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Consider a linear model with design matrix

$$X = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & \frac{1}{\sqrt{3}} \\ 0 & -\frac{2}{\sqrt{3}} \\ -1 & \frac{1}{\sqrt{3}} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ 0 & -\frac{2}{\sqrt{6}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \end{pmatrix}$$

and parameter vector  $\beta = (\beta_1, \beta_2)'$ . Denote the lasso and ridge regression estimators as solutions to the objective function

$$\min_{\beta \in \mathbb{R}^2} \left\{ \frac{1}{2} \|y - X\beta\|^2 + \lambda \|\beta\|_q^q \right\}$$

for  $q = 1$  or  $q = 2$ , respectively, as  $\hat{\beta}^{lasso}$  and  $\hat{\beta}^{ridge}$ . Suppose we have observed

$$y = \frac{1}{\sqrt{2}} \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}.$$

**Hint:** The following tasks have to be solved theoretically (see related theoretical exercise). In particular, *do not* use **R**, since the objective function in **glmnet** - and thus the tuning parameter  $\lambda$  - are scaled differently than above.

Find the missing numerical values with a precision of two decimals.

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(a) Find the smallest value  $\lambda_{\min}$ , say, for the tuning parameter  $\lambda$ , such that *at least* one of the lasso estimates  $\hat{\beta}_i^{lasso}$ ,  $i \in \{1, 2\}$ , equals exactly zero (with a precision of two decimal places).

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$\lambda_{\min} =$

0.5 ✓

(b) Calculate the corresponding ridge regression estimates for the value of  $\lambda_{\min}$  derived in (a) with a precision two decimal places.

1 of 1 point

(i) The corresponding ridge regression estimate of the first component is

$\hat{\beta}_1^{ridge} =$

0.33 ✓

(ii) The corresponding ridge regression estimate of the second component is

1 of 1 point

$\hat{\beta}_2^{ridge} =$

-0.58 ✓

(c) Suppose the underlying model is normal, that is

1 point

$$Y = X\beta + \epsilon$$

with  $\epsilon \sim N_2(0, 4 \cdot I_2)$ , where  $I_2$  denotes the 2-dimensional identity matrix.

Consider the mean squared error (MSE) of the ridge estimator of the first component  $\beta_1$  of  $\beta$ ,

$$\text{MSE}(\hat{\beta}_1^{ridge}) = \text{bias}(\hat{\beta}_1^{ridge})^2 + \text{Var}(\hat{\beta}_1^{ridge}).$$

Let  $\lambda = 1$  and denote by

$$(\beta_1^{lower}, \beta_1^{upper})$$

the interval of values for  $\beta_1$ , for which

$$\text{MSE}(\hat{\beta}_1^{ridge}) < \text{Var}(\hat{\beta}_1^{LS}),$$

i.e., the MSE of  $\hat{\beta}_1^{ridge}$  is smaller than the variance of the least squares estimator  $\hat{\beta}_1^{LS}$  of  $\beta_1$ . Find this interval (with a precision of two decimal places).

**Hint:** Consult Exercise 27(a) and Theorem I.4.25 of the lecture to derive the distribution of the estimator  $\hat{\beta}_1^{LS}$ .

$\beta_1^{lower} =$

-3 ✗ -3.46 ✓

$$\beta_1^{upper} =$$

1 point

3 ✖ 3.46 🔑

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