#import 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
#load dataset diabetec to pandas dataframe
diabetes\_dataset=pd.read\_csv('/content/diabetes.csv')

pd.read\_csv?

#printing the first five rows of dataset
diabetes\_dataset.head()

₹		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
	0	6	148	72	35	0	33.6	0.627	50	1
	1	1	85	66	29	0	26.6	0.351	31	0
	2	8	183	64	0	0	23.3	0.672	32	1
	3	1	89	66	23	94	28.1	0.167	21	0
	4	0	137	40	35	168	43.1	2.288	33	1

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

**→** (768, 9)

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

<del></del>		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
	count	768.000000	768.000000	768.000000	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	0.471876	33.240885	0.348958
	std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.331329	11.760232	0.476951
	min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.078000	21.000000	0.000000
	25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	0.243750	24.000000	0.000000
	50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	0.372500	29.000000	0.000000
	75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	0.626250	41.000000	1.000000
	max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.420000	81.000000	1.000000
	1									

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



count

dtunar int64

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

```
<del>_</del>__
               Pregnancies
                               Glucose BloodPressure SkinThickness
                                                                         Insulin
                                                                                        BMI DiabetesPedigreeFunction
                                                                                                                             Age
      Outcome
         0
                   3.298000 109.980000
                                             68.184000
                                                            19.664000
                                                                       68.792000 30.304200
                                                                                                              0.429734 31.190000
                   4.865672 141.257463
                                             70.824627
                                                            22.164179 100.335821 35.142537
                                                                                                              0.550500 37.067164
#separating data and labels
x=diabetes_dataset.drop(columns='Outcome',axis=1)
#for coliumn axis=1 ,row=axis=0
y=diabetes_dataset['Outcome']
print(x)
                                BloodPressure SkinThickness Insulin
                                                                          BMI \
₹
          Pregnancies Glucose
     0
                    6
                           148
                                            72
                                                           35
                                                                      0
                                                                        33.6
     1
                    1
                            85
                                            66
                                                           29
                                                                      0
                                                                        26.6
                    8
                                                            0
                                                                        23.3
                           183
                                            64
     3
                            89
                                            66
                                                           23
                                                                     94
                                                                        28.1
                    1
     4
                    0
                           137
                                            40
                                                           35
                                                                    168
                                                                        43.1
                   10
                            101
                                                                    180
                                                                        32.9
     763
                                            76
                                                           48
     764
                    2
                           122
                                            70
                                                           27
                                                                     a
                                                                        36.8
     765
                    5
                            121
                                            72
                                                           23
                                                                    112 26.2
     766
                                            60
                                                            0
                                                                      0
                                                                         30.1
                    1
                            126
                                                           31
                                                                      0
     767
                    1
                                            70
                                                                        30.4
                            93
          DiabetesPedigreeFunction
                                    Age
     0
                                      50
                             0.627
                             0.351
                                      31
     1
     2
                             0.672
                                      32
     3
                              0.167
                                      21
     4
                              2.288
                                     33
     763
                              0.171
                                     63
                             0.340
                                      27
     764
     765
                             0.245
                                      30
     766
                             0.349
                                      47
     767
                             0.315
     [768 rows x 8 columns]
print(y)
 <del>_</del>__
     0
            1
            0
     2
            1
     3
     4
            1
     763
            0
     764
            0
     765
            0
     766
            1
     767
            0
     Name: Outcome, Length: 768, dtype: int64
#data standardization
scaler=StandardScaler()
scaler.fit(x)
      ▼ StandardScaler ① ?
     StandardScaler()
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]
     [ \ 1.23388019 \ \ 1.94372388 \ \ -0.26394125 \ \dots \ \ -1.10325546 \ \ 0.60439732
      -0.10558415]
     [ \ 0.3429808 \quad 0.00330087 \quad 0.14964075 \ \dots \ -0.73518964 \ -0.68519336
      -0.27575966]
     1.17073215]
     -0.87137393]]
x=standardized_data
y=diabetes_dataset['Outcome']
x--- data ,y----model
print(x)
→ [[ 0.63994726 0.84832379 0.14964075 ... 0.20401277 0.46849198
       1.4259954 ]
     [-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 \quad 0.00330087 \quad 0.14964075 \ \dots \ -0.73518964 \ -0.68519336
      -0.27575966]
     [-0.84488505 \quad 0.1597866 \quad -0.47073225 \ \dots \ -0.24020459 \ -0.37110101
       1.17073215]
     -0.87137393]]
print(y)
#all data in x and all labels in y
₹
    0
           1
           0
    2
           1
           0
    3
    4
           1
    763
           0
    764
           a
    765
    766
           1
    767
    Name: Outcome, Length: 768, dtype: int64
#train test spilt
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 model
classifier=svm.SVC(kernel='linear')
#training suppot vector machine classifier
classifier.fit(x_train,y_train)
₹
                   (i) (?)
           SVC
     SVC(kernel='linear')
#evaluation of model how many time model is correctly
#model_evaluation
#accuracy score on training data
x_train_prediction=classifier.predict(x_train)
training_data_accuracy=accuracy_score(x_train_prediction,y_train)
```

```
print('accuracy score',training_data_accuracy)
⇒ accuracy score 0.7866449511400652
#accuracy score on test data
x_test_prediction=classifier.predict(x_test)
test_data_accuracy=accuracy_score(x_test_prediction,y_test)
print('accuracy score',test_data_accuracy)
⇒ accuracy score 0.7727272727272727
#overtraining:overfitting model work on training data but not good on test data
#making predictive system
input_data=(1,85,66,29,0,26.6,0.351,31)
#change input_data to numpy array
input_data_as_numpy_array=np.asarray(input_data)
#reshape array as we are predicting for one instancee
input_data_reshaped=input_data_as_numpy_array.reshape(1,-1)
#standardize input data
std_data=scaler.transform(input_data_reshaped)
print(std_data)
→ [[-0.84488505 -1.12339636 -0.16054575 0.53090156 -0.69289057 -0.68442195
       -0.36506078 -0.19067191]]
     /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:2739: UserWarning: X does not have valid feature names, but Standard
      warnings.warn(
prediction=classifier.predict(std_data)
print(prediction)
→ [0]
if(prediction[0]==0):
 print('person is not diabetic')
else:
 print('person is diabetic')
→ person is not diabetic
Start coding or generate with AI.
```