

6.006 Lecture 15: Shortest Paths (Intro)

▢ Motivation

▢ ~~unweighted~~ Weighted graphs

▢ General Structure of S.P. algorithms

▢ Optimal substructure

Motivation

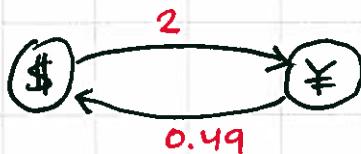
- PS5: find shortest path from Caltech To MIT.

- Map is a graph of road segments

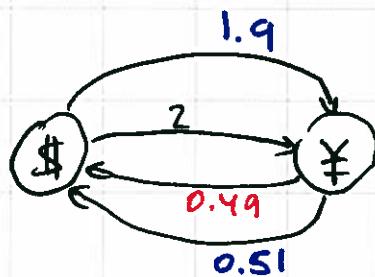
- last segment is 0.7 mile on Mass Ave, but some interstate segments are many miles long; we clearly have to represent length of road segment in the graph.

- Arbitrage

I have \$1 and can convert it to Yens at the local bank; I get ¥2. For each ¥, I can get \$0.49 (the bank needs to make money).



If I traverse the cycle, I end back with \$'s, but only $\$0.98$: $1 \times 2 \times 0.49 = 0.98$. Not good. But suppose that a bank in Japan has slightly different exchange rates:



Now there is a path from \$ to \$ (a cycle) that makes me \$0.02 on the dollar. People exploit that and this is what causes exchange rates to converge. Clearly, weighted graphs are useful...

Definitions

A directed graph $G(V, E)$

Edges have weight $w(u, v)$ ($u, v \in V$,
 $(u, v) \in E$)

Weight is a function $w: E \rightarrow \mathbb{R}$

Path $p = \langle v_0, v_1, v_2, \dots, v_k \rangle$ where

$(v_i, v_{i+1}) \in E$ for $0 \leq i \leq k-1$; $v_0 \xrightarrow{p} v_k$

Weight of a path $w(p) = \sum_{i=0}^{k-1} w(v_i, v_{i+1})$

Shortest paths:

$$\delta(u, v) = \begin{cases} \min_p \{w(p) : u \xrightarrow{p} v\} \\ \infty \quad \text{if } v \text{ is unreachable from } u \\ \text{(set above is empty).} \end{cases}$$

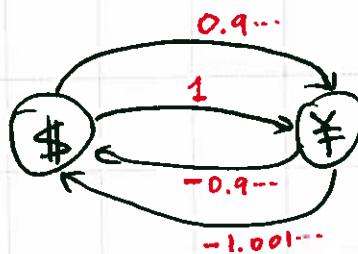
Back to Modelling

This clearly works for geographic routes:

we want to minimize the sum of the segment lengths on the path.

But not for arbitrage: there we multiplied the rates.

Solution: use $w(u, v) = \log(\text{exchange rate})$



Now an additive cycle with negative weight means we can make money.

Single-Source Shortest Paths

Input: $G = (V, E)$, w , a source vertex s

Goal: Find $\delta(s, v)$ for every $v \in V$

and the best path from s to v

Data structures:

$d[v]$ = length of best path to v

so far

initialization: $d[v] = \begin{cases} 0 & v=s \\ \infty & \text{otherwise} \end{cases}$

