

- Formulate problems as LPs.
- Equivalence theorem
- Minimum cost trading to achieve target sector exposures.

Midterm 25 points x 4.

- Least squares
- Multi-objective / constrained LS.
- Linear optimization modeling
- Linear optimization geometry  
(equivalence theorem)

no coding / no simplex

You are allowed to use all course materials (lecture slides, precept, homework, books)

No internet No communication

NO precept next week!

(Q1) Formulate problems as LPs

(a)  $A \in \mathbb{R}^{M \times n}$   $b \in \mathbb{R}^M$  find the vector  $x \in \mathbb{R}^n$  that minimizes.

$$\sum_{i=1}^M \max\{0, a_i^T x + b_i\}. \quad (1)$$

$$\underline{c_i = \max\{0, a_i^T x + b_i\}}.$$

$$\begin{cases} c_i \geq 0 \\ c_i \geq a_i^T x + b_i \end{cases}$$

$$\text{minimize } \mathbf{1}^T \cdot \mathbf{c} \quad \mathbf{c} = \begin{pmatrix} c_1 \\ \vdots \\ c_M \end{pmatrix}$$

$$\underset{(c, x)}{\text{minimize}} \quad (\mathbb{1}^T, 0) \begin{pmatrix} c \\ x \end{pmatrix}$$

$$\boxed{\mathbb{1}^T C}$$

s.t.  $\begin{cases} c_i \geq 0 \\ c_i - a_i^T x \geq b_i \end{cases}$

(2)

Let  $X_1$  minimize ①

$(C^*, X_2)$  minimize ②

~~$(c, x)$~~

By definition of ①.

$$\{\max\{0, a_i^T x_1 + b_i\}\} \leq \{\max\{0, a_i^T x_2 + b_i\}\}$$

for any  $c$  satisfies  $\begin{cases} c_i \geq 0 \\ c_i - a_i^T x \geq b_i \end{cases}$

$(C, X_1)$  is a feasible solution of ②

$$\Rightarrow \mathbb{1}^T C \geq \mathbb{1}^T C^* \quad \text{if } x = x^*$$

$$(\mathbb{1}^T, 0) \begin{pmatrix} c \\ x_1 \end{pmatrix} \geq (\mathbb{1}^T, 0) \begin{pmatrix} c^* \\ x_2 \end{pmatrix}$$

$$\underset{(c, x)}{\text{minimize}} \quad (\mathbb{1}^T, 0) \begin{pmatrix} c \\ x \end{pmatrix}$$

$$\text{s.t.} \quad \begin{cases} c_i \geq 0 \\ c_i - a_i^T x \geq b_i \end{cases}$$



$$c_i \geq \max\{0, a_i^T x + b_i\}.$$

$$\Rightarrow \mathbb{1}^T c = \sum \max\{0, a_i^T x_1 + b\}.$$

$$\mathbb{1}^T c^* = \sum \max\{0, a_i^T x_2 + b\}.$$

$$\underline{\sum \max\{0, a_i^T x_1 + b\}} \geq \underline{\sum \max\{0, a_i^T x_2 + b\}}$$

$$\leq$$

$$\Rightarrow \underline{\sum \max\{0, a_i^T x_1 + b\}}$$

$$= \underline{\sum \max\{0, a_i^T x_2 + b\}}.$$

## C2 equivalence theorem.

$$P = \{ x \mid Ax \leq b \}.$$

(1) Show that every vertex is an extreme point.

vertex: if  $\exists c$  such that  $x$  is the unique optimum of  
minimize  $c^T y$   
s.t.  $y \in P$ .

extreme point:  $x$  not on a straight line between any other points in the set  $P$

vertex  $\Rightarrow$  extreme point.

(proof by contradiction)

Assume  $x$  is vertex, but not extreme point.

$$\exists y, z \neq x \quad y, z \in P \quad \underline{x = \lambda y + (1-\lambda)z}$$

vertex  $\exists c$  such that

$$c^T x < c^T w \quad \forall w \in P$$

$$c^T y, c^T z > c^T x$$

$$\underline{c^T x} < \underline{c^T y}, \underline{c^T z}$$

$$\Rightarrow \underline{\underline{c^T x}} = \underline{\underline{\lambda c^T y + (1-\lambda) c^T z}} > \underline{\underline{c^T x}}$$

contradiction!

$$I(X) = \left\{ \begin{array}{l} 1, 2, 4, 6 \\ \{a_1, a_2, a_4, a_6\} \end{array} \right\}$$

$$P = \{x \mid Ax \leq b\}$$
$$a_i^T x \leq b_i$$

② extreme point  $\Rightarrow$  feasible solution

$\{a_i \mid i \in I(X)\}$  has  $n$  linear independent vectors.

$$I(X) = \{i \in \{1, \dots, m\} \mid a_i^T x = b_i\}$$

(proof by contradiction)

Assume  $x$  is extreme point but not basic feasible solution.

Then  $\{a_i : i \in I(x)\}$  does not span

$$\mathbb{R}^n \Rightarrow \exists d \in \mathbb{R}^n \quad a_i^T d = 0 \quad i \in I(x)$$

Let  $y = x + \varepsilon d$        $z = x - \varepsilon d$ .  $a_i^T d = 0$

$$\left\{ \begin{array}{l} \forall i \in I(x) \quad a_i^T x = b_i \Rightarrow a_i^T y = b_i \\ \forall i \notin I(x) \quad a_i^T x < b_i \end{array} \right. \quad \left\{ \begin{array}{l} a_i^T z = b_i \\ a_i^T y = b_i \end{array} \right.$$

$$\Rightarrow a_i^T y < b_i \quad a_i^T z < b_i \quad \varepsilon \text{ small enough}$$

$$x = 0.5y + 0.5z \quad \text{and } y, z \in P.$$

$\Rightarrow x$  not extreme point

contradiction!

$$a_i^T y = a_i^T x + \underbrace{\varepsilon - a_i^T d}_{\leq b_i} \quad n$$

$$\overbrace{a_i^T x}^{\text{ent}} \quad \overbrace{a_i^T X}^{\text{b}_i} \quad \overbrace{b_i}^{a_i^T X}$$

Q3.  $h^{\text{curr}}$   $\rightarrow$  n-vector (portfolio)  
with entries giving the dollar  
value invested in the n assets.

$$\text{total value} = 1^T h^{\text{curr}} \quad \Delta$$

we seek a new portfolio.  $h$ .

$$\text{s.t. } 1^T h = 1^T h^{\text{curr}}$$

Same total ~~vector~~ value.

m-industry sectors.

(Tech / bio).

$s$   $m$ -vector denotes the  
sector exposures (dollar value)  
to the  $m$  sectors.  $s = Sh$ .

$S$  is  $m \times n$  matrix

$$\begin{cases} S_{ij} = 1 & \text{asset } j \text{ in sector } i \\ S_{ij} = 0 & \text{asset } j \text{ not in} \\ & \text{sector } i. \end{cases}$$

The new portfolio must  
have a given sector exposure  
values.

Among all portfolios that  
have the same value as  
our current portfolio  $h^{curr}$  and  
achieve the desire exposures

we wish to minimize the  
trading cost.

$$\sum_{i=1}^n k_i (h_i - h_i^{curr})^2 = \infty$$

$$k_i > 0$$

Explain how to find  $h$ .  
give KKT equations. LP

$$\underline{h}^{\text{curr}} = (v_1, v_2, \dots, v_n) \cdot n$$

$$h = (h_1, h_2, \dots, h_n) \cdot n$$

$$\text{constraints} = \left\{ \begin{array}{l} Sh = s^{\text{des}} \\ \underline{U}^T h = \underline{I}^T h^{\text{curr}} \end{array} \right.$$

$$\text{minimize } \sum_{i=1}^n k_i (h_i - h_i^{\text{curr}})^2$$

$$\text{Let } K = \begin{pmatrix} k_1 & & \\ & \ddots & \\ & & k_n \end{pmatrix}$$

$$\underline{U}^T h^{\text{curr}} = \underline{V}$$

LP:

$$\underset{h}{\text{minimize}} \|Kh - Kh^{\text{curr}}\|_2$$

$$[S \\ I^T] h = [S^{\text{des}} \\ v]$$

[ $kT$  equation:

$$\begin{aligned} & \min \|Ax - b\|^2 \\ \text{s.t. } & Cx = d \end{aligned}$$

$$A = \underline{K}, \quad b = \underline{Kh^{\text{curr}}}$$

$$C = \begin{bmatrix} S \\ I^T \end{bmatrix}, \quad d = \begin{bmatrix} S^{\text{des}} \\ v \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 2A^TA & C^T \\ C & 0 \end{bmatrix} \begin{bmatrix} x^* \\ z \end{bmatrix} = \begin{bmatrix} 2A^Tb \\ d \end{bmatrix}$$

$x^*, z$