

# Numerical Optimization, 2023 Fall

## Homework 3

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Problem 1. Prove the dual of the dual of a linear programming (standard form) is itself. [25pts]

We can prove this with the help of the Duality Scheme.

Consider a linear programming that is in standard form:

$$\begin{aligned} \min_{\mathbf{x}} \quad & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} \quad & A\mathbf{x} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned} \tag{1}$$

The Lagrangian of the above linear programming is  $L(\mathbf{x}, \boldsymbol{\lambda}) = \mathbf{c}^T \mathbf{x} - \boldsymbol{\lambda}^T (A\mathbf{x} - \mathbf{b})$ .

Since for the primal question, the variables are  $\mathbf{x} \geq \mathbf{0}$ , so for  $x_i \geq 0$ , the dual constrain is that:

$$\mathbf{a}_i^T \boldsymbol{\lambda} \leq c_i$$

where  $\mathbf{a}_i$  is the  $i$ -th column of  $A$ .

So for the dual problem, the constraints are  $A^T \boldsymbol{\lambda} \leq \mathbf{c}$ .

And since for the primal question, the constrain it that  $A\mathbf{x} = \mathbf{b}$ , i.e.  $\sum_{j=1}^n a_{ij}x_j = b_i$ .

So for the dual question, the variables  $\boldsymbol{\lambda}$  is free.

So the dual problem is that:

$$\begin{aligned} \max_{\boldsymbol{\lambda}} \quad & \boldsymbol{\lambda}^T \mathbf{b} \\ \text{s.t.} \quad & A^T \boldsymbol{\lambda} \leq \mathbf{c} \\ & \boldsymbol{\lambda} \text{ is free} \end{aligned} \tag{2}$$

To easily get the dual of the dual problem, we can first convert the objective function of the dual problem to a minimization problem. And take  $M = A^T$ , the dual problem becomes:

$$\begin{aligned} \min_{\boldsymbol{\lambda}} \quad & \boldsymbol{\lambda}^T (-\mathbf{b}) \\ \text{s.t.} \quad & M\boldsymbol{\lambda} \leq \mathbf{c} \\ & \boldsymbol{\lambda} \text{ is free} \end{aligned} \tag{3}$$

The Lagrangian of the dual problem is:

$$L(\boldsymbol{\lambda}, \mathbf{y}) = \boldsymbol{\lambda}^T (-\mathbf{b}) - \mathbf{y}^T (M\boldsymbol{\lambda} - \mathbf{c})$$

Since for the dual question, the variables are  $\boldsymbol{\lambda}$  is free, so for  $\lambda_i$  is free, the dual of the dual constrain is that:

$$m_i^T \boldsymbol{\lambda} = (-\mathbf{b})_i = -b_i$$

where  $m_i$  is the  $i$ -th column of  $M$ .

So for the dual problem, the constraints are

$$M^T \mathbf{y} = -\mathbf{b}$$

And since for the dual question, the constraint is that  $M\boldsymbol{\lambda} \leq \mathbf{c}$ , i.e.  $\sum_{j=1}^n m_{ij}\lambda_j \leq c_i$ .  
so for the dual question, the variables are  $\mathbf{y} \leq \mathbf{0}$ .

So the dual of the dual problem is that:

$$\begin{aligned} \max_{\mathbf{y}} \quad & \mathbf{c}^T \mathbf{y} \\ \text{s.t.} \quad & M^T \mathbf{y} = -\mathbf{b} \\ & \mathbf{y} \leq \mathbf{0} \end{aligned} \tag{4}$$

We can take  $\mathbf{x} = -\mathbf{y}$ . And since  $M = A^T$ , so  $M^T = (A^T)^T = A$ .

Consider the objective function is

$$\max_{\mathbf{y}} \quad \mathbf{c}^T \mathbf{y}$$

We can convert it into a minimization problem by taking  $-\mathbf{y}$  as the variable.  
i.e.

$$\min_{\mathbf{y}} \quad \mathbf{c}^T (-\mathbf{y})$$

And since we have  $\mathbf{x} = -\mathbf{y}$ , so we can convert the above minimization problem into:

$$\min_{\mathbf{x}} \quad \mathbf{c}^T \mathbf{x}$$

And for the first constraint, we have:

$$M^T \mathbf{y} = -\mathbf{b}$$

Since  $M^T = A$ , and move the "-" from the right to the left, we have:

$$A(-\mathbf{y}) = \mathbf{b}$$

i.e.

$$A\mathbf{x} = \mathbf{b}$$

For the second constraint, we have:

$$\mathbf{y} \leq \mathbf{0}$$

We can convert it into:

$$-\mathbf{y} \geq \mathbf{0}$$

i.e.

$$\mathbf{x} \geq \mathbf{0}$$

So with the conversions above, we can get that the dual of the dual problem is that:

$$\begin{aligned} \min_{\mathbf{x}} \quad & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} \quad & A\mathbf{x} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned} \tag{5}$$

Which is exactly same with the primal problem.

So above all, the dual of the dual of a linear programming (standard form) is itself.

Problem 2. Prove the dual objective increases after a pivot of the dual simplex method. [25pts]

Consider the dual simplex method.

Suppose that the current state is feasible, and after the pivot, it is also feasible.

Then for the pivot process, we have:

$r^T \geq 0$  always satisfies, this is to make sure the dual problem is feasible.

As for choosing the pivot element, we choose the  $p$ -th row s.t. the current  $b_p$  in the tableau is negative.

i.e.  $b_p < 0$ .

Suppose that the  $p$ -th row is  $y_{p1}, y_{p2}, \dots, y_{pn}, b_p$ , and we choose the pivot element  $a_{pq}$ .

s.t.  $\hat{\epsilon} = \frac{r_q}{-y_{pq}} = \min\left\{\frac{r_i}{-y_{pi}} \mid a_{pi} < 0, i = 1, \dots, n\right\}$ .

Then we pivot with  $y_{pq}$ , and we need to update the tableau to make the  $r_q$  become 0 by:

Let the last line of the simplex tableau adds  $\hat{\epsilon}$  times the  $p$ -th row.

So we have

$$\begin{aligned} r'_q &= r_q + \hat{\epsilon}y_{pq} = r_q + \frac{r_q}{-y_{pq}}y_{pq} = 0 \\ -f' &= -f + \hat{\epsilon}b_p = -f + \frac{r_q}{-y_{pq}}b_p \end{aligned}$$

We know that the lower-right corner is  $-f$ , where  $f$  is the dual objective value of the dual problem.

So after the pivot, the lower-right corner becomes:

$$-f' = -f - \frac{r_q}{y_{pq}}b_p$$

Since  $r_q > 0, y_{pq} < 0, b_p < 0$ , so  $\frac{r_q}{y_{pq}}b_p > 0$  So

$$-f = -f - \frac{r_q}{y_{pq}}b_p < -f'$$

i.e.

$$f' > f$$

So above all, the dual objective increases after a pivot of the dual simplex method.

Problem 3. Let  $L(\mathbf{x}, \boldsymbol{\lambda})$  be the Lagrangian of a linear programming problem, and  $(\mathbf{x}^*, \boldsymbol{\lambda}^*)$  be the optimal primal-dual solution. Prove that

$$L(\mathbf{x}, \boldsymbol{\lambda}^*) \geq L(\mathbf{x}^*, \boldsymbol{\lambda}^*) \geq L(\mathbf{x}^*, \boldsymbol{\lambda}),$$

for any primal feasible  $\mathbf{x}$  and dual feasible  $\boldsymbol{\lambda}$ . [25pts]

Suppose that the primal problem is that:

$$\begin{aligned} \min_{\mathbf{x}} \quad & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} \quad & A\mathbf{x} = \mathbf{b} \\ & \mathbf{x} \geq 0 \end{aligned} \tag{6}$$

And the dual problem is that:

$$\begin{aligned} \max_{\boldsymbol{\lambda}} \quad & \boldsymbol{\lambda}^T \mathbf{b} \\ \text{s.t.} \quad & A^T \boldsymbol{\lambda} \leq \mathbf{c} \end{aligned} \tag{7}$$

And the Lagrangian of the primal problem is that:

$$L(\mathbf{x}, \boldsymbol{\lambda}) = \mathbf{c}^T \mathbf{x} + \boldsymbol{\lambda}^T (A\mathbf{x} - \mathbf{b})$$

Since we can considering the primal feasible  $\mathbf{x}$  and dual feasible  $\boldsymbol{\lambda}$ , so we have:

$$A\mathbf{x} = \mathbf{b}$$

$$\mathbf{c}^T \mathbf{x} \geq \mathbf{c}^T \mathbf{x}^*$$

If we put  $A\mathbf{x} - \mathbf{b} = \mathbf{0}$  into the Lagrangian, we have:

$$L(\mathbf{x}, \boldsymbol{\lambda}) = \mathbf{c}^T \mathbf{x} + \boldsymbol{\lambda}^T (A\mathbf{x} - \mathbf{b}) = \mathbf{c}^T \mathbf{x} + \boldsymbol{\lambda}^T \mathbf{0} = \mathbf{c}^T \mathbf{x}$$

And since  $\mathbf{c}^T \mathbf{x} \geq \mathbf{c}^T \mathbf{x}^*$ , and  $L(\mathbf{x}, \boldsymbol{\lambda})$  do not contain  $\boldsymbol{\lambda}$ , so we have:

$$L(\mathbf{x}, \boldsymbol{\lambda}) \geq L(\mathbf{x}^*, \boldsymbol{\lambda}) = L(\mathbf{x}^*, \boldsymbol{\lambda})$$

So above all, we have prove that

$$L(\mathbf{x}, \boldsymbol{\lambda}^*) \geq L(\mathbf{x}^*, \boldsymbol{\lambda}^*) \geq L(\mathbf{x}^*, \boldsymbol{\lambda})$$

Problem 4. Construct a linear programming problem for which both the primal and the dual problem has no feasible solution. [25pts]

Construct a linear programming problem that is:

$$\begin{aligned}
 \min_{x_1, x_2} \quad & x_1 - 2x_2 \\
 \text{s.t.} \quad & x_1 - x_2 \leq 1 \\
 & x_1 - x_2 \geq 2 \\
 & x_1, x_2 \leq 0
 \end{aligned} \tag{8}$$

Since it is impossible to satisfy  $x_1 - x_2 \leq 1$  and  $x_1 - x_2 \geq 2$  at the same time, so the primal problem has no feasible solution.

And the dual problem is that:

$$\begin{aligned}
 \max_{\lambda_1, \lambda_2} \quad & \lambda_1 + 2\lambda_2 \\
 \text{s.t.} \quad & \lambda_1 + \lambda_2 \leq 1 \\
 & -\lambda_1 - \lambda_2 \leq -2 \\
 & \lambda_1 \leq 0, \lambda_2 \geq 0
 \end{aligned} \tag{9}$$

The second constrain  $-\lambda_1 - \lambda_2 \leq -2$  can be written as  $\lambda_1 + \lambda_2 \geq 2$ .

Since it is impossible to satisfy  $\lambda_1 + \lambda_2 \leq 1$  and  $\lambda_1 + \lambda_2 \geq 2$  at the same time, so the dual problem has no feasible solution.

So above all, the above construction's primal and dual problem has no feasible solution.