MVE165/MMG631 Linear and Integer Optimization with Applications Lecture 3 Extreme points of convex polyhedra; reformulations; basic feasible solutions; the simplex method

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## Course evaluation

- The first meeting was held on Friday, March 24 at 9.30.
- The second meeting will be held during week 17 (April, 24–28)
- Notes will be published in the course's PingPong event
- Any voluntary representative from GU is also welcome! Anyone?

Contact any student representative to present your opinion:

- Arvid Bjurklint (TKTEM)
- Frida Eriksson (TKTEM)
- Oskar Holmstedt (TKTEM)
- Stefanus Ivarsson Bergenhem (MPSYS)

A linear optimization model – a linear program	
minimize $z = \sum_{j=1}^{n} c_j x_j$	In
subject to $\sum_{j=1}^{n} a_{ij} x_j \leq b_i, i = 1, \dots, m$	
$x_j \ge 0,  j = 1, \dots, n$	С,

In	vector	notation
	min	$z = \mathbf{c}^{\mathrm{T}} \mathbf{x}$
	s.t.	$\mathbf{A}\mathbf{x} \leq \mathbf{b}$
		$\mathbf{x} \geq 0^n$

$$\mathbf{c}, \mathbf{x} \in \mathbb{R}^n$$
,  $\mathbf{b} \in \mathbb{R}^m$ ,  
 $\mathbf{A} \in \mathbb{R}^{m imes n}$ 

 $c_j$ ,  $a_{ij}$ ,  $b_i$ : constant parameters

The feasible region is a polyhedron, 
$$X \subset \mathbb{R}^n_+$$
  
$$X := \left\{ \mathbf{x} \ge \mathbf{0}^n \ \left| \ \sum_{j=1}^n a_{ij} x_j \le b_i, i = 1, \dots, m \right. \right\} = \left\{ \mathbf{x} \ge \mathbf{0}^n \ \left| \ \mathbf{A} \mathbf{x} \le \mathbf{b} \right. \right\}$$

# Linear programs, convex polyhedra and extreme points (Ch. 4.1)

#### Definition (Convex combination)

A convex combination of the points  $\mathbf{x}^{p}$ , p = 1, ..., P, is a point  $\mathbf{x}$  that can be expressed as

$$\mathbf{x} = \sum_{p=1}^{P} \lambda_p \mathbf{x}^p; \qquad \sum_{p=1}^{P} \lambda_p = 1; \qquad \lambda_p \ge 0, \quad p = 1, \dots, P$$

[DRAW ON THE BOARD]

# Linear programs, convex polyhedra and extreme points (Ch. 4.1)

#### Intersection of linear constraints form a convex set

The feasible region of a linear program is a *convex set*, since for any two feasible points  $\mathbf{x}^1$  and  $\mathbf{x}^2$  and any  $\lambda \in [0, 1]$  it holds that

$$\sum_{j=1}^{n} a_{ij} \left( \lambda x_j^1 + (1-\lambda) x_j^2 \right) = \lambda \sum_{j=1}^{n} a_{ij} x_j^1 + (1-\lambda) \sum_{j=1}^{n} a_{ij} x_j^2$$

$$\leq \lambda b_i + (1-\lambda) b_i$$

$$= b_i, \qquad i = 1, \dots, m$$

and

$$\lambda x_j^1 + (1-\lambda) x_j^2 \geq 0, \qquad j=1,\ldots,n$$

[DRAW ON THE BOARD]

# Linear programs, convex polyhedra and extreme points (Ch. 4.1)

### Definition (Extreme point (Def. 4.2))

The point  $\mathbf{x}^k$  is an *extreme point* of the polyhedron X if  $\mathbf{x}^k \in X$  and it is *not* possible to express  $\mathbf{x}^k$  as a *strict convex combination* of two distinct points in X.

**I.e:** Given 
$$\mathbf{x}^1 \in X$$
,  $\mathbf{x}^2 \in X$ , and  $0 < \lambda < 1$ , it holds that  
 $\mathbf{x}^k = \lambda \mathbf{x}^1 + (1 - \lambda)\mathbf{x}^2$  only if  $\mathbf{x}^k = \mathbf{x}^1 = \mathbf{x}^2$  hold.  
[DRAW ON THE BOARD]

### Theorem (Optimal solution in an extreme point (Th. 4.2))

Assume that the feasible region  $X = \{ \mathbf{x} \ge \mathbf{0}^n \mid \mathbf{A}\mathbf{x} \le \mathbf{b} \}$  is non-empty and bounded. Then, the minimum value of the objective  $\mathbf{c}^T \mathbf{x}$  is attained at (at least) one extreme point  $\mathbf{x}^k$  of X.

## A general linear program – notation

#### Definition (Notation of linear programs)

minimize or maximize  $c_1x_1 + \ldots + c_nx_n$ 

subject to 
$$a_{i1}x_1 + \ldots + a_{in}x_n \begin{cases} \leq \\ = \\ \geq \end{cases} b_i, \quad i = 1, \ldots, m$$
  
$$x_j \begin{cases} \leq 0 \\ \text{unrestricted in sign} \\ \geq 0 \end{cases}, \quad j = 1, \ldots, n$$

#### The blue notation corresponds to the standard form

# The standard form and the simplex method for linear programs (Ch. 4.2)

- Every linear program can be reformulated such that:
  - all constraints are expressed as *equalities* with *non-negative right hand sides*
  - all variables involved are restricted to be *non-negative*
- Referred to as the *standard form*
- These requirements streamline the calculations of the *simplex method*
- Software solvers (e.g., Cplex, GLPK, Clp, Gurobi, SCIP) handle also inequality constraints and unrestricted variables – the reformulations are made automatically

### • Slack variables:

$$\left[\begin{array}{ccc}\sum_{j=1}^n a_{ij}x_j &\leq b_i, \ \forall i\\ x_j &\geq 0, \ \forall j\end{array}\right] \Longleftrightarrow \left[\begin{array}{ccc}\sum_{j=1}^n a_{ij}x_j &+s_i &=b_i, \ \forall i\\ x_j &\geq 0, \ \forall j\\ s_i &\geq 0, \ \forall i\end{array}\right]$$

• The lego example:

$$\begin{bmatrix} 2x_1 & +x_2 \le & 6\\ 2x_1 & +2x_2 \le & 8\\ & x_1, x_2 \ge & 0 \end{bmatrix} \iff \begin{bmatrix} 2x_1 & +x_2 & +s_1 & = & 6\\ 2x_1 & +2x_2 & +s_2 & = & 8\\ & & x_1, x_2, s_1, s_2 \ge & 0 \end{bmatrix}$$

 s<sub>1</sub> and s<sub>2</sub> are called *slack variables*—they "fill out" the (positive) distances between the left and right hand sides

• Surplus variables:

$$\left[\begin{array}{ccc}\sum_{j=1}^n a_{ij}x_j &\geq b_i, \ \forall i\\ x_j &\geq 0, \ \forall j\end{array}\right] \Longleftrightarrow \left[\begin{array}{ccc}\sum_{j=1}^n a_{ij}x_j &-s_i &= b_i, \ \forall i\\ x_j &\geq 0, \ \forall j\\ s_i &\geq 0, \ \forall i\end{array}\right]$$

• Surplus variable *s*<sub>3</sub> (another instance):

$$\left[\begin{array}{cccc} x_1 & + & x_2 & \ge & 800 \\ & x_1, x_2 & \ge & 0 \end{array}\right] \iff \left[\begin{array}{cccc} x_1 & + & x_2 - & \mathbf{s_3} & = & 800 \\ & & x_1, x_2, \mathbf{s_3} & \ge & 0 \end{array}\right]$$

• Suppose that b < 0:

$$\left[\begin{array}{c}\sum_{j=1}^{n}a_{j}x_{j}\leq b\\x_{j}\geq 0,\forall j\end{array}\right] \Longleftrightarrow \left[\begin{array}{c}\sum_{j=1}^{n}(-a_{j})x_{j}\geq -b\\x_{j}\geq 0,\forall j\end{array}\right] \Longleftrightarrow \left[\begin{array}{c}-\sum_{j=1}^{n}a_{j}x_{j}&-s&=-b\\x_{j}&\geq 0,\forall j\\s&\geq 0\end{array}\right]$$

• Non-negative right hand side:

$$\begin{bmatrix} x_1 - x_2 \leq -23 \\ x_1, x_2 \geq 0 \end{bmatrix} \Longleftrightarrow \begin{bmatrix} -x_1 + x_2 \geq 23 \\ x_1, x_2 \geq 0 \end{bmatrix} \Longleftrightarrow \begin{bmatrix} -x_1 + x_2 - s_4 = 23 \\ x_1, x_2, s_4 \geq 0 \end{bmatrix}$$

• Suppose that some of the variables are unconstrained (here: k < n). Replace  $x_j$  with  $x_j^1 - x_j^2$  for the corresponding indices:

$$\begin{bmatrix} \sum_{j=1}^{n} a_j x_j \le b \\ x_j \ge 0, j = 1, \dots, k \end{bmatrix} \iff \begin{bmatrix} \sum_{j=1}^{k} a_j x_j + \sum_{j=k+1}^{n} a_j (x_j^1 - x_j^2) + s &= b \\ x_j \ge 0, \ j = 1, \dots, k, \\ x_j^1 \ge 0, x_j^2 \ge 0, \ j = k+1, \dots, n \\ s \ge 0 \end{bmatrix}$$

• Sign-restricted (non-negative) variables:

$$\begin{bmatrix} x_1 + x_2 \le 10 \\ x_1 \ge 0 \end{bmatrix} \Longleftrightarrow \begin{bmatrix} x_1 + x_2^1 - x_2^2 \le 10 \\ x_1, x_2^1, x_2^2 \ge 0 \end{bmatrix} \Longleftrightarrow \begin{bmatrix} x_1 + x_2^1 - x_2^2 + s_5 = 10 \\ x_1, x_2^1, x_2^2, s_5 \ge 0 \end{bmatrix}$$

## Basic feasible solutions (Ch. 4.3)

- Consider *m* equations with *n* variables, where  $m \le n$
- Set n m variables to zero and solve (if possible) the remaining  $(m \times m)$  system of equations
- If the solution is *unique*, it is called a *basic* solution

### Definition (Def. 4.3)

A *basic* solution to the  $m \times n$  system of equations  $\mathbf{A}\mathbf{x} = \mathbf{b}$  is obtained if n - m of the variables are set to 0 and the remaining variables get their unique values from the solution to the remaining  $m \times m$  system of equations.

The variables that are set to 0 are called *nonbasic variables* and the remaining *m* variables are called *basic variables*.

## Basic feasible solutions (Ch. 4.3)

- A basic solution **x** corresponds to the *intersection* of *m* hyperplanes in  $\mathbb{R}^m$ 
  - It is feasible if  $\mathbf{x} \geq \mathbf{0}$
  - It is infeasible if  $x \not\geq 0$
- Each extreme point of the feasible set is an intersection of m hyperplanes such that all variable values are ≥ 0
- Basic feasible solution  $\iff$  extreme point of the feasible set

$$\begin{array}{ll} a_{11}x_1 + \ldots + a_{1n}x_n = b_1 & x_1 \ge 0 \\ a_{21}x_1 + \ldots + a_{2n}x_n = b_2 & x_2 \ge 0 \\ & \ddots & & \ddots \\ a_{m1}x_1 + \ldots + a_{mn}x_n = b_m & x_n \ge 0 \end{array}$$

### Basic feasible solutions

Assume that 
$$m < n$$
 and that  $b_i \ge 0$ ,  $i = 1, ..., m$ , and let  
 $\mathbf{c} = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$ ,  $\mathbf{A} = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix}$ ,  $\mathbf{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$ ,  $\mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$ 

Consider the linear program to

 $\begin{array}{ll} \underset{\mathbf{x}}{\textit{minimize}} & z = \mathbf{c}^{\mathrm{T}} \mathbf{x} \\ \textit{subject to} & \mathbf{A} \mathbf{x} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array}$ 

Partition x into m basic variables x<sub>B</sub> and n - m non-basic variables x<sub>N</sub>, such that x = (x<sub>B</sub>, x<sub>N</sub>).

• Analogously, let  $\mathbf{c} = (\mathbf{c}_B, \mathbf{c}_N)$  and  $\mathbf{A} = (\mathbf{A}_B, \mathbf{A}_N) \equiv (\mathbf{B}, \mathbf{N})$ 

• The matrix  $\mathbf{B} \in \mathbb{R}^{m \times m}$  with inverse  $\mathbf{B}^{-1}$  (if it exists)

#### Rewrite the linear program as

$$\begin{array}{ll} \mbox{minimize} & z = \mathbf{c}_B^{\mathrm{T}} \mathbf{x}_B + \mathbf{c}_N^{\mathrm{T}} \mathbf{x}_N & (1a) \\ \mbox{subject to} & \mathbf{B} \mathbf{x}_B + \mathbf{N} \mathbf{x}_N &= \mathbf{b} & (1b) \\ & \mathbf{x}_B \geq \mathbf{0}^m, \ \mathbf{x}_N &\geq \mathbf{0}^{n-m} & (1c) \end{array}$$

• Multiply the system of equations (1b) by  $\mathbf{B}^{-1}$  from the left:

$$\mathbf{B}^{-1}\mathbf{B}\mathbf{x}_B + \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N = \mathbf{x}_B + \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N = \mathbf{B}^{-1}\mathbf{b}$$
$$\implies \mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b} - \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N = \mathbf{B}^{-1}(\mathbf{b} - \mathbf{N}\mathbf{x}_N)$$
(2)

• Replace  $\mathbf{x}_B$  in (1a) by the expression in (2):

$$\mathbf{c}_B^{\mathrm{T}}\mathbf{x}_B + \mathbf{c}_N^{\mathrm{T}}\mathbf{x}_N = \mathbf{c}_B^{\mathrm{T}}\mathbf{B}^{-1}(\mathbf{b} - \mathbf{N}\mathbf{x}_N) + \mathbf{c}_N^{\mathrm{T}}\mathbf{x}_N = \mathbf{c}_B^{\mathrm{T}}\mathbf{B}^{-1}\mathbf{b} + (\mathbf{c}_N^{\mathrm{T}} - \mathbf{c}_B^{\mathrm{T}}\mathbf{B}^{-1}\mathbf{N})\mathbf{x}_N$$

$$\begin{array}{ll} \Rightarrow & \textit{minimize} \quad z = \mathbf{c}_B^{\mathrm{T}} \mathbf{B}^{-1} \mathbf{b} + (\mathbf{c}_N^{\mathrm{T}} - \mathbf{c}_B^{\mathrm{T}} \mathbf{B}^{-1} \mathbf{N}) \mathbf{x}_N \\ & \textit{subject to} \qquad \qquad \mathbf{B}^{-1} \mathbf{b} - \mathbf{B}^{-1} \mathbf{N} \mathbf{x}_N \quad \geq \mathbf{0}^m, \ \mathbf{x}_N \geq \mathbf{0}^{n-m} \end{array}$$

#### The rewritten program

$$\begin{array}{ll} \mbox{minimize} & z = \mathbf{c}_B^{\mathrm{T}} \mathbf{B}^{-1} \mathbf{b} + (\mathbf{c}_N^{\mathrm{T}} - \mathbf{c}_B^{\mathrm{T}} \mathbf{B}^{-1} \mathbf{N}) \mathbf{x}_N & (3a) \\ \mbox{subject to} & \mathbf{B}^{-1} \mathbf{b} - \mathbf{B}^{-1} \mathbf{N} \mathbf{x}_N & \geq \mathbf{0}^m & (3b) \\ & \mathbf{x}_N & \geq \mathbf{0}^{n-m} & (3c) \end{array}$$

