### LINEAR PROGRAMMING

#### [V. CH6]: THE SIMPLEX METHOD IN MATRIX NOTATION

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#### LP IN MATRIX NOTATION

#### SIMPLEX METHOD IN MATRIX NOTATION Primal Simplex Algorithm Dual Simplex Algorithm Two-Phase Methods

## STANDARD FORM LP

As usual, we begin our discussion with the standard-form linear programming problem:

$$\begin{array}{ll} \max_{x} & \sum_{j=1}^{n} c_{j} x_{j} \\ \text{subject to} & \sum_{j=1}^{n} a_{ij} x_{j} \leq b_{i}, \quad i=1,2,\cdots,m \\ & x_{j} \geq 0, \quad j=1,2,\cdots,n \end{array}$$

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It is convenient to introduce slack variables as follows:

$$\frac{x_{n+i}}{x_{n+i}} = b_i - \sum_{j=1}^n a_{ij} x_j, \qquad i = 1, \dots, m$$

 $w_i$  renamed as  $x_{n+i}$ 

# MATRIX FORM

With these slack variables, we now write our problem in matrix form:

$$\max_{x} c^{T} x$$
  
subject to  $Ax = b$   
 $x \ge 0$ 

where

$$A = \begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} & 1 & & \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} & 1 & & \\ \vdots & \vdots & \ddots & \vdots & & \ddots & \\ a_{m,1} & a_{m,2} & \cdots & a_{m,n} & & & 1 \end{pmatrix}, b = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}, c = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \\ 0 \\ \vdots \\ 0 \end{pmatrix}, x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \\ x_{n+1} \\ \vdots \\ x_{n+m} \end{pmatrix}$$

## SEPARATING BASIC AND NON-BASIC PARTS

Consider an iteration of the Simplex algorithm where  ${\cal B}$  and  ${\cal N}$  are the set of basic and non-basic indices.

The *i*th component of *Ax* can be broken up into a *basic* and a *nonbasic* part:

$$\sum_{j=1}^{n+m} a_{ij}x_j = \sum_{j\in\mathcal{B}} a_{ij}x_j + \sum_{j\in\mathcal{N}} a_{ij}x_j.$$

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To break up the matrix product *Ax* analogously:

- Let *B* denote an  $m \times m$  matrix whose columns are indexed by  $\mathcal{B}$ .
- Similarly, let N denote an  $m \times n$  matrix whose columns are indexed by  $\mathcal{N}$ .

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Now, one could write *A* and *x* in a partitioned-matrix form as:

$$A = \begin{bmatrix} B & N \end{bmatrix}, x = \begin{bmatrix} x_{\mathcal{B}} \\ x_{\mathcal{N}} \end{bmatrix}$$

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This is a rearrangement (relabeling of the variables) so that basic columns/variables are listed first, followed by the nonbasic columns/variables. So technically, the equality is not correct.

### SIMPLIFYING

Now, we can write:

$$Ax = \begin{bmatrix} B & N \end{bmatrix} \begin{bmatrix} x_{\mathcal{B}} \\ x_{\mathcal{N}} \end{bmatrix} = Bx_{\mathcal{B}} + Nx_{\mathcal{N}}.$$

Check that this equality holds (and matrices have the right format).

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$$c^{T}x = \begin{bmatrix} c_{\mathcal{B}} \\ c_{\mathcal{N}} \end{bmatrix}^{T} \begin{bmatrix} x_{\mathcal{B}} \\ x_{\mathcal{N}} \end{bmatrix} = c_{\mathcal{B}}^{T}x_{\mathcal{B}} + c_{\mathcal{N}}^{T}x_{\mathcal{N}}.$$

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**Example:** Write the following LP in matrix form, then calculate the above values for  $\mathcal{B} = \{1, 2\}$ 

$$\max_{x} 3x_{1} + 4x_{2} - 2x_{3}$$
subject to
$$x_{1} + 0.5x_{2} - 5x_{3} \le 2$$

$$2x_{1} - x_{2} + 3x_{3} \le 3$$

$$x_{1}, x_{2}, x_{3} \ge 0$$

SIMPLEX METHOD IN MATRIX NOTATION

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Primal Simplex Algorithm Dual Simplex Algorithm Two-Phase Methods

# DICTIONARIES IN MATRIX FORM: BASIC VARIABLES

A dictionary has the property that the

basic variables are written as functions of the nonbasic variables.

In matrix notation, we see that the constraint equations Ax = b can be written as

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 $x_{\mathcal{B}}$  can be written as a function of the nonbasic variables  $x_{\mathcal{N}}$  iff the matrix B is *invertible*,

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The fact that *B* must be invertible means that its *m* column vectors are linearly independent and therefore form a basis for  $\mathbb{R}^m$ . This is why the basic variables are called *basic*.

## DICTIONARIES IN MATRIX FORM: OBJECTIVE

Similarly, the objective function can be written as

$$\begin{aligned} \zeta &= c_{\mathcal{B}}^T x_{\mathcal{B}} + c_{\mathcal{N}}^T x_{\mathcal{N}} \\ &= c_{\mathcal{B}}^T \left( B^{-1} b - B^{-1} N x_{\mathcal{N}} \right) + c_{\mathcal{N}}^T x_{\mathcal{N}} \\ &= c_{\mathcal{B}}^T B^{-1} b - \left( \left( B^{-1} N \right)^T c_{\mathcal{B}} - c_{\mathcal{N}} \right)^T x_{\mathcal{N}} \end{aligned}$$

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Putting all together, we can write the dictionary associated with basis  $\mathcal B$  as

$$\zeta = c_{\mathcal{B}}^T B^{-1} b - \left( \left( B^{-1} N \right)^T c_{\mathcal{B}} - c_{\mathcal{N}} \right)^T x_{\mathcal{N}}$$
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Comparing against the component-form notation, we make the following identifications:

$$c_{\mathcal{B}}^{T}B^{-1}b = \bar{\zeta}$$
$$c_{\mathcal{N}} - (B^{-1}N)^{T}c_{\mathcal{B}} = [\bar{c}_{j}]$$
$$B^{-1}b = [\bar{b}_{i}]$$
$$B^{-1}N = [\bar{a}_{ij}]$$

bracketed expressions on the right denote vectors and matrices with the index *i* running over  $\mathcal{B}$  and the index *j* running over  $\mathcal{N}$ .

## MATRIX FORM: BASIC SOLUTIONS

Recall: We rewrote dictionaries in matrix form as

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The basic solution associated with this dictionary is obtained by setting  $x_N$  equal to zero.

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Solving  $Bx_{\mathcal{B}} = b$  for some *b* may give us a feasible (or infeasible) solution:  $x_{\mathcal{B}}^*$  might be negative!

## EXAMPLE TIME

As an example, consider the same LP

$\max_x$	$3x_1 +$	$4x_2 -$	$2x_3$
subject to	$x_1 + $	$0.5x_2 - $	$5x_3 \le 2$
	$2x_1 - $	$x_2 +$	$3x_3 \leq 3$
	$x_1,$	$x_2,$	$x_3 \ge 0$

Write the initial dictionary. Do first pivot. You get a new basis  $\mathcal{B} = \{2, 5\}$ . Compute different part of this dictionary in matrix form and compare.

## TOWARDS DUAL SIMPLEX

$$\zeta = c_{\mathcal{B}}^T B^{-1} b - \left( \left( B^{-1} N \right)^T c_{\mathcal{B}} - c_{\mathcal{N}} \right)^T x_{\mathcal{N}}$$
$$x_{\mathcal{B}} = B^{-1} b \qquad -B^{-1} N x_{\mathcal{N}}$$

To write the associated dual dictionary using the negative transpose property, it is important to correctly associate *complementary pairs of variables*.

Recall that, we have appended the primal slack variables to the end of the original variables:

$$(x_1,\ldots,x_n,w_1,\ldots,w_m) \rightarrow (x_1,\ldots,x_n,x_{n+1},\ldots,x_{n+m})$$

Recall that,

- dual slacks are complementary to the primal originals, and
- dual originals are complementary to the primal slacks

using similar index for complementary variables,

$$(z_1,\ldots,z_n,y_1,\ldots,y_m) \rightarrow (z_1,\ldots,z_n,z_{n+1},\ldots,z_{n+m})$$

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## DUAL DICTIONARY: NEGATIVE TRANSPOSE PROPERTY

Primal dictionary in matrix form:

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and the corresponding dual dictionary:

$$-\xi = -c_{\mathcal{B}}^T B^{-1} b \qquad - (B^{-1}b)^T z_{\mathcal{B}}$$
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$$z_{\mathcal{N}} = \left( (B^{-1}N)^T c_{\mathcal{B}} - c_{\mathcal{N}} \right) + (B^{-1}N)^T z_{\mathcal{B}}$$

The dual solution associated with this is obtained by setting  $z_{\mathcal{B}}$  equal to zero:

$$z_{\mathcal{B}}^* = 0, \quad z_{\mathcal{N}}^* = \left(B^{-1}N\right)^T c_{\mathcal{B}} - c_{\mathcal{N}}$$

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## SUCCINCT PRIMAL AND DUAL DICTIONARY

Using the shorthands

$$\begin{aligned} z^*_{\mathcal{N}} &= \left(B^{-1}N\right)^T c_{\mathcal{B}} - c_{\mathcal{N}}, \\ x^*_{\mathcal{B}} &= B^{-1}b \text{ and} \\ \zeta^* &= c^T_{\mathcal{B}}B^{-1}b, \end{aligned}$$

we write the primal dictionary as

$$\zeta = \zeta^* - (z_{\mathcal{N}}^*)^T x_{\mathcal{N}}$$
$$x_{\mathcal{B}} = x_{\mathcal{B}}^* - B^{-1} N x_{\mathcal{N}}$$

and dual dictionary as

$$-\xi = -\zeta^* - (x_{\mathcal{B}}^*)^T z_{\mathcal{B}}$$
$$z_{\mathcal{N}} = z_{\mathcal{N}}^* + (B^{-1}N)^T z_{\mathcal{B}}$$

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**Assumptions:** Given initial  $\mathcal{B}$  with m elements, B invertible,  $x_{\mathcal{B}}^* = B^{-1}b \ge 0$ . Let us discuss the details of a single Simplex iteration (move to adjacent basis) in matrix notation.

#### STEP 1. Check for optimality

if  $z_N^* \ge 0$  stop // Primal feasibility and complementary is already maintained, just check dual feasibility.

#### STEP 2. Select entering variable

Else: pick  $j \in \mathcal{N}$  with  $z_j^* < 0$ . *|| variable*  $x_j$  *is the entering variable.* 

#### Primal Dictionary

$$\zeta = \zeta^* - (z_{\mathcal{N}}^*)^T x_{\mathcal{N}}$$
$$x_{\mathcal{B}} = x_{\mathcal{B}}^* - B^{-1} N x_{\mathcal{N}}$$

**Dual Dictionary** 

$$-\xi = -\zeta^* - (x_{\mathcal{B}}^*)^T z_{\mathcal{B}}$$
$$z_{\mathcal{N}} = z_{\mathcal{N}}^* + (B^{-1}N)^T z_{\mathcal{B}}$$

#### STEP 3. Compute primal step direction $\Delta x_{\mathcal{B}}$

Having selected the entering variable, we let

 $x_{\mathcal{N}} = te_j$ 

where  $e_j$  denote the unit vector that is zero in every component except for a one in the position associated with index j,

$$x_{\mathcal{B}} = x_{\mathcal{B}}^* - B^{-1}Nte_j$$

Hence, the step direction  $\Delta x_{\mathcal{B}}$  for the primal basic variables is given by

$$\Delta x_{\mathcal{B}} = B^{-1} N e_{j}$$

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#### STEP 4. Compute primal step length t

We wish to pick the largest  $t \ge 0$  for which

$$x_{\mathcal{B}}^* \ge t\Delta x_{\mathcal{B}}.$$

(every component of  $x_{\mathcal{B}}$  remains nonnegative)

Since, for each  $i \in \mathcal{B}^*, x_i^* \ge 0$  and  $t \ge 0$ 

$$\frac{1}{t} \geq \frac{\Delta x_i}{x_i^*}, \quad \forall i \in \mathcal{B}$$

Hence, the largest t for which all of the inequalities hold is given by

$$t = \left(\max_{i \in \mathcal{B}} \frac{\Delta x_i}{x_i^*}\right)^{-1} /\!\!/ \text{ convention for } \frac{0}{0} \text{ is to set such ratios to } 0$$

if  $t \leq 0$  stop *// primal is unbounded* 

STEP 5. Select leaving variable The leaving variable is chosen as any variable  $x_i, i \in B$ , for which the maximum in the calculation of t is obtained.

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#### Primal Dictionary

**Dual Dictionary** 

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$$x_{\mathcal{B}} = x_{\mathcal{B}}^* - B^{-1} N x_{\mathcal{N}}$$

$$\begin{aligned} -\xi &= -\zeta^* - \left(x_{\mathcal{B}}^*\right)^T z_{\mathcal{B}} \\ z_{\mathcal{N}} &= z_{\mathcal{N}}^* + \left(B^{-1}N\right)^T z_{\mathcal{B}} \end{aligned}$$

Essentially all that remains is to explain changes to the objective function.

#### STEP 6. Compute dual step direction $\Delta z_N$

Since in dual dictionary  $z_i$  is the entering variable, we see that

$$\Delta z_{\mathcal{N}} = -\left(B^{-1}N\right)^T e_i$$

#### STEP 7. Compute dual step length s

Since we know that *j* is the leaving variable in the dual dictionary,

$$s = \frac{z_j^*}{\Delta z_j}$$

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Primal Dictionary

**Dual Dictionary** 

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$$x_{\mathcal{B}} = x_{\mathcal{B}}^* - B^{-1} N x_{\mathcal{N}}$$

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$$z_{\mathcal{N}} = z_{\mathcal{N}}^* + (B^{-1}N)^T z_{\mathcal{B}}$$

We now have everything we need to update the data in the dictionary:

STEP 8. Update current primal and dual solutions

$$\begin{aligned} x_j^* &\leftarrow t \\ x_{\mathcal{B}}^* &\leftarrow x_{\mathcal{B}}^* - t\Delta x_{\mathcal{B}} \end{aligned}$$

$$\begin{aligned} z_i^* &\leftarrow s \\ z_{\mathcal{N}}^* &\leftarrow z_{\mathcal{N}}^* - s\Delta z_{\mathcal{N}} \end{aligned}$$

STEP 9. Update basis

$$\mathcal{B} \leftarrow (\mathcal{B} \setminus \{i\}) \cup \{j\}$$

As an example, let us solve the following LP:

$$\begin{array}{rcl}
\max_{x} & 4x_{1} + & 3x_{2} \\
\text{subject to} & x_{1} - & x_{2} \leq 1 \\
& 2x_{1} - & x_{2} \leq 3 \\
& + & x_{2} \leq 5 \\
& x_{1}, & x_{2} \geq 0
\end{array}$$

Lets now derive *Dual simplex* in matrix notation.

RECALL: Dual Simplex is to apply simplex to the dual LP but (indirectly) on a sequence of primal pivots

Here, instead of assuming that the primal dictionary is feasible ( $x_B \ge 0$ ), we now assume that the dual dictionary is feasible ( $z_N \ge 0$ ) and perform the analogous steps

#### Primal Dictionary

$$\zeta = \zeta^* - (z_{\mathcal{N}}^*)^T x_{\mathcal{N}}$$
$$x_{\mathcal{B}} = x_{\mathcal{B}}^* - B^{-1} N x_{\mathcal{N}}$$

STEP 1. Check for optimality

 $\text{if} \ x^*_{\mathcal{B}} \geq 0 \ \text{stop} \\$ 

#### STEP 2. Select Entering Variable

Else: pick  $i \in \mathcal{B}$  with  $x_i^* < 0$ .  $|| variable z_i is the entering variable.$ 

STEP 3. Compute Dual Step Direction  $\Delta z_N$ 

$$\Delta z_{\mathcal{B}} = -\left(B^{-1}N\right)^T e_i$$

STEP 4. Compute Dual Step Length, s

$$s = \left(\max_{j \in \mathcal{N}} \frac{\Delta z_j}{z_j^*}\right)^{-1}$$

if  $s \leq 0$  stop *|| dual is unbounded, primal infeasible*  Dual Dictionary

$$\begin{vmatrix} -\xi &= -\zeta^* - (x_{\mathcal{B}}^*)^T z_{\mathcal{B}} \\ z_{\mathcal{N}} &= z_{\mathcal{N}}^* + (B^{-1}N)^T z_{\mathcal{B}} \end{vmatrix}$$

STEP 5. Select Leaving Variable Pick  $z_j$ , where the maximum obtained.

STEP 6. Compute Primal Step Direction  $\Delta x_{\mathcal{B}}$  $\Delta x_{\mathcal{B}} = B^{-1} N e_j$ 

STEP 7. Compute Primal Step Length, t

$$t = \frac{x_i^*}{\Delta x_i}$$

STEP 8. Update Current Primal and Dual Solutions  $z_i^* \leftarrow s \qquad x_j^* \leftarrow t$  $z_N^* \leftarrow z_N^* - s\Delta z_N \qquad x_B^* \leftarrow x_B^* - t\Delta x_B$ 

STEP 9. Update Basis  $\mathcal{B} \leftarrow (\mathcal{B} \setminus \{i\}) \cup \{j\}$ 

Primal Simplex	Dual Simplex	
Suppose $x_{\mathcal{B}}^* \ge 0$	Suppose $z^*_{\mathcal{N}} \geq 0$	
while $(z_N^* \not\geq 0)$ {	while $(x_{\mathcal{B}}^* \not\geq 0)$ {	
pick $j \in \{j \in \mathcal{N} : z_j^* < 0\}$	pick $i \in \{i \in \mathcal{B} : x_i^* < 0\}$	
$\Delta x_{\mathcal{B}} = B^{-1} N e_j$	$\Delta z_{\mathcal{N}} = -(B^{-1}N)^T e_i$	
$t = \left(\max_{i \in \mathcal{B}} \frac{\Delta x_i}{x_i^*}\right)^{-1}$	$s = \left( \max_{j \in \mathcal{N}} \frac{\Delta z_j}{z_j^*} \right)^{-1}$	
pick $i \in \operatorname{argmax}_{i \in \mathcal{B}} \frac{\Delta x_i}{x_i^*}$	pick $j \in \operatorname{argmax}_{j \in \mathcal{N}} \frac{\Delta z_j}{z_i^*}$	
$\Delta z_{\mathcal{N}} = -(B^{-1}N)^T e_i$	$\Delta x_{\mathcal{B}} = B^{-1} N e_j$	
$s = \frac{z_j}{\Delta z_j}$	$t = \frac{x_i^*}{\Delta x_i}$	
$x_j^* \leftarrow t$	$x_j^* \leftarrow t$	
$x^*_{\mathcal{B}} \leftarrow x^*_{\mathcal{B}} - t\Delta x_{\mathcal{B}}$	$x_{\mathcal{B}}^* \leftarrow x_{\mathcal{B}}^* - t\Delta x_{\mathcal{B}}$	
$z_i^* \leftarrow s$	$z_i^* \leftarrow s$	
$z_{\mathcal{N}}^{*} \leftarrow z_{\mathcal{N}}^{*} - s\Delta z_{\mathcal{N}}$	$z_{\mathcal{N}}^{*} \leftarrow z_{\mathcal{N}}^{*} - s\Delta z_{\mathcal{N}}$	
$\mathcal{B} \leftarrow \mathcal{B} \setminus \{i\} \cup \{j\}$	$\mathcal{B} \leftarrow \mathcal{B} \setminus \{i\} \cup \{j\}$	
}	}	

Initially we set

$$\mathcal{N} = \{1, 2, \dots, n\}$$
,  $\mathcal{B} = \{n + 1, n + 2, \dots, n + m\}$ 

So we have

$$N = \begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m,1} & a_{m,2} & \cdots & a_{m,n} \end{pmatrix} \quad , \quad B = \begin{pmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{pmatrix}$$

and

$$c_{\mathcal{N}} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix} \quad , \quad c_{\mathcal{B}} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

Substituting these expressions into the definitions of  $x^*_{\mathcal{B}}, z^*_{\mathcal{N}}$  and  $\zeta^*$ , we find that

$$x_{\mathcal{B}}^* = B^{-1}b = b$$
 ,  $z_{\mathcal{N}}^* = (B^{-1}N)^T c_{\mathcal{B}} - c_{\mathcal{N}} = -c_{\mathcal{N}}$  ,  $\zeta^* = 0$ 

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$$\zeta = \zeta^* - (z_{\mathcal{N}}^*)^T x_{\mathcal{N}}$$
$$x_{\mathcal{B}} = x_{\mathcal{B}}^* - B^{-1} N x_{\mathcal{N}}$$

$$x_{\mathcal{B}}^* = B^{-1}b = b$$
 ,  $z_{\mathcal{N}}^* = (B^{-1}N)^T c_{\mathcal{B}} - c_{\mathcal{N}} = -c_{\mathcal{N}}$  ,  $\zeta^* = 0$ 

Hence, the initial primal dictionary reads

$$\zeta = c_{\mathcal{N}}^T x_{\mathcal{N}}$$
$$x_{\mathcal{B}} = b - N x_{\mathcal{N}}$$

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LINEAR PROGRAMMING

$$\zeta = c_{\mathcal{N}}^T x_{\mathcal{N}}$$
$$x_{\mathcal{B}} = b - N x_{\mathcal{N}}$$

If  $b \ge 0$  and  $c \le 0$ *// primal feasibility* + optimality observed  $\Rightarrow$  stop

If  $b \ge 0$  and  $c \le 0$ // primal feasibility observed  $\Rightarrow$  start with primal simplex

If  $b \not\geq 0$  and  $c \leq 0$ //dual feasibility observed  $\Rightarrow$  start with dual simplex

If  $b \not\geq 0$  and  $c \not\leq 0$ *// we need to do a Phase I* 

$$\zeta = c_{\mathcal{N}}^T x_{\mathcal{N}} x_{\mathcal{B}} = b - N x_{\mathcal{N}}$$

If  $b \not\geq 0$  and  $c \not\leq 0$ 

- a Phase I could be :
- → Define the *auxiliary problem*, as in Ch. 2, and apply primal simplex. // In Phase II, proceed with primal simplex.
- → Replace  $c_N$  with a *non-positive* one and apply dual simplex. // In Phase II, proceed with primal simplex.
- → Replace b with a non-negative one and apply primal simplex. // In Phase II, proceed with dual simplex.