

# week2\_lec3

## Main Ideas

- Nomenclature Of Problems
- Calculating the  $n^{th}$  Fibonacci Number  $F_n$
- Algorithms For Multiplying Large Integers

## Nomenclature of Problems

1. Tractable - Any problem that is solvable by a polynomial-time algorithm.

Upper Bound : Polynomial in nature

2. Intractable - Any problem that cannot be solved by a polynomial-time algorithm.

Lower Bound : Exponential in nature

## Problems

### Calculating the $n^{th}$ Fibonacci Number $F_n$

#### 1. Recursion Algorithm

Fibonacci number  $n$  is calculated using:  $F_N = F_{N-1} + F_{N-2}$

Time Complexity :  $T(n) = T(n-1) + T(n-2)$  : Exponential

The problem: WAP to take  $N$  and give  $F_N$

#### 2. Memoization Algorithm

This significantly lowers the complexity of time and space.

Time Complexity :  $O(n^2)$

```
int fib(int n) {
    int f[n + 2];
    int i;
    f[0] = 0;
    f[1] = 1;
```

```

for (i = 2; i <= n; i++)
{
    f[i] = f[i - 1] + f[i - 2];
}
return f[n];
}

```

### 3.Using Matrix Multiplication

$$\begin{pmatrix} F_n \\ F_{n+1} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}^n * \begin{pmatrix} F_0 \\ F_1 \end{pmatrix}$$

Time Complexity :  $O(M(n)\log n)$  [M(n) is the time taken for n bit matrix multiplication]

### 4.Direct Formula

$$F_n = \frac{1}{\sqrt{5}} \left( \frac{1+\sqrt{5}}{2} \right)^n - \frac{1}{\sqrt{5}} \left( \frac{1-\sqrt{5}}{2} \right)^2$$

Time Complexity:  $O(n\log^2 n)$ .

Problem in this method is that precision is very low.

Even if precise storage of 5 is achievable, it will still have to be raised to the nth power, which is worse than method 3.

## Algorithms For Multiplying Large Integers

### Traditional Method

Integers are expressed in binary form and multiplied directly with repeated addition.

**Time Complexity:**  $O(n^2)$

### Karatsuba Algorithm

The Karatsuba algorithm is a rapid multiplication method that uses the divide and conquer paradigm to multiply two  $n$ -digit integers.

Multiplying two complex no.s

$$(a + ib)(c + id) = (ac - bd) + i(ad + bc)$$

This operation naively takes four multiplications namely:  $ac, bd, ad, bc$

Instead we can:

- Compute  $ac$
- Compute  $bd$
- Compute  $(a + b)(c + d)$

$$\text{and then obtain } (ad + bc) = (a + b)(c + d) - ac - bd$$

So to multiply two  $n$ -bit integers  $x$  and  $y$ , partition each of them into parts that contain  $n/2$  of the bits each i.e.

$$x = 2^{n/2}a + b$$

$$y = 2^{n/2}c + d$$

therefore

$$x \cdot y = (2^{n/2}a + b) \cdot (2^{n/2}c + d)$$

Which is similar to the above method of computing complex products but instead of  $i$  we have  $2^{n/2}$

Time Complexity:  $O(n \log_2^3) = O(n^{1.585})$

**Algorithms with better complexities than  $O(n^{1.585})$**

1. Fast Fourier Transform -  $O(n \log n \log(\log n))$
2. Faster -  $O(n \log n \cdot 2O(\log n))$
3. Fastest -  $O(n \log n)$