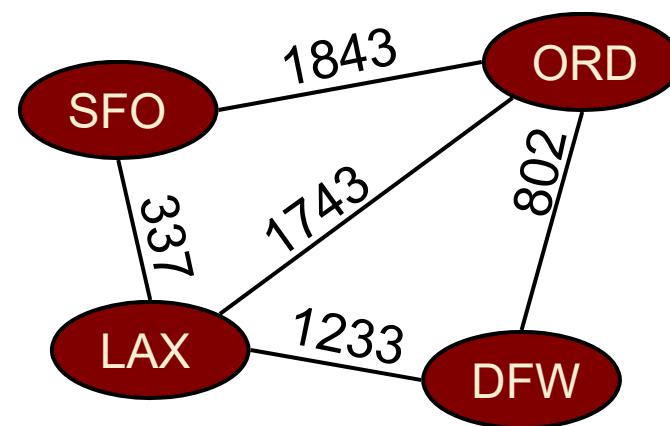


Graphs – Shortest Path (Weighted Graph)



Outline

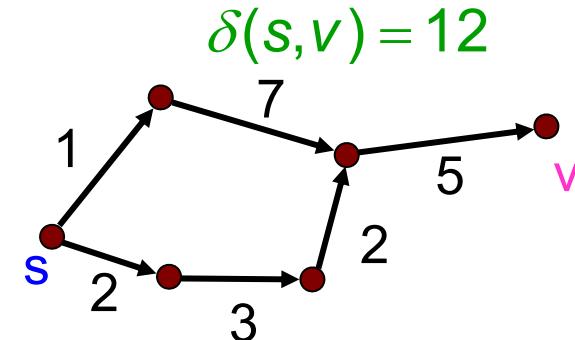
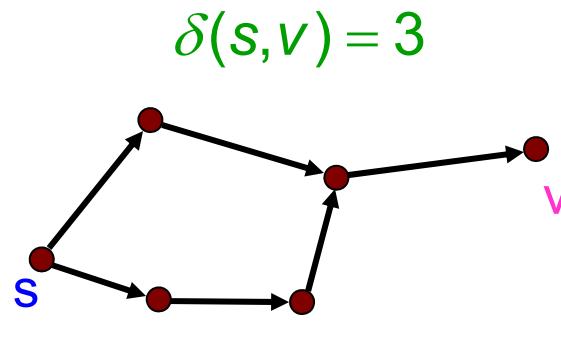
- The shortest path problem
- Single-source shortest path
 - Shortest path on a directed acyclic graph (DAG)
 - Shortest path on a general graph: Dijkstra's algorithm

Outline

- **The shortest path problem**
- Single-source shortest path
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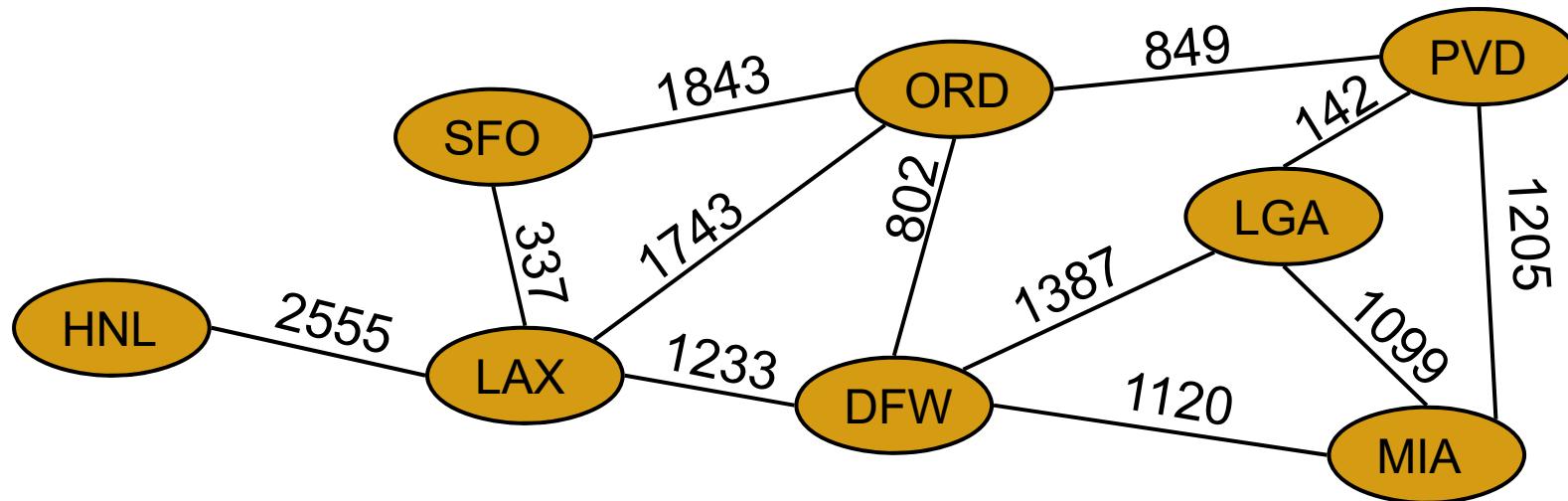
Shortest Path on Weighted Graphs

- BFS finds the **shortest paths** from a source node **s** to every vertex **v** in the graph.
- Here, the **length** of a path is simply the number of edges on the path.
- But what if edges have different ‘costs’?



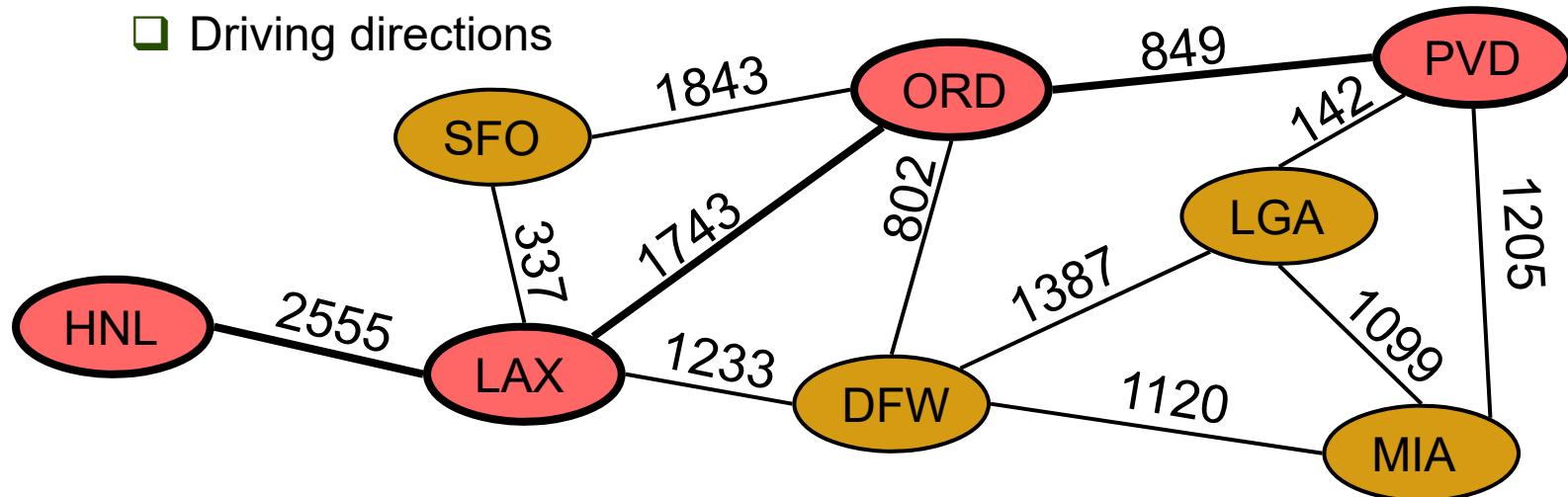
Weighted Graphs

- In a weighted graph, each edge has an associated numerical value, called the weight of the edge
- Edge weights may represent, distances, costs, etc.
- Example:
 - In a flight route graph, the weight of an edge represents the distance in miles between the endpoint airports



Shortest Path on a Weighted Graph

- Given a weighted graph and two vertices u and v , we want to find a path of minimum total weight between u and v .
 - Length of a path is the sum of the weights of its edges.
- Example:
 - Shortest path between Providence and Honolulu
- Applications
 - Internet packet routing
 - Flight reservations
 - Driving directions



Shortest Path: Notation

- Input:

Directed Graph $G = (V, E)$

Edge weights $w : E \rightarrow \mathbb{N}$

Weight of path $p = \langle v_0, v_1, \dots, v_k \rangle = \sum_{i=1}^k w(v_{i-1}, v_i)$

Shortest-path weight from u to v :

$$\delta(u, v) = \begin{cases} \min\{w(p) : u \rightarrow \dots \rightarrow v\} & \text{if } \exists \text{ a path } u \rightarrow \dots \rightarrow v, \\ \infty & \text{otherwise.} \end{cases}$$

Shortest path from u to v is any path p such that $w(p) = \delta(u, v)$.

Shortest Path Properties

Property 1 (Optimal Substructure):

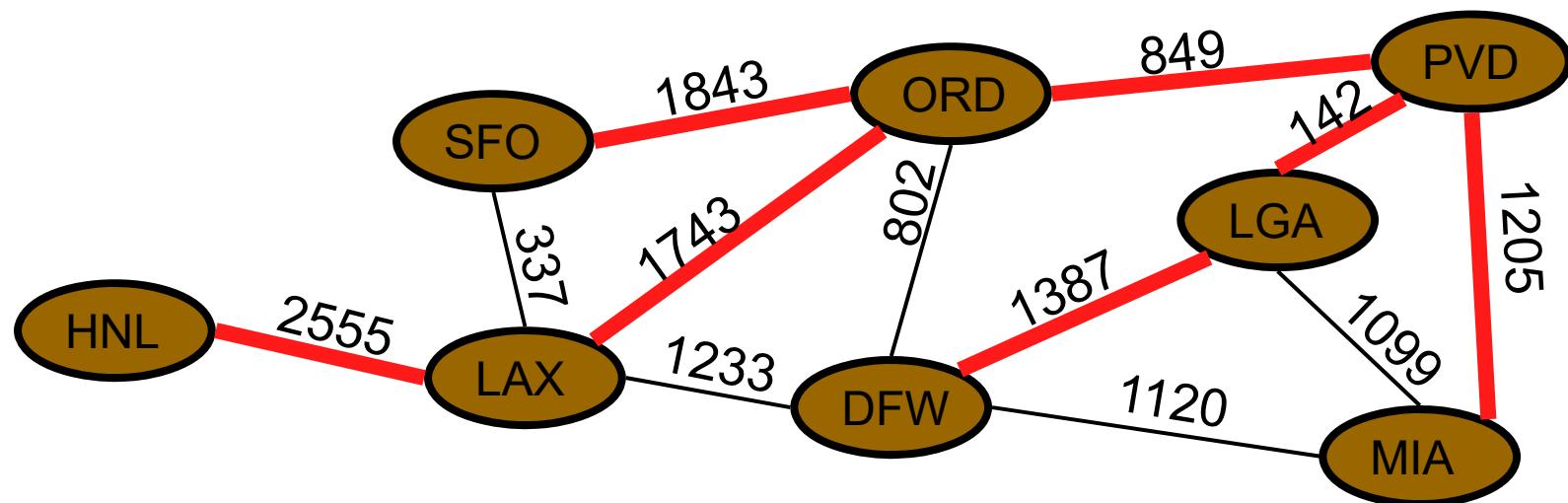
A subpath of a shortest path is itself a shortest path

Property 2 (Shortest Path Tree):

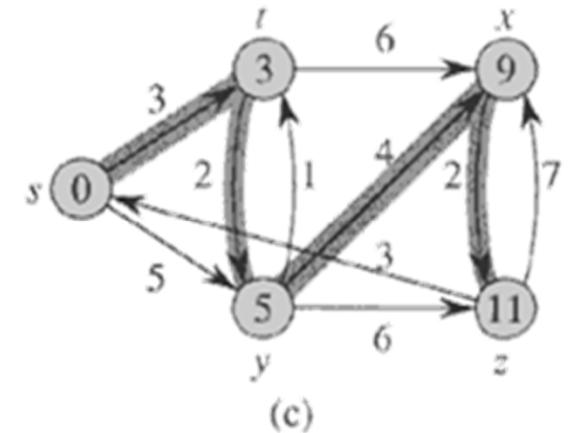
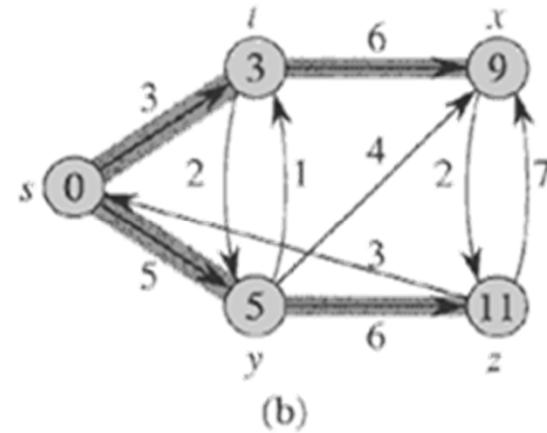
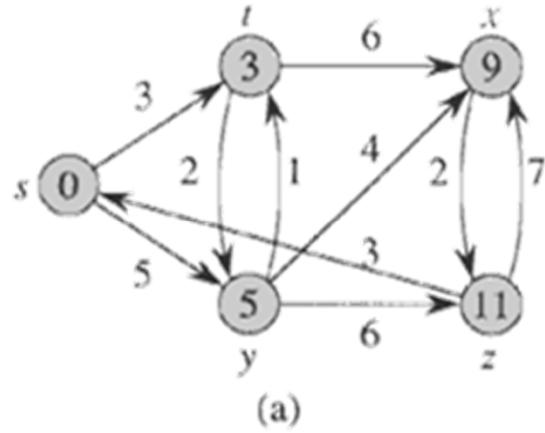
There is a tree of shortest paths from a start vertex to all the other vertices

Example:

Tree of shortest paths from Providence



Shortest path trees are not necessarily unique



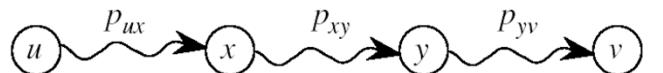
Single-source shortest path search induces a search tree rooted at s .

This tree, and hence the paths themselves, are not necessarily unique.

Optimal substructure: Proof

- Lemma: Any subpath of a shortest path is a shortest path
- Proof: Cut and paste.

Suppose this path p is a shortest path from u to v .



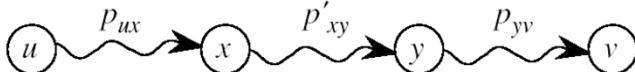
Then $\delta(u, v) = w(p) = w(p_{ux}) + w(p_{xy}) + w(p_{yv})$.

p'_{xy}

Now suppose there exists a shorter path $x \rightarrow \dots \rightarrow y$.

Then $w(p'_{xy}) < w(p_{xy})$.

Construct p' :



Then $w(p') = w(p_{ux}) + w(p'_{xy}) + w(p_{yv}) < w(p_{ux}) + w(p_{xy}) + w(p_{yv}) = w(p)$.

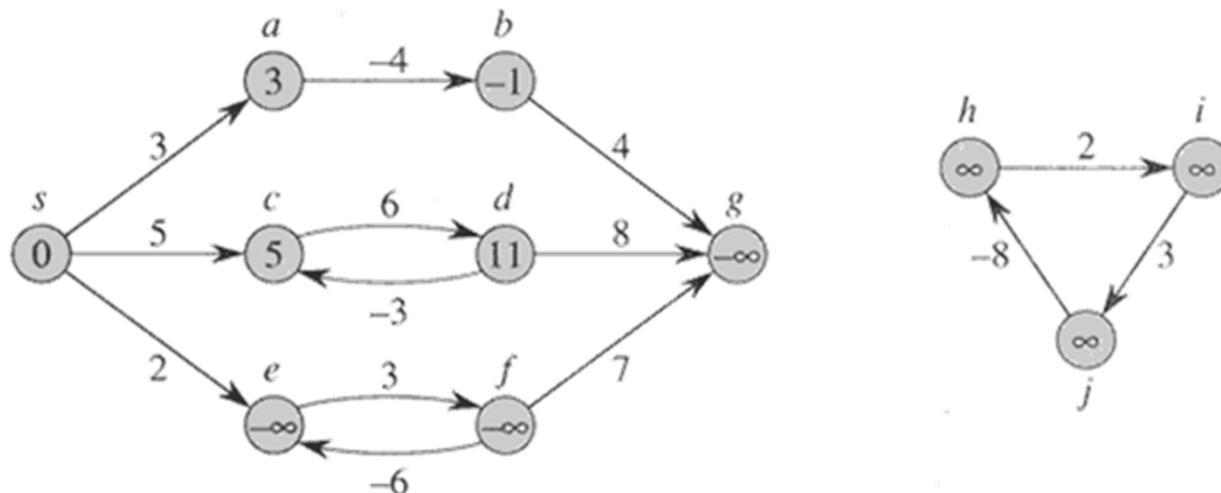
So p wasn't a shortest path after all!

Shortest path variants

- **Single-source shortest-paths problem:** – the shortest path from s to each vertex v .
- **Single-destination shortest-paths problem:** Find a shortest path to a given ***destination*** vertex t from each vertex v .
- **Single-pair shortest-path problem:** Find a shortest path from u to v for given vertices u and v .
- **All-pairs shortest-paths problem:** Find a shortest path from u to v for every pair of vertices u and v .

Negative-weight edges

- OK, as long as no negative-weight cycles are reachable from the source.
 - ❑ If we have a negative-weight cycle, we can just keep going around it, and get $w(s, v) = -\infty$ for all v on the cycle.
 - ❑ But OK if the negative-weight cycle is not reachable from the source.
 - ❑ Some algorithms work only if there are no negative-weight edges in the graph.



Cycles

- Shortest paths can't contain cycles:
 - Already ruled out negative-weight cycles.
 - Positive-weight: we can get a shorter path by omitting the cycle.
 - Zero-weight: no reason to use them → assume that our solutions won't use them.

Outline

- The shortest path problem
- **Single-source shortest path**
 - Shortest path on a directed acyclic graph (DAG)
 - Shortest path on a general graph: Dijkstra's algorithm

Output of a single-source shortest-path algorithm

➤ For each vertex v in V :

□ $d[v] = \delta(s, v)$.

- ❖ Initially, $d[v]=\infty$.
- ❖ Reduce as algorithm progresses.
But always maintain $d[v] \geq \delta(s, v)$.
- ❖ Call $d[v]$ a shortest-path estimate.

□ $\pi[v] = \text{predecessor of } v \text{ on a shortest path from } s$.

- ❖ If no predecessor, $\pi[v] = \text{NIL}$.
- ❖ π induces a tree — **shortest-path tree**.

Initialization

- All shortest-path algorithms start with the same initialization:

INIT-SINGLE-SOURCE(V, s)

for each v in V

do $d[v] \leftarrow \infty$

$\pi[v] \leftarrow \text{NIL}$

$d[s] \leftarrow 0$

Relaxing an edge

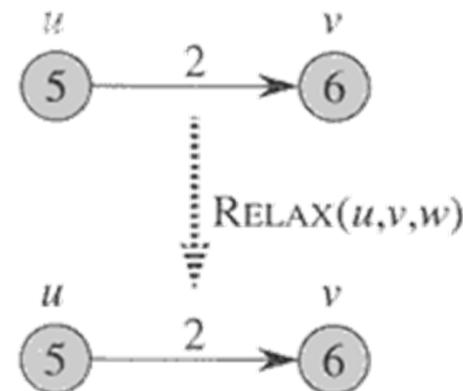
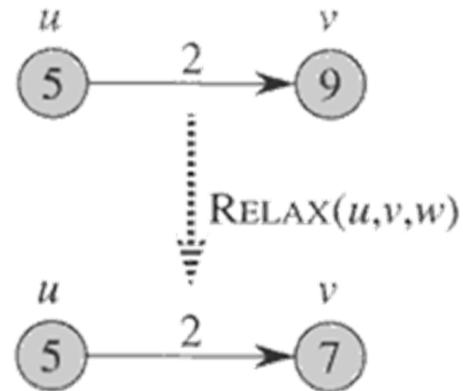
- Can we improve shortest-path estimate for v by first going to u and then following edge (u,v) ?

$\text{RELAX}(u, v, w)$

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

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

$\pi[v] \leftarrow u$



General single-source shortest-path strategy

1. Start by calling INIT-SINGLE-SOURCE
2. Relax Edges

Algorithms differ in the order in which edges are taken and how many times each edge is relaxed.

Outline

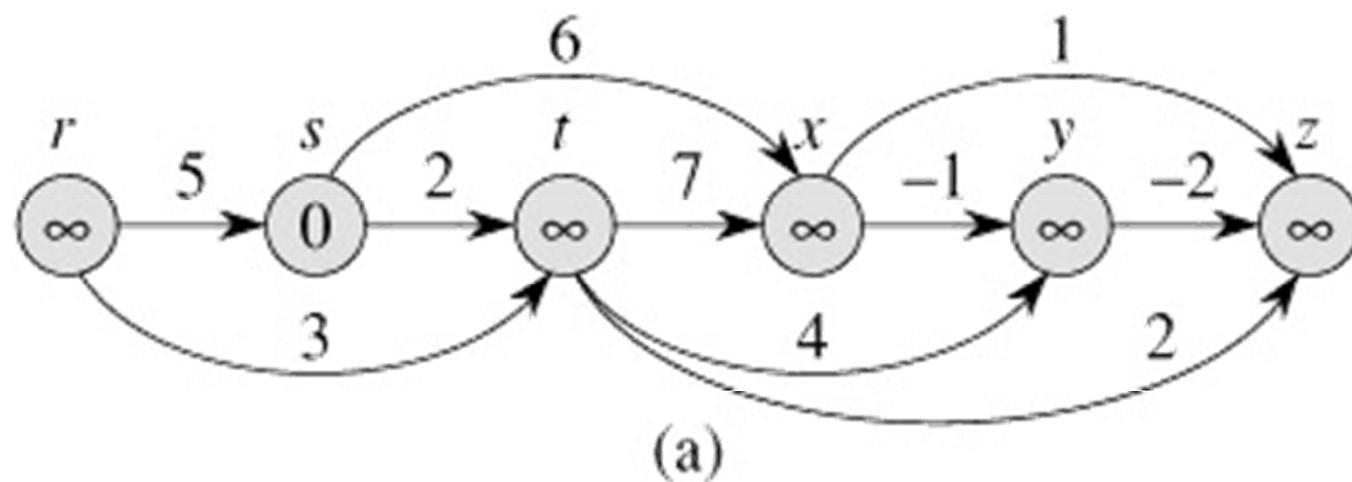
- The shortest path problem
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Example 1. Single-Source Shortest Path on a Directed Acyclic Graph

- Basic Idea: topologically sort nodes and relax in linear order.
- Efficient, since $\delta[u]$ (shortest distance to u) has already been computed when edge (u,v) is relaxed.
- Thus we only relax each edge once, and never have to backtrack.

Example: Single-source shortest paths in a directed acyclic graph (DAG)

- Since graph is a DAG, we are guaranteed no negative-weight cycles.
- Thus algorithm can handle negative edges



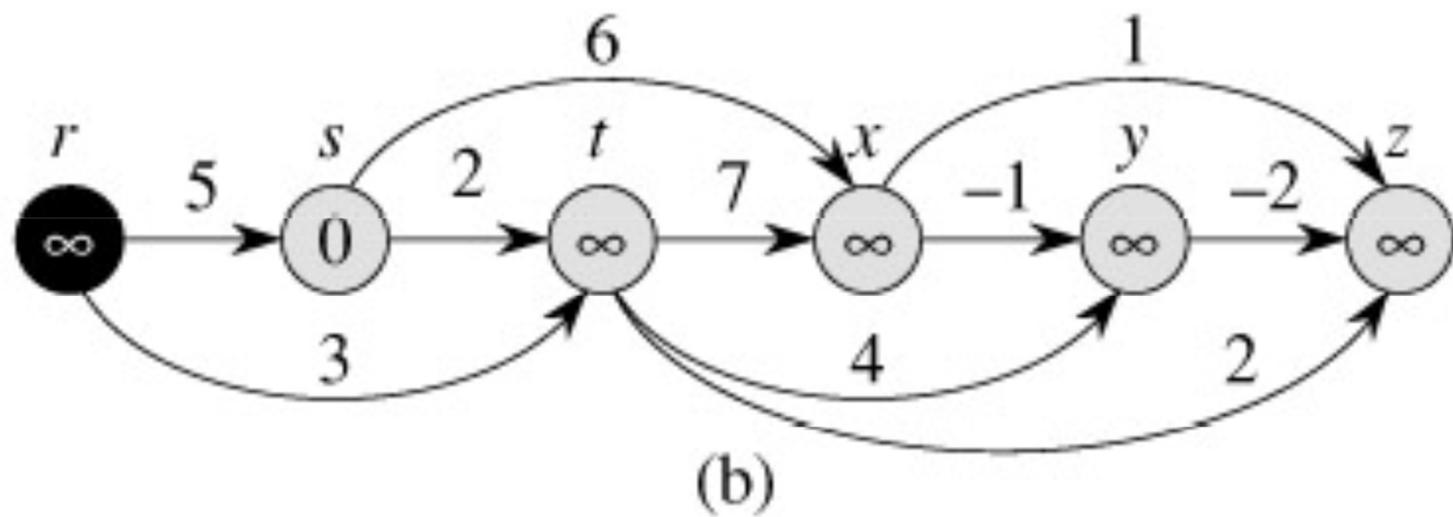
Algorithm

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

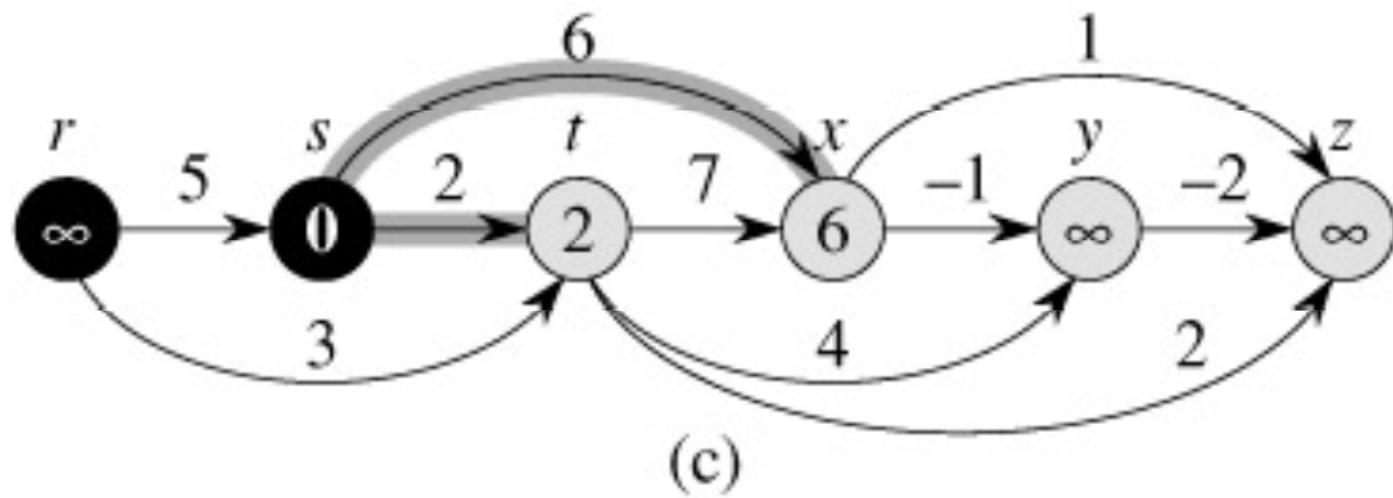
- 1 topologically sort the vertices of G
- 2 INITIALIZE-SINGLE-SOURCE(G, s)
- 3 **for** each vertex u , taken in topologically sorted order
- 4 **do for** each vertex $v \in Adj[u]$
- 5 **do** RELAX(u, v, w)

Time: $\Theta(V + E)$

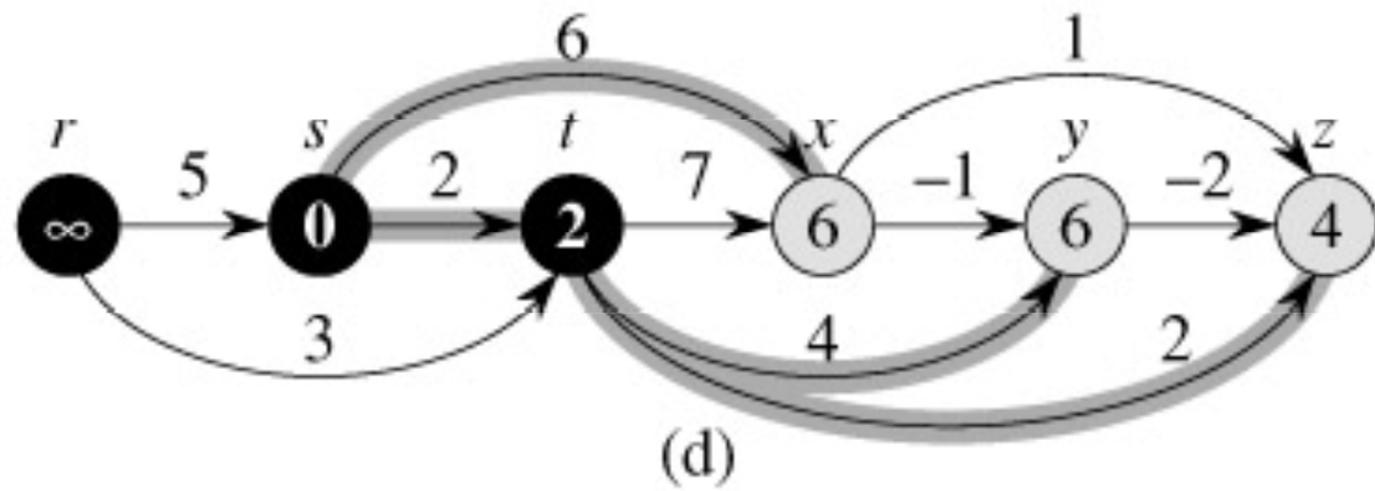
Example



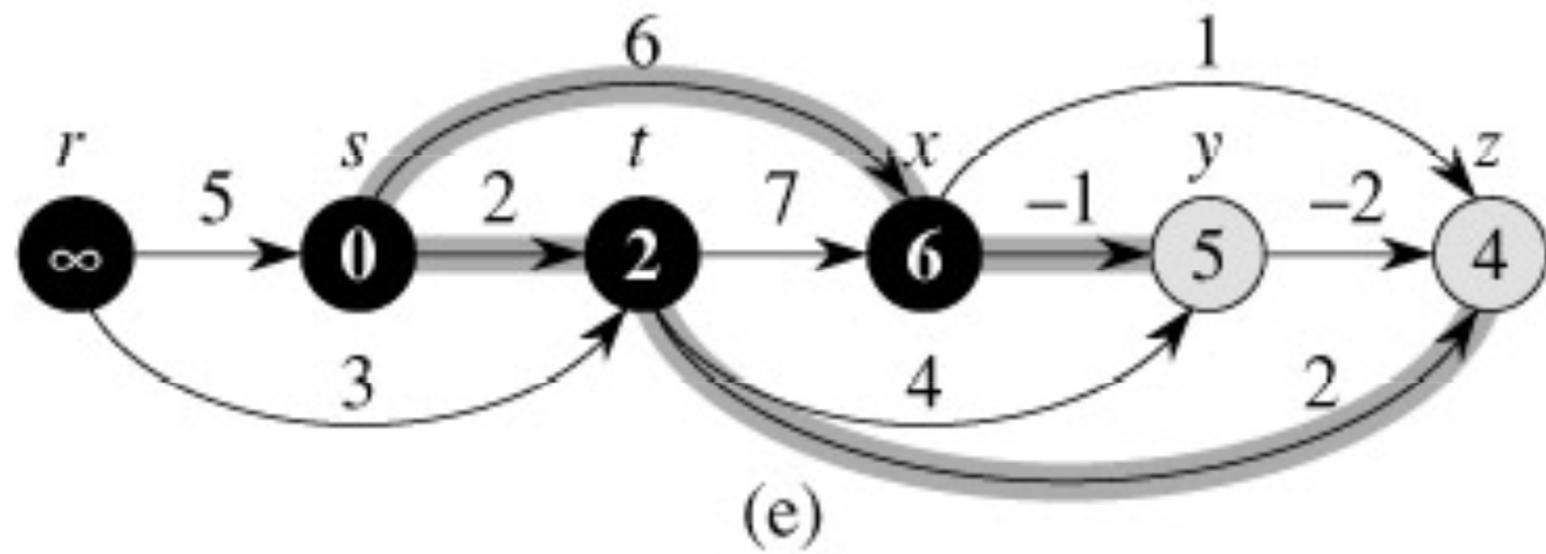
Example



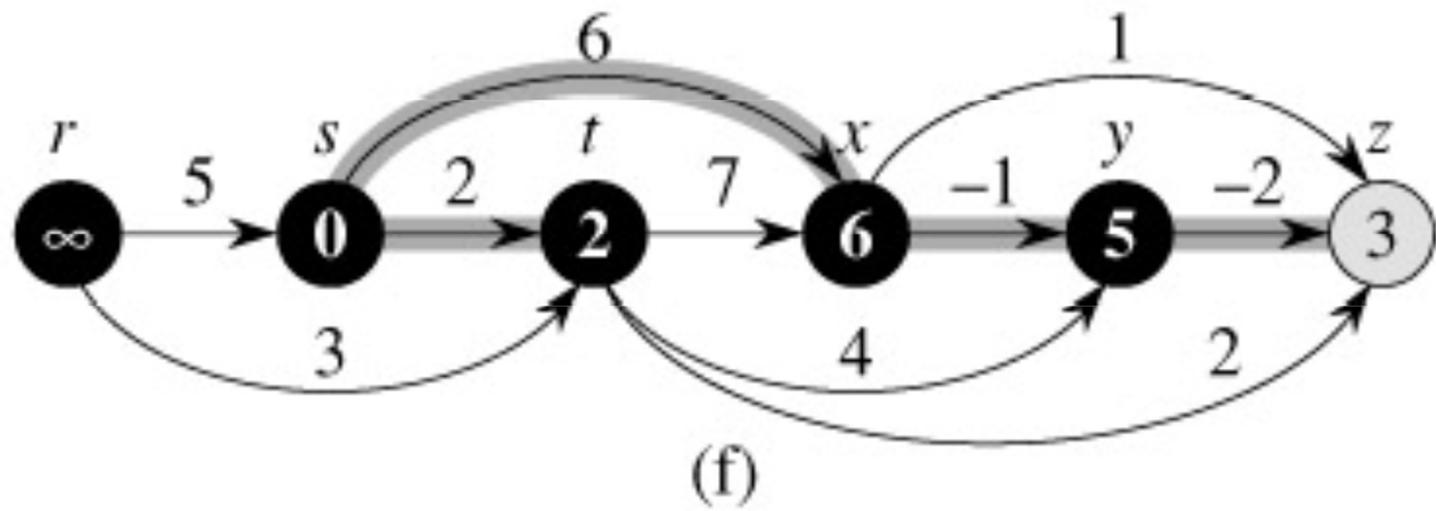
Example



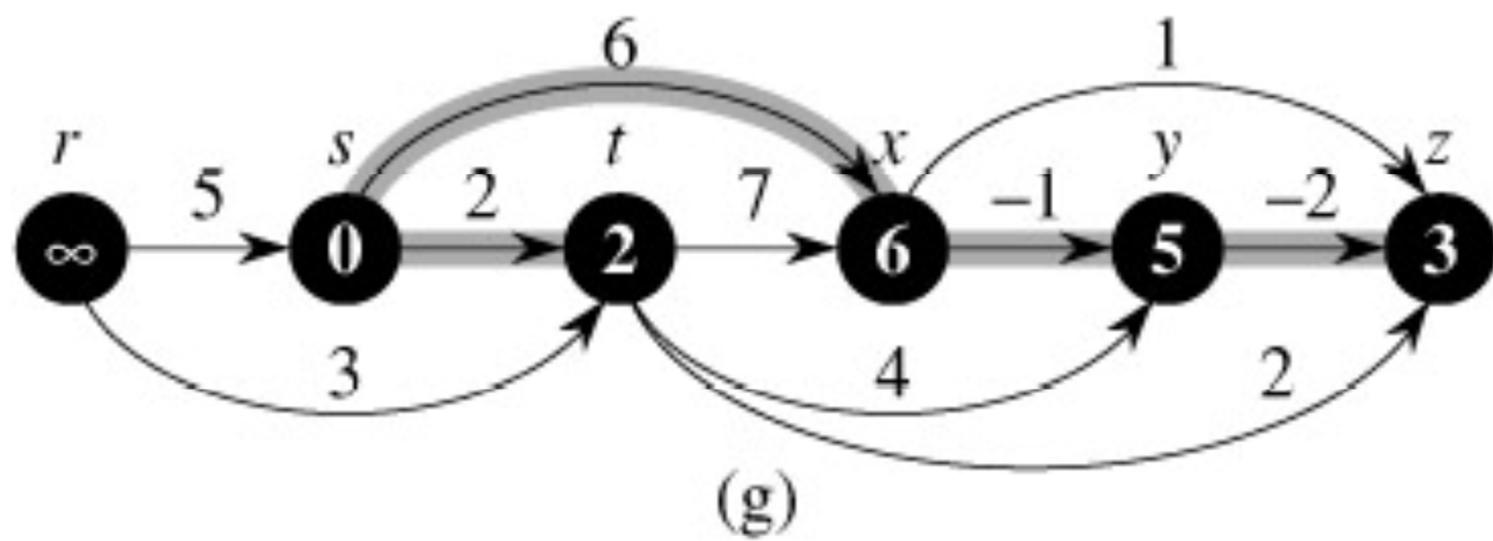
Example



Example



Example



Correctness: 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) , \dots , (v_{k-1}, v_k) ,

even intermixed with other relaxations,

then $d[v_k] = \delta(s, v_k)$.

Correctness of DAG Shortest Path Algorithm

- Because we process vertices in topologically sorted order, edges of *any* path are relaxed in order of appearance in the path.
 - → Edges on any shortest path are relaxed in order.
 - → By path-relaxation property, correct.

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Example 2. Single-Source Shortest Path on a General Graph (May Contain Cycles)

- This is fundamentally harder, because the first paths we discover may not be the shortest (not monotonic).

Dijkstra's algorithm (E. Dijkstra, 1959)

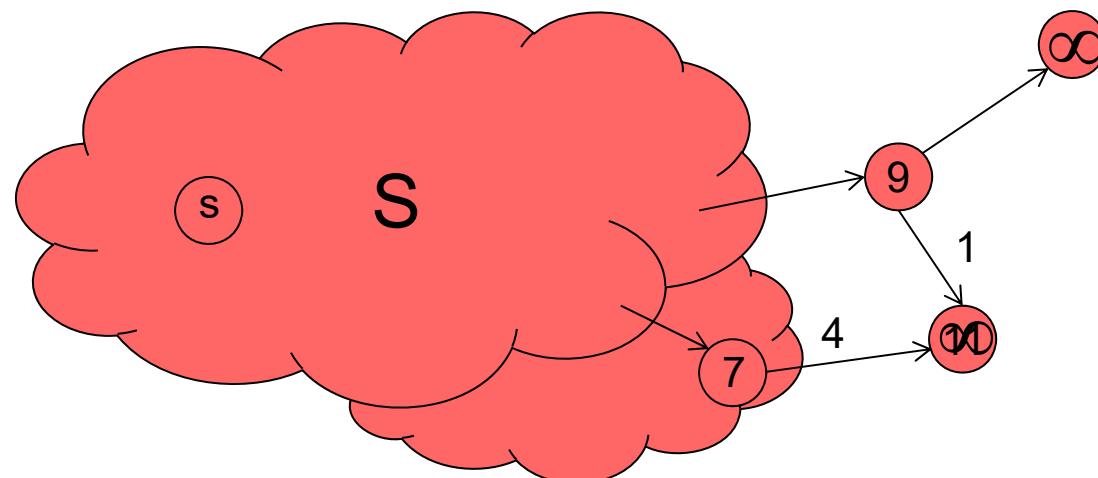
- Applies to general, weighted, directed or undirected graph (may contain cycles).
- But weights must be non-negative. (But they can be 0!)
- Essentially a weighted version of BFS.
 - Instead of a FIFO queue, uses a priority queue.
 - Keys are shortest-path weights ($d[v]$).
- Maintain 2 sets of vertices:
 - S = vertices whose final shortest-path weights are determined.
 - Q = priority queue = $V-S$.



Edsger Dijkstra

Dijkstra's Algorithm: Operation

- We grow a “**cloud**” S of vertices, beginning with s and eventually covering all the vertices
- We store with each vertex v a label $d(v)$ representing the distance of v from s in the subgraph consisting of the cloud S and its adjacent vertices
- At each step
 - We add to the cloud S the vertex u outside the cloud with the smallest distance label, $d(u)$
 - We update the labels of the vertices adjacent to u



Dijkstra's algorithm

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

- Dijkstra's algorithm can be viewed as greedy, since it always chooses the "lightest" vertex in $V - S$ to add to S .

Dijkstra's algorithm: Analysis

- Analysis:
 - Using minheap, queue operations takes $O(\log V)$ time

DIJKSTRA(G, w, s)

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )  $O(V)$ 
2   $S \leftarrow \emptyset$ 
3   $Q \leftarrow V[G]$ 
4  while  $Q \neq \emptyset$ 
5    do  $u \leftarrow \text{EXTRACT-MIN}(Q)$        $O(\log V) \times O(V)$  iterations
6     $S \leftarrow S \cup \{u\}$ 
7    for each vertex  $v \in \text{Adj}[u]$ 
8      do RELAX( $u, v, w$ )       $O(\log V) \times O(E)$  iterations
```

→ Running Time is $O(E \log V)$

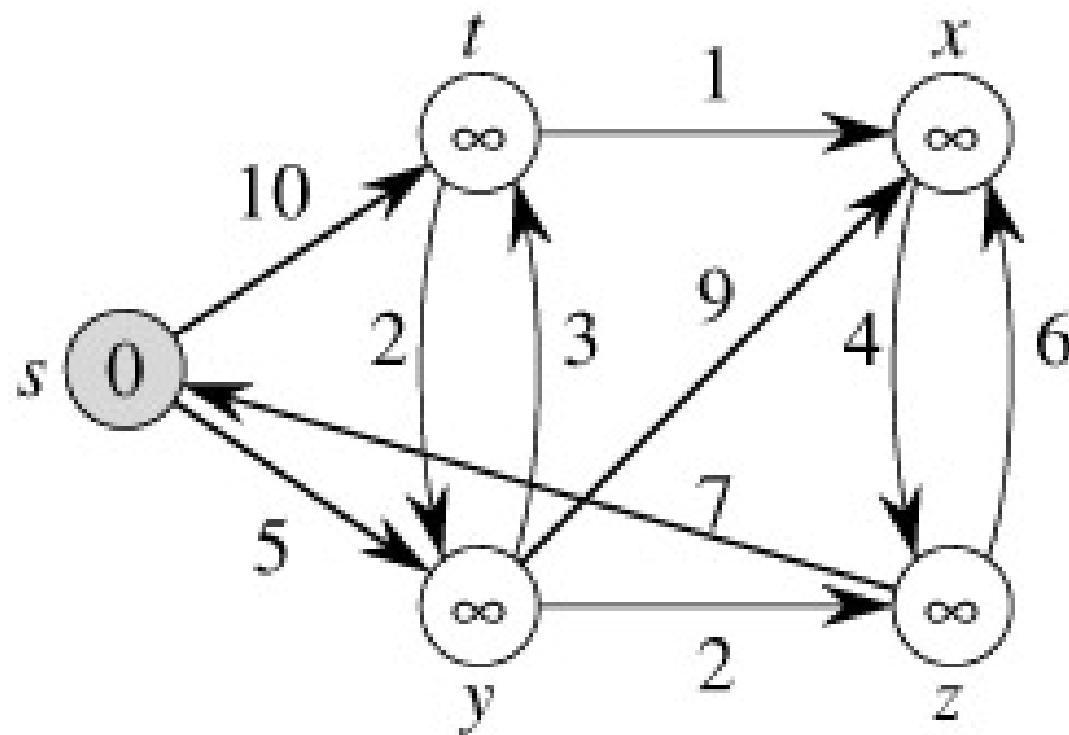
Example

Key:

White \Leftrightarrow Vertex $\in Q = V - S$

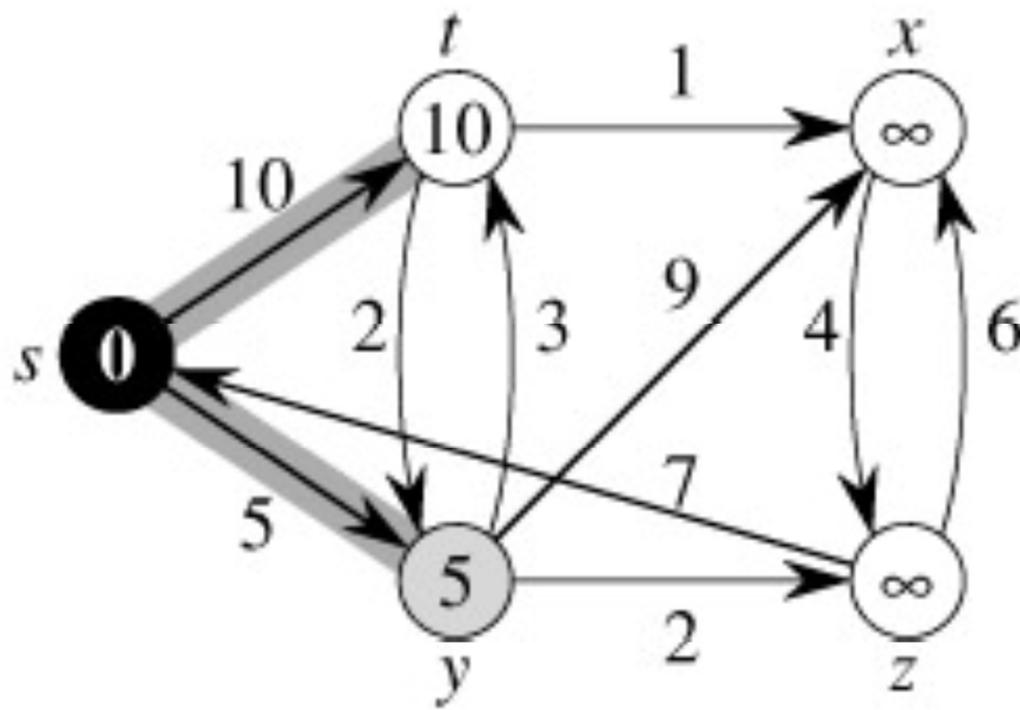
Grey \Leftrightarrow Vertex $= \min(Q)$

Black \Leftrightarrow Vertex $\in S$, Off Queue



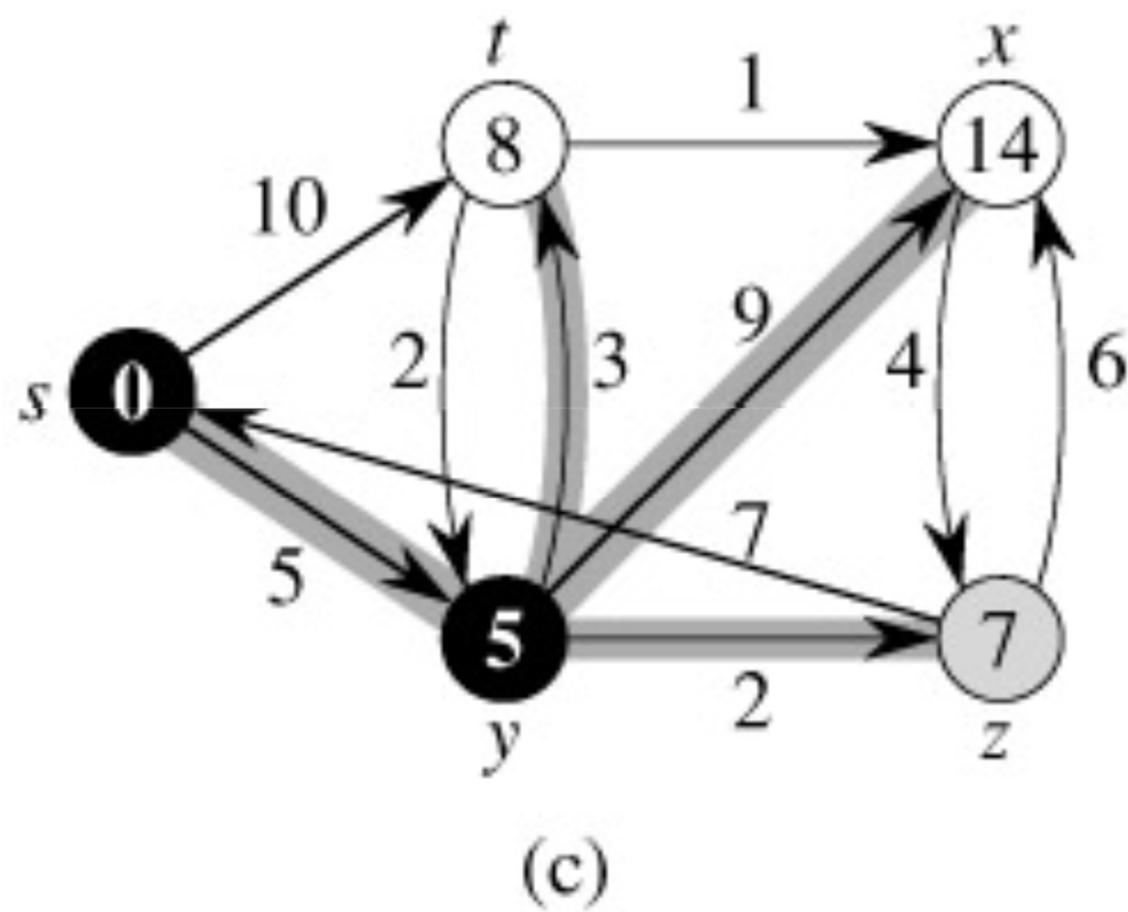
(a)

Example

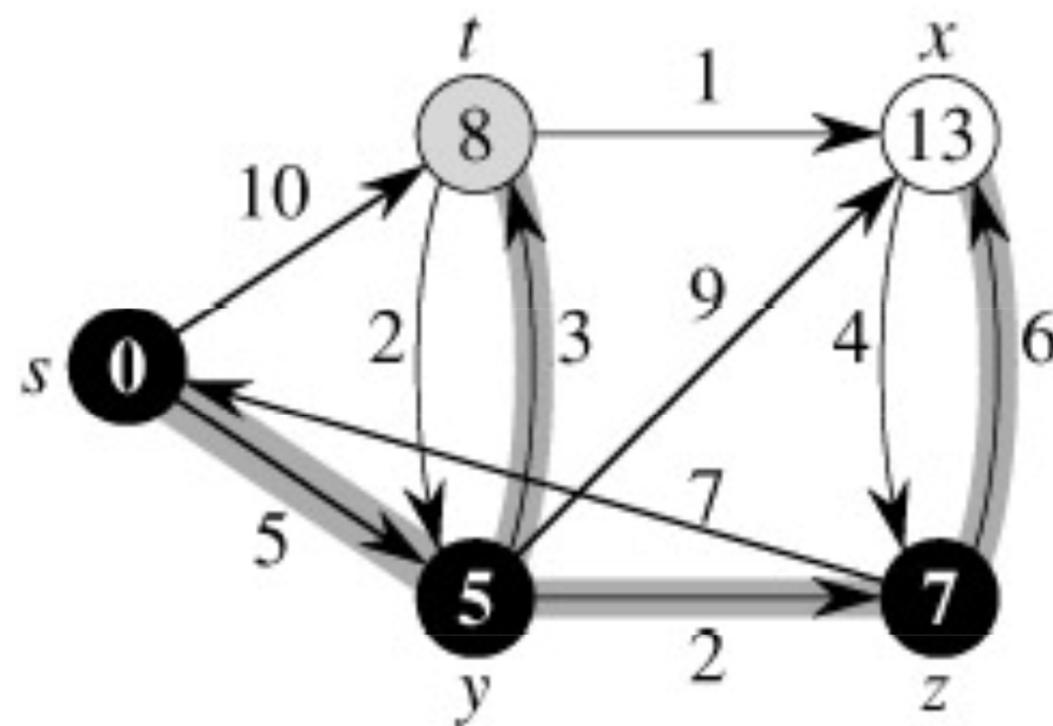


(b)

Example

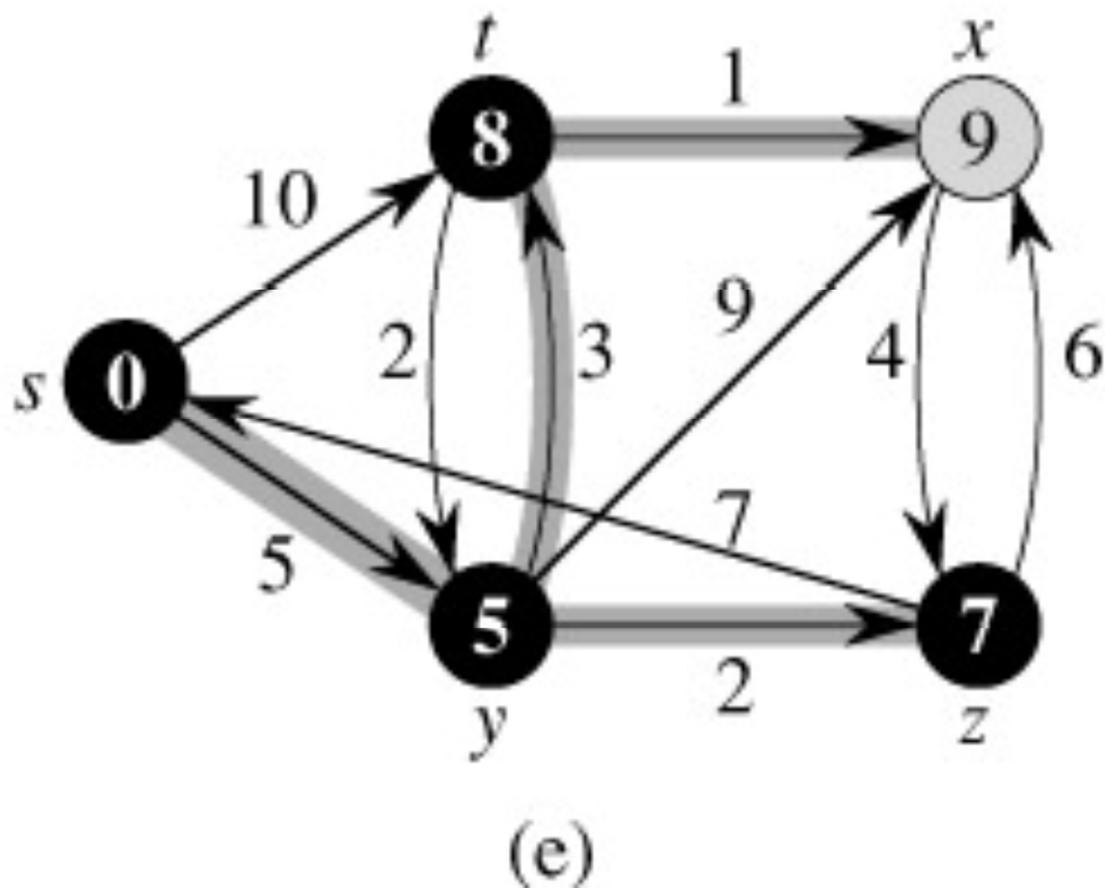


Example

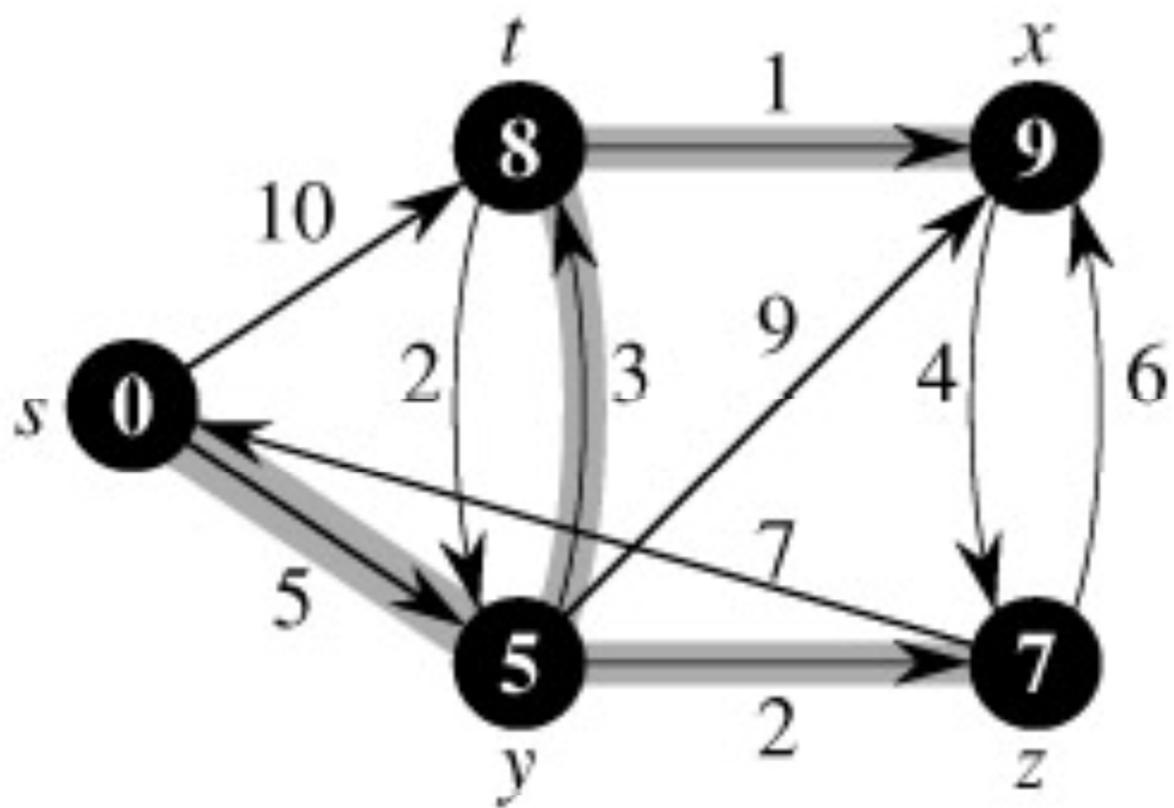


(d)

Example

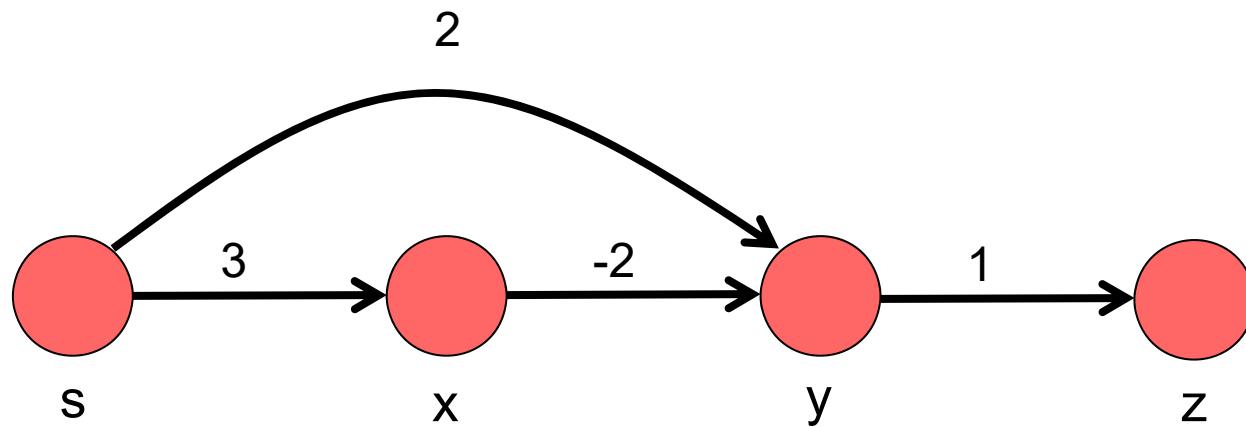


Example



(f)

Dijkstra's Algorithm Cannot Handle Negative Edges



Correctness of Dijkstra's algorithm

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

➤ **Loop invariant:** $d[v] = \delta(s, v)$ for all v in S .

- **Initialization:** Initially, S is empty, so trivially true.
- **Termination:** At end, Q is empty $\rightarrow S = V \rightarrow d[v] = \delta(s, v)$ for all v in V .
- **Maintenance:**
 - ◊ Need to show that
 - ❖ $d[u] = \delta(s, u)$ when u is added to S in each iteration.
 - ❖ $d[u]$ does not change once u is added to S .

Correctness of Dijkstra's Algorithm: Upper Bound Property

- Upper Bound Property:

1. $d[v] \geq \delta(s, v) \forall v \in V$
2. Once $d[v] = \delta(s, v)$, it doesn't change

- Proof:

By induction.

Base Case: $d[v] \geq \delta(s, v) \forall v \in V$ immediately after initialization, since

$$d[s] = 0 = \delta(s, s)$$

$$d[v] = \infty \forall v \neq s$$

Inductive Step:

Suppose $d[x] \geq \delta(s, x) \forall x \in V$

Suppose we relax edge (u, v) .

If $d[v]$ changes, then $d[v] = d[u] + w(u, v)$

$$\geq \delta(s, u) + w(u, v) \leftarrow$$

$$\geq \delta(s, v)$$

A valid path from s to v !

Correctness of Dijkstra's Algorithm

Claim: When u is added to S , $d[u] = \delta(s, u)$

Proof by Contradiction: Let u be the first vertex added to S such that $d[u] \neq \delta(s, u)$ when u is added.

Let y be first vertex in $V - S$ on shortest path to u

Let x be the predecessor of y on the shortest path to u

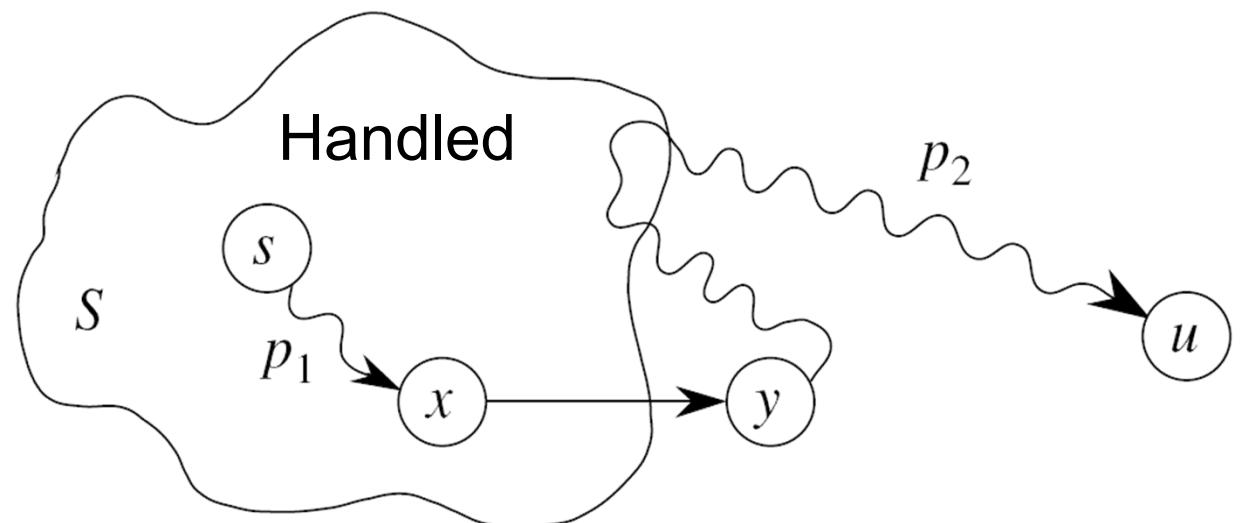
Claim: $d[y] = \delta(s, y)$ when u is added to S .

Proof:

$d[x] = \delta(s, x)$, since $x \in S$.

(x, y) was relaxed when x was added to $S \rightarrow d[y] = \delta(s, x) + w(x, y) = \delta(s, y)$

Optimal substructure
property!



Correctness of Dijkstra's Algorithm

Thus $d[y] = \delta(s, y)$ when u is added to S .

$\rightarrow d[y] = \delta(s, y) \leq \delta(s, u) \leq d[u]$ (upper bound property)

But $d[u] \leq d[y]$ when u added to S

Thus $d[y] = \delta(s, y) = \delta(s, u) = d[u]!$

Thus when u is added to S , $d[u] = \delta(s, u)$

DIJKSTRA(G, w, s)

```

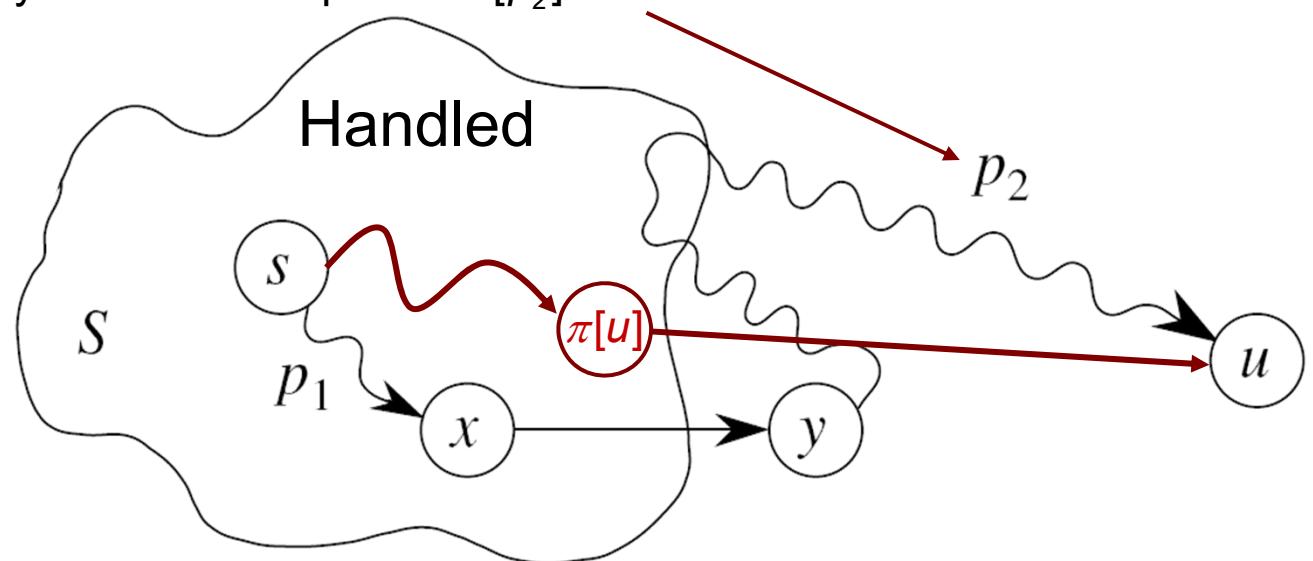
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6       $S \leftarrow S \cup \{u\}$ 
7      for each vertex  $v \in \text{Adj}[u]$ 
8        do RELAX( $u, v, w$ )

```

Consequences:

There is a shortest path to u such that the predecessor of u $\pi[u] \in S$ when u is added to S .

The path through y can only be a shortest path if $w[p_2] = 0$.



Correctness of Dijkstra's algorithm

DIJKSTRA(G, w, s)

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
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7      for each vertex  $v \in \text{Adj}[u]$ 
8        do RELAX( $u, v, w$ )
```

Relax(u, v, w) can only decrease $d[v]$.

By the **upper bound property**, $d[v] \geq \delta(s, v)$.

Thus once $d[v] = \delta(s, v)$, it will not be changed.

➤ **Loop invariant:** $d[v] = \delta(s, v)$ for all v in S .

□ **Maintenance:**

❖ Need to show that

❖ $d[u] = \delta(s, u)$ when u is added to S in each iteration. ✓

❖ $d[u]$ does not change once u is added to S . ?

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