

Announcements: quiz Friday, equatio, Plickers

## Course Organization

- Algorithmic Paradigms
  - Divide & Conquer \*
  - Dynamic Programming
  - Greedy
- Applications to graph Problems
- What problems can we not solve efficiently

## Recall goal:

Appreciate beauty, creativity, surprising cleverness of algorithm design. Next algorithm hopefully touches on some of these...

example:

Multiplication - basic operation used frequently in many applications

Input: 2  $n$ -digit integers  $a, b$ .  $a = a_{n-1} \dots a_1 a_0$   $\left( a = \sum_{i=0}^{n-1} a_i 10^i \right)$   
 $b = b_{n-1} \dots b_1 b_0$

Output: Integer  $c = a \times b$

① Suppose multiplying/adding 2 one-digit numbers takes  $O(1)$  time.

Using "grade school" algorithm, what is runtime of multiplying  $a$  and  $b$ ?

- A)  $O(n)$     B)  $O(n^2)$     C)  $O(\log(n))$     D) Not enough information

Use this example to remind ourselves about runtime and big- $O$  notation

ex:

$$\begin{array}{r}
 \text{3} \quad 512 \\
 \text{2} \quad 629 \\
 \hline
 4608 \\
 1024 \\
 3072 \\
 \hline
 322048
 \end{array}$$

Step 1

Produce these numbers.

Each digit requires 1 multiplication &  $\leq 1$  addition  
 $n^2$  times  $\rightarrow 2n^2$  operations

from carryover  
 $\downarrow$

Step 2

Produce this number (adding  $n$  digit numbers uses  
 $n$  time, keep a running sum)  $\rightarrow \frac{3}{2}(n^2 + n)$  operations

Total:  $\frac{7}{2}n^2 + \frac{3}{2}n \rightarrow O(n^2)$  runtime  
 operations

Teaser: We'll be able to do in  $O(n^{1.58})$  using Karatsuba multiplication!

Example brings up 2 guiding principles:

## Principles of Algorithm Analysis

1. Use worst-case analysis. e.g. Counted 1 addition operation for carryover for every multiplication, even though might not actually use.
  - upper bounds hold for all instances
  - don't need extra info about problem
  - easier

2. Use asymptotic analysis (big-O notation)

Q. Why do we use big-O notation?

- Only care about big problems (otherwise do by hand or using brute force algorithm) and for big problems, lower order terms don't matter  
e.g. Insertion Sort:  $\frac{1}{2}n^2$  Merge Sort:  $6n(\log_2 n + 1)$

- Independent of architecture
- Easier
- Predicts how time will scale as problem size increases

# Divide & Conquer (& Combine)

↓  
Split big problem into smaller versions of same problem.

↓  
Solve smaller problems via recursion

↓  
combine solutions to get solution to big problem

Already Seen!

## Merge Sort

Input: Array A of integers of size n

Output: Sorted array

MergeSort(A)  
if  $\text{length}(A) = 1$ : return A } base case

$A1 = \text{MergeSort}(A[1 : \frac{n}{2}])$

$A2 = \text{MergeSort}(A[\frac{n}{2} + 1 : n])$

$p1 = p2 = 1$

for  $i = 1$  to  $n$

if  $A1[p1] < A2[p2]$

$A[i] = A1[p1]$

$p1++$

else

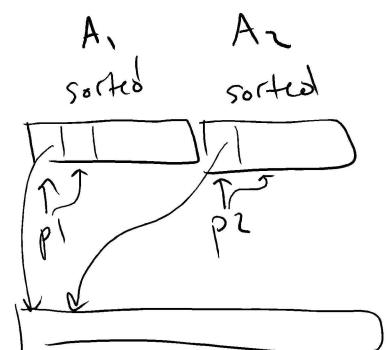
$A[i] = A2[p2]$

$p2++$

return A

} DIVIDE & CONQUER

COMBINE



Q: Create a Divide & Conquer Algorithm for Multiplication  
( $a, b$   $n$  digit numbers,  $n$  is a power of 2)

Base Case: If 1 digit numbers, can do in  $O(1)$  time

Divide:  $a = \overbrace{(a_{n-1} \dots a_{n/2})}^{a'} \cdot 10^{n/2} + \overbrace{(a_{n/2-1} \dots a_0)}^{a''}$   $\neq a', a'', b', b''$   
 $b = \overbrace{(b_{n-1} \dots b_{n/2})}^{b'} \cdot 10^{n/2} + \overbrace{(b_{n/2-1} \dots b_0)}^{b''}$  each  $\frac{n}{2}$  digits!

