

Churn model

October 30, 2023

```
[1]: import pandas as pd
      from sklearn.model_selection import train_test_split
```

```
[3]: data=pd.read_csv("churn.csv")
```

```
[4]: data.head()
```

```
[4]:  customerID  gender  SeniorCitizen  Partner  Dependents  tenure  PhoneService  \
0  7590-VHVEG  Female                0      Yes           No         1           No
1  5575-GNVDE   Male                0      No            No        34           Yes
2  3668-QPYBK   Male                0      No            No         2           Yes
3  7795-CFOCW   Male                0      No            No        45           No
4  9237-HQITU   Female              0      No            No         2           Yes
```

```
      MultipleLines  InternetService  OnlineSecurity  ...  DeviceProtection  \
0  No phone service                DSL              No  ...              No
1                No                DSL              Yes  ...              Yes
2                No                DSL              Yes  ...              No
3  No phone service                DSL              Yes  ...              Yes
4                No      Fiber optic              No  ...              No
```

```
      TechSupport  StreamingTV  StreamingMovies      Contract  PaperlessBilling  \
0                No           No              No  Month-to-month              Yes
1                No           No              No      One year              No
2                No           No              No  Month-to-month              Yes
3                Yes           No              No      One year              No
4                No           No              No  Month-to-month              Yes
```

```
      PaymentMethod  MonthlyCharges  TotalCharges  Churn
0      Electronic check           29.85          29.85   No
1      Mailed check           56.95         1889.5   No
2      Mailed check           53.85          108.15  Yes
3  Bank transfer (automatic)         42.30         1840.75   No
4      Electronic check           70.70          151.65  Yes
```

[5 rows x 21 columns]

```
[11]: x=pd.get_dummies(data.drop(['Churn','customerID'],axis=1))
      y=data['Churn'].apply(lambda x: 1 if x=='Yes'else 0)
```

```
[15]: y.head()
```

```
[15]: 0    0
      1    0
      2    1
      3    0
      4    1
      Name: Churn, dtype: int64
```

```
[16]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)
```

```
[17]: x_train.head()
```

```
[17]:      SeniorCitizen  tenure  MonthlyCharges  gender_Female  gender_Male  \
6768              0      72          117.50              1              0
924               1      22           81.70              1              0
3668              0      22           59.75              0              1
6016              0      68          108.45              0              1
2857              0      47           63.80              1              0

      Partner_No  Partner_Yes  Dependents_No  Dependents_Yes  PhoneService_No  \
6768           0           1              0              1              0
924            1           0              1              0              0
3668           0           1              1              0              0
6016           1           0              1              0              0
2857           0           1              0              1              0

      ...  TotalCharges_995.35  TotalCharges_996.45  TotalCharges_996.85  \
6768  ...                   0                   0                   0
924   ...                   0                   0                   0
3668  ...                   0                   0                   0
6016  ...                   0                   0                   0
2857  ...                   0                   0                   0

      TotalCharges_996.95  TotalCharges_997.65  TotalCharges_997.75  \
6768                    0                    0                    0
924                     0                    0                    0
3668                    0                    0                    0
6016                    0                    0                    0
2857                    0                    0                    0

      TotalCharges_998.1  TotalCharges_999.45  TotalCharges_999.8  \
6768                   0                   0                   0
924                    0                   0                   0
```

3668	0	0	0
6016	0	0	0
2857	0	0	0

	TotalCharges_999.9
6768	0
924	0
3668	0
6016	0
2857	0

[5 rows x 6575 columns]

```
[18]: y_train.head()
```

```
[18]: 6768    0
      924    0
      3668   0
      6016   1
      2857   0
      Name: Churn, dtype: int64
```

```
[19]: from tensorflow.keras.models import Sequential,load_model
      from tensorflow.keras.layers import Dense # Weighted sum and activation
      ↪function
      from sklearn.metrics import accuracy_score
```

Sequential is a type of model in TensorFlow used to create a linear stack of layers for building neural networks.

load_model is a function that allows you to load a pre-trained model from a file

The Dense layer is used to create fully connected layers in a neural network, where every neuron is connected to every neuron in the previous and subsequent layers.

```
[20]: 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')) #output

      # activation='relu': The activation function for this layer is
      # Rectified Linear Unit (ReLU), which is commonly used in hidden layers
      # of neural networks. It introduces non-linearity by outputting the input
      # if it's positive and 0 if it's negative.
```

```
[21]: model.compile(loss='binary_crossentropy',optimizer='sgd',metrics='accuracy')
```

```
[30]: model.fit(x_train,y_train,epochs=300,batch_size=32)
```

Epoch 1/300

```

177/177 [=====] - 1s 3ms/step - loss: 0.4164 -
accuracy: 0.8058
Epoch 2/300
177/177 [=====] - 1s 3ms/step - loss: 0.4185 -
accuracy: 0.8048
Epoch 3/300
177/177 [=====] - 1s 3ms/step - loss: 0.4171 -
accuracy: 0.7998
Epoch 4/300
177/177 [=====] - 1s 3ms/step - loss: 0.4178 -
accuracy: 0.8023
Epoch 5/300
177/177 [=====] - 1s 3ms/step - loss: 0.4166 -
accuracy: 0.8028
Epoch 6/300
177/177 [=====] - 1s 3ms/step - loss: 0.4171 -
accuracy: 0.8049
Epoch 7/300
177/177 [=====] - 1s 3ms/step - loss: 0.4152 -
accuracy: 0.8088
Epoch 8/300
177/177 [=====] - 1s 3ms/step - loss: 0.4146 -
accuracy: 0.8040
Epoch 9/300
177/177 [=====] - 1s 3ms/step - loss: 0.4180 -
accuracy: 0.8056
Epoch 10/300
177/177 [=====] - 1s 3ms/step - loss: 0.4161 -
accuracy: 0.8051
Epoch 11/300
177/177 [=====] - 1s 3ms/step - loss: 0.4169 -
accuracy: 0.8055
Epoch 12/300
177/177 [=====] - 1s 3ms/step - loss: 0.4222 -
accuracy: 0.8019
Epoch 13/300
177/177 [=====] - 1s 3ms/step - loss: 0.4183 -
accuracy: 0.8035
Epoch 14/300
177/177 [=====] - 1s 3ms/step - loss: 0.4190 -
accuracy: 0.8012
Epoch 15/300
177/177 [=====] - 1s 3ms/step - loss: 0.4154 -
accuracy: 0.8016
Epoch 16/300
177/177 [=====] - 1s 3ms/step - loss: 0.4121 -
accuracy: 0.8049
Epoch 17/300

```

```

177/177 [=====] - 1s 3ms/step - loss: 0.4158 -
accuracy: 0.8009
Epoch 18/300
177/177 [=====] - 1s 3ms/step - loss: 0.4145 -
accuracy: 0.8053
Epoch 19/300
177/177 [=====] - 1s 3ms/step - loss: 0.4161 -
accuracy: 0.8053
Epoch 20/300
177/177 [=====] - 1s 3ms/step - loss: 0.4175 -
accuracy: 0.8024
Epoch 21/300
177/177 [=====] - 1s 3ms/step - loss: 0.4167 -
accuracy: 0.8042
Epoch 22/300
177/177 [=====] - 1s 3ms/step - loss: 0.4150 -
accuracy: 0.8035
Epoch 23/300
177/177 [=====] - 1s 3ms/step - loss: 0.4165 -
accuracy: 0.8046
Epoch 24/300
177/177 [=====] - 1s 3ms/step - loss: 0.4132 -
accuracy: 0.8055
Epoch 25/300
177/177 [=====] - 1s 3ms/step - loss: 0.4140 -
accuracy: 0.8069
Epoch 26/300
177/177 [=====] - 1s 3ms/step - loss: 0.4178 -
accuracy: 0.7996
Epoch 27/300
177/177 [=====] - 1s 3ms/step - loss: 0.4157 -
accuracy: 0.8062
Epoch 28/300
177/177 [=====] - 1s 3ms/step - loss: 0.4139 -
accuracy: 0.8083
Epoch 29/300
177/177 [=====] - 1s 3ms/step - loss: 0.4140 -
accuracy: 0.8051
Epoch 30/300
177/177 [=====] - 1s 3ms/step - loss: 0.4137 -
accuracy: 0.8067
Epoch 31/300
177/177 [=====] - 1s 3ms/step - loss: 0.4210 -
accuracy: 0.8010
Epoch 32/300
177/177 [=====] - 1s 3ms/step - loss: 0.4181 -
accuracy: 0.8007
Epoch 33/300

```

```

177/177 [=====] - 1s 4ms/step - loss: 0.4146 -
accuracy: 0.8049
Epoch 34/300
177/177 [=====] - 1s 3ms/step - loss: 0.4165 -
accuracy: 0.8032
Epoch 35/300
177/177 [=====] - 1s 3ms/step - loss: 0.4169 -
accuracy: 0.8035
Epoch 36/300
177/177 [=====] - 1s 4ms/step - loss: 0.4168 -
accuracy: 0.8019
Epoch 37/300
177/177 [=====] - 1s 3ms/step - loss: 0.4160 -
accuracy: 0.8039
Epoch 38/300
177/177 [=====] - 1s 3ms/step - loss: 0.4130 -
accuracy: 0.8081
Epoch 39/300
177/177 [=====] - 1s 3ms/step - loss: 0.4143 -
accuracy: 0.8051
Epoch 40/300
177/177 [=====] - 1s 3ms/step - loss: 0.4158 -
accuracy: 0.8003
Epoch 41/300
177/177 [=====] - 1s 3ms/step - loss: 0.4160 -
accuracy: 0.8039
Epoch 42/300
177/177 [=====] - 1s 3ms/step - loss: 0.4152 -
accuracy: 0.8040
Epoch 43/300
177/177 [=====] - 1s 3ms/step - loss: 0.4153 -
accuracy: 0.8056
Epoch 44/300
177/177 [=====] - 1s 3ms/step - loss: 0.4167 -
accuracy: 0.8056
Epoch 45/300
177/177 [=====] - 1s 3ms/step - loss: 0.4129 -
accuracy: 0.8056
Epoch 46/300
177/177 [=====] - 1s 3ms/step - loss: 0.4142 -
accuracy: 0.8055
Epoch 47/300
177/177 [=====] - 1s 3ms/step - loss: 0.4139 -
accuracy: 0.8071
Epoch 48/300
177/177 [=====] - 1s 3ms/step - loss: 0.4154 -
accuracy: 0.8030
Epoch 49/300

```

177/177 [=====] - 1s 3ms/step - loss: 0.4157 -
 accuracy: 0.8064
 Epoch 50/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4146 -
 accuracy: 0.8014
 Epoch 51/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4131 -
 accuracy: 0.8048
 Epoch 52/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4150 -
 accuracy: 0.8033
 Epoch 53/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4125 -
 accuracy: 0.8072
 Epoch 54/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4156 -
 accuracy: 0.8005
 Epoch 55/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4173 -
 accuracy: 0.8035
 Epoch 56/300
 177/177 [=====] - 1s 4ms/step - loss: 0.4155 -
 accuracy: 0.8074
 Epoch 57/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4250 -
 accuracy: 0.7930
 Epoch 58/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4137 -
 accuracy: 0.8033
 Epoch 59/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4128 -
 accuracy: 0.8051
 Epoch 60/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4135 -
 accuracy: 0.8081
 Epoch 61/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4115 -
 accuracy: 0.8039
 Epoch 62/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4123 -
 accuracy: 0.8055
 Epoch 63/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4141 -
 accuracy: 0.8049
 Epoch 64/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4140 -
 accuracy: 0.8062
 Epoch 65/300

```

177/177 [=====] - 1s 3ms/step - loss: 0.4154 -
accuracy: 0.8051
Epoch 66/300
177/177 [=====] - 1s 3ms/step - loss: 0.4132 -
accuracy: 0.8076
Epoch 67/300
177/177 [=====] - 1s 3ms/step - loss: 0.4158 -
accuracy: 0.8003
Epoch 68/300
177/177 [=====] - 1s 3ms/step - loss: 0.4175 -
accuracy: 0.8014
Epoch 69/300
177/177 [=====] - 1s 3ms/step - loss: 0.4137 -
accuracy: 0.8062
Epoch 70/300
177/177 [=====] - 1s 3ms/step - loss: 0.4146 -
accuracy: 0.8042
Epoch 71/300
177/177 [=====] - 1s 3ms/step - loss: 0.4131 -
accuracy: 0.8032
Epoch 72/300
177/177 [=====] - 1s 3ms/step - loss: 0.4141 -
accuracy: 0.8044
Epoch 73/300
177/177 [=====] - 1s 3ms/step - loss: 0.4152 -
accuracy: 0.8032
Epoch 74/300
177/177 [=====] - 1s 3ms/step - loss: 0.4142 -
accuracy: 0.8064
Epoch 75/300
177/177 [=====] - 1s 3ms/step - loss: 0.4142 -
accuracy: 0.7994
Epoch 76/300
177/177 [=====] - 1s 3ms/step - loss: 0.4146 -
accuracy: 0.8039
Epoch 77/300
177/177 [=====] - 1s 3ms/step - loss: 0.4188 -
accuracy: 0.7996
Epoch 78/300
177/177 [=====] - 1s 3ms/step - loss: 0.4149 -
accuracy: 0.8024
Epoch 79/300
177/177 [=====] - 1s 3ms/step - loss: 0.4123 -
accuracy: 0.8076
Epoch 80/300
177/177 [=====] - 1s 3ms/step - loss: 0.4175 -
accuracy: 0.8021
Epoch 81/300

```


177/177 [=====] - 1s 3ms/step - loss: 0.4144 -
 accuracy: 0.8078
 Epoch 82/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4148 -
 accuracy: 0.8033
 Epoch 83/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4118 -
 accuracy: 0.8046
 Epoch 84/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4127 -
 accuracy: 0.8071
 Epoch 85/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4140 -
 accuracy: 0.8049
 Epoch 86/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4132 -
 accuracy: 0.8076
 Epoch 87/300
 177/177 [=====] - 1s 4ms/step - loss: 0.4148 -
 accuracy: 0.8032
 Epoch 88/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4153 -
 accuracy: 0.8064
 Epoch 89/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4121 -
 accuracy: 0.8062
 Epoch 90/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4144 -
 accuracy: 0.8049
 Epoch 91/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4105 -
 accuracy: 0.8069
 Epoch 92/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4138 -
 accuracy: 0.8062
 Epoch 93/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4198 -
 accuracy: 0.8035
 Epoch 94/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4156 -
 accuracy: 0.8071
 Epoch 95/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4121 -
 accuracy: 0.8076
 Epoch 96/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4120 -
 accuracy: 0.8058
 Epoch 97/300

177/177 [=====] - 1s 3ms/step - loss: 0.4129 -
accuracy: 0.8053
Epoch 98/300
177/177 [=====] - 1s 3ms/step - loss: 0.4110 -
accuracy: 0.8055
Epoch 99/300
177/177 [=====] - 1s 3ms/step - loss: 0.4121 -
accuracy: 0.8055
Epoch 100/300
177/177 [=====] - 1s 3ms/step - loss: 0.4142 -
accuracy: 0.8064
Epoch 101/300
177/177 [=====] - 1s 4ms/step - loss: 0.4138 -
accuracy: 0.8046
Epoch 102/300
177/177 [=====] - 1s 4ms/step - loss: 0.4114 -
accuracy: 0.8080
Epoch 103/300
177/177 [=====] - 1s 3ms/step - loss: 0.4138 -
accuracy: 0.8023
Epoch 104/300
177/177 [=====] - 1s 4ms/step - loss: 0.4128 -
accuracy: 0.8053
Epoch 105/300
177/177 [=====] - 1s 4ms/step - loss: 0.4135 -
accuracy: 0.8069
Epoch 106/300
177/177 [=====] - 1s 4ms/step - loss: 0.4129 -
accuracy: 0.8049
Epoch 107/300
177/177 [=====] - 1s 3ms/step - loss: 0.4138 -
accuracy: 0.8044
Epoch 108/300
177/177 [=====] - 1s 4ms/step - loss: 0.4118 -
accuracy: 0.8065
Epoch 109/300
177/177 [=====] - 1s 3ms/step - loss: 0.4113 -
accuracy: 0.8069
Epoch 110/300
177/177 [=====] - 1s 4ms/step - loss: 0.4121 -
accuracy: 0.8085
Epoch 111/300
177/177 [=====] - 1s 4ms/step - loss: 0.4122 -
accuracy: 0.8067
Epoch 112/300
177/177 [=====] - 1s 3ms/step - loss: 0.4104 -
accuracy: 0.8072
Epoch 113/300

177/177 [=====] - 1s 3ms/step - loss: 0.4123 -
 accuracy: 0.8049
 Epoch 114/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4106 -
 accuracy: 0.8071
 Epoch 115/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4120 -
 accuracy: 0.8064
 Epoch 116/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4117 -
 accuracy: 0.8030
 Epoch 117/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4116 -
 accuracy: 0.8051
 Epoch 118/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4116 -
 accuracy: 0.8049
 Epoch 119/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4123 -
 accuracy: 0.8019
 Epoch 120/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4126 -
 accuracy: 0.8033
 Epoch 121/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4166 -
 accuracy: 0.8032
 Epoch 122/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4121 -
 accuracy: 0.8087
 Epoch 123/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4112 -
 accuracy: 0.8040
 Epoch 124/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4080 -
 accuracy: 0.8083
 Epoch 125/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4128 -
 accuracy: 0.8051
 Epoch 126/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4117 -
 accuracy: 0.8044
 Epoch 127/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4124 -
 accuracy: 0.8024
 Epoch 128/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4126 -
 accuracy: 0.8046
 Epoch 129/300

