

# DIVIDE & CONQUER

Divide & Conquer (3)  
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Algorithm Design and Analysis

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Slides credited from Hsueh-I Lu, Hsu-Chun Hsiao, & Michael Tsai

# Outline



- Recurrence (遞迴)
- Divide-and-Conquer
- D&C #1: Tower of Hanoi (河內塔)
- D&C #2: Merge Sort
- D&C #3: Bitonic Champion
- D&C #4: Maximum Subarray
- Solving Recurrences
  - Substitution Method
  - Recursion-Tree Method
  - Master Method
- D&C #5: Matrix Multiplication
- D&C #6: Selection Problem
- D&C #7: Closest Pair of Points Problem

Divide-and-Conquer 首部曲

Divide-and-Conquer 之神乎奇技



# D&C #5: Matrix Multiplication

Textbook Chapter 4.2 – Strassen's algorithm for matrix multiplication

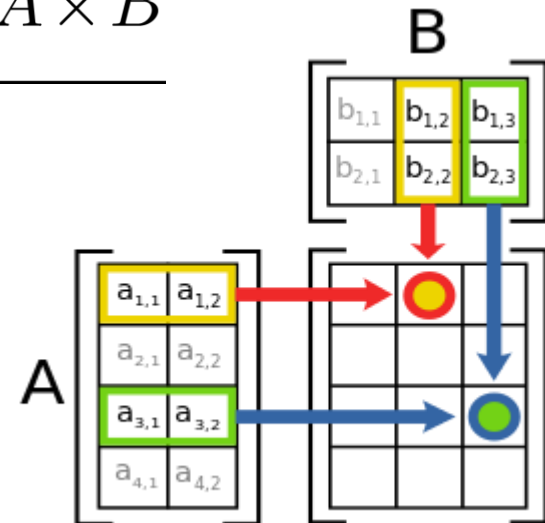
# Matrix Multiplication Problem

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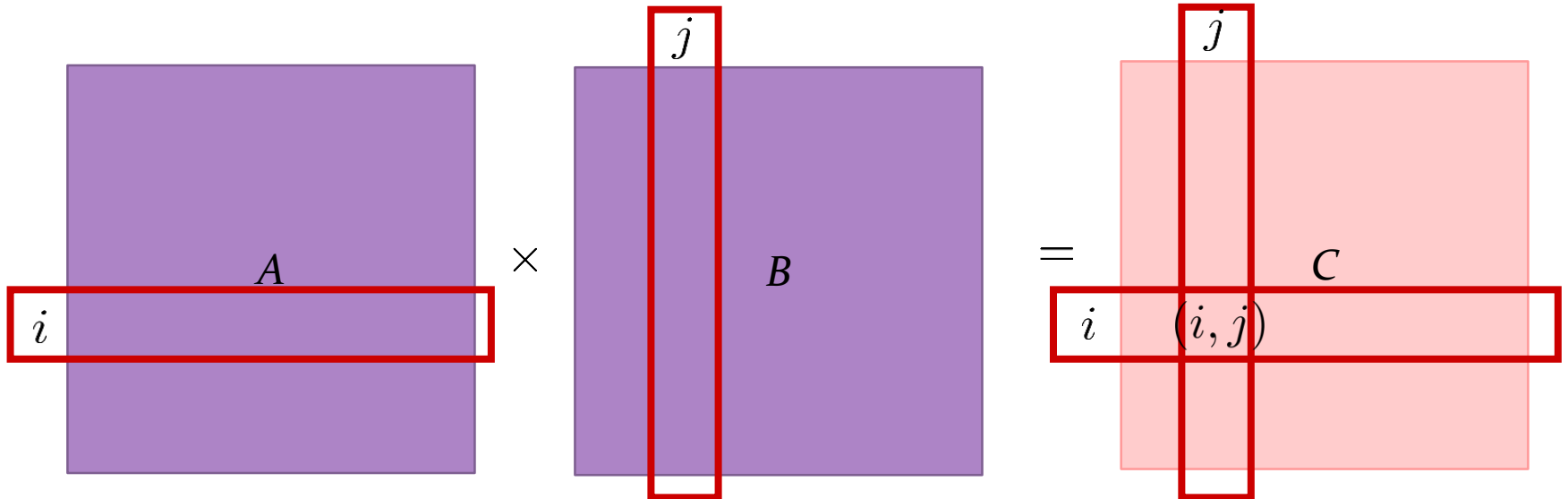
Input: two  $n \times n$  matrices  $A$  and  $B$ .

Output: the product matrix  $C = A \times B$

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# Naïve Algorithm

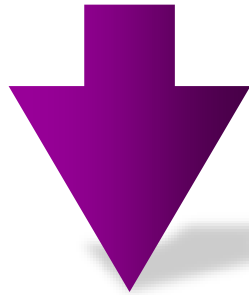


$$C(i, j) = \sum_{k=1}^n A(i, k) \cdot B(k, j)$$

- Each entry takes  $n$  multiplications
- There are total  $n^2$  entries

➔  $\Theta(n)\Theta(n^2) = \Theta(n^3)$

# Matrix Multi. Problem Complexity



Upper bound =  $O(n^3)$



Lower bound =  $\Omega(n^2)$

Why?

# Divide-and-Conquer

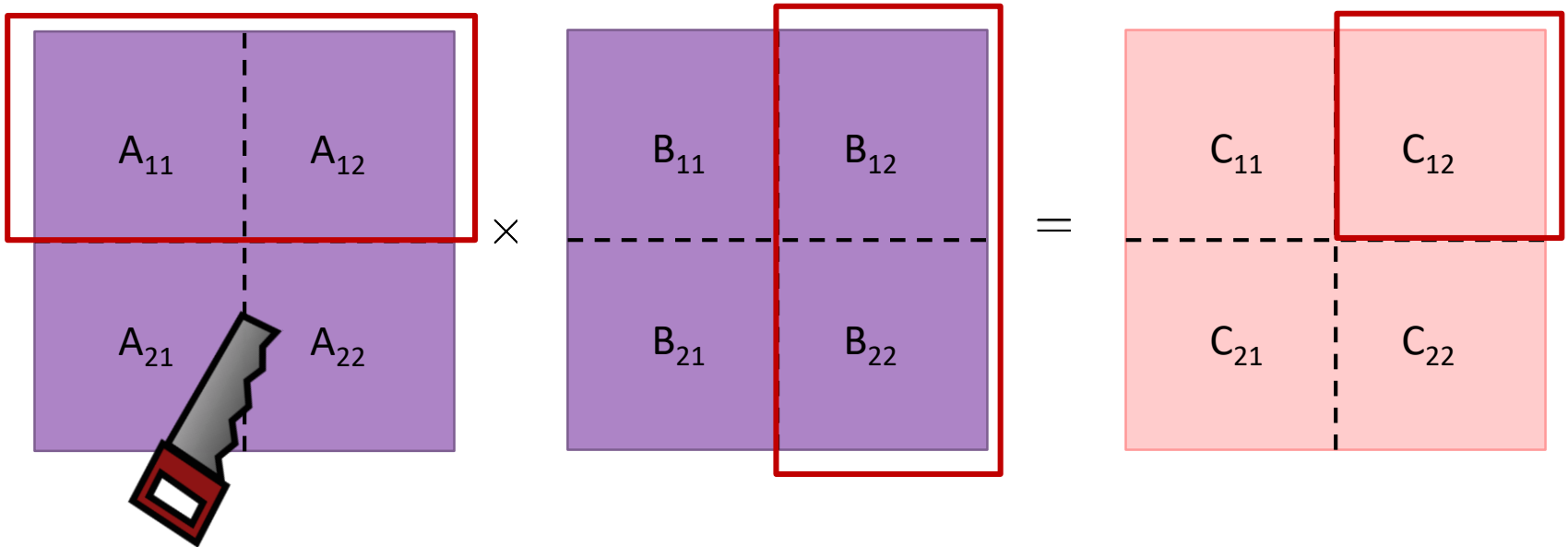
- We can assume that  $n = 2^k$  for simplicity
  - Otherwise, we can increase  $n$  s.t.  $n = 2^{\lceil \log_2 n \rceil}$
  - $n$  may not be twice large as the original in this modification

$$C_{11} = A_{11}B_{11} + A_{12}B_{21}$$

$$C_{12} = A_{11}B_{12} + A_{12}B_{22}$$

$$C_{21} = A_{21}B_{11} + A_{22}B_{21}$$

$$C_{22} = A_{21}B_{12} + A_{22}B_{22}$$



# Algorithm Time Complexity

```
MatrixMultiply(n, A, B)
```

```
//base case
```

```
if n == 1
```

```
    return AB  $\Theta(1)$ 
```

```
//recursive case
```

```
Divide A and B into n/2 by n/2 submatrices Divide  $\Theta(1)$ 
```

```
C11 = MatrixMultiply(n/2, A11, B11) + MatrixMultiply(n/2, A12, B21)
```

```
C21 = MatrixMultiply(n/2, A11, B12) + MatrixMultiply(n/2, A12, B22)
```

```
C21 = MatrixMultiply(n/2, A21, B11) + MatrixMultiply(n/2, A22, B21)
```

```
C22 = MatrixMultiply(n/2, A21, B12) + MatrixMultiply(n/2, A22, B22)
```

```
return C
```

**Combine**  $4\Theta((n/2)^2) = \Theta(n^2)$

Conquer

$8T(n/2)$

- $T(n)$  = time for running `MatrixMultiply(n, A, B)`

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ 8T(n/2) + \Theta(n^2) & \text{if } n \geq 2 \end{cases} \Rightarrow \Theta(n^{\log_2 8}) = \Theta(n^3)$$







# Strassen's Technique

- Important theoretical breakthrough by Volker Strassen in 1969
- Reduces the running time from  $\Theta(n^3)$  to  $\Theta(n^{\log^2 7}) \approx \Theta(n^{2.807})$
- The key idea is to reduce the number of recursive calls
  - From 8 recursive calls to 7 recursive calls  $T(n/2)$
  - At the cost of extra addition and subtraction operations  $\Theta((n/2)^2)$

轉換調整  
加加減減  
兜出答案

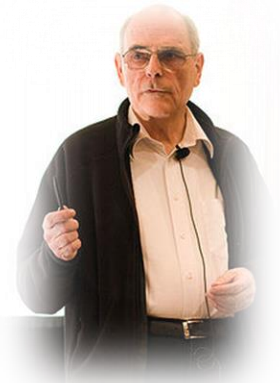
## Intuition:

$$ac + ad + bc + bd = (a + b)(c + d)$$

4 multiplications  
3 additions

➔

1 multiplication  
2 additions



# Strassen's Algorithm

▪  $C = A \times B$

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$

$$B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$C = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

$$C_{11} = M_1 + M_4 - M_5 + M_7 \quad \mathbf{2 + 1 -}$$

$$C_{12} = M_3 + M_5 \quad \mathbf{1 +}$$

$$C_{21} = M_2 + M_4 \quad \mathbf{1 +}$$

$$C_{22} = M_1 - M_2 + M_3 + M_6 \quad \mathbf{2 + 1 -}$$

$$M_1 = (A_{11} + A_{22})(B_{11} + B_{22}) \quad \mathbf{2 + 1 \times}$$

$$M_2 = (A_{21} + A_{22})B_{11} \quad \mathbf{1 + 1 \times}$$

$$M_3 = A_{11}(B_{12} - B_{22}) \quad \mathbf{1 - 1 \times}$$

$$M_4 = A_{22}(B_{21} - B_{11}) \quad \mathbf{1 - 1 \times}$$

$$M_5 = (A_{11} + A_{12})B_{22} \quad \mathbf{1 + 1 \times}$$

$$M_6 = (A_{21} - A_{11})(B_{11} + B_{12}) \quad \mathbf{1 + 1 - 1 \times}$$

$$M_7 = (A_{12} - A_{22})(B_{21} + B_{22}) \quad \mathbf{1 + 1 - 1 \times}$$

$$18\Theta((n/2)^2) + 7T(n/2)$$

---


$$\mathbf{12 + 6 - 7 \times}$$

# Verification of Strassen's Algorithm

$$\begin{aligned}C_{12} &= M_3 + M_5 \\ &= A_{11}(B_{12} - B_{22}) + (A_{11} + A_{12})B_{22} \\ &= A_{11}B_{12} + A_{12}B_{22}\end{aligned}$$

$$\begin{aligned}C_{21} &= M_2 + M_4 \\ &= (A_{21} + A_{22})B_{11} + A_{22}(B_{21} - B_{11}) \\ &= A_{21}B_{11} + A_{22}B_{21}\end{aligned}$$

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$

$$B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$C = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

- Practice

$$C_{11} = M_1 + M_4 - M_5 + M_7$$

$$C_{22} = M_1 - M_2 + M_3 + M_6$$

# Strassen's Algorithm Time Complexity

```
Strassen(n, A, B)
```

```
// base case
```

```
if n == 1
```

```
    return AB  $\Theta(1)$ 
```

```
// recursive case
```

Divide A and B into  $n/2$  by  $n/2$  submatrices **Divide**  $\Theta(1)$

```
 $M_1 = \text{Strassen}(n/2, A_{11}+A_{22}, B_{11}+B_{22})$ 
```

```
 $M_2 = \text{Strassen}(n/2, A_{21}+A_{22}, B_{11})$ 
```

```
 $M_3 = \text{Strassen}(n/2, A_{11}, B_{12}-B_{22})$ 
```

```
 $M_4 = \text{Strassen}(n/2, A_{22}, B_{21}-B_{11})$ 
```

```
 $M_5 = \text{Strassen}(n/2, A_{11}+A_{12}, B_{22})$ 
```

```
 $M_6 = \text{Strassen}(n/2, A_{11}-A_{21}, B_{11}+B_{12})$ 
```

```
 $M_7 = \text{Strassen}(n/2, A_{12}-A_{22}, B_{21}+B_{22})$ 
```

```
 $C_{11} = M_1 + M_4 - M_5 + M_7$ 
```

```
 $C_{12} = M_3 + M_5$ 
```

```
 $C_{21} = M_2 + M_4$ 
```

```
 $C_{22} = M_1 - M_2 + M_3 + M_6$ 
```

```
return C
```

**Conquer**

$$7T(n/2) + \Theta((n/2)^2)$$

- $T(n)$  = time for running `Strassen(n, A, B)`

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ 7T(n/2) + \Theta(n^2) & \text{if } n \geq 2 \end{cases}$$

$$\Rightarrow \Theta(n^{\log_2 7}) \sim \Theta(n^{2.807})$$



# Practicability of Strassen's Algorithm

- Disadvantages

1. Larger constant factor than it in the naïve approach

$$c_1 n^{\log_2 7}, c_2 n^3 \rightarrow c_1 > c_2$$

2. Less numerical stable than the naïve approach

- Larger errors accumulate in non-integer computation due to limited precision

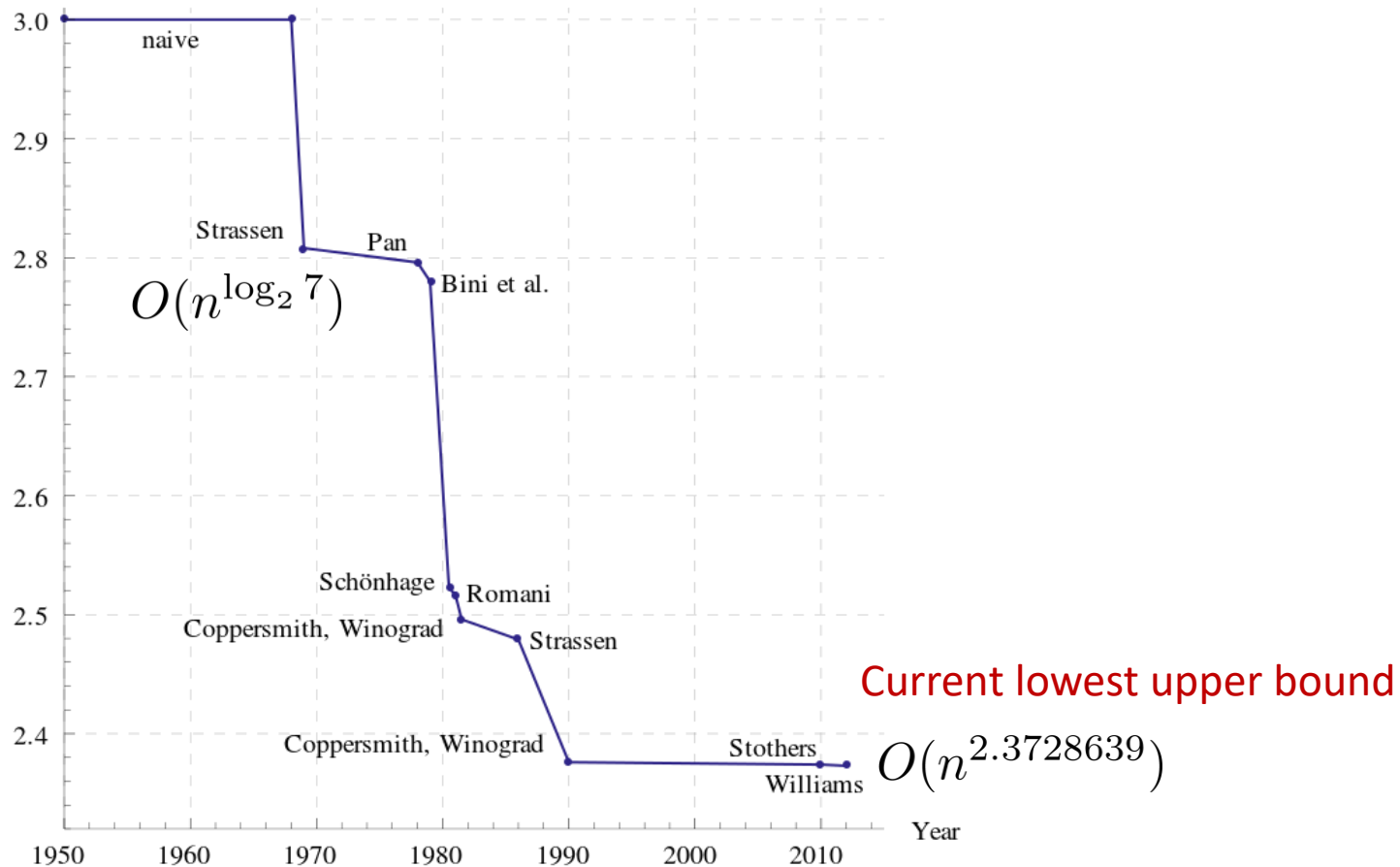
3. The submatrices at the levels of recursion consume space

4. Faster algorithms exist for sparse matrices

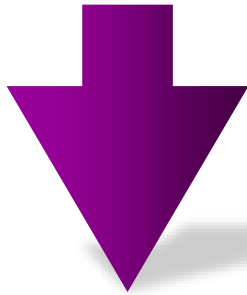
- Advantages: find the crossover point and combine two subproblems

# Matrix Multiplication Upper Bounds

- Each algorithm gives an upper bound



# Matrix Multi. Problem Complexity



Upper bound =  $O(n^{2.3728639})$



Lower bound =  $\Omega(n^2)$



# D&C #6: Selection Problem

Textbook Chapter 9.3 – Selection in worst-case linear time



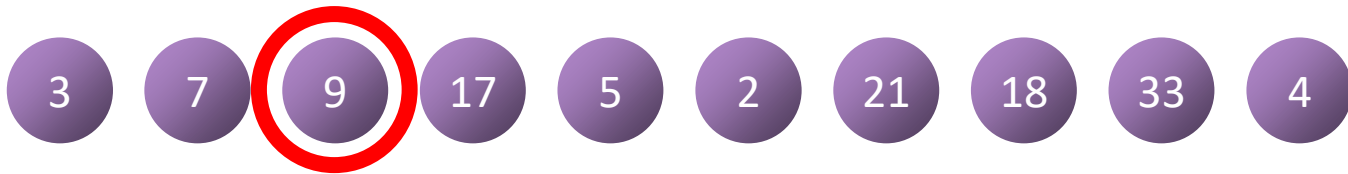
# Selection Problem

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- Input:
    - An array  $A$  of  $n$  distinct integers.
    - An index  $k$  with  $1 \leq k \leq n$ .
  - Output:

The  $k$ -th largest number in  $A$ .
-

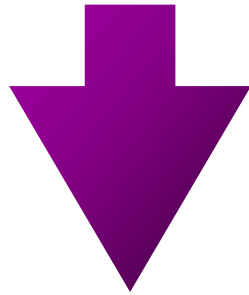
$n = 10, k = 5$



# Selection Problem $\leq$ Sorting Problem

- If the sorting problem can be solved in  $O(f(n))$ , so can the selection problem based on the algorithm design
  - Step 1: sort  $A$  into increasing order
  - Step 2: output  $A[n - k + 1]$

# Selection Problem Complexity



Upper bound =  $O(n \log n)$



Can we make the upper bound better if we do not sort them?



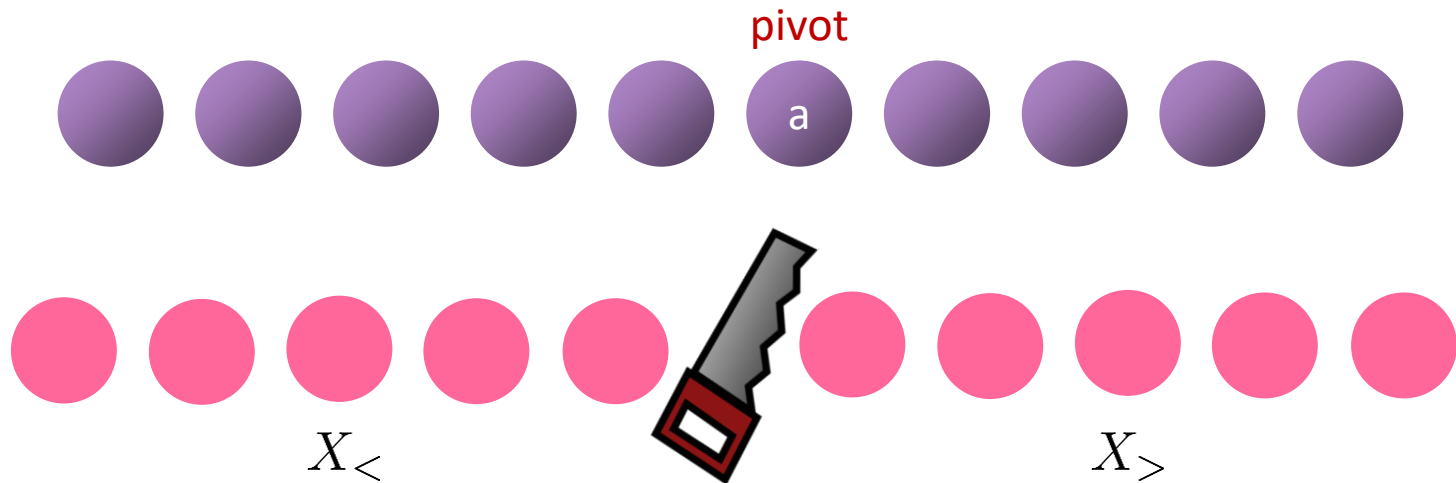
Lower bound =  $\Omega(n)$

# Hardness of Selection Problem

- Upper bounds in terms of #comparisons
  - $3n + o(n)$  by Schonhage, Paterson, and Pippenger (*JCSS* 1975).
  - $2.95n$  by Dor and Zwick (*SODA* 1995, *SIAM Journal on Computing* 1999).
- Lower bounds in terms of #comparisons
  - $2n + o(n)$  by Bent and John (*STOC* 1985)
  - $(2 + 2^{-80})n$  by Dor and Zwick (*FOCS* 1996, *SIAM Journal on Discrete Math* 2001).

# Divide-and-Conquer

- Idea
  - Select a pivot and divide the inputs into two subproblems
  - If  $k \leq |X_{>}|$ , we find the  $k$ -th largest
  - If  $k > |X_{>}|$ , we find the  $(k - |X_{>}|)$ -th largest



We want these subproblems to have similar size  
→ The better pivot is the medium in the input array

# Homework Practice

認真想一想!





# D&C #7: Closest Pair of Points Problem

Textbook Chapter 33.4 – Finding the closest pair of points



# Closest Pair of Points Problem

- Input:  $n \geq 2$  points, where  $p_i = (x_i, y_i)$  for  $0 \leq i < n$
- Output: two points  $p_i$  and  $p_j$  that are closest
  - “Closest”: smallest Euclidean distance
  - Euclidean distance between  $p_i$  and  $p_j$ :  $d(p_i, p_j) = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$



- Brute-force algorithm
  - Check all pairs of points:  
 $\Theta(C_2^n) = \Theta(n^2)$

# Closest Pair of Points Problem

- 1D:
  - Sort all points  $\Theta(n \log n)$
  - Scan the sorted points to find the closest pair in one pass  $\Theta(n)$ 
    - We only need to examine the adjacent points

➔  $T(n) = \Theta(n \log n)$



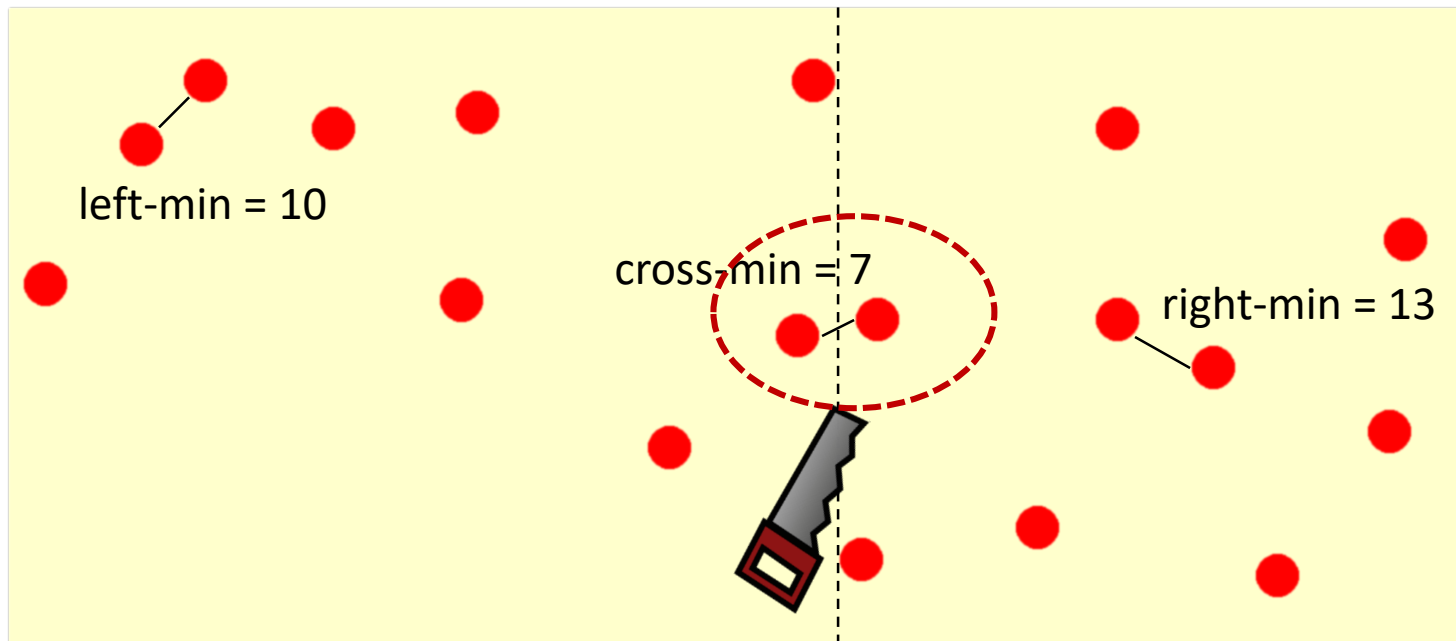
- 2D:

我想想~



# Divide-and-Conquer Algorithm

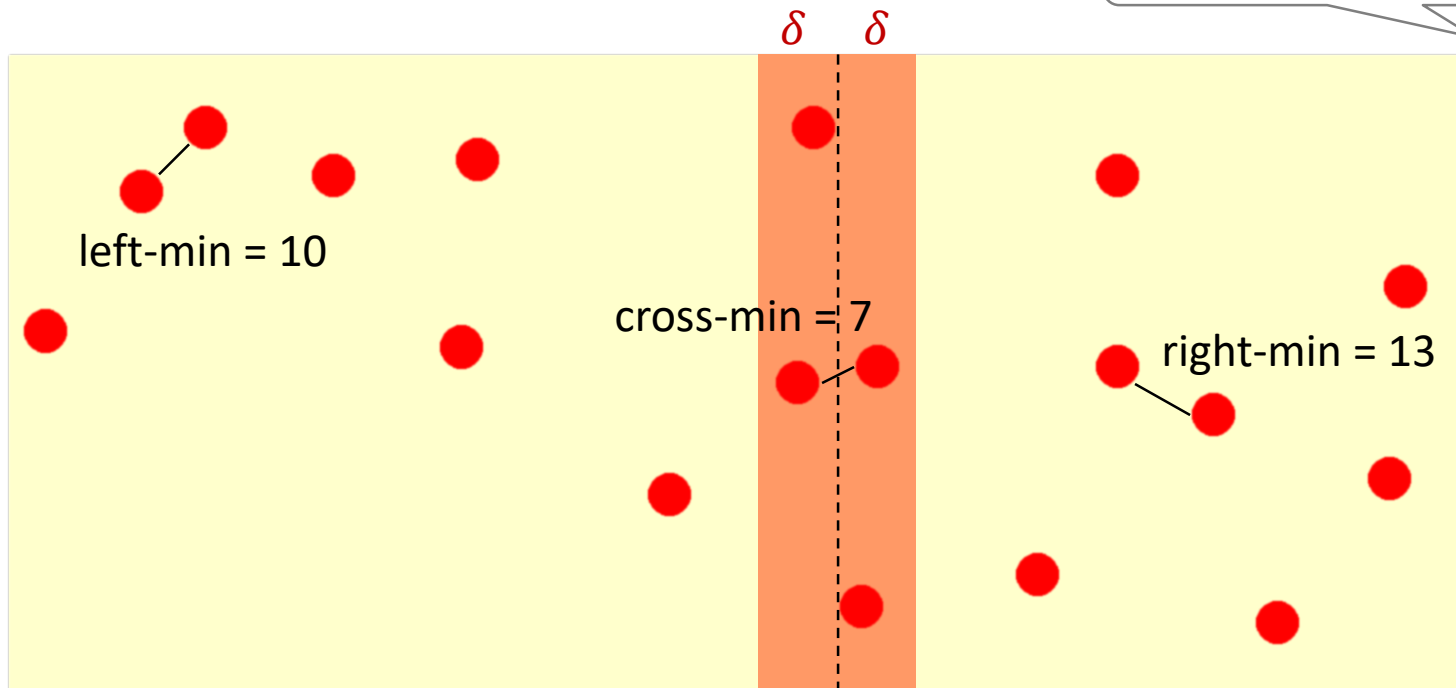
- **Divide**: divide points evenly along x-coordinate
- **Conquer**: find closest pair in each region recursively
- **Combine**: find closet pair with one point in each region, and return the best of three solutions



# Cross Two Regions

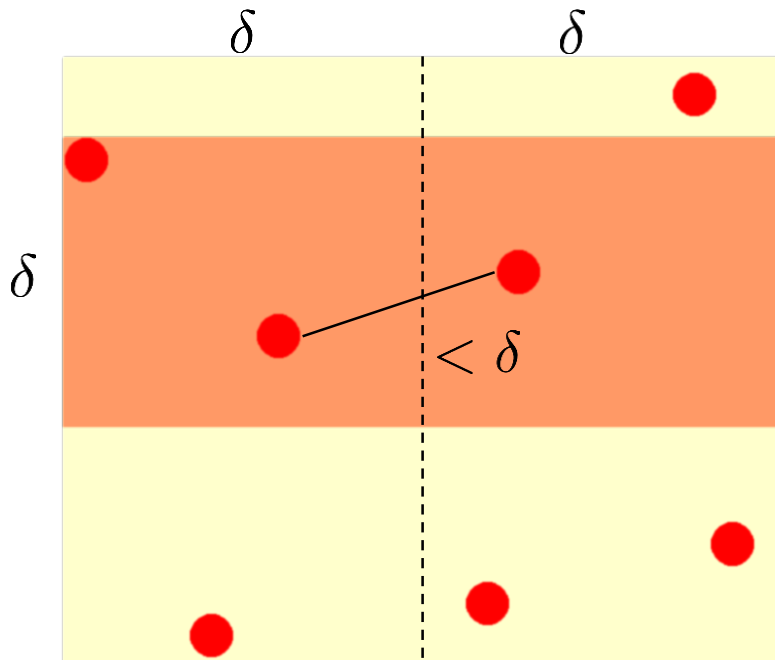
- Algo 1: check all pairs that cross two regions  $\rightarrow n/2 \times n/2$  combinations
- Algo 2: only consider points within  $\delta$  of the cut,  $\delta = \min\{l-\min, r-\min\}$ 
  - Other pairs of points must have distance larger than  $\delta$

縮小搜尋範圍!



# Cross Two Regions

- Algo 1: check all pairs that cross two regions  $\rightarrow n/2 \times n/2$  combinations
- Algo 2: only consider points within  $\delta$  of the cut,  $\delta = \min\{l-\min, r-\min\}$
- Algo 3: only consider pairs within  $\delta \times 2\delta$  blocks
  - Obs 1: every pair with smaller than  $\delta$  distance must appear in a  $\delta \times 2\delta$  block



縮小搜尋範圍!

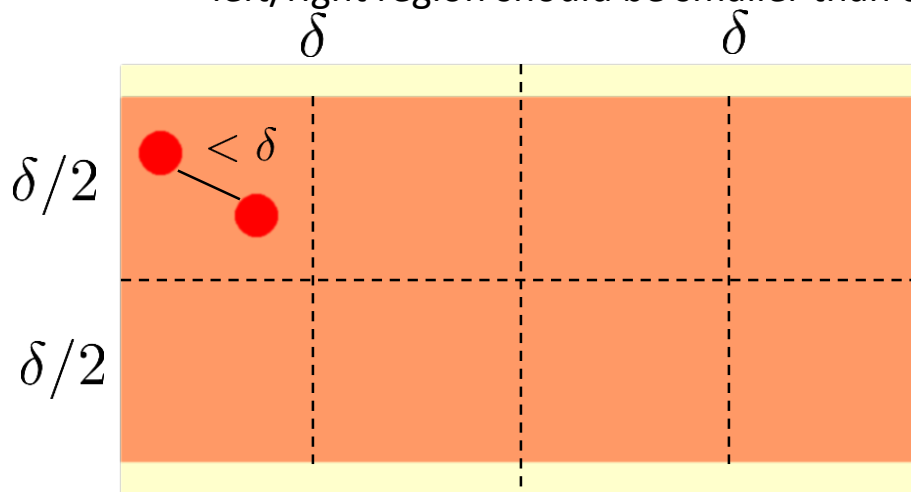


要是很倒霉，所有的點都聚集在某個  $\delta \times 2\delta$  區塊內怎麼辦



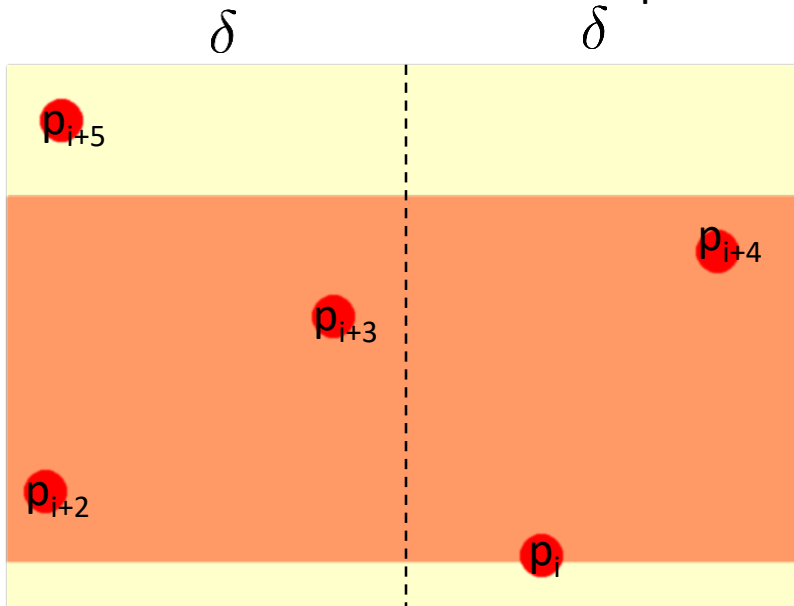
# Cross Two Regions

- Algo 1: check all pairs that cross two regions  $\rightarrow n/2 \times n/2$  combinations
- Algo 2: only consider points within  $\delta$  of the cut,  $\delta = \min\{l-\min, r-\min\}$
- Algo 3: only consider pairs within  $\delta \times 2\delta$  blocks
  - Obs 1: every pair with smaller than  $\delta$  distance must appear in a  $\delta \times 2\delta$  block
  - Obs 2: there are at most 8 points in a  $\delta \times 2\delta$  block
    - Each  $\delta/2 \times \delta/2$  block contains at most 1 point, otherwise the distance returned from left/right region should be smaller than  $\delta$



# Cross Two Regions

- Algo 1: check all pairs that cross two regions  $\rightarrow n/2 \times n/2$  combinations
- Algo 2: only consider points within  $\delta$  of the cut,  $\delta = \min\{l-\min, r-\min\}$
- Algo 3: only consider pairs within  $\delta \times 2\delta$  blocks
  - Obs 1: every pair with smaller than  $\delta$  distance must appear in a  $\delta \times 2\delta$  block
  - Obs 2: there are at most 8 points in a  $\delta \times 2\delta$  block



## Find-closest-pair-across-regions

1. Sort the points by y-values within  $\delta$  of the cut (yellow region)
2. For the sorted point  $p_i$ , compute the distance with  $p_{i+1}, p_{i+2}, \dots, p_{i+7}$
3. Return the smallest one

At most 7 distance calculations needed

# Algorithm Complexity

```
Closest-Pair(P)
// termination condition (base case)
if |P| <= 3 brute-force finding closest pair and return it
// Divide
find a vertical line L s.t. both planes contain half of the points
// Conquer (by recursion)
left-pair, left-min = Closest-Pair(points in the left)
right-pair, right-min = Closest-Pair(points in the right)
// Combine
delta = min{left-min, right-min}
remove points that are delta or more away from L // Obs 1
sort remaining points by y-coordinate into p0, ..., pk
for point pi:
    compute distances with pi+1, pi+2, ..., pi+7 // Obs 2
    update delta if a closer pair is found
return the closest pair and its distance
```

$\Theta(1)$

$\Theta(n \log n)$

$2T(n/2)$

$\Theta(n \log n)$

$\Theta(n)$

- $T(n)$  = time for running `Closest-Pair(P)` with  $|P| = n$

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq 3 \\ 2T\left(\frac{n}{2}\right) + \Theta(n \log n) & \text{if } n > 3 \end{cases} \Rightarrow T(n) = \Theta(n \log^2 n)$$

Exercise 4.6-2



# Preprocessing

- Idea: do not sort inside the recursive case

Closest-Pair(P)

```
sort P by x- and y-coordinate and store in Px and Py            $\Theta(n \log n)$ 
// termination condition (base case)
if |P| <= 3 brute-force finding closest pair and return it     $\Theta(1)$ 
// Divide
find a vertical line L s.t. both planes contain half of the points  $\Theta(n)$ 
// Conquer (by recursion)
left-pair, left-min = Closest-Pair(points in the left)          $2T(n/2)$ 
right-pair, right-min = Closest-Pair(points in the right)
// Combine
delta = min{left-min, right-min}
remove points that are delta or more away from L // Obs 1
for point  $p_i$  in sorted candidates                              $\Theta(n)$ 
    compute distances with  $p_{i+1}, p_{i+2}, \dots, p_{i+7}$  // Obs 2
    update delta if a closer pair is found
return the closest pair and its distance
```

$$T'(n) = \begin{cases} \Theta(1) & \text{if } n \leq 3 \\ 2T'(\frac{n}{2}) + \Theta(n) & \text{if } n > 3 \end{cases} \rightarrow \begin{cases} T'(n) = \Theta(n \log n) \\ T(n) = \Theta(n \log n) \end{cases}$$

# Closest Pair of Points Problem

- $O(n)$  algorithm
  - Taking advantage of randomization
    - Chapter 13.7 of Algorithm Design by Kleinberg & Tardos
    - Samir Khuller and Yossi Matias. 1995. A simple randomized sieve algorithm for the closest-pair problem. Inf. Comput. 118, 1 (April 1995), 34-37.

# Concluding Remarks

- When to use D&C
  - Whether the problem with small inputs can be solved directly
  - Whether subproblem solutions can be combined into the original solution
  - Whether the overall complexity is better than naïve
- Note
  - Try different ways of dividing
  - D&C may be suboptimal due to repetitive computations
  - Example.
    - D&C algo for Fibonacci:  $\Omega\left(\left(\frac{1+\sqrt{5}}{2}\right)^n\right)$
    - Bottom-up algo for Fibonacci:  $\Theta(n)$

Our next topic: **Dynamic Programming**  
“a technique for solving problems with overlapping subproblems”

```
Fibonacci(n)
  if n < 2
    return 1
  a[0]=1
  a[1]=1
  for i = 2 ... n
    a[i]=a[i-1]+a[i-2]
  return a[n]
```

1. Divide

2. Conquer

3. Combine



# Question?

Important announcement will be sent to @ntu.edu.tw mailbox  
& post to the course website

Course Website: <http://ada.miulab.tw>

Email: [ada-ta@csie.ntu.edu.tw](mailto:ada-ta@csie.ntu.edu.tw)