

Follow the instructions below carefully:

- Do not load any package but the default ones (like `base`).
- Load the `assignmentA_grpB.Rdata` file with the assignment and group numbers substituted for A and B, respectively, into R. The file is available via Brightspace. Verify that you are using the correct file!
- Verify that two objects, named `X` and `Y`, have been loaded into R's memory. These objects contain the data on the response vector (`Y`) and design matrix (`X`).
- Consider the linear regression model  $\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$  without (!) intercept and  $\boldsymbol{\varepsilon} \sim \mathcal{N}(\mathbf{0}_n, \sigma^2 \mathbf{I}_{nn})$ , to explain the variation in the response `Y` by a linear combination of the columns of the design matrix `X`.
- Evaluate the generalized ridge estimator of regression parameter  $\boldsymbol{\beta}$  of the linear regression model with target  $\boldsymbol{\beta}_0$  equal to the zero vector with the last  $p - B$  elements replaced by one and penalty parameter  $\boldsymbol{\Delta} = (1 - \rho)\mathbf{I}_{pp} + \rho\mathbf{1}_{pp}$  with  $\rho = 1/(41 + B)$  in which the group number is substituted for `B`.
- Estimate the error variance  $\sigma^2$  by  $\frac{1}{n}\|\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}(\boldsymbol{\beta}_0, \boldsymbol{\Delta})\|_2^2$ , where  $n$  denotes the sample size and  $\hat{\boldsymbol{\beta}}(\boldsymbol{\beta}_0, \boldsymbol{\Delta})$  the generalized ridge estimator with  $\boldsymbol{\beta}_0$  and  $\boldsymbol{\Delta}$  as above. This yields part one of the solution of the bonus exercise that is to be send in.
- In the remainder replace  $\boldsymbol{\Delta}$  by  $\lambda\boldsymbol{\Delta}$  with  $\lambda \in \{10, 20, 30, \dots, 200\}$ . Evaluate the mean squared error (MSE) of the obtained generalized ridge regression estimator, defined as:

$$\text{MSE}[\hat{\boldsymbol{\beta}}(\boldsymbol{\beta}_0, \boldsymbol{\Delta})] = \text{tr}\{\text{Var}[\hat{\boldsymbol{\beta}}(\boldsymbol{\beta}_0, \boldsymbol{\Delta})]\} + \{\mathbb{E}[\hat{\boldsymbol{\beta}}(\boldsymbol{\beta}_0, \boldsymbol{\Delta})] - \boldsymbol{\beta}\}^\top \{\mathbb{E}[\hat{\boldsymbol{\beta}}(\boldsymbol{\beta}_0, \boldsymbol{\Delta})] - \boldsymbol{\beta}\},$$

in which the true regression parameter  $\boldsymbol{\beta}$  is assumed to equal the zero vector and the error variance  $\sigma^2 = 1$ . Find the  $\lambda$  that minimizes the MSE of  $\hat{\boldsymbol{\beta}}(\boldsymbol{\beta}_0, \lambda\boldsymbol{\Delta})$ . This minimizing  $\lambda$  is part two of the solution that is to be send in.

- Send your answer in before 23:59 CET, October 7, 2024. Instructions for composing the email can be found in the pdf-file with information on the hand-in assignments.