



Chapter 7

Approximation Algorithms

Algorithm Theory
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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:

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

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

- 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 of the algorithms to at least compute an approximate solution?

Approximation Algorithm

$$\left\lceil \frac{2v_i}{\varepsilon V} \cdot v_i \right\rceil \cdot \frac{\varepsilon V}{2n}$$



- The algorithm has a parameter $0 < \underline{\varepsilon} \leq 1$
- We assume that each item alone fits into the knapsack
- We define:

$$\underline{V} := \max_{1 \leq i \leq n} v_i, \quad \forall i: \underline{\hat{v}}_i := \left\lceil \frac{2v_i n}{\varepsilon V} \right\rceil, \quad \underline{\hat{V}} := \max_{1 \leq i \leq n} \underline{\hat{v}}_i$$

- We solve the problem with **integer** values $\underline{\hat{v}}_i$ and weights \underline{w}_i using dynamic programming in time $O(n^2 \underline{\hat{V}})$

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

Proof:

$$\underline{\hat{V}} = \max_{1 \leq i \leq n} \underline{\hat{v}}_i = \max_{1 \leq i \leq n} \left\lceil \frac{2v_i n}{\varepsilon V} \right\rceil = \left\lceil \frac{2V n}{\varepsilon V} \right\rceil = \left\lceil \frac{2n}{\varepsilon} \right\rceil = \Theta\left(\frac{n}{\varepsilon}\right)$$

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 (subsets of $[n]$)

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

• $\rightarrow v(S)$: value of solution S w.r.t. values v_1, v_2, \dots

• $\rightarrow \hat{v}(S)$: value of solution S w.r.t. values $\hat{v}_1, \hat{v}_2, \dots$

- Let S^* be an optimal solution and \hat{S} be the solution found by the approximation algorithm.
- Weights are not changed at all, hence, \hat{S} is a feasible solution

$$\text{approx. ratio} : \max_{\text{instances}} \frac{v(S^*)}{v(\hat{S})} \leq \underline{1 + \varepsilon}$$

Approximation Algorithm

$$v(S^*) \leftarrow v(\hat{S})$$



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

Proof:

- We have

$$v(S^*) = \sum_{i \in S^*} v_i = \max_{S \in \mathcal{S}} \sum_{i \in S} v_i,$$

$$\hat{v}(\hat{S}) = \sum_{i \in \hat{S}} \hat{v}_i = \max_{S \in \mathcal{S}} \sum_{i \in S} \hat{v}_i$$

- Because every item fits into the knapsack, we have

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

- Also: $\hat{v}_i = \left\lceil \frac{2v_i n}{\varepsilon V} \right\rceil \Rightarrow v_i \leq \frac{\varepsilon V}{2n} \cdot \hat{v}_i$, and $\hat{v}_i \leq \frac{2v_i n}{\varepsilon V} + 1$

Approximation Algorithm $v(\hat{S}) \geq v(S^*) - \frac{\epsilon V}{2} \geq v(S^*) \left(1 - \frac{\epsilon}{2}\right)$



Theorem: The approximation algorithm computes a feasible solution with approximation ratio at most $1 + \epsilon$. $v(S^*) \geq V$

Proof:

- We have

$$v(S^*) = \sum_{i \in S^*} v_i \leq \frac{\epsilon V}{2n} \cdot \sum_{i \in S^*} \hat{v}_i \leq \frac{\epsilon V}{2n} \cdot \sum_{i \in \hat{S}} \hat{v}_i \leq \frac{\epsilon V}{2n} \cdot \sum_{i \in \hat{S}} \left(1 + \frac{2v_i n}{\epsilon V}\right)$$

\hat{S} is an opt. sol. w.r.t. \hat{V}

- Therefore

$$v(S^*) = \sum_{i \in S^*} v_i \leq \frac{\epsilon V}{2n} \cdot |\hat{S}| + \sum_{i \in \hat{S}} v_i \leq \frac{\epsilon V}{2} + v(\hat{S})$$

- V is a lower bound on $v(S^*)$:

$$\left(1 - \frac{\epsilon}{2}\right) \cdot v(S^*) \leq v(\hat{S}), \quad 0 < \epsilon \leq 1 \Rightarrow 1 - \frac{\epsilon}{2} \geq \frac{1}{1 + \epsilon}$$

$\frac{1}{1 - \epsilon/2} \approx 1 + \frac{\epsilon}{2}$

Approximation Schemes

- 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) $O(n^3 \cdot e^{1/\varepsilon})$
- 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