```
In [1]: import pandas as pd
In [2]: from sklearn.preprocessing import StandardScaler
         from sklearn.model selection import train test split
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.linear model import LogisticRegression
         from sklearn.svm import SVC
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.model selection import cross val score
         from sklearn import metrics
         from sklearn.metrics import accuracy_score
         from sklearn.metrics import f1 score
In [3]: pwd
Out[3]: 'C:\\Users\\USER'
In [4]: cd C:\Users\USER\Downloads
         C:\Users\USER\Downloads
         df=pd.read csv('creditcard.csv')
In [5]:
In [6]: df.head()
Out[6]:
                      V1
                               V2
                                       V3
                                                V4
                                                         V5
                                                                 V6
                                                                          V7
                                                                                   V8
            Time
                                                                              0.098698
              0.0 -1.359807 -0.072781 2.536347
                                           1.378155 -0.338321
                                                            0.462388
                                                                     0.239599
                 1.191857
                          0.266151 0.166480
                                           0.448154
                                                            -0.082361
                                                                     -0.078803
                                                                              0.085102 -0
                                                    0.060018
                 -1.358354
                          -1.340163 1.773209
                                           0.379780 -0.503198
                                                            1.800499
                                                                     0.791461
                                                                              0.247676 -1
              1.0 -0.966272 -0.185226 1.792993 -0.863291 -0.010309
                                                            1.247203
                                                                     0.237609
                                                                              0.377436 -1
```

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V1
                                                                                    V8
             Time
                                V2
                                        V3
                                                 V4
                                                          V5
                                                                  V6
                                                                           V7
              0.095921
                                                                      0.592941 -0.270533
          5 rows × 31 columns
         df.drop('Time',axis=1,inplace=True)
 In [8]: df.head()
 Out[8]:
                  V1
                           V2
                                   V3
                                            V4
                                                    V5
                                                             V6
                                                                      V7
                                                                              V8
                                                                                       ٧٤
           0 -1.359807 -0.072781 2.536347
                                      1.378155 -0.338321
                                                        0.462388
                                                                 0.239599
                                                                          0.098698
                                                                                  0.363787
           1 1.191857 0.266151 0.166480
                                       0.448154
                                               0.060018 -0.082361
                                                                -0.078803
                                                                          0.085102 -0.255425
           2 -1.358354 -1.340163 1.773209
                                       0.379780 -0.503198
                                                        1.800499
                                                                 0.791461
                                                                          0.247676 -1.514654
           3 -0.966272 -0.185226 1.792993
                                      -0.863291 -0.010309
                                                                          0.377436 -1.387024
                                                        1.247203
                                                                 0.237609
           4 -1.158233 0.877737 1.548718 0.403034 -0.407193
                                                       0.095921
                                                                 0.592941 -0.270533 0.817739
          5 rows × 30 columns
 In [9]:
          cases=len(df)
          non fraud count=len(df[df.Class==0])
          fraud count=len(df[df.Class==1])
In [10]: print('Fraud Cases: {}'.format(len(df[df['Class'] == 1])))
          print('Valid Transactions: {}'.format(len(df[df['Class'] == 0])))
          Fraud Cases: 492
          Valid Transactions: 284315
         fraud percentage=round(fraud count/non fraud count*100,3)
In [11]:
          fraud percentage
```

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Out[11]: 0.173
In [12]: sc=StandardScaler()
In [13]: amount=df['Amount'].values
         df['Amount']=sc.fit transform(amount.reshape(-1,1))
In [14]: df['Amount']
Out[14]: 0
                   0.244964
         1
                  -0.342475
         2
                   1.160686
                   0.140534
         4
                  -0.073403
                     . . .
         284802
                  -0.350151
         284803
                  -0.254117
                  -0.081839
         284804
         284805
                  -0.313249
         284806
                   0.514355
         Name: Amount, Length: 284807, dtype: float64
In [15]: x=df.drop('Class',axis=1).values
         y=df['Class'].values
In [16]: from imblearn.over sampling import SMOTE
         smote = SMOTE()
         # fit predictor and target variable
         x smote, y smote = smote.fit resample(x, y)
In [18]: from collections import Counter
         print('Original dataset shape', Counter(y))
         print('Resample dataset shape', Counter(y_smote))
         Original dataset shape Counter({0: 284315, 1: 492})
```

```
Resample dataset shape Counter({0: 284315, 1: 284315})
In [ ]:
In [19]: x train,x test,y train,y test=train test split(x,y,test size=0.3,random
         state=0)
In [20]: tree model=DecisionTreeClassifier(max depth=4, criterion='entropy')
In [21]: tree model.fit(x train,y train)
Out[21]: DecisionTreeClassifier(criterion='entropy', max depth=4)
In [23]: tree_yhat=tree_model.predict(x_test)
         tree yhat
Out[23]: array([0, 0, 0, ..., 0, 0, 0], dtype=int64)
In [33]: knn=KNeighborsClassifier(n neighbors=5)
         knn.fit(x train,y train)
         knn yhat=knn.predict(x test)
In [ ]: svm = SVC(kernel="rbf", random state=0)
         svm.fit(x train, y train)
         svm yhat = svm.predict(x test)
In [37]: lg=LogisticRegression()
         lg.fit(x train,y train)
         lg yhat=lg.predict(x test)
In [25]: from sklearn.metrics import classification report, accuracy score
         from sklearn.metrics import precision score, recall score
         from sklearn.metrics import f1 score
In [39]: accuracy=accuracy score(y test,lg yhat)
```

```
accuracy
Out[39]: 0.999204147794436
In [26]: accuracy=accuracy score(y test, tree yhat)
In [27]: accuracy
Out[27]: 0.9993679997191109
In [30]: Precision=precision score(y test, tree yhat)
In [31]: report=classification report(y test, tree yhat)
In [32]: report
Out[32]: '
                        precision
                                     recall f1-score
                                                        support\n\n
                                                                              0
         1.00
                   1.00
                             1.00
                                      85296\n
                                                                0.85
                                                                          0.76
                                                                  1.00
         0.81
                    147\n\n
                                                                           85443
                               accuracy
         \n macro avg
                              0.93
                                        0.88
                                                  0.90
                                                           85443\nweighted avg
                             1.00
                                      85443\n'
         1.00
                   1.00
In [35]: cMatrix=metrics.confusion matrix(tree yhat,y test)
In [36]: cMatrix
Out[36]: array([[85277,
                           35],
                          112]], dtype=int64)
                [ 19,
In [40]: report=classification report(y test,lg yhat)
In [41]: report
Out[41]: '
                                     recall f1-score
                        precision
                                                        support\n\n
                                                                              0
                   1.00
                             1.00
                                      85296\n
                                                                0.88
                                                                          0.62
         1.00
         0.73
                    147\n\n
                                                                  1.00
                                                                           85443
                               accuracy
```