

Operations Research I

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Linear Programming

Plan

- 1 Introduction to Linear Programming
- 2 Geometrical interpretation
- 3 Basis and extreme points
- 4 The simplex algorithm

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Linear Programming

Framework

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finite number of real variables, linear constraints, linear objective

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Real variables x_1, x_2, \dots, x_n

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Real variables $x_1, x_2 \dots x_n$

Constraint (i):

$$\sum_{j=1}^n a_{ij} x_j \leq b_i$$

Linear Programming

Framework

Linear Programming

finite number of real variables, linear constraints, linear objective

Real variables $x_1, x_2 \dots x_n$

Constraint (i):

$$\sum_{j=1}^n a_{ij} x_j \leq b_i$$

Objective function (maximize / minimize):

$$f(x_1, x_2 \dots x_n) = \sum_{j=1}^n c_j x_j$$

Linear Programming

Example: Cucumber and onions culture

Constraints about the quantities of fertilizer and anti-parasite

- 8ℓ of fertilizer A available
→ $2\ell/m^2$ for cucumbers, $1\ell/m^2$ for onions
- 7ℓ of fertilizer B available
→ $1\ell/m^2$ for cucumbers, $2\ell/m^2$ for onions
- 3ℓ of anti-parasites available
→ $1\ell/m^2$ for onions

Objective: produce the maximum weight of vegetables knowing that the yield is $4kg/m^2$ for cucumbers, $5kg/m^2$ for onions

Linear Programming

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Decision variables

- x_c : area of cucumbers
- x_o : area of onions

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- $2x_c + x_o \leq 8$ (fertilizer A)
- $x_c + 2x_o \leq 7$ (fertilizer B)
- $x_o \leq 3$ (anti-parasites)
- $x_c \geq 0$ and $x_o \geq 0$

Linear Programming

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General optimization problem with constraints

⇒ **NO GENERAL solution method!!**

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Any linear problem

⇒ general and efficient solution methods

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Those methods are efficient in theory and in practice

⇒ existence of numerous solution softwares:

Excel, CPLEX, Mathematica, LP-Solve...

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Restrictive framework

- real variables
- linear constraints
- linear objective

Linear Programming

In extenso representation

- $\max 4x_c + 5x_o$
- $2x_c + x_o \leq 8$ (fertilizer A)
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Linear Programming

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matrix representation

$$\max \quad (4 \quad 5) \begin{pmatrix} x_c \\ x_o \end{pmatrix}$$

$$\begin{pmatrix} 2 & 1 \\ 1 & 2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_c \\ x_o \end{pmatrix} \leq \begin{pmatrix} 8 \\ 7 \\ 3 \end{pmatrix}$$

$$x_c \geq 0 \quad x_o \geq 0$$

Linear Programming

in extenso representation

$$\max z = \sum_j c_j x_j$$

$$s.t. \quad \sum_j a_{ij} x_j \quad \left\{ \begin{array}{c} \leq \\ \geq \\ = \end{array} \right\} b_i \quad i = 1, 2 \dots m$$

$$x_j \geq 0 \quad j = 1, 2 \dots n$$

Linear Programming

- second member $b = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}$

- matrix $m \times n$

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ & & \ddots & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

- cost (or profit) $c = (c_1, c_2 \dots c_n)$

- n decision variables $X = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$

Linear Programming

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Matrix representation

$$\max z = CX$$

$$s.t. \quad Ax \begin{cases} \leq \\ \geq \\ = \end{cases} b$$

$$x \geq 0$$

Linear Programming

Vocabulary

- x_i **decision variable** of the problem
- $x = (x_1, \dots, x_n)$ **feasible solution**
iff it satisfies all constraints
- set of all feasible solutions = **admissible region**
- $x = (x_1, \dots, x_n)$ **optimal solution**
iff it is feasible and it optimizes the objective function

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- **constraints** linear equalities or inequalities
 - $a_{11}x_1 + a_{12}x_2 \dots + a_{1n}x_n \leq b_1$
 - $a_{21}x_1 + a_{22}x_2 \dots + a_{2n}x_n \geq b_2$
 - $a_{31}x_1 + a_{32}x_2 \dots + a_{3n}x_n = b_3$

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 - $a_{31}x_1 + a_{32}x_2 \dots + a_{3n}x_n = b_3$
- linear **objective function** (or economical function)
 - $\max / \min c_1x_1 + c_2x_2 \dots + c_nx_n$

Programmation linéaire

Applications

Exercice sheet: Linear programming

- Wine production
- Advertizing
- Olive oil production
- Bergamote

Caseine: Lab Linear Programming

- Perfumes
- Wines Q1 et Q2
- Dairy Products

Linear Programming

Canonical form of an LP

- maximization
- all variables are non negative
- all constraints are inequalities of the type " \leq "

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- matrix form

$$\max z = cx$$

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Standard form of an LP

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Linear Programming

Changing form

- equality \rightarrow inequality

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$$ax = b \iff \begin{cases} ax \leq b \\ ax \geq b \end{cases}$$

- $\max \leftrightarrow \min$

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- equality \rightarrow inequality

$$ax = b \iff \begin{cases} ax \leq b \\ ax \geq b \end{cases}$$

- $\max \leftrightarrow \min$ $\max f(x) = -\min -f(x)$
- inequality \rightarrow equality:

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- inequality \rightarrow equality: add a slack variable

$$\begin{aligned} ax \leq b &\iff ax + s = b, & s \geq 0 \\ ax \geq b &\iff ax - s = b, & s \geq 0 \end{aligned}$$

- unconstrained variable \rightarrow positive variable

Linear Programming

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- unconstrained variable \rightarrow positive variable

$$x \leq 0 \iff \begin{cases} x = x^+ - x^- \\ x^+, x^- \geq 0 \end{cases}$$

Linear Programming

Changing form

Exercices sheet: Linear Programming

- Linear and canonical forms

Linear Programming

Linearize a non linear formulation

e_i : linear expression of the decision variables

- **obj:** $\min \max\{e_1, e_2 \dots e_n\}$

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$$\begin{cases} \min y \\ y \geq e_i \quad i = 1, 2 \dots n \end{cases}$$

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- **obj:** $\min |e_1|$

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$$\begin{cases} \max y \\ y \leq e_i & i = 1, 2 \dots n \end{cases}$$

- **obj:** $\min |e_1|$

$$|e| = \max(e, -e) \quad \begin{cases} \min y \\ y \geq e_1 \\ y \geq -e_1 \end{cases} \quad \begin{cases} \min e^+ + e^- \\ e_1 = e^+ - e^- \\ e^+, e^- \geq 0 \end{cases}$$

Linear Programming

Linearize a non linear formulation

Exercices sheet: Linear Programming

- Linearization

Linear Programming

A little history

- 30's-40's: Kantorovitch, soviet economist
⇒ linear models for production planning and optimization
- 40's-50's: Dantzig, american mathematician
⇒ simplex algorithm

Linear Programming

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⇒ simplex algorithm
- historical application
 - Operation Vittles and Plainfare for supply of the trizone during the blockade of Berlin by airlift (June 23, 1948 has – May 12, 1949)
 - simplex executed by hand (thousands of variables), up to 12 000 tons of hardware each day!

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- 1975: Kantorovitch has the Nobel price in economy
- XXIème century: software with LP available everywhere, use of LP in all industrial domains...

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Geometrical interpretation

Example: Cucumber and onions culture

Decision variables

- x_c : area of cucumbers
- x_o : area of onions

Objective function $\max 4x_c + 5x_n$

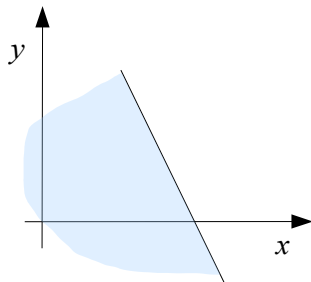
Constraints

- $2x_c + x_n \leq 8$ (fertilizer A)
- $x_c + 2x_n \leq 7$ (fertilizer B)
- $x_n \leq 3$ (anti-parasites)
- $x_c \geq 0$ and $x_n \geq 0$

Geometrical interpretation

Interpret the constraints cucumbers and onions

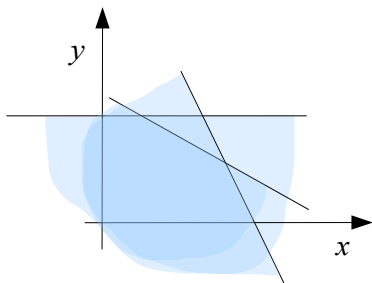
- $2x + y \leq 8 \Rightarrow$ half-plane of \mathbb{R}^2



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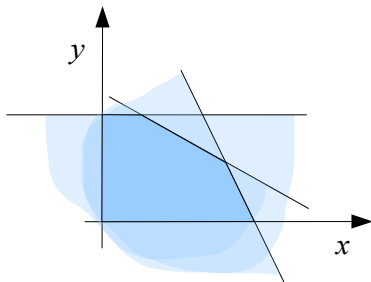
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- $x + 2y \leq 7 \Rightarrow$ half-plane
- $y \leq 3 \Rightarrow$ half-plane



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- $2x + y \leq 8 \Rightarrow$ half-plane of \mathbb{R}^2
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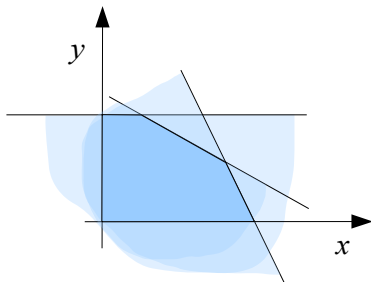


Geometrical interpretation

Interpret the constraints cucumbers and onions

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Set of feasible solutions = intersection of half-planes: **polyedron**



Geometrical interpretation

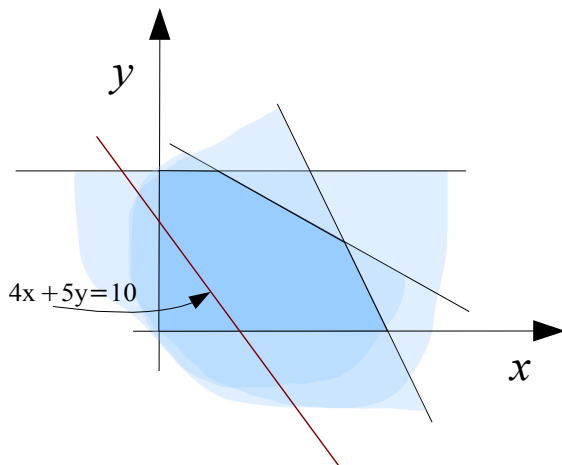
Optimize the objective

The **level lines** $\{4x + 5y = \text{constant}\}$ are parallel lines

Geometrical interpretation

Optimize the objective

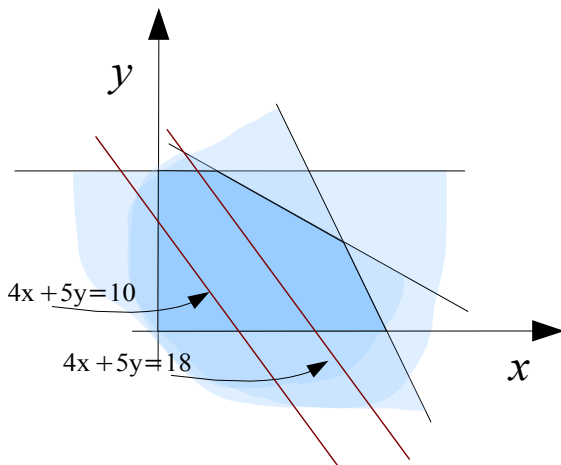
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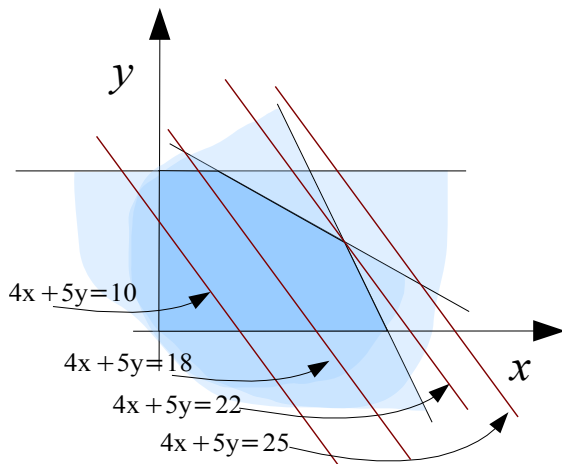
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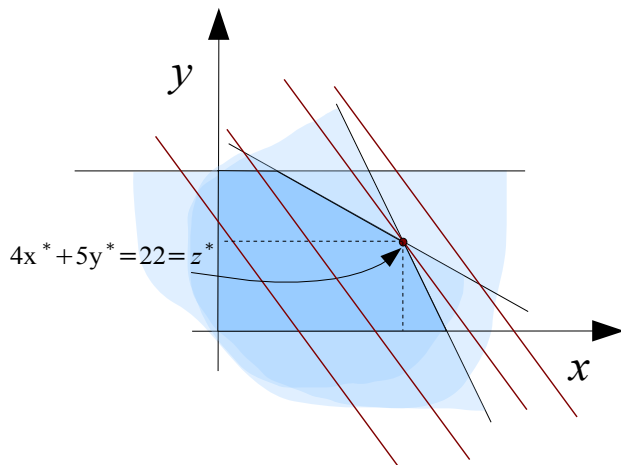
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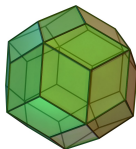
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Geometrical interpretation

Geometry of a PL

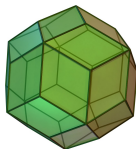
The set of feasible solutions is always a **polyhedron** (intersection of half-spaces)



Geometrical interpretation

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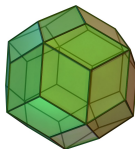
The set of feasible solutions is always a **polyhedron** (intersection of half-spaces)



The level lines $\{f = \text{constant}\}$ of the objective function f are **affine hyperplanes** ($n = 2 \Rightarrow$ line, $n = 3 \Rightarrow$ plan...)

Geometrical interpretation

Geometry of a PL



Optimum is reached on the edge

The optimum of the objective function, if it exists, is reached on (at least) one of the **vertices** of the polyhedron

Mathematical justification:

the partial derivatives of $f(x) = c \cdot x$ are never zero,

and the domain $\{x \mid \sum_{j=1}^n a_{ij}x_j \leq b_i, i = 1, \dots, m\}$ is compact

\Rightarrow the optimum is reached on the edge...

Linear Programming

Solutions of an LP

The feasible region can be

- empty
- non empty, bounded
- non empty, unbounded

Propose examples of LP for each case

Linear Programming

Solutions of an LP

The feasible region can be

- empty
 - nb of optimal solutions: 0
- non empty, bounded
 - nb of optimal solutions: 1 or ∞
- non empty, unbounded
 - nb of optimal solutions: 0 or 1 or ∞

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Linear Programming

Solutions of an LP

The feasible region can be

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Propose examples of LP for each case

Exercice sheet: Linear programming

- **Graphical solution**

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Basis and extreme points

Recall

$$\begin{array}{llll} \max & z & = & cx \\ \text{s.t.} & Ax & \leq & b \\ & x & \geq & 0 \end{array}$$

- A matrix $m \times n$
- $x = (x_1 \ x_2 \ \dots \ x_n)$
- $b = (b_1 \ b_2 \ \dots \ b_m)$
- $c = (c_1 \ c_2 \ \dots \ c_n)$

- The constraints define a polyhedron
- An optimal solution is a vertex of the polyhedron

How to enumerate the vertices of a polyhedron?

Basis and extreme points

Change to the standard form

Standard form

Add **slack variables**:

$$\sum_{j=1}^n a_{ij}x_j \leq b_i \Leftrightarrow \sum_{j=1}^n a_{ij}x_j + e_i = b_i, e_i \geq 0$$

Standard LP:

$$\begin{array}{lll} \max & z(x) & = \quad c \cdot x \\ \text{s.c} & Ax & = \quad b \\ & x & \geq \quad 0 \end{array}$$

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We work on a space of higher dimension but all constraints are equalities

► Easier algebraic manipulations

Basis and extreme points

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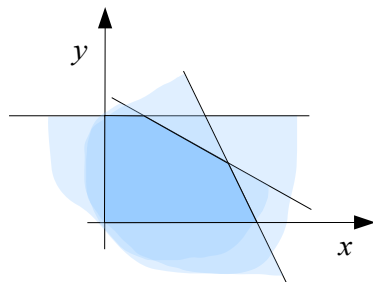
$$\max z = 4x + 5y$$

