## Importing the dependencies

import numpy as np
#for creating numpy arrays
import pandas as pd
#for creating data frames
from sklearn.preprocessing import StandardScaler #to standardise our data
from sklearn.model\_selection import train\_test\_split #to split data into training and test data
from sklearn import svm
from sklearn.metrics import accuracy\_score

## Data Collection and Analysis

## PIMA Diabetes Dataset

#loading the diabetes data set to a pandas Dataframe
diabetes\_dataset = pd.read\_csv('/content/diabetes.csv')

pd.read\_csv?
#a google collab feature

#printing the first five 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	
							<b>→</b>

#number of rows and columns in this dataset
diabetes\_dataset.shape

#meaning the data is taken from 768 people and 9 columns worth data was recorded

(768, 9)

#getting statistical measures of the data
diabetes\_dataset.describe()

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Dia
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	<b>&gt;</b>

diabetes\_dataset['Outcome'].value\_counts()

0 5001 268

Name: Outcome, dtype: int64

0 --> Non diabetic 1 --> Diabetic

diabetes\_dataset.groupby('Outcome').mean()

```
Pregnancies
                               Glucose BloodPressure SkinThickness
                                                                         Insulin
                                                                                        BMI DiabetesPedigreeF
      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 the labels
X = diabetes_dataset.drop(columns= 'Outcome', axis=1)
#to drop a column, drop=1, if a row then drop=0
Y = diabetes_dataset['Outcome']
print(X)
          Pregnancies Glucose BloodPressure SkinThickness Insulin
                                                                         BMI \
     0
                           148
                                                                        33.6
                    6
                                            72
                                                           35
                                                                     0
     1
                    1
                            85
                                            66
                                                           29
                                                                     0
                                                                        26.6
     2
                    8
                           183
                                                            0
                                                                     0
                                                                        23.3
     3
                    1
                            89
                                            66
                                                           23
                                                                    94
                                                                        28.1
     4
                    0
                           137
                                           40
                                                           35
                                                                   168
                                                                       43.1
     763
                                                                   180
                                                                        32.9
                   10
                           101
                                            76
                                                           48
     764
                    2
                           122
                                            70
                                                           27
                                                                     0
                                                                        36.8
     765
                    5
                                            72
                                                           23
                                                                   112 26.2
                           121
     766
                    1
                           126
                                            60
                                                            0
                                                                     0
                                                                        30.1
     767
                    1
                            93
                                            70
                                                           31
                                                                     0
                                                                        30.4
          DiabetesPedigreeFunction
     0
                             0.627
                                     31
     1
                             0.351
     2
                             0.672
                                     32
     3
                             0.167
                                     21
     4
                             2.288
     763
                             0.171
                                     63
     764
                             0.340
                                     27
     765
                             0.245
                                     30
                             0.349
     766
                                     47
                             0.315
                                     23
     [768 rows x 8 columns]
print(Y)
     0
            1
            0
     1
     2
            1
     3
            0
     4
            1
     763
            9
     764
            0
     765
            0
     766
            1
     767
     Name: Outcome, Length: 768, dtype: int64
Data Standardisation
scaler = StandardScaler()
#implies we are taking one instance of Standard Scaler func
Double-click (or enter) to edit
scaler.fit(X)
      ▼ StandardScaler
     StandardScaler()
```

https://colab.research.google.com/drive/1dcJUsvSny6Q2CnDvm9j2nDtB-jUpASyR#scrollTo=7BujBL krDD8&printMode=true

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]
    -0.27575966]
     [-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 ...  0.20401277  0.46849198
      1.4259954 1
     [-0.84488505 -1.12339636 -0.16054575 ... -0.68442195 -0.36506078
     -0.19067191]
     [ 1.23388019 1.94372388 -0.26394125 ... -1.10325546 0.60439732
     -0.10558415]
    [ 0.3429808
               -0.27575966]
     [-0.84488505 \quad 0.1597866 \quad -0.47073225 \ \dots \ -0.24020459 \ -0.37110101
      1.17073215]
    -0.87137393]]
    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
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') #support vector classifier
#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: 0.7866449511400652
#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 a 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/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but StandardScaler was fi
      warnings.warn(
    - 4 ■
```

✓ 0s completed at 1:07 PM

X