# Random Forest

December 13, 2017

## 1 Random Forest

Created on Thu Dec 7 09:38:42 2017

Author: tpin3694

#### 1.0.1 Import all the packages

```
In [1]: import pandas as pd
    import logging
    from sklearn.model_selection import train_test_split, cross_val_score
    from sklearn.ensemble import RandomForestClassifier
    from sklearn.metrics import confusion_matrix, roc_curve
    import scikitplot as skplt
    import glob
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
    from imblearn.combine import SMOTETomek
    from collections import Counter
    import random
```

# 1.0.2 Define data reader function, to return the orederd\_product\_key and campaign\_key columns from our data

### 1.0.3 Checks for NaNs and if any fraud actually exists in the file

#### 1.0.4 Uses SMOTE and Tomek Links to under and over sample to combat the class imbalance

### 1.0.5 Plot socres for different trees and save the plot in the local file

#### 1.0.6 Calculate the model recall, precision, f1 and auc value

```
In [6]: def data_scorer(model, features, labels, folds):
    recall = np.mean(cross_val_score(model, X = features, y = labels, cv = folds, score
    precision = np.mean(cross_val_score(model, X = features, y = labels, cv = folds, score
    accuracy = np.mean(cross_val_score(model, X = features, y = labels, cv = folds, score
    f1 = np.mean(cross_val_score(model, X = features, y = labels, cv = folds, scoring
    auc = np.mean(cross_val_score(model, X = features, y = labels, cv = folds, scoring
    return accuracy, recall, precision, f1, auc
```

#### 1.0.7 Define test model with features and labels

#### 1.0.8 Set a seed for reproducability

```
In [8]: random.seed(123)
```

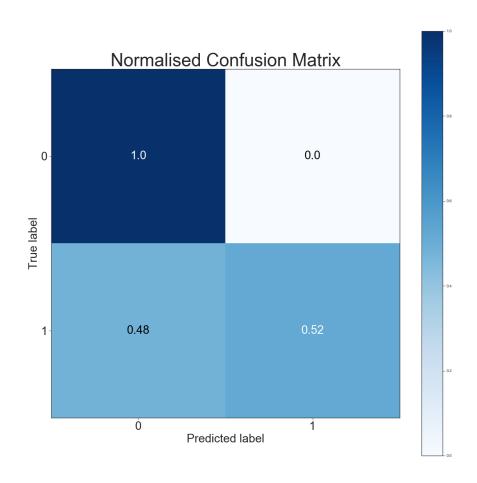
## 1.0.9 Setup File Structure and create a data list

1.0.10 For each file in the file list, the main process would be: read in data, then check for any fraud, for each site, split out into test and train data set, and seperate labels and features, train classifier, define variable importances, remove the umimportant variables, retrain model, check there are enough minority cases to run 10 fold cross-validation and return matrix, finally calculate prediction probabilities

```
In [24]: import warnings
         warnings.filterwarnings('ignore')
         for file in files:
             filename = file.strip(file_dir)
             site = data_reader(file)
             file_checks = file_checker(site)
             if file_checks[1] > 0:
                 train, test = train_test_split(site, test_size = 0.5, random_state = 42)
                 train_x = train.loc[:, train.columns != "fraud_status"]
                 train_y = train['fraud_status']
                 test_x = test.loc[:, test.columns != "fraud_status"]
                 test_y = test['fraud_status']
                 classifier = RandomForestClassifier(80)
                 classifier.fit(train_x, train_y)
                 importances = pd.Series(classifier.feature_importances_, name = "Importances"
                 var_importance = pd.concat([pd.Series(test_x.columns, name = "Names"), importance
                 var_importance = var_importance.sort_values(by = "Importances", ascending = Fe
                 var_importance = var_importance.drop("index", 1)
                 sns.set_style('ticks')
                 fig, ax = plt.subplots()
                 fig.set_size_inches(w = 20, h = 20)
                 sns.barplot(y = "Names", x = "Importances", data = var_importance, ax = ax)
                 sns.despine()
                 fig.savefig("/Users/sunmengnan/Desktop/FD/plot/" +
                             filename+ "variable_importance_no_smote.png")
                 unimportant = var_importance.drop(var_importance[var_importance.Importances <
                 unimportant = unimportant["Names"]
                 unimportant.to_csv("/Users/sunmengnan/Desktop/FD/result/"+
                                    filename + "important_variables.csv",index = True)
                 train_x = pd.DataFrame(train_x)
                 train_x.columns = train.loc[:, train.columns != "fraud_status"].columns
                 train_x = train_x[unimportant]
                 train_x = train_x.as_matrix()
                 test_x = test[unimportant].values
                 test_y = test['fraud_status'].values
                 clf = RandomForestClassifier(80, n_jobs = -1, random_state=42)
                 clf.fit(train_x, train_y)
                 clf.predict(test_x)
```

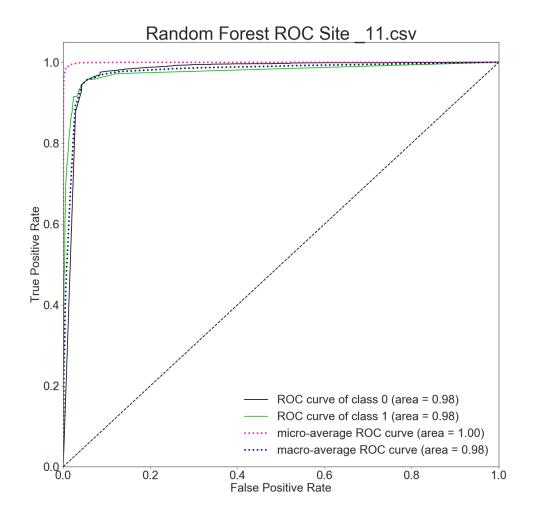
```
if sum(test_y) < 10:</pre>
        cv = sum(test_y)
    else:
        cv = 10
   accuracy, recall, precision, f_score, auc = data_scorer(clf, test_x, test_y,
   results = [filename, accuracy, recall, precision, f_score, auc]
   y_prob = clf.predict_proba(test_x)
   y_pred = clf.predict(test_x)
   fig, ax = plt.subplots()
   fig.set_size_inches(w = 20, h = 20)
    skplt.metrics.plot_roc_curve(test_y, y_prob, title = "Random Forest ROC Site")
                                 ax = ax, title_fontsize=46, text_fontsize = 30)
   fig.savefig("/Users/sunmengnan/Desktop/FD/plot/" + filename + "rf_roc_no_smote
   fig, ax = plt.subplots()
   fig.set_size_inches(w = 20, h = 20)
    skplt.metrics.plot_confusion_matrix(test_y, y_pred, normalize = True, ax = ax
                                        text_fontsize=30, title_fontsize=46,
                                        title="Normalised Confusion Matrix")
    fig.savefig("/Users/sunmengnan/Desktop/FD/plot/"+ filename + "rf_confusion_no
   data_list.append(results)
else:
   pass
```

# 1.0.11 From the file and plots that generate, we can view the confusion matrix, taking site 11 for example,



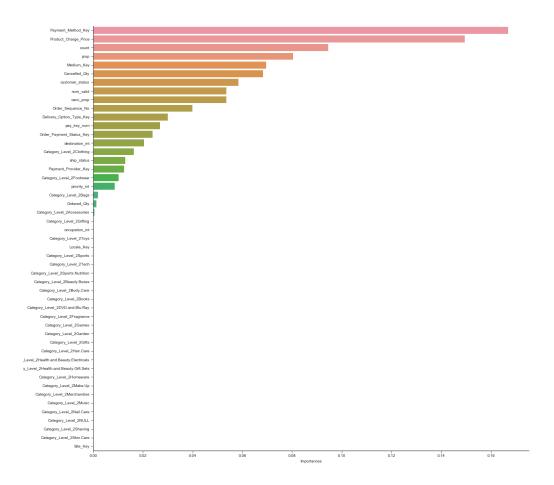
# 1.0.12 Also, we can visualize the random forest ROC curve

In [30]: Image(filename="/Users/sunmengnan/Desktop/FD/plot/\_11.csvrf\_roc\_no\_smote.png",width=6
Out[30]:



# 1.0.13 Finally, we can find the most important features as followed

In [34]: Image(filename="/Users/sunmengnan/Desktop/FD/plot/\_11.csvvariable\_importance\_no\_smote
Out[34]:



## 1.0.14 Output our result as a csv file into the local file