```
!pip install tensorflow
Requirement already satisfied: tensorflow in
/usr/local/lib/python3.10/dist-packages (2.15.0)
Requirement already satisfied: absl-py>=1.0.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.4.0)
Requirement already satisfied: astunparse>=1.6.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.6.3)
Requirement already satisfied: flatbuffers>=23.5.26 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (23.5.26)
Requirement already satisfied: gast!=0.5.0,!=0.5.1,!=0.5.2,>=0.2.1
in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0.5.4)
Requirement already satisfied: google-pasta>=0.1.1 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (0.2.0)
Requirement already satisfied: h5py>=2.9.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (3.9.0)
Requirement already satisfied: libclang>=13.0.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (16.0.6)
Requirement already satisfied: ml-dtypes~=0.2.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (0.2.0)
Requirement already satisfied: numpy<2.0.0,>=1.23.5 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.23.5)
Requirement already satisfied: opt-einsum>=2.3.2 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (3.3.0)
Requirement already satisfied: packaging in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (23.2)
Requirement already satisfied: 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
/usr/local/lib/python3.10/dist-packages (from tensorflow) (3.20.3)
Requirement already satisfied: setuptools in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (69.0.3)
Requirement already satisfied: six>=1.12.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.16.0)
Requirement already satisfied: termcolor>=1.1.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (2.4.0)
Requirement already satisfied: typing-extensions>=3.6.6 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (4.5.0)
Requirement already satisfied: wrapt<1.15,>=1.11.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.14.1)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (0.35.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.60.0)
Requirement already satisfied: tensorboard<2.16,>=2.15 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (2.15.1)
Requirement already satisfied: tensorflow-estimator<2.16,>=2.15.0
in /usr/local/lib/python3.10/dist-packages (from tensorflow) (2.15.0)
Requirement already satisfied: keras<2.16,>=2.15.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (2.15.0)
Requirement already satisfied: wheel<1.0,>=0.23.0 in
```

```
/usr/local/lib/python3.10/dist-packages (from astunparse>=1.6.0-
>tensorflow) (0.42.0)
Requirement already satisfied: google-auth<3,>=1.6.3 in
/usr/local/lib/python3.10/dist-packages (from tensorboard<2.16,>=2.15-
>tensorflow) (2.17.3)
Requirement already satisfied: google-auth-oauthlib<2,>=0.5 in
/usr/local/lib/python3.10/dist-packages (from tensorboard<2.16,>=2.15-
>tensorflow) (1.2.0)
Requirement already satisfied: markdown>=2.6.8 in
/usr/local/lib/python3.10/dist-packages (from tensorboard<2.16,>=2.15-
>tensorflow) (3.5.2)
Requirement already satisfied: requests<3,>=2.21.0 in
/usr/local/lib/python3.10/dist-packages (from tensorboard<2.16,>=2.15-
>tensorflow) (2.31.0)
Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0
in /usr/local/lib/python3.10/dist-packages (from
tensorboard<2.16,>=2.15->tensorflow) (0.7.2)
Requirement already satisfied: werkzeug>=1.0.1 in
/usr/local/lib/python3.10/dist-packages (from tensorboard<2.16,>=2.15-
>tensorflow) (3.0.1)
Requirement already satisfied: cachetools<6.0,>=2.0.0 in
/usr/local/lib/python3.10/dist-packages (from google-auth<3,>=1.6.3-
>tensorboard<2.16,>=2.15->tensorflow) (5.3.2)
Requirement already satisfied: pyasn1-modules>=0.2.1 in
/usr/local/lib/python3.10/dist-packages (from google-auth<3,>=1.6.3-
>tensorboard<2.16,>=2.15->tensorflow) (0.3.0)
Requirement already satisfied: rsa<5,>=3.1.4 in
/usr/local/lib/python3.10/dist-packages (from google-auth<3,>=1.6.3-
>tensorboard<2.16,>=2.15->tensorflow) (4.9)
Requirement already satisfied: requests-oauthlib>=0.7.0 in
/usr/local/lib/python3.10/dist-packages (from google-auth-
oauthlib < 2, >= 0.5 - stensor board < 2.16, >= 2.15 - stensor flow) (1.3.1)
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0-
>tensorboard<2.16,>=2.15->tensorflow) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in
/usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0-
>tensorboard<2.16,>=2.15->tensorflow) (3.6)
Requirement already satisfied: urllib3<3,>=1.21.1 in
/usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0-
>tensorboard<2.16,>=2.15->tensorflow) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in
/usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0-
>tensorboard<2.16,>=2.15->tensorflow) (2023.11.17)
Requirement already satisfied: MarkupSafe>=2.1.1 in
/usr/local/lib/python3.10/dist-packages (from werkzeug>=1.0.1-
>tensorboard<2.16,>=2.15->tensorflow) (2.1.4)
Requirement already satisfied: pyasn1<0.6.0,>=0.4.6 in
/usr/local/lib/python3.10/dist-packages (from pyasn1-modules>=0.2.1-
```

```
>google-auth<3,>=1.6.3->tensorboard<2.16,>=2.15->tensorflow) (0.5.1)
Requirement already satisfied: oauthlib>=3.0.0 in
/usr/local/lib/python3.10/dist-packages (from requests-
oauthlib>=0.7.0->google-auth-oauthlib<2,>=0.5-
>tensorboard<2.16,>=2.15->tensorflow) (3.2.2)
WARNING: Running pip as the 'root' user can result in broken
permissions and conflicting behaviour with the system package manager.
It is recommended to use a virtual environment instead:
https://pip.pypa.io/warnings/venv
!nvidia-smi
Thu Feb 1 08:01:13 2024
+-----
-----+
Version: 12.2
|-----
| GPU Name Persistence-M | Bus-Id
                                      Disp.A |
Volatile Uncorr. ECC |
             Pwr:Usage/Cap | Memory-Usage |
| Fan Temp Perf
GPU-Util Compute M. |
MIG M. |
| 0 Tesla T4
                         Off | 00000000:00:04.0 Off |
| N/A 39C P8
               9W / 70W | 0MiB / 15360MiB |
0% Default |
N/A |
+-----
+----+
+-----
| Processes:
GPU GI CI PID Type Process name
GPU Memory
    ID
         ID
Usage |
No running processes found
```

```
----+
!pip install tensorflow-gpu
Collecting tensorflow-gpu
 Using cached tensorflow-gpu-2.12.0.tar.gz (2.6 kB)
 error: subprocess-exited-with-error
 x python setup.py egg info did not run successfully.
  exit code: 1
  See above for output.
  note: This error originates from a subprocess, and is likely not a
problem with pip.
 Preparing metadata (setup.py) ... error: metadata-generation-failed
× Encountered error while generating package metadata.
└-> See above for output.
note: This is an issue with the package mentioned above, not pip.
hint: See above for details.
import tensorflow as tf
print("TensorFlow version:", tf.__version__)
print("GPU available:", tf.config.list physical devices('GPU'))
TensorFlow version: 2.15.0
GPU available: [PhysicalDevice(name='/physical device:GPU:0',
device type='GPU')]
import tensorflow as tf
# Create a random matrix
a = tf.random.normal((1000, 1000))
# Perform a matrix multiplication on GPU
b = tf.matmul(a, a)
print("GPU computation time:", b)
GPU computation time: tf.Tensor(
[[ 23.644073 -53.09984 4.4855604 ... 44.762833 15.747591
 -27.254604 1
 [ 17.465328 -10.247104 -48.122856 ... 9.175959
                                                     20.489845
  40.124718 ]
 [ -5.5265207 32.026146 31.498093 ... 40.996006
                                                     36,62998
  69.20412
 [ 3.4558582 45.48342
                           5.707222 ... -24.857723
                                                     46.83235
```

```
-18.666574 ]
 [ -1.7733288 -16.848602    9.835978    ...    57.113823    7.558341
  -18.197208 ]
                           66.461 ...
