

## Programming Exercise 2

Due on 27th November

Pick a binary classification dataset from the LIBSVM repository:

<http://www.csie.ntu.edu.tw/~cjlin/libsvmtools/datasets/>

### Tasks:

1. **Decision Trees:** Partition the dataset into a training and a testing set. Run a decision tree learning algorithm using the training set. Test the decision tree on the testing dataset and report the total classification error (i.e. 0/1 error). Repeat the experiment with a different partition. Plot the resulting trees. Are they very similar, or very different? Explain why.  
*Advice: it can be convenient to set a maximum depth for the tree.*
2. **Support Vector Machines:** Run SVM to train a classifier, using radial basis as kernel function. Apply cross-validation to evaluate different combinations of values of the model hyper-parameters (box constraint  $C$  and kernel parameter  $\gamma$ ). How sensitive is the cross-validation error to changes in  $C$  and  $\gamma$ ? Choose the combination of  $C$  and  $\gamma$  that minimizes the cross-validation error, train the SVM on the entire dataset and report the total classification error.  
*Advice: use a logarithmic range for  $\gamma$ .*
3. **Neural Networks:** Train a Multi-Layer perceptron using the cross-entropy loss with  $\ell_2$  regularization (weight decay penalty). In other words, the activation function equals the logistic function. Plot curves of the training and validation error as a function of the penalty strength  $\alpha$ . How do the curves behave? Explain why.  
*Advice: use a logarithmic range for hyper-parameter  $\alpha$ . Experiment with different sizes of the training/validation sets and different model parameters (network layers).*

Attach the source code for each section. You are free to use the programming language/library of your choice. We recommend, scikit-learn <https://scikit-learn.org/stable/index.html>