```
In [1]: import pandas as pd
                  import seaborn as sns
                  import numpy as np
  In [2]: data=pd.read_csv("Documents/MLProject/Project3/creditcard.csv")
                  data.head()
  Out[2]:
                                                                                                                                                                   V9 ...
                        Time
                                          V1
                                                         V2
                                                                        ٧3
                                                                                       V4
                                                                                                      V5
                                                                                                                     V6
                                                                                                                                    V7
                                                                                                                                                   V8
                                                                                                                                                                                      V21
                         0.0 -1.359807 -0.072781 2.536347 1.378155 -0.338321 0.462388
                                                                                                                            0.239599
                                                                                                                                           0.098698
                                                                                                                                                          0.363787 ... -0.018307 0.277
                          0.0 1.191857 0.266151 0.166480
                                                                              0.085102 -0.255425 ... -0.225775 -0.638
                         1.0 -1.358354 -1.340163 1.773209 0.379780 -0.503198 1.800499
                                                                                                                            1.0 -0.966272 -0.185226 1.792993 -0.863291 -0.010309
                                                                                                            1.247203
                                                                                                                            0.237609
                                                                                                                                           0.377436 -1.387024 ... -0.108300 0.005
                          2.0 \quad -1.158233 \quad 0.877737 \quad 1.548718 \quad 0.403034 \quad -0.407193 \quad 0.095921 \quad 0.592941 \quad -0.270533 \quad 0.817739 \quad \dots \quad -0.009431 \quad 0.7981239 \quad 0.009431 \quad 0.798129 \quad 0.009431 \quad 0.798129 \quad 0.999129 \quad 0.999
                  5 rows × 31 columns
  In [3]: fraud=data.loc[data['Class']==1]
                  normal=data.loc[data['Class']==0]
  In [4]: len(fraud)
  Out[4]: 492
  In [5]: len(normal)
  Out[5]: 284315
  In [6]: sns.relplot(x='Amount', y='Class', hue='Class', data=data)
  Out[6]: <seaborn.axisgrid.FacetGrid at 0x1a221bc510>
                       1.0
                       0.8
                       0.6
                    Class
                                                                                                        Class
                                                                                                       0
                       0.4
                                                                                                       1
                       0.2
                       0.0
                                         5000
                                                               15000
                                                   10000
                                                                            20000
                                                                                        25000
                                                          Amount
  In [7]: from sklearn.model_selection import train_test_split
                  from sklearn.linear_model import LogisticRegression
                  from sklearn.svm import SVC
  In [8]: x=data.drop(['Class'], axis='columns')
                  y=data.Class
  In [9]: x_train, x_test, y_train, y_test=train_test_split(x,y,test_size=0.25)
In [10]: model1=LogisticRegression()
                  model1.fit(x_train,y_train)
                  /opt/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/_logistic.py:940: Convergence
                  Warning: lbfgs failed to converge (status=1):
                  STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
                  Increase the number of iterations (max_iter) or scale the data as shown in:
                         https://scikit-learn.org/stable/modules/preprocessing.html
                  Please also refer to the documentation for alternative solver options:
                         https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
                     extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG)
Out[10]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                                                      intercept_scaling=1, l1_ratio=None, max_iter=100,
                                                      multi_class='auto', n_jobs=None, penalty='12',
                                                      random_state=None, solver='lbfgs', tol=0.0001, verbose=0,
                                                      warm_start=False)
In [11]: model2=SVC()
                  model2.fit(x_train,y_train)
Out[11]: SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
                          decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
                          max_iter=-1, probability=False, random_state=None, shrinking=True,
                          tol=0.001, verbose=False)
In [12]: y_predict=model1.predict(x_test)
                  y=np.array(y_test)
In [13]: from sklearn.metrics import confusion_matrix, classification_report, accuracy_score
In [14]: print(confusion_matrix(y_test, y_predict))
                  [[71030
                                       47]
                                       94]]
                           31
In [15]: | print(accuracy_score(y_test, y_predict))
                  0.9989045251537878
In [16]: model1.score(x_test, y_test)
Out[16]: 0.9989045251537878
In [18]: model2.score(x_test, y_test)
Out[18]: 0.9982444313361984
In [17]: print(classification_report(y_test,y_predict))
                                                                                                         support
                                             precision
                                                                     recall f1-score
                                       0
                                                                                                              71077
                                                      1.00
                                                                         1.00
                                                                                            1.00
                                                                                            0.71
                                       1
                                                      0.67
                                                                         0.75
                                                                                                                 125
                                                                                                              71202
                                                                                            1.00
                         accuracy
                                                      0.83
                                                                         0.88
                                                                                            0.85
                       macro avg
                                                                                                              71202
                  weighted avg
                                                                         1.00
                                                                                            1.00
                                                                                                              71202
                                                      1.00
```

In []: