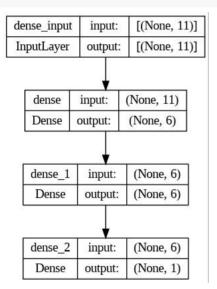
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
dataset = pd.read_csv("/content/drive/MyDrive/Colab Notebooks/Bank Customer Churn Prediction.csv")
dataset.head()
 С→
         customer_id credit_score country gender age tenure balance products_n
            15634602
                                  619
                                         France Female
                                                                             0.00
      1
            15647311
                                  608
                                          Spain
                                                 Female
                                                           41
                                                                    1 83807.86
      2
             15619304
                                  502
                                                           42
                                                                     8 159660.80
                                                 Female
                                         France
                                                 Female
      3
             15701354
                                  699
                                                           39
                                                                     1
                                                                             0.00
                                         France
      4
             15737888
                                                                   2 125510 82
                                  850
                                                           43
                                          Spain Female
X = dataset.drop(['churn'],axis = 1).values
y = dataset['churn'].values
X.shape
     (10000, 11)
y.shape
     (10000,)
     array([[15634602, 619, 'France', ..., 1, 1, 101348.88], [15647311, 608, 'Spain', ..., 0, 1, 112542.58], [15619304, 502, 'France', ..., 1, 0, 113931.57],
             [15584532, 709, 'France', ..., 0, 1, 42085.58],
[15682355, 772, 'Germany', ..., 1, 0, 92888.52],
[15628319, 792, 'France', ..., 1, 0, 38190.78]], dtype=object)
X[:,2]
     array(['France', 'Spain', 'France', ..., 'France', 'Germany', 'France'],
            dtype=object)
from sklearn.preprocessing import LabelEncoder, OneHotEncoder
labelencoded_x1 = LabelEncoder()
X[:,2] = labelencoded_x1.fit_transform(X[:,2])
X[:,2]
     array([0, 2, 0, ..., 0, 1, 0], dtype=object)
X[:,3]
     array(['Female', 'Female', 'Female', ..., 'Female', 'Male', 'Female'],
            dtype=object)
labelencoded_x2 = LabelEncoder()
X[:,3] = labelencoded_x2.fit_transform(X[:,3])
X[:,3]
     array([0, 0, 0, ..., 0, 1, 0], dtype=object)
# split data into train and test
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)
# feature scaling
{\tt from \ sklearn.preprocessing \ import \ StandardScaler}
sc = StandardScaler()
X train = sc.fit transform(X train)
X_test = sc.transform(X_test)
# import required libraries for model creation
```

```
import keras
from keras.models import Sequential
from keras.layers import Dense
# initialize neural network
classifier = Sequential()
classifier.add(Dense(units = 6,kernel initializer='uniform',activation = 'relu',input dim = 11))
classifier.add(Dense(units = 6,kernel_initializer='uniform',activation = 'relu'))
classifier.add(Dense(units = 1,kernel initializer='uniform',activation = 'sigmoid'))
# compiling neural network
classifier.compile(optimizer = 'adam',loss = 'binary crossentropy',metrics = ['accuracy'])
# model fitting
classifier.fit(X_train,y_train,batch_size = 10, epochs = 100)
                                        ID IMD/DUCP
                                                    1000. 0.1000
   Epoch 38/100
   800/800 [============== ] - 1s 1ms/step - loss: 0.4030 - accuracy: 0.8363
   Epoch 39/100
   800/800 [====
                     ========= 1 - 1s 1ms/step - loss: 0.4032 - accuracy: 0.8356
   Epoch 40/100
   800/800 [======
                   Epoch 41/100
   800/800 [============= ] - 1s 1ms/step - loss: 0.4029 - accuracy: 0.8357
   Epoch 42/100
   800/800 [=====
                       ======== ] - 1s 1ms/step - loss: 0.4024 - accuracy: 0.8356
   Epoch 43/100
   800/800 [=====
                      ========] - 1s 2ms/step - loss: 0.4024 - accuracy: 0.8360
   Epoch 44/100
   800/800 [=====
                      ========= 1 - 1s 2ms/step - loss: 0.4025 - accuracy: 0.8354
   Epoch 45/100
   800/800 [=====
                    Epoch 46/100
   800/800 [=====
                      =============== | - 1s 1ms/step - loss: 0.4028 - accuracy: 0.8354
   Epoch 47/100
   800/800 [====
                       ======== ] - 1s 1ms/step - loss: 0.4022 - accuracy: 0.8355
   Epoch 48/100
   800/800 [====
                      ======== ] - 1s 1ms/step - loss: 0.4020 - accuracy: 0.8360
   Epoch 49/100
   800/800 [=====
                    =================== | - 1s 1ms/step - loss: 0.4026 - accuracy: 0.8364
   Epoch 50/100
   800/800 [====
                      ========= | - 1s 1ms/step - loss: 0.4022 - accuracy: 0.8351
   Epoch 51/100
   800/800 [=====
                      ========] - 1s 1ms/step - loss: 0.4020 - accuracy: 0.8372
   Epoch 52/100
   800/800 [======
                   ======== ] - 1s 1ms/step - loss: 0.4021 - accuracy: 0.8370
   Epoch 53/100
   800/800 [====
                       ======== ] - 2s 2ms/step - loss: 0.4019 - accuracy: 0.8349
   Epoch 54/100
   800/800 [============] - 3s 4ms/step - loss: 0.4018 - accuracy: 0.8355
   Epoch 55/100
                     ========== ] - 2s 2ms/step - loss: 0.4019 - accuracy: 0.8361
   800/800 [=====
   Epoch 56/100
   800/800 [============] - 1s 1ms/step - loss: 0.4008 - accuracy: 0.8371
   Epoch 57/100
   800/800 [=====
                   Epoch 58/100
   800/800 [====
                         ========] - 1s 1ms/step - loss: 0.4012 - accuracy: 0.8375
   Epoch 59/100
   800/800 [====
                       ========] - 1s 1ms/step - loss: 0.4016 - accuracy: 0.8350
   Epoch 60/100
   800/800 [====
                      ========== ] - 1s 1ms/step - loss: 0.4012 - accuracy: 0.8359
   Epoch 61/100
   800/800 [===========] - 1s 1ms/step - loss: 0.4020 - accuracy: 0.8350
   Epoch 62/100
   800/800 [============ ] - 1s 1ms/step - loss: 0.4014 - accuracy: 0.8355
   Epoch 63/100
   800/800 [============] - 1s 1ms/step - loss: 0.4014 - accuracy: 0.8351
   Epoch 64/100
   800/800 [======
                   Epoch 65/100
   800/800 [=====
                      ========== 1 - 1s 1ms/step - loss: 0.4013 - accuracy: 0.8346
   Epoch 66/100
   800/800 [============== ] - 1s 2ms/step - loss: 0.4010 - accuracy: 0.8350
```

```
y_pred = classifier.predict(X_test)
y_pred = (y_pred>0.5)
y_pred
    63/63 [========== ] - 0s 911us/step
    array([[ True],
           [False],
           [False],
           [False],
           [False],
           [False]])
y_test = _y_test
# confusion matrix
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test,y_pred)
print(cm)
    [[1540 67]
[ 256 137]]
#from confusion_matrix import accuracy
from sklearn.metrics import accuracy_score
accuracy_score(y_test, y_pred)
```

0.8385

from keras.utils.vis_utils import plot_model
plot_model(classifier,to_file = 'model_plot.jpg',show_shapes = True,show_layer_names = True)



classifier.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 6)	72
dense_1 (Dense)	(None, 6)	42
dense_2 (Dense)	(None, 1)	7
Total params: 121		
-		
Trainable params: 121		
Non-trainable params: 0		
·		