# Linear-Time Approximation Algorithms for Unit Disk Graphs

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**Abstract.** Numerous approximation algorithms for unit disk graphs have been proposed in the literature, exhibiting sharp trade-offs between running times and approximation ratios. We propose a method to obtain linear-time approximation algorithms for unit disk graph problems. Our method yields linear-time  $(4+\varepsilon)$ -approximations to the maximum-weight independent set and the minimum dominating set, bringing dramatic performance improvements when compared to previous algorithms that achieve the same approximation factors. Furthermore, we present an alternative linear-time approximation scheme for the minimum vertex cover, which could be obtained by an indirect application of our method.

# 1 Introduction

A unit disk graph is the intersection graph of n unit disks in the plane. Unit disk graphs are often represented using the coordinates of the disk centers instead of explicit adjacency information. In this geometric setting, two vertices are adjacent if the corresponding points (the disk centers) are within Euclidean distance at most 2 from one another.

Owing to their applicability in wireless networks [10,13], numerous approximation algorithms for unit disk graphs have been proposed in the literature. Such approximations are either graph-based algorithms, when they receive as input solely the adjacency representation of the graph, or geometric algorithms, when the input consists of a geometric representation of the graph. While the m edges of a graph can be obtained from the vertices' coordinates in O(n+m) time under the real-RAM model with floor function and constant-time hashing [3], obtaining a geometric representation of a given unit disk graph is NP-hard [4].

Linear- and near-linear-time approximation algorithms are an active topic of research, even for problems that can be solved exactly in polynomial time, such as maximum flow and matching (see [5] for references). We note that, when the goal is to design O(n)-time algorithms, the geometric representation is required, since the number m of edges in a unit disk graph can be as high as  $O(n^2)$ .

The shifting strategy [7] gave rise to geometric PTASs for several problems for unit disk graphs [8]. Essentially, the shifting strategy reduces the original problem to a set of subproblems of constant diameter. Such reduction takes O(n)

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time and yields a  $(1+\varepsilon)$ -approximation to the original problem, given the exact solutions to the subproblems. However, the running times of the PTASs are polynomials of high degree because each subproblem is solved exactly by exploiting the fact that the point set has constant diameter. Graph-based PTASs for these problems are also known [13]. While they do not use the shifting strategy, their running times are even higher than those of their geometric counterparts.

The minimum dominating set problem (MDSP) admits some PTASs [8,13], the fastest of which is geometric and provides a 4-approximation in roughly  $O(n^{10})$  time. Such high running times have motivated the study of faster constant-factor approximation algorithms. Examples of graph-based algorithms include a 44/9-approximation that runs in O(n+m) time and a 43/9-approximation that runs in  $O(n^2m)$  time [5]. Among the geometric algorithms, we cite the original 5-approximation, which can be implemented in O(n) time if the floor function and constant-time hashing are available [10]; a 44/9-approximation that uses local improvements and runs in  $O(n \log n)$  time [5]; a 4-approximation that uses grids and runs in  $O(n^8 \log n)$  time [6]; and a recent 4-approximation that uses hexagonal grids and runs in  $O(n^6 \log n)$  time [9].

The maximum-weight independent set problem (MWISP) also admits some PTASs, the fastest of which attains a  $(1 + \varepsilon)$ -approximation in  $O(n^{4\lceil 2/\varepsilon\sqrt{3}\rceil})$  time [8,12,13]. A 5-approximation can be obtained in  $O(n\log n)$  time by a greedy approach that considers the vertices in decreasing order of weights. In contrast, for the unweighted version, a greedy approach that considers the vertices from left to right [10] can be implemented to give a 3-approximation in O(n) time with floor function and constant-time hashing.

Some efficient PTASs for unit disk graph problems are also known, as the one given in [11] for the minimum vertex cover problem (MVCP).

Our results. We introduce a method to obtain linear-time approximation algorithms for problems on unit disk graphs and other geometric intersection graphs (Section 2). Our method is based on approximating the input point set, which can be arbitrarily dense, by a *sparse* set of points, that is, a set of points such that any sufficiently small square contains at most a constant number of points.

To convert the general idea into efficient algorithms, we need to investigate the fundamental question of how well a sparse point set—generated using only local information—can approximate a denser one for each considered problem. Although our algorithms share the same basic idea, their analyses differ significantly. For example, the MWISP analysis applies the Four-Color Theorem for planar graphs [2], while the MDSP analysis applies packing arguments.

By using our method, we obtained linear-time  $(4+\varepsilon)$ -approximation algorithms for the MWISP (Section 3) and the MDSP (Section 4). The proposed algorithms provide significant improvements when compared not only to existing linear-time algorithms, but also to sub-quartic-time algorithms (see Table 1 in Section 6). We have also included (Section 5) a linear-time  $(1+\varepsilon)$ -approximation obtained independently for the MVCP, illustrating an indirect application of our method. Open problems and lower bounds to the approximation ratios of our algorithms are also discussed in Section 6.

#### $\mathbf{2}$ Our Method

The shifting strategy [7] is the main idea behind the existing geometric PTASs for problems on unit disk graphs such as the minimum dominating set, maximum independent set, and minimum vertex cover [8]. Generally, the shifting strategy reduces the original problem with n points to a set of subproblems whose inputs have constant diameter and the sum of the input sizes is O(n). Such reduction is based on partitioning the points according to a number of iteratively shifted grids and takes O(n) time (by using the floor function and constant-time hashing). Exploiting the inputs' constant diameter, each subproblem is solved exactly in polynomial time. The solutions to the subproblems are then combined appropriately (normally in O(n) time) to yield feasible solutions to the original problem, the best of which is returned. The high complexities of these geometric PTASs are due to the exact algorithms that are employed to solve each subproblem.

We propose a method that is based on the shifting strategy. It presents, however, a crucial difference. Rather than obtaining exact, costly solutions for the subproblems, we solve each subproblem approximately. To do that, we employ the coresets paradigm [1], where only a subset with a constant number of input points is considered. For a problem whose input is a set P of n points, our method can be briefly described as follows:

- 1. Apply the shifting strategy to construct a set of r subproblems with inputs
- $P_1, \ldots, P_r$  such that  $\sum_{i=1}^r |P_i| = O(n)$  and  $\operatorname{diam}(P_i) = O(1)$  for all i. 2. For each subproblem instance  $P_i$ , obtain a coreset  $Q_i \subseteq P_i$  with  $|Q_i| = O(1)$ , such that the optimal solution for instance  $Q_i$  is an  $\alpha$ -approximation to the optimal solution for instance  $P_i$ .
- 3. Solve the problem exactly for each  $Q_i$ .
- 4. Combine the solutions into an  $(\alpha+\varepsilon)$ -approximation for the original problem.

Coresets for different problems must be devised appropriately. For the MWISP, we create a grid with cells of diameter 0.29 and consider only one point of maximum weight inside each cell. For the MDSP, we create a grid with cells of diameter 0.24 and consider only the (at most four) points, inside each cell, with minimum or maximum coordinate in either dimension (breaking ties arbitrarily). Finally, we solve the MVCP by breaking each subproblem into two cases. In the first one, the number of input points is already bounded by a constant. In the second one, we use the same coreset as in the MWISP.

We assume a real-RAM computation model with floor function and constanttime hashing (as in [3]), so we can partition the input points into grid cells efficiently, yielding an overall O(n) running time for our method. Without these operations, the running time of our algorithms becomes  $O(n \log n)$ . We also assume that  $\varepsilon$  is constant. Otherwise, the running time becomes  $2^{O(1/\varepsilon^2)}n$  for the WIS and the DS on UDGs, and  $2^{O(1/\varepsilon^3)}n$  for the VC.

### 3 Maximum-Weight Independent Set

In this section, we show how to obtain a linear-time  $(4 + \varepsilon)$ -approximation to the MWISP. We start by presenting a 4-approximation for point sets of constant diameter, and then we use the shifting strategy to obtain the desired  $(4 + \varepsilon)$ -approximation.

Given a point p and a set S of points, let w(p) denote the weight of p, and let  $w(S) = \sum_{p \in S} w(p)$ . We say two or more points are *independent* if their minimum distance is strictly greater than 2.

**Theorem 1** Given a set P of n points with real weights as input, with diam(P) = O(1), the MWISP can be 4-approximated in O(n) time in the real-RAM.

*Proof.* Our algorithm proceeds as follows. First, we find the points of P with minimum or maximum coordinates in either dimension. That defines a bounding box of constant size for P. Within this bounding box, we create a grid with cells of diameter  $\gamma = 0.29$  (any value  $\gamma < (2 - \sqrt{2})/2$  suffices). Note that the number of grid cells is constant, and therefore the points of P can be partitioned among the grid cells in O(n) time (even without using the floor function or hashing). Then, we build the subset  $Q \subseteq P$  as follows. For each non-empty grid cell C, we add to Q a point of maximum weight in  $P \cap C$ . Afterwards, we determine the maximum-weight independent set  $I^*$  of Q. Since |Q| = O(1), this can be done in constant time. We return the solution  $I^*$ .

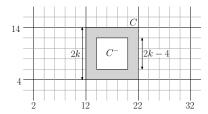
Next, we show that  $I^*$  is indeed a 4-approximation. We argue that, given an independent set  $I\subseteq P$ , there is an independent set  $I'\subseteq Q$  with 4  $w(I')\ge w(I)$ . Given a point  $p\in P$ , let q(p) denote the point from Q that is contained in the same grid cell as p. Consider the set  $S=\{q(p):p\in I\}$ . Note that  $w(q(p))\ge w(p)$  and  $w(S)\ge w(I)$ . The set S may not be independent, but since I is independent, the minimum distance in S is at least  $2-2\gamma=1.42>\sqrt{2}$ . We claim that the unit disk graph formed by S is a planar graph. To prove the claim, we show that a planar drawing can be obtained by connecting the points of S within distance at most 2 by straight line segments. Given a pair of points  $p_1, p_2$  with distance  $\|p_1p_2\|\le 2$ , the Pythagorean Theorem shows that a unit disk centered within distance greater than  $\sqrt{2}$  from both  $p_1$  and  $p_2$  cannot intersect the segment  $p_1p_2$ . By the Four-Color Theorem [2], S admits a partition into four independent sets  $S_1,\ldots,S_4$ . The set I' of maximum weight among  $S_1,\ldots,S_4$  must have weight at least w(I)/4.

Since  $I^*$  is the maximum-weight independent set of Q, we have that  $I^*$  is a 4-approximation for the MWISP.

The following theorem uses the shifting strategy to obtain a  $(4 + \varepsilon)$ -approximation for point sets of arbitrary diameter. The proof uses the ideas from [8], presented in a different manner and including details about an efficient implementation of the strategy.

**Theorem 2** Given a set P of n points in the plane as input, the MWISP can be  $(4+\varepsilon)$ -approximated in O(n) time on a real-RAM with constant-time hashing and the floor function. Without these operations, it can be done in  $O(n \log n)$  time.

<sup>&</sup>lt;sup>3</sup> Note that the Four-Color Theorem is only used in the argument, and no coloring is ever computed by the algorithm.



**Fig. 1.** Grid rooted at (2,4) with k=5 and the contraction of a cell

*Proof.* Let k be the smallest integer such that

$$\left(\frac{k-2}{k}\right)^2 \ge \frac{4}{4+\varepsilon}.\tag{1}$$

Throughout this proof, we consider grids with square cells of side 2k. We say a grid is rooted at a point (x,y) if there is a grid cell with corner at (x,y). Given a cell C, the square region  $C^- \subset C$ , called the *contraction* of C, is formed by removing from C the points within distance at most 2 from the boundary of C. Figure 1 illustrates these concepts.

