

## SYDE552/BIOL487 Assignment: Regression, Classification, and Neural Networks

15 marks total

Due Feb 3

**This assignment should be done in pairs.** This assignment can be done in Matlab or Python. You will need .mat or .pkl files *perceptron-data*, *regression-data*, and *backprop-data*, from Learn, in the file *assignment1-data.zip*.

Submit (via Learn):

- A PDF containing the requested figures and answers.
- Your code.

1. [5 marks] Load the *perceptron-data* file. Implement the perceptron learning rule. Train a perceptron to discriminate the *vectors* with *labels* 0 from those with *labels* 1. Present the samples to the perceptron in random order. Plot the fraction correct vs. the step number for 10 separate runs (in a single plot). Stop each run after 300 steps.
2. [5 marks] Load the *regression-data* file. Plot 20 Gaussian functions with peak values of 1, widths of  $\sigma=0.2$ , and centres evenly spaced from -1 to 1. Calculate least-squares optimal regression weights for these basis functions, for approximation of the training data *trainx* and *trainy* (where the latter is a function of the former). Use the regularizer that we discussed in class. Plot the training data along with regularized regression models for the following values of  $\lambda$  ( $10^{-8}$ ,  $10^{-5}$ ,  $10^{-2}$ , and 10). (Plot the models' output in steps of 0.05.) Calculate the sum-squared error for both training and test data, for each case, and plot as a function of  $\ln(\lambda)$ .
3. [5 marks] Load the *backprop-data* file. Implement a 2-layer feedforward network with 10 hidden-layer units and a single output unit. Implement this from scratch, i.e. don't use a neural network library. The hidden units should have nonlinearity  $z = \tanh(a)$  and the output unit should have nonlinearity  $z = 1/(1 + e^{-a})$ . Train the network using gradient descent and backpropagation to discriminate the *vectors* with *labels* 0 from those with *labels* 1. Plot contours of the network output for inputs over the range  $[-8, 8]$ , along with the training data. Use different colours for data points that the network assigns to different classes. Include plots for the network before training and after 1000 iterations. [Bonus: 1 mark] Describe a procedure for estimating  $\partial E_n / \partial w_{ji}$  for each weight in a network using small perturbations to the **activation** of each node (i.e. an alternative to backpropagation). Roughly how much computation time would this take relative to backpropagation?

