

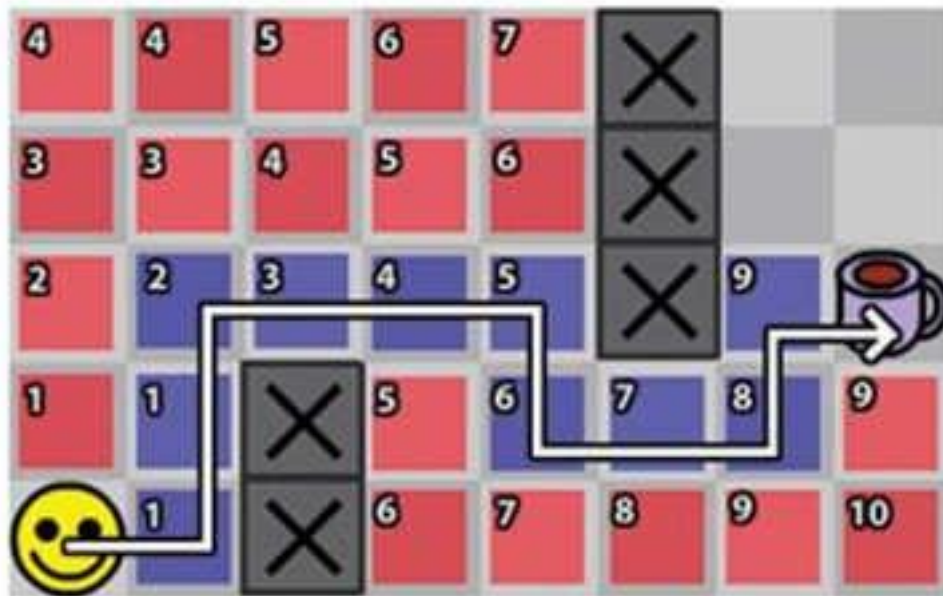
# DATA STRUCTURES AND ALGORITHMS

## Spring 2025

Graph Part II

Lecturer: Do Thuy Duong

# Pathfinding in a graph



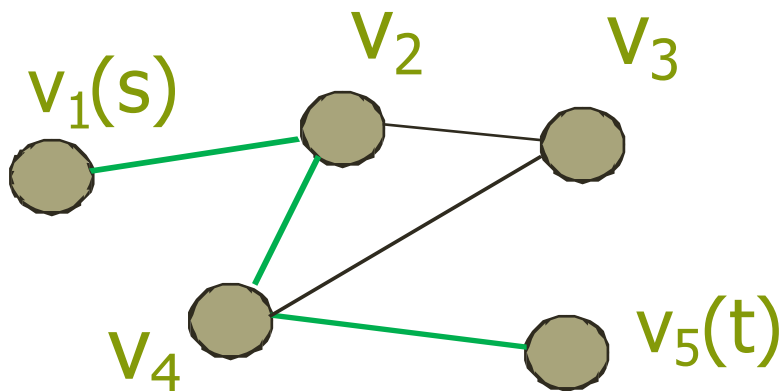
# Content

- Pathfinding in a graph
- Shortest path problem
  - Dijkstra algorithm
  - Bellman-Ford algorithm
  - Floyd-Warshall algorithm

# Pathfinding in a graph

- **Pathfinding problem**

- In a graph  $G=(V,E)$ , find a path from vertex  $s$  to vertex  $t$ .
  - Find a path
  - Find all possible paths
- Graph traversal algorithms can be used
  - **BFS**
  - **DFS**



# Pathfinding using BFS [1]

## Code

**Algorithm** BFS( $V, E, s, t$ ): Boolean

*Input:* A graph with  $V$  &  $E$  are the set of vertices and edges,  $s$  is the source vertex,  $t$  is the destination vertex.

*Output:* A path from  $s$  to  $t$  if exist, otherwise return false.

**For each**  $v$  *in*  $V$  **do**

$\text{Color}[v] \leftarrow \text{Black}$

$\text{Path}[v] \leftarrow \text{null}$

$\text{Color}[s] \leftarrow \text{Yellow}$

$Q \leftarrow \text{new empty queue}$

$Q.\text{Enqueue}(s)$

...

# Pathfinding using BFS [2]

## Code

```
while Q is not empty do  
     $u \leftarrow Q.\text{Dequeue}()$   
  
    If u equals to t then  
        showPath(s, t)  
        return true  
  
    For each v adjacent to u do  
        If Color[v] is Black then  
             $\text{Color}[v] \leftarrow \text{Yellow}$   
             $Q.\text{Enqueue}(v)$   
             $\text{path}[v] \leftarrow u$   
  
    endwhile  
return false
```

# Pathfinding using BFS [3]

## Code

**Algorithm** showPath(s, t)

*Input:* Source vertex s, Destination vertex t.

*Output:* print the path from s to t if exist.

$u \leftarrow t$

**while**  $u \neq s$  **do**

    Print u

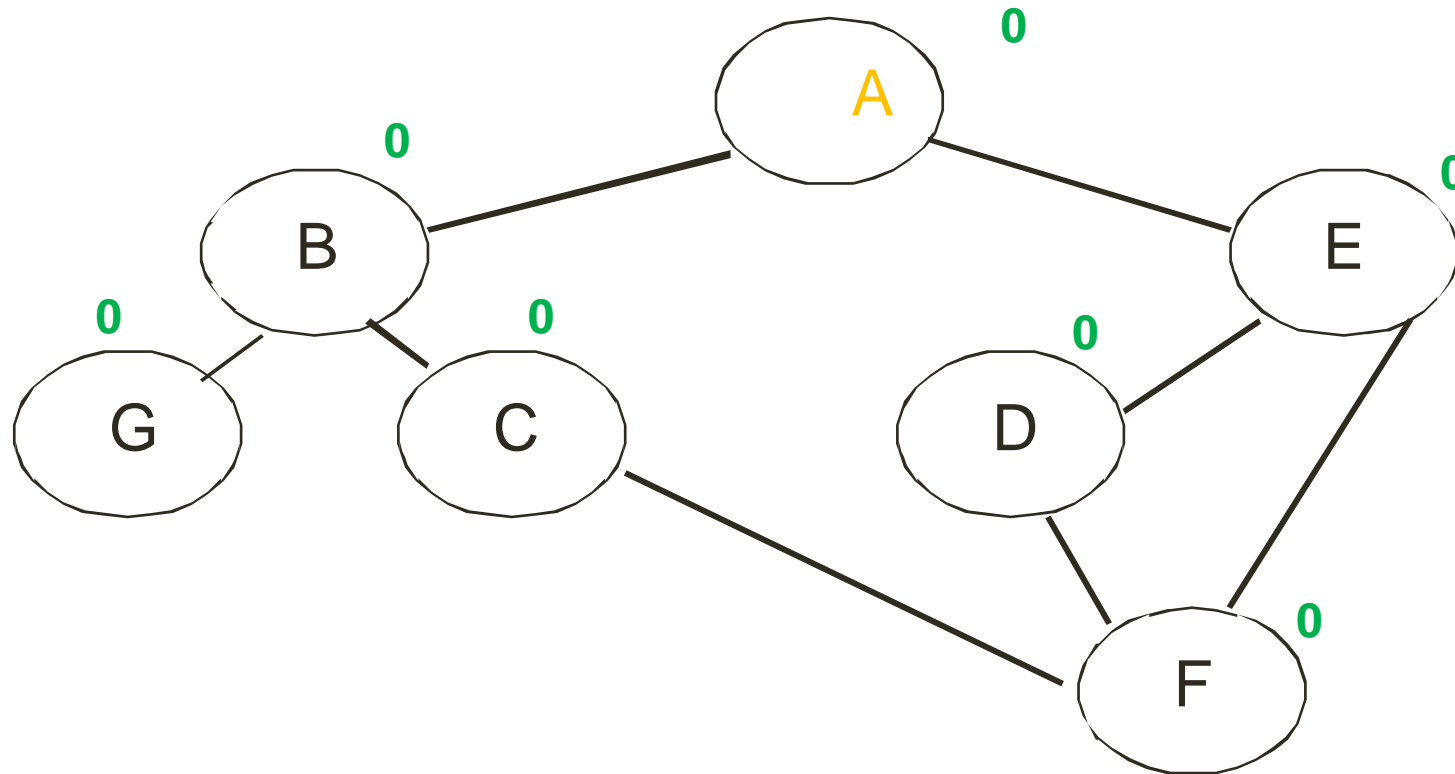
$u \leftarrow \text{path}[u]$

# Pathfinding using BFS [4]

Find path from A to D

Start at A. Mark A yellow. Put A to the queue

Queue: A

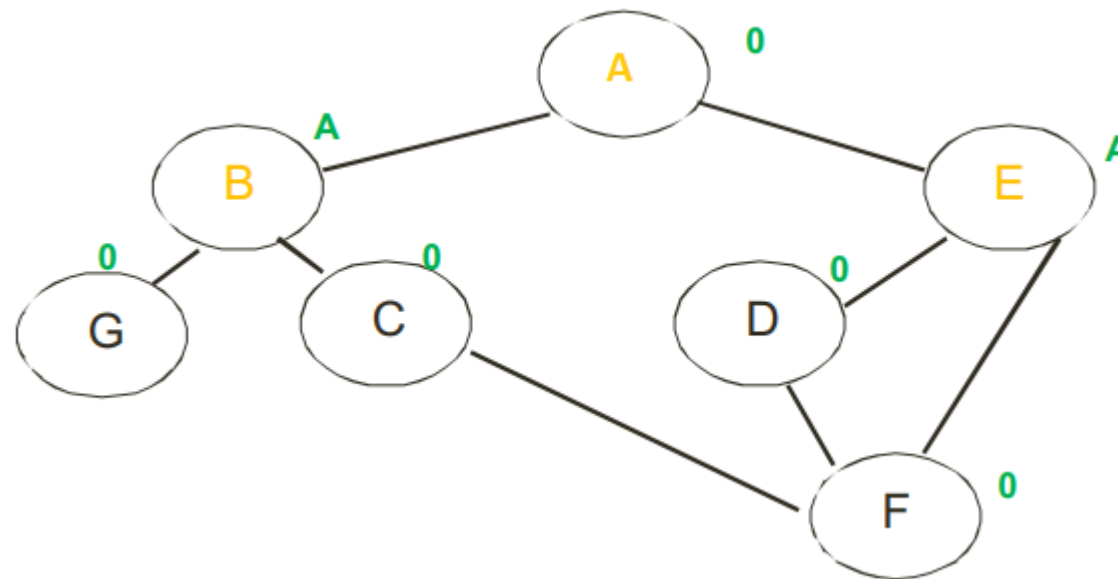




# Pathfinding using BFS [5]

- Take A from Queue,  $A \neq D$
- Found 2 black vertices B and E, that are adjacent to A
- Mark B & E yellow. Put B & E to queue, update path[B] & path[E]

