

Importing the dependencies

```
In [1]: 1 import numpy as np
2 import pandas as pd
3 from sklearn.preprocessing import StandardScaler
4 from sklearn.model_selection import train_test_split
5 from sklearn import svm
6 from sklearn.metrics import accuracy_score
7 %matplotlib inline
8 import warnings
9 warnings.filterwarnings('ignore')
10 from sklearn.metrics import adjusted_rand_score
```

Data Collection and Analysis Diabetes dataset

```
In [2]: 1 #Load the Dataset to a pandas dataframe
2 diabetes_data = pd.read_csv(r"C:\Users\user\Desktop\Machine learning p
```

```
In [3]: 1 # printing the first 10 rowa of the data
2 diabetes_data.head(10)
```

Out[3]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFun
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	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	
5	5	116	74	0	0	25.6	
6	3	78	50	32	88	31.0	
7	10	115	0	0	0	35.3	
8	2	197	70	45	543	30.5	
9	8	125	96	0	0	0.0	

```
In [4]: 1 # Lets look at the number of rows and colums in the dataset
2 diabetes_data.shape
```

Out[4]: (768, 9)

```
In [5]: 1 # statistical measures of the data
        2 diabetes_data.describe()
```

Out[5]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Diabetes
count	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	
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	

```
In [6]: 1 # checking counts of output
        2 diabetes_data['Outcome'].value_counts()
```

Out[6]: 0 500
1 268
Name: Outcome, dtype: int64

0 --> Non-diabetic

1 --> diabetic

```
In [7]: 1 # mean value for both ccaes
        2 diabetes_data.groupby('Outcome').mean()
```

Out[7]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Diabetes
Outcome							
0	3.298000	109.980000	68.184000	19.664000	68.792000	30.304200	
1	4.865672	141.257463	70.824627	22.164179	100.335821	35.142537	

```
In [8]: 1 # separating data and labels
        2 x = diabetes_data.drop(columns= 'Outcome', axis=1)
        3 y = diabetes_data['Outcome']
```

In [9]:

```
1 print(x)
2 print(y)
```

	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	
3	1	89	66	23	94	28.1	
4	0	137	40	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	1	126	60	0	0	30.1	
767	1	93	70	31	0	30.4	

	DiabetesPedigreeFunction	Age
0	0.627	50
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]

0	1
1	0
2	1
3	0
4	1
..	
763	0
764	0
765	0
766	1
767	0

Name: Outcome, Length: 768, dtype: int64

Data Standardization

In [10]:

```
1 # Performing standardization
2 scaler = StandardScaler()
```

In [11]:

```
1 scaler.fit(x)
```

Out[11]: StandardScaler()

In [12]:

```
1 Sd_data = scaler.transform(x)
```

```
In [13]: 1 print(Sd_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]
 ...
 [ 0.3429808  0.00330087  0.14964075 ... -0.73518964 -0.68519336
   -0.27575966]
 [-0.84488505  0.1597866  -0.47073225 ... -0.24020459 -0.37110101
    1.17073215]
 [-0.84488505 -0.8730192   0.04624525 ... -0.20212881 -0.47378505
   -0.87137393]]
```

```
In [14]: 1 x = Sd_data
          2 y = diabetes_data['Outcome']
```

```
In [15]: 1 print(x)
          2 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  0.00330087  0.14964075 ... -0.73518964 -0.68519336
   -0.27575966]
 [-0.84488505  0.1597866  -0.47073225 ... -0.24020459 -0.37110101
    1.17073215]
 [-0.84488505 -0.8730192   0.04624525 ... -0.20212881 -0.47378505
   -0.87137393]]
0      1
1      0
2      1
3      0
4      1
..
763    0
764    0
765    0
766    1
767    0
Name: Outcome, Length: 768, dtype: int64
```

Train Test Split

Splitting dataset into train and test set

```
In [16]: 1 x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0
```

```
In [17]: 1 # Let's see the shapes
2 print(x.shape,x_train.shape, x_test.shape)

(768, 8) (614, 8) (154, 8)
```

Training a model

```
In [18]: 1 classifier = svm.SVC(kernel='linear')
```

```
In [19]: 1 #training the suport vector Machine Classifier
2 classifier.fit(x_train, y_train)
```

```
Out[19]: SVC(kernel='linear')
```

Model Evaluation

Lets check classification report,confusion marix and accuracy score of above model

```
In [20]: 1 #importing tools
2 from sklearn.metrics import classification_report, accuracy_score, co
```

```
In [21]: 1 # prediction classes
2 y_pred = classifier.predict(x_test)
```

```
In [22]: 1 print("CLASSIFICATION REPORT: ")
2 print(classification_report(y_test, y_pred))
```

```
CLASSIFICATION REPORT:
              precision    recall  f1-score   support

      0              0.78        0.91        0.84         100
      1              0.76        0.52        0.62          54

   accuracy                   0.77         154
  macro avg              0.77        0.71        0.73         154
 weighted avg              0.77        0.77        0.76         154
```

```
In [23]: 1 #Accuracy score of training data
2 x_train_prediction = classifier.predict(x_train)
3 training_data_accuracy = accuracy_score(x_train_prediction,y_train )*
```

```
In [24]: 1 print("ACCURACY OF TRAINING DATA: ",training_data_accuracy.round(2))
```

```
ACCURACY OF TRAINING DATA:  78.66
```

```
In [25]: 1 #Accuracy score of test data
2 x_test_prediction = classifier.predict(x_test)
3 test_data_accuracy = accuracy_score(x_test_prediction,y_test )*100
```

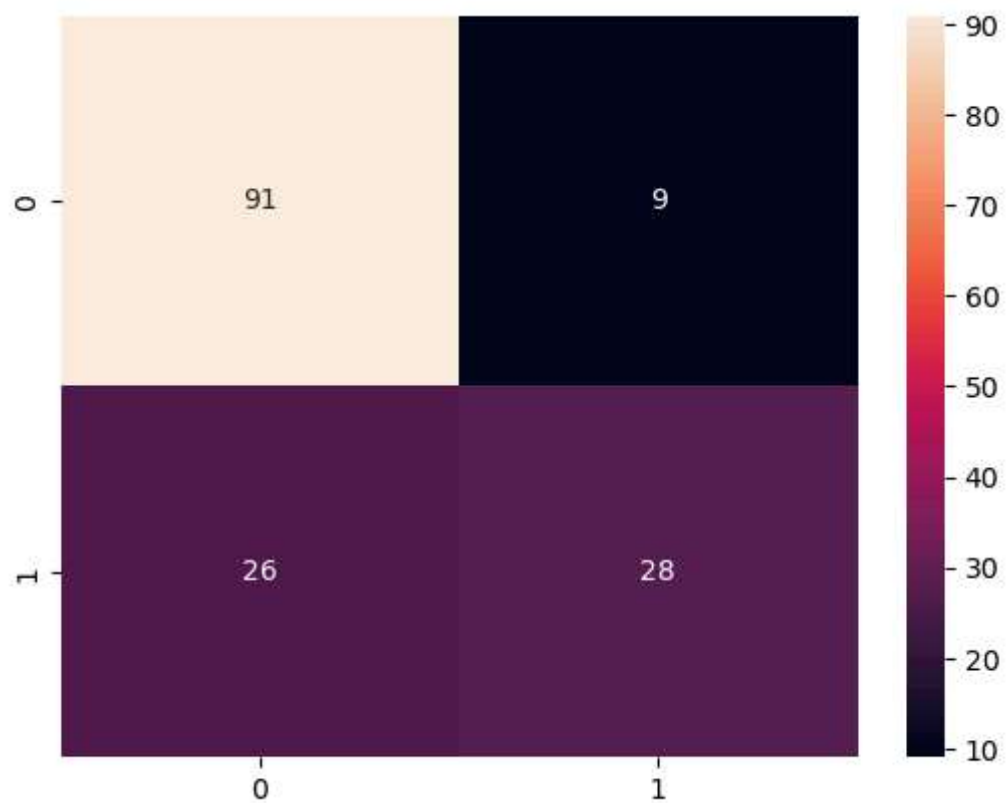
```
In [26]: 1 print("ACCURACY OF Test DATA: ",test_data_accuracy.round(2))
```

ACCURACY OF Test DATA: 77.27

Confusion Matrix

```
In [27]: 1 # Confusion Matrix
2 cm = confusion_matrix(y_test, y_pred)
3 #Visualizing the Confusion Matrix
4 import seaborn as sns
5 sns.heatmap(cm,annot=True)
```

Out[27]: <AxesSubplot:>



Making a predictive System

```
In [28]: ▶ 1 # input_data = (4,110,92,0,0,37.6,0.191,30) not diabetic
2 input_data = (5,166,72,19,175,25.8,0.587,51) # diabetic
3
4 # input data into numpy array
5 input_data_as_numpy_array = np.asarray(input_data)
6
7 # reshape the array as we are predicting for one instance
8 input_data_resaped = input_data_as_numpy_array.reshape(1,-1)
9
10 # standardize the inputdata
11 std_data = scaler.transform(input_data_resaped)
12 # print(std_data)
13
14 prediction = classifier.predict(std_data) # imp line
15 # print(prediction)
16
17 if (prediction[0]==0):
18     print('The person is not diabetic')
19 else:
20     print('The person is diabetic')
```

The person is diabetic

```
In [ ]: ▶ 1
```