!pip install tensorflow-gpu

```
Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Requirement already satisfied: tensorflow-gpu in /usr/local/lib/python3.8/dist-packages (2.11.0)
     Requirement already satisfied: protobuf<3.20,>=3.9.2 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (3.19.6)
     Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (2.1.1)
     Requirement already satisfied: absl-py>=1.0.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (1.3.0)
     Requirement already satisfied: libclang>=13.0.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (14.0.6)
     Requirement already satisfied: numpy>=1.20 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (1.21.6)
     Requirement already satisfied: gast<=0.4.0,>=0.2.1 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (0.4.0)
     Requirement already satisfied: setuptools in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (57.4.0)
     Requirement already satisfied: google-pasta>=0.1.1 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (0.2.0)
    Requirement already satisfied: keras<2.12,>=2.11.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (2.11.0)
     Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (1.15.0)
     Requirement already satisfied: packaging in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (21.3)
     Requirement already satisfied: flatbuffers>=2.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (23.1.4)
     Requirement already satisfied: tensorboard<2.12,>=2.11 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (2.11.0)
     Requirement already satisfied: typing-extensions>=3.6.6 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (4.4.0)
     Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (3.3.0)
     Requirement already satisfied: grpcio<2.0,>=1.24.3 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (1.51.1)
     Requirement already satisfied: h5py>=2.9.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (3.1.0)
     Requirement already satisfied: tensorflow-estimator<2.12,>=2.11.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (2.11.
     Requirement already satisfied: astunparse>=1.6.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (1.6.3)
     Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (0.2
     Requirement already satisfied: wrapt>=1.11.0 in /usr/local/lib/python3.8/dist-packages (from tensorflow-gpu) (1.14.1)
     Requirement already satisfied: wheel<1.0,>=0.23.0 in /usr/local/lib/python3.8/dist-packages (from astunparse>=1.6.0->tensorflow-gpu) (0.
    Requirement already satisfied: google-auth<3,>=1.6.3 in /usr/local/lib/python3.8/dist-packages (from tensorboard<2.12,>=2.11->tensorflow
     Requirement already satisfied: tensorboard-data-server<0.7.0,>=0.6.0 in /usr/local/lib/python3.8/dist-packages (from tensorboard<2.12,>=
     Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in /usr/local/lib/python3.8/dist-packages (from tensorboard<2.12,>=2.11->te
     Requirement already satisfied: requests<3,>=2.21.0 in /usr/local/lib/python3.8/dist-packages (from tensorboard<2.12,>=2.11->tensorflow-g
     Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in /usr/local/lib/python3.8/dist-packages (from tensorboard<2.12,>=2.11-
     Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.8/dist-packages (from tensorboard<2.12,>=2.11->tensorflow-gpu)
     Requirement already satisfied: werkzeug>=1.0.1 in /usr/local/lib/python3.8/dist-packages (from tensorboard<2.12,>=2.11->tensorflow-gpu)
     Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /usr/local/lib/python3.8/dist-packages (from packaging->tensorflow-gpu) (3.0.
     Requirement already satisfied: pyasn1-modules>=0.2.1 in /usr/local/lib/python3.8/dist-packages (from google-auth<3,>=1.6.3->tensorboard<
     Requirement already satisfied: rsa<5,>=3.1.4 in /usr/local/lib/python3.8/dist-packages (from google-auth<3,>=1.6.3->tensorboard<2.12,>=2
     Requirement already satisfied: cachetools<6.0,>=2.0.0 in /usr/local/lib/python3.8/dist-packages (from google-auth<3,>=1.6.3->tensorboarc
     Requirement already satisfied: requests-oauthlib>=0.7.0 in /usr/local/lib/python3.8/dist-packages (from google-auth-oauthlib<0.5,>=0.4.1
     Requirement already satisfied: importlib-metadata>=4.4 in /usr/local/lib/python3.8/dist-packages (from markdown>=2.6.8->tensorboard<2.12
     Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.8/dist-packages (from requests<3,>=2.21.0->tensorboard<2.12,>=2.11
    Requirement already satisfied: urllib3<1.27,>=1.21.1 in /usr/local/lib/python3.8/dist-packages (from requests<3,>=2.21.0->tensorboard<2.
     Requirement already satisfied: chardet<5,>=3.0.2 in /usr/local/lib/python3.8/dist-packages (from requests<3,>=2.21.0->tensorboard<2.12,>
     Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.8/dist-packages (from requests3,>=2.21.0->tensorboard<2.12,
     Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.8/dist-packages (from importlib-metadata>=4.4->markdown>=2.6.8->tensc
     Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /usr/local/lib/python3.8/dist-packages (from pyasn1-modules>=0.2.1->google-auth<?
     Requirement already satisfied: oauthlib>=3.0.0 in /usr/local/lib/python3.8/dist-packages (from requests-oauthlib>=0.7.0->google-auth-oau
#importing tensorflow
import tensorflow as tf
#checking techsorflow version
print(tf.__version__)
     2.11.0
# import necessary libraries
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

data = pd.read_csv('/content/Churn_Modelling.csv')

import warnings

data.head()

reading the dataset

#checking first five rows

warnings.filterwarnings('ignore')

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	Esti
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	

#checking null values if any
data.isnull().sum()

RowNumber CustomerId 0 Surname 0 ${\tt CreditScore}$ 0 Geography Gender Age a Tenure 0 Balance NumOfProducts 0 HasCrCard 0 IsActiveMember 0 EstimatedSalary 0 Exited 0 dtype: int64

#observation: No null values found here

#checking duplicate values
data.duplicated().sum()

0

#observation: No duplicate values found here

#checking columns
data.columns

