## **Programming Assignment #1**

## Logistic Regression

## DUE: February 11th at 11:59 PM

**Problem:** Implement the **Logistic Regression** algorithm. The code skeleton as well as data sets for this assignment can be found on e-Learning.

**Data Set:** The data set (in the folder ./data/) is obtained from the UCI Repository and are collectively the MONK's Problem. These problems were the basis of a first international comparison of learning algorithms<sup>1</sup>. The training and test files for are named monks-3.train and monks-3.test. There are six attributes/features (columns 2–7 in the raw files), and the class labels (column 1). There are 2 classes. Refer to the file ./data/monks.names for more details.

- a. (**Reporting Error Rates**, 30 points) For the following learning rates {0.01, 0.1, 0.33} and iterations {10, 100, 1000, 10000} fit Logistic Regression Models using the MONKS-3 train set and compute the average training and test errors. Report what the best parameters are for the model.
- b. (Saving the Model, 50 points) Retrain the models on the best parameters and save it as a pickle file in the format 'NETID\_lr.obj'. The object file will be loaded for grading and the class functions will be individually tested.
- c. (scikit-learn, 10 points) For monks datasets, use scikit-learns's default Logistic Regression Algorithm<sup>2</sup>. Compute the train and test errors using sklearn's Logistic Regression Algorithm. Compare the results with the version you implemented and speculate on why there is a difference in performances between the two algorithms. Do not change the default parameters.
- d. (**Plotting curves**, 10 points) For each of the learning rates, fit a Logistic Regression model that runs for 1000 iterations. Store the training and testing loss at every 100th iteration and create three plots with epoch number on the x-axis and loss on the *y*-axis.

<sup>1</sup>https://archive.ics.uci.edu/ml/datasets/MONK's+Problems

<sup>&</sup>lt;sup>2</sup> https://scikit-learn.org/stable/modules/generated/sklearn.linear\_model.LogisticRegression.html