CENG 218 Design and Analysis of Algorithms

Izmir Institute of Technology

Lecture 13: Shortest paths

Slides were mostly prepared using the material provided by Prof. Charles E. Leiserson and Prof. Erik Demaine from MIT

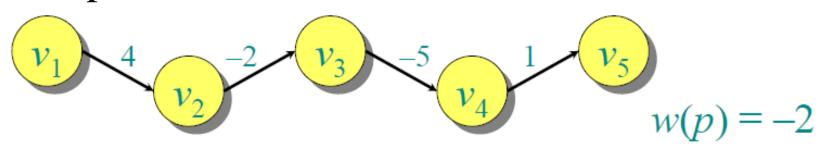
Paths in graphs

Consider a digraph G=(V, E) with edge-weight function $w: E \rightarrow \mathbb{R}$.

The *weight* of path $p = v_1 \rightarrow v_2 \rightarrow ... \rightarrow v_k$ can be defined as

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

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

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

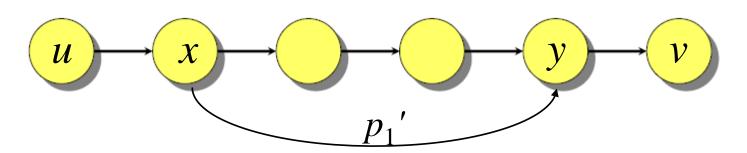
Note: $\delta(u, v) = \infty$ if no path from u to v exists.

Optimal substructure

Theorem. A sub-path of a shortest path is a shortest path.

Proof. Let p be the shortest path from u to v and let p contain p_1 which is the shortest path from x to y.

If there was a p_1' that is shorter than p_1 , then $w(p') = \delta(u,x) + w(p_1') + \delta(y,v)$ would be less than w(p), which is a contradiction since p is the shortest path.

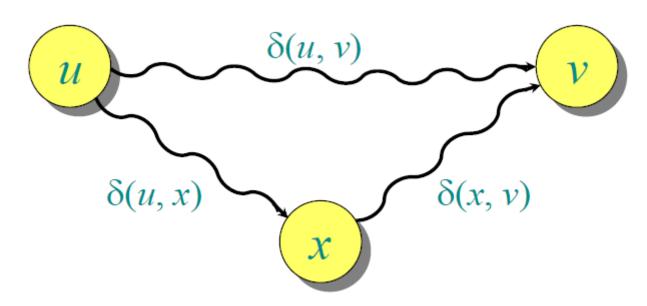


Triangle inequality

Theorem. For all $u, v, x \in V$, we have

$$\delta(u, v) \le \delta(u, x) + \delta(x, v).$$

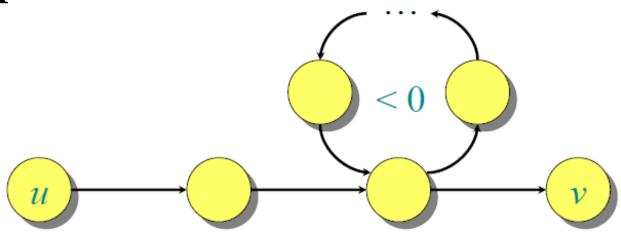
Proof. If $\delta(u, v)$ was bigger than $\delta(u, x) + \delta(x, v)$, then it would no longer be a shortest path from u to v.



When shortest paths do not exist?

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

Example:



You can keep with the cycle until reaching $-\infty$ weight.

Single-source shortest paths

Problem. From a given source vertex $s \in V$, find the shortest-path weights $\delta(s, v)$ for all $v \in V$. If all edge weights w(u, v) are *nonnegative*, all shortest-paths must exist.

IDEA: Greedy.

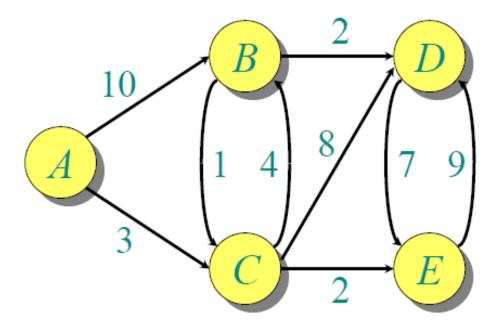
- 1. Maintain a set *S* of vertices whose shortest-path distances from *s* are known.
- 2. At each step add to S the vertex $v \in V S$ whose distance estimate from s is minimal.
- 3. Update the distance estimates of vertices adjacent to *v*.

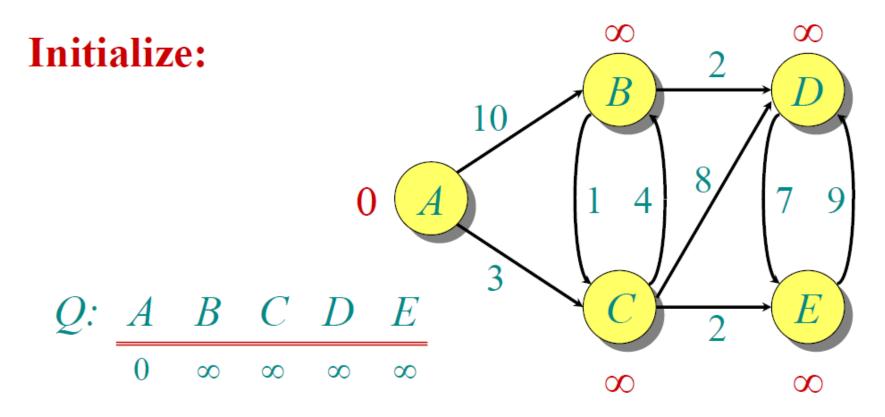
Dijkstra's algorithm

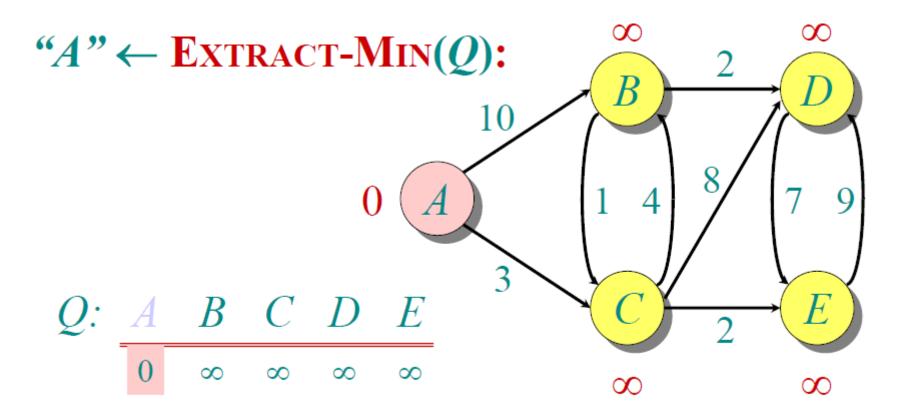
```
d[s] \leftarrow 0
for each v \in V - \{s\}
    do d[v] \leftarrow \infty
S \leftarrow \emptyset
Q \leftarrow V \triangleright Q is a priority queue maintaining V - S
while Q \neq \emptyset
    do u \leftarrow \text{Extract-Min}(Q)
        S \leftarrow S \cup \{u\}
        for each v \in Adj[u]
             do| if d[v] > d[u] + w(u, v)
                      then d[v] \leftarrow d[u] + w(u, v)
                             Relaxation Step
```

