#### COM S 783 / OR&IE 634 Approximation Algorithms

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Lecturer: Yuval Rabani

In today's lecture, we will present a randomized algorithm that embeds any arbitrary metric into a dominating tree metric such that each edge is distorted, in expectation, in the worst case by a factor of  $O(\log n)$  and such that each edge is not contracted.

## 1 Overview and Notation

Recall that a metric on graph G=(V,E) is a function  $d:E\to\mathbb{R}$ . One can define a metric  $\mathcal{M}=(X,d)$  on an arbitrary set of vertices X with a distance function d such that the following properties hold:

- $d: X \times X \to \mathbb{R}$
- $d(i, j) \ge 0$
- $d(i,j) = 0 \Leftrightarrow i = j$
- d(i, j) = d(j, i)
- d(i, j) + d(j, k) > d(i, k)

A tree metric is the shortest path metric of a weighted tree. In other words, d(i, j) is the length of the unique shortest path between node i and node j.

A metric (V', d') is said to dominate metric (V, d) if  $\forall u, v \in V, d'(u, v) \geq d(u, v)$ .

WLOG,  $\forall i, j, d(i, j) \geq 1$ .

Let  $\Delta$  denote the diameter of the metric (V, d). WLOG,  $\Delta = 2^{\delta}$ .

Let S be a family of metrics over V, and D be a distribution of S. We say that (S, D)  $\alpha$ -probabilistically approximates a metric (V, d) if every metric in S dominates d and for every pair of vertices  $(u, v) \in V$ ,  $E_{d' \in (S, D)}[d'(u, v)] \leq \alpha \cdot d(u, v)$ . Formally, we're interested in  $O(\log n)$ -probabilistically approximating an arbitrary metric (V, d) by a distribution over tree metrics.

For a parameter r, an r-cut decomposition of (V, d) is a partitioning of V into clusters, each centered around a node and having radius at most r. Thus each cluster will have diameter at most 2r.

A hierarchical cut decomposition of (V, d) is a sequence of  $\delta + 1$  nested cut decompositions  $D_0, D_1, \ldots, D_\delta$  such that

- $D_{\delta} = \{V\}$ , the trivial partition where all nodes are in a single cluster
- $D_i$  is a  $2^i$ -cut decomposition, and a refinement of  $D_{i+1}$ .

Note that each cluster in  $D_0$  has radius at most 1. Hence, each cluster consists of just a unique node.

Recall that a laminar family  $\mathcal{F} \subseteq 2^V$  is a family of subsets of V such that for any  $A, B \in \mathcal{F}$ , it si the case that  $A \subseteq B$  or  $B \subseteq A$  or  $A \cap B = \emptyset$ .

A hierarchical cut decomposition defines a laminar family. A root tree can be generated from the decomposition as follows: Each set in the laminar family is a node in the tree and the children of a node corresponding to a set S are the nodes corresponding to maximal subsets of S in the family.

**Remark 1:** The node corresponding to V is the root and the singletons are the leaves.

**Remark 2:** The children of a set in  $D_{i+1}$  are sets in  $D_i$ .

Define a distance function on this tree as follows: The links from a node S in  $D_i$  to each of its children in the tree have length equal to the  $2^i$  (which is an upper bound on the radius of S). Hence, define the distance function  $d^T(\cdot, \cdot)$  on the tree to be the length of the shortest (unique) path distance in T from node u to node v. It's obvious that  $\forall u, v, d^T(u, v) \geq d(u, v)$ .

An edge (u, v) is at level i if u and v are first separated in the decomposition  $D_i$ . Note that if (u, v) is at level i, then  $d^T(u, v) = 2 \sum_{j=0}^{i} 2^j \le 2^{i+2}$ .

# 2 Algorithm

Below is the random process that defines a hierarchical cut decomposition of (V, d), such that the probability that an edge (u, v) is at level i decreases geometrically with i:

- Pick a random permutation  $\pi$  of  $\{v_1, \ldots, v_n\}$ ,
- Select  $\beta$  uniformly at random in the interval [1, 2]
- For each i, we compute  $D_i$  from  $D_{i+1}$  as follows. First, set  $\beta_i$  to be  $2^{i-1}\beta$ . Let S be a cluster in  $D_{i+1}$ . Assign a node  $u \in S$  to the first (as defined by  $\pi$ ) node  $v \in V$  closer than  $\beta_i$  to u. Each child cluster of S in  $D_i$  then consists of the set of vertices in S assigned to a single center v.

