

# Optimization

## Homework 10

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**Exercise 1** Use duality to check if  $x = (1, 0, 1, 0)$  is an optimal solution of the following linear program.

$$\begin{array}{ll} \min & -x_1 + 2x_2 - x_3 - x_4 \\ \text{s.t.} & x_1 + x_2 - x_3 + 2x_4 \geq -2 \\ & x_1 + 2x_2 - x_3 + x_4 = 0 \\ & -x_1 - x_2 - x_3 - x_4 \geq -2 \\ & x_1, x_2, x_3 \geq 0 \\ & x_4 \text{ free} \end{array}$$

First we do a bit of work to reduce the problem. Note  $x_4$  is free and is a term in the equality constraint. So we replace  $x_4$  with  $x_4 = -x_1 - 2x_2 + x_3$  everywhere in the problem.

$$\begin{array}{ll} \min & 4x_2 - 2x_3 \\ \text{s.t.} & x_1 + 3x_2 - x_3 \leq 2 \\ & -x_2 + 2x_3 \leq 2 \\ & x_1, x_2, x_3 \geq 0 \end{array}$$

We notice an interesting feature of this linear program. The variable  $x_1$  appears only in the first constraint. Suppose we have some feasible point  $\hat{x}$  with  $\hat{x}_1 + 3\hat{x}_2 - \hat{x}_3 < 2$ . We know that  $\hat{x}_1 \geq 0$  because it is feasible. We can increase  $\hat{x}_1$  without changing this fact. So we increase  $\hat{x}$  to  $\hat{x} + \delta > 0$  so that  $\hat{x}_1 + \delta + 3\hat{x}_2 - \hat{x}_3 = 2$  without affecting any other constraints or the value of the objective function. That is,  $x_1$  acts as a slack variable in this problem. We will write  $\bar{x}_1$  to represent  $x_1 + \delta$  to emphasize the special nature of this variable. For the second constraint, we do need to introduce a separate slack variable to put this into standard form. Then the problem becomes

$$\begin{array}{ll} \min & 4x_2 - 2x_3 \\ \text{s.t.} & \bar{x}_1 + 3x_2 - x_3 = 2 \\ & -x_2 + 2x_3 + u = 2 \\ & \bar{x}_1, x_2, x_3, u \geq 0 \end{array}$$

So the point in question,  $(1, 0, 1, 0)$  has  $\delta = 2$  and  $u = 0$ , so that  $\bar{x}_1 = 1 + 2 = 3$ . We can see this when we

manipulate the tableau corresponding to this LP.

$$T_0 = \left[ \begin{array}{c|cccc} 0 & 0 & 4 & -2 & 0 \\ \hline 2 & 1 & 3 & -1 & 0 \\ 2 & 0 & -1 & 2 & 1 \end{array} \right]$$

$$T_1 = \left[ \begin{array}{ccc} 1 & 0 & 1 \\ 0 & 1 & 0.5 \\ 0 & 0 & 0.5 \end{array} \right] T_0 = \left[ \begin{array}{c|cccc} 2 & 0 & 3 & 0 & 1 \\ \hline 3 & 1 & 2.5 & 0 & 0.5 \\ 1 & 0 & -0.5 & 1 & 0.5 \end{array} \right]$$

This tableau corresponds to  $\bar{x}_1 = x_1 + \delta = 3$ ,  $x_2 = 0$ ,  $x_3 = 1$ , and  $u = 0$ . This then corresponds to a parameterized family of points to the original problem;  $(x_1, x_2, x_3, x_4) = (3 - \delta, 0, 1, \delta - 2)$  for all  $0 \leq \delta \leq 3$ .

But all of that was an aside. We want to use duality in this problem. First we write down the dual problem for reference.

$$\begin{aligned} \max \quad & 2w_1 + 2w_2 \\ \text{s.t.} \quad & w_1 \leq 0 \\ & 3w_1 - w_2 \leq 4 \\ & -w_1 + 2w_2 \leq -2 \\ & w_1, w_2 \leq 0 \end{aligned}$$

We want  $x_1$  and  $x_3$  to be basic variables. Take

$$B = \begin{bmatrix} 1 & -1 \\ 0 & 2 \end{bmatrix} \quad \text{and} \quad B^{-1} = \frac{1}{2} \begin{bmatrix} 2 & 1 \\ 0 & 1 \end{bmatrix}.$$

Since  $b = (2, 2)^\top$ , then

$$B^{-1}b = \frac{1}{2} \begin{bmatrix} 2 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

which is what we got with the tableau manipulations above. Also define

$$w = B^{-\top} c_B = \frac{1}{2} \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ -2 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}.$$

The cost at this  $w$  is  $2w_1 + 2w_2 = -2$ , which is the same as the primal cost of  $(1, 0, 1, 0)$  by construction. Next we note that  $w$  is feasible for the dual problem. Clearly  $w \leq 0$ . Moreover, we note

$$\begin{aligned} 3w_1 - w_2 &= 1 \leq 4 \\ -w_1 + 2w_2 &= -2 \leq -2 \end{aligned}$$

Thus  $w$  is feasible for the dual problem with the same dual objective function value as the primal objective function value at  $(1, 0, 1, 0)$ . Then the weak duality theorem tells us that  $w$  is optimal for the dual and  $(1, 0, 1, 0)$  is optimal for the primal problem. By the comments above, we know that this is not the unique optimal solution to the primal problem, but only one optimal solution in a parameterized set of optimal solutions.

