Subquadratic diameter computation in graphs of bounded VC-dimension

Timothy Chan, Hsien-Chih Chang, Jie Gao, Hung Le, Sándor Kisfaludi-Bak, Da Wei Zheng

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Given a simple undirected graph G on n vertices and m edges, and the corrsponding shortest-path metric,

Diameter

Find the maximum distance among vertices

Eccentricites

For each vertex v find the minimum radius of ball at v that covers G.

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Lower bound by Roditty and V. Williams (2013): No $O(n^{2-\varepsilon})$ algo to decide diameter 2 vs 3 in sparse graphs under SETH.

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Cabello '18, Gawrychowski et al. '21: diameter of **planar** graphs in $\widetilde{O}(n^{5/3})$

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Is there a subquadratic algorithm for diameter in unit disk graphs?

Results: Subquadratic algorithms for diameter

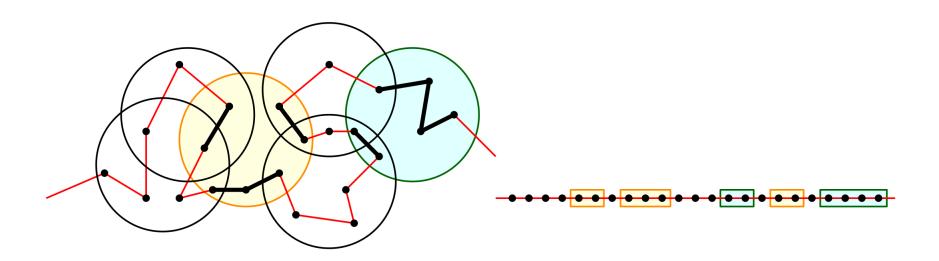
graph class	best previous		new
planar	$\widetilde{O}(n^{5/3})$	[Cab18, GKM ⁺ 21]	
K_h -minor-free	$\widetilde{O}(n^{2-1/(3h-1)})$	[DHV22, LW24]	$\widetilde{O}(n^{2-1/(2h-2)})$
VC-dimbounded	$\widetilde{O}(\min\{Dmn^{1-1/d}, mn\})$	[DHV22, DKP24]	$\widetilde{O}(mn^{1-1/(2d)})$
unit square	$\widetilde{O}(\min\{Dn^{7/4}, n^2\})$	[DKP24]	$O^*(n^{2-1/8})$
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- We don't need separators!
- More efficient ball growing process based on very simple LDD
- New data structures and ideas for geom. intersection graphs

Implicit representations



Diameter $\leq D$ iff all balls in $\mathcal{N}^D := \{N^D(v) \mid v \in G\}$ cover V(G). Strategy: grow balls to compute all balls $\mathcal{N} = \bigcup_r \mathcal{N}^r$.

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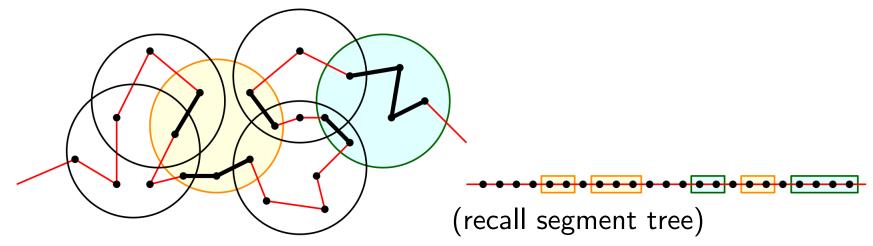
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Implicit representation of disks via geometric stabbing path:



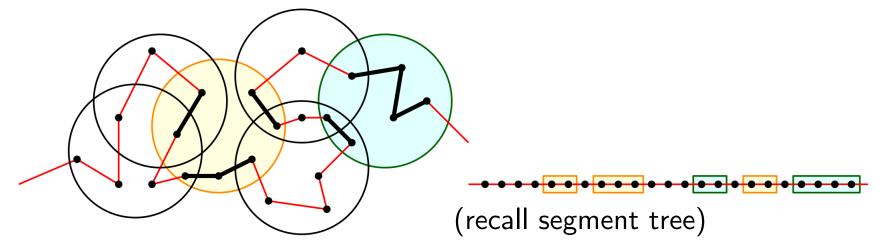
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A stabbing path of G is any complete ordering λ of V(G). To represent $X\subset V(G)$, list its elements as intervals of λ (denoted: $\operatorname{Rep}_{\lambda}(X)$)

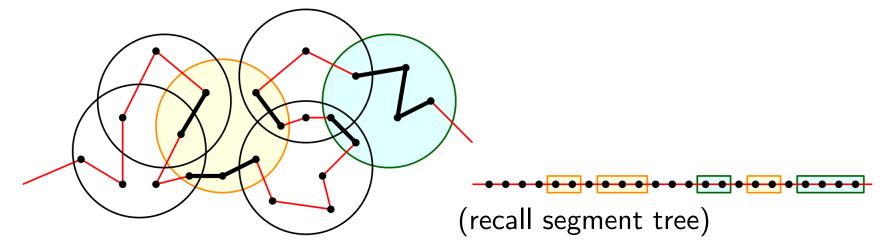
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We want subquadratic representation in amortized sense:

$$\sum_{X\in\mathcal{N}}|\mathsf{Rep}_{\lambda}(X)|=O(n^{1.999}).$$

Given a set family \mathcal{X} over universe V, the set $A \subset V$ is *shattered* by \mathcal{X} if: for all $S \subset A$ there is $X \in \mathcal{X}$ s.t. $X \cap A = S$.

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The *VC dimension* of \mathcal{X} is the maximum |A| such that A is shattered by \mathcal{X} .

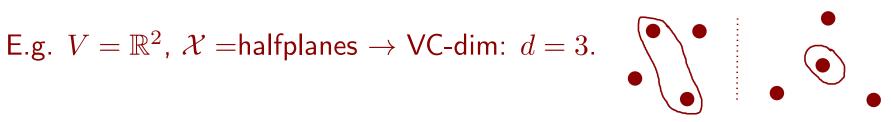
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G:= undirected unweighted graph, $N^r(v)=$ ball of radius r at $v\in V(G)$.

The distance VC-dimension d of G is the VC dimension of $\mathcal{N} := \bigcup_r \mathcal{N}^r(v)$.

Lemma (Informal)

Given* a set system (V, \mathcal{N}) of VC-dimension** d we can construct a stabbing path λ in $\widetilde{O}(n^{1+1/d})$ time such that with high*** probability:

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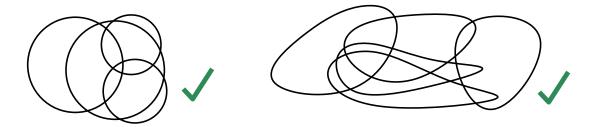
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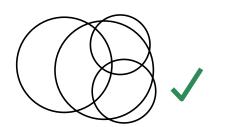
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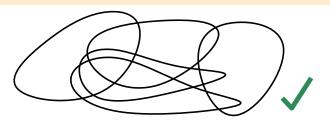
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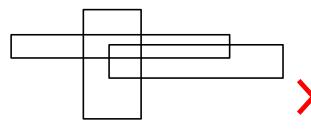
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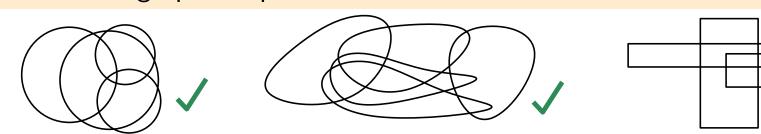
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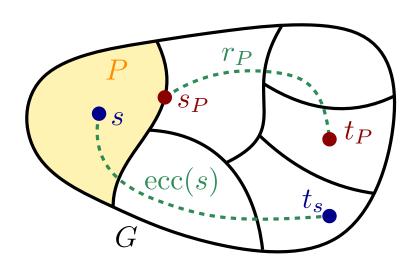
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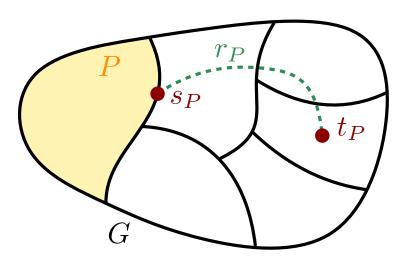


Generic framework and data structures



Step 1. Fix Δ , decompose G into $\widetilde{O}(n/\Delta)$ pieces P of (strong) diameter $\leq \Delta$ s.t. total boundary is small: $\sum_P |\partial P| = \widetilde{O}(n/\Delta)$.

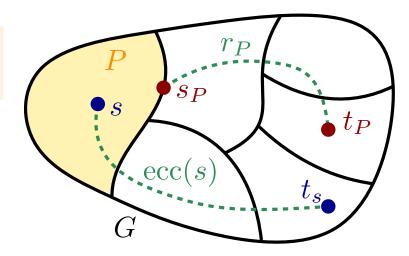
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By triangle-ineq:

 $\operatorname{dist}_{G}(s_{P}, t) - \Delta \leq \operatorname{dist}_{G}(s, t) \leq \operatorname{dist}_{G}(s_{P}, t) + \Delta$



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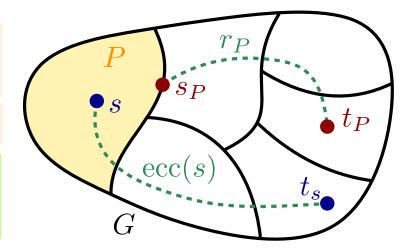
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The relevant region of P is

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The modified r-ball of $s \in P$ is $\widehat{N}^r(s) := N^r(s) \cap R_P$. Let $\widehat{\mathcal{N}} := \{\widehat{N}^r(s)\}_{r,s}$.

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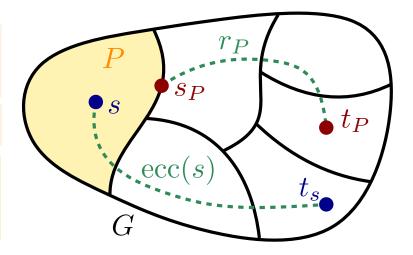
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Lemma

The modified ball system $\widehat{\mathcal{N}}$ has the same VC-dim as \mathcal{N} and $|\widehat{\mathcal{N}}| = O(\Delta n)$

Step 3. Compute stabbing path λ for $\hat{\mathcal{N}}$. For each P and each $s \in P$, from $r = r_P - 3\Delta - 1$ to $r = r_P + \Delta$, compute $\operatorname{Rep}_{\lambda}(\hat{N}^r(s))$.

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In dense graphs, we need to solve:

Interval Searching

Given set of objects \mathcal{O} , each associated with some integer intervals of [1:n], design a data structure that:

for query $q \in \mathcal{O}$ returns the union of the representations of the objects $o \in \mathcal{O}$ that intersect q.

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We need*: $\widetilde{O}(\text{input size})$ preprocessing and $\widetilde{O}(\text{output size})$ query time

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Reduction with polylog preprocessing and query overhead

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Hard and non-decomposable range searching...

 \sim Reduction with $n^{o(1)}$ preprocessing and query overhead Slice into blocks of size b.

DSP3: Rainbow colored intersection searching

Given objects \mathcal{O} and $C: \mathcal{O} \to [n]$ (color), design a data structure that: for query $q \in \mathcal{O}$ decides if all colors appear in $\{C(o) \mid o \in \mathcal{O} \text{ intersects } q\}$.

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for query $q \in \mathcal{O}$ returns the union of the representations of the objects $o \in \mathcal{O}$ that intersect q.

Reduction with polylog preprocessing and query overhead

DSP2: Interval Cover

Given objects \mathcal{O} and $I:\mathcal{O}\to [1:n]$, design a data structure that: for query $q\in\mathcal{O}$ decides if the union of $\{I(o)\mid o\in\mathcal{O} \text{ intersects } q\}$ covers I(q).

Hard and non-decomposable range searching...

 \sim Reduction with $n^{o(1)}$ preprocessing and query overhead Slice into blocks of size b.

DSP3: Rainbow colored intersection searching

Given objects \mathcal{O} and $C: \mathcal{O} \to [n]$ (color), design a data structure that: for query $q \in \mathcal{O}$ decides if all colors appear in $\{C(o) \mid o \in \mathcal{O} \text{ intersects } q\}$.

Theorem

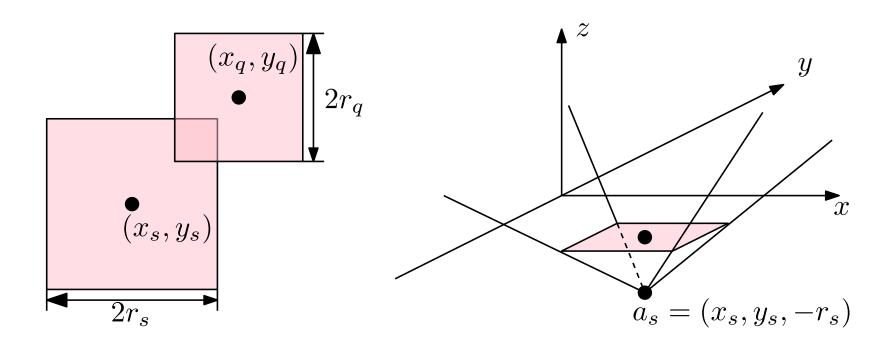
If we can construct in $\widetilde{O}(|\mathcal{O}_{RC}|)$ time a data structure \mathcal{D}_{RC} with $\widetilde{O}(1)$ query for DSP3, then for any $b \in [1,n]$, we can construct a data structure for DSP2 with total run time $\widetilde{O}(N_{IC} \cdot b + L_{IC}/b)$.

Rainbow colored intersection searching

Given squares \mathcal{O} and coloring $C:\mathcal{O}\to[n]$, design a data structure that: given square $q\in\mathcal{O}$ decides if all colors appear in $\{C(o)\mid o\in\mathcal{O} \text{ intersects } q\}$.

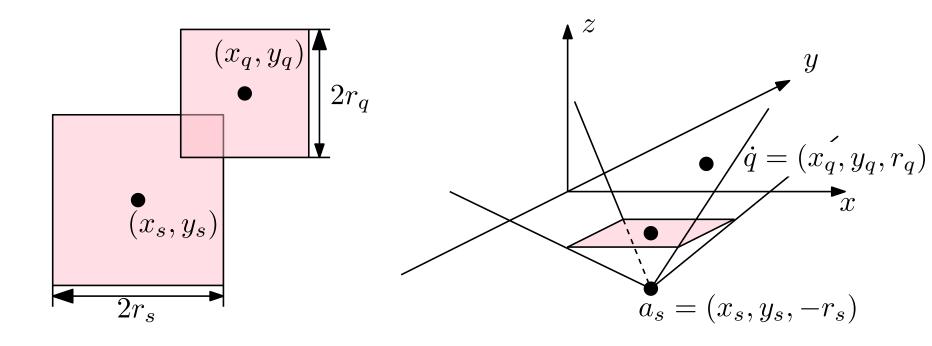
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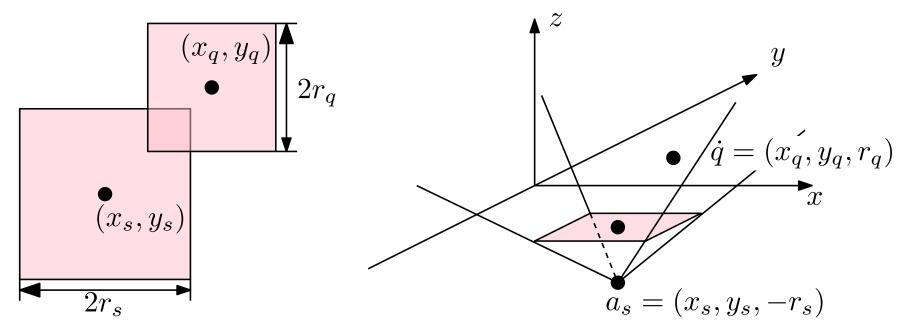
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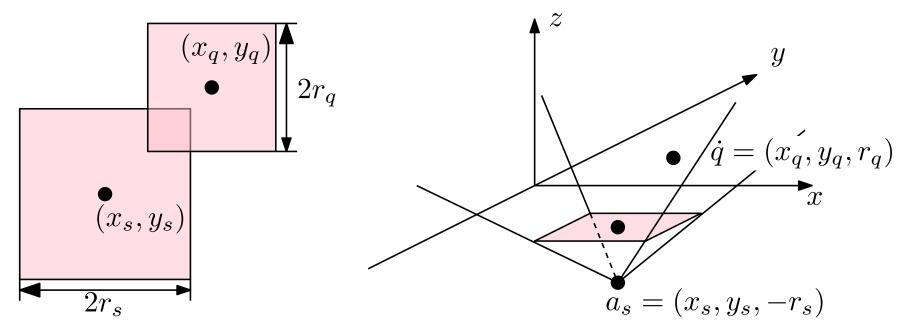


Rough idea:

• q has intersection with color class i iff \dot{q} is above lower envelope of color-i cones.

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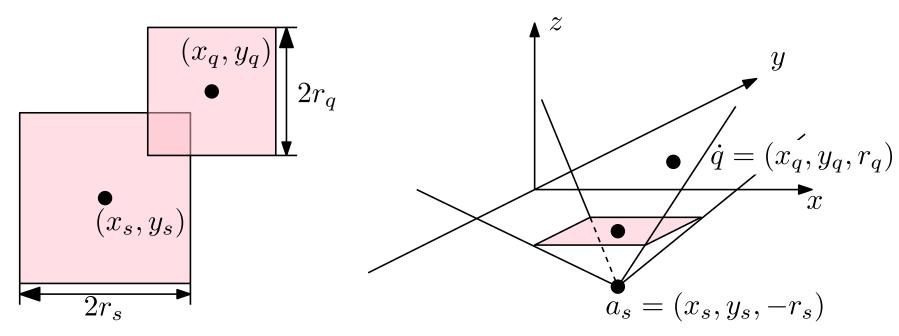


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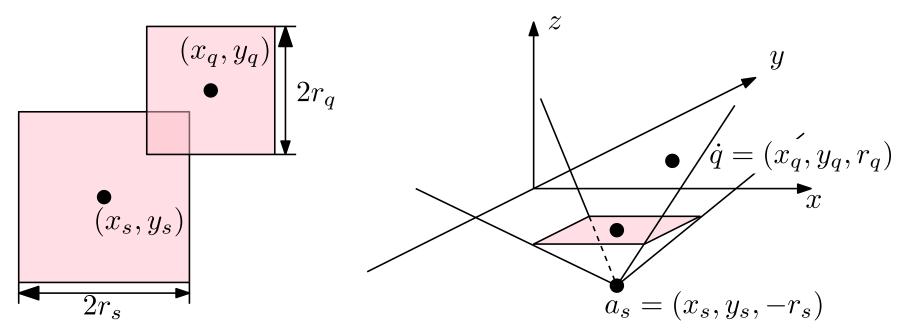


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- for each color class, compute lower envelope of cones
- slice the space above lower envelope into pw disjoint slabs of fixed directions
- build (reverse) range counting data structure on all slabs

Running time analysis for squares

Stabbing path λ gives:

$$\sum_{P} \sum_{s \in P} \sum_{r=r_P-2\Delta}^{r_P+\Delta} |\mathsf{Rep}_{\lambda}(\hat{N}^r[s])| \ = \ \widetilde{O}(\Delta \cdot n^{2-1/d}) = \widetilde{O}(\Delta \cdot n^{7/4}) \quad (d=4)$$

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The total length of all the intervals for $s \in P$ is at most $2|P| \cdot |R_P| = O(|P| \cdot n)$. Ball growing running time for fixed P, r:

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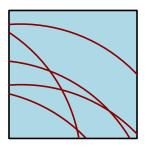
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BFSes on
$$\partial P$$
 Constructing λ
$$\widetilde{O}(n^2/\Delta + n^{5/4}) + \sum_{P} \sum_{r=r_P-2\Delta}^{r_P+\Delta} \widetilde{O}\left(b \cdot \sum_{s \in P} \left((|\mathsf{Rep}_{\lambda_P}(\hat{N}^{r-1}[s])| + |\mathsf{Rep}_{\lambda_P}(\hat{N}^{r}[s])|\right) + |P|n \right)$$

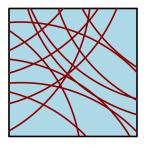
$$= \widetilde{O}(n^2/\Delta + n^{5/4}) + \widetilde{O}(b\Delta \cdot n^{7/4}) + \widetilde{O}(n^2\Delta/b) \quad (d=4)$$

$$= \widetilde{O}(n^{2-1/16}). \quad \text{(for optimal choices of } b = \Delta^2 \text{ and } \Delta = n^{1/16})$$

Base data strucutre problem is related to Hopcroft's problem, $\Omega(n^{1/3})$ query time likely.

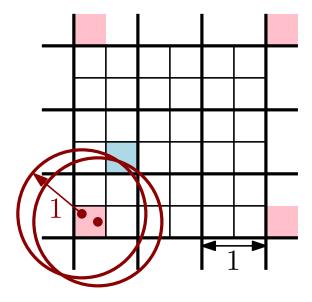


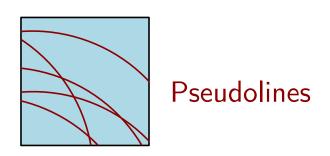
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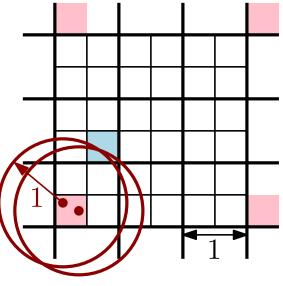
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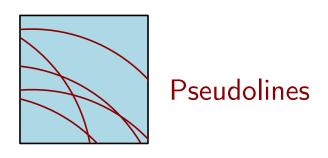




