

# Single - Sourced Shortest Paths

- Directed graph  $G = (V, E)$
- Weight function  $w : E \rightarrow \mathbb{R}$
- Weight of path  $p = \langle v_0, v_1, \dots, v_k \rangle :$

$$\sum_{i=1}^k w(v_{i-1}, v_i)$$

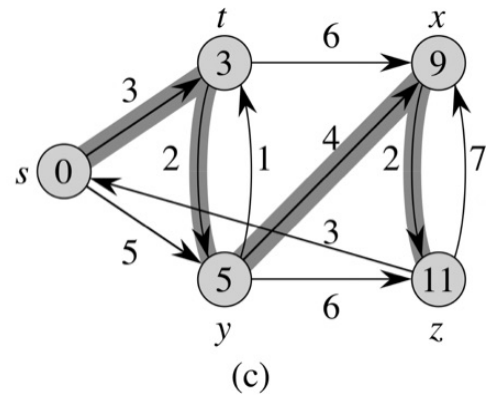
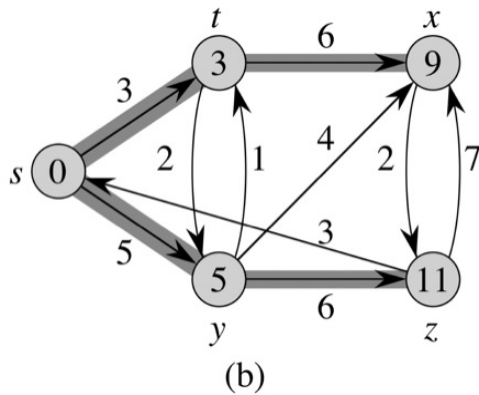
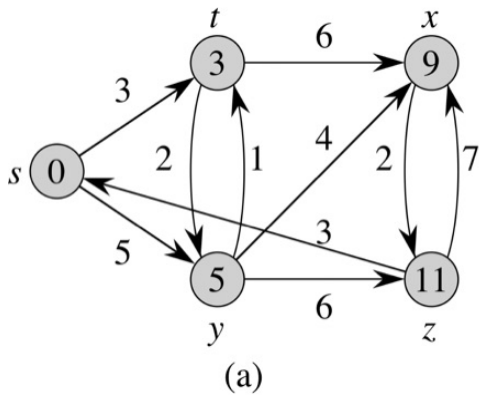
- Shortest path weight  $\delta(u, v)$

$$\delta(u, v) = \begin{cases} \min \{w(p) : u \xrightarrow{p} v\} & \text{if there exists} \\ & u \rightsquigarrow v \\ \infty & \text{otherwise} \end{cases}$$

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

- There may be more than one shortest path from  $s$  to some  $v \in V$

- Looking at all shortest paths from  $s$ , they form a tree



b + c both show shortest paths from  $s$

- What might the weights represent?

- Any measure that

- Accumulates linearly along a path

- We want to minimize

- Travel time - edges are roads or airplane routes, etc.

- Cost

- Penalties

- Variants of shortest path problems

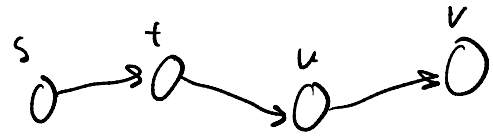
- Single-source - find shortest paths from a given source  $s \in V$  to all vertices  $v \in V$

- Single-destination

- Single-pair - shortest path from  $u$  to  $v$

- all-pairs - find shortest path from  $u$  to  $v$  for all  $u, v$

- Problem exhibits optimal substructure



- Lemma - any subpath of a shortest path is a shortest path

- Proof idea - If a subpath of a shortest path  $p$  was not a shortest path, we could replace it with a shorter subpath to improve  $p$

- Output of a single-source shortest-path algorithm
  - For each vertex  $v \in V$ , find  $v.d = \delta(s, v)$ 
    - Initialize  $v.d = \infty$
    - Reduce  $v.d$  as the algorithm progresses
      - maintaining  $v.d \geq \delta(s, v)$
    - $v.d$  is a shortest path estimate during execution

-  $V.\pi =$  predecessor of  $V$  on a shortest path from  $S$

- Using  $V.\pi$  for all  $V$ , we can build a shortest path tree



- Initialization for all single-source shortest path algorithms in the book

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

**for each  $v \in G.V$**

$$v.d = \infty$$

$$v.\pi = \text{NIL}$$

$$s.d = 0$$

$$\Theta(V)$$

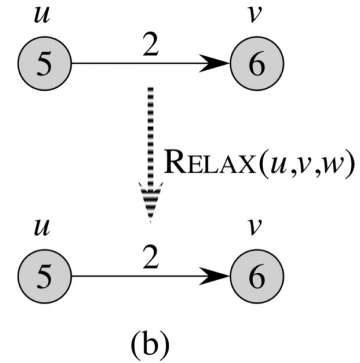
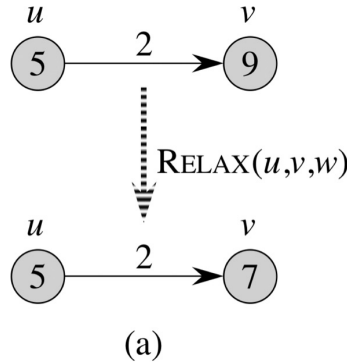
- All algorithms also involve relaxing

**RELAX**( $u, v, w$ )

**if**  $v.d > u.d + w(u, v)$

$v.d = u.d + w(u, v)$

$v.\pi = u$



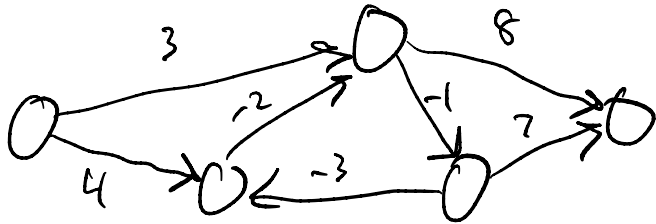
$O(1)$

Estimate of  $d(s, v)$  can be improved using edge  $(u, v)$

$(u, v)$  does not improve the estimate

- All algorithms in the book start with Init-Single-Source, then relax edges
- Some algorithms support negative weights, others do not
- Never possible to find shortest paths if there is a negative cycle reachable from  $s$

Negative cycles cause path lengths of  $-\infty$



## - Bellman - Ford Algorithm

- Allows negative weights

- Computes  $v.d$  and  $v.\pi$  for all  $v \in V$

- Returns True if there are no negative cycles, False otherwise

- Relaxes all edges  $|V| - 1$  times

# BELLMAN-FORD( $G, w, s$ )

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

**for**  $i = 1$  to  $|G.V| - 1$

**for** each edge  $(u, v) \in G.E$

        RELAX( $u, v, w$ )

**for** each edge  $(u, v) \in G.E$

**if**  $v.d > u.d + w(u, v)$

**return** FALSE

**return** TRUE

Subpath information  
propagates throughout  
the graph

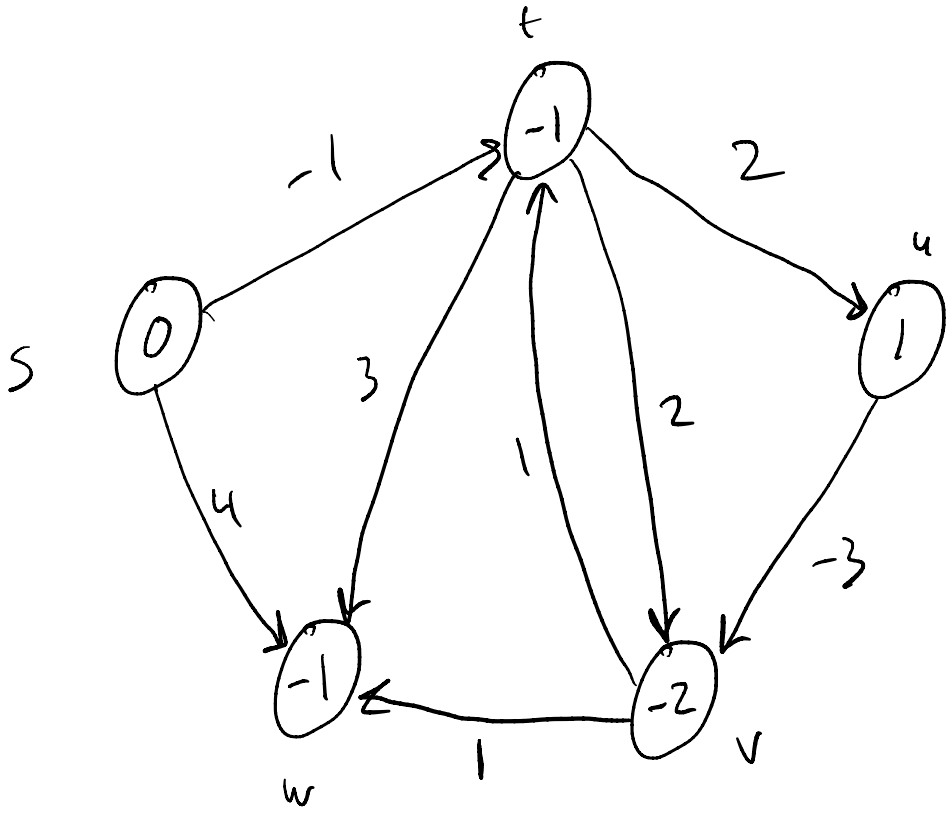
checks for  
negative cycles

Init-Single-Source  $\Theta(V)$

Final loop  $\Theta(E)$

Nested loop  $\Theta(VE)$

whole thing  $\Theta(VE)$



<u>V</u>	<u><math>\pi</math></u>
s	NIL
t	s
u	t
v	u
w	v