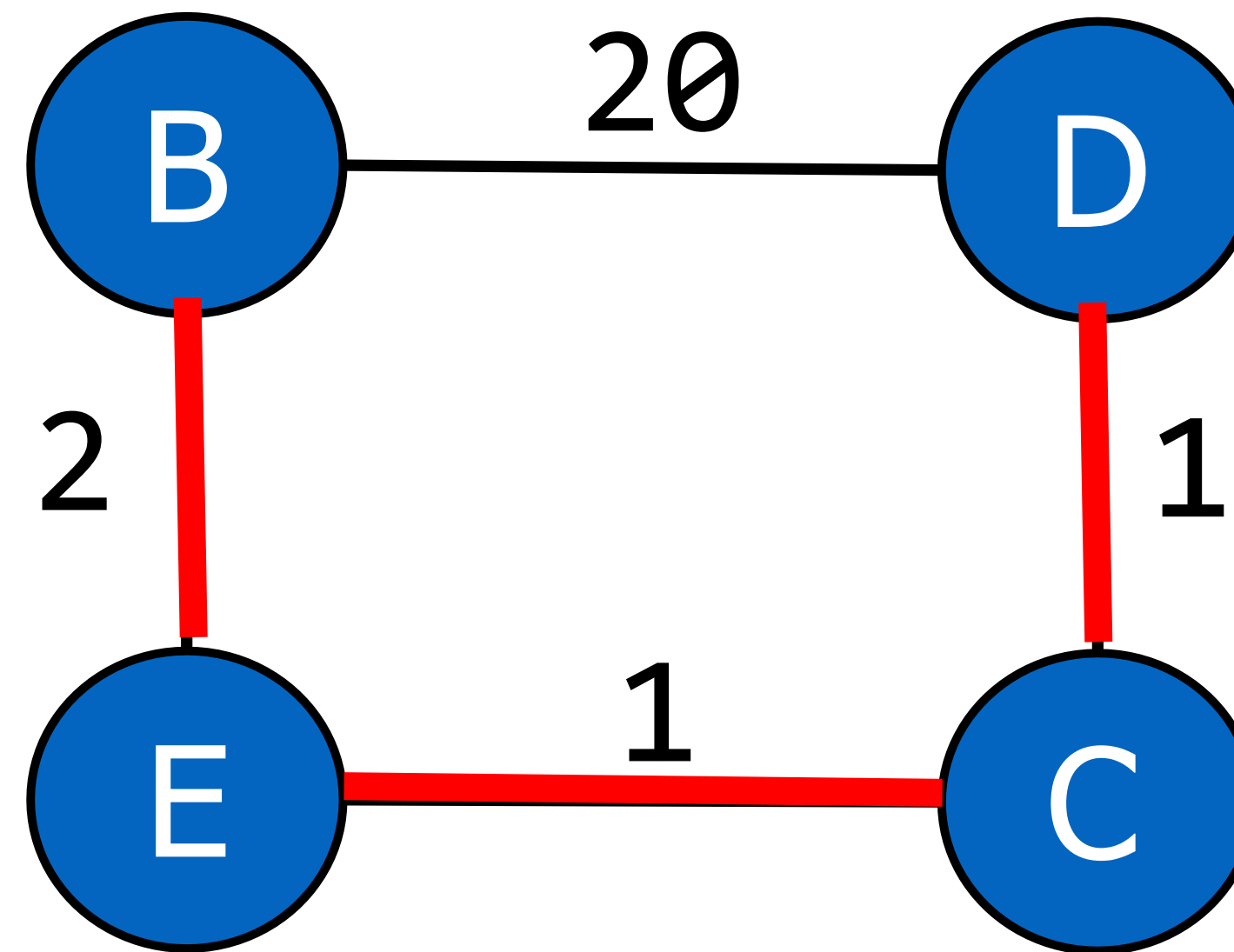
Any **sub-path** of an **optimal path** is also **optimal**



If the path (S, \dots, x, \dots, D) is the shortest path from S to D then Path B (S, \dots, x) is the shortest path from S to x

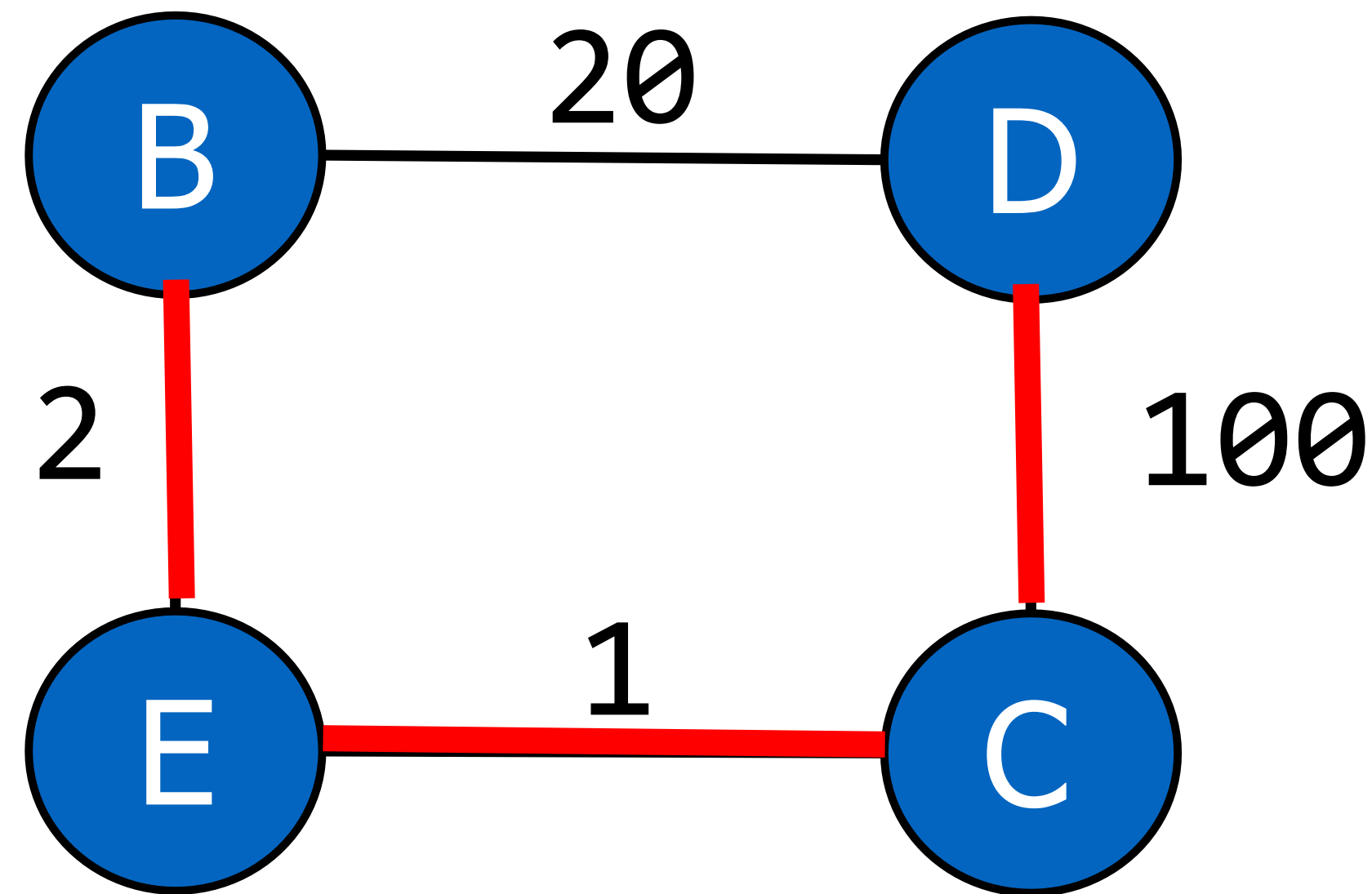
Example – 1

- Examine the shortest path from **B** to **D** and all **its sub-paths**



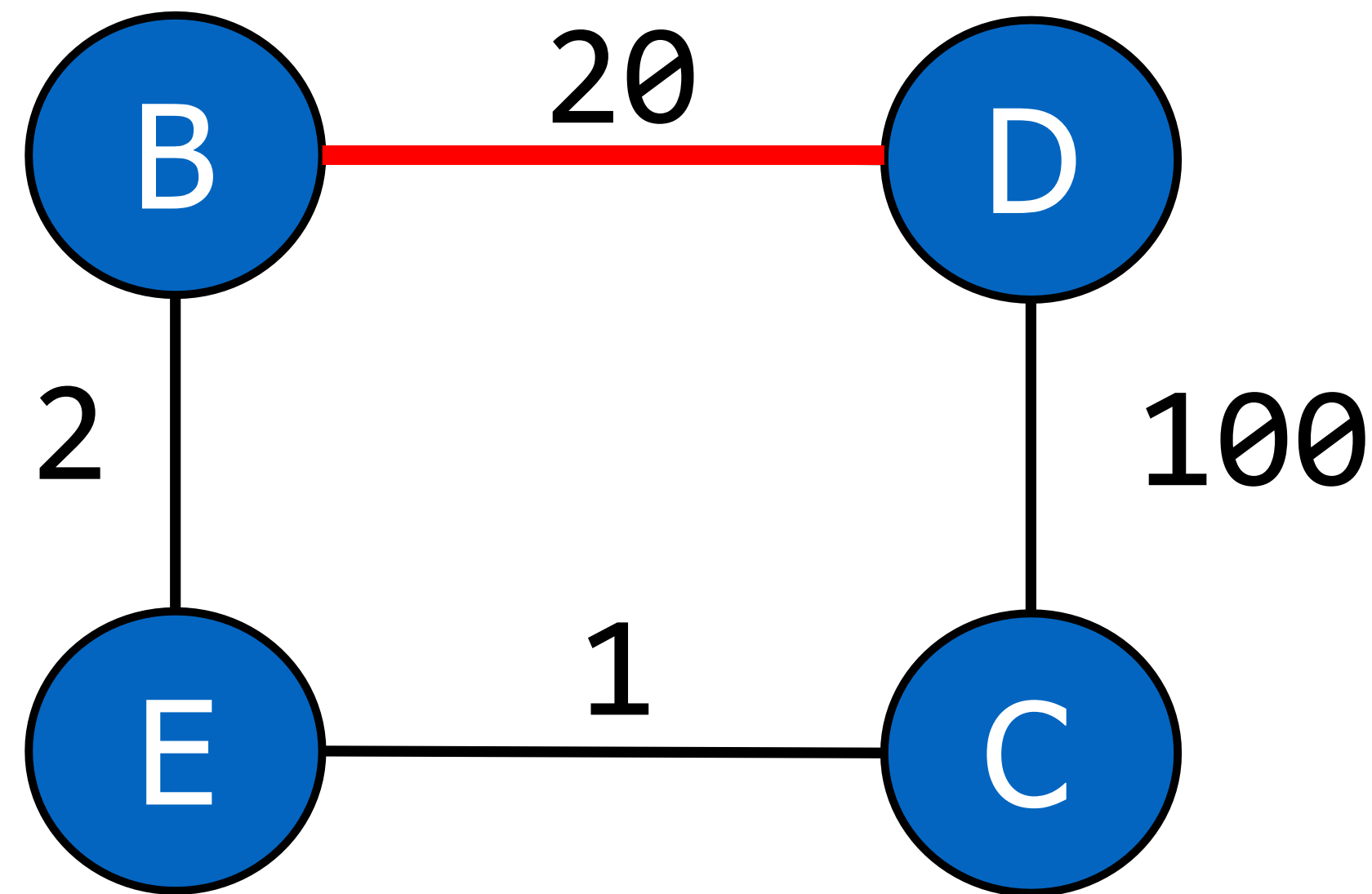
Example – 2

- Examine the shortest path from **B** to **D** and all **its sub-paths**



Example – 2

- Examine the shortest path from **B** to **D** and all **its sub-paths**

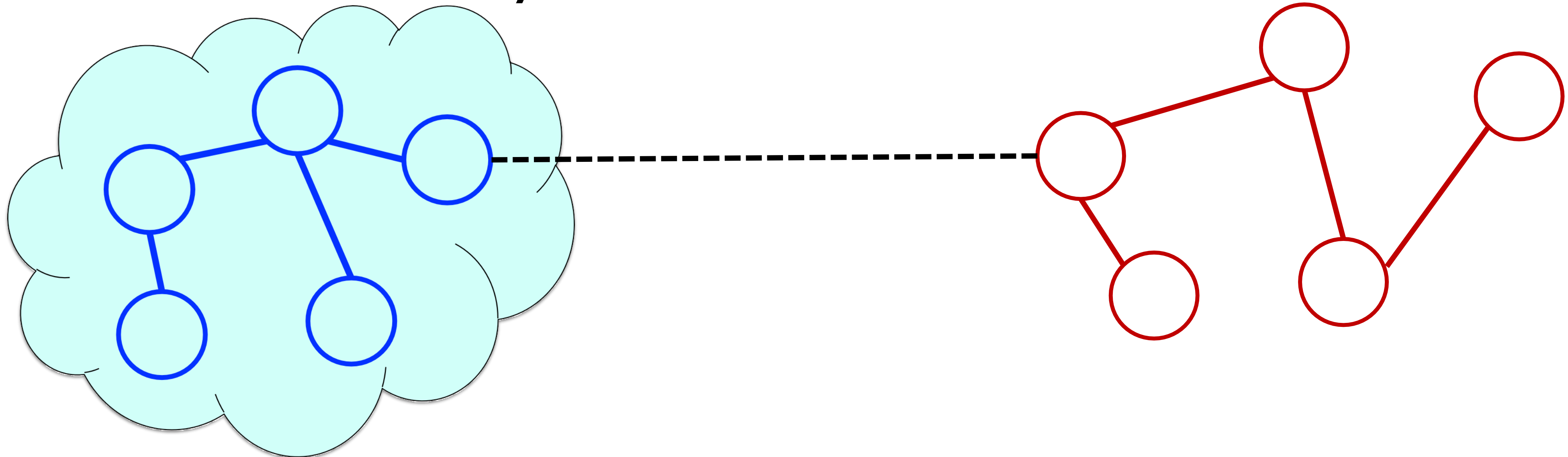


Key Insight

- Adapt BFS to handle weighted graphs (with no negative edge weights) and keep track of learnt paths

Dijkstra's Algorithm: Big Picture

- Two Kinds of Vertices
 - Vertices **inside** the cloud
 - Vertices for whom the shortest path is known
- Vertices **outside** the cloud
 - Vertices for whom only tentative distances are known

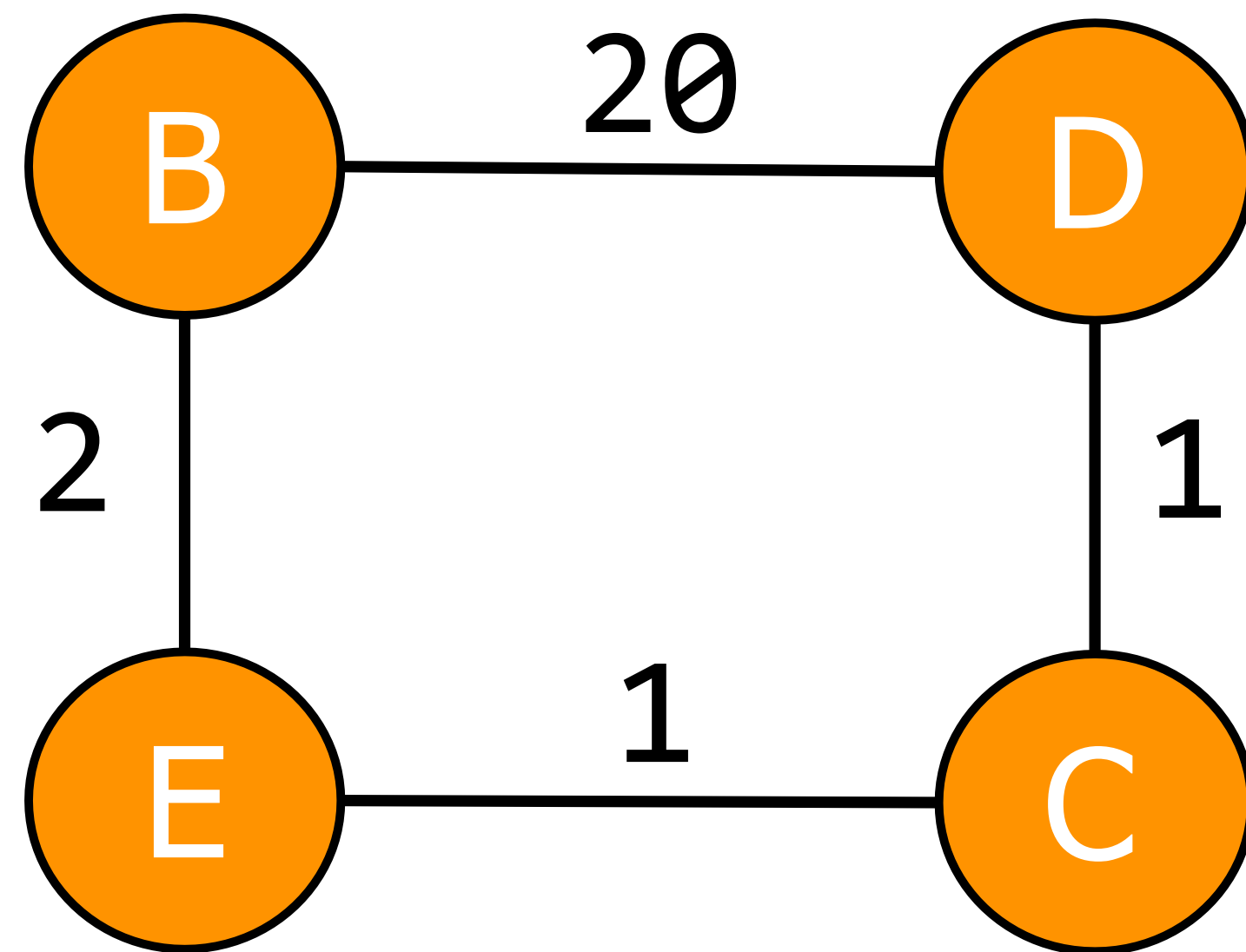


At Each Step/Iteration

- Pick the **closest vertex outside** the cloud
- Add it to the cloud
- Update distances
- **Greedy algorithm:** makes choices that currently seem **best**

Let's see how Dijkstra would work here

- Optimal substructure property + non-negative edge weights + keeping track



Goal: Find the Shortest Path (SP) from B to D

To find the SP from B to D, we need to find the SP from B to C

To find the SP from B to C, we need to find the SP from B to E

So, let's start by find the SP from B to any of the adjacent vertices!

Initialization and Update

- The shortest path from the starting vertex v to v , $D[v]=0$
- If all edge weights are positive, then the least cost edge incident to v , say (v, u) , defines $D[u]$
- When we establish the shortest path from v to a new node u , we go through each of its incident edges to see if there is a better way from v to other nodes through u

More Details

$D[u]$: Stores the **length of the best path** found so far from v to u

Initially $D[v]=0$, $D[u]=\infty$ for all $u \neq v$

Define the set $C=\emptyset$ (called **cloud**)

At each iteration

1. **Select vertex u not in C with the smallest $D[u]$ and pull u into C**
2. We update the label $D[z]$ for each vertex z adjacent to u and outside of C . This update is called '**Edge Relaxation**'

Building Block – Initialize Single Source

Initially $D[v]=0$, $D[u]=\infty$ for all $u \neq v$

INITIALIZE-SINGLE-SOURCE(G, s)

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


Building Block – Relax Edge

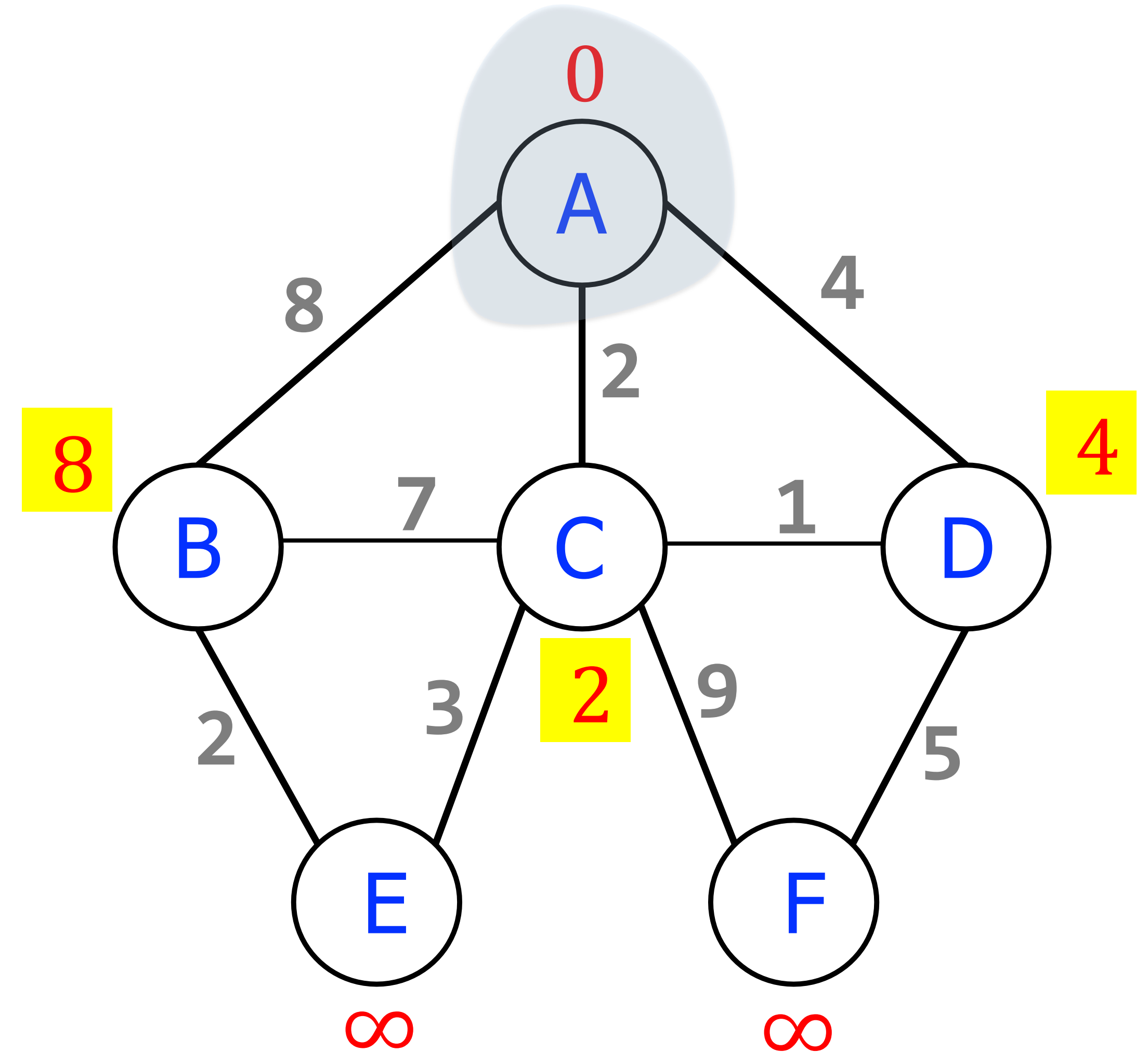
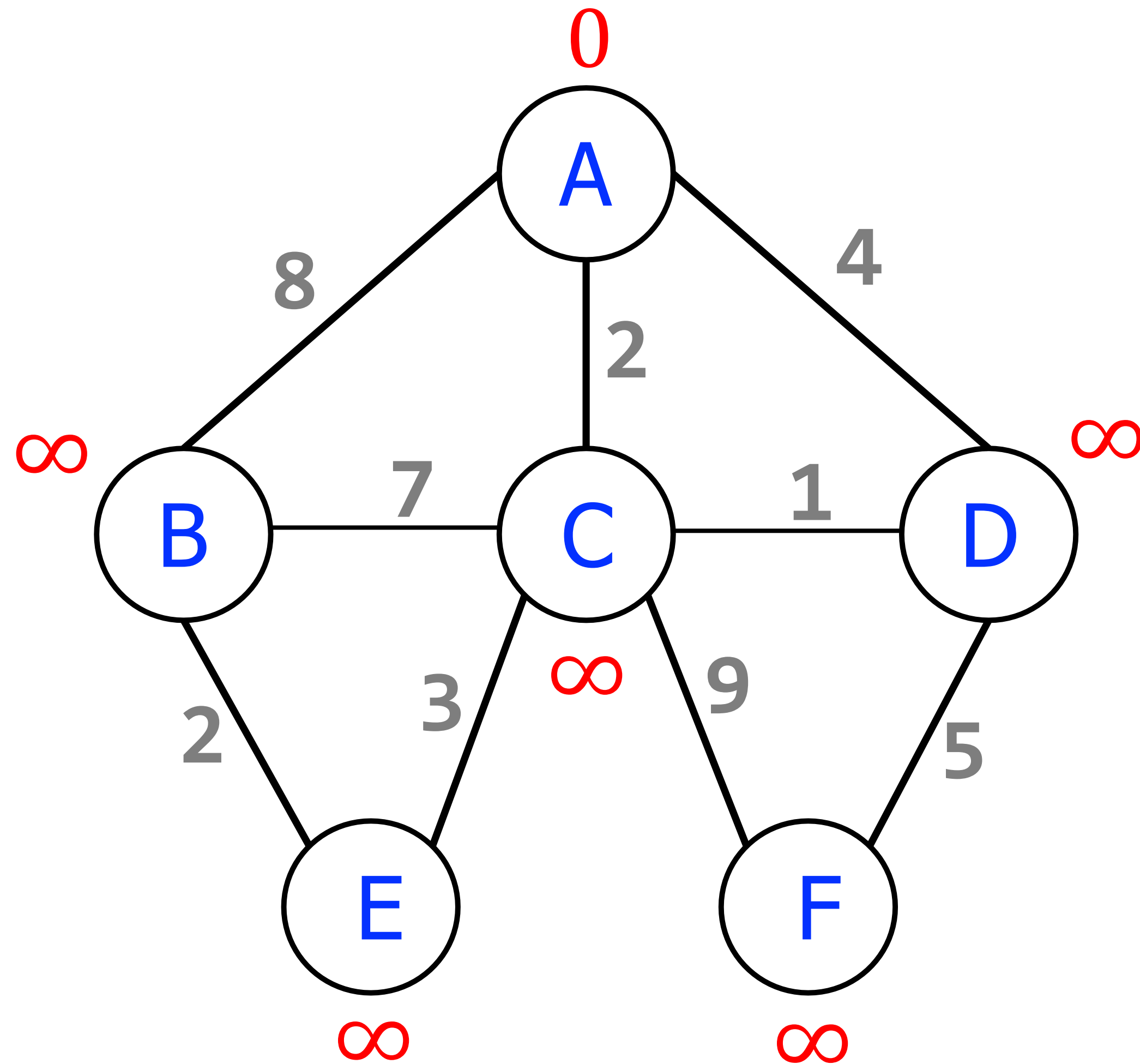
At each iteration