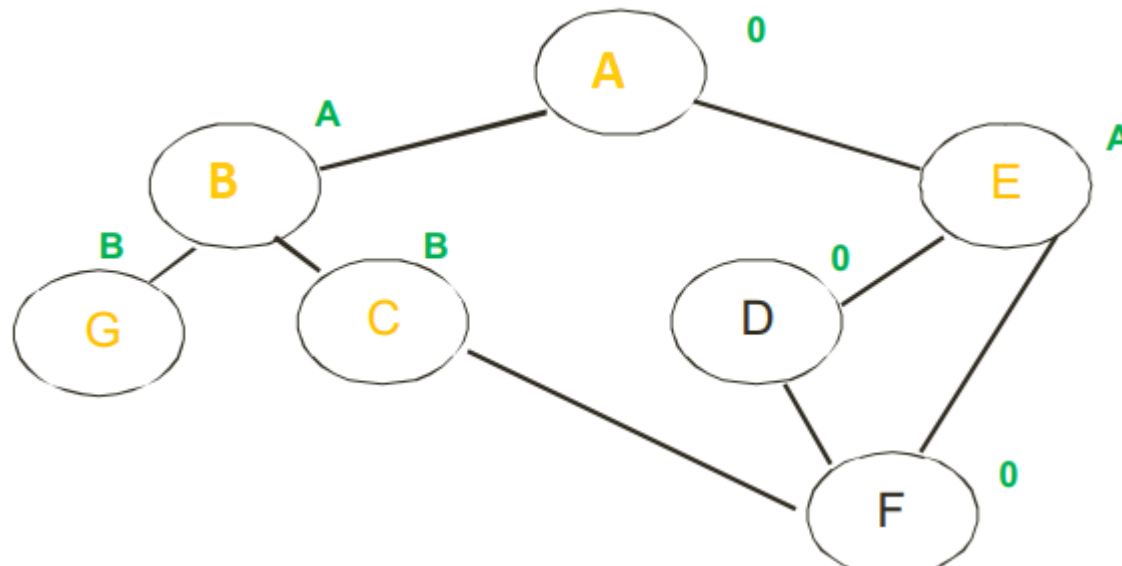
Queue: ~~A~~ B E



# Pathfinding using BFS [6]

- Take B from Queue,  $B \neq D$
- Found 2 black vertices G and C, that are adjacent to B
- Mark G & C yellow. Put G & C to queue, update path[G] & path[C]

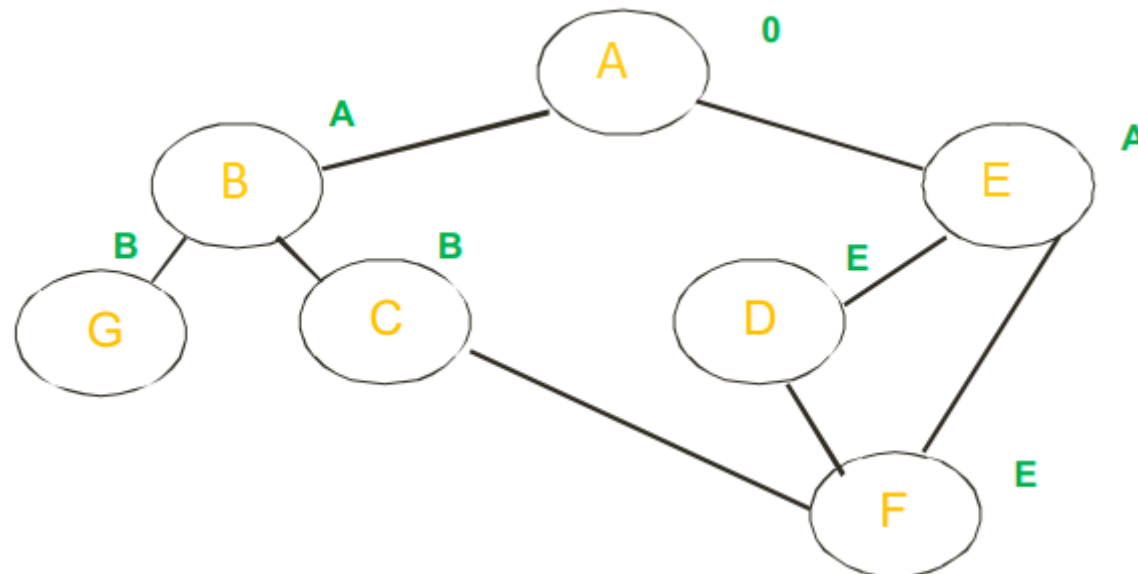
Queue: ~~A~~ B E G C



# Pathfinding using BFS [7]

- Take E from Queue,  $E \neq D$
- Found 2 black vertices D and F, that are adjacent to E
- Mark D & F yellow. Put D & F to queue, update path[D] & path[F]

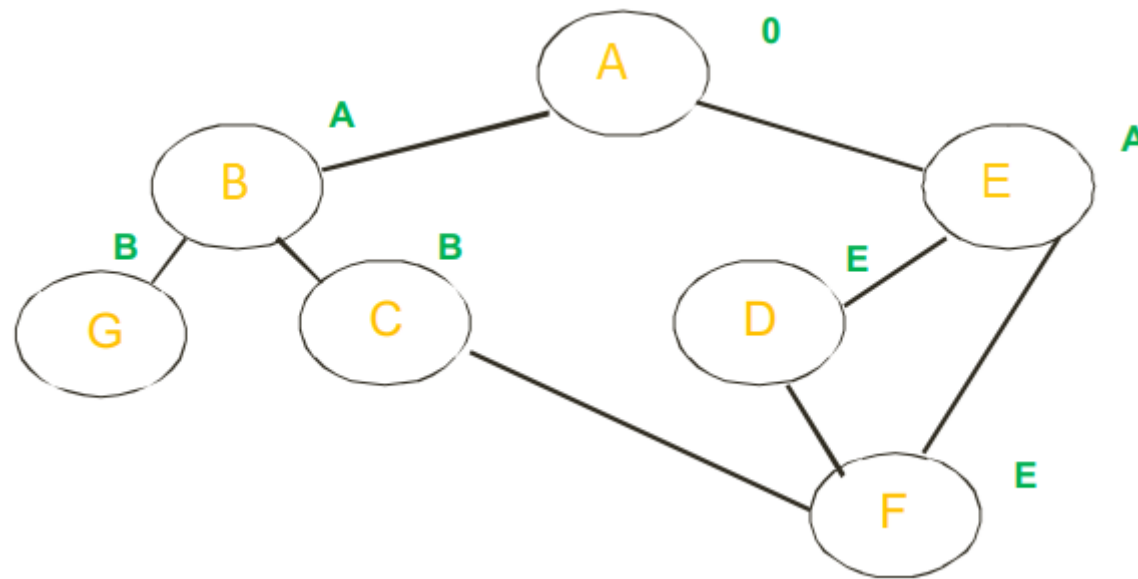
Queue: ~~A~~~~B~~~~E~~ G C D F



# Pathfinding using BFS [8]

- Take G from Queue,  $G \neq D$
- Nothing put to the queue

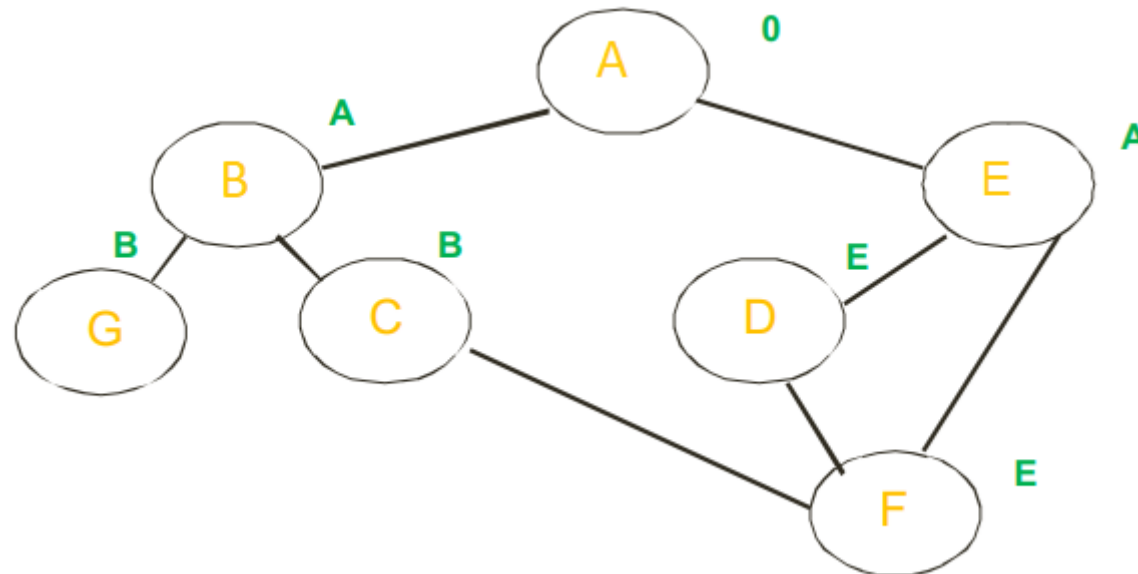
Queue: ~~A~~~~B~~~~E~~~~G~~C D F



# Pathfinding using BFS [9]

- Take C from Queue,  $C \neq D$
- No black vertices adjacent to C. Nothing put to the queue

Queue: ~~A~~~~B~~~~E~~~~G~~~~C~~~~D~~ F

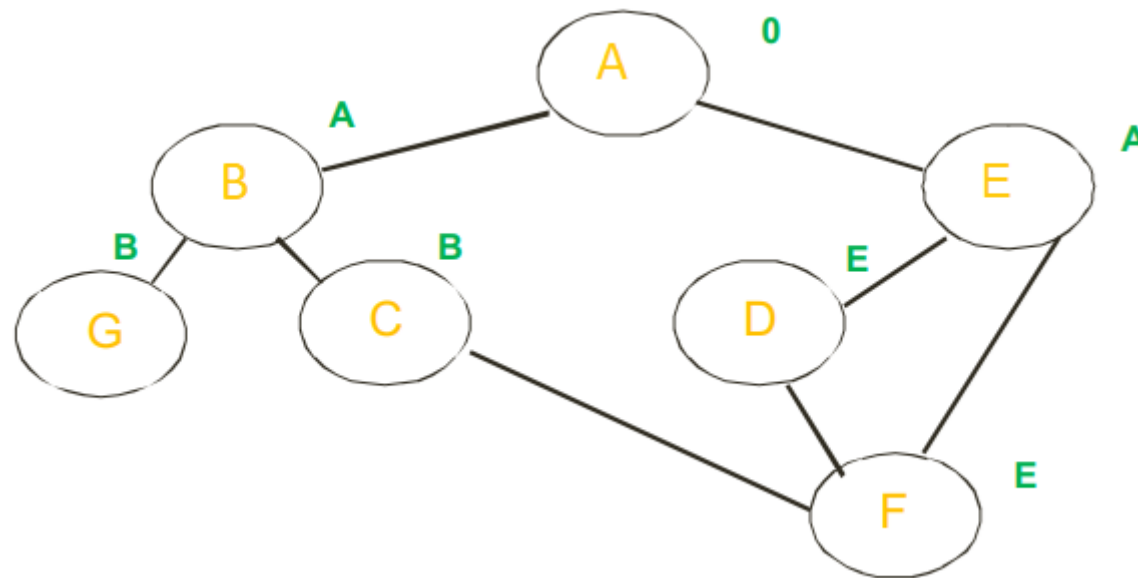


# Pathfinding using BFS [10]

Take D from the queue. D is the destination vertex. Show path and return true.

