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# ECE-374-B: Lecture 11 - Divide and Conquer Algorithms

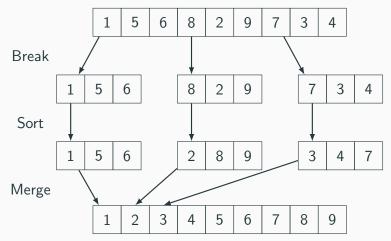
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February 23, 2023

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Simpler case: Break into 3 lists:



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So why don't we use smaller lists?

- 1. Pick a pivot element from array
- 2. Split array into 3 subarrays: those smaller than pivot, those larger than pivot, and the pivot itself.
- 3. Recursively sort the subarrays, and concatenate them.

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## **Quick Sort: Example**

array: 16, 12, 14, 20, 5, 3, 18, 19, 1

• pivot: 16

See visualizer:

https://www.hackerearth.com/practice/algorithms/sorting/quick-sort/visualize/

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- Typically, pivot is the first or last element of array. Then,

$$T(n) = \max_{1 \le k \le n} (T(k-1) + T(n-k) + O(n))$$

In the worst case T(n) = T(n-1) + O(n), which means  $T(n) = O(n^2)$ . Happens if array is already sorted and pivot is always first element.

# Selecting in Unsorted Lists

#### The Selection Problem

Big problem with QuickSort is that the pivot might not be the median.

How long would it take us to find the median of an unsorted list?

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How long would it take us to find the median of an unsorted list?

Sort, then A[n/2]. Is this the optimal way?

## Rank of element in an array

A: an unsorted array of n integers

For  $1 \le j \le n$ , element of rank j is the j-th smallest element in A.



### **Problem - Selection**

**Input** Unsorted array A of n integers **and** integer j **Goal** Find the j-th smallest number in A (rank j number)

Median: 
$$j = \lfloor (n+1)/2 \rfloor$$

#### **Problem - Selection**

**Input** Unsorted array A of n integers **and** integer j **Goal** Find the j-th smallest number in A (rank j number)

Median: 
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Simplifying assumption for sake of notation: elements of A are distinct

## Algorithm I

- Sort the elements in A
- Pick jth element in sorted order

Time taken = 
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Time taken =  $O(n \log n)$ 

Do we need to sort? Is there an O(n) time algorithm?

## Algorithm II

If j is small or n - j is small then

- Find j smallest/largest elements in A in O(jn) time. (How?)
- Time to find median is  $O(n^2)$ .

## Quick select

### QuickSelect

- Pick a pivot element a from A
- Partition *A* based on *a*.  $A_{less} = \{x \in A \mid x \leq a\}$  and  $A_{greater} = \{x \in A \mid x > a\}$
- $|A_{\text{less}}| = j$ : return a
- $|A_{\text{less}}| > j$ : recursively find jth smallest element in  $A_{\text{less}}$
- $|A_{less}| < j$ : recursively find kth smallest element in  $A_{greater}$  where  $k = j |A_{less}|$ .

## Example

16	14	34	20	12	5	3	19	11
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- Partitioning step: O(n) time to scan A
- How do we choose pivot? Recursive running time?

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Suppose we always choose pivot to be A[1].

Say A is sorted in increasing order and j=n. How long does this new algorithm take?

Should we combine this with QuickSort

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Of course not! It takes  $O(n^2)$  which is already the worse case of QuickSort. Need another method....

Looking at the quicksort recurrence again:

$$T(n) = T(k-1) + T(n-k) + O(n)$$

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What if  $k = \frac{3}{5}n$ ?

What if  $k = \frac{7}{10}n$ ?

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What if  $k = \frac{3}{5}n$ ?

What if  $k = \frac{7}{10}n$ ?

we only need to be able to find a rough median! .... How do we do that?

# Median of Medians

# **Divide and Conquer Approach**

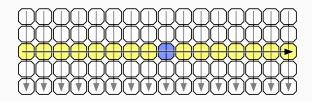
#### Idea

- Break input A into many subarrays:  $L_1, \ldots L_k$ .
- Find median m<sub>i</sub> in each subarray L<sub>i</sub>.
- Find the median x of the medians  $m_1, \ldots, m_k$ .
- Intuition: The median x should be close to being a good median of all the numbers in A.
- Use x as pivot in previous algorithm.

# Example

# Example

11     7     3     42     174     310     1     92     87     12     19     15
--



# Choosing the pivot

- Partition array A into  $\lceil n/5 \rceil$  lists of 5 items each.  $L_1 = \{A[1], A[2], \dots, A[5]\}, L_2 = \{A[6], \dots, A[10]\}, \dots, L_i = \{A[5i+1], \dots, A[5i-4]\}, \dots, L_{\lceil n/5 \rceil} = \{A[5\lceil n/5 \rceil 4, \dots, A[n]\}.$
- For each i find median  $b_i$  of  $L_i$  using brute-force in O(1) time. Total O(n) time
- Let  $B = \{b_1, b_2, \dots, b_{\lceil n/5 \rceil}\}$
- Find median b of B

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- Let  $B = \{b_1, b_2, \dots, b_{\lceil n/5 \rceil}\}$
- Find median b of B

Median of B is an approximate median of A. That is, if b is used a pivot to partition A, then  $|A_{\text{less}}| \leq 7n/10$  and  $|A_{\text{greater}}| \leq 7n/10$ .

# Algorithm for Selection

```
 \begin{split} \textbf{select}(A, \ j) : \\ & \quad \text{Form lists } L_1, L_2, \dots, L_{\lceil n/5 \rceil} \text{ where } L_i = \{A[5i-4], \dots, A[5i]\} \\ & \quad \text{Find median } b_i \text{ of each } L_i \text{ using brute-force} \\ & \quad \text{Find median } b \text{ of } B = \{b_1, b_2, \dots, b_{\lceil n/5 \rceil}\} \\ & \quad \text{Partition } A \text{ into } A_{\text{less}} \text{ and } A_{\text{greater}} \text{ using } b \text{ as pivot} \\ & \quad \textbf{if } (|A_{\text{less}}|) = j \text{ return } b \\ & \quad \textbf{else if } (|A_{\text{less}}|) > j) \\ & \quad \textbf{return select}(A_{\text{less}}, \ j) \\ & \quad \textbf{else} \\ & \quad \textbf{return select}(A_{\text{greater}}, \ j - |A_{\text{less}}|) \end{aligned}
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How do we find median of B?

# Algorithm for Selection

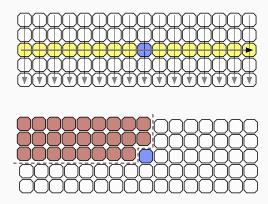
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How do we find median of B? Recursively!

Median of medians is a good median

#### Median of Medians: Proof of Lemma

There are at least 3n/10 elements smaller than the median of medians b.



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At least half of the  $\lfloor n/5 \rfloor$  groups have at least 3 elements smaller than b, except for the group containing b which has 2 elements smaller than b. Hence number of elements smaller than b is:

$$3\lfloor \frac{\lfloor n/5\rfloor + 1}{2} \rfloor - 1 \ge 3n/10$$

#### Median of Medians: Proof of Lemma

There are at least 3n/10 elements smaller than the median of medians b.

$$|A_{\rm greater}| \le 7n/10$$
.

Via symmetric argument,

$$|A_{\mathsf{less}}| \le 7n/10.$$

$$T(n) \le T(\lceil n/5 \rceil) + \max\{T(|A_{less}|), T(|A_{greater})|\} + O(n)$$

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From Lemma,

$$T(n) \le T(\lceil n/5 \rceil) + T(\lfloor 7n/10 \rfloor) + O(n)$$

and

$$T(n) = O(1) \qquad n < 10$$

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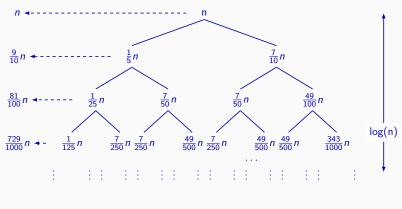
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$$T(n) = O(1) \qquad n < 10$$

**Exercise:** show that T(n) = O(n)

#### Recursion tree fill-in

If the workload is decreasing at every level, then total work is dominated by the root.



