```
K-Nearest Neighbour
           Importing Necessary Packages
 In [94]: import pandas as pd
           import numpy as np
           import seaborn as sns
           from sklearn.model_selection import train_test_split
           from sklearn.preprocessing import StandardScaler
           from sklearn.neighbors import KNeighborsClassifier
           from sklearn.metrics import accuracy score , classification report , confusion matrix
           import matplotlib.pyplot as plt
 In [55]: data = pd.read csv('KNN Project Data', sep = ',')
           Exploratory Data Analysis
 In [57]: data.head()
 Out[57]:
                   XVPM
                              GWYH
                                          TRAT
                                                    TLLZ
                                                                IGGA
                                                                          HYKR
                                                                                      EDFS
                                                                                               GUUB
                                                                                                           MGJM
           0 1636.670614
                          817.988525 2565.995189 358.347163 550.417491 1618.870897 2147.641254 330.727893 1494.878631
                          577.587332 2644.141273 280.428203 1161.873391 2084.107872 853.404981 447.157619 1193.032521
           1 1013.402760
           2 1300.035501
                          820.518697 2025.854469 525.562292
                                                           922.206261 2552.355407
                                                                                 818.676686 845.491492 1968.367513 1
           3 1059.347542 1066.866418 612.000041 480.827789
                                                           419.467495 685.666983
                                                                                 852.867810 341.664784 1154.391368 1
           4 1018.340526 1313.679056 950.622661 724.742174 843.065903 1370.554164
                                                                                 905.469453 658.118202 539.459350 1
 In [58]: data.dtypes
 Out[58]: XVPM
                            float64
                            float64
                            float64
           TRAT
                            float64
           TLLZ
                            float64
           IGGA
           HYKR
                            float64
           EDFS
                            float64
           GUUB
                            float64
           MGJM
                            float64
                            float64
           JHZC
           TARGET CLASS
                             int64
           dtype: object
 In [80]: print(len(data['TARGET CLASS'] == 0))
           print(len(data['TARGET CLASS'] == 1))
           1000
           1000
 In [59]: data.isnull().sum()
 Out[59]: XVPM
                            0
           GWYH
                            0
                            0
           TRAT
           TLLZ
           IGGA
           HYKR
           EDFS
           GUUB
           MGJM
           JHZC
           TARGET CLASS
           dtype: int64
 In [62]: corr = data.corr()
           sns.heatmap (corr , annot = True , square = True , linewidths = 0.5 , cmap = 'Blues_r')
 Out[62]: <matplotlib.axes. subplots.AxesSubplot at 0x1f9a42890f0>
                  XVPM - 1 .00480140530007012902
                                                       - 0.75
                       05$9.120.03 1 .093$1904806$405$409$106$9.1
                                                       - 0.50
                                                       - 0.25
                  EDFS
                  GUUB
                                                       - 0.00
                       02 0.110.10.060.0670064048060.01 1 -0.3
                                                        -0.25
            TARGET CLASS
                              TLLZ
IGGA
HYKR
EDFS
           Feature Scaling
 In [64]: scaler = StandardScaler()
 In [67]: | scaled = scaler.fit_transform(data.drop('TARGET CLASS', axis = 1))
 In [68]: data feat = pd.DataFrame(scaled , columns = data.columns[:-1])
 In [70]: data feat.head()
 Out[70]:
                 XVPM
                          GWYH
                                   TRAT
                                             TLLZ
                                                     IGGA
                                                             HYKR
                                                                       EDFS
                                                                               GUUB
                                                                                        MGJM
                                                                                                  JHZC
           0 1.568522 -0.443435
                                1.619808 -0.958255 -1.128481
                                                           0.138336
                                                                    0.980493
           1 -0.112376 -1.056574
                                1.741918 -1.504220
                                                  0.640009
                                                           1.081552 -1.182663
                                                                            -0.461864
           2 0.660647 -0.436981 0.775793
                                        0.213394 -0.053171
                                                           2.030872 -1.240707
                                                                            1.149298 2.184784
                                                                                               0.342811
           3 0.011533
                      0.191324 -1.433473 -0.100053 -1.507223 -1.753632 -1.183561 -0.888557
                                                                                      0.162310 -0.002793
           4 -0.099059 0.820815 -0.904346 1.609015 -0.282065 -0.365099 -1.095644 0.391419 -1.365603
           Train Test Split
 In [72]: x_train, x_test , y_train , y_test = train_test_split (data_feat , data['TARGET CLASS'] , test_size
           = 0.3)
           Implementing KNN
 In [90]: knn = KNeighborsClassifier (n_neighbors = 1)
 In [91]: knn.fit(x_train , y_train)
 Out[91]: KNeighborsClassifier(algorithm='auto', leaf size=30, metric='minkowski',
                                 metric_params=None, n_jobs=None, n_neighbors=1, p=2,
 In [81]: pred = knn.predict(x_test)
           Classification Report
 In [85]: confusion_matrix(pred, y_test)
 Out[85]: array([[110, 47],
                  [ 40, 103]], dtype=int64)
 In [86]: classification report(pred, y test)
                                                                                    0
 Out[86]: '
                                       recall f1-score
                                                            support\n\n
                                                                                            0.73
                                                                                                       0.70
                          precision
           0.72
                      157\n
                                       1
                                               0.69
                                                          0.72
                                                                     0.70
                                                                                143\n\n
                                                                                            accuracy
                      300\n macro avg
           0.71
                                               0.71
                                                          0.71
                                                                     0.71
                                                                                300\nweighted avg
                                                                                                          0.71
           0.71
                                 300\n'
                     0.71
 In [87]: accuracy score (pred, y test)
 Out[87]: 0.71
           Choosing a K Value
In [107]: acrate = []
           for i in range (1,100):
               knn = KNeighborsClassifier(n_neighbors = i)
               knn.fit(x_train , y_train)
               pred = knn.predict(x test)
               acc = accuracy score(pred , y test)
               acrate.append(acc.mean())
In [108]: plt.figure(figsize= (10,6))
           plt.plot (range(1,100) , acrate , marker = '*' , markerfacecolor = 'red')
           plt.title('Accuracy Vs. K-Value')
           plt.xlabel('K-Value')
           plt.ylabel('Accuracy Score')
Out[108]: Text(0, 0.5, 'Accuracy Score')
                                              Accuracy Vs. K-Value
              0.84
             0.82
              0.80
           Accuracy Score
              0.74
              0.72
                                 20
                                                            60
                                                                          80
                                                                                       100
                                                    K-Value
           From the above plot we can see that the value of k was stable somewhere in the range of (60,85) Hence choosing k as 70
```

In [110]: knn = KNeighborsClassifier (n neighbors = 70)

[32, 128]], dtype=int64)

knn.fit(x_train,y_train)
prd = knn.predict(x_test)
confusion_matrix (prd, y_test)

Out[110]: array([[118, 22],

In [111]: accuracy_score (prd, y_test)