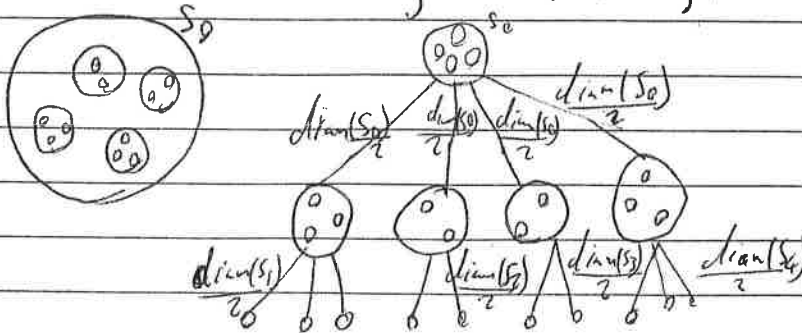


A place to network and exchange ideas.

Here we outline the probabilistically embedding an  $n$  point metric into a tree with expected distortion  $O(\log n)$  based on Bartal & FRT results.

- First decompose the graph hierarchically
- Then convert the resulting laminar family to a tree.



Let the input metric be  $(V, d)$

- The minimum distance is strictly more than 1 (by scaling)
- $\Delta$  is the diameter of  $(V, d)$
- $\Delta = 2^k$

- an  $r$ -cut decomposition of  $(V, d)$  is a partition of  $V$  into clusters, each centered around a vertex and having radius at most  $r$  (thus the diameter of each cluster is at most  $2r$ ). A hierarchical cut decomposition of  $(V, d)$  is a sequence of  $k+1$  nested cut decompositions  $D_0, D_1, \dots, D_k$  such that [note that the diameter of a cluster in  $D_j$  is at most  $2^{j+1}$ ]

- $D_0 = \{V\}$ , i.e., the trivial partition (that puts all vertices in a single cluster).
- $D_i$  is a  $2^i$ -cut decomposition, and a refinement of  $D_{i-1}$ .

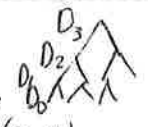
since all the distances are at least one, and since each cluster in  $D_0$  has radius at most 1, each cluster in  $D_0$  must be a singleton vertex.

A hierarchical cut decomposition defines a laminar family, and naturally corresponds to a rooted tree 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.

(Recall 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 is the case that  $A \subseteq B$ , or  $B \subseteq A$  or  $A \cap B = \emptyset$ ).

The distance  $d^T(u, v)$  is equal to the length of the shortest path distance in  $T$  from node  $u$  to node  $v$ .

Lm 1: For any tree  $T$  as generated above dominates  $d$ , i.e.,  $d_T(x, y) \geq d(x, y)$   
 Pf: let  $x, y$  be such that  $\frac{\Delta}{2^j} < d_G(x, y) \leq \frac{\Delta}{2^{j-1}}$ . Since  $x$  and  $y$  are at distance greater than  $\frac{\Delta}{2^j}$ , they cannot lie in the ~~same~~ level  $j$  component. Hence they were separated at some level  $j' \leq j$  and  $\left( \begin{array}{l} \text{from the bottom of the tree} \end{array} \right)$  the edges of length  $\frac{\Delta}{2^{j'}}$  added at this level to connect their subtrees ensure that  $x$  and  $y$  are at distance at least  $2 \frac{\Delta}{2^{j'}} \Rightarrow \frac{\Delta}{2^{j-1}} \geq d_G(x, y)$  ✓



Now, we want to put an upper bound on  $d_T(u, v)$ .

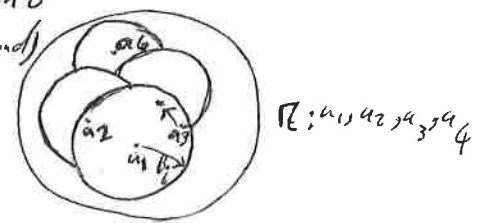
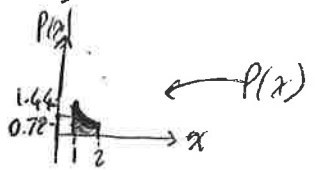
We say an edge  $(u, v)$  is at level  $i$  if  $u$  and  $v$  are first separated in the decomposition <sup>cut</sup>

$D_i$ . Note that if  $(u, v)$  is at level  $i$ , then  $d_T(u, v) = 2 \sum_{j=0}^i 2^j \leq 2^{i+2}$

Also, note the difference of diameter (the maximum distance in the graph) and weak diameter (the maximum distance in the original graph).  
 e.g. the weak diameter (and not diameter) of each cluster in  $D_j$  is at most  $2^{j+1}$

