

# CS 401

## Master Theorem / Closest Points

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# Divide and Conquer

**Divide:** We reduce a problem to several subproblems.

Typically, each sub-problem is **at most a constant  $c < 1$  fraction** of the size of the original problem

**Conquer:** Recursively solve each subproblem

**Combine:** Merge the solutions


**Examples:**

- Mergesort, Counting Inversions, Binary Search

# Master Theorem

# Master Theorem

Suppose  $T(n) = a T\left(\frac{n}{b}\right) + cn^k$  for all  $n > b$ . Then,



$c$ : absolute constant

- If  $a < b^k$  then  $T(n) = \Theta(n^k)$
- If  $a = b^k$  then  $T(n) = \Theta(n^k \log n)$
- If  $a > b^k$  then  $T(n) = \Theta(n^{\log_b a})$

Works even if it is  $\lceil \frac{n}{b} \rceil$  instead of  $\frac{n}{b}$ .

We also need  $a \geq 1, b > 1, k \geq 0$  and  $T(n) = O(1)$  for  $n \leq b$ .

# Question

Consider the following recurrence. Which case of the master theorem?

$$T(n) = \begin{cases} 0 & \text{if } n \leq 1 \\ T\left(\left\lfloor \frac{n}{5} \right\rfloor\right) + T\left(n - 3\left\lfloor \frac{n}{10} \right\rfloor\right) + \frac{11}{5}n & \text{if } n > 1 \end{cases}$$

- A.  $T(n) = \Theta(n)$
- B.  $T(n) = \Theta(n \log n)$
- C.  $T(n) = \Theta(n^2)$
- D. Master theorem not applicable

Master Theorem

Suppose  $T(n) = a T\left(\frac{n}{b}\right) + cn^k$  for all  $n > b$ .

$$T(n) = \begin{cases} \Theta(n^k) & \text{if } a < b^k \\ \Theta(n^k \log n) & \text{if } a = b^k \\ \Theta(n^{\log_b a}) & \text{if } a > b^k \end{cases}$$

Akra–Bazzi theorem  
Wiki!

# How to use master theorem?

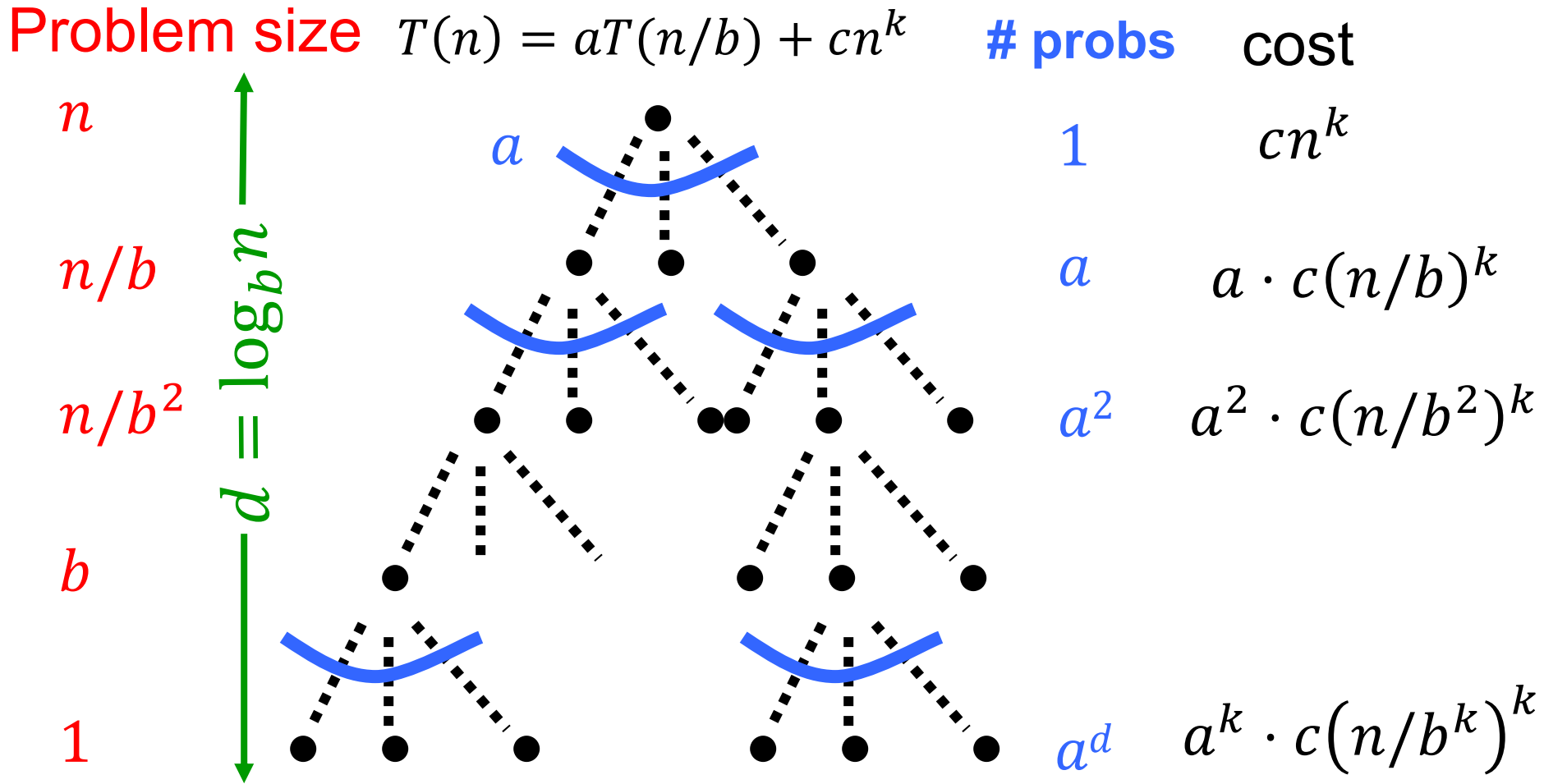
For a divide and conquer algorithm

- $a$ : number of subproblems
- $b$ : ratio of problem size / subproblem size
- $c \cdot n^k$ : running time of divide and combine step

We have recurrence  $T(n) = a T\left(\frac{n}{b}\right) + cn^k$  for all  $n > b$ .

Example: Mergesort have two subproblems of half size of the original problem, and the cost of divide and combine step is  $O(n)$ , so  $T(n) = 2 T(n / 2) + c n$ , which implies  $T(n) = \Theta(n \log n)$

# Understand Master Theorem



$$T(n) = \sum_{i=0}^{d=\log_b n} a^i c \left(\frac{n}{b^i}\right)^k$$

# Understand Master Theorem

Suppose  $T(n) = a T\left(\frac{n}{b}\right) + cn^k$  for all  $n > b$ . Then,

- If  $a < b^k$  then  $T(n) = \Theta(n^k)$  # of problems increases **slower** than the decreases of cost.  
**Top** term dominates.
- If  $a = b^k$  then  $T(n) = \Theta(n^k \log n)$
- If  $a > b^k$  then  $T(n) = \Theta(n^{\log_b a})$  # of problems increases **faster** than the decreases of cost  
**Bottom** term dominates.

$$T(n) = \sum_{i=0}^{d=\log_b n} a^i c \left(\frac{n}{b^i}\right)^k$$



# A Useful Identity

**Theorem:**  $1 + x + x^2 + \dots + x^d = \frac{x^{d+1} - 1}{x - 1}$

**Proof:** Let  $S = 1 + x + x^2 + \dots + x^d$

Then,  $xS = x + x^2 + \dots + x^{d+1}$

So,  $xS - S = x^{d+1} - 1$

i.e.,  $S(x - 1) = x^{d+1} - 1$

Therefore,  $S = \frac{x^{d+1} - 1}{x - 1}$

**Corollary:**

$$1 + x + x^2 + \dots + x^d = \begin{cases} O_x(1) & \text{if } x < 1 \\ d + 1 & \text{if } x = 1 \\ O_x(x^{d+1}) & \text{if } x > 1 \end{cases}$$

$O_x$  means the hidden constant depends on  $x$

$$\text{Solve: } T(n) = aT\left(\frac{n}{b}\right) + cn^k$$

Corollary:

$$1 + x + x^2 + \dots + x^d = \begin{cases} \Theta_x(1) & \text{if } x < 1 \\ \Theta(d) & \text{if } x = 1 \\ \Theta_x(x^{d+1}) & \text{if } x > 1 \end{cases}$$

