

## AI ASS 12.4

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Batch:43

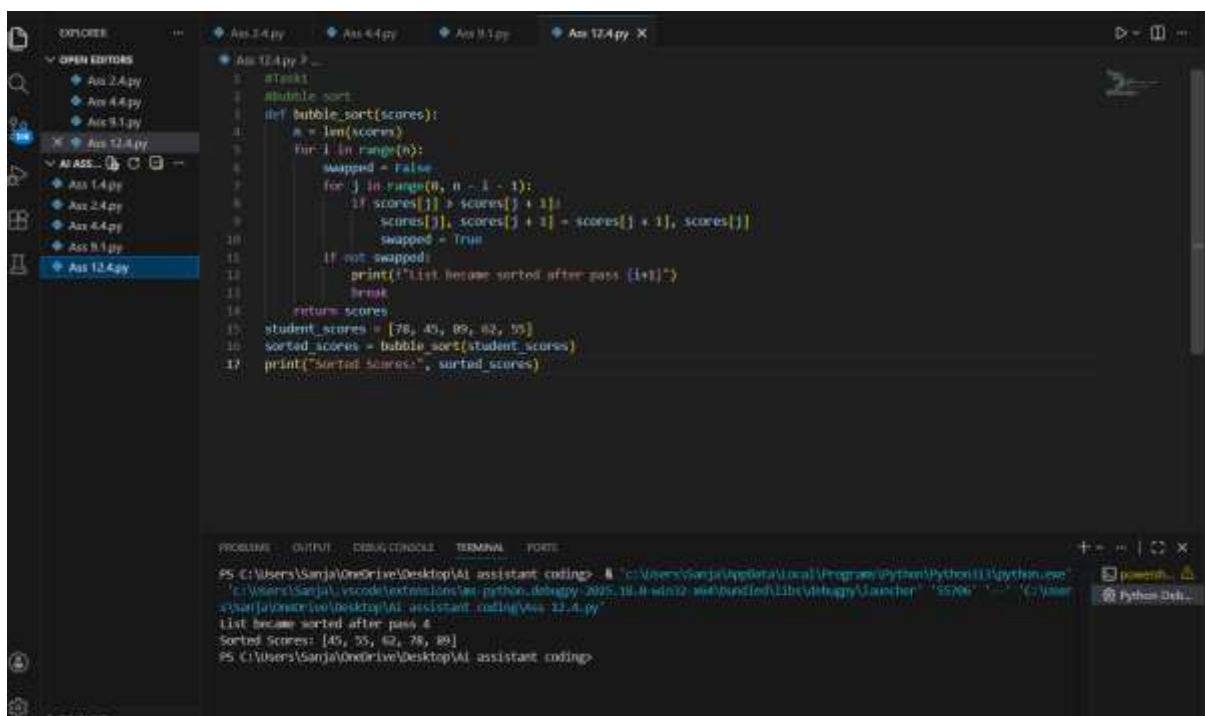
### Task1:

Write a Python program to implement Bubble Sort for sorting student exam scores.

Add inline comments explaining comparisons, swaps, and passes.

Include early termination using a swapped flag.

Provide sample input/output and briefly explain best, average, and worst-case time complexity.



```
1 #Task1
2 #Bubble sort
3 def bubble_sort(scores):
4     n = len(scores)
5     for i in range(n):
6         swapped = False
7         for j in range(0, n - i - 1):
8             if scores[j] > scores[j + 1]:
9                 scores[j], scores[j + 1] = scores[j + 1], scores[j]
10                swapped = True
11            if not swapped:
12                print(f'List became sorted after pass {i+1}')
13                break
14        return scores
15 student_scores = [78, 45, 89, 82, 55]
16 sorted_scores = bubble_sort(student_scores)
17 print("Sorted Scores:", sorted_scores)
```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

```
PS C:\Users\Sarja\OneDrive\Desktop\AI assistant coding> "c:\Users\Sarja\AppData\Local\Programs\Python\Python113\python.exe"
c:\Users\Sarja\vscode\extensions\ms-python.debugpy-2025.10.0-wn32-wsl-bundled\Lib\debugpy\launcher 55706 -c "c:\Users\Sarja\OneDrive\Desktop\AI assistant coding\Ass 12.4.py"
List became sorted after pass 4
Sorted Scores: [45, 55, 82, 78, 89]
PS C:\Users\Sarja\OneDrive\Desktop\AI assistant coding>
```

### Observation:

Bubble Sort was implemented to sort student scores after internal assessments.

The algorithm shows  $O(n)$  time complexity in the best case when the list is already sorted due to early termination.

In the average case, when scores are randomly arranged, the time complexity is  $O(n^2)$ .

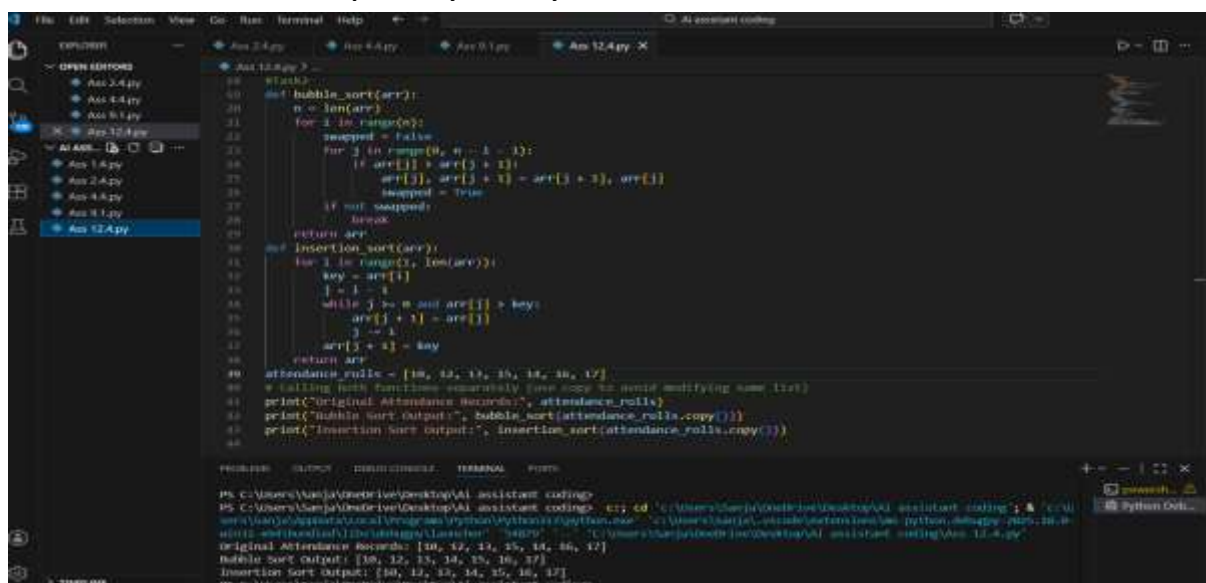
In the worst case, when scores are in reverse order, it also takes  $O(n^2)$  time due to maximum comparisons and swaps.

Thus, Bubble Sort is suitable for small datasets but not efficient for large datasets.

## Task2:

I have a nearly sorted list of student roll numbers (attendance records).

1. Implement Bubble Sort in Python.
2. Suggest a more suitable sorting algorithm for nearly sorted data.
3. Implement Insertion Sort in Python.
4. Explain why Insertion Sort performs better on nearly sorted datasets.
5. Compare their behavior on a nearly sorted input example.
6. Include time complexity analysis.



```
def bubble_sort(arr):
    n = len(arr)
    for i in range(n):
        swapped = False
        for j in range(0, n - i - 1):
            if arr[j] > arr[j + 1]:
                arr[j], arr[j + 1] = arr[j + 1], arr[j]
                swapped = True
        if not swapped:
            break
    return arr

def insertion_sort(arr):
    for i in range(1, len(arr)):
        key = arr[i]
        j = i - 1
        while j > 0 and arr[j] > key:
            arr[j + 1] = arr[j]
            j -= 1
        arr[j + 1] = key
    return arr

attendance_roll = [10, 12, 13, 15, 14, 16, 17]
# Calling both functions separately (we copy to avoid modifying same list)
print("Original Attendance Records:", attendance_roll)
print("Bubble Sort Output:", bubble_sort(attendance_roll.copy()))
print("Insertion Sort Output:", insertion_sort(attendance_roll.copy()))
```

```
PS C:\Users\Ajay\Documents\Desktop\AI assistant coding>
PS C:\Users\Ajay\Documents\Desktop\AI assistant coding> cd "C:\Users\Ajay\Documents\Desktop\AI assistant coding" & python .\Ass 11.4.py
original Attendance Records: [10, 12, 13, 15, 14, 16, 17]
Bubble Sort Output: [10, 12, 13, 14, 15, 16, 17]
Insertion Sort Output: [10, 12, 13, 14, 15, 16, 17]
PS C:\Users\Ajay\Documents\Desktop\AI assistant coding>
```

