

## Correction on Clicker question for Lecture 25

Given the LP problem  $\max_{A\vec{x} \leq \vec{b}, \vec{x} \geq \vec{0}} \vec{c} \cdot \vec{x}$ , consider its primal and dual

dictionaries determined by  $B, N, \vec{b}, \vec{c}$ :

$$\begin{array}{lcl} \text{primal} & & \begin{array}{l} z = \vec{c}_B^T B^{-1} \vec{b} + [\vec{c}_N^T - \vec{c}_B^T B^{-1} N] \vec{x}_N \\ \vec{x}_B = B^{-1} \vec{b} - B^{-1} N \vec{x}_N \end{array} \end{array}$$

$$\begin{array}{lcl} \text{dual} & & \begin{array}{l} -w = -\vec{c}_B^T B^{-1} \vec{b} - (B^{-1} \vec{b})^T \vec{y}_B \\ \vec{y}_N = -[\vec{c}_N^T - \vec{c}_B^T B^{-1} N]^T + (B^{-1} N)^T \vec{y}_B \end{array} \end{array}$$

Let  $\vec{x}^*, \vec{y}^*$  be the corresponding **primal/dual basic solutions**, respectively. Find a wrong statement:

- A. If  $\vec{x}^*, \vec{y}^*$  are primal/dual optimal, respectively, then  $\vec{x}_B^* \geq 0$  and  $\vec{y}_N^* \geq 0$ . **[I didn't mean to include  $\vec{y}^*$  in this sentence. After this correction, A. is the answer.]**
- B. If  $\vec{x}_B^* \geq 0$  and  $\vec{y}_N^* \geq 0$ , then  $\vec{x}^*, \vec{y}^*$  are primal/dual optimal, respectively.
- C.  $\vec{x}_N^* = \vec{0}$ , and  $\vec{y}_B^* = \vec{0}$ .
- D. One of A, B, and C is wrong. **[In the original problem before the correction, D was the answer.]**

# Comparison with strong duality theorem

*primal*

$$\begin{array}{rcl} z & = & \vec{c}_B^T \vec{x}_B^* - (\vec{y}_N^*)^T \vec{x}_N \\ \vec{x}_B & = & \vec{x}_B^* - B^{-1} N \vec{x}_N \end{array}$$

*dual*

$$\begin{array}{rcl} -w & = & -(\vec{y}^*)^T \vec{b} - (\vec{x}_B^*)^T \vec{y}_B \\ \vec{y}_N & = & \vec{y}_N^* + (B^{-1} N)^T \vec{y}_B \end{array}$$

- ▶ At the feasible and dual feasible dictionary (that is,  $\vec{x}_B^* \geq \vec{0}, \vec{y}_N^* \geq \vec{0}$ ), the corresponding primal/dual basic solutions  $\vec{x}^*, \vec{y}^*$  are primal/dual optimal, respectively.
- ▶ Optimality of the primal basic solution  $\vec{x}^*$  **does not necessarily** imply dual optimality of the dual basic solution  $\vec{y}^*$  as it may not be dual feasible (it may happen  $\vec{y}_N^* \not\geq \vec{0}$ ).
- ▶ **Strong duality** says if there is a primal optimal solution then there is a dual optimal solution.
  - ▶ It **does not mean** that for an optimal basic solution, the corresponding dual basic solution is dual optimal. The corresponding dual basic solution may not be feasible, **unless** the dictionary is **final**, that is, both primal/dual feasible.
  - ▶ In the strong duality theorem, you can read off the dual optimal solution from the dictionary, **only when the dictionary is final, where the corresponding dual basic solution has  $\vec{y}_N^* \geq \vec{0}$ .**

# Basic idea for sensitivity analysis

How can we find the new optimal solution in the new situation, using the old optimal solution of the problem, **without having to solve the new problem from scratch?**

- ▶ Changing  $\vec{c}$ 
  - ▶ affects  $\vec{c}_B, \vec{c}_N$  (so  $[\vec{c}_N^T - \vec{c}_B^T B^{-1} N]$ ),
  - ▶ but it does not affect  $B, N, \vec{x}_B^* = B^{-1} \vec{b}$ .
  - ▶ It does not affect feasibility of the primal basic solution.
- ▶ Changing  $\vec{b}$ 
  - ▶ affects  $\vec{x}_B^* = B^{-1} \vec{b}$
  - ▶ but it does not affect  $B, N, \vec{c}_B, \vec{c}_N, [\vec{c}_N^T - \vec{c}_B^T B^{-1} N]$ , &  $(\vec{y}^*)^T = \vec{c}_B^T B^{-1}$ .
  - ▶ It does not affect dual feasibility of the dual basic solution.
- ▶ Under the changes, we may **recycle unaffected items**.

## Example

$$\begin{array}{ll} \text{Maximize} & z = x_1 + x_2 \\ \text{subject to} & x_1 + 2x_2 + x_3 \leq 2 \quad \& \quad x_1, x_2, x_3 \geq 0, \\ & 2x_1 + x_2 + x_3 \leq 2 \end{array}$$

We got an optimal feasible dictionary with

$$\vec{x}_B^* = \begin{bmatrix} x_2^* \\ x_1^* \end{bmatrix} = \begin{bmatrix} 2/3 \\ 2/3 \end{bmatrix}, \quad N = \begin{bmatrix} x_5 & x_4 & x_3 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} x_2 & x_1 \\ 2 & 1 \\ 1 & 2 \end{bmatrix},$$
$$\vec{c}_N^T = \begin{bmatrix} x_5 & x_4 & x_3 \\ 0 & 0 & 0 \end{bmatrix}, \quad \vec{c}_B^T = \begin{bmatrix} x_2 & x_1 \\ 1 & 1 \end{bmatrix},$$

and  $(\vec{y}^*)^T = \vec{c}_B^T B^{-1} = [y_1^* \quad y_2^*] = [1/3 \quad 1/3]$  is the optimal dual solution.

**Question 2:** In the same problem, how much can we change  $\vec{b}$  to  $\vec{b} = (2, 2 + s)$  still keeping the same dual optimal  $\vec{y}^*$ ? Find the range for  $s$ . What happens to the primal optimal solution?



# Dual Simplex Method

Recall **primal /dual correspondence**.

For  $\max_{A\vec{x} \leq \vec{b}, \vec{x} \geq 0} \vec{c}^T \vec{x}$ , with  $A$  is  $m \times n$ :

- ▶  $x_j \longleftrightarrow y_{m+j}, \quad j = 1, \dots, n$
- ▶  $x_{n+i} \longleftrightarrow y_i, \quad i = 1, \dots, m$
- ▶  $\vec{x}_B \longleftrightarrow \vec{y}_B$
- ▶  $\vec{x}_N \longleftrightarrow \vec{y}_N$

*primal*

$$\begin{array}{rcl} z & = & \vec{c}_B^T \vec{x}_B^* - (\vec{y}_N^*)^T \vec{x}_N \\ \hline \vec{x}_B & = & \vec{x}_B^* - B^{-1} N \vec{x}_N \end{array}$$

*dual*

$$\begin{array}{rcl} -w & = & -(\vec{y}^*)^T \vec{b} - (\vec{x}_B^*)^T \vec{y}_B \\ \hline \vec{y}_N & = & \vec{y}_N^* + (B^{-1} N)^T \vec{y}_B \end{array}.$$

# Dual Simplex Method: Dual Pivot

$$Z = -2X_1 - 2X_2 \quad \leftarrow \text{dual feasible}$$

$$X_3 = -1 + X_1 + 2X_2$$

$$X_4 = -1 + 2X_1 + X_2$$

$$X_5 = -0 + X_1 + X_2$$

$X_3$  leaving  
 $X_2$  entering

dual pivot

$$Z = -1 - X_1 - X_3$$

$$X_2 = \frac{1}{2} - \frac{1}{2}X_1 + \frac{1}{2}X_3$$

$$X_4 = -\frac{1}{2} + \frac{3}{2}X_1 + \frac{1}{2}X_3$$

$$X_5 = \frac{1}{2} + \frac{1}{2}X_1 + \frac{1}{2}X_3$$

$X_3$  leaving  
 $X_1$  entering

dual pivot

$$Z = -\frac{4}{3} - \frac{2}{3}X_4 - \frac{2}{3}X_3$$

$$X_2 = \frac{1}{3} + \frac{1}{3}X_4 + \frac{2}{3}X_3$$

$$X_1 = \frac{1}{3} + \frac{2}{3}X_4 - \frac{1}{3}X_3$$

$$X_5 = \frac{2}{3} + \frac{1}{3}X_4 + \frac{1}{3}X_3$$

optimal

$$\vec{x}^* = \left( \frac{1}{3}, \frac{1}{3}, 0, 0, \frac{2}{3} \right)$$

$$-w = Y_1 + Y_2 + 0$$

$$Y_4 = 2 - Y_1 - 2Y_2 - Y_3$$

$$Y_5 = 2 - 2Y_1 - Y_2 - Y_3$$

mirror

$$X_2 + 1 = X_3 \rightarrow Y_1$$

$$X_2 \leftarrow Y_5 = Y_5 + 2$$

pivot  $\downarrow$   $Y_1$  entering  
 $Y_5$  leaving

$$-w = 1 - \frac{1}{2}Y_5 + \frac{1}{2}Y_2 - \frac{1}{2}Y_3$$

$$Y_4 = 1 + \frac{1}{2}Y_5 - \frac{3}{2}Y_2 - \frac{1}{2}Y_3$$

$$Y_1 = 1 - \frac{1}{2}Y_5 - \frac{1}{2}Y_2 - \frac{1}{2}Y_3$$

mirror

$\leftarrow$

$$X_4 = X_2 + 2 \rightarrow Y_2$$

$$X_1 \leftarrow Y_4 = Y_4 + 1$$

pivot  $\downarrow$   $Y_2$  entering  
 $Y_4$  leaving

$$-w = \frac{4}{3} - \frac{1}{3}Y_5 - \frac{1}{3}Y_4 - \frac{2}{3}Y_3$$

$$Y_2 = \frac{2}{3} - \frac{1}{3}Y_5 - \frac{2}{3}Y_4 - \frac{1}{3}Y_3$$

$$Y_1 = \frac{2}{3} - \frac{2}{3}Y_5 + \frac{1}{3}Y_4 - \frac{1}{3}Y_3$$

optimal

$$\vec{y}^* = \left( \frac{2}{3}, \frac{2}{3}, 0, 0, 0 \right)$$

# Dual Simplex Method: Dual Pivot

*primal*

$$\begin{array}{rcl} z & = & \vec{c}_B^T \vec{x}_B^* - (\vec{y}_N^*)^T \vec{x}_N \\ \hline \vec{x}_B & = & \vec{x}_B^* - B^{-1} N \vec{x}_N \end{array}$$

*dual*

$$\begin{array}{rcl} -w & = & -(\vec{y}_N^*)^T \vec{b} - (\vec{x}_B^*)^T \vec{y}_B \\ \hline \vec{y}_N & = & \vec{y}_N^* + (B^{-1} N)^T \vec{y}_B \end{array}$$

- ▶ Dual Pivot  
= the operation **on the primal dictionary** corresponding to the usual pivot on the dual dictionary.
- ▶ From a feasible dictionary ( $\vec{x}_B^* \geq \vec{0}$ ),
  - ▶ Pivot tries to achieve dual feasibility ( $\vec{y}_N^* \geq \vec{0}$ ).
  - ▶ Pivot keeps feasibility of the dictionary.
- ▶ From a dual feasible dictionary (i.e.  $\vec{y}_N^* \geq \vec{0}$ , dual dictionary is feasible).
  - ▶ Dual pivot tries to achieve primal feasibility ( $\vec{x}_B^* \geq \vec{0}$ ).
  - ▶ Dual pivot keeps dual feasibility of the dictionary.



# Primal and dual simplex methods

*primal*

$$\begin{array}{rcl} z & = & \vec{c}_B^T \vec{x}_B^* - (\vec{y}_N^*)^T \vec{x}_N \\ \vec{x}_B & = & \vec{x}_B^* - B^{-1} N \vec{x}_N \end{array}$$

*dual*

$$\begin{array}{rcl} -w & = & -(\vec{y}^*)^T \vec{b} - (\vec{x}_B^*)^T \vec{y}_B \\ \vec{y}_N & = & \vec{y}_N^* + (B^{-1} N)^T \vec{y}_B \end{array}.$$

## For the primal problem:

- ▶ Starting from a feasible dictionary ( $\vec{x}_B^* \geq \vec{0}$ ), can apply the simplex method to arrive at a dual feasible dictionary ( $\vec{y}_N^* \geq \vec{0}$ ) (or get unboundedness in the end).
- ▶ Starting from a dual feasible dictionary ( $\vec{y}_N^* \geq \vec{0}$ ), can apply the dual simplex method to arrive at a feasible dictionary ( $\vec{x}_B^* \geq \vec{0}$ ) (or get infeasibility in the end).
- ▶ From a dictionary neither feasible or dual feasible, can apply two phase method either to the primal or the dual problem.

# Dual Pivot. Example without optimal solution

*primal*

$$\begin{array}{rclcl} z & = & 3 & -2x_1 & -x_2 \\ \hline x_4 & = & 1 & -x_1 & -x_2 \\ x_3 & = & -2 & -3x_1 & -2x_2 \end{array}$$

$x_3$  leaving but no one entering  
the primal LP is infeasible

*dual*

$$\begin{array}{rclcl} -w & = & -3 & -y_2 & +2y_1 \\ \hline y_3 & = & 2 & +y_2 & +3y_1 \\ y_4 & = & 1 & +y_2 & +2y_1 \end{array}$$

$y_1$  entering but no one leaving  
the dual LP is unbounded

## Clicker question for Lecture 26

For **Dual Pivot**, find a correct statement:

A. For

$$\begin{array}{rcll} z & = & -4/3 & -2/3x_4 & -2/3x_3 \\ \hline x_2 & = & 1/3 & +1/3x_4 & +2/3x_3 \\ x_1 & = & -1/3 & +2/3x_4 & -1/3x_3 \\ x_5 & = & 2/3 & +1/3x_4 & +1/3x_3 \end{array}$$

$x_1$  is leaving and  
 $x_3$  is entering.

B. For

$$\begin{array}{rcll} z & = & 0 & -2x_1 & -2x_2 \\ \hline x_3 & = & -1 & +x_1 & +2x_2 \\ x_4 & = & -1 & +2x_1 & +x_2 \\ x_5 & = & -0 & +x_1 & +x_2 \end{array}$$

$x_3$  is leaving and  
 $x_1$  is entering.

C. For

$$\begin{array}{rcll} z & = & -1 & -x_1 & -x_3 \\ \hline x_2 & = & 1/2 & -1/2x_1 & +1/2x_3 \\ x_4 & = & -1/2 & +3/2x_1 & +1/2x_3 \\ x_5 & = & 1/2 & +1/2x_1 & +1/2x_3 \end{array}$$

$x_4$  is leaving and  
 $x_1$  is entering.