

Single-Source Shortest Paths

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2017 Spring

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Outline

- **Shortest Paths**
- Shortest-Paths Properties
- The Bellman-Ford Algorithm
- Single-Source Shortest Paths in Directed Acyclic Graphs
- Dijkstra's Algorithm
- Difference Constraints and Shortest Paths

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Overview (1/2)

How to find the shortest route between two points on a map?

- Input:
 - Directed graph $G = (V, E)$
 - Weight function $w : E \rightarrow \mathbb{R}$
- Weight of path $p = \langle v_0, v_1, \dots, v_k \rangle$
$$= \sum_{i=1}^k w(v_{i-1}, v_i)$$
 - = sum of edge weights on path p .

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Overview (2/2)

- Shortest-path weight u to v :

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

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

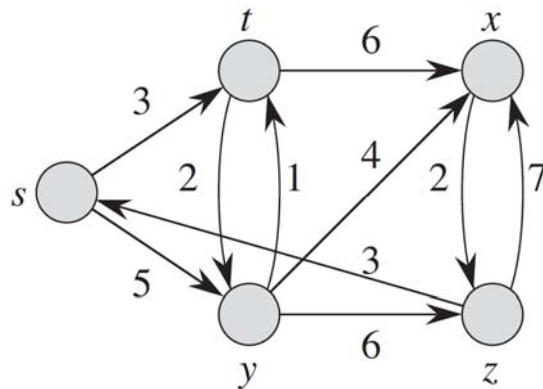
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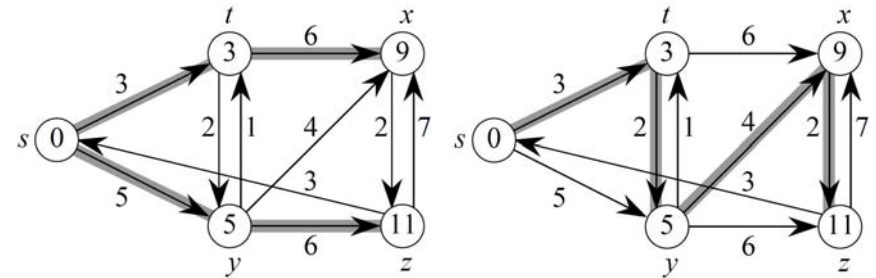
Example (1/2)

- Find shortest paths from s



Example (2/2)

- Shortest paths from s



- This example shows that the shortest path might not be unique.
- It also shows that when we look at shortest paths from one vertex to all other vertices, the shortest paths are organized as a tree.

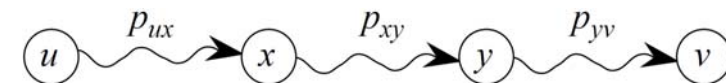
Variants

- Single-source:** Find shortest paths from a given source vertex $s \in V$ to every vertex $v \in V$.
- Single-destination:** Find shortest paths to a given destination vertex.
- Single-pair:** Find shortest path from u to v . If we solve the single-source problem with source vertex u , we solve this problem also.
- All-pairs:** Find shortest path from u to v for all $u, v \in V$. We'll see algorithms for all-pairs in the next chapter.

Optimal Substructure (1/2)

- Lemma 24.1** Any subpath of a shortest path is a shortest path.

Proof. Cut-and-paste.

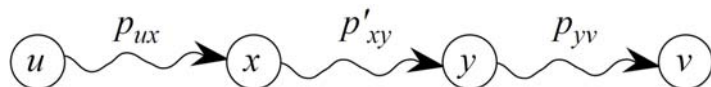


Suppose this path p is a shortest path from u to v . Then $\delta(u, v) = w(p) = w(p_{ux}) + w(p_{xy}) + w(p_{yv})$.

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

Optimal Substructure (2/2)

Construct p' :



Then

$$\begin{aligned}
 w(p') &= w(p_{ux}) + w(p'_{xy}) + w(p_{yv}) \\
 &< w(p_{ux}) + w(p_{xy}) + w(p_{yv}) \\
 &= w(p) .
 \end{aligned}$$

Contradicts the assumption that p is a shortest path.

Negative-Weight Edges

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

Cycles

Shortest paths can't contain cycles:

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

Output of Single-Source Shortest-Path Algorithm

For each vertex $v \in V$:

- $v.d = \delta(s, v)$.
 - Initially, $v.d = \infty$.
 - Reduces as algorithms progress. But always maintain $v.d \geq \delta(s, v)$.
 - Call $v.d$ a **shortest-path estimate**.
- $v.\pi$ = predecessor of v on a shortest path from s .
 - If no predecessor, $v.\pi = \text{NIL}$.
 - π induces a tree – **shortest-path tree**.
 - We won't prove properties of π in lecture – see text.

Initialization

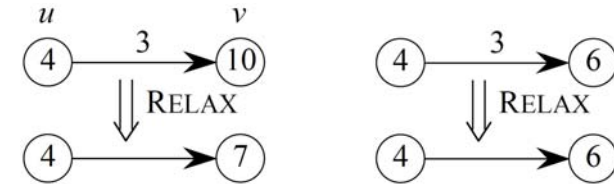
- All the shortest-paths algorithms start with INIT-SINGLE-SOURCE.

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 (1/2)

- Can we improve the shortest-path estimate for v by going through u and taking (u, v) ?



RELAX(u, v, w)

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

Relaxation (2/2)

For all the single-source shortest-paths algorithms we'll look at,

- start by calling INIT-SINGLE-SOURCE,
- then relax edges.
- ➡ The algorithms differ in the order and how many times they relax each edge.

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Shortest-Paths Properties (1/2)

- **Lemma 24.10 (Triangle inequality)**
For any edge $(u, v) \in E$, we have $\delta(s, v) \leq \delta(s, u) + w(u, v)$.
- **Lemma 24.11 (Upper-bound property)**
Always have $v.d \geq \delta(s, v)$ for all v . Once $v.d = \delta(s, v)$, it never changes.
- **Corollary 24.12 (No-path property)**
If there is no path from s to v , then we always have $v.d = \delta(s, v) = \infty$.

Shortest-Paths Properties (2/2)

- **Lemma 24.14 (Convergence property)**
If $s \rightsquigarrow u \rightarrow v$ is a shortest path in G for some $u, v \in V$, and if $u.d = \delta(s, u)$ at any time prior to relaxing edge (u, v) , then $v.d = \delta(s, v)$ at all times afterward.
- **Lemma 24.15 (Path-relaxation property)**
If $p = \langle v_0, v_1, \dots, v_k \rangle$ 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 $v_k.d = \delta(s, v_k)$. This property holds regardless of any other relaxation steps that occur, even if they are intermixed with relaxations of the edges of p .

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The Bellman-Ford Algorithm (1/3)

