# Asymptotic Analysis

- Runtime Analysis
- Big Oh  $O(\cdot)$
- Complexity Classes and Curse of Exponential Time
- $\Omega(\cdot), \Theta(\cdot), o(\cdot), \omega(\cdot)$  Relational properties

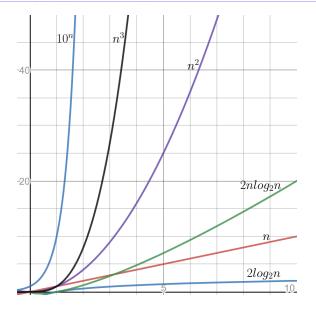
## Imdadullah Khan

## Asymptotic-Complexity Classes

Class Name	Class Symbol	Example	
Constant	O(1)	Comparison of two integers	
Logarithmic	O(log(n))	Binary Search, Exponentiation	
Linear	<i>O</i> ( <i>n</i> )	Linear Search	
Log-Linear	On(log(n))	Merge Sort	
Quadratic	$O(n^2)$	Integer multiplications	
Cubic	<i>O</i> ( <i>n</i> <sup>3</sup> )	Matrix multiplication	
Polynomial	$O(n^a),\ a\in\mathbb{R}$		
Exponential	$O(a^n),~a\in\mathbb{R}$	Print all subsets	
Factorial	O(n!)	Print all permutations	

 $n! \gg 2^n \gg n^3 \gg n^2 \gg n \log n \gg n \gg \log n \gg 1$ 

#### Growth Rates of Functions



#### Fibonacci Sequence

 $0, 1, 1, 2, 3, 5, 8, 13, 21, \ldots$ 

$$F_n = \begin{cases} 0 & \text{if } n = 0\\ 1 & \text{if } n = 1\\ F_{n-1} + F_{n-2} & \text{if } n > 2 \end{cases}$$

## Find $F_n$ : The curse of Exponential time

Implementation of the recursive definition of  $F_n$ 

```
function FIB1(n)

if n = 0 then

return 0

else if n = 1 then

return 1

else

return FIB1(n - 1) + FIB1(n - 2)
```

Is it correct?

- How much time it takes to compute  $F_n$ ?
- Can we do better?

#### Find $F_n$ : The curse of Exponential time

Let T(n) be the number of ops (comparisons and additions) on input n

$$T(n) = \begin{cases} 1 & \text{if } n = 0\\ 2 & \text{if } n = 1\\ T(n-1) + T(n-2) + 3 & \text{if } n > 2 \end{cases}$$

By definition, we have  $T(n) > F_n$ 

The running time of FIB1(n) grows as fast as  $F_n$ 

$$T(n) \geq 2^{.69n}$$

**exponential** in *n* (prove by induction)

#### Find $F_n$ : The curse of Exponential time

 $T(n) \geq 2^{.69n}$ 

- For n = 300, computing  $F_{300}$  takes (much) more than  $2^{150}$  ops
- On a 64*THz* computer (64  $\times$  2<sup>40</sup> operations per second)
- It needs  $2^{104}s > 10^{27}h > 10^{23}$  years

Another perspective to see growth of exponential time

• Runtime of FIB1(n) is  $\geq 2^{0.694n} \approx (1.6)^n$ 

• it takes 1.6 times longer to compute  $F_{n+1}$  than  $F_n$ 

- Moore's law  $\implies$  computers get roughly 1.6 times faster each year
- If we can compute  $F_{100}$  with this year's technology, next year we will manage  $F_{101}$ , the year after,  $F_{102}$ , ...

▷ one more Fibonacci number every year

#### Such is the curse of exponential time

#### How can we improve it?

Imdadullah Khan (LUMS)

#### Exponential vs Polynomial Growth rates

Sizes of problems that can be solved within  $10^{12}$  operations on today's computer and next years computer with double speed

Complexity	Increase	Problem Size (today)	Problem Size (next year)
n	$n \rightarrow 2n$	10 <sup>12</sup>	$2\times 10^{12}$
n <sup>2</sup>	$n \rightarrow \sqrt{2}n$	10 <sup>6</sup>	$1.4  imes 10^6$
n <sup>3</sup>	$n \rightarrow \sqrt[3]{2}n$	10 <sup>4</sup>	$1.25  imes 10^4$
2 <sup><i>n</i>/10</sup>	$n \rightarrow n + 10$	400	410
2 <sup><i>n</i></sup>	$n \rightarrow n+1$	40	41