```
In [1]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]:

```
## Data Collection and analysis
diabetes_dataset = pd.read_csv('diabetes.csv')
```

In [3]:

diabetes_dataset.head()

Out[3]:

	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

In [4]:

diabetes_dataset.isnull()

Out[4]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Age	Outcome
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
							•••		
763	False	False	False	False	False	False	False	False	False
764	False	False	False	False	False	False	False	False	False
765	False	False	False	False	False	False	False	False	False
766	False	False	False	False	False	False	False	False	False
767	False	False	False	False	False	False	False	False	False

768 rows × 9 columns

In [5]:

```
diabetes_dataset.shape
```

Out[5]:

(768, 9)

In [6]:

diahetes dataset describe()

diaseces_databet.debetise()

Out[6]:

	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							1	

In [7]:

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

Out[7]:

0 500 1 268

Name: Outcome, dtype: int64

0 --> Non-Diabetic

1 --> Diabetic

In [8]:

```
diabetes_dataset.groupby('Outcome').mean()
```

Out[8]:

		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Ag
Outco	ome								
	0	3.298000	109.980000	68.184000	19.664000	68.792000	30.304200	0.429734	31.19000
	1	4.865672	141.257463	70.824627	22.164179	100.335821	35.142537	0.550500	37.06716
4									•

In [9]:

```
# separating the data and labels
X = diabetes_dataset.drop('Outcome', axis=1)
y = diabetes_dataset['Outcome']
```

In [10]:

```
## Now Scale our Data
from sklearn.preprocessing import StandardScaler
```

In [11]:

```
scale = StandardScaler()
```

In [12]:

```
stand_scaler = scale.fit_transform(X)
```

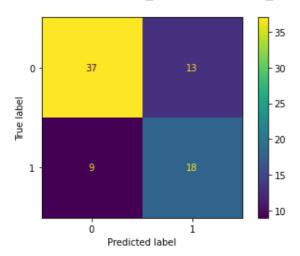
```
In [13]:
stand scaler
Out[13]:
array([[ 0.63994726, 0.84832379, 0.14964075, ..., 0.20401277,
         0.46849198, 1.4259954],
       [-0.84488505, -1.12339636, -0.16054575, ..., -0.68442195,
        -0.36506078, -0.19067191],
       [1.23388019, 1.94372388, -0.26394125, ..., -1.10325546,
         0.60439732, -0.10558415],
       [0.3429808, 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]:
X = stand scaler
y = diabetes dataset['Outcome']
In [15]:
print(X,y)
[[0.63994726 \quad 0.84832379 \quad 0.14964075 \dots \quad 0.20401277 \quad 0.46849198]
   1.4259954 ]
 [-0.84488505 -1.12339636 -0.16054575 ... -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 ... \quad -0.24020459 \quad -0.37110101
  1.17073215]
 [-0.84488505 -0.8730192
                          0.04624525 ... -0.20212881 -0.47378505
  -0.87137393]] 0
1
       0
2
       1
3
       0
4
       1
      . .
763
      0
764
      0
765
       0
766
       1
767
Name: Outcome, Length: 768, dtype: int64
In [16]:
## Now split into train test data
from sklearn.model selection import train test split
In [17]:
X train , X test, y train, y test = train test split(X,y,test size = 0.1, random state =
42)
In [18]:
len(X train)
Out[18]:
```

691

```
X train
Out[19]:
array([[-0.25095213, -0.77912776, -0.67752325, ..., -0.92556851,
         1.4258696 , 0.49003012],
       [-1.14185152, -0.12188771, -0.16054575, ..., -0.15136112,
         0.06379607, -0.95646168],
       [0.04601433, -0.34096773, 1.18359575, ..., 0.71168975,
        -0.84827977, -0.27575966],
       [1.82781311, -0.62264204, 0.87340925, ..., 1.72704372,
        2.00573238, 0.404942371,
       [-1.14185152, 0.62924378, -3.57259724, ..., 1.32090213,
       -0.8059981 , -0.36084741],
       [-1.14185152, 0.12848945, 1.39038675, ..., -1.20479085,
        -0.63385134, -1.04154944]])
Now train our data by SVM Classifier
In [20]:
from sklearn.svm import SVC
In [21]:
classifier = SVC(kernel= 'linear', C = 1000)
In [22]:
classifier.fit(X_train,y_train)
Out[22]:
SVC(C=1000, kernel='linear')
Now Model Evalution
In [23]:
pred = classifier.predict(X test)
In [24]:
from sklearn.metrics import classification report, plot confusion matrix
In [25]:
print(classification report(y test,pred))
              precision
                         recall f1-score
                                              support
                             0.74
                                       0.77
                                                   50
           0
                   0.80
                   0.58
                             0.67
                                       0.62
                                                   27
                                                   77
                                       0.71
   accuracy
                   0.69
                             0.70
                                       0.70
                                                   77
   macro avq
                            0.71
                                      0.72
                                                   77
weighted avg
                  0.73
In [26]:
plot confusion matrix(classifier, X test, y test)
```

In [19]:

Out[26]:



Making a Predictive System

```
In [27]:
```

In [28]:

```
if (prediction[0] == 0):
    print('The Person is not Diabitic')
else:
    print('The Person is Diabitic')
```

The Person is not Diabitic

In [31]:

```
input_data = (5,166,72,19,175,25.8,0.587,51)

## Change input into numpy array
input_data_as_numpy_array = np.asarray(input_data)

## Reshape data
input_data_reshape = input_data_as_numpy_array.reshape(1,-1)

## Standardize the input data

std_data = scale.transform(input_data_reshape)
```