

1. What is the time/space efficiency of each of your algorithms?

**Time Complexity for brute force algorithm(Exhaustive Search) is  $O(N \cdot 2^n)$**

Besides the memory used to store the values and weights of all items, this algorithm requires a two one dimensional arrays

$$\begin{aligned} \sum_{i=1}^{2^n} \left[ \sum_{j=1}^1 + \sum_{k=1}^n \right] &= \sum_{i=1}^{2^n} [\{1+..+1\}(n \text{ times}) + \{1+..+1\}(n \text{ times})] \\ &= (2^n) * [1+1+1+....+1] (2^n \text{ times}) \\ &= O(2^n * 2^n) \\ &= O(n * 2^n) \end{aligned}$$

**Time Complexity for dynamic programming is  $O(N \cdot \text{Capacity})$**

In terms of memory, Dynamic Programming requires a two dimensional array with rows equal to the number of items and columns equal to the capacity of the knapsack

$$\begin{aligned} \sum_{i=0}^N \sum_{j=0}^{\text{Capacity}} 1 &= \sum_{i=0}^N [1+1+1+....+1] (\text{Capacity times}) \\ &= \text{Capacity} * [1+1+1+....+1] (N \text{ times}) \\ &= \text{Capacity} * N \\ &= O(N * \text{Capacity}) \end{aligned}$$

**Time Complexity for Greedy Algorithm is  $O(N \log N)$**

In terms of memory, this algorithm only requires a one dimensional array to record the solution string.

Complexity for Sorting by non-increasing order  $O(N \log N)$

Complexity for

$$\sum_{i=0}^N 1 = [1+1+1+....+1] (N \text{ times}) = N \approx O(N)$$

Total :  $O(N \log N) + O(N) = O(N \log N)$

2. Do all three of your solutions provide an optimal solution? Why or Why not?

Both the Exhaustive Search and dynamic programming provide an optimal solution from the given example. However, they don't include fractions during the calculation. In this case, **Only the Greedy Algorithm provides the optimal solution.**

For the given example, both the Exhaustive Search and dynamic programming provide 198 but Greedy Algorithm provides 226 which is calculated by the multiplication of the following array. Noticed it has 0.8.

[1, 1, 0.8, 0, 0]

[120, 74, 40, 25, 4]

optimal value: 226.0

For my own example 1, **Only the Greedy Algorithm provides the optimal solution.**

3. Create and supply two other knapsack examples to test your 3 functions.

example 1

input file: input2.txt is:

The Capacity: 60

Weights of each item: ['40', '10', '20', '24']

Values of each item: ['280', '100', '120', '120']

Number of items: 4

Testing Brute Force

optimal value: 340

Time use: 2.6941299438476562e-05 seconds.

Testing Dynamic Programming

optimal value: 400

Time use: 0.00036716461181640625 seconds.

**Testing Greedy Algorithm**

**optimal value: 440.0**

**Time use: 0.0001201629638671875 seconds.**

example 2

Solution for Knapsack Problem

input file: input3.txt is:

The Capacity: 5

Weights of each item: ['4', '2', '3']

Values of each item: ['10', '4', '7']

Number of items: 3

**Testing Brute Force**

**optimal value: 11**

**Time use: 2.09808349609375e-05 seconds.**

**Testing Dynamic Programming**

**optimal value: 11**

**Time use: 2.7179718017578125e-05 seconds.**

Testing Greedy Algorithm

optimal value: 12.3333333333

Time use: 0.00011491775512695312 seconds.

4. Which is the better knapsack solution and why? Is this true for all knapsack examples?

The Greedy Algorithm is a better solution because it consider the fraction and give the most optimize solution. Dynamic programming is probably one of the easiest to implement because it does not require the use of any additional structures. Since the complexity of Exhaustive Search grows exponentially, it can only be used for small instances of the KP. In conclusion, for small value, all three solution works well. For later value, the greedy algorithm is the best solution.