

TODAY: All-pairs shortest paths

- dynamic programming
- matrix multiplication
- Floyd-Warshall algorithm
- Johnson's algorithm
- difference constraints

Recall: single-source shortest paths [6.006]

- given directed graph $G=(V,E)$, vertex set V , & edge weights $w: E \rightarrow \mathbb{R}$
- find $S(s,v)$ = shortest-path weight $s \rightarrow v \quad \forall v \in V$
(or $-\infty$ if neg.-weight cycle along the way,
or ∞ if no path)

<u>situation</u>	<u>algorithm</u>	<u>time</u>
unweighted ($w=1$)	BFS	$O(V+E)$
nonneg. edge weights	Dijkstra	$O(E+V \lg V)$
general	Bellman-Ford	$O(VE)$
acyclic graph (DAG)	topological sort + 1 pass Bellman-Ford	$O(V+E)$

using Fibonacci
heaps

all of these results are the best known

All-pairs shortest paths:

given edge-weighted graph $G=(V, E, w)$,
find $S(u, v)$ for all $u, v \in V$

<u>situation</u>	<u>algorithm</u>	<u>time</u>	<u>$E=O(V^2)$</u>
unweighted	$ V \times \text{BFS}$	$O(VE)$	$O(V^3)$
nonneg. weights	$ V \times \text{Dijkstra}$	$O(VE + V^2 \lg V)$	$O(V^3)$
general	$ V \times \text{B-F}$	$O(V^2 E)$	$O(V^4)$
general	Johnson's (TODAY)	$O(VE + V^2 \lg V)$	$O(V^3)$

these results (except third) are also best known — don't know how to beat $|V| \times \text{Dijkstra}$

Application: Google Maps preprocessing (between waypoints)
Internet routing

— define $w(u, v) = \infty$ for $(u, v) \notin E$

Dynamic program (#1):

① subproblems: $d_{uv}^{(m)}$ = weight of shortest path $u \rightarrow v$ using $\leq m$ edges

② guessing: what's the last edge (x, v) ?

③ recurrence: $d_{uv}^{(m)} = \min(d_{ux}^{(m-1)} + w(x, v) \text{ for } x \text{ in } V)$
 $d_{uv}^{(0)} = \begin{cases} 0 & \text{if } u=v \\ \infty & \text{else} \end{cases}$

④ topolog. order: for $m=0, 1, \dots, n-1$: for $u \& v$ in V :

⑤ original problem: $\hookrightarrow |V|$

if no neg.-weight cycles then (by B-F analysis) shortest path is simple $\Rightarrow S(u, v) = d_{uv}^{(n-1)} = d_{uv}^{(n)} = \dots$
(neg.-weight cycle $\Leftrightarrow d_{vv}^{(n-1)} < 0$ for some $v \in V$)

Time: V^3 subproblems $\cdot V$ choices $\cdot O(1)$ time/choice
= $O(V^4)$ - no better than $|V| \times$ Bellman-Ford

Bottom-up via relaxation steps: (like Dijkstra & Bellman-Ford)

for m in range $(1, n)$:

for u in V :

for v in V :

for x in V :

if $d_{uv} > d_{ux} + d_{xv}$:

$d_{uv} = d_{ux} + d_{xv}$

omit superscripts because

more relaxation never hurts

instead of $w(x, v)$ - only helps

} relaxation step
(Δ inequality)

OR: $d_{uv}^{(m)} = \min(d_{ux}^{\lceil m/2 \rceil} + d_{xv}^{\lceil m/2 \rceil} \text{ for } x \in V) \Rightarrow O(n^3 \lg n)$ time! (student suggest.)

Matrix multiplication: (recall)

given $n \times n$ matrices A & B ,
compute $C = A \cdot B$: $c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$

- $O(n^3)$ via standard algorithm
- $O(n^{2.807})$ via Strassen's algorithm
- $O(n^{2.376})$ via Coppersmith-Winograd algorithm
- $O(n^{2.3728})$ via Vassilevska Williams algorithm

Connection to shortest paths:

- define $\oplus = \min$ & $\odot = +$

- then $C = A \odot B$ is $c_{ij} = \min_k (a_{ik} + b_{kj})$

- define $D^{(m)} = (d_{ij}^{(m)})$, $W = (w_{ij})$, $V = \{1, 2, \dots, n\}$

$\Rightarrow D^{(m)} = D^{(m-1)} \odot W$ (by ③ above)

$= W^{(m)}$

where $W^{(0)} = \begin{pmatrix} 0 & \infty & \infty & \infty \\ \infty & 0 & \infty & \infty \\ \infty & \infty & 0 & \infty \\ \infty & \infty & \infty & 0 \end{pmatrix}$

[$W^{(m)}$ makes sense because \odot is associative, which follows from $(\mathbb{R}, \min, +)$ being closed semiring]

Matrix multiplication algorithm:

- $n-2$ multiplications $\Rightarrow O(n^4)$ time (still no better)

- repeated squaring: $((W^2)^2)^2 \dots = W^{2^{\lceil \lg n \rceil}} = W^{n-1}$

$= (S_{ij})$ if no negative-weight cycles

- time: $O(n^3 \lg n)$

- neg.-weight cycles \Leftrightarrow neg. diagonal entries in W

- can't use Strassen etc. \because (no negation)

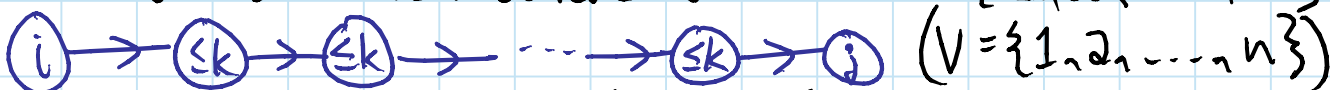
Transitive closure: $t_{ij} = \begin{cases} 1 & \text{if there's a path } i \rightarrow j \\ 0 & \text{else} \end{cases}$

$= [\text{is } \delta(i,j) < \infty?] \Rightarrow$ special case of APSP

- $(\{0,1\}, \text{or, and})$ is a ring \Rightarrow can use Strassen etc.
 $\Rightarrow O(n^{2.3728} \lg n)$ time

Floyd-Warshall algorithm: faster dynamic program

① subproblem $c_{uv}^{(k)}$ = weight of shortest path $u \rightarrow v$ whose intermediate vertices $\in \{1, 2, \dots, k\}$



② guessing = does shortest path use vertex k ?

③ $c_{uv}^{(k)} = \min \{ c_{uv}^{(k-1)}, c_{uk}^{(k-1)} + c_{kv}^{(k-1)} \}$

$c_{uv}^{(0)} = w(u,v)$

④ for k : for u, v :

⑤ $\delta(u,v) = c_{uv}^{(n)}$, neg.-weight cycle \Leftrightarrow neg. $c_{uu}^{(n)}$

no neg.-weight cycles \Rightarrow use vertex k only once

Time: $O(V^3)$ subproblems \cdot 2 choices \cdot $O(1)$
 $= O(V^3)$

Bottom up via relaxation:

simple & efficient in practice

$C = (w(u,v))$

for $k = 1, 2, \dots, n$:

for u in V :

for v in V :

if $c_{uv} > c_{uk} + c_{kv}$:

$c_{uv} = c_{uk} + c_{kv}$

} relaxation again

again OK to omit subscripts

Johnson's algorithm:

- ① find function $h: V \rightarrow \mathbb{R}$ such that $w_h(u,v) = w(u,v) + h(u) - h(v) \geq 0$ for all $u,v \in V$ or determine that a negative-weight cycle exists
- ② run Dijkstra's algorithm on (V, E, w_h) from every source vertex $s \in V$
 \Rightarrow get $\delta_h(u,v)$ for all $u,v \in V$
- ③ claim $\delta(u,v) = \delta_h(u,v) - h(u) + h(v)$

Proof of claim:

