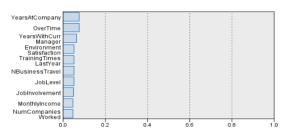


Artificial Neural Network

Figure 2. Importance of Variables According ANN analysis



Artificail Neural Network was implemented on the dataset using Python's Keras library and TensorFlow as the backend. Without any data cleaning the important variables and confusion matrix by ANN was given. The accuracy came out to be 85.33%. To improve the accuracy, a number of steps were taken 1. data cleaning

- 2. adding derived information
- 3. Using dummy variables and dummy traps
- 4. Hyperparameters
- 5. Training the neural network with K fold cross validation
- The accuracy increased and came out out to be 97.5%

Figure 3. ANN Confusion Matrix

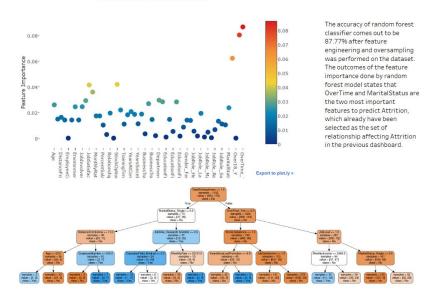
Comparing \$N-Nattrition with Nattrition 'Partition' Correct 922 90.39% 384 85.33% 9.61% 14.67% Wrong 1,020 450 Total Coincidence Matrix for \$N-Nattrition (rows show actuals) 'Partition' = 1_Training 0.000000 852 1.000000 79 'Partition' = 2_Testing 0.000000 0.000000 352 1.000000

print("Best accuracy: ",max)

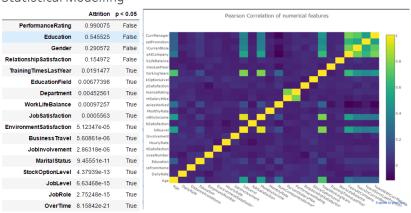
Best accuracy: 0.974999979138

Random Forest

Random Forest Feature Importance



Statistical Modelling



After performing EDA, the next logical step was to figure out which of the attributes have a statistically significant categorical relationship with Attrition, which is our target variable. To do this, Chi-square test was performed and the p-values of the attributes were obtained. Attributes having p-values > 0.05 had a non-significant categorical relationship and the others had a statistically significant categorical relationship

From te correlation plot, it was observed that quite a lot of columns were poorly correlated with one another. Generally, when making a predictive model, it would be preferable to train a model with features that are not too correalted with one another so that we do not have to deal with redundant features.