494. Target Sum

Dynamic Programming Backtracking

Medium

Problem Description

adding either a '+' (plus sign) or a '-' (minus sign) before each number in nums, then concatenate (join together) these numbers with their signs to form an expression. The end goal is to find out how many different expressions you can create that, once evaluated, equal the target.

For instance, if nums is [2, 1] and your target is 1, you could create the expression "+2-1", which equals 1. You want to find out

You're given a list of numbers called nums and a single number called target. Your task is to form a mathematical expression by

all such possible expressions that result in the target.

Here's what you need to consider:

• You can't reorder the numbers – their order in the expression must match their order in the given nums list.

• The objective is to count the number of valid expressions, not necessarily to generate each one.

You can only use '+' or '-' before each number.

- Intuition

where the difference between the sums of the subsets equals target.

except it allows for both positive and negative summations. The crux is to figure out how to partition nums into two subsets

First step is to notice a helpful fact: if we sum all numbers with a '+' and then subtract the sum of numbers with a '-', we should obtain the target. This is basically the same as finding two subsets with a particular sum difference.

Now, how do we solve this with dynamic programming? We can use a technique similar to the 0-1 Knapsack problem. The idea is

Understanding the problem, we see that it resembles a classic problem in computer science known as the subset sum problem,

to iteratively build up an array dp where each index represents a possible sum of numbers, starting from 0 up to some value that depends on nums and target. Each cell in dp will tell us the number of ways to achieve that sum.

But how do we compute the size of dp and its initial values? We define a new variable n which is the sum that we want one subset to have so that the other subset has a sum of n + target (given that the total sum of nums is s). It turns out n should be

(s - target) / 2. We only proceed if s - target is even, otherwise, it's not possible to split nums into two such subsets.

Once we have dp setup starting at dp[0] = 1 (there's one way to achieve a sum of 0: by choosing no numbers), we start populating dp by iterating through each number in nums. For each v in nums, we iterate backwards through dp from n down to v (since v is the smallest sum that including v can achieve) and update dp[j] to include the number of ways we can achieve the sum j - v.

In the end, <code>dp[n]</code> gives us the number of different expressions we can form that equal <code>target</code>.

The code provided implements this <code>dynamic programming</code> approach efficiently.

Solution Approach

The provided solution uses a <u>dynamic programming</u> approach to solve the problem efficiently, much like the "0-1 Knapsack"

problem. The solution involves some pre-processing steps and then iteratively filling out a dp array to count the number of ways to reach different sums.

• If the total sum s is less than target, it's not possible to create any expression equal to target.

1. Calculate the sum s of all numbers in nums. This is done to determine the sum of one subset (n) so that the difference with

Here's a step-by-step breakdown of the implementation:

the other subset's sum will be target.

Our dp array is [1, 0, 0, 0, 0] initially.

2. Check for two conditions that might make it impossible to create expressions equal to target:

(by choosing no numbers).

4. Iterate over each number v in nums. For each v, iterate backwards over the dp array from n down to v to avoid recomputing

Initialize the dp array of size n + 1 with zeros and set dp[0] to 1 - this signifies that there is one way to create a sum of zero

Update the dp[j] value by adding the number of combinations that exist without the current number v (dp[j - v]). This

step accumulates the number of ways there are to achieve a sum of \mathbf{j} by either including or excluding the current number \mathbf{v} .

values that rely on the current index. This is crucial because we must not count any combination twice.

∘ If s - target is odd, we can't split the array into two subsets with integer sums that have the needed difference.

to equal the target.

Example of dp Array Update

By the end of the iteration, dp[-1] (which is dp[n]) will hold the number of possible expressions that can be created using nums

During the iteration process:
When v = 1, we update dp to [1, 1, 0, 0, 0].
When v = 2, we update dp for j from 2 to 1, resulting in [1, 1, 1, 0, 0].

 \circ When v = 3, we update dp for j from 2 to 1, but this time there are no changes as 3 is too large to affect dp[1] or dp[2].

Suppose nums = [1,2,3] and target = 1. We compute n = (s - target) / 2, which would be 2 in this case.

The method described balances the need to consider both including and excluding each number in nums while ensuring each sum is only counted once. It utilizes memory efficiently by only maintaining a one-dimensional array rather than a full matrix that

even, so we proceed.

The dp array is now [1, 1, 0].

The dp array is now [1, 1, 1].

Solution Implementation

total_sum = sum(nums)

return 0

dp[0] = 1

from typing import List

Python

class Solution:

For v = 2, update dp[j] from j = 2 to 2:

to the target without redundantly computing combinations.

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

Compute the subset sum we need to find to partition the array

into two subsets that give the desired target on applying + and - operations

There is always 1 way to achieve a sum of 0, which is by selecting no elements

For each number in nums, update the count of ways to achieve each sum <= subset_sum

Add the number of ways to achieve a sum of j before num was considered

Return the number of ways to achieve subset sum, which indirectly gives us the number of ways

// If the sum is less than the target or (sum - target) is odd, it's not possible to partition

Initialize a list for dynamic programming, with a size of subset_sum + 1

Calculate the sum of all numbers in the array

subset_sum = (total_sum - target) // 2

Update the dynamic programming table

dp[j] += dp[j - num]

for i in range(subset sum, num - 1, -1):

public int findTargetSumWays(int[] nums, int target) {

if (sum < target || (sum - target) % 2 != 0) {</pre>

// Iterate over every number in the input array

for (int j = newTarget; j >= num; --j) {

dp[j] += dp[j - num];

* Calculates the number of wavs to assign '+' and '-'

* @param {number[]} nums - Array of numbers to be used.

// Calculate the sum of the input array elements

if (sum < target || (sum - target) % 2 !== 0) {</pre>

* @param {number} target - The target sum to be achieved.

* @return {number} - Number of ways to achieve the target sum.

let sum: number = nums.reduce((acc, val) => acc + val, 0);

const findTargetSumWays = (nums: number[], target: number): number => {

* to make the sum of nums be equal to target.

// For each number, iterate backwards from the new target to the number's value

// Update the dp array to count additional ways to reach current sum (i)

// by adding the number of ways to reach the sum without the current number (j - num)

// This is to ensure we do not count any subset more than once

// Return the total number of ways to reach the 'newTarget', which corresponds

// to the number of ways to reach the original 'target' considering +/- signs

// If the sum is less than the target, or the difference is odd, there is no solution

waysToSum[0] = 1; // Base case - there's one way to have a sum of zero (using no elements)

for (int num : nums) {

return dp[newTarget];

// Initialize the sum of all numbers in nums

to achieve the target sum using '+' and '-' operations

 $dp = [0] * (subset_sum + 1)$

 \circ dp[2] gets updated to dp[2] + dp[2 - 2] which is dp[2] = 0 + 1 = 1.

6.

Example Walkthrough

calculate n = (s - target) / 2, which is (6 - 2) / 2 = 2.

would be required in a naive implementation of dynamic programming for this problem.

In the end, dp[n] gives us the count of the number of expressions that sum up to target.

Here's a hypothetical example to illustrate the dynamic programming update process:

Let's walk through a small example to illustrate the provided solution approach. Suppose we have nums = [1, 2, 3] and target = 2.

Now we initialize our dp array with zeros and set dp[0] to 1. So initially, our dp array is [1, 0, 0].

Next, we iterate over each number v in nums. For each v, we update the dp array from n down to v:
For v = 1, update dp[j] from j = 2 to 1. The dp array changes as follows:
dp[1] gets updated to dp[1] + dp[1 - 1] which is dp[1] = 0 + 1 = 1.

According to the solution approach, our first step is to calculate the sum s of all numbers in nums. Here, the sum is 1 + 2 + 3 =

Then we check if the total sum s is less than target or if s - target is odd. In this case, 6 is not less than 2, and 6 - 2 is

Our next step is to determine the size of the dp array, which will help us count the number of ways to make up different sums. We

Finally, our dp array is [1, 1, 1], and the dp[-1] or dp[2] indicates that there is 1 possible expression that can be created using nums to sum up to target, which is "1 + 2 - 3".

This walkthrough demonstrates how the dynamic programming approach effectively counts the number of expressions equating

For v = 3, since our dp array size is only 3, we don't need to update it for v = 3 because 3 is larger than n.

Check if the target is achievable or not # If the sum of nums is less than target, or the difference between sum and target # is not an even number, then return 0 because target can't be achieved if total sum < target or (total_sum - target) % 2 != 0:

```
# Example usage:
# solution = Solution()
# result = solution.findTargetSumWays([1, 1, 1, 1, 1], 3)
# print(result) # Outputs: 5
```

int sum = 0;

for (int num : nums) {

sum += num;

return 0;

Java

class Solution {

return dp[-1]

for num in nums:

```
// Compute the subset sum needed for one side of the partition
        int subsetSum = (sum - target) / 2;
        // Initialize a DP array to store the number of ways to reach a particular sum
        int[] dp = new int[subsetSum + 1];
       // There's one way to reach the sum of 0 — by not including any numbers
       dp[0] = 1;
        // Go through every number in nums
        for (int num : nums) {
            // Update the DP table from the end to the start to avoid overcounting
            for (int i = subsetSum; i >= num; i--) {
                // Increase the current dp value by the value from dp[j - num]
                dp[j] += dp[j - num];
        // Return the number of ways to reach the target sum
        return dp[subsetSum];
C++
#include <vector>
#include <numeric> // For using accumulate function
class Solution {
public:
    int findTargetSumWays(vector<int>& nums, int target) {
        // Calculate the sum of all numbers in the vector
        int sum = std::accumulate(nums.begin(), nums.end(), 0);
       // If the sum is less than the target or the adjusted sum is odd, return 0
        if (sum < target || (sum - target) % 2 != 0) return 0;</pre>
        // Calculate the new target (n) which is the sum to be found
        int newTarget = (sum - target) / 2;
        // Initialize the dynamic programming array with zeros
        // and set dp[0] to 1 since there's one way to get sum 0: using no numbers
        vector<int> dp(newTarget + 1, 0);
        dp[0] = 1;
```

```
const totalNums: number = nums.length;
// Calculate the subset sum that we need to find
const subsetSum: number = (sum - target) / 2;
// Initialize a DP array for storing number of ways to sum up to j with array elements
let waysToSum: number[] = new Array(subsetSum + 1).fill(0);
```

return 0;

// Fill the DP array

return dp[-1]

print(result) # Outputs: 5

Time and Space Complexity

result = solution.findTargetSumWays([1, 1, 1, 1, 1], 3)

Example usage:

solution = Solution()

};

/**

TypeScript

```
for (let i = 1; i <= totalNums; ++i) {</pre>
        for (let i = subsetSum; i >= nums[i - 1]; --i) {
            waysToSum[j] += waysToSum[j - nums[i - 1]]; // Update the ways to sum to j
   // Return the total ways to achieve the subset sum, which is equivalent to the target
   return waysToSum[subsetSum];
};
export { findTargetSumWays }; // Exporting the function to be used in other modules
from typing import List
class Solution:
   def findTargetSumWays(self, nums: List[int], target: int) -> int:
       # Calculate the sum of all numbers in the array
        total_sum = sum(nums)
       # Check if the target is achievable or not
       # If the sum of nums is less than target, or the difference between sum and target
       # is not an even number, then return 0 because target can't be achieved
        if total sum < target or (total_sum - target) % 2 != 0:</pre>
            return 0
       # Compute the subset sum we need to find to partition the array
       # into two subsets that give the desired target on applying + and - operations
        subset_sum = (total_sum - target) // 2
       # Initialize a list for dynamic programming, with a size of subset_sum + 1
       dp = [0] * (subset_sum + 1)
       # There is always 1 way to achieve a sum of 0, which is by selecting no elements
       dp[0] = 1
       # Update the dynamic programming table
       # For each number in nums, update the count of ways to achieve each sum <= subset_sum
        for num in nums:
            for j in range(subset sum, num - 1, -1):
                # Add the number of ways to achieve a sum of j before num was considered
                dp[j] += dp[j - num]
       # Return the number of ways to achieve subset sum, which indirectly gives us the number of ways
        # to achieve the target sum using '+' and '-' operations
```

n is the computed value (sum(nums) - target) // 2. This is because the algorithm consists of a nested loop, where the outer loop runs for each element in nums, and the inner loop runs from n down to the value of the current element v.

The space complexity of the algorithm is O(n + 1), which simplifies to O(n), as there is a one-dimensional list dp of size n + 1 elements used to store the intermediate results for the subsets that sum up to each possible value from O(n) to O(n).

The time complexity of the algorithm is O(len(nums) * n), where len(nums) is the number of elements in the input list nums, and