$$\text{s.t. } 2x + y \leq 8$$

$$x + 2y \leq 7$$

$$y \leq 3$$

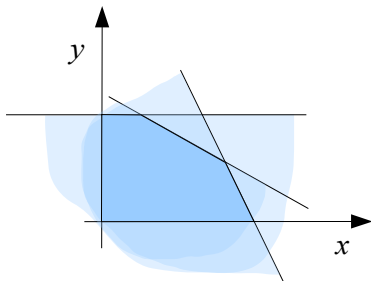
$$x, y \geq 0$$



Basis and extreme points

Change to the standard form

$$\begin{aligned}\max z &= 4x + 5y \\ \text{s.t. } 2x + y &\leq 8 \\ x + 2y &\leq 7 \\ y &\leq 3 \\ x, y &\geq 0\end{aligned}$$



$$\begin{aligned}\max z &= 4x + 5y \\ \text{s.t. } 2x + y + e_1 &= 8 \\ x + 2y + e_2 &= 7 \\ y + e_3 &= 3 \\ x, y, e_1, e_2, e_3 &\geq 0\end{aligned}$$

9 interesting points
(intersection of constraints)

5 feasible points

enumeration of those 9 points
as solutions of the standard
form (basic solution)

Basis and extreme points

$$\begin{array}{rclclclclcl} \text{s.t.} & 2x & + & y & + & e_1 & & & = & 8 \\ & x & + & 2y & & & + & e_2 & = & 7 \\ & & & y & & & & + & e_3 & = & 3 \\ & x, & & y, & & e_1, & & e_2, & & e_3 & \geq & 0 \end{array}$$

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 \text{s.t.} & 2x & + & y & + & e_1 & & = & 8 \\
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x	y	e_1	e_2	e_3	basic solution	admiss.	extreme pt
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Basis and extreme points

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 \end{array}$$

x	y	e_1	e_2	e_3	basic solution	admiss.	extreme pt
<u>0</u>	<u>0</u>	8	7	3	✓	✓	(0,0)

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<u>0</u>	<u>0</u>	8	7	3	✓	✓	(0,0)
<u>0</u>	8	<u>0</u>	-9	-5	✓	✗	
<u>0</u>	3.5	4.5	<u>0</u>	-0.5	✓	✗	
<u>0</u>	3	5	1	<u>0</u>	✓	✓	(0,3)
4	<u>0</u>	<u>0</u>	3	3	✓	✓	(4,0)
7	<u>0</u>	-6	<u>0</u>	3	✓	✗	
	<u>0</u>			<u>0</u>	✗	✗	
3	2	<u>0</u>	<u>0</u>	1	✓	✓	(3,2)
2.5	3	<u>0</u>	-1.5	<u>0</u>	✓	✗	
1	3	3	<u>0</u>	<u>0</u>	✓	✓	(1,3)

Basis and extreme points

$$\begin{array}{rclclclclcl}
 \text{s.t.} & 2x & + & y & + & e_1 & & & = & 8 \\
 & x & + & 2y & & & + & e_2 & = & 7 \\
 & & & y & & & & + & e_3 & = & 3 \\
 & x, & & y, & & e_1, & & e_2, & & e_3 & \geq & 0
 \end{array}$$

x	y	e ₁	e ₂	e ₃	basic solution	admiss.	extreme pt
<u>0</u>	<u>0</u>	8	7	3	✓	✓	(0,0)
<u>0</u>	8	<u>0</u>	-9	-5	✓	✗	
<u>0</u>	3.5	4.5	<u>0</u>	-0.5	✓	✗	
<u>0</u>	3	5	1	<u>0</u>	✓	✓	(0,3)
4	<u>0</u>	<u>0</u>	3	3	✓	✓	(4,0)
7	<u>0</u>	-6	<u>0</u>	3	✓	✗	
	<u>0</u>			<u>0</u>	✗	✗	
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1	3	3	<u>0</u>	<u>0</u>	✓	✓	(1,3)

{extreme points} \iff {feasible basic solutions}

Basis and extreme points

- Linear system $Ax = b$
- A format $m \times n$, rank $A = m \leq n$
- **Basis** of A : invertible submatrix $B(m \times m)$ of A
 $A = (B, N)$

$$(B, N) \begin{pmatrix} x_B \\ x_N \end{pmatrix} = b \quad \text{or} \quad Bx_B + Nx_N = b$$

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

- **Basic solution** associated to B :
 - $x_N = 0$ non-basic variables
 - $x_B = B^{-1}b$ basic variables

Basis and extreme points

Applications

Exercise sheet: Linear programming

-
- Exercice Bases *2

Basis and extreme points

Basis and basic solutions

$$\begin{cases} 2x + y + e_1 = 8 \\ x + 2y + e_2 = 7 \\ y + e_3 = 3 \\ x, y, e_1, e_2, e_3 \geq 0 \end{cases}$$

Initial basis ? $\{e_1, e_2, e_3\}$ for example:

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e_1, e_2, e_3 = basic variables, x, y = non-basic variables

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- number of extreme points \approx

Basis and extreme points

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- $\{\textbf{extreme points of the polyhedron}\} \iff \{\text{feasible basic solutions of the corresponding linear system}\}$
- number of extreme points $\approx C_n^m = \frac{n!}{m!(n-m)!}$
- basic degenerated solutions: some basic variables are zero
- if A is invertible: a single basic solution

Basis and extreme points

Neighboring basis and pivoting

Neighboring basis

Two neighboring vertices correspond to two bases B and B' such that a variable of B is replaced to obtain B'

Basis and extreme points

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Neighboring basis

Two neighboring vertices correspond to two bases B and B' such that a variable of B is replaced to obtain B'

- pass to a neighboring vertex: change basis (neighboring basis)

pivoting principle

Basis and extreme points

Which variable enters the basis?

Let's try with y : what is the max value for y ?

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► candidate for a new basis: $\{e_1, e_2, e_3\} \cup \{y\} \setminus \{e_3\} = \{e_1, e_2, y\}$

$(x, y, e_1, e_2, e_3) = (0, 3, 5, 1, 0)$

Plan

- 1 Introduction to Linear Programming
- 2 Geometrical interpretation
- 3 Basis and extreme points
- 4 The simplex algorithm**

The simplex algorithm

Towards a solution algorithm

► A naive solution method: enumerate all vertices, calculate f on these points, choose the vertex with optimal f :

- works: finite number of vertices
- limitation: this number can be very large in general...

The simplex algorithm (G. B. Dantzig 1947) iterative algorithm allowing to solve a linear program.

The simplex algorithm

Local improvement principle

From a current vertex, find a neighboring vertex that improves the objective.

Local improvement principle (maximization):

Let x_0 be a non optimal vertex. Then, there exists x , a **neighboring** vertex of x_0 , such that $f(x) > f(x_0)$.

The simplex algorithm

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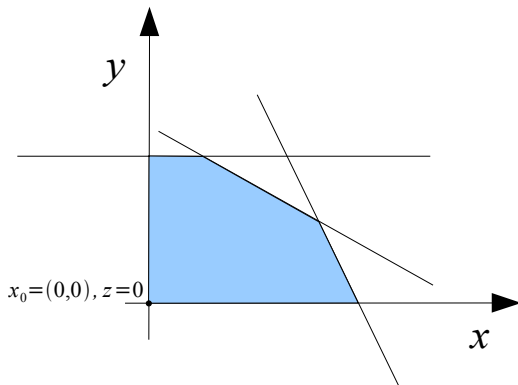
Remark: we change from a **continuous** problem (real variables) to a **discrete** problem (finite number of vertices)...

The simplex algorithm

2D Illustration: Cucumber and onions

$$x_0 = (0, 0), z = 0$$

$$z = 4x + 5y$$

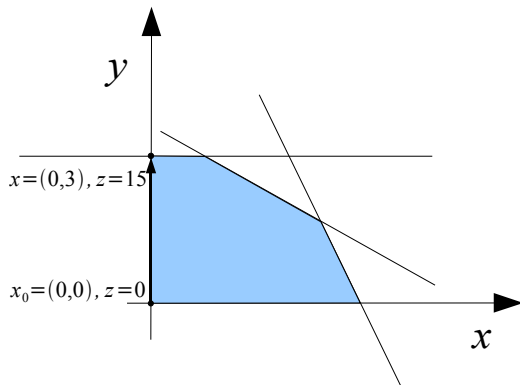


The simplex algorithm

2D Illustration: Cucumber and onions

$x_0 = (0, 0), z = 0 \rightarrow x = (0, 3), z = 15$

$$z = 4x + 5y$$



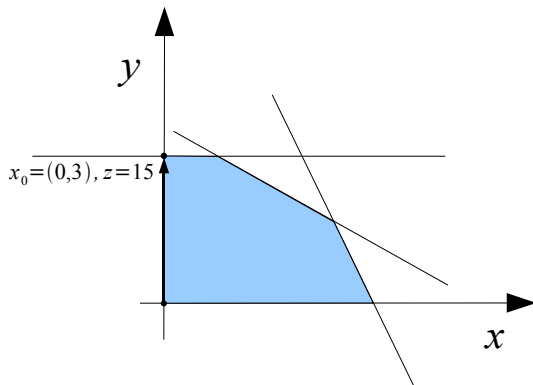
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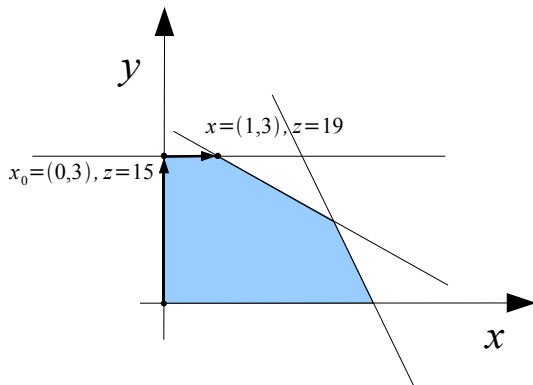
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$$x_0 = (0, 3), z = 15 \rightarrow x = (1, 3), z = 19$$



$$x_0 = (1, 3), z = 19 \rightarrow x = (3, 2), z = 22$$


The simplex algorithm

Concrete illustration

► Standardization:

Maximize $z = 4x + 5y$

$$\text{s.t. } \begin{cases} 2x + y \leq 8 \\ x + 2y \leq 7 \\ y \leq 3 \\ x, y \geq 0 \end{cases}$$

The simplex algorithm

Concrete illustration

► Standardization:

$$\begin{array}{ll}\text{Maximize } z = 4x + 5y \\ \text{s.t. } \begin{cases} 2x + y \leq 8 \\ x + 2y \leq 7 \\ y \leq 3 \\ x, y \geq 0 \end{cases}\end{array}$$

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Initial basis? $\{e_1, e_2, e_3\}$ for example:

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The simplex algorithm

Associated basic solution

- ▶ set the non-basic variables to 0
- ▶ We deduce:
 - value of the basic variables
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The simplex algorithm

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$$\text{here: } x = y = 0 \Rightarrow \begin{cases} e_1 = 8 - 2x - y = 8 \\ e_2 = 7 - x - 2y = 7 \\ e_3 = 3 - y = 3 \end{cases} \quad \text{and } z = 4x + 5y = 0$$

The simplex algorithm

Basis change

Essential observation: $z = 4x + 5y = 0 \Rightarrow$ we can improve z if x or y enters in the basis.

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The simplex algorithm

New basis $\{e_1, e_2, y\}$

$$\begin{cases} e_1 = 8 - 2x - y \\ e_2 = 7 - x - 2y \\ e_3 = 3 - y \end{cases} \Rightarrow \begin{cases} e_1 = 8 - 2x - y = 5 - 2x + e_3 \\ e_2 = 7 - x - 2y = 1 - x + 2e_3 \\ y = 3 - e_3 \end{cases}$$

The simplex algorithm

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Describe z with the non-basic variables

► $z = 4x + 5y = 15 + 4x - 5e_3$

The simplex algorithm

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Describe z with the non-basic variables

► $z = 4x + 5y = 15 + 4x - 5e_3$

Associated basic solution

$$x = e_3 = 0 \Rightarrow \begin{cases} e_1 = 5 - 2x + e_3 = 5 \\ e_2 = 1 - x + 2e_3 = 1 \\ y = 3 - e_3 = 3 \end{cases} \quad \text{and} \quad z = 15$$

The simplex algorithm

Iteration

$z = 15 + 4x - 5e_3$ can still increase if x enters the basis

The simplex algorithm

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If x enters, which variable leaves the basis?

Max value for x :

The simplex algorithm

Iteration

$z = 15 + 4x - 5e_3$ can still increase if x enters the basis

If x enters, which variable leaves the basis?

Max value for x :

- $e_1 = 5 - 2x + e_3 \geq 0 \Rightarrow x \leq 2.5$
- $e_2 = 1 - x + 2e_3 \geq 0 \Rightarrow x \leq 1$
- $y = 3 - e_3 \geq 0 \Rightarrow$ no constraint on x

The simplex algorithm

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Finally: $x_{\max} = 1$ and e_2 leaves the basis

New basis $\{e_1, y, x\}$

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$$\begin{cases} e_1 = 3 + 2e_2 - 3e_3 \\ x = 1 - e_2 + 2e_3 \\ y = 3 - e_3 \\ z = 19 - 4e_2 + 3e_3 \end{cases}$$

The simplex algorithm

Iteration (cont.)

