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Approximation and Parametrized Algorithms for Segment Set Cover

Master's thesis in COMPUTER SCIENCE

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10 Supervisor's statement

- Hereby I confirm that the presented thesis was prepared under my supervision and that it fulfils the requirements for the degree of Master of Computer Science.
- Date Supervisor's signature

$_{14}$ Author's statement

Hereby I declare that the presented thesis was prepared by me and none of its contents was obtained by means that are against the law.

The thesis has never before been a subject of any procedure of obtaining an academic degree.

Moreover, I declare that the present version of the thesis is identical to the attached electronic version.

21 Date Author's signature

22	${f Abstract}$
23 24	The work presents a study of different geometric set cover problems. It mostly focuses on segment set cover and its connection to the polygon set cover.
25	${f Keywords}$
26 27	set cover, geometric set cover, FPT, W[1]-completeness, APX-completeness, PCP theorem, NP-completeness
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odcinkami na płaszczyźnie

38 Contents

39	1.	Introduction	5				
40	2.	initions					
41		2.1. Geometric Set Cover	7				
42		2.2. Approximation	7				
43		2.3. δ -extensions	7				
44	3.	Geometric Set Cover with segments	9				
45		3.1. FPT for segments	9				
46			9				
47			9				
48		Ů,	11				
49			11				
50			 [3				
51			13				
52		0 0	۱4				
53		0 0	L 6				
54			L8				
55		v	18				
		v	LO [9				
56		1	21				
57			23				
58							
59		3.5. What is missing	25				
60	4.		27				
61		±	27				
62		v	27				
63		ı v	27				
64		1 1	28				
65		4.5. Connection with general set cover	28				
66	5 .	Geometric Set Cover with polygons	29				
67		5.1. State of the art	29				
	G	Conclusions	1				

69 Chapter 1

Introduction

The Set Cover problem is one of the most common NP-complete problems. [tutaj referencja]
We are given a family of sets and have to choose the smallest subfamily of these sets that cover
all their elements. This problem naturally extends to settings were we put different weights
on the sets and look for the subfamily of the minimal weight. This problem is NP-complete
even without weights and if we put restrictions on what the sets can be. One of such variants
is Vertex Cover problem, where sets have size 2 (they are edges in a graph).

In this work we focus on another such variant where the sets correspond to some geometric shapes and only some points of the plane have to be covered. When these shapes are rectangles with edges parallel to the axis, the problem can be proven to be W[1]-complete (solution of size k cannot be found in $n^o(k)$ time), APX-complete (for suffciently small $\epsilon > 0$, the problem does not admit $1 + \epsilon$ -approximation scheme) [refrencje].

Some of these settings are very easy. Set cover with lines parallel to one of the axis can be solved in polynomial time.

There is a notion of δ -expansions, which loosen the restrictions on geometric set cover. We allow the objects to cover the points after δ -expansion and compare the result to the original setting. This way we can produce both FPT and EPTAS for the rectangle set cover with δ -extensions [referencie].

Our contribution. In this work, we prove that unweighted geometric set cover with segments is fixed parameter tractable (FPT).

Moreover, we show that geometric set cover with segments is APX-complete for unweighted axis-parallel segments, even with 1/2-extensions. So the problem for very thin rectangles also can't admit PTAS. Therefore, in the efficient polynomial-time approximation scheme (EPTAS) for *fat polygons* by [Har-Peled and Lee, 2009], the assumption about polygons being fat is necessary.

Finally, we show that geometric set cover with weighted segments in 3 directions is W[1]-complete. However, geometric set cover with weighted segments is FPT if we allow δ -extension.

This result is especially interesting, since it's counter-intuitive that the unweighed setting is FPT and the weighted setting is W[1]-complete. Most of such problems (like vertex cover or [wiecej przykladow]) are equally hard in both weighted and unweighted settings.

Chapter 2

Definitions

103 2.1. Geometric Set Cover

In the geometric set cover problem we are are given \mathcal{P} – a set of objects, which are connected subsets of the plane, \mathcal{C} – a set of points in the plane. The task is to choose $\mathcal{R} \subseteq \mathcal{P}$ such that every point in \mathcal{C} is inside some element from \mathcal{R} and $|\mathcal{R}|$ is minimized.

In the parametrized setting for a given k, we only look for a solution \mathcal{R} such that $|\mathcal{R}| \leq k$.

In the weighted setting, there is some given weight function $f: \mathcal{P} \to \mathbb{R}^+$, and we would like to find a solution \mathcal{R} that minimizes $\sum_{R \in \mathcal{R}} f(R)$.

110 2.2. Approximation

Let us recall some definitions related to optimization problems that are used in the following sections.

Definition 1. A polynomial-time approximation scheme (PTAS) for a minimization problem Π is a family of algorithms \mathcal{A}_{ϵ} for every $\epsilon > 0$ such that \mathcal{A}_{ϵ} takes an instance I of Π and in polynomial time finds a solution that is within a factor $(1+\epsilon)$ of being optimal. That means the reported solution has weight at most $(1+\epsilon)opt(I)$, where opt(I) is the weight of an optimal solution for I.

Definition 2. A problem Π is **APX-hard** if assuming $P \neq NP$, there exists $\epsilon > 0$ such that there is no polynomial-time $(1 + \epsilon)$ -approximation algorithm for Π .

$_{ ext{0}}$ 2.3. δ -extensions

121 TODO PLACEHOLDER for introductory text

 δ -extensions is one of the modifications to a problem, that makes geometric set cover problem easier, it has been already used in literature (place some refrence here).

Definition 3 (δ -extensions for center-symmetric objects). For any $\delta > 0$ and a center-symmetric object L with centre of symmetry $S = (x_s, y_s)$, the δ -extension of L is the object $L^{+\delta} = \{(1+\delta) \cdot (x-x_s, y-y_s) + (x_s, y_s) : (x,y) \in L\}$, that is, $L^{+\delta}$ is the image of L under homothety centered at S with scale $(1+\delta)$

The geometric set cover problem with δ -extensions is a modified version of geometric set cover where:

- We need to cover all the points in C with objects from $\{P^{+\delta}: P \in P\}$ (which always include no fewer points than the objects before δ -extensions);
- We look for a solution that is no larger than the optimum solution for the original problem. Note that it does not need to be an optimal solution in the modified problem.
- Formally, we have the following.

Definition 4 (Geometric set cover problem with δ-extensions). The geometric set cover problem with δ-extensions is the problem where for an input instance $I = (\mathcal{P}, \mathcal{C})$, the task is to output a solution $\mathcal{R} \subseteq \mathcal{P}$ such that the δ-extended set $\{R^{+\delta} : R \in \mathcal{R}\}$ covers \mathcal{C} and is no larger than the optimal solution for the problem without extensions, i.e. $|\mathcal{R}| \leq |opt(I)|$.

TODO: Some text

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Definition 5 (Geometric set cover PTAS with δ-extensions). We define a PTAS for geometric set cover with δ-extensions as a family of algorithms $\{\mathcal{A}_{\delta,\epsilon}\}_{\delta,\epsilon>0}$ that each takes as an input instance $I = (\mathcal{P}, \mathcal{C})$, and in polynomial-time outputs a solution $\mathcal{R} \subseteq \mathcal{P}$ such that the δ-extended set $\{R^{+\delta}: R \in \mathcal{R}\}$ covers \mathcal{C} and is within a $(1+\epsilon)$ factor of the optimal solution for this problem without extensions, i.e. $(1+\epsilon)|\mathcal{R}| \leq |opt(I)|$.

⁴⁵ Chapter 3

Geometric Set Cover with segments

3.1. FPT for segments

In this section we consider the fixed-parameter tractable algorithms for unweighted geometric set cover with segments. Setting where segments are limited to be axis-parallel (or limited to constant number of directions) has an FPT algorithm already present in literature. We present an FPT algorithm for unweighted geometric set cover with segments, where segments are in arbitrary directions.

3.1.1. Axis-parallel segments

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You can find this in Platypus book. (TODO add referece)

We show an $\mathcal{O}(2^k)$ -time branching algorithm. The algorithm covers one of the points with a segment in one of the directions greedily until it covers all points or decide the is no solution of size k.

Let us take the point a which is the smallest among points that are not yet covered in the lexicographic ordering of points in \mathbb{R}^2 . We need to cover a with some of the remaining segments.

Branch over choice of one of the coordinates x or y. Among segments that share this coordinate with a, we greedily choose the segment that covers the most points. As a was the smallest in the lexicographical order, then all points that one of the coordinates with a, they have larger second coordinate. Therefore if we choose the segment that covers the most points s, it would be the segment that have the end with highest second coordinate. It also hold that every other segment covering a in this direction does not cover any point that is not covered by s. Therefore greedy choice of this segments is optimal.

If after k steps we do not cover all of the points, there is no solution of size at most k.

TODO: Maybe split it into theorem + algorithm + explanation like in section 3.1.2

TODO: Should it use Remark template?

The same algorithm can be used for segments in d directions, where we branch over d directions and it runs in complexity $\mathcal{O}(d^k)$.

3.1.2. Segments in arbitrary directions

174 In this section we consider setting where segments are not constrained to only d directions.

We present a fixed-parameter tractable algorithm, where parameter is the size of the solution.

Theorem 1. (FPT for segment cover). There exists an algorithm that given a family \mathcal{P} of n segments (in any direction), a set of m points \mathcal{C} and a parameter k, runs in time

178 $f(k) \cdot (nm)^c$ for some computable function f and constant c, and outputs a subfamily $\mathcal{R} \subseteq \mathcal{P}$ 179 such that $|\mathcal{R}| \leq k$ and \mathcal{R} covers all points in \mathcal{C} , or determines that such a set \mathcal{R} does not 180 exist.

We will need the following lemmas.

Lemma 1. Given an instance $(\mathcal{P}, \mathcal{C})$ of the segment cover problem, without a loss of generality we can assume that no segment covers a superset of what another segment covers. That is, for any distinct $A, B \in \mathcal{P}$, we have $A \cap C \not\subseteq B \cap C$ and $A \cap C \not\supseteq B \cap C$.

185 Proof. Trivial.

Lemma 2. Given an instance $(\mathcal{P}, \mathcal{C})$ of the segment cover problem, if there exists a line L with at least k+1 points on it, then there exists a subset $\mathcal{A} \subseteq \mathcal{P}$, $|\mathcal{A}| \leq k$, such that every solution \mathcal{R} with $|\mathcal{R}| \leq k$ satisfies $|\mathcal{A} \cap \mathcal{R}| \geq 1$. Moreover, such a subset can be found in polynomial time.

190 Proof. First we use Lemma 1.

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Let us enumerate the points from C that lie on L as $x_1, x_2, \ldots x_t$ in the order in which they appear on L. Every segment that is not collinear with L can cover at most one of these points. Therefore, in any solution of size not larger than k, among any k of these points at least one must be covered with segment collinear with L.

Therefore, every solution needs to take one of the segments collinear with L that covers any of the points $x_1, x_2, \ldots x_k$. After using reduction from Lemma 1, there are at most k such segments that are distinct.

We are ready to prove Theorem 1.

199 Proof of Theorem 1.

Algorithm. First we use Lemma 1.

Next, we present a recursive algorithm. Given an instance of the problem:

- (1) If there exist a line with at least k+1 points from \mathcal{C} , we branch over adding to the solution one of the at most k possible segments provided by Lemma 2; name this segment S. Then we find a solution \mathcal{R} for the problem for points $\mathcal{C} S$, segments $\mathcal{P} \{S\}$, and parameter k-1. We return $\mathcal{R} \cup \{S\}$.
- 206 (2) If every line has at most k points on it and $|\mathcal{C}| > k^2$, then answer NO.
- 207 (3) If $|\mathcal{C}| \leq k^2$, solve the problem by brute force: check all subsets of \mathcal{P} of size at most k.

Correctness. Lemma 2 proves that at least one segment that we branch over in (1) must be present in every solution \mathcal{R} with $|\mathcal{R}| \leq k$. Therefore, the recursive call can find a solution, provided there exists one.

In (2) the answer is no, because every line covers no more than k points from \mathcal{C} , which implies the same about every segment from \mathcal{P} . Under this assumption we can cover only k^2 points with a solution of size k, which is less than $|\mathcal{C}|$.

Checking all possible solutions in (3) is trivially correct.

- Complexity. In the leaves of recursion we have $|\mathcal{C}| \leq k^2$, so $|\mathcal{P}| \leq k^4$, because every segments can be uniquely identified by the two extreme points it covers (by Lemma 1). Therefore, there are $\binom{k^4}{k}$ possible solutions to check, each can be checked in time $O(k|\mathcal{C}|)$. Therefore, (3) takes time $k^{O(k)}$.
- In this branching algorithm our parameter k is decreased with every recursive call, so we have at most k levels of recursion with branching over k possibilites. Candidates to branch over can be found on each level in time $O(nm \log(nm))$.
- Reduction from Lemma 1 can be implemented in time $O(n^2m)$.
- It follows that the overall complexity is $O(n^2m + nm\log(nm) \cdot f(k))$

4 3.2. APX-completeness for segments parallel to axes

- In this section we analyze whether there exists PTAS for geometric set cover for rectangles.
- We show that we can restrict this problem to a very simple setting: segments parallel to axes
- 227 and allow (1/2)-extension, and the problem is still APX-hard. Note that segments are just
- degenerated rectangles with one side being very narrow.
- Our results can be summarized in the following theorem and this section aims to prove it.
- Theorem 2. (axis-parallel segment set cover with 1/2-extension is APX-hard).
- Unweighted geometric set cover with axis-parallel segments in 2D (even with 1/2-extension)
- is APX-hard. That is, assuming $P \neq NP$, there does not exist a PTAS for this problem.
- Theorem 2 implies the following.
- Corollary 1. (rectangle set cover is APX-hard). Unweighted geometric set cover with rectangles (even with 1/2-extension) is APX-hard.
- We prove Theorem 2 by taking a problem that is APX-hard and showing a reduction. For this problem we choose MAX-(3,3)-SAT which we define below.

3.2.1. MAX-(3,3)-SAT and statement of reduction

- Definition 6. MAX-3SAT is the following maximization problem. We are given a 3-CNF formula, and need to find an assignment of variables that satisfies the most clauses.
- Definition 7. MAX-(3,3)-SAT is a variant of MAX-3SAT with an additional restriction that every variable appears in exactly 3 clauses. Note that thus, the number of clauses is equal to the number of variables.
- In our proof of Theorem 2 we use hardness of approximation of MAX-(3,3)-SAT proved in [Håstad, 2001] and described in Theorem 3 below.
- **Definition 8** (α-satisfiable MAX-3SAT formula). MAX-3SAT formula of size n is at most α-satisfiable, if every assignment of variables satisfies no more than αn clauses.
- 248 Theorem 3. [Håstad, 2001]
- For any $\epsilon > 0$, it is NP-hard to distinguish satisfiable (3,3)-SAT formulas from at most (7/8 + ϵ)-satisfiable (3,3)-SAT formulas.

Given an instance I of MAX-(3,3)-SAT, we construct an instance J of axis-parallel segment set cover problem, such that for a sufficiently small $\epsilon > 0$, a polynomial time $(1 + \epsilon)$ approximation algorithm for J would be able to distinguish whether an instance I of MAX(3,3)-SAT is fully satisfiable or is at most $(7/8+\epsilon)$ -satisfiable. However, according to (Theorem
3) the latter problem is NP-hard. This would imply P = NP, contradicting the assumption.

The following lemma encapsulates the properties of the reduction described in this section, and it allows us to prove Theorem 2.

Lemma 3. Given an instance S of MAX-(3,3)-SAT with n variables and optimum value opt(S), we can construct an instance I of geometric set cover with axis-parallel segments in 2D, such that:

- (1) For every solution X of instance I, there exists a solution of S that satisfies at least 15n |X| clauses.
- 263 (2) For every solution of instance S that satisfies w clauses, there exists a solution of I of size 15n w.
- 265 (3) Every solution with 1/2-extensions of I is also a solution to the original instance I.

 266 Therefore, the optimum size of a solution of I is opt(I) = 15n opt(S).

We prove Lemma 3 in subsequent sections, but meanwhile let us prove Theorem 2 using Lemma 3 and Theorem 3.

TODO: This below can't use current template

270 Proof of Theorem 2. Consider any $0 < \epsilon < 1/(15 \cdot 8)$.

Let us assume that there exists a polynomial-time $(1 + \epsilon)$ -approximation algorithm for unweighted geometric set cover with axis-parallel segments in 2D with (1/2)-extensions. We construct an algorithm that solves the problem stated in Theorem 3, thereby proving that P = NP.

Take an instance S of MAX-(3,3)-SAT to be distinguished and construct an instance of geometric set cover I using Lemma 3. We now use the $(1 + \epsilon)$ -approximation algorithm for geometric set cover on I. Denote the size of the solution returned by this algorithm as approx(I). We prove that if in S one can satisfy at most $(\frac{7}{8} + \epsilon)n$ clauses, then $approx(I) \ge 15n - (\frac{7}{8} + \epsilon)n$ and if S is satisfiable, then $approx(I) < 15n - (\frac{7}{8} + \epsilon)n$.

Assume S satisfiable. From the definition of S being satisfiable, we have:

$$opt(S) = n.$$

From Lemma 3 we have:

$$opt(I) = 14n.$$

Therefore,

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$$approx(I) \leq (1+\epsilon)opt(I) = 14n(1+\epsilon) = 14n + 14\epsilon \cdot n =$$

$$= 14n + (15\epsilon - \epsilon)n < 14n + \left(\frac{1}{8} - \epsilon\right)n = 15n - \left(\frac{7}{8} + \epsilon\right)n$$

Assume S is at most $(\frac{7}{8} + \epsilon)$ satisfiable. From the defintion of S being at most $(\frac{7}{8} + \epsilon)$ n satisfiable, we have:

$$opt(S) \le \left(\frac{7}{8} + \epsilon\right)n$$

From Lemma 3 we have:

$$opt(I) \ge 15n - \left(\frac{7}{8} + \epsilon\right)n$$

Since a solution to I with $\frac{1}{2}$ -extensions is also a solution without extentions, by Lemma 3 (3.), we have:

$$approx(I) \ge opt(I) = 15n - \left(\frac{7}{8} + \epsilon\right)n$$

Therefore, by using the assumed $(1 + \epsilon)$ -approximation algorithm, it is possible to distinguish the case when S is satisfiable from the case when it is at most $(\frac{7}{8} + \epsilon)n$ satisfiable, it suffices to compute approx(I) with $15n - (\frac{7}{8} + \epsilon)n$. Hence, the assumed approximation algorithm cannot exist, unless P = NP.

3.2.2. Reduction

We proceed to the proof of Lemma 3. That is, we show a reduction from MAX-(3,3)-SAT problem to geometric set cover with segments parallel to axis. Moreover, the obtained instance of geometric set cover will be robust to 1/2-extensions (have the same optimal solution after 1/2-extension).

The construction will be composed of 2 types of gadgets: **VARIABLE-gadgets** and **CLAUSE-gadgets**. CLAUSE-gadgets would be constructed using two **OR-gadgets** connected together. Every gadget consists of a point set and a segment set.

295 3.2.2.1. VARIABLE-gadget

VARIABLE-gadget is responsible for choosing the value of a variable in a CNF formula. It allows two minimum solutions of size 3 each. These two choices correspond to the two Boolean values of the variable.

Points. Define points a, b, c, d, e, f, g, h as follows, where L = 12n:

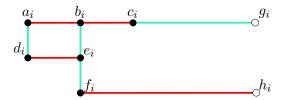


Figure 3.1: **VARIABLE-gadget.** We denote the set of points marked with black circles as C_{var}^i , and they need to be covered (are part of the set \mathcal{C}). Note that some of the points are not marked as black dots and exists only to name segments for further reference. We denote the set of red segments as X_{false}^i and the set of blue segments as X_{true}^i .

$$\begin{array}{ll} a = (-L,0) & b = (-\frac{2}{3}L,0) & c = (-\frac{1}{3}L,0) & d = (-L,1) \\ e = (-\frac{2}{3}L,1) & f = (-\frac{2}{3}L,2) & g = (L,0) & h = (L,2) \end{array}$$

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Let us define:

$$C_{var} = \{a, b, c, d, e, f\}$$

and

$$C_{var}^i = C_{var} + (0,4i)$$

We denote $a_i = a + (0, 4i)$ etc.

 $\mathbf{Segments.}$ Let us define:

$$X_{true}^{i} = \{(a_i, d_i), (b_i, f_i), (c_i, g_i)\}$$
$$X_{false}^{i} = \{(a_i, c_i), (d_i, e_i), (f_i, h_i)\}$$

$$P_{var}^i = X_{true}^i \cup X_{false}^i$$

Lemma 4. For any $1 \le i \le n$, points in C^i_{var} can be covered using 3 segments from P^i_{var} .

Proof. We can use either set X^i_{true} or X^i_{false} .

Lemma 5. For any $1 \le i \le n$, points in C^i_{var} can not be covered with fewer than 3 segments from P^i_{var} .

Proof. No segment of P_{var}^i covers more than one point from $\{d_i, f_i, c_i\}$, therefore C_{var}^i can not be covered with fewer than 3 segments.

Lemma 6. For every set $A \subseteq P_{var}^i$ such that A covers C_{var}^i and $(c_i, g_i), (f_i, h_i) \in A$, it holds that $|A| \ge 4$.

Proof. No segment from P_{var}^i covers more than one point from $\{a_i, e_i\}$, therefore C_{var}^i - $\{c_i, f_i, g_i, h_i\}$ can not be covered with fewer than 2 segments.

313 3.2.2.2. OR-gadget

OR-gadget has 3 important segments -x, y, result. x and y don't count to the weight of solution of OR-gadget (they are part of different gadgets). It has a minimal solution of weight w and result can be chosen only if x or y are also chosen for the solution. If none of them are chosen, then solution choosing result segment has weight at least w+1. Therefore the following formula holds for a solution R assuming that R uses only w from this OR-gadget:

$$(x \in R) \lor (y \in R) \iff result \in R$$

Points.

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$$\begin{array}{lll} l_0 = (0,0) & m_0 = (0,1) & n_0 = (0,2) & o_0 = (0,3) \\ p_0 = (0,4) & q_0 = (1,1) & r_0 = (1,3) & s_0 = (2,1) \\ t_0 = (2,2) & u_0 = (2,3) & v_0 = (3,2) \end{array}$$

$$vec_{i,j} = (10i + 3 + 3j, 4n + 2j)$$

Define $\{l_{i,j}, m_{i,j} \dots v_{i,j}\}$ as $\{l_0, m_0 \dots v_0\}$ shifted by $vec_{i,j}$ Note that $v_{i,0} = l_{i,1}$ (see Figure 3.3)

$$C$$
 or $gadget_{i,j} = \{l_{i,j}, m_{i,j}, n_{i,j}, o_{i,j}, p_{i,j}, q_{i,j}, r_{i,j}, s_{i,j}, t_{i,j}, u_{i,j}\}$

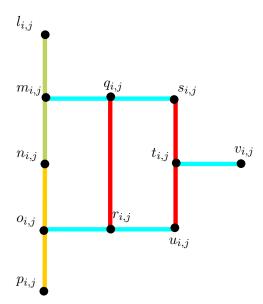


Figure 3.2: **OR-gadget.** We denote these point as $or_gadget_{i,j}$. We denote set of red segments as $or_{i,j}^{false}$, set of blue segments as $or_{i,j}^{true}$, green and yellow segments as $or_move_variable_{i,j}$.

323 Segments. We define names subsets of segments, to refer to them in lemmas.

$$or_{i,j}^{false} = \{(q_{i,j}, r_{i,j}), (s_{i,j}, u_{i,j})\}$$

$$or_{i,j}^{true} = \{(m_{i,j}, s_{i,j}), (o_{i,j}, u_{i,j}), (t_{i,j}, v_{i,j})\}$$

or move
$$variable_{i,j} = \{(l_{i,j}, n_{i,j}), (n_{i,j}, p_{i,j})\}$$

Segments in OR-gadget:

$$P_or_gadget_{i,j} = or_{i,j}^{false} \cup or_{i,j}^{true} \cup or_move_variable_{i,j}$$

Lemma 7. For any $1 \leq i \leq n, j \in \{0,1\}$ and $x \in \{l_{i,j}, p_{i,j}\}$ we can cover points in $C_{or_gadget_{i,j}} - \{x\} \cup \{v_{i,j}\}$ with 4 segments from $P_{or_gadget_{i,j}}$.

Proof. We can do that using one segment from $or_move_variable_{i,j}$ (chosen depending on the value of x) and all segments from $or_{i,j}^{true}$.

Lemma 8. For any $1 \le i \le n, j \in \{0,1\}$, we can cover points in $C_or_gadget_{i,j}$ with 4 segments from $P_or_gadget_{i,j}$.

Proof. We can do that using $or_move_variable_{i,j}$ and $or_{i,j}^{false}$.

332 3.2.2.3. CLAUSE-gadget

CLAUSE-gadget is responsible for calculating if choice of the variable values meets the clause in formula. It has minimal solution of weight w if at least one variable in the clause has a correct value. Otherwise it has minimal solution w+1. This way by the minimal solution for the whole problem, we can tell how many clauses were satisfiable.

The CLAUSE-gadgets consist of two OR-gadgets. We don't want the CLAUSE-gadgets to be crammed somewhere between the very long variable segments. That's why we have a simple gadget to pass the value of the segment, ie. segments $(x_{i,0}, x_{i,1}), (y_{i,0}, y_{i,1}), (z_{i,0}, z_{i,1})$. Two segments and one of them is chosen if x was chosen in the solution and the other one if x wasn't.

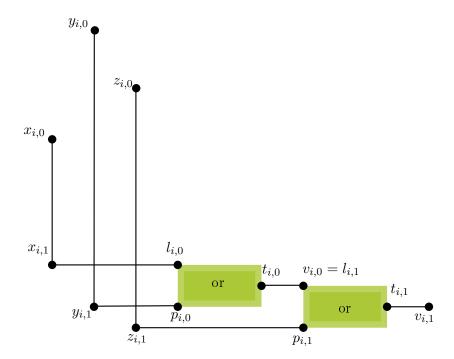


Figure 3.3: **CLAUSE-gadget.** We denote set of these points as C_clause_i . Every green rectangle is an OR-gadget. y-coordinates of $x_{i,0}$, $y_{i,0}$ and $z_{i,0}$ depend on the values of variables in the i-th clause.

Points. TODO: Rephrase it

Assuming clause $C_i = x_i \vee y_i \vee z_i$, function idx(w) is returning index of the variable w, function neg(w) is returning whether variable w is negated in a clause.

$$x_{i,0} = (10i+1, 4 \cdot idx(x_i) + 2 \cdot neg(x_i)) \quad x_{i,1} = (10i+1, 4n)$$

$$y_{i,0} = (10i+2, 4 \cdot idx(y_i) + 2 \cdot neg(y_i)) \quad y_{i,1} = (10i+2, 4n+4)$$

$$z_{i,0} = (10i+3, 4 \cdot idx(z_i) + 2 \cdot neg(z_i)) \quad z_{i,1} = (10i+3, 4n+6)$$

$$move_variable_i = \{x_{i,j} : j \in \{0,1\}\} \cup \{y_{i,j} : j \in \{0,1\}\} \cup \{z_{i,j} : j \in \{0,1\}\}$$

$$C_clause_i = move_variable_i \cup C_or_gadget_{i,0} \cup C_or_gadget_{i,1} \cup \{v_{i,1}\}$$

Segments.

$$P_clause_{i} = \{(x_{i,0}, x_{i,1}), (y_{i,0}, y_{i,1}), (z_{i,0}, z_{i,1}), (x_{i,1}, l_{i,0}), (y_{i,1}, p_{i,0}), (z_{i,1}, p_{i,1}), \} \cup \\ \cup P_or_gadget_{i,0} \cup P_or_gadget_{i,1}$$

