

# Linear Programming

## Lecture 6: The Simplex Algorithm Language, Notation, and Linear Algebra

Math Dept, University of Washington

- 1 Dictionaries for LPs in Standard Form
- 2 The Simplex Algorithm via Matrix Multiplication
- 3 The Block Structure of the Simplex Algorithm
- 4 Block Structure and Matrix Multiplication
- 5 The Block Structure of an Optimal Tableau
- 6 Block Structure and Duality

# Dictionaries for LPs in Standard Form

# Dictionaries for LPs in Standard Form

$$\begin{aligned} \mathcal{P} : \quad & \text{maximize} && c^T x \\ & \text{subject to} && Ax \leq b \\ & && 0 \leq x, \end{aligned}$$

where  $c \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$ ,  $A \in \mathbb{R}^{m \times n}$

# Dictionaries for LPs in Standard Form

$$\begin{aligned} \mathcal{P} : \quad & \text{maximize} && c^T x \\ & \text{subject to} && Ax \leq b \\ & && 0 \leq x, \end{aligned}$$

where  $c \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$ ,  $A \in \mathbb{R}^{m \times n}$

The initial dictionary:

# Dictionaries for LPs in Standard Form

$$\begin{aligned} \mathcal{P} : \quad & \text{maximize} && c^T x \\ & \text{subject to} && Ax \leq b \\ & && 0 \leq x, \end{aligned} \quad \left[ \sum_{j=1}^n a_{ij} x_j \leq b_i, \quad i = 1, \dots, m \right]$$

where  $c \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$ ,  $A \in \mathbb{R}^{m \times n}$

The initial dictionary:

$$\text{slack variables} \quad x_{n+i} := b_i - \sum_{j=1}^n a_{ij} x_j \quad i = 1, 2, \dots, m$$

# Dictionaries for LPs in Standard Form

$$\begin{aligned} \mathcal{P} : \quad & \text{maximize} && c^T x \\ & \text{subject to} && Ax \leq b \\ & && 0 \leq x, \end{aligned} \quad \left[ \sum_{j=1}^n a_{ij} x_j \leq b_i, \quad i = 1, \dots, m \right]$$

where  $c \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$ ,  $A \in \mathbb{R}^{m \times n}$

The initial dictionary:

$$\text{slack variables} \quad x_{n+i} := b_i - \sum_{j=1}^n a_{ij} x_j \quad i = 1, 2, \dots, m$$

$$\text{objective} \quad z := \sum_{j=1}^n c_j x_j$$

# Dictionaries for LPs in Standard Form

$$\begin{array}{ll} \mathcal{P} : & \text{maximize} \quad c^T x \\ & \text{subject to} \quad Ax \leq b \\ & \quad \quad \quad 0 \leq x, \end{array} \quad \left[ \sum_{j=1}^n a_{ij} x_j \leq b_i, \quad i = 1, \dots, m \right]$$

where  $c \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$ ,  $A \in \mathbb{R}^{m \times n}$

The initial dictionary:

$$\text{slack variables} \quad x_{n+i} := b_i - \sum_{j=1}^n a_{ij} x_j \quad i = 1, 2, \dots, m$$

( $D_I$ )

$$\text{objective} \quad z := \sum_{j=1}^n c_j x_j$$



# General Dictionaries

A dictionary for  $\mathcal{P}$  is any system of the form

$$\begin{aligned}x_i &= \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j & i \in B & \\z &= \hat{z} + \sum_{j \in N} \hat{c}_j x_j & & (D_B)\end{aligned}$$

A dictionary for  $\mathcal{P}$  is any system of the form

$$\begin{aligned}x_i &= \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j & i \in B \\z &= \hat{z} + \sum_{j \in N} \hat{c}_j x_j\end{aligned} \quad (D_B)$$

where  $B$  and  $N$  are index sets partitioning  $\{1, \dots, n + m\}$  and satisfying

A dictionary for  $\mathcal{P}$  is any system of the form

$$\begin{aligned}x_i &= \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j & i \in B \\z &= \hat{z} + \sum_{j \in N} \hat{c}_j x_j\end{aligned} \quad (D_B)$$

where  $B$  and  $N$  are index sets partitioning  $\{1, \dots, n + m\}$  and satisfying

- (1)  $B$  contains  $m$  elements and  $N$  contains  $n$  elements,

A dictionary for  $\mathcal{P}$  is any system of the form

$$\begin{aligned}x_i &= \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j & i \in B \\z &= \hat{z} + \sum_{j \in N} \hat{c}_j x_j\end{aligned} \quad (D_B)$$

where  $B$  and  $N$  are index sets partitioning  $\{1, \dots, n + m\}$  and satisfying

- (1)  $B$  contains  $m$  elements and  $N$  contains  $n$  elements,
- (2)  $B \cap N = \emptyset$

A dictionary for  $\mathcal{P}$  is any system of the form

$$\begin{aligned}x_i &= \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j & i \in B \\z &= \hat{z} + \sum_{j \in N} \hat{c}_j x_j\end{aligned} \quad (D_B)$$

where  $B$  and  $N$  are index sets partitioning  $\{1, \dots, n + m\}$  and satisfying

- (1)  $B$  contains  $m$  elements and  $N$  contains  $n$  elements,
- (2)  $B \cap N = \emptyset$
- (3)  $B \cup N = \{1, 2, \dots, n + m\}$ ,

A dictionary for  $\mathcal{P}$  is any system of the form

$$\begin{aligned}x_i &= \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j & i \in B \\z &= \hat{z} + \sum_{j \in N} \hat{c}_j x_j\end{aligned} \quad (D_B)$$

where  $B$  and  $N$  are index sets partitioning  $\{1, \dots, n + m\}$  and satisfying

- (1)  $B$  contains  $m$  elements and  $N$  contains  $n$  elements,
- (2)  $B \cap N = \emptyset$
- (3)  $B \cup N = \{1, 2, \dots, n + m\}$ ,

and such that the systems  $(D_I)$  and  $(D_B)$  have identical solution sets.

# Properties of Dictionaries

$$x_i = \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j \quad i \in B \quad (D_B)$$

$$z = \hat{z} + \sum_{j \in N} \hat{c}_j x_j$$

- $B \sim$  basic variables     $N \sim$  nonbasic variables

# Properties of Dictionaries

$$x_i = \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j \quad i \in B \quad (D_B)$$

$$z = \hat{z} + \sum_{j \in N} \hat{c}_j x_j$$

- $B \sim$  basic variables     $N \sim$  nonbasic variables
- Basic solution identified by the dictionary is

$$x_i = \hat{b}_i \quad i \in B$$

$$x_j = 0 \quad j \in N.$$



# Properties of Dictionaries

$$x_i = \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j \quad i \in B \quad (D_B)$$

$$z = \hat{z} + \sum_{j \in N} \hat{c}_j x_j$$

- $B \sim$  basic variables     $N \sim$  nonbasic variables
- Basic solution identified by the dictionary is

$$x_i = \hat{b}_i \quad i \in B$$

$$x_j = 0 \quad j \in N.$$

- Dictionary is feasible if  $0 \leq \hat{b}_i$  for  $i \in B$ .

# Properties of Dictionaries

$$x_i = \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j \quad i \in B \quad (D_B)$$

$$z = \hat{z} + \sum_{j \in N} \hat{c}_j x_j$$

- $B \sim$  basic variables     $N \sim$  nonbasic variables
- Basic solution identified by the dictionary is

$$\begin{aligned} x_i &= \hat{b}_i & i \in B \\ x_j &= 0 & j \in N. \end{aligned}$$

- Dictionary is feasible if  $0 \leq \hat{b}_i$  for  $i \in B$ .
- If feasible, then the basic solution is a *basic feasible solution* (BFS).

# Properties of Dictionaries

$$x_i = \hat{b}_i - \sum_{j \in N} \hat{a}_{ij} x_j \quad i \in B \quad (D_B)$$

$$z = \hat{z} + \sum_{j \in N} \hat{c}_j x_j$$

- $B \sim$  basic variables     $N \sim$  nonbasic variables
- Basic solution identified by the dictionary is

$$\begin{aligned} x_i &= \hat{b}_i & i \in B \\ x_j &= 0 & j \in N. \end{aligned}$$

- Dictionary is feasible if  $0 \leq \hat{b}_i$  for  $i \in B$ .
- If feasible, then the basic solution is a *basic feasible solution* (BFS).
- A feasible dictionary is *optimal* if  $\hat{c}_j \leq 0$   $j \in N$ .

# The Simplex Algorithm via Matrix Multiplication

We have already seen that Gaussian elimination can be performed by matrix multiplication.

# The Simplex Algorithm via Matrix Multiplication

We have already seen that Gaussian elimination can be performed by matrix multiplication.

How does this look in the context of the simplex algorithm?

# The Simplex Algorithm via Matrix Multiplication

We have already seen that Gaussian elimination can be performed by matrix multiplication.

How does this look in the context of the simplex algorithm?

First recall that

$$\begin{bmatrix} I_{s \times s} & -\alpha^{-1}a & 0 \\ 0 & \alpha^{-1} & 0 \\ 0 & -\alpha^{-1}b & I_{t \times t} \end{bmatrix} \begin{pmatrix} a \\ \alpha \\ b \end{pmatrix} = \begin{bmatrix} a - a \\ \alpha^{-1}\alpha \\ -b + b \end{bmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}.$$

# The Simplex Algorithm via Matrix Multiplication

We have already seen that Gaussian elimination can be performed by matrix multiplication.

How does this look in the context of the simplex algorithm?

First recall that

$$\begin{bmatrix} I_{s \times s} & -\alpha^{-1}a & 0 \\ 0 & \alpha^{-1} & 0 \\ 0 & -\alpha^{-1}b & I_{t \times t} \end{bmatrix} \begin{pmatrix} a \\ \alpha \\ b \end{pmatrix} = \begin{bmatrix} a - a \\ \alpha^{-1}\alpha \\ -b + b \end{bmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}.$$

The elimination matrix and its inverse.

$$G = \begin{bmatrix} I_{s \times s} & -\alpha^{-1}a & 0 \\ 0 & \alpha^{-1} & 0 \\ 0 & -\alpha^{-1}b & I_{t \times t} \end{bmatrix} \quad G^{-1} = \begin{bmatrix} I & a & 0 \\ 0 & \alpha & 0 \\ 0 & b & I \end{bmatrix}$$

# The Simplex Algorithm via Matrix Multiplication

The elimination matrices also have the following important property.

$$\begin{bmatrix} I_{s \times s} & -\alpha^{-1}a & 0 \\ 0 & \alpha^{-1} & 0 \\ 0 & -\alpha^{-1}b & I_{t \times t} \end{bmatrix} \begin{pmatrix} x \\ 0 \\ y \end{pmatrix} = \begin{pmatrix} x \\ 0 \\ y \end{pmatrix}$$



# The Simplex Algorithm via Matrix Multiplication

The elimination matrices also have the following important property.

$$\begin{bmatrix} I_{s \times s} & -\alpha^{-1}a & 0 \\ 0 & \alpha^{-1} & 0 \\ 0 & -\alpha^{-1}b & I_{t \times t} \end{bmatrix} \begin{pmatrix} x \\ 0 \\ y \end{pmatrix} = \begin{pmatrix} x \\ 0 \\ y \end{pmatrix}$$

We call these matrices *Gauss-Jordan elimination* or pivot matrices.

# The Simplex Algorithm via Matrix Multiplication

The elimination matrices also have the following important property.

$$\begin{bmatrix} I_{s \times s} & -\alpha^{-1}a & 0 \\ 0 & \alpha^{-1} & 0 \\ 0 & -\alpha^{-1}b & I_{t \times t} \end{bmatrix} \begin{pmatrix} x \\ 0 \\ y \end{pmatrix} = \begin{pmatrix} x \\ 0 \\ y \end{pmatrix}$$

We call these matrices *Gauss-Jordan elimination* or pivot matrices.

These matrices perform precisely the operations required in order to execute a simplex pivot.

# The Simplex Algorithm via Matrix Multiplication

The elimination matrices also have the following important property.

$$\begin{bmatrix} I_{s \times s} & -\alpha^{-1}a & 0 \\ 0 & \alpha^{-1} & 0 \\ 0 & -\alpha^{-1}b & I_{t \times t} \end{bmatrix} \begin{pmatrix} x \\ 0 \\ y \end{pmatrix} = \begin{pmatrix} x \\ 0 \\ y \end{pmatrix}$$

We call these matrices *Gauss-Jordan elimination* or pivot matrices.

These matrices perform precisely the operations required in order to execute a simplex pivot.

Each simplex pivot can be realized as left multiplication of the simplex tableau by the appropriate Gaussian-Jordan pivot matrix.

# The Simplex Algorithm via Matrix Multiplication

$$\left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & \textcircled{2} & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right]$$

# The Simplex Algorithm via Matrix Multiplication

$$\left[ \begin{array}{cccc|cccc|c} 1 & -2 & 0 & 0 & 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 0 & \frac{1}{2} & 0 & 0 & 3 & \textcircled{2} & 1 & 0 & 1 & 0 & 5 \\ 0 & -1 & 1 & 0 & 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ 0 & \frac{-5}{2} & 0 & 1 & 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right]$$

# The Simplex Algorithm via Matrix Multiplication

$$\begin{bmatrix} 1 & -2 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & \frac{-5}{2} & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 4 & 2 & 1 & 0 & 0 & | & 11 \\ 3 & \textcircled{2} & 1 & 0 & 1 & 0 & | & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & | & 8 \\ 4 & 5 & 3 & 0 & 0 & 0 & | & 0 \end{bmatrix} = \begin{bmatrix} -5 & 0 & 0 & 1 & -2 & 0 & | & 1 \\ \frac{3}{2} & 1 & \frac{1}{2} & 0 & \frac{1}{2} & 0 & | & \frac{5}{2} \\ 1 & 0 & \textcircled{1} & 0 & -1 & 1 & | & 3 \\ -\frac{7}{2} & 0 & \frac{1}{2} & 0 & \frac{-5}{2} & 0 & | & \frac{-25}{2} \end{bmatrix}$$

# The Simplex Algorithm via Matrix Multiplication

$$\left[ \begin{array}{cccc|cccc|c} 1 & -2 & 0 & 0 & 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 0 & \frac{1}{2} & 0 & 0 & 3 & \textcircled{2} & 1 & 0 & 1 & 0 & 5 \\ 0 & -1 & 1 & 0 & 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ 0 & \frac{-5}{2} & 0 & 1 & 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right] = \left[ \begin{array}{cccc|cccc|c} -5 & 0 & 0 & 1 & -2 & 0 & 1 & -2 & 0 & 1 & 1 \\ \frac{3}{2} & 1 & \frac{1}{2} & 0 & \frac{1}{2} & 0 & 1 & -1 & 1 & 1 & 3 \\ 1 & 0 & \textcircled{1} & 0 & -1 & 1 & 1 & -1 & 1 & 1 & 3 \\ -\frac{7}{2} & 0 & \frac{1}{2} & 0 & \frac{-5}{2} & 0 & 1 & -2 & 0 & 1 & 1 \end{array} \right]$$

$$\left[ \begin{array}{cccc|cccc|c} 1 & 0 & 0 & 0 & -5 & 0 & 0 & 1 & -2 & 0 & 1 \\ 0 & 1 & \frac{-1}{2} & 0 & \frac{3}{2} & 1 & \frac{1}{2} & 0 & \frac{1}{2} & 0 & \frac{5}{2} \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & -1 & 1 & 3 \\ 0 & 0 & \frac{-1}{2} & 1 & \frac{-7}{2} & 0 & \frac{1}{2} & 0 & \frac{-5}{2} & 0 & \frac{-25}{2} \end{array} \right]$$

# The Simplex Algorithm via Matrix Multiplication

$$\left[ \begin{array}{cccc|cccc|c} 1 & -2 & 0 & 0 & 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 0 & \frac{1}{2} & 0 & 0 & 3 & \textcircled{2} & 1 & 0 & 1 & 0 & 5 \\ 0 & -1 & 1 & 0 & 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ 0 & \frac{-5}{2} & 0 & 1 & 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right] = \left[ \begin{array}{cccc|cccc|c} -5 & 0 & 0 & 1 & -2 & 0 & 1 & -2 & 0 & 1 & 1 \\ \frac{3}{2} & 1 & \frac{1}{2} & 0 & \frac{1}{2} & 0 & 1 & 0 & 1 & 0 & \frac{5}{2} \\ 1 & 0 & \textcircled{1} & 0 & -1 & 1 & 1 & 0 & 1 & 0 & 3 \\ -\frac{7}{2} & 0 & \frac{1}{2} & 0 & \frac{-5}{2} & 0 & 1 & 0 & 1 & 0 & \frac{-25}{2} \end{array} \right]$$

