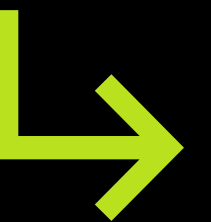


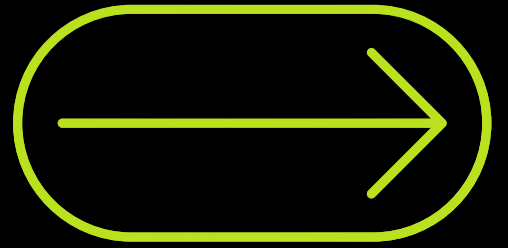
Project 1

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Integration of
Mergesort &
Insertionsort



01 - Algorithm Implementation



02 - Time Complexity Analysis



03 - MergeSort Comparison



04 - Conclusion



01 - Algorithm Implementation

```
def hybrid_sort(arr, l, r, S):
    key_comparison = 0

    if r - l + 1 <= S: # If subarray is lesser than subarray
        subarray = arr[l:r + 1] # Extract the subarray
        key_comparison += insertion_Sort(subarray) # Sort the subarray with Insertion Sort
        arr[l:r + 1] = subarray # Insert sorted subarray back into array
    else:
        if l < r:
            # Use standard merge for larger subarrays
            m = l + (r - l) // 2

            # Sort 1st and 2nd halves
            key_comparison += hybrid_sort(arr, l, m, S)
            key_comparison += hybrid_sort(arr, m + 1, r, S)

            # Merge both the sorted halves
            key_comparison += merge(arr, l, m, r)
        return key_comparison
```

Tips



```
def insertion_Sort(arr): # Returns number of key comparisons.
    key_comparison = 0
    for i in range(1, len(arr)):
        temp = arr[i]
        j = i - 1
        # Move elements of arr[0..i-1], that are
        # greater than temp, to one position ahead
        # of their current position
        while j >= 0 and temp < arr[j]:
            key_comparison += 1
            arr[j + 1] = arr[j]
            j -= 1
        # One more comparison happens after the while loop, if j >= 0
        if j >= 0:
            key_comparison += 1
        arr[j + 1] = temp
    return key_comparison
```

Hybird_sort function enhances sorting efficiency by dynamically switching between insertion sort for smaller subarray and merge for large ones



01 - Algorithm Implementation

Algorithm



```
def merge(arr, l, m, r): # Not sorted inplace.
    key_comparison = 0
    # Find sizes of two subarrays to be merged
    n1 = m - l + 1
    n2 = r - m

    # Create temp arrays
    L = [0] * n1
    R = [0] * n2

    # Copy data to temp arrays L[] and R[]
    for i in range(0, n1):
        L[i] = arr[l + i]

    for j in range(0, n2):
        R[j] = arr[m + 1 + j]

    # Merge the temp arrays back into arr[l..r]
    i = 0 # Initial index of first subarray
    j = 0 # Initial index of second subarray
    k = l # Initial index of merged subarray

    while i < n1 and j < n2:
        key_comparison += 1;
        if L[i] <= R[j]:
            arr[k] = L[i]
            i += 1
        else:
            arr[k] = R[j]
            j += 1
        k += 1

    # Copy the remaining elements of L[], if any
    while i < n1:
        arr[k] = L[i]
        i += 1
        k += 1

    # Copy the remaining elements of R[], if any
    while j < n2:
        arr[k] = R[j]
        j += 1
        k += 1

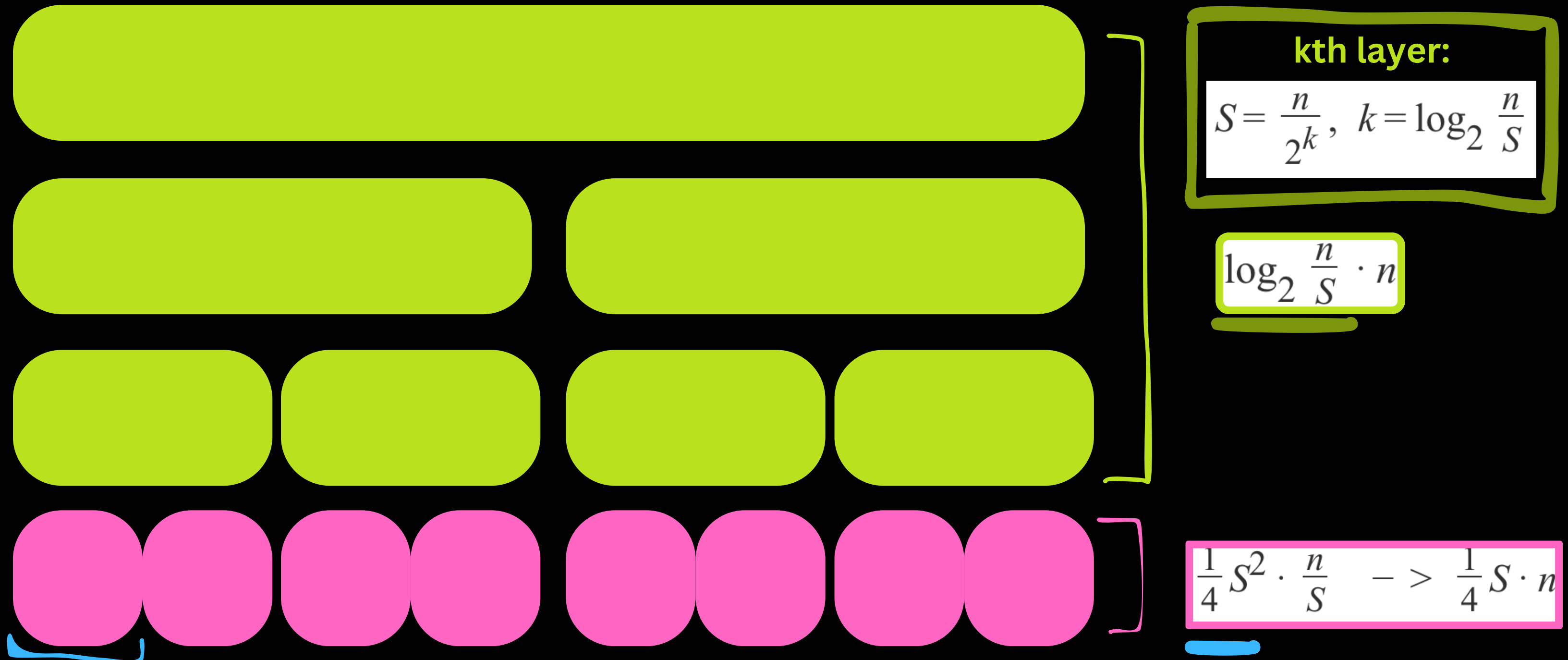
    return key_comparison
```

Flow of Hybrid Sort:

1. Check if Array Size is lesser than threshold
 - a. Extract the sub array from current array
 - b. Sort subarray with insertion sort
 - c. insert the subarray back to current array
2. If array size is bigger than threshold
 - a. If left pointer is smaller than right pointer
 - i. Call hybrid sort with 1st half of array
 - ii. Call hybrid sort with 2nd half of array
 - iii. Merge and sort both halves of the array

02 - Time Complexity Analysis

Theoretical Analysis for Hybrid Sort ↴



02 - Time Complexity Analysis

Theoretical Analysis for Hybrid Sort Continued... ↴

Putting it together....

$$n \cdot \log_2 \frac{n}{S} + \frac{1}{4}S \cdot n = O\left(n \cdot \log_2 \frac{n}{S} + S \cdot n\right)$$

Key Comparison (Fixed Threshold S, Varying Input N)

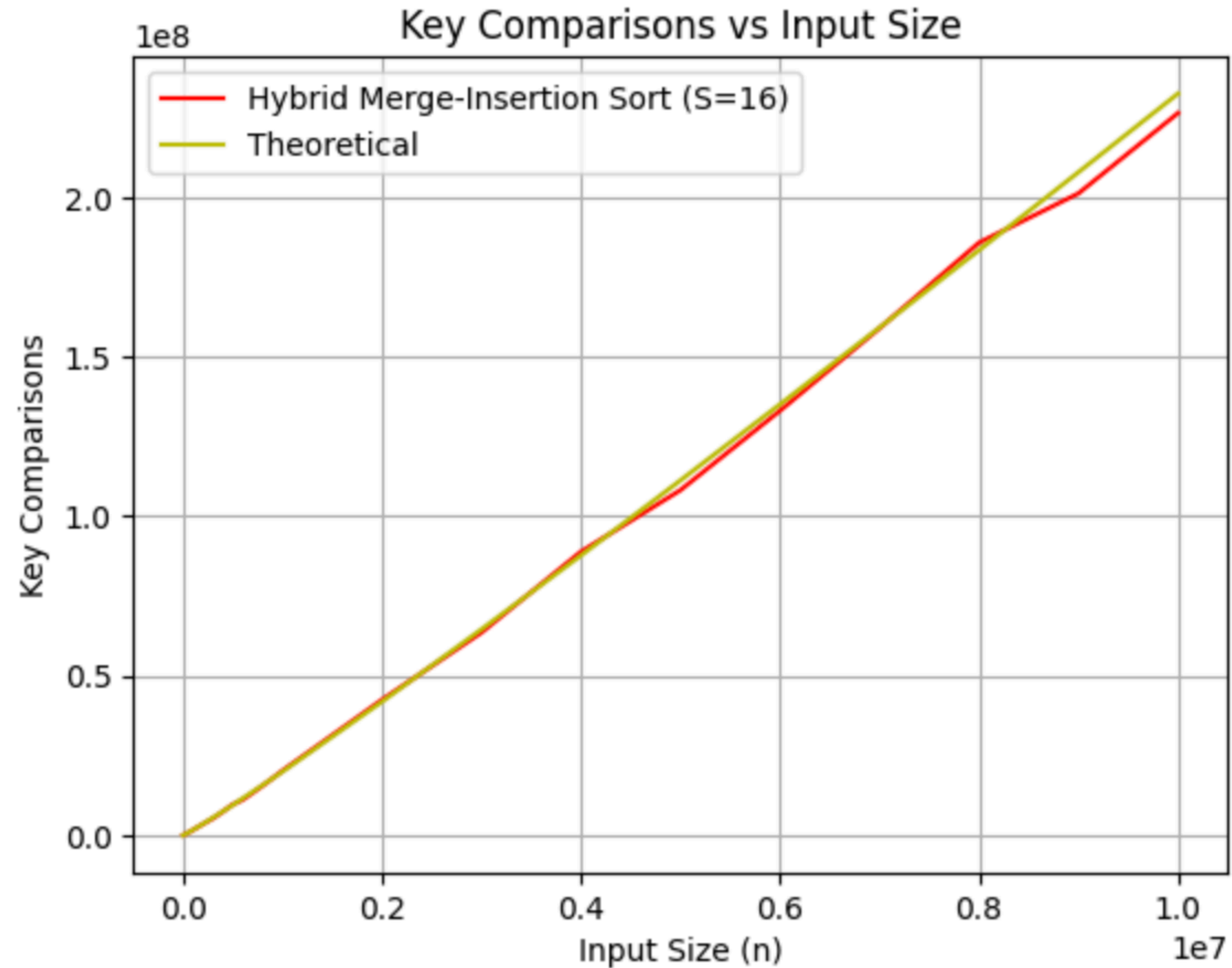
Method

1. Fixed S and varying n -> S threshold fixed at 16
2. Loop over different n sizes
3. Record Key Comparisons and Time for each
4. Plot graph to compare

Key Comparison (Fixed Threshold S, Varying Input N)

✓ both theoretical & empirical models match closely

$$n \cdot \log_2\left(\frac{n}{16}\right) + \frac{16}{4} \cdot n$$

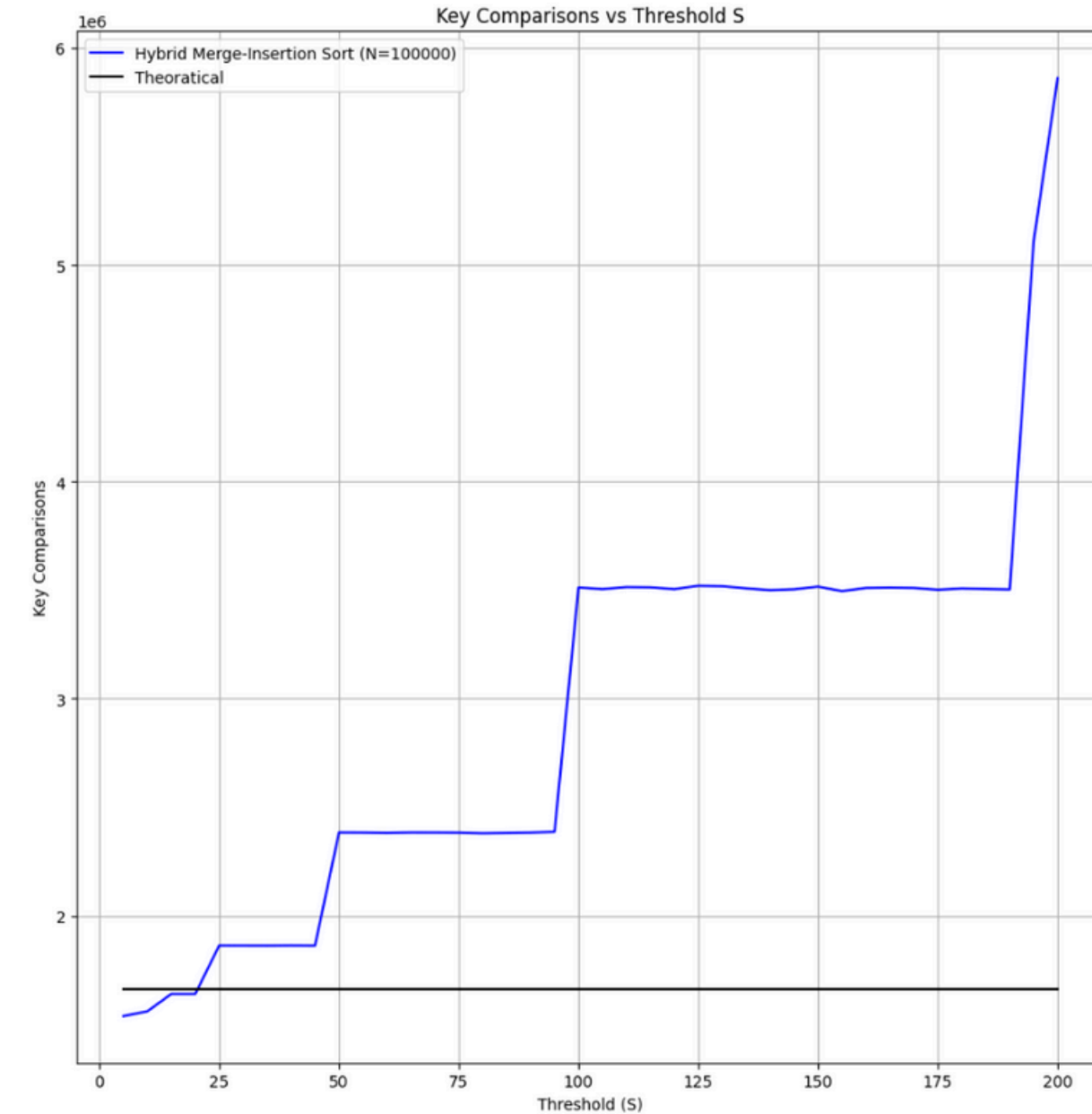
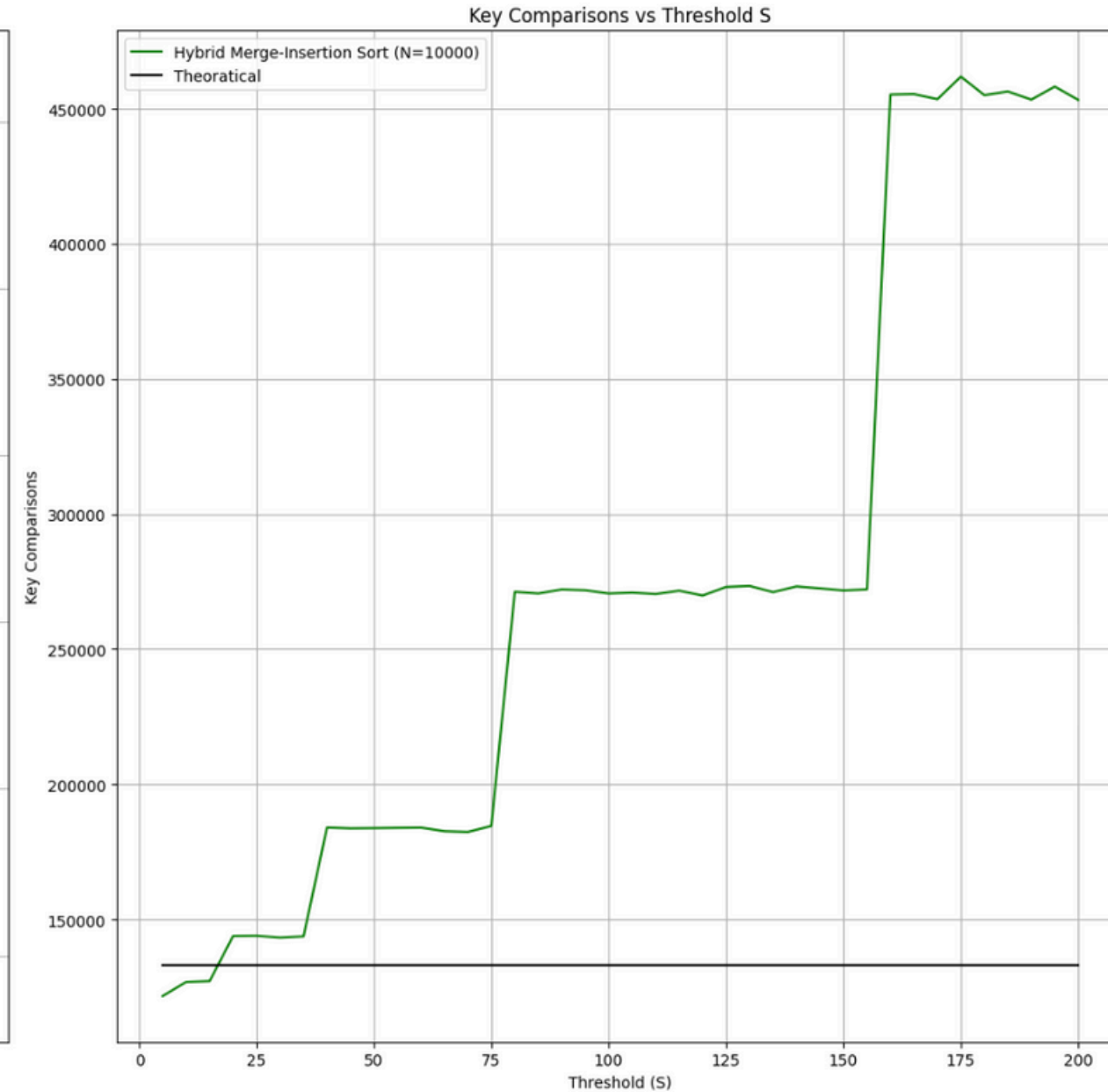
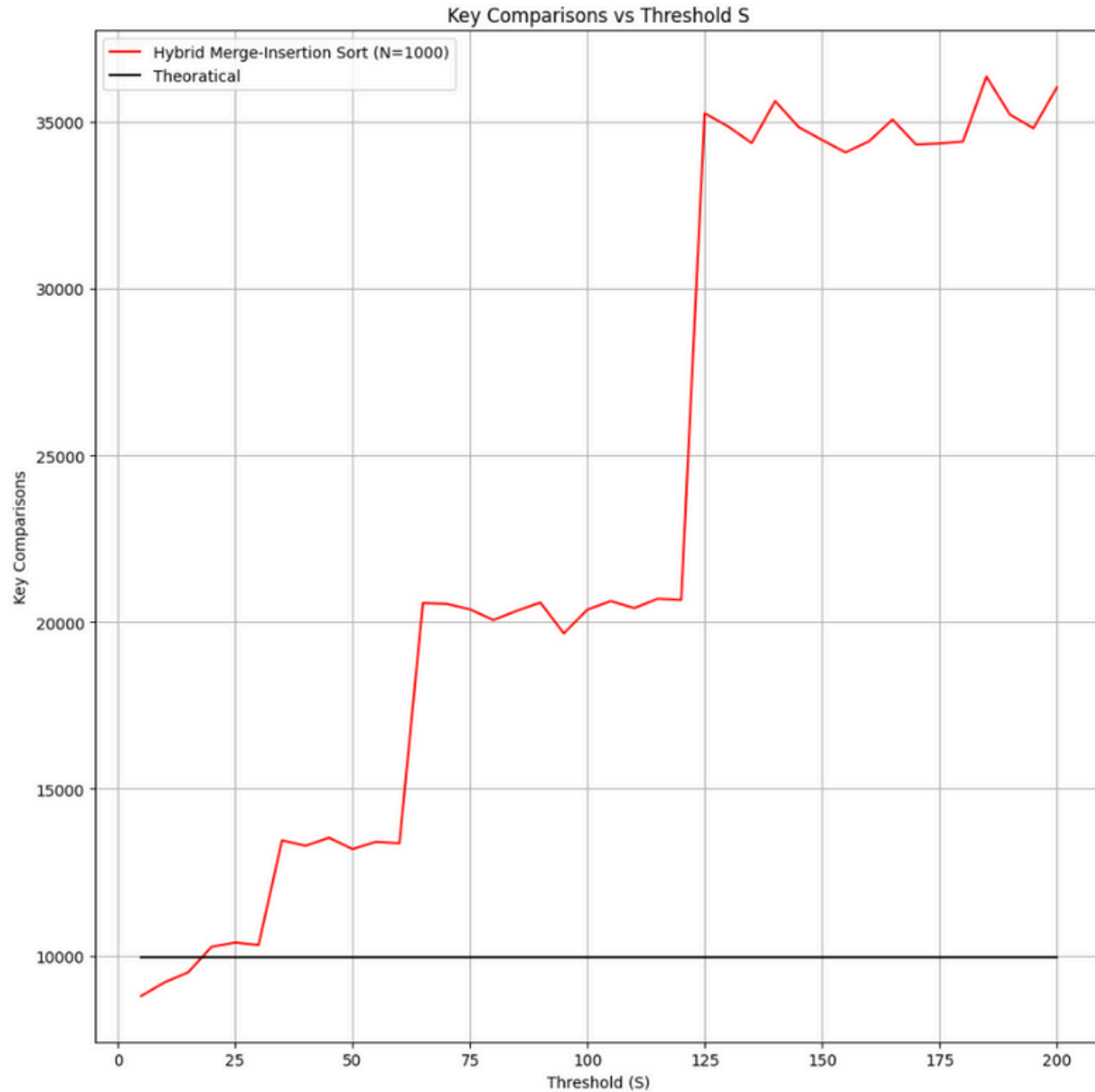


Key Comparison (Varying Threshold S , Fixed Input N)

Method

1. Fixed N size at 100,000
2. Loop and run over hybrid sort for each S
3. Record Key Comparisons and Time
4. Plot Graph

Key Comparison (Varying Threshold S, Fixed Input N)



- Both theoretical & empirical models match closely are quite different
- As n is fixed, input size doesn't increase, thus theoretical model remains a straight line

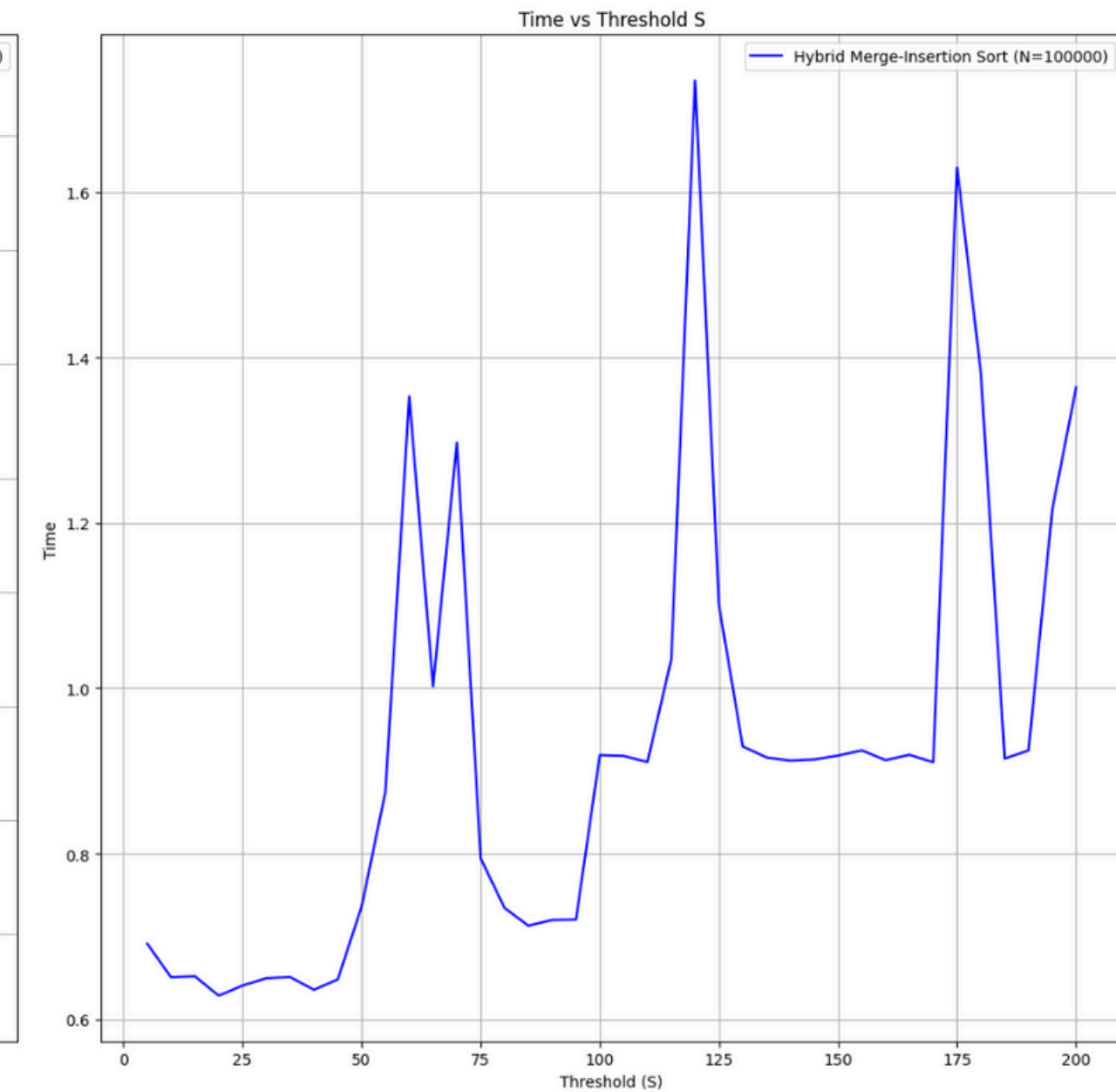
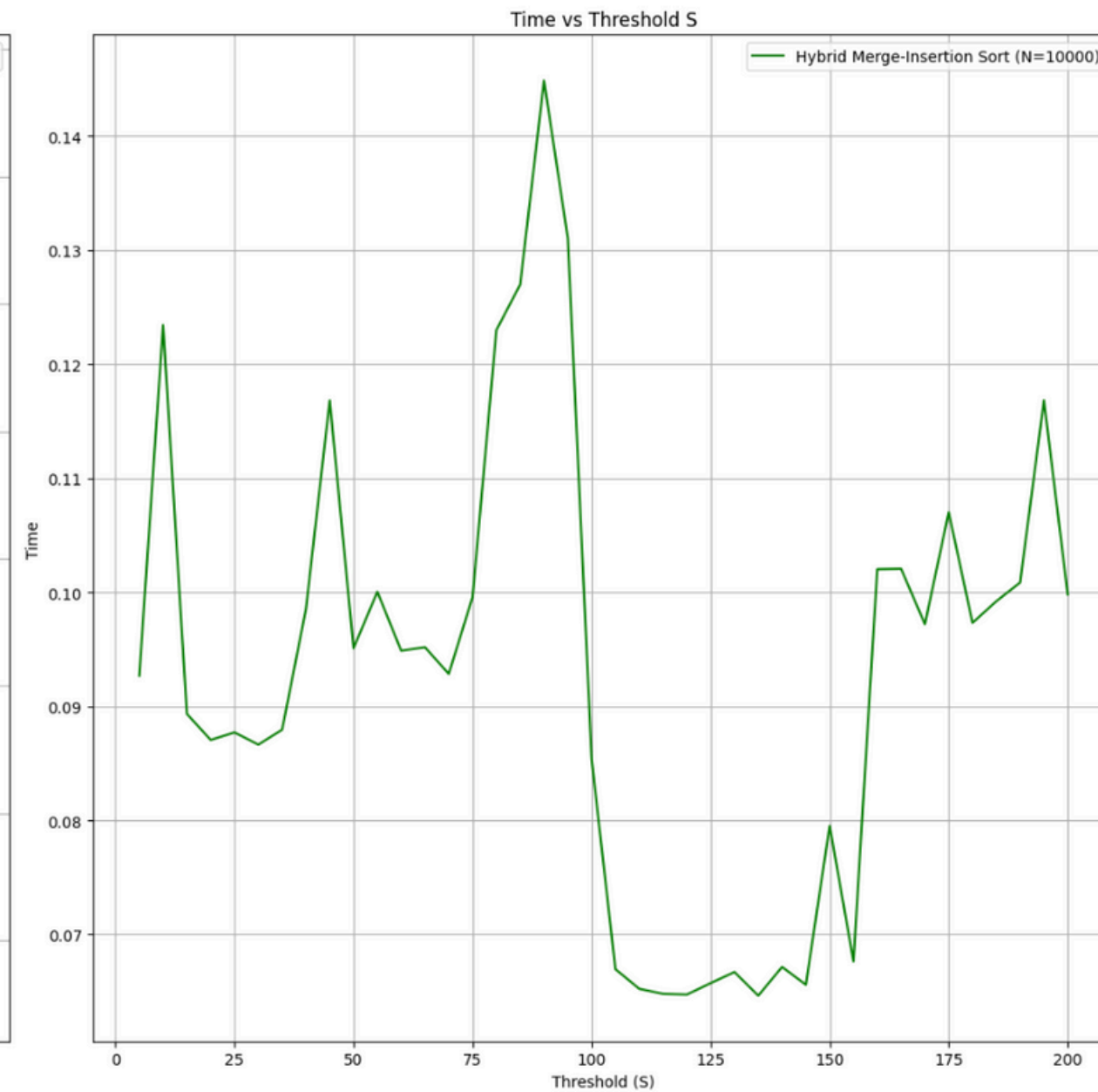
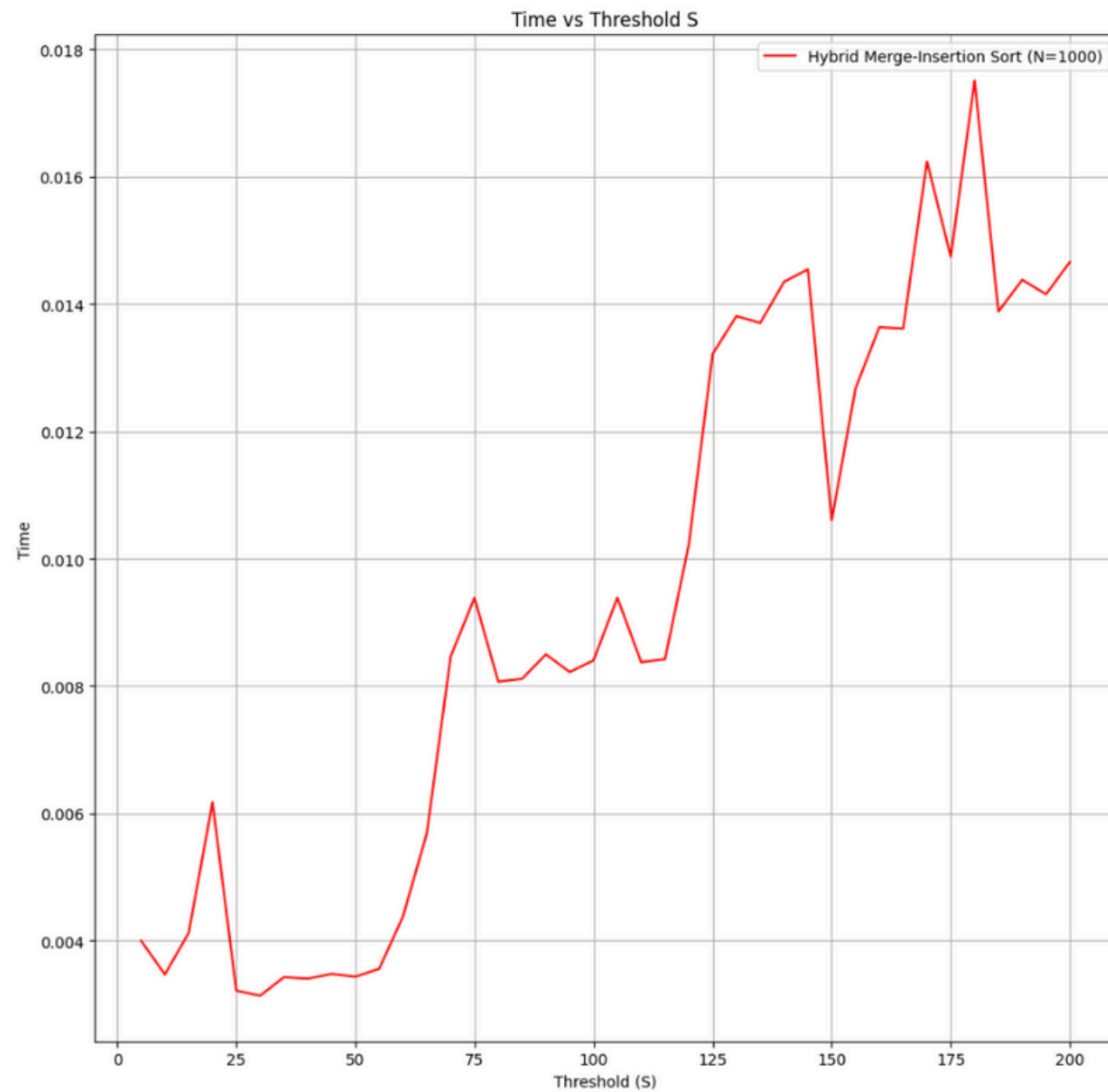
$$n \cdot \log_2\left(\frac{n}{16}\right) + \frac{16}{4} \cdot n$$

