INM427 Neural Computing

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**A Comparison Between Python and MATLAB for Binary Classification in Predicting Bank Churn**

1. **Initial Analysis of the Data Set**

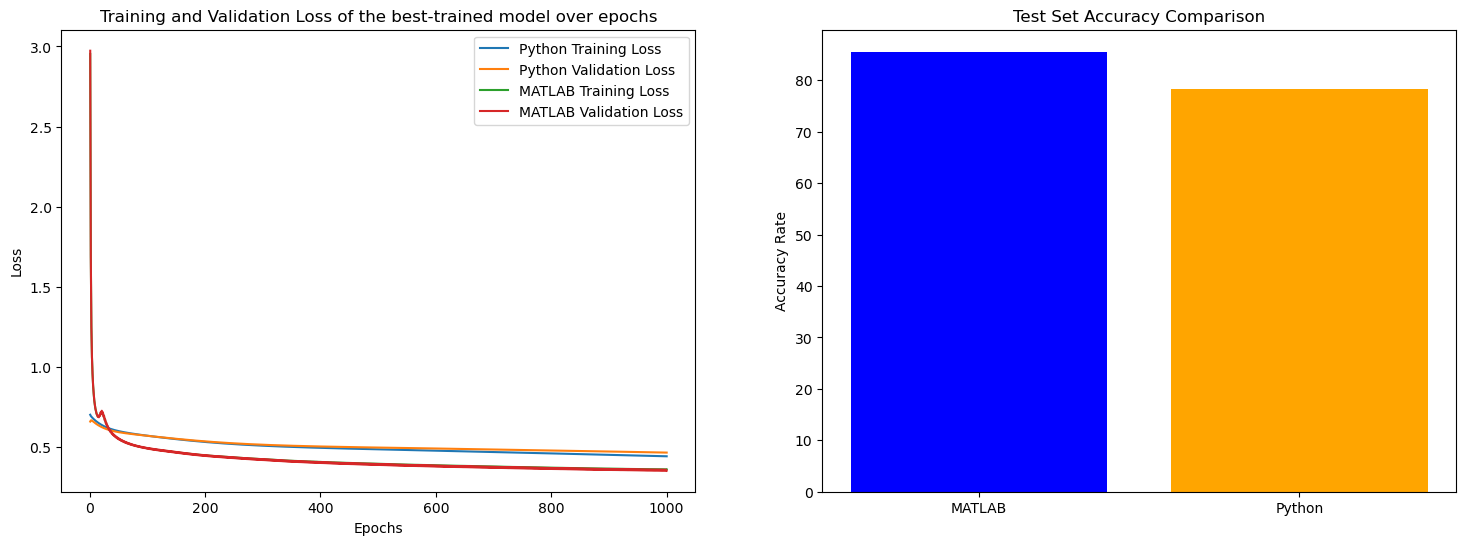
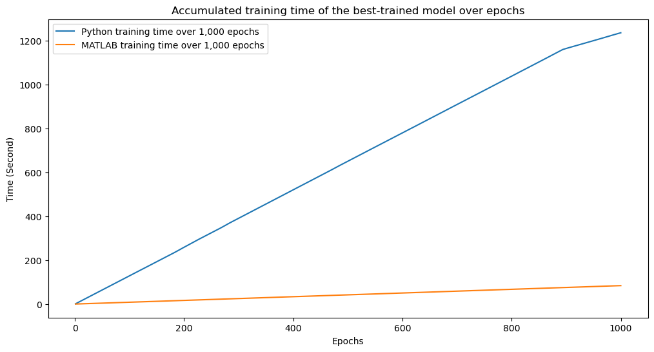
The data set [1] from Kaggle has 14 features including a target and 165,034 samples.10 features remained after removing 3 features deemed unnecessary for classification. Normalisation was applied to the features for better accuracy. After checking the histogram of each feature and finding that some features and the target have unbalanced data, SMOTE was applied to ensure that the training set is balanced.

1. **A Comparison Between Python and MATLAB in Model Implementation**

A multilayer perceptron was implemented for this task. However, Support Vector Machines can be used for this task since it was originally made for binary classification [2].ReLU activation was applied to the hidden layer nodes, and sigmoid to the output layer nodes. Grid search was implemented to find the best combination of the learning rate and the number of hidden neurons. The learning rate range is [0.01, 0.1, 0.3, 0.4] and the number of neurons is [10, 25, 50, 100, 200]. Although the number of hidden neurons should lie between the size of the input layer and the size of the output layer [3], there is another suggestion to use more than 100 neurons in the network [4]. Our model performed better with a larger number of neurons and a higher learning rate in our experiment (Table 1).

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| --- | --- | --- |
| **Programming Language** | **Python** | **MATLAB** |
| Optimised Hyperparameters  from Grid Search | Learning rate: 0.2  The number of neurons: 200 | Learning rate: 0.3  The number of neurons: 25 |
| Training Speed (Second) | 1235.00 | 84.14 |
| Test Set Accuracy of the Final Model | 78.22% | 85.53% |
| Precision, Recall, and F1 of the Final Model | 48.78%, 69.30%, 57.26% | 69.81%, 53.48%, 60.57% |

**Table 1**



**Figure 1 Figure 2 Figure 3**

MATLAB is faster than Python when training the best model because MATLAB specializes in matrix computation [5] (Figure 1). The loss graph is one way to compare the training speed between MATLAB and Python as well. MATLAB started with a larger loss when training and validating in the first epoch (Figure 2). Then, MATLAB decreased the loss after passing 14 epochs whereas Python started with relatively low losses, and the magnitude of loss reduction is relatively small compared to MATLAB. We can say MATLAB converges faster than Python. The hypothesis behind this conclusion is that the initial weights in MATLAB are set higher than in Python in the first epoch. Despite the large weight setting, MATLAB's optimizer finds the local minima faster than Python in non-convex scenarios. Additionally, differences in the randomness utilized during training, such as weight initialization, may contribute to disparities despite similar settings between MATLAB and Python. MATLAB's faster performance may stem from its neural network with 25 neurons, compared to Python's 200, facilitating quicker forward and backward propagations. MATLAB achieves higher test accuracy than Python (Figure 3), influenced by differing hyperparameters, loss reduction rates during training, and variations in weight settings.

**References**

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