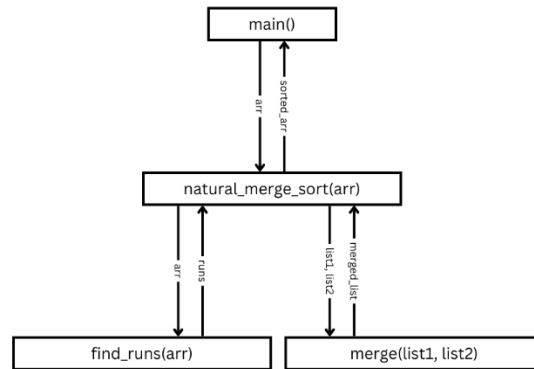


1. Modularization Metrics



`natural_merge_sort()`

Cohesion: High - This function focuses solely on sorting the list by repeatedly merging naturally ordered runs.

Coupling: Moderate - This function depends on `find_runs()` and `merge()` to perform its operations. Although there is dependency, it does not modify or interfere with the internal workings of the other functions.

`find_runs()`

Cohesion: High - This function's sole responsibility is to identify and separate the list into runs, so it has high cohesion.

Coupling: Low - This function is self-contained and does not rely on any other function to achieve its task.

`merge()`

Cohesion: High - The function only merges two sorted lists, so its purpose is clear and cohesion is high.

Coupling: Low - This function is self-contained and does not rely on any other function to achieve its task.

2. Algorithmic Metrics

PSEUDOCODE:

```
FUNCTION natural_merge_sort(arr):
```

```
    sorted <- False
```

```
    WHILE !sorted
```

```
        runs <- []
```

```
        start <- 0
```

```
        WHILE start < len(arr)
```

```
            end <- start + 1
```

```
            WHILE end < len(arr) AND arr[end] >= arr[end - 1]
```

```
                end <- end + 1
```

```
            runs.append(arr[start:end])
```

```
            start <- end
```

```
        IF len(runs) == 1
```

```
            RETURN runs[0]
```

```
        ELSE
```

```
            arr <- find_runs(runs)
```

```
FUNCTION find_runs(runs):
```

```
    WHILE len(runs) > 1
```

```
        merged_runs <- []
```

```
        FOR i IN range(0, len(runs), 2)
```

```
            IF i + 1 < len(runs)
```

```
                merged_runs.append(merge(runs[i], runs[i + 1]))
```

```
            ELSE
```

```
                merged_runs.append(runs[i])
```

```
        runs <- merged_runs
```

```
    RETURN runs[0]
```

```
FUNCTION merge(list1, list2):
```

```

merged <- []
i, j <- 0, 0
WHILE i < len(list1) AND j < len(list2)
    IF list1[i] <= list2[j]:
        merged.append(list1[i])
        i <- i + 1
    ELSE:
        merged.append(list2[j])
        j <- j + 1
merged.extend(list1[i:])
merged.extend(list2[j:])
RETURN merged

```

ALGORITHMIC EFFICIENCY:

natural_merge_sort()

- In the worst case, the entire list may be treated as individual elements requiring a typical $O(n \log n)$ merge complexity. However, with naturally ordered runs, efficiency improves when the input is partially sorted.

find_runs()

- Runs in $O(n)$ because it only scans through the list once, grouping elements into natural runs.

merge()

- Each merge operation between two sorted sublists of sizes a and b takes $O(a + b)$ time, as it only makes a single pass through each sublist. In total, the merging step has $O(n \log n)$ complexity.

OVERALL, THE WORST CASE COMPLEXITY IS $O(N \log N)$!!!

3. Test

- Already Sorted List
 - Input: [1, 2, 3, 4, 5, 6]
 - Expected Output: [1, 2, 3, 4, 5, 6]
- Reversed Sorted List
 - Input: [6, 5, 4, 3, 2, 1]
 - Expected Output: [1, 2, 3, 4, 5, 6]

- c. Random Order List
 - i. Input: [3, 1, 4, 1, 5, 9]
 - ii. Expected Output: [1, 1, 3, 4, 5, 9]
- d. List with Duplicate Elements
 - i. Input: [2, 3, 3, 1, 5, 2]
 - ii. Expected Output: [1, 2, 2, 3, 3, 5]
- e. Single Element List
 - i. Input: [7]
 - ii. Expected Output: [7]
- f. Empty List
 - i. Input: []
 - ii. Expected Output: []

AUTOMATION DRIVER:

```
def test_natural_merge_sort():
```

```
    test_cases = [  
        ([1, 2, 3, 4, 5, 6], [1, 2, 3, 4, 5, 6]),  
        ([6, 5, 4, 3, 2, 1], [1, 2, 3, 4, 5, 6]),  
        ([3, 1, 4, 1, 5, 9], [1, 1, 3, 4, 5, 9]),  
        ([2, 3, 3, 1, 5, 2], [1, 2, 2, 3, 3, 5]),  
        ([7], [7]),  
        ([], [])  
    ]
```

```
    for i, (input_data, expected) in enumerate(test_cases):  
        result = natural_merge_sort(input_data)  
        assert result == expected, f"Test case {i+1} failed: expected  
{expected}, got {result}"
```

```
    print("All test cases passed!")
```