2208. Minimum Operations to Halve Array Sum

Array Heap (Priority Queue) Medium Greedy

Problem Description

The goal of this problem is to reduce the sum of a given array nums of positive integers to at least half of its original sum through a series of operations. In each operation, you can select any number from the array and reduce it exactly to half of its value. You are allowed to select reduced numbers in subsequent operations.

Leetcode Link

Intuition

The objective is to determine the minimum number of such operations needed to achieve the goal of halving the array's sum.

To solve this problem, a greedy approach is best, as taking the largest number and halving it will most significantly impact the sum in

the shortest number of steps. Using a max heap data structure is suitable for efficiently retrieving and halving the largest number at each step. The solution involves:

1. Calculating the target sum, which is half the sum of the array nums. 2. Creating a max heap to access the largest element quickly in each operation.

- 3. Continuously extracting the largest element from the heap, halving it, and adding the halved value back to the heap.
- 4. Accumulating the count of operations until the target sum is reached or passed.
- This approach guarantees that each step is optimal in terms of reducing the total sum and reaching the goal using the fewest

Solution Approach The solution uses the following steps, algorithms, and data structures:

largest element in the array.

- value back on to the heap. Since we're using a min heap, every value is negated when it's pushed onto the heap and negated again when it's popped off to maintain the original sign.
- 4. Counting Operations: A variable ans is used to count the number of operations. It is incremented by one with each halving operation.
- original array has been reduced by at least half. In terms of algorithm complexity, this solution operates in O(n log n) time where n is the number of elements in nums. The log n
- The code implementation based on these steps is shown in the reference solution with proper comments to highlight each step: 1 class Solution:

while s > 0: # Continue until we've halved the array sum t = -heappop(h) / 2 # Extract and halve the largest value s -= t # Reduce the target sum by the halved value 10

```
heappush(h, -t) # Push the negated halved value back onto the heap
11
              ans += 1 # Increment operation count
12
          return ans # Return the total number of operations needed
13
Example Walkthrough
Let's walk through an example to illustrate the solution approach.
Suppose we have the following array of numbers: nums = [10, 20, 30].
 1. Calculating the Target Sum:
```

We now perform operations which will involve halving the largest element and keeping a tally of operations.

The current sum we need to track is 30.

Extract the largest element (-30) and halve its value (15).

Push the negated halved value back onto the heap (-15).

The heap is now h = [-20, -15, -10].

The heap is now h = [-15, -10, -10].

Increment the operation count (ans = 2).

Subtract 10 from the target sum, leaving us with 5.

Subtract 15 from the target sum, leaving us with 15. Increment the operation count (ans = 1).

Second Operation:

Third Operation:

Python Solution

class Solution:

from heapq import heappush, heappop

for value in nums:

target_sum = sum(nums) / 2

3. Reducing Sum with Operations:

First Operation:

Extract the largest element (-15) and halve its value (7.5). Push the negated halved value back onto the heap (-7.5). The heap is now h = [-10, -10, -7.5]. Subtract 7.5 from the target sum, which would make it negative (-2.5). Increment the operation count (ans = 3).

Initialize a max heap (using negative values because Python has a min heap by default) $max_heap = []$ # Add negative values of nums to the heap to simulate max heap

def halveArray(self, nums: List[int]) -> int:

heappush(max_heap, -value)

heappush(max_heap, -largest)

Increment the operation count

14 15 16

13

20

21

22

24

25

26

while target_sum > 0: # Retrieve and negate the largest element, then halve it largest = -heappop(max_heap) / 2 # Subtract the halved value from the target sum

30 31 Java Solution class Solution { public int halveArray(int[] nums) { // Initial sum of the array elements. double sum = 0; // Priority queue to store the array elements in descending order. PriorityQueue<Double> maxHeap = new PriorityQueue<>(Collections.reverseOrder()); // Add all elements to the priority queue and calculate the total sum. 9 10 for (int value : nums) { maxHeap.offer((double) value); 11 12 sum += value; 13 14 // The target is to reduce the sum to less than half of its original value.

return operations; 35 36 } 37

31

32

33

34

23

24

25

26

27

28

29

30

33

28

35

36

37

39

40

41

43

47

42 }

```
using namespace std;
   class Solution {
   public:
       int halveArray(vector<int>& nums) {
           // Use a max heap to keep track of the largest numbers in the array
           priority_queue<double> maxHeap;
10
11
           double total = 0; // Original total sum of the array elements
            for (int value : nums) {
13
                total += value; // Accumulate total sum
14
15
               maxHeap.push(value); // Add current value to the max heap
16
17
18
           double targetHalf = total / 2.0; // Our target is to reduce the total to this value or less
            int operations = 0; // Initialize the number of operations performed to 0
20
21
           // Continue reducing the total sum until it's less than or equal to targetHalf
22
           while (total > targetHalf) {
```

maxHeap.pop(); // Remove this largest number from max heap

total -= topValue; // Subtract the halved value from the total sum

maxHeap.push(topValue); // Push the halved value back into max heap

double topValue = maxHeap.top() / 2.0; // Halve the largest number in max heap

```
function halveArray(nums: number[]): number {
       // Calculate the target sum which is half the sum of the input array
       let targetSum: number = nums.reduce((accumulator, currentValue) => accumulator + currentValue) / 2;
10
       // Initialize a max priority queue to facilitate the retrieval of the largest element
11
       const maxPriorityQueue = new MaxPriorityQueue<number>();
12
13
14
       // Enqueue all numbers in the array into the max priority queue with their values as priorities
       for (const value of nums) {
15
16
           maxPriorityQueue.enqueue(value, value);
17
18
       // Initialize the operation counter
19
       let operationCount: number = 0;
21
22
       // Continue until the remaining sum is reduced to targetSum or less
23
       while (targetSum > 0) {
           // Dequeue the largest element
24
           let dequeuedItem = maxPriorityQueue.dequeue().value;
25
26
27
           // Halve the dequeued element
```

```
Time Complexity
```

Time and Space Complexity

2. Heap construction: Inserting all elements into a heap has an overall complexity of 0(n * log(n)) as each insertion operation into the heap is $O(\log(n))$, and it is performed n times for n elements.

3. Halving elements until sum is reduced: The complexity of this part depends on the number of operations we need to reduce the sum by half. In the worst-case scenario, every element is divided multiple times. For each halving operation, we remove the

The given algorithm consists of multiple steps that contribute to the overall time complexity:

needed.

1. Sum calculation: The sum of the array is calculated with a complexity of O(n), where n is the number of elements in nums.

- Sum computation: O(n) Heapification: O(n * log(n)) Halving operations: 0(m * log(n)) Since the number of halving operations m is not necessarily linear and depends on the values in the array, we cannot directly relate it
- **Space Complexity**
- Hence, the space complexity is O(n).

operations possible.

1. Max Heap (Priority Queue): The Python code uses a max heap, which is implemented using a min heap with negated values (the heapq library in Python only provides a min heap). A max heap allows us to easily and efficiently access and remove the

2. Halving the Largest Element: At each step, the code pops the largest value in the heap, halves it, and pushes the negated half

- 3. Sum Reduction Tracking: We keep track of the current sum of nums we need to halve, starting with half of the original sum of nums. After halving and pushing the largest element back onto the heap, we decrement this sum by the halved value.
- 5. Condition Check: The loop continues until the sum we are tracking is less than or equal to zero, meaning the total sum of the
- def halveArray(self, nums: List[int]) -> int: s = sum(nums) / 2 # Target sum to achieve h = [] # Initial empty heap for v in nums: heappush(h, -v) # Negate and push all values to the heap ans = 0 # Operation counter
- The sum of nums is 60, so halving the sum gives us a target of 30. 2. Creating a Max Heap: We create a max heap with the negated values of nums: h = [-30, -20, -10].

factor comes from the push and pop operations of the heap, which occur for each of the n elements.

- Extract the largest element (-20) and halve its value (10). Push the negated halved value back onto the heap (-10).
- Since the sum we are tracking is now less than 0, the loop ends. 4. Result: The minimum number of operations required to reduce the sum of nums to at least half its original sum is 3.

Calculate the sum of half the elements to determine the target

- # Counter for the number of operations performed operations = 0
 - # Keep reducing the target sum until it reaches 0 or below target_sum -= largest # Push the halved negative value back to maintain the max heap
- 27 operations += 1 28 29 # Return the total number of operations needed to reach the target sum return operations
- double halfSum = sum / 2.0; 16 17 // Counter for the number of operations performed. 18 19 int operations = 0; 20 // Continue until we reduce the sum to less than half. 22 while (halfSum > 0) { // Retrieve and remove the largest element from the queue. double largest = maxHeap.poll(); 24 25 // Divide it by 2 (halving the element) and subtract from halfSum. 26 halfSum -= largest / 2.0; // Add the halved element back to the priority queue. 27 28 maxHeap.offer(largest / 2.0); 29 // Increment the operation counter. 30 operations++;
- C++ Solution 1 #include <vector> 2 #include <queue> // Function to find the minimum number of operations to reduce array sum to less than or equal to half of the initial sum.

// Return the number of operations required to achieve the target.

- operations++; // Increment the number of operations return operations; // Return the total number of operations performed 31 32 }; Typescript Solution // Importing PriorityQueue to use in the implementation import { MaxPriorityQueue } from 'typescript-collections'; // This function takes an array of numbers and returns the minimum number of // operations to reduce the sum of the array to less than or equal to half its original sum by performing // operations that halve the value of any element in the array.
- 29 30 // Subtract the halved value from the remaining sum targetSum -= dequeuedItem; 31 33 // Re-enqueue the halved element to ensure correct ordering in the priority queue 34 maxPriorityQueue.enqueue(dequeuedItem, dequeuedItem);

dequeuedItem /= 2;

operationCount += 1;

// const result = halveArray([10, 20, 7]);

return operationCount;

// Example usage:

// Increment the operation counter

// Return the total number of operations performed

// console.log(result); // Outputs the number of operations needed

- - maximum element $(0(\log(n)))$ complexity for removal) and insert it back into the heap $(0(\log(n)))$ complexity for insertion). The number of such operations could vary, but it could potentially be 0(m * log(n)), where m is the number of halving operations

Therefore, the time complexity of the algorithm is determined by summing these complexities:

to n. As a result, the overall worst-case time complexity of the algorithm is O(n * log(n) + m * log(n)).

The space complexity of the algorithm is determined by:

1. **Heap storage**: We store all n elements in the heap, which requires 0(n) space. 2. Auxiliary space: Aside from the heap, the algorithm uses a constant amount of extra space for variables like s, t, and ans.