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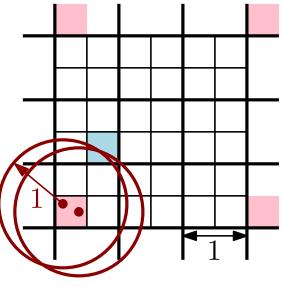


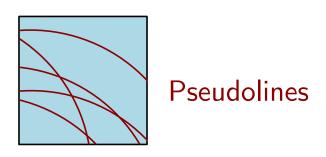


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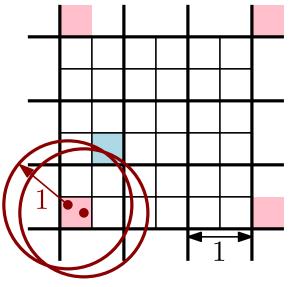


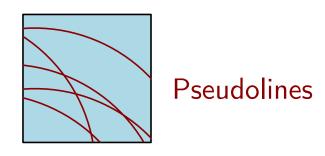


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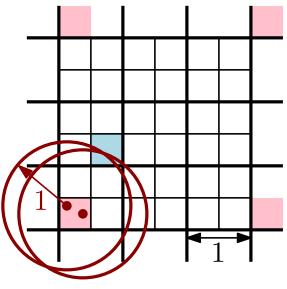


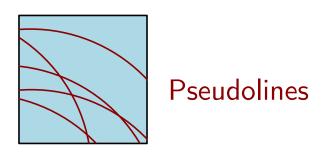


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 - \rightarrow Fortunately, the combined set systems still have VC-dimension ≤ 8 .
- Switching stabbing paths is too costly if pieces are small.
 - \rightarrow Work only with pieces larger than a threshold; for small pieces, we switch to a different algo (based on distance compression)

VC-dim d + efficient interval cover DS $\leadsto O(n^{2-f(d)})$ diameter computation/distance oracle

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There is some f(d) > 0 such that if \mathcal{G} is a (nice?) graph class with distance VC-dim $\leq d$, then there is a $O(n^{2-f(d)})$ time algorithm for diameter.

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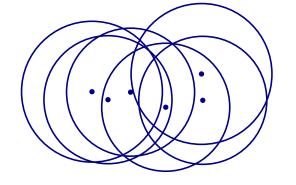
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Diameter-2 for disks or radii $[1, 1 + \varepsilon]$?



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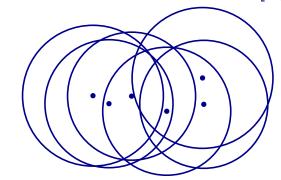
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Thanks for listening!

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Diameter:

graph class	best previous		new
planar	$\widetilde{O}(n^{5/3})$	[Cab18, GKM ⁺ 21]	
K_h -minor-free	$\widetilde{O}(n^{2-1/(3h-1)})$	[DHV22, LW24]	$\widetilde{O}(n^{2-1/(2h-2)})$
VC-dimbounded	$\widetilde{O}(\min\{Dmn^{1-1/d}, mn\})$	[DHV22, DKP24]	$\widetilde{O}(mn^{1-1/(2d)})$
unit square	$\widetilde{O}(\min\{Dn^{7/4}, n^2\})$	[DKP24]	$O^*(n^{2-1/8})$
arbitrary square	$\widetilde{O}(n^2)$	[CS19]	$\widetilde{O}(n^{2-1/12})$
unit disk	$O(n^2\sqrt{\frac{\log\log n}{\log n}})$	[CS16]	$O^*(n^{2-1/18})$

Distance oracle (construction time/space) Query: $\widetilde{O}(1)$.

graph class	best previous		new
planar	$n^{3/2+o(1)}, n^{1+o(1)}$	[CGL ⁺ 23]	
K_h -minor-free	$\widetilde{O}(n^{2-1/(3h-1)})$	[LW24]	
VC-dimbounded	$O(mn), O(n^2)$	folklore	$\widetilde{O}(mn^{1-1/(4d+1)})$
unit square	$\widetilde{O}(n^2)$	[CS19]	$O^*(n^{2-1/16})$
arbitrary square	$\widetilde{O}(n^2)$	[CS19]	$\widetilde{O}(n^{2-1/20})$
unit disk	$O(n^2\sqrt{\frac{\log\log n}{\log n}})$	[CS16]	$O^*(n^{2-1/20})$