1561. Maximum Number of Coins You Can Get

Game Theory

Sorting

Problem Description

Medium

In this problem, we have 3n piles of coins of different sizes, which means the total number of piles is divisible by 3. You are playing a game with two other friends, Alice and Bob. You will take turns choosing any 3 piles of coins. During each turn: 1. Alice will always take the pile with the largest number of coins.

- 2. You will take the next largest pile.
- 3. Bob will take the pile with the smallest number of coins.

Greedy Array Math

the game. The question provides us with an array of integers piles, where piles[i] represents the number of coins in the i-th pile, and we need to return the maximum coins you can get under the game's rules.

This process repeats until there are no more piles left. Your objective is to maximize the number of coins you get by the end of

The intuition behind the solution is to use a greedy approach. A greedy algorithm is one that always makes the choice that looks

Intuition

largest pile and Bob the smallest, the best strategy for you is to ensure that each time you choose 3 piles, the second largest pile is as large as possible, without ever being the largest. To ensure that, the best approach is to sort the piles in ascending order. By doing so, we can then pair the largest available pile for Alice with the next largest pile for ourselves, which leaves the smallest pile in this subset for Bob. By choosing from the larger

best at the moment without considering the future outcomes. Applying this to our scenario, since Alice will always take the

end of the sorted piles, we can maximize the coins we get. After sorting the piles, we can iterate through the sorted array taking the second largest element from the end in each step until we have picked n piles ourselves. To do so:

1. We start from the second element from the end, as this would have been the largest pile if we picked any three piles. 2. We move two steps towards the front of the array each time to ensure we are always picking the next pile with the maximum number of coins available for us.

- 3. We stop picking after we have selected n piles for ourselves, which happens after n iterations as there are 3n piles and we are picking one out of every three piles.
- Therefore, the sum of the selected piles would give us the maximum coins we can have.
- **Solution Approach**

The solution to this problem leverages the sort function to order the array of piles and then accumulate the coins that you would

receive according to the rules established by the game.

Here is a step-by-step breakdown of the implementation: **Sorting the Piles:** First, we use Python's built-in sort function piles.sort() which sorts the array of piles in ascending order.

Sorting is a common pre-processing step in greedy algorithms. No additional data structure is used here; the array is sorted in place. Slicing the Sorted Piles: After sorting, the largest piles are at the end of the array. By employing Python's slicing, we can get

every other pile starting from the second-to-last element using the slice piles[-2 : len(piles) // 3 - 1 : -2]. • piles [-2] is the starting point, which represents the second largest pile because the largest pile would be taken by Alice.

maximum number of coins you can achieve.

the piles, which are meant for Bob. ∘ :-2 is the step, which allows us to skip every alternate pile to simulate the take turns in which you would always end up with the second largest pile out of every chosen set of three piles.

Summation: Lastly, we apply the sum() function on the sliced array to get the total number of coins accumulated across all

○ len(piles) // 3 - 1 is the stopping point, which ensures that only the piles meant for you are included, leaving out the smallest third of

The algorithm exhibits a greedy property since in every turn, it picks the local optimum (the second biggest pile of every chosen set of three piles) without considering the rest of the array. This local optimum choice leads to a global optimum, which is the

chosen piles. This sum represents the maximum number of coins you can obtain by following this strategy.

the two-thirds of the piles. So, the overall time complexity is dominated by the sorting step, making it O(n log n).

The complexity of this algorithm is primarily determined by the sorting step. Python's sort function has a time complexity of O(n log n). The slicing step is O(n), where n is the total number of piles divided by 3, since we're summing every second pile out of

Assume we have an array of piles piles = [2, 4, 1, 2, 7, 8]. There are 3n = 6 piles, which implies that n = 2. According to the rules, Alice will choose the largest pile, you will choose the second-largest pile, and Bob will choose the smallest pile each

Start from the second-to-last element to avoid the largest pile, which Alice takes. This gives us a starting point of piles [-2], which is 7.

Summation: Summing the slice [7, 4] gives us 11. This is the maximum number of coins you can gather following this

∘ End at len(piles) // 3 - 1 which is 1 here, as you only need to collect n piles and the last third are the smallest ones taken by Bob.

Slicing the Sorted Piles: To simulate taking turns:

turn.

Example Walkthrough

Following this slicing, we would pick the piles at indices -2 and -4 which correspond to the values [7, 4] in the original sorted piles.

Let's go through an example to illustrate the solution approach.

Here is how you can follow the steps outlined in the solution approach:

Sorting the Piles: First, sort piles to get [1, 2, 2, 4, 7, 8].

strategy.

We start taking coins from the second largest pile (hence -2 index)

This allows us to simulate the process of taking piles from the end,

skipping the smallest one each time and ignoring the smallest third.

the point which is just beyond one third of the piles count from the end.

and continue with every second pile from the end until we reach

for i in range(len(piles) // 3, len(piles), 2):

import java.util.Arrays; // Import necessary class for sorting

#include <algorithm> // Include algorithm for std::sort

// Sort the piles in ascending order.

// Return the total number of coins.

Initialize the total coins collected.

total_coins += piles[i]

Return the total coins collected.

for i in range(len(piles) // 3, len(piles), 2):

We start taking coins from the second largest pile (hence -2 index)

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and continue with every second pile from the end until we reach

total_coins = 0

return total_coins

Time and Space Complexity

std::sort(piles.begin(), piles.end());

int maxCoins(vector<int>& piles) {

return totalCoins;

// Function to maximize the number of coins you can get.

// each time to get to the next pile you can take.

int totalCoins = 0; // This will store the total number of coins you can get.

totalCoins += piles[i]; // Add the coins from the current pile to your total.

// Start from the second—largest pile and move two steps backward

for (int i = piles.size() - 2; i >= piles.size() / 3; i -= 2) {

total_coins += piles[i]

Therefore, by applying the greedy algorithm, we were able to maximize your share of the coins to 11. This example validates the approach described and shows how sorting, slicing, and summing in a strategic manner results in an optimal solution for you in

○ Use a step of -2 to simulate the turns, meaning you take every other pile from the end of the sorted piles.

this coin pile game.

Sort the piles in ascending order. piles.sort() # Initialize the total coins collected. total_coins = 0

Return the total coins collected. return total_coins

Solution Implementation

def maxCoins(self, piles):

Python

Java

class Solution {

#include <vector>

class Solution {

public:

class Solution:

```
/**
* Calculate the maximum number of coins you can get.
* @param piles Array of piles where each pile has a certain number of coins.
* @return The maximum coins that can be taken.
public int maxCoins(int[] piles) {
   // Sort the array to arrange the piles in ascending order
   Arrays.sort(piles);
    // Initialize answer to 0, this will hold the sum of coins picked
    int maxCoins = 0;
    // Start from the second largest pile and move backwards in steps of 2
   // This approach ensures that we take the second largest piles leaving the smallest ones
    for (int i = piles.length - 2; i >= piles.length / 3; i -= 2) {
        // Add the number of coins from the current selected pile to the answer
       maxCoins += piles[i];
    // Return the calculated maximum number of coins
    return maxCoins;
```

function maxCoins(piles: number[]): number { // Sort the array in ascending order.

TypeScript

};

```
piles.sort((a, b) => a - b);
   // Get the total number of piles.
   const totalPiles = piles.length;
   // Initialize the maximum number of coins.
   let maxCoins = 0;
   // Calculate the number of times we can pick piles according to the rules.
   const timesToPick = Math.floor(totalPiles / 3);
   // Iterate to pick the second-largest pile from each triplet from the end.
   for (let i = 1; i <= timesToPick; i++) {</pre>
       // The index for the second-largest pile in each triplet
        const index = totalPiles - 2 * i;
       // Accumulate the number of coins from the chosen piles.
       maxCoins += piles[index];
   // Return the maximum number of coins we can collect.
   return maxCoins;
class Solution:
   def maxCoins(self, piles):
       # Sort the piles in ascending order.
       piles.sort()
```

Sorting the piles list is the most computationally heavy operation. The sort function in Python typically uses Timsort, which has a time complexity of $O(n \log n)$, where n is the number of elements in the list to sort.

The slicing and summing operations occur after sorting. Slicing in Python is O(k), where k is the number of elements in the slice. However, in this case, the code is slicing from the second-to-last element down to a third of the list's length, creating a subsequence. The sum function operates over this subsequence. Its complexity is O(m), where m is the length of the

Time Complexity

subsequence.

The time complexity of the given code is determined primarily by the sort operation and the slicing to sum up the required coins.

The value of m is approximately (2/3)n/2 elements because we start at the second to last element and take every second element till we reach the first third of the array. Thus, the slice and sum operation would have a complexity of O(n).

Combining both the sorting and the summation complexities, the time complexity of the code is O(n log n) + O(n), which

- simplifies to O(n log n) because the sorting complexity dominates the overall time complexity. **Space Complexity**
- The space complexity of the code consists of the additional space required for the sorted array and space for the variables used

proportional to the input size is created.

in summing operation: The sort() method sorts the list in place and therefore does not require additional space, so this contributes O(1) space

- complexity. Slicing does not create a new list; it creates a view on the existing list, which also is O(1) space complexity since it's not
- duplicating the entire list. Variable space for the sum operation is negligible and constant, so adds O(1).
- Considering all the extra space required, the overall space complexity of the function is O(1), since no additional space