At the basic solution defined by  $B \subset \{1, \ldots, n\}$ :

- Each non-basic variable takes the value 0, i.e.,  $\mathbf{x}_N = \mathbf{0}$
- The basic variables take the values  $\mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b} - \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N = \mathbf{B}^{-1}\mathbf{b}$
- The value of the objective function is  $z = \mathbf{c}_B^{\mathrm{T}} \mathbf{B}^{-1} \mathbf{b}$
- The basic solution is feasible if  $\mathbf{B}^{-1}\mathbf{b} \ge \mathbf{0}^m$

# The simplex method: Optimality and feasibility and change of basis (Ch. 4.4)

#### Optimality condition (for minimization)

The basis *B* is optimal if  $\mathbf{c}_N^{\mathrm{T}} - \mathbf{c}_B^{\mathrm{T}} \mathbf{B}^{-1} \mathbf{N} \ge \mathbf{0}^{n-m}$ (marginal values = reduced costs  $\ge 0$ ) If not, choose as entering variable  $j \in N$  the one with the

lowest (negative) value of the reduced cost  $c_i - \mathbf{c}_B^{\mathrm{T}} \mathbf{B}^{-1} \mathbf{A}_i$ 

#### Feasibility condition

For all  $i \in B$  it holds that  $x_i = (\mathbf{B}^{-1}\mathbf{b})_i - (\mathbf{B}^{-1}\mathbf{A}_j)_i x_j$ 

Choose the leaving variable  $i^* \in B$  according to

$$i^* = \arg\min_{i\in B} \left\{ \left. rac{(\mathbf{B}^{-1}\mathbf{b})_i}{(\mathbf{B}^{-1}\mathbf{A}_j)_i} 
ight| (\mathbf{B}^{-1}\mathbf{A}_j)_i > 0 
ight\}$$

# Simplex search for linear optimization (Ch. 4.6)

Overview of the simplex algorithm for linear optimization (minimization)

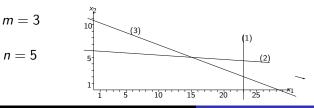
- Initialization: Choose any *feasible basis*, construct the corresponding *basic solution* x<sup>0</sup>, let t = 0
- Step direction: Select a variable to *enter the basis* using the *optimality condition* (negative marginal value).
   Stop if no entering variable exists
- Step length: Use the *feasibility condition* (smallest non-negative quotient) to select a variable to *leave the basis*
- New iterate: Compute the new basic solution x<sup>t+1</sup> by performing matrix operations
- Solution Let t := t + 1 and repeat from step 2

## Basic feasible solutions, example

• Constraints:

$$egin{array}{rcrcr} x_1 & \leq & 23 & (1) \ 0.067 x_1 & + & x_2 & \leq & 6 & (2) \ 3 x_1 & + & 8 x_2 & \leq & 85 & (3) \ & & x_1, x_2 & \geq & 0 \end{array}$$

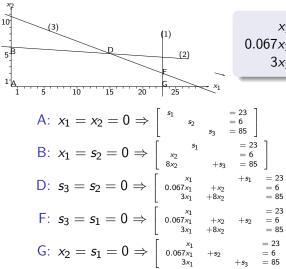
• Add slack variables:



# Basic and non-basic variables and solutions

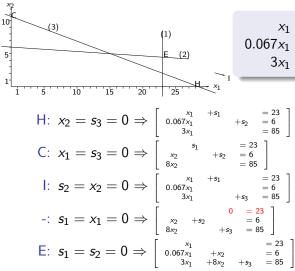
basic	_ ba	sic solu	tion	non-basic	point	feasible?	
variables				variables (0,0)			
$s_1, s_2, s_3$	23	6	85	$x_1, x_2$	А	yes	
$s_1, s_2, x_1$	$-5\frac{1}{3}$	$4\frac{1}{9}$	$28\frac{1}{3}$	<i>s</i> <sub>3</sub> , <i>x</i> <sub>2</sub>	Н	no	
$s_1, s_2, x_2$	23	$-4\frac{5}{8}$	$10\frac{5}{8}$	$x_1, s_3$	С	no	
<i>s</i> <sub>1</sub> , <i>x</i> <sub>1</sub> , <i>s</i> <sub>3</sub>	-67	90	-185	<i>s</i> <sub>2</sub> , <i>x</i> <sub>2</sub>	1	no	
$s_1, x_2, s_3$	23	6	37	<i>s</i> <sub>2</sub> , <i>x</i> <sub>1</sub>	В	yes	
$x_1, s_2, s_3$	23	$4\frac{7}{15}$	16	$s_1, x_2$	G	yes	
$x_2, s_2, s_3$	-	-	-	$s_1, x_1$	-	-	
<i>x</i> <sub>1</sub> , <i>x</i> <sub>2</sub> , <i>s</i> <sub>1</sub>	15	5	8	$s_2, s_3$	D	yes	
$x_1, x_2, s_2$	23	2	$2\frac{7}{15}$	$s_1, s_3$	F	yes	
$x_1, x_2, s_3$	23	$4\frac{7}{15}$	$-19\frac{11}{15}$	$s_1, s_2$	E	no	
	×2						
	10	(2)					
	-	(3)		(1	L)		
	-						
	5_ <sup>B</sup>			E	(2)		
	-			F		~	
				6	н		
	1	5	10	15 20	25	< <u>x1</u>	
			Lectur	e 3 Linear and Inte	eger Optimiz	ation with Applic	ations

# Basic **feasible** solutions correspond to solutions to the system of equations that **fulfil non-negativity**



$x_1$	+	<i>s</i> <sub>1</sub>	= 23
$0.067x_1$	$+x_{2}$	$+s_2$	= 6
3 <i>x</i> <sub>1</sub>	$+8x_{2}$	+s	i <sub>3</sub> = 85

# Basic **infeasible** solutions corresp. to solutions to the system of equations with one or more variables < 0



### Basic feasible solutions and the simplex method

Express the *m* basic variables in terms of the *n* – *m* non-basic variables

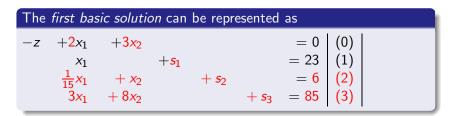
Example: Start at $x_1 = x_2 = 0 \Rightarrow s_2$	$_1, s_2, s_3$ are <i>basic</i>
$x_1$ $+s_1$	= 23
	+ <b>s</b> <sub>2</sub> = 6
$3x_1 + 8x_2$	+ <b>s</b> <sub>3</sub> = 85

Express  $s_1$ ,  $s_2$ , and  $s_3$  in terms of  $x_1$  and  $x_2$  (non-basic):

• We wish to maximize the value of the objective function  $2x_1 + 3x_2$ 

Express the objective in terms of the non-basic variables:(maximize) $z = 2x_1 + 3x_2$  $\Leftrightarrow$  $z - 2x_1 - 3x_2 = 0$ Lecture 3Linear and Integer Optimization with Applications

## Basic feasible solutions and the simplex method



- Marginal values for increasing the non-basic variables x<sub>1</sub> and x<sub>2</sub> from zero: 2 and 3, resp.
- $\Rightarrow Choose x_2 let x_2 enter the basis DRAW GRAPH!!$ 
  - One basic variable  $(s_1, s_2, \text{ or } s_3)$  must *leave the basis*. Which?

