

Chapitre 01: Introduction, simple models

Requirements of an LP problem:

- LP problems seek to maximize or minimize a quantity expressed as an objective function
- Real-world restrictions or constraints limit how much the objective can be achieved
- Both the objective function and constraints must be written as linear equations or inequalities

Definition:

LP is a mathematical method to efficiently allocate limited resources and determine the optimal solution by maximizing or minimizing the objective function.

Components of LP:

- A set of decision variables
- An objective function
- A set of constraints

Essentials of LP model:

- Limited resources: limited number of labor, material equipment and finance
- Objective: refers to the optimization aim (maximize profits or minimize costs)
- Linearity: the increase of labor input will cause a proportionate increase of output
- Homogeneity: products, workers, efficiency and machines are assumed to be identical
- Divisibility: it is assumed that resources can be divided into fractions

LP Models:

- Blending models (diet problem)
- Production model
- Transportation model

Canonic LP model:

• Maximization canonic model:

- Objective function is maximization
- Constraints are "less or equal" inequalities
- Variables are non-negative
($x \geq 0$)

• Minimization canonic model:

- Objective function is minimization
- Constraints are "greater or equal" inequalities
- Variables are non-negative
($x \geq 0$)

• Mixed cononic model:

- Objective function is min or max
- Constraints are mixed ($<=$ or $>=$ or $=$)
- Variables can include unrestricted or non-negative variables.

LP standard form:

It is a specific, constrained format of an LP and is commonly used for computational purposes

- Objective Function: is min or max
- Constraints: all constraints must be equalities
- Variable restrictions: All variables must be non-negative
($x \geq 0$)

Transformation rules:

- Objective function:

$$\text{Min}(Z) = \text{Max}(-Z)$$

- Constraints:

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

$$ax \geq b \Leftrightarrow \begin{cases} ax - s = b \\ s \geq 0 \end{cases}$$

- Variables:

$$x \leq 0 \Leftrightarrow x' = -x, x' \geq 0$$

$$x \in \mathbb{R} \Leftrightarrow x = x' - x'' \\ x', x'' \geq 0$$

Remarque:

If we want to transform into a cononic form and we have an equality:

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

Chapitre 02: Geometry Graphic method

Convex:

A convex set is a region in which, for any two points within the set, the straight line segment connecting them lies entirely within the set

Feasible region:

Feasible region or feasible set or solution space is the set of all possible points of an optimization problem that satisfy the problem's constraints

Theorems:

- The feasible region of an LP is a convex
- if the feasible region is non-empty and bounded, then there is an optimal solution

Corner point:

also called an extreme point of the feasible region is a point that is not the middle point of two other points of the feasible region

Theorems:

- If every variable is non-negative, and if the feasible region is non-empty then there is a corner point
- In 2D, a corner point is at the intersection of two equality constraints
- Corner points illustrate optimality
- if a feasible region has a corner point and it has an optimal solution then there is an optimal solution that

is a corner point

Graphical solution:

- Solution: A set of real values of $x = (x_1, x_2, \dots)$ which satisfies the constraints.
- Feasible solution: a set of real values of $x = (x_1, x_2, \dots, x_n)$ which satisfies the constraints and satisfies the non negativity restriction $x_j \geq 0$
- Feasible region: The collection of all the feasible solutions, is called feasible region

Solution case:

- Case 01: "Unique Solution"
One optimal solution, objective line touches the feasible region at one point
- Case 02: "Multiple solutions"
Infinite solutions along a line segment. Objective line is parallel to one edge of the feasible region

- Case 03: "No solution (infeasible)"

No feasible solution, the constraints define a region with no overlapping points (no feasible region)

- Case 04: "Unbounded"

Feasible region extends infinitely in the direction of optimization.

Chapitre 03: solving, simplex method

Simplex for Maximization:

- 1) All inequalities must be \leq
- 2) Convert to standard form:
transform inequalities into equations using slack variables (RHS must be positive)
- 3) Standard objective functions:
 $Z = C_1x_1 + C_2x_2 + \dots + C_nx_n$
 becomes:
 $Z - C_1x_1 - C_2x_2 - \dots - C_nx_n = 0$

- 4) Create the initial table To:

- Each constraint forms a row. Coefficients of variables in each constraint go into the corresponding columns.
- The RHS of each constraint goes in the solution column.

- Example:

Basic	Z	x_1	s_1	s_2	Solution
s_1	0	1	1	0	40
s_2	0	4	0	1	120
Z	1	-40	0	0	0

- 5) Pivot Column:

Identify the most negative coefficient in the (Z-row)

- 6) Pivot Row:

Divide the RHS (solution column) by the corresponding positive values in the in the pivot column and we choose the smallest positive value.