177/177 [=====] - 1s 3ms/step - loss: 0.4114 -
accuracy: 0.8069
Epoch 130/300
177/177 [=====] - 1s 3ms/step - loss: 0.4127 -
accuracy: 0.8078
Epoch 131/300
177/177 [=====] - 1s 3ms/step - loss: 0.4114 -
accuracy: 0.8080
Epoch 132/300
177/177 [=====] - 1s 3ms/step - loss: 0.4118 -
accuracy: 0.8021
Epoch 133/300
177/177 [=====] - 1s 3ms/step - loss: 0.4131 -
accuracy: 0.8046
Epoch 134/300
177/177 [=====] - 1s 3ms/step - loss: 0.4100 -
accuracy: 0.8044
Epoch 135/300
177/177 [=====] - 1s 3ms/step - loss: 0.4102 -
accuracy: 0.8076
Epoch 136/300
177/177 [=====] - 1s 4ms/step - loss: 0.4111 -
accuracy: 0.8076
Epoch 137/300
177/177 [=====] - 1s 4ms/step - loss: 0.4112 -
accuracy: 0.8056
Epoch 138/300
177/177 [=====] - 1s 4ms/step - loss: 0.4144 -
accuracy: 0.8033
Epoch 139/300
177/177 [=====] - 1s 3ms/step - loss: 0.4097 -
accuracy: 0.8051
Epoch 140/300
177/177 [=====] - 1s 3ms/step - loss: 0.4115 -
accuracy: 0.8056
Epoch 141/300
177/177 [=====] - 1s 3ms/step - loss: 0.4125 -
accuracy: 0.8060
Epoch 142/300
177/177 [=====] - 1s 3ms/step - loss: 0.4113 -
accuracy: 0.8051
Epoch 143/300
177/177 [=====] - 1s 3ms/step - loss: 0.4104 -
accuracy: 0.8055
Epoch 144/300
177/177 [=====] - 1s 3ms/step - loss: 0.4081 -
accuracy: 0.8055
Epoch 145/300

177/177 [=====] - 1s 3ms/step - loss: 0.4118 -
 accuracy: 0.8071
 Epoch 146/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4095 -
 accuracy: 0.8104
 Epoch 147/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4109 -
 accuracy: 0.8087
 Epoch 148/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4100 -
 accuracy: 0.8060
 Epoch 149/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4127 -
 accuracy: 0.8035
 Epoch 150/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4123 -
 accuracy: 0.8033
 Epoch 151/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4137 -
 accuracy: 0.8053
 Epoch 152/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4117 -
 accuracy: 0.8039
 Epoch 153/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4091 -
 accuracy: 0.8058
 Epoch 154/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4085 -
 accuracy: 0.8078
 Epoch 155/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4083 -
 accuracy: 0.8081
 Epoch 156/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4116 -
 accuracy: 0.8064
 Epoch 157/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4081 -
 accuracy: 0.8078
 Epoch 158/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4093 -
 accuracy: 0.8076
 Epoch 159/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4103 -
 accuracy: 0.8072
 Epoch 160/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4103 -
 accuracy: 0.8037
 Epoch 161/300

177/177 [=====] - 1s 3ms/step - loss: 0.4109 -
 accuracy: 0.8067
 Epoch 162/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4090 -
 accuracy: 0.8067
 Epoch 163/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4105 -
 accuracy: 0.8064
 Epoch 164/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4093 -
 accuracy: 0.8044
 Epoch 165/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4122 -
 accuracy: 0.8048
 Epoch 166/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4103 -
 accuracy: 0.8064
 Epoch 167/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4100 -
 accuracy: 0.8088
 Epoch 168/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4102 -
 accuracy: 0.8051
 Epoch 169/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4098 -
 accuracy: 0.8051
 Epoch 170/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4099 -
 accuracy: 0.8035
 Epoch 171/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4073 -
 accuracy: 0.8072
 Epoch 172/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4112 -
 accuracy: 0.8071
 Epoch 173/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4097 -
 accuracy: 0.8053
 Epoch 174/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4118 -
 accuracy: 0.8060
 Epoch 175/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4107 -
 accuracy: 0.8058
 Epoch 176/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4125 -
 accuracy: 0.8046
 Epoch 177/300

177/177 [=====] - 1s 3ms/step - loss: 0.4096 -
 accuracy: 0.8078
 Epoch 178/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4125 -
 accuracy: 0.8021
 Epoch 179/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4102 -
 accuracy: 0.8085
 Epoch 180/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4099 -
 accuracy: 0.8053
 Epoch 181/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4101 -
 accuracy: 0.8081
 Epoch 182/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4095 -
 accuracy: 0.8117
 Epoch 183/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4090 -
 accuracy: 0.8072
 Epoch 184/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4094 -
 accuracy: 0.8080
 Epoch 185/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4078 -
 accuracy: 0.8095
 Epoch 186/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4099 -
 accuracy: 0.8080
 Epoch 187/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4089 -
 accuracy: 0.8062
 Epoch 188/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4103 -
 accuracy: 0.8058
 Epoch 189/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4090 -
 accuracy: 0.8044
 Epoch 190/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4107 -
 accuracy: 0.8028
 Epoch 191/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4105 -
 accuracy: 0.8053
 Epoch 192/300
 177/177 [=====] - 1s 3ms/step - loss: 0.4092 -
 accuracy: 0.8080
 Epoch 193/300