Queue: ~~A~~~~B~~~~E~~~~G~~~~C~~~~D~~~~F~~

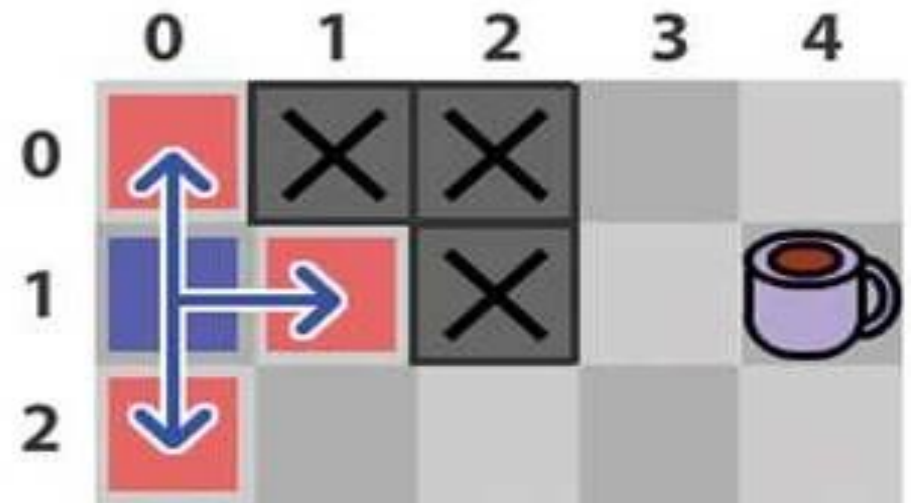
→ Path: A E D



# Find my coffee game [1]

## Introduction

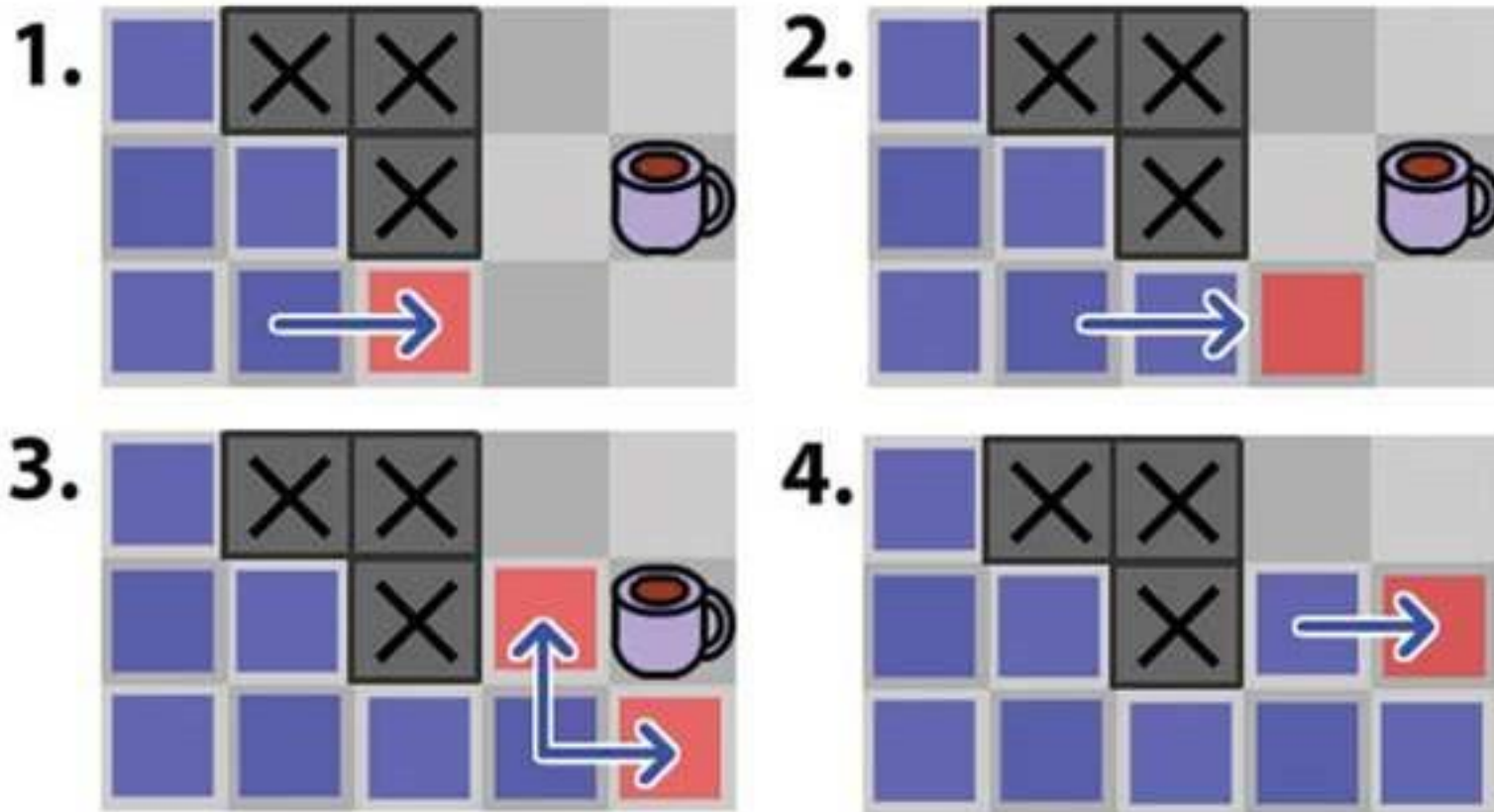
- Start from a position in a map
- Find the cup of coffee



# Find my coffee game [2]

## Introduction

- BFS is a solution





# Find my coffee game [3]

## Data representation

- The map
- Using a matrix  $M \times N$  (M rows, N columns)

	0	1	2	3	4
0	G	X	X	G	G
1	Y	G	X	G	C
2	G	G	G	G	G

# Find my coffee game [4]

## Data representation

- The graph
- Each cell in the map is a node of the graph. Nodes are indexed from 0 to  $M*N - 1$ .
- With node  $v$ , the corresponding row and column are:  
 $\text{row} = v / N$ ;  $\text{column} = v \% N$ ;

	0	1	2	3	4
0	0	1	2	3	4
1	5	6	7	8	9
2	10	11	12	13	14

# Find my coffee game [5]

## Data representation

- Each node has a list of adjacent nodes.

- Node 0:

- {5}

- Node 5:

- {0, 6, 10}

- Node 9:

- {8, 4, 14}

	0	1	2	3	4
0	G	X	X	G	G
1	Y	G	X	G	C
2	G	G	G	G	G

	0	1	2	3	4
0	0	1	2	3	4
1	5	6	7	8	9
2	10	11	12	13	14

# Find my coffee game [6]

- **Vertex ADT**

## GVertex

- index : int //Position in the vertices list
- row, column : int //Which cell in the map
- marked : boolean //Used in BFS
- numOfAdjVertex : int
- adjVertex : int[ ] //List of adjacent vertices

---

+ GVertex(index, map, m, n) : void

+ getIndex() : int

+ getRow() : int

+ getColumn() : int

+ ...

# Find my coffee game [7]

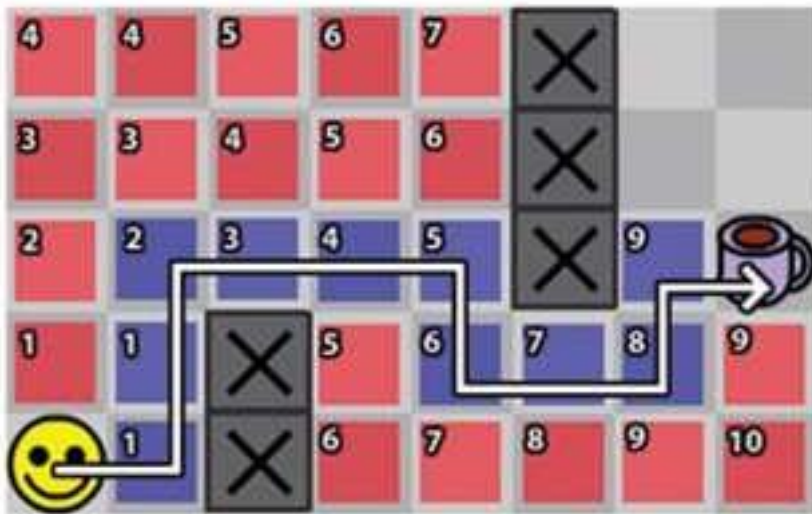
- **Other variables**

- char[][] map (The map)
- GVertex [] listVertex (Array contains all  $M \times N$  vertices)
- GVertex startVertex, endVertex
  - Your position and position of the coffee cup.
- ArrayQueue q
  - Vertex queue used in BFS(each queue's item is a vertex)
- GVertex [] path
  - Path information

# Find my coffee game [8]

- **Implementation**

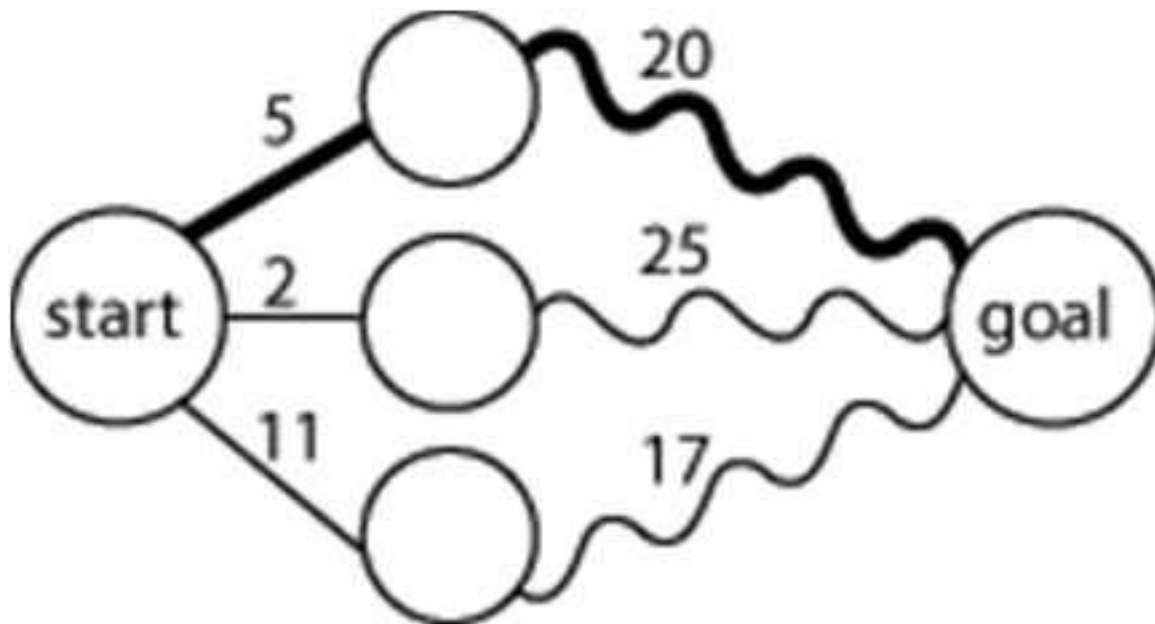
- See Week 12 Example Code for more detail.



	0	1	2	3	4	5	6	7
0	G	G	G	G	G	X	G	G
1	G	G	G	G	G	X	G	G
2	G	G	G	G	G	X	G	C
3	G	G	X	G	G	G	G	G
4	Y	G	X	G	G	G	G	G

(4,0) → (4,1) → (3,1) → (2,1) → (2,2) → (2,3) → (2,4) → (3,4) → (3,5)  
→ (3,6) → (2,6) → (2,7)

# Shortest path



# Recall a weighted graph [1]

- In a weighted graph, each edge has an associated numerical value, called the weight of the edge.
- Edge weights may represent distances, costs, etc.
- Example:
  - In a train route graph, the weight of an edge represents the distance between the endpoint cities





# Recall a weighted graph [2]

	STU	FRA	HAN	COP	BER	LEI
STU	0	210	$\infty$	$\infty$	$\infty$	480
FRA	210	0	350	$\infty$	545	395
HAN	$\infty$	350	0	475	290	$\infty$
COP	$\infty$	$\infty$	475	0	435	$\infty$
BER	$\infty$	545	290	435	0	190
LEI	480	395	$\infty$	$\infty$	190	0

$w[i][i]=0$

$w[i][j]=w[j][i]$  if  $G$  is undirected

$w[i][j]=\infty$  if there is no edge  
connect vertex  $i$  to vertex  $j$



# Shortest path problem [1]

- **Definition**

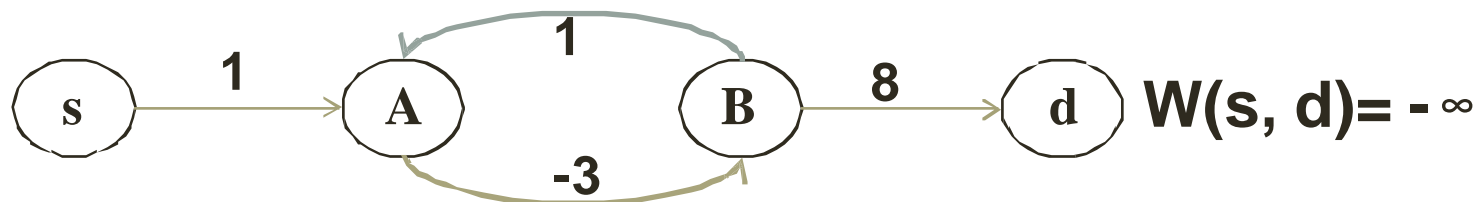
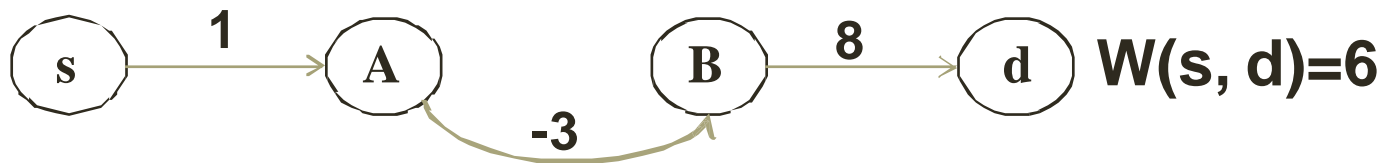
- In a graph  $G=(V,E)$ , consider to a **path from s to t**.
  - $P_k=\{V_1 \rightarrow V_2 \rightarrow V_3 \rightarrow \dots \rightarrow V_k\}$ .
- Let  $W(P_k)$  is the **weight of the path**  $P_k$  we have:
- If  $G$  is an **un-weighted graph**, then:  
 $W(P_k)$ =The number of vertices in  $P_k=k$
- If  $G$  is **a weighted graph**, then:
$$W(P_k) = \sum_{i=1}^{k-1} w(V_i, V_{i+1})$$
- Find a  $P_k$  so that  **$W(P_k)$  is minimum**.

# Shortest path problem [2]

- Shortest path properties
  - A **sub-path** of a shortest path is **itself a shortest path**.
  - For example:  
If  $P_{sp} = \{V_1 \rightarrow V_3 \rightarrow V_6 \rightarrow V_4 \rightarrow V_5 \rightarrow V_7\}$  is the shortest path from  $V_1$  to  $V_7$ , then the path  $\{V_3 \rightarrow V_6 \rightarrow V_4\}$  is the shortest path from  $V_3$  to  $V_4$ .
  - There is a **tree of shortest path** from a vertex to all other vertices that connect to this vertex.

# Shortest path problem [3]

- Shortest path in an un-weighted graph
  - **BFS is a solution**
- Shortest path in a weighted graph
  - Every edge has **positive weight**
  - Edge may have **negative weight**
    - Graph may have **negative weight cycle**



# Shortest path problem [4]

- Shortest path algorithm types:
  - **Single-pair shortest path (SPSP)**
    - Find shortest paths from a given vertex to a given vertex.
  - **Single-source shortest path (SSSP)**
    - Find shortest paths from a given vertex to all other vertices.
  - **All-pairs shortest path (APSP)**
    - Find shortest paths for every pair of vertices.

# Dijkstra algorithm [1]

- A **SSSP algorithm**
  - Find shortest paths from **vertex  $s$**  to all other vertices.
  - **Assumption**
    - The graph is connected.
    - The edge **weights are non-negative**.
  - Definition of the **distance** from  **$s$  to  $v$** 
    - $D[v]$  is the total weight of the shortest path from  $s$  to  $v$ .
    - With a given vertex  $s$ , Dijkstra algorithm will compute the  $D[v]$  for all  $v$  in  $V$ .

# Dijkstra algorithm [2]

- **Dijkstra algorithm idea**

- We grow a “**cloud**” of **vertices**, beginning with  $s$  and eventually covering all the vertices.
- We store with each vertex  $v$  a label  $d(v)$  representing **the distance of  $v$  from  $s$  in the sub-graph** consisting of the cloud and its adjacent vertices
- At each step:
  - We **add** to the cloud the vertex  $u$  outside the cloud with the smallest distance label,  $d(u)$
  - We **update** the labels of the vertices adjacent to  $u$

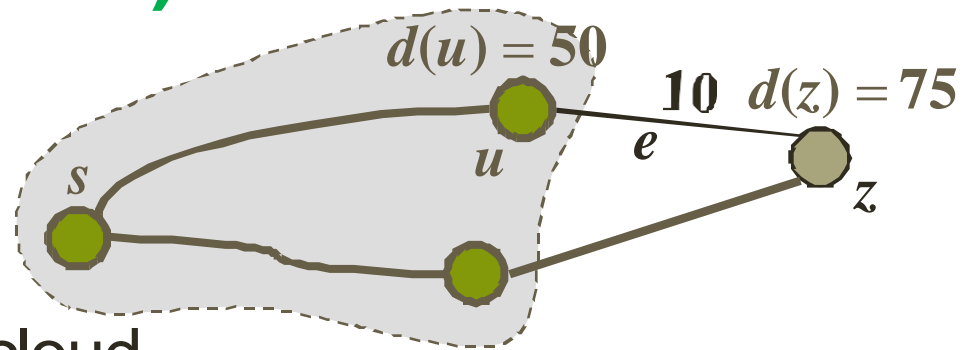
# Dijkstra algorithm[3]

- **Update (Edge relaxation)**

- Consider an edge

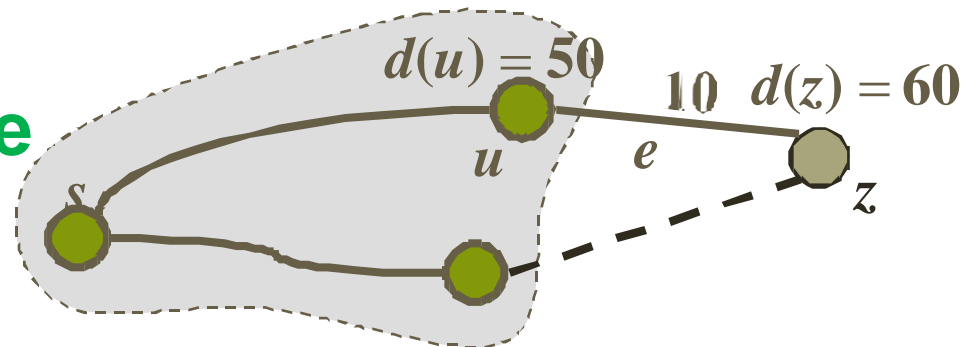
**$e = (u, z)$**  such that:

- **$u$**  is the vertex most recently added to the cloud
- **$z$**  is not in the cloud



- The relaxation of edge  **$e$**  updates distance  **$d(z)$**  as follows:

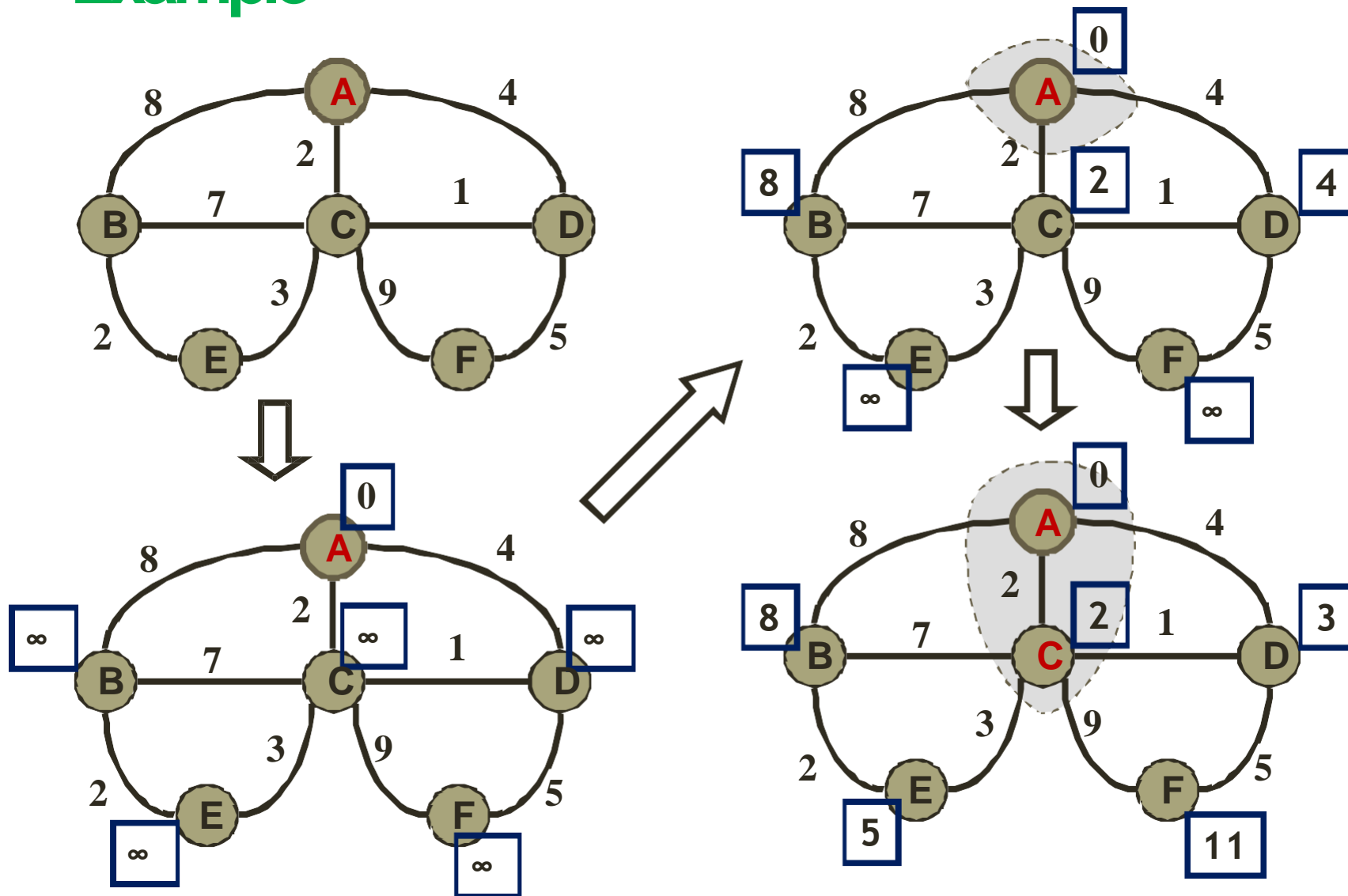
$$d(z) \leftarrow \min\{d(z), d(u) + \text{weight}(u, z)\}$$