**Remark 1:** The center v itself need not be in S. Thus, one center v may correspond to more than one cluster, each inside a different level (i + 1) cluster.

**Remark 2:** Since  $\beta_i \leq 2^i$ , the radius of each cluster is at most  $2^i$ . Thus, we get a  $2^i$ -cut decomposition.

## 3 Analysis

For a fixed edge (u, v), we'll show  $E[d^T(u, v)] \leq O(\log n) \cdot d(u, v)$ .

From previous remarks after the algorithm presented above, it follows that

$$E[d^T(u,v)] \le \sum_{i=0}^{\delta} Pr[(u,v) \text{ is at level } i] \cdot 2^{i+2}$$

If nodes u and v are in separate clusters in  $D_i$ , we say that  $D_i$  separates (u, v).

Based on that definition, (u, v) is at level i if

- (a)  $D_i$  separates (u, v).
- (b)  $D_j$  does not separate (u, v) for any j > i.

Clearly if  $d(u, v) > 2^{i+2}$ , then u and v cannot be in the same cluster in  $D_{i+1}$ , i.e.  $D_{i+1}$  separates (u, v). From (b) above, (u, v) cannot be at level i. Let  $j^*$  be the smallest i such that  $d(u, v) \leq 2^{i+2}$ . Thus,  $\Pr[(u, v) \text{ is at level } i] = 0$  for any  $i < j^*$ . For  $i \geq j^*$ , we'll find an upper bound of the probability that (u, v) is at level i.

From (a) and (b) above, for any  $i \ge j^*$ ,

Pr[(u, v) is at level i]

=  $Pr[D_i \text{ separates } (u,v)] \cdot Pr[\nexists j > i : D_j \text{ separates } (u,v)|D_i \text{ separates } (u,v)]$ 

 $\leq Pr[D_i \text{ separates } (u, v)]$ 

For any  $j^* \leq j \leq \delta$ , let  $K^u_j$  be the set of vertices in V closer than  $2^j$  to node u, and let  $k^u_j := |K^u_j|$ . Define  $K^v_j$  and  $k^v_j$  similarly. For  $j < j^*$ , let  $k^u_j = 0$ .

Now consider the clustering step at level  $i \geq j^*$ . In each iteration, all unassigned nodes v such that  $d(v, \pi(l)) \leq \beta_i$  assign themselves to  $\pi(l)$ . For some initial iterations of this procedure, both u and v remain unassigned. Then at some step l, at least one of u and v gets assigned to the center  $\pi(l)$ .

Center  $\pi(l)$  settles the edge (u, v) at level i if its is the first center to which at least one of u and v gets assigned. Note that exactly one center settles any edge (u, v) at any particular level.

Center  $\pi(l)$  cuts the edge e = (u, v) at level i if it settles e at this level, but exactly one of u and v is assigned to  $\pi(l)$  at level i. Clearly,  $D_i$  separates (u, v) iff some center w cuts it at this level. Hence  $Pr[D_i$  separates  $(u, v)] = \sum_{w} Pr[w \text{ cuts } (u, v) \text{ at level } i]$ .

Center w cuts u out of (u, v) at level i if w cuts (u, v) at this level and u is assigned to w (and v is not assigned to w) at this level.

For each center w, we'll find an upper bound for the probability that w cuts u out of (u, v) at level i. Arrange the centers in  $K_i^u$  in increasing order of distance from u, say  $w_1, w_2, \ldots, w_{k_i^u}$ . For a center  $w_s$  to cut (u, v) such that only u is assigned to  $w_s$ , it must be the case that

- (a)  $d(u, w_s) \leq \beta_i$ .
- (b)  $d(v, w_s) > \beta_i$ .
- (c)  $w_s$  settles e.

Thus  $\beta_i$  must lie in  $[d(u, w_s), d(v, w_s)]$ . By the triangle inequality,  $d(v, w_s) \leq d(v, u) + d(u, w_s)$ , and hence the interval  $[d(u, w_s), d(v, w_s)]$  is of length at most d(u, v). Since  $\beta_i$  is distributed uniformly in  $[2^{i-1}, 2^i]$ , the probability that  $\beta_i$  falls in the bad interval is at most  $d(u, v)/2^{i-1}$ . Moreover, for such a value of  $\beta_i$ , any of  $w_1, w_2, \ldots, w_s$  can settle (u, v) at level i and hence the first amongst these in the permutation  $\pi$  will. Since  $\pi$  is a random permutation, the probability that  $w_s$  is the one to settle (u, v) at level i is at most 1/s.