- Lemma 9. For any $1 \le i \le n$ and $a \in \{x_{i,0}, y_{i,0}, z_{i,0}\}$, points in $C_clause_i \{a\}$ can be covered with a set of segments $P_true_i^a$, a subset of P_clause_i such that $|P_true_i^a| = 11$.
- Proof. For $a = x_{i,0}$ (analogous proof for $y_{i,0}$): First we use Lemma 7 twice with excluded
- $x = l_{i,0}$ and $x = l_{i,1} = v_{i,0}$, resulting with 8 segments condots condots cover all required
- points apart from $x_{i,1}, y_{i,0}, y_{i,1}, z_{i,0}, z_{i,1}, l_{i,0}$. We cover those using additional 3 segments:
- 351 $\{(x_{i,1}, l_{i,0}), (y_{i,0}, y_{i,1}), (z_{i,0}, z_{i,1})\}$
- For $a=z_{0,i}$: Using Lemma 8 and Lemma 7 with $x=p_{i,1}$, resulting with 8 segments
- or $c_{i,0}^{false} \cup or_{i,1}^{true}$ which cover all required points apart from $x_{i,0}, x_{i,1}, y_{i,0}, y_{i,1}, z_{i,1}, p_{i,1}$. We cover
- those using additional 3 segments: $\{(x_{i,0}, x_{i,1}), (y_{i,0}, y_{i,1}), (z_{i,1}, p_{i,1})\}$.
- Lemma 10. For any $1 \le i \le n$, points in C_clause_i can be covered with a set of segments $P_false^a_i$, a subset of P_clause_i such that $|P_false^a_i| = 12$.
- Proof. Using Lemma 8 twice we can cover $or_gadget_{i,0}$ and $or_gadget_{i,1}$ with 8 segments.
- To cover the remaining points we additionally use: $\{(x_{i,0}, x_{i,1}), (y_{i,0}, y_{i,1}), (z_{i,0}, z_{i,1}), (t_{i,1}, v_{i,1})\}$
- 360 Lemma 11. For any $1 \le i \le n$:
- 361 (1) points in $C_clause_i \{x_{i,0}, y_{i,0}, z_{i,0}\}$ can not be covered using any subset of segments 362 from P_clause_i of size less than 11;
- 263 (2) points in C_clause_i can not be covered using any subset of segments from P_clause_i 264 of size less than 12.
- 265 Proof of no cover with fewer than 12 segments. There is independent set of 12 points in $C_clause_i \supseteq \{x_{i,0}, y_{i,0}, z_{i,0}, l_{i,0}, p_{i,0}, q_{i,0}, v_{i,0} = l_{i,1}, p_{i,1}, q_{i,1}, u_{i,1}, v_{i,1}\}.$
- Proof of no cover with fewer than 11 segments. We can choose disjoint sets X, Y, Z such that
- 368 $X \cup Y \cup Z \subseteq C$ clause_i $-\{x_{i,0}, y_{i,0}, z_{i,0}\}$ and there are no segments covering points from
- different sets. And we prove lower bounds for each of these sets.

$$X = \{x_{i,1}, y_{i,1}, z_{i,1}\}$$

Set X is an indendent set, so it must be covered with 3 segments.

$$Y = or \ gadget_{i,0} - \{l_{i,0}, p_{i,0}\}$$

$$Z = or_gadget_{i,1} - \{l_{i,1}, p_{i,1}\}$$

- For both Y and Z we can check all of the subsets of 3 segments with brutforce that none of them cover, so they have to be covered with 4 segments.
- TODO: Funny fact, neither Y nor Z doesn't have independent set of size 4.
- Therefore C_clause_i must be covered with at least 3+4+4=11 segments.

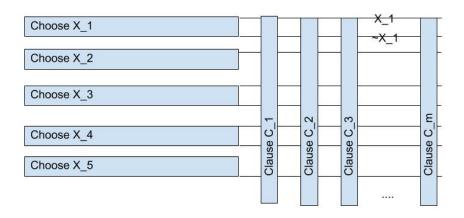


Figure 3.4: General schema.

General layout of VARIABLE-gadget and CLAUSE-gadget and how they interact with each other.

TODO: Rename Choose X to VARIABLE-gadget and Clause C to CLAUSE-gadget.

375 3.2.2.4. Summary

376 Add some smart lemmas that sets will be exclusive to each other.

Lemma 12. Robustness to 1/2-extensions. For every segment $s \in \mathcal{P}$, s and $s^{+1/2}$ cover the same points from \mathcal{C} .

Proof. We can just check every segment. Most of the segments s are collinear only with points that lay on s, so trivially $s^{+1/2}$ cannot cover more points than s does.

TODO: list problematic segments here

3.2.2.5. Summary of construction

We define:

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$$\mathcal{C} := \bigcup_{1 \leq i \leq n} C_variable_i \cup C_clause_i$$

$$\mathcal{P} := \bigcup_{1 \leq i \leq n} P_variable_i \cup P_clause_i$$

The subsequent sections define these sets.

We prove some properties of different gadgets. Every segment for a gadget will only cover points in this gadget (won't interact with any different gadget), so we can prove lemmas *locally*. TODO: y axis is increasing values downward on figures (not upwards like in normal).

3.2.3. Construction lemmas and proof of Lemma 3

Lemma 13. Given an instance S of MAX-(3,3)-SAT of size n with optimum solution satisfying k clauses opt(S) = k. Instance of geometric set cover, constructed for S as described in Section 3.2.2, can be solved with a solution of size 15n - k.

Proof. Let us name the assignments of the variables in the optimum solution of an instance S as $y_1, y_2 \ldots y_n$ and clauses as $c_1, c_2 \ldots c_n$.

We cover every VARIABLE-gadget with solution described in Lemma 4, in the *i*-th gadget choosing the set of segments corresponding to the value of y_i . CLAUSE-gadgets that are satisfied, let us name the variable that is true in them a, are covered with set $P_true_i^a$ described in Lemma 9 and unsatisfied with set P_false_i described in Lemma 10.

$$R_{i} = \begin{cases} X_{i}^{true} & \text{if } y_{i} \\ X_{i}^{false} & \text{if } \neg y_{i} \end{cases}$$

$$C_{i} = \begin{cases} P_{-}true_{i}^{a} & \text{if } c_{i} \text{ satisfied} \\ P_{-}false_{i} & \text{if } c_{i} \text{ not satisfied} \end{cases}$$

$$\mathcal{R} = \bigcup_{i=1}^{n} \{R_{i} \cup C_{i} : 1 \leq i \leq n\}$$

This set covers all points form C, because the smaller sets individually cover their corresponding gadgets (proved in respective lemmas).

All of these sets are disjoint, so the size of the solution is:

$$|\mathcal{R}| = \sum_{i=1}^{n} R_i + \sum_{i=1}^{n} C_i = 3n + 11k + 12(n-k) = 15n - k.$$

Lemma 14. Given an instance S of MAX-(3,3)-SAT of size n and a solution of an instance of geometric set cover, as described in Section 3.2.2, that is of size w. There exists a solution of an instance S that satisfied at least 15n - w clauses.

404 Proof

 Given a solution \mathcal{R} of the instance of geometric set cover, we construct a solution of the instance S by constructing an assignment of variables that satisfies at least 15n - w clauses in S.

Variables We need to use at least 3 segments to cover VARIABLE-gadget (Lemma 5).

If we have chosen both segments (c_i, g_i) and (f_i, h_i) , then we have used at least 4 segments

(Lemma 6).

$$\begin{cases} |C_{var}^{i} \cap \mathcal{R}| \geq 4 & \text{if } (c_{i}, g_{i}) \in \mathcal{R} \land (f_{i}, h_{i}) \in \mathcal{R} \\ |C_{var}^{i} \cap \mathcal{R}| \geq 3 & \text{otherwise} \end{cases}$$

If we chose at most one of the segments (c_i, g_i) and (f_i, h_i) , choose the corresponding variable value to the solution. If we chose both segments, choose the value that appears in most clauses. Every variable is in exactly 3 clauses, so one value appears in at least 2 of them.

If we have chosen none of the segments, set value to false. Formally, we define the value of the x_i variable as follows:

$$\begin{cases} x_{i} = majority(X_{i}) & \text{if } (c_{i}, g_{i}) \in \mathcal{R} \land (f_{i}, h_{i}) \in \mathcal{R} \\ x_{i} = true & \text{if } (c_{i}, g_{i}) \in \mathcal{R} \\ x_{i} = false & \text{if } (f_{i}, h_{i}) \in \mathcal{R} \\ x_{i} = false & \text{otherwise} \end{cases}$$

$$(3.1)$$

TODO: Maybe remove section below, because we do this calculation at the end anyway To cover $\bigcup_{1 \leq i \leq n} C_{var}^i$ we have used at least 3n + a segments, where a is the number of i such that we have chosen both values (c_i, g_i) and (f_i, h_i) .

Clauses For a clause $C_i = x \vee y \vee z$, we need to use at least 11 segments to cover $C_{20} \quad C_{20} \quad C_{20}$

TODO: maybe put something with cases and names of sets as above

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Moreover, if all of the points $\{x_{i,0}, y_{i,0}, z_{i,0}\}$ are not covered by the segments from P_{var}^i , then we need to cover C_clause_i with at least 12 segments by Lemma 11.

then we need to cover C_clause_i with at least 12 segments by Lemma 11.

TODO: Maybe remove section below, because we do this calculation at the end anyway

We covered CLAUSE-gadget with at least 11 or at least 12 segments: $|\bigcup_{i=1}^{n} P_clause_i \cap \mathcal{R}| \geq$ 11n + b, where b is the number of clauses where none of the segments covering the points $x_{i,0}, y_{i,0}, z_{i,0}$ were chosen in P_{var}^j .

Satisfied clauses with chosen variables assignment Clauses for which none of the points $x_{i,0}, y_{i,0}, z_{i,0}$ were covered by segments in P_{var}^j , are not satisfied in our variables assignment, but not all clauses that cover one of these points with segment in P_{var}^j are satisfied.

Let us look at such clause C_i and of points $x_{i,0}, y_{i,0}, z_{i,0}$ that are covered in P_{var}^j . Consider the cases of choosing variable value in equation (3.1).

If only one of the segments (c_i, g_i) and (f_i, h_i) are chosen in P_{var}^j , then the value of x_j is the same as the one satisfying C_i and clause is satisfied.

If we chose neither (c_i, g_i) or (f_i, h_i) , then it is impossible that this point is covered in P_{var}^j .

If we chose both (c_i, g_i) and (f_i, h_i) , then there are 3 clauses for which this point is covered by P_{var}^j . We chose variable value in a way that only one clause using x_j is not satisfied by the value of x_j . Therefore there are at most a clauses that are covered with 11 segments from CLAUSE-gadget, but are not satisfied.

So in the solution to this MAX-(3,3)-SAT instance that we have shown, there are at most a + b unsatsfied clauses.

Conclusions We proved that given a solution of size w we have the variables assignment that satisfies at least n - (a + b) clauses of S. At last we prove that $n - (a + b) \ge 15n - w$.

$$w \ge 3(n-a) + 4a + 11(n-b) + 12b = 3n + a + 11n + b = 14n + a + b$$
$$15n - w < 15n - 14n - a - b = n - (a+b)$$

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satisfying k clauses. Let us construct an instance of geometric set cover for S as described in

Section 3.2.2 and name it I.

Given the Lemma 13, we know that there exists a solution of I of size 15n - k, so:

$$opt(I) < 15n - k$$
.

Proof of Lemma 3. Given an instance S of MAX-(3,3)-SAT of size n with optimum solution

Since the optimum solution of S satisfies k clauses, then according to Lemma 14:

$$opt(I) \ge 15n - k$$
.

Therefore solution from Lemma 13 of size 15n-k is an optimum solution for instance I.

3.3. FPT for weighted segments with δ -extensions

Theorem 4 (FPT for weighted segment cover with δ -extensions). There exists an algorithm \mathcal{A} that given a family \mathcal{P} of n weighted segments (in any direction), a set of m points \mathcal{C} , and parameters k and δ , runs in time $f(k,\delta)\cdot(nm)^c$ for some computable function f and a constant c, and outputs a set $\mathcal{R}\subseteq\mathcal{P}$ such that $|\mathcal{R}|\leq k$ and $\mathcal{R}^{+\delta}$ covers all points in \mathcal{C} or determines that such a set \mathcal{R} does not exist.

To solve this problem we will introduce a lemma about choosing *good* subsets of points.

Definition 9. For a set of collinear points C, a subset $A \subseteq C$ is (k, δ) -good if for any set of segments R that covers set A such that $|R| \le k$, it holds that $R^{+\delta}$ covers C.

Lemma 15. There exists an algorithm that for any set of collinear points C, $\delta > 0$ and $k \geq 1$, outputs a (k, δ) -good set of size at most $f(k, \delta)$ for some computable function f. This algorithm runs in time $O(|C| \cdot f(k, \delta))$.

Proof. We prove this for a fixed δ by induction over k for any set of collinear points C.

Inductive hypothesis For any set of collinear points C, there exists an algorithm that runs in time $O(|C|k(1+\frac{1}{\delta}))$ and finds a set A such that:

- A is $(l, \delta) good$ for every $1 \le l \le k$,
- A has size $|A| < f(\delta, k)$ for some computable function f,
- extreme points from C are in A.

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Base case for k = 1 It is sufficient that A consists of 2 points: extreme points from C or a single point if |C| = 1.

If they are covered with one segment, it must be a segment that includes the extreme points from C, so it covers whole set C.

Inductive step Assuming inductive hypothesis for any set of collinear points C and for 473 k, we will prove hypothesis for k+1. 474

Let us name s the minimal segment that includes all points from C.

We define $M = [1 + \frac{2}{\delta}]$ subsegments of s in the following way. We split s into M parts v_i of equal length, that is $|v_i| = \frac{|s|}{M}$ for any $1 \le i \le M$.

 C_i is a subset of C such that they lay on v_i .

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 t_i is a segment connecting leftmost and rightmost point in C_i (it might be degenerated segment if $|C_i| = 1$ or it might be empty if C_i is empty).

TODO: Add a picture with v_i and t_i here

We use inductive hypothesis to choose (k, δ) -good sets A_i for sets C_i . If $|C_i| \leq 1$, then 482 $A_i = C_i$ and it's still a (k, δ) -good set. Then we define $A = \bigcup_{i=1}^M A_i$. It includes ends of s, because they are in sets A_1 and A_M . 483

Proof that A is (k, δ) -good for C Let us take any cover of A with k+1 segments and name it \mathcal{R} .

For every segment t_i , if there exists a segment x from \mathcal{R} such that it is disjoint with t_i , then we have a cover of A_i with at most k segments using $\mathcal{R} - \{x\}$. Since A_i is (k, δ) -good for t_i and C_i , then $(\mathcal{R} - \{x\})^{+\delta}$ covers C_i .

If there exists a segment t_i for which a segment x as defined above does not exist, then all k+1 segments that cover A_i intersect with t_i . (Note: There exists only one such segment t_i). From the inductive hypothesis ends of s are in A_1 and A_M respectively, so $\mathcal R$ must cover them. Hence there must exist segments starting in the ends of s and ending somewhere in t_i . Let us name these two segments y and z. It follows that: $|y|+|z|+|t_i|\geq |s|$. Since $|t_i|\leq |v_i|=\frac{|s|}{M}\leq \frac{|s|}{1+\frac{2}{\delta}}=\frac{|s|\delta}{\delta+2}$, therefore $\max(|y|,|z|)>|s|(1-\frac{\delta}{\delta+2})/2=\frac{|s|}{\delta+2}$.

TODO: Add a picture with such segments here

After δ -extension, the longer of these segments will lengthen both ways by at least:

$$\frac{|s|\delta}{\delta+2} = \frac{|s|}{1+\frac{2}{\delta}} > \frac{|s|}{M} = v_i > t_i.$$

Therefore the longer of segments y and z will cover the segment t_i after δ -extension, 497 therefore $\mathcal{R}^{+\delta}$ covers C_i . Since $C = \bigcup_{i=1}^M C_i$, then $\mathcal{R}^{+\delta}$ covers C_i . 498

Complexity We use the recursive algorithm for subsets C_i . Every point from C belongs to at most 2 sets C_i .

Apart from recursive algorithm we perform operations linear in size of |C|+M to calculate 502 the sets C_i . 503

Therefore it has complexity:

$$O(|C|+M) + \sum_{i=0}^{M} O(|C_{i}|k(1+\frac{1}{\delta})) = O(|C|+(1+\frac{1}{\delta})) + O((\sum_{i=0}^{M} |C_{i}|)k(1+\frac{1}{\delta})) \le O(|C|k(1+\frac{1}{\delta})).$$

Proof of Theorem 4. To construct an algorithm for this problem let us formulate some claims 505 about the problem first.

Definition 10. Line is **long** if there are at least k+1 points from \mathcal{C} on it.

Claim 1. If there are more than k long lines, then C can not be covered with k segments.

Claim 2. If there is more than k^2 points from C that do not lie on any long line, then C can not be covered with k segments.

Applying the above claims, if we have more than k long lines or more than k^2 points form \mathcal{C} that do not lie on any long line, then we answer that there is no solution of size at most k.

Otherwise, we can split C into at most k+1 sets: D, at most k^2 points that do not lie on any long line and C_i – points that lay on i-th long line. Sets C_i do not need to be disjoint.

Then for every set C_i , we can use Lemma 15 to get (k, δ) -good set A_i for C_i .

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Then we have set $D \cup \bigcup A_i$ of size at most $f(k, \delta)$ for some computable function f, that if we have a solution \mathcal{R} of size at most k that covers $D \cup \bigcup A_i$, then $\mathcal{R}^{+\delta}$ covers \mathcal{C} . This is because \mathcal{R} already covers points D, they cover C_i , because they cover (k, δ) -good set A_i with at most k segments, so $\mathcal{R}^{+\delta}$ covers C_i .

After that we shrunk down size of \mathcal{C} to size of $f(k,\delta)$ for some computable function f. Then we would like to shrink down size of \mathcal{P} . For every collinear subset of D, we can choose one segment from \mathcal{P} that covers these points and have the lowest weight or decide there is no segment that cover them. There are at most $|D|^2$ different segments, because we can distinguish these collinear sets by their extreme points.

This has complexity $O(|D|^2|\mathcal{P}|)$ and produce shrunk down set \mathcal{P} of size $f(k,\delta)$ for some computable function f.

Then we can iterate over all subsets of shrunk down set \mathcal{P} and choose the set with the lowest sum of weights that cover D. This solution would have weight not larger than optimal solution for the problem without extension, because we iterate over all posibilities of covering the subset of \mathcal{C} .

3.4. W[1]-completeness for weighted segments in 3 directions

Theorem 5. W[1]-completeness for weighted segments in 3 directions. Consider the problem of covering a set C of points by selecting k axis-pararell or right-diagonal weighted segments with weights from a set P with minimal weight. Assuming ETH, there is no algorithm for this problem with running time $f(k) \cdot (|C| + |P|)^{o(\sqrt{k})}$ for any computable function f.

We will show reduction from grid tiling problem.

Let's have an instance of grid tiling problem – size of the gird k, number of elements available n and k^2 sets of available pairs in every tile $S_{i,j} \subseteq \{1,n\} \times \{1,n\}$.

40 Construction. We construct a set \mathcal{P} of segments and a set \mathcal{C} of points.

First let's choose any ordering of n^2 elements $\{1, n\} \times \{1, n\}$ and name this sequence $a_1 \dots a_{n^2}$.

$$match_v(i, j) \iff a_i = \{x_i, y_i\} \land a_j = \{x_j, y_j\} \land x_i = x_j$$

$$match_h(i,j) \iff a_i = \{x_i, y_i\} \land a_j = \{x_j, y_j\} \land y_i = y_j$$

Points. Define points:

$$h_{i,j,t} = (j \cdot (n^2 + 1) + t, (n^2 + 1) \cdot i)$$
$$v_{i,j,t} = ((n^2 + 1) \cdot i, j \cdot (n^2 + 1) + t)$$

Let's define sets H and V as:

$$H = \{h_{i,j,t} : 1 \le i, j, \le k, 1 \le t \le n^2\}$$
$$V = \{v_{i,j,t} : 1 \le i, j, \le k, 1 \le t \le n^2\}$$

Let's define $\epsilon = 0.1$. For a point $\{x,y\} = p$ we define points $p^L = \{x-\epsilon,y\}, p^R = \{x+\epsilon,y\},$ $p^U = \{x,y-\epsilon\},$ and $p^D = \{x,y+\epsilon\}.$

Then we define:

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$$\mathcal{C} := H \cup \{ p^L : p \in H \} \cup \{ p^R : p \in H \} \cup V \cup \{ p^U : p \in V \} \cup \{ p^D : p \in V \}$$

Segments. Define horizontal segments.

$$\begin{aligned} hor_{i,j,t_{1},t_{2}} &= (h_{i,j,t_{1}}^{R}, h_{i,j+1,t_{2}}^{L}) \\ ver_{i,j,t_{1},t_{2}} &= (v_{i,j,t_{1}}^{D}, v_{i,j+1,t_{2}}^{U}) \\ horbeg_{i,t} &= (h_{i,1,1}^{L}, h_{i,1,t}^{L}) \\ horend_{i,t} &= (h_{i,n,t}^{R}, h_{i,n,n^{2}}^{R}) \\ \end{aligned}$$

$$verbeg_{i,t} &= (v_{i,1,1}^{U}, v_{i,1,t}^{U}) \\ verend_{i,t} &= (v_{i,n,t}^{D}, v_{i,n,n^{2}}^{D}) \end{aligned}$$

$$HOR = \{hor_{i,j,t_1,t_2} : 1 \le i \le k, 1 \le j < k, 1 \le t_1, t_2 \le n^2, match_h(t_1, t_2)\}$$

$$\cup \{horbeg_{i,t} : 1 \le i \le k, 1 \le t \le n^2\}$$

$$\cup \{horend_{i,t} : 1 \le i \le k, 1 \le t \le n^2\}$$

$$VER = \{ver_{i,j,t_1,t_2} : 1 \le i \le k, 1 \le j < k, 1 \le t_1, t_2 \le n^2, match_v(t_1, t_2)\}$$

$$\cup \{verbeg_{i,t} : 1 \le i \le k, 1 \le t \le n^2\}$$

$$\cup \{verend_{i,t} : 1 \le i \le k, 1 \le t \le n^2\}$$

$$DIAG := \{(h_{i,j,t}, v_{j,i,t}) : 1 \le i, j \le k, 1 \le t \le n^2, a_t \in S_{i,j}\}$$

TODO: explain that these segments are in fact diagonal

$$\mathcal{P} := HOR \cup VER \cup DIAG$$

Lemma 16. If there exists solution for grid tiling, then there exists solution for our construction using $2(k+1)k+k^2$ segments with weight exactly $2k \cdot (k(n^2+1)-2-2\epsilon(k-1))$. **Claim 3.** If there exists a solution to the grid tiling $c_1 ldots c_k$ and $r_1 ldots r_k$, then there exists a solution covering all points

$$\{h_{i,j,t}: 1 \le i, j \le k, t = (c_i, r_j)\} \cup \{v_{i,j,t}: 1 \le i, j \le k, t = (c_j, r_i)\}$$

- with segments in DIAG and the rest in VER or HOR and has weight $2k \cdot (k(n^2+1)-550 \quad 2-2\epsilon(k-1))$.
- 551 **Proof.** TODO: jakiś prosty z definicji
- Lemma 17. If there exists solution for our construction using $2(k+1)k + k^2$ segments with weight exactly $2k \cdot (k(n^2+1) 2 2\epsilon(k-1))$, then there exists a solution for grid tiling
- Proof. This follows from Lemma 18, because we just take which points are covered with DIAG.
- Claim 4. Points p^L, p^R, p^U, p^D cannot be covered with DIAG.
- Claim 5. Points in $H \cup \{p^L : p \in H\} \cup \{p^R : p \in H\}$ cannot be covered with VER. Points in $V \cup \{p^U : p \in V\} \cup \{p^D : p \in V\}$ cannot be covered with HOR.
- Claim 6. For given i, j if none of the points $h_{i,j,t}$ $(v_{i,j,t})$ for $1 \le t \le n^2$ are covered with DIAG, then some spaces between neighbouring points were covered twice.
- Claim 7. For given i, j two points h_{i,j,t_1}, h_{i,j,t_2} (v_{i,j,t_1}, v_{i,j,t_2}) for $1 \le t_1 < t_2 \le n^2$ are covered with DIAG, then one of them had to be also covered with a segment from HOR (VER).
- **Proof.** Point v_{i,j,t_2}^L had to be covered with VER from Claims 4 and 5. And every segment in VER covering v_{i,j,t_2}^L , covers also v_{i,j,t_1}^L .
- Lemma 18. If there exists solution for our construction with weight at most (exactly) $2k \cdot (k(n^2+1)-2-2\epsilon(k-1))$, then for every i,j there must be exactly one t such that $h_{i,j,t}$ ($v_{i,j,t}$) is covered with DIAG and moreover if h_{i,j,t_1} and $h_{i,j+1,t_2}$ are uncovered, then $math_h(t_1,t_2)$. Analogically for v.
- **Proof.** Only k^2 points can be covered only in DIAG, the rest has to be covered with $VER \cup HOR$. Therefore every result must be at least $ALL_LINES 2k^2\epsilon$, because only spaces of length ϵ can be uncovered in this axis.
- Of course if h_{i,j,t_1} and $h_{i,j+1,t_2}$ are uncovered, then there must exist a segment in HOR between h_{i,j,t_1}^R and $h_{i,j+1,t_2}^L$, so $math_h(t_1,t_2)$ must be true.

574 3.5. What is missing

We don't know FPT for axis-pararell segments without δ -extensions.

576 Chapter 4

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Geometric Set Cover with lines

$_{578}$ 4.1. Lines parallel to one of the axis

When \mathcal{R} consists only of lines parallel to one of the axis, the problem can be solved in polynomial time.

We create bipartial graph G with node for every line on the input split into sets: H – horizontal lines and V – vertical lines. If any two lines cover the same point from C, then we add edge between them.

Of course there will be no edges between nodes inside H, because all of them are pararell and if they share one point, they are the same lines. Similar argument for V. So the graph is bipartial.

Now Geometric Set Cover can be solved with Vertex Cover on graph G. Since Vertex Cover (even in weighted setting) on bipartial graphs can be solved in polynomial time.

Short note for myself just to remember how to this in polynomial time:

Non-weighted setting - Konig theorem + max matching

Weighted setting - Min cut in graph of $\neg A$ or $\neg B$ (edges directed from V to H)

592 4.2. FPT for arbitrary lines

You can find this is Platypus book. We will show FPT kernel of size at most k^2 .

(Maybe we need to reduce lines with one point/points with one line).

For every line if there is more than k points on it, you have to take it. At the end, if there is more than k^2 points, return NO. Otherwise there is no more than k^4 lines.

In weighted settings among the same lines with different weights you leave the cheapest one and use the same algorithm.

99 4.3. APX-completeness for arbitrary lines

We will show a reduction from Vertex Cover problem. Let's take an instance of the Vertex Cover problem for graph G. We will create a set of |V(G)| pairwise non-pararell lines, such that no three of them share a common point.

Then for every edge in $(v, w) \in E(G)$ we put a point on crossing of lines for vertices v and w. They are not pararell, so there exists exactly one such point and any other line don't cover this point (any three of them don't cross in the same point).

Solution of Geometric Set Cover for this instance would yield a sound solution of Vertex Cover for graph G. For every point (edge) we need to choose at least one of lines (vertices) v or w to cover this point.

Vertex Cover for arbitrary graph is APX-complete, so this problem in also APX-complete.

4.4. 2-approximation for arbitrary lines

Vertex Cover has an easy 2-approximation algorithm, but here very many lines can cross through the same point, so we can do d-approximation, where d is the biggest number of lines crossing through the same point. So for set where any 3 lines don't cross in the same point it yields 2-approximation.

The problematic cases are where through all points cross at least k points and all lines have at least k points on them. It can be created by casting k-grid in k-D space on 2D space.

Greedy algorithm yields $\log |\mathcal{R}|$ -approximation, but I have example for this for bipartial graph and reduction with taking all lines crossing through some point (if there are no more than k) would solve this case. So maybe it works.

Unfortunaly I haven't done this:(

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I can link some papers telling it's hard to do.

4.5. Connection with general set cover

Problem with finite set of lines with more dimensions is equivalent to problem in 2D, because we can project lines on the plane which is not perpendicular to any plane created by pairs of (point from C, line from P).

Of course every two lines have at most one common point, so is every family of sets that have at most one point in common equivalent to some geometric set cover with lines?

No, because of Desargues's theorem. Have to write down exactly what configuration is banned.

630 Chapter 5

Geometric Set Cover with polygons

5.1. State of the art

Covering points with weighted discs admits PTAS [Li and Jin, 2015] and with fat polygons with δ -extensions with unit weights admits EPTAS [Har-Peled and Lee, 2009].

Although with thin objects, even if we allow δ -expansion, the Set Cover with rectangles is APX-complete (for $\delta = 1/2$), it follows from APX-completeness for segments with δ -expansion in Section 3.2.

Covering points with squares is W[1]-hard [Marx, 2005]. It can be proven that assuming SETH, there is no $f(k) \cdot (|\mathcal{C}| + |\mathcal{P}|)^{k-\epsilon}$ time algorithm for any computable function f and $\epsilon > 0$ that decides if there are k polygons in \mathcal{P} that together cover \mathcal{C} , Theorem 1.9 in [Marx and Pilipczuk, 2015].

- 642 Chapter 6
- 643 Conclusions

Bibliography

- [Har-Peled and Lee, 2009] Har-Peled, S. and Lee, M. (2009). Weighted geometric set cover problems revisited. *Journal of Computational Geometry*, 3.
- [Håstad, 2001] Håstad, J. (2001). Some optimal inapproximability results. J. ACM, 48(4):798-859.
- [Li and Jin, 2015] Li, J. and Jin, Y. (2015). A PTAS for the weighted unit disk cover problem. CoRR, abs/1502.04918.
- [Marx, 2005] Marx, D. (2005). Efficient approximation schemes for geometric problems? In
 Brodal, G. S. and Leonardi, S., editors, Algorithms ESA 2005, pages 448-459, Berlin,
 Heidelberg. Springer Berlin Heidelberg.
- [Marx and Pilipczuk, 2015] Marx, D. and Pilipczuk, M. (2015). Optimal parameterized algorithms for planar facility location problems using voronoi diagrams. *CoRR*, abs/1504.05476.