

Single-Source Shortest Paths

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single-source shortest-paths problem

- **single-source shortest-paths problem:**
- Given a graph $G = (V, E)$, we want to find a shortest path from a given source vertex $s \in V$ to each vertex $v \in V$.
- *The algorithm for the single-source problem can solve many other problems, including the following variants.*
 - Single-destination shortest-paths problem
 - Single-pair shortest-path problem
 - All-pairs shortest-paths problem

Outline

- **Bellman-Ford algorithm.**
 - uses **dynamic programming**
- Single-source shortest paths in directed acyclic graphs
- **Dijkstra's algorithm**
 - uses the **greedy approach**

optimal-substructure property

Lemma 24.1: Let $p = \langle v_1, v_2, \dots, v_k \rangle$ be a SP from v_1 to v_k . Then, $p_{ij} = \langle v_i, v_{i+1}, \dots, v_j \rangle$ is a SP from v_i to v_j , where $1 \leq i \leq j \leq k$.

So, we have the **optimal-substructure property**.

Bellman-Ford's algorithm uses **dynamic programming**.

Dijkstra's algorithm uses the **greedy approach**.

Let $\delta(u, v)$ = weight of SP from u to v .

Corollary: Let p = SP from s to v , where $p = s \xrightarrow{p'} u \rightarrow v$. Then, $\delta(s, v) = \delta(s, u) + w(u, v)$.

Lemma 24.10: Let $s \in V$. For all edges $(u, v) \in E$, we have $\delta(s, v) \leq \delta(s, u) + w(u, v)$.

Relaxation

- For each vertex $v \in V$, we maintain an attribute $v.d$
- $v.d$:
 - ♦ is an upper bound on the weight of a shortest path from source s to v .
- $v.\pi$:
 - ♦ a predecessor vertex of v .

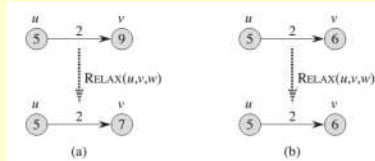
Relaxation

- Algorithms keep track of $d[v]$, $\pi[v]$.
- **Initialized** as follows:

```
INITIALIZE-SINGLE-SOURCE( $G, s$ )
1  for each vertex  $v \in G.V$ 
2       $v.d = \infty$ 
3       $v.\pi = \text{NIL}$ 
4   $s.d = 0$ 
```

Relaxation

- The process of relaxing an edge (u, v) consists of testing whether we can improve the shortest path to v found so far by going through u and, if so, updating $v.d$ and $v.\pi$.



Relaxation

```

RELAX( $u, v, w$ )
1  if  $v.d > u.d + w(u, v)$ 
2     $v.d = u.d + w(u, v)$ 
3     $v.\pi = u$ 
    
```

a relaxation step on edge (u, v) in $O(1)$ time

Properties of Relaxation

- $d[v]$, if not ∞ , is the length of *some* path from s to v .
- $d[v]$ either stays the same or decreases with time
- Therefore, if $d[v] = \delta(s, v)$ at any time, this holds thereafter
- Note that $d[v] \geq \delta(s, v)$ always
- After i iterations of relaxing on all (u, v) , if the shortest path to v has i edges, then $d[v] = \delta(s, v)$.

Properties of Relaxation

Consider any algorithm in which $d[v]$, and $\pi[v]$ are first initialized by calling $\text{Initialize}(G, s)$ [s is the source], and are only changed by calling Relax . We have:

Lemma 24.11: $(\forall v :: d[v] \geq \delta(s, v))$ is an invariant.

Implies $d[v]$ doesn't change once $d[v] = \delta(s, v)$.

Proof:

$\text{Initialize}(G, s)$ establishes invariant. If call to $\text{Relax}(u, v, w)$ changes $d[v]$, then it establishes:

$$\begin{aligned}
 d[v] &= d[u] + w(u, v) \\
 &\geq \delta(s, u) + w(u, v) && \text{, invariant holds before call.} \\
 &\geq \delta(s, v) && \text{, by Lemma 24.10.}
 \end{aligned}$$

Corollary 24.12: If there is no path from s to v , then $d[v] = \delta(s, v) = \infty$ is an invariant.

- Bellman-Ford returns a compact representation of the set of shortest paths from s to all other vertices in the graph reachable from s . This is contained in the predecessor subgraph.

Predecessor Subgraph

Lemma 24.16: Assume given graph G has no negative-weight cycles reachable from s . Let G_π = predecessor subgraph. G_π is always a tree with root s (i.e., this property is an invariant).

Proof:

Two proof obligations:

- (1) G_π is acyclic.
- (2) There exists a unique path from source s to each vertex in V_π .

Proof of (1):

Suppose there exists a cycle $c = \langle v_0, v_1, \dots, v_k \rangle$, where $v_0 = v_k$. We have $\pi[v_i] = v_{i-1}$ for $i = 1, 2, \dots, k$.

Assume relaxation of (v_{k-1}, v_k) created the cycle.

We show cycle has a negative weight.

Note: Cycle must be reachable from s . (Why?)

Bellman-Ford Algorithm

- Bellman-Ford algorithm solves the single-source shortest-paths problem in the general case in which *edge weights may be negative*.
- returns *a boolean value indicating* whether or not there is a negative-weight cycle that is reachable from the source.
- *If there is such a cycle*, the algorithm indicates that *no solution exists*.
- *If there is no such cycle*, the algorithm produces the shortest paths and their weights.

Bellman-Ford Algorithm

```

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

Bellman-Ford Algorithm

INITIALIZE-SINGLE-SOURCE(G, s)

```

1  for each vertex  $v \in G.V$ 
2       $v.d = \infty$ 
3       $v.\pi = \text{NIL}$ 
4   $s.d = 0$ 
    
```

RELAX(u, v, w)

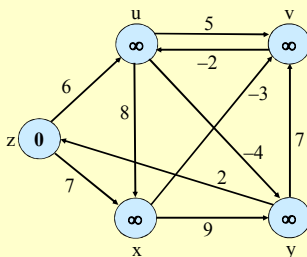
```

1  if  $v.d > u.d + w(u, v)$ 
2       $v.d = u.d + w(u, v)$ 
3       $v.\pi = u$ 
    
```

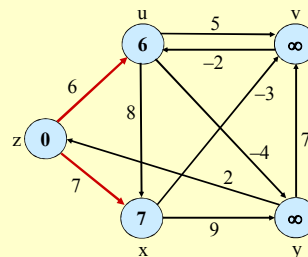
Bellman-Ford Algorithm

- So if Bellman-Ford has not converged after $V(G) - 1$ iterations, then there cannot be a shortest path tree, so there must be a negative weight cycle.

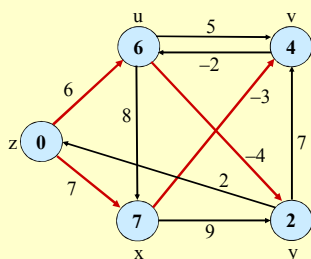
Example



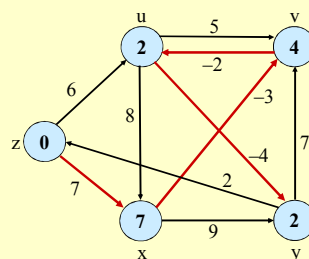
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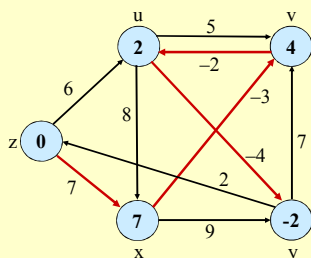
Example



Example



Example



dynamic programming

Note: This is essentially **dynamic programming**.

Let $d(i, j)$ = cost of the shortest path from s to i that is at most j hops.

$$d(i, j) = \begin{cases} 0 & \text{if } i = s \wedge j = 0 \\ \infty & \text{if } i \neq s \wedge j = 0 \\ \min(\{d(k, j-1) + w(k, i) : i \in \text{Adj}(k)\} \cup \{d(i, j-1)\}) & \text{if } j > 0 \end{cases}$$

		i →				
		z	u	v	x	y
j ↓	0	0	∞	∞	∞	∞
	1	0	6	∞	7	∞
	2	0	6	4	7	2
	3	0	2	4	7	2
	4	0	2	4	7	-2

Analysis of Bellman-Ford Algorithm

The Bellman-Ford algorithm runs in time $O(VE)$, since the initialization in line 1 takes $\Theta(V)$ time, each of the $|V| - 1$ passes over the edges in lines 2-4 takes $\Theta(E)$ time, and the **for** loop of lines 5-7 takes $O(E)$ time.

Single-source shortest paths in directed acyclic graphs

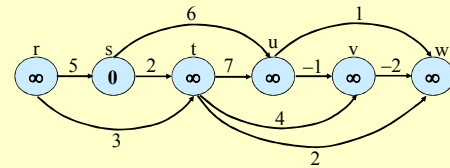
Single-source shortest paths in DAGs

• Shortest paths are always well defined in a dag, since even **if there are negative-weight edges, no negative-weight cycles can exist**

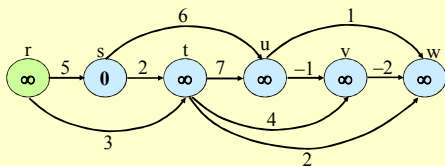
DAG-SHORTEST-PATHS(G, w, s)

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

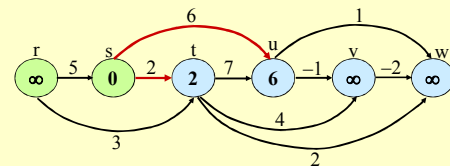
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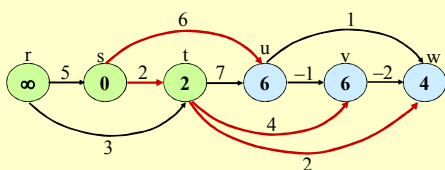
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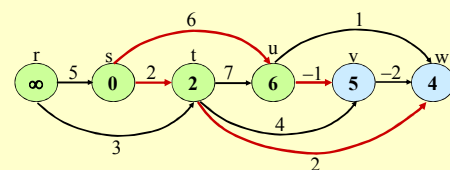
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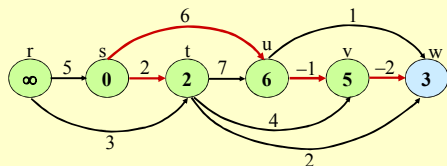
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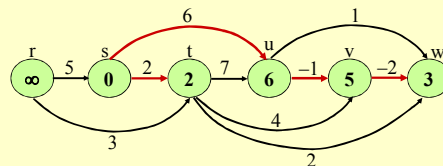
Example



Example



Example



- can compute shortest paths from a single source in $\Theta(V+E)$ time.

Single-source shortest paths in DAGs

■ Analysis

topological sort of line 1 takes $\Theta(V + E)$ time

- The for loop of lines 3–5 makes one iteration per vertex.
- Altogether, the for loop of lines 4–5 relaxes each edge exactly once.
- Because each iteration of the inner for loop takes $O(1)$ time, the total running time is $\Theta(V + E)$
- which is *linear in the size of an adjacency-list* representation of the graph.

Jim Anderson

Dijkstra's Algorithm

Dijkstra's Algorithm

- Assumes **no negative-weight edges**.
- Greedy approach
- Maintains a set S of vertices whose *Shortest Path from s has been determined*.
- Repeatedly selects u in $V-S$ with minimum *Shortest Path* estimate (greedy choice).
- Store $V-S$ in priority queue Q .

Dijkstra's Algorithm

```

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

Dijkstra's Algorithm

INITIALIZE-SINGLE-SOURCE(G, s)

```

1  for each vertex  $v \in G.V$ 
2     $v.d = \infty$ 
3     $v.\pi = \text{NIL}$ 
4   $s.d = 0$ 

```

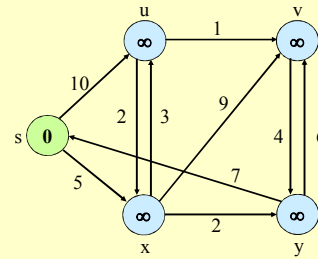
RELAX(u, v, w)

```

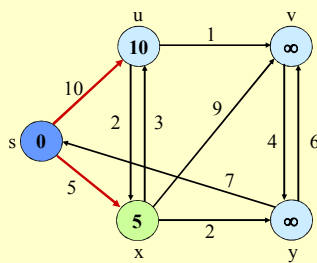
1  if  $v.d > u.d + w(u, v)$ 
2     $v.d = u.d + w(u, v)$ 
3     $v.\pi = u$ 

```

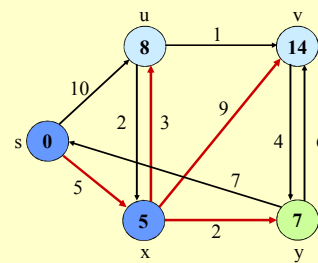
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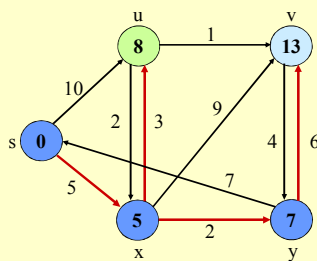
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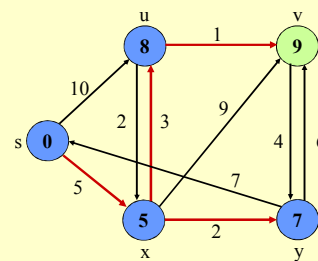
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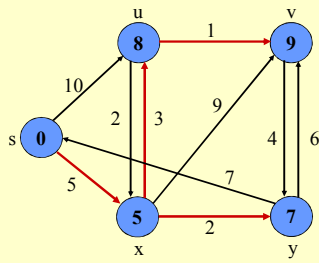
Example



Example



Example



Complexity

Running time is

$O(V^2)$ using linear array for priority queue.

$O((V + E) \lg V)$ using binary heap.

$O(V \lg V + E)$ using Fibonacci heap.

(See book.)