



Combinatorics of Efficient Computations

Approximation Algorithms

Lecture 8: FPTAS for Knapsack via Scaling

Joachim Spoerhase

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Approximation Scheme

Let Π be an optimization problem. An algorithm \mathcal{A} is called polynomial time approximation scheme (PTAS), if it computes for every (I, ϵ) with $I \in D_{\Pi}$ and $\epsilon > 0$ a solution $s \in S_{\Pi}(I)$ with the following properties:

- $obj_{\Pi}(I,s) \leq (1+\epsilon) \cdot OPT$, if Π is a minimization problem,
- $\operatorname{obj}_{\Pi}(I,s) \ge (1-\epsilon) \cdot \operatorname{OPT}$, if Π is a maximization problem. The running time of \mathcal{A} is polynomial in |I| for every fixed $\epsilon > 0$.

 \mathcal{A} is called **fully polynomial time approximation scheme** (FPTAS), if its running time is polynomial in |I| and $1/\epsilon$. Example running times

- $O(n^{1/\epsilon}) \rightsquigarrow$ polynomial time approximation scheme
- $O(2^{1/\epsilon}n^4) \rightsquigarrow$ polynomial time approximation scheme
- $O(n^3/\epsilon^2) \rightsquigarrow$ fully polynomial time approximation scheme (FPTAS)

Knapsack Problem

We are given a set $S = \{a_1, \ldots, a_n\}$ of **objects**. For every object a_i , $i = 1, \ldots, n$ two quantities size $(a_i) \in \mathbb{N}^+$ and profit $(a_i) \in \mathbb{N}^+$ are specified. Moreover, we are given a knapsack **capacity** $B \in \mathbb{N}^+$. We are looking for a subset of objects whose total size is at most B and whose total profit is maximized.

NP-hard

Pseudopolynomial Algorithm

Let Π be an optimization problem whose instances are specified by discrete **objects** (for example sets, graphs, or strings) and **numbers** (such as costs, weights, profits). By |I|we denote (as usual) the size of the instance $I \in D_{\Pi}$ where all numbers in I are encoded in **binary**. By $|I_u|$ we denote the size of I when all numbers in I are encoded in **unary**.

- The running time of a polynomial algorithm for Π is polynomial in |I|.
- The running time of a **pseudo-polynomial algorithm** is polynomial in $|I_u|$
- The running time of a pseudo-polynomial algorithm is not always polynomial in $\left|I\right|$

$\label{eq:pseudopolynomial} Pseudopolynomial algorithm for ${\rm KNAPSACK}$$

- $P := \max_i \operatorname{profit}(a_i) \rightsquigarrow \operatorname{OPT} \le nP$
- For every i = 1,...,n and every p ∈ {1,...,nP} let S_{i,p} be a subset of {a₁,...,a_i} whose total profit is exactly p and whose total size is minimum among all subsets with these properties.
- A(i, p) denotes the total size of the set $S_{i,p}$ (we set $A(i, p) = \infty$ if such a set does not exist).
- If all A(i, p) are known then OPT can be determined by max{ p | A(n, p) ≤ B }

$\label{eq:polynomial} \textbf{Pseudo-Polynomial Algorithm for $KNAPSACK$}$

- A(1,p) is known for all $p \in \{0,\ldots,nP\}$
- We set $A(i,p) := \infty$ for p < 0
- $A(i+1,p) = \min\{A(i,p), \operatorname{size}(a_{i+1}) + A(i,p-\operatorname{profit}(a_{i+1}))\}$
- \rightsquigarrow All values A(i, p) and therefore OPT can be computed in $O(n^2P)$ time

KNAPSACK can be solved in pseudo-polynomial time $O(n^2P)$.

FPTAS for $\operatorname{KNAPSACK}$ via Scaling

- Running time O(n²P) polynomial in n, if P is polynomial in n
- FPTAS idea: Scale profits to polynomial size (depending on the required error parameter ϵ).

FPTAS for KNAPSACK via Scaling

KnapsackFPTAS(I,ϵ) $K \leftarrow \frac{\epsilon P}{n}$ profit' $(a_i) := \left\lfloor \frac{\text{profit}(a_i)}{K} \right\rfloor$ compute optimum solution S' for I with respect to profit'(\cdot) **return** S'

Lemma The solution S' satisfies $\operatorname{profit}(S') \ge (1 - \epsilon) \cdot \operatorname{OPT}$.

Theorem KnapsackFPTAS is an FPTAS for KNAPSACK with running time $O(n^3/\epsilon)$.

Strong NP-Hardness

An optimization problem is **strongly NP-hard**, if it remains NP-hard also with unary numbers.

Theorem A strongly NP-hard problem has no pseudo-polynomial algorithm unless P = NP.

FPTAS and Pseudo-Polynomial Algorithms

Theorem

Let p be a polynomial. Let Π be an NP-hard minimization problem with integer objective function and with $OPT(I) < p(|I_u|)$ for all instances I of Π . If Π admits an FPTAS then there is also a pseudo-polynomial algorithm for Π .

FPTAS und Strong NP-Hardness

Corollary

Let Π be an NP-hard optimization problem, that satisfies the requirements of the previous theorem. If Π is strongly NP-hard then there is no FPTAS for Π unless P = NP.