```
Practical No:05
In [1]: import pandas as pd
         import seaborn as sns
In [2]: df = pd.read csv("diabetes.csv")
         df
                          Glucose BloodPressure SkinThickness Insulin BMI
                                                                               DiabetesPedigreeFunction Age Outcome
Out[2]:
              Pregnancies
           0
                        6
                               148
                                               72
                                                              35
                                                                      0 33.6
                                                                                                         50
                                                                                                 0.627
                                                                                                                    1
                                                              29
                                                                                                                    0
           1
                                85
                                               66
                                                                      0 26.6
                                                                                                 0.351
                                                                                                         31
           2
                        8
                               183
                                               64
                                                              0
                                                                      0 23.3
                                                                                                 0.672
                                                                                                         32
                                                                                                                    1
           3
                                89
                                               66
                                                              23
                                                                     94
                                                                         28.1
                                                                                                 0.167
                                                                                                         21
                                                                                                                    0
                        0
           4
                               137
                                               40
                                                              35
                                                                                                 2.288
                                                                    168 43.1
                                                                                                         33
                                                                                                                    1
         763
                       10
                               101
                                               76
                                                              48
                                                                    180 32.9
                                                                                                 0.171
                                                                                                         63
                                                                                                                    0
         764
                        2
                               122
                                               70
                                                              27
                                                                      0 36.8
                                                                                                 0.340
                                                                                                         27
                                                                                                                    0
         765
                        5
                               121
                                               72
                                                              23
                                                                    112 26.2
                                                                                                         30
                                                                                                                    0
                                                                                                 0.245
                                                              0
                                                                      0 30.1
                                                                                                         47
         766
                               126
                                               60
                                                                                                 0.349
                                                                                                                    1
         767
                                93
                                               70
                                                              31
                                                                      0 30.4
                                                                                                 0.315
                                                                                                         23
                                                                                                                    0
        768 rows × 9 columns
In [3]: x = df.drop('Outcome', axis = 1)
         y = df['Outcome']
In [5]: sns.countplot(x=y)
Out[5]: <Axes: xlabel='Outcome', ylabel='count'>
           500
           400
           300
           200
           100
             0
```

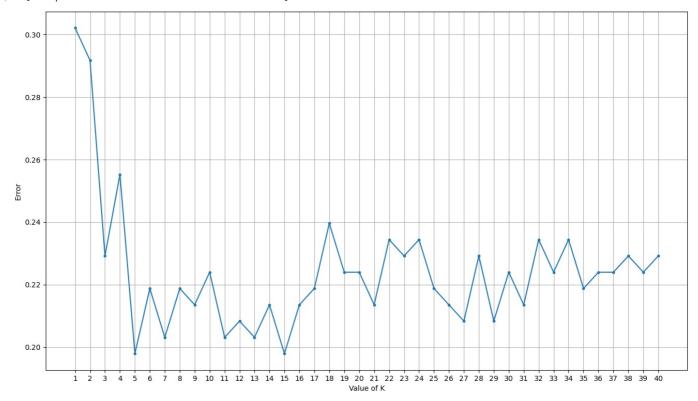
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In [9]: x.shape

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Out[9]: (768, 8)
In [10]: x_train.shape
Out[10]: (576, 8)
In [11]: x_test.shape
Out[11]: (192, 8)
In [12]: from sklearn.neighbors import KNeighborsClassifier
In [13]: knn = KNeighborsClassifier(n_neighbors = 5)
In [14]: knn.fit(x_train, y_train)
Out[14]: VKNeighborsClassifier
         KNeighborsClassifier()
In [15]: from sklearn.metrics import accuracy_score , ConfusionMatrixDisplay
         from sklearn.metrics import classification_report
In [17]: y_pred = knn.predict(x_test)
In [19]: ConfusionMatrixDisplay.from_predictions(y_test,y_pred)
Out[19]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x21f23f3b4d0>
                                                                    100
           0 -
                        110
                                                13
                                                                    80
        True label
                                                                    60
                                                                    40
           1 -
                         0
                               Predicted label
In [18]: print(classification_report(y_test,y_pred))
                                   recall f1-score
                      precision
                                                       support
                   0
                                      0.89
                                                0.85
                           0.81
                                                           123
                   1
                           0.77
                                      0.64
                                                0.70
                                                            69
                                                0.80
                                                           192
            accuracy
                           0.79
                                      0.77
                                                0.78
                                                           192
           macro avg
        weighted avg
                           0.80
                                      0.80
                                                0.80
                                                           192
In [20]: import matplotlib.pyplot as plt
         import numpy as np
In [21]: error = []
         for k in range (1,41):
             knn = KNeighborsClassifier(n_neighbors = k)
             knn.fit(x_train, y_train)
             pred=knn.predict(x_test)
             error.append(np.mean(pred!=y_test))
In [22]: plt.figure(figsize=(16,9))
         plt.xlabel('Value of K')
         plt.ylabel('Error')
```

```
plt.grid()
plt.xticks(range(1,41))
plt.plot(range(1,41),error,marker='.')
```

Out[22]: [<matplotlib.lines.Line2D at 0x21f2518fd90>]



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In [23]: knn = KNeighborsClassifier(n_neighbors = 33)
    knn.fit(x_train, y_train)
```

Out[23]: 🔻

KNeighborsClassifier 000

KNeighborsClassifier(n_neighbors=33)

In [24]: y_pred=knn.predict(x_test)

In [25]: print(classification_report(y_test,y_pred))

	precision	recall	f1-score	support
0 1	0.77 0.80	0.93 0.51	0.84 0.62	123 69
accuracy macro avg weighted avg	0.78 0.78	0.72 0.78	0.78 0.73 0.76	192 192 192

In []: