import numpy as np
import pandas as pd

from sklearn.preprocessing import StandardScaler

 $from \ sklearn.model_selection \ import \ train_test_split$

from sklearn import svm

 $from \ sklearn.metrics \ import \ accuracy_score$

Data Collection and Analysis

loading the diabetes dataset to a pandas DataFrame
diabetes_dataset = pd.read_csv('/content/diabetes.csv')

pd.read_csv?

printing the first 5 rows of the dataset
diabetes_dataset.head()

₽		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction
	0	6	148	72	35	0	33.6	0.627
	1	1	85	66	29	0	26.6	0.351
	2	8	183	64	0	0	23.3	0.672
	3	1	89	66	23	94	28.1	0.167
	4	0	137	40	35	168	43.1	2.288
	◀ 🎚							•

number of rows and Columns in this dataset
diabetes_dataset.shape

(768, 9)

getting the statistical measures of the data
diabetes_dataset.describe()

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Diabetes
count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	
mean	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578	
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	

diabetes_dataset['Outcome'].value_counts()

0 500

1 268

Name: Outcome, dtype: int64

0 --> Non-Diabetic

1 --> Diabetic

diabetes_dataset.groupby('Outcome').mean()

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Diabete
Outcome							
0	3.298000	109.980000	68.184000	19.664000	68.792000	30.304200	
1	4.865672	141.257463	70.824627	22.164179	100.335821	35.142537	

```
# separating the data and labels
X = diabetes_dataset.drop(columns = 'Outcome', axis=1)
Y = diabetes_dataset['Outcome']
print(X)
         Pregnancies Glucose BloodPressure SkinThickness Insulin BMI \
    0
                        148
                                                              0 33.6
                 6
                                       72
                                                     35
                                                              0 26.6
    1
                         85
                                                     29
                  1
                                       66
    2
                  8
                        183
                                       64
                                                      0
                                                              0 23.3
    3
                        89
                                      66
                                                    23
                                                             94 28.1
                 1
    4
                  0
                       137
                                       40
                                                    35
                                                            168 43.1
                                                    48
     763
                 10
                        101
                                       76
                                                            180 32.9
    764
                  2
                        122
                                       70
                                                     27
                                                             0 36.8
                                                    23
    765
                  5
                        121
                                       72
                                                            112 26.2
     766
                  1
                         126
                                       60
                                                     0
                                                             0 30.1
                                                     31
                                                             0 30.4
    767
         DiabetesPedigreeFunction Age 0.627 50
    0
    1
                          0.351
                                  31
    2
                          0.672
                                  32
    3
                          0.167
                                  21
    4
                          2.288
                                  33
    763
                          0.171 63
     764
                          0.340
                                  27
                          0.245
     765
                                  30
     766
                          0.349
                                  47
                          0.315
    767
                                  23
    [768 rows x 8 columns]
print(Y)
    1
           0
    2
          1
    3
           0
    4
          1
     763
          a
     764
           0
     765
           0
     766
           1
     767
           0
    Name: Outcome, Length: 768, dtype: int64
Data Standardization
scaler = StandardScaler()
scaler.fit(X)
     ▼ StandardScaler
     StandardScaler()
standardized_data = scaler.transform(X)
print(standardized_data)
     [[ 0.63994726  0.84832379  0.14964075 ...  0.20401277  0.46849198
       1.4259954 ]
     [-0.84488505 -1.12339636 -0.16054575 ... -0.68442195 -0.36506078
      -0.19067191]
     -0.10558415]
     [ \ 0.3429808 \quad 0.00330087 \quad 0.14964075 \ \dots \ -0.73518964 \ -0.68519336
       -0.27575966]
     [-0.84488505 \quad 0.1597866 \quad -0.47073225 \ \dots \ -0.24020459 \ -0.37110101
       1.17073215]
     -0.87137393]]
X = standardized_data
Y = diabetes_dataset['Outcome']
print(X)
print(Y)
     [[ 0.63994726  0.84832379  0.14964075 ...  0.20401277  0.46849198
```

1.4259954]

```
-0.19067191]
     -0.10558415]
     -0.27575966]
     1.170732151
     -0.87137393]]
    a
    1
          0
    2
          1
    3
          0
    4
          1
    763
    764
          0
    765
          0
    766
          1
    767
          0
    Name: Outcome, Length: 768, dtype: int64
Train Test Split
X_train, X_test, Y_train, Y_test = train_test_split(X,Y, test_size = 0.2, stratify=Y, random_state=2)
print(X.shape, X_train.shape, X_test.shape)
    (768, 8) (614, 8) (154, 8)
Training the Model
classifier = svm.SVC(kernel='linear')
#training the support vector Machine Classifier
classifier.fit(X_train, Y_train)
            SVC
    SVC(kernel='linear')
Model Evaluation
Accuracy Score
# accuracy score on the training data
X train prediction = classifier.predict(X train)
training_data_accuracy = accuracy_score(X_train_prediction, Y_train)
print('Accuracy score of the training data : ', training_data_accuracy)
    Accuracy score of the training data : 0.7866449511400652
# accuracy score on the test data
X_test_prediction = classifier.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)
print('Accuracy score of the test data : ', test_data_accuracy)
    Accuracy score of the test data : 0.7727272727272727
Making a Predictive System
input_data = (5,166,72,19,175,25.8,0.587,51)
# changing the input_data to numpy array
input_data_as_numpy_array = np.asarray(input_data)
# reshape the array as we are predicting for one instance
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
# standardize the input data
std_data = scaler.transform(input_data_reshaped)
print(std_data)
prediction = classifier.predict(std_data)
```

 $[-0.84488505 \ -1.12339636 \ -0.16054575 \ \dots \ -0.68442195 \ -0.36506078$