### artificial-neural-network

#### December 5, 2023

## [3]: | pip install tensorflow

```
Requirement already satisfied: tensorflow in c:\users\nikee\anaconda3\lib\site-
packages (2.15.0)
Requirement already satisfied: tensorflow-intel==2.15.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow) (2.15.0)
Requirement already satisfied: absl-py>=1.0.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (2.0.0)
Requirement already satisfied: astunparse>=1.6.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (1.6.3)
Requirement already satisfied: flatbuffers>=23.5.26 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (23.5.26)
Requirement already satisfied: gast!=0.5.0,!=0.5.1,!=0.5.2,>=0.2.1 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0 - tensorflow) (0.5.4)
Requirement already satisfied: google-pasta>=0.1.1 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (0.2.0)
Requirement already satisfied: h5py>=2.9.0 in c:\users\nikee\anaconda3\lib\site-
packages (from tensorflow-intel==2.15.0->tensorflow) (3.7.0)
Requirement already satisfied: libclang>=13.0.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (16.0.6)
Requirement already satisfied: ml-dtypes~=0.2.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (0.2.0)
Requirement already satisfied: numpy<2.0.0,>=1.23.5 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0 \rightarrow tensorflow) (1.24.3)
Requirement already satisfied: opt-einsum>=2.3.2 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (3.3.0)
Requirement already satisfied: packaging in c:\users\nikee\anaconda3\lib\site-
packages (from tensorflow-intel==2.15.0->tensorflow) (23.0)
Requirement already satisfied:
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protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.3,!=4.21.4,!=4.21.5,<5.0.0dev,>=3.20.3
in c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0 \rightarrow tensorflow) (4.23.4)
Requirement already satisfied: setuptools in c:\users\nikee\anaconda3\lib\site-
packages (from tensorflow-intel==2.15.0->tensorflow) (68.0.0)
Requirement already satisfied: six>=1.12.0 in c:\users\nikee\anaconda3\lib\site-
packages (from tensorflow-intel==2.15.0->tensorflow) (1.16.0)
Requirement already satisfied: termcolor>=1.1.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (2.4.0)
Requirement already satisfied: typing-extensions>=3.6.6 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (4.7.1)
Requirement already satisfied: wrapt<1.15,>=1.11.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (1.14.1)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (0.31.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (1.59.3)
Requirement already satisfied: tensorboard<2.16,>=2.15 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (2.15.1)
Requirement already satisfied: tensorflow-estimator<2.16,>=2.15.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (2.15.0)
Requirement already satisfied: keras<2.16,>=2.15.0 in
c:\users\nikee\anaconda3\lib\site-packages (from tensorflow-
intel==2.15.0->tensorflow) (2.15.0)
Requirement already satisfied: wheel<1.0,>=0.23.0 in
c:\users\nikee\anaconda3\lib\site-packages (from astunparse>=1.6.0->tensorflow-
intel==2.15.0->tensorflow) (0.38.4)
Requirement already satisfied: google-auth<3,>=1.6.3 in
c:\users\nikee\anaconda3\lib\site-packages (from
tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow) (2.24.0)
Requirement already satisfied: google-auth-oauthlib<2,>=0.5 in
c:\users\nikee\anaconda3\lib\site-packages (from
tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow) (1.1.0)
Requirement already satisfied: markdown>=2.6.8 in
c:\users\nikee\anaconda3\lib\site-packages (from
tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow) (3.4.1)
Requirement already satisfied: requests<3,>=2.21.0 in
c:\users\nikee\anaconda3\lib\site-packages (from
tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow) (2.31.0)
Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0 in
c:\users\nikee\anaconda3\lib\site-packages (from
```

```
tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow) (0.7.2)
Requirement already satisfied: werkzeug>=1.0.1 in
c:\users\nikee\anaconda3\lib\site-packages (from
tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow) (2.2.3)
Requirement already satisfied: cachetools<6.0,>=2.0.0 in
c:\users\nikee\anaconda3\lib\site-packages (from google-
auth<3,>=1.6.3->tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow)
(5.3.2)
Requirement already satisfied: pyasn1-modules>=0.2.1 in
c:\users\nikee\anaconda3\lib\site-packages (from google-
auth<3,>=1.6.3->tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow)
(0.2.8)
Requirement already satisfied: rsa<5,>=3.1.4 in
c:\users\nikee\anaconda3\lib\site-packages (from google-
auth<3,>=1.6.3->tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow)
(4.9)
Requirement already satisfied: requests-oauthlib>=0.7.0 in
c:\users\nikee\anaconda3\lib\site-packages (from google-auth-
oauthlib<2,>=0.5->tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow)
(1.3.1)
Requirement already satisfied: charset-normalizer<4,>=2 in
c:\users\nikee\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorboard<2.16,>=2.15->tensorflow-
intel==2.15.0->tensorflow) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in
c:\users\nikee\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorboard<2.16,>=2.15->tensorflow-
intel==2.15.0->tensorflow) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in
c:\users\nikee\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorboard<2.16,>=2.15->tensorflow-
intel==2.15.0->tensorflow) (1.26.16)
Requirement already satisfied: certifi>=2017.4.17 in
c:\users\nikee\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorboard<2.16,>=2.15->tensorflow-
intel==2.15.0->tensorflow) (2023.7.22)
Requirement already satisfied: MarkupSafe>=2.1.1 in
c:\users\nikee\anaconda3\lib\site-packages (from
werkzeug>=1.0.1->tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow)
(2.1.1)
Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in
c:\users\nikee\anaconda3\lib\site-packages (from pyasn1-modules>=0.2.1->google-
auth<3,>=1.6.3->tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow)
(0.4.8)
Requirement already satisfied: oauthlib>=3.0.0 in
c:\users\nikee\anaconda3\lib\site-packages (from requests-
oauthlib>=0.7.0->google-auth-
oauthlib<2,>=0.5->tensorboard<2.16,>=2.15->tensorflow-intel==2.15.0->tensorflow)
```

(3.2.2)

```
[2]: #Importing necessary Libraries
     import numpy as np
     import pandas as pd
     import tensorflow as tf
     import keras
     from keras.models import Sequential # for all deep learning neural network
     from keras.layers import Dense #for hidden layer
     from keras.layers import LeakyReLU,PReLU,ELU # for Activation function
     from keras. layers import Dropout # to avoid overfittin---> if neural network_
      →is too deep we usually used this
[3]: #Import data set
     dataset= pd.read_csv('Churn_Modelling.csv')
     dataset.head()
[3]:
       RowNumber CustomerId
                               Surname CreditScore Geography Gender
                                                                        Age \
                     15634602 Hargrave
                                                        France Female
                                                                         42
               1
                                                 619
                                                         Spain Female
     1
               2
                     15647311
                                   Hill
                                                 608
                                                                         41
                     15619304
     2
               3
                                   Onio
                                                 502
                                                        France Female
                                                                         42
     3
               4
                     15701354
                                   Boni
                                                 699
                                                        France Female
                                                                         39
     4
               5
                     15737888 Mitchell
                                                         Spain Female
                                                 850
                                                                         43
       Tenure
                 Balance NumOfProducts HasCrCard IsActiveMember
     0
                    0.00
                 83807.86
                                       1
                                                  0
                                                                  1
     1
             1
     2
              159660.80
                                       3
                                                                  0
            8
                                                  1
     3
             1
                    0.00
                                       2
                                                  0
                                                                  0
     4
               125510.82
                                       1
                                                  1
                                                                  1
       EstimatedSalary Exited
     0
             101348.88
     1
             112542.58
     2
             113931.57
                              1
               93826.63
     3
                              0
              79084.10
[4]: X= dataset.iloc[:,3:-1]
     y=dataset.iloc[:,-1]
     X.head()
[4]:
       CreditScore Geography Gender Age Tenure
                                                      Balance NumOfProducts \
                      France Female
                                                         0.00
     0
               619
                                        42
                                                 2
                                                                           1
     1
               608
                        Spain Female
                                        41
                                                 1
                                                     83807.86
                                                                           1
     2
               502
                      France Female
                                        42
                                                 8 159660.80
                                                                           3
     3
               699
                      France Female
                                                         0.00
                                                                           2
                                        39
                                                 1
```

```
4
                 850
                         Spain Female
                                         43
                                                  2 125510.82
                                                                             1
                                    EstimatedSalary
        HasCrCard IsActiveMember
      0
                                          101348.88
                                 1
      1
                 0
                                 1
                                          112542.58
                                 0
      2
                 1
                                          113931.57
      3
                 0
                                 0
                                           93826.63
      4
                 1
                                 1
                                           79084.10
[5]: geography=pd.get_dummies(X["Geography"],drop_first=True)
      gender= pd.get dummies(X['Gender'],drop first=True)
[6]: # concat
      X=pd.concat([X,geography,gender],axis=1)
      # Dropping
      X.drop(['Geography','Gender'],axis=1,inplace=True)
      X.head()
[6]:
                                     Balance NumOfProducts HasCrCard \
         CreditScore
                     Age
                           Tenure
      0
                 619
                       42
                                2
                                        0.00
                                                           1
                                                                      1
                 608
                       41
                                1
                                    83807.86
                                                           1
                                                                      0
      1
                                8 159660.80
                                                           3
      2
                 502
                       42
                                                                      1
      3
                 699
                       39
                                1
                                        0.00
                                                           2
                                                                      0
                 850
                                2 125510.82
                       43
                                                           1
                                                                      1
         IsActiveMember EstimatedSalary Germany Spain
                                                            Male
                               101348.88
                                            False False False
      0
      1
                      1
                               112542.58
                                            False True False
                                            False False False
      2
                      0
                               113931.57
                                93826.63
                                            False False False
      3
                      0
                      1
                                79084.10
                                            False True False
[7]: #splitting into training and test
      from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.
       →20,random_state=0)
[8]: from sklearn.preprocessing import StandardScaler
      sc=StandardScaler()
      X_train=sc.fit_transform(X_train)
      X test=sc.transform(X test)
[9]: classifier= Sequential()
[10]: # Adding the input layer and the first hidden layer
      classifier.
       -add(Dense(units=6,kernel_initializer='he_uniform',activation='relu',input_dim=11))
```

### Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 6)	72
dense_1 (Dense)	(None, 6)	42
dense_2 (Dense)	(None, 1)	7

Total params: 121 (484.00 Byte)
Trainable params: 121 (484.00 Byte)
Non-trainable params: 0 (0.00 Byte)

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```
[11]: # compiling ANN classifier.

→compile(loss='binary_crossentropy',optimizer='adam',metrics=['accuracy'])
```

```
Epoch 6/100
accuracy: 0.8205 - val_loss: 0.4156 - val_accuracy: 0.8239
Epoch 7/100
accuracy: 0.8212 - val_loss: 0.4105 - val_accuracy: 0.8243
accuracy: 0.8268 - val_loss: 0.4071 - val_accuracy: 0.8217
Epoch 9/100
accuracy: 0.8263 - val_loss: 0.4035 - val_accuracy: 0.8239
Epoch 10/100
accuracy: 0.8278 - val_loss: 0.4005 - val_accuracy: 0.8251
Epoch 11/100
536/536 [============ ] - 1s 2ms/step - loss: 0.3911 -
accuracy: 0.8287 - val_loss: 0.3978 - val_accuracy: 0.8228
Epoch 12/100
accuracy: 0.8293 - val_loss: 0.3945 - val_accuracy: 0.8281
Epoch 13/100
accuracy: 0.8304 - val_loss: 0.3925 - val_accuracy: 0.8307
Epoch 14/100
accuracy: 0.8313 - val_loss: 0.3908 - val_accuracy: 0.8285
Epoch 15/100
accuracy: 0.8326 - val_loss: 0.3896 - val_accuracy: 0.8266
Epoch 16/100
accuracy: 0.8336 - val_loss: 0.3880 - val_accuracy: 0.8273
Epoch 17/100
accuracy: 0.8354 - val_loss: 0.3860 - val_accuracy: 0.8281
Epoch 18/100
accuracy: 0.8336 - val_loss: 0.3832 - val_accuracy: 0.8304
Epoch 19/100
accuracy: 0.8330 - val_loss: 0.3821 - val_accuracy: 0.8285
Epoch 20/100
accuracy: 0.8341 - val_loss: 0.3814 - val_accuracy: 0.8285
Epoch 21/100
accuracy: 0.8332 - val_loss: 0.3811 - val_accuracy: 0.8273
```

```
Epoch 22/100
accuracy: 0.8384 - val_loss: 0.3804 - val_accuracy: 0.8398
Epoch 23/100
accuracy: 0.8418 - val_loss: 0.3794 - val_accuracy: 0.8391
Epoch 24/100
accuracy: 0.8436 - val_loss: 0.3794 - val_accuracy: 0.8387
Epoch 25/100
accuracy: 0.8474 - val_loss: 0.3768 - val_accuracy: 0.8413
Epoch 26/100
accuracy: 0.8485 - val_loss: 0.3773 - val_accuracy: 0.8421
Epoch 27/100
accuracy: 0.8494 - val_loss: 0.3766 - val_accuracy: 0.8421
Epoch 28/100
accuracy: 0.8494 - val_loss: 0.3773 - val_accuracy: 0.8482
Epoch 29/100
accuracy: 0.8520 - val_loss: 0.3771 - val_accuracy: 0.8444
Epoch 30/100
536/536 [============== ] - 1s 2ms/step - loss: 0.3575 -
accuracy: 0.8496 - val_loss: 0.3754 - val_accuracy: 0.8482
Epoch 31/100
accuracy: 0.8511 - val_loss: 0.3729 - val_accuracy: 0.8482
Epoch 32/100
accuracy: 0.8518 - val_loss: 0.3732 - val_accuracy: 0.8478
Epoch 33/100
accuracy: 0.8509 - val_loss: 0.3716 - val_accuracy: 0.8497
Epoch 34/100
accuracy: 0.8509 - val_loss: 0.3725 - val_accuracy: 0.8482
Epoch 35/100
accuracy: 0.8543 - val_loss: 0.3721 - val_accuracy: 0.8463
Epoch 36/100
accuracy: 0.8552 - val_loss: 0.3706 - val_accuracy: 0.8493
Epoch 37/100
accuracy: 0.8565 - val_loss: 0.3726 - val_accuracy: 0.8463
```

```
Epoch 38/100
accuracy: 0.8550 - val_loss: 0.3679 - val_accuracy: 0.8516
Epoch 39/100
accuracy: 0.8554 - val_loss: 0.3685 - val_accuracy: 0.8508
Epoch 40/100
accuracy: 0.8580 - val_loss: 0.3696 - val_accuracy: 0.8504
Epoch 41/100
accuracy: 0.8589 - val_loss: 0.3675 - val_accuracy: 0.8504
Epoch 42/100
accuracy: 0.8604 - val_loss: 0.3657 - val_accuracy: 0.8474
Epoch 43/100
accuracy: 0.8574 - val_loss: 0.3671 - val_accuracy: 0.8504
Epoch 44/100
accuracy: 0.8600 - val_loss: 0.3666 - val_accuracy: 0.8493
Epoch 45/100
accuracy: 0.8586 - val_loss: 0.3722 - val_accuracy: 0.8410
Epoch 46/100
536/536 [============= ] - 1s 2ms/step - loss: 0.3410 -
accuracy: 0.8610 - val_loss: 0.3664 - val_accuracy: 0.8455
Epoch 47/100
accuracy: 0.8610 - val_loss: 0.3649 - val_accuracy: 0.8463
Epoch 48/100
accuracy: 0.8606 - val_loss: 0.3651 - val_accuracy: 0.8489
Epoch 49/100
accuracy: 0.8623 - val_loss: 0.3654 - val_accuracy: 0.8519
Epoch 50/100
accuracy: 0.8623 - val_loss: 0.3659 - val_accuracy: 0.8489
Epoch 51/100
536/536 [============ ] - 1s 2ms/step - loss: 0.3393 -
accuracy: 0.8612 - val_loss: 0.3653 - val_accuracy: 0.8485
Epoch 52/100
accuracy: 0.8628 - val_loss: 0.3648 - val_accuracy: 0.8497
Epoch 53/100
accuracy: 0.8608 - val_loss: 0.3634 - val_accuracy: 0.8508
```

```
Epoch 54/100
accuracy: 0.8628 - val_loss: 0.3649 - val_accuracy: 0.8512
Epoch 55/100
accuracy: 0.8610 - val_loss: 0.3637 - val_accuracy: 0.8489
accuracy: 0.8653 - val_loss: 0.3644 - val_accuracy: 0.8493
Epoch 57/100
accuracy: 0.8623 - val_loss: 0.3639 - val_accuracy: 0.8485
Epoch 58/100
accuracy: 0.8632 - val_loss: 0.3638 - val_accuracy: 0.8523
Epoch 59/100
accuracy: 0.8617 - val_loss: 0.3636 - val_accuracy: 0.8527
Epoch 60/100
accuracy: 0.8640 - val_loss: 0.3636 - val_accuracy: 0.8497
Epoch 61/100
accuracy: 0.8628 - val_loss: 0.3637 - val_accuracy: 0.8504
Epoch 62/100
536/536 [============= ] - 1s 2ms/step - loss: 0.3363 -
accuracy: 0.8634 - val_loss: 0.3650 - val_accuracy: 0.8459
Epoch 63/100
accuracy: 0.8643 - val_loss: 0.3658 - val_accuracy: 0.8470
Epoch 64/100
accuracy: 0.8612 - val_loss: 0.3644 - val_accuracy: 0.8523
Epoch 65/100
accuracy: 0.8640 - val_loss: 0.3634 - val_accuracy: 0.8527
Epoch 66/100
accuracy: 0.8608 - val_loss: 0.3646 - val_accuracy: 0.8493
Epoch 67/100
accuracy: 0.8658 - val_loss: 0.3643 - val_accuracy: 0.8504
Epoch 68/100
accuracy: 0.8630 - val_loss: 0.3639 - val_accuracy: 0.8519
Epoch 69/100
accuracy: 0.8640 - val_loss: 0.3676 - val_accuracy: 0.8474
```

```
Epoch 70/100
accuracy: 0.8630 - val_loss: 0.3650 - val_accuracy: 0.8501
Epoch 71/100
accuracy: 0.8612 - val_loss: 0.3640 - val_accuracy: 0.8523
Epoch 72/100
accuracy: 0.8647 - val_loss: 0.3657 - val_accuracy: 0.8504
Epoch 73/100
accuracy: 0.8617 - val_loss: 0.3655 - val_accuracy: 0.8474
Epoch 74/100
accuracy: 0.8647 - val_loss: 0.3632 - val_accuracy: 0.8489
Epoch 75/100
536/536 [=========== ] - 1s 2ms/step - loss: 0.3334 -
accuracy: 0.8643 - val_loss: 0.3645 - val_accuracy: 0.8485
Epoch 76/100
accuracy: 0.8645 - val_loss: 0.3662 - val_accuracy: 0.8512
Epoch 77/100
accuracy: 0.8606 - val_loss: 0.3632 - val_accuracy: 0.8535
Epoch 78/100
536/536 [============== ] - 1s 2ms/step - loss: 0.3339 -
accuracy: 0.8612 - val_loss: 0.3632 - val_accuracy: 0.8519
Epoch 79/100
accuracy: 0.8642 - val_loss: 0.3627 - val_accuracy: 0.8554
Epoch 80/100
accuracy: 0.8614 - val_loss: 0.3636 - val_accuracy: 0.8512
Epoch 81/100
accuracy: 0.8600 - val_loss: 0.3668 - val_accuracy: 0.8455
Epoch 82/100
accuracy: 0.8630 - val_loss: 0.3652 - val_accuracy: 0.8519
Epoch 83/100
accuracy: 0.8628 - val_loss: 0.3647 - val_accuracy: 0.8489
Epoch 84/100
accuracy: 0.8634 - val_loss: 0.3632 - val_accuracy: 0.8504
Epoch 85/100
accuracy: 0.8638 - val_loss: 0.3643 - val_accuracy: 0.8516
```

```
Epoch 86/100
accuracy: 0.8612 - val_loss: 0.3637 - val_accuracy: 0.8497
Epoch 87/100
accuracy: 0.8627 - val_loss: 0.3629 - val_accuracy: 0.8519
Epoch 88/100
accuracy: 0.8612 - val_loss: 0.3625 - val_accuracy: 0.8550
Epoch 89/100
accuracy: 0.8632 - val_loss: 0.3654 - val_accuracy: 0.8489
Epoch 90/100
accuracy: 0.8617 - val_loss: 0.3640 - val_accuracy: 0.8527
Epoch 91/100
accuracy: 0.8638 - val_loss: 0.3646 - val_accuracy: 0.8504
Epoch 92/100
accuracy: 0.8623 - val_loss: 0.3638 - val_accuracy: 0.8493
Epoch 93/100
accuracy: 0.8621 - val_loss: 0.3627 - val_accuracy: 0.8538
Epoch 94/100
536/536 [============= ] - 1s 2ms/step - loss: 0.3316 -
accuracy: 0.8621 - val_loss: 0.3644 - val_accuracy: 0.8516
Epoch 95/100
accuracy: 0.8617 - val_loss: 0.3637 - val_accuracy: 0.8501
Epoch 96/100
accuracy: 0.8647 - val_loss: 0.3654 - val_accuracy: 0.8493
Epoch 97/100
accuracy: 0.8645 - val_loss: 0.3656 - val_accuracy: 0.8497
Epoch 98/100
accuracy: 0.8630 - val_loss: 0.3638 - val_accuracy: 0.8527
Epoch 99/100
536/536 [============= ] - 1s 2ms/step - loss: 0.3321 -
accuracy: 0.8619 - val_loss: 0.3626 - val_accuracy: 0.8523
Epoch 100/100
accuracy: 0.8623 - val_loss: 0.3624 - val_accuracy: 0.8550
```

```
[13]: # predicting the model
      y_pred= classifier.predict(X_test)
      y_pred= (y_pred>0.5)
      y_pred
     63/63 [======== ] - Os 1ms/step
[13]: array([[False],
             [False],
             [False],
             [False],
             [False],
             [False]])
[14]: from sklearn.metrics import
       →accuracy_score,confusion_matrix,classification_report
      cm= confusion_matrix(y_test,y_pred)
      print(cm)
      accuracy=accuracy_score(y_test,y_pred)
      print('The accuracy of the model is',accuracy)
      cl_report = classification_report(y_test,y_pred)
      print(cl_report)
     [[1515
              801
      [ 200 205]]
     The accuracy of the model is 0.86
                                recall f1-score
                   precision
                                                   support
                0
                        0.88
                                  0.95
                                            0.92
                                                      1595
                1
                        0.72
                                  0.51
                                            0.59
                                                       405
                                            0.86
                                                      2000
         accuracy
        macro avg
                        0.80
                                  0.73
                                            0.75
                                                      2000
                        0.85
                                  0.86
                                            0.85
     weighted avg
                                                      2000
[16]: import matplotlib.pyplot as plt
      print(model_history.history.keys())
      # summarize history for accuracy
      plt.plot(model_history.history['accuracy'])
      plt.plot(model_history.history['val_accuracy'])
      plt.title('model accuracy')
      plt.ylabel('accuracy')
      plt.xlabel('epoch')
      plt.legend(['train', 'test'], loc='upper left')
```

# plt.show()

dict\_keys(['loss', 'accuracy', 'val\_loss', 'val\_accuracy'])