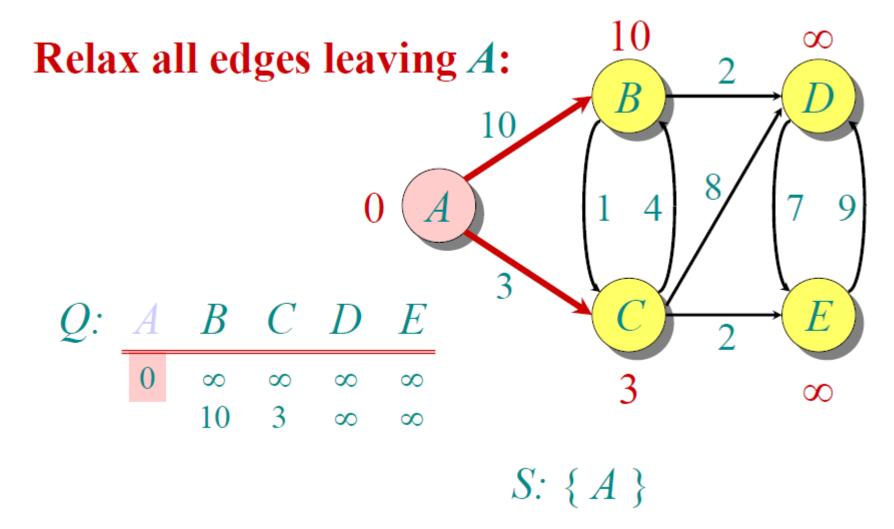
Similar to DECREASE-KEY in Prim's algorithm for MST

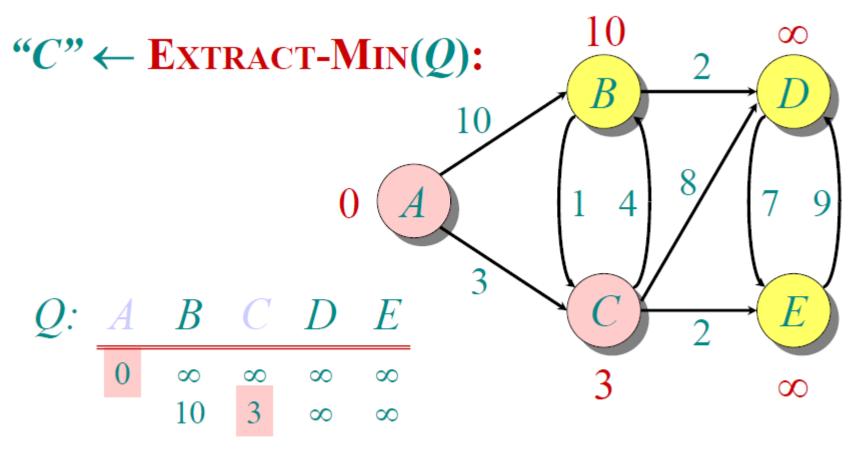
Graph with nonnegative edge weights:



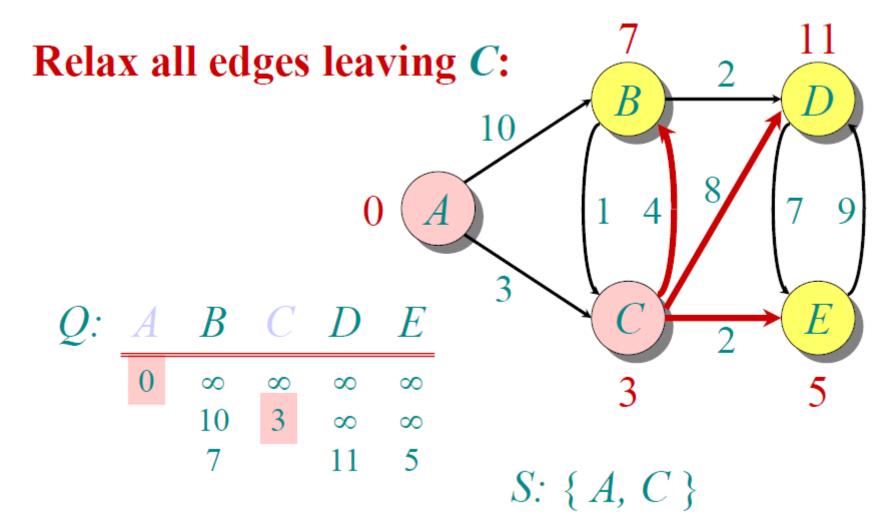


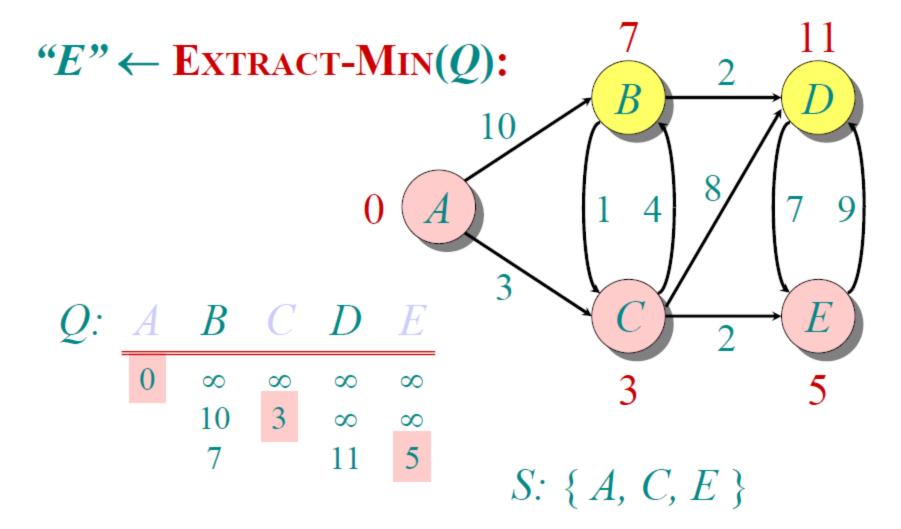


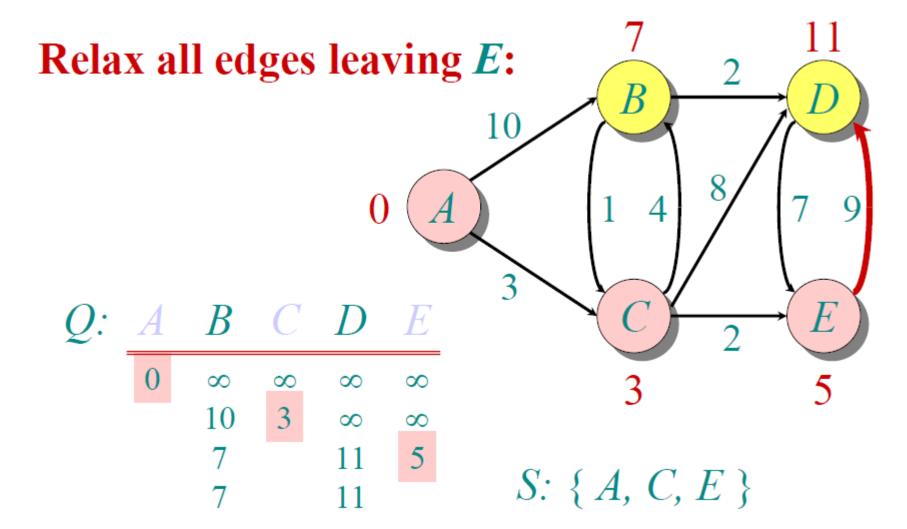


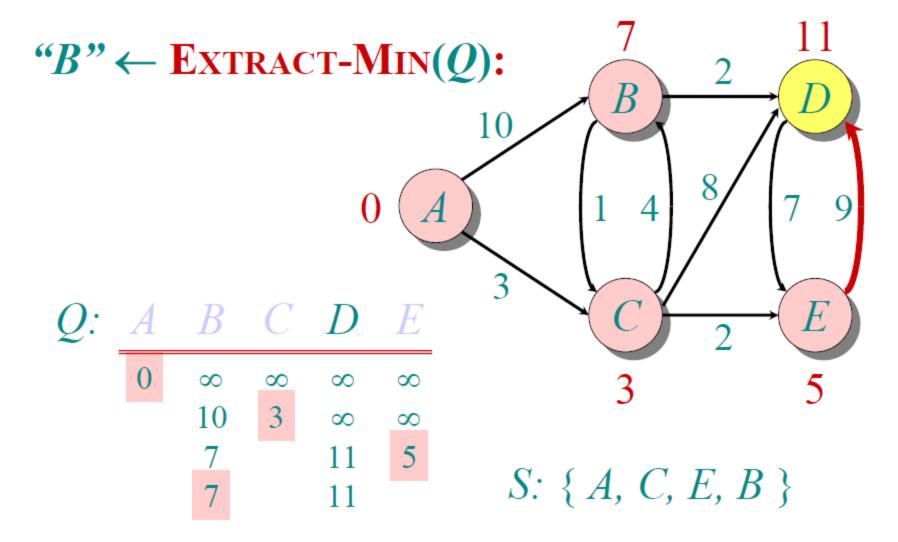


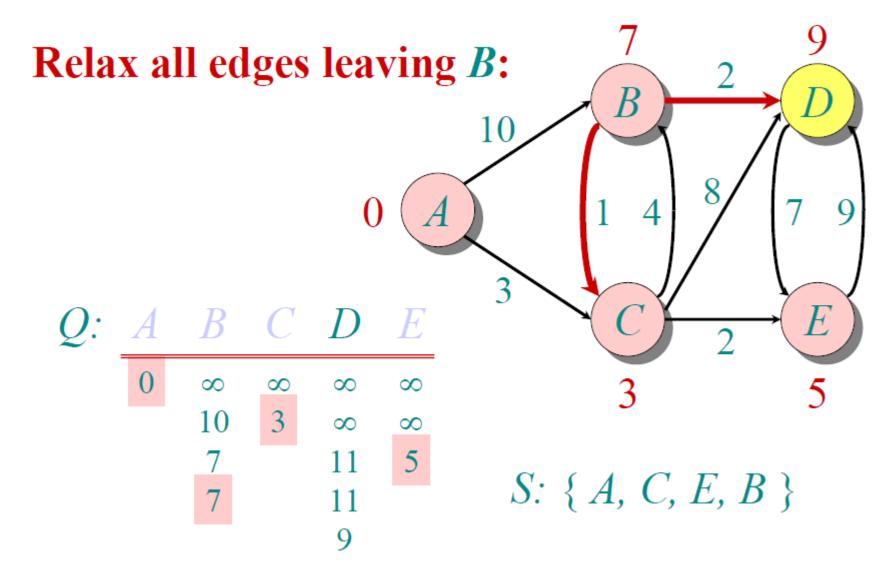
S: { *A*, *C* }

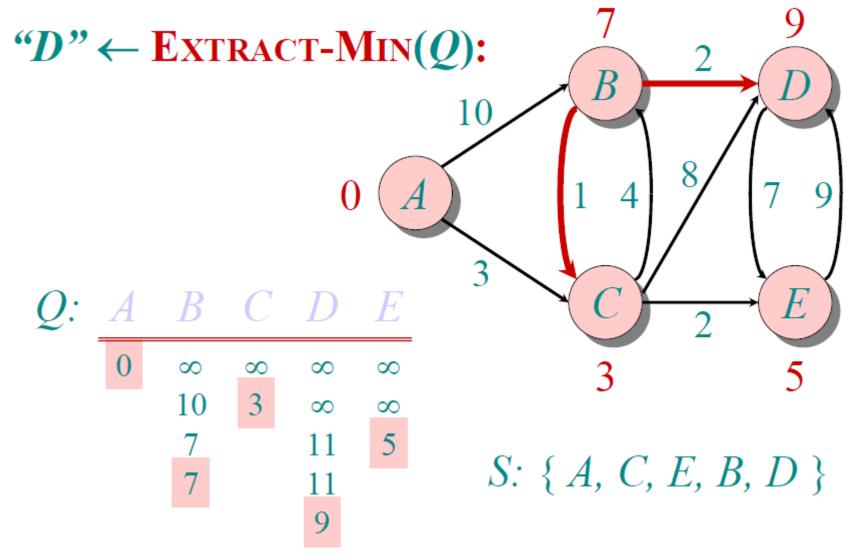












Analysis of Dijkstra

times while
$$Q \neq \emptyset$$

$$do u \leftarrow \text{Extract-Min}(Q)$$

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

$$for each $v \in Adj[u]$$$

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

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

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

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

Note: Same as the analysis of Prim's MST algorithm.

Analysis of Dijkstra

Time =
$$\Theta(V) \cdot T_{\text{EXTRACT-MIN}} + \Theta(E) \cdot T_{\text{DECREASE-KEY}}$$
assume we keep
 Q as an array
$$= \Theta(V) \cdot O(V) + O(V^2) \cdot O(1)$$

$$= O(V^2)$$

Analysis of Dijkstra

When it is dense graph, implementing priority queue as an array is OK since the best we can get is $O(V^2)$.

But when the graph is sparse, $E << V^2$, then use binary heap for priority queue of vertices:

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

Q is a binary heap.

Reorganization of binary heap

after deleting min

$$= \Theta(V) \cdot \Theta(\log V) + \Theta(E) \cdot \Theta(\log V)$$

$$= \Theta(E \log V)$$

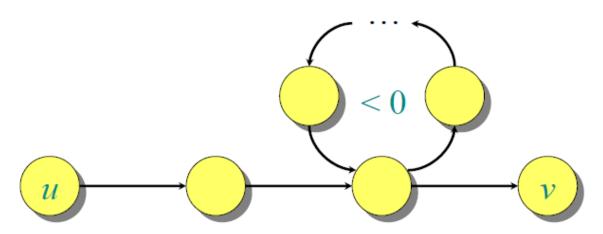
Bellman-Ford algorithm

Finds all shortest-path lengths from a *source* $s \in V$ to all $v \in V$

OR

reports that a negative-weight cycle exists.

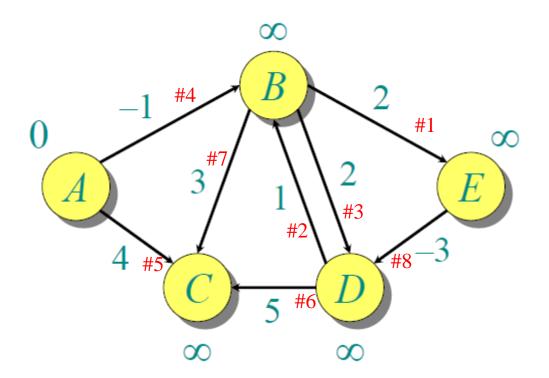
Recall negative-weight cycles:



Bellman-Ford algorithm

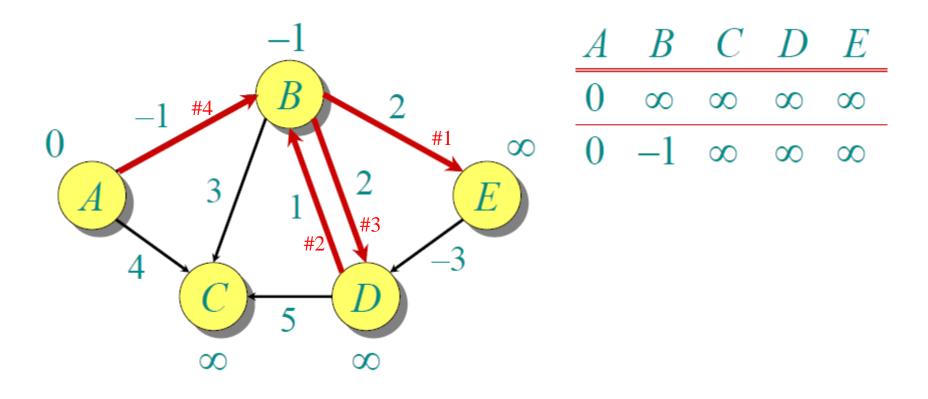
```
d[s] \leftarrow 0
for each v \in V - \{s\}
do \ d[v] \leftarrow \infty
initialization
for i \leftarrow 1 to |V| - 1
    do for each edge (u, v) \in E
         do if d[v] > d[u] + w(u, v) relaxation
then d[v] \leftarrow d[u] + w(u, v) step
for each edge (u, v) \in E
    do if d[v] > d[u] + w(u, v)
             then report that a negative-weight cycle exists
```

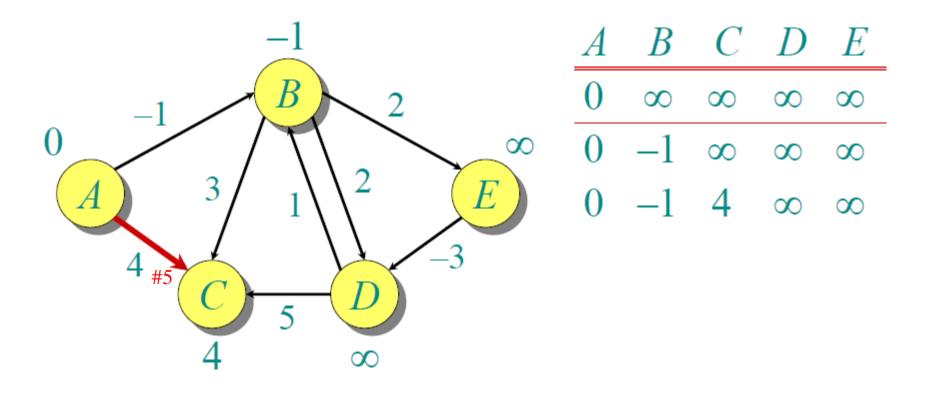
At the end, $d[v] = \delta(s, v)$. Running time = O(VE).

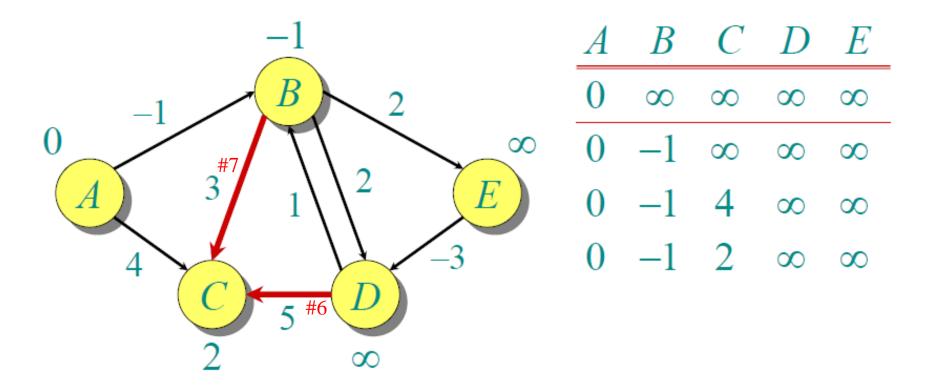


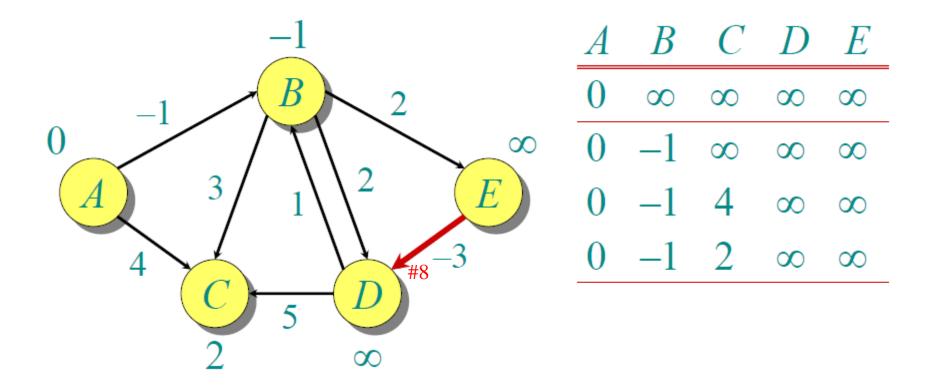


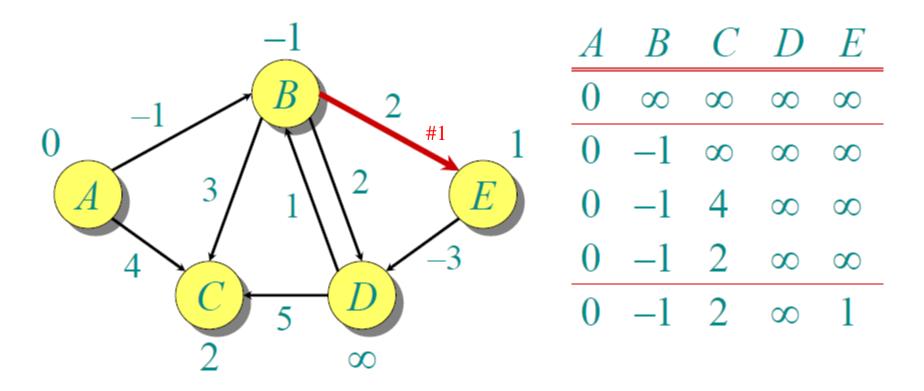
Note: Order of edges has no importance.

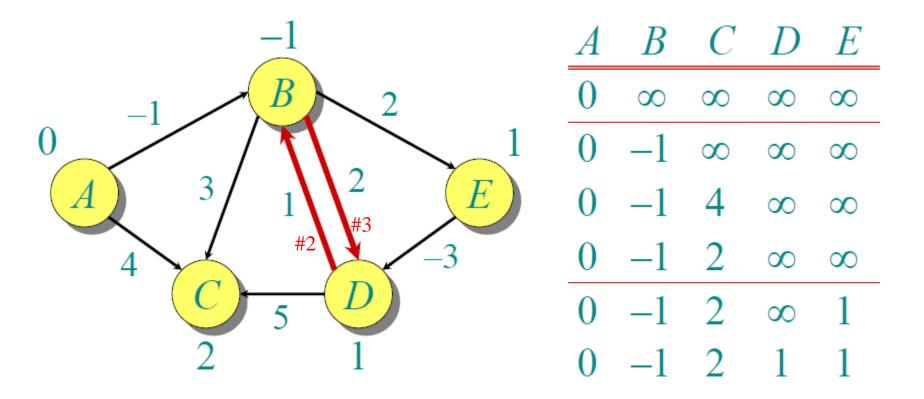


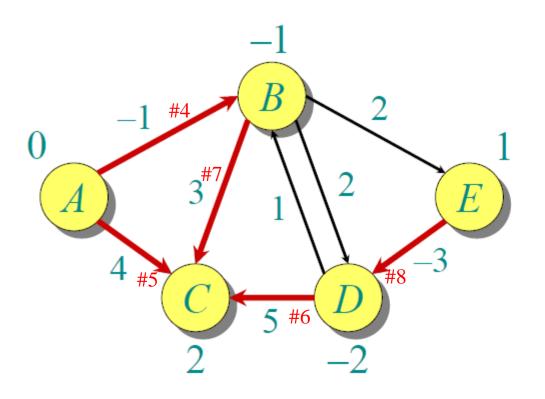








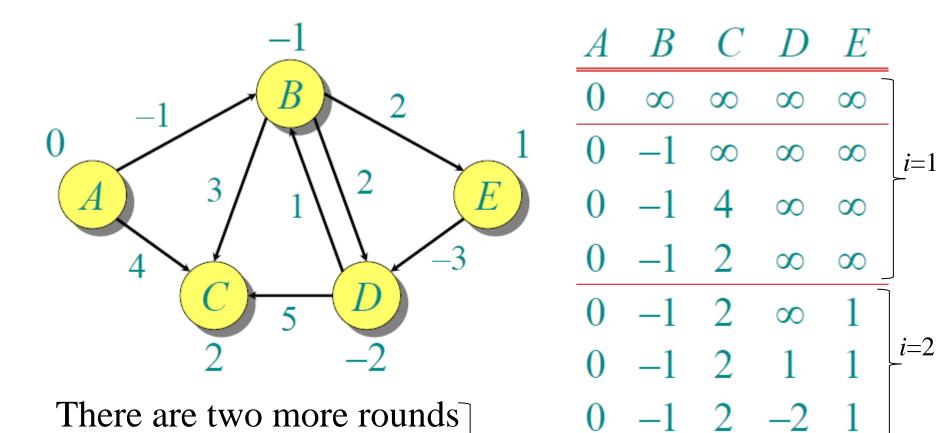




A	B	C	D	E
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞
0	-1	2	∞	∞
0	-1	2	∞	1
0	-1	2	1	1
0	-1	2	-2	1

here since |V|-1 is 4,

but result does not change.



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 $v \in V$.

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

Conclusion: $O(V \cdot E)$ is slower than Dijkstra's, but we can report if negative weight exists.

The End

Textbook Section 24.1, 24.3.