



中山大學  
SUN YAT-SEN UNIVERSITY

# Lecture 3

## Divide-and-Conquer

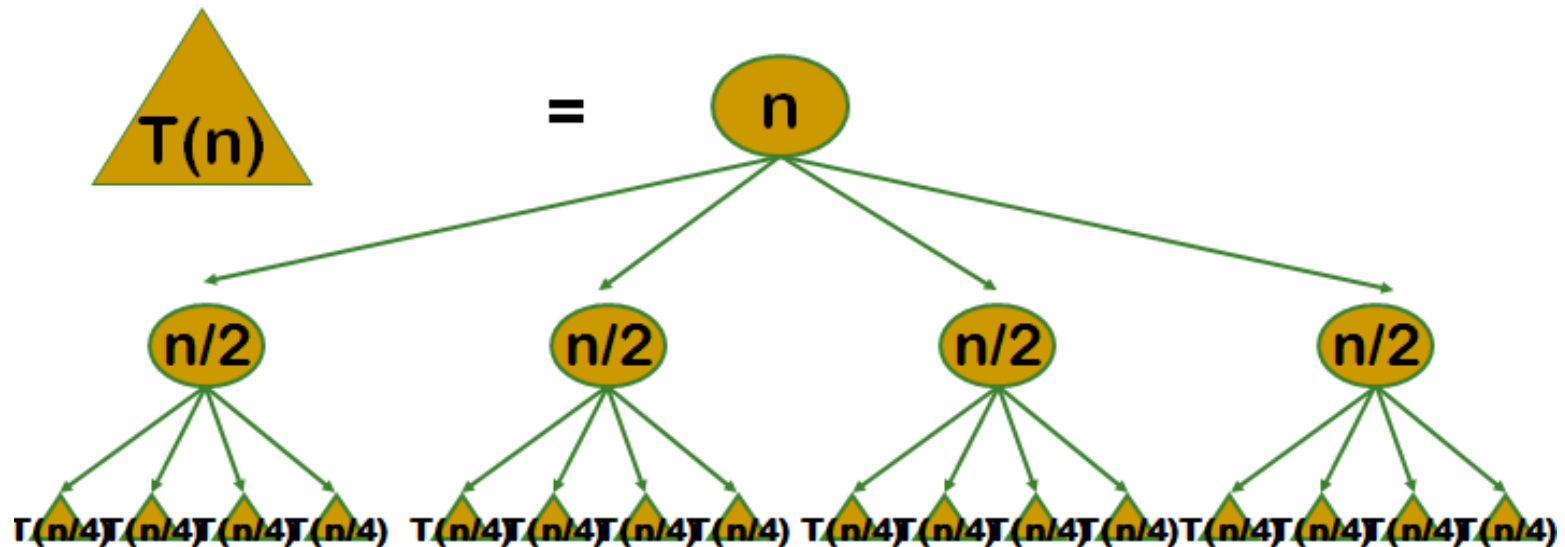
Algorithm Design

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# Definition

- A divide and conquer algorithm works by recursively breaking down a problem into two or more sub-problems of the same or related type, until these become simple enough to be solved directly. The solutions to the sub-problems are then combined to give a solution to the original problem.



# Merge-sort

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```

procedure MERGE-SORT( $A, p, r$ )
  if  $p < r$ 
    then  $q \leftarrow \lfloor (p + r)/2 \rfloor$ 
      MERGE-SORT( $A, p, q$ )
      MERGE-SORT( $A, q + 1, r$ )
      MERGE( $A, p, q, r$ )

```

```

procedure MERGE( $A, p, q, r$ )
   $n_1 \leftarrow q - p + 1$ ;  $n_2 \leftarrow r - q$ 
  allocate arrays  $L[1 \dots n_1 + 1]$  and  $R[1 \dots n_2 + 1]$ 
  for  $i \leftarrow 1$  to  $n_1$ 
    do  $L[i] \leftarrow A[p + i - 1]$ 
  for  $j \leftarrow 1$  to  $n_2$ 
    do  $R[j] \leftarrow A[q + j]$ 
   $L[n_1 + 1] \leftarrow \infty$ ;  $R[n_2 + 1] \leftarrow \infty$ 
   $i \leftarrow 1$ ;  $j \leftarrow 1$ 
  for  $k \leftarrow p$  to  $r$ 
    do if  $L[i] \leq R[j]$ 
      then  $A[k] \leftarrow L[i]$ 
         $i \leftarrow i + 1$ 
      else  $A[k] \leftarrow R[j]$ 
         $j \leftarrow j + 1$ 

```

# Analysis of the Merge-sort algorithm

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- Described by recursive equation
- Suppose  $T(n)$  is the running time on a problem of size  $n$ .
- $$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq n_c \\ aT(n/b) + D(n) + C(n) & \text{if } n > n_c \end{cases}$$

where  $a$ : number of subproblems

$n/b$ : size of each subproblem

$D(n)$ : cost of divide operation

$C(n)$ : cost of combination operation

# Analysis of the Merge-sort algorithm

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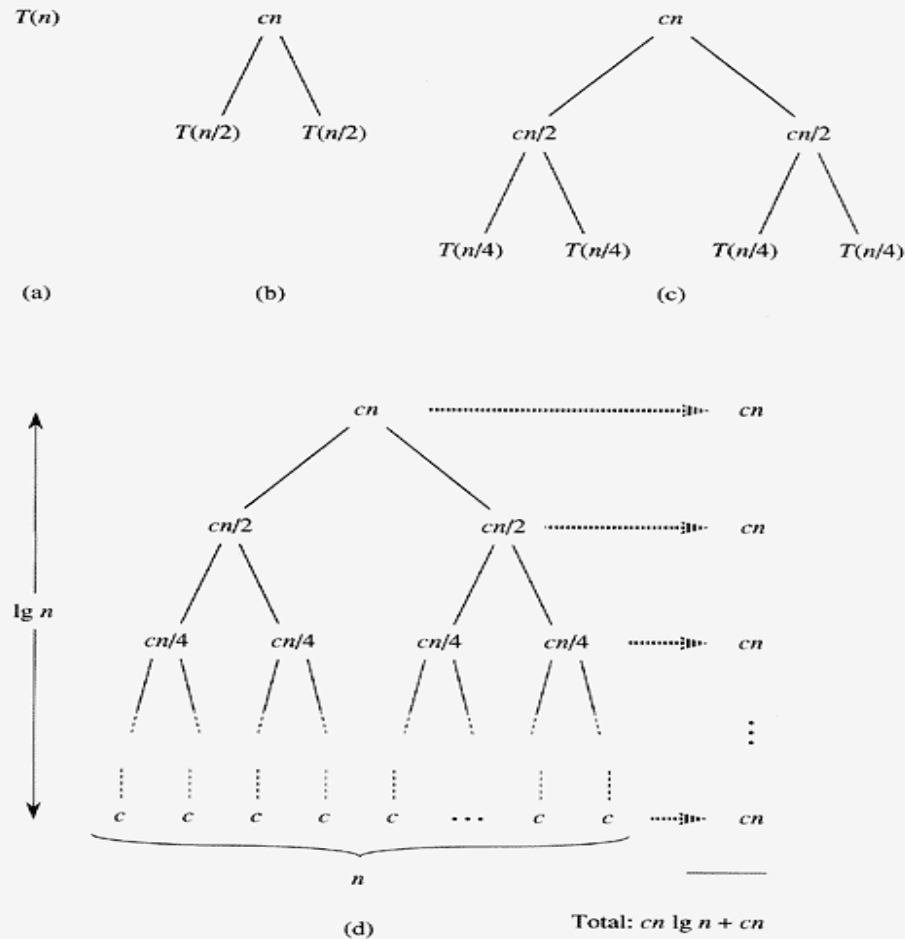
- **Divide:**  $D(n) = \Theta(1)$
- **Conquer:**  $a=2, b=2$ , so  $2T(n/2)$
- **Combine:**  $C(n) = \Theta(n)$
- $$T(n) = \begin{cases} \Theta(1) & \text{if } n=1 \\ 2T(n/2) + \Theta(n) & \text{if } n>1 \end{cases}$$
- $$T(n) = \begin{cases} c & \text{if } n=1 \\ 2T(n/2) + cn & \text{if } n>1 \end{cases}$$

# Analysis of the Merge-sort algorithm

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- The recursive equation can be solved by recursive tree.
- $T(n) = 2T(n/2) + cn$ ,
- $\lg n + 1$  levels,  $cn$  at each level, thus
- Total cost for merge sort is:  
 $T(n) = cn \lg n + cn = \Theta(n \lg n)$ .
- Question: best, worst, average?

# Analysis of the Merge-sort algorithm



**Figure 2.5** The construction of a recursion tree for the recurrence  $T(n) = 2T(n/2) + cn$ . Part (a) shows  $T(n)$ , which is progressively expanded in (b)–(d) to form the recursion tree. The fully expanded tree in part (d) has  $\lg n + 1$  levels (i.e., it has height  $\lg n$ , as indicated), and each level contributes a total cost of  $cn$ . The total cost, therefore, is  $cn \lg n + cn$ , which is  $\Theta(n \lg n)$ .

# Master theorem

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## *Theorem 4.1 (Master theorem)*

Let  $a \geq 1$  and  $b > 1$  be constants, let  $f(n)$  be a function, and let  $T(n)$  be defined on the nonnegative integers by the recurrence

$$T(n) = aT(n/b) + f(n) ,$$

where we interpret  $n/b$  to mean either  $\lfloor n/b \rfloor$  or  $\lceil n/b \rceil$ . Then  $T(n)$  has the following asymptotic bounds:

1. If  $f(n) = O(n^{\log_b a - \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$ .
2. If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$ .
3. If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ , and if  $af(n/b) \leq cf(n)$  for some constant  $c < 1$  and all sufficiently large  $n$ , then  $T(n) = \Theta(f(n))$ . ■



# Exercise

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Use the master method to give tight asymptotic bounds for the following recurrences.

(a)  $T(n) = 2T(n/4) + 1$

(b)  $T(n) = 2T(n/4) + \sqrt{n}$

(c)  $T(n) = 2T(n/4) + n$

(d)  $T(n) = 2T(n/4) + n^2$

# Multiplication of Large Integers

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Consider the problem of multiplying two (large)  $n$ -digit integers represented by arrays of their digits such as:

$$A = 12345678901357986429 \quad B = 87654321284820912836$$

The grade-school algorithm:

$$\begin{array}{r}
 a_1 \ a_2 \ \dots \ a_n \\
 b_1 \ b_2 \ \dots \ b_n \\
 \hline
 (d_{10}) \ d_{11} \ d_{12} \ \dots \ d_{1n} \\
 (d_{20}) \ d_{21} \ d_{22} \ \dots \ d_{2n} \\
 \dots \ \dots \ \dots \ \dots \ \dots \ \dots \\
 \hline
 (d_{n0}) \ d_{n1} \ d_{n2} \ \dots \ d_{nn}
 \end{array}$$

Efficiency:  $\Theta(n^2)$  single-digit multiplications

# First Divide-and-Conquer Algorithm

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A small example:  $A * B$  where  $A = 2135$  and  $B = 4014$

$$A = (21 \cdot 10^2 + 35), \quad B = (40 \cdot 10^2 + 14)$$

$$\begin{aligned} \text{So, } A * B &= (21 \cdot 10^2 + 35) * (40 \cdot 10^2 + 14) \\ &= 21 * 40 \cdot 10^4 + (21 * 14 + 35 * 40) \cdot 10^2 + 35 * 14 \end{aligned}$$

In general, if  $A = A_1A_2$  and  $B = B_1B_2$  (where  $A$  and  $B$  are  $n$ -digit,  $A_1, A_2, B_1, B_2$  are  $n/2$ -digit numbers),

$$A * B = A_1 * B_1 \cdot 10^n + (A_1 * B_2 + A_2 * B_1) \cdot 10^{n/2} + A_2 * B_2$$

Recurrence for the number of one-digit multiplications  $M(n)$ :

$$M(n) = 4M(n/2), \quad M(1) = 1$$

Solution:  $M(n) = n^2$

## Second Divide-and-Conquer Algorithm

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$$A * B = A_1 * B_1 \cdot 10^n + (A_1 * B_2 + A_2 * B_1) \cdot 10^{n/2} + A_2 * B_2$$

The idea is to decrease the number of multiplications from 4 to 3:

$$(A_1 + A_2) * (B_1 + B_2) = A_1 * B_1 + (A_1 * B_2 + A_2 * B_1) + A_2 * B_2,$$

i.e.,  $(A_1 * B_2 + A_2 * B_1) = (A_1 + A_2) * (B_1 + B_2) - A_1 * B_1 - A_2 * B_2$ , which requires only 3 multiplications at the expense of 3 extra add/sub.

Recurrence for the number of multiplications  $M(n)$ :

$$M(n) = 3M(n/2), \quad M(1) = 1$$

Solution:  $M(n) = 3^{\log_2 n} = n^{\log_2 3} \approx n^{1.585}$

# Karatsuba Multiplication Algorithm

KARATSUBA-MULTIPLY( $x, y, n$ )

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IF ( $n = 1$ )

    RETURN  $x \times y$ .

ELSE

$m \leftarrow \lceil n / 2 \rceil$ .

$a \leftarrow \lfloor x / 2^m \rfloor$ ;  $b \leftarrow x \bmod 2^m$ .

$c \leftarrow \lfloor y / 2^m \rfloor$ ;  $d \leftarrow y \bmod 2^m$ .

$e \leftarrow \text{KARATSUBA-MULTIPLY}(a, c, m)$ .

$f \leftarrow \text{KARATSUBA-MULTIPLY}(b, d, m)$ .

$g \leftarrow \text{KARATSUBA-MULTIPLY}(a - b, c - d, m)$ .

    RETURN  $2^{2m} e + 2^m (e + f - g) + f$ .

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# Example of Large-Integer Multiplication

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$$2135 * 4014$$

$$= (21 * 10^2 + 35) * (40 * 10^2 + 14)$$

$$= (21 * 40) * 10^4 + c1 * 10^2 + 35 * 14$$

where  $c1 = (21 + 35) * (40 + 14) - 21 * 40 - 35 * 14$ , and

$$21 * 40 = (2 * 10 + 1) * (4 * 10 + 0)$$

$$= (2 * 4) * 10^2 + c2 * 10 + 1 * 0$$

where  $c2 = (2 + 1) * (4 + 0) - 2 * 4 - 1 * 0$ , etc.

- This process requires 9 digit multiplications as opposed to 16.

# Matrix multiplication

- Given two  $n$ -by- $n$  matrices  $A$  and  $B$ , compute  $C = AB$ .
- Grade-school.  $\Theta(n^3)$  arithmetic operations.

$$\begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \times \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{nn} \end{bmatrix}$$

$$\begin{bmatrix} .59 & .32 & .41 \\ .31 & .36 & .25 \\ .45 & .31 & .42 \end{bmatrix} = \begin{bmatrix} .70 & .20 & .10 \\ .30 & .60 & .10 \\ .50 & .10 & .40 \end{bmatrix} \times \begin{bmatrix} .80 & .30 & .50 \\ .10 & .40 & .10 \\ .10 & .30 & .40 \end{bmatrix}$$

# Block matrix multiplication

$$\begin{bmatrix} 152 & 158 & 164 & 170 \\ 504 & 526 & 548 & 570 \\ 856 & 894 & 932 & 970 \\ 1208 & 1262 & 1316 & 1370 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 2 & 3 \\ 4 & 5 & 6 & 7 \\ 8 & 9 & 10 & 11 \\ 12 & 13 & 14 & 15 \end{bmatrix} \times \begin{bmatrix} 16 & 17 & 18 & 19 \\ 20 & 21 & 22 & 23 \\ 24 & 25 & 26 & 27 \\ 28 & 29 & 30 & 31 \end{bmatrix}$$

$$C_{11} = A_{11} \times B_{11} + A_{12} \times B_{21} = \begin{bmatrix} 0 & 1 \\ 4 & 5 \end{bmatrix} \times \begin{bmatrix} 16 & 17 \\ 20 & 21 \end{bmatrix} + \begin{bmatrix} 2 & 3 \\ 6 & 7 \end{bmatrix} \times \begin{bmatrix} 24 & 25 \\ 28 & 29 \end{bmatrix} = \begin{bmatrix} 152 & 158 \\ 504 & 526 \end{bmatrix}$$



# Matrix multiplication

- To multiply two  $n$ -by- $n$  matrices  $A$  and  $B$ :
  - Divide: partition  $A$  and  $B$  into  $1/2n$ -by- $1/2n$  blocks.
  - Conquer: multiply 8 pairs of  $1/2n$ -by- $1/2n$  matrices, recursively.
  - Combine: add appropriate products using 4 matrix additions.

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{aligned} C_{11} &= (A_{11} \times B_{11}) + (A_{12} \times B_{21}) \\ C_{12} &= (A_{11} \times B_{12}) + (A_{12} \times B_{22}) \\ C_{21} &= (A_{21} \times B_{11}) + (A_{22} \times B_{21}) \\ C_{22} &= (A_{21} \times B_{12}) + (A_{22} \times B_{22}) \end{aligned}$$

- Running time

$$T(n) = \underbrace{8T(n/2)}_{\text{recursive calls}} + \underbrace{\Theta(n^2)}_{\text{add, form submatrices}} \Rightarrow T(n) = \Theta(n^3)$$

# Strassen's method

- **Key idea:** multiply 2-by-2 blocks with only 7 multiplications. (plus 11 additions and 7 subtractions)

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$C_{11} = P_5 + P_4 - P_2 + P_6$$

$$C_{12} = P_1 + P_2$$

$$C_{21} = P_3 + P_4$$

$$C_{22} = P_1 + P_5 - P_3 - P_7$$

$$P_1 \leftarrow A_{11} \times (B_{12} - B_{22})$$

$$P_2 \leftarrow (A_{11} + A_{12}) \times B_{22}$$

$$P_3 \leftarrow (A_{21} + A_{22}) \times B_{11}$$

$$P_4 \leftarrow A_{22} \times (B_{21} - B_{11})$$

$$P_5 \leftarrow (A_{11} + A_{22}) \times (B_{11} + B_{22})$$

$$P_6 \leftarrow (A_{12} - A_{22}) \times (B_{21} + B_{22})$$

$$P_7 \leftarrow (A_{11} - A_{21}) \times (B_{11} + B_{12})$$

Pf.  $C_{12} = P_1 + P_2$

$$= A_{11} \times (B_{12} - B_{22}) + (A_{11} + A_{12}) \times B_{22}$$

$$= A_{11} \times B_{12} + A_{12} \times B_{22}. \quad \checkmark$$

# Strassen's algorithm

assume  $n$  is  
a power of 2

STRASSEN( $n, A, B$ )

IF ( $n = 1$ ) RETURN  $A \times B$ .

Partition  $A$  and  $B$  into 2-by-2 block matrices.

$P_1 \leftarrow \text{STRASSEN}(n/2, A_{11}, (B_{12} - B_{22}))$ .

$P_2 \leftarrow \text{STRASSEN}(n/2, (A_{11} + A_{12}), B_{22})$ .

$P_3 \leftarrow \text{STRASSEN}(n/2, (A_{21} + A_{22}), B_{11})$ .

$P_4 \leftarrow \text{STRASSEN}(n/2, A_{22}, (B_{21} - B_{11}))$ .

$P_5 \leftarrow \text{STRASSEN}(n/2, (A_{11} + A_{22}) \times (B_{11} + B_{22}))$ .

$P_6 \leftarrow \text{STRASSEN}(n/2, (A_{12} - A_{22}) \times (B_{21} + B_{22}))$ .

$P_7 \leftarrow \text{STRASSEN}(n/2, (A_{11} - A_{21}) \times (B_{11} + B_{12}))$ .

$C_{11} = P_5 + P_4 - P_2 + P_6$ .

$C_{12} = P_1 + P_2$ .

$C_{21} = P_3 + P_4$ .

$C_{22} = P_1 + P_5 - P_3 - P_7$ .

RETURN  $C$ .

keep track of indices of submatrices  
(don't copy matrix entries)

# Analysis of Strassen's algorithm

- **Theorem.** Strassen's algorithm requires  $O(n^{2.81})$  arithmetic operations to multiply two  $n$ -by- $n$  matrices.

$$T(n) = \underbrace{7T(n/2)}_{\text{recursive calls}} + \underbrace{\Theta(n^2)}_{\text{add, subtract}} \Rightarrow T(n) = \Theta(n^{\log_2 7}) = O(n^{2.81})$$

- Q. What if  $n$  is not a power of 2 ?
- A. Could pad matrices with zeros.

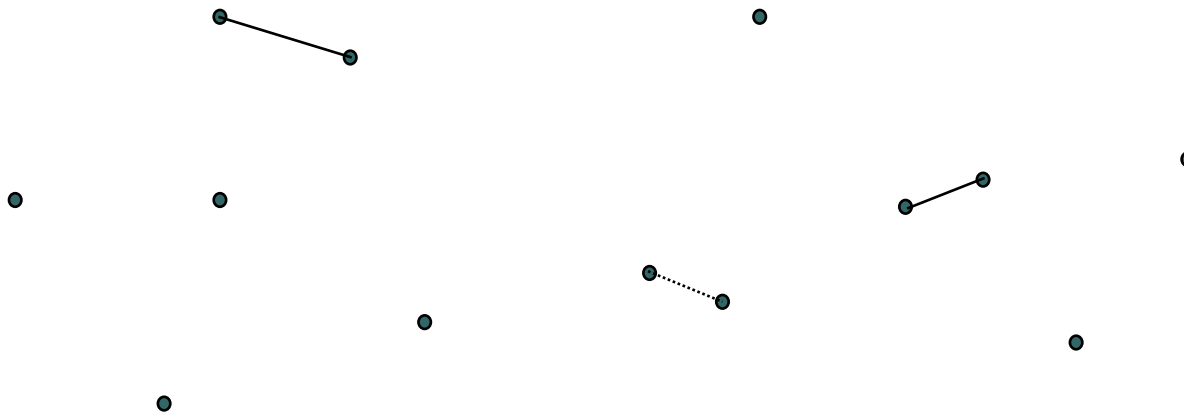
$$\begin{bmatrix} 1 & 2 & 3 & 0 \\ 4 & 5 & 6 & 0 \\ 7 & 8 & 9 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{bmatrix} 10 & 11 & 12 & 0 \\ 13 & 14 & 15 & 0 \\ 16 & 17 & 18 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 84 & 90 & 96 & 0 \\ 201 & 216 & 231 & 0 \\ 318 & 342 & 366 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

# Closest Pair

- Given a set  $S = \{p_1, p_2, \dots, p_n\}$  of  $n$  points in the plane find the two points of  $S$  whose distance is the smallest.
- 1-d



- 2-d



# Closest Pair – Naïve Algorithm

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Pseudo code

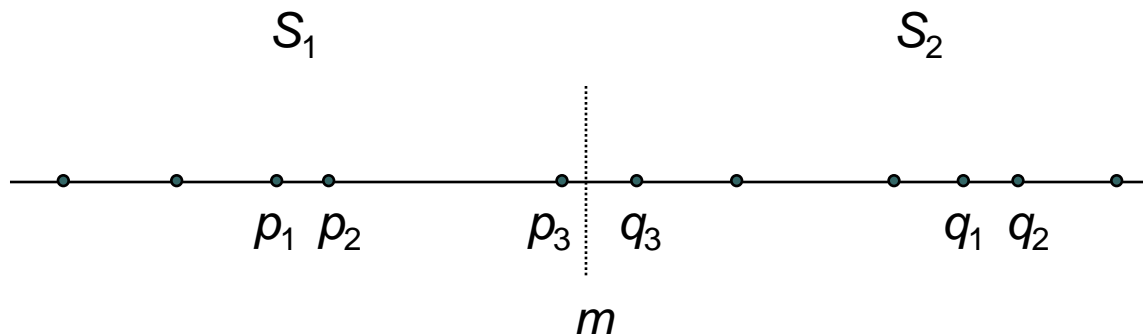
```
for each pt  $i \in S$ 
  for each pt  $j \in S$  and  $i \neq j$ 
  {
    compute distance of  $i, j$ 
    if distance of  $i, j < \text{min\_dist}$ 
       $\text{min\_dist} = \text{distance } i, j$ 
  }
return  $\text{min\_dist}$ 
```

- Time Complexity–  $O(n^2)$
- Can we do better?

# 1-D Closest Pair – Divide & Conquer

- We consider a divide-and-conquer algorithm for CLOSEST-PAIR in 1 dimension ( $d = 1$ ).
- Partition  $S$ , a set of points on a line, into two sets  $S_1$  and  $S_2$  at some point  $m$  such that for every point  $p \in S_1$  and  $q \in S_2$ ,  $p < q$ .
- Solving CLOSEST-PAIR recursively on  $S_1$  and  $S_2$  separately produces  $\{p_1, p_2\}$ , the closest pair in  $S_1$ , and  $\{q_1, q_2\}$ , the closest pair in  $S_2$ .
- Let  $\delta$  be the smallest distance found so far:  

$$\delta = \min(|p_2 - p_1|, |q_2 - q_1|)$$
- The closest pair in  $S$  is either  $\{p_1, p_2\}$  or  $\{q_1, q_2\}$  or some  $\{p_3, q_3\}$  with  $p_3 \in S_1$  and  $q_3 \in S_2$ .



# 1-D Closest Pair – Divide & Conquer

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- To check for such a point  $\{p_3, q_3\}$ , is it necessary to test every possible pair of points in  $S_1$  and  $S_2$ ?
- Note that if  $\{p_3, q_3\}$  is to be closer than  $\delta$  (i.e.,  $|q_3 - p_3| < \delta$ ), then both  $p_3$  and  $q_3$  must be within  $\delta$  of  $m$ .
- Because  $\delta$  is the distance between the closest pair in either  $S_1$  or  $S_2$ , a semi-closed interval of length  $\delta$  can contain at most 1 point.
- For the same reason, there can be at most 1 point of  $S_2$  within  $\delta$  of  $m$ .
- So, the number of distance computations needed to check for a closest pair  $\{p_3, q_3\}$  with  $p_3 \in S_1$  and  $q_3 \in S_2$  is 1, not  $O(N^2)$ .
- Thus a divide-and-conquer algorithm can solve 1-dimensional CLOSEST-PAIR in  $O(N \log N)$  time.



# 1-D Closest Pair – Divide & Conquer

## Divide-and-conquer for $d = 1$

procedure CPAIR1( $S$ )

Input:  $X[1:N]$ ,  $N$  points of  $S$  in one dimension.

Output:  $\delta$ , the distance between the two closest points.

```

1  begin
2      if ( $|S| = 2$ ) then
3           $\delta = |X[2] - X[1]|$ 
4      else if ( $|S| = 1$ ) then
5           $\delta = \infty$ 
6      else
7          begin
8              Construct( $S_1, S_2$ ) /*  $S_1 = \{p: p \leq m\}, S_2 = \{p: p > m\}$  */
9               $\delta_1 = \text{CPAIR1}(S_1)$ 
10              $\delta_2 = \text{CPAIR1}(S_2)$ 
11              $p = \max(S_1)$ 
12              $q = \min(S_2)$ 
13              $\delta = \min(\delta_1, \delta_2, q - p)$ 
14         end
15     endif
16     return  $\delta$ 
17 end
```

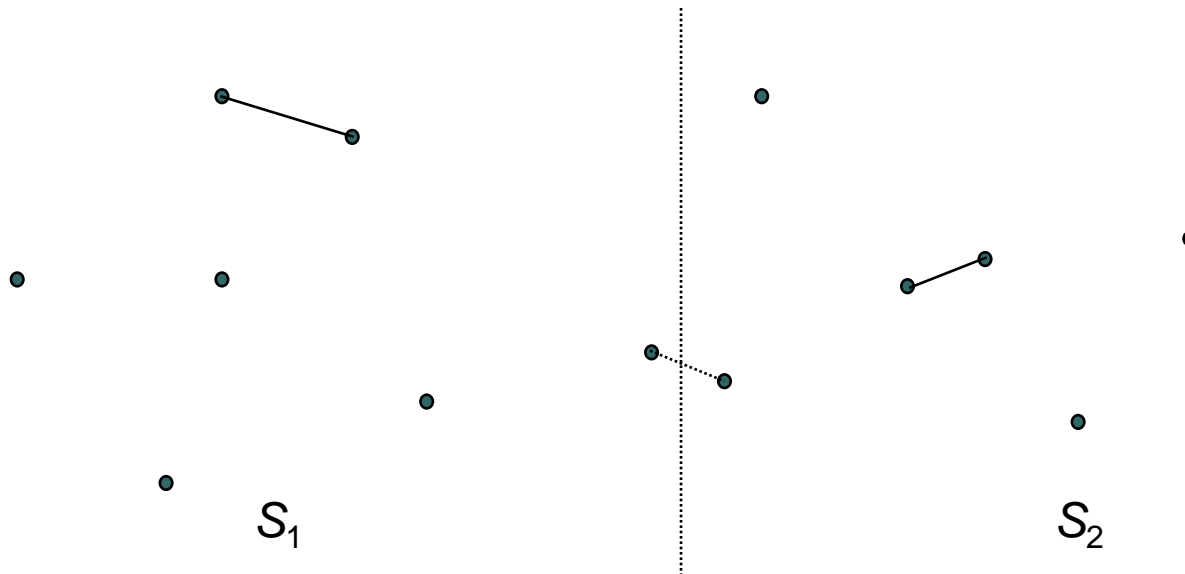
# Closest Pair – Divide & Conquer

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- Divide the problem into two equal-sized sub problems
- Solve those sub problems recursively
- Merge the sub problem solutions into an overall solution

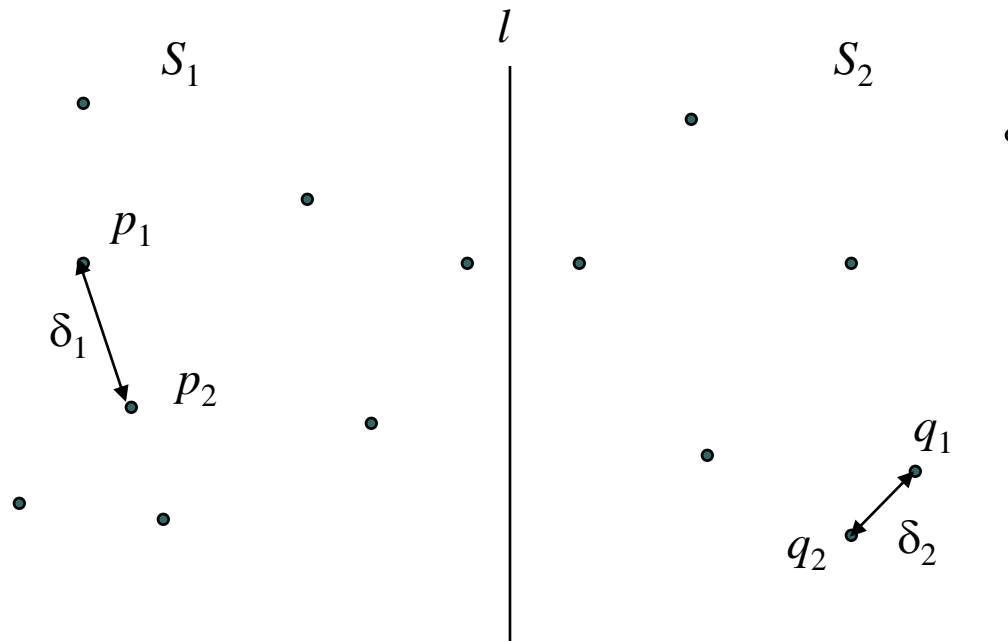
# Closest Pair – Divide & Conquer

- Assume that we have solutions for sub problems  $S_1$ ,  $S_2$ .
- How can we merge in a time-efficient way?
  - The closest pair can consist of one point from  $S_1$  and another from  $S_2$
  - Testing all possibilities requires:  $O(n/2) \cdot O(n/2) \in O(n^2)$
  - Not good enough



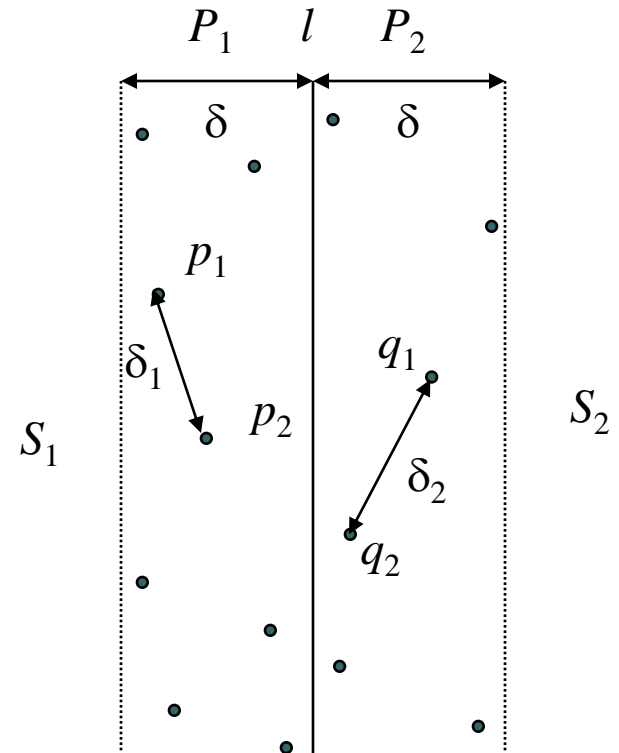
# Closest Pair – Divide & Conquer

- Partition two dimensional set  $S$  into subsets  $S_1$  and  $S_2$  by a vertical line  $l$  at the median  $x$  coordinate of  $S$ .
- Solve the problem recursively on  $S_1$  and  $S_2$ .
- Let  $\{p_1, p_2\}$  be the closest pair in  $S_1$  and  $\{q_1, q_2\}$  in  $S_2$ .
- Let  $\delta_1 = \text{distance}(p_1, p_2)$  and  $\delta_2 = \text{distance}(q_1, q_2)$
- Let  $\delta = \min(\delta_1, \delta_2)$



# Closest Pair – Divide & Conquer

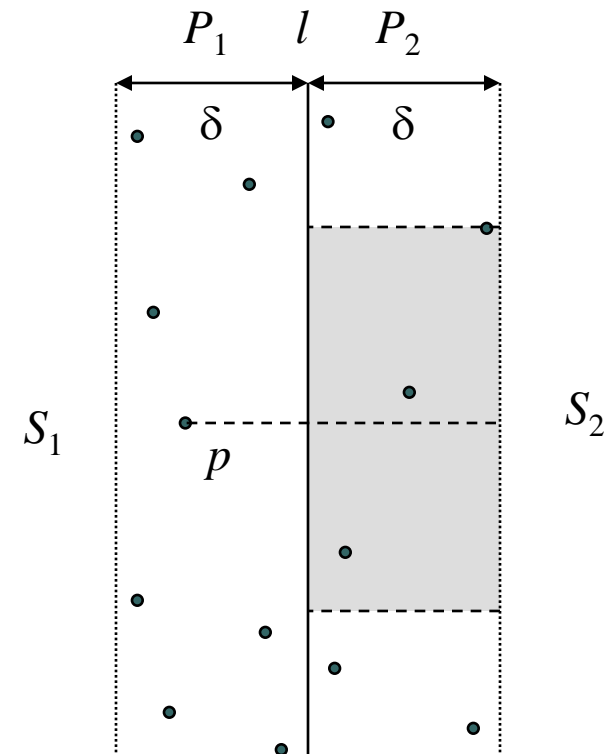
- In order to merge we have to determine if exists a pair of points  $\{p, q\}$  where  $p \in S_1$ ,  $q \in S_2$  and  $\text{distance}(p, q) < \delta$ .
- If so,  $p$  and  $q$  must both be within  $\delta$  of  $l$ .
- Let  $P_1$  and  $P_2$  be vertical regions of the plane of width  $\delta$  on either side of  $l$ .
- If  $\{p, q\}$  exists,  $p$  must be within  $P_1$  and  $q$  within  $P_2$ .
- However, every point in  $S_1$  and  $S_2$  may be a candidate, as long as each is within  $\delta$  of  $l$ , which implies:  $O(n/2) \cdot O(n/2) = O(n^2)$



Can we do better ?

# Closest Pair – Divide & Conquer

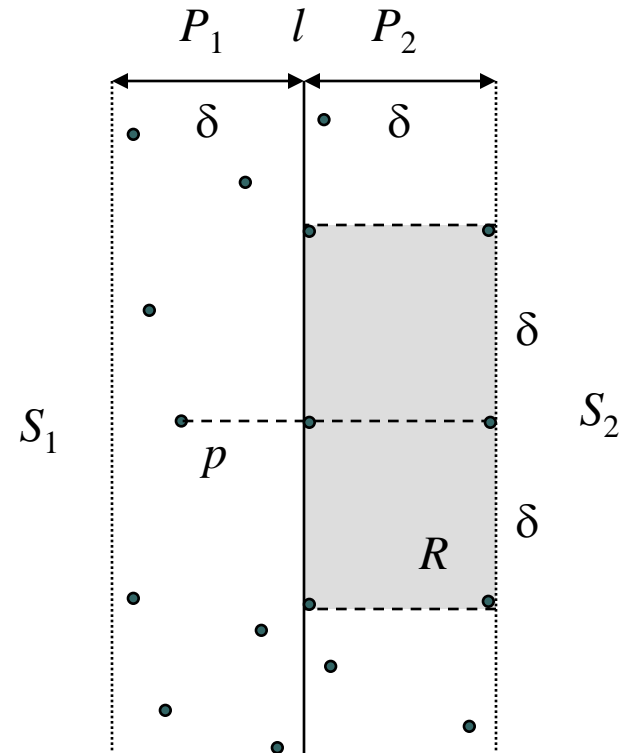
- For a point  $p$  in  $P_1$ , which portion of  $P_2$  should be checked?
- We only need to check the points that are within  $\delta$  of  $p$ .
- Thus we can limit the portion of  $P_2$ .
- The points to consider for a point  $p$  must lie within  $\delta \times 2\delta$  rectangle  $R$ .
- At most, how many points are there in rectangle  $R$ ?



# Closest Pair – Divide & Conquer

How many points are there in rectangle  $R$ ?

- Since no two points can be closer than  $\delta$ , there can only be at most 6 points
- Therefore,  $6 \cdot O(n/2) \in O(n)$
- Thus, the time complexity is
  - $O(n \log n)$



How do we know which 6 points to check?

# Closest Pair – Divide & Conquer

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How do we know which 6 points to check?

- Project  $p$  and all the points of  $S_2$  within  $P_2$  onto  $l$ .
- Only the points within  $\delta$  of  $p$  in the  $y$  projection need to be considered (max of 6 points).
- After sorting the points on  $y$  coordinate we can find the points by scanning the sorted lists. Points are sorted by  $y$  coordinates.
- To prevent resorting in  $O(n \log n)$  in each merge, two previously sorted lists are merged in  $O(n)$ .

Time Complexity:  $O(n \log n)$



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# Thank you!

