

## Homework 7

### Problem 14.1

Modify algorithm 14.1 so that it picks all sets that are nonzero in the fractional solution (we call the modified algorithm  $R$ ). Show that the algorithm also achieves a factor of  $f$ .

We first give the LP relaxation and its dual for set cover.

Primal:

$$\text{minimize} \quad \sum_{S \in \mathcal{S}} c(S)x_S \tag{1}$$

$$\text{s.t.} \quad \forall e \in U, \sum_{S: e \in S} x_S \geq 1 \tag{2}$$

$$\forall S \in \mathcal{S}, x_S \geq 0 \tag{3}$$

Dual:

$$\text{maximize} \quad \sum_{e \in U} y_e \tag{4}$$

$$\text{s.t.} \quad \forall S \in \mathcal{S}, \sum_{e: e \in S} y_e \leq c(S) \tag{5}$$

$$\forall e \in U, y_e \geq 0 \tag{6}$$

Suppose we have an optimal solution  $y^* = \sum_{e \in U} y_e$  for the dual problem. We know that this objective value is the sum of the  $y_e$  values for each element in the universe. We also know that each constraint in the dual corresponds to a set; specifically, each constraint says that all elements within a set must be less than or equal to the cost of taking that set.

If we summed up all the constraints of the dual, we would get the following sum:

$$\sum_{S \in \mathcal{S}} \sum_{e \in S} y_e \tag{7}$$

This is similar to the objective value for the dual, except that it recounts elements that appear in multiple sets. However, by definition of  $f$ , each element can be obtained in at most  $f$  sets, so the most the objective value of the dual could be is

$$OPT_D^* = \sum_{e \in U} y_e \leq \frac{\sum_{S \in \mathcal{S}} \sum_{e \in S} y_e}{f} \tag{8}$$

We will multiply both sides of this inequality by  $f$  to get

$$f * OPT_D^* \leq \sum_{S \in \mathcal{S}} \sum_{e \in S} y_e \quad (9)$$

Given an optimal primal solution  $x^*$ , for every nonzero value  $x_j^*$ , we know from the primal complementary slackness conditions that  $\forall \{S : x_s^* > 0\}, \sum_{e \in S} y_e = c(S)$ . If we break down the sum of all dual constraints into two disjoint sets (based on whether or not the constraint's corresponding  $x_s^*$  value is equal to 0), we have:

$$\sum_{S \in \mathcal{S}} \sum_{e \in S} y_e = \sum_{S: x_s^*=0} \sum_{e \in S} y_e + \sum_{S: x_s^*>0} \sum_{e \in S} y_e \quad (10)$$

Getting rid of the first term on the RHS (which can't be negative), and plugging in the known cost from complementary slackness conditions, we have

$$\sum_{S \in \mathcal{S}} \sum_{e \in S} y_e \geq \sum_{S: x_s^*>0} \sum_{e \in S} y_e = \sum_{S: x_s^*>0} c(S) \quad (11)$$

The RHS of this equation is the output value of our modified algorithm  $R$ , because for every  $x_s^* > 0$ , we round  $x_s^*$  up to 1 and pay the entire  $c(s)$ . So we know that

$$output(R) = \sum_{S: x_s^*>0} c(S) \leq \sum_{S \in \mathcal{S}} \sum_{e \in S} y_e \leq f * OPT_D^* \quad (12)$$

Because of LP duality theorem, and because the optimal value of the relaxed LP cannot be larger than the optimal value of the integer problem,

$$output(R) \leq f * OPT_D^* = f * OPT_P^* \leq f * OPT \quad (13)$$