

Complexity Theory of Polynomial-Time Problems

Lecture 9: Dynamic Algorithms I

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Today's Plan

- 1. Decremental SSSP via Even-Shiloach tree
- 2. Decremental APSP



Floyd-Warshall Algorithm

 $O(n^3)$ algorithm for computing All-Pairs Shortest Paths (APSP) n: number of nodes

Put some order $v_1, ..., v_n$ on the nodes Set $d(v_i, v_j) = w(v_i, v_j)$ for every pair of nodes $v_i \neq v_j$ For k = 1 to n: For every pair of nodes v_i, v_j : $d(v_i, v_j) \leftarrow \min(d(v_i, v_j), d(v_i, v_k) + d(v_k, v_j))$

Running Time: *n* iterations, each takes time $O(n^2)$

Correctness: After iteration *i*, $d(\cdot, \cdot)$ gives correct distance in graph restricted to $\{v_1, \dots, v_k\}$ \Rightarrow Correct in full graph after iteration *n*



Dynamic View

Why stop after n iterations?

Floyd-Warshall allows insertions of new nodes

 $\frac{\text{Insert}(v, In_v, Out_v):}{\text{Set } d(v', v) = w(v', v) \text{ for every incoming neighbor } v' \text{ of } v}$ Set d(v, v') = w(v, v') for every outgoing neighbor v' of v

For every incoming neighbor *s* of *v* and every node *t* $d(s,t) \leftarrow \min(d(s,t), d(s,v) + d(v,t))$ For every node *s* outgoing neighbor *t* of *v* $d(s,t) \leftarrow \min(d(s,t), d(s,v) + d(v,t))$ For every other pair of nodes *s*, *t*: $d(s,t) \leftarrow \min(d(s,t), d(s,v) + d(v,t))$

Update Time: $O(n^2)$ per insertion



Dynamic Algorithms

A dynamic graph algorithm is a **data structure** supporting:

- **Preprocess**(*G*): preprocess the graph *G*
- Insert(*u*,*v*): insert the edge (*u*,*v*) into *G*
- Delete(*u*, *v*): delete the edge (*u*, *v*) from *G*
- Query(G): return result of algorithm for **current** graph G

Terminology:

- Incremental: only insertions are supported
- Decremental: only deletions are supported
- Fully dynamic: both insertions and deletions are supported

Some algorithms also support insertions and deletions of nodes

Goal:

- Time spent per update or query less than recomputing from scratch
- (Polynomial preprocessing time)



Measuring Update Time

Two Measures

- Worst-case update time Fixed upper bound on running time per update
- Amortized update time "On average" upper bound on running time per update

Formally:	Amortized update time $u(n, m)$ if total time spent for a
	sequence of t updates is at most $t \cdot u(n, m)$.

Very common in incremental/decremental algorithms:

- Amortize update time over *m* insertions/deletions
- "Total update time"



1. Decremental SSSP



Even-Shiloach Algorithm

Goal: Decremental SSSP in unweighted graphs from source *s*

Example of shortest path tree from *s*:





Deletion Procedure I



- c loses its parent
- *c* finds no new parent at level 2
- *c* increases level to 3
- *c* informs neighbors about level increase
- Children of c lose their parent

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- e finds new parent b
- *f* finds new parent *bg* finds no new parent at level 3
- g increases level to 4

Deletion Procedure II





- g finds new parent c
- Now we are done because all nodes have a parent again



Internal Data Structures and Initialization

Data Structures:

For every node v:

- Number neighbors of v from 1 to deg(v) (initial degree of v)
- $n_i(v)$ Pointer to *i*-th neighbor of v
- p(v) Index of parent of v (among neighbors) in tree
- $\ell(v)$ Level of v in tree (will correspond to distance from root)

<u>Global:</u>

 Q Priority queue with levels as keys (used in update procedure)

Initialization:

Compute BFS tree from source *s* such that each node takes parent with minimum index among neighbors.

Time: *0(m)*



Pseudocode





Correctness I

Claim 1: Initially, and after each update is finished: $\ell(v) \ge \operatorname{dist}(s, v) \forall v$

Proof: If $\ell(v) = \infty$, then certainly true

Otherwise:

Consider path π from v induced by following parents Levels of nodes on π are strictly decreasing:

- When parent of a node is set, parent has strictly smaller level
- When level of a node changes it informs all potential children

Thus, π ends at *s* because *s* is the only node at level 0

 $\ell(v) = \text{length of } \pi$ π cannot be shorter than shortest path from *s* to *v* Thus, $\ell(v) \ge \text{dist}(s, v)$



Correctness II

Claim 2: At any time: For every node v with neighbor u, $\ell(v) \le \ell(u) + 1$ if $\ell(u) + 1 \le n - 1$.

Proof:

By induction on #level increases of v (in total over all deletions) Induction Base: True after initialization

Induction Step: $\ell(v)$: level of v directly before level increase $\ell'(v)$: level of v directly after level increase

By IH: $\ell(v) \le \ell(u) + 1$ Algorithm guarantees: $\ell(v) < \ell(u) + 1$ (otherwise no level increase of v)

(Detail: no candidate parent for v at level $\ell(v)$ anymore by processing order according to levels)

Thus: $\ell(v) + 1 \le \ell(u) + 1$ Since $\ell'(v) = \ell(v) + 1$ we have $\ell'(v) \le \ell(u) + 1$

Inequality remains true until next level increase of v because level of u never decreases



Correctness III

Lemma: Initially and after each update is finished, $\ell(v) = \text{dist}(s, v) \forall v$

Proof by induction on distance to s If dist $(s, v) = \infty$: Then $\ell(v) \ge \text{dist}(s, v) = \infty$ by Claim 1 If dist $(s, v) < \infty$: Consider successor u of v on shortest path from s to v



When algorithm finished update: $\ell(u) = \operatorname{dist}(s, u)$ by IH In particular: $\operatorname{dist}(s, u) \le n - 2$ and thus $\ell(u) + 1 \le n - 1$

By Claim 2: $\ell(v) \le \ell(u) + 1 = \operatorname{dist}(s, v)$ By Claim 1: $\ell(v) \ge \operatorname{dist}(s, v)$ $\Rightarrow \ell(v) = \operatorname{dist}(s, v)$

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Running Time

Lemma: The total update time over **all** deletions is *O(mn)*

(where m is the number of edges at initialization)

Amortized analysis!

Idea: Every time the level of some node v increases, we charge running time of $O(\deg(v))$ to that level increase (see next slide).

(where deg(v) is the degree of v at initialization)

The level of every node can increase at most n - 1 times (max. distance).

Additionally, charge time O(1) to every deletion

Total time: $O(\# del + \sum_{v \in V} n deg(v)) = O(m + n \cdot \sum_{v \in V} deg(v)) = O(n \cdot m)$

Remember from kindergarten: sum of degrees \leq twice #edges



Running Time Analysis



Total: $O(\# del + \sum_{v \in V} n deg(v)) +$



Banker's View



Every node v receives:

- $10 \deg(v)$ coins at initialization
- 3 coins when deleting incident edges

Observation: Sufficient number of coins to pay 1 coin per operation.

(Note: give constant number of coins to each neighbor at level increase)

Total number of coins spent: $O(\#del + \sum_{v \in V} n \deg(v))$



Implementing Priority Queue

Standard heap: Time $O(\log n)$ per operation

In our application we can get O(1) per operation

Array A of size n, where A[i] contains pointer to list of nodes at level i



In unweighted undirected graphs:

- At most two lists non-empty
- at consecutive levels



Extensions

Theorem: Maintaining SSSP under deletions takes total time

- *O(mn)* in unweighted undirected graphs
- O(mn) in unweighted directed graphs
- O(mnW) in directed graphs with weights $\{1, 2, ..., W\}$.

[Even/Shiloach '81, King '99, King/Thorup '01]

Theorem: Maintaining SSSP under deletions up to depth *D* takes total time O(mD) in directed graphs with integer weights.



2. Decremental APSP



Hitting Set for Long Paths

Random process for picking a set of nodes *S*:

- Set $p = \min\left(\frac{10\log n}{h}, 1\right)$
- Iterate over all nodes
- Pick each node with probability *p* independently (flip biased coin)
- Expected size of S: $O\left(\frac{n \log n}{h}\right)$

Lemma: For every pair of nodes *s* and *t*, if the shortest path from *s* to *t* contains at least *h* nodes, then one of them is from *S* with probability at least $1 - \frac{1}{n}$ (i.e., 'with high probability').

Caveat: There could be many shortest paths from *s* to *t*. We only guarantee to hit one of them (e.g. lexicographic shortest path).



Lemma also holds for all graphs during a sequence of deletions (if sequence of deletions is independent from random choices of algorithm)



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Maintaining shortest paths in range $2^i \dots 2^{i+1}$



Decremental APSP algorithm

For i = 1 to $\lfloor \log n \rfloor$:

- Pick *i*-centers S_i with sampling probability $p = \min\left(\frac{10 \log n}{2^i}, 1\right)$
- For every *i*-center $c \in S_i$: Maintain ES-tree to and from *c* of depth 2^{i+1}

Total update time: $O\left(\sum_{i=1}^{\lfloor \log n \rfloor} |S_i| m 2^i\right) = O\left(\sum_{i=1}^{\lfloor \log n \rfloor} m n \log n\right) = O(m n \log^2 n)$

Query Algorithm:

- Question: What is the distance from s to t
- Return minimum value of $\hat{d}(s,c) + \hat{d}(c,t)$ among all centers $c \in \bigcup S_i$
- Query time: O(n) (= number of centers)

Correctness:

- Let π be shortest path from s to t
- π has between 2^i and 2^{i+1} nodes for some i = 1 to $\lfloor \log n \rfloor$
- π contains a center $c \in S_i$ with high probability
- Subpaths from s to c and from c to s are also shortest paths and both have length $\leq 2^{i+1}$
- Thus, $\hat{d}(s,c) + \hat{d}(c,t) = \operatorname{dist}(s,t)$
- Other centers can never report a smaller value for $\hat{d}(s,c) + \hat{d}(c,t)$

Extensions

Result we just showed:

Theorem: There is a decremental algorithm for maintaining APSP in unweighted, directed graphs with total update time $O(mn \log^2 n)$ and query time O(n).

By explicitly maintaining distances after each update, one can reduce query time.

Theorem: There is a decremental algorithm for maintaining APSP in unweighted, directed graphs with total update time $O(n^3 \log^2 n)$ and **constant** query time.

[Baswana et al. '02]

