

### 3.4 Simplex method

Given a LP in standard form:

$$\begin{aligned} \min \quad & z = c^T x \\ \text{s. t.} \quad & Ax = b \\ & x \geq 0 \end{aligned}$$

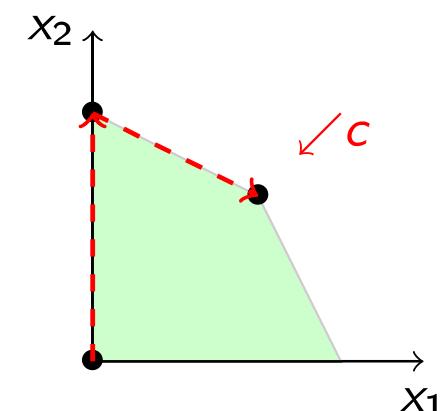
Examine a **sequence of basic feasible solutions** with non-increasing objective function values until an optimal solution is reached or the LP is found to be unbounded (G. Dantzig 1947).

At each iteration, we move from a basic feasible solution to a “**neighboring**” basic feasible solution.

Generate a path (i.e., a sequence of adjacent vertices) **along the edges** of the polyhedron of the feasible solutions until an **optimal vertex** is reached.



George Dantzig  
(1914-2005)



## 3.4 Simplex method

Given the correspondence between the basic feasible solutions and the vertices, we need to describe how to :

- Find an **initial vertex** or establish that the LP is infeasible. (Achieved by applying the same method to another LP, see later.)
- Determine whether the current vertex is **optimal**.
- Move from a current vertex to a **better adjacent vertex** (in terms of objective function value) or establish that the LP is **unbounded**.

## 3.4 Simplex method

### 3.4.1 Optimality test

Given a LP  $\min\{c^T x : Ax = b, x \geq 0\}$  and a feasible basis  $B$  of  $A$ ,  $Ax = b$  can be rewritten as

$$Bx_B + Nx_N = b \Rightarrow x_B = B^{-1}b - B^{-1}Nx_N$$

with  $B^{-1}b \geq 0$ .

Then, a **basic feasible solution** is such that  $x_B = B^{-1}b$ ,  $x_N = 0$ .

By substitution we express the objective function in terms of only the non basic variables (for the current basis  $B$ ):

$$\begin{aligned} c^T x &= (c_B^T \ c_N^T) \begin{bmatrix} x_B \\ x_N \end{bmatrix} = (c_B^T \ c_N^T) \begin{bmatrix} B^{-1}b - B^{-1}Nx_N \\ x_N \end{bmatrix} \\ &= c_B^T B^{-1}b - c_B^T B^{-1}Nx_N + c_N^T x_N \\ &= c_B^T B^{-1}b + (c_N^T - c_B^T B^{-1}N)x_N \end{aligned}$$

- Here,  $z_0 = c_B^T B^{-1}b$  is the cost of  $x = (x_B \ x_N)$ , where  $x_B = B^{-1}b$ ,  $x_N = 0$ .
- The **reduced costs** of the non basic variables  $x_N$  are denoted by  
 $\bar{c}_N^T := c_N^T - c_B^T B^{-1}N$ .

## 3.4 Simplex method

### Definition 17

$$\bar{c}^T = \underbrace{c^T - c_B^T B^{-1} A}_{\text{is the vector of reduced costs with respect to the basis } B.} = [\underbrace{c_B^T - c_B^T B^{-1} B}_{0^T}, \underbrace{c_N^T - c_B^T B^{-1} N}_{\bar{c}_N^T}]$$

- The reduced costs are defined also for basic variables, but  $\bar{c}_B = 0$ .
- $\bar{c}_j$  represents the change in the objective function value if non basic  $x_j$  would be increased from 0 to 1 while keeping all other non basic variables to 0.
- The solution value changes by  $\Delta z = \theta^* \bar{c}_j$ .

## 3.4 Simplex method

Consider a LP  $\min\{c^T x : Ax = b, x \geq 0\}$  and a feasible basis  $B$ .

### Proposition 1

If  $\bar{c}_N \geq 0$  then the basic feasible solution  $(x_B^T, x_N^T)$ , where  $x_B = B^{-1}b \geq 0$  and  $x_N = 0$ , of cost  $c_B^T B^{-1}b$  is a **global optimum**.

### Proof.

$\bar{c}^T \geq 0^T$  implies that

$$c^T x = c_B^T B^{-1}b + \bar{c}_N^T x_N \geq c_B^T B^{-1}b, \forall x \geq 0, Ax = b.$$



### Note:

- For maximization problems we check whether  $\bar{c}_N \leq 0$
- This optimality condition is sufficient but in general not necessary.

## 3.4 Simplex method

### Example

$$\min -x_1 - x_2$$

$$\text{s. t. } x_1 - x_2 + s_1 = 1$$

$$x_1 + x_2 + s_2 = 3$$

$$x_1, x_2, s_1, s_2 \geq 0$$

$$A = \begin{bmatrix} 1 & -1 & 1 & 0 \\ 1 & 1 & 0 & 1 \end{bmatrix}$$

$$c^T = [-1 \underline{-1} 0 \underline{0}]$$

Consider  $x_B = (\underline{x_1}, \overline{s_2}) = (1, 2)$ ,  $x_N = (\overline{x_2}, \underline{s_1})$ ,  $z = -1$ , so

value of objective function :  $z = C^T * x$   
meaning this is the value of the objective function for  
the current basic feasible solution.

$$B = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}, N = \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix}^{\text{non basis}}$$

Therefore,

$$c_B^T = c_N^T = \underline{-1} \overline{0}, B^{-1} = \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix}^{\text{inverse of the matrix}}$$

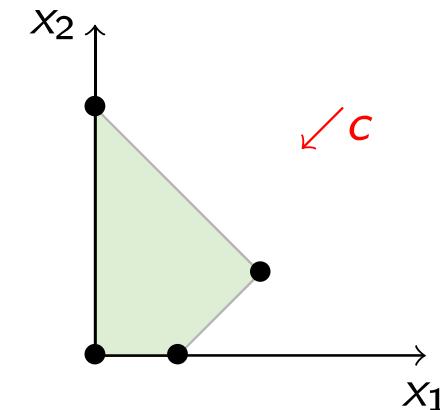
and the reduced costs for  $x_2, s_1$  are

The reduced cost for  $x_2$  is 2  
The reduced cost for  $s_1$  is 0

$$\bar{c}_N^T = c_N^T - c_B^T B^{-1} N = [-2 \ 0]$$

Since the reduced cost for  $x_2$  is negative

Since  $\bar{c}_2 = -2 < 0$ , increasing  $x_2$  to 1 (keeping the other non-basic variables to 0) we improve the solution by -2.



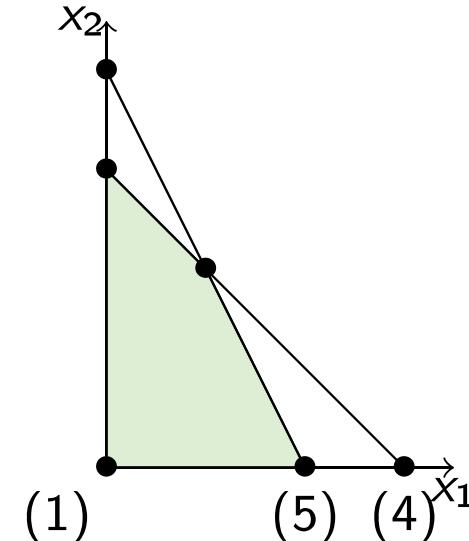
Increasing  $x_2$  from 0 to 1  
keeping  $s_1 = 0$ , we move  
along the edge to  $(x_1, x_2) = (2, 1)$ , and  $z = -3$ .

Remember: In the context of the Simplex method, when we initially set up a basic feasible solution, we assign the non-basic variables to zero. In this example, the non-basic variables are  $x_2, s_2$  and  $s_1$ , so they are set to 0 in the current basic feasible solution.

## 3.4 Simplex method

### 3.4.2 Move to an adjacent vertex (basic feasible solution)

$$\begin{aligned}x_1 + x_2 + s_1 &= 6 \\2x_1 + x_2 + s_2 &= 8 \\x_1, x_2, s_1, s_2 &\geq 0\end{aligned}$$



Move from vertex (1) to vertex (5):

- In (1)  $x_1 = 0, x_2 = 0 \Rightarrow s_1 = 6, s_2 = 8$  with  $x_B = (s_1, s_2)$ ,  $x_N = (x_1, x_2)$
- In (5)  $x_2 = 0, s_2 = 0 \Rightarrow x_1 = 4, s_1 = 2$  with  $x_B = (x_1, s_1)$ ,  $x_N = (x_2, s_2)$

Thus,  $x_1$  enters the basis  $B$  and  $s_2$  exits the basis  $B$ .

**Note:** When moving from the current vertex to an adjacent vertex, we substitute one column of  $B$  (that of  $s_2$ ) with one column of  $N$  (that of  $x_1$ ).

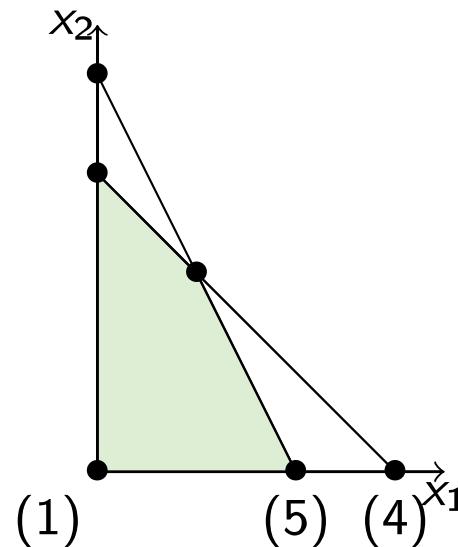
### 3.4 Simplex method

By expressing the basic variables in terms of the non basic variables, we obtain

$$s_1 = 6 - x_1 - x_2$$

$$s_2 = 8 - 2x_1 - x_2$$

Now we increase  $x_1$ , while keeping  $x_2 = 0$ .



Since  $s_1 = 6 - x_1 \geq 0$  implies  $x_1 \leq 6$  and  $s_2 = 8 - 2x_1 \geq 0$  implies  $x_1 \leq 8/2 = 4$ , the upper limit on the increase of  $x_1$  is:  $x_1 \leq \min\{6, 4\} = 4$ .

We move from vertex (1) to vertex (5) by letting  $x_1$  enter the basis and  $s_2$  exit the basis ( $s_1 = 2$  and  $s_2 = 0$ ).

**Note:** When  $x_1 = 6$ , we obtain the infeasible basic solution (4).

## 3.4 Simplex method

### General case

Given a basis  $B$ , the system

$$Ax = b \Leftrightarrow \sum_{j=1}^m a_{ij}x_i = b_j, \text{ for } i = 1, \dots, m$$

can be expressed in **canonical form**

Identità  $x_B + \underbrace{B^{-1}N}_{\text{Identità}} x_N = \underbrace{B^{-1}b}_{\text{Identità}} \Leftrightarrow x_B + \bar{N}x_N = \bar{b}$

which emphasizes the basic feasible solution  $(x_B, x_N) = (B^{-1}b, 0)$ . This amounts to pre-multiply the system by  $B^{-1}$ :

$$\underbrace{B^{-1}B}_{=I} x_B + \underbrace{\cancel{B^{-1}N}}_{=0} x_N = \underbrace{B^{-1}b}_{=\bar{b}}$$

$(m \times m) * (m * (n-m)) = (m * (n-m))$

### 3.4 Simplex method

In the canonical form  $\bar{N}$

$$x_{B_i} + \sum_{j=1}^{n-m} \bar{a}_{ij} x_{N_j} = \bar{b}_i, \text{ for } i = 1, \dots, m \quad (\bar{I}x_B + \bar{N}x_N = \bar{b})$$

**basic variables** are expressed in terms of **non basic variables**:  $x_B = \bar{b} - \bar{N}x_N$ .

If we increase the value of a non basic  $x_s$  (from value 0) while keeping all other non basic variables to 0, the system becomes

the others non basic variables are all 0s

$$x_{B_i} + \bar{a}_{is} x_s = \bar{b}_i \Leftrightarrow x_{B_i} = \bar{b}_i - \bar{a}_{is} x_s, \text{ for } i = 1, \dots, m$$

To guarantee  $x_{B_i} \geq 0$  for each  $i$ , we need to satisfy

$$\bar{b}_i - \bar{a}_{is} x_s \geq 0 \Rightarrow x_s \leq \bar{b}_i / \bar{a}_{is}, \text{ for } \bar{a}_{is} \geq 0.$$

The value of  $x_s$  can be increased up to

$$\theta^* = \min_{i=1, \dots, m} \left\{ \bar{b}_i / \bar{a}_{is} : \bar{a}_{is} \geq 0 \right\}$$

(If  $\bar{a}_{is} \leq 0$  for any  $i$ , there is no limit to the increase of  $x_s$ .) The value of  $x_r$  where

$$r = \operatorname{argmin}_{i=1, \dots, m} \left\{ \bar{b}_i / \bar{a}_{is} : \bar{a}_{is} \geq 0 \right\}$$

decreases to 0 and exits the basis.

## 3.4 Simplex method

### 3.4.3 Change of basis (for minimization LP)

Let  $B$  be a feasible basis and  $x_s$  (in  $x_N$ ) a non basic with reduced cost  $\bar{c}_s < 0$ .

- Increase  $x_s$  as much as possible ( $x_s$  “enters the basis”) while keeping the other non basic variables equal to 0.
- The basic variable  $x_r$  (in  $x_B$ ) such that  $x_r \geq 0$  imposes the tightest upper bound  $\theta^*$  on the increase of  $x_s$  ( $x_r$  leaves the basis).
- If  $\theta^* > 0$ , the new basic feasible solution has a better objective function value.
- The new basis differs w.r.t. the previous one by a single column (adjacent vertices).

To go from the canonical form of the current basic feasible solution

$$B^{-1}Bx_B + B^{-1}Nx_N = B^{-1}b$$

to that of an adjacent basic feasible solution, it is not necessary to compute  $B^{-1}$  from scratch.

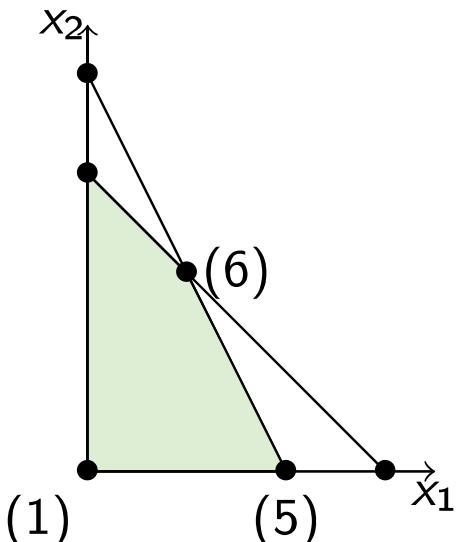
$B^{-1}$  of the new basis  $B$  can be obtained by applying to the inverse of the previous basis (which differs w.r.t a single column) a unique “**pivoting**” operation.

## 3.4 Simplex method

### Example

$$\begin{aligned} \min \quad & -x_1 - x_2 \\ \text{s. t.} \quad & x_1 + x_2 + s_1 = 6 \\ & 2x_1 + x_2 + s_2 = 8 \\ & x_1, x_2, s_1, s_2 \geq 0 \end{aligned}$$

Move from vertex (1) to vertex (5).



System in canonical form w.r.t. the basis with  $s_1$  and  $s_2$  basic (vertex (1)):

$$\begin{array}{lcl} s_1 & = & 6 - x_1 - x_2 \\ s_2 & = & 8 - 2x_1 - x_2 \end{array} \xrightarrow{\text{pivot}} \begin{array}{lcl} x_1 & + x_2 & + s_1 & = 6 \\ 2x_1 & + x_2 & + s_2 & = 8 \end{array}$$

pivot

columns of basic variables

$x_1$  enters the basis and  $s_2$  exits

In the canonical form I want to express the basic variables in terms of non basic variables

System in canonical form w.r.t. the basis with  $x_1$  and  $s_1$  basic (vertex (5)):

I need the basic variable  $s_1$  to be expressed in terms of only non basic variables --> need to "eliminate" dependency from the other basic variable  $x_1$  by performing a substitution

$$\begin{array}{lcl} (4 - \frac{1}{2}x_2 - \frac{1}{2}s_2) + x_2 + s_1 = 6 \\ x_1 = 4 - \frac{1}{2}x_2 - \frac{3}{2}s_2 \end{array} \Rightarrow \begin{array}{lcl} \frac{1}{2}x_2 & + s_1 & - \frac{1}{2}s_2 = 2 \\ x_1 & + \frac{1}{2}x_2 & + \frac{1}{2}s_2 = 4 \end{array}$$

## 3.4 Simplex method

### “Pivoting” operation

Same operations used in the Gaussian elimination method to solve systems of linear equations.

Given  $Ax = b$ :

- ① Select a coefficient  $\bar{a}_{rs} \neq 0$  (the “pivot”)
- ② Divide the  $r$ -th row by  $\bar{a}_{rs}$ .
- ③ For each row  $i$  with  $i \neq r$  and  $\bar{a}_{is} \neq 0$ , subtract the resulting  $r$ -th row multiplied by  $\bar{a}_{is}$ .

$$\begin{array}{c|ccccc|c} s & A & & b \\ \hline 1 & 1 & 1 & 0 & 6 \\ r \rightarrow & 2 & 1 & 0 & 1 & 8 \end{array} \longrightarrow \begin{array}{c|ccccc|c} & 0 & 1/2 & 1 & -1/2 & 2 \\ & 1 & 1/2 & 0 & 1/2 & 4 \end{array}$$

pivot  
columns of basic variables

$\therefore \quad \left( \begin{array}{ccccc|c} 1 & 1 & 1 & 0 & | & 2 \\ 1 & 1/2 & 0 & 1/2 & | & 4 \end{array} \right) \quad (0 \quad 1/2 \quad 1 \quad -1/2) - \underbrace{(1 \quad 1/2 \quad 0 \quad 1/2)}_{2 \times 1} \cdot (1 \quad 1/2 \quad 0 \quad 1/2)$

Note: This does not affect the feasible solutions.

## 3.4 Simplex method

### Moving to an adjacent basic feasible solution (vertex)

Goals :

- improve the objective function value
  - preserve feasibility
- ① Which non basic variable **enters the basis**? (Ie, choice of pivot column  $s$ .)
- ▶ Any one with reduced cost  $\bar{c}_j < 0$ .
  - ▶ One that yields the maximum  $\Delta z$  w.r.t.  $z = c_B^T B^{-1} b$  (the actual decrement  $\Delta z$  also depends on  $\theta^*$ ).
  - ▶ **Bland's rule:**  $s = \min\{j : \bar{c}_j < 0\}$ . (Or  $\bar{c}_j > 0$  for maximization problems.)

- ② Which basic variable **leaves the basis**?

Bland's rule for picking the variable entering the basis: if they have the same reduced cost, I will pick the one with the smallest index

**Min ratio test:** index  $i$  with smallest  $\bar{b}_i / \bar{a}_{is} = \theta^*$  (**tightest upper bound on increase of  $x_s$** ) among those with  $\bar{a}_{is} > 0$  (**otherwise no limit!**).

- ▶ **Bland's rule:**  $r = \min\{i : \bar{b}_i / \bar{a}_{is} = \theta^*, \bar{a}_{is} > 0\}$
- ▶ randomly...

Bland's rule for variable exiting the basis: can happen that when increasing a non basic variable up to the upper bound theta\*, more than one basic variable goes to zero, in this case I will need to pick one who will enter the basis. That can be established by the bland's rule too. The variable with the smallest index will exit the basis.

**Unboundedness:** If  $\exists \bar{c}_j < 0$  with  $\bar{a}_{ij} \leq 0 \forall i$ , no element of the  $j$ -th column can play the role of a pivot.  $\Rightarrow$  The minimization problem is unbounded!

## 3.4 Simplex method

### 3.4.4 “Tableau” representation

System  $\begin{cases} z = c^T x \\ Ax = b \end{cases}$  with (implicit) nonnegativity constraints

Initial tableau: Column of the righten sight on the left

-z	$x_1$	$\dots$	$x_n$	$z$
0	$c'$			-1 → objective function
$b$	$A$			0 → $m$ rows 0

The first column contains the:

- right hand side of the objective function
- right hand side vector

First column is associated to:

- the objective function value moved in the right side (first row)
- the right hand side vector  $b$  (second row)

### 3.4 Simplex method

Consider a basis  $B$  and a partition  $A = [B \ N]$ , with  $0 = c^T x - z$ , we have:

	$x_1$	$\dots$	$x_m$	$x_{m+1}$	$\dots$	$x_n$	$z$
0	$c_B^I$			$c_N^I$			-1
$\bar{b}$		$\bar{B}^{-1}$	$\bar{B}$			$\bar{N}$	0
$b$							0
							$\vdots$
							0

by “pivoting” operations (or pre-multiplying by  $B^{-1}$ ) we put the **tableau** in **canonical form** with respect to  $B$ :

	$x_1$	$\dots$	$x_m$	$x_{m+1}$	$\dots$	$x_n$	
$-z$	$-z_0$	0	$\dots$	0		$\bar{c}_N^I$	
$x_{B[1]}$	$\bar{b}$					$\bar{N}$	
$\vdots$							
$x_{B[m]}$							

basic variables

$\bar{z} = c_B^T x_B + c_N^T x_N$        $x_B = \bar{b} - \bar{N} x_N$   
 $C^T x = z_0 + c_N^T x_N$

$z = c_B^T B^{-1} b + \bar{c}_N^T x_N = z_0 + \bar{c}_N^T x_N$   
 $\bar{b} = B^{-1} b$

## 3.4 Simplex method

### Example

*zero because initially  
 $x_1, x_2 \in X^r \Rightarrow x_1 = x_2 = 0 \Rightarrow z(x_1, x_2) = 0$*

$$\begin{array}{ll} \min & -x_1 - x_2 \\ \text{s. t.} & 6x_1 + 4x_2 + x_3 = 24 \\ & 3x_1 - 2x_2 + x_4 = 6 \\ & x_1, x_2, x_3, x_4 \geq 0 \end{array}$$

*First basis, trivial in this case*

Tableau w.r.t. the basis with columns 3, 4:

	$x_1$ reduce cost	$x_2$	$x_3$	$x_4$
$-z$	0	-1	-1	0
$x_3$	24	6	4	1
$x_4$	6	3	-2	0

*basic variables*

**pivot**

**basic variables**

$x_1$  enters in the basis and  $x_4$  exits the basis

Pivot w.r.t. 3 amounts to deriving an expression for  $x_1$  from the pivot row and substituting it in all other rows.

$$\theta^* = \min \left\{ \frac{24}{6}, \frac{6}{3} \right\} = \frac{6}{3}$$

*RATIO FOR  $x_3$*       *RATIO FOR  $x_4$*       *PIVOT*

Tableau w.r.t. the new basis with columns:

	$x_1$	$x_2$	$x_3$	$x_4$
$-z$	2	0	$-5/3$	0
$x_3$	12	0	8	1
$x_1$	2	1	$-2/3$	0

*This is  $x_B$   
since  $\bar{b} = B^{-1}b = x_B$*

*THIS ROW IS ALSO OBTAINED BY INVOLVING IT IN THE PIVOTING OPERATION, SINCE WE WANT IT TO BE INDEPENDENT FROM  $x_B$ , WE NEED TO SUBSTITUTE  $x_1$  ( $\bar{c}_x$ ) IN IT*

Corresponding basic feasible solution:

$$x^T = (2, 0, 12, 0), \text{ with } z = -2.$$

$x_2$  is the only non basic variable can “enter” the basis ( $\bar{c}_2 = -5/3 < 0$ )

$x_3$  is the only basic variable can “exit” the basis ( $\bar{a}_{rs} = 8 > 0$ )

$2/-2/3 < 0 \Rightarrow x_1$  doesn't impose an upper bound on the variable entering the basis

## 3.4 Simplex method

	$x_1$	$x_2$	$x_3$	$x_4$	
$-z$	2	0	$-5/3$	0	$1/3$
$x_3$	12	0	$8$	1	$-2$
$x_1$	2	1	$-2/3$	0	$1/3$

	$x_1$	$x_2$	$x_3$	$x_4$	
$-z$	$9/2$	0	0	$5/24$	$-1/12$
$x_2$	$3/2$	0	1	$1/8$	$-1/4$
$x_1$	3	1	0	$1/12$	$1/6$

	$x_1$	$x_2$	$x_3$	$x_4$	
$-z$	6	$1/2$	0	$1/4$	0
$x_2$	6	$3/2$	1	$1/4$	0
$x_4$	18	6	0	$1/2$	1

All reduced costs  $\geq 0$ , then, the optimal basic (feasible) solution is:

$$x^*{}^T = (0, 6, 0, 18) \text{ with } z^* = -6.$$

## 3.4 Simplex method

### Algorithm 1: Simplex algorithm (LP with minimization)

```
1 Let  $B[1], \dots, B[m]$  be the column indices of the initial feasible basis  $B$  --> we need a first  
basis given to us
2 Construct the initial tableau  $\bar{A} = \{\bar{a}[i,j] : 0 \leq i \leq m, 0 \leq j \leq n\}$  in canonical  
form w.r.t.  $B$ 
3 unbounded  $\leftarrow$  false --> if the problem is unbounded
4 optimal  $\leftarrow$  false --> if the problem is optimal
5 while optimal = false and unbounded = false do
6   if  $\bar{a}[0,j] \geq 0 \forall j = 1, \dots, m$  then optimal  $\leftarrow$  true /* for LP with min */  
check whether all coefficients are greater or equal to zero, if so the optimality test is satisfied
7
8   else
9     Select a non basic variable  $x_s$  with  $\bar{a}[0,s] < 0$  /* negative reduced cost */  
select reduced cost associated to  $x_s$ 
10    if ALL the ratios are negative, then the problem is unbounded.
11    if  $\bar{a}[i,s] \leq 0 \forall i = 1, \dots, m$  then unbounded  $\leftarrow$  true
12    else otherwise we can pick the minimum from the non negative ones
13       $r \leftarrow \operatorname{argmin} \left\{ \frac{\bar{a}[i,0]}{\bar{a}[i,s]} \text{ with } 1 \leq i \leq m \text{ and } \bar{a}[i,s] > 0 \right\}$ 
14      pivot(r, s) /* update tableau pick the pivot based on which is the ratio chosen */
15       $B[r] \leftarrow s$ 
```

## 3.4 Simplex method

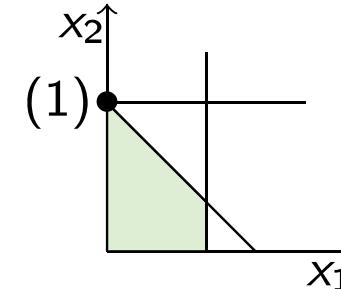
### 3.4.5 Degenerate basic feasible solutions and convergence

$$x_1 + s_1 = 2$$

$$+x_2 + s_2 = 3$$

$$x_1 + x_2 + s_3 = 3$$

$$x_1, x_2, s_1, s_2, s_3 \geq 0$$



Point (1) corresponds to the basic solution  
 $x = (0, 3, 2, 0, 0)$

#### Definition 18

A **basic feasible solution**  $x$  is **degenerate** if it contains at least one basic variable  $x_j = 0$ .

that means not only the non basic variables are 0, but also at least a basic variable

A solution  $x$  with more than  $n - m$  zeros corresponds to several distinct bases!

We would have many basic feasible solutions that correspond to the same vertex, since by making the basic variable that equals to zero exit and making a non basic variable entering, the vertex is the same.

