

# Data Structures and Algorithms

## Lecture 1



# Why Data Structure & Algorithms:

Students gain a solid foundation in designing, implementing, and analyzing algorithms and data structures, essential for efficient problem-solving in programming and software development.





# Course Syllabus

- **1. Introduction to Data Structures and Algorithms**
  - Overview of fundamental concepts.
  - Importance of data structures in software development.
  - Basics of algorithmic thinking.
- **2. Arrays and Linked Lists**
  - Implementation and manipulation of arrays.
  - Singly and doubly linked lists.
  - Time and space complexity analysis.
- **3. Stacks and Queues**
  - Understanding stack and queue data structures.
  - Applications and practical implementation.
  - Evaluating efficiency in different scenarios.





# Course Syllabus (Cont.)

- **4. Trees and Graphs**
  - Binary trees and their variants.
  - Graph representation and traversal.
  - Tree and graph algorithms.
- **5. Sorting and Searching Algorithms**
  - Comparison of sorting algorithms (e.g., Bubble Sort, Quick Sort, Merge Sort).
  - Searching techniques (e.g., Linear Search, Binary Search).
  - Analysis of algorithmic complexity.
- **6. Algorithmic Paradigms**
  - Divide and conquer strategies.
  - Dynamic programming principles.
  - Greedy algorithms and their applications.
- **7. Practical Application**
  - Coding exercises applying learned concepts.





# Our Roadmap

Type	Marks	Week	Discussion
Mini Project	10	Week 5	Week 7
Final Project	10	Week 8	Week 13
Midterm	15	-	-
Final Exam	25	-	-
Lab Participation	15	-	-
Assessments (Lab)	15	-	-
Assessments (Lecture)	10	-	-





# Teaching Assistants

1- Aya Abd Elnabi

2-Hagar Sobeh



Textbooks:

"Introduction to Algorithms" by Thomas H. Cormen et al.

"Data Structures and Algorithms" by Michael T. Goodrich et al.



# What is an algorithm?







## Definition:

- An algorithm is a Step By Step process to solve a problem, where each step indicates an intermediate task.
- Algorithm contains finite number of steps that leads to the solution of the problem.





# What is an algorithm?

- It's a set of well-defined rules for solving some computational problem.
  - a bunch of numbers and you want to **rearrange them** so that they're in **sorted order**.
  - you have a road map and you want to **compute the shortest path** from some origin to some destination.
  - you need to complete several tasks before certain deadlines, and you want to **know in what order you should finish the tasks** so that you complete them all by their respective deadlines.



# Why Study Algorithms?



# Why Study Algorithms?

- Important for all other branches of computer science
  1. Routing protocols in communication networks take credit of **classical shortest path algorithms**.
  2. Public-key cryptography relies on **efficient number-theoretic algorithms**.
  3. Computer graphics requires the **computational primitives** supplied by **geometric algorithms**.
  4. Database indices rely on **balanced search tree data structures**.
  5. Computational biology uses **dynamic programming algorithms** to measure genome similarity.

Challenging (Good for the brain)



# Properties / Characteristics of an Algorithm





## Algorithm has the following basic properties:

1. **Input-Output:** Algorithm takes input and produces the required output. This is the basic characteristic of an algorithm.
2. **Finiteness:** An algorithm must terminate in countable number of steps.
3. **Definiteness:** Each step of an algorithm must be stated **clearly** and **unambiguously**.
4. **Effectiveness:** Each and every step in an algorithm **can be converted** in to programming language statement.
5. **Generality:** Algorithm is generalized one. It works on all set of inputs and provides the required output. In other words it is **not restricted** to a single input value.



# Categories of Algorithm:

- Based on the different types of steps in an Algorithm, it can be **divided into three categories**, namely The written program is buggy.
  - Sequence
  - Selection and
  - Iteration





## Sequence:

- The steps described in an algorithm are performed successively one by one without skipping any step.
- The sequence of steps defined in an algorithm should be simple and easy to understand.
- Each instruction of such an algorithm is executed, because no selection procedure or conditional branching exists in a sequence algorithm.







## Sequence Example:

// adding two numbers

Step 1: start

Step 2: read a,b

Step 3:  $\text{Sum} = a + b$

Step 4: write Sum

Step 5: stop





# Selection:

- The sequence type of algorithms are not sufficient to solve the problems, which involves decision and conditions.
- In order to solve the problem which involve decision making or option selection, we go for Selection type of algorithm.
- The general format of Selection type of statement is as shown below:

```
if(condition)
    Statement 1;
else
    Statement 2;
```

- The above syntax specifies that **if the condition is true**, **statement 1 will be executed** **otherwise statement 2 will be executed**.





## Selection Examples:

Example1:

// Person eligibility for vote

Step 1 : start

Step 2 : read age

Step 3 : if age  $\geq$  18 then step\_4 else step\_5

Step 4 : write "person is eligible for vote"

Step 5 : write " person is not eligible for vote"

Step 6 : stop

Example2:

// biggest among two numbers

Step 1 : start

Step 2 : read a,b

Step 3 : if a  $>$  b then

Step 4 : write "a is greater than b"

Step 5 : else

Step 6 : write "b is greater than a"

Step 7 : stop





# Iteration:

- Iteration type algorithms are used in solving the problems which involves repetition of statement.
- In this type of algorithms, a particular number of statements are repeated 'n' no. of times.



# Iteration Example:

Step 1 : start

Step 2 : **counter** =1

Step 3 : **sum**=0

Step 2 : read grade

Step 3 : repeat step 4 until **counter** > 5

Step 4 : (a)  $\text{sum} = \text{grade} + \text{sum}$

(b)  $\text{Average} = \text{sum} / 5$

(c)  $\text{counter} = \text{counter} + 1$

Step 5 : write Average

Step 6 : stop

The input: 10, 25, 30, 40, 50



The output:  $215/5 = 43$





# Algorithm analysis Importance

- To detect
  - The written `program` is `buggy`.
  - The `program` may be `inefficient`
  - If the program is running on a large data set, then `the running time becomes an issue`





# Analysis Aspects

- Correctness
  - Does the input/output relation match algorithm requirement?
- Amount of work done (complexity)
  - Basic operations to do task in a finite amount of time
- Amount of space used e.g. Memory used



# Performance Analysis of Algorithm







## Performance Analysis an Algorithm:

- The Efficiency of an Algorithm can be measured by the following metrics:
  - i. Time Complexity and
  - ii. Space Complexity.





# Performance Analysis an Algorithm:

## i. Time Complexity:

- The amount of time required for an algorithm to complete its execution is its time complexity.
- An algorithm is said to be efficient if it takes the minimum (reasonable) amount of time to complete its execution.

## ii. Space Complexity:

- The amount of space occupied by an algorithm is known as Space Complexity.
- An algorithm is said to be efficient if it occupies less space and required the minimum amount of time to complete its execution



# Asymptotic analysis





## Asymptotic analysis:

- Asymptotic analysis refers to computing the running time of any operation in mathematical units of computation.
- For example, the running time of one operation is computed as  $O(n)$  and may be for another operation it is computed as  $O(n^2)$ .
- This means the first operation running time will increase linearly with the increase in  $n$  and the running time of the second operation will increase exponentially when  $n$  increases. Similarly, the running time of both operations will be nearly the same if  $n$  is significantly small.

## ASYMPTOTIC ANALYSIS (High Level Idea)

*We'll express the asymptotic runtime of an algorithm using*

# BIG-O NOTATION

THE POINT OF ASYMPTOTIC NOTATION

**suppress constant factors and lower-order terms**

*too system dependent*

*irrelevant for large inputs*



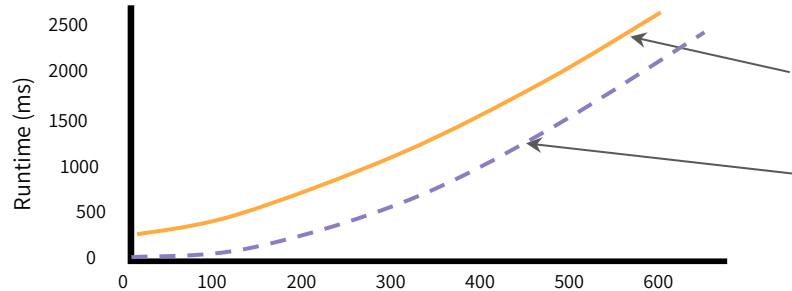
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THE POINT OF ASYMPTOTIC NOTATION

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$$0.1n^{1.6} + 300$$

$$.008n^2$$

n (input size)



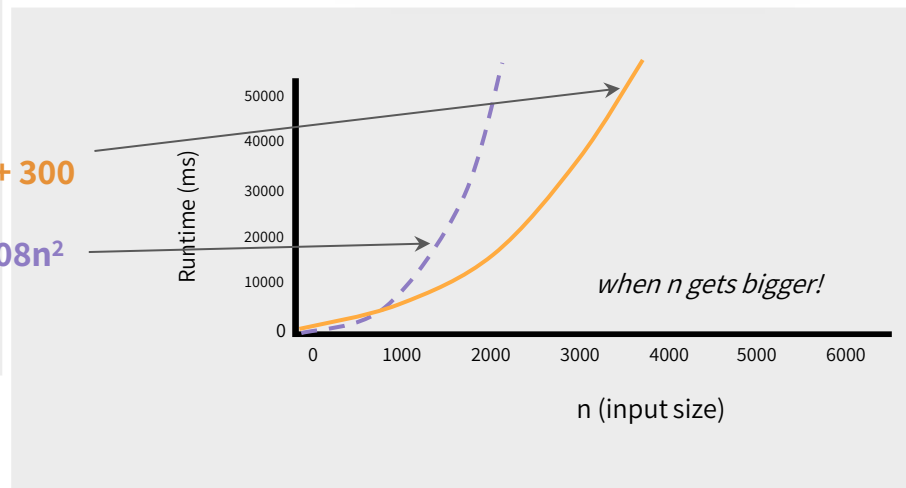
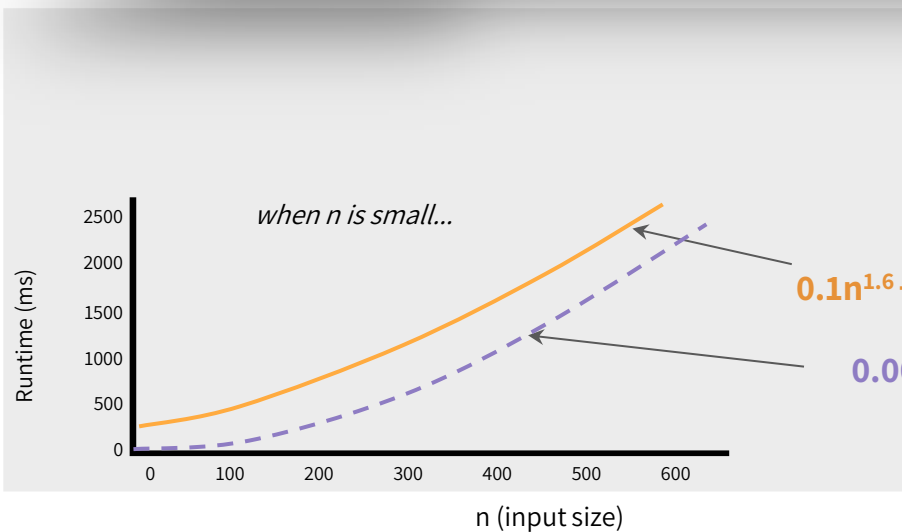
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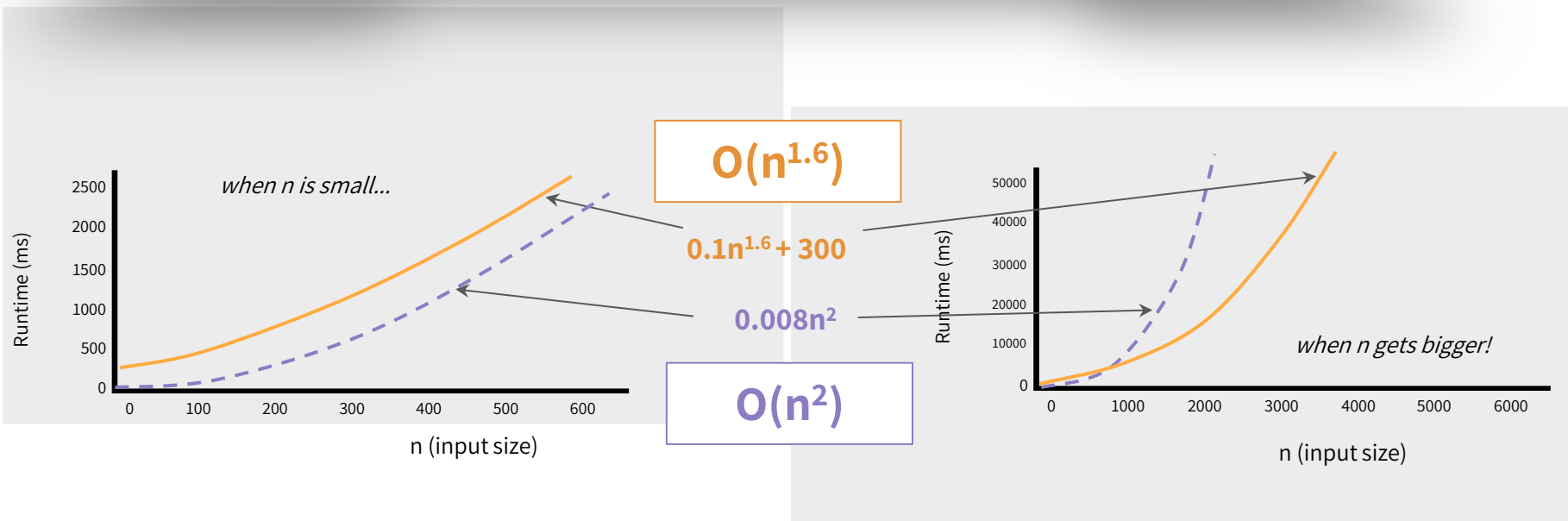
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# ASYMPTOTIC ANALYSIS (High Level Idea)

- To **compare algorithm runtimes** in this class, **we compare their Big-O runtimes**
  - Ex: a runtime of  $O(n^2)$  is considered “better” than a runtime of  $O(n^3)$
  - Ex: a runtime of  $O(n^{1.6})$  is considered “better” than a runtime of  $O(n^2)$
  - Ex: a runtime of  $O(1/n)$  is considered “better” than  $O(1)$ 

$O(1/n) = O(1 * n^{-1})$

$O(1) = (1 * n^0)$



## The time required by an algorithm falls under three types:

- **Best Case:**
  - Minimum time required for program execution.
- **Average Case:**
  - Average time required for program execution.
- **Worst Case:**
  - Maximum time required for program execution.

# ASYMPTOTIC NOTATIONS



## **Asymptotic Notations:**

Following are the commonly used asymptotic notations to calculate the running time complexity of an algorithm.

- O Notation
- $\Omega$  Notation
- $\theta$  Notation

**Worst Case**

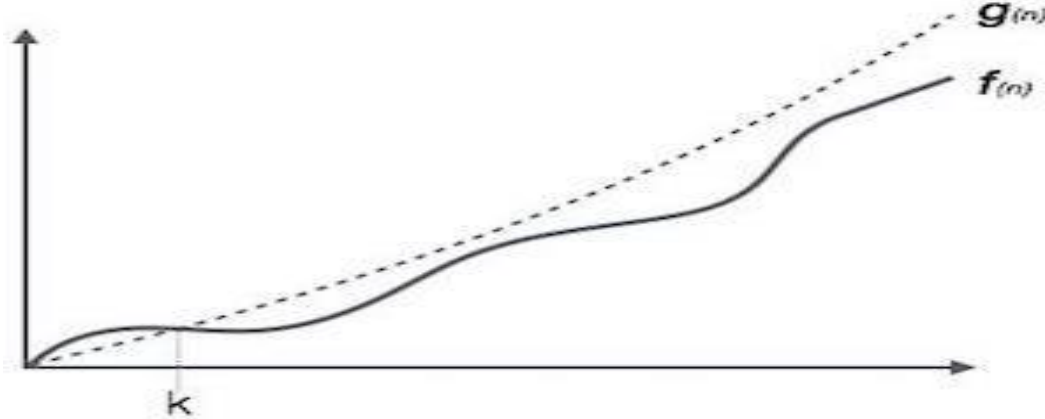
**Big Oh Notation, 0**



## Big Oh Notation, O:

- The notation  $O(n)$  is the formal way to express the upper bound of an algorithm's running time.
- It measures the worst case time complexity or the longest amount of time an algorithm can possibly take to complete.

## Big Oh Notation, O:



For example, for a function  $f(n)$

$$O(f(n)) = \{ g(n) : \text{there exists } c > 0 \text{ and } n_0 \text{ such that } f(n) \leq c \cdot g(n) \text{ for all } n > n_0. \}.$$

Best Case

Omega Notation,  $\Omega$

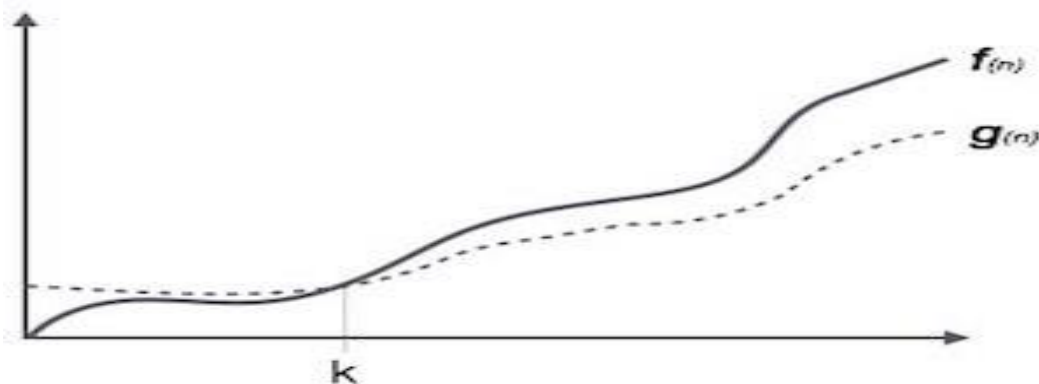




## Omega Notation, $\Omega$ :

- The notation  $\Omega(n)$  is the formal way to express the lower bound of an algorithm's running time.
- It measures the best case time complexity or the best amount of time an algorithm can possibly take to complete.

## Omega Notation, $\Omega$ :



For example, for a function  $f(n)$

$$\Omega(f(n)) = \{ g(n) : \text{there exists } c > 0 \text{ and } n_0 \text{ such that } g(n) \leq c \cdot f(n) \text{ for all } n > n_0. \}.$$

Average Case

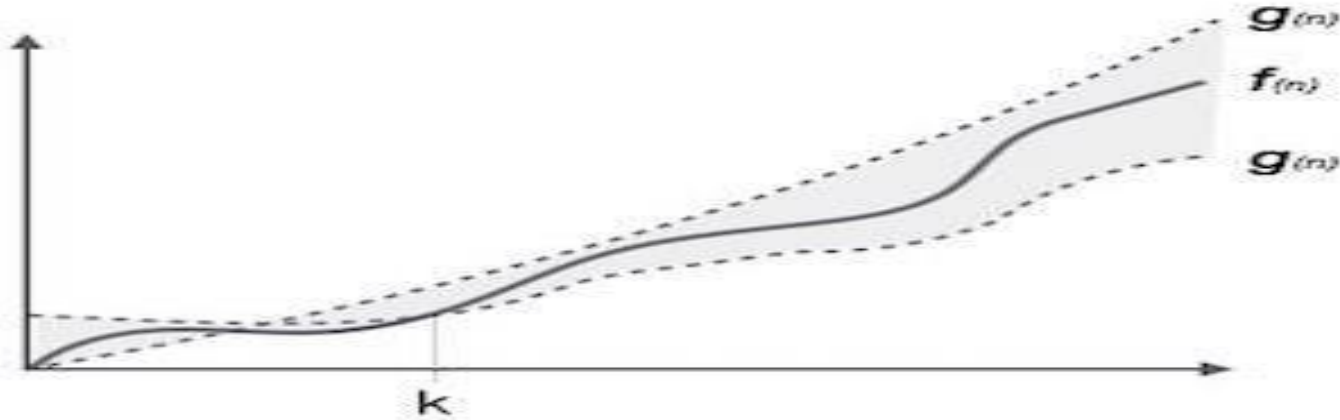
Theta Notation,  $\theta$



## Theta Notation, $\theta$ :

The notation  $\theta(n)$  is the formal way to express both the lower bound and the upper bound of an algorithm's running time.

## Big Oh Notation, O:



It is represented as follows

$$\theta(f(n)) = \{ g(n) \text{ if and only if } g(n) = O(f(n)) \text{ and } g(n) = \Omega(f(n)) \text{ for all } n > n_0. \}.$$



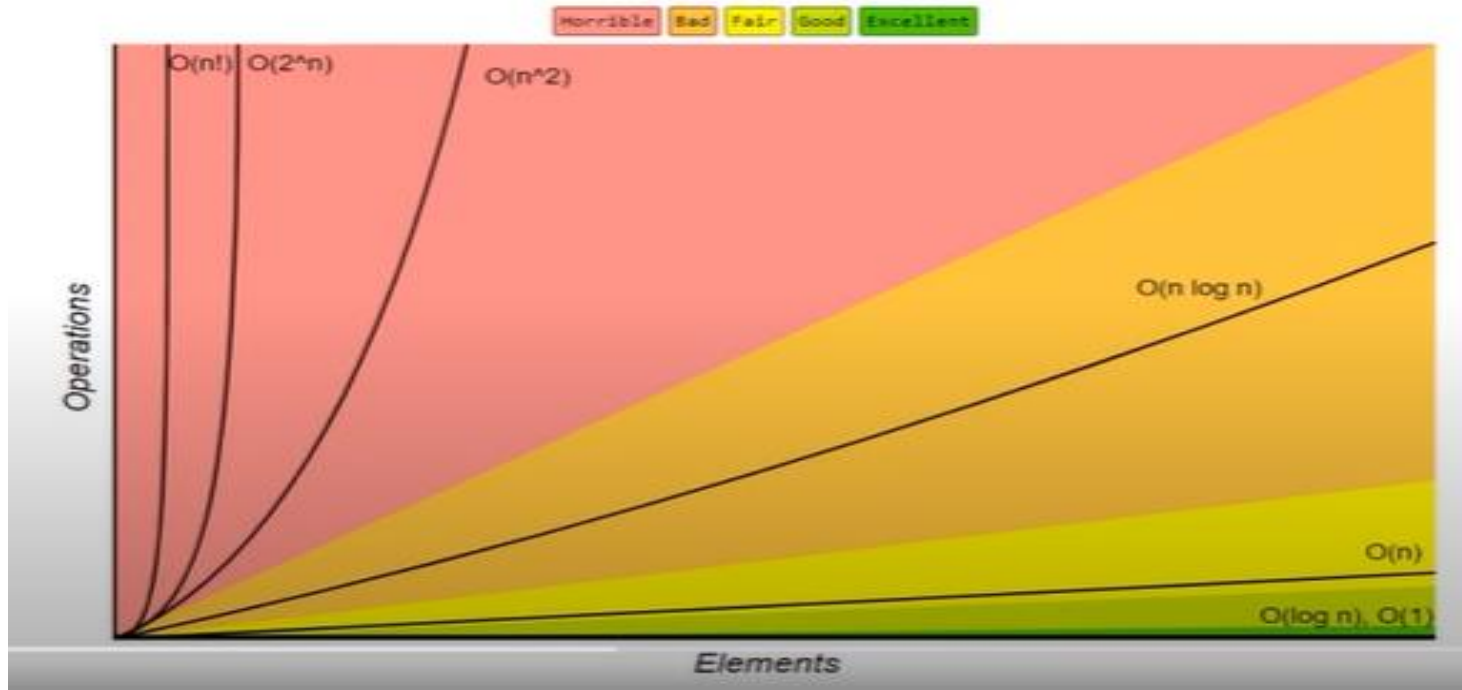
# Some common Big O run times

- $O(\log n)$ , also known as **log time**. Example: **Binary search**.
- $O(n)$ , also known as **linear time**. Example: **Simple search**.
- $O(n * \log n)$ . Example: A **fast sorting algorithm**, like **quicksort**.
- $O(n^2)$ . Example: A **slow sorting algorithm**, like **selection sort**.
- $O(n!)$ . Example: A **really slow algorithm**, like the **traveling salesperson**.





# Growth Rates of Common Functions





# Growth Rates of Common Functions

$n$	$f(n)$	$\lg n$	$n$	$n \lg n$	$n^2$	$2^n$	$n!$
10		0.003 $\mu s$	0.01 $\mu s$	0.033 $\mu s$	0.1 $\mu s$	1 $\mu s$	3.63 ms
20		0.004 $\mu s$	0.02 $\mu s$	0.086 $\mu s$	0.4 $\mu s$	1 ms	77.1 years
30		0.005 $\mu s$	0.03 $\mu s$	0.147 $\mu s$	0.9 $\mu s$	1 sec	$8.4 \times 10^{15}$ yrs
40		0.005 $\mu s$	0.04 $\mu s$	0.213 $\mu s$	1.6 $\mu s$	18.3 min	
50		0.006 $\mu s$	0.05 $\mu s$	0.282 $\mu s$	2.5 $\mu s$	13 days	
100		0.007 $\mu s$	0.1 $\mu s$	0.644 $\mu s$	10 $\mu s$	$4 \times 10^{13}$ yrs	
1,000		0.010 $\mu s$	1.00 $\mu s$	9.966 $\mu s$	1 ms		
10,000		0.013 $\mu s$	10 $\mu s$	130 $\mu s$	100 ms		
100,000		0.017 $\mu s$	0.10 ms	1.67 ms	10 sec		
1,000,000		0.020 $\mu s$	1 ms	19.93 ms	16.7 min		
10,000,000		0.023 $\mu s$	0.01 sec	0.23 sec	1.16 days		
100,000,000		0.027 $\mu s$	0.10 sec	2.66 sec	115.7 days		
1,000,000,000		0.030 $\mu s$	1 sec	29.90 sec	31.7 years		







# Growth Rates of Common Functions

$n$	$f(n)$	$\lg n$	$n$	$n \lg n$	$n^2$	$2^n$	$n!$
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# Thank You

