

# ASSIGNMENT 1

## Q2.3.1, Q2.3.2, Q2.3.3, Q2.4.1, Q2.4.2, Q2.4.3

### Section 2: Time and Space Complexity Analysis

#### 2.3 Time Complexity Analysis Questions

##### Problem 1

##### Question:

**For the given code snippet, find the time complexity in terms of Big O (O), Theta ( $\Theta$ ), and Omega ( $\Omega$ ). Provide a step-by-step explanation of how you arrived at the time complexity.**

##### Answer:

- Analyze the nested loops. The outer loop runs `n` times, and for each iteration of the outer loop, the inner loop also runs `n` times. This results in `n \* n = n^2` operations for adjusting `a`.
- The second loop (for `k`) is independent and runs `n` times for adjusting `b`.

##### Time Complexity:

- Big O (O):** The worst-case scenario is  $O(n^2)$  due to the nested loops.
- Theta ( $\Theta$ ):** Since the dominant term in the total operations is  $n^2$  and there are no scenarios where the algorithm would perform better or worse than this rate of growth, the average-case complexity is also  $\Theta(n^2)$ .
- Omega ( $\Omega$ ):** The best-case scenario, which is also  $\Omega(n^2)$ , aligns with the worst-case since the loops will always perform  $n^2$  operations for adjusting “a plus n” operations for adjusting ‘b’, but  $n^2$  is the dominant term.

## Problem 2

### Question:

**Determine the time complexity for the following function:**

### Answer:

#### - Analysis:

- The first loop runs approximately  $n/2$  times.
- The second loop is a geometric progression starting from 1 and multiplying by 2 each time until  $n$ . This runs in  $O(\log n)$  time.
- The third loop, identical to the second loop, also runs in  $O(\log n)$  time.

#### - Overall Time Complexity:

- **Big O (O):** Considering the combination of loops, the worst-case time complexity is  $O(n \log^2 n)$  because the outer loop runs  $n/2$  times, and each of the two inner loops runs  $\log n$  times.

- **Theta ( $\Theta$ ):** The average-case complexity, considering the nature of loops and their execution path, remains  $\Theta(n \log^2 n)$ .

- **Omega ( $\Omega$ ):** The best-case scenario, given the lower bounds of the loops, is  $\Omega(n \log^2 n)$ .

## Problem 3

### Question:

**Evaluate the time complexity for this recursive function:**

### Answer:

#### - Analysis:

- This function makes two recursive calls for each value of 'a' greater than 0, effectively doubling the number of calls with each decrement of 'a'.

#### -Overall Time Complexity:

- **Big O (O):** The time complexity is  $O(2^n)$  because each function call spawns two additional calls.

- **Theta ( $\Theta$ ):** The average-case complexity mirrors the worst-case, so it is also  $\Theta(2^n)$ .

- **Omega ( $\Omega$ ):** The best-case scenario, where  $a \leq 0$ , executes in constant time,  $\Omega(1)$ , but considering the growth rate as 'a' increases, the lower bound in terms of growth behavior is  $\Omega(2^n)$  for  $a > 0$ .

## 2.4 Space Complexity Analysis Questions

### Problem 1

#### Question:

For the given code snippet, analyze the space complexity in terms of Big O (O), Theta ( $\Theta$ ), and Omega ( $\Omega$ ). Present a step-by-step explanation of your analysis.

#### - Analysis:

- This function performs an in-place reversal of an array. The variables `start`, `end`, and `temp` are used for index tracking and swapping elements.

#### - Space Complexity:

- **Big O (O):** The space complexity is  $O(1)$  as the memory usage does not scale with the size of the input array. Only a constant amount of extra space (`start`, `end`, `temp`) is used regardless of the input size.

- **Theta ( $\Theta$ ):** Since the algorithm consistently uses a fixed amount of space, the average-case space complexity is also  $\Theta(1)$ .

- **Omega ( $\Omega$ ):** The best-case scenario also has a space complexity of  $\Omega(1)$ , reflecting that even in the most efficient case, the algorithm requires a constant amount of space.

### Problem 2

#### Question:

Analyze the space complexity of the following function:

#### - Analysis:

- This function calculates the nth Fibonacci number using dynamic programming. The `dp` array of size  $n + 1$  stores intermediate Fibonacci values.

#### -Space Complexity:

- **Big O (O):** The space complexity is  $O(n)$  as the algorithm allocates an array of size  $n + 1$  to store the Fibonacci sequence up to  $n$ .

- **Theta ( $\Theta$ ):** Given the allocation of memory scales linearly with  $n$ , the average-case space complexity is  $\Theta(n)$ .

- **Omega ( $\Omega$ ):** The best-case space complexity is also  $\Omega(n)$ , indicating that the minimum space required by the algorithm grows linearly with the input size.

### Problem 3

#### Question:

Evaluate the space complexity for this function:

#### -Analysis:

- This function computes the length of the longest common subsequence between two strings `s1` and `s2` using dynamic programming. The `dp` 2D array of size  $(m + 1) \times (n + 1)$  is used to store the lengths of the longest common subsequences for all subproblems.

#### - Space Complexity:

- **Big O (O):** The space complexity is  $O(mn)$  due to the allocation of a 2D array whose size is directly proportional to the lengths of the input strings `s1` and `s2`.

- **Theta ( $\Theta$ ):** The average-case space complexity is  $\Theta(mn)$  as the space used by the algorithm scales with the product of the lengths of the two input strings.

- **Omega ( $\Omega$ ):** The best-case scenario still requires a space complexity of  $\Omega(mn)$  for the dynamic programming table.