

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:

$$\begin{array}{ll}
\text{Primal} & \text{Duality} \\
\min \sum_i c_i x_i & \max \sum_j b_j y_j \\
\text{s.t. } \sum_i a_{ij} x_i \geq b_j & \text{s.t. } \sum_j a_{ij} y_j \leq c_i \\
x_i \geq 0 & y_j \geq 0
\end{array}$$

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:

$$\begin{aligned}
\text{Primal} : x_i > 0 \Rightarrow \sum_j a_{ij} y_j = c_i \\
\text{Dual} : y_j > 0 \Rightarrow \sum_i a_{ij} x_i = b_j
\end{aligned}$$

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

$$\begin{aligned}\sum_i c_i x_i &= \sum_i (\sum_j a_{ij} y_j) * x_i \\ \sum_j b_j y_j &= \sum_j (\sum_i a_{ij} x_i) * y_j\end{aligned}$$

It is easy to see that the RHS of the two inequations 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 we can use the complementary slackness to solve LP problems for an optimal solution. While for integer LP, since it is NP-Hard, we need to get an approximately optimal solution in polynomial time. In order to apply the complementary slackness theorem into the approximate situation, we define a **relaxed slackness** as follows:

Definition 2.

$$\begin{aligned}Primal &: x_i > 0 \Rightarrow \frac{c_i}{\alpha} \leq \sum_j a_{ij} y_j \leq c_i \\ Dual &: y_j > 0 \Rightarrow b_j \geq \sum_i a_{ij} x_i \geq \beta b_j\end{aligned}$$

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

$$\sum_i c_i x_i \leq \alpha \sum_i (\sum_j a_{ij} y_j) * x_i \tag{1}$$

$$\sum_j b_j y_j \geq \frac{1}{\beta} \sum_j (\sum_i a_{ij} x_i) * y_j \tag{2}$$

Then we can get

$$\frac{\text{cost of } x}{\text{cost of } y} = \frac{\sum_i c_i x_i}{\sum_j b_j y_j} \leq \alpha\beta$$

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 reaches satisfied 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:

<i>Primal</i>	<i>Duality</i>
$\min \sum_{S \in C} c(S)x_S$	$\max \sum_{e \in U} y_e$
<i>s.t.</i> $\sum_{S: e \in S} x_S \geq 1, \forall e \in U$	<i>s.t.</i> $\sum_{e \in S} y_e \leq c(S), \forall S \in C$
$x_j \geq 0, \forall j$	$y_e \geq 0, \forall e$

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

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

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

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 get a f -approximate solution, where f is the maximum frequency of any element.