

## Tutorial 2

1. If  $\mathbb{X}$  is orthonormal, then  $\mathbb{X}^\top \mathbb{X} = nI_p$ , the optimal choice of  $\lambda$  minimizing

$E\|\hat{\beta}_{ridge} - \beta\|^2$  and the expected prediction error is:

$$\lambda^* = \frac{p\sigma^2}{n \sum_{j=1}^p \beta_j^2}$$

where  $(\beta_1, \beta_2, \dots, \beta_p)$  is the true coefficient vector

2. For the simple linear regression model,

$$y_i = \beta_0 + \beta_1 \mathbf{x}_i + \varepsilon_i, \quad i = 1, \dots, n$$

Suppose we use the weighted least square estimation and minimize

$$\sum_{i=1}^n w_i (Y_i - \beta_0 - \beta_1 \mathbf{x}_i)^2$$

Find the weighted estimator of  $\beta_0$  and  $\beta_1$ .

3. Suppose  $\sum_{i=1}^n \mathbf{x}_{ik}^2 = c_k, k = 1, \dots, p$  and  $\sum_{i=1}^n \mathbf{x}_{ik}\mathbf{x}_{il} = 0$  if  $k \neq l$ . For model

$$Y_i = \beta_1 \mathbf{x}_{i1} + \dots + \beta_k \mathbf{x}_{ip} + \varepsilon_i,$$

prove that the ridge regression estimator and the least squares estimator have the following relation

$$\hat{\beta}_{ridge} = \begin{pmatrix} c_1/(c_1 + \lambda) & 0 & 0 & \dots & 0 \\ 0 & c_2/(c_2 + \lambda) & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & c_p/(c_p + \lambda) \end{pmatrix} \beta_{LSE}$$

4. For a fixed  $\lambda$ , please find the solution to

$$\min_{\beta} \left\{ \sum_{i=1}^n w_i (Y_i - \beta_0 - \beta_1 \mathbf{x}_{i1} - \dots - \beta_p \mathbf{x}_{ip})^2 + \lambda(\beta_1^2 + \dots + \beta_p^2) \right\}$$

5. For [Data A](#), the first 50 columns are the predictors and the last the response.

Please fit a linear regression model to the data and use the ridge regression method to estimate the model.

- (a) Please calculate the estimators of the coefficients with  $\lambda = 0.1, 1, 10, 100, 1000$  respectively, and plot how the estimators change with  $\lambda$ .
- (b) Use leave-one-out CV to choose the tuning parameter  $\lambda$ .
- (c) Use the  $\lambda$  selected based on the above CV to estimate the model, predict the response in [Data B](#), and calculate the mean prediction error square.
- (d) Select  $\lambda$  by AIC and BIC, and 5-fold CV respectively [Hint: when AIC and BIC are used, please replace  $p$  in the penalties by  $\text{trace}\{\mathbb{X}(\mathbb{X}^\top \mathbb{X} + \lambda I)^{-1} \mathbb{X}^\top\}$ ]