$$T(n) \le T(\lceil n/5 \rceil) + T(\lfloor 7n/10 \rfloor) + O(n) = O(n)$$

### What about QuickSort?

How would we use the median of medians approach for quicksort?

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How would we use the median of medians approach for quicksort?

Just use MoM if find pivot!

- Original recurrence: T(n) = T(k-1) + T(n-k) + O(n)
- With MoM:  $T(n) = T(\frac{3}{10}n) + T(\frac{7}{10}n) + O(n) + O(n)$

Due to:M. Blum, R. Floyd, D. Knuth, V. Pratt, R. Rivest, and R.

Tarjan.

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All except Vaughan Pratt! **Favorite Knuth quote**: He once warned a correspondent, "Beware of bugs in the above code; I have only proved it correct, not tried it."

# **Takeaway Points**

- Recursion tree method and guess and verify are the most reliable methods to analyze recursions in algorithms.
- Recursive algorithms naturally lead to recurrences.
- Some times one can look for certain type of recursive algorithms (reverse engineering) by understanding recurrences and their behavior.

# Problem statement: Multiplying numbers + a slow algorithm

# The Problem: Multiplying numbers

Given two large positive integer numbers b and c, with n digits, compute the number b \* c.

# Egyptian multiplication: 1850BC (3870 years ago?)

76 35

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76	35	
76	34 + 1	76
76	34	
152	17	
152	16 + 1	152
152	16	
304	8	
608	4	
1216	2	

76	35	
76	34 + 1	76
76	34	
152	17	
152	16 + 1	152
152	16	
304	8	
608	4	
1216	2	
2432	1	2432

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152	16	
304	8	
608	4	
1216	2	
2432	1	2432
		2660

### The problem: Multiplying Numbers

**Problem** Given two n-digit numbers x and y, compute their product.

#### **Grade School Multiplication**

Compute "partial product" by multiplying each digit of y with x and adding the partial products.

 $\begin{array}{r}
 3141 \\
 \times 2718 \\
 \hline
 25128 \\
 3141 \\
 21987 \\
 \underline{6282} \\
 8537238
\end{array}$ 

#### Time Analysis of Grade School Multiplication

- Each partial product:  $\Theta(n)$
- Number of partial products:  $\Theta(n)$
- Addition of partial products:  $\Theta(n^2)$
- Total time:  $\Theta(n^2)$

Multiplication using Divide and

Conquer

#### **Divide and Conquer**

Assume n is a power of 2 for simplicity and numbers are in decimal.

Split each number into two numbers with equal number of digits

- $b = b_{n-1}b_{n-2}...b_0$  and  $c = c_{n-1}c_{n-2}...c_0$
- $b = b_{n-1} \dots b_{n/2} \dots 0 + b_{n/2-1} \dots b_0$
- $b(x) = b_L x + b_R$ , where  $x = 10^{n/2}$ ,  $b_L = b_{n-1} \dots b_{n/2}$  and  $b_R = b_{n/2-1} \dots b_0$
- Similarly  $c(x) = c_L x + c_R$  where  $c_L = c_{n-1} \dots c_{n/2}$  and  $c_R = c_{n/2-1} \dots c_0$

### **E**xample

$$1234 \times 5678 = (12x + 34) \times (56x + 78)$$
$$= 12 \cdot 56 \cdot x^2 + (12 \cdot 78 + 34 \cdot 56)x + 34 \cdot 78.$$

$$1234 \times 5678 = (100 \times 12 + 34) \times (100 \times 56 + 78)$$
$$= 10000 \times 12 \times 56$$
$$+100 \times (12 \times 78 + 34 \times 56)$$
$$+34 \times 78$$

for x=1

#### **Divide and Conquer for multiplication**

Assume n is a power of 2 for simplicity and numbers are in decimal.

- $b = b_{n-1}b_{n-2}...b_0$  and  $c = c_{n-1}c_{n-2}...c_0$
- $b \equiv b(x) = b_L x + b_R$ where  $x = 10^{n/2}$ ,  $b_L = b_{n-1} \dots b_{n/2}$  and  $b_R = b_{n/2-1} \dots b_0$
- $c \equiv c(x) = c_L x + c_R$  where  $c_L = c_{n-1} \dots c_{n/2}$  and  $c_R = c_{n/2-1} \dots c_0$

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- $c \equiv c(x) = c_L x + c_R$  where  $c_L = c_{n-1} \dots c_{n/2}$  and  $c_R = c_{n/2-1} \dots c_0$

Therefore, for  $x = 10^{n/2}$ , we have

$$bc = b(x)c(x) = (b_L x + b_R)(c_L x + c_R)$$
  
=  $b_L c_L x^2 + (b_L c_R + b_R c_L)x + b_R c_R$   
=  $10^n b_L c_L + 10^{n/2}(b_L c_R + b_R c_L) + b_R c_R$ 

#### **Time Analysis**

$$bc = 10^n b_L c_L + 10^{n/2} (b_L c_R + b_R c_L) + b_R c_R$$

4 recursive multiplications of number of size n/2 each plus 4 additions and left shifts (adding enough 0's to the right)

#### Time Analysis

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

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4 recursive multiplications of number of size n/2 each plus 4 additions and left shifts (adding enough 0's to the right)

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

 $T(n) = \Theta(n^2)$ . No better than grade school multiplication!

Faster multiplication: Karatsuba's

**Algorithm** 

#### A Trick of Gauss

Carl Friedrich Gauss: 1777-1855 "Prince of Mathematicians"

Observation: Multiply two complex numbers: (a + bi) and (c + di)

$$(a+bi)(c+di) = ac - bd + (ad + bc)i$$

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$$(a+bi)(c+di) = ac-bd+(ad+bc)i$$

How many multiplications do we need?

Only 3! If we do extra additions and subtractions.

Compute ac, bd, (a+b)(c+d). Then

## Gauss technique for polynomials

$$p(x) = ax + b$$
 and  $q(x) = cx + d$ .

$$p(x)q(x) = acx^2 + (ad + bc)x + bd.$$

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$$p(x)q(x) = acx^2 + ((a+b)(c+d) - ac - bd)x + bd.$$

$$bc = b(x)c(x) = (b_L x + b_R)(c_L x + c_R)$$

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=  $b_L c_L x^2 + (b_L c_R + b_R c_L)x + b_R c_R$ 

$$bc = b(x)c(x) = (b_L x + b_R)(c_L x + c_R)$$

$$= b_L c_L x^2 + (b_L c_R + b_R c_L)x + b_R c_R$$

$$= (b_L * c_L)x^2 + ((b_L + b_R) * (c_L + c_R) - b_L * c_L - b_R * c_R)x$$

$$+ b_R * c_R$$

$$bc = b(x)c(x) = (b_L x + b_R)(c_L x + c_R)$$

$$= b_L c_L x^2 + (b_L c_R + b_R c_L)x + b_R c_R$$

$$= (b_L * c_L)x^2 + ((b_L + b_R) * (c_L + c_R) - b_L * c_L - b_R * c_R)x$$

$$+ b_R * c_R$$

Recursively compute only  $b_L c_L$ ,  $b_R c_R$ ,  $(b_L + b_R)(c_L + c_R)$ .

$$bc = b(x)c(x) = (b_L x + b_R)(c_L x + c_R)$$

$$= b_L c_L x^2 + (b_L c_R + b_R c_L)x + b_R c_R$$

$$= (b_L * c_L)x^2 + ((b_L + b_R) * (c_L + c_R) - b_L * c_L - b_R * c_R)x$$

$$+ b_R * c_R$$

Recursively compute only  $b_L c_L$ ,  $b_R c_R$ ,  $(b_L + b_R)(c_L + c_R)$ .

#### Time Analysis

Running time is given by

$$T(n) = 3T(n/2) + O(n)$$
  $T(1) = O(1)$ 

which means 
$$T(n) = O(n^{\log_2 3}) = O(n^{1.585})$$

#### State of the Art

Schönhage-Strassen 1971:  $O(n \log n \log \log n)$  time using Fast-Fourier-Transform (FFT)

Martin Fürer 2007:  $O(n \log n2^{O(\log^* n)})$  time

Conjecture: There is an  $O(n \log n)$  time algorithm.