

# CS583 Lecture 08

## Single Source Shortest Path

Jyh-Ming Lien

some materials here are based on Prof. Shehu, and Prof. Wang's past lecture notes

# Path Weight

- *Weight of path*  $p = \langle v_0, v_1, \dots, v_k \rangle$

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

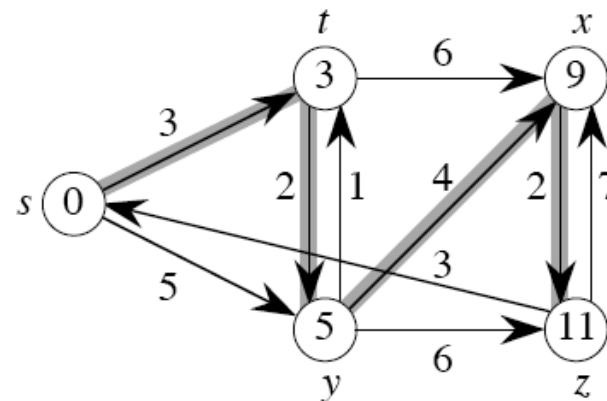
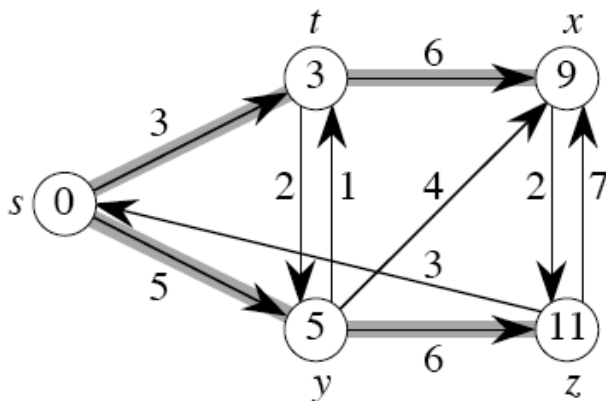
= sum of edge weights on path  $p$  .

- *Shortest-path weight*  $u$  to  $v$ :

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

# Shortest Path

- Shortest path  $u$  to  $v$  is any path  $p$  such that  $w(p) = \delta(u, v)$ .
- may not be unique



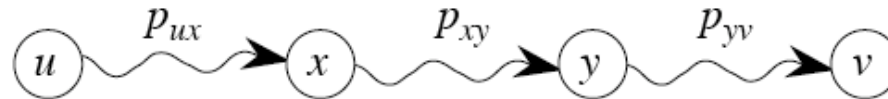
- the shortest paths from one vertex to all the other vertices form a **tree**

# Shortest Path

- Variants
  - single source
  - single destination
  - single pair
  - all pairs (next week)
- negative edge weight
  - assume there are no negative cycles

# Optimal Structure

- Any subpath of a shortest path is a shortest path



- proof

# Cycles

- Shortest path cannot contain cycles  
(assume that no negative cycles)
  - proof

# Shortest Path

- output of the shortest path algorithm for each vertex  $v$  from source  $s$ 
  - $d[v]$ , distance from  $s$ 
    - ▶ this is initially  $+\infty$
    - ▶ reduces as algorithm progress
  - $\pi[v]$ , the predecessor of  $v$  on a shortest path from  $s$ 
    - ▶ this is initially null

# The RELAX function

- we can always improve the value of  $d[v]$  for  $v$  by **relaxing** an edge  $(u,v)$

$\text{RELAX}(u, v, w)$

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

**then**  $d[v] \leftarrow d[u] + w(u, v)$

$\pi[v] \leftarrow u$

- RELAX will never hurt your solution!



# A Framework

- A framework for single-source shortest paths (SSSP) algorithms
  - make all  $d[v] = \infty$  and  $\pi[v] = \text{null}$
  - call RELAX
- The algorithms differ in
  - the order RELAX is called
  - the number of times RELAX is called

# Important Properties

- **triangle inequality**  $\delta(s, v) \leq \delta(s, u) + w(u, v)$
- **upper-bound property**

Always have  $d[v] \geq \delta(s, v)$  for all  $v$ . Once  $d[v] = \delta(s, v)$ , it never changes.

- **no-path property** If  $\delta(s, v) = \infty$ , then  $d[v] = \infty$  always.
- **convergence property**

if  $d[u] = \delta(s, u)$ , then after  $\text{RELAX}(u, v, w)$ ,  $d[v] = \delta(s, v)$

- **path relaxation property**

Let  $p = \langle v_0, v_1, \dots, v_k \rangle$  be a shortest path from  $s = v_0$  to  $v_k$ . If we relax, *in order*,  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$ , even intermixed with other relaxations, then  $d[v_k] = \delta(s, v_k)$ .

# bellman-ford algorithm

- allow negative edge weights
- can detect negative cycles

```
BELLMAN-FORD( $V, E, w, s$ )
INIT-SINGLE-SOURCE( $V, s$ )
for  $i \leftarrow 1$  to  $|V| - 1$ 
    do for each edge  $(u, v) \in E$ 
        do RELAX( $u, v, w$ )
for each edge  $(u, v) \in E$ 
    do if  $d[v] > d[u] + w(u, v)$ 
        then return FALSE
return TRUE
```



# DAG

- algorithm

DAG-SHORTEST-PATHS( $V, E, w, s$ )

topologically sort the vertices

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

**for** each vertex  $u$ , taken in topologically sorted order

**do for** each vertex  $v \in Adj[u]$

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

- Example:
- Time complexity? Why does it work?

# Dijkstra's Algorithm

- no negative weights
- a weighted version of BFS
  - instead of a queue, uses a priority queue
  - keys are  $d[v]$
- very similar to Prim's algorithm
  - greedy
  - iteratively expand the shortest path tree
  - differences?

# Dijkstra's algorithm

- algorithm

DIJKSTRA( $V, E, w, s$ )

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

$S \leftarrow \emptyset$

$Q \leftarrow V$       $\triangleright$  i.e., insert all vertices into  $Q$

**while**  $Q \neq \emptyset$

**do**  $u \leftarrow \text{EXTRACT-MIN}(Q)$

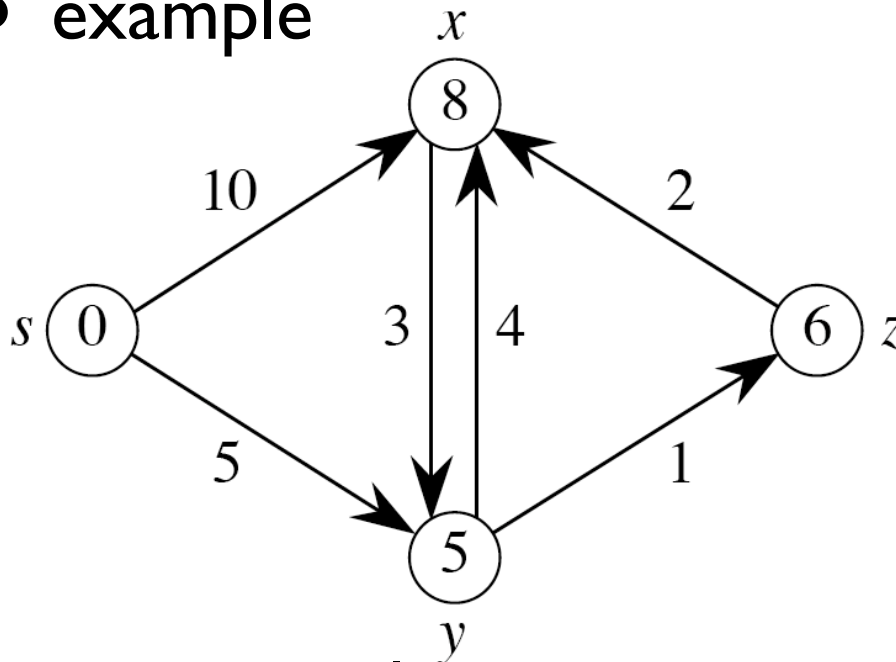
$S \leftarrow S \cup \{u\}$

**for** each vertex  $v \in \text{Adj}[u]$

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

# Dijkstra's algorithm

- example



- time complexity
- why does it work?



# Other Algorithms

- Best-first search
- A\* (and D\*) search
  - Both best-first search and Dijkstra's algorithm are a special case of A\*
- Iterative deepening depth-first search
- We will also look at algorithms that find all-pairs shortest-paths next week