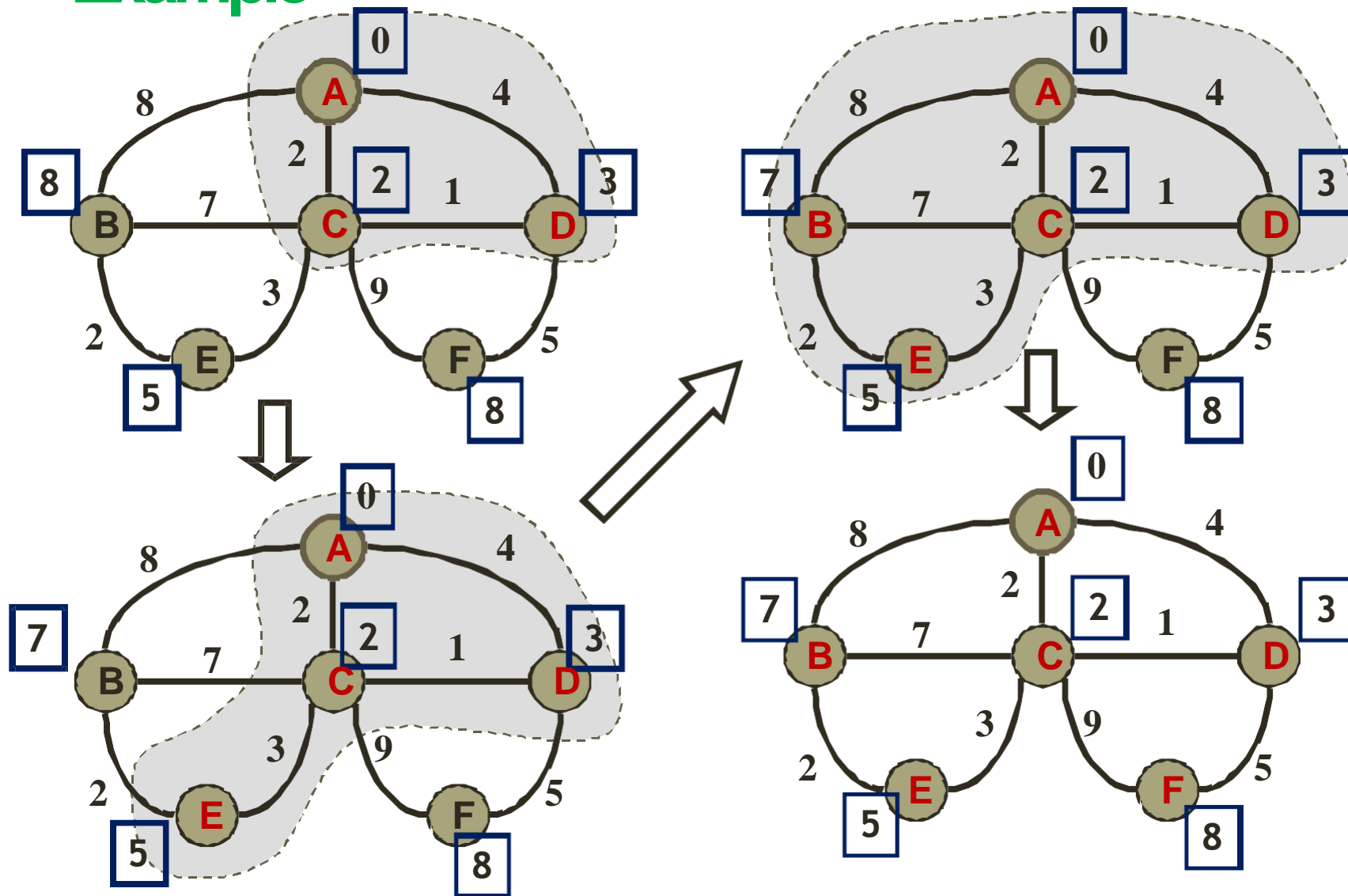
# Dijkstra algorithm[4]

- Example



# Dijkstra algorithm[5]

- Example



# Dijkstra algorithm[6]

- Dijkstra algorithm

## Code

**Algorithm** Dijkstra( $V, E, s$ ):

*Input:* A graph  $G=(V, E)$ ,  $w$  is the weighted matrix,  $s$  is the source vertex.

*Output:* The shortest paths from  $s$  to other vertices.

**For each**  $v$  *in*  $V$  **do**

$D[v] \leftarrow \infty$

$\text{path}[v] \leftarrow -1$

$D[s] \leftarrow 0$

Create empty  $S$  set      //The “cloud” set

**Do**

Find vertex  $u$  in  $V \setminus S$  so that  $D[u]$  is minimum

Move  $u$  from  $V$  to  $S$

**For each**  $z$  *in*  $V \setminus S$  **do**

$D[z] \leftarrow \min\{D[z], D[u] + w[u][z]\}$

$\text{path}[z] \leftarrow u$  //Only update if  $D[u] + w[u][z] < D[z]$

**While** could not find any  $u$

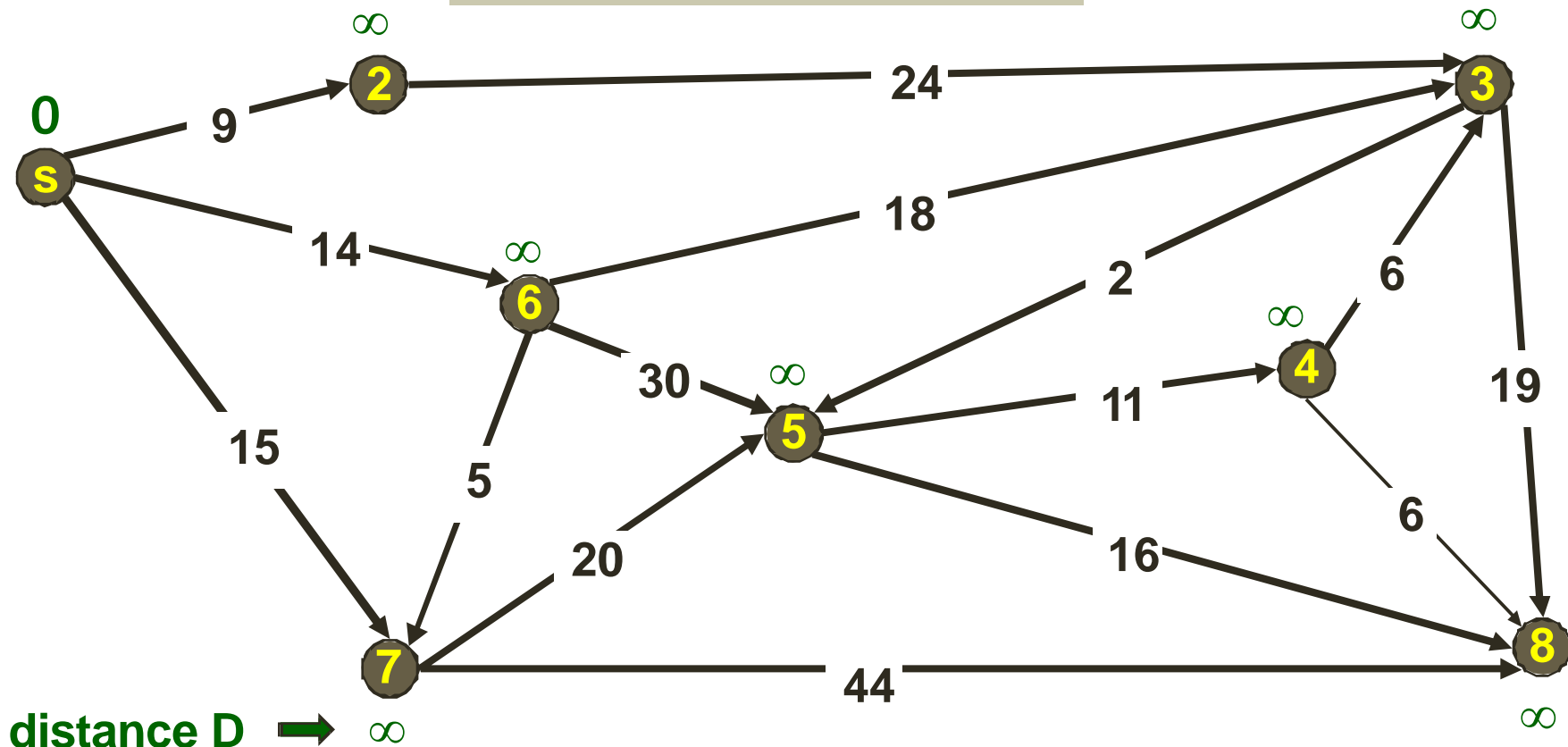
# Dijkstra algorithm[7]

- Dijkstra algorithm demonstration

$S = \{ \}$

$V = \{ \textcolor{red}{s}, 2, 3, 4, 5, 6, 7, 8 \}$

$D = \{ \textcolor{red}{0}, \infty, \infty, \infty, \infty, \infty, \infty, \infty \}$



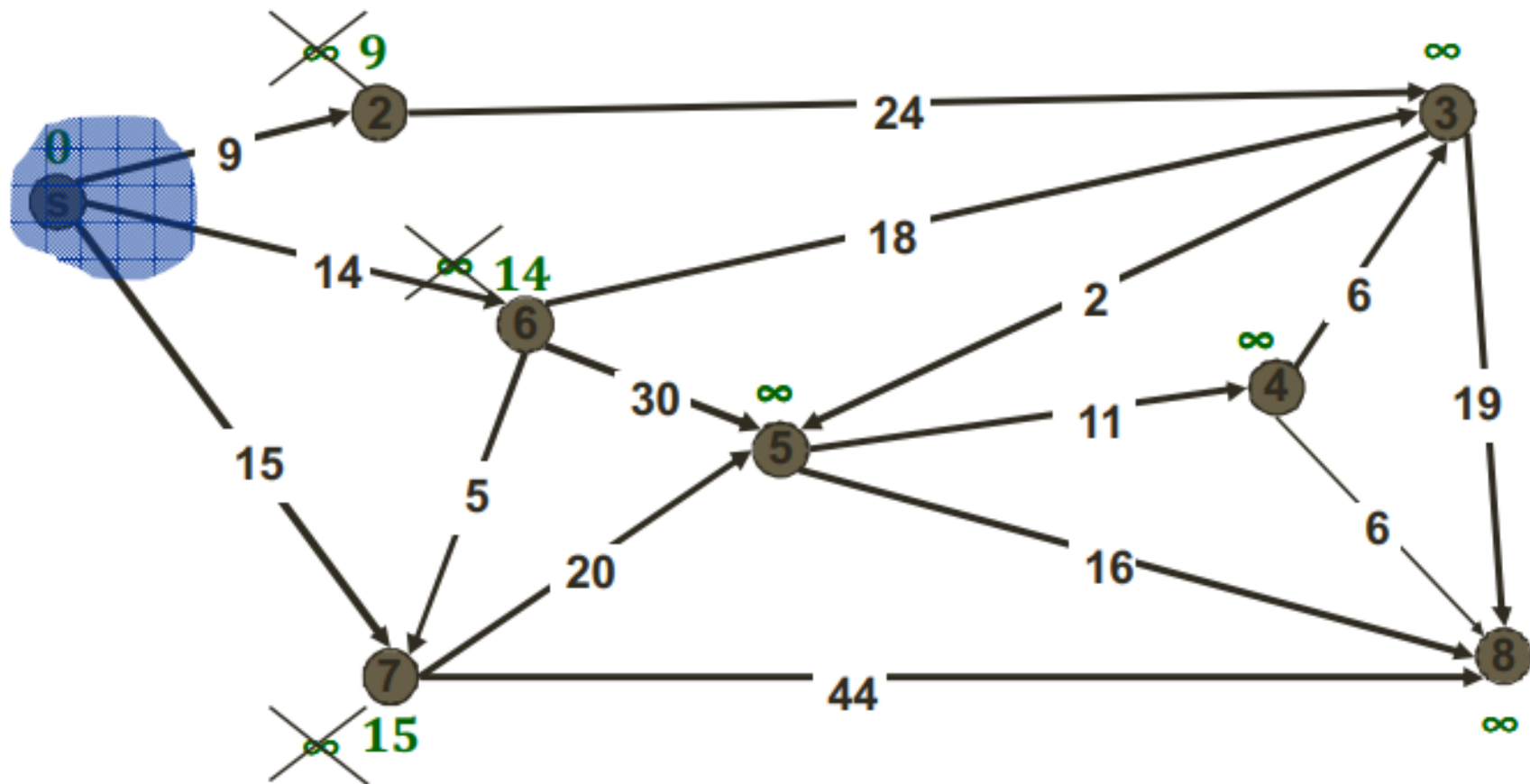
# Dijkstra algorithm[8]

