

## 6.4: Single-Source Shortest Paths

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Lent 2016



UNIVERSITY OF  
CAMBRIDGE

Introduction

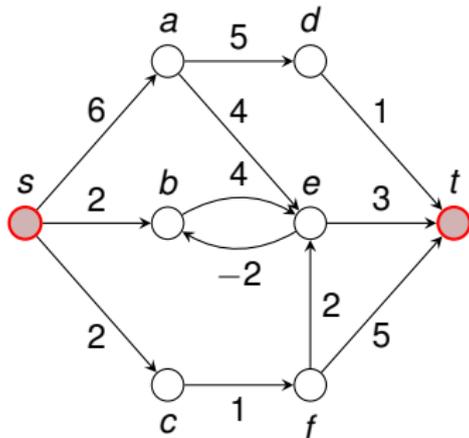
Bellman-Ford Algorithm



## Shortest Path Problem

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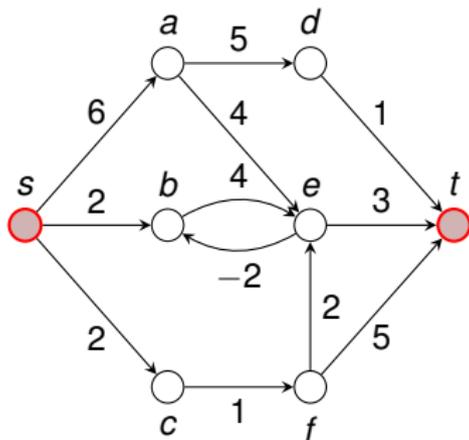
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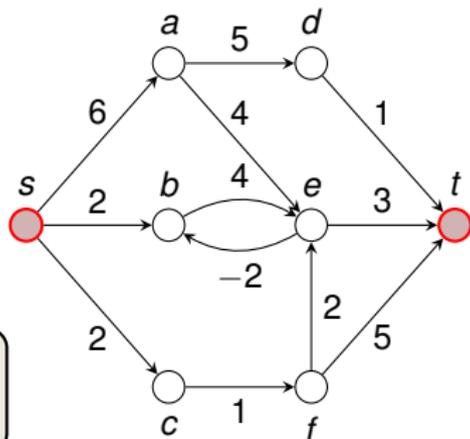


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$p = (v_0 = s, v_1, \dots, v_k = t)$  such that  $w(p) = \sum_{i=1}^k w(v_{k-1}, v_k)$  is minimized.

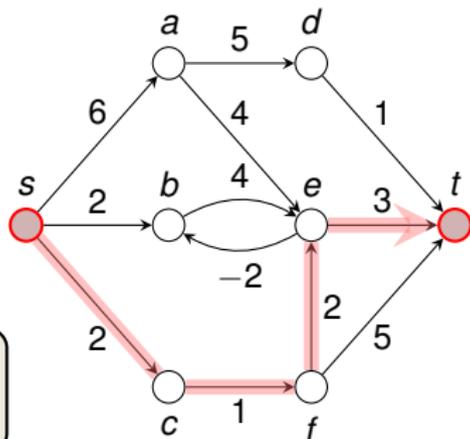


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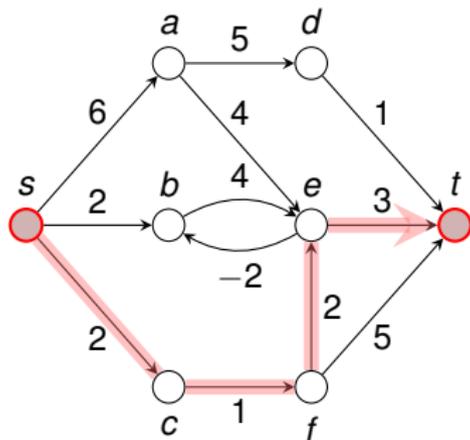
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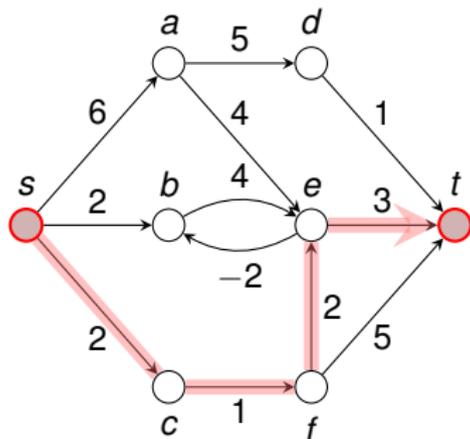
What if  $G$  is **unweighted**?



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What if  $G$  is **unweighted**?

Two possible answers are:

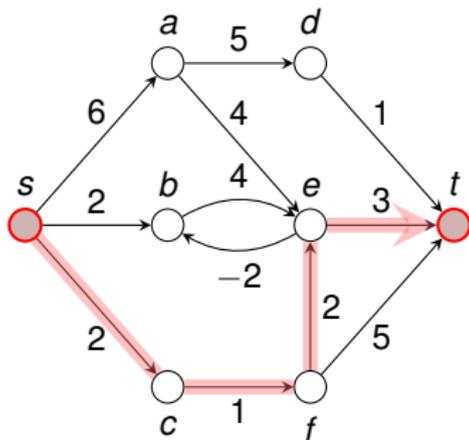
1. Run BFS (computes shortest paths in unweighted graphs)
2. Set the weight of all edges to 1



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### Applications

- Car Navigation, Internet Routing, Arbitrage in Concurrency Exchange

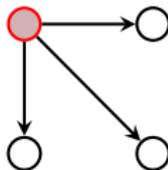


## Variants of Shortest Path Problems

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Single-source shortest-paths problem (SSSP)

- Bellman-Ford Algorithm
- Dijkstra Algorithm

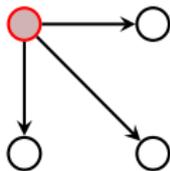


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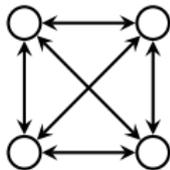
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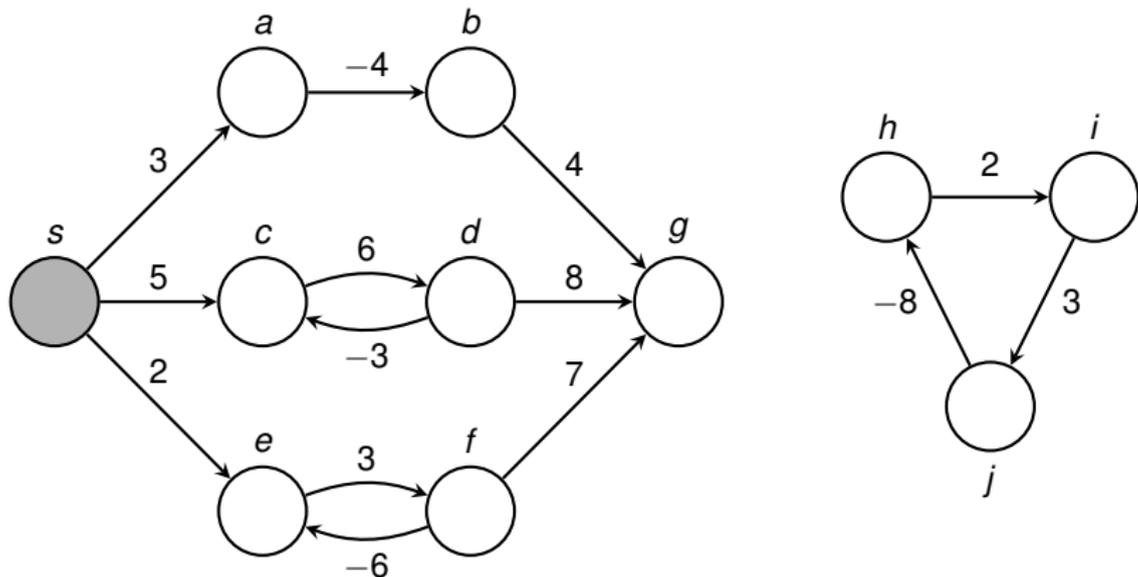


### All-pairs shortest-paths problem (APSP)

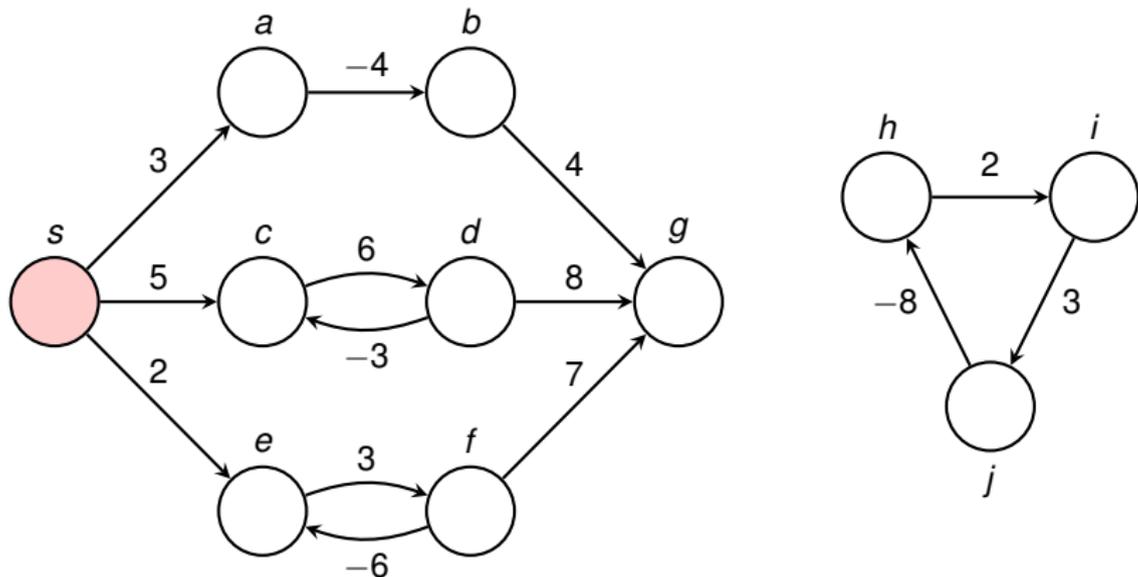
- Shortest Paths via Matrix Multiplication
- Johnson's Algorithm



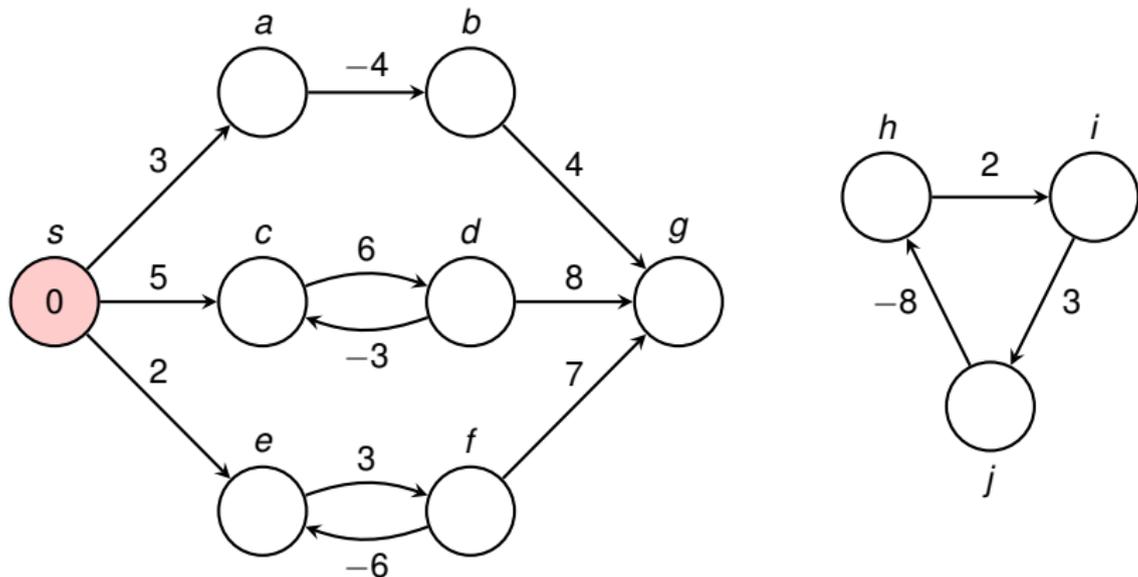
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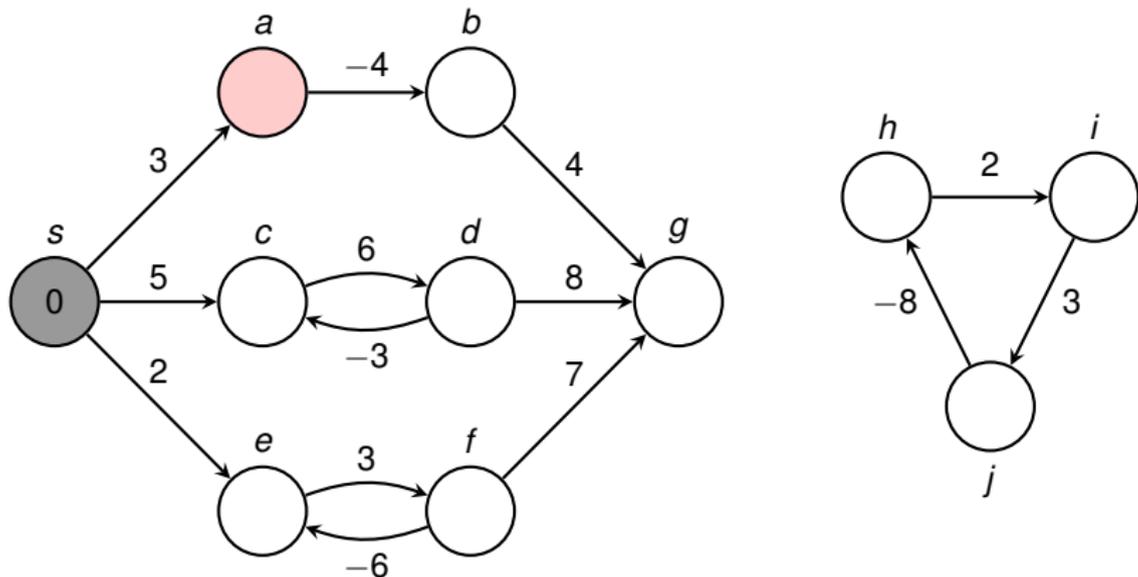
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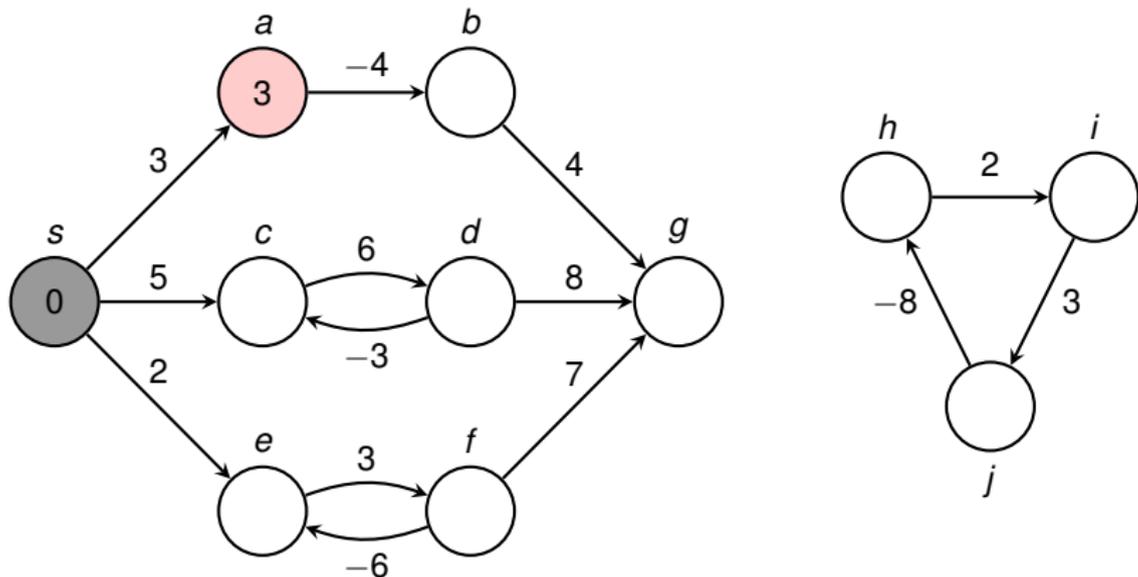
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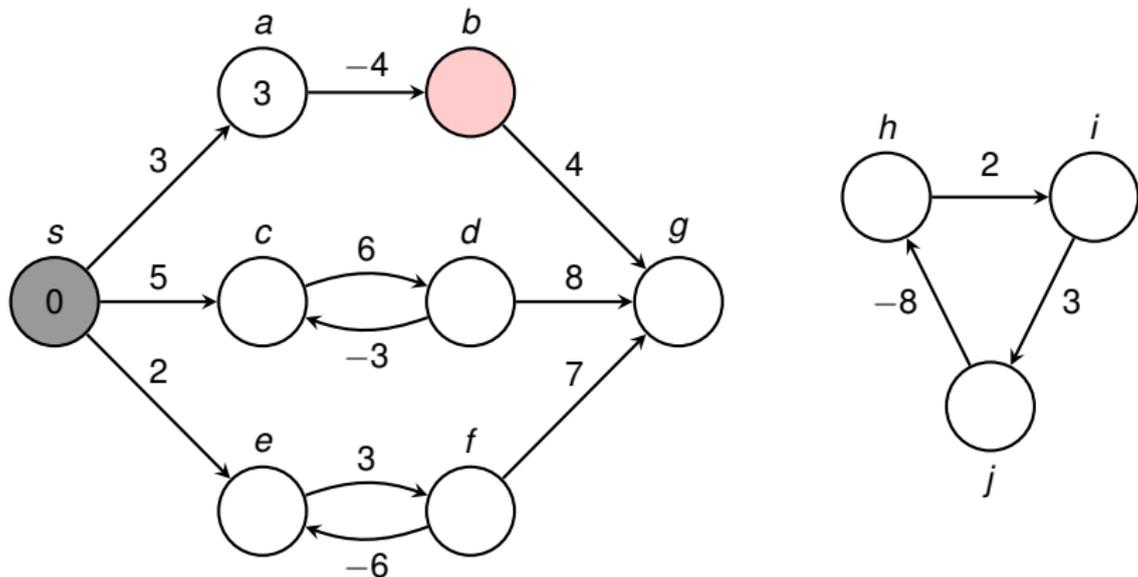
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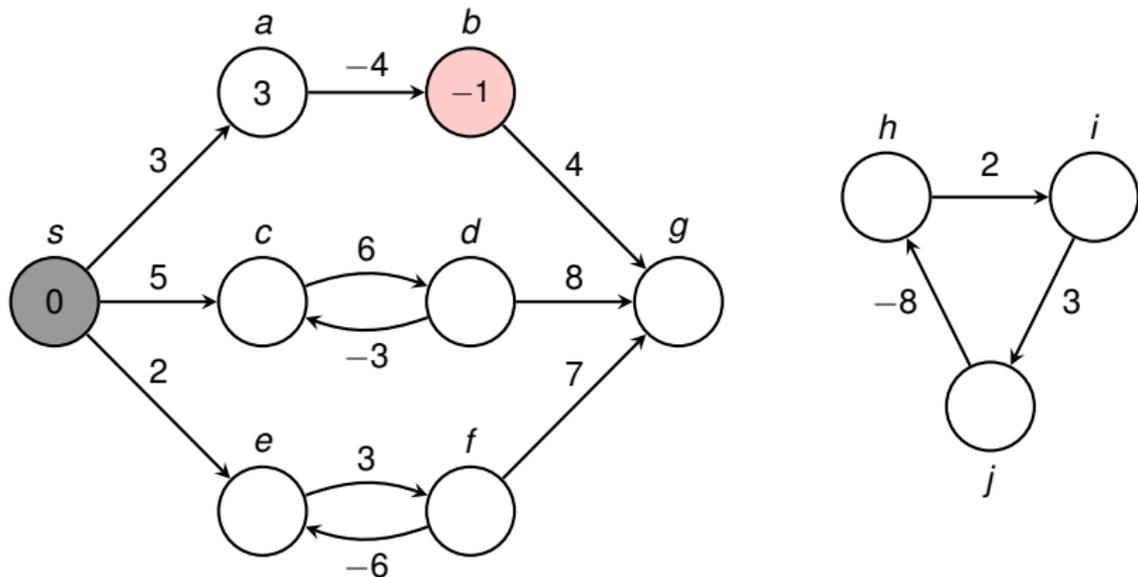
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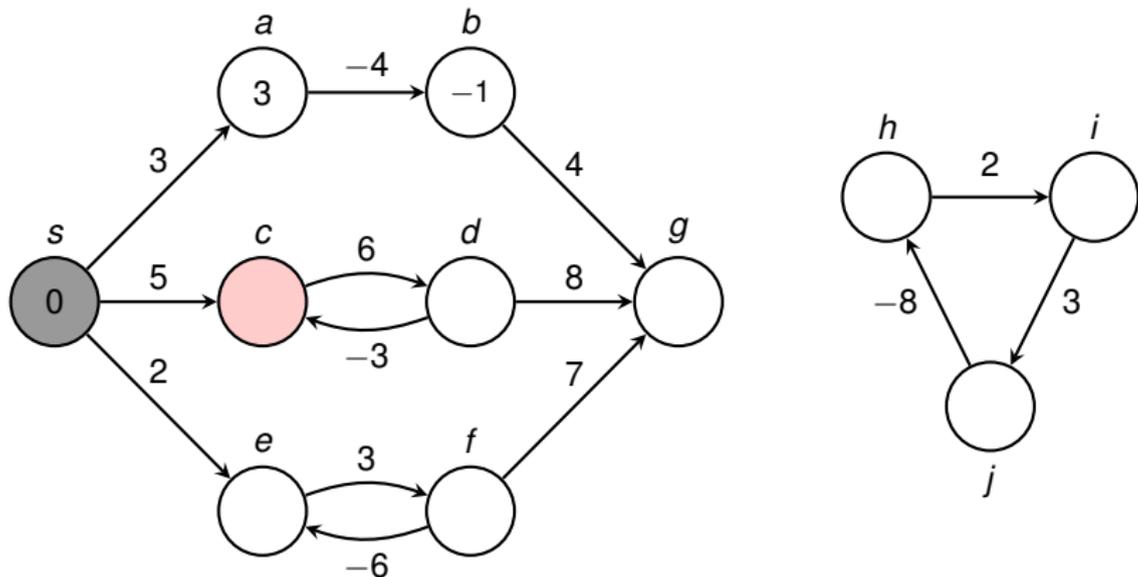
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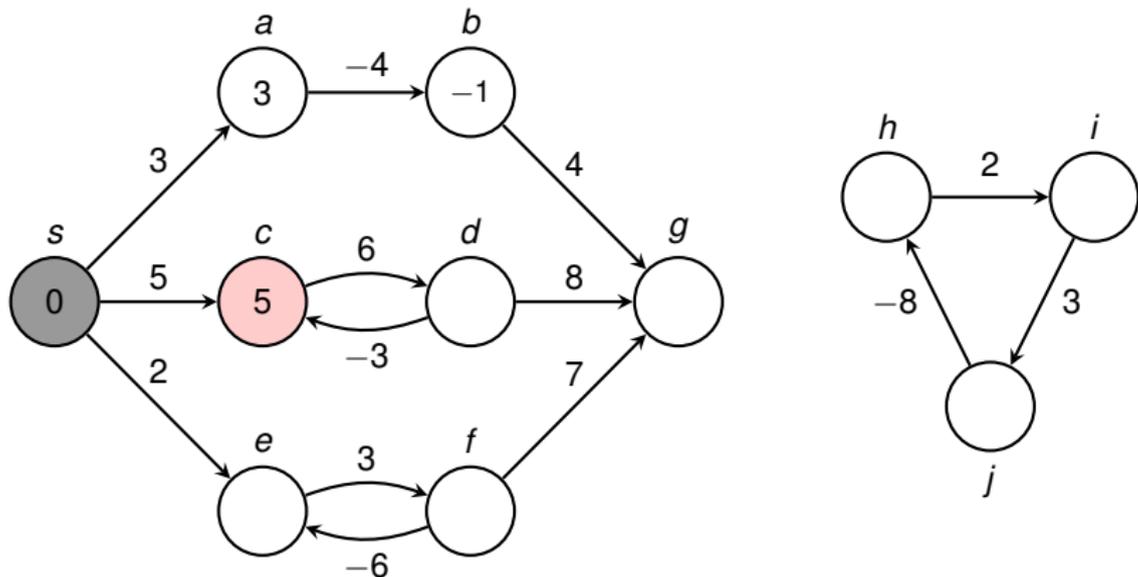
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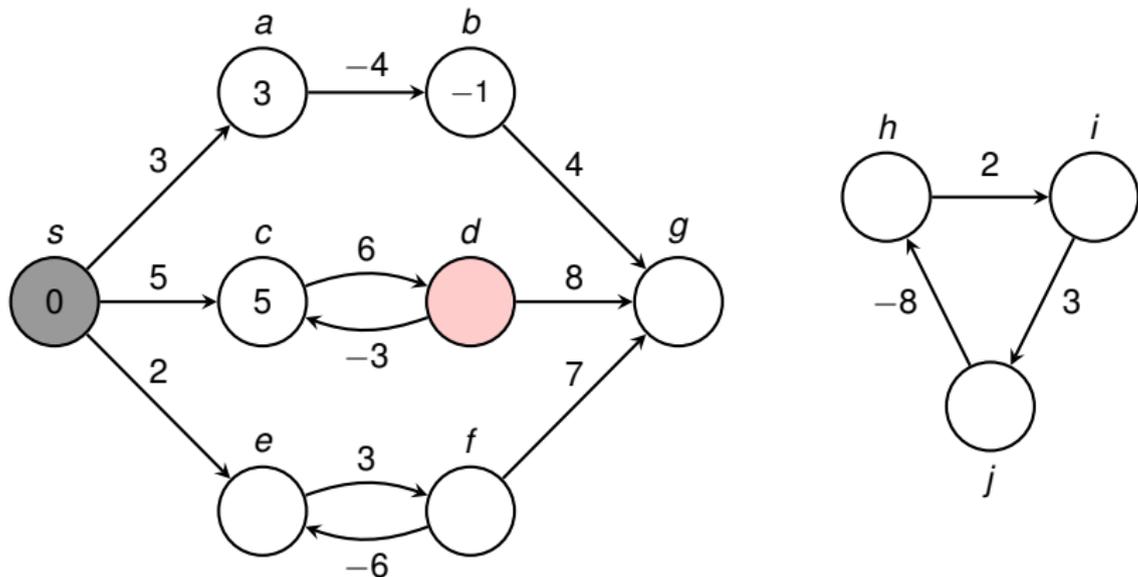
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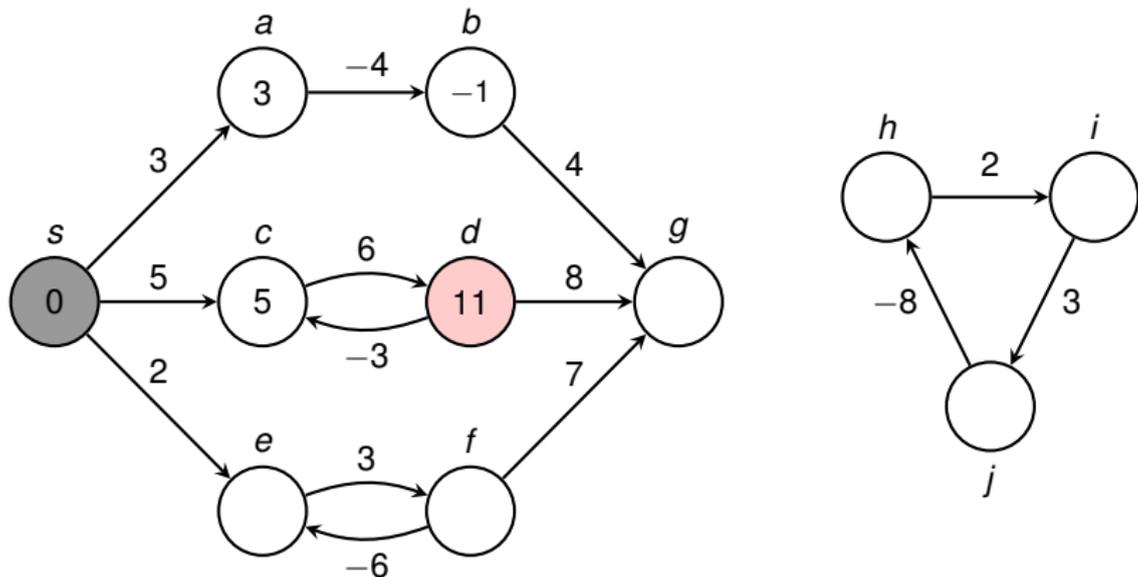
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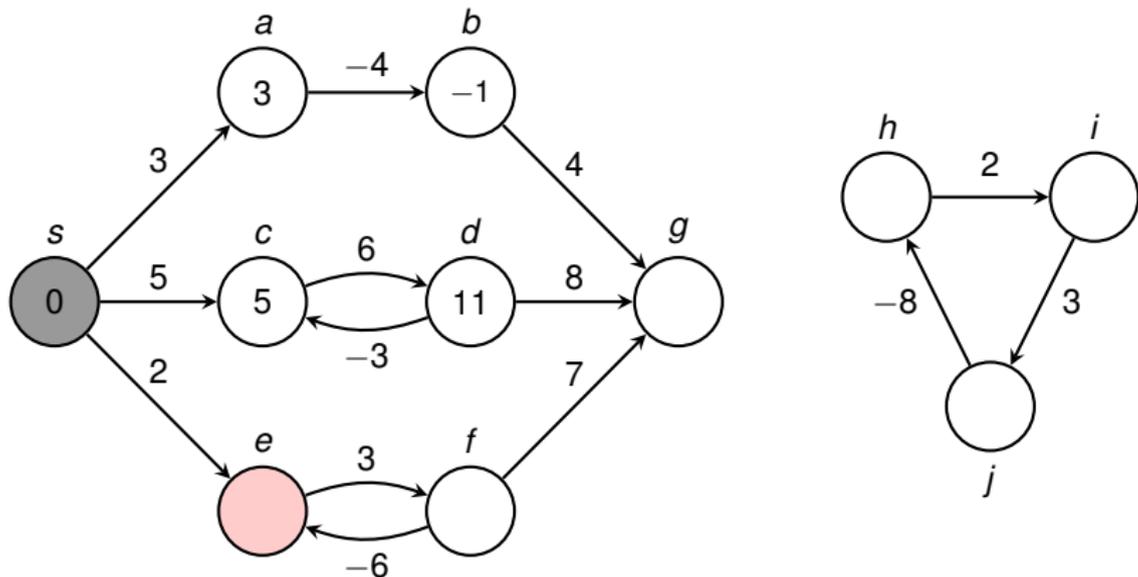
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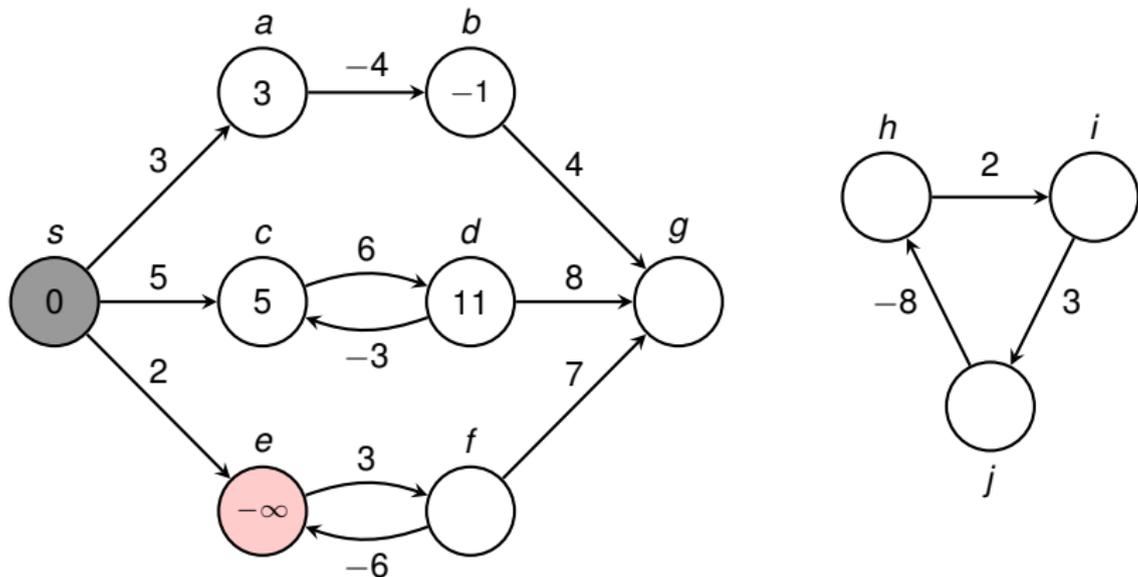
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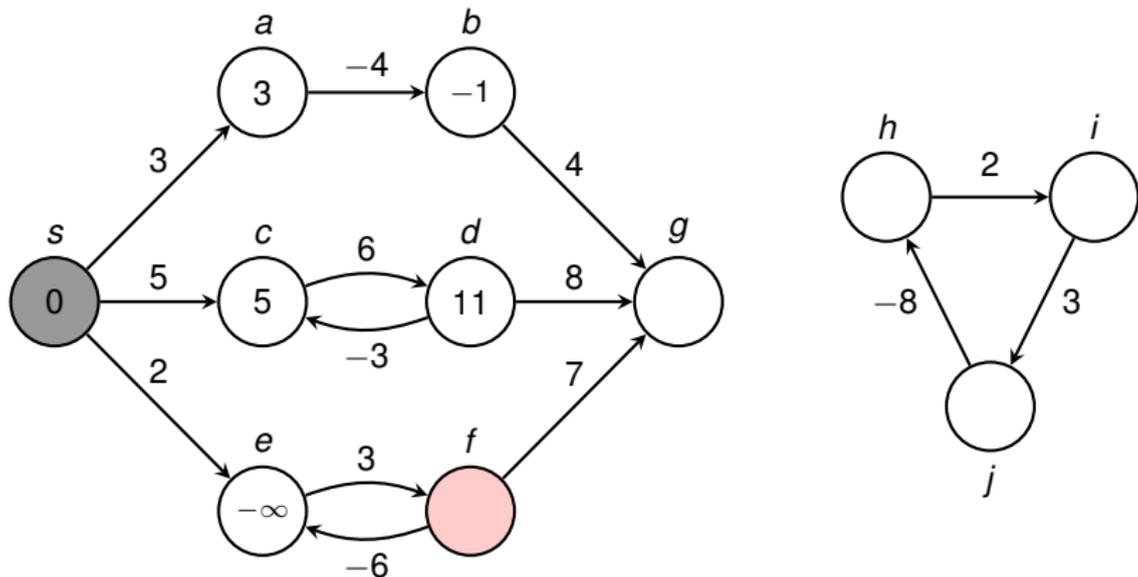
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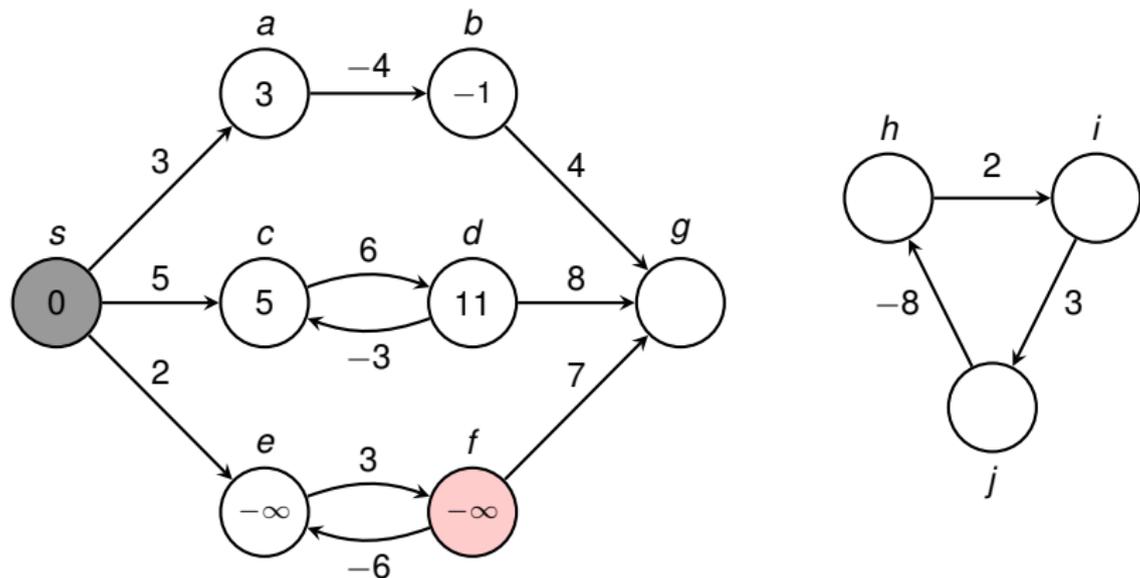
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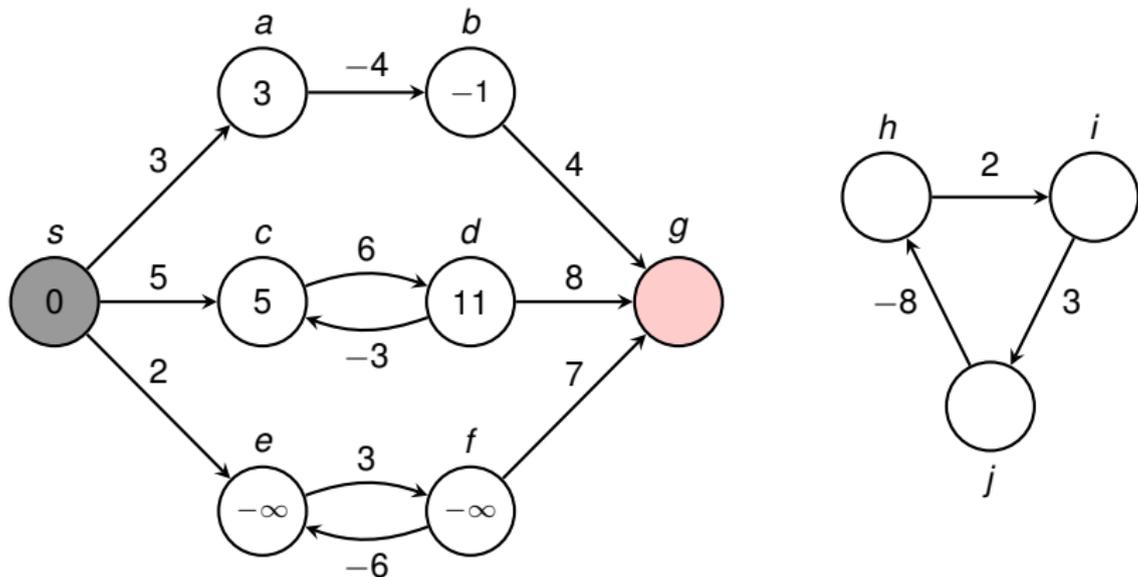
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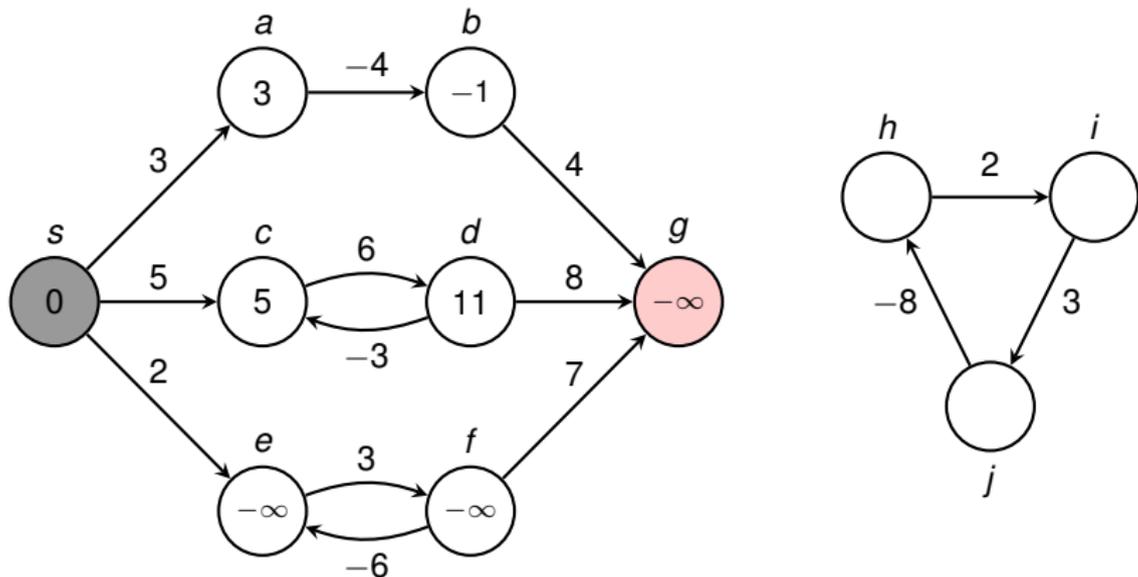
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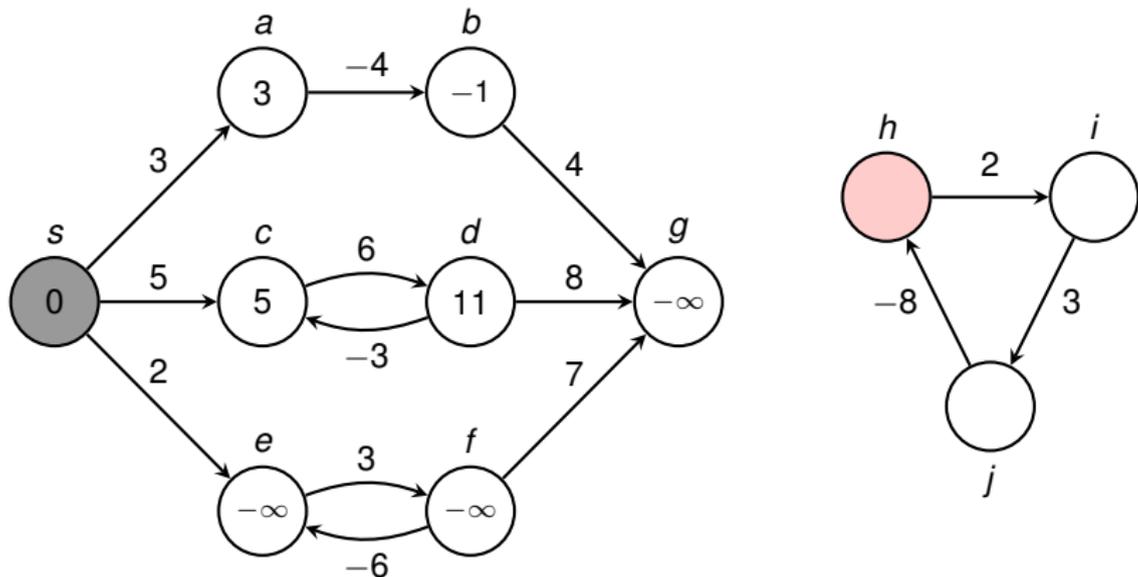
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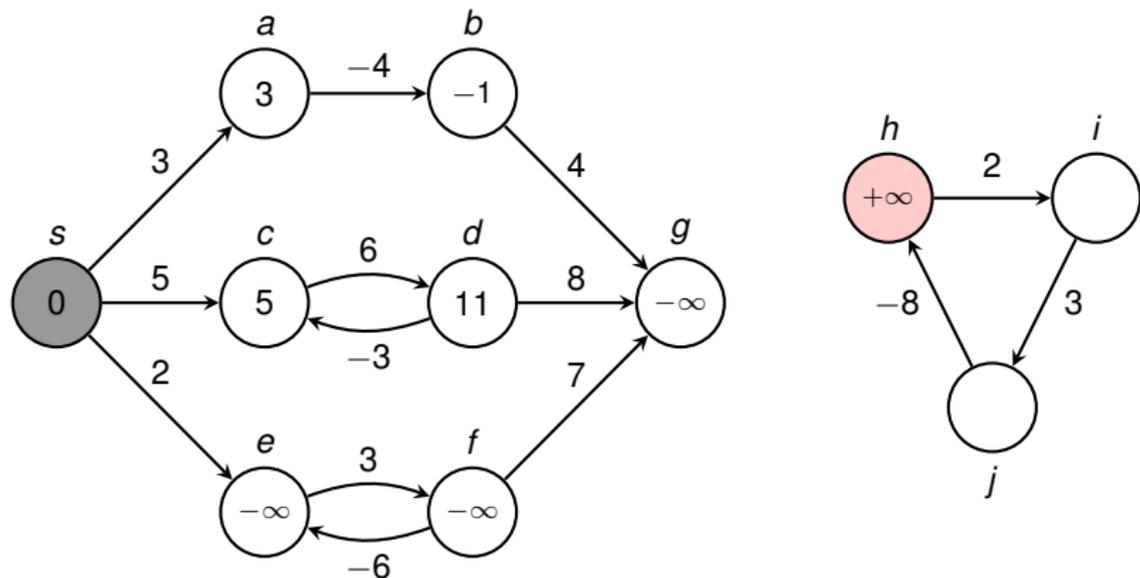
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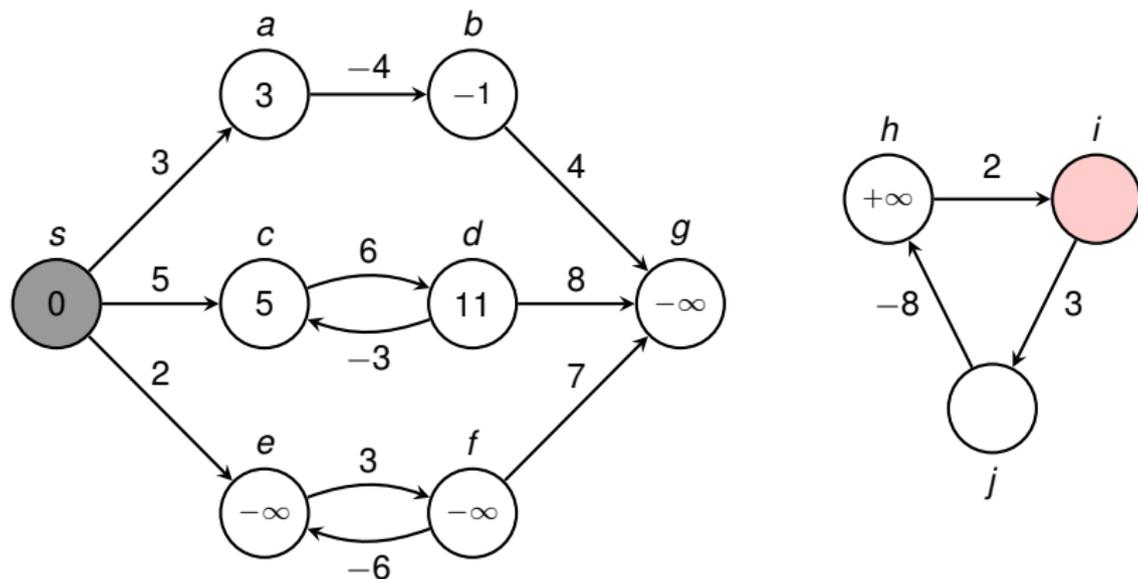
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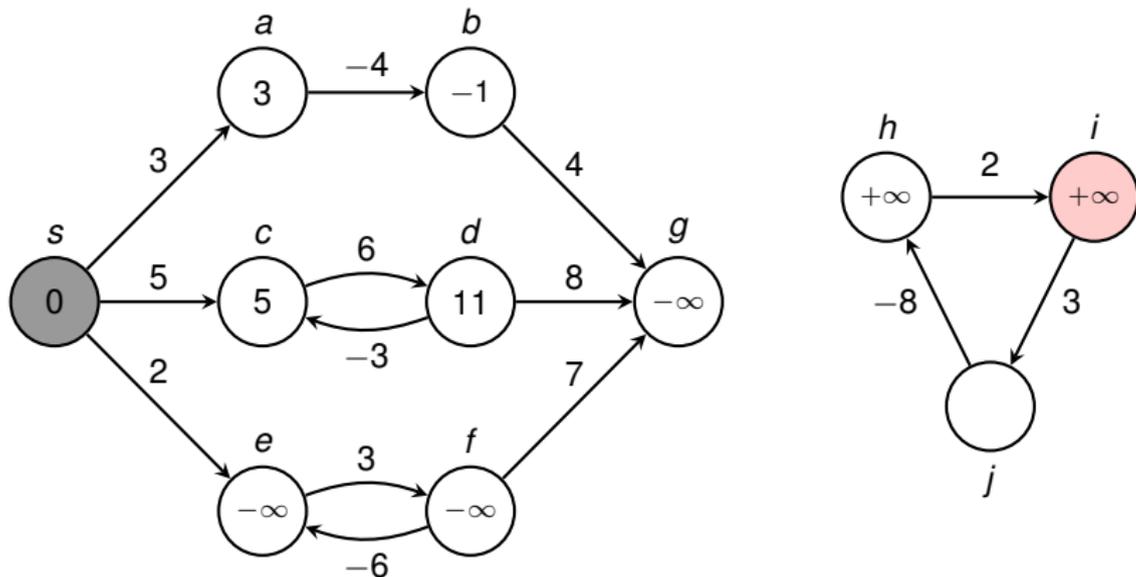
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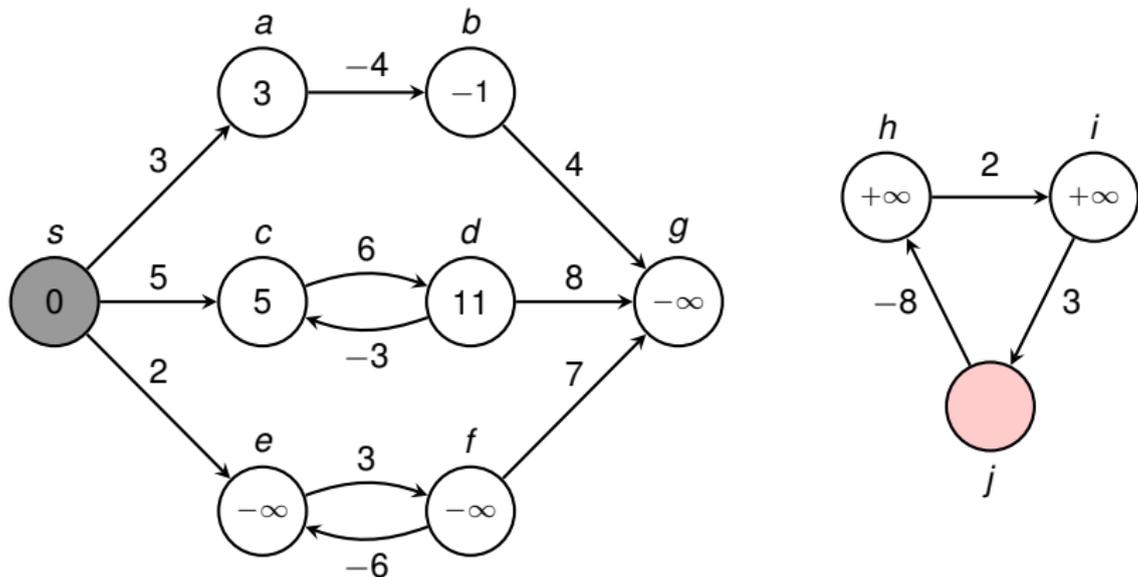
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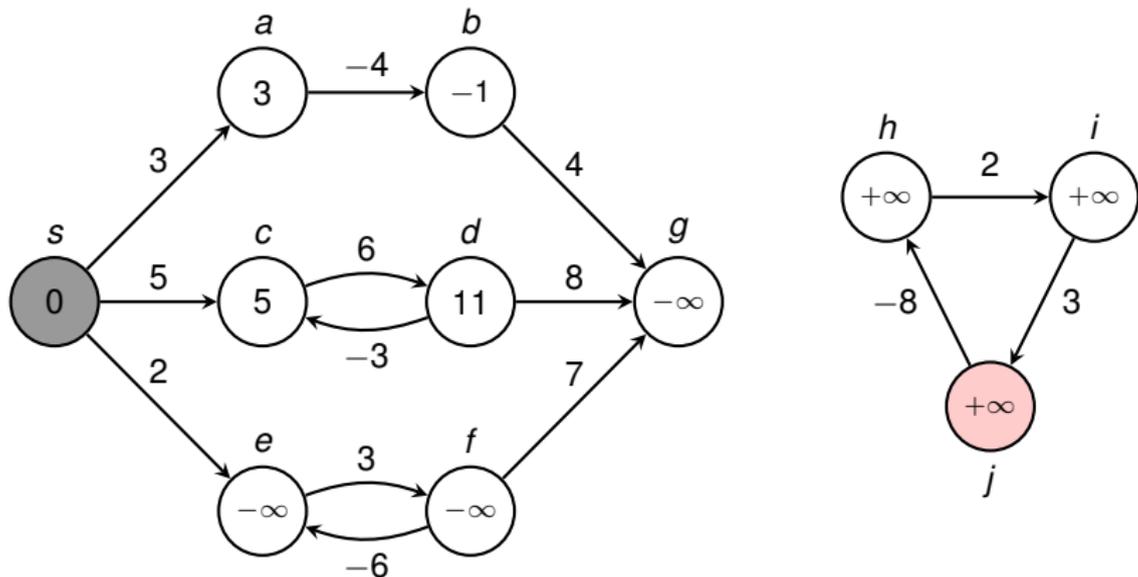
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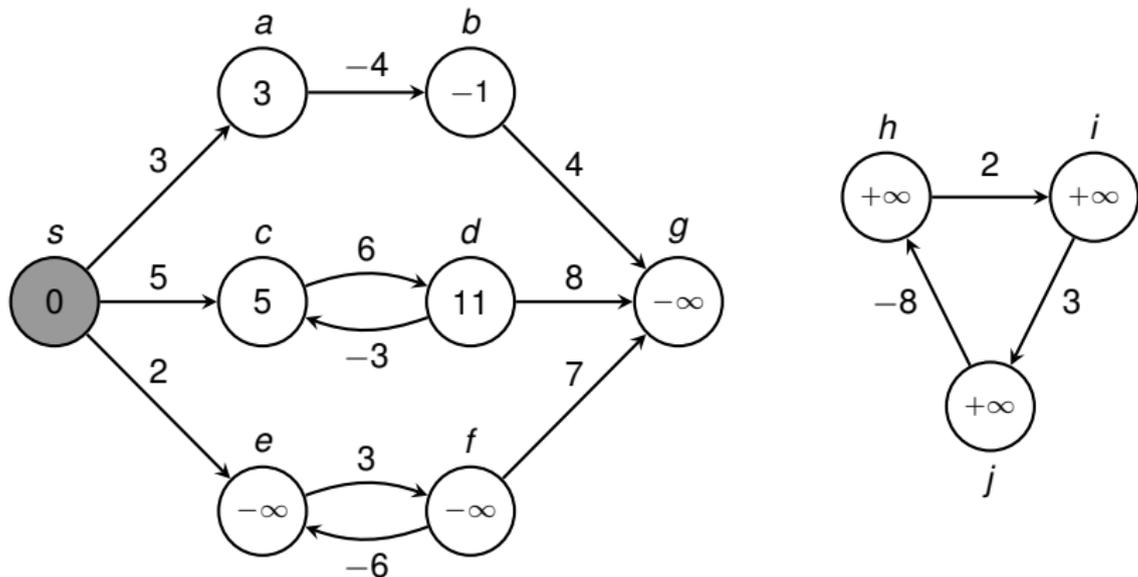
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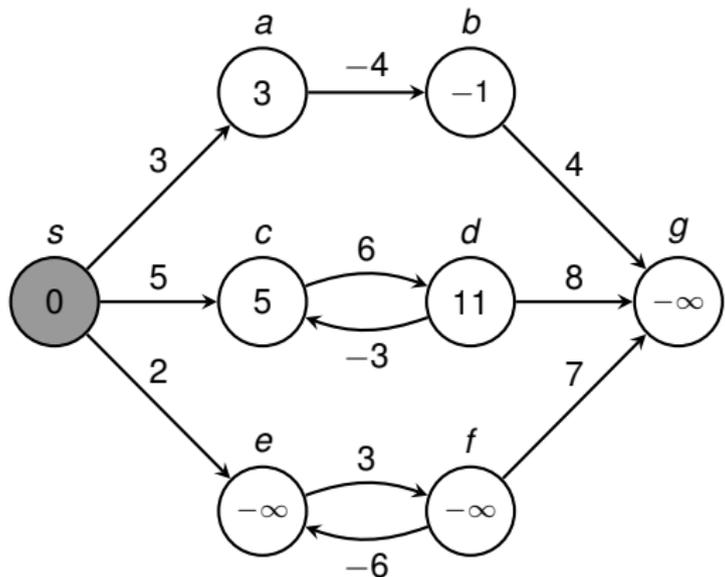
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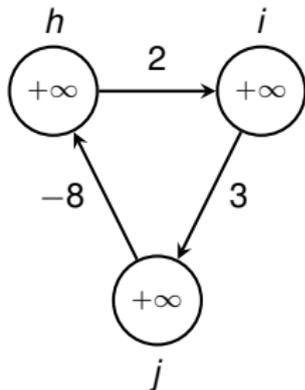
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Negative-Weight Cycle  
(reachable from  $s$ )



