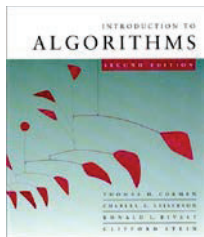


# Shortest Path Algorithms

**The slides source: Prof. Erik Demaine**

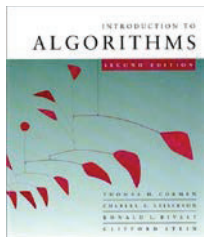
**Link: <https://engineeringppt.com/category/electrical-engineering/>**



# Paths in graphs

Consider a digraph  $G = (V, E)$  with edge-weight function  $w : E \rightarrow \mathbb{R}$ . The **weight** of path  $p = v_1 \rightarrow v_2 \rightarrow \cdots \rightarrow v_k$  is defined to be

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

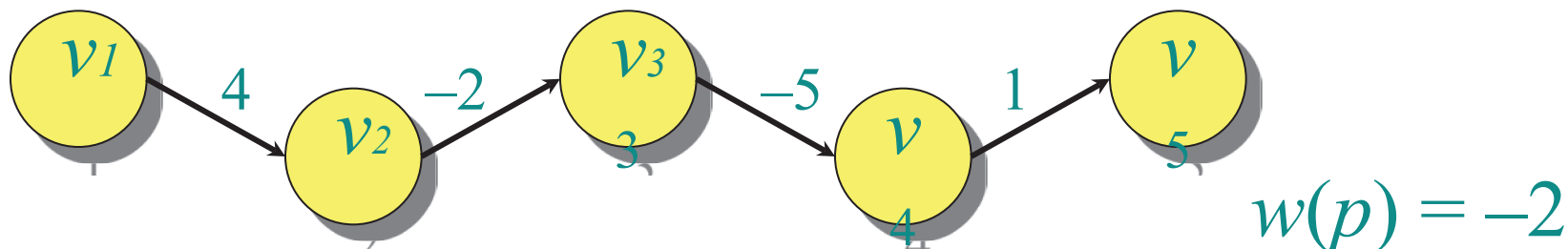


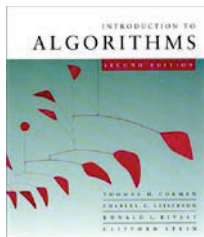
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**Example:**



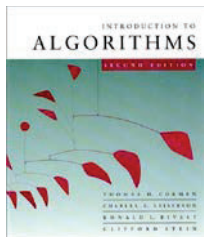


# Shortest paths

A *shortest path* from  $u$  to  $v$  is a path of minimum weight from  $u$  to  $v$ . The *shortest-path weight* from  $u$  to  $v$  is defined as

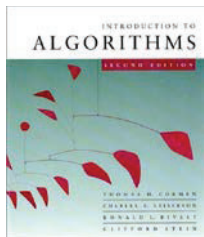
$$\delta(u, v) = \min\{w(p) : p \text{ is a path from } u \text{ to } v\}.$$

**Note:**  $\delta(u, v) = \infty$  if no path from  $u$  to  $v$  exists.



# Well-definedness of shortest paths

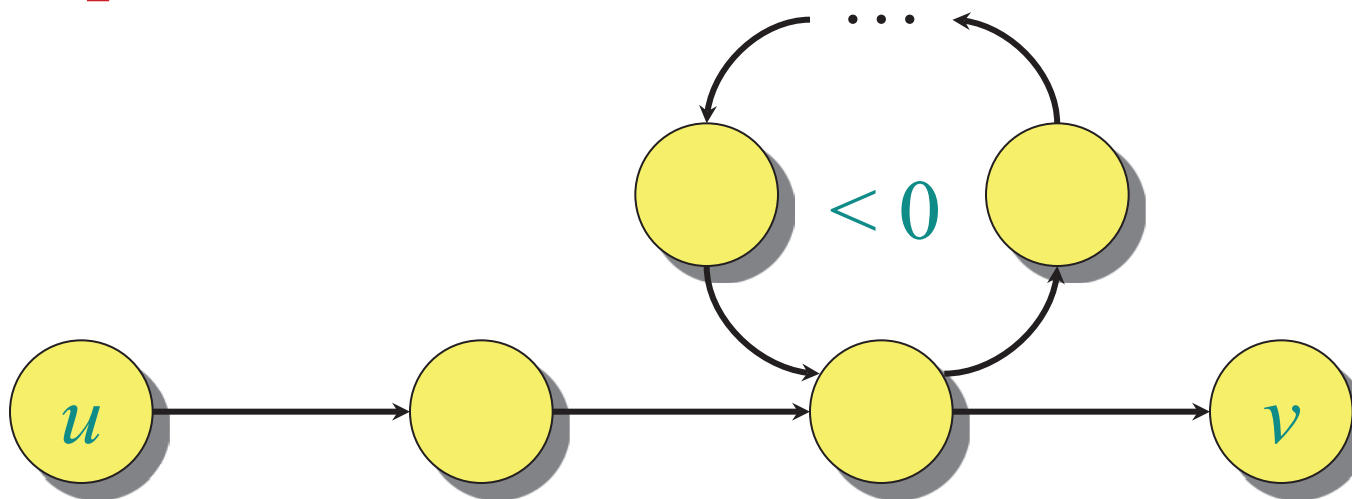
If a graph  $G$  contains a negative-weight cycle, then some shortest paths do not exist.

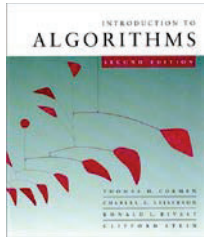


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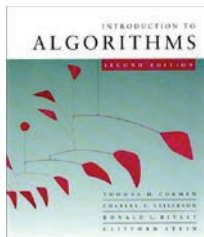
**Example:**





# Optimal substructure

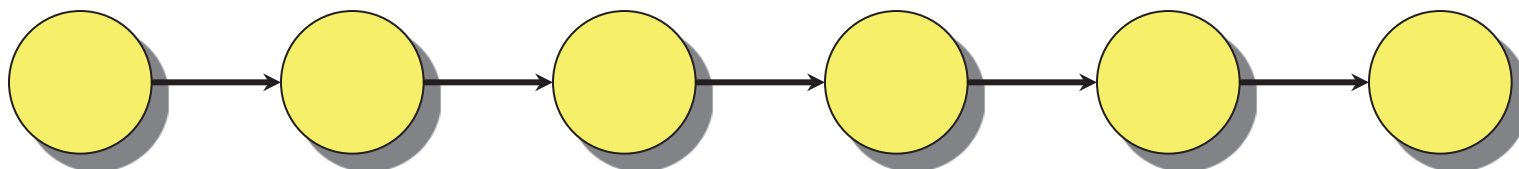
**Theorem.** A subpath of a shortest path is a shortest path.



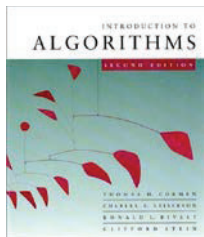
# Optimal substructure

**Theorem.** A subpath of a shortest path is a shortest path.

*Proof.* Cut and paste:



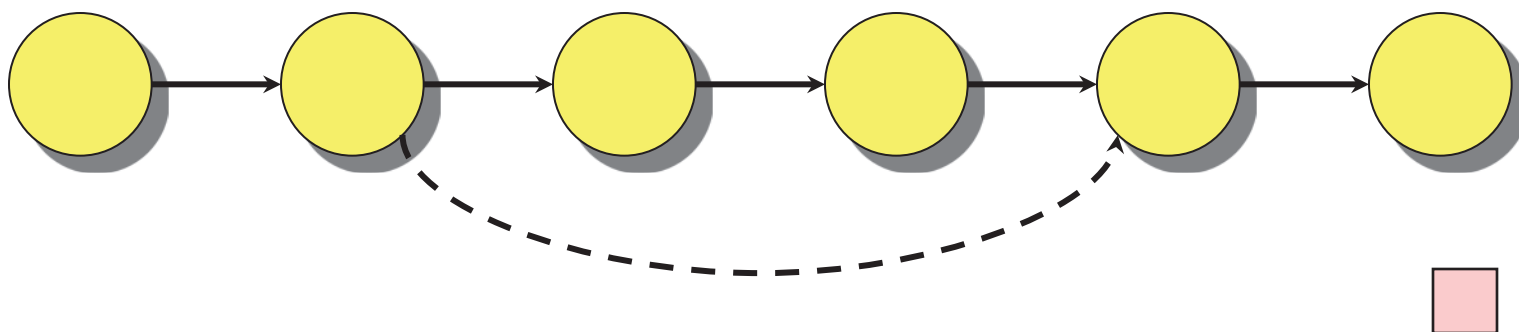


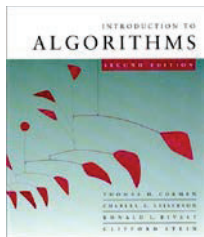


# Optimal substructure

**Theorem.** A subpath of a shortest path is a shortest path.

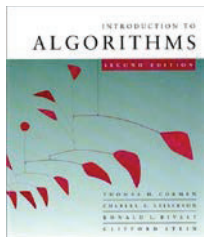
*Proof.* Cut and paste:





# Triangle inequality

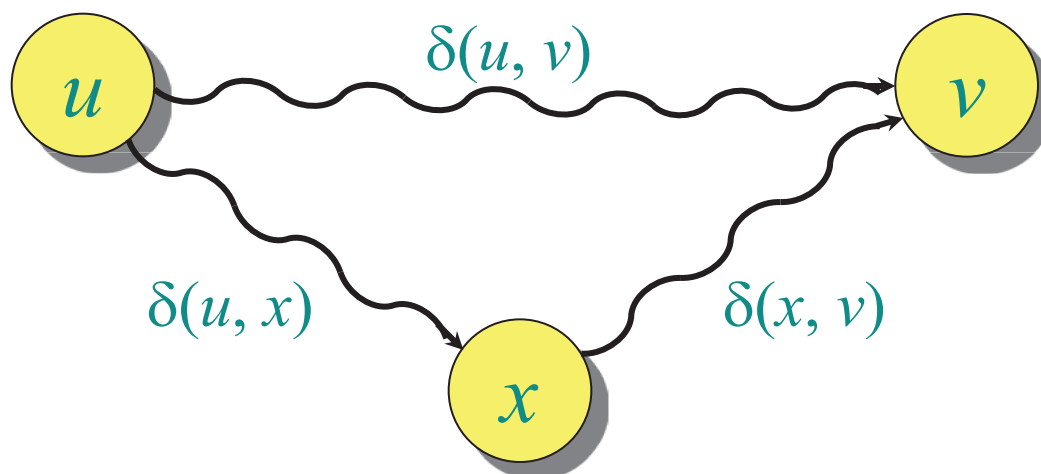
**Theorem.** For all  $u, v, x \in V$ , we have  
$$\delta(u, v) \leq \delta(u, x) + \delta(x, v).$$

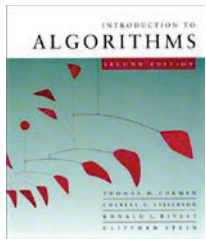


# Triangle inequality

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*Proof.*





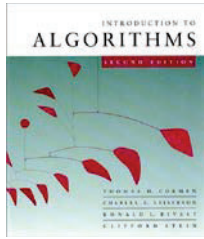
# Single-source shortest paths (nonnegative edge weights)

**Problem.** Assume that  $w(u, v) \geq 0$  for all  $(u, v) \in E$ . (Hence, all shortest-path weights must exist.) From a given source vertex  $s \in V$ , find the shortest-path weights  $\delta(s, v)$  for all  $v \in V$ .

---

**IDEA:** Greedy.

1. Maintain a set  $S$  of vertices whose shortest-path distances from  $s$  are known.
2. At each step, add to  $S$  the vertex  $v \in V - S$  whose distance estimate from  $s$  is minimum.
3. Update the distance estimates of vertices adjacent to  $v$ .



# Dijkstra's algorithm

$d[s] \leftarrow 0$

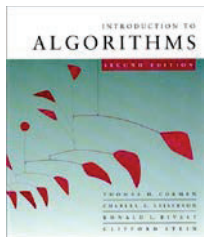
**for** each  $v \in V - \{s\}$

**do**  $d[v] \leftarrow \infty$

$S \leftarrow \emptyset$

$Q \leftarrow V$

▷  $Q$  is a priority queue maintaining  $V - S$ ,  
keyed on  $d[v]$



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**while**  $Q \neq \emptyset$

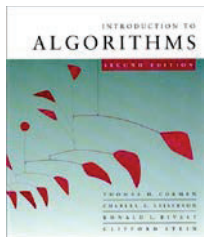
**do**  $u \leftarrow \text{EXTRACT-MIN}(Q)$

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

**for** each  $v \in \text{Adj}[u]$

**do if**  $d[v] > d[u] + w(u, v)$

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



# Dijkstra's algorithm

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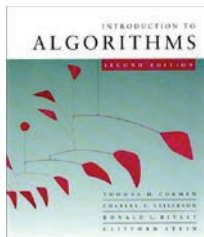
**for** each  $v \in \text{Adj}[u]$

**do** **if**  $d[v] > d[u] + w(u, v)$

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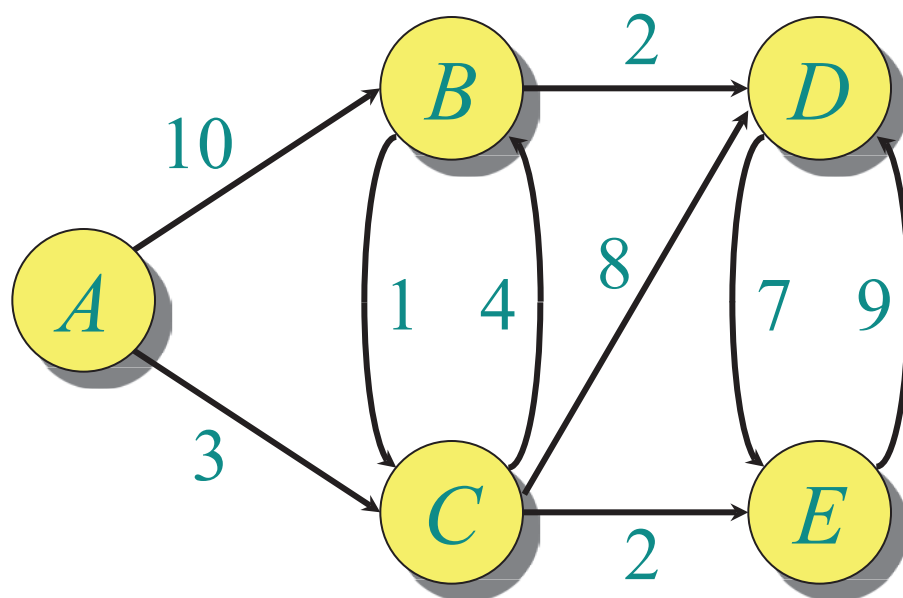
*relaxation  
step*

