

**Your Ultimate Guide To Landing
Top AI roles**

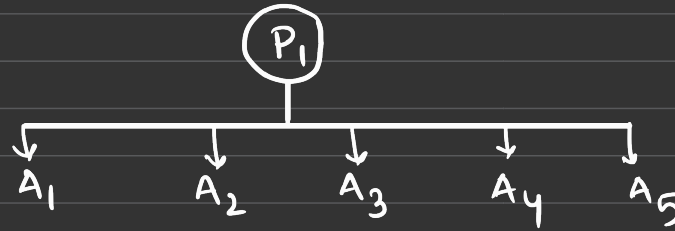


2.3.1

Time and space complexity



- For Solving any problem P_1 , there will be many solutions
- Before implementing any algorithm as a Program we must have to find out which algorithm is good in terms of time and memory.



- After analysis, we will choose the best algorithm in terms of Time and Space
- How to calculate time and space complexity?
 - ↳ Asymptotic notation
 - ↳ Big-O

Asymptotic Notation



→ Asymptotic notation describes the performance of algorithms in terms of input size (n), particularly for large inputs.

↑
time and space
Complexity.

→ It gives a mathematical way to express the growth rate of an algorithm's resource (Time and Space) Consumption

→ Types of Asymptotic Notation

① Big- $O(O)$ → Worst-case time and space

② Omega (Ω) → Best-case time and space

③ Theta (Θ) → Average-case time and space

Q. Given a List of numbers of size n . Find the best, average and worst Case time Complexity to search for number x .



NOTE - List is not sorted.

Time Complexity = $T(n)$

→ Worst case time Complexity $\Rightarrow O(n) \leftarrow$ last/donot exists

Average case time complexity $\Rightarrow Q(n) \leftarrow$ target at middle

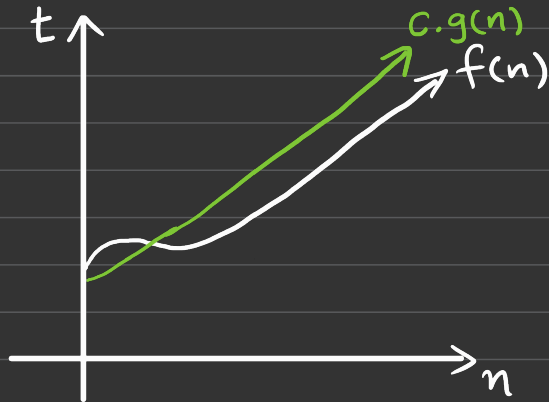
Best case time Complexity $\Rightarrow \Omega(1) \leftarrow$ first position

↳ what about Space Complexity?

Big-O Notation (O)



- Big-O defines the maximum time/space an algorithm can take
- Focus on worst-case Scenario.
- Analogy : Going to Airport
- we will find another function for a given function Such that after some input n_0 , the value of $c.g(n)$ is always greater than $f(n)$



$$f(n) \leq c.g(n)$$

for $n \geq n_0$

$c > 0, n_0 \geq 1$

$$f(n) = O(g(n))$$

→ Big-O is an upper bound, not necessarily a tighter upper bound.



Example : $f(n) = 3n + 2$

→ $g(n) = n$

$$f(n) \leq c \cdot g(n), \quad c > 0 \\ n_0 \geq 1$$

$$3n + 2 \leq c \cdot n$$

Put $c = 4$

$$3n + 2 \leq 4 \cdot n$$

$$2 \leq n \Rightarrow n \geq 2$$

$$\therefore f(n) = O(g(n)) = O(n)$$

Shortcut

☆☆

① Keep the highest order term

→ Drop all the lower-order terms — they become insignificant for large input size ($n \rightarrow \infty$)

② Ignore Constants

→ Coefficients don't affect asymptotic growth

$$T(n) = 100n \rightarrow O(n)$$

Comparison of Functions → To find highest order term



→ when we compare two functions, we are interested in values at $n \rightarrow \infty$ (i.e. when n grows very large, which one wins)

Examples

① $n \square n^5$

② $3^n \square 2^n$

③ $2^n \square n^2$

④ $n^2 \square n \cdot \log n$

⑤ $n \square (\log n)^{100}$

⑥ $n^{\log n} \square n \cdot \log n$

⑦ $\sqrt{\log n} \square \log \log n$

TRICK

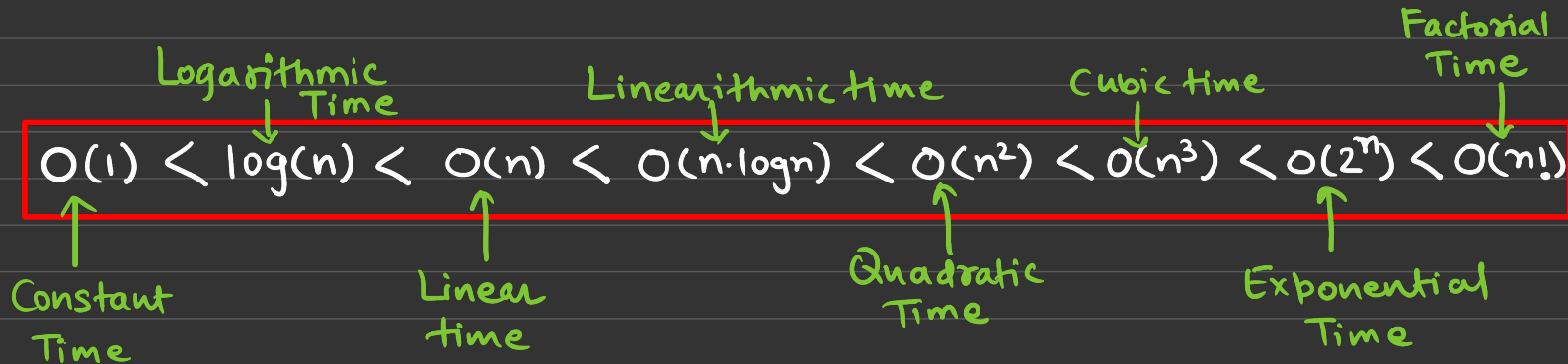
- ① Take larger n value
- ② Apply log and then ①

⑧ $n^{\sqrt{n}} \square n^{\log n}$

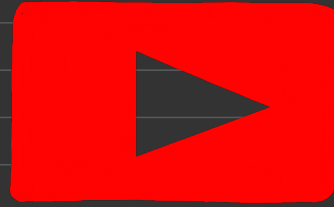
⑨ $f(n) = \begin{cases} n^3 & 0 < n < 10000 \\ n^2 & n \geq 10000 \end{cases}$

$g(n) = \begin{cases} n & 0 < n < 100 \\ n^3 & n \geq 100 \end{cases}$

Function Growth (Fastest to Slowest)



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