

EOSC 410/510 Assignment 4:

[12 points in total; grading scheme is given in square brackets]

Objective:

The given data file (data.mat or data.csv) contains two predictors x_1 and x_2 and target data y , each with 80 observations in time. Following the guidelines below, develop a MLP NN model of y as a function of x_1 and x_2 . Briefly describe the approach taken. Also given are the forecast predictors: x_{1test} and x_{2test} (each with 40 observations in time), i.e. as a separate testing input dataset. Your task is to predict y_{test} using these test predictors and submit the results for **y_{test} as a spreadsheet (e.g. csv file)** together with your assignment. The TA will calculate the root mean squared error (RMSE) between your modelled y_{test} and true y_{test} which is not given to you. The smaller the RMSE is between your modelled y_{test} and true y_{test} , the higher will be your mark for this question (see point 5 below).

Note: no need to standardize the data. Consider that this data is already given in the standardized form.

Guidelines:

- 1) Set the model design, i.e. divide the data into training/validation and testing, select the model parameters etc. Provide some rationale for your choice. *[2 points for the answers and rationale.]*
- 2) Provide several runs of the model (try ≥ 10 runs). For each run, use different NN model design, i.e. different number of hidden neurons per one or two hidden layers. Plot the results of the best model for each run, i.e. how many hidden layers and how many hidden neurons per layer *[2 points for the plots]*.
- 3) Plot the RMSE, calculated over your testing sample, for each model run as well as RMSE from the model ensemble mean *[1 point for the plot(s) with RMSE of model runs and 1 point for the ensemble-mean RMSE]*
- 4) Plot the results of the ensemble runs and ensemble mean for your testing sample (y_{model} vs time), as well as the target (true) y for your testing sample. *[3 points for the plot]*.
- 5) Predict y_{test} as an ensemble mean of your trained model runs with x_{1test} and x_{2test} as an input and submit y_{test} values as a **spreadsheet**. *[1 point for a reasonable estimate; additional 2 points according to how small your RMSE is. These points will be linearly scaled relative to the students' RMSE between their modelled and true y_{test} . E.g. student with the smallest RSME will get 2 points, and student with the largest RMSE will get 1 point, while the others will get their points scaled accordingly between $min=1$ and $max=2$].*