

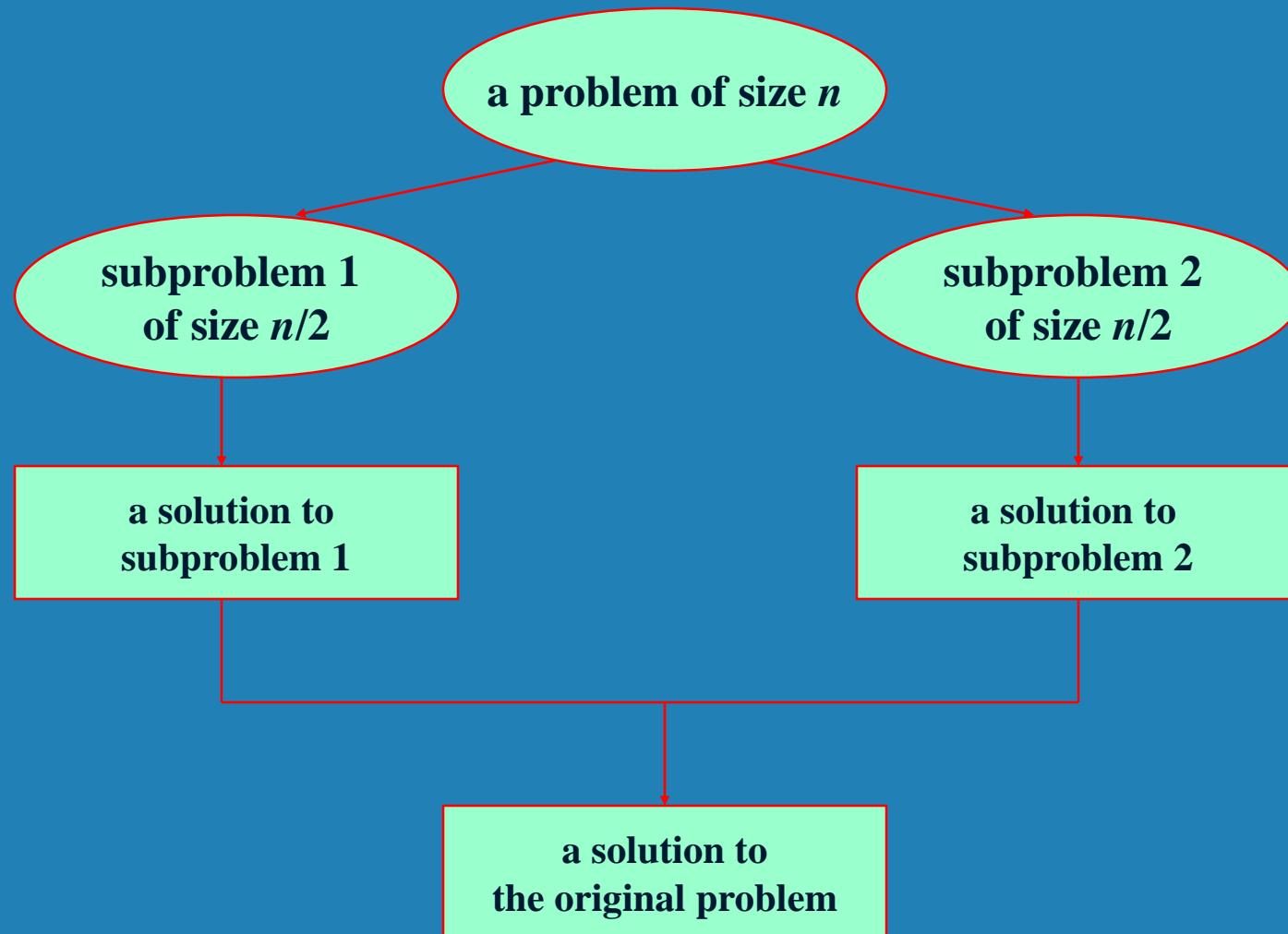
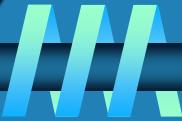
# Divide-and-Conquer



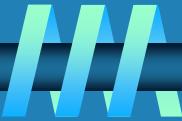
**The most-well known algorithm design strategy:**

- 1. Divide instance of problem into two or more smaller instances**
- 2. Solve smaller instances recursively**
- 3. Obtain solution to original (larger) instance by combining these solutions**

# Divide-and-Conquer Technique (cont.)



# Divide-and-Conquer Examples



- **Sorting: mergesort and quicksort**
- **Binary tree traversals**
- **Multiplication of large integers**
- **Matrix multiplication: Strassen's algorithm**
- **Closest-pair and convex-hull algorithms**

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- **Binary search: decrease-by-half (or degenerate divide&conq.)**

# General Divide-and-Conquer Recurrence

$$T(n) = aT(n/b) + f(n) \quad \text{where } f(n) \in \Theta(n^d), \quad d \geq 0$$



**Master Theorem:**

- If  $a < b^d$ ,  $T(n) \in \Theta(n^d)$
- If  $a = b^d$ ,  $T(n) \in \Theta(n^d \log n)$
- If  $a > b^d$ ,  $T(n) \in \Theta(n^{\log_b a})$

**Note:** The same results hold with  $O$  instead of  $\Theta$ .

**Examples:**  $T(n) = 4T(n/2) + n \Rightarrow T(n) \in ?$

$T(n) = 4T(n/2) + n^2 \Rightarrow T(n) \in ?$

$T(n) = 4T(n/2) + n^3 \Rightarrow T(n) \in ?$

# Mergesort



- Split array A[0..n-1] in two about equal halves and make copies of each half in arrays B and C
- Sort arrays B and C recursively
- Merge sorted arrays B and C into array A as follows:
  - Repeat the following until no elements remain in one of the arrays:
    - compare the first elements in the remaining unprocessed portions of the arrays
    - copy the smaller of the two into A, while incrementing the index indicating the unprocessed portion of that array
  - Once all elements in one of the arrays are processed, copy the remaining unprocessed elements from the other array into A.

# Pseudocode of Mergesort

**ALGORITHM** *Mergesort( $A[0..n - 1]$ )*

//Sorts array  $A[0..n - 1]$  by recursive mergesort  
//Input: An array  $A[0..n - 1]$  of orderable elements  
//Output: Array  $A[0..n - 1]$  sorted in nondecreasing order

**if**  $n > 1$

copy  $A[0..\lfloor n/2 \rfloor - 1]$  to  $B[0..\lfloor n/2 \rfloor - 1]$   
copy  $A[\lfloor n/2 \rfloor ..n - 1]$  to  $C[0..\lceil n/2 \rceil - 1]$   
*Mergesort( $B[0..\lfloor n/2 \rfloor - 1]$ )*  
*Mergesort( $C[0..\lceil n/2 \rceil - 1]$ )*  
*Merge( $B, C, A$ )*

# Pseudocode of Merge



**ALGORITHM** *Merge( $B[0..p - 1]$ ,  $C[0..q - 1]$ ,  $A[0..p + q - 1]$ )*

//Merges two sorted arrays into one sorted array

//Input: Arrays  $B[0..p - 1]$  and  $C[0..q - 1]$  both sorted

//Output: Sorted array  $A[0..p + q - 1]$  of the elements of  $B$  and  $C$

$i \leftarrow 0$ ;  $j \leftarrow 0$ ;  $k \leftarrow 0$

**while**  $i < p$  **and**  $j < q$  **do**

**if**  $B[i] \leq C[j]$

$A[k] \leftarrow B[i]$ ;  $i \leftarrow i + 1$

**else**  $A[k] \leftarrow C[j]$ ;  $j \leftarrow j + 1$

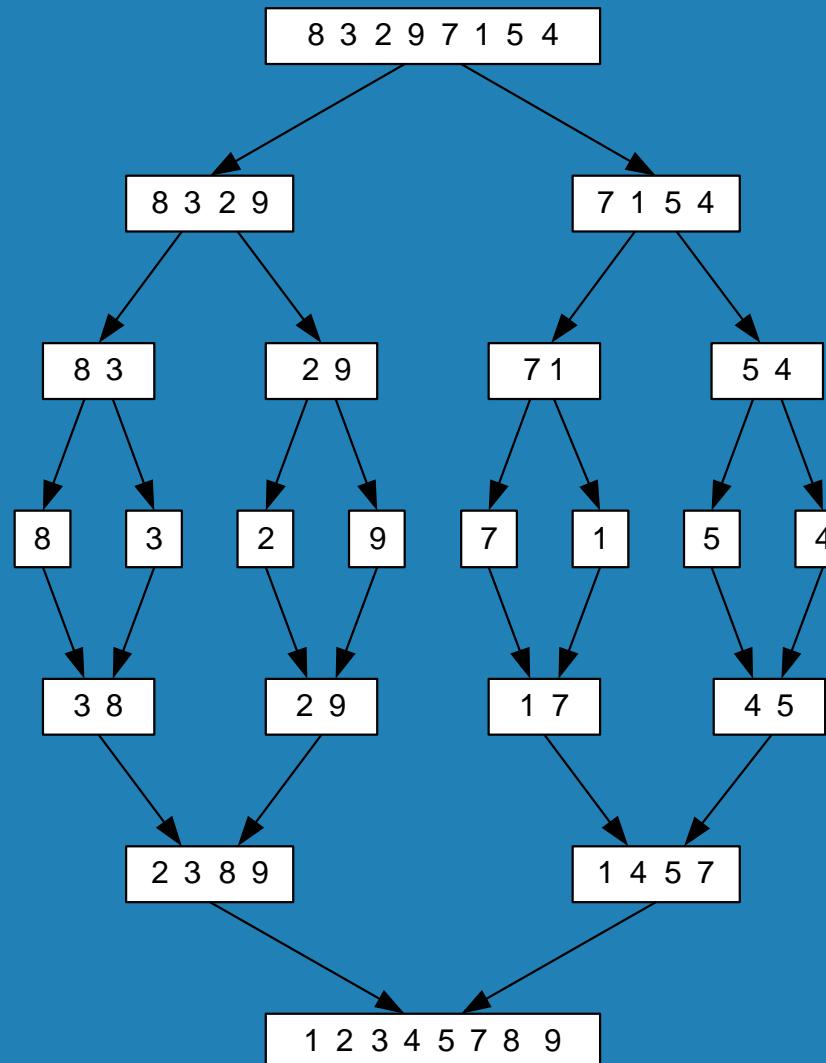
$k \leftarrow k + 1$

**if**  $i = p$

        copy  $C[j..q - 1]$  to  $A[k..p + q - 1]$

**else** copy  $B[i..p - 1]$  to  $A[k..p + q - 1]$

# Mergesort Example



# Analysis of Mergesort

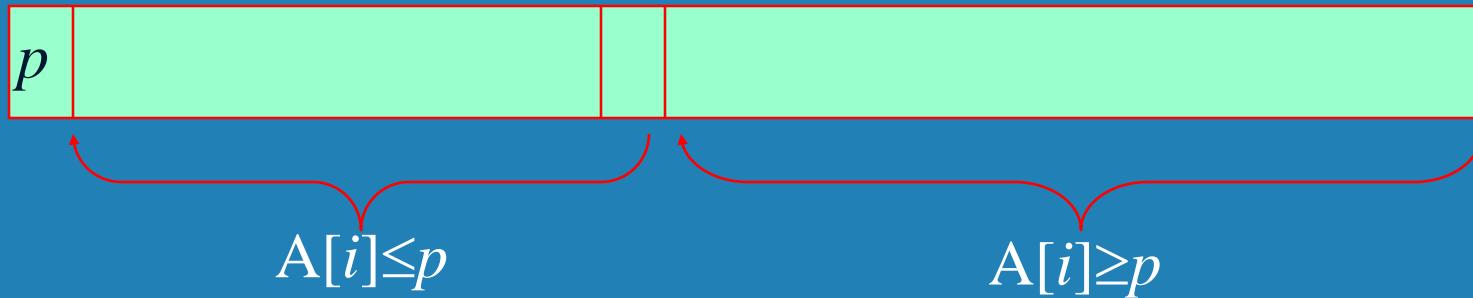


- All cases have same efficiency:  $\Theta(n \log n)$
- Number of comparisons in the worst case is close to theoretical minimum for comparison-based sorting:
$$\lceil \log_2 n! \rceil \approx n \log_2 n - 1.44n$$
- Space requirement:  $\Theta(n)$  (not in-place)
- Can be implemented without recursion (bottom-up)

# Quicksort



- Select a *pivot* (partitioning element) – here, the first element
- Rearrange the list so that all the elements in the first  $s$  positions are smaller than or equal to the pivot and all the elements in the remaining  $n-s$  positions are larger than or equal to the pivot (see next slide for an algorithm)



- Exchange the pivot with the last element in the first (i.e.,  $\leq$ ) subarray — the pivot is now in its final position
- Sort the two subarrays recursively

# Hoare's Partitioning Algorithm

```
Algorithm Partition( $A[l..r]$ )
//Partitions a subarray by using its first element as a pivot
//Input: A subarray  $A[l..r]$  of  $A[0..n - 1]$ , defined by its left and right
//      indices  $l$  and  $r$  ( $l < r$ )
//Output: A partition of  $A[l..r]$ , with the split position returned as
//        this function's value


$p \leftarrow A[l]$



$i \leftarrow l; j \leftarrow r + 1$



repeat



repeat  $i \leftarrow i + 1$  until  $A[i] \geq p$



repeat  $j \leftarrow j - 1$  until  $A[j] < p$



swap( $A[i], A[j]$ )



until  $i \geq j$



swap( $A[i], A[j]$ ) //undo last swap when  $i \geq j$



swap( $A[l], A[j]$ )



return  $j$


```

# Quicksort Example



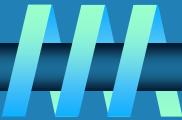
5 3 1 9 8 2 4 7

# Analysis of Quicksort



- Best case: split in the middle —  $\Theta(n \log n)$
- Worst case: sorted array! —  $\Theta(n^2)$
- Average case: random arrays —  $\Theta(n \log n)$
  
- Improvements:
  - better pivot selection: median of three partitioning
  - switch to insertion sort on small subfiles
  - elimination of recursionThese combine to 20-25% improvement
  
- Considered the method of choice for internal sorting of large files ( $n \geq 10000$ )

# Binary Tree Algorithms



Binary tree is a divide-and-conquer ready structure!

Ex. 1: Classic traversals (preorder, inorder, postorder)

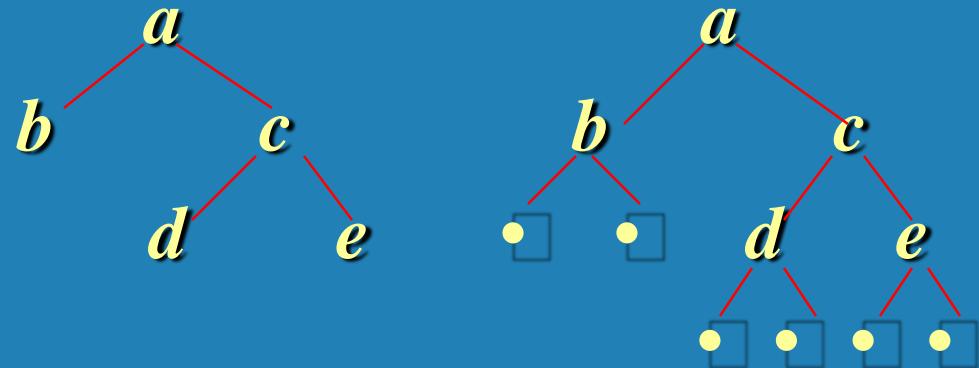
Algorithm *Inorder*( $T$ )

if  $T \neq \emptyset$

*Inorder*( $T_{left}$ )

print(root of  $T$ )

*Inorder*( $T_{right}$ )

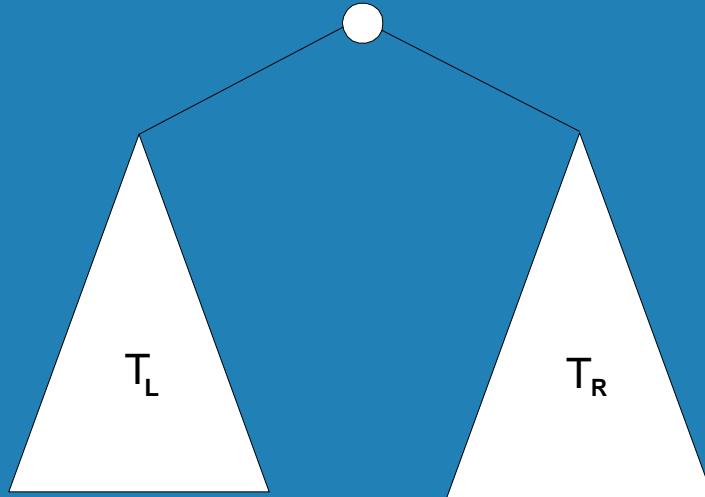


Efficiency:  $\Theta(n)$

# Binary Tree Algorithms (cont.)



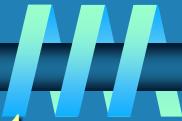
Ex. 2: Computing the height of a binary tree



$$h(T) = \max\{h(T_L), h(T_R)\} + 1 \text{ if } T \neq \emptyset \text{ and } h(\emptyset) = -1$$

Efficiency:  $\Theta(n)$

# Multiplication of Large Integers



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:  $n^2$  one-digit multiplications

# First Divide-and-Conquer Algorithm

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

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

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

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

In general, if  $A = A_1 A_2$  and  $B = B_1 B_2$  (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



$$\mathbf{A} * \mathbf{B} = \mathbf{A}_1 * \mathbf{B}_1 \cdot 10^n + (\mathbf{A}_1 * \mathbf{B}_2 + \mathbf{A}_2 * \mathbf{B}_1) \cdot 10^{n/2} + \mathbf{A}_2 * \mathbf{B}_2$$

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

$$(\mathbf{A}_1 + \mathbf{A}_2) * (\mathbf{B}_1 + \mathbf{B}_2) = \mathbf{A}_1 * \mathbf{B}_1 + (\mathbf{A}_1 * \mathbf{B}_2 + \mathbf{A}_2 * \mathbf{B}_1) + \mathbf{A}_2 * \mathbf{B}_2,$$

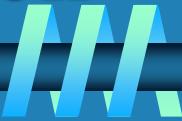
I.e.,  $(\mathbf{A}_1 * \mathbf{B}_2 + \mathbf{A}_2 * \mathbf{B}_1) = (\mathbf{A}_1 + \mathbf{A}_2) * (\mathbf{B}_1 + \mathbf{B}_2) - \mathbf{A}_1 * \mathbf{B}_1 - \mathbf{A}_2 * \mathbf{B}_2$ , which requires only 3 multiplications at the expense of (4-1) extra add/sub.

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

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

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

# Example of Large-Integer Multiplication



**2135 \* 4014**

# Strassen's Matrix Multiplication



Strassen observed [1969] that the product of two matrices can be computed as follows:

$$\begin{pmatrix} C_{00} & C_{01} \\ \hline C_{10} & C_{11} \end{pmatrix} = \begin{pmatrix} A_{00} & A_{01} \\ \hline A_{10} & A_{11} \end{pmatrix} * \begin{pmatrix} B_{00} & B_{01} \\ \hline B_{10} & B_{11} \end{pmatrix}$$
$$= \begin{pmatrix} M_1 + M_4 - M_5 + M_7 & M_3 + M_5 \\ M_2 + M_4 & M_1 + M_3 - M_2 + M_6 \end{pmatrix}$$

# Formulas for Strassen's Algorithm



$$\mathbf{M}_1 = (\mathbf{A}_{00} + \mathbf{A}_{11}) * (\mathbf{B}_{00} + \mathbf{B}_{11})$$

$$\mathbf{M}_2 = (\mathbf{A}_{10} + \mathbf{A}_{11}) * \mathbf{B}_{00}$$

$$\mathbf{M}_3 = \mathbf{A}_{00} * (\mathbf{B}_{01} - \mathbf{B}_{11})$$

$$\mathbf{M}_4 = \mathbf{A}_{11} * (\mathbf{B}_{10} - \mathbf{B}_{00})$$

$$\mathbf{M}_5 = (\mathbf{A}_{00} + \mathbf{A}_{01}) * \mathbf{B}_{11}$$

$$\mathbf{M}_6 = (\mathbf{A}_{10} - \mathbf{A}_{00}) * (\mathbf{B}_{00} + \mathbf{B}_{01})$$

$$\mathbf{M}_7 = (\mathbf{A}_{01} - \mathbf{A}_{11}) * (\mathbf{B}_{10} + \mathbf{B}_{11})$$

# Analysis of Strassen's Algorithm



If  $n$  is not a power of 2, matrices can be padded with zeros.

Number of multiplications:

$$M(n) = 7M(n/2), \quad M(1) = 1$$

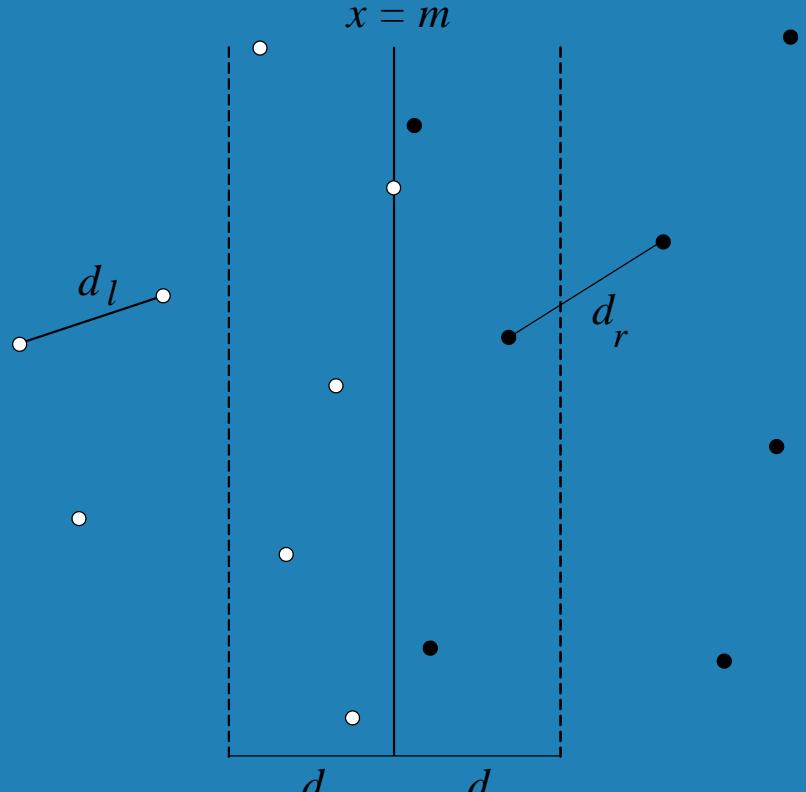
Solution:  $M(n) = 7^{\log_2 n} = n^{\log_2 7} \approx n^{2.807}$  vs.  $n^3$  of brute-force alg.

Algorithms with better asymptotic efficiency are known but they are even more complex.

# Closest-Pair Problem by Divide-and-Conquer



Step 1 Divide the points given into two subsets  $P_l$  and  $P_r$  by a vertical line  $x = m$  so that half the points lie to the left or on the line and half the points lie to the right or on the line.



# Closest Pair by Divide-and-Conquer (cont.)

Step 2 Find recursively the closest pairs for the left and right subsets.

Step 3 Set  $d = \min\{d_l, d_r\}$

We can limit our attention to the points in the symmetric vertical strip  $S$  of width  $2d$  as possible closest pair. (The points are stored and processed in increasing order of their  $y$  coordinates.)

Step 4 Scan the points in the vertical strip  $S$  from the lowest up. For every point  $p(x,y)$  in the strip, inspect points in the strip that may be closer to  $p$  than  $d$ . There can be no more than 5 such points following  $p$  on the strip list!

# Efficiency of the Closest-Pair Algorithm



**Running time of the algorithm is described by**

$$T(n) = 2T(n/2) + M(n), \text{ where } M(n) \in O(n)$$

**By the Master Theorem (with  $a = 2, b = 2, d = 1$ )**

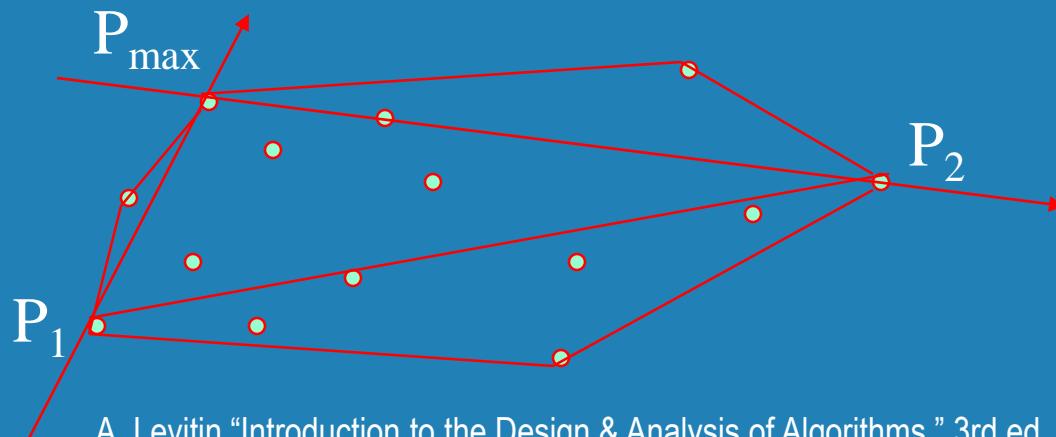
$$T(n) \in O(n \log n)$$

# Quickhull Algorithm

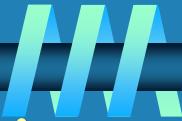


**Convex hull:** smallest convex set that includes given points

- Assume points are sorted by  $x$ -coordinate values
- Identify *extreme points*  $P_1$  and  $P_2$  (leftmost and rightmost)
- Compute *upper hull* recursively:
  - find point  $P_{\max}$  that is farthest away from line  $P_1P_2$
  - compute the upper hull of the points to the left of line  $P_1P_{\max}$
  - compute the upper hull of the points to the left of line  $P_{\max}P_2$
- Compute *lower hull* in a similar manner



# Efficiency of Quickhull Algorithm



- Finding point farthest away from line  $P_1P_2$  can be done in linear time
- Time efficiency:
  - worst case:  $\Theta(n^2)$  (as quicksort)
  - average case:  $\Theta(n)$  (under reasonable assumptions about distribution of points given)
- If points are not initially sorted by  $x$ -coordinate value, this can be accomplished in  $O(n \log n)$  time
- Several  $O(n \log n)$  algorithms for convex hull are known