

```
In [1]: import numpy as np
import pandas as pd
import tensorflow as tf
```

```
In [2]: tf.__version__
```

```
Out[2]: '2.16.2'
```

```
In [3]: dataset=pd.read_csv(r"D:\Data Science with AI\Data Science With AI\6th, 7th-octo
```

```
In [4]: dataset
```

```
Out[4]:
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure
0	1	15634602	Hargrave	619	France	Female	42	
1	2	15647311	Hill	608	Spain	Female	41	
2	3	15619304	Onio	502	France	Female	42	
3	4	15701354	Boni	699	France	Female	39	
4	5	15737888	Mitchell	850	Spain	Female	43	
...
9995	9996	15606229	Obijiaku	771	France	Male	39	
9996	9997	15569892	Johnstone	516	France	Male	35	
9997	9998	15584532	Liu	709	France	Female	36	
9998	9999	15682355	Sabbatini	772	Germany	Male	42	
9999	10000	15628319	Walker	792	France	Female	28	

10000 rows × 14 columns

```
In [5]: x=dataset.iloc[:,3:-1].values
y=dataset.iloc[:, -1].values
```

```
In [6]: x.head()
```

```
-----
AttributeError                                Traceback (most recent call last)
Cell In[6], line 1
----> 1 x.head()

AttributeError: 'numpy.ndarray' object has no attribute 'head'
```

```
In [7]: dataset.head()
```

Out[7]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure
0	1	15634602	Hargrave	619	France	Female	42	2
1	2	15647311	Hill	608	Spain	Female	41	1
2	3	15619304	Onio	502	France	Female	42	8
3	4	15701354	Boni	699	France	Female	39	1
4	5	15737888	Mitchell	850	Spain	Female	43	2

In [8]: `x.head()`

```
AttributeError
Cell In[8], line 1
----> 1 x.head()
```

Traceback (most recent call last)

```
AttributeError: 'numpy.ndarray' object has no attribute 'head'
```

In [9]: `y`Out[9]: `array([1, 0, 1, ..., 1, 1, 0], dtype=int64)`In [10]: `x`

```
Out[10]: array([[619, 'France', 'Female', ..., 1, 1, 101348.88],
   [608, 'Spain', 'Female', ..., 0, 1, 112542.58],
   [502, 'France', 'Female', ..., 1, 0, 113931.57],
   ...,
   [709, 'France', 'Female', ..., 0, 1, 42085.58],
   [772, 'Germany', 'Male', ..., 1, 0, 92888.52],
   [792, 'France', 'Female', ..., 1, 0, 38190.78]], dtype=object)
```

```
In [11]: from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
x[:,2]=le.fit_transform(x[:,2])
```

In [12]: `print(x)`

```
[[619 'France' 0 ... 1 1 101348.88]
 [608 'Spain' 0 ... 0 1 112542.58]
 [502 'France' 0 ... 1 0 113931.57]
 ...
 [709 'France' 0 ... 0 1 42085.58]
 [772 'Germany' 1 ... 1 0 92888.52]
 [792 'France' 0 ... 1 0 38190.78]]
```

```
In [13]: from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
ct=ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [1])], remainder='p')
x=np.array(ct.fit_transform(x))
```

In [14]: `print(x)`

```
[[1.0 0.0 0.0 ... 1 1 101348.88]
 [0.0 0.0 1.0 ... 0 1 112542.58]
 [1.0 0.0 0.0 ... 1 0 113931.57]
 ...
 [1.0 0.0 0.0 ... 0 1 42085.58]
 [0.0 1.0 0.0 ... 1 0 92888.52]
 [1.0 0.0 0.0 ... 1 0 38190.78]]
```

```
In [15]: from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
x=sc.fit_transform(x)
```

```
In [16]: print(x)

[[ 0.99720391 -0.57873591 -0.57380915 ...  0.64609167  0.97024255
  0.02188649]
 [-1.00280393 -0.57873591  1.74273971 ... -1.54776799  0.97024255
  0.21653375]
 [ 0.99720391 -0.57873591 -0.57380915 ...  0.64609167 -1.03067011
  0.2406869 ]
 ...
 [ 0.99720391 -0.57873591 -0.57380915 ... -1.54776799  0.97024255
 -1.00864308]
 [-1.00280393  1.72790383 -0.57380915 ...  0.64609167 -1.03067011
 -0.12523071]
 [ 0.99720391 -0.57873591 -0.57380915 ...  0.64609167 -1.03067011
 -1.07636976]]
```

```
In [17]: 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=0)
```

```
In [18]: ann=tf.keras.models.Sequential()
```

```
In [20]: ann.add(tf.keras.layers.Dense(units=6,activation='relu'))
```

```
In [21]: ann.add(tf.keras.layers.Dense(units=6,activation='relu'))
```

```
In [22]: ann.add(tf.keras.layers.Dense(units=1,activation='sigmoid'))
```

```
In [26]: ann.compile(optimizer='adagrad',loss='binary_crossentropy',metrics=[ 'accuracy'])
```

```
In [27]: ann.fit(x_train,y_train,batch_size=32,epochs=100)
```

Epoch 1/100
250/250 1s 2ms/step - accuracy: 0.8647 - loss: 0.3284
Epoch 2/100
250/250 0s 1ms/step - accuracy: 0.8588 - loss: 0.3334
Epoch 3/100
250/250 1s 1ms/step - accuracy: 0.8625 - loss: 0.3359
Epoch 4/100
250/250 0s 1ms/step - accuracy: 0.8613 - loss: 0.3355
Epoch 5/100
250/250 1s 1ms/step - accuracy: 0.8656 - loss: 0.3328
Epoch 6/100
250/250 0s 1ms/step - accuracy: 0.8679 - loss: 0.3295
Epoch 7/100
250/250 1s 1ms/step - accuracy: 0.8614 - loss: 0.3390
Epoch 8/100
250/250 1s 2ms/step - accuracy: 0.8643 - loss: 0.3384
Epoch 9/100
250/250 0s 1ms/step - accuracy: 0.8634 - loss: 0.3347
Epoch 10/100
250/250 1s 2ms/step - accuracy: 0.8585 - loss: 0.3451
Epoch 11/100
250/250 0s 2ms/step - accuracy: 0.8613 - loss: 0.3409
Epoch 12/100
250/250 1s 1ms/step - accuracy: 0.8610 - loss: 0.3353
Epoch 13/100
250/250 0s 2ms/step - accuracy: 0.8645 - loss: 0.3171
Epoch 14/100
250/250 0s 2ms/step - accuracy: 0.8696 - loss: 0.3275
Epoch 15/100
250/250 1s 2ms/step - accuracy: 0.8618 - loss: 0.3348
Epoch 16/100
250/250 0s 1ms/step - accuracy: 0.8662 - loss: 0.3237
Epoch 17/100
250/250 1s 2ms/step - accuracy: 0.8610 - loss: 0.3298
Epoch 18/100
250/250 0s 2ms/step - accuracy: 0.8618 - loss: 0.3330
Epoch 19/100
250/250 0s 2ms/step - accuracy: 0.8639 - loss: 0.3304
Epoch 20/100
250/250 0s 1ms/step - accuracy: 0.8611 - loss: 0.3345
Epoch 21/100
250/250 0s 1ms/step - accuracy: 0.8632 - loss: 0.3373
Epoch 22/100
250/250 1s 2ms/step - accuracy: 0.8593 - loss: 0.3349
Epoch 23/100
250/250 0s 1ms/step - accuracy: 0.8678 - loss: 0.3265
Epoch 24/100
250/250 0s 2ms/step - accuracy: 0.8684 - loss: 0.3278
Epoch 25/100
250/250 0s 1ms/step - accuracy: 0.8617 - loss: 0.3337
Epoch 26/100
250/250 1s 2ms/step - accuracy: 0.8637 - loss: 0.3308
Epoch 27/100
250/250 1s 2ms/step - accuracy: 0.8653 - loss: 0.3307
Epoch 28/100
250/250 0s 2ms/step - accuracy: 0.8615 - loss: 0.3379
Epoch 29/100
250/250 0s 2ms/step - accuracy: 0.8643 - loss: 0.3362
Epoch 30/100
250/250 0s 2ms/step - accuracy: 0.8597 - loss: 0.3358

Epoch 31/100
250/250 0s 1ms/step - accuracy: 0.8626 - loss: 0.3390
Epoch 32/100
250/250 1s 2ms/step - accuracy: 0.8557 - loss: 0.3435
Epoch 33/100
250/250 0s 1ms/step - accuracy: 0.8611 - loss: 0.3363
Epoch 34/100
250/250 0s 2ms/step - accuracy: 0.8653 - loss: 0.3375
Epoch 35/100
250/250 0s 2ms/step - accuracy: 0.8638 - loss: 0.3323
Epoch 36/100
250/250 1s 2ms/step - accuracy: 0.8617 - loss: 0.3400
Epoch 37/100
250/250 0s 1ms/step - accuracy: 0.8615 - loss: 0.3330
Epoch 38/100
250/250 0s 2ms/step - accuracy: 0.8656 - loss: 0.3308
Epoch 39/100
250/250 1s 2ms/step - accuracy: 0.8650 - loss: 0.3273
Epoch 40/100
250/250 0s 2ms/step - accuracy: 0.8642 - loss: 0.3343
Epoch 41/100
250/250 0s 1ms/step - accuracy: 0.8587 - loss: 0.3386
Epoch 42/100
250/250 0s 1ms/step - accuracy: 0.8643 - loss: 0.3305
Epoch 43/100
250/250 0s 2ms/step - accuracy: 0.8645 - loss: 0.3322
Epoch 44/100
250/250 0s 1ms/step - accuracy: 0.8600 - loss: 0.3342
Epoch 45/100
250/250 0s 2ms/step - accuracy: 0.8670 - loss: 0.3260
Epoch 46/100
250/250 1s 1ms/step - accuracy: 0.8571 - loss: 0.3449
Epoch 47/100
250/250 1s 2ms/step - accuracy: 0.8631 - loss: 0.3368
Epoch 48/100
250/250 0s 1ms/step - accuracy: 0.8622 - loss: 0.3358
Epoch 49/100
250/250 0s 2ms/step - accuracy: 0.8666 - loss: 0.3247
Epoch 50/100
250/250 1s 1ms/step - accuracy: 0.8623 - loss: 0.3371
Epoch 51/100
250/250 1s 2ms/step - accuracy: 0.8660 - loss: 0.3240
Epoch 52/100
250/250 1s 2ms/step - accuracy: 0.8616 - loss: 0.3354
Epoch 53/100
250/250 1s 1ms/step - accuracy: 0.8608 - loss: 0.3368
Epoch 54/100
250/250 1s 2ms/step - accuracy: 0.8601 - loss: 0.3349
Epoch 55/100
250/250 0s 1ms/step - accuracy: 0.8631 - loss: 0.3335
Epoch 56/100
250/250 1s 2ms/step - accuracy: 0.8582 - loss: 0.3413
Epoch 57/100
250/250 0s 1ms/step - accuracy: 0.8650 - loss: 0.3292
Epoch 58/100
250/250 1s 2ms/step - accuracy: 0.8659 - loss: 0.3288
Epoch 59/100
250/250 1s 2ms/step - accuracy: 0.8566 - loss: 0.3439
Epoch 60/100
250/250 1s 2ms/step - accuracy: 0.8632 - loss: 0.3366

Epoch 61/100
250/250 0s 1ms/step - accuracy: 0.8609 - loss: 0.3427
Epoch 62/100
250/250 1s 1ms/step - accuracy: 0.8614 - loss: 0.3352
Epoch 63/100
250/250 1s 2ms/step - accuracy: 0.8670 - loss: 0.3237
Epoch 64/100
250/250 1s 2ms/step - accuracy: 0.8693 - loss: 0.3210
Epoch 65/100
250/250 0s 2ms/step - accuracy: 0.8668 - loss: 0.3339
Epoch 66/100
250/250 0s 2ms/step - accuracy: 0.8672 - loss: 0.3294
Epoch 67/100
250/250 0s 2ms/step - accuracy: 0.8676 - loss: 0.3231
Epoch 68/100
250/250 0s 2ms/step - accuracy: 0.8719 - loss: 0.3250
Epoch 69/100
250/250 1s 2ms/step - accuracy: 0.8657 - loss: 0.3307
Epoch 70/100
250/250 0s 2ms/step - accuracy: 0.8639 - loss: 0.3296
Epoch 71/100
250/250 1s 2ms/step - accuracy: 0.8654 - loss: 0.3381
Epoch 72/100
250/250 0s 2ms/step - accuracy: 0.8670 - loss: 0.3293
Epoch 73/100
250/250 0s 2ms/step - accuracy: 0.8630 - loss: 0.3389
Epoch 74/100
250/250 0s 2ms/step - accuracy: 0.8653 - loss: 0.3293
Epoch 75/100
250/250 0s 2ms/step - accuracy: 0.8598 - loss: 0.3433
Epoch 76/100
250/250 1s 1ms/step - accuracy: 0.8635 - loss: 0.3287
Epoch 77/100
250/250 0s 2ms/step - accuracy: 0.8626 - loss: 0.3352
Epoch 78/100
250/250 0s 1ms/step - accuracy: 0.8628 - loss: 0.3375
Epoch 79/100
250/250 0s 2ms/step - accuracy: 0.8659 - loss: 0.3220
Epoch 80/100
250/250 1s 1ms/step - accuracy: 0.8614 - loss: 0.3361
Epoch 81/100
250/250 0s 2ms/step - accuracy: 0.8571 - loss: 0.3366
Epoch 82/100
250/250 0s 2ms/step - accuracy: 0.8649 - loss: 0.3371
Epoch 83/100
250/250 0s 2ms/step - accuracy: 0.8586 - loss: 0.3468
Epoch 84/100
250/250 1s 2ms/step - accuracy: 0.8650 - loss: 0.3320
Epoch 85/100
250/250 1s 2ms/step - accuracy: 0.8686 - loss: 0.3268
Epoch 86/100
250/250 0s 2ms/step - accuracy: 0.8656 - loss: 0.3343
Epoch 87/100
250/250 0s 2ms/step - accuracy: 0.8672 - loss: 0.3269
Epoch 88/100
250/250 0s 2ms/step - accuracy: 0.8641 - loss: 0.3251
Epoch 89/100
250/250 0s 2ms/step - accuracy: 0.8644 - loss: 0.3315
Epoch 90/100
250/250 0s 2ms/step - accuracy: 0.8699 - loss: 0.3255

