2263. Make Array Non-decreasing or Non-increasing Greedy **Dynamic Programming** Hard

Problem Description

of operations needed to transform nums into a non-decreasing (every element is greater than or equal to the previous one) or nonincreasing (every element is less than or equal to the previous one) sequence. An operation consists of selecting any element in the array and either incrementing it by 1 or decrementing it by 1.

In this LeetCode problem, we are given an array nums of integers with a 0-based index. The goal is to determine the minimum number

Leetcode Link

element multiple times if required.

Intuition

increasing orders. We will calculate the minimum operations required for both scenarios and then return the smaller result.

When looking for the minimum number of operations, it's important to keep in mind that we can use an operation on the same

For both cases, the approach is to use dynamic programming (DP) to efficiently calculate the minimum number of operations. The

To do this, the solution iterates over each element x of the input array and updates the DP table to reflect the minimum number of

operations considering both the possibility of incrementing and decrementing. Specifically, during the update process for each number x, it considers the possible outcomes for having each value from 0 to 1000 at position i. For each of these values, it finds the minimum operations from the previous step and the current cost of making x equal

Finally, the solution iterates over the values of the last element (the last row of the DP table), collecting the minimum number of operations required to achieve both non-decreasing and non-increasing order, respectively. The result is the minimum of these two

values. In summary, the intuition is to explore all possible target values for each element separately for both non-decreasing and non-

increasing cases, combine them smartly using dynamic programming to keep track of the minimum cost at each step, and finally

Solution Approach The solution uses dynamic programming as its core approach, which is a method usually used to solve problems by breaking them down into smaller and simpler sub-problems, solving each sub-problem just once, and saving their solutions - typically using a

order with an array ending in value j at index i. The size of f is (n+1) x 1001, where n is the length of nums and 1001 accommodates

• For each element x, we iterate from 0 to 1000 to consider all possible values j that the current element could take. mi stores the

minimum operations needed to get to the previous element's state, stored in mi. This dynamic programming step ensures that

• Iterate through each element x in nums, starting at index 1 for the array f since index 0 is used for the base case where no

Dynamic Programming.

Data Structures:

for each element, f stores the accumulated minimum operations to adjust the sequence to non-decreasing/non-increasing up to that element. After updating f[i][j] for all values from 0 to 1000, it retains the minimum value reached for the entire sequence by taking the

• For every possible value j, it calculates the number of operations required to adjust x to j (abs(x - j)) and adds it to the

minimum number of operations seen so far for transforming the sequence up to the previous element.

Algorithm Used:

Post the solve function, the main part of the convertArray method applies this process twice: once for nums as is, and once for the

• A 2D list f to serve as the DP table.

Example Walkthrough

the determination of the minimum number of operations needed dynamically as the array is processed.

Let's take a small example to illustrate the solution approach with the given array nums:

considering all possible values that each element can take post operations (0 to 1000).

1 nums = [3, 2, 5, 1, 6]

Converting to Non-Decreasing:

3. Continue filling f using the same approach for elements 5, 1, and 6.

We repeat the same process, but before starting, we reverse the array:

ending with value j. The answer will be the minimum value in f[5][0..1000].

0 to 1000. For example, to make 3 a 0, we would need f[1][0] = 3 operations. 2. Element 2: Now, for each target value j from 0 to 1000, we calculate the minimum of f[1][0..1000] and add the cost to convert 2 to j, updating f[2][j] accordingly.

At the end of this process, f[5][j] contains the minimum operations needed to convert the array into a non-decreasing sequence

1. Element 3: The initial minimum operations matrix f at index 1 will be set considering the cost of making 3 all possible values from

- 1. Element 6: Similar to the non-decreasing process, f[1][0..1000] is set according to the cost to change 6 into all possible ending values. 2. Element 1: Just as in the non-decreasing case, we iterate over all possible target values j, and for each, add the minimum of
- After this, the f[5][j] row reflects the minimum operations required to make the original array non-increasing. The final step is to take the minimum value from f[5] [0..1000].

Finally, the solution will return the smaller value between the two minimums obtained for both the non-decreasing and non-

increasing scenarios. This represents the minimum number of operations required to convert the initial array into either a non-

min_value = float('inf') # Initialize the minimum value to be infinity for j in range(1001): 14 # Update the minimum value for the previous row if min_value > min_ops[i - 1][j]: 16 17 min_value = min_ops[i - 1][j] # Compute the minimum operations for current element to be j 18

min_ops[i][j] = min_value + abs(num - j)

// Find the minimum of the original array and the reversed array

int[][] dp = new int[n + 1][1001]; // DP table to store minimal operations

int minPrevious = Integer.MAX_VALUE; // Initialize to max value

// Calculate min operations for getting value j at index i

minPrevious = Math.min(minPrevious, dp[i - 1][j]);

// Function to find the minimum cost to make all elements of the array equal

// Create a 2D array to store subproblem solutions, where f[i][j] will

minCostPrevious = min(minCostPrevious, f[i - 1][j]);

// with the sum of minCostPrevious and current cost

f[i][j] = minCostPrevious + abs(nums[i - 1] - j);

int minCostPrevious = INT_MAX; // Initialize the minimum cost to a large value

// Update minCostPrevious to the minimum cost found in previous row

// Calculate cost of making nums[i-1] equal to j and update f[i][j]

// Return the minimum cost of making all elements of the array equal until the last element

// Helper function to calculate the minimum cost of making all elements equal by incrementing or decrementing

// Reversing the nums vector to calculate cost from both ends

// Return the minimum cost of the two possible scenarios

// Helper function to calculate the minimum cost of making all

// Initialize the 2D array with zero using memset

// hold minimum cost to make the first i elements equal to j

// Method to compute the minimum operations to satisfy the condition.

// Update minimum from the previous row

Find and return the minimum value from the last row

Initialize a 2D array to store the minimum number of operations

needed to make all elements equal to any value between 0 and 1000

for i, num in enumerate(arr, 1): # Enumerate from 1 for 1-based indexing

Call the solve function on both the original and reversed array, and return the minimum result

return Math.min(computeMinimumOperations(nums), computeMinimumOperations(reverseArray(nums)));

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                     dp[i][j] = minPrevious + Math.abs(j - nums[i - 1]);
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             // Find the minimal operations among all possibilities for the last element
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             int answer = Integer.MAX_VALUE; // Initialize to max value
 26
             for (int cost : dp[n]) {
 27
                 answer = Math.min(answer, cost); // Update the answer with the minimal cost
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             return answer; // Return the final minimal cost
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         // Helper method to reverse the given array.
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         private int[] reverseArray(int[] nums) {
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             int[] copiedNums = nums.clone(); // Clone the array to avoid modifying the original one
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             // Reverse the array in place
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             for (int start = 0, end = copiedNums.length - 1; start < end; ++start, --end) {</pre>
 39
                 // Swap elements
 40
                 int temp = copiedNums[start];
 41
                 copiedNums[start] = copiedNums[end];
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                 copiedNums[end] = temp;
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             return copiedNums; // Return the reversed array
 46
 47 }
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C++ Solution
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12 let cost2: number = calculateMinimumCost(nums); 13 14 // Return the minimum cost of the two possible scenarios 15 return Math.min(cost1, cost2); 16 } 17

Typescript Solution

2 const MAX_VALUE: number = 1000;

nums.reverse();

41 // Return the minimum cost of making all elements of the array equal until the last element 42 43 return Math.min(...dp[n]); 44 45 // Example usage: 47 // let nums = [1, 2, 3]48 // let cost = convertArray(nums) // console.log(cost); 50 Time and Space Complexity Time Complexity The given function convertArray includes a nested function solve which computes the minimum cost to make all elements of the array equal by only increments or only decrements. The outer function convertArray calls solve twice, once for the original array and once for the reversed array. Within solve, there is a nested loop. The outer loop runs n times, where n is the length of nums.

of convertArray is 0(2 * n * 1001) which simplifies to 0(n) because constants are dropped in Big O notation and 1001 is a constant. **Space Complexity**

No other significant data structures are used that scale with the input size.

The space complexity of the solve function is 0((n + 1) * 1001) which simplifies to 0(n), again dropping the constant 1001.

Thus, the overall space complexity of convertArray is O(n) as only one instance of the 2D array f is maintained throughout the calls to solve.

The intuition behind the solution is to consider separately the scenarios where we convert the array to non-decreasing and non-

The solve function in the code defines the process for transforming the array to either non-decreasing or non-increasing order by considering one direction at a time. The function sets up a 2D array f to store the minimum operations needed to achieve the desired

memory-based data structure (array, map, etc.).

Let's walk through the main steps of the solve function:

operations have been applied (f[0] is initialized with zeros).

minimum of the last row of f, essentially, min(f[n]).

take the minimum of the costs from each of these two scenarios.

all possible values an element can take after the operations (assuming the limit as 1000).

reversed nums. Reversing nums effectively solves for a non-increasing sequence, as the reversed-non-decreasing sequence is the original non-increasing sequence. The final answer is the minimum number of operations between the non-decreasing and non-increasing transformations.

