## **Big O Notation**

Big O notation describes the upper bound of an algorithm's running time as a function of the input size. It provides a high-level understanding of the algorithm's efficiency and scalability, helping to identify potential performance bottlenecks.

- O(1): Constant time; the operation's duration is independent of the input size.
- **O(n)**: Linear time; the operation's duration grows linearly with the input size.
- O(log n): Logarithmic time; the operation's duration grows logarithmically with the input size.
- O(n log n): Log-linear time; common in efficient sorting algorithms like merge sort and quicksort.
- **O(n²)**: Quadratic time; the operation's duration grows quadratically with the input size, often seen in simple nested loops.

## **Best, Average, and Worst-Case Scenarios**

- Best Case: The minimal time an algorithm takes to complete, usually occurring with optimal input.
- Average Case: The expected time taken for random input, giving a realistic performance expectation.
- Worst Case: The maximum time an algorithm could take, providing an upper bound on performance, crucial for ensuring the system's worst-case efficiency.

## **Time Complexity**

- Linear Search:
  - **Best Case**: O(1) When the target is the first element.
  - Average Case: O(n) On average, half the array is searched.
  - o Worst Case: O(n) When the target is the last element or not present.
- Binary Search:
  - Best Case: O(1) When the target is the middle element on the first check.
  - Average/Worst Case: O(log n) Divides the search space in half each time.

**Binary Search** is generally more suitable for e-commerce platforms due to its O(log n) complexity, making it highly efficient for large datasets. However, it requires the data to be sorted, which adds overhead when inserting new products.

**Linear Search** can be used for unsorted data or when the dataset is relatively small, as it doesn't require sorting. However, its O(n) complexity makes it less efficient for large datasets.