```

Epoch 91/100
250/250 ━━━━━━━━ 0s 2ms/step - accuracy: 0.8679 - loss: 0.3274
Epoch 92/100
250/250 ━━━━ 1s 2ms/step - accuracy: 0.8685 - loss: 0.3265
Epoch 93/100
250/250 ━━━━ 0s 2ms/step - accuracy: 0.8662 - loss: 0.3296
Epoch 94/100
250/250 ━━━━ 1s 2ms/step - accuracy: 0.8643 - loss: 0.3307
Epoch 95/100
250/250 ━━━━ 0s 2ms/step - accuracy: 0.8602 - loss: 0.3287
Epoch 96/100
250/250 ━━━━ 1s 2ms/step - accuracy: 0.8705 - loss: 0.3256
Epoch 97/100
250/250 ━━━━ 0s 2ms/step - accuracy: 0.8633 - loss: 0.3346
Epoch 98/100
250/250 ━━━━ 1s 2ms/step - accuracy: 0.8617 - loss: 0.3355
Epoch 99/100
250/250 ━━━━ 1s 2ms/step - accuracy: 0.8629 - loss: 0.3357
Epoch 100/100
250/250 ━━━━ 1s 2ms/step - accuracy: 0.8726 - loss: 0.3147

```

Out[27]: <keras.src.callbacks.history.History at 0x19a37c02810>

```
In [28]: y_pred=ann.predict(x_test)
y_pred=(y_pred>0.5)
print(np.concatenate((y_pred.reshape(len(y_pred),1),y_test.reshape(len(y_test),1
```

```
63/63 ━━━━━━━━ 0s 4ms/step
[[0 0]
 [0 1]
 [0 0]
 ...
 [0 0]
 [0 0]
 [0 0]]
```

```
In [29]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
print(cm)
```

```
[[1511  84]
 [ 204 201]]
```

```
In [30]: from sklearn.metrics import accuracy_score
ac=accuracy_score(y_test,y_pred)
ac
```

Out[30]: 0.856

In []: