### **Problem Description**

the item, while the weight represents its physical weight. We are also given a capacity, which is the maximum weight limit of a bag we want to fill with items (or portions of items) to maximize the total price of items inside the bag.

A key aspect of the problem is that items can be divided into portions, with each portion keeping the same price-to-weight ratio as

The problem provides us with a list of items, each characterized by two values: price and weight. The price represents the value of

the original item. That means we can take part of an item if taking the whole item would exceed the bag's weight limit. The challenge is to find the maximum possible total price we can achieve given the bag's capacity. Since the result must be within 10^-5 of the actual answer, we're dealing with an approximation and a floating-point number for the

end result. Intuition

#### The solution approach to this problem is similar to the classic knapsack problem which is generally solved using dynamic programming. However, since this problem allows for dividing the items (thus making it a fractional knapsack problem), we can use a

greedy algorithm instead.

To maximize the total price, we intuitively want to prioritize taking items or portions of items that have the highest price-to-weight ratio first, since they provide the most value for the least amount of weight.

Here's how we arrive at the solution: 1. We sort the items based on their price-to-weight ratio in ascending order. Sorting by x[1] / x[0] implies sorting by weight-toprice ratio, which is the inverse of our desired price-to-weight ratio, effectively sorting by cheapest cost efficiency first.

2. Then, we iterate over the sorted items and keep adding them to our bag.

weight without exceeding the capacity of our bag.

- 3. For each item, v = min(w, capacity) determines the actual weight we can add to the bag. This will either be the full weight w of the item if it fits, or the remaining capacity of the bag if there isn't enough space for the whole item.
- 6. If the loop terminates and capacity is not zero, it means we were unable to fill the bag completely, which should not happen

5. We deduct the weight of the item or portion of the item from the remaining capacity.

4. ans += v / w \* p calculates the price of the portion added to the bag. We update the total price in ans accordingly.

- given we can take portions of items. Therefore, if capacity is zero, we return ans which holds the maximum total price, and if capacity is greater than zero (it shouldn't be according to the problem statement), we return -1.
- **Solution Approach**

By employing a greedy strategy, we ensure that we always take the item or portion of the item that has the most value relative to

The solution approach leverages a greedy algorithm, which is often used when we're looking to make a sequence of choices that are locally optimal (maximizing or minimizing some criteria) with the goal of finding a global optimum.

The algorithm sorts the array of items based on a key, which is the ratio of weight to price (x[1] / x[0]). This sorting step is crucial

as it allows us to later iterate through the items in order of what provides the least value per weight unit, due to the ascending order.

### Once the items are sorted, the implementation uses a for loop to iterate over each item. The min function determines how much of

only for the remaining capacity.

remaining capacity.

the current item's weight can be used without exceeding the bag's capacity. This means that if the bag's remaining capacity is greater than or equal to the current item's weight, we can take the full item. Otherwise, we can only take a portion of it that fits the

weight with the item's total price (v / w \* p). This product gives us the value of the portion of the item being considered, which is accumulated in the ans variable representing the current total price. The capacity of the bag is decreased by the weight we just decided to take. This ensures that in the next iteration we're accounting

Post this evaluation, it calculates the price of the amount taken by multiplying the ratio of the weight we could fit to the item's total

After the loop, the conditional -1 if capacity else ans serves as a final check. The intent here is that if we have any remaining capacity, the algorithm should return -1, indicating that the bag was not filled correctly. However, based on the problem description, this condition should not occur because we are allowed to take any fractional part of an item, which implies that we should always

that leads to redundancy. If the capacity is indeed zero, ans is returned, giving us the maximum total price for filling the bag.

structures are used outside of the variables for accumulating the total price and maintaining remaining capacity.

end up with capacity reaching zero before we run out of items. Still, this is included perhaps as a safety check or due to an oversight

In terms of data structures, the input array items and the iteration for accessing each (p, w) are straightforward. No additional data

Example Walkthrough Let's consider an example to illustrate the solution approach. Imagine we have a list of items with the following price and weight:

3. Item 3: price = 120, weight = 30 And let's say the capacity of our bag is 50.

### ○ Item 1: 60 / 10 = 6.0

maximum value:

∘ Item 3: 120 / 30 = 4.0

○ Item 2: ratio = 5.0

∘ Item 3: ratio = 4.0

4. Move to the second item:

how much we can take:

Python Solution

class Solution:

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from typing import List

Total price = 60 + 100 + 80 = 240.

# Loop through the sorted items

weight\_to\_take = min(weight, capacity)

# Add to the total maximum price

# Decrease the remaining capacity

# Break if the capacity is filled

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

1 #include <algorithm> // Required for std::sort

// Iterate through each item

for (const auto& item : items) {

capacity -= weightToTake;

if (capacity == 0) {

// Function to calculate maximum price of items fit into a given capacity

double totalValue = 0.0; // Initialize total value to accumulate

// Add value of the current item fraction to total value

totalValue += static\_cast<double>(weightToTake) / weight \* price;

// Sorting the items based on the price-to-weight ratio in descending order

std::sort(items.begin(), items.end(), [&](const auto& item1, const auto& item2) {

int weightToTake = std::min(weight, capacity); // Weight to take of current item

double maxPrice(std::vector<std::vector<int>>& items, int capacity) {

return item1[1] \* item2[0] < item1[0] \* item2[1];</pre>

int price = item[0]; // Price of the current item

// Decrease the capacity by the weight taken

// Otherwise return the total value of items taken

return capacity > 0 ? -1 : totalValue;

int weight = item[1]; // Weight of the current item

// Break the loop if the capacity is fully utilized

#include <vector>

class Solution {

});

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max\_price += price\_for\_weight

capacity -= weight\_to\_take

# Calculate the price for the weight taken

for price, weight in items:

1. Item 1: price = 60, weight = 10

2. Item 2: price = 100, weight = 20

∘ Item 2: 100 / 20 = 5.0

2. Next, we sort the items based on their price-to-weight ratio in descending order, so we first fill our bag with items that give the

5. Finally, for the third item, we can't add it in full because the remaining capacity is only 20 and the weight is 30. So, we calculate

Since we want to sort based on highest value first we have: ○ Item 1: ratio = 6.0

Now, here is how we would apply the solution approach step-by-step:

1. First, we calculate the price-to-weight ratio for each item:

3. Now, we start to fill the bag. We take the first item in full because the capacity allows it.

The items are already sorted in descending order of price-to-weight ratio.

Add Item 1: price = 60, weight = 10. Remaining capacity = 40.

Add Item 2: price = 100, weight = 20. Remaining capacity = 20.

We take (\frac{20}{30}) of Item 3: price contribution = (\frac{20}{30}\times 120 = 80). 6. We calculate the final price:

Total price = price of Item 1 + price of Item 2 + fraction of price from Item 3.

- 7. The remaining capacity of the bag is now zero. We've maximized the total price of items in the bag without exceeding the weight capacity, successfully applying the greedy strategy to this fractional knapsack problem.
  - def max\_price(self, items: List[List[int]], capacity: int) -> float: # Initialize the maximum price achieved to zero max\_price = 0 # Sort items by their value to weight ratio in ascending order items.sort(key=lambda item: item[0] / item[1], reverse=True)

# Take the minimum of item's weight or remaining capacity

price\_for\_weight = (weight\_to\_take / weight) \* price

if capacity == 0: 21 22 break 23 24 # Return the total maximum price if the capacity has been completely used, else return -1 25 return max\_price if capacity == 0 else -1

The resulting maximum total price to fill the bag is 240, following the greedy algorithm based on the price-to-weight ratio.

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27 # Example usage:
28 # solution = Solution()
29 # items = [[60, 10], [100, 20], [120, 30]] # Each item is [price, weight]
30 # capacity = 50
31 # print(solution.max_price(items, capacity)) # Expected output is the maximum price that fits into the capacity
```

**Java Solution** 

class Solution {

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// Function to calculate the maximum price achievable within the given capacity
       public double maxPrice(int[][] items, int capacity) {
           // Sort the items array based on value—to—weight ratio in descending order
            Arrays.sort(items, (item1, item2) \rightarrow item2[0] * item1[1] \rightarrow item1[0] * item2[1]);
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           // Variable to store the cumulative value of chosen items
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           double totalValue = 0;
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           // Iterate through each item
            for (int[] item : items) {
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                int price = item[0];
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                int weight = item[1];
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                // Determine the weight to take, up to the remaining capacity
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                int weightToTake = Math.min(weight, capacity);
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               // Compute value contribution of this item based on the weight taken
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                double valueContribution = (double) weightToTake / weight * price;
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               // Add the value contribution to the total value
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                totalValue += valueContribution;
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                // Subtract the weight taken from the remaining capacity
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                capacity -= weightToTake;
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               // If no capacity is left, break the loop as no more items can be taken
                if (capacity == 0) {
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                    break;
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           // If there is unused capacity, the requirement to fill the exact capacity is not met
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            // In this context, return -1 to indicate the requirement is not fulfilled
            return capacity > 0 ? -1 : totalValue;
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40 }
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C++ Solution
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#### 29 break; 30 31 32 33 // Return -1 if there's remaining capacity, indicating incomplete filling

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37 };
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Typescript Solution
 1 // Define the maxPrice function which calculates the maximum price that can
 2 // be achieved given a set of items and a capacity constraint
   function maxPrice(items: number[][], capacity: number): number {
       // Sort the items based on the unit price in descending order
       items.sort((a, b) => b[1] / b[0] - a[1] / a[0]);
       // Initialize the maximum price achievable to 0
       let maxPrice = 0;
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       // Loop through each item
       for (const [price, weight] of items) {
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           // Determine how much of the item's weight can be used, up to the remaining capacity
13
           const usableWeight = Math.min(weight, capacity);
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           // Increment the maximum price by the value of the current item,
           // prorated by the fraction of usable weight to its full weight
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           maxPrice += (usableWeight / weight) * price;
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           // Decrease the remaining capacity by the weight of the current item used
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           capacity -= usableWeight;
21
22
           // If no capacity is left, break out of the loop
23
           if (capacity === 0) break;
24
       // If there is no capacity left (i.e., the knapsack is filled to its limit),
       // return the maximum price, otherwise, return -1 indicating not all capacity was used
       return capacity === 0 ? maxPrice : -1;
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29 }
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```

# Time and Space Complexity

## The time complexity of the provided code consists of two main operations: sorting the list and iterating through the list.

operation is the dominant term.

**Time Complexity** 

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- Sorting the list of items has a time complexity of O(NlogN), where N is the length of the items. This is because the built-in sorted() function in Python uses TimSort (a combination of merge sort and insertion sort) which has this time complexity for sorting an array.
- The iteration through the sorted list has a time complexity of O(N) since each item is being accessed once to calculate the proportional value and update the remaining capacity. Combining these two operations, the overall time complexity is O(NlogN) + O(N), which simplifies to O(NlogN) as the sorting

**Space Complexity** • The space complexity of sorting in Python is O(N) because the sorted() function generates a new list.

- The additional space used in the code for variables like ans and v are constant 0(1).
- Therefore, the overall space complexity is O(N) due to the sorted list that is created and used for iteration.