# Machine Learning 2014: Project 1 - Regression Report

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# **Experimental Protocol**

As first analysis, each feature was plotted against the response variable in order to spot possible non-linearities and non-significant parameters.

Afterwards, the parameters were estimated as described in the next sections.

#### 1 Tools

Both the analysis and the estimation were carried out in *MATLAB*. Git was used to keep track of progress and versioning of our algorithm.

## 2 Algorithm

Ridge regression has been used to estimate the model's parameters. The optimal penalizing parameter  $\lambda$  was chosen by minimizing the prediction error estimated using *Cross Validation* (the number of subsets was set to  $\sqrt{n}$ , where n is the sample size.) (Figure 1).

### 3 Features

In a first step, we exponentiated some of the given features (e.g.  $x_i \to x_i^a$  with  $a \in [-2:0.5:5]$  and  $a \neq 0$ ).

Next, we combined some of the transformed features in order to consider possible interactions. Hence, our final model has the form  $y=\beta_0+...+\beta_ix_i^a+...+\beta_zx_j^bx_k^c+....$ 

### 4 Parameters

Cross Validation was used to find how to transform and combine the parameters. In details:

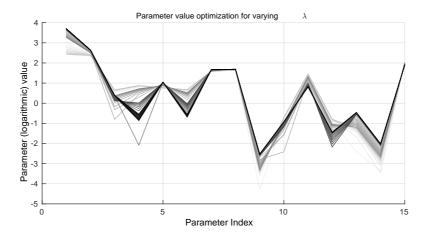


Figure 1: Parameter evolution during prediction error minimization. Darker values are closer to the optimal  $\lambda$ .

- $\bullet$  For each feature  $x_i$ , the exponent a is chosen so that the estimated prediction error is minimized.
- For each pair of feature  $x_i$  and  $x_j$ , the combination term  $x_ix_j$  is included only if the estimated prediction error decreases by a certain percentage.

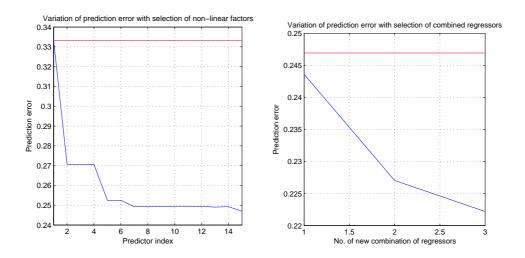


Figure 2: Error reduction through feature transformation and combination.

### 5 Lessons Learned

Other strategies that we tried include feature discarding<sup>1</sup>, response transformation and different feature transformations (e.g.  $log(\cdot)$ ). The resulting worse performance is most likely given by a loss of information (in the case of the discarded feature) or wrong model assumptions.

<sup>&</sup>lt;sup>1</sup>We computed the parameters  $\beta$  on normalized data and excluded the features corresponding to the lowest  $|\beta|$ .