

```
In [7]: import pandas as pd
import numpy as np
```

```
In [8]: df=pd.read_csv('diabetes.csv')
df.head()
```

```
Out[8]:
```

	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 [9]: df.shape
```

```
Out[9]: (768, 9)
```

```
In [10]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Pregnancies           768 non-null    int64
1   Glucose               768 non-null    int64
2   BloodPressure         768 non-null    int64
3   SkinThickness         768 non-null    int64
4   Insulin               768 non-null    int64
5   BMI                   768 non-null    float64
6   DiabetesPedigreeFunction 768 non-null    float64
7   Age                   768 non-null    int64
8   Outcome               768 non-null    int64
```

dtypes: float64(2), int64(7)  
memory usage: 54.1 KB

In [11]: `df.describe()`

Out[11]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
<b>count</b>	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000
<b>mean</b>	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578	0.471876	33.240885	0.348958
<b>std</b>	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.331329	11.760232	0.476951
<b>min</b>	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.078000	21.000000	0.000000
<b>25%</b>	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	0.243750	24.000000	0.000000
<b>50%</b>	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	0.372500	29.000000	0.000000
<b>75%</b>	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	0.626250	41.000000	1.000000
<b>max</b>	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.420000	81.000000	1.000000

In [12]: `#Checking for null values`  
`df.isna().sum()`

Out[12]:

Pregnancies	0
Glucose	0
BloodPressure	0
SkinThickness	0
Insulin	0
BMI	0
DiabetesPedigreeFunction	0
Age	0
Outcome	0

dtype: int64

## Extract input and target features

In [13]: `x=df.iloc[:, :-1]`  
`x.head()`

```
Out[13]:
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age
<b>0</b>	6	148	72	35	0	33.6	0.627	50
<b>1</b>	1	85	66	29	0	26.6	0.351	31
<b>2</b>	8	183	64	0	0	23.3	0.672	32
<b>3</b>	1	89	66	23	94	28.1	0.167	21
<b>4</b>	0	137	40	35	168	43.1	2.288	33

```
In [14]: y=df.iloc[:,-1]
y.head()
```

```
Out[14]: 0    1
1    0
2    1
3    0
4    1
Name: Outcome, dtype: int64
```

## Train-Test splitting

```
In [40]: from sklearn.model_selection import train_test_split
```

```
In [16]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

```
In [17]: x_train.head()
```

```
Out[17]:
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age
<b>765</b>	5	121	72	23	112	26.2	0.245	30
<b>350</b>	4	92	80	0	0	42.2	0.237	29
<b>208</b>	1	96	64	27	87	33.2	0.289	21
<b>638</b>	7	97	76	32	91	40.9	0.871	32

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age
621	2	92	76	20	0	24.2	1.698	28

In [18]: `y_train.head()`

Out[18]:

```
765    0
350    0
208    0
638    1
621    0
Name: Outcome, dtype: int64
```

## Normalization of input values

In [19]: `from sklearn.preprocessing import StandardScaler`

In [20]:

```
scaler=StandardScaler()
scaler.fit(x_train)
x_train=scaler.transform(x_train)
x_test=scaler.transform(x_test)
```

In [21]: `x_train`

Out[21]:

```
array([[ 0.32087331, -0.02224577,  0.13503557, ..., -0.72173529,
        -0.65749046, -0.30195314],
       [ 0.02632807, -0.93151964,  0.55540717, ...,  1.35049517,
        -0.6808989 , -0.38609887],
       [-0.85730766, -0.80610255, -0.28533603, ...,  0.18486553,
        -0.52874406, -1.05926476],
       ...,
       [-0.85730766, -0.68068547,  0.13503557, ..., -0.83829826,
         0.55097009, -0.47024461],
       [-1.1518529 ,  1.26327936,  0.34522137, ...,  2.30890175,
        -0.30928997, -0.63853608],
       [-1.1518529 , -0.33578848, -0.23278958, ..., -0.92895834,
         0.5568222 , -0.2178074 ]])
```

In [22]: `x_test`

Out[22]: `array([[ -0.56276241, 1.04379946, -0.91589343, ..., 0.89719475,  
 -0.67212074, -0.72268182],  
 [ -0.56276241, 0.25994267, 1.39615038, ..., -1.18798715,  
 -0.59019121, -1.05926476],  
 [ -0.26821717, 1.13786227, -0.28533603, ..., -0.07416328,  
 -0.51118773, -0.80682755],  
 ...,  
 [ 0.02632807, 0.69890247, 0.66050007, ..., 0.87129187,  
 0.24666041, 0.28706702],  
 [ 0.90996379, -0.24172567, -0.28533603, ..., -0.56631801,  
 0.76749813, 0.03462981],  
 [ 0.61541855, 1.92171905, 1.29105747, ..., 1.169175 ,  
 2.90059194, 0.96023291]])`

## Model Creation-Naive-bayes

In [23]: `from sklearn.naive_bayes import GaussianNB`

In [24]: `model=GaussianNB()  
model.fit(x_train,y_train)`

Out[24]: `GaussianNB()`

In [25]: `y_pred=model.predict(x_test)  
y_pred`

Out[25]: `array([1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0,  
 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,  
 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0,  
 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,  
 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0,  
 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0])`

```
1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0,  
0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1], dtype=int64)
```

```
In [26]: from sklearn.metrics import classification_report, accuracy_score, confusion_matrix, ConfusionMatrixDisplay
```

```
In [27]: score=accuracy_score(y_test,y_pred)  
score
```

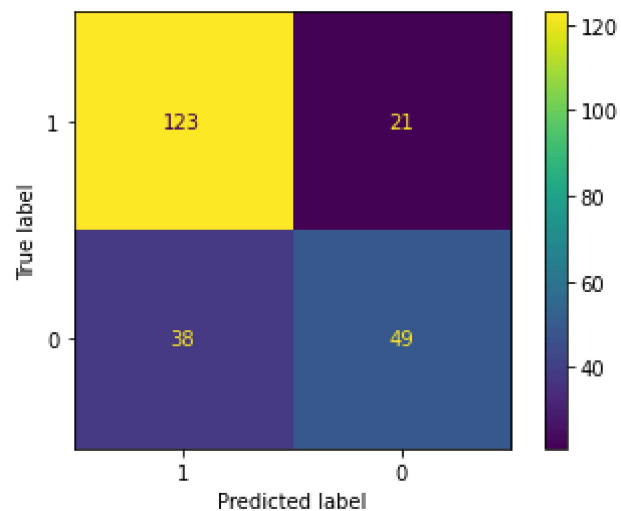
```
Out[27]: 0.7445887445887446
```

```
In [28]: cm=confusion_matrix(y_test,y_pred)  
cm
```

```
Out[28]: array([[123,  21],  
               [ 38,  49]], dtype=int64)
```

```
In [29]: cmd=ConfusionMatrixDisplay(cm,display_labels=['1','0'])  
cmd.plot()
```

```
Out[29]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x190d03f23d0>
```



```
In [30]: nb_cr=classification_report(y_test,y_pred)
print(nb_cr)
```

	precision	recall	f1-score	support
0	0.76	0.85	0.81	144
1	0.70	0.56	0.62	87
accuracy			0.74	231
macro avg	0.73	0.71	0.72	231
weighted avg	0.74	0.74	0.74	231

## Model Creation-KNN

```
In [31]: from sklearn.neighbors import KNeighborsClassifier
```

```
In [33]: knn_model=KNeighborsClassifier(n_neighbors=5)
knn_model.fit(x_train,y_train)
```

```
Out[33]: KNeighborsClassifier()
```

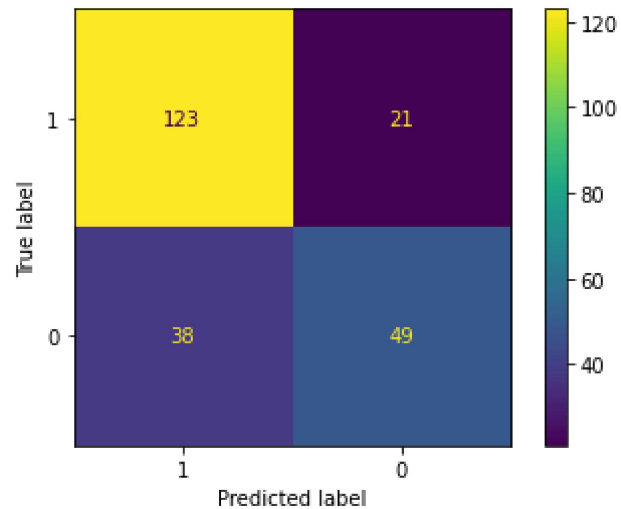
```
In [34]: knn_model.predict(x_test)
```

```
Out[34]: array([1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0,
0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0,
1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0,
0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0,
1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1,
1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1,
1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1], dtype=int64)
```

```
In [36]: score=accuracy_score(y_test,y_pred)
score
```

Out[36]: 0.7445887445887446

```
In [38]: knn_cmd=ConfusionMatrixDisplay(confusion_matrix(y_test,y_pred),display_labels=[1,0])
knn_cmd.plot();
```



```
In [39]: knn_cr=classification_report(y_test,y_pred)
print(knn_cr)
```

	precision	recall	f1-score	support
0	0.76	0.85	0.81	144
1	0.70	0.56	0.62	87
accuracy			0.74	231
macro avg	0.73	0.71	0.72	231
weighted avg	0.74	0.74	0.74	231

In [ ]: