

# Sliding Window Technique - GeeksforGeeks

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Courses Tutorials Practice Jobs DSA Tutorial Interview Questions Quizzes Must Do Advanced DSA System Design Aptitude Puzzles Interview Corner DSA Python Technical Scripter 2026 Explore DSA Fundamentals Logic Building Problems Analysis of Algorithms Data Structures Array Data Structure String in Data Structure Hashing in Data Structure Linked List Data Structure Stack Data Structure Queue Data Structure Tree Data Structure Graph Data Structure Trie Data Structure Algorithms Searching Algorithms Sorting Algorithms Introduction to Recursion Greedy Algorithms Tutorial Graph Algorithms Dynamic Programming or DP Bitwise Algorithms Advanced Segment Tree Binary Indexed Tree or Fenwick Tree Square Root (Sqrt) Decomposition Algorithm Binary Lifting Geometry Interview Preparation Interview Corner GfG160 Practice Problem GeeksforGeeks Practice - Leading Online Coding Platform Problem of The Day - Develop the Habit of Coding DSA Course 90% Refund Sliding Window Technique Last Updated : 2 Sep, 2025 Sliding Window Technique is a method used to solve problems that involve subarray or substring or window. Instead of repeatedly iterating over the same elements, the sliding window maintains a range (or "window") that moves step-by-step through the data, updating results incrementally. The main idea is to use the results of previous window to do computations for the next window. Commonly used for problems like finding subarrays with a specific sum, finding the longest substring with unique characters, or solving problems that require a fixed-size window to process elements efficiently. Table of Content Example Problem - Maximum Sum of a Subarray with K Elements Naive Approach - O(n<sup>k</sup>) Time and O(1) Space Using the Sliding Window Technique - O(n) Time and O(1) Space How to use Sliding Window Technique? How to Identify Sliding Window Problems? More Example Problems Example Problem - Maximum Sum of a Subarray with K Elements Given an array arr[] and an integer k, we need to calculate the maximum sum of a subarray having size exactly k . Input : arr[] = [5, 2, -1, 0, 3], k = 3 Output : 6 Explanation : We get maximum sum by considering the subarray [5, 2, -1] Input : arr[] = [1, 4, 2, 10, 23, 3, 1, 0, 20], k = 4 Output : 39 Explanation : We get maximum sum by adding subarray [4, 2, 10, 23] of size 4. Try it on GfG Practice Naive Approach - O(n<sup>k</sup>) Time and O(1) Space C++ #include <iostream> #include <vector> #include <climits> using namespace std ; int maxSum ( vector < int >& arr , int k ) { int n = arr . size () ; int max\_sum = INT\_MIN ; // Consider all blocks starting with i for ( int i = 0 ; i <= n - k ; i ++ ) { int current\_sum = 0 ; // Calculate sum of current subarray of size k for ( int j = 0 ; j < k ; j ++ ) current\_sum += arr [ i + j ] ; // Update result if required max\_sum = max ( current\_sum , max\_sum ) ; } return max\_sum ; } int main () { vector < int > arr = { 5 , 2 , -1 , 0 , 3 } ; int k = 3 ; cout << maxSum ( arr , k ) << endl ; return 0 ; } C #include <limits.h> #include <math.h> #include <stdio.h> // Find maximum between two numbers. int max ( int num1 , int num2 ){ return ( num1 > num2 ) ? num1 : num2 ; } int maxSum ( int arr [] , int n , int k ){ // Initialize result int max\_sum = INT\_MIN ; // Consider all blocks starting with i. for ( int i = 0 ; i < n - k + 1 ; i ++ ) { int current\_sum = 0 ; for ( int j = 0 ; j < k ; j ++ ) current\_sum = current\_sum + arr [ i + j ] ; // Update result if required. max\_sum = max ( current\_sum , max\_sum ) ; } return max\_sum ; } // Driver code int main (){ int arr [] = { 5 , 2 , -1 , 0 , 3 } ; int k = 3 ; int n = sizeof ( arr ) / sizeof ( arr [ 0 ] ) ; printf ( "%d" , maxSum ( arr , n , k )) ; return 0 ; } Java class GFG { // Returns maximum sum in // a subarray of size k. static int maxSum ( int arr [] , int n , int k ){ // Initialize result int max\_sum = Integer . MIN\_VALUE ; // Consider all blocks starting with i. for ( int i = 0 ; i < n - k + 1 ; i ++ ) { int current\_sum = 0 ; for ( int j = 0 ; j < k ; j ++ ) current\_sum = current\_sum + arr [ i + j ] ; // Update result if required. max\_sum = Math . max ( current\_sum , max\_sum ) ; } return max\_sum ; } public static void main ( String [] args ){ int arr [] = { 5 , 2 , -1 , 0 , 3 } ; int k = 3 ; int n = arr . length ; System . out . println ( maxSum ( arr , n , k )) ; } } Python import sys def maxSum ( arr , n , k ): # Initialize result max\_sum = - sys . maxsize # Consider all blocks starting with i. for i in range ( n - k + 1 ): current\_sum = 0 for j in range ( k ): current\_sum += arr [ i + j ] # Update result if required. max\_sum = max ( current\_sum , max\_sum ) return max\_sum if \_\_name\_\_ == "\_\_main\_\_" : arr = [ 5 , 2 , -1 , 0 , 3 ] k = 3 n = len ( arr ) print ( maxSum ( arr , n , k )) C# using System ; class GFG { // Returns maximum sum in a // subarray of size k. static int maxSum ( int [] arr , int n , int k ){ // Initialize result int max\_sum = int . MinValue ; // Consider all blocks starting // with i. for ( int i = 0 ; i < n - k + 1 ; i ++ ) { int current\_sum = 0 ; for ( int j = 0 ; j < k ; j ++ ) current\_sum = current\_sum + arr [ i + j ] ; // Update result if required. max\_sum = Math . Max ( current\_sum , max\_sum ) ; } return max\_sum ; } public static void Main (){ int [] arr = { 5 , 2 , -1 , 0 , 3 } ; int k = 3 ; int n = arr . Length ; int maxSum = - int . MaxValue ; for ( int i = 0 ; i < n - k + 1 ; i ++ ) { int currentSum = 0 ; for ( int j = 0 ; j < k ; j ++ ) currentSum += arr [ i + j ] ; maxSum = Math . Max ( currentSum , maxSum ) ; } Console . WriteLine ( maxSum ) ; } }

, 3 }; int k = 3 ; int n = arr . Length ; Console . WriteLine ( maxSum ( arr , n , k )); } } JavaScript function maxSum ( arr , n , k ) { // Initialize result let max\_sum = Number . MIN\_SAFE\_INTEGER ; // Consider all blocks starting with i for ( let i = 0 ; i < n - k + 1 ; i ++ ) { let current\_sum = 0 ; for ( let j = 0 ; j < k ; j ++ ) { current\_sum += arr [ i + j ]; } // Update result if required max\_sum = Math . max ( current\_sum , max\_sum ); } return max\_sum ; } // Driver code const arr = [ 5 , 2 , -1 , 0 , 3 ]; const k = 3 ; const n = arr . length ; console . log ( maxSum ( arr , n , k )); Output 6 Using the Sliding Window Technique - O(n) Time and O(1) Space We compute the sum of the first k elements out of n terms using a linear loop and store the sum in variable window\_sum . Then we will traverse linearly over the array till it reaches the end and simultaneously keep track of the maximum sum. To get the current sum of a block of k elements just subtract the first element from the previous block and add the last element of the current block. The below representation will make it clear how the window slides over the array. Consider an array arr[] = {5, 2, -1, 0, 3} and value of k = 3 and n = 5 This is the initial phase where we have calculated the initial window sum starting from index 0 . At this stage the window sum is 6. Now, we set the maximum\_sum as current\_window i.e 6. Now, we slide our window by a unit index. Therefore, now it discards 5 from the window and adds 0 to the window. Hence, we will get our new window sum by subtracting 5 and then adding 0 to it. So, our window sum now becomes 1. Now, we will compare this window sum with the maximum\_sum. As it is smaller, we won't change the maximum\_sum. Similarly, now once again we slide our window by a unit index and obtain the new window sum to be 2. Again we check if this current window sum is greater than the maximum\_sum till now. Once, again it is smaller so we don't change the maximum\_sum. Therefore, for the above array our maximum\_sum is 6. C++ #include <iostream> #include <vector> using namespace std ; int maxSum ( vector < int >& arr , int k ){ int n = arr . size (); // n must be greater if ( n <= k ) { cout << "Invalid" ; return -1 ; } // Compute sum of first window of size k int max\_sum = 0 ; for ( int i = 0 ; i < k ; i ++ ) max\_sum += arr [ i ]; // Compute sums of remaining windows by // removing first element of previous // window and adding last element of // current window. int window\_sum = max\_sum ; for ( int i = k ; i < n ; i ++ ) { window\_sum += arr [ i ] - arr [ i - k ]; max\_sum = max ( max\_sum , window\_sum ); } return max\_sum ; } int main (){ vector < int > arr = { 5 , 2 , -1 , 0 , 3 }; int k = 3 ; cout << maxSum ( arr , k ); return 0 ; } Java class GFG { static int maxSum ( int arr [] , int n , int k ){ // n must be greater if ( n <= k ) { System . out . println ( "Invalid" ); return -1 ; } // Compute sum of first window of size k int max\_sum = 0 ; for ( int i = 0 ; i < k ; i ++ ) max\_sum += arr [ i ]; // Compute sums of remaining windows by // removing first element of previous // window and adding last element of // current window. int window\_sum = max\_sum ; for ( int i = k ; i < n ; i ++ ) { window\_sum += arr [ i ] - arr [ i - k ]; max\_sum = Math . max ( max\_sum , window\_sum ); } return max\_sum ; } public static void main ( String [] args ){ int arr [] = { 5 , 2 , -1 , 0 , 3 }; int k = 3 ; int n = arr . length ; System . out . println ( maxSum ( arr , n , k )); } } Python def maxSum ( arr , k ): # length of the array n = len ( arr ) # n must be greater than k if n <= k : print ( "Invalid" ) return -1 # Compute sum of first window of size k window\_sum = sum ( arr [: k ]) # first sum available max\_sum = window\_sum # Compute the sums of remaining windows by # removing first element of previous # window and adding last element of # the current window. for i in range ( n - k ): window\_sum = window\_sum - arr [ i ] + arr [ i + k ] max\_sum = max ( window\_sum , max\_sum ) return max\_sum if \_\_name\_\_ == "\_\_main\_\_" : arr = [ 5 , 2 , -1 , 0 , 3 ] k = 3 print ( maxSum ( arr , k )) C# using System ; class GFG { static int maxSum ( int [] arr , int n , int k ){ // n must be greater if ( n <= k ) { Console . WriteLine ( "Invalid" ); return -1 ; } // Compute sum of first window of size k int max\_sum = 0 ; for ( int i = 0 ; i < k ; i ++ ) max\_sum += arr [ i ]; // Compute sums of remaining windows by // removing first element of previous // window and adding last element of // current window. int window\_sum = max\_sum ; for ( int i = k ; i < n ; i ++ ) { window\_sum += arr [ i ] - arr [ i - k ]; max\_sum = Math . Max ( max\_sum , window\_sum ); } return max\_sum ; } public static void Main (){ int [] arr = { 5 , 2 , -1 , 0 , 3 }; int k = 3 ; int n = arr . Length ; Console . WriteLine ( maxSum ( arr , n , k )); } } JavaScript function maxSum ( arr , k ){ const n = arr . length ; if ( n < k ) { console . log ( "Invalid" ); return -1 ; } // Compute sum of first window of size k let windowSum = 0 ; for ( let i = 0 ; i < k ; i ++ ) { windowSum += arr [ i ]; } let maxSum = windowSum ; // Slide the window from start to end of the array for ( let i = k ; i < n ; i ++ ) { windowSum += arr [ i ] - arr [ i - k ]; maxSum = Math . max ( maxSum , windowSum ); } return maxSum ; } // Driver Code const arr = [ 5 , 2 , -1 , 0 , 3 ]; const k = 3 ; console . log ( maxSum ( arr , k )); Output 6 How to use Sliding Window Technique? There are basically two types of sliding window: 1. Fixed Size Sliding Window: The general steps to solve these questions by following below steps: Find the size of the window required, say K. Compute the result for 1st window, i.e. include the first K elements of the data structure. Then use a loop to slide the window by 1 and keep computing the result window by window. 2. Variable Size Sliding Window: The general steps to solve these questions by following below steps:

In this type of sliding window problem, we increase our right pointer one by one till our condition is true. At any step if our condition does not match, we shrink the size of our window by increasing left pointer. Again, when our condition satisfies, we start increasing the right pointer and follow step 1. We follow these steps until we reach to the end of the array. How to Identify Sliding Window Problems? These problems generally require Finding Maximum/Minimum Subarray , Substrings which satisfy some specific condition. The size of the subarray or substring 'k' will be given in some of the problems. These problems can easily be solved in  $O(n^2)$  time complexity using nested loops, using sliding window we can solve these in  $O(n)$  Time Complexity. Required Time Complexity:  $O(n)$  or  $O(n \log n)$   
Constraints:  $n \leq 10$  6 More Example Problems Top Problems on Sliding Window Technique Practice Questions on Sliding Window Comment Article Tags: Article Tags: DSA sliding-window