

# Ch. 7: Primal simplex algorithm

Standard form:

$$\begin{cases} \min c_B^T x_B + c_N^T x_N \\ B x_B + N x_N = b \quad | B^{-1} \\ x_B, x_N \geq 0 \end{cases}$$

$$\rightarrow x_B = B^{-1} b - B^{-1} N x_N, \quad x_N = 0$$

$$\rightarrow z = \underbrace{c_B^T \cdot B^{-1} b}_{y^T} + \underbrace{(c_N^T - c_B^T B^{-1} N)}_{\text{slack of dual constraints}} x_N$$

$y^T$ : dual variables

slack of dual constraints.

- dual LP:

$$\begin{cases} \max y^T b \\ y^T A \leq c^T \end{cases}$$

$$\text{For } y^T = c_B^T B^{-1}:$$

$$y^T B = c_B^T$$

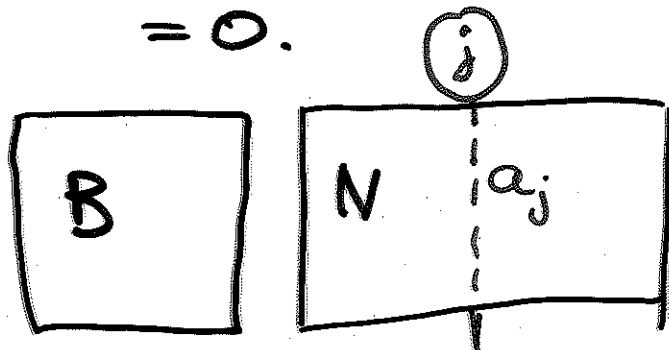
$$(c_N^T - y^T N) \begin{cases} \geq 0 \\ < 0 \end{cases}$$

optimal  
opportunity to improve  
the cost of primal.

## Primal simplex (c'ed)

If  $B$  is the current basis, we write  $z(B)$  for the cost corresponding to  $B$ .

$$z(B) = y(B)^T b + \underbrace{(c_N^T - y(B)^T N)}_{=0} \cdot x_N$$



$$P_j = d_j^1 - y(B)^T a_j < 0$$

If  $\theta > 0$  then we hope to reduce  $z(B)$  in value.



How to choose  $\theta$ ?

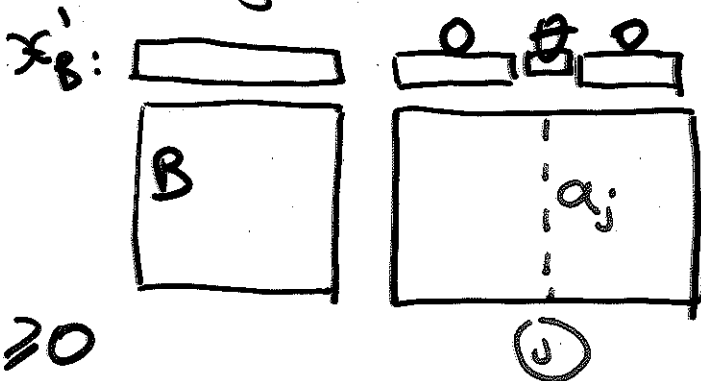
$$x'_B = B^{-1} b - B^{-1} a_j \theta \geq 0$$

new value for variables  $B$ .

maintain primal feasibility.

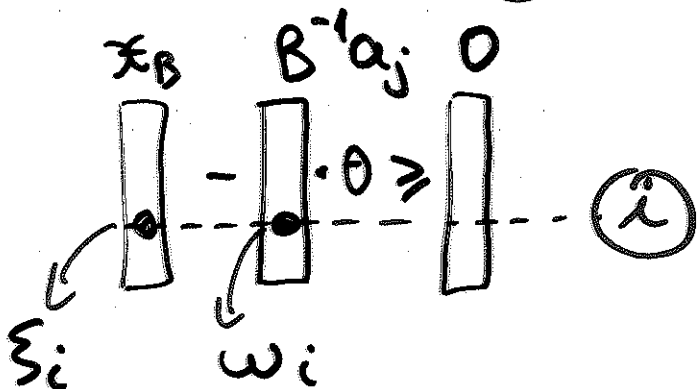
# Primal simplex (c'ed)

Maintain primal feasibility (c'ed):



$$x'_B = x_B - B^{-1}a_j \theta \geq 0$$

$$\theta = \min_{i \in B} \frac{\xi_i}{\omega_i}, \quad \omega_i > 0$$



Q: If  $\omega_i \leq 0$  for all  $i \in B$ ?

