

$\hat{\mu}_1 = \frac{1}{396} (0.05 \times 31 + 0.1 \times 5 + 0.15 \times 18 + 0.2 \times 12 + 0.3 \times 144) = 0.202$
 $\hat{\mu}_2 = \frac{1}{604} (0.05 \times 169 + 0.1 \times 148 + 0.15 \times 132 + 0.2 \times 99 + 0.3 \times 36) = 0.132$
 $\hat{\sigma}_1^2 = \frac{1}{396-1} \sum_{i=1}^n (x_i - 0.202)^2 = \frac{2.824}{395} = 0.0071$
 $\hat{\sigma}_2^2 = \frac{1}{604-1} \sum_{i=1}^n (x_i - 0.132)^2 = \frac{3.368}{603} = 0.0055$

$$\ln \mu(x) = -\frac{x^2}{2\sigma^2} + \frac{x\mu}{\sigma^2} - \frac{\mu^2}{2\sigma^2} + \log \pi \mu$$

106064701
蔡明霖
(HW2)

$5\% = \int_1(0.05) \geq 0.17$
 $10\% = \int_1(0.1) \geq 0.23$
 $15\% = \int_1(0.15) \geq 0.33$
 $20\% = \int_1(0.2) \geq 0.46$
 $25\% = \int_1(0.25) \geq 0.63$
 $30\% = \int_1(0.3) \geq 0.78$ #

← 查折扣所对应的兑换率。
(代入公式算得)

2. $\hat{\sigma}^2 = 25$ ($\bar{x}_1 = 10, \bar{x}_2 = 0$)

$$P_{YB}(4) = \frac{0.8 \cdot \exp(-\frac{1}{2 \times 25} \cdot (4-10)^2)}{\frac{1}{5} \cdot 0.8 \cdot \exp(-\frac{1}{2 \times 25} \cdot (4-10)^2) + (1-0.8) \cdot \exp(-\frac{1}{2 \times 25} \cdot (4-0)^2)} = 0.774$$
 #

$$NRE_{(VCM)} = \frac{1}{n} \sum_{i=1}^n \left(\frac{y_i - \hat{y}_i}{1 - h_i} \right)^2, \quad h_i = x_i^T (X^T X)^{-1} x_i$$

$$\text{Proof of } y = X(X^T X)^{-1} X^T y \Rightarrow \hat{y}_i = x_i^T (X^T X)^{-1} X^T y$$

$$\textcircled{2} \hat{y}_i^{(-i)} = x_i^T (X^{(-i)T} X^{(-i)})^{-1} X^{(-i)T} y^{(-i)}$$
 ← Removing the i th term.

$$\text{hence } (A - X X^T)^{-1} = A^{-1} + \frac{A^{-1} X X^T A^{-1}}{1 - X^T A^{-1} X} \Rightarrow \hat{y}_i = x_i^T \left[(X^T X)^{-1} + \frac{X X^T A^{-1} X X^T}{1 - x_i^T (X^T X)^{-1} x_i} \right] (X^T y - x_i \hat{y}_i)$$

$$\textcircled{1} \Rightarrow A \cdot x_i = (X^T X)^{-1} x_i + (X^T X)^{-1} x_i \left(\frac{x_i^T (X^T X)^{-1} x_i}{1 - x_i^T (X^T X)^{-1} x_i} \right)$$

$$= \frac{1}{1 - h_i} \cdot (X^T X)^{-1} x_i \Rightarrow x_i^T \frac{(X^T X)^{-1} x_i}{1 - h_i} \cdot \hat{y}_i = \frac{h_i \hat{y}_i}{1 - h_i} \quad \textcircled{1}$$

$$\textcircled{2} \Rightarrow (X^T y - x_i \hat{y}_i) \cdot x_i^T + (X^T y - x_i \hat{y}_i) \cdot \left(\frac{x_i^T (X^T X)^{-1} x_i}{1 - h_i} \right) = (X^T X)^{-1} \left[\frac{1 - h_i}{1 - h_i} + \frac{x_i x_i^T (X^T X)^{-1} x_i}{1 - h_i} \right] \cdot x_i^T = \frac{(X^T X)^{-1} x_i^T}{1 - h_i}$$

$$\Rightarrow x_i^T \frac{(X^T X)^{-1} x_i^T}{1 - h_i} \cdot y = \frac{\hat{y}_i}{1 - h_i} \quad \textcircled{2} \Rightarrow \textcircled{2} - \textcircled{1} = \frac{\hat{y}_i - h_i \hat{y}_i}{1 - h_i} = \hat{y}_i^{(-i)}$$

③ NRE is

$$VCM = \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i^{(-i)})^2 = \frac{1}{n} \sum_{i=1}^n \left(\frac{y_i - \hat{y}_i}{1 - h_i} \right)^2$$
 #

$$(a) \quad L = \|Y - XB\|^2 + \lambda \|B\|^2 \\ = (Y - XB)^T (Y - XB) + \lambda B^T B = Y^T Y - Y^T X B - B^T X^T Y + B^T X^T X B + \lambda B^T B$$

对 B 偏微:

$$\begin{aligned} \text{得} \quad & 0 = -X^T Y - X^T Y + (X^T X + (X^T X)^T) B + 2\lambda B \\ & = 2X^T X B - 2X^T Y + 2\lambda B = 0 \end{aligned}$$

$$\Rightarrow 2X^T (XB - Y) + 2\lambda B = 0$$

$$\Rightarrow \hat{B}_\lambda = (X^T X + \lambda I)^{-1} X^T Y \quad \#$$

$\because \lambda > 0$ \therefore 当 $\lambda > 0$ 时, 如同分母变大, 则 B 会缩小.

$$\Rightarrow \|B_\lambda\|_{\lambda \rightarrow \infty} \leq \|B_\lambda\|_{\lambda=0} \quad \#$$

$$\begin{aligned} (b) \quad \overline{err} &= \frac{1}{n} \times (Y - X(X^T X + \lambda I)^{-1} X^T Y)^2 \\ &= \frac{1}{n} \times (I_m Y - X(X^T X + \lambda I)^{-1} X^T Y)^2 \\ &= \frac{1}{n} \times \left[(I_m - X(X^T X + \lambda I)^{-1} X^T) \cdot Y \right]^2 \\ &= \frac{1}{n} \cdot Y^T \cdot (I_m - X(X^T X + \lambda I)^{-1} X^T)^T \cdot Y \quad \# \end{aligned}$$

$\because \lambda > 0$ \therefore 当 λ 增大时, 例权会变小, $\Rightarrow I_m -$ 变小则变大.
 $\Rightarrow \overline{err}$ 变大 $\#$



5.

例

- ⑥ $(10, 3, 5) \rightarrow (0.846, 1.192)$
- ⑤ $(10, 3, 3) \rightarrow (0.37, 1.14)$
- ④ $(10, 5, 5) \rightarrow (1, 1.14)$
- ③ $(3, 10, 10) \rightarrow (0.57, 1.14)$
- ② $(3, 5, 5) \rightarrow (3.5, 0.5)$
- ① $(5, 10, 10) \rightarrow (1, 1.14)$
- ⑦ $(5, 3, 3) \rightarrow (5.5, 0.5)$

加权平均

加权平均

$$\bar{B} = (3.29, 1.058)$$

(a)

$$RSS = 0.35 + 0.269 + 1.41$$

$$KSE = \sqrt{\frac{RSS}{3-2}} = 1.454 = \sigma$$

$$SE(\hat{\beta}_0) = \sqrt{\hat{\sigma}^2 \sum x_i^2} = \sqrt{3.47} = 1.86$$

$$SE(\hat{\beta}_1) = \sqrt{\hat{\sigma}^2 \sum x_i^2} = \sqrt{0.57} = 0.57$$

(b)

$$\hat{\beta}_0 = 2.84$$

$$\hat{\beta}_1 = 1.192$$

$$SE(\hat{\beta}_0) = 1.33$$

$$SE(\hat{\beta}_1) = 0.2. \quad \#$$

例

- ⑥ $(10, 3, 5) \rightarrow (2.846, 1.192)$
 ⑦ $(10, 3, 3) \rightarrow (2.37, 1.14)$
 ⑧ $(10, 5, 5) \rightarrow (1, 1.14)$
 ⑨ $(3, 10, 10) \rightarrow (2.57, 1.14)$
 ⑩ $(3, 5, 5) \rightarrow (2.5, 0.5)$
 ⑪ $(5, 10, 10) \rightarrow (1, 1.14)$
 ⑫ $(5, 3, 3) \rightarrow (2.5, 0.5)$

加權平均

$$\bar{B} = (3.29, 1.058)$$

(a)

$$SE(B) = \left(\frac{1}{B-1} \sum_{k=1}^B (B_k - \bar{B})^2 \right)^{0.5} \Big|_{B \rightarrow \infty}$$

$B \rightarrow \infty$, 可用相同比例看待.

$$\Rightarrow SE(B_0) = \sqrt{1.397} = 1.61$$

$$SE(B_1) = \sqrt{0.1132} = 0.33 \#$$

(b)

$$B_0 = 2.84, \quad RSS = 1.038$$

$$B_1 = 1.192$$

$$SE(B_0) = 1.33$$

$$SE(B_1) = 0.2 \#$$

$$6. (a) (y_1 - \beta_1 x_{11} - \beta_2 x_{12})^2 + (y_2 - \beta_1 x_{21} - \beta_2 x_{22})^2 + \lambda (\beta_1^2 + \beta_2^2) \quad \# \quad (ridge)$$

$$(b) \text{展開上式: } = y_1^2 + \beta_1^2 x_{11}^2 + \beta_2^2 x_{12}^2 - 2\beta_1 x_{11} y_1 - 2\beta_2 x_{12} y_1 + 2\beta_1 \beta_2 x_{11} x_{12} \\ + y_2^2 + \beta_1^2 x_{21}^2 + \beta_2^2 x_{22}^2 - 2\beta_1 x_{21} y_2 - 2\beta_2 x_{22} y_2 + 2\beta_1 \beta_2 x_{21} x_{22}$$

$$(2) \frac{\partial}{\partial \beta_1} = (2\beta_1 x_{11}^2 - 2x_{11} y_1 + 2\beta_2 x_{11} x_{12}) + (2\beta_1 x_{21}^2 - 2x_{21} y_2 + 2\beta_2 x_{21} x_{22}) + 2\lambda \beta_1 = 0$$

$$(3) x_{11} = -x_{12} = a \Rightarrow \text{上式} = (\beta_1 a^2 - a y_1 - \beta_2 a^2) + (\beta_1 b^2 - b y_2 - \beta_2 b^2) + \lambda \beta_1 = 0 \\ x_{21} = -x_{22} = b \\ = \beta_1 (a^2 + b^2) - \beta_2 (a^2 + b^2) + \lambda \beta_1 = a y_1 + b y_2$$

$$(4) \text{Add. } 2\beta_1 ab \rightarrow 2\beta_2 ab$$

$$\Rightarrow \beta_1 (a+b)^2 = \beta_2 (a+b)^2 + \lambda \beta_1 = a y_1 + b y_2 + 2\beta_1 ab - 2\beta_2 ab$$

$$\Rightarrow \lambda \beta_1 = a y_1 + b y_2 + 2\beta_1 ab - 2\beta_2 ab$$

$$\text{類似地可得} \Rightarrow \lambda \beta_2 = -a y_1 - b y_2 - 2\beta_1 ab + 2\beta_2 ab$$

$$\Rightarrow \beta_1 = -\beta_2 \quad \#$$

$$(c) \text{lasso } (y_1 - \beta_1 x_{11} - \beta_2 x_{12})^2 + (y_2 - \beta_1 x_{21} - \beta_2 x_{22})^2 + \lambda (|\beta_1| + |\beta_2|) \quad \#$$

(d) 僅變動正則化項而已。

$$\Rightarrow \frac{\partial \lambda |\beta|}{\partial \beta} = \lambda \frac{|\beta|}{\beta} \Rightarrow \lambda \frac{|\beta_1|}{\beta_1} = -\lambda \frac{|\beta_2|}{\beta_2} \quad \#$$

$$(\text{hint } \frac{\partial}{\partial x} |x| = \frac{|x|}{x})$$