

# CS 381

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- Alternative proof without using integrals for upper bound of the sum  $\sum_{k=1, n-1} k \cdot \log(k)$
- Closest pair of points
- Farthest pair of points

# Alternative proof without using integrals

## Exercise:

Bound  $\sum_{k=1, n-1} k \cdot \log(k)$  by  $\frac{1}{2}n^2\log(n) - c_1 \cdot n^2$

# Alternative proof without using integrals

## Exercise:

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Write

$$\begin{aligned}\sum_{k=1, n-1} k \cdot \log(k) &\leq \sum_{k=1, \frac{n}{2}} k \cdot \log\left(\frac{n}{2}\right) + \sum_{k=\frac{n}{2}+1, n-1} k \cdot \log(n) \\ &\leq \frac{n(n-1)}{2} \cdot \log(n) - \frac{n}{2} \cdot \left(\frac{n}{2} - 1\right) \cdot \log(2) \\ &\leq \frac{n(n-1)}{2} \cdot \log(n) - \frac{n}{2} \cdot \left(\frac{n}{2} - 1\right) \cdot 0.69\end{aligned}$$

Can take  $c_1 = 0.15$ .

## 5.4 Closest Pair of Points

# Closest Pair of Points

**Closest pair.** Given  $n$  points in the plane, find a pair with smallest Euclidean distance between them.

**Fundamental geometric primitive.**

- Graphics, computer vision, geographic information systems, molecular modeling, air traffic control.
- Special case of nearest neighbor, Euclidean MST, Voronoi.

↑ fast closest pair inspired fast algorithms for these problems

**Algorithm?**

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**Brute force.** Check all pairs of points  $p$  and  $q$  with  $\Theta(n^2)$  comparisons.

**1-D version?**

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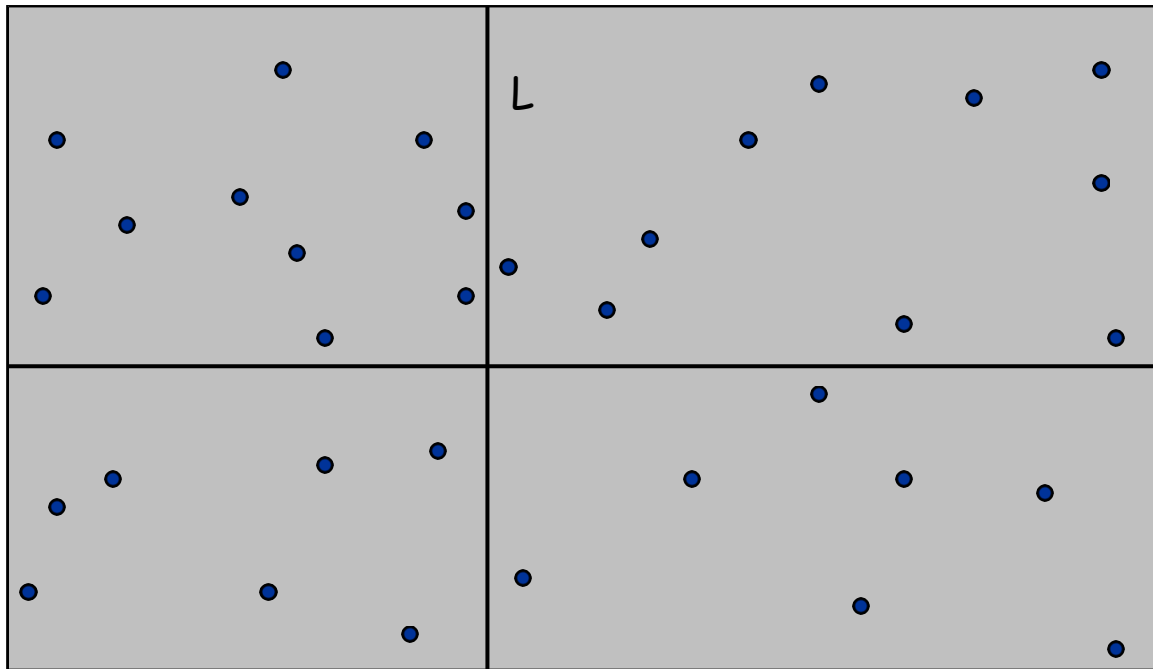
**1-D version.**  $O(n \log n)$  easy if points are on a line.

**Assumption.** No two points have same  $x$  coordinate.

↑  
to make presentation cleaner

# Closest Pair of Points: First Attempt

**Divide.** Sub-divide region into 4 quadrants.

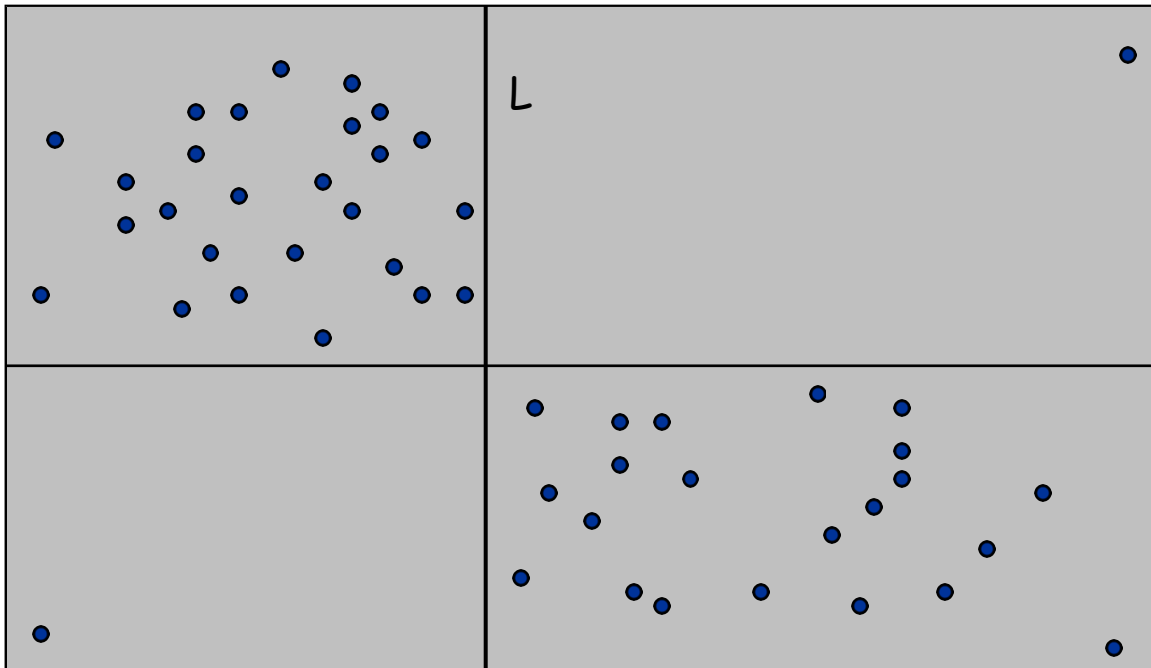




## Closest Pair of Points: First Attempt

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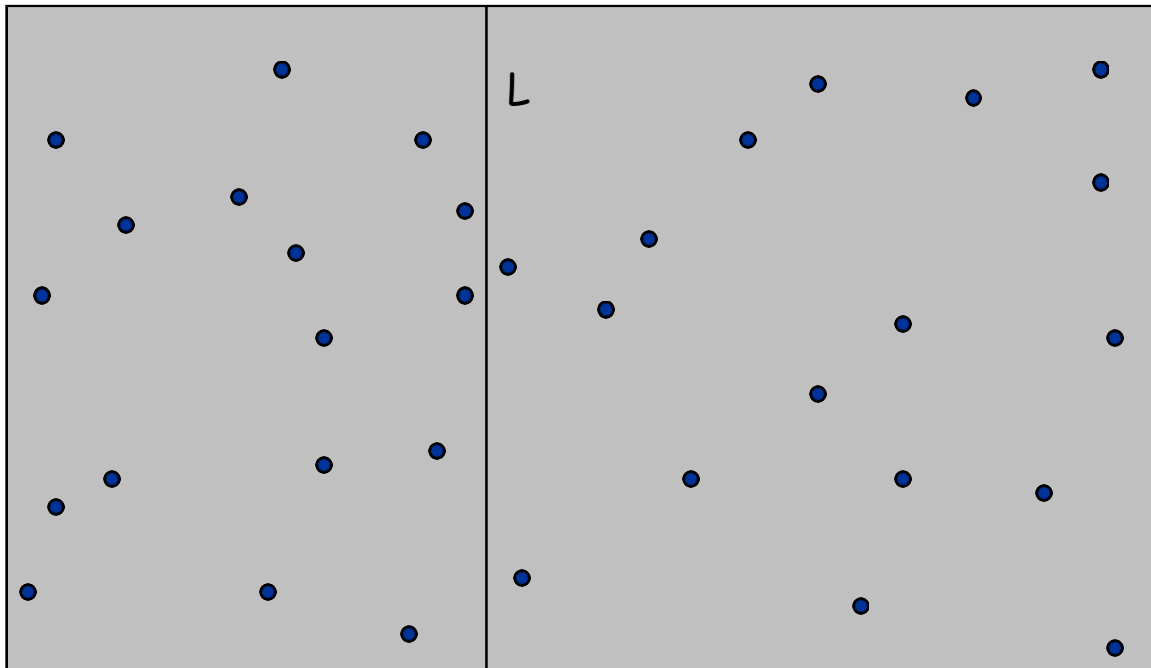
**Obstacle.** Impossible to ensure  $n/4$  points in each piece.



# Closest Pair of Points

## Algorithm.

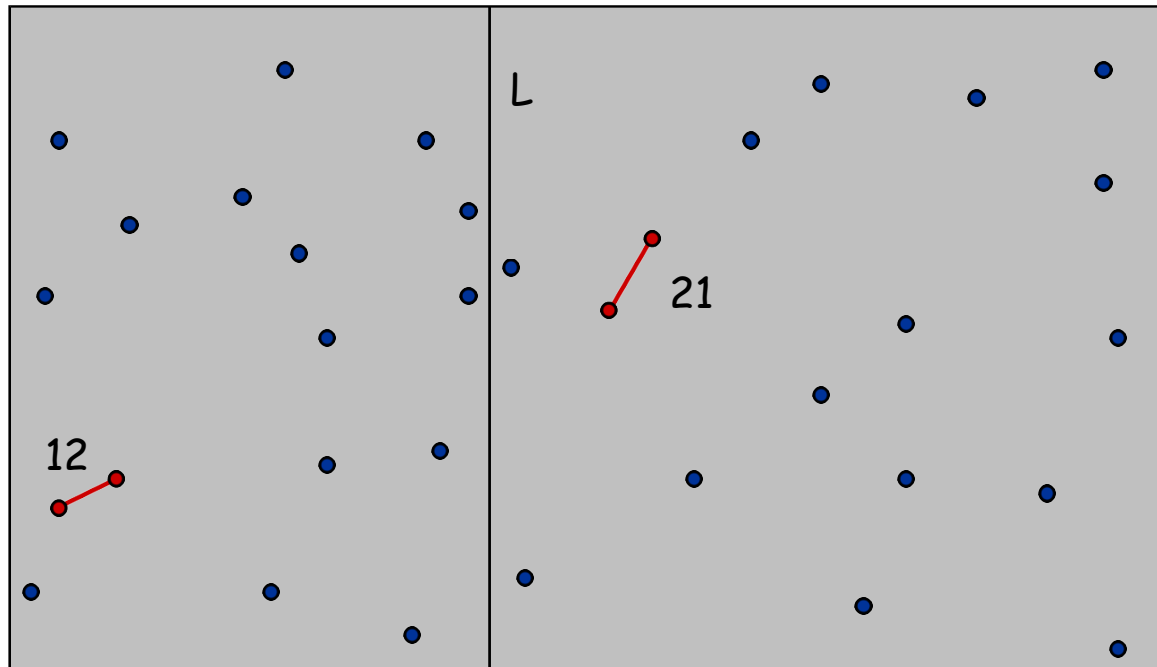
- **Divide:** draw vertical line  $L$  so that roughly  $\frac{1}{2}n$  points on each side.



# Closest Pair of Points

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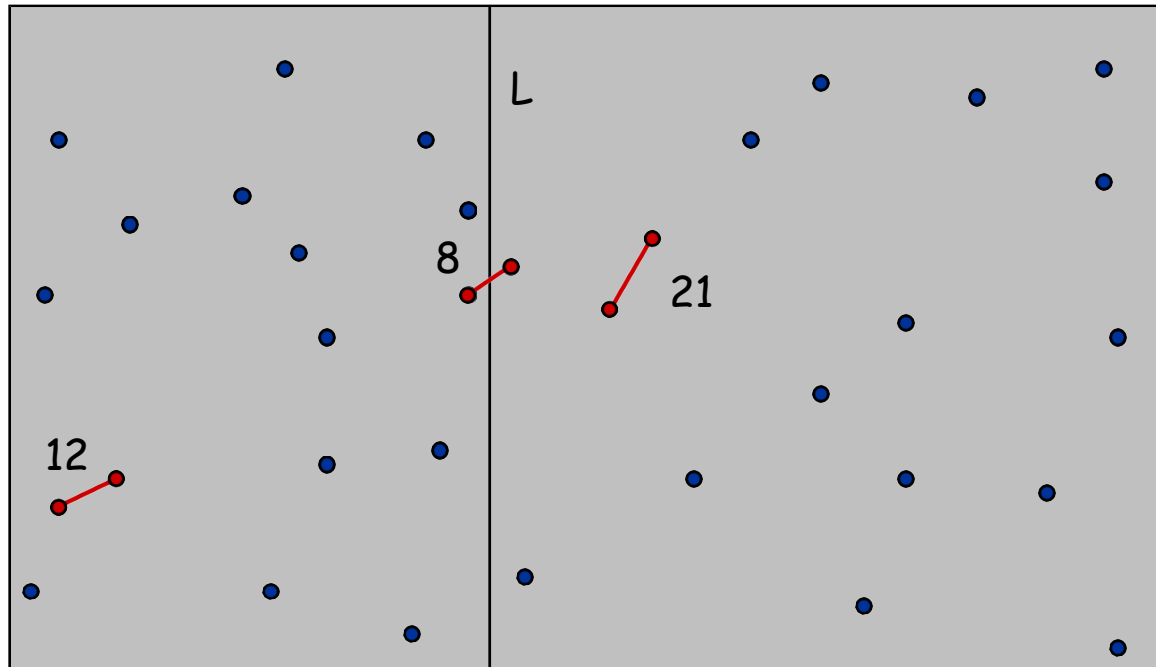
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- **Conquer**: find closest pair in each side recursively.



# Closest Pair of Points

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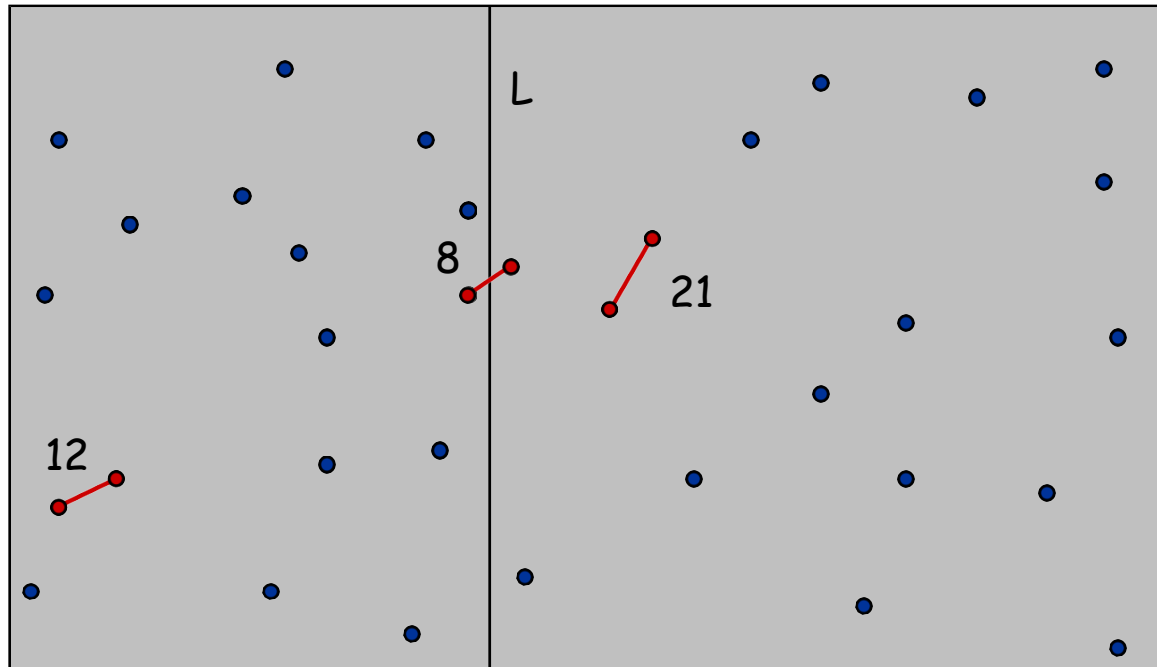
- Divide: draw vertical line  $L$  so that roughly  $\frac{1}{2}n$  points on each side.
- Conquer: find closest pair in each side recursively.
- **Combine**: find closest pair with one point in each side. ← *How much work is this?*
- Return best of 3 solutions.



# Closest Pair of Points

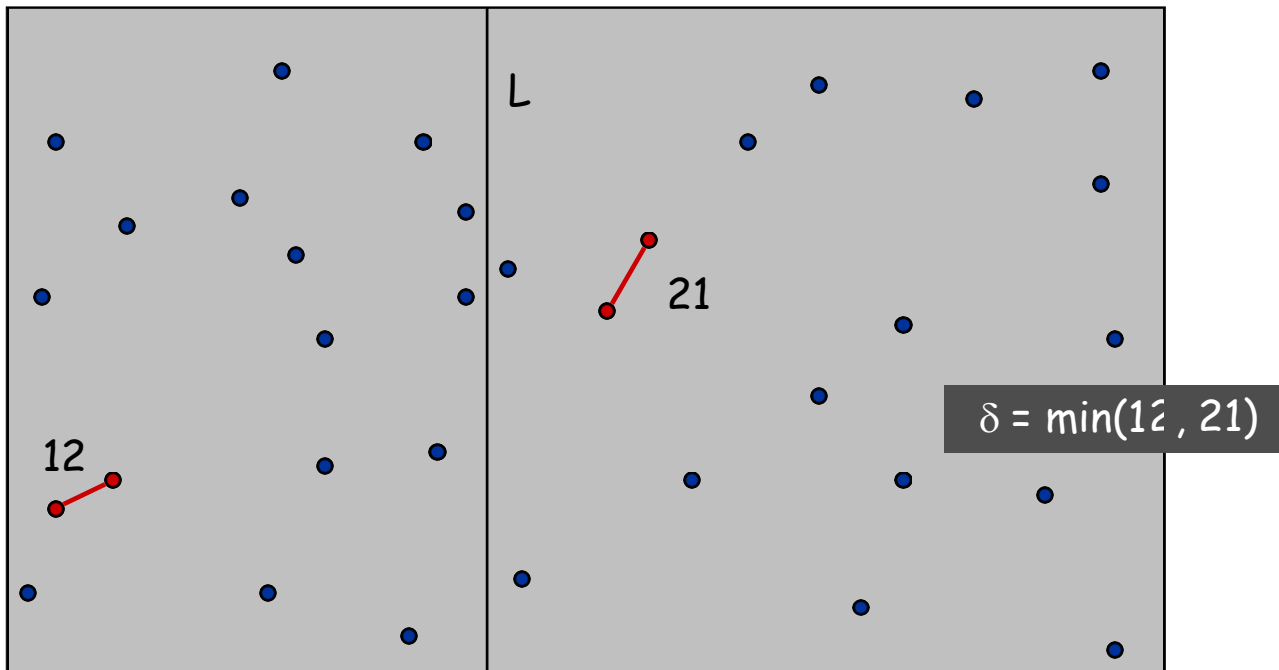
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- Conquer: find closest pair in each side recursively.
- **Combine**: find closest pair with one point in each side. ← seems like  $\Theta(n^2)$
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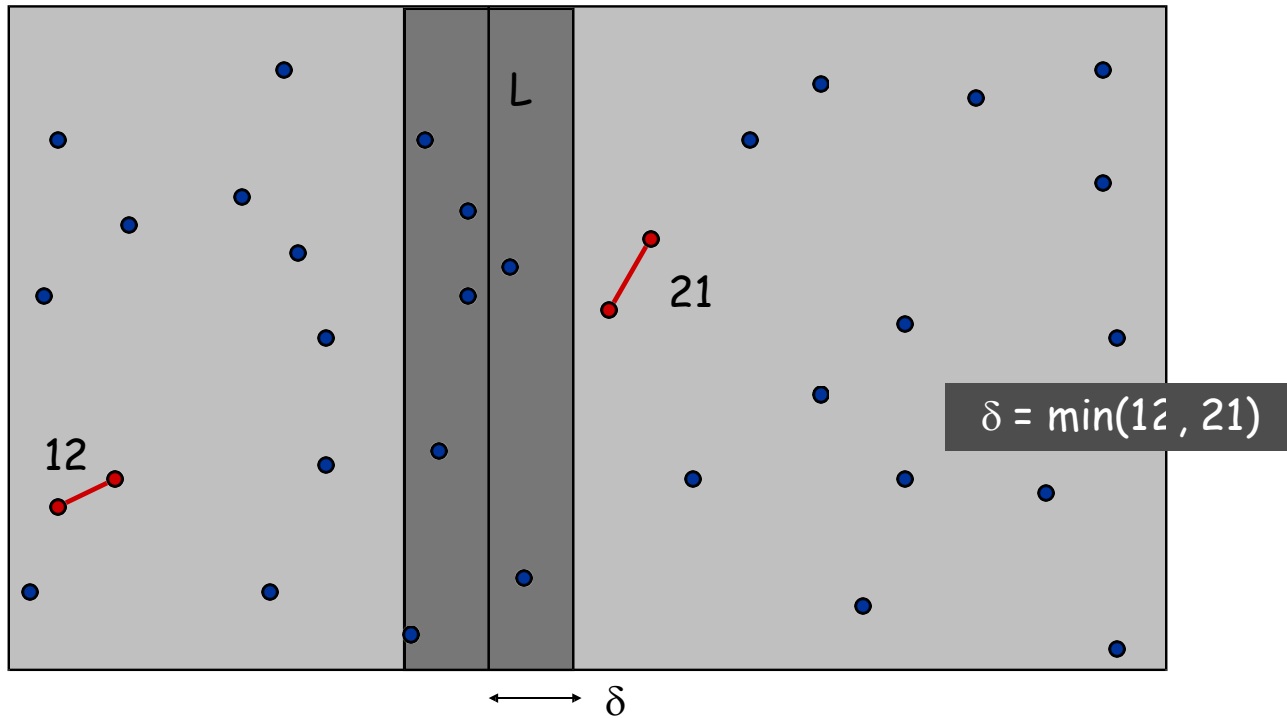
Find closest pair with one point in each side, **assuming that distance  $< \delta$** .



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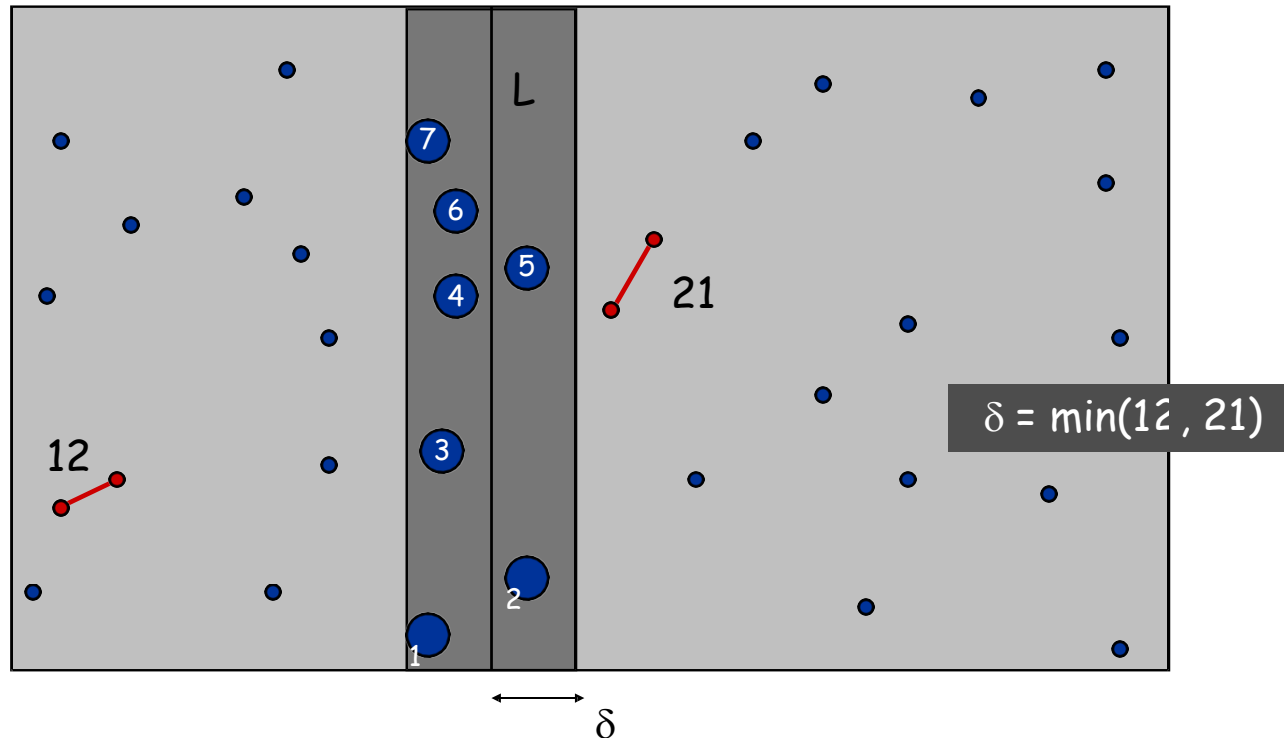
- Observation: only need to consider points within  $\delta$  of line  $L$ .



# Closest Pair of Points

Find closest pair with one point in each side, **assuming that distance  $< \delta$** .

- Observation: only need to consider points within  $\delta$  of line  $L$ .
- Sort points in  $2\delta$ -strip by their  $y$  coordinate.

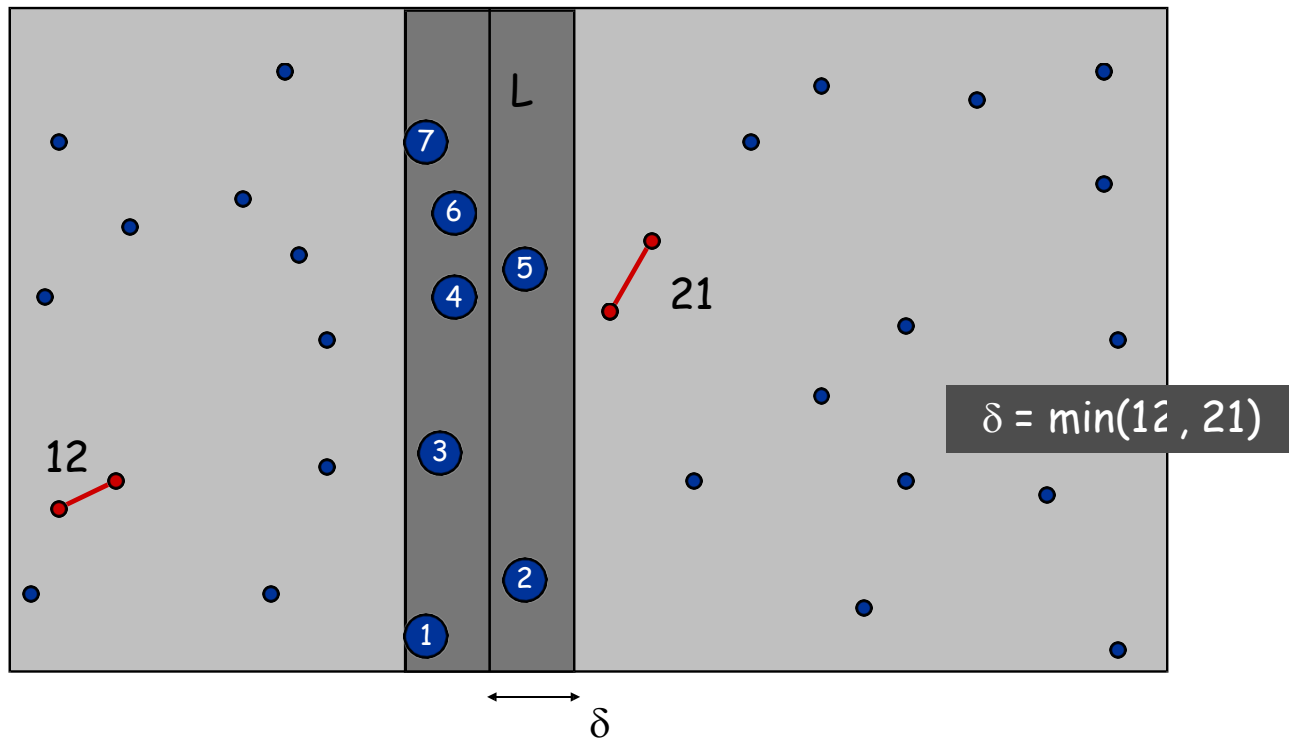




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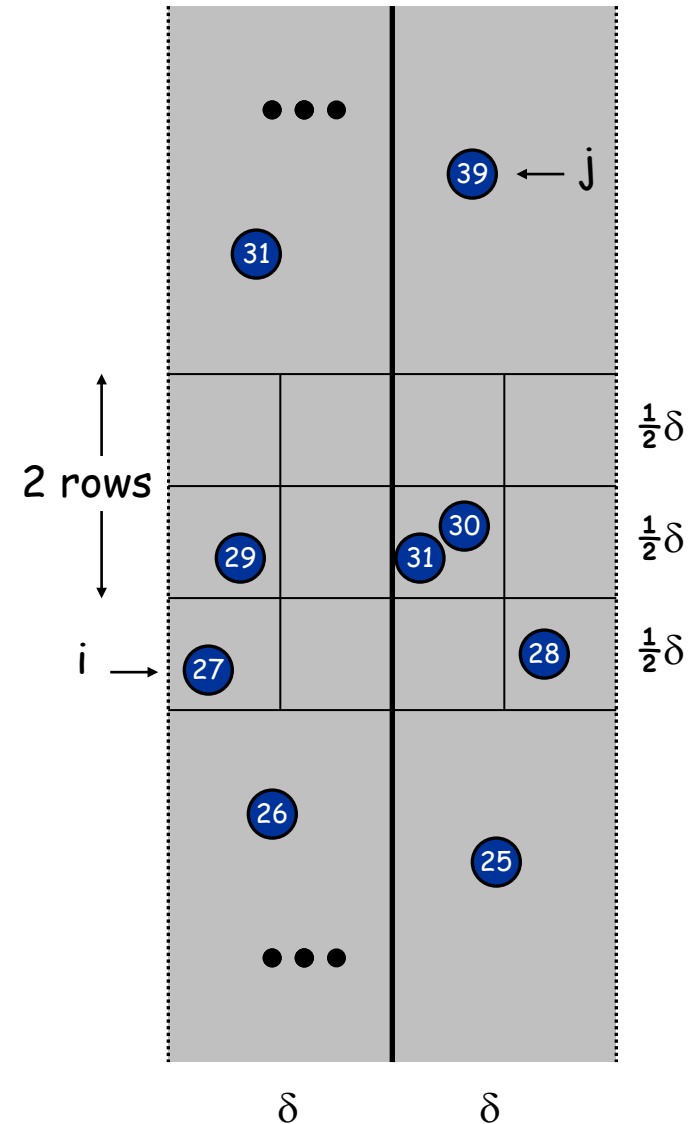
- Observation: only need to consider points within  $\delta$  of line  $L$ .
- Sort points in  $2\delta$ -strip by their  $y$  coordinate.
- Only check distances of those within 11 positions in sorted list!



# Closest Pair of Points

**Def.** Let  $s_i$  be the point in the  $2\delta$ -strip, with the  $i^{\text{th}}$  smallest y-coordinate.

**Claim.** If  $|i - j| \geq 12$ , then the distance between  $s_i$  and  $s_j$  is at least  $\delta$ .



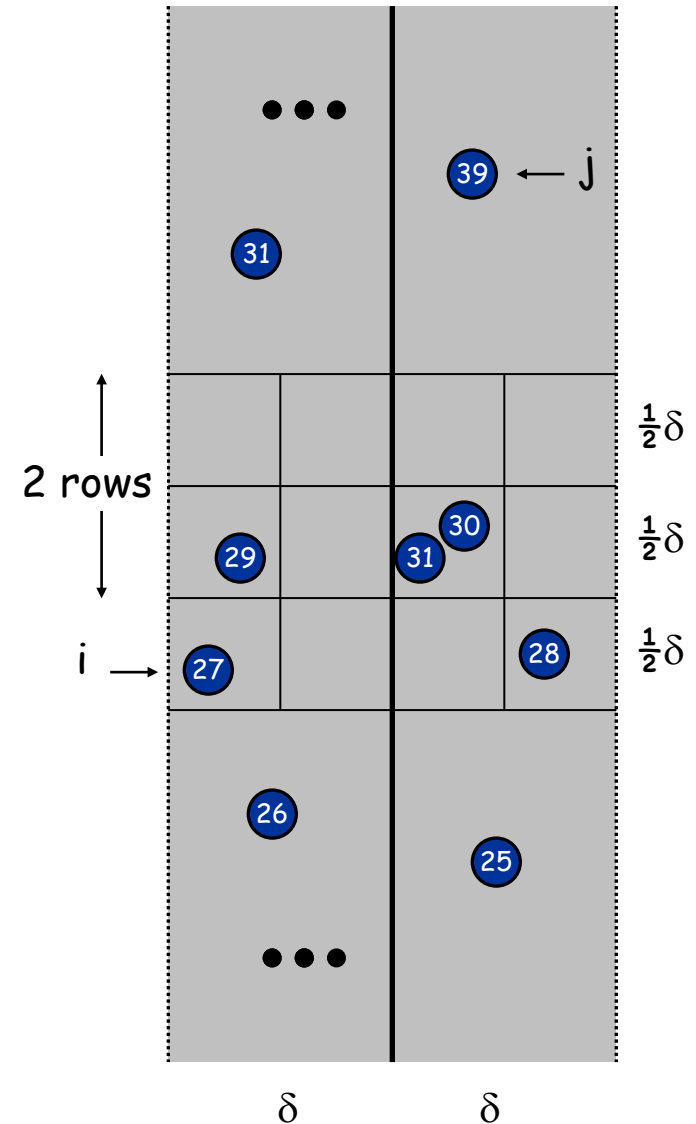
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**Pf.**

- How many points in the same box?



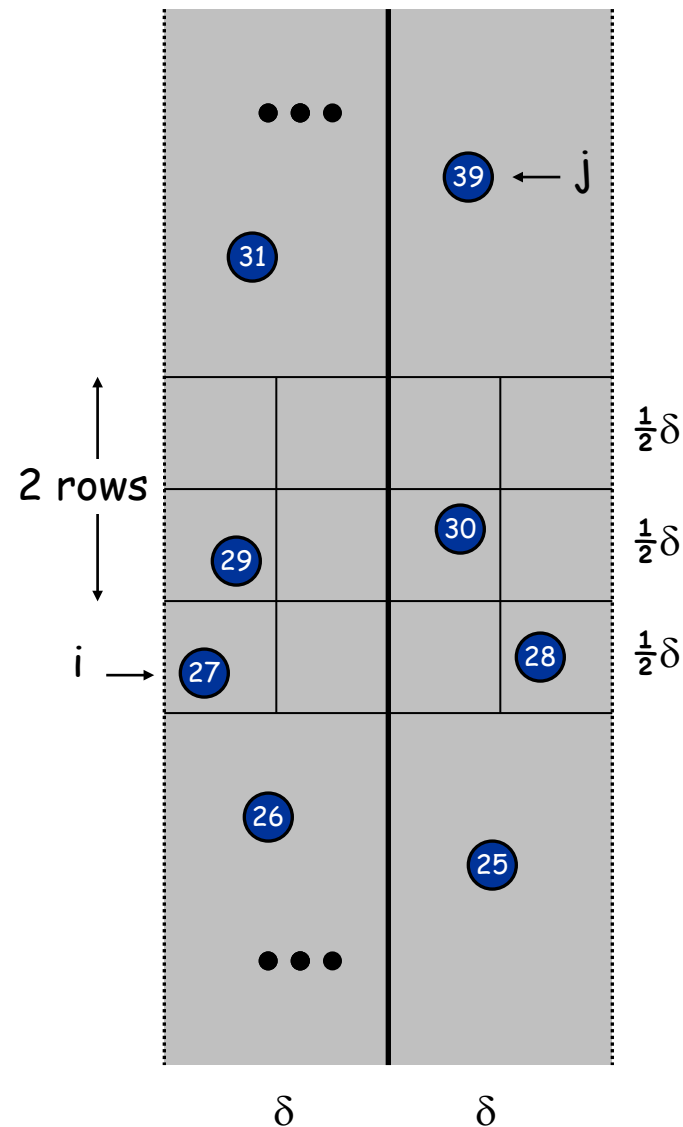
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Pf.

