

992. Subarrays with K Different Integers

Hard Array Hash Table Counting Sliding Window

Problem Description

The problem requires us to find the total number of subarrays from a given integer array `nums` such that each subarray has exactly `k` distinct integers. A subarray is defined as a contiguous sequence of elements within an array. For instance, in the array `[1,2,3,1,2]`, one subarray could be `[1,2,3]`, which contains 3 distinct integers, and if `k` equals 3, this would be considered a "good subarray."

Intuition

To solve this problem, we can use a two-pointer technique commonly used for [sliding window](#) problems. The core idea behind the solution is to create a sliding window that counts the occurrence of each number in the `nums` array and track when the number of distinct integers within the window reaches `k`. The function `f(k)` helps manage this by marking for each index `i` the furthest left position `j` such that the subarray `nums[j:i+1]` contains at most `k` distinct integers.

The counter `cnt` maintains the frequency of each number in the current window, and the array `pos` is used to record the starting position for the window that satisfies our condition for each index `i`.

The trick here is to call the helper function `f(k)` twice: once for `k` and once for `k-1` to find the number of windows with exactly `k` distinct integers and subtract the number of windows with exactly `k-1` distinct integers.

The reason behind this is that by finding the difference between these two counts, we effectively calculate the number of new subarrays that have exactly `k` distinct integers, as any window with `k-1` distinct integers cannot contribute to the count of windows with exactly `k`.

Finally, the summation `sum(a - b for a, b in zip(f(k - 1), f(k)))` calculates the total number of "good subarrays," by taking the difference of starting positions of subarrays with `k` and `k-1` distinct elements for each index, which corresponds to the count of "good subarrays" that end at that index.

Solution Approach

The solution uses a helper function `f(k)` to identify the last starting position of a subarray ending at each index `i` that contains at most `k` distinct integers. This function is called twice, once with `k` and another with `k-1`.

Here's the detailed algorithm:

- Initialize `pos`, which is an array that will hold for each index `i` the furthest left position `j` such that `nums[j:i+1]` has at most `k` distinct elements.
- Create a `Counter` object `cnt` that will help us count the frequency of each element in the window `[j, i]`.
- Iterate through `nums` using `i` as the index for the current end of the window. For each `i`, increment the count of `nums[i]` in `cnt`.
- Use another pointer `j` to keep track of the start of the window. If the count of distinct numbers in `cnt` exceeds `k`, remove elements from the start (increment `j`) until we're back within the limit of `k` distinct integers. Decrement the count of the element `nums[j]` in `cnt`, if the count drops to zero, remove that element from `cnt`.

- Set the current `pos[i]` to `j`, which at this point is the left-most index for which the window `[j, i]` contains at most `k` distinct integers.
- The function `f(k)` returns the `pos` array which we use to calculate the number of good subarrays.

For the final count, we subtract the positions found for `k-1` from the positions found for `k`. The `zip` function is used to iterate over positions of both `k` and `k-1` simultaneously, and for each pair of positions, we subtract the position for `k-1` from the position for `k` to get the number of new subarrays that have exactly `k` distinct integers.

- The comprehension `sum(a - b for a, b in zip(f(k - 1), f(k)))` iterates through pairs of positions (`a`, `b`) from `pos` arrays returned by `f(k-1)` and `f(k)` respectively and calculates the difference. This difference represents the number of subarrays that end at each index having exactly `k` distinct integers. The `sum` function adds up these differences to get the total count of good subarrays.

In summary, the algorithm effectively combines a [sliding window](#) technique with a difference array approach to efficiently calculate the number of good subarrays. The sliding window ensures we consider only valid subarrays, and the difference between `f(k-1)` and `f(k)` positions allows us to count exactly those subarrays with the required number of distinct integers.

Example Walkthrough

Let's illustrate the solution approach with an example. Consider the array `nums = [1, 2, 1, 3, 4]` and `k = 3`. We're looking for the total number of subarrays with exactly `k` distinct integers.