Optimal Value of S

Method

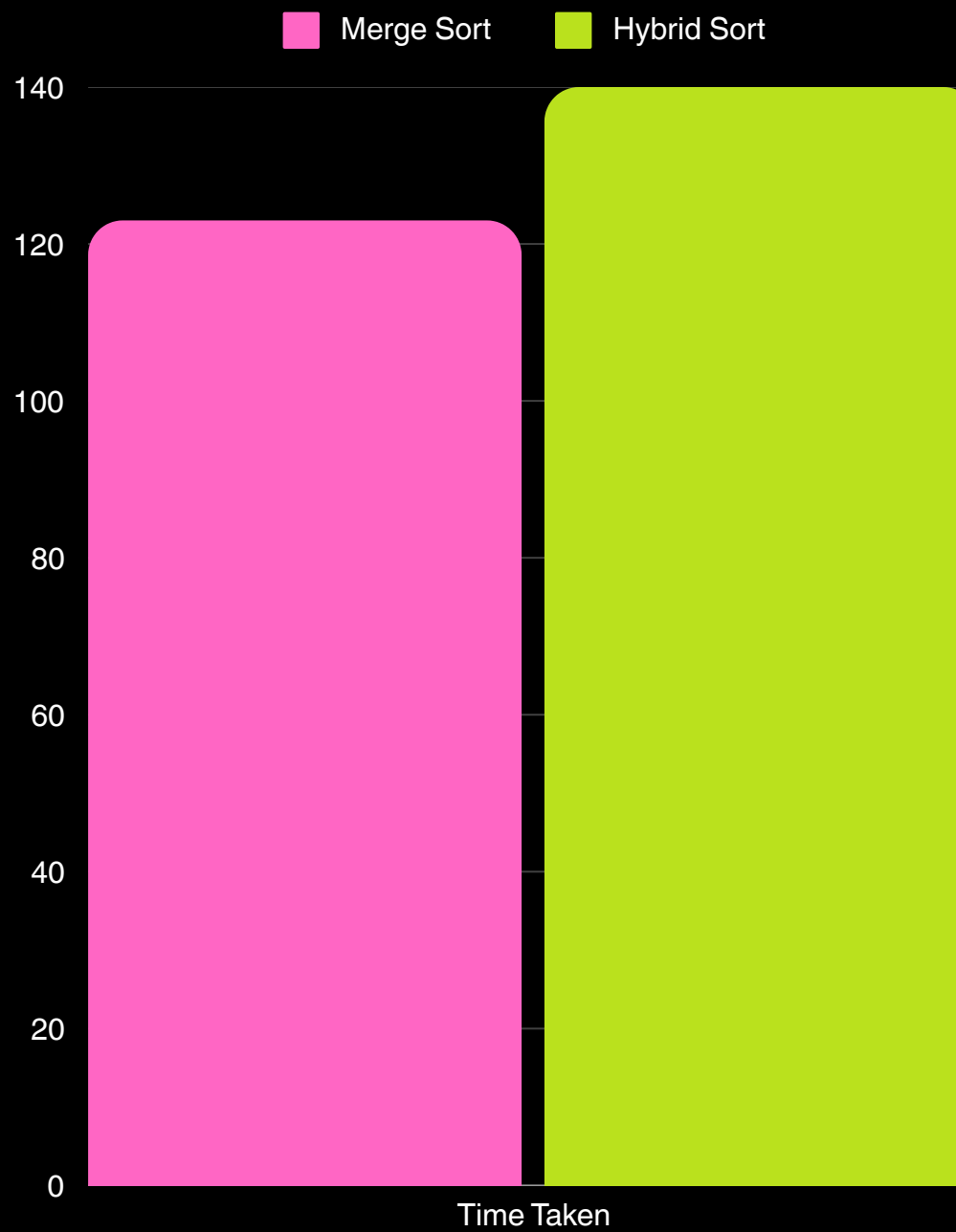
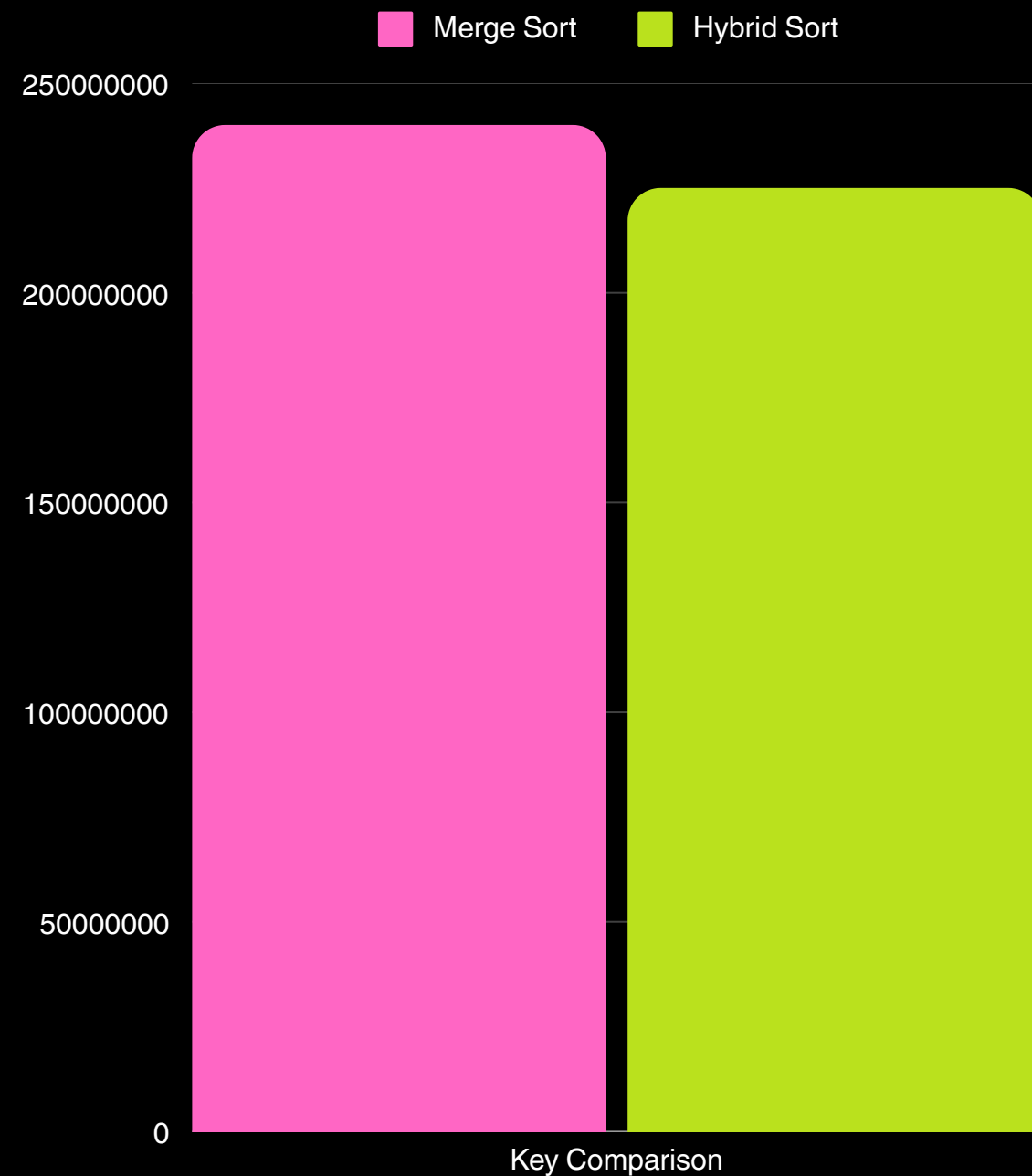
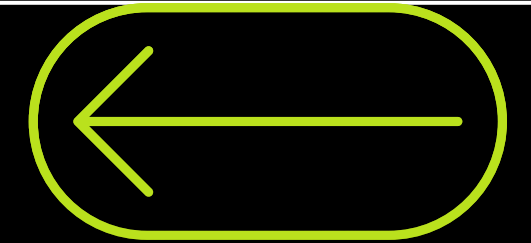
1. Run for $n = 1\ 000, 10\ 000, 100\ 000$
2. Plot against Threshold and Time
3. Derive Optimal S from graph

Optimal Value of S



- The optimal S is the value where the time is the lowest.
- In the graph, there is a low point around $S \approx 30-35$, where the time is minimal.
- Optimal S seems to be 31

03 - MergeSort Comparison



- Optimal S: 31
- Dataset: 10 Million Integers
- Merge Sort
 - Key Comparisons: ~230 000 000
 - CPU Time: ~140s
- Hybrid Sort
 - Key Comparisons: ~240 000 000
 - CPU Time: ~130s

04 - Conclusions

- Hybrid Sort is better in terms of merging:
 - More efficient; lower CPU time usage
- Hybrid Sort had best CPU time performance with varying amounts of input data:
 - Merging small sorted subarrays to minimise comparisons and swaps

Thanks

Q&A