If a basic variable  $x_i$  in  $X_B$  is zero, than its ratio will be zero so the upper bound on the entering variable is zero. That is, the exiting variable is zero and the entering variable is 0. Different bases (since different variables are involved) but same vertex.

Same vertex.

- More than  $n$  constraints (the  $m$  of  $Ax = b$  and more than  $n - m$  among the  $n$  of  $x \geq 0$ ) are satisfied with equality ("active").

In this case we have a mapping basis  $\rightarrow$  vertex which is not one to one, so the simplex method, which only is based on changing of the basis, doesn't know the vertex is always the same, and can get stuck in a cycle

## 3.4 Simplex method

### 3.4.5 Degenerate basic feasible solutions and convergence

In the presence of degenerate basic feasible solutions (BFSs), a basis change **may not decrease the objective function** value:

- if the current BFS is degenerate, one can have  $\theta^* = 0$  and hence the new BFS is identical (same vertex).

The simplex method changes the bases because it sees that the reduced cost of a variable is good for upgrading the objective function, but then the upper bound of the increasing of the variable is zero, then there is no real increase of objective function

Note that a degenerate BFS can arise from a non degenerate one: even if  $\theta^* > 0$ , several basic variables may go to 0 when  $x_s$  is increased to  $\theta^*$ .

- One can **cycle** through a **sequence of “degenerate” basis** associated to the same vertex.

When I am making a non basic variable entering the basis by increasing its value up to  $\theta^* > 0$ , it can be that that  $\theta^*$  is imposed at the same time by two basis variables (the ratio of the two variables is the same). Usually, it happens that the variable that imposes the upper bound becomes zero after the increase of the entering value, but in this case two basic variables go to zero. So we need to choose which one will exit the base, and the other one will stay in the base and impose the upper bound for the next entering variable to 0. That is, we reached a degenerate BFS since there is a basic variable which is =0.

## 3.4 Simplex method

Several “anticycling” rules have been proposed for the choice of the variables that enter and exit the bases (indices  $r$  and  $s$ ).

**Bland's rule:** Among all candidate variables to enter/exit the basis ( $x_s/x_r$ ) always select the one with smallest index.



Robert Bland

### Proposition 2

The Simplex algorithm with Bland's rule terminates after  $\leq \binom{n}{m}$  iterations.

**Note:** The number of pivots is finite.

- In some “pathological” cases (see e.g. Klee & Minty 72), the number of iterations grows exponentially w.r.t.  $n$  or  $m$ .  $\rightarrow$  so in theory it is not polynomial, in the practical application it is very efficient
- However, the Simplex algorithm is overall **very efficient** and **extensive experimental campaigns** showed a number of iterations that grows linearly w.r.t.  $m$  ( $m \leq \dots \leq 3m$ ) and very slowly ( $\approx$  logarithmically) w.r.t.  $n$ .  
 $m$  is usually very smaller than  $n$

## 3.4 Simplex method

### 3.4.6 Two-phase simplex method

The overall simplex method is called a two-phase method:  
- First phase to get the first basis if it is not straightforward  
- Second phase we apply the iterations we saw in the previous examples

Phase 1: Determine an initial basic feasible solution.

Example:

$$\begin{array}{lll} \min & x_1 & + x_3 \\ \text{s. t.} & x_1 + 2x_2 & \leq 5 \\ & x_2 + 2x_3 = 6 \\ & x_1, x_2, x_3 \geq 0 \end{array}$$

There is no sub-matrix  $I_{2 \times 2}$  of  $A$ !

$$\begin{array}{lll} \min & x_1 & + x_3 \\ \text{s. t.} & x_1 + 2x_2 & + x_4 = 5 \\ & x_2 + 2x_3 & = 6 \\ & x_1, x_2, x_3, x_4 \geq 0 \end{array}$$

If not so I can multiply both sides by -1

↑

Given the problem  $(P)$ : s. t.  $Ax = b$ , with  $b \geq 0$ , an auxiliary LP with artificial variables  $y_i$ ,  $i = 1, \dots, m$ , is written:

$$(P_A) \quad \begin{array}{ll} \min & v = \sum_{i=1}^m y_i \\ \text{s. t.} & Ax + ly = b \\ & x \geq 0, y \geq 0 \end{array}$$

--> we forget the original obj function, we only focus on constraint (auxiliary obj funct)

There is an obvious initial basic feasible solution  $y = b \geq 0$  and  $x = 0$ .

