

Lecture 8

Divide and conquer

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Principles of divide and conquer

- Identity a division of the problem into a self-similar parts of size n/b
- Recursively solve each subpart of the problem
- Stitch the solutions from each subpart together
- Runtime is defined by the following recursively defined formula:

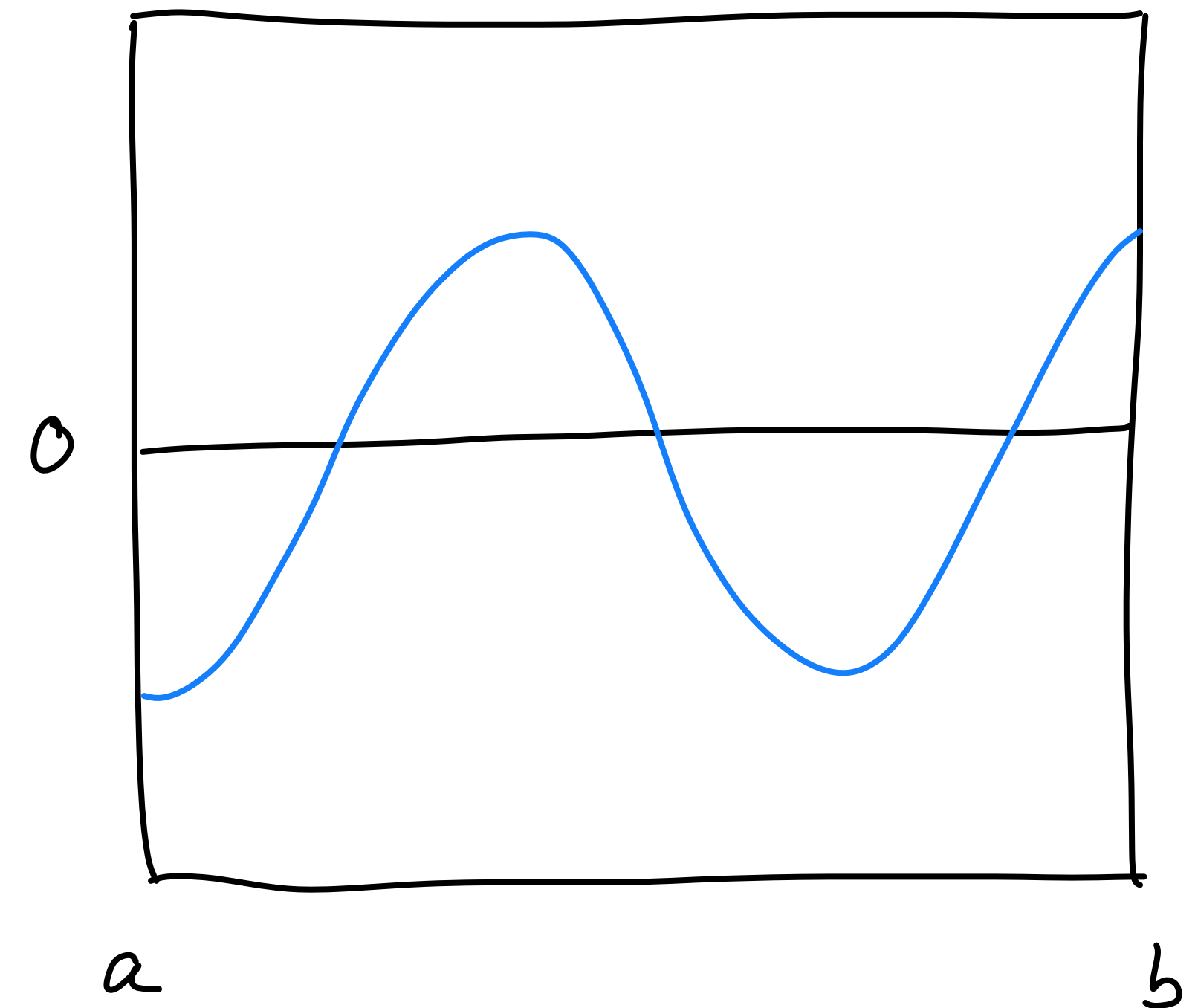
$$T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n) \text{ and } T(n < b) = O(1)$$

Examples of divide and conquer

- Mergesort, Quicksort (sort of)
- Binary search (today)
- Euclidean closest pair (today)
- Rank selection, Median finding

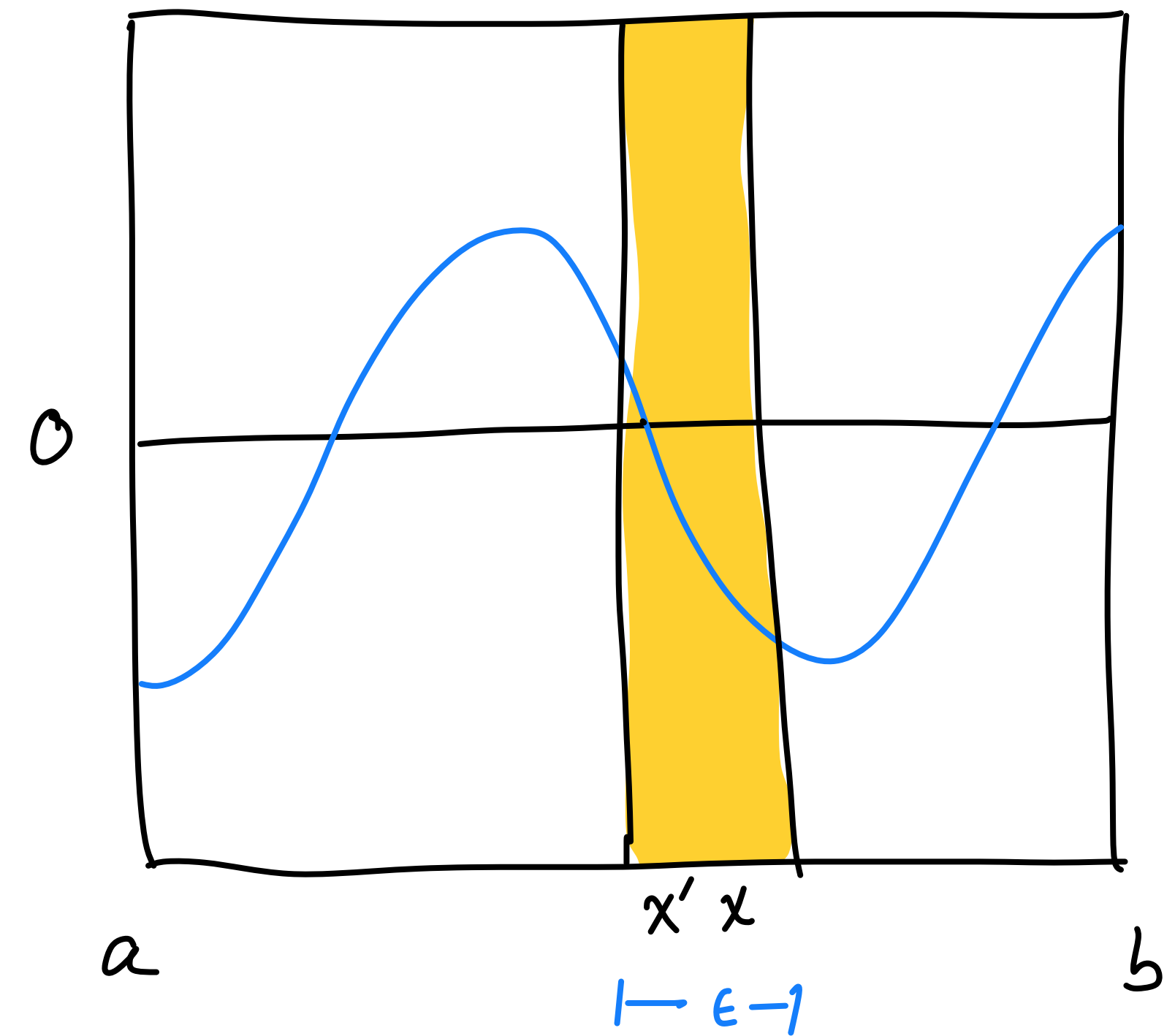
Binary search for roots of a function

- **Input:** Description of
 - a continuous $f : \mathbb{R} \rightarrow \mathbb{R}$,
 - $a < b \in \mathbb{R}$ such that $f(a) \leq 0 < f(b)$
 - and $\epsilon > 0$



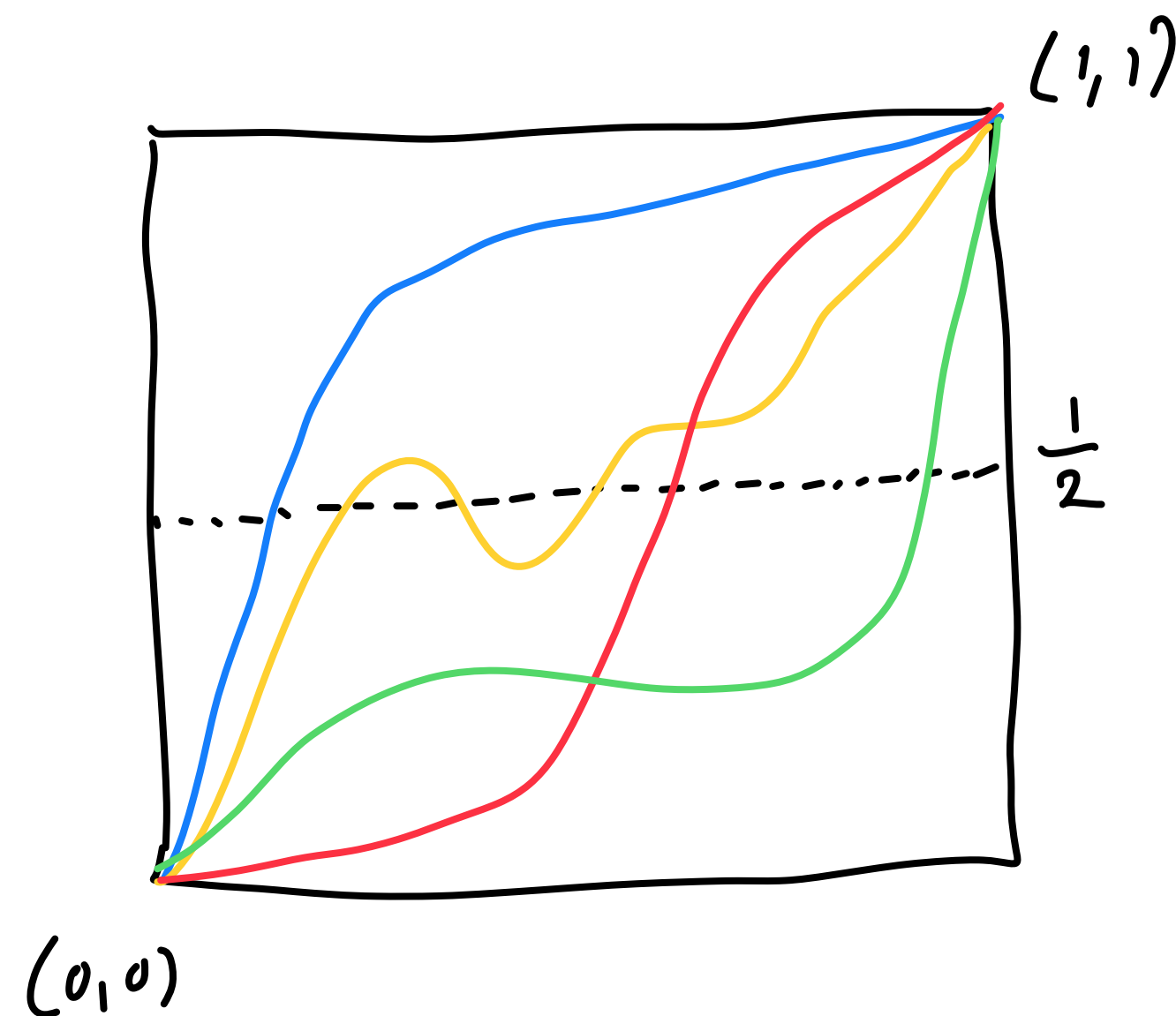
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 - a continuous $f : \mathbb{R} \rightarrow \mathbb{R}$,
 - $a < b \in \mathbb{R}$ such that $f(a) \leq 0 < f(b)$
 - and $\epsilon > 0$
- **Output:** A value $x \in [a, b]$ such that $f(x') = 0$ for some $|x' - x| \leq \epsilon$.



Bisection method

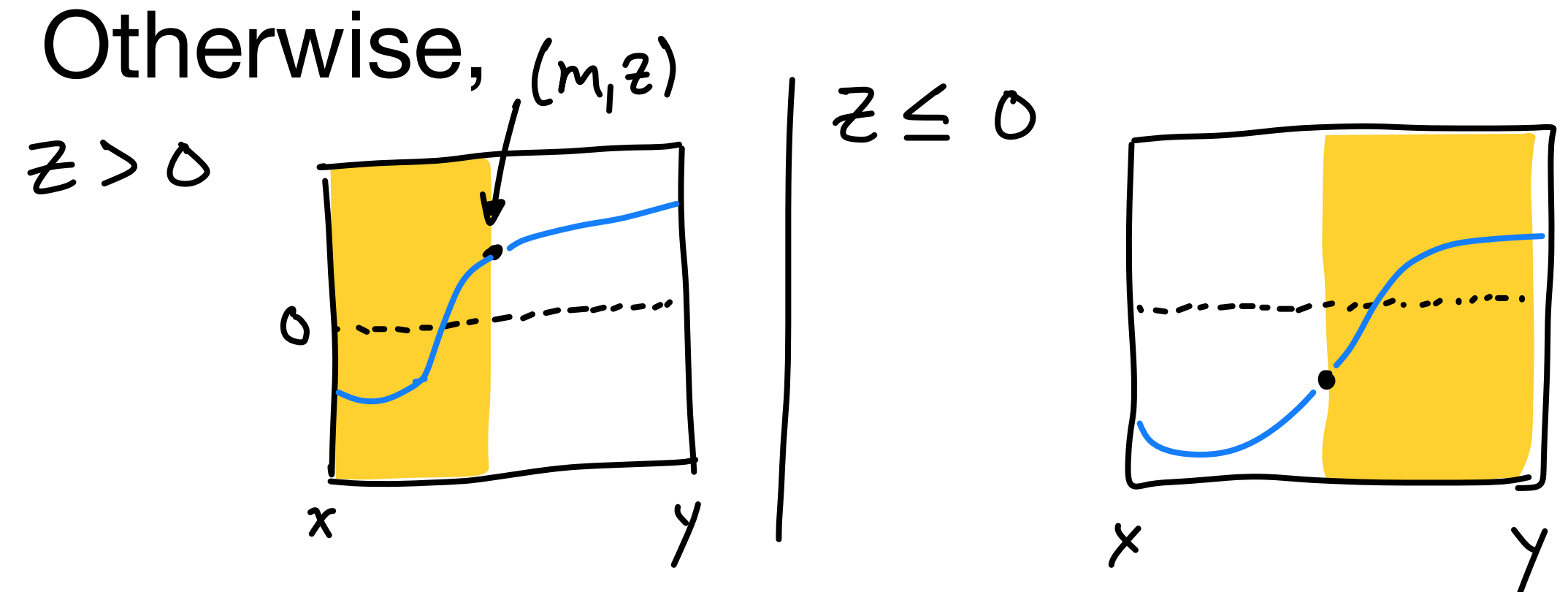
- **Intermediate value theorem (IVT):** If $f(0) = 0$, $f(1) = 1$ and f is continuous, there exists an $x \in (0,1)$ such that $f(x) = 1/2$.
- **Proof by picture:**



Any function must cross
the midline at some point
by continuity.

Bisection method

- **Algorithm** $g(x, y)$:
 - Let $m \leftarrow (x + y)/2$.
 - If $y - x \leq 2\epsilon$, return m .
 - Let $z \leftarrow f(m)$.
 - If $z > 0$, return $g(x, m)$.
 - Else, return $g(m, y)$.
- **Claim:** If $f(x) \leq 0 < f(y)$ for $x < y$, then $g(x, y)$ outputs an m such that $f(m') = 0$ for $|m' - m| \leq \epsilon$.
- **Proof:**
 - Base case, follows from IVT.
 - Otherwise,



Runtime analysis

Binary search problem

- Therefore, running $g(a, b)$ will solve the bisection problem.
- Each iteration of g is on an interval of half the length
 - starting from $b - a$ until the length is $\leq 2\epsilon$
 - Therefore, $\log_2 \left(\frac{b - a}{2\epsilon} \right)$ recursions.
- Each recursion costs $O(1)$ arithmetic operations plus 1 query to f .
- Runtime: $O(\log(b - a) + \log(1/\epsilon))$ queries to f .

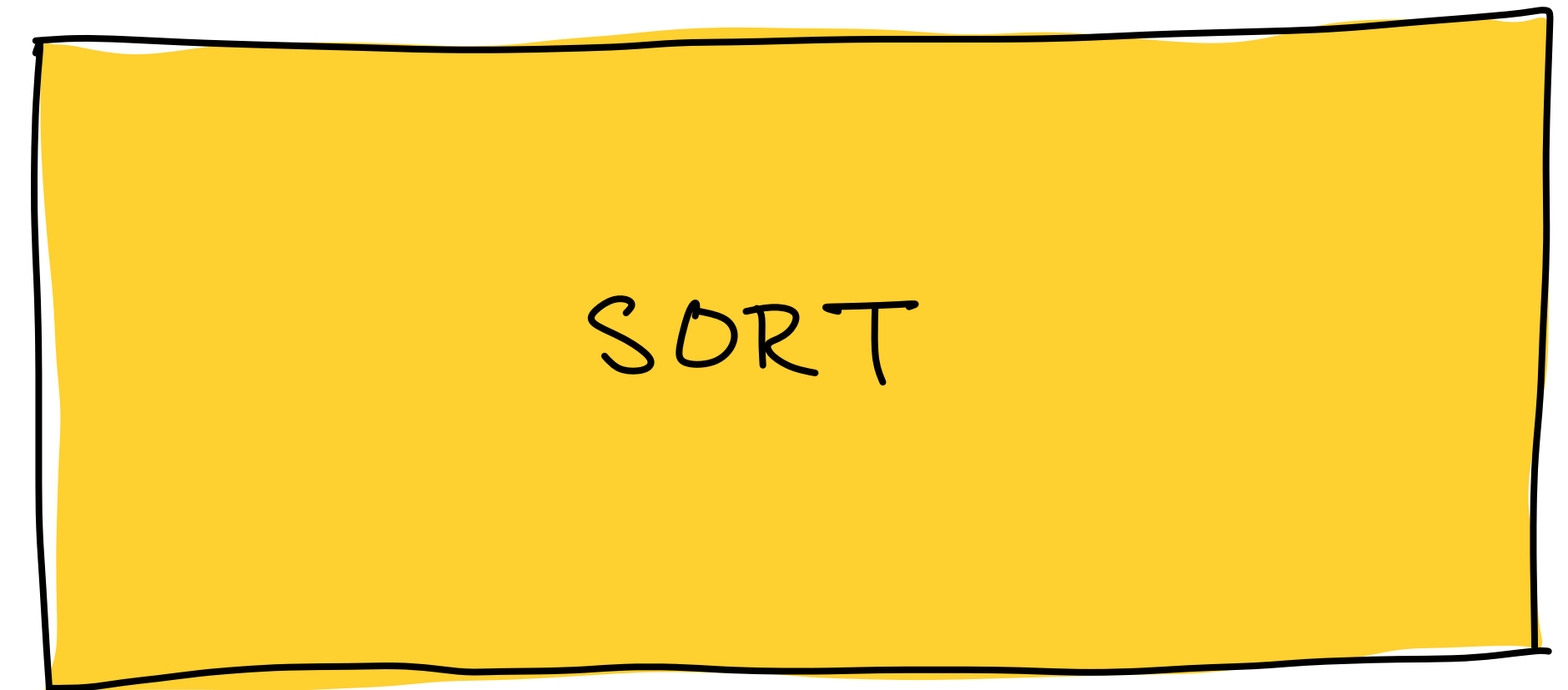
Runtime analysis

- Simple version of generalized runtime analysis.
- Let $k = (b - a)/(2\epsilon)$.
- Then, $T(k) = T(k/2) + 1$ and $T(1) = 0$ for number of queries.
- Solves to $T(k) = \lceil \log_2 k \rceil + 1$.

Another classic divide and conquer problem

Mergesort

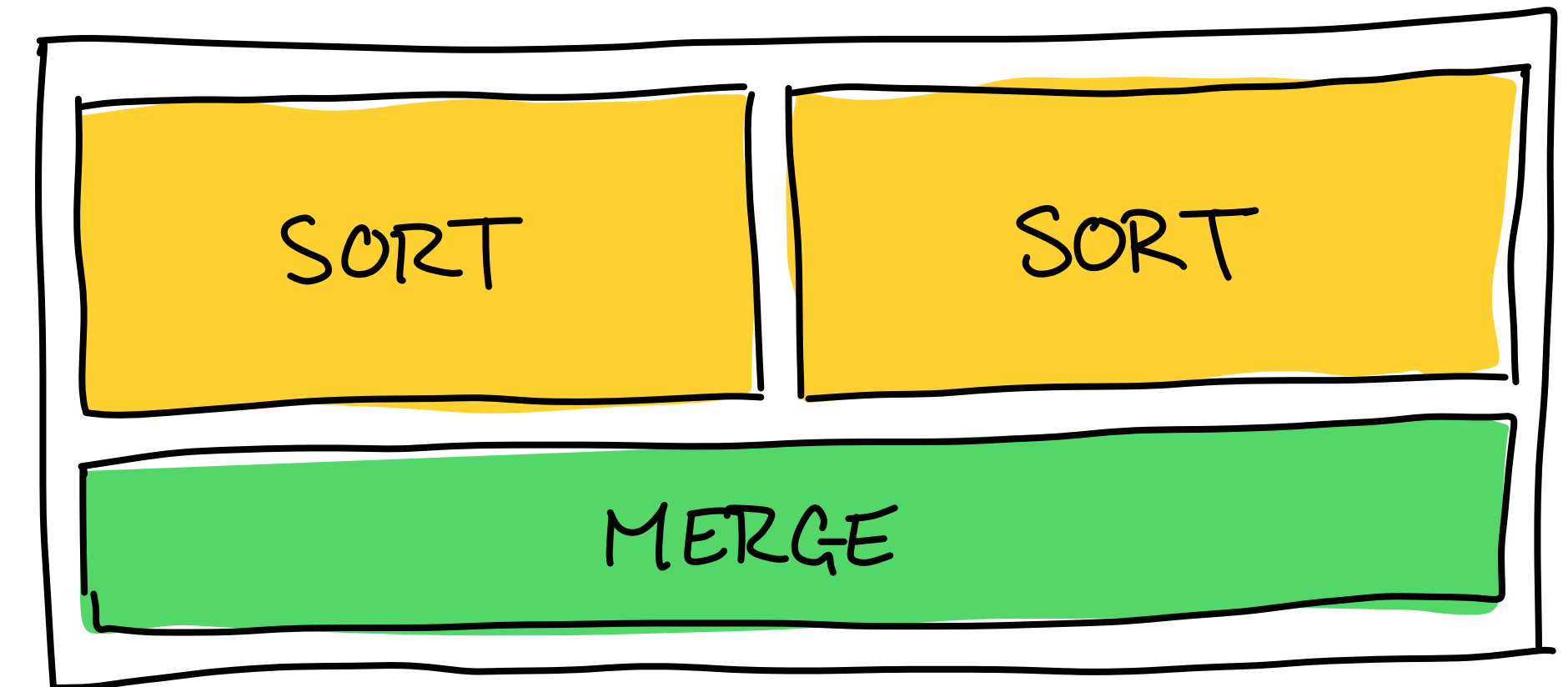
- To sort an array of n entries, recursively sort the first half and recursively sort the second half. Then *merge* the two sorted lists.
- Merging two sorted arrays takes $O(n)$ time as we only have to compare current elements as we iterate through both arrays
- Recursive time equation:
 $T(n) \leq 2T(n/2) + O(n)$ with $T(1) = 0$.
- Solution: $T(n) \leq O(n \log n)$



Another classic divide and conquer problem

Mergesort

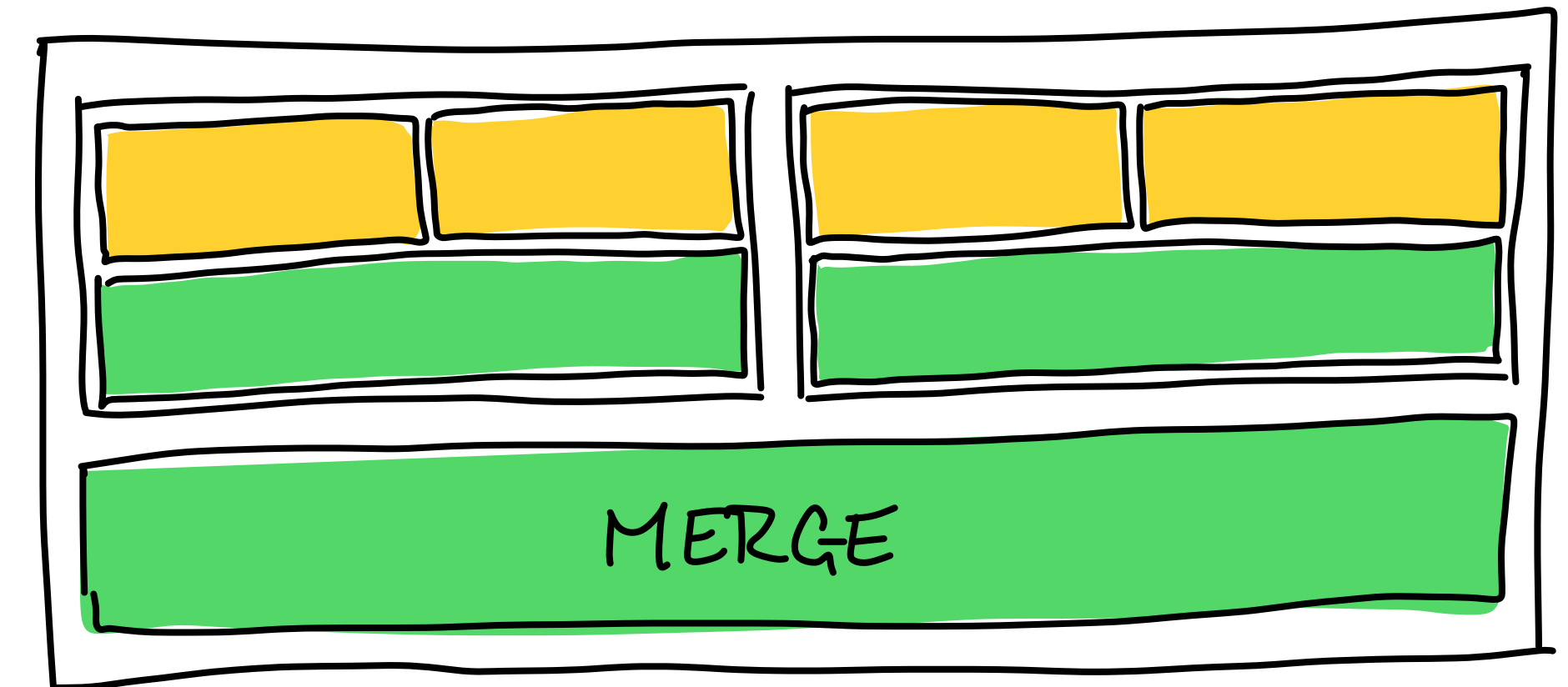
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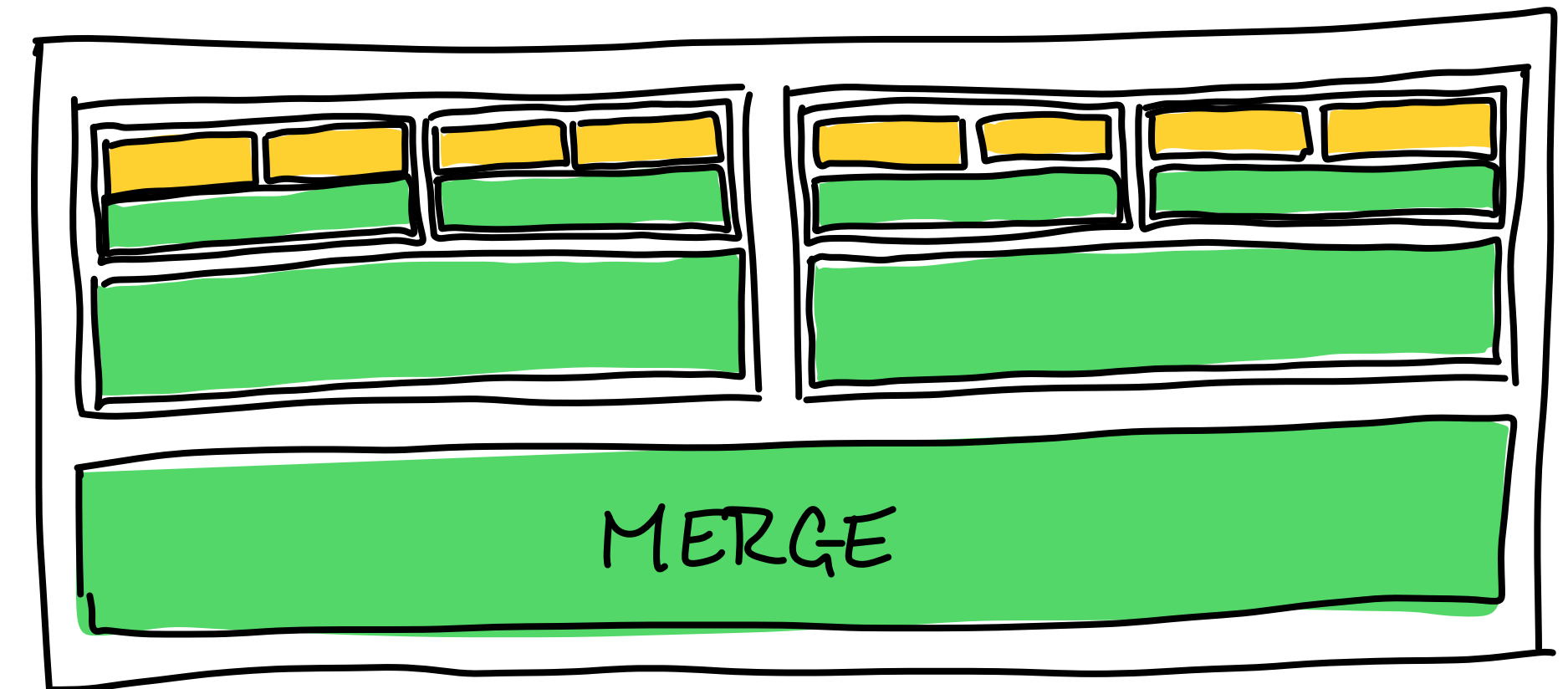
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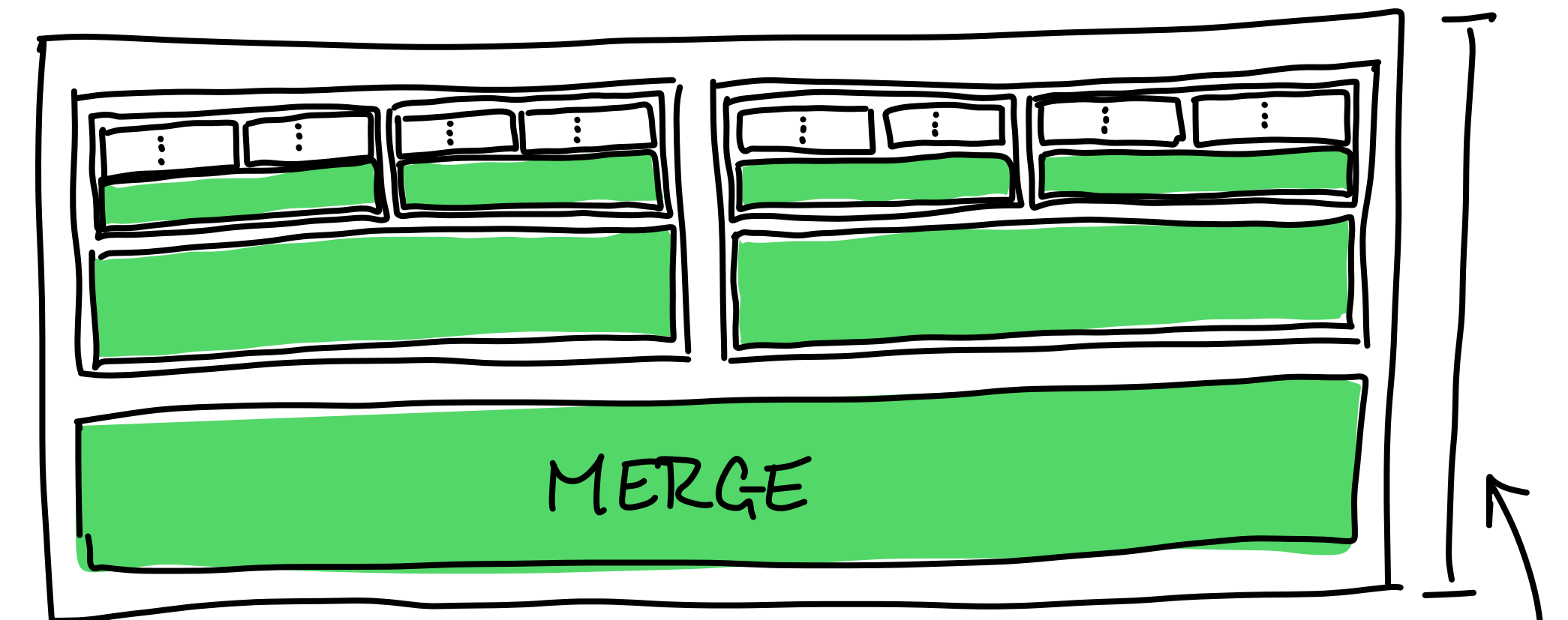
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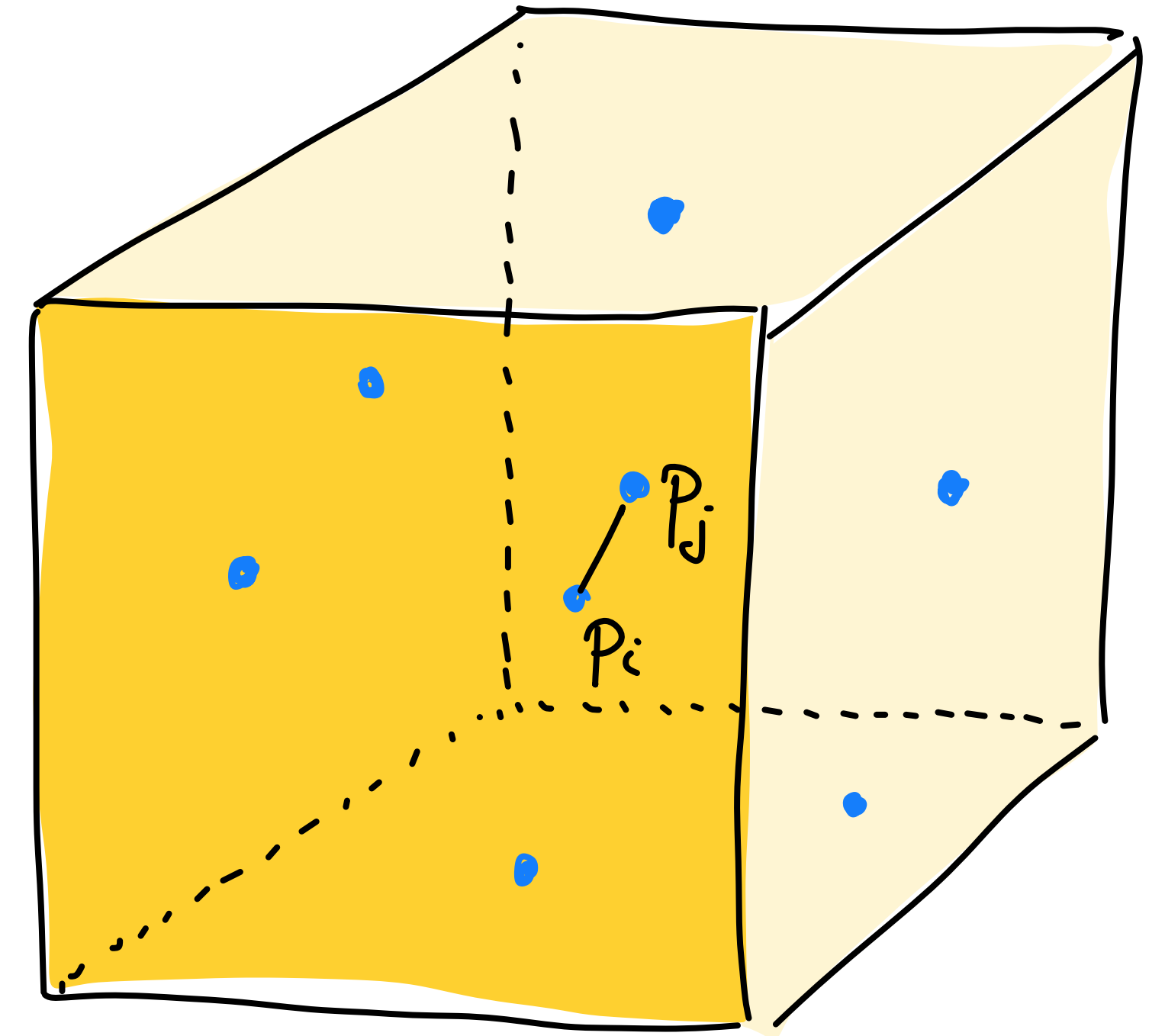


Total compute: $O(n \log n)$

$O(\log n)$
height

Euclidean closest pair

- **Input:** A sequence of n points $p_1, \dots, p_n \in \mathbb{R}^d$
- **Find:** The pair p_i, p_j minimizing $\|p_i - p_j\|$.
- **Brute force algorithm:** Try all pairs. $O(n^2d)$ time.
- Is there a better algorithm for small d ?
 - In 1D for example, we can sort and then compare nearest neighbors for $O(n \log n)$.
 - Can we do better?



2D Euclidean closest pair

- Sorting on first coordinate will not work
- No single direction for sorting guarantees success
- **Divide and conquer algorithm:**
 - Need to figure out a way to subdivide the problem
 - Then build solution from best solutions to both halves. This will require extra processing

B • (1, 10)

A • (0, 0)
C • (7, 2)

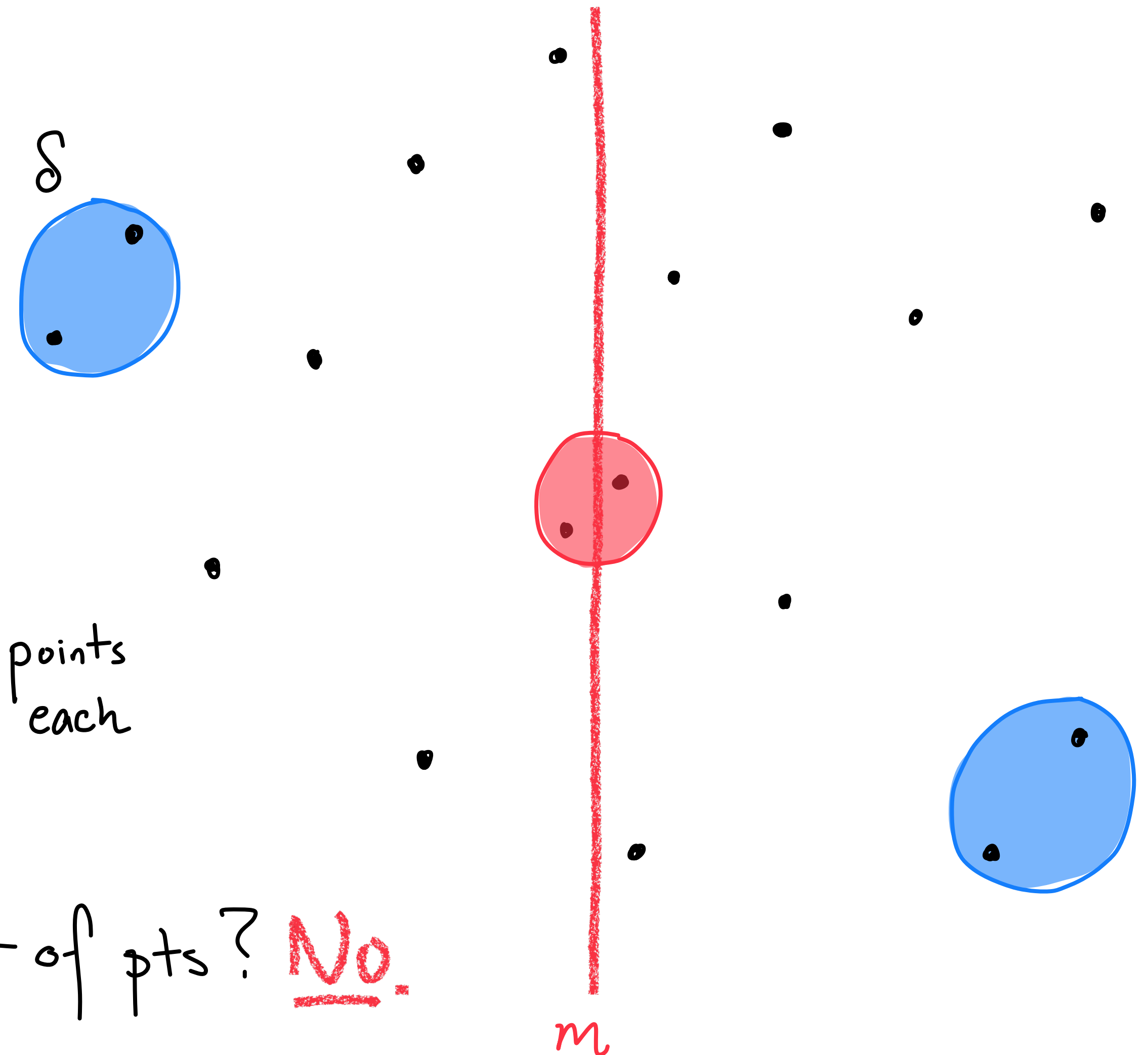
Sorting gives
A, B, C while shortest
pair is A - C.

Split across x -coordinate anyways

- Let's split according to x -coordinate anyways
- Let m be the median x -coordinate
- Divide the set into points $\{p : p_1 \leq m\}$ and $\{p : p > m\}$
- Let δ be the minimum of the two solutions

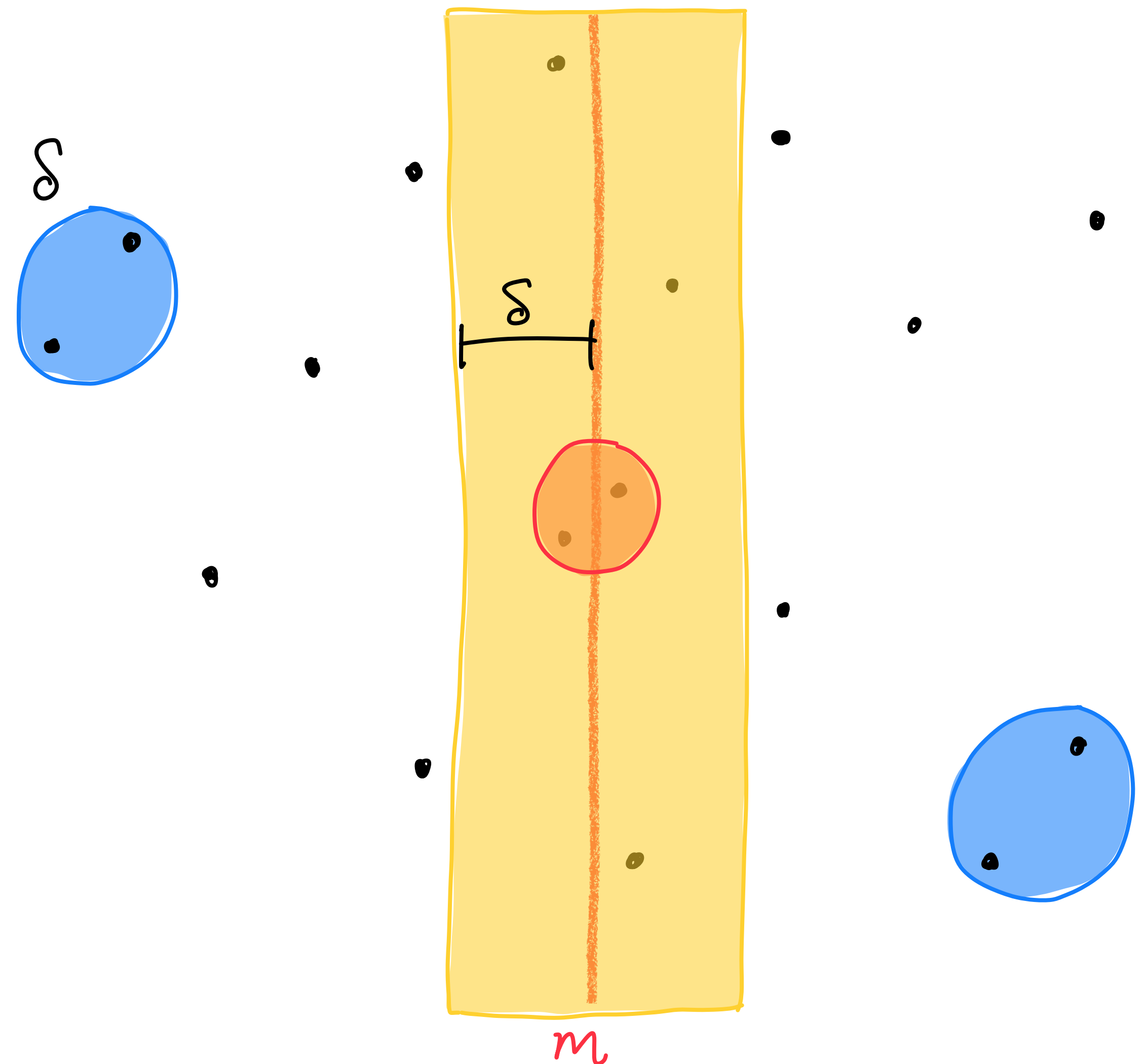
$\frac{n}{2}$ points each

Is this guaranteed to be the closest set of pts? No.



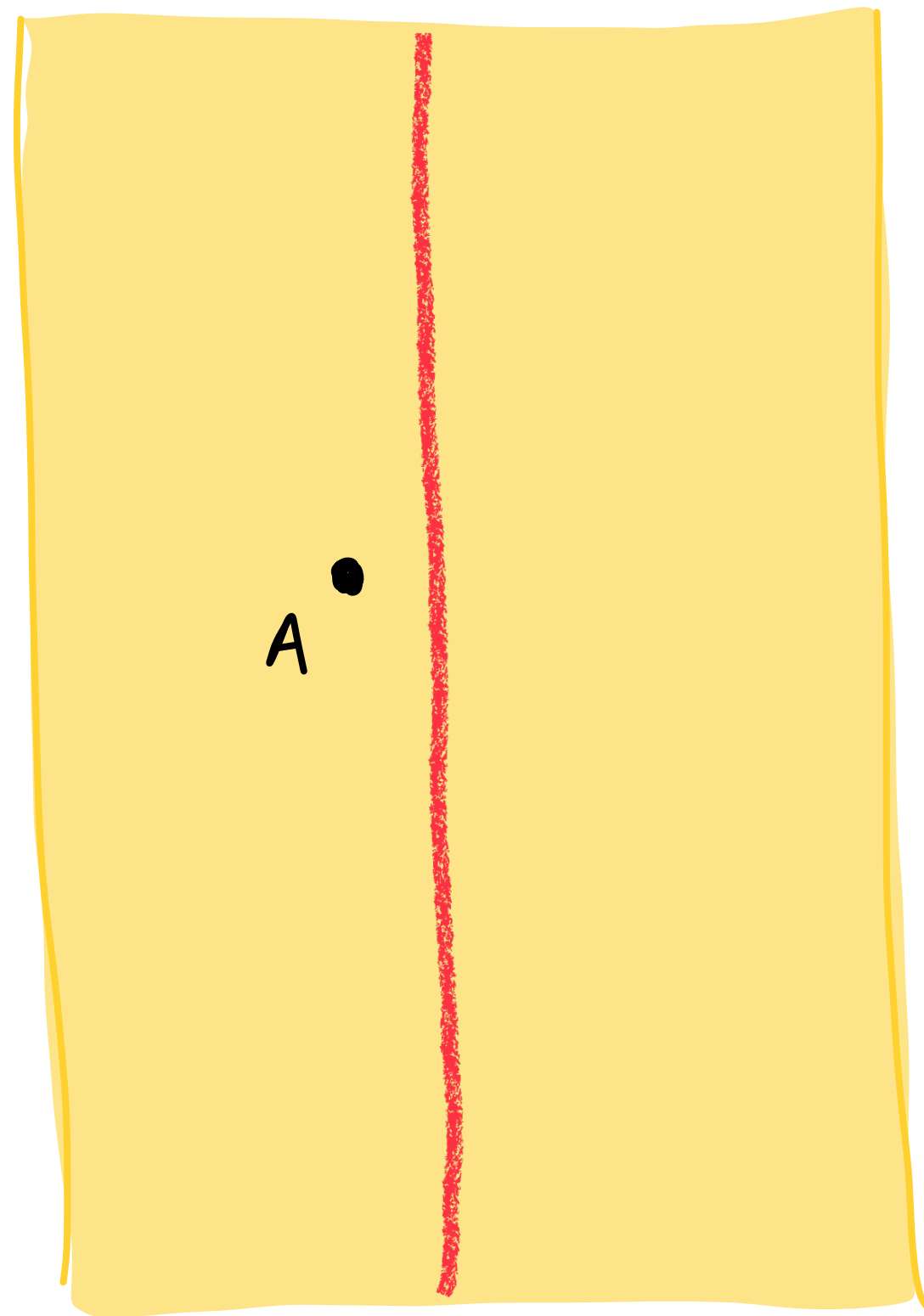
The “conquer” aspect of the algorithm

- We only need to worry about pairs that are **both** split by the median and $< \delta$ distance apart
- During “conquer” step, only need to look at vertices in the δ -width band
- Within the band, only need to compare points with y -coordinates that differ by $< \delta$



Close-up analysis

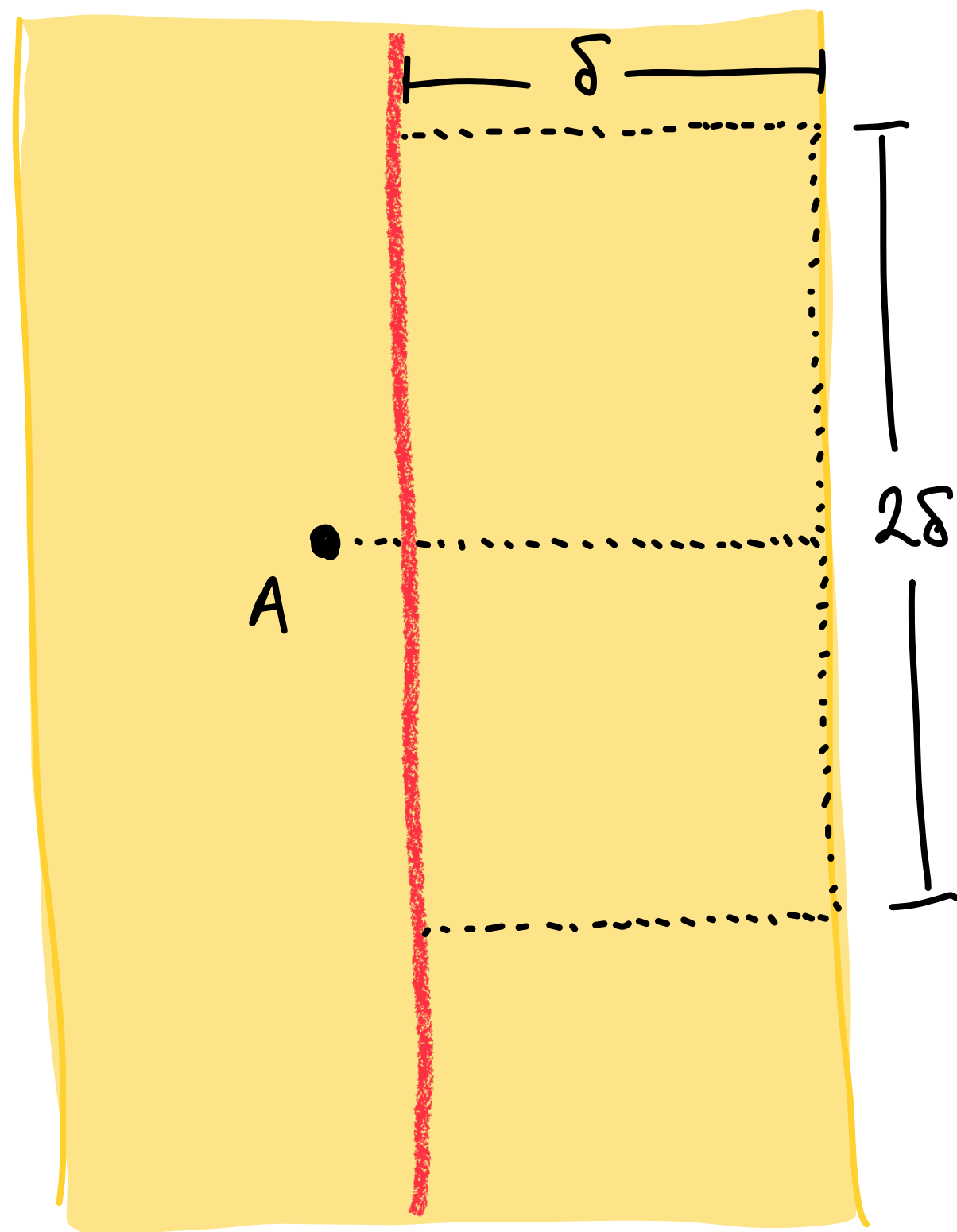
How many pts across the median do we have to compare with A?



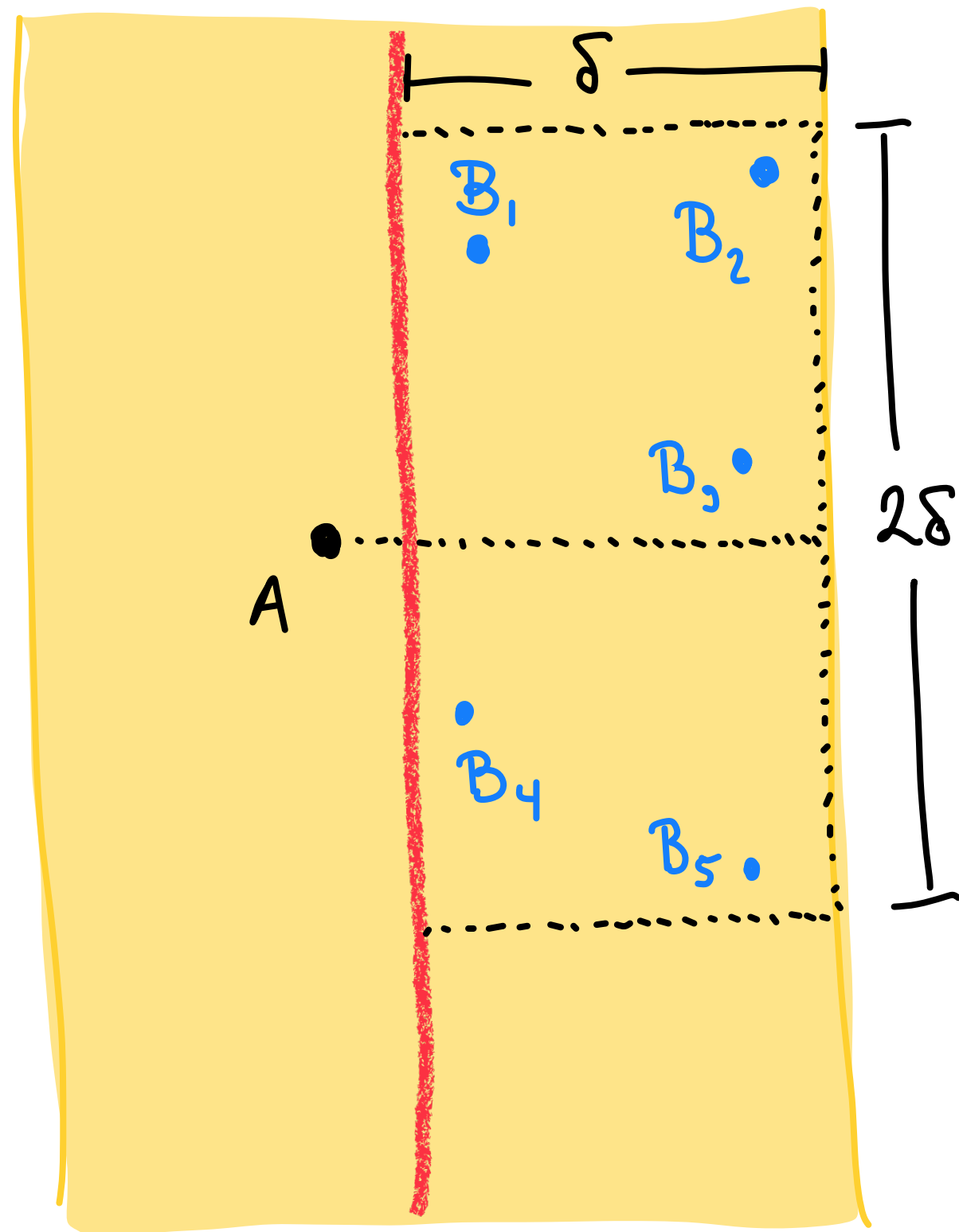
Close-up analysis

How many pts across the median do we have to compare with A ?

In order to be distance $\leq \delta$ from A ,
a pt B must lie in this box.



Close-up analysis



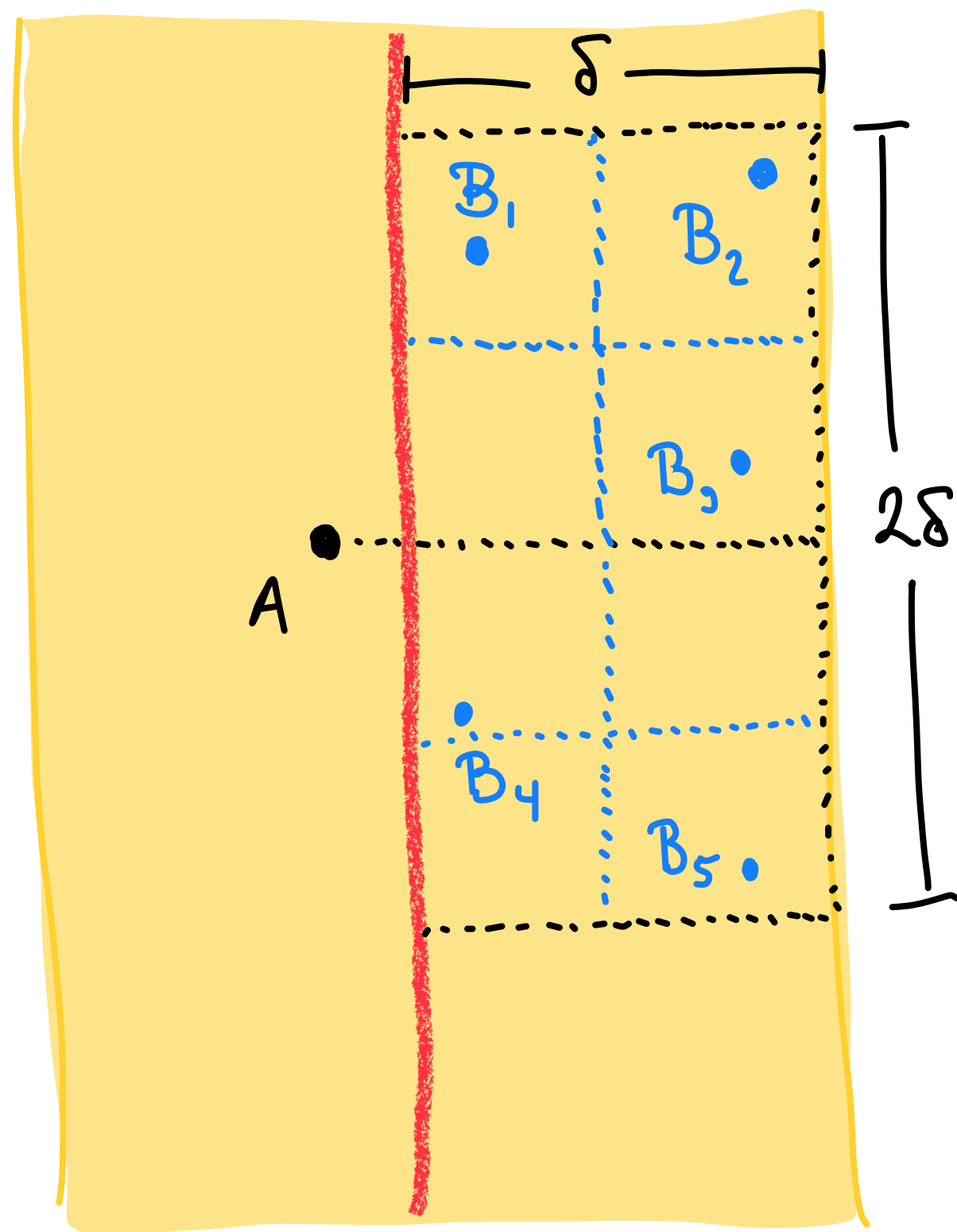
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In order to be distance $\leq \delta$ from A , a pt B must lie in this box.

How many such points B_i can exist?

Note: $\|B_i - B_j\| \geq \delta$ since on the same side.

Close-up analysis



How many pts across the median do we have to compare with A?

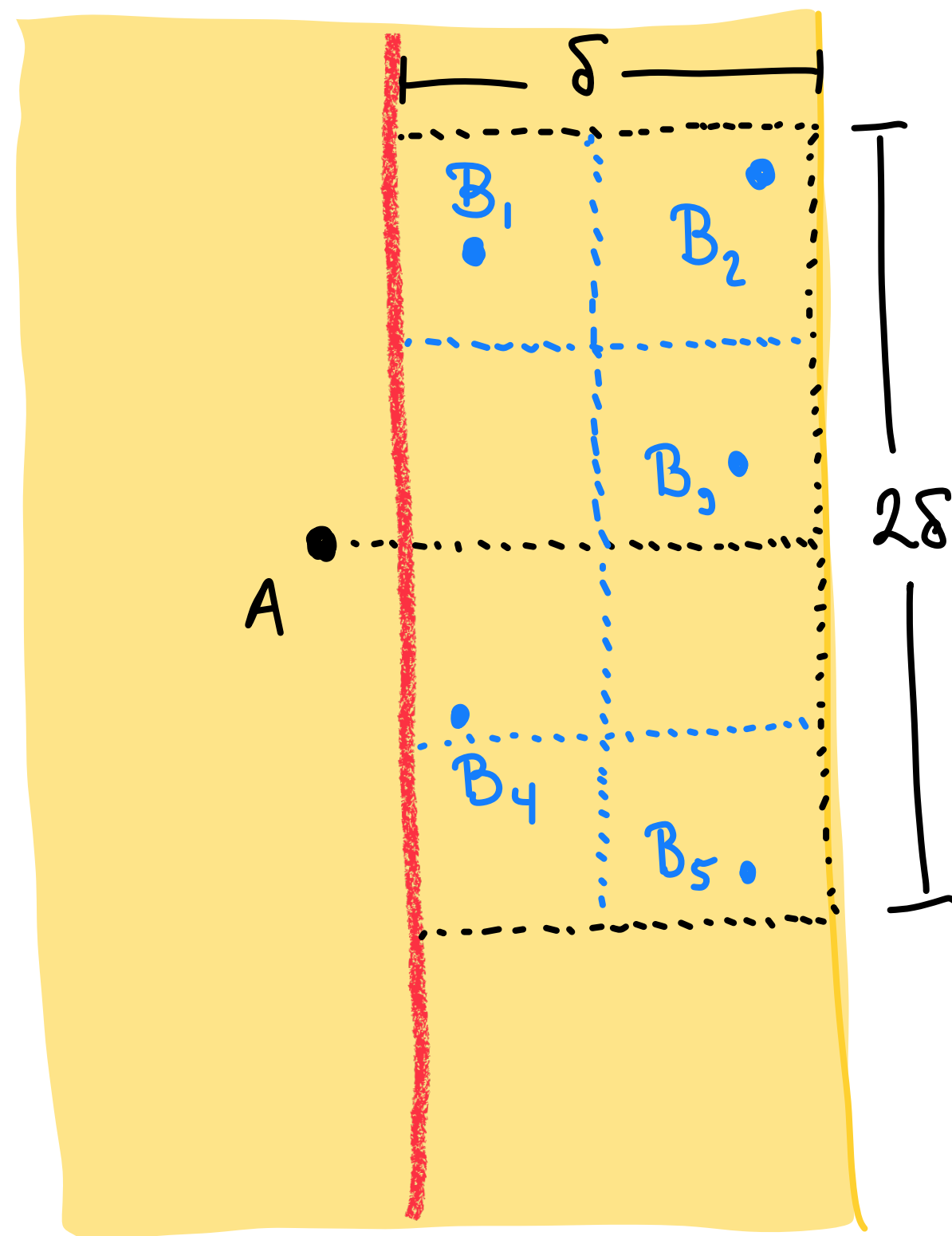
In order to be distance $\leq \delta$ from A, a pt B must lie in this box.

How many such points B_i can exist?

Note: $\|B_i - B_j\| \geq \delta$ since on the same side.

Each $\frac{\delta}{2} \times \frac{\delta}{2}$ box can have at most 1 point B_i . So at most 8 points.

The full conquer subroutine



- Let M be the set of points in band.
- Sort the points in M by y -coordinate
- For each point $a \in M$, compare a to the 8 points before and 8 points after a in the sorting.
- By analysis, this checks all possible pairs of distance $< \delta$.

Divide and conquer algorithm

$$\text{Total: } T(n) = 2T\left(\frac{n}{2}\right) + O(n \log n)$$

- **Divide step:**

- Compute median m and divide into two subproblems $\leftarrow O(n \log n)$ time with sorting
- Recursively calculate shortest distance for subproblems $\leftarrow 2T\left(\frac{n}{2}\right)$ by recursion

- **Conquer step:**

- Compute the set of points in the band $M \leftarrow O(1)$ time since we sorted for the median
- Sort M by y -coordinate $\leftarrow O(n \log n)$ time as potentially n points in band.
- Compare points in sorted M with the next 8 points and update if closer pair found.

$\uparrow O(n)$ time since sorted.

Better divide and conquer algorithm

- **Preprocessing:**

- Sort points according to x -coordinate for list X
 - Sort points according to y -coordinate for list Y
- $\left. \begin{array}{l} \text{Sort } X \\ \text{Sort } Y \end{array} \right\} O(n \log n) \text{ time. Only once.}$

- **Divide step** (sorted lists X, Y):

- Compute median m by x -coordinate $\leftarrow O(1)$ time since sorted
- Divide X into X_L, X_R . Filter Y into Y_L and Y_R . $\leftarrow O(n)$ time, once-through the list
- Recursively solve (X_L, Y_L) and (X_R, Y_R) problems for $\delta \leftarrow 2 T(\frac{n}{2})$ by recursion

- **Conquer step:**

- Filter Y into the band M of x -coordinates $m \pm \delta \leftarrow O(n)$ time
- Compare M to the next 8 points and update if closer point is found. $\leftarrow O(n)$ time

Total time:

$$T(n) = 2T\left(\frac{n}{2}\right) + O(n)$$

$\Downarrow$

$$T(n) = O(n \log n)$$

Analysis divide and conquer runtimes

The master theorem

- For solving recursive equations of the form

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n) \text{ and } T(n < b) = O(1)$$

- Different cases based on how $f(n)$, a , and b compare:

Analysis divide and conquer runtimes

The master theorem

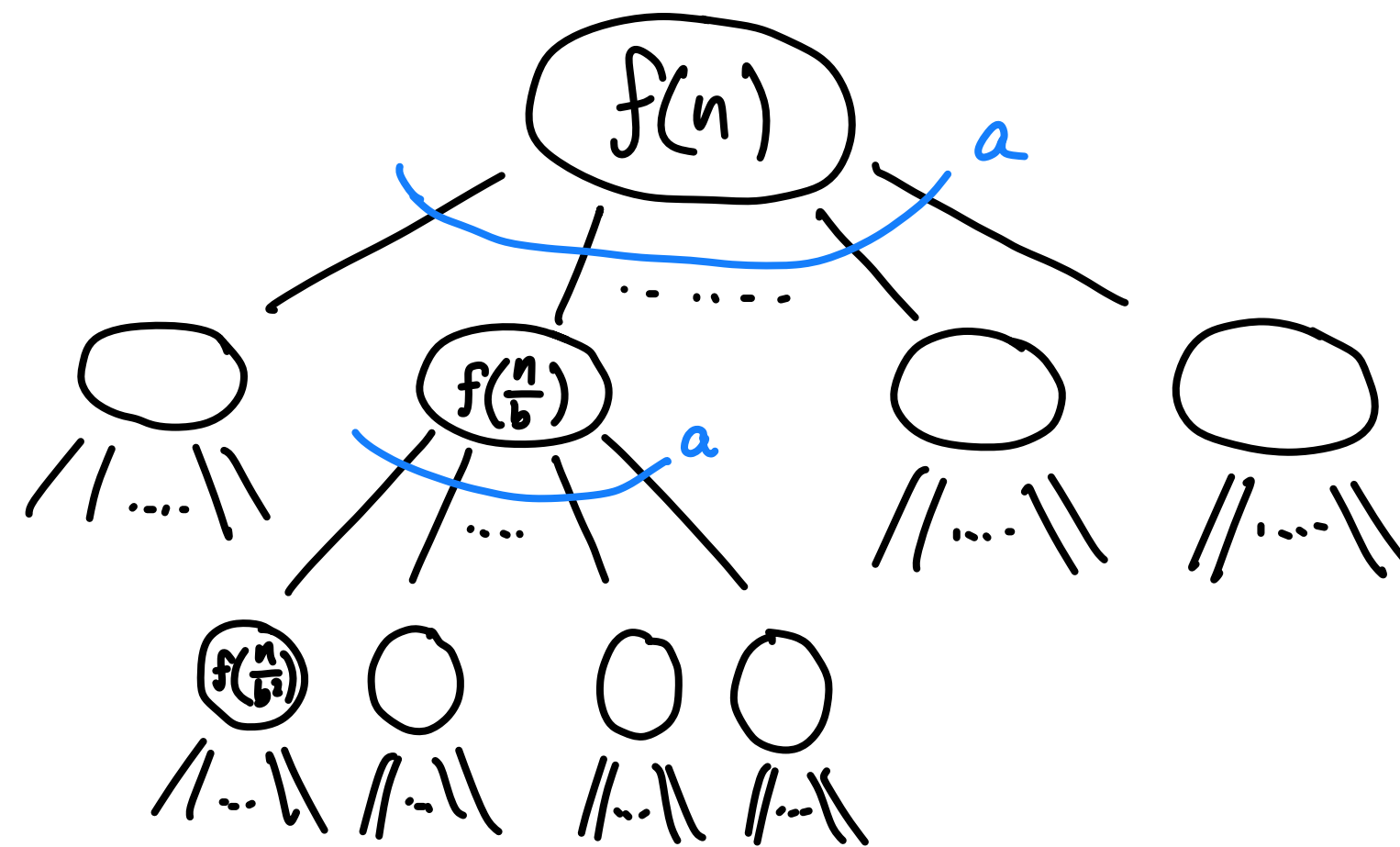
- For solving recursive equations of the form

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + O(n^k) \text{ and } T(n < b) = O(1)$$

- Different cases based on how $f(n)$, a , and b compare:
 - If $a < b^k$, then $T(n) = O(n^k)$
 - If $a = b^k$, then $T(n) = O(n^k \log n)$
 - If $a > b^k$, then $T(n) = O(n^{\log_b a})$

Proof of the master theorem

- **Proof strategy:**
 - Due to recursion, the problem has a tree like structure



- Calculate the amount of work done by the “conquer” step at each level
- Count how many levels of computation there are

Proof the master theorem

- Let $d = \lceil \log_b n \rceil$ so $n \leq b^d$

<u>level</u>	<u># of problems</u>	<u>compute per conquer</u>	<u>total compute at level</u>
d	1	n^k	n^k
$d-1$	a	$(n/b)^k$	$a(n/b)^k = (a/b^k) \cdot n^k$
$\vdots$	$\vdots$	$\vdots$	$\vdots$
$d-j$	a^j	$(n/b^j)^k$	$a^j(n/b^j)^k = (a/b^k)^j \cdot n^k$
$\vdots$	$\vdots$	$\vdots$	$\vdots$
1	a^d	1	a^d

Proof the master theorem

- Let $d = \lceil \log_b n \rceil$ so $n \leq b^d$

$$\text{Total compute} = \sum_{j=0}^d \left(\frac{a}{b^k}\right)^j \cdot n^k$$

– If $a < b^k$, then $\sum_{j=0}^d \left(\frac{a}{b^k}\right)^j \leq \sum_{j=0}^{\infty} \left(\frac{a}{b^k}\right)^j = \left(1 - \frac{a}{b^k}\right)^{-1} \Rightarrow O(n^k)$.

– If $a = b^k$, then $\sum_{j=0}^d \left(\frac{a}{b^k}\right)^j = \sum_{j=0}^d 1 = d+1 \Rightarrow O(n^k \log n)$.

– If $a > b^k$, then $\sum_{j=0}^d \left(\frac{a}{b^k}\right)^j = \frac{\left(\frac{a}{b^k}\right)^{d+1} - 1}{\left(\frac{a}{b^k}\right) - 1} \Rightarrow O\left(\left(\frac{a}{b^k}\right)^d \cdot n^k\right) = O(a^d) = O(n^{\log_b a})$

last row of
prev table.

Analysis divide and conquer runtimes

The master theorem

- For solving recursive equations of the form

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + O(n^k) \text{ and } T(n < b) = O(1)$$

- Different cases based on how $f(n)$, a , and b compare:
 - If $a < b^k$, then $T(n) = O(n^k)$ $\leftarrow$ most of the compute is in the largest conquer step
 - If $a = b^k$, then $T(n) = O(n^k \log n)$ $\leftarrow$ each level has a commensurate amount of compute
 - If $a > b^k$, then $T(n) = O(n^{\log_b a})$ $\leftarrow$ the number of leaves dominates the computation