

Algorithms: Design and Analysis, Part II

Single-Source Shortest Paths Revisited

The Single-Source Shortest Path Problem

Input: Directed graph G = (V, E), edge lengths c_e for each $e \in E$, source vertex $s \in V$. [Can assume no parallel edges.]

Goal: For every destination $v \in V$, compute the length (sum of edge costs) of a shortest s-v path.

On Dijkstra's Algorithm

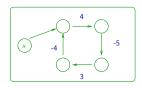
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Good news: O(m \log n) running time using heaps (n = \text{number of vertices}, m = \text{number of edges})
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Bad news:

- (1) Not always correct with negative edge lengths [e.g. if edges \mapsto financial transactions]
- (2) Not very distributed (relevant for Internet routing)

Solution: The Bellman-Ford algorithm

On Negative Cycles



Question: How to define shortest path when *G* has a negative cycle?

Solution #1: Compute the shortest s-v path, with cycles allowed.

Problem: Undefined or $-\infty$. [will keep traversing negative cycle]

Solution #2: Compute shortest cycle-free s-v path.

Problem: NP-hard (no polynomial algorithm, unless P=NP)

Solution #3: (For now) Assume input graph has no negative cycles.

Later: Will show how to quickly check this condition.

Quiz

Quiz: Suppose the input graph G has no negative cycles. Which of the following is true? [Pick the strongest true statement.] [n = #] of vertices, m = #] of edges]

- A) For every v, there is a shortest s-v path with $\leq n-1$ edges.
- B) For every v, there is a shortest s-v path with $\leq n$ edges.
- C) For every v, there is a shortest s-v path with $\leq m$ edges.
- D) A shortest path can have an arbitrarily large number of edges in it.



Algorithms: Design and Analysis, Part II

Optimal Substructure

Single-Source Shortest Path Problem, Revisited

Input: Directed graph G = (V, E), edge lengths c_e [possibly negative], source vertex $s \in V$.

Goal: either

(A) For all destinations $v \in V$, compute the length of a shortest s-v path \rightarrow focus of this + next video

OR

(B) Output a negative cycle (excuse for failing to compute shortest paths) \rightarrow later

Optimal Substructure (Informal)

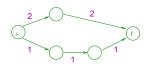
Intuition: Exploit sequential nature of paths. Subpath of a shortest path should itself be shortest.

Issue: Not clear how to define "smaller" & "larger" subproblems.

Key idea: Artificially restrict the number of edges in a path.

Subproblem size ← Number of permitted edges

Example:



Optimal Substructure (Formal)

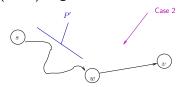
Lemma: Let G = (V, E) be a directed graph with edge lengths c_e and source vertex s.

[G might or might not have a negative cycle]

For every $v \in V$, $i \in \{1, 2, ...\}$, let $P = \text{shortest } s\text{-}v \text{ path } \underline{\text{with at}}$ most i edges. (Cycles are permitted.)

Case 1: If P has $\leq (i-1)$ edges, it is a shortest s-v path with $\leq (i-1)$ edges.

Case 2: If P has i edges with last hop (w, v), then P' is a shortest s-w path with $\leq (i-1)$ edges.



Proof of Optimal Substructure

Case 1: By (obvious) contradiction.

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Case 2: If Q (from s to w, \leq (i-1) edges) is shorter than P' then Q + (w, v) (from s to v, \leq i edges) is shorter than P' + (w, v) (= P) which contradicts the optimality of P. QED!
```

Quiz

Question: How many candidates are there for an optimal solution to a subproblem involving the destination v?

- A) 2
- B) 1 + in-degree(v)
- C) n-1
- D) n

1 from Case 1+1 from Case 2 for each choice of the final hop (w,c)



Algorithms: Design and Analysis, Part II

The Basic Algorithm

The Recurrence

Notation: Let $L_{i,v} = \text{minimum length of a } s-v \text{ path with } \leq i \text{ edges.}$

- With cycles allowed
- Defined as $+\infty$ if no s-v paths with $\leq i$ edges

Recurrence: For every $v \in V$, $i \in \{1, 2, ...\}$

$$L_{i,v} = \min \left\{ \begin{array}{l} L_{(i-1),v} & \text{Case 1} \\ \min_{(u,v) \in E} \{L_{(i-1),w} + c_{wv}\} & \text{Case 2} \end{array} \right\}$$

Correctness: Brute-force search from the only (1+in-deg(v)) candidates (by the optimal substructure lemma).

If No Negative Cycles

Now: Suppose input graph G has no negative cycles.

- ⇒ Shortest paths do not have cycles [removing a cycle only decreases length]
- \Rightarrow Have $\leq (n-1)$ edges

Point: If G has no negative cycle, only need to solve subproblems up to i = n - 1.

Subproblems: Compute $L_{i,v}$ for all $i \in \{0,1,\ldots,n-1\}$ and all $v \in V$.

Let A = 2-D array (indexed by i and v)

Base case:
$$A[0, s] = 0$$
; $A[0, v] = +\infty$ for all $v \neq s$.

For
$$i = 1, 2, ..., n - 1$$

For each $v \in V$

$$A[i, v] = \min \left\{ \begin{array}{l} A[i-1, v] \\ \min_{(w,v) \in E} \{A[i-1, w] + c_{wv}\} \end{array} \right\}$$

As discussed: If G has no negative cycle, then algorithm is correct [with final answers $= A[n-1, \nu]$'s]

Example

$$A[i, v] = \min \left\{ \begin{array}{l} A[i-1, v] \\ \min_{(w,v) \in E} \{A[i-1, w] + c_{wv}\} \end{array} \right\}$$

i = 1 i = 2 i = 3 i = 4

Quiz

Question: What is the running time of the Bellman-Ford algorithm? [Pick the strongest true statement.] [m = # of edges, n = # of vertices]

- A) $O(n^2) \to \#$ of subproblems, but might do $\Theta(n)$ work for one subproblem
- B) *O(mn)*
- C) $O(n^3)$
- D) $O(m^2)$

```
Reason: Total work is O(n) \sum_{v \in V} \text{in-deg}(v) = O(mn)
# iterations of outer loop (i.e. choices of i) work done in one iteration = m
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Stopping Early

Note: Suppose for some j < n-1, A[j, v] = A[j-1, v] for all vertices v.

- \Rightarrow For all v, all future A[i, v]'s will be the same
- \Rightarrow Can safely halt (since A[n-1, v]'s = correct shortest-path distances)



Algorithms: Design and Analysis, Part II

Detecting Negative Cycles

Checking for a Negative Cycle

Question: What if the input graph G has a negative cycle? [Want algorithm to report this fact]

Claim:

G has no negative-cost cycle (that is reachable from s) \iff In the extended Bellman-Ford algorithm, A[n-1,v]=A[n,v] for all $v \in V$.

Consequence: Can check for a negative cycle just by running Bellman-Ford for one extra iteration (running time still O(mn)).

Proof of Claim

- (⇒) Already proved in correctness of Bellman-Ford
- (\Leftarrow) Assume A[n-1,v] = A[n,v] for all $v \in V$. (Assume also these are finite (< + ∞))

Let d(v) denote the common value of A[n-1, v] and A[n, v].

Recall algorithm:
$$d(v)$$
 $d(w)$

$$A[n, v] = \min \left\{ \begin{array}{l} A[i-1, v] \\ \min_{(w,v) \in E} \{A[n-1, w] + c_{wv}\} \end{array} \right\}$$

Thus: $d(v) \le d(w) + c_{wv}$ for all edges $(w, v) \in E$

Equivalently: $d(v) - d(w) \le c_{wv}$

Now: Consider an arbitrary cycle C.

$$\sum_{(w,v)\in C} \ge \sum_{(w,v)\in C} (d(w) - d(v)) = 0$$
 QED!



Algorithms: Design and Analysis, Part II

Space Optimization

Quiz

Question: How much space does the basic Bellman-Ford algorithm require? [Pick the strongest true statement.] [m = # of edges, n = # of vertices]

- A) $\Theta(n^2) \to \Theta(1)$ for each of n^2 subproblems
- B) ⊖(*mn*)
- C) $\Theta(n^3)$
- D) $\Theta(m^2)$

Predecessor Pointers

$$A[i, v] = \min \left\{ \begin{array}{l} A[i-1, v] \\ \min_{(w,v) \in E} \{A[i-1, w] + c_{wv}\} \end{array} \right\}$$

Note: Only need the A[i-1, v]'s to compute the A[i, v]'s.

 \Rightarrow Only need O(n) to remember the current and last rounds of subproblems [only O(1) per destination!]

Concern: Without a filled-in table, how do we reconstruct the actual shortest paths?

Exercise: Find analogous optimizations for our previous DP algorithms.

Computing Predecessor Pointers

Idea: Compute a second table B, where B[i, v] = 2nd-to-last vertex on a shortest $s \to v$ path with $\leq i$ edges (or NULL if no such paths exist)

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("Predecessor pointers")
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Reconstruction: Assume the input graph G has no negative cycles and we correctly compute the B[i, v]'s.

Then: Tracing back predecessor pointers – the B[n-1, v]'s (= last hop of a shortest s-v path) – from v to s yields a shortest s-v path.

[Correctness from optimal substructure of shortest paths]

Computing Predecessor Pointers

Recall:

$$A[i, v] = \min \left\{ \begin{array}{l} (1) \ A[i-1, v] \\ (2) \ \min_{(w,v) \in E} \{A[i-1, w] + c_{wv}\} \end{array} \right\}$$

Base case: B[0, v] = NULL for all $v \in V$

To compute B[i, v] with i > 0:

Case 1: B[i, v] = B[i - 1, v]

Case 2: B[i, v] = the vertex w achieving the minimum (i.e., the new last hop)

Correctness: Computation of A[i, v] is brute-force search through the (1+in-deg(v)) possible optimal solutions, B[i, v] is just caching the last hop of the winner.

To reconstruct a negative-cost cycle: Use depth-first search to check for a cycle of predecessor pointers after each round (must be a negative cost cycle). (Details omitted)



Algorithms: Design and Analysis, Part II

Internet Routing

From Bellman-Ford to Internet Routing

Note: The Bellman-Ford algorithm is intuitively "distributed".

Toward a routing protocol:

(1) Switch from source-driven to destination driven

[Just reverse all directions in the Bellman-Ford algorithm]

- Every vertex v stores shortest-path distance from v to destination t and the first hop of a shortest path

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[For all relevant destinations t]
("Distance vector protocols")
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Handling Asynchrony

(2) Can't assume all A[i, v]'s get computed before all A[i - 1, v]'s

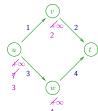
Fix: Switch from "pull-based" to "push-based": As soon as A[i,v] < A[i-1,v], v notifies all of its neighbors.

Fact: Algorithm guaranteed to converge eventually. (Assuming no negative cycles)

[Reason: Updates strictly decrease sum of shortest-path estimates]

 \Rightarrow RIP, RIP2 Internet routing protocols very close to this algorithm [see RFC 1058]

Example:



Handling Failures

Problem: Convergence guaranteed only for static networks (not true in practice).

Counting to Infinity:

$$\begin{array}{c|cccc}
s & \hline & v & \hline & t \\
 & \chi & 1 & 0 \\
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 &$$

Fix: Each V maintains entire shortest path to t, not just the next hop.

"Path vector protocol" "Border Gateway Protocol (BGP)"

Con: More space required.

Pro#1: More robust to failures.

Pro#2: Permits more sophisticated route selection (e.g., if you care about intermediate stops).