- 7) Pivot:

The pivot is the intersection of the pivot row and column.

8) Create the tables T_1 :

- Divide the pivot row by the pivot (pivot core becomes 1)
- All the values in the pivot column becomes 0 (except pivot)
- The rest of the values are filled using pivot operations:

$$a = a_{T_0} - \frac{b \times c}{p}$$

a = The new value

a_{T_0} = The value in the previous Table

b = The projection of a_{T_0} on the pivot column

c = The projection of a_{T_0} on the pivot row

p = pivot

- 9) Repeat: repeat steps 5 to 8 until all Z-row coefficients are non-negative

10) Solution: The RHS (solution column) gives the solution.

Non basic variables are set to 0.

Remark:

- The leaving variable in each table is the one of the pivot row
- The entering variable is the one of the pivot column.

Big M methode!

1) Standard form: (RHS > 0)

- For equal constraint:
 - Add Artificial variable.
- For " $>$ " constraint:
 - Add surplus variable
 - Add Artificial variable
- For " \leq " constraint:
 - Add slack variable

2) Standard objective function:

- Add a large penalty (M) for artificial variables in the objective function.

$$Z = c_1x_1 + c_2x_2 + \dots + c_nx_n - M_1a_1 - M_2a_2 - \dots - M_na_n + 0s_1 + 0s_2 + \dots + 0s_n$$

3) Create the initial table T_0 :

- C_j : The coefficients in the objective function
- X_B : corresponds to the RHS of the equations
- B : Basic variables
- C_B : Coefficients of basic
- Z_j : Sum of products of C_B and X_i coefficients

$$Z_j = \sum C_B \times x_i$$

4) Pivot column:

- Identify the most negative value in the $(Z_j - C_j)$ row.
- The variable of the pivot column is the entering variable.

5) Pivot row:

- We divide the X_B column (solution column) by the correspondent positive value in the pivot column and we choose

the smallest positive value.

- The variable of the pivot row is the leaving variable.

6) Pivot:

The pivot is the intersection of the pivot row and column.

7) Create table T_1 :

- The column of the variable leaving the basis can be removed since it's no longer contributes to the solution
- Divide the pivot row by the pivot

- All values in the pivot column becomes 0 (except pivot)

- The rest of the values are filled using pivot operation

8) Repeat: repeat steps 4 to 7 until all $(Z_j - C_j)$ row values are non-negative

9) Solution: The X_B column gives the solution.

Two Phase method

Phase 01:

- Standard Form: we convert the inequalities into equations using slack, surplus and artificial variables the same way in the Big M method

- Define the auxiliary objective function:

This function only contains the artificial variables

$$Z' = -M_1 A_1 - M_2 A_2 - \dots - M_n A_n$$

- Constructing the table T_{01}

We use the auxiliary objective function instead of the objective function

The rest of the table remains the same as in the Big M method.

- Solving for Z' :

we solve for Z' in the same way in Big M method

- End of the phase:

To be able to move to phase two there are 3 conditions:

- The value of $Z' = 0$
- All $Z_j - C_j$ are non-negative
- All the artificial variables are removed from the table

Phase 2:

- Constructing the initial table:

We use the original objective function (without artificial variables), while keeping the rest values from the last table of phase 1

• Solving for Z:

we solve the same way we did in Big M method, we stop when all the values in $(Z_j - C_j)$ row are non-negative

Special case:

Degeneracy:

• Detection in table:

When determining the pivot row, two or more rows have the same minimum ratio $(\text{solution} / \text{pivot column})$.

• Interpretation:

This can result in degeneracy where the program might cycle without any progress

• Resolution:

- Choose any row arbitrarily from the tied rows
- Use techniques like Bland's Rule to prevent cycling.

Multiple Optimal Solutions:

• Detection in table:

When a non-basis variable takes a 0 value in the Z-row (or $Z_j - C_j$ row) in the optimal table.

• Interpretation:

The objective function is parallel to one of the constraints and the PL has multiple optimal solutions (alternative solutions)

Unbounded:

• Detection in table:

When selecting a pivot row, all ratios $(\text{solution} / \text{pivot column})$ are non-positive or undefined

• Interpretation:

The objective function can increase (in maximization) or decrease (in minimization) indefinitely without violating constraints which make the LP unbounded

Infeasibility:

• Detection in table:

At the end table (or the last table in phase I in the two phase method) if artificial variable remains in the basis (in the two phase method if the artificial variables are not removed) with a non-zero value.

• Interpretation:

The problem is infeasible (has no feasible solution) as the artificial variables represent a violation of the original constraints