Problem Description

The problem presents us with an array of integers, nums, and asks us to find the kth smallest sum of any subarray within this array. A subarray is defined as a contiguous part of the original array—which means any sequence of elements that are next to each other without gaps—and the sum of a subarray is the total sum of its elements. So, if nums is [1, 2, 3], then [1, 2], [2, 3], and [1, 2, 3] are all subarrays, with sums 3, 5, and 6, respectively. In simple terms, we want to look through all possible continuous sequences of numbers in the list, calculate the sum for each, and

find out which is the kth smallest among these sums.

Intuition

The challenge lies in efficiently finding the kth smallest sum without having to compute and sort all possible subarray sums, which would be a very time-consuming process, especially for large arrays.

The intuition behind this kind of problem is to use Binary Search in an innovative way. Instead of searching for an explicit value in an array, we use Binary Search to find the smallest sum s such that there are at least k subarrays in nums whose sums are less than or equal to s. This is not a straightforward application of Binary Search because there is no sorted list of subarray sums to begin with.

Here's how the solution works: First, we define a function f(s) which counts how many subarrays have a sum less than or equal to s. We use a two-pointer

technique to maintain a sliding window of elements whose sum is less than or equal to s. As we iterate through nums, we keep

expanding our window by adding new elements to the sum t, and we shrink the window from the left by removing elements whenever the total sum exceeds s. This way, f(s) returns true if there are at least k such subarrays.

Next, we set up our binary search range between 1 and r where 1 is the minimum element in nums (since no subarray sum can be smaller than the smallest element) and r is the sum of all elements in nums (since no subarray sum can be larger than this). Using the bisect_left method from the bisect module, we perform our Binary Search within the range [1, r]. The key argument in bisect_left allows us to use the function f to check if a mid-value is feasible (i.e., the count of subarrays with sum less than or

equal to this value is at least k). The method zeroes in on the smallest s that satisfies f(s), which will be the kth smallest subarray sum in nums.

This approach is efficient because each iteration of Binary Search narrows down the range of possible sums by half, and for each

such guess, counting the subarrays takes linear time, resulting in a total time complexity of O(n log x), where x is the maximum subarray sum. Solution Approach

less than or equal to a given value. The code defines a helper function f(s) to use within the binary search. We can describe what each part of this function does and how it integrates with the binary search mechanism:

problems involving sorted arrays or monotonic functions. In this case, the monotonic function is the number of subarrays with sums

f(s) uses a sliding window with two pointers, i and j, which represent the start and end of the current subarray being

The kthSmallestSubarraySum function in the provided Python code leverages binary search, which is a classic optimization for

If the sum t exceeds the value s, we subtract elements starting from nums [j] and increment j to reduce t. This maintains the window where the subarray sum is less than or equal to s.

considered.

2. Counting Subarrays

sum.

linearly with the two-pointer technique.

2. Sliding Window via Two Pointers:

Example Walkthrough

sum.

1. Sliding Window via Two Pointers

- The line cnt += i j + 1 is crucial. It adds the number of subarrays ending at i for which the sum is less than or equal to s. This works because for every new element added to the window, there are i - j + 1 subarrays. For instance, if our window is [nums[j], ..., nums[i]], then [nums[i]], [nums[i-1], nums[i]], up to [nums[j], ..., nums[i]] are all valid subarrays.
- 3. Binary Search

As we iterate over the array with i, we add each nums [i] to the temporary sum t.

 The bisect_left function from the bisect module is used with a custom key function, which is the f(s) we defined earlier. The key function transforms the value we're comparing against in the binary search. o bisect_left will find the smallest value within the range [1, r] for which f(s) returns True, indicating it is the kth smallest

• The actual search takes place over a range of potential subarray sums, from min(nums) to sum(nums). We use the binary

search algorithm to efficiently find the smallest sum that has at least k subarrays less than or equal to it.

- The combination of these techniques results in an efficient solution where, rather than having to explicitly sort all the subarray sums, we can deduce the kth smallest sum by narrowing down our search space logarithmically with binary search and counting subarrays
- First, we initialize our helper function f(s) that counts subarrays with sums less than or equal to s. Now let's illustrate this function: 1. Initializing Variables:

• We would start with i = 0, meaning our subarray only includes nums [0] for now which is [1]. The temporary sum t starts at

Let's apply the solution approach to a small example. Suppose nums = [1, 2, 3] and we are looking for the 2nd smallest subarray

• At i = 0, we add nums [0] to t so, t = 1. Since $t \le 3$, we continue to i = 1.

4. Binary Search:

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3 again. However, with numbers [1, 2, 3], no subarray ending at index 2 has a sum ≤ 3 other than [3] itself. 3. Counting Subarrays:

∘ For each 1, the number of subarrays ending at 1 with a sum ≤ 3 is added up. We use cnt += 1 - 1 + 1 to do this count.

• We set l = 1 (smallest element) and r = 6 (sum of all elements). We then search for the smallest sum s where f(s) gives us

subarrays with sum < 3.5, which is 4. Since we are looking for the 2nd smallest sum, 4 subarrays mean we can try a smaller

 \circ We adjust our binary search boundaries to 1 = 1 and r = 3. We check for s = 2. The subarrays within this constraint are [1]

The binary search will now try to see if there is a sum smaller than 2 that also satisfies the condition, but since 2 is the sum

○ If we move to i = 2, t would become 6 which exceeds 3. Therefore, we adjust j to remove elements from the start until t ≤

Let's say we're checking for s = 3, we want to count how many subarrays have sum ≤ 3.

• At i = 1, t becomes t + nums[1] = 3. This sum is still ≤ 3 and we can include [1, 2].

and [2], which gives us count 2. Here, f(2) would return True since it matches our k value.

Helper function to check if the count of subarrays with sum less than or equal to 'limit'

count = 0 # Count of subarrays with sum less than or equal to 'limit'

33 # 1. A binary search is applied to find the kth smallest sum within the range of the minimum

37 # 3. The bisect_left function is used to find the insertion point (the kth smallest sum) for

which the condition in `has_k_or_more_subarrays_with_sum_at_most` returns True.

possible subarrays with a sum less than or equal to the passed limit is at least k.

35 # 2. The `has_k_or_more_subarrays_with_sum_at_most` function checks if the number of all

element (left) and the sum of all elements (right), inclusively.

Shrink the window from the left if the total sum exceeds the limit

The count is increased by the number of subarrays ending with nums[end]

smallest subarray sum without having to explicitly calculate every possible subarray sum.

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

start = 0 # Start index for the current subarray

def has_k_or_more_subarrays_with_sum_at_most(limit):

total_sum = 0 # Sum of the current subarray

Iterate over the numbers in the array

total_sum -= nums[start]

Check if we have at least k subarrays

for end, num in enumerate(nums):

while total_sum > limit:

count += end - start + 1

total_sum += num

start += 1

return count >= k

Subarrays are [1], [2], [3], [1, 2]. Thus, the count here is 4.

at least k subarrays. In this case, k = 2. • Using binary search, we check s = (1 + r) / 2 which is 3.5 initially, then we apply the helper function to count how many

is at least k

of our smallest individual element and there's no smaller sum that could give us 2 subarrays, 2 is indeed our 2nd smallest sum.

By using this approach of binary search combined with a two-pointer technique to count subarrays, we efficiently find the 2nd

Python Solution 1 from bisect import bisect_left 2 from typing import List class Solution:

24 25 # Binary search to find the kth smallest subarray sum 26 left, right = min(nums), sum(nums) 27 # Perform binary search with a custom key function by using the bisect_left function 28 kth_smallest_sum = left + bisect_left(range(left, right + 1), True,

key=has_k_or_more_subarrays_with_sum_at_most)

29 30 return kth_smallest_sum 31 32 # Explanation:

```
Java Solution
   class Solution {
       public int kthSmallestSubarraySum(int[] nums, int k) {
           // Initialize the left and right boundaries for the binary search.
           // Assume the smallest subarray sum is large, and find the smallest element of the array.
            int left = Integer.MAX_VALUE, right = 0;
            for (int num : nums) {
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                left = Math.min(left, num);
                right += num;
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           // Perform binary search to find the kth smallest subarray sum.
           while (left < right) {</pre>
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                int mid = left + (right - left) / 2;
13
               // If there are more than k subarrays with a sum <= mid, move the right pointer.
                if (countSubarraysWithSumAtMost(nums, mid) >= k) {
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                    right = mid;
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                } else {
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                    // Otherwise, move the left pointer.
18
                    left = mid + 1;
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           // The left pointer points to the kth smallest subarray sum.
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            return left;
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25
       // Helper method to count the number of subarrays with a sum at most 's'.
       private int countSubarraysWithSumAtMost(int[] nums, int s) {
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27
            int currentSum = 0, start = 0;
28
            int count = 0;
29
            for (int end = 0; end < nums.length; ++end) {</pre>
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                currentSum += nums[end];
31
               // If the current sum exceeds 's', shrink the window from the left.
               while (currentSum > s) {
32
33
                    currentSum -= nums[start++];
34
35
               // Add the number of subarrays ending at index 'end' with a sum at most 's'.
                count += end - start + 1;
36
37
            return count;
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```

12 13 // Lambda function to count the subarrays with sum less than or equal to 'sum' auto countSubarraysLEQ = [&](int sum) { 14 int totalCount = 0; 16

C++ Solution

class Solution {

// Function to find the kth smallest subarray sum

int leftBound = INT_MAX, rightBound = 0;

leftBound = min(leftBound, x);

for (int x : nums) {

rightBound += x;

int kthSmallestSubarraySum(vector<int>& nums, int k) {

// Initialize the left and right boundaries for binary search

// Find the smallest number in the nums array and the total sum

2 public:

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```
int subarraySum = 0;
                for (int i = 0, j = 0; i < nums.size(); ++i) {
17
                    subarraySum += nums[i];
18
                    // Increment j to maintain the condition that subarraySum <= sum
19
                    while (subarraySum > sum) {
20
                        subarraySum -= nums[j++];
21
22
23
                    // Update the total count of subarrays ending at index i
24
                    totalCount += i - j + 1;
25
26
                return totalCount;
27
           };
28
29
           // Binary search to find the kth smallest subarray sum
30
           while (leftBound < rightBound) {</pre>
                int mid = leftBound + (rightBound - leftBound) / 2;
31
                if (countSubarraysLEQ(mid) >= k) {
                    rightBound = mid;
34
                } else {
35
                    leftBound = mid + 1;
36
37
38
           // The left bound will be the kth smallest subarray sum
            return leftBound;
41
42 };
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Typescript Solution
 1 // Function to find the kth smallest subarray sum
   function kthSmallestSubarraySum(nums: number[], k: number): number {
       // Initialize the left and right boundaries for binary search
       let leftBound = Number.MAX_SAFE_INTEGER;
       let rightBound = 0;
       // Find the smallest number in the nums array and the total sum
       nums.forEach((x) \Rightarrow {
 8
            leftBound = Math.min(leftBound, x);
 9
            rightBound += x;
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       });
11
12
13
       // Function to count the subarrays with sum less than or equal to 'sum'
14
       const countSubarraysLEQ = (sum: number): number => {
            let totalCount = 0;
15
            let subarraySum = 0;
16
            for (let i = 0, j = 0; i < nums.length; ++i) {
17
                subarraySum += nums[i];
               // Increment j to maintain the condition that subarraySum <= sum
19
20
               while (subarraySum > sum) {
21
                    subarraySum -= nums[j++];
```

37 38 39 // The left bound will be the kth smallest subarray sum return leftBound; 40

} else { 34 35 36

Time Complexity

};

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41 }

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29 // Binary search to find the kth smallest subarray sum while (leftBound < rightBound)</pre> let mid = leftBound + Math.floor((rightBound - leftBound) / 2); if (countSubarraysLEQ(mid) >= k) { 32 rightBound = mid; leftBound = mid + 1;

// Update the total count of subarrays ending at index i

totalCount += i - j + 1;

return totalCount;

Time and Space Complexity

an integer list nums. 1. The binary search is performed on a range from the minimum element in nums to the sum of all elements in nums. This range is r -1+1, where 1 is the minimum value (i.e., min(nums)) and r is the sum of the numbers in nums. Since we are effectively halving the search space with each iteration, the time complexity for the binary search alone is $O(\log(r-1))$.

The given Python code implements a binary search combined with a two-pointer technique to find the k-th smallest subarray sum in

2. Inside the binary search, we use function f() to count the number of subarray sums that are less than or equal to a given sum s by using a sliding window technique with the two pointers 1 and 1. This is done by iterating through the array once for each value of s. The inner loop moves j appropriately but does not iterate more than n times across all iterations of 1 since it only subtracts the values that were previously added. Therefore, for each s, the function f() has a time complexity of O(n) where n is the length of nums. Combining the binary search and the sliding window, the total time complexity is $0(n * \log(r - 1))$.

Space Complexity

The space complexity of the code is 0(1), not counting the input nums. This is because the code only uses a fixed number of integer variables (1, r, s, t, j, and cnt) and does not create additional data structures that grow with the size of the input.