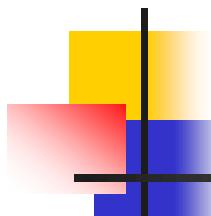


Properties of Shortest Paths and Relaxation

- Triangle inequality (Lemma 24.10)
- Upper-bound property (Lemma 24.11)
- No-path property (Corollary 24.12)
- Convergence property (Lemma 24.14)
- Path-relaxation property (Lemma 24.15)
- Predecessor-subgraph property (Lemma 24.17)

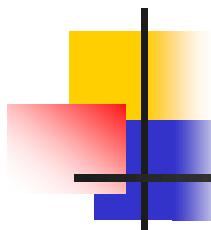




Triangle Inequality (Lemma 24.10)

- Let $G=(V,E)$ be a weighted, directed graph with weight function $w : E \rightarrow \mathbb{R}$ and source vertex s . Then, for all edges $(u,v) \in E$, we have $\delta(s,v) \leq \delta(s,u) + w(u,v)$
- Proof:
 - Suppose that there is a shortest path p from source s to v . Then p has no more weight than any other path from s to v . Specifically, path p has no more weight than the particular path that takes a shortest path from source s to vertex u and then takes edge (u,v) .
 - Otherwise, (there is no shortest path from s to v). Exercise 24.5-3.

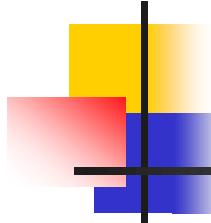




Upper-bound Property (Lemma 24.11)

- Let
 - $G=(V,E)$ be a weighted, directed graph with weight function $w : E \rightarrow \mathbb{R}$
 - $s \in V$ the source vertex
- The graph G is initialized by **INITILIZE-SINGLE-SOURCE**(G,s).
- Then, we have $v.d \geq \delta(s,v)$ for all $v \in V$. This invariant is maintained over any sequence of relaxation steps on the edges of G . Moreover, once $v.d$ achieves its lower bound $\delta(s,v)$, it never changes.

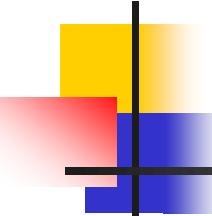




Upper-bound Property (Lemma 24.11)

- Proof by induction over the number of relaxation steps
 - Basis case (0 relaxation)
 - $v.d = \infty \geq \delta(s, v)$ for all vertices $v \in V - \{s\}$
 - $s.d = 0 = \delta(s, s)$





Upper-bound Property (Lemma 24.11)

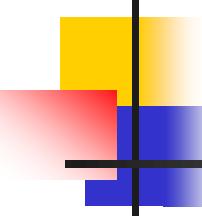
- Proof by induction over the number of relaxation steps
 - Induction step
 - Induction hypothesis: $v.d \geq \delta(s,v)$ for all $v \in V$ prior to the relaxation of an edge (u,v)
 - The only d value that may change is $v.d$. If it changes, we have
$$v.d = u.d + w(u,v) \geq \delta(s,u) + w(u,v) \geq \delta(s,v)$$
 - To see that the value of $v.d$ never changes once $v.d = \delta(s,v)$
 - $v.d$ cannot decrease because $v.d \geq \delta(s,v)$
 - $v.d$ cannot increase because relaxation steps do not increase d values



No-path Property (Corollary 24.12)

- Suppose that in a weighted, directed graph $G=(V,E)$ with weight function $w:E \rightarrow \mathbb{R}$, no path connects a source s to a given vertex v .
- Then, after the graph is initialized by $\text{INITIALIZE-SINGLE-SOURCE}(G,s)$, we have $v.d = \delta(s,v) = \infty$ and this equality is maintained as an invariant over any sequence of relaxation steps on the edges of G .
- Proof: By the upper-bound property, we always have $\infty = \delta(s,v) \leq v.d$ and thus we have $v.d = \infty = \delta(s,v)$.

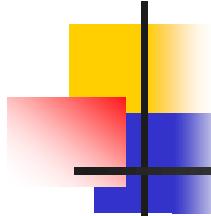




Lemma 24.13

- Let $G=(V,E)$ be a weighted, directed graph with weight function $w:E \rightarrow \mathbb{R}$, and let $(u,v) \in E$.
- Then, immediately after relaxing edge (u,v) by executing $\text{RELAX}(u,v,w)$, we have $v.d \leq u.d + w(u,v)$.
- Proof: if, just prior to relaxing edge (u,v) , we have $v.d > u.d + w(u,v)$, then before $v.d = u.d + w(u,v)$ afterward. If, instead, $v.d \leq u.d + w(u,v)$ just before the relaxation, then neither $u.d$ nor $v.d$ changes, and so $v.d \leq u.d + w(u,v)$ afterward.

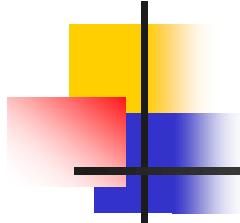




Convergence Property (Lemma 24.14)

- Let $G=(V,E)$ be a weighted, directed graph with weight function $w : E \rightarrow \mathbb{R}$.
- Let $s \in V$ the source vertex.
- Let $s \rightsquigarrow u \rightarrow v$ be a shortest path in G for some vertices $u,v \in V$.
- Let the graph is initialized by **INITILIZE-SINGLE-SOURCE**(G,s) and then a sequence of relaxation steps that includes the call **RELAX**(u,v,w) is executed on the edge of G .
- If $u.d = \delta(s,u)$ at any time prior to the call, then we have $v.d = \delta(s,v)$ at all times after the call.

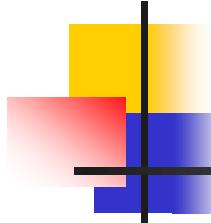




Convergence Property (Proof)

- By the upper-bound property, if $u.d = \delta(s,u)$ at some point prior to relaxing edge (u,v) , this equality holds thereafter.
- In particular, after relaxing edge (u,v) , we have $v.d \leq u.d + w(u,v) = \delta(s,u) + w(u,v) = \delta(s,v)$.
- By the upper-bound property, $v.d \geq \delta(s,v)$ from which we conclude that $v.d = \delta(s,v)$.
- Thus, this equality is maintained thereafter.

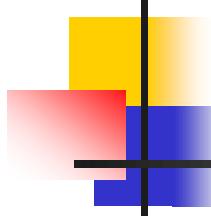




Shortest-paths Properties

- Triangle Inequality
 - For any edge $(u,v) \in E$, we have $\delta(s,v) \leq \delta(s,u) + w(u,v)$.
- Upper-bound property
 - We always have $v.d \geq \delta(s,v)$ for all $v \in V$, and once $v.d$ achieves the value $\delta(s,v)$, it never changes.
- No-path property
 - If there is no path from s to v , then we always have $v.d = \infty = \delta(s,v)$.
- Convergence property
 - If $s \rightsquigarrow u \rightarrow v$ be a shortest path in G for some vertices $u, v \in V$, 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.

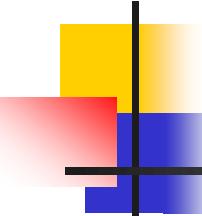




Path-relaxation Property (Lemma 24.15)

- If $p = \langle v_0, v_1, \dots, v_k \rangle$ is a shortest path from $s = v_0$ to v_k , and the edges of p are relaxed in the order $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$, then $v_k.d = \delta(s, v_k)$.
- This property holds regardless of any other relaxation steps that occur, even if they are intermixed with relaxations of the edges of p .

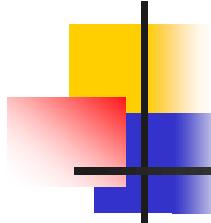




Path-relaxation Property (Proof)

- We show by induction that after i -th edge of path $p = \langle v_0, v_1, \dots, v_k \rangle$ is relaxed, we have $v_i.d = \delta(s, v_i)$.
- For the basis, $i=0$, and before any edges of p have been relaxed, we have from the initialization that $v_0.d = s.d = 0 = \delta(s, s)$. By the upper-bound property, the value of $s.d$ never changes after initialization.
- For the induction step, we assume that $v_{i-1}.d = \delta(s, v_{i-1})$, and we examine the relaxation of edge (v_{i-1}, v_i) . By the convergence property, after relaxation, we have $v_i.d = \delta(s, v_i)$, and this equality is maintained at all times thereafter.





Predecessor-subgraph Property (Lemma 24.17)

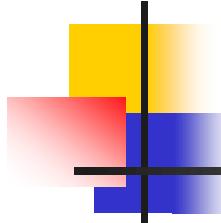
- Let $G=(V,E)$ be a weighted, directed graph with weight function $w:E \rightarrow \mathbb{R}$, let $s \in V$ be a source vertex, and assume that G contains no negative-weight cycles that are reachable from s .
- Let us call **INITIALIZE-SINGLE-SOURCE**(G, s) and then execute any sequence of relaxation steps on edges of G that produces $v.d = \delta(s,v)$ for all $v \in V$.
- Then, the predecessor subgraph G is a shortest-path tree rooted at s .
- Proof is omitted.





Bellman-Ford Algorithm



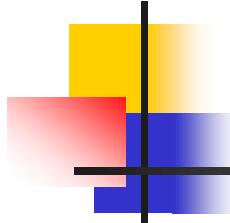


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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





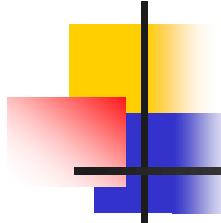
Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s)
 2. **for** $i=1$ **to** $|G.V|-1$
 3. **for** each edge $(u,v) \in G.E$
 4. RELAX(u, v, w)
 5. **for** each edge $(u,v) \in G.E$
 6. **if** $v.d > u.d + w(u,v)$
 7. **return** FALSE
 8. **return** TRUE
- Edge weights may be negative
- Relaxation:
Make $|V|-1$ passes,
relaxing each edge

Test whether negative
-weight cycle exists



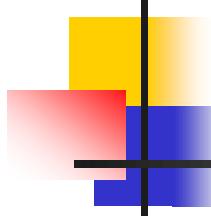


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s) $\leftarrow O(|V|)$
 2. **for** $i=1$ **to** $|G.V|-1$ $\leftarrow O(|V||E|)$
 for each edge $(u,v) \in G.E$
 3. RELAX(u, v, w)
 4. **for** each edge $(u,v) \in G.E$ $\leftarrow O(|E|)$
 if $v.d > u.d + w(u,v)$
 return FALSE
 5. **return** TRUE
- **Running time $O(|V||E|)$**





Lemma 24.2

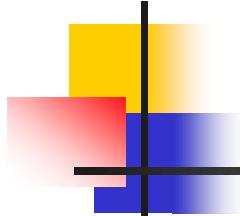
- Let $G=(V,E)$ be a weighted, directed graph with source s and weight function $w:E \rightarrow \mathbb{R}$.
- Assume that G contains no negative-weight cycles that are reachable from s .
- Then, after the $|V|-1$ iterations of the **for** loop of lines 2 - 4 of BELLMAN-FORD, we have $v.d = \delta(s,v)$ for all vertices that are reachable from s .



Lemma 24.2 (Proof)

- We prove the lemma by appealing to the path-relaxation property.
- Consider any vertex v that is reachable from s , and let $p = \langle v_0, v_1, \dots, v_k \rangle$, where $v_0 = s$ and $v_k = v$, be any shortest path from s to v .
- Because shortest paths are simple, p has at most $|V|-1$ edges, and so $k \leq |V|-1$. Each of the $|V|-1$ iterations of the **for** loop of lines 2–4 relaxes all $|E|$ edges.
- The edge (v_{i-1}, v_i) is one among the edges relaxed in the i -th iteration, for $i = 1, 2, \dots, k$.
- By the path-relaxation property, therefore, $v.d = v_k.d = \delta(s, v_k) = \delta(s, v)$.

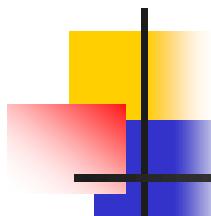




Corollary 24.3

- Let $G=(V,E)$ be a weighted, directed graph with source vertex s and weight function $w:E\rightarrow R$.
- Assume that G contains no negative-weight cycles that are reachable from s .
- Then, for each vertex $v \in V$, there is a path from s to v if and only if BELLMAN-FORD terminates with $v.d < \infty$ when it is run on G .
- The proof is left as Exercise 24.1-2.

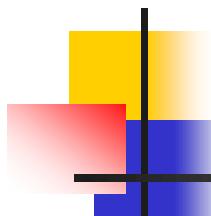




Correctness of the Bellman-Ford Algorithm

- Let BELLMAN-FORD be run on a weighted, directed graph $G=(V,E)$ with source s and weight function $w:E \rightarrow \mathbb{R}$.
- If G contains no negative-weight cycles that are reachable from s , then the algorithm returns TRUE, we have $v.d = \delta(s,v)$ for all vertices $v \in V$, and the predecessor subgraph G is a shortest-paths tree rooted at s .
- If G does contain a negative-weight cycle reachable from s , then the algorithm returns FALSE





Correctness of the Bellman-Ford Algorithm (Proof)

- Suppose that G contains no negative-weight cycles that are reachable from s .
- We first prove the claim that at termination, $v.d = \delta(s, v)$ for all vertices $v \in V$.
 - If vertex v is reachable from s , then Lemma 24.2 proves this.
 - If v is not reachable from s , then the claim follows from the no-path property.
 - Thus, the claim is proven.
- The predecessor-subgraph property, along with the claim, implies that $G.\pi$ is a shortest-paths tree.
- Now we use the claim to show that BELLMAN-FORD returns TRUE.
- At termination, we have for all edges $(u, v) \in E$,
 - $v.d = \delta(s, v)$
 $\leq \delta(s, u) + w(u, v)$ (by the triangle inequality)
 $= u.d + w(u, v)$
- and so none of the tests in line 6 causes BELLMAN-FORD to return FALSE.
- Thus, it returns TRUE.



Correctness of the Bellman-Ford Algorithm (Proof)

- Suppose G contains a negative-weight cycle that is reachable from s.
- Let this cycle $c = \langle v_0, v_1, \dots, v_k \rangle$, where $v_0 = v_k$ and $\sum_{i=1}^k w(v_{i-1}, v_i) < 0$ (**24.1**)
- Assume Bellman-Ford algorithm returns TRUE.
 $v_i \cdot d \leq v_{i-1} \cdot d + w(v_{i-1}, v_i)$ for $i = 1, 2, \dots, k$
- Summing the inequalities around cycle c

$$\begin{aligned}\sum_{i=1}^k v_i \cdot d &\leq \sum_{i=1}^k (v_{i-1} \cdot d + w(v_{i-1}, v_i)) \\ &= \sum_{i=1}^k v_{i-1} \cdot d + \sum_{i=1}^k w(v_{i-1}, v_i)\end{aligned}$$

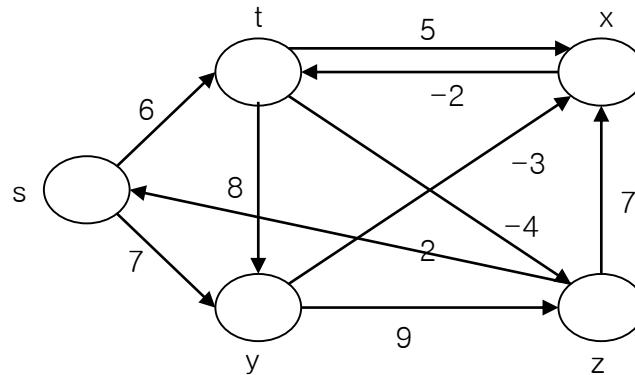
- Since $v_0 = v_k$, each vertex in c appears exactly once in each of the summations, $\sum_{i=1}^k v_i \cdot d = \sum_{i=1}^k v_{i-1} \cdot d$.
- Moreover, by Corollary 24.3, $v_i \cdot d$ is finite for $i = 1, 2, \dots, k$.
- Thus, $0 \leq \sum_{i=1}^k w(v_{i-1}, v_i)$ which contradicts inequality (**24.1**)



Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

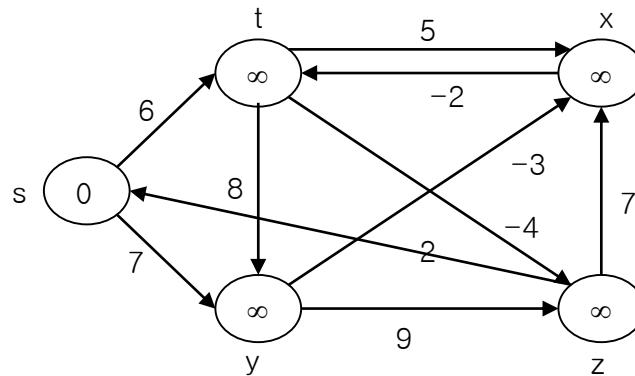
1. INITIALIZE-SINGLE-SOURCE(G, s)
2. **for** $i=1$ **to** $|G.V|-1$
 - 3. **for** each edge $(u,v) \in G.E$
 - 4. RELAX(u, v, w)
 - 5. **for** each edge $(u,v) \in G.E$
 - 6. **if** $v.d > u.d + w(u,v)$
 - 7. **return** FALSE
 - 8. **return** TRUE



Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s)
2. **for** $i=1$ **to** $|G.V|-1$
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7. **return** FALSE
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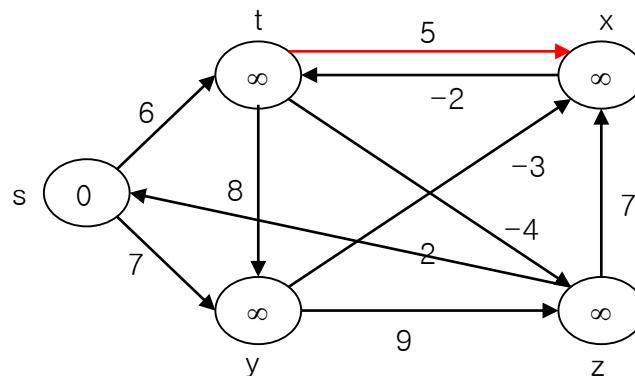


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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2. **for** $i=1$ **to** $|G.V|-1$
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4. RELAX(u, v, w)
5. **for** each edge $(u,v) \in G.E$
6. **if** $v.d > u.d + w(u,v)$
7. **return** FALSE
8. **return** TRUE

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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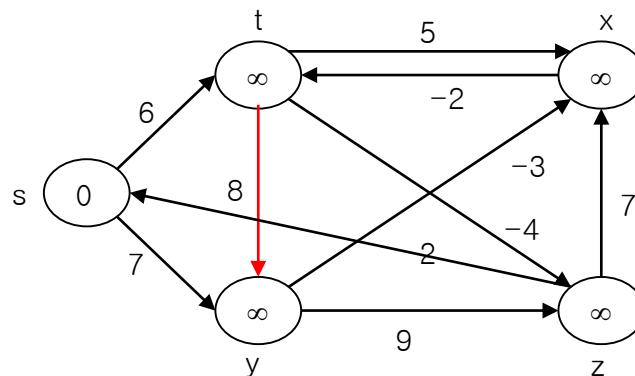


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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2. **for** $i=1$ **to** $|G.V|-1$
3. **for** each edge $(u,v) \in G.E$
4. RELAX(u, v, w)
5. **for** each edge $(u,v) \in G.E$
6. **if** $v.d > u.d + w(u,v)$
7. **return** FALSE
8. **return** TRUE

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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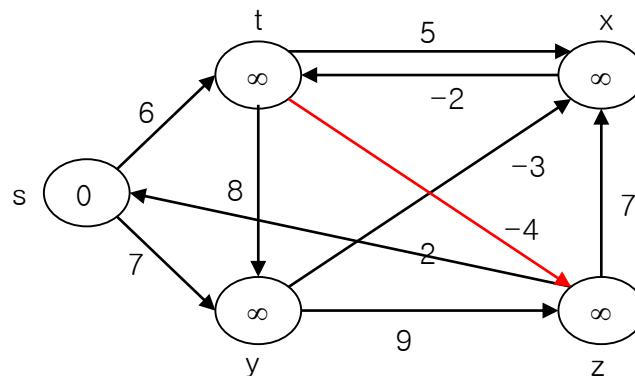


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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5. **for** each edge $(u,v) \in G.E$
6. **if** $v.d > u.d + w(u,v)$
7. **return** FALSE
8. **return** TRUE

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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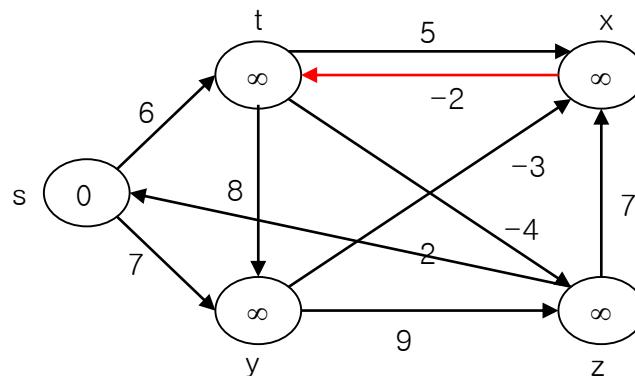


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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2. **for** $i=1$ **to** $|G.V|-1$
3. **for** each edge $(u,v) \in G.E$
4. RELAX(u, v, w)
5. **for** each edge $(u,v) \in G.E$
6. **if** $v.d > u.d + w(u,v)$
7. **return** FALSE
8. **return** TRUE

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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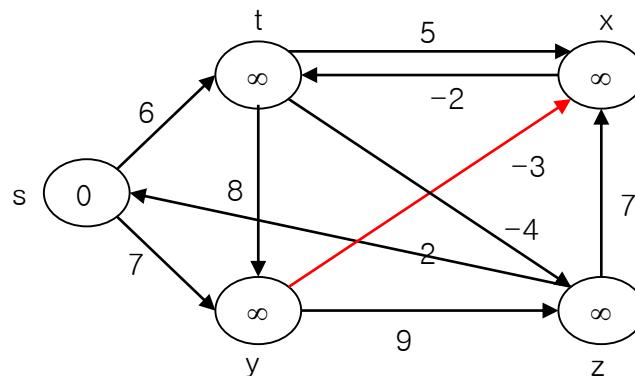


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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2. **for** $i=1$ **to** $|G.V|-1$
3. **for** each edge $(u,v) \in G.E$
4. RELAX(u, v, w)
5. **for** each edge $(u,v) \in G.E$
6. **if** $v.d > u.d + w(u,v)$
7. **return** FALSE
8. **return** TRUE

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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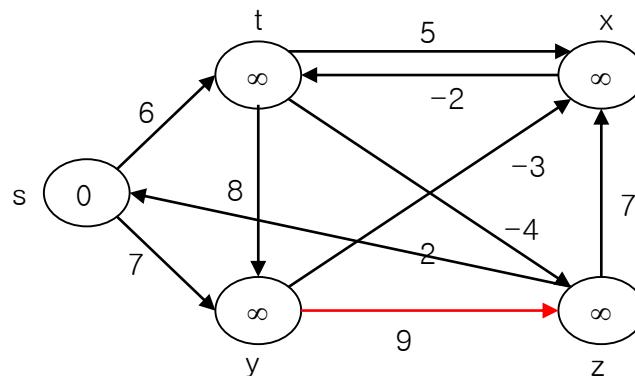


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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2. **for** $i=1$ **to** $|G.V|-1$
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4. RELAX(u, v, w)
5. **for** each edge $(u,v) \in G.E$
6. **if** $v.d > u.d + w(u,v)$
7. **return** FALSE
8. **return** TRUE

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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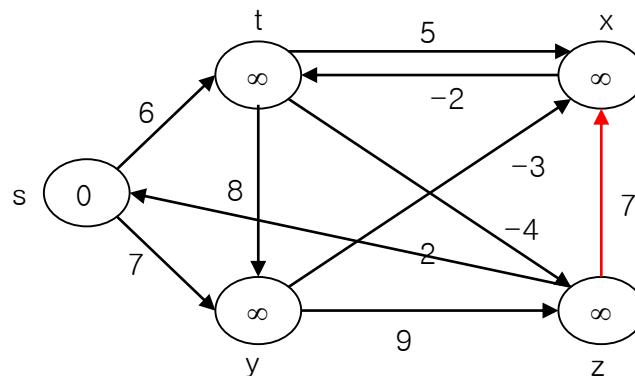


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	---------	---------	---------	---------	---------	---------	---------	---------	---------	---------

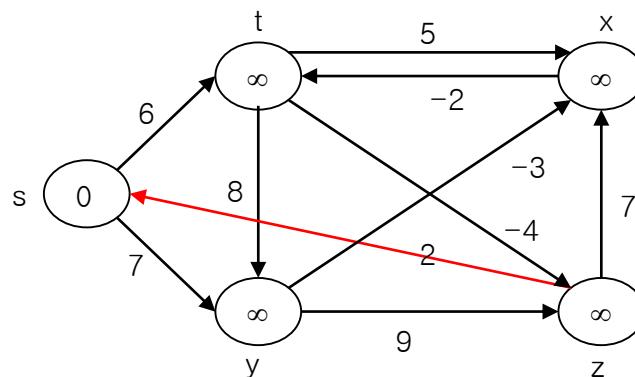


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

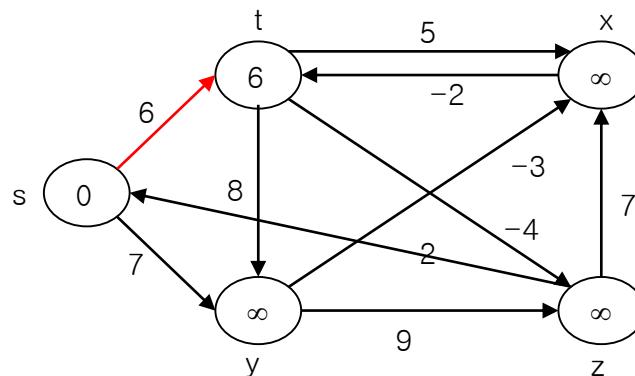


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	---------	---------	---------	---------	---------	---------	---------	---------	---------	---------

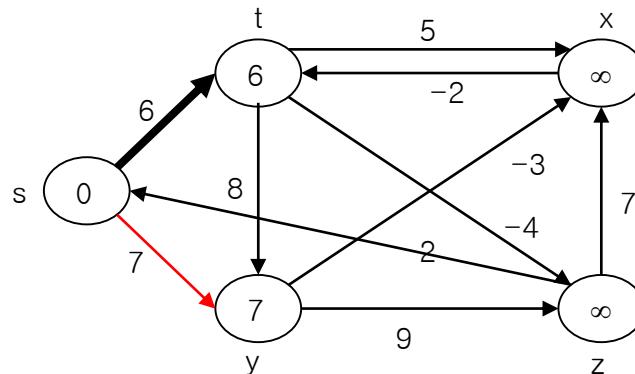


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 1$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

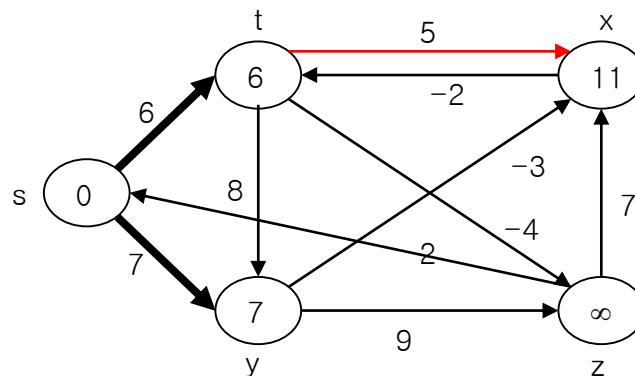


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

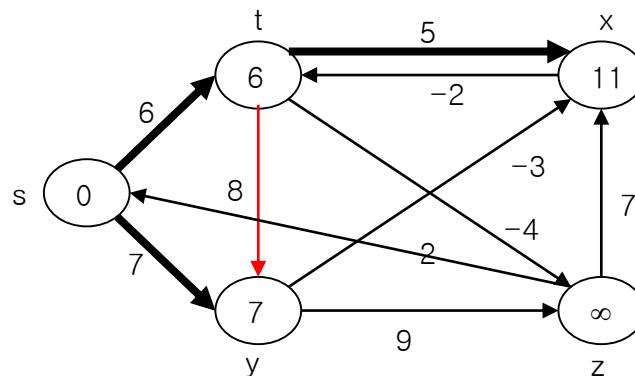


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

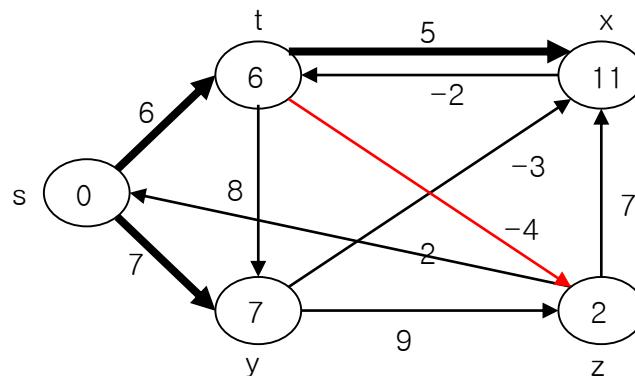


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	----------------	-------	-------	-------	-------	-------	-------	-------

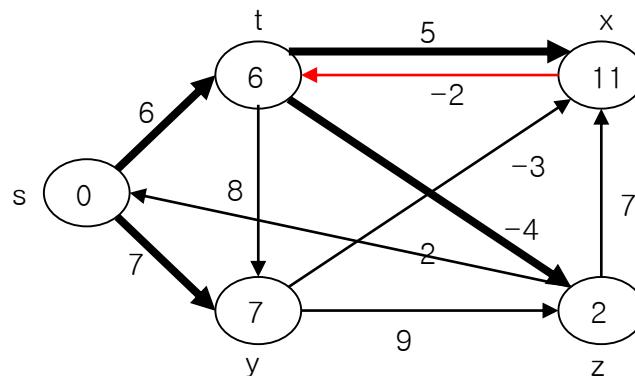


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

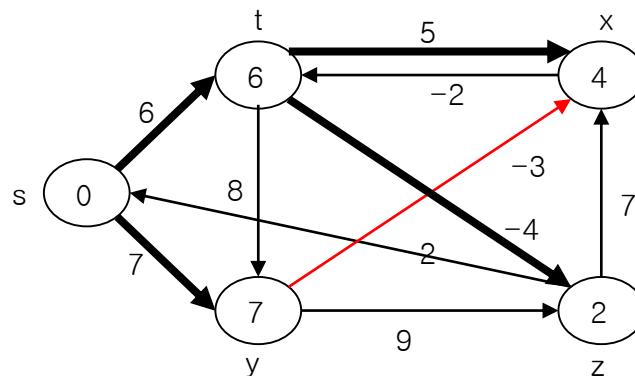


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

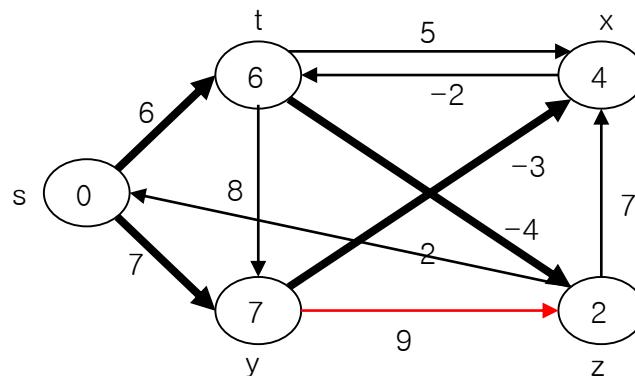


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

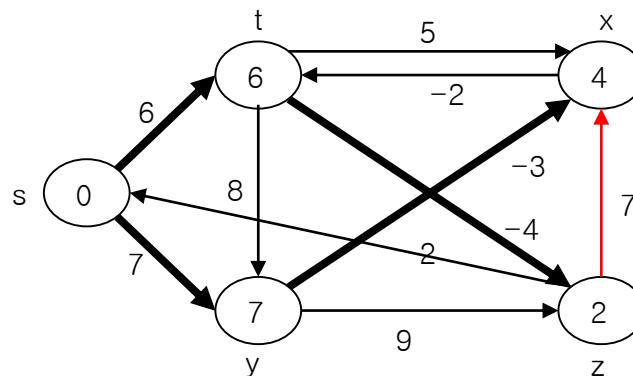


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

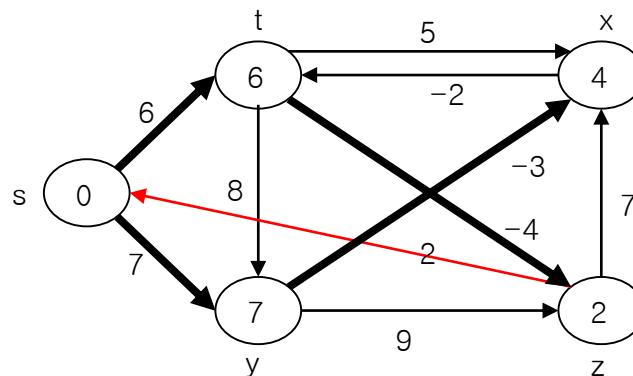


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

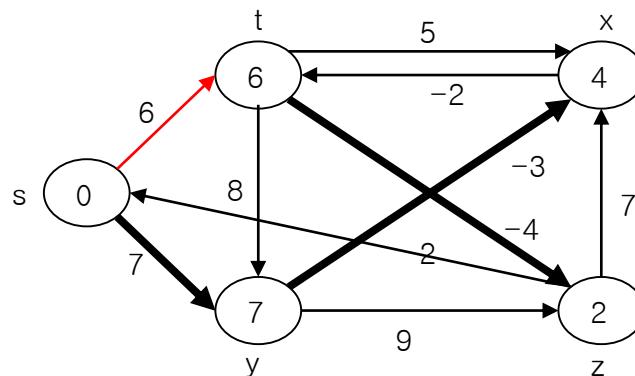


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

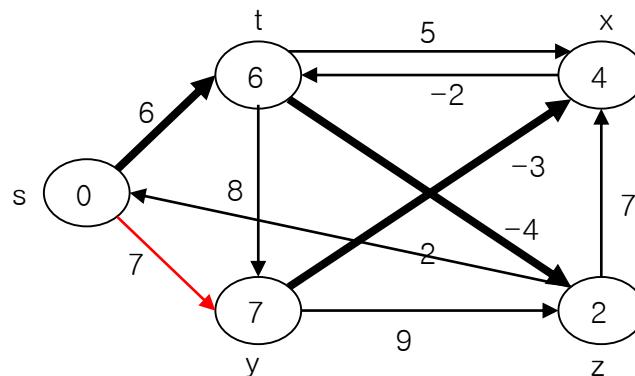


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 2$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

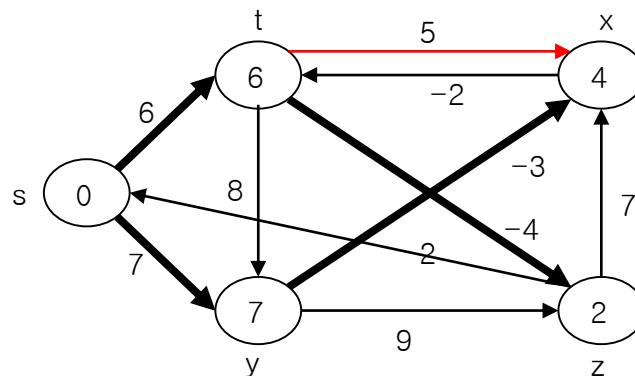


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

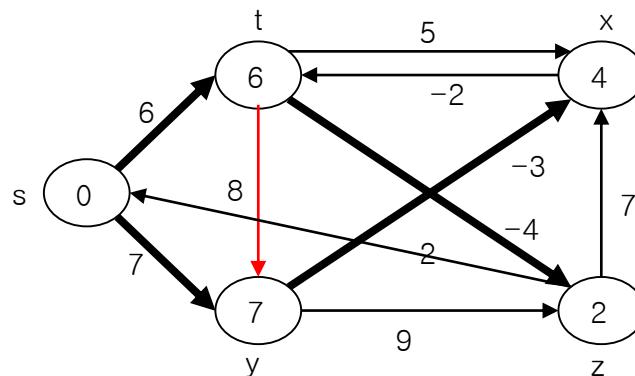


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

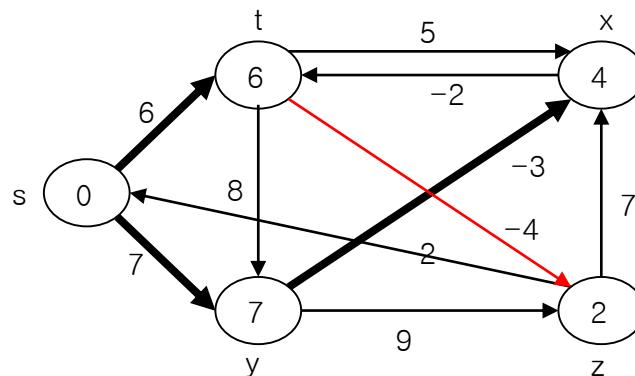


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	----------------	-------	-------	-------	-------	-------	-------	-------

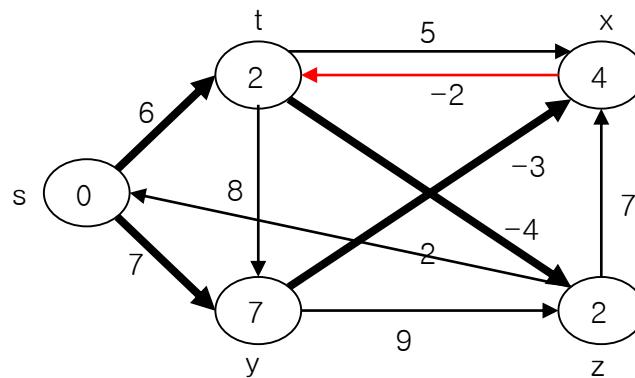


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

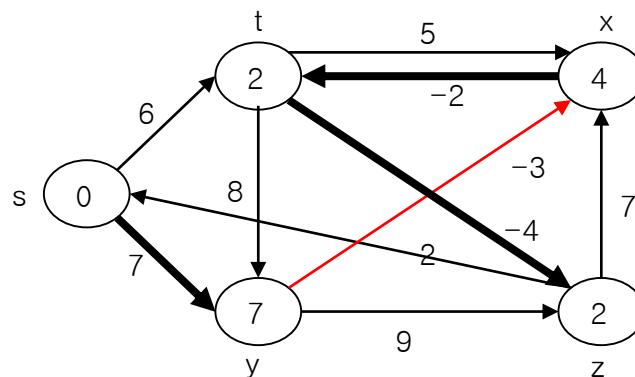


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

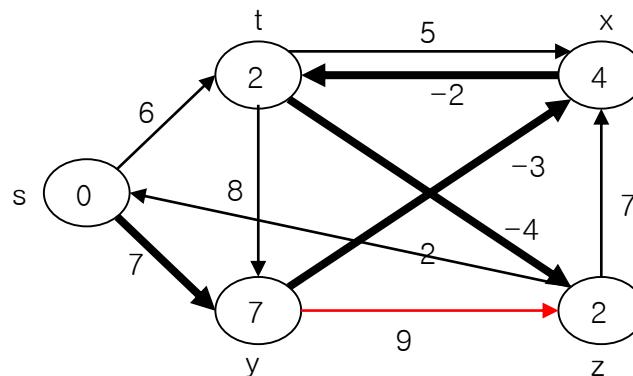


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

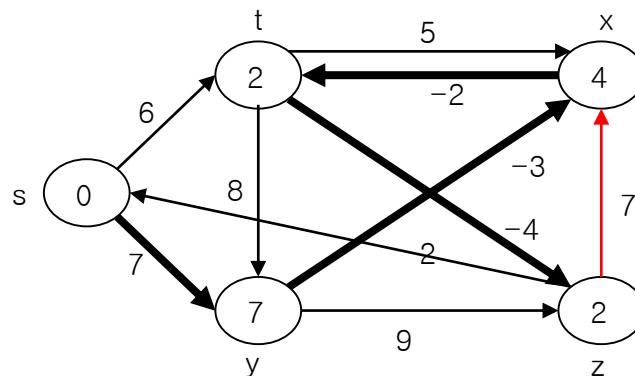


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

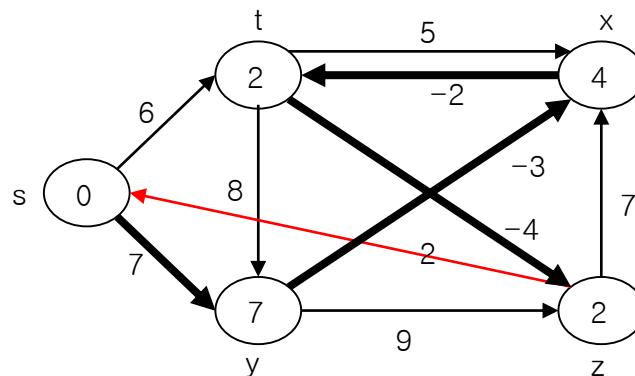


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

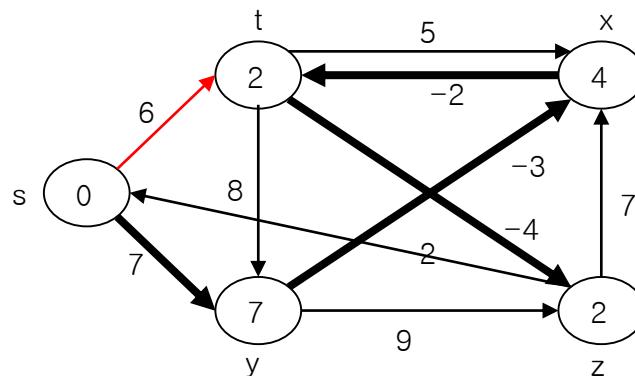


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

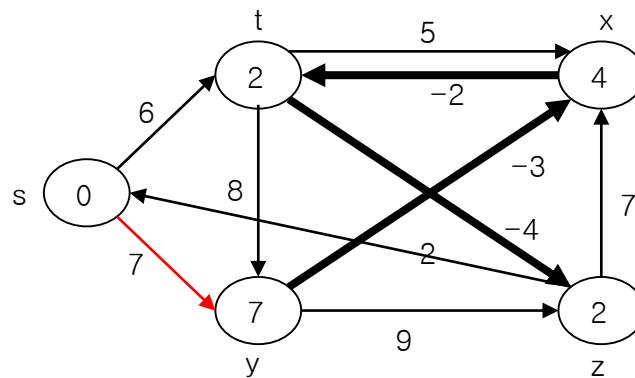


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 3$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

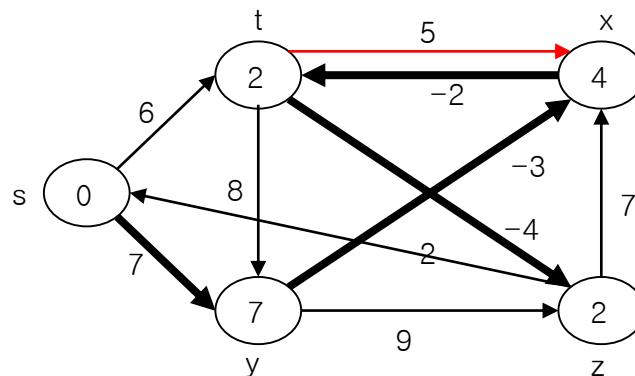


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 4$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

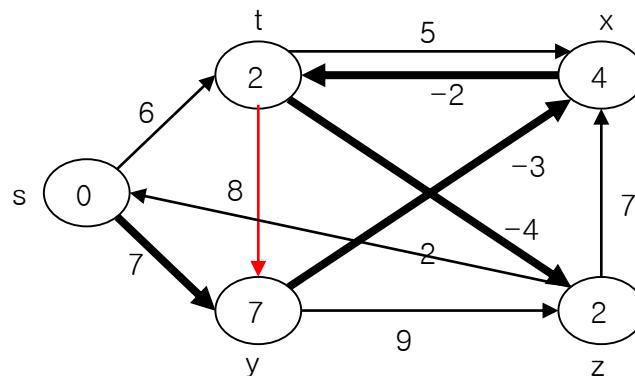


Bellman-Ford Algorithm

BELLMAN-FORD(G, w, s)

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

$i = 4$



G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

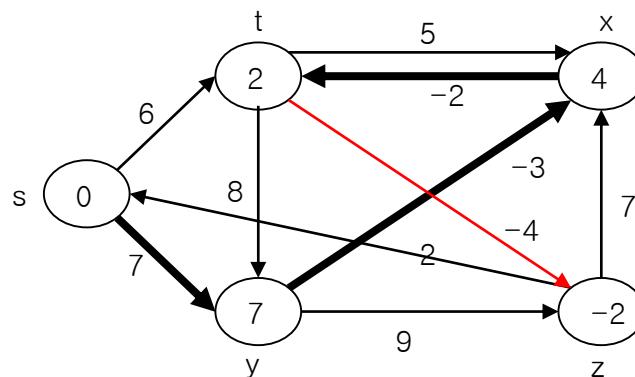


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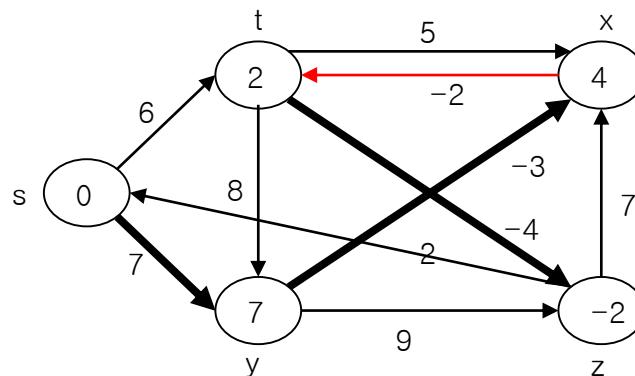


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-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

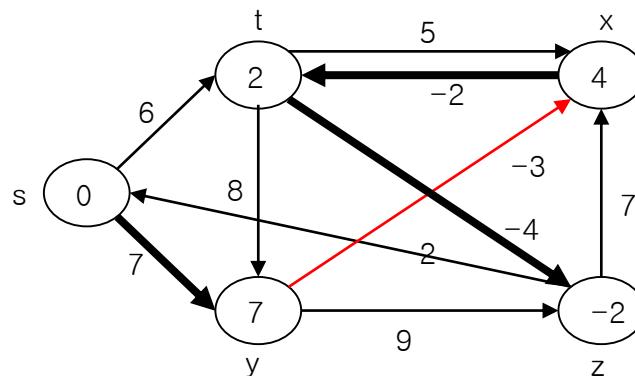


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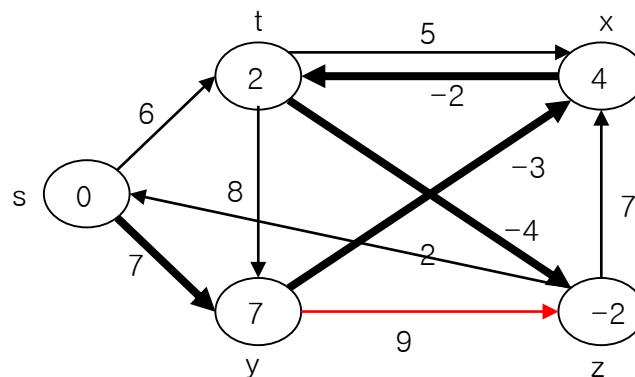


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-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

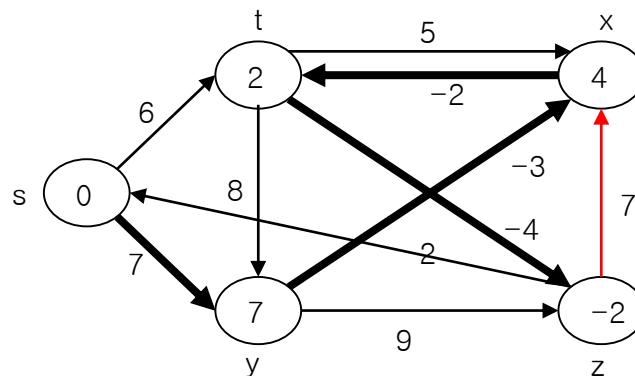


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G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

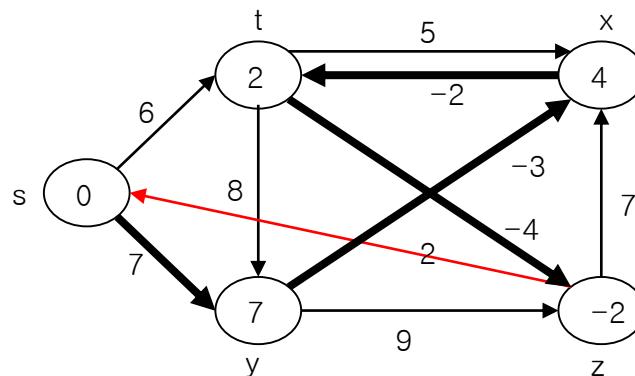


Bellman-Ford Algorithm

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G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

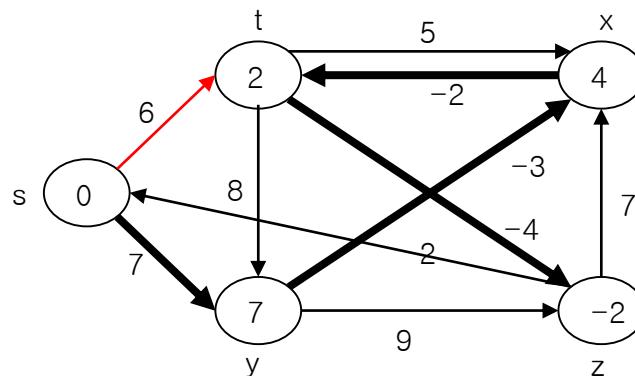


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-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

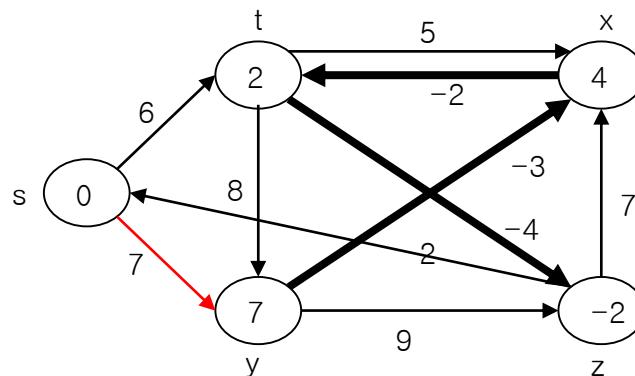


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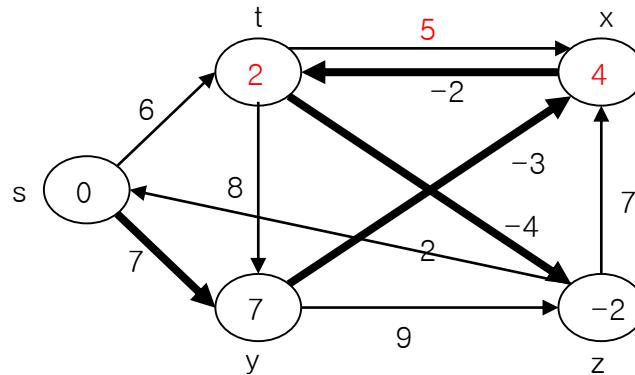
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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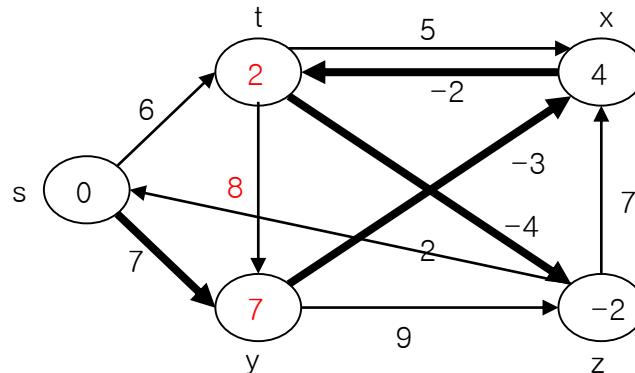
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
-----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------



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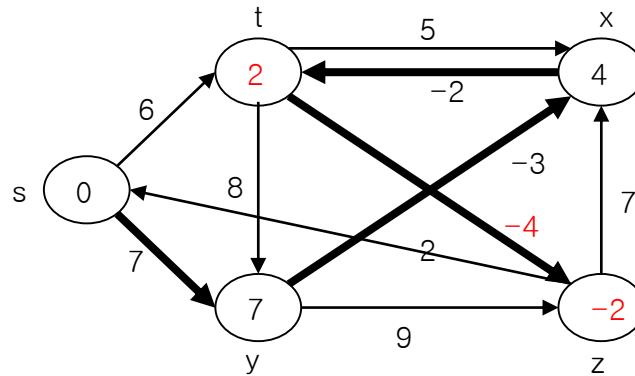
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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Bellman-Ford Algorithm

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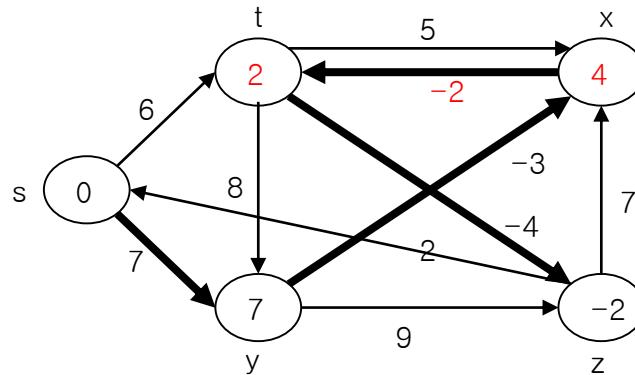
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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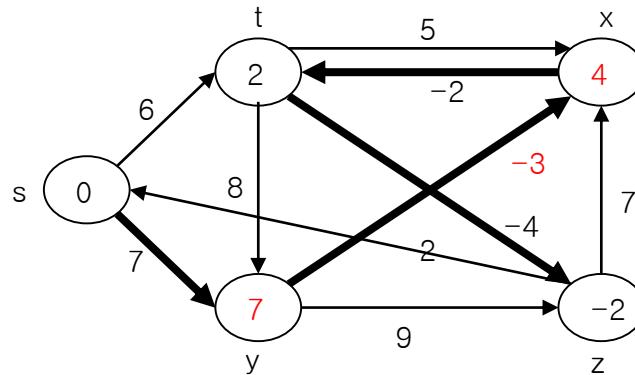
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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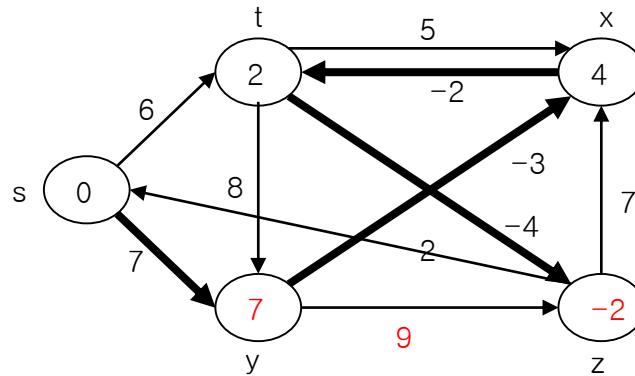
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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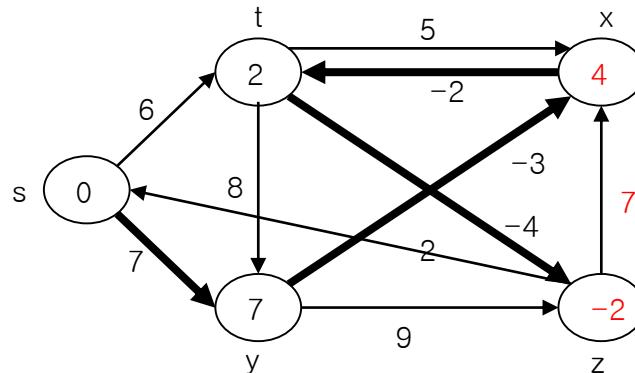
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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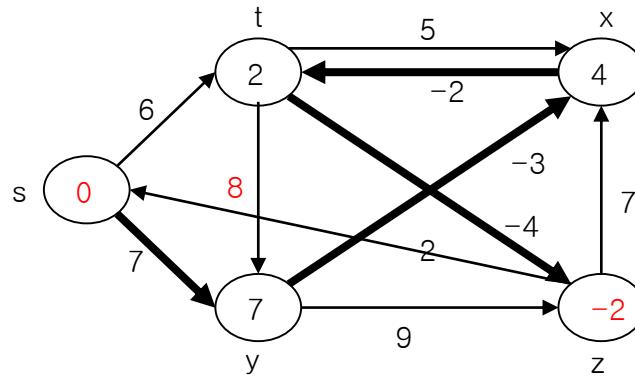
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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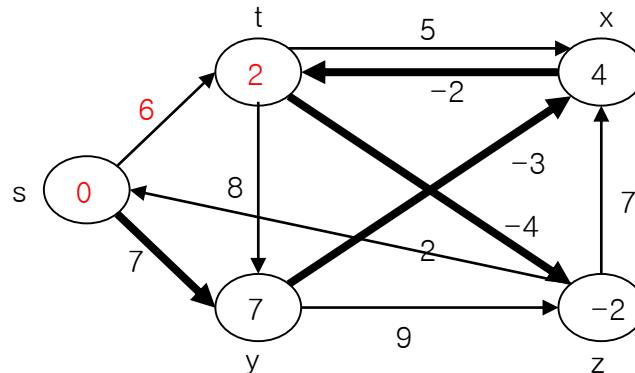
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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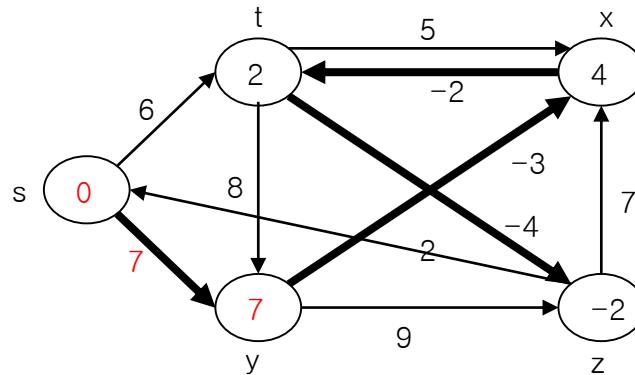
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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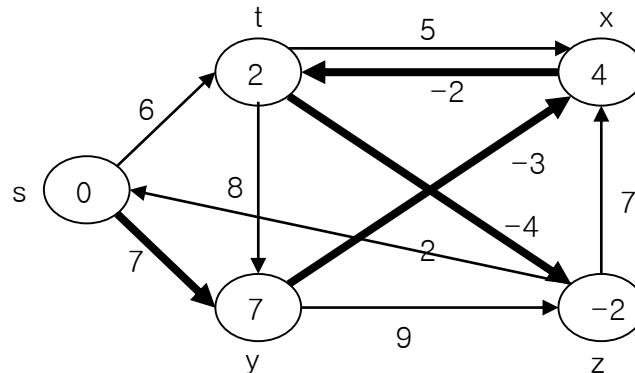
G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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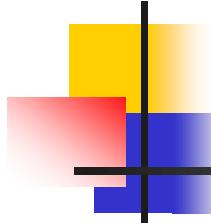


G.E	(t,x)	(t,y)	(t,z)	(x,t)	(y,x)	(y,z)	(z,x)	(z,s)	(s,t)	(s,y)
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Single-source Shortest Paths in Directed Acyclic Graphs (DAG)

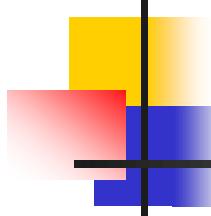




Single-source Shortest Paths in Directed Acyclic Graphs

- Problem: finding shortest paths in DAG
 - Bellman-Ford takes $O(|V||E|)$ time.
 - How can we do better?
 - Idea: use the topological sort.





DAG Shortest Path

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

1. topologically sort the vertices of G
2. INITIALIZE-SINGLE-SOURCE(G, s)
3. **for** each vertex u , taken in topologically sorted order
4. **for** each vertex $v \in G.\text{Adj}[u]$
5. RELAX(u, v, w)



DAG shortest path

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

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

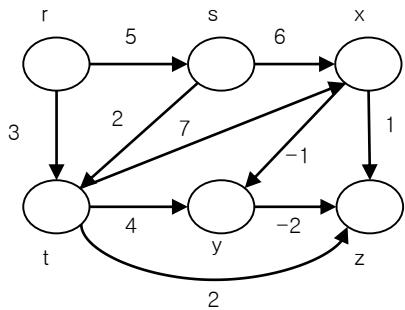
running time is $O(|V| + |E|)$



DAG Shortest Path

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

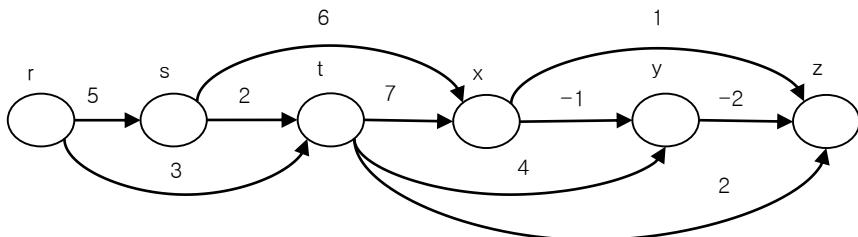
1. topologically sort the vertices of G
2. INITIALIZE-SINGLE-SOURCE(G, s)
3. **for** each vertex u , taken in topologically sorted order
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DAG Shortest Path

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

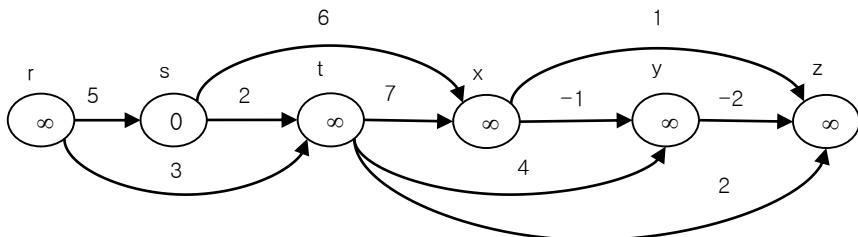
1. topologically sort the vertices of G
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3. **for** each vertex u , taken in topologically sorted order
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DAG Shortest Path

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

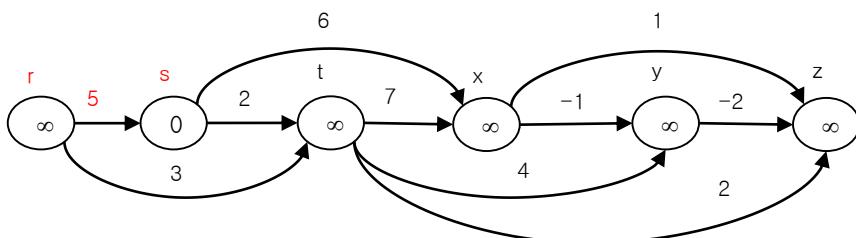
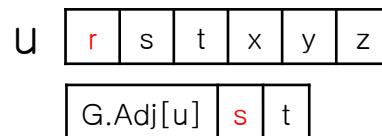
1. topologically sort the vertices of G
2. **INITIALIZE-SINGLE-SOURCE(G, s)**
3. **for** each vertex u , taken in topologically sorted order
4. **for** each vertex $v \in G.\text{Adj}[u]$
5. RELAX(u, v, w)



DAG Shortest Path

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

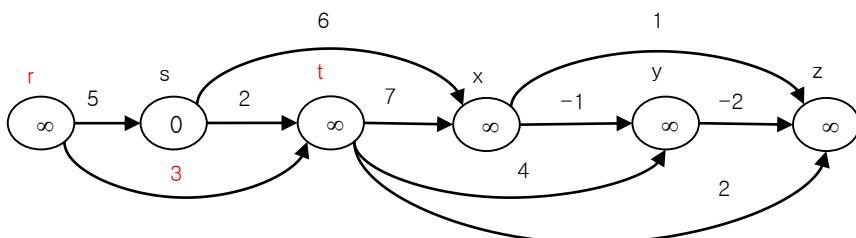
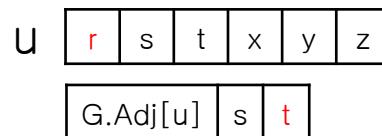
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4. **for** each vertex $v \in G.\text{Adj}[u]$
5. RELAX(u, v, w)



DAG Shortest Path

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

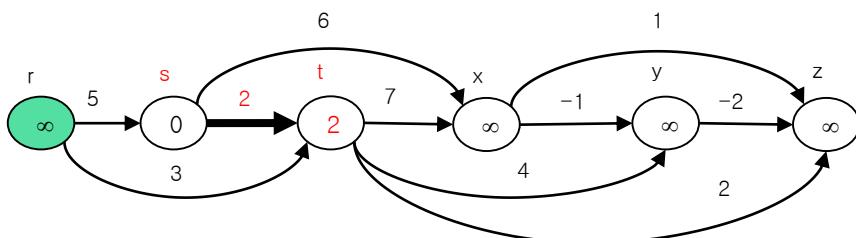
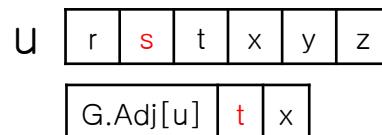
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DAG Shortest Path

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

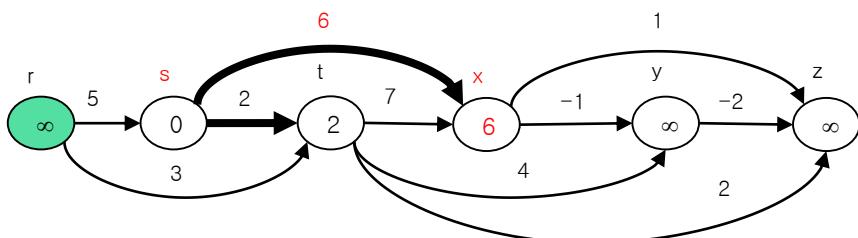
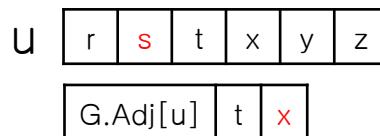
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DAG Shortest Path

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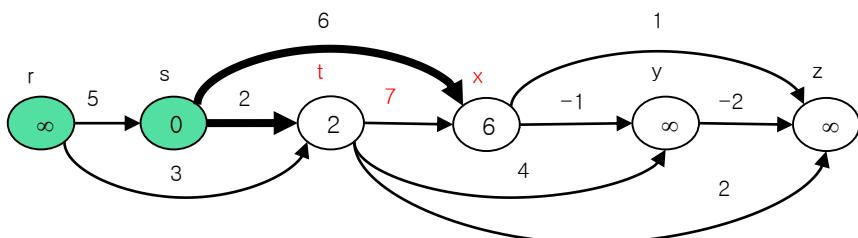
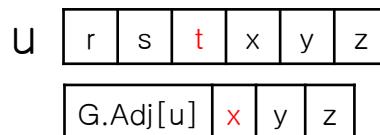
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DAG Shortest Path

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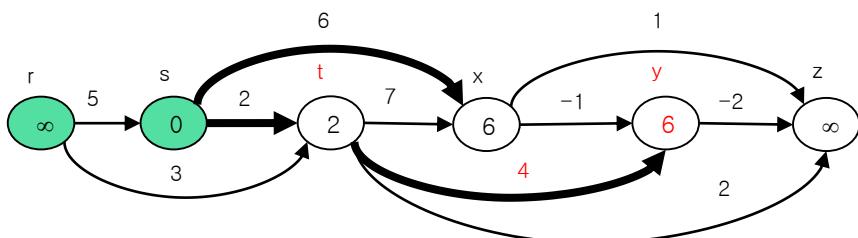
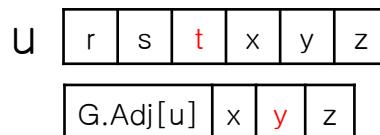
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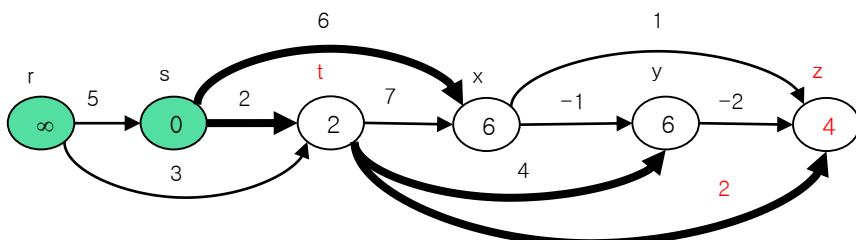


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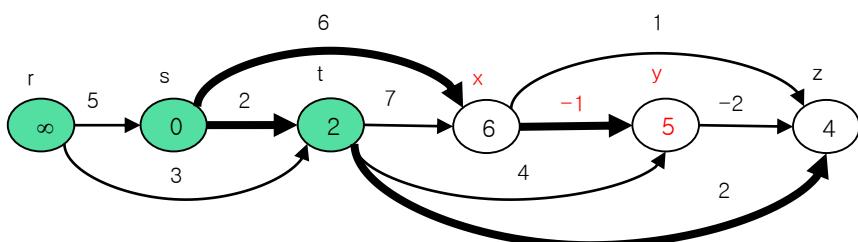
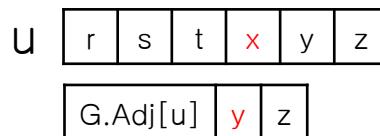
u	r	s	t	x	y	z
	G.Adj[u]	x	y			z



DAG Shortest Path

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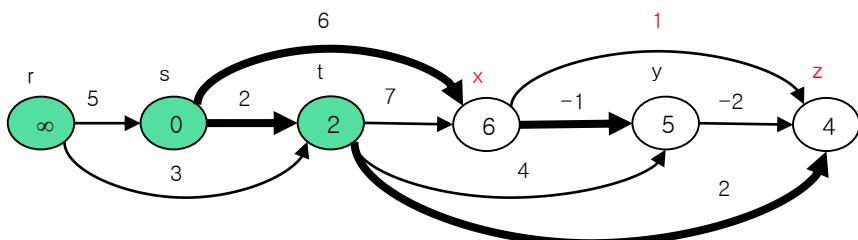


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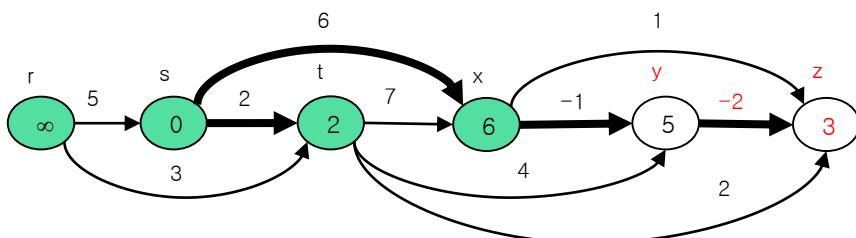
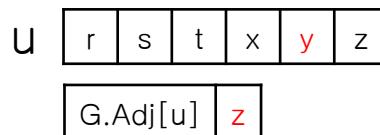
u	r	s	t	x	y	z
	G.Adj[u]	y	z			



DAG Shortest Path

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

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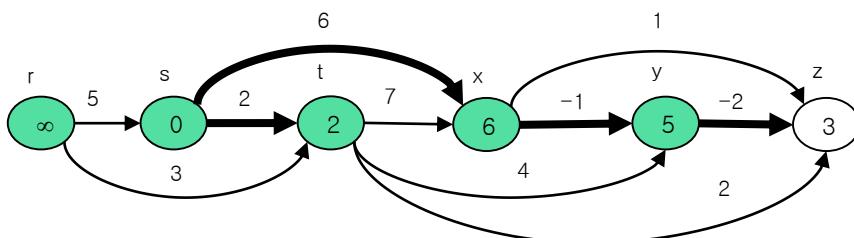
DAG Shortest Path

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5. RELAX(u, v, w)

u

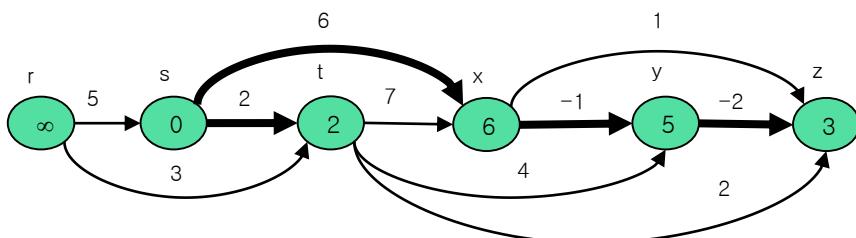
r	s	t	x	y	z
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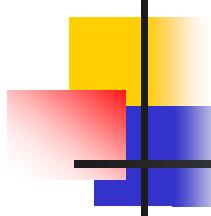


DAG Shortest Path

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

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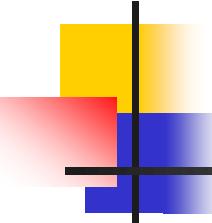




Theorem 24.5

- If a weighted, directed graph $G=(V,E)$ has source vertex s and no cycles, then at the termination of the DAG-SHORTEST-PATHS procedure, $v.d=\delta(s,v)$ for all vertices $v \in V$, and the predecessor subgraph G_π is a shortest-paths tree.

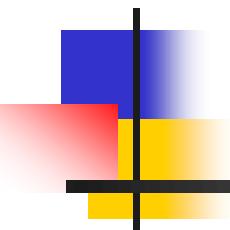




Theorem 24.5 (Proof)

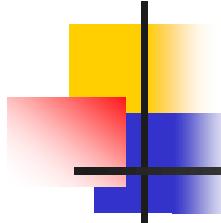
- We first show that $v.d = \delta(s, v)$ for all vertices $v \in V$ at termination.
- If v is not reachable from s , then $v.d = \delta(s, v) = \infty$ by the no-path property.
- Now, suppose that v is reachable from s , so that there is a shortest path $p = \langle v_0, v_1, \dots, v_k \rangle$ where $v_0 = s$ and $v_k = v$.
- Because we process the vertices in topologically sorted order, we relax the edges on p in the order $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$.
- The path-relaxation property implies that $v.d = \delta(s, v)$ at termination for $i = 1, 2, \dots, k$.
- Finally, by the predecessor subgraph property, G_π is a shortest-paths tree.





Dijkstra's Algorithm





Dijkstra's Algorithm

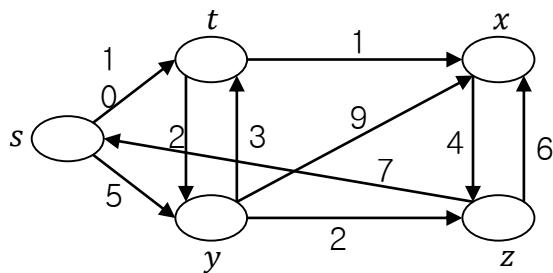
- On weighted, directed graph $G=(V,E)$ for which all edge weights are nonnegative.
- The running time of Dijkstra's algorithm is lower than that of the Bellman-Ford algorithm.



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

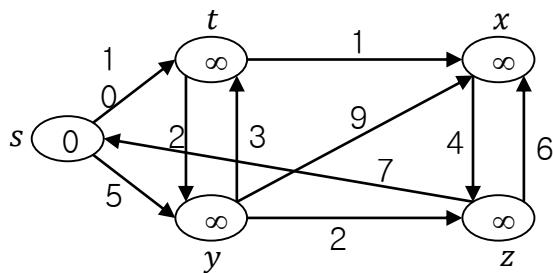
1. INITIALIZE-SINGLE-SOURCE(G, s)
2. $S = \emptyset$
3. $Q = G.V$
4. **while** $Q \neq \emptyset$
5. $u = \text{Extract-Min}(Q)$
6. $S = S \cup \{u\}$
7. **for** each vertex $v \in G.\text{Adj}[u]$
 RELAX(u, v, w)
- 8.



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

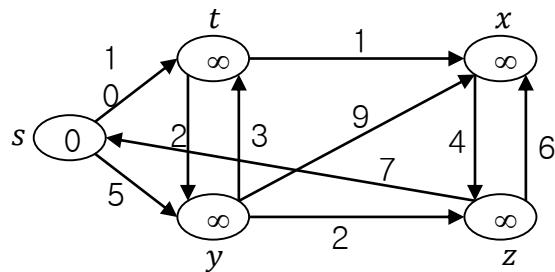
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Dijkstra's Algorithm

DIJKSTRA(G, w, s)

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2. $S = \emptyset$
3. $Q = G.V$
4. **while** $Q \neq \emptyset$ $S = \emptyset$
5. $u = \text{Extract-Min}(Q)$
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8. RELAX(u, v, w)



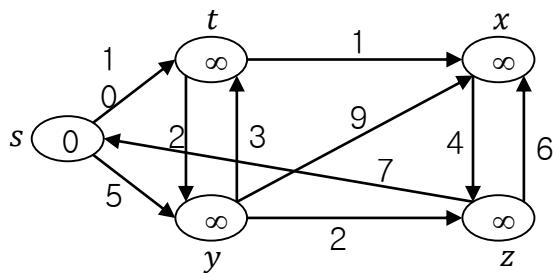
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$S = \emptyset$

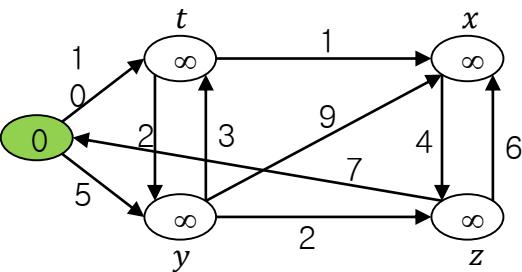
Q	$G.V$	s	t	x	y	z
d	0	∞	∞	∞	∞	∞



Dijkstra's Algorithm

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7. **for** each vertex $v \in G.\text{Adj}[u]$
 RELAX(u, v, w)
- 8.



$$S=\{s\}$$

Q	G.V	t	x	y	z
d	∞	∞	∞	∞	∞

$$u=s$$

$$G.\text{adj}[s] = \{t, y\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

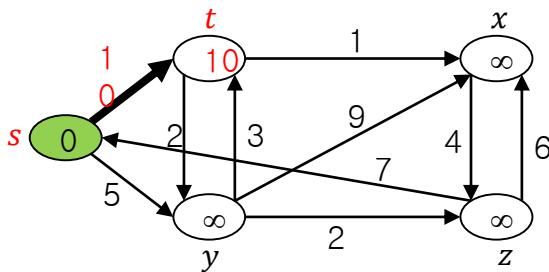
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7. **for each vertex** $v \in G.\text{Adj}[u]$
8. RELAX(u, v, w)

$S=\{s\}$

Q	$G.V$	t	x	y	z
d	10	∞	∞	∞	

$u=s$

$G.\text{adj}[s] = \{t, y\}$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

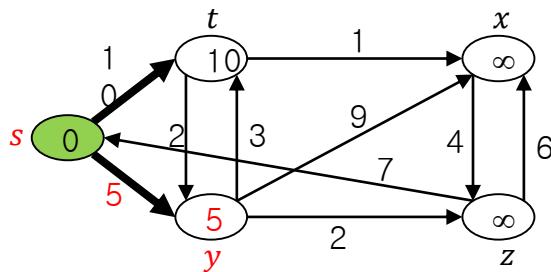
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8. RELAX(u, v, w)

$S=\{s\}$

Q	$G.V$	t	x	y	z
d	10	∞	5	∞	

$u=s$

$G.\text{adj}[s] = \{t, y\}$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

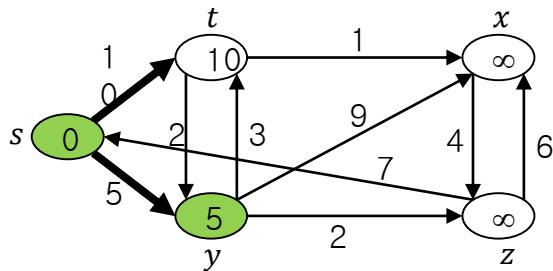
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$$S=\{s, y\}$$

Q	$G.V$	t	x	z
d	10	∞	∞	

$$u=y$$

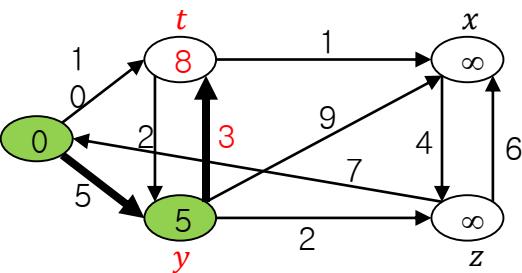
$$G.\text{adj}[y] = \{t, x, z\}$$



Dijkstra's Algorithm

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$$S = \{s, y\}$$

Q	$G.V$	t	x	z
d	8	∞	∞	

$$u = y$$

$$G.\text{adj}[y] = \{t, x, z\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

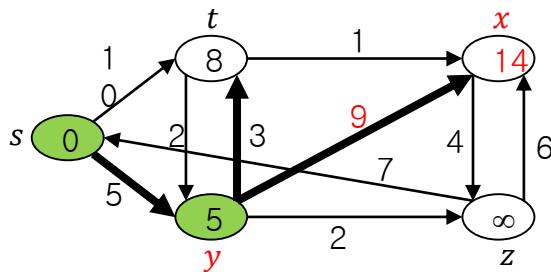
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$$S = \{s, y\}$$

Q	$G.V$	t	x	z
d	8	14	∞	

$$u = y$$

$$G.\text{adj}[y] = \{t, x, z\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

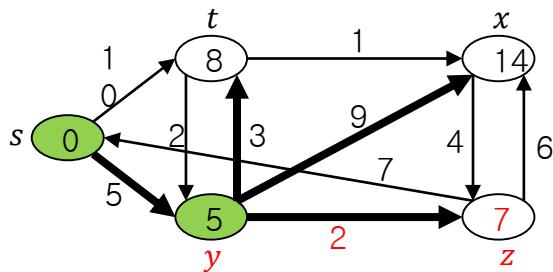
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8. RELAX(u, v, w)

$$S=\{s, y\}$$

Q	$G.V$	t	x	z
	d	8	14	7

$$u=y$$

$$G.\text{adj}[y] = \{t, x, z\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

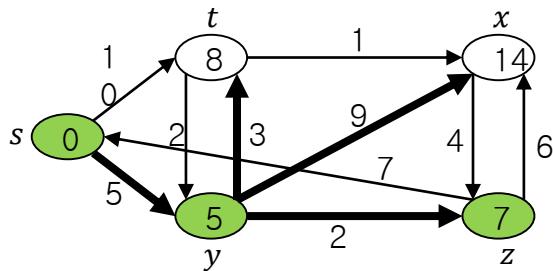
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- 8.

$$S = \{s, y, z\}$$

Q	$G.V$	t	x
	d	8	14

$$u = z$$

$$G.\text{adj}[z] = \{x, s\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

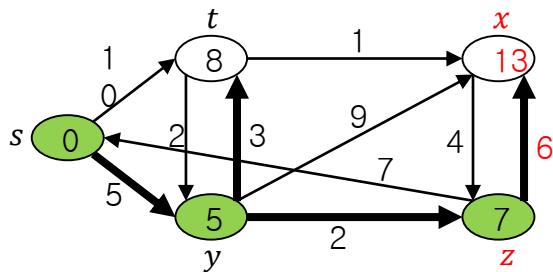
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8. RELAX(u, v, w)

$$S = \{s, y, z\}$$

Q	$G.V$	t	x
	d	8	13

$$u = z$$

$$G.\text{adj}[z] = \{x, s\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

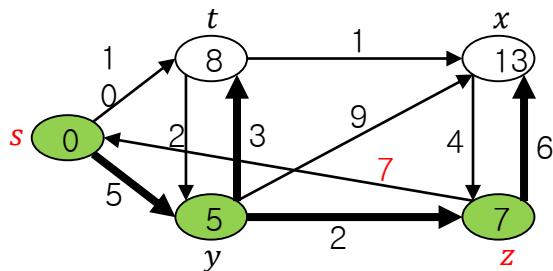
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$$S = \{s, y, z\}$$

Q	$G.V$	t	x
	d	8	13

$$u = z$$

$$G.\text{adj}[z] = \{x, s\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

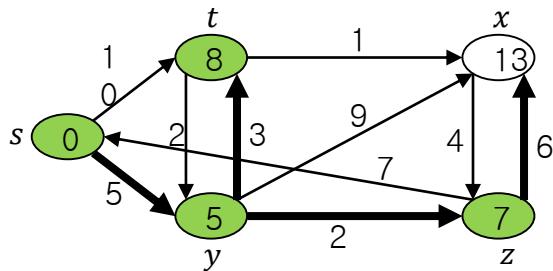
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6. $S = S \cup \{u\}$
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8. RELAX(u, v, w)

$$S = \{s, y, z, t\}$$

$G.V$	x
d	13

$$u = t$$

$$G.\text{adj}[t] = \{x, y\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

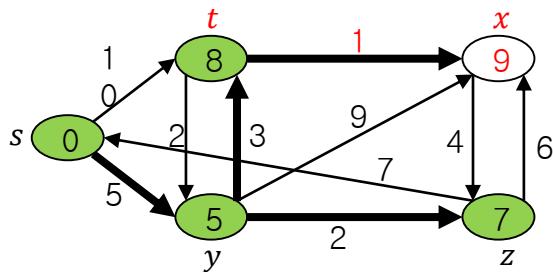
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$$S = \{s, y, z, t\}$$

$G.V$	x
d	9

$$u = t$$

$$G.\text{adj}[t] = \{x, y\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

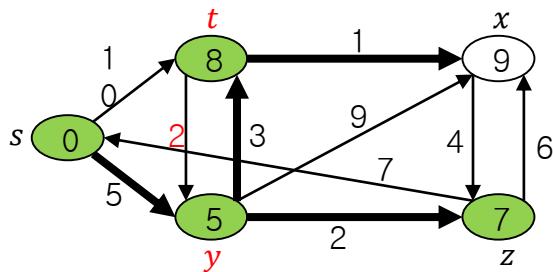
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$$S = \{s, y, z, t\}$$

$G.V$	x
d	9

$$u = t$$

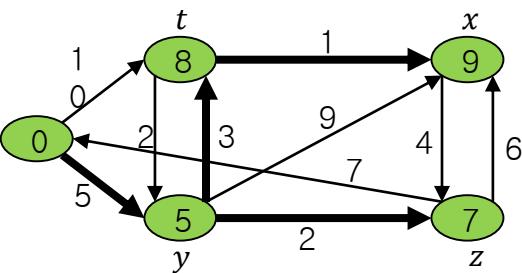
$$G.\text{adj}[t] = \{x, y\}$$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

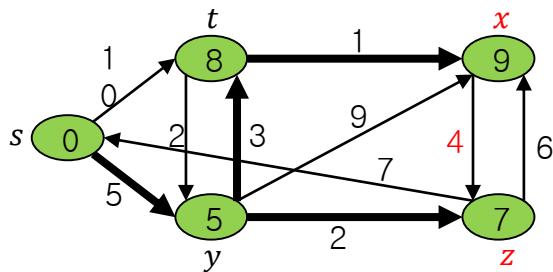
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3. $Q = G.V$
4. **while** $Q \neq \emptyset$
5. $u = \text{Extract-Min}(Q)$
6. $S = S \cup \{u\}$
7. **for** each vertex $v \in G.\text{Adj}[u]$
8. RELAX(u, v, w)



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

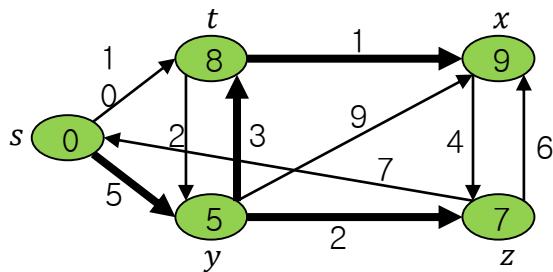
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6. $S = S \cup \{u\}$
 $Q = \emptyset$
7. **for each vertex** $v \in G.\text{Adj}[u]$
8. RELAX(u, v, w)
 $u = x$
 $G.\text{adj}[x] = \{z\}$



Dijkstra's Algorithm

DIJKSTRA(G, w, s)

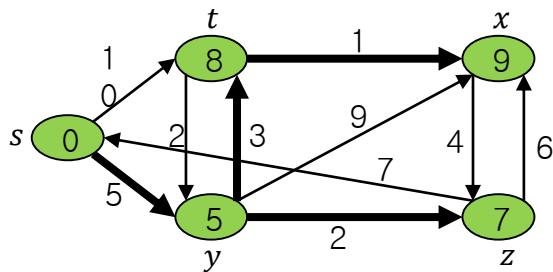
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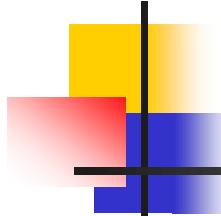


Dijkstra's Algorithm

DIJKSTRA(G, w, s)

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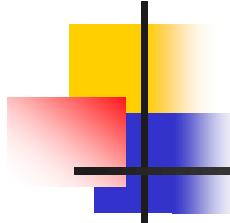




Running Time of Dijkstra's Algorithm

- It depends on implementations of the min-priority queue Q.
- If we implement Q as a binary min-heap,
 - EXTRACT-MIN takes $O(\lg |V|)$ time.
 - DECREASE-KEY takes $O(\lg |V|)$ time.
- If we implement Q as a simple array,
 - EXTRACT-MIN takes $O(|V|)$ time.
 - DECREASE-KEY $O(1)$ time.
- If we implement Q as a Fibonacci heap,
 - EXTRACT-MIN takes $O(\lg |V|)$ amortized time.
 - DECREASE-KEY $O(1)$ amortized time.



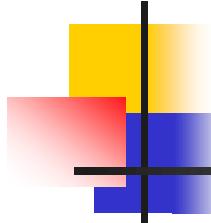


Dijkstra's Algorithm

DIJKSTRA(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s)
2. $S = \emptyset$
3. $Q = G.V$
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 5. $u = \text{Extract-Min}(Q)$
 6. $S = S \cup \{u\}$
 7. **for** each vertex $v \in G.\text{Adj}[u]$
 8. RELAX(u, v, w)





Dijkstra's Algorithm

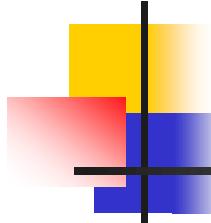
- min-priority queue : array

DIJKSTRA(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s) $\leftarrow O(|V|)$
2. $S = \emptyset$ $\leftarrow O(1)$
3. $Q = G.V$ $\leftarrow O(|V|)$
4. **while** $Q \neq \emptyset$
5. $u = \text{Extract-Min}(Q)$ $\leftarrow O(|V|^2)$
6. $S = S \cup \{u\}$ $\leftarrow O(|V|)$
7. **for** each vertex $v \in G.\text{Adj}[u]$ $\leftarrow O(|E|)$
 RELAX(u, v, w)

Dijkstra's algorithm running time is $O(|V|^2)$





Dijkstra's Algorithm

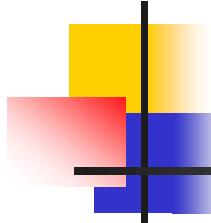
- min-priority queue : binary min-heap

DIJKSTRA(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s) $\leftarrow O(|V|)$
2. $S = \emptyset \leftarrow O(1)$
3. $Q = G.V \leftarrow O(|V|)$
4. **while** $Q \neq \emptyset$
5. $u = \text{Extract-Min}(Q) \leftarrow O(|V| \lg |V|)$
6. $S = S \cup \{u\} \leftarrow O(|V|)$
7. **for** each vertex $v \in G.\text{Adj}[u]$ $\leftarrow O(|E| \lg |V|)$
8. RELAX(u, v, w)

- Running time:
 - $O((|V|+|E|) \lg |V|)$, if all vertices are reachable $\rightarrow O(|E| \lg |V|)$.
 - Better than $O(|V|^2)$, if the graph is sufficiently sparse: $|E| = o(|V|^2 / \lg |V|)$.





Dijkstra's Algorithm

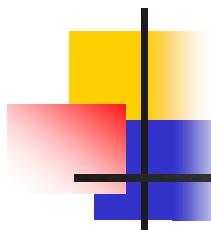
- min-priority queue : Fibonacci heap

DIJKSTRA(G, w, s)

1. INITIALIZE-SINGLE-SOURCE(G, s) $\leftarrow O(|V|)$
2. $S = \emptyset$ $\leftarrow O(1)$
3. $Q = G.V$ $\leftarrow O(|V|)$
4. **while** $Q \neq \emptyset$
5. $u = \text{Extract-Min}(Q)$ $\leftarrow O(|V| \lg |V|)$
6. $S = S \cup \{u\}$ $\leftarrow O(|V|)$
7. **for** each vertex $v \in G.\text{Adj}[u]$ $\leftarrow O(|E|)$
 RELAX(u, v, w)

Dijkstra's algorithm running time is $O(|V| \lg |V| + |E|)$

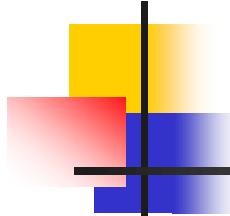




Theorem 24.6 (Correctness of Dijkstra's Algorithm)

- Dijkstra's algorithm, run on a weighted, directed graph $G=(V,E)$ with non-negative weight function w and source s , terminates with $u.d=\delta(s,u)$ for all vertices $u \in V$.





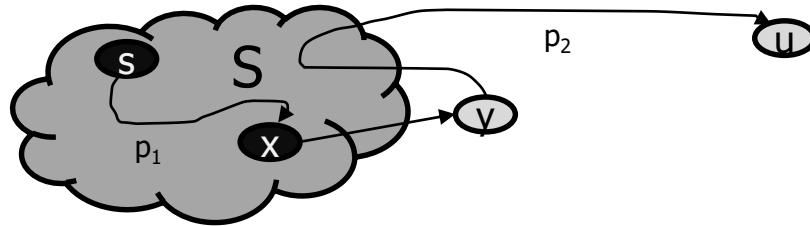
Theorem 24.6 (Proof)

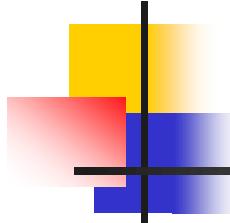
- Loop invariant
- At the start of each iteration of the while loop of lines 4-8, $v.d = \delta(s, v)$ for each vertex v in S
- Initialization: $S = \{\}$, so true
- Maintenance:
 - Let u be the first vertex for which $u.d \neq \delta(s, u)$ when it is added to set S
 - $u \neq s$ because s is the first vertex added to set S and $s.d = \delta(s, s) = 0$
 - Because $u \neq s$, we also have that $S \neq \{\}$ just before u is added to S
 - There must be some path from s to u , for otherwise $u.d = \delta(s, u) = \infty$ by no-path property
 - There is a shortest path p from s to u
 - Prior to adding u to S , path p connects a vertex in S , namely s , to a vertex in $V - S$, namely u .



Theorem 24.6 (Proof)

- Let us consider the first vertex y along p such that $y \in V-S$, and let $x \in S$ be y 's predecessor along p
- Figure shown below illustrates, we can decompose path p into $s \rightsquigarrow x \rightarrow y \rightsquigarrow u$ ($s \rightsquigarrow x$: p_1 , $y \rightsquigarrow u$: p_2)
- Claim: $y.d = \delta(s,y)$ when u is added to S
 - $x.d = \delta(s,x)$ when x was added to S
(\because we chose u as the first vertex for which $u.d \neq \delta(s,u)$)
 - Edge (x,y) was relaxed at that time, and the claim follows from the convergence property

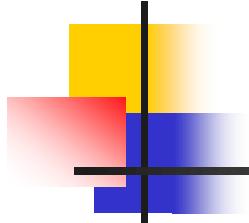




Theorem 24.6 (Proof)

- We can now obtain a contradiction to prove $u.d = \delta(s,u)$
 - $\delta(s,y) \leq \delta(s,u)$ (\because y appears before u on a shortest path from s to u and all edge weights are non-negative)
 - $y.d = \delta(s,y) \leq \delta(s,u) \leq u.d$ (by the upper-bound property)
 - But because both vertices u and y were in $V-S$ when u was chosen in line 5, $u.d \leq y.d$
 - $y.d = \delta(s,y) = \delta(s,u) = u.d$
 - Consequently $u.d = \delta(s,u)$, which contradicts our choice of u .
- $u.d = \delta(s,u)$ when u is added to S , and that this equality is maintained at all times thereafter
- Termination : At termination, $Q = \{\}$ which, along with our earlier invariant that $Q = V-S$, implies that $S = V$. $u.d = \delta(s,u)$ for all vertices $u \in V$





Shortest-Path Algorithms

	Bellman-Ford	Dijkstra
Negative Edge	O	X
Positive Cycle	O	O
Negative Cycle	X	X
Time Complexity	$O(V E)$	Array: $O(V ^2)$ Min-heap: $O((V + E)\lg V)$ Fibonacci heap: $O(V \lg V + E)$

