Implement K-Nearest Neighbors algorithm on diabetes.csv dataset. Compute confusion matrix, accuracy, error rate, precision and recall on the given dataset

```
In [1]:
              import numpy as np
              import pandas as pd
In [2]: data = pd.read_csv('./diabetes.csv')
         data.head()
Out[2]:
             Pregnancies
                         Glucose BloodPressure
                                                 SkinThickness Insulin BMI Pedigree Age
          0
                      6
                                                                      33.6
                                                                                                 1
                              148
                                             72
                                                           35
                                                                               0.627
                                                                                       50
          1
                      1
                              85
                                             66
                                                           29
                                                                    0 26.6
                                                                               0.351
                                                                                       31
                                                                                                 0
          2
                      8
                              183
                                             64
                                                            0
                                                                    0
                                                                      23.3
                                                                               0.672
                                                                                       32
                                                                                                 0
                              89
                                                           23
                                                                   94
                                                                      28.1
                                                                               0.167
                                                                                       21
                      0
                                             40
                                                                               2.288
                             137
                                                           35
                                                                  168 43.1
                                                                                       33
                                                                                                 1
In [3]: |#Check for null or missing values
         data.isnull().sum()
Out[3]:
         Pregnancies
                             0
         Glucose
         BloodPressure
                             0
         SkinThickness
                             0
         Insulin
                             0
         BMI
                             0
         Pedigree
                             0
                             0
         Age
         Outcome
                             0
         dtype: int64
In [4]:
         #Replace zero values with mean values
         for column in data.columns[1:-3]:
              data[column].replace(0, np.NaN, inplace = True)
              data[column].fillna(round(data[column].mean(skipna=True)), inplace = True)
         data.head(10)
Out[4]:
             Pregnancies
                         Glucose BloodPressure SkinThickness Insulin
                                                                      BMI Pedigree
                                                                                    Age
                                                                                          Outcome
          0
                            148 0
                                           72.0
                                                                               0.627
                                                                                                 1
                      6
                                                          35.0
                                                                156.0
                                                                      33.6
                                                                                       50
          1
                       1
                             85.0
                                           66.0
                                                          29.0
                                                                156.0 26.6
                                                                               0.351
                                                                                       31
                                                                                                 0
          2
                      8
                            183.0
                                           64.0
                                                          29.0
                                                                156.0 23.3
                                                                               0.672
                                                                                       32
                                                                                                 1
          3
                             89.0
                                           66.0
                                                          23.0
                                                                 94.0 28.1
                                                                               0.167
                                                                                                 0
                      1
                                                                                       21
          4
                      0
                            137.0
                                           40.0
                                                          35.0
                                                                168.0 43.1
                                                                               2.288
                                                                                       33
                                                                                                 1
                                                          29.0
          5
                      5
                            116.0
                                           74.0
                                                                156.0 25.6
                                                                               0.201
                                                                                       30
                                                                                                 0
          6
                      3
                             78.0
                                           50.0
                                                          32.0
                                                                 88.0 31.0
                                                                               0.248
                                                                                       26
                      10
                            115.0
                                           72.0
                                                          29.0
                                                                156.0 35.3
                                                                               0.134
                                                                                       29
                                                                                                 0
                            197.0
          8
                      2
                                           70.0
                                                          45.0
                                                                543.0 30.5
                                                                               0.158
                                                                                       53
                                                                                                 1
                      8
                            125.0
                                                                                                 1
                                           96.0
                                                          29.0
                                                                156.0 32.0
                                                                               0.232
                                                                                       54
In [5]: |X = data.iloc[:, :8] #Features
         Y = data.iloc[:, 8:] #Predictor
         #Perform Spliting
In [6]:
         from sklearn.model_selection import train_test_split
```

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state=0)

```
In [7]: #KNN
        from sklearn.neighbors import KNeighborsClassifier
        knn = KNeighborsClassifier()
        knn_fit = knn.fit(X_train, Y_train.values.ravel())
        knn_pred = knn_fit.predict(X_test)
```

In [8]: from sklearn.metrics import confusion_matrix, precision_score, recall_score, f1_score, ac print("Confusion Matrix")
print(confusion_matrix(Y_test, knn_pred)) print("Accuracy Score:", accuracy_score(Y_test, knn_pred))
print("Reacal Score:", recall_score(Y_test, knn_pred)) print("F1 Score:", f1_score(Y_test, knn_pred))
print("Precision Score:",precision_score(Y_test, knn_pred))

Confusion Matrix [[88 19] [19 28]]

Accuracy Score: 0.7532467532467533 Reacal Score: 0.5957446808510638 F1 Score: 0.5957446808510638

Precision Score: 0.5957446808510638