

Optimal parameterized algorithms for planar facility location problems using Voronoi diagrams and sphere cut decompositions

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The square root phenomenon

Most parameterized problems can be solved faster on planar graphs:

	General graphs	Planar graphs
VERTEX COVER, k -PATH, ...	$2^{O(k)} \cdot n^{O(1)}$	$2^{O(\sqrt{k})} \cdot n^{O(1)}$
INDEPENDENT SET, DOMINATING SET, ...	$n^{O(k)}$	$2^{O(\sqrt{k})} \cdot n^{O(1)}$
STRONGLY CONNECTED STEINER SUBGRAPH, ...	$n^{O(k)}$	$n^{O(\sqrt{k})}$

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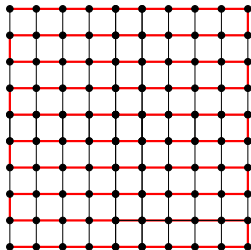
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This talk:

A general family of packing/covering problems on planar graphs and on 2D geometric objects that can be solved in time $n^{O(\sqrt{k})}$.

Bidimensionality for k -PATH

- Observation:** If the treewidth of a planar graph G is at least $5\sqrt{k}$
- \Rightarrow It has a $\sqrt{k} \times \sqrt{k}$ grid minor (Planar Excluded Grid Theorem)
 - \Rightarrow The grid has a path of length at least k .
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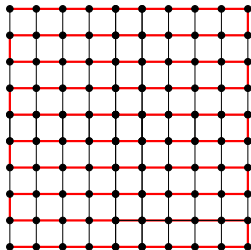


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We use this observation to find a path of length at least k on planar graphs:

- Set $w := 5\sqrt{k}$.
- Find an $O(1)$ -approximate tree decomposition.
 - If treewidth is at least w : we answer “there is a path of length at least k .”
 - If we get a tree decomposition of width $O(w)$, then we can solve the problem in time $2^{O(w \log w)} \cdot n^{O(1)} = 2^{O(\sqrt{k} \log k)} \cdot n^{O(1)}$.

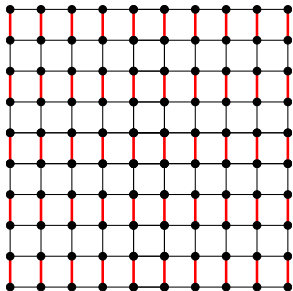


Bidimensionality

Definition

A graph invariant $x(G)$ is **minor-bidimensional** if

- $x(G') \leq x(G)$ for every minor G' of G , and
- If G_k is the $k \times k$ grid, then $x(G_k) \geq ck^2$ (for some constant $c > 0$).



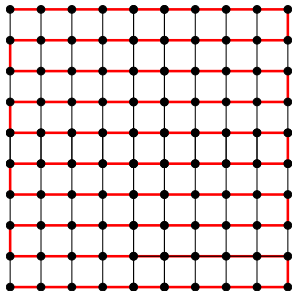
Examples: **minimum vertex cover**, length of the longest path, feedback vertex set are minor-bidimensional.

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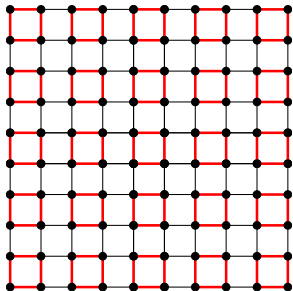
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Examples: minimum vertex cover, length of the longest path, **feedback vertex set** are minor-bidimensional.

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Algorithms based on bidimensionality:

- 1 If treewidth is $\Omega(\sqrt{k})$, then we can find a $\Omega(\sqrt{k}) \times \Omega(\sqrt{k})$ grid minor.
- 2 The problem is trivial if there is a $\Omega(\sqrt{k}) \times \Omega(\sqrt{k})$ grid minor.
- 3 If treewidth is $O(\sqrt{k})$, we can solve the problem in time $2^{O(\sqrt{k})} \cdot n^{O(1)}$.

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Variant of theory works for **contraction-bidimensional** problems, e.g., **INDEPENDENT SET**, **DOMINATING SET**.

However, for some problems, large treewidth (e.g., **MULTIWAY CUT**, **SUBSET TSP**) is not of any apparent help.

General principle

Exploit the fact that some auxiliary planar graph related to the solution has size $O(k)$ and hence treewidth $O(\sqrt{k})$.

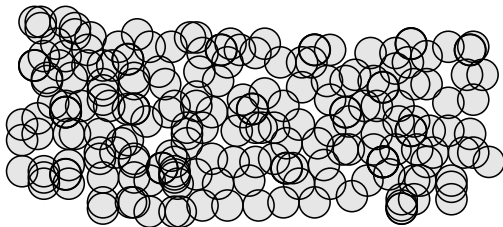
Outline

- A 2D geometric problem: **INDEPENDENT SET** problem for unit disks.
 - A simple $n^{O(\sqrt{k})}$ algorithm using shifting.
 - A more complicated algorithm via Voronoi diagrams (idea essentially comes from recent work on QPTASs for geometric problems, e.g., [Har-Peled SOCG 2014]).
- Several generalizations/variants.
- Planar graphs.
- Some lower bounds.

Independent Set for Unit Disks

Theorem [Alber and Fiala 2004]

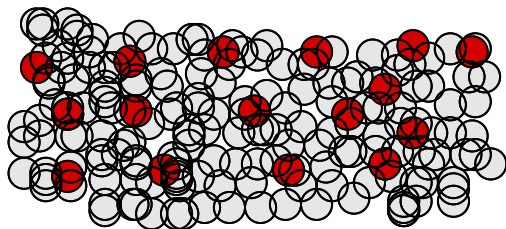
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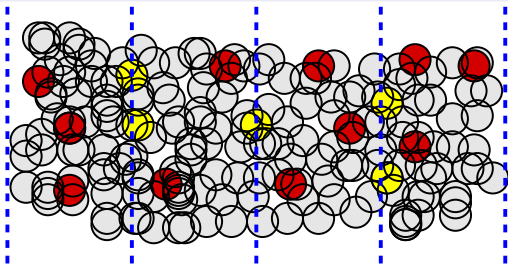
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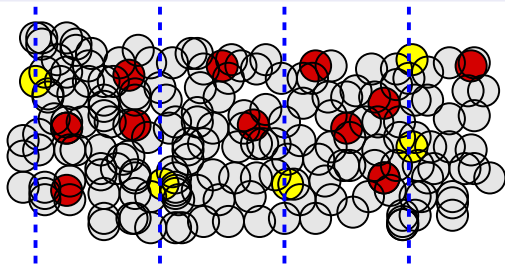
Simple solution by shifting strategy. Consider a family of vertical lines at distance $\lfloor \sqrt{k} \rfloor$ from each other, going through $(i, 0)$ for some integer $0 \leq i < \lfloor \sqrt{k} \rfloor$.

Claim: Exists i such that the lines hit $O(\sqrt{k})$ disks of the solution.

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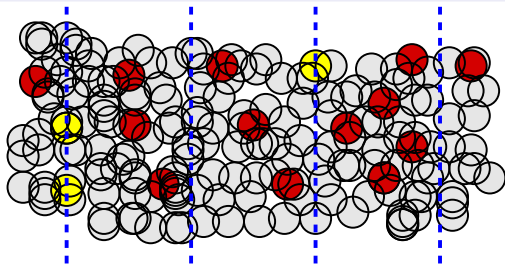
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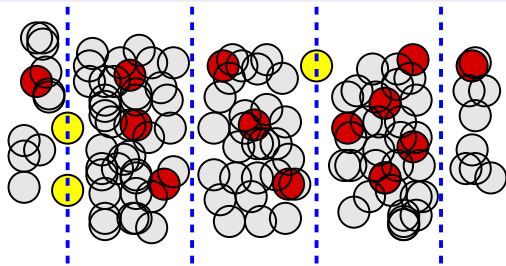
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Algorithm: Guess i and the $O(\sqrt{k})$ disks hit by the lines \Rightarrow Remove every disk intersected by the lines or disks \Rightarrow Problem falls apart into strips of height $O(\sqrt{k})$; can be solved optimally in time $n^{O(\sqrt{k})}$.

Challenges

Key idea

We were able to find a separator that hits $O(\sqrt{k})$ disks of the solution and breaks the instance in a nice way.

Two natural directions:

- 1 Can we solve **INDEPENDENT SET** for disks with arbitrary radius in time $n^{O(\sqrt{k})}$?
- 2 Can we solve **SCATTERED SET** (find k vertices that are at distance at least d from each other) on planar graphs in time $n^{O(\sqrt{k})}$, if d is part of the input?

Problem:

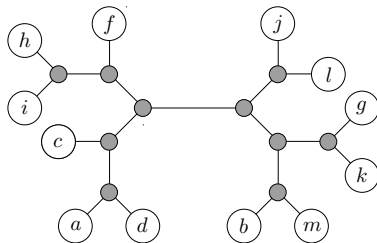
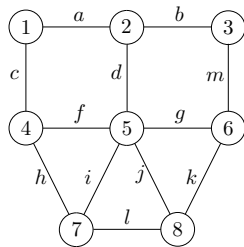
The algorithm for unit disks crucially uses the fact that the disks have similar area.

Branch Decompositions

Definition

A **branch decomposition** of a graph $G = (V, E)$ is a tuple (T, μ) where

- T is a tree with degree 3 for all internal nodes.
- μ is a bijection between the leaves of T and $E(G)$.

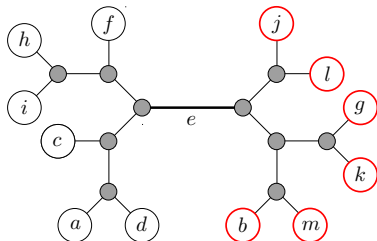
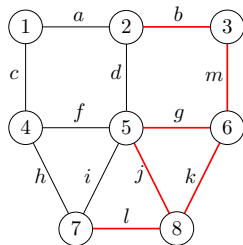


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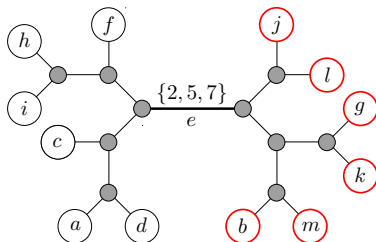
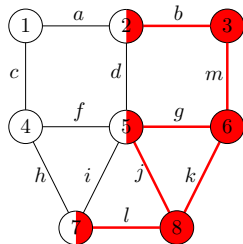
Edge $e \in T$ partitions the edge set of G into A_e and B_e

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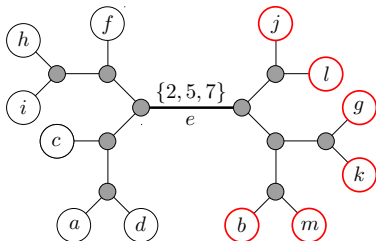
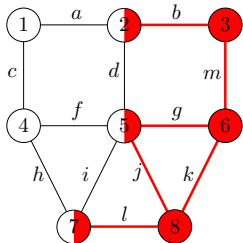
Middle set: $\text{mid}(e) = V(A_e) \cap V(B_e)$

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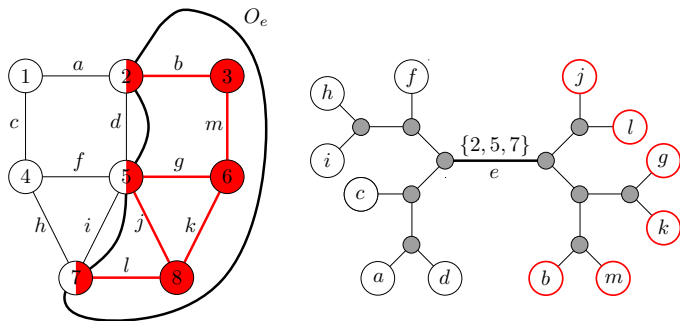


- The **width** of a branch decomposition is $\max_{e \in T} |\text{mid}(e)|$.
- The **branch width** of a graph G is the minimum width over all branch decompositions of G .

Sphere cut decomposition

Let G be a **planar** graph embedded on the sphere (or a plane) \mathcal{S}_0

A **sphere cut decomposition** of G is a branch decomposition (T, τ) where for every $e \in E(T)$, the vertices in $\text{mid}(e)$ are the vertices in a Jordan curve of \mathcal{S}_0 that meets no edges of G and goes through every face at most once (a noose).



Sphere cut decompositions

Theorem [Seymour-Thomas 1994, Dorn et al. 2005]

Every 2-edge connected planar graph G of branchwidth ℓ has a sphere cut decomposition of width ℓ . This decomposition can be constructed in $O(n^3)$ steps.

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α -edge-balanced noose: at most α fraction of the edges are inside/outside the noose.

Corollary

Every connected 3-regular planar multigraph with n vertices has a $\frac{2}{3}$ -edge-balanced noose of length $O(\sqrt{n})$.

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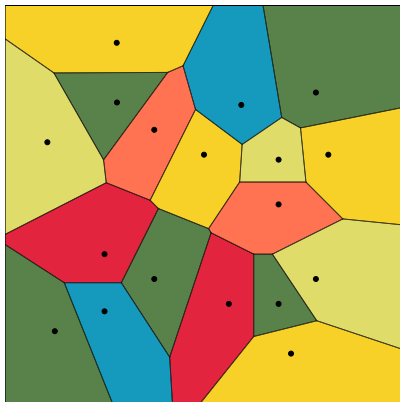
α -face-balanced noose: at most α fraction of the faces are strictly inside/outside the noose.

Corollary

Every connected 3-regular planar multigraph with n vertices has a $\frac{2}{3}$ -face-balanced noose of length $O(\sqrt{n})$.

Voronoi diagrams

Voronoi diagram: we partition the points of the plane according to the closest center.

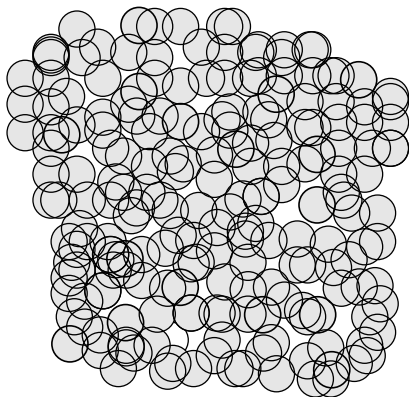


Observation: every cell is convex.

- Assume that the branch points of the diagram have degree 3.
- Ignore what happens at infinity.

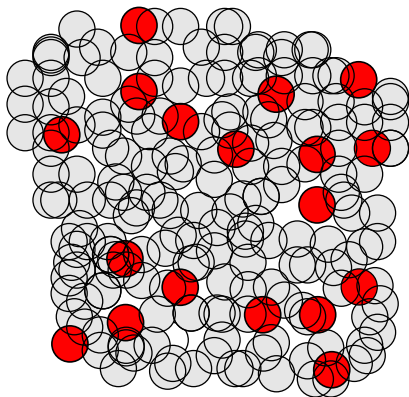
Voronoi separators

Consider the Voronoi diagram of the centers of the solution disks.



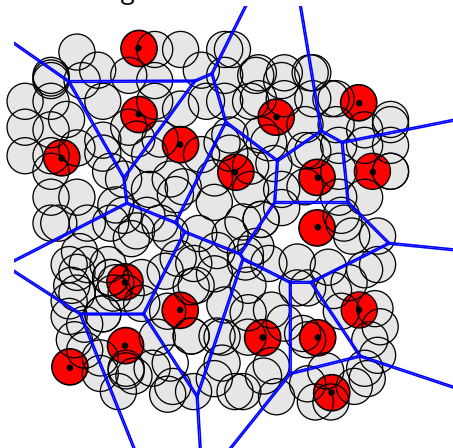
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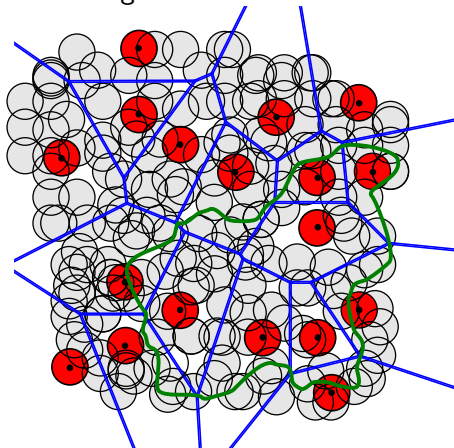
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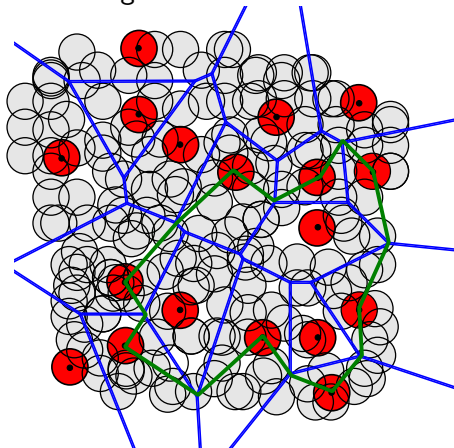
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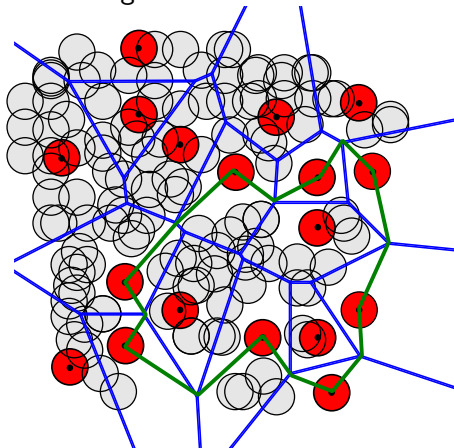


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\Rightarrow There is a corresponding polygon of length $O(\sqrt{k})$.

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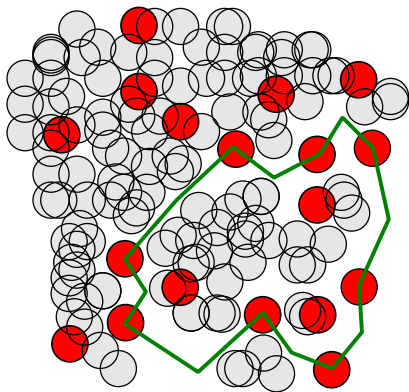
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Algorithm: guess $O(\sqrt{k})$ disks and a polygon going through them, remove any disks intersecting the polygon or the guessed disks, recursion on the inside and the outside.

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Algorithm: guess $O(\sqrt{k})$ disks and a polygon going through them, remove any disks intersecting the polygon or the guessed disks, recursion on the inside and the outside.

Running time

Number of candidate polygons

Number of centers: n .

Potential locations of Voronoi branch points: n^3 .

\Rightarrow Number of polygons of length $O(\sqrt{k})$: $n^{O(\sqrt{k})}$.

Recursion

$T(n, k)$: running time with n centers and solution size at most k .

$$T(n, k) = n^{O(\sqrt{k})} T(n, \frac{2}{3}k)$$

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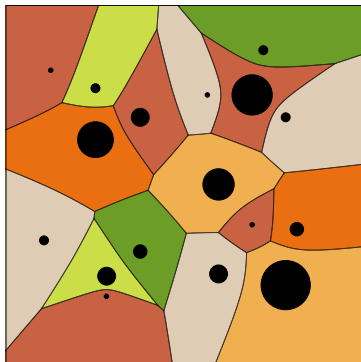
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This gives another $n^{O(\sqrt{k})}$ time algorithm for INDEPENDENT SET for unit disks. But what about general disks?

Additively weighted Voronoi diagrams

Each center c_i has a weight w_i and we classify each point v according to $\text{dist}(v, c_i) - w_i$.



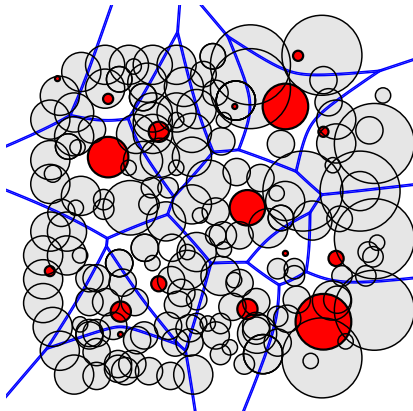
Alternative view: distance is measured from a disk of radius w_i centered at c_i .

Observation: The cells are star convex, that is, the segment between the center and a point of the cell is in the cell.

INDEPENDENT SET for general disks

Theorem

The **INDEPENDENT SET** problem for disks (of arbitrary radius) can be solved in time $n^{O(\sqrt{k})}$.

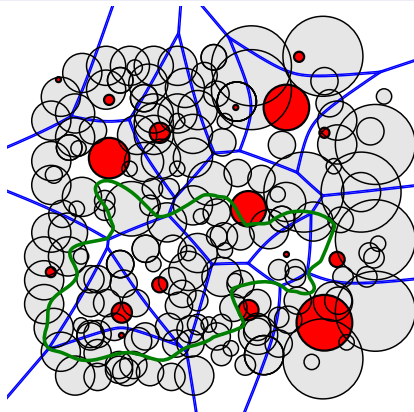


Use the additively weighted Voronoi diagram. Algorithm for unit disks go through.

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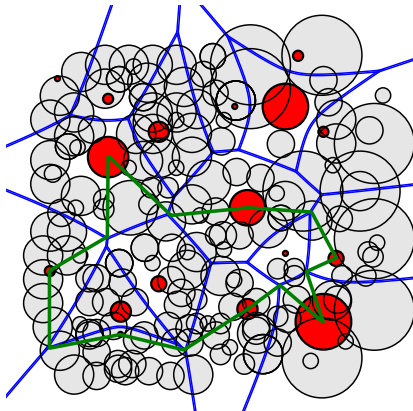


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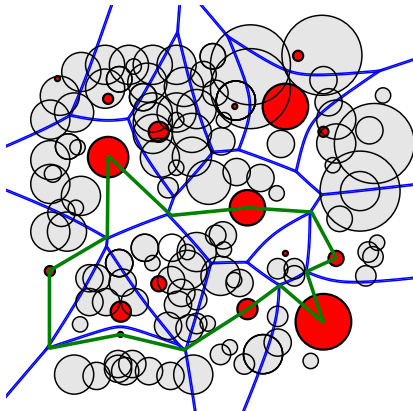


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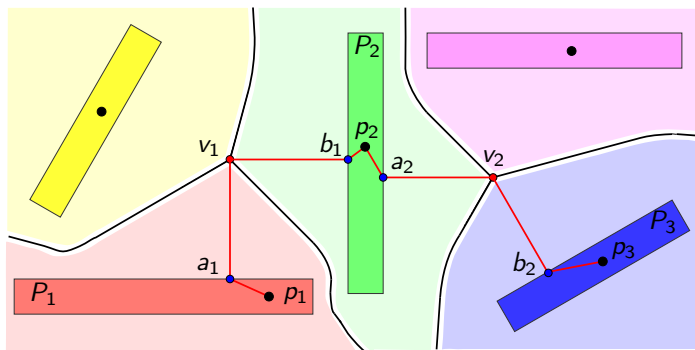


Use the additively weighted Voronoi diagram. Algorithm for unit disks go through.

INDEPENDENT SET for convex polygons

We can define Voronoi diagrams for arbitrary objects.

Problem: The cells are not necessarily star convex.



Fix: the polygon is now of the form

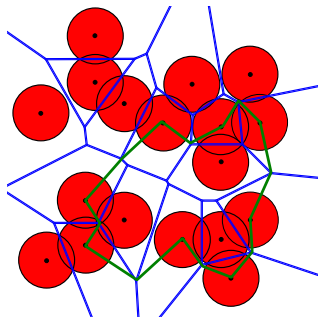
\dots — branch point — polygon boundary — polygon center — polygon boundary — branch point — \dots .

Can be further generalized to nonconvex polygons.

Covering points with unit disks

Task: Given n unit disks and m points, select k disks covering the maximum number of points in total.

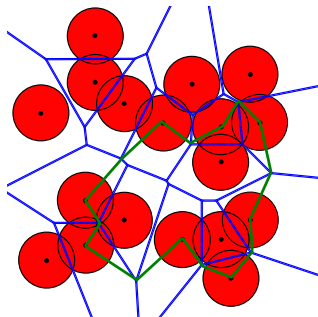
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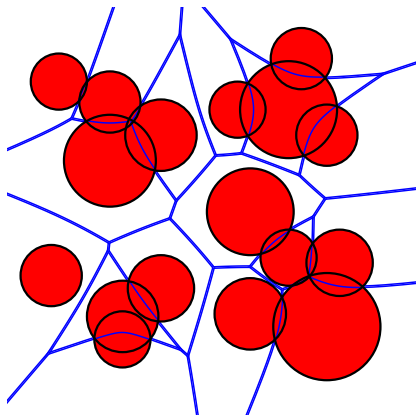


We guess $O(\sqrt{k})$ centers and a polygon.

- Remove the points covered by the selected centers.
- Remove any center that is closer to a point of the polygon than the selected centers.
- We solve recursively the inside/outside (why?)

Covering points with arbitrary disks

Task: Given n arbitrary disks and m points, select k disks covering the maximum number of points in total.



Can be solved in time $n^{O(\sqrt{k})} m^{O(1)}$ using the additively weighted Voronoi diagram.

Planar graphs

So far we have considered

- selecting k disjoint connected objects in the plane,
- selecting k disks covering the maximum number of points.

Similar algorithms can be worked out for planar graphs.

Packing problems:

Theorem

Given a set \mathcal{D} of connected subgraphs in a planar graph, we can find k vertex-disjoint subgraphs in time $|\mathcal{D}|^{O(\sqrt{k})} \cdot n^{O(1)}$.

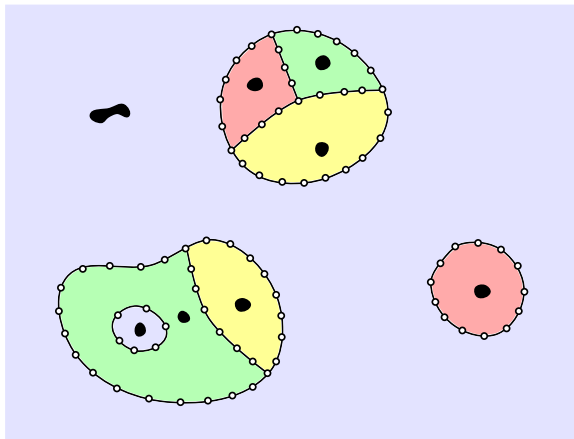
Covering problems:

Theorem

Given a set \mathcal{D} of center points equipped with a radius and a set \mathcal{C} of client points, we can find k center points satisfying the maximum number of client points in time $|\mathcal{D}|^{O(\sqrt{k})} \cdot n^{O(1)}$.

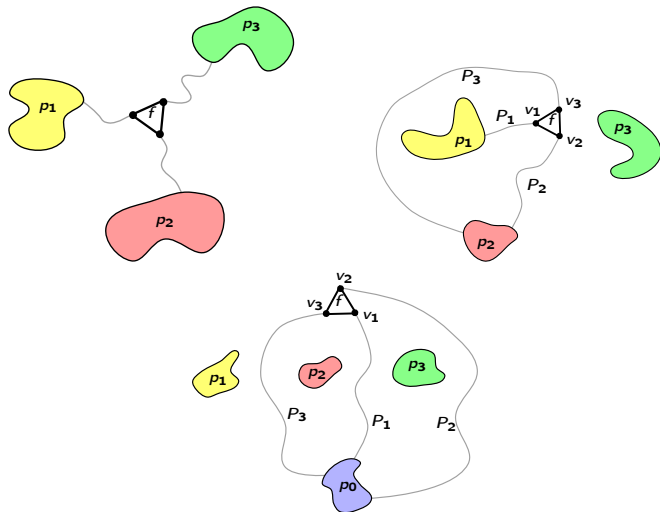
Planar graphs — challenges

The (analog of) Voronoi diagram is not necessarily connected.



Planar graphs — challenges

We need to define the analog of branch points and find candidate branch points efficiently (to have $|\mathcal{D}|^{O(\sqrt{k})}$ instead of $n^{O(\sqrt{k})}$).



The general planar graph problem

We have the following setting:

- A planar graph with weighted edges.
- Set \mathcal{D} of connected subgraphs, equipped with cost and radius.
- Set \mathcal{C} of client points, equipped with sensitivity and prize.
- Task: find a set of exactly k vertex-disjoint objects from \mathcal{D} .
- An object satisfies a client point if

$$\text{distance} \leq \text{radius} + \text{sensitivity}.$$

- We want to maximize the total prize minus the total cost.

Theorem

The general problem on planar graphs can be solved in time $|\mathcal{D}|^{O(\sqrt{k})} \cdot n^{O(1)}$ (where n is the length of the input).

Consequences

Packing problems:

Theorem

SCATTERED SET on planar graphs can be solved in time $n^{O(\sqrt{k})}$.

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Covering problems:

Theorem

d -DOMINATING SET on planar graphs can be solved in $n^{O(\sqrt{k})}$ (d is part of the input).

Theorem

Given a set \mathcal{D} of metric balls in a planar graph, we can find k vertex-disjoint balls in time $|\mathcal{D}|^{O(\sqrt{k})} \cdot n^{O(1)}$.

Consequences

Packing problems:

Theorem

SCATTERED SET on planar graphs can be solved in time $n^{O(\sqrt{k})}$.

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Given a set \mathcal{D} of connected subgraphs in a planar graph, we can find k vertex-disjoint subgraphs in time $|\mathcal{D}|^{O(\sqrt{k})} \cdot n^{O(1)}$.

Covering problems:

What about covering points by arbitrary objects?

Lower bounds

Strong Exponential Time Hypothesis (SETH):

No $(2 - \epsilon)^n$ time algorithm for CNF-SAT.

Theorem

Given a set \mathcal{D} of convex polygons and a set \mathcal{C} of points, there is no $f(k)(|\mathcal{D}| + |\mathcal{C}|)^{k-\epsilon}$ time algorithm for the problem of covering \mathcal{C} with k polygons from \mathcal{D} (unless SETH fails).

Lower bounds

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We use the following result:

Theorem [Pătraşcu and Williams 2010]

There is no $f(k)n^{k-\epsilon}$ time algorithm for DOMINATING SET (unless SETH fails).

Lower bounds

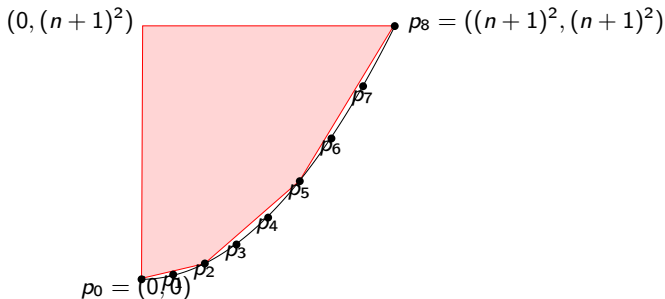
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Transparent reduction from DOMINATING SET:



Lower bounds

Exponential Time Hypothesis (ETH):

No $2^{o(n)}$ time algorithm for n -variable 3SAT.

Theorem

Given a set \mathcal{D} of axis-parallel rectangles and a set \mathcal{C} of points, there is no $f(k)(|\mathcal{D}| + |\mathcal{C}|)^{o(k)}$ time algorithm for the problem of covering \mathcal{C} with k rectangles from \mathcal{D} (unless ETH fails).

Lower bounds

Exponential Time Hypothesis (ETH):

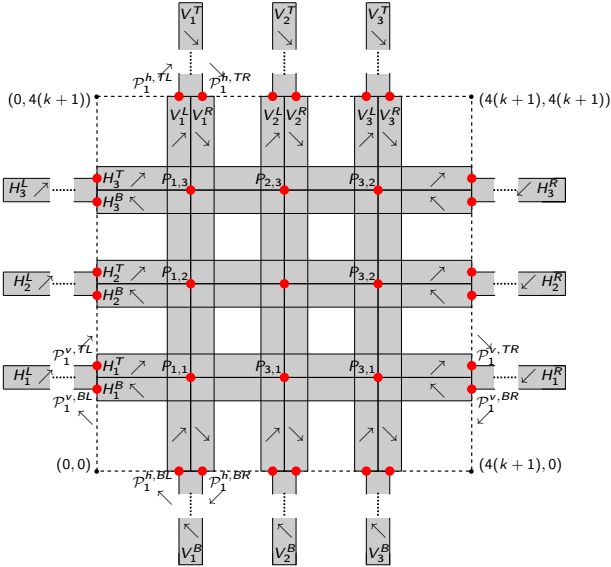
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Theorem

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- Remains true if every rectangle is either $1 \times k$ or $k \times 1$.
- Slightly weaker bound $f(k)(|\mathcal{D}| + |\mathcal{C}|)^{o(k/\log k)}$ remains true if every rectangle has width and height in the range $[1, 1 + \epsilon]$.

Lower bounds



Summary

- Main result: $n^{O(\sqrt{k})}$ time algorithms for natural packing/covering problems.
- Main technical tool: find balanced separators in the Voronoi diagram of the solution (comes from recent results on QPTASs for geometric problems, e.g., [Har-Peled SOCG 2014]).
- Lower bounds show that the square root phenomenon does not hold for certain natural covering problems.
- It seems essential that the covering problems are defined in a metric way.