

From binary search through games to graphs

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Outline



- 1 (A little dubious) motivation via multi-criteria optimisation
- 2 Multidimensional binary search as a game
- 3 Switch to graphs: edge and pair queries

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Let us start with a definition!



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Definition (Renyi)

Mathematician is a device that turns coffee into theorems.

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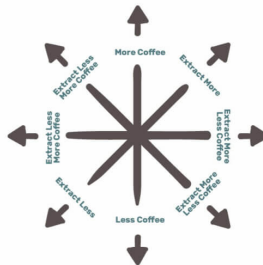
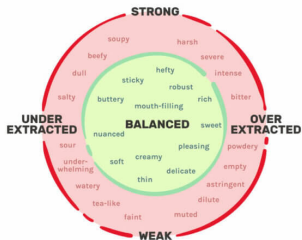
Definition (Renyi)

Mathematician is a device that turns coffee into theorems.

Hence, we need a good coffee.
(Computer Scientists too).

Recipe for good coffee

COFFEE COMPASS



1. Use the map to find where your brew sits. This is where to place the center of the compass.
2. Use the compass to help you move towards the green zone.

e.g. To correct an underwhelming and watery brew, you'll need to extract more.

To correct a salty and sour brew, you'll need to extract more using less coffee.

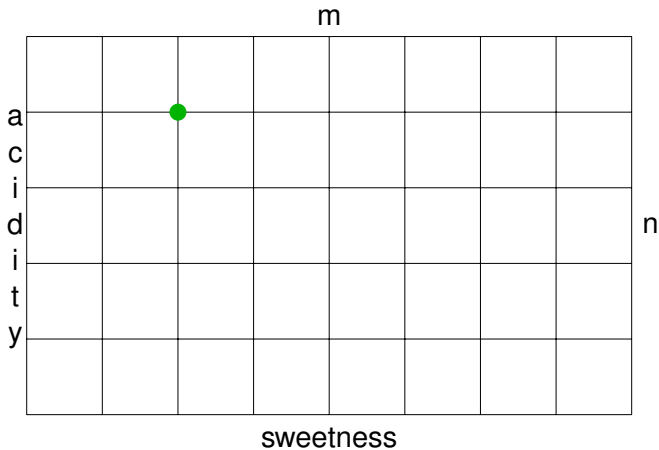
Extract More
Finer grind and/or longer brew time

Extract Less
Coarser grind and/or shorter brew time

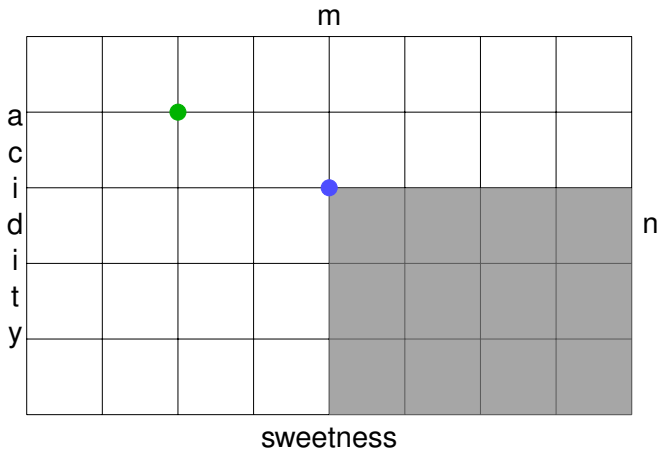
Less Coffee
Reduce coffee or increase water

More Coffee
Increase coffee or reduce water

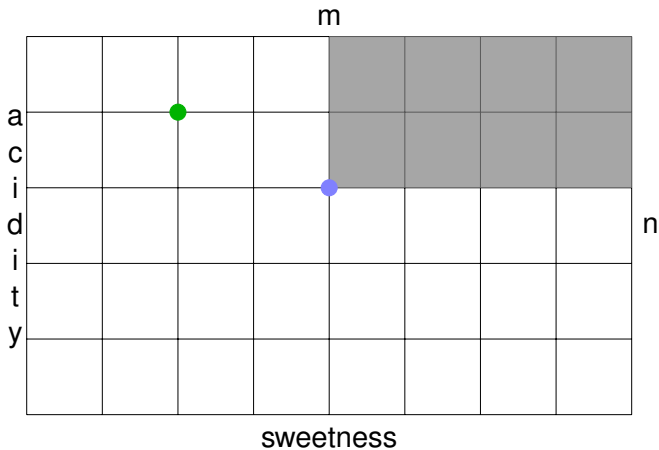
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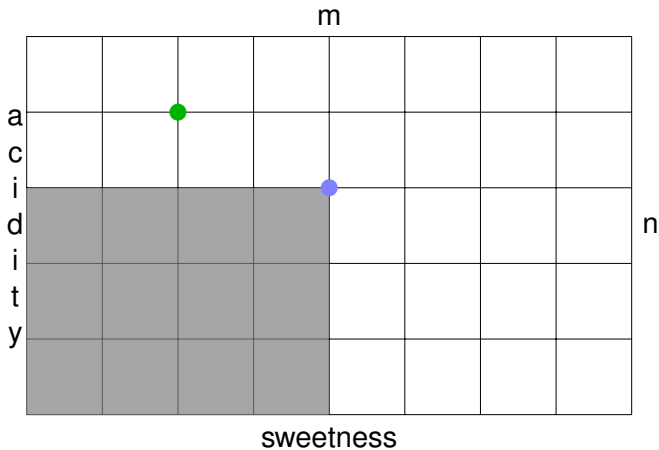
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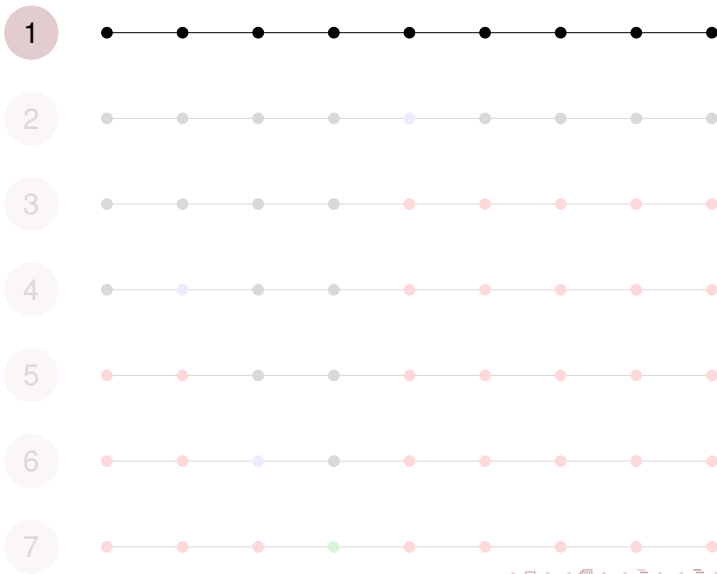
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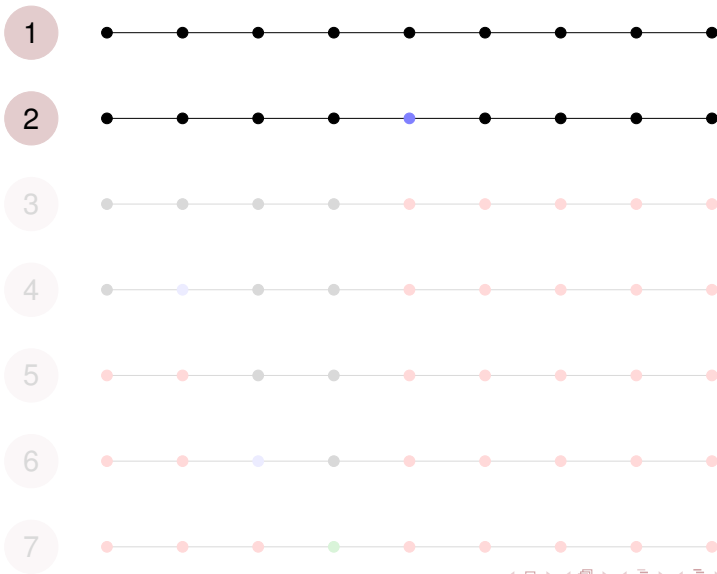
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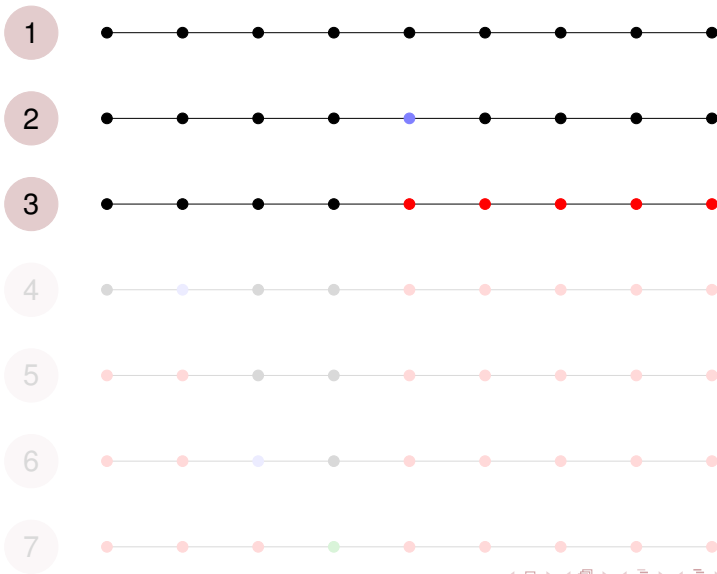
One dimensional solution — a binary search



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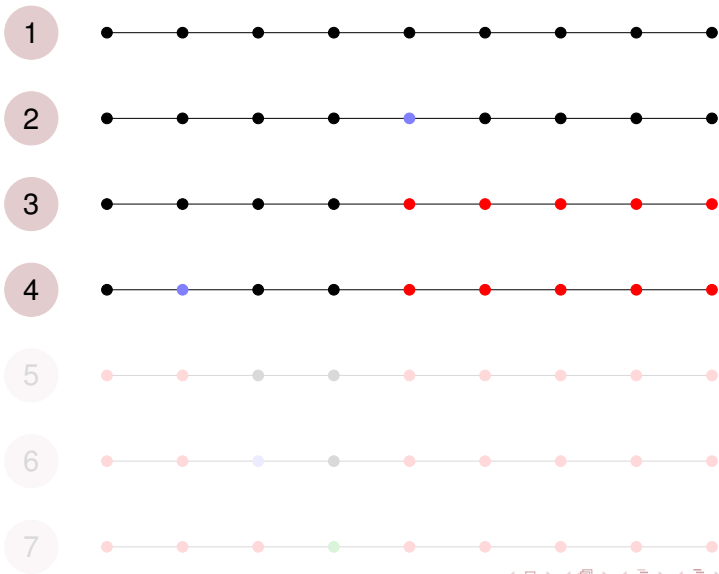


One dimensional solution — a binary search



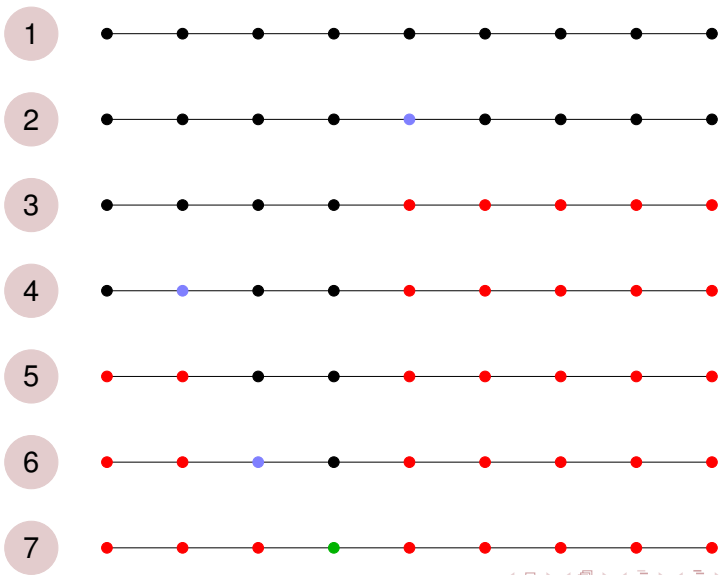


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Outline



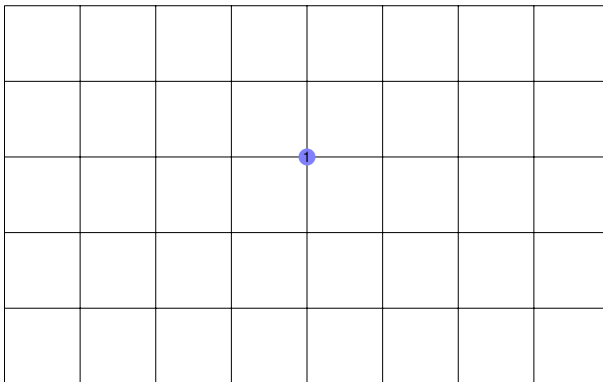
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Multidimensional binary search as a game

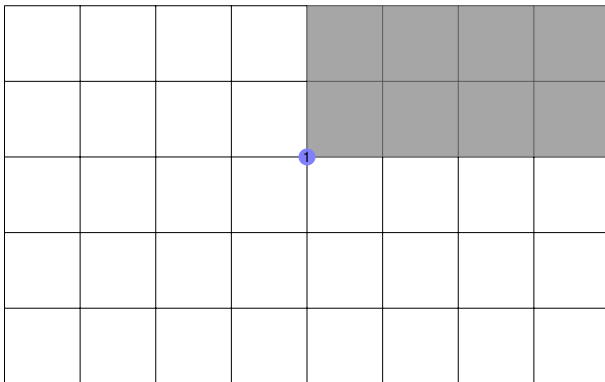
- 1 A game between Algorithm and Adversary
- 2 A board: d -dimensional grid $n_1 \times n_2 \times \dots \times n_d$
- 3 Adversary hides a target in some point of the grid
- 4 At each step Algorithm picks a point (x_1, x_2, \dots, x_d)
- 5 Adversary answers with (d -dimensional) interval with one end at (x_1, x_2, \dots, x_d) and another at some corner, **where there is no target.**
- 6 The Algorithm' goal is to find the target as soon as possible
- 7 Adversary wants to play long
- 8 It is good to think that there is no target point, but the area of possible target places shrinks at each step
- 9 Let $T(n_1, n_2 \dots n_d)$ be the number of steps provided both player play optimally. Surely $T(n) = \Theta(\log n)$.

2D example

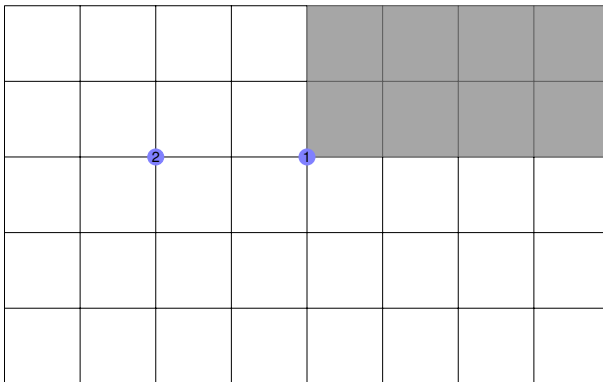




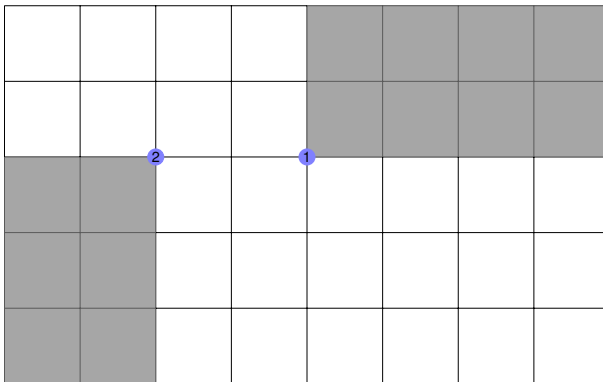
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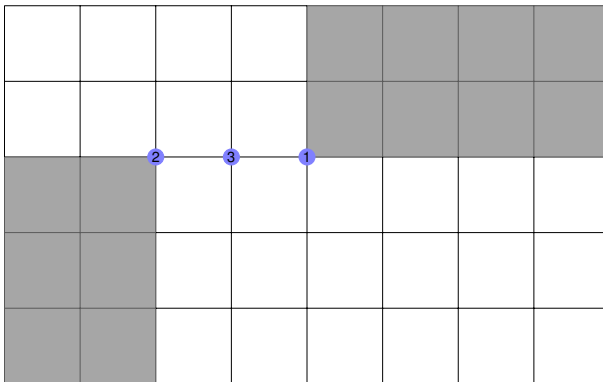


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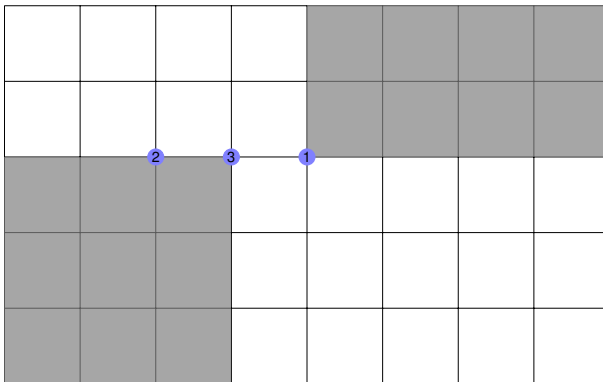


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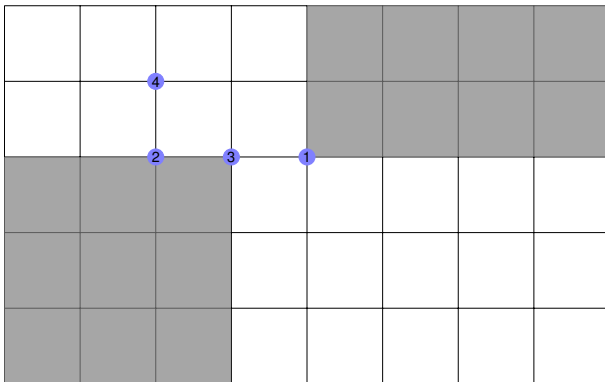


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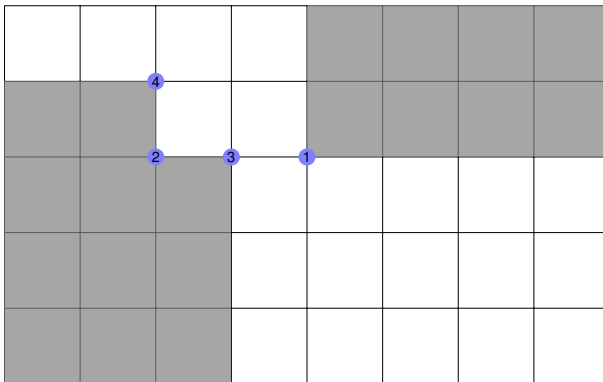




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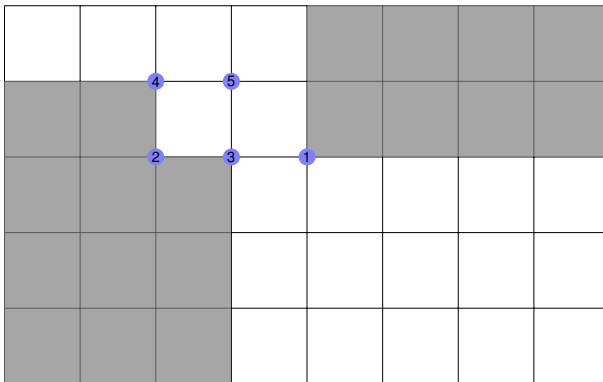


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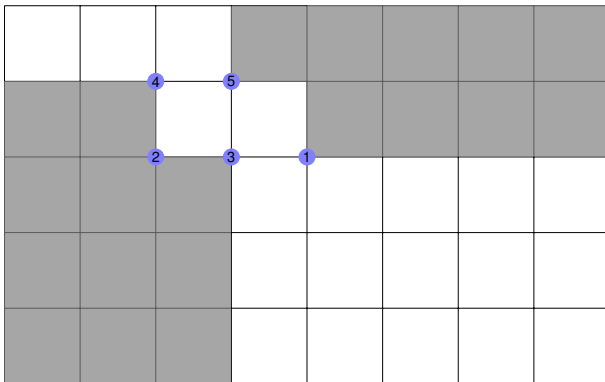




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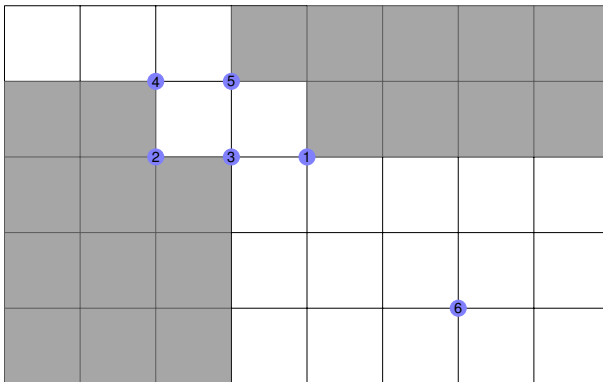


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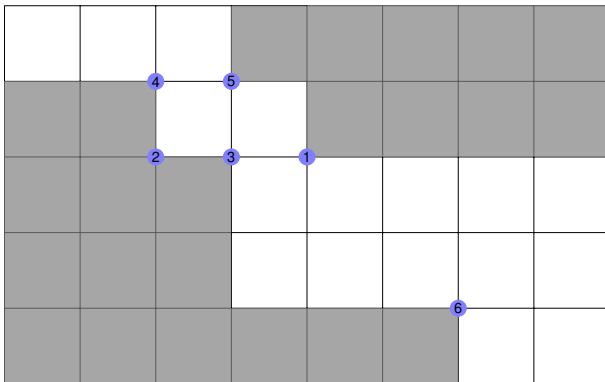


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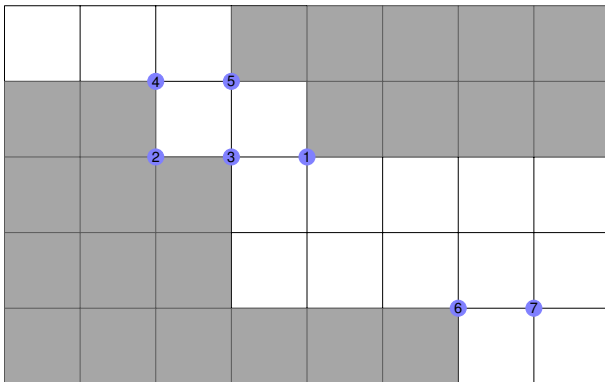


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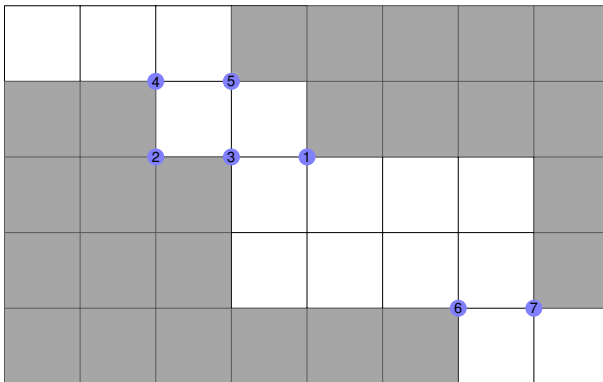


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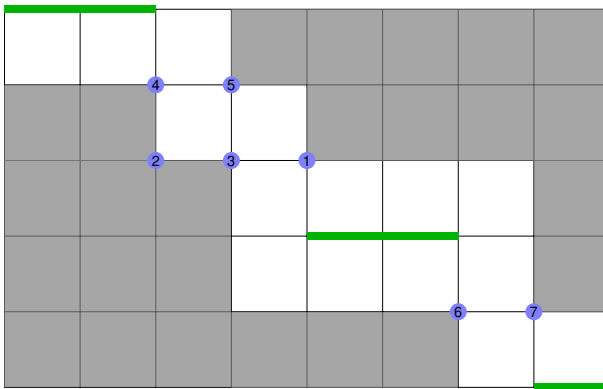




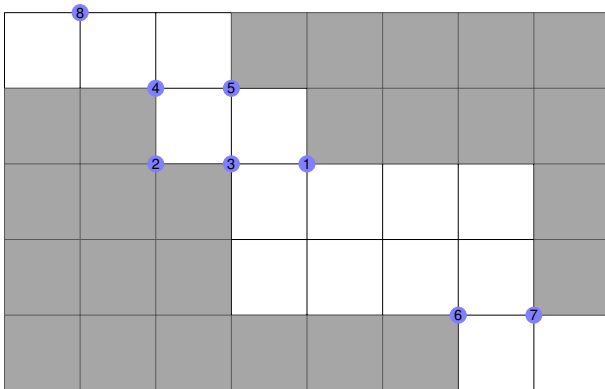
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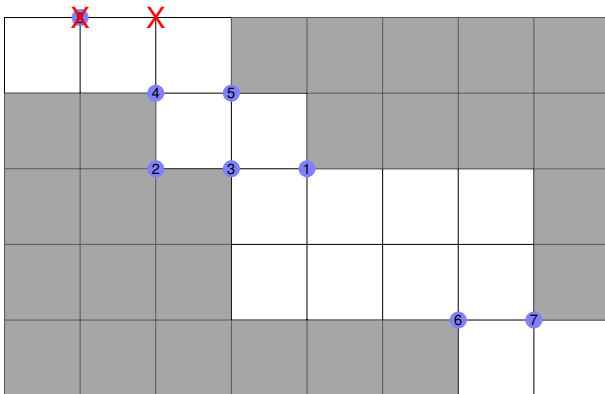


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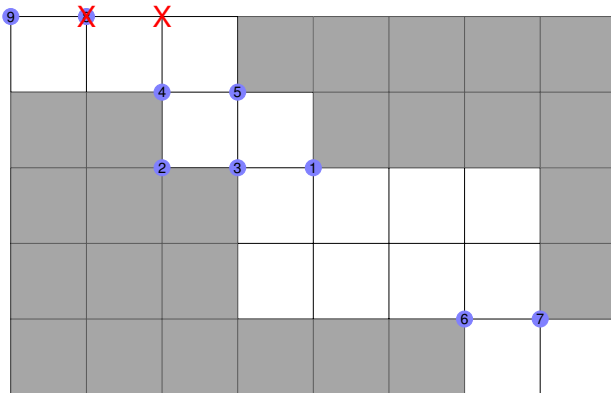


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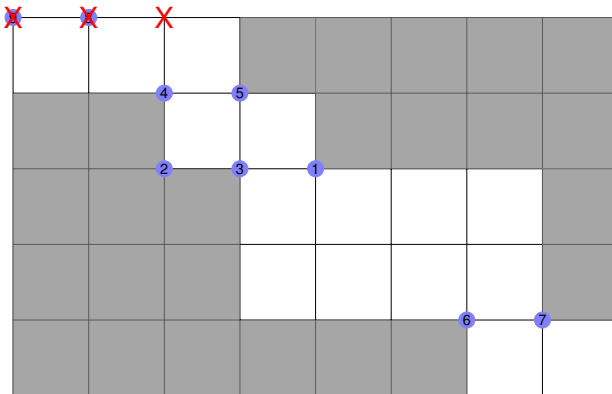




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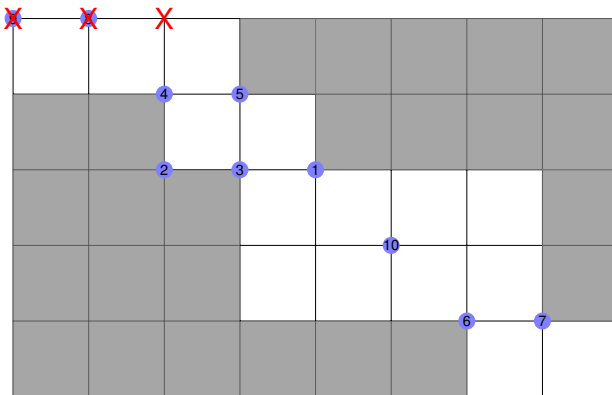


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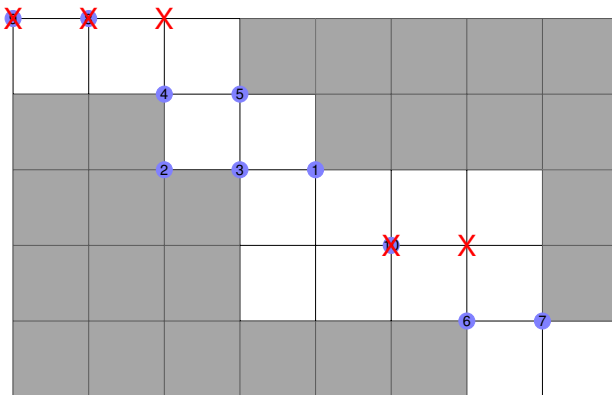


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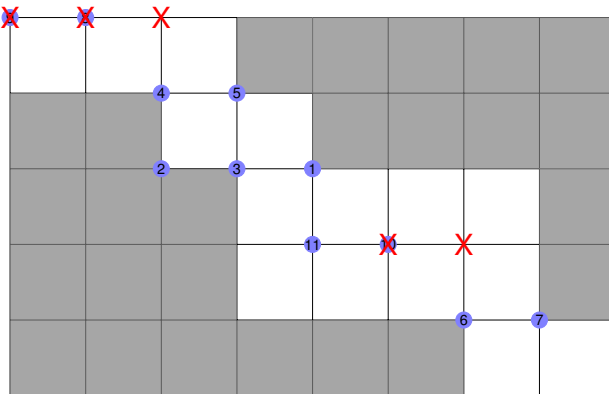




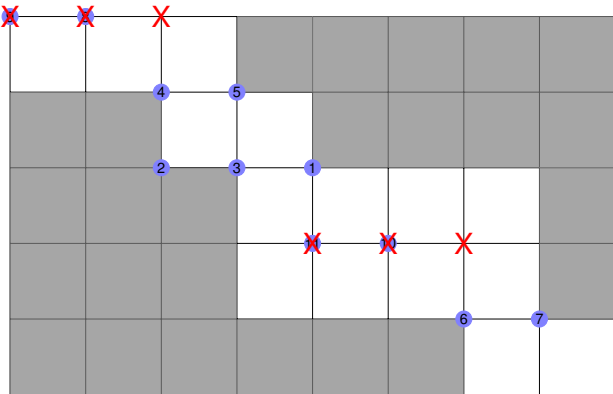
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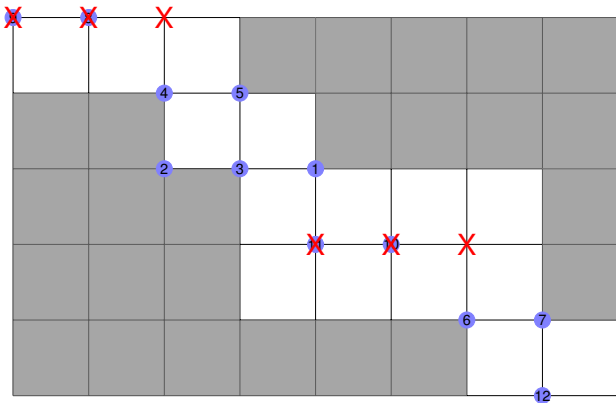
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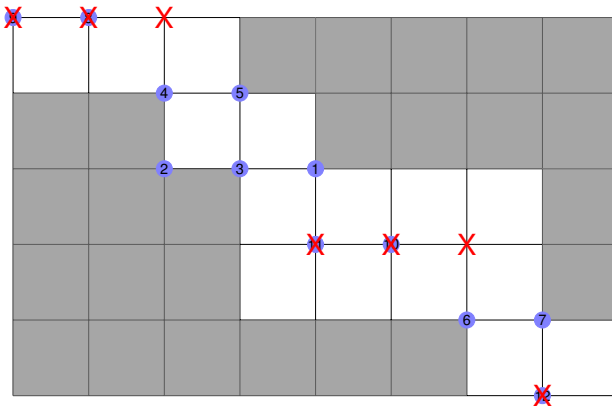


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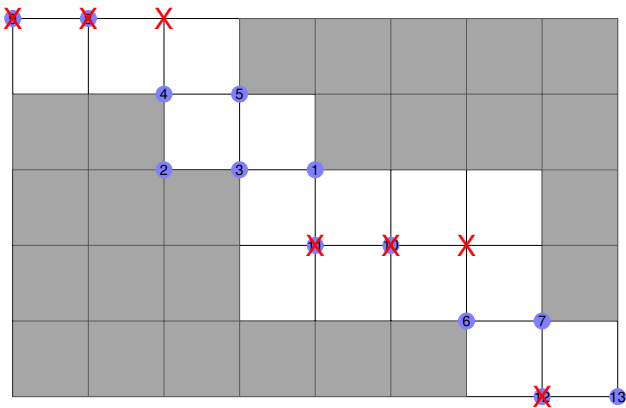


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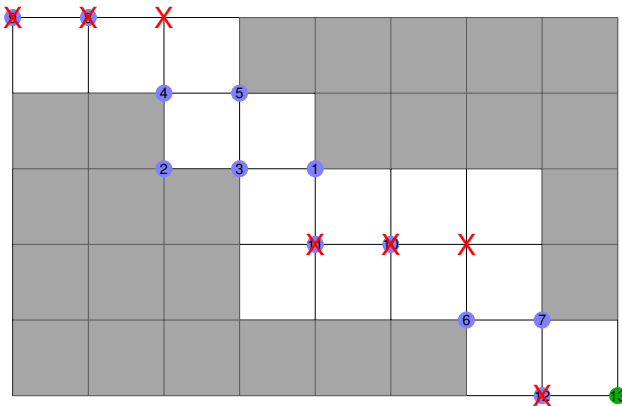


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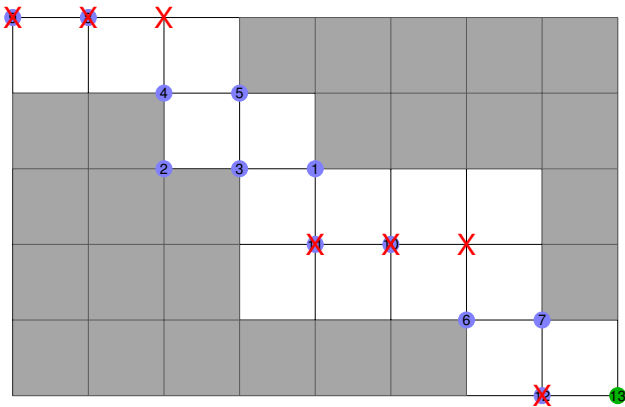


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2D example



Theorem (Dereniowski, G., Wróbel)

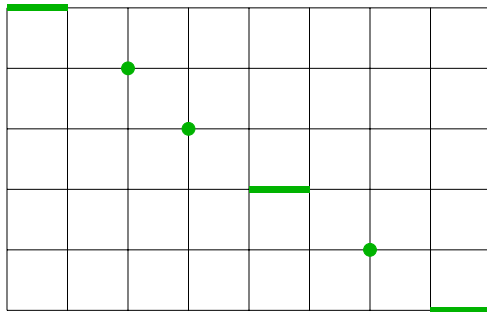
Given $n \leq m$, $T(n, m) = \Theta(n \log(\frac{m}{n} + 1))$.





Informal proof (2D) $T(n, m) = \Theta(n \log(\frac{m}{n} + 1))$

From below $T(n, m) = \Omega(n \log(\frac{m}{n} + 1))$:

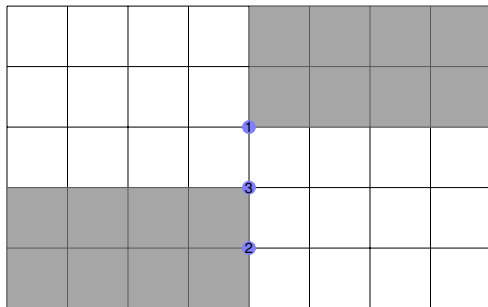


- n (independent) segments with roughly $\frac{m}{n}$ elements,
- for each perform binary search with $\log_2 \frac{m}{n} + 1$ queries,



Informal proof (2D) $T(n, m) = \Theta(n \log(\frac{m}{n} + 1))$

From above $T(n, m) = \mathcal{O}(n \log(\frac{m}{n} + 1))$:

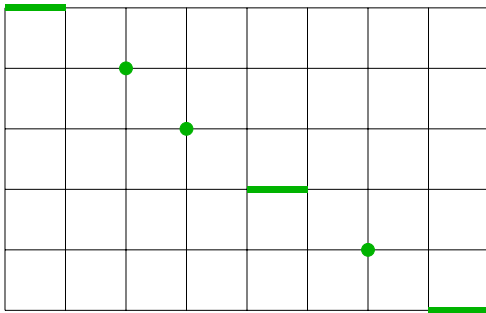


- $\log n$ queries to split the board into 2 (independent) pieces,
- for Adversary it is optimal to have them of the same size,
- hence $T(n, m) \leq \log n + 2T(\frac{n}{2}, \frac{m}{2})$
- after $\mathcal{O}(n)$ steps we are left with diagonal from the lower bound



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d -dimensional case

Theorem (Dereniowski, G., Wróbel)

Given d and n_1, \dots, n_d , such that $n_i \leq n_{i+1}$ there is

$$T(n_1, \dots, n_d) = \mathcal{O} \left(\left(\prod_{i=1}^{d-1} n_i \right) \log \left(\frac{n_d}{n_{d-1}} + 1 \right) \right).$$

For proof: Algorithm consider n_1 $(d-1)$ -dimensional cubes.

Theorem (Dereniowski, G., Wróbel)

For fixed d there is $T(n, \dots, n) \geq \frac{2}{d-1} n^{d-1}$,
i.e $T(n, \dots, n) = \Omega \left(\frac{n^{d-1}}{d} \right)$.

For proof: Adversary hides target at the hyperplane given by

$$x_d = n - \left\lfloor \frac{x_1 + \dots + x_{d-1}}{d-1} \right\rfloor.$$



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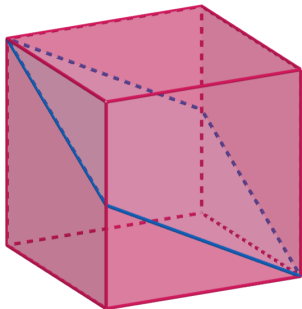
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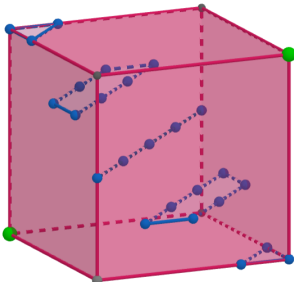


d -dimensional case

Theorem (Dereniowski, G., Wróbel)

Given d and n_1

$$T(n_1, \dots, n_{d-1}, n_d)$$



there is

$$\left(\frac{n_d}{n_{d-1}} + 1 \right)^{n_{d-1}}$$

For proof: Algorithm

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d -dimensional case — further work



Theorem (Dereniowski, G., Wróbel)

For fixed d there is $T(n, \dots, n) = \Omega\left(\frac{n^{d-1}}{d}\right)$ and
 $T(n, \dots, n) = \mathcal{O}(n^{d-1})$.

To do:

- Fill the gap.
- Lower bound when $(n_1, \dots, n_d) \neq (n, \dots, n)$ (we did it for $d \leq 3$).

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Why graphs? Graphical embeddings of data



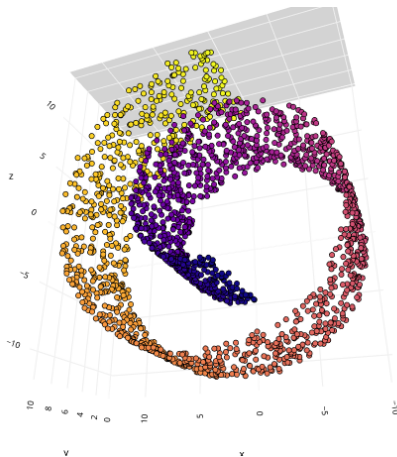


Why graphs? Graphical embeddings of data





Why graphs? Graphical embeddings of data





Why graphs? Graphical embeddings of data



Edge and pair queries (Dereniowski, G., Prałat [2023])



- 1 A game between Algorithm and Adversary
- 2 A board: graph G
- 3 Adversary hides a target at some vertex $x \in V(G)$
- 4 At each step Algorithm picks two (**adjacent**) vertices u, v
- 5 Adversary answers which vertex is closer to x breaking ties arbitrary
- 6 The Algorithm' goal is to find the target as soon as possible
- 7 Adversary wants to play long
- 8 It is good to think that there is no target vertex, but the area of possible target places shrinks at each step
- 9 Let $pq(G)$ (**$eq(G)$**) be the number of rounds provided both player play optimally. Surely $pq(G) \leq eq(G)$.



Edge and pair queries — simple properties

Observation

For any connected graph G on n vertices,

$$\log_2 n \leq \text{pq}(G) \leq \text{eq}(G) \leq n - 1.$$

In fact, there exists a strategy of the algorithm that in each round eliminates at least one vertex from the search space.

Lower bound is achieved by P_n (binary search), upper — by K_n and $K_{1,n-1}$.





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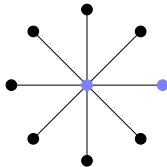
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Observation

For any $n \in \mathbb{N} \setminus \{1\}$, there exists a graph G on $\Theta(n)$ vertices such that $\text{eq}(G) = \Omega(n)$ and $\text{pq}(G) = \mathcal{O}(\log n)$.



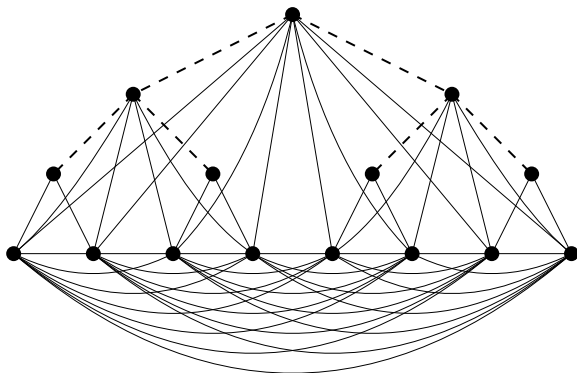
Edge and pair queries — simple properties

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In fact, the round elimination argument shows that

Lower bounds for edge and pair queries are $\Omega(n)$ and $\mathcal{O}(\log n)$, respectively.



such
edge.

— by K_n

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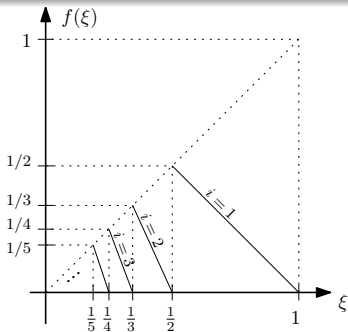


Edge and pair queries — results

Theorem (Dereniowski, G., Prałat [2023])

Suppose that $pn = n^{\xi+o(1)}$, where $\xi \in (\frac{1}{i+1}, \frac{1}{i})$ for some $i \in \mathbb{N}$.
Then, a.a.s. for $G \in \mathcal{G}(n, p)$.

$$pq(G) = \Theta(\text{eq}(G)) = \Theta\left(n^{f(\xi)+o(1)}\right), \text{ for } f$$





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Theorem (Dereniowski, G., Prałat [2023])

For fixed $p \in (0, 1)$ a.a.s. for $G \in \mathcal{G}(n, p)$.

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Theorem (Dereniowski, G., Prałat [2023])

Calculating $pq(G)$ or $\text{eq}(G)$ is \mathcal{NP} -hard, even for graphs of diameter at most 3.

Edge and pair queries — questions



Question

What happens for particular graph classes?
(eg. [grids](#), intersection graphs)



Edge and pair queries — questions

Question

What happen for particular graph classes?
(eg. [grids](#), intersection graphs)

Thank you for the attention!

Source of pictures:
www.homegrounds.co (Coffee Compass)
some commercials (cold drink samples)
towardsdatascience.com (graph embeddings)