

# Dynamic Maximum Depth of Geometric Objects

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**Based on joint work with**



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UCSB



**Jie Xue**  
NYU Shanghai



**Jiumu Zhu**  
NYU Shanghai

# **Depth Problem**

#objects that contain a point

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## MaxDepth

**Input:** a set of geometric objects  $\mathcal{O}$  (in  $\mathbb{R}^d$ )

**Output:** a point  $p^*$  s.t.

$$\text{dep}(p^*, \mathcal{O}) = \text{dep}(\mathcal{O}) = \max_{p \in \mathbb{R}^d} \text{dep}(p, \mathcal{O})$$

and the value of  $\text{dep}(p^*, \mathcal{O})$

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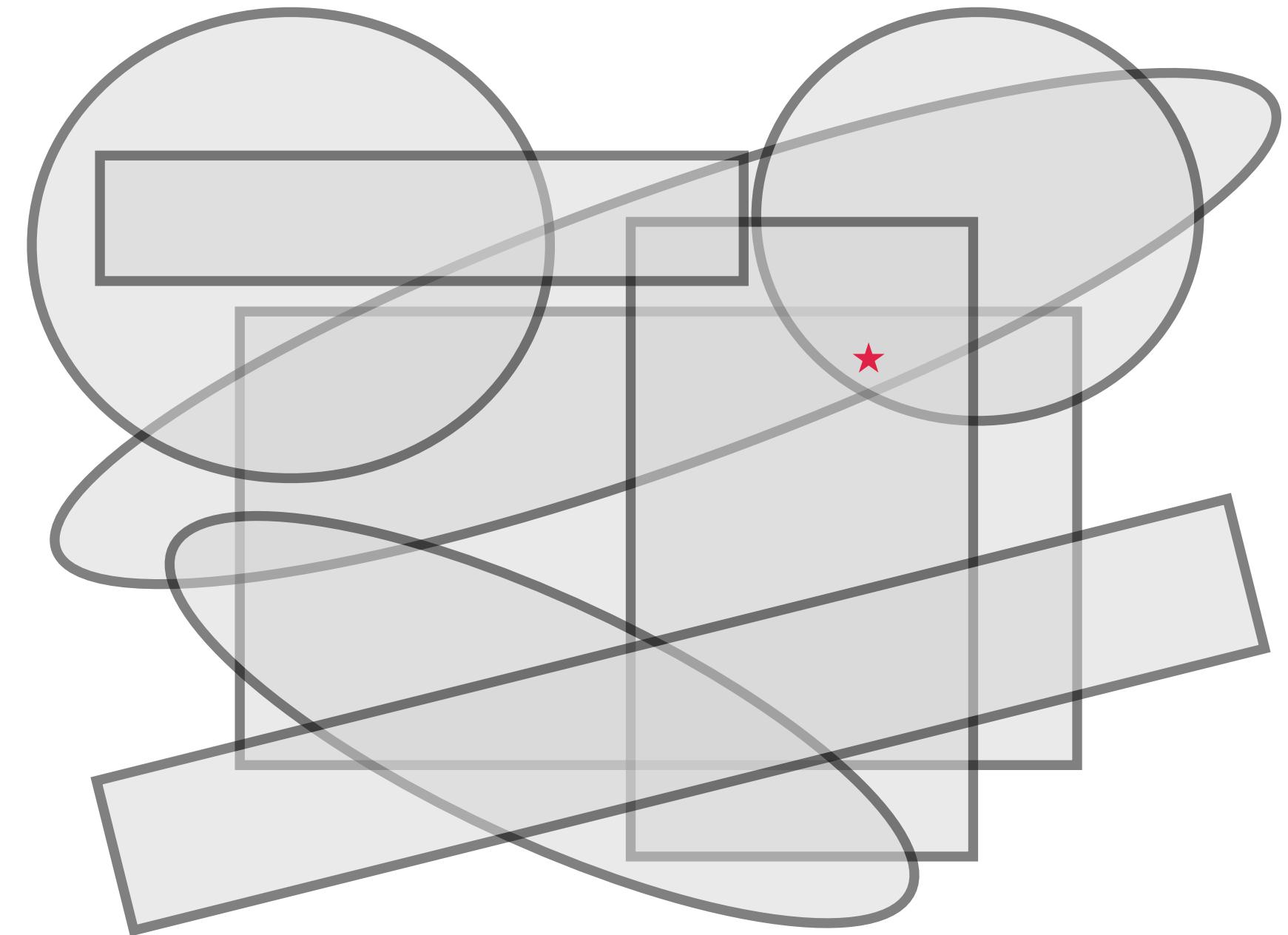
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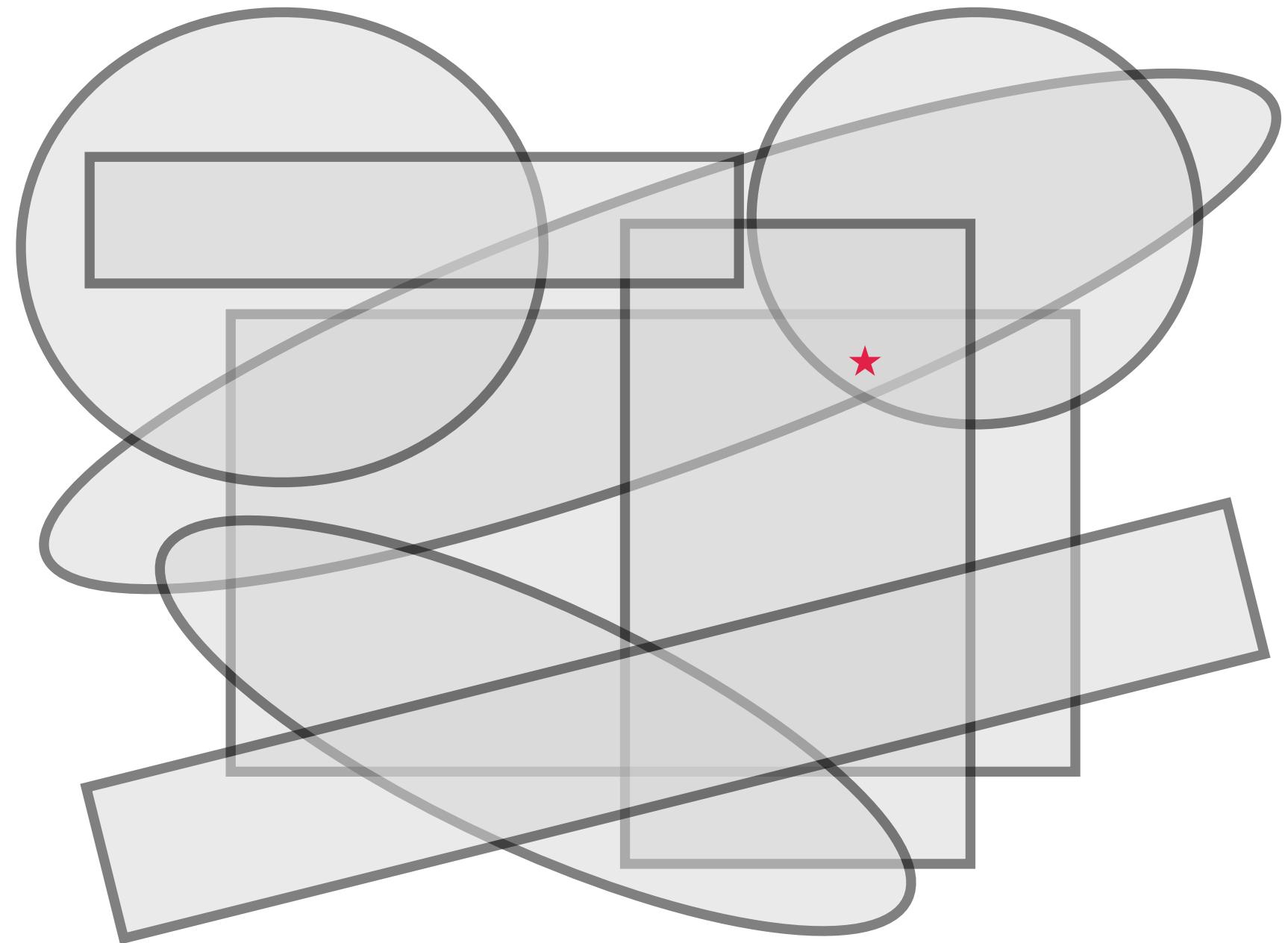
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interval, half-plane, rectangle (box), disk (ball), etc.

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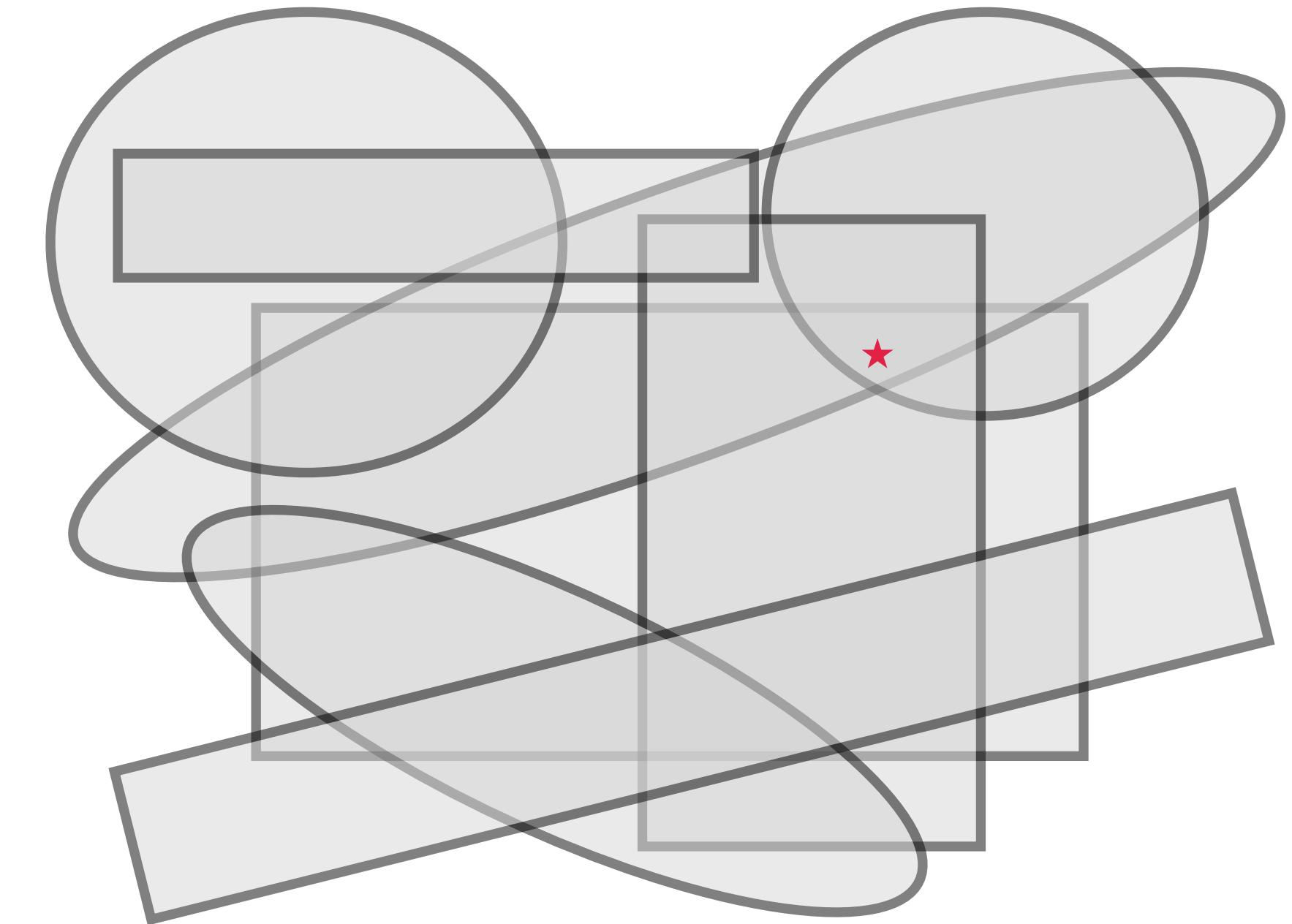
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## Dynamic version

Insert or delete an object is allowed  
Goal: **fast** maintain the maximum depth

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Dynamic + range search, point location, nearest neighbor, bichromatic closest pair,  
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## Dynamic data structure for geometric optimization problem

On geometric intersection graph:

minimum set cover [[Agarwal, Chang, Suri, Xiao and Xue 2020; Chan and He 2021; Chan, He, Suri and Xue 2022](#)]

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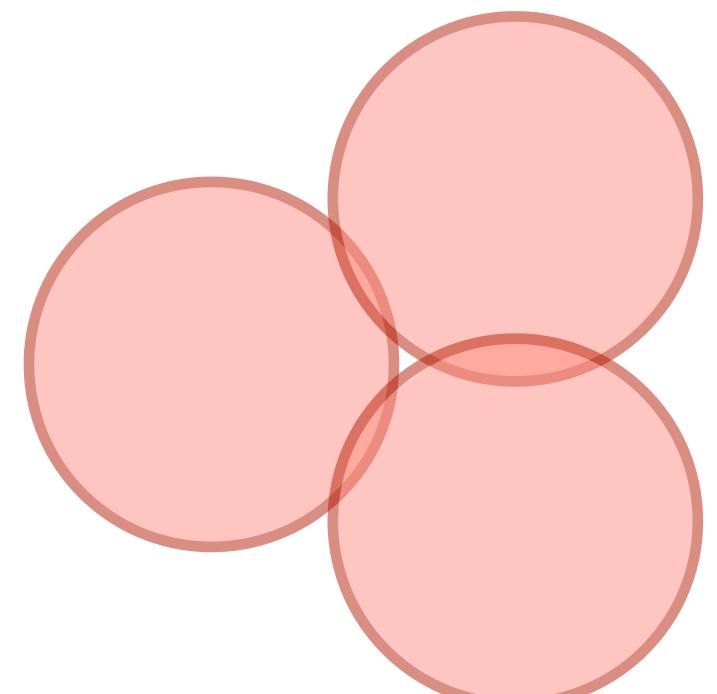
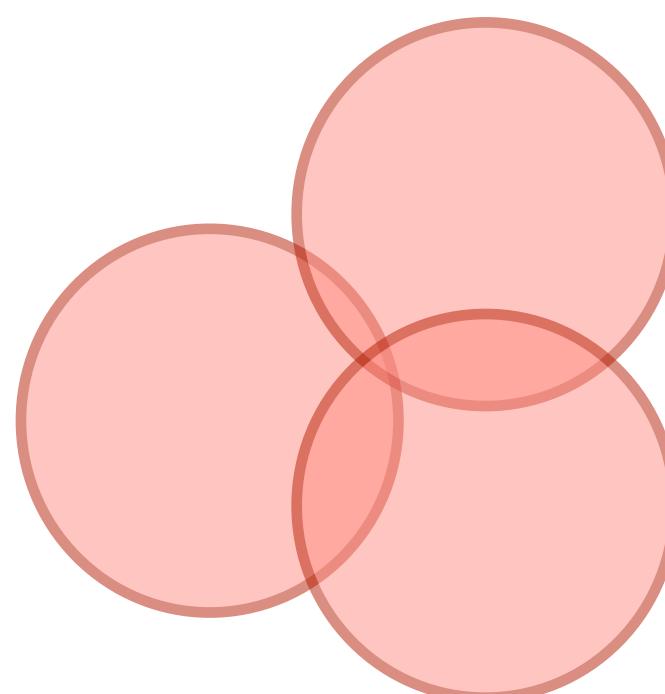
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## Dynamic Maximum Depth

- Static maximum depth is polynomial time solvable
- Maximum depth is a constant approx. of maximum clique



# **Known Results for MaxDep**

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## Static MaxDep Problem

For  $d$ -dimensional (axis-parallel) boxes:

- best known running time:  $O((n^{d/2}/\log^{d/2} n)(\log \log n)^{O(1)})$  [Chan 2013]
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Exact depth with  $\tilde{O}(\sqrt{n})$  update time,  
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Unlikely have sublinear update time for exact depth

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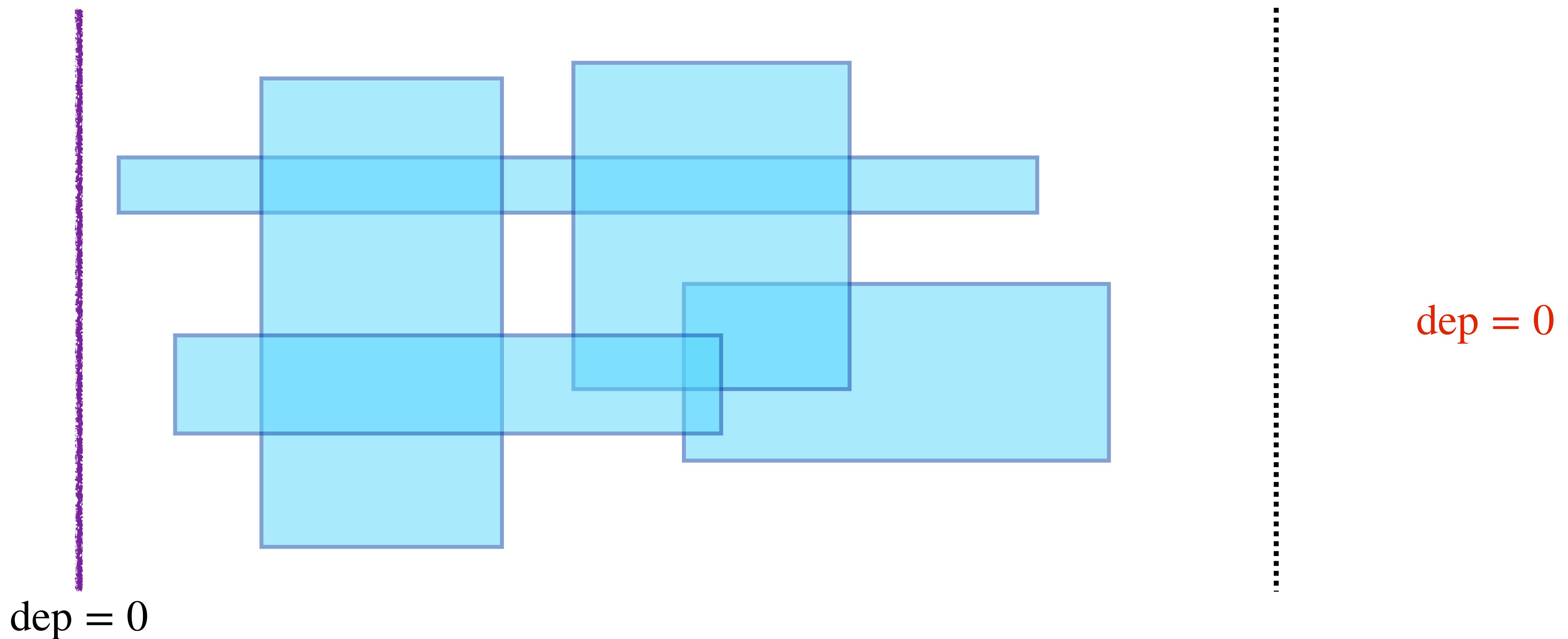
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For rectangles:  $O(n \log n)$ -time algorithm via sweep line [Imai and Asano 1977]