$$\left[ \begin{array}{cccc|cccc|c} 1 & 0 & 0 & 0 & -5 & 0 & 0 & 1 & -2 & 0 & 1 \\ 0 & 1 & \frac{-1}{2} & 0 & \frac{3}{2} & 1 & \frac{1}{2} & 0 & \frac{1}{2} & 0 & \frac{5}{2} \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & -1 & 1 & 3 \\ 0 & 0 & \frac{-1}{2} & 1 & \frac{-7}{2} & 0 & \frac{1}{2} & 0 & \frac{-5}{2} & 0 & \frac{-25}{2} \end{array} \right] = \left[ \begin{array}{cccc|cccc|c} -5 & 0 & 0 & 1 & -2 & 0 & 1 & -2 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 & 1 & \frac{-1}{2} & 1 & 0 & 1 & 0 & \frac{5}{2} \\ 1 & 0 & 1 & 0 & -1 & 1 & 1 & 0 & 1 & 0 & 3 \\ -4 & 0 & 0 & 0 & -2 & \frac{-1}{2} & 1 & 0 & 1 & 0 & \frac{-25}{2} \end{array} \right]$$



# The Simplex Algorithm via Matrix Multiplication

$$\left[ \begin{array}{cc|c} A & I & b \\ \hline c^T & 0 & 0 \end{array} \right]$$

# The Simplex Algorithm via Matrix Multiplication

$$G_2 G_1 \left[ \begin{array}{cc|c} A & I & b \\ \hline c^T & 0 & 0 \end{array} \right]$$

# The Simplex Algorithm via Matrix Multiplication

$$G_2 G_1 \left[ \begin{array}{ccc|c} A & I & b \\ \hline c^T & 0 & 0 \end{array} \right]$$

where

$$G_2 G_1 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & \frac{-1}{2} & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & \frac{-1}{2} & 1 \end{bmatrix} \begin{bmatrix} 1 & -2 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & \frac{-5}{2} & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & -2 & 0 & 0 \\ 0 & 1 & \frac{-1}{2} & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -2 & \frac{-1}{2} & 1 \end{bmatrix}$$

# The Simplex Algorithm via Matrix Multiplication

$$G_2 G_1 \left[ \begin{array}{cc|c} A & I & b \\ \hline c^T & 0 & 0 \end{array} \right]$$

# The Simplex Algorithm via Matrix Multiplication

$$G_2 G_1 \left[ \begin{array}{cccc|cc} A & I & & & b & \\ \hline c^T & 0 & & & 0 & \end{array} \right]$$
$$= \left[ \begin{array}{cccc} 1 & -2 & 0 & 0 \\ 0 & 1 & \frac{-1}{2} & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -2 & \frac{-1}{2} & 1 \end{array} \right] \left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & 2 & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right]$$

# The Simplex Algorithm via Matrix Multiplication

$$\begin{aligned} & G_2 G_1 \left[ \begin{array}{cccc|cc} A & I & & & b \\ \hline c^T & 0 & & & 0 \end{array} \right] \\ &= \left[ \begin{array}{cccc} 1 & -2 & 0 & 0 \\ 0 & 1 & \frac{-1}{2} & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -2 & \frac{-1}{2} & 1 \end{array} \right] \left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & 2 & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right] \\ &= \left[ \begin{array}{cccc|ccc} -5 & 0 & 0 & 1 & -2 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & \frac{-1}{2} & 1 \\ 1 & 0 & 1 & 0 & -1 & 1 & 3 \\ \hline -4 & 0 & 0 & 0 & -2 & \frac{-1}{2} & -14 \end{array} \right] \end{aligned}$$

# The Block Structure of the Simplex Algorithm

Let  $T_0$  be the initial tableau:

$$T_0 = \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right].$$

# The Block Structure of the Simplex Algorithm

Let  $T_0$  be the initial tableau:

$$T_0 = \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right].$$

Let  $T_k$  denote the tableau after  $k$  pivots:

$$T_k = \left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right]$$



# The Block Structure of the Simplex Algorithm

Let  $T_0$  be the initial tableau:

$$T_0 = \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right].$$

Let  $T_k$  denote the tableau after  $k$  pivots:

$$T_k = \left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right]$$

$T_k$  is obtained from  $T_0$  by multiplying it on the left by a product of Gaussian pivot matrices  $G := G_k G_{k-1} \cdots G_1$ :

$$GT_0 = T_k,$$

where  $G$  is invertible ( $G^{-1} = G_1^{-1} G_2^{-1} \cdots G_k^{-1}$ ).

# The Block Structure of the Simplex Algorithm

Let's investigate the structure of  $T_k$  by examining the consequence of this product in terms of the block structure of  $T_0$  and  $T_k$ .

$$T_0 = \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] \quad T_k = \left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right]$$

Here we use the fact that the first column of the simplex tableau remains unchanged by pivoting.

# The Block Structure of the Simplex Algorithm

Let's investigate the structure of  $T_k$  by examining the consequence of this product in terms of the block structure of  $T_0$  and  $T_k$ .

$$T_0 = \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] \quad T_k = \left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right]$$

Here we use the fact that the first column of the simplex tableau remains unchanged by pivoting.

First we must decompose  $G$  into a block structure that is conformal to that of  $T_0$ :

# The Block Structure of the Simplex Algorithm

Let's investigate the structure of  $T_k$  by examining the consequence of this product in terms of the block structure of  $T_0$  and  $T_k$ .

$$T_0 = \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] \quad T_k = \left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right]$$

Here we use the fact that the first column of the simplex tableau remains unchanged by pivoting.

First we must decompose  $G$  into a block structure that is conformal to that of  $T_0$ :

$$G = \begin{bmatrix} M & u \\ v^T & \beta \end{bmatrix},$$

where  $M \in \mathbb{R}^{m \times m}$ ,  $u, v \in \mathbb{R}^m$ , and  $\beta \in \mathbb{R}$ .

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right] = T_k$$

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right] = T_k = GT_0$$

# Block Structure and Matrix Multiplication

$$\begin{aligned} \left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^\top & -y^\top & \hat{z} \end{array} \right] &= T_k = GT_0 \\ &= \begin{bmatrix} M & u \\ v^\top & \beta \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^\top & 0 & 0 \end{array} \right] \end{aligned}$$

# Block Structure and Matrix Multiplication

$$\begin{aligned} \left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right] &= T_k = GT_0 \\ &= \begin{bmatrix} M & u \\ v^T & \beta \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] \\ &= \left[ \begin{array}{ccc|c} -u & MA + uc^T & M & Mb \\ -\beta & v^T A + \beta c^T & v^T & v^T b \end{array} \right]. \end{aligned}$$



# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^\top & -y^\top & \hat{z} \end{array} \right] = \left[ \begin{array}{ccc|c} -u & MA + uc^\top & M & Mb \\ -\beta & v^\top A + \beta c^\top & v^\top & v^\top b \end{array} \right]$$

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^\top & -y^\top & \hat{z} \end{array} \right] = \left[ \begin{array}{ccc|c} -u & MA + uc^\top & M & Mb \\ -\beta & v^\top A + \beta c^\top & v^\top & v^\top b \end{array} \right]$$

Equating terms on the left and right gives

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right] = \left[ \begin{array}{ccc|c} -u & MA + uc^T & M & Mb \\ -\beta & v^T A + \beta c^T & v^T & v^T b \end{array} \right]$$

Equating terms on the left and right gives

$$u = 0$$

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right] = \left[ \begin{array}{ccc|c} -u & MA + uc^T & M & Mb \\ -\beta & v^T A + \beta c^T & v^T & v^T b \end{array} \right]$$

Equating terms on the left and right gives

$$u = 0 \quad \beta = 1$$

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{cc|c} 0 & \hat{A} & R \\ -1 & \hat{c}^T & -y^T \end{array} \middle| \begin{array}{c} \hat{b} \\ \hat{z} \end{array} \right] = \left[ \begin{array}{cc|c} -u & MA + uc^T & M \\ -\beta & v^T A + \beta c^T & v^T \end{array} \middle| \begin{array}{c} Mb \\ v^T b \end{array} \right]$$

Equating terms on the left and right gives

$$\begin{aligned} u &= 0 & \beta &= 1 \\ M &= R \end{aligned}$$

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^T & -y^T & \hat{z} \end{array} \right] = \left[ \begin{array}{ccc|c} -u & MA + uc^T & M & Mb \\ -\beta & v^T A + \beta c^T & v^T & v^T b \end{array} \right]$$

Equating terms on the left and right gives

$$\begin{aligned} u &= 0 & \beta &= 1 \\ M &= R & \text{and } v &= -y. \end{aligned}$$

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^\top & -y^\top & \hat{z} \end{array} \right] = \left[ \begin{array}{ccc|c} -u & MA + uc^\top & M & Mb \\ -\beta & v^\top A + \beta c^\top & v^\top & v^\top b \end{array} \right]$$

Equating terms on the left and right gives

$$\begin{aligned} u &= 0 & \beta &= 1 \\ M &= R & \text{and } v &= -y. \end{aligned}$$

Therefore,

$$T_k = \begin{bmatrix} R & 0 \\ -y^\top & 1 \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^\top & 0 & 0 \end{array} \right] = \left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^\top - y^\top A & -y^\top & -y^\top b \end{array} \right],$$

where the matrix  $R$  is necessarily invertible.

# Block Structure and Matrix Multiplication

$$\left[ \begin{array}{ccc|c} 0 & \hat{A} & R & \hat{b} \\ -1 & \hat{c}^\top & -y^\top & \hat{z} \end{array} \right] = \left[ \begin{array}{ccc|c} -u & MA + uc^\top & M & Mb \\ -\beta & v^\top A + \beta c^\top & v^\top & v^\top b \end{array} \right]$$

Equating terms on the left and right gives

$$\begin{aligned} u &= 0 & \beta &= 1 \\ M &= R & \text{and } v &= -y. \end{aligned}$$

Therefore,

$$T_k = \begin{bmatrix} R & 0 \\ -y^\top & 1 \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^\top & 0 & 0 \end{array} \right] = \left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^\top - y^\top A & -y^\top & -y^\top b \end{array} \right],$$

where the matrix  $R$  is necessarily invertible. ( $R \sim$  record matrix)



# The Block Structure of an Optimal Tableau

$$T_k = \begin{bmatrix} R & 0 \\ -y^T & 1 \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] = \left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

# The Block Structure of an Optimal Tableau

$$T_k = \begin{bmatrix} R & 0 \\ -y^T & 1 \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] = \left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

$T_k$  is an optimal tableau if and only if

# The Block Structure of an Optimal Tableau

$$T_k = \begin{bmatrix} R & 0 \\ -y^T & 1 \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] = \left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

$T_k$  is an optimal tableau if and only if

- it is feasible:  $0 \leq Rb$ , and

# The Block Structure of an Optimal Tableau

$$T_k = \begin{bmatrix} R & 0 \\ -y^T & 1 \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] = \left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

$T_k$  is an optimal tableau if and only if

- it is feasible:  $0 \leq Rb$ , and
- the z-row has non-positive entries:

$$\begin{aligned} c - A^T y &\leq 0 && \text{or equivalently} && A^T y \geq c \\ -y &\leq 0 && \text{or equivalently} && 0 \leq y. \end{aligned}$$

# The Block Structure of an Optimal Tableau

$$T_k = \begin{bmatrix} R & 0 \\ -y^T & 1 \end{bmatrix} \left[ \begin{array}{ccc|c} 0 & A & I & b \\ -1 & c^T & 0 & 0 \end{array} \right] = \left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

$T_k$  is an optimal tableau if and only if

- it is feasible:  $0 \leq Rb$ , and
- the z-row has non-positive entries:

$$\begin{aligned} c - A^T y &\leq 0 && \text{or equivalently} && A^T y \geq c \\ -y &\leq 0 && \text{or equivalently} && 0 \leq y. \end{aligned}$$

In this case the optimal value  $= z = b^T y$ .

# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

with

$$0 \leq Rb,$$

# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

with

$$0 \leq Rb, \quad A^T y \geq c,$$



# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

with

$$0 \leq Rb, \quad A^T y \geq c, \quad 0 \leq y,$$

# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

with

$$0 \leq Rb, \quad A^T y \geq c, \quad 0 \leq y, \quad c^T x^* = z = b^T y,$$

# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

with

$$0 \leq Rb, \quad A^T y \geq c, \quad 0 \leq y, \quad c^T x^* = z = b^T y,$$

where  $x^*$  is the optimal solution to

$$\begin{array}{ll} \mathcal{P} & \max \quad c^T x \\ & \text{s.t.} \quad Ax \leq b \\ & \quad \quad 0 \leq x \end{array}$$

# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

with

$$0 \leq Rb, \quad A^T y \geq c, \quad 0 \leq y, \quad c^T x^* = z = b^T y,$$

where  $x^*$  is the optimal solution to

$$\begin{array}{ll} \mathcal{P} & \max \quad c^T x \\ & \text{s.t.} \quad Ax \leq b \\ & \quad \quad 0 \leq x \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \min \quad b^T y \\ & \text{s.t.} \quad A^T y \geq c \\ & \quad \quad 0 \leq y \end{array}$$

# The Block Structure of an Optimal Tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

with

$$0 \leq Rb, \quad A^T y \geq c, \quad 0 \leq y, \quad c^T x^* = z = b^T y,$$

where  $x^*$  is the optimal solution to

$$\begin{array}{ll} \mathcal{P} & \max \quad c^T x \\ & \text{s.t.} \quad Ax \leq b \\ & \quad \quad 0 \leq x \end{array} \qquad \begin{array}{ll} \mathcal{D} & \min \quad b^T y \\ & \text{s.t.} \quad A^T y \geq c \\ & \quad \quad 0 \leq y \end{array}$$

**WEAK DUALITY THM.  $\Rightarrow$  Y SOLVES  $\mathcal{D}$  !!!**

# Optimal Tableaus Yield Optimal Solutions

$$\begin{array}{ll} \mathcal{P} & \max \quad c^T x \\ & \text{s.t.} \quad Ax \leq b \\ & \quad \quad 0 \leq x \end{array} \qquad \begin{array}{ll} \mathcal{D} & \min \quad b^T y \\ & \text{s.t.} \quad A^T y \geq c \\ & \quad \quad 0 \leq y \end{array}$$

**Theorem:** If the simplex tableau

$$\left[ \begin{array}{ccc|c} 0 & RA & R & Rb \\ -1 & c^T - y^T A & -y^T & -y^T b \end{array} \right]$$

is optimal for  $\mathcal{P}$ , i.e. if  $x^*$  is the associated BFS and

$$0 \leq Rb, \quad A^T y \geq c, \quad 0 \leq y, \quad c^T x^* = z = b^T y,$$

then  $x^*$  is an optimal solution to  $\mathcal{P}$  and  $y$  is an optimal solution to  $\mathcal{D}$ .

# Plastic Cup Factory Reprised

$$\begin{array}{ll} \mathcal{P} & \text{maximize} \quad 25B + 20C \\ & \text{subject to} \quad 20B + 12C \leq 1800 \\ & \quad \quad \quad \frac{1}{15}B + \frac{1}{15}C \leq 8 \\ & \quad \quad \quad 0 \leq B, C \end{array}$$

# Plastic Cup Factory Reprised

$$\begin{array}{ll} \mathcal{P} & \text{maximize} \quad 25B + 20C \\ & \text{subject to} \quad 20B + 12C \leq 1800 \\ & \quad \quad \quad \frac{1}{15}B + \frac{1}{15}C \leq 8 \\ & \quad \quad \quad 0 \leq B, C \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \text{minimize} \quad 1800R + 8L \\ & \text{subject to} \quad 20R + \frac{1}{15}L \geq 25 \\ & \quad \quad \quad 12R + \frac{1}{15}L \geq 20 \\ & \quad \quad \quad 0 \leq R, L \end{array}$$



# Plastic Cup Factory Reprised

$$\begin{array}{ll} \mathcal{P} & \text{maximize} \quad 25B + 20C \\ & \text{subject to} \quad 20B + 12C \leq 1800 \\ & \quad \quad \quad \frac{1}{15}B + \frac{1}{15}C \leq 8 \\ & \quad \quad \quad 0 \leq B, C \end{array}$$

$$\left[ \begin{array}{cccc|c} 20 & 12 & 1 & 0 & 1800 \\ \frac{1}{15} & \frac{1}{15} & 0 & 1 & 8 \\ \hline 25 & 20 & 0 & 0 & 0 \end{array} \right]$$

$$\begin{array}{ll} \mathcal{D} & \text{minimize} \quad 1800R + 8L \\ & \text{subject to} \quad 20R + \frac{1}{15}L \geq 25 \\ & \quad \quad \quad 12R + \frac{1}{15}L \geq 20 \\ & \quad \quad \quad 0 \leq R, L \end{array}$$

# Plastic Cup Factory Reprised

$$\begin{array}{ll} \mathcal{P} & \text{maximize} \quad 25B + 20C \\ & \text{subject to} \quad 20B + 12C \leq 1800 \\ & \quad \quad \quad \frac{1}{15}B + \frac{1}{15}C \leq 8 \\ & \quad \quad \quad 0 \leq B, C \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \text{minimize} \quad 1800R + 8L \\ & \text{subject to} \quad 20R + \frac{1}{15}L \geq 25 \\ & \quad \quad \quad 12R + \frac{1}{15}L \geq 20 \\ & \quad \quad \quad 0 \leq R, L \end{array}$$

$$\left[ \begin{array}{cccc|c} 20 & 12 & 1 & 0 & 1800 \\ \frac{1}{15} & \frac{1}{15} & 0 & 1 & 8 \\ \hline 25 & 20 & 0 & 0 & 0 \end{array} \right] \Rightarrow \left[ \begin{array}{cccc|c} 1 & 0 & 1/8 & -75/2 & 45 \\ 0 & 1 & -1/8 & 75/2 & 75 \\ \hline 0 & 0 & -5/8 & -375/2 & -2625 \end{array} \right]$$

# Plastic Cup Factory Reprised

$$\begin{array}{ll} \mathcal{P} & \text{maximize} \quad 25B + 20C \\ & \text{subject to} \quad 20B + 12C \leq 1800 \\ & \quad \quad \quad \frac{1}{15}B + \frac{1}{15}C \leq 8 \\ & \quad \quad \quad 0 \leq B, C \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \text{minimize} \quad 1800R + 8L \\ & \text{subject to} \quad 20R + \frac{1}{15}L \geq 25 \\ & \quad \quad \quad 12R + \frac{1}{15}L \geq 20 \\ & \quad \quad \quad 0 \leq R, L \end{array}$$

$$\left[ \begin{array}{cccc|c} 20 & 12 & 1 & 0 & 1800 \\ \frac{1}{15} & \frac{1}{15} & 0 & 1 & 8 \\ \hline 25 & 20 & 0 & 0 & 0 \end{array} \right] \Rightarrow \left[ \begin{array}{cccc|c} 1 & 0 & 1/8 & -75/2 & 45 \\ 0 & 1 & -1/8 & 75/2 & 75 \\ \hline 0 & 0 & -5/8 & -375/2 & -2625 \end{array} \right]$$

$$(B, C)^* = (45, 75),$$

# Plastic Cup Factory Reprised

$$\begin{array}{ll} \mathcal{P} & \text{maximize} \quad 25B + 20C \\ & \text{subject to} \quad 20B + 12C \leq 1800 \\ & \quad \quad \quad \frac{1}{15}B + \frac{1}{15}C \leq 8 \\ & \quad \quad \quad 0 \leq B, C \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \text{minimize} \quad 1800R + 8L \\ & \text{subject to} \quad 20R + \frac{1}{15}L \geq 25 \\ & \quad \quad \quad 12R + \frac{1}{15}L \geq 20 \\ & \quad \quad \quad 0 \leq R, L \end{array}$$

$$\left[ \begin{array}{cccc|c} 20 & 12 & 1 & 0 & 1800 \\ \frac{1}{15} & \frac{1}{15} & 0 & 1 & 8 \\ \hline 25 & 20 & 0 & 0 & 0 \end{array} \right] \Rightarrow \left[ \begin{array}{cccc|c} 1 & 0 & 1/8 & -75/2 & 45 \\ 0 & 1 & -1/8 & 75/2 & 75 \\ \hline 0 & 0 & -5/8 & -375/2 & -2625 \end{array} \right]$$

$$(B, C)^* = (45, 75), \quad (R, L)^* = \left( \frac{5}{8}, \frac{375}{2} \right),$$

# Plastic Cup Factory Reprised

$$\begin{array}{ll} \mathcal{P} & \text{maximize} \quad 25B + 20C \\ & \text{subject to} \quad 20B + 12C \leq 1800 \\ & \quad \quad \quad \frac{1}{15}B + \frac{1}{15}C \leq 8 \\ & \quad \quad \quad 0 \leq B, C \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \text{minimize} \quad 1800R + 8L \\ & \text{subject to} \quad 20R + \frac{1}{15}L \geq 25 \\ & \quad \quad \quad 12R + \frac{1}{15}L \geq 20 \\ & \quad \quad \quad 0 \leq R, L \end{array}$$

$$\left[ \begin{array}{cccc|c} 20 & 12 & 1 & 0 & 1800 \\ \frac{1}{15} & \frac{1}{15} & 0 & 1 & 8 \\ \hline 25 & 20 & 0 & 0 & 0 \end{array} \right] \Rightarrow \left[ \begin{array}{cccc|c} 1 & 0 & 1/8 & -75/2 & 45 \\ 0 & 1 & -1/8 & 75/2 & 75 \\ \hline 0 & 0 & -5/8 & -375/2 & -2625 \end{array} \right]$$

$$(B, C)^* = (45, 75), \quad (R, L)^* = \left( \frac{5}{8}, \frac{375}{2} \right), \quad z^* = 2625$$

