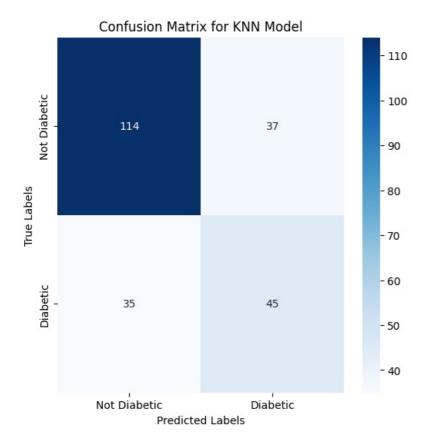
Implement K-Nearest Neighbors algorithm on diabetes.csv dataset. Compute confusion matrix, accuracy, error rate, precision and recall on the given dataset. Dataset link: https://www.kaggle.com/datasets/abdallamahgoub/diabetes

```
In [17]: import pandas as pd
         from sklearn.model_selection import train_test_split
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score
         import seaborn as sns
         import matplotlib.pyplot as plt
In [18]: # Load the dataset
         df = pd.read_csv("C:/Users/samik/Downloads/diabetes (1).csv")
         print(df.head(10)) # Inspect the first few rows
           Pregnancies Glucose BloodPressure SkinThickness Insulin
                                                                        BMI \
        0
                           148
                                            72
                                                                    0
                                                                       33.6
                                                                    0 26.6
                                            66
        1
                    1
                            85
                                                           29
        2
                    8
                           183
                                            64
                                                           0
                                                                   0 23.3
        3
                            89
                                            66
                                                                   94 28.1
                    1
                                                           23
        4
                     0
                           137
                                            40
                                                           35
                                                                  168 43.1
        5
                    5
                                            74
                                                           0
                                                                   0 25.6
                           116
        6
                    3
                            78
                                            50
                                                          32
                                                                  88 31.0
        7
                   10
                           115
                                            0
                                                           0
                                                                    0
                                                                       35.3
        8
                    2
                           197
                                           70
                                                          45
                                                                   543 30.5
        9
                    8
                                           96
                                                          0
                                                                    0
                           125
                                                                        0.0
           Pedigree Age Outcome
        0
             0.627
                     50
                               1
        1
             0.351
                     31
                                0
        2
             0.672
                     32
                                1
        3
             0.167
                     21
                               0
             2.288
        4
                     33
                               1
        5
             0.201
                     30
                               0
        6
             0.248
                     26
                               1
        7
             0.134
                     29
                               0
        8
             0.158
                      53
                               1
             0.232
                    54
                               1
In [19]: # Separate features and target variable
         X = df.drop(columns=['Outcome'])
         y = df['Outcome']
         # Split the dataset into training and testing sets
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
In [20]: # Initialize and train the KNN model
         knn = KNeighborsClassifier(n_neighbors=5) # Default n_neighbors is 5; can be tuned
         knn.fit(X_train, y_train)
         # Make predictions on the test set
         y_pred = knn.predict(X_test)
In [21]: # Calculate performance metrics
         conf matrix = confusion matrix(y test, y pred)
         accuracy = accuracy_score(y_test, y_pred)
         error_rate = 1 - accuracy
         precision = precision_score(y_test, y_pred)
         recall = recall score(y test, y pred)
In [22]: # Display results
         print("Confusion Matrix:\n", conf_matrix)
         print(f"Accuracy: {accuracy:.2f}")
         print(f"Error Rate: {error_rate:.2f}")
         print(f"Precision: {precision:.2f}")
         print(f"Recall: {recall:.2f}")
        Confusion Matrix:
         [[114 37]
         [ 35 45]]
       Accuracy: 0.69
        Error Rate: 0.31
       Precision: 0.55
        Recall: 0.56
In [23]: # Plot confusion matrix as heatmap
         plt.figure(figsize=(6, 6))
         sns.heatmap(conf_matrix, annot=True, fmt="d", cmap="Blues", xticklabels=['Not Diabetic', 'Diabetic'], yticklabe
         plt.xlabel("Predicted Labels")
         plt.ylabel("True Labels")
         plt.title("Confusion Matrix for KNN Model")
         plt.show()
```



In []:

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