

# Unit 1

## **Analysis of Non-Recursive Algorithms**

### **•Steps:**

- Identify the basic operations
- Count the number of basic operations
- Express the count as a function of input size

### **•Example:** Linear search

## **Understanding Non-Recursive Algorithms**

### **•Definition:**

- Algorithms that do not call themselves during their execution.

### **•Examples:**

- Iterative processes like looping through an array.
- Algorithms like linear search, bubble sort, insertion sort, etc.

## **Slide 6b: Steps to Analyze Non-Recursive Algorithms**

### **•Identify the Basic Operations:**

- Determine the most significant operation that contributes to the running time (e.g., comparisons, assignments).

### **•Count the Number of Basic Operations:**

- Establish how many times the basic operation is executed based on the input size.

### **•Express the Count as a Function of Input Size:**

- Create a mathematical expression that represents the total number of operations in terms of the input size  $n$ .

## **Example - Linear Search**

### **•Problem Statement:**

- Given an array of  $n$  elements, find a target element.

### **•Algorithm:**

- Start from the first element.
- Compare the target element with each element in the array.
- If the target is found, return its position.
- If the target is not found after checking all elements, return -1.

```
def linear_search(arr, target):  
    for i in range(len(arr)):  
        if arr[i] == target:  
            return i  
    return -1
```

## **Time Complexity Analysis of Linear Search**

- **Identify Basic Operation:**

- Comparison of elements ( $arr[i] == target$ )

- **Count Basic Operations:**

- In the worst case, all  $n$  elements are compared.

- **Express as a Function of Input Size:**

- Worst-case time complexity:  $T(n)=n$
- Therefore,  $T(n)$  is  $O(n)$ .

## **Analysis of Recursive Algorithms**

### **•Steps:**

- Define the recurrence relation
- Solve the recurrence relation
- Use the solution to express time complexity

### **•Example:** Binary search

## **Understanding Recursive Algorithms**

### **•Definition:**

- Algorithms that call themselves with a subset of the original problem.

### **•Examples:**

- Divide and conquer algorithms like merge sort, quicksort.
- Recursive algorithms like factorial calculation, Fibonacci sequence.

## Steps to Analyze Recursive Algorithms

- **Define the Recurrence Relation:**

- Establish a mathematical relation that describes the time complexity in terms of the input size.

- **Solve the Recurrence Relation:**

- Use methods such as substitution, recursion tree, or the Master theorem to solve the recurrence.

- **Express Time Complexity:**

- Derive the overall time complexity from the solution of the recurrence relation.

## Binary Search

- **Problem Statement:**

- Given a sorted array of  $n$  elements, find the target element.

- **Algorithm:**

- Compare the target with the middle element.
- If equal, return the position.
- If less, repeat the search on the left subarray.
- If more, repeat the search on the right subarray.

```
def binary_search(arr, low, high, target):
    if high >= low:
        mid = (high + low) // 2
        if arr[mid] == target:
            return mid
        elif arr[mid] > target:
            return binary_search(arr, low, mid - 1, target)
        else:
            return binary_search(arr, mid + 1, high, target)
    else:
        return -1
```

- Define the Recurrence Relation:
  - Each call reduces the problem size by half.
  - $T(n) = T(n/2) + O(1)$
- Solve the Recurrence Relation:
  - Using the Master theorem:  $T(n) = T(n/2) + O(1)$
  - Here,  $a = 1$ ,  $b = 2$ , and  $f(n) = O(1)$ .
  - $f(n) = O(n^c)$  where  $c = 0$ .
  - Since  $f(n)$  is  $O(n^{\log_b a}) = O(n^0) = O(1)$ , we use Case 2 of the Master theorem:  $T(n) = O(\log n)$ .
- Express Time Complexity:
  - Therefore,  $T(n) = O(\log n)$ .

## Understanding Amortized Analysis

### •Definition:

- Amortized analysis provides the average time per operation over a sequence of operations, ensuring that occasional expensive operations are accounted for over time.

### •Why Amortized Analysis?

- To give a more accurate overall performance metric than worst-case or average-case analysis for each operation.

## Types of Amortized Analysis

### •Aggregate Method:

- Calculate the total cost of  $n$  operations and divide by  $n$  to find the average cost per operation.

### •Accounting Method:

- Assign different costs to operations, charging more than the actual cost for cheaper operations to cover the cost of expensive ones.

### •Potential Method:

- Use a potential function to represent the pre-paid work stored in the data structure, which can be used to pay for future operations.



## Dynamic Array Resizing (Aggregate Method):

- Problem Statement:
  - Implement a dynamic array that resizes itself when full.
- Operations:
  1. Insertion
  2. Resizing (doubling the array size)
- Analysis:
  - Total Cost:
    - Insertions:  $n$  operations
    - Resizings:  $\log n$  resizing operations (since array size doubles each time)
  - Cost Breakdown:
    - Each insertion typically costs  $O(1)$ .
    - Each resizing operation costs  $O(n)$  but occurs less frequently.
  - Amortized Cost:
    - Total insertion cost:  $O(n)$
    - Total resizing cost:  $O(n)$  (since each element is copied once for each doubling)
    - Total cost:  $O(n) + O(n) = O(2n) = O(n)$
    - Amortized cost per operation:  $\frac{O(n)}{n} = O(1)$ .

## Writing Characteristic Polynomial Equations

### •Purpose:

- To solve linear recurrence relations.

### •Example:

- Recurrence relation:  $T(n) = 2T(n/2) + n$
- Characteristic equation:  $x^2 - 2x + 1 = 0$

## Solving Recurrence Equations

### •Definition:

- A recurrence equation defines a sequence where each term is a function of preceding terms.
- To find a closed-form expression for the sequence.

### •Why Solve Recurrence Equations?

- To analyze the performance of recursive algorithms.

## **Solving Recurrence Equations Methods:**

- Substitution method
- Master theorem
- Iteration method

### **Proof Techniques: By Contradiction**

#### **•Steps:**

- Assume the opposite of what you want to prove.
- Show that this assumption leads to a contradiction.

#### **•Example:**

- Proving the irrationality of  $\sqrt{2}$

### **Proof Techniques: By Mathematical Induction**

#### **•Steps:**

- Base case: Prove the statement for the initial value.
- Induction step: Assume the statement for  $k$ , prove for  $k+1$ .

#### **•Example:**

- Proving the formula for the sum of the first  $n$  natural numbers