1. Select vertex u not in C with the smallest $D[u]$ and pull u into C
2. We update the label $D[z]$ for each vertex z adjacent to u and outside of C . This update is called ‘Edge Relaxation’

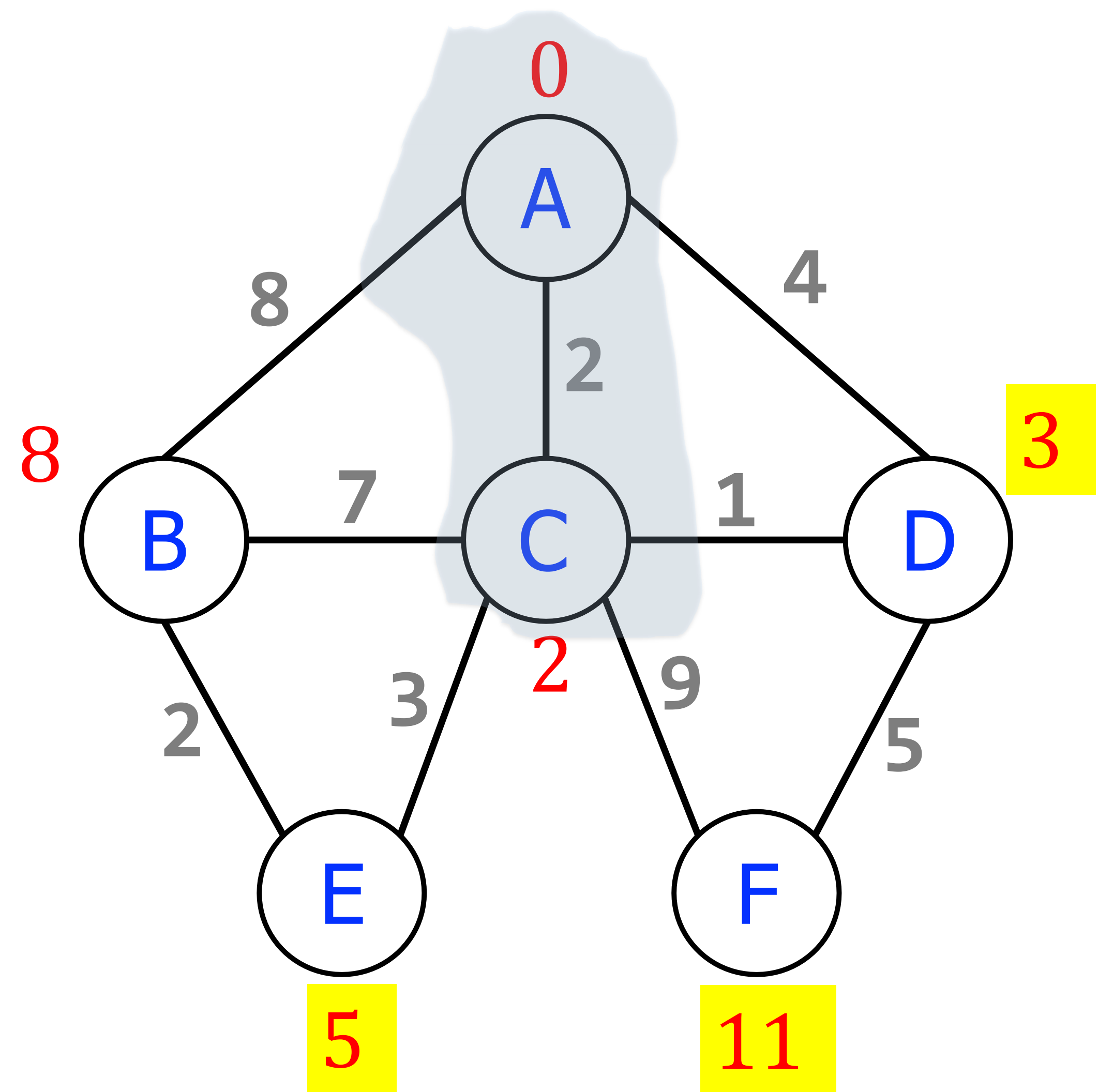
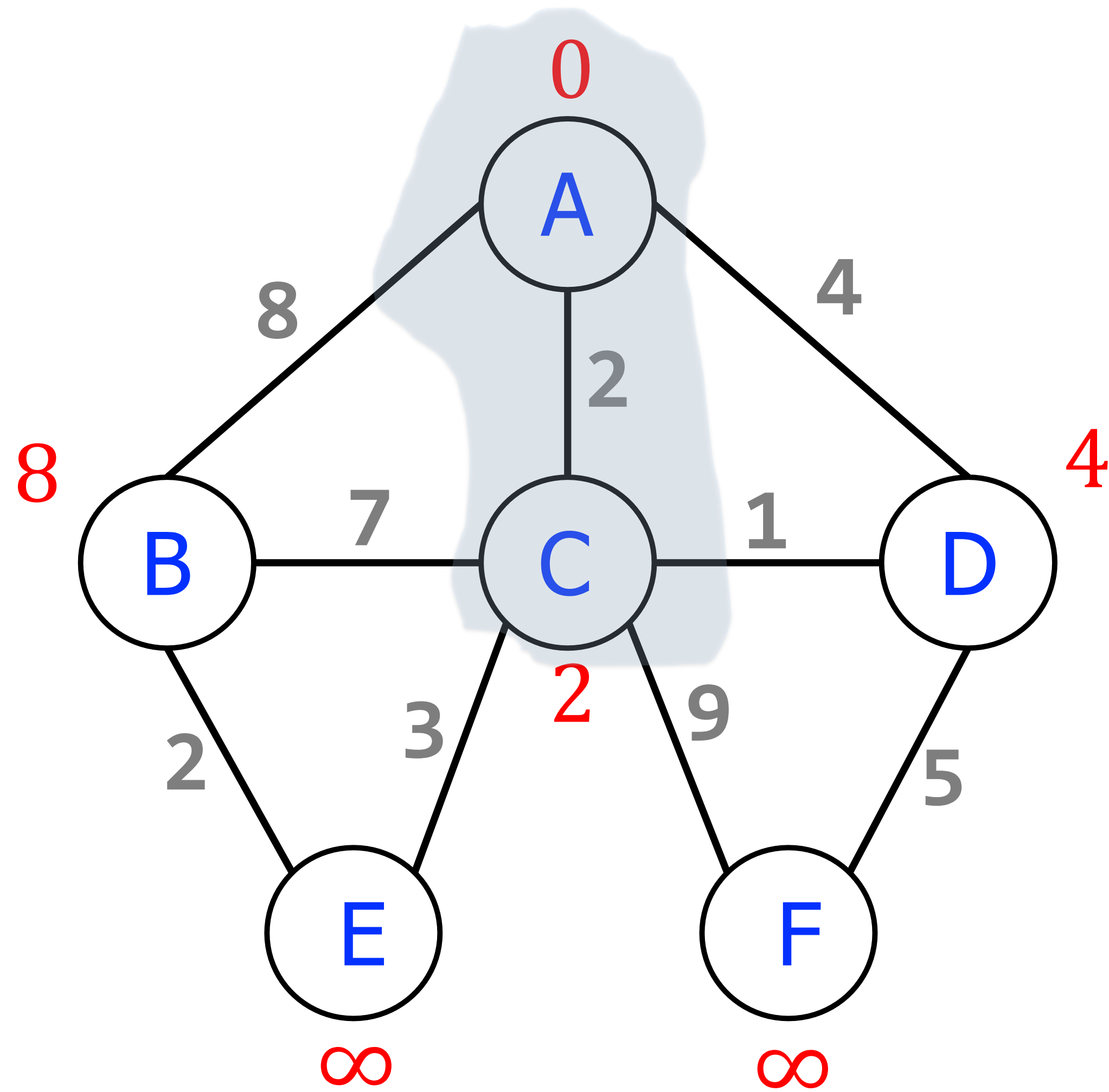
$\text{RELAX}(u, v, w)$

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

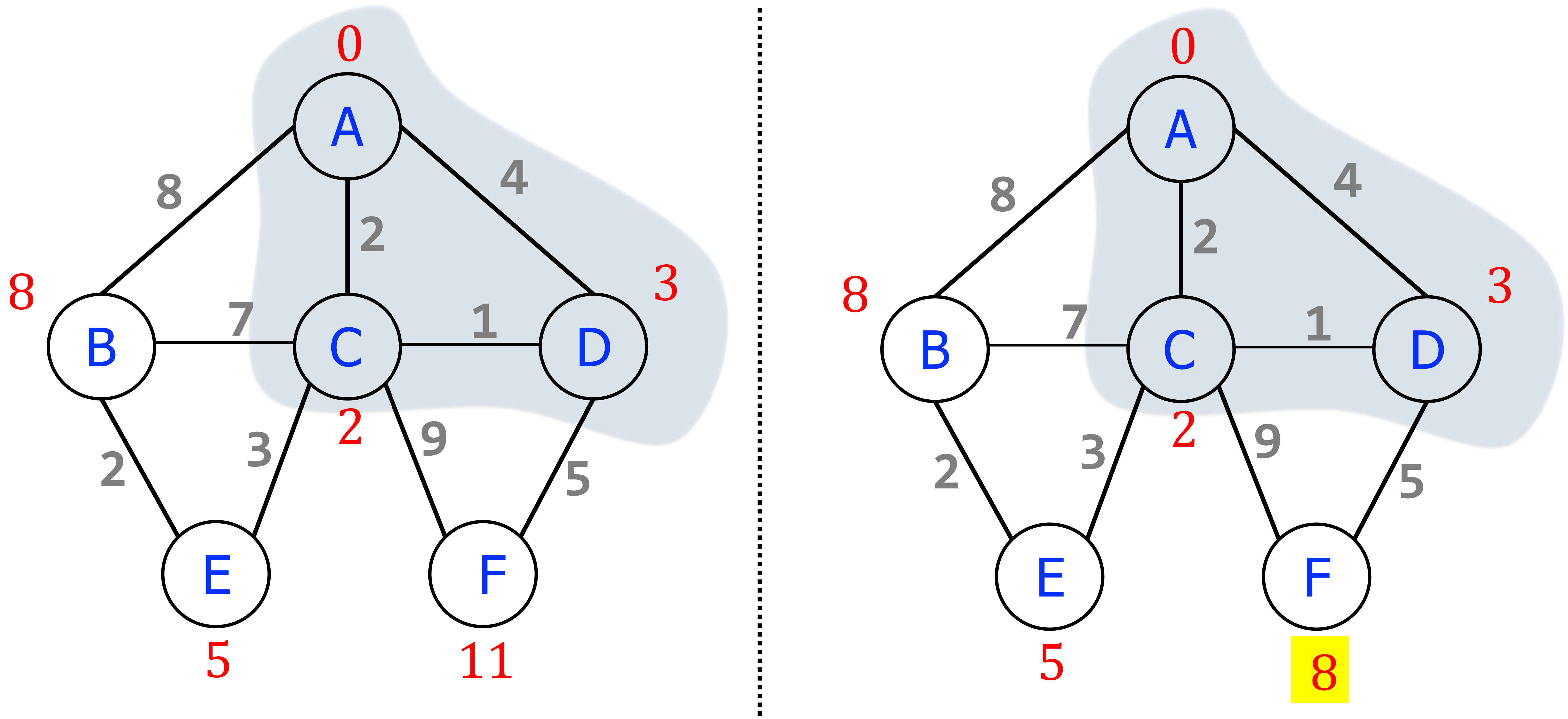
Dijkstra's Algorithm Example



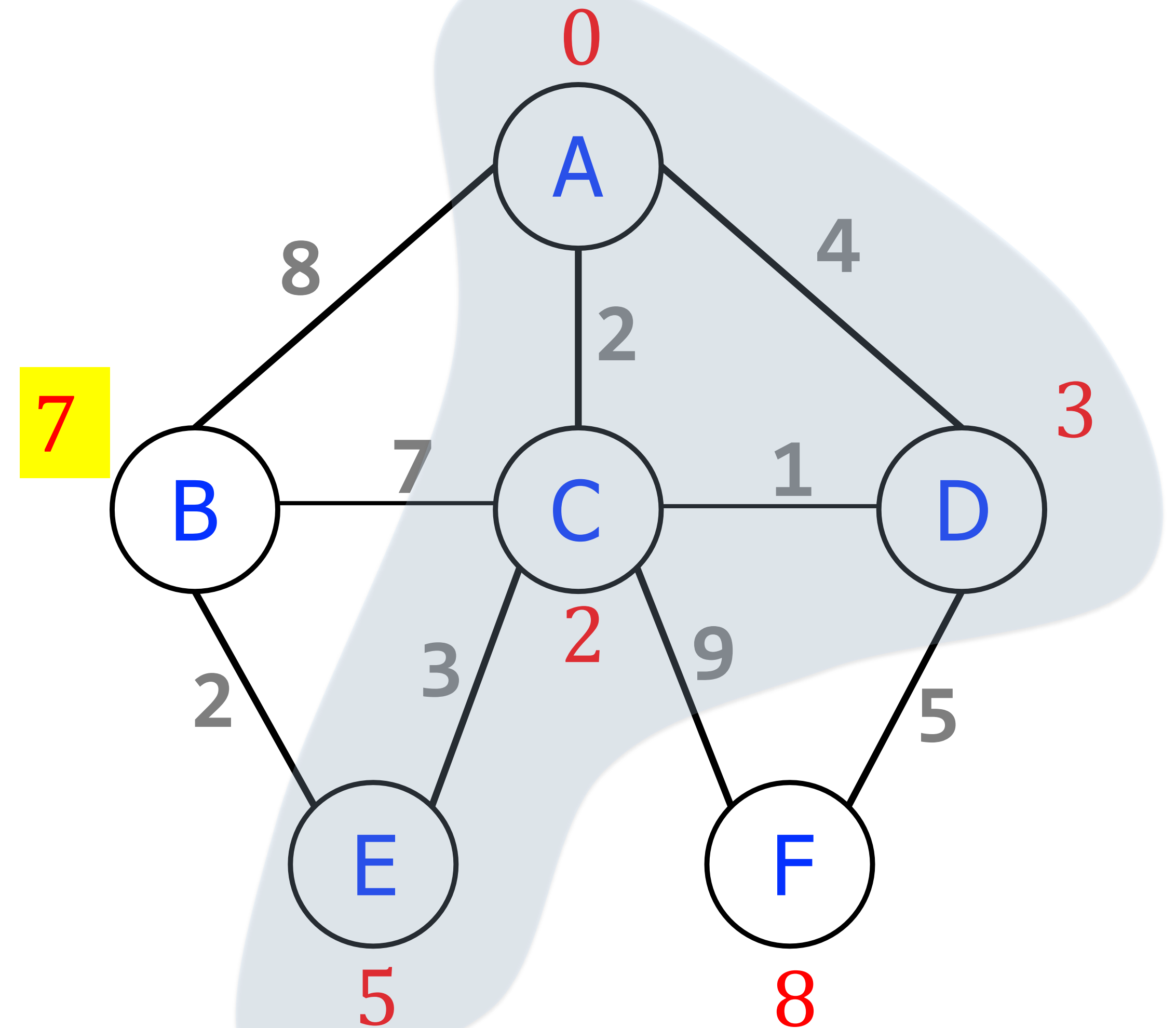
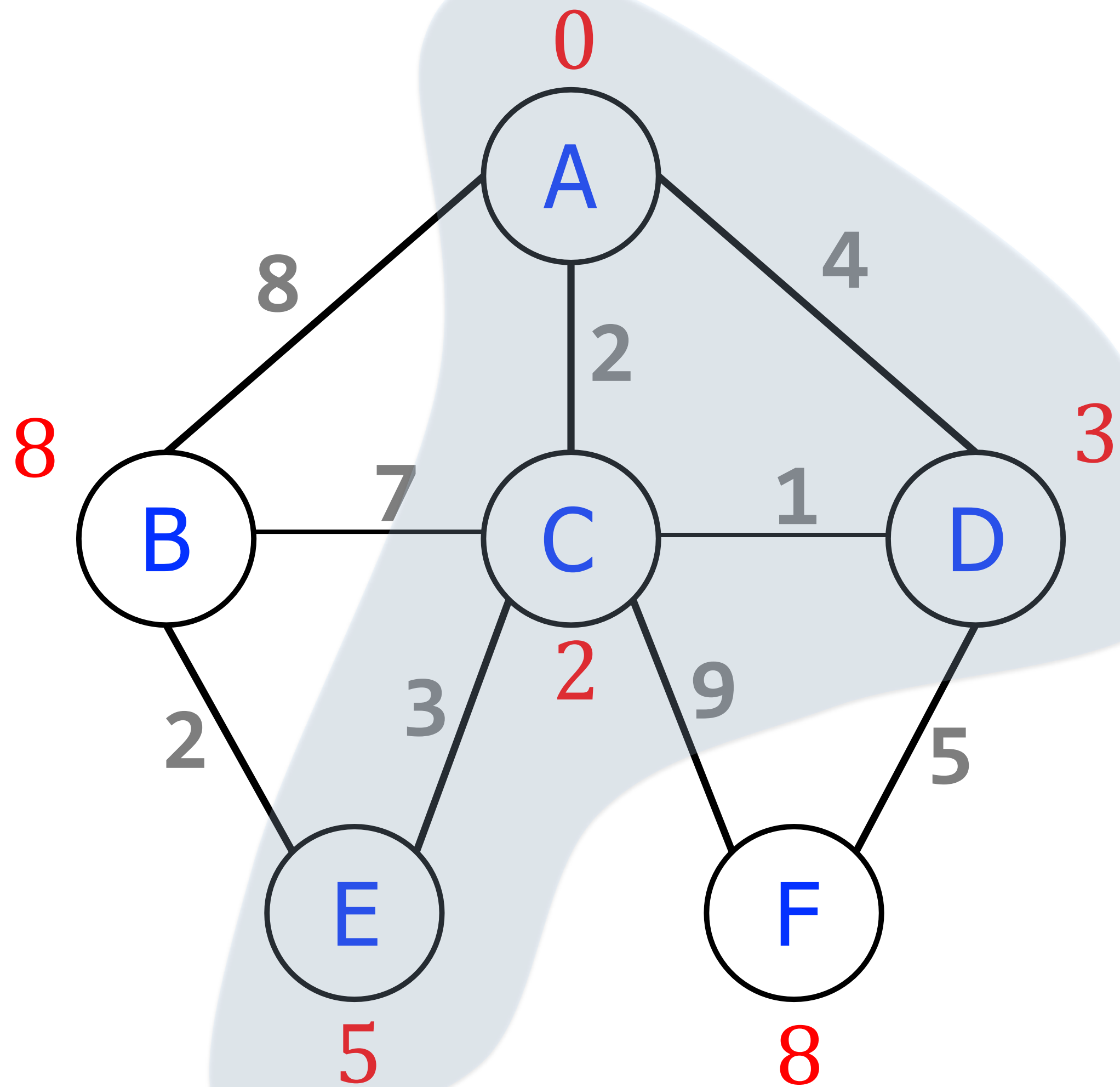
Dijkstra's Algorithm Example



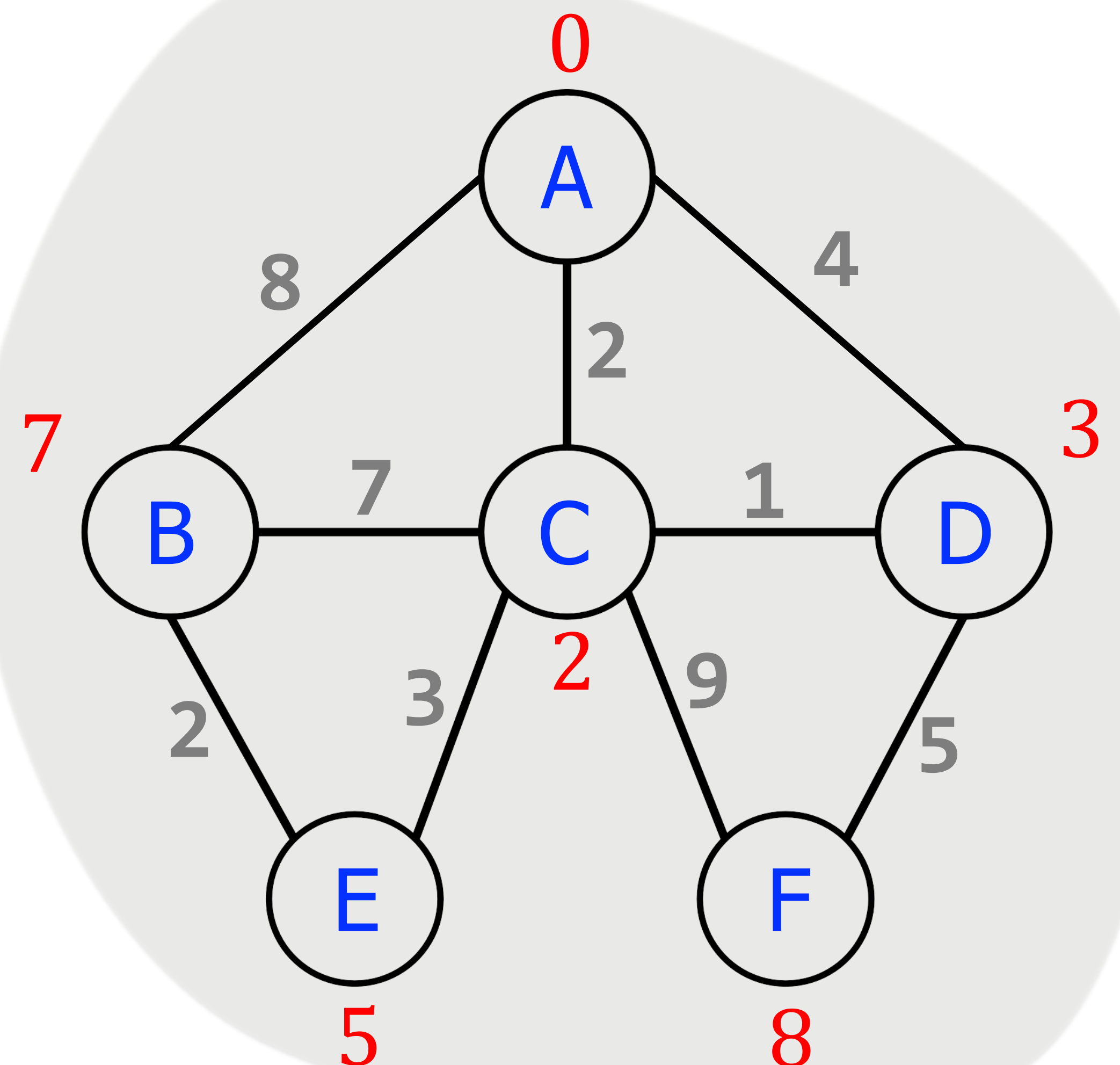
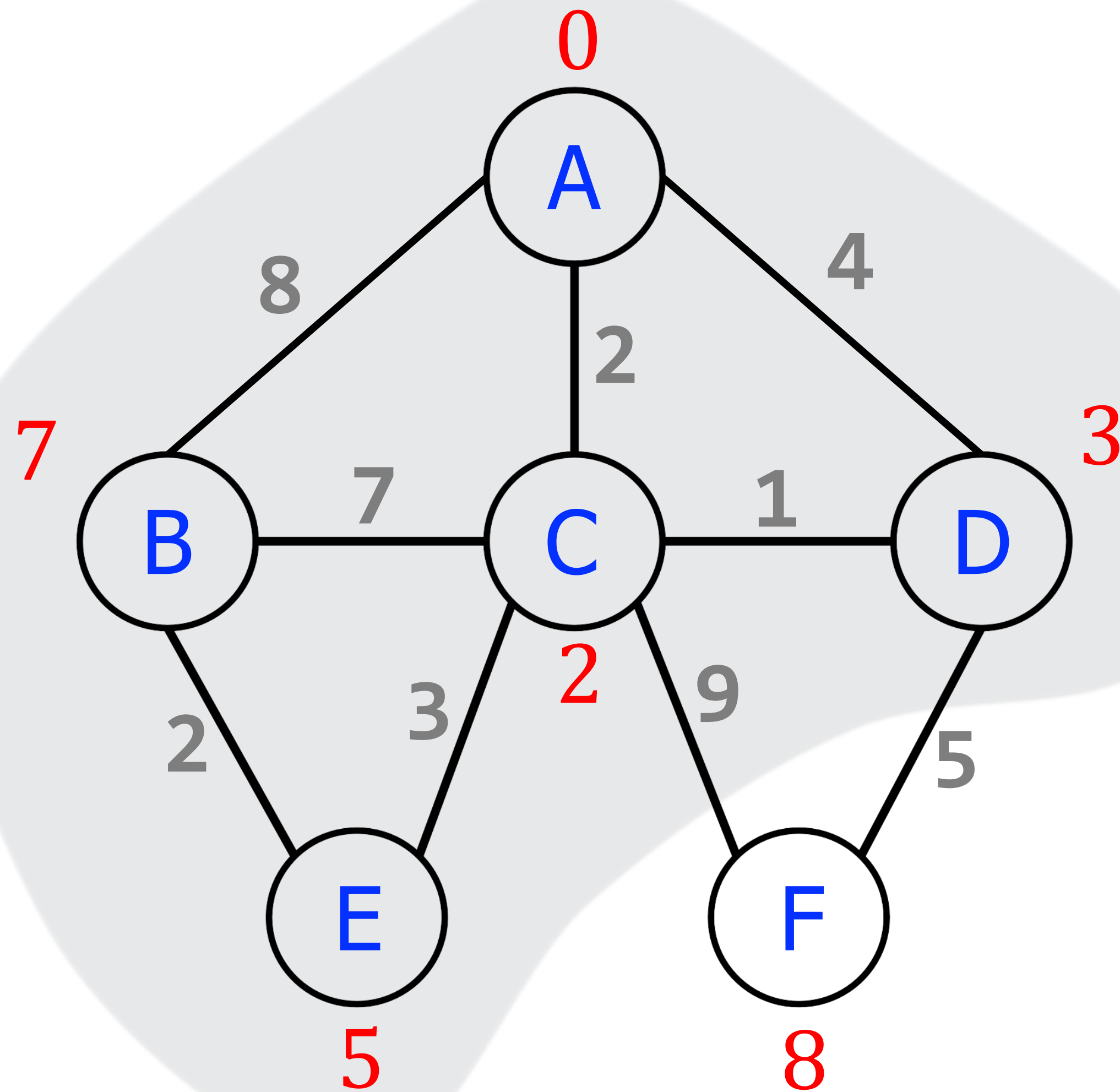
Dijkstra's Algorithm Example



