

In [1]:

```
1 import pandas as pd
2 import numpy as np
```

In [2]:

```
1 df=pd.read_csv('heart.csv')
2 df.head()
```

Out[2]:

|   | Age | Sex | RestingBP | Cholesterol | FastingBS | RestingECG | MaxHR | ExerciseAngina | Oldpeak | ST_Slope |
|---|-----|-----|-----------|-------------|-----------|------------|-------|----------------|---------|----------|
| 0 | 40  | M   | 140       | 289         | 0         | Normal     | 172   | N              | 0.0     |          |
| 1 | 49  | F   | 160       | 180         | 0         | Normal     | 156   | N              | 1.0     |          |
| 2 | 37  | M   | 130       | 283         | 0         | ST         | 98    | N              | 0.0     |          |
| 3 | 48  | F   | 138       | 214         | 0         | Normal     | 108   | Y              | 1.5     |          |
| 4 | 54  | M   | 150       | 195         | 0         | Normal     | 122   | N              | 0.0     |          |

In [3]:

```
1 df.isna().sum()
```

Out[3]:

```
Age          0
Sex          0
RestingBP    0
Cholesterol  0
FastingBS    0
RestingECG   0
MaxHR        0
ExerciseAngina 0
Oldpeak      0
ST_Slope     0
HeartDisease 0
dtype: int64
```

In [4]:

```
1 x=df.drop('HeartDisease',axis=1)
2 y=df.HeartDisease
3
```

In [5]:

```
1 from sklearn.preprocessing import OneHotEncoder,StandardScaler
2 from sklearn.compose import ColumnTransformer
```

In [6]:

```

1 encoder=OneHotEncoder()
2 features=['Sex','RestingECG','ExerciseAngina','ST_Slope']
3 values=['Age','RestingBP','Cholesterol','MaxHR','Oldpeak']
4 tranformer=ColumnTransformer([
5     ('one_hot',encoder,features),
6     ('values',StandardScaler(),values)]
7     ,remainder='passthrough')
8 transformed_x=tranformer.fit_transform(x)
9 transformed_x

```

Out[6]:

```

array([[ 0.          ,  1.          ,  0.          , ...,  1.38292822,
        -0.83243239,  0.          ],
       [ 1.          ,  0.          ,  0.          , ...,  0.75415714,
         0.10566353,  0.          ],
       [ 0.          ,  1.          ,  0.          , ..., -1.52513802,
        -0.83243239,  0.          ],
       ...,
       [ 0.          ,  1.          ,  0.          , ..., -0.85706875,
         0.29328271,  0.          ],
       [ 1.          ,  0.          ,  1.          , ...,  1.4615246 ,
        -0.83243239,  0.          ],
       [ 0.          ,  1.          ,  0.          , ...,  1.42222641,
        -0.83243239,  0.          ]])

```

In [7]:

```

1 from sklearn.model_selection import train_test_split
2 x_train,x_test,y_train,y_test=train_test_split(transformed_x,y,test_size=0.2,random_state=4.

```

In [8]:

```
1 x_train.shape , x_test.shape
```

Out[8]:

```
((734, 16), (184, 16))
```

In [9]:

```
1 y_train.shape , y_test.shape
```

Out[9]:

```
((734,), (184,))
```

## Addign layers

In [10]:

```

1 from keras.models import Sequential
2 from keras.layers import Dense

```

In [11]:

```

1 model=Sequential()
2 model.add(Dense(128,activation='relu',input_shape=(16,)))
3 model.add(Dense(128,activation='relu'))
4 model.add(Dense(2,activation='softmax'))
5 model.compile(loss='binary_crossentropy',optimizer='Adam',metrics=['Accuracy'])
6 model.summary()

```

Model: "sequential"

| Layer (type)    | Output Shape | Param # |
|-----------------|--------------|---------|
| dense (Dense)   | (None, 128)  | 2176    |
| dense_1 (Dense) | (None, 128)  | 16512   |
| dense_2 (Dense) | (None, 2)    | 258     |

=====  
 Total params: 18946 (74.01 KB)  
 Trainable params: 18946 (74.01 KB)  
 Non-trainable params: 0 (0.00 Byte)

In [36]:

```

1 from keras.utils import to_categorical
2 y_train_encoded=to_categorical(y_train)
3 y_test_encoded=to_categorical(y_test)
4 model.fit(x_train,y_train_encoded,epochs=100,batch_size=10,shuffle=True,validation_split=0.1)

```

```

59/59 [=====] - 0s 3ms/step - loss: 6.0987e-06 - Accur
acy: 1.0000 - val_loss: 0.1680 - val_Accuracy: 0.9592
Epoch 83/100
59/59 [=====] - 0s 3ms/step - loss: 5.7809e-06 - Accur
acy: 1.0000 - val_loss: 0.1788 - val_Accuracy: 0.9592
Epoch 84/100
59/59 [=====] - 0s 3ms/step - loss: 5.7678e-06 - Accur
acy: 1.0000 - val_loss: 0.1748 - val_Accuracy: 0.9592
Epoch 85/100
59/59 [=====] - 0s 3ms/step - loss: 5.3662e-06 - Accur
acy: 1.0000 - val_loss: 0.1748 - val_Accuracy: 0.9592
Epoch 86/100
59/59 [=====] - 0s 3ms/step - loss: 5.2182e-06 - Accur
acy: 1.0000 - val_loss: 0.1791 - val_Accuracy: 0.9592
Epoch 87/100
59/59 [=====] - 0s 3ms/step - loss: 5.0639e-06 - Accur
acy: 1.0000 - val_loss: 0.1829 - val_Accuracy: 0.9592
Epoch 88/100
59/59 [=====] - 0s 3ms/step - loss: 5.2929e-06 - Accur
acy: 1.0000 - val_loss: 0.1790 - val_Accuracy: 0.9592

```

In [33]:

```
1 y_pred=model.predict(x_test)
2 y_preds=np.argmax(y_pred,axis=1)
3 y_preds
```

6/6 [=====] - 0s 0s/step

Out[33]:

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

In [34]:

```
1 from sklearn.metrics import accuracy_score, confusion_matrix
```

In [37]:

```
1 accuracy=accuracy_score(y_preds,y_test)
2 accuracy
3
```

Out[37]:

0.7934782608695652

In [38]:

```
1 confusion_matrix=confusion_matrix(y_preds,y_test)
2 confusion_matrix
```

Out[38]:

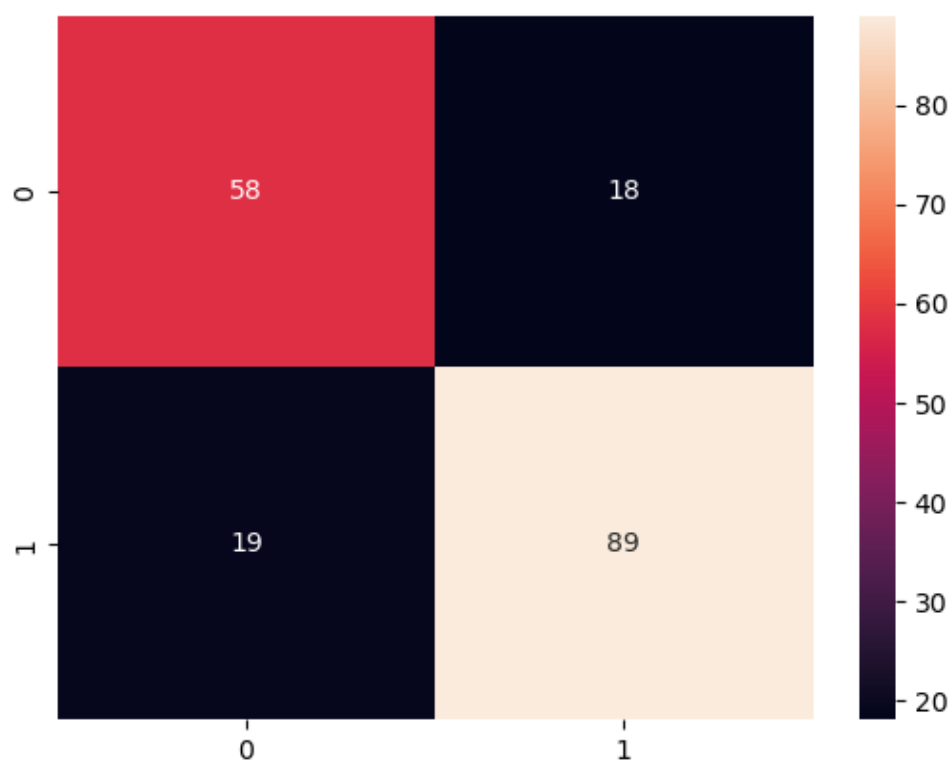
```
array([[60, 21],
       [17, 86]], dtype=int64)
```

In [31]:

```
1 import seaborn as sns
2 sns.heatmap(confusion_matrix,annot=True)
```

Out[31]:

&lt;Axes: &gt;



In [ ]:

```
1
```