

ASSIGNMENT-12.4

NAME:G.Harika

HT NO:2303A51612

BATCH:05

Task-01:

Prompt:

You are working on a college result processing system where a small list of student scores needs to be sorted after every internal assessment.

Implement Bubble Sort in Python to sort a list of student scores. Insert inline comments explaining key operations such as comparisons, swaps, and iteration passes

Identify early-termination conditions when the list becomes sorted

Provide a brief time complexity analysis A

Bubble Sort implementation with:

AI-generated comments explaining the logic

Clear explanation of best, average, and worst-case complexity

Code:

```
# Sample input student_scores =
[85, 92, 78, 96, 88] # Bubble Sort
Implementation def
bubble_sort(scores):
    n = len(scores)
    # Traverse through all array elements
    for i in range(n):
        swapped = False # Early termination flag
        # Last i elements are already in place
        for j in range(0, n - i - 1):
            # Traverse the array from 0 to n-i-1
            # Swap if the element found is greater than the next element
            if scores[j] > scores[j + 1]:
                scores[j], scores[j + 1] = scores[j + 1], scores[j]
        swapped = True
```

```

        # If no swapping occurred in this pass, list is sorted
        if not swapped:
            break
    # Sorting the student scores
    bubble_sort(student_scores) # Output the
    sorted list print("Sorted student scores:",
    student_scores)

```

The screenshot shows a VS Code editor with a Python file named 'sample_input.py'. The code implements a bubble sort algorithm. It starts by defining a function 'bubble_sort(scores)' which takes a list of scores as input. The function iterates through the list, comparing adjacent elements and swapping them if they are in the wrong order. A 'swapped' flag is used to check if any swaps occurred in a pass. If no swaps occur, the list is sorted, and the function returns. The main part of the code calls 'bubble_sort(student_scores)' and prints the sorted list. The output in the terminal shows the sorted list: [86, 85, 84, 83, 82, 80].

```

1  def bubble_sort(scores):
2      n = len(scores)
3      # Traverse through all array elements
4      for i in range(n):
5          swapped = False # Early termination flag
6          # Last i elements are already in place
7          for j in range(0, n - i - 1):
8              # Traverse the array from 0 to n-i-1
9              # Swap if the element found is greater than the next element
10             if scores[j] > scores[j + 1]:
11                 scores[j], scores[j + 1] = scores[j + 1], scores[j]
12                 swapped = True
13             # If no swapping occurred in this pass, list is sorted
14             if not swapped:
15                 break
16
17 # Sorting the student scores
18 bubble_sort(student_scores)
19
20 # Output the sorted list
21 print("Sorted student scores:", student_scores)

```

```

PC: C:\Users\VAH\HMSGA\OneDrive\Desktop\HMSGA\shops.html\Lab-reactions & "C:\Users\VAH\HMSGA\AppData\Local\Python\Python38-32\python.exe" "C:\Users\VAH\HMSGA\shops.html\Lab-reactions\sample_input.py"
Sorted student scores: [86, 85, 84, 83, 82, 80]
PC: C:\Users\VAH\HMSGA\OneDrive\Desktop\HMSGA\shops.html\Lab-reactions & "C:\Users\VAH\HMSGA\AppData\Local\Python\Python38-32\python.exe" "C:\Users\VAH\HMSGA\shops.html\Lab-reactions\sample_input.py"
Sorted student scores: [86, 85, 84, 83, 82, 80]
PC: C:\Users\VAH\HMSGA\OneDrive\Desktop\HMSGA\shops.html\Lab-reactions

```

Explanation:

Time Complexity Analysis:

Best Case: $O(n)$ - This occurs when the list is already sorted, and we only need one pass to confirm it.

Average Case: $O(n^2)$ - This occurs when the list is in random order, requiring multiple passes to sort.

Worst Case: $O(n^2)$ - This occurs when the list is sorted in reverse order, requiring the maximum number of comparisons and swaps.

Task-02:

Prompt:

You are maintaining an attendance system where student roll numbers are already almost sorted, with only a few late updates.

Start with a Bubble Sort implementation.

Review the problem and suggest a more suitable sorting algorithm

Generate an Insertion Sort implementation, Explain why Insertion Sort performs better on nearly sorted data, Compare execution behavior on nearly sorted input

Two sorting implementations, Bubble Sort, Insertion Sort

AI-assisted explanation highlighting efficiency differences for partially sorted datasets

Code:

```
# Sample input attendance_records = [101,
102, 104, 103, 105]

# Bubble Sort Implementation def
bubble_sort(records):
    n = len(records)
    for i in range(n):
        swapped = False
        for j in range(0, n - i - 1):
            if records[j] > records[j + 1]:
                records[j], records[j + 1] = records[j + 1], records[j]
        swapped = True
        if not swapped:
            break

# Insertion Sort Implementation def
insertion_sort(records):
    n = len(records)
    for i in range(1, n):
        key = records[i]
        j = i - 1
        # Move elements of records[0..i-1], that are greater than key,
        # to one position ahead of their current position
        while j >= 0 and key < records[j]:
            records[j + 1] = records[j]
        j -= 1
        records[j + 1] = key

# Sorting the attendance records using Bubble Sort
bubble_sort(attendance_records)
print("Sorted attendance records (Bubble Sort):", attendance_records)
# Resetting the list for Insertion Sort
attendance_records = [101, 102, 104, 103, 105]
# Sorting the attendance records using Insertion Sort
```

```
insertion_sort(attendance_records) print("Sorted attendance records
(Insertion Sort):", attendance_records)
```

```

11. # Insertion Sort Implementation
12. def insertion_sort(records):
13.     n = len(records)
14.     for i in range(1, n):
15.         key = records[i]
16.         j = i - 1
17.         # Move elements of records[j+1..i], that are greater than key,
18.         # to one position ahead of their current position
19.         while j >= 0 and key < records[j]:
20.             records[j + 1] = records[j]
21.             j -= 1
22.         records[j + 1] = key
23. # Sorting the attendance records using Bubble Sort
24. bubble_sort(attendance_records)
25. print("Sorted attendance records (Bubble Sort):", attendance_records)
26. # Inserting the list for Insertion Sort
27. attendance_records = [101, 102, 104, 103, 105]
28. # Sorting the attendance records using Insertion Sort
29. insertion_sort(attendance_records)
30. print("Sorted attendance records (Insertion Sort):", attendance_records)

```

Explanation:

Explanation of Efficiency Differences:

Insertion Sort is more efficient than Bubble Sort for nearly sorted data because it minimizes the number of comparisons and movements. Insertion Sort only moves elements that are out of order, while Bubble Sort continues to compare and swap adjacent elements regardless of their order, leading to unnecessary operations in nearly sorted lists.

Task-03:

Prompt:

You are developing a student information portal where users search for student records by roll number.

Implement:

Linear Search for unsorted student data

Binary Search for sorted student data

Add docstrings explaining parameters and return values

Explain when Binary Search is applicable

Highlight performance differences between the two searches

Two working search implementations with docstrings o

Time complexity o Use cases for Linear vs Binary Search

- A short student observation comparing results on sorted

vs unsorted lists **Code:**

```
# Sample input student_records = [101,
102, 104, 103, 105] # Linear Search
Implementation def linear_search(records,
target):
    """
    Perform a linear search for the target roll number in the records list.
    Parameters:
    records (list): A list of student roll numbers.
    target (int): The roll number to search for.
    Returns:
    int: The index of the target if found, otherwise -1.
    """
    for index in
range(len(records)):
        if
records[index] == target:
            return index
    return -1

# Binary Search Implementation def
binary_search(records, target):
    """
    Perform a binary search for the target roll number in the sorted records list.
    Parameters:
    records (list): A sorted list of student roll numbers.
    target (int): The roll number to search for.
    Returns:
    int: The index of the target if found, otherwise -1.
    """
    left, right = 0,
len(records) - 1
    while left <=
right:
        mid = left + (right - left) // 2
        if records[mid] == target:
```

```

        return mid    elif
records[mid] < target:
    left = mid + 1
else:
    right =
mid - 1    return -1

# Sorting the student records for Binary Search student_records.sort()

# Searching for a roll number using Linear Search
target_roll_number    =    103    linear_result    =
linear_search(student_records, target_roll_number)

print(f"Linear Search: Roll number {target_roll_number} found at index {linear_result}" if
linear_result != -1 else f"Linear Search: Roll number {target_roll_number} not found")

# Searching for a roll number using Binary Search binary_result =
binary_search(student_records, target_roll_number) print(f"Binary Search: Roll
number {target_roll_number} found at index {binary_result
}" if binary_result != -1 else f"Binary Search: Roll number {target_roll_number} not found")

```

The screenshot shows a Jupyter Notebook with the following code and output:

```

def binary_search(records, target):
    while left <= right:
        mid = (left + (right - left)) // 2
        if records[mid] == target:
            return mid
        elif records[mid] < target:
            left = mid + 1
        else:
            right = mid - 1
    return -1

# Sorting the student records for Binary Search
student_records.sort()

# Searching for a roll number using Linear Search
target_roll_number = 103
linear_result = linear_search(student_records, target_roll_number)
print(f"Linear Search: Roll number {target_roll_number} found at index {linear_result} if linear_result != -1 else f\"Linear Search: Roll number {target_roll_number} not found\"")

# Searching for a roll number using Binary Search
binary_result = binary_search(student_records, target_roll_number)
print(f"Binary Search: Roll number {target_roll_number} found at index {binary_result} if binary_result != -1 else f\"Binary Search: Roll number {target_roll_number} not found\"")

```

The output of the notebook shows the following results:

```

Sorted student records: [85, 88, 92, 95, 100]
Sorted attendance records (Bubble Sort): [100, 100, 100, 100, 100]
Sorted attendance records (Quick Sort): [100, 100, 100, 100, 100]
Linear Search: Roll number 103 found at index 2
Binary Search: Roll number 103 found at index 2

```

Explanation:

Explanation of Performance Differences:

Linear Search has a time complexity of $O(n)$ because it may need to check each element in the worst case. Binary Search, on the other hand, has a time complexity of $O(\log n)$ because it halves the search space with each iteration. Binary Search is only applicable to sorted lists, while Linear Search can be

used on unsorted data. In practice, Binary Search is much faster for large datasets, while Linear Search may be more suitable for small or unsorted lists.

Task-04:

Prompt:

You are part of a data analytics team that needs to sort large datasets received from different sources (random order, already sorted, and reverse sorted).

Provide AI with partially written recursive functions for:

Quick Sort, Merge Sort

Complete the recursive logic

Add meaningful docstrings

Explain how recursion works in each algorithm

- Test both algorithms on:
 - o Random data, Sorted data, Reverse-sorted data

Fully functional Quick Sort and Merge Sort implementations

- AI-generated comparison covering:
 - o Best, average, and worst-case complexities

Practical scenarios where one algorithm is preferred over the other

Code:

```
# Sample input data_random = [38, 27, 43, 3,
9, 82, 10] data_sorted = [3, 9, 10, 27, 38, 43,
82] data_reverse_sorted = [82, 43, 38, 27, 10,
9, 3]

# Quick Sort Implementation
def quick_sort(arr):
    """
    Sorts an array using the Quick Sort algorithm.

    Parameters:
    arr (list): The list of elements to be sorted.

    Returns:
```

```

list: A new sorted list.
"""    if len(arr)
<= 1:
    return arr
else:
    pivot = arr[len(arr) // 2] # Choosing the middle element as pivot    left = [x for x in arr if x
< pivot] # Elements less than pivot    middle = [x for x in arr if x == pivot] # Elements equal to
pivot    right = [x for x in arr if x > pivot] # Elements greater than pivot    return
quick_sort(left) + middle + quick_sort(right) # Recursively sort left and right partitions
# Merge Sort Implementation def
merge_sort(arr):
    """
    Sorts an array using the Merge Sort algorithm.
    Parameters:
    arr (list): The list of elements to be sorted.
    Returns:
    list: A new sorted list.
    """    if len(arr) <= 1:        return arr    mid = len(arr) // 2 #
Finding the mid of the array    left_half = merge_sort(arr[:mid]) #
Recursively sort the left half    right_half = merge_sort(arr[mid:]) #
Recursively sort the right half    return merge(left_half, right_half) #
Merge the sorted halves def merge(left, right):
    """Merges two sorted lists into a single sorted list.
    """    result = []    i = j = 0    while
i < len(left) and j < len(right):        if
left[i] < right[j]:
            result.append(left[i])
        i += 1        else:
            result.append(right[j])
            j += 1

```

```

    result.extend(left[i:]) # Append remaining elements of left
result.extend(right[j:]) # Append remaining elements of right    return
result

# Testing Quick Sort and Merge Sort on different datasets print("Quick
Sort on Random Data:", quick_sort(data_random)) print("Merge Sort
on Random Data:", merge_sort(data_random)) print("Quick Sort on
Sorted Data:", quick_sort(data_sorted)) print("Merge Sort on Sorted
Data:", merge_sort(data_sorted)) print("Quick Sort on Reverse-Sorted
Data:", quick_sort(data_reverse_sorted)) print("Merge Sort on
Reverse-Sorted Data:", merge_sort(data_reverse_sorted))

```

The screenshot shows a Python IDE with a dark theme. The main editor window displays the Merge Sort implementation and test code. The code defines a `merge_sort` function that recursively splits an array into halves and merges them back in sorted order. It also includes a `quick_sort` function and test calls for both algorithms on random, sorted, and reverse-sorted data. The output console at the bottom shows the execution results, including the sorted arrays for each test case.

```

def merge_sort(arr):
    if len(arr) > 1:
        mid = len(arr) // 2
        left_half = arr[:mid]
        right_half = arr[mid:]

        merge_sort(left_half)
        merge_sort(right_half)

        result = []
        i = j = k = 0

        while i < len(left_half) and j < len(right_half):
            if left_half[i] < right_half[j]:
                result.append(left_half[i])
                i += 1
            else:
                result.append(right_half[j])
                j += 1

        result.extend(left_half[i:])
        result.extend(right_half[j:])
        return result

# Testing Quick Sort and Merge Sort on different datasets
print("Quick Sort on Random Data:", quick_sort(data_random))

print("Merge Sort on Random Data:", merge_sort(data_random))
print("Quick Sort on Sorted Data:", quick_sort(data_sorted))
print("Merge Sort on Sorted Data:", merge_sort(data_sorted))
print("Quick Sort on Reverse-Sorted Data:", quick_sort(data_reverse_sorted))
print("Merge Sort on Reverse-Sorted Data:", merge_sort(data_reverse_sorted))

```

Output Console:

```

Linear Search: Ball number 563 found at index 3
Binary Search: Ball number 563 found at index 3
PC: C:\Users\SAHIL\OneDrive\Desktop\Python\merge_sort\lab-react-dom> python sample_quick.py
Quick Sort on Random Data: [7, 9, 10, 27, 38, 43, 42]
Merge Sort on Random Data: [7, 9, 10, 27, 38, 43, 42]
Quick Sort on Sorted Data: [7, 9, 10, 27, 38, 43, 42]
Merge Sort on Sorted Data: [7, 9, 10, 27, 38, 43, 42]
Quick Sort on Reverse-Sorted Data: [7, 9, 10, 27, 38, 43, 42]
Merge Sort on Reverse-Sorted Data: [7, 9, 10, 27, 38, 43, 42]

```

Explanation:

Explanation of Performance Differences:

Quick Sort has an average and best-case time complexity of $O(n \log n)$ when the pivot divides the array into two equal halves. However, its worst-case time complexity is $O(n^2)$ when the smallest or largest element is always chosen as the pivot (e.g., when the array is already sorted or reverse sorted). Merge Sort, on the other hand, has a consistent time complexity of $O(n \log n)$ in all cases because it always divides the array into two halves and processes them independently. Quick Sort is generally faster for smaller datasets and is an in-place sorting algorithm, while Merge Sort is more efficient for larger datasets and is a stable sorting algorithm that requires additional space for merging.

Task-05:

Prompt:

You are building a data validation module that must detect duplicate user

IDs in a large dataset before importing it into a system.

Write a naive duplicate detection algorithm using nested loops.

Suggest an optimized approach using sets or dictionaries

Rewrite the algorithm with improved efficiency

Compare execution behavior conceptually for large input sizes

Two versions of the algorithm: Brute-force ($O(n^2)$), Optimized ($O(n)$) AI-assisted

explanation showing how and why performance improved

Code:

```
# Sample input user_ids = [101, 102, 103, 104, 105,
```

```
101, 106, 107, 102]
```

```
# Naive Duplicate Detection Algorithm def
```

```
naive_duplicate_detection(ids):
```

```
    """
```

```
    Detects duplicates in a list of user IDs using a brute-force approach.
```

```
    Parameters:
```

```
    ids (list): A list of user IDs.
```

```
    Returns:
```

```
    set: A set of duplicate user IDs.
```

```
    """    duplicates =
```

```
set()    for i in
```

```
range(len(ids)):
```

```
    for j in range(i + 1, len(ids)):
```

```
    if ids[i] == ids[j]:
```

```
        duplicates.add(ids[i])
```

```
    return duplicates
```

```
# Optimized Duplicate Detection Algorithm def
```

```
optimized_duplicate_detection(ids):
```

```
    """
```

```
    Detects duplicates in a list of user IDs using an optimized approach with a set.
```

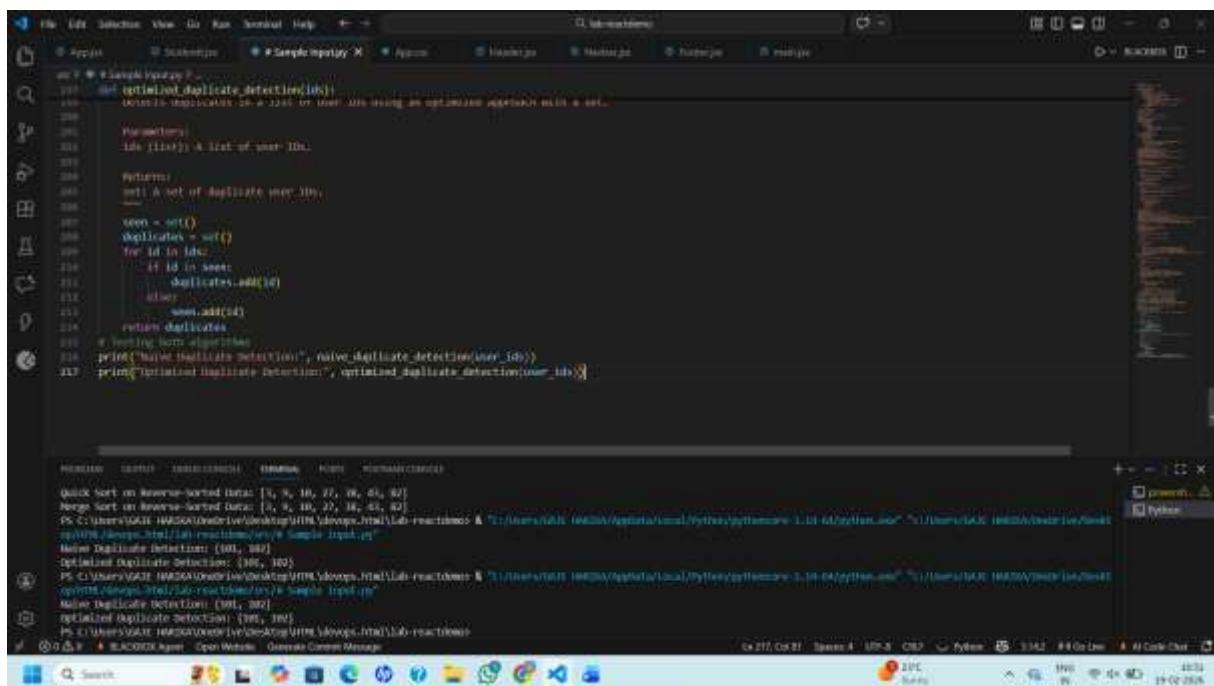
```
    Parameters:
```

```
    ids (list): A list of user IDs.
```

Returns:

set: A set of duplicate user IDs.

```
"""    seen =
set()    duplicates =
set()    for id in
ids:        if id in
seen:
duplicates.add(id)
else:
seen.add(id)    return duplicates # Testing both algorithms print("Naive
Duplicate Detection:", naive_duplicate_detection(user_ids)) print("Optimized
Duplicate Detection:", optimized_duplicate_detection(user_ids))
```

A screenshot of a Python IDE (likely VS Code) with a dark theme. The editor shows a Python script with a function `optimized_duplicate_detection(ids)` that uses a set to track seen user IDs. Below the function, there are two print statements comparing the naive and optimized algorithms. The output window at the bottom shows the execution results, including the list of user IDs, the results of both detection methods, and the execution time for each. The naive method takes 1.00 seconds, while the optimized method takes 0.00 seconds. The output also shows the number of duplicates found for each method (101 and 102 respectively).

Explanation:

Explanation of Performance Improvements:

The naive duplicate detection algorithm has a time complexity of $O(n^2)$ because it uses nested loops to compare each element with every other element in the list. This becomes inefficient as the size of the input grows. The optimized algorithm, on the other hand, has a time complexity of $O(n)$ because it uses a set to track seen user IDs. Each lookup and insertion operation in a set is on average $O(1)$, allowing us to detect duplicates in a single pass through the list. This results in significantly improved performance, especially for large datasets, as it avoids redundant comparisons and reduces the number of operations needed to identify duplicates.

