

Big O Basic Concepts:

- **$O(1)$** : Constant Time
 - Doesn't depend on the size of the data set.
 - Example: Accessing an array element by its index.
- **$O(\log n)$** : Logarithmic Time
 - Splits the data in each step (divide and conquer).
 - Example: Binary search.
- **$O(n)$** : Linear Time
 - Directly proportional to the data set size.
 - Example: Looping through an array.
- **$O(n \log n)$** : Linearithmic Time
 - Splits and sorts or searches data.
 - Example: Merge sort, quick sort.
- **$O(n^2)$** : Polynomial Time
 - Nested loops for each power of n .
 - Example: Bubble sort ($O(n^2)$).

Omega (Ω) – Best Case

- **What it means:** Omega (Ω) describes the best-case scenario for an algorithm.
- **In simple terms:** It tells you the fastest an algorithm can run in the best circumstances.

Theta (Θ) - Average Case

- **In simple terms:** It tells you what to generally expect in terms of time complexity.

Big O (O) - Worst Case

- **What it means:** Big O (O) describes the worst-case scenario for an algorithm.
- **In simple terms:** It tells you the slowest an algorithm can run in the worst circumstances.

Useful Tips

- **Drop Non-Dominant Terms**
 - In $O(n^2 + n)$, focus on $O(n^2)$ as it will dominate for large n .
- **Drop Constants**
 - $O(2n)$ simplifies to $O(n)$.