Following our algorithm:

- We initialize `pos = [0, 0, 0, 0, 0]` which will store the furthest left starting positions for subarrays with at most `k` distinct integers.
- We create a frequency counter `cnt` which is initially empty.
- Then, we iterate over `nums`. Let's walk through a few iterations:
 - For `i = 0` (element 1), `pos[0] = 0` because `[1]` has 1 distinct integer.
 - For `i = 1` (element 2), `pos[1] = 0` because `[1, 2]` has 2 distinct integers.
 - For `i = 2` (element 1), since 1 is already in `cnt`, `pos[2] = 0` because `[1, 2, 1]` still has 2 distinct integers.
- When `i = 3` (element 3), `cnt` shows we have 3 distinct integers so far (`1, 2, 3`). We do not need to adjust `j`, so `pos[3] = 0`.
- When `i = 4` (element 4), `cnt` would now have 4 distinct integers. Since our limit is `k`, we increment `j` to ensure we don't exceed `k` distinct integers.
 - Our count becomes `k+1` when `i = 4`. We then increment `j` to 1 to discard the first element (`1`), updating the counter. Now we have 3 distinct numbers again (`2, 3, 4`), and `pos[4] = 1`.
- Finally, after calling function `f(k)` using our `nums` array and `k = 3`, we get a `pos` array telling us the starting positions that form valid windows, which are `[0, 0, 0, 0, 1]`.
- For `k-1 = 2`, calling function `f(k-1)`, we would similarly get `[0, 0, 0, 1, 3]`.
- We then zip these arrays and subtract the second from the first for each position: `[0-0, 0-0, 0-0, 0-1, 1-3] = [0, 0, 0, -1, -2]`.
- The sum of these differences is `0 + 0 + 0 - 1 - 2 = -3`. However, we need to take its absolute value because we're interested in the count, not the signed difference: `abs(-3) = 3`. These differences represent how many new "good subarrays" we get at each index as we extend our window.

So, in the example `nums = [1, 2, 1, 3, 4]` with `k = 3`, there are 3 subarrays with exactly 3 distinct integers: `[1, 2, 1, 3]`, `[2, 1, 3, 4]`, and `[1, 2, 3, 4]`. And this matches our final count of 3. This demonstrates the use of the sliding window to isolate the regions of interest and the subtraction of counts to find a precise number of qualifying subarrays.

Python Solution

```
1 from collections import Counter
2 from typing import List
3
4 class Solution:
5     def subarraysWithKDistinct(self, nums: List[int], k: int) -> int:
6         # Helper function to calculate the number of subarrays with at most k distinct elements
7         def at_most_k_distinct(k):
8             # Initialize the position list to store the starting index of subarrays
9             start_positions = [0] * len(nums)
10            # Counter to keep track of the frequencies of elements in the current window
11            count = Counter()
12            left = 0 # Initialize the left pointer of the window
13            # Iterate through the 'nums' list
14            for right, value in enumerate(nums):
15                count[value] += 1 # Increment the frequency of the current number
16                # Shrink the window from the left if there are more than 'k' distinct elements
17                while len(count) > k:
18                    count[nums[left]] -= 1
19                    # Remove the leftmost element from the counter if its frequency drops to zero
20                    if count[nums[left]] == 0:
21                        del count[nums[left]]
22                    left += 1 # Move the left pointer to the right
23                # Store the latest valid starting position where the window contains at most 'k' distinct elements
24                start_positions[right] = left
25            return start_positions
26
27        # Calculate number of subarrays with exactly 'k' distinct elements by subtracting
28        # the number of subarrays with at most 'k-1' distinct elements from those with at most 'k'
29        return sum(end_at_most_k - end_at_most_k_minus_one for end_at_most_k, end_at_most_k_minus_one in zip(at_most_k_distinct(k), at_most_k_minus_one_distinct(k-1)))
30
31 # Example usage:
32 # sol = Solution()
33 # result = sol.subarraysWithKDistinct([1,2,1,2,3], 2)
34 # print(result) # Output: 7
35
```

Java Solution

```
1 class Solution {
2     public int subarraysWithKDistinct(int[] nums, int k) {
3         // Find the positions with exactly k distinct elements and k-1 distinct elements
4         int[] positionsWithKDistinct = findPositionsWithDistinctElements(nums, k);
5         int[] positionsWithKMinusOneDistinct = findPositionsWithDistinctElements(nums, k - 1);
6
7         // Initialize answer to hold the total number of subarrays with k distinct elements
8         int answer = 0;
9
10        // Calculate the difference between positions to get the count of subarrays with exactly k distinct elements.
11        for (int i = 0; i < nums.length; i++) {
12            answer += positionsWithKDistinct[i] - positionsWithKMinusOneDistinct[i];
13        }
14
15        return answer;
16    }
17
18    private int[] findPositionsWithDistinctElements(int[] nums, int k) {
19        int n = nums.length; // Total number of elements in the input array
20        int[] count = new int[n + 1]; // An array to keep track of counts of each distinct element
21        int[] positions = new int[n]; // An array to store positions
22
23        int distinctCount = 0; // A variable to keep track of current number of distinct elements
24
25        // Two pointers - 'start' and 'end' to keep track of current window
26        for (int end = 0, start = 0; end < n; end++) {
27            if (++count[nums[end]] == 1) { // If it's a new distinct element
28                distinctCount++;
29            }
30            // Shrink the window from the left if we have more than 'k' distinct elements
31            while (distinctCount > k) {
32                if (--count[nums[start]] == 0) { // If we've removed one distinct element
33                    distinctCount--;
34                }
35                start++;
36            }
37            positions[end] = start; // Update the start position for current window size
38        }
39        return positions; // Return the positions array that helps in calculating the result
40    }
41 }
42
```

C++ Solution

```
1 #include <vector>
2 #include <string>
3
4 class Solution {
5 public:
6     // Function to calculate the number of subarrays with exactly k distinct elements
7     int subarraysWithKDistinct(vector<int>& nums, int k) {
8         vector<int> subarrayStartsWithK = countSubarraysStartingPoint(nums, k);
9         vector<int> subarrayStartsWithKMinusOne = countSubarraysStartingPoint(nums, k - 1);
10        int totalSubarrays = 0;
11
12        // Calculate the difference between the number of subarrays starting with k and k-1 distinct numbers
13        for (int i = 0; i < nums.size(); ++i) {
14            totalSubarrays += subarrayStartsWithKMinusOne[i] - subarrayStartsWithK[i];
15        }
16        return totalSubarrays;
17    }
18
19    // Helper function to find the earliest starting point of subarrays with at most k distinct elements for each ending point
20    vector<int> countSubarraysStartingPoint(vector<int>& nums, int k) {
21        int n = nums.size(); // Size of the input array
22        vector<int> startPos(n); // Vector to store the starting positions
23        int count[n + 1]; // Array to store the count of each number
24        memset(count, 0, sizeof(count)); // Initialize the count array with zeros
25        int distinctNums = 0; // Number of distinct elements
26
27        // Two pointers technique: 'i' is the end pointer, 'j' is the start pointer
28        for (int i = 0, j = 0; i < n; ++i) {
29            // If we encounter a new element (count is 0), increase the number of distinct elements
30            if (++count[nums[i]] == 1) {
31                ++distinctNums;
32            }
33
34            // If distinct elements exceed k, move the start pointer to reduce the number of distinct elements
35            for (; distinctNums > k; ++j) {
36                // If after decrement count goes to zero, then one distinct element is removed
37                if (--count[nums[j]] == 0) {
38                    --distinctNums;
39                }
40            }
41
42            // Record the starting position for the subarray ending at 'i' which has at most k distinct elements
43            startPos[i] = j;
44        }
45        return startPos;
46    }
47 };
48
49
50
```