                                           9.252407
 6.84245
                9.575052
                                                       49.378075
   30.718555 ]], shape=(1000, 1000), dtype=float32)
import tensorflow as tf
print(tf.__version__)
2.15.0
## import some basics libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
dataset=pd.read csv('Churn Modelling.csv')
dataset.head()
   RowNumber CustomerId Surname CreditScore Geography Gender Age
/
0
                15634602 Hargrave
                                            619
                                                   France Female
                                                                    42
           2
                15647311
                              Hill
                                            608
1
                                                    Spain Female
                                                                    41
2
           3
                15619304
                              Onio
                                            502
                                                   France Female
                                                                    42
                15701354
                              Boni
                                            699
                                                   France Female
                                                                    39
                15737888 Mitchell
                                            850
                                                    Spain Female
                                                                    43
                      NumOfProducts HasCrCard
  Tenure
             Balance
                                                IsActiveMember \
0
        2
                0.00
                                  1
                                             1
                                                             1
1
        1
           83807.86
                                  1
                                             0
                                                             1
2
        8 159660.80
                                  3
                                             1
                                                             0
3
                                  2
        1
                0.00
                                             0
                                                             0
4
           125510.82
                                                             1
   EstimatedSalary Exited
0
         101348.88
                         1
         112542.58
1
                         0
2
         113931.57
                         1
3
          93826.63
                         0
          79084.10
                         0
##Divide the dataset into independent and dependent features
X=dataset.iloc[:,3:13]
y=dataset.iloc[:,13]
X.head()
```

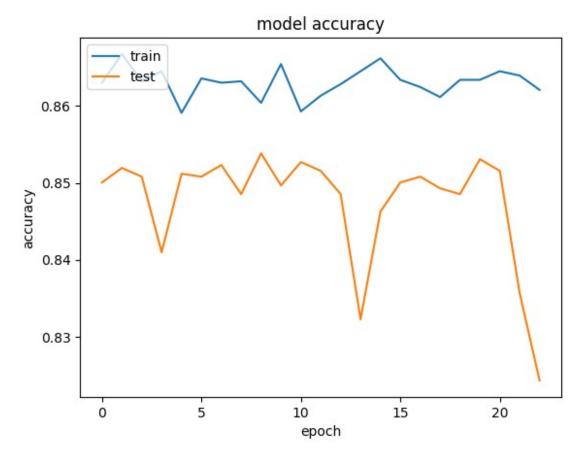
```
CreditScore Geography
                           Gender
                                   Age Tenure
                                                    Balance
NumOfProducts
           619
                   France
                           Female
                                     42
                                              2
                                                       0.00
1
1
                                                  83807.86
           608
                    Spain
                           Female
                                     41
                                              1
1
2
           502
                   France
                           Female
                                     42
                                                 159660.80
                                              8
3
3
           699
                   France
                           Female
                                     39
                                                       0.00
2
4
                                                 125510.82
           850
                    Spain Female
                                     43
1
   HasCrCard
              IsActiveMember
                               EstimatedSalary
0
                                      101348.88
1
           0
                            1
                                      112542.58
2
           1
                            0
                                      113931.57
3
           0
                            0
                                       93826.63
4
           1
                            1
                                       79084.10
У
0
        1
1
        0
2
        1
3
        0
4
        0
9995
        0
9996
        0
9997
        1
9998
        1
9999
Name: Exited, Length: 10000, dtype: int64
##Feature Engineering
geography=pd.get dummies(X['Geography'],drop first=True)
gender=pd.get dummies(X['Gender'],drop first=True)
##Concatenate these variables with dataframe
X=X.drop(['Geography','Gender'],axis=1)
X.head()
                                          NumOfProducts HasCrCard \
   CreditScore
                 Age Tenure
                                 Balance
0
                  42
                                    0.00
           619
                           1
                               83807.86
                                                       1
                                                                  0
1
           608
                  41
2
           502
                  42
                           8
                              159660.80
                                                       3
                                                                  1
3
           699
                  39
                           1
                                    0.00
                                                       2
                                                                  0
```

```
4
           850
                 43
                          2 125510.82
   IsActiveMember
                   EstimatedSalary
                         101348.88
0
                1
1
                1
                         112542.58
2
                0
                         113931.57
3
                0
                          93826.63
4
                1
                          79084.10
X=pd.concat([X,geography,gender],axis=1)
#Splitting the dataset into Training set and Test Set
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,rando
m state=0)
##Feature scaling
from sklearn.preprocessing import StandardScaler
sc= StandardScaler()
X train=sc.fit transform(X train)
X test=sc.transform(X test)
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:432:
UserWarning: X has feature names, but StandardScaler was fitted
without feature names
 warnings.warn(
X train
array([[ 0.16958176, -0.46460796, 0.00666099, ..., -0.5698444 ,
         1.74309049, -1.09168714],
                     0.30102557, -1.37744033, ..., 1.75486502,
       [-2.30455945,
        -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([[ 5.97000000e+02, 3.50000000e+01, 8.00000000e+00, ...,
         1.00000000e+00, -7.10542736e-18, -4.17443857e-17],
       [ 5.23000000e+02, 4.00000000e+01, 2.00000000e+00, ...,
```

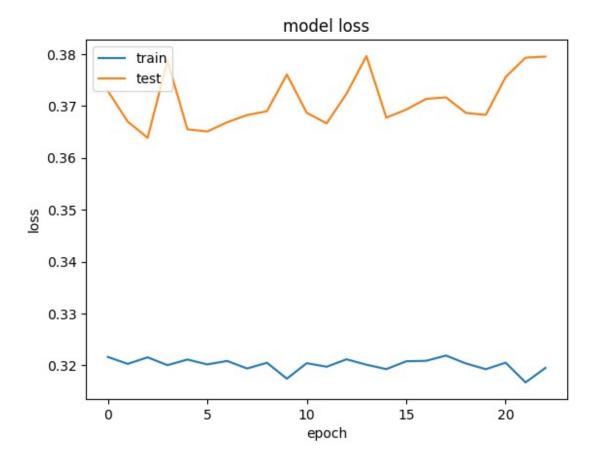
```
5.77315973e-17, -7.10542736e-18, -4.17443857e-17],
       [ 7.06000000e+02, 4.20000000e+01, 8.00000000e+00, ...,
         5.77315973e-17, 1.00000000e+00, -4.17443857e-17],
       [ 5.7800000e+02,
                          3.60000000e+01,
                                           1.00000000e+00, ...,
         5.77315973e-17,
                          1.00000000e+00,
                                           1.00000000e+00],
                                           4.00000000e+00, ...,
       [ 6.50000000e+02, 3.40000000e+01,
         1.00000000e+00, -7.10542736e-18,
                                           1.00000000e+00],
                                           8.00000000e+00, ...,
       [ 5.73000000e+02, 3.00000000e+01,
         1.00000000e+00, -7.10542736e-18, 1.00000000e+00]
X train.shape
(8000, 11)
## Part2- Now lets create the ANN
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import LeakyReLU,PReLU,ELU,ReLU
from tensorflow.keras.layers import Dropout
###lets initialize the ANN
Classifier=Sequential()
## Adding the imput Layer
Classifier.add(Dense(units=11,activation='relu'))
##Adding the first hidden layer
Classifier.add(Dense(units=7,activation='relu'))
#adding the second hudden layer
Classifier.add(Dense(units=6,activation='relu'))
#adding output layer
Classifier.add(Dense(1,activation='sigmoid'))
Classifier.compile(optimizer=opt,loss='binary crossentropy',metrics=['
accuracy'])
import tensorflow
opt=tensorflow.keras.optimizers.Adam(learning rate=0.01)
##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,bat
ch size=10,epochs=1000,callbacks=early stopping)
Epoch 1/1000
- accuracy: 0.8630 - val loss: 0.3729 - val accuracy: 0.8501
Epoch 2/1000
536/536 [============== ] - 3s 6ms/step - loss: 0.3203
- accuracy: 0.8668 - val loss: 0.3670 - val accuracy: 0.8519
Epoch 3/1000
- accuracy: 0.8632 - val loss: 0.3639 - val accuracy: 0.8508
Epoch 4/1000
- accuracy: 0.8645 - val loss: 0.3787 - val_accuracy: 0.8410
Epoch 5/1000
- accuracy: 0.8591 - val loss: 0.3655 - val accuracy: 0.8512
Epoch 6/1000
- accuracy: 0.8636 - val loss: 0.3651 - val accuracy: 0.8508
Epoch 7/1000
- accuracy: 0.8630 - val loss: 0.3669 - val accuracy: 0.8523
Epoch 8/1000
- accuracy: 0.8632 - val loss: 0.3682 - val accuracy: 0.8485
Epoch 9/1000
- accuracy: 0.8604 - val loss: 0.3690 - val accuracy: 0.8538
Epoch 10/1000
536/536 [============= ] - 2s 4ms/step - loss: 0.3175
- accuracy: 0.8655 - val_loss: 0.3760 - val_accuracy: 0.8497
Epoch 11/1000
- accuracy: 0.8593 - val loss: 0.3687 - val accuracy: 0.8527
Epoch 12/1000
536/536 [============== ] - 3s 5ms/step - loss: 0.3198
- accuracy: 0.8614 - val loss: 0.3666 - val accuracy: 0.8516
Epoch 13/1000
536/536 [============== ] - 2s 4ms/step - loss: 0.3212
- accuracy: 0.8628 - val loss: 0.3723 - val accuracy: 0.8485
Epoch 14/1000
- accuracy: 0.8645 - val_loss: 0.3796 - val_accuracy: 0.8323
Epoch 15/1000
```

```
- accuracy: 0.8662 - val loss: 0.3677 - val accuracy: 0.8463
Epoch 16/1000
536/536 [============== ] - 3s 6ms/step - loss: 0.3208
- accuracy: 0.8634 - val loss: 0.3693 - val accuracy: 0.8501
Epoch 17/1000
- accuracy: 0.8625 - val loss: 0.3713 - val accuracy: 0.8508
Epoch 18/1000
- accuracy: 0.8612 - val loss: 0.3716 - val accuracy: 0.8493
Epoch 19/1000
536/536 [============= ] - 3s 5ms/step - loss: 0.3204
- accuracy: 0.8634 - val loss: 0.3687 - val accuracy: 0.8485
Epoch 20/1000
- accuracy: 0.8634 - val loss: 0.3683 - val accuracy: 0.8531
Epoch 21/1000
- accuracy: 0.8645 - val loss: 0.3756 - val accuracy: 0.8516
Epoch 22/1000
536/536 [============== ] - 3s 5ms/step - loss: 0.3167
- accuracy: 0.8640 - val loss: 0.3793 - val accuracy: 0.8357
Epoch 23/1000
- accuracy: 0.8621 - val loss: 0.3795 - val accuracy: 0.8243
Epoch 23: early stopping
model history.history.keys()
dict keys(['loss', 'accuracy', 'val loss', 'val accuracy'])
#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()
```



```
# summarize history for loss
plt.plot(model_history.history['loss'])
plt.plot(model_history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```



```
# Part 3 - Making the predictions and evaluating the model
# Predicting the Test set results
y_pred = Classifier.predict(X_test)
y_pred = (y_pred > 0.5)
63/63 [========= ] - Os 4ms/step
## make the confusion metrix
from sklearn.metrics import confusion matrix
cm=confusion matrix(y test,y pred)
\mathsf{cm}
array([[730, 865],
       [139, 266]])
# Calculate the Accuracy
from sklearn.metrics import accuracy_score
score=accuracy_score(y_pred,y_test)
score
0.498
```

```
## get the weights
Classifier.get weights()
[array([[-6.72509551e-01, -1.84281871e-01,
                                             1.12500034e-01,
         -1.81433892e+00, -8.09982002e-01,
                                            1.76370156e+00,
          2.31279278e+00, -5.47874928e-01,
                                            2.78982568e+00,
         -8.15676376e-02, -2.19630742e+00],
        [ 4.70728397e+00, -3.89807850e-01, -4.09404993e+00,
                          2.88580871e+00,
         -2.07181364e-01,
                                           -1.16329718e+00,
                          -2.11800814e+00,
         -3.05263829e+00,
                                            1.81204987e+00,
          1.05808757e-03, -2.98090363e+00],
                          6.20624185e-01,
                                            4.66723025e-01,
        [-2.85697073e-01,
          1.75992215e+00, 4.37421829e-01,
                                             1.16838582e-01,
                          2.99453640e+00,
         -8.64167035e-01,
                                             3.56366873e+00,
         -1.84047893e-01,
                          2.27878761e+00],
        [-1.02047884e+00, -4.06635571e+00, -8.81114721e-01,
          1.57106626e+00, -1.12913764e+00, -2.75600582e-01,
          2.89875650e+00, 9.60043252e-01,
                                            7.88069069e-01.
          1.57982016e+00,
                          3.26408362e+00],
                                             9.12325382e-01,
        [-3.52437019e+00, -5.48682070e+00,
          3.52715421e+00, -1.69791281e-01,
                                            3.48134851e+00,
          2.25111961e+00, -1.78163588e+00,
                                            1.97066784e+00,
          5.48902845e+00, -1.61501467e+00],
        [-2.34298706e+00, -6.70243979e-01, -1.40832591e+00,
         -7.77164519e-01, -6.49422407e-01,
                                            1.55517626e+00,
          9.26993638e-02, -1.32727683e+00,
                                             1.07449472e+00,
         -2.03437224e-01, -1.94583404e+00],
        [-1.00281692e+00, 6.86129570e-01,
                                            9.08577815e-03,
         -9.94411290e-01,
                          4.81117868e+00,
                                            1.80351293e+00,
          3.01582432e+00, 3.72218132e+00,
                                            9.33219045e-02,
          1.67438760e-01, -9.32064414e-01],
        [ 2.16485828e-01,
                          1.30235478e-01, -6.87126100e-01,
         -3.62331939e+00, -5.15279591e-01,
                                            3.00986767e+00,
          1.08365655e+00, -9.81289268e-01, -1.16710246e+00,
          1.47214696e-01, -1.26870525e+00],
        [ 3.76084298e-01,
                          2.87395501e+00, -1.94282091e+00,
          4.97392595e-01, -7.32620060e-01,
                                            3.05467248e-01,
         -2.46112287e-01,
                          3.67432326e-01,
                                            1.04490042e+00,
         -1.77958167e+00, -1.45957160e+00],
        [-6.41324759e-01, -2.74460584e-01, -7.71972775e-01,
         -6.38196409e-01, 5.41176915e-01, 5.50887465e-01,
          4.25735079e-02,
                          1.83118045e+00, -1.89027357e+00,
          3.71938318e-01,
                          1.40154719e+00],
        [-1.10428102e-01, -1.19963318e-01,
                                            1.17026711e+00,
                          1.42019665e+00,
         -6.04610741e-01,
                                            1.30864692e+00,
          1.63064048e-01,
                          5.19107401e-01, -1.28584743e-01,
          2.91206777e-01, -1.20407677e+00]], dtype=float32),
array([-0.18619463, -2.4928243 , 5.1261234 , -2.9508593 , -4.0754614
        -3.6271155 , 0.01662346, -2.4706192 , -4.8959513 , -4.277336
```

```
-1.7559004 ], dtype=float32),
array([[-0.92239684, 0.37103936, -1.3784449 , -7.501407 , -
0.8615784 ,
         1.4035454 ],
        [-0.7046321 , -2.8108506 , -4.5866075 , 0.36675444, -
0.21041748,
         -0.20127408],
        [ 1.8358886 , 0.53034216, 1.8308427 , 0.35606104, -
0.50012803,
        -2.221816 ],
        [-0.09373069, -2.33576 , 1.568848 , -3.0296187 , -
0.01621949,
         0.9282872],
        [-0.53145057, 1.7057312, 0.65117496, -1.2876368, -
0.02406495,
         -3.862307 ],
        [-0.4172956 , 1.6052992 , 1.5952787 , 1.3609582 , -
0.46453026,
         2.1103034 ],
        [-3.4181988 , -0.96016973 , 0.24366865 , 0.5355145 , -
0.53002965,
        -2.3825674],
        [ 1.9648957 , -2.0153368 , 0.90328765, 1.0305336 , -
0.4076225
         0.985752051,
        [-2.371814 , 1.8528981 , -1.8947605 , -5.2339053 , -
0.4787469
         0.085101461,
        [ 0.06613193, -1.8432208 , -2.4533746 , -1.3874533 , -
0.05306538,
         4.3157873 ],
        [-0.39467245, -2.9727259 , -1.0642824 , -3.4855046 , -
0.27543354,
         -1.4027644 ]], dtype=float32),
array([ 1.6702005 , -6.5485625 , 3.44491 , -0.92504317, -0.6787864
        -3.050061 ], dtype=float32),
array([[-2.3024502 , -0.5547372 , 0.27347672, -0.00821126, -
0.06514273,
         -0.45833448, -1.4710728 ],
        [ 0.10025582, -0.00645378, 1.5795236 , -0.7306909 , -
0.2858292
         -0.76354474, -1.0672929 ],
        [-0.72517025, 0.0380387, 0.8398574, -0.10435436, -
0.00438156,
         -0.20708339, -0.9149316 ],
        [ 0.65167207, -0.30606323, 1.3041909 , -0.77460784, -
0.1830079 ,
```

```
0.02212565, -1.0740519 ],
        [-0.6097542, 0.58453035, -0.39472187, -0.52702916,
0.4870162
          0.06253976, -0.28517613],
        [-0.86673176, -0.11276332, -1.7392315 , -0.2503454 , -
0.42943946,
         -0.6649588 , 0.883663 ]], dtype=float32),
array([-3.9933953 , -3.5602343 , -0.5923214 , -0.66967607, -0.1898797
        -0.53514206, 1.9213245 ], dtype=float32),
array([[-0.21315075, -1.8544565 , -0.60921353, -0.28420576, -
0.15054648,
         -0.313075361,
        [-0.05213704, -0.4249323, -0.20810893, -0.32871756, -0.4249323]
0.62844694,
          0.42495382],
        [ 0.36464247, -0.94677836, -0.17811972, -0.56175035, -
0.83719474,
         -0.39334235],
        [0.20393509, -0.16592333, -0.60960275, 0.04187553, -0.60960275]
0.2376461
          0.01504612],
        [-0.38531703, 0.07927102, 0.07064295, 0.20331907, -
0.36400813.
          0.47444415],
        [-0.22164404, 0.18270329, -0.1682508, 0.1844728, -
0.07435948,
         -0.53298396],
        [-3.442308 , 0.5775392 , -0.39769346, -0.0858357 , -
0.5590735
         -0.5944817 ]], dtype=float32),
 array([ 0.12267611, 0.82916784, 0. , -0.13760321, -
0.42449898,
        -0.39950797], dtype=float32),
 array([[-0.32271966],
        [ 0.5535804 ],
        [-0.40824097],
        [-0.30143762],
        [-0.2743126],
        [ 0.08382902]], dtype=float32),
 array([-0.99778324], dtype=float32)]
```