

CSC 8301- Design and Analysis of Algorithms

Lecture 6

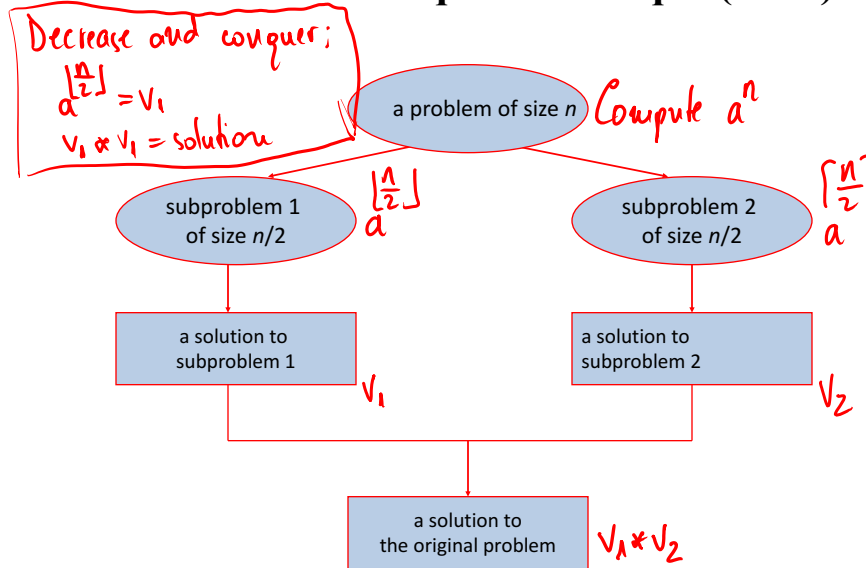
Divide and Conquer Algorithm Design Technique

Divide-and-Conquer

The most-well known algorithm design strategy:

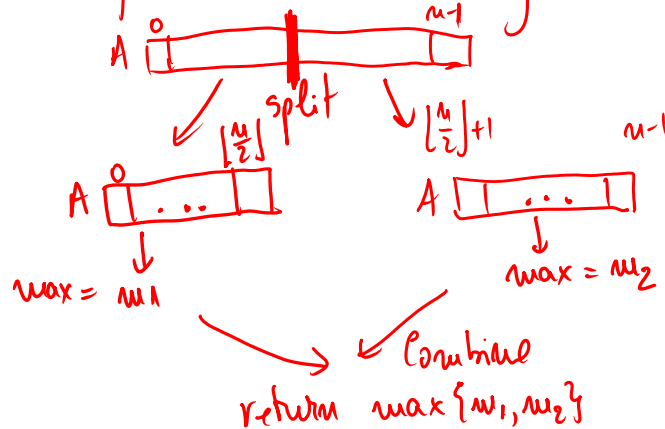
1. Divide a problem instance into two or more smaller instances (ideally of about the same size)
2. Solve the smaller instances (usually recursively)
3. Obtain a solution to the original instance by combining these solutions to the smaller instances

Divide-and-Conquer Technique (cont.)



Divide-and-Conquer Examples

Find the largest value in an array $A[0..n-1]$



Time Complexity - recurrence formula (assume $n=2^k$)
 $C(n) = 2C\left(\frac{n}{2}\right) + 1$

General Divide-and-Conquer Recurrence

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

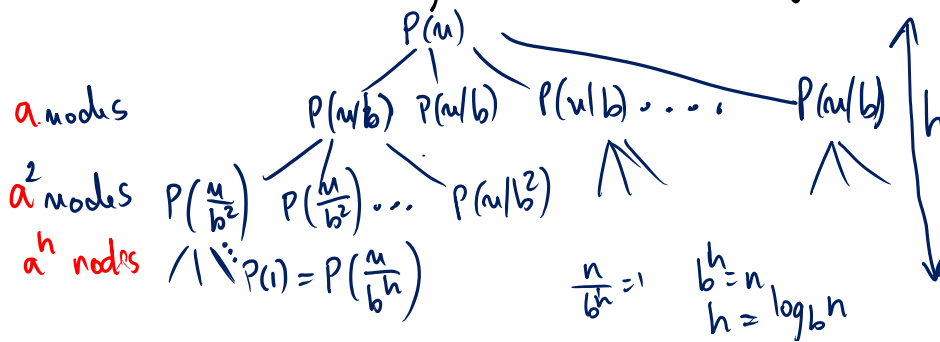
n = size of the problem

a = constant = number of subproblems

n/b = size of the subproblems

$f(n)$ = cost of dividing the problem and merging solutions

leaves = a^h
 $= a^{\log_b n}$
 $= n^{\log_b a}$



General Divide-and-Conquer Recurrence

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

leaves = $n^{\log_b a}$

- 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 Θ

Examples:

