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6.006 Introduction to Algorithms  
Spring 2008

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## Lecture 15: Shortest Paths I: Intro

### Lecture Overview

- Homework Preview
- Weighted Graphs
- General Approach
- Negative Edges
- Optimal Substructure

### Readings

[CLRS, Sections 24 \(Intro\)](#)

### Motivation:

Shortest way to drive from A to B (Google maps “get directions”)

Formulation: Problem on a weighted graph  $G(V, E)$   $W : E \rightarrow \mathfrak{R}$

Two algorithms: Dijkstra  $O(V \lg V + E)$  assumes non-negative edge weights  
Bellman Ford  $O(VE)$  is a general algorithm

### Problem Set 5 Preview:

- Use Dijkstra to find shortest path from CalTech to MIT
  - See “CalTech Cannon Hack” photos (search web.mit.edu)
  - See Google Maps from CalTech to MIT
- Model as a weighted graph  $G(V, E), W : E \rightarrow \mathfrak{R}$ 
  - $V$  = vertices (street intersections)
  - $E$  = edges (street, roads); directed edges (one way roads)
  - $W(U, V)$  = weight of edge from  $u$  to  $v$  (distance, toll)

$$\begin{aligned} \text{path } p &= \langle v_0, v_1, \dots, v_k \rangle \\ (v_i, v_{i+1}) &\in E \quad \text{for } 0 \leq i < k \\ w(p) &= \sum_{i=0}^{k-1} w(v_i, v_{i+1}) \end{aligned}$$

**Weighted Graphs:****Notation:**

$v_0 \xrightarrow{p} v_k$  means  $p$  is a path from  $v_0$  to  $v_k$ .  $(v_0)$  is a path from  $v_0$  to  $v_0$  of weight 0.

**Definition:**

Shortest path weight from  $u$  to  $v$  as

$$\delta(u, v) = \begin{cases} \min \left\{ w(p) : u \xrightarrow{p} v \right\} & \text{if } \exists \text{ any such path} \\ \infty & \text{otherwise (} v \text{ unreachable from } u \text{)} \end{cases}$$

**Single Source Shortest Paths:**

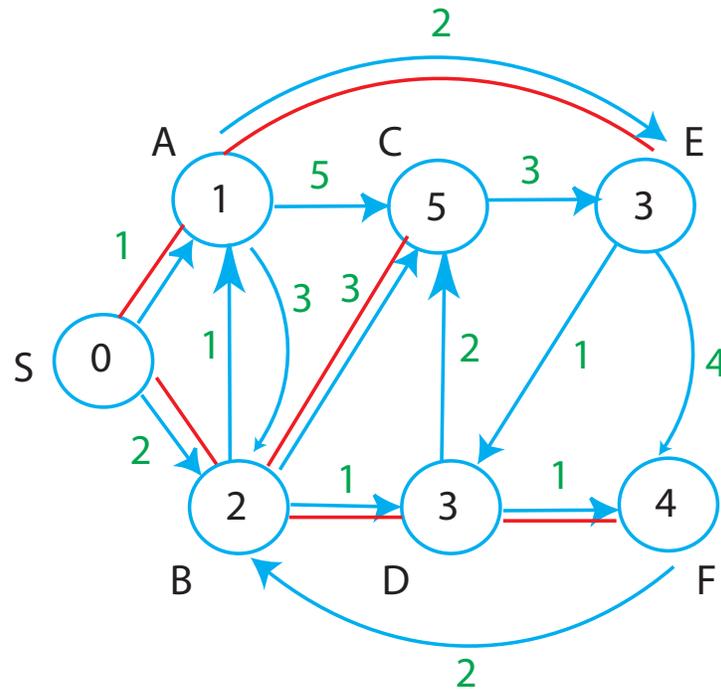
Given  $G = (V, E)$ ,  $w$  and a source vertex  $S$ , find  $\delta(S, V)$  [and the best path] from  $S$  to each  $v \in V$ .

Data structures:

$$\begin{aligned} d[v] &= \text{value inside circle} \\ &= \begin{cases} 0 & \text{if } v = s \\ \infty & \text{otherwise} \end{cases} \leftarrow \text{initially} \\ &= \delta(s, v) \leftarrow \text{at end} \\ d[v] &\geq \delta(s, v) \quad \text{at all times} \end{aligned}$$

$d[v]$  decreases as we find better paths to  $v$

$\Pi[v]$  = predecessor on best path to  $v$ ,  $\Pi[s] = \text{NIL}$

**Example:**Figure 1: Shortest Path Example: Bold edges give predecessor  $\Pi$  relationships**Negative-Weight Edges:**

- Natural in some applications (e.g., logarithms used for weights)
- Some algorithms disallow negative weight edges (e.g., Dijkstra)
- If you have negative weight edges, you might also have negative weight cycles  $\implies$  may make certain shortest paths undefined!

**Example:**

See Figure 2

$B \rightarrow D \rightarrow C \rightarrow B$  (origin) has weight  $-6 + 2 + 3 = -1 < 0!$

Shortest path  $S \rightarrow C$  (or  $B, D, E$ ) is undefined. Can go around  $B \rightarrow D \rightarrow C$  as many times as you like

Shortest path  $S \rightarrow A$  is defined and has weight 2

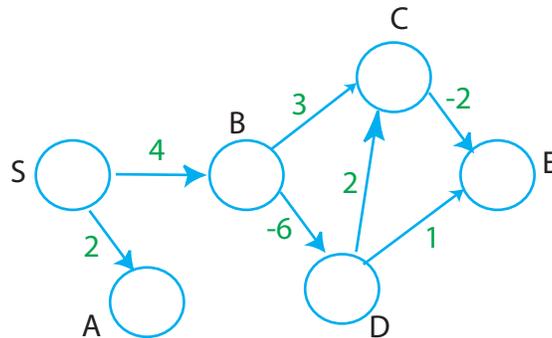


Figure 2: Negative-weight Edges

If negative weight edges are present, s.p. algorithm should find negative weight cycles (e.g., Bellman Ford)

### General structure of S.P. Algorithms (no negative cycles)

```

Initialize:      for  $v \in V$ :  $d[v] \leftarrow \infty$ 
                   $\Pi[v] \leftarrow \text{NIL}$ 
                   $d[S] \leftarrow 0$ 
Main:           repeat
                  select edge  $(u, v)$  [somehow]
                  "Relax" edge  $(u, v)$ 
                  [ if  $d[v] > d[u] + w(u, v)$  :
                     $d[v] \leftarrow d[u] + w(u, v)$ 
                     $\pi[v] \leftarrow u$ 
                  ]
                  until all edges have  $d[v] \leq d[u] + w(u, v)$ 
  
```

**Complexity:**

Termination? (needs to be shown even without negative cycles)

Could be exponential time with poor choice of edges.

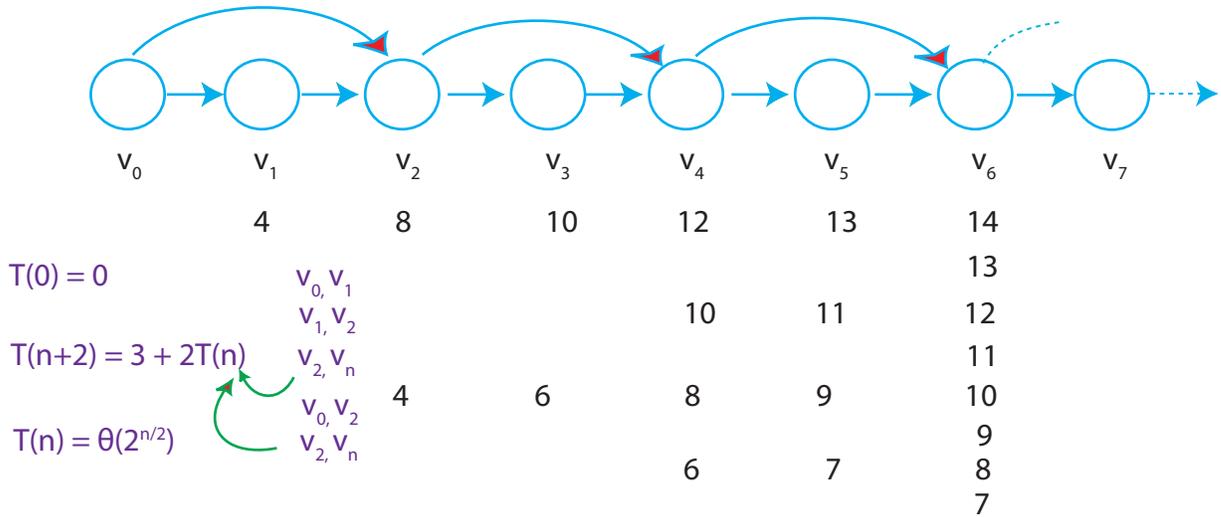


Figure 3: Running Generic Algorithm

**Optimal Substructure:**

**Theorem:** Subpaths of shortest paths are shortest paths

Let  $p = \langle v_0, v_1, \dots, v_k \rangle$  be a shortest path

Let  $p_{ij} = \langle v_i, v_{i+1}, \dots, v_j \rangle \quad 0 \leq i \leq j \leq k$

Then  $p_{ij}$  is a shortest path.

**Proof:**

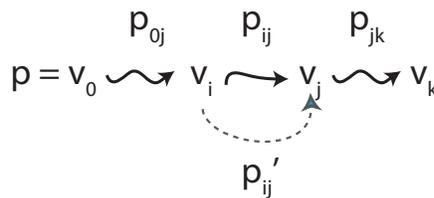


Figure 4: Optimal Substructure Theorem

If  $p'_{ij}$  is shorter than  $p_{ij}$ , cut out  $p_{ij}$  and replace with  $p'_{ij}$ ; result is shorter than  $p$ .

**Contradiction.**

**Triangle Inequality:**

**Theorem:** For all  $u, v, x \in X$ , we have

$$\delta(u, v) \leq \delta(u, x) + \delta(x, v)$$

**Proof:**

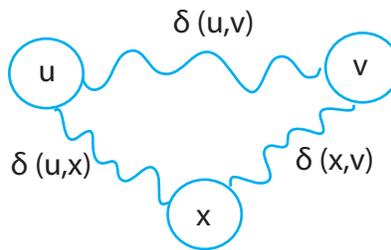


Figure 5: Triangle inequality