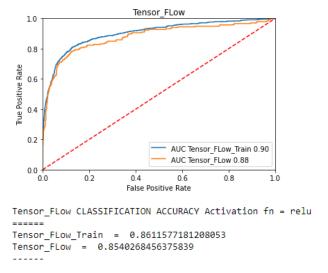
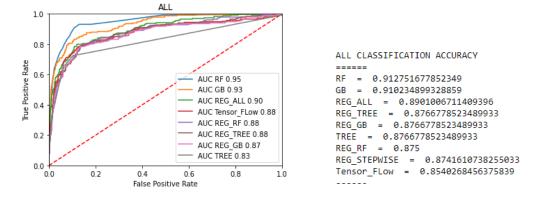
MSDS 422 Machine Learning: Assignment 4 – Neural Networks

In addition to our models that we had created in prior assignments (decision tree, random forest, gradient boosting, logistic regression models using different set of variables), we have created an additional model using neural networks. We created an ROC curve of our neural network model to give a sense for how well it performs at predicting loan default. We limited our model to use the variables that our Gradient Boosting model found to be predictive. Below is the ROC curve for our neural network model:



Based on these metrics, this appears to be a very strong model for predicting loan defaults as the accuracy against the training dataset is not that far off from the accuracy of the test dataset. Here is how our neural network model (Tensor_Flow) compares to the other loan default predicting models:



While the Tensor Flow model rates high in terms of AUC (.88) and Accuracy (.85), it is not amongst the top models in these rankings. We would recommend going with the Regression Model with Tree-based variables. Even though it did not have the best AUC or accuracy, regression models tend to be easy to implement. Additionally, the fact that it has a lesser number of variables (than say the Regression Model with ALL variables) makes it easy to work with.

We also built a model for predicting the loss amount using neural networks. Below are our model rankings:

```
ALL LOSSES MODEL ACCURACY
======

GB = 2422.2769707799052

RF = 2951.0813707195125

REG_ALL = 3634.676632630516

REG_TREE = 4239.965700169976

REG_RF = 4358.06472084949

REG_GB = 4358.06472084949

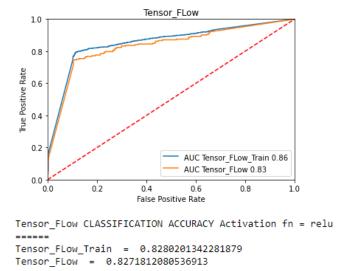
REG_STEPWISE = 4796.838013306385

TREE = 5722.46895603711

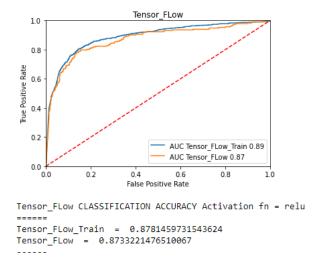
Tensor_FLow = 9467.696005753261
```

With its Root Mean Square Error of 2422.28, we would elect to go with the Gradient Boosting method to predict loss amounts. Even though regression models can be easier to work with, the difference in RMSE between Gradient Boosting and the closest regression model (Regression with all variables) is fairly large. Our Tensor Flow model did not do a great job at predicting loss amounts based on its accuracy.

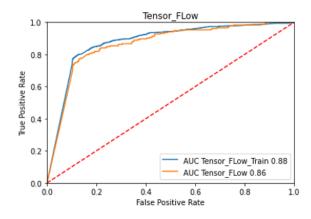
BINGO BONUS: We'll explore how these settings impacted the neural network results. As a baseline, here how the model performed with 200 epochs, relu activation function, and only 1 layer and no dropout:



<u>Adding hidden layer</u> – this seemed to smoothen our curve somewhat while also increasing the AUC for both datasets. Also saw significant increases in the accuracy:



Adding dropout: saw a very slight decrease in AUC for both datasets and small reduction in accuracy:

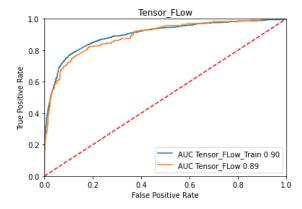


```
Tensor_FLow CLASSIFICATION ACCURACY Activation fn = relu =====

Tensor_FLow_Train = 0.8687080536912751

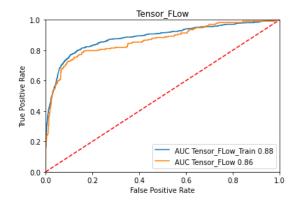
Tensor_FLow = 0.8624161073825504
```

<u>Increase epochs from 200 -> 800:</u> jump in AUC as well as accuracy.



Tensor_FLow CLASSIFICATION ACCURACY Activation fn = relu
=====
Tensor_FLow_Train = 0.8873741610738255
Tensor_FLow = 0.8791946308724832

Selu Activation function: dip in AUC as compared to relu function

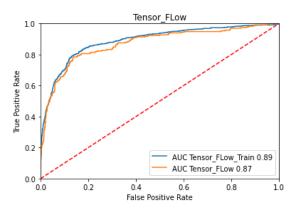


Tensor_FLow CLASSIFICATION ACCURACY Activation fn = selu =====

Tensor_FLow_Train = 0.881501677852349

Tensor_FLow = 0.8825503355704698

Sigmoid Activation function: slightly less AUC and accuracy then relu:



Tensor_FLow CLASSIFICATION ACCURACY Activation fn = sigmoid
======
Tensor_FLow_Train = 0.8733221476510067
Tensor_FLow = 0.8741610738255033