



Antonios Antoniadis and Marvin Künnemann

Winter 2018/19

## Exercises for Randomized and Approximation Algorithms

[www.mpi-inf.mpg.de/departments/algorithms-complexity/teaching/winter18/rand-apx-algo/](http://www.mpi-inf.mpg.de/departments/algorithms-complexity/teaching/winter18/rand-apx-algo/)

### Exercise Sheet 9: The Primal-Dual Method I

To be handed in by **December 18th, 2018** via e-mail to André Nusser  
(CC to Antonios Antoniadis and Marvin Künnemann)

**Exercise 1** (8 Points) Consider the following linear program:

$$\begin{array}{ll} \text{minimize} & -5x_1 + 8x_2 + 4x_3 \\ \text{subject to} & x_1 + x_2 = 2 \\ & x_2 - x_3 \leq 3 \\ & 2x_1 - x_3 \geq -1 \\ & x_1 \geq 0 \\ & x_3 \leq 0 \end{array}$$

- (i) Formulate the dual of the linear program. (3 Points)
- (ii) Rewrite the (primal) linear program, so that the constraints are in standard form ( $Ax \geq b$ ). (3 Points)

**Exercise 2** (11 Points) Consider the following maximization problem:

$$\begin{array}{ll} \text{maximize} & c^T x \\ \text{subject to} & Ax \leq b, \\ & x \geq 0 \\ & x \in \mathbb{R}^n \end{array}$$

and its corresponding dual:

$$\begin{array}{ll} \text{minimize} & b^T y \\ \text{subject to} & A^T y \geq c, \\ & y \geq 0, \\ & y \in \mathbb{R}^m \end{array}$$

and consider some feasible solutions  $x^*$  and  $y^*$  for the respective problems. Assume that there exist  $\lambda, \mu > 0$  so that the following approximate complementary slackness conditions hold:

$$x_i^* > 0 \Rightarrow \sum_{j=1}^m a_{ij} y_j^* \leq \mu c_i,$$

and

$$y_i^* > 0 \Rightarrow \sum_{i=1}^n a_{ij} x_i^* \geq \lambda b_j.$$

Prove that  $x^*$  is a  $\frac{\lambda}{\mu}$ -approximation for the primal.

**Exercise 3** (21 Points) The *local ratio* technique is closely related to the primal dual method; however, it only uses duality in an implicit way. Consider the following local ratio algorithm for the set cover problem. We compute a collection  $I$  of indices of a set cover, where  $I$  is initially empty. In each iteration, we find some element  $e_i$  not covered by the current collection  $I$ . Let  $\epsilon$  be the minimum weight of any set containing  $e_i$ . We subtract  $\epsilon$  from the weight of each set containing  $e_i$ ; some such set has now weight zero and we add its index to  $I$ .

Let  $\epsilon_j$  be the value of  $\epsilon$  in the  $j$ 'th iteration of the algorithm.

- (i) Show that the cost of the solution returned is at most  $f \sum_j \epsilon_j$ . (3 Points)
- (ii) Show that the cost of the optimal solution must be at least  $\sum_j \epsilon_j$ . (3 Points)
- (iii) Conclude that the algorithm is an  $f$ -approximation algorithm. (3 Points)

The local ratio technique depends upon the *local ratio theorem*. For a minimization problem  $\Pi$  with weights  $w$ , we say that a feasible solution  $F$  is  $\alpha$ -approximate with respect to  $w$  if the weight of  $F$  is at most  $\alpha$  times the weight of an optimal solution given weights  $w$ .

**Theorem 1** (Local Ratio Theorem). *If there are nonnegative weights  $w$  such that  $w = w^1 + w^2$ , where  $w^1$  and  $w^2$  are also nonnegative weights, and we have a feasible solution  $F$  such that  $F$  is  $\alpha$ -approximate with respect to both  $w^1$  and  $w^2$ , then  $F$  is  $\alpha$ -approximate with respect to  $w$ .*

- (iv) Prove the local ratio theorem. (5 Points)
- (v) Explain how the set cover algorithm above can be analyzed in terms of the local ratio theorem to prove that it is an  $f$ -approximation algorithm. (7 Points)