

Importing Libraries

```
In [1]: 1 import pandas as pd
2 import numpy as np
3 import seaborn as sns
4 import matplotlib.pyplot as plt
5
6 from sklearn.model_selection import train_test_split
7 from sklearn.metrics import classification_report
8
9 from tensorflow.keras import Sequential
10 from tensorflow.keras.layers import Dense,Dropout
11 from tensorflow.keras.callbacks import EarlyStopping
```

Loading Data

```
In [2]: 1 df = pd.read_excel("/content/customer_churn_large_dataset.xlsx")
2 df.head()
```

Out[2]:

	CustomerID	Name	Age	Gender	Location	Subscription_Length_Months	Monthly_Bill	Total_Usage_GB	Churn
0	1	Customer_1	63	Male	Los Angeles	17	73.36	236	0
1	2	Customer_2	62	Female	New York	1	48.76	172	0
2	3	Customer_3	24	Female	Los Angeles	5	85.47	460	0
3	4	Customer_4	36	Female	Miami	3	97.94	297	1
4	5	Customer_5	46	Female	Miami	19	58.14	266	0

Data Encoding

```
In [3]: 1 # Ecoded 0 for Femal and 1 for Male
2 df["Gender"] = df["Gender"].apply(lambda x: 0 if x == "Female" else 1)
3 df.sample(5)
```

Out[3]:

	CustomerID	Name	Age	Gender	Location	Subscription_Length_Months	Monthly_Bill	Total_Usage_GB	Churn
86607	86608	Customer_86608	56	1	Chicago	21	55.91	83	1
68308	68309	Customer_68309	47	1	Los Angeles	13	65.99	305	1
54253	54254	Customer_54254	50	1	Los Angeles	14	95.26	217	1
96836	96837	Customer_96837	47	0	New York	4	41.45	213	1
56035	56036	Customer_56036	32	0	New York	12	71.66	65	1

```
In [4]: 1 # Encoding categorial variable ie.Location
2 from sklearn.preprocessing import OrdinalEncoder
3 OE = OrdinalEncoder()
4
5
6 df["Location"] = OE.fit_transform(df[["Location"]])
7 Loc = OE.categories_
8 Loc
```

Out[4]: array(['Chicago', 'Houston', 'Los Angeles', 'Miami', 'New York'],
dtype=object))

```
In [5]: 1 df.sample(5)
```

Out[5]:

	CustomerID	Name	Age	Gender	Location	Subscription_Length_Months	Monthly_Bill	Total_Usage_GB	Churn
74315	74316	Customer_74316	69	0	4.0	22	73.62	324	0
85621	85622	Customer_85622	37	1	0.0	11	73.83	454	1
50514	50515	Customer_50515	28	0	2.0	11	93.79	364	1
17853	17854	Customer_17854	52	1	4.0	13	40.90	85	1
50203	50204	Customer_50204	46	1	4.0	14	58.18	93	1

Data Segregation

```
In [6]: 1 X = df.iloc[:,2:-1].values
2 X
```

Out[6]: array([[63. , 1. , 2. , 17. , 73.36, 236.],
[62. , 0. , 4. , 1. , 48.76, 172.],
[24. , 0. , 2. , 5. , 85.47, 460.],
...,
[64. , 1. , 0. , 17. , 96.11, 251.],
[51. , 0. , 4. , 20. , 49.25, 434.],
[27. , 0. , 2. , 19. , 76.57, 173.]])

```
In [7]: 1 Y = df.iloc[:, -1].values
2 Y
```

Out[7]: array([0, 0, 0, ..., 1, 1, 1])

Data Scaling

```
In [8]: 1 from sklearn.preprocessing import StandardScaler
2 SS = StandardScaler()
3 X = SS.fit_transform(X)
4 X
```

Out[8]: array([[1.24167039, 1.00432937, 0.00294695, 0.65111499, 0.41060598,
-0.29428898],
[1.17622625, -0.99568929, 1.41974758, -1.65887854, -0.80537409,
-0.78485174],
[-1.31065114, -0.99568929, 0.00294695, -1.08138015, 1.0092043 ,
1.42268068],
...,
[1.30711454, 1.00432937, -1.41385369, 0.65111499, 1.5351404 ,
-0.17931334],
[0.45634069, -0.99568929, 1.41974758, 1.08423877, -0.78115335,
1.22338955],
[-1.11431871, -0.99568929, 0.00294695, 0.93986418, 0.56927655,
-0.7771867]])

Data Splitting

```
In [9]: 1 X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.2,random_state=1)
2 X_train,X_val,Y_train,Y_val = train_test_split(X_train,Y_train,test_size=0.1,random_state=1)
```

Creating an object of early stopping to save the resources

```
In [24]: 1 ES = EarlyStopping(monitor='val_loss', mode='min', patience=1000)
```

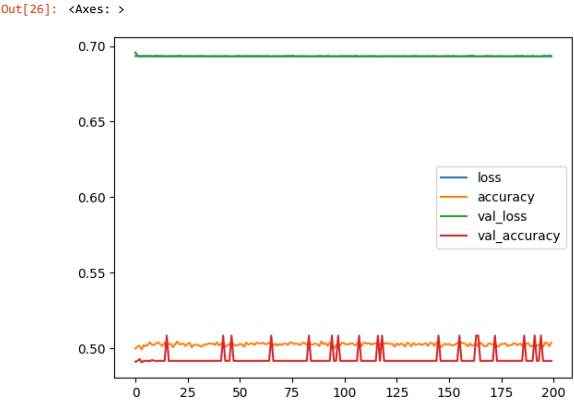
ANN Modelling

```
In [25]: 1 # obj of Sequential
2 ann = Sequential()
3
4 # adding input/hidden/dropout Layers
5 ann.add(Dense(128,activation='relu'))
6 ann.add(Dropout(0.6))
7
8 ann.add(Dense(64,activation='relu'))
9 ann.add(Dropout(0.4))
10
11 ann.add(Dense(48,activation='relu'))
12 ann.add(Dropout(0.4))
13
14 # Output Layers
15 ann.add(Dense(1,activation='sigmoid'))
16
17 # compiling the model
18 ann.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
19
20 # Fitting the model
21 ann.fit(X_train, Y_train, batch_size = 32, epochs = 200, callbacks=ES, validation_data=(X_val,Y_val))
```

Epoch 1/200
2250/2250 [=====] - 10s 4ms/step - loss: 0.6958 - accuracy: 0.4998 - val_loss: 0.6934 - val_accuracy: 0.4911
Epoch 2/200
2250/2250 [=====] - 8s 4ms/step - loss: 0.6935 - accuracy: 0.5012 - val_loss: 0.6933 - val_accuracy: 0.4915
Epoch 3/200
2250/2250 [=====] - 8s 3ms/step - loss: 0.6933 - accuracy: 0.5016 - val_loss: 0.6932 - val_accuracy: 0.4929
Epoch 4/200
2250/2250 [=====] - 8s 4ms/step - loss: 0.6933 - accuracy: 0.4991 - val_loss: 0.6932 - val_accuracy: 0.4906
Epoch 5/200
2250/2250 [=====] - 7s 3ms/step - loss: 0.6933 - accuracy: 0.5021 - val_loss: 0.6935 - val_accuracy: 0.4916
Epoch 6/200
2250/2250 [=====] - 8s 3ms/step - loss: 0.6932 - accuracy: 0.5013 - val_loss: 0.6934 - val_accuracy: 0.4916
Epoch 7/200
2250/2250 [=====] - 8s 3ms/step - loss: 0.6933 - accuracy: 0.5026 - val_loss: 0.6933 - val_accuracy: 0.4916
Epoch 8/200
2250/2250 [=====] - 7s 3ms/step - loss: 0.6932 - accuracy: 0.5039 - val_loss: 0.6936 - val_accuracy: 0.4915
Epoch 9/200
2250/2250 [=====] - 8s 3ms/step - loss: 0.6933 - accuracy: 0.5022 - val_loss: 0.6932 - val_accuracy: 0.4922
Epoch 10/200

Visualization of Loss/Accuracy w.r.t Epochs

```
In [26]: 1 # as the loss and accuracy goes parallelly they will take muchmore time to converge with each other
2 lossdf = pd.DataFrame(ann.history.history)
3 lossdf.plot()
```



```
In [27]: 1 Y_pred = ann.predict(X_test)

625/625 [=====] - 1s 1ms/step
```

```
In [28]: 1 # Set Threshold
2 Y_pred = np.where(Y_pred>0.5,1,0)
```

```
In [29]: 1 Y_pred
```

Out[29]: array([[0],
[0],
[0],
...,
[0],
[0],
[0]])

Classification report

```
In [30]: 1 # evaluation
2 from sklearn.metrics import classification_report
3 print( classification_report(Y_test,Y_pred) )
```

	precision	recall	f1-score	support
0	0.50	1.00	0.67	10020
1	0.00	0.00	0.00	9980
accuracy			0.50	20000
macro avg	0.25	0.50	0.33	20000
weighted avg	0.25	0.50	0.33	20000

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

```
In [ ]: 1
```