## Another Duality Example

$$\begin{array}{ll} \mathcal{P} & \max \quad 4x_1 + 5x_2 + 3x_3 \\ & \text{s.t.} \quad x_1 + 4x_2 + 2x_3 \leq 11 \\ & \quad \quad 3x_1 + 2x_2 + x_3 \leq 5 \\ & \quad \quad 4x_1 + 2x_2 + 2x_3 \leq 8 \\ & \quad \quad 0 \leq x_1, x_2, x_3 \end{array}$$

## Another Duality Example

$$\begin{array}{ll} \mathcal{P} & \max \quad 4x_1 + 5x_2 + 3x_3 \\ & \text{s.t.} \quad x_1 + 4x_2 + 2x_3 \leq 11 \\ & \quad \quad 3x_1 + 2x_2 + x_3 \leq 5 \\ & \quad \quad 4x_1 + 2x_2 + 2x_3 \leq 8 \\ & \quad \quad 0 \leq x_1, x_2, x_3 \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \min \quad 11y_1 + 5y_2 + 8y_3 \\ & \text{s.t.} \quad y_1 + 3y_2 + 4y_3 \geq 4 \\ & \quad \quad 4y_1 + 2y_2 + 2y_3 \geq 5 \\ & \quad \quad 2y_1 + y_2 + 2y_3 \geq 3 \\ & \quad \quad 0 \leq y_1, y_2, y_3 \end{array}$$

# Another Duality Example

$$\begin{array}{ll} \mathcal{P} & \max \quad 4x_1 + 5x_2 + 3x_3 \\ & \text{s.t.} \quad x_1 + 4x_2 + 2x_3 \leq 11 \\ & \quad \quad 3x_1 + 2x_2 + x_3 \leq 5 \\ & \quad \quad 4x_1 + 2x_2 + 2x_3 \leq 8 \\ & \quad \quad 0 \leq x_1, x_2, x_3 \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \min \quad 11y_1 + 5y_2 + 8y_3 \\ & \text{s.t.} \quad y_1 + 3y_2 + 4y_3 \geq 4 \\ & \quad \quad 4y_1 + 2y_2 + 2y_3 \geq 5 \\ & \quad \quad 2y_1 + y_2 + 2y_3 \geq 3 \\ & \quad \quad 0 \leq y_1, y_2, y_3 \end{array}$$

$$T_0 = \left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & 2 & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right]$$



# Another Duality Example

$$T_0 = \left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & 2 & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right]$$

## Another Duality Example

$$T_0 = \left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & 2 & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right] \Rightarrow$$

## Another Duality Example

$$T_0 = \left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & 2 & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right] \Rightarrow$$
$$T_{\text{opt}} = \left[ \begin{array}{cccccc|c} -5 & 0 & 0 & 1 & -2 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & \frac{-1}{2} & 1 \\ 1 & 0 & 1 & 0 & -1 & 1 & 3 \\ \hline -4 & 0 & 0 & 0 & -2 & \frac{-1}{2} & -14 \end{array} \right]$$

## Another Duality Example

$$T_0 = \left[ \begin{array}{cccccc|c} 1 & 4 & 2 & 1 & 0 & 0 & 11 \\ 3 & 2 & 1 & 0 & 1 & 0 & 5 \\ 4 & 2 & 2 & 0 & 0 & 1 & 8 \\ \hline 4 & 5 & 3 & 0 & 0 & 0 & 0 \end{array} \right] \Rightarrow$$

$$T_{\text{opt}} = \left[ \begin{array}{cccccc|c} -5 & 0 & 0 & 1 & -2 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & \frac{-1}{2} & 1 \\ 1 & 0 & 1 & 0 & -1 & 1 & 3 \\ \hline -4 & 0 & 0 & 0 & -2 & \frac{-1}{2} & -14 \end{array} \right]$$

$$x^* = (0, 1, 3), \quad y^* = (0, 2, 1/2), \quad z^* = 14$$

# Another Duality Example

Check

$$\begin{array}{ll} \mathcal{P} & \max \quad 4x_1 + 5x_2 + 3x_3 \\ & \text{s.t.} \quad x_1 + 4x_2 + 2x_3 \leq 11 \\ & \quad \quad 3x_1 + 2x_2 + x_3 \leq 5 \\ & \quad \quad 4x_1 + 2x_2 + 2x_3 \leq 8 \\ & \quad \quad 0 \leq x_1, x_2, x_3 \end{array}$$

$$\begin{array}{ll} \mathcal{D} & \min \quad 11y_1 + 5y_2 + 8y_3 \\ & \text{s.t.} \quad y_1 + 3y_2 + 4y_3 \geq 4 \\ & \quad \quad 4y_1 + 2y_2 + 2y_3 \geq 5 \\ & \quad \quad 2y_1 + y_2 + 2y_3 \geq 3 \\ & \quad \quad 0 \leq y_1, y_2, y_3 \end{array}$$

$$x^* = (0, 1, 3), \quad y^* = (0, 2, 1/2), \quad z^* = 14$$

# Strong Duality

If we can now show that the simplex algorithm works, then we have an algorithm that simultaneously solves both the primal and dual problems.

# Strong Duality

If we can now show that the simplex algorithm works, then we have an algorithm that simultaneously solves both the primal and dual problems.

Moreover, the optimal value in the primal and dual coincides giving equality in the weak duality inequality.

# Strong Duality

If we can now show that the simplex algorithm works, then we have an algorithm that simultaneously solves both the primal and dual problems.

Moreover, the optimal value in the primal and dual coincides giving equality in the weak duality inequality.

We now focus on the details of the simplex algorithm to determine if and when it works.