

Grokking the Coding Interview: Patterns for Coding Questions



Characters (medium)
Fruits into Baskets (medium)

sum (easv)

Smallest Subarray with a given

Longest Substring with K Distinct

- No-repeat Substring (hard)

 Longest Substring with Same

 Letters after Replacement (hard)
- Longest Subarray with Ones after Replacement (hard)
- Problem Challenge 1 Solution Review: Problem Challenge 1
- Problem Challenge 2
 Solution Review: Problem
- Challenge 2
 Problem Challenge 3
- Solution Review: Problem Challenge 3
- Problem Challenge 4

 Solution Review: Problem Challenge 4

Pattern: Two Pointers

Introduction

Pair with Target Sum (easy) Remove Duplicates (easy) Squaring a Sorted Array (easy) Triplet Sum to Zero (medium) Triplet Sum Close to Target (medium) Triplets with Smaller Sum (medium) Subarrays with Product Less than a Target (medium) Dutch National Flag Problem (medium) Problem Challenge 1 Solution Review: Problem Challenge 1 Problem Challenge 2

> Solution Review: Problem Challenge 2

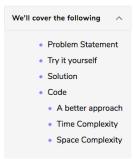
> Solution Review: Problem Challenge 3

Problem Challenge 3

Pattern: Fast & Slow pointers

Introduction
LinkedList Cycle (easy)
Start of LinkedList Cycle (medium)

Maximum Sum Subarray of Size K (easy)



Problem Statement

Given an array of positive numbers and a positive number 'k', find the maximum sum of any contiguous subarray of size 'k'.

Example 1:

```
Input: [2, 1, 5, 1, 3, 2], k=3
Output: 9
Explanation: Subarray with maximum sum is [5, 1, 3].
```

Example 2:

```
Input: [2, 3, 4, 1, 5], k=2
Output: 7
Explanation: Subarray with maximum sum is [3, 4].
```

Try it yourself

Try solving this question here:



Solution

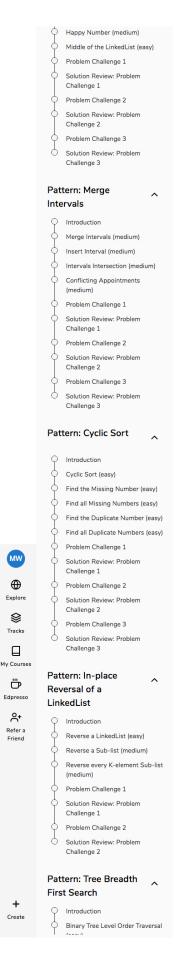
A basic brute force solution will be to calculate the sum of all 'k' sized subarrays of the given array, to find the subarray with the highest sum. We can start from every index of the given array and add the next 'k' elements to find the sum of the subarray. Following is the visual representation of this algorithm for Example-1:



Code

Here is what our algorithm will look like:





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```
(i = 0; i < arr.length - k + 1; i++) {
      for (j = i; j < i + k; j++) {
  windowSum += arr[j];</pre>
      maxSum = Math.max(maxSum, windowSum);
  return maxSum;
console.log(`Maximum sum of a subarray of size K: $\{max_sub_array_of_size_k(3, [2, 1, 5, 1, 3, 2])\}`); \\ console.log(`Maximum sum of a subarray of size K: $\{max_sub_array_of_size_k(2, [2, 3, 4, 1, 5])\}`); \\ \\
                                                                                                                               SAVE
                                                                                                                                               RESET
                                                                                                                                                            03
```

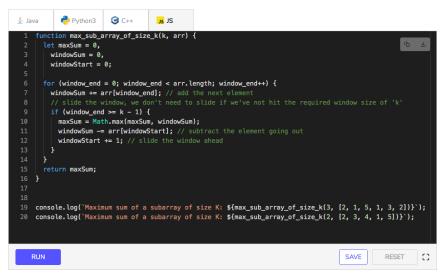
The time complexity of the above algorithm will be O(N*K), where 'N' is the total number of elements in the given array. Is it possible to find a better algorithm than this?

A better approach

If you observe closely, you will realize that to calculate the sum of a contiguous subarray we can utilize the sum of the previous subarray. For this, consider each subarray as a Sliding Window of size 'k'. To calculate the sum of the next subarray, we need to slide the window ahead by one element. So to slide the window forward and calculate the sum of the new position of the sliding window, we need to do two things:

- 1. Subtract the element going out of the sliding window i.e., subtract the first element of the window.
- 2. Add the new element getting included in the sliding window i.e., the element coming right after the end of the window.

This approach will save us from re-calculating the sum of the overlapping part of the sliding window. Here is what our algorithm will look like:



Time Complexity

The time complexity of the above algorithm will be O(N).

Space Complexity

The algorithm runs in constant space O(1).

