#### Unit -2

## 1. Introduction to Recursion and Divide & Conquer

- A **recursive algorithm** calls itself to solve smaller versions of the same problem.
- If the problem can be broken into **smaller sub-problems**, solved individually, and combined back, this is called **Divide and Conquer (D&C)**.
- Steps in D&C:
  - 1. **Divide** the problem into sub-problems.
  - 2. **Conquer** by solving them (recursively).
  - 3. **Combine** their results to get the final solution.

**Examples:** Merge Sort, Quick Sort, Binary Search, Strassen's Matrix Multiplication.

#### 2. Recurrence Relations

• When analyzing recursive algorithms, we often get equations like:

$$T(n) = aT(n/b) + f(n)$$

- o a = number of sub-problems
- o n/b = size of each sub-problem
- o f(n) = extra work (divide + combine)

This is called a **recurrence relation**. We solve it to find time complexity.

#### 3. Methods to Solve Recurrence

### (a) Substitution Method

- Guess the solution (like O(n log n)) and prove it using induction.
- Example: Binary Search  $\rightarrow$  T(n) = T(n/2) + O(1)  $\rightarrow$  solution = O(log n).

# (b) Recursion Tree Method

- Draw a tree where each level shows the work done in recursive calls.
- Add up the cost at each level.
- Example: Merge Sort  $\rightarrow$  tree has log n levels, each level costs O(n). Total = O(n log n).

### (c) Master Theorem

For T(n) = aT(n/b) + f(n):

- Compare f(n) with n^(log\_b a).
- 1. If f(n) grows slower  $\rightarrow O(n^{(\log_b a)})$

- 2. If f(n) grows same  $\rightarrow O(n^{(\log_b a)} * \log n)$
- 3. If f(n) grows faster  $\rightarrow O(f(n))$  (if regularity condition holds).

## **Example:**

- Merge Sort:  $T(n) = 2T(n/2) + O(n) \rightarrow Case 2 \rightarrow O(n log n)$ .
- Binary Search:  $T(n) = T(n/2) + O(1) \rightarrow Case 1 \rightarrow O(log n)$ .

# 4. Binary Search

- Works on a sorted array.
- Steps:
  - 1. Find middle element.
  - 2. If target = mid  $\rightarrow$  found.
  - 3. If target < mid  $\rightarrow$  search left half.
  - 4. Else → search right half.
- Time Complexity = **O(log n)**, Space = O(1) (iterative) or O(log n) (recursive).

## **5. Multiplying Large Integers**

- Normal multiplication takes O(n²).
- Using Divide & Conquer (Karatsuba's algorithm), we reduce work.
- Idea: Split numbers into halves, do fewer multiplications + additions.
- Time Complexity = O(n^1.59) (faster than O(n²)).

## 6. Matrix Multiplication

- Normal multiplication: O(n³).
- **Strassen's Algorithm**: reduces number of multiplications from 8 to 7 (by using clever formulas).
- Time Complexity = O(n^2.81).

# 7. Merge Sort

- Divide array into halves until single elements remain.
- Merge sorted halves step by step.
- Example:  $[5,2,8,4] \rightarrow$  split into [5,2] and  $[8,4] \rightarrow$  sort individually  $\rightarrow$  merge into  $[2,5,4,8] \rightarrow$  final [2,4,5,8].

- Time Complexity: O(n log n)
- Space Complexity: O(n)

#### 8. Quick Sort

- Pick a **pivot** element.
- Partition array into two parts:
  - Left  $\rightarrow$  elements < pivot
  - o Right → elements > pivot
- Recursively sort partitions.
- Example:  $[5,3,8,1] \rightarrow \text{pivot}=5 \rightarrow \text{partition } [3,1] \text{ and } [8] \rightarrow \text{sort} \rightarrow [1,3,5,8].$
- Time Complexity:
  - Best / Avg: O(n log n)
  - Worst: O(n²) (when pivot is always smallest/largest).
- Space: O(log n) (stack).

### Quick Revision for Viva:

- Recursion = function calls itself.
- Recurrence Relation = formula for time complexity.
- Methods: Substitution, Recursion Tree, Master Theorem.
- Binary Search = O(log n).
- Merge Sort = O(n log n).
- Quick Sort = O(n log n) avg, O(n<sup>2</sup>) worst.
- Strassen's Matrix = O(n^2.81).
- Karatsuba Multiplication = O(n^1.59).