Importing the Dependencies

```
In [1]: 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

Out[4]:

	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
4							•

```
In [5]: # number of rows and Columns in this dataset
diabetes_dataset.shape
```

Out[5]: (768, 9)

In [6]: # getting the statistical measures of the data
diabetes_dataset.describe()

Out[6]:

	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	
4							•

In [7]: diabetes_dataset['Outcome'].value_counts()

Out[7]: 0 500

1 268

Name: Outcome, dtype: int64

0 --> Non-Diabetic

1 --> Diabetic

In [8]: diabetes_dataset.groupby('Outcome').mean()

Out[8]:

		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	Diabe
0	utcome							
	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	
4								•

```
In [9]: # separating the data and labels
X = diabetes_dataset.drop(columns = 'Outcome', axis=1)
Y = diabetes_dataset['Outcome']
```

```
In [10]: print(X)
                                                          SkinThickness
                Pregnancies
                              Glucose
                                        BloodPressure
                                                                           Insulin
                                                                                      BMI \
                                   148
                                                                                     33.6
          0
                           6
                                                     72
                                                                      35
                                                                                  0
          1
                           1
                                    85
                                                     66
                                                                      29
                                                                                  0
                                                                                     26.6
          2
                           8
                                                                       0
                                                                                  0
                                                                                     23.3
                                   183
                                                     64
          3
                           1
                                    89
                                                     66
                                                                      23
                                                                                 94
                                                                                     28.1
          4
                           0
                                   137
                                                     40
                                                                      35
                                                                               168
                                                                                     43.1
          763
                          10
                                   101
                                                     76
                                                                      48
                                                                               180
                                                                                     32.9
          764
                           2
                                   122
                                                     70
                                                                      27
                                                                                  0
                                                                                     36.8
                           5
                                                                                     26.2
          765
                                   121
                                                     72
                                                                      23
                                                                               112
          766
                           1
                                   126
                                                     60
                                                                                  0
                                                                                     30.1
                                                                       0
          767
                           1
                                    93
                                                     70
                                                                      31
                                                                                  0
                                                                                     30.4
                DiabetesPedigreeFunction
                                             Age
          0
                                     0.627
                                               50
          1
                                     0.351
                                               31
          2
                                     0.672
                                              32
          3
                                     0.167
                                              21
          4
                                     2.288
                                              33
                                        . . .
          763
                                     0.171
                                              63
                                     0.340
          764
                                              27
          765
                                     0.245
                                              30
          766
                                     0.349
                                              47
          767
                                     0.315
                                              23
          [768 rows x 8 columns]
In [11]: | print(Y)
          0
                  1
          1
                  0
          2
                  1
          3
                  0
          4
                  1
          763
                  0
          764
                  0
          765
                  0
          766
                  1
          767
          Name: Outcome, Length: 768, dtype: int64
```

Data Standardization

```
In [12]: scaler = StandardScaler()
In [13]: scaler.fit(X)
Out[13]: StandardScaler()
```

```
In [14]:
      standardized data = scaler.transform(X)
In [15]: | 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
                 -0.27575966]
       1.17073215]
       -0.87137393]]
In [16]: | X = standardized_data
      Y = diabetes_dataset['Outcome']
In [17]:
      print(X)
      print(Y)
      [ 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
       -0.27575966]
       [-0.84488505 0.1597866 -0.47073225 ... -0.24020459 -0.37110101
        1.17073215]
                         0.04624525 ... -0.20212881 -0.47378505
       [-0.84488505 -0.8730192
        -0.87137393]]
      0
           1
           0
      1
      2
           1
      3
           0
      4
           1
      763
           0
      764
           0
      765
           0
      766
           1
      767
      Name: Outcome, Length: 768, dtype: int64
```

Train Test Split

Training the Model

```
In [20]: classifier = svm.SVC(kernel='linear')
In [21]: #training the support vector Machine Classifier
      classifier.fit(X_train, Y_train)
Out[21]: SVC(kernel='linear')
```

Model Evaluation

Accuracy Score

```
In [22]: # accuracy score on the training data
    X_train_prediction = classifier.predict(X_train)
    training_data_accuracy = accuracy_score(X_train_prediction, Y_train)

In [23]: print('Accuracy score of the training data : ', training_data_accuracy)
    Accuracy score of the training data : 0.7866449511400652

In [24]: # accuracy score on the test data
    X_test_prediction = classifier.predict(X_test)
    test_data_accuracy = accuracy_score(X_test_prediction, Y_test)

In [25]: print('Accuracy score of the test data : ', test_data_accuracy)
    Accuracy score of the test data : 0.7727272727272727
```

Making a Predictive System

```
In [26]: 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
```

In []: