



Maximum Sum Subarray of Size K (easy)

We'll cover the following

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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#

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Try solving this question here:

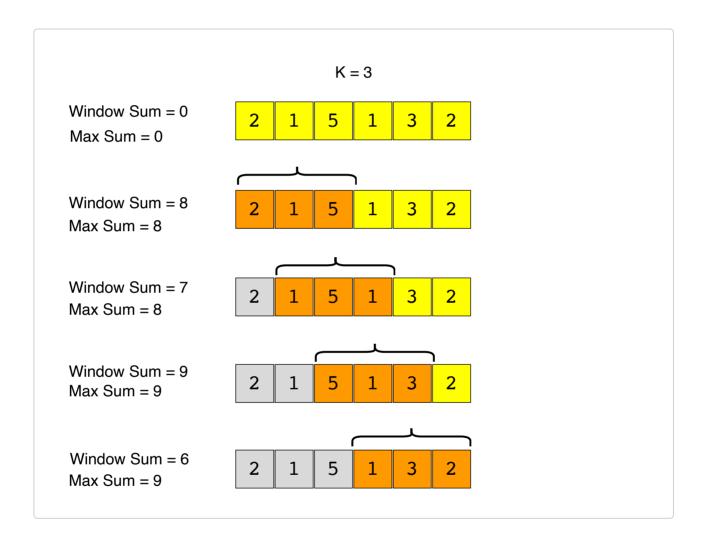


Solution#





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 subarray's sum. Following is the visual representation of this algorithm for Example-1:



Code#

Here is what our algorithm will look like:

```
Java

Python3

C++

Js

JS

1 def max_sub_array_of_size_k(k, arr):

2 max_sum = 0

3 window_sum = 0

4
```

```
5
      for i in range(len(arr) - k + 1):
                                                          (%)
 6
        window sum = 0
 7
        for j in range(i, i+k):
          window_sum += arr[j]
 9
        max sum = max(max sum, window sum)
10
      return max_sum
11
12
13
    def main():
14
      print("Maximum sum of a subarray of size K: " + str(max_sub_array_of_siz
      print("Maximum sum of a subarray of size K: " + str(max sub array of siz
15
16
17
18
    main()
19
                                                         X
                                                                   0.88s
Output
Maximum sum of a subarray of size K: 9
 Maximum sum of a subarray of size K: 7
```

The above algorithm's time complexity 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:

= 155 Voeract the element going out of the sliding window, i.e., subtract the

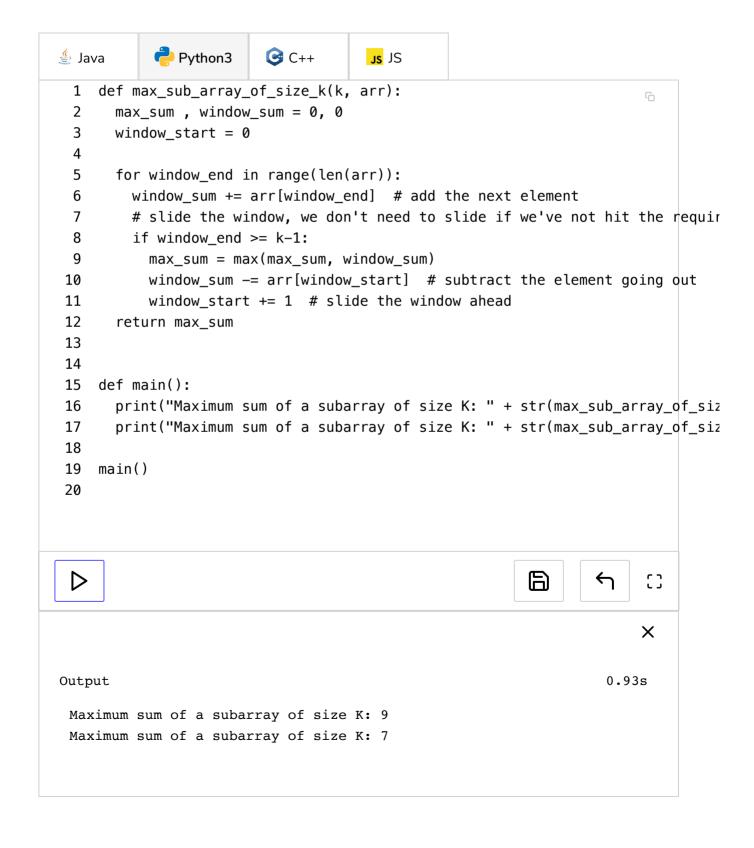
first element of the window.





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 ${\cal O}(N)$.

Space Complexity#

The algorithm runs in constant space O(1).

