

# Foundations of Operations Research

Master of Science in Computer Engineering

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Lesson 10: Introduction to Linear Programming

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# A diet model

*In 1945, George J. Stigler studied the problem of determining the minimum cost diet for the soldiers of the United States Army.*

*Given a set of 77 available foods, he wanted to compute the yearly amount of each food which would have provided the adequate quantity of 9 basic nutrients (calories, proteins, minerals and vitamins).*

*Among the feasible solutions, he wanted to compute one with the minimum total cost.*

*At the time, no algorithm existed for this problem.*

*He devised a heuristic method, which required 10 individuals working for 12 days on hand-operated desk calculators.*

*They found a solution costing 39.93 Dollars/year (the optimum is 39.69).*



# The general diet problem

In all **diet problems**, we have:

- a **set of foods**  $F$
- a **set of nutrients**  $N$
- a **unitary cost**  $c_j$  for each food  $j \in F$
- a **required amount**  $r_i$  for each nutrient  $i \in N$
- a **content**  $a_{ij}$  of each nutrient  $i \in N$  in each food  $j \in F$

The natural decision variables are the amounts  $x_j$  for each food  $j \in F$

The general model for the diet problem is therefore:

$$\begin{aligned} \min f &= \sum_{j \in F} c_j x_j \\ \sum_{j \in F} a_{ij} x_j &\geq r_i && i \in N \\ x_j &\geq 0 && j \in F \end{aligned}$$

# A transportation model

*A freight transportation company must carry the production of a set  $M$  of factories to a set  $N$  of customers.*

*The productive capacity  $p_i$  of each factory  $i \in M$  and the demand  $d_j$  of each customer  $j \in N$  are known.*

*A maximum amount  $q_{ij}$  can be transported from each factory  $i \in M$  to each customer  $j \in N$  at a unitary cost equal to  $c_{ij}$  Euros/unit.*

We adopt as decision variables  $x_{ij}$  the number of product units transported from factory  $i \in M$  to customer  $j \in N$

The general model for the transportation problem is:

$$\min f = \sum_{i \in M} \sum_{j \in N} c_{ij} x_{ij}$$

$$\sum_{j \in N} x_{ij} \leq p_i \quad i \in M$$

$$\sum_{i \in M} x_{ij} \geq d_j \quad j \in N$$

$$0 \leq x_{ij} \leq q_{ij} \quad i \in M, j \in N$$

# Linear Programming: fundamental assumptions

All these models fall into the framework of **Linear Programming** (*LP*)

They can be immediately recognized from the following assumptions:

- ① **divisibility**: the decision variables can assume any fractional value
- ② **linearity**: the objective and the constraint functions satisfy
  - **proportionality**: each decision variable  $x_j$  gives a contribution proportional to its value:  $c_j x_j$
  - **additivity**: the contributions of the decision variables are summed

$$\sum_{j \in N} c_j x_j$$

except for a suitable constant term (all functions involved are **affine**)

$$\sum_{j \in N} c_j x_j + d$$

# Algebraic and matrix form

Since

- the objective function  $f(x)$  is affine:  $f(x) = \sum_{j=1}^n c_j x_j + d$
- the constraint functions  $g_i(x)$  are affine:  $g_i(x) = \sum_{j=1}^n a_{ij} x_j + b_i$

the general mathematical programming form reduces to:

$$\begin{aligned} \min f &= \sum_{j=1}^n c_j x_j + d \\ \sum_{j=1}^n a_{ij} x_j + b_i &\leq 0 \quad i = 1, \dots, m \end{aligned}$$

or to the compact **matrix form** with a **decision variable vector**  $x \in \mathbb{R}^n$

$$\begin{aligned} \min f &= c^T x + d \\ Ax + b &\leq 0 \end{aligned}$$

where  $c \in \mathbb{R}^n$ ,  $d \in \mathbb{R}$ ,  $A \in \mathbb{R}^{m,n}$  and  $b \in \mathbb{R}^m$

# Standard form

However, LP problems are commonly represented in the **standard form**:

$$\begin{aligned} \min f &= \sum_{j \in N} c_j x_j \\ \sum_{j \in N} a_{ij} x_j &= b_i && i \in N \\ x_j &\geq 0 && j \in N \end{aligned}$$

where

- the **objective function  $f$**  is **minimized**
- all **constraints** are **equalities**
- all **decision variables** are **nonnegative**

*What if the problem has not this form?*

# Reduction into standard form

Any LP problem can be reduced to an equivalent problem in standard form applying the following transformations

- 1 maximization is turned into minimization reverting the objective's sign

$$\max f(x) = c^T x \Leftrightarrow \min \hat{f}(x) = \hat{c}^T x \text{ with } \hat{c} = -c$$

- 2 inequalities are turned into equalities introducing nonnegative variables
  - a **slack variable**  $s_i$  for each  $\leq$  inequality

$$a_i^T x \leq b_i \Leftrightarrow \begin{cases} a_i^T x + s_i = b_i \\ s_i \geq 0 \end{cases}$$

- a **surplus variable**  $s_i$  for each  $\geq$  inequality

$$a_i^T x \geq b_i \Leftrightarrow \begin{cases} a_i^T x - s_i = b_i \\ s_i \geq 0 \end{cases}$$

