# Machine Learning for Applications in Computer Vision: Week 1

Date: Friday, 17. April 2015

Please work in groups of 2–3 people. We will check your solutions next week after the lecture. Please be prepared to present your solution and explain the code.

## Download the MNIST Dataset

Please download MNIST Digit Recognition Dataset from: http://yann.lecun.com/exdb/mnist/Download all 4 .gz files.

# Coding

Please code in Python.

You can use scikit-learn Python package for machine learning algorithms.

The documentation is on http://scikit-learn.org

# Exercise 1: Dataset Preparation

- 1. Once you have downloaded the datasets extract the images and labels.
- 2. Use the image pixels as features. Reshape the  $28 \times 28$  images to 1D vectors (with length 784).

You can ask the tutors how to read database in case you could not do it yourself.

# Exercise 2: Support Vector Machines

- 1. Make yourself familiar with  ${\tt scikit-learn}$  and the SVM tool
- 2. Load the training data (images and labels) into your program. You might have to unpack the data set and convert it to a reasonable format.
- 3. Train the default SVM on the first 1,000 samples.
- 4. Check the model you trained against the test set (NEVER do training with the test set!).
- 5. Tune the parameters (particularly kernel) to find a good set of hyperparameters.
- 6. Now run your training on the entire dataset and check it against the test set.
- 7. Report the methods used and your results (cf. lecture slide).

# Exercise 3: Decision Trees and Random Forests

#### 1. Decision Trees

- (a) Train a decision tree with default parameters and output the classification score for the test images.
- (b) Tune the parameters of the decision tree and find the best hyperparameters:
  - \* criterion (gini and entropy)
  - \* max depth
  - \* max\_features
- (c) Visualize the constructed tree using pydot.
- (d) Visualize the pixel importances on a  $28 \times 28$  image.

### 2. Random Forests

- (a) Construct a Random Forests classifier with default parameters and output the classification score for the test images.
- (b) Tune the parameters of the decision tree and find the best hyperparameters:
  - \*  $n_{estimators}$
  - \* criterion (gini and entropy)
  - \* max depth
  - \* max\_features
- (c) Visualize the pixel importances on a  $28 \times 28$  image.

## 3. Cross-Validation

Apply cross-validation on a decision tree and random forests using the training images and output the mean scores. You can use the best hyperparameters.