Negative-Weight Cycle  
(not reachable from  $s$ )



Introduction

Bellman-Ford Algorithm



## Relaxing Edges

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Definition

Fix the source vertex  $s \in V$

- $v.\delta$  is the length of the shortest path (distance) from  $s$  to  $v$
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Given estimates  $u.d$  and  $v.d$ , can we find a better path from  $v$  using the edge  $(u, v)$ ?



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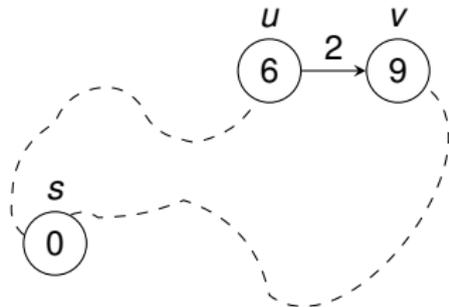
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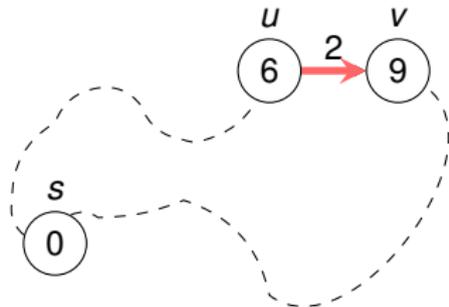
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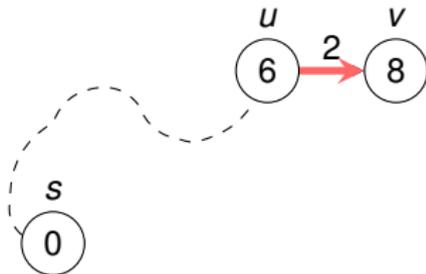
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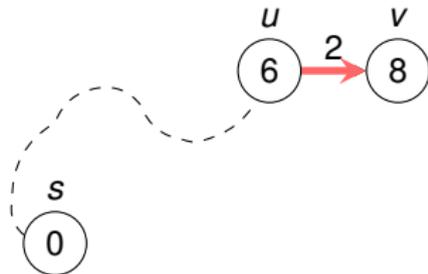
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After relaxing  $(u, v)$ , regardless of whether we found a shortcut:  
 $v.d \leq u.d + w(u, v)$



## Properties of Shortest Paths and Relaxations

### Toolkit

Triangle inequality (Lemma 24.10)

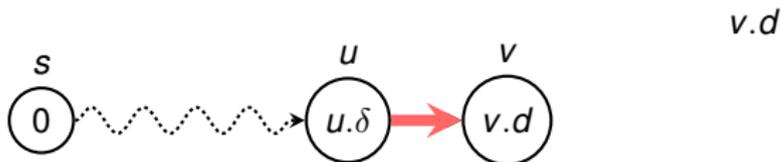
- For any edge  $(u, v) \in E$ , we have  $v.\delta \leq u.\delta + w(u, v)$

Upper-bound Property (Lemma 24.11)

- We always have  $v.d \geq v.\delta$  for all  $v \in V$ , and once  $v.d$  achieves the value  $v.\delta$ , it never changes.

Convergence Property (Lemma 24.14)

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



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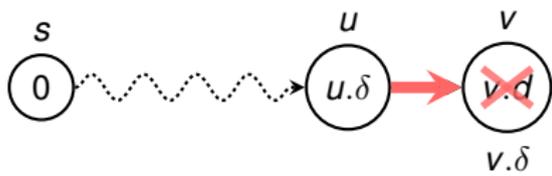
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Since  $v.d \geq v.\delta$ , we have  $v.d = v.\delta$ .  $\square$



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If  $p = (v_0, v_1, \dots, v_k)$  is a **shortest path** from  $s = v_0$  to  $v_k$ , and we **relax the edges of  $p$**  in the order  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$ , then  $v_k.d = v_k.\delta$  (regardless of the order of other relaxation steps).



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Proof:

- By induction on  $i$ ,  $0 \leq i \leq k$ :  
After the  $i$ th edge of  $p$  is relaxed, we have  $v_i.d = v_i.\delta$ .



## Path-Relaxation Property

### Path-Relaxation Property (Lemma 24.15)

If  $p = (v_0, v_1, \dots, v_k)$  is a **shortest path** from  $s = v_0$  to  $v_k$ , and we **relax the edges of  $p$**  in the order  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$ , then  $v_k.d = v_k.\delta$  (regardless of the order of other relaxation steps).

Proof:

- By induction on  $i$ ,  $0 \leq i \leq k$ :  
After the  $i$ th edge of  $p$  is relaxed, we have  $v_i.d = v_i.\delta$ .
- For  $i = 0$ , by the initialization  $s.d = s.\delta = 0$ .  
**Upper-bound Property**  $\Rightarrow$  the value of  $s.d$  never changes after that.



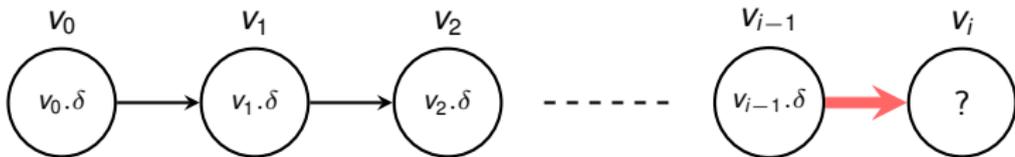
## Path-Relaxation Property

### Path-Relaxation Property (Lemma 24.15)

If  $p = (v_0, v_1, \dots, v_k)$  is a shortest path from  $s = v_0$  to  $v_k$ , and we relax the edges of  $p$  in the order  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$ , then  $v_k.d = v_k.\delta$  (regardless of the order of other relaxation steps).

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- By induction on  $i$ ,  $0 \leq i \leq k$ :  
After the  $i$ th edge of  $p$  is relaxed, we have  $v_i.d = v_i.\delta$ .
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Upper-bound Property  $\Rightarrow$  the value of  $s.d$  never changes after that.
- Inductive Step ( $i - 1 \rightarrow i$ ): Assume  $v_{i-1}.d = v_{i-1}.\delta$  and relax  $(v_{i-1}, v_i)$ .



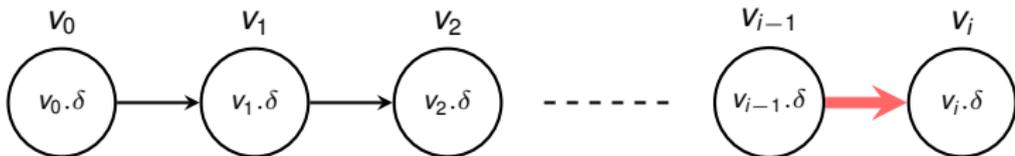
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**Convergence Property**  $\Rightarrow v_i.d = v_i.\delta$  (now and at all later steps) □



## Path-Relaxation Property

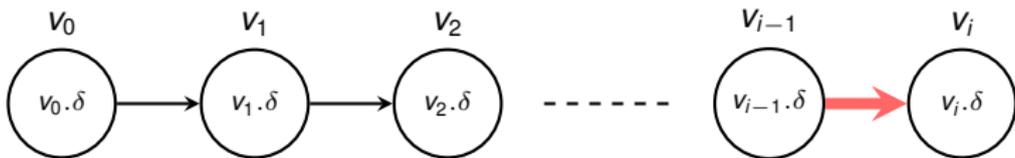
“Propagation”: By relaxing proper edges, set of vertices with  $v.\delta = v.d$  gets larger

### Path-Relaxation Property (Lemma 24.15)

If  $p = (v_0, v_1, \dots, v_k)$  is a **shortest path** from  $s = v_0$  to  $v_k$ , and we **relax the edges of  $p$**  in the order  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$ , then  $v_k.d = v_k.\delta$  (regardless of the order of other relaxation steps).

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Convergence Property  $\Rightarrow v_i.d = v_i.\delta$  (now and at all later steps) □



## The Bellman-Ford Algorithm

```
BELLMAN-FORD( $G, w, s$ )
0: assert( $s$  in  $G.vertices()$ )
1: for  $v$  in  $G.vertices()$ 
2:    $v.predecessor = None$ 
3:    $v.d = Infinity$ 
4:  $s.d = 0$ 
5:
6: repeat  $|V|-1$  times
7:   for  $e$  in  $G.edges()$ 
8:     Relax edge  $e=(u,v)$ : Check if  $u.d + w(u,v) < v.d$ 
9:       if  $e.start.d + e.weight < e.end.d$ :
10:          $e.end.d = e.start.d + e.weight$ 
11:          $e.end.predecessor = e.start$ 
12:
13: for  $e$  in  $G.edges()$ 
14:   if  $e.start.d + e.weight < e.end.d$ :
15:     return FALSE
16: return TRUE
```



# The Bellman-Ford Algorithm

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Time Complexity



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## Time Complexity

- A single call of line 9-11 costs  $\mathcal{O}(1)$



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### Time Complexity

- A single call of line 9-11 costs  $\mathcal{O}(1)$
- In each pass every edge is relaxed  $\Rightarrow \mathcal{O}(E)$  time per pass



## The Bellman-Ford Algorithm

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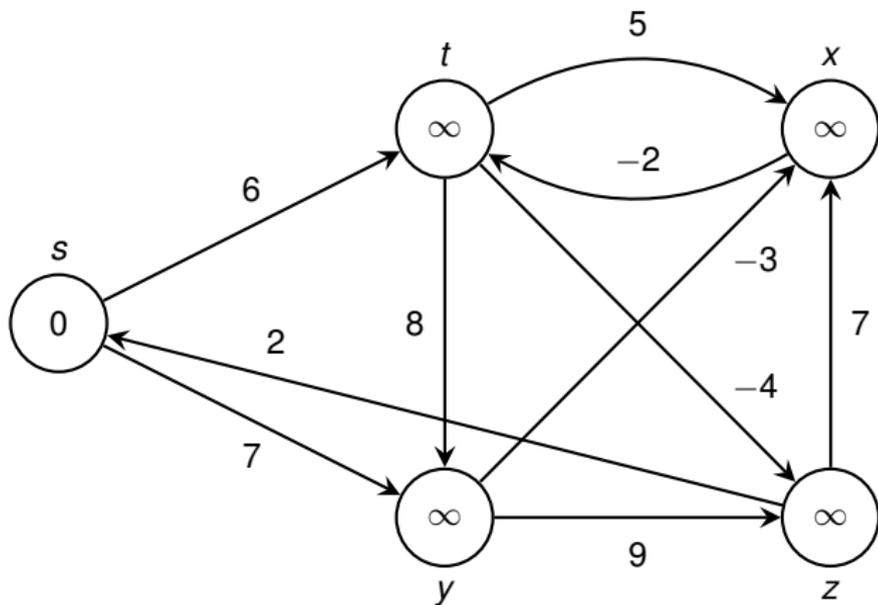
### Time Complexity

- A single call of line 9-11 costs  $\mathcal{O}(1)$
- In each pass every edge is relaxed  $\Rightarrow \mathcal{O}(E)$  time per pass
- Overall  $(V - 1) + 1 = V$  passes  $\Rightarrow \mathcal{O}(V \cdot E)$  time



## Execution of Bellman-Ford (Figure 24.4)

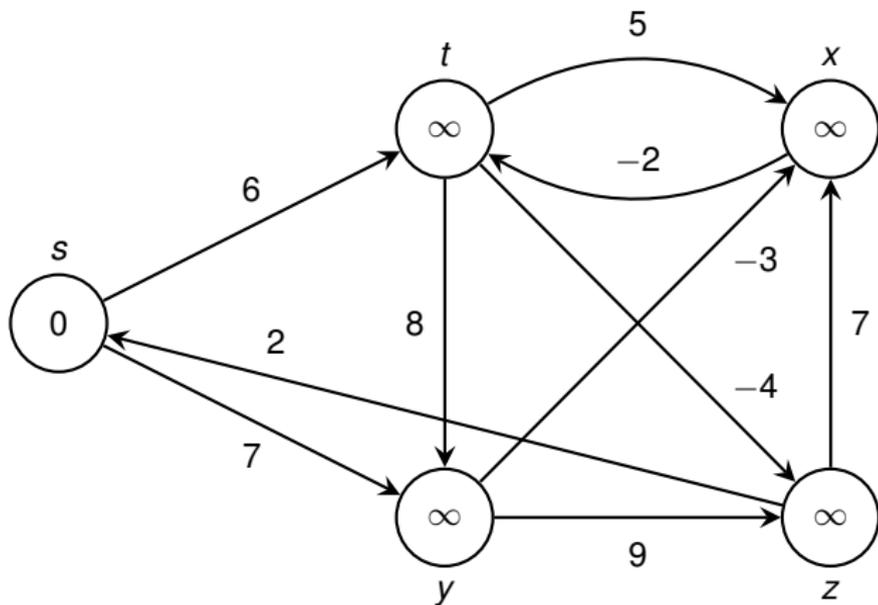
Pass: 1



## Execution of Bellman-Ford (Figure 24.4)

Pass: 1

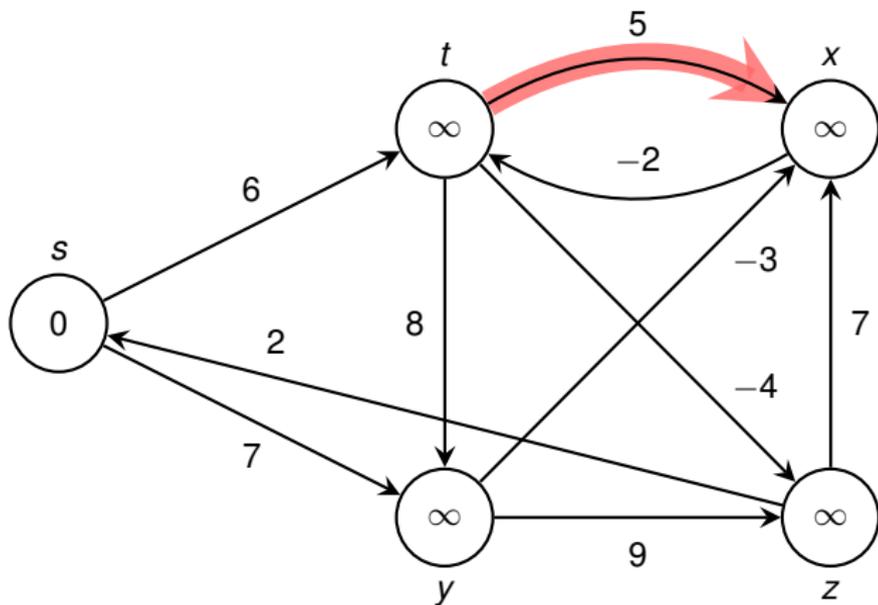
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 1

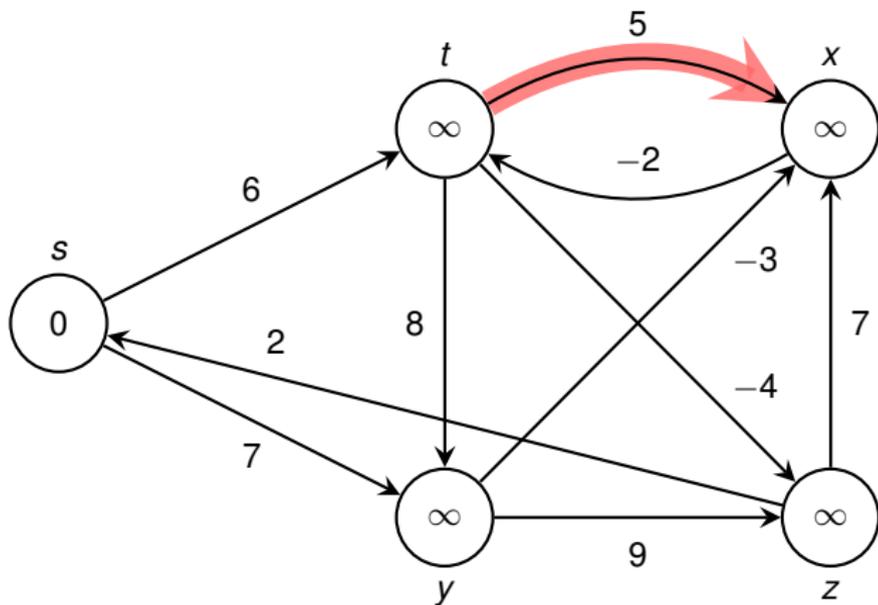
Relaxation Order:  $(t,x)$ ,  $(t,y)$ ,  $(t,z)$ ,  $(x,t)$ ,  $(y,x)$ ,  $(y,z)$ ,  $(z,x)$ ,  $(z,s)$ ,  $(s,t)$ ,  $(s,y)$



## Execution of Bellman-Ford (Figure 24.4)

Pass: 1

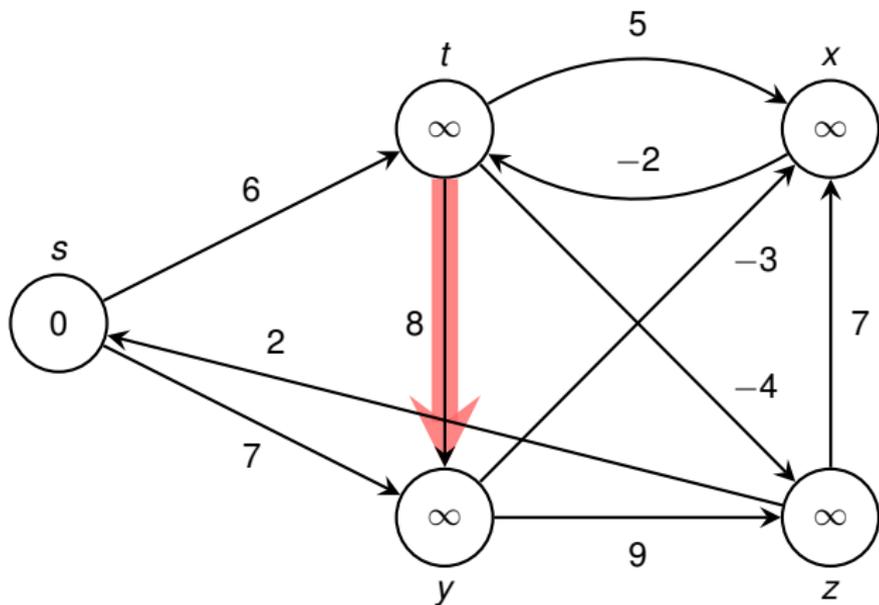
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 1

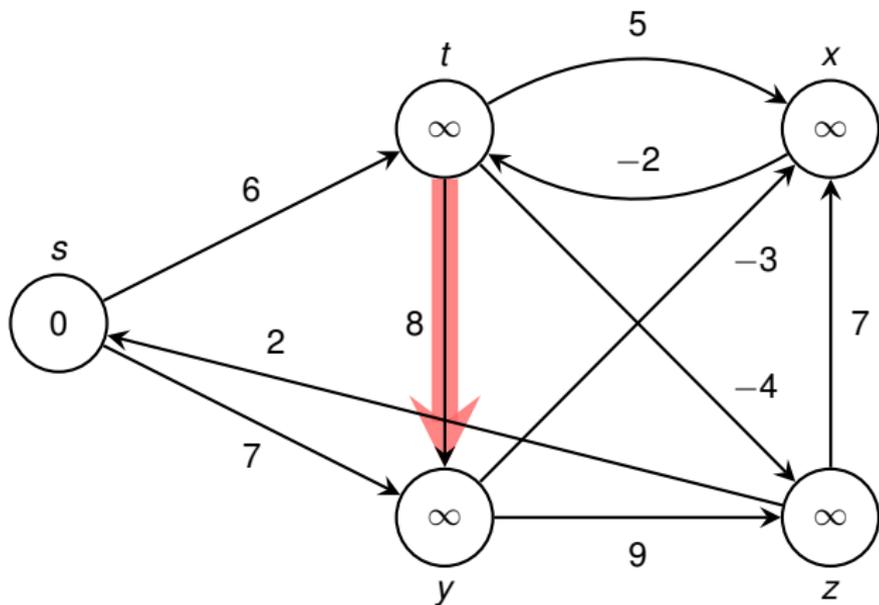
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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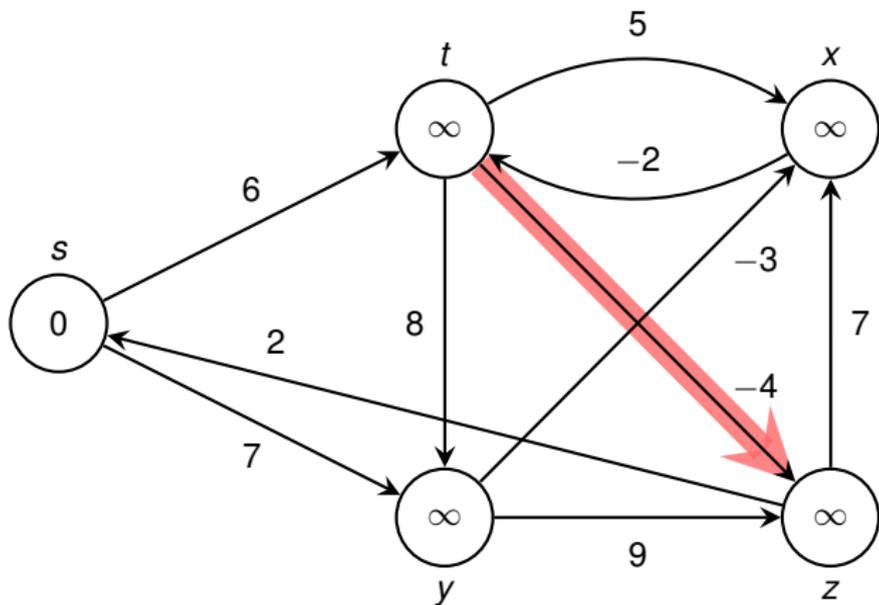
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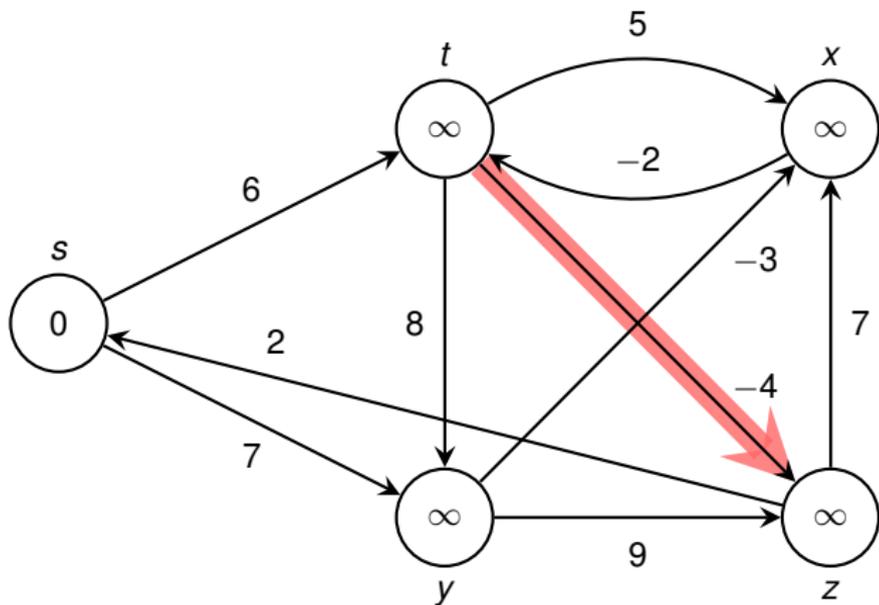
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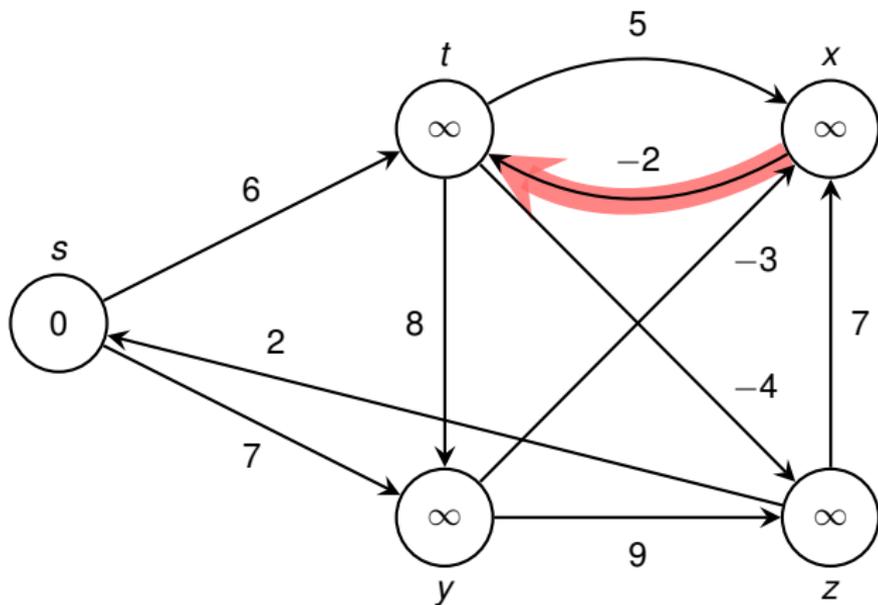
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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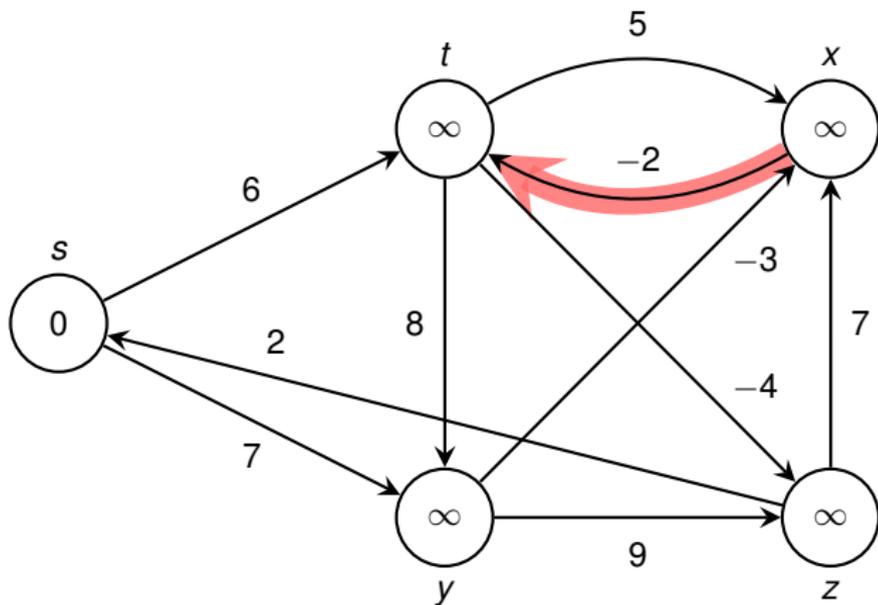
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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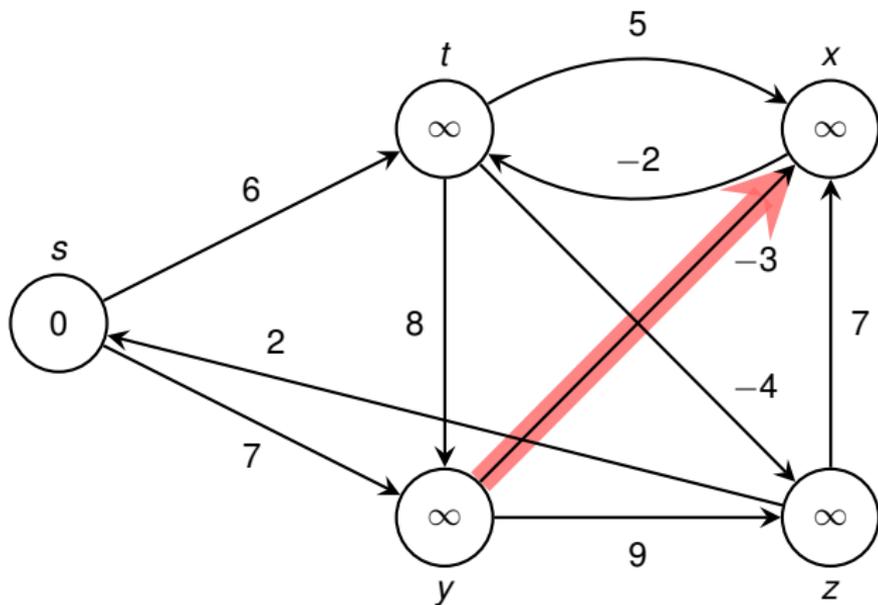
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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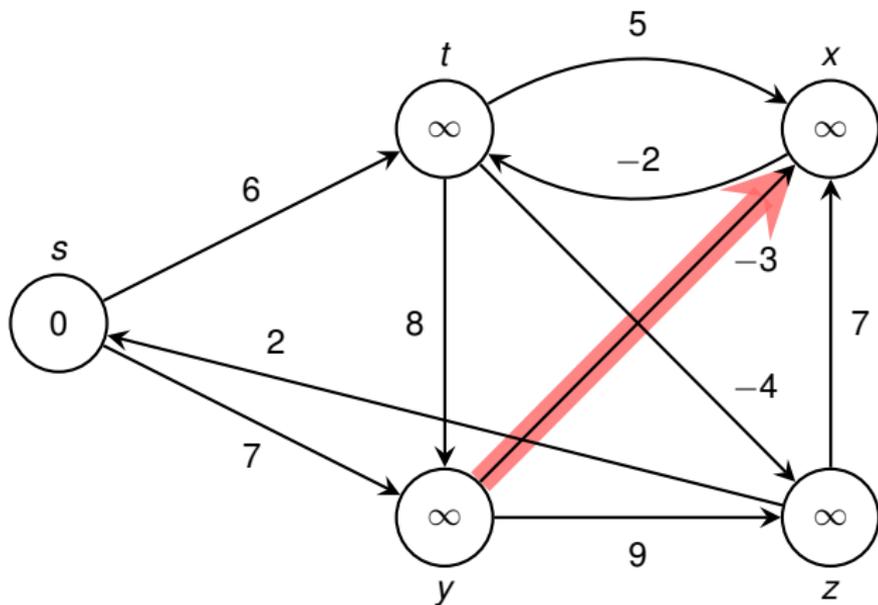
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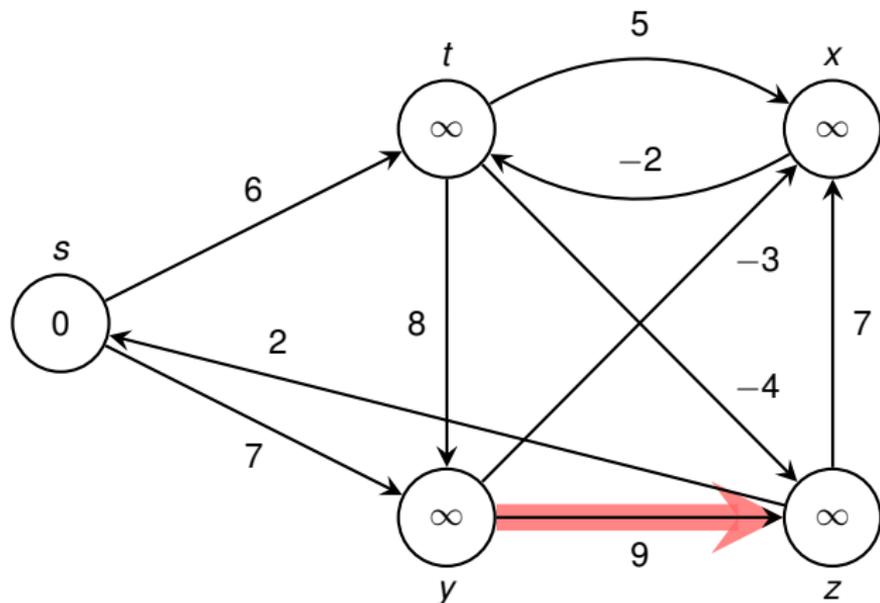
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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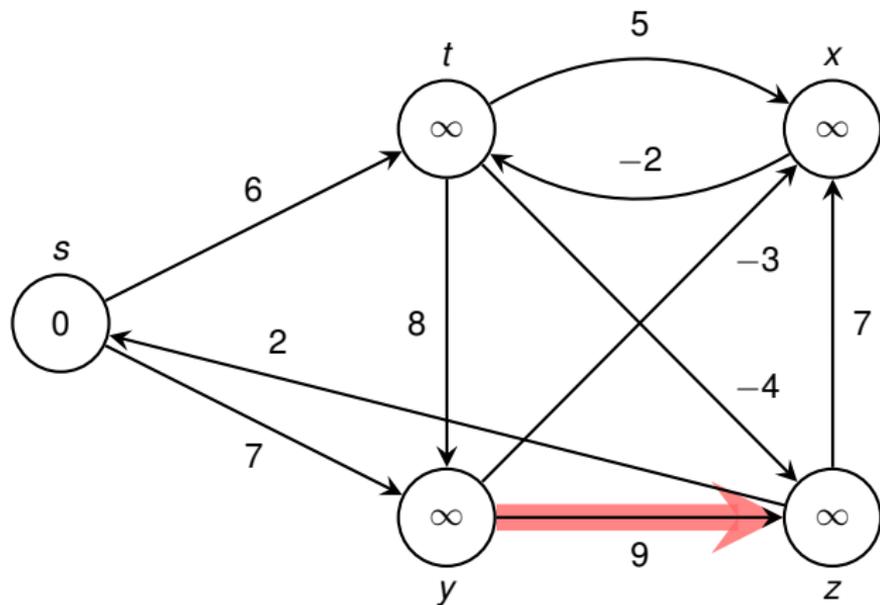
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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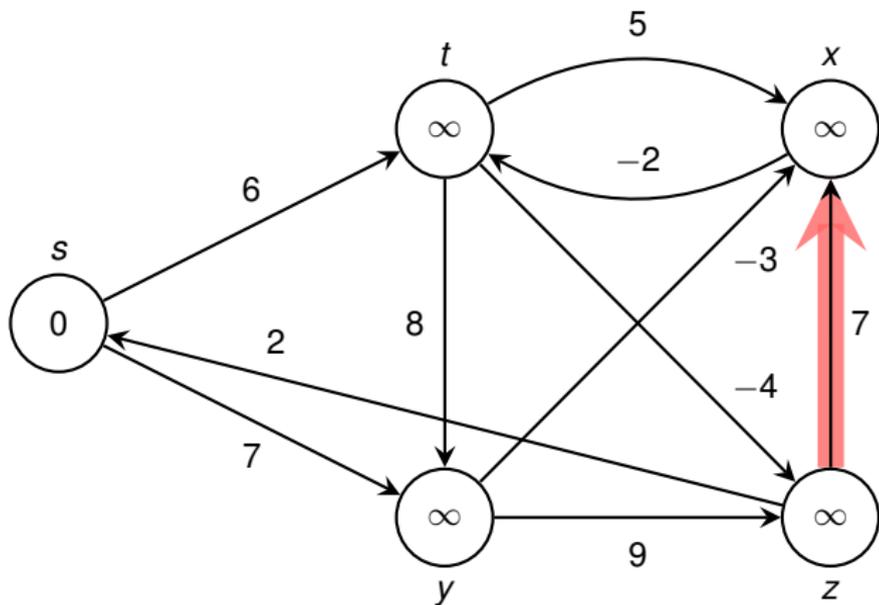
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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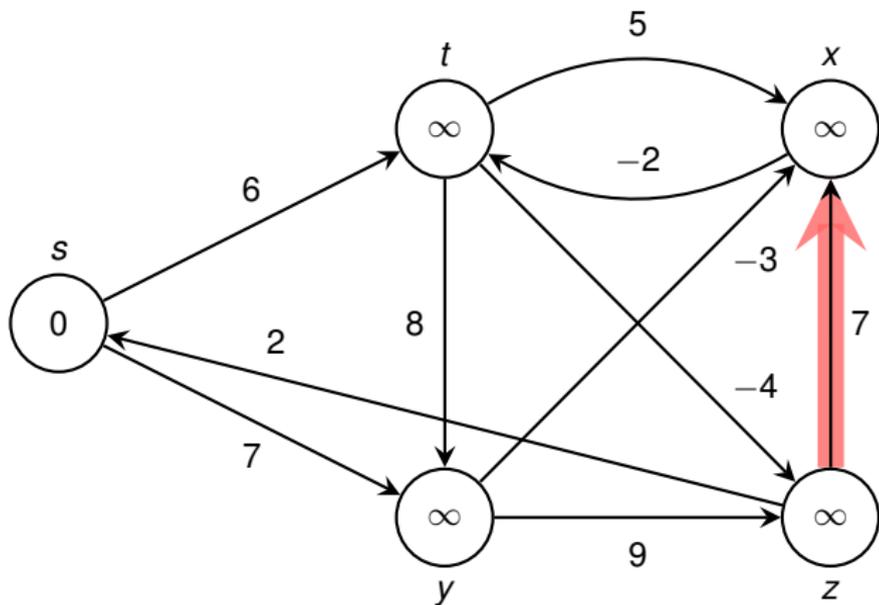
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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Pass: 1

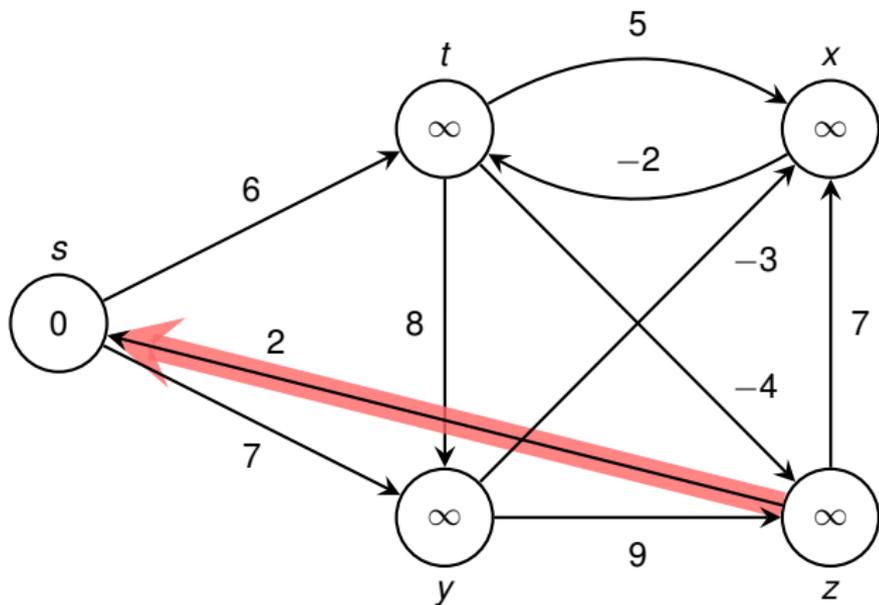
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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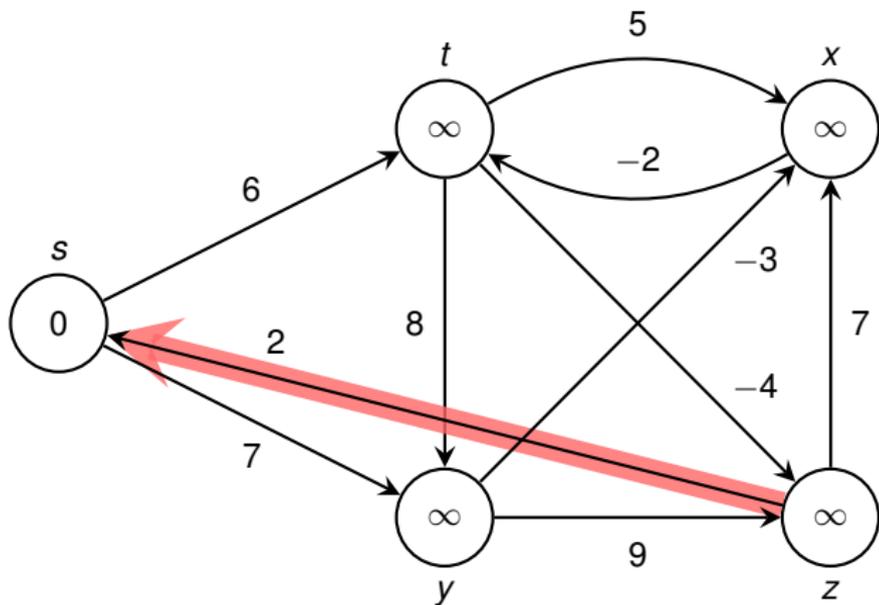
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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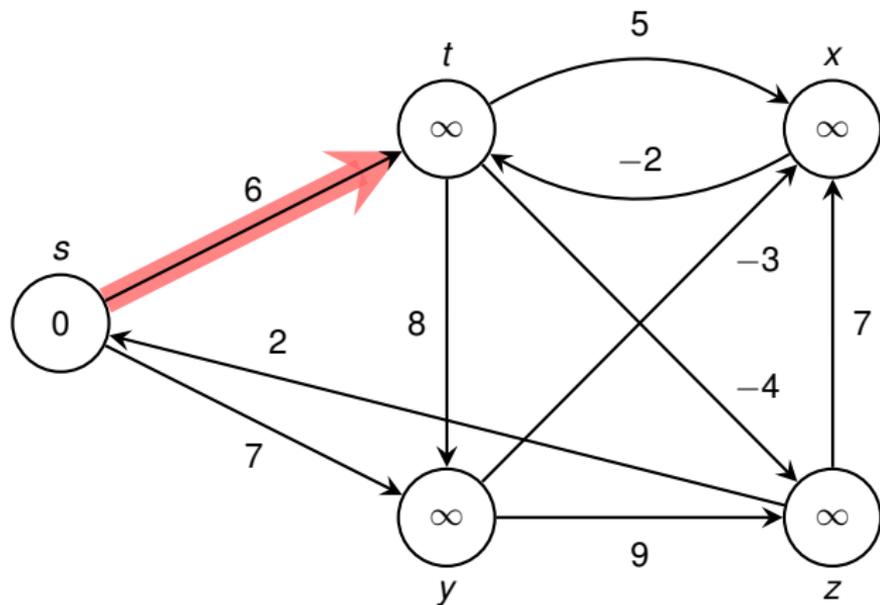
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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Pass: 1

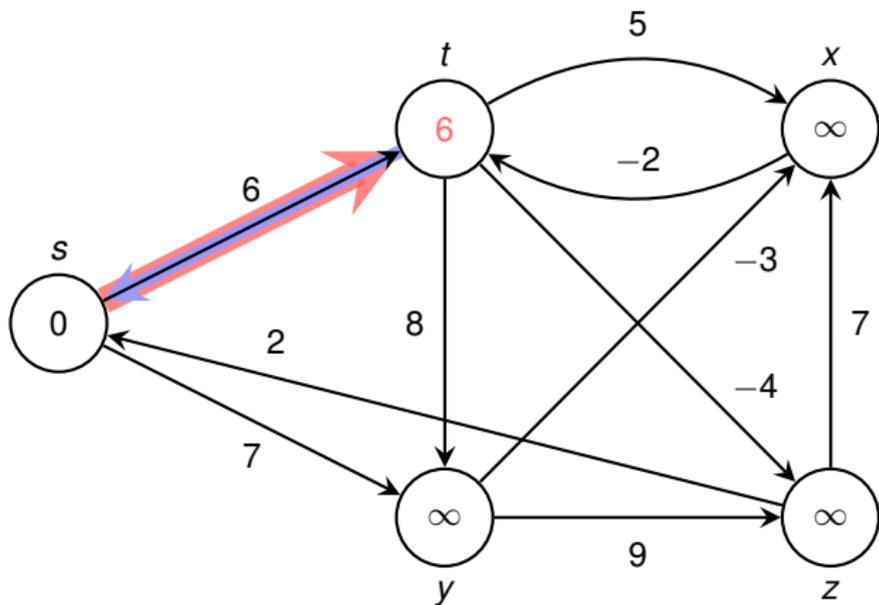
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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Pass: 1

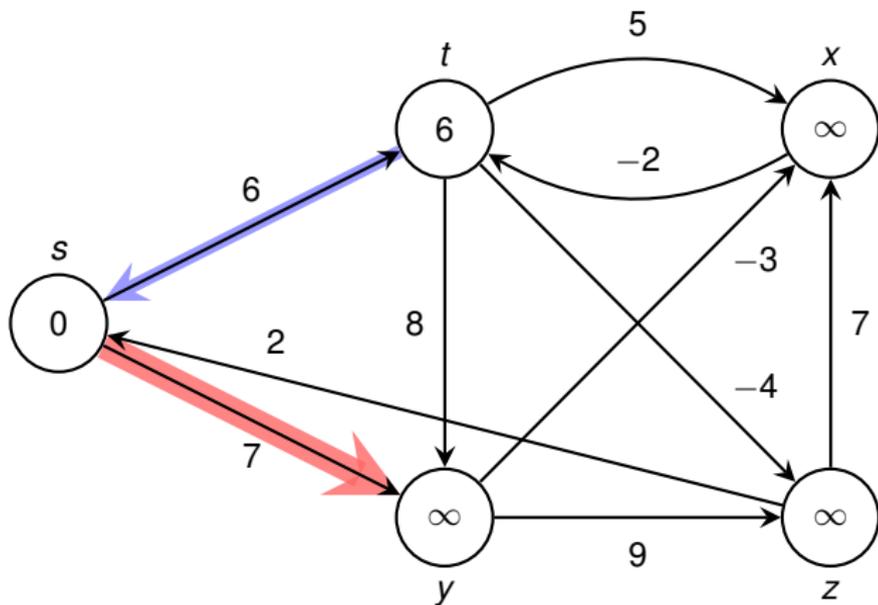
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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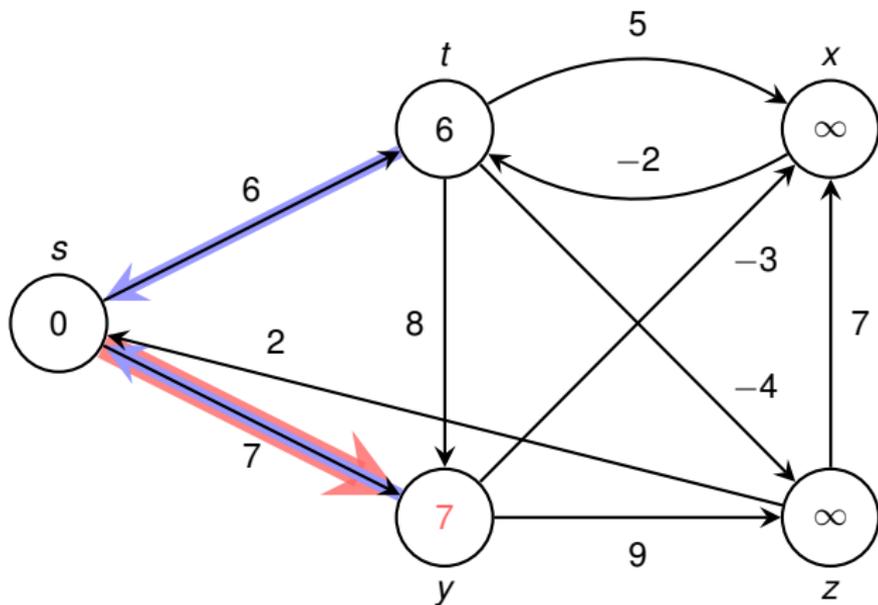
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 1

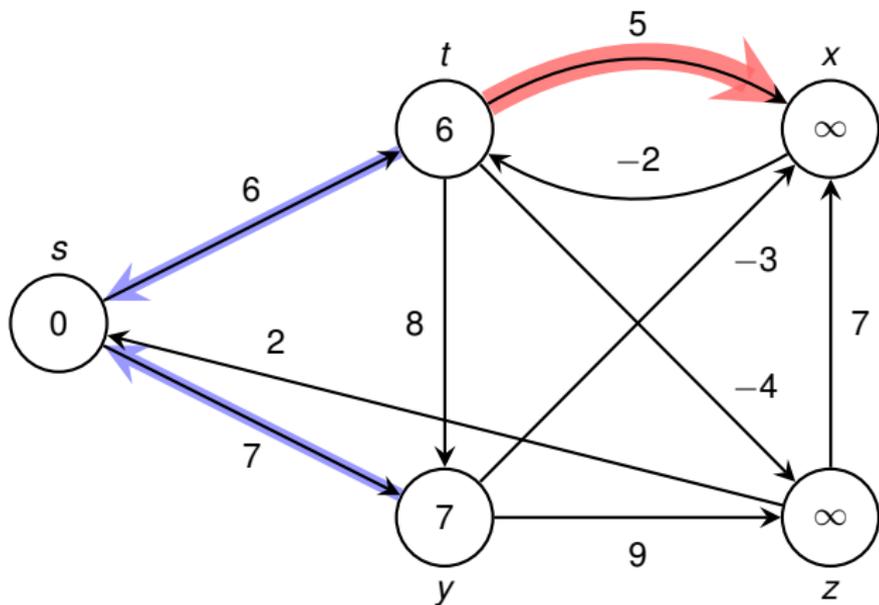
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

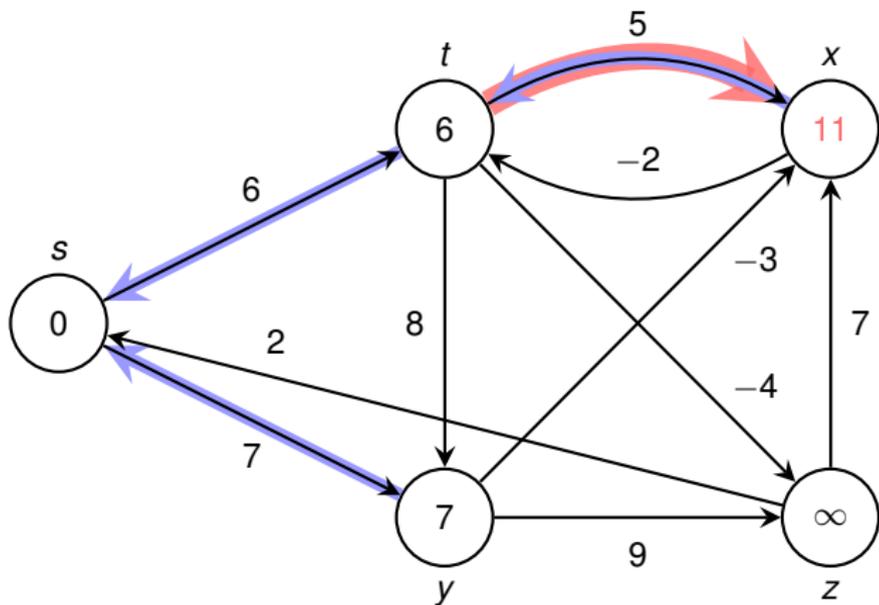
Relaxation Order:  $(t,x)$ ,  $(t,y)$ ,  $(t,z)$ ,  $(x,t)$ ,  $(y,x)$ ,  $(y,z)$ ,  $(z,x)$ ,  $(z,s)$ ,  $(s,t)$ ,  $(s,y)$



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

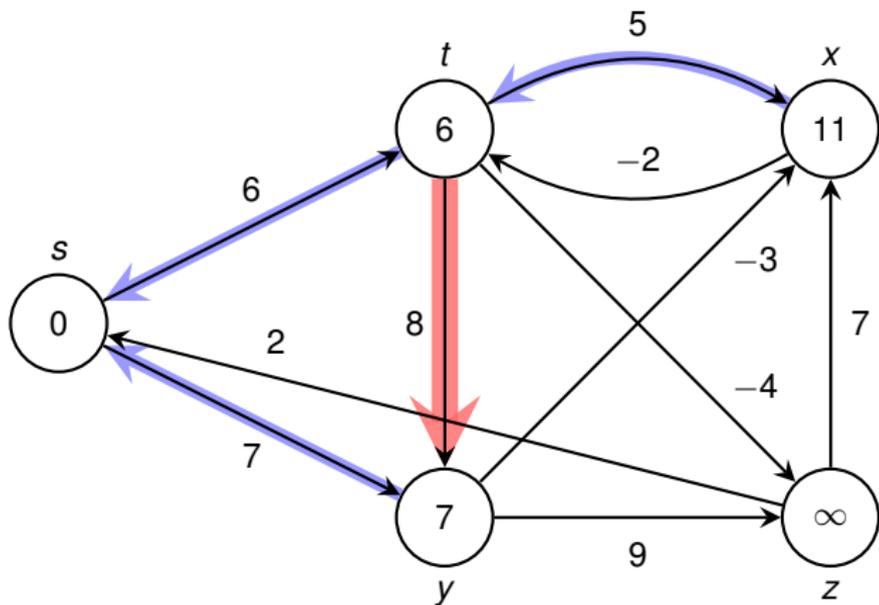
Relaxation Order:  $(t,x)$ ,  $(t,y)$ ,  $(t,z)$ ,  $(x,t)$ ,  $(y,x)$ ,  $(y,z)$ ,  $(z,x)$ ,  $(z,s)$ ,  $(s,t)$ ,  $(s,y)$



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

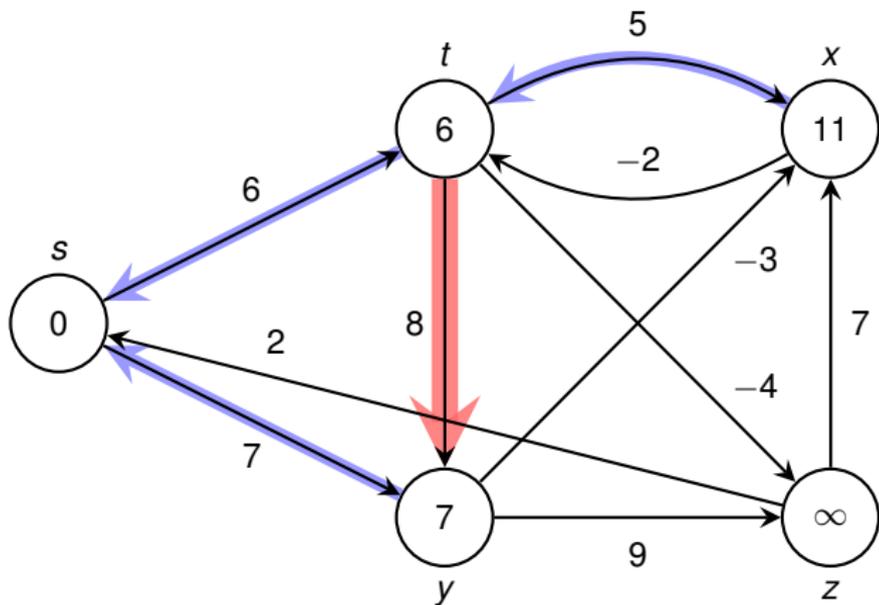
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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Pass: 2

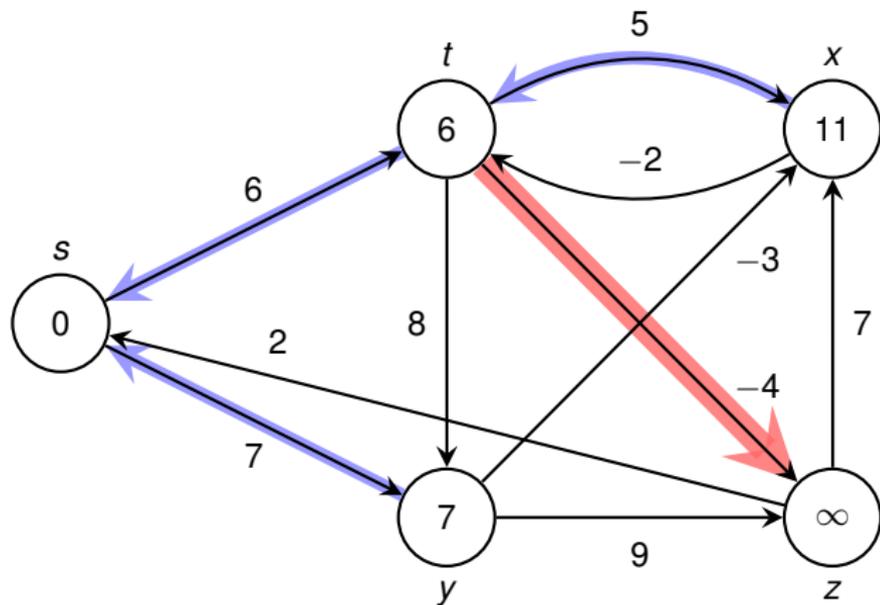
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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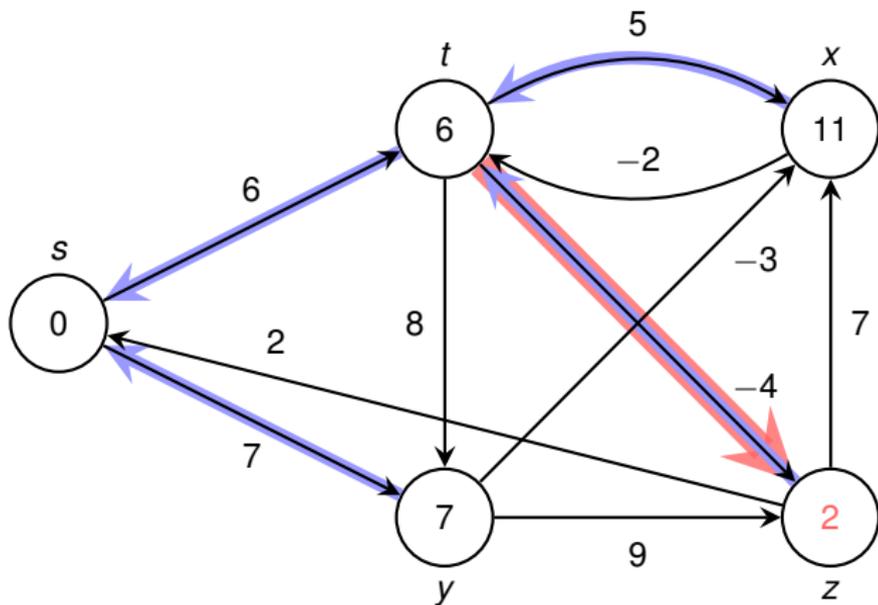
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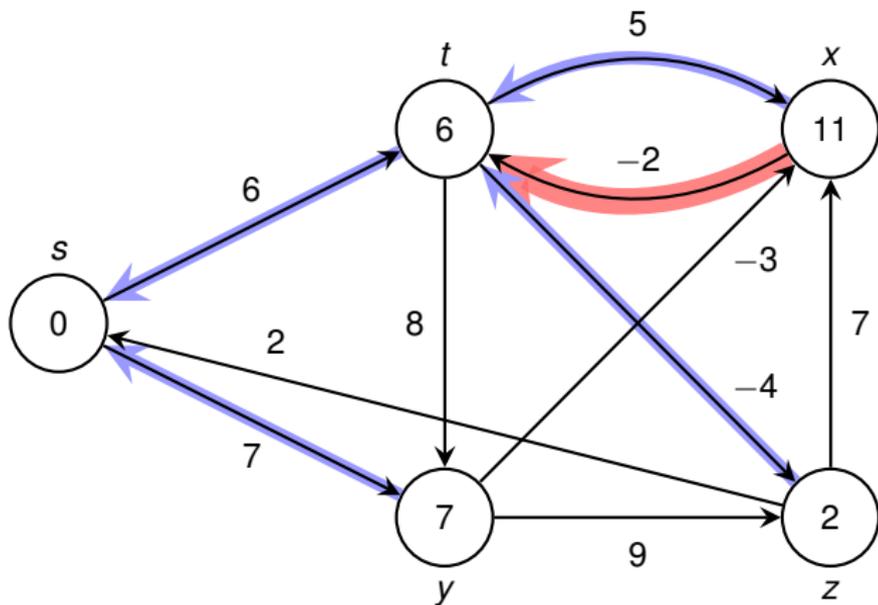
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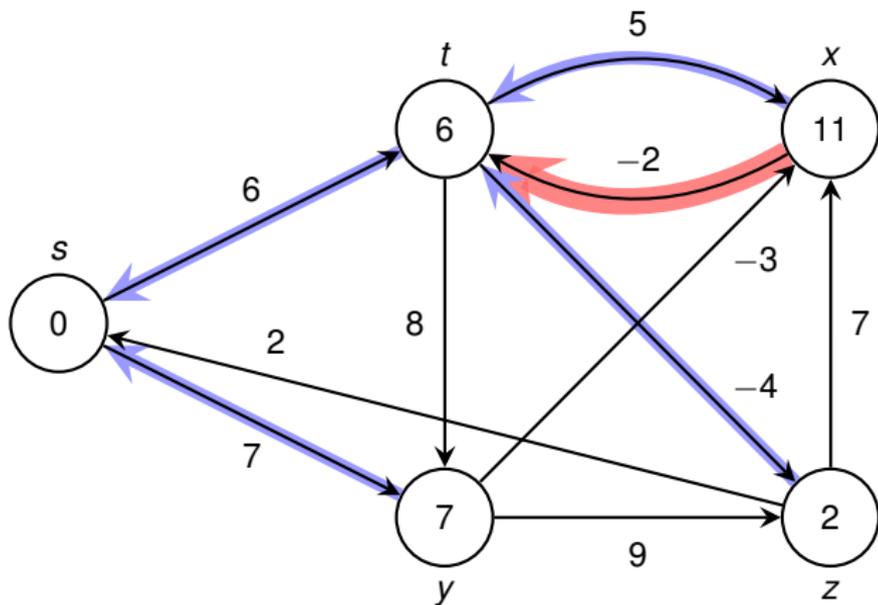
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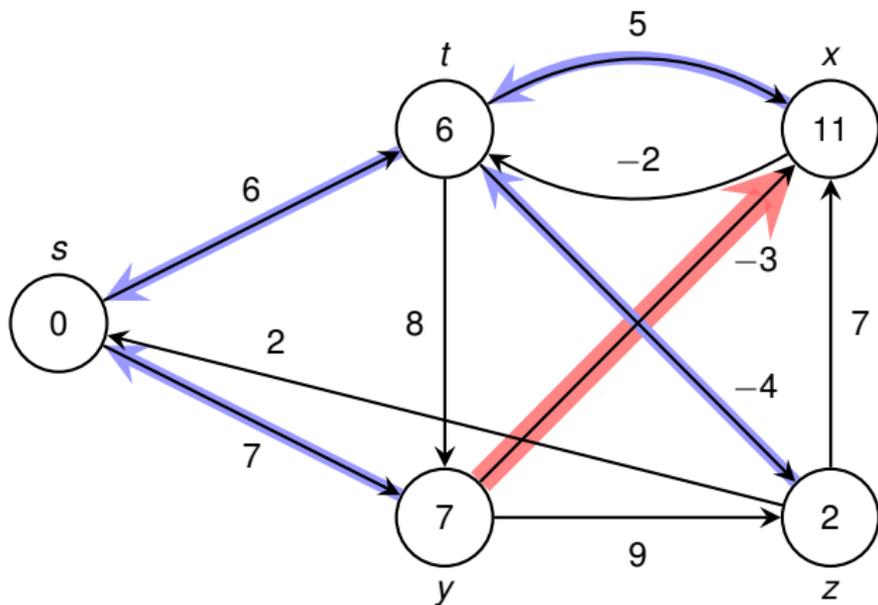
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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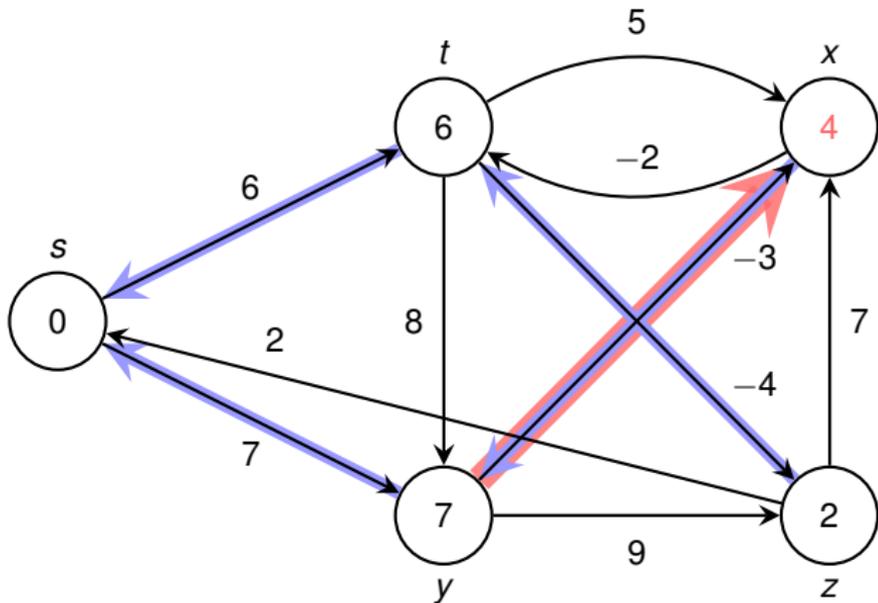
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

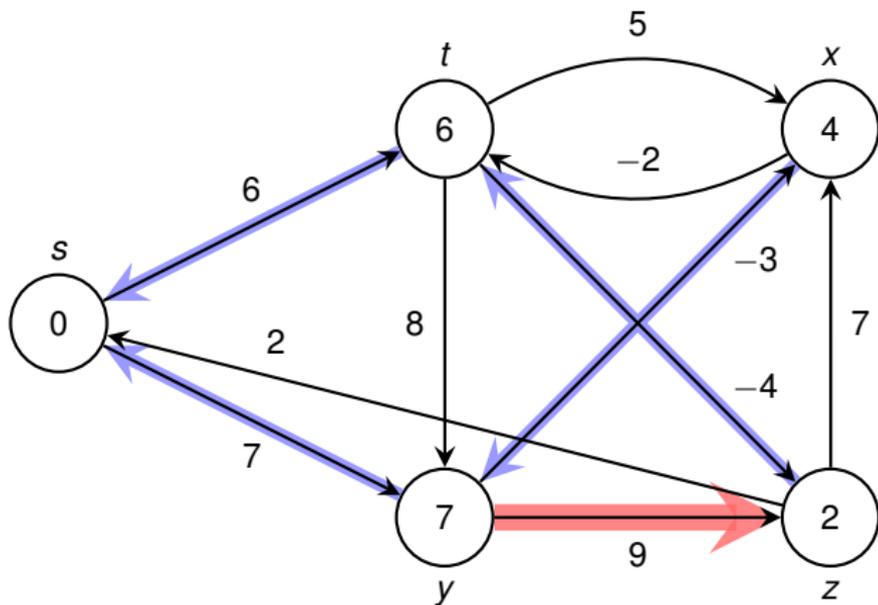
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

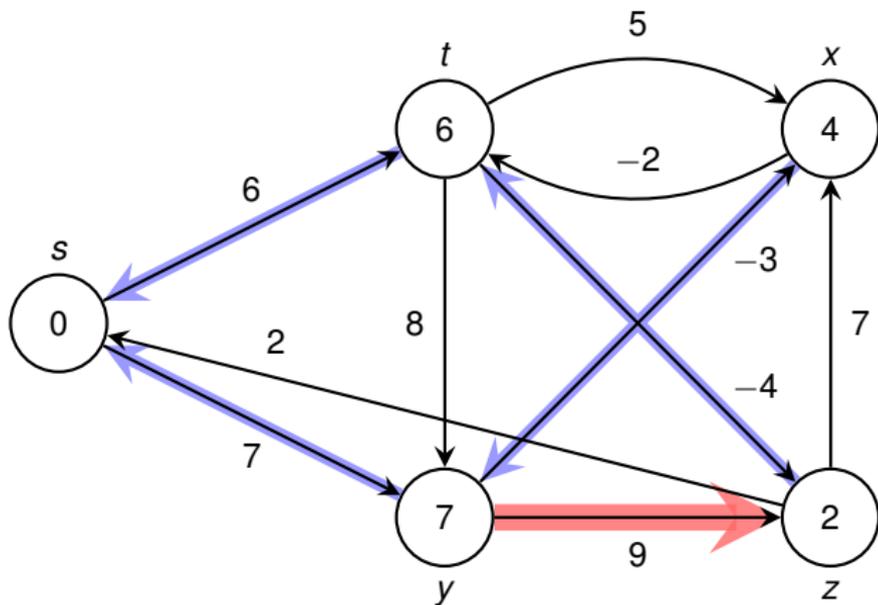
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

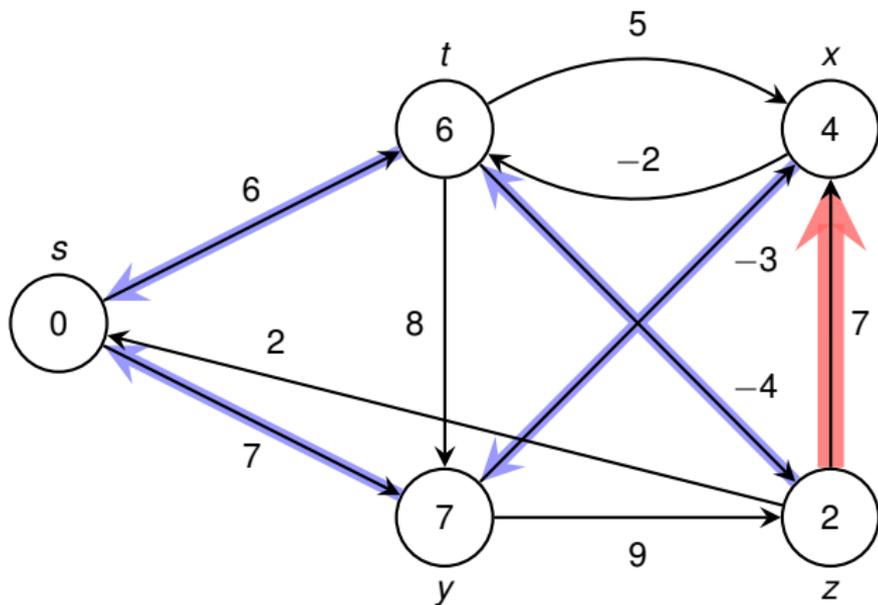
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

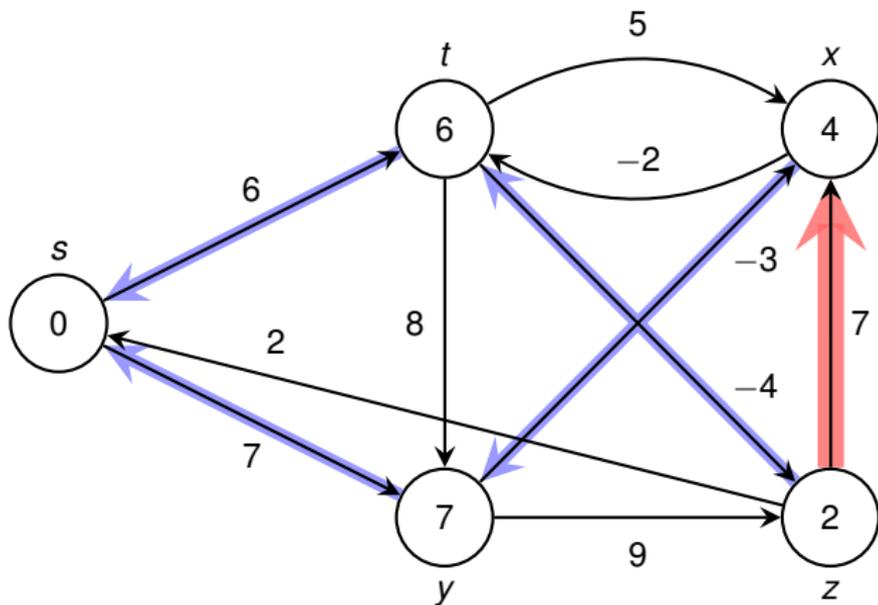
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

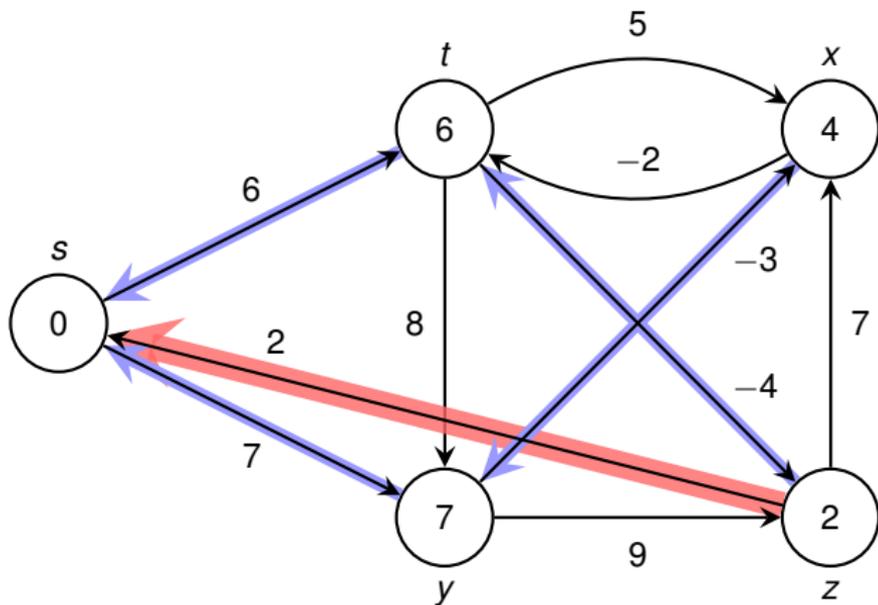
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

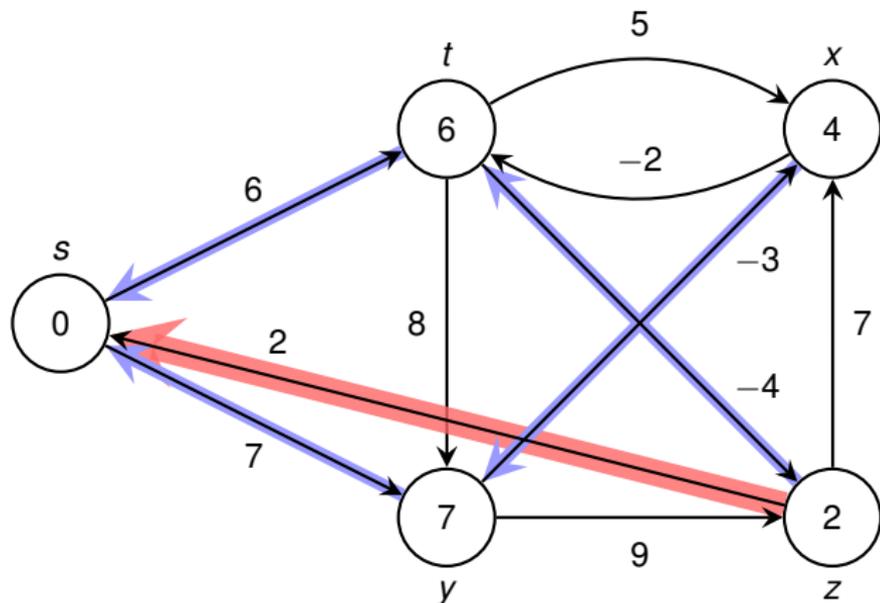
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

