11/5/23, 11:42 PM ml practical 5

In [3]: import pandas as pd

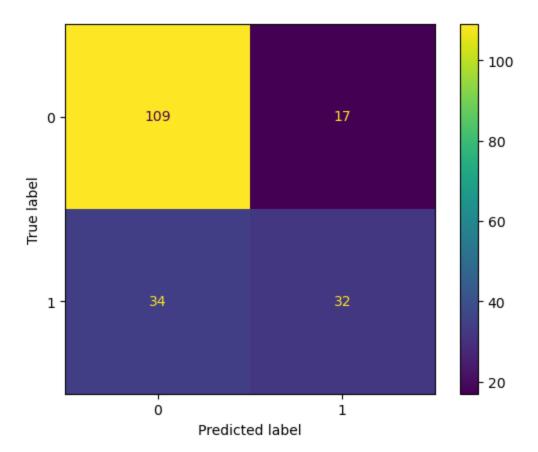
import seaborn as sns

```
In [4]: df = pd.read_csv('diabetes.csv')
         df
 In [5]:
 Out[5]:
               Pregnancies Glucose
                                     BloodPressure SkinThickness Insulin BMI Pedigree Age
                                                                                                Οι
            0
                         6
                                148
                                                                           33.6
                                                                                    0.627
                                                72
                                                               35
                                                                        0
                                                                                             50
            1
                                 85
                                                66
                                                               29
                                                                           26.6
                                                                                    0.351
                                                                                            31
            2
                         8
                                183
                                                64
                                                                0
                                                                           23.3
                                                                                    0.672
                                                                                            32
                                                                        0
                                 89
                                                               23
                                                                       94
                                                                           28.1
                                                                                    0.167
                                                                                            21
                                                66
            4
                         0
                                                40
                                                                      168
                                                                                             33
                                137
                                                               35
                                                                           43.1
                                                                                    2.288
          763
                        10
                                101
                                                                           32.9
                                                76
                                                               48
                                                                      180
                                                                                    0.171
                                                                                            63
          764
                                122
                                                 70
                                                                          36.8
                                                                                    0.340
                                                               27
                         5
          765
                                121
                                                72
                                                               23
                                                                      112 26.2
                                                                                    0.245
                                                                                             30
                                                                                    0.349
          766
                                126
                                                60
                                                                           30.1
          767
                         1
                                 93
                                                70
                                                               31
                                                                        0
                                                                           30.4
                                                                                    0.315
                                                                                            23
         768 rows × 9 columns
 In [6]: # input data
          x = df.drop('Outcome',axis = 1)
          # output dta
          y = df['Outcome']
 In [7]: y.value_counts()
Out[7]: 0
               500
               268
          Name: Outcome, dtype: int64
 In [8]: # Feature scaling
          from sklearn.preprocessing import MinMaxScaler as mms
          scaler = mms()
 In [9]: x_scaled = scaler.fit_transform(x)
In [19]: # cross-validation
          from sklearn.model_selection import train_test_split as tts
```

11/5/23, 11:42 PM ml practical 5

```
xtrain, xtest, ytrain, ytest = tts(x_scaled, y, test_size = 0.25)
In [20]: x.shape
Out[20]: (768, 8)
In [21]: xtrain.shape
Out[21]: (576, 8)
In [22]: xtest.shape
Out[22]: (192, 8)
In [23]: from sklearn.neighbors import KNeighborsClassifier as knc
         knn = knc(n_neighbors = 5)
In [24]: knn.fit(xtrain, ytrain)
Out[24]: ▼ KNeighborsClassifier
         KNeighborsClassifier()
In [25]: from sklearn.metrics import accuracy_score as score
         from sklearn.metrics import ConfusionMatrixDisplay as matrix
         from sklearn.metrics import classification_report as report
In [26]: y_pred = knn.predict(xtest)
In [27]: matrix.from_predictions(ytest, y_pred)
Out[27]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x2a9a1da5330>
```

11/5/23, 11:42 PM ml_practical_5



```
In [29]: print(report(ytest, y_pred))
                      precision
                                    recall f1-score
                                                        support
                   0
                           0.76
                                      0.87
                                                0.81
                                                            126
                   1
                           0.65
                                      0.48
                                                0.56
                                                             66
                                                0.73
                                                            192
            accuracy
                           0.71
                                      0.67
                                                0.68
                                                            192
           macro avg
        weighted avg
                           0.72
                                      0.73
                                                0.72
                                                            192
In [30]:
         score(ytest, y_pred)
Out[30]: 0.734375
In [31]: import matplotlib.pyplot as plt
          import numpy as np
In [61]: error = []
          err = 1
          i = 1
          for k in range(1,500):
              knn = knc(n\_neighbors = k)
              knn.fit(xtrain, ytrain)
              pred = knn.predict(xtest)
              error.append(np.mean(pred != ytest))
              if(np.mean(pred != ytest) < err):</pre>
                  err = np.mean(pred != ytest)
```

11/5/23, 11:42 PM

```
ml_practical_5
                   i = k
In [62]: plt.figure(figsize=(16,9))
          plt.xlabel('Value of k')
          plt.ylabel('Error')
          plt.grid()
          plt.xticks(range(1,500))
          plt.plot(range(1,500), error, marker='.')
Out[62]: [<matplotlib.lines.Line2D at 0x2a9a82a4ca0>]
         0.34
         0.32
         0.30
         0.26
         0.24
         0.22
                                                     Value of k
In [46]: print(err)
        0.2135416666666666
In [47]: i
Out[47]: 12
In [48]:
          knn = knc(n_neighbors = i)
          knn.fit(xtrain, ytrain)
          y_pred = knn.predict(xtest)
In [49]: score(ytest, y_pred)
Out[49]: 0.78645833333333334
```

In [63]: print(report(ytest, y_pred))

11/5/23, 11:42 PM ml_practical_5

	precision	recall	f1-score	support
0	0.78	0.94	0.85	126
	0.82	0.48	0.61	66
accuracy			0.79	192
macro avg	0.80	0.71	0.73	192
weighted avg	0.79	0.79	0.77	192

In []: