

STAT 542 HW2

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Question 1

(a)(b)(c)

Question 1

$$\begin{aligned} (1) \quad \frac{\partial J(\beta, \beta_0)}{\partial \beta} &= \frac{\partial}{\partial \beta} \left\{ \frac{1}{2n} \|y - X\beta - \beta_0 \mathbf{1}\|^2 + \lambda \sum_{j=1}^p |\beta_j| \right\} \\ &= \frac{1}{2n} \frac{\partial}{\partial \beta} \langle y - X\beta - \beta_0 \mathbf{1}, y - X\beta - \beta_0 \mathbf{1} \rangle \\ &= \frac{1}{2n} \left(-\sum_{i=1}^n y_i + \sum_{j=1}^p x_{ij} \beta_j + n\beta_0 \right) = 0 \end{aligned}$$

$$\Rightarrow \beta_0 = \frac{1}{n} \sum_{i=1}^n \left(y_i - \sum_{j=1}^p x_{ij} \beta_j \right)$$

$$\begin{aligned} (2) \quad \frac{\partial J(\beta, \beta_0)}{\partial \beta_j} &= \frac{\partial}{\partial \beta_j} \left\{ \frac{1}{2n} \|y - X\beta - \beta_0 \mathbf{1}\|^2 + \lambda \sum_{j=1}^p |\beta_j| \right\} \\ &= \frac{1}{n} \sum_{i=1}^n (-x_{ij}) \left(y_i - \beta_0 - \sum_{k=1}^p x_{ik} \beta_k \right) + \begin{cases} \lambda, & \beta_j > 0 \\ -\lambda, & \beta_j < 0 \end{cases} \\ &= -\frac{1}{n} \sum_{i=1}^n \left(x_{ij} y_i - \beta_0 x_{ij} - \sum_{k=1}^p x_{ij} x_{ik} \beta_k \right) \pm \lambda \\ &= -\frac{1}{n} \sum_{i=1}^n \left(x_{ij} y_i - \beta_0 x_{ij} + \sum_{k=1, k \neq j}^p x_{ij} x_{ik} \beta_k \right) \pm \lambda \\ &\quad + \frac{1}{n} \sum_{i=1}^n (x_{ij}^2 \beta_j) = 0 \end{aligned}$$

$$\Rightarrow \beta_j = \frac{1}{\sum_{i=1}^n x_{ij}^2} \left\{ \sum_{i=1}^n \left(x_{ij} y_i - \beta_0 x_{ij} - \sum_{k=1, k \neq j}^p x_{ij} x_{ik} \beta_k \right) \right. \\ \left. + \begin{cases} -\lambda n, & \beta_j > 0 \\ +\lambda n, & \beta_j < 0 \end{cases} \right\}$$

$$\text{let } A = \sum_{i=1}^n \left(x_{ij} y_i - \beta_0 x_{ij} - \sum_{k=1, k \neq j}^p x_{ij} x_{ik} \beta_k \right), \quad B = \sum_{i=1}^n x_{ij}^2$$

$$\beta_j = \begin{cases} \frac{A - \lambda n}{B}, & A > \lambda n \\ 0, & -\lambda n \leq A \leq \lambda n \\ \frac{A + \lambda n}{B}, & A < -\lambda n \end{cases}$$

13) Let $\beta_1, \dots, \beta_p = 0$, $\beta_0 = \bar{y}$

$$|\lambda \cdot n| = A$$

$$\lambda_{\max} = \max_{j=1, \dots, p} \left| \frac{\sum_{i=1}^n (x_{ij} y_i - x_{ij} \bar{y})}{n} \right|$$

(d)

I use the beta of different lambda from the training data to predict Y in the test data.

Then I got the prediction errors of different lambda. To be the best lambda, the error should be the minimum. Thus, I got the best lambda=0.05372383.

Question 2

(a)

Number of nonzero parameters: 61(including intercept)

Best lambda: 0.04173215

(b)

Estimated degree of freedom: 60.31393, which is close to the number of non-zero parameter in part a.

(c)

Number of nonzero parameters: 201(including intercept)

Best lambda: 29.291

theoretical value of the degrees of freedom: 170.1506, from the following formula:

$$df(\lambda) = \text{Trace}(\mathbf{X}(\mathbf{X}^T \mathbf{X} + \lambda \mathbf{I})^{-1} \mathbf{X}^T) = \sum_{j=1}^p \frac{d_j^2}{d_j^2 + \lambda}$$

Estimated degree of freedom: 173.9836, which matches the theoretical value