

6.4: Single-Source Shortest Paths

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Thomas Sauerwald

Lent 2015



UNIVERSITY OF
CAMBRIDGE

Introduction

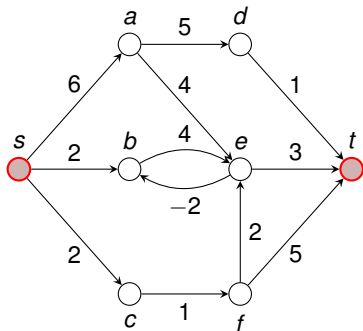
Bellman-Ford Algorithm



Shortest Path Problem

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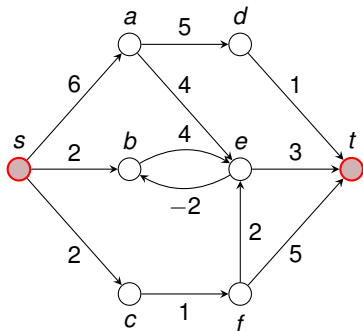
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- Goal: Find a path of **minimum weight** from s to t in G

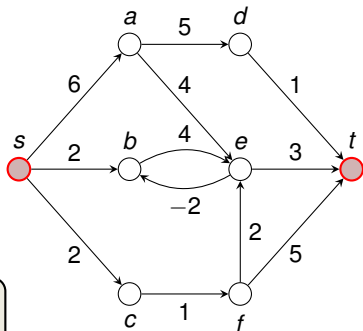


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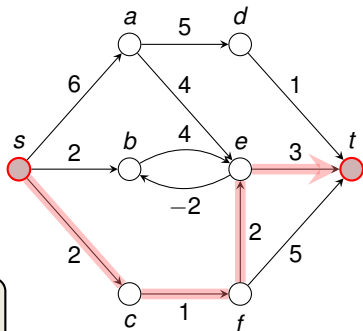


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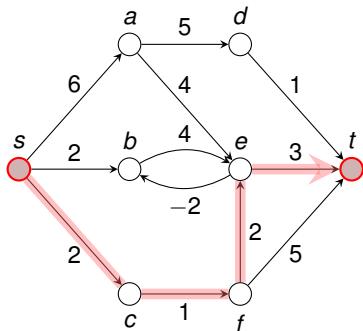
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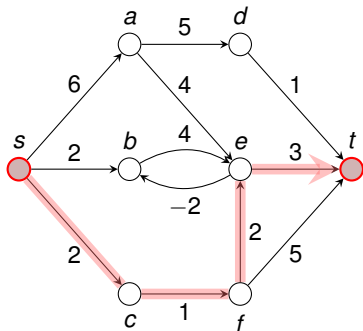
How to cope with an **unweighted** graph G ?



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How to cope with an **unweighted** graph G ?

Two possible answers are:

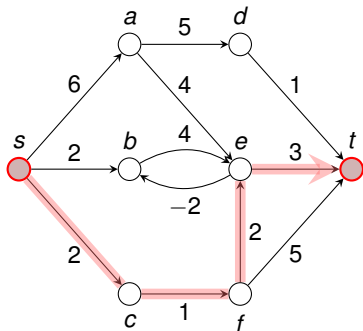
1. Run BFS (computes shortest paths in unweighted graphs)
2. Add a weight of 1 to all edges



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Applications

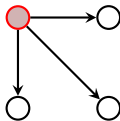
- Car Navigation, Traffic Planning, Internet Routing, Arbitrage in Concurrency Exchange, ...



Variants of Shortest Path Problems

Single-source shortest-paths problem (SSSP)

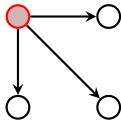
- Bellman-Ford Algorithm
- Dijkstra Algorithm



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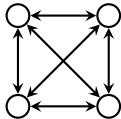
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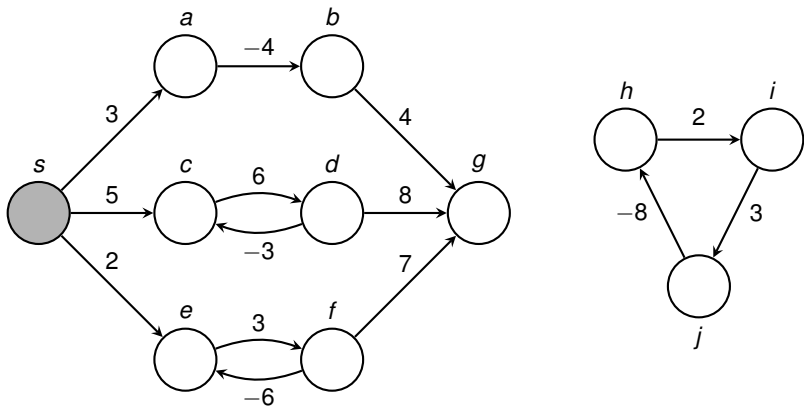


All-pairs shortest-paths problem (APSP)

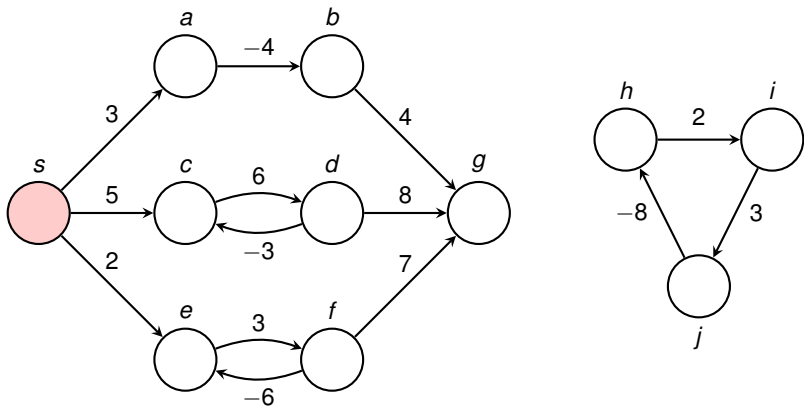
- Shortest Paths via Matrix Multiplication
- Johnson's Algorithm



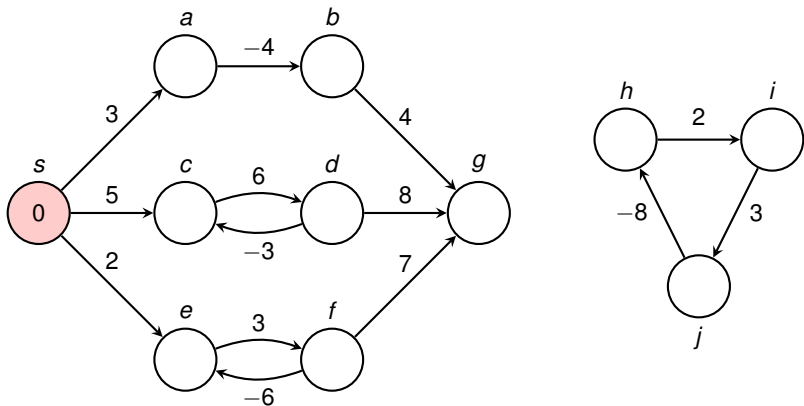
Distances and Negative-Weight Cycles (Figure 24.1)



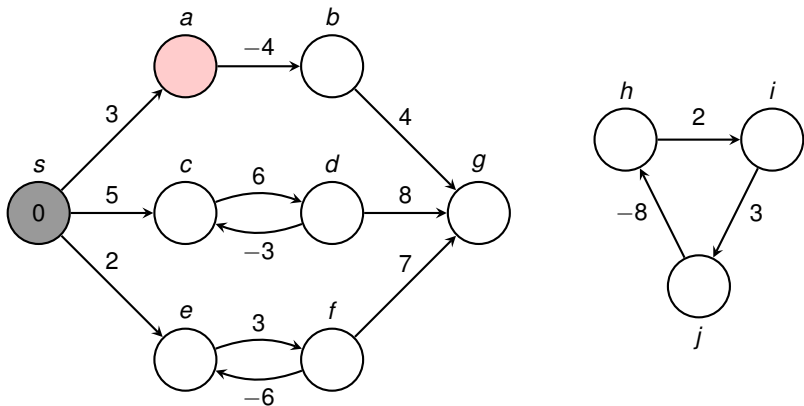
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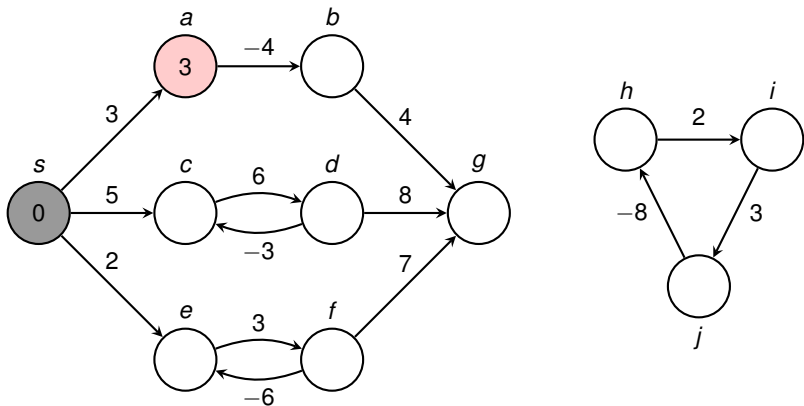
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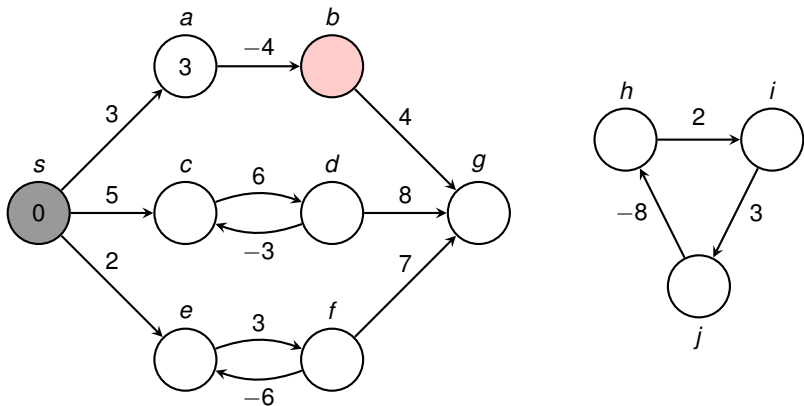
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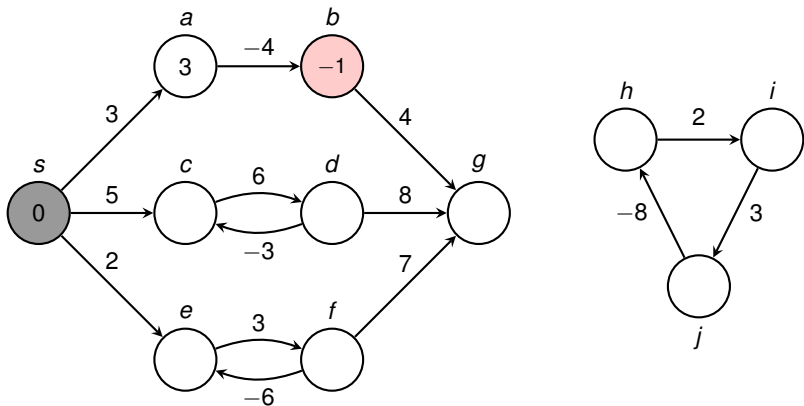
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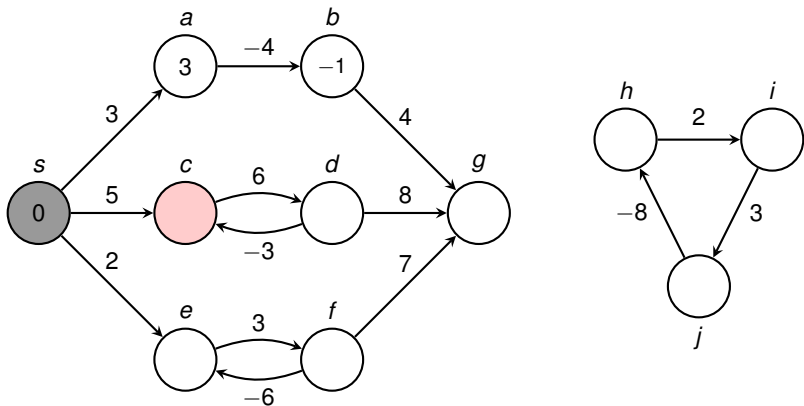
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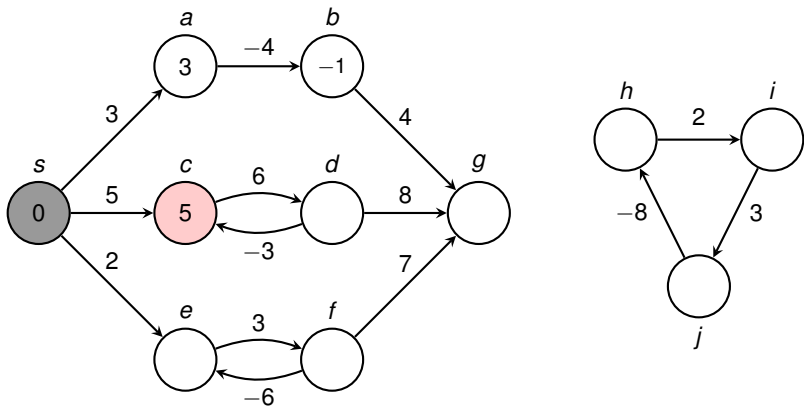
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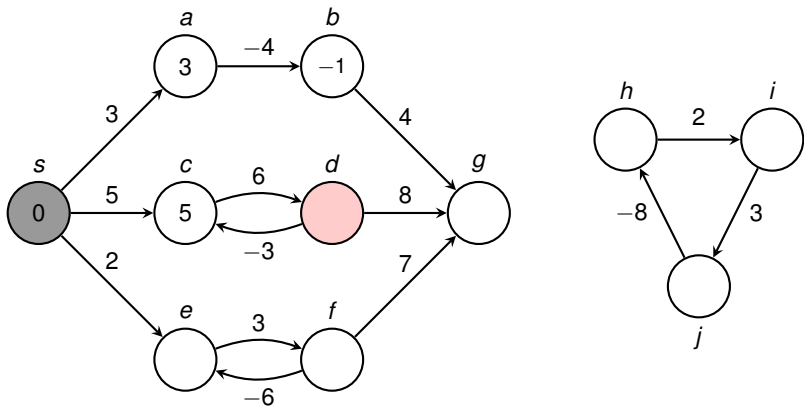
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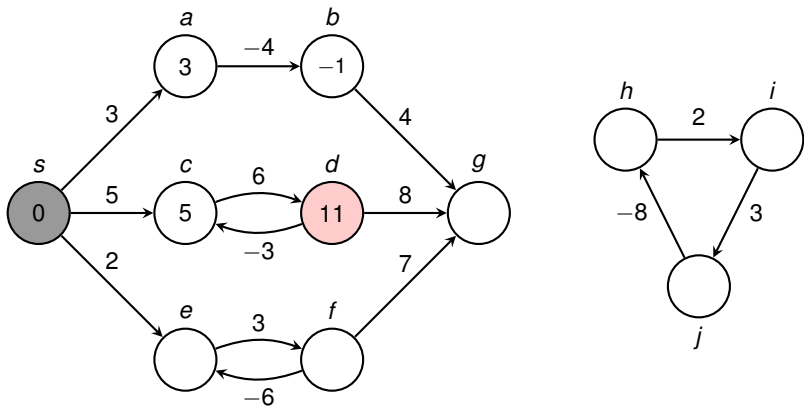
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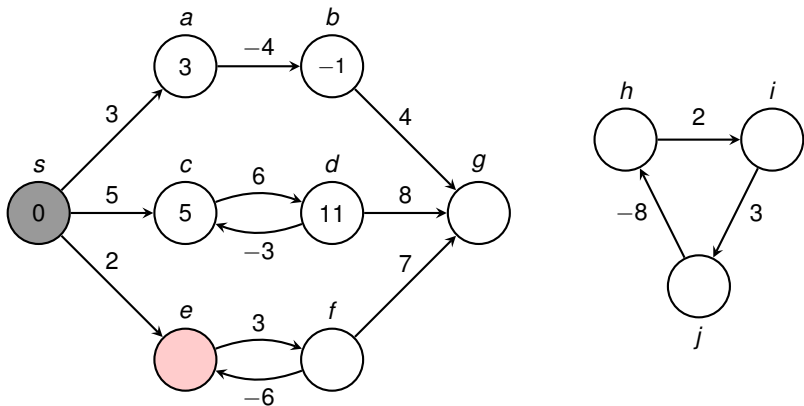
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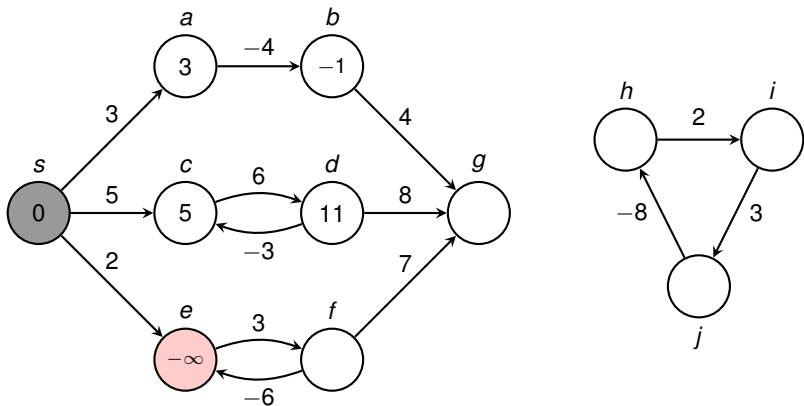
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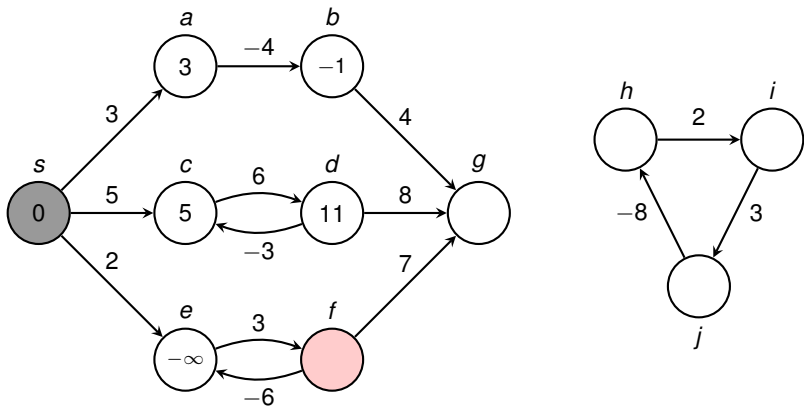
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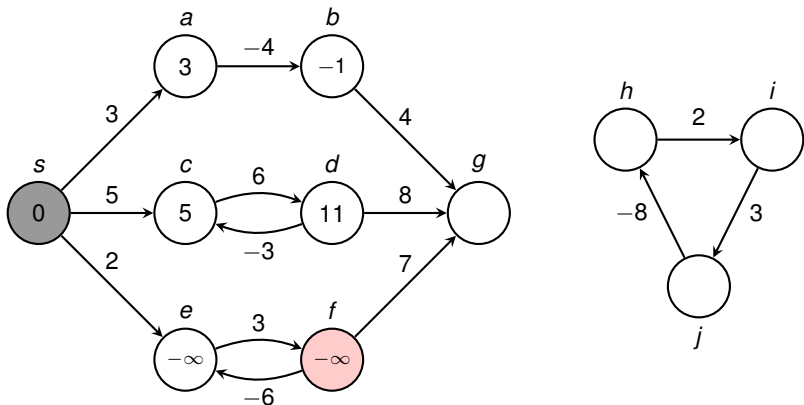
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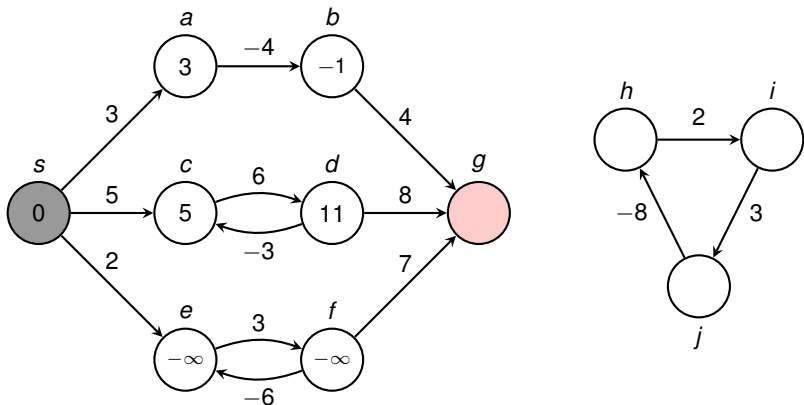
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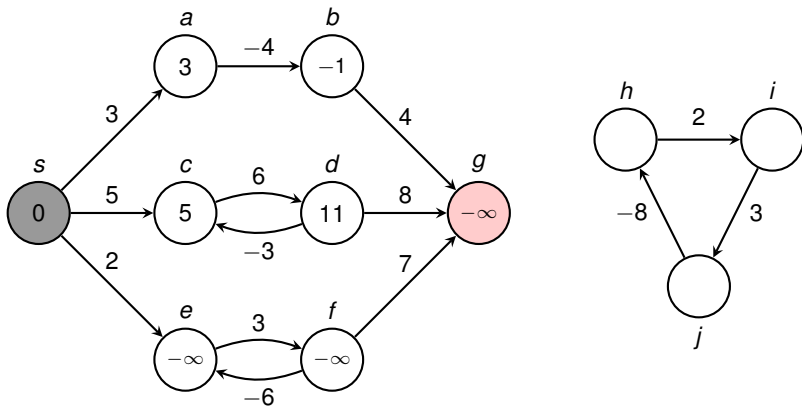
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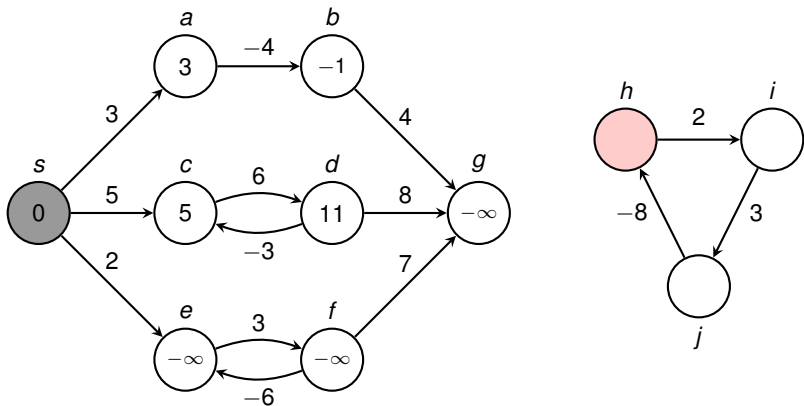
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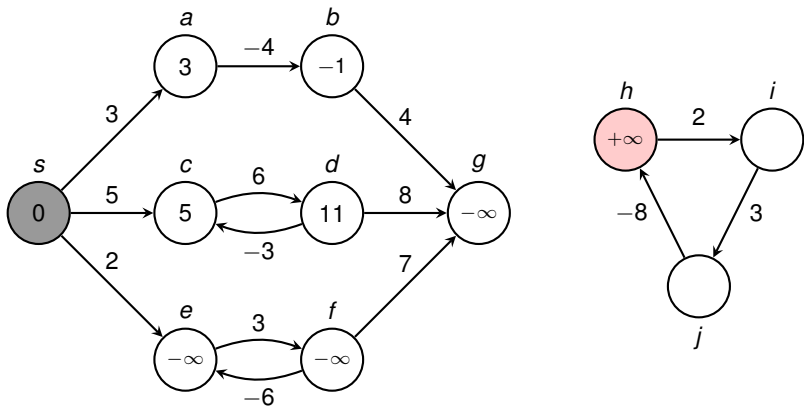
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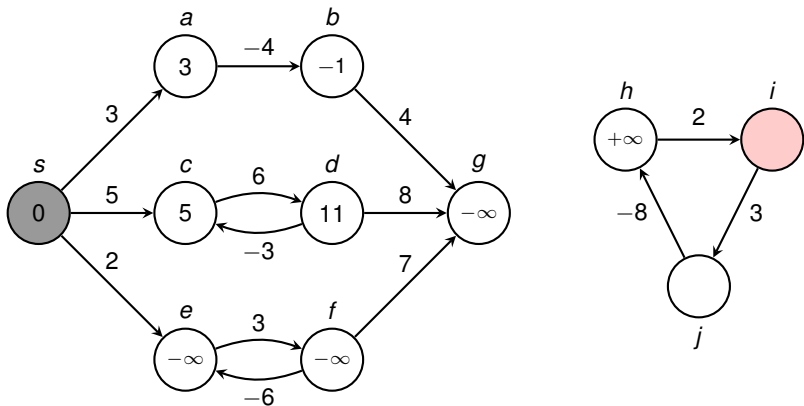
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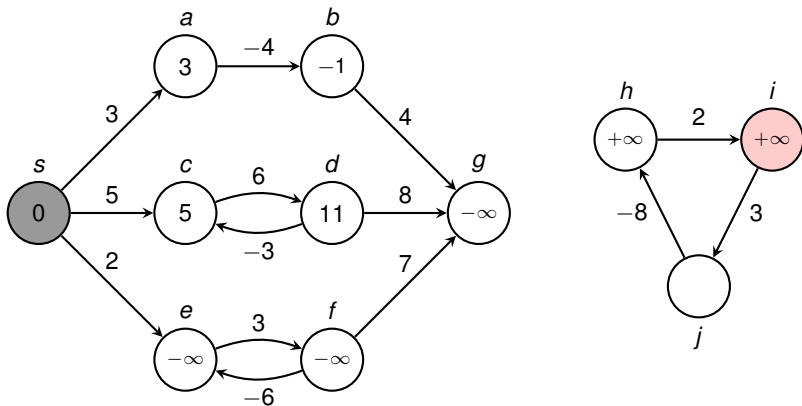
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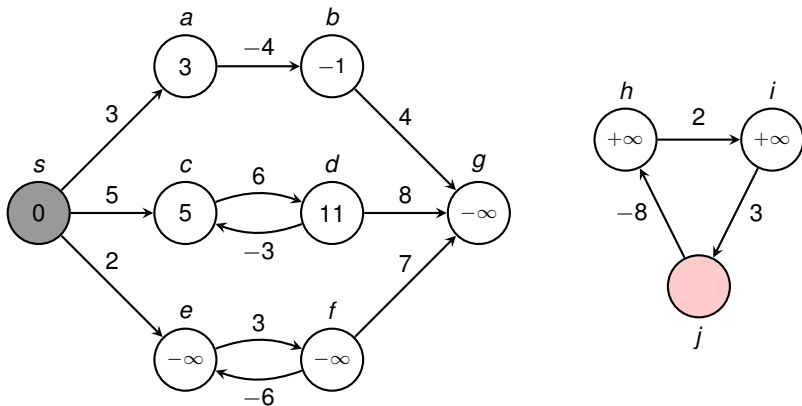
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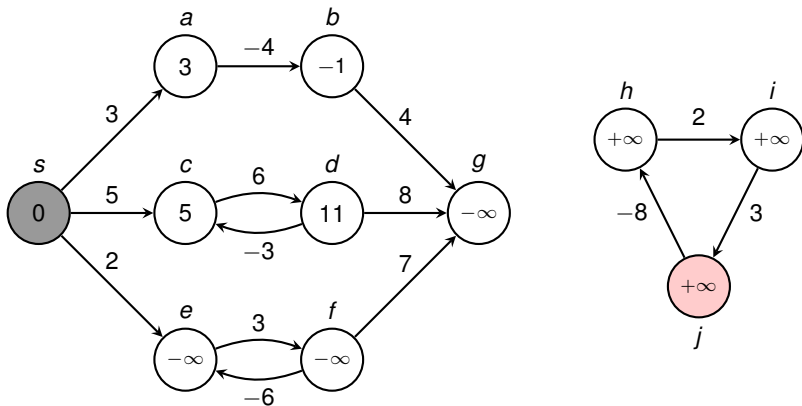
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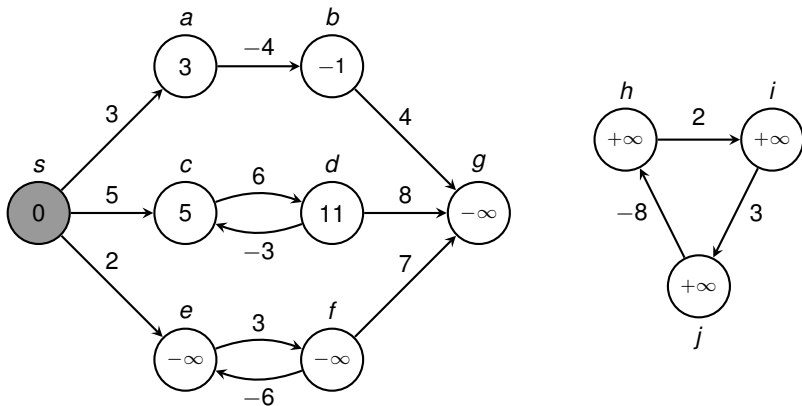
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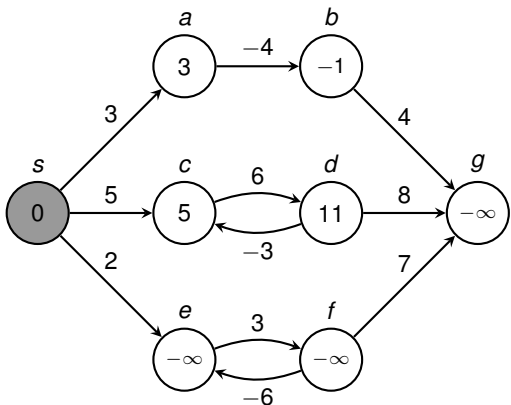
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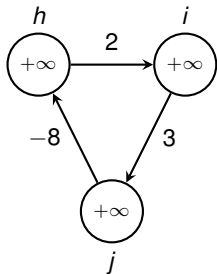
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Negative-Weight Cycle
(reachable from s)



Negative-Weight Cycle
(not reachable from s)



Introduction

Bellman-Ford Algorithm



Relaxing Edges

Definition

Fix the source vertex $s \in V$

- $v.\delta$ is the length of the shortest path (distance) from s to v
- $v.d$ is the length of the shortest path discovered so far



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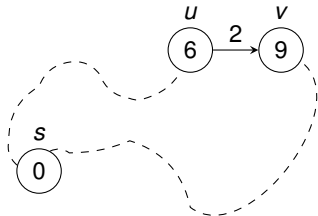
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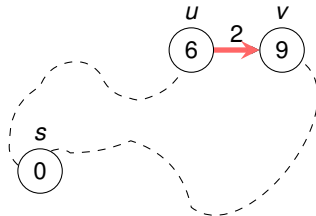
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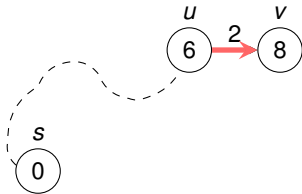
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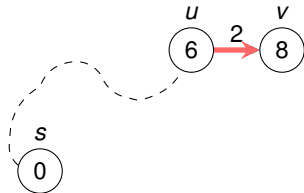
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After relaxing (u, v) , regardless of whether we found a shortcut:
 $v.d \leq u.d + w(u, v)$



Properties of Shortest Paths and Relaxations

Toolkit

Triangle inequality (Lemma 24.10)

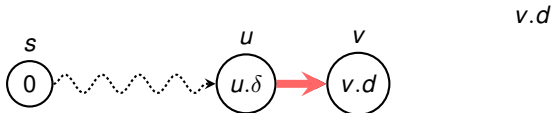
- For any edge $(u, v) \in E$, we have $v.\delta \leq u.\delta + w(u, v)$

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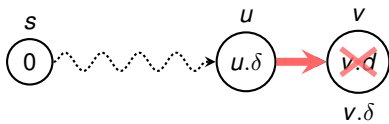
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Since $v.d \geq v.\delta$, we have $v.d = v.\delta$. \square



Path-Relaxation Property

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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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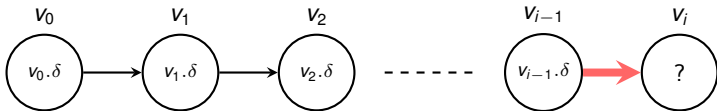
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- Inductive Step ($i - 1 \rightarrow i$):** Assume $v_{i-1}.d = v_{i-1}.\delta$ and relax (v_{i-1}, v_i) .



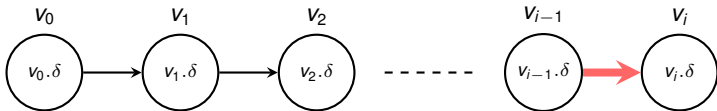
Path-Relaxation Property

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 $s.d = s.\delta = 0$.
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- Inductive Step ($i - 1 \rightarrow i$):** Assume $v_{i-1}.d = v_{i-1}.\delta$ and relax (v_{i-1}, v_i) .
Convergence Property $\Rightarrow v_i.d = v_i.\delta$ (now and at all later steps) □



Path-Relaxation Property

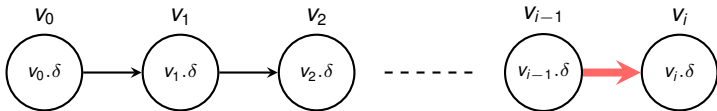
“Propagation”: By relaxing proper edges, set of vertices with $v.\delta = v.d$ gets larger

