Importing the Dependencies

```
import numpy as np
import pandas as pd
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('/content/diabetes.csv')
```

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

Pregnancies		Glucose	BloodPressure	 DiabetesPedigreeFunction
Age	Outcome			-
0	6	148	72	 0.627
50	1			
1	1	85	66	 0.351
31	0			
2	8	183	64	 0.672
32	1			
3	1	89	66	 0.167
21	0			
4	0	137	40	 2.288
33	1			

[5 rows x 9 columns]

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	 Age	Outcome
count	768.000000	768.000000	 768.000000	768.000000
mean	3.845052	120.894531	 33.240885	0.348958
std	3.369578	31.972618	 11.760232	0.476951
min	0.000000	0.00000	 21.000000	0.000000
25%	1.000000	99.000000	 24.000000	0.000000
50%	3.000000	117.000000	 29.000000	0.000000
75%	6.000000	140.250000	 41.000000	1.000000
max	17.000000	199.000000	 81.000000	1.000000

```
[8 rows x 9 columns]
diabetes dataset['Outcome'].value counts()
0
     500
     268
1
Name: Outcome, dtype: int64
0 --> Non-Diabetic
1 --> Diabetic
diabetes_dataset.groupby('Outcome').mean()
         Pregnancies
                           Glucose ... DiabetesPedigreeFunction
Age
Outcome
0
             3.298000
                       109.980000
                                                            0.429734
31.190000
             4.865672
                       141.257463
                                                            0.550500
37.067164
[2 rows x 8 columns]
# separating the data and labels
X = diabetes dataset.drop(columns = 'Outcome', axis=1)
Y = diabetes dataset['Outcome']
print(X)
     Pregnancies Glucose BloodPressure
                                                    BMI
DiabetesPedigreeFunction Age
                        148
                                         72
                                              . . .
                                                   33.6
                6
0.627
        50
1
                1
                         85
                                         66
                                             . . .
                                                   26.6
0.351
        31
                8
                        183
                                         64
                                              . . .
                                                   23.3
0.672
        32
                1
                         89
                                                   28.1
3
                                         66
                                              . . .
0.167
        21
                0
                        137
                                         40
                                                   43.1
                                              . . .
2.288
        33
                        . . .
              . . .
                                        . . .
. . .
               10
                                         76
                                                   32.9
763
                        101
                                             . . .
0.171
        63
764
                2
                        122
                                         70
                                                   36.8
                                             . . .
0.340
        27
                                         72 ...
765
                5
                        121
                                                   26.2
0.245
        30
```

```
766
               1
                      126
                                       60 ...
                                                30.1
0.349
        47
                                       70 ... 30.4
767
               1
                       93
0.315
        23
[768 rows x 8 columns]
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
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(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: 0.7833876221498371
```

```
# 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)
prediction = classifier.predict(input data reshaped)
print(prediction)
if (prediction[0] == 0):
  print('The person is not diabetic')
else:
  print('The person is diabetic')
[1]
The person is diabetic
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:446:
UserWarning: X does not have valid feature names, but SVC was fitted
with feature names
  "X does not have valid feature names, but"
Saving the trained model
import pickle
filename = 'trained model.sav'
pickle.dump(classifier, open(filename, 'wb'))
# loading the saved model
loaded model = pickle.load(open('trained model.sav', 'rb'))
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)
```

```
prediction = loaded_model.predict(input_data_reshaped)
print(prediction)

if (prediction[0] == 0):
    print('The person is not diabetic')

else:
    print('The person is diabetic')

[1]
The person is diabetic

/usr/local/lib/python3.7/dist-packages/sklearn/base.py:446:
UserWarning: X does not have valid feature names, but SVC was fitted with feature names
    "X does not have valid feature names, but"
```