```
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

```
In [2]:
```

```
diabetes_dataset = pd.read_csv('B:\VSCODE\Multiple Disease Prediction System\Diabetes Pre
diction System\diabetes.csv')
diabetes_dataset.head()
```

Out[2]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	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

In [3]:

```
diabetes_dataset.shape
```

Out[3]:

(768, 9)

In [4]:

```
diabetes dataset.describe()
```

Out[4]:

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

In [5]:

```
diabetes_dataset['Outcome'].value_counts()
```

```
0
     500
     268
1
Name: Outcome, dtype: int64
In [6]:
diabetes dataset.groupby('Outcome').mean()
Out[6]:
         Pregnancies
                      Glucose BloodPressure SkinThickness
                                                          Insulin
                                                                     BMI DiabetesPedigreeFunction
                                                                                                    Ag
Outcome
           3.298000 109.980000
                                              19.664000
                                                       68.792000 30.304200
                                                                                       0.429734 31.19000
      0
                                 68.184000
            4.865672 141.257463
                                 70.824627
                                              22.164179 100.335821 35.142537
                                                                                       0.550500 37.06716
In [7]:
#sepating data and labels
X = diabetes dataset.drop(columns = 'Outcome', axis = 1)
Y = diabetes dataset['Outcome']
In [8]:
print(X)
     Pregnancies
                    Glucose BloodPressure SkinThickness
                                                               Insulin
                                                                           BMI
0
                        148
                                          72
                                                            35
                                                                          33.6
                                                                      0
                 6
                          85
                                           66
                                                            29
                                                                          26.6
1
                 1
                                                                       0
2
                 8
                         183
                                           64
                                                            0
                                                                       0
                                                                          23.3
3
                                                            23
                                                                      94
                 1
                         89
                                           66
                                                                          28.1
4
                 0
                         137
                                           40
                                                            35
                                                                     168 43.1
               . . .
                         . . .
                                          . . .
                                                           . . .
. .
                                                                     . . .
                                                                           . . .
763
               10
                         101
                                           76
                                                            48
                                                                     180
                                                                          32.9
                                                                          36.8
764
                 2
                        122
                                           70
                                                            27
                                                                      0
765
                5
                         121
                                           72
                                                            23
                                                                     112 26.2
766
                 1
                                           60
                                                                      0 30.1
                         126
                                                            0
767
                                           70
                                                            31
                                                                       0 30.4
                 1
                         93
     DiabetesPedigreeFunction Age
0
                           0.627
1
                           0.351
                                    31
2
                           0.672
                                   32
3
                           0.167
                                   21
4
                           2.288
                                   33
763
                           0.171
                                   63
764
                           0.340
                                    27
765
                           0.245
                                    30
766
                           0.349
                                    47
767
                           0.315
                                    23
[768 rows x 8 columns]
In [9]:
print(Y)
0
        1
1
        0
2
       1
3
       0
4
       1
763
       0
764
       0
```

.

```
767
       0
Name: Outcome, Length: 768, dtype: int64
Data Standarization
In [10]:
scaler = StandardScaler()
In [11]:
scaler.fit(X)
Out[11]:
▼ StandardScaler
StandardScaler()
In [12]:
standardized data = scaler.transform(X)
In [13]:
print(standardized data)
[[0.63994726 \quad 0.84832379 \quad 0.14964075 \quad ... \quad 0.20401277 \quad 0.46849198]
   1.4259954 ]
 [-0.84488505 -1.12339636 -0.16054575 \dots -0.68442195 -0.36506078]
  -0.19067191]
 [ \ 1.23388019 \ \ 1.94372388 \ -0.26394125 \ \dots \ -1.10325546 \ \ 0.60439732
  -0.10558415]
 [ 0.3429808
               -0.275759661
 [-0.84488505 \quad 0.1597866 \quad -0.47073225 \quad \dots \quad -0.24020459 \quad -0.37110101
   1.17073215]
 [-0.84488505 - 0.8730192 \quad 0.04624525 \dots -0.20212881 -0.47378505
  -0.87137393]]
In [14]:
X = standardized data
Y = diabetes_dataset['Outcome']
In [15]:
print(X)
print(Y)
[[0.63994726 \quad 0.84832379 \quad 0.14964075 \dots \quad 0.20401277 \quad 0.46849198]
   1.4259954 ]
 [-0.84488505 -1.12339636 -0.16054575 \dots -0.68442195 -0.36506078]
  -0.19067191]
 [1.23388019 \ 1.94372388 \ -0.26394125 \ \dots \ -1.10325546 \ 0.60439732
  -0.10558415]
 [ 0.3429808
                -0.27575966]
 [-0.84488505 \quad 0.1597866 \quad -0.47073225 \quad \dots \quad -0.24020459 \quad -0.37110101
   1.17073215]
                           0.04624525 ... -0.20212881 -0.47378505
 [-0.84488505 -0.8730192
  -0.87137393]]
0
       1
1
       0
2
       1
3
       0
4
       1
      . .
```

```
764
       0
765
       0
766
       1
767
       0
Name: Outcome, Length: 768, dtype: int64
Train Test Split
In [16]:
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test size = 0.2, stratify = Y,
random state = 2)
In [17]:
print(X.shape, X train.shape, X test.shape)
(768, 8) (614, 8) (154, 8)
Train the model
In [18]:
classifier = svm.SVC(kernel = 'linear')
In [19]:
# training the support vector Machine Classifier
classifier.fit(X_train, Y_train)
Out[19]:
         SVC
SVC(kernel='linear')
Model Evaluation
Accuracy Score
In [20]:
# accuracy score on the training data
X train prediction = classifier.predict(X train)
training data accuracy = accuracy score(X train prediction, Y train)
In [21]:
print('Accuracy score of the training data: ', training data accuracy)
Accuracy score of the training data: 0.7866449511400652
In [22]:
# accuracy score on the test data
X test prediction = classifier.predict(X test)
test data accuracy = accuracy score(X test prediction, Y test)
In [23]:
print('Accuracy score of the test data : ', test data accuracy)
```

Accuracy score of the test data: 0.7727272727272727

763

0

Prediction System that the patient has diabetes or not

```
In [24]:
```

```
input data = (2,197,70,45,543,30.5,0.158,53)
# 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)
[-0.54791859 2.38188392 0.04624525 1.53455054 4.02192191 -0.18943689
 -0.94794368 1.68125866]]
[1]
C:\Users\SATYAM\AppData\Roaming\Python\Python39\site-packages\sklearn\base.py:439: UserWa
rning: X does not have valid feature names, but StandardScaler was fitted with feature na
mes
  warnings.warn(
In [25]:
if (prediction[0] == 0):
 print('The person is not diabetic')
else:
   print('The person is diabetic')
result = classifier.predict([[2,197,70,45,543,30.5,0.158,53]])
The person is diabetic
In [ ]:
```