Chapter 3: ANN - Demo

```
In [1]: # from google.colab import drive
        # drive.mount("/content/gdrive", force_remount=True)
In [2]: # %cd '/content/gdrive/My Drive/LDS8 DeepLearning/Practice/Chapter3/'
In [3]: import warnings
        warnings.filterwarnings('ignore')
In [4]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import tensorflow as tf
        from tensorflow import keras
In [5]: print(tf.__version__)
        print(keras. version )
        2.5.0
        2.5.0
        Đọc dữ liệu, Tiền xử lý dữ liệu
In [6]: df = pd.read csv("Churn Modelling.csv")
```

```
In [7]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):

| # | Column | Non-Null Count | Dtype | | |
|--|-----------------|----------------|---------|--|--|
| | | | | | |
| 0 | RowNumber | 10000 non-null | int64 | | |
| 1 | CustomerId | 10000 non-null | int64 | | |
| 2 | Surname | 10000 non-null | object | | |
| 3 | CreditScore | 10000 non-null | int64 | | |
| 4 | Geography | 10000 non-null | object | | |
| 5 | Gender | 10000 non-null | object | | |
| 6 | Age | 10000 non-null | int64 | | |
| 7 | Tenure | 10000 non-null | int64 | | |
| 8 | Balance | 10000 non-null | float64 | | |
| 9 | NumOfProducts | 10000 non-null | int64 | | |
| 10 | HasCrCard | 10000 non-null | int64 | | |
| 11 | IsActiveMember | 10000 non-null | int64 | | |
| 12 | EstimatedSalary | 10000 non-null | float64 | | |
| 13 | Exited | 10000 non-null | int64 | | |
| <pre>dtypes: float64(2), int64(9), object(3)</pre> | | | | | |
| memory usage: 1.1+ MB | | | | | |

In [8]: df.head()

Out[8]:

| | RowNumber | CustomerId | Surname | CreditScore | Geography | Gender | Age | Tenure | Balance |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|-----------|
| 0 | 1 | 15634602 | Hargrave | 619 | France | Female | 42 | 2 | 0.00 |
| 1 | 2 | 15647311 | Hill | 608 | Spain | Female | 41 | 1 | 83807.86 |
| 2 | 3 | 15619304 | Onio | 502 | France | Female | 42 | 8 | 159660.80 |
| 3 | 4 | 15701354 | Boni | 699 | France | Female | 39 | 1 | 0.00 |
| 4 | 5 | 15737888 | Mitchell | 850 | Spain | Female | 43 | 2 | 125510.82 |
| | | | | | | | | | |

In [9]: df.groupby(by='Geography').count()

Out[9]:

RowNumber Customerld Surname CreditScore Gender Age Tenure Balance Num Geography **France** 5014 5014 5014 5014 5014 5014 5014 5014 Germany 2509 2509 2509 2509 2509 2509 2509 2509 Spain 2477 2477 2477 2477 2477 2477 2477 2477

In [10]: X = df.iloc[:, 3:13].values # inputs
y = df.iloc[:, 13].values # output

```
In [11]: # khach hang ra di
          exited = np.count nonzero(y)
          exited
Out[11]: 2037
In [12]: # khach hang o lai
          no_exited = y.shape[0] - exited
          no exited
Out[12]: 7963
In [13]: from scipy.stats import itemfreq
          # khach hang o Lai/ra di
          itemfreq(y)
Out[13]: array([[
                     0, 7963],
                     1, 2037]], dtype=int64)
In [14]: # in thong tin 10 khach hang dau tien
          X[:10]
Out[14]: array([[619, 'France', 'Female', 42, 2, 0.0, 1, 1, 101348.88],
                 [608, 'Spain', 'Female', 41, 1, 83807.86, 1, 0, 1, 112542.58],
                 [502, 'France', 'Female', 42, 8, 159660.8, 3, 1, 0, 113931.57],
                 [699, 'France', 'Female', 39, 1, 0.0, 2, 0, 0, 93826.63],
                 [850, 'Spain', 'Female', 43, 2, 125510.82, 1, 1, 1, 79084.1],
                 [645, 'Spain', 'Male', 44, 8, 113755.78, 2, 1, 0, 149756.71], [822, 'France', 'Male', 50, 7, 0.0, 2, 1, 1, 10062.8],
                 [376, 'Germany', 'Female', 29, 4, 115046.74, 4, 1, 0, 119346.88],
                 [501, 'France', 'Male', 44, 4, 142051.07, 2, 0, 1, 74940.5],
                 [684, 'France', 'Male', 27, 2, 134603.88, 1, 1, 1, 71725.73]],
                dtype=object)
```