Then  $\theta \rightarrow \infty$  maintains feasibility of the primal (problem is unbounded; its solution cost =  $-\infty$ )

Q: If  $\xi_i = 0$  for some  $i \in B$ ?

Degenerate solution,  $\theta = 0$ ; cost of primal is not improved (only basis changes).

# Primal simplex (class) implementation:

a) Find a basic feasible solution:

$$\text{basis } B \text{ and } x_B = B^{-1} \cdot b \geq 0.$$

b) Recap what we need:

- $x_B = B^{-1} b$  : primal
- $z(B) = c_B^T B^{-1} b$  : cost
- $y(B) = c_B^T B^{-1}$  : dual

## implementation 1

- compute  $B^{-1}$
- compute  $y(B), P_j(y(B))$   
OPT ← entering variable in B.
- let  $j =$  entering variable.
- compute  $\theta$ , exit variable  $i$
- $B \leftarrow (B \setminus \{i\}) \cup \{j\}$ .
- 

Obs: time consuming; compute  $B^{-1}$  from scratch while  $B'$  not too different from  $B$ .

# Primal simplex (c'ed)

## implementation 2

- use only row matrix operations!

$$\begin{cases} Bx_B = b \\ c_B^T x_B + z = 0, \quad -z = \text{objective} \end{cases}$$

$$\left( \begin{array}{c|c} B & 0 \\ \hline c_B^T & 1 \end{array} \right) \begin{bmatrix} x_B \\ z \end{bmatrix} = \begin{bmatrix} b \\ 0 \end{bmatrix}, \quad z: \text{considered unknown.}$$

↓  
solve by computing the inverse of coefficient matrix.

Digression 1 Computing inverse of a matrix  $Q$ :

$$(Q | I_n)$$

⚡ row operations

$$(I_n | Q^{-1})$$

why?

Primal simplex;

implementation 2 (c'col)

pivot rows  
do NOT  
affect  
columns

$$\left( \begin{array}{c|c} B & 0 \\ \hline c_B^T & 1 \end{array} \right)^{-1} = \left( \begin{array}{c|c} Q & 0 \\ \hline q^T & 1 \end{array} \right)$$

, therefore we get the inverse.

Q: What are Q, q<sup>T</sup>?

$$\left( \begin{array}{cc} B & 0 \\ c_B^T & 1 \end{array} \right) \left( \begin{array}{cc} Q & 0 \\ q^T & 1 \end{array} \right) = \left( \begin{array}{cc} I & 0 \\ 0 & 1 \end{array} \right)$$

$$= \left( \begin{array}{cc} B \cdot Q & 0 \\ c_B^T Q + q^T & 1 \end{array} \right)$$

Therefore

$$Q = B^{-1}$$

$$q^T = -c_B^T B^{-1} = -y(B)^T$$

nice!

# Primal simplex

## implementation 2 (c'col)

Step 1: initialize the tableau & compute its inverse by row operations.

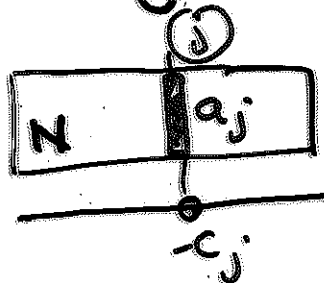
$$\left( \begin{array}{cc|c} B & 0 & b \\ c_B^T & 1 & 0 \end{array} \right) \quad (\text{together with the RHS})$$

inverse:  $\begin{matrix} \swarrow \\ \downarrow \\ \searrow \end{matrix}$  row op. (Gauss-Jordan elim)

$$\left( \begin{array}{cc|c} B^{-1} & 0 & x_B \\ -y^T(B) & 1 & -z \end{array} \right)$$

Step 2: use  $y^T(B)$  to find reduced costs of non-basic variables.

$$\forall j \in N, \quad P_j = c_j - y^T(B) \cdot a_j$$



$\nearrow$  OPT

$\searrow$  j: var. entering B.

# Primal simplex

## implementation 2 (ced)

Step 3:  $j$  enters the basis, update the tableau.

$$\left( \begin{array}{cc|c|c} B^{-1} & 0 & B^{-1}a_j & x_B \\ -y^i(B) & 1 & p_j & -z \end{array} \right)$$

↑  
entering column ( $j$ )

Recall that

if  $B = (b_1, b_2, \dots, b_m)$ , then  
↓  
columns of  $B$

$$B^{-1}a_j = \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_m \end{pmatrix} \text{ where } a_j = w_1 b_1 + w_2 b_2 + \dots + w_m b_m$$

Step 4: compute the basic var.  $i$  leaving the basis:

$$\theta = \min_{l \in B} \frac{x_l}{w_l}, w_l > 0$$

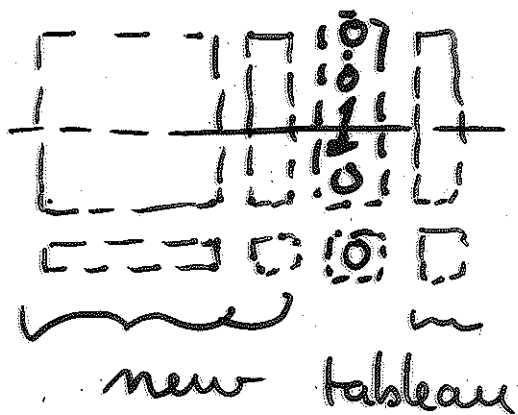
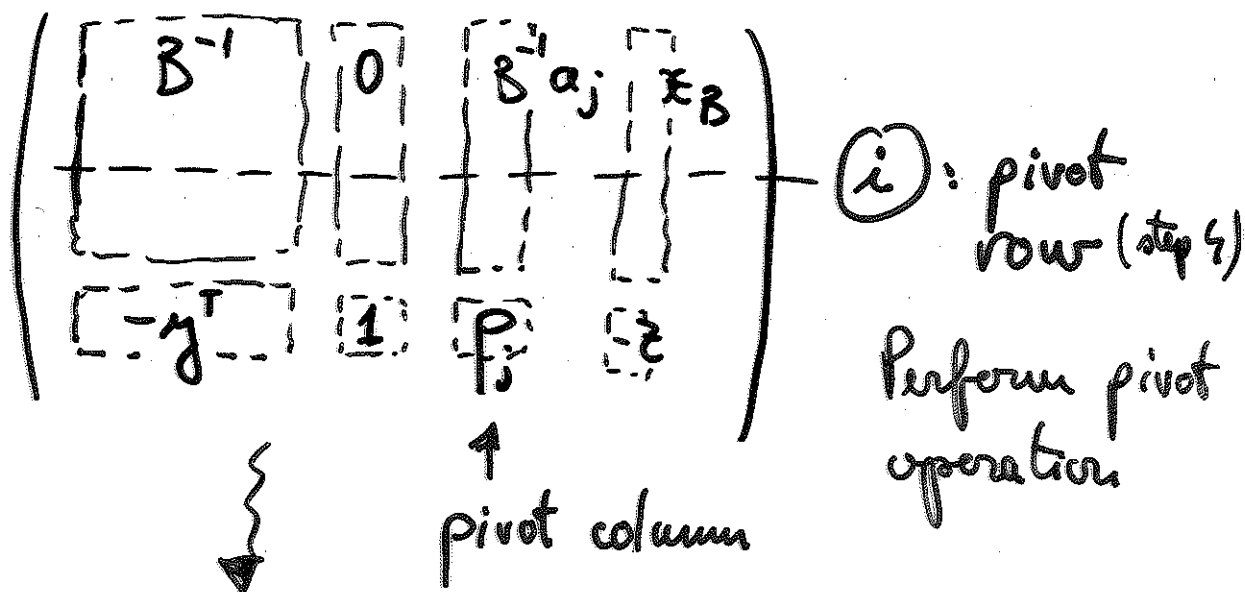
the basis:  $\left( \begin{array}{c|c|c} \dots & B^{-1}a_j & x_B \\ \hline & p_j & -z \end{array} \right)$



# Primal simplex

## Implementation 2 (c'ed)

Step 5: Update the inverse of the new tableau:



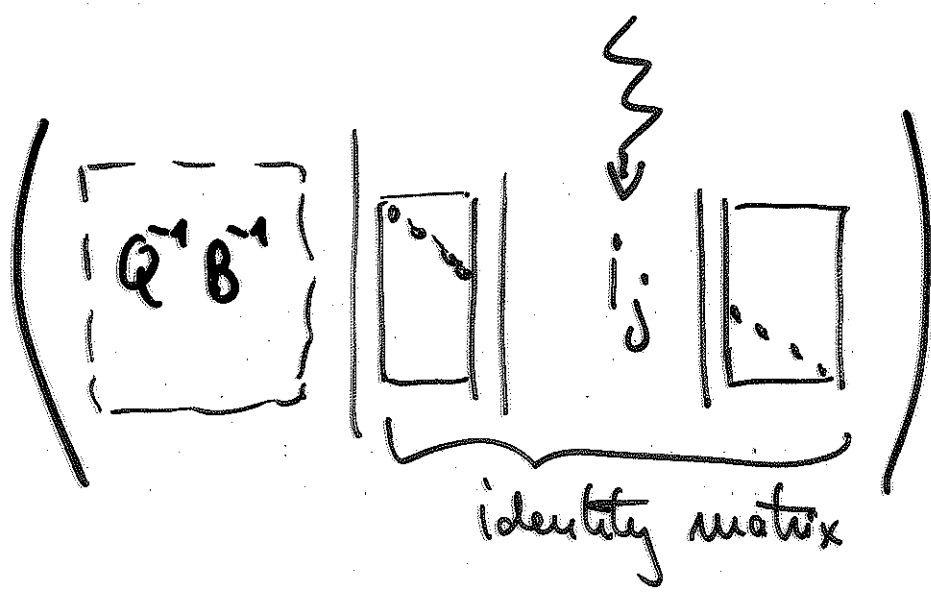
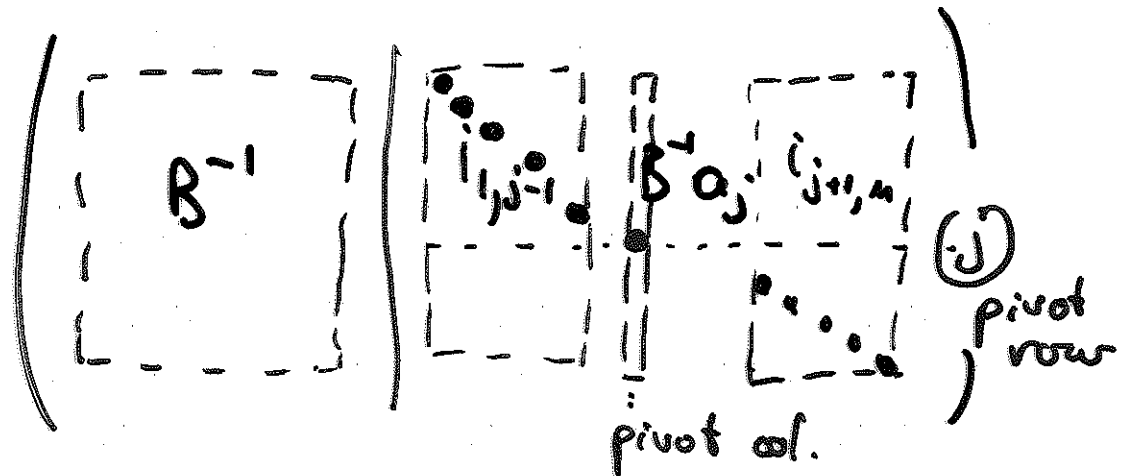
Note: One pivot operation is faster than  $O(n)$  pivots necessary for calculating the inverse from scratch.



# Primal - simplex (ced)

## Diagram 2 (ced)

Computing  $C^{-1} = Q^{-1} \cdot B^{-1}$



(see diagram 1)

This explains Step 5 of primal (revised) simplex algorithm.

# Primal simplex (cont)

Q: How to find a basic feasible solution to start simplex?

$$\begin{bmatrix} A & \dots \end{bmatrix} \begin{matrix} x \\ \vdots \end{matrix} = \begin{matrix} b \\ \vdots \end{matrix}$$

$x \geq 0$

a) Section 3.3 explains how to obtain a basis from the set of column vectors of  $A$ .

Problem:  $x_B \neq 0$ !

b)

$$\begin{bmatrix} A & I_m \end{bmatrix} \begin{matrix} x \\ t \end{matrix} = \begin{matrix} b \\ \vdots \end{matrix}$$

nice basis

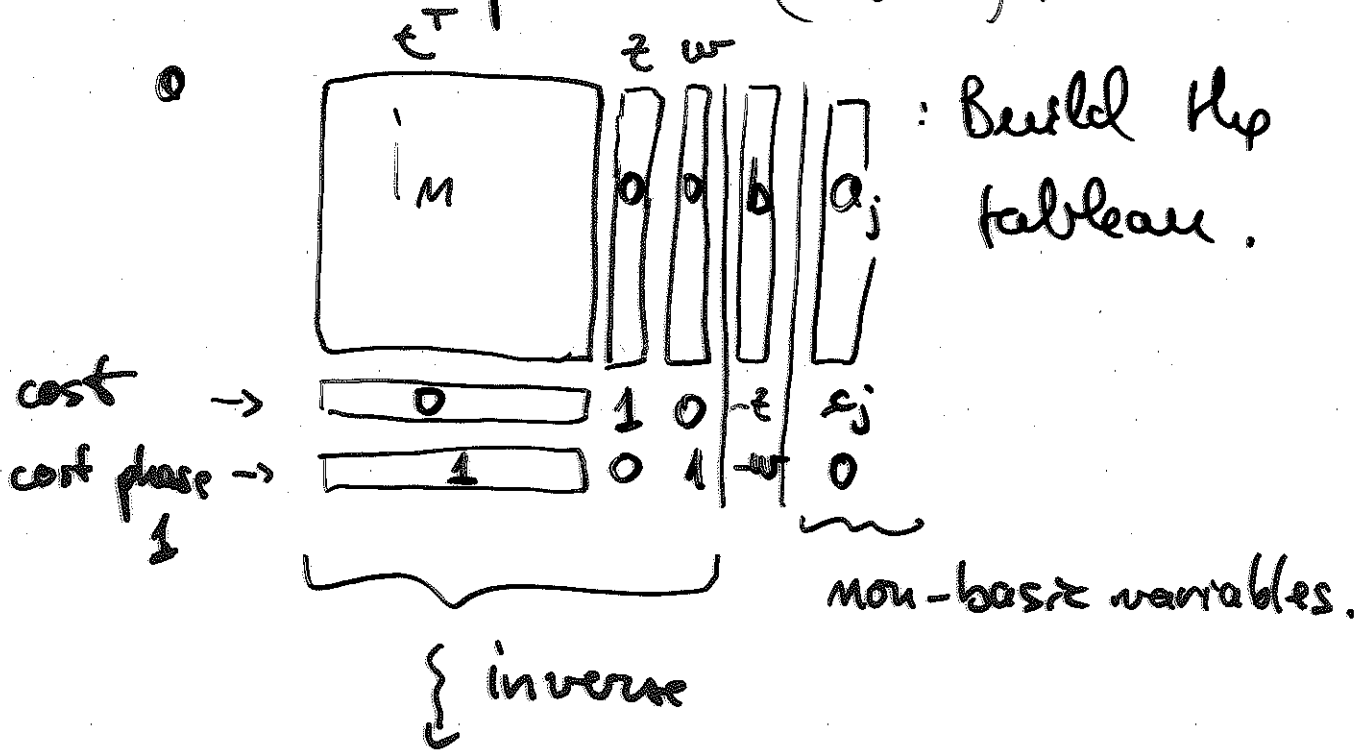
Cost (objective):  $\min 1^T t$ .

... this is "Phase I" of simplex algorithm...

# Primal simplex (ced)

## Phase I

- Make  $b$  positive ( $\times (-1)$ ).



$$\begin{pmatrix} I_n & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix}$$

duals

- Execute simplex on this tableau.  
Objective: eliminate all artificial variables  $t$  from basis and obtain objective cost  $w = 0$ .