Patterns: Separating the problem into two scenarios (non-decreasing and non-increasing) and taking the minimum of the two results. Iterative approach for dynamic programming with memory optimization properties.

The clever part of this algorithm is its ability to handle all possible end values for each subarray in an efficient manner, allowing for

We want to transform this array into either a non-decreasing or non-increasing sequence using the minimum number of operations.

We loop over each element in nums starting from index 1 in f as index 0 represents the base case where the array is empty (no operations needed).

Initialize 2D array f with size (n+1) x 1001, where n is the length of nums. In our case, n is 5, so we have f of size 6 x 1001. We're

Converting to Non-Increasing:

Summary:

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27 # Usage

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Java Solution

class Solution {

1 class Solution {

int convertArray(vector<int>& nums) {

return min(cost1, cost2);

int n = nums.size();

int f[n + 1][1001];

memset(f, 0, sizeof(f));

for (int i = 1; $i \le n$; ++i) {

int cost1 = calculateMinimumCost(nums);

// Cost calculated after reversing array

// elements equal by incrementing or decrementing

for (int j = 0; $j \le 1000$; ++j) {

return *min_element(f[n], f[n] + 1001);

1 // Global variable to define the maximum value for calculation

let cost1: number = calculateMinimumCost(nums);

function calculateMinimumCost(nums: number[]): number {

function convertArray(nums: number[]): number {

// Cost calculated after reversing array

let n: number = nums.length;

// Function to find the minimum cost to make all elements of the array equal

// Create a 2D array to store subproblem solutions, where dp[i][j] will

// Initialize the subproblem solutions for base cases here if necessary

// with the sum of minCostPrevious and current cost

dp[i][j] = minCostPrevious + Math.abs(nums[i - 1] - j);

let dp: number[][] = Array.from({length: n + 1}, () => Array(MAX_VALUE + 1).fill(0));

minCostPrevious = Math.min(minCostPrevious, dp[i - 1][j]);

// Calculate cost of making nums[i-1] equal to j and update dp[i][j]

// hold minimum cost to make the first i elements equal to j

// Reversing the nums array to calculate cost from both ends

int calculateMinimumCost(vector<int>& nums) {

int cost2 = calculateMinimumCost(nums);

reverse(nums.begin(), nums.end());

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44 };

1 Reversed nums = [6, 1, 5, 2, 3]

f[1] [0..1000] and the cost to convert 1 to j into f[2][j]. 3. Proceed with each element (5, 2, 3) in the reversed array filling f accordingly.

class Solution: def convertArray(self, nums: List[int]) -> int: # Helper function to solve the problem for both original and reversed arrays def solve(arr):

return min(min_ops[n])

// Main method to convert the array.

// Populate the dp table

for (int i = 1; $i \le n$; ++i) {

public int convertArray(int[] nums) {

private int computeMinimumOperations(int[] nums) {

int n = nums.length; // Length of the array

for (int j = 0; $j \le 1000$; ++j) {

print(sol.convertArray([1, 5, 3, 3]))

return min(solve(nums), solve(nums[::-1]))

n = len(arr) # Length of the array

 $min_{ops} = [[0] * 1001 for _ in range(n + 1)]$

decreasing or a non-increasing sequence.

Python Solution

from typing import List

27 28 for (let i: number = 1; i <= n; ++i) {</pre> 29 let minCostPrevious: number = Number.MAX_SAFE_INTEGER; // Initialize the minimum cost to a large value 30 for (let j: number = 0; j <= MAX_VALUE; ++j) {</pre> 31 // Update minCostPrevious to the minimum cost found in previous row **if** (i > 1) {

• The inner loop runs for a constant 1001 times since it iterates over a preset range from 0 to 1000. • Inside the inner loop, the computation is done in constant time, namely the minimum comparison mi > f[i - 1][j], the minimum update mi = f[i - 1][j], and the cost calculation f[i][j] = mi + abs(x - j).

The space complexity is determined by the amount of memory used by the function relative to the input size: • The function solve uses a 2D array f of size (n + 1) * 1001, where n is the length of the input array nums.

Putting it together, the time complexity of the solve function is 0(n * 1001), and since it is called twice, the overall time complexity

basic idea of the DP solution is to create a 2D array f where each entry f[i][j] represents the minimum number of operations to convert the first i elements of the array into a sequence that ends with the number j.

to the considered value. The cost is simply the absolute difference between x and the considered value.