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## Exercises for Randomized and Approximation Algorithms

www.mpi-inf.mpg.de/departments/algorithms-complexity/teaching/winter18/rand-apx-algo/

## **Exercise Sheet 4: Dynamic Programming and PTAS's**

To be handed in by **November 13th, 2018** via e-mail to André Nusser (CC to Antonios Antoniadis and Marvin Künnemann)

**Exercise 1** (12 Points) In the weighted interval scheduling problem we are given a set J of n jobs, where each job j comes with a starttime  $s_i$ , a finishing time  $f_i$  and some value  $v_i$  (you may assume that all these values are integers). A feasible solution to the problem is a subset  $S \subseteq J$  of the jobs, so that no two jobs in S overlap, in other words, for any  $i, j \in S$ ,  $[s_i, f_i] \cap [s_j, f_j] = \emptyset$ . Goal is to find such a feasible set S of jobs of maximum total value  $\sum_{j \in S} v_j$ .

- (i) Show that the following two greedy algorithms can produce solutions that are arbitrarily worse than the optimal one.
  - *starttime-based greedy:* Go through the jobs in order of ascending starttimes, while adding them to S if and only if they do not overlap with any other job already in S.
  - weight-based-greedy: Go through the jobs in order of decreasing values, while adding them to S if and only if they do not overlap with any other job already in S.

(5 Points)

(ii) Give an optimal, polynomial-time algorithm for the problem using dynamic programming.
(7 Points) (In case you have difficulties solving this exercise, a pseudopolynomial-time dynamic programming algorithm gives part of the points.)

**Exercise 2** (8 Points) Consider the following greedy algorithm for the knapsack problem. Assume that the items are indexed in order of non-increasing ratio of value to size, i.e.,  $v_1/s_1 \ge v_2/s_2 \ge \cdots \ge v_n/s_n$ , and let  $i^*$  be the index of an item of maximum value, i.e.,  $v_{i^*} = \max_{i \in I} v_i$ . The algorithm identifies the largest k so that  $\sum_{i=1}^k s_i \le B$ . It then outputs either  $\{1, 2, \ldots, k\}$  or  $\{i^*\}$ , whatever has greater value. Prove that this algorithm is a 2-approximation algorithm for the knapsack problem.

**Exercise 3** (10 Points) Suppose we are given a directed acyclic graph with a specified source vertex s and a sink vertex t, and each edge e has an associated cost  $c_e$  and length  $\ell_e$ . You

may assume that  $c_e$  and  $\ell_e$  are positive integers. Give a fully polynomial-time approximation scheme for the problem of finding a minimum-cost path from s to t of total length at most L.

**Exercise 4** (10 Points) Consider some minimization problem  $\Pi$  such that:

- any feasible solution has a non-negative, integer objective function value, and
- there is some polynomial p, such that if it takes n bits to encode the input instance I in unary, OPT(I) < p(n).

Prove that if there is a fully polynomial-time approximation scheme for  $\Pi$ , then there is a pseudopolynomial algorithm for  $\Pi$ .

Note: Since there is no pseudopolynomial-time algorithm for a strongly NP-hard problem unless P=NP, you proved that unless P=NP there cannot be any such problem  $\Pi$  that is strongly NP-hard.