DATA.ML.200 Pattern Recognition and Machine Learning

Exercise Set 2: Multi-layer Percepteron (MLP)

1. **pen&paper** Count the number of parameters in a neural network (2 points)

a) Consider the conventional full-connected neural network architecture of Figure 1. Suppose our inputs are 64×64 RGB images of two different traffic signs.

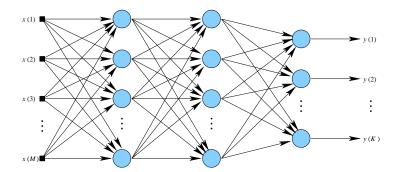


Figure 1: Vanilla neural network.

Let the network structure be the following:

- The input is "flattened" $3 \times 64 \times 64$ -dimensional
- On the 1st layer there are 100 nodes (marked in blue)
- On the 2nd layer there are 100 nodes (marked in blue)
- On the 3rd (output) layer there are 10 nodes (marked in blue; one for each class)

Compute the number of parameters (weights) in the net.

- b) An old rule of thumb states that the number of training samples should be at least 5 times the number of coefficients. Compute the desired sample size based on this rule for (a).
- 2. **python** Load Traffic sign data for deep neural network processing.

Download an extended version of the two class German Traffic Sign Recognition Benchmark (GTSRB) dataset (GTSRB_subset_2.zip) from the course Moodle page.

This time, images are in color and there are about 400 from both classes. Split your data into two parts - 80% for training and 20% for testing. Note that there are ready-made functions for that.

3. **python** Define the network in Keras.

Define the above network in your code. You may in the beginning reduce the number of neurons from 100 to 10 in the two layers.

4. **python** Compile and train the net. (8 pts)

Use the following parameters:

- Loss: binary/categorial crossentropy
- Optimizer: stochastic gradient descent (SGD)
- Number of epochs: 10

Compute the test set accuracy for your network.

NOTE: if you have GPU in your laptop/desktop report training times per epoch for the both GPU training and CPU training. In the case of GPU you may need to adjust the *batch_size* option to make sure the data fits to your GPU memory.