```
Index(['RowNumber', 'CustomerId', 'Surname', 'CreditScore', 'Geography',
    'Gender', 'Age', 'Tenure', 'Balance', 'NumOfProducts', 'HasCrCard',
    'IsActiveMember', 'EstimatedSalary', 'Exited'],
    dtype='object')
```

#dropping unnecessary columns
data = data.drop(['RowNumber','CustomerId','Surname'],axis=1)

from pandas.io.pytables import DataCol
#splitting the data
x = data.iloc[:,:-1]

y = data.iloc[:,-1]

x.head()

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	7
0	619	France	Female	42	2	0.00	1	1	1	101348.88	
1	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	
2	502	France	Female	42	8	159660.80	3	1	0	113931.57	
3	699	France	Female	39	1	0.00	2	0	0	93826.63	
4	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	

у

```
0 1
1 0
2 1
3 0
4 0
```

```
9995 0
9996 0
9997 1
9998 1
9999 0
```

Name: Exited, Length: 10000, dtype: int64

geography = pd.get_dummies(x['Geography'],drop_first=True)

geography

	Germany	Spain	
0	0	0	
1	0	1	
2	0	0	
3	0	0	
4	0	1	
9995	0	0	
9996	0	0	
9997	0	0	
9998	1	0	
9999	0	0	

10000 rows × 2 columns

gender = pd.get_dummies(x['Gender'],drop_first=True)

gender

	Male	
0	0	
1	0	
2	0	
3	0	
4	0	
9995	1	
9996	1	
9997	0	
9998	1	
9999	0	
10000	_	

10000 rows × 1 columns

x = x.drop(['Geography','Gender'],axis=1)

x.head()

#concatinating the geography and gender
x = pd.concat([x,geography,gender],axis=1)

x.head()

	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Germany	Spain	Male	1
0	619	42	2	0.00	1	1	1	101348.88	0	0	0	
1	608	41	1	83807.86	1	0	1	112542.58	0	1	0	
2	502	42	8	159660.80	3	1	0	113931.57	0	0	0	
3	699	39	1	0.00	2	0	0	93826.63	0	0	0	
4	850	43	2	125510.82	1	1	1	79084.10	0	1	0	

```
#splitting the dataset into x_train, y_train
from sklearn.model selection import train test split
```

```
x_train, x_test, y_train, y_test = train_test_split(
  x, y, test_size=0.2, random_state=0)
```

x_train.shape

(8000, 11)

#performing feature scaling

from sklearn.preprocessing import StandardScaler

```
sc = StandardScaler()
```

x_train = sc.fit_transform(x_train)

x_test = sc.transform(x_test)

x train

```
array([[ 0.16958176, -0.46460796, 0.00666099, ..., -0.5698444 , 1.74309049, -1.09168714], [-2.30455945, 0.30102557, -1.37744033, ..., 1.75486502, -0.57369368, 0.91601335], [-1.19119591, -0.94312892, -1.031415 , ..., -0.5698444 , -0.57369368, -1.09168714], ..., [ 0.9015152 , -0.36890377, 0.00666099, ..., -0.5698444 , -0.57369368, 0.91601335], [-0.62420521, -0.08179119, 1.39076231, ..., -0.5698444 , 1.74309049, -1.09168714], [-0.28401079, 0.87525072, -1.37744033, ..., 1.75486502, -0.57369368, -1.09168714]])
```

x_test

```
array([[-0.55204276, -0.36890377, 1.04473698, ..., 1.75486502, -0.57369368, -1.09168714],
[-1.31490297, 0.10961719, -1.031415 , ..., -0.5698444 , -0.57369368, -1.09168714],
[ 0.57162971, 0.30102557, 1.04473698, ..., -0.5698444 , 1.74309049, -1.09168714],
...,
[ -0.74791227, -0.27319958, -1.37744033, ..., -0.5698444 , 1.74309049, 0.91601335],
[ -0.00566991, -0.46460796, -0.33936434, ..., 1.75486502, -0.57369368, 0.91601335],
[ -0.79945688, -0.84742473, 1.04473698, ..., 1.75486502, -0.57369368, 0.91601335]])
```

```
x_train.shape
     (8000, 11)
#part2 = now let's create the ANN
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import ReLU, PReLU, LeakyReLU, ELU
from tensorflow.keras.layers import Dropout
# let's initialize the ANN
classifier = Sequential()
#adding the input layer
classifier.add(Dense(units=11,activation='relu'))
#adding the first hidden layer
classifier.add(Dense(units=7,activation='relu'))
\#classifier.add(Dropout(0.2)), we can also check with this
#adding the 2nd hidden layer
classifier.add(Dense(units=5,activation='relu'))
\#classifier.add(Dropout(0.2)), we can also check with this
#adding the output layer
classifier.add(Dense(units=1,activation='sigmoid'))
classifier.compile(optimizer = 'adam',loss = 'binary_crossentropy', metrics = ['accuracy'])
#Observations: in adam optimizer we will get the learning rate automatically as 0.01
#creating our own learning rate
import tensorflow
tensorflow.keras.optimizers.Adam(learning_rate-0.01)
    NameError
                                               Traceback (most recent call last)
    <ipython-input-169-d162903c8642> in <module>
           1 #creating our own learning rate
           2 import tensorflow
     ---> 3 tensorflow.keras.optimizers.Adam(learning rate-0.01)
    NameError: name 'learning_rate' is not defined
      SEARCH STACK OVERFLOW
model_history = classifier.fit(x_train,y_train,validation_split=0.33,batch_size=10,epochs=100)
```

```
======================== - בככא.ט :msoss - val_loss: סככא.ט - val_accuracy: סככא.ט - val_accuracy: סככא.ט
    )

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    Epoch 83/100
    536/536 [==============] - 2s 4ms/step - loss: 0.3238 - accuracy: 0.8675 - val_loss: 0.3627 - val_accuracy: 0.8591
    Epoch 84/100
                       ==========] - 2s 3ms/step - loss: 0.3229 - accuracy: 0.8662 - val_loss: 0.3633 - val_accuracy: 0.8546
    536/536 [====
    Epoch 85/100
    536/536 [=====
                      ==========] - 2s 4ms/step - loss: 0.3229 - accuracy: 0.8651 - val_loss: 0.3629 - val_accuracy: 0.8535
    Epoch 86/100
    536/536 [====
                         :========] - 2s 4ms/step - loss: 0.3233 - accuracy: 0.8651 - val_loss: 0.3615 - val_accuracy: 0.8599
    Epoch 87/100
                                    ===] - 2s 4ms/step - loss: 0.3238 - accuracy: 0.8679 - val_loss: 0.3631 - val_accuracy: 0.8561
    536/536 [====
    Fnoch 88/100
    536/536 [====
                                        - 2s 3ms/step - loss: 0.3247 - accuracy: 0.8658 - val_loss: 0.3652 - val_accuracy: 0.8550
    Epoch 89/100
                       ==========] - 2s 3ms/step - loss: 0.3233 - accuracy: 0.8649 - val_loss: 0.3635 - val_accuracy: 0.8561
    536/536 [====
    Epoch 90/100
    536/536 [==============] - 2s 4ms/step - loss: 0.3213 - accuracy: 0.8698 - val_loss: 0.3791 - val_accuracy: 0.8383
    Epoch 91/100
    536/536 [====
                            ========] - 2s 3ms/step - loss: 0.3229 - accuracy: 0.8670 - val_loss: 0.3651 - val_accuracy: 0.8557
    Epoch 92/100
    536/536 [======
                      ==========] - 2s 3ms/step - loss: 0.3226 - accuracy: 0.8683 - val_loss: 0.3652 - val_accuracy: 0.8565
    Epoch 93/100
    536/536 [====
                                     ==] - 2s 3ms/step - loss: 0.3233 - accuracy: 0.8640 - val_loss: 0.3657 - val_accuracy: 0.8516
    Epoch 94/100
    536/536 [====
                              =======] - 2s 3ms/step - loss: 0.3220 - accuracy: 0.8653 - val_loss: 0.3654 - val_accuracy: 0.8542
    Epoch 95/100
    536/536 [====
                                      ==] - 2s 4ms/step - loss: 0.3226 - accuracy: 0.8655 - val_loss: 0.3660 - val_accuracy: 0.8523
    Epoch 96/100
    536/536 [====
                             ========] - 3s 5ms/step - loss: 0.3217 - accuracy: 0.8642 - val_loss: 0.3690 - val_accuracy: 0.8565
    Epoch 97/100
    536/536 [=============] - 2s 4ms/step - loss: 0.3217 - accuracy: 0.8673 - val_loss: 0.3735 - val_accuracy: 0.8527
    Epoch 98/100
    536/536 [====
                         =========] - 5s 9ms/step - loss: 0.3228 - accuracy: 0.8668 - val_loss: 0.3738 - val_accuracy: 0.8527
    Epoch 99/100
    536/536 [=====
                      ==========] - 3s 5ms/step - loss: 0.3213 - accuracy: 0.8651 - val_loss: 0.3646 - val_accuracy: 0.8550
    Fnoch 100/100
                      536/536 [=======
#early_stopping
import tensorflow as tf
early_stopping = tf.keras.callbacks.EarlyStopping(
   monitor="val loss",
   min_delta=0.0001,
   patience=20,
   verbose=1.
   mode="auto",
   baseline=None,
   restore_best_weights=False,
   start_from_epoch=0,
)
model history = classifier.fit(x train,y train,validation split=0.33,batch size=10,epochs=100,callbacks = early stopping)
    Epoch 1/100
    536/536 [============] - 2s 4ms/step - loss: 0.3216 - accuracy: 0.8660 - val_loss: 0.3692 - val_accuracy: 0.8557
    Epoch 2/100
                                     ===] - 2s 4ms/step - loss: 0.3211 - accuracy: 0.8653 - val_loss: 0.3687 - val_accuracy: 0.8542
    536/536 [==:
    Epoch 3/100
    536/536 [============] - 2s 3ms/step - loss: 0.3212 - accuracy: 0.8671 - val_loss: 0.3670 - val_accuracy: 0.8538
    Epoch 4/100
    536/536 [===
                       Enoch 5/100
    536/536 [=====
                      Epoch 6/100
    536/536 [===
                            ========] - 2s 4ms/step - loss: 0.3205 - accuracy: 0.8679 - val_loss: 0.3643 - val_accuracy: 0.8550
    Epoch 7/100
    536/536 [==:
                                     :==] - 2s 4ms/step - loss: 0.3223 - accuracy: 0.8684 - val_loss: 0.3664 - val_accuracy: 0.8516
    Epoch 8/100
    536/536 [============] - 2s 3ms/step - loss: 0.3207 - accuracy: 0.8681 - val_loss: 0.3677 - val_accuracy: 0.8523
    Epoch 9/100
    536/536 [===
                                 :======] - 2s 3ms/step - loss: 0.3206 - accuracy: 0.8671 - val_loss: 0.3720 - val_accuracy: 0.8497
    Epoch 10/100
    536/536 [=============] - 2s 3ms/step - loss: 0.3216 - accuracy: 0.8675 - val_loss: 0.3690 - val_accuracy: 0.8573
    Epoch 11/100
    536/536 [====
                        ==========] - 2s 3ms/step - loss: 0.3194 - accuracy: 0.8664 - val_loss: 0.3656 - val_accuracy: 0.8516
    Epoch 12/100
    536/536 [==========] - 2s 3ms/step - loss: 0.3188 - accuracy: 0.8662 - val_loss: 0.3661 - val_accuracy: 0.8550
    Epoch 13/100
    536/536 [====
                        ==========] - 2s 4ms/step - loss: 0.3191 - accuracy: 0.8673 - val_loss: 0.3682 - val_accuracy: 0.8523
    Epoch 14/100
                        :===========] - 2s 4ms/step - loss: 0.3194 - accuracy: 0.8690 - val_loss: 0.3711 - val_accuracy: 0.8538
    536/536 [==:
```

```
Epoch 15/100
    536/536 [====
                      :=========] - 2s 4ms/step - loss: 0.3213 - accuracy: 0.8673 - val_loss: 0.3723 - val_accuracy: 0.8542
    Epoch 16/100
    Epoch 17/100
    536/536 [====
                      ===========] - 2s 3ms/step - loss: 0.3179 - accuracy: 0.8686 - val_loss: 0.3701 - val_accuracy: 0.8504
    Epoch 18/100
    536/536 [====
                                      - 2s 4ms/step - loss: 0.3201 - accuracy: 0.8655 - val_loss: 0.3666 - val_accuracy: 0.8542
    Epoch 19/100
    536/536 [====
                      ==========] - 2s 4ms/step - loss: 0.3184 - accuracy: 0.8692 - val_loss: 0.3665 - val_accuracy: 0.8554
    Epoch 20/100
    536/536 [====
                            ========] - 2s 3ms/step - loss: 0.3186 - accuracy: 0.8664 - val_loss: 0.3708 - val_accuracy: 0.8595
    Epoch 21/100
    536/536 [============] - 2s 4ms/step - loss: 0.3187 - accuracy: 0.8671 - val_loss: 0.3657 - val_accuracy: 0.8546
    Epoch 22/100
    536/536 [====
                        :=========] - 2s 3ms/step - loss: 0.3179 - accuracy: 0.8711 - val_loss: 0.3755 - val_accuracy: 0.8535
    Epoch 23/100
    Epoch 24/100
    536/536 [====
                      :===========] - 2s 4ms/step - loss: 0.3191 - accuracy: 0.8699 - val_loss: 0.3741 - val_accuracy: 0.8482
    Epoch 25/100
    536/536 [=====
                           ========] - 2s 4ms/step - loss: 0.3182 - accuracy: 0.8694 - val_loss: 0.3724 - val_accuracy: 0.8474
    Epoch 26/100
    Epoch 26: early stopping
#observation: when we apply early stopping, the models has stopped in 26th epochs where we got the high accuracy continuously
model_history.history.keys()
    dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])
#summerize history of accuracy
plt.plot(model_history.history['accuracy'])
plt.plot(model_history.history['val_accuracy'])
plt.title('model accuracy')
plt.ylable('accuracy')
plt.xlable('epoch')
plt.legend(['train','test'],loc='upper left')
plt.show()
                                       Traceback (most recent call last)
    AttributeError
    <ipython-input-178-6c93442be282> in <module>
         4 plt.plot(model_history.history['val_accuracy'])
         5 plt.title('model accuracy')
    ---> 6 plt.ylable('accuracy')
         7 plt.xlable('epoch')
         8 plt.legend(['train','test'],loc='upper left')
    AttributeError: module 'matplotlib.pyplot' has no attribute 'ylable'
     SEARCH STACK OVERFLOW
                     model accuracy
    0.870
    0.865
     0.860
    0.855
    0.850
    0.845
                       10
                             15
                                    20
#summerize history of loss
```

```
https://colab.research.google.com/drive/17Sv6d9k5Z10BYT5swbigKApMoG2rbl8Y#scrollTo=FjR0HTSk1cXT&printMode=true
```

plt.plot(model_history.history['loss'])
plt.plot(model_history.history['val_loss'])

plt.title('model loss')
plt.ylable('loss')
plt.xlable('epoch')

classifier.get_weights()