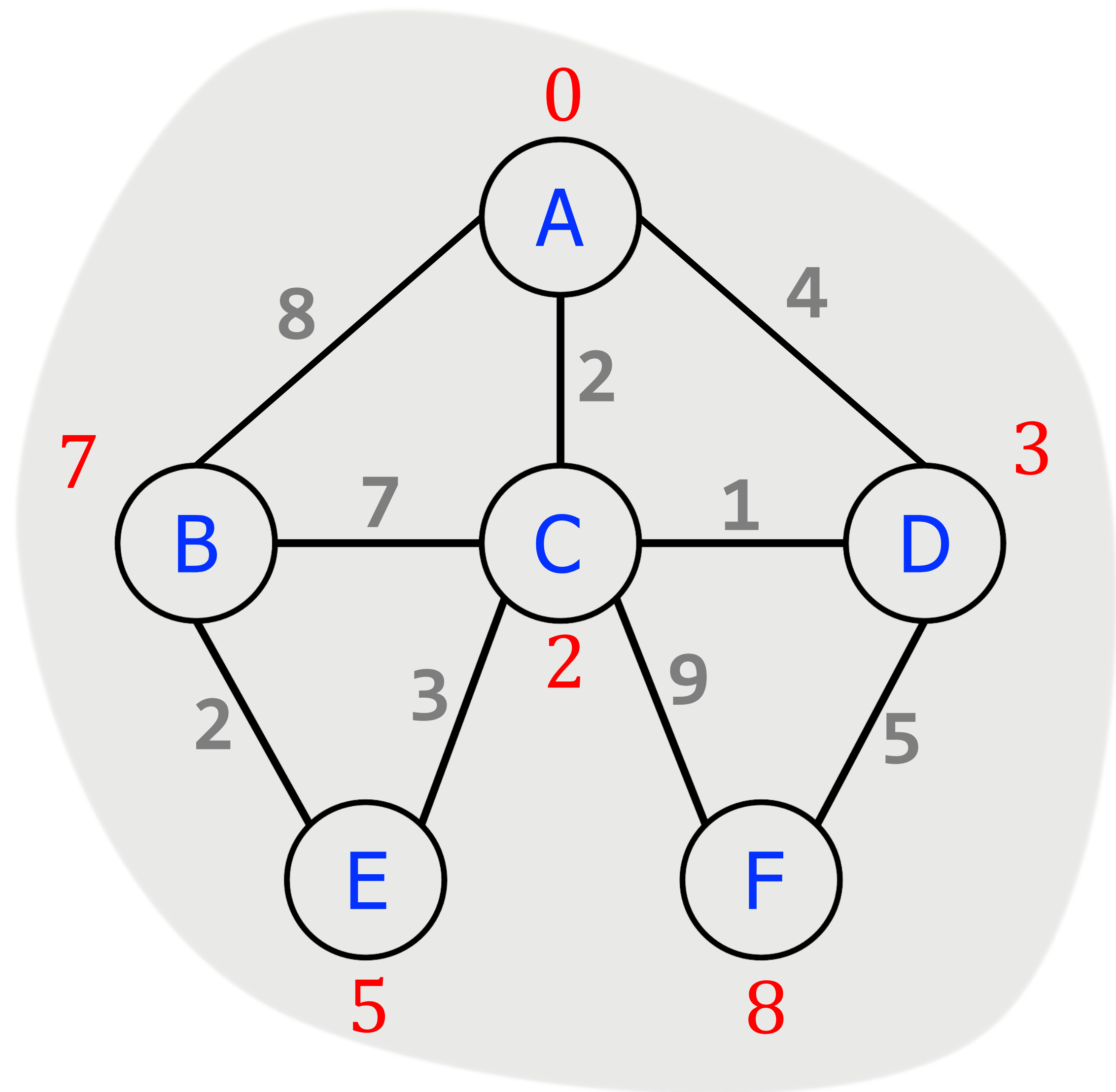
Dijkstra's Algorithm Example



Dijkstra's Algorithm Example



Dijkstra's Algorithm Example



| Path | Shortest Path & Cost |
|-------------------|--|
| $A \rightarrow B$ | $A \rightarrow C \rightarrow E \rightarrow B \mid 7$ |
| $A \rightarrow C$ | $A \rightarrow C \mid 2$ |
| $A \rightarrow D$ | $A \rightarrow C \rightarrow D \mid 3$ |
| $A \rightarrow E$ | $A \rightarrow C \rightarrow E \mid 5$ |
| $A \rightarrow F$ | $A \rightarrow C \rightarrow D \rightarrow F \mid 8$ |