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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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:        $e.end.d = e.start.d + e.weight$ 
11:        $e.end.predecessor = e.start$ 
12:
13: for  $e$  in  $G.edges()$ 
14:   if  $e.start.d + e.weight.d < e.end.d$ :
15:     return FALSE
16: return TRUE
```



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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

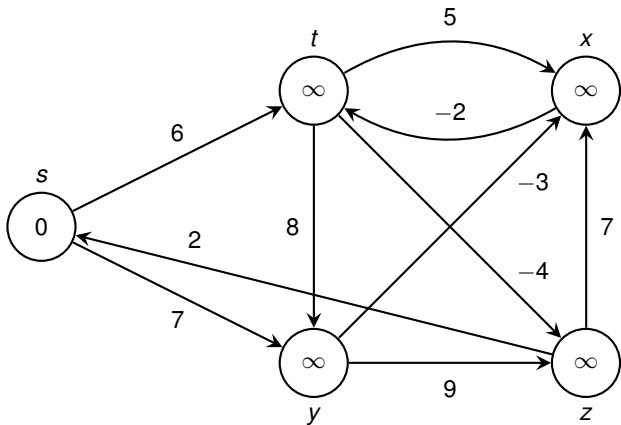
- A single call of line 9-11 costs $\mathcal{O}(1)$
- In each pass every edge is relaxed $\Rightarrow \mathcal{O}(E)$ time per pass
- Overall $(V - 1) + 1 = V$ passes $\Rightarrow \mathcal{O}(V \cdot E)$ time



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

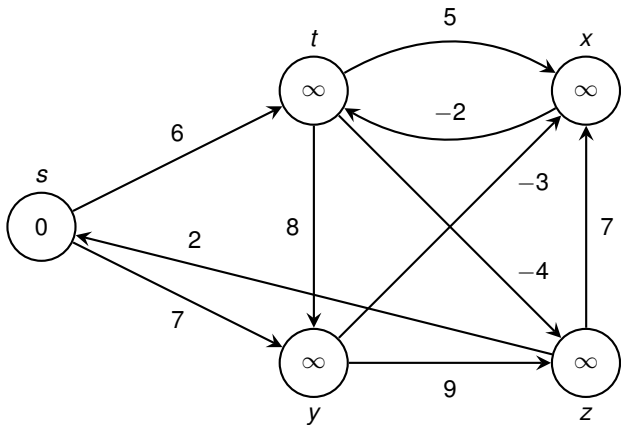
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

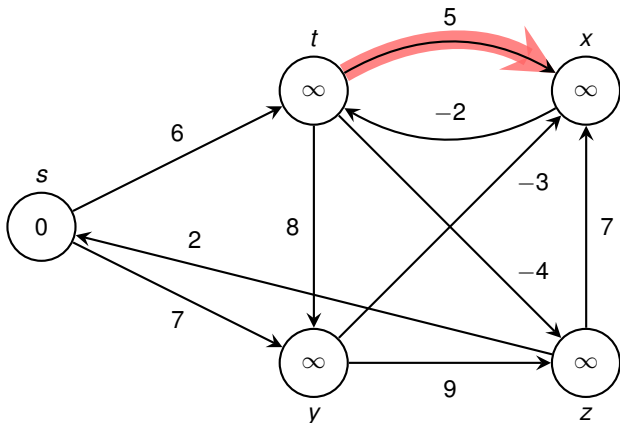
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

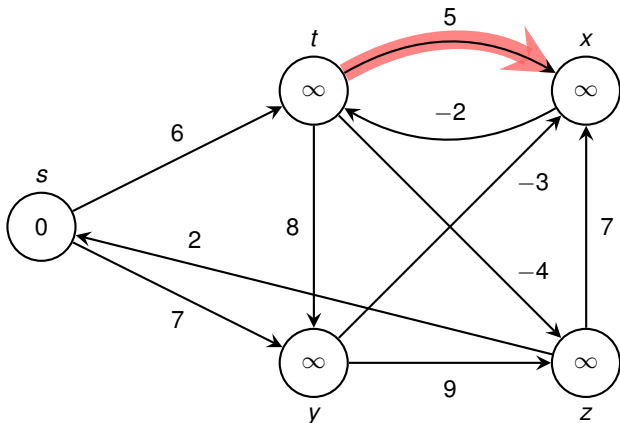
Relaxation Order: (t,x) , (t,y) , (t,z) , (x,t) , (y,x) , (y,z) , (z,x) , (z,s) , (s,t) , (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

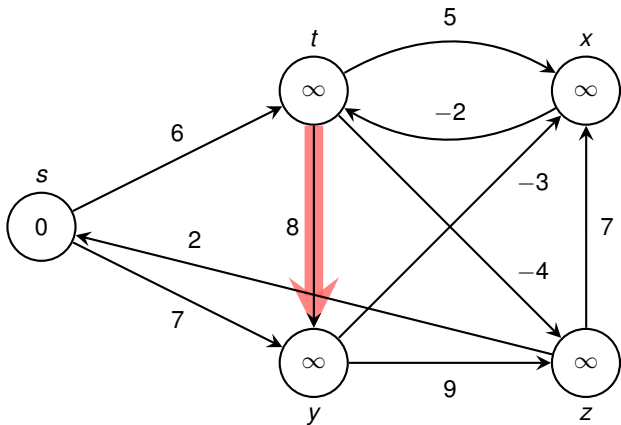
Relaxation Order: (t,x) , (t,y) , (t,z) , (x,t) , (y,x) , (y,z) , (z,x) , (z,s) , (s,t) , (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

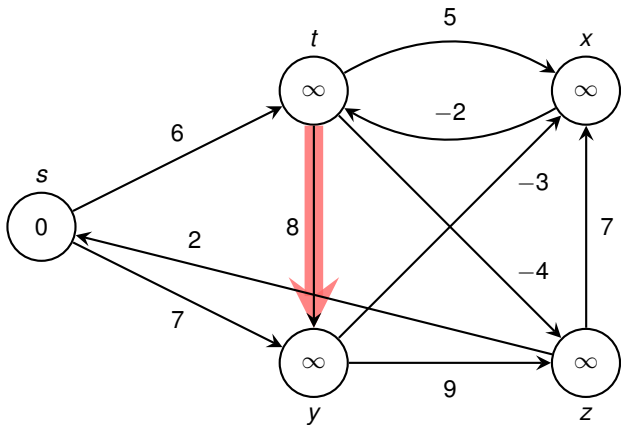
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

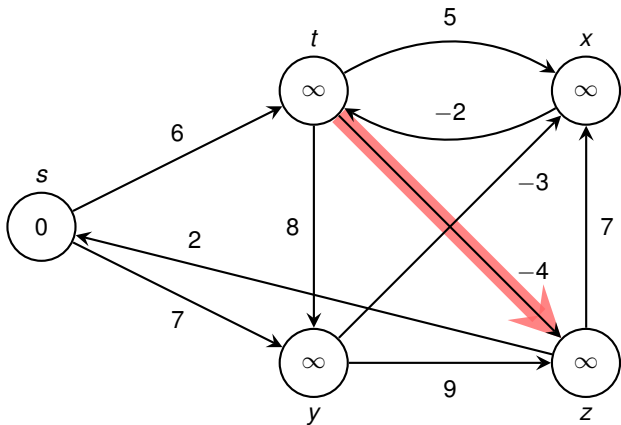
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

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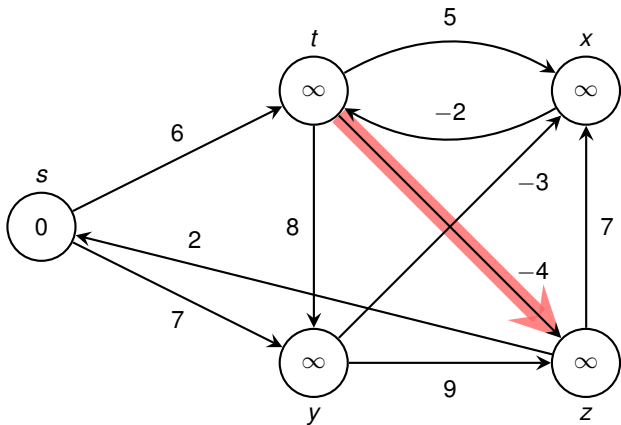
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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Pass: 1

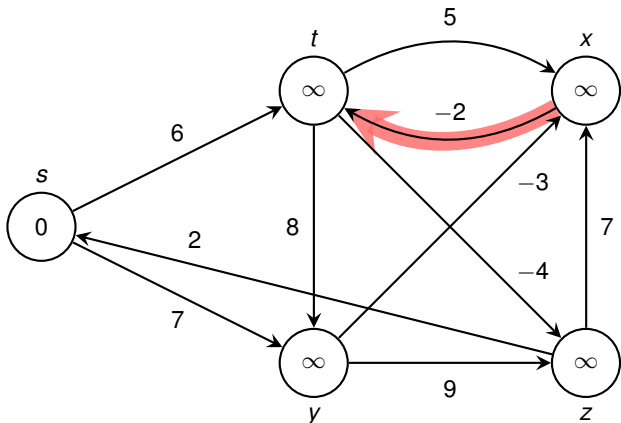
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

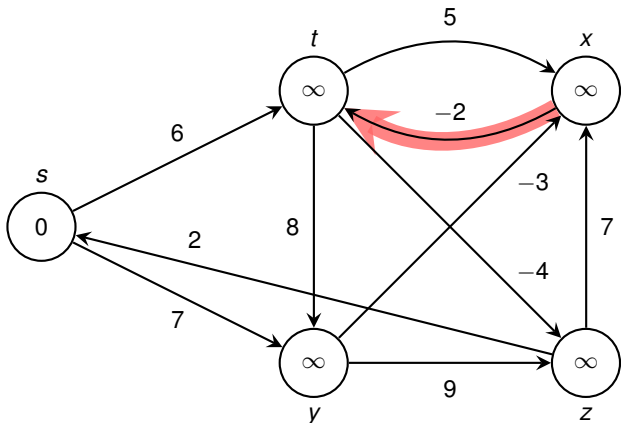
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

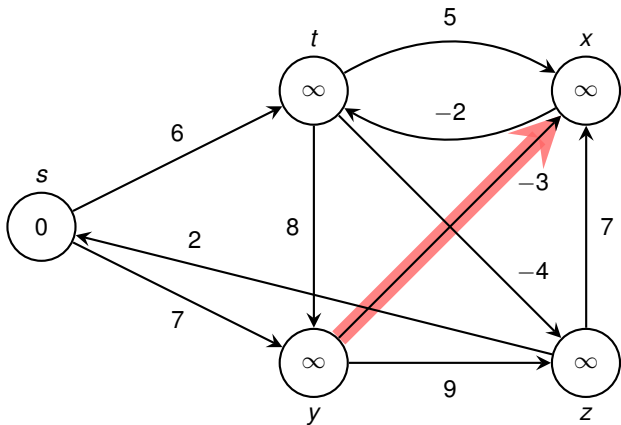
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

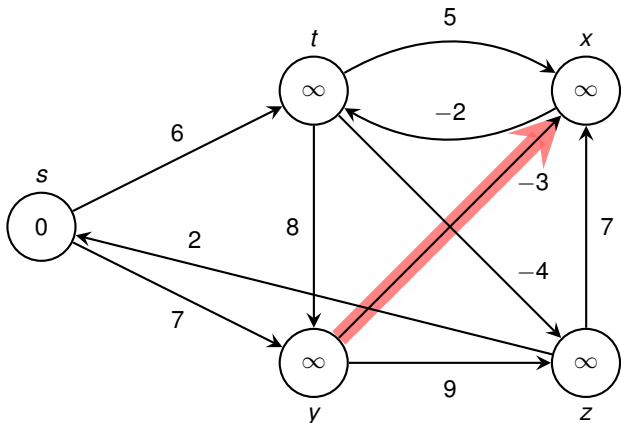
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

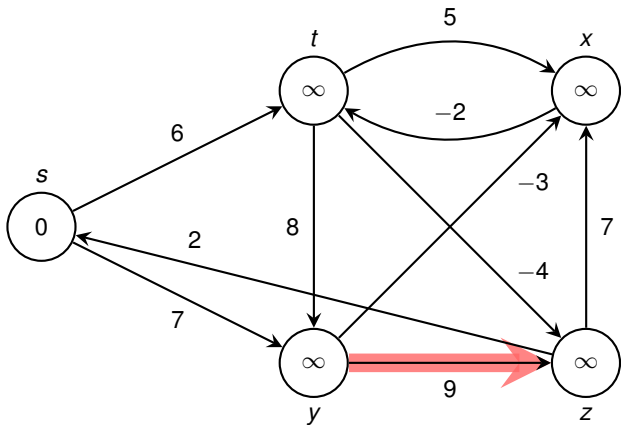
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

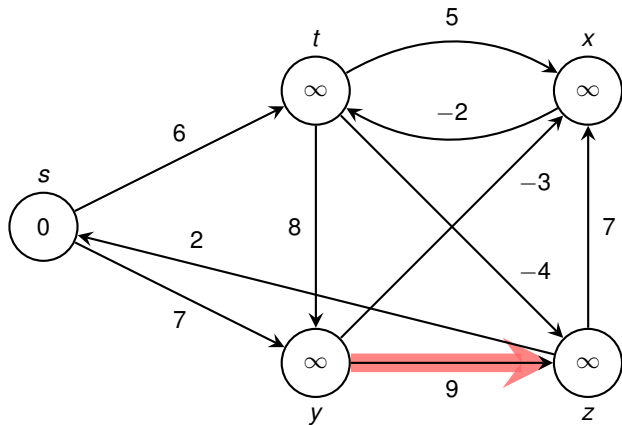
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

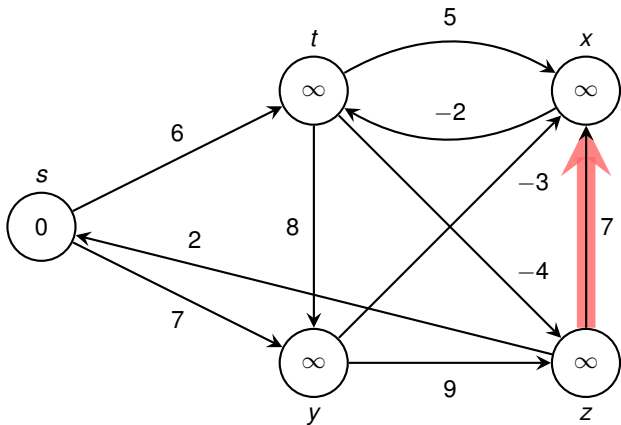
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

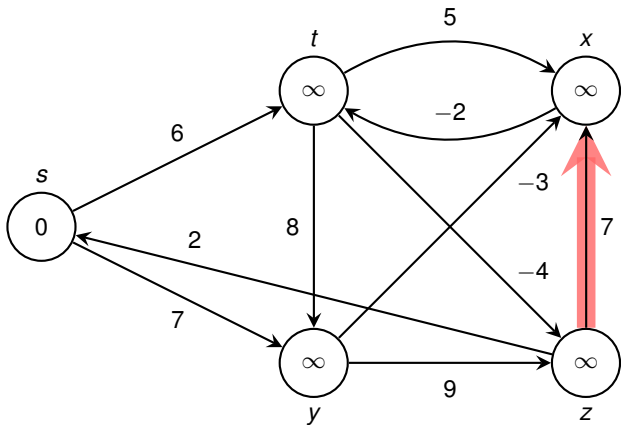
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

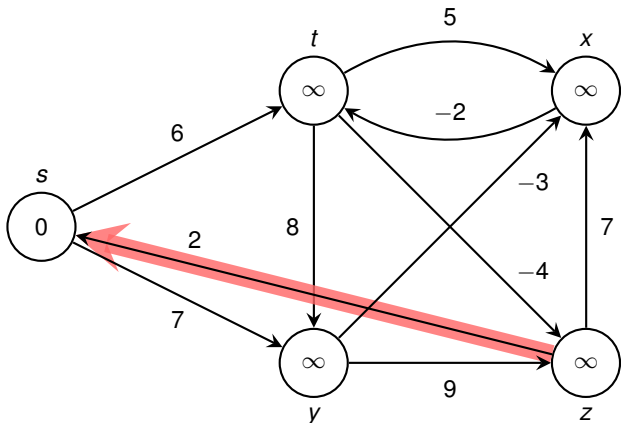
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

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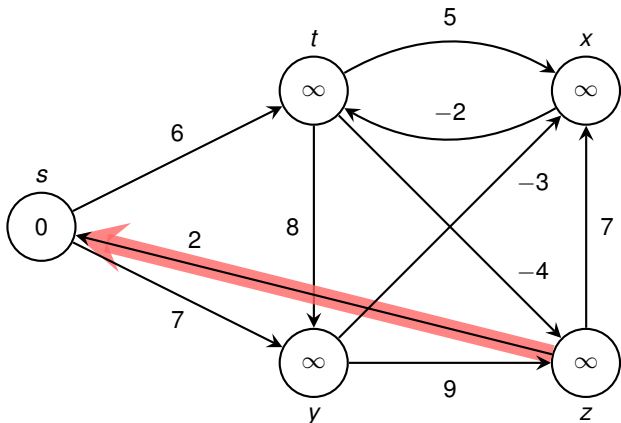
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

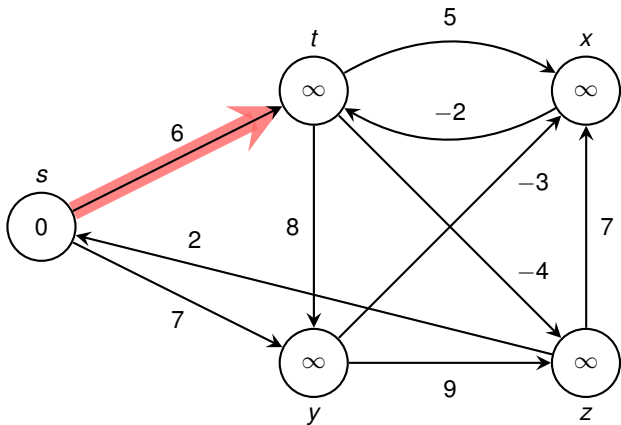
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

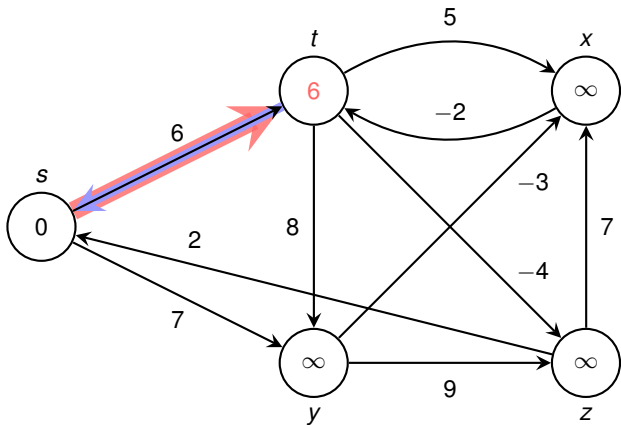
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

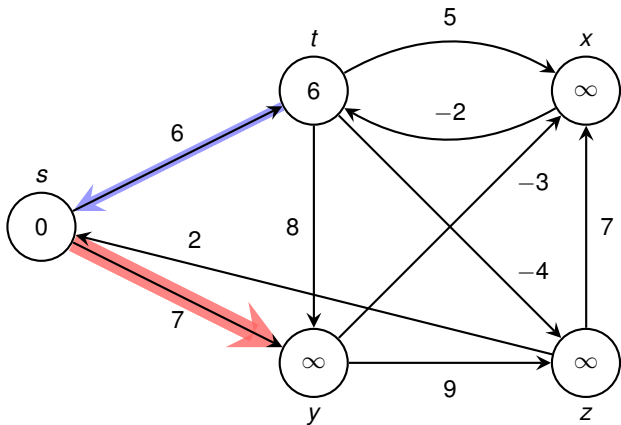
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

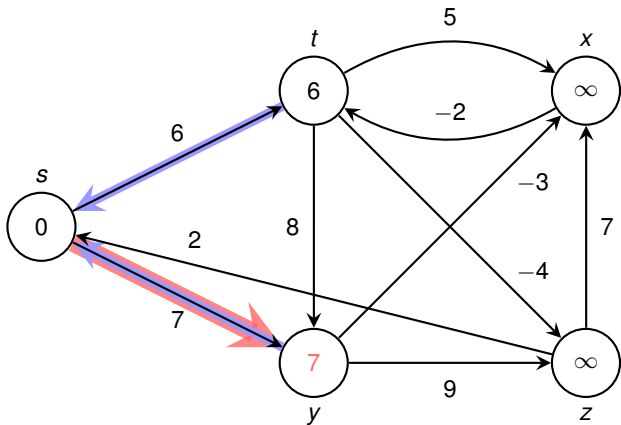
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 1

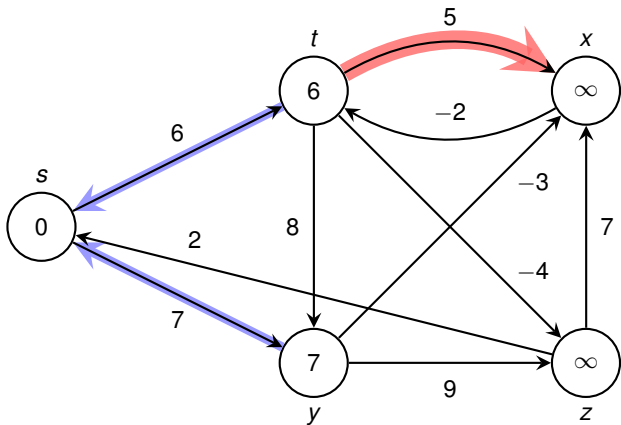
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

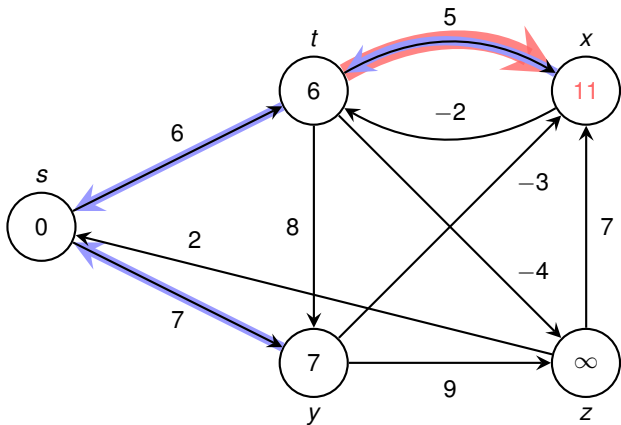
Relaxation Order: (t,x) , (t,y) , (t,z) , (x,t) , (y,x) , (y,z) , (z,x) , (z,s) , (s,t) , (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

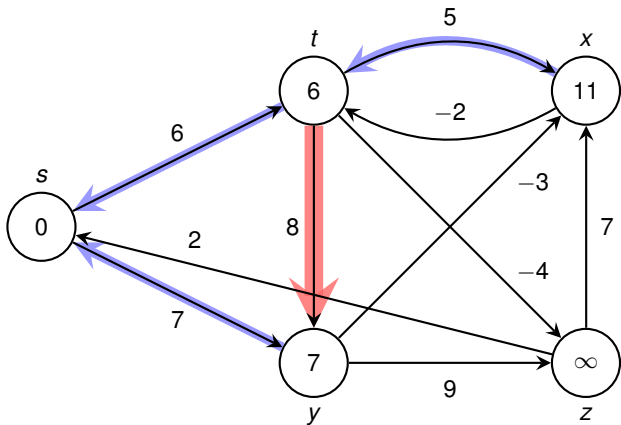
Relaxation Order: $(t,x), (t,y), (t,z), (x,t), (y,x), (y,z), (z,x), (z,s), (s,t), (s,y)$



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

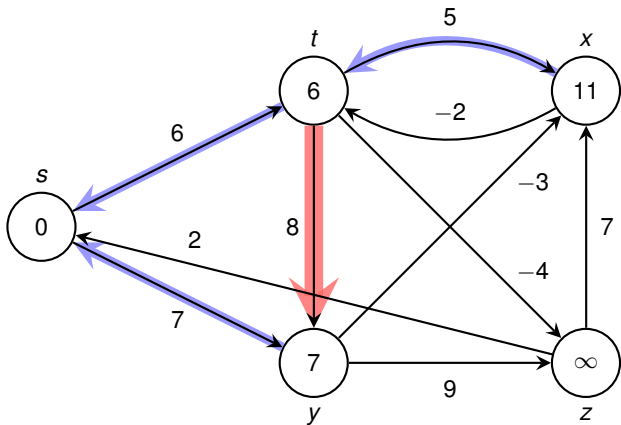
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

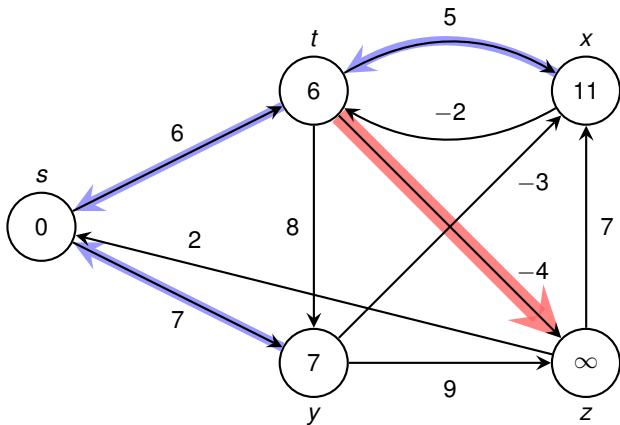
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

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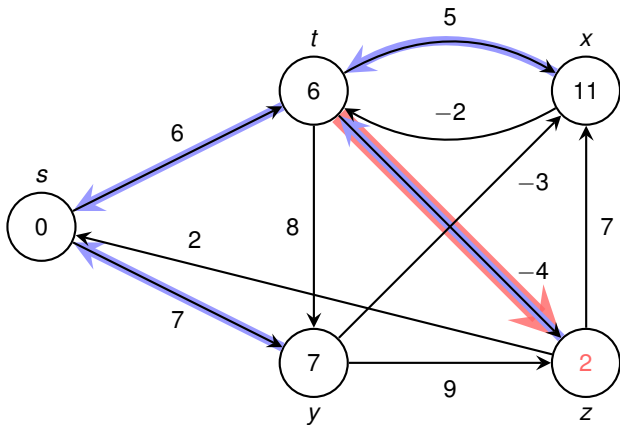
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Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

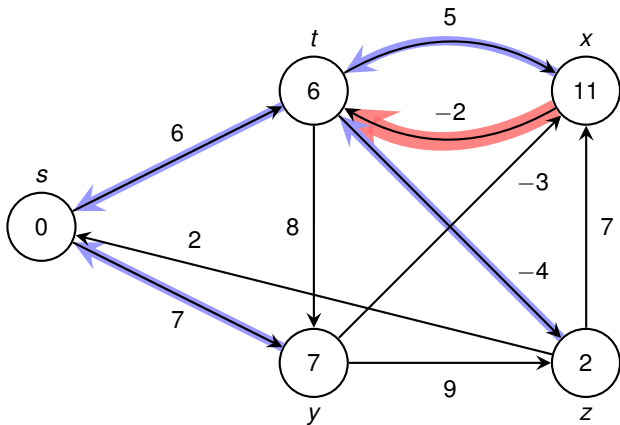
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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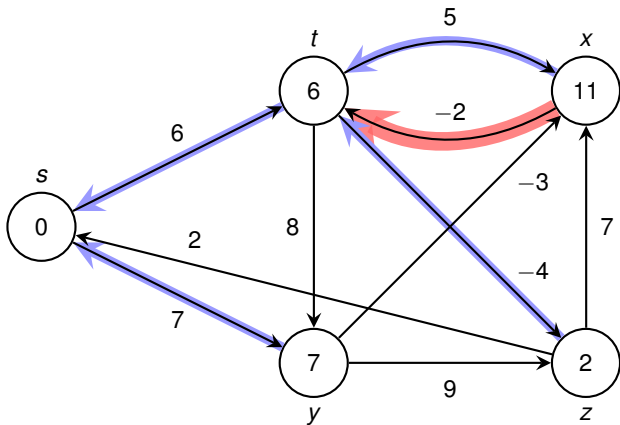
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

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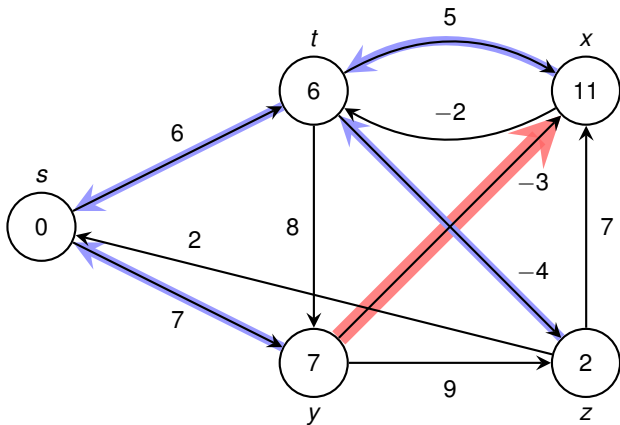
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

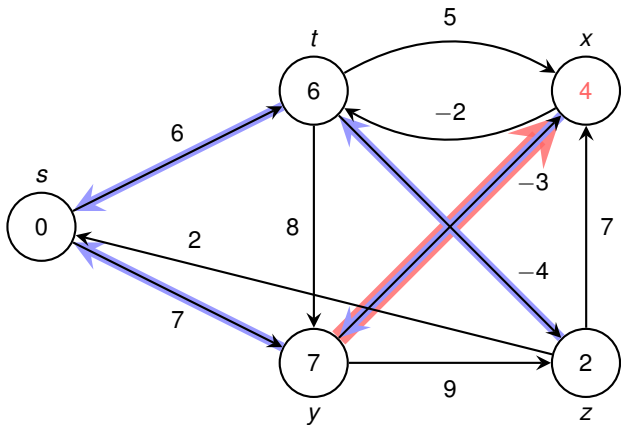
Relaxation Order: (t,x),(t,y),(t,z),(x,t), (y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

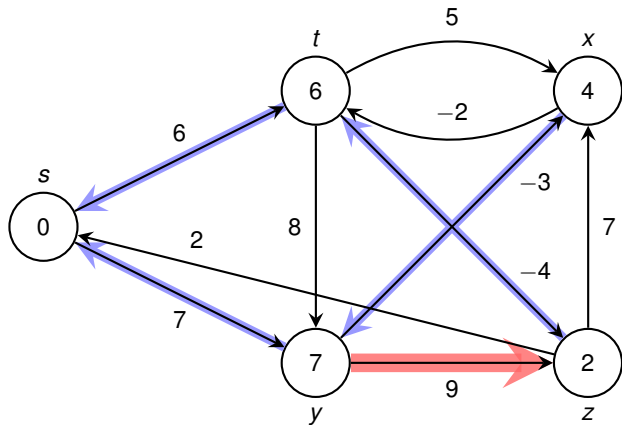
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

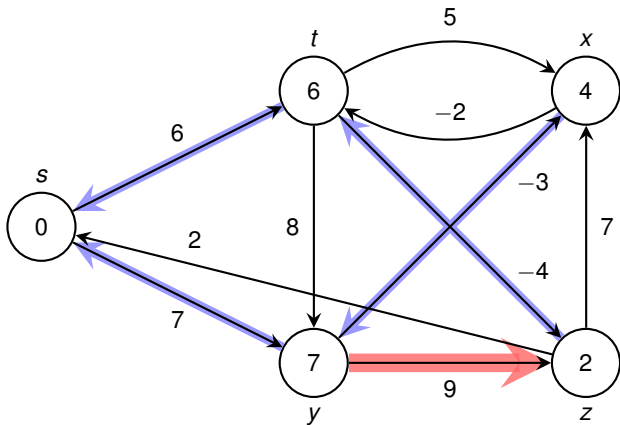
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

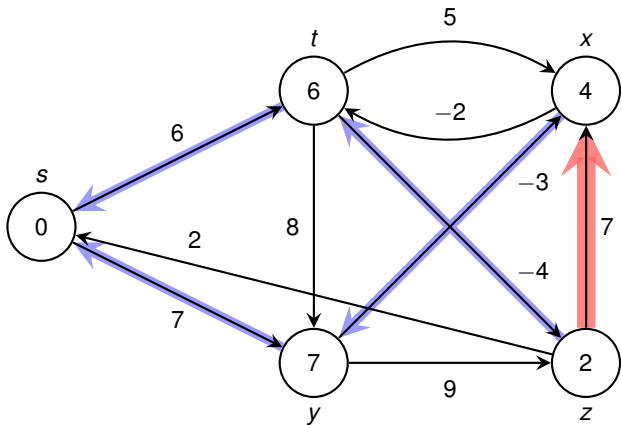
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

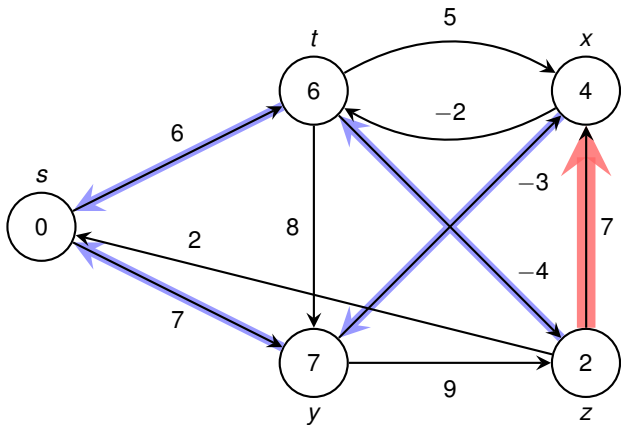
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

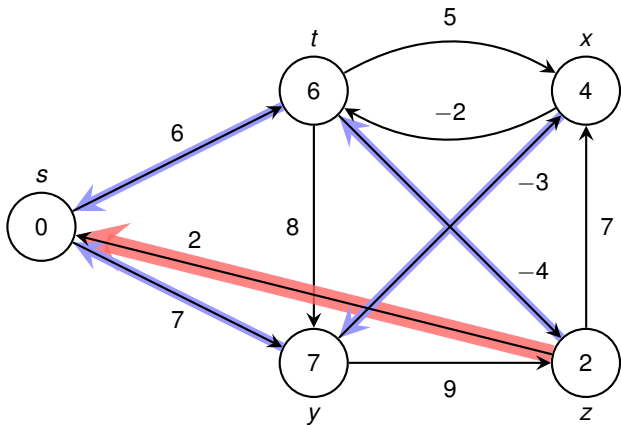
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

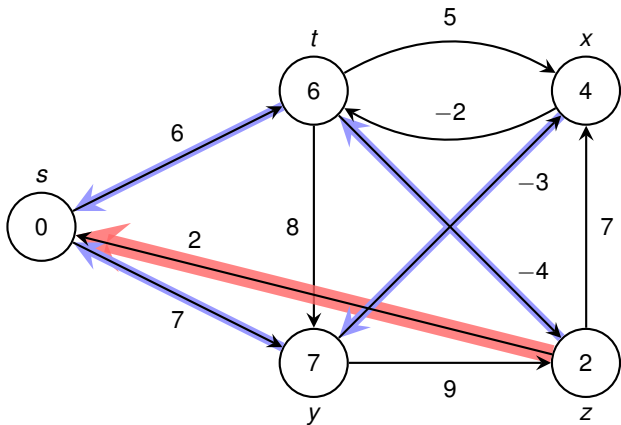
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

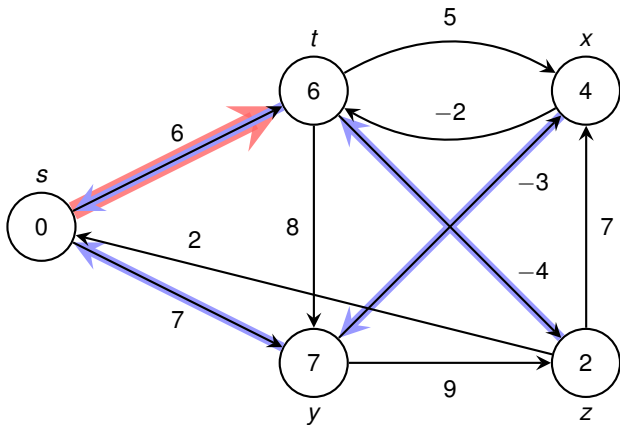
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

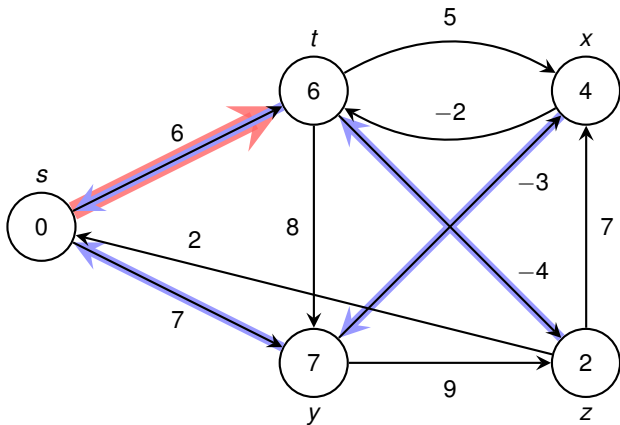
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

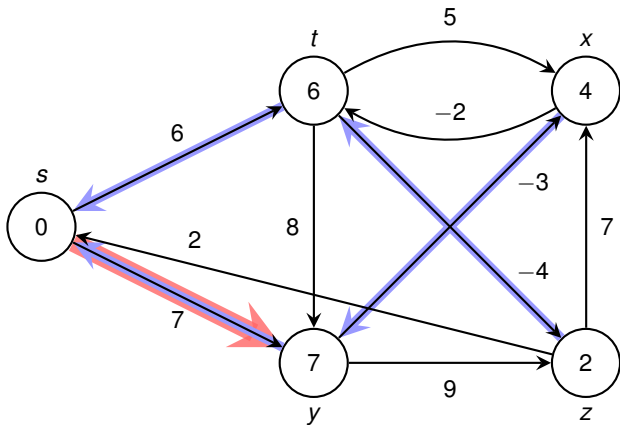
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

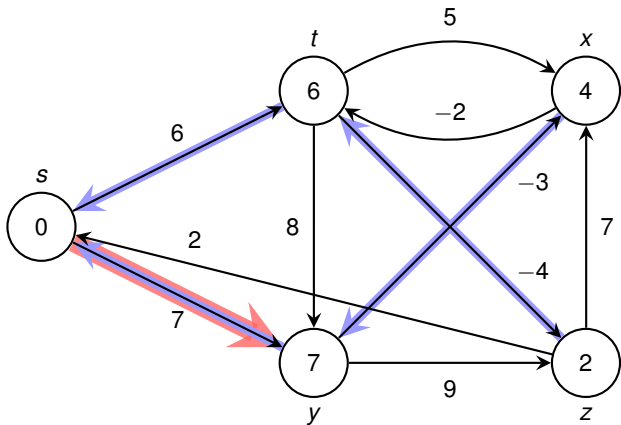
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 2

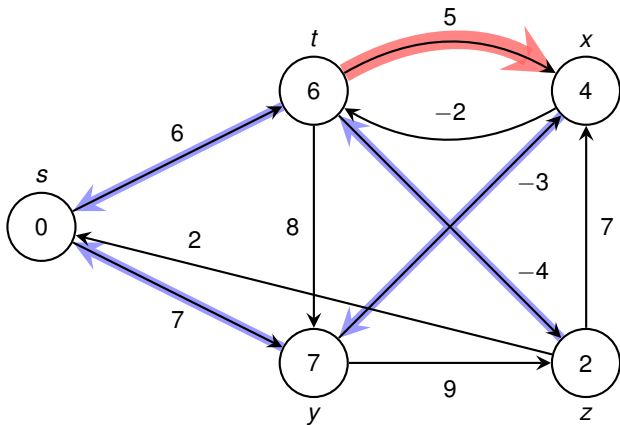
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

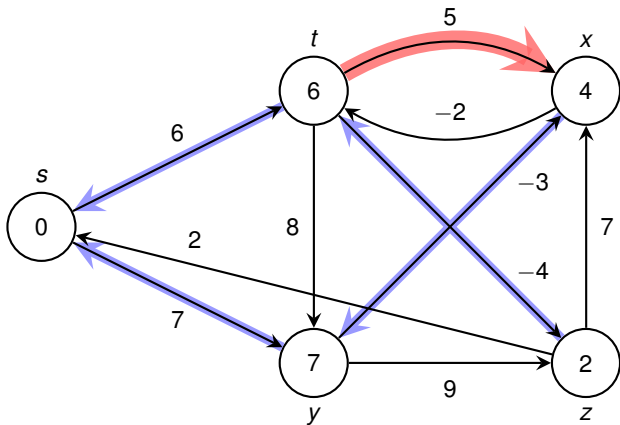
Relaxation Order: (t,x) , (t,y) , (t,z) , (x,t) , (y,x) , (y,z) , (z,x) , (z,s) , (s,t) , (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

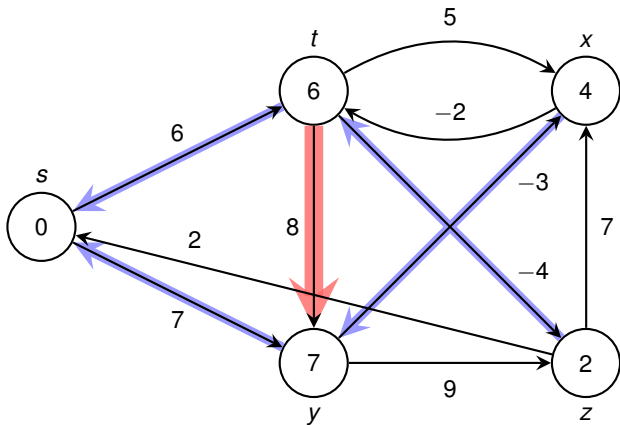
Relaxation Order: $(t,x), (t,y), (t,z), (x,t), (y,x), (y,z), (z,x), (z,s), (s,t), (s,y)$



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

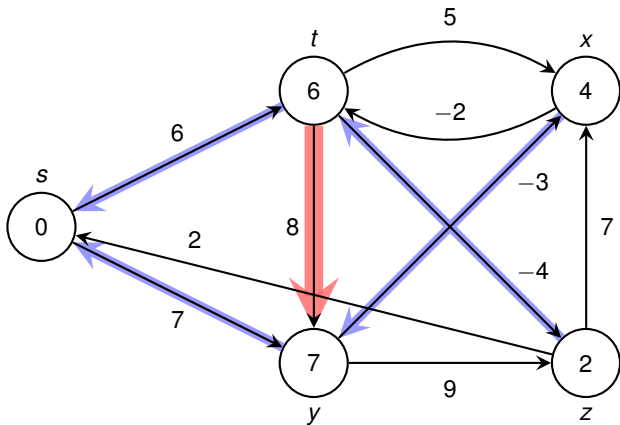
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

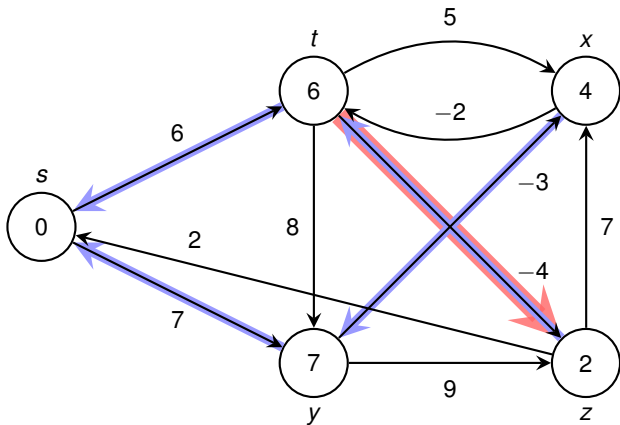
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

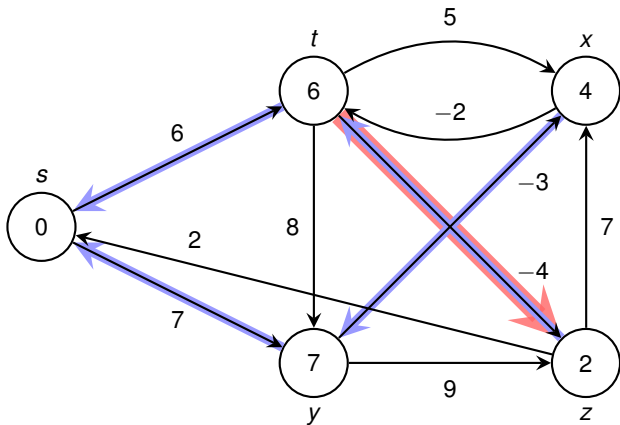
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

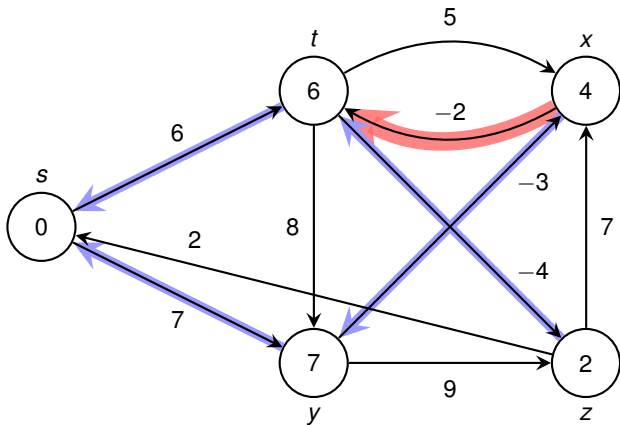
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

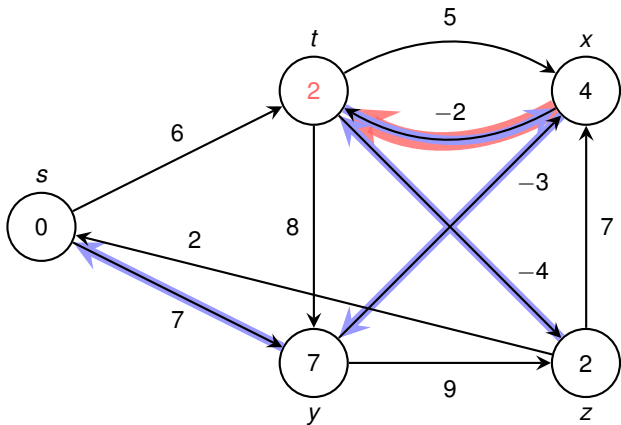
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

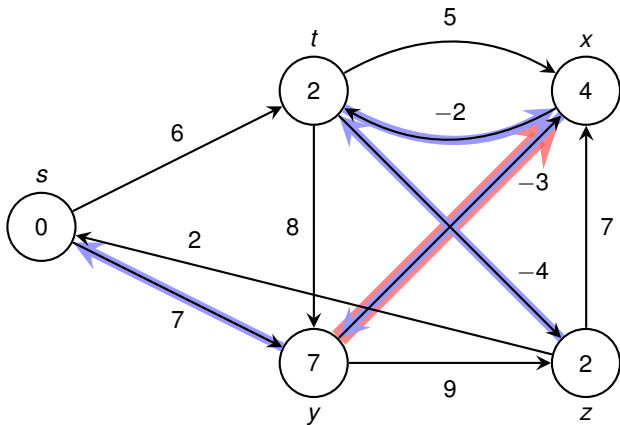
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

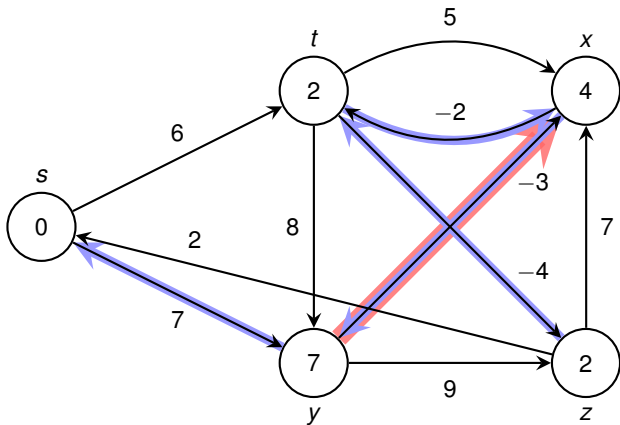
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

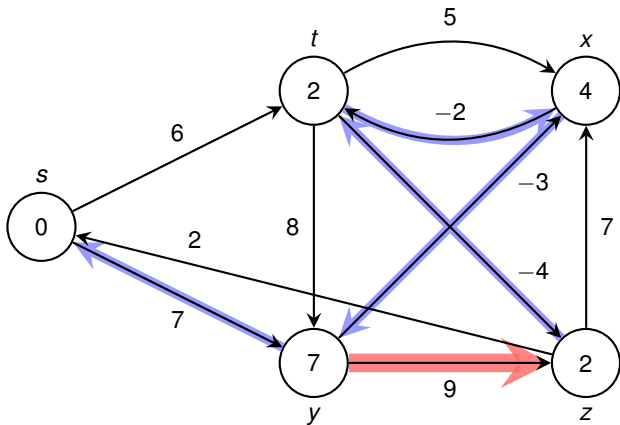
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

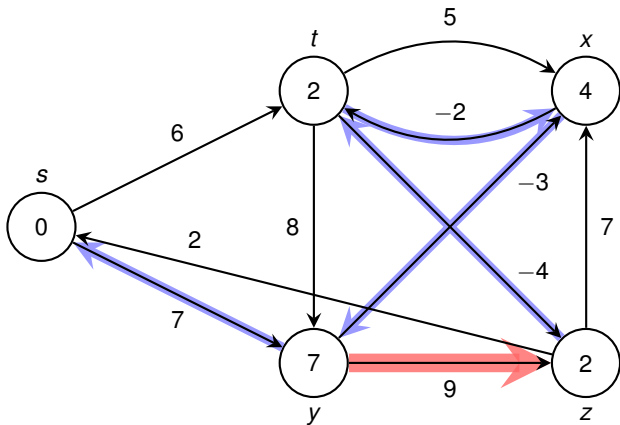
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

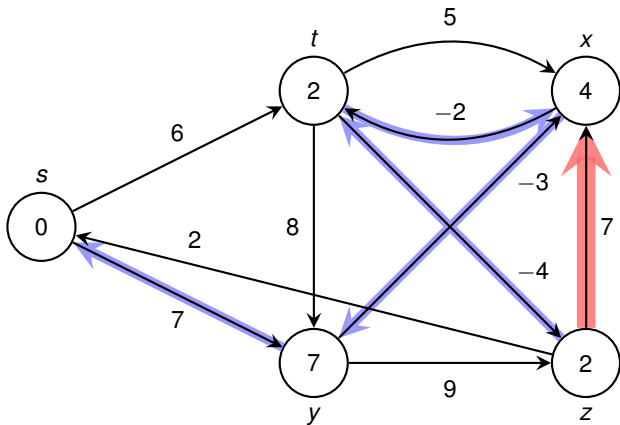
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

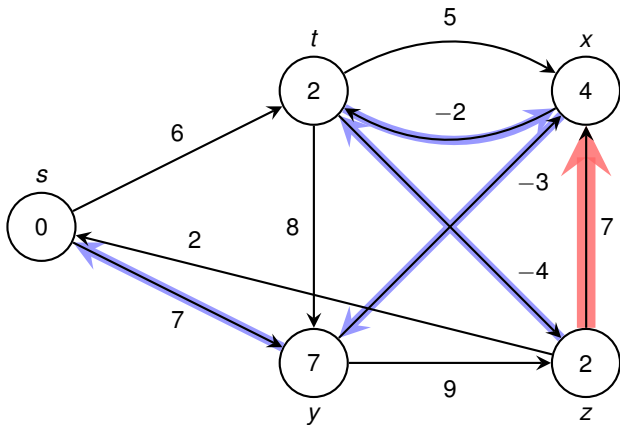
