

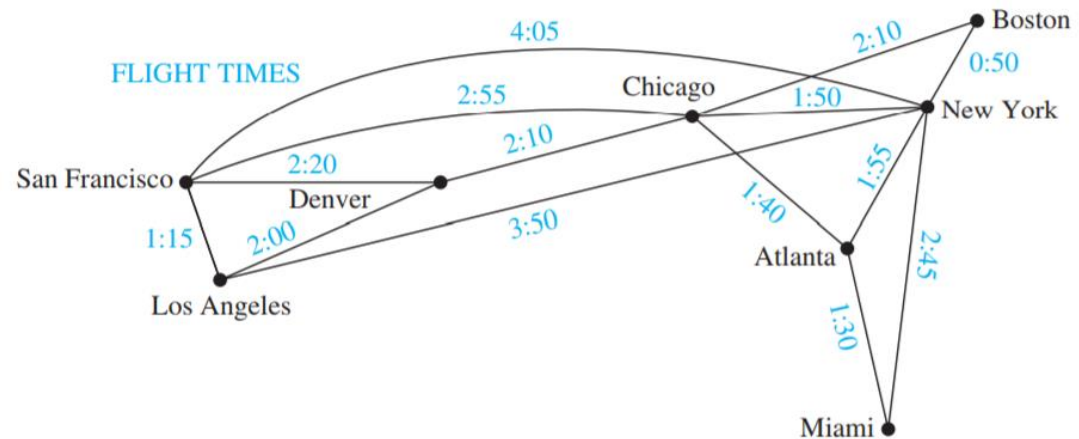
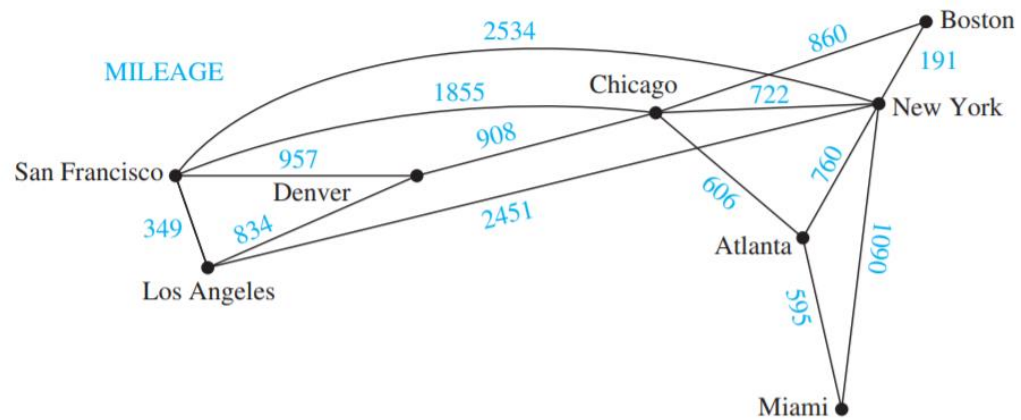
Single Source Shortest Path

Shortest Path Problems

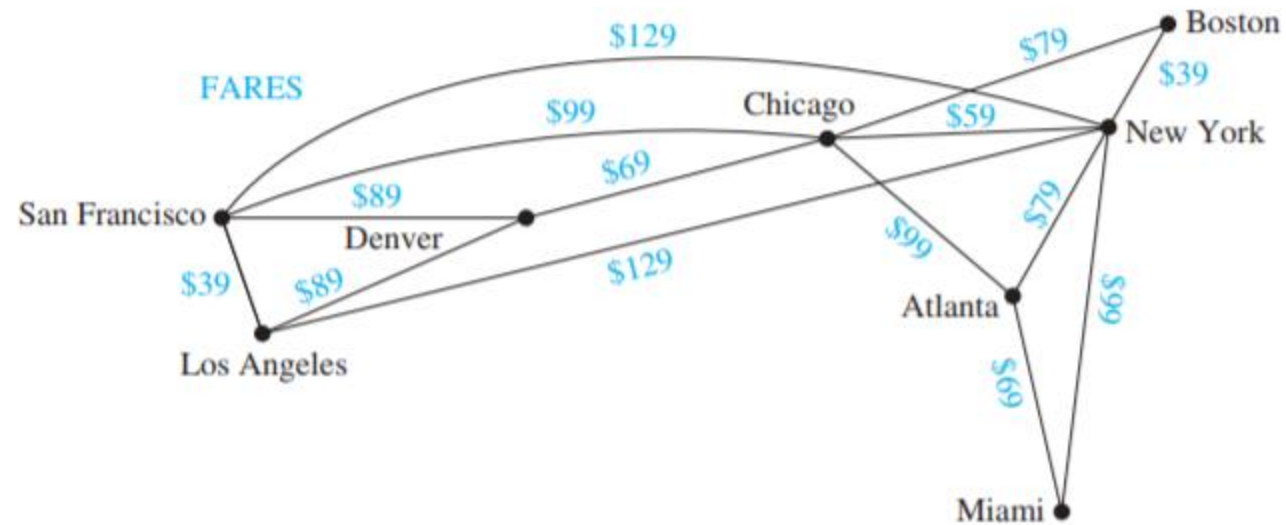
Weighted Graphs

- Many problems can be modeled using graphs with **weights** assigned to their edges
- representing cities by vertices and flights by edges
- assigning distances between cities to the edges
- assigning flight times to edges
- assigning fares to the edges

Weighted Graphs



Weighted Graphs



Weighted Graphs

- Graphs that have a number assigned to each edge

Used to model:

- Communication networks
- Communication costs
- Response times of the computers over these lines
- Distance between computers

Shortest Path Algorithm

- Generalize distance to weighted setting
- Digraph $G = (V, E)$ with weight function $W: E \rightarrow \mathbb{R}$ (assigning real values to edges)
- Weight of path $p = v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_k$ is

$$w(p) = \sum_{i=1}^{k-1} w(v_i, v_{i+1})$$

- Shortest path = a path of the minimum weight
- Applications
 - static/dynamic network routing
 - robot motion planning
 - map/route generation in traffic

Shortest Path Problems

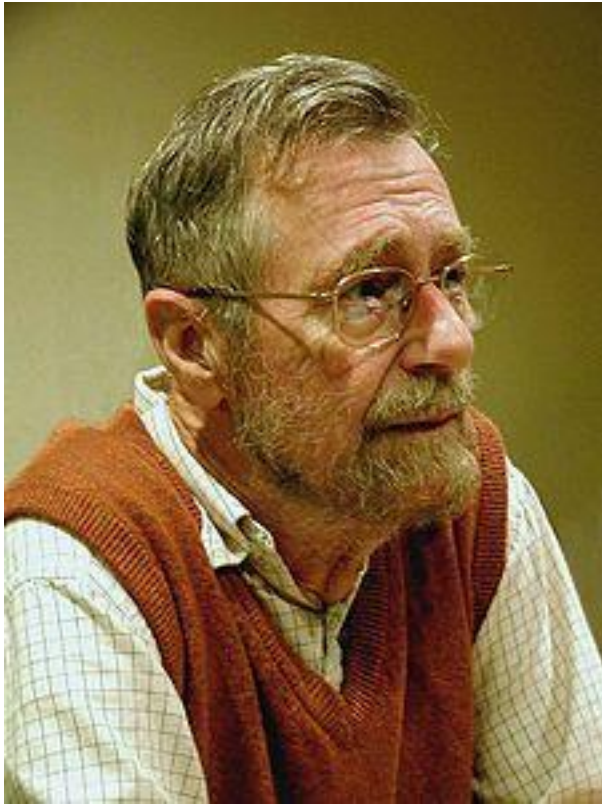
- **Single-source (single-destination).** Find a shortest path from a given source (vertex s) to each of the vertices. *The topic of this lecture.*
- **Single-pair.** Given two vertices, find a shortest path between them. Solution to single-source problem solves this problem efficiently, too.
- **All-pairs.** Find shortest-paths for every pair of vertices. Dynamic programming algorithm.
- **Unweighted shortest-paths** – BFS.

For this...

We'll have Dijkstra's Algorithm as an example.

Dijkstra's Algorithm

Edsger Wybe Dijkstra



May 11, 1930 – August 6, 2002

Dutch computer scientist from
Netherlands

Received the 1972 A. M. Turing Award

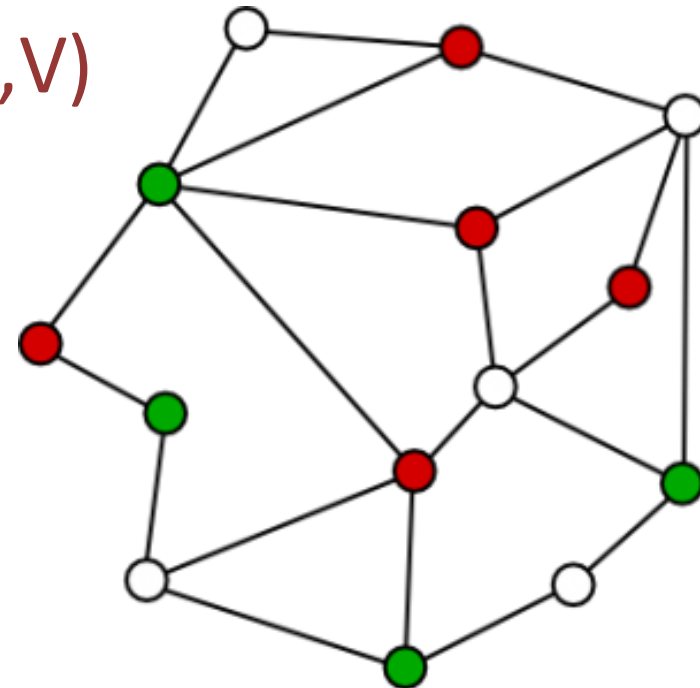
Known for his many essays on
programming

Single-Source Single Path Problem

- The problem of finding shortest paths from a source vertex v to all other vertices in the graph.

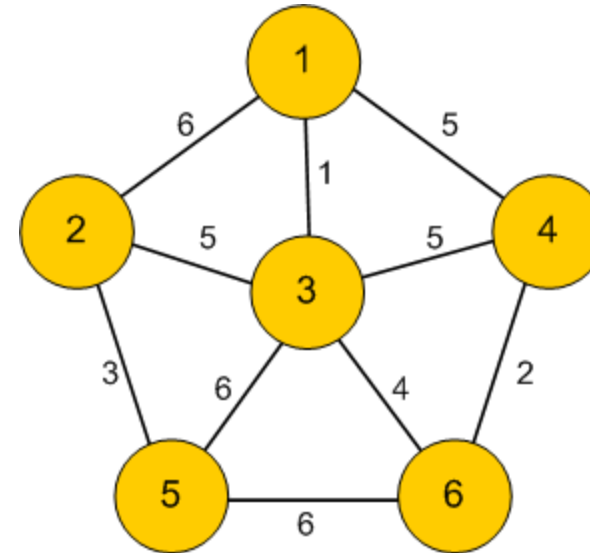
Weighted graph $G = (E, V)$

Source vertex $s \in V$ to all vertices $v \in V$



Solution to Single-Source Single Path Problem

- Works on both **directed** and **undirected** graphs.
- All edges must have nonnegative weights.
- Graph must be connected

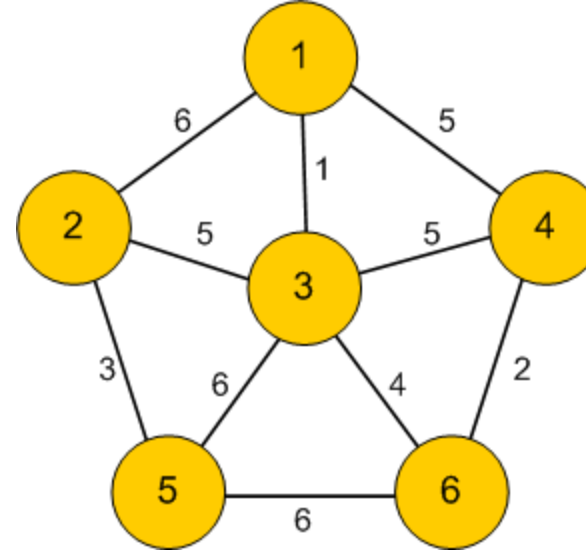


Output of Dijkstra's Algorithm

Original algorithm outputs value of shortest path
not the path itself

Value: $\delta(1,5) = 7$

Path: {1,3,5}



Implementation

Edge Relaxation

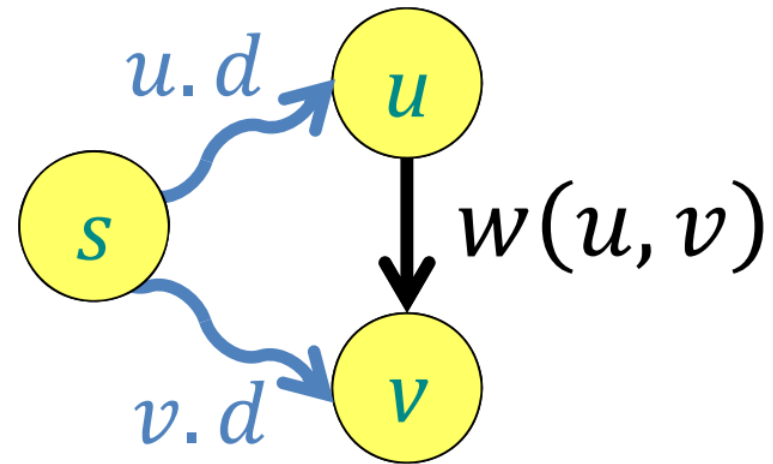
- Relaxing an edge, (a concept you can find in other shortest-path algorithms as well) is trying to lower the cost of getting to a vertex by using another vertex.

Edge Relaxation

Consider an edge (u,v)

If $D[v] > D[u] + w(u,v)$ then

$D[v] = D[u] + w(u,v)$



Edge Relaxation

Maintain value $D[u]$ for each vertex

Each starts at infinity, and decreases as we find out about a shorter path from v to u ($D[v] = 0$)

Maintain priority queue, Q , of vertices to be relaxed

Use $D[u]$ as key for each vertex

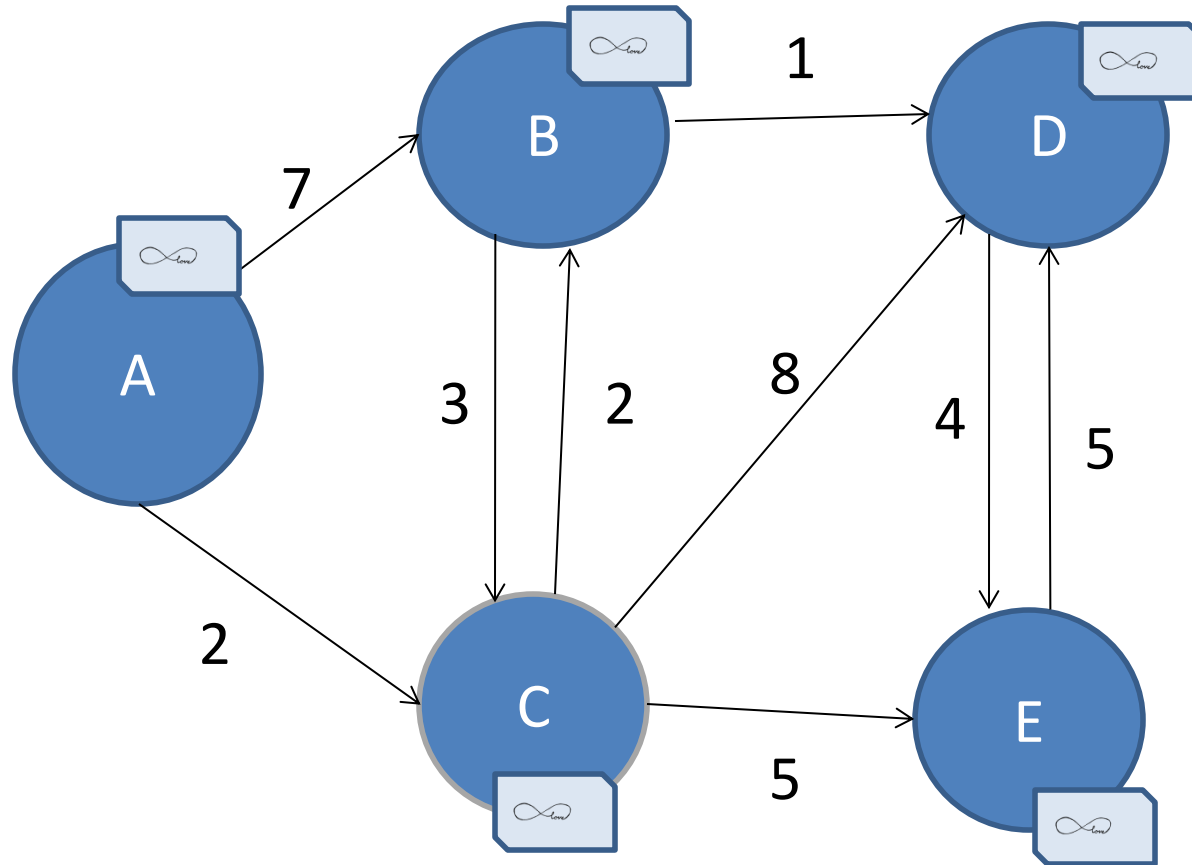
Remove min vertex from Q , and relax its neighbors

Dijkstra's

```
dist[s] ← 0                                (distance to source vertex is zero)
for all v ∈ V − {s}
    do dist[v] ← ∞                          (set all other distances to infinity)
S ← ∅                                       (S, the set of visited vertices is initially empty)
Q ← V                                       (Q, the queue initially contains all vertices)
while Q ≠ ∅                                (while the queue is not empty)
do u ← mindistance(Q, dist)                (select the element of Q with the min. distance)
    S ← S ∪ {u}                             (add u to list of visited vertices)
    for all v ∈ neighbors[u]
        do if dist[v] > dist[u] + w(u, v)    (if new shortest path found)
            then d[v] ← d[u] + w(u, v)       (set new value of shortest path)
                                                (if desired, add traceback code)

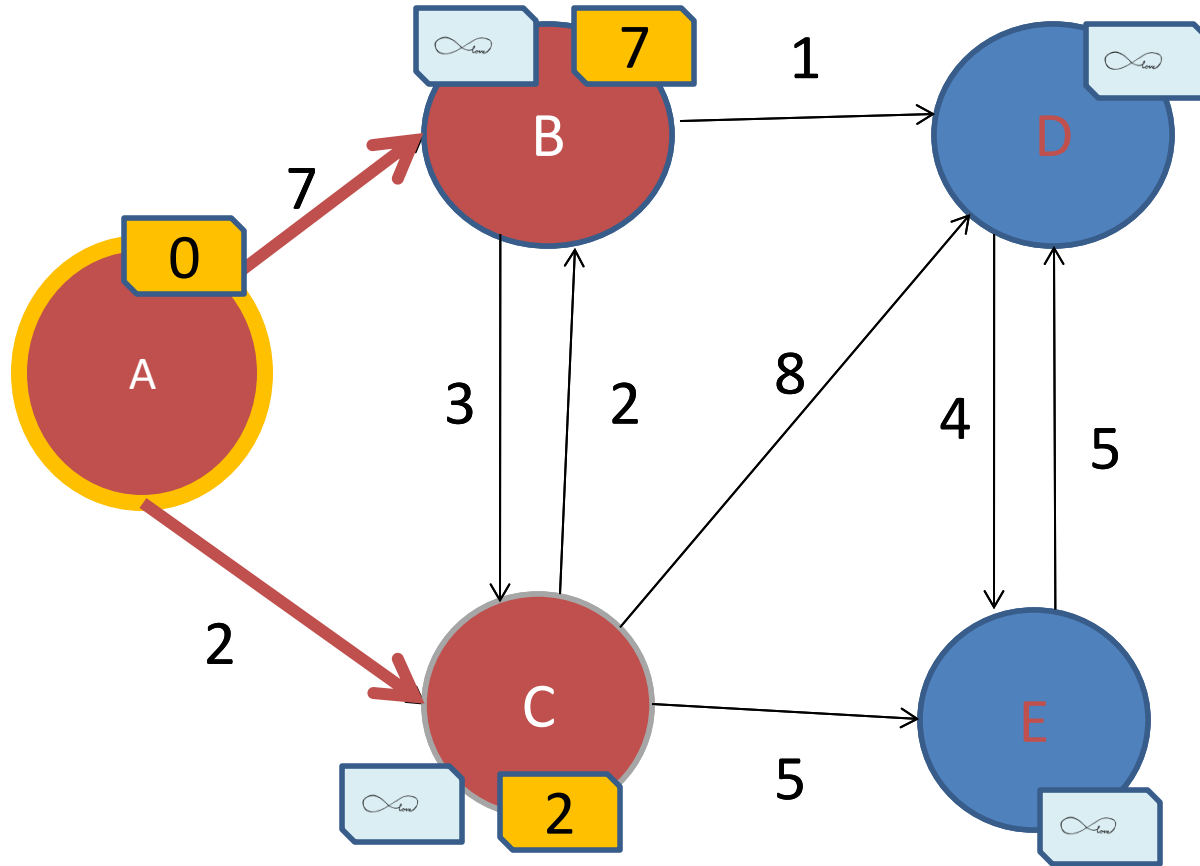
return dist
```

Shortest Path



	A	B	C	D	E
A	∞	∞	∞	∞	∞
B					
C					
D					
E					

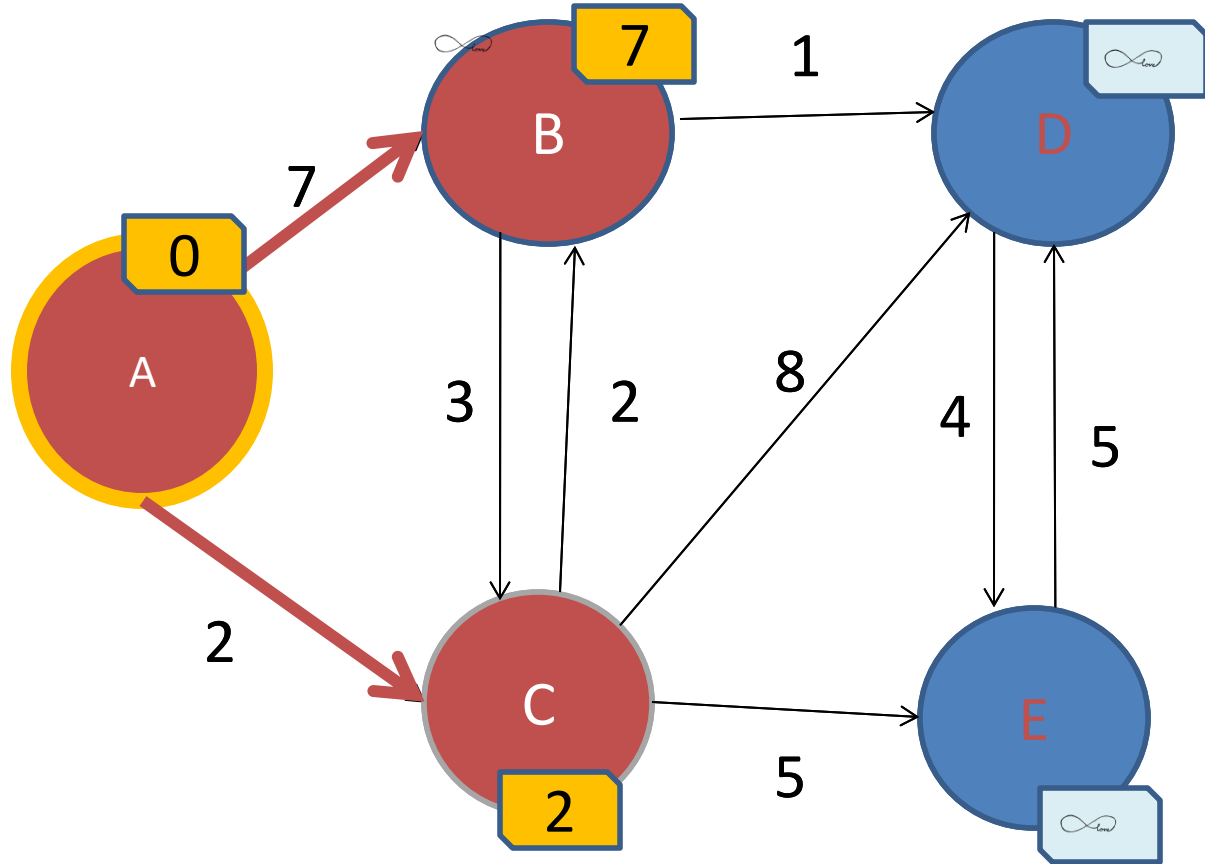
Shortest Path



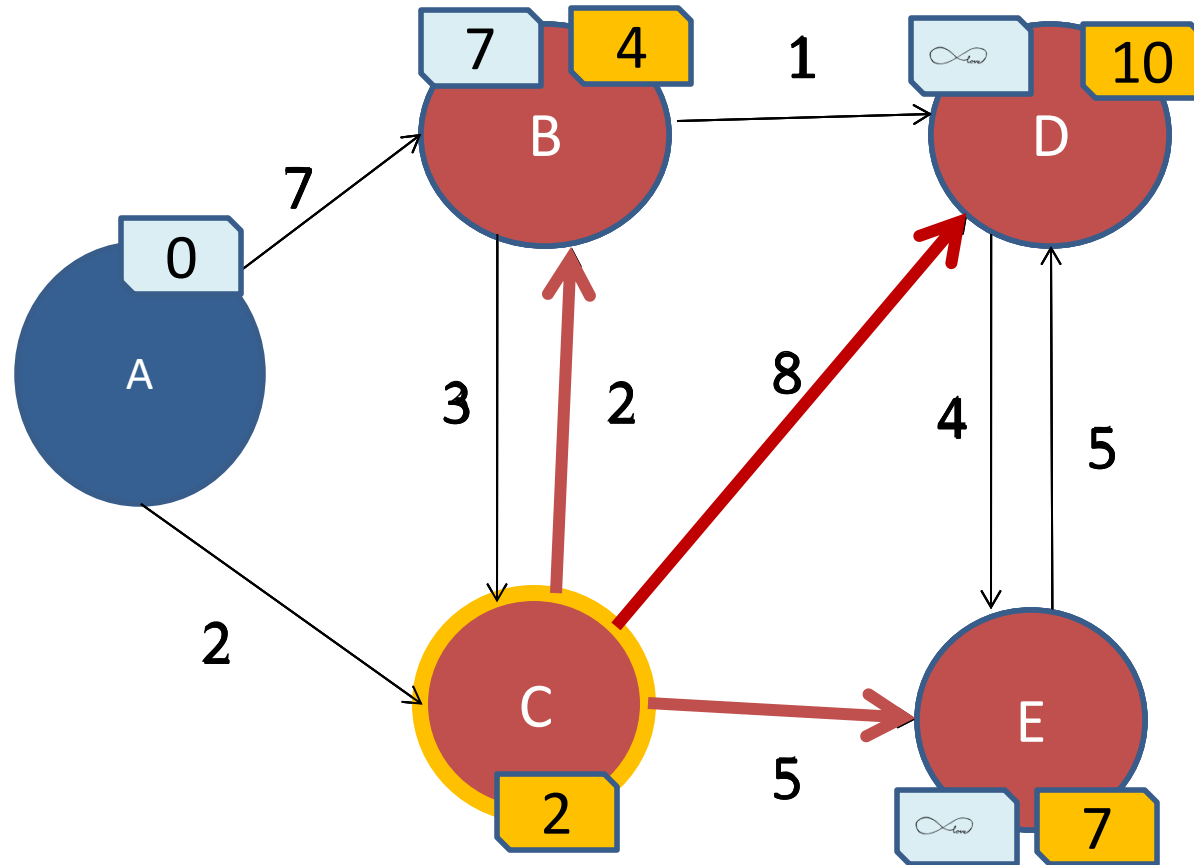
	A	B	C	D	E
A	0	7	2	∞	∞
B	7	0	2	∞	∞
C	2	2	0	∞	∞
D	∞	∞	∞	0	7
E	∞	∞	∞	7	0

	A	B	C	D	E
A	0	7	2	∞	∞
B	7	0	2	∞	∞
C	2	2	0	∞	∞
D	∞	∞	∞	0	7
E	∞	∞	∞	7	0

Shortest Path

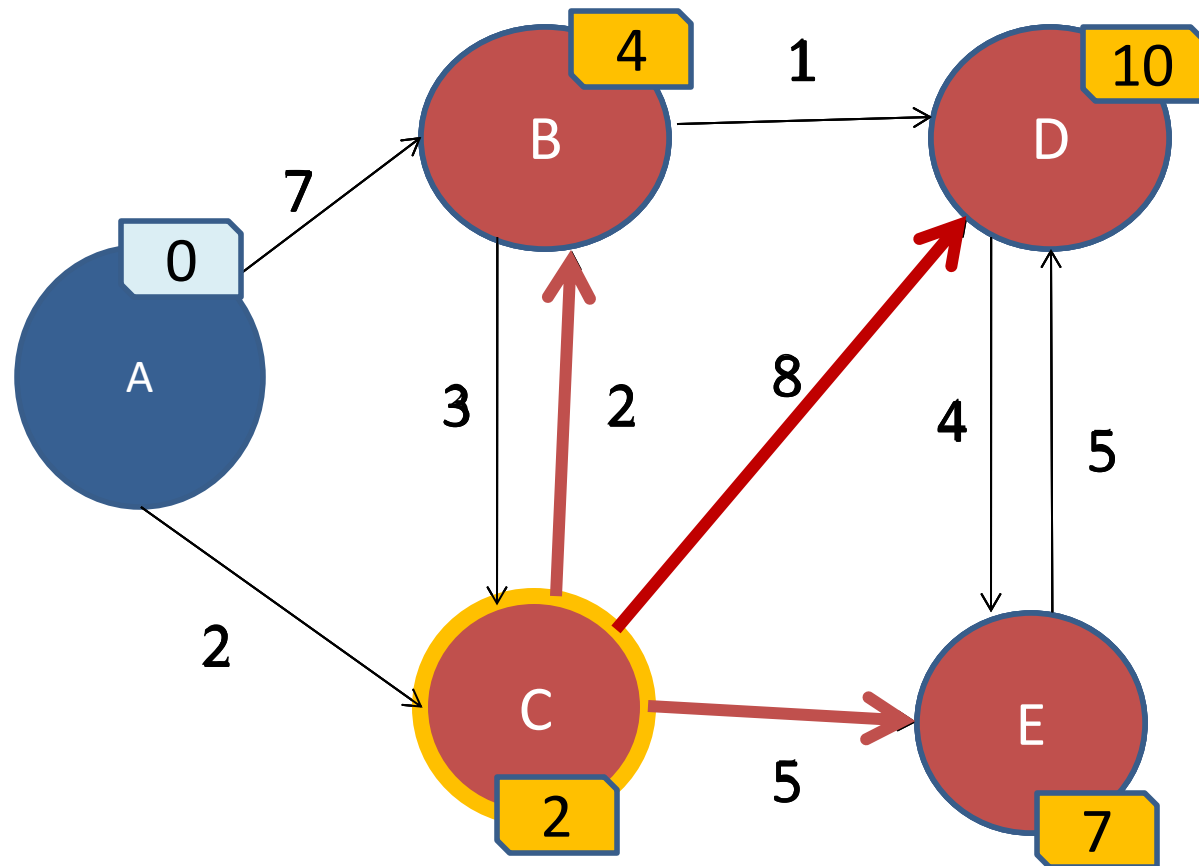
[illegible][illegible]

Shortest Path



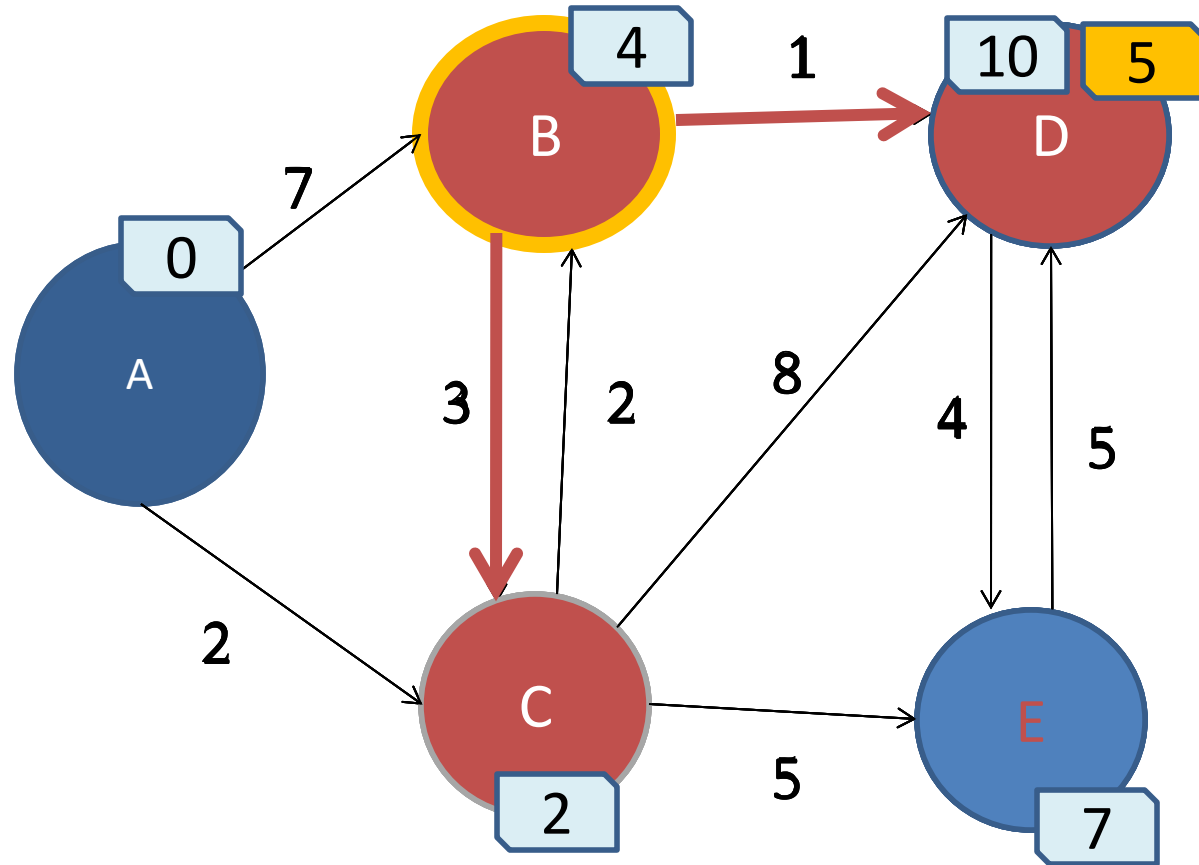
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C

Shortest Path



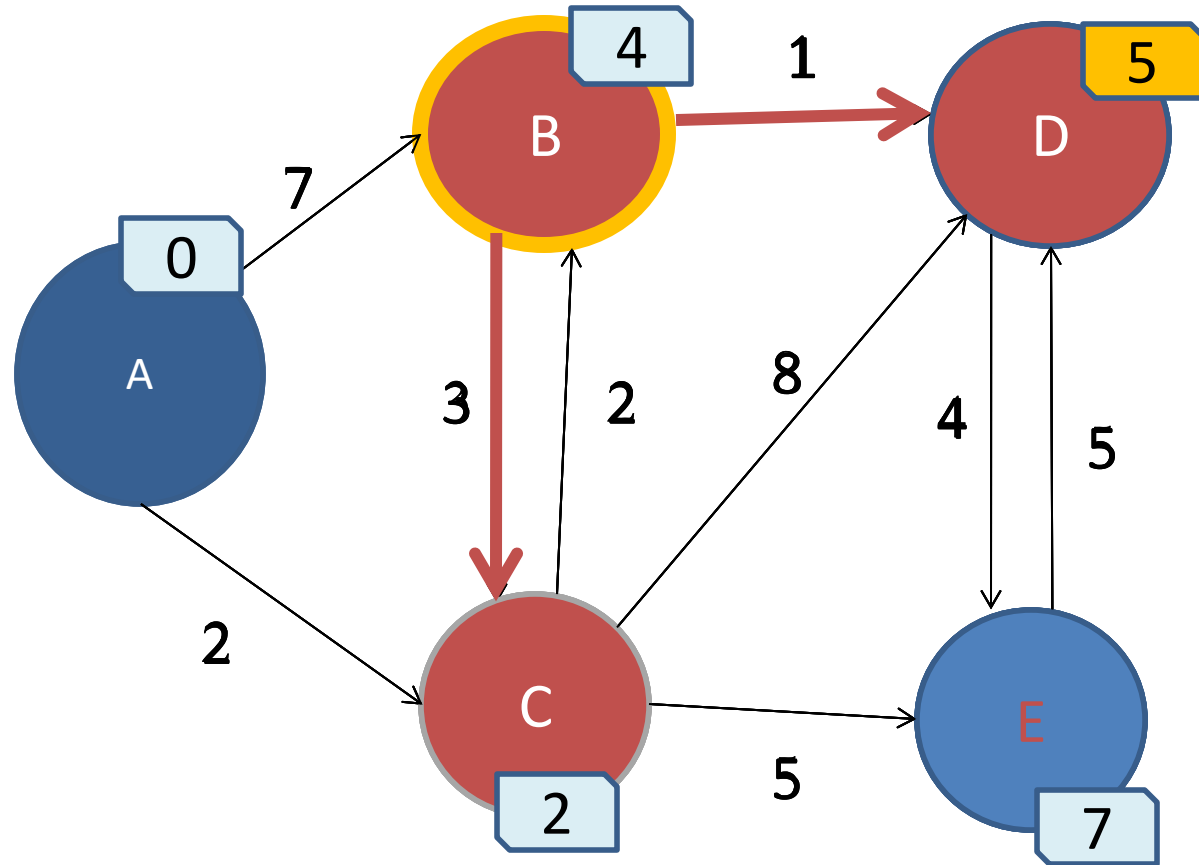
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C

Shortest Path



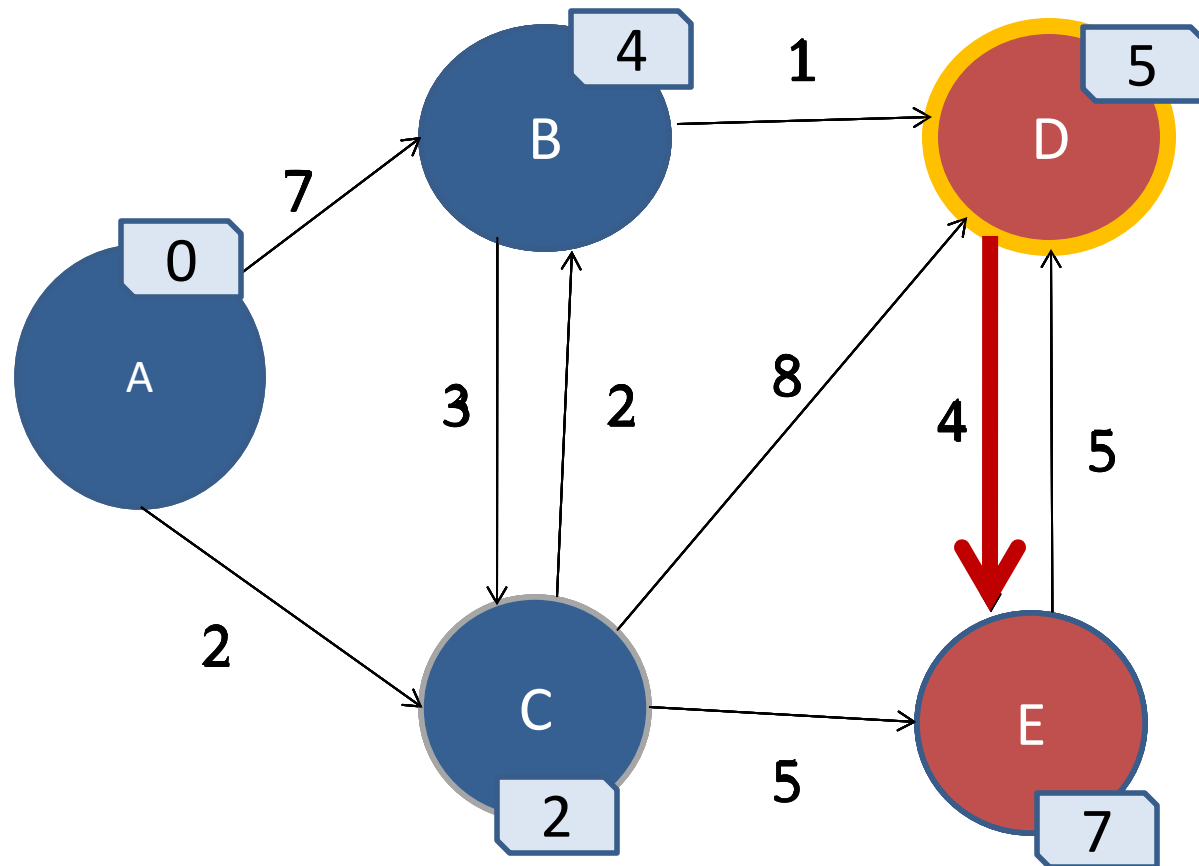
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C

Shortest Path



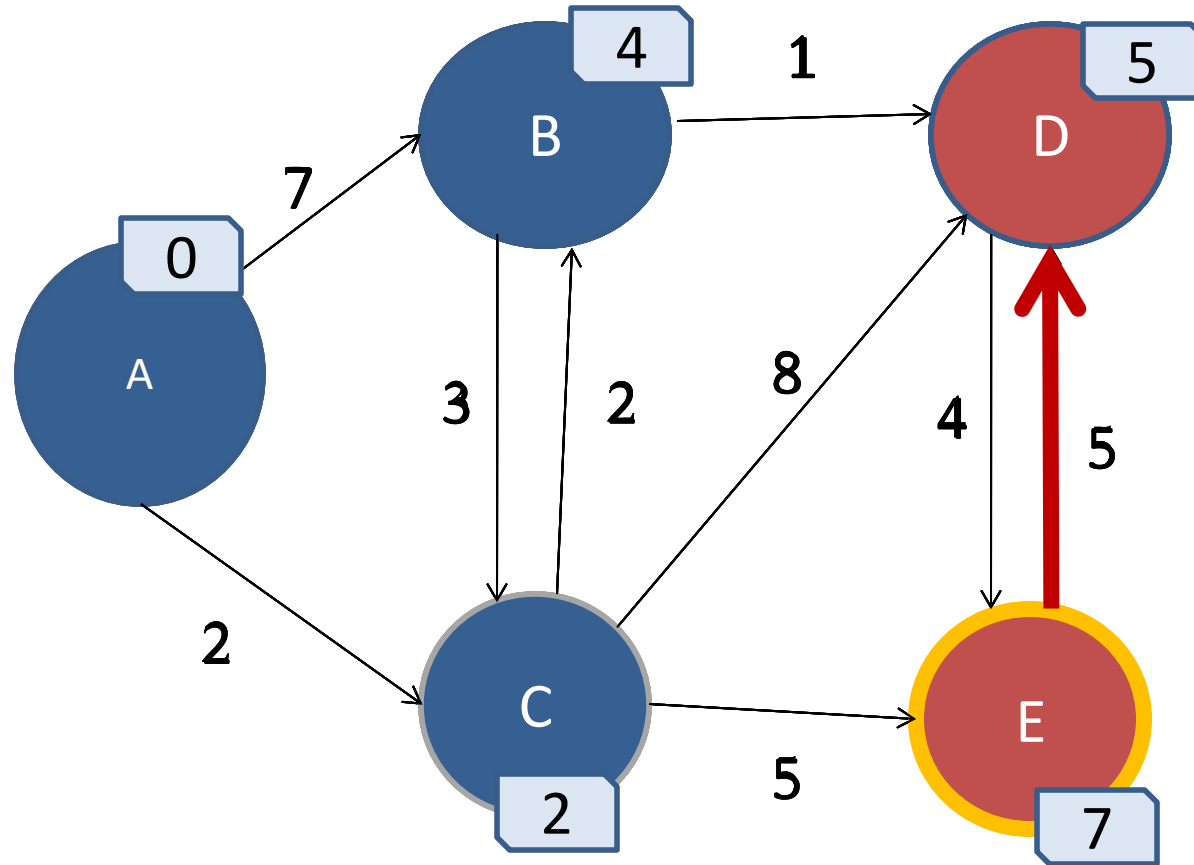
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C

Shortest Path



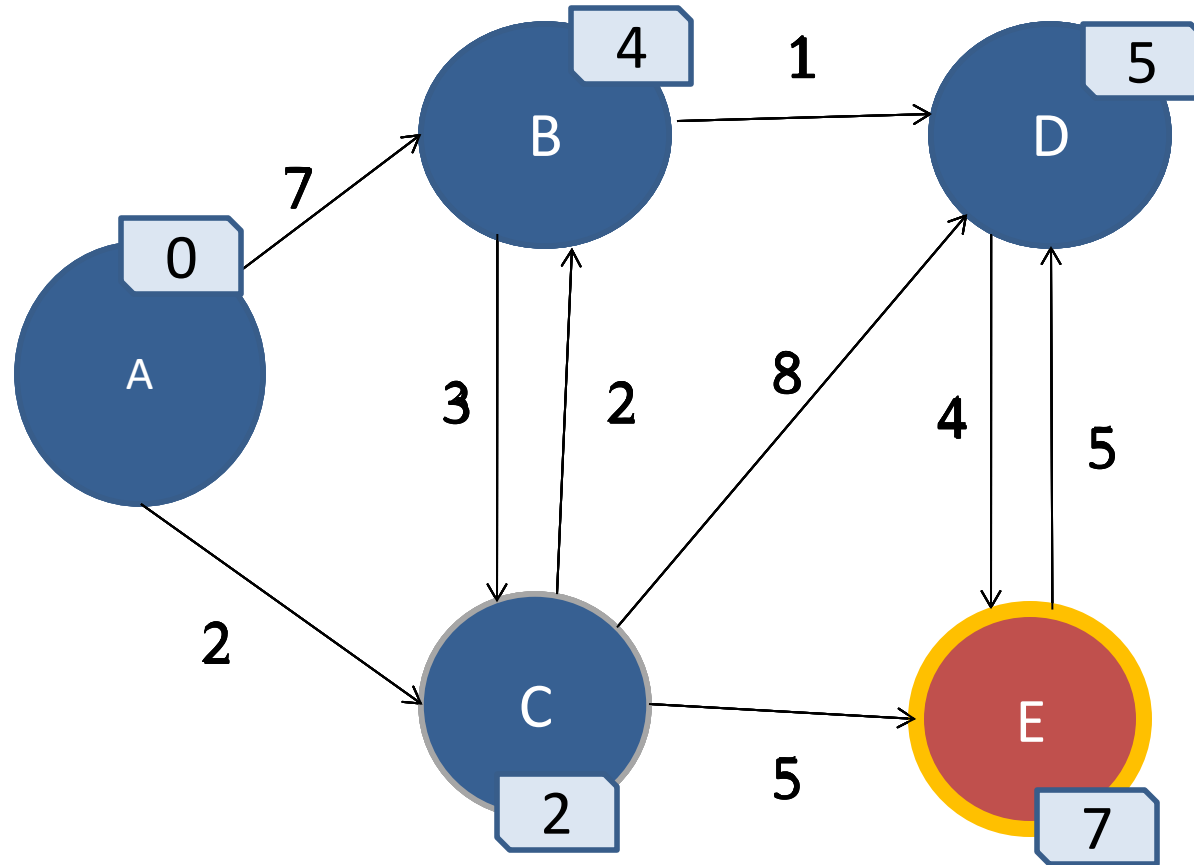
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C

Shortest Path



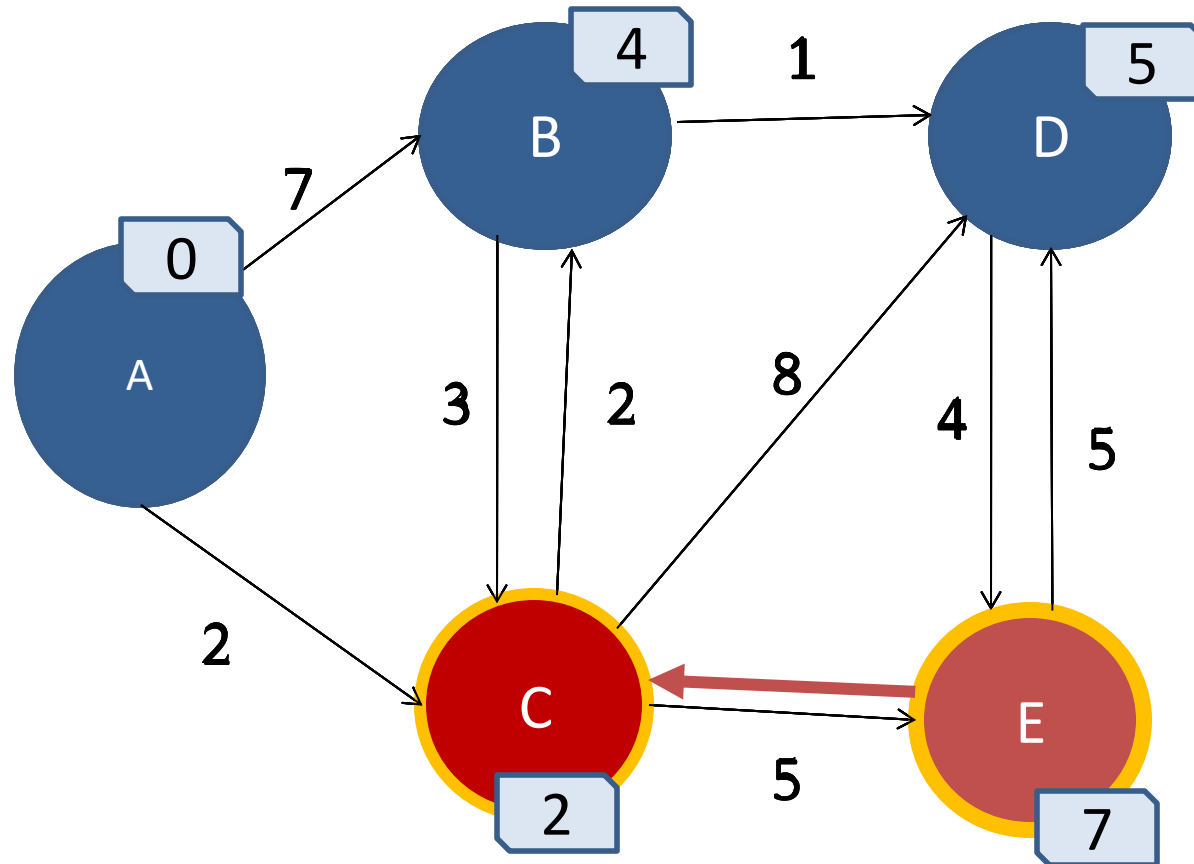
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C
E	0	4 C	2 A	5 B	7 C

Shortest Path



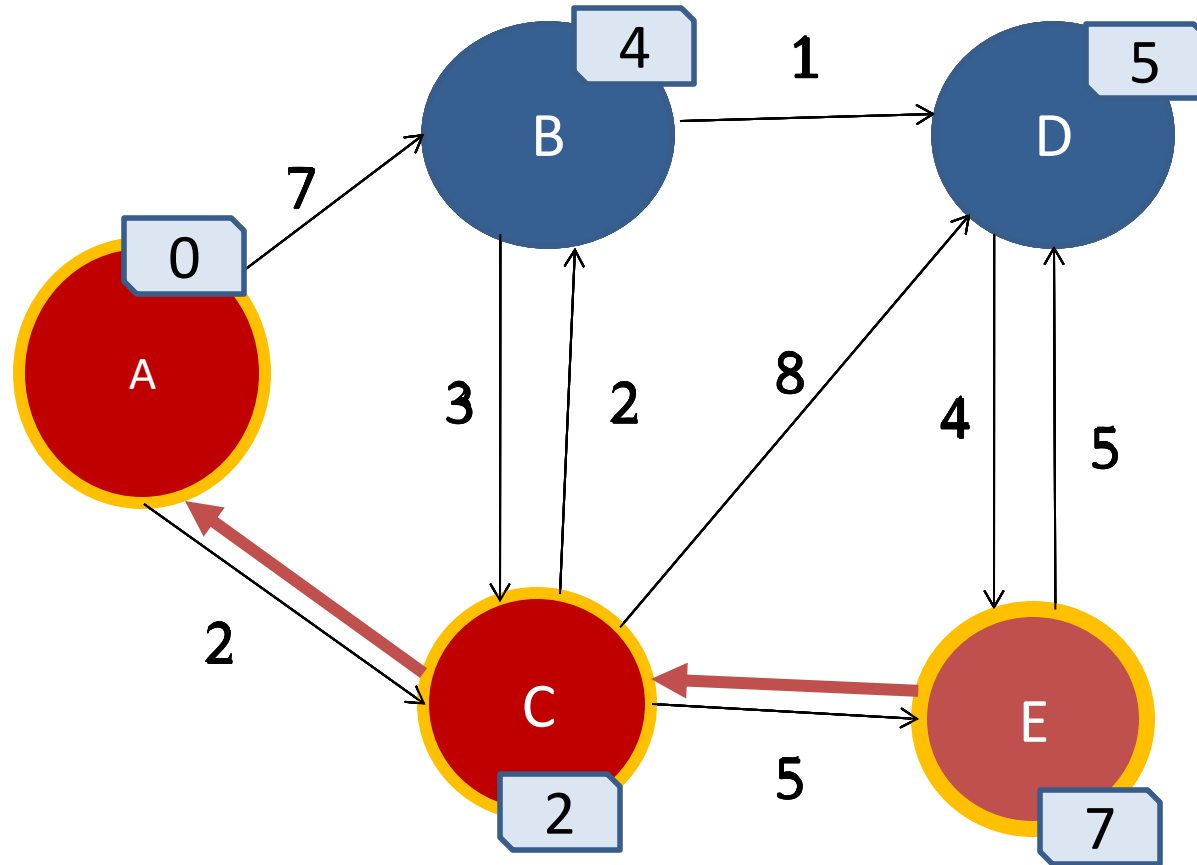
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C
E	0	4 C	2 A	5 B	7 C

Shortest Path



	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C
E	0	4 C	2 A	5 B	7 C

Shortest Path

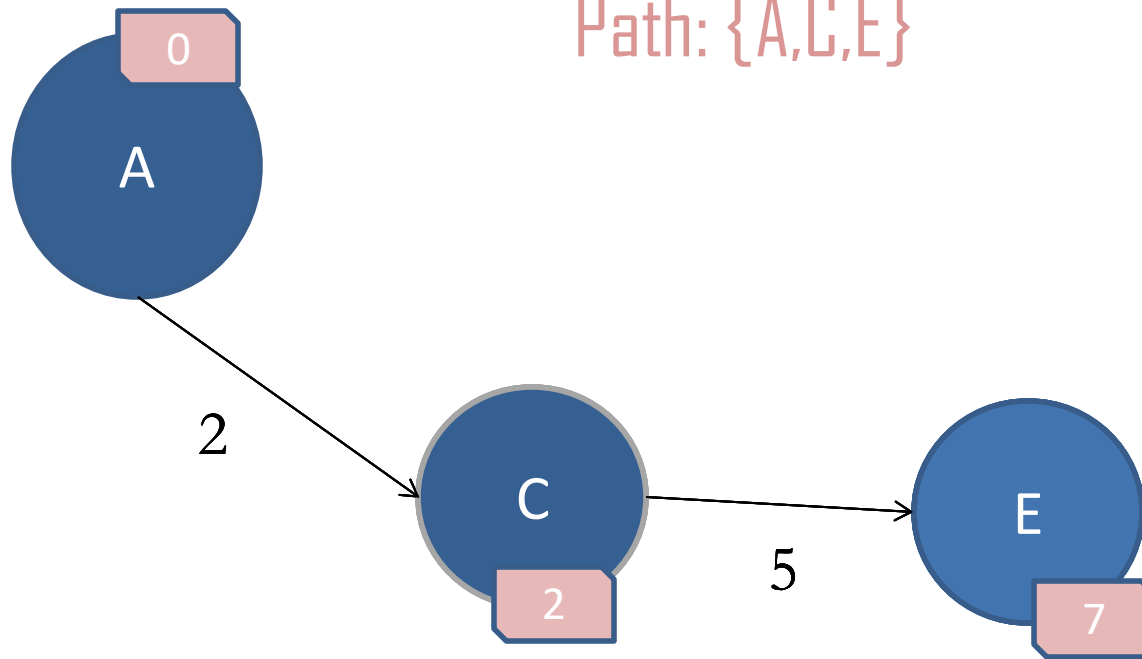


	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C
E	0	4 C	2 A	5 B	7 C

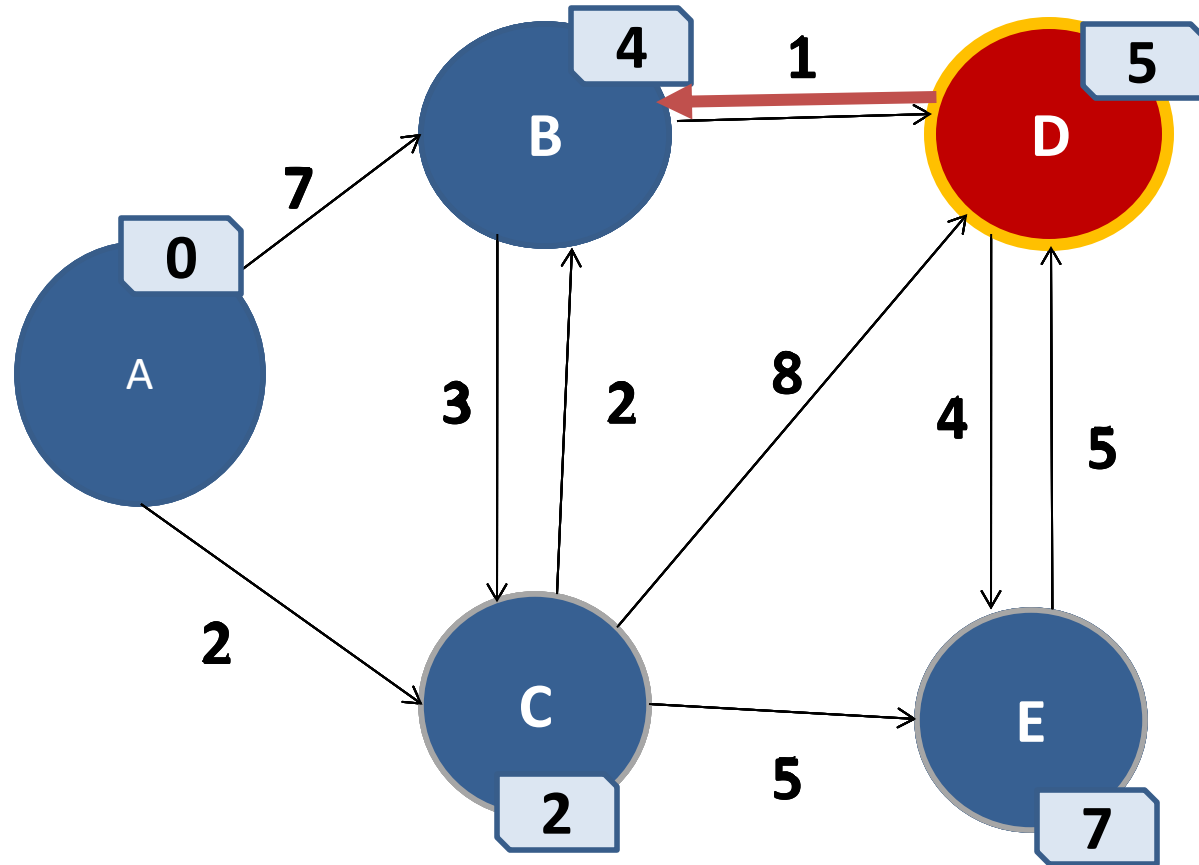
Solution

Value: $\delta(A,E) = 7$

Path: $\{A,C,E\}$

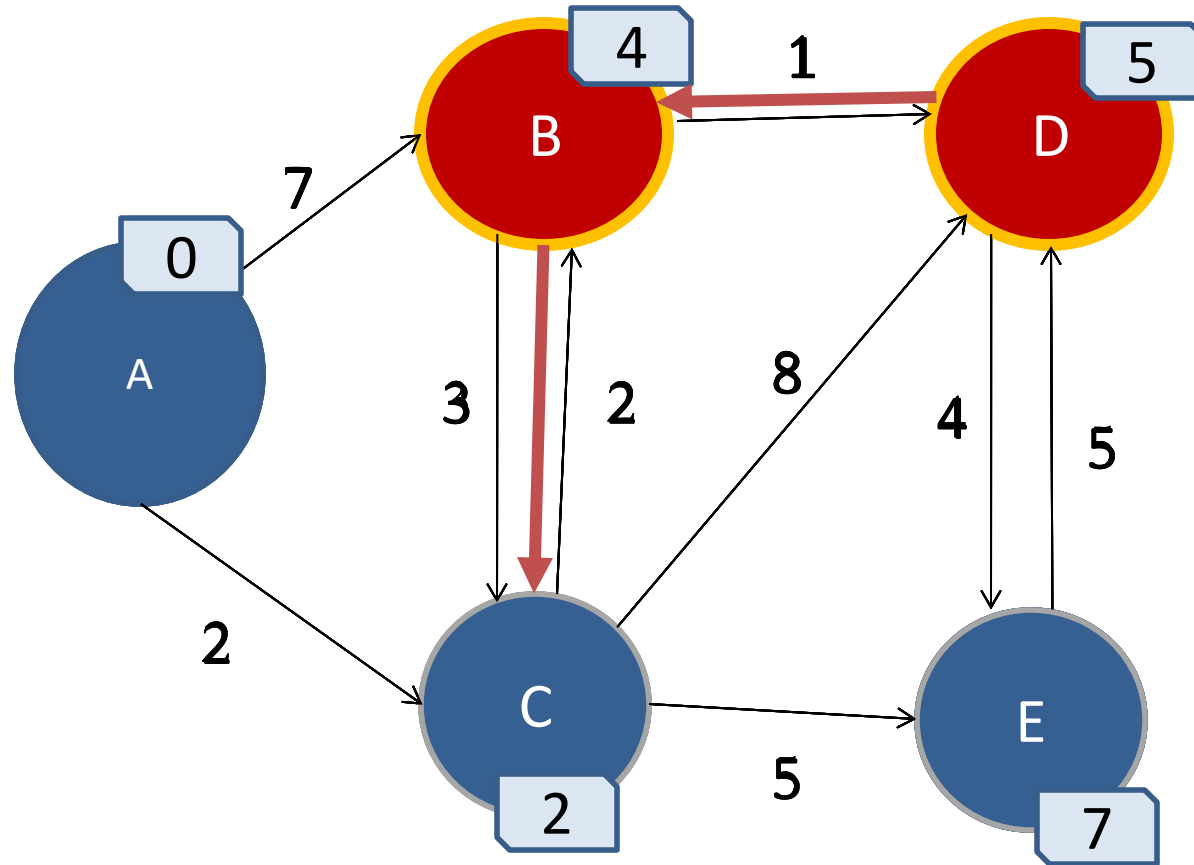


Shortest Path



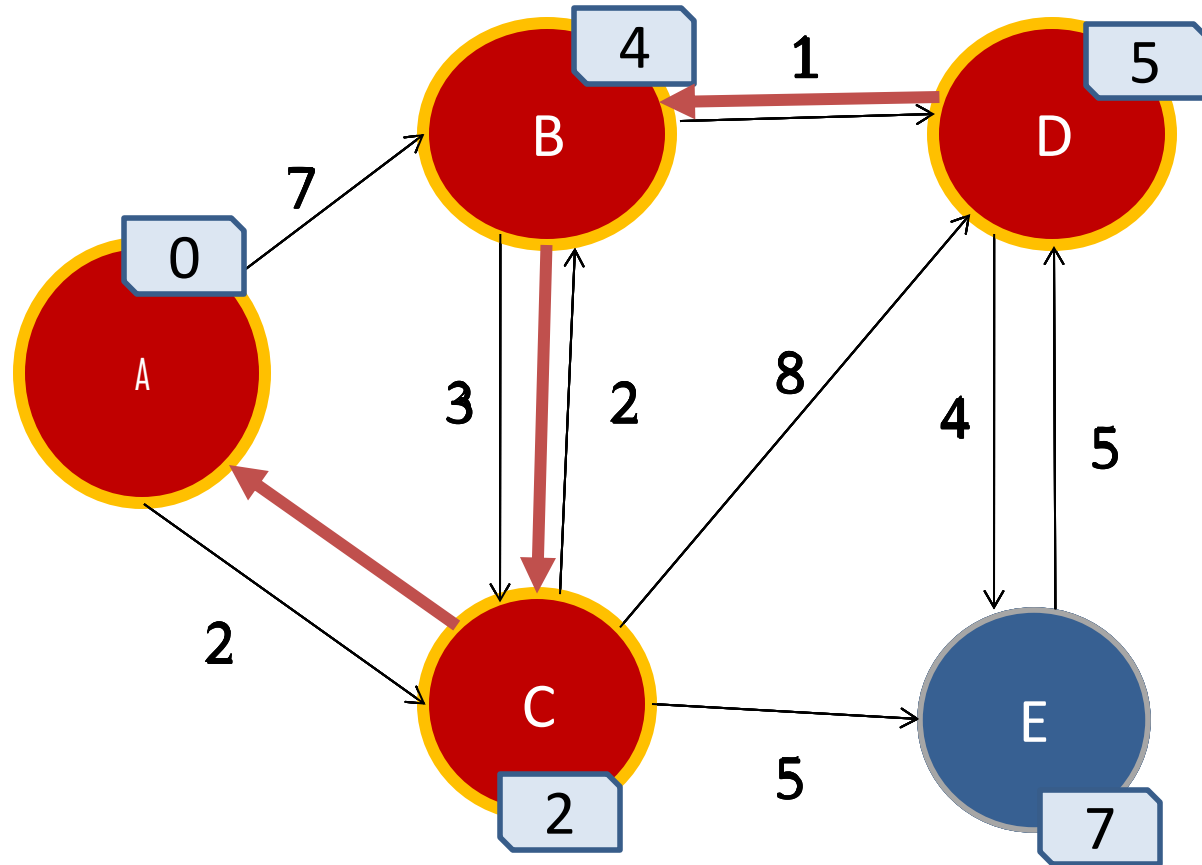
	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C
E	0	4 C	2 A	5 B	7 C

Shortest Path



	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C
E	0	4 C	2 A	5 B	7 C

Shortest Path



	A	B	C	D	E
	∞	∞	∞	∞	∞
A	0	7 A	2 A	∞	∞
C	0	4 C	2 A	10 C	7 C
B	0	4 C	2 A	5 B	7 C
D	0	4 C	2 A	5 B	7 C
E	0	4 C	2 A	5 B	7 C

Edge Relaxation

Looking for the minDistance is $O(V)$

$O(V)$ $\text{dist}[s] \leftarrow 0$ (distance to source vertex is zero)
for all $v \in V - \{s\}$
do $\text{dist}[v] \leftarrow \infty$ (set all other distances to infinity)
 $S \leftarrow \emptyset$ (S , the set of visited vertices is initially empty)
 $Q \leftarrow V$ (Q , the queue initially contains all vertices)
 $O(V)$ while $Q \neq \emptyset$ (while the queue is not empty)
do $u \leftarrow \text{mindistance}(Q, \text{dist})$ (select the element of Q with the min. distance)
 $S \leftarrow S \cup \{u\}$ (add u to list of visited vertices)
 for all $v \in \text{neighbors}[u]$
 do if $\text{dist}[v] > \text{dist}[u] + w(u, v)$ (if new shortest path found)
 then $d[v] \leftarrow d[u] + w(u, v)$ (set new value of shortest path)
 (if desired, add traceback code)
return dist

Edge Relaxation

O(V) $\text{dist}[s] \leftarrow 0$ (distance to source vertex is zero)

O(V) for all $v \in V - \{s\}$
do $\text{dist}[v] \leftarrow \infty$ (set all other distances to infinity)

$S \leftarrow \emptyset$ (S, the set of visited vertices is initially empty)

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O(V) while $Q \neq \emptyset$ (while the queue is not empty)
do $u \leftarrow \text{mindistance}(Q, \text{dist})$ (select the element of Q with the min. distance)
 $S \leftarrow S \cup \{u\}$ (add u to list of visited vertices)
for all $v \in \text{neighbors}[u]$
do if $\text{dist}[v] > \text{dist}[u] + w(u, v)$ (if new shortest path found)
then $\text{dist}[v] \leftarrow \text{dist}[u] + w(u, v)$ (set new value of shortest path)
(if desired, add traceback code)

return dist

Continue
looping
until queue
is empty

Edge Relaxation

Looking for the minDistance is $O(V)$

$O(V)$ $\text{dist}[s] \leftarrow 0$ (distance to source vertex is zero)

for all $v \in V - \{s\}$
do $\text{dist}[v] \leftarrow \infty$ (set all other distances to infinity)

$S \leftarrow \emptyset$ (S, the set of visited vertices is initially empty)

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$O(V)$ while $Q \neq \emptyset$ (while the queue is not empty)
do $u \leftarrow \text{mindistance}(Q, \text{dist})$ (select the element of Q with the min. distance)
 $S \leftarrow S \cup \{u\}$ (add u to list of visited vertices)
for all $v \in \text{neighbors}[u]$
do if $\text{dist}[v] > \text{dist}[u] + w(u, v)$ (if new shortest path found)
then $d[v] \leftarrow d[u] + w(u, v)$ (set new value of shortest path)
(if desired, add traceback code)

return dist

Edge Relaxation

$\text{dist}[s] \leftarrow 0$ (distance to source vertex is zero)

$O(V)$ for all $v \in V - \{s\}$
do $\text{dist}[v] \leftarrow \infty$ (set all other distances to infinity)

$S \leftarrow \emptyset$ (S , the set of visited vertices is initially empty)

$Q \leftarrow V$ (Q , the queue initially contains all vertices)

$O(V)$ while $Q \neq \emptyset$ ($O(V)$ while the queue is not empty)
do $u \leftarrow \text{mindistance}(Q, \text{dist})$ (select the element of Q with the min. distance)
 $S \leftarrow S \cup \{u\}$ (add u to list of visited vertices)

$O(E)$ for all $v \in \text{neighbors}[u]$
do if $\text{dist}[v] > \text{dist}[u] + w(u, v)$ (if new shortest path found)
then $d[v] \leftarrow d[u] + w(u, v)$ (set new value of shortest path)
(if desired, add traceback code)

return dist

Total time $O(|V|^2 + |E|) = O(|V|^2)$

$O(V)$

$\text{dist}[s] \leftarrow 0$

(distance to source vertex is zero)

for all $v \in V - \{s\}$

do $\text{dist}[v] \leftarrow \infty$

(set all other distances to infinity)

$S \leftarrow \emptyset$

(S, the set of visited vertices is initially empty)

$Q \leftarrow V$

(Q, the queue initially contains all vertices)

while $Q \neq \emptyset$

(while the queue is not empty)

do $u \leftarrow \text{mindistance}(Q, \text{dist})$

(select the element of Q with the min. distance)

$S \leftarrow S \cup \{u\}$

(add u to list of visited vertices)

for all $v \in \text{neighbors}[u]$

do if $\text{dist}[v] > \text{dist}[u] + w(u, v)$

(if new shortest path found)

then $d[v] \leftarrow d[u] + w(u, v)$

(set new value of shortest path)

(if desired, add traceback code)

return dist

Looking for the minDistance is $O(\log(V))$
using removeMin()

$O(V)$ $\text{dist}[s] \leftarrow 0$ (distance to source vertex is zero)
 $\text{for all } v \in V - \{s\}$
 $\text{do } \text{dist}[v] \leftarrow \infty$ (set all other distances to infinity)
 $S \leftarrow \emptyset$ (S, the set of visited vertices is initially empty)
 $Q \leftarrow V$ (Q, the queue initially contains all vertices)
 $O(V)$ $\text{while } Q \neq \emptyset$ (while the queue is not empty)
 $\text{do } u \leftarrow \text{mindistance}(Q, \text{dist})$ (select the element of Q with the min. distance)
 $S \leftarrow S \cup \{u\}$ (add u to list of visited vertices)
 $\text{for all } v \in \text{neighbors}[u]$
 $\text{do if } \text{dist}[v] > \text{dist}[u] + w(u, v)$ (if new shortest path found)
 $\text{then } \text{dist}[v] \leftarrow \text{dist}[u] + w(u, v)$ (set new value of shortest path)
 (if desired, add traceback code)

return dist

Edge Relaxation

$\text{dist}[s] \leftarrow 0$ (distance to source vertex is zero)

$O(V)$ for all $v \in V - \{s\}$
do $\text{dist}[v] \leftarrow \infty$ (set all other distances to infinity)

$S \leftarrow \emptyset$ (S , the set of visited vertices is initially empty)

$Q \leftarrow V$ (Q , the queue initially contains all vertices)

$O(V)$ while $Q \neq \emptyset$ ($O(\log V)$ (while the queue is not empty))
do $u \leftarrow \text{mindistance}(Q, \text{dist})$ (select the element of Q with the min. distance)
 $S \leftarrow S \cup \{u\}$ (add u to list of visited vertices)

$O(E)$ for all $v \in \text{neighbors}[u]$
do if $\text{dist}[v] > \text{dist}[u] + w(u, v)$ (if new shortest path found)
then $\text{dist}[v] \leftarrow \text{dist}[u] + w(u, v)$ (set new value of shortest path)
(if desired, add traceback code)

return dist

Edge Relaxation

$\text{dist}[s] \leftarrow 0$ (distance to source vertex is zero)

$O(V)$ for all $v \in V - \{s\}$
do $\text{dist}[v] \leftarrow \infty$ (set all other distances to infinity)

$S \leftarrow \emptyset$ (S , the set of visited vertices is initially empty)

$Q \leftarrow V$ (Q , the queue initially contains all vertices)

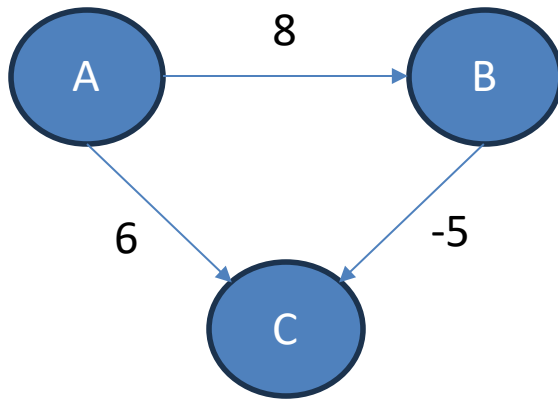
$O(V)$ while $Q \neq \emptyset$ $O(\log V)$ (while the queue is not empty)
do $u \leftarrow \text{mindistance}(Q, \text{dist})$ (select the element of Q with the min. distance)
 $S \leftarrow S \cup \{u\}$ (add u to list of visited vertices)
for all $v \in \text{neighbors}[u]$
do if $\text{dist}[v] > \text{dist}[u] + w(u, v)$ (if new shortest path found)
then $d[v] \leftarrow d[u] + w(u, v)$ (set new value of shortest path)
 $O(\log V)$ (if desired, add traceback code)

return dist

Total time $O(V \log V + E \log V) = O((V+E) \log V)$

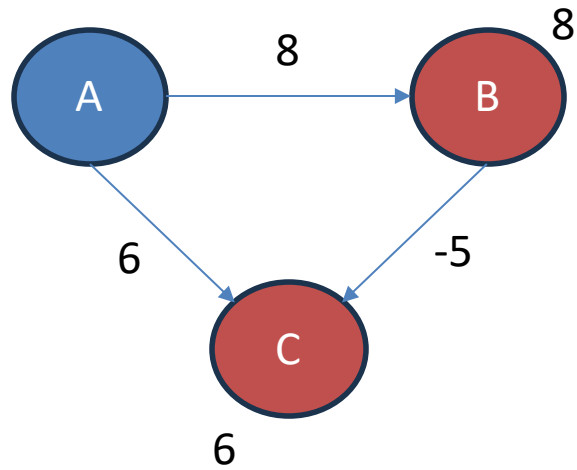
Bellman-Ford

Why can't Dijkstra handle negative weights?

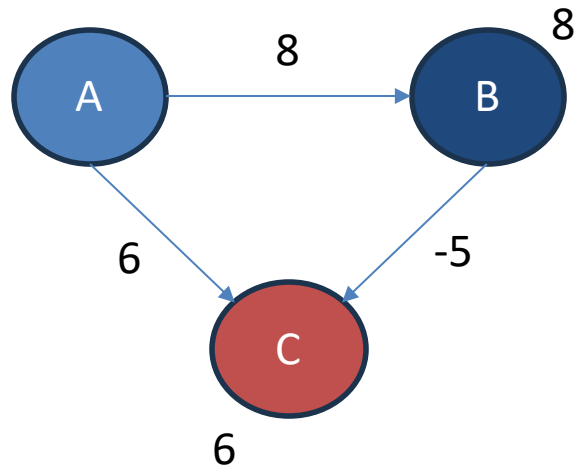


Mark A as visited

Why can't Dijkstra handle negative weights?

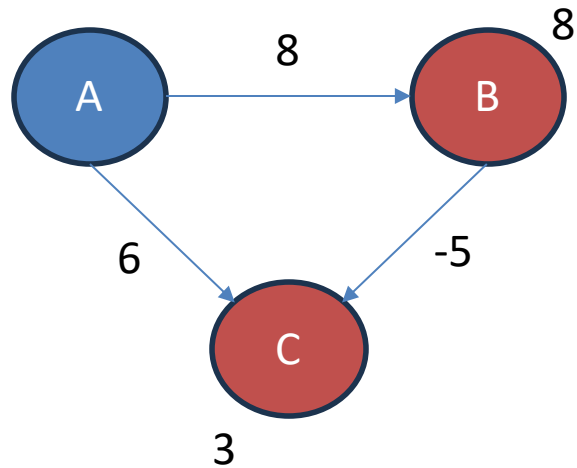


Why can't Dijkstra handle negative weights?



Mark C as Visited

Why can't Dijkstra handle negative weights?



Mark B as Visited

$$8 + (-5) = 3$$

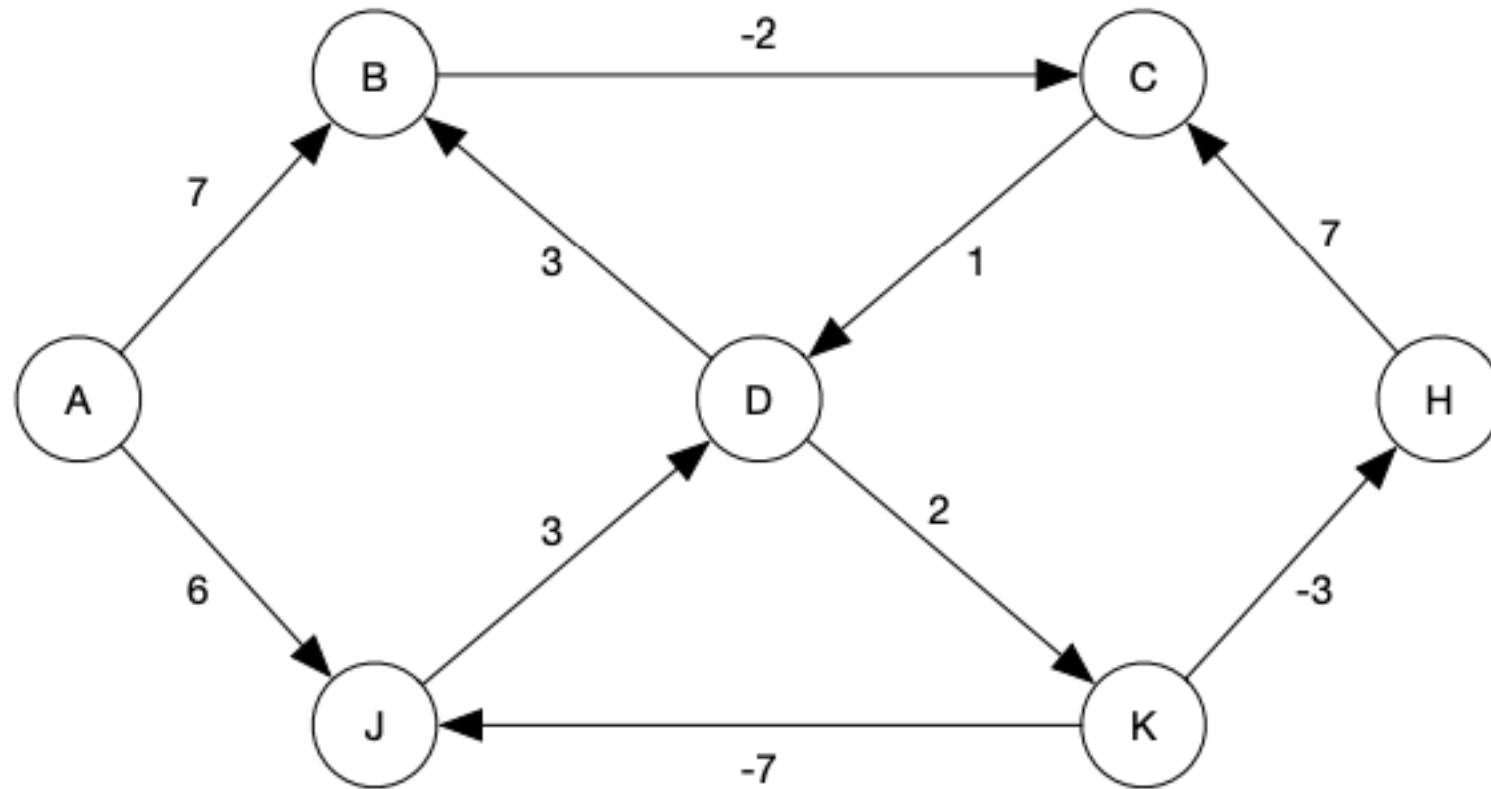
Should be a shorter but will not be reconsidered anymore

Bellman-Ford

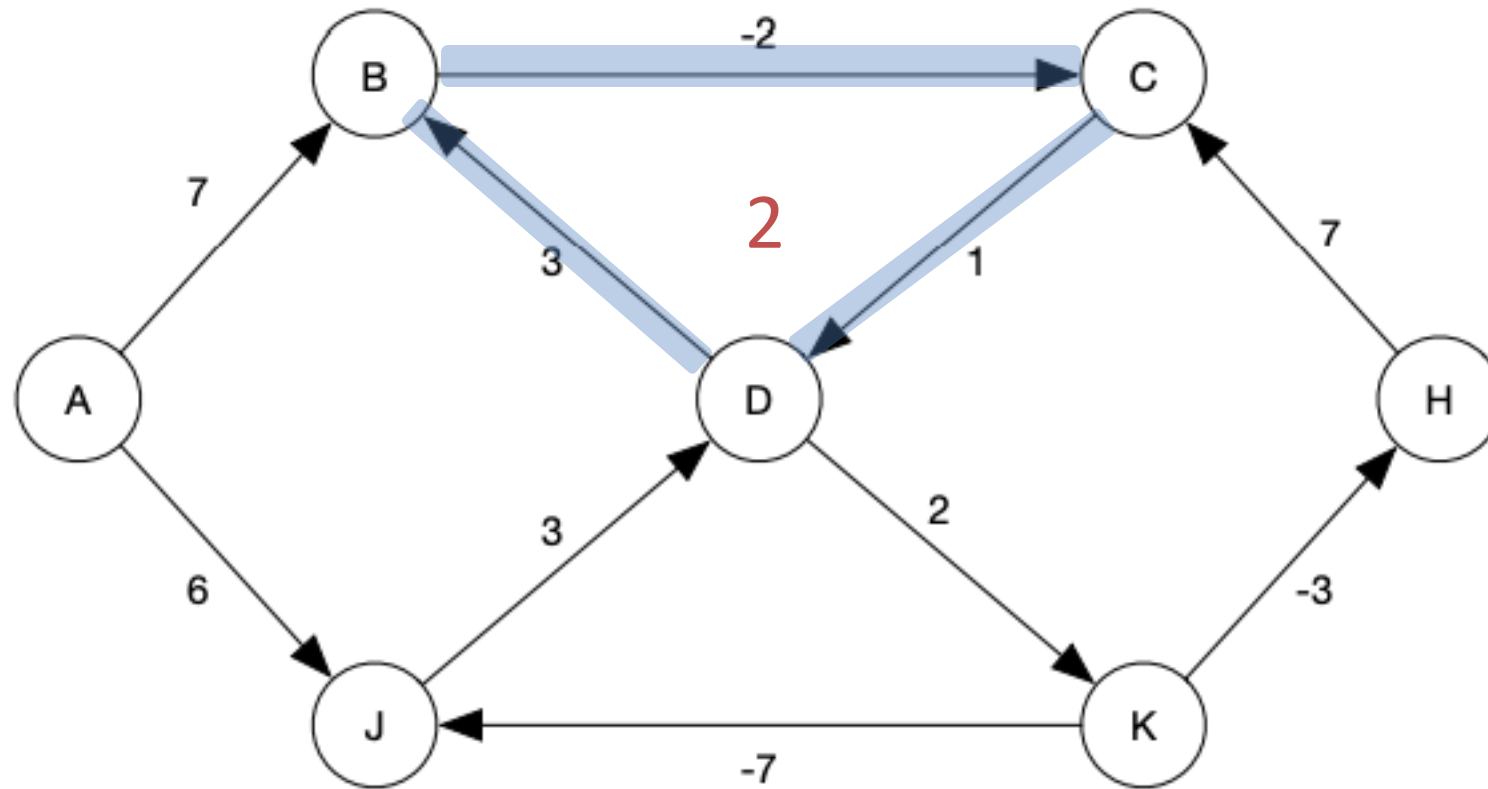
- Given some graph, $G = (V, E)$, and some starting node $S \in V$, the BellmanFord algorithm will find the shortest paths (or paths with minimum weight) from S to all other nodes in V . Note that G must not contain any negative weight cycles.

What is a negative cycle?

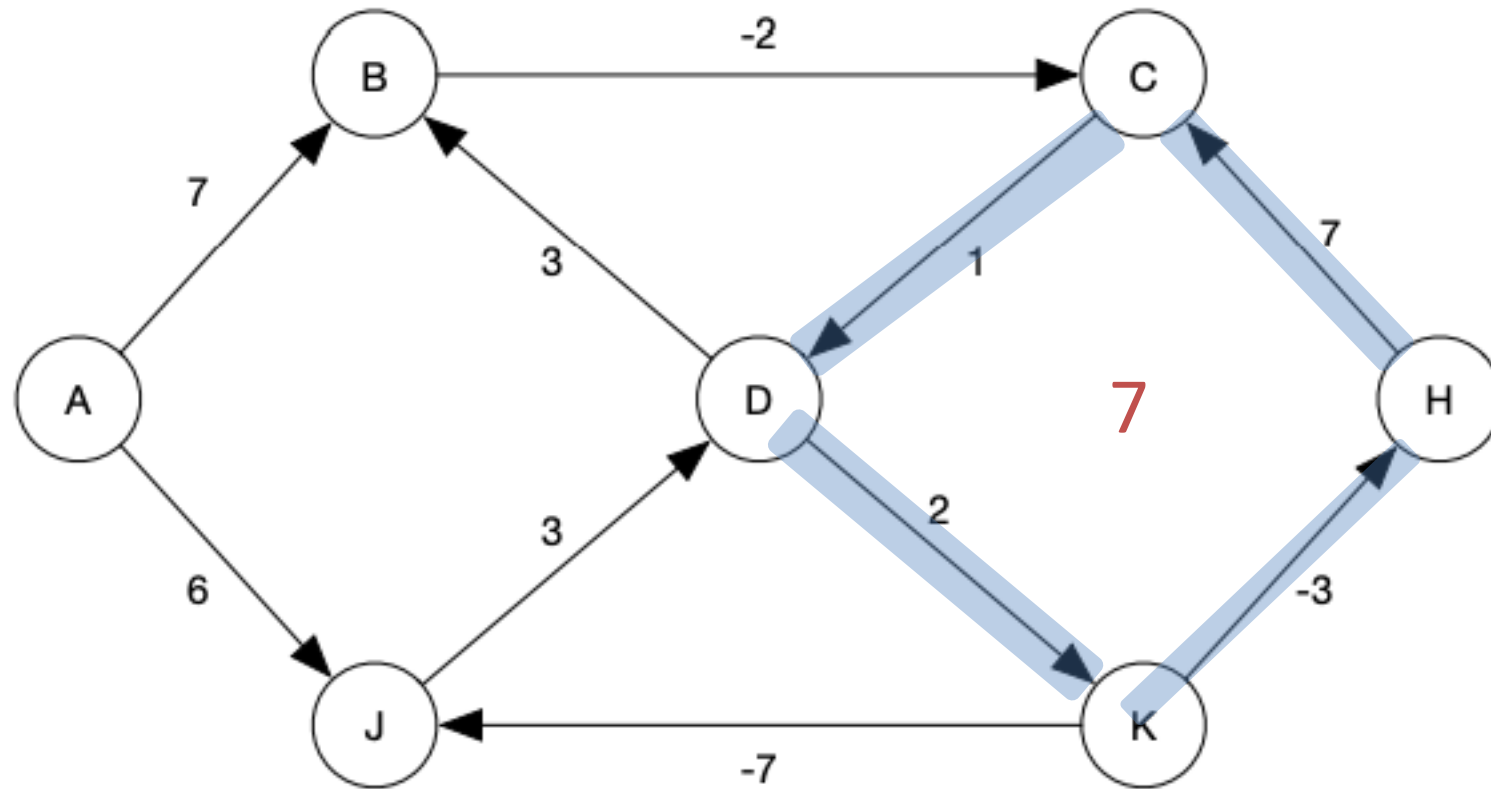
What is a negative cycle?



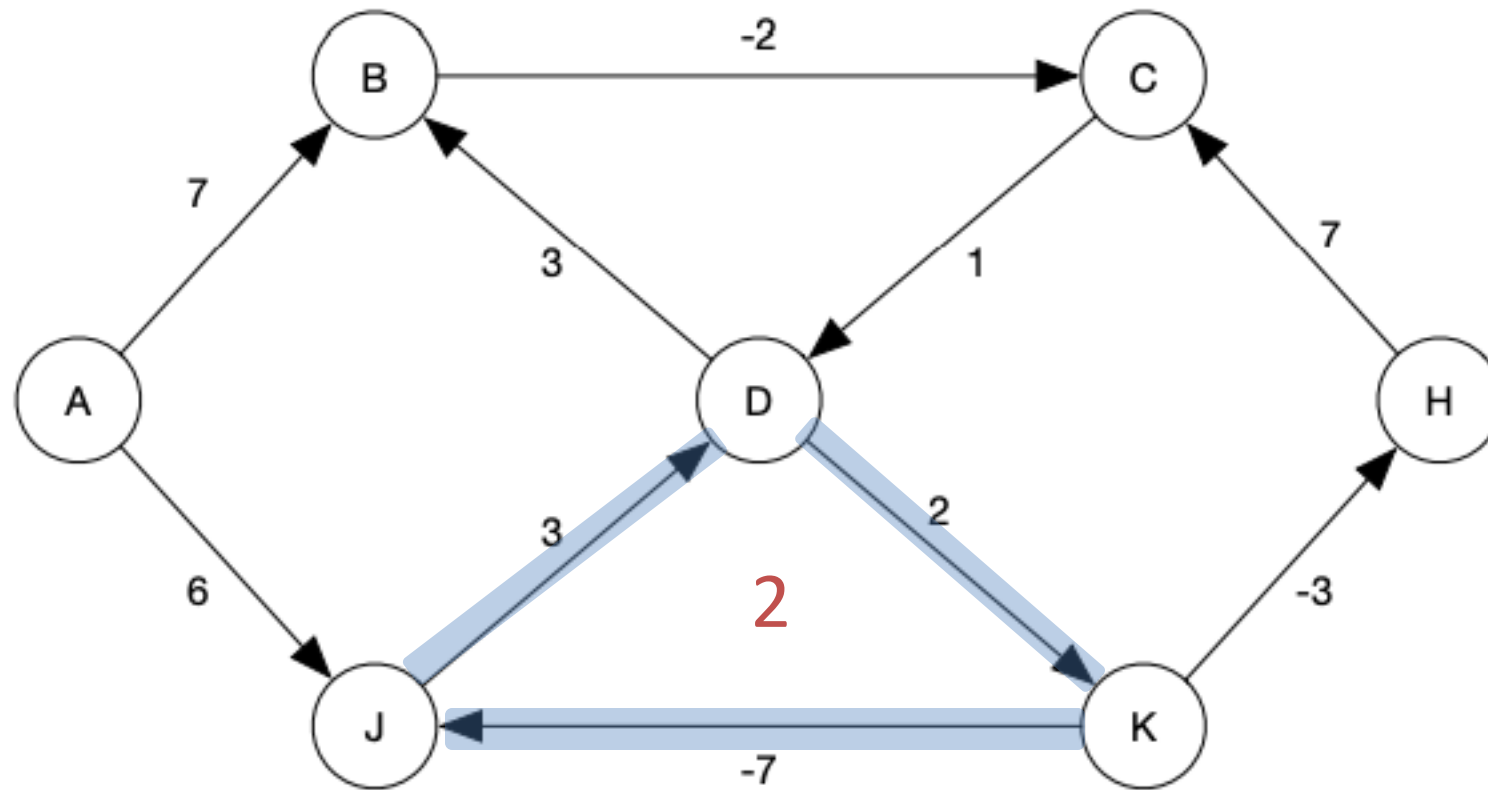
What is a negative cycle?



What is a negative cycle?



What is a negative cycle?



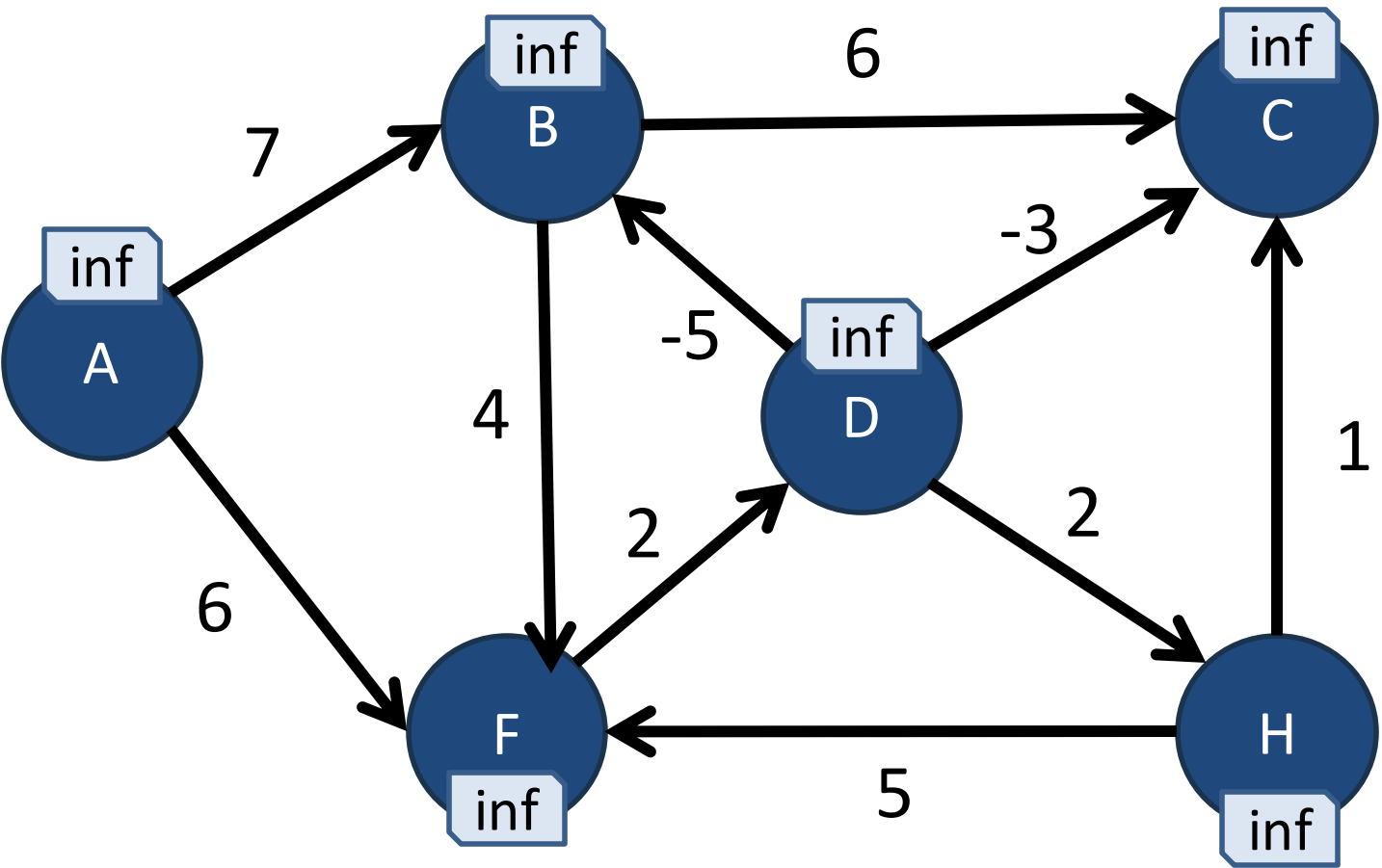
No Well-Defined Shortest Path

- Because you can keep lowering the total cost indefinitely, there is **no "shortest" path** to nodes reachable through the negative cycle — the cost can be made arbitrarily negative.

Bellman-Ford algorithm

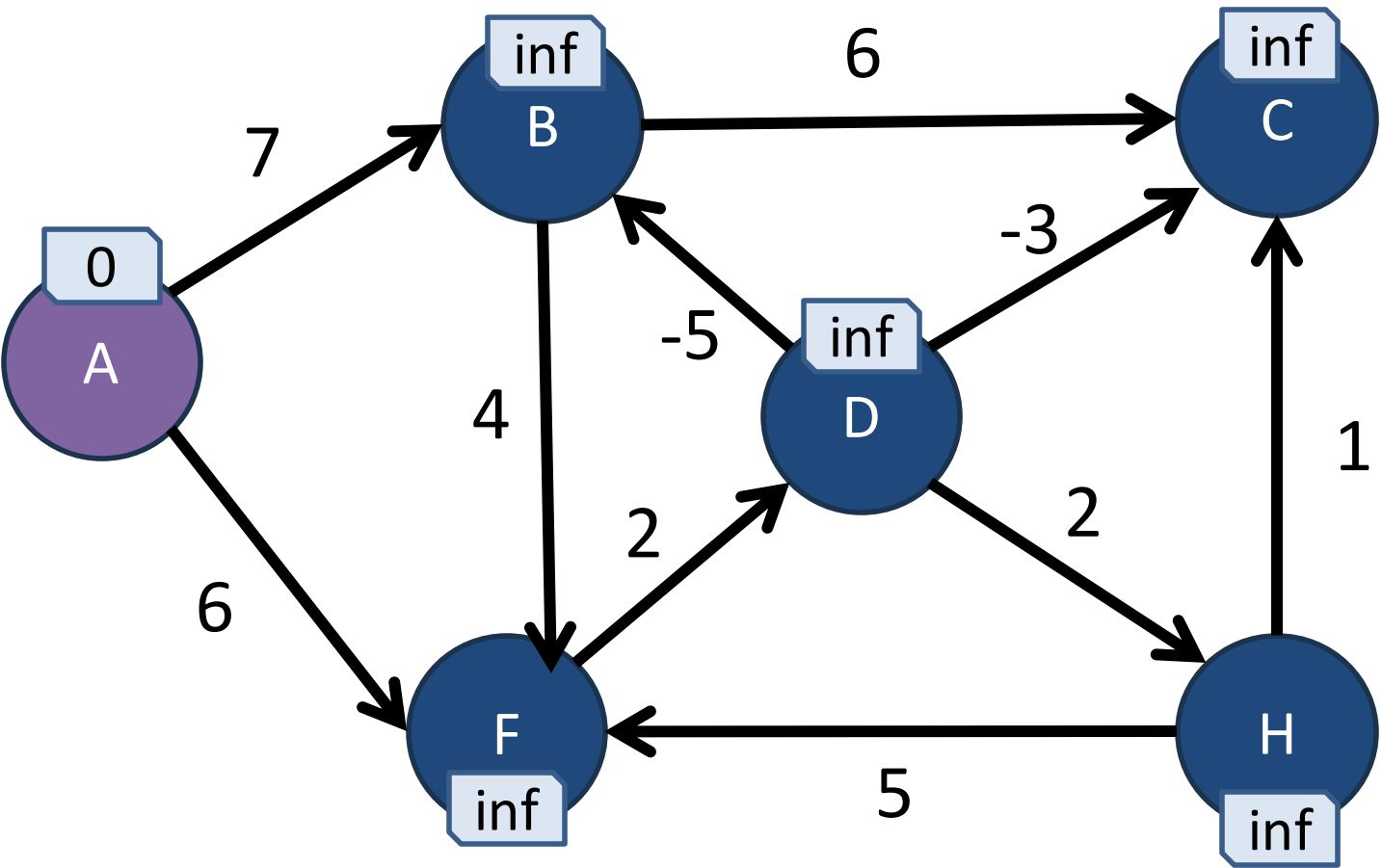
BELLMAN-FORD(G, s)

```
1  for all  $v \in V$ 
2       $dist[v] \leftarrow \infty$ 
3       $prev[v] \leftarrow null$ 
4   $dist[s] \leftarrow 0$ 
5  for  $i \leftarrow 1$  to  $|V| - 1$ 
6      for all edges  $(u, v) \in E$ 
7          if  $dist[v] > dist[u] + w(u, v)$ 
8               $dist[v] \leftarrow dist[u] + w(u, v)$ 
9               $prev[v] \leftarrow u$ 
10 for all edges  $(u, v) \in E$ 
11     if  $dist[v] > dist[u] + w(u, v)$ 
12         return false
```



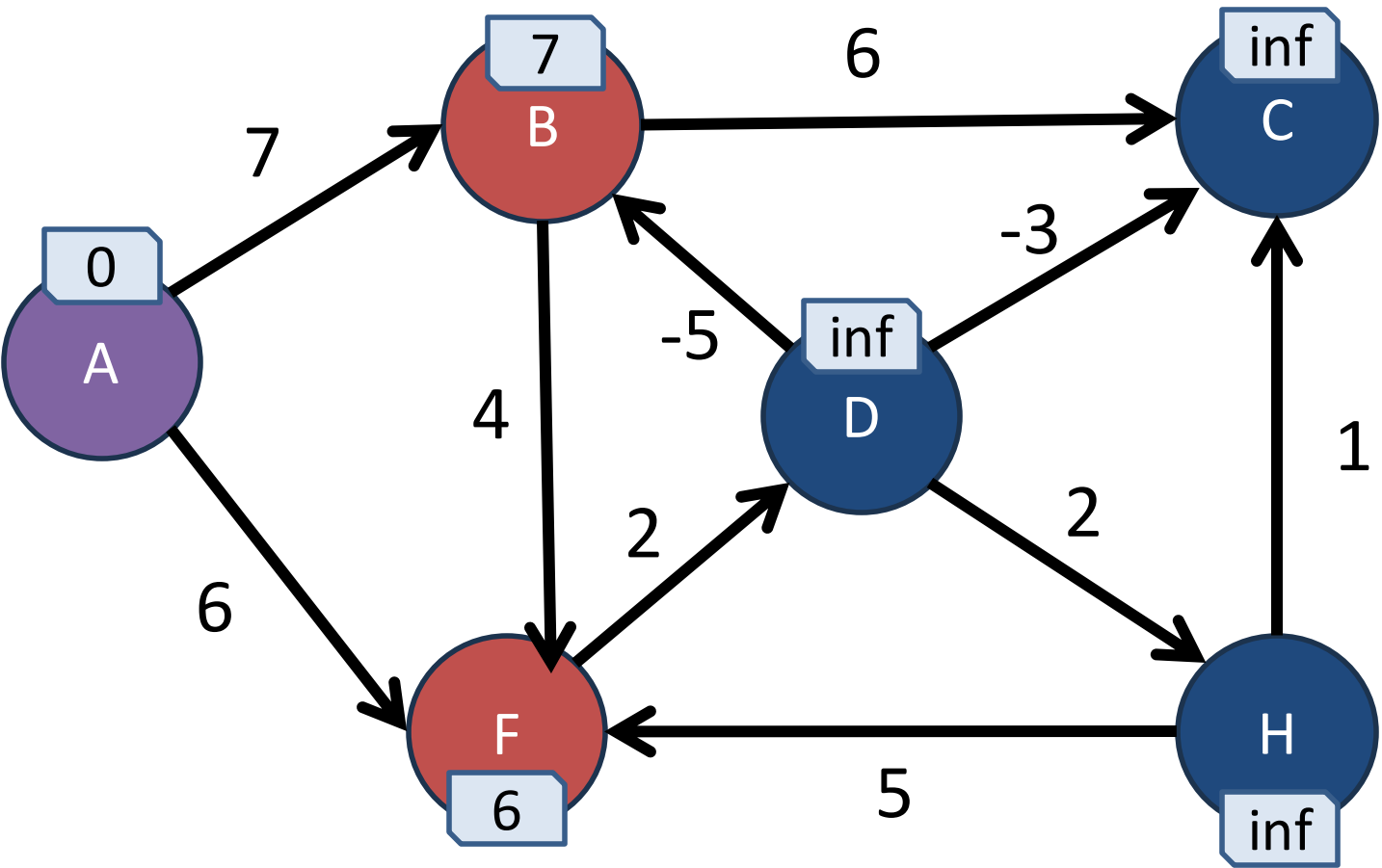
	Cost	Previous
A		
B		
C		
D		
F		
H		

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



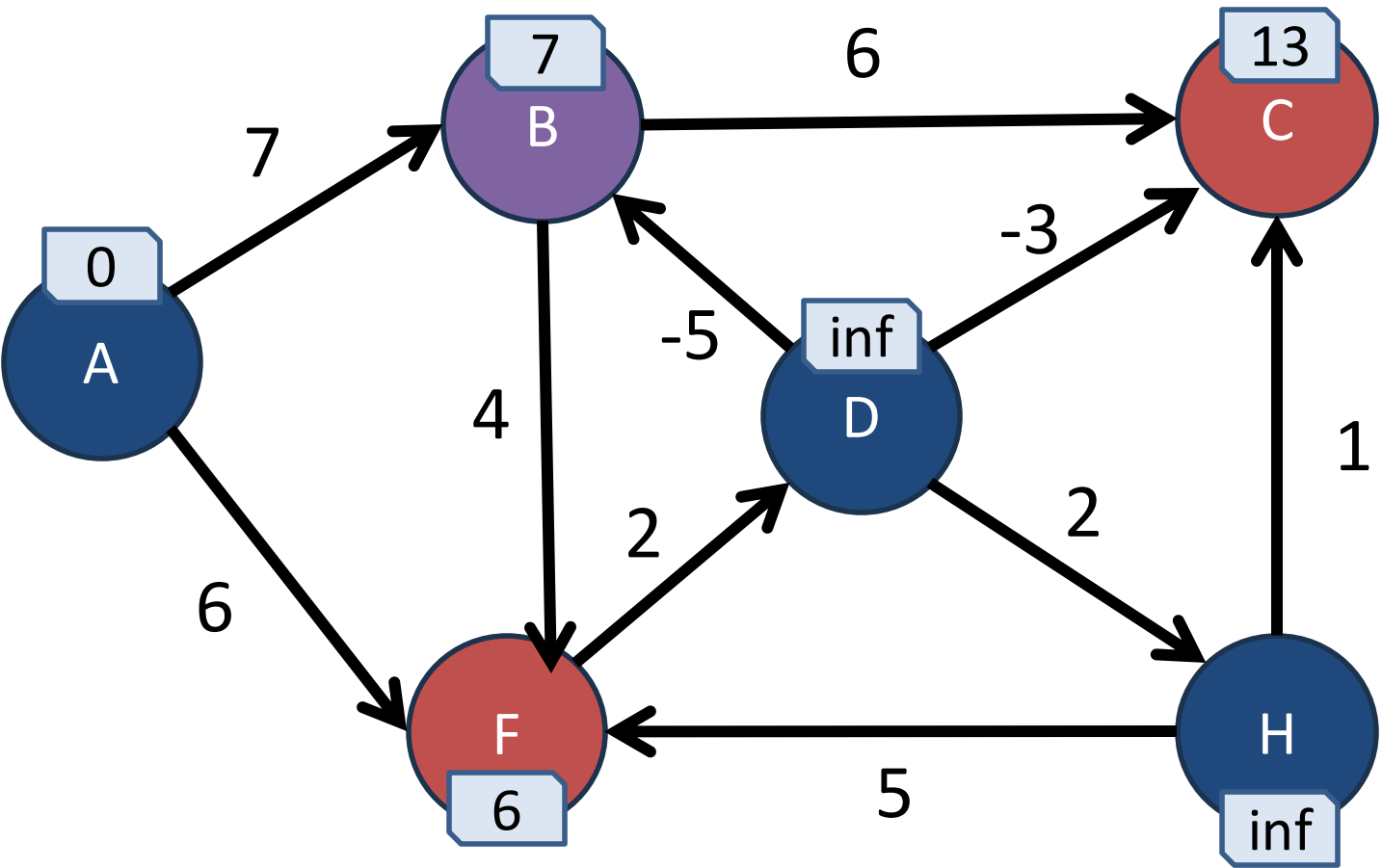
	Cost	Previous
A	0	
B		
C		
D		
F		
H		

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



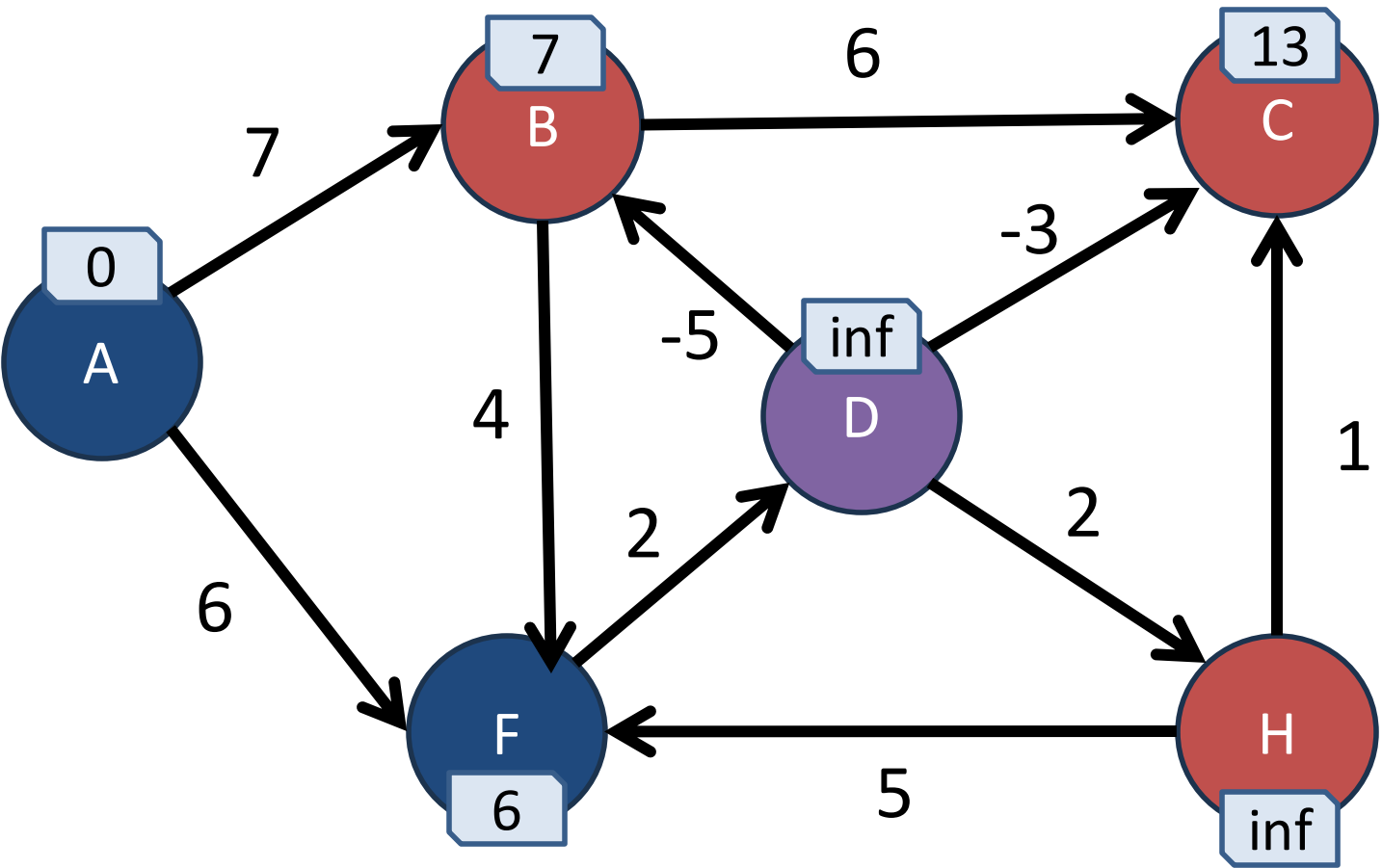
	Cost	Previous
A	0	
B	7	A
C		
D		
F	6	A
H		

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



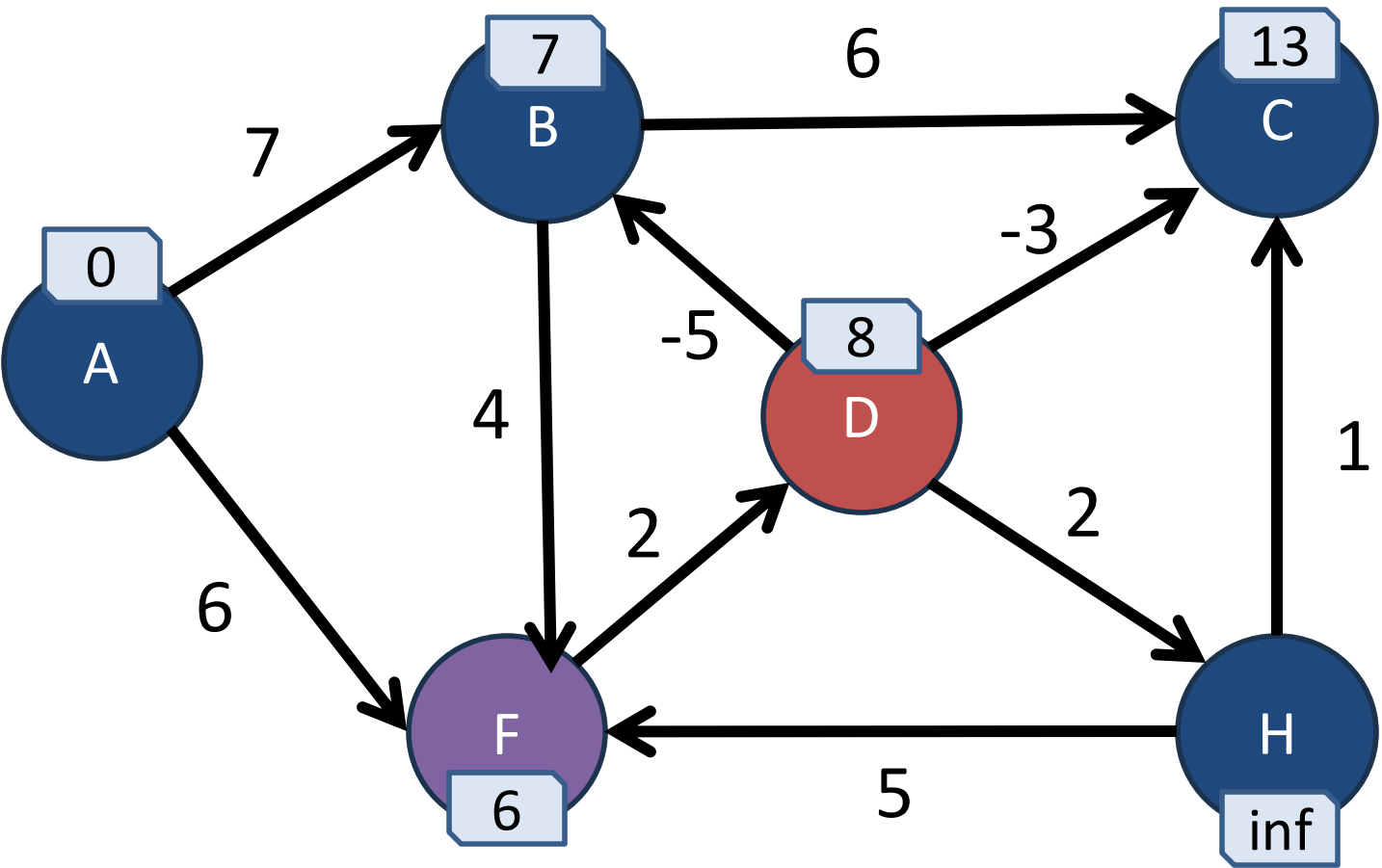
	Cost	Previous
A	0	
B	7	A
C	13	B
D		
F	6	A
H		

