PTAS for connected vertex cover in unit disk graphs
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ABSTRACT
This paper gives the first polynomial time approximation scheme for the connected vertex cover problem in unit disk graphs.
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1. Introduction

Minimum Vertex Cover Problem (MVC) is a classical combinatorial optimization problem. For an undirected graph \( G = (V, E) \), a subset \( C \subseteq V \) is a vertex cover (VC) of \( G \) if every edge of \( G \) has at least one end in \( C \). MVC is to find a vertex cover \( C \) of \( G \) with the minimum number of vertices. This problem has many real-world applications, including many in the field of bioinformatics [2]. It can be used in the construction of polygenetic trees, in phenotype identification, and in analysis of microarray data. MVC has been studied extensively in the literature [7]. It is well known that MVC is \( \mathcal{NP} \)-hard [8]. Papadimitriou et al. [12] proved that VC is APX-complete. Bar-Yehuda and Even [2] and Monien et al. [11] gave approximation algorithms for VC with performance ratio \( 1 - (\log \log n)/(2 \log n) \), where \( n \) is the number of vertices.

If furthermore, the subgraph \( G[C] \) of \( G \) induced by a vertex cover \( C \) is connected, then \( C \) is a connected vertex cover (CVC). The Minimum Connected Vertex Cover Problem (MCVC) is to find a CVC with minimum cardinality. MCVC problem is an enforced version of MVC when certain connectivity constraints are needed in some applications. For example, in routing and wavelength assignment (RWA) problem for optical networks, people select a suitable path and wavelength among the many possible choices with the help of CVC. MCVC is also \( \mathcal{NP} \)-hard [6]. The currently best known approximation algorithms for MCVC have performance ratio 2 [1,5,13].

In this paper, we consider the MCVC problem in Unit Disk Graphs (UDG). A graph \( G \) is a UDG if each vertex of \( G \) is associated with the center of a disk with diameter 1 on the plane, and two vertices \( u, v \) of \( G \) are adjacent if and only if the two disks corresponding to \( u \) and \( v \) have non-empty intersection. In other words, \((u, v) \in E(G)\) if and only if the Euclidean distance between the centers corresponding to \( u \) and \( v \) is at most 1. Such a set of unit disks on the plane is called the geometric representation of \( G \). When talking about a unit disk graph in this paper, we assume that the geometric representation is given, since it has been proved in [9] that determining whether a graph is a UDG is \( \mathcal{NP} \)-complete. UDG is widely used in wireless networks, where each vertex represents an idealized multi-hop radio based station, and the corresponding disk is the communication range of the station. For MVC in UDG, there exists a polynomial time approximation scheme (PTAS). That is, for any positive real number \( \epsilon \), there exists a \((1 + \epsilon)\)-approximation algorithm. In fact, Erlebach et al. [4,10] presented a PTAS for Minimum Weight Vertex Cover (MWVC) in Disk Graph (DG), where DG is a generalization of UDG, in which disks have different radiiuses.

In this paper, we present the first PTAS for CVC on UDG, using partition technique and shifting strategy. Such an approach was used for Steiner trees in the plane [14]. A more complicated approach was used for Minimum Connected Dominating Set [3]. It should be noted that in [3], the technique is heavily based on the property of 2-dimension. In other words, the
technique cannot be applied to higher-dimensional space, e.g., unit ball graphs, which is also an important model for wireless sensor networks. The technique presented in this paper can be applied to any dimension. Therefore, it is actually proved in this paper that there exists PTAS in unit n-dimensional ball graphs for any n.

The idea of the algorithm is: First, we take an area containing all vertices of the graph, and partition it into small squares. For each small square, define the inner area and the boundary area, such that the inner area and the boundary area of the same small square has an overlap. For each component of the inner area, compute a minimum CVC. To cover edges not in the inner area, use a constant-approximation algorithm to compute a connected vertex cover \( C_0 \) of \( G \), and unite those vertices of \( C_0 \) which belong to the ‘boundary area’ of the partition into the above CVCs. The overlap of the inner area and the boundary area ensures the connectivity of the output. The shifting strategy is used to select a partition such that the number of vertices of \( C_0 \) falling into the boundary area of this partition is small enough relative to \( \epsilon \). From appearance, our algorithm is similar to that in [3]. However, the analysis of performance ratio is much different, which is why our method can be applied to any dimension.

The rest of this paper is organized as follows. In Section 3, we introduce some terminologies used to describe the algorithm. In Section 3, the algorithm is presented. In Section 4 we show the correctness of our algorithm, analyze the time complexity, and prove that it is a PTAS. A conclusion is given in Section 5.

2. Preliminaries

In this section we introduce the symbols and definitions used for the description of the algorithm.

For a given UDG \( G = (V, E) \), where \( |V| = n \), we assume that all the disks are located in a plane square \( Q = (x, y) | 0 \leq x \leq q, 0 \leq y \leq q \) where \( q \) is related to \( n \). Using partition strategy, we divide \( Q \) into small squares each with side length \( m \times m \). Set \( m = \lceil \frac{48p}{\epsilon} \rceil \), where \( p \) is a constant which is the approximation ratio of an APX for CVC (for example, \( p \) can be taken as 2 if we use the 2-approximation algorithm in [5]), and \( \epsilon \) is an arbitrary positive number. Let \( p = \lceil \frac{48p}{\epsilon} \rceil + 1 \). Since we shall use shifting policy, we widen \( Q \) into a bigger region \( \tilde{Q} = (x, y) | m \leq x \leq pm, -m \leq y \leq pm \). Name this partition as \( P(0) \), and denote by \( P(a) \) the partition obtained from \( P(0) \) by shifting it such that the left-bottom corner of \( P(a) \) is at \( (a - m, a - m) \), for \( a = 0, 1, \ldots, m - 1 \).

For each square \( e \), we define the inner area \( I_e \) and boundary area \( B_e \) as follows. Suppose \( e = (x, y) | im \leq x \leq (i + 1)m, jm \leq y \leq (j + 1)m \). Define

\[
I_e = (x, y) | im + 1 \leq x \leq (i + 1)m - 1, jm + 1 \leq y \leq (j + 1)m - 1,
\]

\[
B_e = e - (x, y) | im + 2 \leq x \leq (i + 1)m - 2, jm + 2 \leq y \leq (j + 1)m - 2.
\]

Note that \( I_e \) and \( B_e \) have an overlap of width 1. This is to ensure the connectedness of the vertex cover computed by our algorithm.

3. Algorithm overview

For a partition \( P(a) \), denote by \( B(P(a)) = \bigcup_{e \in P(a)} B_e \). The algorithm is executed in two phases.

**Phase I** Use a \( \rho \)-approximation to compute a CVC \( C_0 \) for graph \( G \). Let \( C_0(a) = C_0 \cap B(P(a)) \) be the set of vertices of \( C_0 \) lying in the boundary area of partition \( P(a) \). Choose \( a^* \) such that \( |C_0(a^*)| = \min |C_0(a)| \).

**Phase II** For any square \( e \in P(a^*) \), denote by \( G_e \) the subgraph of \( G \) induced by the vertices in \( I_e \), and \( \text{Comp}(G_e) \) the set of connected components in \( G_e \). For each square \( e \) and each component \( H \in \text{Comp}(G_e) \), use exhaust search to find a minimum CVC \( C_H \) of \( H \). Set \( C_e = \bigcup_{H \in \text{Comp}(G_e)} C_H \).

**Final result** Output \( C = C_0(a^*) \cup \bigcup_{e \in P(a^*)} C_e \).

4. Analysis of the algorithm

In this section, we firstly prove the correctness of our algorithm. Then we analyze the time complexity, showing that our algorithm runs in polynomial time. Finally, we prove the performance ratio of the algorithm, which is \( (1 + \epsilon) \).

4.1. Correctness

To prove that the output \( C \) of our algorithm is a CVC for graph \( G \), we firstly prove that \( C \) is a vertex cover for \( G \), then prove that the induced subgraph \( G[C] \) is connected.

**Lemma 1.** \( C \) is a vertex cover for \( G = (V, E) \).

**Proof.** For each square \( e \), the inner area \( I_e \) and the boundary area \( B_e \) have an overlap with width 1. Since for any edge \( (v, w) \), the Euclidean distance between \( v \) and \( w \) is less than or equal to 1, we see that either both \( v \) and \( w \) belong to the inner area \( I_e \) for some square \( e \), or both \( v \) and \( w \) belong to the boundary area \( B(P(a^*)) \). In the former case, the edge \( (v, w) \) is in a component \( H \) of \( G_e \). By Phase II of the algorithm, either \( v \in C_H \) or \( w \in C_H \), meaning that \( (v, w) \) is covered by \( C_e \subseteq C \). In
the second case, by Phase I of the algorithm, since $C_0$ is a CVC of $G$, we have either $v \in C_0(a^*)$ or $w \in C_0(a^*)$, meaning that $(v, w)$ is covered by $C_0(a^*) \subseteq C$. Thus we have proved that any edge in $G$ is covered by $C$. So $C$ is a vertex cover of $G$. □

**Lemma 2.** The induced subgraph $G[C]$ is connected.

**Proof.** We prove this lemma by two steps. In step 1, we show that distinct connected components in $G[C_0(a^*)]$ (if they exist) can be connected through vertices in $\bigcup_{e \in P(a^*)} C_e$. In step 2, we show that there are no other components of $G[C]$ left after step 1.

**Step 1.** Let $H_1$ and $H_2$ be two distinct components in $G[C_0(a^*)]$ which are ‘closest’ in $G[C_0]$ with each other. Then, there is a path $P = (v_1, v_2, \ldots, v_t)$ of $G[C_0]$ connecting $H_1$ and $H_2$ through the inner area of ‘one’ square $e$. Without loss of generality, we may assume that $v_1 \in V(H_1)$, $v_t \in V(H_2)$ and $\{v_2, \ldots, v_{t-1}\} \subseteq I_e$ (see Fig. 1).

It is easy to see that $v_1$ and $v_t$ belong to $I_e \cap B_e$, so $P$ is in a connected component $H$ of $G$. Based on Phase II of our algorithm, $P$ is covered by $C_H$. It follows that $\{v_1 \in C_H \text{ or } v_2 \in C_H\}$, and $\{v_t \in C_H \text{ or } v_1 \in C_H\}$. Since $G[C_H]$ is connected, we see that $H_1$ and $H_2$ are connected through $G[C_H]$.

**Step 2.** Let $\tilde{G}$ be the component of $G[C]$ containing all vertices of $C_0(a^*)$. Such a $\tilde{G}$ exists because of step 1. Suppose $\tilde{G} \neq G[C]$, then there exist a square $e$ and a connected component $H$ of $G_e$ such that:

(i) $C_H \cap C_0(a^*) = \emptyset$ and

(ii) no vertex of $C_H$ is adjacent with any vertex in $C_0(a^*)$.

Let $x$ be a vertex in $C_H$. Then either $x \in C_0$ or $x$ is adjacent with a vertex $y \in C_0$. We firstly assume that $x \in C_0$. From (i), we know that $x \notin C_0(a^*)$, so $x \notin e \cap B_e$. Since $G[C_0]$ is connected, there is a path $P$ in $G[C_0]$ connecting $x$ to the other parts of $G$ outside of $e$. Suppose $P = (v_0, v_1, \ldots, v_t)$, where $v_0 = x$, $v_t \notin e$, and $\{v_1, \ldots, v_{t-1}\} \subseteq e$. Let $i$ be the index such that $v_i$ is the first vertex on $P$ with $v_i \in B_e$. Then

(iii) $v_i \in C_0(a^*)$;

(iv) $v_i \in I_e$ and thus $v_i$ and $x$ belong to a same component of $G_e$, which is $H$;

(v) both $v_{i-1}$ and $v_i$ are in $I_e$, and hence the edge $(v_{i-1}, v_i)$ is in $H$ (note that $i \geq 1$ since $v_0 = x \notin B_e$).

By (v) and Phase II of the algorithm, either $v_i \in C_H$ or $v_{i-1} \in C_H$. But this contradicts (i), (ii) and (iii).

The case that $x \notin C_0$ but is adjacent with a vertex $y \in C_0$ can be proved similarly.

Therefore, we have proved that $\tilde{G} = G[C]$. □

Based on the conclusions from Lemmas 1 and 2, we obtain the following theorem showing the correctness of our algorithm.

**Theorem 1.** The output $C$ of our algorithm is a connected vertex cover for $G$.

4.2. Time complexity

In this subsection we consider the time complexity of our algorithm. Phase I of the algorithm uses a polynomial time $\rho$-approximation to compute $C_0$. Phase II uses exhaust search which is the most time consuming part. We shall prove that this part can also be executed in polynomial time, by implementing the relation between vertex cover and independent set.

**Lemma 3.** The number of independent unit disks in an $m \times m$ square is at most \( \left\lceil \frac{(m+2)^2}{\pi} \right\rceil \).

**Proof.** Enlarge the $m \times m$ square to an $(m + 2) \times (m + 2)$ square by adding a boundary with width one. Then all the disks whose centers are in the $m \times m$ square lie completely in the $(m + 2) \times (m + 2)$ square. Since each unit disk occupies area $\pi$, the result follows from the independence assumption. □

With the help of Lemma 3, we have the following theorem.

**Theorem 2.** The running time of our algorithm is $n^{O(1/\varepsilon^2)}$, where $n$ is the number of vertices in the graph.
Proof. It is well known that a vertex set $S$ is a vertex cover of a graph if and only if its complement is an independent set. Thus by Lemma 3, each $V(H) \setminus C_H$ contains at most $\lceil \frac{(m+2)^2}{\pi} \rceil$ independent vertices, and therefore the exhaust search for $C_H$ (which can be done by considering the complement of each independent set in $H$) takes time at most $\sum_{k=0}^{\lceil \frac{(m+2)^2}{\pi} \rceil} \left( \begin{array}{c} n_H \vspace{1mm} \\ k \end{array} \right) = n_H^{O(m^2)}$, where $n_H$ is the number of vertices in $H$, and the total running time for phase II is at most $\sum_{e,H} n_H^{O(m^2)} = \left( \sum_{e,H} n_H \right)^{O(m^2)} = n^{O(1/\epsilon^2)}$. \hfill $\Box$

4.3. Performance

Here we prove that our algorithm is a $(1 + \epsilon)$-approximation.

**Definition 1.** For two subgraphs $G_1$, $G_2$ of $G$, the distance between $G_1$ and $G_2$ is the length of a shortest path of $G$ connecting $G_1$ and $G_2$ (where 'length' means the number of edges on the path), denoted by $\text{dist}(G_1, G_2)$.

In other words, if $\text{dist}(G_1, G_2) = k$, then $G_1$ and $G_2$ can be connected through $k − 1$ vertices. If a vertex cover of a connected graph is not a connected vertex cover, the distance between connected components of the subgraph induced by the vertex cover is not far, as can be seen from the following lemma.

**Lemma 4.** Suppose $H$ is a connected graph, and $C$ is a vertex cover of $H$. If $H[C]$ is not connected, then there exist two components $R_1$, $R_2$ of $H[C]$ such that $\text{dist}(R_1, R_2) = 2$.

**Proof.** Let $H$, $R_1$, $R_2$ be two ‘closest’ connected components of $H$, and $P = (v_0, v_1, \ldots, v_k)$ be a shortest path of $H$ connecting $R_1$ and $R_2$, where $v_0 \in V(R_1)$ and $v_k \in V(R_2)$. If $k \geq 3$, consider the edge $(v_1, v_2)$. Since $C$ covers $H$, we have either $v_1 \in C$ or $v_2 \in C$. Suppose, without loss of generality, that $v_1 \in C$. Let $R_3$ be the component of $G[C]$ containing $v_1$. Then $R_3 \neq R_1$ and $R_2$, and $\text{dist}(R_3, R_2) < \text{dist}(R_1, R_2)$, contradicting our choice of $R_1$ and $R_2$. \hfill $\Box$

The following lemma is well known for unit disk graph.

**Lemma 5.** Let $G$ be a unit disk graph and $u$ be a vertex in $V(G)$. The there are at most 5 independent vertices in $N(u)$, where $N(u)$ is the set of vertices adjacent with $u$ in $G$.

The following theorem shows that our algorithm is a PTAS.

**Theorem 3.** Let $C^*$ be an optimal CVC for $G$, and $C$ be the output of our algorithm. Then $|C| \leq (1 + \epsilon)|C^*|$. 

**Proof.** Firstly, we prove that

$$|C_0(a^*)| \leq \frac{\epsilon}{6}|C^*|. \hfill (1)$$

When the partition shifts, a vertex of $C_0$ belongs to at most 8 boundary areas of $B(P(a))$'s. Therefore, we have,

$$|C_0(0)| + |C_0(1)| + \cdots + |C_0(m - 1)| \leq 8|C_0|,$n$$

and thus

$$|C_0(a^*)| \leq \frac{8\rho|C^*|}{m} \leq \frac{\epsilon}{6}|C^*|.$$

Next, we shall add some vertices to $C^*$ such that the resulting set $\tilde{C}$ satisfies: for each square $e$ and for each component $H \in \text{Comp}(C_0)$, $\tilde{C} \cap V(H)$ is a CVC of $H$. \hfill (2)

For a square $e$, let $\tilde{C}_e = C^* \cap I_e$. It is easy to see that for each component $H \in \text{Comp}(C_e)$, $\tilde{C}_e \cap V(H)$ covers every edge of $H$. Suppose there exist a component $H \in \text{Comp}(G_e)$ such that requirement (2) is not satisfied. By Lemma 4, there are two components $R_1$, $R_2$ of $G(\tilde{C}_e \cap V(H))$ such that $R_1$ and $R_2$ can be connected through one vertex in $V(H) \setminus \tilde{C}_e$. Add this vertex to $\tilde{C}_e$. If the new $\tilde{C}_e$ still does not satisfy requirement (2), continue as above to add vertices to merge components. Suppose this is done $k$ times before $\tilde{C}_e$ satisfies (2), then

$$|\tilde{C}_e| \leq |C^* \cap I_e| + k. \hfill (3)$$

On the other hand, we can show that

$$|C_0(a^*) \cap I_e| \geq \frac{k}{5}. \hfill (4)$$

For this purpose, we suppose that the components merged are in the order, $R_1$ with $R_2$, $R_3$ with $R_4$, \ldots, $R_{2k−1}$ with $R_{2k}$. For simplicity of presenting the idea, we firstly assume that all the above $R_i$’s are distinct components of $G[C^* \cap I_e]$. For each $i = 1, 2, \ldots, k$, let $x_i$ be a vertex in $V(R_{2i−1}) \cap B_e \cap I_e$, such that $x_i$ is adjacent with a vertex $y_i \in B_e \setminus I_e$. Such an $x_i$ exists since $R_{2i−1}$ is connected to the outer parts of $e$ through $C^*$. Then either $x_i \in C_0$ or $y_i \in C_0$. Set $z_i = x_i$ if $x_i \in C_0$ and $z_i = y_i$ otherwise. Note that both $x_i, y_i \in B_e$. Hence $z_i \in C_0(a^*) \cap e$. A vertex may serve more than once as $z_i$'s. For example, it is
possible that there are two indices $i \neq j$ such that the vertex of $C_0$ covering edges $(x_i, y_i)$ and $(x_j, y_j)$ is the same $y_i = y_j \in C_0$. In this case, we see that $x_i$ and $x_j$ are independent since they belong to different components of $G[C^* \cap I_e]$. Then by Lemma 5, such a vertex serves at most 5 times as $z_i$'s, and inequality (4) follows. Next, suppose the $R_e$'s are not all distinct. For example, suppose $R_1$ is the component obtained by merging $R_1$ and $R_2$. Then $x_1$ can be chosen such that $x_1 \in V(H_2) \cap B_e \cap I_e$, which is independent with $x_1$. In general we can find $k$ independent vertices $x_1, x_3, \ldots, x_{2k-1}$ and thus (4) also holds in this case.

Combining inequalities (3) and (4), we have

$$|\tilde{C}_e| \leq |C^* \cap e| + 5|C_0(a^*) \cap e|. \quad (5)$$

Since in Phase II of the algorithm, $C_e$ is a 'minimum' vertex set satisfying requirement (2) for each square $e$, we have $|C_e| \leq |C_0|$. Combining this with (1) and (5), we have

$$\left\lfloor \sum_{e \in P(a^*)} |C_e| \right\rfloor \leq \sum_{e \in P(a^*)} \left( |C^* \cap e| + 5|C_0(a^*) \cap e| \right) \leq |C^*| + 5|C_0(a^*)| \leq \left( 1 + \frac{5\varepsilon}{6} \right) |C^*|. $$

Hence

$$|C| \leq |C_0(a^*)| + \sum_{e \in P(a^*)} |C_e| \leq (1 + \varepsilon)|C^*|. \quad \square$$

5. Conclusion

In this paper, we presented the first polynomial time approximation scheme to compute a connected vertex cover of a graph. The method used in this paper can be applied to CVC problems in $n$-dimensional ball graphs for any integer $n \geq 1$. In a unit ball graph, each vertex corresponds to the center of a unit ball in the $n$-dimensional space, and two vertices are adjacent if and only if the Euclidean distance between them is at most 1.

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