Machine Learning Laboratory - Assignment 5

• NAME: - ANURAG AVINASH SHEVALE

• CLASS :- BE COMP I

• ROLL NO :- 20

```
In [ ]:
#Name - Anurag Avinash Shevale
#Class - BE Comp I
#Roll No - 20
In [1]:
import numpy as np
import pandas as pd
In [2]:
data = pd.read csv('/home/admin1/Anurag/diabetes.csv')
In [3]:
data.head(5)
Out[3]:
   Pregnancies Glucose BloodPressure SkinThickness Insulin BMI Pedigree Age Outco
            6
                  148
                                                     0 33.6
                                                                       50
                                                                0.627
1
            1
                   85
                                66
                                              29
                                                     0 26.6
                                                                0.351
                                                                       31
            8
                  183
                                64
                                                     0 23.3
                                                                0.672
                                              23
                                                    94 28.1
            1
                   89
                                66
                                                                0.167
                                                                       21
 4
            0
                  137
                                 40
                                              35
                                                    168 43.1
                                                                2.288
                                                                       33
In [4]:
data.tail(5)
Out[4]:
     Pregnancies Glucose BloodPressure SkinThickness Insulin BMI Pedigree Age Outo
763
             10
                    101
                                  76
                                                     180 32.9
                                                                 0.171
                                                                        63
             2
                    122
                                  70
                                               27
                                                       0 36.8
764
                                                                 0.340
                                                                        27
                    121
                                  72
                                               23
                                                     112 26.2
765
                                                                 0.245
                                                                         30
             1
                                                0
766
                    126
                                  60
                                                       0 30.1
                                                                 0.349
                                                                        47
              1
                     93
                                  70
                                               31
                                                       0 30.4
                                                                 0.315 23
```

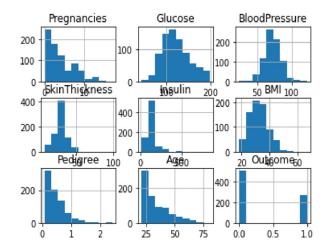
```
In [5]:
data.describe()
Out[5]:
        Pregnancies
                       Glucose BloodPressure SkinThickness
                                                                   Insulin
                                                                                 вмі
                                                                                         Ре
                                                               768.000000
                                                                           768.000000
                                                                                       768.0
 count
         768.000000
                     768.000000
                                    768.000000
                                                   768.000000
 mean
           3.845052 120.894531
                                     69.105469
                                                    20.536458
                                                                79.799479
                                                                            31.992578
                                                                                         0.4
           3.369578
                      31.972618
                                     19.355807
                                                    15.952218 115.244002
                                                                                         0.3
   std
                                                                             7.884160
   min
           0.000000
                       0.000000
                                      0.000000
                                                     0.000000
                                                                 0.000000
                                                                             0.000000
                                                                                         0.0
  25%
           1.000000
                      99.000000
                                     62.000000
                                                     0.000000
                                                                 0.000000
                                                                            27.300000
                                                                                         0.2
                                     72.000000
                                                    23.000000
  50%
           3.000000 117.000000
                                                                30.500000
                                                                            32.000000
                                                                                         0.3
  75%
                                     80.000000
           6.000000 140.250000
                                                    32.000000 127.250000
                                                                            36.600000
                                                                                         0.6
          17.000000 199.000000
                                    122.000000
                                                    99.000000 846.000000
                                                                            67.100000
                                                                                         2.4
4
In [6]:
data.shape
Out[6]:
(768, 9)
In [8]:
data.boxplot()
Out[8]:
<AxesSubplot:>
 800
                               φ
                               0
 600
 400
 200
                   8
             0
   0 -
   Pregnanci@sucBszodPr@sixuThicknessulin
                                    BMI Pedigree Age Outcome
```

```
In [12]:
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[12]:
               Glucose BloodPressure SkinThickness Insulin BMI Pedigree Age
   Pregnancies
                                                                                Outco
0
             6
                                                      156.0
                                                            33.6
                                                                     0.627
1
             1
                   85.0
                                 66.0
                                                29.0
                                                      156.0 26.6
                                                                     0.351
                                                                             31
2
             8
                  183.0
                                 64.0
                                                29.0
                                                      156.0 23.3
                                                                     0.672
                                                                             32
3
             1
                   89.0
                                 66.0
                                                23.0
                                                       94.0 28.1
                                                                     0.167
                                                                             21
4
             o
                  137.0
                                 40.0
                                                35.0
                                                      168.0 43.1
                                                                     2.288
                                                                             33
             5
                  116.0
                                  74.0
                                                29.0
                                                       156.0 25.6
                                                                     0.201
                                                                             30
             3
6
                   78.0
                                 50.0
                                                32.0
                                                       88.0 31.0
                                                                     0.248
                                                                             26
            10
                                                      156.0 35.3
                                                                     0.134
                                                                             29
                  115.0
                                  72.0
                                                29.0
8
             2
                  197.0
                                  70.0
                                                45.0
                                                       543.0 30.5
                                                                     0.158
                                                                             53
9
             8
                  125.0
                                 96.0
                                                29.0
                                                      156.0 32.0
                                                                     0.232
                                                                             54
```

In [16]:

```
data.hist()
```

Out[16]:



In [17]:

```
X = data.iloc[:, :8] #Features
Y = data.iloc[:, 8:] #Predictor
```

In [18]:

```
#This step performs splitting of data for training and testing
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_s
```

In [19]:

```
#Execution of K-Nearest Neighbor
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 [21]:

```
#We import all metrics that are needed for calculation from sklearn.metrics import confusion_matrix, precision_score, recall_score, f1_s
```

```
In [22]:
#After executing this we will get Confusion matrix
print("Confusion Matrix")
print(confusion_matrix(Y_test, knn_pred))
Confusion Matrix
[[88 19]
[19 28]]
In [23]:
#After executing this we will get the Accuracy Score
print("Accuracy Score:", accuracy_score(Y_test, knn_pred))
Accuracy Score: 0.7532467532467533
In [25]:
#After executing this we will get the Recall Score
print("Recall Score:", recall_score(Y_test, knn_pred))
Recall Score: 0.5957446808510638
In [26]:
#After executing this we will get the F1 Score
print("F1 Score:", f1 score(Y test, knn pred))
F1 Score: 0.5957446808510638
In [27]:
#After executing this we will get the Precision Score
print("Precision Score:",precision score(Y test, knn pred))
Precision Score: 0.5957446808510638
```