$z = 19 - 4e_2 + 3e_3$ can still increase if e_3 enters the basis

The simplex algorithm

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The simplex algorithm

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The simplex algorithm

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Finally: $e_{3_{\max}} = 1$, e_1 leaves. New basis $\{e_3, y, x\}$:

The simplex algorithm

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Finally: $e_{3_{\max}} = 1$, e_1 leaves. New basis $\{e_3, y, x\}$:

$$\begin{cases} e_3 = 1 + 2/3e_2 - 1/3e_1 \\ x = 3 + 1/3e_2 - 2/3e_1 \\ y = 2 - 2/3e_2 + 1/3e_1 \\ z = 22 - 2e_2 - e_1 \end{cases}$$

The simplex algorithm

Termination

One has $z = 22 - 2e_2 - e_1$, hence $z^* \leq 22$

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but the basic solution $x = 3, y = 2, e_3 = 1$ leads to $z = 22$

► optimum

The simplex algorithm

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One has $z = 22 - 2e_2 - e_1$, hence $z^* \leq 22$

but the basic solution $x = 3, y = 2, e_3 = 1$ leads to $z = 22$

► optimum

The termination conditions concern the coefficients of z expressed with the non-basic variables

The simplex algorithm

Reduced costs

B , a basis of $Ax = b$

the objective function

$$\begin{aligned}z &= CX = c_B x_B + c_N x_N \\&= c_B B^{-1} b - (c_B B^{-1} N - c_N) x_N \\&= z_0 - \sum_{j=1}^n (c_B B^{-1} a^j - c_j) x_j \\&= z_0 - \sum_{j=1}^n (z_j - c_j) x_j\end{aligned}$$

$z_j - c_j = c_B B^{-1} a^j - c_j$ are the reduced costs of the non-basic variable x_j

The simplex algorithm

At each iteration

		x_N	x_B
z	$= z_0$	reduced costs	0

$$x_B = \oplus \quad \dots$$

For the optimum

		x_N	x_B
z	$= z_0^*$	\ominus	0

$$x_B = \oplus \quad \dots$$

The simplex algorithm

Heuristic: the variable with the higher coefficient enters the basis

Which variable leaves the basis?

Minimal quotient principle

pivot column x_1 right member ≥ 0 quotient

$a_1 \leq 0$	b_1	-
$a_2 > 0$	b_2	$\frac{b_2}{a_2}$
$a_3 > 0$	b_3	$\frac{b_3}{a_3}$
$a_4 = 0$	b_4	-

$$\text{row } r \quad \frac{b_r}{a_r} = \min \left\{ \frac{b_i}{a_i} \mid a_i > 0 \right\}$$

The simplex algorithm

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$a_4 = 0$	b_4	-

$$\text{row } r \quad \frac{b_r}{a_r} = \min \left\{ \frac{b_i}{a_i} \mid a_i > 0 \right\}$$

The simplex algorithm

Pivoting

$$\begin{array}{ccc}
 & c & \cdots & a \\
 & \vdots & & \vdots \\
 \text{pivot row} - & p & \cdots & b \\
 & | & & \\
 & \text{pivot} & & \\
 & \text{column} & &
 \end{array}
 \implies a \rightarrow a - \frac{b}{p}c$$

The simplex algorithm

Phase II

Input: a linear program and a feasible basic solution

Output: an optimal feasible solution or declare "non-bounded linear program"

- ① Choose an entering column (variable)
 - choose a non-basic variable x_j (column) with a negative reduced cost
 - if there is no entering variable: STOP, the basic solution is optimal
- ② Choose a leaving row (variable)
 - choose a row r minimizing the quotient
 - if there is no leaving row: STOP the current linear program is unbounded
- ③ Update the basis and the program
 - pivot around a_{rj} and go to (1)

The simplex algorithm

- A basic solution is degenerated if at least one basic variable is zero (in this case, there isn't a bijection between the feasible basic solutions and the extreme points)
- If all basic solutions are non degenerated, then the simplex algorithm terminates after a finite number of iterations

The simplex algorithm

Phase I

- Exercice on Chamilo