Week 7 - Wednesday





- What did we talk about last time?
- Closest pair of points

#### **Questions?**

# Logical warmup

- A man has two 10 gallon jars
- The first contains 6 gallons of wine and the second contains 6 gallons of water
- He poured 3 gallons of wine into the water jar and stirred
- Then he poured 3 gallons of the mixture in the water jar into the wine jar and stirred
- Then he poured 3 gallons of the mixture in the wine jar into the water jar and stirred
- He continued the process until both jars had the same concentration of wine
- How many pouring operations did he do?



# Three-Sentence Summary of Integer Multiplication

# **Integer Multiplication**

# **Multiplication by hand**

How long does it take to do multiplication by hand when we count single digit addition or multiplication as an operation?

> 123 <u>x 456</u> 738 615 <u>492</u> 56088

- Let's assume that the length of the numbers is n digits
- (n multiplications + n carries) x n digits + (n + 1 digits) x n additions
- Running time: O(n<sup>2</sup>)

#### Can we do better?

- Imagine that we're in base 2, because it keeps things simple
- I want to find the product  $x \cdot y$
- We can break *x* (and similarly *y*) into high-order part *x*<sub>1</sub> and low-order part *x*<sub>0</sub> such that *x* = *x*<sub>1</sub> · 2<sup>*n*/<sub>2</sub></sup> + *x*<sub>0</sub> *xy* = (*x*<sub>1</sub> · 2<sup>*n*/<sub>2</sub></sup> + *x*<sub>0</sub>)(*y*<sub>1</sub> · 2<sup>*n*/<sub>2</sub></sup> + *y*<sub>0</sub>)
  = *x*<sub>1</sub>*y*<sub>1</sub> · 2<sup>*n*</sup> + (*x*<sub>1</sub>*y*<sub>0</sub> + *x*<sub>0</sub>*y*<sub>1</sub>) · 2<sup>*n*/<sub>2</sub></sup> + *x*<sub>0</sub>*y*<sub>0</sub>

# Did that help things?

- Not really!
- We turned the multiplication of two *n*-bit numbers into the multiplication (and then addition) of four *n*/2-bit numbers
- $T(n) \le 4T\left(\frac{n}{2}\right) + cn$
- Which is  $O(n^{\log_2 4}) = O(n^2)$ , which is ... the same

#### We need a trick

- We want  $x_1y_1 \cdot 2^n + (x_1y_0 + x_0y_1) \cdot 2^{\frac{n}{2}} + x_0y_0$
- What if we compute
  - $a = (x_1 + x_0) \cdot (y_1 + y_0) = x_1y_1 + x_0y_1 + x_1y_0 + x_0y_0$
  - $b = x_1 y_1$
  - $c = x_0 y_0$
- Then,  $b \cdot 2^n + (a b c) \cdot 2^{\frac{n}{2}} + c =$
- $x_1y_1 \cdot 2^n + (x_1y_0 + x_0y_1) \cdot 2^{\frac{n}{2}} + x_0y_0$

# **Running time**

- We do two additions before the multiplies: O(n)
- We do three recursive multiplies of n/2-bit numbers
- We do two additions and two subtractions after the multiplies: O(n)
- $T(n) \leq 3T\left(\frac{n}{2}\right) + cn$
- Which is  $O(n^{\log_2 3}) \approx O(n^{1.59})$ , which is better!

## Note about multiplication

- Integer multiplication can be viewed as:
  - Polynomial multiplication
  - Vector convolution
- We will not cover section 5.6, but it describes the Fast Fourier Transform (FFT)
- The FFT can perform polynomial multiplication in O(n log n) time: even better than O(n<sup>1.59</sup>)
- However, the FFT has a big constant
- Using the FFT to multiply two integers only makes sense when the numbers are large, e.g., millions of digits

#### **Master Theorem**

#### **Master Theorem**

- Has a great name ...
- Allows us to determine the Big Theta running time of many recursive functions that would otherwise take more effort to determine

#### Basic form the recurrence relation must take

# $T(n) = aT\left(\frac{n}{b}\right) + f(n),$ where $a \ge 1$ and b > 1

- **a** is the number of recursive calls made
- b is how much the quantity of data is divided by each recursive call
- f(n) is the non-recursive work done at each step

#### Case 1

• If 
$$f(n)$$
 is  $O(n^{\log_b(a)-\epsilon})$   
for some constant  $\epsilon > 0$ , then  
 $T(n)$  is  $O(n^{\log_b(a)})$ 

#### Case 2

# • If f(n) is $\Theta(n^{\log_b(a)} \log^k n)$ for some constant $k \ge 0$ , then T(n) is $\Theta(n^{\log_b(a)} \log^{k+1} n)$



• If 
$$f(n)$$
 is  $\Omega(n^{\log_b(a)+\epsilon})$   
for some constant  $\epsilon > 0$ , and if  
 $af\left(\frac{n}{b}\right) \le cf(n)$   
for some constant  $c < 1$  and sufficiently large  $n$ ,  
then

$$T(n)$$
 is  $\Theta(f(n))$ 

# Stupid Sort

#### Stupid Sort algorithm (recursive)

- Base case: List has size less than 3
  - Swap out of order items if necessary
- Recursive case:
  - Recursively sort the first 2/3 of the list
  - Recursively sort the second 2/3 of the list
  - Recursively sort the first 2/3 of the list again



- How long does Stupid Sort take?
- We need to know log<sub>b</sub> a
- **a** = 3

Because I'm a nice guy, I'll tell you that the log<sub>1.5</sub> 3 is about 2.7



We know that binary search takes O(log *n*) time
Can we use the Master Theorem to check that?

### **Practicing the Master Theorem**

- One way to practice is to try to create a problems that different cases of the Master Theorem apply to
- Give a recurrence relation that uses Case 1
- Give a recurrence relation that uses Case 2
- Give a recurrence relation that uses Case 3



# Upcoming



- More Master Theorem examples
- Solved Exercises

### Reminders

- Work on Assignment 4
  - Due next Monday
- Exam 2 is next Wednesday
  - Review Chapters 4 and 5