## Lecture Outline:

- Primal-dual schema
- Primal-dual for set cover

This lecture describes the primal-dual schema, and also an application of this method to design approximation algorithms.

## 1 Primal-dual schema

In a previous lecture we have introduced the concept of duality for LP problems. A common form of the LP problem with its duality is showed below:

$$Primal$$
  $Dual$  
$$\min \sum_{i} c_{i}x_{i} \qquad \max \sum_{j} b_{j}y_{j}$$
  $s.t.$  
$$\sum_{i} a_{ij}x_{i} \geq b_{j} \qquad s.t.$$
 
$$\sum_{j} a_{ij}y_{j} \leq c_{i}$$
 
$$x_{i} \geq 0 \qquad y_{j} \geq 0$$

From the principle of duality, a feasible solution of the dual in fact sets a lower bound for the primal problem. We here use a concept of **complementary slackness** to help us illustrate the connection between the two solutions of the primal and duality problem:

**Definition 1. The complementary slackness** condition is as follows:

Primal : 
$$x_i > 0 \Rightarrow \sum_j a_{ij} y_j = c_i$$
  
Dual :  $y_j > 0 \Rightarrow \sum_j a_{ij} x_i = b_j$ 

Then the following theorem will show this connection:

**Theorem 1.** If (x, y) satisfies **complementary slackness**, then x and y are optimal solutions for primal and dual problems, respectively.

**Proof:** From the forms of the primal and duality, if complementary slackness is satisfied, we will have

$$\sum_{i} c_{i} x_{i} = \sum_{i} \left(\sum_{j} a_{ij} y_{j}\right) * x_{i}$$

$$\sum_{j} b_{j} y_{j} = \sum_{j} \left(\sum_{i} a_{ij} x_{i}\right) * y_{j}$$

It is easy to see that the RHS of the two equations are equal. So we have

$$\sum_{i} c_i x_i = \sum_{j} b_j y_j$$

implying that x and y are both optimal by duality.

So if we obtain an LP solution that satisfies the complementary slackness, then the solution is optimal. For integer LPs in general, however, it is unlikely that the optimal solution is integral. We apply the complementary slackness approach to approximation algorithms by defining **relaxed slackness** as follows:

## Definition 2.

Primal : 
$$x_i > 0 \Rightarrow \frac{c_i}{\alpha} \le \sum_j a_{ij} y_j \le c_i$$

Dual :  $y_j > 0 \Rightarrow b_j \le \sum_i a_{ij} x_i \le \beta b_j$ 

And accordingly, we will fit 1 into the approximate situation with the description below:

**Theorem 2.** If (x, y) satisfies relaxed complementary slackness, then x and y are  $\alpha\beta$ -optimal for both primal and dual problem.

**Proof:** Based on the definition of relaxed complementary slackness, we will have

$$\sum_{i} c_i x_i \leq \alpha \sum_{i} (\sum_{j} a_{ij} y_j) * x_i \tag{1}$$

$$\sum_{j} b_{j} y_{j} \geq \frac{1}{\beta} \sum_{j} \left( \sum_{i} a_{ij} x_{i} \right) * y_{j} \tag{2}$$

Then we can get

$$\frac{\text{cost of x}}{\text{cost of y}} = \frac{\sum_{i} c_{i} x_{i}}{\sum_{j} b_{j} y_{j}} \le \alpha \beta$$

which indicates that the duality yields a  $\alpha\beta$ -approximation solution.

Based on the property of the relaxed complementary slackness, the primal-dual schema for LP solution can be done in an iterative process: start with a dual feasible solution, then we try to improve this dual problem, until the improved solution satisfies the relaxed complementary slack conditions.

## 2 Primal-dual for set cover

In this section we will apply the primal-dual schema to the set cover problem. Recall the primal problem of set cover with universe U of elements and collection C of sets, and its dual:

$$Primal \qquad Dual$$
 
$$\min \sum_{S \in C} c(S)x_S \qquad \max \sum_{e \in U} y_e$$
 
$$s.t. \quad \sum_{S: \ e \in S} x_S \ge 1, \ \forall e \in U \qquad \qquad s.t. \quad \sum_{e \in s} y_e \le c(S), \ \forall S \in C$$
 
$$x_j \ge 0, \ \forall j \qquad \qquad y_e \ge 0, \ \forall e$$

As mentioned above, we will take an iterative process to get the approximate solution. First of all, we start with a primal unfeasible but dual feasible solution

$$x = 0, y = 0$$

The iterative algorithm is as following:

- While any element is uncovered
  - 1. Pick any uncovered elements e
  - 2. Increase  $y_e$  until some set S become tight, i.e.,  $\sum_{e \in S} y_e = c(S)$ .
  - 3. Add S to solution.

According to the definition of the relaxed complementary slackness, we will have

$$x_S > 0 \Rightarrow \sum_{e \in S} y_e = c(S)$$
  $\alpha = 1$  (3)

$$y_e > 0 \Rightarrow \sum_{S: e \in S} x_S \le f_e \qquad \beta = \max_e f_e$$
 (4)

Here we let  $f_e$  be the number of the sets that contain e. So from the above analysis we see that the primal-dual algorithm will yield an f-approximate solution, where f is the maximum frequency of any element.