

# COMP251: Single source shortest paths

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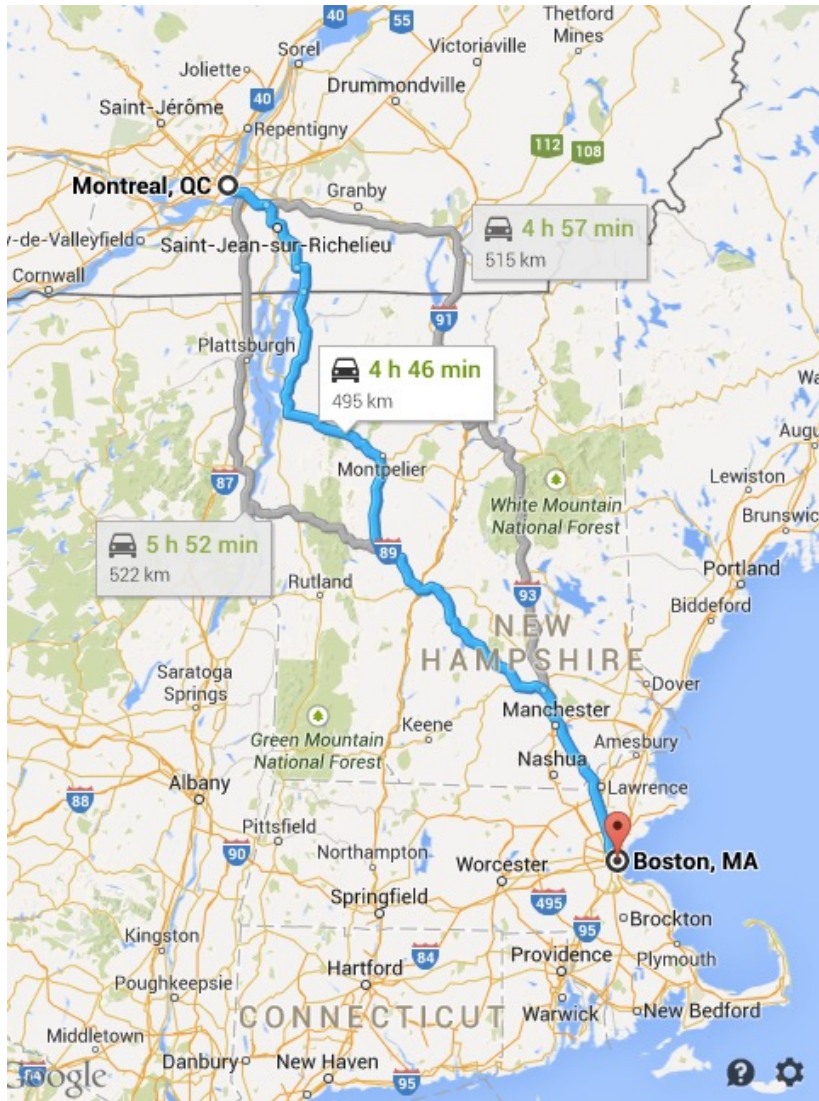
McGill University

Based on (Cormen *et al.*, 2002)

# Outline

- Introduction
- Optimal Substructure
- Definitions & properties
  - Edge relaxation
  - Triangle inequality
  - Upper bound property
  - No-path property
  - Convergence property
  - Path relaxation property
- Single source shortest path in DAGs
- Dijkstra's Algorithm

# Problem



What is the shortest road to go from one city to another?

Example: Which road should I take to go from Montréal to Boston (MA)?

Variants:

- What is the fastest road?
- What is the cheapest road?

# Modeling as graphs

## Input:

- Directed graph  $G = (V, E)$
- Weight function  $w: E \rightarrow \mathbb{R}$

**Weight of path**  $p = \langle v_0, v_1, \dots, v_n \rangle$

$$w(p) = \sum_{k=1}^n w(v_{k-1}, v_k)$$

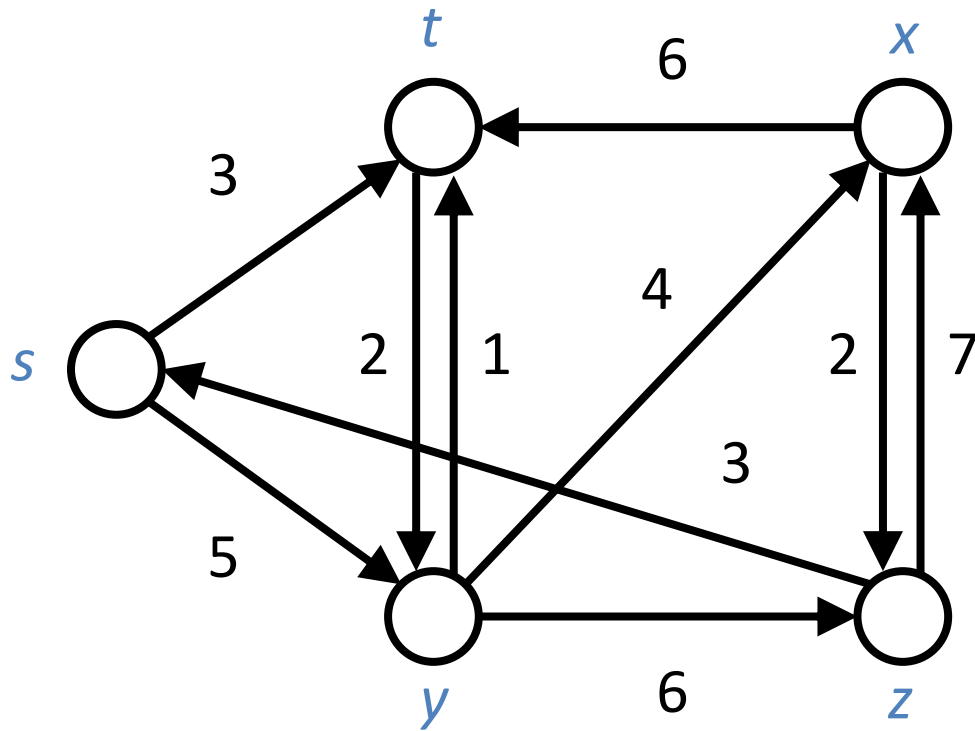
= sum of edges weights on path  $p$

**Shortest-path weight  $u$  to  $v$ :**

$$\delta(u, v) = \begin{cases} \min \left\{ w(p) : u \xrightarrow{p} v \right\} & \text{If there exists a path } u \rightsquigarrow v. \\ \infty & \text{Otherwise.} \end{cases}$$

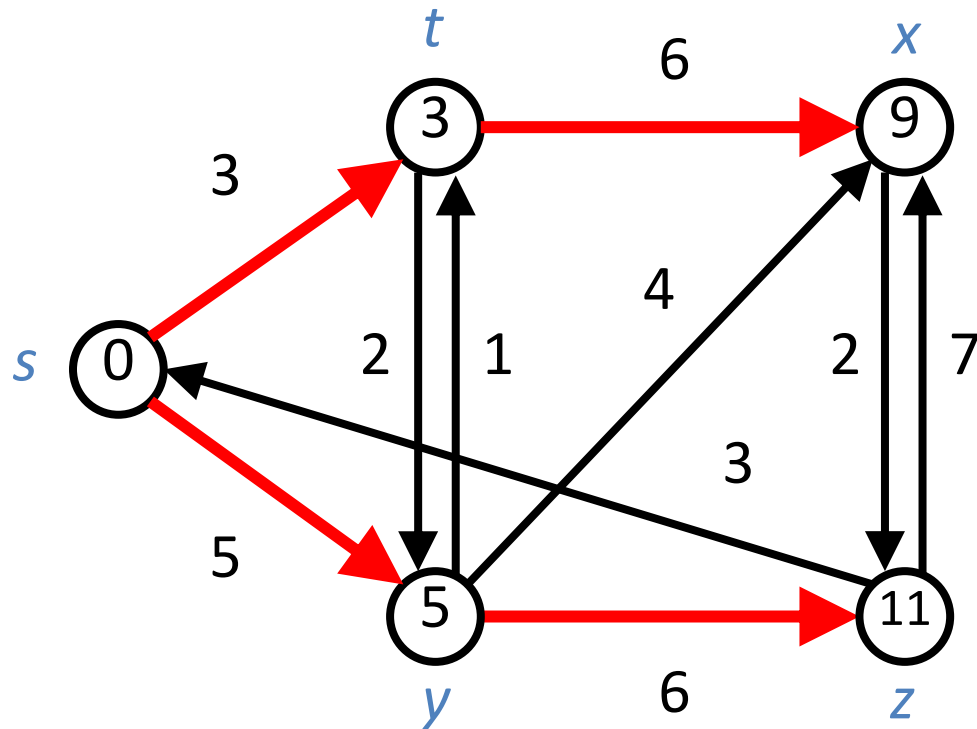
Shortest path  $u$  to  $v$  is any path  $p$  such that  $w(p) = \delta(u, v)$   
Generalization of breadth-first search to weighted graphs

# Example



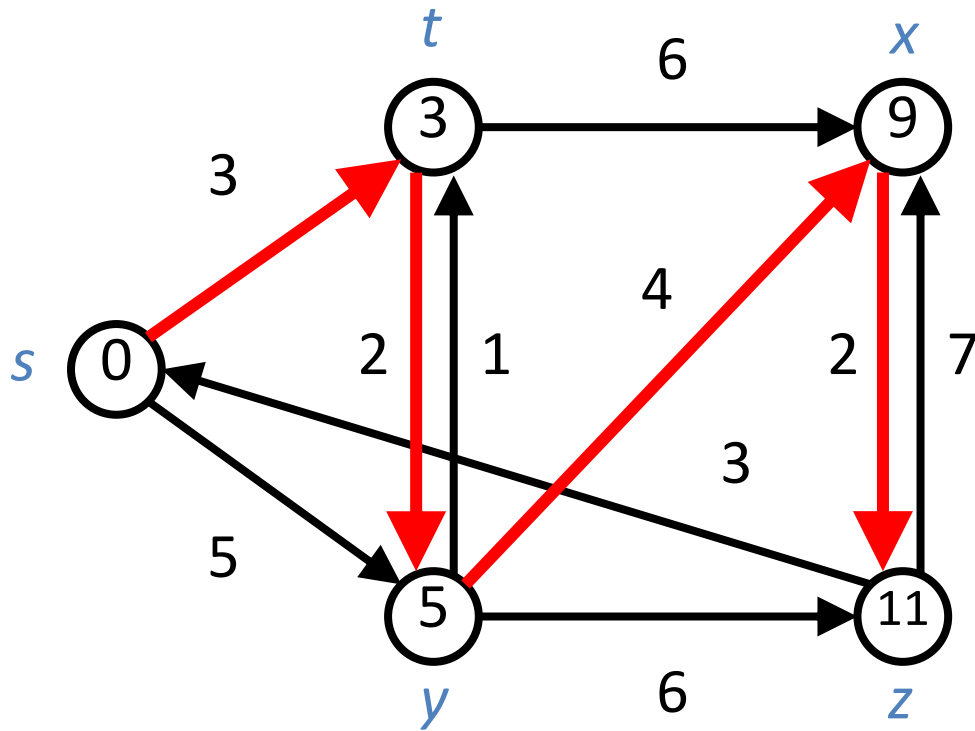
Shortest path from  $s$ ?

# Example



Shortest paths are organized as a tree.  
Vertices store the length of the shortest path from s.

# Example



Shortest paths are not necessarily unique!

# Variants

- **Single-source:** Find shortest paths from a given source vertex  $s \in V$  to every vertex  $v \in V$ .
- **Single-destination:** Find shortest paths to a given destination vertex.
- **Single-pair:** Find shortest path from  $u$  to  $v$ .

*Note: No way to know that is better in worst case than solving the single-source problem!*

- **All-pairs:** Find shortest path from  $u$  to  $v$  for all  $u, v \in V$ .



# Negative weight edges

Negative weight edges can create issues.

**Why?** 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.

**When?** If they are reachable from the source (Corollary: It is OK to have a negative-weight cycles if it is not reachable from the source).

**What?** Some algorithms work only if there are no negative-weight edges in the graph. We must specify when they are allowed and not.

# Cycles

Shortest paths cannot contain cycles:

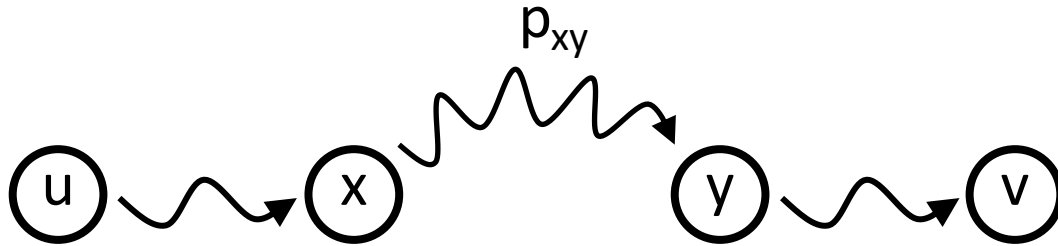
- Negative-weight: Already ruled out.
- Positive-weight: we can get a shorter path by omitting the cycle.
- Zero-weight: no reason to use them  $\Rightarrow$  assume that our solutions will not use them.

# Optimal substructure

***Lemma***

Any subpath of a shortest path is a shortest path.

**Proof** (contradiction using cut and paste approach):



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

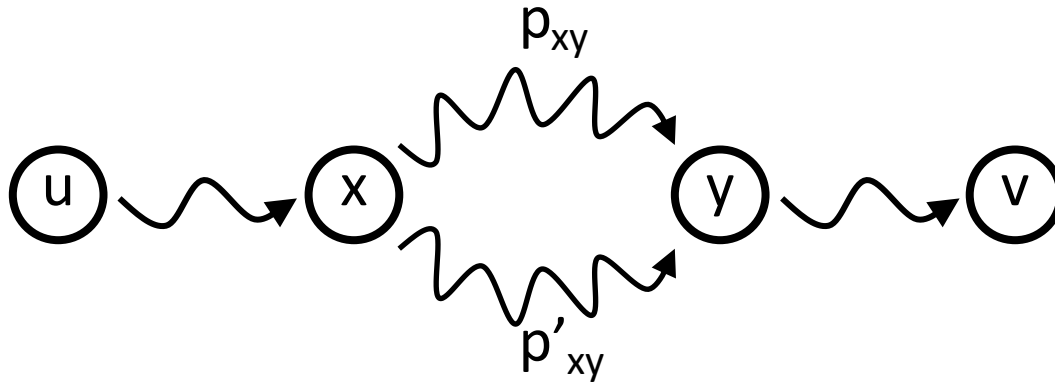
$$\text{Then } \delta(u, v) = w(p) = w(p_{ux}) + w(p_{xy}) + w(p_{yv}).$$

# Optimal substructure

## ***Lemma***

Any subpath of a shortest path is a shortest path.

**Proof:** (cont'd)



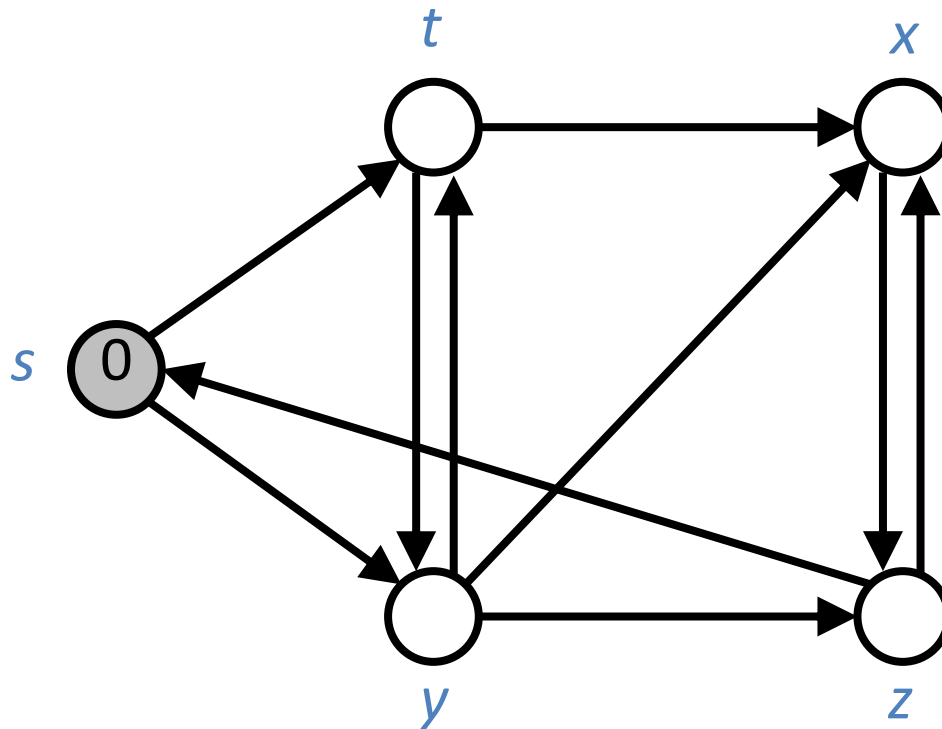
Now suppose there exists a shorter path  $x \overset{p'_{xy}}{\rightsquigarrow} y : w(p'_{xy}) < w(p_{xy})$ .

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

$$< w(p_{ux}) + w(p_{xy}) + w(p_{yv}) = w(p)$$

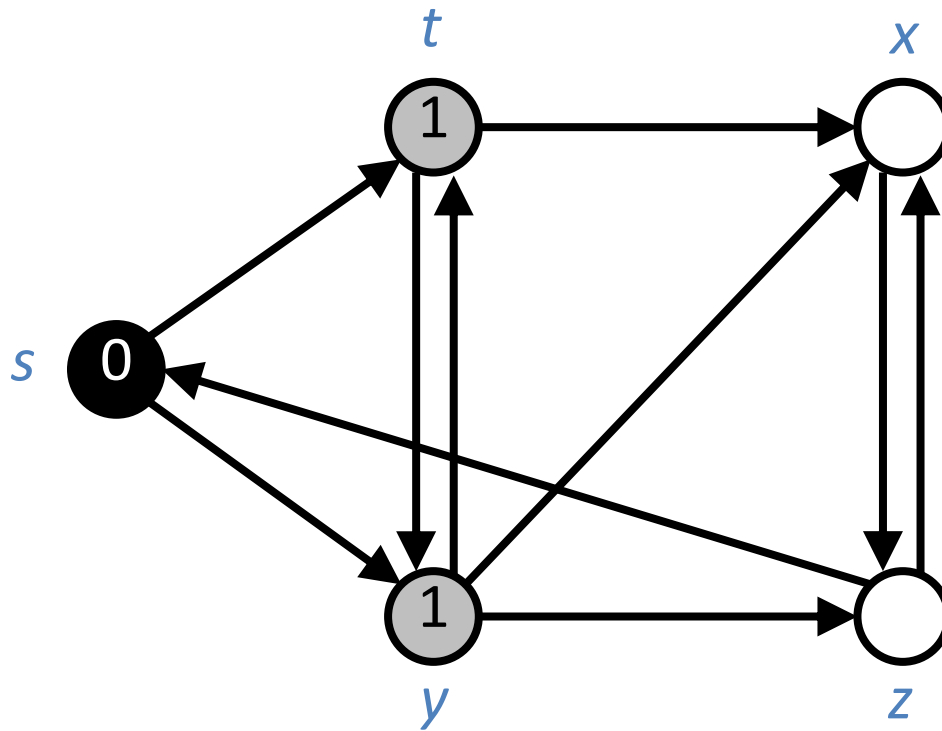
*Contradiction of the hypothesis that  $p$  is the shortest path!*

# Customized breadth-first search

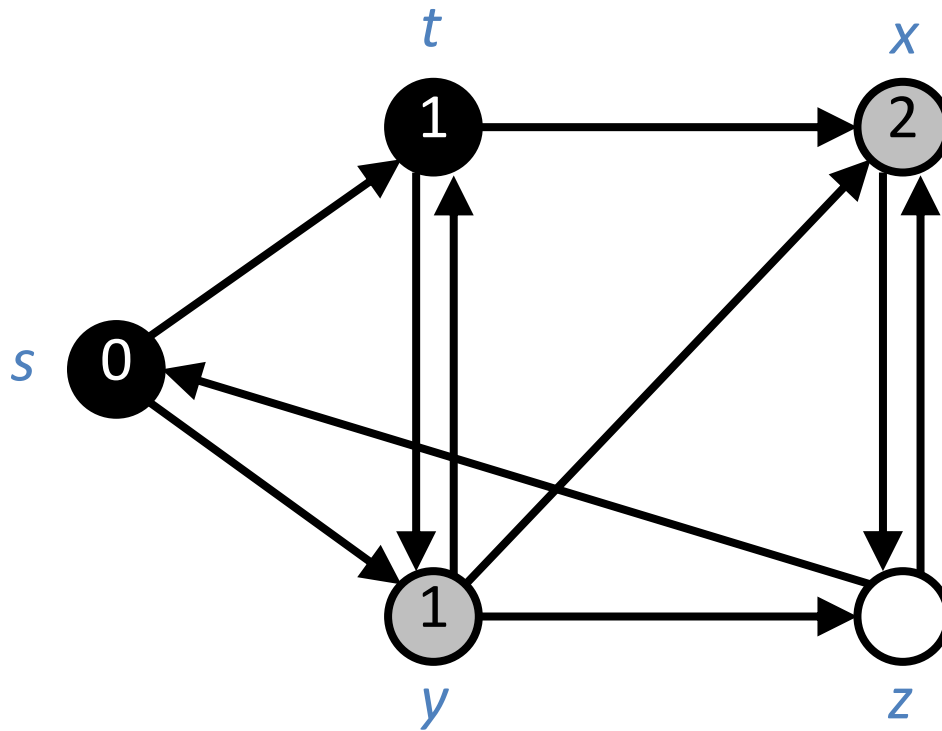


Vertices count the number of edges used to reach them.

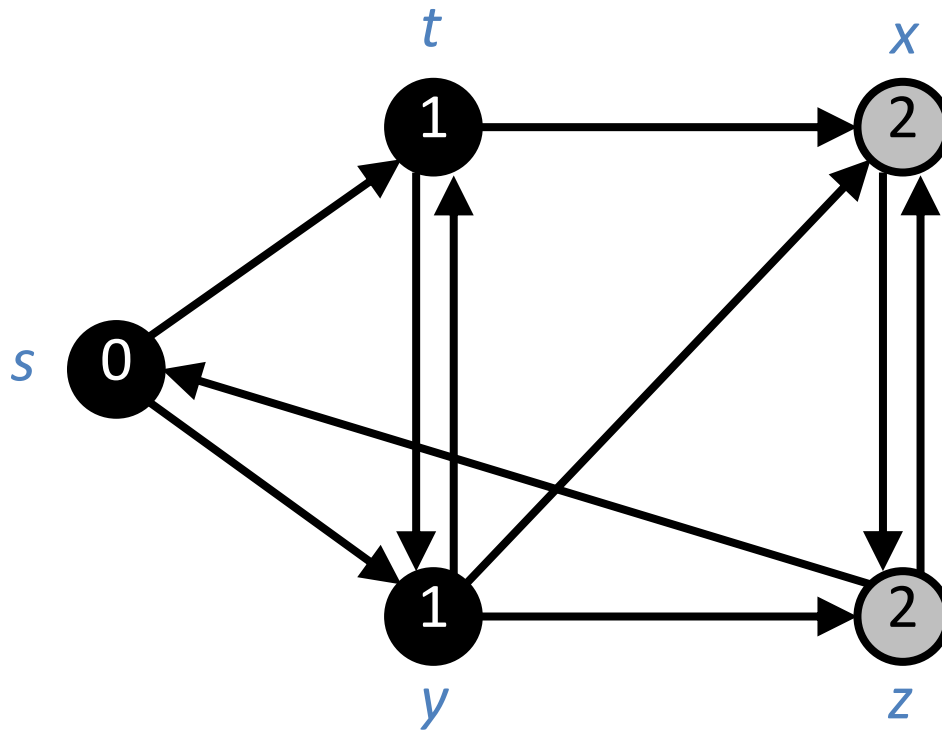
# Customized breadth-first search



# Customized breadth-first search

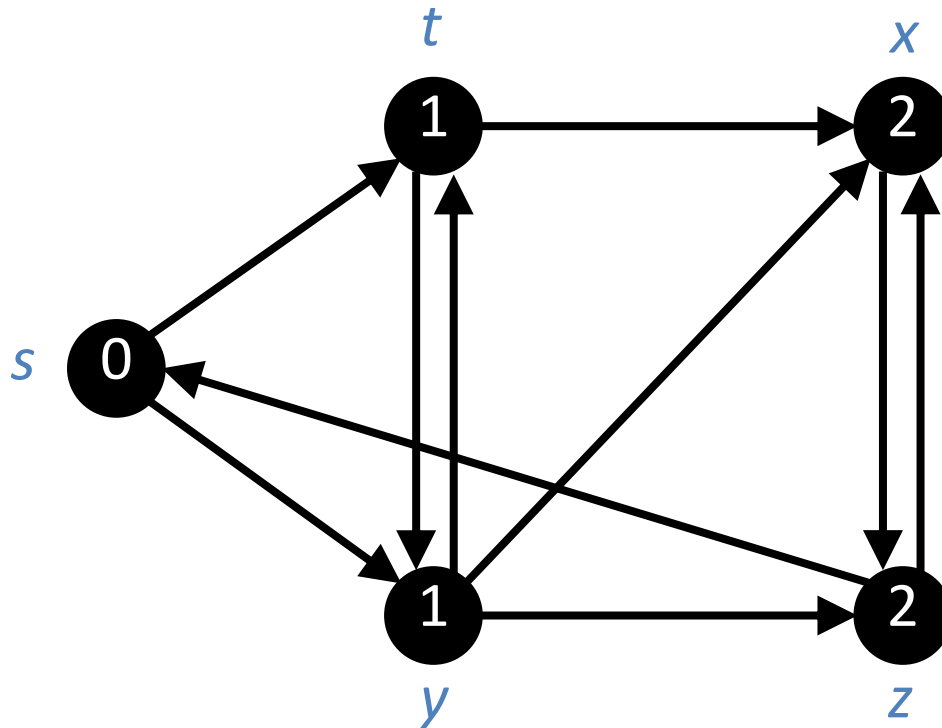


# Customized breadth-first search





# Customized breadth-first search



Can we generalize BFS to use edge weights?

# Principle of a single-source shortest-path algorithm

For each vertex  $v \in V$  :

- $d[v] = \delta(s, v)$ .
    - Initially,  $d[v] = \infty$ .
    - Reduces as algorithms progress, but always maintain  $d[v] \geq \delta(s, v)$ .
    - Call  $d[v]$  a **shortest-path estimate**.
  - $\pi[v] =$  predecessor of  $v$  on a shortest path from  $s$ .
    - If no predecessor,  $\pi[v] = NIL$ .
    - $\pi$  induces a tree - **shortest-path tree** (see proof in textbook).
- $\delta(s, v)$  is the absolute shortest path  
 $d[v]$  is our current estimate of the shortest path

# Generic algorithm structure

1. Initialization
2. Scan vertices and relax edges

The algorithms differ in the order and how many times they relax each edge.

# 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

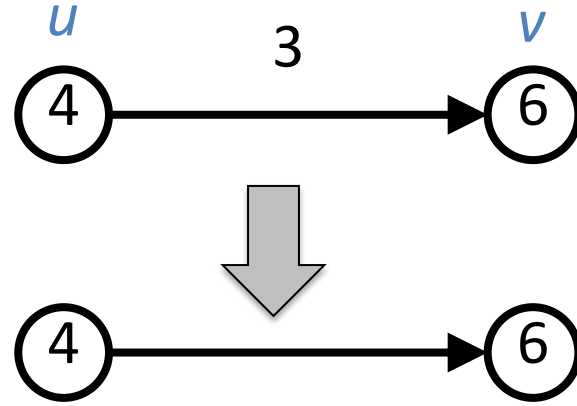
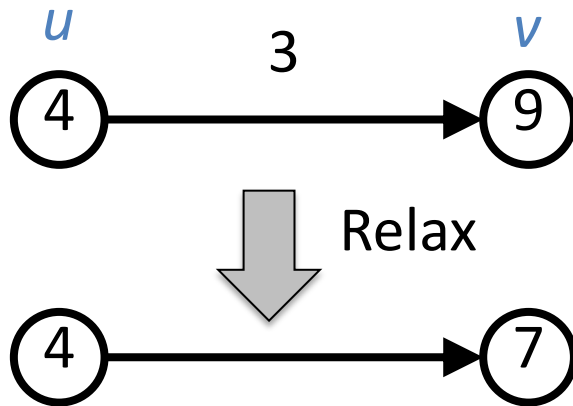
This is used to reduce  $d[v]$  during the execution of the algorithm.

```
RELAX(u, v, w)
```

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

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

```
     $\pi[v] \leftarrow u$ 
```



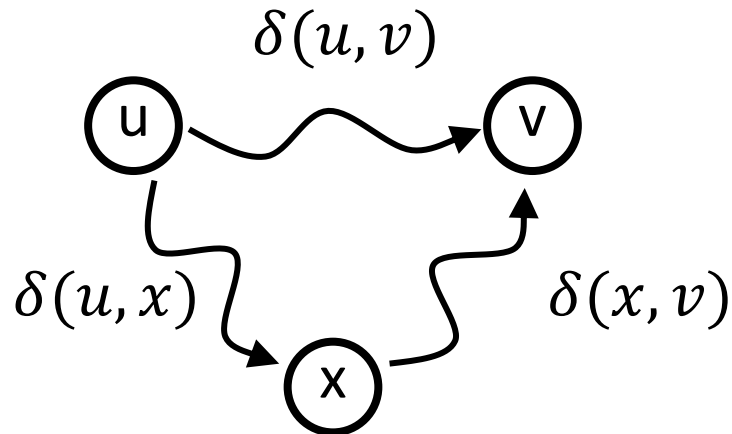
# Triangle inequality

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

## Proof:

Weight of shortest path  $u \rightsquigarrow v$  is  $\leq$  weight of any path  $u \rightsquigarrow v$ .

Path  $u \rightsquigarrow x \rightsquigarrow v$  is a path  $u \rightsquigarrow v$ , and if we use a shortest path  $u \rightsquigarrow x$  and  $x \rightsquigarrow v$ , its weight is  $\delta(u, x) + \delta(x, v)$ .



# Upper bound property

Always have  $\delta(s, v) \leq d[v]$  for all  $v$ .  
Once  $d[v] = \delta(s, v)$ , it never changes.

## Proof:

WLOG = Without Loss Of Generality

- Initially true.
- Then, assume it exists a vertex  $v$  such that  $d[v] < \delta(s, v)$ .  
WLOG,  $v$  is the first vertex for which this happens.  
Let  $u$  be the vertex that causes  $d[v]$  to change.  
Then, after relaxation  $d[v] = d[u] + \delta(u, v)$ . But we also have:  
$$d[v] < \underbrace{\delta(s, v)}_{\text{(triangle inequality)}} \leq \delta(s, u) + \underbrace{\delta(u, v)}_{\text{(v is first violation, thus inequality is valid for u)}} \leq d[u] + \delta(u, v)$$
  
 $\Rightarrow d[v] < d[u] + \delta(u, v)$ .  
Contradicts  $d[v] = d[u] + \delta(u, v)$ .

Note the strict inequality!

# No-path property

If  $\delta(s, v) = \infty$ , then  $d[v] = \infty$  always.

**Proof:**  $d[v] \geq \delta(s, v) = \infty \Rightarrow d[v] = \infty$ .



# Convergence property

We need all 3 conditions!

In 1:  $(u, v)$  is the last edge on the shortest path

If:

1.  $s \rightsquigarrow u \rightarrow v$  is a shortest path,
  2.  $d[u] = \delta(s, u)$ ,
  3. we call  $RELAX(u, v, w)$ ,
- then  $d[v] = \delta(s, v)$  afterward.

In other words, after  $RELAX$   $d[v]$  is guaranteed to be the shortest path value.

**Proof:**

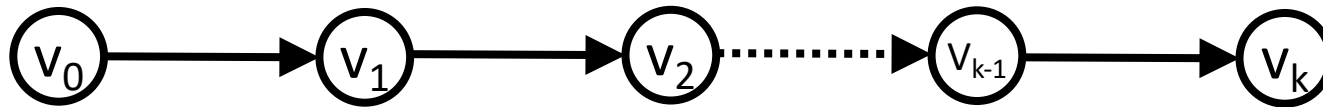
After relaxation:

$$\begin{aligned} d[v] &\leq d[u] + w(u, v) && \text{(RELAX code)} \\ &= \delta(s, u) + w(u, v) && (d[u] = \delta(s, u) \text{ by 2.}) \\ &= \delta(s, v) && (s \rightsquigarrow u \rightarrow v \text{ is a shortest path} \\ &&& \text{\& lemma sub-optimal structure}) \end{aligned}$$

Since  $d[v] \geq \delta(s, v)$ , must have  $d[v] = \delta(s, v)$ .

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



## Proof:

Induction to show that  $d[v_i] = \delta(s, v_i)$  after  $(v_{i-1}, v_i)$  is relaxed.

**Base Case:**  $i = 0$ . Initially,  $d[v_0] = 0 = \delta(s, v_0) = \delta(s, s)$ .

**Inductive step:** Assume  $d[v_{i-1}] = \delta(s, v_{i-1})$ . Relax  $(v_{i-1}, v_i)$ . By *convergence property*,  $d[v_i] = \delta(s, v_i)$  afterward and  $d[v_i]$  never changes.

# Single-source shortest paths in a DAG

**DAG  $\Rightarrow$  no negative-weight cycles.**

**DAG-SHORTEST-PATHS ( $V, E, w, s$ )**

**topologically sort the vertices**

**INIT-SINGLE-SOURCE ( $V, s$ )**

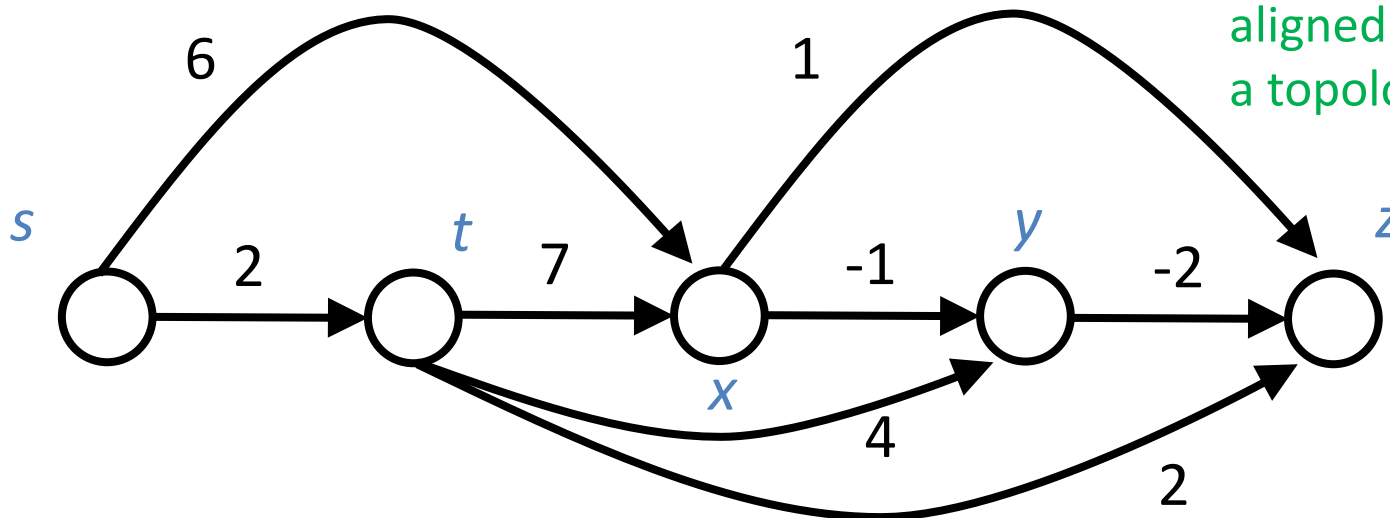
**for** each vertex  $u$  in topological order **do**

**for** each vertex  $v \in Adj[u]$  **do**

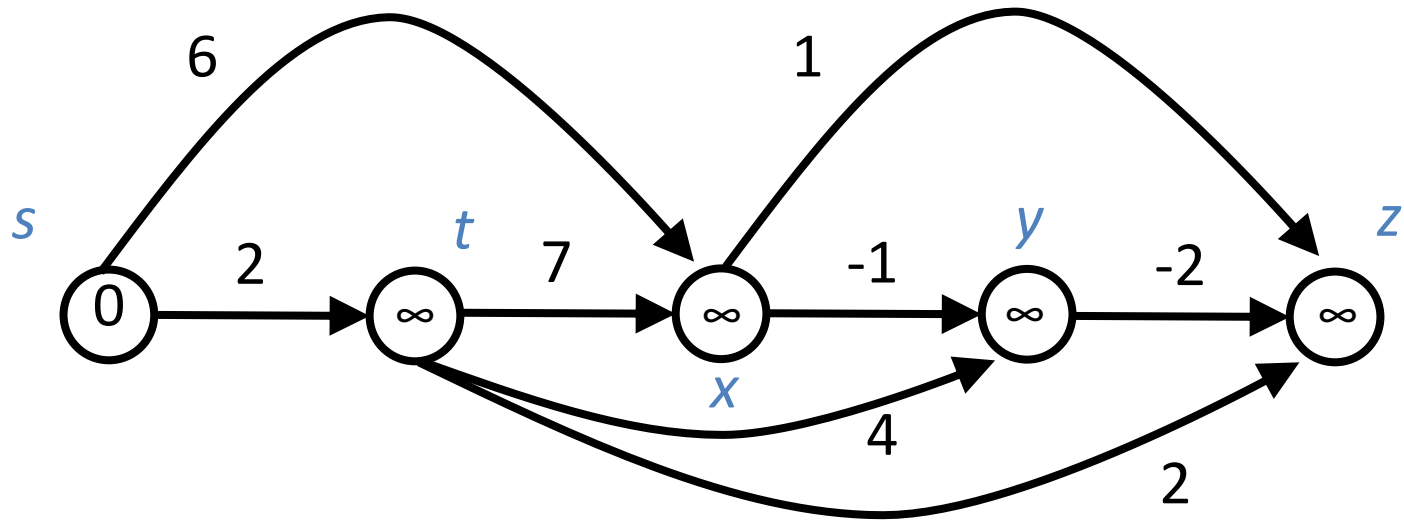
        RELAX( $u, v, w$ )

Note: the edges  
can have any  
weight you want.

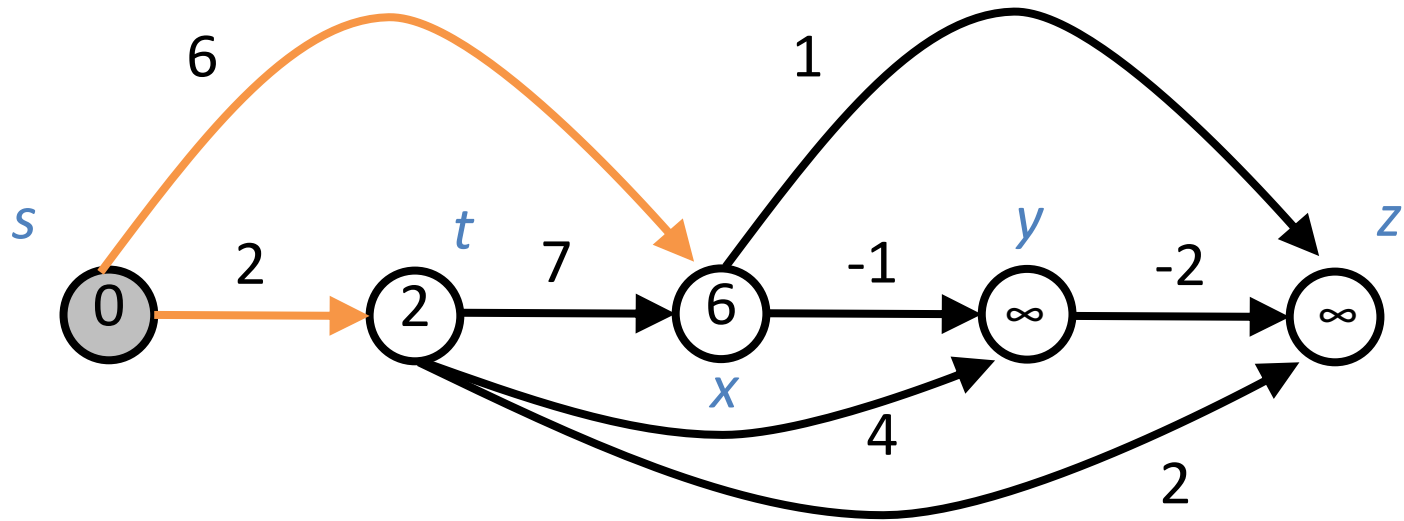
The vertices are  
aligned according to  
a topological order.



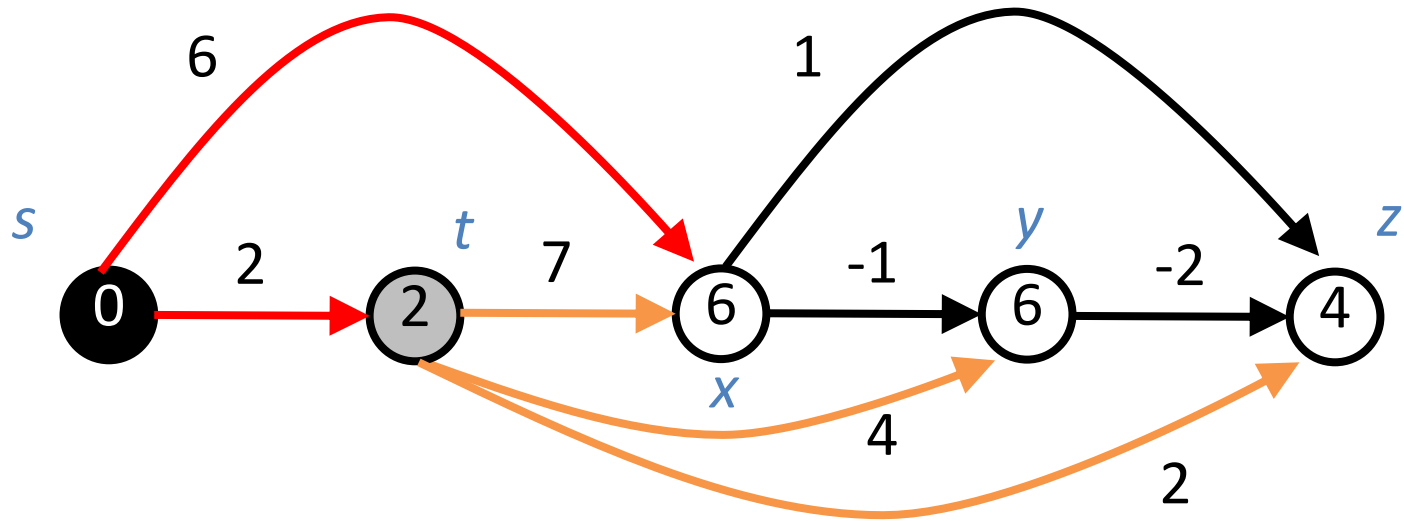
# Example



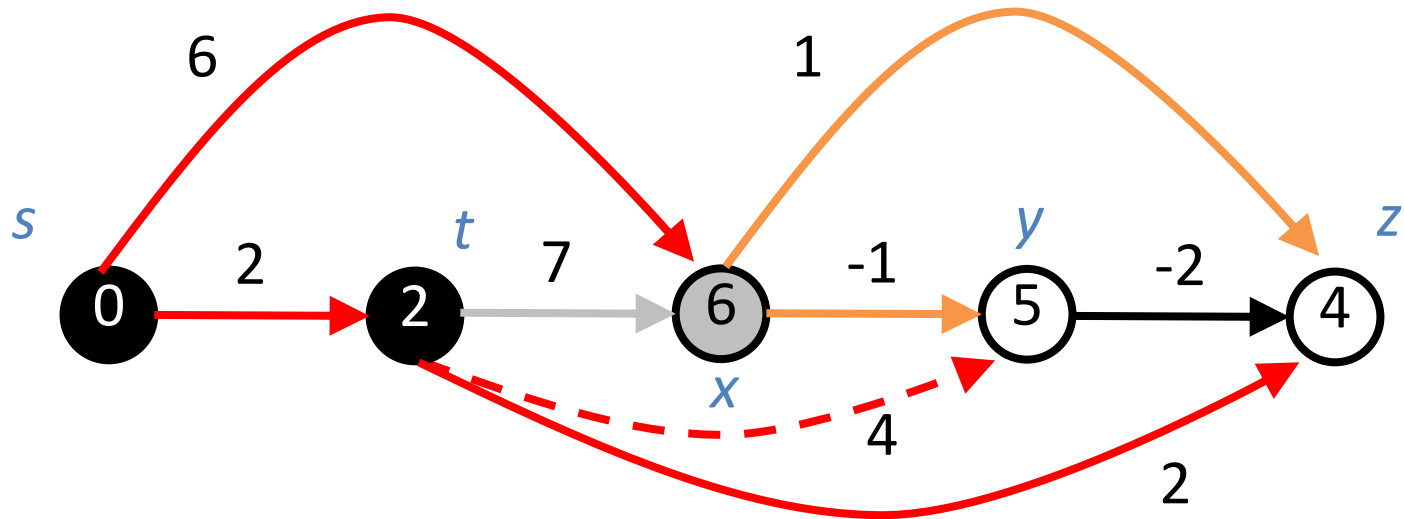
# Example



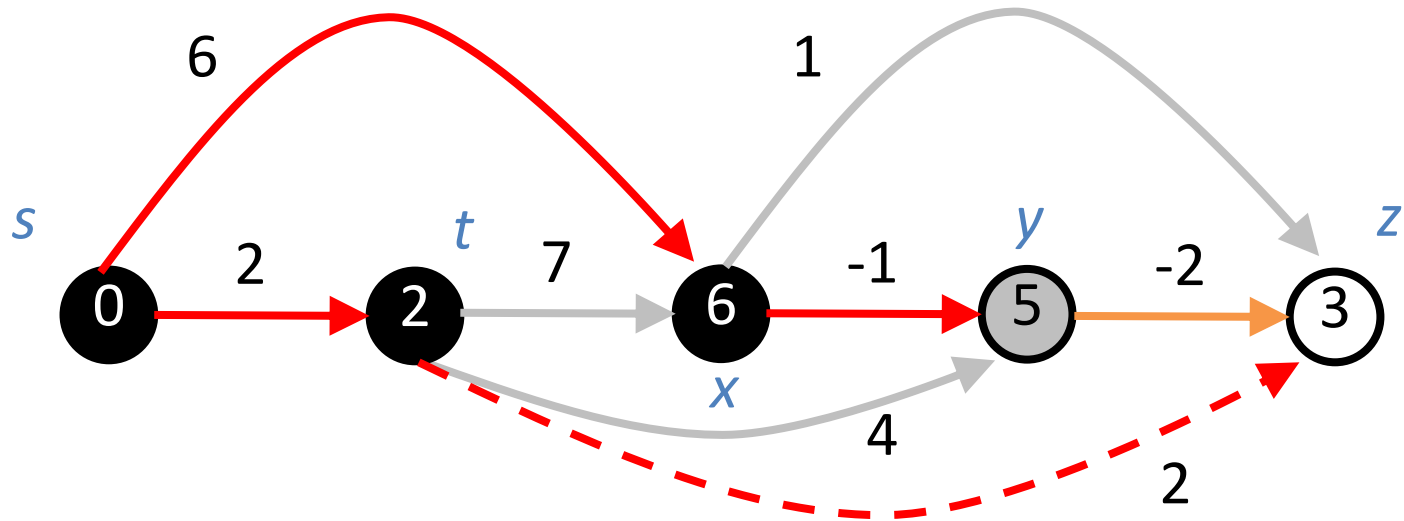
# Example



# Example

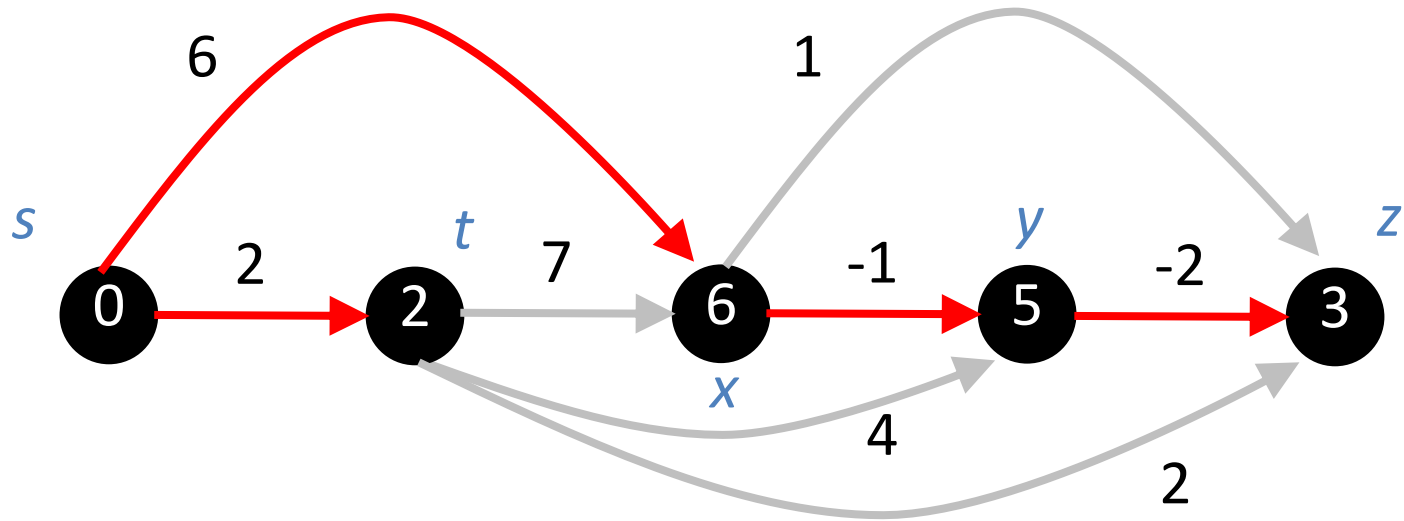


# Example





# Example



# Single-source shortest paths in a DAG

DAG-SHORTEST-PATHS ( $V, E, w, s$ )

topologically sort the vertices

INIT-SINGLE-SOURCE ( $V, s$ )

**for** each vertex  $u$  in topological order **do**

**for** each vertex  $v \in Adj[u]$  **do**

        RELAX ( $u, v, w$ )

**Time:**  $O(V + E)$ .

No edge leads you to a vertex already processed.

**Correctness:**

Because we process vertices in topologically sorted order, edges of **any** path must be relaxed in order of appearance in the path.

⇒ Edges on any shortest path are relaxed in order.

⇒ By the path-relaxation property, the result is correct.

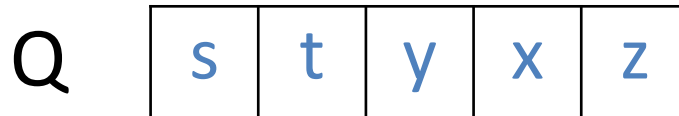
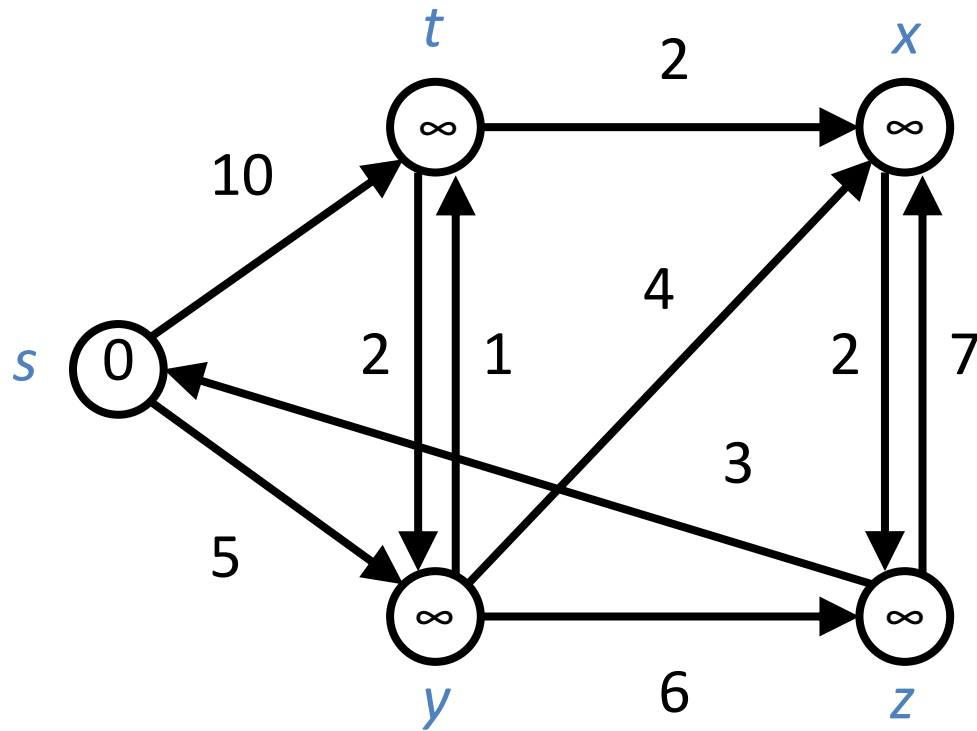
# Dijkstra's algorithm

- **No negative-weight edges.**
- Weighted version of BFS:
  - **Instead of a FIFO queue, uses a priority queue.**
  - Keys are shortest-path weights ( $d[v]$ ).
- Have two sets of vertices:
  - $S$  = vertices whose final shortest-path weights are determined,
  - $Q$  = priority queue =  $V \setminus S$ .
- Similar Prim's algorithm, but computing  $d[v]$ , and using shortest-path weights as keys.
- Greedy choice: At each step we choose the light edge.

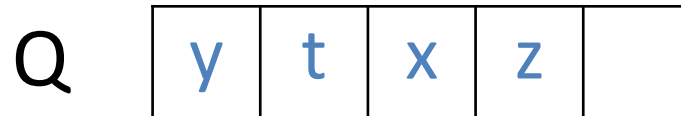
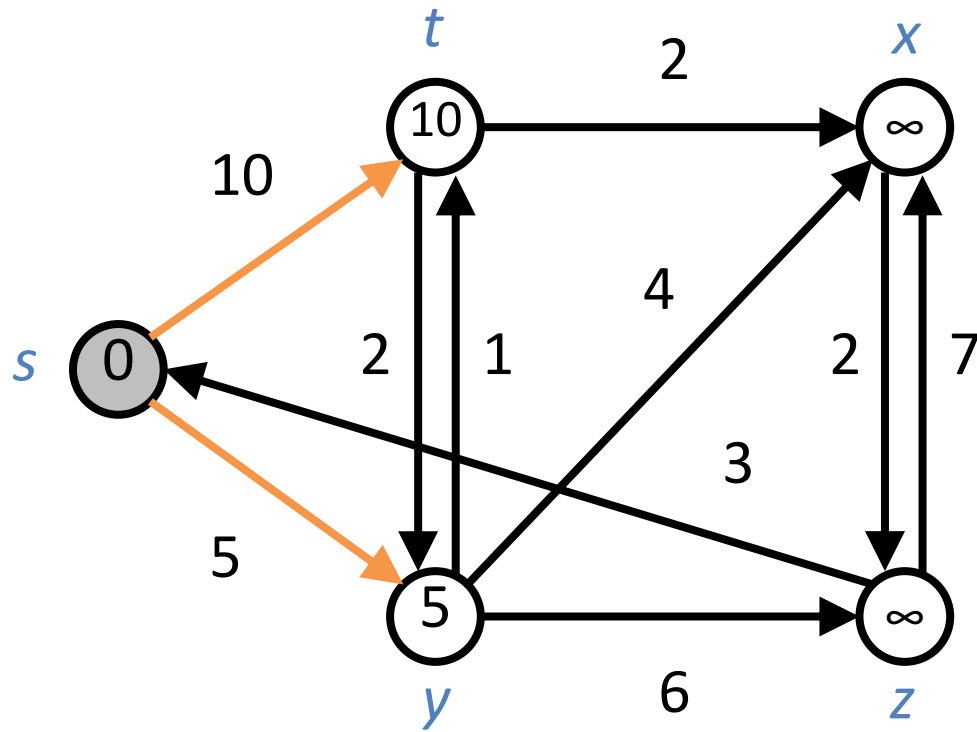
# Dijkstra's algorithm

```
DIJKSTRA( $V, E, w, s$ )  
INIT-SINGLE-SOURCE( $V, s$ )  
 $S \leftarrow \emptyset$   
 $Q \leftarrow V$   
while  $Q \neq \emptyset$  do  
     $u \leftarrow \text{EXTRACT-MIN}(Q)$   
     $S \leftarrow S \cup \{u\}$   
    for each vertex  $v \in \text{Adj}[u]$  do  
        RELAX( $u, v, w$ )
```

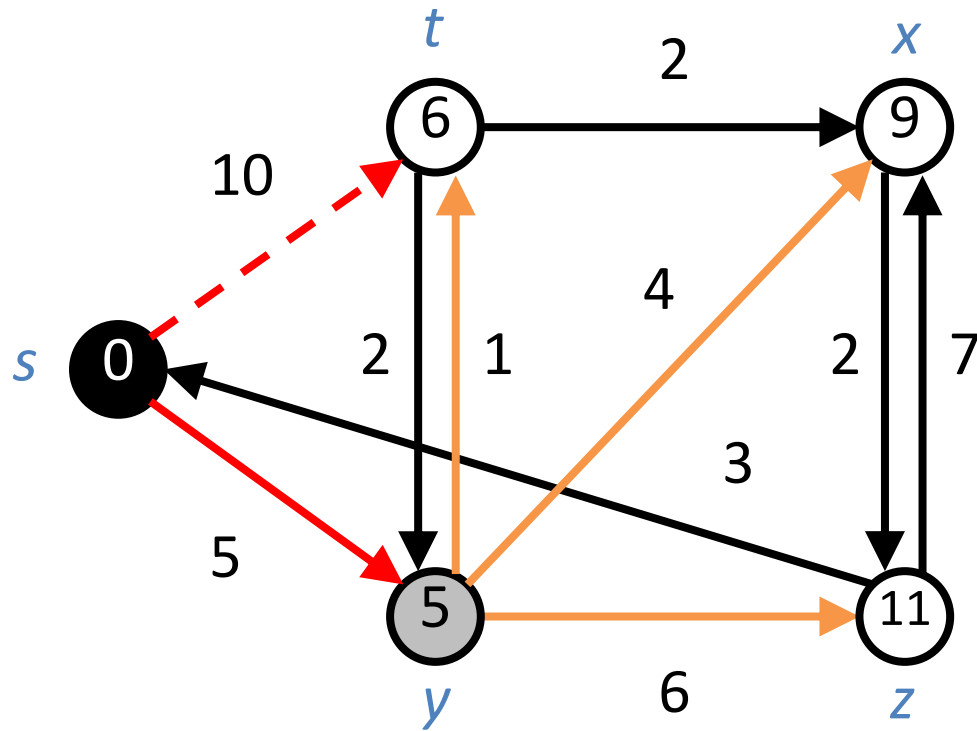
# Example



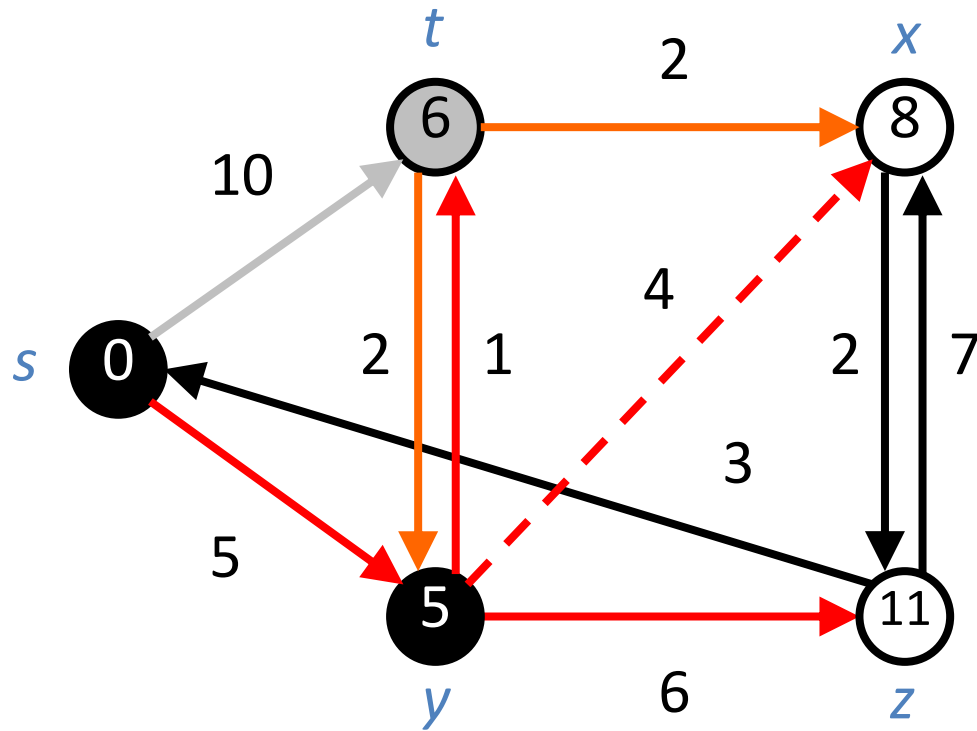
# Example



# Example

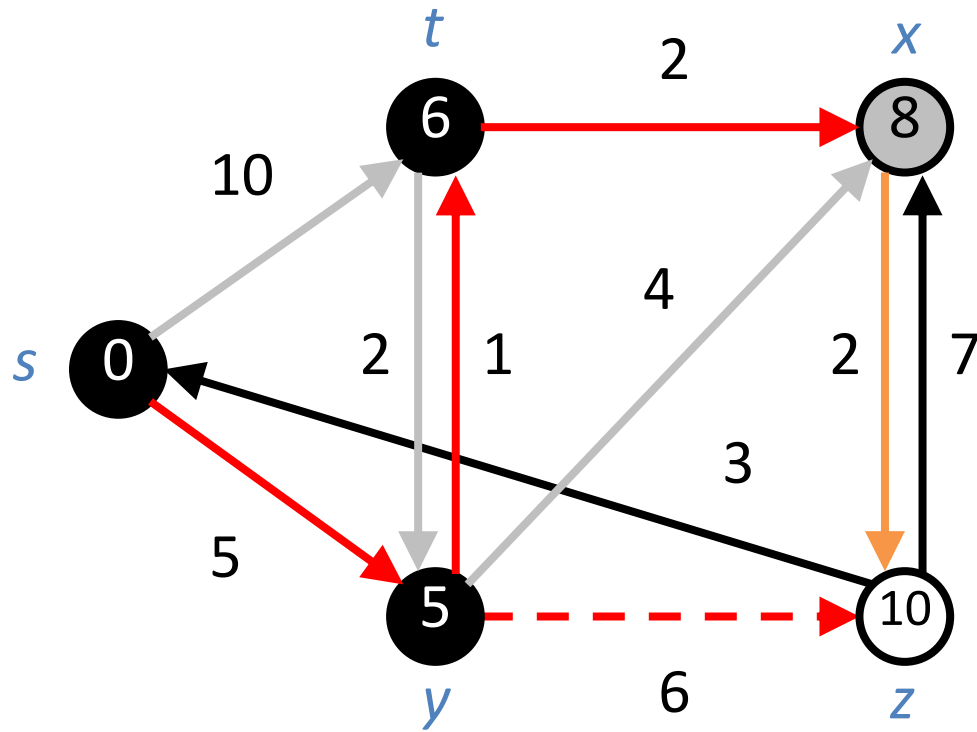


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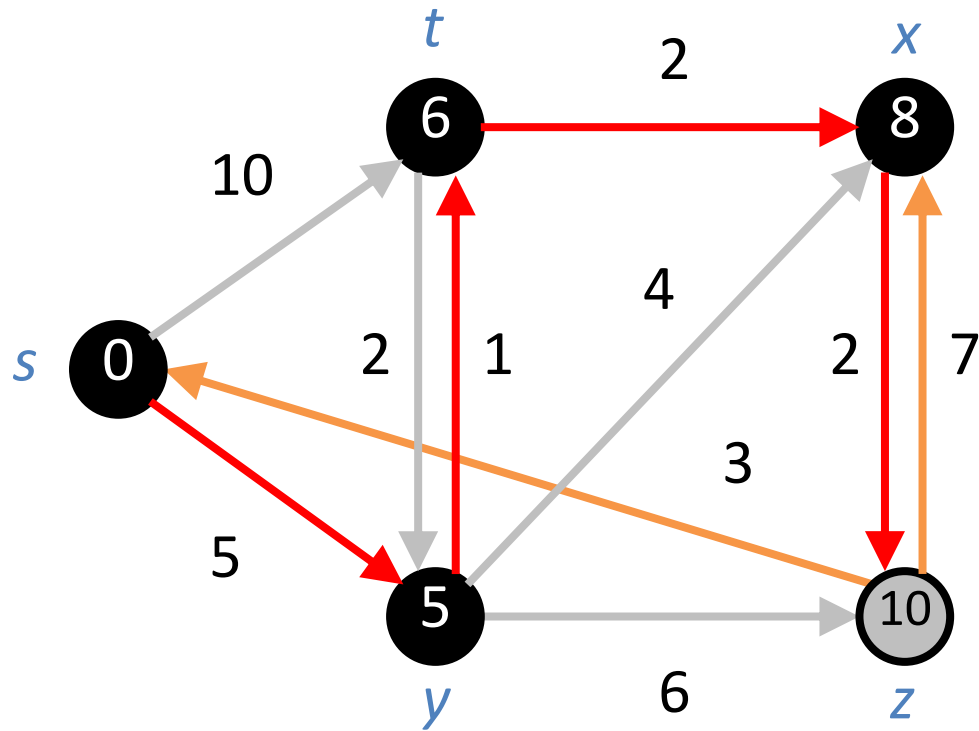




