

# CSE 102 Spring 2021

## Advanced Homework Assignment 4

Jaden Liu  
University of California at Santa Cruz  
Santa Cruz, CA 95064 USA

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### 1 AdvHW4

**Greedy Approximation to 0-1 Knapsack Problem: (up to 10 points: simple proof without external resources)** Consider the 0-1 knapsack problem: Given  $n$  items with weights  $w_1, \dots, w_n$  and value  $v_1, \dots, v_n$  and a total weight of  $W$ , where each  $w_i, v_i$  and  $B$  are positive integers, find a subset  $S$  of items that a thief would like to steal so that the total weight is smaller than  $W$  and the total value is maximum.

Greedy strategy:

1. Calculate the value of each item's unit weight  $v_i/w_i$
2. Pack items with the highest unit weight into the backpack.
3. If total weight of the items in the backpack does not exceed give weight maximum  $W$ , then select items with the second highest value per unit weight.
4. Loop process 2 and 3 to pack as many backpacks as possible until the backpack is full.
5. After the looping, we need to compare  $v_{k+1}$  and  $v_1 + v_2 + v_3 + \dots + v_k$  unless we could get wrong other. Our final answer would be  $\max(v_{k+1}, v_1 + v_2 + v_3 + \dots + v_k)$  for  $1 \leq k \leq n$

*Proof.* First the we sort the knapsack's items in decreasing order of the value density  $v_i/w_i$ , then item  $n_1$  has biggest value density with value  $v_1$  and  $w_1$ . Assume we have the optimal total value  $v_{max}$ , then obviously  $v_1 + v_2 + v_3 + \dots + v_k \leq v_{max}$ . Also we can have  $v_{max} \leq v_1 + v_2 + v_3 + \dots + v_k + v_{k+1}$ . This situation is like partial Knapsack Problem. When we add partial  $v_{k+1}$  to fill the pack completely. The pack has already been at the maximum value density, hence adding partial  $v_{k+1}$  is equal to  $v_{max}$ . Not to mention adding whole  $v_{k+1}$  is larger or equal to  $v_{max}$ . Then we have to two inequalities:

$$\begin{cases} v_{max} \leq v_1 + v_2 + v_3 + \dots + v_k + v_{k+1} \\ v_1 + v_2 + v_3 + \dots + v_k + v_{k+1} \leq 2\max(v_{k+1}, v_1 + v_2 + v_3 + \dots + v_k) \end{cases}$$

It's easy to prove the second inequality if we consider  $v_{k+1} = a$ ,  $v_1 + v_2 + v_3 + \dots + v_k = b$ , then  $\max(a, b) \geq a$  and  $\max(a, b) \geq b$ . Thus  $2\max(a, b) \geq a + b$ .

Combine these two inequalities, we can get  $v_{max} \leq 2\max(v_{k+1}, v_1 + v_2 + v_3 + \dots + v_k)$ . Thus the

solution of our greedy algorithm is at least  $\frac{V_{max}}{2}$ .

For further convenience, I want to change the name " $v_{max}$ " to " $v_{optimal}$ " or " $v_{opt}$ ", and name the solution of our greedy algorithm " $\max(v_k + 1, v_1 + v_2 + \dots v_k)$ " as " $v_{max}$ ".

1. It's hard to compare  $v_{max}$  and  $v_{opt}$ . Therefore, we need to come up with a "bridge" to connect these two variables. It's natural to think of  $\max(a, b) \geq a$  or  $b$ . In order to get the coefficient "2" in the final result, we can easily come up with inequality that  $2 * \max(a, b) \geq a + b$ . Then we find  $a + b$ , which is " $v_1 + v_2 + \dots + v_{k+1}$ " is always larger than  $v_{opt}$ .

2. 0-1 Knapsack problem is different from general Knapsack problem mainly because of its element divisibility. In general Knapsack problem, for example, we have fulfilled the pack with  $v_1$  to  $v_k$  with only  $w_{rem}$ . However, we have  $w_{k+1}$  for the last element which cannot be packed in the bag. Thus, we choose  $w_{rem}/w_{k+1}$  of the last element, so the total value is  $v_1 + \dots + v_{k+1} * \frac{w_{rem}}{w_{k+1}}$ . It's easy to prove  $v_{max} = v_{opt}$  in this situation since we have chosen from most valuable to least valuable to let the final pack density reaching maximum. However, in the 0-1 Knapsack problem, the elements are not divisible. Hence,  $v_1 + \dots + v_{k+1}$  is always larger to  $v_{opt}$ .  $\square$