### Asymptotic Analysis

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

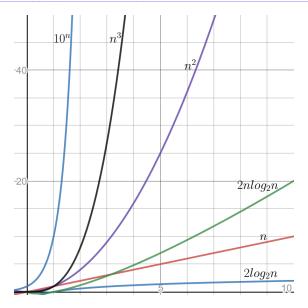
### IMDAD ULLAH KHAN

## Asymptotic-Complexity Classes

Class Name	Class Symbol Example		
Constant	O(1)	Comparison of two integers	
Logarithmic	O(log(n))	Binary Search, Exponentiation	
Linear	O(n)	Linear Search	
Log-Linear	On(log(n))	Merge Sort	
Quadratic	$O(n^2)$	Integer multiplications	
Cubic	$O(n^3)$	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

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

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

Implementation of the recursive definition of  $F_n$ 

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function {\rm FiB1}(n)

if n=0 then

return 0

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return {\rm FiB1}(n-1)+{\rm FiB1}(n-2)
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- Is it correct?
- How much time it takes to compute  $F_n$ ?
- Can we do better?

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

▷ exponential in n (prove by induction)

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- Runtime of FIB1(n) is  $\geq 2^{0.694n} \approx (1.6)^n$ 
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How can we improve it?



### 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>	$2\times10^{12}$
n <sup>2</sup>	$n  o \sqrt{2}n$	$10^{6}$	$1.4  imes 10^6$
$n^3$	$n  o \sqrt[3]{2}n$	10 <sup>4</sup>	$1.25\times10^4$
$2^{n/10}$	$n \rightarrow n + 10$	400	410
2 <sup>n</sup>	$n \rightarrow n+1$	40	41