```
In [1]: import pandas as pd
         import numpy as np
         from sklearn.preprocessing import StandardScaler
         from sklearn.model_selection import train_test_split
         from sklearn import svm
         from sklearn.metrics import accuracy_score
In [2]: df = pd.read_csv('diabetes.csv')
In [3]: df.head()
                                                                        BMI DiabetesPedigreeFunction
Out[3]:
            Pregnancies
                         Glucose BloodPressure SkinThickness
                                                                Insulin
                                                                                                       Age
                                                                                                            Outcome
         0
                      6
                                             72
                                                            35
                                                                       33 6
                                                                                                 0.627
                                                                                                        50
                                                                                                                    1
                             148
                                                                     0
                      1
                                                            29
                                                                     0 26.6
                                                                                                        31
                                                                                                                   0
         1
                              85
                                             66
                                                                                                 0.351
         2
                      8
                                                             0
                                                                       23.3
                                                                                                        32
                             183
                                             64
                                                                     0
                                                                                                 0.672
         3
                              89
                                             66
                                                            23
                                                                    94
                                                                        28.1
                                                                                                 0.167
                                                                                                        21
                                                                                                                    0
                                                                                                 2.288
         4
                      0
                             137
                                             40
                                                            35
                                                                   168 43.1
                                                                                                        33
In [4]: df.shape
Out[4]: (768, 9)
In [5]: df.describe()
Out[5]:
                                        BloodPressure
                                                       SkinThickness
                                                                                              DiabetesPedigreeFunction
                Pregnancies
                               Glucose
                                                                           Insulin
                                                                                                                              Age
                 768.000000
                             768.000000
                                            768.000000
                                                           768.000000
                                                                      768.000000
                                                                                  768.000000
                                                                                                            768.000000
                                                                                                                       768.000000
                                                                                                                                   768
         count
         mean
                   3.845052
                             120.894531
                                             69.105469
                                                            20.536458
                                                                        79.799479
                                                                                   31.992578
                                                                                                              0.471876
                                                                                                                         33.240885
                                             19.355807
                                                            15.952218 115.244002
                                                                                    7.884160
           std
                   3.369578
                              31.972618
                                                                                                              0.331329
                                                                                                                         11.760232
                   0.000000
                               0.000000
                                              0.000000
                                                             0.000000
                                                                         0.000000
                                                                                     0.000000
                                                                                                              0.078000
                                                                                                                         21.000000
           min
          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
                  17.000000
                            199.000000
                                             122.000000
                                                            99.000000
                                                                       846.000000
                                                                                   67.100000
                                                                                                              2.420000
                                                                                                                         81.000000
          max
In [7]: df['Outcome'].value_counts()
Out[7]:
         0
               500
```

0 - non diabetic

Name: Outcome, dtype: int64

1 - diabetic

```
In [8]: X = df.drop(columns='Outcome',axis=1)
    y = df['Outcome']
In [9]: X
```

2 8 183 64 0 0 23.3 0.672 32 3 1 89 66 23 94 28.1 0.667 21 4 0 137 40 35 168 43.1 2.288 33
10
763 10 101 76 48 180 32.9 0.171 63 764 2 122 70 27 0 36.8 0.340 27 765 5 121 72 23 112 26.2 0.245 30 766 1 126 60 0 0 3.01 0.349 47 767 1 93 70 31 0 30.4 0.315 23 768 rows × 8 columns
763 10 101 76 48 180 32.9 0.171 63 764 2 122 70 27 0 38.8 0.340 27 765 5 121 72 23 112 26.2 0.245 30 766 1 126 60 0 0 3.0.1 0.349 47 767 1 93 70 31 0 30.4 0.315 23 768 rows × 8 columns
763 10 101 76 48 180 32.9 0.171 63 764 2 122 70 27 0 36.8 0.340 27 765 5 121 72 23 112 26.2 0.245 30 766 1 126 60 0 0 30.1 0.349 47 767 1 93 70 31 0 30.4 0.315 23 768 rows × 8 columns
765
766
766
767 1 93 70 31 0 30.4 0.315 23 768 rows × 8 columns y
768 rows × 8 columns 768 rows × 8 columns 769
0
1
2]: scaler.fit(X) 2]: StandardScaler() 3]: standardized_data = scaler.transform(X) 4]: standardized_data 4]: array([[0.63994726,
<pre>2]: StandardScaler() 3]: standardized_data = scaler.transform(X) 4]: standardized_data 4]: array([[0.63994726,</pre>
<pre>StandardScaler() standardized_data = scaler.transform(X) standardized_data array([[0.63994726, 0.84832379, 0.14964075,, 0.20401277,</pre>
standardized_data # array([[0.63994726, 0.84832379, 0.14964075,, 0.20401277,
standardized_data # array([[0.63994726, 0.84832379, 0.14964075,, 0.20401277,
array([[0.63994726, 0.84832379, 0.14964075,, 0.20401277,
Til: X array([[0.63994726, 0.84832379, 0.14964075,, 0.20401277,
array([[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],,
0.46849198, 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],,
-0.68519336, -0.27575966], [-0.84488505, 0.1597866 , -0.47073225,, -0.24020459, -0.37110101, 1.17073215],

Out[9]: Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age

35

29

0 33.6

0 26.6

0.627

0.351 31

50

72

66

6

In [22]: y

148

85

```
Out[22]: 0
         2
                1
         4
                1
         763
               0
         764
               0
         765
                0
         766
                1
         767
                0
         Name: Outcome, Length: 768, dtype: int64
In [23]: X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.2)
In [24]: X_train.shape
Out[24]: (614, 8)
In [25]: X_test.shape
Out[25]: (154, 8)
```

Train the model

```
In [26]: clf = svm.SVC(kernel='linear')
In [27]: clf.fit(X_train,y_train)
Out[27]: SVC(kernel='linear')
In [28]: X_train_prediction = clf.predict(X_train)
accuracy_score(X_train_prediction,y_train)
Out[28]: 0.7850162866449512
```

```
Accuracy on test data
 In [29]: X test prediction = clf.predict(X test)
                             accuracy score(X test prediction,y test)
Out[29]: 0.77272727272727
 In [30]: input sample = (5,166,72,19,175,22.7,0.6,51)
 In [31]: input np array = np.asarray(input sample)
 In [32]: input np array_reshaped = input np array.reshape(1,-1)
 In [33]: std data = scaler.transform(input np array reshaped)
                         \verb|c:\users| admin \appdata \local \programs \python 39 \lib\site-packages \\ | sklearn \base.py: 450: User \warming: X does \\ | does \begin{tabular}{l} A does \end{tabular} | doe
                         not have valid feature names, but StandardScaler was fitted with feature names
                          warnings.warn(
 In [34]: std data
Out[34]: array([[ 0.3429808 , 1.41167241, 0.14964075, -0.09637905, 0.82661621,
                                                       -1.179407 , 0.38694877, 1.51108316]])
 In [38]: prediction = clf.predict(std data)
 In [39]: prediction
Out[39]: array([1], dtype=int64)
 In [41]: if (prediction[0]==0):
                                        print("Person is not diabetic")
                             else:
                                        print("Person is diabetic")
                         Person is diabetic
```