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

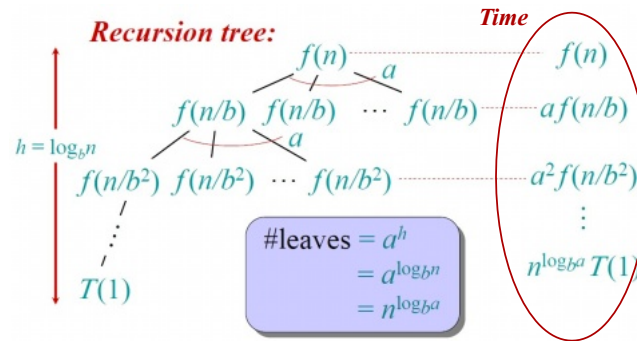
Handwritten notes for the first example:

- $a=4, b=2, d=1$
- $a=4 > b^d=2$
- $T(n) = \Theta(n^{\log_2 4}) = \Theta(n^2)$
- $\Theta(n^2 \log n)$
- $\Theta(n^3)$

Master Theorem – Recursion Tree

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

Visualize this as a recursion tree (branch factor a):



Total time depends on how fast $f(n)$ grows compared with the number of leaves (a compared with $\log_b a$)

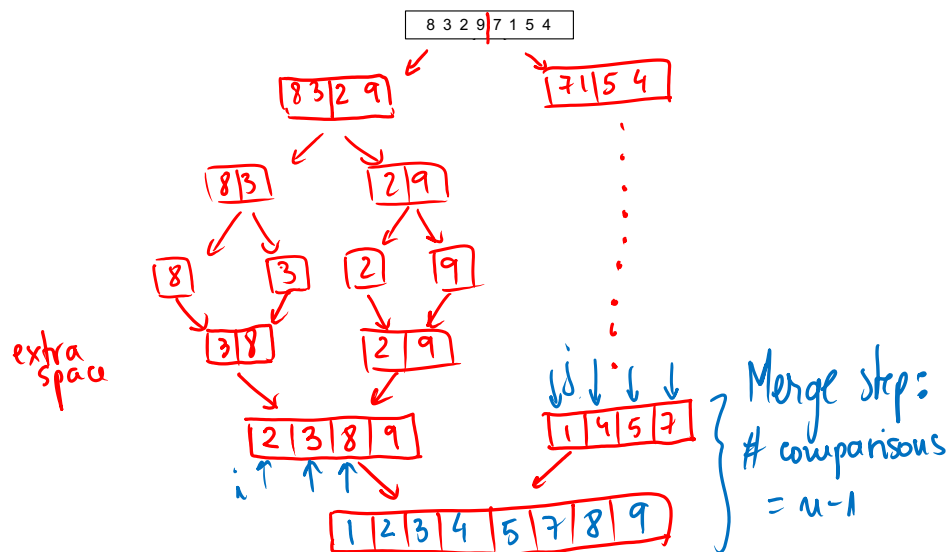
Divide-and-Conquer Examples

- Sorting: mergesort and quicksort
- Binary tree traversals
- Binary search (?)
- Multiplication of large integers
- Closest-pair algorithm

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 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

Mergesort Example



Merging of Two Sorted Arrays

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 $A[k..p+q-1]$

Analysis of Mergesort

- Recurrence for number of comparisons in the worst case:

$$C(n) = 2C(n/2) + C_{merge}(n) = 2C(n/2) + n - 1$$

$$C(1) = 0$$

Master Theorem:

$$a = b = 2, d = 1 = \log_b a$$

$$C(n) = \Theta(n \log n)$$

Solve exactly ($n = 2^k$)

$$C(n) = 2C\left(\frac{n}{2}\right) + n - 1$$

$$= 2 \left[2C\left(\frac{n}{2^2}\right) + \frac{n}{2} - 1 \right] + n - 1$$

$$= 2^2 C\left(\frac{n}{2^2}\right) + (n-2) + (n-1)$$

$$= 2^2 \left[2C\left(\frac{n}{2^3}\right) + \frac{n}{2^2} - 1 \right] + (n-2) + (n-1)$$

$$= 2^3 C\left(\frac{n}{2^3}\right) + (n-2^2) + (n-2) + (n-1)$$

$$= (n-2^{k-1}) + (n-2^{k-2}) + \dots + (n-1) = k \cdot n - (2^k - 1)$$

$$\rightarrow = n \log_2 n - (n-1) = \Theta(n \log n)$$

Analysis of Mergesort

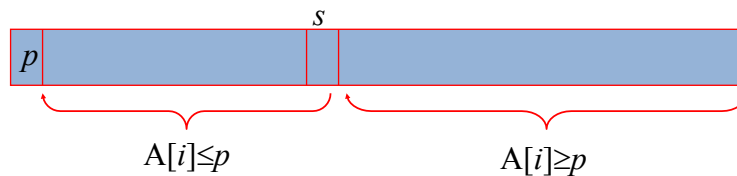
- Recurrence for number of comparisons in the worst case:

$$\begin{aligned} C(n) &= 2 C(n/2) + C_{merge}(n) = 2 C(n/2) + n-1 \\ C(1) &= 0 \end{aligned} \quad \left. \vphantom{\begin{aligned} C(n) \\ C(1) \end{aligned}} \right\} \Theta(n \log n)$$

- All cases have same efficiency: $\Theta(n \log n)$
- 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)



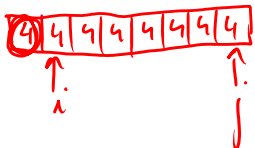
- Exchange the pivot with the last element in the first (i.e., \leq) subarray — *the pivot is now in its final position*
- Sort the two subarrays recursively

Two-Way (Hoar's) Partitioning Algorithm

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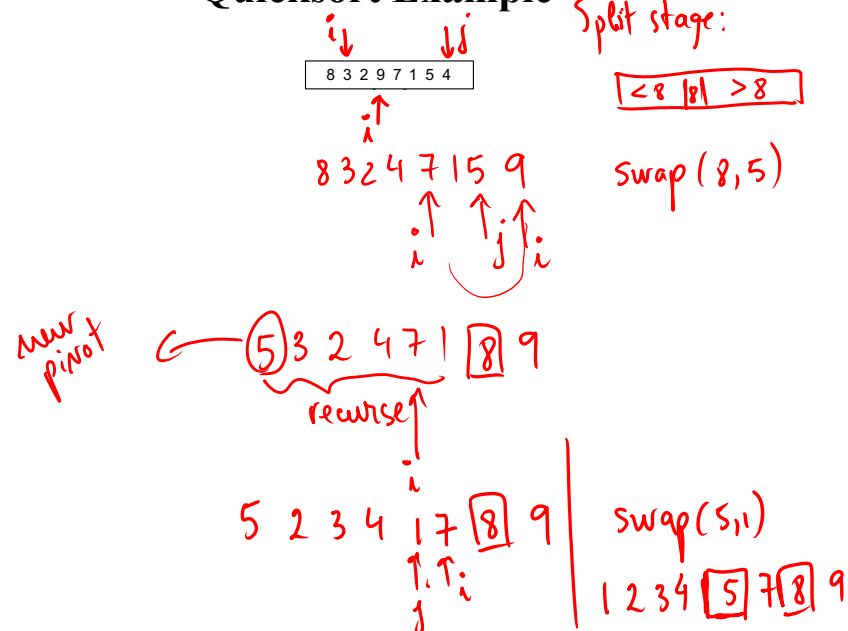
Algorithm Partition( $A[l..r]$ )
//Partitions a subarray by using its first element as a pivot
//Input: A subarray  $A[l..r]$  of  $A[0..n-1]$ , defined by its left and right
//       indices  $l$  and  $r$  ( $l < r$ )
//Output: A partition of  $A[l..r]$ , with the split position returned as
//        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  $A[j] \leq p$ 
  swap( $A[i], A[j]$ )
until  $i \geq j$ 
swap( $A[i], A[j]$ ) //undo last swap when  $i \geq j$ 
swap( $A[l], A[i]$ )
return  $j$ 

```



 $\rightarrow A[l]$ will be in its final position

Quicksort Example



Analysis of Quicksort

- Best-case time efficiency: split in the middle — $\Theta(n \log n)$

How many comparisons?

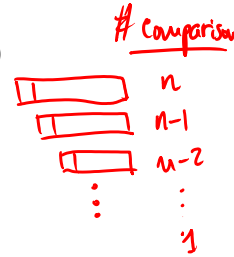
$$C_{\text{best}}(n) = 2C_{\text{best}}\left(\frac{n}{2}\right) + \underline{O(n)} \rightarrow n$$

- Worst-case time efficiency: sorted array! — $\Theta(n^2)$

All splits skewed to one extreme:

$$C_{\text{worst}}(n) = C_{\text{worst}}(n-1) + n$$

$$= (n) + (n-1) + \dots + 1 = \Theta(n^2)$$



- Average case time efficiency: random arrays — $\Theta(n \log n)$

Analysis of Quicksort

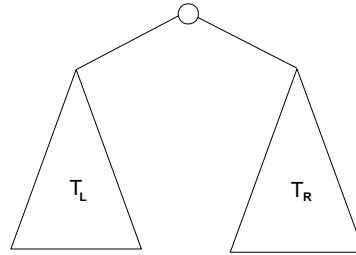
- Not stable
- Space efficiency: $\Theta(\log n)$ on average (recursive calls)
- Improvements:
 - better pivot selection: median-of-three partitioning
 - stop recursive calls when unsorted subarrays become small (say, <10 elements) and finish sorting with insertion sort
 These yields about 20% improvement
- Considered the method of choice for sorting random files of nontrivial sizes

Binary Tree Algorithms

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

Example 1: Classic traversals

- *Preorder*: root, left, right
- *Inorder*: left, root, right
- *Postorder*: left, right, root



Algorithm *Inorder*(T)

if $T \neq \emptyset$

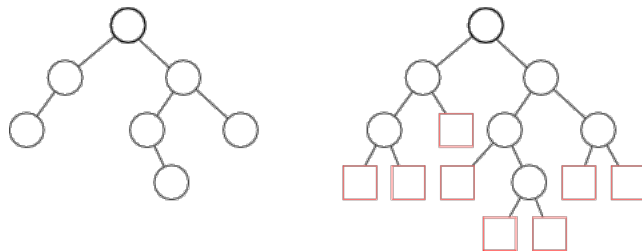
Inorder(T_{left})

print(root of T)

Inorder(T_{right})

Extended Binary Tree

- Replace every null subtree of the original tree with special extra nodes (called *extended* or *external*)

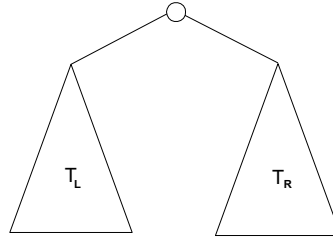


- How many extended nodes in a tree with n (original) nodes?

Binary Tree Algorithms (cont.)

Example 2: Computing the height of a binary tree

- The height is the length of the longest path (counting edges) on the way from the root to a leaf
- The height of a single node is 0



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

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:

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

Efficiency: n^2 one-digit multiplications

First Divide-and-Conquer Algorithm

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

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

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

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

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

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.

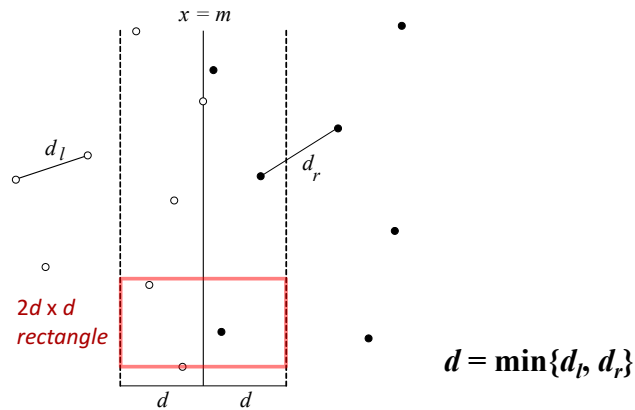
Recurrence for the number of multiplications $M(n)$:

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

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

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

Recurrence for the Running time of the algorithm

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

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

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

Homework

Reading: Chapter 5, Appendix B (pp. 487-491)

Exercises:

- 5.1: 1, 2, 3, 6, 8
- 5.2: 1, 7, 8, 9
- 5.3: 1, 2, 5, 8
- 5.4: 2, 3
- 5.5: 2

Next: Transform and Conquer (Ch. 6)