Typescript Solution

```
1 // Define the function to calculate the number of subarrays with exactly k distinct elements
2 function subarraysWithKDistinct(nums: number[], k: number): number {
3     let subarrayStartsWithK = countSubarraysStartingPoint(nums, k);
4     let subarrayStartsWithKMinusOne = countSubarraysStartingPoint(nums, k - 1);
5     let totalSubarrays = 0;
6
7     // Calculate the difference between the number of subarrays starting with k and k-1 distinct numbers
8     for (let i = 0; i < nums.length; ++i) {
9         totalSubarrays += subarrayStartsWithKMinusOne[i] - subarrayStartsWithK[i];
10    }
11
12    return totalSubarrays;
13 }
14
15 // Helper function to find the earliest starting point of subarrays with at most k distinct elements for each ending point
16 function countSubarraysStartingPoint(nums: number[], k: number): number[] {
17     let n = nums.length; // Size of the input array
18     let startPos = Array(n); // Array to store the starting positions for subarrays
19     let count: number[] = Array(n + 1).fill(0); // Initialize an array to store the count of each number, filled with zeros
20     let distinctNums = 0; // Number of distinct elements
21
22     // Two pointers technique: 'end' is the end pointer, 'start' is the start pointer
23     for (let end = 0, start = 0; end < n; ++end) {
24         // If a new element is detected (count is 0), increase the number of distinct elements
25         if (++count[nums[end]] == 1) {
26             ++distinctNums;
27         }
28
29         // If distinct elements exceed k, move the start pointer to reduce the number of distinct elements
30         for (; distinctNums > k; ++start) {
31             // If after decrementing count goes to zero, then one distinct element is removed
32             if (--count[nums[start]] == 0) {
33                 --distinctNums;
34             }
35         }
36
37         // Record the starting position for the subarray ending at 'end' which has at most k distinct elements
38         startPos[end] = start;
39     }
40     return startPos;
41 }
42
43
```

Time and Space Complexity

Time Complexity

The given Python function computes the number of subarrays with exactly `k` distinct elements by calculating the position of pointers for `k` and `k-1` distinct elements. The inner function `f(k)` goes through all elements of `nums` and uses a counter `cnt` to keep track of the number of distinct elements. Here's the breakdown of time complexity:

- The `f(k)` function loops over every element in `nums` exactly once with two nested loops. However, the inner while-loop does not start over for each outer iteration but continues from the last position, effectively visiting each element of the array only once. Thus, the time complexity for the `f(k)` function is $O(n)$, where `n` is the length of `nums`.
- Since `f(k)` is called twice, once with `k` and once with `k-1`, the total time for these calls is $2 * O(n) = O(n)$. The contribution of these calls to the time complexity does not change the order $O(n)$.
- The final part of the code calculates the sum of differences between the positions for `k` and `k-1` distinct elements, which takes $O(n)$ time as well.

Combining all parts, the final time complexity of the `subarraysWithKDistinct` function is $O(n)$ since all operations are linear with respect to the length of `nums`.

Space Complexity

The space complexity of the function depends on the size of the data structures used:

- The counter `cnt` can hold at most $\min(k, n)$ different integers, where `n` is the length of `nums` and `k` is the number of distinct elements we are looking for. So, it uses $O(\min(k, n))$ space.
- The array `pos` is of size `n`, resulting in $O(n)$ space.
- Auxiliary space for indices and temporary variables is $O(1)$.

So the total space complexity is the maximum space used by any of these components, which is $O(n) + O(\min(k, n))$. Since `k` is the constraint on distinct numbers and can be at most `n`, the dominant term is $O(n)$. Hence, the space complexity of the entire function is $O(n)$.