The Algorithm Partition  $(V, d)$

1. choose a random permutation  $\pi$  of  $v_1, v_2, \dots, v_n$
2. choose  $\beta$  in  $[1, 2]$  randomly from the distribution  $p(x) = \frac{1}{x \ln 2}$   
(instead we could choose  $\beta$  uniformly at random from  $[1, 2]$  with slightly worse bounds)
3.  $D_0 \leftarrow \{V\}$   $j \leftarrow 0$
4. while  $D_{i+1}$  has non-singleton clusters do
  - 4.1  $\beta_i \leftarrow 2^{i-1} \beta$
  - 4.2 For  $L=1, 2, \dots, n$  do
    - 4.2.1 for every cluster  $S$  in  $D_{i+1}$ 
      - 4.2.1.1 create a new cluster consisting of all unassigned vertices in  $S$  closer than  $\beta_i$  to  $\pi(L)$
  - 4.3.  $i \leftarrow i+1$



Illustrating the construction to  $\pi(L)$

Note that our randomized algorithm above obtains the laminar decomposition, which will implicitly define the distribution of trees that are used in the embedding.

Note that by the definition of our distribution  
 Observation: For any  $x \geq 1$ ,  $PR[\text{some } \beta_i \text{ lies in } [x, x+dx]] \leq \frac{1}{x \ln 2} dx = \frac{1}{x \ln 2} \frac{dx}{2^i} \rightarrow$  scale  $\beta_i$  chooses the desired value

A place to network and exchange ideas.

Fix an arbitrary edge  $(u, v)$ , we show the expected value of  $R_T(u, v)$  is in  $O(\log n) d(u, v)$ .

- Consider level  $i$  of clustering. In each iteration, all unassigned vertices  $w$  such that  $d(v, R(w)) \leq \beta_i$  assigns themselves to  $R(w)$ .

- First  $u$  and  $v$  are both assigned.  
Then at some step  $l$   $u$  is assigned to  $w$  ( $v$  may or may not) for the first time (we say  $w$  settles  $(u, v)$ ).

We say  $w$  cuts  $(u, v)$  if  $w$  settles  $(u, v)$  and exactly one is assigned to  $w$ .

- When ever a vertex  $w$  cuts edge  $(u, v)$  at level  $i$ , the tree length of the edge  $(u, v)$  is about  $2^{i+2}$ , as we discuss: we ~~change~~ <sup>blame w for</sup> this length.

- Let  $d_w^T(u, v) = \sum_{i=1}^T \mathbb{1}(w \text{ cuts } (u, v) \text{ at level } i) \cdot 2^{i+2}$  ( $\mathbb{1}(\cdot)$  is the indicator function)

Thus  $d^T(u, v) \leq \sum_w d_w^T(u, v)$ .

- Sort the vertices of  $V$  in order of increasing distance from edge  $(u, v)$  (from either end and breaking ties arbitrarily). Let  $u_s$  be the  $s$ th vertex and say  $d(u_s, u) \leq d(u_s, v)$ .

→ We compute  $d_{u_s}^T(u, v)$ .

$u_s$  cuts  $(u, v)$  if  $\begin{cases} d_{u_s}(u, v) \leq \beta_i < d(u_s, v) \text{ for some } i \\ \text{and if it happens the } w \text{ settles } (u, v) \text{ at level } i. \end{cases}$

Contribution to  $d_{u_s}^T(u, v)$  is at most  $2^{i+2} \leq 8\beta_i$ .

- The probability that some  $\beta_i$  falls in  $(x, x+\Delta x)$  for a particular  $x \in [d(u_s, u), d(u_s, v)]$  is at most  $\frac{1}{x \ln 2} \Delta x$  according to observation 1.

- Conditioned on  $\beta_i$  taking this value  $x$ , one of  $u_1, u_2$ , —  $u_s$  can settle  $(u, v)$  at level  $i$  (trivial) then the probability is at most  $\frac{1}{5}$ . Thus the expected cost of  $d_{u_s}^T(u, v)$  is at most  $\int_{d(u_s, u)}^{d(u_s, v)} \frac{1}{x \ln 2} \cdot 8x \cdot \frac{1}{5} dx = \frac{8}{5 \ln 2} (d(u_s, v) - d(u_s, u)) \leq \frac{8}{5 \ln 2} d(u, v)$

- Thus by linearity of Expectation  $E[d^T(u, v)] \leq \sum_s \frac{8}{5 \ln 2} d(u, v) = \frac{8}{5 \ln 2} \ln n d(u, v) = 8 \log n d(u, v)$ .  
(it was uniform in  $[1, 2]$  then we had  $\frac{d(u_s, v) - d(u_s, u)}{5 \cdot 2^i} \leq \frac{8 d(u, v)}{5 \cdot 2^i}$  by Triangle Ineq.)

★ (can be derived by the method of conditional expectation and give  $O(\log n)$  trees.)  
★ The steiner points of trees can be removed by an extra factor 8 in cost (by Gupta's Lemma)

A place to network and exchange ideas.

- Uniform buy-at-bulk network design from Averbuch - Alzav 2007 = A Graph  $G(X, E)$  where  $|X| = n$
- we are given pairs  $(s_1, t_1), (s_2, t_2), \dots, (s_k, t_k)$  where the demand  $s_i, t_i$  is dem <sub>$i$</sub>
  - Each edge  $e$  has length  $l_e$  - we want to send all the traffic from  $u$  to  $v$  on a single path  $P_{uv}$  (i.e. it is indivisible) - w.l.o.g. assume  $l_e$  is a metric.
  - The edge of the network must be installed by purchasing zero or more copies from the set of cables where each cable type  $i$  has specified capacity  $u_i$  and cost  $c_i$ .
  - No assumption about  $c_i, u_i$ , but usually if  $u_i < u_j$  then  $c_i < c_j$  (otherwise get rid of  $i$ ) and  $\frac{c_j}{u_j} < \frac{c_i}{u_i}$  (economy of scale). These are not necessary for our algorithms:

Algorithm: first we decide about the route  $P_{s_i t_i}$ . Then we buy enough capacity. Let  $C(\text{dem})$  be the minimum cost to cover a total demand of dem for unit distance.

lm1:  $\min_i c_i \lceil \frac{\text{dem}}{u_i} \rceil$  is a 2 approximation for  $C(\text{dem})$ . (has PTAS by Knapsack)

pf: By an averaging argument we need at least  $\min_{u_i < \text{dem}} c_i \frac{\text{dem}}{u_i}$ . we have two approximation since  $\lceil \frac{\text{dem}}{u_i} \rceil < \frac{\text{dem}}{u_i} + 1 \leq 2 \frac{\text{dem}}{u_i}$

lm2:  $C(\text{dem})$  is subadditive, i.e.  $C(x+y) \leq C(x) + C(y)$ .

pf: by definition

Our Algorithm: Use FRT, route in a tree with min total cost, map (smear) the solution back into the graph and install enough capacity (use 2-approximation). (note that during this mapping the total cost can only decrease with respect to the tree cost)  $C$ , due to shorter distances or subadditivity.

So we only need to prove

Thm: Embedded OPT on a set of trees has expected cost at most  $OPT_G$ .

Suppose opt routes  $u, v$  via path  $Q_{uv}$  in  $G$ . then for each edge  $e$  in  $G$  let  $f_e$  be the flow that passes through  $e$  in  $G$ . Thus

$$f_e = \sum_{(u,v) | e \in Q_{uv}} \text{dem}_{uv}$$

and thus  $opt_G = \sum_{e \in E(G)} l(e) C(f_e)$ .

consider an edge  $e = (x, y)$  in  $G$ . We associate with this edge in each tree  $T \in \text{Support}(D)$  a path  $P_T$  between  $x$  and  $y$  of length  $d_T(x, y) = l(e) \alpha_T$  where  $E(G) = O(\log n)$  by FRT. Note that want to send a flow  $f_e$  through this path  $P_T$ .

By linearity of expectation we are focusing only on one tree  $T$ .

The cost of designing a path  $P_T$  with a flow of  $f_e$  on each edge  $e' \in P_T$  satisfies  $\sum_{e' \in P_T} l(e') C(f_{e'}) = d_T(x, y) C(f) = l(e) \alpha_T C(f)$ .

Now, if we design a network in  $T$  for all paths  $P_T$  for all  $e \in E(G)$  each with demand  $f_e$  then the flow  $f_{e'}$  on each edge  $e' \in E(T)$  is  $f_{e'} = \sum_{e \in E(G) | e' \in P_T} f_e$

and its cost due to subadditivity of  $C$  is

$$\begin{aligned} \sum_{e' \in E(T)} l(e') C(f_{e'}) &= \sum_{e' \in E(T)} l(e') C\left(\sum_{e \in E(G) | e' \in P_T} f_e\right) \leq \sum_{e' \in E(T)} l(e') \sum_{e \in E(G) | e' \in P_T} C(f_e) \\ &= \sum_{e' \in E(T)} \sum_{e \in E(G) | e' \in P_T} l(e') C(f_e) = \sum_{e \in E(G)} \sum_{e' \in P_T} l(e') C(f_e) = \sum_{e \in E(G)} l(e) \alpha_T C(f_e) = \alpha_T \sum_{e \in E(G)} l(e) C(f_e) = \alpha_T opt_G \end{aligned}$$

Note that a path  $Q_{uv}$  in  $G$  is associated with a (maybe non-simple) path  $Q_{T, (u, v)}$  in  $T$  which consists of concatenation of paths in  $T$  associated with the edges of  $Q_{uv}$  and

$$\begin{aligned} f_{e'} &= \sum_{e \in E(G) | e' \in P_T} f_e = \sum_{e \in E(G) | e' \in P_T} \sum_{(u, v) | e \in Q_{uv}} \text{dem}_{u, v} \\ &= \sum_{(u, v) | e' \in P_T, e \in Q_{uv}} \text{dem}_{u, v} = \sum_{(u, v) | e' \in Q_{T, (u, v)}} \text{dem}_{u, v} \end{aligned}$$

which implies that the network that we designed for each  $T$  is a feasible solution for demand.

In next sessions, we show how we can obtain polylog approximation for non-uniform bag-at-bulk. If the  $C$  and  $l$  of different edges are different.