## Problem Set 3 – Ridge and Lasso Regression

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## Simulating Data

To simulate data we must draw random variables with some prespecified distribution. In R for every distribution usually four functions are implemented which are useful for working with distributions and differ by their prefix: r (random), d (density), p (probability), and q (quantile). The prefix is combined with a name for the distribution, e.g. norm for the normal distribution: dnorm for the density of a normal distribution, pnorm for the probability, qnorm for the quantiles, and rnorm to draw from a normal distribution. (Check out the help page of the functions!)

Here we want to simulate a linear model of the form

$$y_i = x_i'\beta + \varepsilon_i, i = 1, \dots, n$$

with  $\beta$  a p-dimensional coefficient vector and  $x_i$  p-dimensional vector of regressors. In vector notation:

$$y = X\beta + \varepsilon$$

with y and  $\varepsilon$  n-dimensional vectors and X a  $n \times p$ -design matrix.

Here the task is to simulate from this model, where we assume that the coefficient vector  $\beta$  has s entries equal to one and all others are zero.

- Set n = 100, p = 10, s = 3
- Create the coefficient vector  $\beta$ . Useful functions: c(), rep()
- Simulate a design matrix and the error. Useful functions: matrix(), rnorm
- Construct the model from above. Useful function: %%\* for matrix multiplication

## Ridge Regression I

- Estimate a ridge regression on simulated data from Exercise 1. Useful function: glmnet from the package glmnet with default  $\alpha = 1$ . Also check out the option lambda in glmnet and the function cv.glmnet to perform cross-validation to determine  $\lambda$ .
- Simulate new data from the same model and make predictions both in- and out-of-sample. Calculate the MSE for the predictions (also for the in-sample fit). Useful function: *predict*
- Repeat the previous steps with different settings on n, p, and s.
- Compare the results with ols regression!

## Lasso Estimation I

- Redo the calculations from above but with Lasso with varying n, p, and s. Hint: Set option  $\alpha$  in glmnet to 0
- Compare the results! (In particular compare a "sparse" with a "dense" setting)
- The package *hdm* contains the function *rlasso* which determines the penalization parameter by some theoretical grounded method. Look up the function in the man pages and / or vignette and analyze now the data set using this function. Compare the results.