# **DNN- Experiment 3**

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### **Dataset - Churn Dataset.**

```
In [11]: import pandas as pd
          from sklearn.model_selection import train_test_split
In [12]: churn_df=pd.read_csv(r'D:/Churn Dataset.csv')
In [13]: X = pd.get_dummies(churn_df.drop(['Churn','customerID'],axis=1))
          y = churn_df['Churn'].apply(lambda x: 1 if x == 'Yes' else 0)
In [14]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=.2)
In [15]: X_train.head()
Out[15]:
               SeniorCitizen tenure MonthlyCharges gender_Female gender_Male Partner_No Partner_Yes Dependents_No Dependents_Yes PhoneService_No ... TotalCharges_
          4602
                         0
                                67
                                                                           0
                                                                                                 0
                                                                                                                1
                                                                                                                               0
                                                                                                                                               0 ...
                                             26.10
                                                                                      1
                                                                                                                                               0 ...
          5398
                         0
                                 2
                                             19.75
                                                                                                 0
                                                                                                                               0
                                                                                      0
                                                                                                                0
                                                                                                                                               0 ...
          539
                         0
                                60
                                             80.60
                                                                                                 1
                                                                                                                               1
          4456
                                68
                                             24.95
                                                                                                                                               0 ...
                                69
                                            105.00
                                                                                                                                               0 ...
          3813
         5 rows × 6575 columns
```

# **Dependencies**

```
In [16]: from tensorflow.keras.models import Sequential, load_model
    from tensorflow.keras.layers import Dense
    from sklearn.metrics import accuracy_score
```

# **Build & Compile model**

```
In [17]: model = Sequential()
    model.add(Dense(units=32, activation ='relu', input_dim =len(X_train.columns)))
    model.add(Dense(units=64, activation ='relu'))
    model.add(Dense(units=1, activation ='sigmoid'))
In [18]: model.compile(loss = 'binary_crossentropy', optimizer = 'sgd', metrics ='accuracy')
```

# Fit, Predict and Evaluate

```
In [19]: model.fit(X_train, y_train, epochs =200, batch_size = 32)
```

```
Epoch 1/200
Epoch 2/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
Epoch 6/200
177/177 [=========== ] - 0s 3ms/step - loss: 0.4736 - accuracy: 0.7806
Epoch 7/200
177/177 [============ ] - 1s 3ms/step - loss: 0.4734 - accuracy: 0.7767
Epoch 8/200
Epoch 9/200
Epoch 10/200
Epoch 11/200
Epoch 12/200
Epoch 13/200
Epoch 14/200
Epoch 15/200
Epoch 16/200
177/177 [============ ] - 0s 3ms/step - loss: 0.4642 - accuracy: 0.7799
Epoch 17/200
Epoch 18/200
Epoch 19/200
Epoch 20/200
Epoch 21/200
Epoch 22/200
Epoch 23/200
Epoch 24/200
Epoch 25/200
Epoch 26/200
Epoch 27/200
Epoch 28/200
Epoch 29/200
Epoch 30/200
Epoch 31/200
Epoch 32/200
Epoch 33/200
Epoch 34/200
Epoch 35/200
177/177 [============== ] - 1s 3ms/step - loss: 0.4497 - accuracy: 0.7856
Epoch 36/200
Epoch 37/200
Epoch 38/200
Epoch 39/200
Epoch 40/200
177/177 [============ - 1s 3ms/step - loss: 0.4517 - accuracy: 0.7829
Epoch 41/200
Epoch 42/200
Epoch 43/200
Epoch 44/200
177/177 [=========== - 1s 3ms/step - loss: 0.4468 - accuracy: 0.7842
Epoch 45/200
Epoch 46/200
177/177 [=========== - 1s 3ms/step - loss: 0.4456 - accuracy: 0.7884
Epoch 47/200
177/177 [===========] - 0s 3ms/step - loss: 0.4481 - accuracy: 0.7787
Epoch 48/200
Epoch 49/200
Epoch 50/200
177/177 [============= - 1s 3ms/step - loss: 0.4449 - accuracy: 0.7824
Epoch 51/200
Epoch 52/200
Epoch 53/200
Epoch 54/200
177/177 [=========== - 1s 3ms/step - loss: 0.4417 - accuracy: 0.7918
Epoch 55/200
Epoch 56/200
```

```
Epoch 57/200
Epoch 58/200
Epoch 59/200
Epoch 60/200
Epoch 61/200
Epoch 62/200
177/177 [=========== ] - 1s 3ms/step - loss: 0.4448 - accuracy: 0.7833
Epoch 63/200
177/177 [============ ] - 1s 3ms/step - loss: 0.4432 - accuracy: 0.7900
Epoch 64/200
Epoch 65/200
Epoch 66/200
Epoch 67/200
Epoch 68/200
Epoch 69/200
Epoch 70/200
Epoch 71/200
Epoch 72/200
177/177 [============ ] - 0s 3ms/step - loss: 0.4406 - accuracy: 0.7911
Epoch 73/200
Epoch 74/200
Epoch 75/200
177/177 [============= - 1s 3ms/step - loss: 0.4406 - accuracy: 0.7923
Epoch 76/200
Epoch 77/200
Epoch 78/200
Epoch 79/200
177/177 [=========== - 1s 3ms/step - loss: 0.4386 - accuracy: 0.7893
Fnoch 80/200
177/177 [=========== - 0s 3ms/step - loss: 0.4378 - accuracy: 0.7870
Epoch 81/200
Epoch 82/200
Epoch 83/200
Epoch 84/200
Epoch 85/200
Epoch 86/200
177/177 [============= - 0s 3ms/step - loss: 0.4358 - accuracy: 0.7914
Epoch 87/200
Epoch 88/200
Epoch 89/200
Epoch 90/200
Epoch 91/200
177/177 [============== ] - 0s 3ms/step - loss: 0.4365 - accuracy: 0.7879
Epoch 92/200
Epoch 93/200
Epoch 94/200
Epoch 95/200
Epoch 96/200
177/177 [============ ] - 0s 3ms/step - loss: 0.4362 - accuracy: 0.7900
Epoch 97/200
Epoch 98/200
Epoch 99/200
Epoch 100/200
177/177 [============ - 1s 3ms/step - loss: 0.4387 - accuracy: 0.7900
Epoch 101/200
Epoch 102/200
Epoch 103/200
177/177 [===========] - 1s 3ms/step - loss: 0.4327 - accuracy: 0.7943
Epoch 104/200
Epoch 105/200
177/177 [============ - 1s 3ms/step - loss: 0.4327 - accuracy: 0.7929
Epoch 106/200
177/177 [============ - 0s 3ms/step - loss: 0.4356 - accuracy: 0.7879
Epoch 107/200
Epoch 108/200
Epoch 109/200
Epoch 110/200
Epoch 111/200
Epoch 112/200
```

```
Epoch 113/200
Epoch 114/200
Epoch 115/200
Epoch 116/200
Epoch 117/200
Epoch 118/200
Epoch 119/200
Epoch 120/200
Epoch 121/200
Epoch 122/200
Epoch 123/200
Epoch 124/200
Epoch 125/200
Epoch 126/200
Epoch 127/200
Epoch 128/200
177/177 [============ ] - 0s 3ms/step - loss: 0.4344 - accuracy: 0.7918
Epoch 129/200
Epoch 130/200
Epoch 131/200
Epoch 132/200
177/177 [============= - 0s 3ms/step - loss: 0.4301 - accuracy: 0.7966
Epoch 133/200
Epoch 134/200
Epoch 135/200
Epoch 136/200
Epoch 137/200
Epoch 138/200
Epoch 139/200
Epoch 140/200
Epoch 141/200
Epoch 142/200
Epoch 143/200
177/177 [============= ] - 1s 3ms/step - loss: 0.4305 - accuracy: 0.7929
Epoch 144/200
Epoch 145/200
Epoch 146/200
Epoch 147/200
177/177 [============== ] - 1s 3ms/step - loss: 0.4288 - accuracy: 0.7936
Epoch 148/200
Epoch 149/200
Epoch 150/200
Epoch 151/200
177/177 [=========== ] - 0s 3ms/step - loss: 0.4287 - accuracy: 0.7969
Epoch 152/200
177/177 [============ ] - 0s 3ms/step - loss: 0.4283 - accuracy: 0.7964
Epoch 153/200
177/177 [============ ] - 1s 3ms/step - loss: 0.4325 - accuracy: 0.7923
Epoch 154/200
Epoch 155/200
Epoch 156/200
177/177 [============ - 0s 3ms/step - loss: 0.4312 - accuracy: 0.7955
Epoch 157/200
Epoch 158/200
Epoch 159/200
Epoch 160/200
Epoch 161/200
Epoch 162/200
177/177 [================= ] - 0s 3ms/step - loss: 0.4280 - accuracy: 0.7991
Epoch 163/200
Epoch 164/200
Epoch 165/200
Epoch 166/200
Epoch 167/200
Epoch 168/200
```

```
Epoch 169/200
  177/177 [============================ - 0s 3ms/step - loss: 0.4302 - accuracy: 0.7996
  Epoch 170/200
  177/177 [============================= - 0s 3ms/step - loss: 0.4257 - accuracy: 0.7966
  Epoch 171/200
  Epoch 172/200
  Epoch 173/200
  Epoch 174/200
  177/177 [=========== ] - 0s 3ms/step - loss: 0.4295 - accuracy: 0.7955
  Epoch 175/200
  Epoch 176/200
  177/177 [============ ] - 0s 3ms/step - loss: 0.4264 - accuracy: 0.8001
  Epoch 177/200
  177/177 [===========] - 1s 3ms/step - loss: 0.4280 - accuracy: 0.7977
  Epoch 178/200
  177/177 [============================= - 0s 3ms/step - loss: 0.4280 - accuracy: 0.7962
  Epoch 179/200
  177/177 [================= ] - 1s 3ms/step - loss: 0.4294 - accuracy: 0.7941
  Epoch 180/200
  Epoch 181/200
  Epoch 182/200
  Epoch 183/200
  177/177 [=========== ] - 1s 3ms/step - loss: 0.4299 - accuracy: 0.7968
  Epoch 184/200
  177/177 [=========== ] - 0s 3ms/step - loss: 0.4264 - accuracy: 0.7938
  Epoch 185/200
  177/177 [============ ] - 1s 3ms/step - loss: 0.4284 - accuracy: 0.7962
  Epoch 186/200
  Epoch 187/200
  177/177 [============ ] - 1s 3ms/step - loss: 0.4334 - accuracy: 0.7904
  Epoch 188/200
  177/177 [================== ] - 0s 3ms/step - loss: 0.4263 - accuracy: 0.7975
  Epoch 189/200
  Epoch 190/200
  177/177 [============] - 1s 3ms/step - loss: 0.4282 - accuracy: 0.7975
  Epoch 191/200
  Epoch 192/200
  177/177 [============================= - 0s 3ms/step - loss: 0.4262 - accuracy: 0.8007
  Epoch 193/200
  Epoch 194/200
  Epoch 195/200
  Epoch 196/200
  Epoch 197/200
  177/177 [============] - 1s 3ms/step - loss: 0.4274 - accuracy: 0.7991
  Epoch 198/200
  Epoch 199/200
  Epoch 200/200
  Out[19]: <keras.callbacks.History at 0x23a800969d0>
In [20]: y_hat = model.predict(X_test)
  y_hat = [0 if val <0.5 else 1 for val in y_hat]</pre>
  45/45 [============ ] - 0s 3ms/step
In [21]: print(y_hat)
  0, 0, 0,
  0, 0, 0,
```

0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0]