658. Find K Closest Elements

Two Pointers

Given a sorted array of integers arr, and two integers k and x, the task is to find the k closest integers to x in the array. The result should be returned in ascending order. To determine which integers are the closest, we follow these rules:

Sliding Window

Heap (Priority Queue)

2. If the absolute differences are the same, then a is closer to x than b if a is smaller than b.

1. An integer a is considered closer to x than an integer b if the absolute difference |a - x| is less than |b - x|.

Sorting

In essence, the problem asks us to find a subsequence of the array that contains integers that are nearest to x, with a special

Binary Search

emphasis on the absolute difference and the value of the integers themselves when the differences are the same.

To solve the problem, we need an efficient way to find the subsequence of length k out of the sorted array that is closest to the

target integer x.

Intuition

Medium Array

Problem Description

Method 1: Sort A brute force approach would be to sort the elements of the array based on their distance from x. After <u>sorting</u>, we can simply take the first k elements. However, this method is not optimal in terms of time complexity because sorting would take O(n

log n) time.

Method 2: Binary search A more efficient approach utilizes the fact that the array is already sorted. We know that the k closest integers we are looking for must form a consecutive subsequence in the array. Instead of looking for each integer separately, we

should look for the starting index of this subsequence. We can use a binary search strategy to find this starting index. The high-level

steps are:
1. Set two pointers: left at the start of the array and right at the position len(arr) - k because any subsequence starting beyond this point does not have enough elements to be k long.
2. Perform a binary search between left and right:

- Check if x is closer to arr[mid] or to arr[mid + k].
 If x is closer to arr[mid] or at the same distance to both, we discard the elements to the right of mid + k because the best
- starting point for our subsequence must be to the left or inclusive of mid.
 - starting point for our subseque

Calculate the middle index mid.

If x is closer to arr[mid + k], we discard the elements to the left of mid since the best starting point must be to the right.
 Keep narrowing down until left is equal to right, indicating that we have found the starting index of the k closest elements.

by-step breakdown of the approach and the code implementation:

between x and the potential upper bound for our sequence at arr[mid + k].

- Solution Approach
- The solution implements a binary search algorithm to efficiently find the starting index for the k closest elements to x. Here's a step-

1. We begin by initializing two pointers, left at 0 and right at len(arr) - k. This is because the k elements we seek must form consecutive entrants in the array, and starting any further to the right would leave insufficient elements to reach a count of k.

pointer up to mid + 1.

2. We enter a loop that continues until left is no longer less than right, indicating that we have narrowed down to the exact starting index for the k closest elements.

3. Inside the loop, we find the middle index between left and right using the formula mid = (left + right) >> 1. The >> 1 is a bitwise right shift that effectively divides the sum by 2.

4. With mid established, we check the distance between x and the current middle value at arr[mid] compared to the distance

- If the element at mid is closer to x (or the elements are equally distant), we know that the k closest elements can't start any index higher than mid. Thus we set right to mid, moving our search space leftward.
- 5. By repeatedly halving our search space, we eventually arrive at a point where left equals right. This index represents the start of the subarray containing the k closest numbers to x.

6. Once we have the starting index (left), we slice the original array from left to left + k to select the k closest elements.

Using this binary search approach, we achieve a time complexity of O(log(n - k)) for finding the subsequence, which is an

Conversely, if the element at mid + k is closer to x, it means the sequence starts further to the right, so we move our left

- 7. This slice is then returned, and since the original array was sorted, this resulting slice is guaranteed to be sorted as well, yielding the correct answer.
- where the difference in time complexity could be substantial.

optimization over a complete sort-based method that would require O(n log n) time. This optimization is significant for large datasets

Example

Assume we have an array arr = [1, 3, 4, 7, 8, 9], and we want to find k = 3 closest integers to x = 5.

• We begin by initializing two pointers, left is set to 0 and right is set to len(arr) - k, which in this case is 6 - 3 = 3. The

Example Walkthrough

subarray [7, 8, 9] is the last possible sequence of k numbers, and we will not consider starting indices beyond this.

• Since 7 is closer to 5 than 1 is, we move the left pointer to mid + 1, which is 1.

Initialize the left and right pointers for binary search.

Calculate the middle index between left and right.

public List<Integer> findClosestElements(int[] arr, int k, int x) {

// Initialize the left and right pointers for binary search.

// we need to move towards the left (lower indices).

// Otherwise, move right (higher indices).

// Continue searching until the search space is reduced to a single element.

// Add the closest elements to the list, starting from the left pointer.

// If the distance to the left is less than or equal to the distance to the right,

left, right = 0, len(arr) - k

mid = (left + right) // 2

right = mid

if x - arr[mid] <= arr[mid + k] - x:</pre>

while left < right:</pre>

The right pointer is set to the highest starting index for the sliding window.

Perform binary search to find the left bound of the k closest elements.

Let us walk through a simple example to illustrate the solution approach.

• The binary search begins. We calculate the middle index between left and right. The initial mid is (0 + 3) / 2 = 1.5, rounded

Binary Search

down to 1.

Initial Setup

= 2 and from 8 is |5 - 8| = 3.
• Since 3 is closer to 5 than 8 is, we set right to mid, which is now 1. Our search space for starting indices is now [0, 1].

Narrowing Down

We perform another iteration of the binary search. Our new mid is (0 + 1) / 2 = 0.5, rounded down to 0.
Again, we compare distances from x = 5: arr[mid] = 1 has a distance of 4, and arr[mid + k] = arr[3] = 7 has a distance of 2.

• We then compare the distance of x = 5 from arr[mid] = 3 and from arr[mid + k] = arr[4] = 8. The distance from 3 is |5 - 3|

Convergence
 Because left now equals right, we have converged to the starting index for the k closest elements. Our starting index is 1.

• We slice the original array from left to left + k, which yields [3, 4, 7].

Conclusion

Obtaining the Result

to find the starting index and the subsequent slice of the array. This method efficiently finds the correct subsequence without fully sorting the array based on proximity to x.

From our example using the array arr = [1, 3, 4, 7, 8, 9] and the values k = 3 and x = 5, we have walked through the binary

search approach and determined that the k closest integers to x in this array are [3, 4, 7], as illustrated by the binary search logic

1 from typing import List 2 3 class Solution: 4 def findClosestElements(self, arr: List[int], k: int, x: int) -> List[int]:

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Python Solution

1 import java.util.List;

class Solution {

import java.util.ArrayList;

int left = 0;

int right = arr.length - k;

right = mid;

result.add(arr[i]);

left = mid + 1;

// Calculate the middle index.

if $(x - arr[mid] \le arr[mid + k] - x) {$

// Create a list to store the k closest elements.

* Finds the k closest elements to a given target x in a sorted array.

* @param {number} x - The target number to find the closest elements to.

function findClosestElements(arr: number[], k: number, x: number): number[] {

// then the closer elements are to the left side of the array.

if (x - arr[midPointer] <= arr[midPointer + k] - x) {</pre>

* @returns {number[]} - An array of k closest elements to the target.

* @param {number[]} arr - The sorted array of numbers.

// Initialize two pointers for binary search.

rightPointer = midPointer;

contiguous elements as a potential answer).

leftPointer = midPointer + 1;

* @param {number} k - The number of closest elements to find.

List<Integer> result = new ArrayList<>();

for (int i = left; i < left + k; ++i) {</pre>

// Return the list of k closest elements.

while (left < right) {</pre>

} else {

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left = mid + 1

2  # Extract the subarray from the left index of size k, which will be the k closest elements.

2  return arr[left:left + k]

25  # Example usage:
26  # sol = Solution()

27  # result = sol.findClosestElements([1, 2, 3, 4, 5], k=4, x=3)

28  # print(result) # Output should be [1, 2, 3, 4]

Java Solution
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int mid = left + (right - left) / 2; // Using left + (right - left) / 2 to avoid potential overflow.

Check the distance from the x to the middle element and the element at mid + k position.

we move the right pointer to mid. Otherwise, we adjust the left pointer to mid + 1.

If the element at mid is closer to x or equal in distance compared to the element at mid + k,

33 return result; 34 } 35 } 36

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C++ Solution
   #include <vector>
 3 class Solution {
   public:
       std::vector<int> findClosestElements(std::vector<int>& arr, int k, int x) {
           // Initialize the binary search bounds
           int left = 0;
           int right = arr.size() - k;
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           // Perform binary search to find the start index of the k closest elements
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           while (left < right) {</pre>
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               // Calculate mid index (avoid potential overflow by using left + (right-left)/2)
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               int mid = left + (right - left) / 2;
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               // Compare the differences between x and elements at mid index and mid+k index
               // The goal is to find the smallest window such that the elements are closest to x
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               if (x - arr[mid] \le arr[mid + k] - x) {
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                   // If the element at mid index is closer to x, or equally close
                   // as the element at mid+k index (prefer the smaller element),
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                   // move the right bound to mid
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                   right = mid;
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               } else {
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                   // Otherwise, if the element at mid+k index is closer to x,
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                   // move the left bound to mid + 1
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                   left = mid + 1;
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           // Create and return a vector of k closest elements starting from the 'left' index
           return std::vector<int>(arr.begin() + left, arr.begin() + left + k);
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32 };
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Typescript Solution
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11 let leftPointer = 0; 12 let rightPointer = arr.length - k; 13 14 // Binary search to find the start index of the k closest elements. 15 while (leftPointer < rightPointer) {</pre>

} else {

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// Slice the array from the left pointer to get the k closest elements.
return arr.slice(leftPointer, leftPointer + k);

Time and Space Complexity

The provided algorithm is a binary search approach for finding the k closest elements to x in the sorted array arr. Here's the analysis:

• Time Complexity: The time complexity of this algorithm is O(log(N - k)) where N is the number of elements in the array. This is because the binary search is performed within a range that's reduced by k elements (since we're always considering k
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const midPointer = (leftPointer + rightPointer) >> 1; // Equivalent to Math.floor((left + right) / 2)

// If the element at the middle is closer to or as close as the element k away from it,

// Otherwise, they are to the right, and we move the left pointer one step past the middle.

Hence, the iterative process of the binary search dominates the running time, which leads to the O(log(N - k)) time complexity.

• Space Complexity: The space complexity of this algorithm is O(1). The code uses a fixed amount of extra space for variables

through all N elements, but narrows down the search to the correct starting point for the sequence of k closest elements. After

The algorithm performs a binary search by repeatedly halving the search space, which is initially N - k. It does not iterate

finding the starting point, it returns the subarray in constant time, as array slicing in Python is done in 0(1) time.

the same elements in the original array, thus no extra space proportional to k or N is used in this operation.

left, right, and mid. The returned result does not count towards space complexity as it is part of the output.

Please note that array slicing in the return statement does not create a deep copy of the elements but instead returns a reference to