#### **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')

pd.read\_csv?

# printing the first 5 rows of the dataset
diabetes\_dataset.head()

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

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

(768, 9)

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

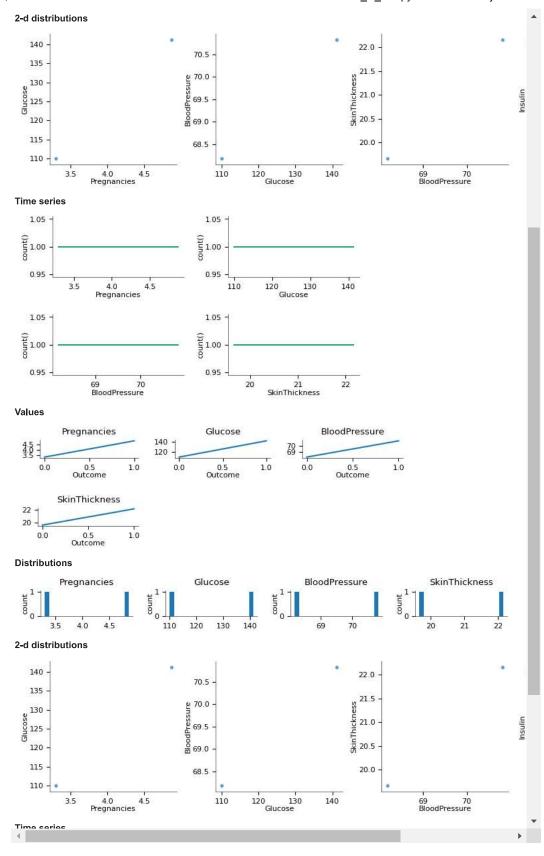
	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFu
count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.
mean	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578	0.
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	0.
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	0.
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	0.
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.

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

0 500 1 268

Name: Outcome, dtype: int64

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



```
# separating the data and labels
X = diabetes_dataset.drop(columns = 'Outcome', axis=1)
Y = diabetes_dataset['Outcome']
print(X)
         Pregnancies Glucose BloodPressure SkinThickness Insulin
                                                                     BMI \
    0
                   6
                          148
                                          72
                                                        35
                                                                  0 33.6
    1
                   1
                          85
                                          66
                                                        29
                                                                  0 26.6
    2
                   8
                          183
                                          64
                                                        0
                                                                 0 23.3
                                                        23
    3
                   1
                          89
                                          66
                                                                 94 28.1
     4
                   0
                          137
                                                        35
                                                                168 43.1
                                         . . .
                                                        . . .
    763
                  10
                          101
                                         76
                                                        48
                                                                180 32.9
    764
                   2
                          122
                                         70
                                                        27
                                                                 0 36.8
    765
                   5
                          121
                                         72
                                                        23
                                                                112 26.2
    766
                                          60
                                                                 0 30.1
                   1
                          126
                                                         0
    767
                   1
                           93
                                          70
                                                        31
                                                                  0 30.4
         DiabetesPedigreeFunction Age
    0
                            0.627
    1
                            0.351
                                    31
    2
                            0.672
                                    32
    3
                            0.167
                                    21
    4
                            2.288
                                   33
                            0.171
    763
                                   63
    764
                            0.340
                                   27
                            0.245
    766
                            0.349
                                    47
    767
                            0.315
                                   23
    [768 rows x 8 columns]
print(Y)
    0
           1
    1
           0
    2
           0
    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
     StandardScaler()
standardized_data = scaler.transform(X)
```

```
-0.87137393]]
```

https://colab.research.google.com/drive/1WA7D8sTxvia0j0RMsVSvzu8dqYYIY4wv#scrollTo=2wiU8pXIDzG1&printMode=true

 $[ [ \ 0.63994726 \ \ 0.84832379 \ \ 0.14964075 \ \dots \ \ 0.20401277 \ \ 0.46849198$ 

[-0.84488505 -1.12339636 -0.16054575 ... -0.68442195 -0.36506078

[ 1.23388019 1.94372388 -0.26394125 ... -1.10325546 0.60439732

 $[ \ 0.3429808 \quad 0.00330087 \quad 0.14964075 \ \dots \ -0.73518964 \ -0.68519336$ 

 $[-0.84488505 \quad 0.1597866 \quad -0.47073225 \ \dots \ -0.24020459 \ -0.37110101$ 

print(standardized\_data)

1.4259954 ]

-0.19067191]

-0.10558415]

-0.27575966]

1.17073215]

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```
X = standardized_data
Y = diabetes_dataset['Outcome']
print(X)
print(Y)
     \hbox{\tt [[ 0.63994726 \ 0.84832379 \ 0.14964075 \dots \ 0.20401277 \ 0.46849198]}
       1.4259954 ]
     [-0.84488505 -1.12339636 -0.16054575 ... -0.68442195 -0.36506078
      -0.19067191]
      -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]]
    a
    1
           0
    2
           1
    3
           a
    4
           1
    763
          0
    764
           a
    765
           0
    766
          1
    767
    Name: Outcome, Length: 768, dtype: int64
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)
```

#### Model Evaluation

▼ SVC SVC(kernel='linear')

### **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 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(
```

✓ 0s completed at 11:12 AM

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