- look at any $u \rightarrow v$ path p in G

- say p is $v_0 \rightarrow v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_k$

$$\begin{aligned} \Rightarrow w_h(p) &= \sum_{i=1}^k w_h(v_{i-1}, v_i) \\ &= \sum_{i=1}^k [w(v_{i-1}, v_i) + h(v_{i-1}) - h(v_i)] \\ &= \sum_{i=1}^k w(v_{i-1}, v_i) + h(v_0) - h(v_k) \quad \text{telescoping} \\ &= w(p) + h(u) - h(v) \end{aligned}$$

- so all $u \rightarrow v$ paths change in weight by the same offset $+h(u) - h(v)$

\Rightarrow shortest path is preserved (but offset) \square

How to find h? (1)

$$w_h(u,v) = w(u,v) + h(u) - h(v) \geq 0$$

$$\Leftrightarrow h(v) - h(u) \leq w(u,v) \quad \text{for all } u,v \in V$$

SYSTEM OF DIFFERENCE CONSTRAINTS

Theorem: if (V, E, w) has a negative-weight cycle then no solution to difference constraints

Proof: say $v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_k \rightarrow v_0$ is neg. weight

$$\text{if } h(v_1) - h(v_0) \leq w(v_0, v_1)$$

$$\& h(v_2) - h(v_1) \leq w(v_1, v_2)$$

$$\vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots$$

$$\& h(v_k) - h(v_{k-1}) \leq w(v_{k-1}, v_k)$$

$$\& h(v_0) - h(v_k) \leq w(v_k, v_0)$$

then sum: $0 \leq w(\text{cycle}) < 0 \quad \text{X} \quad \square$

Good
Will
Hunting

Theorem: if (V, E, w) has no negative-weight cycle then can solve difference constraints

Proof: add to G a new vertex s
& add weight-0 edges (s, v) for all $v \in V$ }

- introduce no (negative-weight) cycles

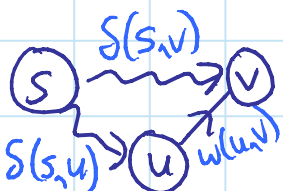
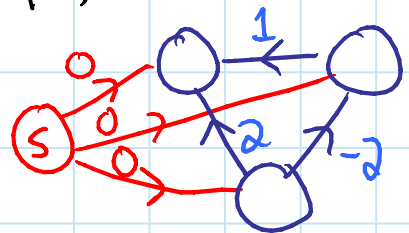
- $s \rightarrow v$ path now exists

$\Rightarrow \delta(s, v)$ is finite for all $v \in V$

- assign $h(v) = \delta(s, v)$

- $h(v) - h(u) \leq w(u, v) \Leftrightarrow \delta(s, v) - \delta(s, u) \leq w(u, v)$

$\Leftrightarrow \delta(s, v) \leq \delta(s, u) + w(u, v)$ TRIANGLE INEQUALITY \square



* { Alternate reduction: for every $(u,v) \in E$,
 add (u,v) with weight $M = |V| \cdot (\text{largest } |w|)$.
 \Rightarrow Strongly connected, still no neg.-weight cycles

Analysis:

① = Bellman-Ford from s
 [+ reweight all edges

$O(VE)$

$O(E)$

② = $|V| \times$ Dijkstra

$\rightarrow O(VE + V^2 \lg V)$

③ = reweight all pairs

$O(V^2)$

$\rightarrow O(VE + V^2 \lg V)$

Also: Bellman-Ford can solve any system of difference constraints $\{x - y \leq c\}$
 (or report unsolvable)
 in $O(VE)$ where $V = \text{variables}$, $E = \text{constraints}$

Exercise: Bellman-Ford minimizes $\max_i x_i - \min_i x_i$

Applications to real-time programming
 multimedia scheduling
 temporal reasoning

e.g. $LB \leq t_{\text{end}} - t_{\text{start}} \leq UB$

$0 \leq t_{\text{start}2} - t_{\text{end}1} \leq \epsilon$

$|t_{\text{start}1} - t_{\text{start}2}| \leq \epsilon \text{ or } 0$

bounds on:
 duration
 gap
 synchrony

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