

# CS261: Exercise Set #7

For the week of May 11–15, 2015

## Instructions:

- (1) *Do not turn anything in.*
- (2) The course staff is happy to discuss the solutions of these exercises with you in office hours or on Piazza.
- (3) While these exercises are certainly not trivial, you should be able to complete them on your own (perhaps after consulting with the course staff or a friend for hints).

## Exercise 31

In Lecture #13 we rounded a linear programming relaxation to obtain a 2-approximation algorithm for the Vertex Cover problem, where the vertices can have arbitrary nonnegative costs. In this exercise we'll consider the special case where every vertex cost  $c_v$  is 1. Prove that the following is a 2-approximation algorithm for this case:

1. Given the input graph  $G = (V, E)$ , compute a maximal matching  $M$  of  $G$ .<sup>1</sup>
2. Let  $S$  denote the  $2|M|$  endpoints of the edges of  $M$ , and return  $S$ .

## Exercise 32

Recall from Lecture #13 our linear programming relaxation of the Vertex Cover problem (with nonnegative edge costs):

$$\min \sum_{v \in V} c_v x_v$$

subject to

$$x_v + x_w \geq 1 \quad \text{for all edges } e = (v, w) \in E$$

and

$$x_v \geq 0 \quad \text{for all vertices } v \in V.$$

Prove that there is always a *half-integral* optimal solution  $\mathbf{x}^*$  of this linear program, meaning that  $x_v^* \in \{0, \frac{1}{2}, 1\}$  for every  $v \in V$ .

[Hint: start from an arbitrary feasible solution and show how to make it “closer to half-integral” while only improving the objective function value.]

## Exercise 33

Prove *Markov's inequality*: if  $X$  is a non-negative random variable with finite expectation and  $c > 1$ , then

$$\Pr[X \geq c \cdot \mathbf{E}[X]] \leq \frac{1}{c}.$$

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<sup>1</sup>A matching  $M$  is maximal if no strict superset of  $M$  is a matching. A maximum matching is always maximal, but a maximal matching need not be maximum.

### Exercise 34

Let  $X$  be a random variable with finite expectation and variance; recall that  $\text{Var}[X] = \mathbf{E}[(X - \mathbf{E}[X])^2]$  and  $\text{StdDev}[X] = \sqrt{\text{Var}[X]}$ . Prove *Chebyshev's inequality*: for every  $c > 1$ ,

$$\Pr[|X - \mathbf{E}[X]| \geq c \cdot \text{StdDev}[X]] \leq \frac{1}{c^2}.$$

[Hint: apply Markov's inequality to the (non-negative!) random variable  $(X - \mathbf{E}[X])^2$ .]

### Exercise 35

There are  $n$  identical bins. Consider the following random process (which is relevant for the analysis of hashing, among other things):

1. Repeat  $n$  independent times:
  - (a) Throw a ball into a bin chosen uniformly at random.

At the conclusion of this process, the expected number of balls in a given bin is 1 (why?). Prove that, with probability at least  $1 - \frac{1}{n}$ , no bin has more than  $\frac{3 \ln n}{\ln \ln n}$  balls in it.

[Hint: this is a special case of the randomized rounding analysis we did in Lecture #14.]