

# Machine Learning

## Exercise 2

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### 1 Linear Ridge Regression

On the course webpage there are two simple data sets `dataLinReg1D.txt` and `dataLinReg2D.txt`. The data files can be plotted, e.g., using `gnuplot` with `plot 'dataLinReg1D.txt'` or `splot 'dataLinReg2D.txt'`. Each line contains a data entry  $(x, y)$  with  $x \in \mathbb{R}^d$  and  $y \in \mathbb{R}$ ; the last entry in a line refers to  $y$ . Compute and report the optimal parameters  $\beta$  for a linear Ridge regression model (just linear features) for both data sets. Tips:

- Write a routine that loads a data file and returns a matrix  $X$  containing all  $x_i$  as rows, and a vector  $y$  containing all  $y_i$ .
- Write a routine that takes the raw  $X$  as input and returns a new  $X$  with a '1' pre-pended to each row. This routine simply computes the "linear features" including the constant 1. This routine can later be replaced by others to work with non-linear features.
- Write a routine that returns the optimal  $\beta$  from  $X$  and  $y$ .
- Generate some test data points (along a grid) and collect them in a matrix  $Z$ . Apply routine b) to compute features. Compute the predictions  $\hat{y} = Z\beta$  (simple matrix multiplication) on the test data and plot it.

### 2 Non-linear features

Test regression with quadratic features on the data sets `dataQuadReg1D.txt` and `dataQuadReg2D.txt`. Compute and report the optimal parameters  $\beta$  for both data sets. In principle, all you have to do is replace routine b) above (see slide 02:8).

### 3 Cross-validation

Implement cross-validation (slide 02:18) to evaluate the *prediction error* of the linear and polynomial regression method for `dataQuadReg2D_noisy.txt`. Report 1) the squared error when training on all data (=training error), and 2) the mean squared error  $\hat{\ell}$  from cross-validation.

Repeat this for different Ridge regularization parameters  $\lambda$ . (Ideally, generate a nice bar plot of the generalization error, including deviation, for various  $\lambda$ .)