Going back, we have

$$T(n) = \sum_{i=0}^{d=\log_b n} a^i c \left(\frac{n}{b^i}\right)^k = cn^k \sum_{i=0}^{d=\log_b n} \left(\frac{a}{b^k}\right)^i$$

Hence, we have

$$T(n) = \Theta(n^k) \begin{cases} 1 & \text{if } a < b^k \\ \log_b n & \text{if } a = b^k \\ \left(\frac{a}{b^k}\right)^{\log_b n} & \text{if } a > b^k \end{cases}$$

$$\text{Solve: } T(n) = aT\left(\frac{n}{b}\right) + cn^k$$

$$T(n) = \Theta(n^k) \begin{cases} 1 & \text{if } a < b^k \\ \log_b n & \text{if } a = b^k \\ \left(\frac{a}{b^k}\right)^{\log_b n} & \text{if } a > b^k \end{cases}$$

For  $a < b^k$ , we simply have  $T(n) = \Theta(n^k)$ .

For  $a = b^k$ , we have  $T(n) = \Theta(n^k \log_b n) = \Theta(n^k \log n)$ .

For  $a > b^k$ , we have  $T(n) = \Theta\left(n^k \left(\frac{a}{b^k}\right)^{\log_b n}\right) = \Theta(n^{\log_b a})$ .

$$\begin{aligned} & b^k \log_b n \\ &= (b^{\log_b n})^k \\ &= n^k \end{aligned}$$

$$\begin{aligned} & a^{\log_b n} \\ &= (b^{\log_b a})^{\log_b n} \\ &= (b^{\log_b n})^{\log_b a} \\ &= n^{\log_b a} \end{aligned}$$

# Finding the Closest Pair of Points

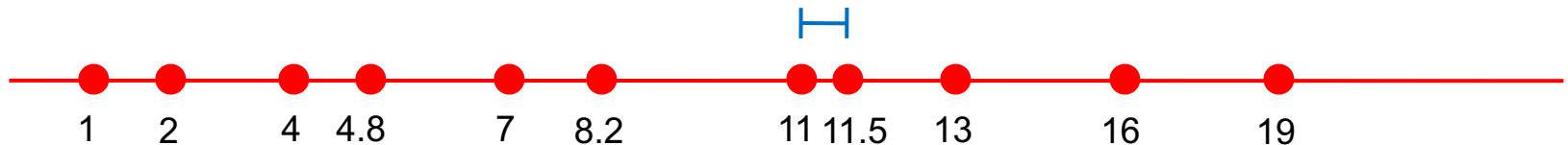
# Closest Pair of Points (1-dim)

Given  $n$  points, find the closest pair.

Brute force: Check all  $\binom{n}{2}$  pairwise distances

1-dim case: 11, 2, 4, 19, 4.8, 7, 8.2, 16, 11.5, 13, 1

find the closest pair



**Fact:** Closest pair is **adjacent** in ordered list

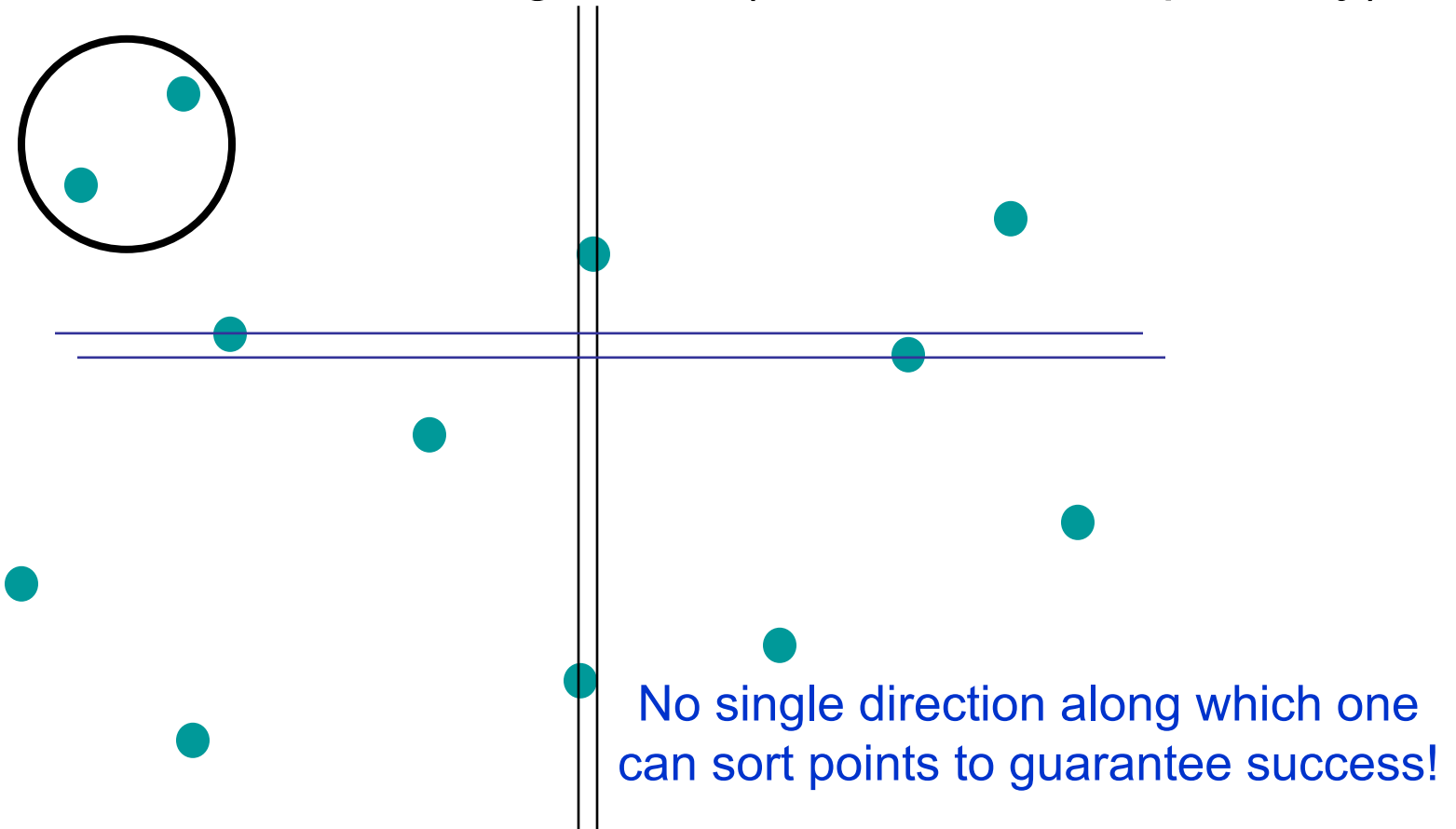
So, first sort, then scan adjacent pairs.

Time  $O(n \log n)$  to sort, if needed, Plus  $O(n)$  to scan adjacent pairs

# Closest Pair of Points (2-dim)

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

Idea: make use of 1-dim algorithm (but not in a simple way)



# Divide & Conquer

**Divide:** draw vertical line  $L$  with  $\approx n/2$  points on each side.

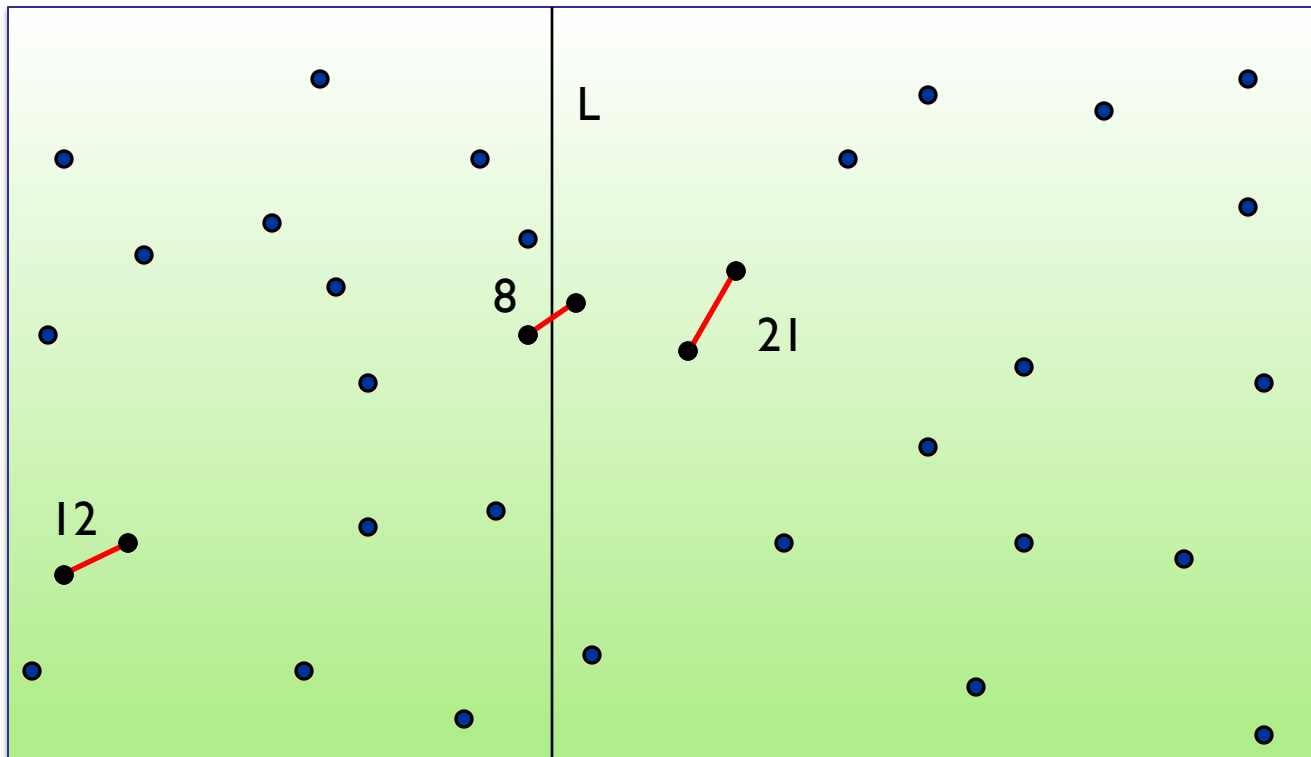
**Conquer:** find closest pair on each side, recursively.

**Combine** to find closest pair overall



How ?

Return best solutions







# Almost 1D Problem

Partition each side of  $L$  into  $\frac{\delta}{2} \times \frac{\delta}{2}$  squares

**Claim:** No two points lie in the same  $\frac{\delta}{2} \times \frac{\delta}{2}$  box.

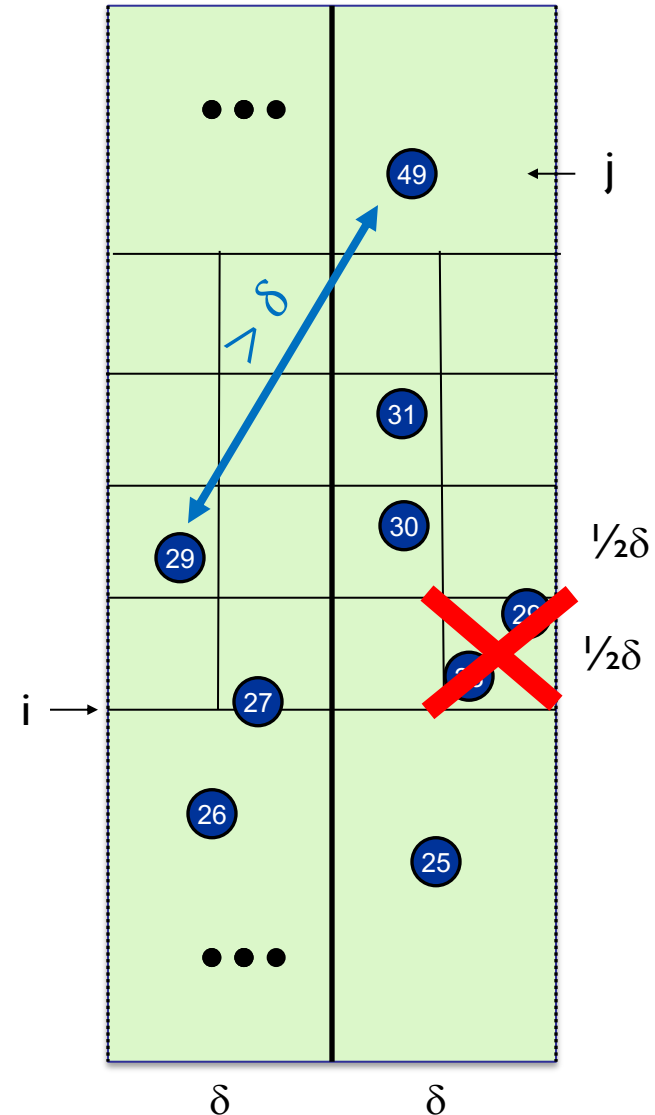
**Proof:** Such points would be within

$$\sqrt{\left(\frac{\delta}{2}\right)^2 + \left(\frac{\delta}{2}\right)^2} = \delta \sqrt{\frac{1}{2}} \approx 0.7\delta < \delta$$

Let  $s_i$  have the  $i^{\text{th}}$  smallest  $y$ -coordinate among points in the  $2\delta$ -width-strip.

**Claim:** If  $|i - j| > 11$ , then the distance between  $s_i$  and  $s_j$  is  $> \delta$ .

**Proof:** only 11 boxes within  $\delta$  of  $y(s_i)$ .



# Closest Pair (2 dimension)

```
Closest-Pair( $p_1, p_2, \dots, p_n$ ) {
```

```
  if ( $n \leq 2$ ) return distance( $p_1, p_2$ )
```

```
  Compute separation line  $L$  such that half the points  
  are on one side and half on the other side.
```

$O(n \log n)$

```
   $\delta_1$  = Closest-Pair(left half)
```

```
   $\delta_2$  = Closest-Pair(right half)
```

```
   $\delta$  = min( $\delta_1, \delta_2$ )
```

$O(1)$

```
  Delete all points further than  $\delta$  from separation line  $L$ 
```

$O(n)$

```
  Sort remaining points  $p[1] \dots p[m]$  by y-coordinate.
```

$O(n \log n)$

```
  for  $i = 1, 2, \dots, m$ 
```

```
    for  $k = 1, 2, \dots, 11$ 
```

```
      if  $i + k \leq m$ 
```

```
         $\delta = \min(\delta, \text{distance}(p[i], p[i+k]));$ 
```

$O(n)$

```
  return  $\delta$ .
```

```
}
```

# Closest Pair Analysis

Running time?

$$T(n) \leq \begin{cases} 1 & \text{if } n \leq 2 \\ 2T\left(\frac{n}{2}\right) + O(n \log n) & \text{o.w.} \end{cases} \Rightarrow T(n) = O(n \log^2 n)$$

Can we do better?

# Closest Pair (2 dimension) Improved

```
Closest-Pair( $p_1, p_2, \dots, p_n$ ) ← {  
  if ( $n \leq 2$ ) return distance( $p_1, p_2$ )
```

Assume: input sorted by x-coordinate  
( $O(n \log n)$  overhead initially)

```
  Compute separation line  $L$  such that half the points  
  are on one side and half on the other side.  $O(1)$ 
```

```
  ( $\delta_1, Q_1$ ) = Closest-Pair(left half)
```

```
  ( $\delta_2, Q_2$ ) = Closest-Pair(right half)  $O(1)$ 
```

```
   $\delta$  = min( $\delta_1, \delta_2$ )  $O(1)$ 
```

```
   $Q_{sorted}$  = merge( $Q_1, Q_2$ ) (merge sort it by y-coordinate)  $O(n)$ 
```

```
  Let  $S$  be points (ordered as  $Q_{sorted}$ ) that is  $\delta$  from line  $L$ .  $O(n)$ 
```

```
  for  $i = 1, 2, \dots, m$ 
```

```
    for  $k = 1, 2, \dots, 11$ 
```

```
      if  $i + k \leq m$ 
```

```
         $\delta = \min(\delta, \text{distance}(S[i], S[i+k]));$   $O(n)$ 
```

```
  return  $\delta$  and  $Q_{sorted}$ .
```

```
}
```

Input sorted by  
y-coordinate

$$T(n) \leq \begin{cases} 1 & \text{if } n = 1 \\ 2T\left(\frac{n}{2}\right) + O(n) & \text{o.w.} \end{cases}$$

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

# Summary

**Closest pair in 2-dimension:** Given  $n$  points in the plane, find a pair with smallest Euclidean distance between them.

**Brute Force:** Check all pairs of points in  $\Theta(n^2)$  time.

**Divide and Conquer:**

- **Divide:** draw vertical line  $L$  with  $\approx n/2$  points on each side.
- **Conquer:** find closest pair on each side, recursively.
- **Combine** to find closest pair overall

**Exercise:** Remove the assumption of “no two points have same  $x$  coordinate”?