importing the dependencies

```
In [4]: import pandas as pd
import numpy as np
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 Data Processing

PIMA Diabetes Dataset

Out[19]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunc
0	6	148	72	35	0	33.6	0
1	1	85	66	29	0	26.6	0
2	8	183	64	0	0	23.3	0
3	1	89	66	23	94	28.1	0
4	0	137	40	35	168	43.1	2
763	10	101	76	48	180	32.9	0
764	2	122	70	27	0	36.8	0
765	5	121	72	23	112	26.2	0
766	1	126	60	0	0	30.1	0
767	1	93	70	31	0	30.4	0

768 rows × 9 columns

 \triangleleft

In [21]: #printing the frist 5 rows
diabetes_data.head()

Out[21]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunctic
0	6	148	72	35	0	33.6	0.62
1	1	85	66	29	0	26.6	0.35
2	8	183	64	0	0	23.3	0.67
3	1	89	66	23	94	28.1	0.1€
4	0	137	40	35	168	43.1	2.28
< ■							•

In [20]: #number of rows and columns in this dataset
diabetes_data.shape

Out[20]: (768, 9)

In [22]: #getting statistcal measures of the dataset
 diabetes_data.describe()

Out[22]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	Diabet
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 [23]: diabetes_data['Outcome'].value_counts()

Out[23]: Outcome

0 500 1 268

Name: count, dtype: int64

0 --> Non-Diabetic

1 --> Diabetic

```
In [25]:
          diabetes_data.groupby('Outcome').mean()
Out[25]:
                    Pregnancies
                                 Glucose BloodPressure SkinThickness
                                                                        Insulin
                                                                                   BMI Diak
           Outcome
                 0
                       3.298000 109.980000
                                              68.184000
                                                           19.664000
                                                                     68.792000 30.304200
                 1
                       4.865672 141.257463
                                              70.824627
                                                                    100.335821 35.142537
                                                           22.164179
In [27]: #separeting data andlabels
          x = diabetes_data.drop(columns='Outcome', axis=1)
          y = diabetes_data['Outcome']
```

```
In [28]: print(x)
print(y)
```

т	,	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BM
I ' 0 6	\	6	148	72	35	0	33.
1		1	85	66	29	0	26.
6 2 3 3 1 4		8	183	64	0	0	23.
		1	89	66	23	94	28.
		0	137	40	35	168	43.
76: 9	3	10	101	76	48	180	32.
76 ₄	4	2	122	70	27	0	36.
765 2 766 1 767	5	5	121	72	23	112	26.
	6	1	126	60	0	Θ	30.
	7	1	93	70	31	0	30.

	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
764	0.340	27
765	0.245	30
766	0.349	47
767	0.315	23

```
[768 rows x 8 columns]
0
        1
1
        0
2
        1
3
        0
        1
763
       0
764
       0
765
       0
766
        1
767
```

Name: Outcome, Length: 768, dtype: int64

Data Standardizition

```
In [29]: | scaler = StandardScaler()
In [30]: scaler.fit(x)
Out[30]:
         ▼ StandardScaler
         StandardScaler()
In [31]:
         standardized data = scaler.transform(x)
In [32]: print(standardized data)
         [[ 0.63994726  0.84832379  0.14964075  ...  0.20401277  0.46849198
            1.4259954 ]
          [-0.84488505 \ -1.12339636 \ -0.16054575 \ \dots \ -0.68442195 \ -0.36506078
           -0.190671911
          -0.10558415]
          [ 0.3429808
                        0.00330087 \quad 0.14964075 \quad \dots \quad -0.73518964 \quad -0.68519336
           -0.27575966]
          [-0.84488505 0.1597866 -0.47073225 ... -0.24020459 -0.37110101
            1.170732151
          [-0.84488505 -0.8730192 0.04624525 ... -0.20212881 -0.47378505
           -0.87137393]]
In [33]: x = standardized data
         y = diabetes data['Outcome']
In [34]:
        print(x)
         [[0.63994726 \quad 0.84832379 \quad 0.14964075 \quad \dots \quad 0.20401277 \quad 0.46849198]
            1.4259954 1
          [-0.84488505 -1.12339636 -0.16054575 ... -0.68442195 -0.36506078
           -0.19067191]
          -0.105584151
          [ 0.3429808
                        0.00330087  0.14964075  ... -0.73518964 -0.68519336
           -0.27575966]
          [-0.84488505 \quad 0.1597866 \quad -0.47073225 \quad \dots \quad -0.24020459 \quad -0.37110101
            1.170732151
          [-0.84488505 -0.8730192  0.04624525 ... -0.20212881 -0.47378505
           -0.87137393]]
```

```
In [35]:
          print(y)
                  1
          1
                  0
          2
                  1
          3
                  0
                  1
          763
                  0
          764
                  0
          765
                  0
          766
                  1
          767
          Name: Outcome, Length: 768, dtype: int64
```

Training and Test split

```
In [42]: x_train,x_test,y_train,y_test= train_test_split(x,y,test_size = 0.2,st
In [43]: print(x.shape,x train.shape,x test.shape)
        (768, 8) (614, 8) (154, 8)
In [44]:
        print(x)
        print (y)
        [[ 0.63994726
                      0.84832379 0.14964075 ... 0.20401277
                                                            0.46849198
           1.4259954
         [-0.84488505 \ -1.12339636 \ -0.16054575 \ \dots \ -0.68442195 \ -0.36506078
          -0.19067191]
          -0.10558415]
                      0.00330087 \quad 0.14964075 \quad \dots \quad -0.73518964 \quad -0.68519336
         [ 0.3429808
          -0.27575966]
         [-0.84488505 \quad 0.1597866 \quad -0.47073225 \quad \dots \quad -0.24020459 \quad -0.37110101
           1.17073215]
          -0.87137393]]
        0
               1
        1
               0
        2
               1
        3
               0
               1
        763
               0
        764
               0
        765
               0
        766
               1
        767
        Name: Outcome, Length: 768, dtype: int64
```

Model training --> SVM Model

Model Evaluation

```
In [50]: # accuracy on training data
    x_train_prediction = classifier.predict(x_train)
    training_data_accuracy = accuracy_score(x_train_prediction,y_train)

In [51]: print("Accuracy on training data : " ,training_data_accuracy)
    Accuracy on training data : 0.7866449511400652

In [52]: # accuracy on test data
    x_test_prediction = classifier.predict(x_test)
    test_data_accuracy = accuracy_score(x_test_prediction,y_test)

In [53]: print("Accuracy on test data : " ,test_data_accuracy)
    Accuracy on test data : 0.7727272727272727
```

Making a Predictive System

```
In [84]: input_data = (5,121,72,23,112,26.2,0.245,30)
# changing the input_data to numpy arrary
input_data_as_numpy_array = np.asarray(input_data)
#reshape the np as we are predicting for one instence
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
#to print the standardiziation data
print(std_data)

#to predict the input data
prediction = classifier.predict(std_data)

if (prediction[0]==0):
    print("The person is not disabetic")
else:
    print("The person is disabetic")
```

```
[[ 1.53084665 - 0.99820778 - 0.36733675 - 1.28821221 - 0.69289057 - 1.2047 9085 - 0.99626558 - 0.0204964 ]] The person is not disabetic
```