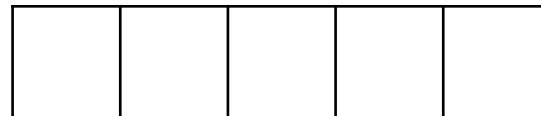
# Example



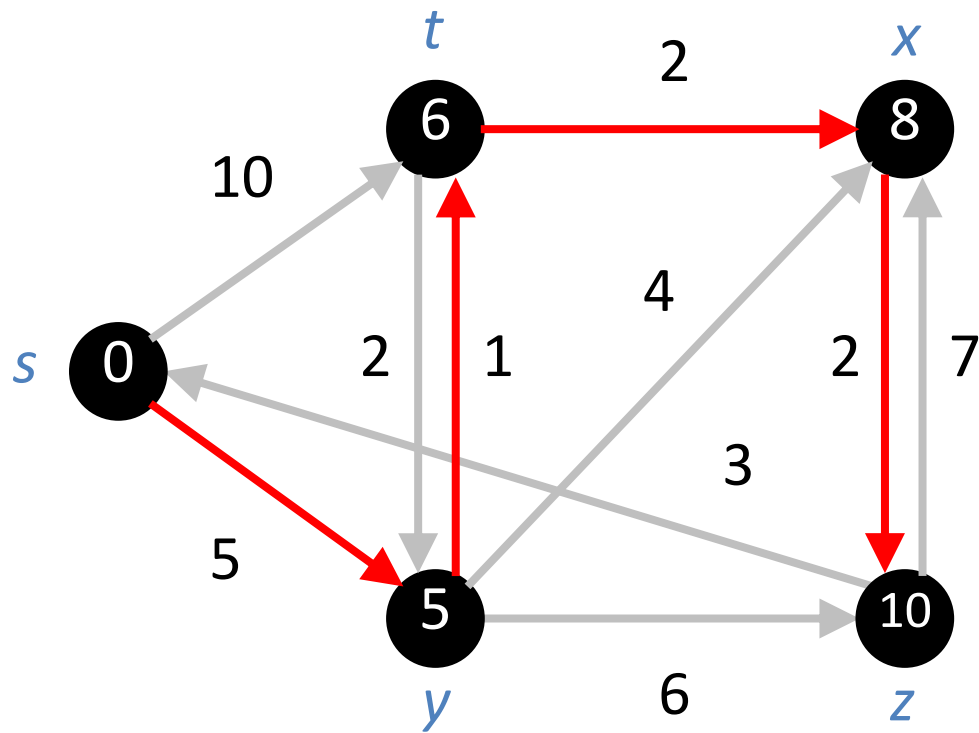
# Example



Q



# Example



# Correctness

## Loop invariant property:

At the end of each iteration of the while loop,  $d[v] = \delta(s, v)$  for all  $v \in S$ .

## Initialization:

Initially,  $S = \emptyset$ , so trivially true.

## Termination:

At each iteration, we remove one vertex from  $Q$  and never add one in. Thus, the algorithm terminates after  $|V|$  iterations.

## Maintenance:

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

# Correctness (cont'd)

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

Suppose there exists  $u$  such that  $d[u] \neq \delta(s, u)$ .

Let  $u$  be the first vertex for which  $d[u] \neq \delta(s, u)$  when  $u$  is added to  $S$ .

- $u \neq s$ , since  $d[s] = \delta(s, s) = 0$ .
- Therefore,  $s \in S$  and  $S \neq \emptyset$ .

Now, we can discuss the general case:

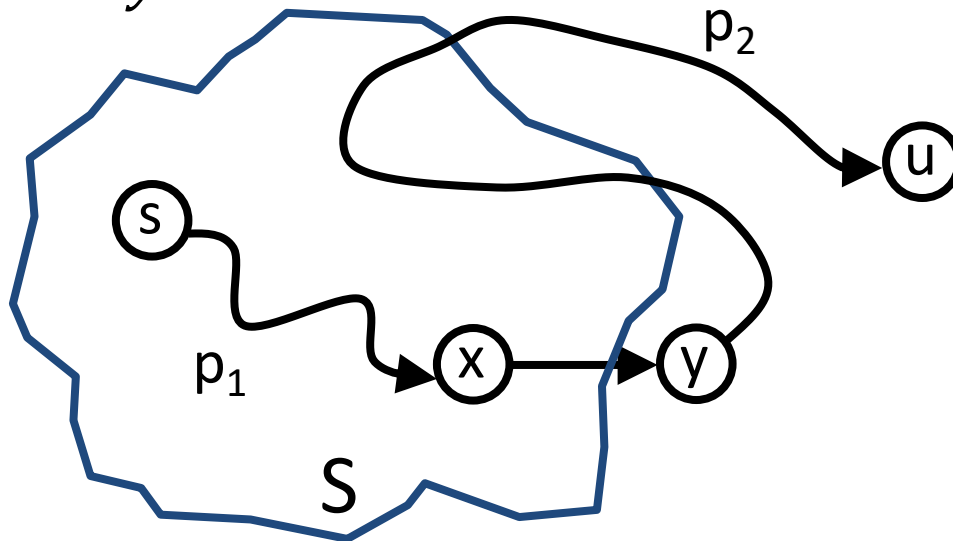
- There must be some path  $s \rightsquigarrow u$ . Otherwise  $d[u] = \delta(s, u) = \infty$  by no-path property.
- Since there is a path  $s \rightsquigarrow u$ , there is also a **shortest  $p$  path  $s \rightsquigarrow u$** .

# Correctness (cont'd)

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

Just before  $u$  is added to  $S$ , the shortest path  $p$  connects a vertex in  $S$  (i.e.,  $s$ ) to a vertex in  $Q$  (i.e.,  $u$ ).

Let  $y$  be the first vertex along  $p$  that is in  $Q$  and let  $x \in S$  be the predecessor of  $y$ .



Note: if  $y = u$ , then  $d[u] = \delta(s, u)$  by the convergence property.

Decompose  $p$  into  $s \xrightarrow{p_1} x \rightarrow y \xrightarrow{p_2} u$ .

# Correctness (cont'd)

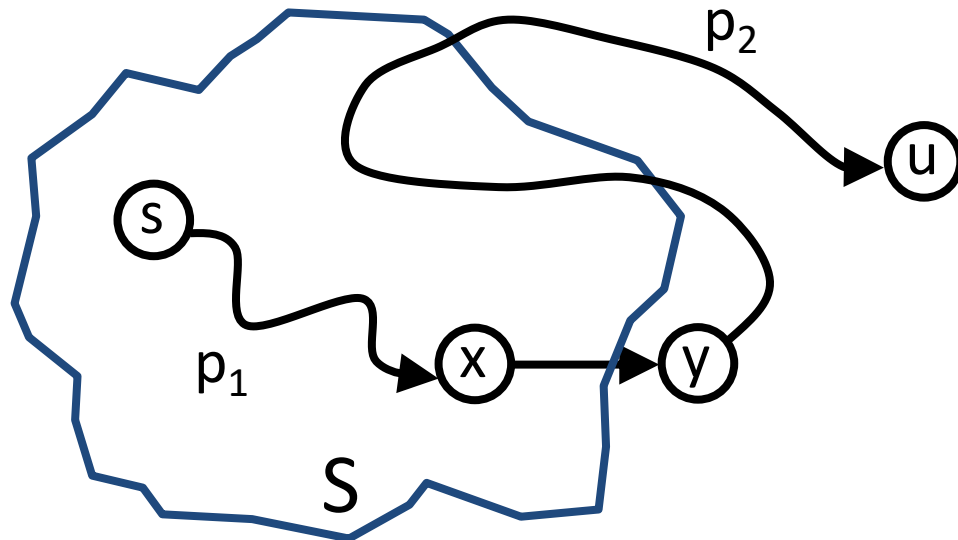
**Claim 1:**  $d[y] = \delta(s, y)$  when  $u$  is added to  $S$ .

We will need this claim to complete the proof

**Proof:**

$x \in S$  and  $u$  is the first vertex such that  $d[u] \neq \delta(s, u)$  when  $u$  is added to  $S \Rightarrow d[x] = \delta(s, x)$  when  $x$  is added to  $S$ . But when  $x$  is added we relax the edge  $(x, y)$ , so by the *convergence property*,  $d[y] = \delta(s, y)$ .

We relax all outgoing edges of a vertex before adding it into  $S$ , and  $y$  is on the shortest path.



# Correctness (cont'd)

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

Now, we can get a contradiction to  $d[u] \neq \delta(s, u)$ :

$y$  is on shortest path  $p(s \rightsquigarrow u)$ , and all edge weights are nonnegative.

$$\Rightarrow \delta(s, y) \leq \delta(s, u)$$

Now, by Claim 1 we have  $d[y] = \delta(s, y)$

$$\leq \delta(s, u) \quad (\text{See inequality above})$$

$$\leq d[u] \quad (\text{upper-bound property})$$

In addition,  $y$  and  $u$  were in  $Q$  when we chose  $u$ , **thus  $d[u] \leq d[y]$** .

*We can now conclude:*

*We have  $d[y] \leq d[u]$  &  $d[u] \leq d[y] \Rightarrow d[u] = d[y]$ .*

This is why we use  
the priority queue!

Therefore,  $d[y] = \delta(s, y) \leq \delta(s, u) \leq d[u] = d[y]$

Contradicts assumption that  $d[u] \neq \delta(s, u)$ . ■



# Correctness

## Loop invariant property:

At the end of each iteration of the while loop,  $d[v] = \delta(s, v)$  for all  $v \in S$ .

**Initialization:** ✓

**Termination:** ✓

**Maintenance:** ✓

**Conclude:** At the end,  $Q = \emptyset \Rightarrow S = V \Rightarrow d[v] = \delta(s, v)$  for all  $v \in V$ .

# Analysis

It depends on implementation of priority queue.

If binary heap, each operation takes  $O(\lg V)$  time  
 $\Rightarrow O(E \lg V)$ .

Note: We can achieve  $O(V \lg V + E)$  with Fibonacci heaps.