**Exercise 2** *It is known that an optimal solution to the following linear program is of the form  $(0, \alpha, \beta)$  with  $\alpha > 0$  and  $\beta > 0$ . Without using any simplex pivots, solve the problem and its dual.*

$$\begin{aligned} \min \quad & -12x_1 + 10x_2 + 2x_3 \\ \text{s.t.} \quad & -4x_1 + x_2 - 8x_3 \geq 1 \\ & -x_1 + x_2 + 12x_3 \geq 3 \\ & x_1, x_2, x_3 \geq 0 \end{aligned}$$

We can put this LP into standard form by the introduction of two slack variables. Having done that, we want  $x_2$  and  $x_3$  to be the basic variables. Let

$$B = \begin{bmatrix} 1 & -8 \\ 1 & 12 \end{bmatrix} \quad \text{and} \quad B^{-1} = \frac{1}{20} \begin{bmatrix} 12 & 8 \\ -1 & 1 \end{bmatrix}.$$

Let  $b = (1, 3)^\top$ . Then

$$B^{-1}b = \frac{1}{20} \begin{bmatrix} 12 & 8 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \frac{1}{10} \begin{bmatrix} 18 \\ 1 \end{bmatrix}.$$

We take  $x = (0, 1.8, 0.1)^\top$ . Then  $x$  has objective function value

$$-12x_1 + 10x_2 + 2x_3 = 18 + \frac{2}{10} = \frac{182}{10}.$$

Also note that  $x$  is feasible for the linear program. Clearly  $x \geq 0$ , and  $x$  meets the other two constraints with equality.

$$\begin{aligned} -4x_1 + x_2 - 8x_3 &= \frac{18}{10} - \frac{8}{10} = 1 \\ -x_1 + x_2 + 12x_3 &= \frac{18}{10} + \frac{12}{10} = 3. \end{aligned}$$

Now that we have the solution to the primal problem, the corresponding point for the dual problem is easy to compute. First we write down the dual problem for reference.

$$\begin{aligned} \max \quad & w_1 + 3w_2 \\ \text{s.t.} \quad & -4w_1 - w_2 \leq -12 \\ & w_1 + w_2 \leq 10 \\ & -8w_1 + 12w_2 \leq 2 \\ & w_1, w_2 \geq 0 \end{aligned}$$

We define

$$w = B^{-\top} c_B = \frac{1}{20} \begin{bmatrix} 12 & -1 \\ 8 & 1 \end{bmatrix} \begin{bmatrix} 10 \\ 2 \end{bmatrix} = \frac{1}{10} \begin{bmatrix} 59 \\ 41 \end{bmatrix}.$$

The objective function value for this  $w$  is

$$w_1 + 3w_2 = \frac{59}{10} + 3 \cdot \frac{41}{10} = \frac{182}{10},$$

which is the same as the cost at  $x$  for the primal problem, by construction. We finally note that this  $w$  is feasible for the dual problem. Clearly,  $w \geq 0$  and also

$$\begin{aligned} -4w_1 - w_2 &= -4 \cdot \frac{59}{10} - \frac{41}{10} = -19.5 < -12 \\ w_1 + w_2 &= \frac{59}{10} + \frac{41}{10} = 10 \\ -8w_1 + 12w_2 &= -8 \cdot \frac{59}{10} + 12 \cdot \frac{41}{10} = 2. \end{aligned}$$

Because  $w$  is feasible for the dual with the same optimal cost as the solution to the primal problem above, the weak duality theorem tells us that they both are optimal solutions.

**Exercise 3** Use Farkas' Lemma to prove the following theorem of alternatives. If  $A$  is an  $m \times n$  matrix and  $b \in \mathbb{R}^m$ , exactly one of the following two systems has a solution.

$$(1) Ax = b \quad (2) A^\top w = 0, b^\top w = 1.$$

It is easiest to prove this directly. First suppose that there is an  $x \in \mathbb{R}^n$  such that  $Ax = b$ . Further suppose that  $A^\top w = 0$ . Then

$$b^\top w = (Ax)^\top w = x^\top A^\top w = x^\top \mathbf{0} = 0 \neq 1.$$

So if (1) has a solution then (2) does not have a solution.

Next suppose that there is a  $w \in \mathbb{R}^m$  such that  $A^\top w = 0$  and  $b^\top w = 1$ . Let  $x \in \mathbb{R}^n$ . Then

$$(Ax)^\top w = x^\top A^\top w = x^\top \mathbf{0} = 0.$$

Since  $b^\top w = 1$ , we cannot have  $Ax = b$  for any  $x \in \mathbb{R}^n$ . To spell this out, if  $Ax = b$  then

$$0 = (Ax)^\top w = b^\top w = 1.$$

So if (2) has a solution then (1) does not have a solution.

Alternatively, we can prove this by way of Farkas' Lemma. Recall that Farkas says

$$(\exists x \geq 0 \text{ s.t. } Ax = b) \iff (b^\top w \leq 0 \forall w \text{ s.t. } A^\top w \leq 0)$$

First suppose that there is an  $x$  such that  $Ax = b$ . We write  $x = x_+ - x_-$ , where  $x_+ \geq 0$  and  $x_- \geq 0$ . Define  $b_+ = Ax_+$  and  $b_- = -Ax_-$ . Then

$$b = Ax = Ax_+ - Ax_- = b_+ + b_-.$$

Then by Farkas, we know two things.

$$b_+^\top w \leq 0 \forall w \text{ s.t. } A^\top w \leq 0 \quad \text{and} \quad b_-^\top w \leq 0 \forall w \text{ s.t. } A^\top w \geq 0$$

Putting these together

$$b^\top w = b_+^\top w + b_-^\top w \leq 0 \forall w \text{ s.t. } A^\top w = 0.$$

In particular, if  $A^\top w = 0$  then we cannot have  $b^\top w = 1$ . Thus if (1) has a solution, then (2) does not have a solution.

Now suppose that (2) has a solution. Then we have a  $w$  such that  $A^\top w = 0$  and  $b^\top w = 1$ . For a proof by contradiction, we suppose that there is an  $x$  such that  $Ax = b$ . Then the argument proceeds as above and we reach a contradiction. Thus if (2) has a solution, then (1) does not have a solution.

So while we can use Farkas' Lemma to prove this statement, it is easier to prove it directly without reference to Farkas.

**Exercise 4** Find all the values of  $\alpha$  and  $\beta$  such that the following linear program has an optimal solution with basic variables  $x_1$  and  $x_4$ .

$$\begin{aligned}
& \min && x_1 + \beta x_2 - x_3 - x_4 \\
& \text{s.t.} && x_1 + \beta x_3 + x_4 = 4 + \alpha \\
& && x_2 + x_3 + x_4 = 2 + 2\alpha \\
& && x_i \geq 0 \quad \forall i.
\end{aligned}$$

Let

$$A = \begin{bmatrix} 1 & 0 & \beta & 1 \\ 0 & 1 & 1 & 1 \end{bmatrix} \quad b = \begin{bmatrix} 4 + \alpha \\ 2 + 2\alpha \end{bmatrix} \quad \text{and} \quad c = \begin{bmatrix} 1 \\ \beta \\ -1 \\ -1 \end{bmatrix}.$$

Then we are trying to minimize  $c^\top x$  subject to  $Ax = b$  and  $x \geq 0$ . If we want to take  $x_1$  and  $x_4$  as basic variables, we take

$$B = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad \text{and} \quad B^{-1} = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}$$

Note

$$\begin{aligned}
B^{-1}A &= \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & \beta & 1 \\ 0 & 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & -1 & \beta - 1 & 0 \\ 0 & 1 & 1 & 1 \end{bmatrix} \\
B^{-1}b &= \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 4 + \alpha \\ 2 + 2\alpha \end{bmatrix} = \begin{bmatrix} 2 - \alpha \\ 2 + 2\alpha \end{bmatrix} \\
c^\top - c_B^\top B^{-1}A &= [1 \quad \beta \quad -1 \quad -1] - [1 \quad -1] \begin{bmatrix} 1 & -1 & \beta - 1 & 0 \\ 0 & 1 & 1 & 1 \end{bmatrix} \\
&= [1 \quad \beta \quad -1 \quad -1] - [1 \quad -2 \quad \beta - 2 \quad -1] \\
&= [0 \quad 2 + \beta \quad 1 - \beta \quad 0]
\end{aligned}$$

For this to be a solution, we need  $B^{-1}b \geq 0$  and we need all the reduced costs to be nonnegative. Then we must have

$$\left. \begin{array}{l} 2 - \alpha \geq 0 \\ 2 + 2\alpha \geq 0 \\ 2 + \beta \geq 0 \\ 1 - \beta \geq 0 \end{array} \right\} \implies \begin{array}{l} -1 \leq \alpha \leq 2 \\ -2 \leq \beta \leq 1 \end{array}$$