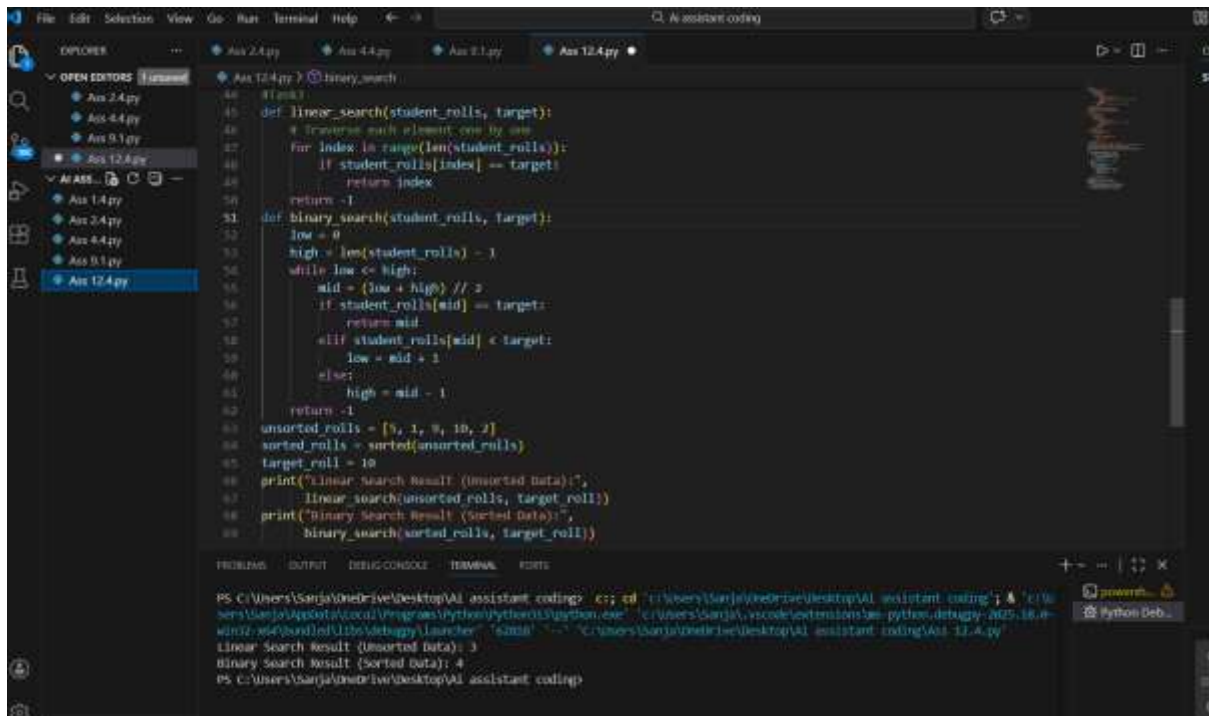
### **Observation:**

- 1.The attendance roll numbers were nearly sorted with only a few misplaced elements.
- 2.Bubble Sort performed multiple comparisons even when only small corrections were needed.
- 3.Insertion Sort shifted only the misplaced elements to their correct positions.
- 4.For nearly sorted data, Insertion Sort behaved close to  **$O(n)$**  time complexity.
5. Bubble Sort still required more passes, making it comparatively slower.
- 6.Therefore, Insertion Sort is more efficient and suitable for partially sorted attendance records.

### **Task3:**

Building a student information portal.

1. Implement Linear Search for unsorted student roll numbers.
2. Implement Binary Search for sorted roll numbers.
3. Add proper Python docstrings explaining parameters and return values.
4. Explain when Binary Search can be used.
5. Compare time complexity and performance differences.
6. Provide sample input and output.
7. Add a short observation comparing results on sorted vs unsorted data.



## Observation:

**Linear Search** checks each element one by one, so its time complexity is  **$O(n)$**  in average and worst cases. It works on both sorted and unsorted lists.

**Binary Search** divides the sorted list into halves repeatedly, so its time complexity is  **$O(\log n)$** . It works only on sorted data.

For unsorted lists, Linear Search is used.

For large sorted lists, Binary Search is faster and more efficient.

## Task4:

work with large datasets that may be random, already sorted, or reverse sorted.

1. Complete recursive implementations of Quick Sort and Merge Sort in Python.
2. Add proper docstrings explaining parameters and return values.

3. Explain how recursion works in both algorithms.
4. Test both algorithms on:
  - Random data
  - Sorted data
  - Reverse sorted data
5. Compare best, average, and worst-case time complexity.
6. Explain practical scenarios where one is preferred over the other.

```

def quick_sort(arr):
    if len(arr) <= 1:
        return arr
    pivot = arr[-1]
    left = [x for x in arr[:-1] if x <= pivot]
    right = [x for x in arr[:-1] if x > pivot]
    return quick_sort(left) + [pivot] + quick_sort(right)

def merge_sort(arr):
    if len(arr) <= 1:
        return arr
    mid = len(arr) // 2
    left_half = merge_sort(arr[:mid])
    right_half = merge_sort(arr[mid:])
    return merge(left_half, right_half)

def merge(left, right):
    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:])
    result.extend(right[j:])
    return result

# Test the algorithms
random_data = [45, 12, 78, 34, 23, 89, 1]
sorted_data = sorted(random_data)
reverse_data = sorted(random_data, reverse=True)

print("Quick Sort - Random:", quick_sort(random_data))
print("Quick Sort - Sorted:", quick_sort(sorted_data))
print("Quick Sort - Reverse:", quick_sort(reverse_data))

print("Merge Sort - Random:", merge_sort(random_data))
print("Merge Sort - Sorted:", merge_sort(sorted_data))
print("Merge Sort - Reverse:", merge_sort(reverse_data))

```

```

def merge(left, right):
    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:])
    result.extend(right[j:])
    return result

# Test the algorithms
random_data = [45, 12, 78, 34, 23, 89, 1]
sorted_data = sorted(random_data)
reverse_data = sorted(random_data, reverse=True)

print("Quick Sort - Random:", quick_sort(random_data))
print("Quick Sort - Sorted:", quick_sort(sorted_data))
print("Quick Sort - Reverse:", quick_sort(reverse_data))

print("Merge Sort - Random:", merge_sort(random_data))
print("Merge Sort - Sorted:", merge_sort(sorted_data))
print("Merge Sort - Reverse:", merge_sort(reverse_data))

