373. Find K Pairs with Smallest Sums

Medium (Array) (Heap (Priority Queue))

can infer that the smallest sums will involve the smallest elements from both arrays.

we are trying to create pairs with the smallest possible sum to begin with.

Leetcode Link

Problem Description

sorted in non-decreasing order. Our task is to find k pairs, each composed of one element from each array. The objective is to identify the pairs with the smallest possible sums. Specifically, we want to find the k pairs whose sums are the smallest among all possible combinations from nums1 and nums2.

If we were to list all possible pairs and their sums, the list would be quite extensive. However, since the arrays are already sorted, we

In this problem, we are working with two key concepts: pairs and sum minimization. We have two arrays, nums1 and nums2, both

Intuition

The solution employs a min-heap, a binary tree where the parent node is always less than or equal to its children. This is an ideal

first element of nums2.

Here's the step-by-step process to understand the solution:

The algorithm initializes by creating a min-heap and pre-filling it with pairs combining each of the first k elements of nums 1 with the

1. Initialize the min-heap q. We start by forming pairs by taking each element from nums1 (up to the first k elements to avoid unnecessary work) and pairing it with the first element in nums2. The reason for pairing with the first element from nums2 is that

Heapify q to ensure the smallest sum pair is at the top of the heap.
 Prepare a list ans to store the resulting k smallest sum pairs.

Pop the smallest sum pair from the heap. This pair is guaranteed to be one of the smallest sum pairs because of the min-

- heap's properties.
- Append this pair to the ans list.

so that it can be considered in the next iteration.

call then converts the list q into a heap data structure.

pairs or there are no more new pairs to consider.

smallest pairs and a list for our result.

Consider the following input:

Following the steps of the solution approach:

data structure for efficiently finding the smallest elements.

Decrease k by 1, as we have successfully found one of the k smallest sum pairs.
 Check if there's a next element in nums2 that can be paired with the same nums1 element. If so, add this new pair to the heap

4. Use a loop to process the min-heap. While the min-heap is not empty and k is greater than 0, we do the following:

5. Return the ans list containing up to k smallest sum pairs.

By utilizing a min-heap, we ensure that at each step, we pop the minimum sum pair without having to check all combinations.

Moreover, since nums1 and nums2 are sorted, we can confidently move to the next element in nums2 for potential pairing without

worrying about missing smaller sums.

Solution Approach

The solution we're discussing uses a min-heap to efficiently find the k pairs with the smallest sums. Let's walk through the implementation step by step, taking advantage of algorithms, data structures, and patterns.

1. Min-Heap Initialization and Heapify: The code snippet q = [[u + nums2[0], i, 0] for i, u in enumerate(nums1[:k])]

nums2[0]), the index of the nums1 element (i), and the index 0 indicating we've used the first element of nums2. The heapify(q)

2. Processing the Min-Heap: The while loop continues as long as there are elements in the heap (i.e., potential pairs to consider)

initializes our min-heap q. For each element in nums1 (limited to the first k elements), we create a list with the sum (u +

and k is greater than 0. In each iteration, we use heappop(q) to remove and return the smallest element from the heap, which is the pair with the current smallest sum.

3. Recording the Result: The current smallest sum pair's indices in nums1 and nums2 are stored in the _, i, j variables. We then append this pair [nums1[i], nums2[j]] to our answer list ans.

element from nums1 and the next element in nums2, then push it into the heap using heappush(q, [nums1[i] + nums2[j + 1], i, j + 1]) to make sure it can be considered in subsequent iterations.
5. Finding the k Smallest Sums: The heap keeps track of the current smallest pairs, and the loop ensures that each time we pop from the heap and push a new pair into it, we're considering the next smallest possible pair. This continues until we've found k

The key algorithms and data structures used here are the heap data structure (a min-heap in particular) for keeping track of the

The pattern is utilizing a greedy-like approach: in each step, we're always taking the smallest available option (the pair with the

smallest sum) and then exploring slightly larger options by considering the next possible element from nums 2.

Let's walk through a small example using the provided solution approach to understand how it works.

4. Generating New Pairs: After finding a pair with a small sum, we check if there is a next element in nums 2 to pair with the same

nums1 element. The if j + 1 < len(nums2) check is used for this purpose, and if true, we form a new pair with the same

This approach results in a time-efficient solution because we avoid generating all possible pairs, which would have resulted in a much higher time complexity.

Example Walkthrough

nums1 = [1, 7, 11]
nums2 = [2, 4, 6]
k = 3

1. Min-Heap Initialization and Heapify: We initialize the min-heap with elements paired as follows: (nums1[i] + nums2[0], i, 0). Since we're only interested in the first k pairs, and k is 3, we consider only the first three elements from nums1. Therefore, our

min-heap q starts with the following elements (assuming we include all three elements of nums1 and k limits other factors):

2. Processing the Min-Heap: We start processing the heap by popping elements while k > 0. The smallest sum pair is at the top,

4. Generating New Pairs: Since nums1[0] was paired with nums2[0], we now pair nums1[0] with the next element in nums2, which is

After heapification, q ensures that the pair with the smallest sum is at the top of the heap.

1 q = [

3 (7 + 2, 1, 0),

4 (11 + 2, 2, 0)

process until k becomes 0.

Python Solution

class Solution:

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Java Solution

class Solution {

class Solution {

};

public:

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};

1 from typing import List

so we pop (3, 0, 0) first.

2 (1 + 4, 0, 1), # sum = 5, new pair

3. Recording the Result: We extract the indices 0 from nums1 and 0 from nums2 and append the pair [1, 2] to our result list ans.

Now, ans = [[1, 2]].

nums2[1] (value 4), and add this new pair to the heap. Now, the heap q looks like:

Heapification ensures the smallest pair stays at the top, which is now (5, 0, 1).

2 (1 + 2, 0, 0), # sum = 3, index in nums1 = 0, index in nums2 = 0

3 (7 + 2, 1, 0), # sum = 9, index in nums1 = 1, index in nums2 = 0

4 (11 + 2, 2, 0), # sum = 13, index in nums1 = 2, index in nums2 = 0

- 5. Finding the k Smallest Sums: We continue the process, now popping (5, 0, 1) from the heap and appending [1, 4] to ans. Subsequently, we form a new pair with nums1[0] and nums2[2], and the heap q gets updated after heapification. We repeat this
- By following this strategy, we avoid generating all possible pairs and efficiently find the k smallest sums using a min-heap to guide our pair selections.

Only consider the first k numbers in nums1 for initial pairing, as we are looking for the k smallest pairs

After these steps, our result list ans is [[1, 2], [1, 4], [1, 6]], which are the k smallest sum pairs from nums1 and nums2.

def k_smallest_pairs(self, nums1: List[int], nums2: List[int], k: int) -> List[List[int]]:

priority_queue = [[u + nums2[0], i, 0] for i, u in enumerate(nums1[:k])]

Iterate until the priority queue is empty or we have found k pairs

public List<List<Integer>> kSmallestPairs(int[] nums1, int[] nums2, int k) {

PriorityQueue<int[]> queue = new PriorityQueue<>((a, b) -> a[0] - b[0]);

heapify(priority_queue) # Convert list into a heap

Pop the smallest sum pair from the heap

return result # Return the list of k smallest pairs

for (int i = 0; i < Math.min(nums1.length, k); ++i) {</pre>

// Function to find k smallest pairs with minimal sum

// Get the sizes of the input arrays

// Variable to store the final pairs

for (int i = 0; i < min(k, size1); ++i) {</pre>

auto [index1, index2] = minHeap.top();

// Add the current smallest pair to the result

minHeap.emplace(index1, index2 + 1);

return nums1[a[0]] + nums2[a[1]] > nums1[b[0]] + nums2[b[1]];

result.push_back({nums1[index1], nums2[index2]});

int size1 = nums1.size();

int size2 = nums2.size();

vector<vector<int>> result;

minHeap.emplace(i, 0);

// Extract the k smallest pairs

if (index2 + 1 < size2) {</pre>

// Return all k smallest pairs found

minHeap.pop();

return result;

// Importing required modules

type IndexPair = [number, number];

// Defining the type for a pair of indices

while (k-- > 0 && !minHeap.empty()) {

vector<vector<int>> kSmallestPairs(vector<int>& nums1, vector<int>& nums2, int k) {

// Priority queue to keep pairs in ascending order based on their sum

// Initialize the priority queue with the first pair from each element in numsl

// If there's a next element in nums2, add the new pair to the priority queue

import { PriorityQueue } from 'typescript-collections'; // Placeholder import for PQ functionality

// Function to compare pairs based on the sum of elements they point to in nums1 and nums2

function comparePairs(nums1: number[], nums2: number[], a: IndexPair, b: IndexPair): boolean {

// Lambda function to compare pairs based on the sum of elements they point to in nums1 and nums2

priority_queue<pair<int, int>, vector<pair<int, int>>, decltype(comparePairs)> minHeap(comparePairs);

auto comparePairs = [&nums1, &nums2](const pair<int, int>& a, const pair<int, int>& b) {

return nums1[a.first] + nums2[a.second] > nums1[b.first] + nums2[b.second];

queue.offer(new int[] {nums1[i] + nums2[0], i, 0});

result = [] # Initialize a list to hold the result pairs

sum_pair, index1, index2 = heappop(priority_queue)

Create a priority queue to hold the sum of pairs along with the indices in nums1 and nums2

Append the corresponding values from nums1 and nums2 to the result
result.append([nums1[index1], nums2[index2]])

k -= 1 # Decrement k as we have found one of the k smallest pairs

k -= 1 # Decrement k as we have found one of the k smallest pairs

// Create a priority queue to hold the arrays with a comparator to prioritize by the sum of pairs.

// Initialize the priority queue with the first k pairs from nums1 and the first element from nums2.

```
k -= 1 # Decrement k as we have found one of the k smallest pairs

# If there are more elements in nums2 to pair with the current nums1 element, push the next pair onto the heap
if index2 + 1 < len(nums2):
    heappush(priority_queue, [nums1[index1] + nums2[index2 + 1], index1, index2 + 1])</pre>
```

from heapq import heapify, heappop, heappush

while priority_queue and k > 0:

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// Prepare a list to store the k smallest pairs.
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           List<List<Integer>> result = new ArrayList<>();
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14
           // Keep polling from the priority queue to find the next smallest pair
15
           while (!queue.isEmpty() && k > 0) {
               // Poll the smallest sum pair from the priority queue.
16
               int[] currentPair = queue.poll();
17
19
               // Add the new pair [nums1[index1], nums2[index2]] to the result list.
                result.add(Arrays.asList(nums1[currentPair[1]], nums2[currentPair[2]]));
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               // Decrease the remaining pairs count.
23
               --k;
24
25
               // If there's a next element in nums2, offer the next pair from nums1 and nums2 into the priority queue.
26
               if (currentPair[2] + 1 < nums2.length) {</pre>
27
                    queue.offer(new int[] {nums1[currentPair[1]] + nums2[currentPair[2] + 1], currentPair[1], currentPair[2] + 1});
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29
30
31
           // Return the list of k smallest pairs.
32
           return result;
33
34 }
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C++ Solution
    #include <vector>
  2 #include <queue>
     using namespace std;
```

Typescript Solution

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12 // Function to find k smallest pairs with minimal sum
13 function kSmallestPairs(nums1: number[], nums2: number[], k: number): number[][] {
     // Get the sizes of the input arrays
15
     const size1 = nums1.length;
     const size2 = nums2.length;
16
17
18
     // Variable to store the final pairs
19
     const result: number[][] = [];
20
21
     // Priority queue to keep pairs in ascending order based on their sum
22
     const minHeap = new PriorityQueue<IndexPair>((a, b) => comparePairs(nums1, nums2, a, b));
23
24
     // Initialize the priority queue with the first pair from each element in numsl
25
     for (let i = 0; i < Math.min(k, size1); i++) {</pre>
26
       minHeap.enqueue([i, 0]);
27
28
29
     // Extract the k smallest pairs
     while (k > 0 && !minHeap.isEmpty()) {
30
       const [index1, index2] = minHeap.dequeue()!;
31
32
33
       // Add the current smallest pair to the result
       result.push([nums1[index1], nums2[index2]]);
34
35
36
       // If there's a next element in nums2, add the new pair to the priority queue
37
       if (index2 + 1 < size2) {</pre>
38
         minHeap.enqueue([index1, index2 + 1]);
39
40
41
       k--;
42
43
44
     // Return all k smallest pairs found
     return result;
45
46 }
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Time and Space Complexity
The given Python code implements a heap to find the k smallest pairs of sums from two integer arrays nums1 and nums2. Here we will
```

discuss its time complexity and space complexity. Time Complexity

The time complexity of the algorithm is as follows:

1. **Heap Initialization**: The code creates a min-heap and initializes it with the sums of elements from nums1 and the first element of nums2. Since the heap is initialized with at most k elements (and not more than the length of nums1), the complexity of this step is 0(k) since each heap insertion is 0(log k) and we do it at most k times.

2. Heap Operations: Then, in each iteration, the algorithm pops an element from the heap and potentially pushes a new element

Overall, the time complexity is $0(k) + 0(k \log k)$, which simplifies to $0(k \log k)$ because as k grows, the k $\log k$ term dominates.

into the heap. Since we perform k iterations (bounded by k and the length of the output), and both heappop and heappush operations have a time complexity of $0(\log k)$, the complexity for this part is $0(k \log k)$.

Space Complexity

The space complexity of the algorithm is as follows:

- 1. Heap Space: The heap size is at most k, as it stores pairs of indices and their corresponding sum, giving us O(k).
- The overall space complexity combines the heap space and the output list, but since both are 0(k), the total space complexity remains 0(k).

2. Output List: The list ans to store the answer pairs. In the worst case, it will contain k pairs, leading to O(k) space complexity.