

**Advanced Algorithms**  
WS 2019  
**Homework Assignment 6**

**Problem 16:**

Let  $G = (V, E)$  be a graph with vertex set  $V = \{v_1, \dots, v_n\}$ . A *vertex cover* of  $G$  is a set  $C \subseteq V$  with the property that for every edge  $\{v_i, v_j\} \in E$ , either  $v_i$  or  $v_j$  is in  $C$ . In the VERTEXCOVER problem the task is to find the smallest possible vertex cover.

- (a) Formulate the vertex cover problem as an ILP, where variable  $x_i \in \{0, 1\}$  indicates whether node  $v_i$  is in the vertex cover or not.
- (b) Propose a randomized rounding strategy for the optimal solution of the LP relaxation to obtain (with probability at least  $1/2$ ) a feasible solution for the original ILP (no proof is needed here). Is there also a simple deterministic rounding strategy?

**Problem 17:**

MAX2SAT is the restriction of the MAXSAT problem to Boolean formulas in CNF that have clauses with at most 2 literals. The decision variant of the MAX2SAT problem is known to be NP-hard. Consider the following arithmetization of MAX2SAT.

For each Boolean variable  $x_i$  we define a variable  $y_i$  that can take the value  $-1$  or  $+1$ . In addition to that we have a variable  $y_0 \in \{-1, +1\}$  with the meaning that  $x_i$  is TRUE if and only if  $y_i$  and  $y_0$  have the same value.

For the arithmetization of a clause  $C$  with one literal, we distinguish between two cases:

- $C = x_i$ : use  $(1 + y_i \cdot y_0)/2$
- $C = \bar{x}_i$ : use  $(1 - y_i \cdot y_0)/2$

For clauses with two literals we can specify similar formulas.

1. Propose an arithmetization for all 4 possibilities for a clause with 2 literals.

Hint: given that  $a(\phi)$  is the arithmetization of a Boolean expression  $\phi$ ,  $a(x_i \vee x_j) = 1 - a(\bar{x}_i \wedge \bar{x}_j) = 1 - a(\bar{x}_i) \cdot a(\bar{x}_j)$ .

2. Formulate a quadratic program for MAX2SAT with the help of these arithmetizations.
3. Formulate a semidefinite program as a relaxation of the quadratic program.

4. Propose an approximation algorithm for MAX2SAT. Do you have an idea how to show that its approximation ratio is at most 1.139?

Hint: use the fact that  $\Pr[\text{sgn}(\vec{r}^T \cdot \vec{u}_i) = \text{sgn}(\vec{r}^T \cdot \vec{u}_j)] = 1 - \arccos(\vec{u}_i^T \cdot \vec{u}_j)/\pi$ .