## 3.4 Simplex method

A partire dalla soluzione banale  $y=b$  e  $x=0$ , possiamo applicare il simplex method, facendo entrare  $x$  nella base e facendo uscire  $y$  dalla base. Se riusciamo ad arrivare alla soluzione ottimale, in cui  $y^*=[0]$  (tutte le  $y$  fuori dalla base) e in cui tutte le  $x$  sono dentro la base  $x^*$ , vuol dire che i constraint  $Ax^*=b$  sono rispettati e quindi quella base formata solo da  $x$  è una basic feasible solution per il problema originale, che possiamo usare come base di partenza per applicare il simplex method.

- ① If  $v^* > 0$ , then (P) is infeasible.   -->  $v^*$  è l'objective function con  $y^*=[0]$ ,  $x^*$  base
- ② If  $v^* = 0$ , clearly  $y^* = 0$  and  $x^*$  is a basic feasible solution of (P).
  - ▶ If  $y_i$  is non basic  $\forall i$ , with  $1 \leq i \leq m$ , delete the corresponding columns and obtain a tableau in canonical form w.r.t. a basis; the row of  $z$  must be determined by substitution.
  - ▶ If there is a basic  $y_i$  (the basic feasible solution is degenerate), we perform a "pivot" operation w.r.t. a coefficient  $\neq 0$  of the row of  $y_i$  so as to "exchange"  $y_i$  with any non basic variable  $x_j$ .

cf. example

SIAMO NEL CASO IN CUI  $y^*=0$  MA UNA SELE  $y_i$  È NELLA BASE,  
QUINDI ABBIAMO UNA BASE DECERERE. POSSIAMO CONSIDERARE  
AD UNA BASE PER IL PROB ORIGINALE

### 3.4 Simplex method

$$(P) \quad \begin{aligned} \min \quad & x_1 + x_2 + 10x_3 \\ \text{s. t.} \quad & x_2 + 4x_3 = 2 \\ & -2x_1 + x_2 - 6x_3 = 2 \\ & x_1, x_2, x_3 \geq 0 \end{aligned}$$

$$(P_A) \quad \begin{aligned} \min \quad & v = y_1 + y_2 \\ \text{s. t.} \quad & x_2 + 4x_3 + y_1 = 2 \\ & -2x_1 + x_2 - 6x_3 + y_2 = 2 \\ & x_1, x_2, x_3, y_1, y_2 \geq 0 \end{aligned}$$

We write  $v = y_1 + y_2$  in canonical form by substituting the expression of  $y_1$  and  $y_2$  in terms of  $x_1$ ,  $x_2$  and  $x_3$ .

	$x_1$	$x_2$	$x_3$	$y_1$	$y_2$
$-v$	-4	2	-2	2	0
$y_1$	2	0	1	4	0
$y_2$	2	-2	1	-6	0

↗ BASIS

	$x_1$	$x_2$	$x_3$	$y_1$	$y_2$
$-v$	0	2	0	10	2
$x_2$	2	0	1	4	1
$y_2$	0	-2	0	-10	-1

→ SINCE WE ARE  
 IN CASE 2  
 OF  
 PREVIOUS  
 SLIDE  
 ( $y_2$  IS  
 BASIC)

By swaping the roles of  $y_2$  and  $x_1$  or  $x_3$ , we identify an optimal solution of the Phase I problem:  $x^* = (0, 2, 0)$ ,  $y^* = (0, 0)$ , with optimal value  $v^* = 0$ .

WE NEED  
 TO PIVOT  
 AS DESCRIBED  
 IN PREVIOUS  
 SLIDE

### 3.4 Simplex method



By selecting as “pivot” the coefficient  $-2$  of the row of  $y_2$ , we obtain:

	$x_1$	$x_2$	$x_3$	$y_1$	$y_2$
$-v$	0	0	0	1	1
$x_2$	2	0	1	4	1
$x_1$	0	1	0	5	$1/2$

$\nwarrow \nearrow$  BASIS

With respect to the original LP,

$$z = x_1 + x_2 + 10x_3$$

We should express the objective function with respect to only the non basic variable, in our case  $x_3$

is not in the canonical form, because  $x_3$  is a non basic variable, but substituting:

$$\begin{cases} x_2 = 2 - 4x_3 \\ x_1 = -5x_3 \end{cases} \Rightarrow z = 2 + x_3$$

we obtain the tableau corresponding to the initial basic feasible solution of  $(P)$ :

	$x_1$	$x_2$	$x_3$
$-z$	-2	0	1
$x_2$	2	0	1
$x_1$	0	1	5

Since the basic feasible solution found is (already) optimal, there is no need for Phase II!

## 3.4 Simplex method

### 3.4.7 Polynomial-time algorithms for LP

- Ellipsoid method (L. Khachiyan 1979)  
Theoretically important.
- Interior point methods (N. Karmarkar 1984, . . .)  
Very efficient variants (e.g. barrier methods) for some types of instances (e.g. sparse and large-scale).



Leonid Khachiyan (1952-2005)



Narendra Karmarkar (1957-)