

Primal-dual schema

Complementary slackness conditions

$$\begin{array}{l|l} P: & \min \sum_{j=1}^n c_j x_j \\ & \text{s.t.} \sum_{j=1}^n a_{ij} x_j \geq b_i \\ & x_j \geq 0 \quad (j=1, \dots, n) \\ y_i & \\ \hline D: & \max \sum_{i=1}^m b_i y_i \\ & \text{s.t.} \sum_{i=1}^m a_{ij} y_i \leq c_j \\ & y_i \geq 0 \quad (i=1, \dots, m) \end{array}$$

Let $x = (x_1, \dots, x_n)$ and $y = (y_1, \dots, y_m)$ be a primal-feasible and a dual-feasible point respectively.

$$\begin{aligned} \sum_{i=1}^m b_i y_i &\leq \sum_{i=1}^m y_i \left(\sum_{j=1}^n a_{ij} x_j \right) = \sum_{i=1}^m \sum_{j=1}^n y_i a_{ij} x_j \\ &= \sum_{j=1}^n x_j \left(\sum_{i=1}^m a_{ij} y_i \right) \leq \sum_{j=1}^n c_j x_j \end{aligned}$$

Let $x^* = (x_1^*, \dots, x_n^*)$ & $y^* = (y_1^*, \dots, y_m^*)$ be a primal optimum point & a dual optimum point respectively.

$$\text{strong duality} \Rightarrow \sum_{i=1}^m b_i y_i^* = \sum_{j=1}^n c_j x_j^*$$

Thus, both the inequalities in the above chain are tight when $y = y^*$ & $x = x^*$.

$$\therefore \sum_{i=1}^m b_i y_i^* = \sum_{i=1}^m y_i^* \sum_{j=1}^n a_{ij} x_j^* \quad \text{--- (1), and}$$

$$\sum_{j=1}^n x_j^* \sum_{i=1}^m a_{ij} y_i^* = \sum_{j=1}^n c_j x_j^* \quad \text{--- (2)}$$

From (1) we have that,

$$\sum_{i=1}^m \underbrace{y_i^*}_{\geq 0} \cdot \underbrace{\left(b_i - \sum_{j=1}^n a_{ij} x_j^* \right)}_{\leq 0} = 0$$

≤ 0

$$\Rightarrow \forall i=1, \dots, m$$

$$\text{Either } y_i^* = 0 \text{ or } \sum_{j=1}^n a_{ij} x_j^* = b_i$$

Dual complementary slackness conditions

Either a dual variable is 0 or the corresponding primal constraint is tight.

Similarly from (2) we have.

$$\forall j=1, \dots, n$$

$$\text{Either } x_j^* = 0 \text{ or } \sum_{i=1}^m a_{ij} y_i^* = c_j$$

Primal complementary slackness conditions.

Related complementary slackness conditions.

Claim 1: Let $\alpha, \beta \geq 1$. Assume that x is primal feasible, y is dual-feasible, and
 $\forall j=1, \dots, n$

$$x_j \neq 0 \Rightarrow \sum_{i=1}^m a_{ij} y_i \geq \frac{c_j}{\alpha}$$

and,

$\forall i=1, \dots, m$

$$y_i \neq 0 \Rightarrow \sum_{j=1}^n a_{ij} x_j \leq b_i \cdot \beta$$

then,
$$\sum_{j=1}^n c_j x_j \leq \alpha \beta \cdot \sum_{i=1}^m b_i x_i$$

Proof:

$$\sum_{j=1}^n c_j x_j \leq \alpha \sum_{j=1}^n \sum_{i=1}^m a_{ij} y_i x_j$$

$x_j \neq 0$ $y_i \neq 0$

$$= \alpha \sum_{i=1}^m y_i \cdot \sum_{j=1}^n a_{ij} x_j$$

$$\leq \alpha \cdot \sum_{i=1}^m y_i \cdot \beta \cdot b_i$$

$$= \alpha \beta \cdot \sum_{i=1}^m b_i y_i$$

□

$$\sum_{j=1}^n c_j x_j \leq \alpha \beta \cdot \sum_{i=1}^m b_i y_i \leq \alpha \beta \cdot \sum_{i=1}^m b_i y_i^*$$

$$= \alpha \beta \cdot \sum c_j x_j^*$$

y^* : A dual optimal pt,
 x^* : A primal optimal point

Primal dual schema

- Assume that Π is a minimization problem that can be formulated as an integer program.
- Let P be a relaxation of that ILP, & let D be the dual of P .
- The algorithm ^{iteratively} constructs an integral feasible point of P (let's call it x) and a feasible point of D (let's call it y) such the pair (x, y) satisfy the relaxed complementary slackness conditions with parameters α, β .
- Claim 1 implies that x is an $\alpha\beta$ -approximate solution.

The set-cover problem

Input: $\mathcal{U} = \{1, \dots, n\}$, $\mathcal{S} = \{S_1, \dots, S_m\}$

P : $c: \mathcal{S} \rightarrow \mathbb{R}^{\geq 0}$.

$$\min \sum_{i=1}^m c(S_i) \cdot x_i \quad \Bigg| \quad \max \sum_{j=1}^n y_j$$

y_j

$$\text{st. } \sum_{i: j \in S_i} x_i \geq 1 \quad \forall j=1, \dots, n$$
$$\text{st. } \sum_{j \in S_i} y_j \leq c(S_i) \quad i=1, \dots, m$$
$$x_i \geq 0 \quad \forall i=1, \dots, m. \quad y_j \geq 0, \quad \forall j=1, \dots, n$$

A primal-dual f -approx. algo. for set cover

(f : max. frequency of an element)

$$x \in \mathbb{R}^m, y \in \mathbb{R}^n$$

- $x \leftarrow 0^m, y \leftarrow 0^n$

- while there exists an uncovered element j

- raise y_j until some dual constraint becomes tight.

- Pick all tight sets (i.e., set the corresponding entries in x to 1).

- Return x .

observe: $x_i \in \{0,1\} \forall i=1, \dots, m.$

② The x returned is primal feasible

The algo returns a valid set-cover

③ y is always dual feasible.

$$(y_j \neq 0) \Rightarrow \left(\sum_{i: j \in S_i} x_i \leq f \cdot 1. \right) \quad \underline{\underline{\beta = f}}$$

$(x_i \neq 0) \Rightarrow$ The algorithm has picked S_i

$\Rightarrow S_i$ is tight

$\Rightarrow \sum_{j \in S_i} y_j = 1$ in the iteration in which S_i is picked, and hence also later

$$\underline{\underline{\alpha = 1}}$$

$$\underline{\underline{\alpha \beta = f}}$$

\Rightarrow By claim 1, the algorithm is an f -approximation algorithm.