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## Importing the 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

# 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()

<b>→</b>		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunc	tion	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

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

**→** (768, 9)

 $\mbox{\tt\#}$  getting the statistical measures of the data diabetes\_dataset.describe()

$\overline{\Rightarrow}$		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	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

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

count

Outcome 500
1 268

dtuma inté A

```
diabetes_dataset.groupby('Outcome').mean()
<del>_</del>
              Pregnancies
                             Glucose BloodPressure SkinThickness
                                                                     Insulin
                                                                                   BMI DiabetesPedigreeFunction
     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 \quad 37.067164
# separating the data and labels
X = diabetes_dataset.drop(columns = 'Outcome', axis=1)
Y = diabetes_dataset['Outcome']
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 ... -1.10325546 0.60439732
       -0.10558415]
     -0.275759661
      [-0.84488505 \quad 0.1597866 \quad -0.47073225 \ \dots \ -0.24020459 \ -0.37110101
       1.17073215]
      [-0.84488505 \ -0.8730192 \quad 0.04624525 \ \dots \ -0.20212881 \ -0.47378505
       -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 ]
      [-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]
     [-0.84488505 \ -0.8730192 \quad 0.04624525 \ \dots \ -0.20212881 \ -0.47378505
       -0.87137393]]
    0
           1
    1
           0
```

Age

766 1 767 0

1

0

1

0

0

2

3

4

763

764

```
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
     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)
\Rightarrow Accuracy score of the test data : 0.77272727272727
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)
# 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]]
     The person is diabetic
    /usr/local/lib/python3.10/dist-packages/sklearn/base.py:493: UserWarning: X does not have valid feature names, but StandardScaler wa
      warnings.warn(
```

## 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)
\ensuremath{\text{\#}} 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 = loaded_model.predict(std_data)
print(prediction)
if (prediction[0] == 0):
 print('The person is not diabetic')
else:
  print('The person is diabetic')

→ [[ 0.3429808  1.41167241  0.14964075 -0.09637905  0.82661621 -0.78595734

        0.34768723 1.51108316]]
     [1]
     The person is diabetic
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:493: UserWarning: X does not have valid feature names, but StandardScaler wa
      warnings.warn(
     4
```

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