2090. K Radius Subarray Averages

Medium Array Sliding Window

Problem Description

defines a radius. The task is to calculate the k-radius average for each element in the array. The k-radius average for an element at index i includes all the elements from index i - k to i + k, inclusive. To qualify for calculating the average, there must be at least k elements before and after index i. If an element doesn't have enough neighbors to satisfy the radius k, the average for that element will be -1.

It is important to note that the average is computed using integer division, meaning you add all the elements within the range and

In this problem, you are given an array nums containing n integers, indexed from 0 to n-1. You are also given an integer k, which

then divide by the total number of elements, truncating the result to remove the decimal part. For example, for the array [1,3,5,7,9] and k = 1, the result would be [-1, 3, 5, 7, -1] because:

elements at indices 0 and 4 don't have enough neighbors for radius 1;
element at index 1 has an average of (1+3+5)/3, which truncates to 3;

- and so on for the elements at indices 2 and 3.
- and so on for the elements at moleces 2 and 3.

Intuition

However, a more efficient method is to use a moving sum (or <u>sliding window</u>). By maintaining a running sum of the last 2k+1 elements, we can compute the k-radius average for the current index by simply dividing this sum by 2k+1. After computing the

elements within the radius for every index, resulting in a time-consuming process with a time complexity of O(n*k).

The straightforward approach would be to calculate the average for each element individually, which would involve summing

average, move the window by increasing the index i, add the next number in the array to the sum, and subtract the number that just left the window (which would be at index i - 2k). This way, we only need one pass through the array, resulting in a time complexity of O(n).

The provided Python function getAverages implements this efficient sliding window approach. An array ans of the same length as nums is initially filled with -1. This will store our results. The variable s is the sliding sum, which is updated as we iterate over nums with index s. If s is large enough (s >= s × 2), it means we can compute an average for the element at index s + s . We

add the current value v to s, and if the window is valid, we compute the average and update ans[i - k]. One important detail is the use of integer division s // (k * 2 + 1) as required by the problem statement. Finally, we return the completed array ans.

Solution Approach

The solution provided uses a sliding window technique to efficiently compute each k-radius average in a single pass through the

Here's an in-depth explanation of the algorithm's steps:

index i.

s = 0

results.

first element, we can't.

 \circ At index 3, s = 12 + 8 = 20.

Thus, we update the running sum s = 30 - 2 = 28.

those at indices 0, 1, 3 and 4 do not have enough neighbors.

def getAverages(self, nums: List[int], k: int) -> List[int]:

input array nums.

Initialize a running sum, s, which will keep track of the sum of elements in the current window. We will also initialize an array ans with the same length as nums filled with -1s, to store our results.

Iterate through the array nums with both index and value (i and v). Increment the running sum s by the value v.

3. Check if the current index i allows us to have a complete window of 2k+1 elements. This is determined by the condition i >= k * 2. If i is less than k * 2, we cannot calculate the average for i - k since we do not have enough elements before

that's falling out of our window as we move forward.

Here's the core code snippet that illustrates the algorithm:

- If we do have enough elements, calculate the average for the element at index i k as the running sum s divided by 2k+1.

 We use integer division // here as specified by the problem constraints. The result is stored in ans [i k].

 Then, to maintain the size of the window, subtract the element at index i 2k from the running sum s. This is the element
- 6. Continue this process until all elements have been visited, and the ans array is fully populated with the k-radius averages where possible or -1 where the average cannot be calculated.

The algorithm makes use of simple data structures which are a running sum variable s and an array ans to hold the results. The

sliding window pattern here avoids redundantly recalculating the sum for overlapping parts of the window, thus optimizing the process to a time complexity of O(n) where n is the length of the input array.

ans = [-1] * len(nums)
for i, v in enumerate(nums):
 s += v
 if i >= k * 2:
 ans[i - k] = s // (k * 2 + 1)
 s -= nums[i - k * 2]

By maintaining the sliding window and updating the running sum in this manner, the algorithm efficiently computes the required

2. **Iterate Through nums**: We begin iterating over nums. At the start, our running sum s will start accumulating values as we iterate:

○ At index 2, s = 6 + 6 = 12. We reach a point where we could compute an average for the first element (i = 2, k = 2), but since it's the

of the array), we subtract the number that's falling out of the window. Here that's nums[4 - 2 * 2], which is nums[0] = 2.

Initialize Variables: We start by initializing the running sum s = 0 and an array ans = [-1, -1, -1, -1, -1] to store the

Calculate k-Radius Averages:

• At index 4, s = 20 + 10 = 30. Since i = 4 >= k * 2, we can now calculate the average for i - k = 2. We compute it as s // (2 * 2 + 10) = 30.

○ At index 0, s = 0 + 2 = 2. We cannot compute an average yet, as we don't have enough elements before this index.

Check For Complete Window: Now, we reach the first index where a complete window 2k+1 is available:

• At index 1, s = 2 + 4 = 6. We still cannot compute an average for i - k = -1.

 \circ As i = 3 is not less than k * 2 (4), we can't calculate the average for i - k = 1 yet.

1) = 30 // 5 = 6. The running sum at this point includes nums[0] through nums[4] (values 2, 4, 6, 8, 10). We store 6 in ans[2].

Slide The Window: Before we move to the next iteration (which we don't have in this example, since we've reached the end

each window separately for each element, which significantly reduces the time complexity from O(nk) to O(n).

If we had more elements in the array, we would continue this process. However, since we've reached the end, we're done, and our result array ans looks like this: [-1, -1, 6, -1, -1]. The k-radius average is only calculated for the element at index 2 since

Initialize the sum variable to keep running total of the elements in the sliding window.

Initialize the result list to -1 for all elements as the default value.

Loop through the list of numbers while calculating the running sum.

averages[index - k] = window sum // (k * 2 + 1)

// Create a new array for storing prefix sums with length of n + 1

// Function to calculate the averages of all possible subarrays of size k*2+1

// Check if the window has enough elements to calculate the average

// Return the answer vector with calculated averages and -1 for rest

Check if the window has reached the size of (k * 2 + 1).

This means we have enough elements to calculate the average for the middle element.

Calculate the average for the middle element in the window and store it in the result list.

// Remove the element that's no longer in the window from the sum

// Calculate the average for the subarray and store it in the answer vector

std::vector<int> getAverages(std::vector<int>& nums, int k) {

// Initialize the answer vector with -1 for all elements

// Variable to store the sum of the elements within the window

// Determine the size of the input vector

// Add the current element to the sum

ans[i - k] = sum / (k * 2 + 1);

function getAverages(nums: number[], k: number): number[] {

// Initialize the length of the input array for easier reference

sum -= nums[i - k * 2];

int n = nums.size();

long sum = 0;

return ans;

std::vector<int> ans(n, -1);

// Iterate over the nums vector

for (int i = 0; i < n; ++i) {

sum += nums[i];

if (i >= k * 2) {

window_sum -= nums[index - k * 2]

Solution Implementation

Python

from typing import List

This example illustrates the efficiency of the sliding window technique in avoiding redundant calculations. We avoid iterating over

window_sum += value

Check if the window has reached the size of (k * 2 + 1).

This means we have enough elements to calculate the average for the middle element.

Subtract the element that's moving out of the sliding window to maintain the correct window size.

Calculate the average for the middle element in the window and store it in the result list.

```
# Return the final list of averages.
return averages

Java
```

int n = nums.length;

class Solution {

window sum = 0

averages = [-1] * len(nums)

if index >= k * 2:

for index, value in enumerate(nums):

public int[] getAverages(int[] nums, int k) {

// Get the length of the input array

long[] prefixSums = new long[n + 1];

class Solution:

```
// Compute the prefix sums
        for (int i = 0; i < n; ++i) {
           prefixSums[i + 1] = prefixSums[i] + nums[i];
       // Initialize the answer array with -1, which signifies positions where
       // the k-range average cannot be computed
        int[] averages = new int[n];
       Arrays.fill(averages, −1);
       // Determine the averages for the k-range for each valid position
        for (int i = 0; i < n; ++i) {
           // Check if current index i allows a full k-range on both sides
           if (i - k >= 0 \&\& i + k < n) {
               // Calculate the sum for this k-range
                long sumForRange = prefixSums[i + k + 1] - prefixSums[i - k];
               // Calculate the average and cast it to int before storing it in the result array
               // (k \ll 1 \mid 1) calculates the size of the range, which is (2 * k + 1)
               averages[i] = (int) (sumForRange / (2 * k + 1));
       // Return the completed array with averages and -1 for non-computable positions
       return averages;
#include <vector>
class Solution {
public:
```

const arrayLength = nums.length; // Initialize the answer array with -1 for all elements // -1 will be used for positions where the average cannot be calculated const averages: number[] = new Array(arrayLength).fill(-1);

};

TypeScript

```
// A running sum of the elements in the current window
    let windowSum = 0;
   // Iterate over the array of numbers
    for (let i = 0; i < arrayLength; ++i) {</pre>
       // Add the current element to the running sum
       windowSum += nums[i];
       // Check if the window has reached the size of k elements on both sides
       if (i >= k * 2) {
           // Calculate the average for the middle element of the current window
            // The window size is k elements before, the element itself, and k elements after (total 2k + 1 elements)
            averages[i - k] = Math.floor(windowSum / (k * 2 + 1));
            // Subtract the element that is moving out of the window from the running sum
            // This is the element i - 2k where i is the current index of the traversal
           windowSum -= nums[i - k * 2];
   // Return the array of calculated averages and -1 for elements where it could not be calculated
   return averages;
from typing import List
class Solution:
   def getAverages(self, nums: List[int], k: int) -> List[int]:
       # Initialize the sum variable to keep running total of the elements in the sliding window.
       window sum = 0
       # Initialize the result list to -1 for all elements as the default value.
       averages = [-1] * len(nums)
       # Loop through the list of numbers while calculating the running sum.
        for index, value in enumerate(nums):
           window sum += value
```

averages[index - k] = window sum // (k * 2 + 1) # Subtract the element that's moving out of the sliding window to maintain the correct window size. window_sum -= nums[index - k * 2]

Time and Space Complexity

return averages

if index >= k * 2:

Return the final list of averages.

The time complexity of the given code is O(n), where n is the length of the input list nums. This is because there is a single loop that iterates over all elements of nums once. Within the loop, operations are done in constant time including calculating the sum for the average and updating the sum by subtracting the element that is falling out of the sliding window.

The space complexity of the code is O(n), where n is the length of the input list nums. This is due to the ans list which is initialized to the same length as nums. There are no other data structures that depend on the size of the input that would increase the space complexity.