

# Chapter 2. Basic Properties of Linear Programming

## Sec. 2.1 Introduction

### Optimization:

Optimize  $z = f(x)$       the **objective** function  
subject to  $h(x) = 0$       the **equality** constraints  
 $g(x) \leq 0$       the **inequality** constraints  
 $x \in S$  in  $\mathbb{R}^n$ .      the **set** constraint (may or may not be included in  $h$  and  $g$ )

### MA420/520 Linear Programming:

(**LP**):  $f(x)$ ,  $h(x)$ ,  $g(x)$  **all linear** functions; straight boundary of  $S$ ;

### MA421/521 Nonlinear Programming:

(**NLP**): At least one of  $f(x)$ ,  $h(x)$ , and  $g(x)$  **is nonlinear**; or curved boundary of  $S$ ;

Chapter 5 (Poly-time)  
Chapter 6 (CLP)

MA420/MA520 LP	MA421/MA521 NLP
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## Examples of nonlinear models:

◇ the Cobb-Douglas production function that relates output (Y) to labor (L) and capital (K) can be written as  $Y = aL^bK^r$ .

◇ free falling  $h(t) = 0.5gt^2$ .

◇ exponential growth (compound interest)  $p(t) = p_0e^{at}$ .

◇ stock price in time (no analytic formula).

◇ shapes of physical objects.

◇ the shortest and equal time curves (inverted cycloid).

◇ least squares fitting of data. (AI)

◇ probability density functions.

◇ ...

# Linear Programming (Linear Optimization)

Ex1

Ex2

**Linear** ----- functions  $f(x) = 2x + 3$ ,

$$x + 4$$

equations:  $2x + 3 = 1$ ,

$$x + 4 = 3$$

inequalities:  $2x + 3 \leq 1$ ,

$$x + 4 \leq 5$$

**Optimization** --- minimization/maximization, both minimization&maximization (game)

say minimize above linear functions  $\rightarrow$  Answer: (optimal f-value)  $f^* = -\infty$

solve equation:  $2x + 3 = 1$

$\rightarrow$  solution  $x = -1$

solve inequality:  $2x + 3 \leq 1$

$\rightarrow$  solutions  $x \leq -1$

$\Rightarrow$  Algebra

Not interesting in LP.

To make the LP more meaningful, we add (on top of  $f(x)$ ):


more variables.  $\min f(x,y) = 2x + 3y + 3$ .

$\rightarrow f^* = -\infty$

(linear) constraints:  $\min f(x,y) = 2x + 3y + 3$  subject to (s.t.)  $x \geq 0$  and  $y \geq 0$ .

$\rightarrow f^* = f(0,0) = 4, x^* = y^* = 0$ .

Def. **Standard LP**:  $\min z = c^T x$  s.t. " $Ax=b, x \geq 0$ ",  $x \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$ ,  $A_{m \times n}$ ,  $c \in \mathbb{R}^n \rightarrow [A, b, c]$

 §2.5

Def.  $x^*$  is an optimal sol. if (1)  $Ax^* = b$ ,  $x^* \geq 0$ . (feasibility) (2) For any feasible  $x$ ,  $c^T x^* \leq c^T x$  (smallest obj. f value)

"Solve LP" — Find  $\underbrace{x^*}$  and  $z^* = f^* = f(x^*)$ .  $\rightarrow$  Your answer to any LP should be  $\begin{cases} x^*, z^* \\ \text{not feasible} \end{cases}$

Def.  $(LP_1)$  is equivalent to  $(LP_2)$  if any opt. sol. of one LP induces an opt. sol. of the other LP.

Claim: Any LP  $\Leftrightarrow$  std LP  $(A, b, c)$

Ex.  $\max w = c^T x$  s.t.  $Ax=b, x \geq 0 \Leftrightarrow$  std LP.  $\min \underbrace{z = -c^T x}_{\wedge}$  s.t.  $Ax=b, x \geq 0$ . Equivalence  $\begin{cases} \text{same } x^* \\ w^* = -z^* \end{cases}$

Ex1.  $LP_1: \min z = c^T x$  s.t.  $\underline{Ax \leq b}, x \geq 0 \Leftrightarrow$  std  $LP_2$ .

Sol.  $LP_1 \Leftrightarrow \min z = c^T x$  s.t.  $Ax + s = b, x \geq 0, s \geq 0$ ,  $s$  is called a slack var.

$x' = \begin{bmatrix} x \\ s \end{bmatrix} \Leftrightarrow \min w = c'^T x'$  s.t.  $\begin{bmatrix} A \\ 0 \end{bmatrix} x' = b', x' \geq 0$

$\uparrow$   
 $c' = \begin{bmatrix} c \\ 0 \end{bmatrix}, [A \ I], b' = b$

std LP with  $c', b', A' = [A \ I]$ .

Equivalence. same  $x^*$ ,  $s^* = b - Ax^*$ ,  $z^* = w^*$   $\left( w = \begin{bmatrix} c \\ 0 \end{bmatrix}'^T \begin{bmatrix} x \\ s \end{bmatrix} = c^T x + 0^T s = c^T x = z \right)$

Ex2.  $\min z = c^T x$  s.t.  $\underline{Ax \geq b, x \geq 0.}$

↓

$$Ax - s = b,$$

$s \geq 0$  called surplus var.

Ex.  $LP_1: \min z = c^T x$  s.t.  $Ax = b, \underline{0 \leq x \leq d.}$  (bounded var.)

So).  $\dots \dots \dots, x + s = d, s \geq 0. \dots \dots \dots$  std LP. ( $LP_2$ ).

$$\Leftrightarrow LP_2: \min w = \begin{bmatrix} c \\ 0 \end{bmatrix}^T \begin{bmatrix} x \\ s \end{bmatrix} \text{ s.t. } \begin{bmatrix} A & 0 \\ I & I \end{bmatrix} \begin{bmatrix} x \\ s \end{bmatrix} = \begin{bmatrix} b \\ d \end{bmatrix}, \quad x = \begin{bmatrix} x \\ s \end{bmatrix} \geq 0.$$

Equivalence. same  $x^*$ ,  $s^* = d - x^*$ ,  $z^* = w^*$

Ex. (free var.) 3-4.

$$\min z = 2x_1 - 3x_2 \quad \text{s.t.} \quad \downarrow -3x_1 + x_2 = 5, \quad x_2 \geq 0. \quad \Leftrightarrow \text{std LP?}$$

Sol. (1). replacement of free  $x_1$  by two nonnegative vars.  $x_1 = u - v, \quad u \geq 0, \quad v \geq 0$

$$\text{std LP.} \quad \min z = 2(u - v) - 3x_2 \quad \text{s.t.} \quad -3(u - v) + x_2 = 5, \quad x_2 \geq 0, \quad u \geq 0, \quad v \geq 0.$$

$$A' = \quad \quad \quad b' = \quad \quad \quad c' =$$

(2). Eliminate  $x_1$ ,  $x_1 = \frac{5 - x_2}{-3} = -\frac{5}{3} + \frac{1}{3}x_2$

$$\text{LP:} \quad 2\left(-\frac{5}{3} + \frac{1}{3}x_2\right) - 3x_2 \quad \text{s.t.} \quad x_2 \geq 0 \quad \Leftrightarrow \quad \min \frac{-10}{3} + \frac{2}{3}x_2 - 3x_2 \quad \text{s.t.} \quad x_2 \geq 0$$

$$\Leftrightarrow \text{std LP.} \quad \min_{\wedge} \overset{w=}{-\frac{7}{3}}x_2 \quad \text{s.t.} \quad x_2 \geq 0 \quad \quad C = -\frac{7}{3}, \quad A = 0, \quad b = 0.$$

$$\text{Equivalence, since } x_2^*, \quad x_1^* = -\frac{5}{3} + \frac{1}{3}x_2^*, \quad z^* = w^* - \frac{10}{3}$$

