Importing the Dependencies

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

PIMA Diabetes Dataset

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

pd.read_csv?

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

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | DiabetesPedigreeFu |
|-----|-------------|---------|---------------|---------------|---------|------|--------------------|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | |
| - 4 | | | | | | |) |

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 | DiabetesPedigreeFu |
|-------|-------------|------------|---------------|---------------|------------|------------|--------------------|
| count | 768.000000 | 768.000000 | 768.000000 | 768.000000 | 768.000000 | 768.000000 | 768. |
| mean | 3.845052 | 120.894531 | 69.105469 | 20.536458 | 79.799479 | 31.992578 | 0. |
| std | 3.369578 | 31.972618 | 19.355807 | 15.952218 | 115.244002 | 7.884160 | 0. |
| min | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0. |
| 25% | 1.000000 | 99.000000 | 62.000000 | 0.000000 | 0.000000 | 27.300000 | 0. |
| 50% | 3.000000 | 117.000000 | 72.000000 | 23.000000 | 30.500000 | 32.000000 | 0. |
| 75% | 6.000000 | 140.250000 | 80.000000 | 32.000000 | 127.250000 | 36.600000 | 0. |
| max | 17.000000 | 199.000000 | 122.000000 | 99.000000 | 846.000000 | 67.100000 | 2. |

diabetes_dataset['Outcome'].value_counts()

0 500

Name: Outcome, dtype: int64

0 --> Non-Diabetic

1 -> Diabetic

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

```
1
               Pregnancies
                              Glucose BloodPressure SkinThickness
                                                                        Insulin
                                                                                       BMI DiabetesPedigreeFunction
                                                                                                                           Age
      Outcome
                                                                      68.792000 30.304200
         0
                  3.298000 109.980000
                                            68.184000
                                                           19.664000
                                                                                                            0.429734 31.190000
                  4.865672 141.257463
                                            70.824627
                                                           22.164179 100.335821 35.142537
                                                                                                            0.550500 37.067164
         1
# 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
                    6
                           148
                                                                        33.6
                                           66
                                                          29
                                                                     0
                                                                       26.6
     1
                    1
                            85
     2
                    8
                           183
                                           64
                                                           a
                                                                    0
                                                                       23.3
     3
                    1
                            89
                                           66
                                                          23
                                                                   94
                                                                       28.1
     4
                    0
                           137
                                                          35
                                                                   168
                                                                       43.1
                                                                  180 32.9
     763
                   10
                           101
                                           76
                                                          48
     764
                    2
                           122
                                           70
                                                          27
                                                                     0 36.8
     765
                    5
                           121
                                           72
                                                          23
                                                                   112 26.2
     766
                                           60
                                                                       30.1
                    1
                           126
                                                           a
                                                                     0
     767
                    1
                            93
                                           70
                                                          31
                                                                     0 30.4
          DiabetesPedigreeFunction Age
     0
                             0.627
     1
                             0.351
                                     31
     2
                             0.672
                                     32
     3
                             0.167
                                     21
     4
                             2.288
                                     33
     763
                             0.171
                                     63
     764
                             0.340
                                     27
     765
                             0.245
                             0.349
     766
                                     47
     767
                             0.315
                                     23
     [768 rows x 8 columns]
print(Y)
     0
            1
     1
            0
     2
            1
     3
            0
     4
            1
     763
            0
     764
     765
            0
     766
            1
     767
     Name: Outcome, Length: 768, dtype: int64
Data Standardization
scaler = StandardScaler()
scaler.fit(X)

▼ StandardScaler

     StandardScaler()
standardized_data = scaler.transform(X)
print(standardized_data)
     \hbox{\tt [[ 0.63994726 \ 0.84832379 \ 0.14964075 \dots \ 0.20401277 \ 0.46849198]}
        1.4259954 ]
      [-0.84488505 \ -1.12339636 \ -0.16054575 \ \dots \ -0.68442195 \ -0.36506078
```

```
-0.19067191]
     [ \ 1.23388019 \ \ 1.94372388 \ \ -0.26394125 \ \dots \ \ -1.10325546 \ \ 0.60439732
     -0.10558415]
     -0.27575966]
     [-0.84488505 0.1597866 -0.47073225 ... -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\ \dots\ \ 0.20401277\ \ 0.46849198
      1.4259954 ]
      \hbox{ $[-0.84488505 \ -1.12339636 \ -0.16054575 \ \dots \ -0.68442195 \ -0.36506078 $ } 
     -0.19067191]
     -0.10558415]
     -0.275759661
     [-0.84488505 \quad 0.1597866 \quad -0.47073225 \ \dots \ -0.24020459 \ -0.37110101
      1.17073215]
     -0.87137393]]
    0
    1
    2
         1
    3
         0
    4
    763
         0
    764
         0
    765
    766
         1
    767
    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)
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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)
print(prediction)
if (prediction[0] == 0):
 print('The person is not diabetic')
else:
 print('The person is diabetic')
    0.34768723 1.51108316]]
    [1]
    The person is diabetic
    /usr/local/lib/python3.9/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but StandardScaler was fit
     warnings.warn(
```

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