Sishir Subedi Assignment -1 February 14, 2018 CSCE -669 Computational Optimization

Problem description - The KNAPSACK problem:

Given a set $Z = \{z_1, z_2, ..., z_n\}$ of n items, where each item z_i has a size s_i and a value v_i , and a knapsack of size S. Pick a subset of the items that can fit into the kanpsack and maximizes the value.

The KNAPSACK problem is NP-hard:

To show the knapsack problem is NP-hard we will use a known NP-Complete PARTITION problem which is described as the following:

Input:Given a set $X = \{a_1, a_2, ..., a_n\}$ of n items, where each a_i is an interger, partition set X into two disjoint sets X' and X" such that the sum of the integers in each subset equals to each other i.e.

$$\sum_{a_i \in X'} a_i = \sum_{a_i \in X''} a_i$$

Now, we take an instance of a knapsack problem where we have the value of items and size of the items in knapsack is equal i.e. $s_i = v_i$ for each item i in a set $Z = \{z_1, z_2, ..., z_n\}$. Now, we assume that each item in a set Z is also equal to each items set X of partition problem i.e. $\sum_{z_i \in Z} v_i = \sum_{a_i \in X} a_i$. Now when there is two disjoint sets X' and X" then one of the sets in partition i.e. $s_i = v_i = a_i \in X'$, and size of the knapsack S will be equal to sum of elements in set X'.

$$\sum_{z_i \in Z} v_i = V = \sum_{z_i \in Z} s_i = S$$

= $\sum_{a_i \in X'} a_i = \sum_{a_i \in X''} a_i = 1/2 \sum_{a_i \in X} a_i$

Thus, finding a subset of items in knapsack problem will be equivalent of searching a disjoint set X" whose sum will be equivalent to X'.

$$\sum_{a_i \in X'} s_i = 1/2 \sum_{a_i \in X} a_i$$

This reduction from known NP-Complete problem shows that the knapsack is NP-Complete problem.

Develop a simple polynomial time approximation algorithm for the KNAPSACK problem that has a constant approximation ration.

Let $Z = \{z_1, z_2, ..., z_n\}$ of n items, where for each i, the item z_i has a size s_i and a value v_i , and the size of a knapsack is S. We can purpose the following greedy algorithm to pick a subset of the items with maximum value that can fit into the knapsack.

Greedy Algorithm:

- 1. Sort n items in a decreasing order of $v_1/s_1 \ge v_2/s_2 \ge ... \ge v_n/s_n$
- 2. Then fill k items in a knapsack until total size is greater than S i.e.

$$\sum_{i=1}^{k} v_i \le S$$

Here, the algorithm will perform poorly in a case where there exists an item with value of S. For example lets consider we have:

- 1. v_1 is 2 and s_1 is 1, then ratio is 2
- 2. v_2 is S and s_2 is S, then ratio is 1

In this case algorithm will pick v_1 which is not good. We can modify this algorithm as the following:

Modified Greedy Algorithm:

- 1. Sort n items in a decreasing order of $v_1/s_1 \ge v_2/s_2 \ge ... \ge v_n/s_n$
- 2. Then fill k-1 items in a knapsack until total size is greater than S i.e.

$$\sum_{i=1}^{k-1} v_i < S \quad and \quad \sum_{i=1}^k v_i > S$$

3. choose max of:

$$\max(\sum_{i=1}^{k-1} v_i; v_k)$$

Here, let $V_{k-1} = \sum_{i=1}^{k-1} v_i$ be the value obtained after putting k-1 items in the knapsack. We consider a case for fractional knapsack then optimal value for fractional knapsack (FOPT) will be as following:

$$FOPT = \sum_{i=1}^{k-1} v_i + \frac{S - \sum_{i=1}^{k-1} s_i}{s_{ik}} v_k$$

From above we know that the optimal solution (OPT) for interger knapsack is bounded by $OPT < \sum_{i=1}^{k-1} v_i + v_k$ because $\sum_{i=1}^k v_i > S$. Since algorithm choses $max(\sum_{i=1}^{k-1} v_i; v_k)$, we can have either

$$\sum_{i=1}^{k-1} v_i \ge OPT/2 \quad or \quad v_k \ge OPT/2$$

Therefore, modified greedy algorithm is a 2 approximation for the integer knapsack problem. Time complexity of this algorithm is bounded by the complexity of sorting algorithm used to sort items.

Develop a pseudo polynomial time approximation algorithm for the KNAPSACK problem.

We consider a dynamic programming solution for a pseudo polynomial time approximation algorithm for knapsack problem. The components of the dynamic program can be described as the following:

Subproblem: For subset of items $\{1, 2, ...i\}$ in set Z the maximum value is V_i .

Substructure:

case I: Optimal solution for items $\{1, 2, ...i\}$ does not contain item i. In this case, the optimal value is the value obtained from a subset $\{1, 2, ...(i-1)\}$. case II: Optimal solution for items $\{1, 2, ...i\}$ contains item i. In this case, the optimal value is the value obtained from a subset $\{1, 2, ...(i-1)\} + v_i$.

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\begin{array}{l} \textbf{Dynamic Algorithm:} \\ \textbf{Input: $N$ items with sizes $s_1, s_2, ..., s_n$ with values $v_1, v_2, ..., v_n$, size of a backpack is $S$.} \\ \textbf{Create a array: $M[0...N,0...S]$} \\ \textbf{FOR s=0 to S: $M[0,s]=0$} \\ \textbf{FOR i = 0 to N:} \\ \textbf{FOR s=0 to S:} \\ \textbf{IF s[i] \leq w:} \\ \textbf{M[i,s] = } max\{M[i-1,s],v_i+M[i-1,s-s_i]\} \\ \textbf{ELSE:} \\ \textbf{M[i,s] = } M[i-1,s] \\ S^* = max\{s: M[N,s] \leq S\} \\ OPT = M[N,S^*] \end{array}
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The time complexiy of this dynamic algorithm is $O(n^2S^*)$, which is a pseudopolynomial because of bound by value of S^* .

Develop a fully polynomial time approximation scheme for the KNAP-SACK problem.

In a pseudo polynomial time approximation algorithm we notice that if size S^* is small and bounded by N then the algorithm will be polynomial i.e. n^2 . However, S^* is not bounded by N and the algorithm is pseudopolynomial. One idea of decreasing complexity of this algorithm is by dividing each item value v_i by k in such a manner that following conditions are kept:

- 1. maintain integer property i.e. $v_i' = \lfloor \frac{v_i}{k} \rfloor$
- 2. value of k should be as small as possible so that precision lost due to division and floor function is minimum
- 3. value of k should be as high as possible so that the time comlexity of the modified dynamic program is bounded by a polynomial.

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FPTAS Algorithm Framework:

1.For i=1 to n do v_i' = \lfloor \frac{v_i}{k} \rfloor

2.Find optimal solution using dynamic programming for items \{v_1', v_2', ... v_n'\} to get set S'

3.Output v_i corresponding to v_i' as set S_{appx}
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Here, let $X = \{z_1, z_2, ..., z_j\}$ be the set of items with values $V = \{v_1, v_2, ..., v_j\}$ where sum of value of all items in this set is an optimal solution $OPT = \sum_{i=1}^{j} v_i$. Similarly, let $X' = \{z_1', z_2', ..., z_{j'}'\}$ be the set of items with values

 $V^{'}=\{v_{1}^{'},v_{2}^{'},...,v_{i^{'}}^{'}\}$ where sum of value of all items in this set is an optimal solution $OPT^{'} = \sum_{i=1}^{j^{'}} v_{i}^{'}$ for modified values problem. Now, first we need to find relationship between $OPT^{'}$ and OPT. To do that we

need to express OPT in terms of $v_i^{'} = \lfloor \frac{v_i}{k} \rfloor$.

$$\begin{aligned} OPT &= \sum_{i=1}^{j} v_i \\ &= k \sum_{i=1}^{j} \frac{v_i}{k} \\ &\leq k \sum_{i=1}^{j} (\lfloor \frac{v_i}{k} \rfloor + 1) \\ &\leq kn + k \sum_{i=1}^{j} \lfloor \frac{v_i}{k} \rfloor \\ &\leq kn + k \sum_{i=1}^{j} v_i' \\ &\leq kn + k \sum_{i=1}^{j} v_i', \text{ because } OPT \leq OPT' \text{ i.e. } \sum_{i=1}^{j} v_i' \leq \sum_{i=1}^{j'} v_i' \\ &\leq kn + k \sum_{i=1}^{j} \lfloor \frac{v_i}{k} \rfloor \end{aligned}$$

$$OPT \le kn + OPT'$$
 ——(i)

So, we have approximation ration of: $\frac{OPT}{OPT'} \leq \frac{kn}{OPT'} + 1$

$$\frac{OPT}{OPT'} \le \frac{kn}{OPT'} + 1$$

i.e.
$$ratio \leq \frac{kn}{OPT'} + 1$$
 ——(ii)

We have a lower bound for OPT that it is $OPT \geq \frac{\sum_{i=1}^n v_i}{n}$ because optimal will be at equal to or larger than the average of all values. So we know that $OPT' \geq OPT \geq \frac{\sum_{i=1}^n v_i}{n} \geq \frac{V_o}{n}$. Using this to the equation (i): $OPT \leq kn + OPT'$

$$OPT^{'} \ge OPT \ge \frac{\sum_{i=1}^{n} v_i}{n} \ge \frac{V_o}{n}$$
. Using this to the equation (i): $OPT \le kn + OPT^{'}$

$$\begin{array}{l}
OPT - kn \leq OPT \\
\frac{V_o}{} - kn \leq OPT'
\end{array}$$

$$\begin{split} OPT &\leq kn + OPT' \\ OPT - kn \leq OPT' \\ \frac{V_o}{n} - kn \leq OPT' \\ \text{so we substitute value of } OPT' \text{ in equation (ii) then we will have} \\ ratio &\leq \frac{kn}{OPT'} + 1 \\ ratio &\leq \frac{kn}{\frac{V_o}{n} - kn} + 1 \\ ratio &\leq \frac{kn^2}{\frac{V_o - kn^2}{n} + 1} \end{split}$$

 $ratio \leq \frac{kn^2}{V_o - kn^2} + 1$ Here, let $\epsilon \geq \frac{kn^2}{V_o - kn^2}$ so that we have approximation ration bounded as $ratio \leq 1$ $1 + \epsilon$, then

$$\begin{array}{l} \epsilon(V_o-kn^2) \geq kn^2 \\ \epsilon V_o kn^2) \geq kn^2 \\ \frac{\epsilon V_o}{n^2(1+\epsilon)} \geq k \\ \frac{V_o}{n^2(1+\frac{1}{\epsilon})} \geq k \end{array}$$

Therefore, we can substitue $k = \frac{V_o}{n^2(1+\frac{1}{\epsilon})}$ in the FPTAS aglorithm above, and the running time algorithm will be $O(n^3(1+\frac{1}{\epsilon}))$ i.e. $O(\frac{n^3}{\epsilon})$. This provides a solution OPT' for items in Z' set, and its corresponding item in Z will be solution for the knapsack problem by FPTAS algorithm with approximation ratio bounded by ϵ .