

## ▼ Artificial Neural Network

```
import numpy as np
import pandas as pd
import tensorflow as tf

dataset = pd.read_csv('dataset.csv')
X = dataset.iloc[:, 3:-1].values
y = dataset.iloc[:, -1].values

print(X)

[[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]]

print(y)

[1 0 1 ... 1 1 0]
```

## ▼ Encoding categorical data

Label Encoding the "Gender" column

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

```
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]]
```

### One Hot Encoding the "Geography" column

```
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
ct = ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [1])], remainder='passthrough')
X = np.array(ct.fit_transform(X))
```

```
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]]
```

### ▼ Splitting the dataset into the Training set and Test set

```
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)
```

## ▼ Feature Scaling

```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

## ▼ Initializing the ANN

```
ann = tf.keras.models.Sequential()
```

## ▼ Adding the input layer and the first hidden layer

```
ann.add(tf.keras.layers.Dense(units=6, activation='relu'))
```

## ▼ Adding the second hidden layer

```
ann.add(tf.keras.layers.Dense(units=6, activation='relu'))
```

## ▼ Adding the output layer

```
ann.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))
```

## ▼ Compiling the ANN

```
ann.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accuracy'])
```

## ▼ Training the ANN on the Training set

```
ann.fit(X_train, y_train, batch_size = 32, epochs = 100)
```

```
Epoch 1/100
250/250 [=====] - 0s 1ms/step - loss: 0.3349 - accuracy: 0.8622
Epoch 2/100
250/250 [=====] - 0s 1ms/step - loss: 0.3346 - accuracy: 0.8626
Epoch 3/100
250/250 [=====] - 0s 1ms/step - loss: 0.3347 - accuracy: 0.8634
Epoch 4/100
250/250 [=====] - 0s 1ms/step - loss: 0.3348 - accuracy: 0.8633
Epoch 5/100
250/250 [=====] - 0s 1ms/step - loss: 0.3346 - accuracy: 0.8626
Epoch 6/100
250/250 [=====] - 0s 1ms/step - loss: 0.3342 - accuracy: 0.8626
Epoch 7/100
250/250 [=====] - 0s 1ms/step - loss: 0.3345 - accuracy: 0.8634
Epoch 8/100
250/250 [=====] - 0s 1ms/step - loss: 0.3344 - accuracy: 0.8654
Epoch 9/100
250/250 [=====] - 0s 1ms/step - loss: 0.3336 - accuracy: 0.8636
Epoch 10/100
250/250 [=====] - 0s 1ms/step - loss: 0.3342 - accuracy: 0.8629
Epoch 11/100
250/250 [=====] - 0s 1ms/step - loss: 0.3342 - accuracy: 0.8616
Epoch 12/100
250/250 [=====] - 0s 1ms/step - loss: 0.3340 - accuracy: 0.8620
Epoch 13/100
250/250 [=====] - 0s 1ms/step - loss: 0.3338 - accuracy: 0.8629
Epoch 14/100
250/250 [=====] - 0s 1ms/step - loss: 0.3337 - accuracy: 0.8641
Epoch 15/100
250/250 [=====] - 0s 1ms/step - loss: 0.3338 - accuracy: 0.8633
Epoch 16/100
250/250 [=====] - 0s 1ms/step - loss: 0.3333 - accuracy: 0.8633
Epoch 17/100
250/250 [=====] - 0s 1ms/step - loss: 0.3336 - accuracy: 0.8644
Epoch 18/100
250/250 [=====] - 0s 1ms/step - loss: 0.3331 - accuracy: 0.8624
Epoch 19/100
```

```

Epoch 89/100
250/250 [=====] - 0s 1ms/step - loss: 0.3325 - accuracy: 0.8637
Epoch 90/100
250/250 [=====] - 0s 1ms/step - loss: 0.3328 - accuracy: 0.8640
Epoch 91/100
250/250 [=====] - 0s 1ms/step - loss: 0.3333 - accuracy: 0.8626
Epoch 92/100
250/250 [=====] - 0s 1ms/step - loss: 0.3332 - accuracy: 0.8637
Epoch 93/100
250/250 [=====] - 0s 1ms/step - loss: 0.3330 - accuracy: 0.8636
Epoch 94/100
250/250 [=====] - 0s 1ms/step - loss: 0.3329 - accuracy: 0.8650
Epoch 95/100
250/250 [=====] - 0s 1ms/step - loss: 0.3327 - accuracy: 0.8648
Epoch 96/100
250/250 [=====] - 0s 1ms/step - loss: 0.3325 - accuracy: 0.8641
Epoch 97/100
250/250 [=====] - 0s 1ms/step - loss: 0.3323 - accuracy: 0.8644
Epoch 98/100
250/250 [=====] - 0s 1ms/step - loss: 0.3320 - accuracy: 0.8646
Epoch 99/100
250/250 [=====] - 0s 1ms/step - loss: 0.3323 - accuracy: 0.8640
Epoch 100/100
250/250 [=====] - 0s 1ms/step - loss: 0.3321 - accuracy: 0.8639

```

```

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)),1))

```

```

[[0 0]
 [0 1]
 [0 0]
 ...
 [0 0]
 [0 0]
 [0 0]]

```

```

from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)

```

```
[[1523  72]  
 [ 198 207]]  
0.865
```

