

# **CS 170 Lecture Notes, Fall 2020**

## **Algorithms and Intractable Problems**

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## §1 September 1st, 2020

### §1.1 Naive Multiplication

Recall the example of Fibonacci: we went from a complexity of  $O(2^n) \rightarrow O(n^2) \rightarrow O(f(n))$ , where  $f(n)$  is the runtime for multiplying  $n$ -bit numbers. The naive algorithm is the usual multiplication algorithm for multiplying by hand.

$$\begin{array}{r} 17 \\ 21 \\ \hline 17 \\ 34x \\ \hline 357 \end{array}$$

Listing 1: Naive  $n$ -bit multiplication

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```

1 function multiply(x, y):
2   Input: n-bit integers, x, y, y >= 0
3   Output: Product
4
5   if (y == 0) return 0;
6   s = multiply(x, floor(y/2))
7   if y is even:
8     return 2s
9   else
10    return x+2s

```

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If we write  $x = \sum_{i=0}^{n-1} x_j 2^i$ ,  $y = \sum_{i=0}^{n-1} y_i 2^i$ , so

$$xy = \left( \sum_{i=0}^{n-1} x_j 2^i \right) \left( \sum_{i=0}^{n-1} y_i 2^i \right) = \sum_{j,k=0}^n x_j y_k 2^{j+k}.$$

### §1.2 Divide and Conquer: Karatsuba's Algorithm

For a Divide and Conquer problem, we do the following:

1. Break problem into pieces.
2. Solve pieces recursively.
3. Glue solutions of pieces to get solution of original problem.

We first have  $x$  an  $n$ -bit number that we break up into  $x_L, x_R$ , each  $n/2$  bit numbers. Similarly, we break  $y$  into  $y_L, y_R$ .

$$x = \underbrace{\begin{array}{cc} \xleftarrow{n/2} & \xrightarrow{n/2} \\ x_L & | & x_R \end{array}}_{n \text{ bit.}}$$

Note that  $x = 2^{n/2}x_L + x_R$ ,  $y = 2^{n/2}y_L + y_R$ , so

$$xy = 2^n x_L x_L + 2^{n/2}(x_L y_R + x_R y_L) + x_R y_R.$$

We now have 4 multiplications, involving  $n/2$ -bit numbers. Multiplication by  $2^m$  can be shifting ( $O(m)$  time), and addition is  $O(1)$ . Hence, we have a recurrence equation for the runtime,

$$T(n) = 4T\left(\frac{n}{2}\right) + O(n).$$



Note that the depth of the recursion tree is  $\log n$  and there are  $4^{\log n} = n^2$  leaves. But this would have the same runtime as the naive algorithm, so more work is required to optimize.

We note the following trick from Gauss:

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

and note that

$$(a + b)(c + d) = (ac + bd) + (ad + bc),$$

so it suffices to compute  $ac, bd, (a + b)(c + d)$ , since  $(a + b)(c + d) - ac - bd = ad + bc$ , which is the other term.

We can apply the trick in the following way:

$$(2^{n/2}x_L + x_R)(2^{n/2}y_L + y_R) = 2^n x_L x_L + 2^{n/2}(x_L y_R + x_R y_L) + x_R y_R$$

and

$$(x_L + x_R)(y_L + y_R) = x_L y_L + x_R y_R + (x_L y_R + x_R y_L).$$

Our new recurrence relation is

$$T(n) = 3T\left(\frac{n}{2}\right) + O(n).$$

Now, the branching factor is only 3, so the number of leaves is given by

$$3^{\log n} = n^{\log 3}.$$

### §1.3 Master Theorem

We have the following method of solving recurrence relations:

#### Theorem 1 (Master Theorem)

If  $T(n) = aT(\lceil n/b \rceil) + O(n^d)$  for constants  $a > 0, b > 1, d \geq 0$ , then

$$T(n) = \begin{cases} O(n^d) & \text{if } d > \log_b a \\ O(n^d \log n) & \text{if } d = \log_b a \\ O(n^{\log_b a}) & \text{if } d < \log_b a \end{cases}$$