

CS 7301.003.20F Lecture 10–September 21, 2020

Main topics are `#multiple-source_shortest_paths`.

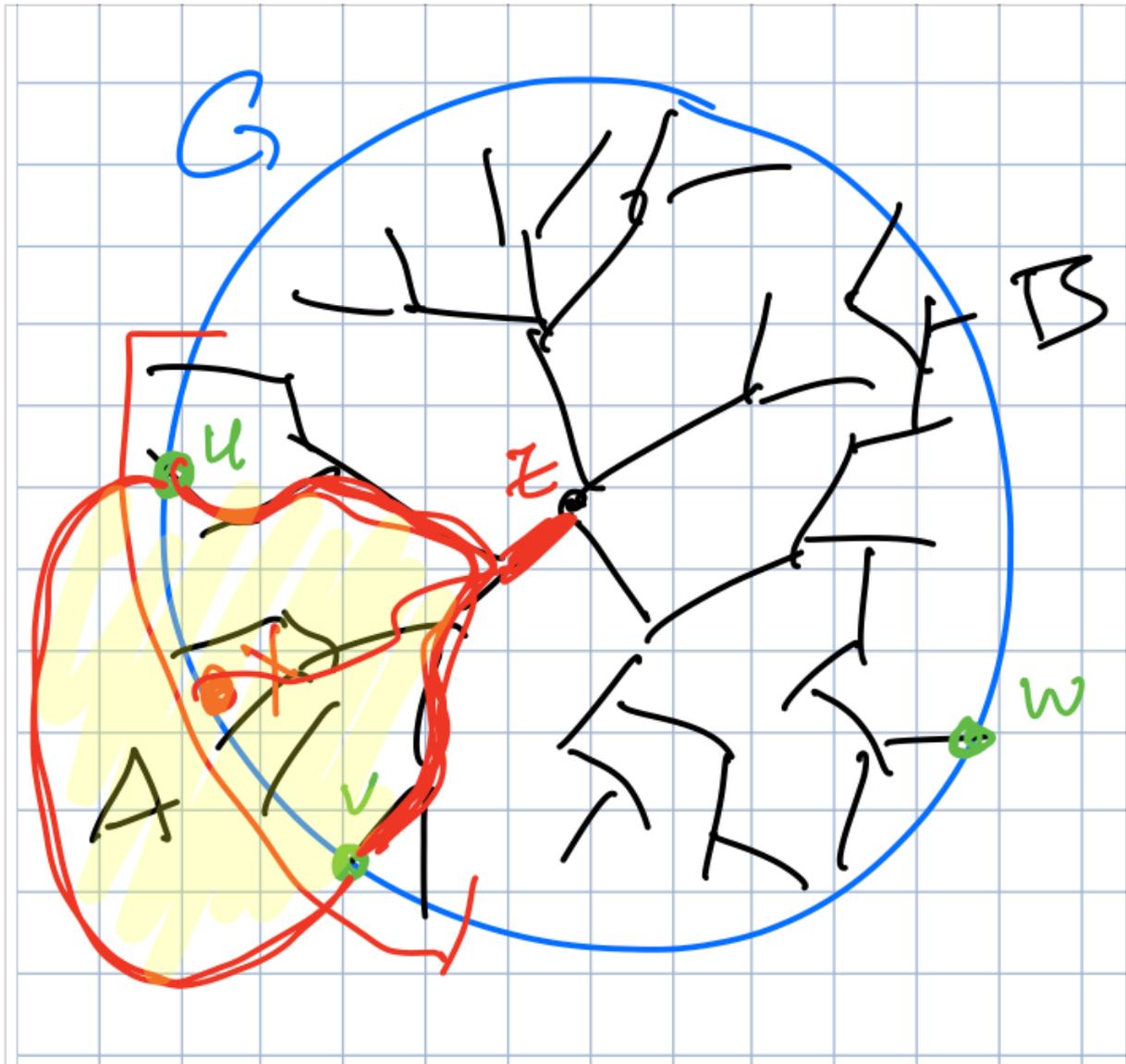
Multiple-Source Shortest Paths

- At the end of the last lecture, I mentioned an algorithm that uses dense distance graphs (DDGs). We're given an r -division of the edges where every piece has at most r edges and $O(\sqrt{r})$ boundary vertices lying on $O(1)$ faces. We need to compute shortest paths within each piece between its boundary vertices, and we need to do so quickly.
- Today, we're going to look at (a modified) algorithm by Klein ['05] that computes so-called *multiple-source shortest paths*.
- Given a directed(!) plane graph $G = (V, E)$ with outer face o and a non-negative weight function $w : D \rightarrow \mathbb{R}^+$, we want to compute all shortest paths rooted at every boundary vertex of o .
- If o has k vertices, then we could run Dijkstra once per vertex for $O(kn \log n)$ time total.
- But Klein claims $O(n \log n)$ time!
- Of course, finding all shortest paths this fast is impossible if $k = \omega(\log n)$. We need to be more restrictive on what we accomplish.
 - If we ask for p source-destination pairs in advance, then we can find all their distances in $O(n \log n + p \log n)$ time.
 - Or we can build a data structure in $O(n \log n)$ time using $O(n \log n)$ space that answers distance queries in $O(\log n)$ time per query.
 - Or we can find the shortest paths in $O(1)$ time per edge if we know the paths in advance or $O(\log \Delta)$ time per edge using the data structure if the graph has max degree Δ .

Trees and Disks

- Our algorithm is going to compute a shortest path tree T from some vertex on o in $O(n \log n)$ time. Then, we're going to move the source vertex-by-vertex around o .
- Every time we move to the next vertex, some edges leave T and are replaced by other edges. The big surprise that makes the result possible is that only $O(n)$ edges leave and enter the tree during one pass around o .
- The small number of changes is largely do to the following observation.
- Pick *any* spanning tree T of G and an edge e in T .
- $T \setminus e$ has 2 component trees. Label them A and B .
- Lemma: The set of outer face vertices in A is either empty, includes all vertices of o , or it induces a consecutive path of vertices in o .

- Suppose u, v , and w are boundary vertices such that u, v in A and w not in A .
- Let z be the least common ancestor of u and v after rooting T on an endpoint of e .
- The path from z to u along T , from u to v along a path just outside o , and from v to z along T bounds a disk.
- For any x between u and v on o , the path from x to e in T begins inside the disk and doesn't leave until hitting z . So x in A .



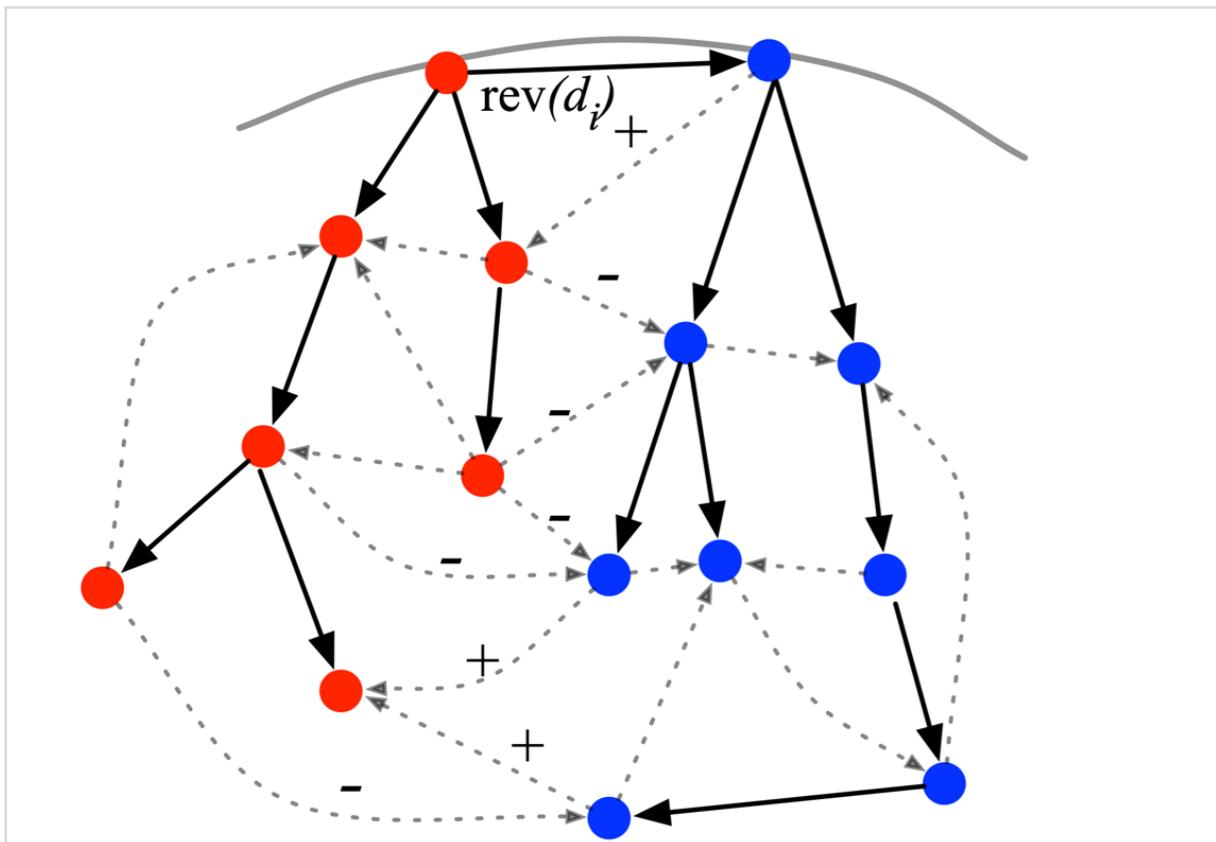
- Let s_0, s_1, \dots be the vertices on o and T_0, T_1, \dots be their shortest path trees.
- Pick any dart $u \rightarrow v$. Which trees T_i contain $u \rightarrow v$?
- Let T_v be a tree of shortest paths into v .
- Assuming each T_i follows the same path s_i, v -path as in T_v , the set of T_i containing $u \rightarrow v$ are those for which s_i lies in the u side of $T_v \setminus uv$.
- But we just saw those s_i are contiguous, so $u \rightarrow v$ will enter the shortest path tree at most once and leave it at most once.
- There are $O(n)$ darts, so the shortest path tree undergoes $O(n)$ changes total.
- We'll assume there is exactly one shortest path between any pair of vertices so that this analysis holds. There are ways to enforce this.

Changing Sources

- Suppose we've computed $T := T_{i-1}$ and we want to compute T_i . There's a couple ways we could do this that all use about the same sequence of computations.
- I'm going to describe a strategy by Eisenstat and Klein [13].
- We're going to incrementally transform T into T_i using a sequence of *pivots* where one dart leaves T , and another takes its place. After all the pivots are complete, we'll have T_i .
- First, we do the *special pivot*. We remove the dart going into s_i and replace it with dart $s_i \rightarrow s_{i-1}$. We now have s_i as the root of T .
- Doing special pivot with no other changes means T may not be a shortest path tree. Therefore, we will temporarily assign $s_i \rightarrow s_{i-1}$ a new weight of $\lambda := -\text{dist}(s_{i-1}, s_i)$.
- Define the *slack* of a dart as $\text{slack}(u \rightarrow v) := \text{dist}(u) + w(u \rightarrow v) - \text{dist}(v)$ (where dist uses distance from the current shortest paths source vertex).
- Ford showed $\text{slack} \geq 0$ for all edges of G , and $\text{slack} = 0$ for all edges appearing on shortest paths. Since we're assuming shortest paths are unique, the slack is 0 if and only if the edge is in the shortest paths tree.
- Claim: Immediately after the special pivot and reweighting of $s_i \rightarrow s_{i-1}$, T is a shortest path tree rooted at s_i .
 - I claim the distance to any vertex x dropped by $\text{dist}(s_{i-1}, s_i)$ when we did the special pivot and reweighted $s_{i-1} \rightarrow s_i$.
 - If the new shortest path to x uses s_{i-1} , then we have prepended the old path with dart $s_{i-1} \rightarrow s_i$.
 - If the new shortest path doesn't use s_{i-1} , then the old path from s_{i-1} to s_i to x just had its prefix from s_{i-1} to s_i removed.
 - Other than $s_{i-1} \rightarrow s_i$, no weights changed, so all other slacks are the same. Finally, $\text{slack}(s_{i-1} \rightarrow s_i) = 0$ by our choice of λ .
 - So T is a shortest path tree rooted at s_i , but only because $s_i \rightarrow s_{i-1}$ has a new weight.
- We need to recover the original weight, so we'll "continuously" increase λ , the current weight of $s_i \rightarrow s_{i-1}$, until $\lambda = w(s_i \rightarrow s_{i-1})$, the original weight of the dart.
- What happens as we increase λ ? $T \setminus (s_{i-1}, s_i)$ has two components. Those in the s_i component retain their distance from s_i . Color these vertices red. Those in the s_{i-1} component have their distances from s_i increase at the same rate λ is increasing. Color those blue. (Here, I'm using Klein and Mozes figures/colors. Erickson uses the opposite colors.)
- Meanwhile, what is happening to the slacks? $\text{blue} \rightarrow \text{blue}$ and $\text{red} \rightarrow \text{red}$ darts don't

change slack. blue \rightarrow red darts have their slack increase at the same rate λ is increasing.

- But red \rightarrow blue darts (except for $s_{i-1} \rightarrow s_i$) have their slacks decrease at the same rate λ increases.



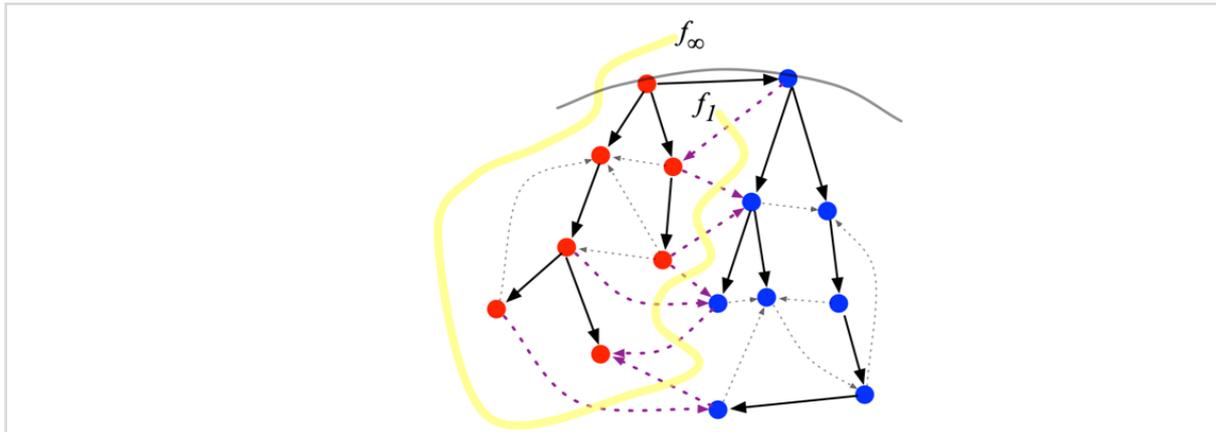
(Here, $\text{rev}(d_i) = s_i \rightarrow s_{i-1}$.)

- Eventually, one of the red \rightarrow blue darts $x \rightarrow y$ hits slack 0 and is about to go negative. Let $z \rightarrow y$ be the predecessor dart of y in T . We can now pivot $z \rightarrow y$ out of T , replacing it with $x \rightarrow y$. And T is still a shortest path tree!
- $z \rightarrow y$ was blue \rightarrow blue but is now blue \rightarrow red. So its 0 slack starts to increase. We may now continue increasing λ .
- After a sequence of pivots, λ eventually reaches $w(s_i \rightarrow s_{i-1})$ and we've computed T_i .

Fast Pivots

- We still need a fast way to find each edge $x \rightarrow y$ that pivots into T and perform the pivot.
- $T \setminus (s_{i-1} \rightarrow s_i)$ has red and blue components. The red-blue edges form an edge cut. In particular, they form a bond.
- The dual of this bond is a cycle.
- Let $C := (G \setminus T)^*$ be the complementary spanning tree. The dual cycle of red-blue edges has exactly one edge outside of C , namely $s_{i-1} \rightarrow s_i$. The red \rightarrow blue darts with decreasing slack form a path in C from $\text{right}(s_{i-1} \rightarrow s_i)^*$ to $\text{left}(s_{i-1} \rightarrow s_i)^*$. The

darts with increasing slack form the reversal of that path.



(Here, the path goes from f_1 to f_{∞} .)

- At this point, I'm going to punt and use a couple data structures as black boxes. Describing how they work would take another week.
- A *dynamic forest* data structure maintains a changing collection of darts between a fixed set of vertices. We need to guarantee that the edges always form a forest.
- We can do the following operations (and more!) in $O(\log n)$ time per operation:
 - Link (insert) an edge between given vertices u and v .
 - Cut (delete) the edge between u and v .
 - Read or write a weight on a given vertex or dart.
 - Add a value to the weight of all darts in the unique path between given vertices u and v .
 - Add a value to the weight of all vertices in a subtree rooted at a given vertex u .
 - Find the minimum weight dart in a path.
- So, we store both T and C in separate dynamic forests. We store distances on vertices of T and slacks on darts of C . To find a normal pivot, we ask for the minimum slack Δ on the path in C .
- To do a pivot, we decrease the slacks on the path by Δ and increase the slacks on the reversal by Δ . We increase distances to all blue vertices by Δ . We do a pair of links and cuts and a pair of slack assignments to pivot the actual edges.
- So that's $O(\log n)$ time doing a constant number of dynamic forest operations each pivot. Earlier, we saw there are $O(n)$ pivots across the whole algorithm, so we can find all the shortest path trees in $O(n \log n)$ time total.
- We can look up distances as we need them in $O(\log n)$ time per distance.
- And if we use something called *persistence*, we can remember how our data structure progressed over time so we can look up shortest path distances between vertices of o and other vertices later.