

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 $\propto T$
 - 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.

Other Concepts:

- 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 $O(n)$.