```
In [15]: # Encoding categorical data
         # Encoding the Independent Variable
         from sklearn.preprocessing import LabelEncoder, OneHotEncoder
         from sklearn.compose import ColumnTransformer
         labelencoder X 1 = LabelEncoder()
         X[:, 1] = labelencoder X 1.fit transform(X[:, 1])
         labelencoder X 2 = LabelEncoder()
         X[:, 2] = labelencoder X 2.fit transform(X[:, 2])
         ct = ColumnTransformer([('encoder', OneHotEncoder(), [1])],
                                remainder='passthrough')
         X = ct.fit_transform(X.tolist())
         # save 3 model nay pickle: labelencoder X 1, labelencoder X 2, ct
         # Save the Modle to file in the current working directory
         # Pkl Filename = "Pickle RL Model.pkl"
         # with open(Pkl_Filename, 'wb') as file:
               pickle.dump(Model, file)
         # Load the Model back from file
         # with open(Pkl Filename, 'rb') as file:
               model load = pickle.load(file)
In [16]: X[:10]
Out[16]: array([[1.0, 0.0, 0.0, 619, 0, 42, 2, 0.0, 1, 1, 1, 101348.88],
                [0.0, 0.0, 1.0, 608, 0, 41, 1, 83807.86, 1, 0, 1, 112542.58],
                [1.0, 0.0, 0.0, 502, 0, 42, 8, 159660.8, 3, 1, 0, 113931.57],
                [1.0, 0.0, 0.0, 699, 0, 39, 1, 0.0, 2, 0, 0, 93826.63],
                [0.0, 0.0, 1.0, 850, 0, 43, 2, 125510.82, 1, 1, 1, 79084.1],
                [0.0, 0.0, 1.0, 645, 1, 44, 8, 113755.78, 2, 1, 0, 149756.71],
                [1.0, 0.0, 0.0, 822, 1, 50, 7, 0.0, 2, 1, 1, 10062.8],
                [0.0, 1.0, 0.0, 376, 0, 29, 4, 115046.74, 4, 1, 0, 119346.88],
                [1.0, 0.0, 0.0, 501, 1, 44, 4, 142051.07, 2, 0, 1, 74940.5],
                [1.0, 0.0, 0.0, 684, 1, 27, 2, 134603.88, 1, 1, 1, 71725.73]],
               dtype=object)
In [17]: # vì chỉ có 3 quốc gia, nên khi tạo ra ma trận OneHotEncoder quốc gia sẽ như sau:
         # France 1, 0, 0
         # Germany 0, 1, 0
         # Spain
                  0, 0, 1
         # => Theo đó có thể bỏ đi 1 cột đầu thì 2 cột sau là đủ thể hiện mã của 3 quốc gi
         # France 0. 0
         # Germany 1, 0
         # Spain 0, 1
         X = X[:, 1:]
```

```
In [18]: X[:10]
Out[18]: array([[0.0, 0.0, 619, 0, 42, 2, 0.0, 1, 1, 1, 101348.88],
                [0.0, 1.0, 608, 0, 41, 1, 83807.86, 1, 0, 1, 112542.58],
                [0.0, 0.0, 502, 0, 42, 8, 159660.8, 3, 1, 0, 113931.57],
                [0.0, 0.0, 699, 0, 39, 1, 0.0, 2, 0, 0, 93826.63],
                [0.0, 1.0, 850, 0, 43, 2, 125510.82, 1, 1, 1, 79084.1],
                [0.0, 1.0, 645, 1, 44, 8, 113755.78, 2, 1, 0, 149756.71],
                [0.0, 0.0, 822, 1, 50, 7, 0.0, 2, 1, 1, 10062.8],
                [1.0, 0.0, 376, 0, 29, 4, 115046.74, 4, 1, 0, 119346.88],
                [0.0, 0.0, 501, 1, 44, 4, 142051.07, 2, 0, 1, 74940.5],
                [0.0, 0.0, 684, 1, 27, 2, 134603.88, 1, 1, 1, 71725.73]],
               dtype=object)
In [19]: # scale truoc => split sau
         from sklearn.preprocessing import StandardScaler
         sc = StandardScaler()
         sc.fit(X)
         X = sc.transform(X)
         import pickle
         scalerfile = 'sc.sav'
         pickle.dump(sc, open(scalerfile, 'wb'))
         # Hoac lam 1 pipeline: Lable Encoder => OnhotEncoder => Stadard scaler => model p
         # o noi khac: goi model pre
In [20]: # 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)
In [21]: X train[:5]
Out[21]: array([[-0.57873591, 1.74273971, 0.17042381, -1.09598752, -0.4693113,
                 -0.00442596, -1.22584767, 0.80773656, 0.64609167, -1.03067011,
                  1.10838187],
                [ 1.72790383, -0.57380915, -2.31280236, 0.91241915, 0.29351742,
                 -1.38753759, -0.01289171, -0.91158349, 0.64609167, 0.97024255,
                 -0.74759209],
                [-0.57873591, -0.57380915, -1.19535058, -1.09598752, -0.94607926,
                 -1.04175968, 0.57507592, -0.91158349, 0.64609167, -1.03067011,
                  1.48746417],
                [-0.57873591, 1.74273971, 0.03591573, 0.91241915, 0.10281024,
                               0.46795537, -0.91158349, 0.64609167, -1.03067011,
                 -0.00442596,
                  1.27855845],
                [-0.57873591, 1.74273971, 2.06388377, -1.09598752, 1.72382128,
                  1.03290776,
                               0.80600989, 0.80773656, 0.64609167, 0.97024255,
                  0.56006929]])
```

```
In [22]: X_train[0].shape
Out[22]: (11,)
```

Xây dựng model ANN

```
In [23]: from tensorflow.keras.layers import Dense, Dropout
    from tensorflow.keras import Sequential
```

```
In [24]: # Intitialing the ANN
classifier = Sequential()
```

- khởi tạo ANN sử dụng keras Sequential class và đặt tên là classifier.
- Tạo một Dense input layer với output_dim là 6 (11 inputs + 1 output)/2, hoặc 12 (11+1).
 Không có một quy tắc nghiêm ngặt nào cho giá trị này,
- activation function: 'relu' (thường được áp dụng cho input layer và các hidden network layers)
- input shape = (11,) (có independent input variables trong X matrix)
- => => add vào hidden layer sử dụng classifier.add()

```
In [25]: # Adding the input layer and the first hidden layer
    classifier.add(Dense(units = 6, activation='relu', input_shape=(11,)))
    classifier.add(Dropout(rate=0.1))
```

Tạo Dense hidden layer tiếp theo với cùng tham số mà không cần input_dim (chỉ cần khai báo input dim ở input layer) => add vào hidden layer sử dụng classifier.add()

 Tạo output layer với output_dim = 1 và sigmoid activation function (kết quả chỉ có 1) => add vào network sử dụng classifier.add().

 Compile neural network sử dụng 'adam' optimizer, 'binary_crossentropy' loss function cho accuracy metrics

In [29]: classifier.summary()

Model: "sequential"

| Layer (type) | Output Shape | Param # |
|---------------------|--------------|---------|
| dense (Dense) | (None, 6) | 72 |
| dropout (Dropout) | (None, 6) | 0 |
| dense_1 (Dense) | (None, 6) | 42 |
| dropout_1 (Dropout) | (None, 6) | 0 |
| dense_2 (Dense) | (None, 1) | 7 |

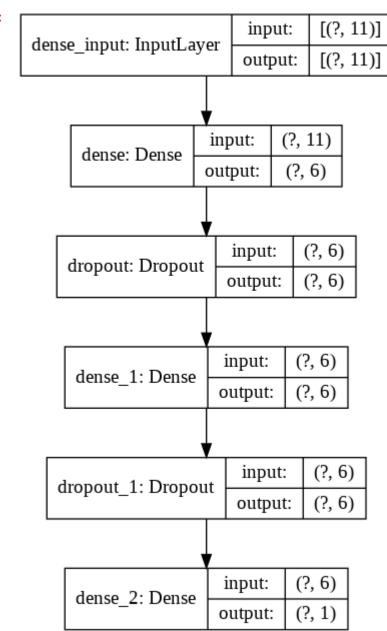
Total params: 121 Trainable params: 121 Non-trainable params: 0

In [30]: from tensorflow.keras.utils import plot_model
from IPython.display import Image

In [31]: plot_model(classifier, to_file='ANN_model.png', show_shapes=True) Image(filename='ANN_model.png')

('You must install pydot (`pip install pydot`) and install graphviz (see instru ctions at https://graphviz.gitlab.io/download/) (https://graphviz.gitlab.io/download/)) ', 'for plot model/model to dot to work.')

Out[31]:



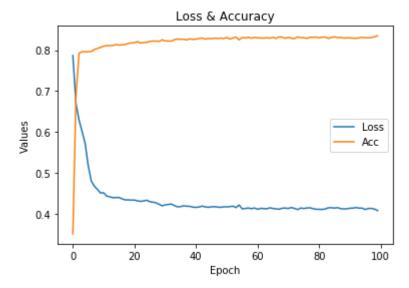
Cách tính params: fully connected: Param # = input * output + output

