Machine Learning Exercise on LDA

In this assignment you will estimate cognitive states from electroencephalogram (EEG) data.

Data Electroencephalography (EEG) data was recorded during a copy-spelling BCI Experiment at the Berlin BCI group (http://www.bbci.de).

The data set consists preprocessed EEG data $X \in \mathbb{R}^{5 \times 62 \times 5322}$ and stimulus labels $Y \in \mathbb{R}^{2 \times 5322}$ during a copy-spelling paradigm with a P300 speller. The data matrix X contains 5 selected time windows of EEG activity at 62 electrodes after a visual stimulus was presented on the screen in front of the subject. If the first row of Y is 1, the stimulus was a target stimulus, if the second row of Y is 1, the stimulus was a non-target stimulus.

- 1. How well can we predict the cognitive state from the EEG data?
 - a) Implement a nearest centroid classifier which predicts +1 if

$$\|\mu_{+} - x\|_{2} < \|\mu_{-} - x\|_{2} \tag{1}$$

where $\mu_{+/-}$ are the means of the positive (respectively negative) class, that is $\mu_{+} = 1/N_{+} \sum_{i}^{N_{+}} x_{+i}$ and x_{+i} is the *i*th data point of the + class.

b) Implement a linear discriminant analysis (LDA) classifier; optimize a vector $w \in \mathbb{R}^{5\cdot 62\times 1}$ such that

$$\underset{w}{\operatorname{argmax}} \frac{w^{\top} S_B w}{w^{\top} S_W w} \tag{2}$$

where

$$S_B = (\mu_+ - \mu_-)(\mu_+ - \mu_-)^{\top}$$
(3)

$$S_W = 1/N_+ \sum_{i} (x_{+i} - \mu_+)(x_{+i} - \mu_+)^{\top}$$

$$+ 1/N_- \sum_{j} (x_{-j} - \mu_-)(x_{-j} - \mu_-)^{\top}$$

$$(4)$$

Train both classifiers on the 5.62 = 310 concatenated (across time and electrodes) features to predict target stimuli from the EEG data. Train both classifiers on 70% of the data and test it on 30% of the data. Select the data points for training and test randomly, but make sure they do not overlap!

- 2. Compare the prediction accuracy for each classifier. For both classifiers: plot histogram of classifier outputs for targets/non-targets using pylab.hist (see fig. 1)
- 3. Estimate the **generalization performance** of each classifier. Test each classifier on the training set data and on the test data data. Compare the prediction accuracies (see fig.2)

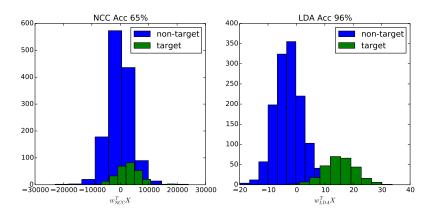


Figure 1: Left: Classifier outputs for nearest-neighbor classifier. Right: Classifier outputs for linear discriminant classifier.

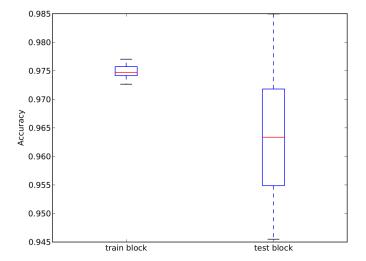


Figure 2: Classification accuracy for training and test blocks in 10-fold cross-validation.