



Chapter 7

Approximation Algorithms

Knapsack Approximation

Algorithm Theory
WS 2012/13

Fabian Kuhn

Knapsack

- n items $1, \dots, n$, each item has weight $w_i > 0$ and value $v_i > 0$
- Knapsack (bag) of capacity W
- Goal: pack items into knapsack such that total weight is at most W and total value is maximized:

$$\max \sum_{i \in S} v_i$$

$$\text{s. t. } S \subseteq \{1, \dots, n\} \text{ and } \sum_{i \in S} w_i \leq W$$

- E.g.: jobs of length w_i and value v_i , server available for W time units, try to execute a set of jobs that maximizes the total value

Knapsack: Dynamic Programming Alg.



We have shown:

- If all item weights w_i are integers, using dynamic programming, the knapsack problem can be solved in time $O(nW)$
- If all values v_i are integers, there is another dynamic progr. algorithm that runs in time $O(n^2V)$, where V is the max. value.

Problems:

- If W and V are large, the algorithms are not polynomial in n
- If the values or weights are not integers, things are even worse (and in general, the algorithms cannot even be applied at all)

Idea:

- Can we adapt one the algorithms to at least compute an approximate solution?

Approximation Algorithm

$O(n^2 V)$



- The algorithm has a parameter $\varepsilon > 0$
- We assume that each item alone fits into the knapsack
- We define:
 - $\underline{V} := \max_{1 \leq i \leq n} v_i$ (with "value" written above v_i and an arrow pointing to it)
 - $\forall i: \underline{\hat{v}}_i := \left\lceil \frac{v_i n}{\varepsilon V} \right\rceil$ (with $v_i \cdot \frac{n}{\varepsilon V}$ written above the fraction)
 - $\underline{\hat{V}} := \max_{1 \leq i \leq n} \underline{\hat{v}}_i$
- We solve the problem with values $\underline{\hat{v}}_i$ and weights \underline{w}_i using dynamic programming in time $O(n^2 \cdot \underline{\hat{V}})$

Theorem: The described algorithm runs in time $O(n^3 / \varepsilon)$.

Proof:

$$\underline{\hat{V}} = \max_{1 \leq i \leq n} \underline{\hat{v}}_i = \max_{1 \leq i \leq n} \left\lceil \frac{v_i n}{\varepsilon V} \right\rceil = \left\lceil \frac{V n}{\varepsilon V} \right\rceil = \left\lceil \frac{n}{\varepsilon} \right\rceil$$

Approximation Algorithm

Theorem: The approximation algorithm computes a feasible solution with approximation ratio at most $1 + \varepsilon$.

Proof:

- Define the set of all feasible solutions

$$\mathcal{S} := \left\{ S \subseteq \{1, \dots, n\} : \sum_{i \in S} w_i \leq W \right\}$$

- Let S^* be an optimal solution and \hat{S} be the solution computed by the approximation algorithm.
- We have

$$S^* = \max_{S \in \mathcal{S}} \sum_{i \in S} v_i, \quad \hat{S} = \max_{S \in \mathcal{S}} \sum_{i \in S} \hat{v}_i$$

- Hence, \hat{S} is a feasible solution

Approximation Algorithm

Theorem: The approximation algorithm computes a feasible solution with approximation ratio at most $1 + \varepsilon$.

Proof:

- Because every item fits into the knapsack, we have

$$\forall i \in \{1, \dots, n\}: \underline{v_i} \leq \sum_{j \in S^*} v_j$$

- For the solution of the algorithm, we get

$$\underline{\hat{v}_i} \geq \left\lceil \frac{v_i n}{\varepsilon V} \right\rceil \Rightarrow \underline{v_i} \leq \frac{\varepsilon V}{n} \cdot \hat{v}_i$$

- Therefore

$$\sum_{i \in S^*} v_i \leq \frac{\varepsilon V}{n} \cdot \sum_{i \in S^*} \hat{v}_i \leq \frac{\varepsilon V}{n} \cdot \sum_{i \in \hat{S}} \hat{v}_i \leq \frac{\varepsilon V}{n} \cdot \sum_{i \in \hat{S}} \left(\frac{v_i n}{\varepsilon V} + 1 \right)$$

value of opt. solution (orig. problem) value of the opt. solution with \hat{v}_i

Approximation Algorithm

Theorem: The approximation algorithm computes a feasible solution with approximation ratio at most $1 + \varepsilon$.

Proof:

- We have

$$\sum_{i \in S^*} v_i \leq \frac{\varepsilon V}{n} \cdot \sum_{i \in S^*} \hat{v}_i \leq \frac{\varepsilon V}{n} \cdot \sum_{i \in \hat{S}} \hat{v}_i \leq \frac{\varepsilon V}{n} \cdot \sum_{i \in \hat{S}} \left(\frac{v_i n}{\varepsilon V} + 1 \right)$$

- Therefore

$$\sum_{i \in S^*} v_i \leq \sum_{i \in \hat{S}} v_i + \frac{\varepsilon V}{n} \cdot |\hat{S}| \leq \varepsilon V + \sum_{i \in \hat{S}} v_i$$

- Because V is a lower bound on the optimal solution:

$$\underbrace{V \leq \sum_{i \in \hat{S}} v_i}_{\text{OPT}} \leq \underbrace{(1 + \varepsilon)}_{\text{ALG}} \cdot \underbrace{\sum_{i \in \hat{S}} v_i}_{\text{ALG}} \quad \frac{\text{OPT}}{\text{ALG}} \leq 1 + \varepsilon$$

Approximation Schemes

$$O(\text{poly}(n) \cdot 2^{1/\varepsilon})$$



- For every parameter $\varepsilon > 0$, the knapsack algorithm computes a $(1 + \varepsilon)$ -approximation in time $O(n^3 / \varepsilon)$.
- For every fixed ε , we therefore get a polynomial time approximation algorithm
- An algorithm that computes an $(1 + \varepsilon)$ -approximation for every $\varepsilon > 0$ is called an approximation scheme.
- If the running time is polynomial for every fixed ε , we say that the algorithm is a polynomial time approximation scheme (PTAS)
- If the running time is also polynomial in $1/\varepsilon$, the algorithm is a fully polynomial time approximation scheme (FPTAS)
- Thus, the described alg. is an FPTAS for the knapsack problem