

# Approximating Densest Subgraph in Geometric Intersection Graphs

Sariel Har-Peled<sup>1</sup>    **Saladi Rahul**<sup>2</sup>

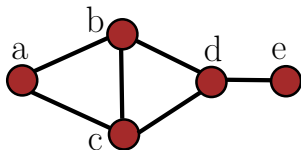
<sup>1</sup>University of Illinois Urbana Champaign (UIUC)

<sup>2</sup>Indian Institute of Science (IISc),

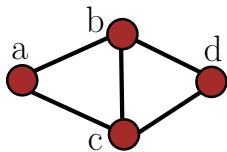
March 4, 2025

# Densest Subgraph

Density  $\frac{6}{5} = 1.2$



- Report the subset of  $V$  with the *maximum density*

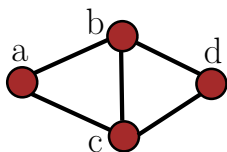
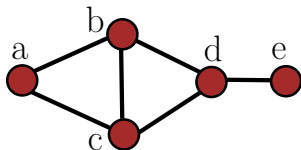


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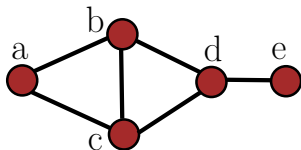
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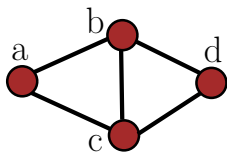
- Report the subset of  $V$  with the *maximum density*
- Undirected graph  $G = (V, E)$
- For any  $S \subseteq V$ , its density =  $\frac{|E_S|}{|S|}$
- Each edge in  $E_S \subseteq E$  has both its vertices in  $S$

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  - Link-spam detection
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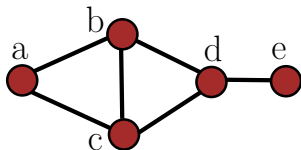


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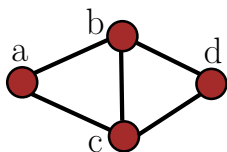
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- Tommaso Lanciano, Atsushi Miyauchi, Adriano Fazzone, and Francesco Bonchi. *A survey on the densest subgraph problem and its variants.*



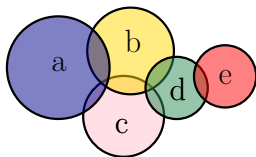
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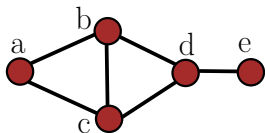
# Densest Subgraph

- Can be solved exactly in polynomial time
  - [Goldberg'84], [Gallo, Grigoriadis, Tarjan'89], [Charikar'00], [Khuller, Saha'09]
- 2-approximation algorithm
  - [Asahiro, Iwama, Tamaki and Tokuyama'00]
  - Analyzed by [Charikar'00]
- $(1 + \epsilon)$ -approximation algorithm
  - [Bahmani, Goel, Munagala'14]
- *Message:*  $\Omega(|E|)$  time taken by all the algorithms

# Densest Subgraph in Disk Intersection Graphs

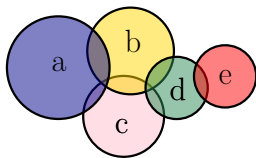


- Collection of  $n$  disks in the plane
- *Disk Intersection Graph*

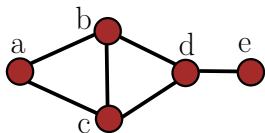


Can near-linear time (in terms of  $n$ ) algorithms be designed?

# Densest Subgraph in Disk Intersection Graphs



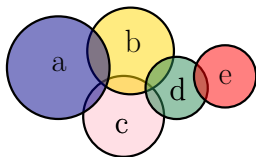
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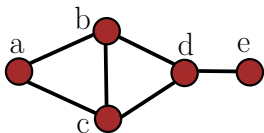
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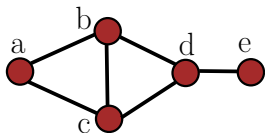
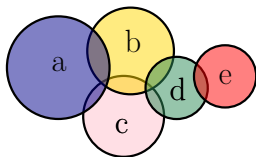


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# Densest Subgraph in Disk Intersection Graphs



- Collection of  $n$  disks in the plane
- *Disk Intersection Graph*
- A vertex associated with each disk.
- Two vertices have an edge iff the corresponding disks intersect.
- *Implicit* disk intersection graph:
  - Only the disks are given as input
  - Edges are not known *explicitly*

Can near-linear time (in terms of  $n$ ) algorithms be designed?

# Our Results for Disks

| <i>Approximation</i> | <i>Running Time</i>   |
|----------------------|---|
| $(2 + \varepsilon)$  | $O\left(\frac{n \log n}{\varepsilon^4}\right) \approx O_\varepsilon(n \log n)$  |
| $(1 + \varepsilon)$  | $O\left(\frac{n \log^2 n}{\varepsilon^2} \left(\frac{1}{\varepsilon^2} + \log \log n\right)\right) \approx \tilde{O}_\varepsilon(n \log^2 n)$ |

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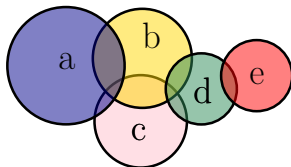
*The  $(2 + \varepsilon)$ -approximation  
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# The $(2 + \varepsilon)$ -approximation algorithm

- *Idea-1*: Connection with *deepest point*
- *Idea-2*: Efficiently throwing away *low-degree* vertices

## Idea-1: Connection with Deepest Point

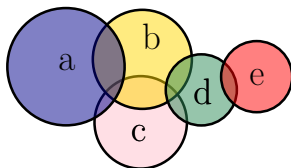
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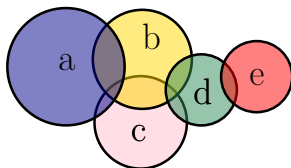
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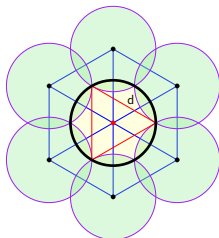
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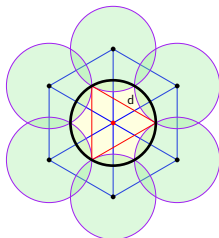


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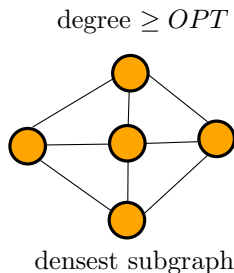
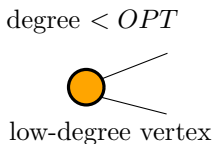


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  - A 1.1-approximation can be computed in  $O(n \log n)$  time [Aronov and Har-Peled'08].
- *Lemma*:  $\frac{\text{deepest}-1}{2} \leq \text{OPT} \leq 7 \cdot \text{deepest}$ .
  - Cute discrete geometry problem.

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- *Idea-1*: Connection with *deepest point*
- *Idea-2*: Efficiently throwing away *low-degree* vertices

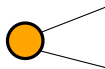
## Idea-2: low-degree vertex



- *Observation:* None of the vertices in the optimal solution are low degree.
- *Algorithm's goal:* Remove low degree vertices quickly from the graph.

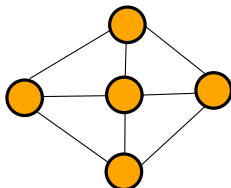
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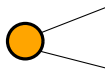


densest subgraph

- *Case 1*: # low-degree vertices  $\leq \epsilon n$ .
- *Most* vertices in the graph have degree  $\geq OPT$ .
- Entire graph is a good approximation
  - density =  $\frac{\#edges}{n} \geq \frac{OPT \cdot (1-\epsilon)n}{2n} = \frac{OPT}{2+\epsilon'}$

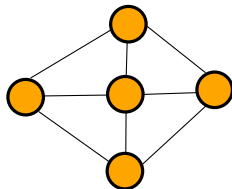
## Idea-2: Many low-degree vertices

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densest subgraph

- *Case 2*: # low-degree vertices  $\geq \epsilon n$ .
- Need to throw away such vertices.
- *Recurse* on the remaining graph.
- *Luckily*, number of recursive steps  $= O(\epsilon^{-1} \log n)$ .

# Running time

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- # *iterations* =  $O(\varepsilon^{-1} \log n)$ .
- *Running time* =  $O(\sum_i n_i \log n_i) = O_\varepsilon(n \log n)$ , since  $n_i \leq (1 - \varepsilon)n_{i-1}$ . *Optimal!*

*The  $(1 + \varepsilon)$ -approximation  
algorithm*

# The $(1 + \varepsilon)$ -approximation algorithm

- *Idea-1: Sampling  $O_\varepsilon(n \log n)$  edges suffices*
- *Idea-2: Efficient *data structure* for sampling edges*

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- Let  $E_S$  be the sampled edges
- Run an approximation algorithm on *sparse graph*  $(V, E_S)$



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- *Key contribution*: Generic technique to use *range reporting* data structures into samples edges (almost)-*uniformly* at random.
- *Disks*: Preprocessing time  $O_\epsilon(n \log^2 n)$ , sample takes  $\tilde{O}_\epsilon(\log n)$  time

*Final Comments*

# Other Geometric Intersection Graphs

- *Prerequisite*: Efficient range reporting data structures.
- Can obtain  $(1 + \varepsilon)$ -approximation and  $(2 + \varepsilon)$ -approximation.
- *Axis-aligned boxes* in  $d$ -dimensions
  - $n \log^{O(d)} n$  running time.
- *Balls* in  $d$ -dimensions
  - $O(n^{2-\lambda})$  running time for some  $\lambda \in (0, 1)$ .

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  - Update time: *sub-linear* in  $n$ ?
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  - Unit-disk graphs or Interval graphs?
- *Dynamic* densest subgraph under insertion and deletion of disks
  - Update time: *sub-linear* in  $n$ ?
  - Several edges are *implicitly* inserted or deleted in each round.
- Other variants of densest subgraph in geometric intersection graphs.

*Thank You*