177/177 [=====] - 1s 3ms/step - loss: 0.4103 -
accuracy: 0.8053
Epoch 194/300
177/177 [=====] - 1s 3ms/step - loss: 0.4098 -
accuracy: 0.8076
Epoch 195/300
177/177 [=====] - 1s 3ms/step - loss: 0.4083 -
accuracy: 0.8103
Epoch 196/300
177/177 [=====] - 1s 3ms/step - loss: 0.4087 -
accuracy: 0.8092
Epoch 197/300
177/177 [=====] - 1s 3ms/step - loss: 0.4090 -
accuracy: 0.8071
Epoch 198/300
177/177 [=====] - 1s 3ms/step - loss: 0.4085 -
accuracy: 0.8065
Epoch 199/300
177/177 [=====] - 1s 3ms/step - loss: 0.4109 -
accuracy: 0.8080
Epoch 200/300
177/177 [=====] - 1s 3ms/step - loss: 0.4086 -
accuracy: 0.8087
Epoch 201/300
177/177 [=====] - 1s 3ms/step - loss: 0.4119 -
accuracy: 0.8062
Epoch 202/300
177/177 [=====] - 1s 3ms/step - loss: 0.4100 -
accuracy: 0.8039
Epoch 203/300
177/177 [=====] - 1s 3ms/step - loss: 0.4093 -
accuracy: 0.8000
Epoch 204/300
177/177 [=====] - 1s 3ms/step - loss: 0.4104 -
accuracy: 0.8053
Epoch 205/300
177/177 [=====] - 1s 3ms/step - loss: 0.4105 -
accuracy: 0.8037
Epoch 206/300
177/177 [=====] - 1s 3ms/step - loss: 0.4104 -
accuracy: 0.8120
Epoch 207/300
177/177 [=====] - 1s 3ms/step - loss: 0.4106 -
accuracy: 0.8032
Epoch 208/300
177/177 [=====] - 1s 3ms/step - loss: 0.4096 -
accuracy: 0.8060
Epoch 209/300

177/177 [=====] - 1s 3ms/step - loss: 0.4078 -
accuracy: 0.8092
Epoch 210/300
177/177 [=====] - 1s 3ms/step - loss: 0.4091 -
accuracy: 0.8088
Epoch 211/300
177/177 [=====] - 1s 3ms/step - loss: 0.4097 -
accuracy: 0.8094
Epoch 212/300
177/177 [=====] - 1s 3ms/step - loss: 0.4159 -
accuracy: 0.8037
Epoch 213/300
177/177 [=====] - 1s 3ms/step - loss: 0.4089 -
accuracy: 0.8074
Epoch 214/300
177/177 [=====] - 1s 3ms/step - loss: 0.4091 -
accuracy: 0.8037
Epoch 215/300
177/177 [=====] - 1s 3ms/step - loss: 0.4078 -
accuracy: 0.8092
Epoch 216/300
177/177 [=====] - 1s 3ms/step - loss: 0.4091 -
accuracy: 0.8021
Epoch 217/300
177/177 [=====] - 1s 3ms/step - loss: 0.4083 -
accuracy: 0.8055
Epoch 218/300
177/177 [=====] - 1s 3ms/step - loss: 0.4086 -
accuracy: 0.8078
Epoch 219/300
177/177 [=====] - 1s 3ms/step - loss: 0.4087 -
accuracy: 0.8108
Epoch 220/300
177/177 [=====] - 1s 3ms/step - loss: 0.4087 -
accuracy: 0.8099
Epoch 221/300
177/177 [=====] - 1s 3ms/step - loss: 0.4099 -
accuracy: 0.8067
Epoch 222/300
177/177 [=====] - 1s 3ms/step - loss: 0.4064 -
accuracy: 0.8083
Epoch 223/300
177/177 [=====] - 1s 3ms/step - loss: 0.4097 -
accuracy: 0.8069
Epoch 224/300
177/177 [=====] - 1s 3ms/step - loss: 0.4090 -
accuracy: 0.8087
Epoch 225/300