The value of  $x_2$  increases until a basic variable reaches the value 0:

$$\begin{array}{c} (2): s_2 = 6 - x_2 \ge 0 & \Rightarrow x_2 \le 6 \\ (3): s_3 = 85 - 8x_2 \ge 0 & \Rightarrow x_2 \le 10\frac{5}{8} \end{array} \right\} \Rightarrow \begin{array}{c} s_2 = 0 \text{ when } x_2 = 6 \\ \text{ (and } s_3 = 37) \end{array}$$

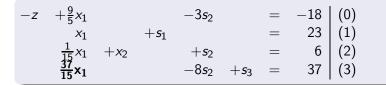
• s<sub>2</sub> will leave the basis

## Change basis through row operations

Eliminate $s_2$ from the basis let $x_2$ enter the basis—use row operations:											
-z	$+2x_{1}$	$+3x_{2}$				=	0	(0)			
	<i>x</i> <sub>1</sub>		$+s_1$			=	23	(1)			
	$\frac{1}{15}x_1$	$+x_{2}$		$+s_{2}$		=	6	(2)			
	$3x_1$	$+8x_{2}$			$+s_{3}$	=	85	(3)			
-z	$+\frac{9}{5}x_1$			-3 <i>s</i> <sub>2</sub>		=	-18	$(0) - 3 \cdot (2)$			
	<i>x</i> <sub>1</sub>		+ <i>s</i> 1			=	23	$(1) - 0 \cdot (2)$			
	$\frac{1}{15}x_1$	$+x_{2}$		$+s_{2}$		=	6	(2)			
	$\frac{\frac{1}{15}x_1}{\frac{37}{15}x_1}$			-8 <i>s</i> <sub>2</sub>	+ <i>s</i> <sub>3</sub>	=	37	(3)-8.(2)			

- Corresponding basic solution:  $s_1 = 23$ ,  $x_2 = 6$ ,  $s_3 = 37$ .
- Nonbasic variables:  $x_1 = s_2 = 0$
- The marginal value of  $x_1$  is  $\frac{9}{5} > 0$ . Let  $x_1$  enter the basis
- Which one should leave?  $s_1$ ,  $x_2$ , or  $s_3$ ?

## Change basis ... $x_1$ enters the basis (marginal value > 0)



The value of  $x_1$  increases until a basic variable reaches the value 0:

$$\begin{array}{ccc} (1): s_1 = 23 - x_1 \ge 0 & \Rightarrow x_1 \le 23 \\ (2): x_2 = 6 - \frac{1}{15} x_1 \ge 0 & \Rightarrow x_1 \le 90 \\ (3): s_3 = 37 - \frac{37}{15} x_1 \ge 0 & \Rightarrow x_1 \le 15 \end{array} \right\} \Rightarrow \begin{array}{c} s_3 = 0 \text{ when} \\ x_1 = 15 \end{array}$$

 $x_1$  enters and  $s_3$  leaves the basis: perform row operations:

-z				+2.84 <i>s</i> <sub>2</sub>	-0.73 <i>s</i> <sub>3</sub>	=	-45	$(0) - (3) \cdot \frac{15}{37} \cdot \frac{9}{5}$
			$s_1$	+3.24 <i>s</i> <sub>2</sub>	-0.41 <i>s</i> <sub>3</sub>	=	8	$ \begin{array}{c} (1) - (3) \cdot \frac{15}{37} \\ (2) - (3) \cdot \frac{15}{37} \cdot \frac{1}{15} \\ \end{array} $
		<i>x</i> <sub>2</sub>		$+1.22s_{2}$	-0.03 <i>s</i> <sub>3</sub>	=	5	$(2)-(3)\cdot\frac{15}{37}\cdot\frac{1}{15}$
	<i>x</i> <sub>1</sub>			-3.24 <i>s</i> <sub>2</sub>	$+0.41s_{3}$	=	15	$(3) \cdot \frac{15}{37}$

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# Change basis ... $s_2$ enters the basis (marginal value > 0)

-z			+2.84 <i>s</i> <sub>2</sub>	-0.73 <i>s</i> <sub>3</sub>	=	-45	(0)
		$s_1$	$+3.24s_{2}$	-0.41 <i>s</i> <sub>3</sub>	=	8	(1)
	<i>x</i> <sub>2</sub>		$+1.22s_{2}$	-0.03 <i>s</i> <sub>3</sub>	=	5	(2)
<i>x</i> <sub>1</sub>			-3.24 <i>s</i> <sub>2</sub>	$+0.41s_{3}$	=	15	(3)

The value of  $s_2$  increases until some basic variable value = 0:

