

Prediction

April 7, 2025

0.1 Prediction With Trained ANN Model

```
[31]: import tensorflow as tf
from tensorflow.keras.models import load_model
import pickle
import pandas as pd
import numpy as np
```

0.1.1 Loading the models

```
[32]: ## Load the trained model, scaler pickle, onehot_encoder
model = load_model('model.h5')

## Load the encoder and scaler
with open('onehot_encoder_geo.pkl', 'rb') as file:
    label_encoder_geo = pickle.load(file)

with open('label_encoder_gender.pkl', 'rb') as file:
    label_encoder_gender = pickle.load(file)

with open('scaler.pkl', 'rb') as file:
    scaler = pickle.load(file)
```

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you train or evaluate the model.

Taking the Input data for model

```
[33]: ## Example input data

input_data = {
    'CreditScore': 600,
    'Geography' : 'France',
    'Gender' : 'Male',
    'Age': 40,
    'Tenure': 3,
    'Balance': 60000,
```

```

        'NumOfProducts': 2,
        'HasCrCard': 1,
        'IsActiveMember': 1,
        'EstimatedSalary': 50000
    }

```

```

[34]: ## converting the geography char data in int data using the one hot encoding
      geo_encoded = label_encoder_geo.transform([[input_data['Geography']]]).toarray()
      geo_encoded_df = pd.DataFrame(geo_encoded, columns=label_encoder_geo.
      ↪get_feature_names_out(['Geography']))
      geo_encoded_df

```

```

c:\Users\Uditya\Desktop\End-to-End-Deep-Learning-Project-Using-
ANN\venv\Lib\site-packages\sklearn\utils\validation.py:2739: UserWarning: X does
not have valid feature names, but OneHotEncoder was fitted with feature names
warnings.warn(

```

```

[34]:   Geography_France  Geography_Germany  Geography_Spain
      0                1.0                0.0                0.0

```

```

[35]: input_df = pd.DataFrame([input_data])
      input_df

```

```

[35]:   CreditScore  Geography  Gender  ...  HasCrCard  IsActiveMember  EstimatedSalary
      0          600    France  Male  ...          1                1            50000

```

```

[1 rows x 10 columns]

```

```

[36]: ## Encode categorical variables
      input_df['Gender']=label_encoder_gender.transform(input_df['Gender'])
      input_df

```

```

[36]:   CreditScore  Geography  Gender  ...  HasCrCard  IsActiveMember
      EstimatedSalary
      0          600    France      1  ...          1                1
      50000

```

```

[1 rows x 10 columns]

```

```

[37]: ## combine one-hot encoded columns with input data
      input_df = pd.concat([input_df.drop("Geography", axis=1), geo_encoded_df],
      ↪axis=1)
      input_df

```

```

[37]:   CreditScore  Gender  ...  Geography_Germany  Geography_Spain
      0          600      1  ...                0.0                0.0

```

```

[1 rows x 12 columns]

```

Scaling the data

```
[38]: input_scaled= scaler.transform(input_df)
input_scaled
```

```
[38]: array([[ -0.53598516,  0.91324755,  0.10479359, -0.69539349, -0.25781119,
          0.80843615,  0.64920267,  0.97481699, -0.87683221,  1.00150113,
          -0.57946723, -0.57638802]])
```

Prediction churn

```
[39]: ## Prediction churn

prediction=model.predict(input_scaled)
prediction
```

1/1 0s 331ms/step

```
[39]: array([[0.02760429]], dtype=float32)
```

```
[40]: prediction_proba = prediction[0][0]
```

```
[41]: prediction_proba
```

```
[41]: np.float32(0.02760429)
```

```
[42]: if prediction_proba > 0.5:
        print("The customer is likely to churn.")
    else:
        print("The customer is not likely to churn.")
```

The customer is not likely to churn.

```
[ ]:
```