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



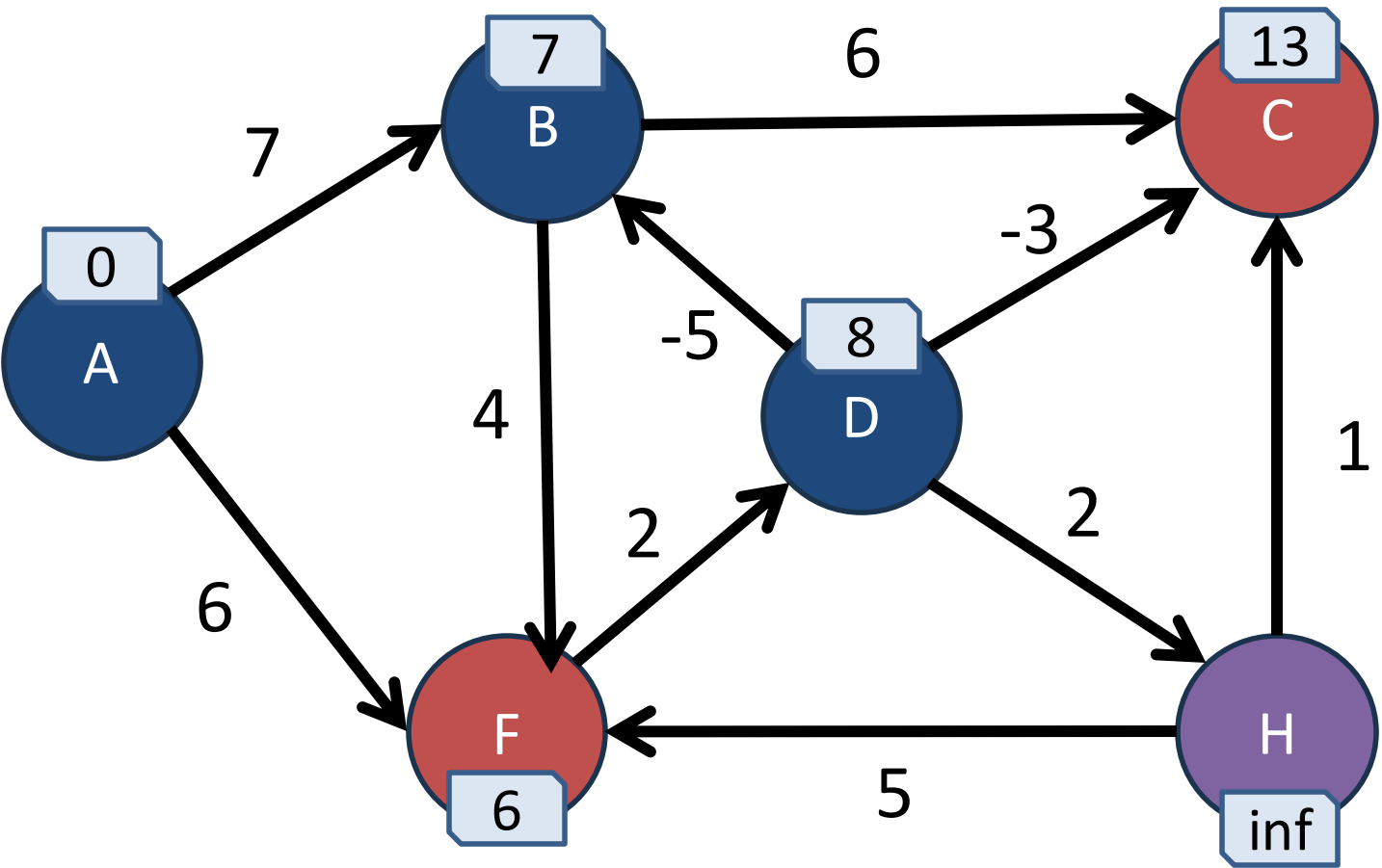
	Cost	Previous
A	0	
B	7	A
C	13	B
D		
F	6	A
H		

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



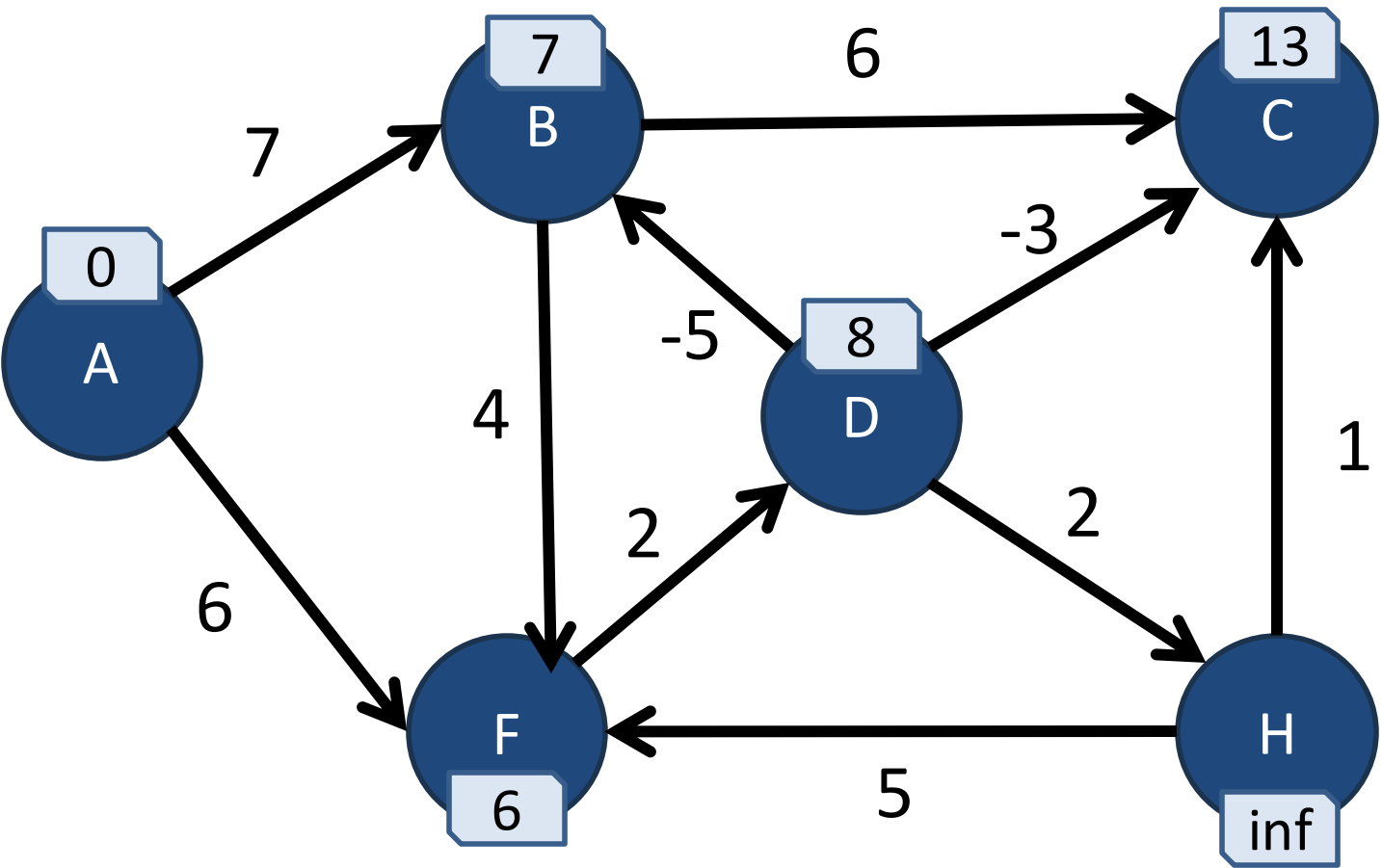
	Cost	Previous
A	0	
B	7	A
C	13	B
D	8	F
F	6	A
H		

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	7	A
C	13	B
D	8	F
F	6	A
H		

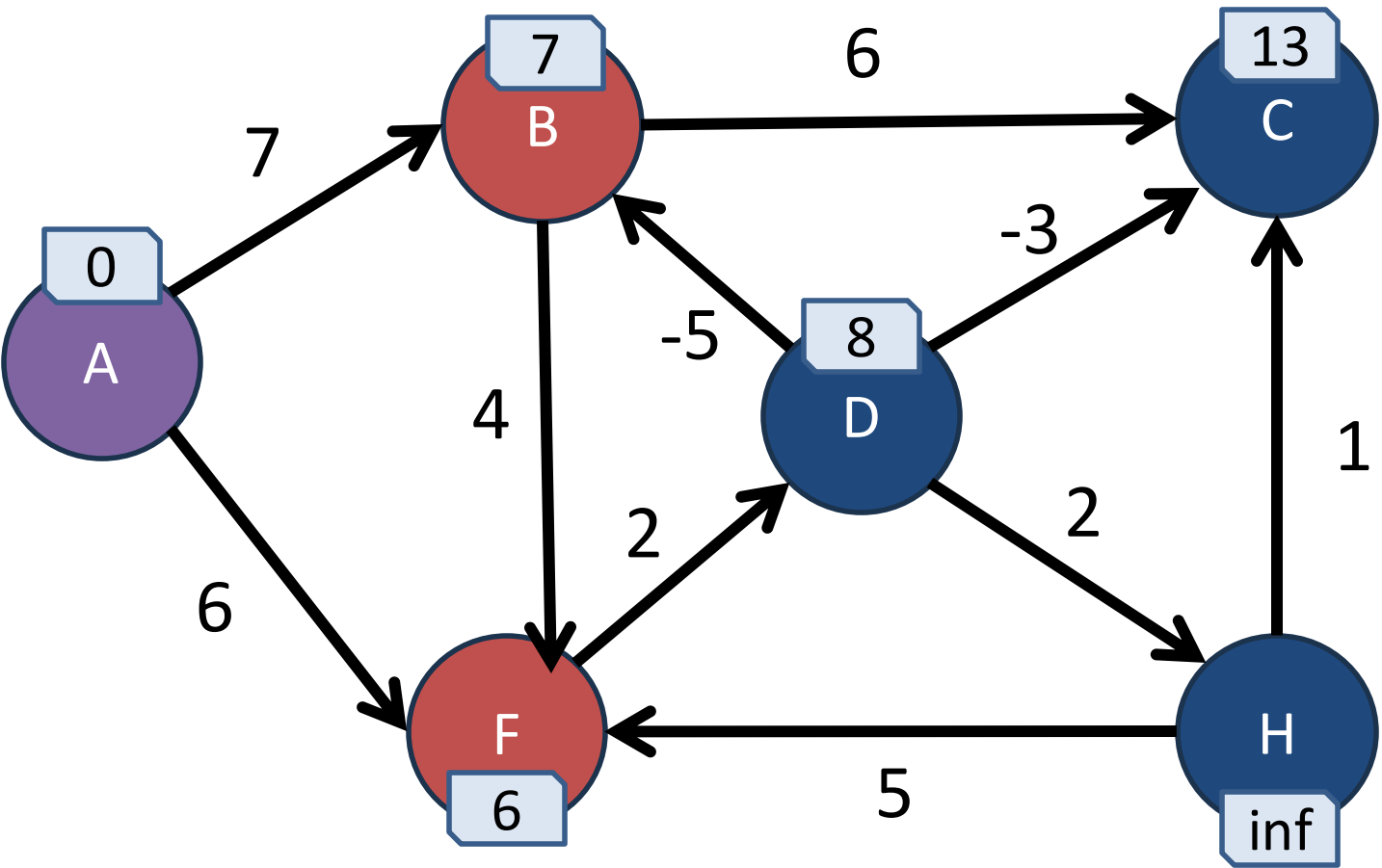
- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	7	A
C	13	B
D	8	F
F	6	A
H		

Second Iteration

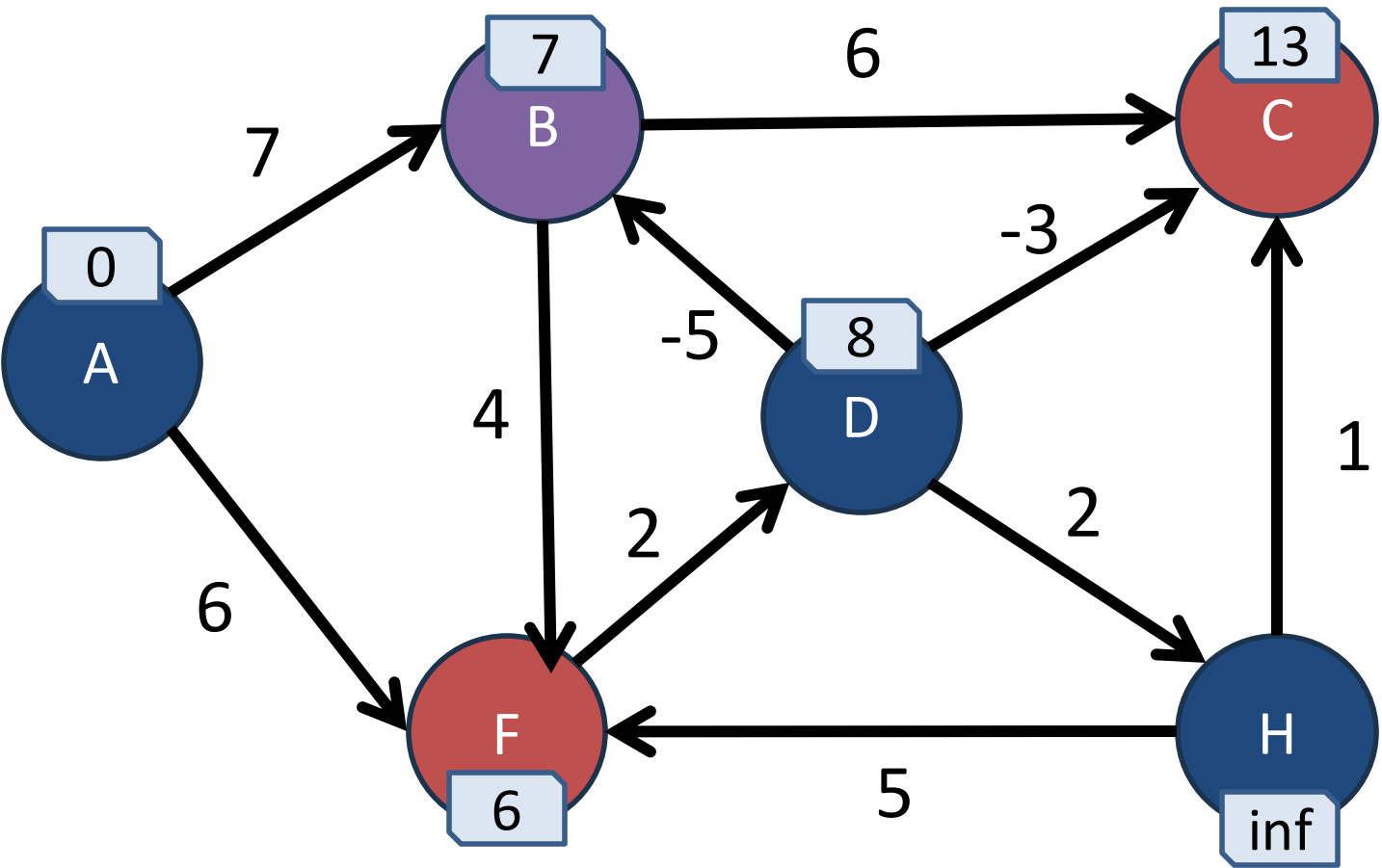
- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	7	A
C	13	B
D	8	F
F	6	A
H		

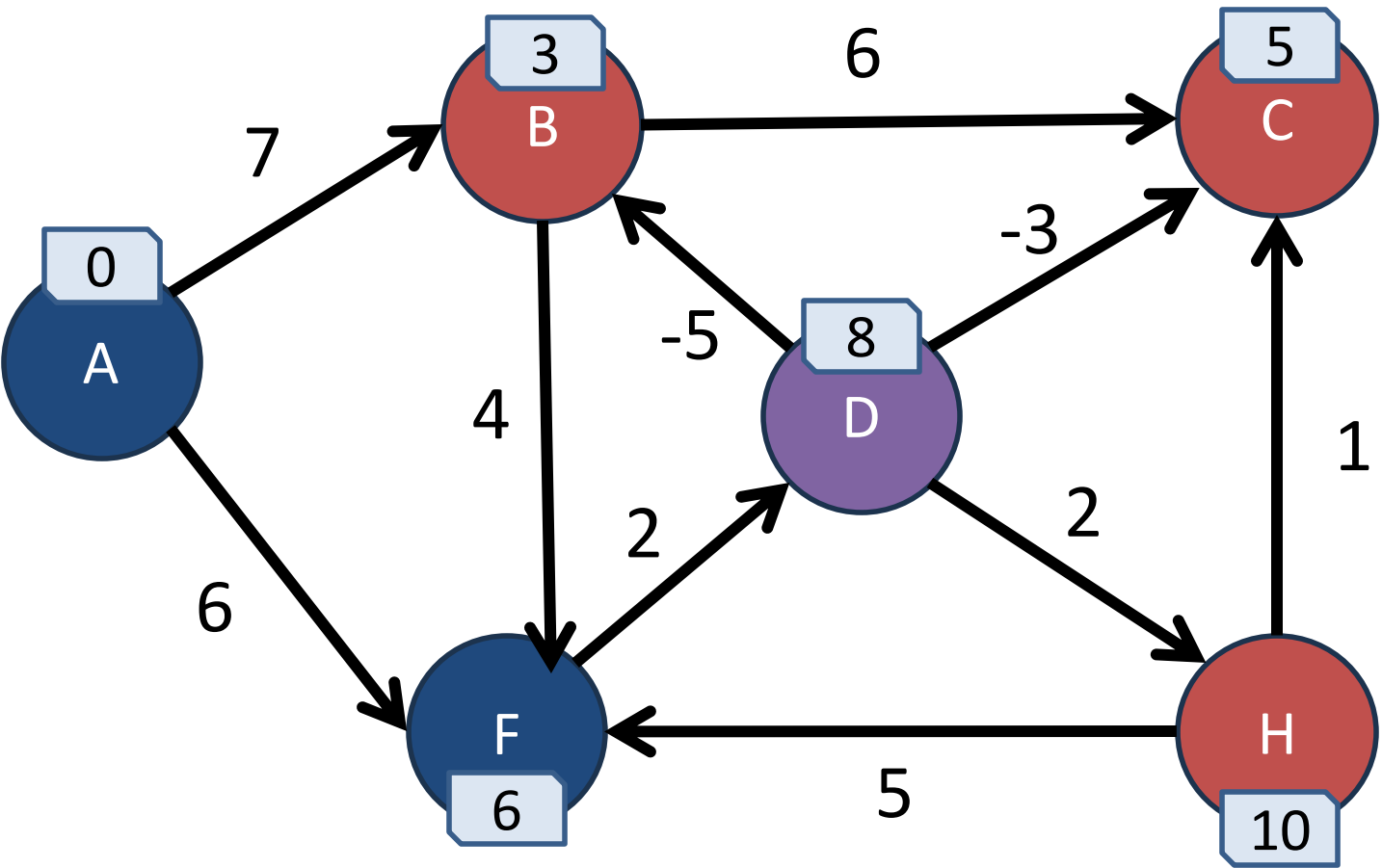
Second Iteration

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



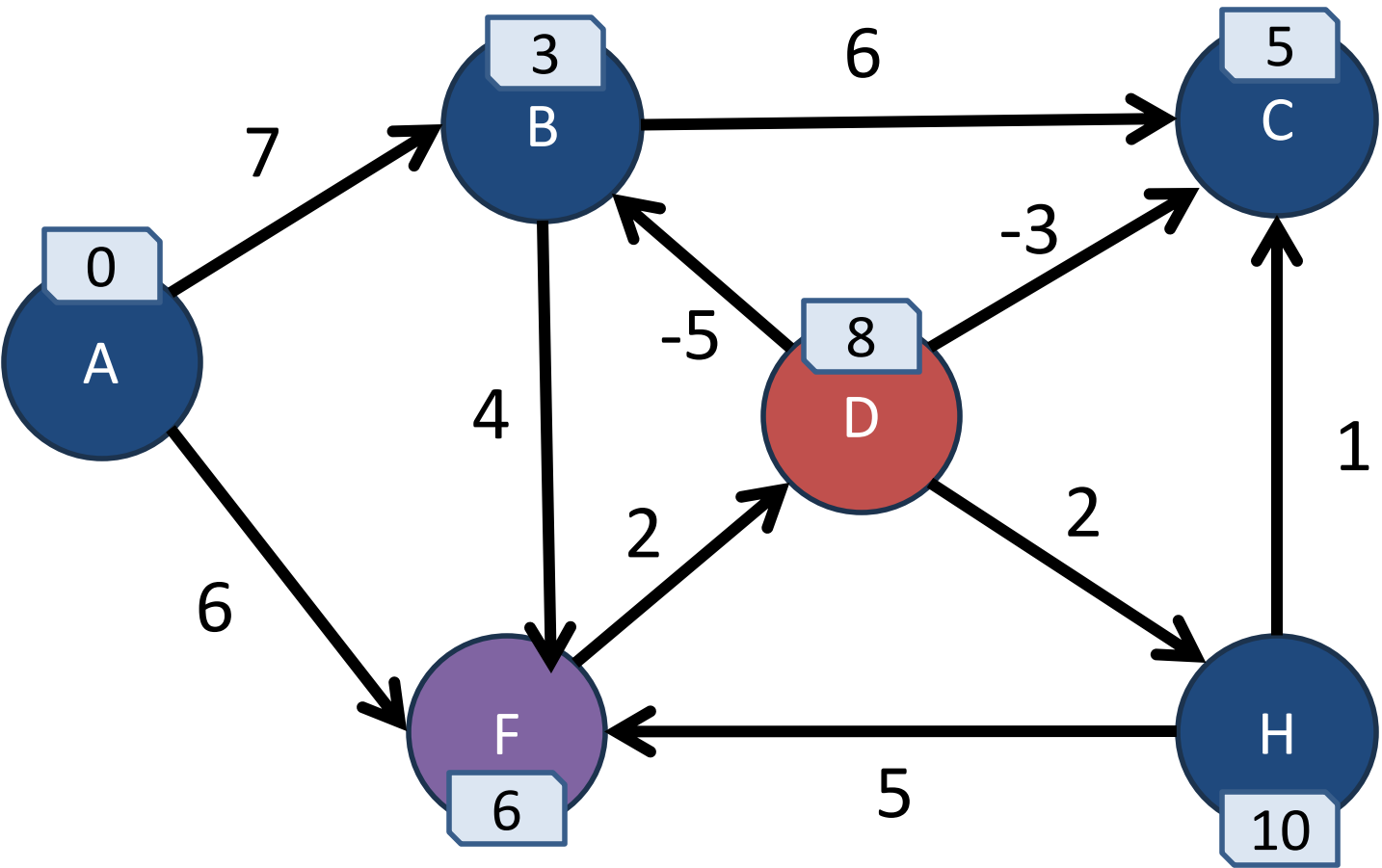
	Cost	Previous
A	0	
B	7	A
C	13	B
D	8	F
F	6	A
H		

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



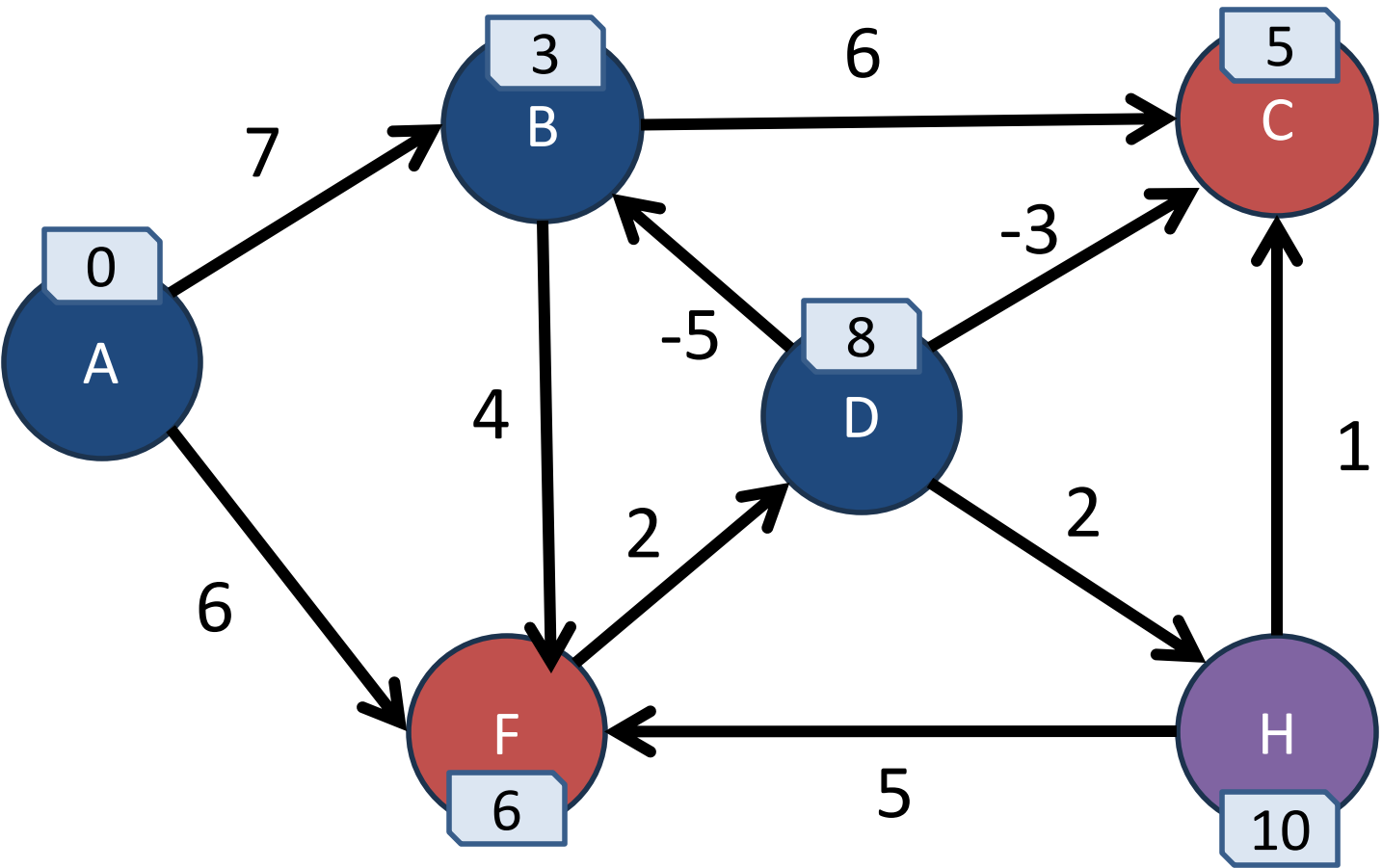
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D





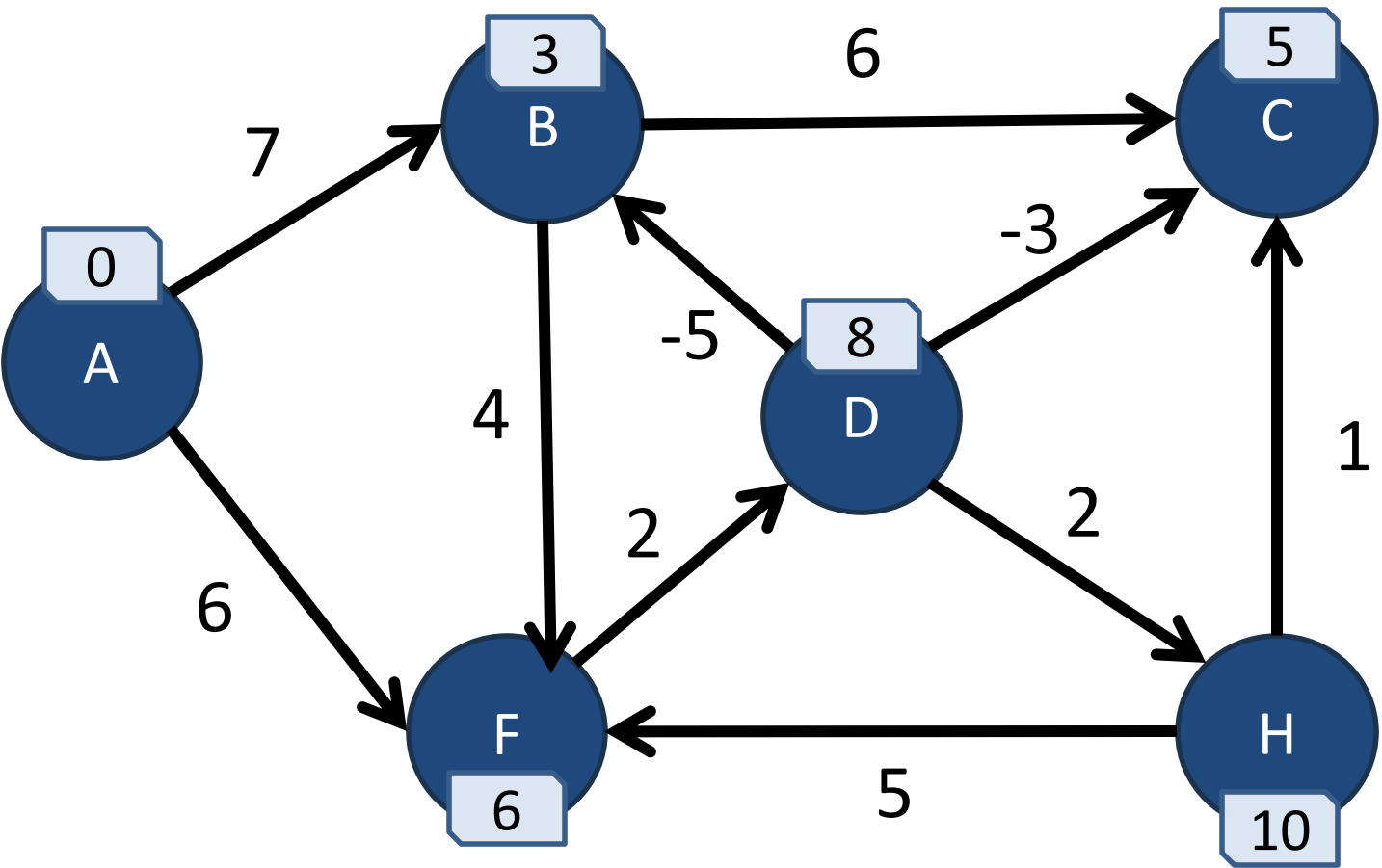
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

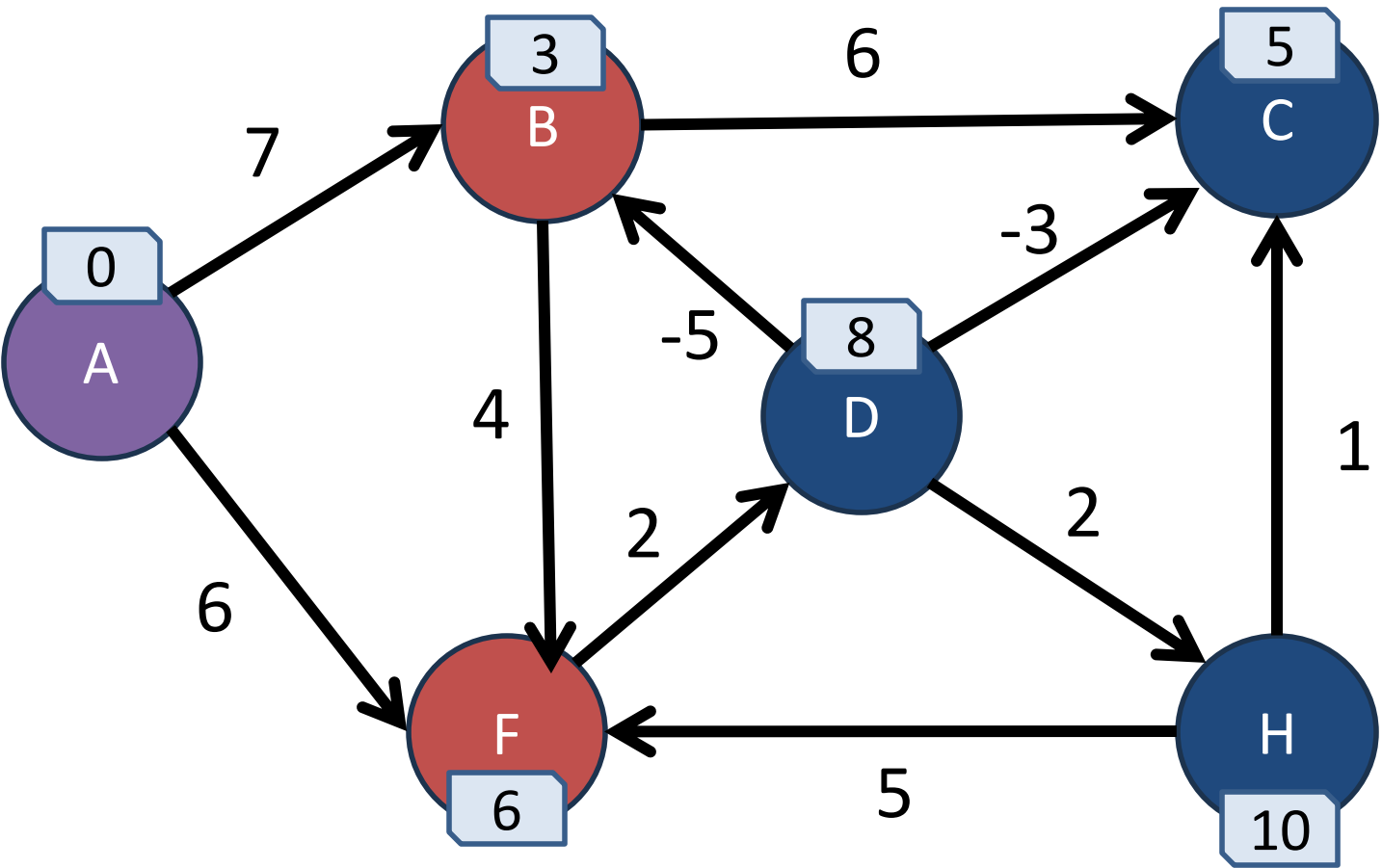
- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

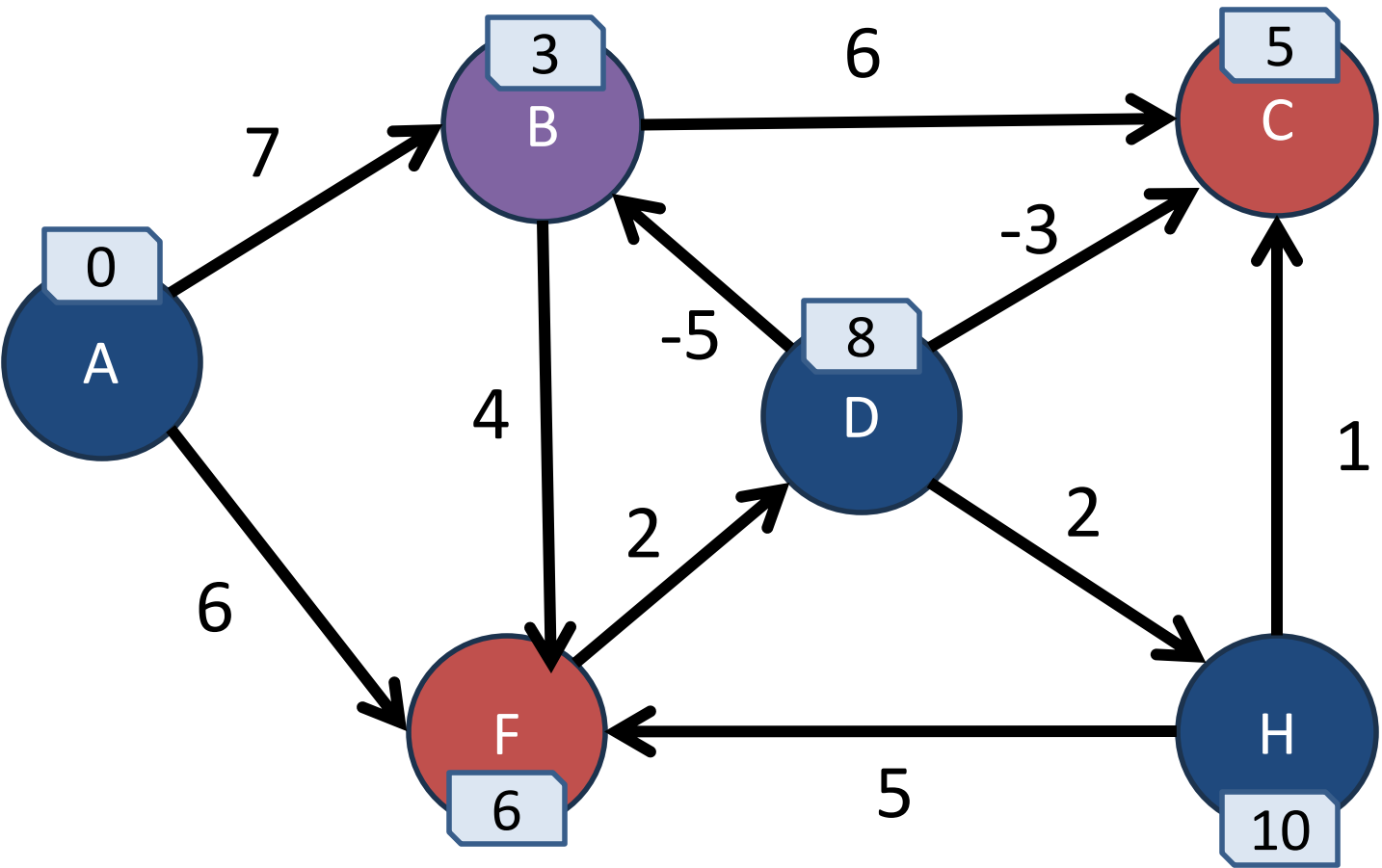
AB	AF	BC	BF	DB	DC	DH	FD	HF	HC
----	----	----	----	----	----	----	----	----	----

Third Iteration



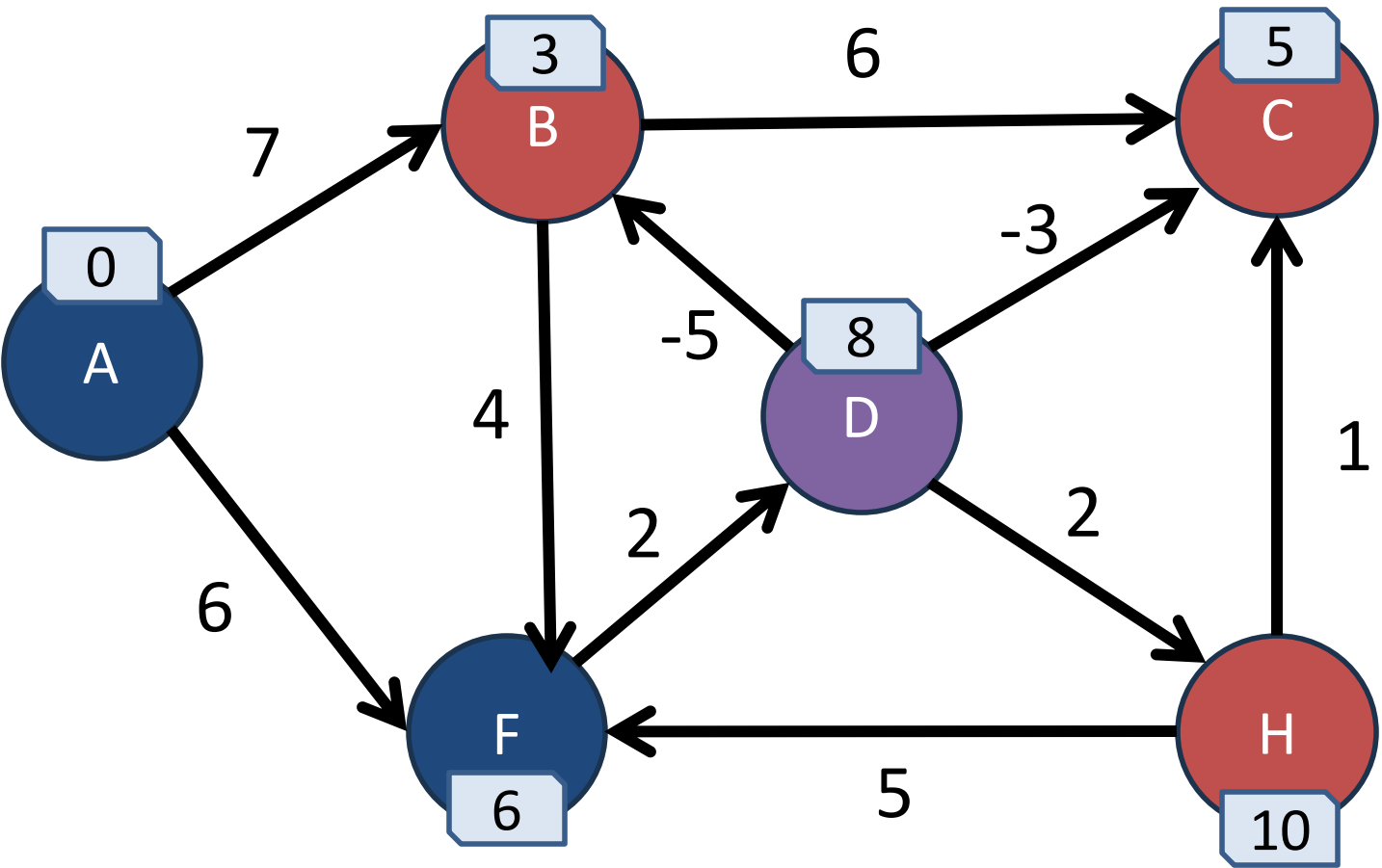
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



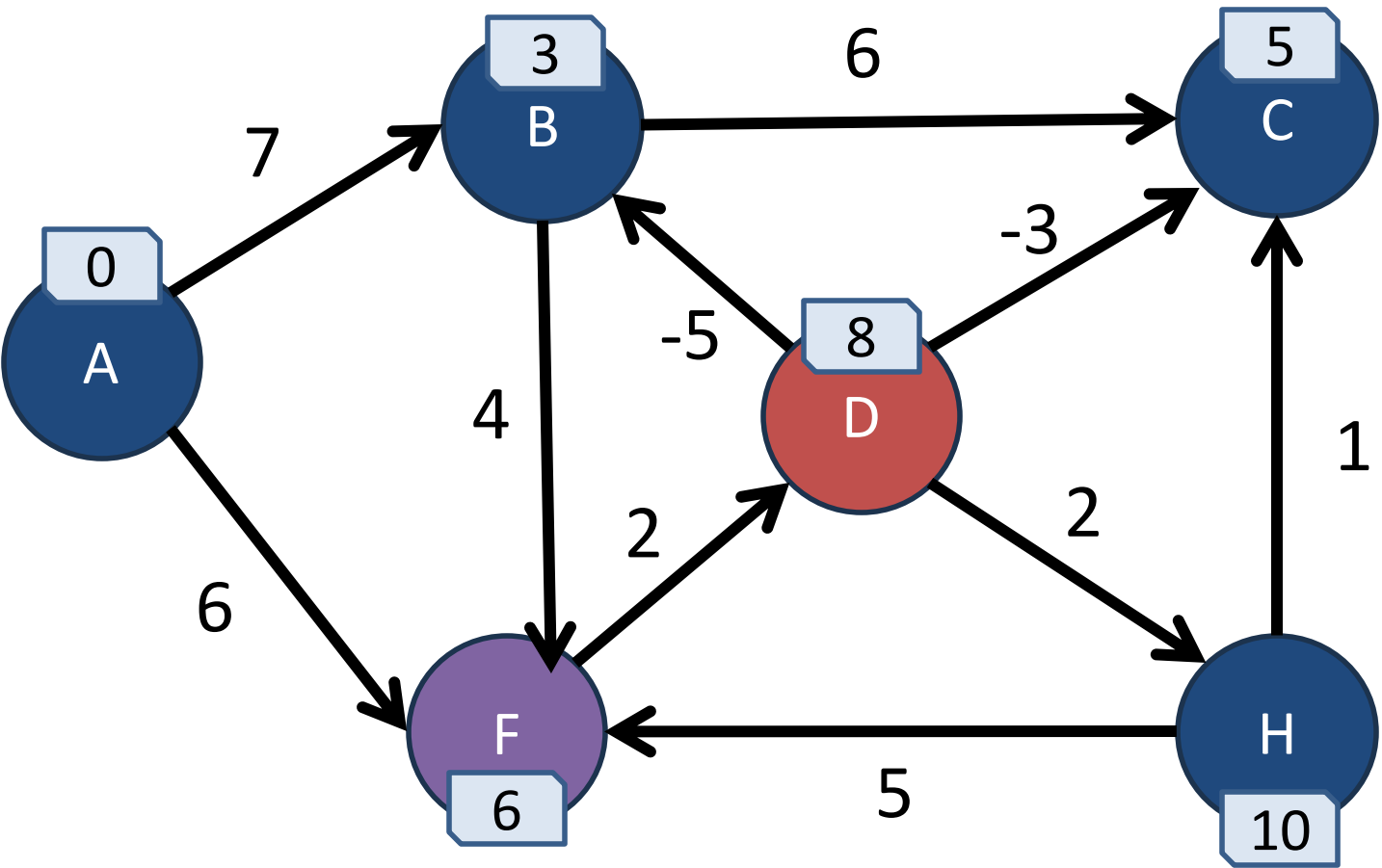
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



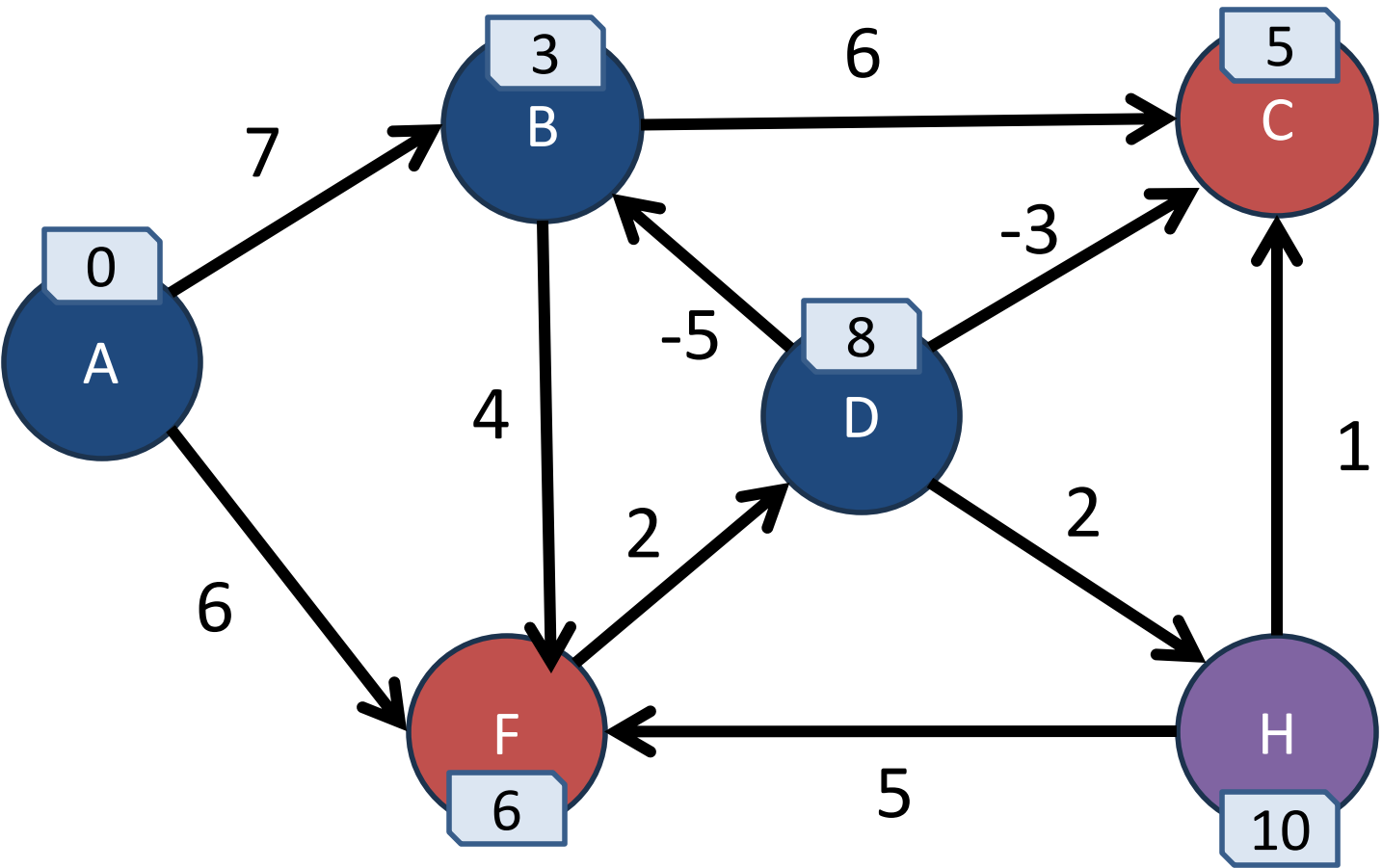
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



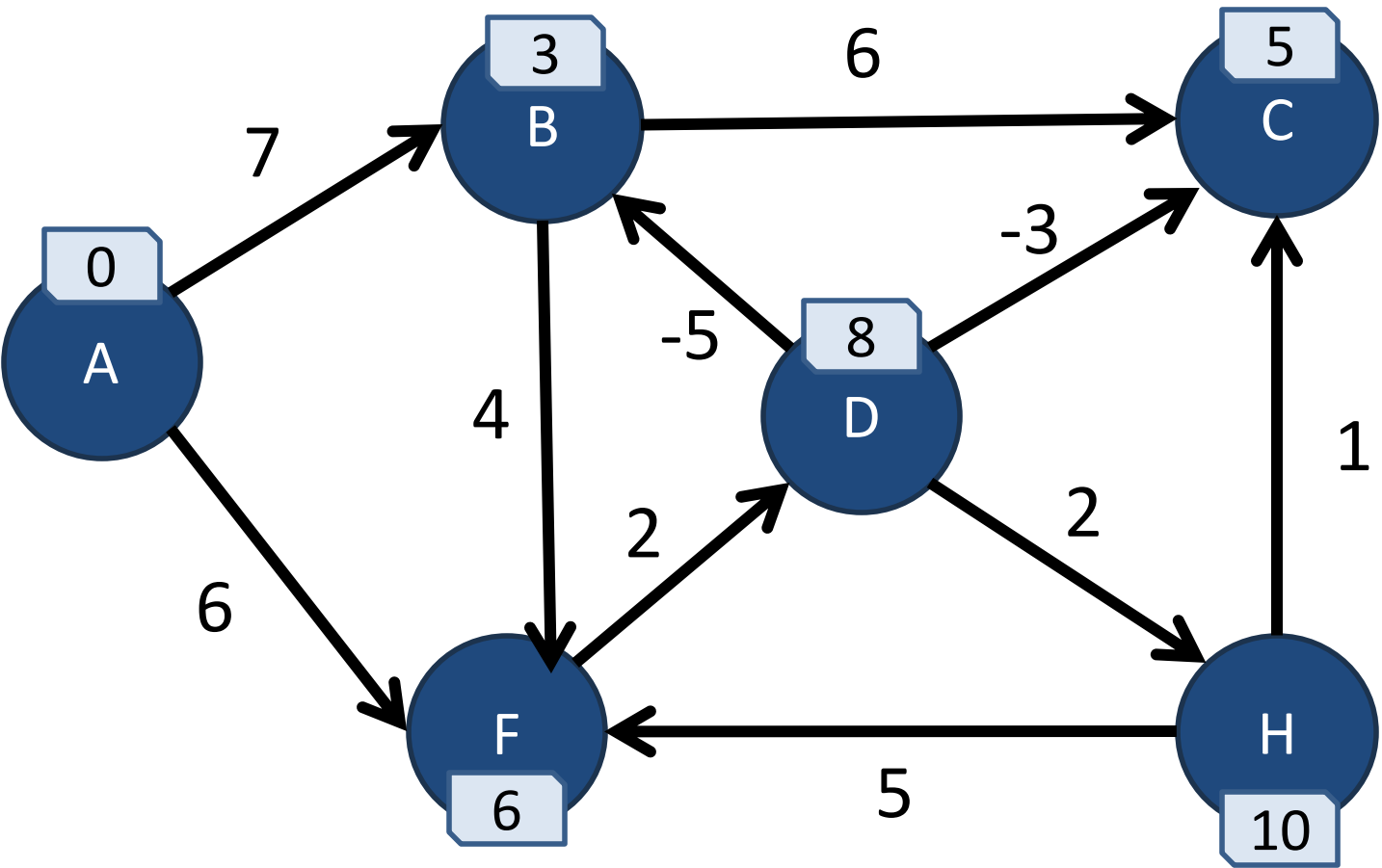
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



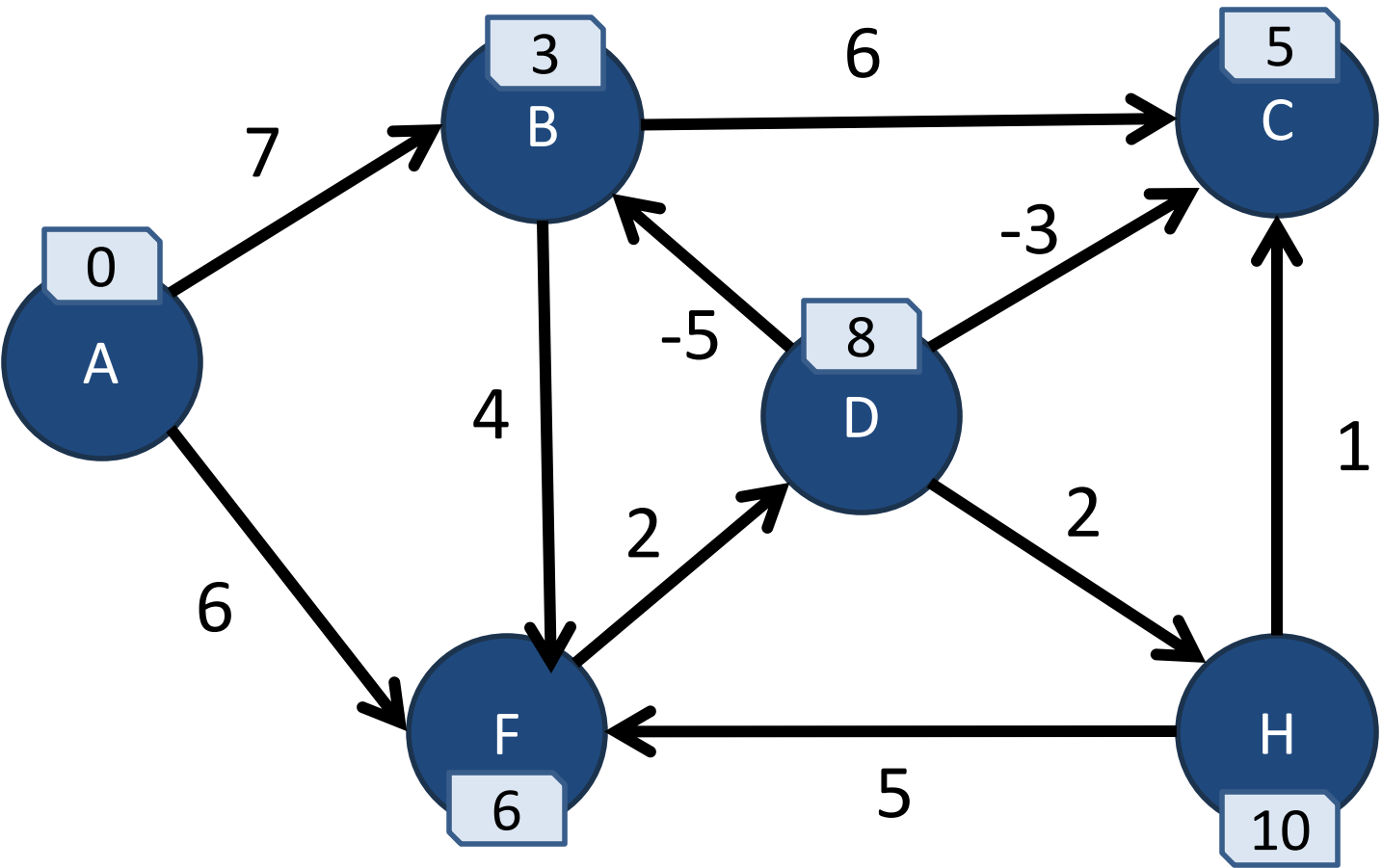
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

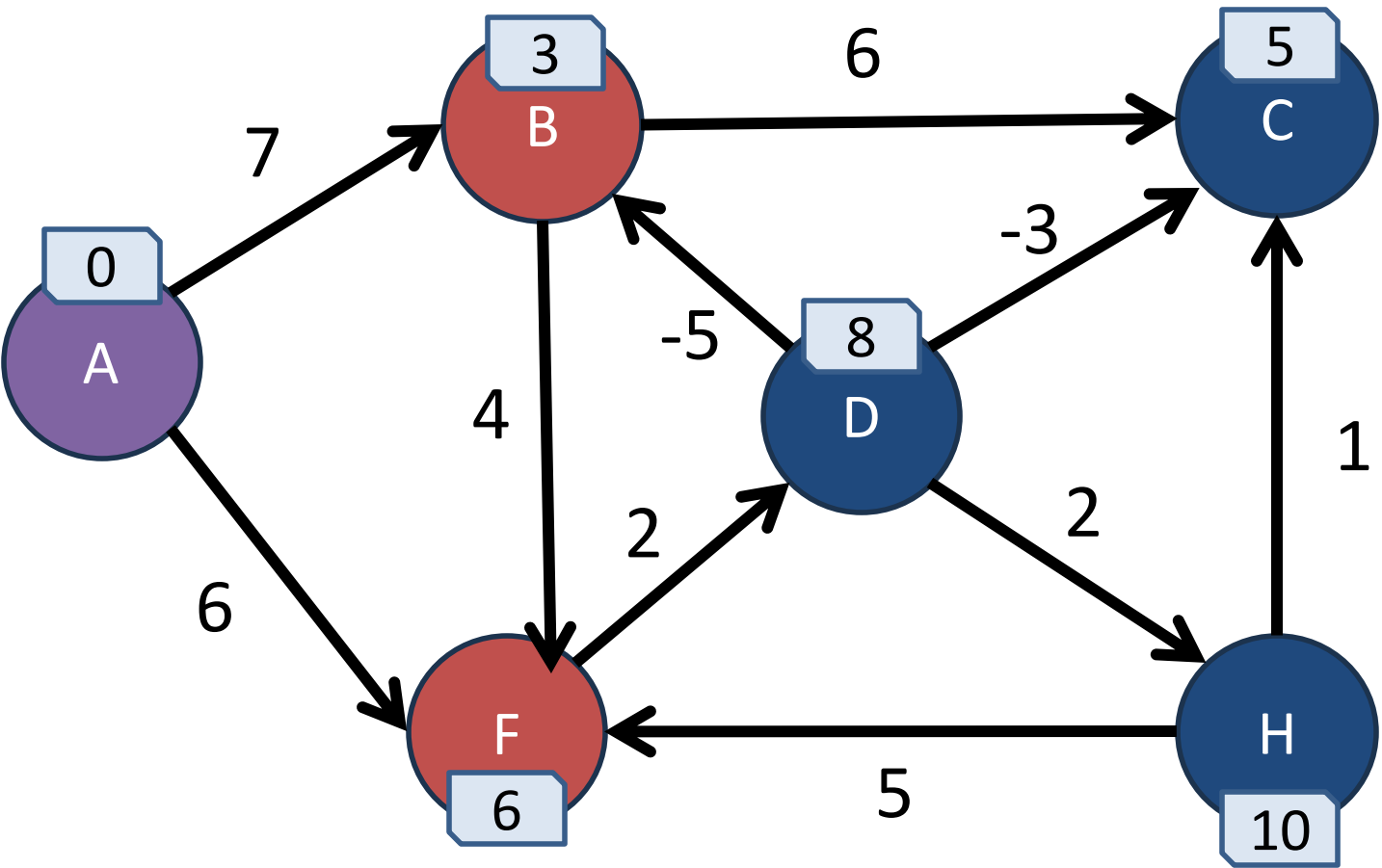
- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

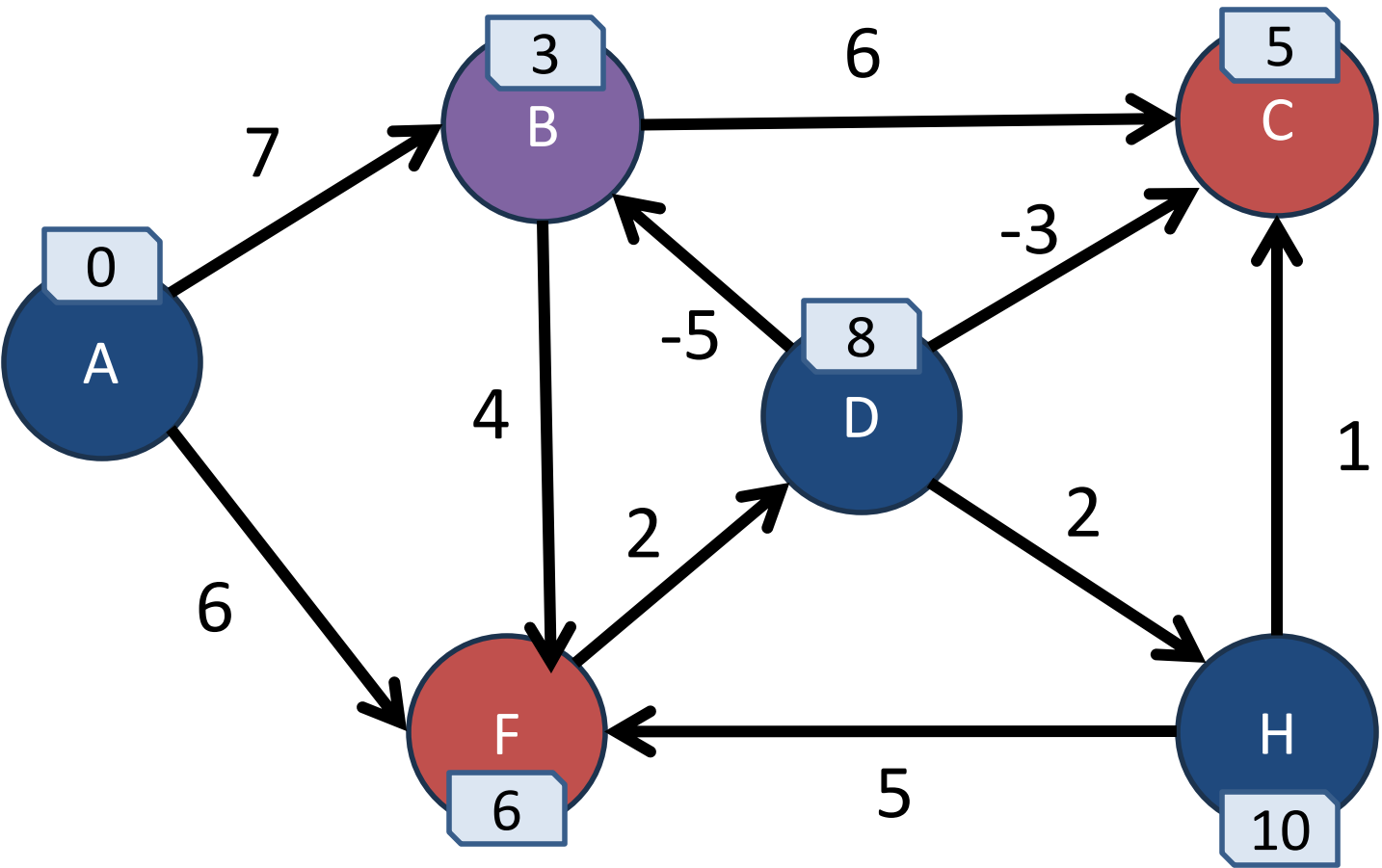
AB	AF	BC	BF	DB	DC	DH	FD	HF	HC
----	----	----	----	----	----	----	----	----	----

Check for negative cycles



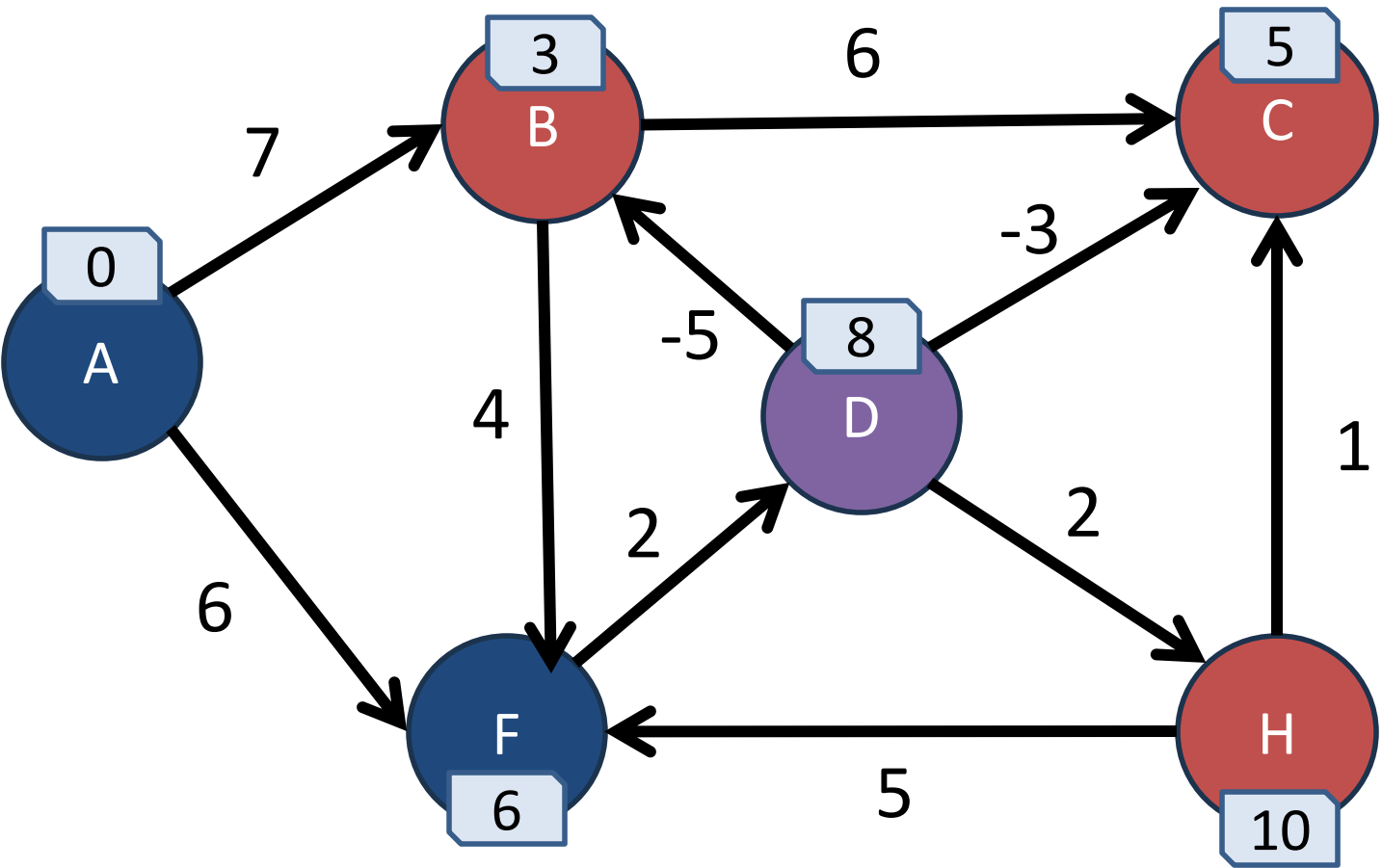
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



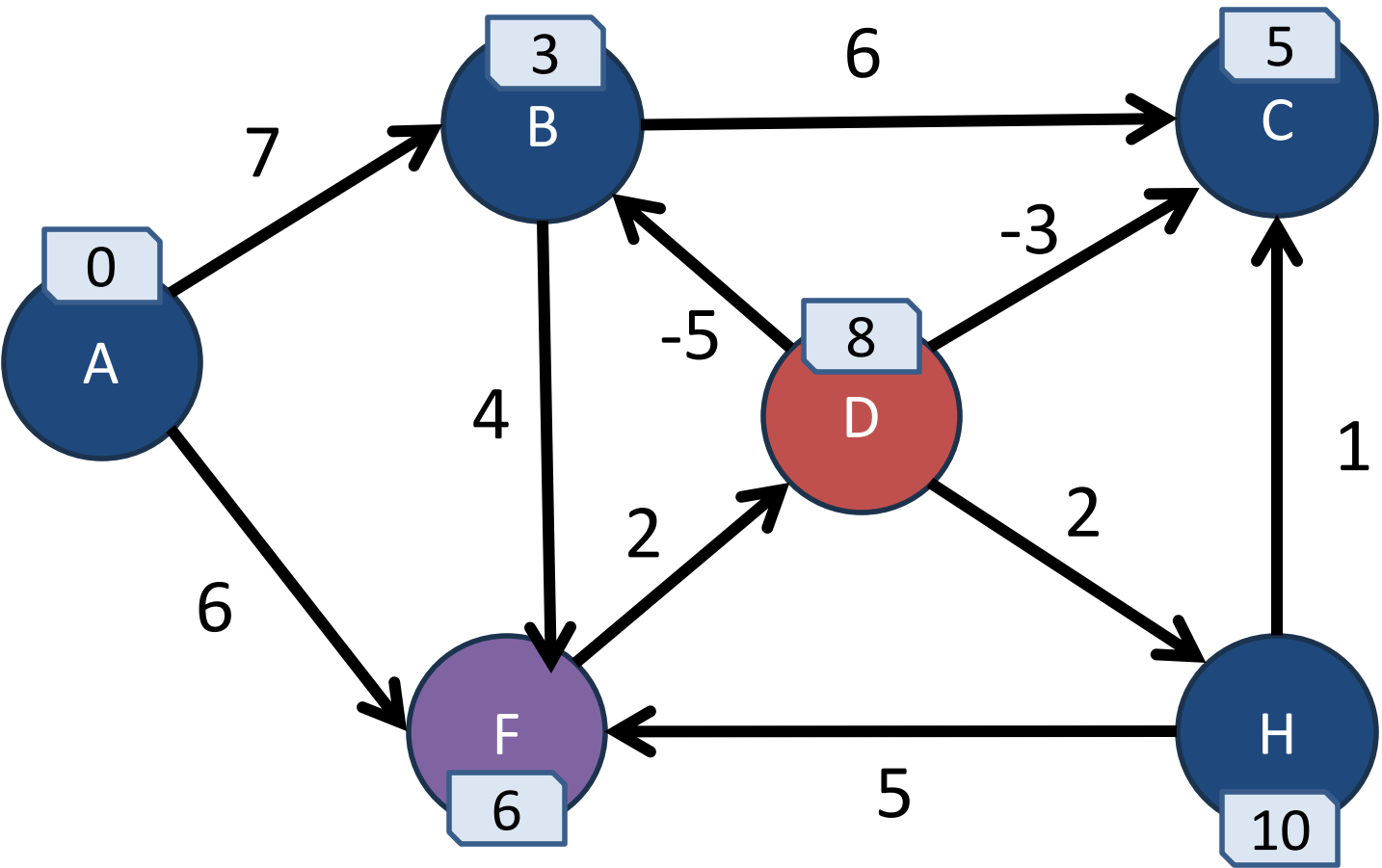
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



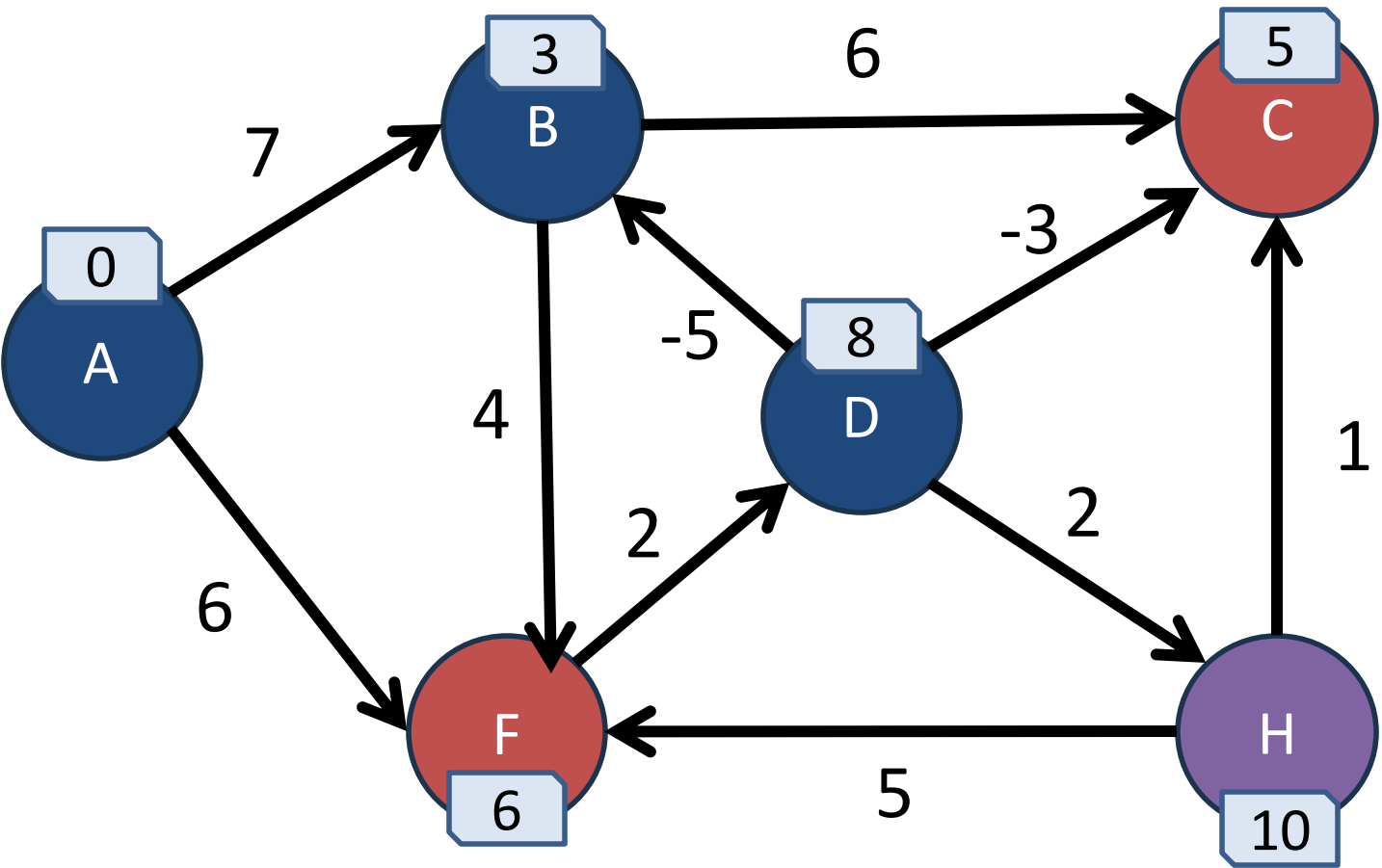
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



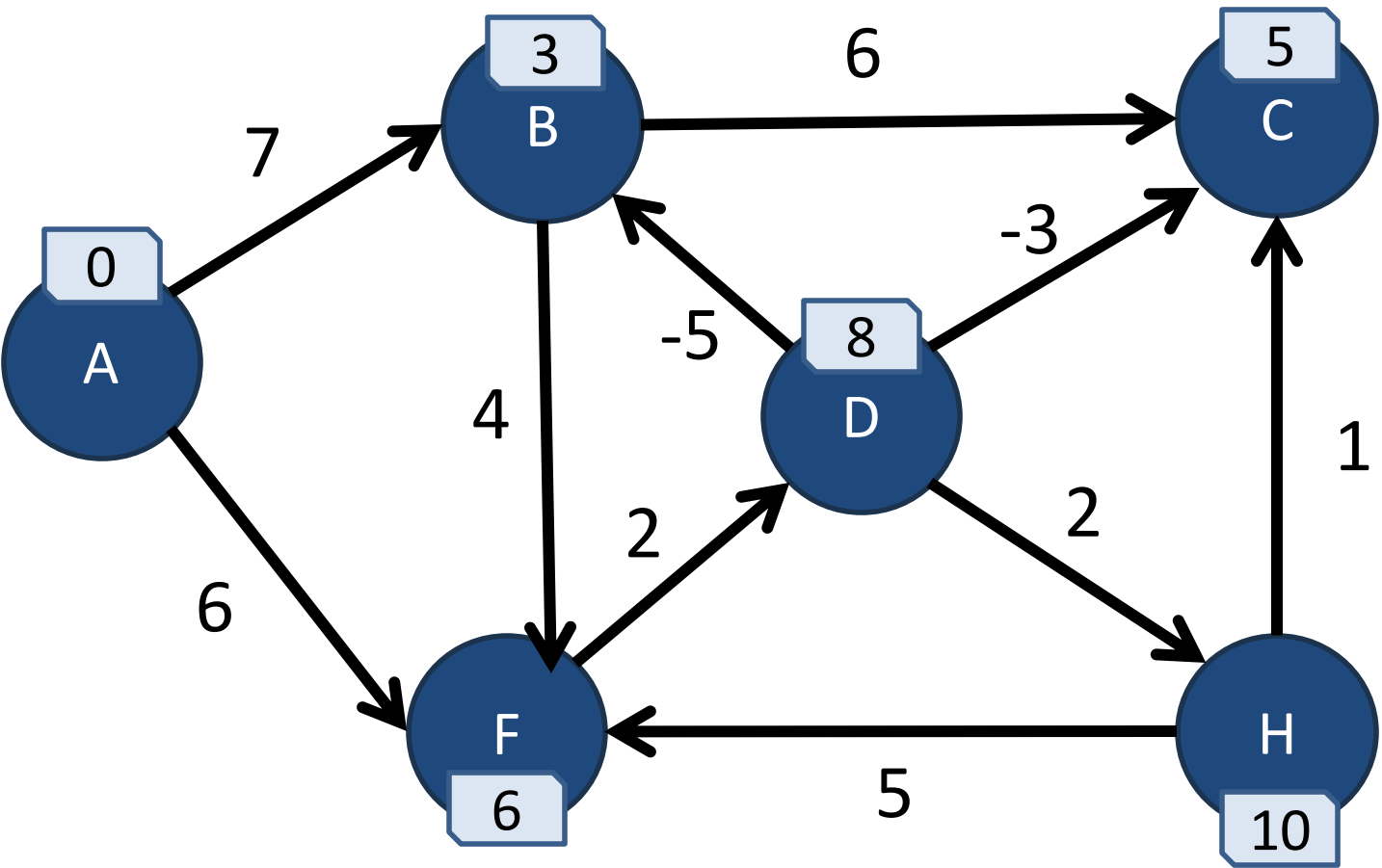
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

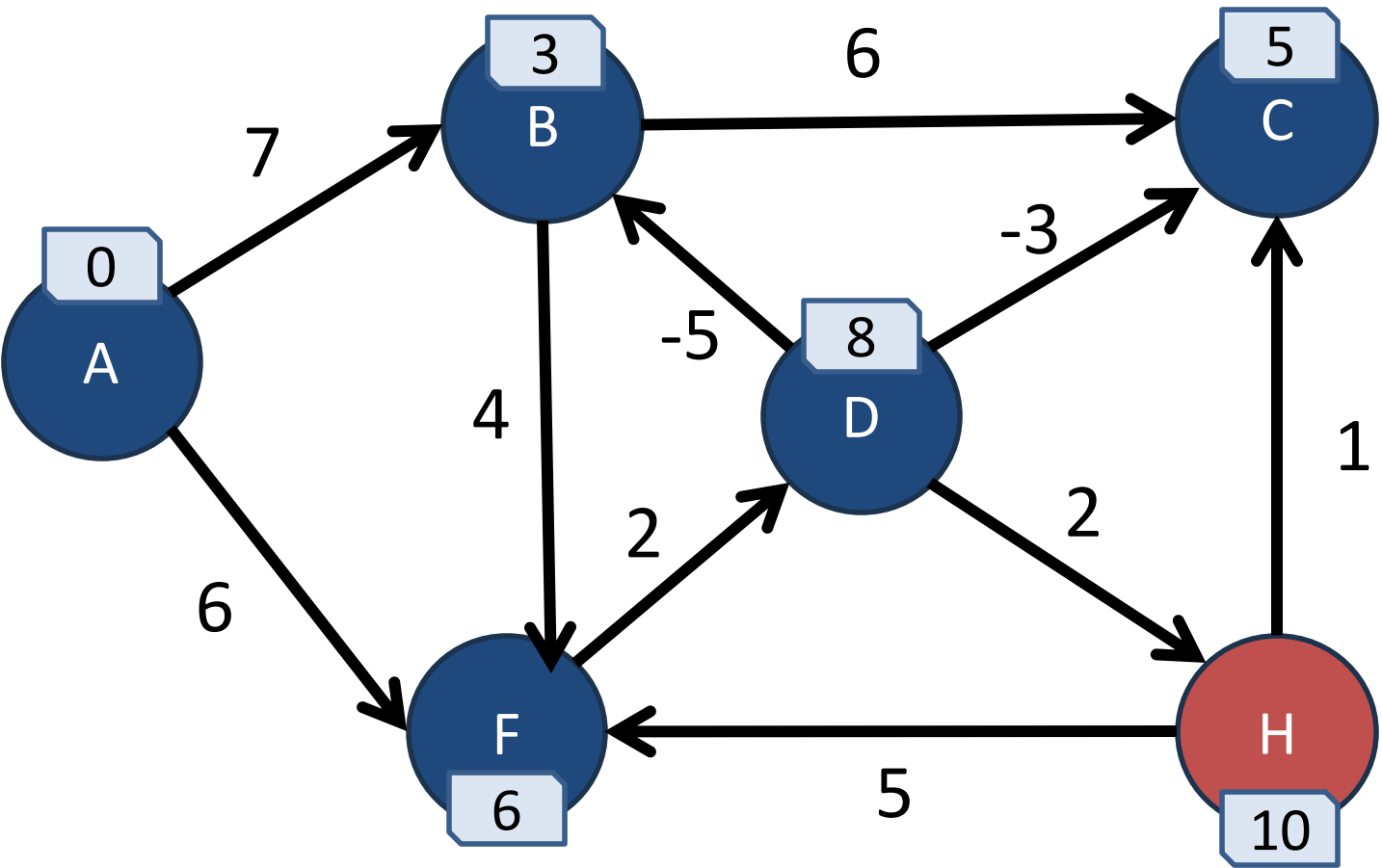
- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

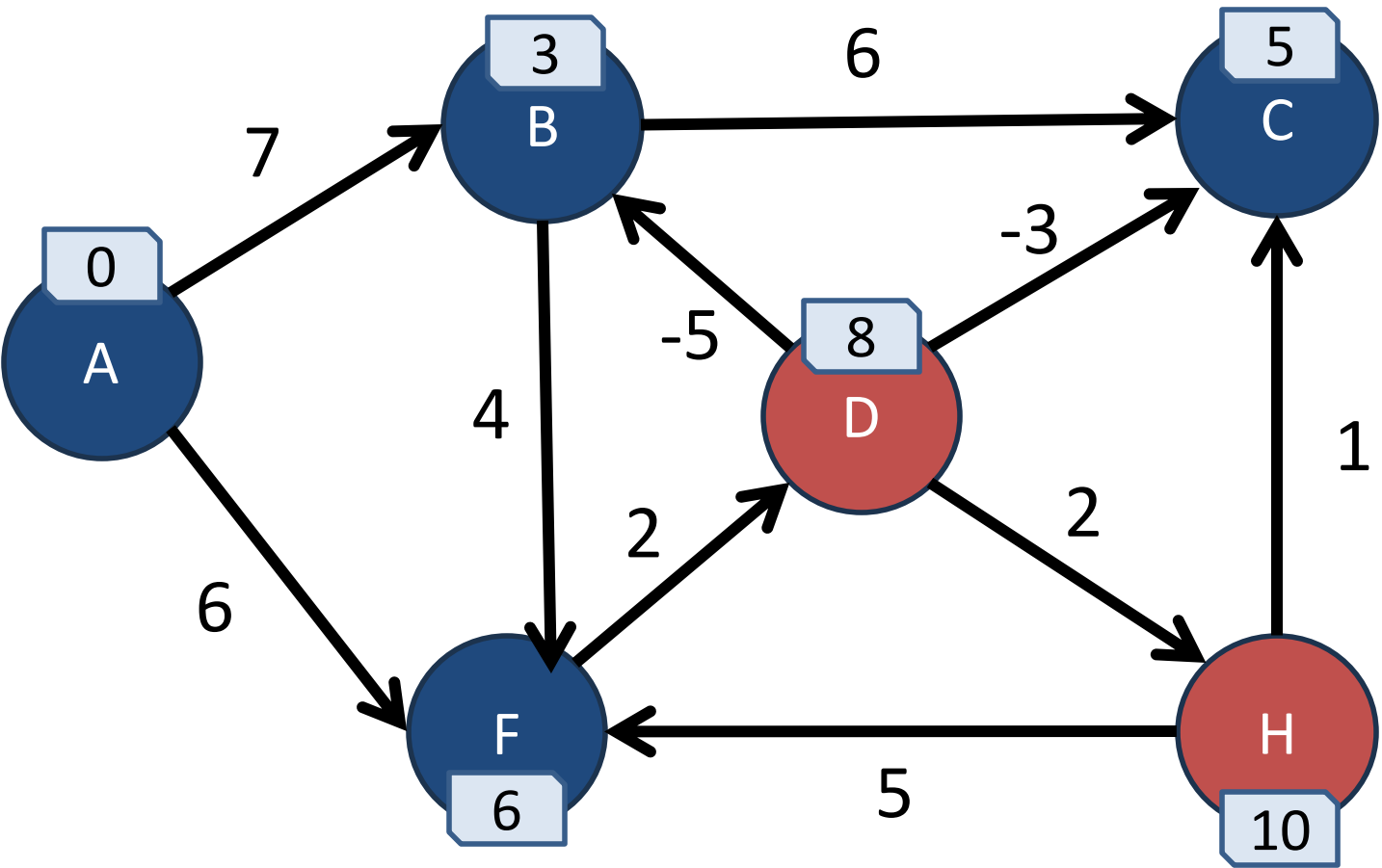
- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC

Let's trace the path



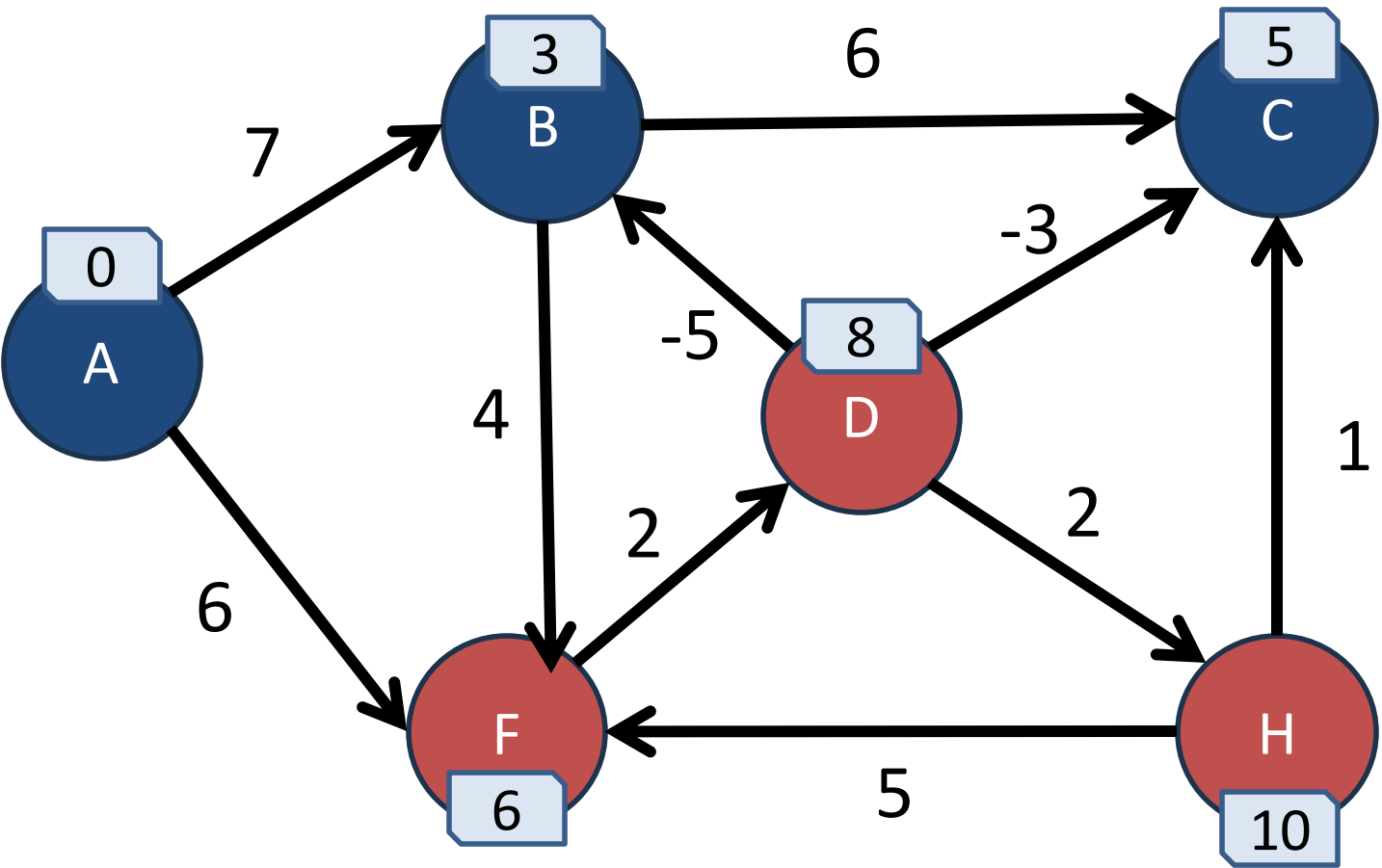
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



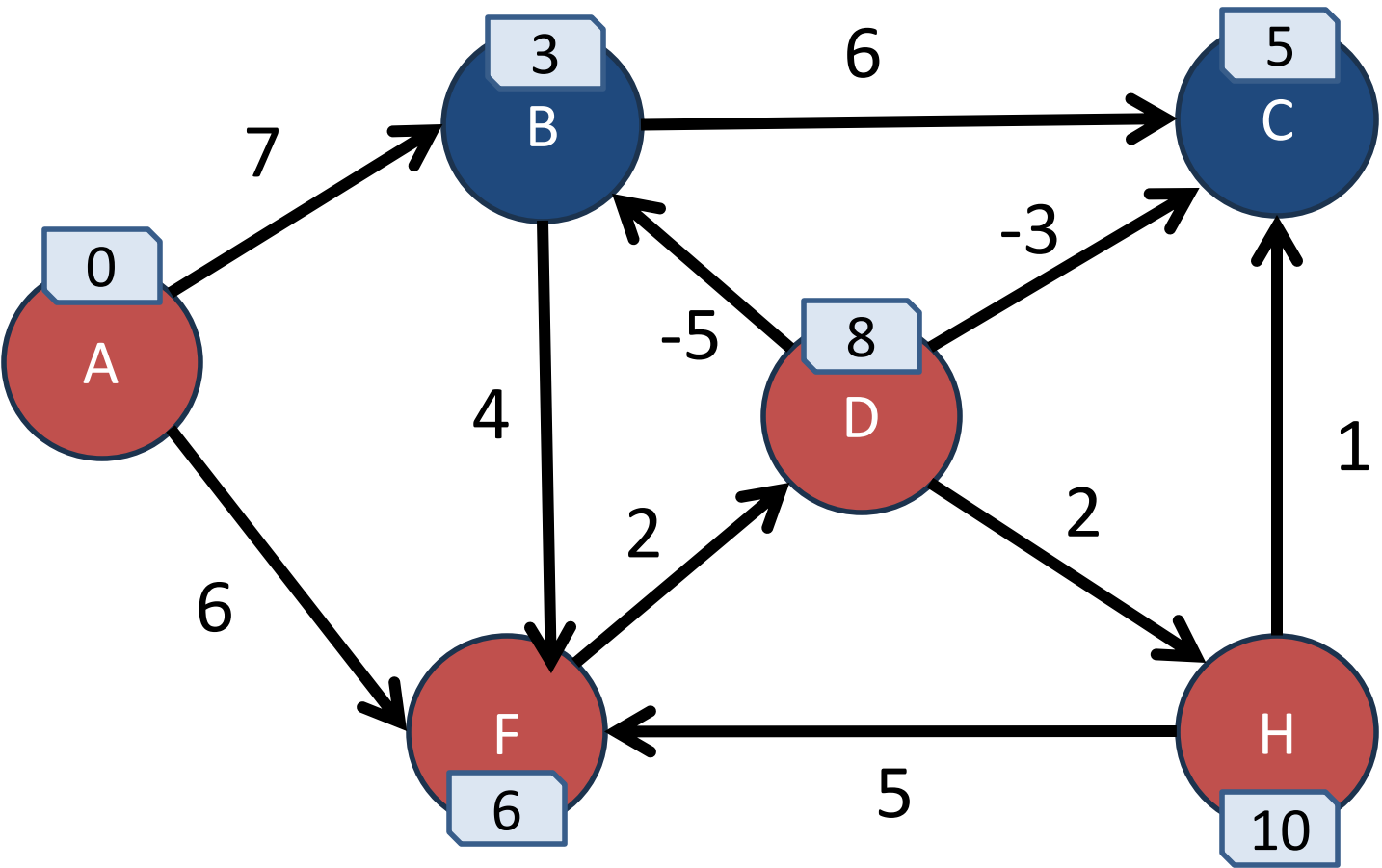
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

AB	AF	BC	BF	DB	DC	DH	FD	HF	HC
----	----	----	----	----	----	----	----	----	----



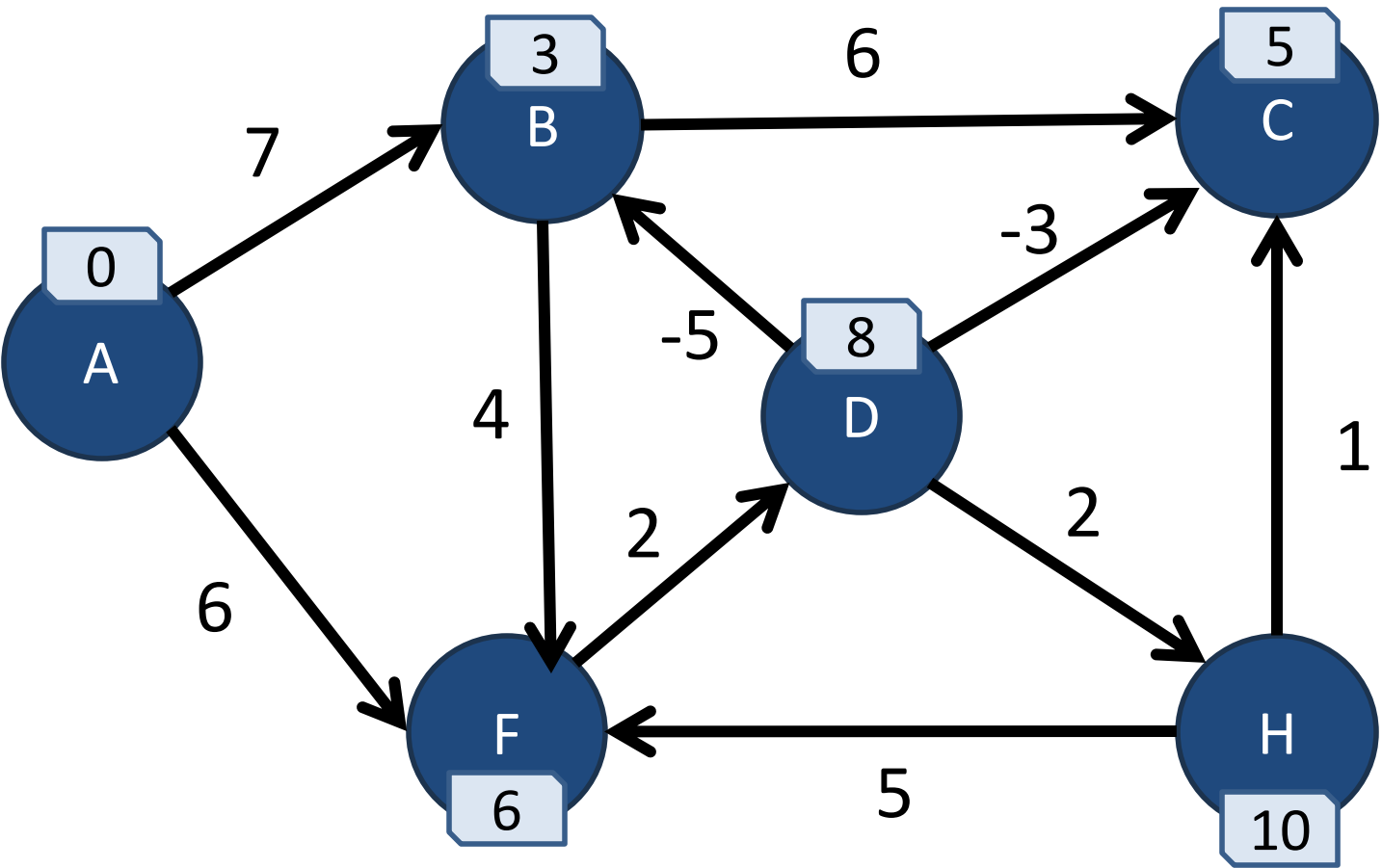
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

AB	AF	BC	BF	DB	DC	DH	FD	HF	HC
----	----	----	----	----	----	----	----	----	----



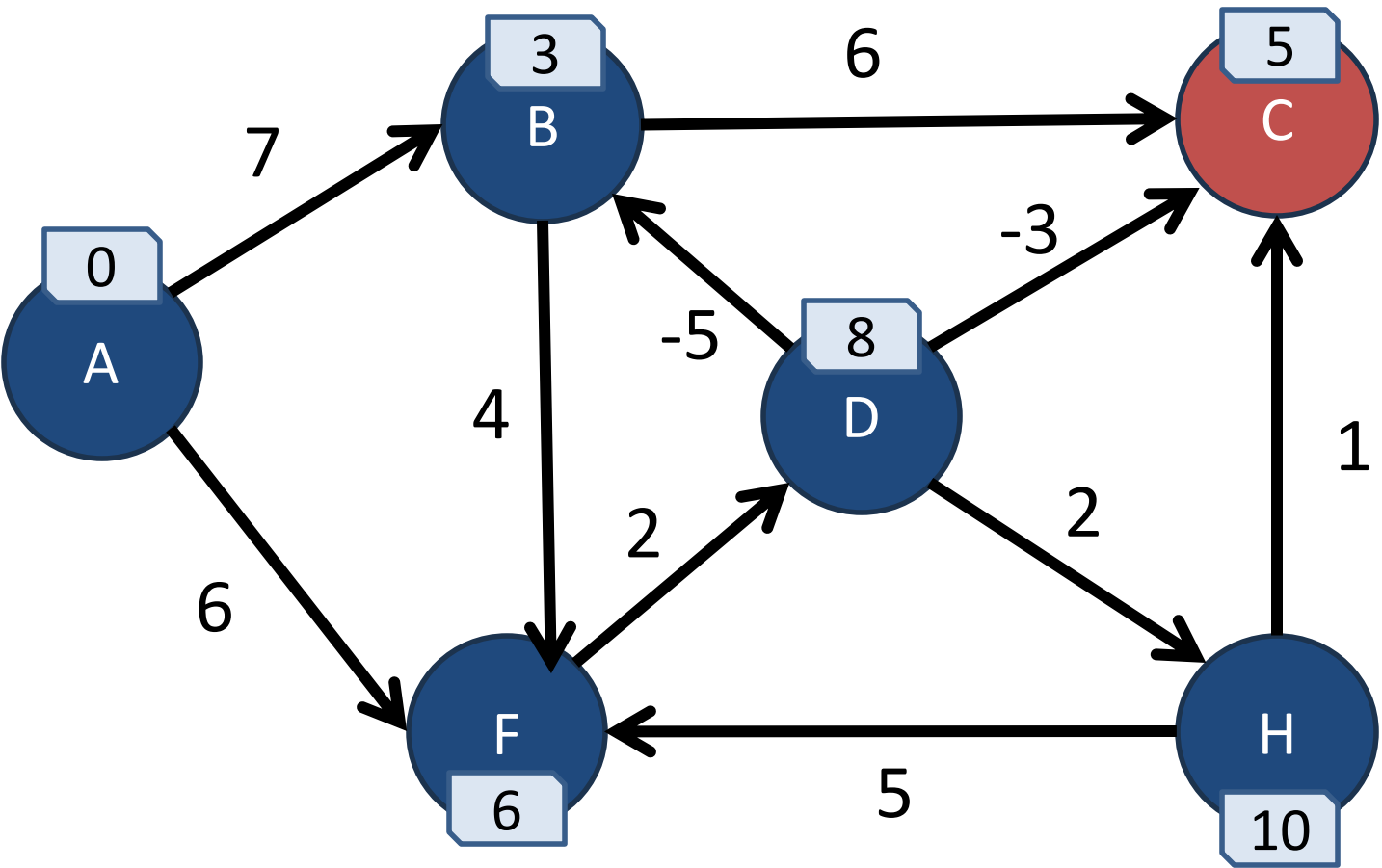
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



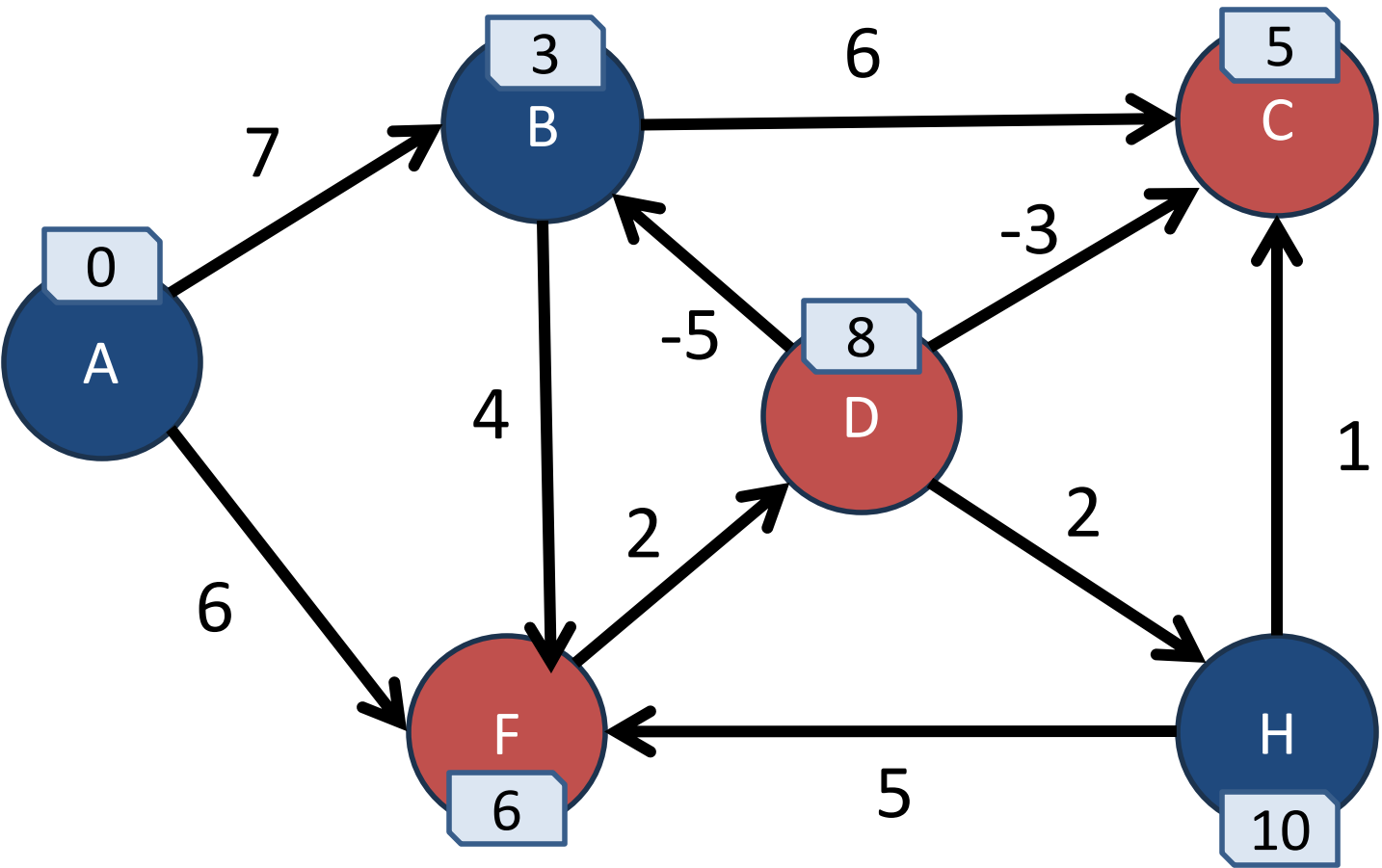
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



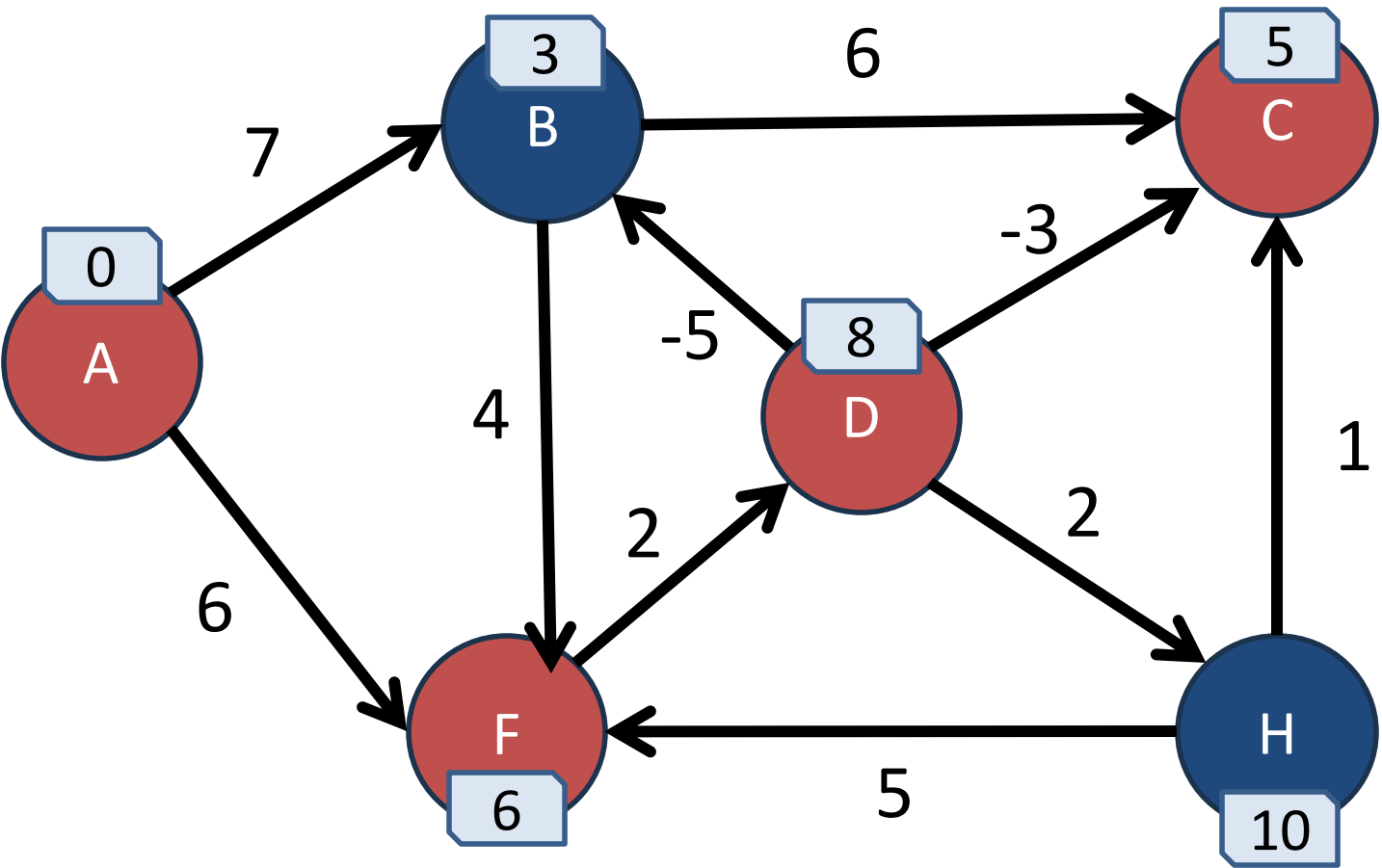
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC



	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

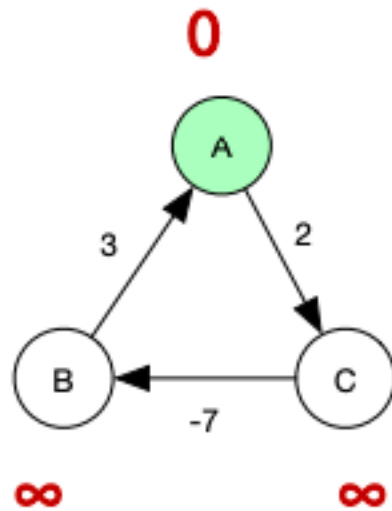
AB	AF	BC	BF	DB	DC	DH	FD	HF	HC
----	----	----	----	----	----	----	----	----	----



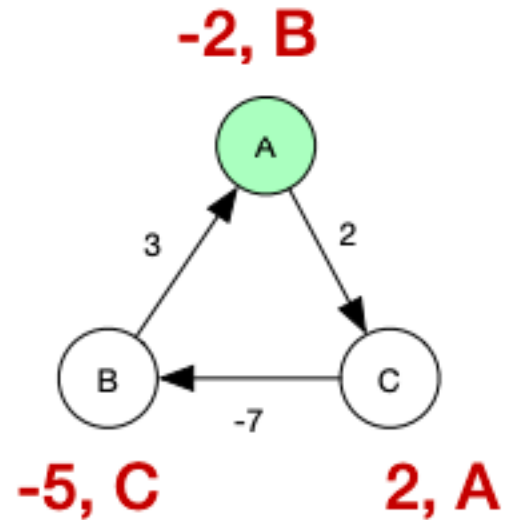
	Cost	Previous
A	0	
B	3	D
C	5	D
D	8	F
F	6	A
H	10	D

- AB
- AF
- BC
- BF
- DB
- DC
- DH
- FD
- HF
- HC

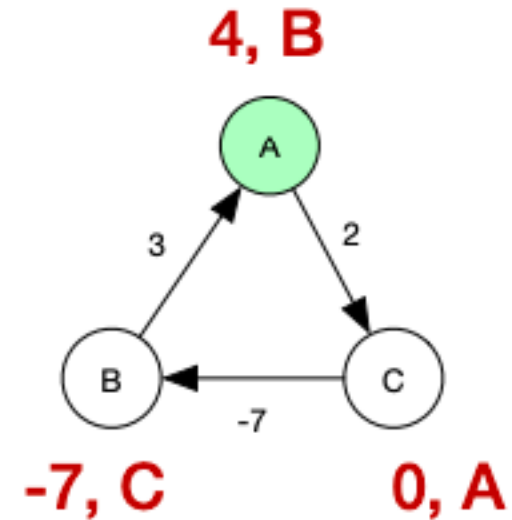
How does bellman-ford detect negative cycles?



initial state



after 1 iteration



after 2 iterations

$$4 > -7 + 3$$

Why this works

- If all shortest paths were found in $V-1$ iterations, no edge should need further relaxation.
- If further relaxation is possible, it means a **path can be made cheaper** — which is only possible if there's a **cycle with a net negative weight**.

Bellman-Ford algorithm

BELLMAN-FORD(G, s)

$O(V)$

```
1  for all  $v \in V$ 
2       $dist[v] \leftarrow \infty$ 
3       $prev[v] \leftarrow null$ 
4   $dist[s] \leftarrow 0$ 
5  for  $i \leftarrow 1$  to  $|V| - 1$ 
6      for all edges  $(u, v) \in E$ 
7          if  $dist[v] > dist[u] + w(u, v)$ 
8               $dist[v] \leftarrow dist[u] + w(u, v)$ 
9               $prev[v] \leftarrow u$ 
10 for all edges  $(u, v) \in E$ 
11     if  $dist[v] > dist[u] + w(u, v)$ 
12         return false
```

Initialize all the
distances

Bellman-Ford algorithm

BELLMAN-FORD(G, s)

```
1  for all  $v \in V$ 
2       $dist[v] \leftarrow \infty$ 
3       $prev[v] \leftarrow null$ 
4   $dist[s] \leftarrow 0$ 
5  for  $i \leftarrow 1$  to  $|V| - 1$   $O(V)$ 
6      for all edges  $(u, v) \in E$   $O(E)$ 
7          if  $dist[v] > dist[u] + w(u, v)$ 
8               $dist[v] \leftarrow dist[u] + w(u, v)$ 
9               $prev[v] \leftarrow u$ 
10 for all edges  $(u, v) \in E$ 
11     if  $dist[v] > dist[u] + w(u, v)$ 
12         return false
```

$O(V)$

$O(VE)$

iterate over all edges/vertices and apply update rule

Bellman-Ford algorithm

BELLMAN-FORD(G, s)

```
1  for all  $v \in V$ 
2       $dist[v] \leftarrow \infty$ 
3       $prev[v] \leftarrow null$ 
4   $dist[s] \leftarrow 0$ 
5  for  $i \leftarrow 1$  to  $|V| - 1$ 
6      for all edges  $(u, v) \in E$ 
7          if  $dist[v] > dist[u] + w(u, v)$ 
8               $dist[v] \leftarrow dist[u] + w(u, v)$ 
9               $prev[v] \leftarrow u$ 
10 for all edges  $(u, v) \in E$ 
11     if  $dist[v] > dist[u] + w(u, v)$ 
12         return false
```

$O(V)$

$O(VE)$

Bellman-Ford algorithm

BELLMAN-FORD(G, s)

```
1  for all  $v \in V$ 
2       $dist[v] \leftarrow \infty$ 
3       $prev[v] \leftarrow null$ 
4   $dist[s] \leftarrow 0$ 
5  for  $i \leftarrow 1$  to  $|V| - 1$ 
6      for all edges  $(u, v) \in E$ 
7          if  $dist[v] > dist[u] + w(u, v)$ 
8               $dist[v] \leftarrow dist[u] + w(u, v)$ 
9               $prev[v] \leftarrow u$ 
10 for all edges  $(u, v) \in E$ 
11     if  $dist[v] > dist[u] + w(u, v)$ 
12         return false
```

O(V)

O(VE)

O(V)

check for negative cycles