Pseudocode & Efficiency

```
dijkstra(Graph G, Node v) {
```

```
  for each vertex:
```

```
     $D[u] = \infty$ ,
```

```
    known[u]=false
```

```
   $D[v] = 0$  //source vertex
```

```
  build-heap with all vertices //using  $D[u]$  as key
```

```
  while(heap is not empty) {
```

```
    b = removeMin()
```

```
    Known[b] = true //moves inside the cloud
```

```
    for each edge (b,a) in G {
```

```
      if(!a.known) //it is outside the cloud
```

```
        if( $D[b] + \text{weight}((b,a)) < D[a]$ ){
```

```
           $D[a] = D[b] + \text{weight}(b,a)$  //decreaseKey in heap
```

```
          path[a] = b //for computing actual paths
```

```
        }
```

```
      }
```

```
    }
```

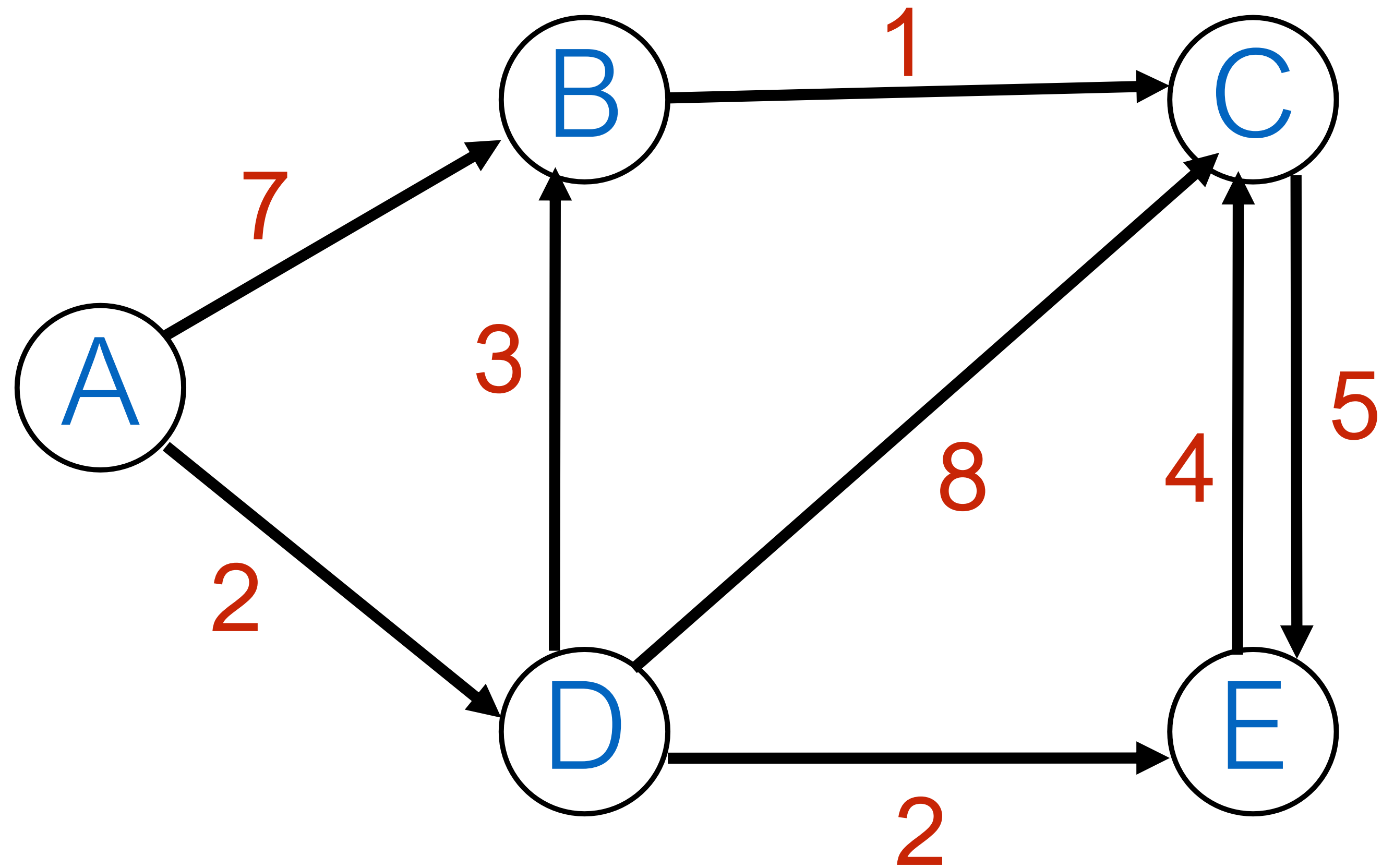
$O(|V|)$

$O(|V| \log |V|)$

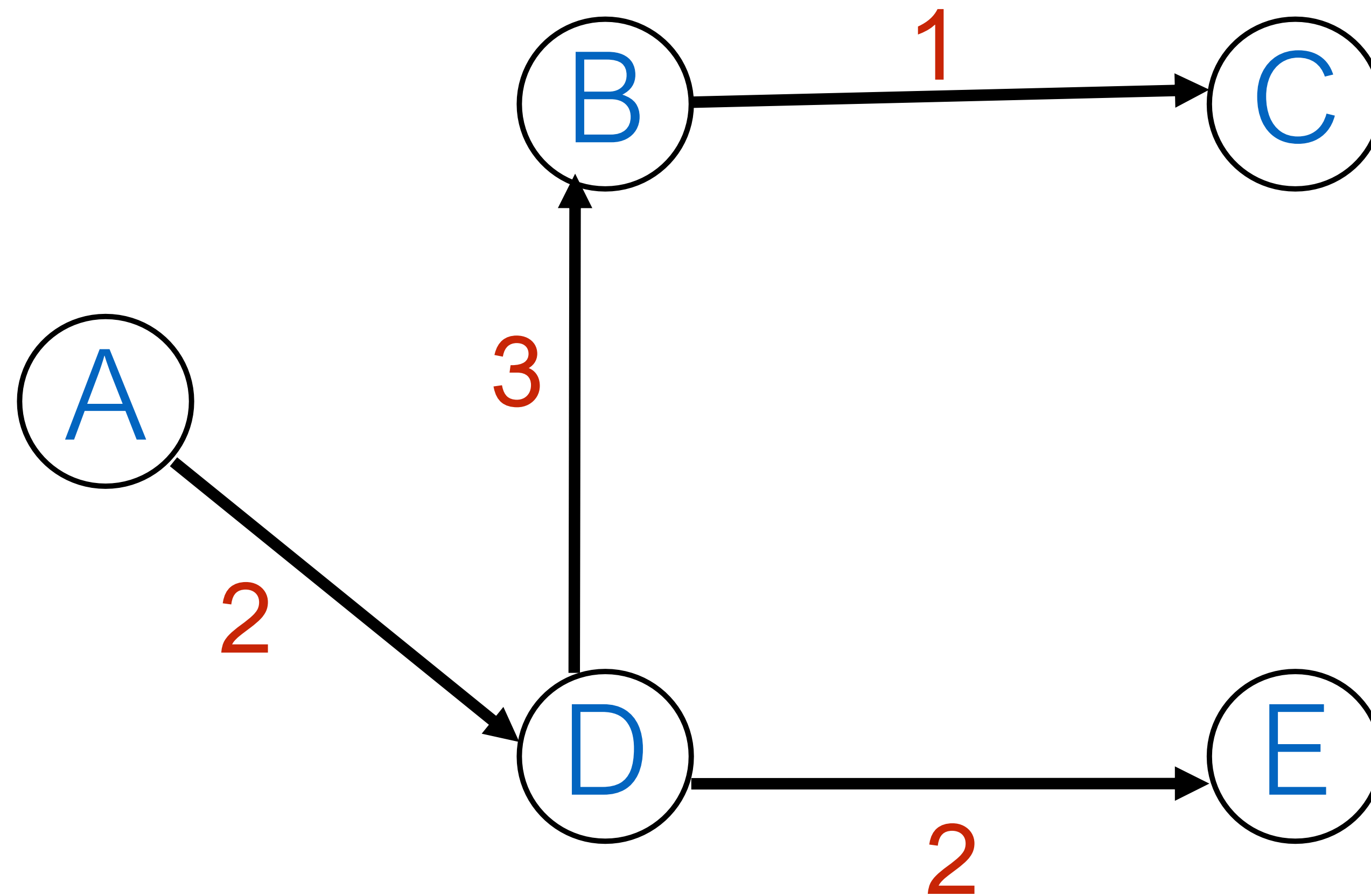
$O(|E| \log |V|)$

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

Let's Try another Example



Example



Questions