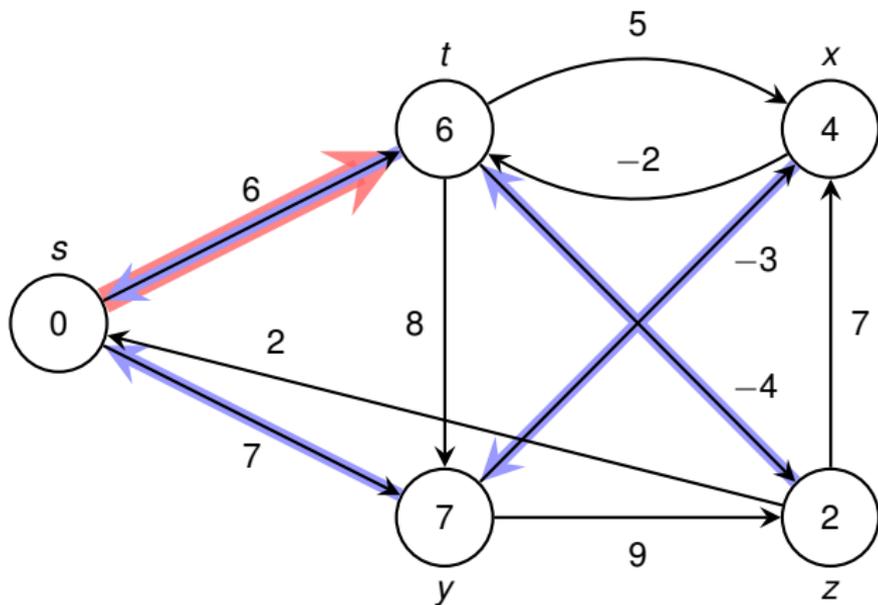
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

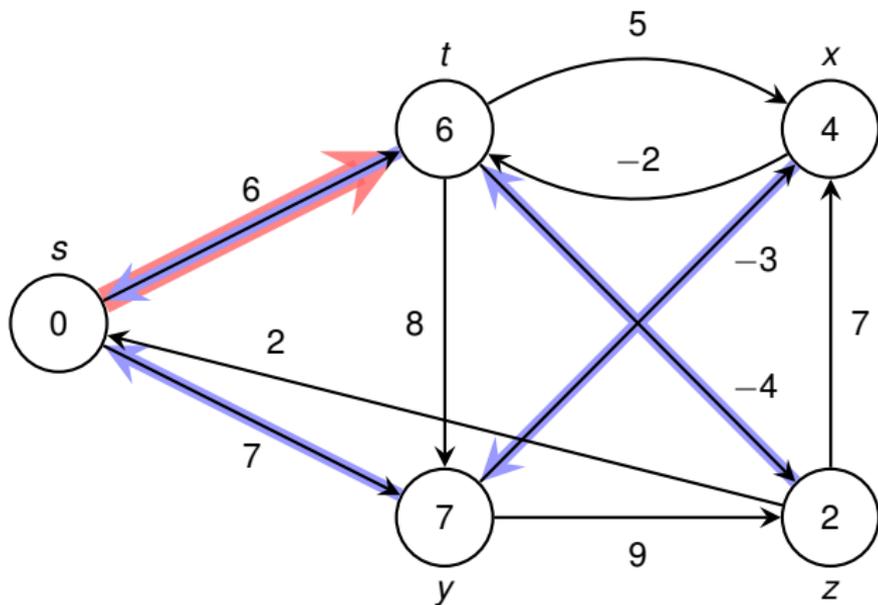
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

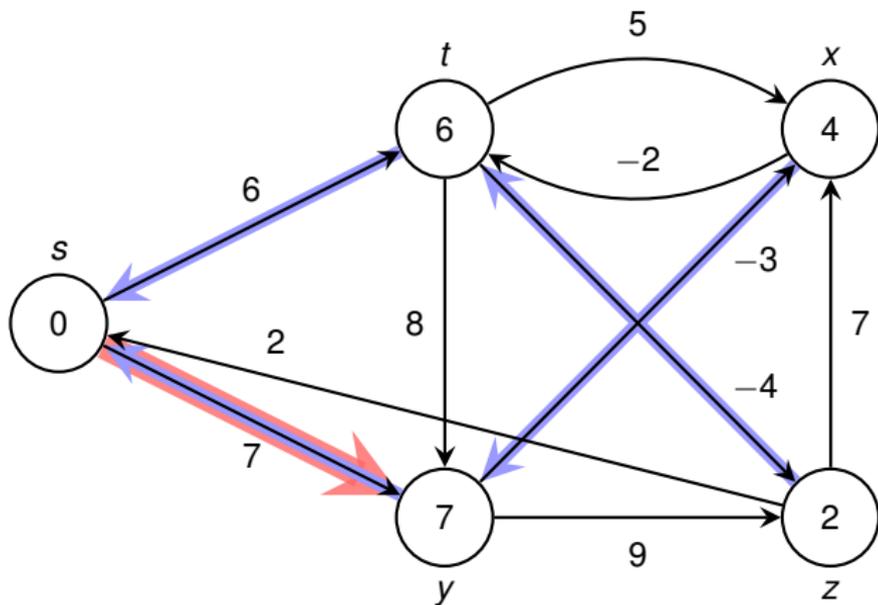
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

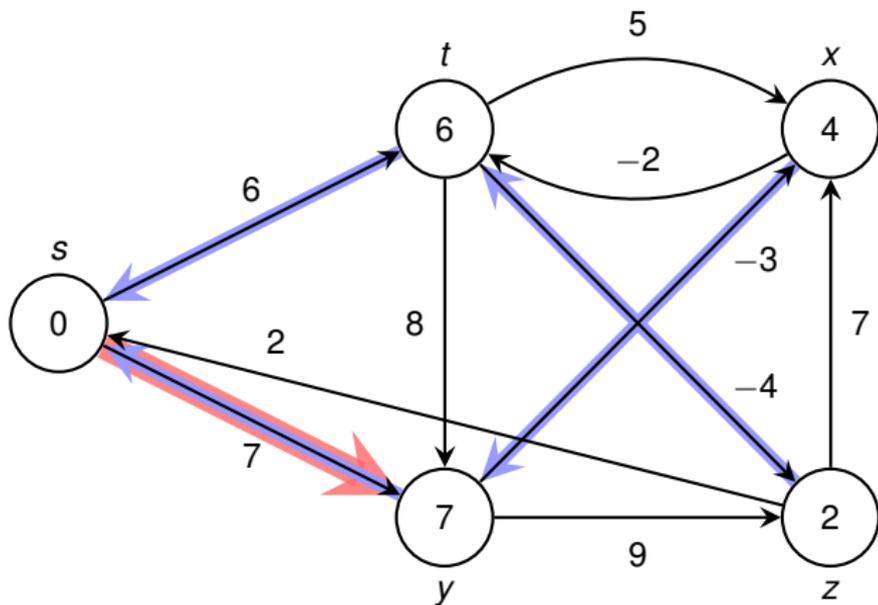
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 2

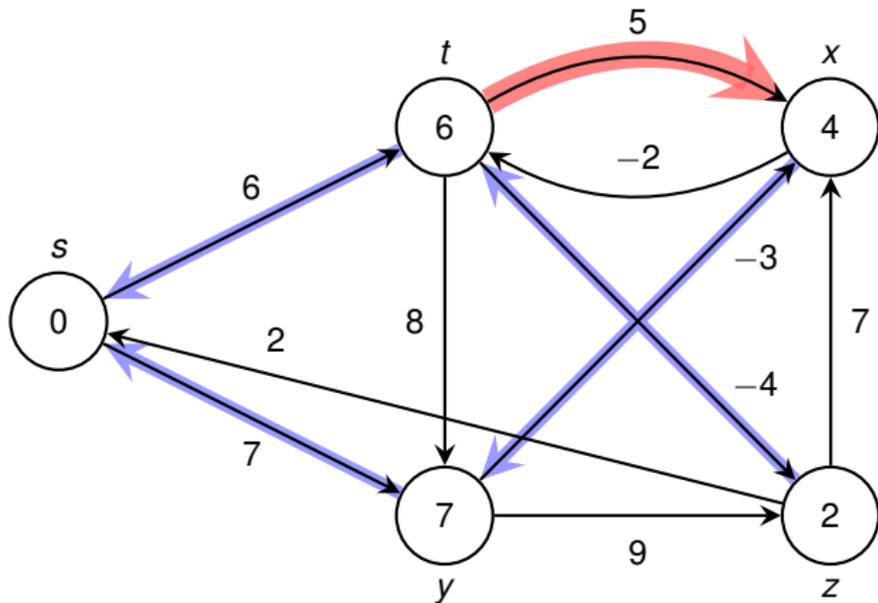
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

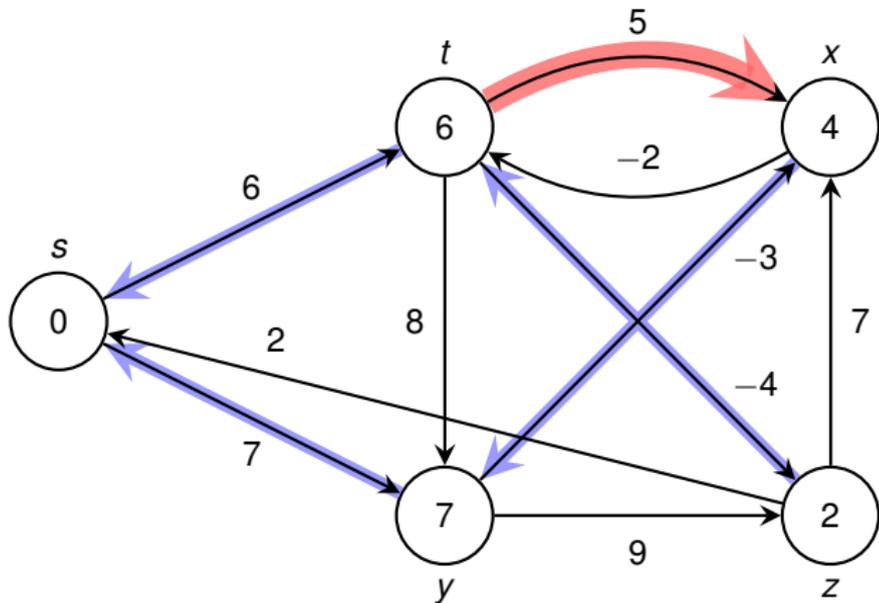
Relaxation Order:  $(t,x)$ ,  $(t,y)$ ,  $(t,z)$ ,  $(x,t)$ ,  $(y,x)$ ,  $(y,z)$ ,  $(z,x)$ ,  $(z,s)$ ,  $(s,t)$ ,  $(s,y)$



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

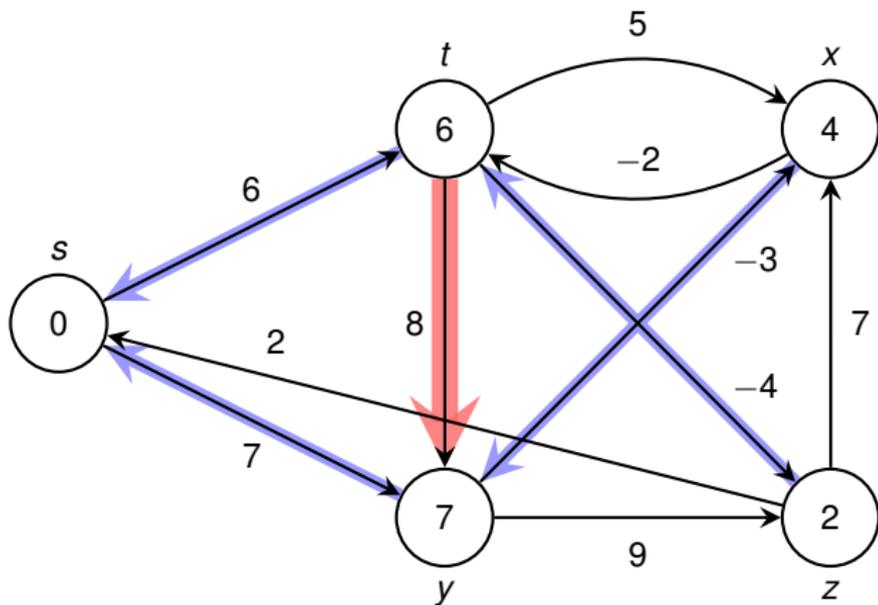
Relaxation Order:  $(t,x)$ ,  $(t,y)$ ,  $(t,z)$ ,  $(x,t)$ ,  $(y,x)$ ,  $(y,z)$ ,  $(z,x)$ ,  $(z,s)$ ,  $(s,t)$ ,  $(s,y)$



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

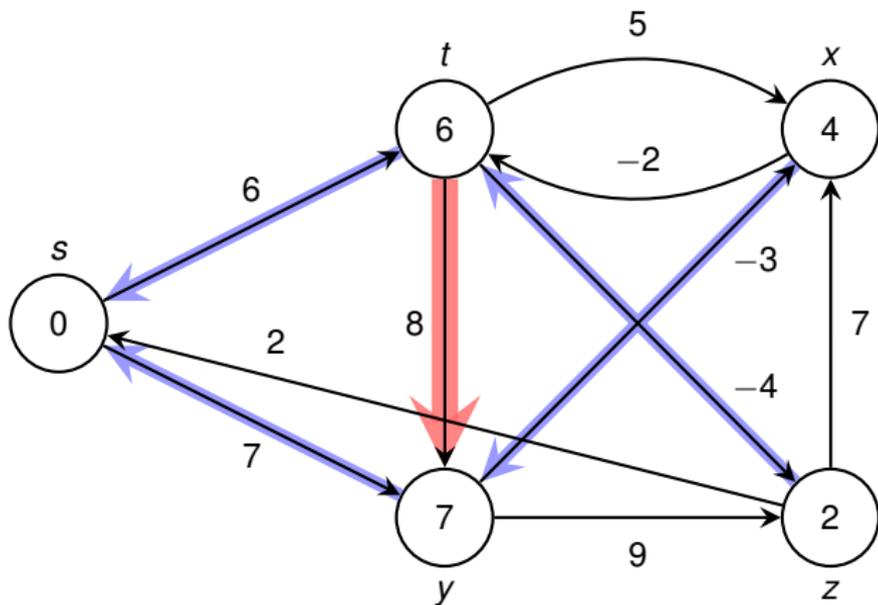
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

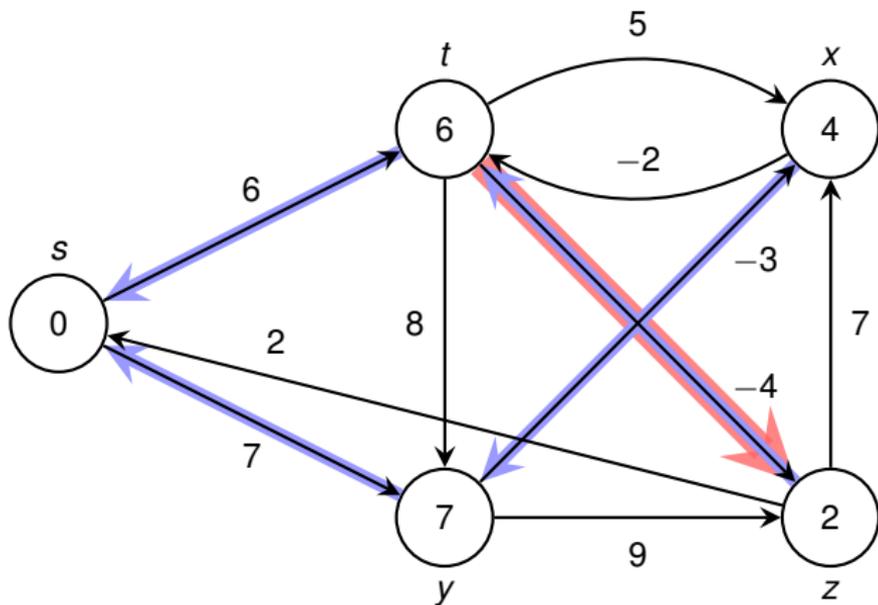
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

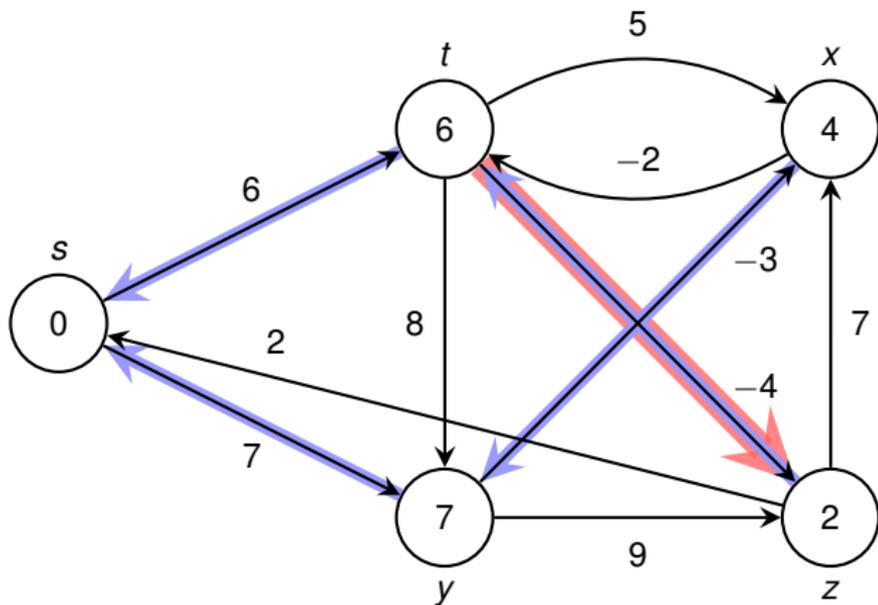
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

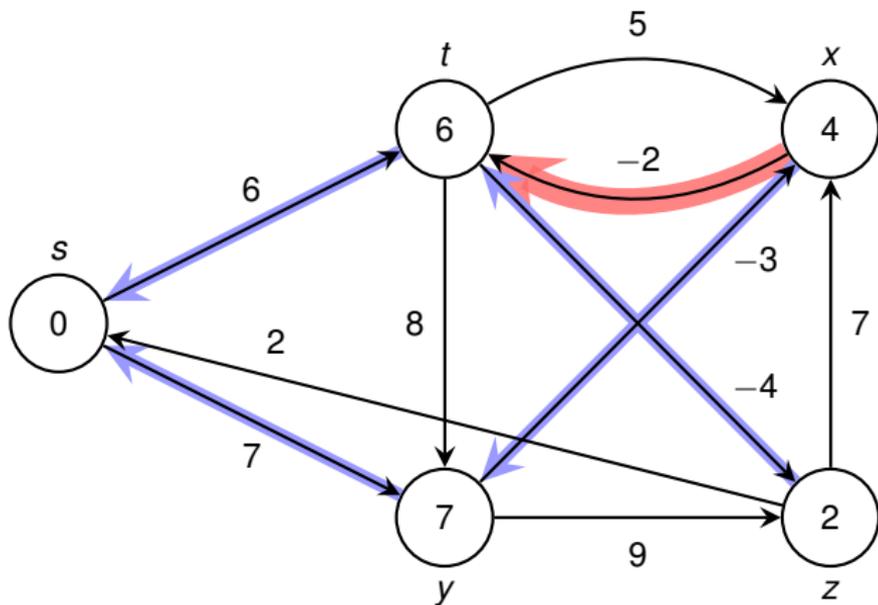
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

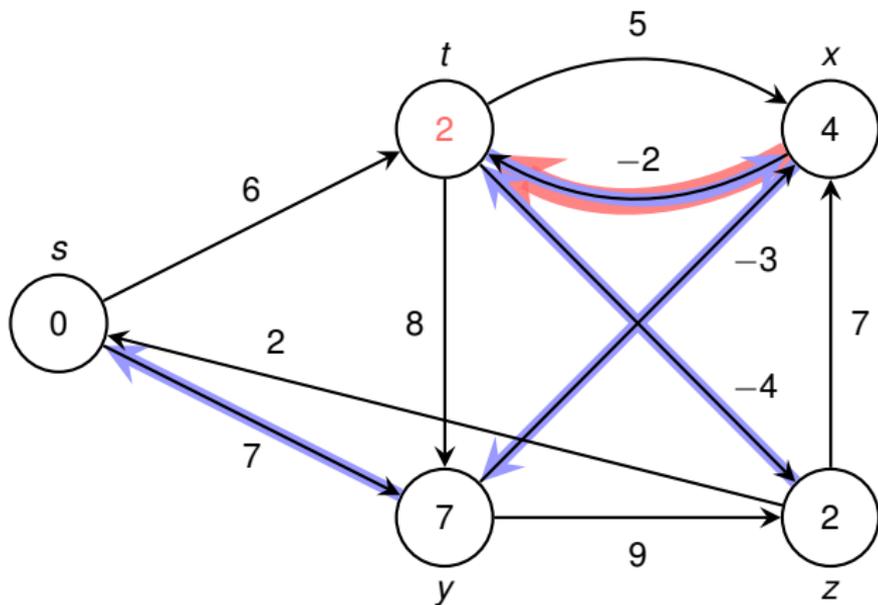
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

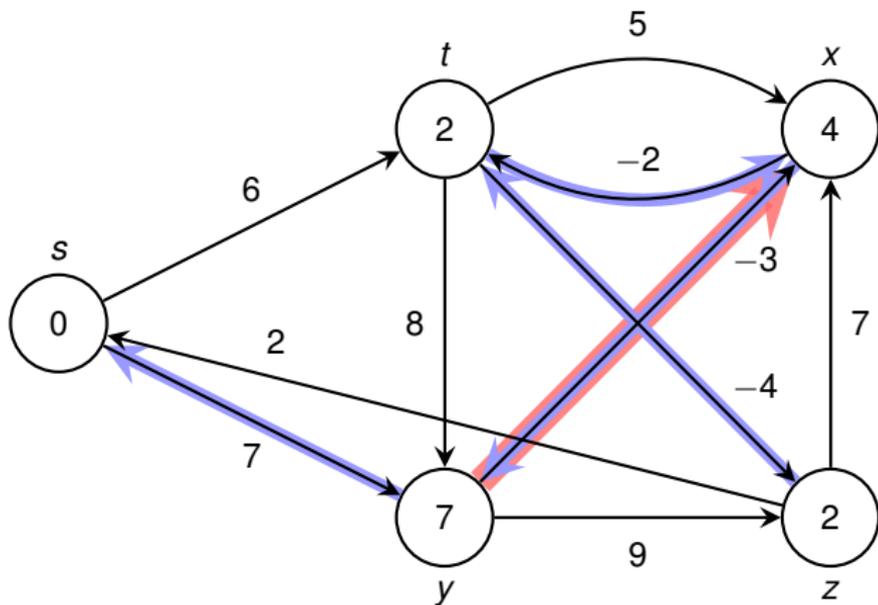
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

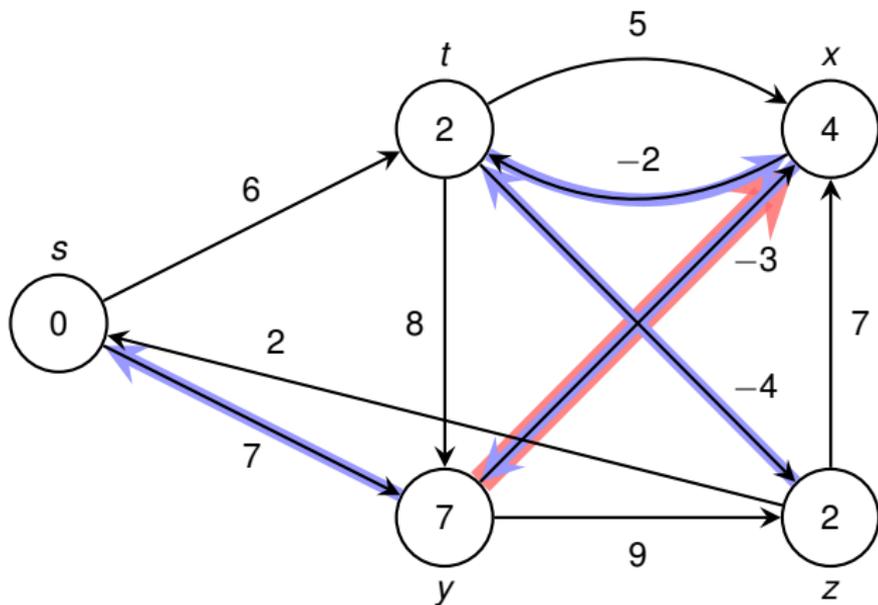
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

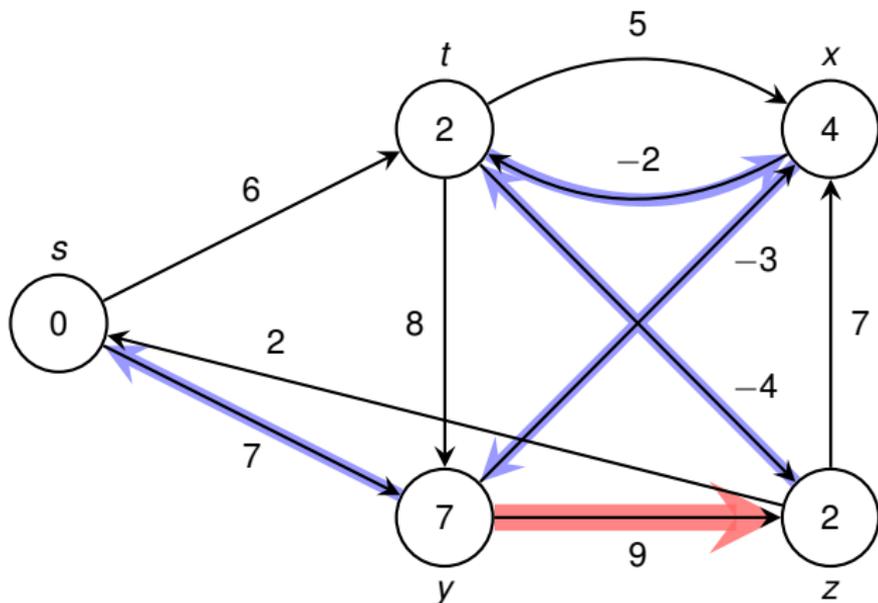
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

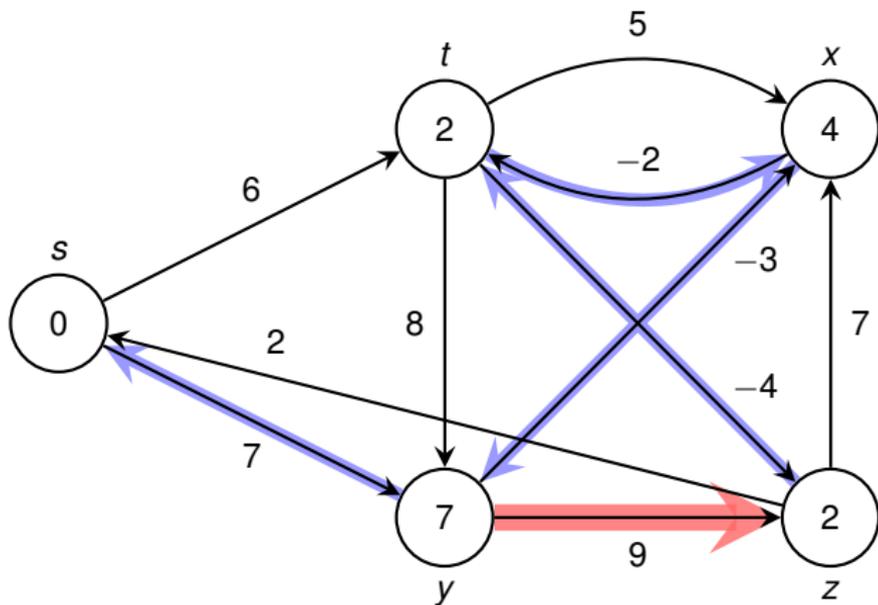
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