- No two points lie in same  $\frac{1}{2}\delta$ -by- $\frac{1}{2}\delta$  box.



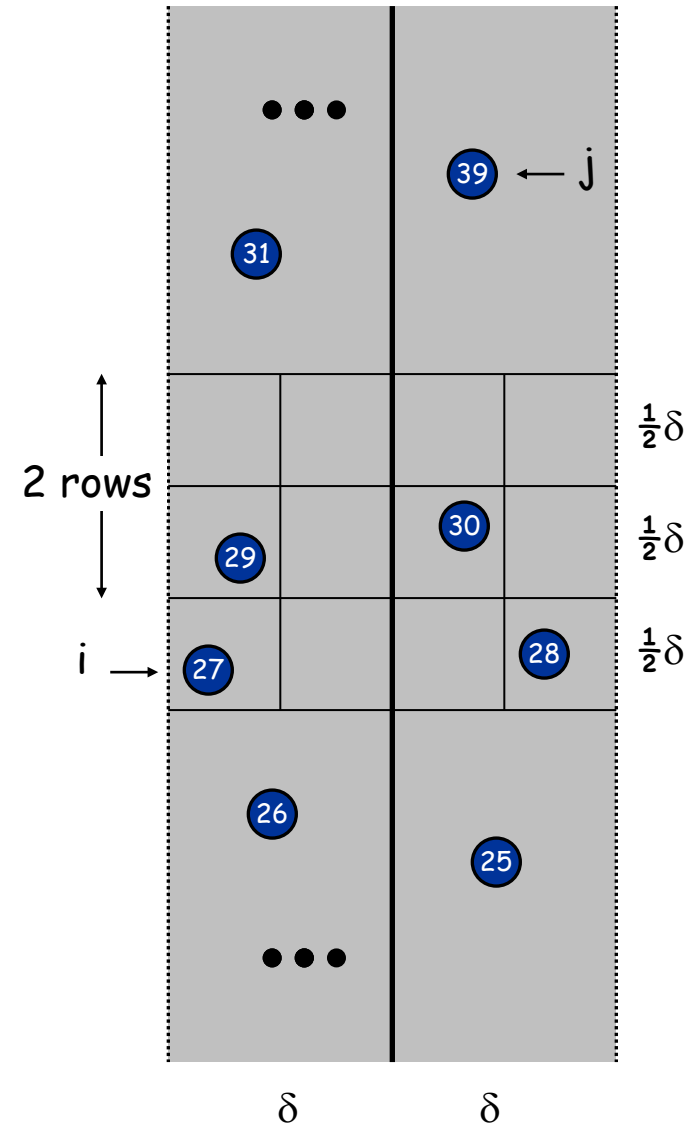
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- How can we use this?



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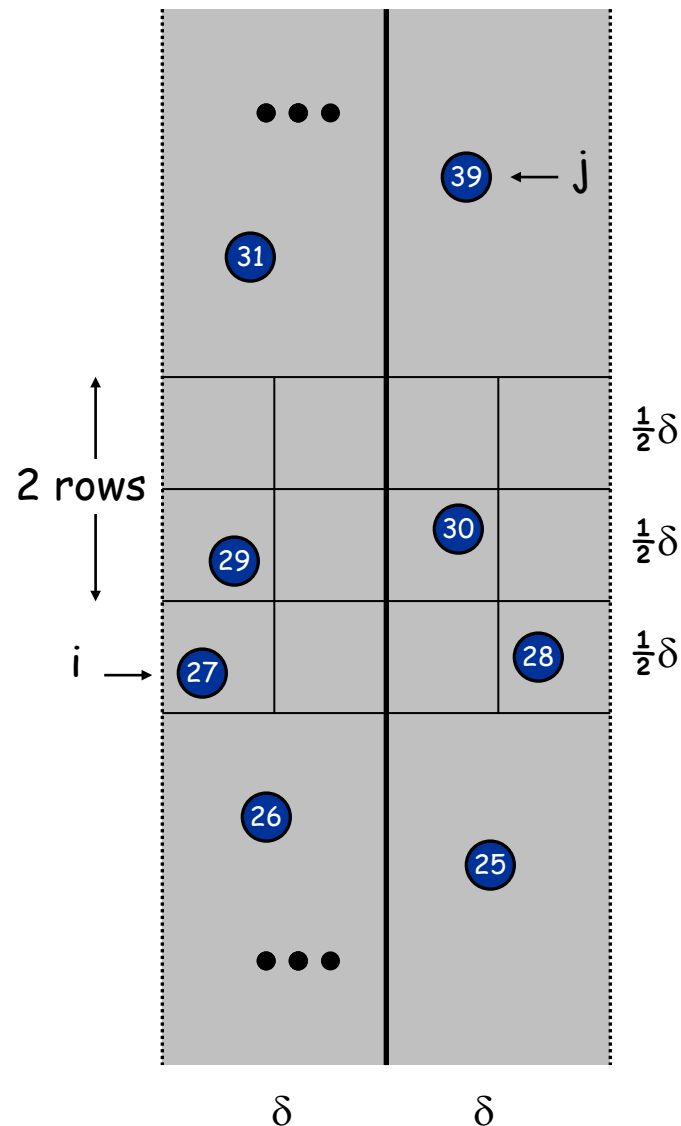
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**Pf.**

- No two points lie in same  $\frac{1}{2}\delta$ -by- $\frac{1}{2}\delta$  box.
- Two points at least 2 rows apart have distance  $\geq 2(\frac{1}{2}\delta)$ . ▫

**Fact.** Still true if we replace 12 with 7.



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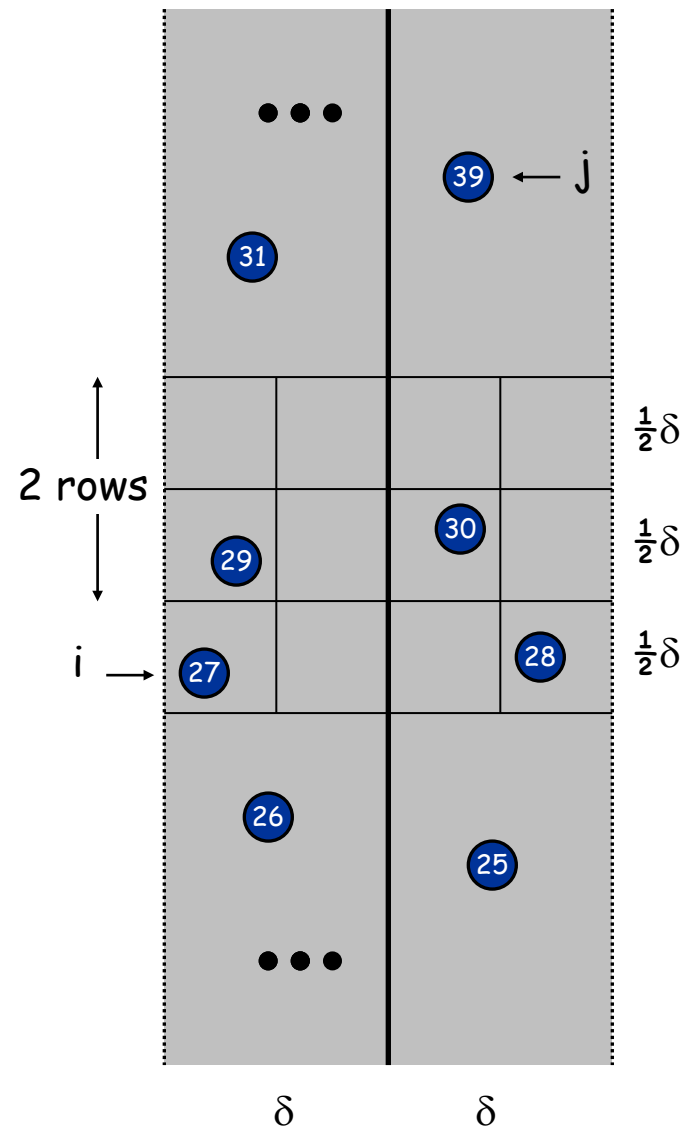
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**Q: How can we use this?**



# Closest Pair Algorithm

```
Closest-Pair( $p_1, \dots, p_n$ ) {  
    Compute separation line  $L$  such that half the points  
    are on one side and half on the other side.  
  
     $\delta_1 = \text{Closest-Pair}(\text{left half})$   
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     $\delta = \min(\delta_1, \delta_2)$   
  
    Delete all points further than  $\delta$  from separation line  $L$   
  
    Sort remaining points by y-coordinate.  
  
    Scan points in y-order and compare distance between  
    each point and next 11 neighbors. If any of these  
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    return  $\delta$ .  
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*Runtime?*

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$O(n \log n)$

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## Closest Pair of Points: Analysis

Running time.

$$T(n) \leq 2T(n/2) + O(n \log n)$$

**Bound?**

## Closest Pair of Points: Analysis

Running time.

$$T(n) \leq 2T(n/2) + O(n \log n) \quad \Rightarrow \quad T(n) = O(n \log^2 n)$$

## Closest Pair of Points: Analysis

Q. Can we achieve  $O(n \log n)$ ?



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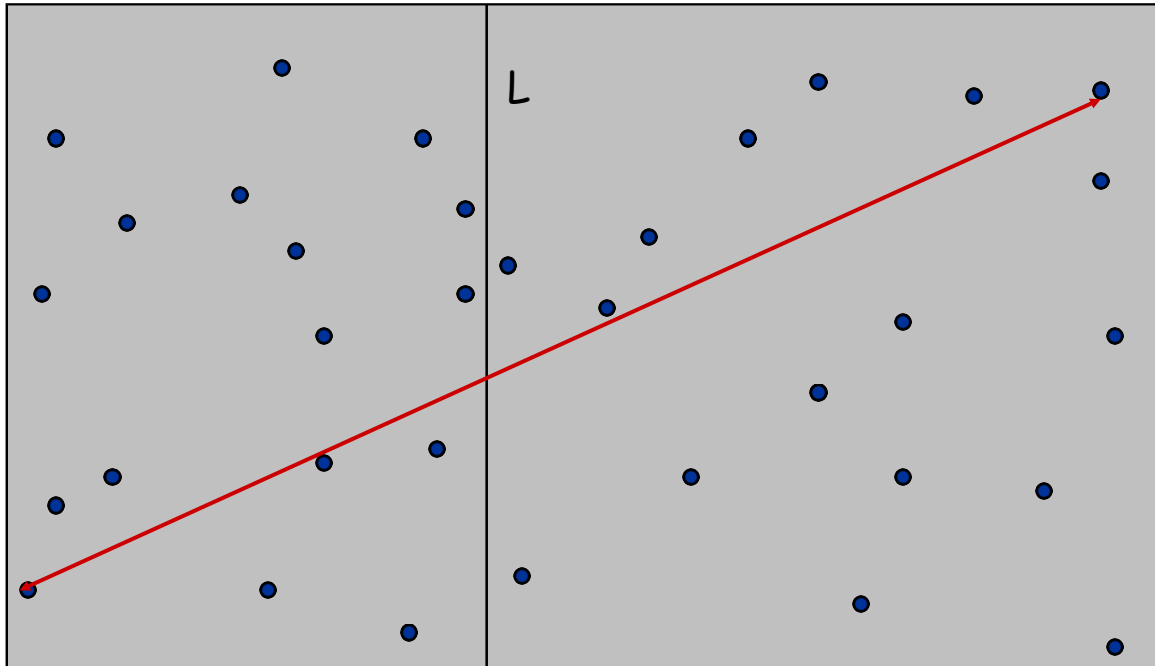
A. Yes. Don't sort points in strip from scratch each time.

- Each recursive call returns two lists: all points sorted by y coordinate, and all points sorted by x coordinate.
- Sort by **merging** two pre-sorted lists.

$$T(n) \leq 2T(n/2) + O(n) \Rightarrow T(n) = O(n \log n)$$

## Farthest Pair of Points

Given data set, find the pair of points that are farthest.



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**Exercise:** Suppose the data is 1D.

## Farthest Pair of Points

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**Exercise:** For 2D data, what is the runtime?

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**Exercise:** For 2D data, can we *find a faster 2-approximation algorithm*?

I.e. Find pair of points  $(p_i, p_j)$  so that

$$d(p_i, p_j) \geq \frac{1}{2} * d(p_k, p_l),$$

where  $(p_k, p_l)$  is the farthest pair of points.

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**Exercise (in class):** Show that this algorithm gives a  $\sqrt{2}$  - approximation to the optimal solution.