## CS602 Applied Algorithms

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## 1 Weak Duality

Let  $A \in \mathbb{R}^{m \times n}$ ,  $b \in \mathbb{R}^m$  and  $w \in \mathbb{R}^n$ .

**Primal LP** -  $\max w^{\mathsf{T}}x : Ax \leq b, x \in \mathbb{R}^n$ 

**Dual LP** - min  $b^{T}y : A^{T}y = w, y > 0, y \in \mathbb{R}^{m}$ 

Weak Duality:  $OPT(Primal LP) \leq OPT(Dual LP)$ .

**Proof:** Let  $\alpha$  be a feasible solution for the primal LP whereas  $\beta$  be a feasible solution for the dual LP.

$$A\alpha \leq b$$
  

$$\beta^{\mathsf{T}} A \alpha \leq \beta^{\mathsf{T}} b \quad (\because \beta \geq 0)$$
  

$$w^{\mathsf{T}} \alpha \leq \beta^{\mathsf{T}} b \quad (\because A^{\mathsf{T}} \beta = w)$$

The above inequality, in particular, is also true when  $\alpha, \beta$  are optimal solutions.

## 2 Strong Duality

Strong duality says that for optimal solutions the above inequality is, in fact, an equality. If LP is feasible and has a bounded optimal value, then there exists a dual feasible solution that has the same value as the primal optimal value.

Claim 8.1. Let  $\alpha^*$  be a optimal solution and  $w^* = w^{\mathsf{T}} \alpha^*$  be the optimal value for the primal LP. Then

$$\exists \beta^* \in \mathbb{R}^m \text{ s.t. } A^{\mathsf{T}}\beta^* = w, \ \beta^* \geq 0 \ (dual \ feasible), \ and \ b^{\mathsf{T}}\beta^* = w^*.$$

*Proof.* Recall the corollary from previous class, that gives a separating hyperplane for a cone. For any cone  $C = \text{cone}(v_1, v_2, ..., v_k) \in \mathbb{R}^n$  and for a vector  $u \in \mathbb{R}^n$ ,

$$u \notin C \implies \exists q \in \mathbb{R}^n \text{ s.t. } q^{\mathsf{T}}u > 0 \text{ and } q^{\mathsf{T}}v_i \leq 0, \ \forall 1 \leq i \leq k.$$

Consider the augmented matrix (A|b). We want to show that there is a  $\beta^* > 0$  such that

$$(A|b)^{\mathsf{T}}\beta^* = (w, w^*).$$

Here  $(w, w^*) \in \mathbb{R}^{n+1}$  denotes a column vector whose first n entries come from w and the last entry is  $w^*$ . Let  $(a_1, b_1), (a_2, b_2), \dots, (a_m, b_m) \in \mathbb{R}^{n+1}$  be the columns of  $(A|b)^T$ . What is want to show is equivalent to

$$(w, w^*) \in \text{cone}((a_1, b_1), (a_2, b_2), \dots, (a_m, b_m)).$$

For the sake of contradiction, let it not be true. Then by above mentioned corollary, there exist  $q \in \mathbb{R}^n$  and  $r \in \mathbb{R}$  such that

$$(q,r)^{\mathsf{T}}(w,w^*) > 0 \text{ and } (q,r)^{\mathsf{T}}(a_i,b_i) \le 0, \ \forall 1 \le i \le m.$$

Equivalently,

$$q^{\mathsf{T}}w > -rw^*$$
 and  $q^{\mathsf{T}}a_i \leq -rb_i, \ \forall 1 \leq i \leq m.$ 

Which is same as

$$w^{\mathsf{T}}q > -rw^* \text{ and } a_i^{\mathsf{T}}q \le -rb_i, \ \forall 1 \le i \le m.$$
 (1)

Now, we consider two cases depending on the sign of r. In both the cases, we will show that there is a primal feasible solution  $\alpha$  with value greater than  $w^*$ , contradicting the optimality of  $\alpha^*$ .

Case 1: r < 0. Divide (1) by -r and define  $\alpha = q/-r$ . We get

$$a_i^{\mathsf{T}} \alpha \leq b_i \ \forall i, \ \text{and} \ w^{\mathsf{T}} \alpha > w^*,$$

which is a contradiction.

Case 2:  $r \ge 0$ , which means r+1 > 0. Now, consider  $\alpha = (r+1)\alpha^* + q$ . Since  $\alpha^*$  is primal optimal, we know

$$(r+1)a_i^{\mathsf{T}}\alpha^* \leq (r+1)b_i, \ \forall i, \ \text{and} \ (r+1)w^{\mathsf{T}}\alpha^* = (r+1)w^*.$$

Combining this with (1), we get

$$a_i^{\mathsf{T}} \alpha = (r+1)a_i^{\mathsf{T}} \alpha^* + a_i^{\mathsf{T}} q \le (r+1)b_i - rb_i = b_i, \ \forall i.$$

And

$$w^{\mathsf{T}}\alpha = (r+1)w^{\mathsf{T}}\alpha^* + w^{\mathsf{T}}q > (r+1)w^* - rw^* = w^*.$$

This gives us a contradiction.

## 3 Various Forms of LP

Primal	Dual
$\max \ w^{T}x : Ax \le b$	$\min b^{T}y : A^{T}y = w, \ y \ge 0$
$\max w^{T}x : Ax = b, \ x \ge 0$	$\min \ b^{T}y : A^{T}y \ge w$
$\max w^{T}x : Ax \le b, \ x \ge 0$	$\min b^{T}y : A^{T}y \ge w, \ y \ge 0$