- Dijkstra algorithm demonstration

$S = \{s\}$

$V = \{-, 2, 3, 4, 5, 6, 7, 8\}$

$D = \{0, 9, \infty, \infty, \infty, 14, 15, \infty\}$



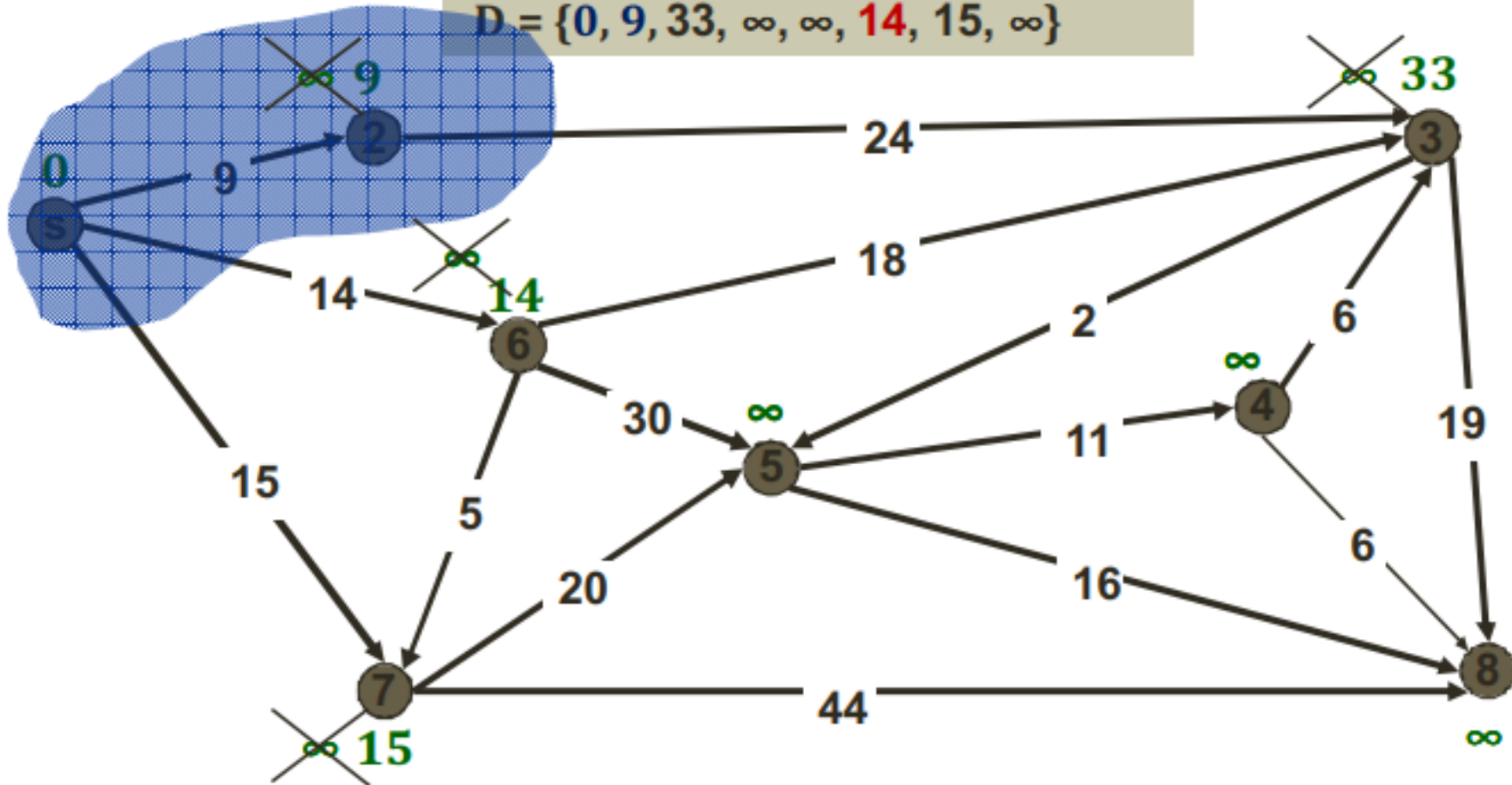
# Dijkstra algorithm[9]

- Dijkstra algorithm demonstration

$S = \{s, 2\}$

$V = \{-, -, 3, 4, 5, 6, 7, 8\}$

$D = \{0, 9, 33, \infty, \infty, 14, 15, \infty\}$



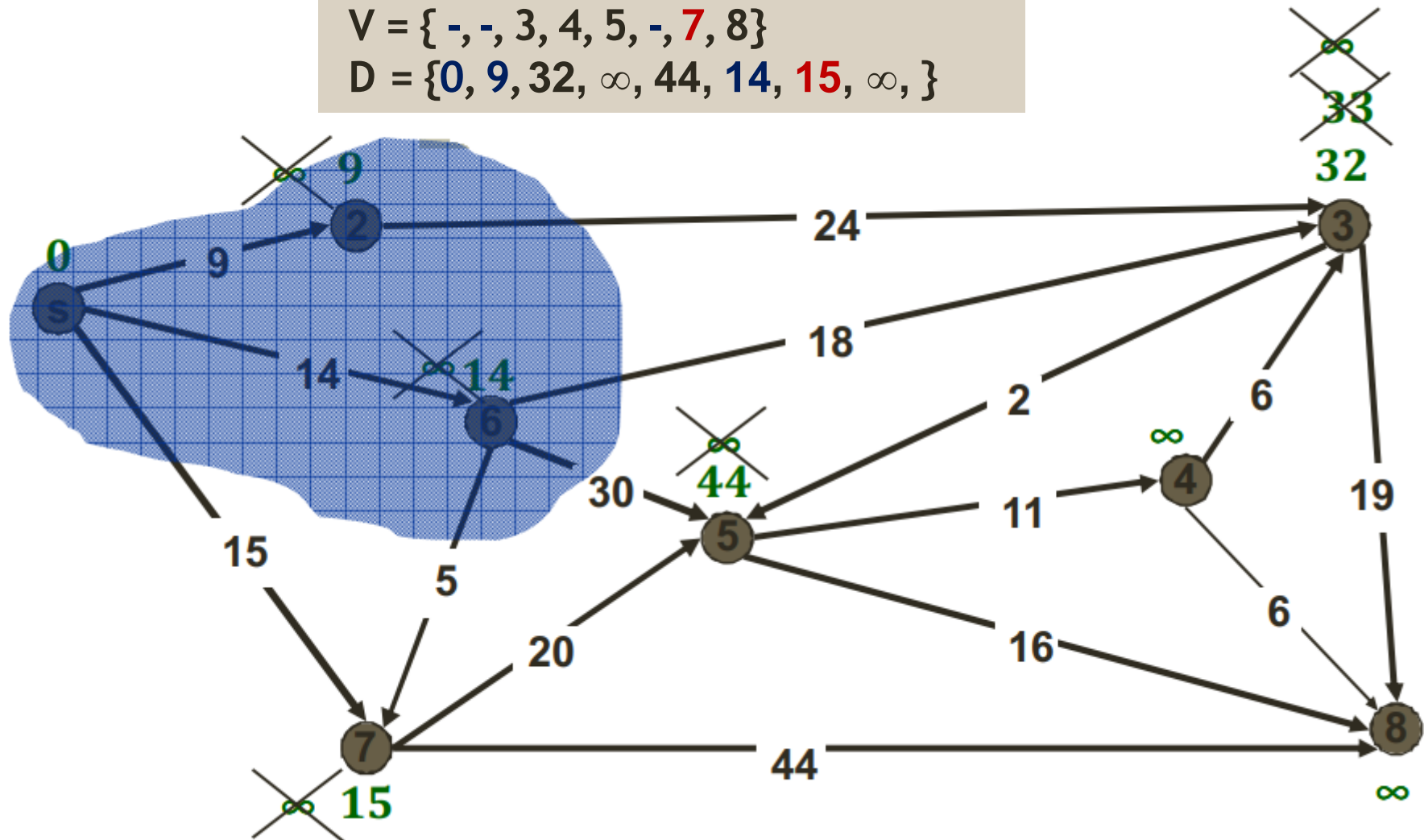
# Dijkstra algorithm[10]

- Dijkstra algorithm demonstration

$S = \{s, 2, 6\}$

$V = \{-, -, 3, 4, 5, -, 7, 8\}$

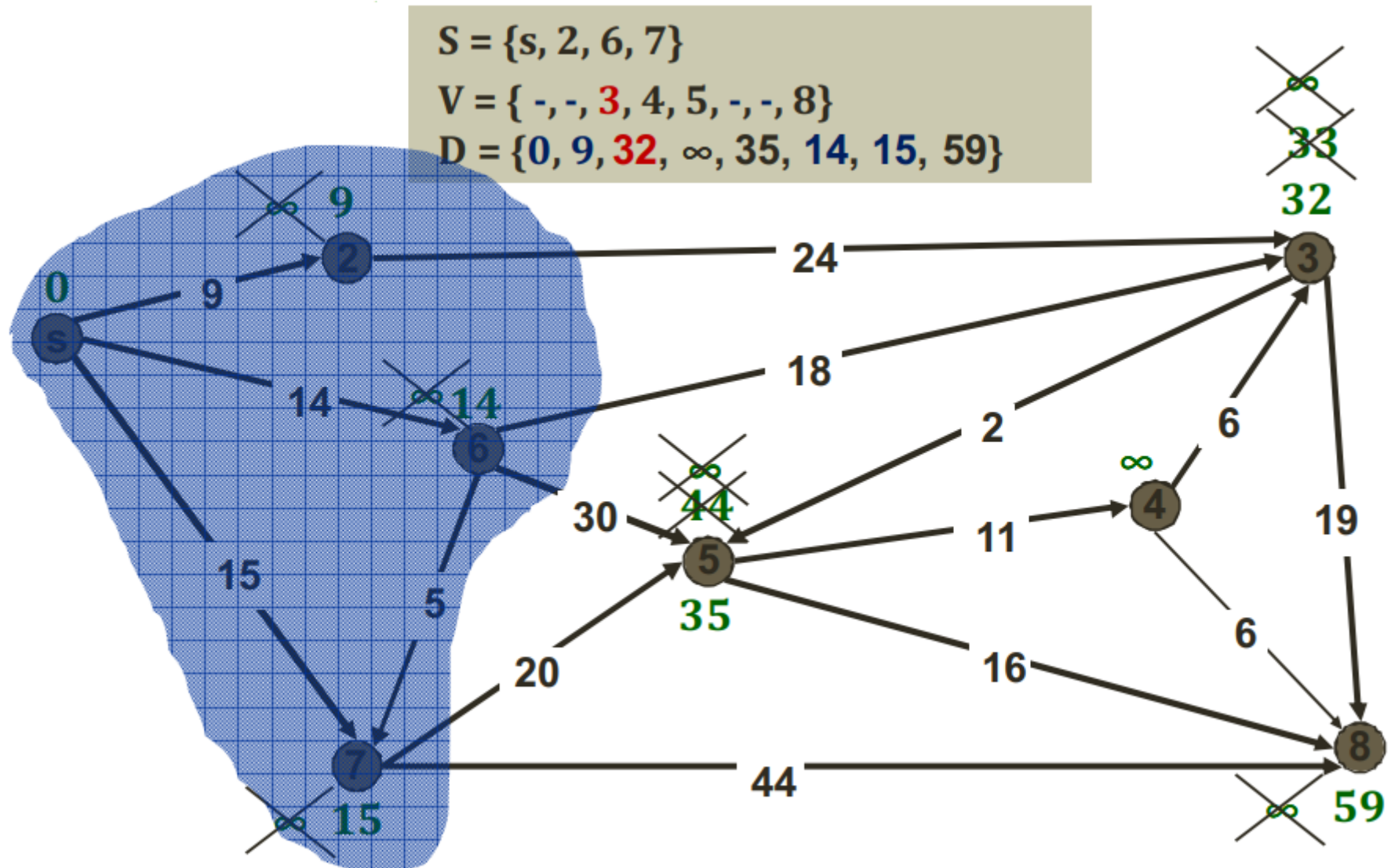
$D = \{0, 9, 32, \infty, 44, 14, 15, \infty, \infty\}$





