

MVE165/MMG631

Linear and Integer Optimization with Applications
Lecture 3

Extreme points of convex polyhedra;
reformulations; basic feasible solutions; the simplex
method

Ann-Brith Strömberg

2015-03-27

Linear programs, convex polyhedra, extreme points

A linear optimization model – a linear program

$$\begin{aligned} & \text{minimize} && z = \sum_{j=1}^n c_j x_j \\ & \text{subject to} && \sum_{j=1}^n a_{ij} x_j \leq b_i, \quad i = 1, \dots, m \\ & && x_j \geq 0, \quad j = 1, \dots, n \end{aligned}$$

c_j, a_{ij}, b_i : constant parameters

Expressed in vector notation

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

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

The feasible region is a polyhedron, $X \subset \mathbb{R}_+^n$

$$X := \left\{ \mathbf{x} \geq \mathbf{0}^n \mid \sum_{j=1}^n a_{ij} x_j \leq b_i, i = 1, \dots, m \right\} = \{ \mathbf{x} \geq \mathbf{0}^n \mid \mathbf{Ax} \leq \mathbf{b} \}$$

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

Definition (Convex combination)

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

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

[DRAW ON THE BOARD]

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

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

$$\begin{aligned}\sum_{j=1}^n a_{ij} (\lambda x_j^1 + (1 - \lambda)x_j^2) &= \lambda \sum_{j=1}^n a_{ij} x_j^1 + (1 - \lambda) \sum_{j=1}^n a_{ij} x_j^2 \\ &\leq \lambda b_i + (1 - \lambda)b_i \\ &= b_i, \quad i = 1, \dots, m\end{aligned}$$

and

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

[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 $\mathbf{x}^k = \mathbf{x}^1 = \mathbf{x}^2$.

[DRAW ON THE BOARD]

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

Assume that the feasible region $X = \{\mathbf{x} \geq \mathbf{0}^n \mid \mathbf{A}\mathbf{x} \leq \mathbf{b}\}$ 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 .

[PROOF ON THE BOARD]

A general linear program – notation

Definition (Notation of linear programs)

minimize or maximize $c_1x_1 + \dots + c_nx_n$

subject to $a_{i1}x_1 + \dots + a_{in}x_n \left\{ \begin{array}{l} \leq \\ = \\ \geq \end{array} \right\} b_i, \quad i = 1, \dots, m$

$x_j \left\{ \begin{array}{l} \leq 0 \\ \text{unrestricted in sign} \\ \geq 0 \end{array} \right\}, \quad j = 1, \dots, n$

The blue notation refers 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) can handle also inequality constraints and unrestricted variables – the reformulations are made automatically

The simplex method—standard form reformulations

- Slack variables:

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

- The lego example:

$$\left[\begin{array}{l} 2x_1 + x_2 \leq 6 \\ 2x_1 + 2x_2 \leq 8 \\ x_1, x_2 \geq 0 \end{array} \right] \Leftrightarrow \left[\begin{array}{l} 2x_1 + x_2 + s_1 = 6 \\ 2x_1 + 2x_2 + s_2 = 8 \\ x_1, x_2, s_1, s_2 \geq 0 \end{array} \right]$$

- s_1 and s_2 are called *slack variables*—they “fill out” the (positive) distances between the left and right hand sides

The simplex method—standard form reformulations

- Surplus variables:

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

- *Surplus variable* s_3 (a different example):

$$\left[\begin{array}{l} x_1 + x_2 \geq 800 \\ x_1, x_2 \geq 0 \end{array} \right] \Leftrightarrow \left[\begin{array}{l} x_1 + x_2 - s_3 = 800 \\ x_1, x_2, s_3 \geq 0 \end{array} \right]$$

The simplex method—standard form reformulations

- Suppose that $b < 0$:

$$\left[\begin{array}{l} \sum_{j=1}^n a_j x_j \leq b \\ x_j \geq 0, \forall j \end{array} \right] \Leftrightarrow \left[\begin{array}{l} \sum_{j=1}^n (-a_j) x_j \geq -b \\ x_j \geq 0, \forall j \end{array} \right] \Leftrightarrow \left[\begin{array}{l} -\sum_{j=1}^n a_j x_j - s = -b \\ x_j \geq 0, \forall j \\ s \geq 0 \end{array} \right]$$

- Non-negative right hand side:

$$\left[\begin{array}{l} x_1 - x_2 \leq -23 \\ x_1, x_2 \geq 0 \end{array} \right] \Leftrightarrow \left[\begin{array}{l} -x_1 + x_2 \geq 23 \\ x_1, x_2 \geq 0 \end{array} \right] \Leftrightarrow \left[\begin{array}{l} -x_1 + x_2 - s_4 = 23 \\ x_1, x_2, s_4 \geq 0 \end{array} \right]$$

The simplex method—standard form reformulations

- 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:

$$\left[\begin{array}{l} \sum_{j=1}^n a_j x_j \leq b \\ x_j \geq 0, j = 1, \dots, k \end{array} \right] \Leftrightarrow \left[\begin{array}{l} \sum_{j=1}^k a_j x_j + \sum_{j=k+1}^n a_j (x_j^1 - x_j^2) + s = b \\ x_j \geq 0, j = 1, \dots, k, \\ x_j^1 \geq 0, x_j^2 \geq 0, j = k+1, \dots, n \\ s \geq 0 \end{array} \right]$$

- Sign-restricted (non-negative) variables:

$$\left[\begin{array}{l} x_1 + x_2 \leq 10 \\ x_1 \geq 0 \end{array} \right] \Leftrightarrow \left[\begin{array}{l} x_1 + x_2^1 - x_2^2 \leq 10 \\ x_1, x_2^1, x_2^2 \geq 0 \end{array} \right] \Leftrightarrow \left[\begin{array}{l} x_1 + x_2^1 - x_2^2 + s_5 = 10 \\ x_1, x_2^1, x_2^2, s_5 \geq 0 \end{array} \right]$$

Basic feasible solutions (Ch. 4.3)

- Consider m equations with n variables, where $m \leq n$
- 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
- A basic solution corresponds to an *intersection* (feasible ($\mathbf{x} \geq \mathbf{0}$) or infeasible ($\mathbf{x} \not\geq \mathbf{0}$)) of m hyperplanes in \mathbb{R}^m
- Each *extreme point* of the feasible set is an intersection of m hyperplanes such that all variable values are ≥ 0
- **Basic feasible solution \Leftrightarrow extreme point of the feasible set**

$$\begin{array}{rcl} a_{11}x_1 + \dots + a_{1n}x_n = b_1 & & x_1 \geq 0 \\ a_{21}x_1 + \dots + a_{2n}x_n = b_2 & & x_2 \geq 0 \\ & \dots & \dots \\ a_{m1}x_1 + \dots + a_{mn}x_n = b_m & & x_n \geq 0 \end{array}$$

Basic feasible solutions

- Assume that $m < n$ and that $b_i \geq 0$, $i = 1, \dots, m$, 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

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

- Partition \mathbf{x} into m basic variables \mathbf{x}_B and $n - m$ non-basic variables \mathbf{x}_N , such that $\mathbf{x} = (\mathbf{x}_B, \mathbf{x}_N)$.
- 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)

Basic feasible solutions (Ch. 4.8)

Rewrite the linear program as

$$\text{minimize } z = \mathbf{c}_B^T \mathbf{x}_B + \mathbf{c}_N^T \mathbf{x}_N \quad (1a)$$

$$\text{subject to } \mathbf{B}\mathbf{x}_B + \mathbf{N}\mathbf{x}_N = \mathbf{b} \quad (1b)$$

$$\mathbf{x}_B \geq \mathbf{0}^m, \mathbf{x}_N \geq \mathbf{0}^{n-m} \quad (1c)$$

Multiply the equation (1b) with \mathbf{B}^{-1} from the left:

$$\begin{aligned} \mathbf{B}^{-1}\mathbf{B}\mathbf{x}_B + \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N &= \mathbf{x}_B + \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N = \mathbf{B}^{-1}\mathbf{b} \\ \Rightarrow \mathbf{x}_B &= \mathbf{B}^{-1}\mathbf{b} - \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N \end{aligned} \quad (2)$$

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

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

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

$$\text{subject to } \mathbf{B}^{-1}\mathbf{b} - \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N \geq \mathbf{0}^m$$

$$\mathbf{x}_N \geq \mathbf{0}^{n-m}$$

The rewritten program

$$\text{minimize } z = \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{b} + (\mathbf{c}_N^T - \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{N}) \mathbf{x}_N \quad (3a)$$

$$\text{subject to } \mathbf{B}^{-1} \mathbf{b} - \mathbf{B}^{-1} \mathbf{N} \mathbf{x}_N \geq \mathbf{0}^m \quad (3b)$$

$$\mathbf{x}_N \geq \mathbf{0}^{n-m} \quad (3c)$$

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

- 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 $z = \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{b}$
- The basic solution is feasible if $\mathbf{B}^{-1} \mathbf{b} \geq \mathbf{0}^m$

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

Optimality condition (for minimization)

The basis B is optimal if $\mathbf{c}_N^T - \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{N} \geq \mathbf{0}^{n-m}$
(marginal values = reduced costs ≥ 0)

If not, choose as *entering* variable $j \in N$ the one with the lowest (negative) value of the reduced cost $c_j - \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{A}_j$

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

$$i^* = \arg \min_{i \in B} \left\{ \frac{(\mathbf{B}^{-1} \mathbf{b})_i}{(\mathbf{B}^{-1} \mathbf{A}_j)_i} \mid (\mathbf{B}^{-1} \mathbf{A}_j)_i > 0 \right\}$$

Basic feasible solutions, example

- Constraints:

$$x_1 \leq 23 \quad (1)$$

$$0.067x_1 + x_2 \leq 6 \quad (2)$$

$$3x_1 + 8x_2 \leq 85 \quad (3)$$

$$x_1, x_2 \geq 0$$

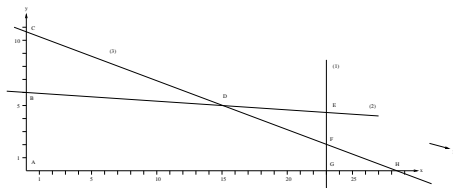
- Add slack variables:

$$x_1 + s_1 = 23 \quad (1)$$

$$0.067x_1 + x_2 + s_2 = 6 \quad (2)$$

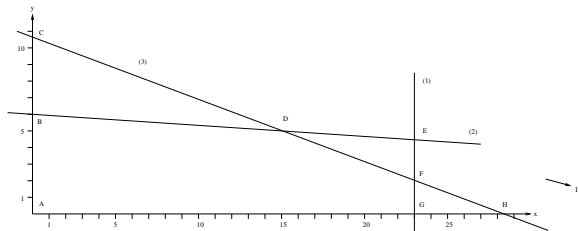
$$3x_1 + 8x_2 + s_3 = 85 \quad (3)$$

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

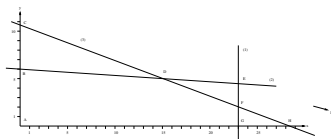


Basic and non-basic variables and solutions

basic variables	basic solution			non-basic variables (0, 0)	point	feasible?
s_1, s_2, s_3	23	6	85	x_1, x_2	A	yes
s_1, s_2, x_1	$-5\frac{1}{3}$	$4\frac{1}{9}$	$28\frac{1}{3}$	s_3, x_2	H	no
s_1, s_2, x_2	23	$-4\frac{5}{8}$	$10\frac{5}{8}$	x_1, s_3	C	no
s_1, x_1, s_3	-67	90	-185	s_2, x_2	I	no
s_1, x_2, s_3	23	6	37	s_2, x_1	B	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	-	-
x_1, x_2, s_1	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



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



$$\begin{bmatrix} x_1 & +s_1 & & = 23 \\ 0.067x_1 & +x_2 & +s_2 & = 6 \\ & 3x_1 & +8x_2 & +s_3 = 85 \end{bmatrix}$$

$$A: x_1 = x_2 = 0 \Rightarrow \begin{bmatrix} s_1 & & = 23 \\ & s_2 & = 6 \\ & & s_3 = 85 \end{bmatrix}$$

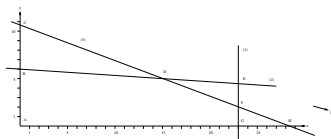
$$B: x_1 = s_2 = 0 \Rightarrow \begin{bmatrix} & s_1 & = 23 \\ x_2 & & = 6 \\ 8x_2 & +s_3 & = 85 \end{bmatrix}$$

$$D: s_3 = s_2 = 0 \Rightarrow \begin{bmatrix} x_1 & +s_1 & = 23 \\ 0.067x_1 & +x_2 & = 6 \\ 3x_1 & +8x_2 & = 85 \end{bmatrix}$$

$$F: s_3 = s_1 = 0 \Rightarrow \begin{bmatrix} x_1 & & = 23 \\ 0.067x_1 & +x_2 & +s_2 = 6 \\ 3x_1 & +8x_2 & = 85 \end{bmatrix}$$

$$G: x_2 = s_1 = 0 \Rightarrow \begin{bmatrix} x_1 & & = 23 \\ 0.067x_1 & +s_2 & = 6 \\ 3x_1 & & +s_3 = 85 \end{bmatrix}$$

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



$$\begin{bmatrix} x_1 & +s_1 & & = 23 \\ 0.067x_1 & +x_2 & +s_2 & = 6 \\ 3x_1 & +8x_2 & & +s_3 = 85 \end{bmatrix}$$

$$H: x_2 = s_3 = 0 \Rightarrow \begin{bmatrix} x_1 & +s_1 & & = 23 \\ 0.067x_1 & & +s_2 & = 6 \\ 3x_1 & & & = 85 \end{bmatrix}$$

$$C: x_1 = s_3 = 0 \Rightarrow \begin{bmatrix} & s_1 & & = 23 \\ x_2 & & +s_2 & = 6 \\ 8x_2 & & & = 85 \end{bmatrix}$$

$$I: s_2 = x_2 = 0 \Rightarrow \begin{bmatrix} x_1 & +s_1 & & = 23 \\ 0.067x_1 & & & = 6 \\ 3x_1 & & +s_3 & = 85 \end{bmatrix}$$

$$\therefore s_1 = x_1 = 0 \Rightarrow \begin{bmatrix} & & & 0 & = 23 \\ x_2 & +s_2 & & & = 6 \\ 8x_2 & & +s_3 & & = 85 \end{bmatrix}$$

$$E: s_1 = s_2 = 0 \Rightarrow \begin{bmatrix} x_1 & & & & = 23 \\ 0.067x_1 & +x_2 & & & = 6 \\ 3x_1 & +8x_2 & +s_3 & & = 85 \end{bmatrix}$$

Basic feasible solutions and the simplex method

- Express the m *basic* variables in terms of the $n - m$ *non-basic* variables
- Example: Start at $x_1 = x_2 = 0 \Rightarrow s_1, s_2, s_3$ are *basic*

$$\begin{bmatrix} x_1 & & +s_1 & & = 23 \\ \frac{1}{15}x_1 & +x_2 & & +s_2 & = 6 \\ 3x_1 & +8x_2 & & & +s_3 = 85 \end{bmatrix}$$

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

$$\begin{bmatrix} s_1 = 23 & -x_1 \\ s_2 = 6 & -\frac{1}{15}x_1 & -x_2 \\ s_3 = 85 & -3x_1 & -8x_2 \end{bmatrix}$$

- We wish to maximize the objective function $2x_1 + 3x_2$
- Express the objective in terms of the *non-basic* variables:

$$\text{(maximize)} \quad z = 2x_1 + 3x_2 \quad \Leftrightarrow \quad z - 2x_1 - 3x_2 = 0$$

Basic feasible solutions and the simplex method

- The *first basic solution* can be represented as

$-z$	$+2x_1$	$+3x_2$		$= 0$	(0)
	x_1		$+s_1$	$= 23$	(1)
	$\frac{1}{15}x_1$	$+x_2$		$+s_2 = 6$	(2)
	$3x_1$	$+8x_2$		$+s_3 = 85$	(3)

- Marginal values** for increasing the non-basic variables x_1 and x_2 from zero: 2 and 3, resp.

⇒ Choose x_2 — let x_2 *enter the basis* DRAW GRAPH!!

- One basic variable (s_1 , s_2 , or s_3) must *leave the basis*. Which?
- The value of x_2 can increase until some basic variable reaches the value 0:

$$\left. \begin{array}{l} (2) : s_2 = 6 - x_2 \geq 0 \Rightarrow x_2 \leq 6 \\ (3) : s_3 = 85 - 8x_2 \geq 0 \Rightarrow x_2 \leq 10\frac{5}{8} \end{array} \right\} \Rightarrow \begin{array}{l} s_2 = 0 \text{ when} \\ x_2 = 6 \\ (\text{and } s_3 = 37) \end{array}$$

- s_2 will leave the basis

Change basis through row operations

- Eliminate s_2 from the basis, let x_2 enter the basis using row operations:

$-z$	$+2x_1$	$+3x_2$			$=$	0	(0)
	x_1		$+s_1$		$=$	23	(1)
	$\frac{1}{15}x_1$	$+x_2$		$+s_2$	$=$	6	(2)
	$3x_1$	$+8x_2$		$+s_3$	$=$	85	(3)
$-z$	$+\frac{9}{5}x_1$			$-3s_2$	$=$	-18	$(0) - 3 \cdot (2)$
	x_1		$+s_1$		$=$	23	$(1) - 0 \cdot (2)$
	$\frac{1}{15}x_1$	$+x_2$		$+s_2$	$=$	6	(2)
	$\frac{37}{15}x_1$			$-8s_2 + s_3$	$=$	37	$(3) - 8 \cdot (2)$

- 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 ...

$-z$	$+\frac{9}{5}x_1$		$-3s_2$	$=$	-18	(0)
	x_1		$+s_1$	$=$	23	(1)
	$\frac{1}{15}x_1$	$+x_2$	$+s_2$	$=$	6	(2)
	$\frac{37}{15}x_1$		$-8s_2 + s_3$	$=$	37	(3)

- The value of x_1 can increase until some basic variable reaches the value 0:

$$\left. \begin{array}{l} (1) : s_1 = 23 - x_1 \geq 0 \Rightarrow x_1 \leq 23 \\ (2) : x_2 = 6 - \frac{1}{15}x_1 \geq 0 \Rightarrow x_1 \leq 90 \\ (3) : s_3 = 37 - \frac{37}{15}x_1 \geq 0 \Rightarrow x_1 \leq 15 \end{array} \right\} \Rightarrow \begin{array}{l} s_3 = 0 \text{ when} \\ x_1 = 15 \end{array}$$

- x_1 enters the basis and s_3 leaves the basis
- Perform row operations:

$-z$		$+2.84s_2$	$-0.73s_3$	$=$	-45	$(0) - (3) \cdot \frac{15}{37} \cdot \frac{9}{5}$
	s_1	$+3.24s_2$	$-0.41s_3$	$=$	8	$(1) - (3) \cdot \frac{15}{37}$
	x_2	$+1.22s_2$	$-0.03s_3$	$=$	5	$(2) - (3) \cdot \frac{15}{37} \cdot \frac{1}{15}$
	x_1	$-3.24s_2$	$+0.41s_3$	$=$	15	$(3) \cdot \frac{15}{37}$

Change basis ...

$-z$		$+2.84s_2$	$-0.73s_3$	$=$	-45	(0)
	s_1	$+3.24s_2$	$-0.41s_3$	$=$	8	(1)
	x_2	$+1.22s_2$	$-0.03s_3$	$=$	5	(2)
	x_1	$-3.24s_2$	$+0.41s_3$	$=$	15	(3)

- Let s_2 enter the basis (marginal value > 0)
- The value of s_2 can increase until some basic variable = 0:

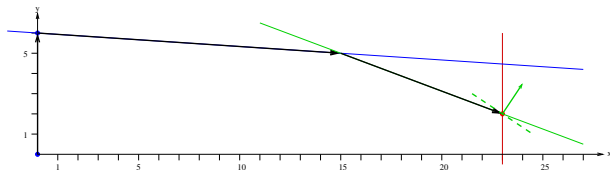
$(1) : s_1 = 8 - 3.24s_2 \geq 0 \Rightarrow s_2 \leq 2.47$	}	\Rightarrow	$s_1 = 0$ when
$(2) : x_2 = 5 - 1.22s_2 \geq 0 \Rightarrow s_2 \leq 4.10$			
$(3) : x_1 = 15 + 3.24s_2 \geq 0 \Rightarrow s_2 \geq -4.63$			
			$s_2 = 2.47$
- s_2 enters the basis and s_1 will leave the basis
- Perform row operations:

$-z$		$-0.87s_1$		$-0.37s_3$	$=$	-52	$(0) - (1) \cdot \frac{2.84}{3.24}$
		$0.31s_1$	$+s_2$	$-0.12s_3$	$=$	2.47	$(1) \cdot \frac{1}{3.24}$
	x_2	$-0.37s_1$		$+0.12s_3$	$=$	2	$(2) - (1) \cdot \frac{1.22}{3.24}$
	x_1	$+s_1$			$=$	23	$(3) + (1)$

Optimal basic solution

$-z$	$-0.87s_1$	$-0.37s_3$	$=$	-52	
	$0.31s_1$	$+s_2$	$-0.12s_3$	$=$	2.47
	x_2	$-0.37s_1$	$+0.12s_3$	$=$	2
	x_1	$+s_1$		$=$	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: $s_1 = s_3 = 0$
- Optimal value: $z = 52$



Summary of the solution course

basis	$-z$	x_1	x_2	s_1	s_2	s_3	RHS
$-z$	1	2	3	0	0	0	0
s_1	0	1	0	1	0	0	23
s_2	0	0.067	1	0	1	0	6
s_3	0	3	8	0	0	1	85
$-z$	1	1.80	0	0	-3	0	-18
s_1	0	1	0	1	0	0	23
x_2	0	0.07	1	0	1	0	6
s_3	0	2.47	0	0	-8	1	37
$-z$	1	0	0	0	2.84	-0.73	-45
s_1	0	0	0	1	3.24	-0.41	8
x_2	0	0	1	0	1.22	-0.03	5
x_1	0	1	0	0	-3.24	0.41	15
$-z$	1	0	0	-0.87	0	-0.37	-52
s_2	0	0	0	0.31	1	-0.12	2.47
x_2	0	0	1	-0.37	0	0.12	2
x_1	0	1	0	1	0	0	23

