

494. Target Sum

Medium Array Dynamic Programming Backtracking

[Leetcode Link](#)

Problem Description

You're given a list of numbers called `nums` and a single number called `target`. Your task is to form a mathematical expression by 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 all such possible expressions that result in the `target`.

Here's what you need to consider:

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

Intuition

Understanding the problem, we see that it resembles a classic problem in computer science known as the subset sum problem, except it allows for both positive and negative summations. The crux is to figure out how to partition `nums` into two subsets where the difference between the sums of the subsets equals `target`.

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 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, `dp[n]` gives us the number of different expressions we can form that equal `target`.

The code provided implements this dynamic programming approach efficiently.

Solution Approach

The provided solution uses a dynamic programming 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.

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

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 the other subset's sum will be `target`.
2. Check for two conditions that might make it impossible to create expressions equal to `target`:
 - If the total sum `s` is less than `target`, it's not possible to create any expression equal to `target`.
 - If `s - target` is odd, we can't split the array into two subsets with integer sums that have the needed difference.
3. 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 (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 values that rely on the current index. This is crucial because we must not count any combination twice.
5. 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 `j` by either including or excluding the current number `v`.

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` to equal the `target`.

Example of dp Array Update

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

- Suppose `nums` = `[1,2,3]` and `target` = 1. We compute `n = (s - target) / 2`, which would be 2 in this case.
- Our `dp` array is `[1, 0, 0, 0, 0]` initially.
- 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]`.
 - 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]`.

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

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

Example Walkthrough

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

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 = 6`.

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 even, so we proceed.

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

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`.
 - The `dp` array is now `[1, 1, 0]`.
- For `v = 2`, update `dp[j]` from `j = 2` to 2:
 - `dp[2]` gets updated to `dp[2] + dp[2 - 2]` which is `dp[2] = 0 + 1 = 1`.
 - The `dp` array is now `[1, 1, 1]`.
- 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`.

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 to the target without redundantly computing combinations.

Python Solution

```
1 from typing import List
2
3 class Solution:
4     def findTargetSumWays(self, nums: List[int], target: int) -> int:
5         # Calculate the sum of all numbers in the array
6         total_sum = sum(nums)
7
8         # Check if the target is achievable or not
9         # If the sum of nums is less than target, or the difference between sum and target
10        # is not an even number, then return 0 because target can't be achieved
11        if total_sum < target or (total_sum - target) % 2 != 0:
12            return 0
13
14        # Compute the subset sum we need to find to partition the array
15        # into two subsets that give the desired target on applying + and - operations
16        subset_sum = (total_sum - target) // 2
17
18        # Initialize a list for dynamic programming, with a size of subset_sum + 1
19        dp = [0] * (subset_sum + 1)
20
21        # There is always 1 way to achieve a sum of 0, which is by selecting no elements
22        dp[0] = 1
23
24        # Update the dynamic programming table
25        # For each number in nums, update the count of ways to achieve each sum <= subset_sum
26        for num in nums:
27            for j in range(subset_sum, num - 1, -1):
28                # Add the number of ways to achieve a sum of j before num was considered
29                dp[j] += dp[j - num]
30
31        # Return the number of ways to achieve subset_sum, which indirectly gives us the number of ways
32        # to achieve the target sum using '+' and '-' operations
33        return dp[-1]
34
35 # Example usage:
36 # solution = Solution()
37 # result = solution.findTargetSumWays([1, 1, 1, 1, 1], 3)
38 # print(result) # Outputs: 5
39
```