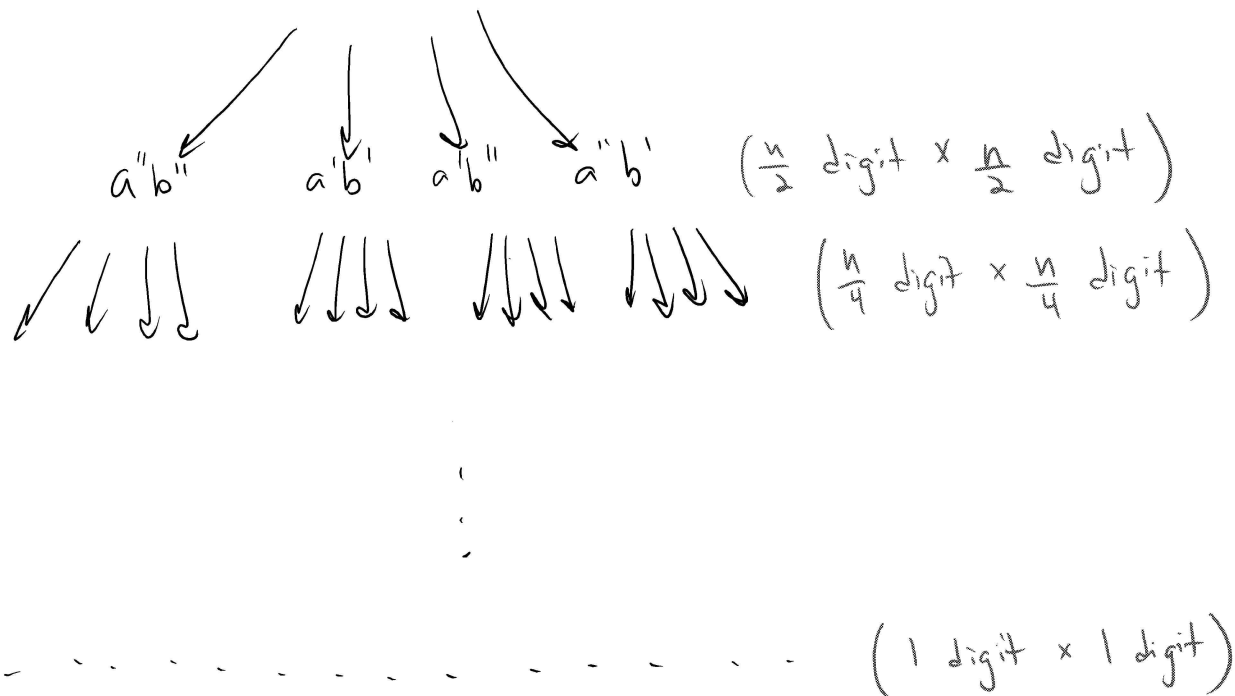
e.g.  $5284 = 5 \times 10^3 + 2 \times 10^2 + 8 \times 10 + 4$   
 $= 52 \times 10^2 + 84$

$$a \cdot b = (a' \cdot 10^{n/2} + a'') \cdot (b' \cdot 10^{n/2} + b'')$$

$$= a' \cdot b' \cdot 10^n + (a''b' + a'b'') 10^{n/2} + a''b''$$

Conquer: Solve smaller problems:  $a'b'$ ,  $a''b'$ ,  $a'b''$ ,  $a''b''$

Combine: Add  $a'b' \cdot 10^n + (a''b' + a'b'') 10^{n/2} + a''b''$

$a \cdot b \quad (n \text{ digit} \times n \text{ digit})$ 


Q: How many 1 digit  $\times$  1 digit multiplication problems will this algorithm require? ( $n \text{ digit} \times n \text{ digit}$ )

A)  $\log_2 n$ B)  $n$ C)  $n^2$ D)  $2^n$ 

Each level increases by factor of 4:

$$\underbrace{4 \cdot 4 \cdot 4 \cdot 4 \cdots 4}_{\log_2 n} = 4^{\log_2 n} = (2^2)^{\log_2 n} = (2^{\log_2 n})^2 = n^2$$

This is bad! There are at least  $n^2$  basic operations, so time complexity is  $\Omega(n^2)$ . No better than regular!