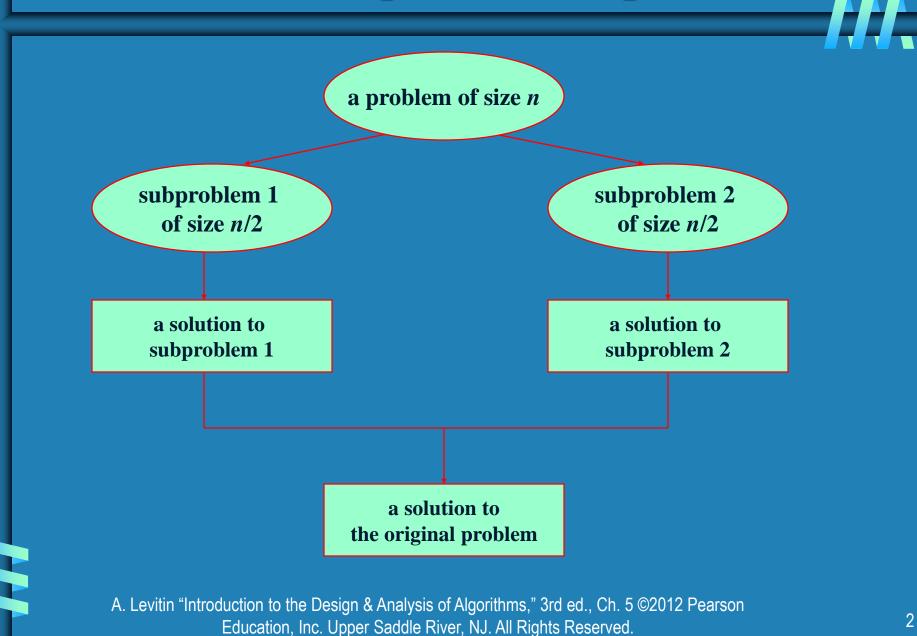
Divide-and-Conquer

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

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



General Divide-and-Conquer Recurrence

T(n) = aT(n/b) + f(n) where $f(n) \in \Theta(n^d)$, $d \ge 0$

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 $a > b^d$, $T(n) \in \Theta(n^{\log b^d})$

Note: The same results hold with O instead of Θ .

Examples: $T(n) = 4T(n/2) + n \implies T(n) \in ?$ $T(n) = 4T(n/2) + n^2 \implies T(n) \in ?$ $T(n) = 4T(n/2) + n^3 \implies T(n) \in ?$

Mergesort



- **Q** 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
- **Q** 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.

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..[n/2] - 1] to B[0..[n/2] - 1]copy A[[n/2]..n - 1] to C[0..[n/2] - 1]Mergesort(B[0..[n/2] - 1])Mergesort(C[0..[n/2] - 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
$$B[i] \leq C[j]$$

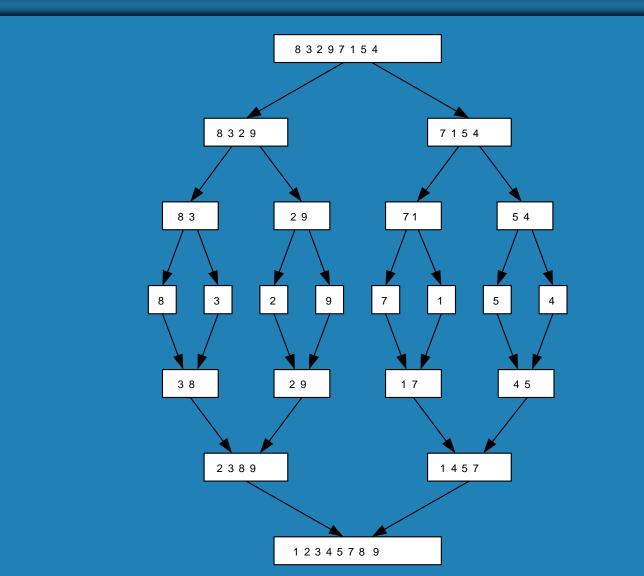
 $A[k] \leftarrow B[i]; i \leftarrow i+1$
else $A[k] \leftarrow C[j]; j \leftarrow j+1$

 $k \leftarrow k+1$

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: $\Theta(n \log n)$

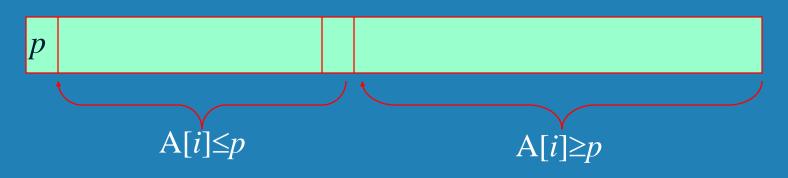
Q Number of comparisons in the worst case is close to theoretical minimum for comparison-based sorting: $\lceil \log_2 n! \rceil \approx n \log_2 n - 1.44n$

Q Space requirement: $\Theta(n)$ (not in-place)

Q 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)



- **Q** 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; \quad j \leftarrow r+1
repeat
    repeat i \leftarrow i+1 until A[i] \ge p
    repeat j \leftarrow j-1 until A[j] \leftarrow p
    swap(A[i], A[j])
until i \geq j
swap(A[i], A[j]) / undo last swap when <math>i \ge j
\operatorname{swap}(A[l], A[j])
return j
```

Quicksort Example

5 3 1 9 8 2 4 7

Analysis of Quicksort

- **Q** Best case: split in the middle $\Theta(n \log n)$
- **Q** Worst case: sorted array! $\Theta(n^2)$
- **Q** 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 recursion

These combine to 20-25% improvement

Q Considered the method of choice for internal sorting of large files ($n \ge 10000$)

Binary Tree Algorithms

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

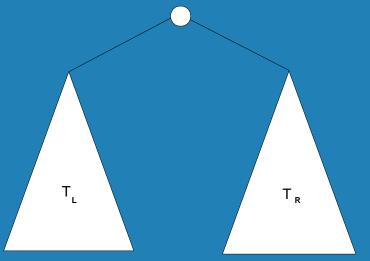
Ex. 1: Classic traversals (preorder, inorder, postorder) Algorithm *Inorder*(*T*) if $T \neq \emptyset$ *Inorder*(T_{left}) print(root of *T*) *d e d e d*

Inorder(T_{right})

Efficiency: $\Theta(n)$

Binary Tree Algorithms (cont.)

Ex. 2: Computing the height of a binary tree



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

 $\begin{array}{c} a_1 \ a_2 \ \dots \ a_n \\ b_1 \ b_2 \ \dots \ b_n \\ (d_{10}) \ d_{11} \ d_{12} \ \dots \ d_{1n} \\ (d_{20}) \ d_{21} \ d_{22} \ \dots \ d_{2n} \end{array}$

••• ••• ••• ••• ••• •••

 $(d_{n0}) d_{n1} d_{n2} \dots d_{nn}$

Efficiency: *n*² one-digit multiplications

First Divide-and-Conquer Algorithm

A small example: A * B where A = 2135 and B = 4014 A = $(21 \cdot 10^2 + 35)$, B = $(40 \cdot 10^2 + 14)$ 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 $A = A_1A_2$ and $B = B_1B_2$ (where A and B are *n*-digit, A₁, A₂, B₁, B₂ are *n*/2-digit numbers), A * B = A₁ * B₁·10^{*n*} + (A₁ * B₂ + A₂ * B₁) ·10^{*n*/2} + A₂ * B₂

Recurrence for the number of one-digit multiplications M(n): M(n) = 4M(n/2), M(1) = 1Solution: $M(n) = n^2$

Second Divide-and-Conquer Algorithm

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

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 (4-1) extra add/sub.

Recurrence for the number of multiplications M(*n*): M(n) = 3M(n/2), M(1) = 1Solution: 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:

$$\begin{pmatrix} C_{00} & C_{01} \\ \hline C_{10} & C_{11} \end{pmatrix} = \begin{pmatrix} A_{00} & A_{01} \\ \hline A_{10} & A_{11} \end{pmatrix} * \begin{pmatrix} B_{00} & B_{01} \\ \hline B_{10} & B_{11} \end{pmatrix}$$

$$= \begin{pmatrix} M_1 & +M_4 - M_5 + M_7 & M_3 + M_5 \\ \hline M_2 + M_4 & M_1 & +M_3 - M_2 + M_6 \end{pmatrix}$$

Formulas for Strassen's Algorithm

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

 $M_2 = (A_{10} + A_{11}) * B_{00}$

$$M_3 = A_{00} * (B_{01} - B_{11})$$

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

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

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

$\mathbf{M}_{7} = (\mathbf{A}_{01} - \mathbf{A}_{11}) * (\mathbf{B}_{10} + \mathbf{B}_{11})$

Analysis of Strassen's Algorithm

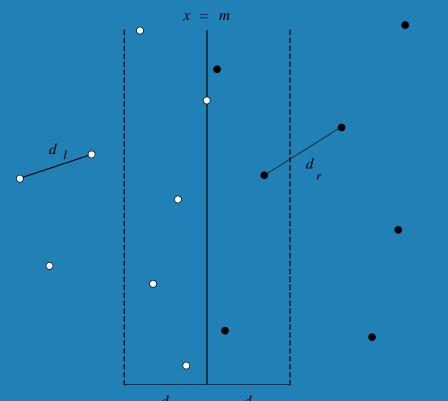
If *n* is not a power of 2, matrices can be padded with zeros.

Number of multiplications: M(n) = 7M(n/2), M(1) = 1Solution: $M(n) = 7^{\log 2^n} = n^{\log 2^7} \approx n^{2.807}$ 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 2*d* 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 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!

Running time of the algorithm is described by

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

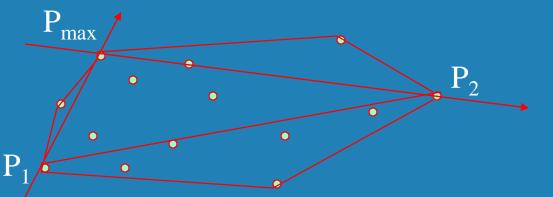
By the Master Theorem (with a = 2, b = 2, d = 1) $T(n) \in O(n \log n)$

Quickhull Algorithm

Convex hull: smallest convex set that includes given points

- **A** Assume points are sorted by *x*-coordinate values
- **Q** 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 $P_1 P_{\text{max}}$
 - compute the upper hull of the points to the left of line $P_{\text{max}}P_2$

Q Compute *lower hull* in a similar manner



Efficiency of Quickhull Algorithm

- **Q** Finding point farthest away from line P_1P_2 can be done in linear time
- **Q** Time efficiency:
 - worst case: $\Theta(n^2)$ (as quicksort)
 - average case: Θ(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

Q Several O(*n* log *n*) algorithms for convex hull are known