

# MATH 464 - Lecture 27 (04/18/2023)

Today: \* Bland's rule in Matlab  
\* Itw and other problems from BT-ILD

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## Bland's rule

Simply put, this is the "minimum index rule" — the non-basic  $x_j$  with  $c_j' < 0$  and smallest  $j$  enters, and in case of a tie, the basic variable  $x_l$  with the smallest  $l$  (that ties) leaves. Since we order the variables  $x_1, \dots, x_n$  by default, choosing the entering variable is easy — just pick the leftmost one. But choosing the leaving variable, while easy to do by hand, will require a bit more work.

We will maintain and update the basis  $B$  (stored as  $Bind$  in Matlab). See the session from today's lecture for details:

[https://www.math.wsu.edu/faculty/bkrishna/FilesMath464/S23/Software/Lec27\\_04182023\\_Session.txt](https://www.math.wsu.edu/faculty/bkrishna/FilesMath464/S23/Software/Lec27_04182023_Session.txt)

Now let's consider a problem not assigned in the homework.

**Exercise 3.19** While solving a standard form problem, we arrive at the following tableau, with  $x_3$ ,  $x_4$ , and  $x_5$  being the basic variables:

		$x_2$				
	-10	$\delta$	$-2=0$	0	0	0
	4	-1	$\eta$	1	0	0
$x_4$	1	$\alpha$	-4	0	1	0
$\beta=0$		$\gamma$	3	0	0	1

$R_6 + \left(\frac{2}{3}\right)R_3$   
 $\delta + \left(\frac{2}{3}\right)r$

The entries  $\alpha$ ,  $\beta$ ,  $\gamma$ ,  $\delta$ ,  $\eta$  in the tableau are unknown parameters. For each one of the following statements, find some parameter values that will make the statement true.

- (a) The current solution is optimal and there are multiple optimal solutions.
- (b) The optimal cost is  $-\infty$ .
- (c) The current solution is feasible but not optimal.

(a) Notice  $x_2$  could enter (as  $c'_2 = -2 < 0$ ). Thus the current bfs is optimal only if  $x_2$  could enter but not change the cost. In other words, we need a degenerate bfs here. If  $\beta=0$  and  $r>0$ , we could get  $\theta^*=0$  (min-ratio). Hence  $x_2$  could enter without changing the cost. The EROs give  $c'_1 = \delta + \left(\frac{2}{3}\right)r$ . If  $\delta + \left(\frac{2}{3}\right)r \geq 0$ , the resulting tableau is optimal, and hence so is the current solution, thus giving multiple optimal solutions.

$\beta=0, r>0, \delta + \frac{2}{3}r \geq 0.$

(b)  $x_2$  cannot improve the cost without bound, as we need  $\beta \geq 0$  for feasibility. We get unbounded LP with  $\delta < 0, \alpha \leq 0, r \leq 0$  ( $\beta \geq 0$  is needed for feasibility).

(c) We need  $\beta > 0$ , as then we get  $\theta^* = \min\left(\frac{\beta}{3}, \frac{4}{\eta}\right)$  if  $\eta > 0$ .  $\theta^* > 0$  here, and hence  $x_2$  could enter to improve the solution. We also need either  $\delta \geq 0$ , or if  $\delta < 0$ , then  $\alpha > 0$  or  $r > 0$ . Another option is  $\delta < 0$ , and  $\alpha < 0, r < 0$ , when the LP will be unbounded.

**Exercise 4.3** The purpose of this exercise is to show that solving linear programming problems is no harder than solving systems of linear inequalities.

Suppose that we are given a subroutine which, given a system of linear inequality constraints, either produces a solution or decides that no solution exists. Construct a simple algorithm that uses a single call to this subroutine and which finds an optimal solution to any linear programming problem that has an optimal solution.

Subroutine : feasibility of a system of linear inequalities.

Use LP duality to come up with a system of linear inequalities that you could input to the subroutine only once.

$$(P) \quad \begin{array}{ll} \min & \bar{c}^T \bar{x} \\ \text{s.t.} & A\bar{x} = \bar{b} \\ & \bar{x} \geq \bar{0} \end{array} \quad \begin{array}{ll} \max & \bar{p}^T \bar{b} \\ \text{s.t.} & \bar{p}^T A \leq \bar{c} \end{array} \quad (D)$$

Hint: weak/strong duality.

$$\hookrightarrow \bar{c}^T \bar{x} \geq \bar{p}^T \bar{b} \quad \text{for feasible } \bar{x}, \bar{p} \\ \text{for (P) and (D).}$$

$$\begin{array}{c} \downarrow \min \bar{c}^T \bar{x} \\ \text{optimality} \times \\ \uparrow \max \bar{p}^T \bar{b} \end{array}$$

If  $\bar{c}^T \bar{x} = \bar{p}^T \bar{b}$ , then  $\bar{x}, \bar{p}$  are optimal for (P) and (D), respectively.

Consider the system

$$\boxed{\begin{array}{l} A\bar{x} = \bar{b} \\ \bar{x} \geq \bar{0} \\ \bar{p}^T A \leq \bar{c} \\ \bar{c}^T \bar{x} = \bar{p}^T \bar{b} \end{array}} \quad \begin{array}{l} \left. \begin{array}{l} A\bar{x} = \bar{b} \\ \bar{x} \geq \bar{0} \end{array} \right\} \text{(P) feasibility} \\ \rightarrow \text{(D) feasibility} \\ \rightarrow \text{optimality} \end{array}$$

Call subroutine once with this system as input.

Bonus: Think about how to handle the situation if the subroutine says the system is infeasible!

Farkas' lemma:

**Exercise 4.26** Let  $A$  be a given matrix. Show that exactly one of the following alternatives must hold.

(a) There exists some  $x \neq 0$  such that  $Ax = 0, x \geq 0$ .

(b) There exists some  $p$  such that  $p'A > 0'$ . we want an equivalent system that is  $\geq$ , not  $>$ .

The crux is to come up with an appropriate pair of primal-dual LPs, and follow arguments presented in class.

$$\max \mathbf{1}^T \bar{x}$$

$$(P) \quad \begin{aligned} A\bar{x} &= \bar{0} \\ \bar{x} &\geq \bar{0} \end{aligned}$$

$$\min \bar{p}^T \bar{0}$$

$$\bar{p}^T A \geq \mathbf{1} \quad (D)$$

Suppose (b) holds. Then (D) is feasible. (D) is not unbounded, as  $\bar{p}^T \bar{0} = 0$  for any  $\bar{p}$ . So (D) has optimal cost = 0. Hence for (P), we get that the only optimal solution is  $\bar{x} = \bar{0}$ . (as  $\mathbf{1}^T \bar{x} = \sum x_i = 0$  optimal  $\Rightarrow \bar{x} = \bar{0}$  is the only optimal solution).

So (a) cannot hold.