# Dijkstra algorithm[11]

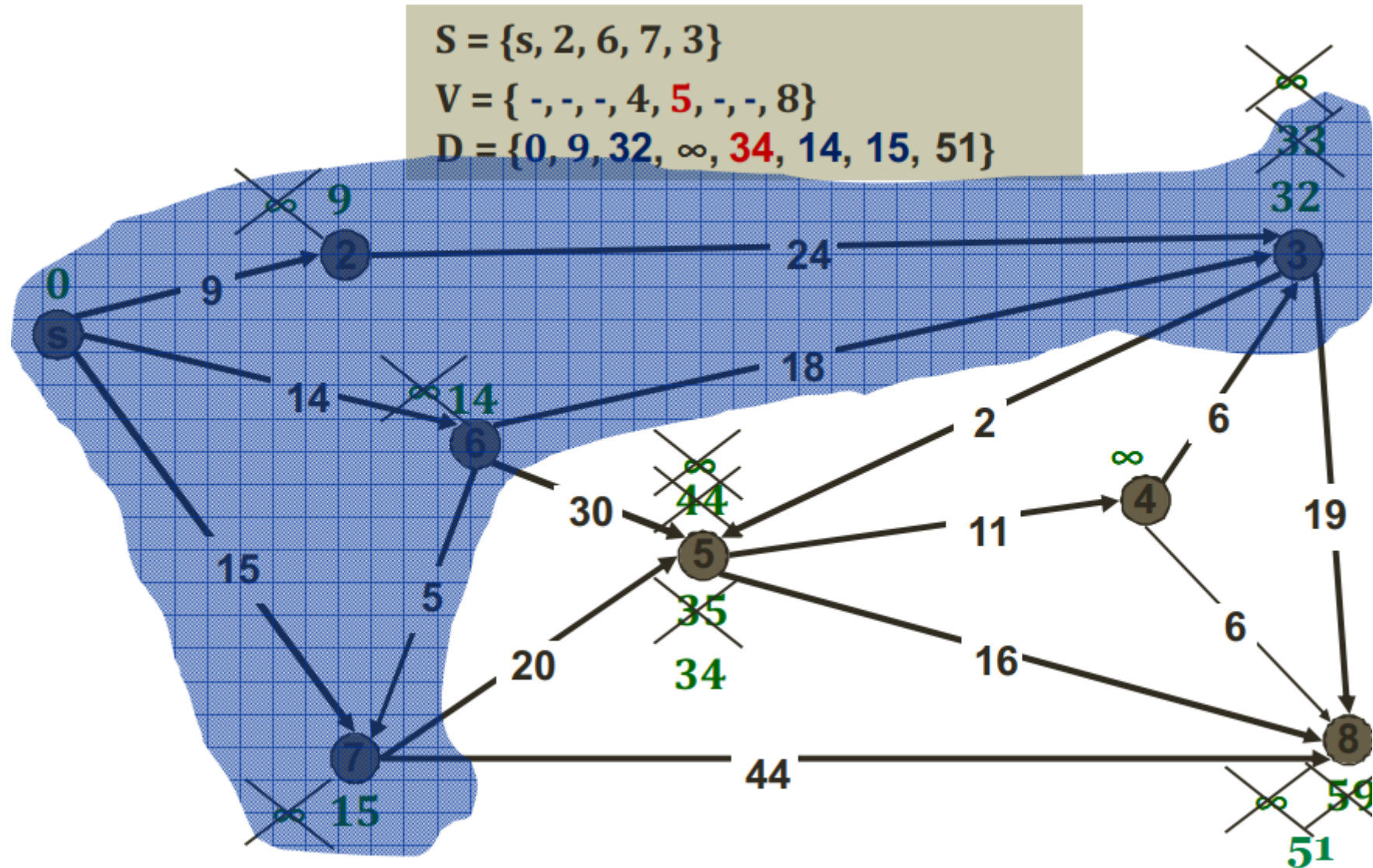
- Dijkstra algorithm demonstration





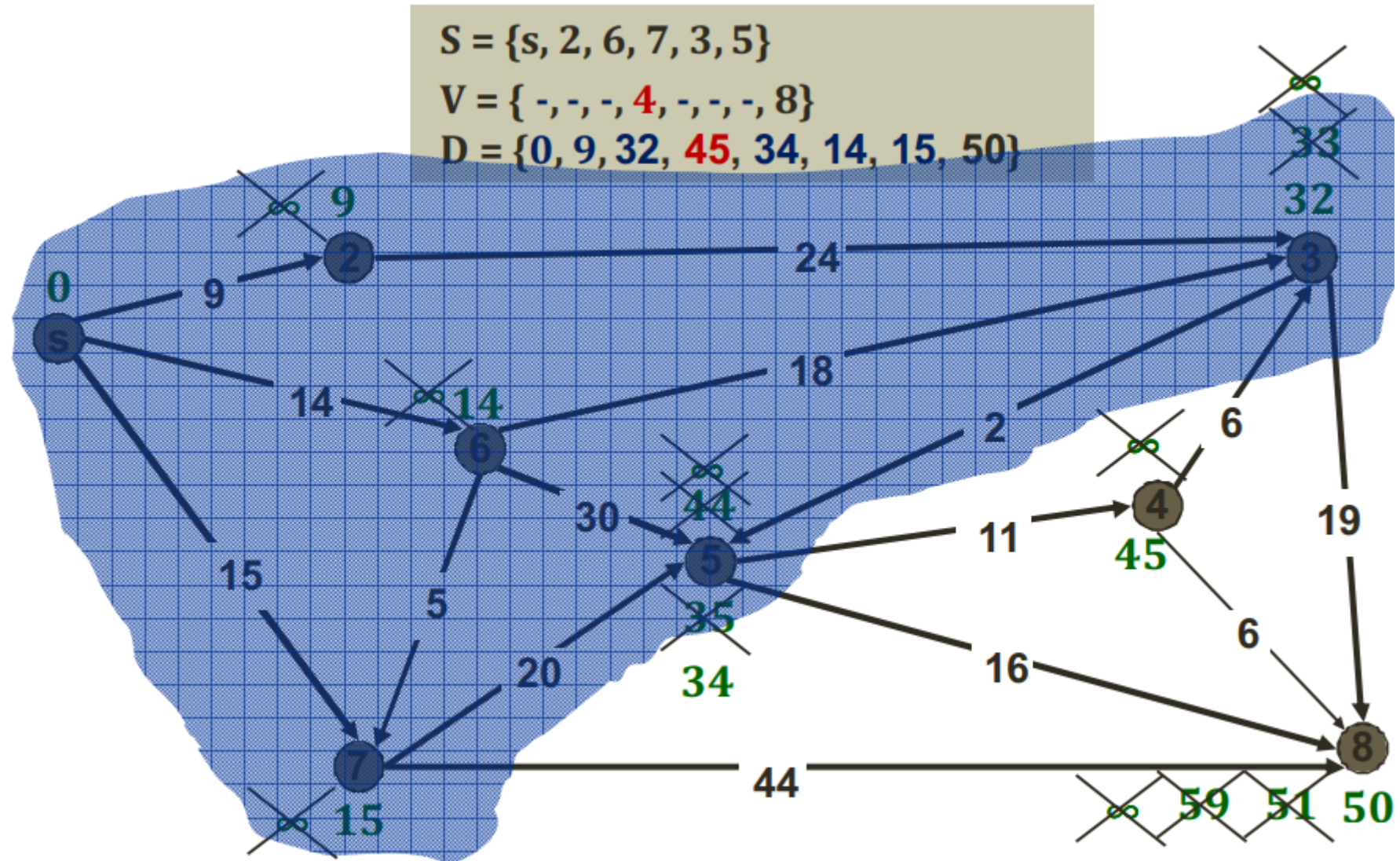
# Dijkstra algorithm[12]

- Dijkstra algorithm demonstration



# Dijkstra algorithm[13]

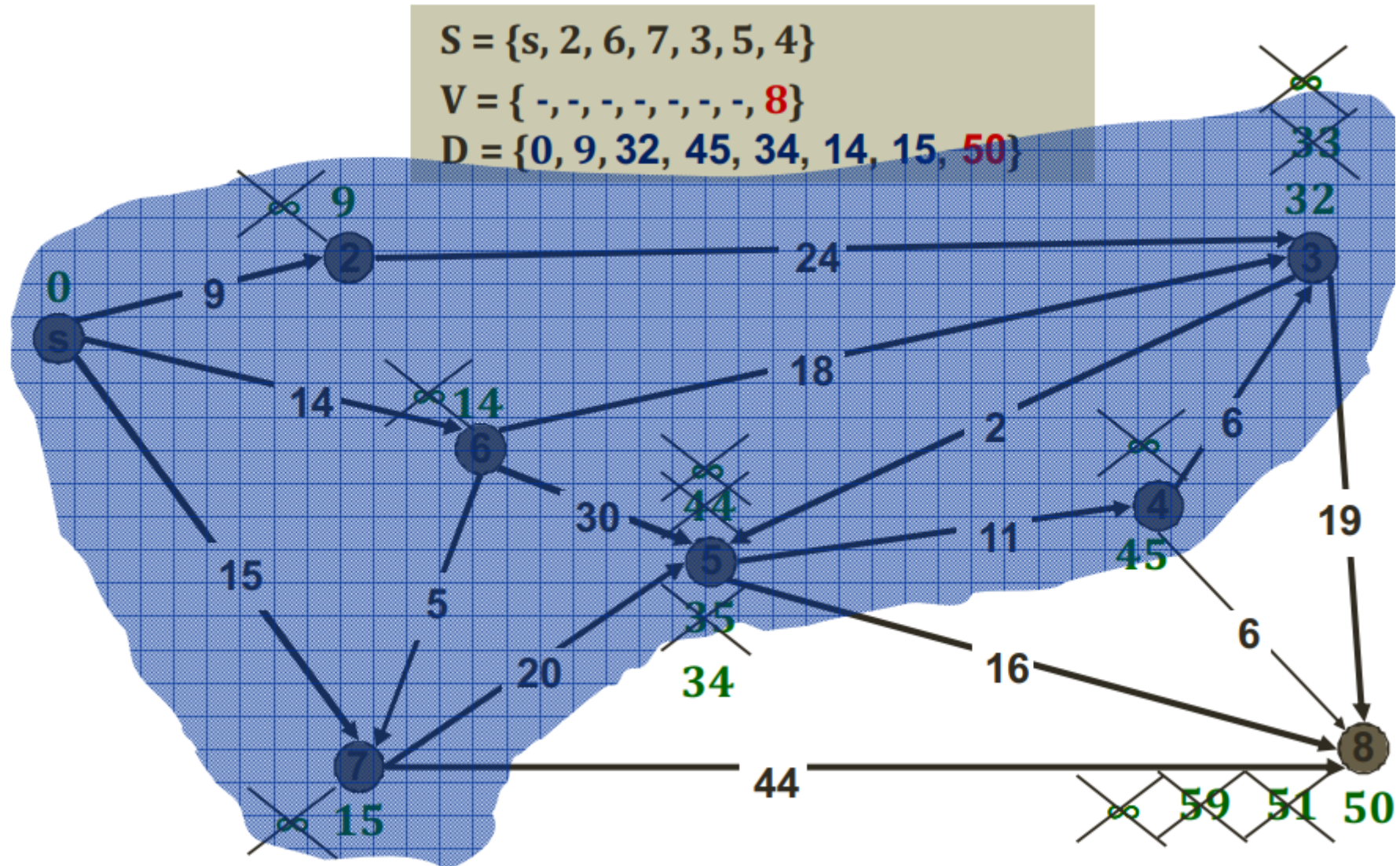
- Dijkstra algorithm demonstration





# Dijkstra algorithm[14]

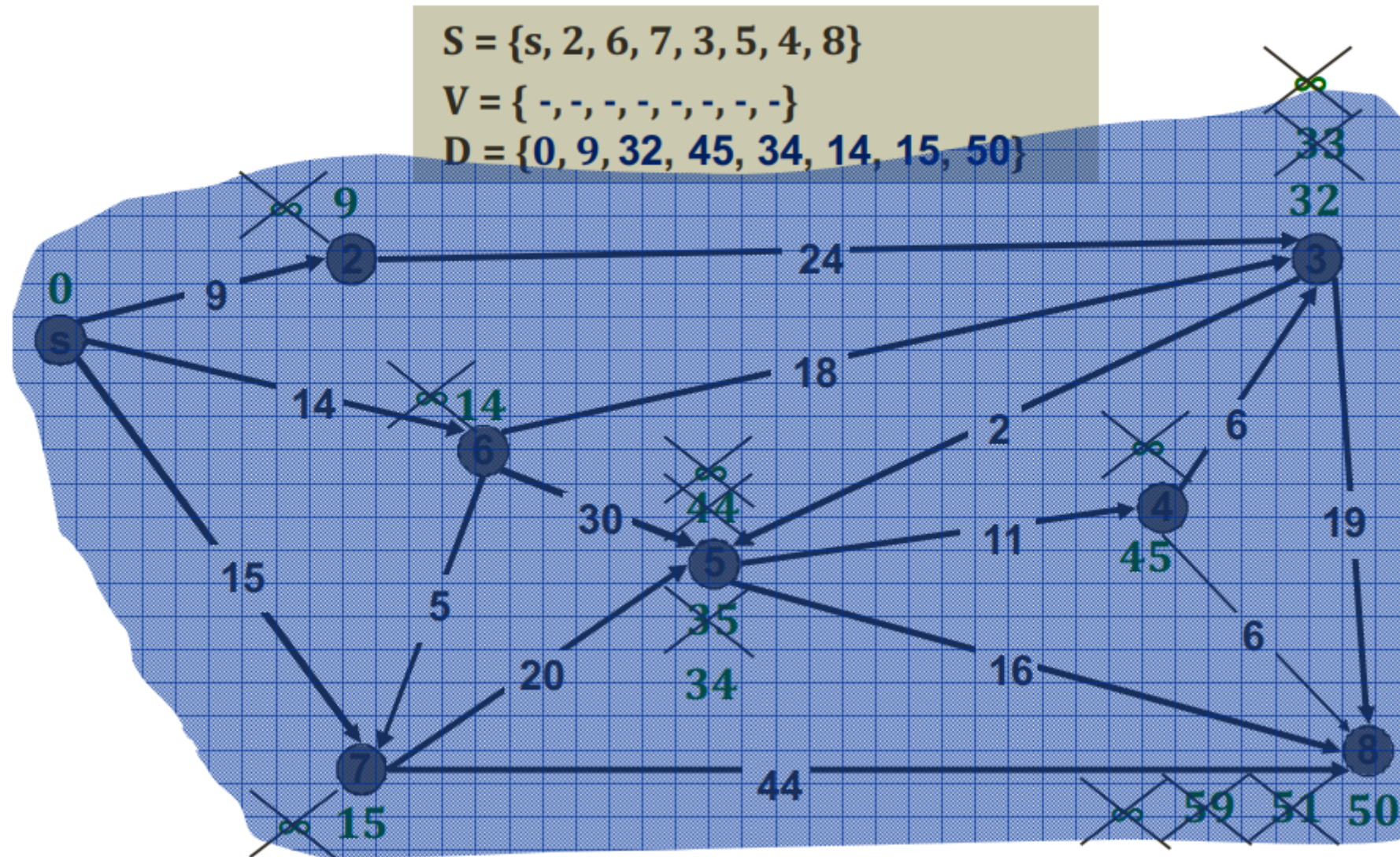
- Dijkstra algorithm demonstration





# Dijkstra algorithm[15]

- Dijkstra algorithm demonstration



# Dijkstra algorithm[16]

## Dijkstra algorithm demonstration:

Summarize steps with table:

D(s)	D(2)	D(3)	D(4)	D(5)	D(6)	D(7)	D(8)
0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
	9(s)	$\infty$	$\infty$	$\infty$	14(s)	15(s)	$\infty$
		33(2)	$\infty$	$\infty$	14(s)	15(s)	$\infty$
		32(6)	$\infty$	44(6)		15(s)	$\infty$
		32(6)	$\infty$	35(7)			59(7)
			$\infty$	34(3)			51(3)
			45(5)				50(5)
							50(5)

# Dijkstra algorithm[17]

- **Dijkstra algorithm analysis**
  - Textbook **page 379**, supplementary textbook **page 658**.
  - The time complexity of Dijkstra algorithm is:
    - **$O(|E| + |V|^2)$  by using array to store  $V$  and  $D[v]$ .**
    - **$O(|E|\log V + |V|\log V)$  by using a priority queue to store  $V$  and  $D[v]$ .**
  - Dijkstra algorithm doesn't work with negative-weighted edge.
  - Graph can be directed or un-directed.

# Bellman-Ford algorithm [1]

- A SSSP algorithm

- Find shortest paths from **vertex s** to all other vertices.
- Works even with **negative-weight edges**.
- Can detect the existence of **negative-weight cycle** reachable from s.

- **Assumption**

- The graph is connected.
- The edges **are directed**.

# Bellman-Ford algorithm [2]

- Definition of  $D[i][v]$ :
  - $D[i][v]$  is the total weight of the shortest path **that use  $i$  or fewer edges**, from  **$s$**  to  **$v$** .
  - Recall that the distance from vertex  $u$  to vertex  $v$  is the total weight of the shortest path from  $u$  to  $v$ .
  - Then  $D[i][v]$  is the **distance** from  **$s$**  to  **$v$**  **using  $i$  or fewer edges**.
    - $D[i][s]=0$
    - $D[i][v]=\infty$  if you can't get to  $v$  within  $i$  edges.
  - Bellman-Ford algorithm **computes the  $D[i][v]$  for any  $v$  in  $V$  and  $i=1 \dots |V|$** .



# Bellman-Ford algorithm [3]

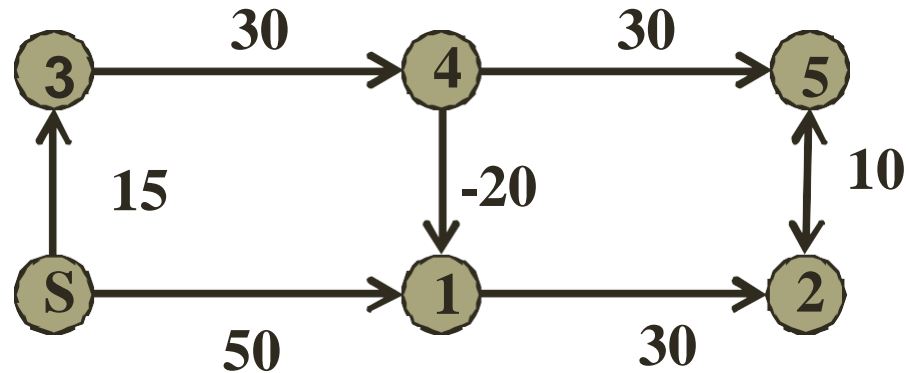
- **Algorithm idea**

- To compute  $D[i][v]$ , we **calculate  $i_{th}$  row from  $i-1_{th}$  row**.
- Consider a vertex  $u$  so that there is an edge from  $u$  to  $v$ .
- We know the shortest path from  $s$  to  $u$  using  $i-1$  or fewer edges,  $D[i-1][u]$ .
- Then:

$$D[i][v] = \min_{w[u][v] \neq \infty} \{D[i-1][u] + w[u][v]\}$$

# Bellman-Ford algorithm [4]

- Example:**



$$D[i][v] = \min_{w[u][v] \neq \infty} \{D[i-1][u] + w[u][v]\}$$

	S	1	2	3	4	5
0	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
1	0	50	$\infty$	15	$\infty$	$\infty$
2	0	50	80	15	45	$\infty$
3	0	25	80	15	45	75
4	0	25	55	15	45	75
5	0	25 (3,4)	55 (3,4,1)	15	45 (3)	65 (3,4,1,2)

# Bellman-Ford algorithm [5]

- **Bellman-Ford algorithm**

## Code

**Algorithm** BellmanFord( $V, E, s$ ): boolean

*Input:* A graph  $G=(V, E)$ ,  $w$  is the weighted matrix,  $s$  is the source vertex.

*Output:* The shortest paths from  $s$  to other vertices or false if  $G$  has a negative-weighted cycle.

**For**  $i \leftarrow 0$  to  $|V|$  **do**

**For each**  $v$  in  $V$  **do**

$D[i][v] \leftarrow \infty$

$path[v] \leftarrow -1$

$D[i][s] \leftarrow 0$

**For**  $i \leftarrow 1$  to  $|V|$  **do**

**For each**  $v$  in  $V$  **do**

**For each**  $e=(u, v)$  in  $E$  **do**

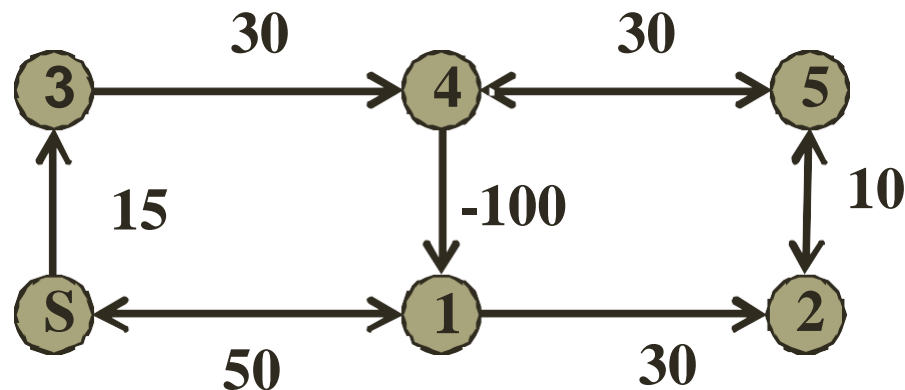
$D[i][v] \leftarrow \text{Min}\{D[i-1][u] + w[u][v]\}$

$path[v] \leftarrow u_{\min} \quad // \quad D[i][v] = D[i-1][u_{\min}] + w[u_{\min}][v]$

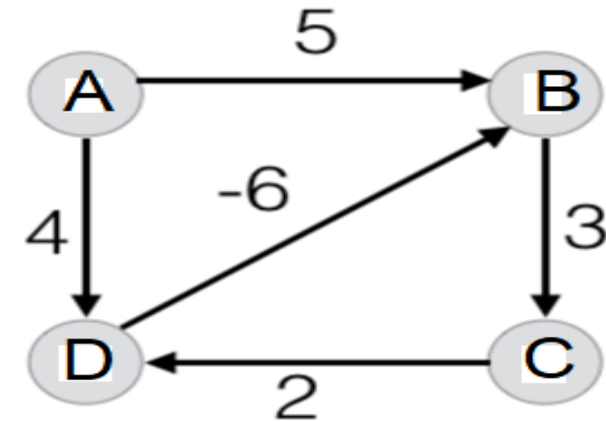
# Bellman-Ford algorithm [6]

- **Algorithm analysis**

- If the graph contains no negative-weight cycles reachable from the source vertex  $s$ , **after  $|V| - 1$  iterations** all distance estimates represent shortest paths.
- If the graph contains a negative-weight cycles reachable from  $s$ .



# Bellman-Ford algorithm[7]



	A	B	C	D	Note
0	0	$\infty$	$\infty$	$\infty$	
1	0	5,A	$\infty$	4,A	
2	0	-2,D	8,B	4,A	
3	0	-2,D (A→D→B =-2)	1,B (A→D→B→C =1)	4,A (A→D=4)	Normally, algorithm stops at step 3.
4	0	-2,D	1,B	3,C (A→D→B→C →D = 3)	Add step 4: The weight is reduced from (4,A) to (3,C) → Graph contains negative cycle