

CSCI203

# Algorithms and Data Structures



## Graphs (II)

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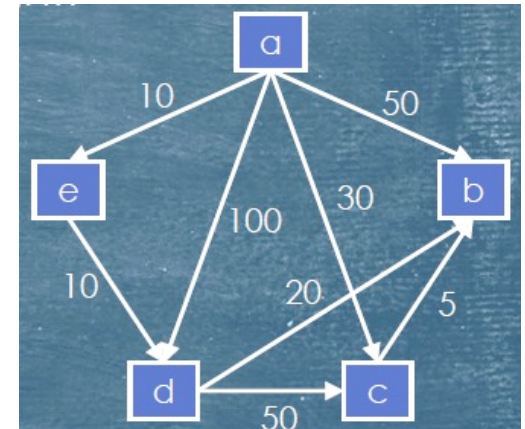
# Weighted Graphs



- ▶ Frequently, we find that travelling along an edge in a graph has some associated cost (or profit) associated with it:
  - The distance along the edge;
  - The cost of petrol;
  - The time of travel;
  - Etc.
- ▶ We call these Weighted Graphs.
- ▶ We call the edge values weights

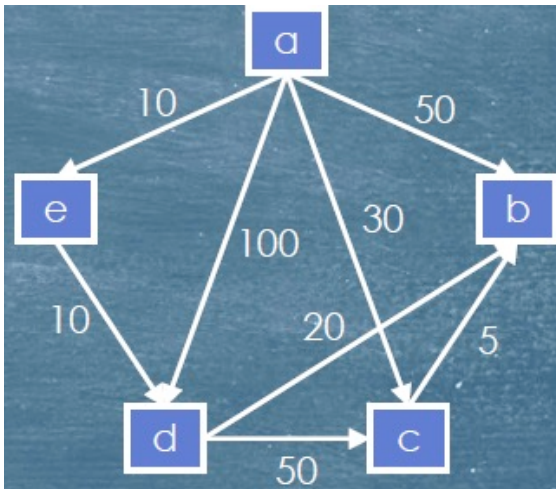
# Definition

- ▶ We extend our previous graph definition as follows:
  - A weighted graph,  $G$ , consists of the ordered sequence,  $(V, E, W)$  where  $V$  and  $E$  are the vertices and edges and  $W$  are the edge weights.
- ▶ Consider the weighted graph shown to the right:
  - $V = \{a, b, c, d, e\}$
  - $E = \{(a, b), (a, c), (a, d), (a, e), (c, b), (d, b), (d, c), (e, d)\}$
- ▶  $W$  is a function that maps edges to weights:
  - e.g  $W((a, b)) = 50$ .



# Representation

- ▶ We can extend our adjacency matrix representation of a graph by replacing the zero-one existence value with the edge weight.



—	50	30	100	10
—	—	—	—	—
—	5	—	—	—
—	20	50	—	—
—	—	—	10	—

# Representation



- ▶ In this example, “—” indicates that no edge exists.
- ▶ The actual value will depend on the nature of the weights:
  - E.g. if all weights are non-zero use 0.
  - We often use  $\infty$  to represent missing edges.
- ▶ We can also use the adjacency list representation:
  - We just need to pair each edge with its corresponding weight.

# Shortest Path



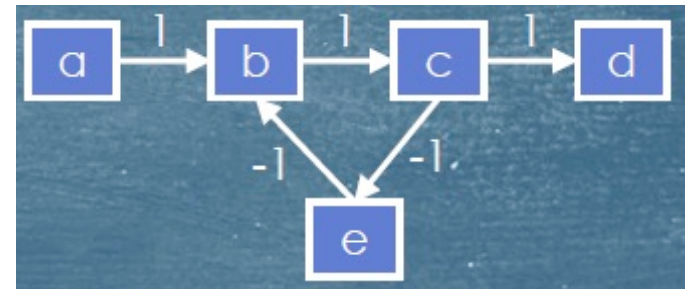
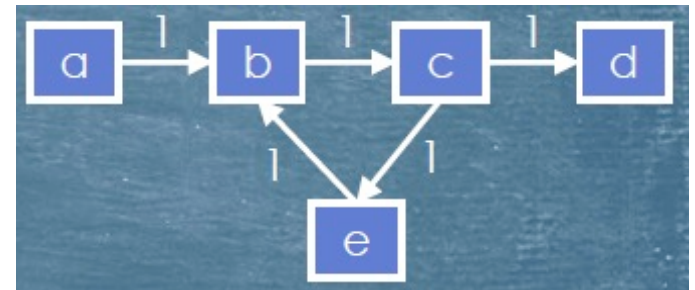
- ▶ A common problem associated with weighted graphs is finding the shortest path between vertices.
- ▶ There are several versions of this problem:
  - Single Source—All destinations;
  - Single Source—Single Destination;
  - All Sources—All Destinations;
  - All Sources—Single Destination.
- ▶ Each has applications in the real world.
- ▶ We will start by looking at the first of these types.

# Single Source—All Destinations

- ▶ This problem is stated as follows:
- ▶ Starting at some source vertex,  $s$ , find the shortest path from  $s$  to each other **reachable** vertex in the graph.
- ▶ As we shall see later, the solution to this problem can be used as a basis for solving all of the other shortest path formulations.
- ▶ We will examine two algorithms for solving this problem:
  - Dijkstra;
  - Bellman Ford.
- ▶ Each has advantages in certain cases.

# Negative Weights

- ▶ There is no a priori reason why the edge weights in a graph must be positive, but negative edge weights can cause problems.
- ▶ Consider the following graph:
  - Clearly the length of the shortest path from a to d is 3.
- ▶ But what if we change the weights?
  - Now what is the shortest path?
- ▶ The problem is that we now have a negative cost cycle.





# What is a Path

- ▶ While it is obvious what a path is, we should define it formally.
- ▶ A path  $p$  from vertex  $v_1$  to vertex  $v_k$  is an ordered sequence  $(v_1, v_2, \dots, v_{k-1}, v_k)$  where each edge  $(v_i, v_{i+1}) \in E$ .
- ▶ The weight of path  $p$ ,  $W(p)$  is the sum of the edge weights:
  - $W(p) = \sum_{i=1}^{k-1} W(v_i, v_{i+1})$

# What doesn't work?



- ▶ We might be tempted to try using a technique we already know for traversing a graph, breadth first search, in our search for shortest paths.
- ▶ Unfortunately, this does not always work.
- ▶ The two definitions of shortest path:
  - Fewest edges;
  - Smallest weight;
- ▶ May not always coincide.

# Dijkstra's Algorithm



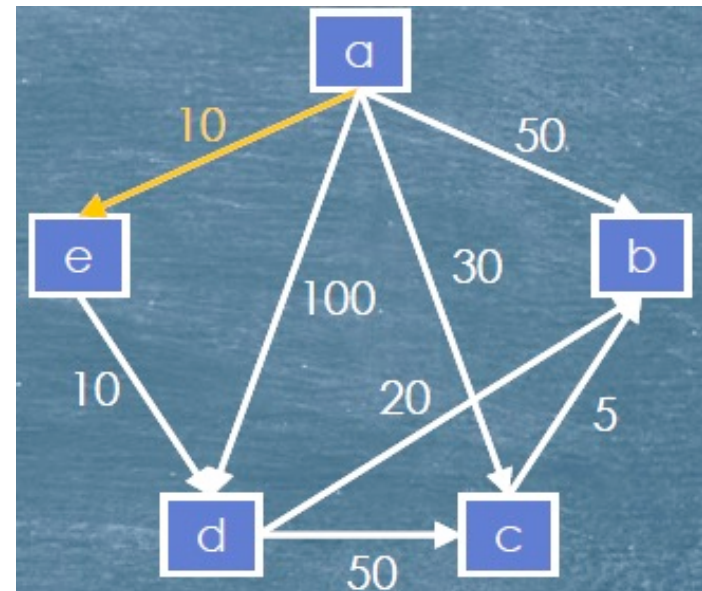
- ▶ This algorithm works by dividing the vertices into two sets,  $S$  and  $C$ .
- ▶ At each iteration,  $S$  contains the set of nodes that have already been chosen.
  - This is the selected set.
- ▶ At each iteration,  $C$  contains the set of nodes that have not yet been chosen:
  - This is the candidate set.
- ▶ At each step we move the node which is cheapest to reach from  $C$  to  $S$ .

# Dijkstra's Algorithm

- ▶ We also need a function  $D$  such that  $D(c_i)$  is the shortest distance we have so far found from vertex  $s$  to vertex  $c_i$  in the candidate set  $C$ .
- ▶ Initially:
  - The selected set,  $S$ , just contains the start vertex;
  - The candidate set,  $C$ , contains all the other vertices;
  - The distance function,  $D()$  has value 0 for vertex  $s$  and is infinite for all other vertices.
- ▶ We start by re-evaluating  $D$  for each vertex directly reachable from vertex  $s$ .

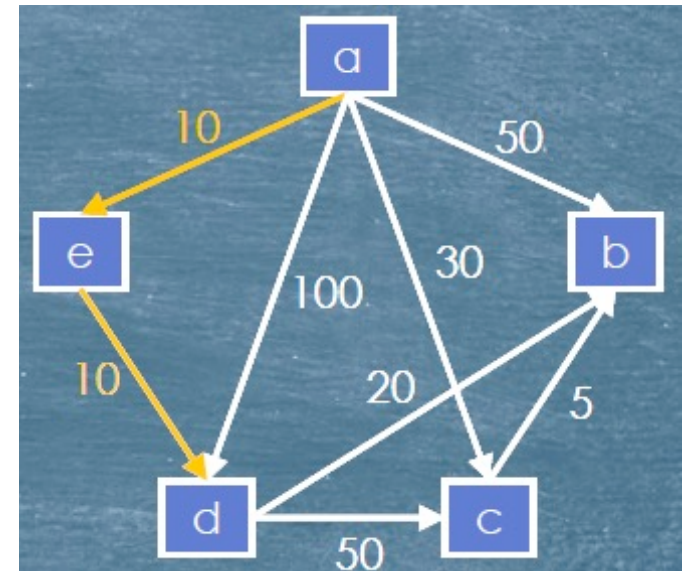
# Dijkstra's Algorithm: an Example

- ▶ Step 0:
- ▶  $S = \{a\}$
- ▶  $C = \{b, c, d, e\}$
- ▶  $D = \{50, 30, 100, 10\}$
- ▶ We now select the minimum value of  $D$ ,  $D(e) = 10$ .



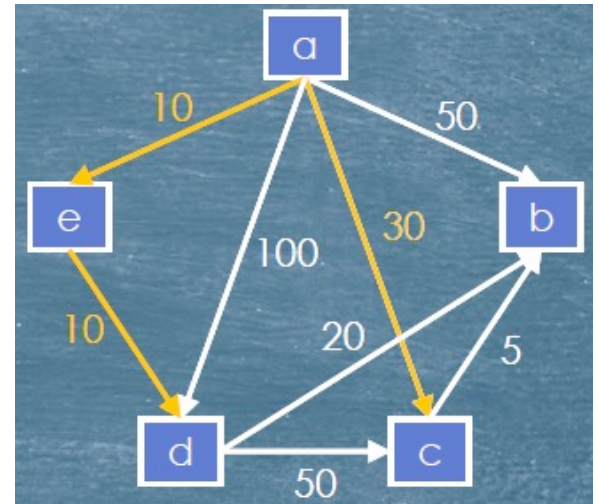
# Dijkstra's Algorithm: an Example

- ▶ Step 1: move vertex  $e$  from  $C$  to  $S$ .
- ▶  $S = \{a, e\}$
- ▶  $C = \{b, c, d\}$
- ▶ We now update  $D$  by looking at vertices we can reach from vertex  $e$ .
- ▶  $D = (50, 30, 100 \rightarrow 20)$
- ▶ We now select the minimum value of  $D$ ,  $D(d) = 20$ .



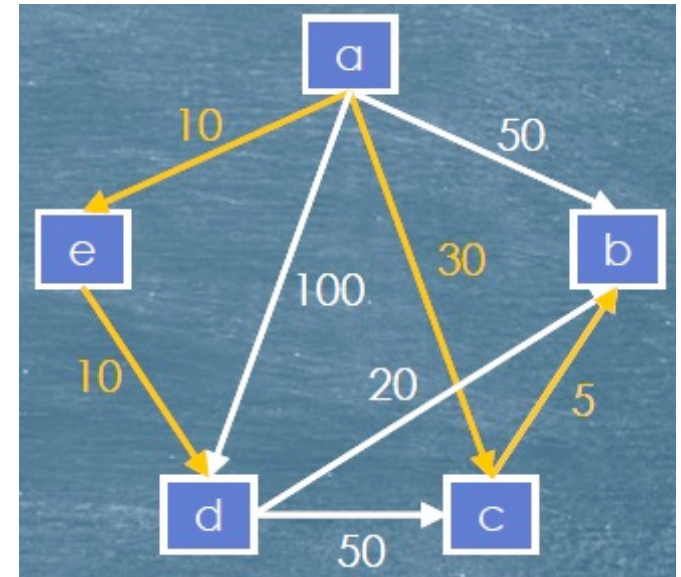
# Dijkstra's Algorithm: an Example

- ▶ Step 2: move vertex  $d$  from  $C$  to  $S$ .
- ▶  $S = \{a, e, d\}$
- ▶  $C = \{b, c\}$
- ▶ We now update  $D$  by looking at vertices we can reach from vertex  $d$ .
- ▶  $D = (50 \rightarrow 40, 30)$
- ▶ We now select the minimum value of  $D$ ,  $D(c) = 30$ .



# Dijkstra's Algorithm: an Example

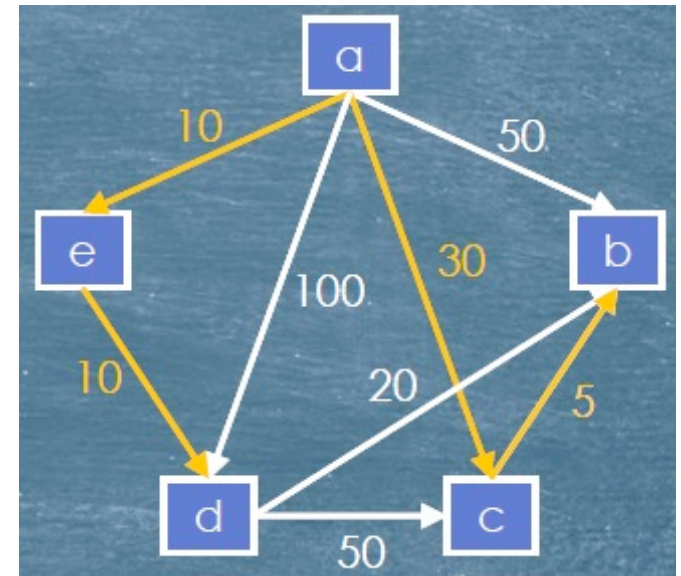
- ▶ Step 3: move vertex  $c$  from  $C$  to  $S$ .
- ▶  $S = \{a, e, d, c\}$
- ▶  $C = \{b\}$
- ▶ We now update  $D$  by looking at vertices we can reach from vertex  $c$ .
- ▶  $D = (40 \rightarrow 35)$
- ▶ We now select the minimum value of  $D$ ,  $D(b) = 35$ .





# Dijkstra's Algorithm: an Example

- ▶ Step 4: move vertex  $b$  from  $C$  to  $S$ .
- ▶  $S = \{a, e, d, c, b\}$
- ▶  $C = \{\}$
- ▶ We now have no remaining candidate vertices; we have finished.
- ▶  $W(a) = 0, W(e) = 10, W(d) = 20, W(c) = 30, W(b) = 35$ .



# Dijkstra's Algorithm: Pseudocode

```
Procedure Dijkstra(G: array[1..n, 1..n]): array [2..n]
  D: array[2..n]
  C: set = {2, 3, ..., n}
  for i = 2 to n do
    D[i] = G[1, i]
  od
  repeat
    v = the index of the minimum D[v] not yet selected
    remove v from C // and implicitly add v to S
    for each u ∈ C do
      D[u] = min(D[u], D[v] + G[v, u])
    rof
  until C contains no reachable nodes
  return D
end Dijkstra
```

# Recording Paths

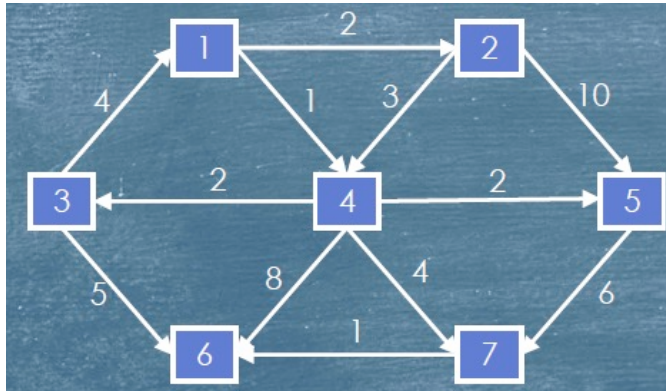


- ▶ Like the basic DFS, Dijkstra's algorithm does not record the shortest path to each vertex, just its total weight.
- ▶ Also, like DFS, we can use a parent record,  $p$ , to keep track of how we reach each vertex.
- ▶ This entails a couple of minor changes to the algorithm...

# Dijkstra's Algorithm: Path Recording

```
Procedure Dijkstra(G: array[1..n, 1..n]): array [2..n]
  D: array[2..n], P: array[2..n]
  C: set = {2, 3, ..., n}, S: set = {}
  for i = 2 to n do
    D[i] = G[1, i]
    P[i]=1
  od
  repeat
    v = the index of the minimum D[v] not yet selected
    move v from C to S
    for each u ∈ C do
      D[u] = min(D[u], D[v] + G[v, u])
      p[u] = v
    rof
  until C contains no reachable nodes
  return D
end Dijkstra
```

# A Larger Example



To

From

0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$
$\infty$	$\infty$	5	8	$\infty$	0	1
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0

► At start

■  $G =$

The diagram illustrates the Floyd-Warshall algorithm for finding shortest paths in a graph with 7 nodes. It shows a 7x7 distance matrix  $D$ , a path matrix  $P$ , and a sequence of updates to  $D$ .

**Initial Distance Matrix  $D$ :**

0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$
$\infty$	$\infty$	5	8	$\infty$	0	1
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0

**Initial Path Matrix  $P$ :**

1	1	1	1	1	1	1
2	2	2	2	2	2	2
3	3	3	3	3	3	3
4	4	4	4	4	4	4
5	5	5	5	5	5	5
6	6	6	6	6	6	6
7	7	7	7	7	7	7

**Sequence of Updates:**

- Update 1:  $D_{1,4} = 2$  (Path 1 → 3 → 4)
- Update 2:  $D_{1,5} = 3$  (Path 1 → 3 → 4 → 5)
- Update 3:  $D_{1,6} = 4$  (Path 1 → 3 → 4 → 5 → 6)
- Update 4:  $D_{1,7} = 0$  (Path 1 → 3 → 4 → 5 → 6 → 7)
- Update 5:  $D_{2,4} = 2$  (Path 2 → 1 → 3 → 4)
- Update 6:  $D_{2,5} = 3$  (Path 2 → 1 → 3 → 4 → 5)
- Update 7:  $D_{2,6} = 4$  (Path 2 → 1 → 3 → 4 → 5 → 6)
- Update 8:  $D_{2,7} = 0$  (Path 2 → 1 → 3 → 4 → 5 → 6 → 7)
- Update 9:  $D_{3,5} = 1$  (Path 3 → 4 → 5)
- Update 10:  $D_{3,6} = 2$  (Path 3 → 4 → 5 → 6)
- Update 11:  $D_{3,7} = 0$  (Path 3 → 4 → 5 → 6 → 7)
- Update 12:  $D_{4,5} = 0$  (Path 4 → 5)
- Update 13:  $D_{4,6} = 1$  (Path 4 → 5 → 6)
- Update 14:  $D_{4,7} = 0$  (Path 4 → 5 → 6 → 7)
- Update 15:  $D_{5,6} = 0$  (Path 5 → 6)
- Update 16:  $D_{5,7} = 0$  (Path 5 → 6 → 7)
- Update 17:  $D_{6,7} = 0$  (Path 6 → 7)

**Final Distance Matrix  $D$ :**

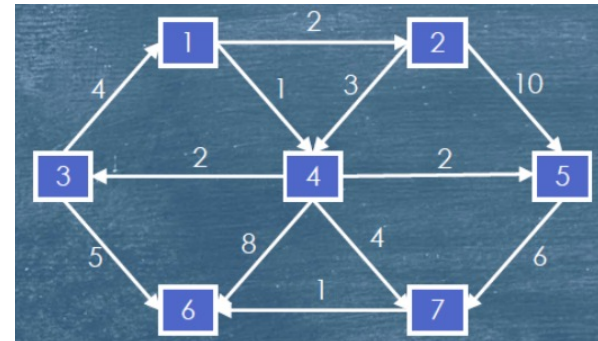
0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$
$\infty$	$\infty$	5	8	$\infty$	0	1
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0

$$C = \{2, 3, 4, 5, 6, 7\}$$

$$S = \{1\}$$

$$v = 4$$

$$D(v) = 1$$



## ► After step 1

■  $G =$

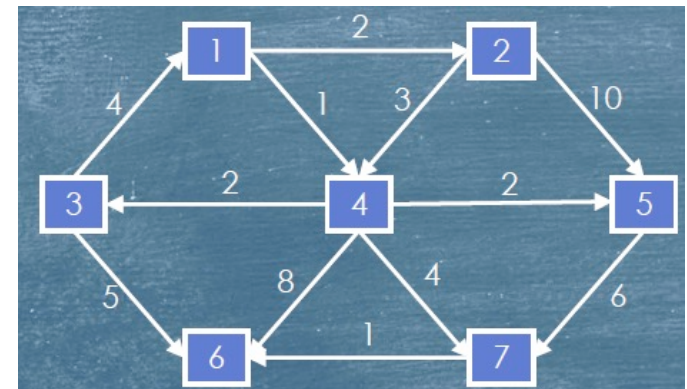
0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$			
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$			
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$	1	4	2
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$	1	4	3
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$	4	4	3
$\infty$	$\infty$	5	8	$\infty$	0	1	4	4	9
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0	4	4	5

$$C = \{2, 3, 5, 6, 7\}$$

$$S = \{1, 4\}$$

$$v = 2$$

$$D(v) = 2$$



## ► After step 2

■  $G =$

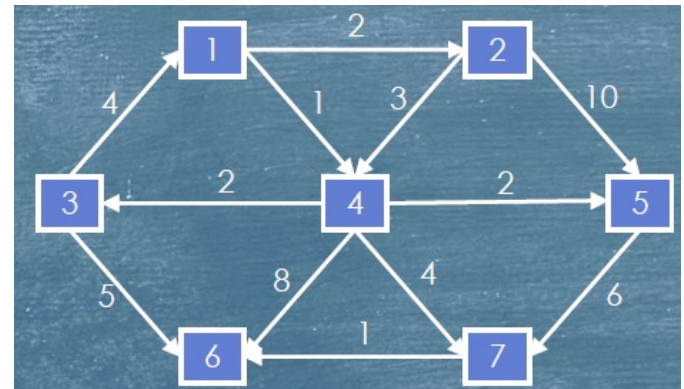
0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$		
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$		
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$		
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$		
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$	4	→
$\infty$	$\infty$	5	5	$\infty$	0	1	3	→
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0	4	→
							P =	D =
							1	2
							4	3
							1	1
							4	3
							3	8
							4	5

$C = \{3, 5, 6, 7\}$

$S = \{1, 4, 2\}$

$v = 3$

$D(v) = 3$





## ► After step 3

■  $G =$

0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$		
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$		
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$		
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$		
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$		
$\infty$	$\infty$	5	8	$\infty$	0	1		
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0		

P =

1

4

1

4

3

4

D =

2

3

1

3

8

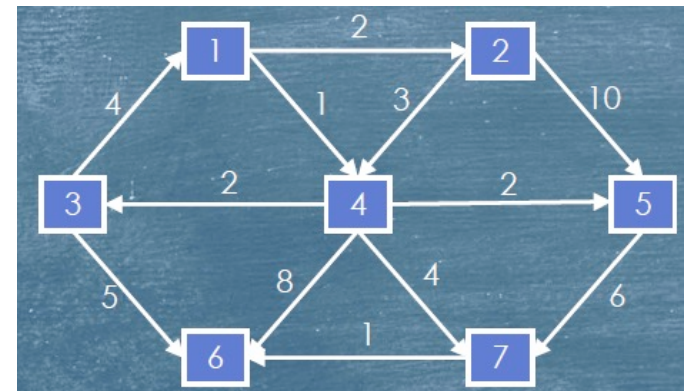
5

$C = \{5, 6, 7\}$

$S = \{1, 4, 2, 3\}$

$v = 5$

$D(v) = 3$



## ► After step 4

■  $G =$

0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$
$\infty$	$\infty$	5	8	$\infty$	0	1
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0

P =

1

4

1

4

7

4

D =

2

3

1

3

6

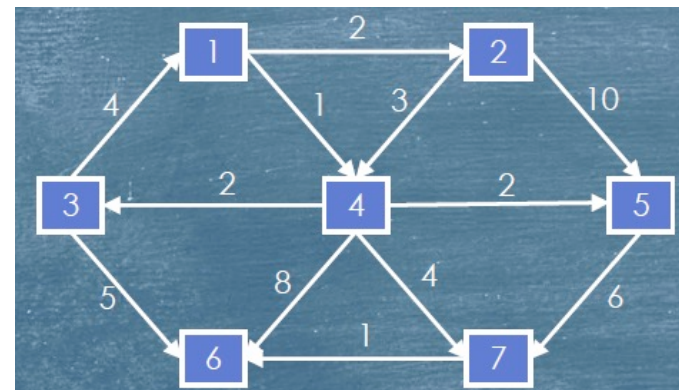
5

$$C = \{6, 7\}$$

$$S = \{1, 4, 2, 3, 5\}$$

$$v = 7$$

$$D(v) = 5$$



## ► After step 5

■  $G =$

0	$\infty$	4	$\infty$	$\infty$	$\infty$	$\infty$
2	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
$\infty$	$\infty$	0	2	$\infty$	$\infty$	$\infty$
1	3	$\infty$	0	$\infty$	$\infty$	$\infty$
$\infty$	10	$\infty$	2	0	$\infty$	$\infty$
$\infty$	$\infty$	5	8	$\infty$	0	1
$\infty$	$\infty$	$\infty$	4	6	$\infty$	0

P =

1

4

1

4

7

4

D =

2

3

1

3

6

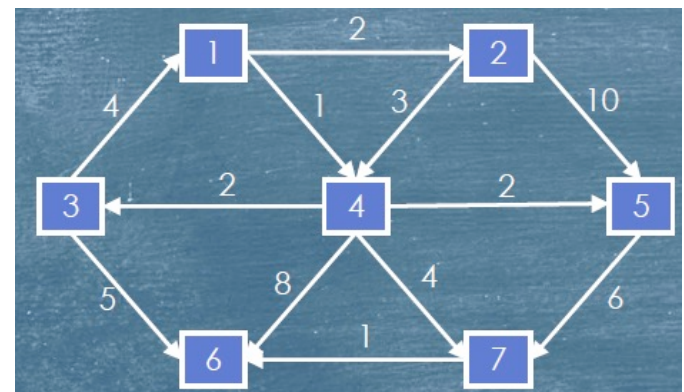
5

$$C = \{6\}$$

$$S = \{1, 4, 2, 3, 5, 7\}$$

$$v = 6$$

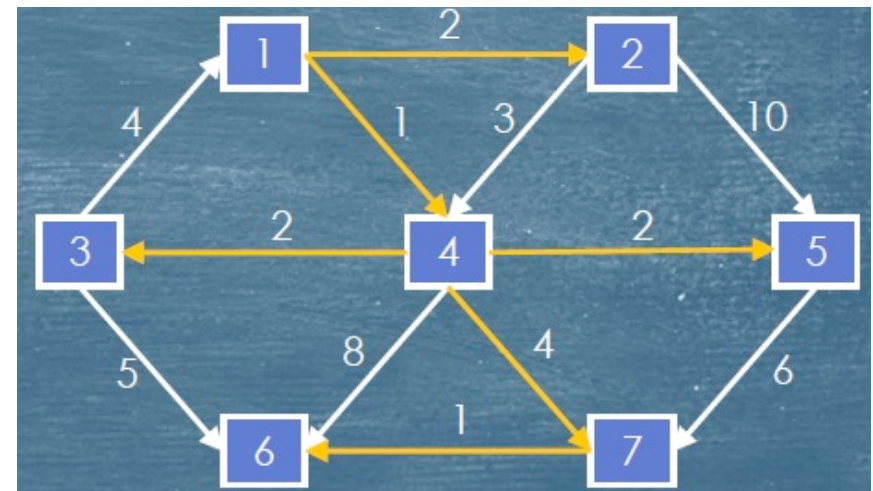
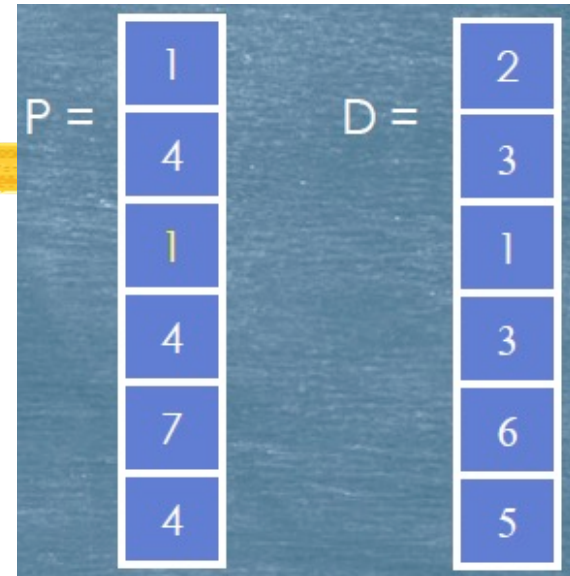
$$D(v) = 6$$



► After step 6

► Paths from vertex 1:

- To vertex 2
  - Path = (1, 2);  $W = 2$
- To vertex 3
  - Path = (1, 4, 3);  $W = 3$
- To vertex 4
  - Path = (1, 4);  $W = 1$
- To vertex 5
  - Path = (1, 4, 5);  $W = 3$
- To vertex 6
  - Path = (1, 4, 7, 6);  $W = 6$
- To vertex 7
  - Path = (1, 4, 7);  $W = 5$



# Analysis

- ▶ The complexity of Dijkstra's Algorithm is  $\Theta(V * \log V + E)$ 
  - How?
- ▶ Usually,  $E > V$ , in fact, in the worst case...
  - ... $E \in \Theta(V^2)$ —for a complete graph.
- ▶ There is one disadvantage:
  - The algorithm only works if all edge weights are non-negative.
- ▶ If we have negative edge weights, and especially negative edge cycles, we need a different algorithm.

# The Bellman-Ford Algorithm

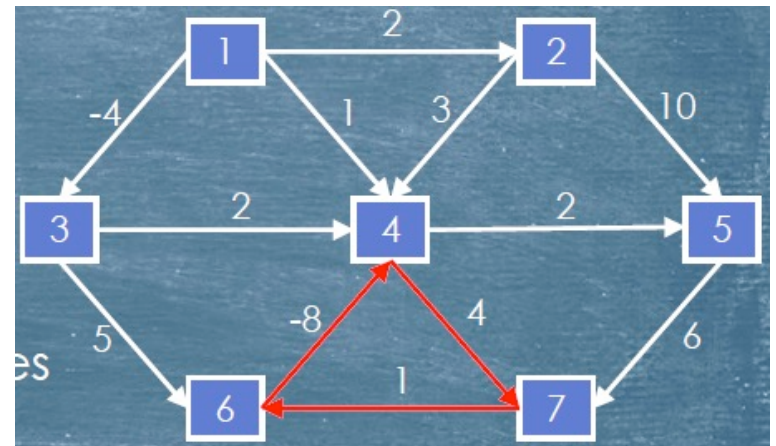


- ▶ Independently invented by both Bellman and Ford.
- ▶ Works with graphs that have negative edge weights.
- ▶ Identifies negative cycles and vertices with a negative cycle on their path.
- ▶ Finds correct path and path length for all other vertices.
- ▶ Let us look at an example graph...



# A Graph with a negative cycle

- ▶ Consider the graph shown:
- ▶ Because the edges between vertices 4, 7 and 6 form a cycle whose total weight is  $-3$ , we can reduce the cost of any vertex with a path through any of these vertices as much as we like.
- ▶ Note that some vertices, namely 2 and 3, still have well defined minimum costs of 2 and  $-4$  respectively.
- ▶ All other vertices have undefined minimum cost



# The Algorithm

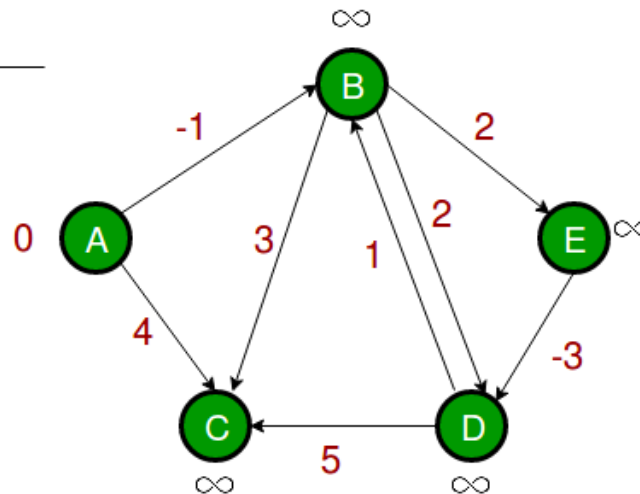
- ▶ **Input:** Graph  $G = (V, E, W)$  and a source vertex  $s$
- ▶ **Output:** Shortest distance to all vertices from  $s$ . If there is a negative weight cycle, then shortest distances are not calculated, negative weight cycle is reported.
  1. Initialization -  $D[v] = \infty, v \in V$ , except  $D[s] = 0$
  2. Calculates shortest distances. Do following  $|V|-1$  times
    - Do following for each edge  $u-v$  in  $E$ 
      - If  $D[v] > D[u] + W((u, v))$ , then update  $D[v]$   
 $D[v] = D[u] + W((u, v))$
  3. Reports if there is a negative weight cycle in graph. Do following for each edge  $u-v$  in  $E$ 
    - If  $D[v] > D[u] + W((u, v))$ , then "Graph contains negative weight cycle"



# An Example

## ► Initialization

A	B	C	D	E
0	$\infty$	$\infty$	$\infty$	$\infty$

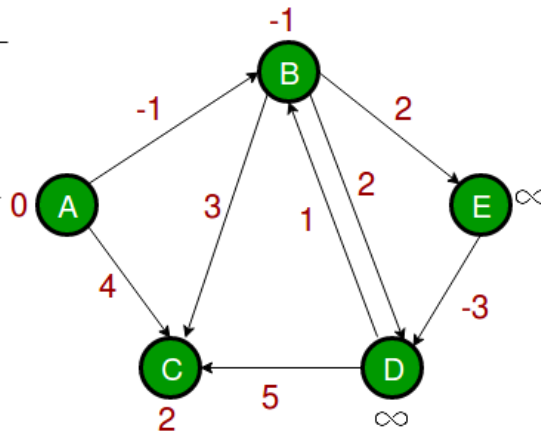


- Let all edges are processed in the following order:
  - (B, E), (D, B), (B, D), (A, B), (A, C), (D, C), (B, C), (E, D).

# An Example

## ► Step 1 - Iteration #1

A	B	C	D	E
0	$\infty$	$\infty$	$\infty$	$\infty$
0	-1	$\infty$	$\infty$	$\infty$
0	-1	4	$\infty$	$\infty$
0	-1	2	$\infty$	$\infty$



(B, E):

$$D(E) > D(B) + W(B, E)?$$

(D, B):

$$D(B) > D(D) + W(D, B)?$$

(B, D):

$$D(D) > D(B) + W(B, D)?$$

(A, B):

$$D(B) > D(A) + W(A, B)? \text{ Yes}$$

$$D(B) = D(A) + W(A, B) = -1$$

(A, C):

$$D(C) > D(A) + W(A, C)? \text{ Yes}$$

$$D(C) = D(A) + W(A, C) = 4$$

(D, C):

$$D(C) > D(D) + W(D, C)?$$

(B, C):

$$D(C) > D(B) + W(B, C)? \text{ yes}$$

$$D(C) = D(B) + W(B, C) = 2$$

(E, D):

$$D(D) > D(E) + W(E, D)?$$

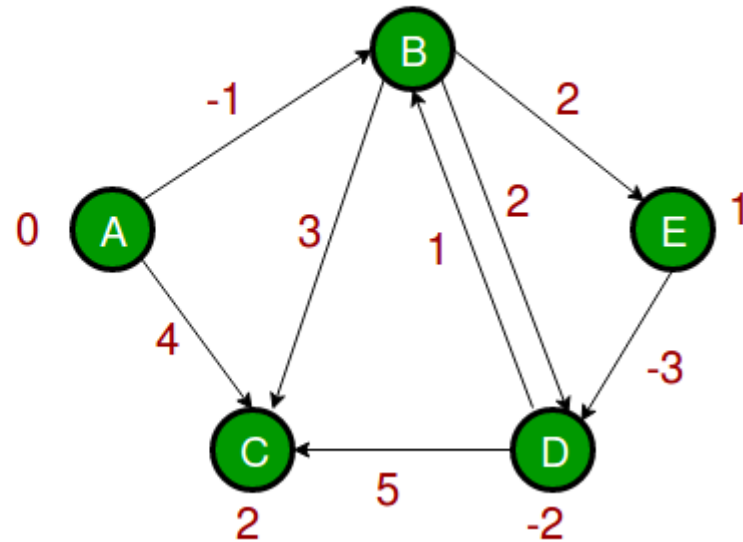
## ► Let all edges are processed in the following order:

- (B, E), (D, B), (B, D), (A, B), (A, C), (D, C), (B, C), (E, D).

# An Example

## ► Step 1 - Iteration #2

	A	B	C	D	E
A	0	$\infty$	$\infty$	$\infty$	$\infty$
B	0	-1	$\infty$	$\infty$	$\infty$
C	0	-1	4	$\infty$	$\infty$
D	0	-1	2	$\infty$	$\infty$
E	0	-1	2	$\infty$	1
	0	-1	2	1	1
	0	-1	2	-2	1



## ► Let all edges are processed in the following order:

- (B, E), (D, B), (B, D), (A, B), (A, C), (D, C), (B, C), (E, D).

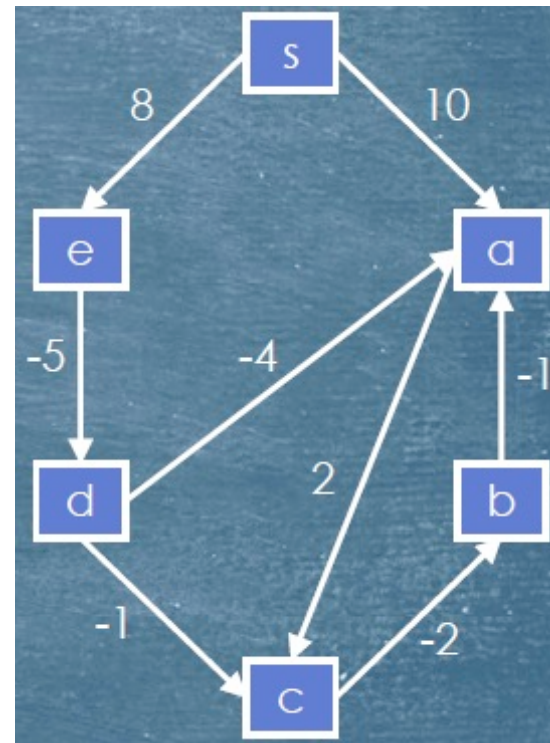
# Different from Dijkstra's



- ▶ Unlike Dijkstra's algorithm, in which we update only the most promising (next lowest cost) vertex at each iteration, Bellman-Ford updates every vertex at each iteration.
- ▶ This means that each iteration of Bellman-Ford involves more work than the corresponding iteration of Dijkstra

# Another Example

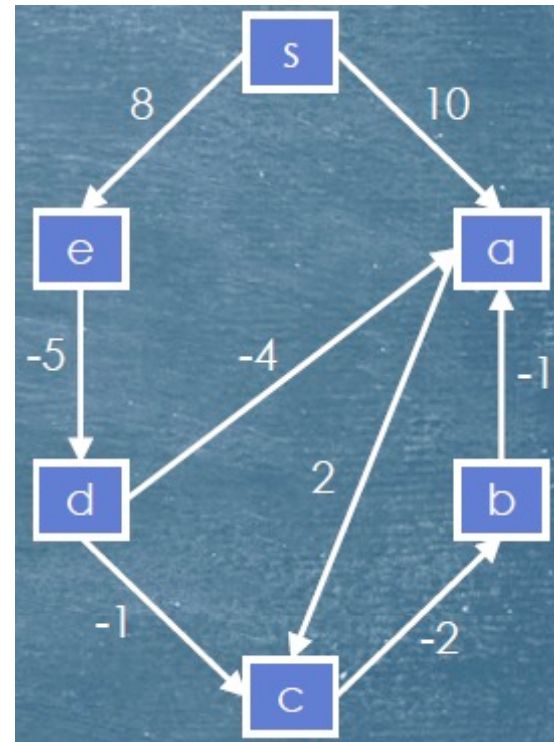
- ▶ Consider the following graph:
- ▶ It has 6 vertices so we will run through the main loop 5 times.
- ▶ Order of edges
  - $(u, v), u = s, a, b, c, d, e$



# Initialization

Set initial values of D

s	a	b	c	d	e
0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$



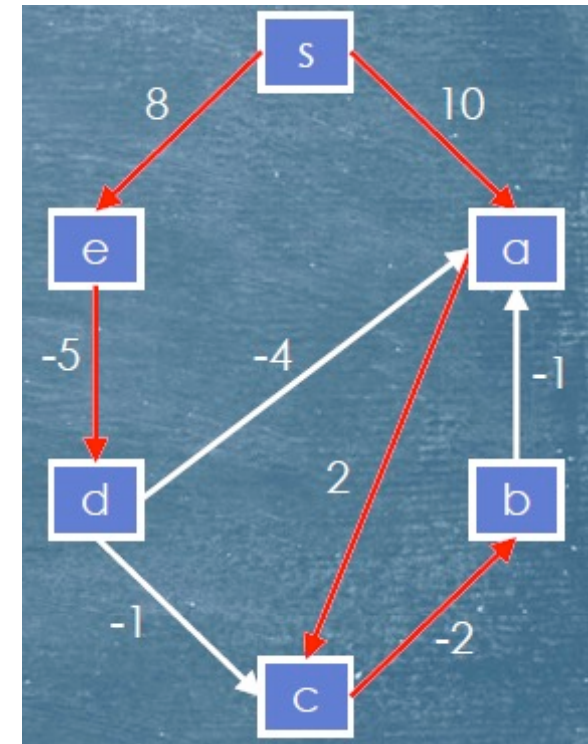
# Step 2 - iteration #1

(s,a), (s,e) - update D[a], D[e]  
(a,c) - update D[c]  
(b,a) - no update  
(c,b) - update D[b]  
(d,a), (d,c) - no update  
(e,d) - update D[d]

s	a	b	c	d	e
0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$



s	a	b	c	d	e
0	10	10	12	3	8



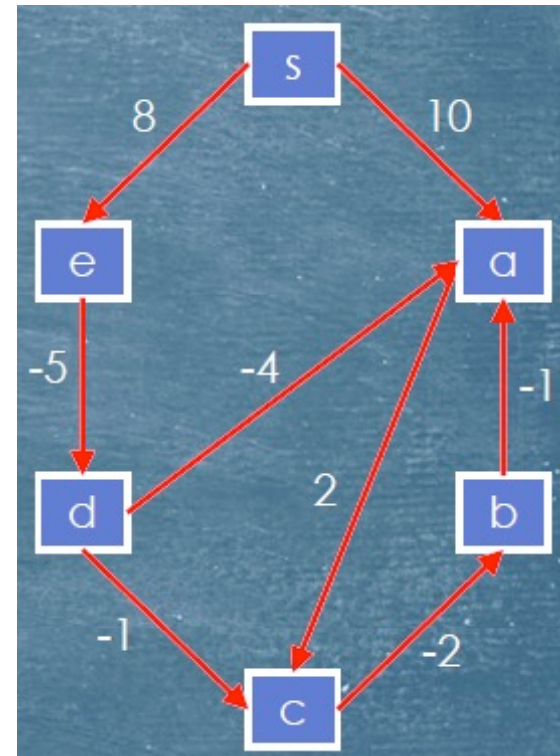
# Step 2 - iteration #2

(s,a), (s,e) - no update  
(a,c) - no update  
(b,a) - no update  
(c,b) - no update  
(d,a), (d,c) - update  $D[a], D[c]$   
(e,d) - no update

s	a	b	c	d	e
0	10	10	12	3	8



s	a	b	c	d	e
0	-1	10	2	3	8





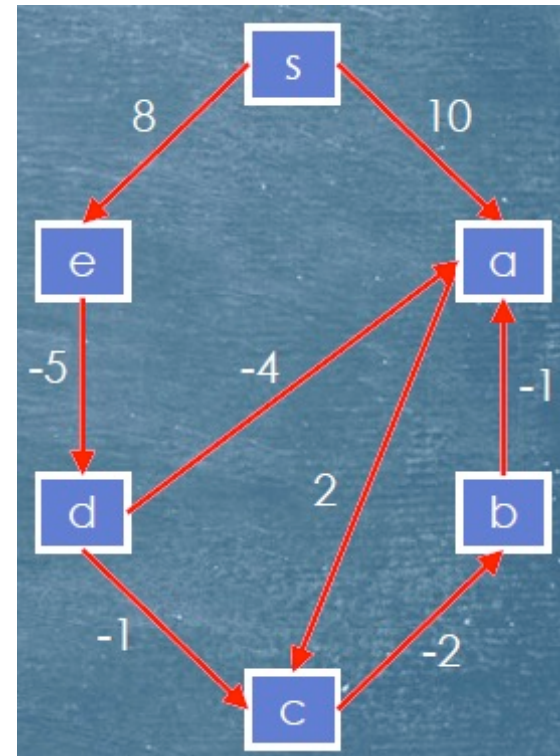
# Step 2 - iteration #3

(s,a), (s,e) - no update  
(a,c) - update D[c]  
(b,a) - no update  
(c,b) - update D[b]  
(d,a), (d,c) - no update  
(e,d) - no update

s	a	b	c	d	e
0	-1	10	2	3	8



s	a	b	c	d	e
0	-1	-1	1	3	8



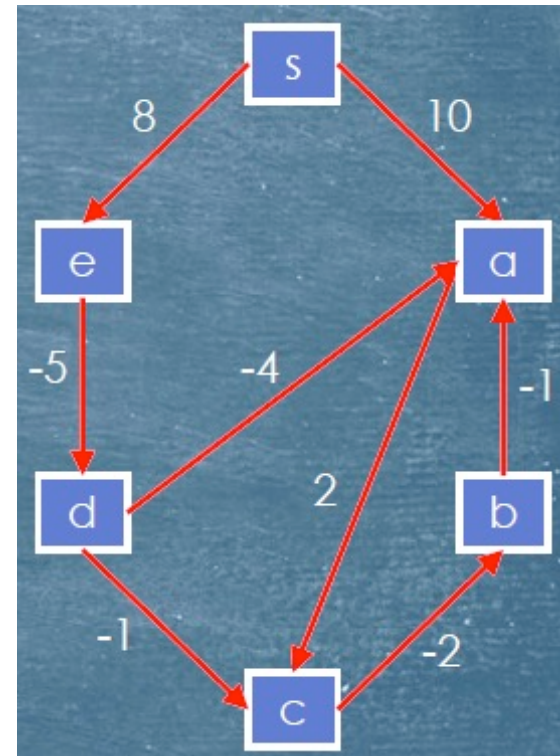
# Step 2 - iteration #4

(s,a), (s,e) - no update  
(a,c) - no update  
(b,a) - update D[a]  
(c,b) - no update  
(d,a), (d,c) - no update  
(e,d) - no update

s	a	b	c	d	e
0	-1	-1	1	3	8



s	a	b	c	d	e
0	-2	-1	1	3	8



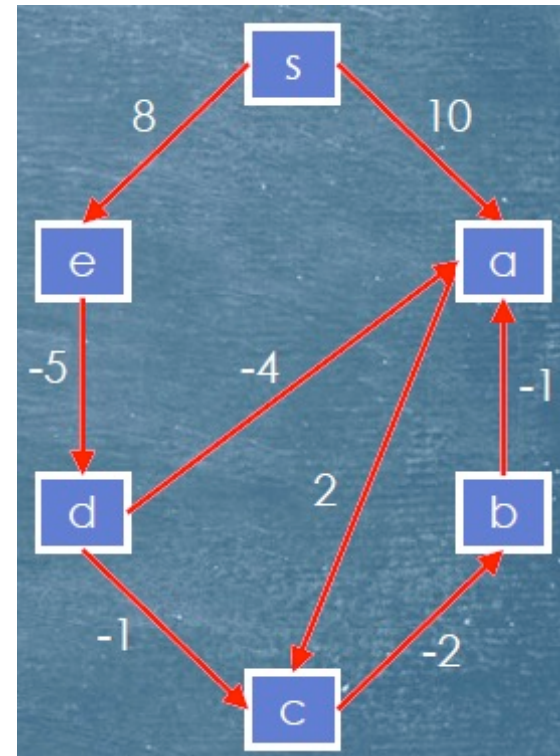
# Step 2 - iteration #5

(s,a), (s,e) - no update  
(a,c) - update D[c]  
(b,a) - no update  
(c,b) - update D[b]  
(d,a), (d,c) - no update  
(e,d) - no update

s	a	b	c	d	e
0	-2	-1	1	3	8



s	a	b	c	d	e
0	-2	-2	0	3	8

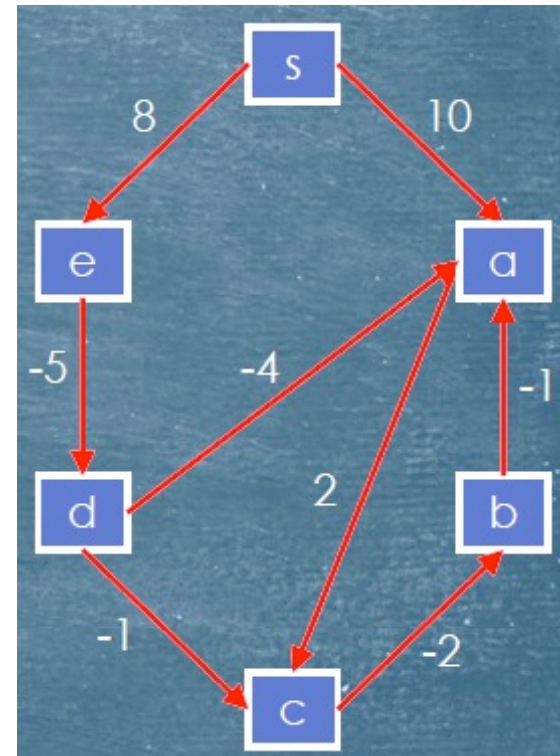


# Step 3 - Check Negative Weight Cycle

$(s, a) - D[a] > D[s] + w(s, a) ?$   
 $(s, e) - D[e] > D[s] + w(s, e) ?$   
 $(a, c) - D[c] > D[a] + w(a, c) ?$   
 $(b, a) - D[a] > D[b] + w(b, a) ?$  yes  
 $(c, b) - D[b] > D[c] + w(c, b) ?$   
 $(d, a) - D[a] > D[d] + w(d, a) ?$   
 $(d, c) - D[c] > D[d] + w(d, c) ?$   
 $(e, d) - D[d] > D[e] + w(e, d) ?$

The graph contains a negative cost cycle that involves vertex a.

s	a	b	c	d	e
0	-2	-2	0	3	8



# Analysis



- ▶ Bellman-Ford performs the major loop  $|V| - 1$  times.
- ▶ Inside this loop it checks every edge;  $|E|$  operations.
- ▶ Finally, it does another  $|E|$  checks for potential cycles.
- ▶ Overall, Bellman-Ford has  $\Theta(|V| \times |E|)$  complexity.

# Notes on Bellman-Ford Algorithm

- ▶ Bellman-Ford algorithm can handle directed and undirected graphs with non-negative weights.
- ▶ However, it can only handle directed graphs with negative weights, as long as we don't have negative cycles.
- ▶ When the graph has a negative cycle, Bellman-Ford algorithm can detect the cycle, but won't be able to find the shortest paths in this case.

# Related References



- ▶ Introduction to the Design and Analysis of Algorithms, A. Levitin, 3rd Ed., Pearson 2011.
  - Chapters 9.3
- ▶ Introduction to Algorithms, T. H. Cormen, 3rd Ed, MIT Press 2009.
  - Chapters 24.1 and 24.3