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

# 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()

| <del>_</del> |   | 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       |

# printing the first 5 rows of the 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       |
|   | <b>A</b> | ^           | 127     | 40            | 35            | 160     | 12 1 | ን ንይይ                    | 33  | 1       |
|   | ١.       |             |         |               |               |         |      |                          |     |         |

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

**→** (768, 9)

#getting the statistical measures of the 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   |
|              | may<br>4 | 17 000000   | 100 000000 | 122 000000    | 00 000000     | 846 000000 | 67 100000  | 2 420000                 | <u> </u>   | 1 000000   |

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

 Outcome
 0

 0
 500

 1
 268

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

```
₹
                              Glucose BloodPressure SkinThickness
                                                                        Insulin
               Pregnancies
                                                                                       BMI DiabetesPedigreeFunction
                                                                                                                            Age
      Outcome
        0
                  3.298000 109.980000
                                            68.184000
                                                           19.664000 68.792000 30.304200
                                                                                                             0.429734 31.190000
                  A 965670 1A1 057A62
                                            70 22/627
                                                           22 16/170 100 225221 25 1/2527
                                                                                                             N EENENN 27 NE716/
# separating the data and labels
X = diabetes_dataset.drop(columns = 'Outcome', axis=1)
Y = diabetes_dataset['Outcome']
print(X)
          Pregnancies Glucose BloodPressure SkinThickness Insulin
\overline{z}
                                                                        BMT
                    6
                           148
                                           72
                                                           35
                                                                     0 33.6
     1
                    1
                            85
                                           66
                                                           29
                                                                     a
                                                                        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
     763
                   10
                           101
                                           76
                                                          48
                                                                   180 32.9
                                                           27
     764
                    2
                           122
                                           70
                                                                     0 36.8
     765
                    5
                                           72
                                                                   112 26.2
                           121
                                                          23
     766
                    1
                           126
                                           60
                                                           a
                                                                     0 30.1
     767
                            93
                                           70
                                                           31
                                                                     0 30.4
          DiabetesPedigreeFunction
     0
     1
                             0.351
                                     31
                             0.672
     3
                             0.167
                                     21
     4
                             2.288
                                     33
    ..
763
                             0.171
                                     63
                             9.349
     764
                                     27
                             0.245
     765
                                     30
     766
                             0.349
                                     47
     767
                             0.315
                                     23
     [768 rows x 8 columns]
print(Y)
<del>_</del>
    0
            1
     1
            0
     2
            1
     3
            a
     4
           1
     763
            0
     764
     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)
classifier = svm.SVC(kernel='linear')
#training the support vector Machine Classifier
classifier.fit(X_train, Y_train)
              SVC
     SVC(kernel='linear')
# accuracy score on the training data
X_train_prediction = classifier.predict(X_train)
{\tt 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
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/python3.10/dist-packages/sklearn/base.py:465: UserWarning: X does not have valid feature names, but SVC was fitted wi
       warnings.warn(
import pickle
filename = 'diabetes_model.sav'
pickle.dump(classifier, open(filename, 'wb'))
# loading the saved model
loaded_model = pickle.load(open('diabetes_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')
 print('The person is diabetic')
₹
     [1]
     The person is diabetic
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:465: UserWarning: X does not have valid feature names, but SVC was fitted wi
       warnings.warn(
for column in X.columns:
 print(column)
\rightarrow
    Pregnancies
     Glucose
     BloodPressure
     SkinThickness
     Insulin
     RMT
     DiabetesPedigreeFunction
```