

# ① Foundations of Linear Programming

## Contents of this Course

- Models      ways to formulate real-world problems for processing by mathematical algorithms
- Algorithms      techniques for solving models
- Software      engines for executing algorithms

# History of Linear Programming

1665 Finding a Minimum Solution of a Function Newton

1788 Lagrangian Multipliers Lagrange

1823 Solution of a Set of Inequalities Fourier

1826 Solution of a Set of Linear Equations Gauss

1896 Solution of a Set of Linear Equations  
as a Combination of Extreme Points Minkowski

1936 Transposition Theorem and Linear Inequalities Motzkin

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↑ before Linear Programming

1939 Mathematical Methods of Organization  
and Production

Kantorovich  
(Nobel Prize 1975)

1941 Transportation Problem

Hitchcock

1947 Linear Programming Model

Dantzig

1951 Simplex Method

Dantzig

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↑ Origins of Linear Programming

1958 Integer Programming

Gomory, Johnson, Balas

1979 Ellipsoid Method

Khachiyan

1984 Interior Point Method

Karmarkar

# Optimization and Calculus

Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $g: \mathbb{R}^n \rightarrow \mathbb{R}^m$ ,  $L: \mathbb{R}^n \rightarrow \mathbb{R}^k$   
be twice continuously differentiable

## general optimization problem

$$\min f(x_1, \dots, x_n)$$

$$\text{subject to } g_i(x_1, \dots, x_n) \geq 0 \quad \forall i = 1, \dots, m$$

$$h_j(x_1, \dots, x_n) = 0 \quad \forall j = 1, \dots, k$$

$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

$$g(x) = \begin{pmatrix} g_1(x) \\ \vdots \\ g_m(x) \end{pmatrix}$$

$$L(x) = \begin{pmatrix} h_1(x) \\ \vdots \\ h_k(x) \end{pmatrix}$$

Let  $y \geq 0 \in \mathbb{R}^m$  and  $z \in \mathbb{R}^k$  free and consider the Lagrangian function:

$$L(x, y, z) = f(x) - \sum_{i=1}^m y_i \cdot g_i(x) - \sum_{j=1}^k z_j \cdot h_j(x)$$

→ instead of  $\min f(x)$  s.t. constraints  
search for a local/global minimum of  $L(x, y, z)$   
with respect to  $\max_{y, z} \min_x L(x, y, z)$

# Optimization and Linear Algebra (and Notation)

Linear systems of equations:  $Ax = b$

$m$  equations in  $n$  variables

$A \in \mathbb{R}^{m \times n}$ : matrix with rows  $a_i \in \mathbb{R}^n$  for  $i=1, \dots, m$

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & \dots & \dots & a_{mn} \end{pmatrix} \in \mathbb{R}^{m \times n}, \quad a_i = \begin{pmatrix} a_{i1} \\ \vdots \\ a_{in} \end{pmatrix}^T \in \mathbb{R}^n$$

$$a_i = (a_{i1} \quad \dots \quad a_{in})$$

$b \in \mathbb{R}^m, x \in \mathbb{R}^n$  : column vectors

$$b = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix} \in \mathbb{R}^m$$

$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n$$

right-hand sides

variables

When does  $Ax=b$  become interesting for optimization?