#Importing all the necessary Libraries In [210]:

import numpy as np import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

from matplotlib import gridspec

In [199]: | data=pd.read\_csv('creditcard.csv') #read data

In [211]: data.head(5)# Understanding the Data

## Out[211]:

	Time	V1	V2	V3	V4	V5	V6	V7	V8	
0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.098698	0
1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.085102	-0
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.247676	-1
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.377436	-1
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.270533	0

5 rows × 31 columns

In [213]: | print(data.shape) data.describe() #Describing the Data

(284807, 31)

## Out[213]:

	Time	V1	V2	V3	V4	<b>V</b> 5
count	284807.000000	2.848070e+05	2.848070e+05	2.848070e+05	2.848070e+05	2.848070e+05
mean	94813.859575	3.918649e <b>-</b> 15	5.682686e <b>-</b> 16	-8.761736e-15	2.811118e <b>-</b> 15	-1.552103e-15
std	47488.145955	1.958696e+00	1.651309e+00	1.516255e+00	1.415869e+00	1.380247e+00
min	0.000000	-5.640751e+01	-7.271573e+01	-4.832559e+01	-5.683171e+00	-1.137433e+02
25%	54201.500000	-9.203734e-01	-5.985499e-01	-8.903648e-01	-8.486401e-01	-6.915971e-01
50%	84692.000000	1.810880e <b>-</b> 02	6.548556e <b>-</b> 02	1.798463e-01	-1.984653e-02	-5.433583e-02
75%	139320.500000	1.315642e+00	8.037239e <b>-</b> 01	1.027196e+00	7.433413e-01	6.119264e-01
max	172792.000000	2.454930e+00	2.205773e+01	9.382558e+00	1.687534e+01	3.480167e+01

8 rows × 31 columns

```
In [214]: #Imbalance in the data
#Time to explain the data we are dealing with.
fraud = data[data['Class'] == 1]
  valid = data[data['Class'] == 0]
  outlierFraction = len(fraud)/float(len(valid))
  print(outlierFraction)
  print('Fraud Cases: {}'.format(len(data[data['Class'] == 1])))
  print('Valid Transactions: {}'.format(len(data[data['Class'] == 0])))
```

0.0017304750013189597 Fraud Cases: 492

Valid Transactions: 284315

Only 0.17% fraudulent transaction out all the transactions. The data is highly Unbalanced. Lets first apply our models without balancing it and if we don't get a good accuracy then we can find a way to balance this dataset. But first, let's implement the model without it and will balance the data only if needed.

```
In [217]: # Print the amount details for Fraudulent Transaction
    print("Amount details of the fraudulent transaction")
    fraud.Amount.describe()

Amount details of the fraudulent transaction
Out[217]: count 492.000000
```

```
      Out[217]:
      count
      492.000000

      mean
      122.211321

      std
      256.683288

      min
      0.000000

      25%
      1.000000

      50%
      9.250000

      75%
      105.890000

      max
      2125.870000
```

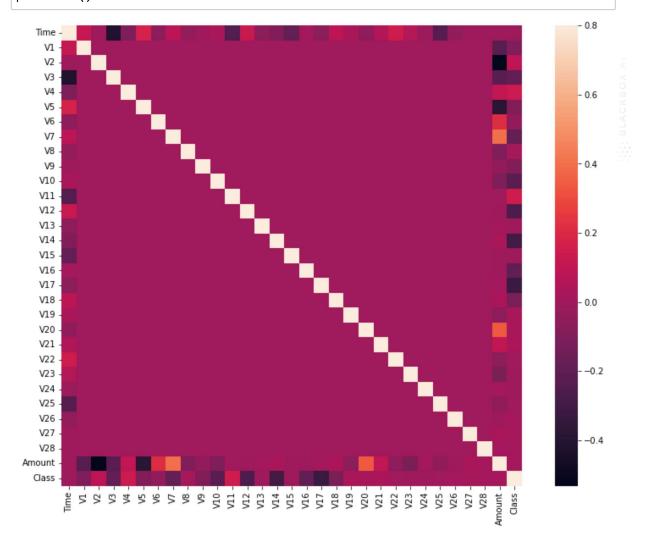
Name: Amount, dtype: float64

## In [218]: #Print the amount details for Normal Transaction print("details of valid transaction") valid.Amount.describe()

details of valid transaction

```
Out[218]: count
                    284315.000000
                        88.291022
          mean
           std
                       250.105092
          min
                         0.000000
           25%
                         5.650000
           50%
                        22.000000
           75%
                        77.050000
                     25691.160000
          max
          Name: Amount, dtype: float64
```

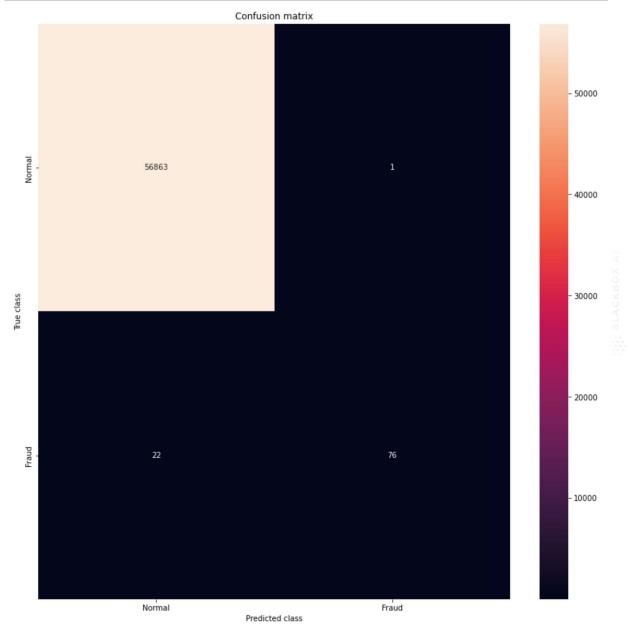
## In [219]: # Correlation matrix #Plotting the Correlation Matrix corrmat = data.corr() fig = plt.figure(figsize = (14, 10)) sns.heatmap(corrmat, vmax = .8, square = True) plt.show()



```
In [221]: | # dividing the X and the Y from the dataset
          #Separating the X and the Y values
          X = data.drop(['Class'], axis = 1)
          Y = data["Class"]
          print(X.shape)
          print(Y.shape)
          # getting just the values for the sake of processing
          # (its a numpy array with no columns)
          xData = X.values
          yData = Y.values
          (284807, 30)
          (284807,)
In [222]: |#Training and Testing Data Bifurcation
          # Using Scikit-learn to split data into training and testing sets
          from sklearn.model selection import train test split
          # Split the data into training and testing sets
          xTrain, xTest, yTrain, yTest = train_test_split(xData, yData, test_size = 0.2,
  In [*]: # Building the Random Forest Classifier (RANDOM FOREST)
          from sklearn.ensemble import RandomForestClassifier
          # random forest model creation
          rfc = RandomForestClassifier()
          rfc.fit(xTrain, yTrain)
          # predictions
          yPred = rfc.predict(xTest)
```

```
In [*]: # Evaluating the classifier
        # printing every score of the classifier
        # scoring in anything
        from sklearn.metrics import classification_report, accuracy_score
        from sklearn.metrics import precision_score, recall_score
        from sklearn.metrics import f1 score, matthews corrcoef
        from sklearn.metrics import confusion_matrix
        n_outliers = len(fraud)
        n_errors = (yPred != yTest).sum()
        print("The model used is Random Forest classifier")
        acc = accuracy_score(yTest, yPred)
        print("The accuracy is {}".format(acc))
        prec = precision_score(yTest, yPred)
        print("The precision is {}".format(prec))
        rec = recall_score(yTest, yPred)
        print("The recall is {}".format(rec))
        f1 = f1_score(yTest, yPred)
        print("The F1-Score is {}".format(f1))
        MCC = matthews corrcoef(yTest, yPred)
        print("The Matthews correlation coefficient is{}".format(MCC))
```

BLACKBOX AI



In [\*]: #link for full details about model
#https://www.geeksforgeeks.org/ml-credit-card-fraud-detection/