

In []:

```
# Credit/Debit Card Fraud Detection
```

In [3]:

```
import pandas as pd
from pandas.plotting import scatter_matrix
import numpy as np
import matplotlib.pyplot as plt
import os
from imblearn.over_sampling import ADASYN
from collections import Counter
import seaborn as sn
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.metrics import roc_curve, roc_auc_score
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn import metrics
%matplotlib inline

flatui = ["#9b59b6", "#3498db", "#95a5a6", "#e74c3c", "#34495e", "#2ecc71"]
sn.set_palette(flatui)
```

Using TensorFlow backend.

In [4]:

```
import warnings
warnings.filterwarnings('ignore')
```

In [5]:

```
def plot_confusion_matrix(cm, classes, title, cmap):
    "plotting confusion matrix"
    plt.clf()
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    classnames = classes
    plt.title(title)
    plt.ylabel('True')
    plt.xlabel('Predicted')
    tick_marks = np.arange(len(classnames))
    plt.xticks(tick_marks, classnames, rotation=0)
    plt.yticks(tick_marks, classnames)
    s = [['TN', 'FP'], ['FN', 'TP']]

    for i in range(2):
        for j in range(2):
            plt.text(j, i, str(s[i][j]) + " = " + str(cm[i][j]))
    plt.show()
```

In [6]:

```
def plot_roc(arg1, arg2, arg3):  
    "a function to plot roc_auc"  
    fig, ax = plt.subplots(figsize=(8, 6))  
    for i, v in arg1:  
        y_score = v.predict_proba(arg2)[:, 1]  
        fpr, tpr, _ = metrics.roc_curve(arg3, y_score)  
        roc_auc = metrics.auc(fpr, tpr)  
        plt.plot(fpr, tpr, lw=2, label= i + ' (area = %0.2f)' % roc_auc)  
    plt.plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--')  
    plt.xlim([0.0, 1.0])  
    plt.ylim([0.0, 1.05])  
    plt.xlabel('False Positive Rate')  
    plt.ylabel('True Positive Rate')  
    plt.title('Receiver Operating Characteristic Curve')  
    plt.legend(loc="lower right")  
    plt.show()
```

In [7]:

```
df = pd.read_csv(r"C:\Users\HP\Desktop\creditcard.csv")  
print('The dataset contains {0} rows and {1} columns.'.format(df.shape[0], df.shape[1]))  
print('Normal transactions count: ', df['Class'].value_counts().values[0])  
print('Fraudulent transactions count: ', df['Class'].value_counts().values[1])
```

The dataset contains 284807 rows and 31 columns.

Normal transactions count: 284315

Fraudulent transactions count: 492

In [8]:

```
X = df.iloc[:, :-1]  
y = df['Class']  
scaler = StandardScaler()  
scaled_X = scaler.fit_transform(X)  
X_train, X_test, y_train, y_test = train_test_split(scaled_X, y, test_size=0.30, random_sta
```

In [9]:

```
X.hist(figsize = (20, 20))
plt.show()
```



In [10]:

```
LGR = LogisticRegression()  
LGR.fit(X_train, y_train);
```

In [11]:

```
RDF = RandomForestClassifier(random_state=0)  
RDF.fit(X_train, y_train);
```

In [12]:

```
modl = [('Logistic Regression', LGR), ('Random Forest Method', RDF)]  
models = [md for md in modl]
```

In [13]:

```

print()

for a,b in models:
    scores = cross_val_score(b, X_train, y_train, cv=10)
    accuracy = metrics.accuracy_score(y_train, b.predict(X_train))
    confusion_matrix = metrics.confusion_matrix(y_train, b.predict(X_train))
    classification_report = metrics.classification_report(y_train, b.predict(X_train))
    print('***** {} *****'.format(a))
    print()
    print("Mean Score: ", '{}%'.format(np.round(scores.mean(), 3) * 100))
    print()
    print("Model Accuracy: ", '{}%'.format(np.round(accuracy, 3) * 100))
    print()
    print("Confusion Matrix:" "\n", confusion_matrix)
    print()
    print("Classification Report:" "\n", classification_report)
    print()

```

***** Logistic Regression *****

Mean Score: 99.9%

Model Accuracy: 99.9%

Confusion Matrix:

```
[[198981    27]
 [   136   220]]
```

Classification Report:

	precision	recall	f1-score	support
'0'	1.00	1.00	1.00	199008
'1'	0.89	0.62	0.73	356
accuracy			1.00	199364
macro avg	0.95	0.81	0.86	199364
weighted avg	1.00	1.00	1.00	199364

***** Random Forest Method *****

Mean Score: 99.9%

Model Accuracy: 100.0%

Confusion Matrix:

```
[[199007     1]
 [    14   342]]
```

Classification Report:

	precision	recall	f1-score	support
'0'	1.00	1.00	1.00	199008
'1'	1.00	0.96	0.98	356
accuracy			1.00	199364
macro avg	1.00	0.98	0.99	199364
weighted avg	1.00	1.00	1.00	199364



In [14]:

```
from sklearn.metrics import roc_curve  
from sklearn.metrics import roc_auc_score
```

In [15]:

```

classnf = {'Normal':0, 'Fraud':1}
print()
print('***** Model Test Results *****' "\n")

for a, b in models:
    accuracy = metrics.accuracy_score(y_test, b.predict(X_test))
    confusion_matrix = metrics.confusion_matrix(y_test, b.predict(X_test))
    classification = metrics.classification_report(y_test, b.predict(X_test))
    print('***** {} *****'.format(a))
    print("Model Accuracy: ", '{}%'.format(np.round(accuracy, 3) * 100))
    print()
    print("Confusion Matrix:" "\n", confusion_matrix)
    print()
    print("Matrix Plot : ")
    plot_confusion_matrix(confusion_matrix, classes = list(classnf.keys()), title='Confusion Matrix')
    print()
    print("Classification Report:" "\n", classification)
    print()

```

***** Model Test Results *****

***** Logistic Regression *****

Model Accuracy: 99.9%

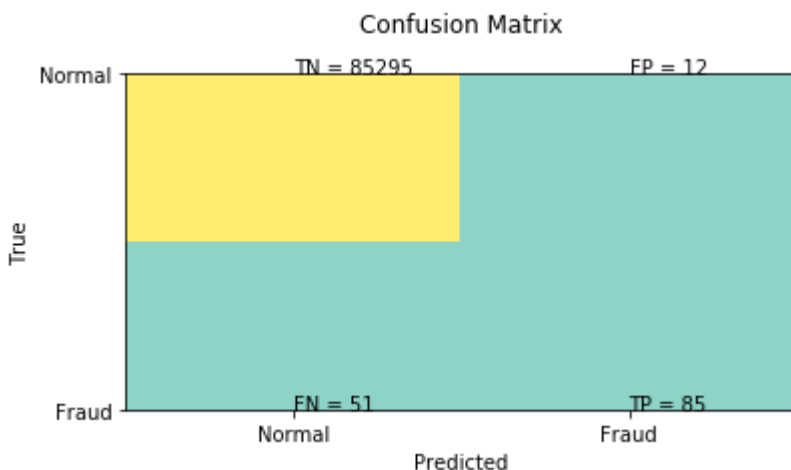
Confusion Matrix:

```

[[85295  12]
 [  51  85]]

```

Matrix Plot :



Classification Report:

	precision	recall	f1-score	support
'0'	1.00	1.00	1.00	85307
'1'	0.88	0.62	0.73	136
accuracy			1.00	85443
macro avg	0.94	0.81	0.86	85443
weighted avg	1.00	1.00	1.00	85443

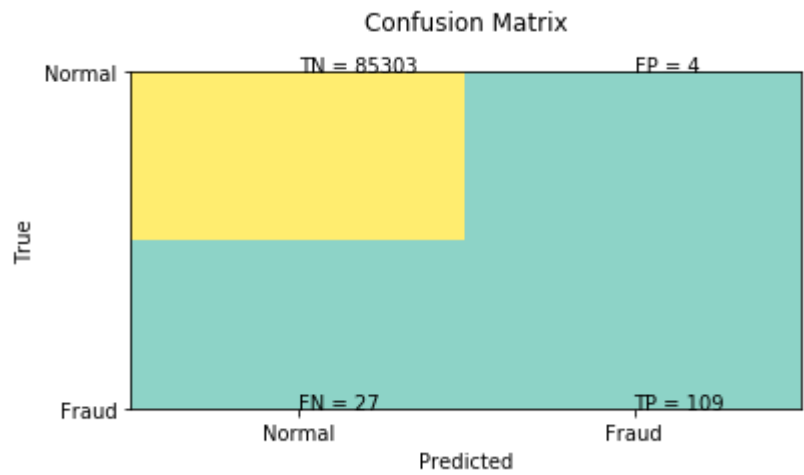
***** Random Forest Method *****

Model Accuracy: 100.0%

Confusion Matrix:

```
[[85303    4]
 [   27  109]]
```

Matrix Plot :



Classification Report:

	precision	recall	f1-score	support
'0'	1.00	1.00	1.00	85307
'1'	0.96	0.80	0.88	136
accuracy			1.00	85443
macro avg	0.98	0.90	0.94	85443
weighted avg	1.00	1.00	1.00	85443

In []: