

① Logic: The algorithm we used solves the problem using greedy strategy. We first sort the profit array (As all weights are equal, we choose the maximum profit object to maximize our profit). After sorting, we choose the highest profit objects until we reach the max. threshold of Knapsack.

Analysis:

Analysis:
Run Time for our Algo : $\{ \underbrace{O(n \log n)}_{\text{(Merge) Sorting}} + \underbrace{O(n)}_{\text{Traversing Profit array}} \} \rightarrow O(n \log n)$
n is Number of Objects

→ $\frac{\text{Max. weight}}{(\text{Traversing port})}$ Weight of each object

Total Time Complexity : $O(n \log n)$
 $\approx O(n^2)$

Space Complexity: $O(n)$ (Arrays for Merge Sort)

② Logic: Here we again use greedy strategy to solve the problem (as 1). We just sort the weight array (As profits are same, we need to include objects with less weight (More no. of objects)). After sorting, we greedily choose low weight objects until we reach max. capacity of knapsack.

Analysis:

Run time for our Algo: $O(n \log n) + O(n) \rightarrow O(n \log n)$
 (Merge) Sorting Traversing
 sort weight
 array

Total Time complexity = $O(n \log n) + O(n^2)$

Space Complexity: $O(n)$ (Arrays for Merge sort)

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PAA Practice EndSem

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③ Logic: Our Algo solves the problem using top down approach of Dynamic Programming using the concept of Memorization where we store the recursive sub problems in a 2D array and use another function to use those subproblems and evaluate our problem and return the solution. We use memory to affect the Time complexity and get lesser time complexity.

Analysis: We run the Algo recursively and store it in a $N \times W$ table. So, the time complexity is $O(N \times W)$ where N is the Number of items and W is the max weight. (For smaller weights, it's polynomial).

Time Complexity: $O(N \times W)$

Space Complexity: As we maintain a $N \times W$ table, the space complexity is $O(N \times W)$
