

# ML: Regularization - Exercises

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Please write down to solution of the exercises in a concise but comprehensible way (including intermediate results). Numerical results should be accurate to 4 digits. Sketches should be correct qualitatively.

At least 75% of the exercises have to be solved satisfactorily. Due time is May 1st, 2018. The first exercise should be handed in via E-Mail to [Josef.Buergler@hslu.ch](mailto:Josef.Buergler@hslu.ch).

## Exercise 1: Simple Example from Lecture (1 Point)

Add a section to the Jupyter Notebook `Ridge+LASSO_Regression.ipynb` that finds the optimal regularization hyperparameter  $\lambda$ . In order to do so, we will use LASSO regression and proceed as follows:

1. For every value  $\lambda_0 = 0$ ,  $\lambda_1 = 0.01$  and from there always double  $\lambda$  until you have reached a point, where the cross validation error starts to grow again (due to overfitting). So  $\lambda_{12} = 10.24$  is the 12th value of  $\lambda$  but it need not be the largest in this example.
  - a) Using the training set compute

$$\theta_k = \underset{\theta}{\operatorname{argmin}} (J_{\text{train}}(\theta) + \lambda \Omega(\theta)) \quad \text{where } k \text{ is one of the numbers } 1, 2, 3, \dots$$

Then compute the error in the cross validation set

$$(J_{\text{cv}}(\theta_k) + \lambda \Omega(\theta_k))$$

- b) Plot these errors for this value of  $\lambda$ .
2. Once you notice the growing tendency of the error in the cross validation set depending on the regularization parameter  $\lambda$  stop after the next larger value of  $\lambda$
  3. Find the value for  $\lambda$  which leads to the smallest cross validation error.
  4. print the corresponding coefficients of the regression polynomial and draw it.

## Exercise 4: Real Word Examples (1 Point)

Fill in the missing sections in the Real World Example of `regularization_gaps.ipynb` and make the Jupyter Notebook runnable. Hand it in via E-Mail to Tim vor der Brück (E-Mail-Address: [tim.vorderbrueck@hslu.ch](mailto:tim.vorderbrueck@hslu.ch)) no later than Tuesday, May 1st, 2018.

**Happy Machine Learning !**