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Approximate Distance Oracles for Unweighted Graphs in Expected $O(n^2)$ time *

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Abstract

Let $G = (V, E)$ be an undirected graph on n vertices, and let $\delta(u, v)$ denote the distance in G between two vertices u and v. Thorup and Zwick showed that for any +ve integer t, the graph G can be preprocessed to build a data-structure that can efficiently report t-approximate distance between any pair of vertices. That is, for any $u, v \in V$, the distance reported $\hat{\delta}(u, v)$ satisfies

$$\delta(u, v) \leq \hat{\delta}(u, v) \leq t\delta(u, v)$$

The remarkable feature of this data-structure is that, for $t > 2$, it occupies sub-quadratic space, i.e., it does not store all-pairs distances information explicitly, and still it can answer any t-approximate distance query in constant time. They named the data-structure "oracle" because of this feature. Furthermore the tradeoff between stretch t and the size of the data-structure is essentially optimal.

In this paper we show that we can actually construct approximate distance oracles in expected $O(n^2)$ time if the graph is unweighted. One of the new ideas used in the improved algorithm also leads to the first linear time algorithm for computing an optimal size $(2, 1)$-spanner of an unweighted graph.

1 Introduction

The all-pairs shortest paths problem is one of the most fundamental algorithmic graph problem. Every computer scientist is aware of this classical problem right from the days he did his first course on algorithms. This problem is phrased most commonly as follows : *Given a graph $G(V, E)$ on $n(= |V|)$ vertices and $m(= |E|)$ edges, compute shortest-paths/distances between each pair of vertices.*

In many applications the aim is not to compute *all* distances, but to have a mechanism (data-structure) through which we can extract distance/shortest-path for any pair of vertices efficiently. Therefore, the following is a useful alternate formulation of the APSP problem.

Preprocess a given graph efficiently to build a data-structure that can answer a shortest-path query or a distance query for any pair of vertices.

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Throughout this paper, we stick to the above formulation of APSP. The objective is to construct a data-structure for this problem such that it is efficient both in terms of the space and the preprocessing time. There is a lower bound of $\Omega(n^2)$ on the space requirement of any data-structure for APSP problem, and space requirement of all the existing algorithms for APSP match this bound. However, there is a huge gap in the preprocessing time. In its most generic version, that is, for directed graph with real edge-weights, the best known algorithm [13] for APSP requires $O(mn + n^2 \log\log n)$ time. However, for graphs with $m = \theta(n^2)$, this algorithm has a running time of $\theta(n^3)$ which matches that of the old and classical algorithm of Floyd and Warshal. In fact the best known upper bound on the worst case time complexity of this problem is $O(n^3/\log n)$ due to Chan [6], which is marginally sub-cubic. Surprisingly, despite the fundamental nature of the problem, the existing lower bound on its time complexity is the trivial lower bound of $\Omega(n^2)$. This has motivated the researchers to explore ways to achieve sub-cubic preprocessing time and/or sub-quadratic space data-structures that report *approximate* instead of exact shortest-paths/distances.

A data-structure is said to compute t-approximate distances for all-pairs of vertices, if for any pair of vertices $u, v \in V$, the distance $\hat{\delta}(u, v)$, that it would report, satisfies

$$\delta(u, v) \leq \hat{\delta}(u, v) \leq t\delta(u, v)$$

In the last ten years, many novel algorithms have been designed for all-pairs approximate shortest paths (APASP) problem that achieve sub-cubic running time and/or sub-quadratic space. The *approximate distance oracle* designed by Thorup and Zwick [15] is a milestone in this area. They show that any given weighted undirected graph can be preprocessed in sub-cubic time to build a data-structure of sub-quadratic size for answering a distance query with stretch 3 or more. Note that 3 is also the least stretch for which we can achieve sub-quadratic space for APASP (see [8]). There are two very impressive features of their data-structure. First, the trade off between stretch and the size of data-structure is essentially optimal and second, inspite of its sub-quadratic size their data-structure can answer any distance query in constant time, hence the name "oracle". In precise words, Thorup and Zwick achieve the following result.

Theorem 1.1 *[15] For any integer $k \geq 1$, and a given undirected weighted graph on n vertices and m edges, it takes expected $O(kmn^{1/k})$ time to build a data-structure of size $O(kn^{1+1/k})$ that can answer any $(2k-1)$ approximate distance query in $O(k)$ time.*

As mentioned in [15] the oracle model for shortest paths has been considered in the past also, at least implicitly, by Awerbuch *et al.* [2], Cohen [7] and by Dor *et al.* [9]. However, the approximate distance oracle of Thorup and Zwick significantly improves all these previous results on oracles for shortest paths. Having achieved essentially optimal query time as well as the space requirement for approximate distance oracles, the only aspect that can be potentially improved is the preprocessing time. Currently, the expected preprocessing time of the $(2k-1)$-approximate distance oracle is $O(mn^{1/k})$ which is certainly sub-cubic. Thorup and Zwick posed the following question : *Can a $(2k-1)$-approximate distance oracle be computed in $O(n^2)$ time?*

In this paper, we answer their question in affirmative for unweighted graphs. The following is the main result of our paper.

Theorem 1.2 *An undirected unweighted graph can be preprocessed in expected $O(n^2)$ time to construct its $(2k-1)$-approximate distance oracle of size $O(kn^{1+1/k})$ for any integer $k > 1$.*

1.1 Summary of related work on APASP

The algorithms for all-pairs approximate shortest paths fall under two categories, depending upon whether the error in the distance is additive or multiplicative. The algorithm that computes all-pairs t-approximate

shortest paths, as defined earlier, is actually an algorithm that achieves a multiplicative error t. An algorithm is said to report distances with additive error k, if for any pair of vertices $u, v \in V$, the distance, that it reports, is at-least $\delta(u,v)$ and at most $\delta(u,v) + k$. The first algorithm for reporting distances with additive error was given by Aingworth *et al.* [1]. Their algorithm reports distances with additive error 2. Dor *et al.* [9] improve and extend this algorithm for arbitrary additive error. Their algorithm requires $O(kn^{2-\frac{1}{k}}m^{\frac{1}{k}}\operatorname{polylog} n)$ time to report distances with additive error $2(k-1)$ for any pair of vertices. They also showed that all-pairs 3-approximate shortest paths can be computed in $O(n^2 \operatorname{polylog} n)$ time. Cohen and Zwick [8] later extended this result to weighted graphs. There are also some algorithms that achieve multiplicative as well as additive errors simultaneously. Elkin [10] presented the first such algorithm. For unweighted graphs, given arbitrarily small $\zeta, \epsilon, \rho > 0$, Elkin's algorithm requires $O(mn^{\rho} + n^{2+\zeta})$ time, and for any pair of vertices $u, v \in V$, reports distance $\hat{\delta}(u,v)$ satisfying the inequality :$\delta(u,v) \leq \hat{\delta}(u,v) \leq (1+\epsilon)\delta(u,v) + \beta$, where β is a function of ζ, ϵ, ρ. If the two vertices $u, v \in V$ are separated by sufficiently long distances in the graph, the stretch $\frac{\hat{\delta}(u,v)}{\delta(u,v)}$ ensured by Elkin's algorithm is quite close to $(1+\epsilon)$. But the stretch factor may be quite huge for short paths. This is because β depends on ζ as $(1/\zeta)^{\log 1/\zeta}$, depends inverse exponentially on ρ and inverse polynomially on ϵ. The output of all the algorithms mentioned above is an $n \times n$ matrix that stores pairwise approximate distance for all the vertices explicitly to answer each distance query in constant time. Recently Baswana *et al.* [3] presented a simple algorithm to compute *nearly* 2-approximate distances for unweighted graphs in expected $O(n^2 \operatorname{polylog} n)$ time. They essentially extend the idea of the 3-approximate distance oracle of Thorup and Zwick [15] to achieve stretch 2 at the expense of introducing a very small additive error (less than 3). All the algorithms for reporting approximate distances that we mentioned above, including the new results of this paper, can also report the corresponding approximate shortest path also, and the time for doing so is proportional to the number of edges in the approximate shortest-path.

1.2 Organization of the paper

Our starting point will be the 3-approximate distance oracle of Thorup and Zwick [15]. The following section explains notations, definitions and lemmas, most of them adapted from [15]. In section 3, we provide a sketch of the existing 3-approximate distance oracle, and identify the most time consuming Steps in their preprocessing algorithm. Subsequently we explore ways to execute them in expected $O(n^2)$ time. We succeed in our goal by providing a tighter analysis of one of the Steps and by performing small but crucial changes in the remaining ones. An important tool used by our preprocessing algorithm is is a special kind of a $(2,1)$-spanner. For an unweighted graph $G = (V,E)$, a sub-graph $G = (V, E_S)$, $E_S \subset E$ is said to be an (α, β)-spanner if for each pair of vertices $u, v \in V$, the distance between the two in the sub-graph is at most $\alpha\delta(u,v) + \beta$. In section 4, we present a *parameterized* $(2,1)$-spanner and explain how its special features are helpful in improving the preprocessing time of the oracle without increasing the stretch. Finally in section 5, we present and analyze our new 3-approximate distance oracle. In a straight forward manner, the techniques used in our 3-approximate distance oracle lead to the expected $O(n^2)$ time preprocessing algorithm for $(2k-1)$-approximate distance oracle in section 6.

A second contribution of the paper, which is important in its own right, is the first expected linear time algorithm for computing a $(2,1)$-spanner of (optimal) size $O(n^{3/2})$. The previous linear time algorithms [4, 12] compute spanners of stretch three or more.

2 Preliminaries

For a given undirected graph $G = (V, E)$, we define the following notations.

- $\delta(u, v)$: the distance between vertex u and vertex v in the graph.
- $\delta(v, Y)$: $\min_{y \in Y} \delta(v, y)$
- $p(v, Y)$: the vertex from the set Y which is nearest to v, that is, at distance $\delta(v, Y)$ from v. In case there are multiple vertices from Y at distance $\delta(v, Y)$, we break the tie arbitrarily to ensure a unique $p(v, Y)$. Moreover, if $v \in Y$, we define $p(v, Y) = v$.

In this paper we deal with unweighted graphs only. It is straightforward to observe that a single source shortest path tree rooted at a vertex $v \in V$ in an unweighted graph is same as the breadth-first-search (BFS) tree rooted at v which can be computed in just $O(m)$ time.

Given a set $Y \subset V$, it is also quite easy to compute $p(v, Y)$ for all $v \in V$ in $O(m)$ time as follows. Connect a dummy vertex ω to all the vertices of set Y, and perform a BFS traversal on the graph starting from ω. Thus the following lemma holds.

Lemma 2.1 *Given a set $Y \subset V$, it requires $O(m)$ time to compute $p(v, Y)$ for all vertices $v \in V$.*

2.1 Ball around a vertex

An important construct of the approximate distance oracles is a $Ball$ which is defined as follows.

Definition 2.1 *For a vertex v in a graph $G = (V, E)$, and subsets X, Y of vertices with $Y \subset X$, the set $Ball(v, X, Y)$ of vertices is defined as*

$$\text{Ball}(v, X, Y) = \{x \in X | \delta(v, x) < \delta(v, Y)\}$$

In simple words, $Ball(v, X, Y)$ consists of all those vertices of the set X whose distance from v is less than the distance from $p(v, Y)$ to v. As a simple observation, it can be noted that $Ball(v, X, \emptyset)$ is the set X itself, whereas $Ball(v, X, X) = \emptyset$.

The following Lemma from [15] suggests that if the set Y is formed by sampling vertices from the set X with suitable probability, the expected size of $Ball(u, X, Y)$ will be sublinear in terms of $|X|$. This observation plays the key role in achieving sub-quadratic bound on the size of the approximate distance oracles of [15]. These oracles basically store distances to a hierarchy of Balls around each vertex.

Lemma 2.2 *[15] Given a graph $G = (V, E)$, let the set Y be formed by picking each vertex of a set $X \subset V$ independently with probability p. The expected size of $Ball(v, X, Y)$ is bounded by $1/p$.*

Proof: Consider the sequence $(x_1, x_2, \cdots)$ of vertices of set X arranged in non-decreasing order of their distance from v. The vertex x_i will belong to $Ball(v, X, Y)$ only if none of $x_1, \cdots, x_i$ are selected for the set Y. Therefore, $x_i \in Ball(v, X, Y)$ with probability at most $(1 - p)^i$. Hence the expected number of vertices in $Ball(v, X, Y)$ would be bounded by $\sum_i (1 - p)^i = 1/p$. □

In order to answer an approximate distance query in constant time, $Ball(v, X, Y)$ can be kept in a hash table [11] so that one can determine in worst case constant time whether or not $x \in Ball(v, X, Y)$, and if so, report the distance $\delta(v, x)$. Once we have the set $Ball(v, X, Y)$, the preprocessing time as well as the space requirement of the corresponding hash-table would be of the order of $|Ball(v, X, Y)|$.

2.2 Computing Balls efficiently

Thorup and Zwick presented a very efficient way to compute Balls wherein for each vertex $x \in X$, we compute all those vertices $v \in V$ (and their distance from x) whose *Ball*(v, X, Y) contains x. The key observation is the following Lemma.

Lemma 2.3 *[15] If a vertex $x \in X$ belongs to $Ball(v, X, Y)$, then for every vertex u lying on a shortest path between x and v, the vertex x will also belong to $Ball(u, X, Y)$.*

Proof: The proof is by contradiction. Given that x belongs to $Ball(v, X, Y)$, let u be a vertex lying on a shortest path between v and x. The vertex x would not belong to $Ball(u, X, Y)$ if and only if there is some vertex $w \in Y$ such that $\delta(u, w) \leq \delta(u, x)$. On the other hand, using triangle inequality it follows that $\delta(v, w) \leq \delta(v, u) + \delta(u, w)$. Therefore, $\delta(u, w) \leq \delta(u, x)$ would imply that $\delta(v, w) \leq \delta(v, x)$. In other words, for the vertex v, the vertex $w \in Y$ is not farther than the vertex x. Hence x does not belong to $Ball(v, X, Y)$ (a contradiction !). □

It follows from Lemma 2.3 that if we look at a shortest-path tree rooted at a vertex $x \in X$, the set of all those vertices $v \in V$ satisfying the condition "$x \in Ball(v, X, Y)$" appears as truncated sub-tree rooted at v. This implies that, for a vertex $x \in X$, computing the set of vertices $v \in V$ satisfying "$x \in Ball(v, X, Y)$", amounts to performing a restricted BFS traversal from x, wherein we have to confine the BFS traversal within this sub-tree only. The algorithm is described below.

Restrict_BFS(x, Y)

Initially all vertices are unvisited.
Visited$(x) \leftarrow$`true`; `Enqueue`(Q, x)
`while` Q `not_empty`
  $w \leftarrow$ `Dequeue`(Q)
  `For each neighbor` w `of` v
  `If not` *Visited*(v) `and` $(\delta(x, w) + 1 < \delta(v, p(v, Y)))$
    Visited$(v) \leftarrow$`true`
    $\delta(x, v) \leftarrow \delta(x, w) + 1$
    `Enqueue`(Q, v)

We can now state the following Lemma whose proof is immediate.

Lemma 2.4 *The algorithm Restrict_BFS(x, Y) begins with exploring the adjacency list of x, and afterwards, it explores the adjacency list of only those vertices v that satisfy "$x \in Ball(v, X, Y)$".*

In order to compute $Ball(v, X, V), \forall v \in V$, it suffices to perform *Restrict_BFS*(x, Y) for all $x \in X$. The running time of the algorithm is of the order of the number of edges explored. For concise analysis, let us associate the cost of exploring an edge to the endpoint of the edge that explores it. In this way, the total time incurred in the computation of $Ball(v, X, Y) \forall v \in V$ is of the order of $\sum_{x \in X} deg(x) + \sum C_v$ where C_v is the amount of computation cost (time) assigned to v. Furthermore, it follows from Lemma 2.4 that $C_v = \sum_{x \in Ball(v, X, Y)} deg(v)$. We now state the following Lemma that would play a crucial role in the analysis of the running time of the new algorithm for approximate distance oracles.

Lemma 2.5 *[15] Given a graph $G = (V, E)$, and sets $X \subset Y$, we can compute $Ball(v, X, Y), \forall v \in V$ by just performing Restrict_BFS(x, Y) on each $x \in X$. Furthermore, Restrict_BFS(x, Y) incurs a computation cost of $\deg(v)$ for a vertex v if $x \in Ball(v, X, Y)$ and nil otherwise.*

3 The 3-approximate distance oracle of Thorup and Zwick

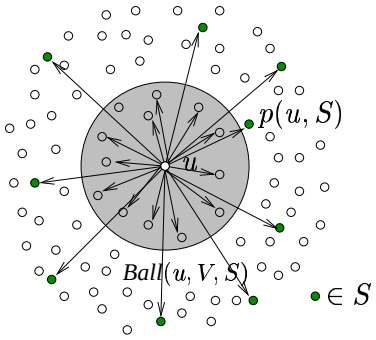


Figure 1: For u, we store distance to the vertices pointed by arrows.

Let $G = (V, E)$ be an undirected unweighted graph. The 3-approximate distance oracle of Thorup and Zwick can be constructed as follows :

Let $S \subset V$ be a set formed by selecting each vertex independently with probability $n^{-1/2}$. For each vertex $v \in V$, compute $p(v, S)$ and the distances to vertices of set $Ball(v, V, S)$. In addition, compute distances $\delta(v, x)$ for every $x \in S$ (see Figure 1).

The final data-structure would keep, for each vertex $v \in V$, the nearest vertex $p(v, S)$ and two hash tables : one hash table storing the distances to vertices of $Ball(u, V, S)$, and another hash table storing the distances to all vertices of the set S. Using these hash-tables, any distance query can be answered as follows.

Answering a distance query :

Let $u, v \in V$ be any two vertices whose approximate distance is to be computed. First it is determined whether or not $u \in Ball(v, V, S)$, and if so, the exact distance $\delta(u, v)$ is reported. Note that $u \notin Ball(v, V, S)$ would imply $\delta(v, p(v, S)) \leq \delta(v, u)$. In this case, we report the distance $\delta(v, p(v, S)) + \delta(u, p(v, S))$, which is at least $\delta(u, v)$ (triangle inequality) and upper-bounded by $3\delta(u, v)$ as shown below.

$$\begin{aligned} \delta(v, p(v, S)) + \delta(u, p(v, S)) &\leq \delta(v, p(v, S)) + (\delta(u, v) + \delta(v, p(v, S))) \text{ (using triangle inequality)} \\ &= 2\delta(v, p(v, S)) + \delta(u, v) \\ &\leq 2\delta(u, v) + \delta(u, v) = 3\delta(u, v) \quad \{\text{since } \delta(v, p(v, S)) \leq \delta(v, u)\} \end{aligned}$$

It follows from Lemma 2.2 that the expected size of $Ball(v, V, S)$ would be $\sqrt{n}$, and hence the total expected size of the data-structure would be $O(n^{3/2})$. Thorup and Zwick showed that it takes $O(m\sqrt{n})$ preprocessing time to build 3-approximate distance oracle for an undirected graph.

3.1 Ideas for Improving the Preprocessing time for Computing the Oracle

Let us explore ways to improve the preprocessing time for 3-approximate distance oracle. There are two main computational tasks involved in building the 3-approximate distance oracle. The first task is the computation of *Balls*() and the second task is the computation of distances from sampled vertices to all the vertices in the graph. Note that there is an additional task of computing $p(v, S)$ for every $v \in V$. But that is quite easily computable in overall $O(m)$ time as mentioned in Lemma 2.1.

In order to compute *Balls*(), Thorup and Zwick build the *Restricted_BFS* trees from the vertices of the set S, and show that the expected time required is $O(m\sqrt{n})$. For dense graphs, the expected time may be as large as $\theta(n^{2.5})$. We provide a tighter analysis of the same algorithm and show that the expected time is bounded by $O(n^2)$. The key idea here is to notice that we charge *deg*(v) units of computation cost to a vertex only if *Ball*(v) is non-empty. Since the vertices are sampled randomly, the heavy degree vertices are likely to have empty Ball around them, and so the computational cost for building their Ball would be nil. This crucial observation was missing in the analysis of Thorup and Zwick.

It is less obvious to perform the second task in $O(n^2)$ time. The second task that computes BFS trees from vertices of set S would require $O(m|S|)$ time, which is certainly not $O(n^2)$ when the graph is dense. To improve its preprocessing time to $O(n^2)$ when the given graph is dense, one approach would be to perform

BFS traversal from vertices of the set S on a sparse sub-graph which, inspite of its sparseness, preserves pairwise (approximate) distances too. Such a subgraph is called a *spanner*.

Definition 3.1 *For a graph $G = (V, E)$, and two real numbers $\alpha \geq 1, \beta \geq 0$, a subgraph $G = (V, E_S)$, $E_S \subset E$ is said to be an (α, β)-spanner if for all $u, v \in V$, the distance $\delta_S(u, v)$ in the spanner satisfies $\delta(u, v) \leq \delta_S(u, v) \leq \alpha\delta(u, v) + \beta$.*

Note that the sparsity of a spanner comes along with the stretching of the distances in the graph. So one has to be careful in employing an (α, β)-spanner (with $\alpha > 1$) in the second step, lest one should end up computing a $(\alpha + 2)$-approximate distance oracle [1] instead of a 3-approximate distance oracles. To explore the possibility of using a spanner in our algorithm, let us carefully revisit the distance reporting scheme of the 3-approximate distance oracle of Thorup and Zwick. For a pair of vertices $u, v \in V$, the distance reported is exact if $u \in Ball(v, V, S)$. So let us analyse carefully the case when u lies outside *Ball*(v, V, S).

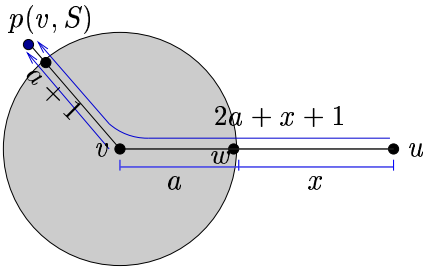


Figure 2: Closely analysing the distance reported by the exisiting oracle of [15].

Let us partition a shortest path between v and u into two sub-paths : the sub-path of lengths a covered by $Ball(v, V, S)$, and the sub-path of length x lying outside the ball. With this partition, the distance $\delta(v, p(v, S))$ would be $a + 1$. Now considering the path between $p(v, S)$ and u that passes through v (see Figure 2), we can conclude that the distance $\delta(u, p(v, S))$ would be bounded by $2a + 1 + x$. Consequently, the distance "$\delta(v, p(v, S)) + \delta(u, p(v, S))$" as reported by 3-approximate distance oracle of Thorup and Zwick is bounded by $3a + x + 2$.

A comparison of this upper bound of $3a + x + 2$ on the distance reported with the actual distance $\delta(u, v) = a + x$ suggests that there is a possibility of stretching the sub-path (of length x) uncovered by *Ball*(v, V, S) by a factor of 3 and still keeping the distance reported to be 3-approximate. So we may employ a spanner for the second task of preprocessing algorithm if for each vertex $v \in V$, the shortest paths from v to all the vertices of *Ball*$(v, V, S) \cup \{p(v, S)\}$ are preserved in the spanner. The reader must note that these features are required in addition to the $O(n^{3/2})$ size and suitable stretch (α, β) that the spanner must have. We introduce a special kind of spanner, called 'parameterized spanner' that has these additional features.

The rest of the paper is organised as follows. In the following section we defi ne a parameterized $(2, 1)$-spannner and present a linear time algorithm for this. The concept and the linear time algorithm for parameterized spanner is of independent interest. However, the reader who is interested in the parameterized spanner as a tool for effi cient computation of approximate distance oracles, requires the features of parameterized spanner mentioned in Lemma 4.1 and Theorem 4.2 only. In section 5, we describe the expected $O(n^2)$ time algorithm for 3-approximate distance oracles. We extend it to $(2k - 1)$-approximate distance oracles in section 6.

4 A parameterized $(2, 1)$-spanner

Definition 4.1 *Given a graph $G = (V, E)$, and a parameter $S \subset V$, a subgraph $G(V, E_S)$ is said to be a parameterized $(2, 1)$-spanner with respect to S if*
(i) $G = (V, E_S)$ is a $(2, 1)$-spanner.

[1] by replacing $\delta(u, p(u, S))$ with $\alpha \cdot \delta(u, p(u, S)$ in the query bound of approximate distance oracle

(ii) If a vertex has no neighbor in set S, then all the edges adjacent to it are present in the parameterized spanner. [2]
(iii) All edges incident to S are present in the spanner.

Let $G = (V, E)$ be the given undirected unweighted graph and S be a subset of vertices from the graph. We shall now present an algorithm that computes a parameterized $(2, 1)$-spanner $G = (V, E_S)$ with respect to S. The algorithm starts with empty spanner, i.e., $E_S = \emptyset$, and adds edges to E_S in the following three steps.

1. *Forming the clusters* :
   We form clusters (of vertices) around the vertices of the set S. Initially the clusters are $\{\{u\}|u \in S\}$. Each $u \in S$ will be referred to as the *center* of its cluster. We process each vertex $v \in V \backslash S$ as follows.

   (a) If v is not adjacent to any vertex in S, we add every edge incident on v to the set E_S.

   (b) If v is adjacent to some vertex $o \in S$, assign v to the cluster centered at $o \in S$ and add the edge $e(v, o)$ to the set E_S. In case there is more than one vertex in S adjacent to v, break ties arbitrarily [3].

   This is the end of the first step of the algorithm. At this stage let E' denote the set of all the edges of the given graph excluding the edges added to the spanner and all the intra-cluster edges. Let V' be the set of vertices corresponding to the end-points of the edges E'. It is evident that each vertex of set V' is either a vertex from S or adjacent to some vertex from S, and the first phase has partitioned V' into disjoint clusters each centered around some vertex from S. The graph $G = (V', E')$ and the corresponding clustering is passed onto the second step.

2. *Adding edges between vertices and clusters* :
   For each vertex $v \in V'$, we group all its neighbors into their respective clusters. There will be at most $|S|$ neighboring clusters of v. For each cluster adjacent to vertex v, we select an (arbitrary) edge from the set of edges between (the vertices of) the cluster and v, and add it to the set E_S.

3. *Adding edges incident on S* :
   Finally, we also add to the set E_S all the edges incident on vertices of S, that did not get added in the two steps given above to complete the construction of the spanner.

Note that the steps 1(a) and 3 ensure that the final spanner is a parameterized spanner with respect to S (refer to Definition 4.1). It is easy to verify that the implementation of the two steps of the algorithm merely requires two traversals of the adjacency list of each vertex. Thus the running time of the algorithm is $O(m)$.

Lemma 4.1 *If $e(u, v)$ is an edge not present in the parameterized spanner $G = (V, E_S)$ as constructed above, there is a one-edge or a two-edge path in the spanner between the vertex u and the center of the cluster containing vertex v, and vice versa.*

Proof: There are two cases depending upon whether u and v belong to same cluster or different clusters.
Case 1 : *(u and v belong to same cluster)*
Let the cluster centered at $o \in S$ contain both the vertices u and v. It follows from step 1(b) of the algorithm that the edges $e(u, o)$ and $e(v, o)$ are present in the spanner. So there is a one edge path from v to the center of the cluster containing u, and vice versa.

[2] Note that it ensures that any edge $e(u, v)$ for which either u or v is not adjacent to S, is present in the spanner.
[3] We will define it more precisely in the context of the actual construction of distance oracles

Case 2 : *(u and v belong to different clusters)*
Let x and y be the centers of the clusters containing u and v respectively. It follows that u is adjacent to the cluster centered at y, therefore, an edge between u and some vertex, say w, of this cluster has been added to the spanner in the second step of the algorithm. It follows from step 1(b) of the algorithm that the edge $e(w, y)$ is also present in the spanner. So there is a two-edge path $u - w - y$ between u and y. Using similar arguments, there is a two-edge path between v and x.

□

We can extend the above observation to $e(u, v) \in E_S$. Let $C(x)$ denote the vertex x if x does not belong to any cluster, otherwise let $C(x)$ denote the center of the cluster containing vertex x (a vertex that belongs to S is its own center). Note that a vertex x is at distance at most one from $C(x)$ implying the necessary result for $e(u, v) \in E_S$. Using the above observations, we shall now prove the following important Theorem.

Theorem 4.1 *For a graph $G = (V, E)$ and a subset $S \subset V$, the parameterized spanner $G(V, E_S)$ with respect to S computed by the algorithm above is a $(2, 1)$-spanner.*

Proof: In order to ensure that $\delta_S(u, w) \leq 2\delta(u, w) + 1$ in the subgraph $G(V, E_S)$, consider the sequence $s_{uw} : \{u(= v_0), v_1, \cdots, v_{j-1}, w(= v_j)\}$ of vertices of a shortest path between u and v in the order they appear on it in the original graph. Now consider another sequence $c_{uw} : \{v_0, C(v_1), v_2, C(v_3), \cdots\}$ formed by replacing each vertex v_{2i+1} of the sequence s_{uw} by $C(v_{2i+1})$, for $i \leq j/2$; It follows that each pair of consecutive vertices in the sequence c_{uw} is connected by a path of length at most two in the spanner. So there is a path of length at most $2j$ between u and the last vertex of the sequence c_{uw} in the spanner. Note that the last vertex of the sequence c_{uw} is w or $C(w)$ depending on whether the path length j is even or odd. If the path length is even, it implies that $\delta_S(u, w) \leq 2j$. Otherwise (j is odd) note that either $C(w)$ is the vertex w itself or there is an edge between $C(w)$ and w in the spanner $G = (V, E_S)$. Thus $\delta_S(u, v) \leq 2j+1$. Combining the two cases (even and odd j), we can conclude that $\delta_S(u, v) \leq 2\delta(u, v) + 1$.

□

Now let us analyze the size of the parameterized $(2, 1)$-spanner $G = (V, E_S)$ when the set S is formed by selecting each vertex independently with probability p. In the first step of the algorithm, the expected number of edges contributed by a vertex v is at most $1 + deg(v)(1 - p)^{deg(v)}$, which is bounded by $O(1/p)$. Hence the expected number of edges added to the spanner during the first phase is $O(n/p)$. In the second phase, the expected number of edges added to the spanner is no more than n^2p since we add at most one edge for each vertex-cluster pair. Hence we can state the following lemma.

Lemma 4.2 *Let a set S be formed by selecting each vertex independently with probability p, then the expected number of edges in the parameterized $(2, 1)$-spanner $G = (V, E_S)$ will be at most $n/p + n^2p$.*

As mentioned earlier, the running time of our algorithm is $O(m)$. We would repeat the algorithm if the size of the spanner is more than $c(n/p + n^2p)$ for any constant $c > 1$. The expected number of iterations is $O(1)$. Thus a $(2, 1)$ spanner of size $O(n/p + n^2p)$ can be computed in $O(m)$ expected time. Now we state the following theorem that would highlight the crucial role played by the parameterized $(2, 1)$-spanner in constructing approximate distance oracles in $O(n^2)$ time.

Theorem 4.2 *Let $G = (V, E)$ be a graph and let S be a set formed by selecting each vertex with probability p. There is an expected $O(m)$ time computable parameterized $(2, 1)$-spanner $G = (V, E_S)$ with the following properties.*

1. *The spanner preserves a shortest path from v to every vertex $u \in Ball(v, V, S) \bigcup \{p(v, S)\}$*
2. *The number of edges in the spanner is $O(n/p + n^2p)$.*

Proof: Let $v(=v_0), v_1, v_2, \cdots, v_{j-1}, u(=v_j)$ be a shortest path between v and any vertex $u \in Ball(v, V, S)$ in the graph $G = (V, E)$. Note that none of the vertices $v_i, i < j$ has any neighbor from the set S. Since the parameterized spanner would keep all those edges whose at least one end-point has no neighbor in the parameter vertex-set S, so all the edges incident on the vertices $v_i, i < j$ are present in the spanner. Hence a shortest path from v to every vertex in $Ball(u, V, S)$ is preserved.

We shall now show that a shortest path from v to $p(v, S)$ is also present in the parameterized $(2, 1)$-spanner. If $p(v, S)$ is adjacent to v in the graph, then the shortest path between the two vertices is just the edge $e(v, p(v, S))$. Since $p(v, S) \in S$ and we add all those edges to the spanner which are incident on S, the edge $e(v, p(v, S))$ is surely present in the parameterized spanner. So let us consider the case when $p(v, S)$ is not adjacent to v. In this case, $Ball(v, V, S) \neq \emptyset$, so a shortest path from v to $p(v, S)$ is a shortest path from v to some vertex, say v', lying at the boundary of $Ball(v, V, S)$ concatenated by the edge $e(v', p(v, S))$. Again, based on the arguments given above, the shortest path $v \rightsquigarrow v'$ as well as the edge $e(v', p(v, S))$ are present in the parameterized spanner, and we are done. □

If we construct the set S by selecting each vertex independently with probability $p = n^{-1/2}$, we get a bound of $O(n^{3/2})$ on the size of $(2, 1)$-spanner. Thorup and Zwick [15] show that $\theta(n^{3/2})$ bound on the size of (3,0)-spanner is indeed optimal. Since a $(2, 1)$-spanner is also a (3,0)-spanner, therefore, $\theta(n^{3/2})$ bound is optimal for $(2, 1)$-spanner too.

Theorem 4.3 *Given an unweighted undirected graph $G(V, E)$, a $(2, 1)$-spanner of size $O(n^{3/2})$ can be computed in expected $O(m)$ time.*

Note that although being of same size asymptotically, a $(2, 1)$-spanner is better than a 3-spanner since it achieves a better ratio of $\delta_S(u, v)/\delta(u, v)$ which is close to 2, for vertices separated by long distances.
Remark : The algorithm for a $(2, 1)$-spanner of expected size $O(n^{3/2})$ might appear similar to the algorithm for a 3-spanner in [4]. However, there is a subtle difference both in the construction and the analysis. The algorithm for a 3-spanner ensures that for each edge (u, v) not in the spanner, either there is a path from u to $C(v)$ of length at most two or a path from v to $C(u)$ of length at most two. But, as the reader may verify in Theorem 4.1, we require the existence of both these paths to prove that the spanner is a $(2, 1)$-spanner.

5 A 3-approximate distance oracle in expected $O(n^2)$ time

We now describe the algorithm for computing a 3-approximate distance oracle that would require expected $O(n^2)$ time. The preprocessing algorithm is essentially the same as that of [15] except that for every $v \in V$, we compute the distance to the vertices of sampled set in a parameterized $(2, 1)$-spanner rather than in the original graph.

Computing a 3-approximate distance oracle

Let $S \subset V$ contain each vertex of V independently with probability $= n^{-1/2}$.

1. Compute $p(v, S)$ and *Ball*(v, V, S) for each $v \in V$.
2. Compute a parameterized (2,1)-spanner with respect to set S. (see Note below)
3. Compute $\delta_S(u, x)$ for each $u \in V, x \in S$.

Note : While computing a $(2,1)$-spanner in the second step of the preprocessing algorithm, we shall adopt the following strategy : in case a vertex v is a neighbor of more than one vertex from the set S, instead of assigning it to the cluster centered at any arbitrarily picked neighbor from S, we shall assign v to the cluster centered at $p(v,S)$. As a result, Lemma 4.1 for the $(2,1)$-spanner can be restated as follows.

Lemma 5.1 *If an edge $e(u,v)$ is not present in the parameterized $(2,1)$-spanner $G=(V,E_S)$, there is a one-edge or a two-edge path in the spanner between the vertex $p(u,S)$ and the vertex v, and vice versa.*

In the next subsection, we shall show that the new oracle is indeed a 3-approximate distance oracle. But prior to that, we shall prove that the expected time to compute the oracle is $O(n^2)$ as follows.

There are three main steps in the construction of our oracle. It follows from Theorem 4.2 that we require expected $O(m)$ time to compute the parameterized $(2,1)$-spanner. So the second step requires expected $O(m)$ time. The third step is accomplished by performing BFS traversal from vertices of S to all the vertices in the parameterized $(2,1)$-spanner. This would require expected $O(\min(m,n^{3/2})\cdot|S|) = O(\min(m\sqrt{n},n^2))$ time since expected size of S is $\sqrt{n}$. Let us now analyze the computation time for the first Step. It follows from Lemma 2.1 that computing computing $p(v,S)$ for all $v\in V$ requires just $O(m)$ time in total. We shall now establish a bound on the expected time required to compute $Ball(v,V,S), \forall v\in V$.

Lemma 5.2 *The expected time for computing $Ball(v,V,S)\forall v\in V$ is at most $O(n^2)$.*

Proof: Recall from subsection 2.2 that we compute $Ball(v,V,S),\forall v\in V$ by executing the subroutine *Restrict_BFS*(x,S) on each $x\in V$. Following the Lemma 2.5, we just need to show that the expected computational cost charged to a vertex v during all these subroutines is bounded by $O(n)$.

Consider a vertex $v\in V$. Let $v_1,v_2,\cdots,v_{n-1}$ be the sequence of vertices arranged in non-decreasing order of their distances from v. It follows from Lemma 2.5 that *Restrict_BFS*(v_j,S) will charge *deg*(v) cost to v if $v_j\in Ball(v,V,S)$. Since set S is a random set of vertices, total computational cost charged to v is a random variable $W_v=\sum_j deg(v)\cdot X_j$, where $X_j=1$ if $v_j\in Ball(v,V,S)$ and 0 otherwise. Hence

$$\mathbf{E}[W_v]=\sum_j deg(v)\cdot \mathbf{Pr}[v_j\in Ball(v,V,S)] \quad \text{(using linearity of expectation)}$$

Now $v_j\in Ball(v,V,S)$ only if none of the vertices at distance at most $\delta(v,v_j)$ from v is present in S. Certainly there are at-least $\max(j,deg(v))$ such vertices. Furthermore, since each vertex is selected in the set S independently with probability $q=n^{-1/2}$,

$$\begin{aligned}
\mathbf{E}[W_v] &= \sum_{j\geq 1} deg(v)\cdot(1-q)^{\max(j,deg(v))} \\
&= \sum_{j\leq deg(v)} deg(v)\cdot(1-q)^{deg(v)} + \sum_{j>deg(v)} deg(v)\cdot(1-q)^j \\
&\leq (deg(v))^2\cdot(1-q)^{deg(v)} + \sum_{j>deg(v)} j\cdot(1-q)^j \\
&\leq (deg(v))^2\cdot(1-q)^{deg(v)} + 1/q^2 \quad \leq \quad 1/q^2+1/q^2 \qquad \{\text{Since } x^2(1-\alpha)^x\leq 1/\alpha^2, \forall\alpha>0\} \\
&= O(n) \qquad \{\text{since } q=n^{-1/2}\ \}
\end{aligned}$$

□

Hence the expected preprocessing time for the new approximate distance oracle is $O(n^2)$. The expected

size of the oracle is $\mathbf{E}[\sum_u |Ball(u,S)| + |S| \cdot n]$ which, using Lemma 2.2, is no more than $2n^{3/2}$. We shall repeat the algorithm if the size exceeds this bound by some constant, the expected number of repetitions will be $O(1)$. So we can state the following theorem.

Theorem 5.1 *The new approximate distance oracle occupies $O(n^{3/2})$ space and requires expected $O(n^2)$ preprocessing time.*

5.1 Reporting distance with stretch at most 3

We describe below the query answering procedure for the new approximate distance oracle. It differs from the algorithm of [15] only for the case when $u \notin Ball(v,V,S)$.

$\mathcal{Q}(u,v)$ *: Reporting approximate distance between u and v*
`If` $u \in Ball(v,V,S)$
  `return` $\delta(u,v)$
`Else report` **minimum** of
  $\delta(u,p(u,S)) + \delta_S(v,p(u,S))$ & $\delta(v,p(v,S)) + \delta_S(u,p(v,S))$

We shall now show that the new oracle ensures a stretch of 3 at most, that is, for any pair of vertices (u,v), the distance reported is at most 3 times $\delta(u,v)$. Clearly, from triangle inequality the reported distance is at least $\delta(u,v)$.

Theorem 5.2 *The query answering procedure $\mathcal{Q}(u,v)$ reports 3-approximate distance between u and v.*

Proof: If $u \in Ball(v,V,S)$, we report the exact distance since we, just like [15], store the exact distance from v to all the vertices of $Ball(v,V,S)$. So it is the case $u \notin Ball(v,V,S)$ that needs to be analyzed carefully. In fact there are the following two sub-cases.
Case 1: $Ball(u,V,S) = \emptyset$
In this case we show that $\delta(u,p(u,S)) + \delta_S(v,p(u,S))$ is at most $3\delta(u,v)$ using Lemma 5.1 and the fact that $\delta_S(v,p(v,S))$ is the distance between v and $p(v,S)$ in a $(2,1)$-spanner. (we don't need the fact that the spanner is a parameterized spanner for this part of the proof).

Let $v_0(=v), v_1, \cdots, v_l(=u)$ be a shortest path between v and u. Observe that $Ball(u,V,S) = \emptyset$ implies that u must be adjacent to $p(u,S)$, so $\delta(u,p(u,S)) = 1$. It follows from Lemma 5.1 that there is a path between $p(u,S)$ and v_1 in the spanner that consists of at most 2 edges. Furthermore, by the property of $(2,1)$-spanner, the distance $\delta_S(v_1,v_l)$ between v_1 and v_l in the spanner is not greater than $3(l-1)$. Hence $\delta_S(v,p(u,S)) \leq 3(l-1)+2$. So

$$\delta(u,p(u,S)) + \delta_S(v,p(u,S)) \leq 3(l-1) + 2 + 1 = 3l = 3\delta(u,v)$$

Case 2: $Ball(u,V,S) \neq \emptyset$
In this case we show that $\delta(v,p(v,S)) + \delta_S(u,p(v,S))$ is at most $3\delta(u,v)$ using the properties of the parameterized $(2,1)$-spanner. The proof is along the lines as sketched in section 3.1.

a shortest path from v to u can be visualized as consisting of two sub-paths (see Figure 3) : the sub-path $\mathcal{P}_{vw}$ of length a lying inside the Ball and the sub-path $\mathcal{P}_{wu}$ of length $x \geq 1$ outside the Ball. (In case, $a = 0$, the vertex w would be same as v.) Let the length of path $\mathcal{P}_{wu}$ be stretched to x' in the parameterized spanner. Since the spanner is a parameterized spanner with respect to S, it follows from Theorem 4.2 that

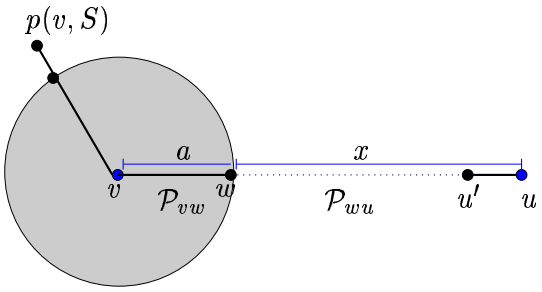


Figure 3: Analyzing the path from v to u when $Ball(u, V, S) \neq \emptyset$

$\delta_S(p(v,S), v) = a + 1$. Now considering the path from u to $p(v,S)$ passing through v it follows that $\delta_S(u, p(v,S)) \leq 2a + 1 + x'$. Hence,

$$\delta(v, p(v,S)) + \delta_S(u, p(v,S)) \leq 3a + 2 + x'$$

To ensure that this distance is no more than three times the actual distance $\delta(u,v) = a + x$, all we need to show is that $x' \leq 3x - 2$. Let $e(u', u)$ be the last edge of the path $\mathcal{P}_{wu}$. Now observe that $Ball(u, V, S) \neq \emptyset$ implies that $u' \in Ball(u, R)$. So it follows from Theorem 4.2 that the edge $e(u', u)$ must be present in the parameterized spanner. Moreover, the part of the sub-path $\mathcal{P}_{wu}$ excluding the edge $e(u', u)$ is of length $x - 1$, and can't be stretched to more than $3(x-1)$ in the $(2,1)$-spanner. Hence $x' \leq 3(x-1)+1 = 3x-2$, and we are done. □

Combining Theorems 5.1 and 5.2, we can state the following theorem.

Theorem 5.3 *An undirected unweighted graph $G = (V, E)$ can be preprocessed in expected $O(n^2)$ time to compute a 3-approximate distance oracle of size $O(n^{3/2})$.*

6 A (2k − 1)-approximate distance oracle in expected $O(n^2)$ time

In this section we shall describe the construction of a $(2k-1)$-approximate distance oracle which can be viewed as a generalization of the new 3-approximate distance oracle.

Computing (2k-1)-approximate distance oracle

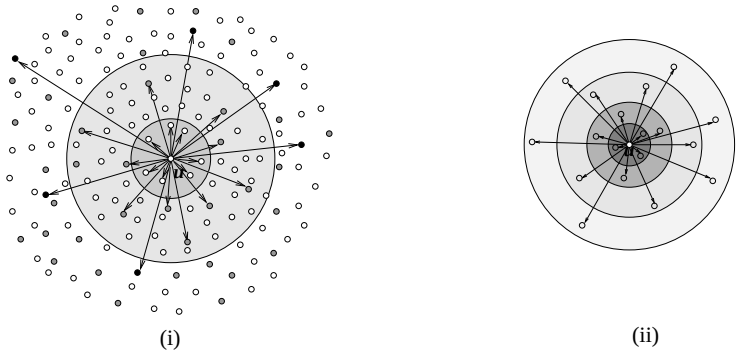
$S_1 \longleftarrow V$

For $i \leftarrow 2$ to k

let S_i contain each element of S_{i-1}, independently, with probability $n^{-1/k}$

1. For $i \leftarrow 2$ to k
   Compute $p(v, S_i)$ and *Ball*(v, S_{i-1}, S_i) for all $v \in V$
2. Compute a parameterized $(2,1)$-spanner (V, E_S) with $S = S_k$.
   (see Note below)
3. Compute $\delta_S(v, x)$ for each $v \in V, x \in S_k$ in the spanner (V, E_S) for $S = S_k$.

Note : While computing a $(2,1)$-spanner in the second step of the preprocessing algorithm, we shall adopt the following strategy : In case a vertex v is neighboring to more than one vertex from set S_k, instead of assigning it to the cluster centered at any arbitrarily picked neighbor from S_k, we shall assign v to the cluster centered at $p(v, S_k)$.



(i) (ii)

Figure 4: (i) Schematic description of $Ball^i(u), i < k$, (ii) hierarchy of balls around u

The final data-structure would keep k hash tables per vertex $v \in V$ as follows. For $1 < i \leq k$, the set of vertices $Ball(v, S_{i-1}, S_i) \cup \{p(u, S_i)\}$ and their distances from v are kept in a hash-table denoted by $Ball^{i-1}(v)$ henceforth. The distances from v to all the vertices of set S_k in the $(2,1)$-spanner is kept in a hash-table denoted as $Ball^k(v)$.

Since S_{i+1} is formed by selecting each vertex of the set S_i with probability $n^{-1/k}$, it follows from Lemma 2.2 that the expected size of $Ball(v, S_i, S_{i+1})$ is $O(n^{1/k})$. Hence, hash-table $Ball^i(v)$ would occupy expected $O(n^{1/k})$ space for each i. It follows that the final data-structure will require expected $O(kn^{1+1/k})$ space.

Let us first analyze the computation time required for the new distance oracle. There are three main tasks in the preprocessing algorithm. The second task involves the computation of a parameterized $(2,1)$-spanner with respect to S_k. It follows from Theorem 4.2 that it would require expected $O(m)$ time and the size of the spanner would be $O(n^{2-1/k})$. The third task requires computation of BFS trees in the spanner from each vertex of set S_k, and hence can be computed in expected $O(n^2)$ time since expected size of S_k is $O(n^{1/k})$. Now let us analyze the first task that requires computation of $p(v, S_i)$ and $Ball(v, S_{i-1}, S_i)$. Similar to the 3-approximate distance oracle, *Ball*$(v, S_{i-1}, S_i) \forall v \in V$ is computed by performing *Restrict_BFS*(x, S_i) on every vertex $x \in S_{i-1}$. Using an improved analysis, which is basically a generalization of Lemma 5.2, we show now bound the expected time required to compute $Ball(v, S_{i-1}, S_i) \forall v \in V$.

Lemma 6.1 *The expected time required to compute $Ball(v, S_{i-1}, S_i) \forall v \in V$ is $O(n^{\frac{k+i}{k}})$.*

Proof: We compute $Ball(v, S_{i-1}, S_i), \forall v \in V$ by executing the subroutine *Restrict_BFS*(x, S_i) on each $x \in S_{i-1}$. Following Lemma 2.5, we just need to show that the expected computational cost charged to a vertex v during all these subroutines is bounded by $O(n^{\frac{i}{k}})$.

Let $v = v_1, v_2, \cdots, v_n$ be the sequence of vertices of the given graph arranged in non-decreasing order of their distance from v. A vertex v_j will charge *deg*(v) to vertex v if $v_j \in Ball(v, S_{i-1}, S_i)$ and zero otherwise. Therefore

$$\mathbf{E}[W_v] = \sum_{j=2}^{j=n} deg(v) \cdot \mathbf{Pr}[v_j \in Ball(v, S_{i-1}, S_i)]$$

Now $v_j \in Ball(v, S_{i-1}, S_i)$ if and only if the following two events happen :
$\mathcal{E}_j$: No vertex at distance at most $\delta(v, v_j)$ from v is present in S_i
$\mathcal{E}'_j : v_j \in S_{i-1}$
Furthermore, every vertex appears in set S_l with probability $n^{\frac{-l+1}{k}}$ independent of any other vertices in the graph. Because of this independence incorporated in selecting vertices to form sets $S_l, l \leq k$, it is easy to see that both the events are independent. Therefore, $\mathbf{Pr}[\mathcal{E}_j \wedge \mathcal{E}'_j] = \mathbf{Pr}[\mathcal{E}_j] \cdot \mathbf{Pr}[\mathcal{E}'_j]$. Hence,

$$\mathbf{E}[W_v] = \sum_j deg(v) \cdot \mathbf{Pr}[\mathcal{E}_j] \cdot \mathbf{Pr}[\mathcal{E}'_j] = n^{\frac{-i+2}{k}} \sum_j deg(v) \cdot \mathbf{Pr}[\mathcal{E}_j]$$

Using arguments similar to those used in Lemma 5.2, it follows that $\mathbf{Pr}[\mathcal{E}_j] = (1-q)^{\max(j, deg(v))}$ with $q = n^{\frac{-i+1}{k}}$, and therefore

$$\mathbf{E}[W_v] \leq n^{\frac{-i+2}{k}} \cdot 1/q^2 = n^{\frac{i}{k}} \quad \{ \text{ for } q = n^{\frac{-i+1}{k}} \}$$

□

The preprocessing algorithm computes $Ball(v, S_{i-1}, S_i) \forall v \in V$ for each $i \leq k$. Hence using Lemma 6.1 and the discussion preceding to that, the following Theorem holds.

Theorem 6.1 *The expected preprocessing time for computing approximate distance oracles is $O(n^2)$.*

6.1 Reporting approximate distance with stretch at most $(2k-1)$

Just like the new 3-approximate distance oracle, there is only a subtle difference between the query answering procedure of the new and the previously existing $(2k-1)$-approximate distance of [15]. First, we provide an overview of the underlying query answering procedure before we formally state it.

Let u and v be two vertices whose approximate distance is to be reported. To analyze the query answering procedure, we introduce the following notation.

For a pair of vertices u and v, a vertex w is said to be *t-near* to u if $\delta(u, w) < t\delta(u, v)$. It follows from the simple triangle inequality that if a vertex is t-near to u then it is $(t+1)$-near to v also.

The entire query answering process is to search for a t-near vertex of u whose distance to both u and and v is known, and $t < k$. The search is performed iteratively as follows : The ith iteration begins with a vertex from set $w \in S_i$ which is $(i-1)$-near to u (and hence i-near to v) and its distance from u is known. It is determined whether or not its distance to v is known. Since $w \in S_i$, we do so by checking whether or not $w \in Ball^i(v)$. Now $w \notin Ball^i(v)$ would happen only if, for the vertex v, the vertex $p(v, S_{i+1})$ is nearer than the vertex $p(u, S_i)$. But this would imply that $p(v, S_{i+1})$ is also i-near to v, and note that its distance from v is known. Therefore, if the distance $\delta(v, w)$ is not known, we continue in the next iteration with vertex $w = p(v, S_{i+1})$, and swap the role of u and v. In this way we proceed gradually, searching over the sets $S_1, \cdots, S_k$. We are bound to find such a vertex within at most k iterations since the distance from S_k is known to all the vertices. We now formally state the query answering procedure which is essentially the same as that of [15] except the `'If'` loop at the end.

$\mathcal{Q}(u,v)$: *Reporting approximate distance between* u *and* v

$i \leftarrow 1, \quad w \leftarrow u$

`While` $(w \notin Ball^i(v))$ `do`

  swap(u,v), $i \leftarrow i+1$

  $w \leftarrow p(u,S_i)$

`If` $i < k$

  `return` $\delta(u,p(u,S_i)) + \delta(v,p(u,S_i))$

`Else`

  `return` **minimum** of

  $\delta(u,p(u,S_k)) + \delta_S(v,p(u,S_k))$ & $\delta(v,p(v,S_k)) + \delta_S(u,p(v,S_k))$

Based on the arguments above, the following Lemma holds.

Lemma 6.2 *[15] In the beginning of the* i*th iteration of* `while` *loop,* $\delta(u,p(u,S_i)) < (i-1)\delta(u,v)$.

Theorem 6.2 *The query answering procedure reports* $(2k-1)$*-approximate distance between* u *and* v.

Proof: The query answering procedure performs iterations of `While` loop until the condition of the loop fails. As mentioned above, there will be at most $(k-1)$ successful iterations before the condition of the loop fails. We shall analyze the distance reported by the oracle on the basis of the final value of i.

Case 1 : $i < k$

If we exit the `While` loop on ith iteration for $i < k$, then we report $\delta(u,p(u,S_i)) + \delta(v,p(u,S_i))$. Using the triangle inequality, this distance is no more than $2\delta(u,p(u,S_i)) + \delta(u,v)$, which, using Lemma 6.2, is no more than $(2i-1)\delta(u,v)$.

Case 2 : $i = k$.

There are the following two sub-cases depending upon which vertex between $p(u,S_k)$ and v is closer to u as shown in Figure 5.

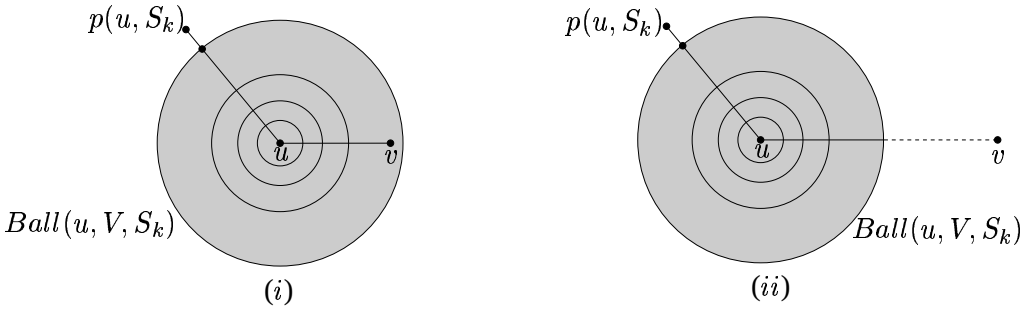


Figure 5: Two sub-cases depending upon whether v lies closer/farther to u relative to $p(u,S_k)$

In sub-case (i), it is the vertex v which is closer to u than the vertex $p(u,S_k)$. Hence the vertex v lies inside $Ball(u,V,S_k)$. Therefore, using Theorem 4.2, it follows that $\delta_S(v,u) = \delta(v,u)$. Also note that $\delta_S(u,p(u,S_k)) = \delta(u,p(u,S_k))$. Combining these two equalities together and using the triangle inequality, it follows that $\delta_S(v,p(u,S_k)) \leq \delta(v,u) + \delta(u,p(u,S_k))$. Hence using Lemma 6.2, the distance reported is at most $(2k-1)\delta(u,v)$ as follows :

$$\begin{aligned} \delta(u,p(u,S_k)) + \delta_S(v,p(u,S_k)) &\leq 2\delta(u,p(u,S_k)) + \delta(u,v) \\ &\leq (2k-1)\delta(u,v) \end{aligned}$$

In sub-case (ii), it is the vertex $p(u, S_k)$ that lies closer to u than the vertex v, that is, $\delta(u, v) > \delta(u, p(u, S_k))$. This implies that $v \notin Ball(u, V, S_k)$. It is not difficult to observe that the sub-case (ii) is exactly the same as that addressed for 3-approximate distance oracle in Theorem 5.2 with S_k playing the role of S. So in this case, the stretch of the distance reported by the oracle is in fact at most 3. □

Combining Theorems 6.1 and 6.2, we can state the following theorem.

Theorem 6.3 *An undirected unweighted graph $G = (V, E)$ can be preprocessed in expected $O(n^2)$ time to compute a $(2k-1)$-approximate distance oracle of size $O(n^{1+1/k})$.*

7 Conclusion and Open Problems

In this paper, we presented an expected $O(n^2)$ time algorithm for computing a $(2k-1)$-approximate distance oracle for unweighted graphs. Recently Roditty *et al.* [14] gave a deterministic algorithm that computes a $(2k-1)$-approximate distance oracle for weighted graphs in $O(mn^{1/k})$ time. They also claim that, for unweighted graphs there exists a deterministic algorithm for computing a $(2k-1)$-approximate distance oracle in $O(n^2)$ time. However, computing a $(2k-1)$-approximate distance oracles for weighted graphs in expected or deterministic $O(n^2 \operatorname{polylog} n)$ time is still an important open problem.

Another result of the paper is a simple and expected linear time algorithm to compute a $(2, 1)$-spanner of $O(n^{3/2})$ size which is optimal. The algorithm was obtained by slight modification in the algorithm [4] for computing a 3-spanner. It is a natural and interesting problem to extend this algorithm for arbitrary k, that is, designing a linear time algorithm for computing a $(k, k-1)$-spanner of size $O(n^{1+1/k})$. Subsequent to the submission of our paper, this result was achieved in [5].

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