

CS 4820, Spring 2017 Homework 11, Problem 2

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(a) A counterexample for GA is an instance with two elements: $(v_1 = 2, w_1 = 1)$ and $(v_2 = 2C + 1, w_2 = 2C + 1)$. If the knapsack capacity is $2C + 1$, then GA will pick the first element and then the second one won't fit. The value obtained will be only 2, whereas value $2C + 1$ would be obtained by taking the second element instead.

A counterexample for EMGA is an instance with $2C + 2$ elements: one element with $(v_1 = 1, w_1 = 2C + 1)$ and $2C + 1$ elements each with $(v_i = 1, w_i = 1)$. If the knapsack capacity is $2C + 1$, then EMGA will pick the first element and then all the remaining ones won't fit. The value obtained will be only 2, whereas value $2C + 1$ would be obtained by taking all elements except the first one.

(b) Assume that there are no items whose weight is greater than W . Such items will never appear in any solution, and can be eliminated without changing the value of the optimum. We try to prove that $GA + EMGA \geq OPT$ from which it follows that $\max\{GA, EMGA\} \geq \frac{1}{2} * OPT$. Let i_1, \dots, i_k denote the sequence of items selected by GA (in the order selected) and let $i(k+1)$ be the next element that GA would have selected, if it had not run out of space. The EMGA gets a value at least as high as the value of item $i(k+1)$, and the value of items i_1, \dots, i_k is the value achieved by the GA. On the other hand, the combined value of items $i_1, \dots, i_k, i(k+1)$ is greater than that of the optimal knapsack solution S_{opt} , since S_{opt} has a smaller combined size, and any elements of S_{opt} that are not among $\{i_1, \dots, i(k+1)\}$ have a value density that is smaller than the least dense element of that set.

Let $S_0 = S_{OPT} \cap \{i_1, \dots, i_{k+1}\}$, $S_1 = \{i_1, \dots, i_{k+1}\} \setminus S_{OPT}$, $S_2 = S_{OPT} \setminus \{i_1, \dots, i_{k+1}\}$, if the value density of each item i by $p_i = v_i/w_i$, and let $p^* = p_{i(k+1)}$, from greedy we know that $p_{i(j)} \geq p^*$ for $j = 1, \dots, k+1$ whereas $p_i \leq p^*$ for every i not belongs to $\{i_1, \dots, i_{k+1}\}$. So $v_i \geq p^* w_i$ for every i belong to S_1 , while $v_i \leq p^* w_i$ for every i belongs to S_2 . It shows that:

$$\begin{aligned} \left(\sum_{j=1}^{k+1} v_{i_j}\right) - \left(\sum_{i \in S_{opt}} v_i\right) &= \left(\sum_{i \in S_1} v_i\right) - \left(\sum_{i \in S_2} v_i\right) \geq \left(\sum_{i \in S_1} p^* w_i\right) - \left(\sum_{i \in S_2} p^* w_i\right) = \\ p^* \left(\sum_{i \in S_1} w_i - \sum_{i \in S_2} w_i\right), &\left(\sum_{j=1}^{k+1} v_{i_j}\right) - \left(\sum_{i \in S_{opt}} v_i\right) \text{ is positive and because } \sum_{i \in S_1} w_i > W - \\ \sum_{i \in S_0} w_i \geq \sum_{i \in S_2} w_i, &\text{ so } p^* \left(\sum_{i \in S_1} w_i - \sum_{i \in S_2} w_i\right) \text{ is also positive and } v_{i_1} + \dots + v_{i_{k+1}} > \\ OPT, &\text{ so } GA + EMGA \geq OPT \text{ is proved.} \end{aligned}$$

(c) It is presented that dynamic programming algorithm for the knapsack problem that runs in time $O(nv^*)$, where v^* is the upper bound of the value of the optimum solution. It is not a polynomial algorithm because the running time is proportional to v^* . Compute a set S^* obtained by running GA and EMGA to get two sets, and taking the one with the higher total value. Let $v_* = \sum_{i \in S_*} v_i$ and $k = \frac{\epsilon v_*}{(1+\epsilon)n}$, for every item i let $v'_i = \frac{v_i}{k}$. Let S be the output of the dynamic program on the instance defined by $\{(v'_i, w_i) | i = 1, \dots, n\}$ and S_{opt} by $\{(v_i, w_i) | i = 1, \dots, n\}$. Because it outputs the optimal solution for the knapsack instance on $\{(v'_i, w_i) | i = 1, \dots, n\}$. We get that $\left(\sum_{i \in S} v'_i\right) \geq \left(\sum_{i \in S_{OPT}} v'_i\right)$, $\left(\sum_{i \in S} v_i\right) \geq \left(\sum_{i \in S} k v'_i\right) \geq \left(\sum_{i \in S_{OPT}} k v'_i\right) \geq \left(\sum_{i \in S_{OPT}} (v_i - k)\right)$, $\left(\sum_{i \in S} v_i\right) \geq \left(\sum_{i \in S_{OPT}} v_i\right) - nk = \left(\sum_{i \in S_{OPT}} v_i\right) - \frac{\epsilon v_*}{(1+\epsilon)} \geq \left(1 - \frac{\epsilon}{1+\epsilon}\right) \left(\sum_{i \in S_{OPT}} v_i\right)$, because $v_* \leq \sum_{i \in S_{OPT}} v_i$, so $\sum_{i \in S} v_i \geq \left(\frac{1}{1+\epsilon}\right) \left(\sum_{i \in S_{OPT}} v_i\right)$, so S is a $(1 + \epsilon)$ approximation for S_{opt} . So based on the value of k , the running time is $O(n^2(v^*/v_*)/\epsilon)$. The

computation of the value density of elements and running GA and EMGA for deciding the set S^* takes $O(n \log n)$, based on the algorithm of $O(n^3 s / \varepsilon)$ presented in class, the overall running time of the algorithm is $O(n^2 s / \varepsilon)$ where s denotes the maximum number of bits in the binary representation of any of the numbers $v_i, w_i (i=1, \dots, n)$.