

- **Greedy is not optimal.**

Pf. A counterexample when greedy strategy does not generate an optimal solution is as follows.

knapsack 1: $B_1 = 5$

knapsack 2: $B_2 = 5$

Item 1: $s_1 = 3, p_1 = 1$

Item 2: $s_2 = 3, p_2 = 1$

Item 3: $s_3 = 2, p_3 = 100$

Item 4: $s_4 = 2, p_4 = 100$

An optimal solution is to pack item 1 and 3 into knapsack 1, and item 2 and item 4 into knapsack 2, which gives a total value of $p_1 + p_2 + p_3 + p_4 = 202$.

Greedy solution, however, will pick item 3 and item 4 for knapsack 1. Then item 3 and 4 cannot fit into knapsack 2 at the same time and the total value will be $p_1 + p_3 + p_4 = 201$, which is smaller than 202 calculated above and therefore not optimal.

- **Greedy gives $O(1)$ -approximation for identical knapsacks.**

Pf.

Suppose x_i is the total value of items covered by i th knapsack, y_i the total elements covered by the time i th knapsack finishes packing, and $z_i = OPT - y_i$ where OPT is the maximum value of packed items at the end. Let k denote the total number of knapsacks.

Because capacity of the knapsacks are identical and every individual solution of a knapsack is solved optionally at its greedy step, we have

$$x_1 \geq x_2 \geq \dots \geq x_i.$$

Therefore we know that the i th knapsack covers at least a fraction of the remaining items that are eventually covered, written as $x_{i+1} \geq \frac{z_i}{k}$.

Then we use mathematical induction to show that $z_{i+1} \leq (1 - \frac{1}{k})^{i+1} * OPT$. The base case is when $i = 1$ and is trivially true. Suppose $z_i \leq (1 - \frac{1}{k})^i * OPT$ is true, then

$$z_{i+1} \leq z_i - x_{i+1} \leq z_i * (1 - \frac{1}{k}) \leq (1 - \frac{1}{k})^{i+1} * OPT.$$

After we plug in $i = k - 1$, it becomes

$$z_k \leq (1 - \frac{1}{k})^k * OPT = (1 - \frac{1}{e}) * OPT.$$

Because $(1 - \frac{1}{e})$ is constant, greedy gives a $O(1)$ -approximation.

- **Greedy gives $O(1)$ -approximation for knapsacks ordered in decreasing size.**

Pf.

Let g_i be the total value of chosen items in i th knapsack by greedy algorithm, OPT be that in an optimal solution. We also denote d_i to be the total value of "discarded" items, or items chosen by an optimal solution but not in g_i , for i th knapsack.

We will first show that $g_i \geq d_i$ for all knapsacks. Suppose $g_i < d_i$, then if we replace the current items in i th knapsack with the "discarded" items, the new total value will be greater than the original. This

contradicts the fact that greedy algorithm picks the combination of items that generates a greatest total value at that specific step. Adding up values for all i , it must be the case that

$$g_{TOT} \geq d_{TOT}.$$

The value of all items chosen in greedy and in optimal solution is $OPT - d_{TOT}$. In addition, greedy may choose other items that are not in optimal solution, which means

$$g_{TOT} \geq OPT - d_{TOT}.$$

Adding up the above two inequalities, we get $2 * g_{TOT} \geq OPT$, or $g_{TOT} \geq \frac{OPT}{2}$. Thus, greedy algorithm gives a 2-approximation whose ratio is constant.

(Discussed with Diqui Zhou)