```
plt.legend(['train','test'],loc='upper left')
plt.show()
 ₽
     AttributeError
                                              Traceback (most recent call last)
     <ipython-input-179-73e2301a56ad> in <module>
          4 plt.plot(model_history.history['val_loss'])
          5 plt.title('model loss')
     ----> 6 plt.ylable('loss')
           7 plt.xlable('epoch')
           8 plt.legend(['train','test'],loc='upper left')
     AttributeError: module 'matplotlib.pyplot' has no attribute 'ylable'
      SEARCH STACK OVERFLOW
                           model loss
      0.37
      0.36
      0.35
      0.34
      0.33
      0.32
                                   15
                          10
                                          20
#making the predictions validating the model
#predicting the test set results
y_pred = classifier.predict(x_test)
     63/63 [======== ] - 0s 3ms/step
y_pred = (y_pred >= 0.5)
#make the confusion matrix
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test,y_pred)
cm
     array([[1499,
                   96],
            [ 191, 214]])
from sklearn.metrics import accuracy_score
score = accuracy_score(y_test, y_pred)
score
     0.8565
# getting the weights
```

```
-עטכאצטנט.ט , כטט/בסטט.ט , סטטטבאבט.ט , יעטכאצטנט.ט , עטכאצטנט.ט ,
         0.0695022 ], dtype=float32),
array([[ 0.05446914, -0.0271513 , -0.49611044, -0.1992463 , 1.0337577 ,
          -0.11983801, 0.08127924],
        \hbox{$[-0.04218744,\ 0.6845132\ ,\ 0.5409455\ ,\ -0.11985739,\ -0.11944453,}
           0.43122166, -0.05408784],
        [ 0.24662411, 0.25438663, -0.16986974, -0.60736585, 0.263576 ,
          0.27018544, -1.0614408],
        \hbox{$[-0.6984371\ ,\ 0.05909414,\ -0.40886375,\ -0.07667132,\ -1.1021444\ ,}
        0.65153563, 0.20674604],
[ 0.49887642, 0.66835535, 0.45850047, 0.5757199, 0.26261824, 0.54127806, -0.05531849],
        [ 0.35465476, 0.29024136, -0.54554623, 0.55335283, -0.12601252,
          0.6390836 , 0.15706158],
        [ 0.4035916 , -0.38544887, -1.0651339 , 0.5149539 , 0.06408582,
          -0.15341984, 0.29750165],
        \hbox{$[-0.6021392\ , \ 0.07206472, \ 0.6419946\ , \ 0.18929596, \ 0.38350916,}
          -0.3143208 , -0.19367662],
        [ \ 0.1587106 \ , \ -0.11467145, \ -0.8828561 \ , \ \ 0.15993945, \ \ 0.11302575,
          -0.03498594, 0.4827117 ],
        [-0.1820494 , -0.2571856 , -0.31944674, -1.4793364 , -0.01877056,
          0.24992792, 0.5423601 ],
        [\ 0.36178312,\ 0.20521167,\ -0.31077975,\ 0.02699341,\ 0.5932222\ ,
0.25637475, 0.6116501 ]], dtype=float32),
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-0.63105196, 0.08686335], dtype=float32),
\verb"array" ([[ \ 0.88825965, \ \ 0.24235576, \ \ 0.2726831 \ , \ -0.1001079 \ , \ \ 0.03549778],
         [ 0.39849687, -0.36546496, 0.477124 , 0.01393195, 0.5723573 ], [ 0.2154519 , 0.83851314, 0.1967897 , -2.218216 , -0.76328075],
        [-0.23566042, 0.5106066, -0.5420726, 0.5148286, -0.44234285], [ 0.26049614, -0.46540615, 0.22289342, 0.5889023, -0.03098134],
        [-0.28139395, 0.6182708, -0.4890441, 0.42980397, -0.31701675],
        [-0.13721392, -0.16644026, 0.46767297, 0.53952897, 0.63457507]],
       dtype=float32),
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       dtype=float32),
array([[ 0.4908445 , -0.47190037, 0.6066686 , 0.07257682, 0.13227247],
         \hbox{$[\,\hbox{-0.60433793, -0.76282346, -0.7777411 , 0.47370833, -0.6816994 \,],}
        [ 0.03423654, -0.46542445, 0.8120542, 0.13453387, -0.6313754 ], [-0.33680356, -0.11890525, -0.18988997, -1.4624531, -0.03748875],
        [ 0.79268265, -0.12498891, 0.6637929 , 0.8535326 , 0.18120623]],
       dtype=float32),
array([ 0.37637845, 0.
                                    , 0.33879015, 0.21803184, -0.01149005],
       dtype=float32),
array([[-1.2212205],
         [ 0.29727983],
         [-0.7134965],
         [-1.2959719],
         [ 0.9470253 ]], dtype=float32),
arrav([2.4014473]. dtvne=float32)]
```

X