- 3 unrestricted variables are replaced by nonnegative ones  
(each unrestricted variable is the difference of two nonnegative variables)

$$x_j \in \mathbb{R} \Leftrightarrow x_j \rightarrow x_j^+ - x_j^- \text{ with } x_j^+ \geq 0 \text{ and } x_j^- \geq 0$$

(alternatively, use an equality constraint to determine the expression of  $x_j$  in terms of the other variables and replace  $x_j$  with that expression)



# Example

Reduce the following *LP* problem into standard form:

$$\begin{aligned}\max f &= 2x_1 - 3x_2 \\ 4x_1 - 7x_2 &\leq 5 \\ 6x_1 - 2x_2 &\geq 4 \\ x_1 &\geq 0 \\ x_2 &\in \mathbb{R}\end{aligned}$$

1) Turn maximization into minimization:  $\hat{f} = -f = -2x_1 + 3x_2$

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3x_2 \\ 4x_1 - 7x_2 &\leq 5 \\ 6x_1 - 2x_2 &\geq 4 \\ x_1 &\geq 0 \\ x_2 &\in \mathbb{R}\end{aligned}$$

# Example

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3x_2 \\ 4x_1 - 7x_2 &\leq 5 \\ 6x_1 - 2x_2 &\geq 4 \\ x_1 &\geq 0 \\ x_2 &\in \mathbb{R}\end{aligned}$$

- 2) Turn the inequalities into equalities introducing a new slack variable  $s_1$  ( $x_3$ ) and a new surplus variable  $s_2$  ( $x_4$ )

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3x_2 \\ 4x_1 - 7x_2 + s_1 &= 5 \\ 6x_1 - 2x_2 - s_2 &= 4 \\ x_1 &\geq 0 \\ x_2 &\in \mathbb{R} \\ s_1 &\geq 0 \\ s_2 &\geq 0\end{aligned}$$

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3x_2 \\ 4x_1 - 7x_2 + x_3 &= 5 \\ 6x_1 - 2x_2 - x_4 &= 4 \\ x_1, x_3, x_4 &\geq 0 \\ x_2 &\in \mathbb{R}\end{aligned}$$

# Example

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3x_2 \\ 4x_1 - 7x_2 + x_3 &= 5 \\ 6x_1 - 2x_2 + x_4 &= 4 \\ x_1, x_3, x_4 &\geq 0 \\ x_2 &\in \mathbb{R}\end{aligned}$$

3a) Turn the unrestricted variable into a difference of nonnegative variables:

$$x_2 = x_2^+ - x_2^- = x_5 - x_6 \quad (x_5 = x_2^+ \text{ and } x_6 = x_2^-)$$

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3(x_2^+ - x_2^-) \\ 4x_1 - 7(x_2^+ - x_2^-) + x_3 &= 5 \\ 6x_1 - 2(x_2^+ - x_2^-) - x_4 &= 4 \\ x_1, x_3, x_4 &\geq 0 \\ x_2^+, x_2^- &\geq 0\end{aligned}$$

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3x_5 - 3x_6 \\ 4x_1 + x_3 - 7x_5 + 7x_6 &= 5 \\ 6x_1 - x_4 - 2x_5 + 2x_6 &= 4 \\ x_1, x_3, x_4, x_5, x_6 &\geq 0\end{aligned}$$

# Example

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3x_2 \\ 4x_1 - 7x_2 + x_3 &= 5 \\ 6x_1 - 2x_2 + x_4 &= 4 \\ x_1, x_3, x_4 &\geq 0 \\ x_2 &\in \mathbb{R}\end{aligned}$$

3b) Alternatively, replace the unrestricted variables with an expression derived from an equality (for example, the second:  $x_2 = 3x_1 - x_4/2 - 2$ ):

$$\begin{aligned}\min \hat{f} &= -2x_1 + 3 \cdot (3x_1 - x_4/2 - 2) & \min \hat{f} &= 7x_1 - 3/2 x_4 - 6 \\ 4x_1 - 7 \cdot (3x_1 - x_4/2 - 2) + x_3 &= 5 & -17x_1 + x_3 + 7/2 x_4 &= -9 \\ x_1, x_3, x_4 &\geq 0 & x_1, x_3, x_4 &\geq 0\end{aligned}$$

*The first equality provides a different, but equivalent, standard form*

# Other transformations

We shall see that any  $LP$  problem also admits other special forms, which allow only  $\leq$  inequalities or only  $\geq$  inequalities

They can be obtained, as well, by simple algebraic transformations

$\leq$ form	= form	$\geq$ form
$ax \leq b$	$\begin{cases} ax + s = b \\ s \geq 0 \end{cases}$	$-ax \geq -b$
$\begin{cases} ax \leq b \\ -ax \leq -b \end{cases}$	$ax = b$	$\begin{cases} ax \geq b \\ -ax \geq -b \end{cases}$
$-ax \leq -b$	$\begin{cases} ax - s = b \\ s \geq 0 \end{cases}$	$ax \geq b$

Two  $LP$  problems are **equivalent** when **corresponding solutions have**

- the same **feasibility status**: both feasible or both unfeasible
- the same **objective value**, possibly except for a constant offset