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

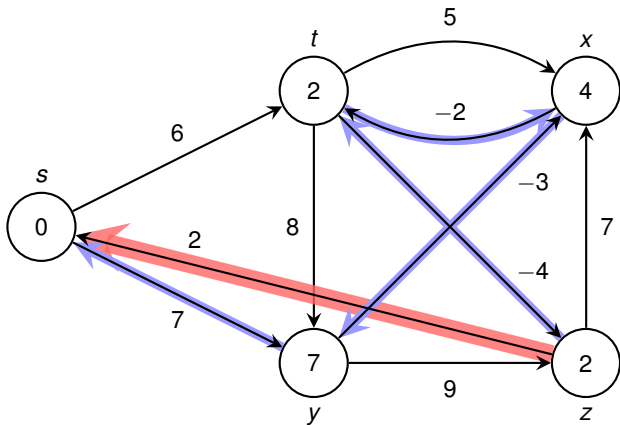
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

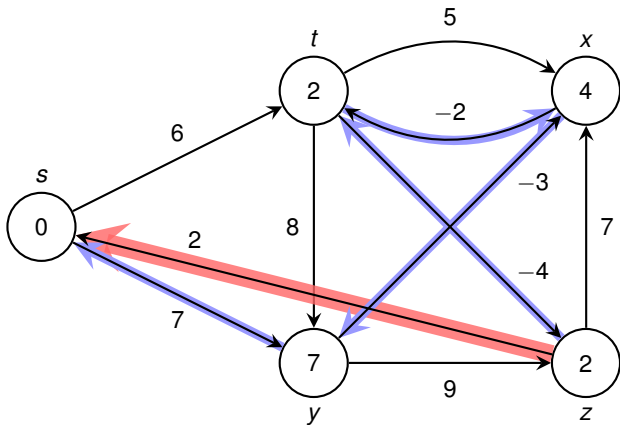
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

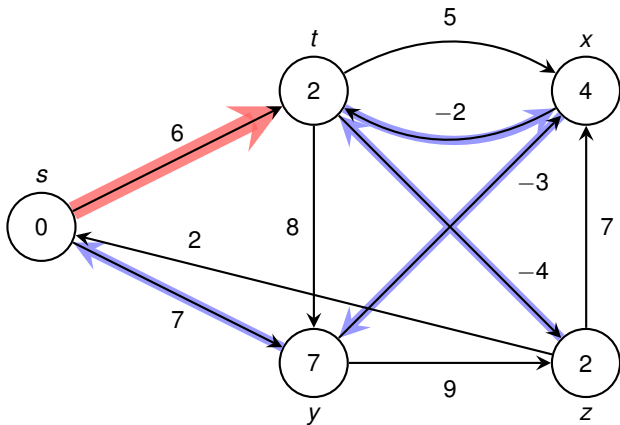
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

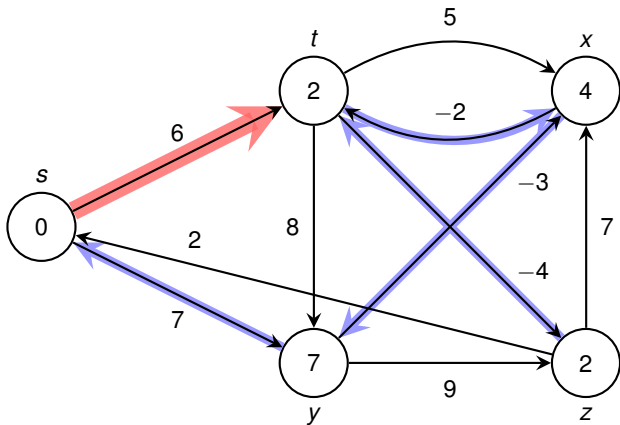
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

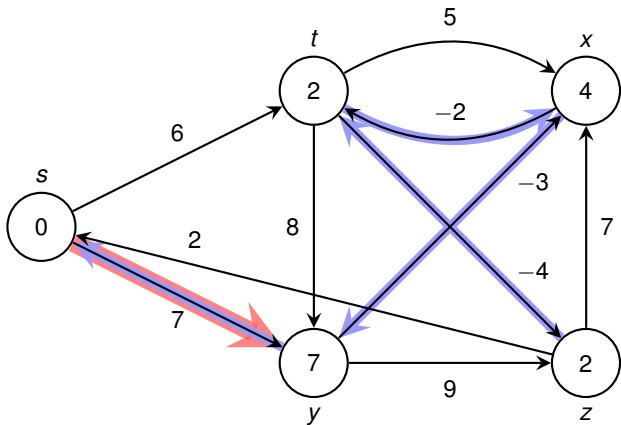
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

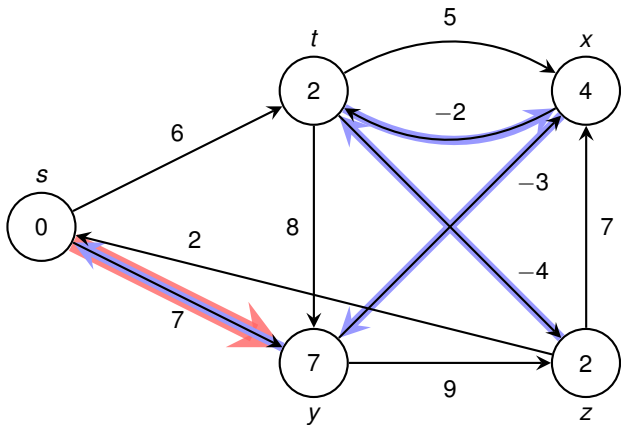
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 3

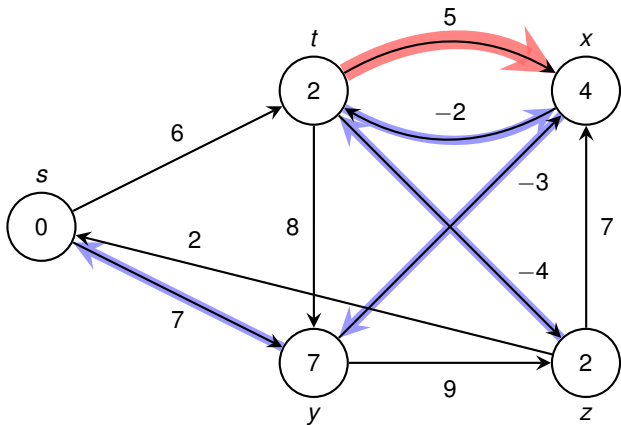
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

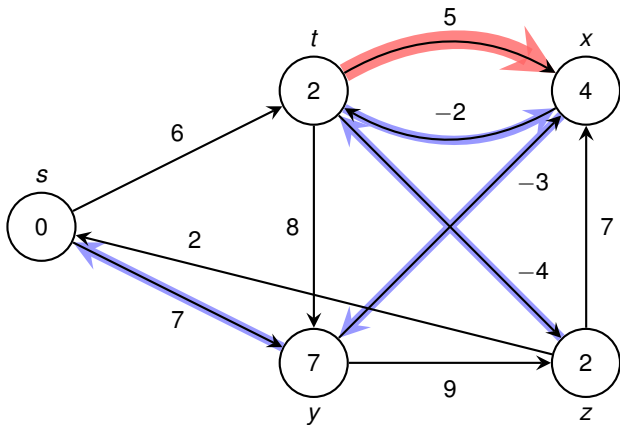
Relaxation Order: (t,x) , (t,y) , (t,z) , (x,t) , (y,x) , (y,z) , (z,x) , (z,s) , (s,t) , (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

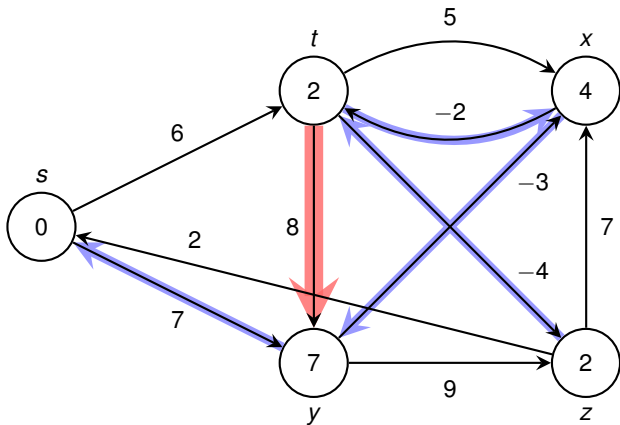
Relaxation Order: (t,x) , (t,y) , (t,z) , (x,t) , (y,x) , (y,z) , (z,x) , (z,s) , (s,t) , (s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

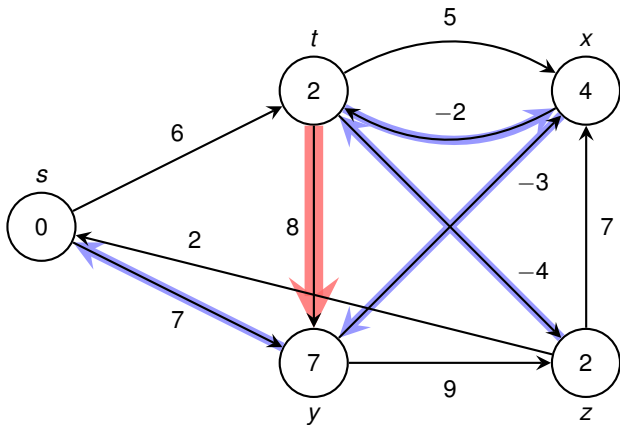
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

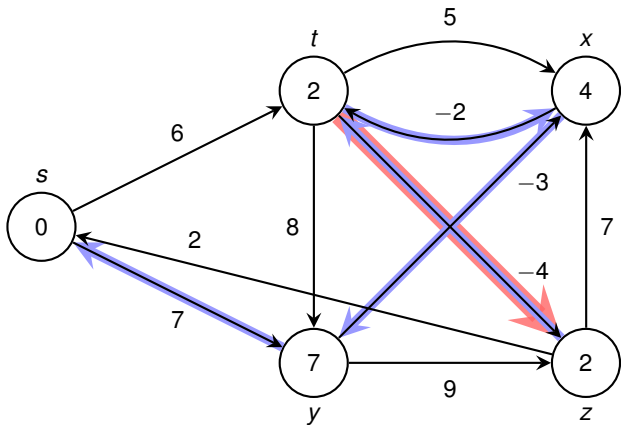
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

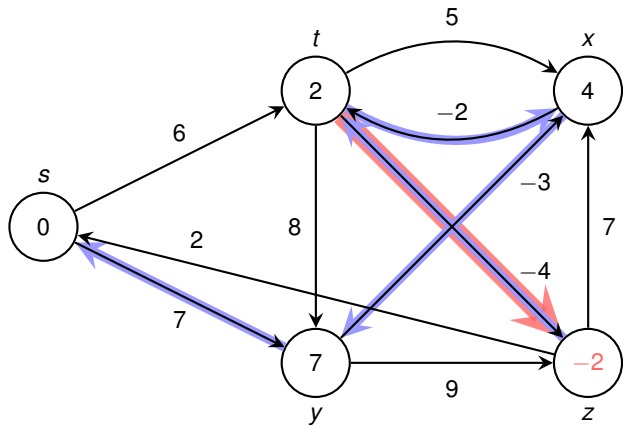
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

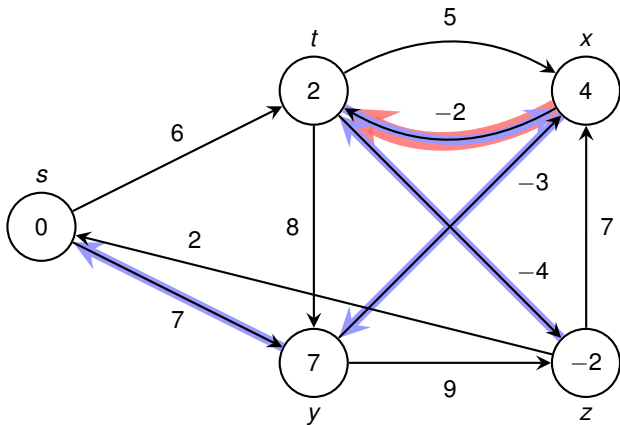
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

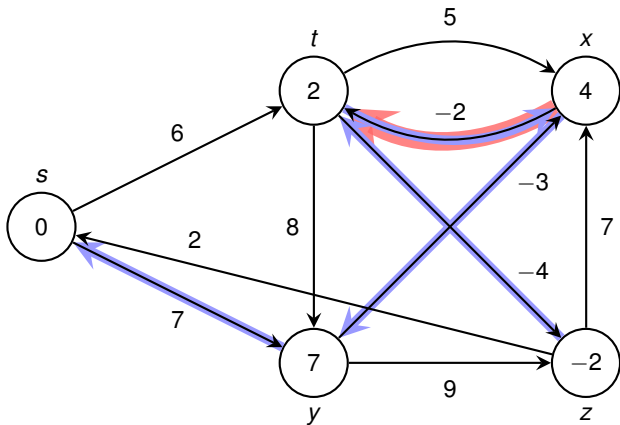
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

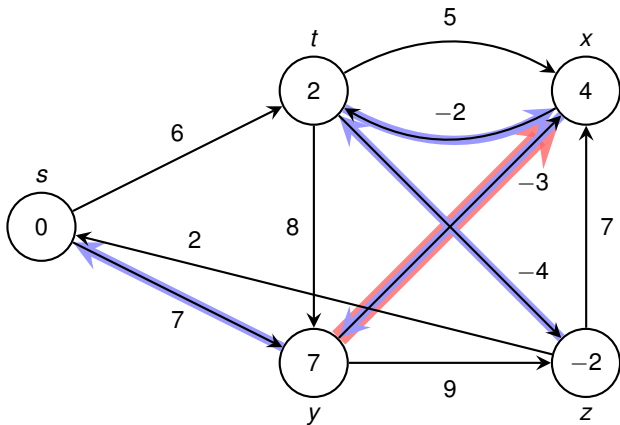
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

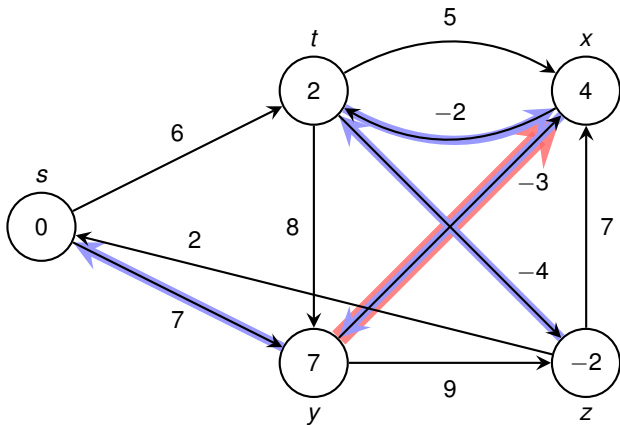
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

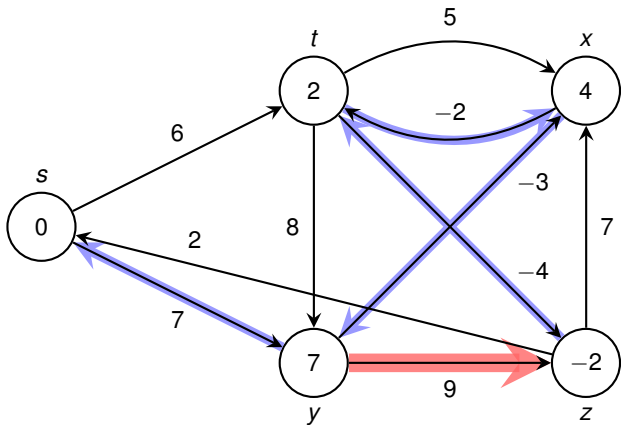
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

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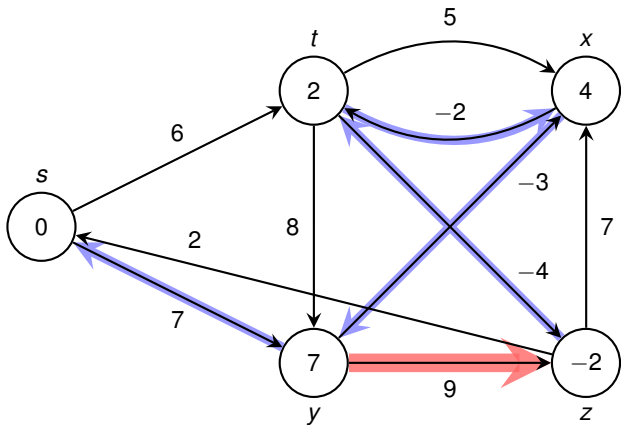
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

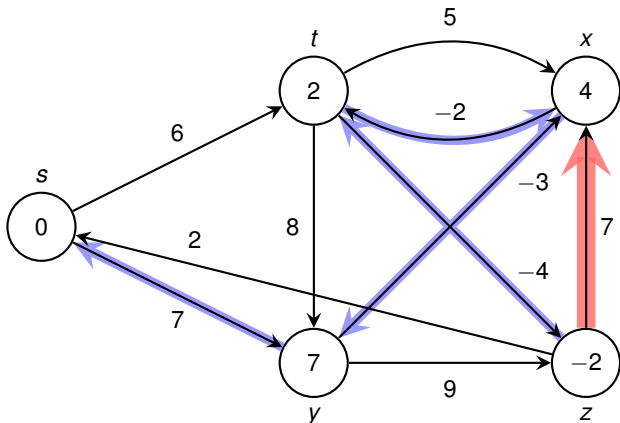
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

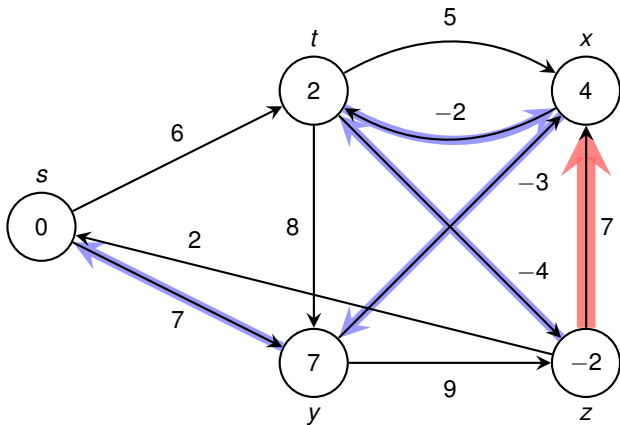
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

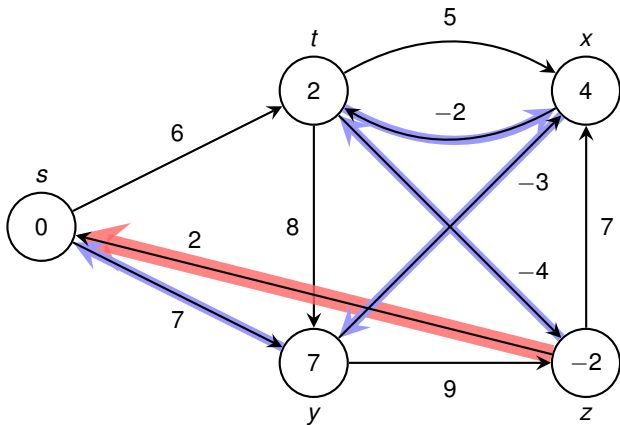
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

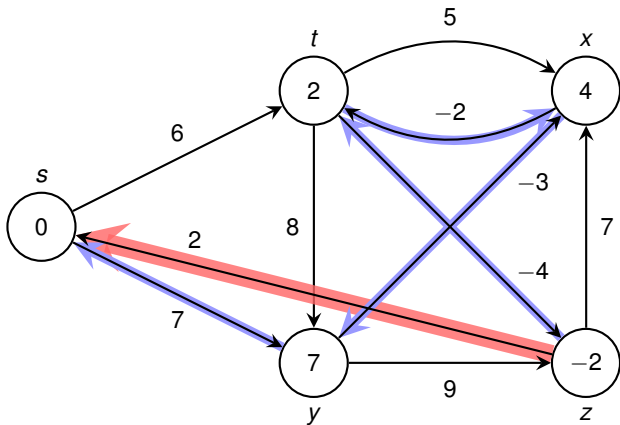
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x), **(z,s)**, (s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

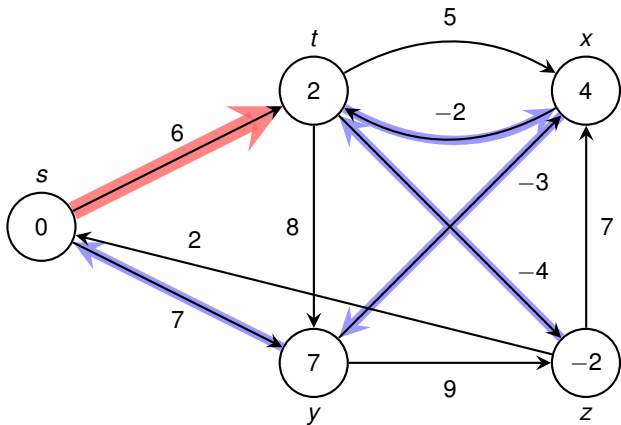
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



Complete Run of Bellman-Ford (Figure 24.4)

Pass: 4

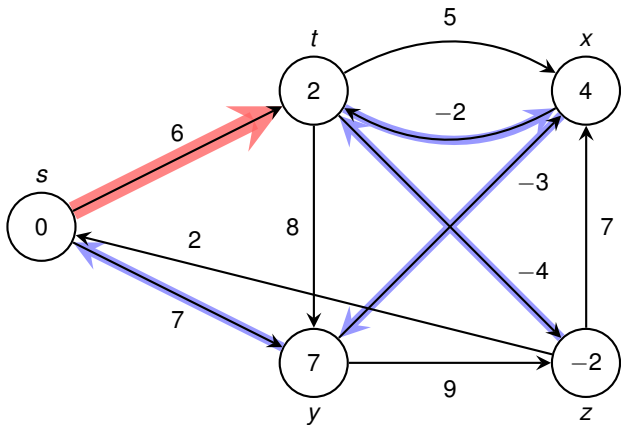
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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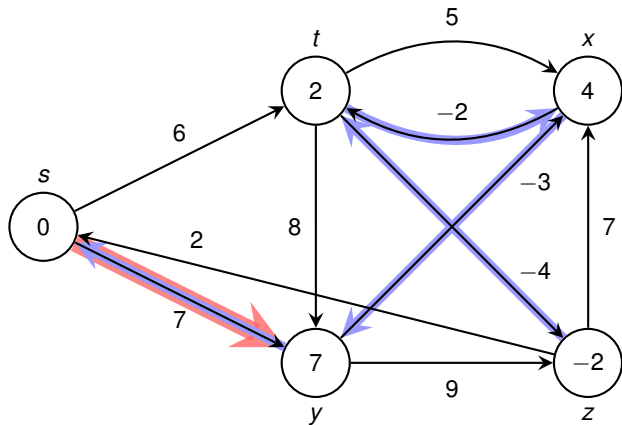
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t),(s,y)



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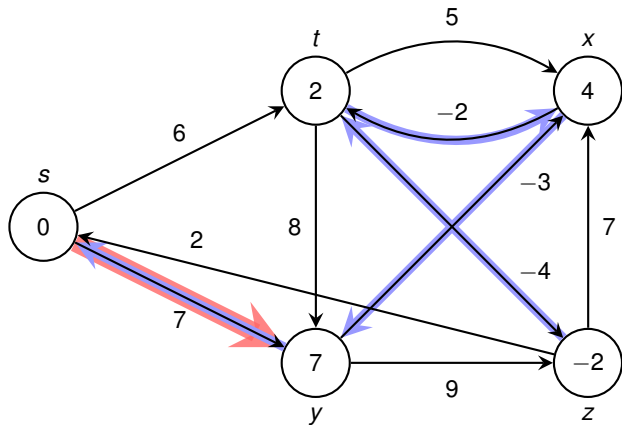
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



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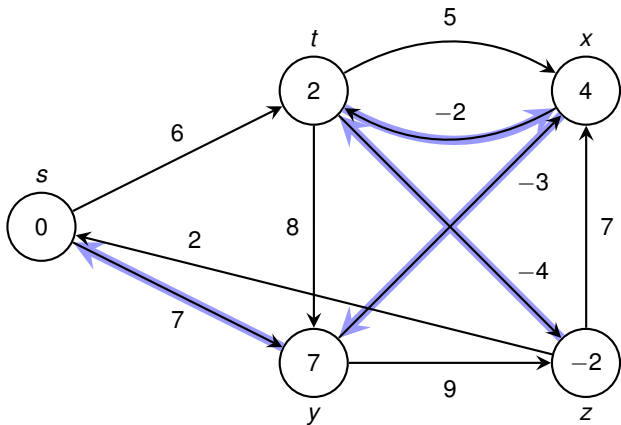
Relaxation Order: (t,x),(t,y),(t,z),(x,t),(y,x),(y,z),(z,x),(z,s),(s,t), (s,y)



Complete Run 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)

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)

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

- Let v be a vertex reachable from s



Bellman-Ford Algorithm: Correctness (1/2)

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- Let v be a vertex reachable from s
- Let $p = (v_0 = s, v_1, \dots, v_k = v)$ be a shortest path from s to v



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- Let v be a vertex reachable from s
- Let $p = (v_0 = s, v_1, \dots, v_k = v)$ be a shortest path from s to v
- p is simple, hence $k \leq |V| - 1$



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Proof that 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
- Let $p = (v_0 = s, v_1, \dots, v_k = v)$ be a shortest path from s to v
- p is simple, hence $k \leq |V| - 1$
- Path-Relaxation Property \Rightarrow after $|V| - 1$ passes, $v.d = v.\delta$

Proof that Bellman-Ford returns TRUE

- Need to prove: $v.d \leq u.d + w(u, v)$ for all edges



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



Bellman-Ford Algorithm: Correctness (2/2)

Theorem 24.3

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.

Proof:

- 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 $1 \leq i < k$,

$$v_i.d \leq v_{i-1}.d + w(v_{i-1}, v_i)$$



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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!** □

