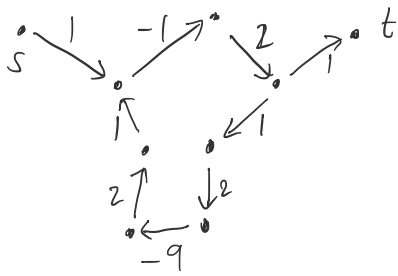


Lec11-Bellman-Ford

Tuesday, September 19, 2023 9:10 PM

KT 6.8

What about negative edges in shortest paths?



Problem: Given a directed graph with weighted edges $d(u,v) \in \mathbb{R}$ that may be pos. or neg., find the shortest path from s to t .

Complication: Negative cycles allow for arbitrarily low-cost paths

Solution idea: Add a big number to all edges to make them positive.

Problem: just turns the problem into shortest unweighted path, since number of edges dominates.

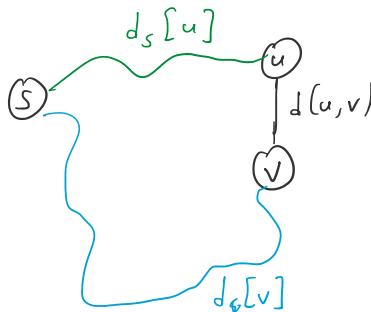
$$\text{cost}(P) = M \times \text{hops}(P) + \text{cost}(P)$$

Modified problem: Given a directed graph with weighted edges $d(u,v) \in \mathbb{R}$ that may be pos. or neg., either
 (1) determine there is a negative cycle
 or (2) find the shortest path from s to t .

Bellman-Ford: Let $d_s[v]$ be the current estimated distance from s to v .
 At start, $d_s[s] = 0$ and $d_s[v] = \infty \quad \forall v \neq s$.

Relaxation step (Ford step):

Find an edge (u,v) s.t. $d_s[u] + d(u,v) < d_s[v]$.
 Set $d_s[v] = d_s[u] + d(u,v)$.



Thm If you cannot relax (via Ford step), then $d_s[u] =$ shortest path distance from s to u for all u .

After i steps: either $d_s[v] = \infty$ or $\exists s \rightarrow v$ path of length $d_s[v]$

distance from s to u for all u .

proof.

Lemma 1:

After any step i , either $d_s[v] = \infty$ or $\exists s \rightarrow v$ path of length $d_s[v]$.

proof.

If $d_s[v] = \infty$, can't say anything at step i .

Let v be a vertex s.t. $d_s[v] < \infty$.

Proof by induction on i .

Base case: When $i=0$, $d_s[s] = 0 < \infty$ and \exists path of length 0 to s .

Induction hypo: Assume true steps $< i$.

Ind. step: In step i , we update distance using some edge (u,v) where $d_s[v] > d_s[u] + d(u,v)$. (setting $d_s[v] = d_s[u] + d(u,v)$)

$\Rightarrow d_s[u] < \infty$, so $d_s[u]$ was updated at an earlier step.

$\Rightarrow \exists$ path P_{su} of length $d_s[u]$ from $s \rightarrow u$.

$\Rightarrow P_{su} + (u,v)$ gives a path $s \rightarrow v$ with length new $d_s[v]$. □

Lemma 2: When no more Ford steps possible, \forall path P_{sv} from $s \rightarrow v$, $\text{length}(P_{sv}) \geq d_s[v]$. (we get the shortest path)

proof. By induction on #edges in P_{sv} .

Base case: When $|P_{sv}| = 1$, only single edge (s,v) , and since can't do Ford step, $d(s,v) \geq d_s[v]$

Induction hypo: Assume true for all P_{sv} of k or fewer edges.

Induction step: Let P_{sv} be a $s \rightarrow v$ path of $k+1$ edges.

$P_{sv} = P_{su} + (u,v)$ for some u .

$\text{length}(P_{sv}) = \text{length}(P_{su}) + d(u,v) \geq d_s[u] + d(u,v) \geq d_s[v]$

otherwise could relax. □

Together, these lemmas imply the Thm. □

Implementation:

$G = (V, E)$

Shortest Path $(G, s, t) =$

Initialization $d[u] = \infty \quad \forall u$.

Shortest Path $(G, s, t) =$

Initialize $d[u] = \infty \quad \forall u.$

$d[s] = 0.$

$Q = [s]$ (a queue)

while queue not empty:

~~while relaxation possible:~~

~~for $u \in V$:~~

$u = Q.dequeue()$ (remove front item)

for $v \in \text{neighbors}(u)$: (relax edges)

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

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

parent $[v] = u$

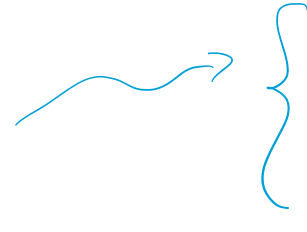
if $v \notin \text{queue}$, $Q.enqueue(v)$

$Q_0 = [s]$

for $i = 0, 1, 2, \dots$

while Q_i not empty:

$u = Q_i.dequeue()$



\rightarrow if $v \notin Q_{i+1}$, $Q_{i+1}.enqueue(v)$

When is relaxation possible? Need $d_s[u] + d(u,v) < d_s[v]$ \leftarrow only ever decrease $d_s[u]$
 \perp this must have decreased.

We can keep track of all nodes that decrease and just check them.

Running time:

Let's consider the queue in stages.

$$Q = [Q_0, Q_1, Q_2, \dots, Q_{|V|-1}]$$

Q_i will capture all updates from paths with i edges, because things are added to Q_i because nodes in Q_{i-1} were updated.

\Rightarrow If we ever reach $Q_{|V|}$, then we have a cycle.

Each subqueue takes $O(|E|)$ time to process because we might have to check the Ford rule on every edge.

Total running time = $O(|V||E|)$. (slower than Dijkstra $O(|E| \log |V|)$)

Alt implementation:

Notice: each edge can only be relaxed $|V|-1$ times without forming a neg cycle.

\Rightarrow can directly count that.

Shortest Path $(G, s, t) =$

Initialize $d[u] = \infty \quad \forall u.$ $d[s] = 0$

Shortest Path $(G, s, t) =$

Initialize $d[u] = \infty \forall u$. $d[s] = 0$

for $i = 1, \dots, |V| - 1$:

for $(u, v) \in E$,

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

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

parent $[v] = u$

$O(|V||E|)$ time.

(slower than Dijkstra)

for $(u, v) \in E$:

if $d[u] + d(u, v) < d[v]$:

report negative-weight cycle

Dynamic Programming view: (DP solves problems by solving subproblems)

Def. Let $dist_s(v, i)$ be min-cost of $s \rightarrow v$ path using $\leq i$ edges

Can recursively define $dist_s(v, i)$:

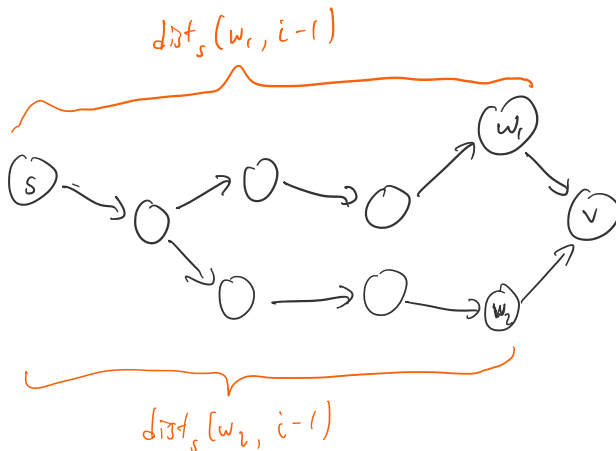
1. If best $s \rightarrow v$ path uses at most $i-1$ edges, then

$$dist_s(v, i) = dist_s(v, i-1)$$

2. If best $s \rightarrow v$ path uses i edges, and the last edge is (w, v) ,

$$then \ dist_s(v, i) = d(w, v) + dist_s(w, i-1).$$

← similar to Ford update rule



Let $N(w)$ be neighbors of w .

Recurrence:

$$dist_s(v, i) = \min \left\{ \begin{array}{l} dist_s(v, i-1) \\ \min_{w \in N(v)} \underline{dist_s(w, i-1) + d(w, v)} \end{array} \right.$$

} $dist_s(v, x)$ only depends on $dist_s(w, y)$ for $y < x$.

Base case: $\text{dist}_s(v, 1) = d(s, v)$ or ∞ if (s, v) does not exist.

Goal: compute $\text{dist}_s(t, n-1)$.

Only $|V| \times (|V|-1)$ possible arguments for $\text{dist}_s(\cdot, \cdot)$

$|V|$ nodes
↓

$|V|-1$ allowed hops

Encode as matrix

allowed hops

7	0							
6	0							
5	0							
4	0							
3	0							
2	0							
1	0	∞	3	∞	∞	7	∞	-2
	s	b	c	d	e	f	g	h

$\text{dist}_s(g, b)$

Can fill from bottom up.

nodes

Bellman-Ford Pseudocode

Bellman-Ford (G, s, t)

Initialize $\text{dist}_s[x, 1]$ to $d(s, x) \forall x$

For $i = 1, \dots, |V|-1$:

For v in V :

best_w = None

for w in $N(v)$:

best_w = $\min(\text{best}_w, \text{dist}_s[w, i-1] + d(w, v))$

$\text{dist}_s[v, i] = \min(\text{best}_w, \text{dist}_s[v, i-1])$

Return $\text{dist}_s[t, n-1]$

Running time:

Simple: $O(|V|^2)$ subproblems

$O(|V|)$ time to compute each entry in table
(have to search all possible neighbors w)

$\Rightarrow O(|V|^3)$ time

Better analysis:

Let n_v be number of in-edges of v .

Actually only $O(n_v)$ time to compute the entry for v .

Total time = $O(|V| \sum_{v \in V} n_v) = O(|V||E|)$.

in each row only $O(|E|)$ computations since
just need to check in-edges once each.