The algorithm proceeds as follows. For i, j from 0 to k-1, we create a grid with cells of side 2k rooted at (2i, 2j). For each cell C in the grid, we run the MWISP 4-approximation algorithm from Theorem 1 with point set  $P \cap C^-$ , obtaining a solution  $I_{i,j}(C)$ . Then, the independent set  $I_{i,j}$  is constructed as the union of the independent sets  $I_{i,j}(C)$  for all grid cells C. We return the maximum-weight set  $I_{i,j}$  that is found, call it  $I^*$ .

To implement the algorithm efficiently, we create a subgrid of subcells of side 2, assigning each point to the subcell that contains it. In order to partition the n points into subcells, we use the floor function and constant-time hashing, taking O(n) time. If these operations are not available, we determine the connected components of the graph (using the Delaunay triangulation, for example) and for each component we partition the points into subcells by sorting them by x coordinate, separating them into columns, and then sorting the points inside each column by y coordinate. The non-empty subcells are stored in a balanced binary search tree. This process takes  $O(n \log n)$  time due to sorting, Delaunay triangulation, and binary search tree operations. Given the partitioning of the point set into subcells, each input to the MWISP algorithm can be constructed as the union of a constant number of subcells. Finally, the total size of the constant-diameter MWISP instances is O(n), since each point from the original point sets appears in a constant number—a function of the fixed  $\varepsilon$ —of such instances.

To prove that the returned solution  $I^*$  is indeed a  $(4 + \varepsilon)$ -approximation, we use a probabilistic argument. Let i, j be picked uniformly at random from  $0, \ldots, k-1$  and let OPT denote the optimal solution. For every cell C, we have

$$w(I_{i,j}(C)) \ge \frac{w(OPT \cap C^-)}{4}.$$

Consequently, by summing over all grid cells,

$$w(I_{i,j}) = \sum_{C} w(I_{i,j}(C)) \ge \frac{1}{4} \sum_{C} w(OPT \cap C^{-}).$$

We now bound  $E[w(I_{i,j})]$ . Let  $\rho(p)$  denote the probability that a given point p is contained in some contracted cell. Since w(p) does not depend on the choice of i, j, we can write

$$4 \operatorname{E}[w(I_{i,j})] \geq \operatorname{E}\left[\sum_{C} w\left(OPT \cap C^{-}\right)\right] = \sum_{p \in OPT} \rho(p)w(p).$$

Note that, for all  $p \in P$ ,  $\rho(p)$  corresponds to the ratio between the areas of  $C^-$  and C, namely

$$\rho(p) = \frac{\operatorname{area}(C^{-})}{\operatorname{area}(C)} = \left(\frac{k-2}{k}\right)^{2}.$$

Therefore, by using inequality (1), we obtain

$$\mathbb{E}[w(I_{i,j})] \ge \frac{1}{4} \left(\frac{k-2}{k}\right)^2 \sum_{p \in OPT} w(p) \ge \frac{1}{4} \left(\frac{4}{4+\varepsilon}\right) w(OPT) = \frac{1}{4+\varepsilon} w(OPT).$$

Since  $I^*$  has maximum weight among the independent sets  $I_{i,j}$ , it follows that  $w(I^*)$  is at least as large as their average weight. Therefore,  $I^*$  satisfies

$$w(I^*) \ge \mathrm{E}[w(I_{i,j})] \ge \frac{1}{4+\varepsilon} \ w(OPT),$$

closing the proof.

# 4 Minimum Dominating Set

In this section, we show how to obtain a linear-time  $(4 + \varepsilon)$ -approximation to the MDSP (in fact, a generalization of it). We start by presenting a 4-approximation for point sets of constant diameter, and then we use the shifting strategy to obtain the desired  $(4 + \varepsilon)$ -approximation. We say that a point p dominates a point q if  $||pq|| \le 2$ . Given two sets of points p and p', we say that p is a p'-dominating set if every point in p' is dominated by some point in p.

We now define a more general version of the MDSP, which we refer to as the minimum partial dominating set problem (MPDSP). Such a generalization is necessary to properly apply the shifting strategy. In the MPDSP, we are given a set P of n points and also a subset  $P' \subseteq P$ . The goal is to find the smallest P'-dominating subset  $D \subseteq P$ .

In order to analyze our algorithm, we prove a geometric lemma that shows that the set-theoretic difference between a unit circle and two unit disks that are sufficiently close to it and form a sufficiently big angle consists of one or two "small" arcs. Given a point p, let  $\bigcirc_p$  denote the unit disk centered at p, and  $\partial \bigcirc_p$  denote its boundary circle.

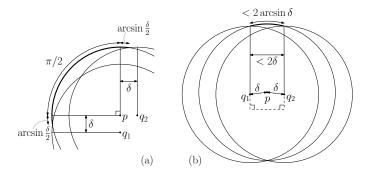


Fig. 2. Proof of Lemma 3

**Lemma 3** Given  $\delta > 0$  and three points  $p, q_1, q_2 \in \mathbb{R}^2$  with (i)  $||pq_1|| \leq \delta$ , (ii)  $||pq_2|| \leq \delta$ , and (iii) the smallest angle  $\angle q_1pq_2$  is greater than or equal to  $\pi/2$ , we have that:

- (1) the portion  $T = (\partial \bigcirc_p) \setminus (\bigcirc_{q_1} \cup \bigcirc_{q_2})$  of the boundary  $\partial \bigcirc_p$  consists of one or two circular arcs;
- (2) if T consists of one circular arc, then the arc length is less than or equal to  $\pi/2 + 2\arcsin(\delta/2)$ ; and
- (3) if T consists of two circular arcs, then each arc length is less than  $2 \arcsin \delta$ .

Proof. Statement (1) is clearly true. We start by proving statement (2). The arc length ||T|| is maximized as the angle  $\angle q_1pq_2$  decreases while the distances  $||pq_1||, ||pq_2||$  are kept constant, therefore it suffices to consider the case when  $\angle q_1pq_2 = \pi/2$ . The arc T centered at p can be decomposed into three arcs by rays in directions  $q_1p$  and  $q_2p$ , as shown in Figure 2(a). The central arc measures  $\pi/2$ , while each of the other two arcs measures  $\arcsin(\delta/2)$ , proving statement (2).

Next, we prove statement (3). Let  $T_1, T_2$  denote the two arcs that form T with  $||T_1|| \ge ||T_2||$ . The arc length  $||T_1||$  is maximized in the limit when  $||T_2|| = 0$ , as shown in Figure 2(b). The rays connecting  $q_1$  and  $q_2$  to the two extremes of  $T_1$  are parallel, and therefore  $||T_1|| < 2 \arcsin \delta$ .

We are now able to prove the following theorem, which presents our 4-approximation algorithm for point sets of constant diameter.

**Theorem 4** Given two sets of points P and P' as input, with  $P' \subseteq P$ , |P| = n, and diam(P) = O(1), the MPDSP can be 4-approximated in O(n) time in the real-RAM.

*Proof.* First, we determine a bounding box of constant size for P, as we did in the algorithm for the MWISP. Within this bounding box, we create a grid with cells of diameter  $\gamma = 0.24$  (any positive  $\gamma$  satisfying

$$\sqrt{8 - 8\cos\left(\frac{\frac{\pi}{2} + 2\arcsin(\frac{\gamma}{2})}{2}\right)} + \gamma < 2$$

suffices). Note that the number of grid cells is constant, and therefore the points of P can be partitioned among the grid cells in O(n) time (even without using the floor function or hashing). Then, we build the subset  $Q \subseteq P$  as follows. For each non-empty grid cell, we add to Q the (at most four) extreme points inside the cell, i.e., those presenting minimum or maximum coordinate in either dimension. Ties are broken arbitrarily. Since there is a constant number of grid cells and we include in Q at most four points per cell, we have |Q| = O(1). Afterwards, we determine the smallest P'-dominating subset  $D^* \subseteq Q$ . To do that, we examine the subsets of Q, from smallest to largest, verifying if all points of P' are dominated, until we find the dominating set  $D^*$ , which is returned as the approximate solution. Since Q has a constant number of points, this procedure takes O(n) time.

Now we show that the returned solution  $D^*$  is indeed a 4-approximation. We argue that, given a P'-dominating set  $D \subseteq P$ , there is a P'-dominating set  $D' \subseteq Q$  with  $|D'| \le 4$  |D|. To build the set D' from D, we proceed as follows. For each point  $p \in D$ , if  $p \in Q$ , we add p to D'. Otherwise, since the set Q contains points of extreme coordinates in both x and y axes, in the cell of p, there are two points  $q_1, q_2 \in Q$  such that (i)  $||pq_1|| \le \gamma$ , (ii)  $||pq_2|| \le \gamma$ , and (iii) the smallest angle  $\angle q_1pq_2$  is at least  $\pi/2$ . We add these two points  $q_1, q_2$  to D'.

By Lemma 3, the portion  $T = (\partial \bigcirc_p) \setminus (\bigcirc_{q_1} \cup \bigcirc_{q_2})$  of  $\partial \bigcirc_p$  consists of one or two circular arcs. We first consider the case where T consists of one circular arc. Let R be the set of points from P' which are dominated by p, but not by  $q_1$  or  $q_2$ . If R is empty, then no extra point needs to be added to D'. Otherwise, the line  $\ell$  which contains p and bisects T separates R into two (possibly empty) sets  $R_1, R_2$ . If  $R_1 \neq \emptyset$ , let  $p_3$  be an arbitrary point of  $R_1$ . Since Q contains a point in the same cell as  $p_3$ , there is a point  $q_3$  with  $||p_3q_3|| \leq \gamma$ . We add the point  $q_3$  to D'. Analogously, if  $R_2 \neq \emptyset$ , let  $p_4$  be an arbitrary point of  $R_2$  and let  $q_4 \in Q$  be a point with  $||p_4q_4|| \leq \gamma$ . We add the point  $q_4$  to D'.

We now show that the four points  $q_1,q_2,q_3,q_4\in Q$  dominate all points dominated by p. Consider a point v that is dominated by p but not by  $q_1$  or  $q_2$ . The point v must be inside the circular crown sector depicted in Figure 3(a) and described as follows. Because v is dominated by p, we have  $\|pv\| \leq 2$ . By Lemma 3, the arc length  $\|T\| < 1.82$ . Also,  $\|pv\| > 1$ , because otherwise the unit circles centered at p and v would intersect forming an arc of length at least  $2\pi/3$ , which is greater than  $\|T\|$ , in which case v is dominated by  $q_1$  or  $q_2$ . Finally, since v is closer to p than it is to  $q_1$  or  $q_2$ , it follows that v must be between the lines that connect p to the endpoints of T. This circular crown sector is bisected by the line  $\ell$ . Using the law of cosines, we calculate the diameter of each circular crown sector as  $d = \sqrt{8-8\cos(\|T\|/2)} < 1.76$ . Therefore, for any point v inside the circular crown sector, the point  $q_3$  (or  $q_4$ , analogously) that is within distance at most  $\gamma$  from a point inside the same sector dominates v, as  $\|vq_3\| \leq d + \gamma < 2$ .

Finally, if T consists of two circular arcs  $T_1, T_2$  centered in p, then we start by adding those same points  $q_1, q_2$  to D', as if T consisted of only one arc. Then, if necessary, we add new points  $q_3, q_4$  to D' as follows. The points that are dominated by p but not by  $q_1$  or  $q_2$  must be within distance 1 of either  $T_1$  or  $T_2$ .

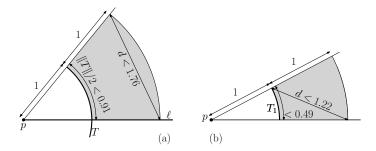


Fig. 3. Proof of Theorem 4

Let  $p_3, p_4$  be arbitrary points that are within distance 1 of  $T_1$  or  $T_2$ , respectively, but are not dominated by  $q_1$  or  $q_2$ . If such points  $p_3, p_4$  exist, then there are two points  $q_3, q_4$  in Q that are within distance at most  $\gamma$  from respectively  $p_3, p_4$ . By Lemma 3, the largest arc among  $T_1, T_2$  measures at most 0.49. The proof that all points dominated by p are dominated by  $q_1, q_2, q_3$ , or  $q_4$  is analogous to the case where T consists of a single arc, using the circular crown sector illustrated in Figure 3(b).

Since  $D^*$  is minimum among all subsets of Q that are P'-dominating sets,  $D^*$  is a 4-approximation for the MPDSP.

The following theorem uses the shifting strategy [8] to obtain a  $(4 + \varepsilon)$ -approximation for point sets of arbitrary diameter.

**Theorem 5** Given two sets of points P and P' as input, with  $P' \subseteq P$  and |P| = n, the MPDSP can be  $(4 + \varepsilon)$ -approximated in O(n) time on a real-RAM with constant-time hashing and the floor function. Without these operations, it can be done in  $O(n \log n)$  time.

*Proof.* Let k be the smallest integer such that

$$\left(\frac{k+2}{k}\right)^2 \le 1 + \frac{\varepsilon}{4}.$$

We consider grids with square cells of side 2k. We say a grid is *rooted at* a point (x, y) if there is a grid cell with corner at (x, y). Given a cell C, the square region  $C^+$ , called the *expansion* of C, is formed by C and all points within  $L_{\infty}$  distance at most 2 from C.

The algorithm proceeds as follows. For i,j from 0 to k-1, we create a grid with cells of side 2k rooted at (2i,2j) and, for each cell C in the grid, we use Theorem 4 to 4-approximate the MPDSP with point sets  $P \cap C^+, P' \cap C$ , obtaining a solution  $D_{i,j}(C)$ . The dominating set  $D_{i,j}$  is constructed as the union of the dominating sets  $D_{i,j}(C)$  for all grid cells C. We return the smallest dominating set  $D_{i,j}$  that is found, call it  $D^*$ . The remainder of the proof is similar to the proof of Theorem 2 and is omitted due to space limitations.  $\square$ 

The MDSP is the special case of the MPDSP in which P' = P, and thus it can be  $(4 + \varepsilon)$ -approximated in linear time by the same algorithm.

## 5 Minimum Vertex Cover

In this section, we show how to obtain a linear-time approximation scheme to the MVCP. We start by presenting an approximation scheme for point sets of constant diameter, and then we use the shifting strategy to generalize the result to arbitrary diameter. Differently than in the previous two problems, the size of a minimum vertex cover for a point set of constant diameter is not upper bounded by a constant. Therefore, strictly speaking, a coreset for the problem does not exist. Nevertheless, it is possible to use coresets to approach the problem indirectly.

Given a graph G=(V,E) with n vertices, it is well known that I is an independent set if and only if  $V\setminus I$  is a vertex cover. While a maximum independent set corresponds to a minimum vertex cover, a constant approximation to the maximum independent set does not necessarily correspond to a constant approximation to the minimum vertex cover. However, in certain cases, an even stronger correspondence holds, as we show in the following proof.

**Theorem 6** Given a set P of n points as input, with  $\operatorname{diam}(P) = O(1)$ , the MVCP can be  $(1+\varepsilon)$ -approximated in O(n) time in the real-RAM, for constant  $\varepsilon > 0$ .

*Proof.* Our algorithm considers two cases, depending on the value of n. If

$$n < \left(1 + \frac{3}{4\varepsilon}\right) \frac{(\operatorname{diam}(P) + 2)^2}{4},$$

then n is constant, and we can solve the MVCP optimally in constant time.

Otherwise, we use Theorem 1 to obtain a 4-approximation I to the maximum independent set. We now show that  $V=P\setminus I$  is a  $(1+\varepsilon)$ -approximation to the minimum vertex cover. Let  $I_{OPT}, V_{OPT}$  respectively be the maximum independent set and the minimum vertex cover. Note that |V|=n-|I| and  $|V_{OPT}|=n-|I_{OPT}|$ . By a simple packing argument, dividing the area of a disk of diameter diam(P)+2 by the area of a unit disk,

$$|I_{OPT}| \le \frac{(\operatorname{diam}(P) + 2)^2}{4},$$

and consequently

$$n \ge \left(1 + \frac{3}{4\varepsilon}\right)|I_{OPT}| = \left(1 + \frac{3}{4\varepsilon}\right)(n - |V_{OPT}|).$$

Manipulating the previous inequality, we obtain

$$n \le \frac{4\varepsilon + 3}{3} |V_{OPT}|. \tag{2}$$

Since I is a 4-approximation to  $I_{OPT}$ ,

$$|V| = n - |I| \le n - \frac{|I_{OPT}|}{4} = \frac{4n - |I_{OPT}|}{4} = \frac{3n + |V_{OPT}|}{4}.$$
 (3)

Combining (2) and (3), we can write  $|V| \leq (1+\varepsilon)|V_{OPT}|$ , as desired.

previous / new results	MWISP		MDSP		MVCP	
previous approximation ratio in $o(n^4)$ time	5	[10]	4.889	[5]	$1 + \varepsilon$	[11]
our approximation ratio in $O(n)$ time	$4+\varepsilon$		$4 + \varepsilon$		$1 + \varepsilon$	
previous time for the same approximation	$O(n^4)$	[12]	$O(n^6 \log n)$	[9]	O(n)	[11]

**Table 1.** Comparison of new and previous approximation algorithms

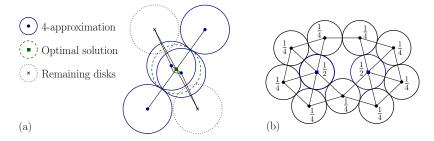
Using the shifting strategy we obtain the following result. The proof is similar to that of Theorem 2 and is omitted due to space limitations.

**Theorem 7** Given a set P of n points in the plane as input, the MVCP can be  $(1 + \varepsilon)$ -approximated in O(n) time on a real-RAM with constant-time hashing and the floor function, for constant  $\varepsilon > 0$ . Without these operations, it can be done in  $O(n \log n)$  time.

# 6 Conclusion

We introduced a method to obtain linear-time approximation algorithms for problems on unit-disk graphs and other geometric intersection graphs. The central idea of the method is a technique to obtain approximate solutions when the inputs are point sets of constant diameter. For the MWISP and the MDSP, the proposed algorithms provide improved approximation factors when compared not only to existing linear-time algorithms, but also to sub-quartic-time algorithms, as shown in Table 1.

While the approximation ratio for the MWISP and the MDSP is 4 (for constant diameter inputs), we only know that the analysis is tight for the MDSP. Figure 4(a) shows an MDSP instance where our algorithm does not achieve an approximation ratio better than 4, even if we reduce the grid size and search for extreme points in a larger number of directions. In contrast, for the MWISP, the best lower bound we are aware of is 3.25, as shown in the following example. Let  $P_1$  be the weighted point set from Figure 4(b), where all adjacent vertices are at



 $\mathbf{Fig. 4.}$  (a) Example where the approximation ratio for the MDSP is exactly 4 (b) Coin graph used in the example where the approximation ratio for the MWISP is 3.25

distance exactly 2. Create another set  $P_2$  by multiplying the coordinates of the points in  $P_1$  by  $1 + \varepsilon$ , while multiplying their weights by  $1 - \varepsilon$ , for arbitrarily small  $\varepsilon > 0$ . The set  $P_2$  forms an independent set of weight just smaller than 3.25, while the maximum independent set in  $P_1$  has weight 1. Since each vertex in  $P_2$  has a smaller weight and is arbitrarily close to a vertex of  $P_1$ , the vertices of  $P_2$  will be disregarded by the algorithm for the input instance  $P_1 \cup P_2$ .

Several open problems remain. Can we obtain an approximation ratio better than 4 in (close to) linear time for the MWISP, or at least for its unweighted version? Can the linear-time approximation scheme for the MVCP be generalized for the weighted version? Are the point coordinates really necessary, or is it possible to devise similar graph-based algorithms? Also, can we use our method to obtain better linear-time approximations to related problems on unit disk graphs such as finding the minimum-weight dominating set or the minimum connected dominating set?

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