## COL 351: Analysis and Design of Algorithms

Instructor: Ragesh Jaiswal

1. (Weak Duality) A linear programming problem in the standard form can be written in short using the following vector notation:

**LP1**: Maximize 
$$(c^T \cdot x)$$
, subject to  $A \cdot x \le b$   $x \ge 0$ 

Here x, c, b are vectors and A is a  $m \times n$  matrix. Consider the following related linear program:

**LP2**: Minimize 
$$(b^T \cdot y)$$
, subject to  $A^T \cdot y \ge c$   $y \ge 0$ 

Here y is vector of size m. Consider the following linear program:

Let  $f_1$  be any feasible solution for LP1 and let  $f_2$  be any feasible solution for LP2. Show that  $f_1 \leq f_2$ .

2. (Randomized rounding) Linear programming can be used to obtain approximation algorithms for NP-hard problems. Consider the Set cover problem. Here we are given subsets  $S_1, ..., S_m \subseteq U$  and we want to find the minimum sized subset  $I \subseteq \{S_1, ..., S_m\}$  such that the union of elements of subsets in S is equal to U. We may solve the following ILP to obtain a solution to the problem:

Minimize 
$$\sum_{i=1}^{m} x_i$$
,  
Subject to:  
 $\sum_{i:e \in S_i} x_i \ge 1$ , for all  $e \in U$   
 $x_i \in \{0,1\}$ , for all  $x_i$ 

The issue here is that we know ILP is NP-hard. We relax the integer constraint to obtain the following LP:

Minimize 
$$\sum_{i=1}^{m} x_i$$
,  
Subject to :  
 $\sum_{i:e \in S_i} x_i \ge 1$ , for all  $e \in U$   
 $x_i \ge 0$ , for all  $x_i$ 

Suppose the optimal solution of the above LP be  $x_1^*, ..., x_m^*$ . Now we pick the set cover using the following randomized algorithm:

- 1.  $I \leftarrow \phi$
- 2. For all  $i: I \leftarrow I \cup S_i$  with probability  $x_i^*$ .
- 3. Repeat (2) until all elements are covered.

## Show the following:

- (a) The number of times step (2) is executed is at at most  $(2 \log n)$  with probability at least (1 1/n).
- (b) The above randomized algorithm with high probability gives an  $O(\log n)$  approximation to the set cover problem.