At this point, it's obvious that

 $Pr[D_i \text{ separates } (u, v)]$ 

$$\leq \sum_{s=1}^{k_i^u} (d(u,v)/2^{i-1}) \cdot \frac{1}{s} + \sum_{s=1}^{k_i^v} (d(u,v)/2^{i-1}) \cdot \frac{1}{s}$$
  
$$\leq (d(u,v)/2^{i-1})(\ln k_i^u + \ln k_i^v)$$

Thus, each i contributes at most  $O(\log n)$  to the expected value of  $d^T(u, v)$ . Hence, the expected length is bounded by  $O(\log n \log \Delta)$ .

However, we want an upper bound of  $O(\log n)$ . Observe that the total number of centers over all  $\delta$  levels is n. A more careful analysis of the above procedure will give the desire result.

First consider some  $i \geq j^* + 4$ . Since the radius of the cluster at level i is at least  $2^{i-1}$ , centers very close to both u and v can never cut the edge (u, v). More precisely, for any w in  $K_{i-2}^u$ , if u is assigned to w, it must be the case that v gets assigned to w also, because  $d(v, w) \leq d(v, u) + d(u, w) \leq 2^{i-2} + 2^{i-2} \leq 2^{i-1} \leq \beta_i$  (since  $i \geq j^* + 4$ ). Thus, no center in  $w_1, w_2, \ldots, w_{k_{i-2}^u}$  can ever cut u out of (u, v). This implies that the probability that u gets cut out of edge e is in fact bounded by

$$\sum_{s=k_{(i-2)}^u+1}^{k_i^u} \frac{1}{s} (d(u,v)/2^{i-1})$$

$$= (d(u,v)/2^{i-1}) \cdot (H_{k_i^u} - H_{k_{(i-2)}^u})$$

Since (u, v) can be cut when either u or v is cut out by some node, the overall probability that  $D_i$  separates (u, v) is then at most  $(d(u, v)/2^{i-1}) \cdot [H_{k_i^u} + H_{k_i^v} - H_{k_{i-2}^u} - H_{k_{i-2}^v}]$ .

For  $i = j^* + 1, j^* + 2, j^* + 3$ , this probability is bounded by  $(d(u, v)/2^{i-1}) \cdot (H_{k_i^u} + H_{k_i^v}) \le (d(u, v)/2^{i-1} \cdot 2H_n)$ .

Hence,

$$E[d^{T}(u,v)]$$

$$\leq \sum_{i=0}^{\delta} Pr[(u,v) \text{ is at level } i] \cdot 2^{i+2}$$

$$\leq \sum_{i=j^{*}}^{\delta} Pr[D_{i} \text{ separates } (u,v)] \cdot 2^{i+2}$$

$$\leq \sum_{i=j^{*}}^{j^{*}+3} 2H_{n} \cdot \frac{d(u,v)}{2^{i-1}} \cdot 2^{i+2} + \sum_{i=j^{*}+4}^{\delta} (H_{k_{i}^{u}} + H_{k_{i}^{v}} - H_{k_{(i-2)}^{u}} - H_{k_{i-2}^{v}}) \cdot \frac{d(u,v)}{2^{i-1}} \cdot 2^{i+2}$$

$$\leq 8d(u,v)(4 \cdot 2H_{n} + H_{k_{\delta}^{u}} + H_{k_{\delta}^{v}} + H_{k_{\delta-1}^{u}} + H_{k_{\delta-1}^{v}})$$

$$\leq 8d(u,v)(12H_{n})$$

$$= 96 \ln n \cdot d(u,v)$$

The third to last inequality follows because alternate terms of the summation  $\sum_{i} (H_{k_i^u} - H_{k_{(i-2)}^u})$  telescope.

### References

[1] J. Fakcharoenphol, S. Rao, K. Talwar. A Tight Bound on Approximating Arbitrary Metrics by Tree Metrics. *Proceedings of the Thirty-fifth ACM Symposium on Theory of Computing*, 448-455, 2003.