

Lecture 14: Shortest Paths I: Intro

Lecture Overview

- 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

Application

- Find shortest path from CalTech to MIT
 - See “CalTech Cannon Hack” photos 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:

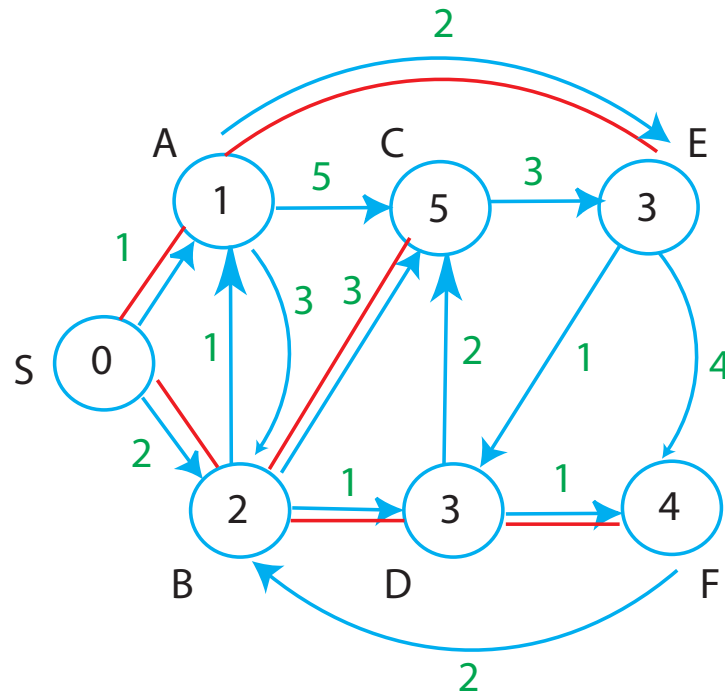
Given $G = (V, E)$, w and a source vertex S , find $\delta(S, V)$ [and the best path] from S to each $v \in V$.

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 Π 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

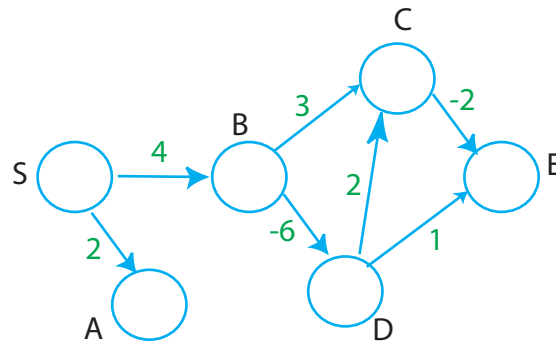


Figure 2: Negative-weight Edges. Error: Edge from B to C should be from C to B .

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)

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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.

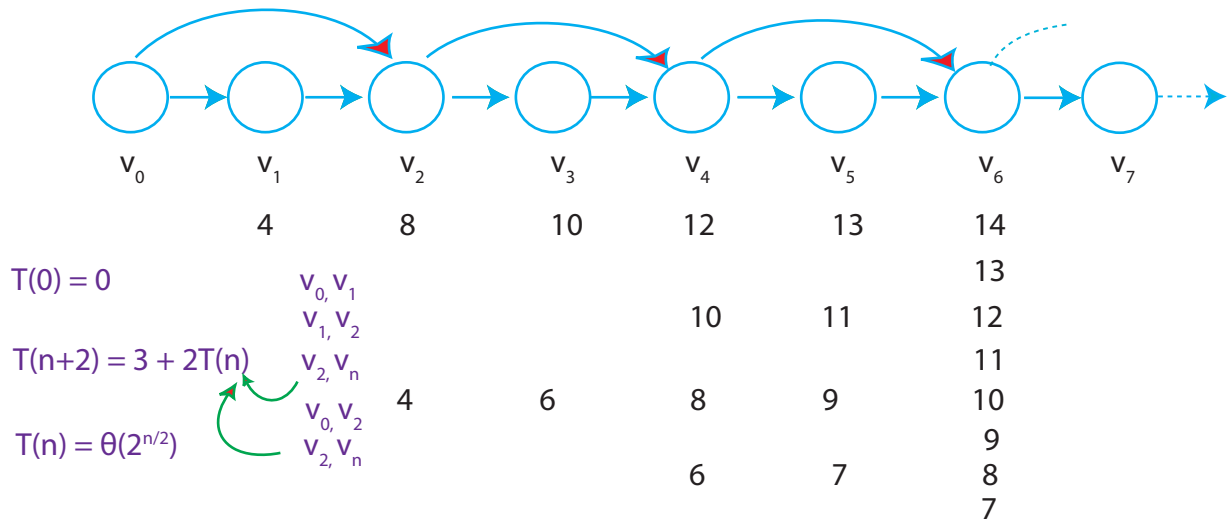


Figure 3: Running Generic Algorithm. The outgoing edges from v_0 and v_1 have weight 4, the outgoing edges from v_2 and v_3 have weight 2, the outgoing edges from v_4 and v_5 have weight 1.

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:

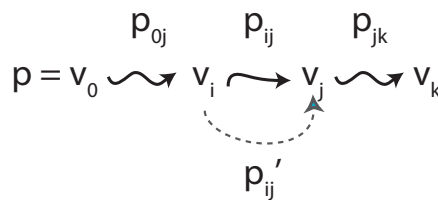


Figure 4: Optimal Substructure Theorem

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:

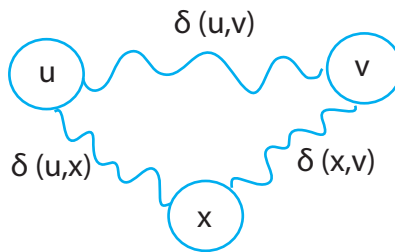


Figure 5: Triangle inequality