↑ Implicit DECREASE-KEY

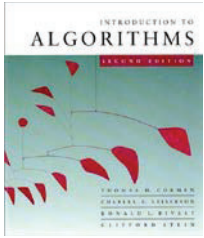


# Example of Dijkstra's algorithm

**Graph with  
nonnegative  
edge weights:**

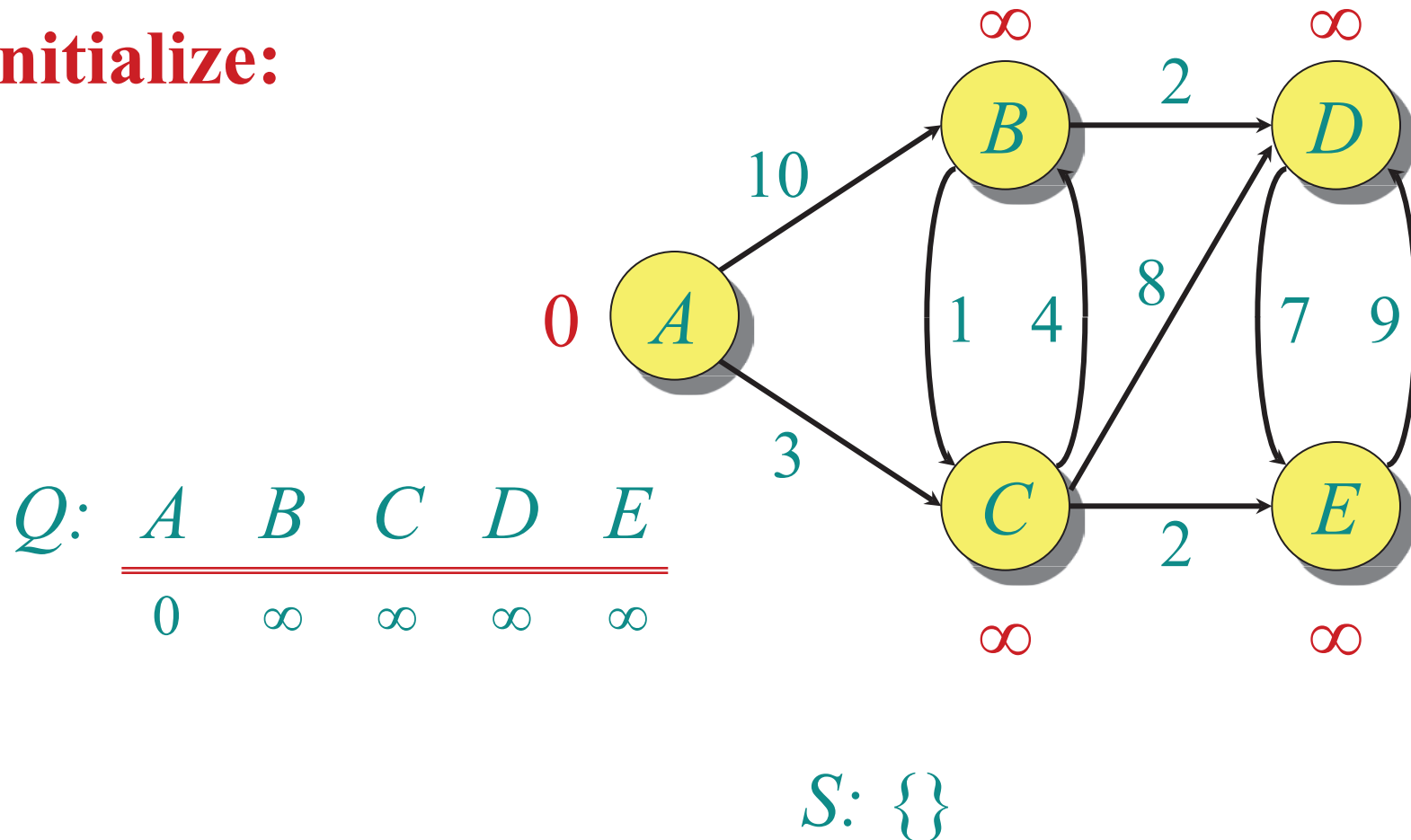


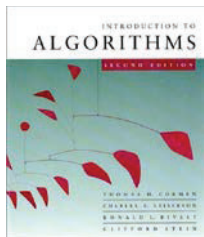




# Example of Dijkstra's algorithm

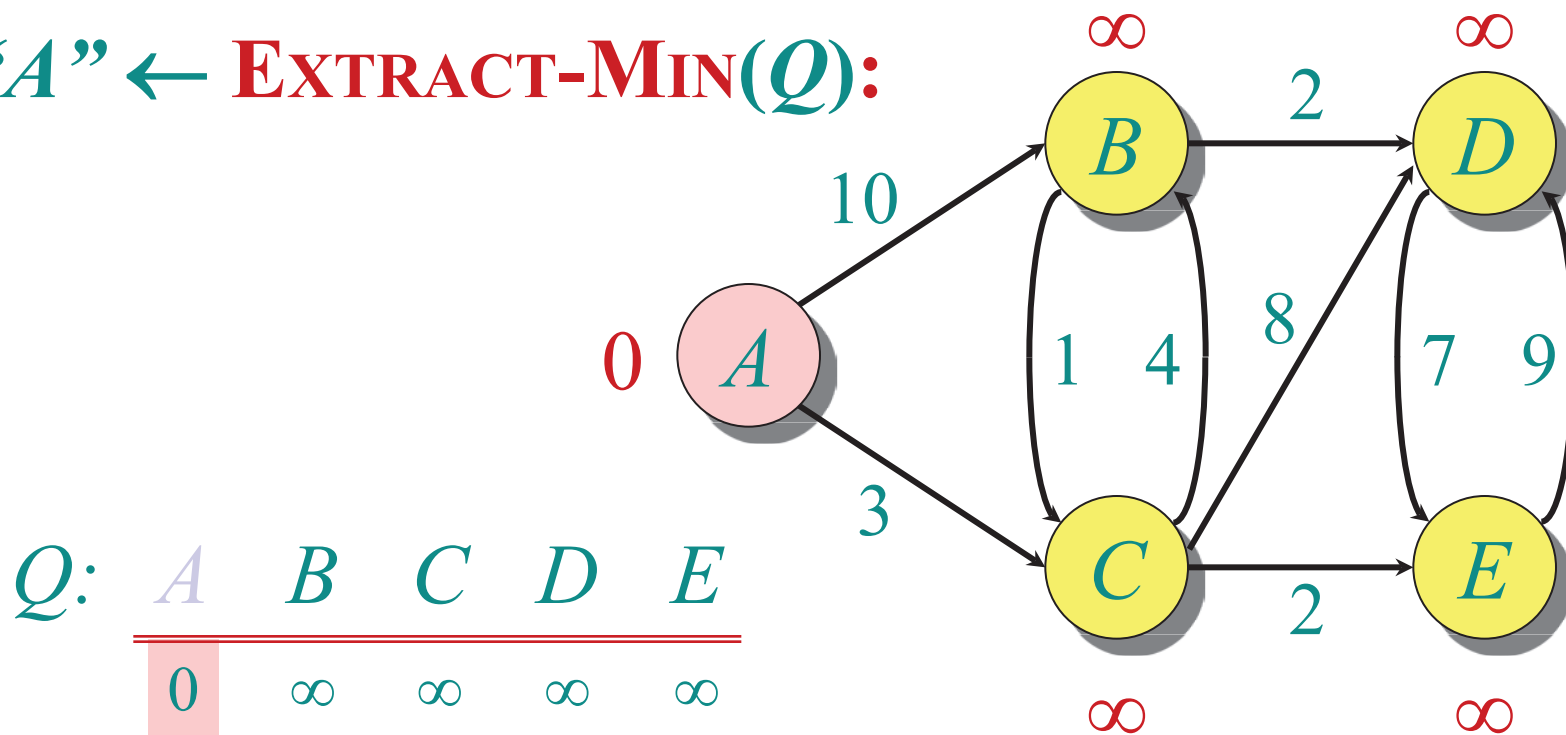
# Initialize:



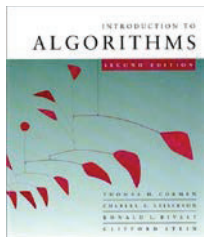


# Example of Dijkstra's algorithm

“A” ← **EXTRACT-MIN**(Q):

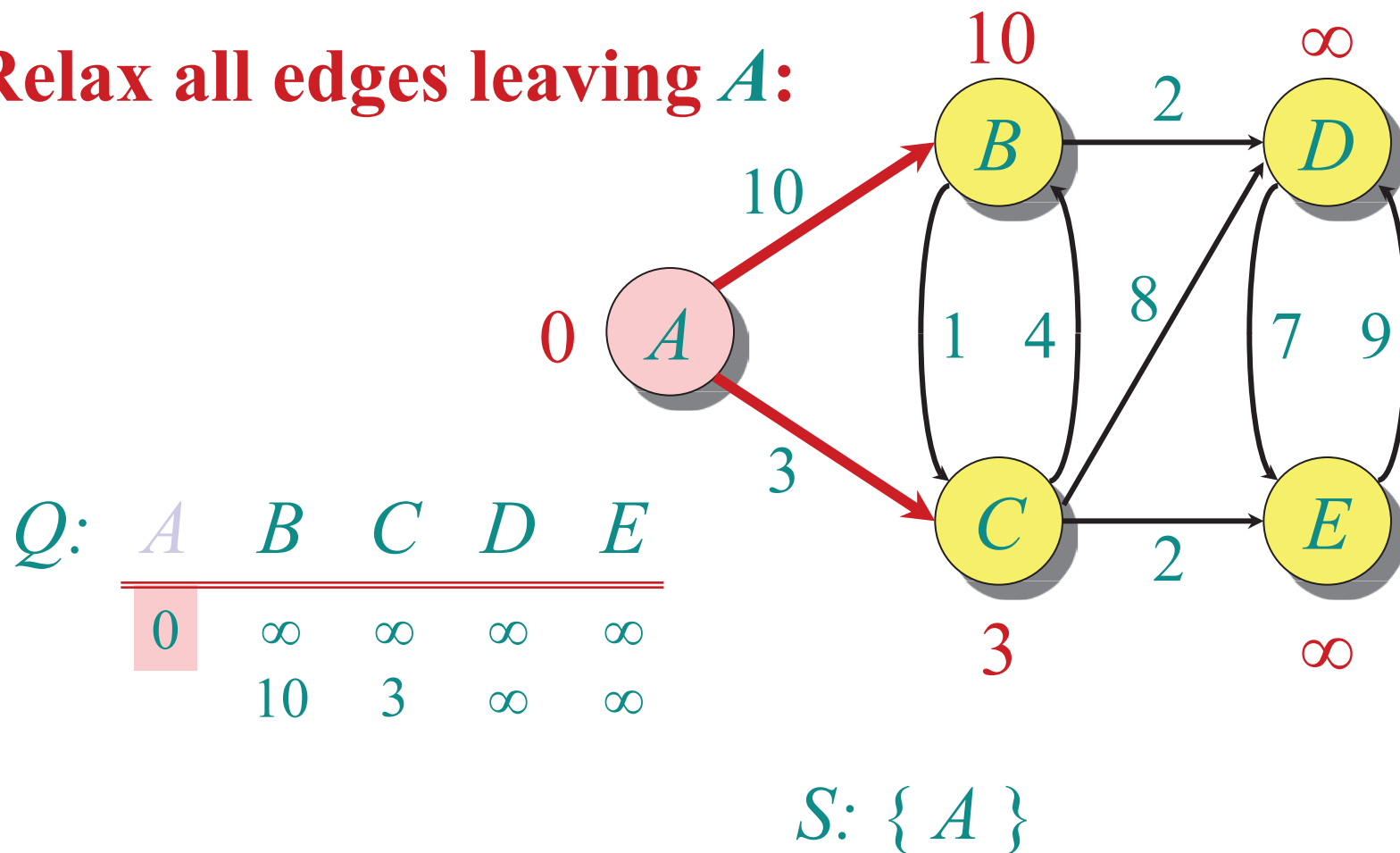


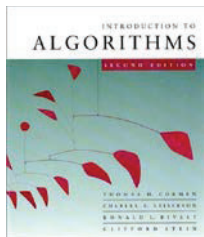
S: { A }



# Example of Dijkstra's algorithm

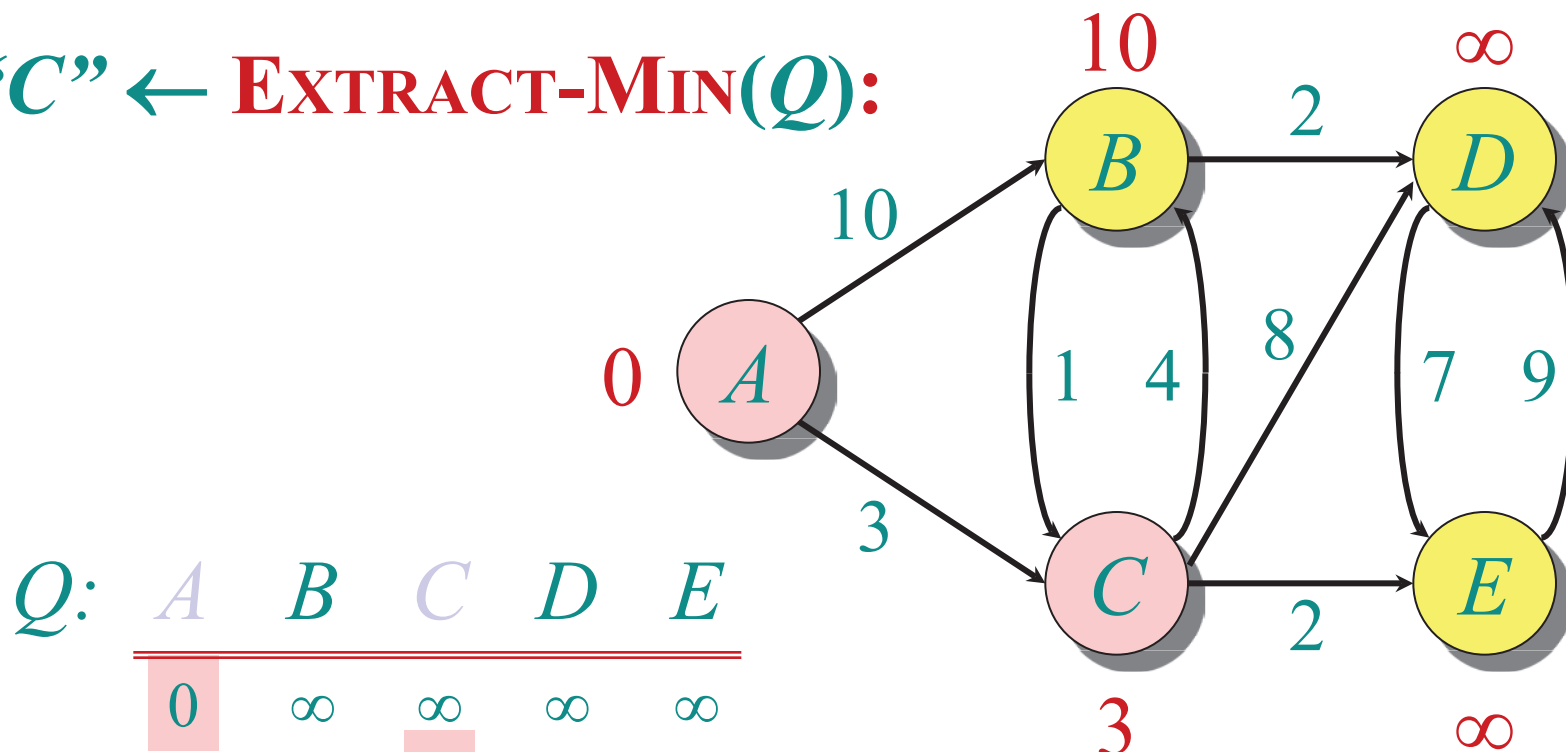
Relax all edges leaving  $A$ :





# Example of Dijkstra's algorithm

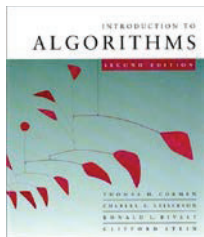
“C” ← **EXTRACT-MIN**(Q):



Q:

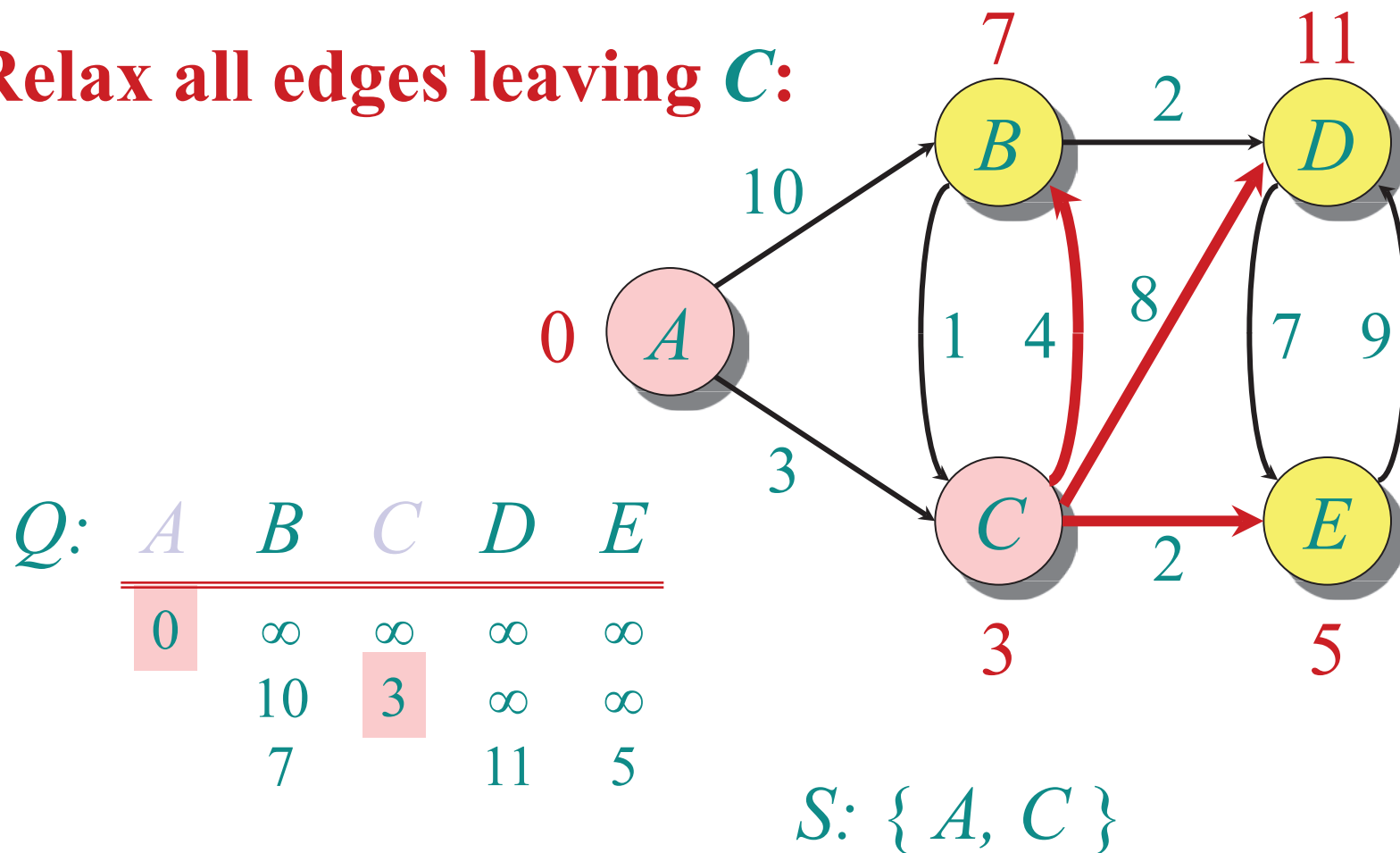
A	B	C	D	E
0	$\infty$	$\infty$	$\infty$	$\infty$
	10	3	$\infty$	$\infty$

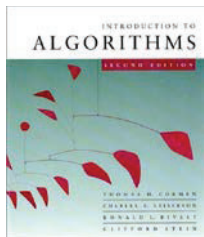
S: { A, C }



# Example of Dijkstra's algorithm

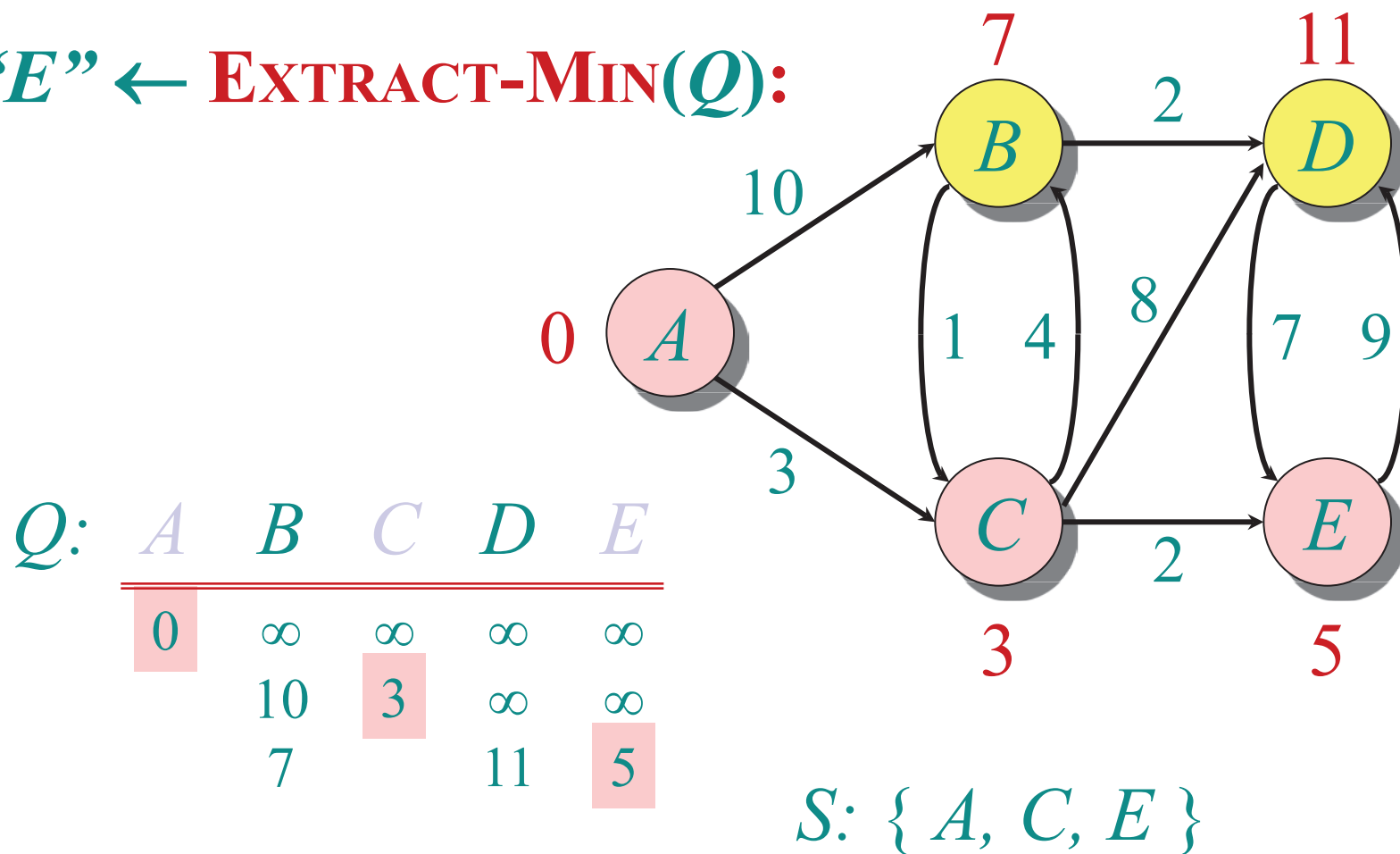
Relax all edges leaving **C**:

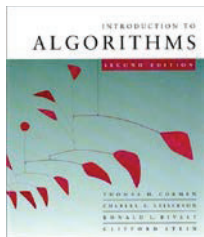




# Example of Dijkstra's algorithm

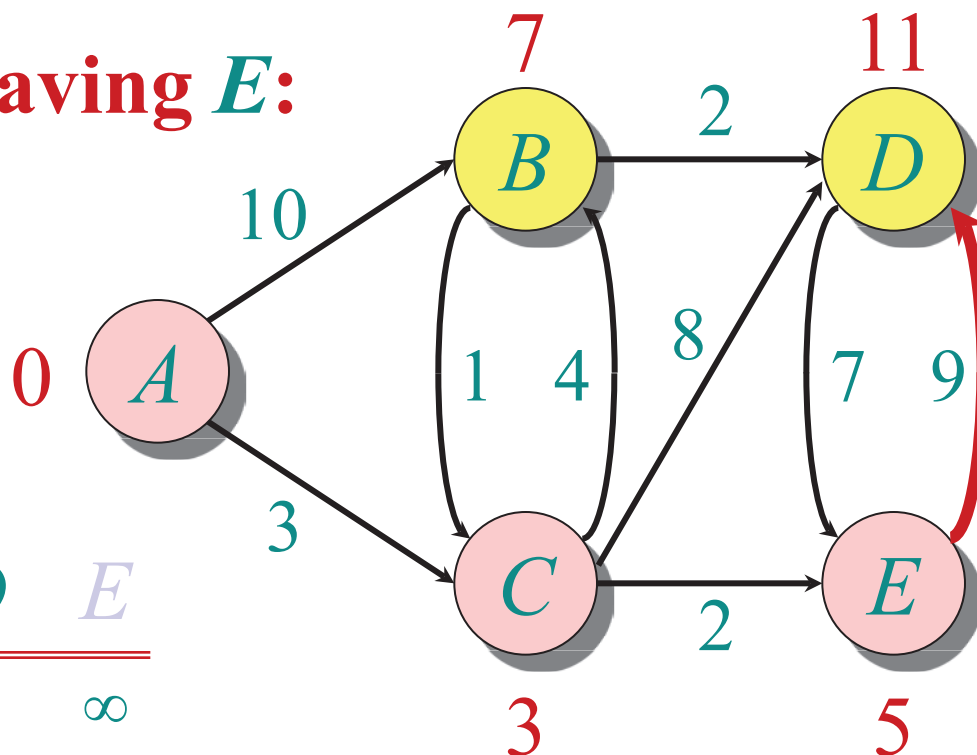
“*E*”  $\leftarrow$  **EXTRACT-MIN**(*Q*):





# Example of Dijkstra's algorithm

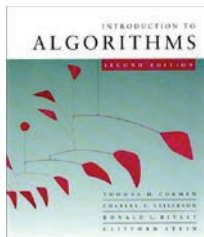
Relax all edges leaving  $E$ :



$Q$ :

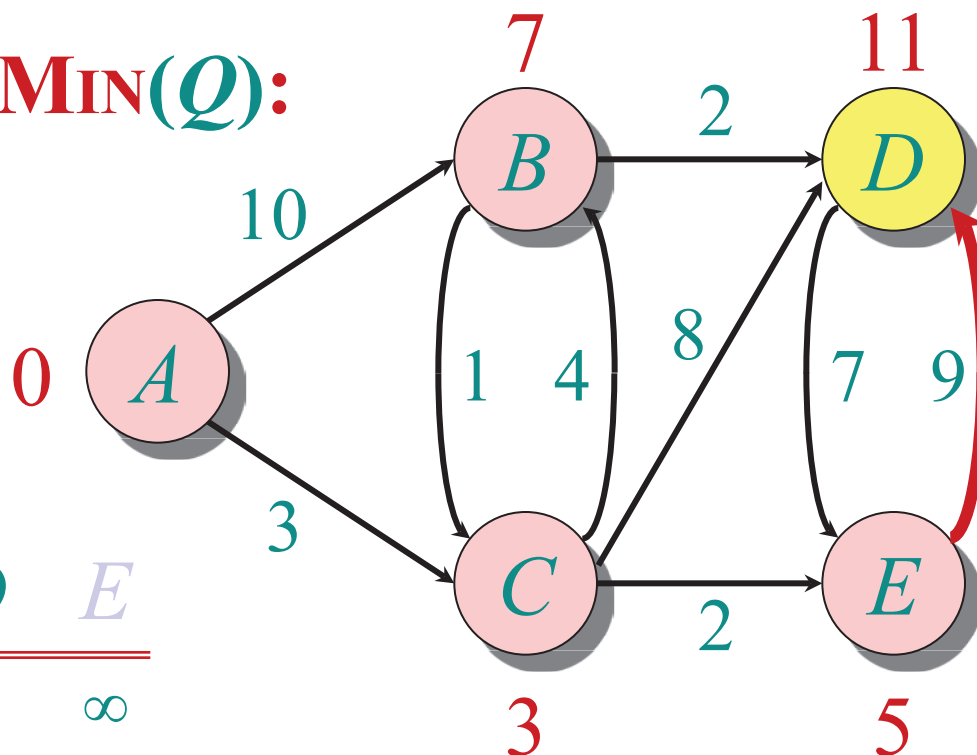
$A$	$B$	$C$	$D$	$E$
0	$\infty$	$\infty$	$\infty$	$\infty$
	10	3	$\infty$	$\infty$
	7		11	5
	7		11	

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



# Example of Dijkstra's algorithm

**"B" ← EXTRACT-MIN(Q):**

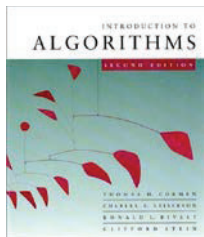


*Q:*

<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
0	∞	∞	∞	∞
	10	3	∞	∞
	7		11	5
	7		11	

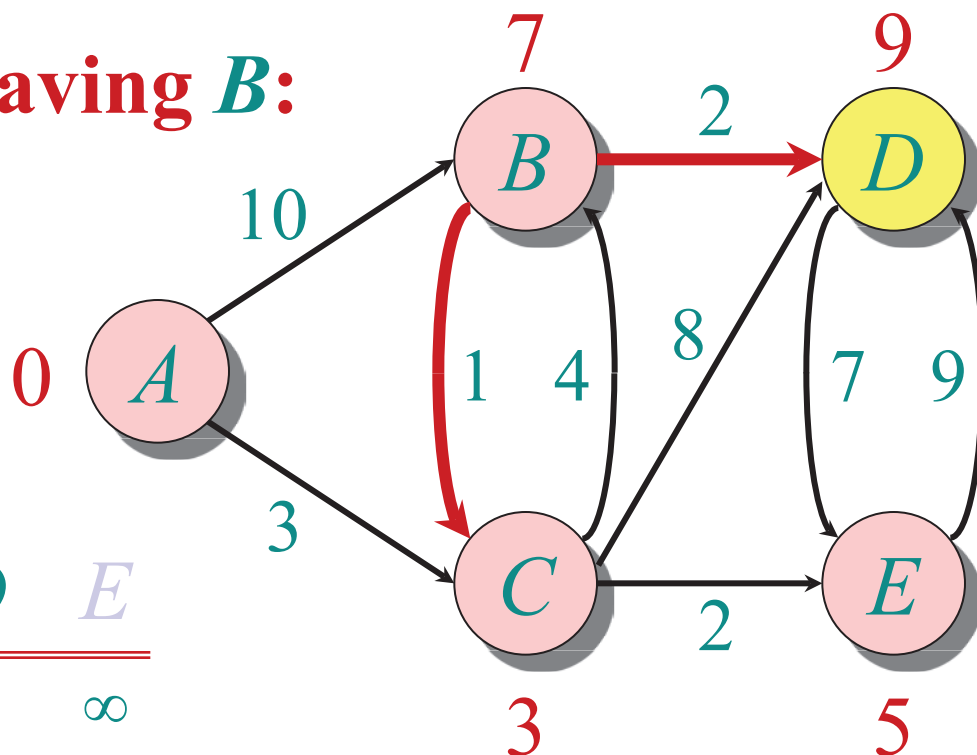
*S:* { *A*, *C*, *E*, *B* }





# Example of Dijkstra's algorithm

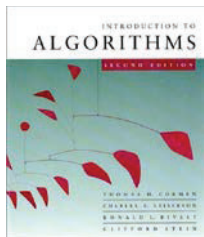
Relax all edges leaving **B**:



*Q*:

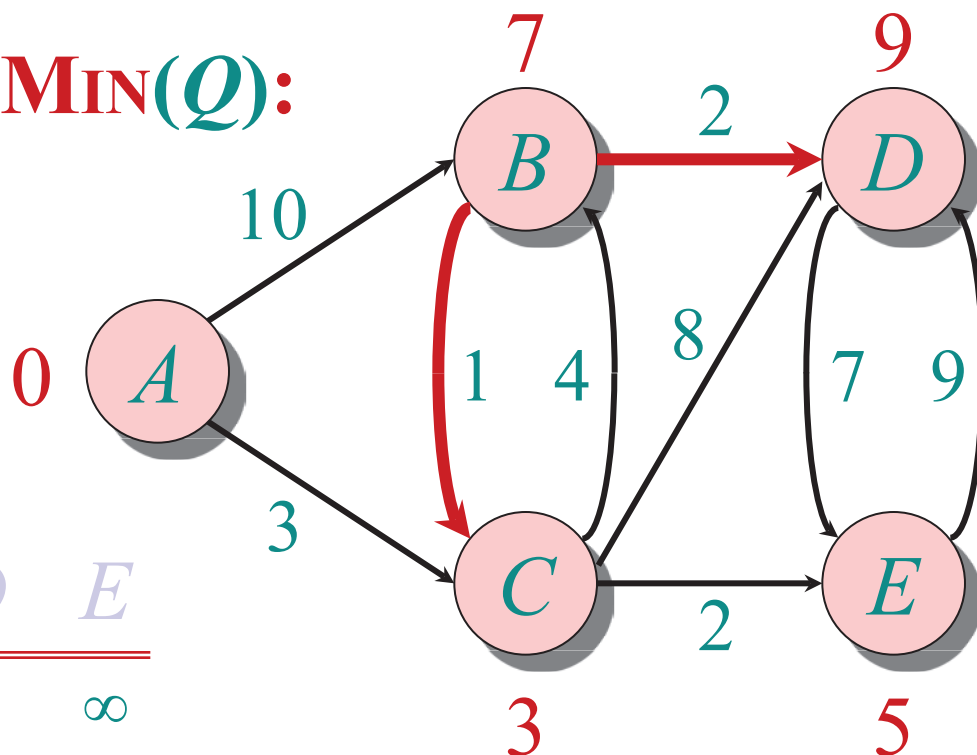
<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
0	$\infty$	$\infty$	$\infty$	$\infty$
	10	3	$\infty$	$\infty$
	7		11	5
	7		11	
			9	

*S*: { *A*, *C*, *E*, *B* }



# Example of Dijkstra's algorithm

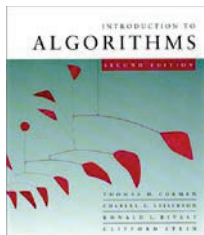
**"D"**  $\leftarrow$  **EXTRACT-MIN**(*Q*):



*Q*:

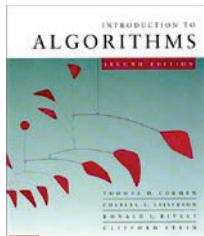
<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
0	$\infty$	$\infty$	$\infty$	$\infty$
	10	3	$\infty$	$\infty$
	7		11	5
	7		11	
			9	

*S*: { *A*, *C*, *E*, *B*, *D* }



# Correctness — Part I

**Lemma.** Initializing  $d[s] \leftarrow 0$  and  $d[v] \leftarrow \infty$  for all  $v \in V - \{s\}$  establishes  $d[v] \geq \delta(s, v)$  for all  $v \in V$ , and this invariant is maintained over any sequence of relaxation steps.



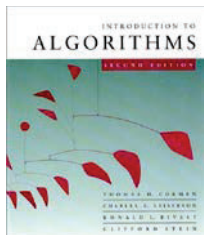
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*Proof.* Suppose not. Let  $v$  be the first vertex for which  $d[v] < \delta(s, v)$ , and let  $u$  be the vertex that caused  $d[v]$  to change:  $d[v] = d[u] + w(u, v)$ . Then,

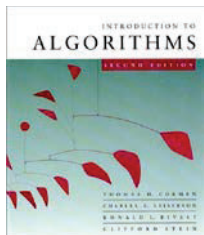
$d[v] < \delta(s, v)$	supposition
$\leq \delta(s, u) + \delta(u, v)$	triangle inequality
$\leq \delta(s, u) + w(u, v)$	sh. path $\leq$ specific path
$\leq d[u] + w(u, v)$	$v$ is first violation

Contradiction. □



## Correctness — Part II

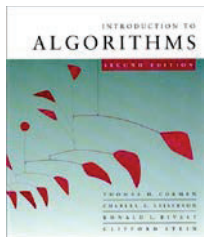
**Lemma.** Let  $u$  be  $v$ 's predecessor on a shortest path from  $s$  to  $v$ . Then, if  $d[u] = \delta(s, u)$  and edge  $(u, v)$  is relaxed, we have  $d[v] = \delta(s, v)$  after the relaxation.



## Correctness — Part II

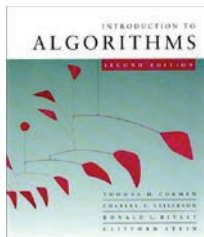
**Lemma.** Let  $u$  be  $v$ 's predecessor on a shortest path from  $s$  to  $v$ . Then, if  $d[u] = \delta(s, u)$  and edge  $(u, v)$  is relaxed, we have  $d[v] = \delta(s, v)$  after the relaxation.

*Proof.* Observe that  $\delta(s, v) = \delta(s, u) + w(u, v)$ . Suppose that  $d[v] > \delta(s, v)$  before the relaxation. (Otherwise, we're done.) Then, the test  $d[v] > d[u] + w(u, v)$  succeeds, because  $d[v] > \delta(s, v) = \delta(s, u) + w(u, v) = d[u] + w(u, v)$ , and the algorithm sets  $d[v] = d[u] + w(u, v) = \delta(s, v)$ . ◻



# Correctness — Part III

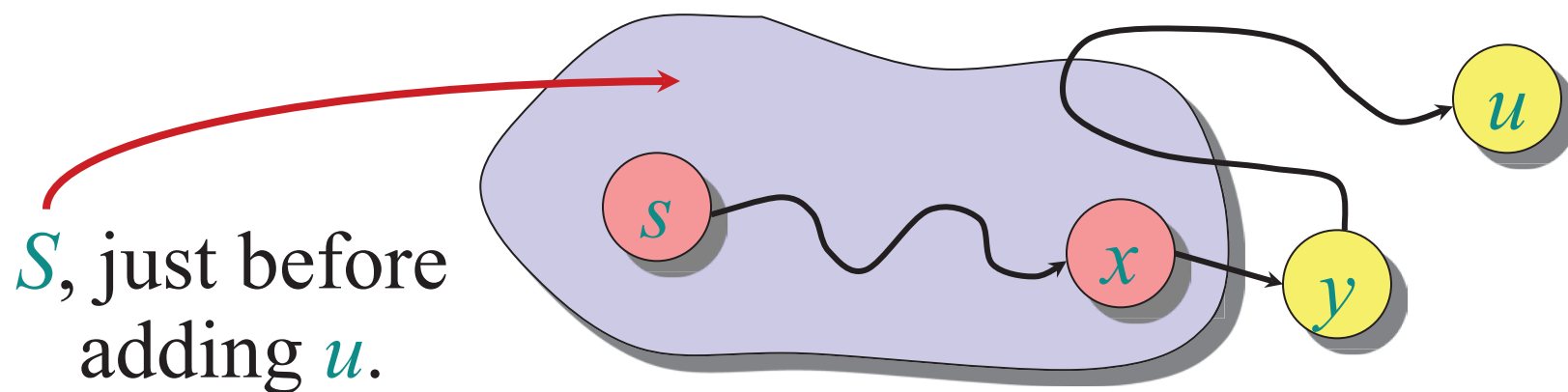
**Theorem.** Dijkstra's algorithm terminates with  $d[v] = \delta(s, v)$  for all  $v \in V$ .



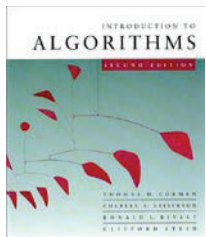
# Correctness — Part III

**Theorem.** Dijkstra's algorithm terminates with  $d[v] = \delta(s, v)$  for all  $v \in V$ .

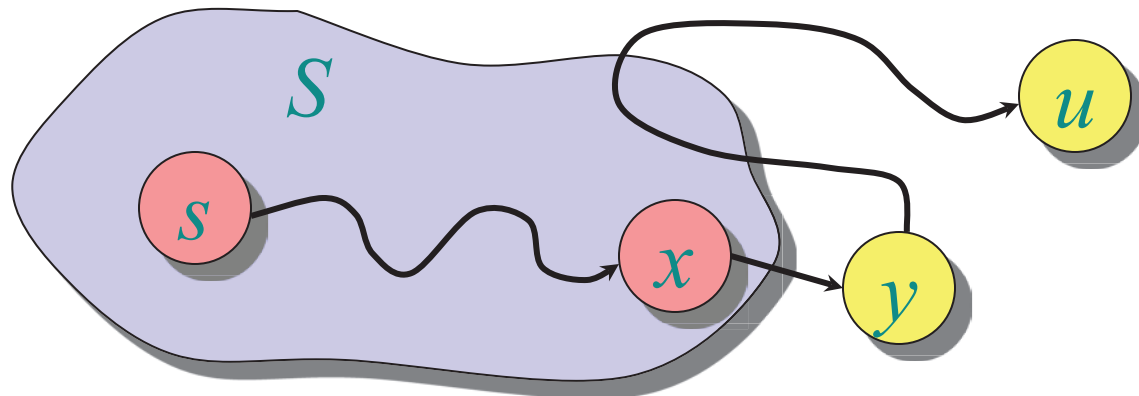
*Proof.* It suffices to show that  $d[v] = \delta(s, v)$  for every  $v \in V$  when  $v$  is added to  $S$ . Suppose  $u$  is the first vertex added to  $S$  for which  $d[u] > \delta(s, u)$ . Let  $y$  be the first vertex in  $V - S$  along a shortest path from  $s$  to  $u$ , and let  $x$  be its predecessor:



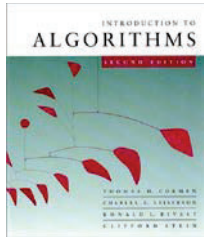




# Correctness — Part III (continued)

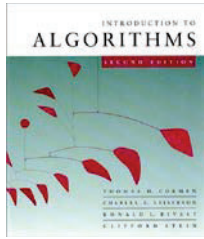


Since  $u$  is the first vertex violating the claimed invariant, we have  $d[x] = \delta(s, x)$ . When  $x$  was added to  $S$ , the edge  $(x, y)$  was relaxed, which implies that  $d[y] = \delta(s, y) \leq \delta(s, u) < d[u]$ . But,  $d[u] \leq d[y]$  by our choice of  $u$ . Contradiction.  $\square$



# Analysis of Dijkstra

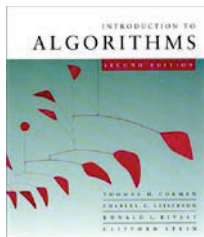
```
while  $Q \neq \emptyset$ 
do  $u \leftarrow \text{EXTRACT-MIN}(Q)$ 
   $S \leftarrow S \cup \{u\}$ 
  for each  $v \in \text{Adj}[u]$ 
    do if  $d[v] > d[u] + w(u, v)$ 
       then  $d[v] \leftarrow d[u] + w(u, v)$ 
```



# Analysis of Dijkstra

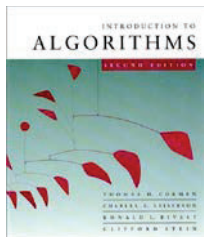
$|V|$   
times

```
while  $Q \neq \emptyset$ 
do  $u \leftarrow \text{EXTRACT-MIN}(Q)$ 
   $S \leftarrow S \cup \{u\}$ 
  for each  $v \in \text{Adj}[u]$ 
    do if  $d[v] > d[u] + w(u, v)$ 
       then  $d[v] \leftarrow d[u] + w(u, v)$ 
```



# Analysis of Dijkstra

$|V|$   
times {  $\text{degree}(u)$   
times { while  $Q \neq \emptyset$   
do  $u \leftarrow \text{EXTRACT-MIN}(Q)$   
 $S \leftarrow S \cup \{u\}$   
for each  $v \in \text{Adj}[u]$   
do if  $d[v] > d[u] + w(u, v)$   
then  $d[v] \leftarrow d[u] + w(u, v)$

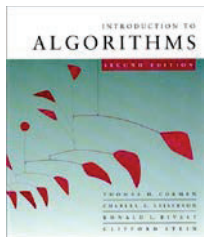


# Analysis of Dijkstra

$|V|$  times { while  $Q \neq \emptyset$   
do  $u \leftarrow \text{EXTRACT-MIN}(Q)$   
 $S \leftarrow S \cup \{u\}$   
for each  $v \in \text{Adj}[u]$   
do if  $d[v] > d[u] + w(u, v)$   
then  $d[v] \leftarrow d[u] + w(u, v)$

$\text{degree}(u)$  times {

Handshaking Lemma  $\Rightarrow \Theta(E)$  implicit DECREASE-KEY's.



# Analysis of Dijkstra

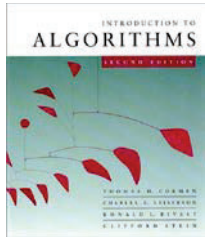
$|V|$  times { while  $Q \neq \emptyset$   
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for each  $v \in \text{Adj}[u]$   
do if  $d[v] > d[u] + w(u, v)$   
then  $d[v] \leftarrow d[u] + w(u, v)$

$\text{degree}(u)$  times {

Handshaking Lemma  $\Rightarrow \Theta(E)$  implicit DECREASE-KEY's.

$$\text{Time} = \Theta(V \cdot T_{\text{EXTRACT-MIN}} + E \cdot T_{\text{DECREASE-KEY}})$$

**Note:** Same formula as in the analysis of Prim's minimum spanning tree algorithm.

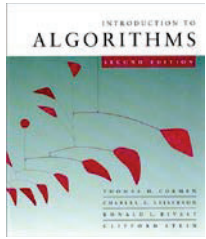


# Analysis of Dijkstra (continued)

$$\text{Time} = \Theta(V) \cdot T_{\text{EXTRACT-MIN}} + \Theta(E) \cdot T_{\text{DECREASE-KEY}}$$

$Q$	$T_{\text{EXTRACT-MIN}}$	$T_{\text{DECREASE-KEY}}$	Total
-----	--------------------------	---------------------------	-------

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# Analysis of Dijkstra (continued)

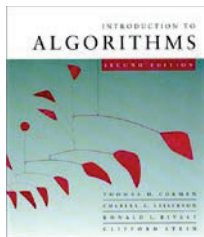
$$\text{Time} = \Theta(V) \cdot T_{\text{EXTRACT-MIN}} + \Theta(E) \cdot T_{\text{DECREASE-KEY}}$$

$Q$	$T_{\text{EXTRACT-MIN}}$	$T_{\text{DECREASE-KEY}}$	Total
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array	$O(V)$	$O(1)$	$O(V^2)$
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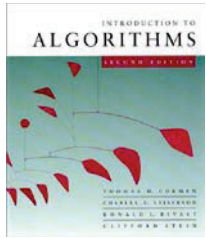




# Analysis of Dijkstra (continued)

$$\text{Time} = \Theta(V) \cdot T_{\text{EXTRACT-MIN}} + \Theta(E) \cdot T_{\text{DECREASE-KEY}}$$

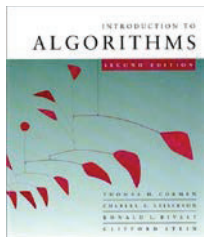
$Q$	$T_{\text{EXTRACT-MIN}}$	$T_{\text{DECREASE-KEY}}$	Total
array	$O(V)$	$O(1)$	$O(V^2)$
binary heap	$O(\lg V)$	$O(\lg V)$	$O(E \lg V)$



# Analysis of Dijkstra (continued)

$$\text{Time} = \Theta(V) \cdot T_{\text{EXTRACT-MIN}} + \Theta(E) \cdot T_{\text{DECREASE-KEY}}$$

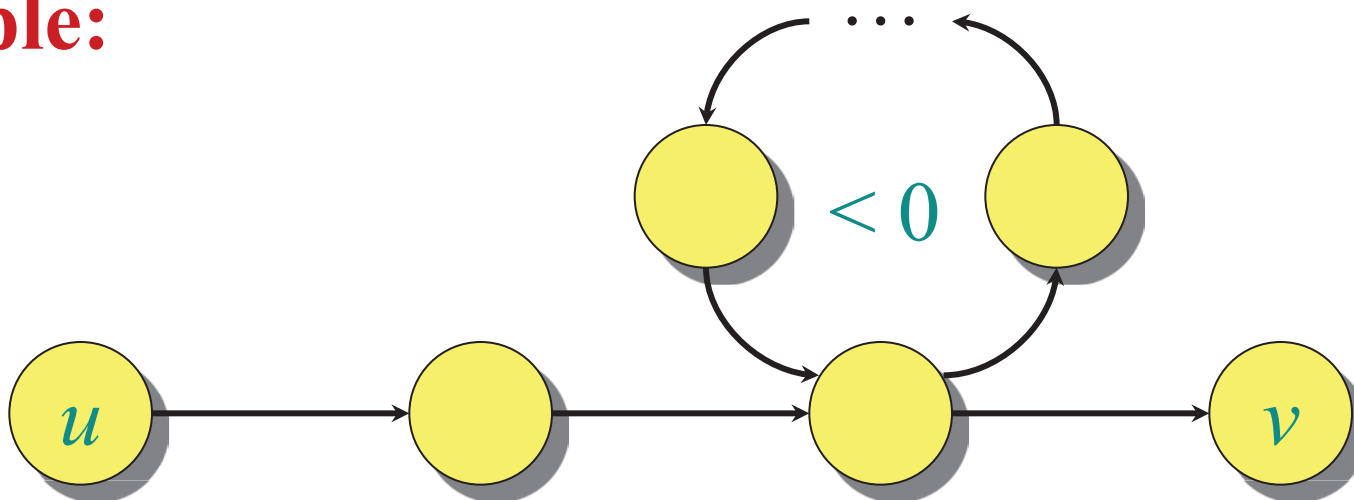
$Q$	$T_{\text{EXTRACT-MIN}}$	$T_{\text{DECREASE-KEY}}$	Total
array	$O(V)$	$O(1)$	$O(V^2)$
binary heap	$O(\lg V)$	$O(\lg V)$	$O(E \lg V)$
Fibonacci heap	$O(\lg V)$ amortized	$O(1)$ amortized	$O(E + V \lg V)$ worst case

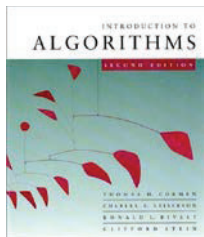


# Negative-weight cycles

**Recall:** If a graph  $G = (V, E)$  contains a negative-weight cycle, then some shortest paths may not exist.

**Example:**

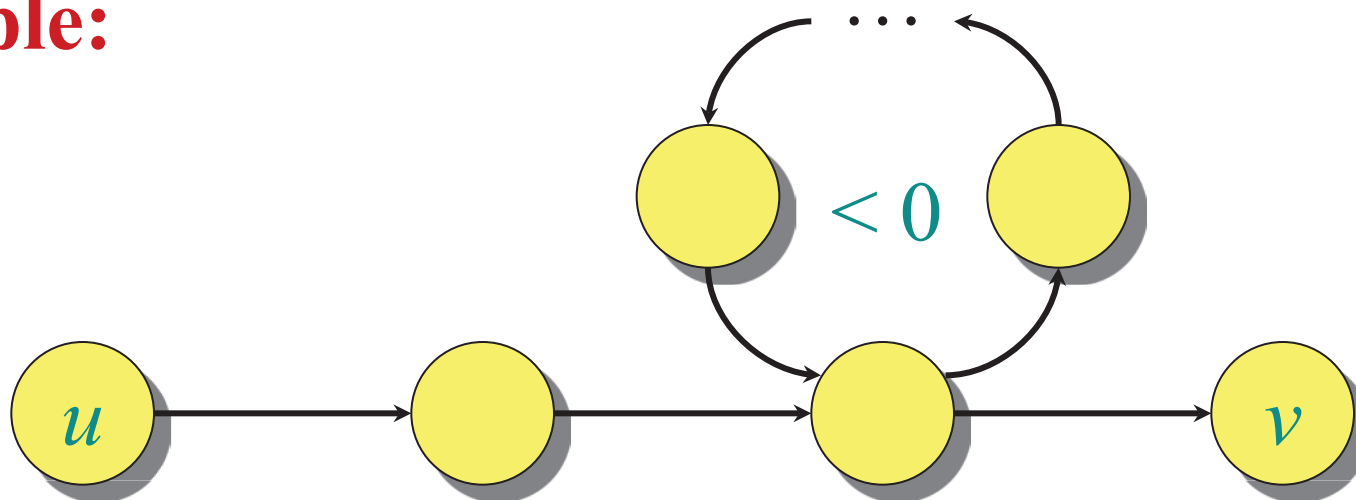




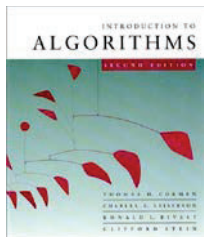
# Negative-weight cycles

**Recall:** If a graph  $G = (V, E)$  contains a negative-weight cycle, then some shortest paths may not exist.

**Example:**



**Bellman-Ford algorithm:** Finds all shortest-path lengths from a **source**  $s \in V$  to all  $v \in V$  or determines that a negative-weight cycle exists.



# Bellman-Ford algorithm

```
 $d[s] \leftarrow 0$   
for each  $v \in V - \{s\}$   
  do  $d[v] \leftarrow \infty$ 
```

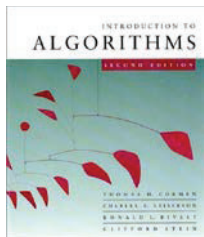
} initialization

```
for  $i \leftarrow 1$  to  $|V| - 1$   
  do for each edge  $(u, v) \in E$   
    do if  $d[v] > d[u] + w(u, v)$   
      then  $d[v] \leftarrow d[u] + w(u, v)$ 
```

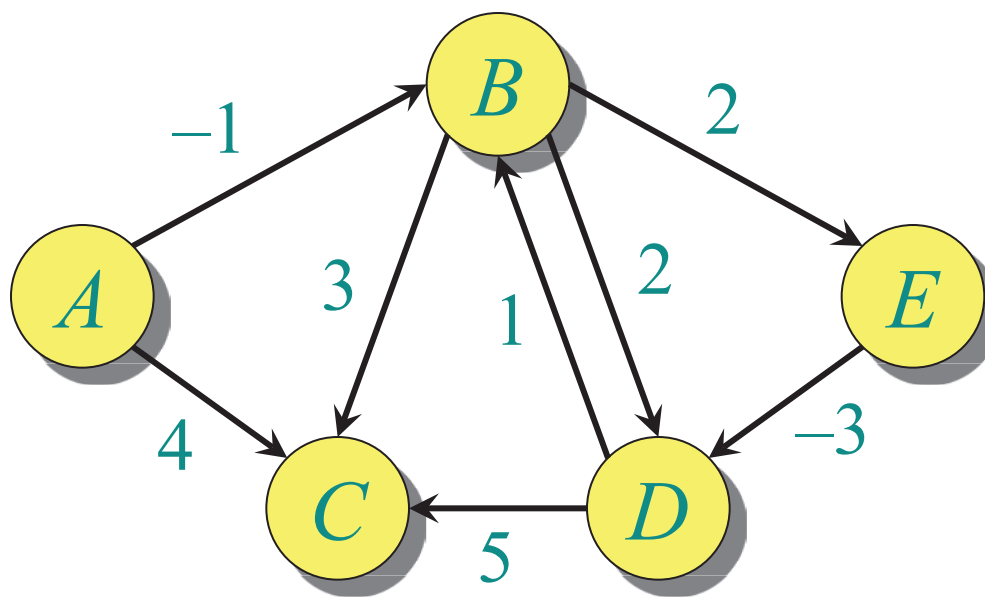
} *relaxation step*

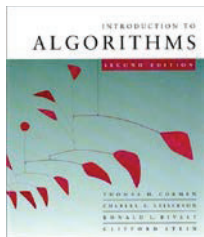
```
for each edge  $(u, v) \in E$   
  do if  $d[v] > d[u] + w(u, v)$   
    then report that a negative-weight cycle exists
```

At the end,  $d[v] = \delta(s, v)$ , if no negative-weight cycles.  
Time =  $O(VE)$ .

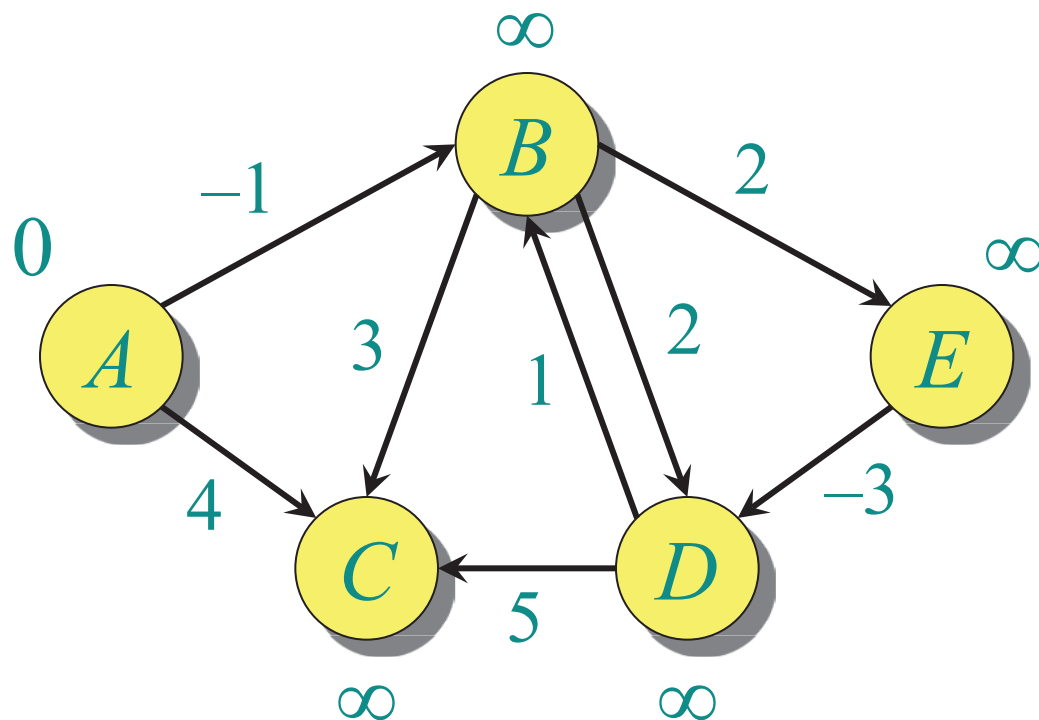


# Example of Bellman-Ford

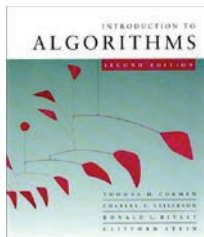




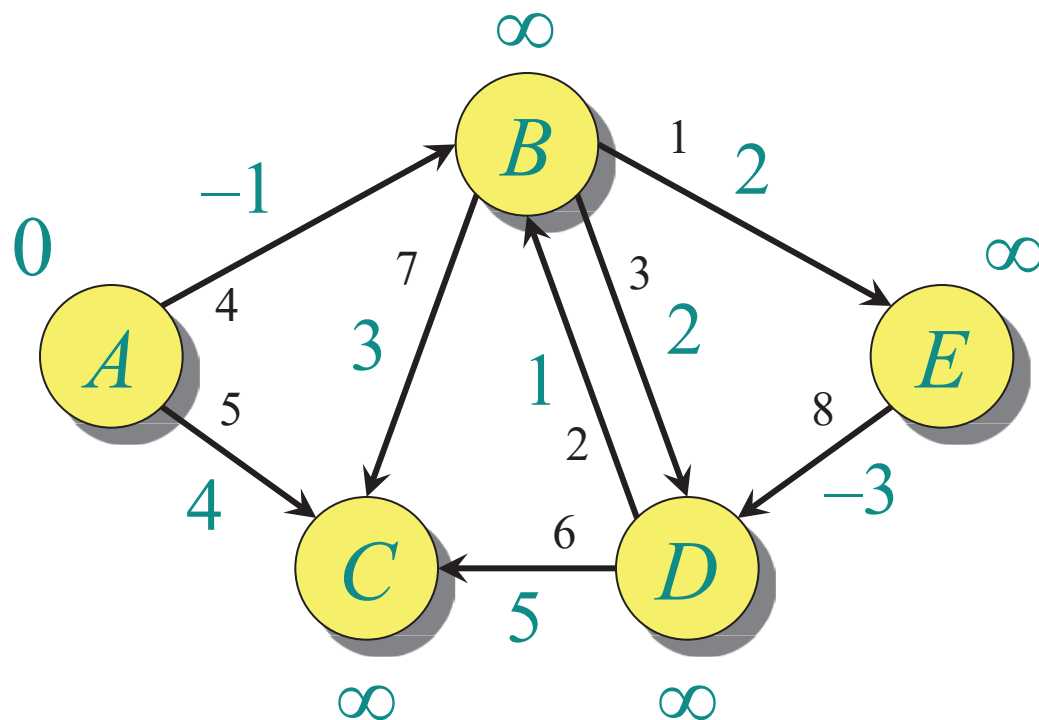
# Example of Bellman-Ford



Initialization.

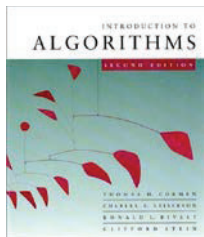


# Example of Bellman-Ford

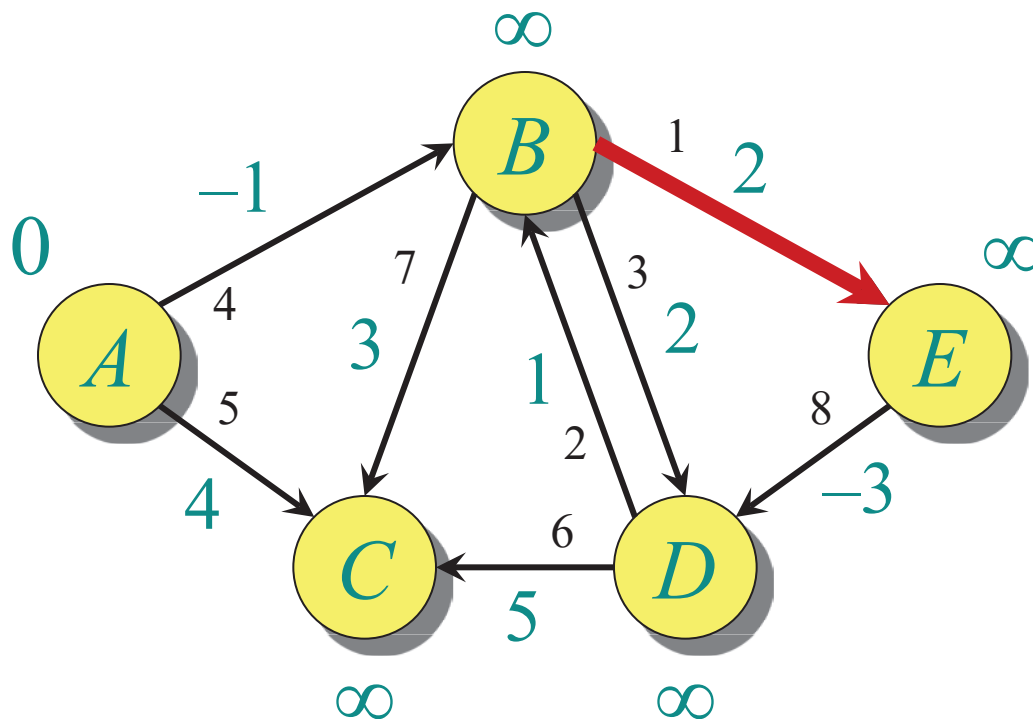


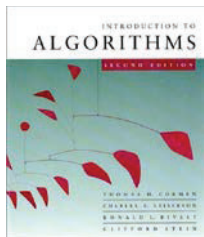
Order of edge relaxation.



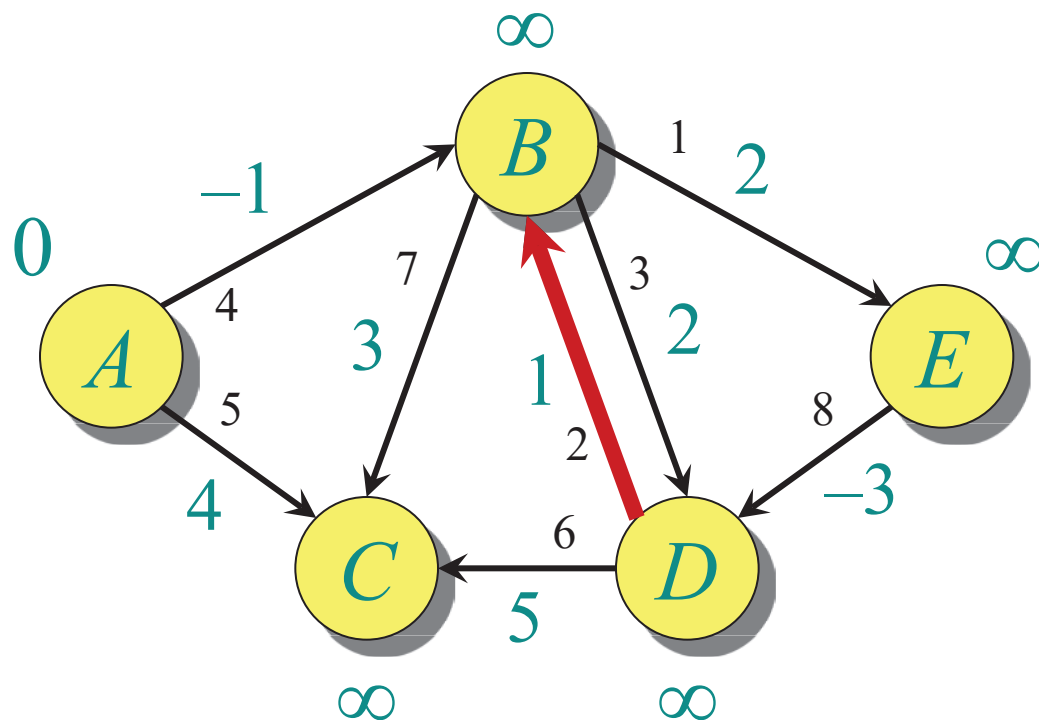


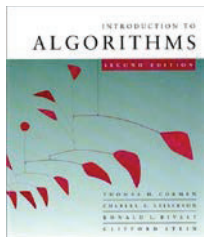
# Example of Bellman-Ford



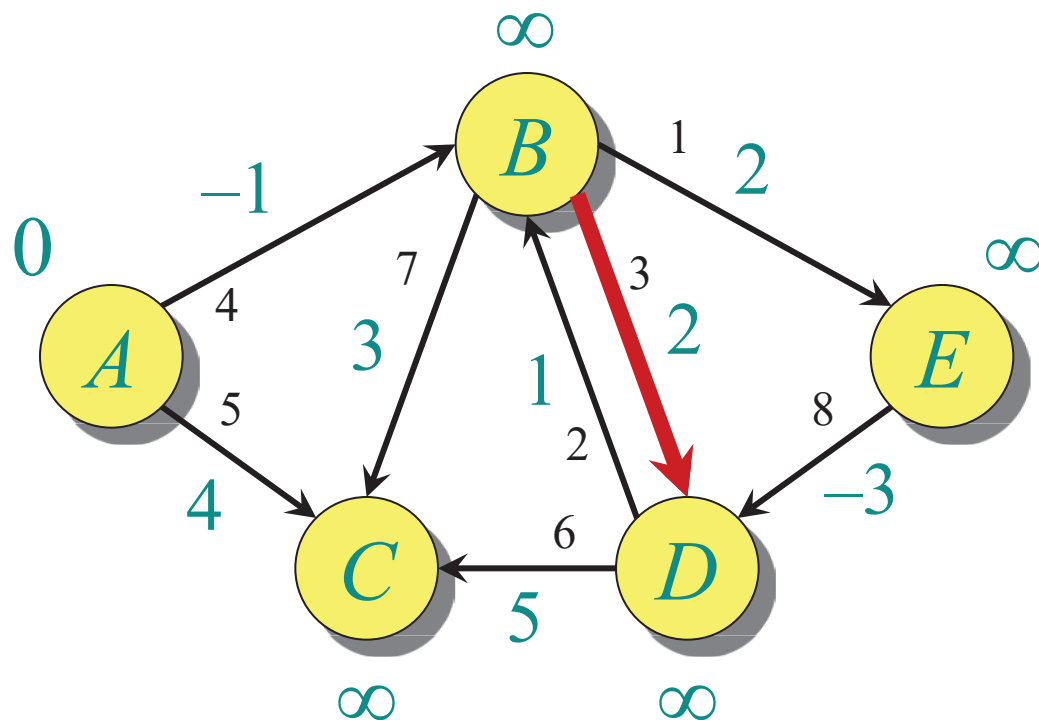


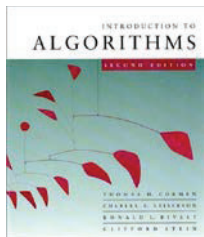
# Example of Bellman-Ford



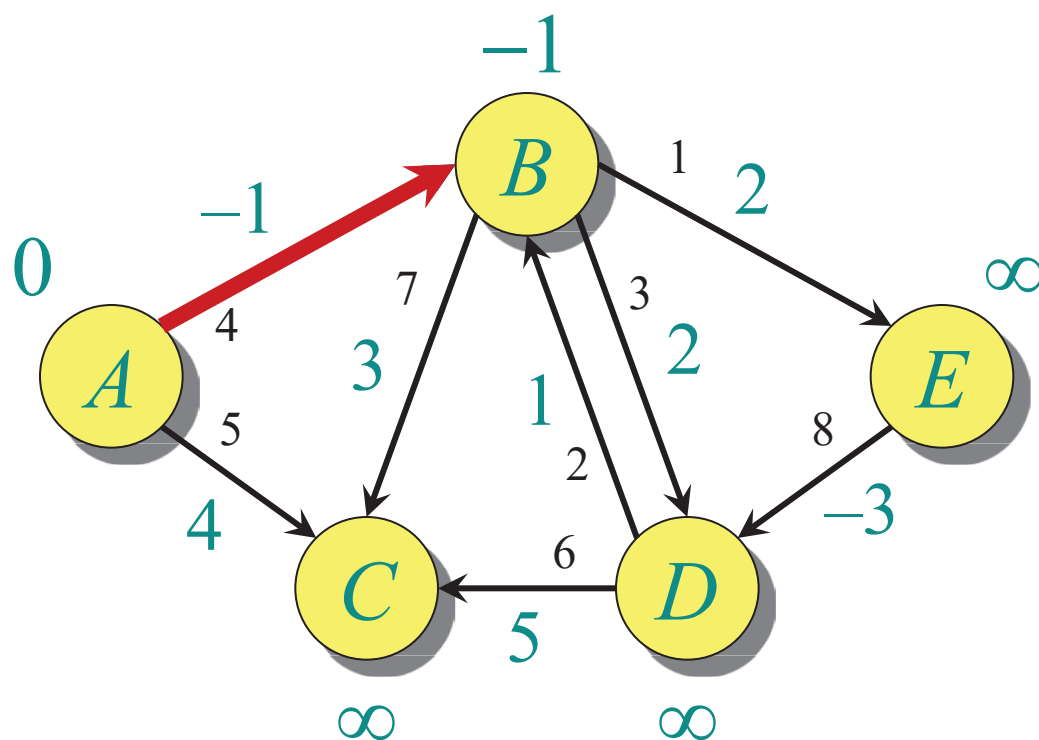


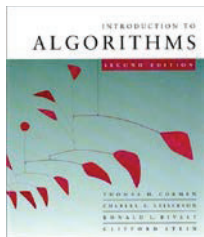
# Example of Bellman-Ford



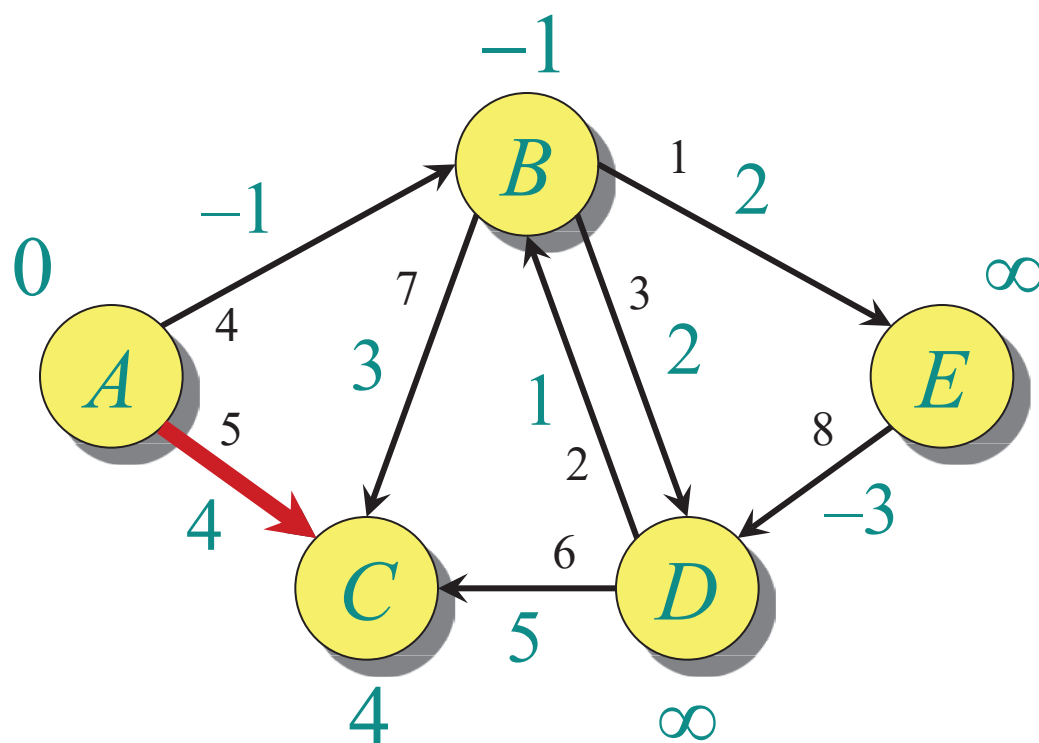


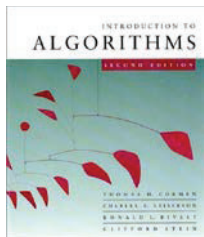
# Example of Bellman-Ford



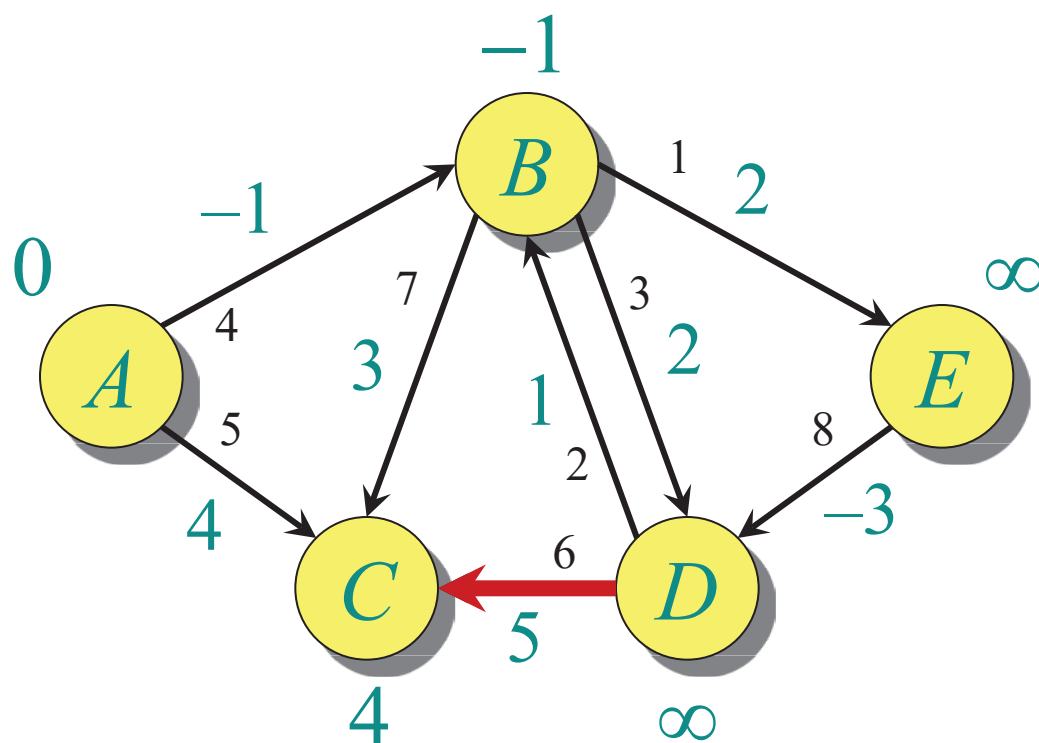


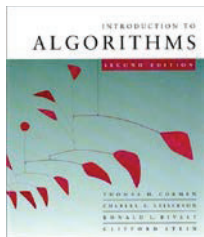
# Example of Bellman-Ford



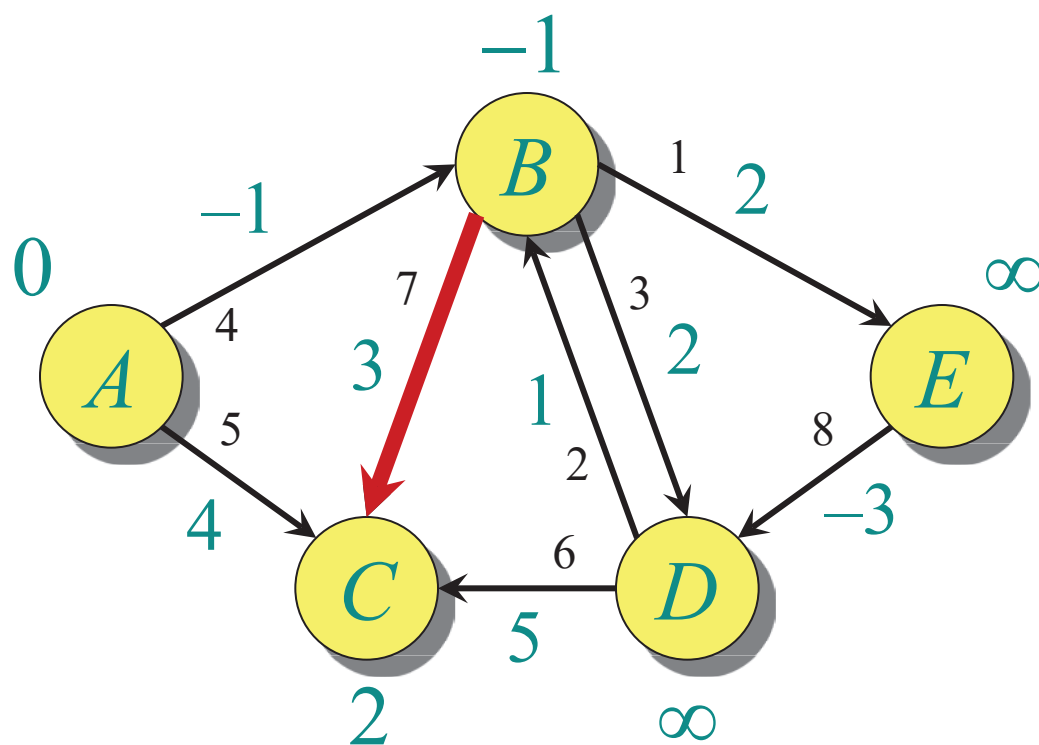


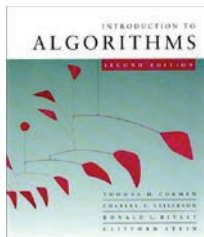
# Example of Bellman-Ford



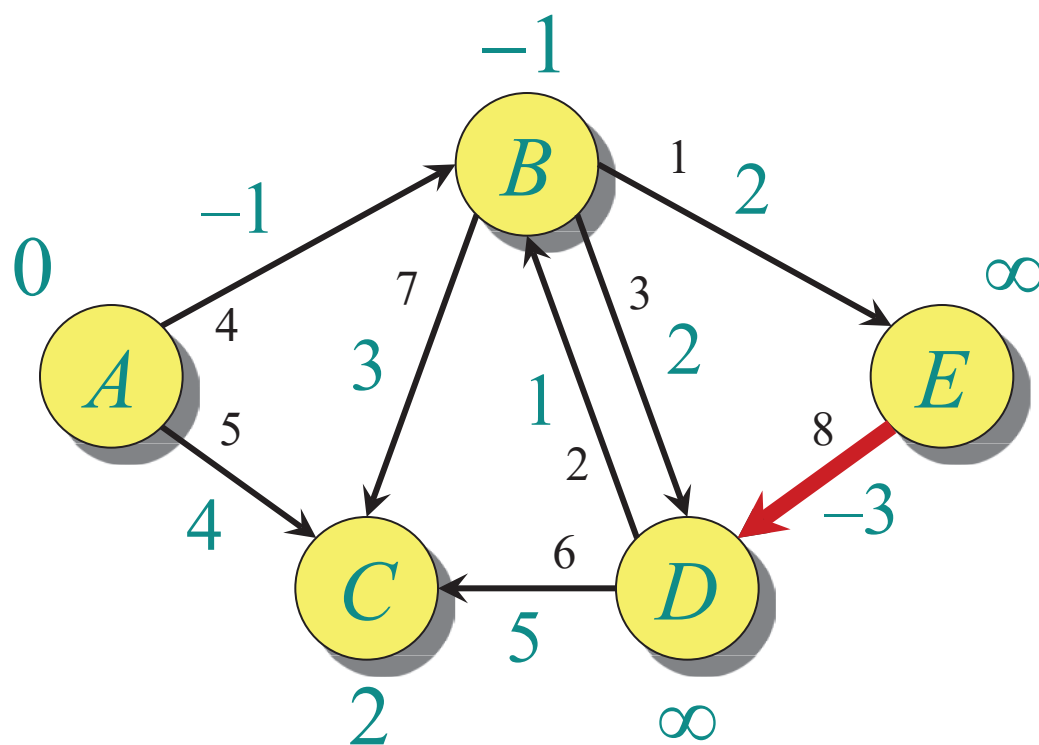


# Example of Bellman-Ford

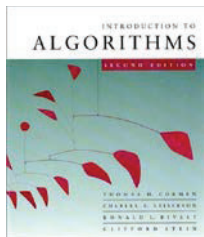




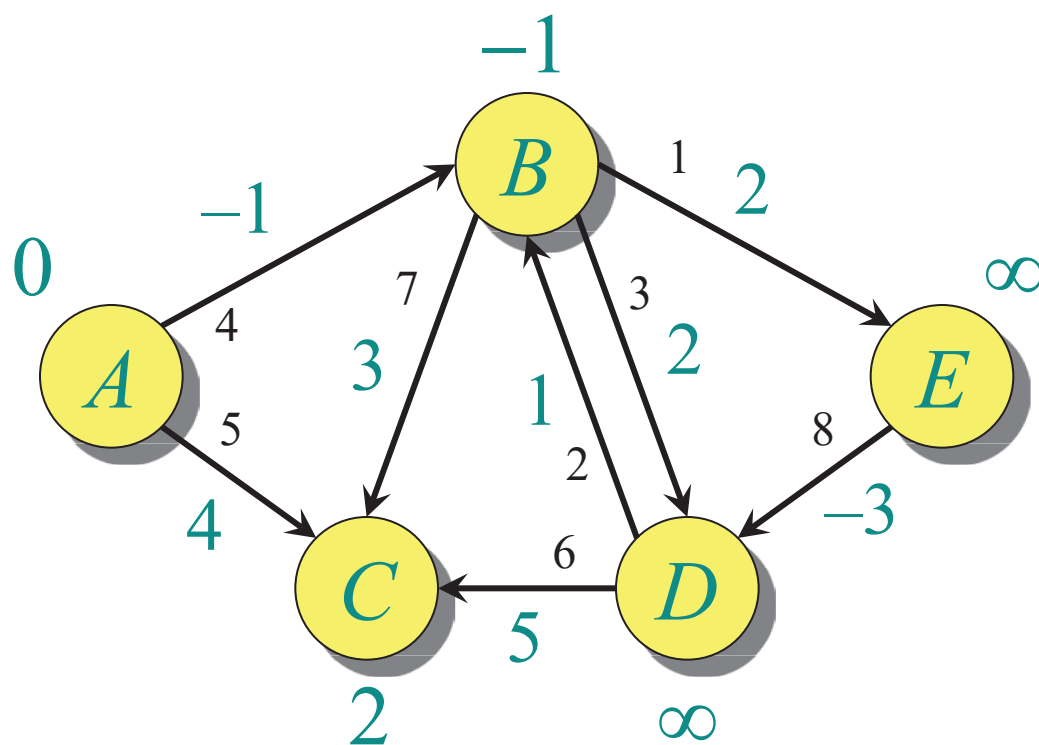
# Example of Bellman-Ford



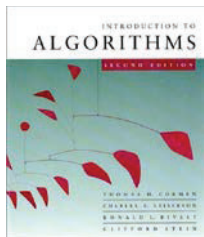




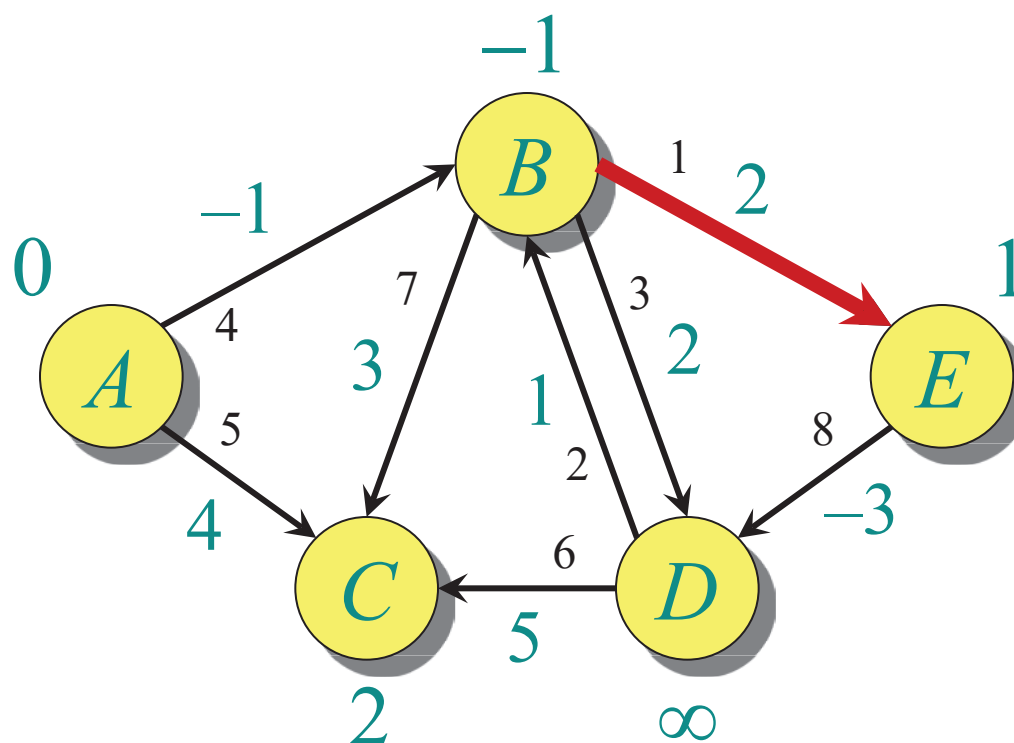
# Example of Bellman-Ford

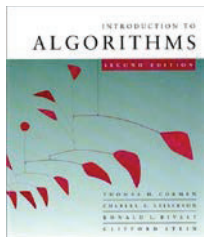


End of pass 1.

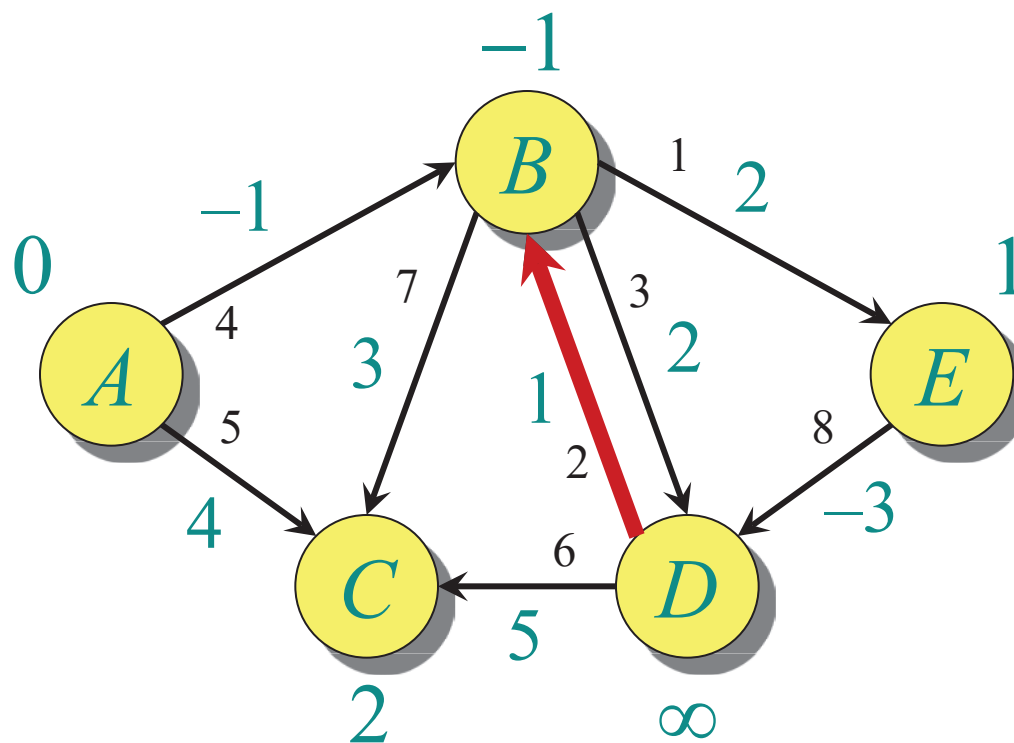


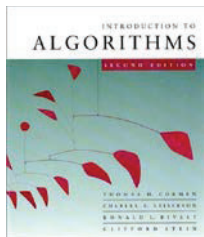
# Example of Bellman-Ford



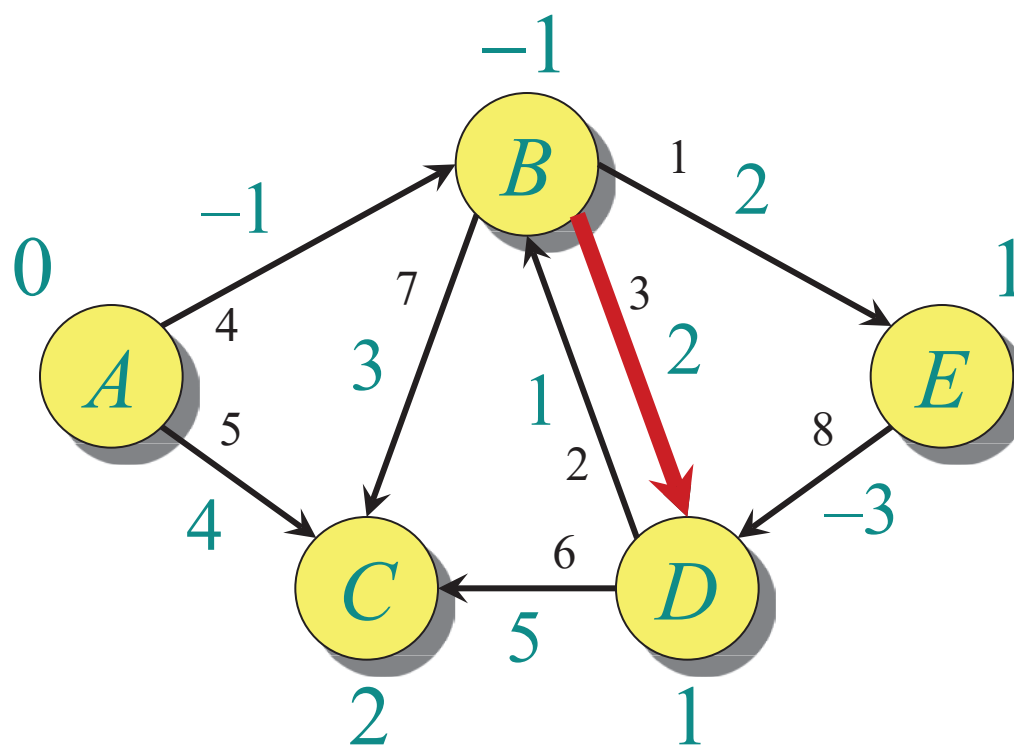


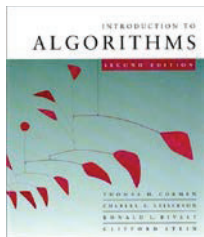
# Example of Bellman-Ford



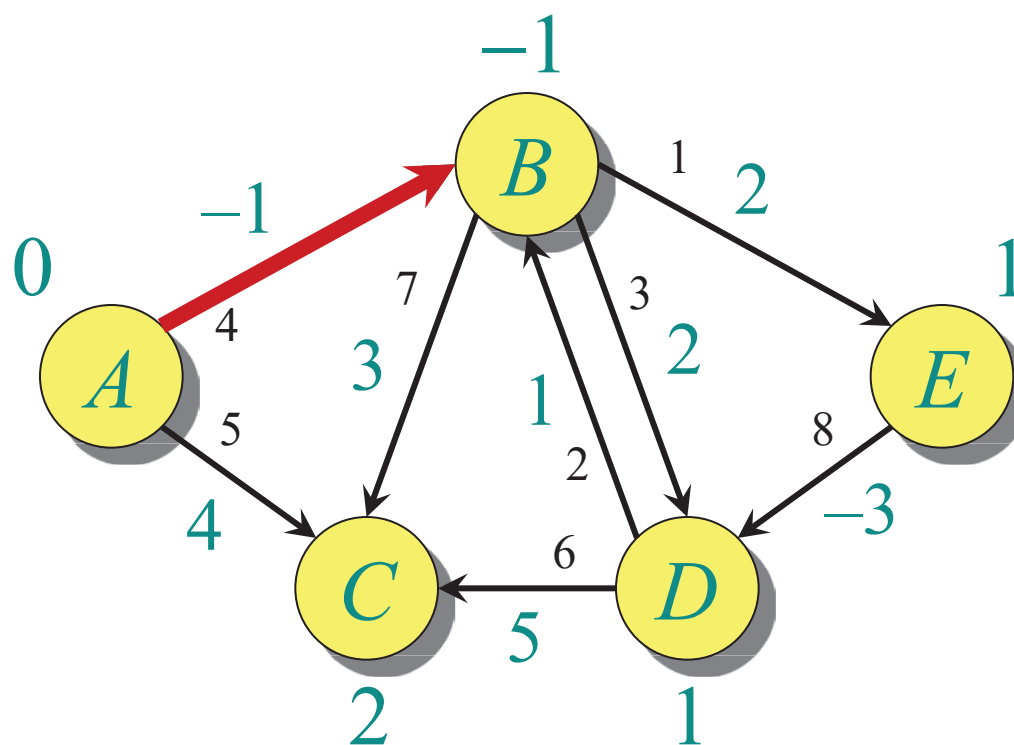


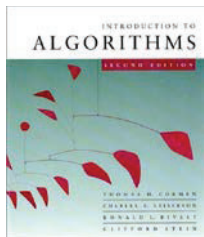
# Example of Bellman-Ford



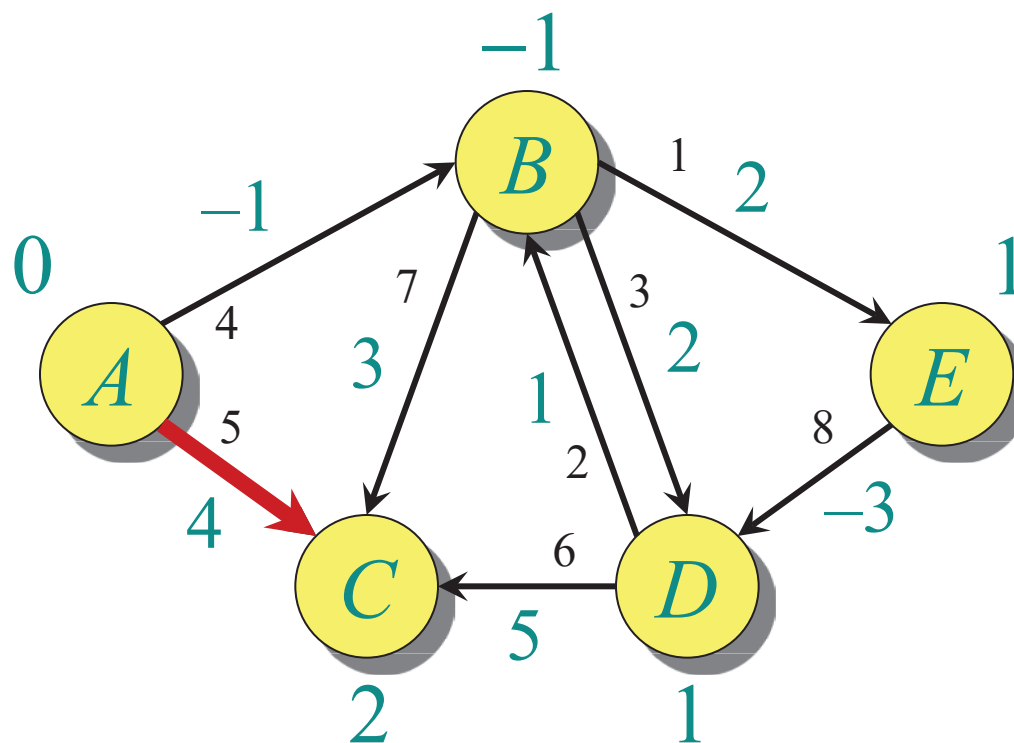


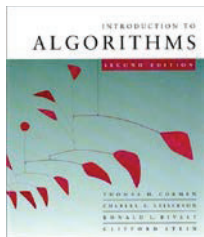
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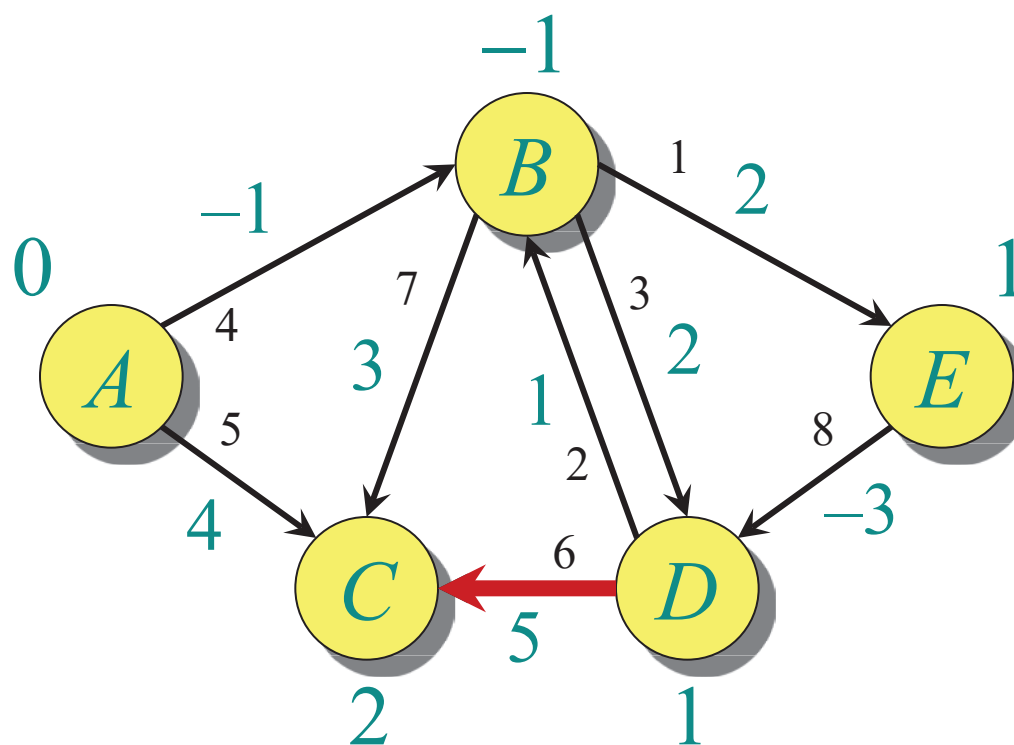


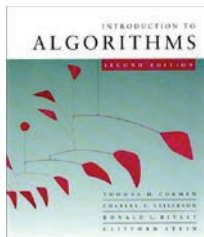
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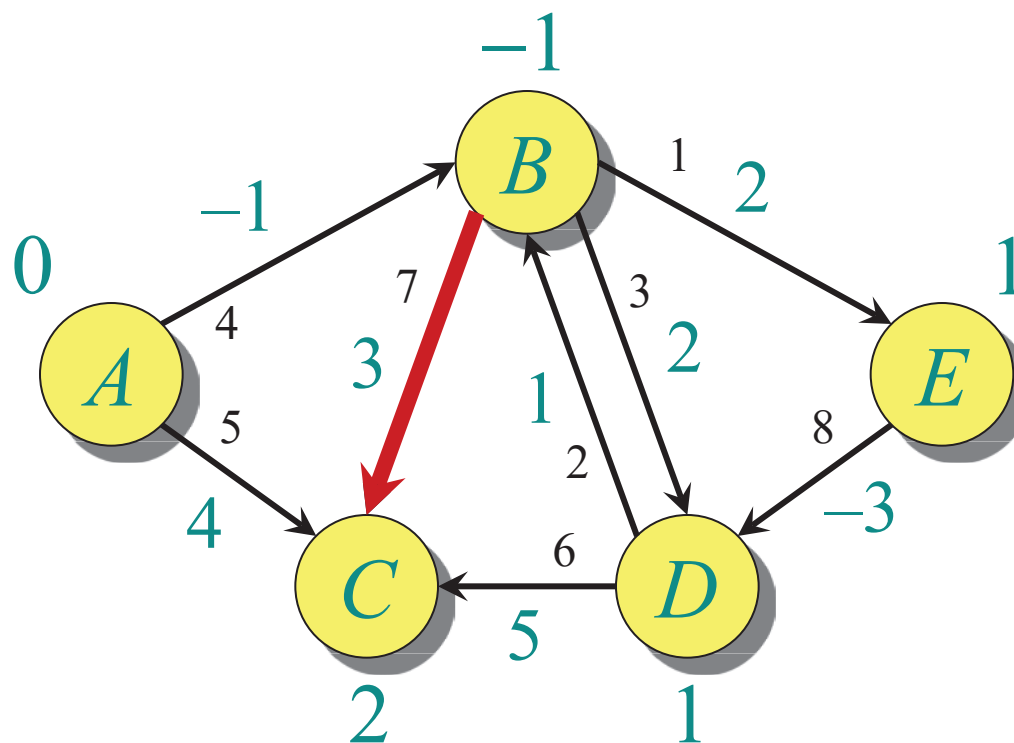


# Example of Bellman-Ford

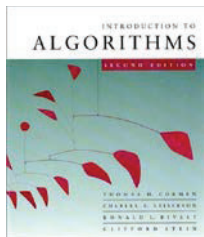




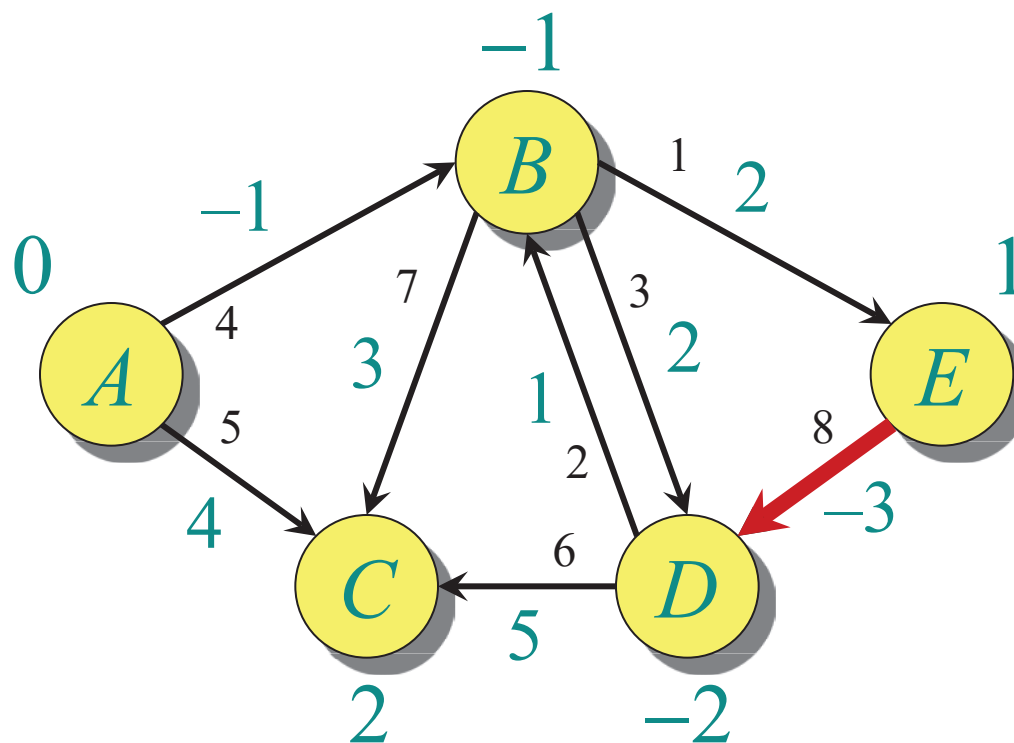
# Example of Bellman-Ford

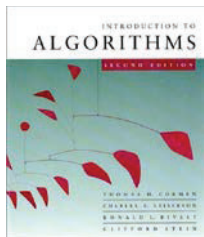




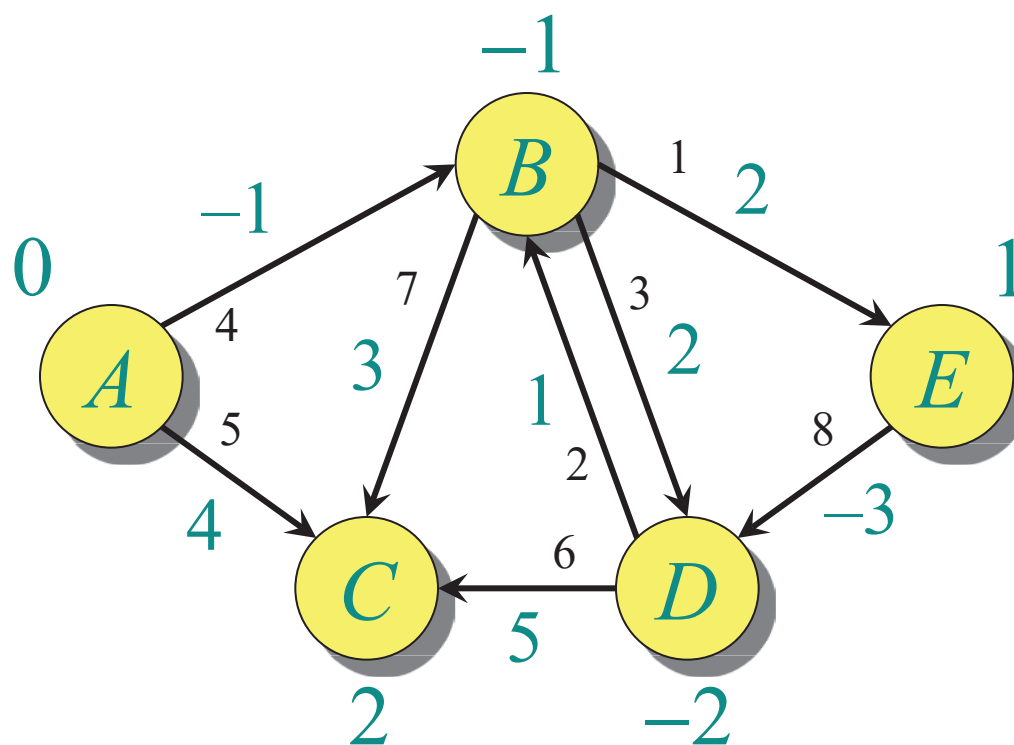


# Example of Bellman-Ford

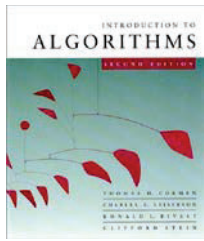




# Example of Bellman-Ford

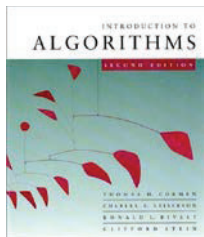


End of pass 2 (and 3 and 4).



# Correctness

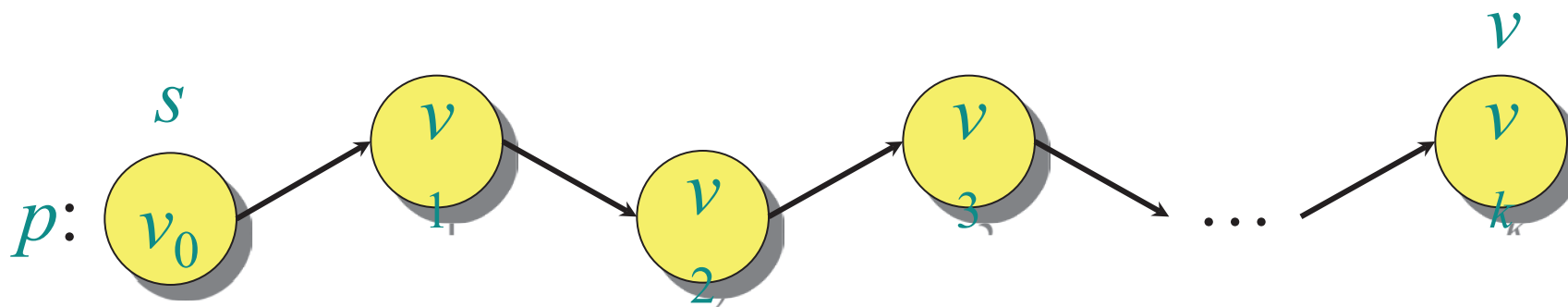
**Theorem.** If  $G = (V, E)$  contains no negative-weight cycles, then after the Bellman-Ford algorithm executes,  $d[v] = \delta(s, v)$  for all  $v \in V$ .



# Correctness

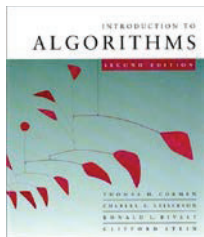
**Theorem.** If  $G = (V, E)$  contains no negative-weight cycles, then after the Bellman-Ford algorithm executes,  $d[v] = \delta(s, v)$  for all  $v \in V$ .

*Proof.* Let  $v \in V$  be any vertex, and consider a shortest path  $p$  from  $s$  to  $v$  with the minimum number of edges.

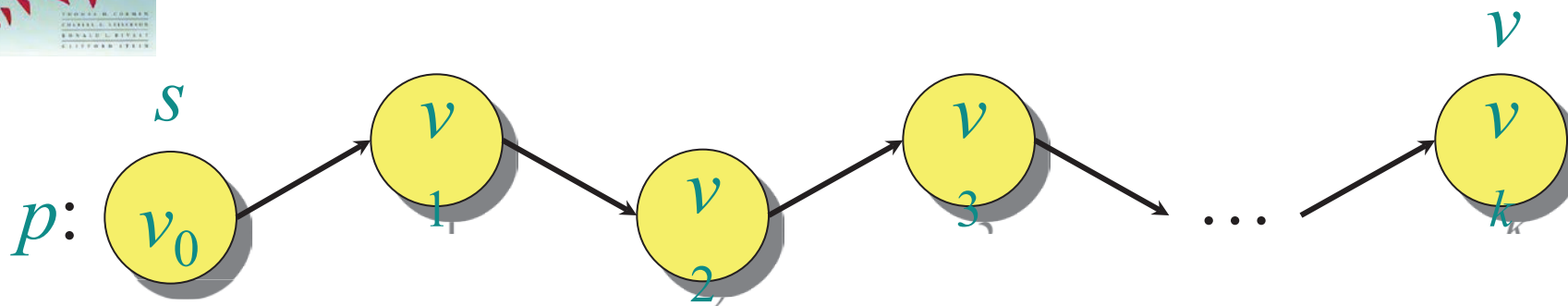


Since  $p$  is a shortest path, we have

$$\delta(s, v_i) = \delta(s, v_{i-1}) + w(v_{i-1}, v_i) .$$



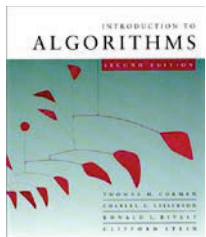
# Correctness (continued)



Initially,  $d[v_0] = 0 = \delta(s, v_0)$ , and  $d[v_0]$  is unchanged by subsequent relaxations (because of the lemma from *Shortest Paths I* that  $d[v] \geq \delta(s, v)$ ).

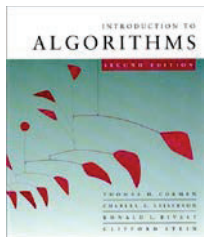
- After 1 pass through  $E$ , we have  $d[v_1] = \delta(s, v_1)$ .
- After 2 passes through  $E$ , we have  $d[v_2] = \delta(s, v_2)$ .
- $\vdots$
- After  $k$  passes through  $E$ , we have  $d[v_k] = \delta(s, v_k)$ .

Since  $G$  contains no negative-weight cycles,  $p$  is simple. Longest simple path has  $\leq |V| - 1$  edges.  $\square$



# Detection of negative-weight cycles

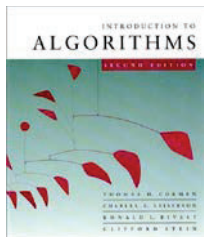
**Corollary.** If a value  $d[v]$  fails to converge after  $|V| - 1$  passes, there exists a negative-weight cycle in  $G$  reachable from  $s$ .  $\square$



# Shortest paths

## Single-source shortest paths

- Nonnegative edge weights
  - ♦ Dijkstra's algorithm:  $O(E + V \lg V)$
- General
  - ♦ Bellman-Ford algorithm:  $O(VE)$
- DAG
  - ♦ One pass of Bellman-Ford:  $O(V + E)$



# Shortest paths

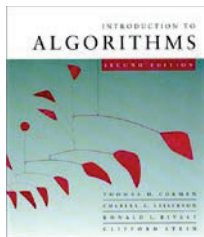
## Single-source shortest paths

- Nonnegative edge weights
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- General
  - ♦ Bellman-Ford algorithm:  $O(VE)$
- DAG
  - ♦ One pass of Bellman-Ford:  $O(V + E)$

## All-pairs shortest paths

- Nonnegative edge weights
  - ♦ Dijkstra's algorithm  $|V|$  times:  $O(VE + V^2 \lg V)$
- General
  - ♦ Three algorithms today.

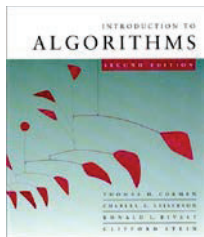




# All-pairs shortest paths

**Input:** Digraph  $G = (V, E)$ , where  $V = \{1, 2, \dots, n\}$ , with edge-weight function  $w : E \rightarrow \mathbb{R}$ .

**Output:**  $n \times n$  matrix of shortest-path lengths  $\delta(i, j)$  for all  $i, j \in V$ .



# All-pairs shortest paths

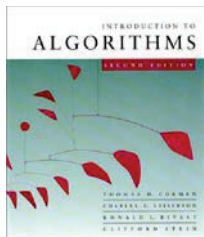
**Input:** Digraph  $G = (V, E)$ , where  $V = \{1, 2, \dots, n\}$ , with edge-weight function  $w : E \rightarrow \mathbb{R}$ .

**Output:**  $n \times n$  matrix of shortest-path lengths  $\delta(i, j)$  for all  $i, j \in V$ .

## IDEA:

- Run Bellman-Ford once from each vertex.
- Time =  $O(V^2E)$ .
- Dense graph ( $\Theta(n^2)$  edges)  $\Rightarrow \Theta(n^4)$  time in the worst case.

*Good first try!*



# Dynamic programming

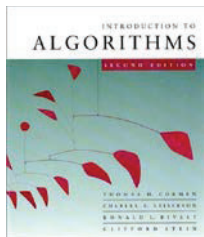
Consider the  $n \times n$  weighted adjacency matrix  $A = (a_{ij})$ , where  $a_{ij} = w(i, j)$  or  $\infty$ , and define  $d_{ij}^{(m)}$  = weight of a shortest path from  $i$  to  $j$  that uses at most  $m$  edges.

**Claim:** We have

$$d_{ij}^{(0)} = \begin{cases} 0 & \text{if } i = j, \\ \infty & \text{if } i \neq j; \end{cases}$$

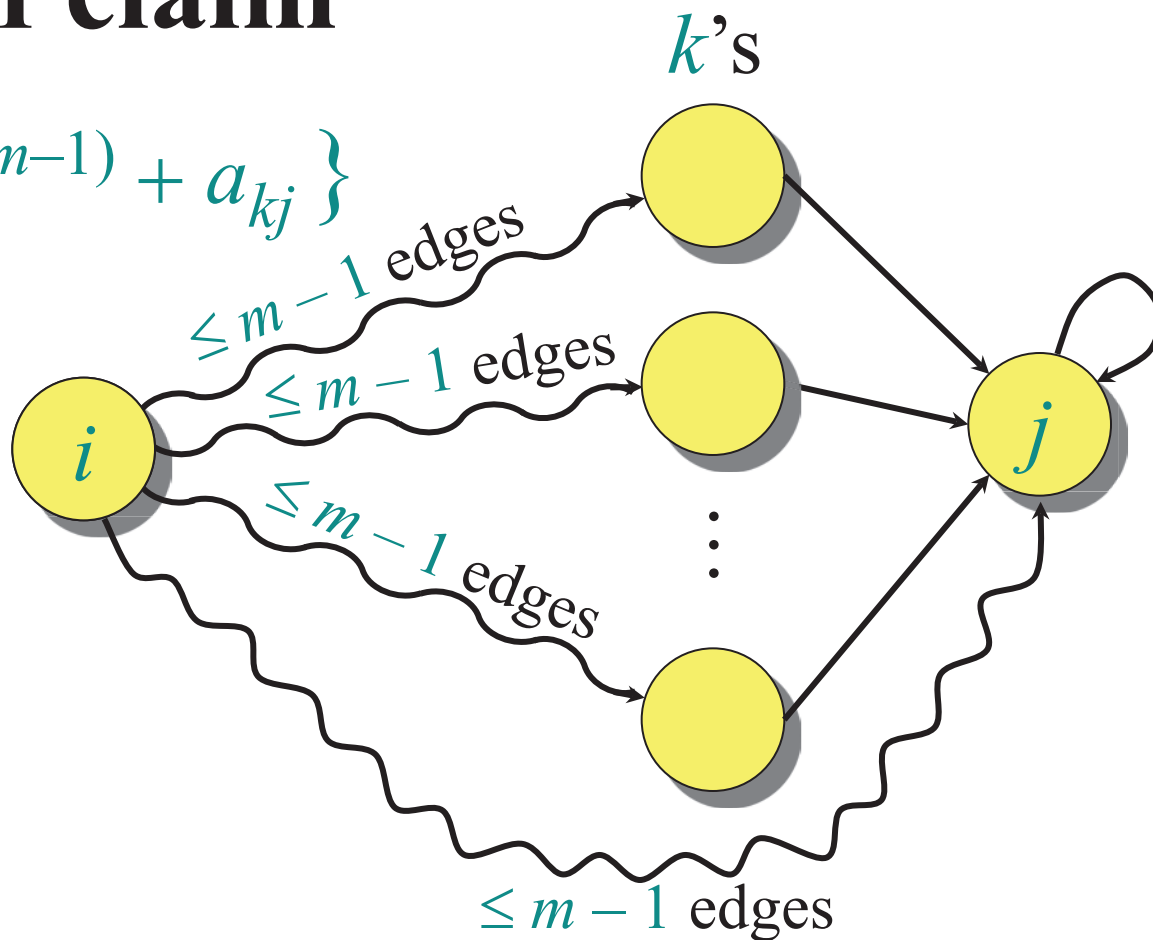
and for  $m = 1, 2, \dots, n - 1$ ,

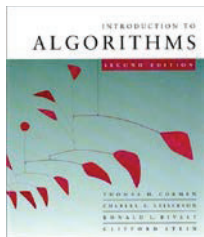
$$d_{ij}^{(m)} = \min_k \{ d_{ik}^{(m-1)} + a_{kj} \}.$$



# Proof of claim

$$d_{ij}^{(m)} = \min_k \{ d_{ik}^{(m-1)} + a_{kj} \}$$





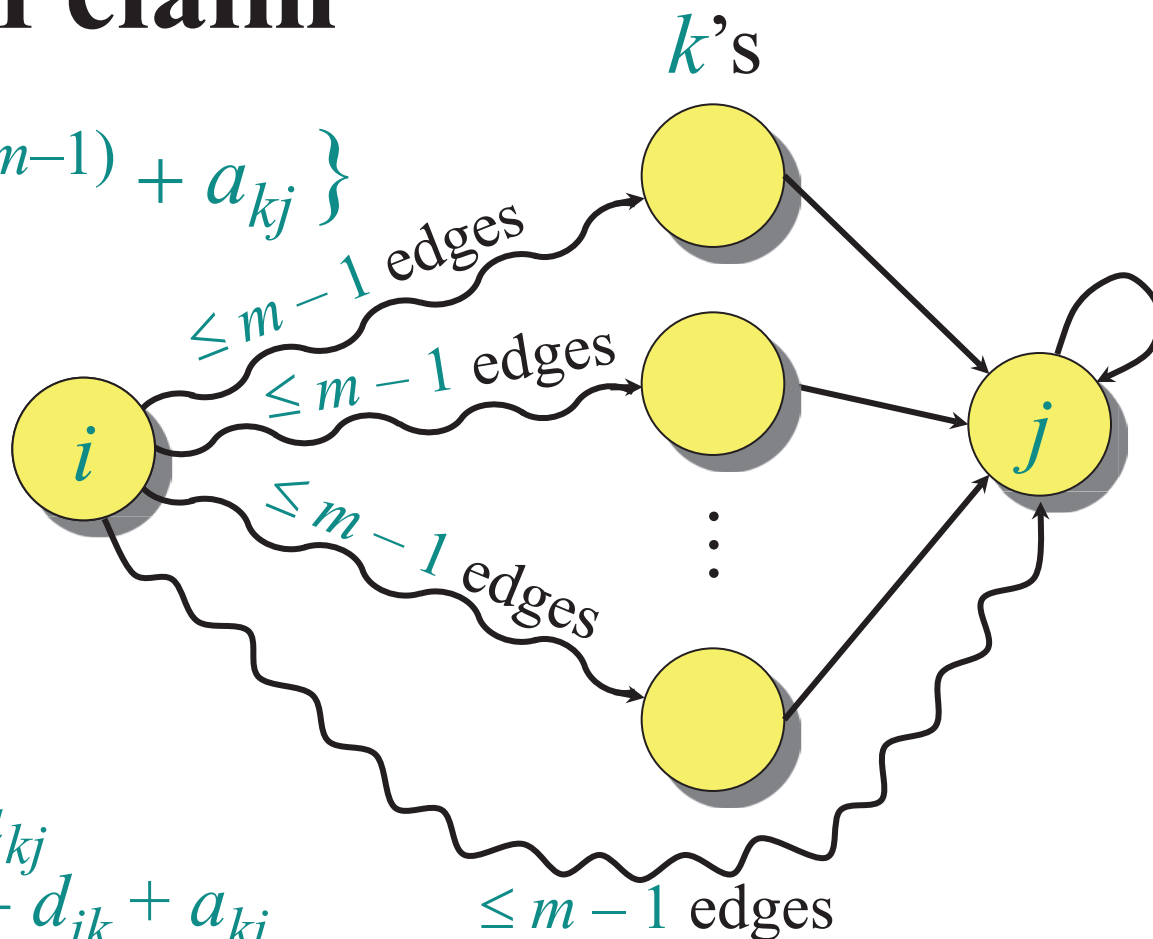
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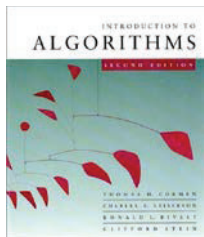
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**Relaxation!**

for  $k \leftarrow 1$  to  $n$

do if  $d_{ij} > d_{ik} + a_{kj}$   
 then  $d_{ij} \leftarrow d_{ik} + a_{kj}$





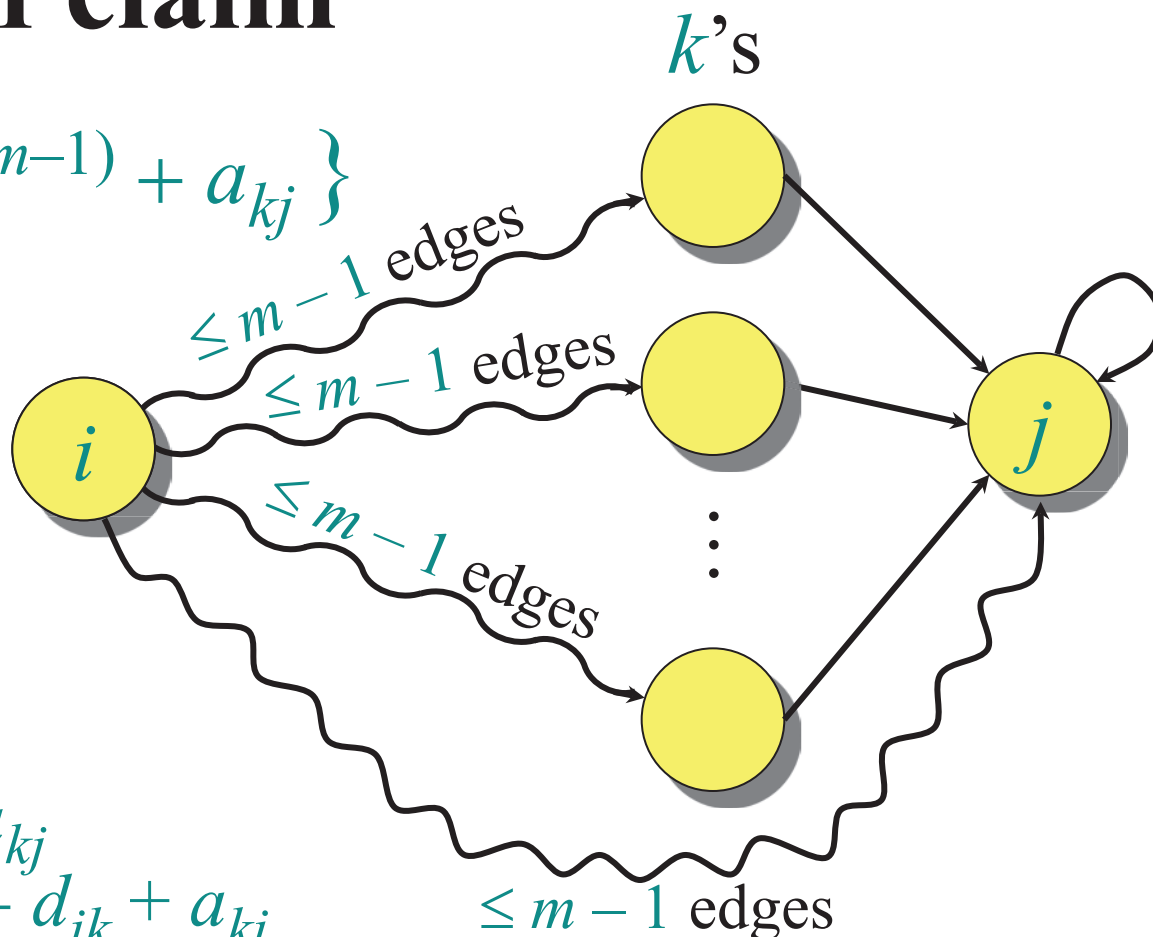
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$$d_{ij}^{(m)} = \min_k \{ d_{ik}^{(m-1)} + a_{kj} \}$$

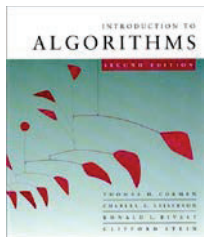
**Relaxation!**

for  $k \leftarrow 1$  to  $n$

do if  $d_{ij} > d_{ik} + a_{kj}$   
 then  $d_{ij} \leftarrow d_{ik} + a_{kj}$



**Note:** No negative-weight cycles implies  
 $\delta(i, j) = d_{ij}^{(n-1)} = d_{ij}^{(n)} = d_{ij}^{(n+1)} = \dots$

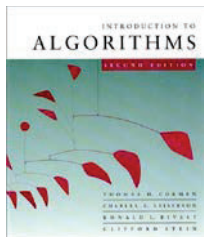


# Matrix multiplication

Compute  $C = A \cdot B$ , where  $C$ ,  $A$ , and  $B$  are  $n \times n$  matrices:

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}.$$

Time =  $\Theta(n^3)$  using the standard algorithm.



# Matrix multiplication

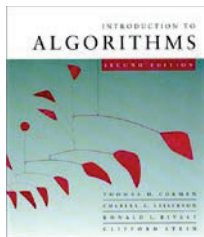
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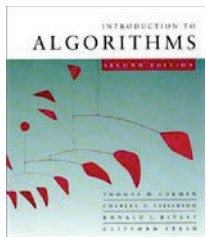
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What if we map “+”  $\rightarrow$  “min” and “.”  $\rightarrow$  “+”?

$$c_{ij} = \min_k \{a_{ik} + b_{kj}\}.$$

Thus,  $D^{(m)} = D^{(m-1)} \text{ “}\times\text{” } A$ .

$$\text{Identity matrix} = I = \begin{pmatrix} 0 & \infty & \infty & \infty \\ \infty & 0 & \infty & \infty \\ \infty & \infty & 0 & \infty \\ \infty & \infty & \infty & 0 \end{pmatrix} = D^0 = (d_{ij}^{(0)}).$$



# Matrix multiplication (continued)

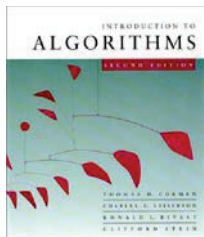
The  $(\min, +)$  multiplication is *associative*, and with the real numbers, it forms an algebraic structure called a *closed semiring*.

Consequently, we can compute

$$\begin{aligned} D^{(1)} &= D^{(0)} \cdot A = A^1 \\ D^{(2)} &= D^{(1)} \cdot A = A^2 \\ &\vdots \\ D^{(n-1)} &= D^{(n-2)} \cdot A = A^{n-1}, \end{aligned}$$

yielding  $D^{(n-1)} = (\delta(i, j))$ .

Time =  $\Theta(n \cdot n^3) = \Theta(n^4)$ . No better than  $n \times$  B-F.



# Improved matrix multiplication algorithm

**Repeated squaring:**  $A^{2k} = A^k \times A^k$ .

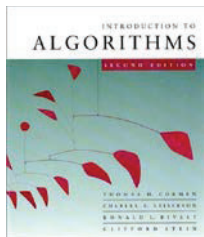
Compute  $A^2, A^4, \dots, A^{2^{\lceil \lg(n-1) \rceil}}$ .

$O(\lg n)$  squarings

**Note:**  $A^{n-1} = A^n = A^{n+1} = \dots$ .

Time =  $\Theta(n^3 \lg n)$ .

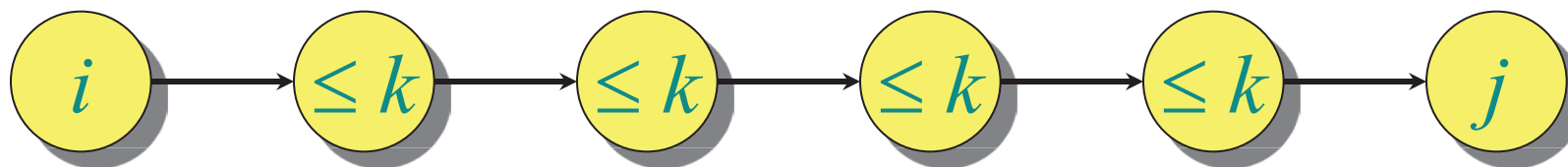
To detect negative-weight cycles, check the diagonal for negative values in  $O(n)$  additional time.



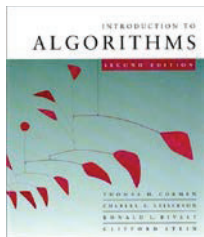
# Floyd-Warshall algorithm

*Also dynamic programming, but faster!*

Define  $c_{ij}^{(k)}$  = weight of a shortest path from  $i$  to  $j$  with intermediate vertices belonging to the set  $\{1, 2, \dots, k\}$ .

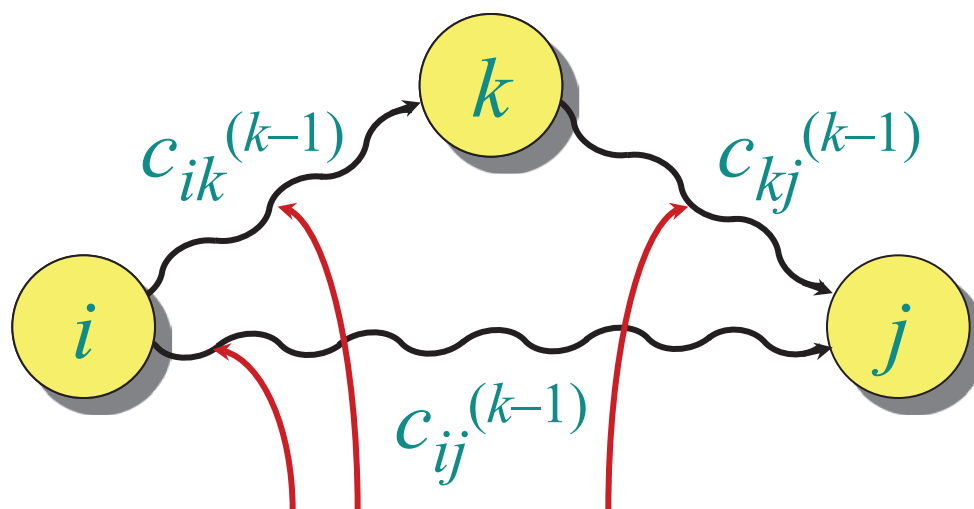


Thus,  $\delta(i, j) = c_{ij}^{(n)}$ . Also,  $c_{ij}^{(0)} = a_{ij}$ .

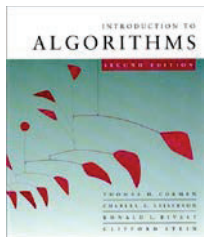


# Floyd-Warshall recurrence

$$c_{ij}^{(k)} = \min \{c_{ij}^{(k-1)}, c_{ik}^{(k-1)} + c_{kj}^{(k-1)}\}$$



intermediate vertices in  $\{1, 2, \dots, k-1\}$

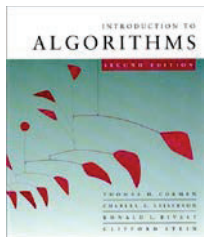


# Pseudocode for Floyd-Warshall

```
for  $k \leftarrow 1$  to  $n$ 
  do for  $i \leftarrow 1$  to  $n$ 
    do for  $j \leftarrow 1$  to  $n$ 
      do if  $c_{ij} > c_{ik} + c_{kj}$ 
        then  $c_{ij} \leftarrow c_{ik} + c_{kj}$  } relaxation
```

## Notes:

- Okay to omit superscripts, since extra relaxations can't hurt.
- Runs in  $\Theta(n^3)$  time.
- Simple to code.
- Efficient in practice.



# Transitive closure of a directed graph

Compute  $t_{ij} = \begin{cases} 1 & \text{if there exists a path from } i \text{ to } j, \\ 0 & \text{otherwise.} \end{cases}$

**IDEA:** Use Floyd-Warshall, but with  $(\vee, \wedge)$  instead of  $(\min, +)$ :

$$t_{ij}^{(k)} = t_{ij}^{(k-1)} \vee (t_{ik}^{(k-1)} \wedge t_{kj}^{(k-1)}).$$

Time =  $\Theta(n^3)$ .