during the algorithm, $d[v] \geq \delta(s, v)$

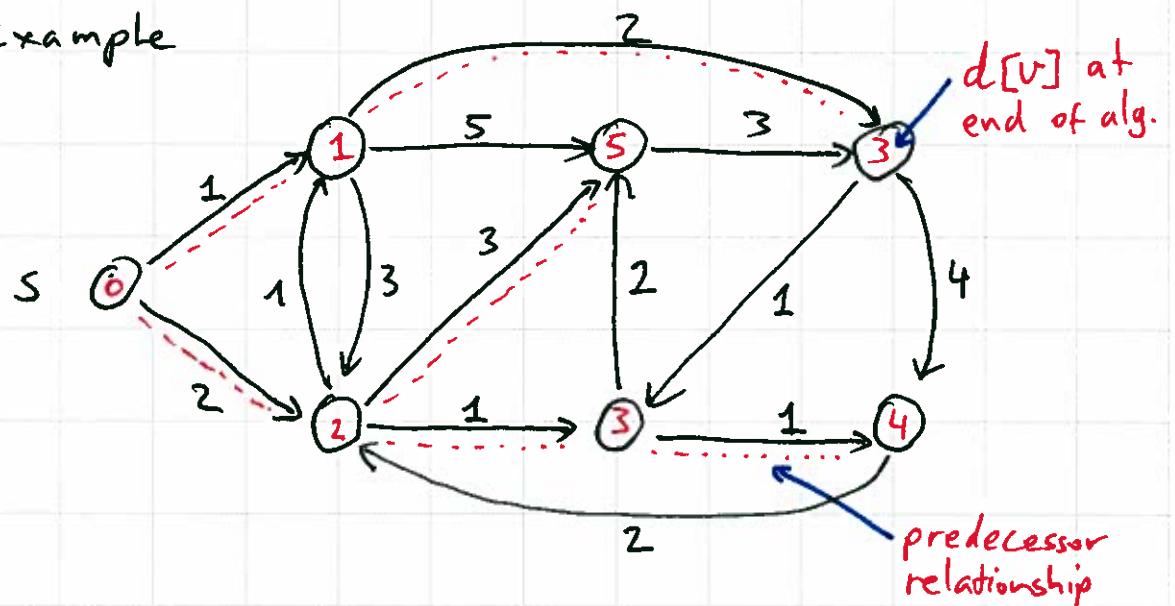
when the algorithm terminates,

$d[v]$ will be exactly $\delta(s, v)$

$\pi[v]$ = predecessor of v on best path

so far, initialize $\pi[v] = \text{None}$.

Example



General structure of SSSP algorithms

Based on the observation that each $(u, v) \in E$ is a constraint

$$\delta(s, v) \leq \delta(s, u) + w(u, v)$$

We start with some $d[v]$ that satisfies constraints (and $\delta(s, s) = 0$) but is not minimal, and relax the constraints until we are (hopefully) done

for v in V :

$$d[v] = \infty$$

$$\pi[v] = \text{None}$$

$$d[s] = 0$$

while $d[v] > d[u] + w(u, v)$ for some v :

$$d[v] = d[u] + w(u, v)$$

$$\pi[v] = u$$

nondeterministic choice,
not a real algorithm yet!!!

Does not terminate if \exists negative cycle (need negative cycle detection)
Complexity may be exponential!

Complexity (bad case):



use \rightarrow to set $d[u]$ and recurse from u , then

use $\rightarrow \rightarrow$ to lower ~~$d[u]$~~ $d[u]$ and recurse again.

$$T(n) = 3 + 2 T(n-2) = \Theta(2^{n/2})$$

We'll see two algorithms that do much better

Bellman-Ford $O(V E)$, detects negative cycles

Dijkstra $O(V \lg V + E)$ but assumes that

there are no negative cycles.

Two Structural Properties

Theorem: subpaths of shortest paths are also shortest paths.

Proof: Let $p = \langle v_0, v_1, \dots, v_i, v_i, v_{i+1}, \dots, v_j, \dots, v_k \rangle$

$$p_{ij} = \langle v_i, v_{i+1}, \dots, v_j \rangle$$



If $w(p'_{ij}) < w(p_{ij})$ then p is not shortest.

we can replace p_{ij} with p'_{ij} and get

$$p' = v_0 \xrightarrow{p'} v_k \text{ with } w(p') < w(p)$$

Theorem: triangle inequality, for all $u, v, x \in V$

$$\text{we have } \delta(u, v) \leq \delta(u, x) + \delta(x, v)$$

Proof:

