

Count of Smaller Numbers After Self

COURSE : DESIGN ANALYSIS AND ALGORITHMS FOR OPEN ADDRESSING

COURSE CODE:CSA0695

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PROBLEM STATEMENT:

Given an integer array `numbers`, return an integer array `counts` where `counts[i]` is the number of smaller elements to the right of `numbers[i]`.

Example 1:

Input: `numbers = [5,2,6,1]`

Output: `[2,1,1,0]`

Explanation:

To the right of 5 there are 2 smaller elements (2 and 1).

To the right of 2 there is only 1 smaller element (1).

To the right of 6 there is 1 smaller element (1).

To the right of 1 there is 0 smaller element.

ABSTRACT:

- Given an integer array `numbers`, this problem requires computing an integer array `counts` where `counts[i]` represents the number of elements smaller than `numbers[i]` to its right. In other words, for each element in the input array, we need to count the number of smaller elements that follow it. This problem can be solved using a combination of sorting and binary search or merge sort techniques

INTRODUCTION

The "Count of Smaller Numbers After Self" problem is a fundamental algorithmic challenge that has far-reaching implications in various fields, including data analysis, software development, and computer science. Given an integer array, this problem requires computing the number of smaller elements that follow each element in the array. This seemingly simple task proves to be a complex and intriguing problem, as it demands an efficient and scalable solution to accommodate large datasets.

DISCRIPTIONS

KEY POINTS:

Use a data structure (like a Binary Indexed Tree or Fenwick Tree) to efficiently count smaller elements to the right of each element in the array.

A more efficient approach uses a data structure like a Binary Indexed Tree (BIT) or Fenwick Tree to reduce the time complexity to $O(n \log n)$.

APPLICATIONS:

1. Data Analysis: Counting smaller elements can help in data analysis, such as finding the number of data points below a certain threshold.
2. Algorithmic Trading: This problem can be used to analyze stock prices and identify trends.
3. Database Query Optimization: Understanding the count of smaller elements can help optimize database queries.

EXTENSIOS:

his problem can be extended by asking for:

- Instead of counting smaller elements, count the number of larger elements to the right of each element.

Count the number of equal elements to the right of each element.

OUTPUT

```
C:\Users\DELL\Documents\ca  X  +  v
Result: [2, 1, 1, 0]

-- Program by B Manohar, 192211557 --

-----
Process exited after 0.04728 seconds with return value 0
Press any key to continue . . . |
```

COMPLEXITY ANALYSIS

► **Time Complexity:** The time complexity for the "Count of Smaller Numbers After Self" problem is $O(n \log n)$, where n is the length of the input array. This is because the sorting step, which takes $O(n \log n)$ time, dominates the overall time complexity. Although the binary search step has a time complexity of $O(\log n)$

► **Space Complexity:** The space complexity for the "Count of Smaller Numbers After Self" problem is $O(n)$, where n is the length of the input array. This is because the algorithm requires additional space to store the output array, sorted indices array, and input array, each of which has a space complexity of $O(n)$.

CASES

BEST CASE

In the best-case scenario, the input array is already sorted in ascending order, allowing the algorithm to skip the sorting step. The binary search then finds the correct count in the first iteration, resulting in a time complexity of $O(n)$. The space complexity remains $O(n)$ for the output array.

WORST CASE

In the worst-case scenario, the input array is in reverse order or has a large number of duplicate elements, leading to a time complexity of $O(n \log n)$ due to the sorting step. Additionally, the binary search takes $O(\log n)$ time, but since it's performed n times, the overall time complexity becomes $O(n \log n)$.

AVERAGE CASE

In the average case, the input array is randomly ordered, and the algorithm performs with a time complexity of $O(n \log n)$ due to the sorting step. The binary search step also takes $O(\log n)$ time on average, and since it's performed n times, the overall time complexity remains $O(n \log n)$.

FUTURE SCOPE

The "Count of Smaller Numbers After Self" problem has a wide range of future scope and applications. It can be extended to handle complex data structures and analyze large datasets, making it a crucial component of advanced data analysis. Additionally, it can be used as a building block for more complex machine learning algorithms and real-time analytics systems. The problem can also be solved in a distributed computing environment, enabling it to handle massive datasets. Furthermore, it can be adapted to handle streaming data, approximation algorithms.

CONCLUSION:

In conclusion, the algorithm's time complexity ranges from $O(n)$ in the best case to $O(n \log n)$ in the average and worst cases, while the space complexity is $O(n)$ in all cases. The algorithm's performance is significantly affected by the input order, with sorted inputs leading to optimal performance and reverse-ordered inputs resulting in the worst performance. Overall, the algorithm provides a reasonable trade-off between time and space complexity for counting smaller numbers after self.