

HW2.

Seongmin Lee (CID: 01247436)

(a) Indirect Method.

$$\text{minimise } 3x_1 + 5x_2 - x_3$$

$$\text{s.t. } x_1 + x_3 = 4$$

$$x_2 - 2x_3 \leq 2$$

$$x_1, x_2 \geq 0, x_3 \text{ unrestricted.}$$

i) equality  $\rightarrow$  inequality

$$x_1 + x_3 = 4 \rightarrow x_1 + x_3 \geq 4$$

$$x_1 + x_3 \leq 4$$

ii) reverse the sign of inequality to  $\geq$

$$x_1 + x_3 \geq 4 \text{ (ok)}$$

$$x_1 + x_3 \leq 4 \xrightarrow{(-1)} -x_1 - x_3 \geq -4$$

$$x_2 - 2x_3 \leq 2 \xrightarrow{(-1)} -x_2 + 2x_3 \geq -2$$

iii)  $x_3 = x_3^+ - x_3^-$  where  $x_3^+, x_3^- \geq 0$

$$\text{objective fn: minimise } 3x_1 + 5x_2 - x_3^+ + x_3^-$$

$$\text{Subject to } x_1 + x_3^+ - x_3^- \geq 4$$

$$\text{Max } CTx \text{ s.t. } Ax \leq b \quad -x_1 - x_3^+ + x_3^- \geq -4$$

$$\text{Min } bTy \text{ s.t. } ATy \geq c \quad -x_2 + 2x_3^+ - 2x_3^- \geq -2$$

$$x_1, x_2, x_3^+, x_3^- \geq 0$$

iv.) The Matrix form of the primal problem. is as below:

$$\begin{pmatrix} b^T \\ 3 \\ 5 \\ -1 \\ -1 \end{pmatrix}^T \begin{pmatrix} x_1 \\ x_2 \\ x_3^+ \\ x_3^- \end{pmatrix} \text{ s.t. } \begin{pmatrix} A^T \\ 1 & 0 & 1 & -1 \\ -1 & 0 & -1 & 1 \\ 0 & -1 & 2 & -2 \end{pmatrix}^T \begin{pmatrix} x_1 \\ x_2 \\ x_3^+ \\ x_3^- \end{pmatrix} \geq \begin{pmatrix} c \\ 4 \\ -4 \\ -2 \end{pmatrix}$$

V) change to the dual form.

$$\begin{array}{c} C^T \quad x \\ \begin{pmatrix} 4 \\ -4 \\ -2 \end{pmatrix}^T \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \end{array} \quad \text{Subject to} \quad \begin{array}{c} A^T -> A \\ \begin{pmatrix} 1 & -1 & 0 \\ 0 & 0 & -1 \\ 1 & -1 & -2 \end{pmatrix} \end{array} \quad \begin{array}{c} x \\ \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \end{array} \leq \begin{array}{c} b \\ \begin{pmatrix} 3 \\ 5 \\ -1 \end{pmatrix} \end{array}$$

∴ Maximise  $4y_1 - 4y_2 - 2y_3$   
 Subject to  $y_1 - y_2 \leq 3$   
 $-y_3 \leq 5$   
 $y_1 - y_2 + 2y_3 \leq -1$   
 $-y_1 + y_2 - 2y_3 \leq 1$   
 $y_1, y_2, y_3 \geq 0$ .

(b) Direct Method.

Maximise  $x_1 - x_3$   
 Subject to  $x_1 + x_2 = 4$   
 $x_3 \leq 2$   
 $x_1, x_2 \leq 0$   $x_3$  unrestricted.

i) change the expressions for the direct method.

maximise  $x_1 + 0 \cdot x_2 - x_3$   
 s.t  $\begin{array}{l} x_1 + x_2 + 0 \cdot x_3 = 4 \quad -> y_1 \text{ (unrestricted)} \\ 0 \cdot x_1 + 0 \cdot x_2 + x_3 \leq 2 \quad -> y_2 > 0 \end{array}$   
 1st con      2nd con      3rd con      objective coef  
 checked by the SOB Table in the lecture note

minimise  $4y_1 + 2y_2$   $x_1, x_2 \leq 0$  in the primal variables.  
 Subject to  $y_1 + 0y_2 \leq 1$   
 $y_1 + 0y_2 \leq 0$   $x_3$  is unrestricted.  
 $0y_1 + y_2 = -1$

∴ minimise  $4y_1 + 2y_2$

Subject to  $y_1 \leq 1$

$y_1 \leq 0$

$y_2 = -1$

$y_1$  unrestricted,  $y_2 \geq 0$ .

$y_2 = -1$  &  $y_2 \geq 0 \Rightarrow$  infeasible.

(4)

(a) Formulate the 1-norm regression problem

By the definition of the 1-norm regression,

minimise  $\|y - X\beta\|$ , where  $\left\| \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \right\| = |x_1| + \dots + |x_n|$

put  $y = \text{comp}_i$

$x_{1i} = \text{years}_i$

$x_{2i} = \text{change stock}_i$

$x_{3i} = \text{change sales}_i$

$x_{4i} = \text{mba}_i$

$$\Rightarrow \text{comp}_i = \beta_1 \cdot \text{years}_i + \beta_2 \cdot \text{change stock}_i + \beta_3 \cdot \text{change sales}_i + \beta_4 \cdot \text{mba}_i$$

$\Rightarrow$  minimise  $| \text{comp}_i - (\beta_1 \cdot \text{years}_i + \beta_2 \cdot \text{change stock}_i + \beta_3 \cdot \text{change sales}_i + \beta_4 \cdot \text{mba}_i) | +$

$\vdots$   
 $+ | \text{comp}_i - (\beta_1 \cdot \text{years}_i + \beta_2 \cdot \text{change stock}_i + \beta_3 \cdot \text{change sales}_i + \beta_4 \cdot \text{mba}_i) |$

subject to  $\beta_1, \beta_2, \beta_3, \beta_4 \in \mathbb{R}$

reformulate as

Minimize.  $\theta_1 + \theta_2 + \dots + \theta_n$

Subject to.  $\theta_1 = |comp_1 - (\beta_1 \cdot years_1 + \beta_2 \cdot change\_stock_1 + \beta_3 \cdot change\_sales_1 + \beta_4 \cdot mba_1)|$

$\vdots$   
 $\theta_n = |comp_n - (\beta_1 \cdot years_n + \beta_2 \cdot change\_stock_n + \beta_3 \cdot change\_sales_n + \beta_4 \cdot mba_n)|$

$\beta_1, \beta_2, \dots, \beta_n \in \mathbb{R}$

To change equality to inequality ( $\geq$ ) in minimisation problems

● We can re-write ex.  $\theta = |a-b|$  to  $\theta \geq (a-b)$  and  $\theta \geq -(a-b)$ .

Therefore, the 1 norm linear program is as below;

Decision Variables:  $\beta_1, \beta_2, \beta_3, \beta_4 \in \mathbb{R}$

Minimize  $\theta_1 + \theta_2 + \dots + \theta_n$

Subject to.  $\theta_1 \geq comp_1 - (\beta_1 \cdot years_1 + \beta_2 \cdot change\_stock_1 + \beta_3 \cdot change\_sales_1 + \beta_4 \cdot mba_1)$

$\theta_i \geq (\beta_1 \cdot years_i + \beta_2 \cdot change\_stock_i + \beta_3 \cdot change\_sales_i + \beta_4 \cdot mba_i) - comp_i$

$i = 1, 2, \dots, n$

● (b) The Ample file is attached in the Hub.

The objective value is 14652.83, which is the total residuals

at  $\beta_1$  for years = 169.274

$\beta_2$  for change stock = 2.41

$\beta_3$  for change sales = -0.1425

$\beta_4$  for mba = 38.5



### (c) Infinite norm regression

Consider the definition of norm. We can extend it to infinite level

$$\|X\|_n = \sqrt[n]{|x_1|^n + |x_2|^n + \dots + |x_n|^n} \text{ [normal form]}$$

if  $n \rightarrow \infty$ , the largest number among  $|x_1|^n, \dots, |x_n|^n$  will be survived.

Therefore, we can generalize it as below.

$$\|X\|_\infty = \max(|x_1|, \dots, |x_n|)$$

ex)  $\|x\|_\infty = \sqrt[\infty]{|1|^{\infty} + |10|^{\infty} + |100|^{\infty}}$   
can be ignored. b/c  
 $= \sqrt[\infty]{|100|^{\infty}} = |100|$

If we apply the above method to the regression, we can re-write

$$\text{minimise } \|y - X\beta\|_\infty \Leftrightarrow \max(|y_1 - x_1\beta|, \dots, |y_n - x_n\beta|)$$

So, we can put  $\theta = |y - X\beta|$

$$\theta = \max(\theta_1, \theta_2, \dots, \theta_n)$$

$$\text{where } \theta_i = |\text{Comp}_i - (\beta_1 \cdot \text{years}_i + \beta_2 \cdot \text{change stock}_i + \beta_3 \cdot \text{change sales}_i + \beta_4 \cdot \text{mka}_i)|$$

To re-summarize

minimize  $\theta$

subject to  $\theta = \max(\theta_1, \theta_2, \dots, \theta_n)$ , where  $\theta_i = |y_i - x_i\beta|$

By relaxing the equality in that  $\theta$  is greater or equal than  $\theta_1, \dots, \theta_n$ .

minimize  $\theta$

subject to  $\theta \geq |y_i - x_i\beta|, \forall i = 1 \dots 50$ .

Finally, we can remove the absolute value operator to re-formulate the LP as below

minimize  $\theta$

subject to  $\theta \geq y_i - x_i\beta$

$\theta \geq -(y_i - x_i\beta) \quad \forall i = 1 \dots 50$ .