```
• dense input: Input layer: shape(None, 11), Param #=0
```

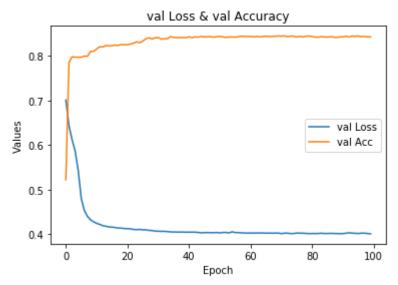
- dense:Dense: shape(?, 6), Param # = input * output + output = 11 * 6 + 6 = 72
- dense 1:Dense: shape(?, 6), Param # = input * output + output = 6 * 6 + 6 = 42
- dense 2:Dense: shape(?, 6), Param # = input * output + output = 6 * 1 + 1 = 7

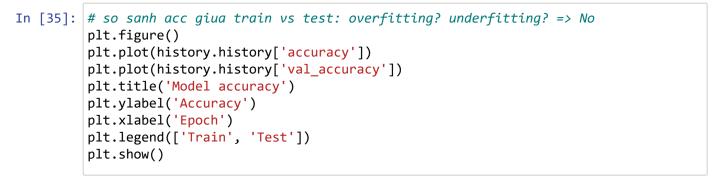
```
In [32]: # Fitting classifier to the Training set
        history = classifier.fit(X_train, y_train,
                                epochs = 100,
                                batch size=64,
                                validation_data=(X_test, y_test)) # default batch size =
        Epoch 1/100
        125/125 [============= ] - 1s 2ms/step - loss: 0.7869 - accur
        acy: 0.3510 - val loss: 0.7013 - val accuracy: 0.5225
        Epoch 2/100
        125/125 [============= ] - 0s 965us/step - loss: 0.6706 - acc
        uracy: 0.6870 - val loss: 0.6443 - val accuracy: 0.7855
        Epoch 3/100
        125/125 [================= ] - 0s 984us/step - loss: 0.6290 - acc
        uracy: 0.7925 - val loss: 0.6128 - val accuracy: 0.7985
        Epoch 4/100
        125/125 [============== ] - 0s 981us/step - loss: 0.6011 - acc
        uracy: 0.7964 - val loss: 0.5874 - val accuracy: 0.7975
        Epoch 5/100
        125/125 [=============== ] - 0s 989us/step - loss: 0.5713 - acc
        uracy: 0.7960 - val loss: 0.5438 - val accuracy: 0.7975
        Epoch 6/100
        125/125 [============= ] - 0s 997us/step - loss: 0.5168 - acc
        uracy: 0.7964 - val loss: 0.4806 - val accuracy: 0.7975
        Epoch 7/100
         40-/40- -
```

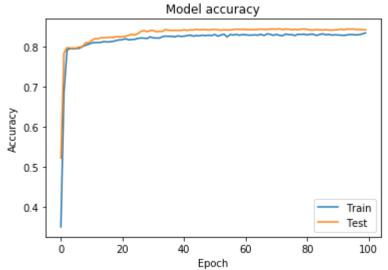
```
In [33]: plt.figure()
plt.plot(history.history['loss'])
plt.plot(history.history['accuracy'])
plt.title('Loss & Accuracy')
plt.ylabel('Values')
plt.xlabel('Epoch')
plt.legend(['Loss', 'Acc'])
plt.show()
```



```
In [34]: plt.figure()
  plt.plot(history.history['val_loss'])
  plt.plot(history.history['val_accuracy'])
  plt.title('val Loss & val Accuracy')
  plt.ylabel('Values')
  plt.xlabel('Epoch')
  plt.legend(['val Loss', 'val Acc'])
  plt.show()
```







```
In [36]: # Part 3: Making a predictions and evaluating the models
# Predicting the Test set results
y_pred = classifier.predict(X_test)
```

Đánh giá model và dự đoán dữ liệu mới

```
In [37]: y_pred[:5]
Out[37]: array([[0.26806527],
                [0.29092318],
                [0.1793733],
                [0.07376492],
                [0.20785925]], dtype=float32)
In [38]: # trên 0.5 => 1, ngược lại => 0
         y_pred = y_pred >= 0.5
         print(y_pred)
         [[False]
          [False]
          [False]
          [False]
          [False]
          [False]]
In [39]: # Making the Confusion Matrix
         from sklearn.metrics import confusion matrix
         cm = confusion matrix(y test, y pred)
         \mathsf{cm}
Out[39]: array([[1559,
                         36],
                [ 278, 127]], dtype=int64)
In [40]: |acc = (cm[0][0] + cm[1][1])/ (cm[0][0] + cm[1][1] + cm[0][1] + cm[1][0])
         print(acc)
         0.843
In [41]: # Evaluation
         print("Accuracy:", classifier.evaluate(X_test, y_test))
         63/63 [============ ] - 0s 708us/step - loss: 0.4012 - accurac
         v: 0.8430
         Accuracy: [0.4012296199798584, 0.8429999947547913]
```

```
In [42]: #Save the result
    from tensorflow.keras.models import load_model
    # Creates a HDF5 file 'my_model.h5'
    classifier.save('ANN_model.h5')
```

Use our ANN model to predict if the customer with the following informations will leave the bank:

Geography: FranceCredit Score: 600Gender: MaleAge: 40 years old

Tenure: 3 yearsBalance: 60000

• Number of Products: 2

- · Does this customer have a credit card ? Yes
- · Is this customer an Active Member: Yes
- Estimated Salary: 50000
- So should we say goodbye to that customer?