

Finding monotone patterns

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ETH-ITS

joint with Omri Ben-Eliezer, Clément Canonne, Erik Waingarten

Mittagsseminar

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Property testing

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Aim: design fast (randomised) algorithms that determine, with probability at least 0.99, if a given (large) object

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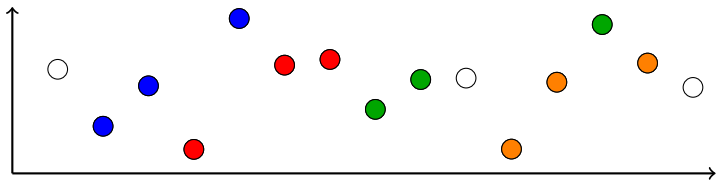
We consider testing with **one-sided error**: if an object is far from having \mathcal{P} , provide evidence.

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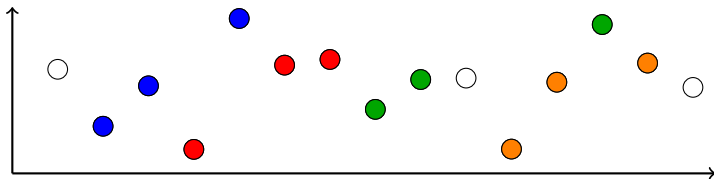
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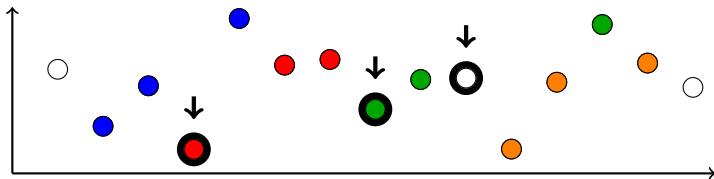
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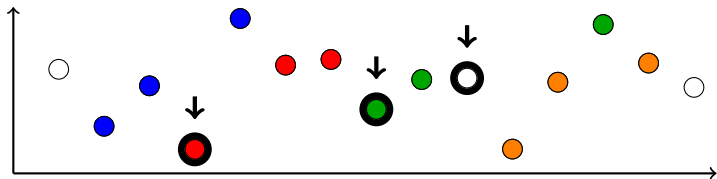
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We sometimes refer to an increasing k -tuple as a $(1\dots k)$ -copy.

History

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Newman, Rabinovich, Rajendraprasad, Sohler '17. For $k \geq 2$, there is a (**non-adaptive**) tester which makes $(\log n)^{O(k^2)}$ queries.

Our results

Theorem (Ben-Eliezer, Canonne, L., Waingarten)

An optimal *non-adaptive* algorithm for testing $(1\dots k)$ -freeness makes $\Theta_k\left((\log n)^{\lfloor \log_2 k \rfloor}\right)$ queries.

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Towards a lower bound for $k = 2$: binary profiles

For $x, y \in \mathbb{N}$: $\mathbf{b}(x, y) = \max \left\{ i : \begin{array}{l} \text{the binary representations} \\ \text{of } x \text{ and } y \text{ differ in } i^{\text{th}} \text{ digit} \end{array} \right\}$.

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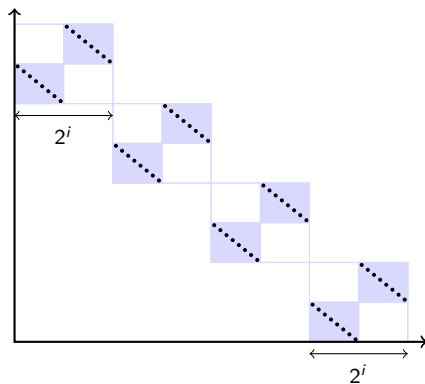
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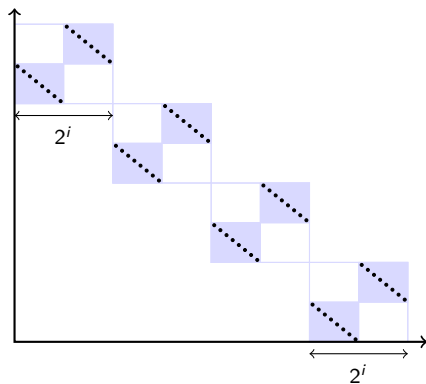
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Observation

If (x, y) is an increasing pair in f_i , then $b(x, y) = i$.

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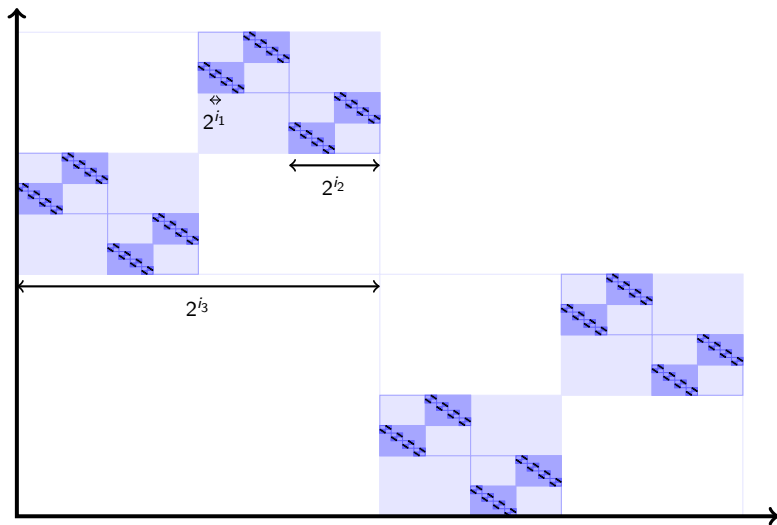
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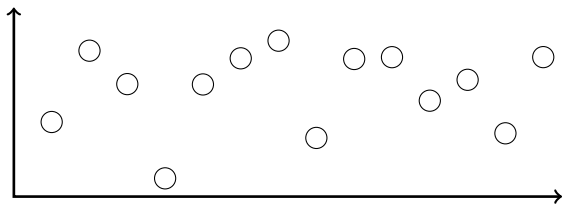
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Thus, to find an increasing pair with probability at least 0.99 need at least $0.99 \log n$ queries.

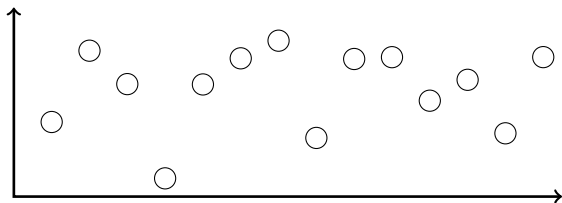
Lower bound for $k = 2^k$: iterated construction



Choosing copies greedily

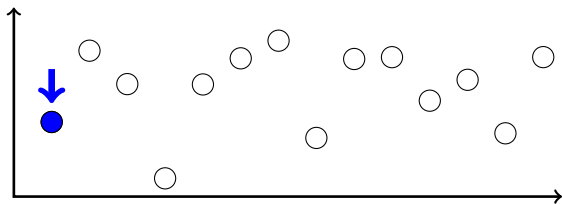


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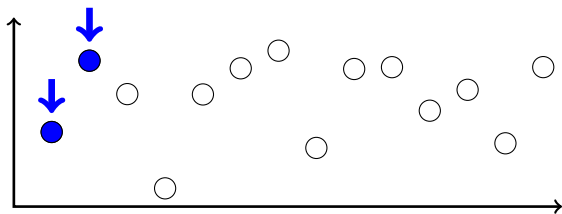
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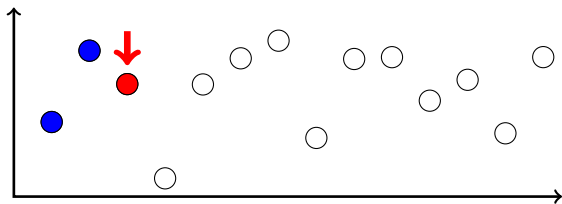
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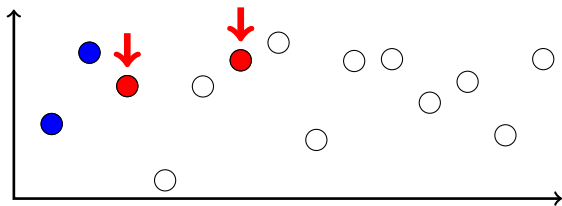
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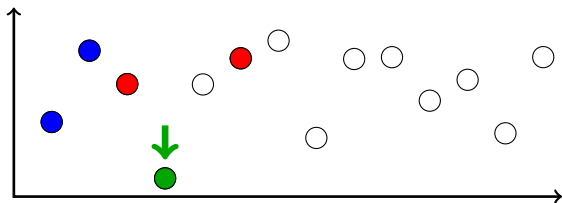
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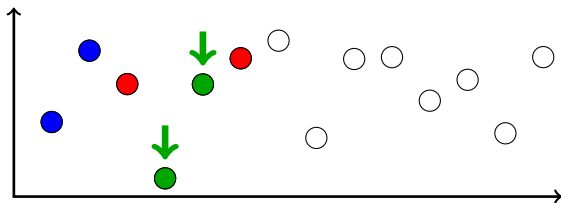
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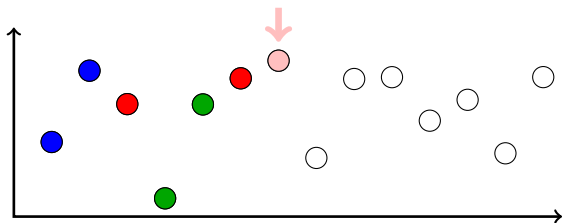
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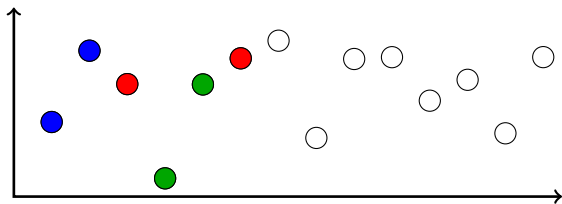
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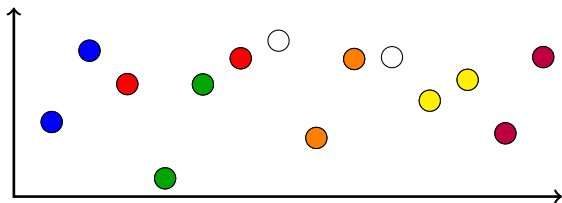
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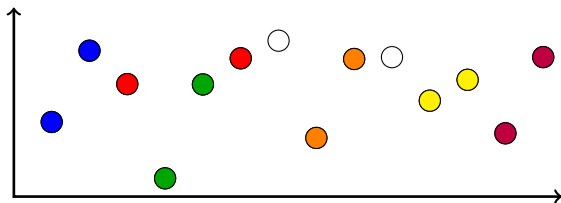
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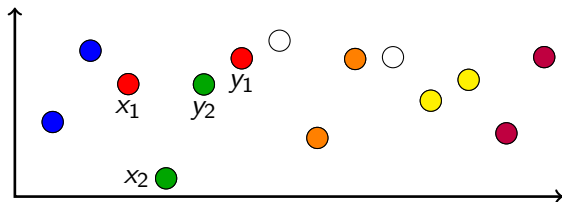


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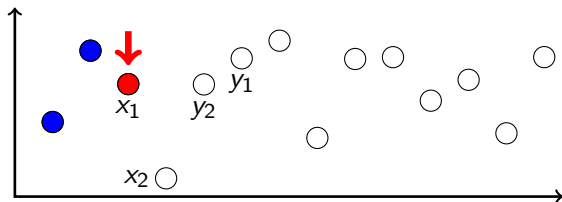


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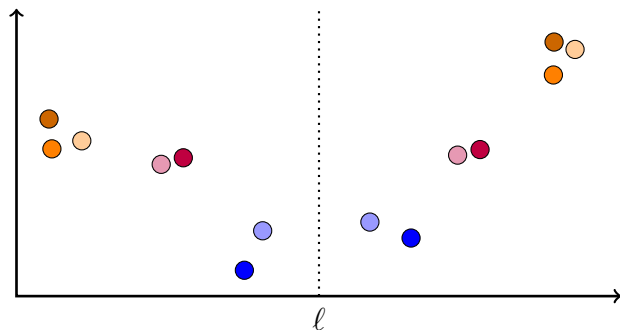
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If $(x_1, y_1), (x_2, y_2) \in \mathcal{F}$, and $x_1 < x_2$, $y_1 > y_2$, then $f(y_1) > f(y_2)$.

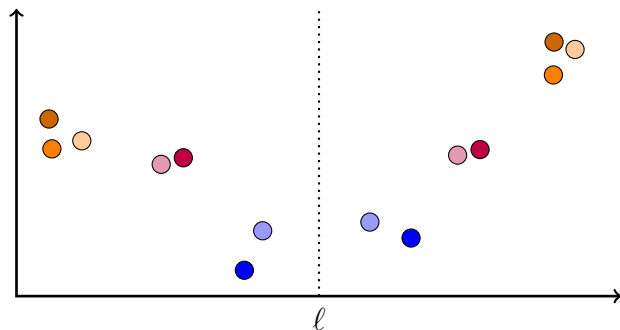
Using the greedy pairing

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Suppose l lies 'roughly in middle' of many pairs of different widths.

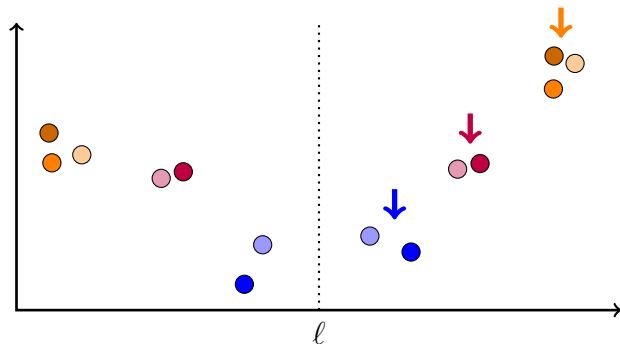
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Suppose ℓ lies 'roughly in middle' of many pairs of different widths.

- If hit entries of k different-widths copies to the right of ℓ , find $(1\dots k)$.

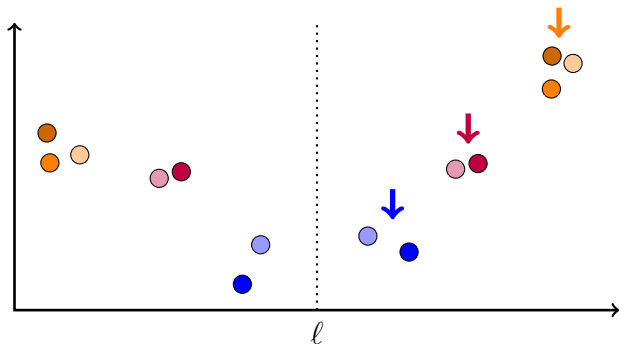
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- If hit entries of k different-widths copies to the right of ℓ , find $(1\dots k)$.
- Can be done with $O(\log n)$ queries: sample $\Theta(1)$ elements from $[\ell, \ell + 2^i]$ for every $i \in [\log n]$.

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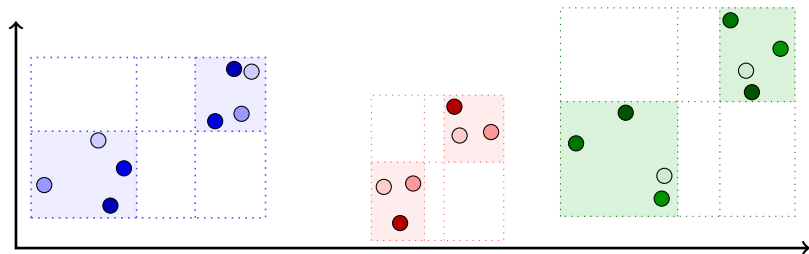
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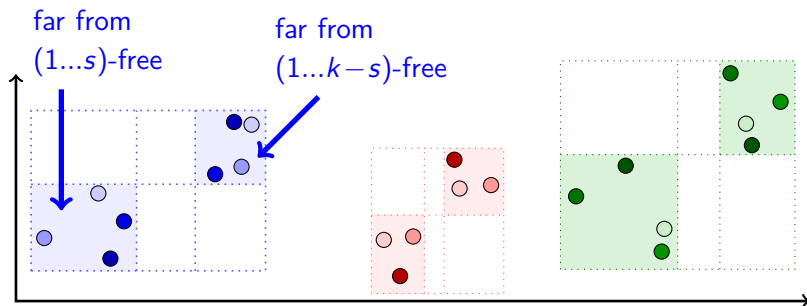
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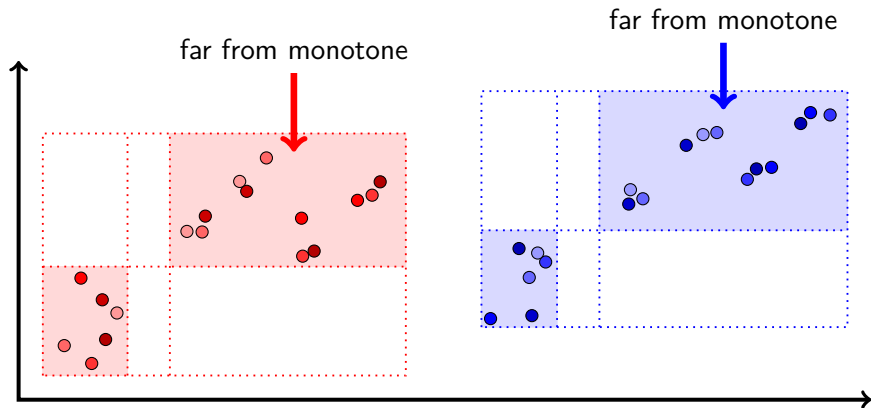
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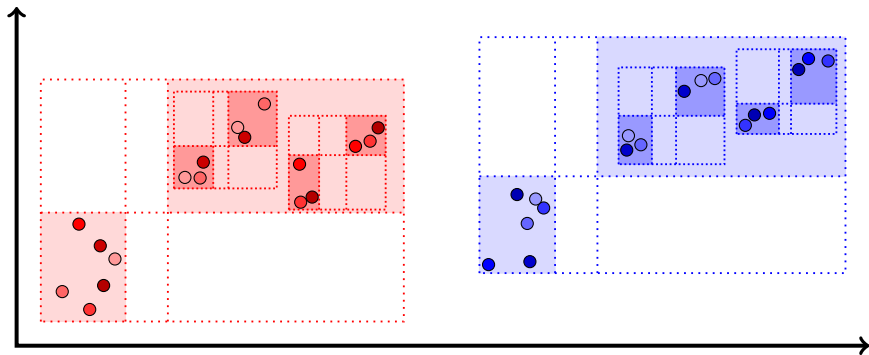


Finding (123) with $O(\log n)$ queries

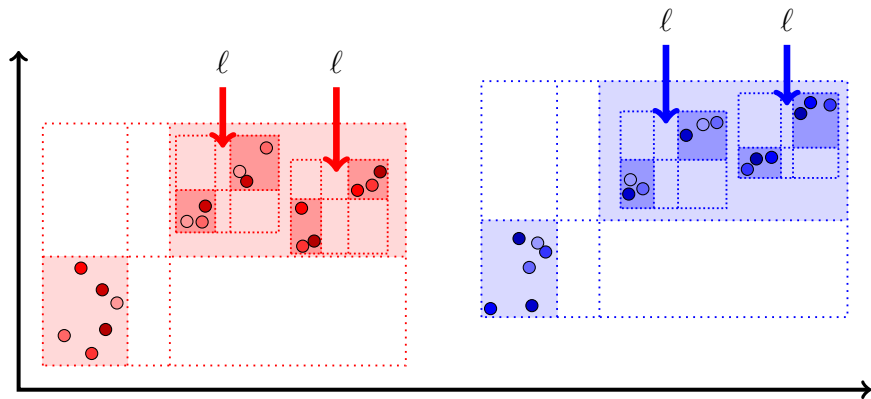
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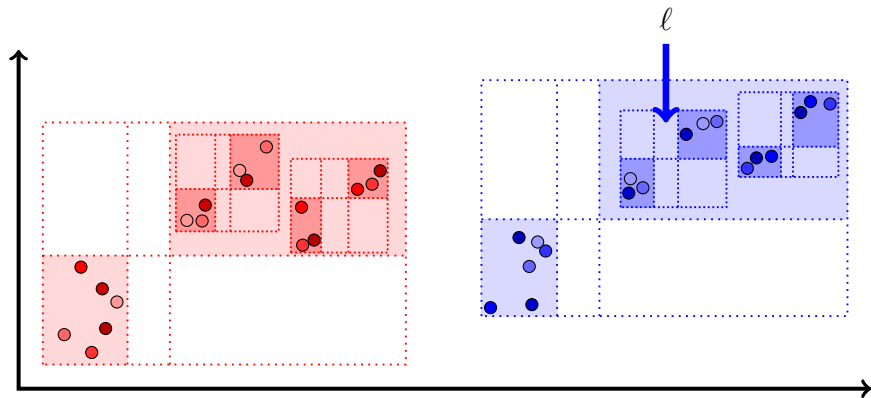


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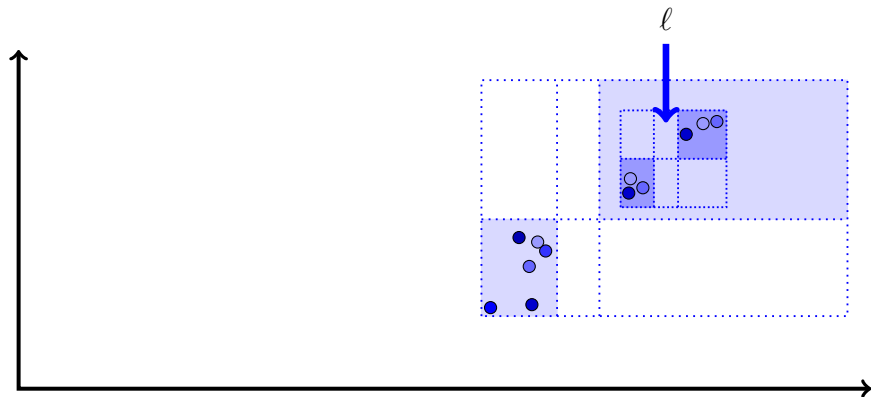
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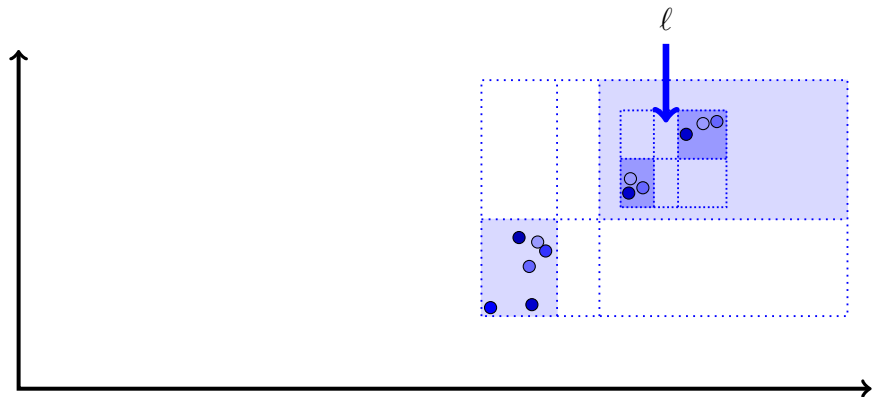
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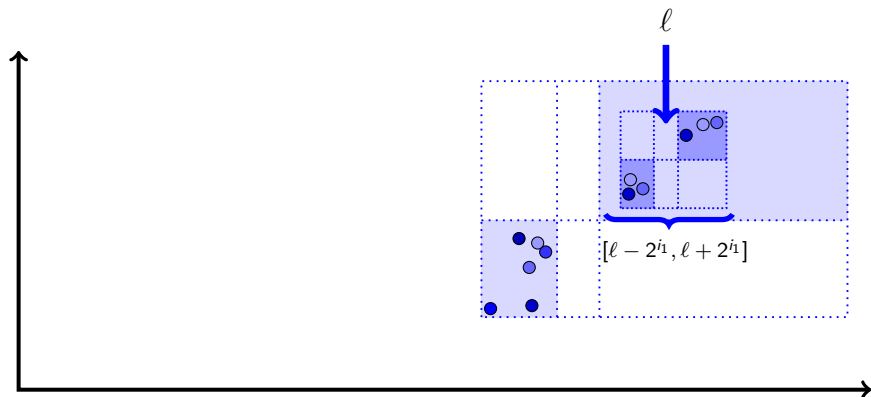
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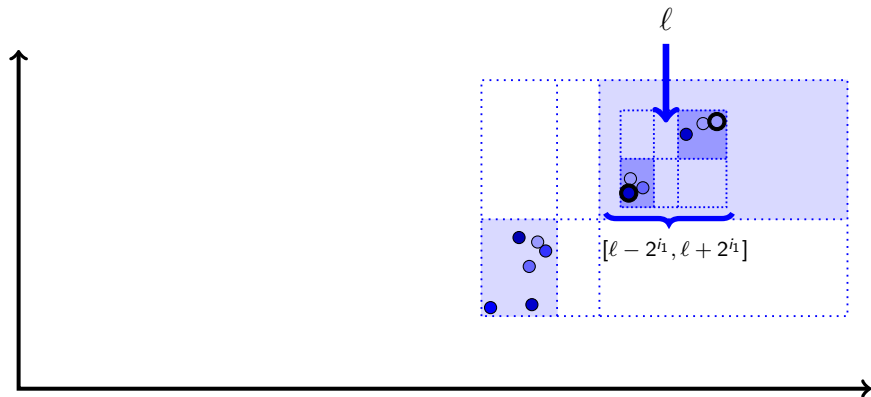
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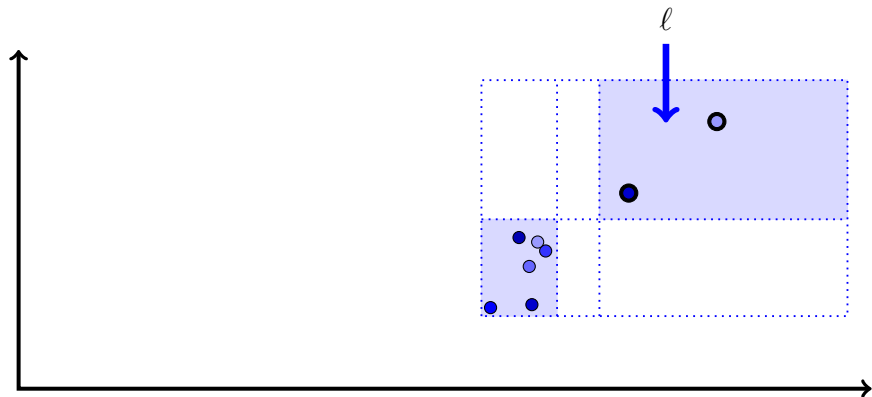
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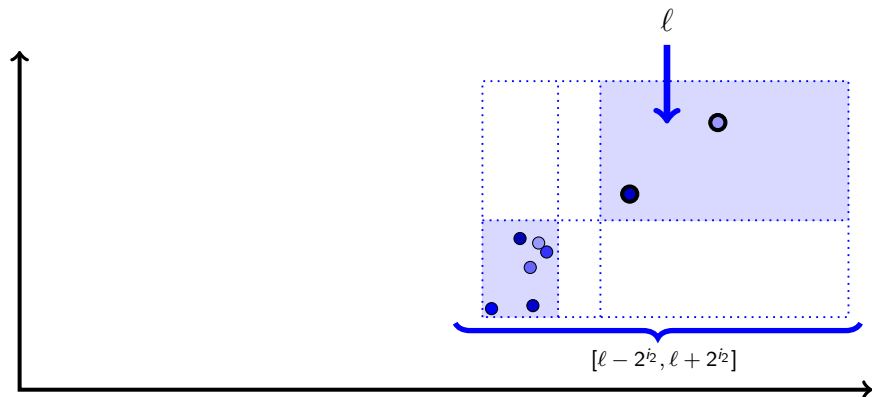
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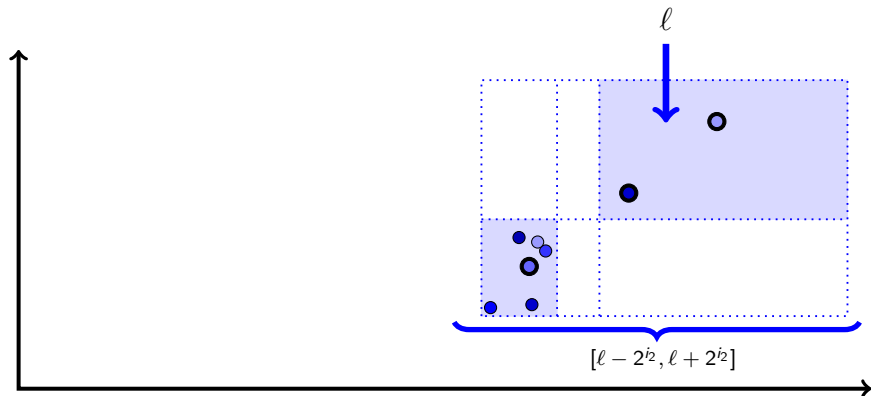
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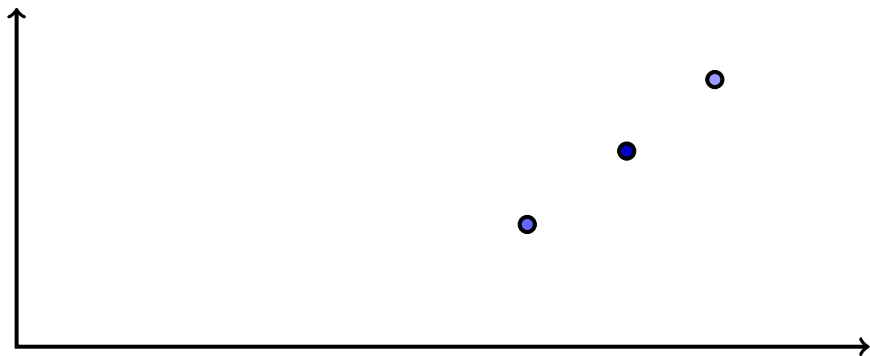
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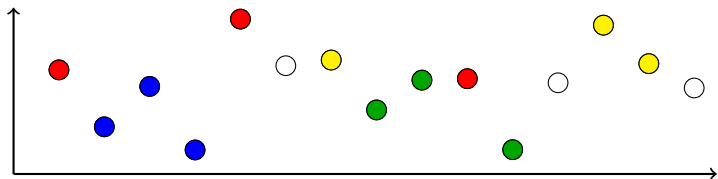
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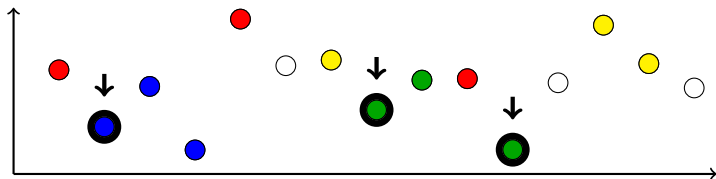
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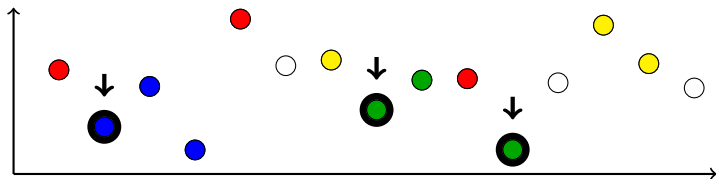
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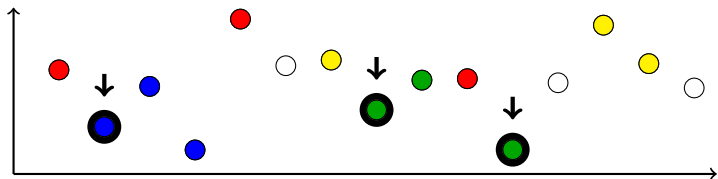
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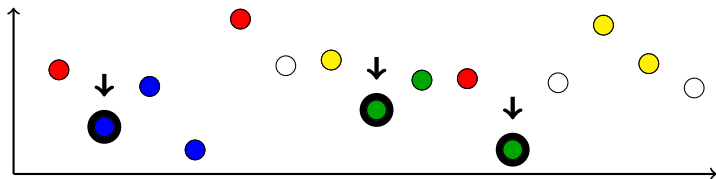


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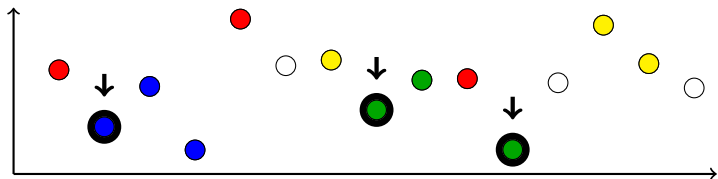
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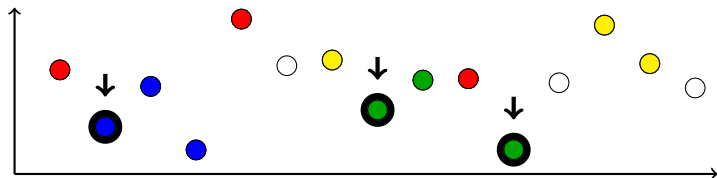
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Thank you!!!