## In [98]:

```
import pandas as pd
import numpy as np
```

## In [22]:

```
df=pd.read_csv('Churn_Modelling.csv')
df.drop(columns=['RowNumber','CustomerId','Surname'],inplace=True)
df.head()
```

#### Out[22]:

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	ls/
0	619	France	Female	42	2	0.00	1	1	
1	608	Spain	Female	41	1	83807.86	1	0	
2	502	France	Female	42	8	159660.80	3	1	
3	699	France	Female	39	1	0.00	2	0	
4	850	Spain	Female	43	2	125510.82	1	1	
4									•

## In [37]:

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

#### Out[37]:

CreditScore	0		
Geography	0		
Gender	0		
Age	0		
Tenure	0		
Balance	0		
NumOfProducts	0		
HasCrCard			
IsActiveMember			
EstimatedSalary			
Exited	0		
dtype: int64			

#### In [44]:

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

## In [45]:

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

```
In [67]:
```

```
encoder=OneHotEncoder()
    features=['Geography','Gender']
    values=['CreditScore','Age','Tenure','Balance','EstimatedSalary']
    tranformer=ColumnTransformer([
 5
        ('one_hot',encoder,features),
 6
        ('values',StandardScaler(),values)]
 7
        ,remainder='passthrough')
    transformed_x=tranformer.fit_transform(x)
 8
 9
    transformed_x
Out[67]:
array([[1., 0., 0., ..., 1., 1., 1.],
       [0., 0., 1., \ldots, 1., 0., 1.],
       [1., 0., 0., ..., 3., 1., 0.],
       [1., 0., 0., ..., 1., 0., 1.],
       [0., 1., 0., \ldots, 2., 1., 0.],
       [1., 0., 0., ..., 1., 1., 0.]])
In [57]:
 1 from sklearn.model_selection import train_test_split
   x_train,x_test,y_train,y_test=train_test_split(transformed_x,y,test_size=0.2,random_
In [63]:
 1 x_train.shape , x_test.shape
Out[63]:
((8000, 13), (2000, 13))
In [64]:
 1 y_train.shape , y_test.shape
Out[64]:
((8000,), (2000,))
```

# Addign layers

```
In [65]:
```

```
from keras.models import Sequential
from keras.layers import Dense
```

#### In [132]:

```
model=Sequential()
model.add(Dense(128,activation='relu',input_shape=(13,)))
model.add(Dense(128,activation='relu'))
model.add(Dense(2,activation='sigmoid'))
model.compile(loss='binary_crossentropy',optimizer='Adam',metrics=['Accuracy'])
model.summary()
```

#### Model: "sequential\_9"

Layer (type)	Output Shape	Param #
dense_25 (Dense)	(None, 128)	1792
dense_26 (Dense)	(None, 128)	16512
dense_27 (Dense)	(None, 2)	258

Total params: 18562 (72.51 KB)
Trainable params: 18562 (72.51 KB)
Non-trainable params: 0 (0.00 Byte)

\_\_\_\_\_

#### In [133]:

```
1 from keras.utils import to_categorical
   y_train_encoded=to_categorical(y_train)
   y_test_encoded=to_categorical(y_test)
 4 model.fit(x_train,y_train_encoded,epochs=100,batch_size=32,validation_split=0.2)
FDOCU A2/IAA
200/200 [============= ] - 1s 4ms/step - loss: 0.1116 -
Accuracy: 0.9573 - val_loss: 0.6128 - val_Accuracy: 0.8138
Epoch 96/100
200/200 [============ ] - 1s 4ms/step - loss: 0.1077 -
Accuracy: 0.9586 - val_loss: 0.5967 - val_Accuracy: 0.8306
Epoch 97/100
200/200 [============= ] - 1s 5ms/step - loss: 0.1106 -
Accuracy: 0.9584 - val_loss: 0.6220 - val_Accuracy: 0.8150
Epoch 98/100
200/200 [============= ] - 1s 5ms/step - loss: 0.1106 -
Accuracy: 0.9572 - val loss: 0.6179 - val Accuracy: 0.8288
Epoch 99/100
200/200 [============= ] - 1s 4ms/step - loss: 0.1074 -
Accuracy: 0.9569 - val_loss: 0.6208 - val_Accuracy: 0.8331
Epoch 100/100
200/200 [============ ] - 1s 3ms/step - loss: 0.1036 -
Accuracy: 0.9595 - val loss: 0.6029 - val Accuracy: 0.8319
Out[133]:
```

```
In [134]:
```

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

```
63/63 [=======] - 0s 2ms/step
```

# Out[134]:

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

## In [135]:

from sklearn.metrics import accuracy\_score,confusion\_matrix

## In [136]:

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

#### Out[136]:

0.8225

#### In [137]:

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

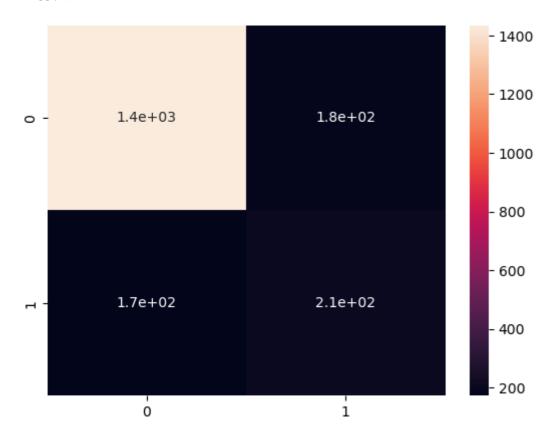
#### Out[137]:

## In [138]:

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

## Out[138]:

#### <Axes: >



## In [ ]: