

Assignment_5

August 21, 2024

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

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

```
[2]:
```

	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	

	Pedigree	Age	Outcome
0	0.627	50	1
1	0.351	31	0
2	0.672	32	1
3	0.167	21	0
4	2.288	33	1

```
[3]: df.isnull().sum()
```

```
[3]: Pregnancies      0
Glucose              0
BloodPressure        0
SkinThickness        0
Insulin              0
BMI                  0
Pedigree             0
Age                  0
Outcome              0
dtype: int64
```

```
[4]: x = df.drop('Outcome', axis=1)
y = df['Outcome']
```

```
[5]: from sklearn.preprocessing import MinMaxScaler
      scaler=MinMaxScaler()
      x=scaler.fit_transform(x)
      x
```

```
[5]: array([[0.35294118, 0.74371859, 0.59016393, ..., 0.50074516, 0.23441503,
             0.48333333],
            [0.05882353, 0.42713568, 0.54098361, ..., 0.39642325, 0.11656704,
             0.16666667],
            [0.47058824, 0.91959799, 0.52459016, ..., 0.34724292, 0.25362938,
             0.18333333],
            ...,
            [0.29411765, 0.6080402 , 0.59016393, ..., 0.390462 , 0.07130658,
             0.15      ],
            [0.05882353, 0.63316583, 0.49180328, ..., 0.4485842 , 0.11571307,
             0.43333333],
            [0.05882353, 0.46733668, 0.57377049, ..., 0.45305514, 0.10119556,
             0.03333333]])
```

```
[6]: from sklearn.model_selection import train_test_split
      x_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.
      ↪2,random_state=42)
      from sklearn.neighbors import KNeighborsClassifier
      clf = KNeighborsClassifier(n_neighbors=3)
      clf
```

```
[6]: KNeighborsClassifier(n_neighbors=3)
```

```
[8]: KNN=clf.fit(x_train,y_train)
      prediction=KNN.predict(X_test)
      prediction
```

```
[8]: array([0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0,
            0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0,
            0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1,
            0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0,
            0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1,
            0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1,
            0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0])
```

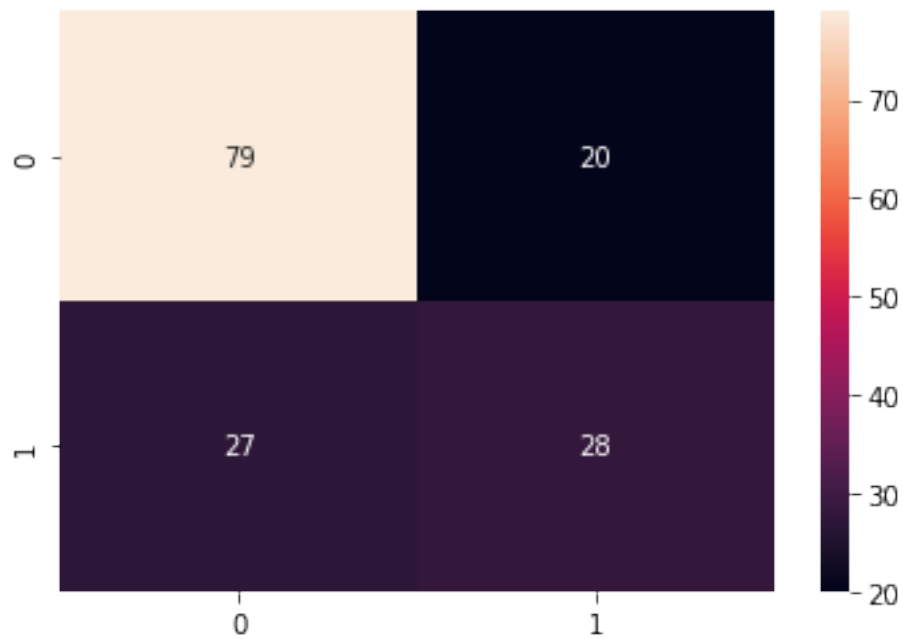
```
[9]: from sklearn import metrics
      print(metrics.classification_report(y_test, prediction))
      print(metrics.confusion_matrix(y_test, prediction))
      cm=metrics.confusion_matrix(y_test, prediction)
      sns.heatmap(cm,annot=True)
```

```
precision    recall  f1-score   support
```

0	0.75	0.80	0.77	99
1	0.58	0.51	0.54	55
accuracy			0.69	154
macro avg	0.66	0.65	0.66	154
weighted avg	0.69	0.69	0.69	154


```
[[79 20]
 [27 28]]
```

[9]: <AxesSubplot:>



```
[10]: print("accuracy:", metrics.accuracy_score(y_test, prediction))
```

accuracy: 0.6948051948051948