$(1): s_1 = 8 - 3.24 s_2 \ge 0$	$\Rightarrow s_2 \leq 2.47$	)	$c_{\rm c} = 0$ when
$(2): x_2 = 5 - 1.22s_2 \ge 0$	$\Rightarrow$ s <sub>2</sub> $\leq$ 4.10	$\rangle \Rightarrow$	$s_1 = 0$ when
$(3): x_1 = 15 + 3.24s_2 \ge 0$	$\Rightarrow s_2 \geq -4.63$	J	$s_2 = 2.47$

 $s_2$  enters and  $s_1$  leaves the basis: perform row operations

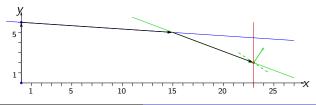
-z		-0.87 <i>s</i> 1		-0.37 <i>s</i> <sub>3</sub>	=	-52	$ \begin{array}{c} (0) - (1) \cdot \frac{2.8}{3.2} \\ (1) \cdot \frac{1}{3.24} \\ (2) - (1) \cdot \frac{1.2}{3.2} \end{array} $	4
		0.31 <i>s</i> 1	$+s_{2}$	$-0.12s_{3}$	=	2.47	$(1) \cdot \frac{1}{3.24}$	1
	<i>x</i> <sub>2</sub>	$-0.37s_1$		$+0.12s_{3}$	=	2	$(2)-(1)\cdot\frac{1.2}{3.2}$	2
<i>x</i> <sub>1</sub>		$+s_1$			=	23	(3)+(1)	

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## Optimal basic solution

-z		-0.87 <i>s</i> 1		-0.37 <i>s</i> 3	=	-52
		0.31 <i>s</i> 1	+ <i>s</i> <sub>2</sub>	-0.12 <i>s</i> <sub>3</sub>	=	2.47
	<i>x</i> <sub>2</sub>	-0.37 <i>s</i> 1		$+0.12s_{3}$	=	2
<i>x</i> <sub>1</sub>		$+s_1$			=	23

- No marginal value is positive. No improvement can be made
- The optimal basis is given by  $s_2 = 2.47$ ,  $x_2 = 2$ , and  $x_1 = 23$
- Non-basic variables:  $s_1 = s_3 = 0$



# Summary of the solution course

basis	-z	<i>x</i> <sub>1</sub>	<i>x</i> <sub>2</sub>	<i>s</i> <sub>1</sub>	<i>s</i> <sub>2</sub>	<i>S</i> 3	RHS
- <i>z</i>	1	2	3	0	0	0	0
$s_1$	0	1	0	1	0	0	23
<i>s</i> <sub>2</sub>	0	0.067	1	0	1	0	6
<b>s</b> 3	0	3	8	0	0	1	85
- <i>z</i>	1	1.80	0	0	-3	0	-18
$s_1$	0	1	0	1	0	0	23
<i>x</i> <sub>2</sub>	0	0.07	1	0	1	0	6
<b>S</b> 3	0	2.47	0	0	-8	1	37
- <i>z</i>	1	0	0	0	2.84	-0.73	-45
$s_1$	0	0	0	1	3.24	-0.41	8
<i>x</i> <sub>2</sub>	0	0	1	0	1.22	-0.03	5
<i>x</i> <sub>1</sub>	0	1	0	0	-3.24	0.41	15
- <i>z</i>	1	0	0	-0.87	0	-0.37	-52
<i>s</i> <sub>2</sub>	0	0	0	0.31	1	-0.12	2.47
<i>x</i> <sub>2</sub>	0	0	1	-0.37	0	0.12	2
$x_1$	0	1	0	1	0	0	23

Solve the lego problem using the simplex method
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maximize	Ζ	=	1600 <i>x</i> 1	+	1000 <i>x</i> <sub>2</sub>		
subject to			$2x_1$	+	<i>x</i> <sub>2</sub>	$\leq$	6
			$2x_1$	+	$2x_2$	$\leq$	8
					$x_1, x_2$	$\geq$	0