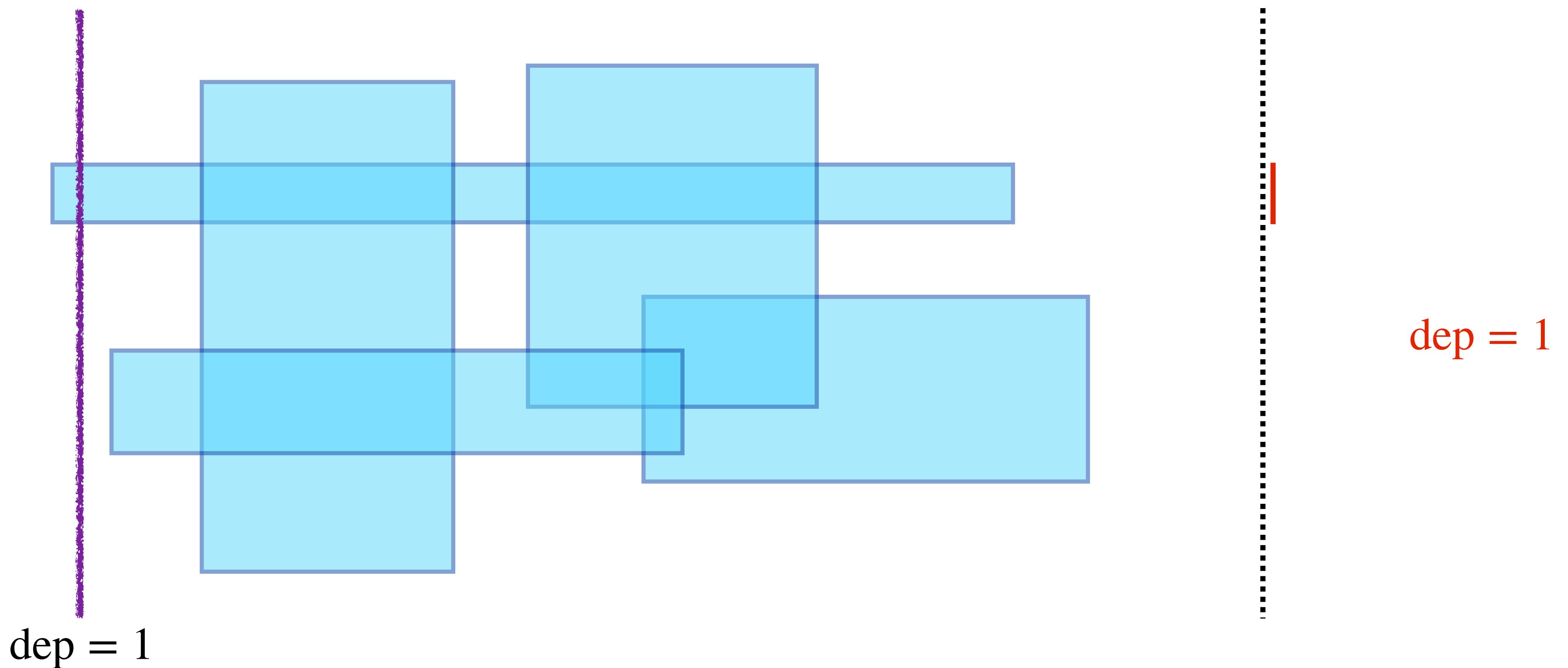
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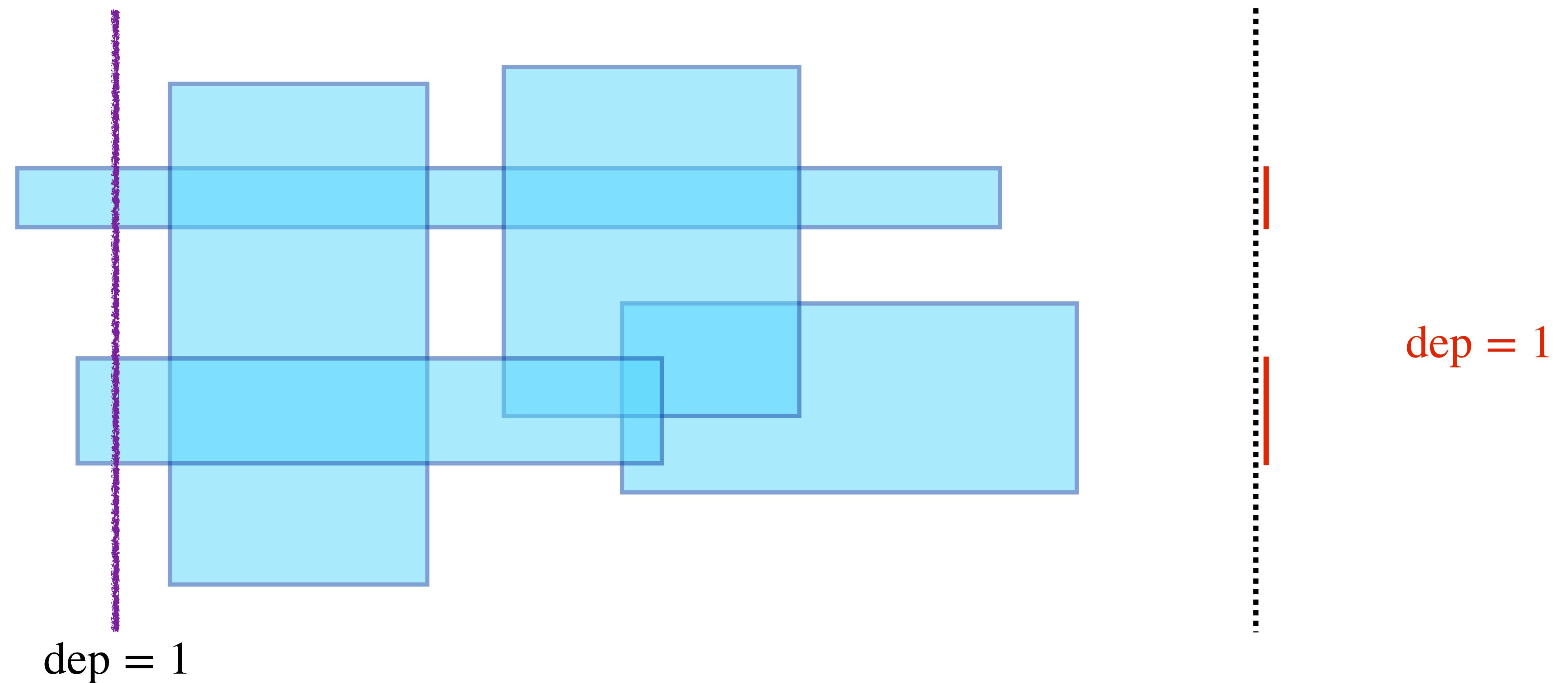
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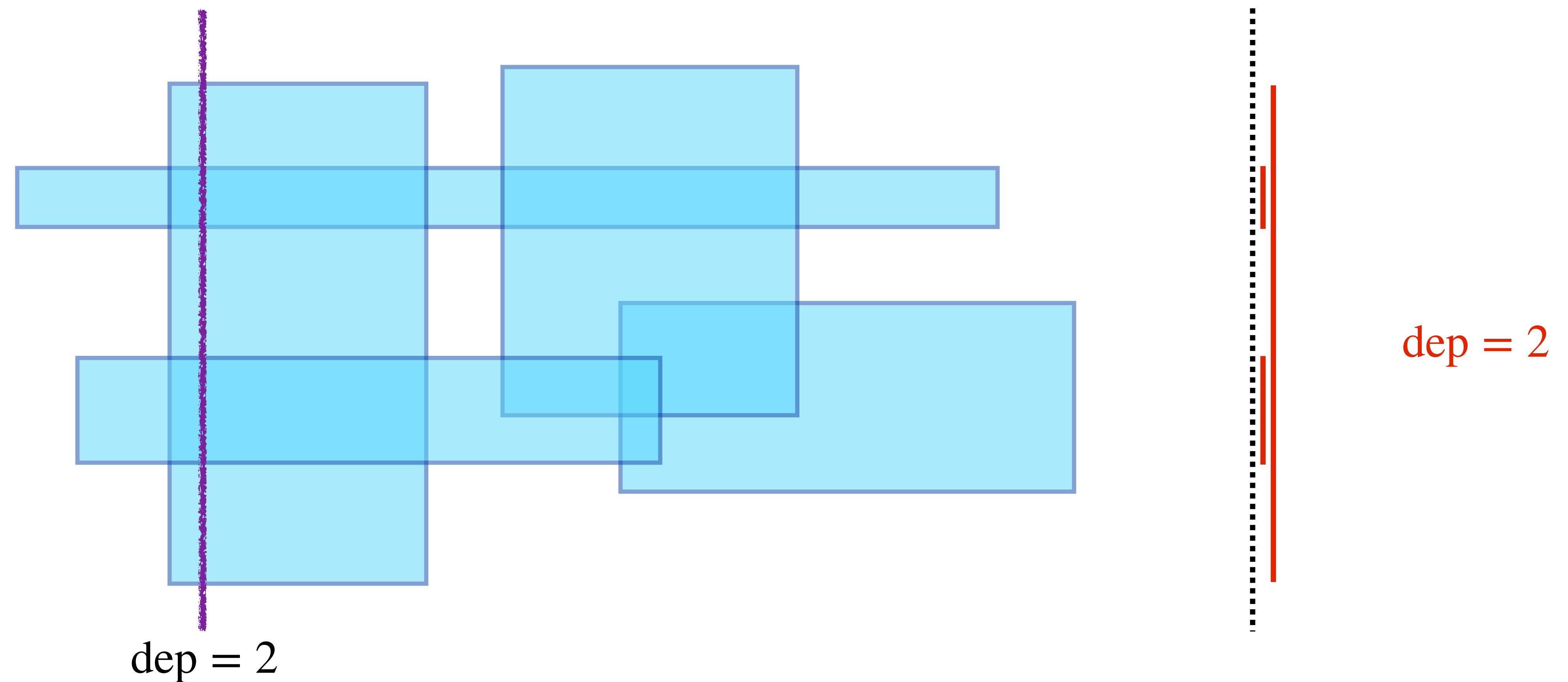
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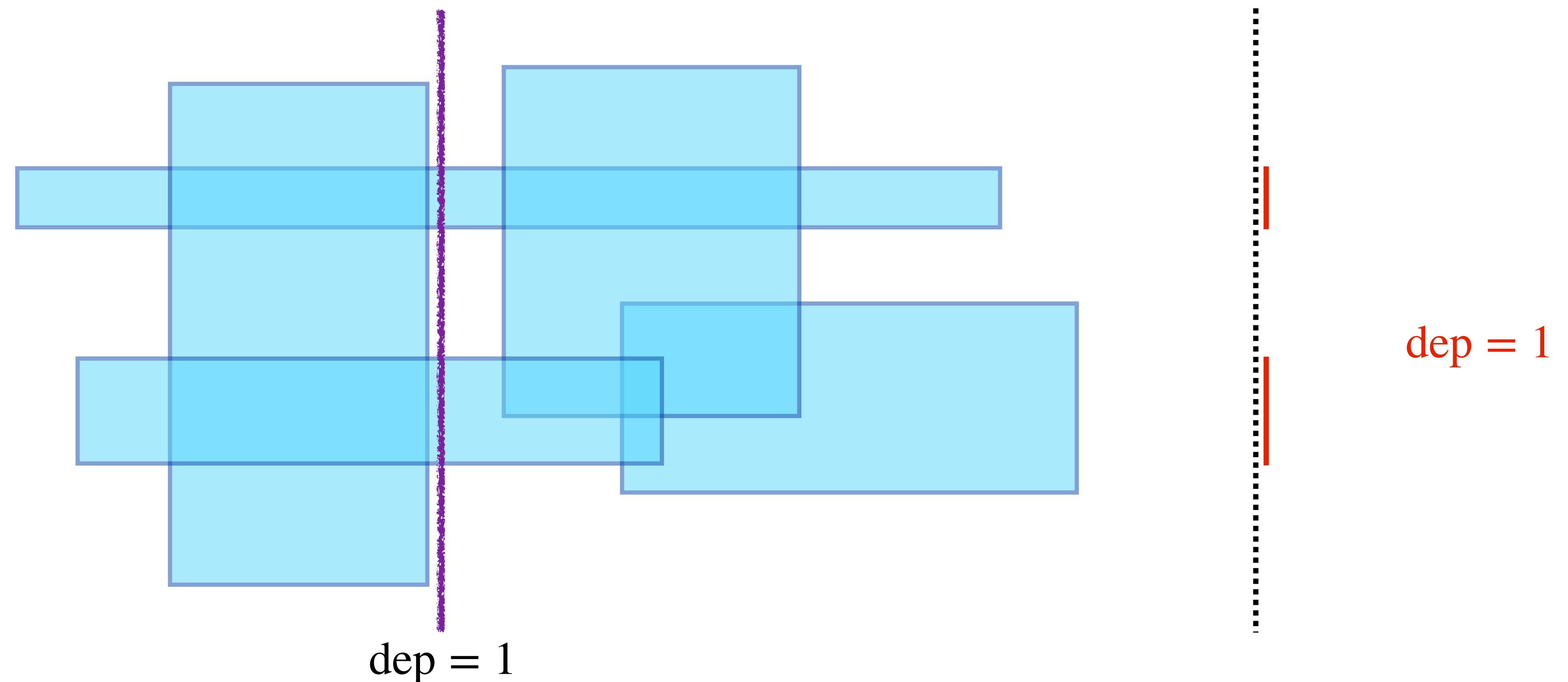
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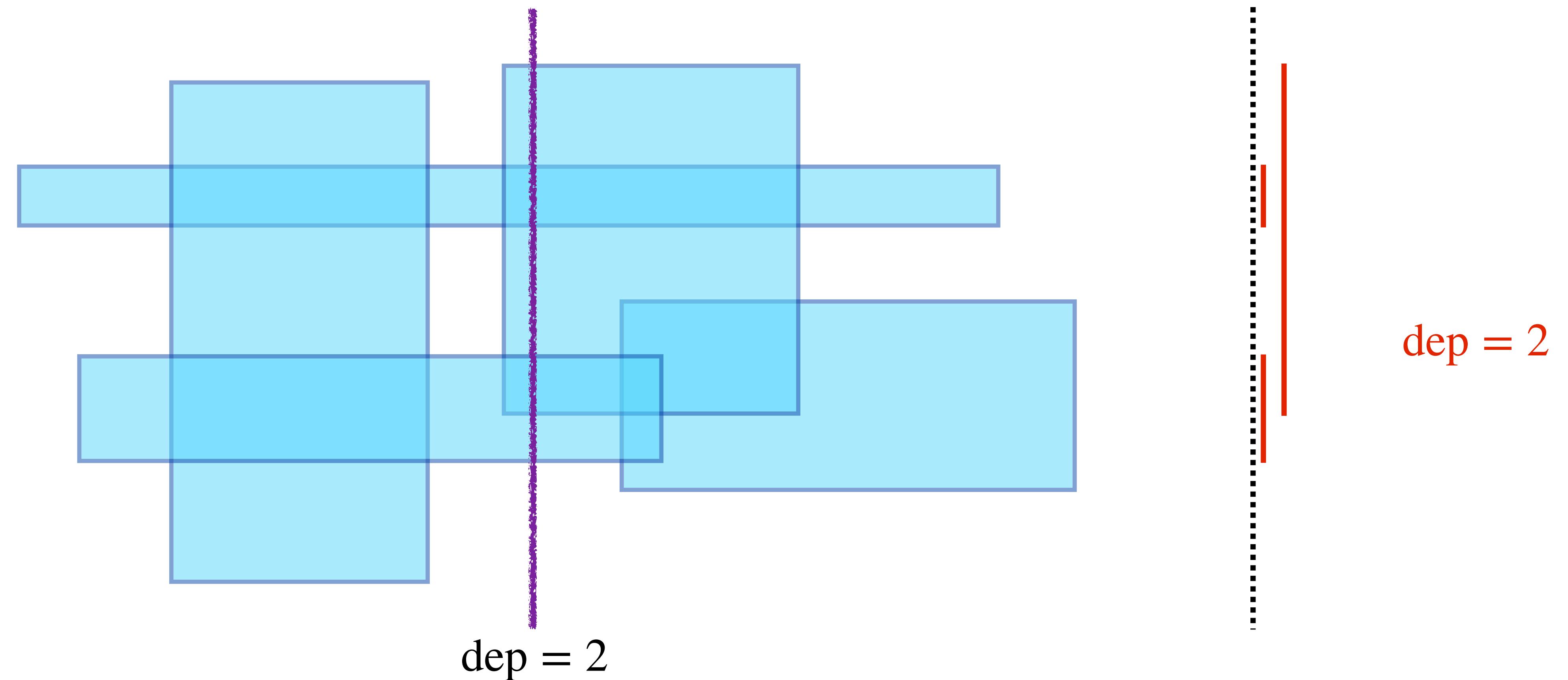
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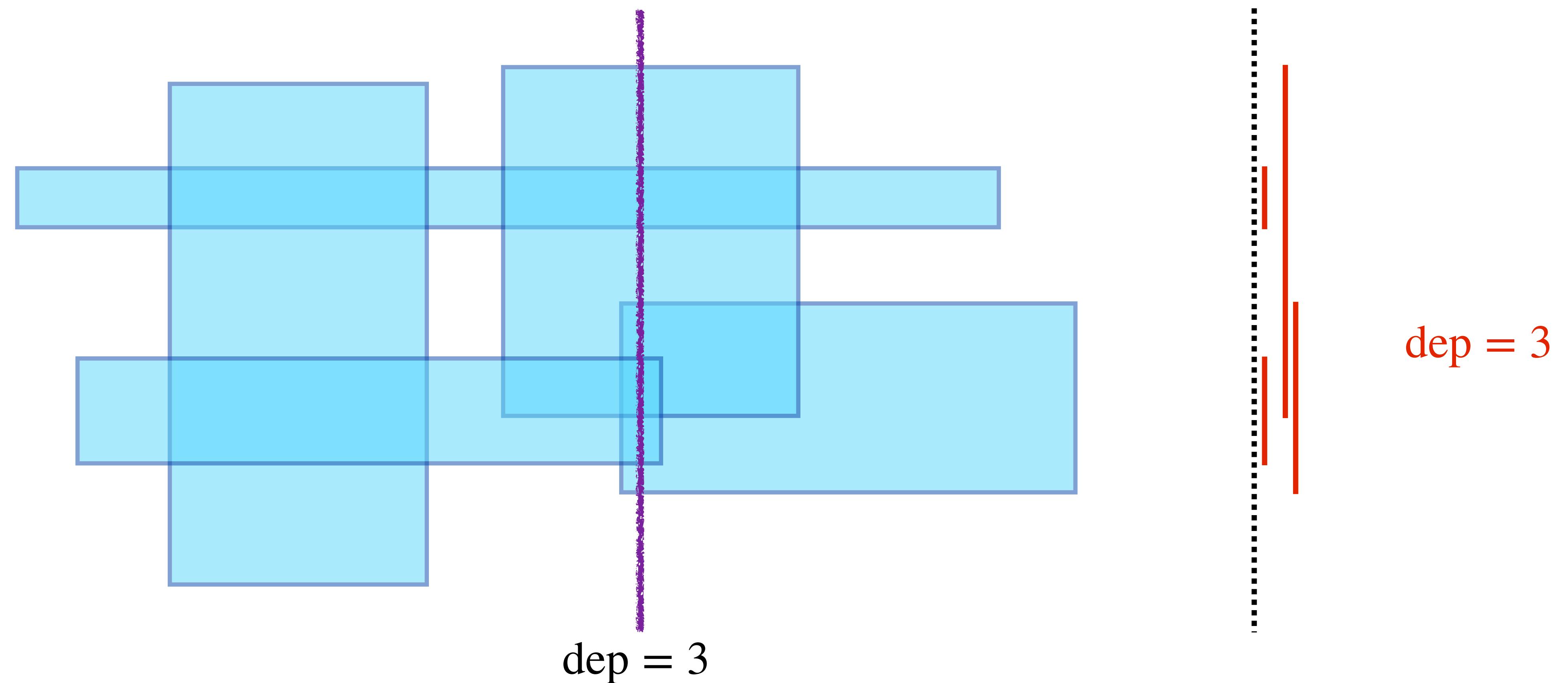
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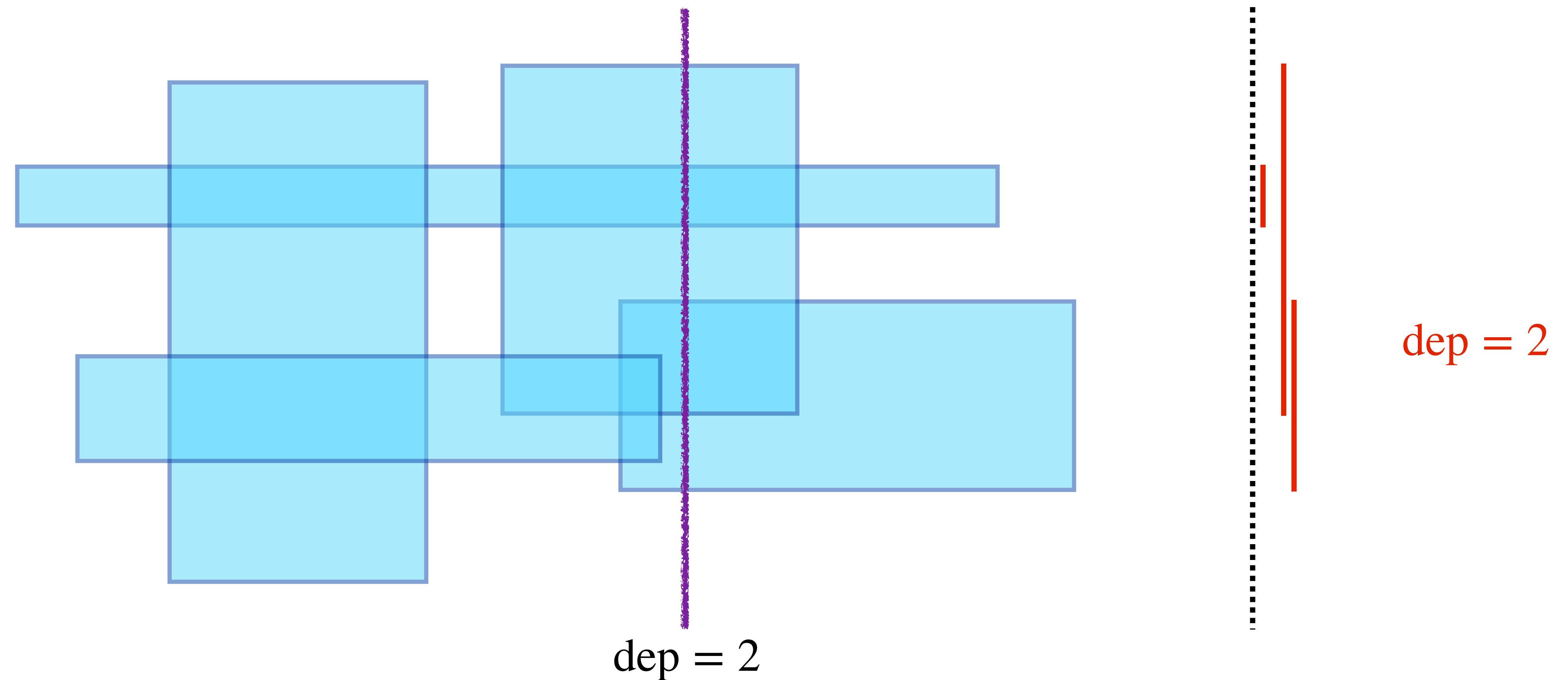
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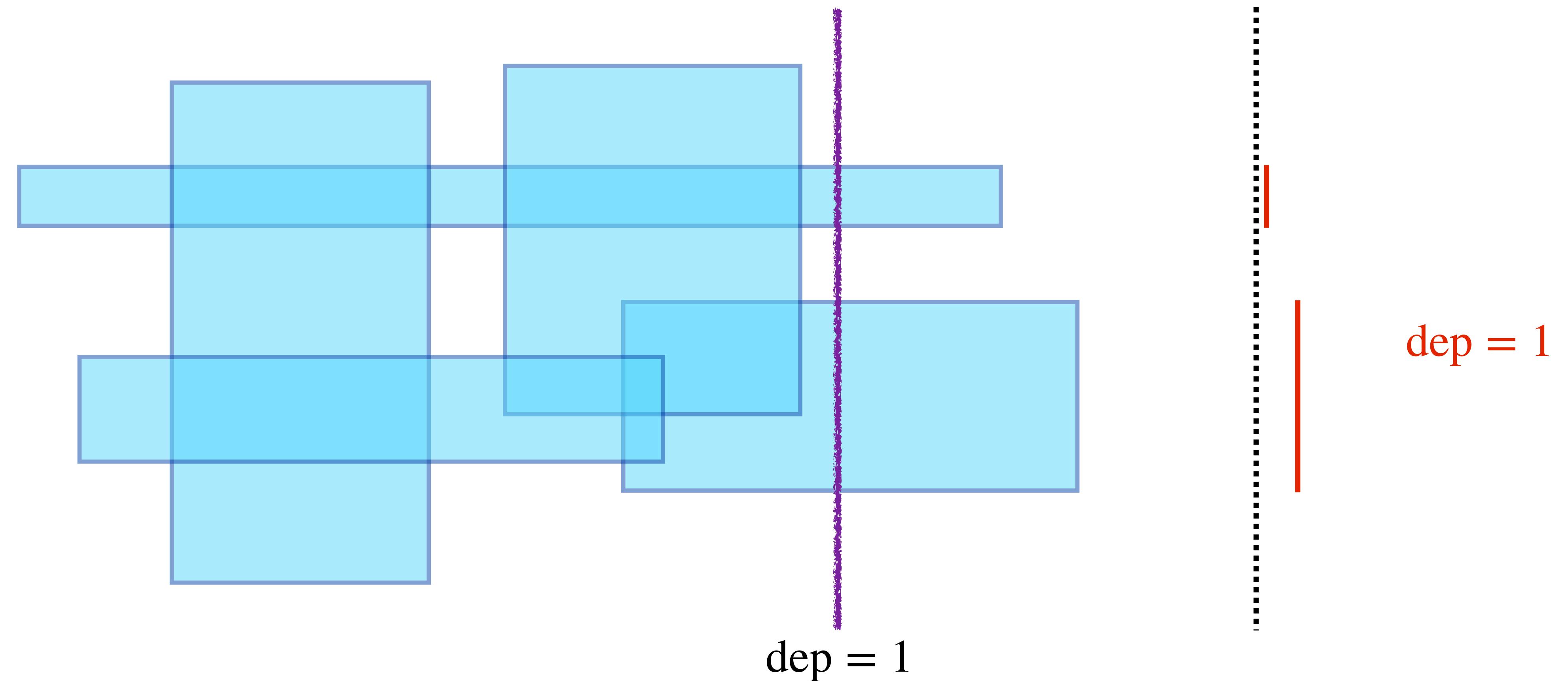
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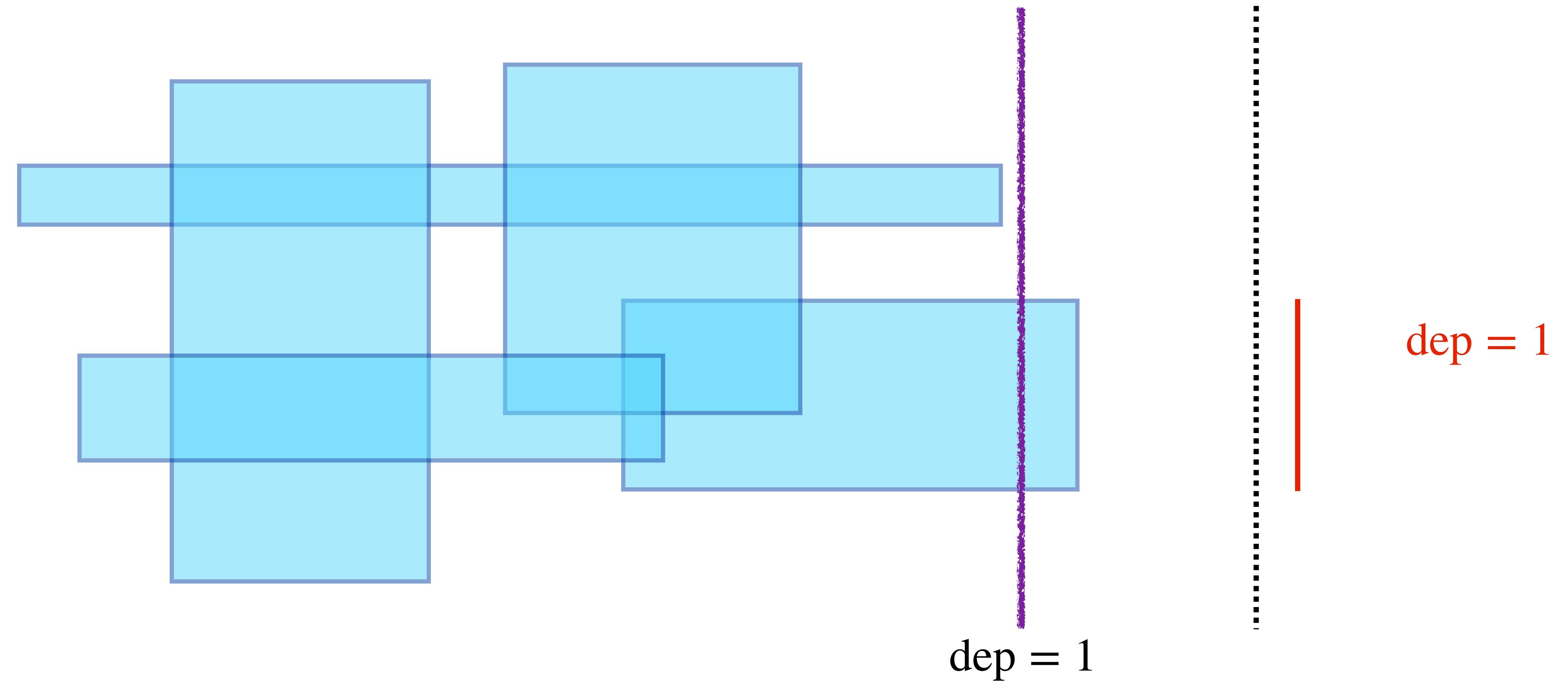
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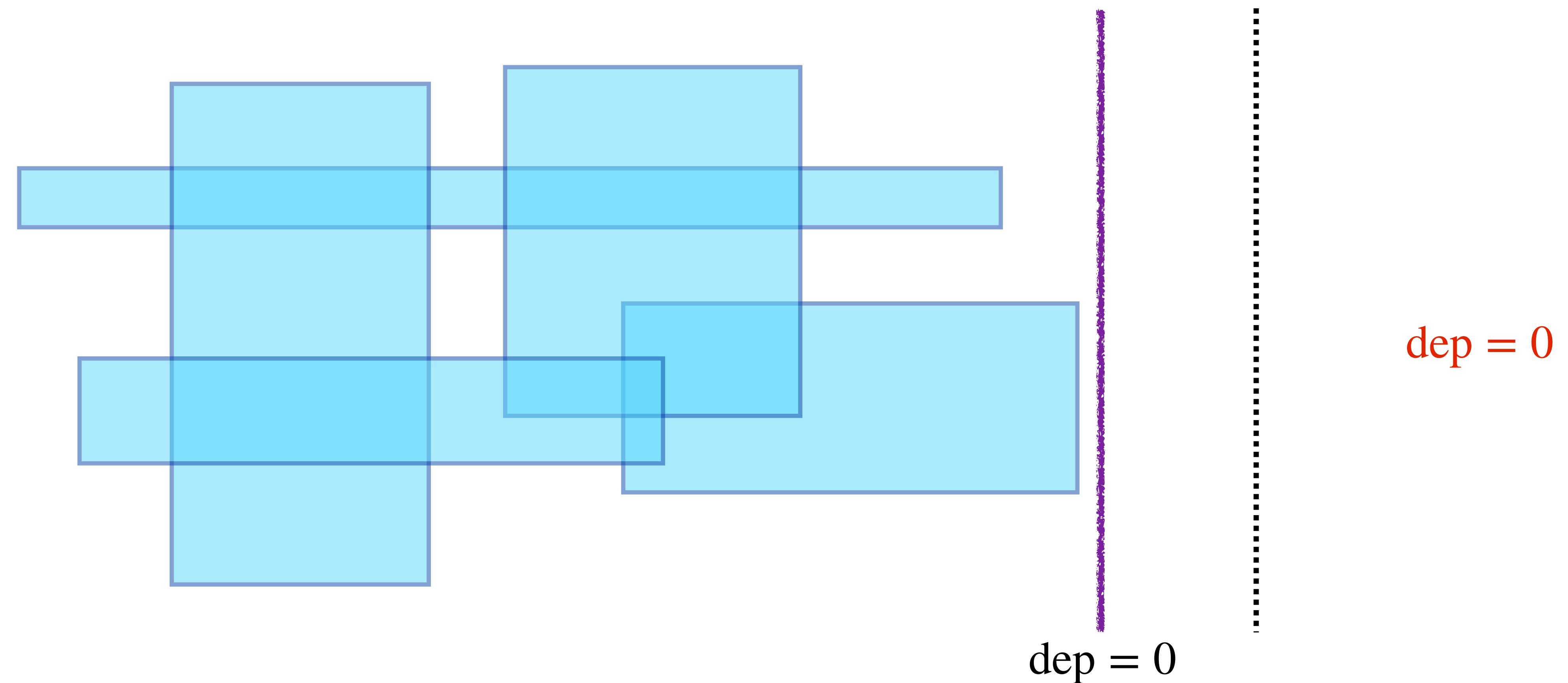
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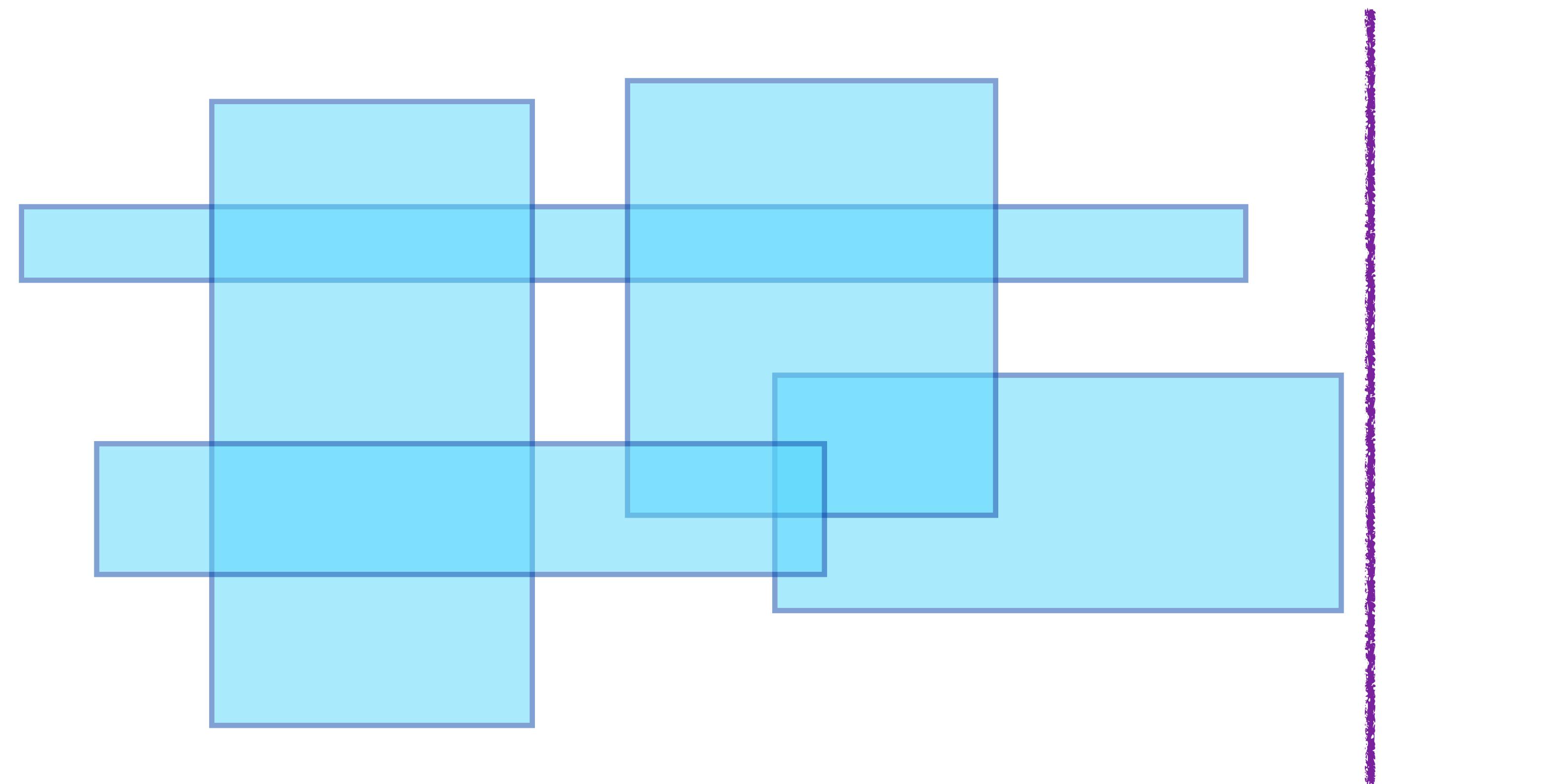
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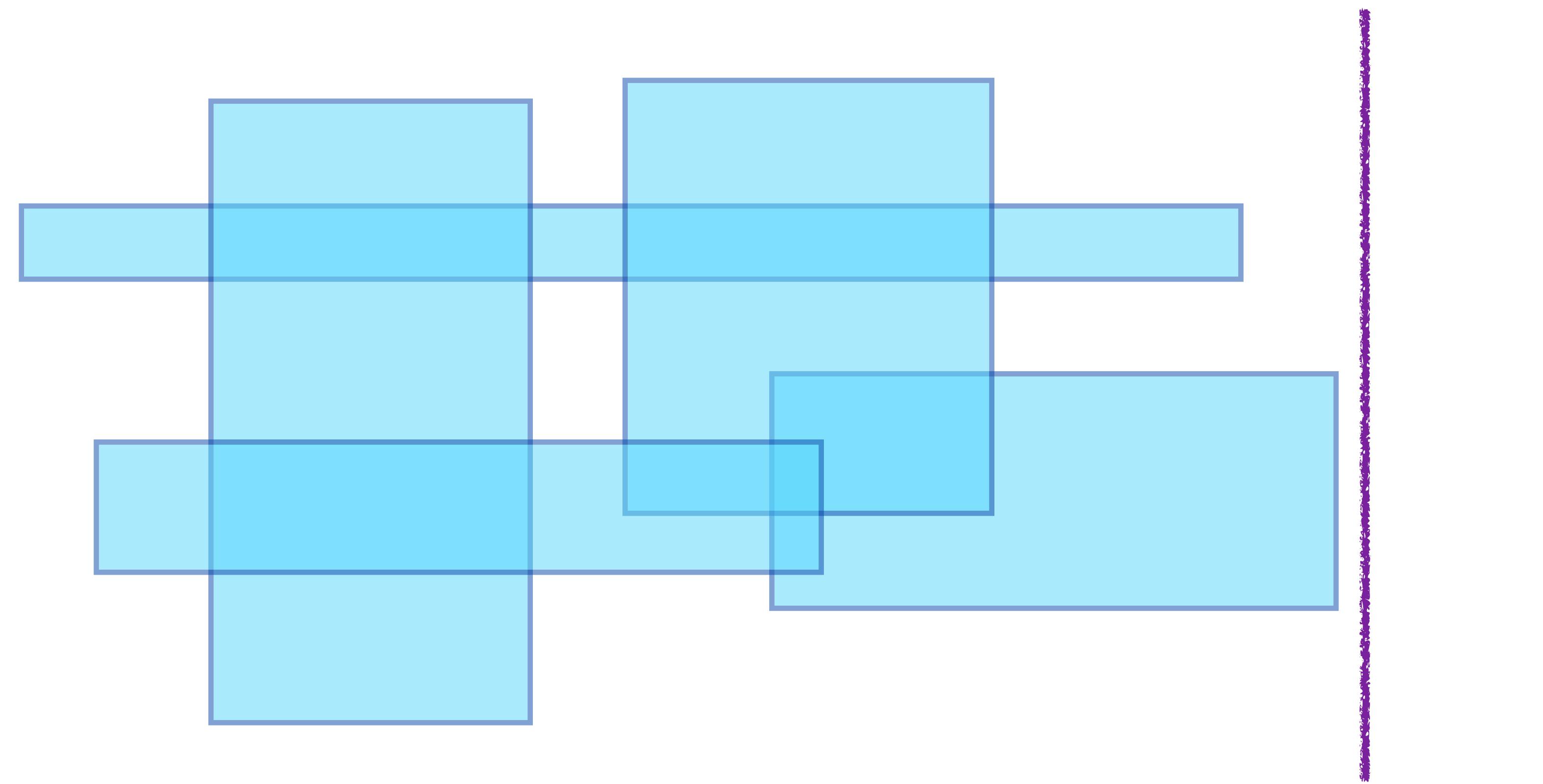
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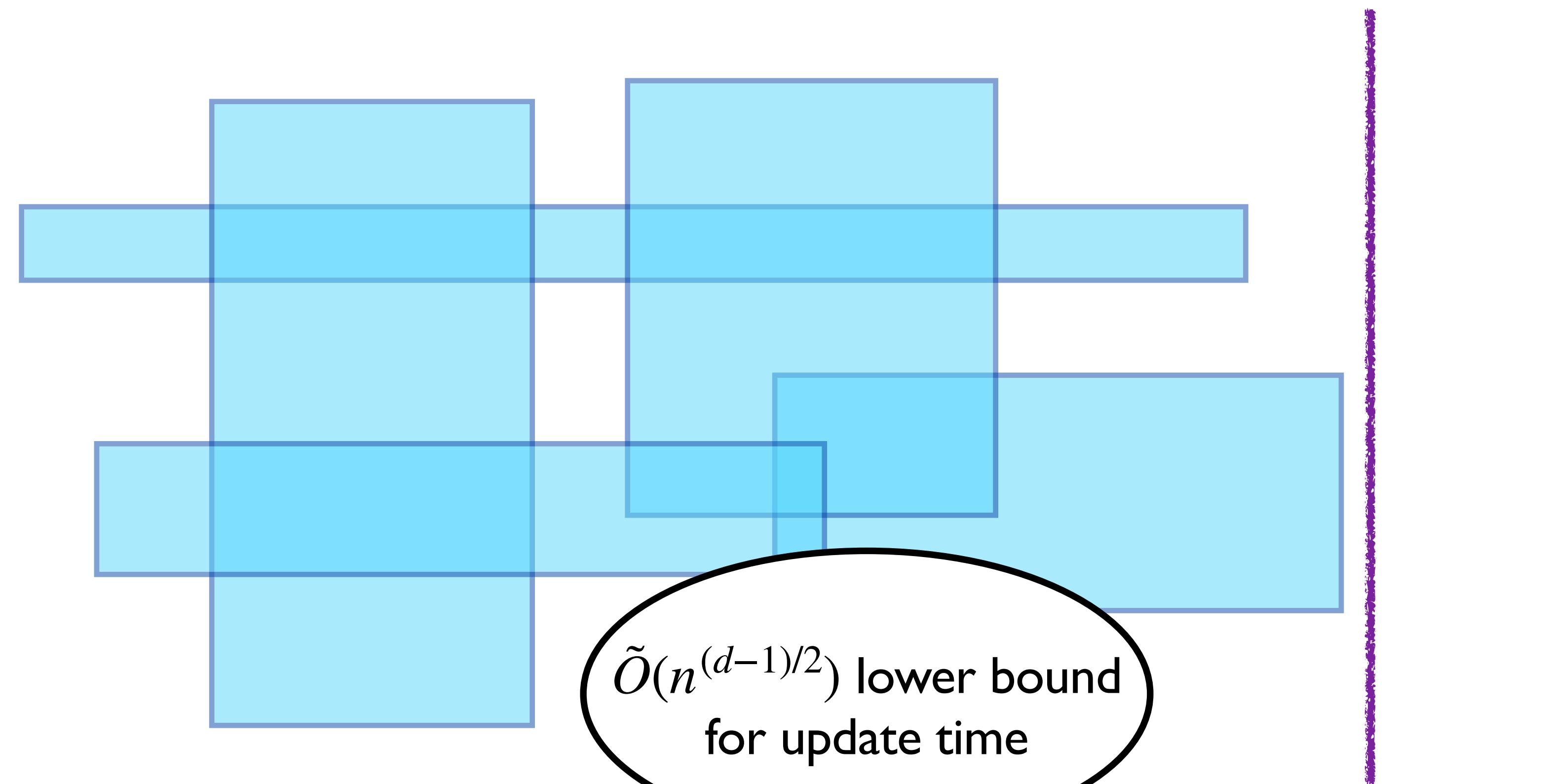


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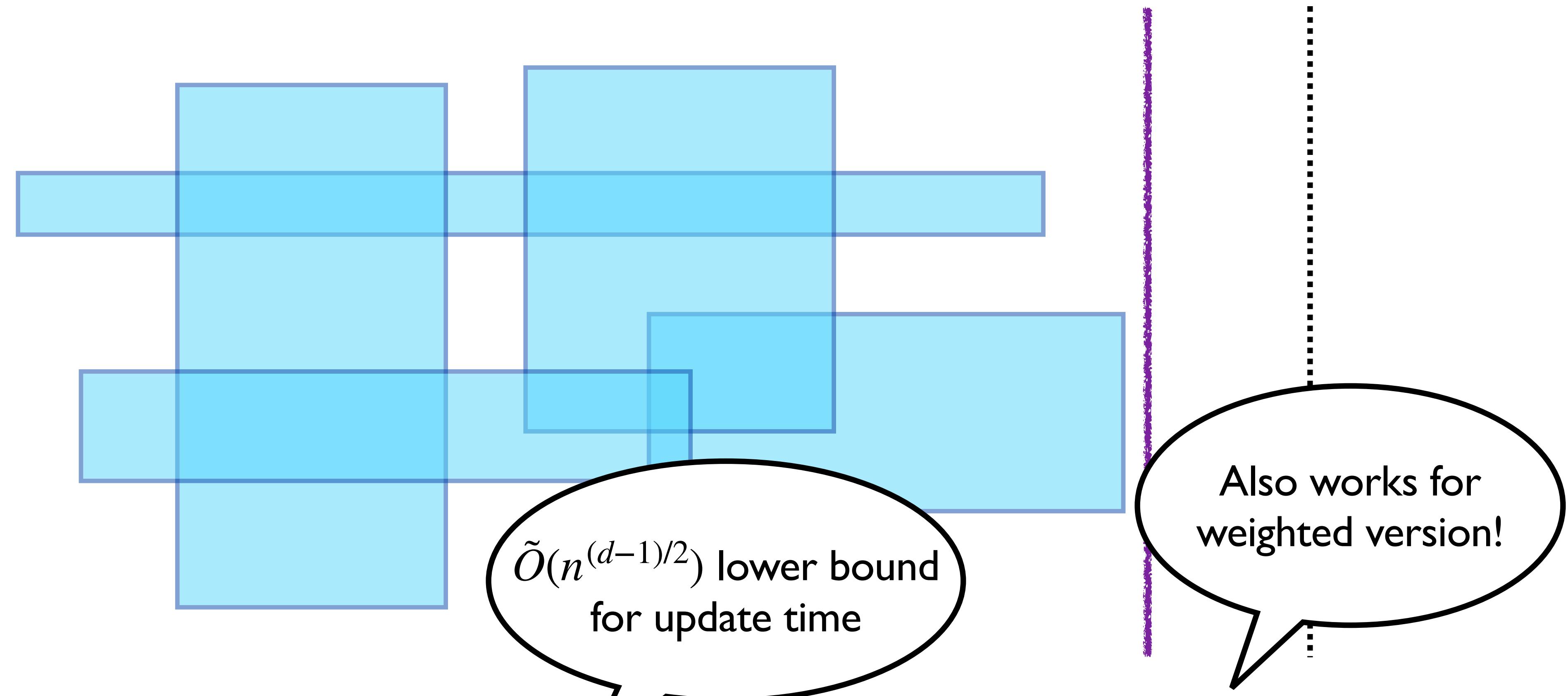


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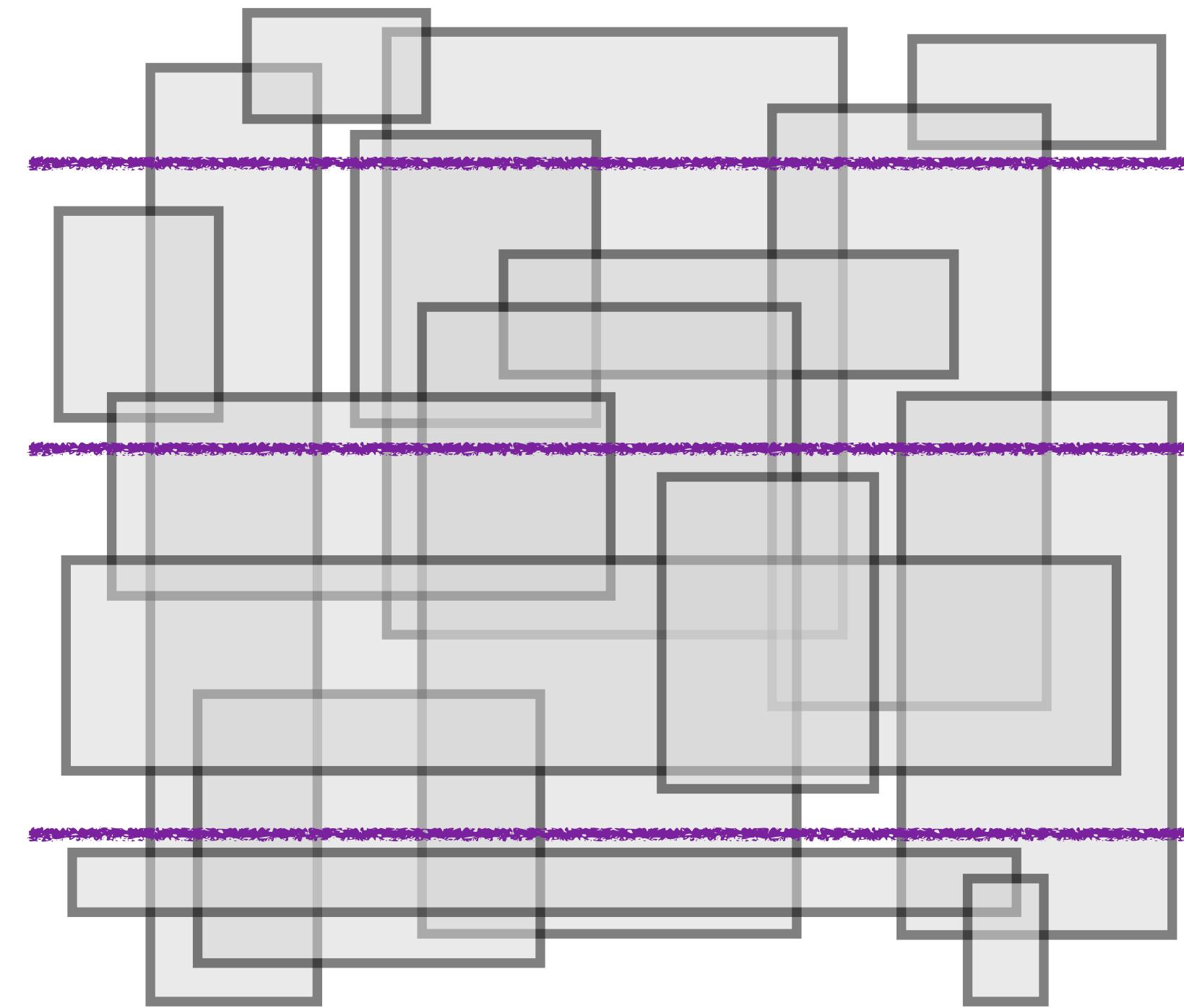
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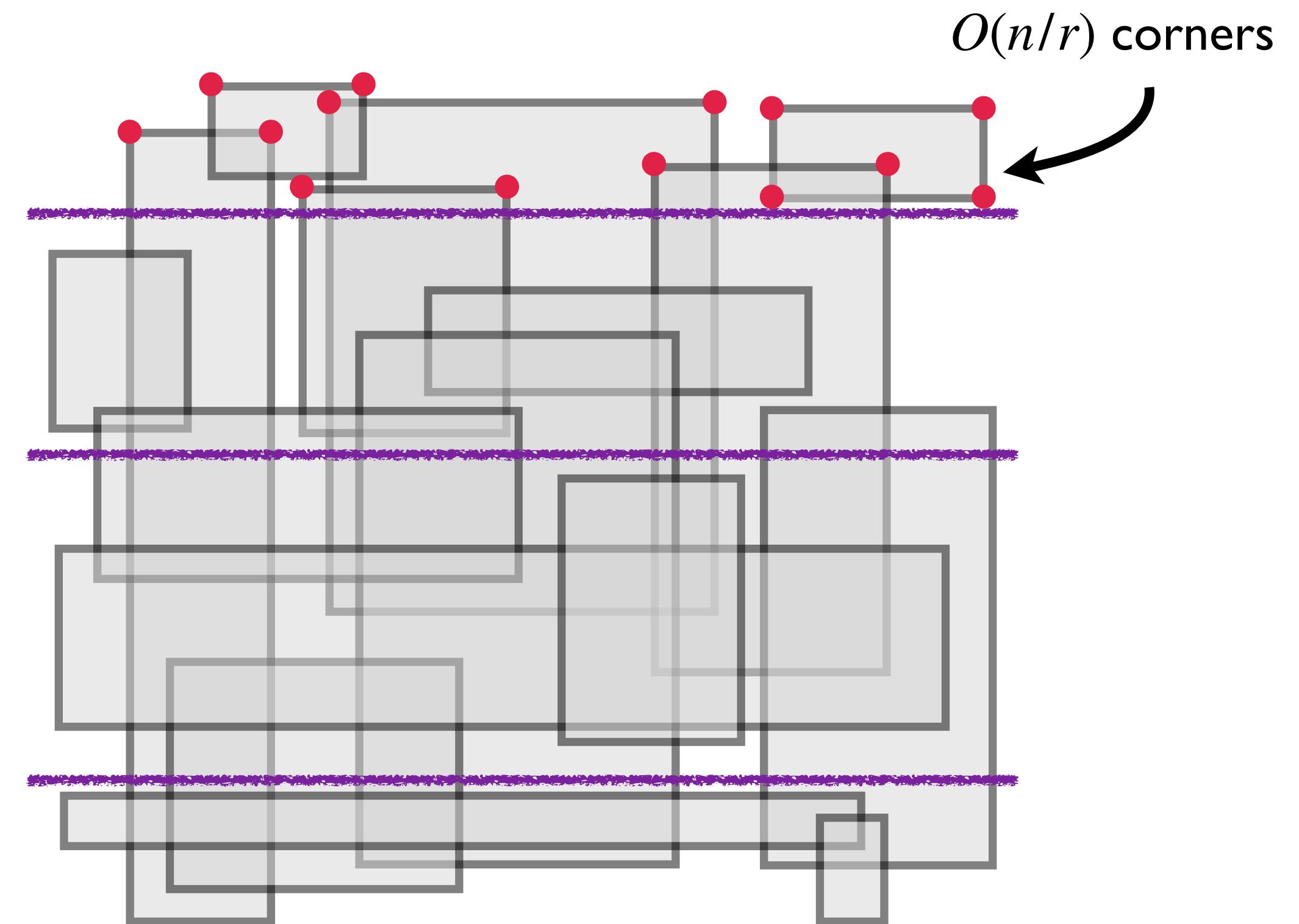
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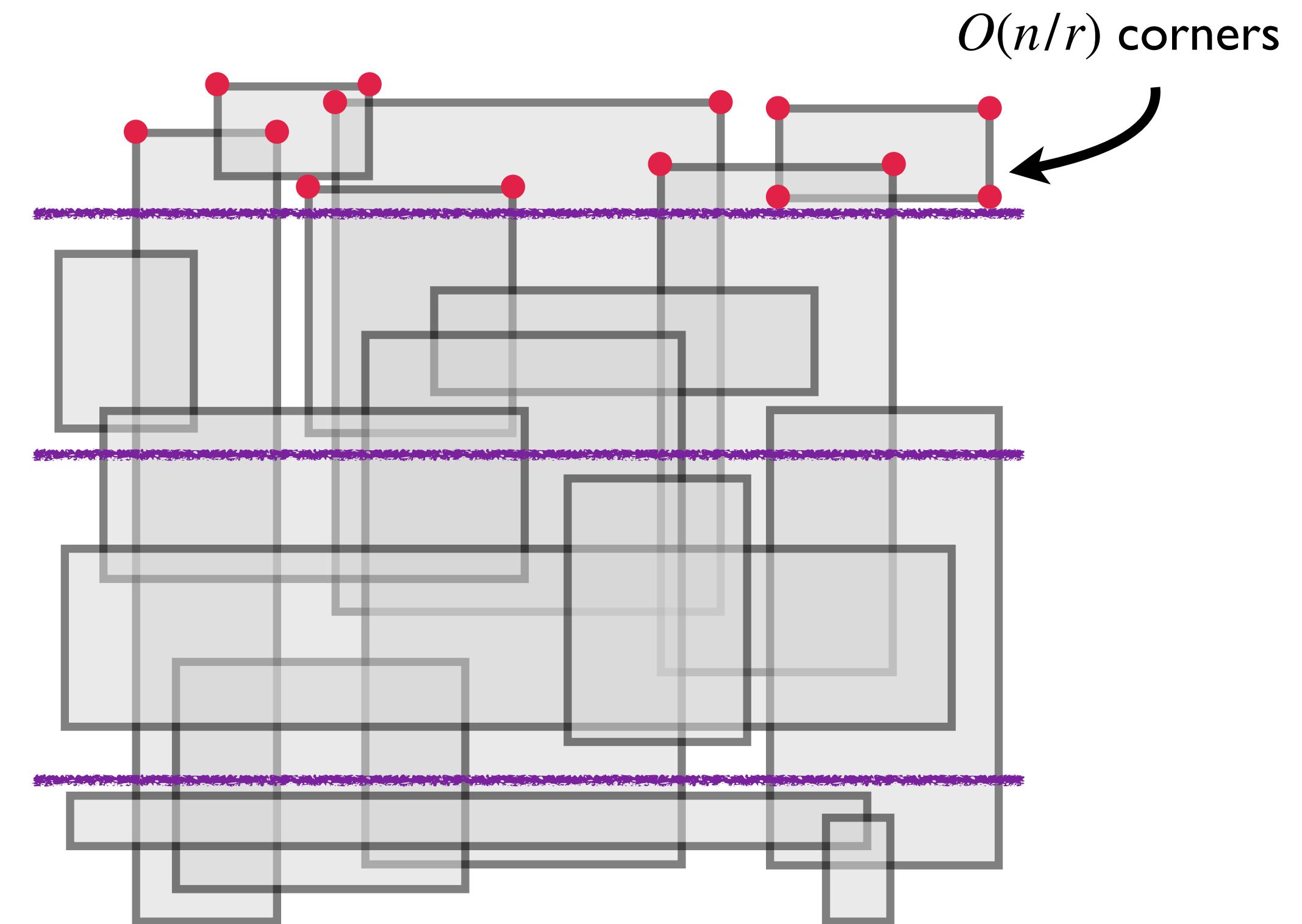
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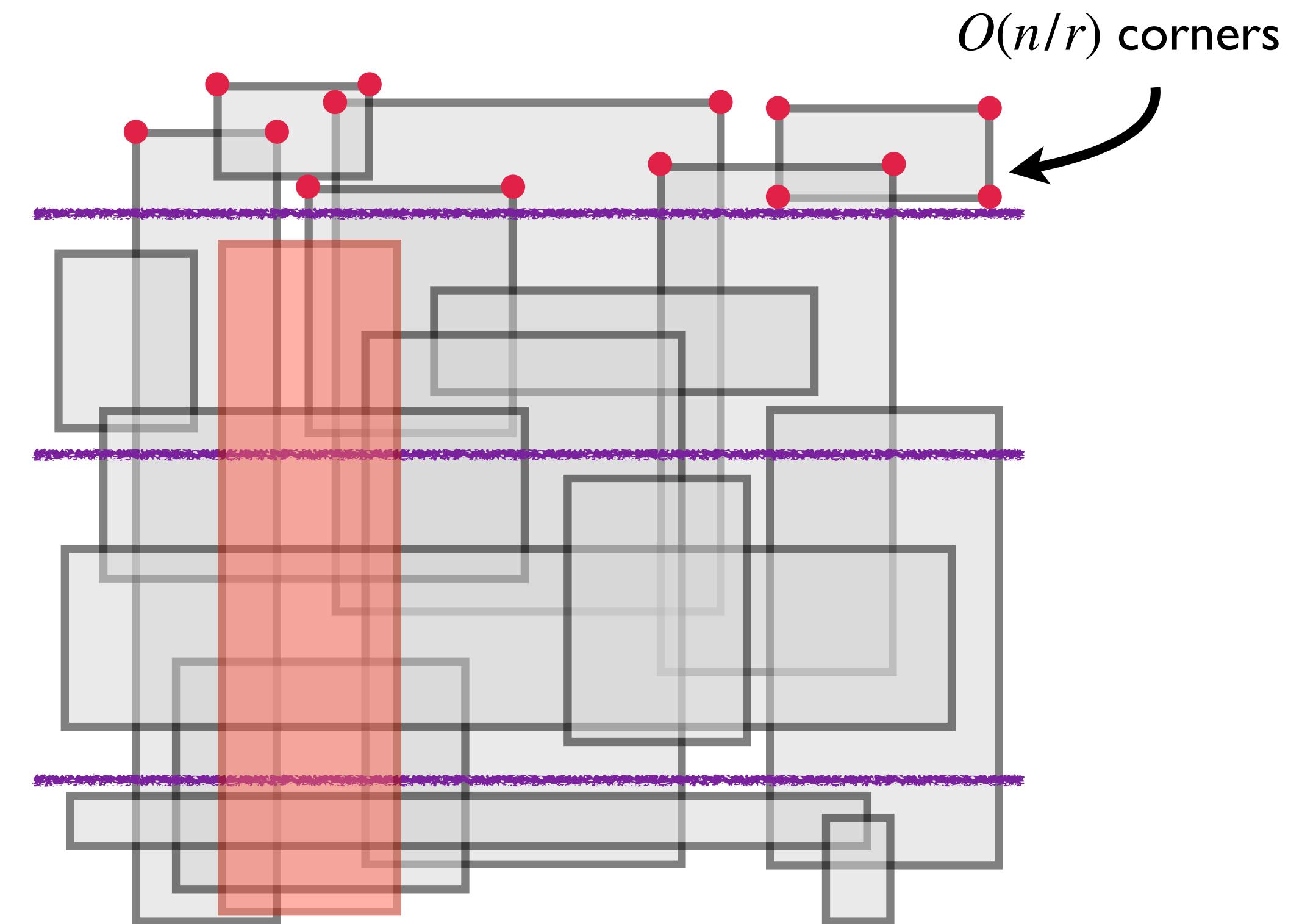
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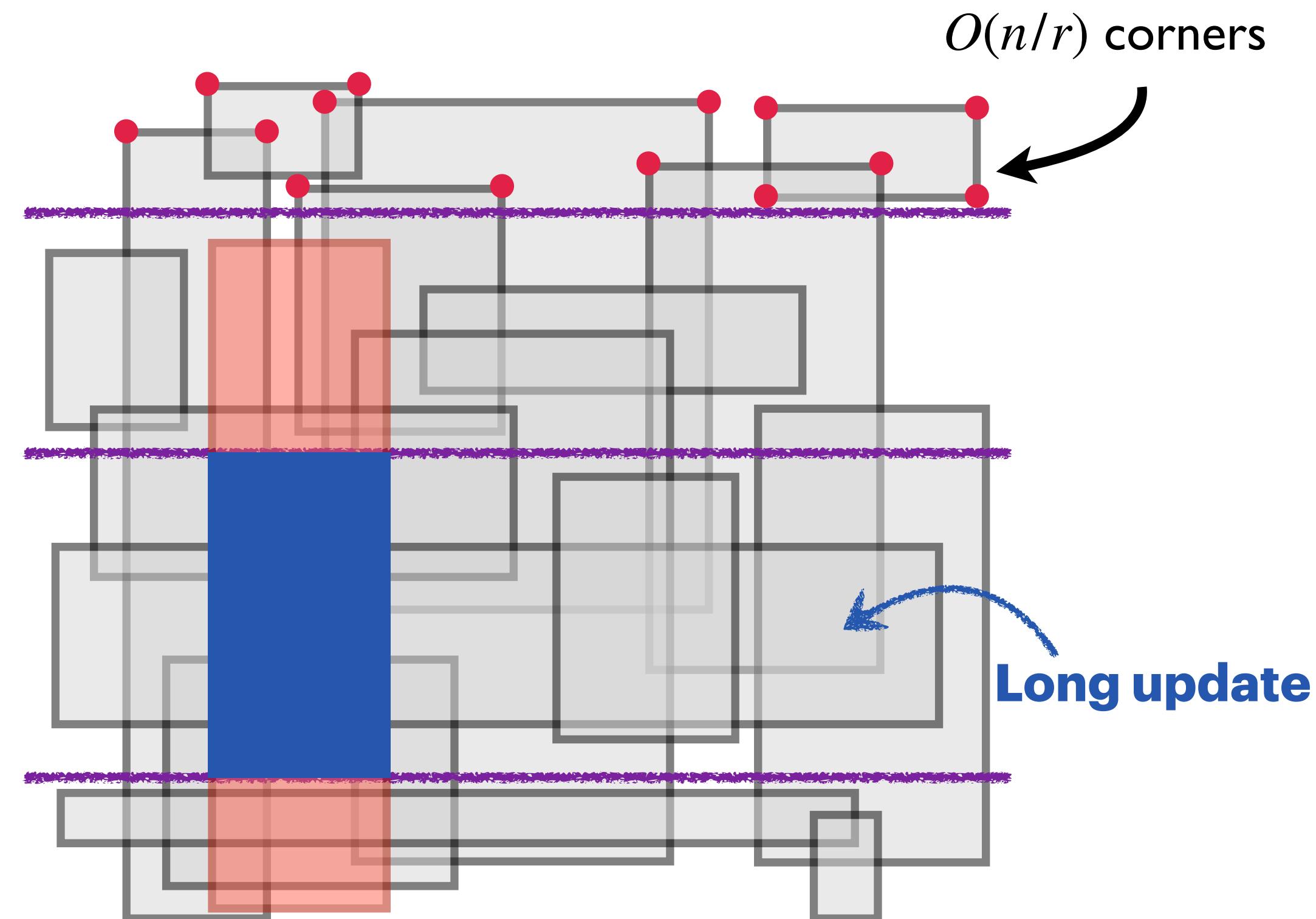
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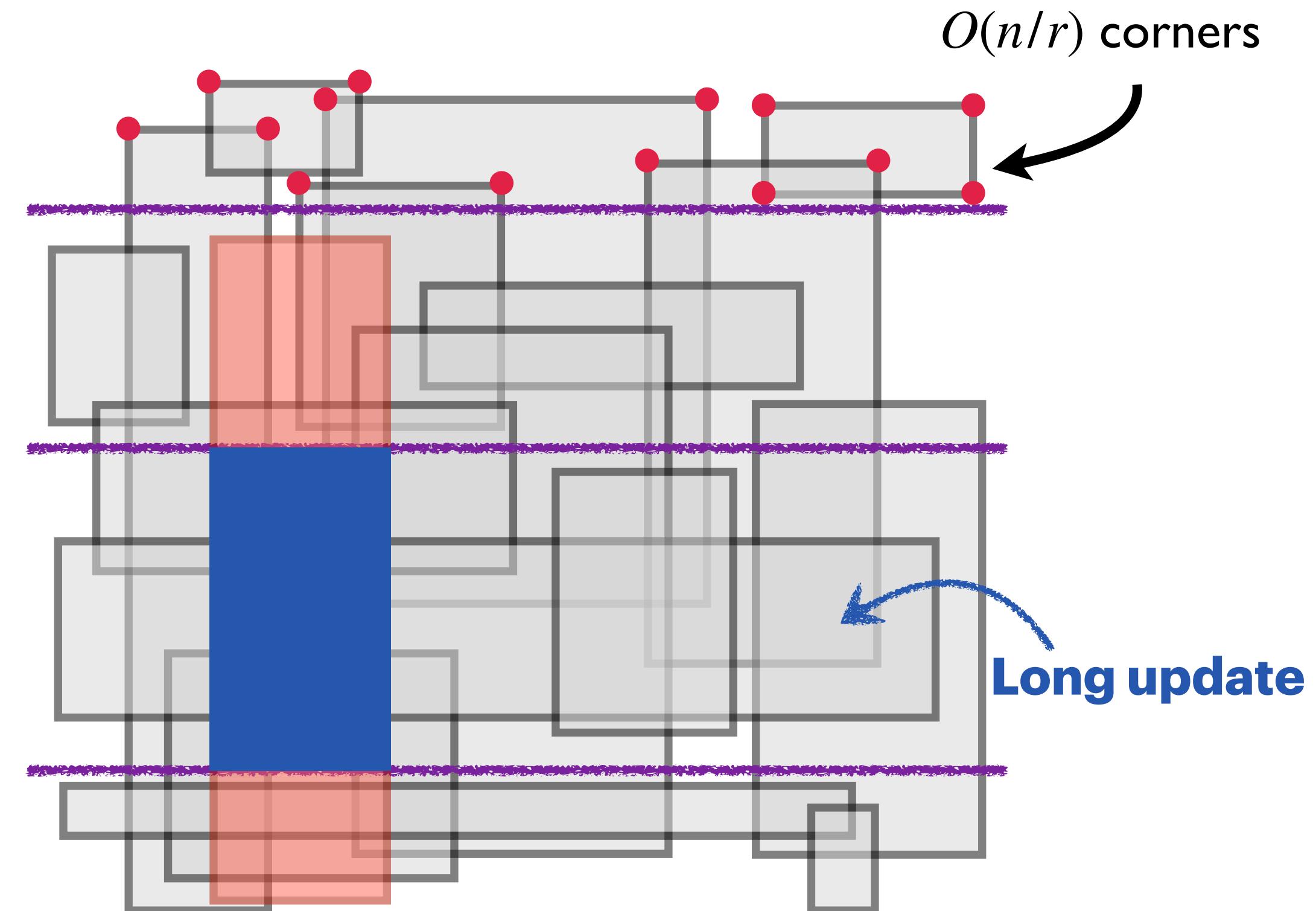
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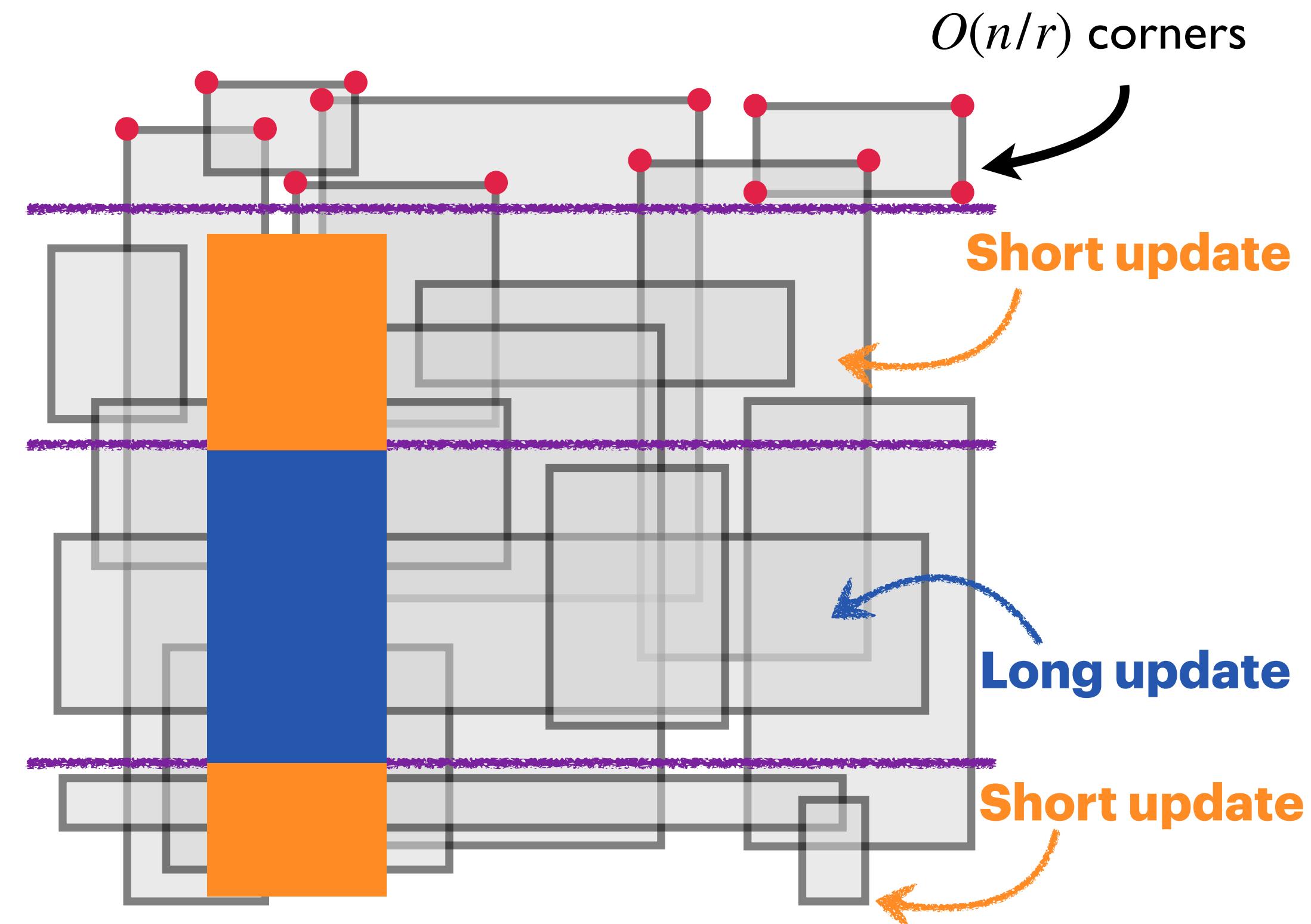
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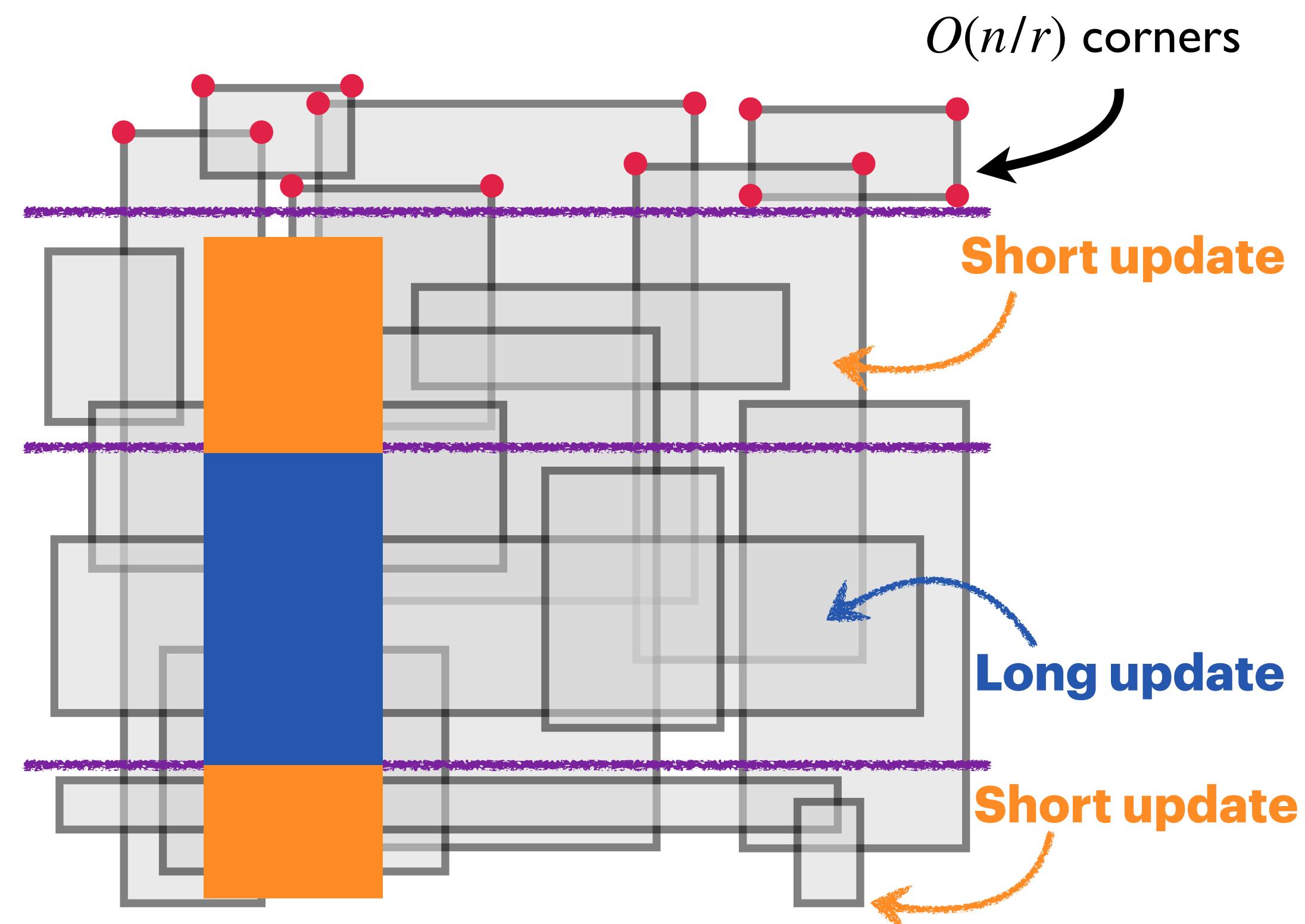
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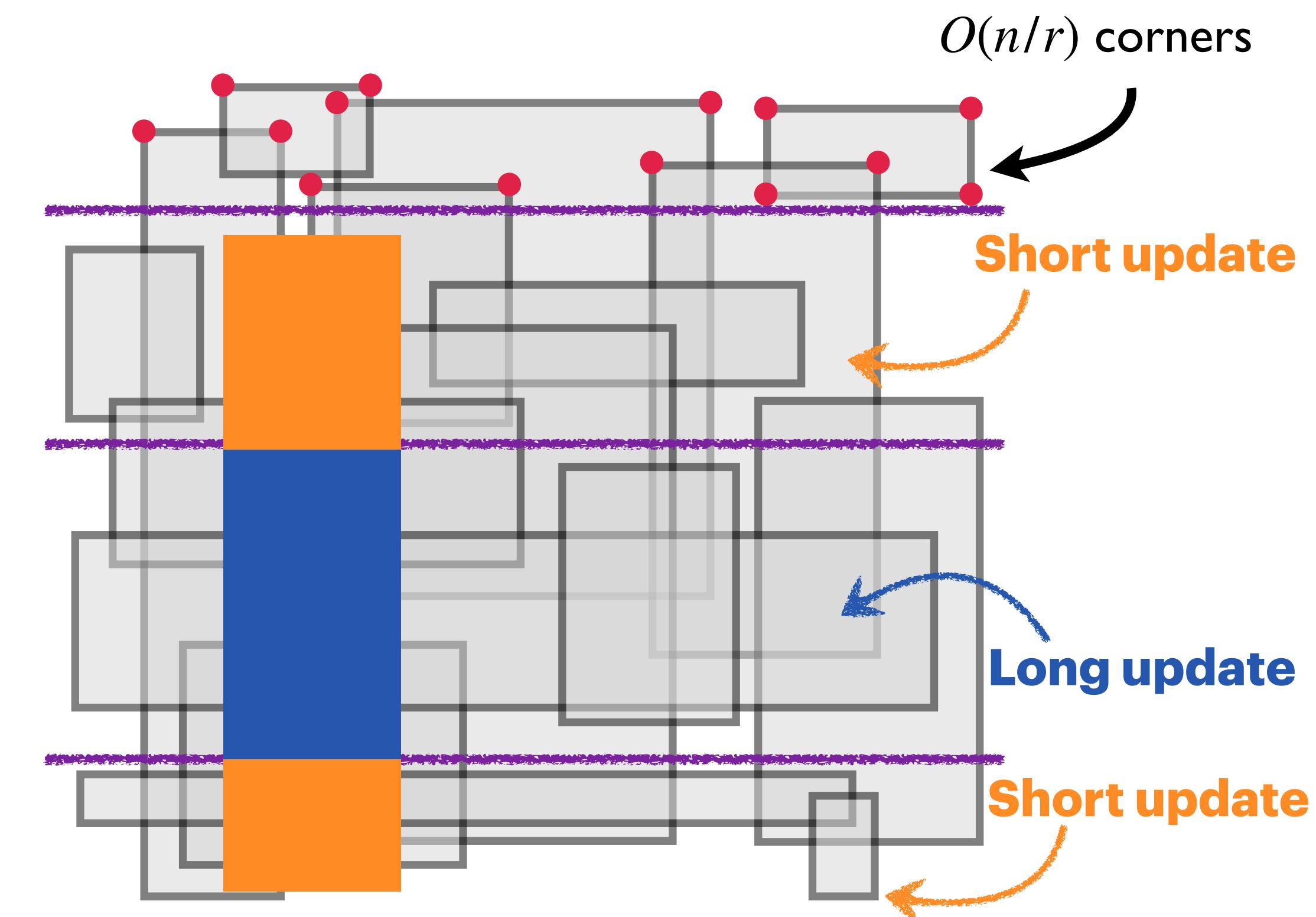
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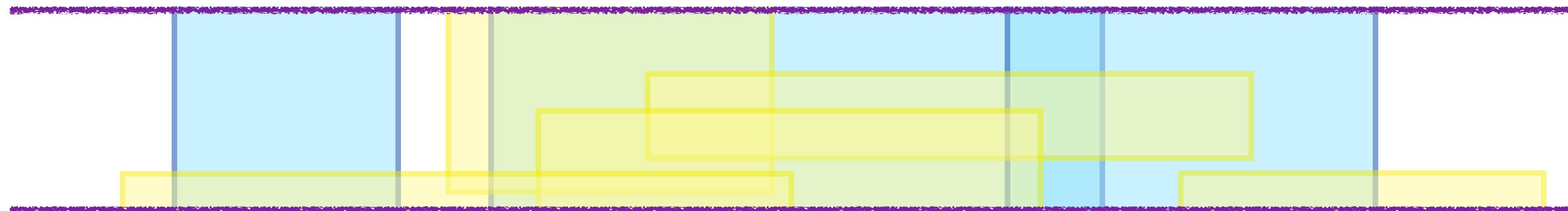


## Total update time

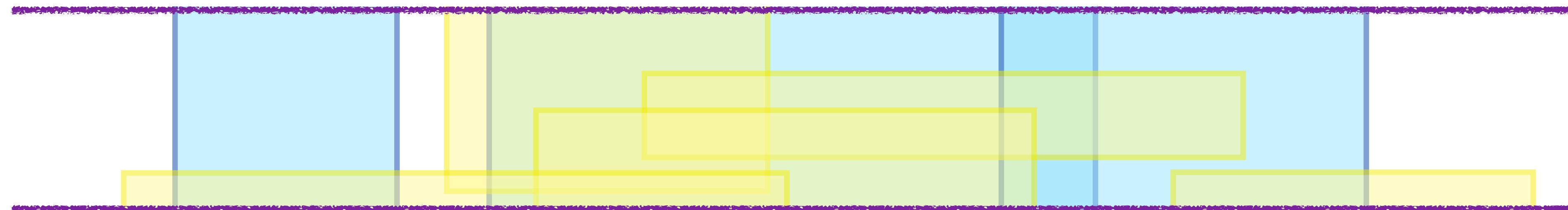
$$O\left(2 \cdot \frac{n}{r} \log n + r \cdot \log n\right) = \tilde{O}\left(\sqrt{n}\right)$$

when  $r = \sqrt{n}$

# Details of Slab Data Structure

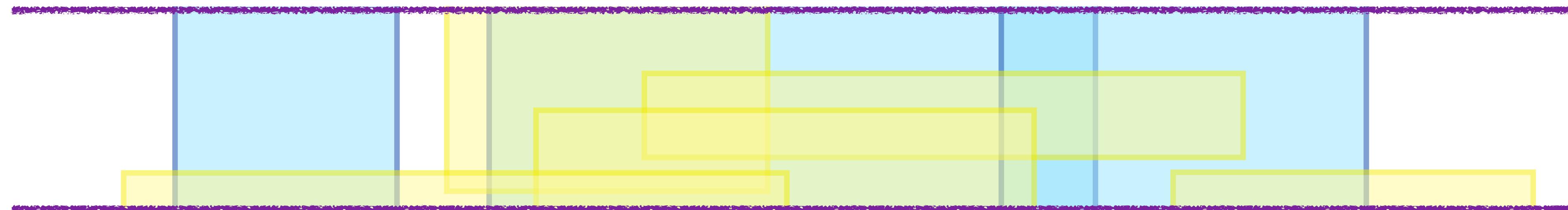


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**Reduce to intervals**

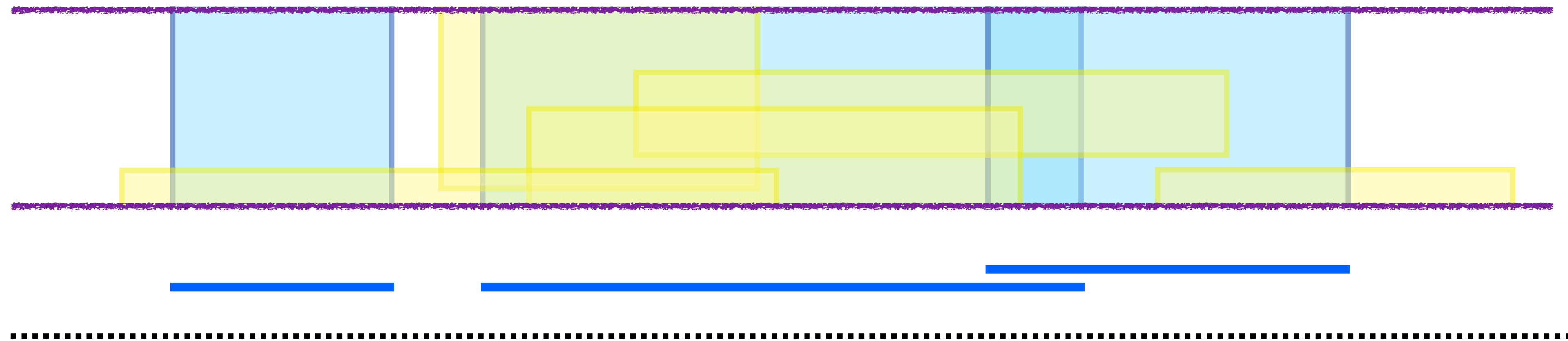
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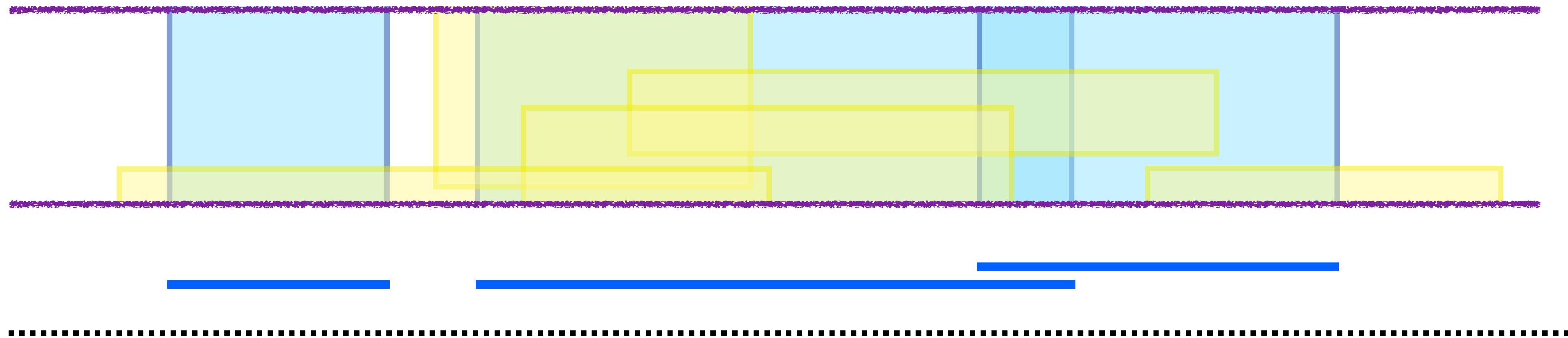
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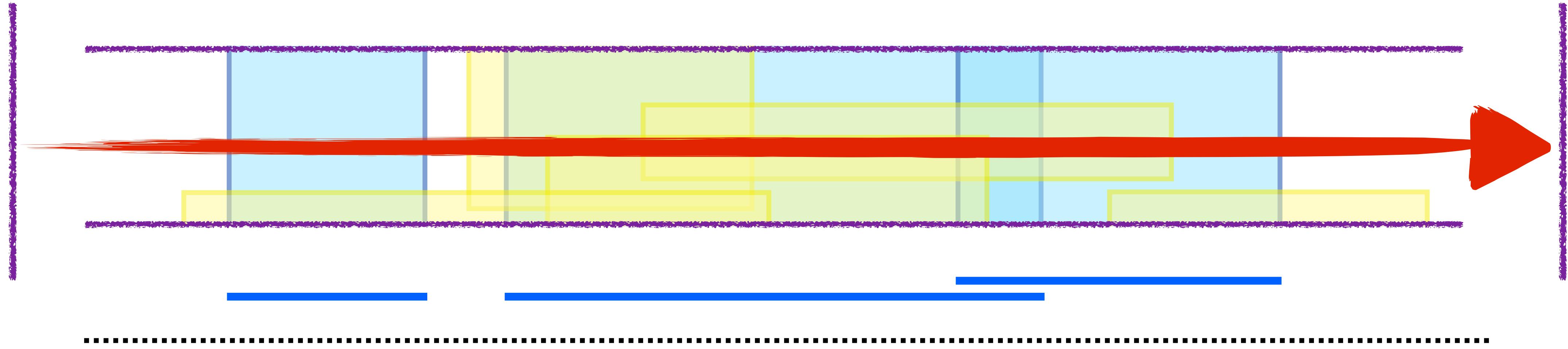
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interval: two adjacent boundaries  
weight: max dep w.r.t. short rectangles

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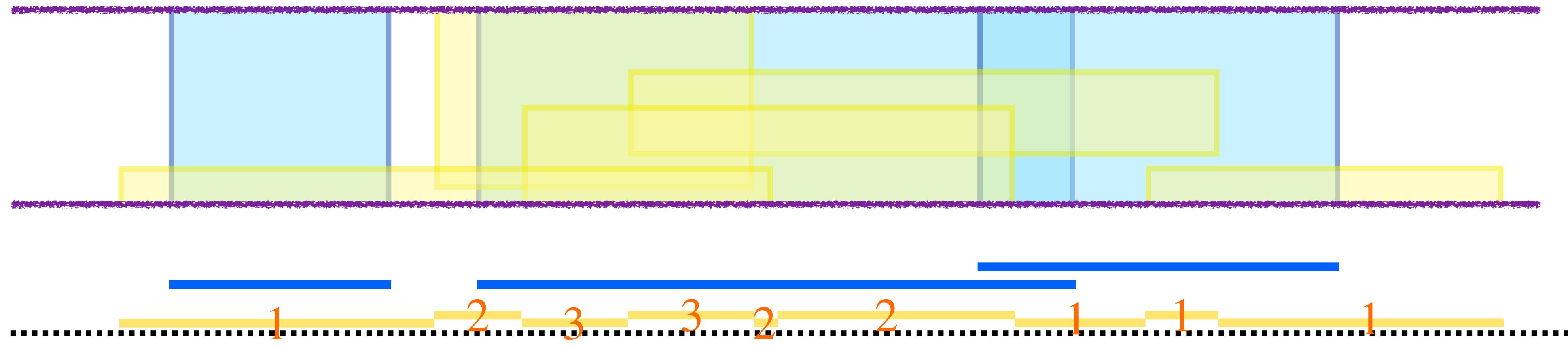
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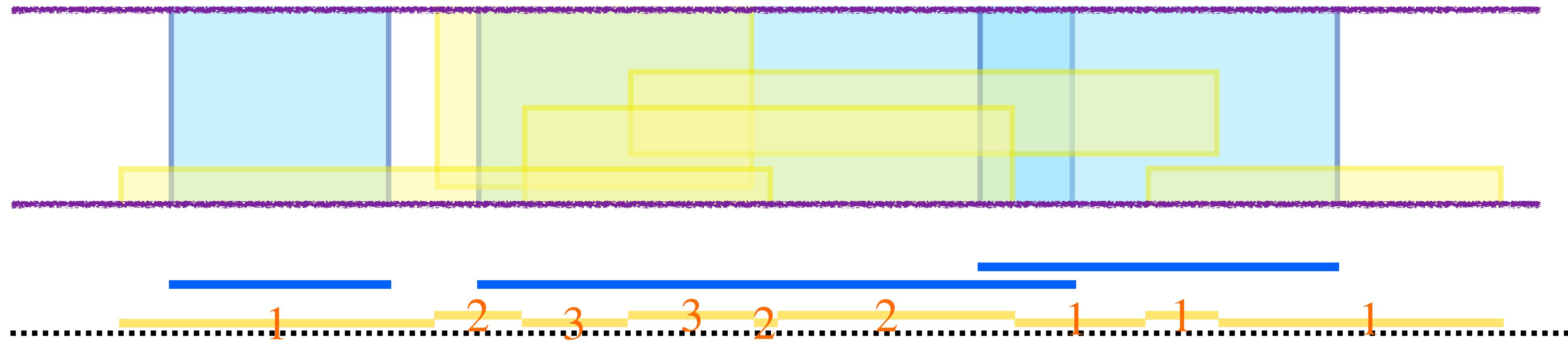
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$O(n)$  long rectangles  $\xrightarrow[\text{O}(n) \text{ time}]{\text{directly}} O(n)$  unweight intervals

$O\left(\frac{n}{r}\right)$  short rectangles  $\xrightarrow[\text{O}\left(\frac{n}{r} \log \frac{n}{r}\right) \text{ time}]{\text{sweep line}} O(n/r)$  weighted intervals

interval: two adjacent boundaries  
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# Details of Slab Data Structure



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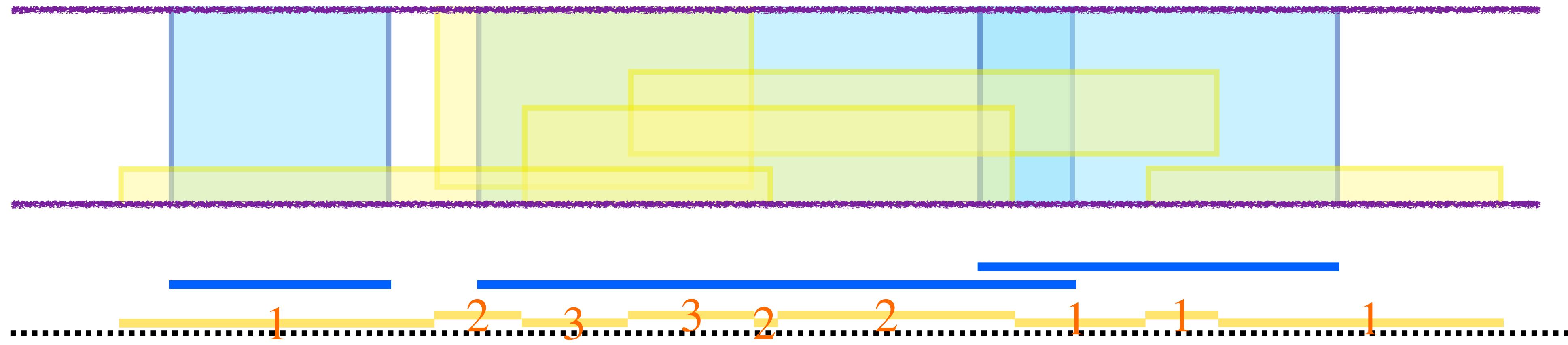
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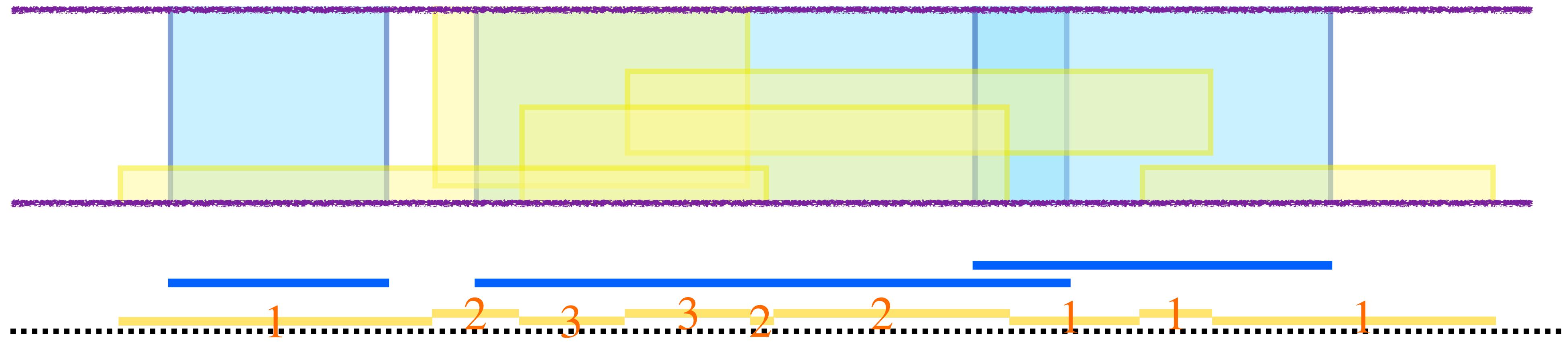
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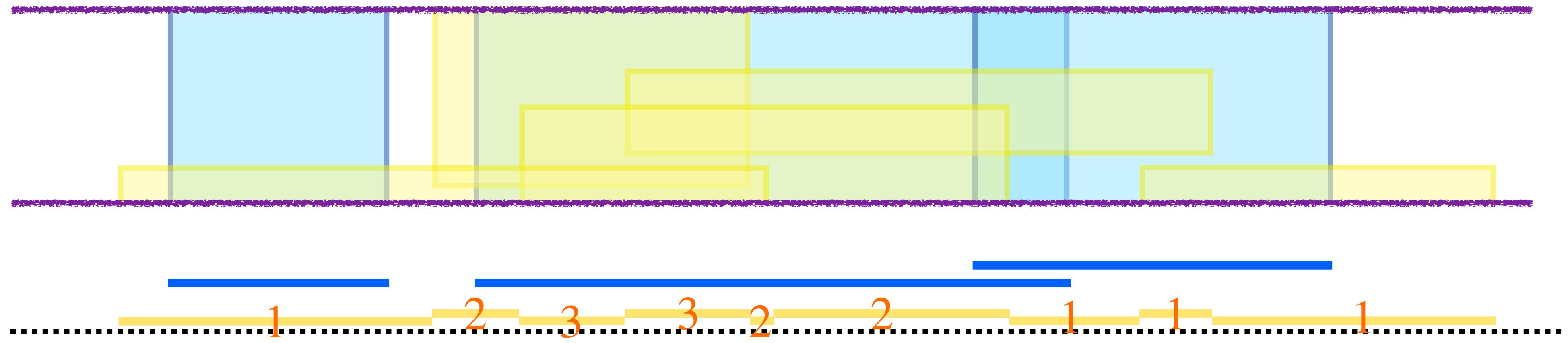
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Total preprocessing time:  $O(n \log n)$

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## Update

Long update:  $O(\log n)$

Short update:

- Delete all weighted intervals
- Recompute weighted intervals
- Insert all weighted intervals

$O((n/r) \cdot \log n)$

# Approx. Ratio vs Update Time



Exact data structure: balancing long and short rectangles

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Speed-up for approx.: maintain long and short rectangles separately

# Approx. Ratio vs Update Time

$D_1$ : exact data structure

$I$ : interval data structure

$\mathcal{L}$ : set of long rect.

$\mathcal{S}$ : set of short rect.



Exact data structure: balancing long and short rectangles



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$D_2$  : Build  $D_1$  on  $\mathcal{S}$ ,  $I$  on  $\mathcal{L}$

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$D_k$  : Build  $D_{k-1}$  on  $\mathcal{S}$ ,  $I$  on  $\mathcal{L}$

$$\max \left\{ \frac{1}{k-1} \text{dep}(\mathcal{S}), \text{dep}(\mathcal{L}) \right\} \geq \frac{1}{k} \text{dep}(\mathcal{O})$$
$$\tilde{O} \left( 2 \cdot (n/r)^{\frac{1}{k}} + r \right) = \tilde{O} \left( n^{\frac{1}{k+1}} \right)$$

$r = n^{1/(k+1)}$

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## Summary

$\forall k \in \mathbb{Z}^+, \exists \frac{1}{k}$ -approx. data structure for dynamic MaxDep with  $\tilde{O} \left( n^{\frac{1}{k+1}} \right)$  update time

# **Approximate via Discretization**

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## DisMaxDepth

**Input:** a set of geometric objects  $\mathcal{O}$

a set of points  $\mathcal{P}$

**Output:** a point  $p^* \in \mathcal{P}$  s.t.

$$\text{dep}(p^*, \mathcal{O}) = \text{dep}(\mathcal{P}, \mathcal{O}) = \max_{p \in \mathcal{P}} \text{dep}(p, \mathcal{O})$$

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## Companion points

Insert/delete some “well-chosen” points  $\Gamma(O)$   
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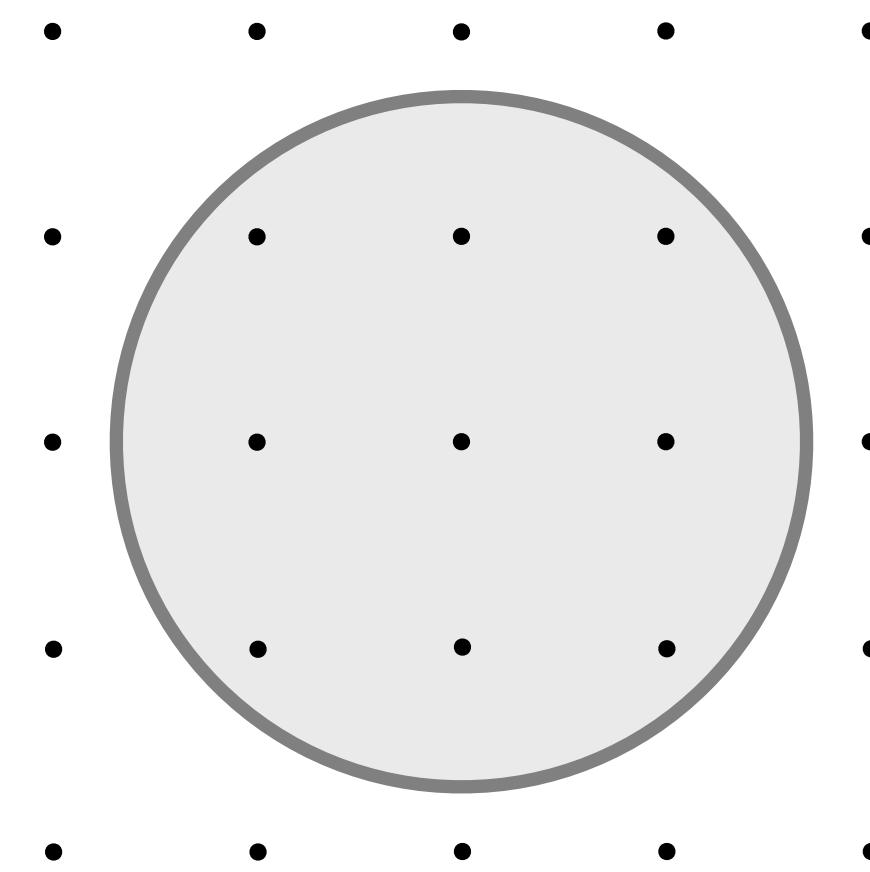
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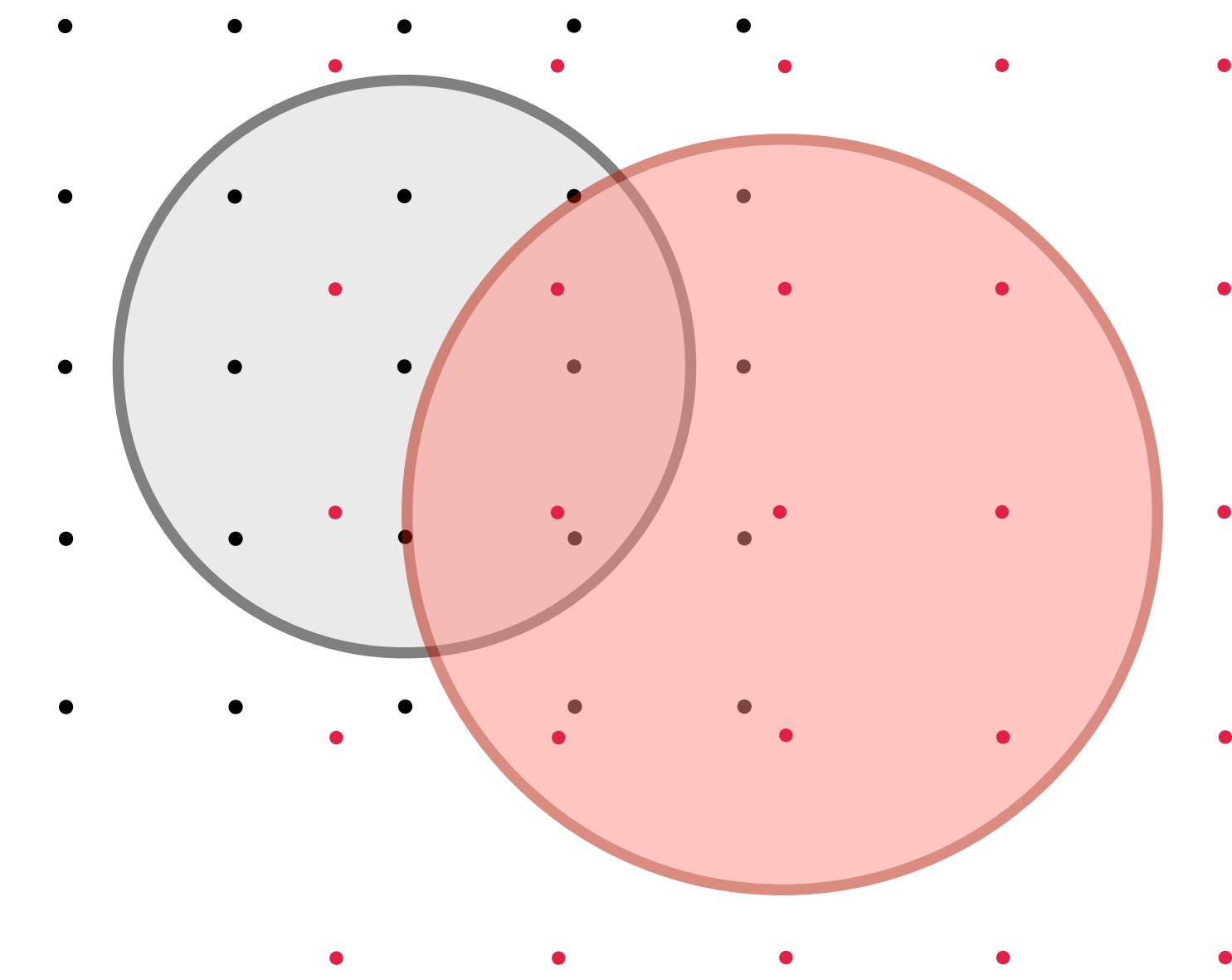
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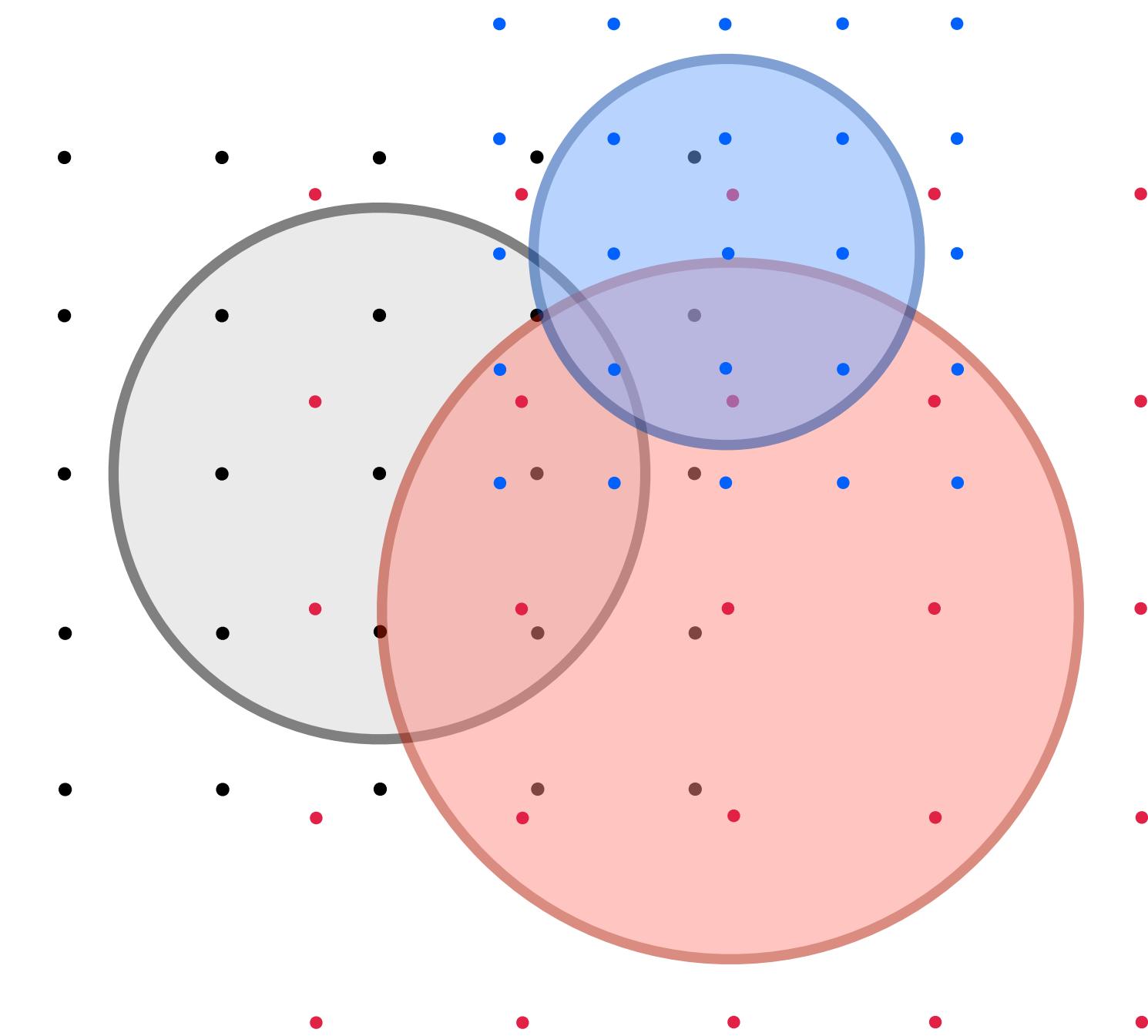
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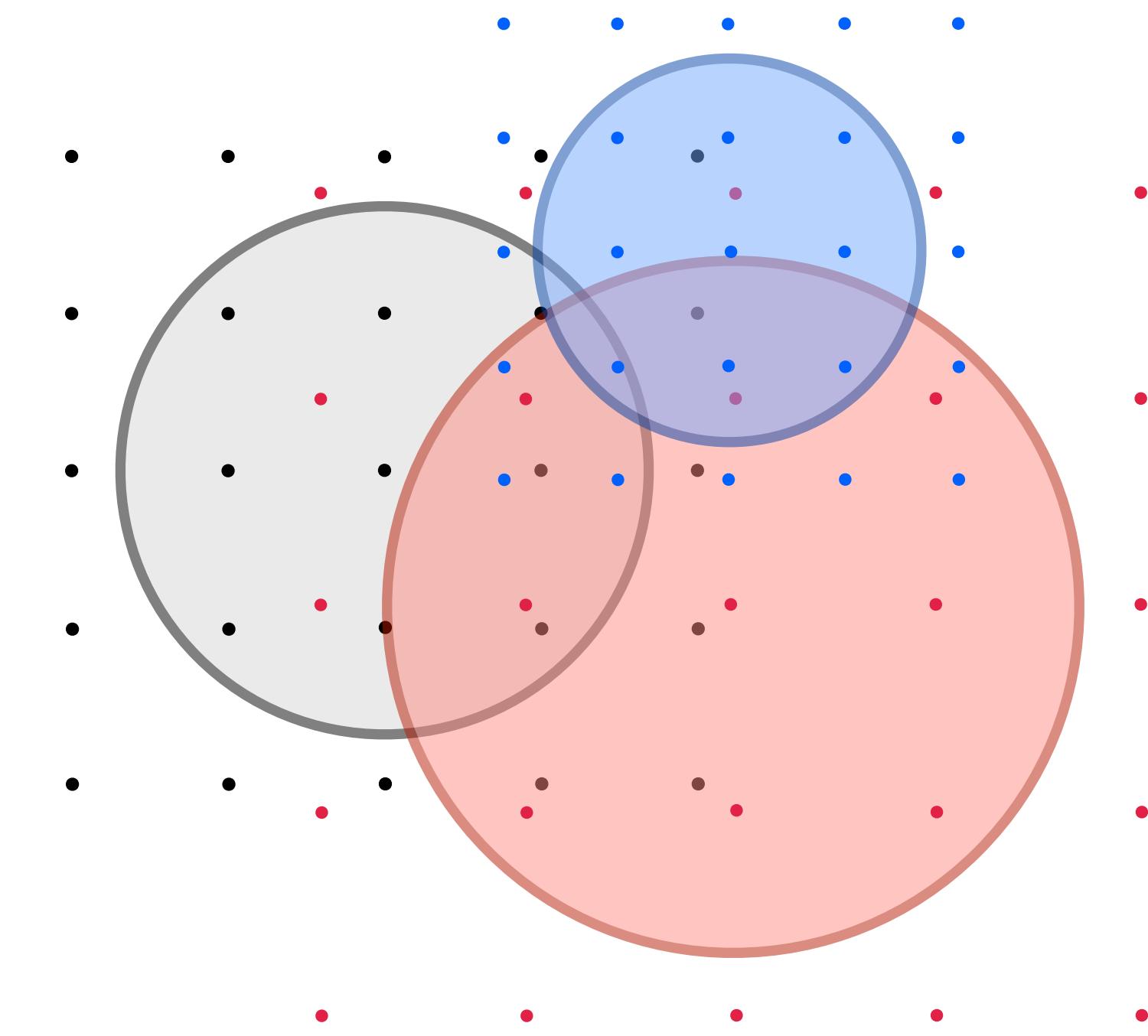
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Insert/delete some “well-chosen” points  $\Gamma(O)$  while insert or delete geometric object  $O$

## “Well-chosen”

Hopefully,  $\text{dep}(\Gamma, \mathcal{O}) \geq \frac{1}{k} \cdot \text{dep}(\mathcal{O})$

# **Approximation for Disks and Fat Objects**

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## Approx. for disks

$$\forall \varepsilon > 0, \exists \Gamma, \text{s.t. } \text{dep}(\Gamma, \mathcal{O}) \geq \left( \frac{1}{2} - \varepsilon \right) \text{dep}(\mathcal{O})$$

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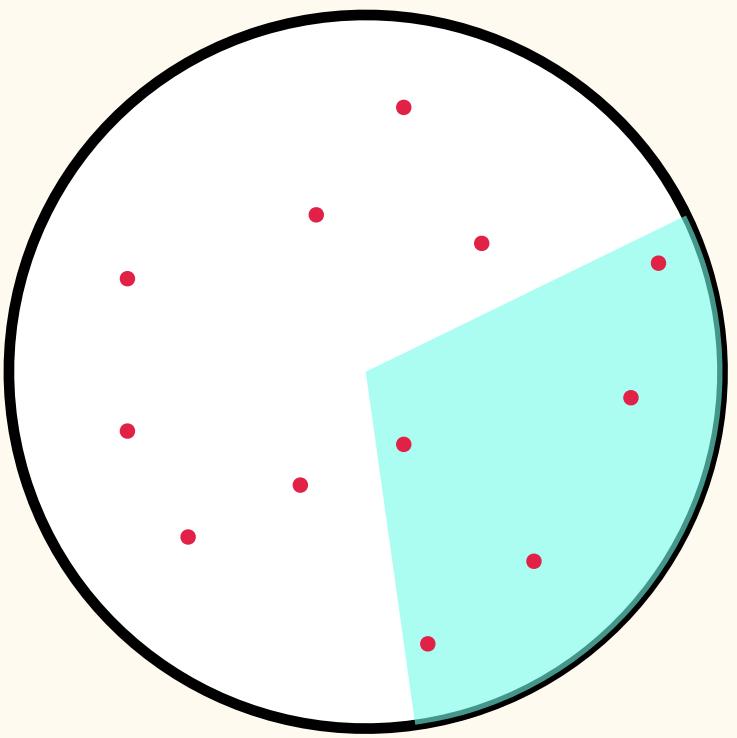
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## Source of approx. ratio

Given a disk  $D$  and a set  $S$  of points in  $D$ ,

$\exists$  circular sector  $X$  with angle  $(1 - 2\varepsilon)\pi$ ,

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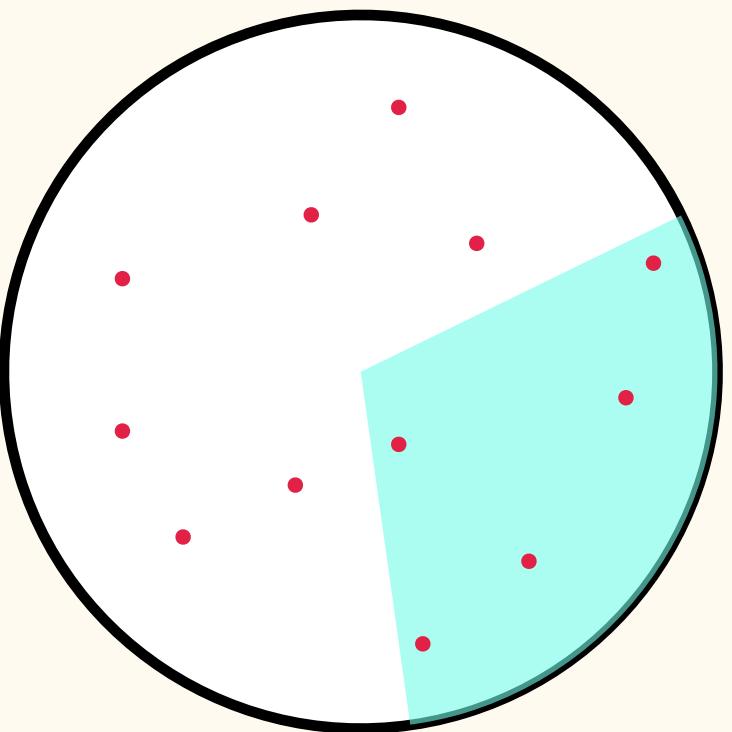
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## Data structure for disks

$\tilde{O}(n^{2/3})$  update time

via dynamic partition tree [Matoušek 1992]

# Approximation for Disks and Fat Objects

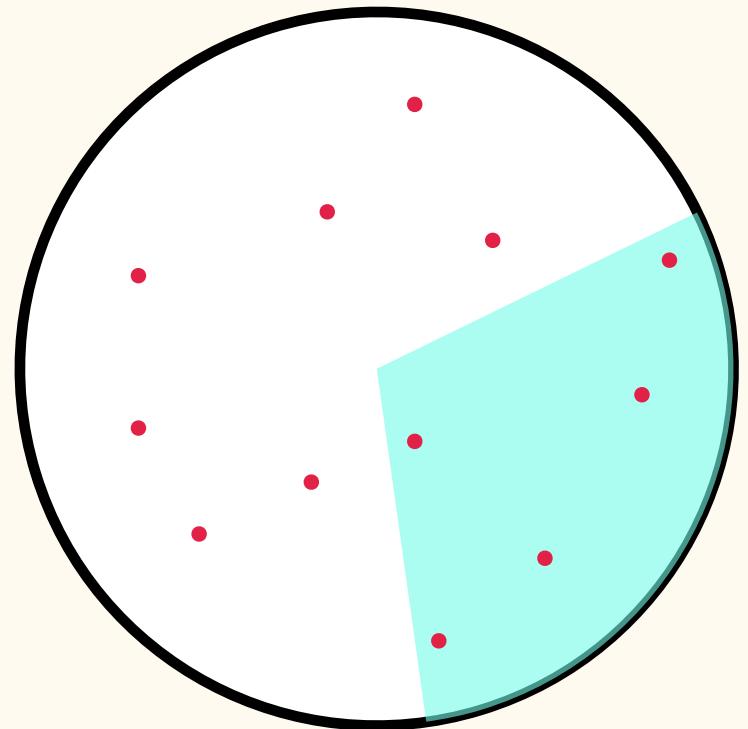
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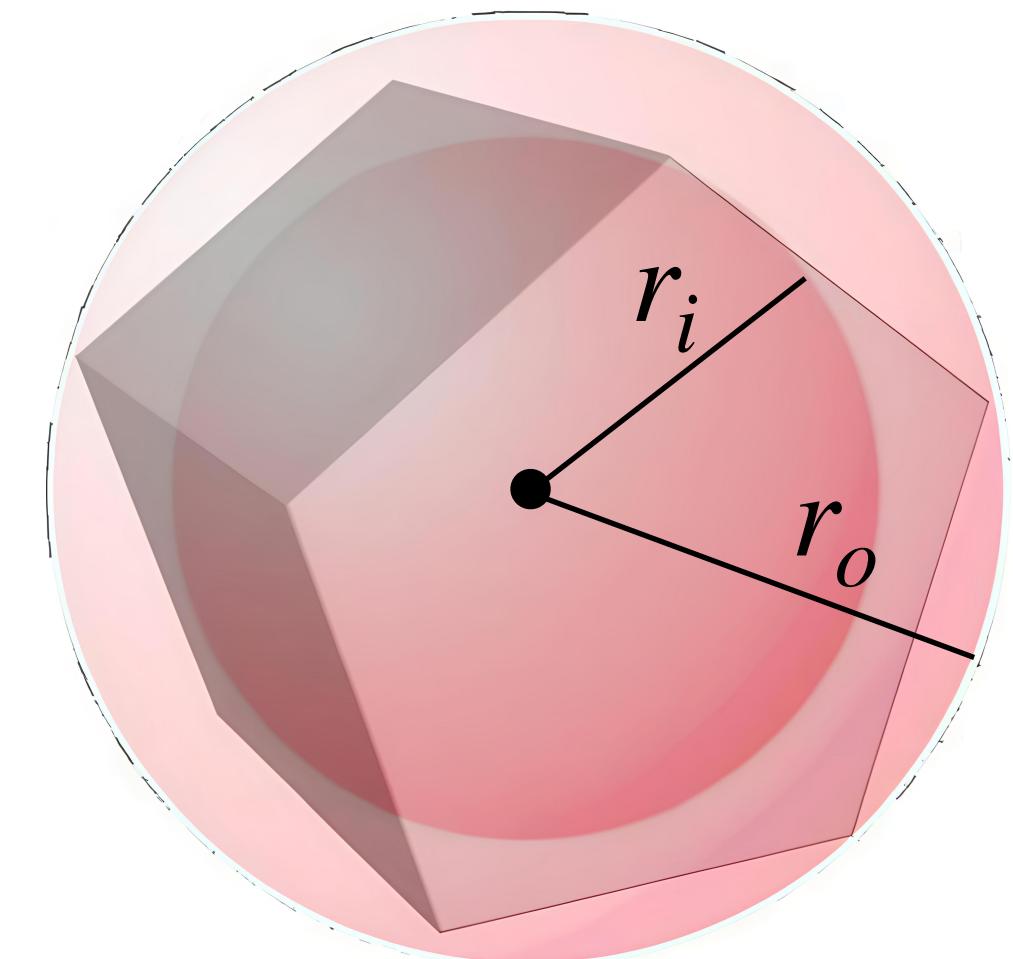
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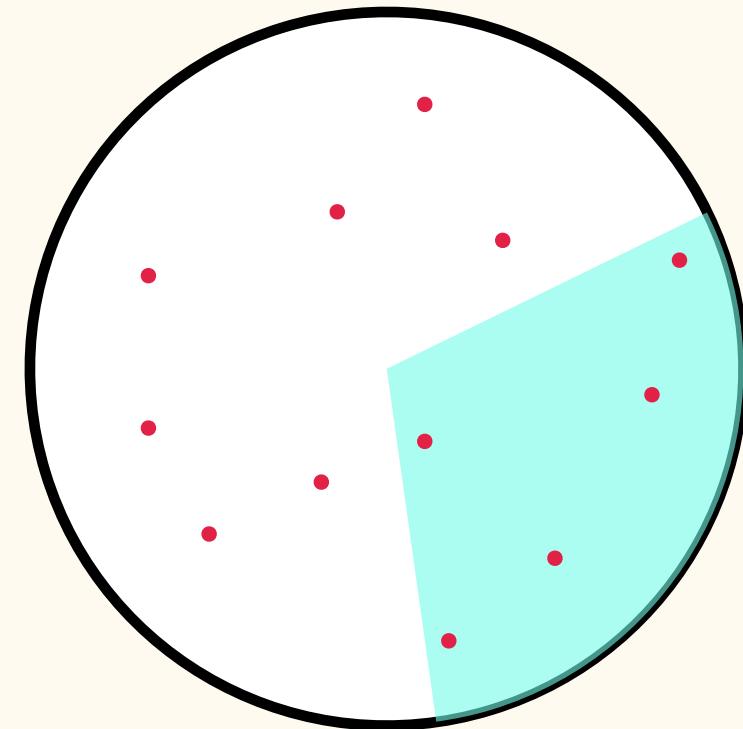
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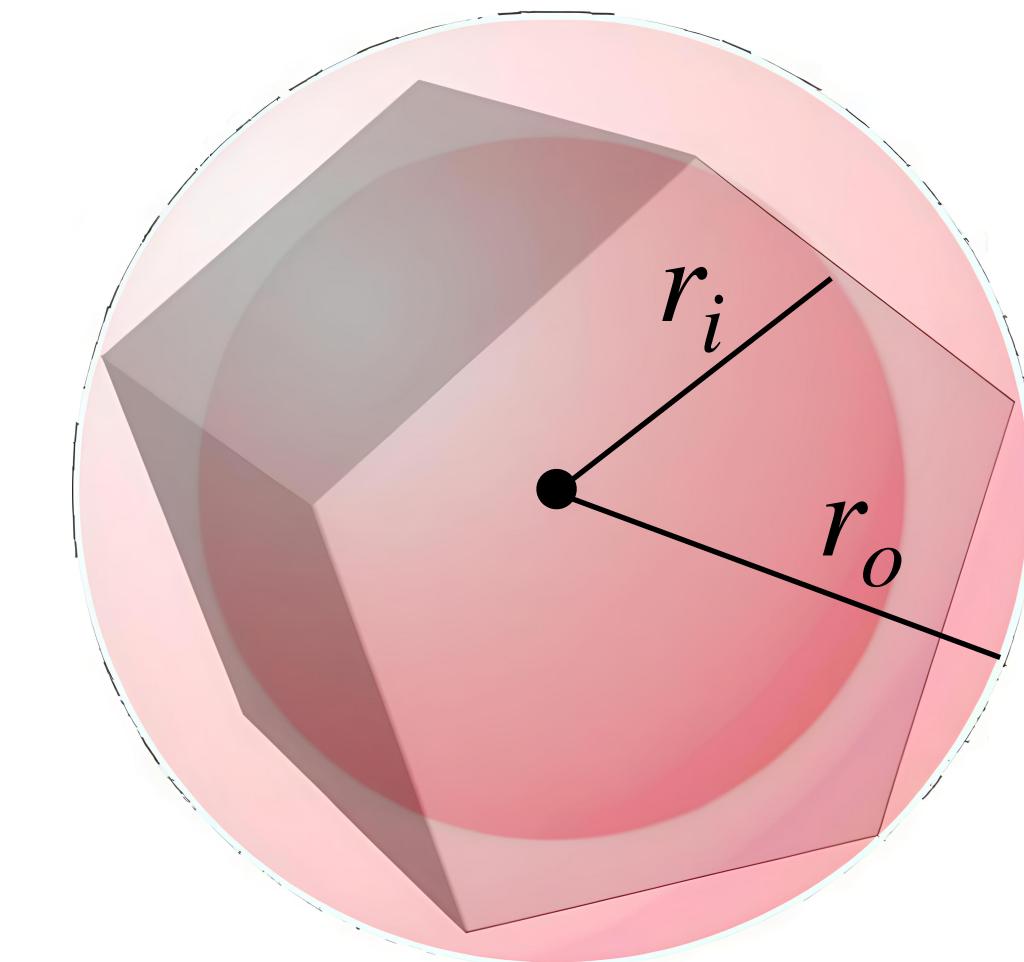
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## Strategy for fat objects (in $\mathbb{R}^d$ )

Reduce to “balls cover a ball” problem:  
If  $k$  unit balls cover a ball with radius  $2\alpha$ ,

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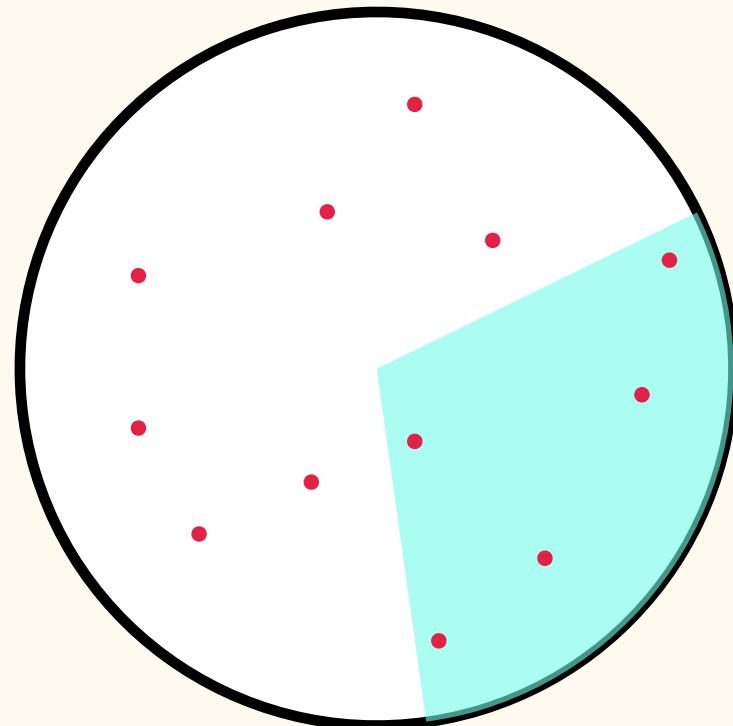
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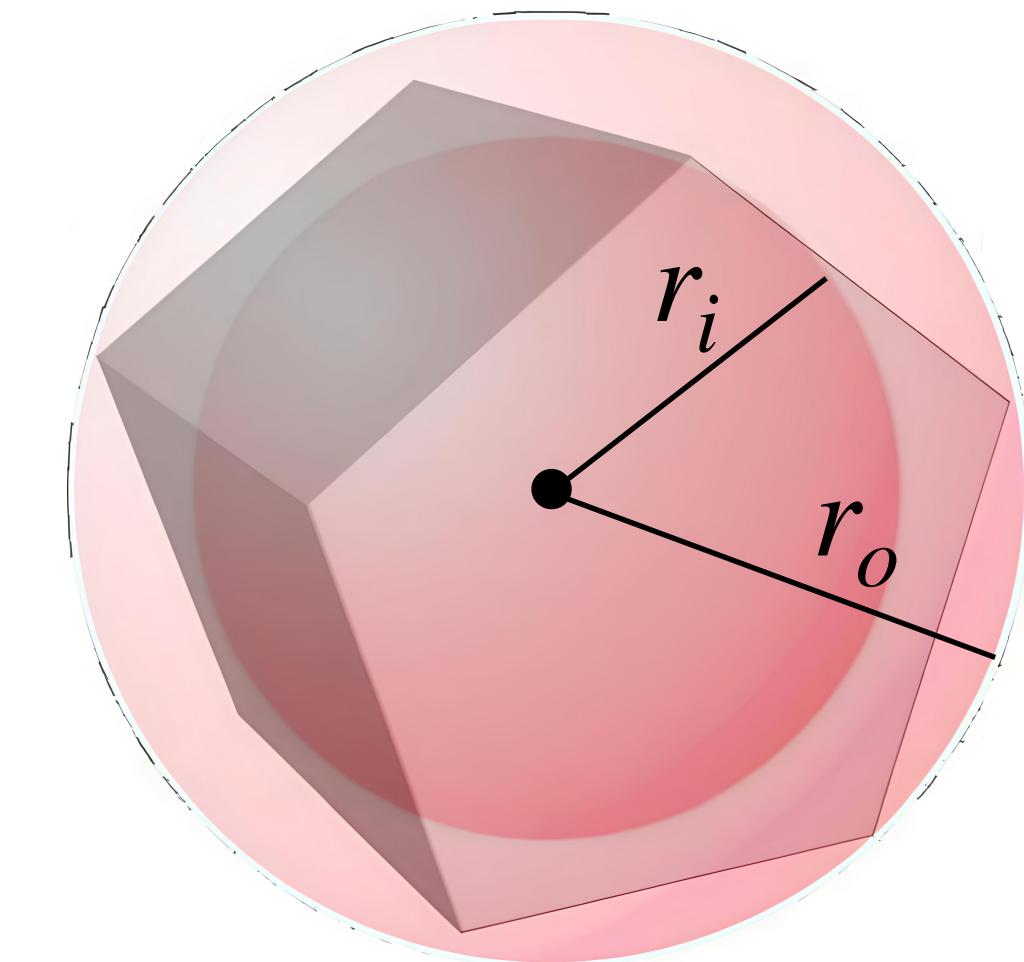


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$$\text{Trivial bound: } k \leq \left( 2 \lceil \sqrt{2}\alpha \rceil \right)^d$$

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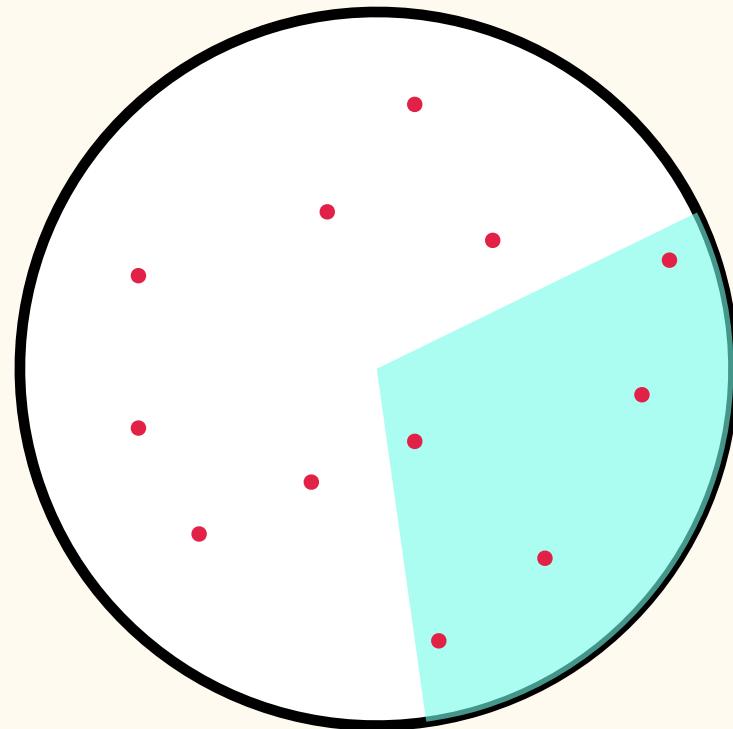
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Given a disk  $D$  and a set  $S$  of points in  $D$ ,  
 $\exists$  circular sector  $X$  with angle  $(1 - 2\varepsilon)\pi$ ,

$$\text{s.t. } |S \cap X| \geq \left( \frac{1}{2} - \varepsilon \right) |S|$$



## Data structure for disks

$\tilde{O}(n^{2/3})$  update time  
via dynamic partition tree [Matoušek 1992]

$\alpha$ -fat object:

$$\frac{r_o}{r_i} \leq \alpha$$

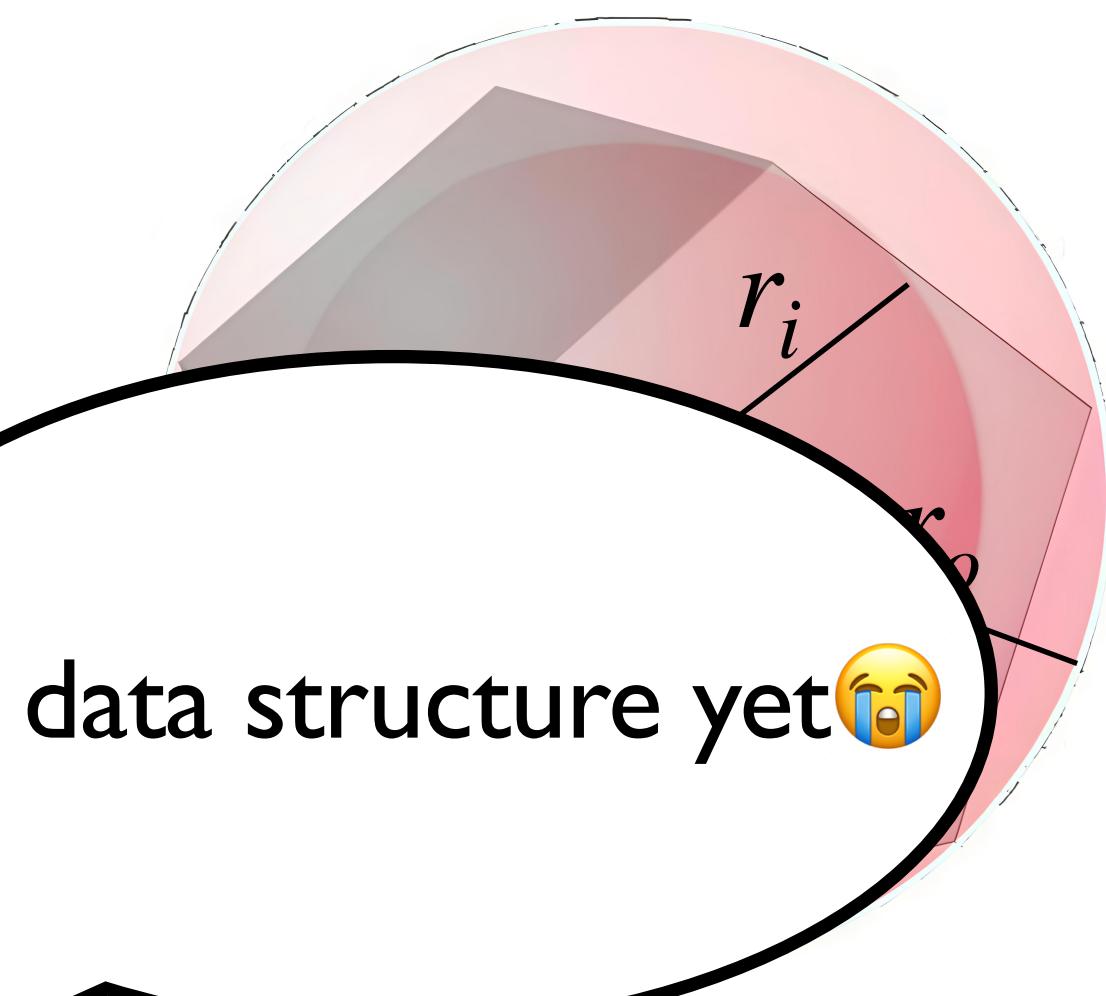
No data structure yet 😱

## Strategy for fat objects (in $\mathbb{R}^d$ )

Reduce to “balls cover a ball” problem:  
If  $k$  unit balls cover a ball with radius  $2\alpha$ ,

$$\text{then } \text{dep}(\Gamma, \mathcal{O}) \geq \frac{1}{k} \text{dep}(\mathcal{O})$$

$$\text{Trivial bound: } k \leq \left( 2 \lceil \sqrt{2}\alpha \rceil \right)^d$$



# Thank you!

**Any Questions?**

## Open Problems

- Data structure for  $d$ -dim. boxes with  $\tilde{O}(n^{(d-1)/2})$  update time
- Improve update time for special objects, e.g., squares
- Data structure for fat objects in discrete setting
- Data structure for other objects, e.g., halfspaces, quadrants