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Towards Sub-Quadratic Diameter Computation in Geometric Intersection Graphs

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Abstract
We initiate the study of diameter computation in geometric intersection graphs from the fine-grained complexity perspective. A geometric intersection graph is a graph whose vertices correspond to some shapes in $d$-dimensional Euclidean space, such as balls, segments, or hypercubes, and whose edges correspond to pairs of intersecting shapes. The diameter of a graph is the largest distance realized by a pair of vertices in the graph.

Computing the diameter in near-quadratic time is possible in several classes of intersection graphs [Chan and Skrepetos 2019], but it is not at all clear if these algorithms are optimal, especially since in the related class of planar graphs the diameter can be computed in $\tilde{O}(n^{5/3})$ time [Cabello 2019, Gawrychowski et al. 2021].

In this work we (conditionally) rule out sub-quadratic algorithms in several classes of intersection graphs, i.e., algorithms of running time $O(n^{2-\delta})$ for some $\delta > 0$. In particular, there are no sub-quadratic algorithms already for fat objects in small dimensions: unit balls in $\mathbb{R}^3$ or congruent equilateral triangles in $\mathbb{R}^2$. For unit segments and congruent equilateral triangles, we can even rule out strong sub-quadratic approximations already in $\mathbb{R}^2$. It seems that the hardness of approximation may also depend on dimensionality: for axis-parallel unit hypercubes in $\mathbb{R}^{12}$, distinguishing between diameter 2 and 3 needs quadratic time (ruling out $(3/2-\varepsilon)$-approximations), whereas for axis-parallel unit squares, we give an algorithm that distinguishes between diameter 2 and 3 in near-linear time.

Note that many of our lower bounds match the best known algorithms up to sub-polynomial factors. Ultimately, this fine-grained perspective may enable us to determine for which shapes we can have efficient algorithms and approximation schemes for diameter computation.

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1 Introduction

The diameter of a simple graph $G = (V, E)$ is the largest distance realized by a pair of its vertices; formally, it is $\text{diam}(G) = \max_{u,v \in V} \text{dist}_G(u,v)$, where $\text{dist}_G(u,v)$ is the number of edges on a shortest path from $u$ to $v$. It is one of the crucial parameters of a graph that can be computed in polynomial time. Geometric intersection graphs are the standard model for wireless communication networks [34], but more abstractly, they can be used to represent networks where the connection of nodes relies on proximity in some metric space. For a (slightly oversimplified) example, consider a set of devices in the plane capable of receiving and transmitting information in a range of radius 2. These devices form a communication network that is a unit disk graph. Indeed, two devices can communicate with each other if and only if their distance is at most 2, i.e., if the unit disks centered at the devices have a non-empty intersection. For our purposes, the underlying metric space will be $d$-dimensional Euclidean space (henceforth denoted by $\mathbb{R}^d$), and we will consider intersection graphs of common objects such as balls and segments. For a set $F$ of objects in $\mathbb{R}^d$ (that is, $F \subset 2^{\mathbb{R}^d}$), the corresponding intersection graph $G[F]$ has vertex set $F$ and edge set $\{uv \mid u,v \in F, u \cap v \neq \emptyset\}$.

Computing the diameter in geometric intersection graphs is an important task: if the graph represents a communication network, then the diameter of the network can help estimate the time required to spread information in the network, as the information needs to go through up to $\text{diam}(G)$ links to reach its destination. In large networks, it is also indispensable to have near-linear time algorithms; it is therefore natural to study if a given class of geometric intersection graphs admits a near-linear time algorithm for exact or approximate diameter computation.

The extensive literature on diameter computation serves as a good starting point. The diameter of an $n$-vertex (unweighted) graph can be computed in $\mathcal{O}(n^{\omega} \log n)$ expected time, where $\omega < 2.37286$ is the exponent of matrix multiplication [41]. If the graph has $m$ edges, then the diameter can also be computed in $\mathcal{O}(mn)$ time [42], which gives a near-quadratic running time of $\tilde{\mathcal{O}}(n^2)$ in case of sparse graphs, i.e., when $m = \mathcal{O}(n)$. In fact, these algorithms are capable of computing not only the diameter, but also all pairwise distances in a graph, known as the all pairs shortest paths problem.

On the negative side, we know that computing the diameter of a graph cannot be done in $\mathcal{O}(n^{2-\varepsilon})$ time under the Orthogonal Vectors Hypothesis\(^1\) (OV); in fact, deciding if the diameter of a sparse graph is at most 2 or at least 3 requires $n^{2-o(1)}$ time under OV [40]\(^2\), which rules out sub-quadratic $(3/2 - \varepsilon)$-approximations for all $\varepsilon > 0$.

In special graph classes however it is possible to compute the diameter in sub-quadratic time. In planar graphs, an algorithm with running time $\mathcal{O}(n^2)$ is very easy: one can just run $n$ breadth-first searches, each of which take linear time because the number of edges is $\mathcal{O}(n)$.

---

\(^1\) See Section 2 for the definitions and some background on the hypotheses used in our lower bounds.

\(^2\) More precisely, Roditty and Vassilevska-Williams [40] give a reduction from $\alpha$-Dominating Set, which can be adapted to a reduction from OV as described in the beginning of Section 4.
It has been a long-standing open problem whether a truly sub-quadratic algorithm exists for diameter computation, until the breakthrough of Cabello [12], who used Voronoi diagrams in planar graphs. The technique was later improved by Gawrychowski et al. [29], who obtained a running time of $\tilde{O}(n^{5/3})$.

Certain geometric intersection graphs often behave similarly to planar graphs. The most widely studied classes, (unit) disk and ball graphs admit approximation schemes for maximum independent set, maximum dominating set, and several other problems [30, 31, 14], with techniques similar to planar graphs. Unlike planar graphs, geometric intersection graphs can have arbitrarily large cliques, but at least the maximum clique can be approximated efficiently [7]. In fact, planar graphs are special disk intersection graphs by the circle packing theorem [33]. When it comes to computing the diameter, the similarity with planar graphs is not so easy to see. Even getting near-quadratic diameter algorithms is non-trivial, as geometric intersection graphs can be arbitrarily dense.

Chan and Skrepetos [16] provide near-quadratic ($\tilde{O}(n^2)$) APSP algorithms for several graph classes, including disks, axis-parallel segments, and fat triangles in the plane, and cubes and boxes in constant-dimensional space. Unit disk graphs have a “weakly” sub-quadratic algorithm (that is poly-logarithmically faster than $O(n^2)$) [15]. We are not aware of any $O(n^2 - \epsilon)$ algorithms for computing the diameter in intersection graphs of any planar shape.

**Further related work.** While computing the diameter is known to require time $n^{2+o(1)}$ already on sparse graphs (assuming the OV Hypothesis), an extensive line of research including [3, 40, 19, 13, 5, 25, 9, 24, 37, 8, 23] studies the (non-)existence of faster approximation algorithms. On the positive side, this includes in particular a folklore 2-approximation in time $\tilde{O}(m)$ and a $3/2$-approximation in time $\tilde{O}(m^{3/2})$ [3, 40, 19], both already for weighted digraphs. Remarkably, these algorithms can be shown to be tight: [40, 5] establish that the $3/2$-approximation in time $\tilde{O}(m^{3/2})$ cannot be improved in either approximation guarantee or running time (assuming the $k$-OV Hypothesis), already for unweighted undirected graphs. The near-linear time $2$-approximation is conditionally optimal as well: For unweighted directed graphs, this has been proven independently in [24, 37]. For unweighted undirected graphs, following further work [8], a resolution has been announced only very recently [23]. Thus, approximating the diameter in sparse graphs is quite well understood, including detailed insights into the full accuracy-time trade-off. In the context of our work, the challenge is to obtain a similar understanding for our setting of unweighted, undirected geometric graphs, which are non-sparse in general.

Note that for graph classes that are non-sparse, a natural question is whether diameter can be computed in $O(m + n)$ time, i.e., linear time in the number of edges plus vertices. The question has been studied by several authors: using a variant of breadth-first search called lexicographic breadth-first search, one can find a vertex of very large eccentricity. In some classes, we now know that there is an $O(m + n)$ algorithm for diameter: notably, this holds in interval graphs as well as $\{\text{claw}, \text{asteroidal triple}\}$-free graphs [27, 10]. In many other graph classes (such as chordal graphs and asteroidal-triple-free graphs) we can get approximations for the diameter that differ only by a small additive constant from the optimum [27, 26, 21]. See [22] for an overview on the connection of lexicographic BFS and diameter, and see [20] for a survey on lexicographic BFS.

Another related direction is to consider edge weighted graph classes. In some classes of geometric intersection graphs there is a natural weighting to consider: for example in ball graphs, it is customary to draw the graph edges with straight segments that connect the centers of the two adjacent disks. The edges then have a natural weighting by their Euclidean
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length. This was considered for unit disk graphs in the plane by Gao and Zhang [28], who obtained a \((1 + \varepsilon)\)-approximation for \(\text{Diameter}\) in \(O(n^{3/2})\) time for any fixed \(\varepsilon > 0\). A faster \((1 + \varepsilon)\)-approximation with running time \(O(n \log^2 n)\) for any fixed \(\varepsilon > 0\) was given by Chan and Skrepetos [17]. Since the underlying graph is not changed by this weighting, it is natural to think that similar results should be possible also for unweighted unit disk graphs.

It remains an open question whether the complexity of diameter computation is influenced by the presence of these Euclidean weights.

Our results. In this article, we show that most of the results of Chan and Skrepetos [16] cannot be significantly improved under standard complexity-theoretic assumptions, even if we are only interested in the diameter instead of all pairs shortest paths. In particular, we rule out sub-quadratic diameter algorithms for fat triangles and axis-aligned segments in the plane, as well as for unit cubes in \(\mathbb{R}^3\), leaving only their \(\tilde{O}(n^{7/3})\) algorithm for arbitrary segments in \(\mathbb{R}^3\) as well as their \(\tilde{O}(n^2)\) algorithm for disks without a matching lower bound.

The \(\text{Diameter}\) problem has as input a set of geometric objects in \(\mathbb{R}^d\) and a number \(k\); the goal is to decide whether the diameter of the intersection graph of the objects is at most \(k\). The \(\text{Diameter}-t\) problem is the same problem, but with \(k\) set to the constant number \(t\). We show the following lower bounds.

\begin{itemize}
  \item \textbf{Theorem 1.} For all \(\delta > 0\) there is no \(O(n^{2-\delta})\) time algorithm for \(\text{Diameter-3}\) in intersection graphs of unit segments in \(\mathbb{R}^2\) under the \(\text{OV}\) Hypothesis.
  \item \(\text{Diameter-3}\) in intersection graphs of congruent equilateral triangles in \(\mathbb{R}^2\) under the \(\text{OV}\) Hypothesis.
  \item \(\text{Diameter}\) in intersection graphs of unit balls in \(\mathbb{R}^3\) under the \(\text{OV}\) Hypothesis.
  \item \(\text{Diameter}\) in intersection graphs of axis-parallel unit cubes in \(\mathbb{R}^3\) under the \(\text{OV}\) Hypothesis.
  \item \(\text{Diameter}\) in intersection graphs of axis-parallel line segments in \(\mathbb{R}^2\) under the \(\text{OV}\) Hypothesis.
  \item \(\text{Diameter-2}\) in intersection graphs of axis-parallel hypercubes in \(\mathbb{R}^{12}\) under the \(\text{Hyperclique}\) Hypothesis.
\end{itemize}

Our results imply lower bounds for approximations. (See Section 4.2 for a short proof.)

\begin{itemize}
  \item \textbf{Corollary 2.} Under the Orthogonal Vectors and Hyperclique Hypotheses, for all \(\delta, \varepsilon > 0\) there is no \(O(n^{2-\delta})\) time \((4/3 - \varepsilon)\)-approximation for \(\text{Diameter}\) in intersection graphs of unit segments or congruent equilateral triangles in \(\mathbb{R}^2\), and no \((3/2 - \varepsilon)\)-approximation in intersection graphs of axis-parallel hypercubes in \(\mathbb{R}^{12}\). Furthermore, for all \(\delta > 0\) there is no \(O(n^{2-\delta} \text{poly}(1/\varepsilon))\) time approximation scheme that provides a \((1 + \varepsilon)\)-approximation for \(\text{Diameter}\) for any \(\varepsilon > 0\) in intersection graphs of axis-parallel unit segments in \(\mathbb{R}^2\), or unit balls or axis-parallel unit cubes in \(\mathbb{R}^3\).
\end{itemize}

Theorem 1 shows that sub-quadratic algorithms in many intersection graphs classes are unlikely to exist; one must wonder if such algorithms are possible at all? A notable case missing from our lower bounds are the case of unit disks; indeed, it is possible that unit disk graphs enjoy sub-quadratic diameter computation. More generally, it is an interesting open question whether intersection graphs of so-called pseudodisks admit sub-quadratic diameter algorithms. (Pseudodisks are objects bounded by Jordan curves such that the boundaries of any pair of objects have at most two intersection points.) We make a step towards resolving this problem with the following theorem for intersection graphs of axis-parallel unit squares – since axis-parallel unit squares are pseudodisks.
Theorem 3. There is an $O(n \log n)$ algorithm for DIAMETER-2 in unit square graphs.

The algorithm is based on the insight that the problem can be simplified to the following: given skylines $A, B$ and a list of axis-parallel squares $S$, check whether each pair $(a, b) \in A \times B$ is covered by some square $s \in S$. Since any axis-parallel square $s \in S$ covers intervals in $A$ and $B$, this problem in turn reduces to checking whether the union of $|S|$ rectangles covers the $A \times B$ grid. Using near-linear skyline computation [35], and a line sweep for the grid covering problem, we obtain a surprisingly simple $O(n \log n)$ time algorithm (in contrast to the quadratic-time hardness in higher dimensions).

Organization. After some preliminaries and the introduction of the complexity-theoretic hypotheses used in the paper, we present our algorithm for unit squares in Section 3. Section 4 showcases our lower bound techniques. The lower bounds for unit segments, congruent equilateral triangles as well as for axis-parallel unit segments have a structure similar to two other lower bounds in Section 4, and they can be found in the full version.

2 Preliminaries

Let $G = (V, E)$ be a graph, and $u$ and $v$ be vertices in $G$. The distance from $u$ to $v$ is denoted by $\text{dist}_G(u, v)$, and equals the number of edges on the shortest path from $u$ to $v$ in $G$. The diameter of $G$ is denoted by $\text{diam}(G)$ and equals to $\max_{u, v \in V} \text{dist}_G(u, v)$. The open and closed neighborhood of a vertex $v$ are $N(v) = \{u \in V \mid uv \in E\}$ and $N[v] = \{v\} \cup N(v)$, respectively. Let $A, B \subseteq V$ be sets of vertices. The diameter of $A$ and $B$ is denoted by $\text{diam}_G(A, B) = \max_{(a, b) \in A \times B} \text{dist}_G(a, b)$. Finally, let $[n]$ denote the set $\{1, \ldots, n\}$.

2.1 Hardness assumptions

We use two hypotheses from fine-grained complexity theory for our lower bounds. For an overview of this field, we refer to the survey [44].

Orthogonal Vectors Hypothesis. Let OV denote the following problem: Given sets $A, B$ of $n$ vectors in $\{0, 1\}^d$, determine whether there exists an orthogonal pair $a \in A, b \in B$, i.e., for all $i \in [d]$ we have $(a)_i = 0$ or $(b)_i = 0$. Exhaustive search yields an $O(n^2d)$ algorithm, which can be improved for small dimension $d = c \log n$ to $O(n^{2-1/O(\log(c))})$ [2, 18]. For larger dimensions $d = \omega(\log n)$, it is known [45] that no $O(n^{2-\epsilon})$-time algorithm can exist unless the Strong Exponential Time Hypothesis [32] fails. Thus, the Strong Exponential Time Hypothesis implies the following (so-called “moderate-dimensional”) OV Hypothesis.

Hypothesis 4 (Orthogonal Vectors Hypothesis). For no $\epsilon > 0$, there is an algorithm that solves OV in time $O(\text{poly}(d)n^{2-\epsilon})$.

By now, there is an extensive list of problems with tight lower bounds (including subquadratic equivalences) based on this assumption, see [44].

Hyperclique Hypothesis. For $k \geq 4$, let $3$-uniform $k$-HYPERCLIQUE denote the following problem: Given a 3-uniform hypergraph $G = (V, E)$, determine whether there exists a hyperclique of size $k$, i.e., a set $S \subseteq V$ such that for all $e \in \binom{S}{k}$, we have $e \in E$. By exhaustive search, we can solve this problem in time $O(n^k)$ where $n = |V|$. Unlike the usual $k$-CLIQUE problem in graphs, for which a $O(n^{\omega(k)/3 + O(1)})$ algorithm exists [39], no techniques are known that would beat exhaustive search by a polynomial factor for the problem in hypergraphs. This has lead to the hypothesis that exhaustive search is essentially best possible.
Hypothesis 5 (Hyperclique Hypothesis). For no \( \epsilon > 0 \) and \( k \geq 4 \), there is an algorithm that would solve \( 3 \)-uniform \( k \)-Hyperclique in time \( O(n^{k-\epsilon}) \).

See [38] for a detailed description of the plausibility of this hypothesis. Tight conditional lower bounds (including fine-grained equivalences) have been obtained, e.g., in [1, 11, 36, 4].

3 Solving the Diameter-2 problem on unit square graphs

In this section, we are going to present an algorithm with running time \( O(n \log n) \) for the Diameter-2 problem for unit square graphs. For each unit square \( v \in V \), we consider the center of \( v \), denoted \( \hat{v} \), as the point representing \( v \) in the plane; for a square set \( X \subset V \), we use \( \hat{X} \) to denote the set of corresponding centers. Let \( \hat{G} = (\hat{V}, E) \) denote the graph on centers of squares in \( G \). Hence, for all \( \{u, v\} \in E(G) \), there is an edge between \( \hat{u} \) and \( \hat{v} \). Note that we will often use \( \hat{G} \) and \( G \) interchangeably.

Notice that a graph has diameter at most two if and only if for every pair of vertices \( u, v \in V \):
\[
N[u] \cap N[v] \neq \emptyset,
\]
where \( N[u] \) denotes the set of neighbors of \( u \). Equivalently, the square of side length 2 centered at \( \hat{w} \) must cover both \( \hat{u} \) and \( \hat{v} \). For a square \( w \), let \( w^2 \) denote the side-length-2 square of center \( \hat{w} \). Thus, in order to decide whether \( \text{diam}(G) \leq 2 \), it is sufficient to check whether for every \( u, v \in V \) there exists \( w \in V \) such that \( \hat{u}, \hat{v} \in w^2 \).

For a set of points \( P \) we define the top-left front, TLF(\( P \)), and bottom-right front, BRF(\( P \)), as follows (see Figure 1a).
\[
\text{TLF}(P) = \{p \in P | \forall q \in P : p_x \leq q_x \text{ or } p_y \geq q_y\}
\]
\[
\text{BRF}(P) = \{p \in P | \forall q \in P : p_x \geq q_x \text{ or } p_y \leq q_y\}
\]

Similarly, we define the top-right front, TRF(\( P \)), and bottom-left front, BLF(\( P \)), as follows (see Figure 1b).
\[
\text{TRF}(P) = \{p \in P | \forall q \in P : p_x \geq q_x \text{ or } p_y \geq q_y\}
\]
\[
\text{BLF}(P) = \{p \in P | \forall q \in P : p_x \leq q_x \text{ or } p_y \leq q_y\}
\]

Lemma 6. The graph \( G \) has diameter at most 2 if and only if
\[
\max \left( \text{diam}_G(\text{BLF}(V), \text{TRF}(V)), \text{diam}_G(\text{TLF}(V), \text{BRF}(V)) \right) \leq 2.
\]
Thus, we can think of the sets $\mathcal{I}$, $\mathcal{J}$, and $\mathcal{Q}$ as sweep $[6, 43]$. The time needed to construct the rectangles in grid of centers, there is a rectangle $R$ of side length. For each square $P \subseteq \mathcal{I}$, $\mathcal{J}$, and $\mathcal{Q}$, we are able to prove Theorem 3.

Using Lemma 6, we are able to prove Theorem 3.

Proof of Theorem 3. We start our algorithm by computing $\mathcal{T}(\mathcal{V}), \mathcal{T}(\mathcal{W}), \mathcal{B}(\mathcal{V}),$ and $\mathcal{B}(\mathcal{W})$ in $O(n \log n)$ time [35]. Let $\mathcal{P} = \mathcal{B}(\mathcal{V})$ and $\mathcal{Q} = \mathcal{B}(\mathcal{W})$. By Lemma 6, it is sufficient to show that in $O(n \log n)$ time we can decide whether $\mathcal{B}(\mathcal{P}, \mathcal{Q}) \subseteq 2$; using the same algorithm for $\mathcal{B}(\mathcal{V})$ and $\mathcal{T}(\mathcal{W})$ will then get the desired running time.

In order to check whether $N[p] \cap N[q] \neq \emptyset$ for all $(p, q) \in \mathcal{P} \times \mathcal{Q}$, we do the following: Consider $\mathcal{P} = \{p_1, \ldots, p_{\mathcal{P}}\}$ and $\mathcal{Q} = \{q_1, \ldots, q_{\mathcal{Q}}\}$ in $x$-order. Also, let GRID $= [[\mathcal{P}] \times [[\mathcal{Q}]]$ be a grid where $p_i$ corresponds to the $i$-th row and $q_j$ corresponds to the $j$-th column.

For each square $v \in \mathcal{V}$, recall that $v^2$ denotes the square with the same center but twice the side length. For each $v \in \mathcal{V}$, define $I_v \subseteq \{1, 2, \ldots, |\mathcal{P}|\}$ such that $i \in I_v$ iff $v^2$ contains $p_i$. Similarly, $J_v \subseteq \{1, 2, \ldots, |\mathcal{Q}|\}$ such that $j \in J_v$ iff $v^2$ contains $q_j$. Since $v^2$ is an axis-parallel square, it covers intervals from both $\mathcal{P}$ and $\mathcal{Q}$, thus $I_v$ and $J_v$ consist of consecutive integers. Therefore, we can think of the sets $I_v \times J_v$ as rectangles in GRID.

Claim 7. We have $N[p] \cap N[q] \neq \emptyset$ for all $(p, q) \in \mathcal{P} \times \mathcal{Q}$ if and only if the union of $I_v \times J_v$ over all squares $v \in \mathcal{V}$ covers GRID.

Proof. If the union of all rectangles covers the whole grid, then for any pair $(p_i, q_j) \in \mathcal{P} \times \mathcal{Q}$ of centers, there is a rectangle $I_v \times J_v$ that covers $(i, j)$. Therefore, $v^2$ covers both $p_i$ and $q_j$. Thus, $v$ is a shared neighbor of $p_i$ and $q_j$.

If $N[p] \cap N[q] \neq \emptyset$ for all $(p, q) \in \mathcal{P} \times \mathcal{Q}$, then for each pair $(p_i, q_j)$ there is at least one square $v_{ij}$ such that $v_{ij}^2$ contains both $p_i$ and $q_j$. Hence, $(i, j) \in I_{v_{ij}} \times J_{v_{ij}}$ for each $(i, j) \in \mathcal{V}$. As a result, the union of $I_v \times J_v$ over all squares $v^2$ covers GRID.

Note that the problem in Claim 7 corresponds to determining whether a union of rectangles covers the full grid. This problem can be solved in $O(n \log n)$ time with a plane sweep [6, 43]. The time needed to construct the rectangles in GRID is $O(n \log n)$ as there are $O(n)$ rectangles. This concludes the proof of Theorem 3. □
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4 Lower bounds based on the Orthogonal Vectors Hypothesis

In this section, we prove lower bounds for finding the diameter in various intersection graphs.

For a comparison to similar results on sparse graphs, let us briefly describe the result ruling out a $(3/2 - \epsilon)$-approximation in time $O(n^{2-\delta})$, for any $\epsilon, \delta > 0$, due to Roditty and Vassilevska-Williams [40]. While it is originally stated as a reduction from $k$-Dominating Set, we adapt it to give a reduction from OV: Given sets $A, B \subseteq \{0, 1\}^d$, introduce vector nodes for each $a \in A$ and $b \in B$ as well as coordinate nodes for $k \in [d]$. Without loss of generality (see Section 4.1), one may assume that all vectors $a \in A$ have $(a)_{d-1} = 1$ and all vectors $b \in B$ have $(b)_{d-1} = 1$. We connect each vector node $v \in A \cup B$ to the coordinate node $k \in [d]$ iff $(v)_k = 1$, and make all coordinate nodes a clique by adding all possible edges between coordinate nodes. The important observation is that (1) a pair $a \in A, b \in B$ has distance at most 2 iff there is a $k \in [d]$ such that $(a)_k = (b)_k = 1$, i.e., $a, b$ do not form an orthogonal pair, and (2) all other types of node pairs have distance at most 2. Thus, $A, B$ contains an orthogonal pair iff the diameter of the constructed graph is at least 3. Since the reduction produces a sparse graph with $O(n + d)$ nodes and $O(nd)$ edges in time $O(n^d)$, any $O(n^{2-\delta})$-time algorithm distinguishing between diameter 2 and 3 would give a $O(n^{2-\delta}poly(d))$-time OV algorithm, refuting the OV Hypothesis.

Generally speaking, implementing this reduction using low-dimensional geometric graphs is problematic: we must be able to implement an arbitrary bipartite graph on a vertex set $L \times R$ where $|L| = n$ and $|R| = d$. Instead, in this section we implement two different types of reductions via geometric graphs; the main ideas are as follows:

Diameter-3 graphs (Section 4.1 and full version). Instead of coordinate nodes, we introduce 1-entry nodes $(v)_k$ for all $v \in A \cup B, k \in [d]$ with $(v)_k = 1$. This increases the number of nodes only to $O(nd)$, while allowing us to geometrically implement edges of the form $(v, (v)_k)$ for all $v \in A \cup B, k \in [d]$ with $(v)_k = 1$ and $\{(v)_k, (v')_k\}$ for all $v, v' \in A \cup B, k \in [d]$ with $(v)_k = (v')_k = 1$. Now, a witness of non-orthogonality of $a, b$ is a 3-path $a - (a)_k - (b)_k - b$. By showing that all other distances are bounded by 3, we obtain hardness for the Diameter-3 problem. See Section 4.1 and the full version for details, including the use of an additional node to make all 1-entry nodes sufficiently close in distance.

(Non-sparse) Diameter-$\Theta(d)$ graphs (Section 4.2 and full version). Instead of coordinate nodes or 1-entry nodes, we introduce vector-coordinate nodes $(v)_k$ for all $v \in A \cup B, k \in [d]$, irrespective of whether $(v)_k = 1$. As opposed to previously, we do not create a constant diameter instance: The idea is to create an instance where the most distant pairs are of the form $(a)_1, (b)_d$ for all $a \in A, b \in B$, and a non-orthogonality witness is a path of the form $(a)_1 \sim \cdots \sim (a)_k \sim (b)_k \sim \cdots \sim (b)_d$ with $(a)_k = (b)_k = 1$. This construction requires us to implement perfect matchings between vector-coordinate gadgets $(a)_k$ for $a \in A$ and $(a')_{k+1}$ for $a' \in A$ if $a = a'$, as well as a gadget for implementing short connections for $(a)_k \sim (b)_k$ that check whether $(a)_k = (b)_k = 1$. Interestingly, this type of reduction generally produces dense graphs with $\Omega(n^2)$ edges, so this approach crucially exploits the expressive power of geometric graphs to give a subquadratic reduction. See Section 4.2 and the full version for details, including a description of auxiliary nodes not mentioned here.

Finally, we remark that the reduction for unit hypercubes given in Section 5 has the most similar structure to the reduction by Roditty and Vassilevska-Williams [40], despite starting from a different hypothesis, and has similarities to [4, Theorem 14]. We crucially exploit properties of the hyperclique problem to implement it using hypercube graphs.
4.1 The Diameter-3 problem for line segment intersection graphs

In this section, we are going to present a lower bound on the running time of the algorithm for the DIAMETER-3 problem for line segment intersection graphs, such that vertices are line segments with any length, and there is an edge between a pair of line segments if they intersect. This serves as a warm-up for the slightly more complicated reductions below.

Theorem 8. For all $\epsilon > 0$, there is no $O(n^{2-\epsilon})$ time algorithm for the DIAMETER-3 problem for line segment intersection graphs, unless the OV Hypothesis fails.

Let $A = \{a_1, a_2, \ldots, a_n\}$ and $B = \{b_1, b_2, \ldots, b_n\}$ be two sets of $n$ vectors in $[0,1]^d$. We construct a set of segments such that the diameter of the corresponding intersection graph is at most 3 if and only if there is no orthogonal pair $(a, b) \in A \times B$.

Without loss of generality, we assume that for each $a_i \in A$ and $b_j \in B$, $(a_i)_{d-1}, (a_i)_{d}) = (1, 0)$ and $(b_j)_{d-1}, (b_j)_{d}) = (0, 1)$, by adding two coordinates to the ends of the vectors. Note that adding these coordinates does not change whether vectors $a, b$ are orthogonal or not.

For each vector $a_i \in A$, let $\bar{a}_i$ denote a zero-length line segment from $(i,1)$ to $(i,1)$. Analogously, for each vector $b_j \in B$, let $\bar{b}_j$ denote a line segment from $(j,1)$ to $(j,1)$. Furthermore, let $\ell$ be a line segment from $(1,0)$ to $(d,0)$, and let $\{w_1, w_2, \ldots, w_d\}$ be $d$ different points on $\ell$ such that for all $k \in [d]$, $w_k$ is located at $(k,0)$. Moreover, for each $a_i \in A$, if $(a_i)_k = 1$, we define a line segment $e_{i,k}$ from $\bar{a}_i$ to $w_k$ (i.e., from $(i,1)$ to $(k,0)$). Analogously, for each $b_j \in B$, if $(b_j)_k = 1$, we define a line segment $e'_{j,k'}$ from $\bar{b}_j$ to $w_{k'}$ (i.e., from $(j,1)$ to $(k',0)$). Let $\bar{V}$ be the set of constructed line segments, and let $G$ be their intersection graph (see Figure 3).

Lemma 9. The sets $A$ and $B$ contain an orthogonal pair if and only if $diam(G) \geq 4$.

Let $\tilde{A}$ be the set of line segments corresponding to vectors in $A$. Analogously, let $\tilde{B}$ be the set of line segments corresponding to vectors in $B$. To prove the lemma, we show that each pair of vertices is within distance at most 3, unless it is in $\tilde{A} \times \tilde{B}$ (see Claim 10 below and see the full version for its proof). The pairs in $\tilde{A} \times \tilde{B}$ have distance 4 or 3 depending on whether their corresponding vectors in $A \times B$ are orthogonal or not.

Claim 10. $dist(\bar{u}, \bar{v}) \leq 3$ for all $(\bar{u}, \bar{v}) \in (\bar{V} \times \bar{V}) \setminus (\tilde{A} \times \tilde{B} \cup \tilde{B} \times \tilde{A})$.

Proof of Lemma 9. If $a_i$ and $b_j$ are not orthogonal, then there is at least one $k \in [d]$ such that $(a_i)_k = (b_j)_k = 1$. Hence, the path $\bar{a}_i - e_{i,k} - e'_{j,k} - \bar{b}_j$ exists, and it has length 3. If $a_i$ and $b_j$ are orthogonal, then there is no index $k$ such that $(a_i)_k = (b_j)_k = 1$. Consequently, there is no path of length 3 from $\bar{a}_i$ to $\bar{b}_j$, and $dist(\bar{a}_i, \bar{b}_j) \geq 4$. Together with Claim 10 this proves the lemma. □
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**Theorem 11.** For all $\epsilon > 0$, there is no $O(n^{2-\epsilon})$ time algorithm for solving Diamater-3 in unit ball graphs in $\mathbb{R}^3$ under the Orthogonal Vectors Hypothesis.

Let $A = \{a_1, a_2, \ldots, a_n\}$ be a given set of vectors from $\{0, 1\}^d$. First, we construct graph $G(A)$ and show that $G(A)$ has diameter $d$ if and only if there is an orthogonal pair of vectors in $A$. Next, we show how $G(A)$ can be realized as an intersection graph of unit balls in $\mathbb{R}^3$. Without loss of generality, assume that the all-one vector is an element of $A$ (if it is not in $A$, then adding the all-one vector does not change whether there is an orthogonal pair.)

We construct a graph $G(A)$ as follows. Let $C_1^T, \ldots, C_d^T$ and $C_1^B, \ldots, C_d^B$ be cliques, such that for all $k \in [2d]$, $C_k^T = \{v_k^T_1, v_k^T_2, \ldots, v_k^T_{\epsilon(k)}\}$, $C_k^B = \{v_k^B_1, v_k^B_2, \ldots, v_k^B_{\epsilon(k)}\}$, and $v_k^T$ and $v_k^B$ correspond to $a_i$ for all $i \in [n]$, see Figure 4. Add a perfect matching between each pair $C_k^T$ and $C_{k+1}^T$ for all $k \in [2d-1]$ such that there is an edge incident to $v_k^T$ and $v_{k+1}^T$ for all $i \in [n]$. Analogously, there is a perfect matching between each pair $C_k^B$ and $C_{k+1}^B$.

Let $M_1^T, \ldots, M_d^T$ be cliques such that if $(a_i)_k = 1$, then there is a vertex $m_{k,i}^T$ in $M_k^T$ that is adjacent to $v_{k,i}^T$. Similarly, let $M_1^B, \ldots, M_d^B$ be cliques such that if $(a_i)_k = 1$, then there is a vertex $m_{k,i}^B$ in $M_k^B$ that is adjacent to $v_{k,i}^B$. Notice that because of the addition of the all ones vector, the cliques $M_k^T$ and $M_k^B$ are all non-empty.

Finally, let $Q = \{q_1, q_2, \ldots, q_d\}$ be a set of vertices such that $q_k$ has edges to all vertices in $M_k^T$ and $M_k^B$ for all $k \in [d]$.

**Lemma 12.** The graph $G(A)$ has diameter at most $2d + 4$ iff $A$ has no orthogonal pair.
Proof. Assume that there is an orthogonal pair \((a_i, a_j)\) \(\in A\) such that \(i \neq j\). Hence, 
\[\sum_{k=1}^{n} (a_k)_i (a_j)_k = 0,\]
which means that there is no \(k \in [n]\) such that \((a_i)_k = (a_j)_k = 1\). Consequently, for all \(k \in [d]\), the distance from \(v^T_{2k,i} \in C^T_{2k}\) to \(v^B_{2k-1,j} \in C^B_{2k-1}\) is at least 5.
Therefore, \(2d + 4 < \text{dist}(v^T_{1,i}, v^B_{2d,j}) \leq \text{diam}(G(A))\).

Now suppose that \(A\) has no orthogonal pair. We want to prove that \(\text{diam}(G(A)) \leq 2d + 4\). Since \(A\) has no orthogonal pair, for each pair \((a_i, a_j)\) there is at least one \(k \in [n]\) such that \((a_i)_k = (a_j)_k = 1\). Therefore, there are cliques \(M^T_k\) and \(M^B_k\) that have the vertices \(v^T_{k,i}\) and \(v^B_{k,j}\) respectively. Since all vertices in \(M^T_k\) and \(M^B_k\) have an edge to \(q_k\), we can reach \(q_k\) from \(v^T_{1,i}\) by a path of length \(2k - 1 + 2\). Simultaneously, we can reach \(q_k\) from \(v^B_{2d,j}\) by a path of length \(2d - 2k + 1 + 2\). In total, this gives a path of length \(2d + 4\) between \(v^T_{1,i}\) and \(v^B_{2d,j}\). Furthermore, it is easy check that the distance of any pair of vertices where at least one vertex is outside \(C^T_1 \cup C^B_{2d}\) is at most \(2d + 4\). As a result, \(\text{diam}(G(A)) \leq 2d + 4\). ▶

Lemma 13. \(G(A)\) can be realized as an intersection graph of unit balls in \(\mathbb{R}^3\).

Proof. For converting \(G(A)\) into an intersection graph of unit balls, we should consider each vertex in \(G(A)\) as the center of a unit diameter ball, and for those vertices that are adjacent, their corresponding unit balls should intersect. To this end, we choose the following coordinates for the centers of the unit balls in \(\mathbb{R}^3\):
- For all \(k \in [2d]\) and \(i \in [n]\), the center point of \(v^T_{k,i} \in C^T_k\) is \((k, \frac{1}{n}, 0)\).
- For all \(k \in [d]\) and \(i \in [n]\), if \(m^T_{k,i} \in M^T_k\) exists, then its center point is \((2k, \frac{i}{n}, -1)\).
- For all \(k \in [d]\), the center point of \(q_k \in Q\) is \((2k, \frac{1}{n}, -1.6)\).
- For all \(k \in [d]\) and \(i \in [n]\), if \(m^B_{k,i} \in M^B_k\) exists, its center point is \((2k, \frac{i}{n}, -2.2)\).
- For all \(k \in [2d]\) and \(i \in [n]\), the center point of \(v^B_{k,i} \in C^B_k\) is \((k, \frac{i}{n}, -3.2)\).

The distance between center points that correspond to adjacent vertices should be at most 1. For each two vertices in the same clique in \(C^T_k, C^B_k, M^T_k\), and \(M^B_k\), their center points differ only in the \(y\)-coordinate. Since this difference is at most \(1 - 1/n < 1\), they form a clique. For each two adjacent vertices in two different cliques, their center points differ either only in the \(x\)-, or only in the \(z\)-coordinate, by exactly 1, hence, they intersect. For a vertex in \(Q\) and \(M^T_k\), if \(m^T_{k,i}\) exists, the distance between \(m^T_{k,i}\) and \(q_k\) is
\[
\sqrt{(2k - 2k)^2 + (\frac{i}{n} - \frac{1}{2})^2 + (-1 - (1.6))^2} = \sqrt{(\frac{i}{n} - \frac{1}{2})^2 + (0.6)^2} \leq \sqrt{(\frac{1}{2})^2 + (0.6)^2} < 1
\]
The same argument holds for adjacent vertices in \(Q\) and \(M^B\). One can easily check that the non-adjacent vertices have distance strictly greater than 1. ▶

Proof of Theorem 11. The construction creates a set of \(N = O(nd)\) balls in \(O(nd)\) time. If there is an algorithm to solve DIAMETER in \(O(N^{2-\delta})\) time in ball graphs, then we could combine this construction with the algorithm, and solve the ORTHOGONAL VECTORS problem in \(O(nd) + O((nd)^{2-\delta}) = O(n^{2-\delta} \cdot \text{poly}(d))\) time. This contradicts the Orthogonal Vectors Hypothesis, and concludes the theorem. ▶

A simple transformation of this construction shows that we can realize \(G(A)\) also as an intersection graph of axis-parallel unit cubes.

Corollary 14. For all \(\epsilon > 0\), there is no \(O(n^{2-\epsilon})\) time algorithm for solving DIAMETER in intersection graphs of axis-parallel unit cubes in \(\mathbb{R}^3\) under the Orthogonal Vectors Hypothesis.

Proof. Let \(P\) denote the set of centers constructed for unit balls. We rotate \(P\) by \(\pi/4\) around the \(y\) axis, and scale \(P\) by a factor of \(\sqrt{2}\). Let \(P'\) be the resulting set of points. Note that in \(P\), all inter-clique edges were realized by a horizontal or vertical point pair of distance
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 exactly 1. In $P'$, the corresponding pairs are diagonal segments in some plane perpendicular to the $y$-axis, therefore the unit side-length cubes centered at the corresponding pair of points will have a touching edge. It is routine to check that the unit side-length cubes centered at $P'$ realize the intersection graph $G(A)$.

**Proof of Corollary 2.** The lower bounds regarding constant-approximations in sub-quadratic time are immediate consequences of our lower bounds for Diameter-2 and Diameter-3. Notice that our proofs for unit balls and axis-parallel unit cubes in $\mathbb{R}^3$, as well as axis-parallel unit segments in $\mathbb{R}^2$ use a construction where the resulting intersection graph has diameter $d^* = \Theta(d)$. Under OV, there exists no $(1 + \varepsilon)$-approximation for these problems that would run in $n^{2-o(1)}\text{poly}(1/\varepsilon)$ time, as setting $\varepsilon = 1/d^* = \Theta(1/d)$ would enable us to decide OV in $n^{2-o(1)}\text{poly}(d)$ time.

**5 The Diameter-2 problem for hypercube graphs: a hyperclique lower bound**

**Theorem 15.** For all $\varepsilon > 0$ there is no $O(n^{2-\varepsilon})$ algorithm for Diameter-2 in unit hypercube graphs in $\mathbb{R}^d$, unless the Hyperclique Hypothesis fails.

**Proof.** Observe that under the Hyperclique Hypothesis, it requires time $n^{6-o(1)}$ to find a hyperclique of size 6 in a given 3-uniform hypergraph $G = (V, E)$. In fact, using a standard color-coding argument, we can assume without loss of generality that $G$ is 6-partite: We have $V = V_1 \cup \cdots \cup V_6$ for disjoint sets $V_i$ of size $n$ each, and any 6-hyperclique must choose exactly one vertex from each $V_i$. By slight abuse of notation, we view each $V_i$ as a disjoint copy of $[n]$, i.e., node $j \in [n]$ in $V_i$ is different from node $j$ in $V_{i'}$ with $i' \neq i$. Furthermore, by complementing the edge set, we arrive at the equivalent task of determining whether $G$ has an independent set of size 6, i.e., whether there are $(v_1, \ldots, v_6) \in V_1 \times \cdots \times V_6$ such that $\{v_i, v_j, v_k\} \notin E$ for all distinct $i, j, k \in [6]$. Finally, for technical reasons, we assume without loss of generality that for each $v_i \in V_i$ and distinct $j, k \in [6 \setminus \{i\}]$, there are $v_j \in V_j, v_k \in V_k$ with $\{v_i, v_j, v_k\} \in E$: To this end, simply add, for every $\ell \in [6]$, a dummy vertex $v'_\ell$ to $V_{\ell}$, and add, for every $i, j, k$ and $v_j \in V_j, v_k \in V_k$, the edge $\{v'_i, v_j, v_k\} \to E$, i.e., each dummy vertex is connected to all other pairs of vertices (including other dummy vertices). Observe that this yields an equivalent instance, since no dummy vertex can be contained in an independent set.

The reduction is given by constructing a set of $O(n^3)$ unit hypercubes in $\mathbb{R}^d$, which we specify by their centers. These (hyper)cubes are of three types: left-half cubes representing a choice of the vertices $(x_1, x_2, x_3) \in V_1 \times V_2 \times V_3$, right-half cubes representing a choice of the vertices $(y_1, y_2, y_3) \in V_4 \times V_5 \times V_6$ and edge cubes representing an edge $\{v_i, v_j, v_k\} \in E$. In particular, the choice of a vertex in $V_i$ will be encoded in the dimensions $2i - 1$ and $2i$.

Specifically, for each $(x_1, x_2, x_3) \in V_1 \times V_2 \times V_3$ such that $\{x_1, x_2, x_3\} \notin E$, we define the center of the left-half cube $X_{x_1, x_2, x_3}$ as

$$\left(\frac{x_1}{n + 1}, 1 - \frac{x_1}{n + 1}, \frac{x_2}{n + 1}, 1 - \frac{x_2}{n + 1}, \frac{x_3}{n + 1}, 1 - \frac{x_3}{n + 1}, 2, \ldots, 2\right).$$

Similarly, for each $(y_1, y_2, y_3) \in V_4 \times V_5 \times V_6$ such that $\{y_1, y_2, y_3\} \notin E$, we define the center of the right-half cube $Y_{y_1, y_2, y_3}$ as

$$\left(2, \ldots, 2, \frac{y_1}{n + 1}, 1 - \frac{y_1}{n + 1}, \frac{y_2}{n + 1}, 1 - \frac{y_2}{n + 1}, \frac{y_3}{n + 1}, 1 - \frac{y_3}{n + 1}\right).$$
Finally, for each edge \( e = \{v_i, v_j, v_k\} \in E \) not already in \( V_1 \times V_2 \times V_3 \cup V_4 \times V_5 \times V_6 \), we define a corresponding edge cube \( E_{v_i, v_j, v_k} \) with the following center point: We set the 2\( i \) - 1-th coordinate to \( 1 + \frac{v_1}{n+1} \), the 2\( i \)-th coordinate to \( 2 - \frac{v_1}{n+1} \), and similarly we set the coordinates 2\( j \) = \( 1, 2, k \) - 1, 2\( k \) = \( 1 + \frac{v_4}{n+1} \) - 1, with \( 2 - \frac{v_4}{n+1} \), respectively, and we set all remaining coordinates to 1. For example, if \( i = 1, j = 2, k = 4 \), the center point of \( E_{v_1, v_2, v_4} \) is

\[
\left( 1 + \frac{v_1}{n+1}, 2 - \frac{v_1}{n+1}, 1 + \frac{v_2}{n+1}, 2 - \frac{v_2}{n+1}, 1, 1, 1 + \frac{v_4}{n+1}, 2 - \frac{v_4}{n+1}, 1, 1, 1 \right).
\]

Let \( S \) denote the set of all unit cubes \( X_{x_1, x_2, x_3, y_4, y_5, y_6, E_{v_i, v_j, v_k}} \) constructed above and let \( G_S \) denote the geometric intersection graph of the unit cubes. We prove that \( \text{diam}(G_S) \leq 2 \) if and only if there is no independent set \( \{v_1, \ldots, v_6\} \in V_1 \times \cdots \times V_6 \) in the 3-uniform hypergraph \( G = (V_1 \cup \cdots \cup V_6, E) \).

1. **Intra-set distances:** We have that the left- and right-half cubes as well as the edge cubes form cliques, i.e., \( \text{dist}_{G_S}(X_{x_1, x_2, x_3}, X'_{x_1', x_2', x_3'}) \leq 1 \), \( \text{dist}_{G_S}(Y_{y_4, y_5, y_6}, Y'_{y_4', y_5', y_6'}) \leq 1 \) and \( \text{dist}_{G_S}(E_{v_i, v_j, v_k}, E'_{v_i', v_j', v_k'}) \leq 1 \): Observe that the center of each \( X_{x_1, x_2, x_3} \) is contained in \([0,1]^6 \times (0,1)^6 \) and thus in a hypercube of side length at most 1. Thus, all cubes \( X_{x_1, x_2, x_3} \) intersect each other, proving \( \text{dist}_{G_S}(X_{x_1, x_2, x_3}, X'_{x_1', x_2', x_3'}) \leq 1 \). The remaining claims follow analogously by observing that the centers of \( Y_{y_4, y_5, y_6} \) and \( E_{v_i, v_j, v_k} \) are contained in \([0,1]^6 \times (0,1)^6 \) and \([1,2]^12 \), respectively, and thus also in hypercubes of side length at most 1.

2. **Equality checks:** Let \( x_1 \in V_1, x_2 \in V_2, x_3 \in V_3 \) and \( v_i \in V_1, v_j \in V_j, v_k \in V_k \). Then \( \text{dist}_{G_S}(X_{x_1, x_2, x_3}, E_{v_i, v_j, v_k}) = 1 \) if \( v_k = x_k \), whenever \( \ell \in \{1, 2, 3\} \cap \{i, j, k\} \). Consider \( \ell \in \{1, 2, 3\} \cap \{i, j, k\} \). Then the dimensions \( (2\ell - 1, 2\ell) \) of \( X_{x_1, x_2, x_3} \) and \( E_{v_i, v_j, v_k} \) are equal to \((\frac{v_{\ell}}{n+1}, 1) - \frac{v_{\ell}}{n+1}) \) and \((1 - \frac{v_{\ell}}{n+1}, 2 - \frac{v_{\ell}}{n+1}) \), respectively. Note that \( 1 + \frac{v_4}{n+1} - \frac{v_{\ell}}{n+1} \leq 1 \) and \( 2 - \frac{v_4}{n+1} - \frac{v_{\ell}}{n+1} \). All other dimensions \( \ell' \notin \{1, 2, 3\} \cap \{i, j, k\} \) are trivially within distance 1, since dimensions \((2\ell - 1, 2\ell') \) of \( X_{x_1, x_2, x_3} \) and \( E_{v_i, v_j, v_k} \) are \((2, 2) \) and \((1, 2) \), respectively (if \( \ell' \notin \{1, 2, 3\} \)), or in \([0,2]^2 \) and \((1, 1) \), respectively (if \( \ell' \notin \{1, 2, 3\} \)). The analogous claim holds for distances between \( Y_{y_4, y_5, y_6} \) and \( E_{v_i, v_j, v_k} \).

3. **Edge distances:** We have that \( \text{dist}_{G_S}(X_{x_1, x_2, x_3}, E_{v_i, v_j, v_k}) \leq 2 \). By our technical assumption, we have that there is an edge \( \{x_1, v'_4, v''_5\} \in E \) for some vertices \( v'_4 \in V_4 \) and \( v''_5 \in V_5 \). Thus, by the previous properties, we obtain that

\[
\text{dist}_{G_S}(X_{x_1, x_2, x_3}, E_{v_i, v_j, v_k}) \leq \text{dist}_{G_S}(X_{x_1, x_2, x_3}, E_{x_1, v'_4, v''_5}) + \text{dist}_{G_S}(E_{x_1, v'_4, v''_5}, E_{v_i, v_j, v_k}) \leq 2.
\]

4. **Distances of left- and right-half cubes:** Let \( x_1 \in V_1, x_2 \in V_2, x_3 \in V_3 \) and \( y_1 \in V_4, y_2 \in V_5, y_3 \in V_6 \) such that \( \{x_1, x_2, x_3\}, \{y_1, y_2, y_3\} \notin E \) (the left-half/right-half cubes for \( \{x_1, x_2, x_3\}, \{y_1, y_2, y_3\} \) exist). Then we have that \( \text{dist}_{G_S}(X_{x_1, x_2, x_3}, Y_{y_1, y_2, y_3}) > 2 \) if \( \{x_1, x_2, x_3, y_1, y_2, y_3\} \) is an independent set in \( G \): If the tuple \( \{x_1, x_2, x_3, y_1, y_2, y_3\} \) is not an independent set, then there must be an edge \( \{x_i, y_j, y_k\} \) or \( \{x_i, x_j, y_k\} \) with \( i, j, k \in [3] \), since \( \{x_1, x_2, x_3\} \) and \( \{y_1, y_2, y_3\} \) are non-edges. Consider the first case, the other is symmetric. Then by the equality-check property, that \( \text{dist}_{G_S}(X_{x_1, x_2, x_3}, E_{x_1, x_2, x_3}) = 1 \) and \( \text{dist}_{G_S}(E_{x_1, x_2, x_3}, Y_{y_1, y_2, y_3}) = 1 \), which yields \( \text{dist}_{G_S}(X_{x_1, x_2, x_3}, Y_{y_1, y_2, y_3}) \leq 2 \). It remains to consider the case that the tuple \( \{x_1, x_2, x_3, y_1, y_2, y_3\} \) is an independent set. Since there cannot be any edge between a left-half cube \( X_{x_1', x_2', x_3'} \) which is contained in \((0,1)^6 \times (2)^6 \) and a right-half cube \( Y_{y_4', y_5', y_6'} \) which is contained in \([2]^6 \times (0,1)^6 \), the only way to reach \( Y_{y_1, y_2, y_3} \) from \( X_{x_1, x_2, x_3} \) via a path of length 2 would have to use some edge cube \( E_{v_i, v_j, v_k} \). However, by the equality-check property, a path \( X_{x_1, x_2, x_3} - E_{v_i, v_j, v_k} - Y_{y_1, y_2, y_3} \) would...
imply that the vertices chosen by \((x_1, x_2, x_3, y_1, y_2, y_3)\) would agree with \(v_i, v_j, v_k\) in the sets \(V_i, V_j, V_k\). Thus, we would have found an edge \(\{v_i, v_j, v_k\}\) among \((x_1, x_2, x_3, y_1, y_2, y_3)\), contradicting the assumption that it is an independent set.

Finally, observe that given a 3-uniform hypergraph \(G\), we can construct the corresponding cube set \(S\), containing \(O(n^3)\) nodes, in time \(O(n^3)\). Thus, if we had an \(O(N^{2-\epsilon})\)-time algorithm for determining whether an \(N\)-vertex unit cube graph \(G_S\) has a diameter of at most 2, we could detect existence of an independent set (or equivalently, hyperclique) of size 6 in \(G\) in time \(O(n^{6-3\epsilon})\), which would refute the Hyperclique Hypothesis.

References


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