177/177 [=====] - 1s 3ms/step - loss: 0.4075 -
accuracy: 0.8080
Epoch 226/300
177/177 [=====] - 1s 3ms/step - loss: 0.4092 -
accuracy: 0.8069
Epoch 227/300
177/177 [=====] - 1s 3ms/step - loss: 0.4071 -
accuracy: 0.8074
Epoch 228/300
177/177 [=====] - 1s 3ms/step - loss: 0.4101 -
accuracy: 0.8037
Epoch 229/300
177/177 [=====] - 1s 3ms/step - loss: 0.4101 -
accuracy: 0.8083
Epoch 230/300
177/177 [=====] - 1s 3ms/step - loss: 0.4086 -
accuracy: 0.8099
Epoch 231/300
177/177 [=====] - 1s 3ms/step - loss: 0.4092 -
accuracy: 0.8053
Epoch 232/300
177/177 [=====] - 1s 3ms/step - loss: 0.4079 -
accuracy: 0.8040
Epoch 233/300
177/177 [=====] - 1s 3ms/step - loss: 0.4077 -
accuracy: 0.8074
Epoch 234/300
177/177 [=====] - 1s 3ms/step - loss: 0.4090 -
accuracy: 0.8048
Epoch 235/300
177/177 [=====] - 1s 3ms/step - loss: 0.4082 -
accuracy: 0.8048
Epoch 236/300
177/177 [=====] - 1s 3ms/step - loss: 0.4093 -
accuracy: 0.8069
Epoch 237/300
177/177 [=====] - 1s 3ms/step - loss: 0.4088 -
accuracy: 0.8053
Epoch 238/300
177/177 [=====] - 1s 3ms/step - loss: 0.4116 -
accuracy: 0.8046
Epoch 239/300
177/177 [=====] - 1s 3ms/step - loss: 0.4075 -
accuracy: 0.8049
Epoch 240/300
177/177 [=====] - 1s 3ms/step - loss: 0.4074 -
accuracy: 0.8074
Epoch 241/300

177/177 [=====] - 1s 3ms/step - loss: 0.4076 -
accuracy: 0.8104
Epoch 242/300
177/177 [=====] - 1s 3ms/step - loss: 0.4089 -
accuracy: 0.8060
Epoch 243/300
177/177 [=====] - 1s 3ms/step - loss: 0.4075 -
accuracy: 0.8080
Epoch 244/300
177/177 [=====] - 1s 3ms/step - loss: 0.4070 -
accuracy: 0.8076
Epoch 245/300
177/177 [=====] - 1s 3ms/step - loss: 0.4072 -
accuracy: 0.8003
Epoch 246/300
177/177 [=====] - 1s 3ms/step - loss: 0.4079 -
accuracy: 0.8074
Epoch 247/300
177/177 [=====] - 1s 3ms/step - loss: 0.4099 -
accuracy: 0.8106
Epoch 248/300
177/177 [=====] - 1s 3ms/step - loss: 0.4073 -
accuracy: 0.8048
Epoch 249/300
177/177 [=====] - 1s 3ms/step - loss: 0.4128 -
accuracy: 0.8048
Epoch 250/300
177/177 [=====] - 1s 3ms/step - loss: 0.4086 -
accuracy: 0.8097
Epoch 251/300
177/177 [=====] - 1s 3ms/step - loss: 0.4082 -
accuracy: 0.8062
Epoch 252/300
177/177 [=====] - 1s 4ms/step - loss: 0.4076 -
accuracy: 0.8067
Epoch 253/300
177/177 [=====] - 1s 3ms/step - loss: 0.4096 -
accuracy: 0.8064
Epoch 254/300
177/177 [=====] - 1s 3ms/step - loss: 0.4068 -
accuracy: 0.8048
Epoch 255/300
177/177 [=====] - 1s 3ms/step - loss: 0.4081 -
accuracy: 0.8097
Epoch 256/300
177/177 [=====] - 1s 4ms/step - loss: 0.4080 -
accuracy: 0.8097
Epoch 257/300

177/177 [=====] - 1s 4ms/step - loss: 0.4087 -
accuracy: 0.8083
Epoch 258/300
177/177 [=====] - 1s 4ms/step - loss: 0.4085 -
accuracy: 0.8065
Epoch 259/300
177/177 [=====] - 1s 3ms/step - loss: 0.4078 -
accuracy: 0.8076
Epoch 260/300
177/177 [=====] - 1s 3ms/step - loss: 0.4100 -
accuracy: 0.8080
Epoch 261/300
177/177 [=====] - 1s 3ms/step - loss: 0.4089 -
accuracy: 0.8065
Epoch 262/300
177/177 [=====] - 1s 3ms/step - loss: 0.4106 -
accuracy: 0.8069
Epoch 263/300
177/177 [=====] - 1s 4ms/step - loss: 0.4072 -
accuracy: 0.8051
Epoch 264/300
177/177 [=====] - 1s 4ms/step - loss: 0.4091 -
accuracy: 0.8062
Epoch 265/300
177/177 [=====] - 1s 3ms/step - loss: 0.4083 -
accuracy: 0.8055
Epoch 266/300
177/177 [=====] - 1s 3ms/step - loss: 0.4110 -
accuracy: 0.8026
Epoch 267/300
177/177 [=====] - 1s 3ms/step - loss: 0.4113 -
accuracy: 0.8064
Epoch 268/300
177/177 [=====] - 1s 4ms/step - loss: 0.4096 -
accuracy: 0.8067
Epoch 269/300
177/177 [=====] - 1s 4ms/step - loss: 0.4111 -
accuracy: 0.8042
Epoch 270/300
177/177 [=====] - 1s 4ms/step - loss: 0.4093 -
accuracy: 0.8080
Epoch 271/300
177/177 [=====] - 1s 4ms/step - loss: 0.4071 -
accuracy: 0.8069
Epoch 272/300
177/177 [=====] - 1s 4ms/step - loss: 0.4086 -
accuracy: 0.8071
Epoch 273/300

177/177 [=====] - 1s 3ms/step - loss: 0.4047 -
accuracy: 0.8076
Epoch 274/300
177/177 [=====] - 1s 4ms/step - loss: 0.4077 -
accuracy: 0.8042
Epoch 275/300
177/177 [=====] - 1s 4ms/step - loss: 0.4092 -
accuracy: 0.8095
Epoch 276/300
177/177 [=====] - 1s 3ms/step - loss: 0.4087 -
accuracy: 0.8033
Epoch 277/300
177/177 [=====] - 1s 3ms/step - loss: 0.4073 -
accuracy: 0.8104
Epoch 278/300
177/177 [=====] - 1s 3ms/step - loss: 0.4107 -
accuracy: 0.8071
Epoch 279/300
177/177 [=====] - 1s 3ms/step - loss: 0.4100 -
accuracy: 0.8080
Epoch 280/300
177/177 [=====] - 1s 3ms/step - loss: 0.4094 -
accuracy: 0.8080
Epoch 281/300
177/177 [=====] - 1s 3ms/step - loss: 0.4080 -
accuracy: 0.8076
Epoch 282/300
177/177 [=====] - 1s 3ms/step - loss: 0.4063 -
accuracy: 0.8087
Epoch 283/300
177/177 [=====] - 1s 3ms/step - loss: 0.4075 -
accuracy: 0.8097
Epoch 284/300
177/177 [=====] - 1s 4ms/step - loss: 0.4095 -
accuracy: 0.8062
Epoch 285/300
177/177 [=====] - 1s 3ms/step - loss: 0.4087 -
accuracy: 0.8056
Epoch 286/300
177/177 [=====] - 1s 3ms/step - loss: 0.4070 -
accuracy: 0.8030
Epoch 287/300
177/177 [=====] - 1s 3ms/step - loss: 0.4069 -
accuracy: 0.8088
Epoch 288/300
177/177 [=====] - 1s 3ms/step - loss: 0.4107 -
accuracy: 0.8065
Epoch 289/300

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177/177 [=====] - 1s 3ms/step - loss: 0.4065 -
accuracy: 0.8049
Epoch 290/300
177/177 [=====] - 1s 3ms/step - loss: 0.4085 -
accuracy: 0.8069
Epoch 291/300
177/177 [=====] - 1s 3ms/step - loss: 0.4059 -
accuracy: 0.8087
Epoch 292/300
177/177 [=====] - 1s 4ms/step - loss: 0.4092 -
accuracy: 0.8090
Epoch 293/300
177/177 [=====] - 1s 3ms/step - loss: 0.4098 -
accuracy: 0.8101
Epoch 294/300
177/177 [=====] - 1s 3ms/step - loss: 0.4076 -
accuracy: 0.8078
Epoch 295/300
177/177 [=====] - 1s 3ms/step - loss: 0.4074 -
accuracy: 0.8067
Epoch 296/300
177/177 [=====] - 1s 3ms/step - loss: 0.4094 -
accuracy: 0.8048
Epoch 297/300
177/177 [=====] - 1s 3ms/step - loss: 0.4075 -
accuracy: 0.8074
Epoch 298/300
177/177 [=====] - 1s 3ms/step - loss: 0.4085 -
accuracy: 0.8037
Epoch 299/300
177/177 [=====] - 1s 3ms/step - loss: 0.4071 -
accuracy: 0.8092
Epoch 300/300
177/177 [=====] - 1s 3ms/step - loss: 0.4065 -
accuracy: 0.8083

```

[30]: <keras.src.callbacks.History at 0x1abf4a22f90>

```

[31]: y_hat= model.predict(x_test)
      y_hat=[0 if val <0.5 else 1 for val in y_hat]

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45/45 [=====] - 0s 2ms/step

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[32]: y_hat

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```
[33]: accuracy_score(y_test,y_hat)
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[33]: 0.7877927608232789
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```
[34]: model.save('tfmodel')
```

```
INFO:tensorflow:Assets written to: tfmodel\assets
```

```
INFO:tensorflow:Assets written to: tfmodel\assets
```

```
[28]: model=load_model('tfmodel') # reloading the model after deleting (del model)
```

```
[29]: model
```

```
[29]: <keras.src.engine.sequential.Sequential at 0x1ab81caab50>
```

```
[ ]:
```