Data Source: https://www.kaggle.com/pankeshpatel/insurance

This project will predict whether a customer buy life insurance or not

Let's import Data

```
In [79]:
```

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
from sklearn.linear_model import LogisticRegression
import seaborn as sns
```

let's read data

```
In [80]:
```

```
data = pd.read_csv('insurance_data.csv')
```

In [81]:

```
data.head()
```

Out[81]:

age		bought_insurance	
0	22	0	
1	25	0	
2	47	1	
3	52	0	
4	46	1	

```
In [82]:
```

```
data.columns
```

Out[82]:

```
Index(['age', 'bought_insurance'], dtype='object')
```

In [83]:

```
data['bought_insurance'].value_counts()

Out[83]:
1    14
0    13
Name: bought_insurance, dtype: int64
```

Observation: 14 customers bought life insurance and 13 customers did not buy life insurance. 1 indicates "Yes" and 0 indicates "NO"

```
In [84]:
```

```
data['age'].value_counts()
Out[84]:
```

```
2
18
25
      2
55
28
      1
23
      1
54
45
      1
29
      1
50
      1
26
      1
46
      1
47
     1
49
      1
19
62
      1
52
      1
21
22
      1
56
      1
58
27
      1
60
    1
61
     1
40
     1
Name: age, dtype: int64
In [85]:
data.isnull().sum()
Out[85]:
                     0
age
bought insurance
dtype: int64
In [86]:
print("Is there any null value in dataset:",data.isnull().values.any())
Is there any null value in dataset: False
Data Visiualization
In [87]:
data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 27 entries, 0 to 26
Data columns (total 2 columns):
                       Non-Null Count Dtype
 # Column
---
0 age 27 non-null
1 bought_insurance 27 non-null
                        27 non-null
                                     int64
dtypes: int64(2)
memory usage: 560.0 bytes
In [88]:
data.describe()
Out[88]:
          age bought insurance
count 27.000000
                     27.000000
```

mean 39.666667

0.518519

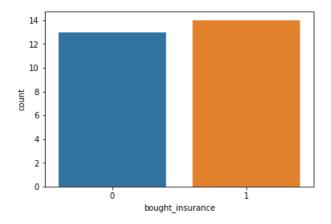
std	15.745 899	bought_inssnance
min	18.000000	0.000000
25%	25.000000	0.000000
50%	45.000000	1.000000
75%	54.500000	1.000000
max	62.000000	1.000000

In [89]:

```
sns.countplot(data['bought_insurance'])
```

Out[89]:

<matplotlib.axes._subplots.AxesSubplot at 0x207d1666ba8>



Taining the Model

```
In [90]:
```

```
X = data.drop(columns = 'bought_insurance')
Y = data['bought_insurance']
```

In [91]:

```
x_train,x_test,y_train,y_test = train_test_split(X,Y,test_size=0.33,random_state=30)
```

In [92]:

```
x_train.shape
```

Out[92]:

(18, 1)

In [93]:

```
x_test.shape
```

Out[93]:

(9, 1)

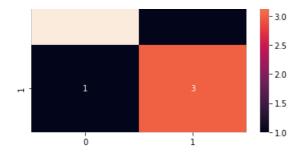
In [94]:

```
y_train.shape
```

Out[94]:

```
In [95]:
y_test.shape
Out[95]:
(9,)
In [96]:
model = LogisticRegression() #initialize the model with model variable
model.fit(x_train,y_train)
Out[96]:
LogisticRegression()
Predicting the Model
In [97]:
y_pred = model.predict(x_test)
print(y_pred)
print(y_test)
[1 0 0 0 0 0 1 1 1]
25
     1
21
11
      0
16
      1
0
18
      0
6
      0
4
Name: bought_insurance, dtype: int64
In [98]:
from sklearn import metrics
print("Accuracy=",metrics.accuracy_score(y_test,y_pred))
Accuracy= 0.7777777777778
Confusion Marix
In [99]:
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification report
cm = confusion matrix(y test, y pred)
In [100]:
sns.heatmap(cm,annot=True)
Out[100]:
<matplotlib.axes._subplots.AxesSubplot at 0x207d1695eb8>
                                        -4.0
```

(⊥∀,)



Observation: We found TPR and TNR is high and 2 mis-classified point

In [101]:

print(classification_report(y_test,y_pred))

support	f1-score	recall	precision	
5	0.80	0.80	0.80	0
4	0.75	0.75	0.75	1
9	0.78			accuracy
9	0.78	0.78	0.78	macro avg
9	0.78	0.78	0.78	weighted avg

Observation : We found model accuracy is 78%

In []: