

### Paths in graphs

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

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

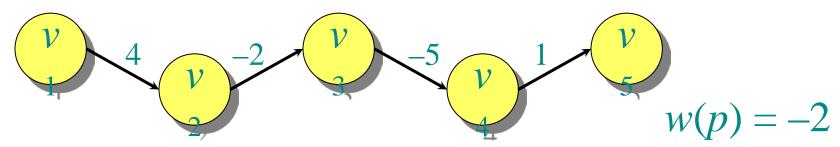


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#### **Example:**





### 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

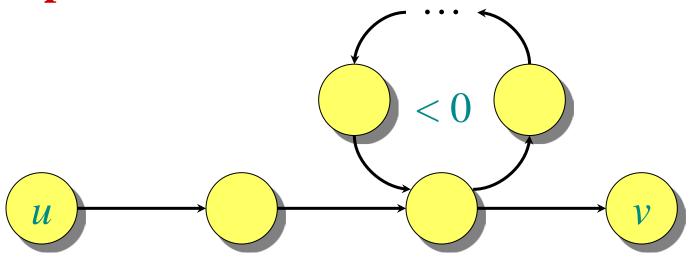
If a graph *G* contains a negative-weight cycle, then some shortest paths do not exist.



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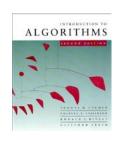
#### **Example:**





### Optimal substructure

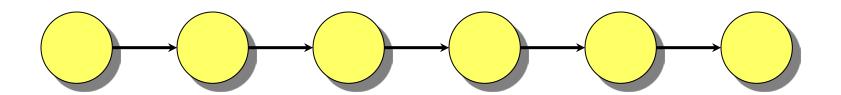
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*Proof.* Cut and paste:

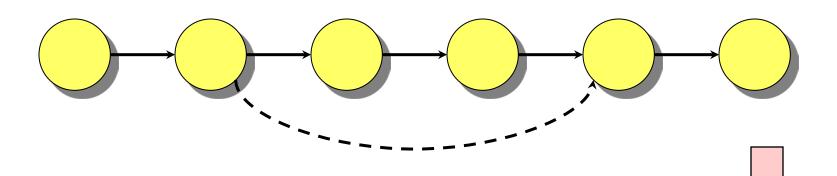


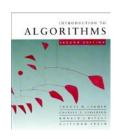


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### Triangle inequality

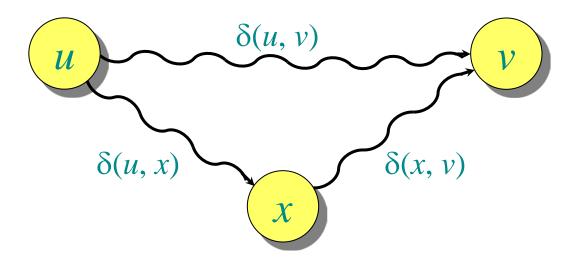
**Theorem.** For all  $u, v, x \in V$ , we have  $\delta(u, v) \leq \delta(u, x) + \delta(x, v)$ .

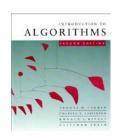


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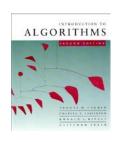


# Single-source shortest paths (nonnegative edge weights)

**Problem.** Assume that  $w(u, v) \ge 0$  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  $\nu$ .



### Dijkstra's algorithm

```
d[s] \leftarrow 0

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

do d[v] \leftarrow \infty

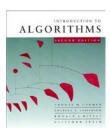
S \leftarrow \emptyset

Q \leftarrow V \triangleright Q is a priority queue maintaining V - S, keyed on d[v]
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                     keyed on d[v]
while Q \neq \emptyset
    do u \leftarrow \text{Extract-Min}(Q)
         S \leftarrow S \cup \{u\}
         for each v \in Adj[u]
             do if d[v] > d[u] + w(u, v)
                      then d[v] \leftarrow d[u] + w(u, v)
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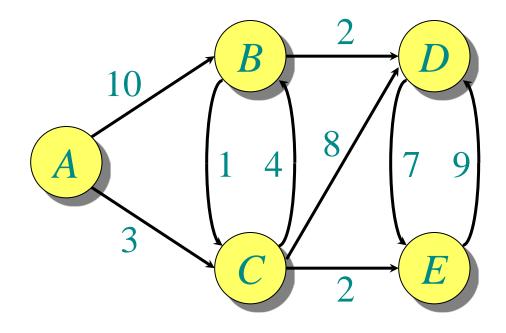


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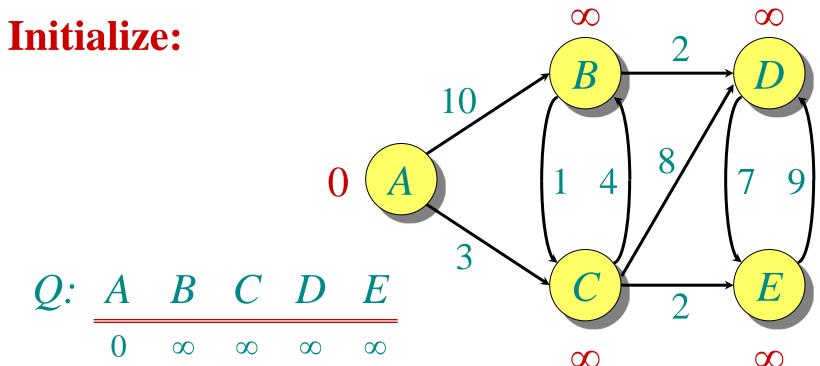
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                                                             relaxation
