CSE 2202 Design and Analysis of Algorithms – I

Single Source Shortest Path (Dijkstra and Bellman Ford)

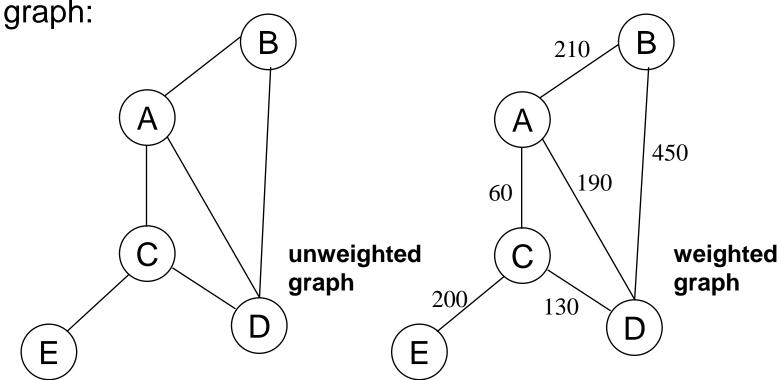
SINGLE SOURCE SHORTEST PATH(DIJKSTRA'S ALGORITHM)

Shortest Path Problems

What is shortest path?

shortest length between two vertices for an unweighted graph:

smallest cost between two vertices for a weighted graph:



Shortest Path Problems

- How can we find the shortest route between two points on a map?
- Model the problem as a graph problem:
 - Road map is a weighted graph:

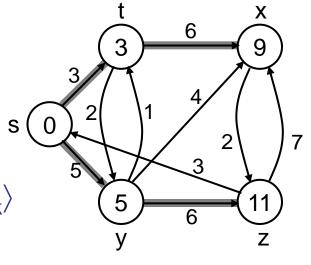
```
vertices = cities
edges = road segments between cities
edge weights = road distances
```

Goal: find a shortest path between two vertices (cities)

Shortest Path Problems

Input:

- Directed graph G = (V, E)
- Weight function w : $E \rightarrow R$
- Weight of path $p = \langle v_0, v_1, \dots, v_k \rangle$ $w(p) = \sum_{i=1}^k w(v_{i-1}, v_i)$



Shortest-path weight from u to v:

$$\delta(u, v) = \begin{cases} \min \{w(p) : u \xrightarrow{p} v\} & \text{if there exists a path from } u \text{ to } v \end{cases}$$

$$\infty \qquad \text{otherwise}$$

Shortest path u to v is any path p such that w(p) = δ(u, v)

Variants of Shortest Paths

Single-source shortest path

G = (V, E) ⇒ find a shortest path from a given source vertex s to each vertex v ∈ V

Single-destination shortest path

- Find a shortest path to a given destination vertex t from each vertex v
- Reverse the direction of each edge ⇒ single-source

Single-pair shortest path

- Find a shortest path from u to v for given vertices u and v
- Solve the single-source problem

All-pairs shortest-paths

Find a shortest path from u to v for every pair of vertices u and v

Optimal Substructure of Shortest Paths

Lemma 24.1 (Subpaths of shortest paths are shortest paths)

Given:

- A weighted, directed graph G = (V, E)
- A weight function w: $E \rightarrow \mathbb{R}$,
- A shortest path $p = \langle v_1, v_2, \dots, v_k \rangle$ from v_1 to v_k
- A subpath of p: $p_{ij} = \langle v_i, v_{i+1}, \dots, v_j \rangle$, with $1 \le i \le j \le k$

Then: p_{ij} is a shortest path from v_i to v_j

Proof:
$$p = v_1 \stackrel{p_{1i}}{\leadsto} v_i \stackrel{p_{ij}}{\leadsto} v_j \stackrel{p_{jk}}{\leadsto} v_k$$

$$w(p) = w(p_{1i}) + w(p_{ij}) + w(p_{jk})$$

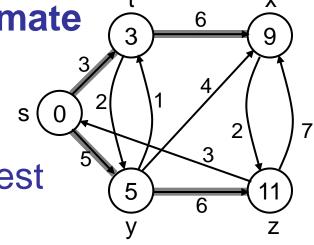
Assume $\exists p'_{ij}$ from v_i to v_j with $w(p'_{ij}) < w(p_{ij})$

$$\Rightarrow$$
 w(p') = w(p_{1i}) + w(p_{ij}') + w(p_{ik}) < w(p) contradiction!

Shortest-Path Representation

For each vertex $v \in V$:

- v.d = $\delta(s, v)$: a **shortest-path estimate**
 - Initially, d[v]=∞
 - Reduces as algorithms progress
- v.π = predecessor of v on a shortest
 path from s
 - If no predecessor, $v.\pi = NIL$
 - $-\pi$ induces a tree—shortest-path tree
- Shortest paths & shortest path trees are not unique



Initialization

INITIALIZE-SINGLE-SOURCE (G, s)

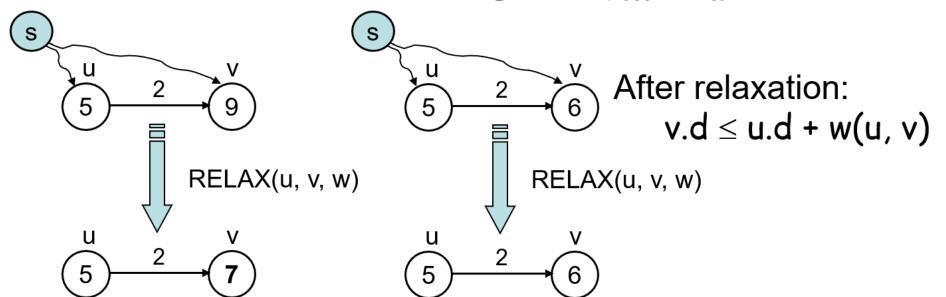
- 1 **for** each vertex $\nu \in G.V$
- $v.d = \infty$
- $\nu.\pi = NIL$
- $4 \quad s.d = 0$
- All the shortest-paths algorithms start with INITIALIZE-SINGLE-SOURCE

After initialization, we have $v.\pi = \text{NIL}$ for all $v \in V$, s.d = 0, and $v.d = \infty$ for $v \in V - \{s\}$.

Relaxation

• **Relaxing** an edge (u, v) = testing whether we can improve the shortest path to v found so far by going through u RELAX(u, v, w)

```
1 if v.d > u.d + w(u, v)
2 v.d = u.d + w(u, v)
3 v.\pi = u
```



RELAX(u, v, w)

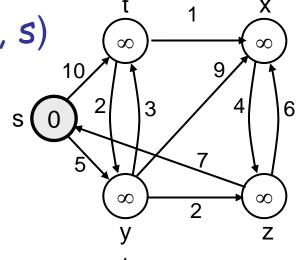
- All the single-source shortest-paths algorithms
 - start by calling INIT-SINGLE-SOURCE
 - then relax edges
- The algorithms differ in the order and how many times they relax each edge

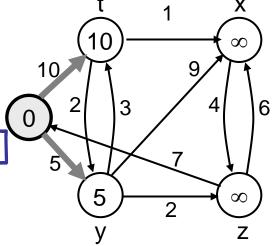
Dijkstra's Algorithm

- Single-source shortest path problem:
 - No negative-weight edges: $w(u, v) > 0 \forall (u, v) \in E$
- Maintains two sets of vertices:
 - S = vertices whose final shortest-path weights have already been determined
 - -Q = vertices in V S: min-priority queue
 - Keys in Q are estimates of shortest-path weights (v.d)
- Repeatedly select a vertex u ∈ V S, with the minimum shortest-path estimate v.d

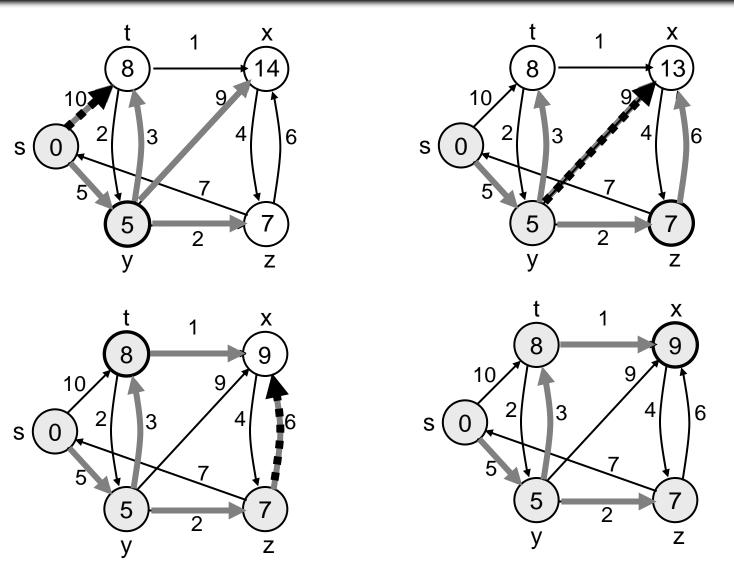
Dijkstra (G, w, s)

- 1. INITIALIZE-SINGLE-SOURCE(V, s)
- 2. S ← Ø
- 3. Q ← G. V
- 4. while $Q \neq \emptyset$
- 5. **do** $u \leftarrow EXTRACT-MIN(Q)$
- 6. $S \leftarrow S \cup \{u\}$
- 7. for each vertex $v \in G.Adj[u]$
- 8. **do** RELAX(u, v, w)





Example



Dijkstra's Pseudo Code

Graph G, weight function w, root s

```
DIJKSTRA(G, w, s)
   1 for each v \in V
   2 \operatorname{do} d[v] \leftarrow \infty
  3 \ d[s] \leftarrow 0
  4 S \leftarrow \emptyset \triangleright \text{Set of discovered nodes}
   5 \ Q \leftarrow V
   6 while Q \neq \emptyset
              \mathbf{do} \ u \leftarrow \text{Extract-Min}(Q)
                  S \leftarrow S \cup \{u\}
      for each v \in Adj[u]
                                                                               relaxing
                          do if d[v] > d[u] + w(u, v)
                                                                               edges
                                  then d[v] \leftarrow d[u] + w(u, v)
```

Dijkstra (G, w, s)

- 1. INITIALIZE-SINGLE-SOURCE(G, s) $\leftarrow \Theta(V)$
- 2. $S \leftarrow \emptyset$ never inserts vertices into Q after line 3
- 3. $Q \leftarrow G.V \leftarrow O(V)$ build min-heap
- 4. while $Q \neq \emptyset \leftarrow$ Executed O(V) times
- 5. do $u \leftarrow EXTRACT-MIN(Q) \leftarrow O(IgV)$
- 6. $S \leftarrow S \cup \{u\}$
- 7. **for** each vertex $v \in G.Adj[u]$
- 8. do RELAX(u, v, w) \leftarrow O(E) times; O(IgV)

Running time: O(VlgV + ElgV) = O(ElgV)

Dijkstra's Running Time

- Extract-Min executed | V time
- Decrease-Key executed |E| time
- Time = $|V| T_{\text{Extract-Min}} + |E| T_{\text{Decrease-Key}}$
- T depends on different Q implementations

Q	T(Extract	T(Decrease-	Total
	-Min)	Key)	
array	<i>O</i> (<i>V</i>)	<i>O</i> (1)	O(V ²)
binary heap	<i>O</i> (lg <i>V</i>)	<i>O</i> (lg <i>V</i>)	<i>O</i> (<i>E</i> lg <i>V</i>)
Fibonacci heap	<i>O</i> (lg <i>V</i>)	O(1) (amort.)	$O(V \lg V + E)$

Question

- Prove that, if there exists negative edge, dijkstra's shortest path algorithm may fail to find the shortest path
- Print the shortest path for dijkstra's algorithm
- Suppose you are given a graph where each edge represents the path cost and each vertex has also a cost which represents that, if you select a path using this node, the cost will be added with the path cost. How can it be solved using Dijkstra's algorithm?

Negative-Weight Edges

s → a: only one path

$$\delta(s, a) = w(s, a) = 3$$

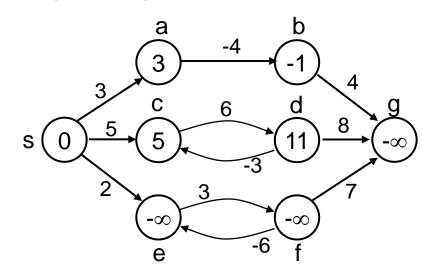
• s \rightarrow b: only one path $\delta(s, b) = w(s, a) + w(a, b) = -1$

s → c: infinitely many paths
 ⟨s, c⟩, ⟨s, c, d, c⟩, ⟨s, c, d, c, d, c⟩

cycle has positive weight (6 - 3 = 3)

 $\langle s, c \rangle$ is shortest path with weight $\delta(s, c) = w(s, c) = 5$

What if we have negativeweight edges?

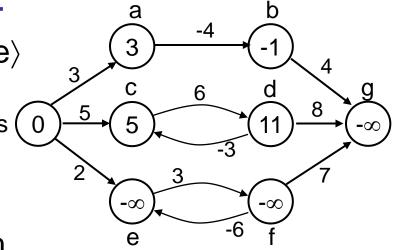


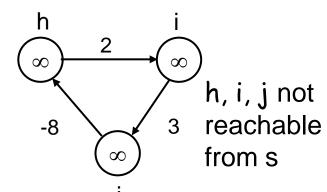
Negative-Weight Edges

- $s \rightarrow e$: infinitely many paths:
 - $-\langle s, e \rangle, \langle s, e, f, e \rangle, \langle s, e, f, e, f, e \rangle$
 - cycle (e, f, e) has negative weight:

$$3 + (-6) = -3$$

- can find paths from s to e with arbitrarily large negative weights
- $-\delta(s, e) = -\infty \Rightarrow$ no shortest path exists between *s* and *e*
- Similarly: $\delta(s, f) = -\infty$, $\delta(s, g) = -\infty$

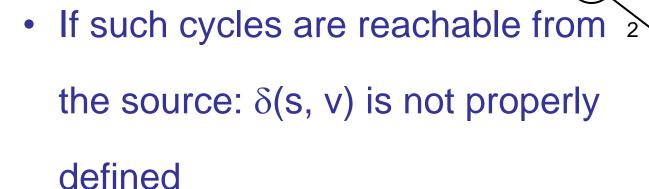




$$\delta(s, h) = \delta(s, i) = \delta(s, j) = \infty$$

Negative-Weight Edges

 Negative-weight edges may form negative-weight cycles



- Keep going around the cycle, and get $w(s, v) = -\infty$ for all v on the cycle

Cycles

- Can shortest paths contain cycles?
- Negative-weight cycles No!
- Positive-weight cycles: No!
 - By removing the cycle we can get a shorter path
- We will assume that when we are finding shortest paths, the paths will have no cycles

BELLMAN FORD

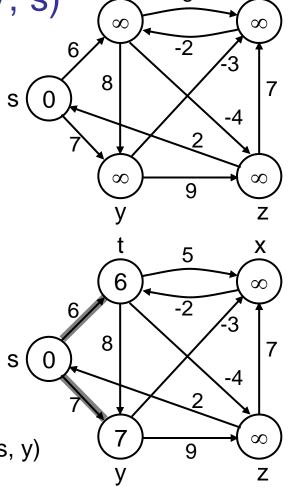
Bellman-Ford Algorithm

- Single-source shortest paths problem
 - Computes v.d and v. π for all v \in V
- Allows negative edge weights
- Returns:
 - TRUE if no negative-weight cycles are reachable from the source s
 - FALSE otherwise ⇒ no solution exists
- Idea:
 - Traverse all the edges |V 1| times, every time performing a relaxation step of each edge
 - This is because, in the worst-case scenario, any vertex's path length can be changed N times to an even shorter path length.

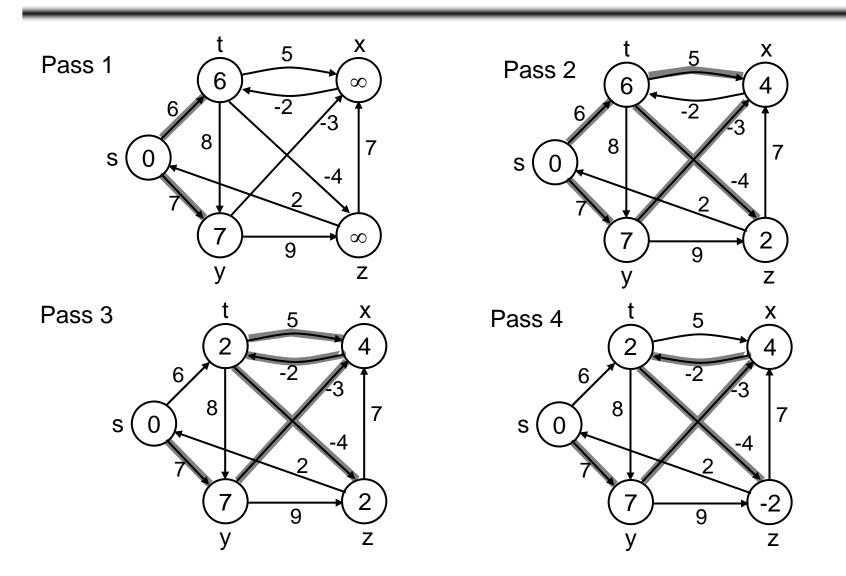
BELLMAN-FORD(V, E, w, s)

- 1. INITIALIZE-SINGLE-SOURCE(V, s)
- 2. **for** $i \leftarrow 1$ to |V| 1
- 3. do for each edge $(u, v) \in E$
- 4. **do** RELAX(u, v, w)
- 5. for each edge $(u, v) \in E$
- 6. **do if** d[v] > d[u] + w(u, v)
- 7. then return FALSE
- 8. return TRUE

E: (t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)

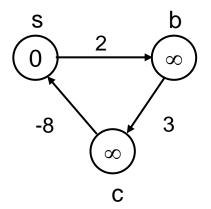


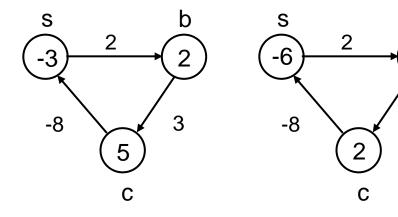
Example (t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)



Detecting Negative Cycles

- for each edge (u, v) ∈ E
- **do if** v.d > u.d + w(u, v)
- then return FALSE
- return TRUE





Look at edge (s, b):

$$\Rightarrow$$
 b.d > s.d + w(s, b)

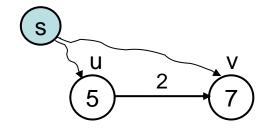
BELLMAN-FORD(V, E, w, s)

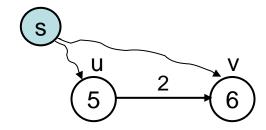
```
INITIALIZE-SINGLE-SOURCE(V, s) \leftarrow \Theta(V)
          i \leftarrow 1 \text{ to } |G.V| - 1 \qquad \leftarrow O(V)
do for each edge (u, v) \in G.E \leftarrow O(E)
2. for i \leftarrow 1 to |G.V| - 1
                   do RELAX(u, v, w)
4.
    for each edge (u, v) ∈ G.E
                                                        ← O(E)
          do if v.d > u.d + w(u, v)
6.
                 then return FALSE
     return TRUE
```

Running time: O(VE)

Triangle inequality

For all
$$(u, v) \in E$$
, we have:
 $\delta(s, v) \le \delta(s, u) + w(u, v)$





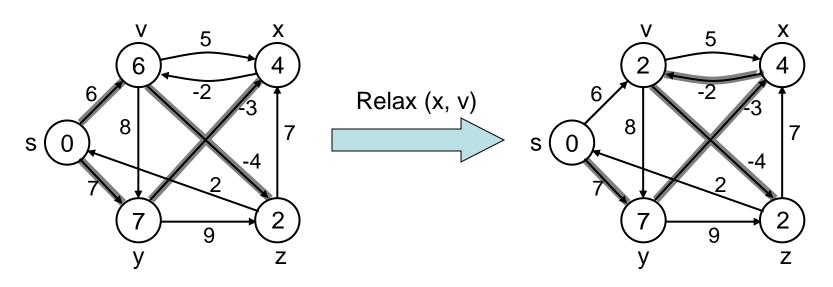
If u is on the shortest path to v we have the equality sign

Upper-bound property

We always have v.d $\geq \delta(s, v)$ for all v.

Once v.d = $\delta(s, v)$, it never changes.

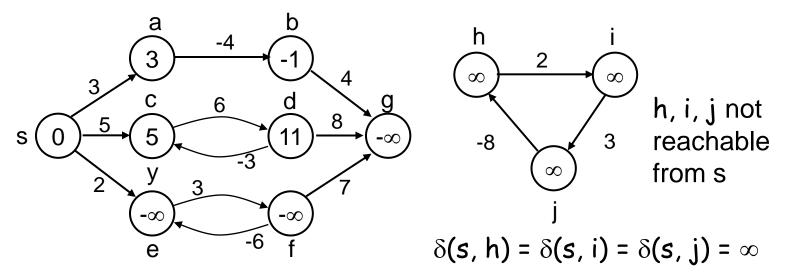
The estimate never goes up – relaxation only lowers the estimate



No-path property

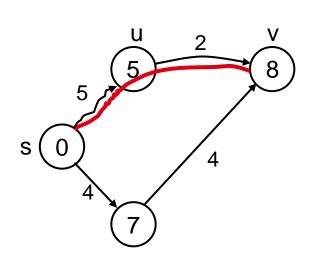
If there is no path from s to v then $v.d = \infty$ always.

 $-\delta(s, h) = \infty$ and $h.d \ge \delta(s, h) \Rightarrow h.d = \infty$



Convergence property

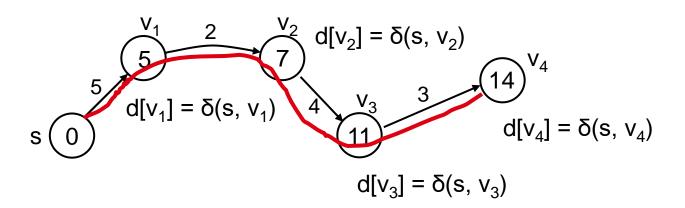
If $s \sim u \rightarrow v$ is a shortest path, and if $u.d = \delta(s, u)$ at any time prior to relaxing edge (u, v), then $v.d = \delta(s, v)$ at all times afterward.



- If v.d > $\delta(s, v) \Rightarrow$ after relaxation: v.d = u.d + w(u, v) v.d = 5 + 2 = 7
- Otherwise, the value remains unchanged, because it must have been the shortest path value

Path relaxation property

Let $p = \langle v_0, v_1, \dots, v_k \rangle$ be a shortest path from $s = v_0$ to v_k . If we relax, in order, (v_0, v_1) , (v_1, v_2) , . . . , (v_{k-1}, v_k) , even intermixed with other relaxations, then $d[v_k] = \delta(s, v_k)$.



SINGLE-SOURCE SHORTEST PATHS IN DAGS

Single-Source Shortest Paths in DAGs

- Given a weighted DAG: G = (V, E)
 - solve the shortest path problem
- Idea:
 - Topologically sort the vertices of the graph⁴
 - Relax the edges according to the order given by the topological sort
 - for each vertex, we relax each edge that starts from that vertex
- Are shortest-paths well defined in a DAG?
 - Yes, (negative-weight) cycles cannot exist

In such setting, we can compute shortest paths from a single source in time:

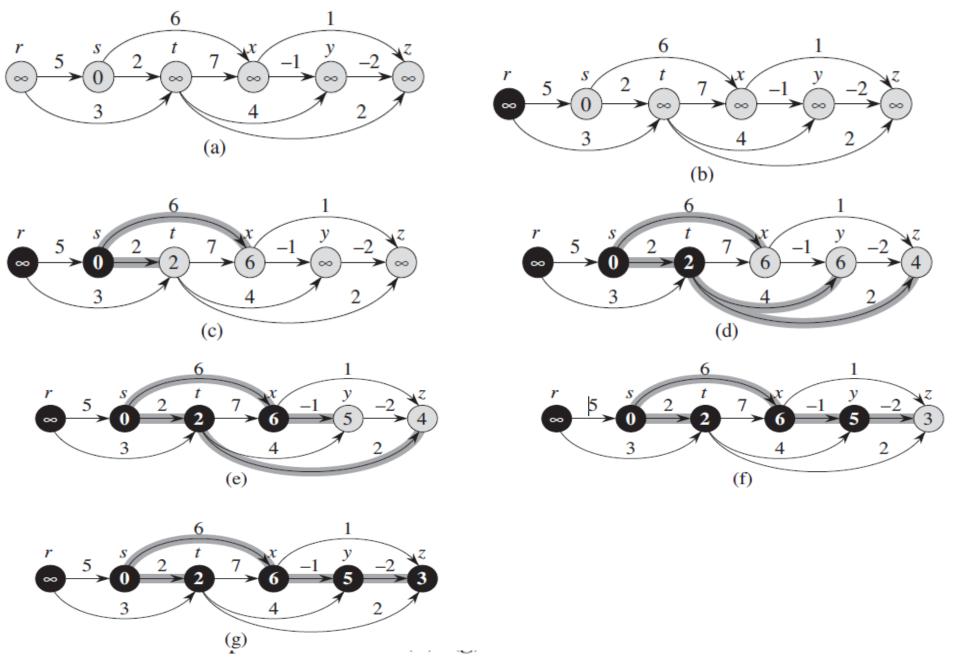
$$\Theta(V+E)$$

DAG-SHORTEST-PATHS(G, w, s)

- 1. topologically sort the vertices of $G \leftarrow \Theta(V+E)$
- 2. INITIALIZE-SINGLE-SOURCE(V, s) $\leftarrow \Theta(V)$
- 3. **for** each vertex **u**, taken in topologically ⊕(V) sorted order
- 4. do for each vertex $v \in G.Adj[u]$
- 5. do RELAX(u, v, w)

Running time: ⊕(V+E)

 $\Theta(\mathsf{E})$



The newly blackened vertex in each iteration was used as u in that iteration.

Readings

- Chapter 24
- Exercise
 - 24.1-6 Find negative cycle
 - 24.2-4 Total Number of paths in a DAG
- Difficult Problems (Solve these if you want):
 - 24.3-6 modify dijkstra
 - 24-2 nesting boxes
 - 24-3 Arbitrage
 - 24.6 Bitonic Shortest path