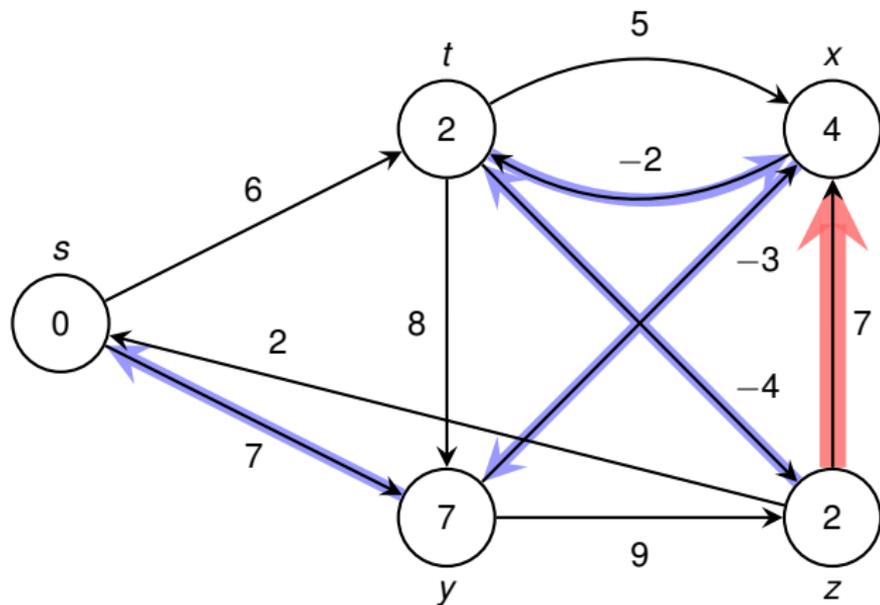
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

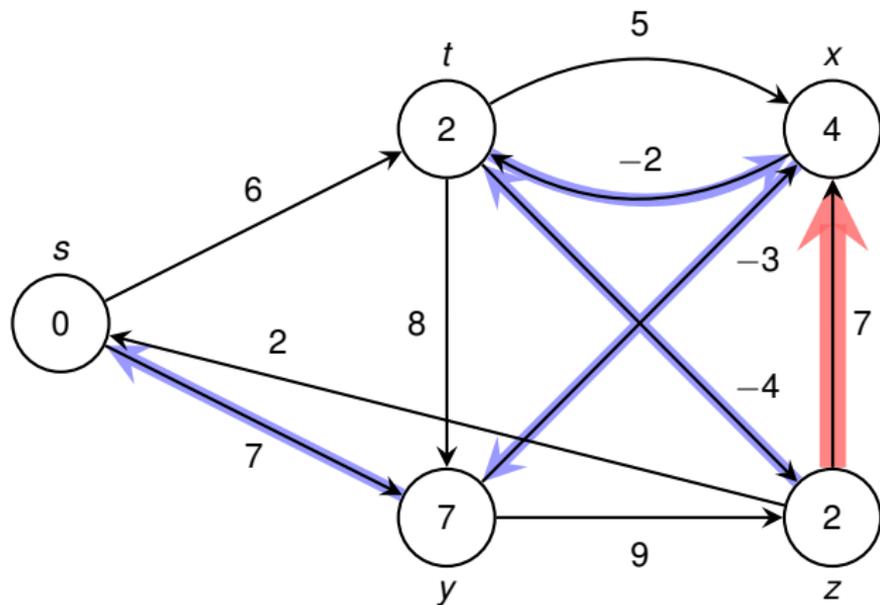
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

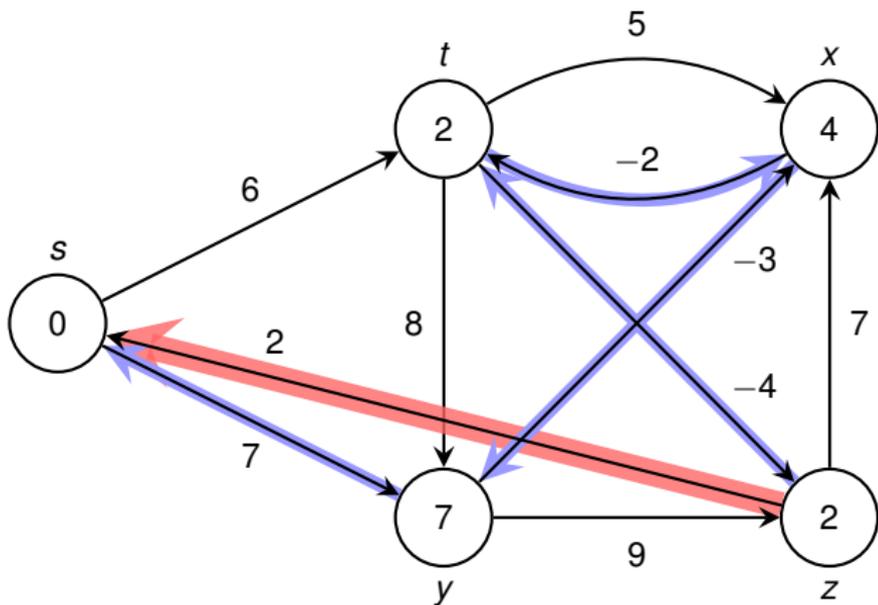
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

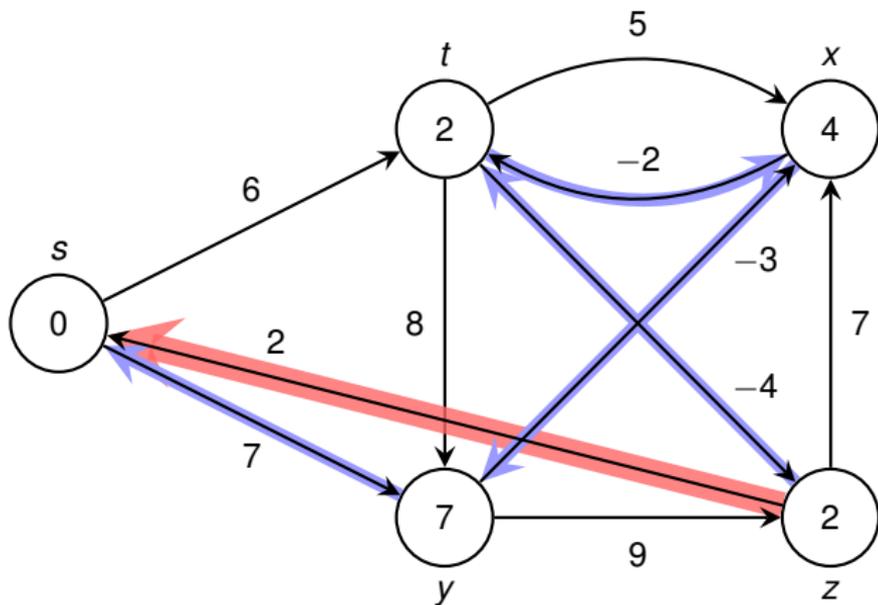
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

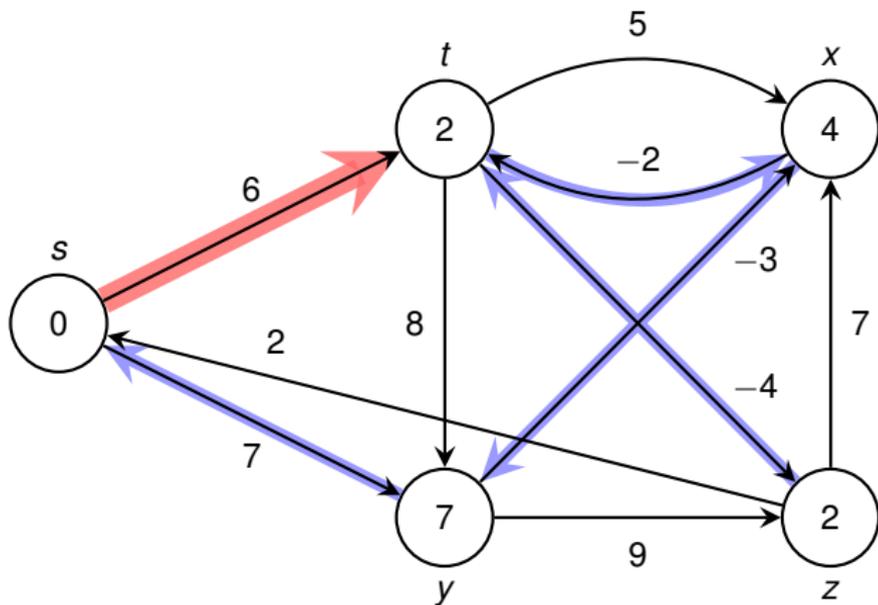
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

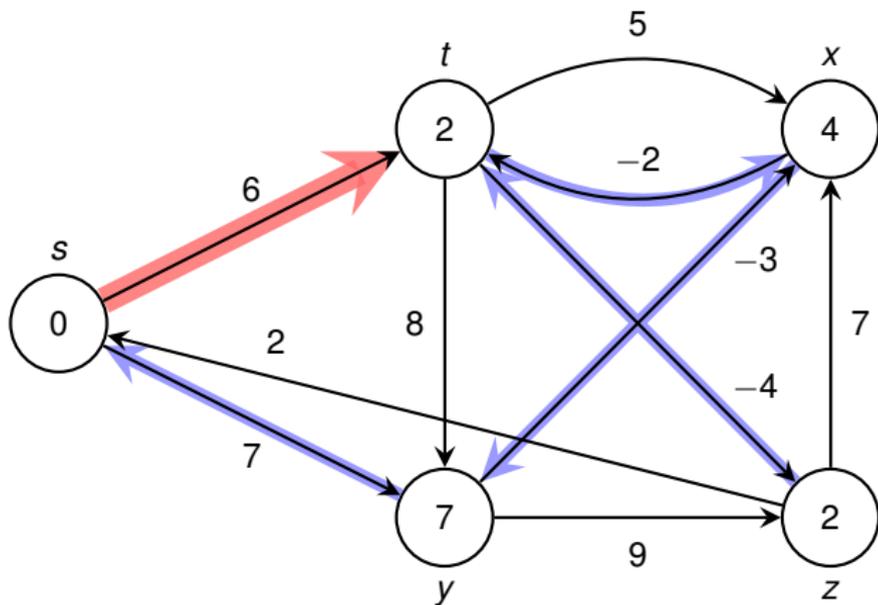
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

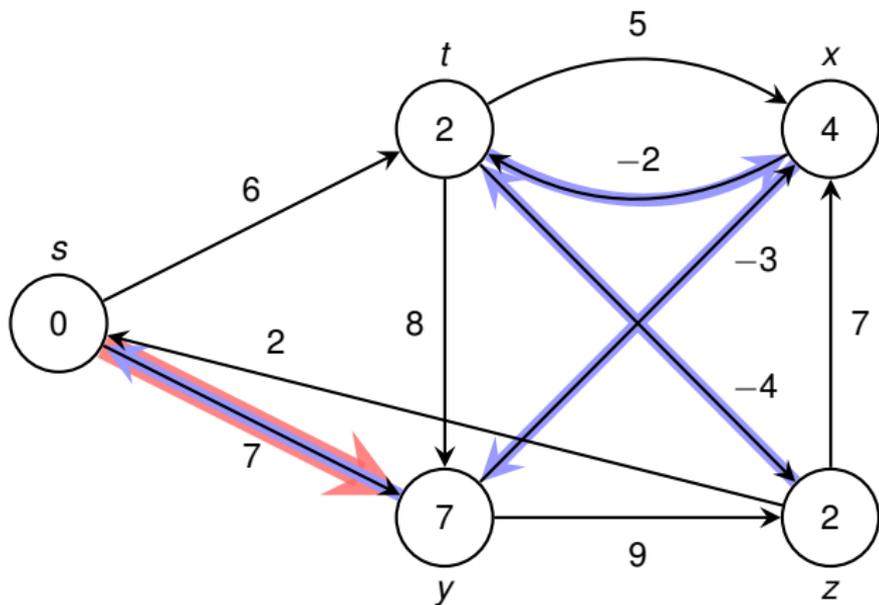
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

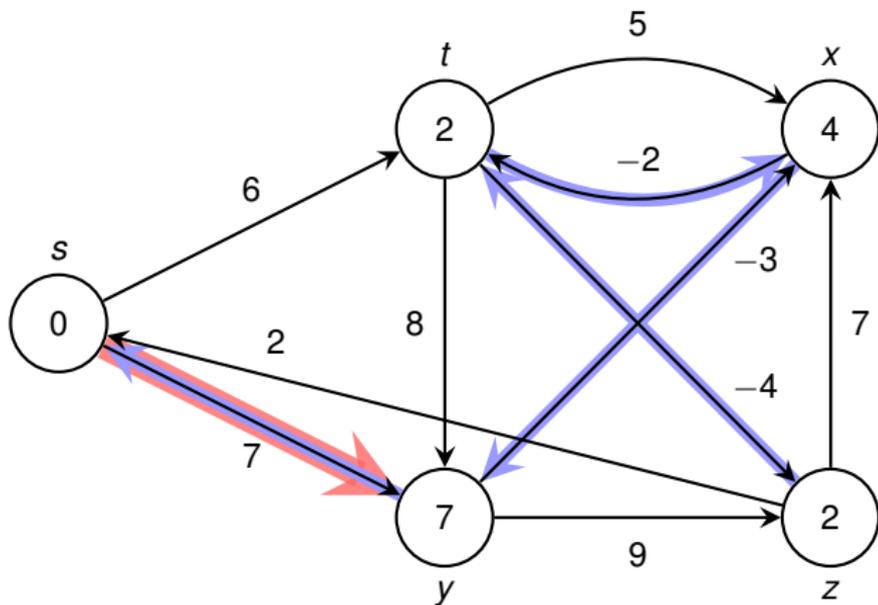
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 3

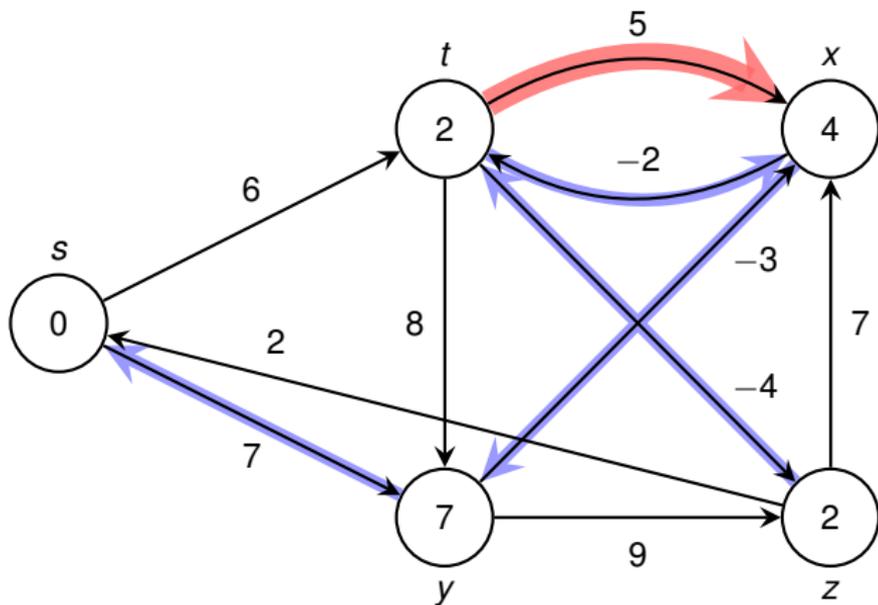
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

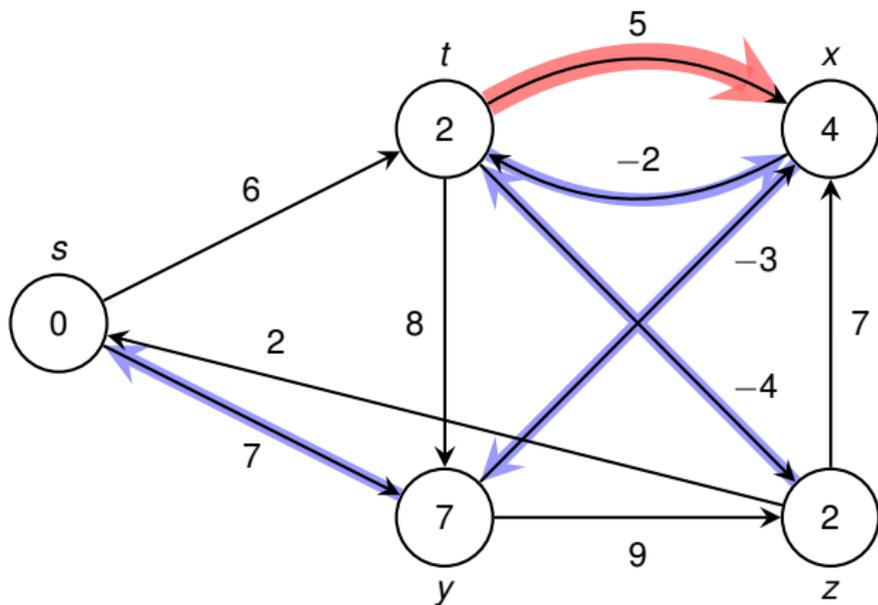
Relaxation Order:  $(t,x)$ ,  $(t,y)$ ,  $(t,z)$ ,  $(x,t)$ ,  $(y,x)$ ,  $(y,z)$ ,  $(z,x)$ ,  $(z,s)$ ,  $(s,t)$ ,  $(s,y)$



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

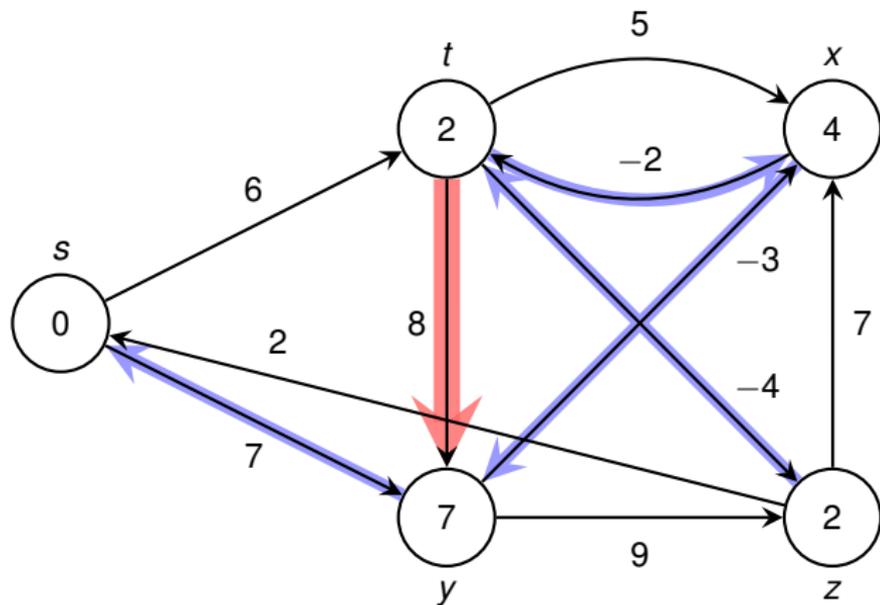
Relaxation Order:  $(t,x)$ ,  $(t,y)$ ,  $(t,z)$ ,  $(x,t)$ ,  $(y,x)$ ,  $(y,z)$ ,  $(z,x)$ ,  $(z,s)$ ,  $(s,t)$ ,  $(s,y)$



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

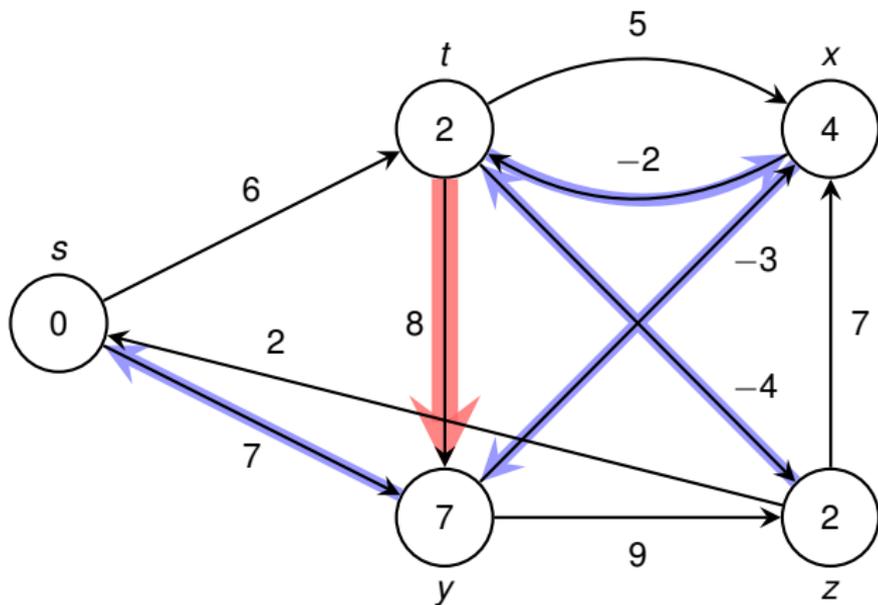
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

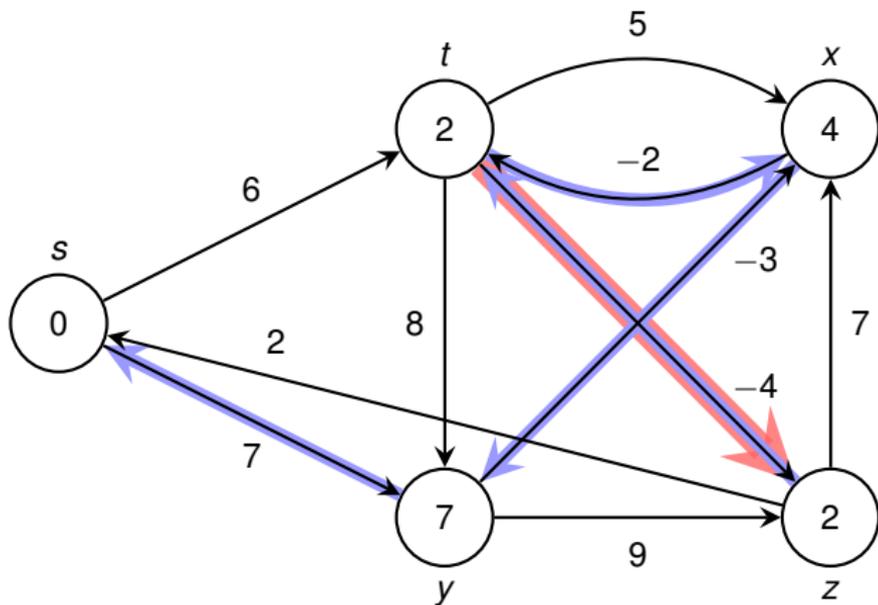
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

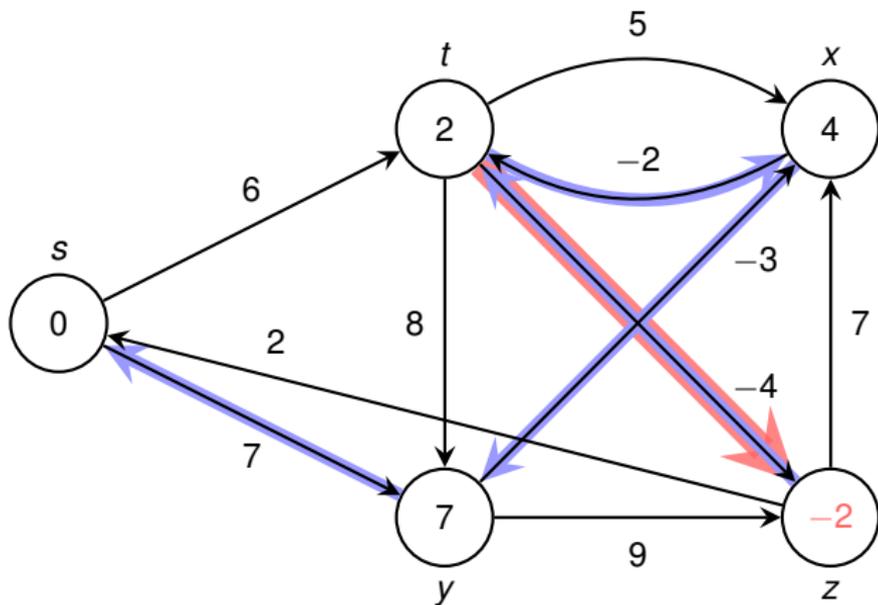
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

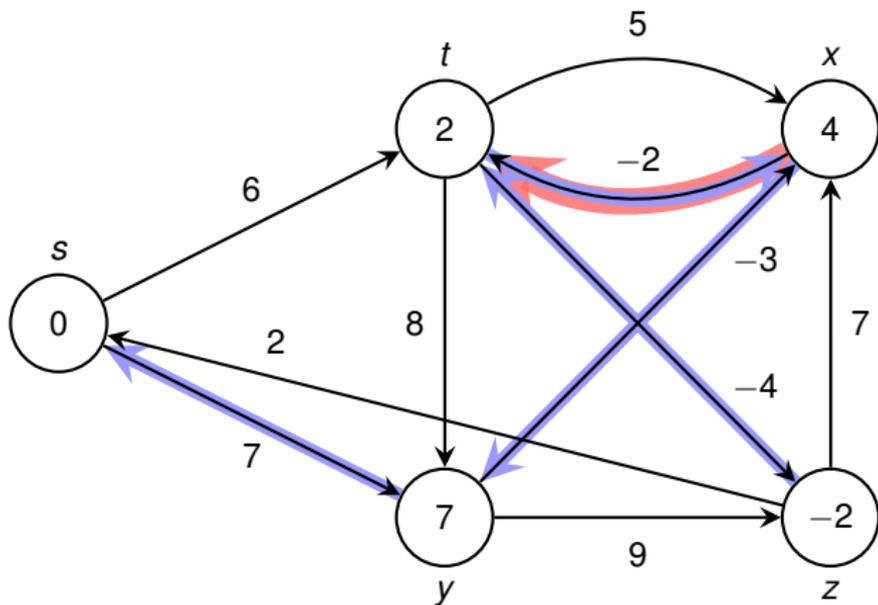
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

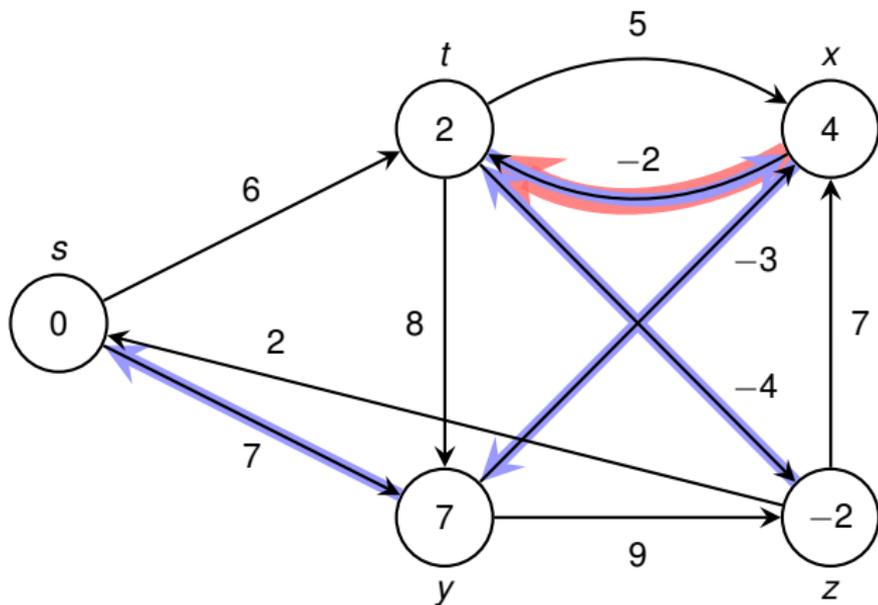
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

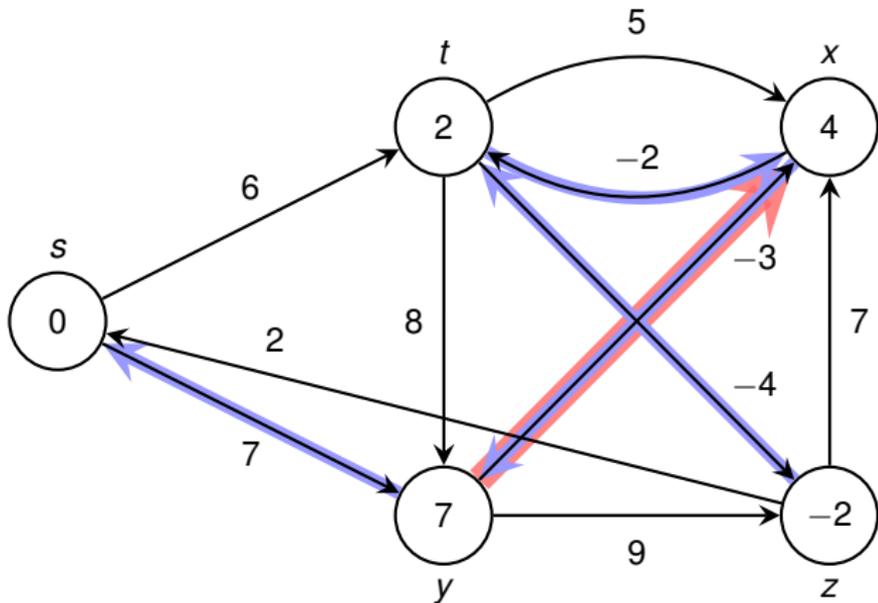
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

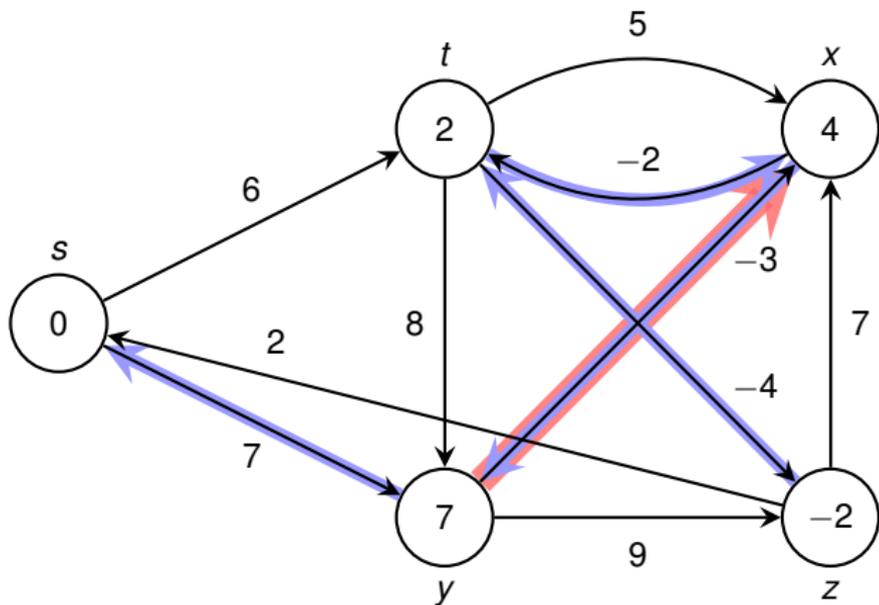
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

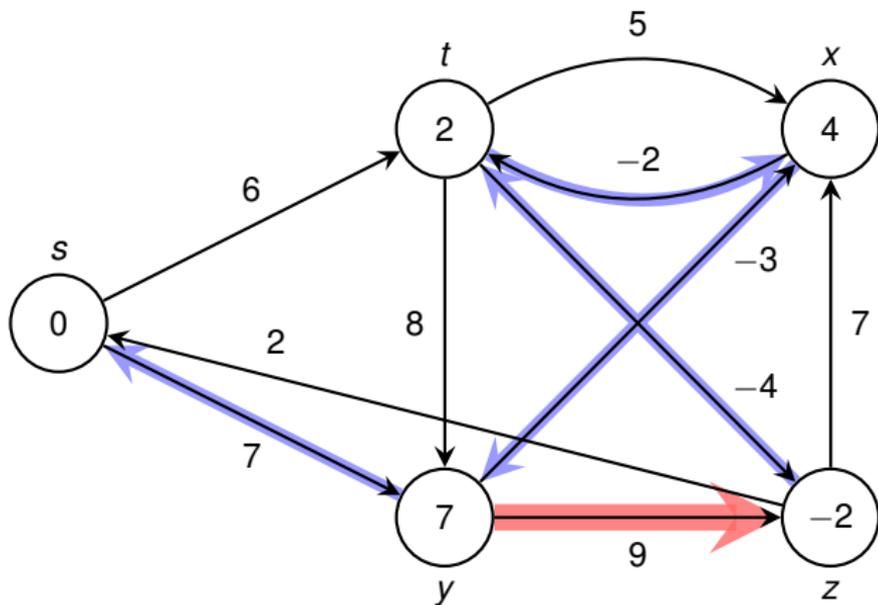
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

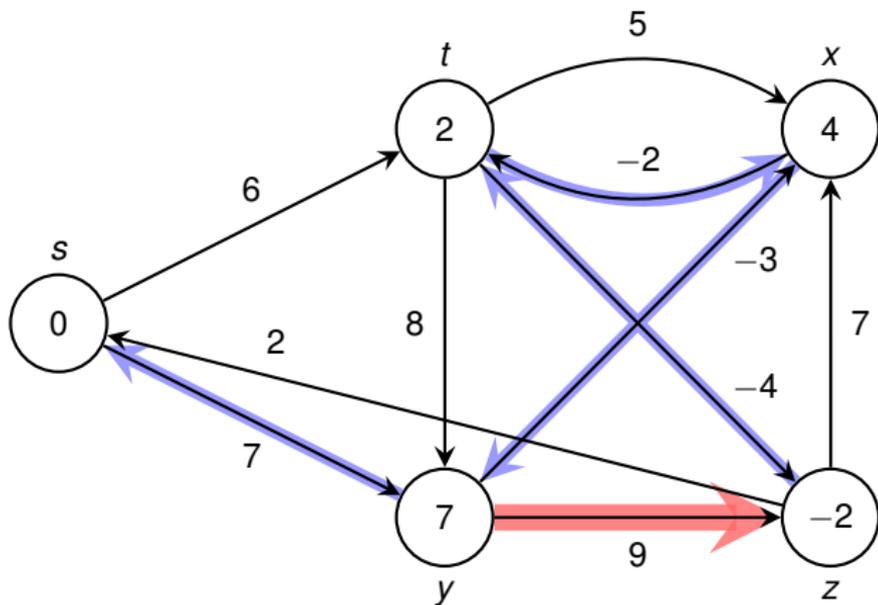
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

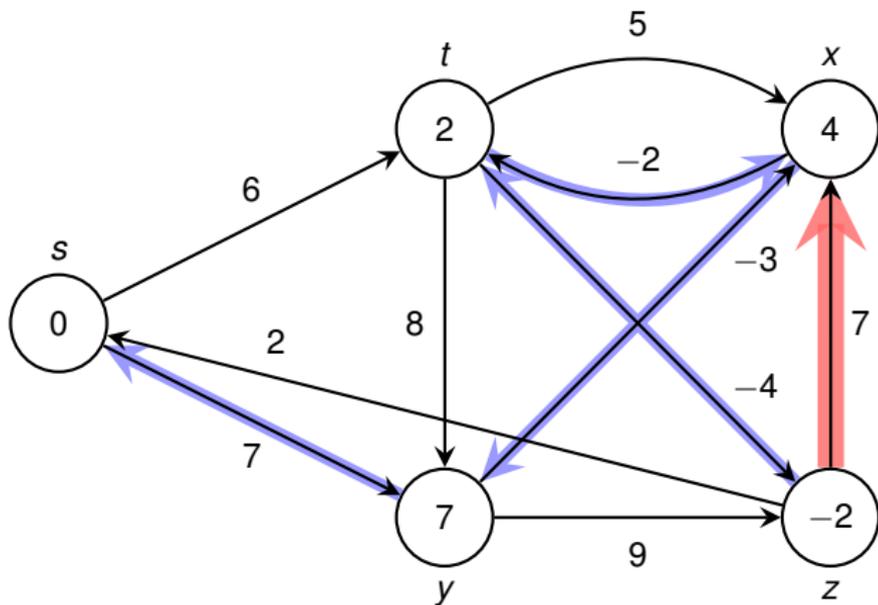
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

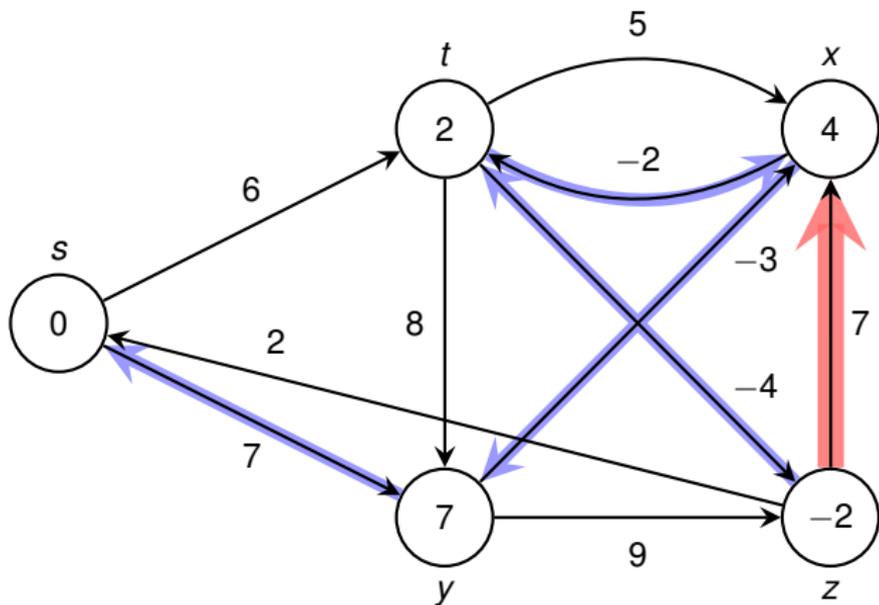
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

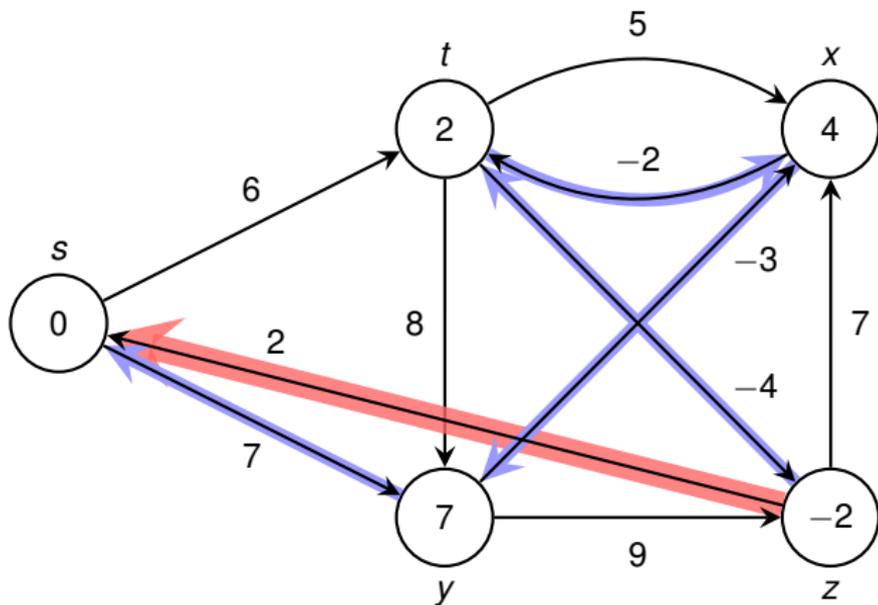
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

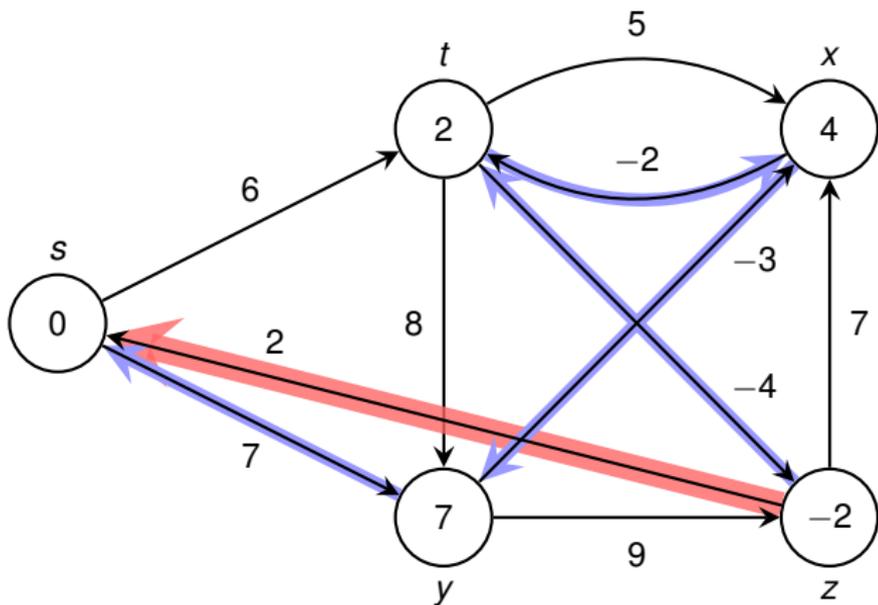
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

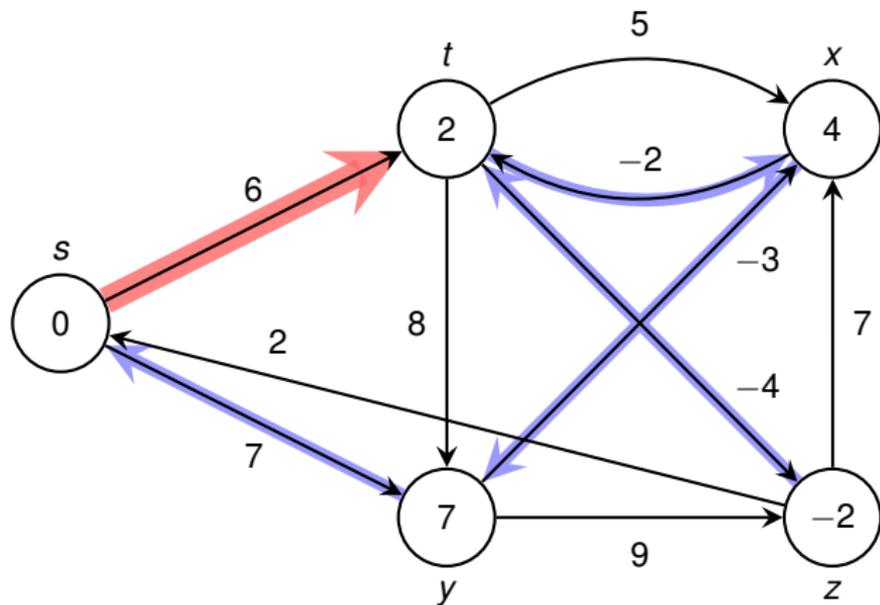
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

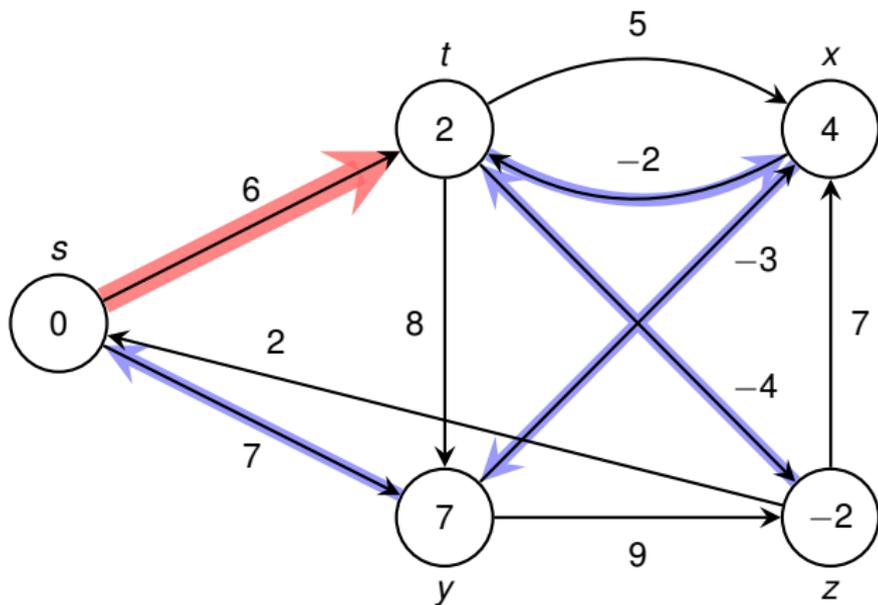
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

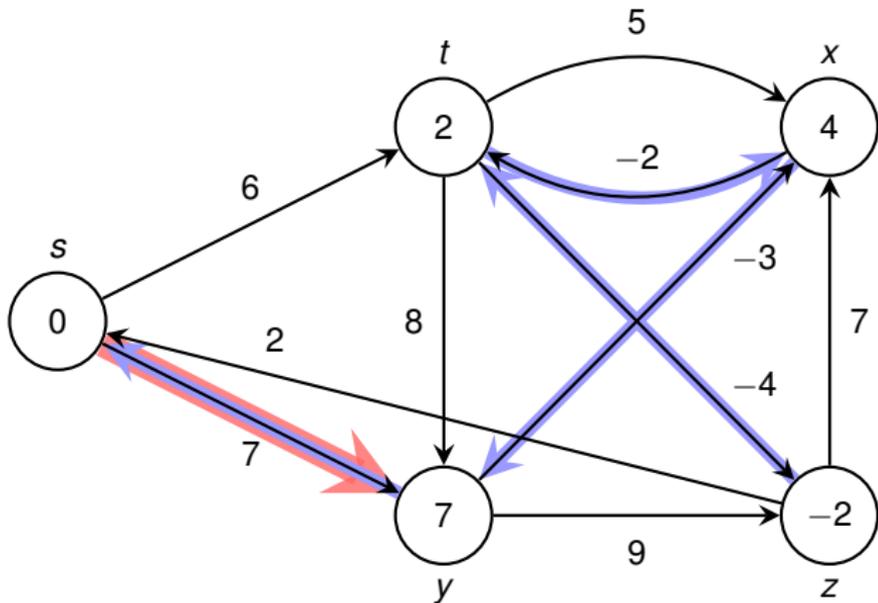
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

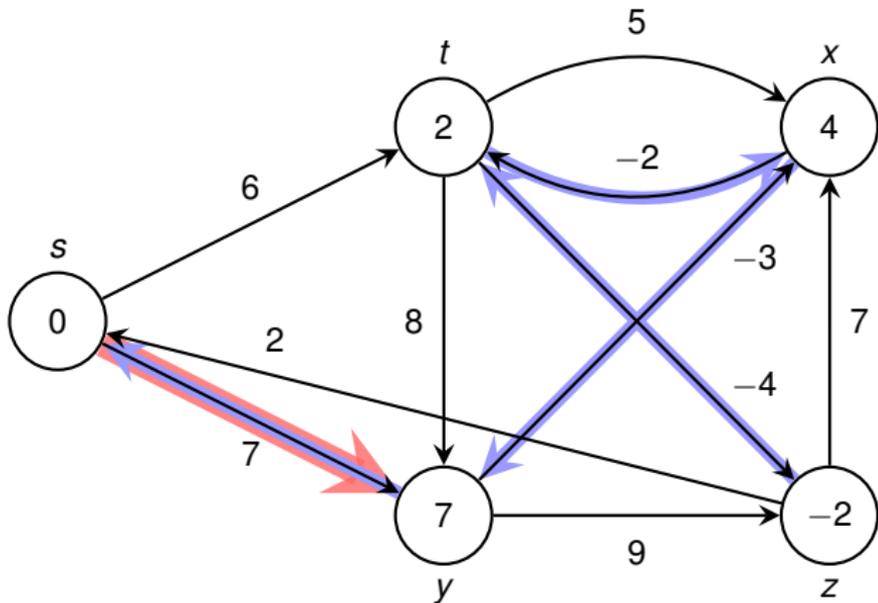
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

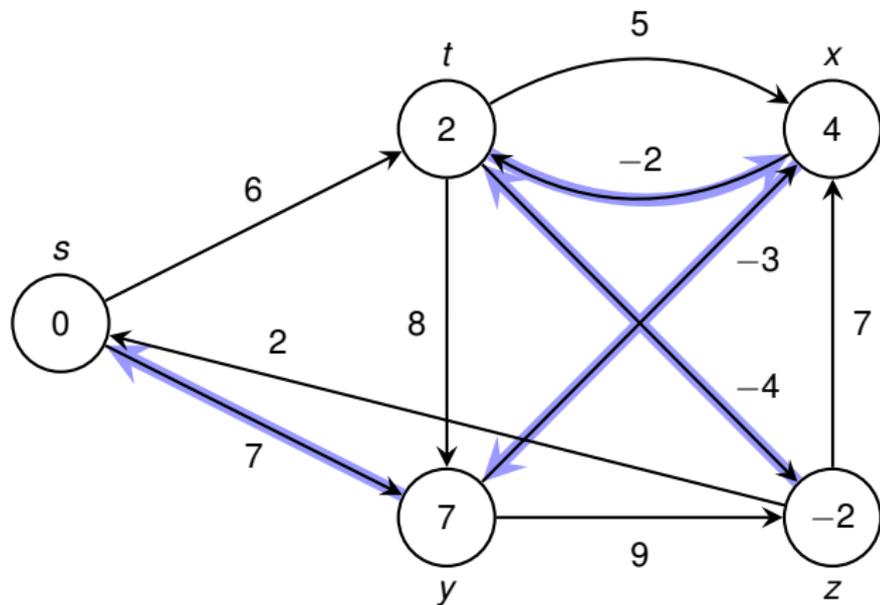
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



## Execution of Bellman-Ford (Figure 24.4)

Pass: 4

Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



## Bellman-Ford Algorithm: Correctness (1/2)

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### Lemma 24.2/Theorem 24.3

Assume that  $G$  contains no negative-weight cycles that are reachable from  $s$ . Then after  $|V| - 1$  passes, we have  $v.d = v.\delta$  for all vertices  $v \in V$  (that are reachable) and Bellman-Ford returns TRUE.



## Bellman-Ford Algorithm: Correctness (1/2)

---

### Lemma 24.2/Theorem 24.3

Assume that  $G$  contains no negative-weight cycles that are reachable from  $s$ . Then after  $|V| - 1$  passes, we have  $v.d = v.\delta$  for all vertices  $v \in V$  (that are reachable) and Bellman-Ford returns TRUE.

Proof that  $v.d = v.\delta$

- Let  $v$  be a vertex reachable from  $s$



## Bellman-Ford Algorithm: Correctness (1/2)

### Lemma 24.2/Theorem 24.3

Assume that  $G$  contains no negative-weight cycles that are reachable from  $s$ . Then after  $|V| - 1$  passes, we have  $v.d = v.\delta$  for all vertices  $v \in V$  (that are reachable) and Bellman-Ford returns TRUE.

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Triangle inequality (holds even if  $w(u, v) < 0!$ )



## Bellman-Ford Algorithm: Correctness (2/2)

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If  $G$  contains a negative-weight cycle reachable from  $s$ , then Bellman-Ford returns FALSE.



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If  $G$  contains a **negative-weight cycle** reachable from  $s$ , then Bellman-Ford returns FALSE.

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- Let  $c = (v_0, v_1, \dots, v_k = v_0)$  be a **negative-weight cycle** reachable from  $s$
- If Bellman-Ford returns TRUE, then for every  $1 \leq i < k$ ,

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



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This cancellation is only valid if all  $.d$ -values are finite!



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- This contradicts the assumption that  $c$  is a **negative-weight cycle**! □



## The Bellman-Ford Algorithm

```
BELLMAN-FORD (G, w, s)
0: assert (s in G.vertices())
1: for v in G.vertices()
2:     v.predecessor = None
3:     v.d = Infinity
4: s.d = 0
5:
6: repeat |V|-1 times
7:     for e in G.edges()
8:         Relax edge e=(u,v): Check if  $u.d + w(u,v) < v.d$ 
9:             if e.start.d + e.weight.d < e.end.d:
10:                 e.end.d = e.start.d + e.weight
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## The Bellman-Ford Algorithm (modified)

```
BELLMAN-FORD-NEW( $G, w, s$ )
0: assert( $s$  in  $G.vertices()$ )
1: for  $v$  in  $G.vertices()$ 
2:    $v.predecessor = None$ 
3:    $v.d = Infinity$ 
4:  $s.d = 0$ 
5:
6: repeat  $|V|$  times
7:   flag = 0
8:   for  $e$  in  $G.edges()$ 
9:     Relax edge  $e=(u,v)$ : Check if  $u.d + w(u,v) < v.d$ 
10:      if  $e.start.d + e.weight < e.end.d$ :
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