

In computer science, when dealing with difficult problems involving graphs and their associated metrics, one technique we usually resort to is to solve the problem on a simple subcase, in particular, on tree. In this lecture, we study the Padded Decomposition of graphs and how to build their Tree metrics with $O(\log n)$ distortion.

1 Tree Metrics

Definition 1 Tree Metric - A metric (V, d) is a Tree Metric if there exists a tree $T = (V, E_T)$ with edge lengths such that $d(u, v) = d_T(u, v)$.

We can build tree metrics to approximate general graph metrics, with an $O(\log n)$ distortion, where n is the number of vertices in the graph, thus the following theorem:

Theorem 1 Given an undirected edge weighted graph $G = (V, E)$, there is a randomized poly-time algorithm that produces an random edge weighted tree $T = (V_T, E_T)$ such that

1. $V \subseteq V_T$;
2. $d_G(u, v) \leq d_T(u, v), \forall u, v$;
3. $E(d_T(u, v)) \leq O(\log |V|)d_G(u, v), \forall u, v$.

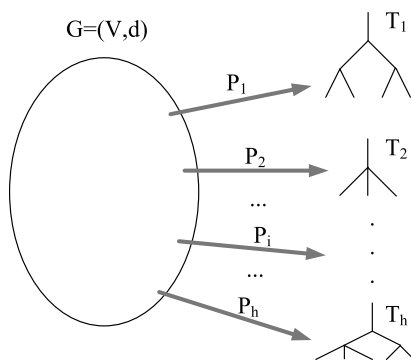


Figure 1: Tree Embedding.

The Problem that Theorem 1 indicates is as follows (refer to Figure 1 for illustration):

Problem 1 Given a graph metric $G = (V, d)$, find a collection of trees $\mathcal{T} = \{T_1, T_2, \dots, T_h\}$ and a distribution $\mu = \{P_1, P_2, \dots, P_h\}$ on them such that

1. $\sum_{i=1}^h P_i = 1$;
2. $d_{T_i}(u, v) \geq d_G(u, v), \forall u, v, \forall i$;
3. $\sum_{i=1}^h P_i d_{T_i}(u, v) \leq O(\log n) d_G(u, v), \forall u, v$.

A tree embedding algorithm is discussed in the next section to show the correctness of Theorem 1.

2 Padded Decomposition and Tree Embeddings

The following algorithm is developed by Bartal in [1] to solve the tree embedding problem. Note that $\Delta(G)$ denotes the diameter of graph G .

TREE EMBEDDING ALGORITHM($G(V, d)$):
 Start with G .
 Randomly decompose G into G_1, G_2, \dots, G_h s.t.

- $\Delta(G) \leq \frac{\Delta(G)}{2}, \forall 1 \leq i \leq h$;
- $Pr[(u, v) \text{ is cut}] \leq \frac{\alpha \cdot d_G(u, v)}{\Delta(G)}$.

Recursively construct “rooted” trees T_1, T_2, \dots, T_h for G_1, G_2, \dots, G_h , respectively.
 Create tree T for G by (see Figure 2)

- adding root r ;
- connecting T_1, T_2, \dots, T_h to r with edge weighted $\Delta(G)$.

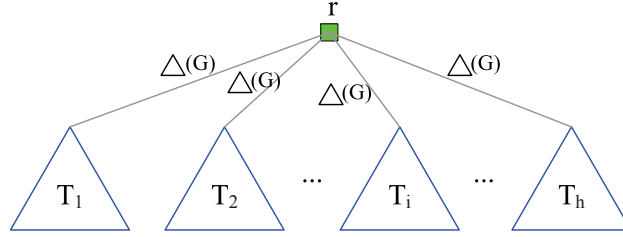


Figure 2: Create tree T from T_1, T_2, \dots, T_h .

The above TREE EMBEDDING ALGORITHM is a high-level algorithm, we first suppose there exists a graph decomposition procedure to randomly decompose G into G_1, G_2, \dots, G_h as in the second step, we then have the following two Propositions that can be easily observed.

Proposition 2 T dominates G .

Proposition 3 $\Delta(T) \leq 4\Delta(G)$.

We subsequently have the following lemma:

Lemma 4 $E(d_T(u, v)) = O(\alpha \log \Delta(G))d_G(u, v)$.

Proof of Lemma 4. Let A_i denote the event that (u, v) is cut in level i , and let \bar{A}_i denote the event that (u, v) is not cut in level i .

$$\begin{aligned}
 E[d_T(u, v)] &\leq 4 \cdot \Delta(G) \cdot Pr[A_1] \\
 &\quad + 4 \cdot \frac{\Delta(G)}{2} \cdot Pr[A_2 | \bar{A}_1] \\
 &\quad + \dots + 4 \cdot \frac{\Delta(G)}{2^i} \cdot Pr[A_i | \bigcap_{j=1}^{i-1} \bar{A}_j] + \dots \\
 &= 4 \cdot \Delta(G) \cdot \frac{\alpha \cdot d_G(u, v)}{\Delta(G)} \\
 &\quad + 4 \cdot \frac{\Delta(G)}{2} \cdot \frac{\alpha \cdot d_G(u, v)}{\Delta(G)/2} \\
 &\quad + \dots + 4 \cdot \frac{\Delta(G)}{2^i} \cdot \frac{\alpha \cdot d_G(u, v)}{\Delta(G)/2^i} + \dots \\
 &\leq 4\alpha \log \Delta(G) \cdot d_G(u, v) \\
 &= O(\alpha \log \Delta(G)) \cdot d_G(u, v)
 \end{aligned}$$

□

Note that now $E[d_T(u, v)]$ is dependent on $\log \Delta(G)$, and this can actually be replaced with $\log n$ if the CKR GRAPH DECOMPOSITION procedure is incorporated into the TREE EMBEDDING ALGORITHM, so that correctness of Theorem 1 can be proved. The CKR GRAPH DECOMPOSITION procedure is introduced in next subsection.

2.1 CKR Graph Decomposition Procedure

CKR GRAPH DECOMPOSITION($G(V, d), \delta$):
 Pick θ at random from $[\frac{\delta}{4}, \frac{\delta}{2}]$.
 Pick a random permutation σ on V .
 For $i = 1$ to n do
 $G_i = B(V_{\sigma(i)}, \theta) \setminus \bigcup_{j < i} B(V_{\sigma(j)}, \theta)$

Note that $B(u, \theta)$ denotes the ball that is centered at u and with radius as θ , and we say $B(u, \theta)$ is cut if any vertex in $B(u, \theta)$ is separated from u .

Theorem 5 *CKR decomposition decomposes G into G_1, G_2, \dots, G_n such that:*

1. $\Delta(G_i) \leq \delta$;
2. $\forall u, Pr[B(u, \varrho) \text{ is cut}] \leq \frac{c\varrho}{\delta} \log n$, where c is a constant.

It is easy to see the following is an immediate Corollary of Theorem 5:

Corollary 6 $\forall u, v, Pr[(u, v) \text{ is cut}] \leq \frac{c \cdot d_G(u, v)}{\delta} \log n$.

Proof of Theorem 5. We fix vertex u , and assume w.l.o.g. $\varrho < \frac{\delta}{4}$ ($\frac{c\varrho}{\delta} > 1$ if $\varrho \geq \frac{\delta}{4}$). Let v_1, v_2, \dots, v_n be an ordering of V s.t. $d(v_1, u) \leq d(v_2, u) \leq \dots \leq d(v_n, u)$. Let A_i be the event that v_i “first” cuts $B(u, \varrho)$, i.e.,

1. $B(v_i, \theta) \cap B(u, \varrho) \neq \emptyset$
2. $B(v_i, \theta) \cap B(u, \varrho) \neq B(u, \varrho)$
3. $\forall j < \arg \sigma(i), v_{\sigma(j)}$ does not cut or captures $B(u, \varrho)$.

As shown in Figure 3, obviously A_i is possible only when $\theta = a \in [d(v_i, u) - \varrho, d(v_i, u) + \varrho]$, and note that σ is a random permutation, so

$$Pr[A_i | \theta = a] \leq \frac{1}{i}, \forall a \in [d(v_i, u) - \varrho, d(v_i, u) + \varrho]$$

recall that $\theta \in [\frac{\delta}{4}, \frac{\delta}{2}]$, thus

$$Pr[A_i] \leq \frac{1}{i} \cdot \frac{2\varrho}{\delta/4} \leq \frac{8}{i} \cdot \frac{\varrho}{\delta}$$

Therefore, $Pr[B(u, \varrho) \text{ is cut}] \leq \sum_{i=1}^n Pr[A_i] \leq \frac{8\varrho}{\delta} \log n$. □

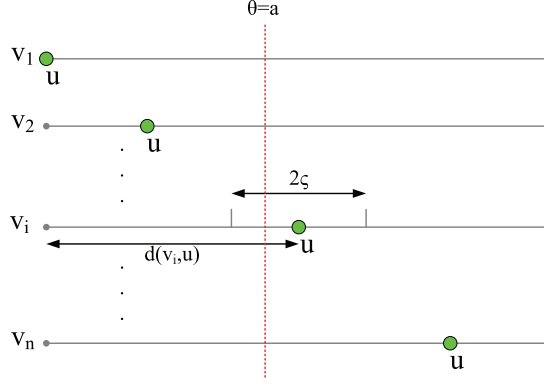


Figure 3: Proof of Theorem 5.

Actually we are able to provide an even tighter bound for $Pr[B(u, \varrho) \text{ is cut}]$. By observation, v_i may cut $B(u, \varrho)$ only when $d(v_i, u) \in [\frac{\delta}{4} - \varrho, \frac{\delta}{2} + \varrho]$, therefore,

$$\begin{aligned}
 Pr[B(u, \varrho) \text{ is cut}] &\leq \sum_{i: d(v_i, u) \in [\frac{\delta}{4} - \varrho, \frac{\delta}{2} + \varrho]} \frac{1}{i} \cdot \frac{8\varrho}{\delta} \\
 &= \frac{8\varrho}{\delta} \cdot \log \frac{|B(u, \frac{\delta}{2} + \varrho)|}{|B(u, \frac{\delta}{4} - \varrho)|} \\
 &\leq \frac{8\varrho}{\delta} \cdot \log \frac{|B(u, \delta)|}{|B(u, \delta/8)|}
 \end{aligned}$$

2.2 Incorporate CKR decomposition into Tree Embedding Algorithm

By applying CKR GRAPH DECOMPOSITION procedure to decompose G into G_1, G_2, \dots, G_h , at the second step in the TREE EMBEDDING ALGORITHM, we are able to improve the bound $E(d_T(u, v)) = O(\log \Delta(G))d_G(u, v)$ in Lemma 4 to $E(d_T(u, v)) = O(\log n)d_G(u, v)$:

$$\begin{aligned}
 E[d_T(u, v)] &\leq 4 \cdot \Delta(G) \cdot Pr[A_1] + 4 \cdot \frac{\Delta(G)}{2} \cdot Pr[A_2 | \bar{A}_1] \\
 &\quad + \dots + 4 \cdot \frac{\Delta(G)}{2^i} \cdot Pr[A_i | \bigcap_{j=1}^{i-1} \bar{A}_j] + \dots \\
 &\leq 4 \cdot \Delta(G) \cdot \frac{8d_G(u, v)}{\Delta(G)/2} \cdot \log \frac{|B(u, \Delta(G)/2)|}{|B(u, \Delta(G)/16)|} \\
 &\quad + 4 \cdot \frac{\Delta(G)}{2} \cdot \frac{8d_G(u, v)}{\Delta(G)/4} \cdot \log \frac{|B(u, \Delta(G)/4)|}{|B(u, \Delta(G)/32)|} \\
 &\quad + \dots \\
 &\quad + 4 \cdot \frac{\Delta(G)}{2^i} \cdot \frac{8d_G(u, v)}{\Delta(G)/2^{i+1}} \cdot \log \frac{|B(u, \Delta(G)/2^{i+1})|}{|B(u, \Delta(G)/2^{i+4})|} \\
 &\quad + \dots \\
 &\leq 64 \cdot d_G(u, v) \cdot 3 \log n \\
 &= O(\log n) \cdot d_G(u, v)
 \end{aligned}$$

Up to now, we have shown the correctness of the sc Tree Embedding Algorithm, thus Theorem 1 has been finally proved.

References

- [1] Yair Bartal. Probabilistic approximations of metric spaces and its algorithmic applications. *IEEE Symposium on Foundations of Computer Science*, pages 184-193, 1996.

- [2] Jittat Fakcharoenphol, Satish Rao and Kunal Talwar. Approximating metrics by tree metrics. *ACM SIGACT News*, Volume 35, Issue 2, pages 60-70, June 2004.