             do if d[v] > d[u] + w(u, v)
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                                        Implicit Decrease-Key
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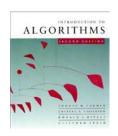


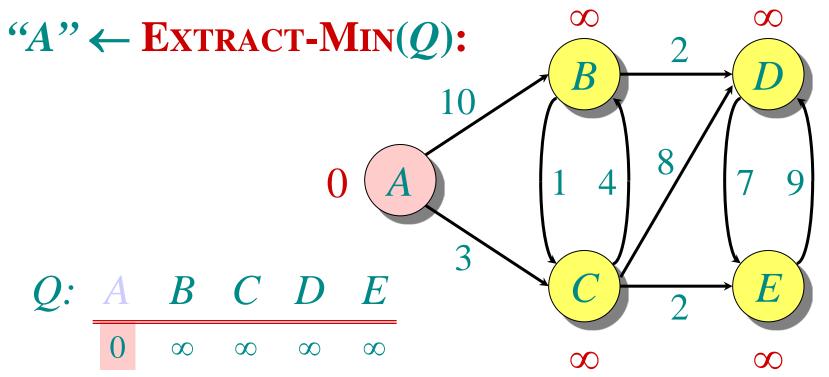
Graph with nonnegative edge weights:





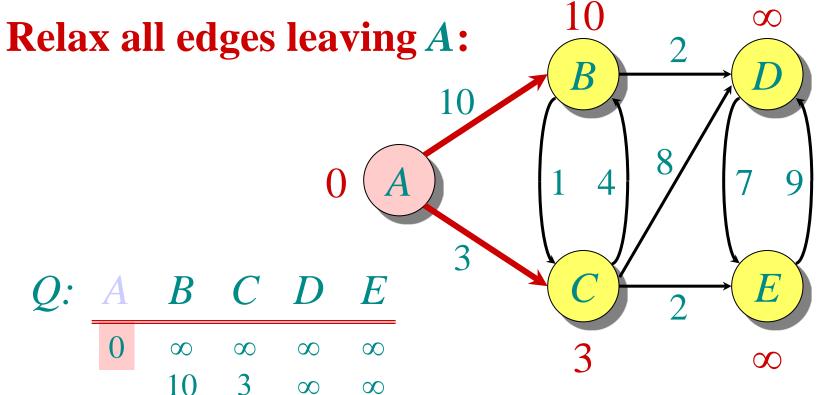




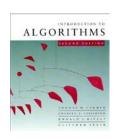


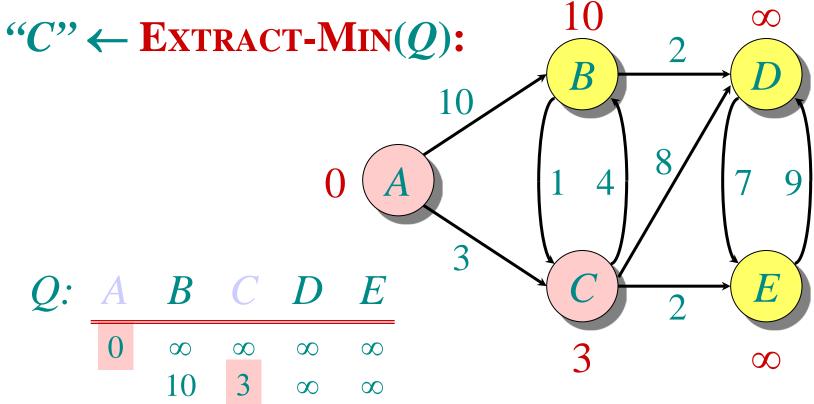
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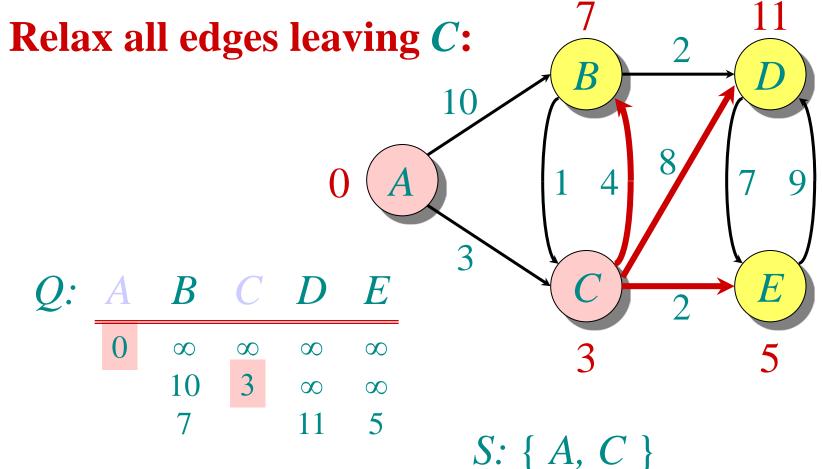
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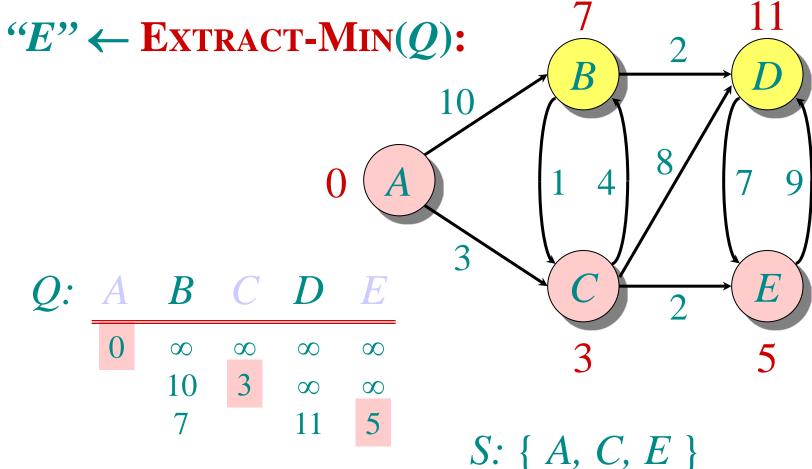


S: { A, C }

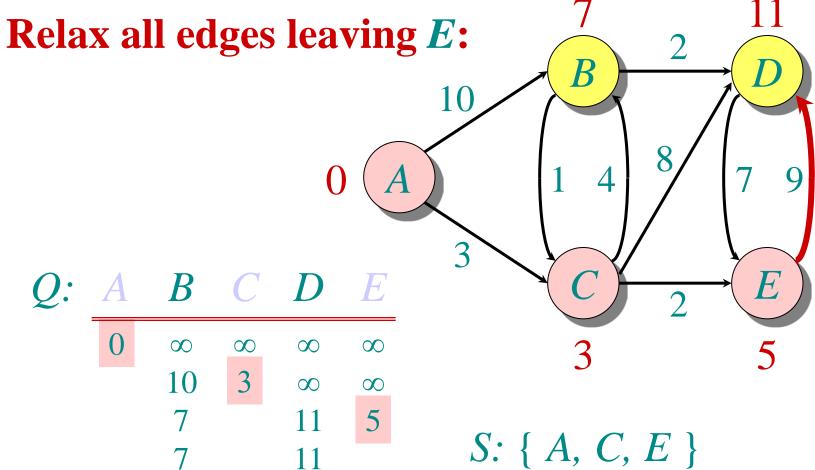


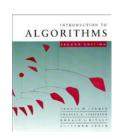


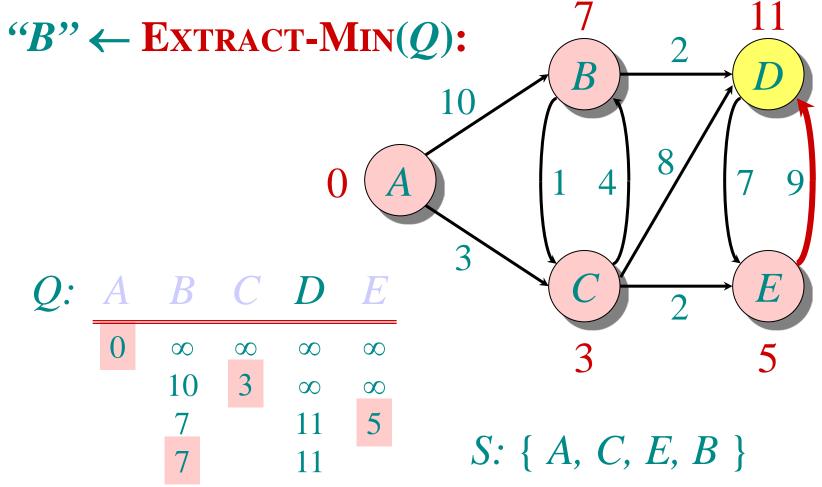




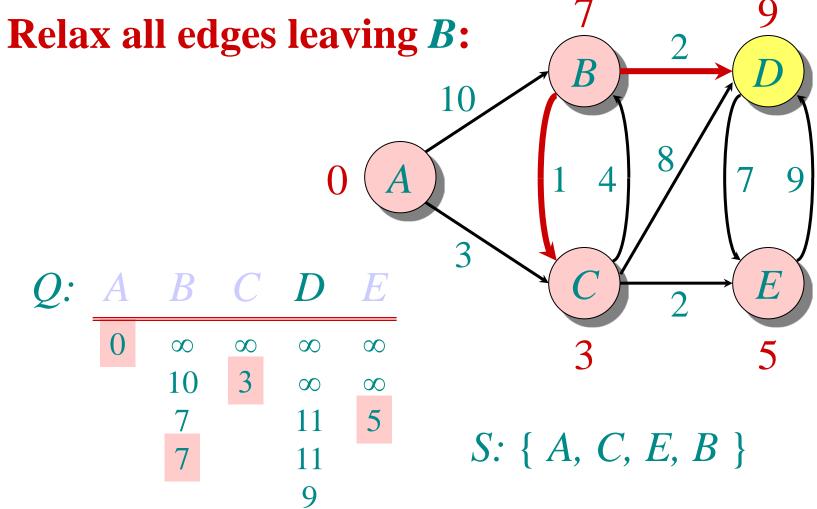


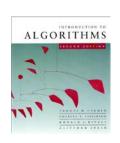


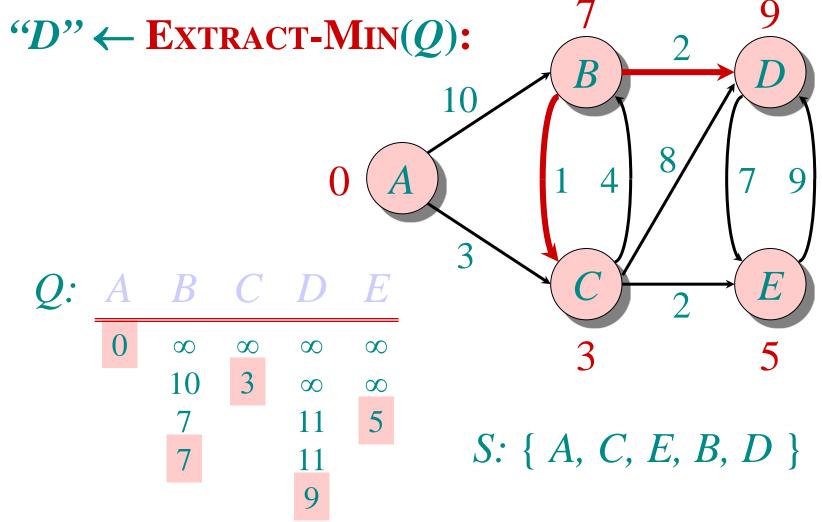














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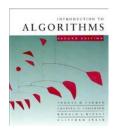
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Handshaking Lemma  $\Rightarrow \Theta(E)$  implicit Decrease-Key's.



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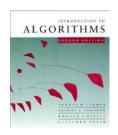
Time = 
$$\Theta(V \cdot T_{\text{EXTRACT-MIN}} + E \cdot T_{\text{DECREASE-KEY}})$$

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

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

T<sub>EXTRACT-MIN</sub> T<sub>DECREASE-KEY</sub>

**Total** 



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

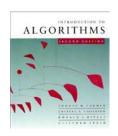
$$Q \quad T_{\text{EXTRACT-MIN}} \quad T_{\text{DECREASE-KEY}} \quad \text{Total}$$

$$\text{array} \quad O(V) \qquad O(1) \qquad O(V^2)$$



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

Q	T <sub>EXTRACT-MIN</sub>	T <sub>DECREASE-KEY</sub>	Total
array	O(V)	<i>O</i> (1)	$O(V^2)$
binary heap	$O(\lg V)$	$O(\lg V)$	$O(E \lg V)$



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binary heap	$O(\lg V)$	$O(\lg V)$	$O(E \lg V)$
Fibonacci heap	i $O(\lg V)$ amortized	O(1) amortized	$O(E + V \lg V)$ worst case

