# Ramsey Spanning Trees and their Applications

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#### Abstract

The metric Ramsey problem asks for the largest subset S of a metric space that can be embedded into an ultrametric (more generally into a Hilbert space) with a given distortion. Study of this problem was motivated as a non-linear version of Dvoretzky theorem. Mendel and Naor [MN07] devised the so called Ramsey Partitions to address this problem, and showed the algorithmic applications of their techniques to approximate distance oracles and ranking problems.

In this paper we study the natural extension of the metric Ramsey problem to graphs, and introduce the notion of Ramsey Spanning Trees. We ask for the largest subset  $S \subseteq V$  of a given graph G = (V, E), such that there exists a spanning tree of G that has small stretch for S. Applied iteratively, this provides a small collection of spanning trees, such that each vertex has a tree providing low stretch paths to all other vertices. The union of these trees serves as a special type of spanner, a tree-padding spanner. We use this spanner to devise the first compact stateless routing scheme with O(1) routing decision time, and labels which are much shorter than in all currently existing schemes.

We first revisit the metric Ramsey problem, and provide a new deterministic construction. We prove that for every k, any n-point metric space has a subset S of size at least  $n^{1-1/k}$  which embeds into an ultrametric with distortion 8k. We use this result to obtain the state-of-the-art deterministic construction of a distance oracle. Building on this result, we prove that for every k, any n-vertex graph G = (V, E) has a subset S of size at least  $n^{1-1/k}$ , and a spanning tree of G, that has stretch  $O(k \log \log n)$  between any point in S and any point in V.

## 1 Introduction

Inspired by the algorithmic success of Ramsey Type Theorems for metric spaces, in this paper we study an analogue Ramsey Type Theorem in a graph setting. The classical Ramsey problem for metric spaces was introduced in [BFM86], and is concerned with finding "nice" structures in arbitrary metric spaces. Following [BLMN03], [MN07] showed that every n-point metric (X,d) has a subset  $M \subseteq X$  of size at least  $n^{1-1/k}$  that embeds into an ultrametric (and thus also into Hilbert space) with distortion at most O(k), for a parameter  $k \ge 1$ . In fact, they construct an ultrametric on X which has O(k) distortion for any pair in  $M \times X$ . Additionally, [MN07] demonstrated the applicability of their techniques, which they denoted Ramsey Partitions, to approximate distance oracles and ranking problems.

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We introduce a new notion that we call Ramsey Spanning Trees, which is a natural extension of the metric Ramsey problem to graphs. In this problem, we wish to find a large subset  $S \subseteq V$ , and a spanning tree of G that has small distortion for  $S \times V$ . Let  $\mathbf{dist}(u, v, G)$  denote the shortest path distance in the graph G between the vertices  $u, v \in V$ , then our main result is the following.

**Theorem 1.** Let G = (V, E) be a weighted graph on n vertices, and fix a parameter  $k \geq 1$ . There is a polynomial time deterministic algorithm that finds a spanning tree T of G and a subset  $S \subseteq V$  of vertices of size at least  $n^{1-1/k}$ , such that for every  $v \in S$  and  $u \in V$  it holds that  $\operatorname{dist}(v, u, T) \leq O(k \log \log n) \cdot \operatorname{dist}(v, u, G)$ .

We remark that the extra factor of  $\log \log n$  in the stretch comes from the state-of-the-art result of  $O(\log n \log \log n)$  for low stretch spanning trees [AN12]. It is quite plausible that if that result is improved to the optimal  $O(\log n)$ , then the stretch in our result would be only O(k). By applying Theorem 1 iteratively, we can obtain a small collection of trees so that each vertex has small stretch to all other vertices in at least one of the trees.

**Theorem 2.** Let G = (V, E) be a weighted graph on n vertices, and fix a parameter  $k \ge 1$ . There is a polynomial time deterministic algorithm that finds a collection  $\mathcal{T}$  of  $k \cdot n^{1/k}$  spanning trees of G, and a mapping  $\mathbf{home} : V \to \mathcal{T}$ , such that for every  $u, v \in V$  it holds that  $\mathbf{dist}(v, u, \mathbf{home}(v)) \le O(k \log \log n) \cdot \mathbf{dist}(v, u, G)$ .

A spanner H with stretch t for a graph G, is a sparse spanning subgraph satisfying  $\mathbf{dist}(v, u, H) \leq t \cdot \mathbf{dist}(v, u, G)$ . Spanners are a fundamental metric and graph-theoretic constructions; they are very well-studied [PS89, ADDJ90, Coh93, EP04, BS03, TZ06, AB16], and have numerous applications [Awe84, ABCP93, Coh93, Elk01, GRTU16]. Theorem 2 can be viewed as providing a spanner which is the union of  $k \cdot n^{1/k}$  spanning trees, such that every vertex has a tree with low stretch paths to all other vertices. We call such a spanner a tree-padding spanner of order  $k \cdot n^{1/k}$ . To the best of our knowledge, no previous construction of spanners can be viewed as a tree-padding spanner of order o(n). Until now even the following weaker question was open: does there exist a spanner which is a union of a sublinear in n number of trees, such that every pair of vertices has a low stretch path in one of these trees.

Having a single tree that provides good stretch for any pair containing the vertex v, suggests that routing messages to or from v could be done on this one tree. Our main application of Ramsey spanning trees is a compact routing scheme that has constant routing decision time and improved label size, see Section 1.1 for more details.

**Deterministic Ramsey Partitions.** As a first step towards our main result, which is of interest in its own right, we provide a new deterministic Ramsey ultrametric construction. In particular, we show a polynomial time deterministic algorithm, that given an n-point metric space (X, d) and a parameter  $k \geq 1$ , finds a set  $M \subseteq X$  of size at least  $n^{1-1/k}$  and an ultrametric  $(M, \rho)$  with distortion at most 8k - 2. That is, for each  $v, u \in M$ ,

$$d(v, u) < \rho(v, u) < (8k - 2) \cdot d(v, u)$$
.

The first result of this flavor was by Bartal et al. [BBM06], who showed a deterministic construction with distortion  $O(k \log \log n)$ . This was improved by [BLMN03] to distortion  $O(k \log k)$ . Mendel and Naor [MN07] developed the so called Ramsey partitions, which had distortion of 128k (using a randomized algorithm). Blelloch et. al [BGS16] showed that the (randomized) algorithm of

[FRT04] constructs an ultrametric with distortion 18.5k (they also provided a near-linear time implementation of it). The best randomized algorithm is by Naor and Tao [NT12], who obtained distortion 2ek. Their techniques are not based on Ramsey partitions, as is ours and other previous works (see Section 3 for more details on Ramsey partitions). In fact, [NT12] declared that a Ramsey partition with distortion better than 16k-2 seems not to be possible with their current techniques. Moreover, [MN07] mention as a drawback that their solution is randomized (while [BLMN03] is deterministic). A deterministic construction similar to ours was obtained by Bartal [Bar11].

An application of our improved deterministic Ramsey ultrametric construction is a new distance oracle that has the best space-stretch-query time tradeoff among deterministic distance oracles. See Section 1.1 below.

**Techniques.** Our construction of Ramsey ultrametrics uses the by-now-standard deterministic ball growing approach, e.g. [Awe84, AKPW95, AP90, FRT04, Bar04]. In this paper we provide tighter and more parameterized analysis of these multi-scale deterministic region growing techniques. Our improved analysis of the deterministic ball growing technique of [FRT04, Bar04] obtains a similar type of improvement as the one obtained by the analysis of Mendel and Naor [MN07] on the randomized partition technique of [CKR04, FRT03].

Our construction of Ramsey spanning trees is based on combining ideas from our Ramsey ultrametric construction, with the Petal Decomposition framework of [AN12]. The optimal multiscale partitions of [FRT04, Bar04] cannot be used in this petal decomposition framework, so we must revert to partitions based on [Sey95, EEST05], which induce an additional factor of  $O(\log \log n)$  to the stretch. In addition, the refined properties required by the Ramsey partition make it very sensitive to constant factors (these constants can be ignored in the [EEST05] analysis of the average stretch, say). In order to alleviate this issue, we consider two possible region growing schemes, and choose between them according to the densities of points that can still be included in M. One of these schemes is a standard one, while the other grows the region "backwards", in a sense that it charges the remaining graph, rather than the cluster being created, for the cost of making a cut. See Section 4.3 for more details.

### 1.1 Applications

**Distance Oracles.** A distance oracle is a succinct data structure that (approximately) answers distance queries. A landmark result of [TZ01a] states that any metric (or graph) with n points has a distance oracle of size  $O(k \cdot n^{1+1/k})$ , that can report any distance in O(k) time with stretch at most 2k-1. A deterministic variant with the same parameters was given by [RTZ05], and this was the state-of-the-art for deterministic constructions. The oracle of [MN07] has improved size  $O(n^{1+1/k})$  and O(1) query time, but larger stretch 128k. This oracle was key for subsequent improvements by [WN13, Che14, Che15], the latter gave a randomized construction of an oracle with size  $O(n^{1+1/k})$ , query time O(1) and stretch 2k-1 (which is asymptotically optimal assuming Erdos' girth conjecture).

Similarly to [MN07], our deterministic construction of Ramsey ultrametrics can provide a deterministic construction of an approximate distance oracle. While the stretch of the oracle of [MN07] is further increased by a large constant factor over the stretch of the Ramsey ultrametric, we use a

<sup>&</sup>lt;sup>1</sup>We measure size in machine words, each word is  $\Theta(\log n)$  bits.

careful analysis and a collection of oracles in various distance scales, in order to increase the stretch by only a  $1 + \epsilon$  factor.

**Theorem 3.** For any metric space on n points, and any k > 1,  $0 < \epsilon < 1$ , there is an efficient deterministic construction of a distance oracle of size  $O(n^{1+1/k})$ , that has stretch  $8(1+\epsilon)k$  and query time  $O(1/\epsilon)$ .

This is the first deterministic construction of an approximate distance oracle with constant query time and small size  $O(n^{1+1/k})$ .

Moreover, our oracle is an essential ingredient towards de-randomizing the recent distance oracles improvements [WN13, Che14, Che15]. Specifically, if we construct [Che14] by replacing the distance oracle of Mendel and Naor [MN07] by our deterministic version, and replacing the distance oracle of Thorup and Zwick [TZ01a] by the deterministic version of Roditty, Thorup, and Zwick [RTZ05], we immediately get a deterministic distance oracle of  $O(k \cdot n^{1+1/k})$  size, 2k-1 stretch and O(1) query time. This is a strict improvement over [RTZ05]. In addition, our oracle can be viewed as a first step towards de-randomizing the [Che15] oracle. A summary of all the previous and current results in a table form can be found at Table 1.

Stateless Routing with Short Labels and Constant Decision Time. A routing scheme in a network is a mechanism that allows packets to be delivered from any node to any other node. The network is represented as a weighted undirected graph, and each node can forward incoming data by using local information stored at the node, often called a routing table, and the (short) packet's header. The routing scheme has two main phases: in the preprocessing phase, each node is assigned a routing table and a short label. In the routing phase, each node receiving a packet should make a local decision, based on its own routing table and the packet's header (which may contain the label of the destination, or a part of it), where to send the packet. The routing decision time is the time required for a node to make this local decision. The stretch of a routing scheme is the worst ratio between the length of a path on which a packet is routed, to the shortest possible path. A routing scheme is called stateless if the routing decision does not depend on the path traversed so far.

The classical routing scheme of [TZ01b], for a graph on n vertices and integer parameters k, b > 1, provides a scheme with routing tables of size  $O(k \cdot b \cdot n^{1/k})$ , labels of size  $(1 + o(1))k \log_b n$ , stretch 4k - 5, and decision time O(1) (but the initial decision time is O(k)). The stretch was improved recently to roughly 3.68k by [Che13], using a similar scheme as [TZ01b]. With Theorem 2, we devise a stateless compact routing scheme with very short labels, of size only  $(1 + o(1)) \log_b n$ , and with constant decision time, while the stretch increases to  $O(k \log \log n)$  (and with the same table size as [TZ01b]).

We wish to point out that our construction of a routing scheme is simpler in some sense that those of [TZ01b, Che13]. In both constructions there is a collection of trees built in the preprocessing phase, such that every pair of vertices has a tree that guarantees small stretch. Routing is then done in that tree. In our construction there are few trees, so every vertex can store information about all of them, and in addition, every vertex  $v \in V$  knows its home tree, and routing towards v from any other vertex on the tree  $\mathbf{home}(v)$  has small stretch. In particular, the header in our construction consists of only the label of the destination. In the [TZ01b] scheme, however, there are n trees, and a certain process is used to find the appropriate tree to route on, which increases the initial decision time, and also some information must be added to the header of the message

after the tree is found. Finally, our routing scheme is stateless, as opposed to [TZ01b]. (We remark that using ideas from [Che14], one can devise a stateless routing scheme based on [TZ01b], but this scheme seems to suffer from larger header and decision time at each node.)

**Theorem 4.** Given a weighted graph G = (V, E) on n vertices and integer parameters k, b > 1, there is a stateless routing scheme with stretch  $O(k \log \log n)$  that has routing tables of size  $O(k \cdot b \cdot n^{1/k})$  and labels of size  $(1 + o(1)) \log_b n$ . The decision time in each vertex is O(1).

Observe that choosing parameters 2k and  $b = n^{1/(2k)}$  for Theorem 4 yields a routing scheme with stretch  $O(k \log \log n)$  that has tables of size  $O(k \cdot n^{1/k})$  and labels of size only O(k). Another interesting choice of parameters is b = 2 and  $k = \frac{100 \log n}{\log \log n}$ , this provides a scheme with stretch  $O(\log n)$  that has tables of size  $O(\log^{1.01} n)$  and labels of size  $O(\log n)$ . Compare this to the [TZ01b] scheme, which for stretch  $O(\log n)$  has tables of size  $O(\log n)$  and labels of size  $O(\log^2 n)$ .

## 1.2 Organization

In Section 3 we present our deterministic Ramsey partitions, that are used for Ramsey ultrametrics and distance oracles. In Section 4 we show the Ramsey spanning trees, and the application to routing. Each section can be read independently.

## 2 Preliminaries

Let G = (V, E) be a weighted undirected graph. We assume that the minimal weight of an edge is 1. For any  $Y \subseteq V$  and  $x, y \in Y$ , denote by  $\mathbf{dist}(x, y, Y)$  the shortest path distance in G[Y] (the graph induced on Y). For  $v \in Y$  and  $r \geq 0$  let  $B(v, r, Y) = \{u \in Y \mid \mathbf{dist}(v, u, Y) \leq r\}$ , when Y = V we simply write B(v, r). We may sometimes abuse notation and not distinguish between a set of vertices and the graph induced by them.

An ultrametric (Z, d) is a metric space satisfying a strong form of the triangle inequality, that is, for all  $x, y, z \in Z$ ,  $d(x, z) \le \max\{d(x, y), d(y, z)\}$ . The following definition is known to be an equivalent one (see [BLMN05]).

**Definition 1.** An ultrametric is a metric space (Z,d) whose elements are the leaves of a rooted labeled tree T. Each  $z \in T$  is associated with a label  $\ell(z) \geq 0$  such that if  $q \in T$  is a descendant of z then  $\ell(q) \leq \ell(z)$  and  $\ell(q) = 0$  iff q is a leaf. The distance between leaves  $z, q \in Z$  is defined as  $d_T(z,q) = \ell(lca(z,q))$  where lca(z,q) is the least common ancestor of z and q in T.

# 3 Ramsey Partitions

Consider an undirected weighted graph G = (V, E), and a parameter  $k \ge 1$ . Let D be the diameter of the graph and let  $\phi = \lceil \log (D+1) \rceil$ . Let  $\rho_i = 2^i/(4k)$ . We start by presenting a construction for a collection S of partial partitions  $\mathcal{X}_i$  satisfying the following key properties.

For a set X, a partial-partition is a set of nonempty subsets of X such that every element  $x \in X$  is in at most one of these subsets. For two partial partitions  $\mathcal{P}_1$  and  $\mathcal{P}_2$ , we say that  $\mathcal{P}_1$  is a refinement of  $\mathcal{P}_2$  if for every  $X_1 \in \mathcal{P}_1$  there is  $X_2 \in \mathcal{P}_2$  such that  $X_1 \subseteq X_2$ .

**Definition 2.** [(G,U,k)-Fully Padded Strong Diameter Hierarchical Partial Partition] Given a graph G = (V,E), an index k and a set of nodes  $U \subseteq V$ , a (G,U,k)-Fully Padded Strong Diameter

Hierarchical Partial Partition (**FPSDHPP**) is a collection  $\{\mathcal{X}_i\}_{i=0}^{\phi}$  of partial partitions of V where each  $X \in \mathcal{X}_i$  has a center  $r(X) \in V$ , such that the following properties hold.

- (i) For every  $0 \le i < \phi$ ,  $\mathcal{X}_i$  is a refinement of  $\mathcal{X}_{i+1}$ .
- (ii) For every  $0 \le i \le \phi$  and every  $X \in \mathcal{X}_i$  and every  $v \in X$ ,  $\mathbf{dist}(v, r(X), X) < 2^i$ .
- (iii) There exists a set  $\hat{V} \subseteq U$  such that  $|\hat{V}| \ge |U|^{1-1/k}$  and for every  $v \in \hat{V}$  and every i, there exists a subset  $X \in \mathcal{X}_i$  such that  $B(v, \rho_i) \subseteq X$ .

For a node v and index i, we say that v is i-padded in S, if there exists a subset  $X \in \mathcal{X}_i$  such that  $B(v, \rho_i) \subseteq X$ . We would like to maximize the number of nodes that are padded on all levels. Note that we do not get to choose the set  $\hat{V}$  of padded vertices. Nevertheless, Definition 2 allows us to specify a subset  $U \subseteq V$  such that the padded vertices will be chosen from U. In such a case, a significant portion of the vertices in U will indeed be padded.

Fully Padded Strong Diameter Hierarchical Partial Partition Construction: Let us now turn to the construction of the collection S of partial partitions  $X_i$  given a set U.

In the beginning of the algorithm, all nodes in U are set as marked. The algorithm iteratively unmarks some of the nodes. The nodes that will remain marked by the end of the process are the nodes that are padded on all levels. For a given graph H, let  $B_M(v,d,H)$  (M stands for marked) be the set of marked nodes at distance at most d from v in H.

For a subgraph G' and a node  $v \in V(G')$ , let  $Z_i(v, G') = |B_M(v, 2^i, G')|/|B_M(v, 2^{i-1}, G')|$ . The construction is given in Algorithm 1.

#### Algorithm 1 S = FPSDHPP(G, U)

```
1: Mark all the nodes in U.
2: Set \mathcal{X}_{\phi} = \{V\} to be the trivial partition. Set r(V) \in V to be the vertex v with maximal
   |B_M(v, 2^{\phi-1}, G)|.
3: for i from \phi - 1 to 0 do
      for every subset X \in \mathcal{X}_{i+1} do
4:
         Set H to be the induced graph on X.
5:
         while H contains a marked vertex do
6:
           Pick a node v \in V(H) with maximal |B_M(v, 2^{i-1}, H)|.
7:
           Let j(v) \geq 0 be the minimal integer such that
8:
           |B_M(v, 2^{i-1} + 2(j(v) + 1)\rho_i, H)| \le |B_M(v, 2^{i-1} + 2j(v)\rho_i, H)| \cdot |Z_i(v, H)|^{1/k}
           Let X(v) = B(v, 2^{i-1} + (2j(v) + 1)\rho_i, H).
9:
           Add X(v) to \mathcal{X}_i.
10:
           Unmark the nodes in B(v, 2^{i-1} + 2(j(v) + 1)\rho_i, H)
                                                                       \ B(v, 2^{i-1} + 2i(v)\rho_i, H).
11:
           Remove all nodes in X(v) from H.
12:
13:
         end while
      end for
14:
15: end for
16: set \hat{V} to be all nodes that remain marked.
```

Let X(v) be a set constructed in Line 9 of Algorithm 1, when partitioning  $X \in \mathcal{X}_{i+1}$ . We say that X is the parent of X(v). Let H(X(v)) denote the graph in Algorithm 1 just before X(v) was constructed (note that this is a graph induced on a subset of X). Similarly, let M(X(v)) denote the set of marked vertices just before X(v) was constructed. (We can also use H(X'), M(X') without (v), if the set X' is clear). We say that  $B(v, 2^{i-1} + 2j(v)\rho_i, H(X(v)))$  is the interior part of X(v). We also say that the set  $B_{M(X(v))}(v, 2^{i-1} + 2(j(v) + 1)\rho_i, H(X(v)))$  (from Line 8 in stage i of the algorithm), is the responsibility set of X(v), hereafter referred to as  $\operatorname{Res}(X(v))$ . Note that every node u that was marked before the processing of X started, belongs to exactly one set  $\operatorname{Res}(X(v))$  for  $X(v) \in \mathcal{X}_i$ .

We now define by induction the term *i-surviving* for  $0 \le i \le \phi$ : All nodes in U are  $\phi$ -surviving. We say that a node is *i*-surviving if it is (i+1)-surviving and it belongs to the interior part of some subset in  $\mathcal{X}_i$ , or equivalently if it remains marked after the construction of  $\mathcal{X}_i$ . We denote by  $S_i$  the set of *i*-surviving vertices. Note that  $S_{\phi} = U$ . Our goal in the analysis is to show that many vertices are 0-surviving, which is exactly the set  $\hat{V}$ . For a subset  $X \in \mathcal{X}_i$ , let  $\mathbf{Sur}(X)$  be the set of nodes in X that are 0-surviving, that is  $S_0 \cap X$ . We now turn to the analysis.

The next auxiliary claim helps in showing that property (ii) holds.

Claim 1. Consider a subset  $X \in \mathcal{X}_i$  centered at some node v = r(X). The index j(v) defined in Line 8 of Algorithm 1, satisfies  $j(v) \leq k - 1$ .

*Proof.* Seeking contradiction, assume that for every  $0 \le j' \le k - 1$ ,  $|B_{M(X)}(v, 2^{i-1} + 2(j' + 1)\rho_i, H(X))| > |B_{M(X)}(v, 2^{i-1} + 2j'\rho_i, H(X))| \cdot |Z_i(v, H(X))|^{1/k}$ . Then applying this for j' = k - 1, k - 2, ..., 0 we get

$$|B_{M(X)}(v, 2^{i}, H(X))| = |B_{M(X)}(v, 2^{i-1} + 2k\rho_{i}, H(X))|$$

$$> |B_{M(X)}(v, 2^{i-1} + 2(k-1)\rho_{i}, H(X))| \cdot |Z_{i}(v, H(X))|^{1/k}$$

$$> \dots > |B_{M(X)}(v, 2^{i-1}, H(X))| \cdot |Z_{i}(v, H(X))|^{k/k}$$

$$= |B_{M(X)}(v, 2^{i}, H(X))|,$$

a contradiction.  $\Box$ 

The next lemma shows that the collection  $\mathcal{S}$  satisfies the basic properties of an **FPSDHPP**.

**Lemma 2.** S is a collection of partial partitions which satisfies properties (i) and (ii).

*Proof.* It is straightforward from Line 12 that S is a collection of partial partitions. Property (i) holds as each X(v) is selected from the graph H(X(v)), which is an induced graph over a subset of X (the parent of X(v)). Finally, property (ii) follows from Claim 1, as the radius of X(v) is bounded by  $2^{i-1} + (2(k-1)+1)\rho_i < 2^{i-1} + 2k\rho_i = 2^i$ .

Next we argue that if a vertex is 0-surviving, then it is padded in all the levels.

**Lemma 3.** Suppose  $x \in Sur(V)$ , then x is padded in all the levels.

*Proof.* Fix some  $x \in \mathbf{Sur}(V)$ . To prove that x is i-padded, we assume inductively that x is j-padded for all  $i < j \le \phi$  (the base case  $i = \phi$  follows as  $B(x, \rho_{\phi}) \subseteq V$ ). Let  $X \in \mathcal{X}_{i+1}$  such that  $x \in X$ . Set  $B = B(x, \rho_i)$ . By the induction hypothesis  $B \subseteq B(x, \rho_{i+1}) \subseteq X$ . Let  $X(v) \in \mathcal{X}_i$  such that  $x \in X(v)$ .

First we argue that  $B \subseteq H(X(v))$ . Seeking contradiction, let  $X(v') \in \mathcal{X}_i$  be the first created cluster such that there is  $u \in B \cap X(v')$ . By the minimality of v', it follows that  $B \subseteq H(X(v'))$ . Thus  $\mathbf{dist}(v, u, H(X(v'))) = \mathbf{dist}(v, u, G) \leq \rho_i$ . Let j(v') the index chosen in Line 8 of Algorithm 1. Then  $X(v') = B(v', 2^{i-1} + (2j(v') + 1)\rho_i, H(X(v')))$ . Using the triangle inequality  $\mathbf{dist}(x, v', H(X(v'))) \leq 2^{i-1} + (2j(v') + 2)\rho_i$ . Therefore x was unmarked in Line 11, a contradiction.

It remains to show that  $B \subseteq X(v)$ . Set j(v) s.t.  $X(v) = B\left(v, 2^{i-1} + (2j(v) + 1)\rho_i, H(X(v))\right)$ . As x is part of the interior of X(v), it holds that  $\mathbf{dist}(x, v, H(X(v))) \leq 2^{i-1} + 2j(v)\rho_i$ . Therefore  $B \subseteq B\left(v, 2^{i-1} + (2j(v) + 1)\rho_i, H(X(v))\right) = X(v)$ .

The next lemma bounds the number of surviving nodes.

**Lemma 4.** For every index  $0 \le i \le \phi$  and every subset  $X = X(v) \in \mathcal{X}_i$ , the 0-surviving nodes satisfy  $|\mathbf{Sur}(X)| \ge |S_i \cap X| / |B_{M(X)}(r(X), 2^{i-1}, H(X))|^{\frac{1}{k}}$ 

Proof. We prove the lemma by induction on i. Consider first the base case where  $X = X(v) \in \mathcal{X}_0$ . As the subsets in  $\mathcal{X}_0$  contain a single node (their radius is less than 1), it holds that  $|\mathbf{Sur}(X)| = 1 = \frac{1}{1} = |S_0 \cap X| / |B_{M(X)}(r(X), 1/2, H(X))|^{\frac{1}{k}}$  (observe that each cluster in  $\mathcal{X}_i$  has at least 1 marked node, for all  $0 \le i \le \phi$ ). Assume the claim holds for every subset  $X' \in \mathcal{X}_i$ , and consider  $X \in \mathcal{X}_{i+1}$ . Let v = r(X). Consider the children  $X_1, ..., X_{j'}$  of X. For every  $1 \le h \le j'$ , set  $v_h = r(X_h)$ .

Note that by definition of  $j(v_h)$  in Line 8 of Algorithm 1 and by the construction of  $X_h$  in Line 9, we have that  $|S_i \cap X_h| \ge |\operatorname{Res}(X_h)|/Z_i(v_h, H(X_h))^{1/k}$ . Moreover, by the induction hypothesis we have that  $|\operatorname{Sur}(X_h)| \ge |S_i \cap X| / |B_{M(X_h)}(v_h, 2^{i-1}, H(X_h))|^{\frac{1}{k}}$  for every h.

We claim that  $|B_{M(X)}(v, 2^i, H(X))| \ge |B_{M(X_h)}(v_h, 2^i, H(X_h))|$  for every  $1 \le h \le j'$ . To see this, note that v is the node with maximal  $|B_{M(X)}(v, 2^i, H(X))|$ , hence  $|B_{M(X)}(v, 2^i, H(X))| \ge |B_{M(X)}(v_h, 2^i, H(X))|$ . In addition, note that  $H(X_h) \subseteq H(X)$  and  $M(X_h) \subseteq M(X)$ , hence  $|B_{M(X)}M(v_h, 2^i, H(X))| \ge |B_{M(X_h)}(v_h, 2^i, H(X_h))|$ .

It follows that  $|B_{M(X)}(v, 2^i, H(X))| \ge |B_{M(X_h)}(v_h, 2^i, H(X_h))|$ . Therefore the number of 0-surviving nodes in  $X_h$  is at least

$$|\mathbf{Sur}(X_{h})| \geq \frac{|S_{i} \cap X_{h}|}{|B_{M(X_{h})}(v_{h}, 2^{i-1}, H(X_{h}))|^{\frac{1}{k}}}$$

$$\geq \frac{|\mathbf{Res}(X_{h})| / |Z_{i}(v_{h}, H(X_{h}))|^{\frac{1}{k}}}{|B_{M(X_{h})}(v_{h}, 2^{i-1}, H(X_{h}))|^{\frac{1}{k}}}$$

$$= \frac{|\mathbf{Res}(X_{h})|}{|B_{M(X_{h})}(v_{h}, 2^{i}, H(X_{h}))|^{\frac{1}{k}}} \cdot \frac{|B_{M(X_{h})}(v_{h}, 2^{i-1}, H(X_{h}))|^{\frac{1}{k}}}{|B_{M(X_{h})}(v_{h}, 2^{i-1}, H(X_{h}))|^{\frac{1}{k}}}$$

$$\geq \frac{|\mathbf{Res}(X_{h})|}{|B_{M(X)}(v, 2^{i}, H(X))|^{\frac{1}{k}}},$$

we conclude that

$$\begin{aligned} |\mathbf{Sur}(X)| &= \sum_{h=1}^{j'} |\mathbf{Sur}(X_h)| \\ &\geq \sum_{h=1}^{j'} \frac{|\mathbf{Res}(X_h)|}{\left|B_{M(X)}(v, 2^i, H(X))\right|^{\frac{1}{k}}} \\ &= \frac{|S_{i+1} \cap X|}{\left|B_{M(X)}(v, 2^i, H(X))\right|^{\frac{1}{k}}} \ . \end{aligned}$$

Using Lemma 4 on V with  $i = \phi$  we have  $|\mathbf{Sur}(X)| \ge |S_{\phi} \cap X| / |U|^{\frac{1}{k}} = |U|^{1-\frac{1}{k}}$ . Combined with Lemma 3, property (iii) follows.

**Lemma 5.** The number of marked nodes  $\hat{V}$  by the end of Algorithm 1 is at least  $|U|^{1-1/k}$ . Moreover, for every  $v \in \hat{V}$  and every i, there exists a subset  $X \in \mathcal{X}_i$  such that  $B(v, \rho_i) \subseteq X$ .

**Theorem 5.** For every n-point metric space and  $k \ge 1$ , there exists a subset of size  $n^{1-1/k}$  that can be embedded into an ultrametric with distortion 8k - 2.

*Proof.* The hierarchical partial partition  $S = \{X_i\}$  naturally induces an ultrametric on  $\hat{V}$ . The singleton sets of  $\hat{V}$  are the leaves, and each  $X \in \mathcal{X}_i$  for  $0 \le i < \phi$  will be a tree-node which is connected to its parent. Each set in  $\mathcal{X}_i$  for  $i \ge 1$  will receive the label  $2^{i+1}(1-1/(4k))$ , while the leaves in  $\mathcal{X}_0$  receive the label 0 (recall Definition 1).

Consider two nodes  $u, v \in \hat{V}$ . Assume the least common ancestor of u, v is  $X \in \mathcal{X}_i$ , for some  $1 \leq i \leq \phi$ . Hence  $\mathbf{dist}(u, v, G) \leq 2 \cdot (2^i - 2^i/4k)$  (they are both in the interior of X - a ball with radius  $\leq 2^i - 2\rho_i$ ). Since this is the label of X, we conclude that distances in the ultrametric are no smaller than those in G.

Next we argue that distances increase by a factor of at most 8k-2. Consider any u,v as above, and seeking a contradiction, assume that  $\operatorname{\mathbf{dist}}(v,u,G)<\frac{2^{i+1}(1-1/(4k))}{8k-2}=\rho_i$ . Let P be the shortest path from v to u in G. As v was padded in X, necessarily  $P\subseteq X$ . Consider the first time a vertex  $z\in P$  was added to a cluster  $X'\in\mathcal{X}_{i-1}$ , then  $P\subseteq H(X')$ . Let j be such that  $X'=B(r(X'),2^{i-2}+(2j+1)\rho_{i-1},H(X'))$ . Since P is a shortest path, at least one of u,v must be within distance less than  $\rho_i/2=\rho_{i-1}$  from z, w.l.o.g assume  $\operatorname{\mathbf{dist}}(v,z,H(X'))\leq\rho_{i-1}$ . This implies that  $v\in\operatorname{\mathbf{Res}}(X')=B(r(X'),2^{i-2}+(2j+2)\rho_{i-1},H(X'))$ , and as v is marked it must lie in the interior of X', which is  $B(r(X'),2^{i-2}+2j\rho_{i-1},H(X'))$ . But then the triangle inequality yields that  $u\in\operatorname{\mathbf{Res}}(X')$ . Yet u and v belong to different clusters of  $\mathcal{X}_{i-1}$ , and so  $u\notin X'$ , which is a contradiction to the fact that  $u\in\hat{V}$ .

#### 3.1 Deterministic Construction of Distance Oracles

We show a distance oracle with  $O(n^{1+1/k})$  size,  $(8+\epsilon)k$  worst case stretch and  $O(1/\epsilon)$  query time (which is O(1) for any fixed epsilon). For simplicity we start by showing a construction similar to the [MN07] oracle, with  $O(k \cdot n^{1+1/k})$  size, 16k stretch and O(1) query time. We will later see how to reduce the size and stretch. Let D be the diameter of the graph.

Our distance oracle is constructed as follows.

- 1. Set  $U \leftarrow V$ .
- 2. Construct the collection of partial partitions  $S(U) = \mathbf{FPSDHPP}(G, U)$  on the graph G and the set U. Remove from U the set of nodes  $\hat{V}$  that were padded in all levels in S(U). Continue this process as long as  $U \neq \emptyset$ .
- 3. Let  $\mathcal{M}$  be the set of all collections  $\mathcal{S}(U)$  that were constructed by this process.
- 4. For every  $S \in \mathcal{M}$  construct a cluster X(S) as follows.
- 5. Let  $S = \{X_0, ..., X_{\phi}\}$ . All nodes V are the leaves (recall that only nodes in  $\hat{V}$  are in  $X_0$ ). For every index i and every set  $X \in \mathcal{X}_i$ , add an intermediate node. Connect X to its parent set. Connect every node  $v \in V$  to the set  $X \in \mathcal{X}_i$  of minimal i such that  $v \in X$ . This completes the construction of X(S).
- 6. In addition, we preprocess X(S) so that least common ancestor (LCA) queries could be done in constant time. In order to do that we invoke any scheme that takes a tree and preprocess it in linear time so that LCA queries can be answered in constant time (see [HT84, BFC00]).
- 7. Finally, note that for every node v there exists a collection  $S \in \mathcal{M}$ , where v is padded in all levels. Denote this collection by  $\mathbf{home}(v)$ .

The query phase is done as follows. Given two nodes s and t. Let  $S = \mathbf{home}(s)$  and let  $S = \{X_i \mid 1 \leq i \leq \phi\}$ . Find the least common ancestor of s and t in X(S) and let i be its level. Namely, let  $\mu \in X(S)$  be the least common ancestor of s and t and let X be the cluster  $\mu$  represents, the index i is the index such that  $X \in \mathcal{X}_i$ . Return  $2^{i+1}$  (denoted by  $\mathbf{dist}(s,t)$ ).

Lemma 6.  $\operatorname{dist}(s,t) \leq \operatorname{dist}(s,t) < 16k \cdot \operatorname{dist}(s,t)$ .

*Proof.* Let  $d = \mathbf{dist}(s, t, G)$  and let j be the index such that  $2^{j-1} < d \le 2^j$ . Let  $X_i \in \mathcal{X}_i \in \mathcal{S} = \mathbf{home}(s)$  be the i level subset such that  $s \in X_i$ . Recall that s is padded in all the subsets  $X_i$  for  $0 \le i \le \phi$ .

Note that  $X_i$  has diameter smaller than  $2 \cdot 2^i$  (follows from property (iii)). Therefore  $t \in X_i$  implies that  $\mathbf{dist}(s,t,G) < 2 \cdot 2^i = 2^{i+1}$ . In particular,  $t \notin X_i$  for every i < j-1. Hence the least common ancestor is at least at level j-1. Hence the minimal distance returned by the algorithm is  $\mathbf{dist}(s,t) \geq 2^j \geq d$ .

It remains to show that  $\mathbf{dist}(s,t) \leq 16k \cdot d$ . Let i be the level of s and t's least common ancestor. Note that  $t \notin X_{i-1}$ . Also recall that s is padded in  $X_{i-1}$  and thus  $B(s, \rho_{i-1}) \subseteq X_{i-1}$ , which implies  $d \geq \rho_{i-1} = 2^{i-1}/(4k) = \mathbf{dist}(s,t)/(16k)$ .

Let  $U_i$  be the set U after constructing the first i collections. Note that  $|U_{i+1}| \leq |U_i| - |U_i|^{1-1/k}$ . By resolving this recurrence relation one can show that the number of phases is  $O(kn^{1/k})$  (see [MN07, Lemma 4.2]). Notice that for every  $S \in \mathcal{M}$ , T(S) is of size O(n). Hence, the size of our data structure is  $O(kn^{1+1/k})$ .

Reducing the size of the data structure: We now show how to reduce the size of the data structure to  $O(n^{1+1/k})$ . We only outline the modifications to the algorithm and the analysis and omit the full details.

Here we will use only the metric structure of the graph G, while ignoring the structure induced by the edges. Specifically, in line (2) of the algorithm, instead of the graph G, we will use the graph  $G_U$  which is the complete graph over U, where the weight of each edge  $\{u, v\}$  is equal to  $\operatorname{dist}(u, v, G)$ . This change allows us to remove the nodes  $\hat{V}$  from  $G_U$  after each iteration.

The query algorithm, given two nodes s and t is as follows. Let  $S_s = \mathbf{home}(s)$  and  $S_t = \mathbf{home}(t)$ , and assume w.l.o.g. that  $S_s$  was constructed before  $S_t$ . Find the least common ancestor of s and t in  $X(S_s)$  and let i be its level. Return  $2^{i+1}$ .

Following the analysis of the previous construction we can show that properties (i) - (iv) are satisfied and that the stretch is bound by 16k. The size of the data structure is bounded by  $O(n^{1+1/k})$  (see [MN07], Lemma 4.2.

Reducing the stretch to  $8(1+\epsilon)k$ : We now explain how to reduce the stretch to  $8(1+\epsilon)k$ . Note that we lose a factor of 2 in the stretch since we look on distances in multiplies of two. Recall that in the algorithm, for a pair of vertices s,t at distance d, we looked on the minimal index j such that  $d \leq 2^j$ . It may happen that d is only slightly larger than  $2^{j-1}$ . Note that by just considering all distances  $(1+\epsilon)^i$  rather than all distances  $2^i$ , we get that the number of nodes that are padded in all levels is a fraction of  $1/n^{1/(\epsilon k)}$  rather than  $1/n^{1/k}$ , which is dissatisfying. So instead we construct  $O(1/\epsilon)$  different copies of our data structure, one for each  $1+\ell\epsilon$  for  $0 \leq \ell < 1/\epsilon$ . In the copy  $\ell$  of the data structure we consider distances  $(1+\ell\epsilon)2^i$  for every  $0 \leq i \leq \phi$ . Specifically, i-clusters have radius bounded by  $(1+\ell\epsilon)2^i$ , while the padding parameter is  $\rho_{\ell,i} = (1+\ell\epsilon)\rho_i$ . We denote by  $\operatorname{home}_{\ell}(s)$  the collection  $\mathcal{S}$ , created for the  $\ell$ 's distance oracle, where s is padded in all levels. The distance estimation of the  $\ell$ 's copy (denoted by  $\operatorname{dist}_{\ell}(s,t)$ ), will be  $(1+(\ell+1)\epsilon)2^{i\ell}$ , where  $i_{\ell}$  is the level of the least common ancestor of s and t in  $\operatorname{home}_{\ell}(s)$ .

Set  $d = \mathbf{dist}(s, t)$ . For every  $\ell$ , we have

$$d > \rho_{\ell, i_{\ell} - 1} = \frac{(1 + \ell \epsilon) 2^{i_{\ell} - 1}}{4k} = \frac{(1 + \ell \epsilon)}{(1 + (\ell + 1)\epsilon)} \cdot \frac{\mathbf{dist}_{\ell}(s, t)}{8k} \ge \frac{\mathbf{dist}_{\ell}(s, t)}{8(1 + \epsilon)k} . \tag{1}$$

On the other hand, there exist indices  $\ell', j$  such that  $(1 + \ell' \epsilon)2^{j-1} < d \le (1 + (\ell' + 1)\epsilon)2^{j-1}$ . Following the analysis above, as t is not separated from s at level  $i_{\ell'}$ , it holds that  $i_{\ell'} \ge j-1$ . Therefore

$$\hat{\mathbf{dist}}_{\ell'}(s,t) = (1 + (\ell+1)\epsilon)2^{i_{\ell'}} \ge (1 + (\ell+1)\epsilon)2^{j-1} \ge d.$$
 (2)

In the query phase we iterate over all  $O(1/\epsilon)$  copies, invoke the query algorithm in each copy and return the largest distance. By equations (1) and (2), the stretch is  $8(1+\epsilon)k$  rather than 16k. The query time is  $O(1/\epsilon)$  which is O(1) for every fixed  $\epsilon$ .

# 4 Ramsey Spanning Trees

In this section we describe the construction of Ramsey spanning trees, each tree will be built using the petal decomposition framework of [AN12]. Roughly speaking, the petal decomposition is an iterative method to build a spanning tree of a given graph. In each level, the current graph is partitioned into smaller diameter pieces, called petals, and a single central piece, which are then connected by edges in a tree structure. Each of the petals is a ball in a certain metric. The main advantage of this framework is that it produces a spanning tree whose diameter is proportional to the diameter of the graph, while allowing large freedom for the choice of radii of the petals. Specifically, if the graph diameter is  $\Delta$ , the spanning tree diameter will be  $O(\Delta)$ , and each radius can be chosen in an interval of length  $\approx \Delta$ . For the specific choice of radii that will ensure a sufficient number of vertices are fully padded, we use a region growing technique based on ideas from [Sey95, EEST05].

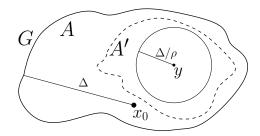


Figure 1: A is a subset of vertices in the graph G with center  $x_0$ . The radius of A (w.r.t  $x_0$ ) is  $\Delta$ . A' is a subset of A (denoted by the dashed line). As  $B(y, \Delta/\rho, G) \subseteq A'$ , we say that the vertex y is padded by A'.

#### 4.1 Preliminaries

For subset  $S \subseteq G$  and a center vertex  $x_0 \in S$ , the radius<sup>2</sup> of S w.r.t  $x_0, \Delta_{x_0}(S)$ , is the minimal  $\Delta$  such that  $B(x_0, \Delta, S) = S$ . (If for every  $\Delta$ ,  $B(x_0, \Delta, S) \neq S$ , (this can happen iff S is not connected) we say that  $\Delta_{x_0}(S) = \infty$ .) When the center  $x_0$  is clear from context or is not relevant, we will omit it.

**Definition 3.** Given a graph G = (V, E), a Strong Diameter Hierarchical Partition (**SDHP**) is a collection  $\{A_i\}_{i\in [\Phi]}$  of partitions of V, where each cluster  $X \in \mathcal{X}_i$  contains a center  $r(X) \in \mathcal{X}_i$ , such that:

- $\mathcal{A}_{\Phi} = \{V\}$  (i.e., the first partition is the trivial one).
- $A_1 = \{\{v\}_{v \in V}\}$  (i.e., in the last partition every cluster is a singleton).
- For every  $1 \le i < \Phi$  and  $A \in \mathcal{A}_i$ , there is  $A' \in \mathcal{A}_{i+1}$  such that  $A \subseteq A'$  (i.e.,  $\mathcal{A}_i$  is a refinement of  $\mathcal{A}_{i+1}$ ). Moreover,  $\Delta(A) \le \Delta(A')$ .

**Definition 4** (Padded, Fully Padded). Given a graph G = (V, E) and a subset  $A \subseteq V$ , we say that a vertex  $y \in A$  is  $\rho$ -padded by a subset  $A' \subseteq A$  (w.r.t A) if  $B(y, \Delta(A)/\rho, G) \subseteq A'$ . See Figure 1 for illustration.

We say that  $x \in V$  is  $\rho$ -fully-padded in the SDHP  $\{A_i\}_{i \in [\Phi]}$ , if for every  $2 \leq i \leq \Phi$  and  $A \in A_i$  such that  $x \in A$ , there exists  $A' \in A_{i-1}$  such that x is  $\rho$ -padded by A' (w.r.t A).

#### 4.2 Petal Decomposition

Here we will give a concise description of the Petal decomposition algorithm, focusing on the main properties we will use. For proofs and further details we refer to [AN12].

The hierarchical-petal-decomposition (see Algorithm 2) is a recursive algorithm. The input is G[X] (a graph G=(V,E) induced over a set of vertices  $X\subseteq V$ ), a center  $x_0\in X$ , a target  $t\in X$ , and the radius  $\Delta=\Delta_{x_0}(X)$ . The algorithm invokes petal-decomposition to partition

<sup>&</sup>lt;sup>2</sup>In the literature, this notion is sometimes called the eccentricity of  $x_0$ .

<sup>&</sup>lt;sup>3</sup>Rather than inferring  $\Delta = \Delta_{x_0}(X)$  from G[X] and  $x_0$  as in [AN12], we can think of  $\Delta$  as part of the input. We shall allow any  $\Delta \geq \Delta_{x_0}(X)$ . We stress that in fact in the algorithm we always use  $\Delta_{x_0}(X)$ , and consider this degree of freedom only in the analysis.

X into  $X_0, X_1, \ldots, X_s$  (for some integer s), and also provides a set of edges  $\{(x_1, y_1), \ldots, (x_s, y_s)\}$  and targets  $t_0, t_1, \ldots, t_s$ . The Hierarchical-petal-decomposition algorithm now recurses on each  $(G[X_j], x_j, t_j, \Delta_{x_j}(X_j))$  for  $0 \le j \le s$ , to get trees  $\{T_j\}_{0 \le j \le s}$ , which are then connected by the edges  $\{(x_j, y_j)\}_{1 \le j \le s}$  to form a spanning tree T for G[X] (the recursion ends when  $X_j$  is a singleton). See Figure 2 for illustration.

## Algorithm 2 T =

hierarchical-petal-decomposition $(G[X], x_0, t, \Delta)$ 

```
1: if |X| = 1 then

2: return G[X].

3: end if

4: Let \left(\{X_j, x_j, t_j, \Delta_j\}_{j=0}^s, \{(y_j, x_j)\}_{j=1}^s\right) = \text{petal-decomposition}(G[X], x_0, t, \Delta);

5: for each j \in [0, \dots, s] do

6: T_j = \text{hierarchical-petal-decomposition}(G[X_j], x_j, t_j, \Delta_j);

7: end for

8: Let T be the tree formed by connecting T_0, \dots, T_s using the edges \{y_1, x_1\}, \dots, \{y_s, x_s\};
```

Next we describe the petal-decomposition procedure, see Algorithm 3. Initially it sets  $Y_0 = X$ , and for j = 1, 2, ..., s it carves out the petal  $X_j$  from the graph induced on  $Y_{j-1}$ , and sets  $Y_j = Y_{j-1} \setminus X_j$  (in order to control the radius increase, the first petal is cut using different parameters). The definition of petal guarantees that  $\Delta_{x_0}(Y_j)$  is non-increasing (see [AN12, Claim 1]), and when at step s it becomes at most  $3\Delta/4$ , define  $X_0 = Y_s$  and then the petal-decomposition routine ends. In carving of the petal  $X_j \subseteq Y_{j-1}$ , the algorithm chooses an arbitrary target  $t_j \in Y_{j-1}$  (at distance at least  $3\Delta/4$  from  $x_0$ ) and a range [lo, hi] of size  $hi - lo \in \{\Delta/8, \Delta/4\}$  which are provided to the sub-routine create-petal.

(One may notice that in Line 15 of the petal-decomposition procedure, the weight of some edges is changed by a factor of 2. This can happen at most once for every edge throughout the hierarchical-petal-decomposition execution, thus it may affect the padding parameter by a factor of at most 2. This re-weighting is ignored for simplicity.)

Both hierarchical-petal-decomposition and petal-decomposition are essentially the algorithms that appeared in [AN12]. The main difference from their work lies in the create-petal procedure, depicted in Algorithm 4. It carefully selects a radius  $r \in [lo, hi]$ , which determines the petal  $X_j$  together with a connecting edge  $(x_j, y_j) \in E$ , where  $x_j \in X_j$  is the center of  $X_j$  and  $y_j \in Y_j$ . It is important to note that the target  $t_0 \in X_0$  of the central cluster  $X_0$  is determined during the creation of the first petal  $X_1$ . It uses an alternative metric on the graph, known as cone-metric:

```
Definition 5 (Cone-metric). Given a graph G = (V, E), a subset X \subseteq V and points x, y \in X, define the cone-metric \rho = \rho(X, x, y) : X^2 \to \mathbb{R}^+ as \rho(u, v) = |(d_X(x, u) - d_X(y, u)) - (d_X(x, v) - d_X(y, v))|.
```

(The cone-metric is in fact a pseudo-metric, i.e., distances between distinct points are allowed to be 0.) The ball  $B_{(X,\rho)}(y,r)$  in the cone-metric  $\rho = \rho(X,x,y)$ , contains all vertices u whose shortest path to x is increased (additively) by at most r if forced to go through y.

In the create-petal algorithm, while working in a subgraph G[Y] with two specified vertices: a center  $x_0$  and a target t, we define  $W_r(Y, x_0, t) = \bigcup_{p \in P_{x_0t}: d_Y(p,t) \le r} B_{(Y,\rho(Y,x_0,p))}(p, \frac{r-d_Y(p,t)}{2})$  which

```
\mathbf{Algorithm\ 3}\ \left(\left\{X_j,x_j,t_j,\Delta_j\right\}_{j=0}^s,\left\{(y_j,x_j)\right\}_{j=1}^s\right) = \mathtt{petal-decomposition}(G[X],x_0,t,\Delta)
 1: Let Y_0 = X; Set j = 1;
 2: if d_X(x_0,t) \geq \Delta/2 then
        Let X_1 = \text{create-petal}
        (G[Y_0], [d_X(x_0,t)-\Delta/2, d_X(x_0,t)-\Delta/4], x_0, t);
        Y_1 = Y_0 \setminus X_1;
        Let \{x_1, y_1\} be the unique edge on the shortest path P_{x_0t} from x_0 to t in Y_0, where x_1 \in X_1
        and y_1 \in Y_1.
        Set t_0 = y_1, t_1 = t; j = 2;
 7: else
        set t_0 = t.
 8:
 9: end if
10: while Y_{j-1} \setminus B_X(x_0, \frac{3}{4}\Delta) \neq \emptyset do
        Let t_j \in Y_{j-1} be an arbitrary vertex satisfying d_X(x_0, t_j) > \frac{3}{4}\Delta;
        Let X_j = \text{create-petal}(G[Y_j], [0, \Delta/8], x_0, t_j);
12:
        Y_j = Y_{j-1} \setminus X_j;
13:
        Let \{x_j, y_j\} be the unique edge on the shortest path P_{x_j t_j} from x_0 to t_j in Y_{j-1}, where
14:
        x_j \in X_j and y_j \in Y_j.
        For each edge e \in P_{x_i t_i}, set its weight to be w(e)/2;
15:
        Let j = j + 1;
16:
17: end while
18: Let s = j - 1;
19: Let X_0 = Y_s;
20: return (\{X_j, x_j, t_j, \Delta_{x_j}(X_j)\}_{j=0}^s, \{(y_j, x_j)\}_{j=1}^s);
```

is union of balls in the cone-metric, where any vertex p in the shortest path from  $x_0$  to t of distance at most r from t is a center of a ball with radius  $\frac{r-d_Y(p,t)}{2}$ .

The following facts are from [AN12].

Fact 1. Running Hierarchical-petal-decomposition on input  $(G[X], x_0, t, \Delta_{x_0}(X))$  will provide a spanning tree T satisfying

$$\Delta_{x_0}(T) \le 4\Delta_{x_0}(X).$$

Fact 2. If the petal-decomposition partitions X with center  $x_0$  into  $X_0, \ldots, X_s$  with centers  $x_0, \ldots, x_s$ , then for any  $0 \le j \le s$  we have  $\Delta_{x_j}(X_j) \le (3/4) \cdot \Delta_{x_0}(X)$ .

We will need the following observation. Roughly speaking, it says that when the petal-decomposition algorithm is carving out  $X_{j+1}$ , it is oblivious to the past petals  $X_1, \ldots, X_j$ , edges and targets – it only cares about  $Y_j$  and the original diameter  $\Delta$ .

Observation 7. Assume that petal-decomposition on input  $(G[X], x_0, t, \Delta_{x_0}(X))$  returns as output  $(X_0, X_1, \ldots, X_s, \{y_1, x_1\}, \ldots, \{y_s, x_s\}, t_0, \ldots, t_s)$ . Then running petal-decomposition on input  $(G[Y_j], x_0, t_0, \Delta_{x_0}(X))$  will output  $(X_0, X_{j+1}, \ldots, X_s, \{y_{j+1}, x_{j+1}\}, \ldots, \{y_s, x_s\}, t_0, t_{j+1}, \ldots, t_s)$ .

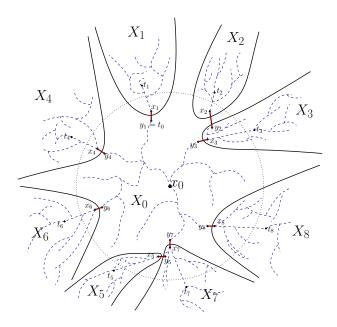


Figure 2: A schematic depiction of a partition done by the petal-decomposition algorithm. The dotted ball contain the vertices within  $3\Delta/4$  from the center  $x_0$ . The algorithm iteratively picks a target  $t_j$  outside this ball, and builds a petal  $X_j$ , that will be connected to the rest of the graph by the edge  $(x_j, y_j)$ . The dotted lines represent the trees created recursively in each  $X_j$ .

### 4.3 Choosing a Radius

Fix some  $1 \leq j \leq s$ , and consider carving the petal  $X_j$  from the graph induced on  $Y = Y_{j-1}$  (i.e. Line 12 in the petal-decomposition algorithm). While the algorithm of [AN12] described a specific way to choose the radius, we require a somewhat more refined choice. The properties of the petal decomposition described above (in Subsection 4.2) together with Fact 2 and Fact 1, hold for any radius picked from a given interval. We will now describe the method to select a radius that suits our needs. The petal-decomposition algorithm provides an interval [lo, hi] of size at least  $\Delta/8$ , and for each  $r \in [lo, hi]$  let  $W_r(Y, x_0, t) \subseteq Y$  denote the petal of radius r (usually we will omit  $(Y, x_0, t)$ ). The following fact demonstrates that petals are similar to balls.

**Fact 3.** For every  $y \in W_r$  and  $l \ge 0$ ,  $B(y, l, Y) \subseteq W_{r+4l}$ .

Note that Fact 3 implies that  $W_r$  is monotone in r, i.e., for  $r \leq r'$  it holds that  $W_r \subseteq W_{r'}$ .

Our algorithm will maintain a set of marked vertices  $M \subseteq V$ , and will update it in any petal creation. Roughly speaking, the marked vertices are those that are fully padded in the (partial) hierarchical partition generated so far by the algorithm. If initially |M| = m, we want that at the end of the process at least  $m^{1-1/k}$  vertices will remain marked. In the partition of X to  $X_0, \ldots, X_s$ , some of the marked vertices will be  $\rho$ -padded by a petal  $X_j$  (w.r.t. X), and some of the others will be unmarked, by the following rule. Fact 3 implies that if we choose a radius r when creating some petal  $X_j = W_r$ , then all marked vertices in  $W_{r-4\Delta/\rho}$  will be  $\rho$ -padded by  $X_j$ , and thus remain marked. All the marked vertices in  $W_{r+4\Delta/\rho} \setminus W_{r-4\Delta/\rho}$  are considered unmarked from now on, since their  $\Delta/\rho$  ball may intersect more than one cluster in the current partition (note that some

of these vertices can be outside  $X_i$ ).

Our algorithm to select a radius is based on region growing techniques, similar to those in Algorithm 1, but rather more involved. Since in the petal decomposition framework we cannot pick as center a vertex maximizing the "small ball", we first choose an appropriate range that mimics that choice (see e.g. Line 9 in the algorithm below) – this is the reason for the extra factor of log log n. The basic idea in region growing is to charge the number of marked vertices whose ball is cut by the partition (those in  $W_{r+4\Delta/\rho} \setminus W_{r-4\Delta/\rho}$ ), to those that are saved (in  $W_{r-4\Delta/\rho}$ ). In our setting we are very sensitive to constant factors in this charging scheme (as opposed to the average stretch considered in [EEST05]), because these constants are multiplied throughout the recursion. In particular, we must avoid a range in [lo, hi] that contains more than half of the marked vertices, a constraint which did not exist in previous manifestation of this region growing scheme. To this end, if the first half [lo, mid] (with mid = (hi + lo)/2) is not suitable, we must "cut backwards" in the regime [mid, hi], and charge the marked vertices that were removed from M to the remaining graph  $Y_j$ , rather than to those saved in the created cluster  $X_j$ .

# **Algorithm 4** $X = \text{create-petal}(G[Y], [lo, hi], x_0, t)$

```
1: Let m = |Y \cap M|;
 2: L = \lceil 1 + \log \log m \rceil;
 3: R = hi - lo; mid = (lo + hi)/2 = lo + R/2;
 4: For every r, denote W_r = W_r(Y, x_0, t), w_r = |M \cap W_r|;
 5: if w_{mid} \leq \frac{m}{2} then
        if w_{lo+\frac{R}{2L}}=0 then
            Set r = lo + \frac{R}{4I};
 7:
        else
 8:
            Choose [a,b] \subseteq [lo,mid] such that b-a=\frac{R}{2L} and w_a \ge w_b^2/m; {see Lemma 11}
 9:
           Pick r \in [a,b] such that w_{r+\frac{b-a}{2k}} \leq w_{r-\frac{b-a}{2k}} \cdot \left(\frac{w_b}{w_a}\right)^{\frac{1}{k}}; {see Lemma 12}
10:
        end if
11:
12: else
        For every r \in [lo, hi], denote q_r = |(Y \setminus W_r) \cap M|;
13:
        if q_{hi-\frac{R}{2L}}=0 then
14:
           Set r = hi - \frac{R}{4L};
15:
        else
16:
            Choose [b,a] \subseteq [mid,hi] such that a-b=\frac{R}{2L} and q_a \ge q_b^2/m; {see Lemma 13}
17:
           Pick r \in [b,a] such that q_{r-\frac{a-b}{2k}} \leq q_{r+\frac{a-b}{2k}} \cdot \left(\frac{q_b}{q_a}\right)^{\frac{1}{k}}; {see Lemma 14}
18:
        end if
19:
20: end if
21: M \leftarrow M \setminus (W_{r+\frac{R}{4Lk}} \setminus W_{r-\frac{R}{4Lk}})
22: return W_r;
```

#### 4.4 Proof of Correctness

Let  $z \in V$  be an arbitrary vertex, given a set  $M \subseteq V$ , let T be the tree returned by calling Hierarchical-petal-decomposition on  $(G[V], z, z, \Delta_z(V))$  and marked vertices M. There is a natural Strong Diameter Hierarchical Partition (SDHP)  $\mathcal{X} = \{\mathcal{X}_i\}_{i=1}^{\Phi}$  associated with the tree T, where  $\mathcal{X}_i$  consists of all the clusters created in level  $\Phi - i$  of the recursion (and  $\mathcal{X}_{\Phi} = \{V\}$ ). By Fact 2, the radius is always non-increasing. Hence  $\mathcal{X}$  is indeed a SDHP. Denote by  $\mathbf{Sur}(M) \subseteq M$  the set of vertices that remained marked throughout the execution of Hierarchical-petal-decomposition.

**Lemma 8.** Suppose  $x \in \mathbf{Sur}(M)$ , then x is  $\rho$ -fully-padded in  $\mathcal{X}$ .

Proof. Fix any  $2 \leq i \leq \Phi$ . Let  $X \in \mathcal{X}_i$  be the cluster containing x in the  $(\Phi - i)$ -th level of the recursion with  $\Delta = \Delta(X)$ . Assume X was partitioned by petal-decomposition into  $X_0, \ldots, X_s$ , and let  $X_j \subseteq X$  be the cluster containing  $x \in X_j$ . Assuming (inductively) that x was  $\rho$ -padded by X, we need to show that it is also  $\rho$ -padded by  $X_j$ , that is,  $B = B(x, \Delta/\rho, G) \subseteq X_j$ . (Note that  $B \subseteq X$  since the radii are non-increasing, so x is padded in all higher levels.)

First we argue that none of the petals  $X_1, \ldots, X_{j-1}$  intersects B. Seeking contradiction, assume it is not the case, and let  $1 \leq j' < j$  be the minimal such that there exists  $y \in X_{j'} \cap B$ . By the minimality of j' it follows that  $B \subseteq Y' = Y_{j'-1}$ , and thus  $\operatorname{\mathbf{dist}}(x,y,Y') = \operatorname{\mathbf{dist}}(x,y,G) \leq \Delta/\rho$ . Let r' be the radius chosen when creating the petal  $X_{j'} = W'_{r'}$ , and Fact 3 implies that

$$x \in B(y, \Delta/\rho, Y') \subseteq W'_{r'+4\Delta/\rho} = W'_{r'+R/(4Lk)}$$
,

where we recall that  $\Delta = 8R$  and  $\rho = 2^7 Lk$ . This is a contradiction to the fact that  $x \in \mathbf{Sur}(X)$ : clearly  $x \notin W'_{r'-R/(4Lk)}$  since it is not included in  $X_{j'} = W'_{r'}$  (and using the monotonicity of  $W'_r$ ), so it should have been removed from M when creating  $X_{j'}$  (in Line 21 of the algorithm).

For the case j=0 the same reasoning shows B does not intersect any petal  $X_1, \ldots, X_s$  and we are done. For j>0, it remains to show that  $B\subseteq X_j$ , but this follows by a similar calculation. Let r be the radius chosen for creating the petal  $X_j=W_r$ , and  $Y=Y_{j-1}$ . We have  $B\subseteq Y$ , and since  $x\in \mathbf{Sur}(X)$  it must be that  $x\in W_{r-R/(4Lk)}$ . Again by Fact 3 we have

$$B = B(x, \Delta/\rho, G) = B(x, \Delta/\rho, Y)$$
  

$$\subseteq W_{r-R/(4Lk)+4\Delta/\rho} = W_r = X_j.$$

**Lemma 9.** Consider a vertex  $v \in \mathbf{Sur}(M)$ , then for every  $u \in V$ ,  $\mathbf{dist}(v, u, T) \leq 8\rho \cdot \mathbf{dist}(v, u, G)$ .

Proof. Let  $\mathcal{X} = \{\mathcal{X}_i\}_{i=1}^{\Phi}$  be the **SDHP** associated with T, and for  $1 \leq i \leq \Phi$  let  $A_i \in \mathcal{X}_i$  be the cluster containing v. Take the minimal  $2 \leq i \leq \Phi$  such that  $u \in A_i$  (there exists such an i since  $u \in A_{\Phi} = V$  and  $u \notin A_1 = \{v\}$ ). By Lemma 8 v is  $\rho$ -fully-padded, so we have that  $B(v, \Delta/\rho, G) \subseteq A_{i-1}$ , where  $\Delta = \Delta(A_i)$ . But as  $u \notin A_{i-1}$ , it must be that  $\operatorname{dist}(u, v, G) > \Delta/\rho$ . Since both  $u, v \in A_i$ , Fact 1 implies that the radius of the tree created for  $A_i$  is at most  $4\Delta$ , so that

$$\mathbf{dist}(u, v, T) \le 2 \cdot 4\Delta \le 8\rho \cdot \mathbf{dist}(u, v, G)$$
.

**Lemma 10.**  $|Sur(M)| \ge |M|^{1-1/k}$ .

*Proof.* We prove by induction on |X| that if a cluster  $X \in \mathcal{X}_i$  (for some  $1 \leq i \leq \Phi$ ) has currently m marked vertices, then at the end of the process at least  $m^{1-1/k}$  of them will remain marked.

The base case when X is a singleton is trivial. For the inductive step, assume we call petal-decomposition on  $(G[X], x_0, t, \Delta)$  with  $\Delta \geq \Delta_{x_0}(X)$  and the current marked vertices  $\hat{M}$ . Assume that the petal-decomposition algorithm does a non-trivial partition of X to  $X_0, \ldots, X_s$  (if it is the case that all vertices are sufficiently close to  $x_0$ , then no petals will be created, and the hierarchical-petal-decomposition will simply recurse on  $(G[X], x_0, t, \Delta_{x_0}(X))$ , so we can ignore this case). Denote by  $M_j$  the marked vertices that remain in  $X_j$  (just before the recursive call on  $X_j$ ), and recall that  $\mathbf{Sur}(X_j)$  is the set of vertices of  $M_j$  that remain marked until the end of the hierarchical-petal-decomposition algorithm. Then  $\mathbf{Sur}(X) = \bigcup_{0 \leq j \leq s} \mathbf{Sur}(X_j)$ , and we want to prove that  $|\mathbf{Sur}(X)| \geq m^{1-1/k}$ .

Let  $X_1 = W_r$  be the first petal created by the petal-decomposition algorithm, and  $Y_1 = X \setminus X_1$ . Denote by  $\mathbf{Res}(X_1) = W_{r+R/(4Lk)} \cap \hat{M}$  the responsibility set for  $X_1$  (i.e. the marked vertices that are either in  $M_1$  or were removed from  $\hat{M}$  when  $X_1$  was created). Define  $M' = \hat{M} \setminus \mathbf{Res}(X_1)$ , the set of marked vertices that remain in  $Y_1$ . By Observation 7, we can consider the remaining execution of petal-decomposition on  $Y_1$  as a new recursive call of petal-decomposition with input  $(G[Y_1], x_0, t_0, \Delta)$  and marked vertices M'. Since  $|X_1|, |Y_1| < |X|$ , the induction hypothesis implies that  $|\mathbf{Sur}(X_1)| \ge |M_1|^{1-1/k}$  and  $|\mathbf{Sur}(Y_1)| \ge |M'|^{1-1/k}$ .

We now do a case analysis according to the choice of radius in Algorithm 4.

- 1. Case 1:  $w_{mid} \leq m/2$  and  $w_{lo+R/(2L)} = 0$ . In this case we set r = lo + R/(4L). Note that  $w_{r+R/(4Lk)} \leq w_{lo+R/(2L)} = 0$ , so  $M' = \hat{M}$ , and by the induction hypothesis on  $Y_1$ , the number of fully padded vertices is  $|\mathbf{Sur}(X)| = |\mathbf{Sur}(Y_1)| \geq |\hat{M}|^{1-1/k} = m^{1-1/k}$ , as required.
- 2. Case 2:  $w_{mid} \leq m/2$  and  $w_{lo+R/(2L)} > 0$ . In this case we pick  $a, b \in [lo, hi]$  so that b-a=R/(2L) and

$$w_a > w_b^2/m (3)$$

and also choose  $r \in [a,b]$  such that  $w_{r+\frac{b-a}{2k}} \le w_{r-\frac{b-a}{2k}} \cdot \left(\frac{w_b}{w_a}\right)^{1/k}$ . As  $\frac{b-a}{2k} = \frac{R}{4Lk}$  and  $|M_1| = w_{r-R/(4Lk)}$  we have that

$$|M_1| \ge \mathbf{Res}(X_1) \cdot \left(\frac{w_a}{w_b}\right)^{1/k}$$
 (4)

By the induction hypothesis on  $X_1$  we have that

$$|\mathbf{Sur}(X_1)| \ge \frac{|M_1|}{|M_1|^{1/k}}$$

$$\stackrel{(4)}{\ge} |\mathbf{Res}(X_1)| \cdot \left(\frac{w_a}{|M_1| \cdot w_b}\right)^{1/k}$$

$$\stackrel{(3)}{\ge} |\mathbf{Res}(X_1)| \cdot \left(\frac{w_b}{m \cdot |M_1|}\right)^{1/k}$$

$$\ge \frac{|\mathbf{Res}(X_1)|}{m^{1/k}},$$

where in the last inequality we use that  $|M_1| = w_{r-(b-a)/(2k)} \le w_b$ . Now by the induction hypothesis on  $Y_1$  we get

$$\begin{aligned} |\mathbf{Sur}(X)| &= |\mathbf{Sur}(Y_1)| + |\mathbf{Sur}(X_1)| \\ &\geq |M'|^{1-1/k} + \frac{|\mathbf{Res}(X_1)|}{m^{1/k}} \\ &\geq \frac{|M'| + |\mathbf{Res}(X_1)|}{m^{1/k}} = \frac{|\hat{M}|}{m^{1/k}} = m^{1-1/k} \ . \end{aligned}$$