```

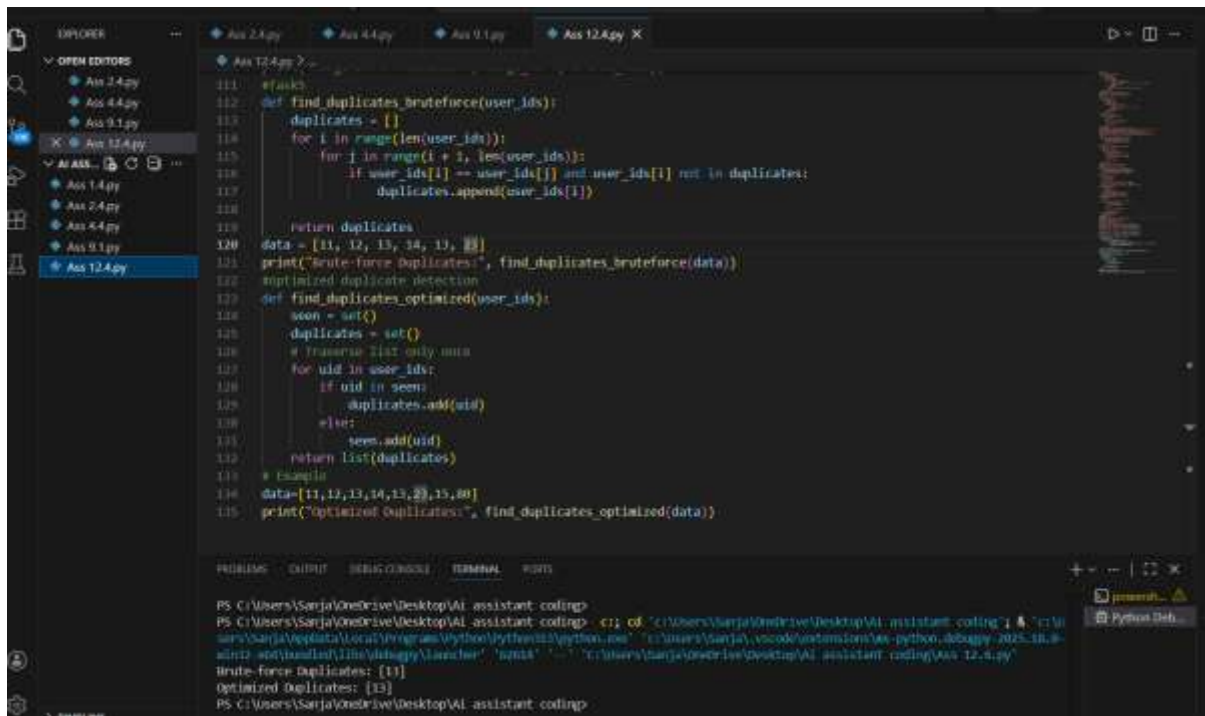
## **Observation:**

1. Quick Sort has best and average time complexity of  $O(n \log n)$ .
2. Quick Sort worst case is  $O(n^2)$  when pivot selection is poor.
3. Merge Sort has  $O(n \log n)$  time complexity in best, average, and worst cases.
4. Quick Sort uses less memory (in-place), while Merge Sort requires extra memory.
5. Quick Sort is usually faster for random datasets.
6. Merge Sort is preferred when guaranteed performance and stable sorting are required.

## **Task5:**

Building a data validation module to detect duplicate user IDs.

1. Write a brute-force duplicate detection algorithm using nested loops.
2. Analyze its time complexity.
3. Suggest an optimized approach using a set or dictionary.
4. Rewrite the algorithm with improved efficiency.
5. Compare performance conceptually for large datasets.
6. Keep explanation simple and clear.



```
111 def find_duplicates_bruteforce(user_ids):
112     duplicates = []
113     for i in range(len(user_ids)):
114         for j in range(i + 1, len(user_ids)):
115             if user_ids[i] == user_ids[j] and user_ids[i] not in duplicates:
116                 duplicates.append(user_ids[i])
117     return duplicates
118
119 data = [11, 12, 13, 14, 13, 13]
120 print("Brute-force Duplicates:", find_duplicates_bruteforce(data))
121
122 # Optimized duplicate detection
123 def find_duplicates_optimized(user_ids):
124     seen = set()
125     duplicates = set()
126     # Traverse list only once
127     for uid in user_ids:
128         if uid in seen:
129             duplicates.add(uid)
130         else:
131             seen.add(uid)
132     return list(duplicates)
133
134 # Example
135 data = [11, 12, 13, 14, 13, 13, 15, 15]
136 print("Optimized Duplicates:", find_duplicates_optimized(data))
```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL RUNS

```
PS C:\Users\Sarja\OneDrive\Desktop\AI assistant coding>
PS C:\Users\Sarja\OneDrive\Desktop\AI assistant coding> cd "c:\Users\Sarja\OneDrive\Desktop\AI assistant coding" & "c:\Users\Sarja\AppData\Local\Programs\Python\Python311\python.exe" "c:\Users\Sarja\vscode\extensions\ms-python.debugpy-2025.18.0-win32-x64\bin\debugpy_launcher" "npml" "..." "c:\Users\Sarja\OneDrive\Desktop\AI assistant coding\Ass 12.4.py"
Brute-force Duplicates: [13]
Optimized Duplicates: [13]
PS C:\Users\Sarja\OneDrive\Desktop\AI assistant coding>
```

## Observation:

- 1.The brute-force algorithm uses nested loops to compare every element with all others, resulting in  $O(n^2)$  time complexity.
- 2.It performs a large number of unnecessary comparisons, especially for large datasets.
- 3.The optimized algorithm uses a set (or dictionary) to store already seen elements.
- 4.Since set lookup takes  $O(1)$  time, the entire list is processed in a single loop, giving  $O(n)$  time complexity.
- 5.The performance improved because repeated comparisons were replaced with fast lookups.
- 6.Therefore, the optimized approach is much faster and more suitable for large datasets.