

Lecture 1

Linear Programming: Formulation and Solution

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Example

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You plan to invest in three companies. Expected returns from investment in companies 1, 2, and 3 are $r_1 = 0.05$, $r_2 = 0.12$, and $r_3 = 0.09$ respectively. For strategic reasons, you do not want to invest more than 40% your budget in company 1, 50% in company 2, and at least 30% in company 3. Determine the percentage of your budget to be invested in each company to maximize your expected return?

- variables: x_1, x_2, x_3 be the fractions invested in three companies.

- objective function:

$$\max_{x_1, x_2, x_3} \{0.05x_1 + 0.12x_2 + 0.09x_3\}$$

Handwritten notes: $x_1 \geq 0$, $x_2 \geq 0$, $x_3 \geq 0$ (circled), $0.04 \times x_3$, $\times 0.09$

- constraints:

$$x_1 + x_2 + x_3 \leq 1,$$

$$x_1 \leq 0.4, \quad x_2 \leq 0.5, \quad x_3 \geq 0.3,$$

$$x_1 \geq 0, \quad x_2 \geq 0, \quad x_3 \geq 0.$$

Handwritten note: ← avoid shorting the stock

More general formulation

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You plan to invest in n different assets with the expected returns r_i ($i = 1, \dots, n$). You are required to invest at least a fraction of s_i and at most S_i of your budget in asset i ($i = 1, \dots, n$). Determine an investment solution to maximize the total expected returns.

x_i : fraction of the budget in excess of s_i invested in asset i (the actual fraction is $x_i + s_i$), $i = 1, \dots, n$.

let

$$\Delta s_i = S_i - s_i, \quad i = 1, \dots, n, \quad \bar{s} = \sum_{i=1}^n s_i, \quad \bar{R} = \sum_{i=1}^n r_i s_i.$$

$$\max_{x_1, \dots, x_n} \{r_1 x_1 + \dots + r_n x_n\} + \bar{R}$$

subject to:

$$x_1 \leq \Delta s_1, \quad x_2 \leq \Delta s_2, \quad \dots, \quad x_n \leq \Delta s_n,$$

$$x_1 + \dots + x_n \leq 1 - \bar{s}, \quad \#$$

$$x_1 \geq 0, \quad x_2 \geq 0, \quad \dots, \quad x_n \geq 0.$$

Definition of Linear Programming Problem

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Linear Programming (LP) is an optimization model in which

- all decision variables take real values,
- the objective is a linear function of decision variables,
- all constraints are linear (in)equalities of decision variables.

$$\begin{aligned} & \max_{x_1, \dots, x_n} \{r_1 x_1 + \dots + r_n x_n\} + \bar{R} \\ \text{subject to: } & x_1 \leq \Delta s_1, x_2 \leq \Delta s_2, \dots, x_n \leq \Delta s_n, \\ & x_1 + \dots + x_n \leq 1 - \bar{s}, \\ & x_1 \geq 0, x_2 \geq 0, \dots, x_n \geq 0. \end{aligned}$$

n decision variables: x_1, \dots, x_n .

$2n + 2$ parameters: \bar{R} , \bar{s} , (r_1, \dots, r_n) and $(\Delta s_1, \dots, \Delta s_n)$.

$(n + 1)$ "regular" constraints and n nonnegativity constraints.

Standard Form

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LP in standard form:

- 1 Maximize an objective.
- 2 Left Hand Side (LHS) of each constraint, which is a linear combination of decision variables, is less than or equal to its Right Hand Side (RHS), which is a constant.
- 3 All decision variables are positive.

LP formulation with n decision variables and $m = n + k$ constraints

$$\begin{aligned} & \max_{x_1, \dots, x_n} \{c_1 x_1 + \dots + c_n x_n\} \\ \text{subject to: } & a_{11} x_1 + \dots + a_{1n} x_n \leq b_1, \\ & \dots\dots\dots \\ & a_{k1} x_1 + \dots + a_{kn} x_n \leq b_k, \\ & x_1 \geq 0, \dots, x_n \geq 0. \end{aligned}$$

Handwritten notes: "rows" with an arrow pointing to the constraint coefficients, and a checkmark next to the RHS term b_1 .

Transform an LP into Its Standard Form

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$$\begin{aligned} & \max_{x_1, \dots, x_n} \{c_1 x_1 + \dots + c_n x_n\} \\ \text{subject to: } & \underline{a_{j1}x_1 + \dots + a_{jn}x_n} \leq \underline{b_j} \quad j = 1, \dots, k \\ & x_1 \geq 0, \dots, x_n \geq 0. \end{aligned}$$

LP Standard form

how to fit following situations into the standard LP form?

① • the objective is $\min_{x_1, \dots, x_n} \{c_1 x_1 + \dots + c_n x_n\}$? $\min = 0 \max$

$$\max_{x_1, \dots, x_n} \{-c_1 x_1 - \dots - c_n x_n\}$$

$\times (-1)$ \Rightarrow use initial value $\times (-1)$

② • constraint $a_{j1}x_1 + \dots + a_{jn}x_n \geq b_j$?

$$-a_{j1}x_1 - \dots - a_{jn}x_n \leq -b_j$$

③ • constraint $a_{j1}x_1 + \dots + a_{jn}x_n = b_j$?

$$a_{j1}x_1 + \dots + a_{jn}x_n \leq b_j \quad \text{and} \quad a_{j1}x_1 + \dots + a_{jn}x_n \geq b_j$$

i.e., $a_{j1}x_1 + \dots + a_{jn}x_n \leq b_j \quad \text{and} \quad -a_{j1}x_1 - \dots - a_{jn}x_n \leq -b_j$

④ • a decision variable x can be negative? $x \text{ can } < 0$

$$x = y_1 - y_2, \quad y_1 \geq 0, y_2 \geq 0.$$

Another Standard Formulation

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Dual of the maximization problem (to be discussed later in the course):

① instead of

$$\max_{x_1, \dots, x_n} \{c_1 x_1 + \dots + c_n x_n\}$$

write the objective function as

$$\min_{y_1, \dots, y_n} \{h_1 y_1 + \dots + h_n y_n\}$$

② instead of

$$a_{j1}x_1 + \dots + a_{jn}x_n \leq b_j$$

make the LHS of constraints greater than or equal to the RHS:

$$a_{11}y_1 + \dots + a_{n1}y_n \geq d_1,$$

③ all decision variables are positive.

$$y_1 \geq 0, \dots, y_n \geq 0.$$

Linear Programming (LP) in Standard Form

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$$\begin{array}{ll}\max_{x_1, \dots, x_n} & \{c_1x_1 + \dots + c_nx_n\} \\ \text{subject to} & a_{11}x_1 + \dots + a_{1n}x_n \leq b_1, \\ & \dots\dots\dots \\ & a_{k1}x_1 + \dots + a_{kn}x_n \leq b_k, \\ & x_1 \geq 0, \dots, x_n \geq 0.\end{array}$$

- all decision variables take non-negative real values;
- the objective is to maximize a linear function of decision variables;
- all constraints are linear inequalities of decision variables;
variables are on the LHS and the constant is on the RHS
 $\text{LHS} \leq \text{RHS}.$

Why Solving LP Needs a New Approach

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Basic calculus, we know how to solve a (simple) optimization problem like

$$\max_{x_1, x_2} \left\{ g(x_1, x_2) = 6x_1 - x_1^2 + x_2 - 2x_2^2 \right\},$$

by the use of the first-order condition

$$\frac{\partial g}{\partial x_1} = 6 - 2x_1 = 0 \quad \text{and} \quad \frac{\partial g}{\partial x_2} = 1 - 4x_2 = 0,$$

the optimal solution $x_1^* = 3$, $x_2^* = 1/4$ and $g(x_1^*, x_2^*) = 9\frac{1}{8}$.

Try this approach on a linear objective function, e.g.,

$$g(x_1, x_2) = 6x_1 + x_2?$$

what happened, and why?

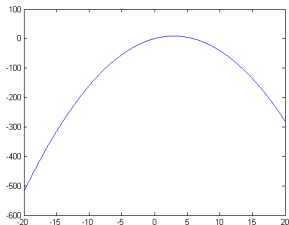
Linear versus Nonlinear Optimization

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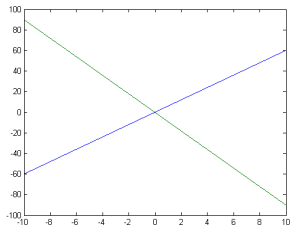
nonlinear function:

$$g(x) = 6x - x^2$$



linear function:

$$g(x) = 6x \text{ or } g(x) = -9x$$



implications of the above comparisons:

- a meaningful LP problem must have constraints.
- optimal solution at the boundary of some constraints.

"Push to the Boundary" in the 2-Dimensional Case

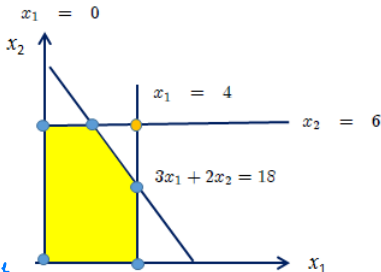
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$$\begin{array}{ll}\max & \{2x_1 + x_2\} \\ \text{subject to} & x_1 \leq 4, \\ & x_2 \leq 6, \\ & 3x_1 + 2x_2 \leq 18, \\ & x_1 \geq 0, x_2 \geq 0.\end{array}$$

how to guarantee that there are more

constant than variable. \rightarrow write in std formatted will guarantee $x_2 = 0$



"Corner Feasible Point" (CFP) solutions

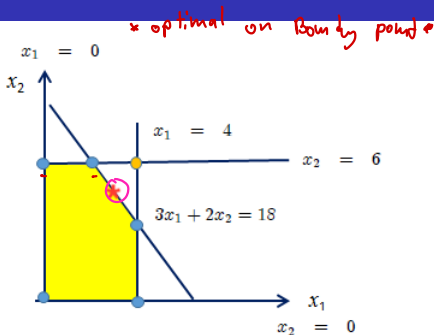
- $(x_1, x_2) = (0, 0) \rightarrow 2x_1 + x_2 = 0 \quad (x_1 = 0, x_2 = 0).$
- $(x_1, x_2) = (0, 6) \rightarrow 2x_1 + x_2 = 6 \quad (x_1 = 0, x_2 = 6).$
- $(x_1, x_2) = (2, 6) \rightarrow 2x_1 + x_2 = 10 \quad (3x_1 + 2x_2 = 18, x_2 = 6).$
- $(x_1, x_2) = (4, 0) \rightarrow 2x_1 + x_2 = 8 \quad (x_1 = 4, x_2 = 0).$
- $(x_1, x_2) = (4, 3) \rightarrow 2x_1 + x_2 = 11 \quad (x_1 = 4, 3x_1 + 2x_2 = 18).$

What about Other Points at the "Boundary"

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$$\begin{array}{ll}\max & \{2x_1 + x_2\} \\ \text{subject to} & x_1 \leq 4, \\ & x_2 \leq 6, \\ & 3x_1 + 2x_2 \leq 18, \\ & x_1 \geq 0, x_2 \geq 0.\end{array}$$



Let point * in the figure be $\mathbf{x} = (x_1, x_2)$.

$$\mathbf{x} = \alpha * (2, 6) + (1 - \alpha) * (4, 3) \quad \text{i.e.} \quad x_1 = 2\alpha + 4(1 - \alpha), \quad x_2 = 6\alpha + 3(1 - \alpha)$$

and

weights avg of 2 points

$$2x_1 + x_2 = \alpha(2 * 2 + 6) + (1 - \alpha) * (2 * 4 + 3) \leq \max\{2 * 2 + 6, 2 * 4 + 3\}.$$

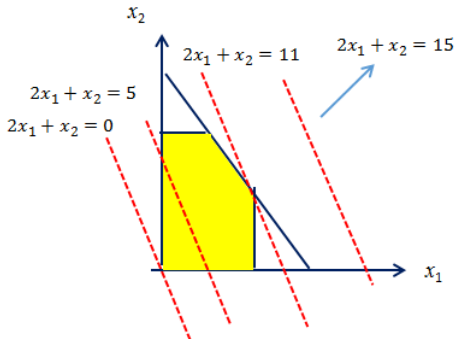
so the objective value at * cannot be better than both CFP solutions.

A Fundamental Fact about LP

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If the optimal objective value of an LP exists and is finite, then this value can always be obtained at one of the “corner points” (solutions to n simultaneous equations).

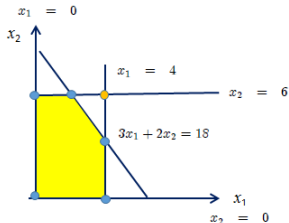


Systems of Equations and Feasible Solutions

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$$\begin{array}{ll}\max & \{2x_1 + x_2\} \\ \text{subject to} & x_1 \leq 4, \\ & x_2 \leq 6, \\ & 3x_1 + 2x_2 \leq 18, \\ & x_1 \geq 0, x_2 \geq 0.\end{array}$$



observation: to solve for $n = 2$ variables, we need a pair of equations.

question: with 3 + 2 (regular+non-negativity) constraints, how many pairs of equations can we possibly have? 10.

question: why we only get five pairs of equations?

- some pairs of equations have no solution: e.g.,

$$x_1 = 0 \text{ and } x_1 = 4.$$

- some solutions violate other constraints: e.g.,

$$x_1 = 4 \text{ and } x_2 = 6 \text{ or } x_1 = 0 \text{ and } 3x_1 + 2x_2 = 18.$$

Possible Procedure

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select n out of $n + m$ (m regular, n non-negativity) constraints to form a system of equations

- if the system has no solution, next!
- if the solution does not satisfy another constraint, next!
- if the solution is feasible for all constraints, evaluate the corresponding objective value.
- compare all objective values to pick the best solution.

Question: how many systems of equations we need to evaluate?

- $n = 2$ and $m + n = 10$?

$$\frac{10 \times 9}{2} = 45$$

- $n = 20$ and $m + n = 30$?

$$\frac{30 \times 29 \times \dots \times 11}{20!} = \text{huge (about 30 million)}$$

- $n = 10k$ and $m + n = 30k$?

don't even think about it!

Searching Optimal Solutions

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$$\begin{array}{ll}\max & \{2x_1 + x_2\} \\ \text{subject to} & x_1 \leq 4, \\ & x_2 \leq 6, \\ & 3x_1 + 2x_2 \leq 18, \\ & x_1 \geq 0, x_2 \geq 0.\end{array}$$

start from 1 corner point and then move to the next one

$$x_1 = 0, x_2 = 0, (2x_1 + x_2 = 0):$$

$$\text{neighbors: } \underline{x_1 = 4}, x_2 = 0 \rightarrow 2x_1 + x_2 = \underline{8}$$

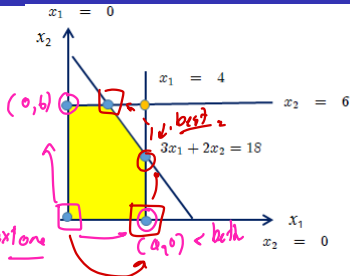
$$\underline{x_1 = 0}, x_2 = 6 \rightarrow 2x_1 + x_2 = 6.$$

$$x_1 = 4, x_2 = 0, (2x_1 + x_2 = 8):$$

$$\text{new neighbor: } x_1 = 4, 3x_1 + 2x_2 = 18 \rightarrow (x_1, x_2) = \underline{(4, 3)} \rightarrow 2x_1 + x_2 = \underline{11}.$$

$$x_1 = 4, x_2 = 3, (2x_1 + x_2 = 11):$$

$$\text{new neighbor: } 3x_1 + 2x_2 = 18, x_2 = 6 \rightarrow (x_1, x_2) = (2, 6) \rightarrow 2x_1 + x_2 = 10.$$



General Idea of SIMPLEX Method

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- ① **Fact 1:** The objective function of an LP is optimized by solutions to one of n simultaneous equations.
- ② **Fact 2:** If a solution is not optimal, then it is always possible to get a better solution by swapping in/out one equation.

SIMPLEX procedure:

- ① start from a feasible solution to a system of n equations, compare its objective value with that of its “adjacent solutions”.
- ② if the current solution is better than all its adjacent solutions, stop! the current solution is optimal.
- ③ otherwise, move to an adjacent solution with higher objective value.
- ④ repeat steps ?? and ?? until the optimal solution is found.

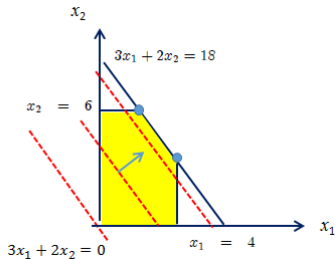
<https://www.youtube.com/watch?v=x8rYyQjEMMs>

Special Case (Non-unique Optimal Solutions)

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$$\begin{array}{ll}\max & \{3x_1 + 2x_2\} \\ \text{subject to} & x_1 \leq 4, \\ & x_2 \leq 6, \\ & 3x_1 + 2x_2 \leq 18, \\ & x_1 \geq 0, x_2 \geq 0.\end{array}$$



- every points on a segment of $3x_1 + 2x_2 = 18$ is an optimal solution.
- both $(2, 6)$ and $(4, 3)$ are optimal.

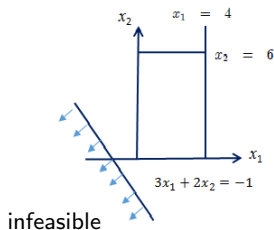
so we can still maximize the LP objective function by going through solutions of n simultaneous equations.

Exceptions (Don't solve, fix the model!)

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$$\begin{array}{ll}\max & \{2x_1 + x_2\} \\ \text{subject to} & x_1 \leq 4, \ x_2 \leq 6, \\ & 3x_1 + 2x_2 \leq -1, \\ & x_1 \geq 0, \ x_2 \geq 0.\end{array}$$



$$\begin{array}{ll}\max & \{2x_1 + x_2\} \\ \text{subject to} & x_1 \leq 4, \\ & x_1 \geq 0, \ x_2 \geq 0.\end{array}$$