- 3. Case 3:  $w_{mid} > m/2$  and  $q_{hi-R/(2L)} = 0$ . In this case we set r = hi R/(4L). Note that  $q_{r-R/(4Lk)} \leq q_{hi-R/(2L)} = 0$  (recall that  $q_r$  is non-increasing in r, by Fact 3), so  $M_1 = \hat{M}$ , and by the induction hypothesis on  $X_1$ ,  $|\mathbf{Sur}(X)| = |\mathbf{Sur}(X_1)| \geq |M_1|^{1-1/k} = m^{1-1/k}$ , as required.
- 4. Case 4:  $w_{mid} > m/2$  and  $q_{hi-R/(2L)} > 0$ . In this case we pick  $a,b \in [lo,hi]$  so that a-b=R/(2L) and

$$q_a > q_b^2/m (5)$$

and also choose  $r \in [b,a]$  such that  $q_{r-\frac{b-a}{2k}} \leq q_{r+\frac{b-a}{2k}} \cdot \left(\frac{q_b}{q_a}\right)^{1/k}$ . In this case when we cut "backwards", we shift the responsibility for the vertices unmarked by the creation of  $X_1$  to  $Y_1$ . This is captured by defining  $\operatorname{Res}(Y_1) = \hat{M} \setminus M_1$ . Since  $|M'| = q_{r+\frac{a-b}{2k}}$  and  $|\operatorname{Res}(Y_1)| = q_{r-\frac{a-b}{2k}}$  we have

$$|M'| \ge |\mathbf{Res}(Y_1)| \cdot \left(\frac{q_a}{q_b}\right)^{1/k}$$
 (6)

By the induction hypothesis on  $Y_1$  we have that

$$\begin{aligned} |\mathbf{Sur}(Y_1)| &\geq \frac{|M'|}{|M'|^{1/k}} \overset{(6)}{\geq} |\mathbf{Res}(Y_1)| \cdot \left(\frac{q_a}{|M'| \cdot q_b}\right)^{1/k} \\ &\stackrel{(5)}{\geq} |\mathbf{Res}(Y_1)| \cdot \left(\frac{q_b}{m \cdot |M'|}\right)^{1/k} \\ &\geq \frac{|\mathbf{Res}(Y_1)|}{m^{1/k}} \ , \end{aligned}$$

where in the last inequality we use that  $|M'| = q_{r+(a-b)/(2k)} \leq q_b$ . Now by the induction hypothesis on  $X_1$  we get

$$\begin{split} |\mathbf{Sur}(X)| &= |\mathbf{Sur}(Y_1)| + |\mathbf{Sur}(X_1)| \\ &\geq \frac{|\mathbf{Res}(Y_1)|}{m^{1/k}} + |M_1|^{1-1/k} \\ &\geq \frac{|\mathbf{Res}(Y_1)| + |M_1|}{m^{1/k}} = \frac{|\hat{M}|}{m^{1/k}} = m^{1-1/k} \;. \end{split}$$

From Lemma 9 and Lemma 10 we derive the following theorem.

**Theorem 6.** Let G = (V, E) be a weighted graph, fix a set  $M \subseteq V$  of size m and a parameter  $k \ge 1$ . There exists a spanning tree T of G, and a set  $\mathbf{Sur}(M) \subseteq M$  of size at least  $m^{1-1/k}$ , such that for every  $v \in \mathbf{Sur}(M)$  and every  $u \in V$  it holds that  $\mathbf{dist}(v, u, T) \le O(k \log \log m) \cdot \mathbf{dist}(v, u, G)$ .

We conclude with the proof of our main results.

*Proof of Theorem 1.* The theorem follows directly from Theorem 6 by choosing M = V.

Proof of Theorem 2. Let  $M_1 = V$ , and for  $i \geq 1$  define  $M_{i+1} = M_i \setminus \mathbf{Sur}(M_i)$ . We shall apply Theorem 6 iteratively, where  $M_i$  is the set of vertices given as input to the i-th iteration, that has size  $|M_i| = m_i$ . Let  $T_i$  be the tree created in iteration i. By Theorem 6 the sizes  $m_1, m_2, \ldots$  obey the recurrence  $m_1 = n$  and  $m_{i+1} \leq m_i - m_i^{1-1/k}$ , which implies that after  $k \cdot n^{1/k}$  iterations we will have  $m_{k \cdot n^{1/k} + 1} < 1$  (see [MN07, Lemma 4.2]), and thus every vertex is in  $\mathbf{Sur}(M_i)$  for some  $1 \leq i \leq k \cdot n^{1/k}$ . For each  $v \in V$ , let  $\mathbf{home}(v)$  be the tree  $T_i$  such that  $v \in \mathbf{Sur}(M_i)$ .

# 4.5 Routing with Short Labels

In this section we prove Theorem 4. We first use a result of [TZ01b] concerning routing in trees.

**Theorem 7** ([TZ01b]). For any tree T = (V, E) (where |V| = n), and integer parameter b, there is a routing scheme with stretch 1 that has routing tables of size O(b) and labels of size  $(1+o(1))\log_b n$ . The decision time in each vertex is O(1).

Combining Theorem 2 and Theorem 7 we can construct a routing scheme. Let  $\mathcal{T}$  be the set of trees from Theorem 2. Each tree  $T \in \mathcal{T}$  is associated with a routing scheme given by Theorem 7. Set  $L_T(x)$  be the label of the vertex x in the routing scheme of the tree T.

In our scheme, the routing table of each vertex will be the collection of its routing tables in all the trees in  $\mathcal{T}$ . Hence the table size is  $O(b) \cdot |\mathcal{T}| = O(k \cdot b \cdot n^{1/k})$ . The label of each  $x \in V$  will be  $(\mathbf{home}(x), L_{\mathbf{home}(x)}(x))$ , i.e., the name of the home tree of x and the label of x in that tree. The label size is  $1 + (1 + o(1)) \log_b n = (1 + o(1)) \log_b n$ .

The routing is done in a straightforward manner, to route from y to x, we extract  $\mathbf{home}(x)$  from the given label of x, and simply use the routing scheme of the tree  $\mathbf{home}(x)$ . Note that this process takes O(1) time, and is independent of the routing path traversed so far. Since all vertices store in their routing table the appropriate routing information for  $\mathbf{home}(x)$ , the routing can be completed.

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Distance Oracle	Stretch	Size	Query time	Is deterministic?
[TZ01a]	2k-1	$O(k \cdot n^{1+1/k})$	O(k)	no
[MN07]	128k	$O(n^{1+1/k})$	O(1)	no
[WN13]	$(2+\epsilon)k$	$O(k \cdot n^{1+1/k})$	$O(1/\epsilon)$	no
[Che14]	2k-1	$O(k \cdot n^{1+1/k})$	O(1)	no
[Che15]	2k-1	$O(n^{1+1/k})$	O(1)	no
[RTZ05]	2k-1	$O(k \cdot n^{1+1/k})$	O(k)	yes
[WN13]	2k-1	$O(k \cdot n^{1+1/k})$	$O(\log k)$	yes
This paper	$8(1+\epsilon)k$	$O(n^{1+1/k})$	$O(1/\epsilon)$	yes
This paper	2k-1	$O(k \cdot n^{1+1/k})$	O(1)	yes

Table 1: Different distance oracles

# A Proof of Correctness for Algorithm 4

In this section we prove that the choices made in the create-petal procedure are all legal. In all the Lemmas that follow, we shall use the notation in Algorithm 4.

**Lemma 11.** If  $w_{mid} \leq \frac{m}{2}$  and  $w_{lo+\frac{R}{2L}} \geq 1$ , then there is  $[a,b] \subseteq [lo,mid]$  such that  $b-a=\frac{R}{2L}$  and  $w_a \geq w_b^2/m$ .

*Proof.* Seeking contradiction, assume that for every such a, b with  $b - a = \frac{R}{2L}$  it holds that  $w_b > \sqrt{m \cdot w_a}$ . Applying this on  $b = mid - \frac{iR}{2L}$  and  $a = mid - \frac{(i+1)R}{2L}$  for every  $i = 0, 1, \dots, L-2$ , we have that

$$w_{mid} > m^{1/2} \cdot w_{mid - \frac{R}{2L}}^{1/2} > \dots > m^{1 - 2^{-(L-1)}} \cdot w_{mid - \frac{(L-1)R}{2L}}^{2^{-(L-1)}} \ge m \cdot 2^{-1} \cdot w_{lo + \frac{R}{2L}}^{1/(2 \log m)} \ge \frac{m}{2} ,$$

where we used that  $1 + \log \log m \le L \le 2 + \log \log m$  and mid = lo + R/2. In the last inequality we also used that  $w_a \ge 1$ , which follows since  $b = a + \frac{R}{2L} \ge lo + \frac{R}{2L}$ , thus  $w_b \ge 1$ , and in particular  $w_a \ge w_b^2/m > 0$ . The contradiction follows.

**Lemma 12.** There is  $r \in [a,b]$  such that  $w_{r+\frac{b-a}{2k}} \leq w_{r-\frac{b-a}{2k}} \cdot \left(\frac{w_b}{w_a}\right)^{\frac{1}{k}}$ .

*Proof.* Seeking contradiction, assume there is no such choice of r, then applying this for  $r = b - (i + 1/2) \cdot \frac{b-a}{k}$  for  $i = 0, 1, \dots, k-1$  we get

$$w_b > w_{b-\frac{b-a}{k}} \cdot \left(\frac{w_b}{w_a}\right)^{1/k} > \dots > w_{b-k \cdot \frac{b-a}{k}} \cdot \left(\frac{w_b}{w_a}\right)^{k/k} = w_a \cdot \frac{w_b}{w_a} = w_b ,$$

a contradiction.  $\Box$ 

The following two lemmas are symmetric to the two lemmas above.

**Lemma 13.** If  $w_{mid} > \frac{m}{2}$  (implies  $q_{mid} \leq \frac{m}{2}$ ) and  $q_{hi-\frac{R}{2L}} \geq 1$ , then there is  $[b,a] \subseteq [mid,hi]$  such that  $a-b=\frac{R}{2L}$  and  $q_a \geq q_b^2/m$ .

**Lemma 14.** There is  $r \in [b,a]$  such that  $q_{r-\frac{a-b}{2k}} \leq q_{r+\frac{a-b}{2k}} \cdot \left(\frac{q_b}{q_a}\right)^{1/k}$ .