Java Solution

```
1 class Solution {
2     public int findTargetSumWays(int[] nums, int target) {
3         // Initialize the sum of all numbers in nums
4         int sum = 0;
5         for (int num : nums) {
6             sum += num;
7         }
8
9         // If the sum is less than the target or (sum - target) is odd, it's not possible to partition
10        if (sum < target || (sum - target) % 2 != 0) {
11            return 0;
12        }
13
14        // Compute the subset sum needed for one side of the partition
15        int subsetSum = (sum - target) / 2;
16
17        // Initialize a DP array to store the number of ways to reach a particular sum
18        int[] dp = new int[subsetSum + 1];
19
20        // There's one way to reach the sum of 0 - by not including any numbers
21        dp[0] = 1;
22
23        // Go through every number in nums
24        for (int num : nums) {
25            // Update the DP table from the end to the start to avoid overcounting
26            for (int j = subsetSum; j >= num; j--) {
27                // Increase the current dp value by the value from dp[j - num]
28                dp[j] += dp[j - num];
29            }
30        }
31
32        // Return the number of ways to reach the target sum
33        return dp[subsetSum];
34    }
35 }
36
```

C++ Solution

```
1 #include <vector>
2 #include <numeric> // For using accumulate function
3
4 class Solution {
5 public:
6     int findTargetSumWays(vector<int>& nums, int target) {
7         // Calculate the sum of all numbers in the vector
8         int sum = std::accumulate(nums.begin(), nums.end(), 0);
9
10        // If the sum is less than the target or (sum - target) % 2 != 0, return 0
11        if (sum < target || (sum - target) % 2 != 0) return 0;
12
13        // Calculate the new target (n) which is the sum to be found
14        int newTarget = (sum - target) / 2;
15
16        // Initialize the dynamic programming array with zeros
17        // and set dp[0] to 1 since there's one way to get sum 0: using no numbers
18        vector<int> dp(newTarget + 1, 0);
19        dp[0] = 1;
20
21        // Iterate over every number in the input array
22        for (int num : nums) {
23            // For each number, iterate backwards from the new target to the number's value
24            // This is to ensure we do not count any subset more than once
25            for (int j = newTarget; j >= num; --j) {
26                // Update the dp array to count additional ways to reach current sum (j)
27                // by adding the number of ways to reach the sum without the current number (j - num)
28                dp[j] += dp[j - num];
29            }
30        }
31
32        // Return the total number of ways to reach the target sum
33        return dp[newTarget];
34    }
35 };
36
```

Typescript Solution

```
1 /**
2  * Calculates the number of ways to assign '+' and '-'
3  * to make the sum of nums be equal to target.
4  *
5  * @param {number[]} nums - Array of numbers to be used.
6  * @param {number} target - The target sum to be achieved.
7  * @return {number} - Number of ways to achieve the target sum.
8  */
9 const findTargetSumWays = (nums: number[], target: number): number => {
10     // Calculate the sum of the input array elements
11     let sum: number = nums.reduce((acc, val) => acc + val, 0);
12
13     // If the sum is less than the target, or the difference is odd, there is no solution
14     if (sum < target || (sum - target) % 2 !== 0) {
15         return 0;
16     }
17
18     const totalNums: number = nums.length;
19     // Calculate the subset sum that we need to find
20     const subsetSum: number = (sum - target) / 2;
21     // Initialize a DP array for storing number of ways to sum up to j with array elements
22     let waysToSum: number[] = new Array(subsetSum + 1).fill(0);
23     waysToSum[0] = 1; // Base case - there's one way to have a sum of zero (using no elements)
24
25     // Fill the DP array
26     for (let i = 1; i <= totalNums; ++i) {
27         for (let j = subsetSum; j >= nums[i - 1]; --j) {
28             waysToSum[j] += waysToSum[j - nums[i - 1]]; // Update the ways to sum to j
29         }
30     }
31
32     // Return the total ways to achieve the subset sum, which is equivalent to the target
33     return waysToSum[subsetSum];
34 };
35
36 export { findTargetSumWays }; // Exporting the function to be used in other modules
37
```

Time and Space Complexity

The time complexity of the algorithm is $O(\text{len}(\text{nums}) * n)$, where `len(nums)` is the number of elements in the input list `nums`, and `n` is the computed value $(\text{sum}(\text{nums}) - \text{target}) // 2$. 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 0 to `n`.