MUSICAL INSTRUMENT CLASSIFICATION USING DEEP LEARNING TECHNIQUES

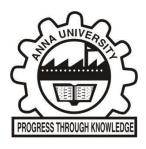
A PROJECT REPORT

Submitted by

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for the course

CS 6301 - MACHINE LEARNING



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ABSTRACT:

Musical Information Retrieval is the interdisciplinary science of retrieving information from music. It is a new field of research; its objective is extracting data and metadata from musical recordings. Effective audio processing can improve speed, reduce errors, and sometimes increase the accuracy of detecting the musical instrument. One of the main tools used for classification tasks in recent years is deep learning. The three deep learning techniques proposed here are Convolutional Neural Network (CNN), Recurrent Neural Network – LSTM (RNN-LSTM) and Deep Belief Network (DBN).

DATASET: Musical Instrument Classification based on Audio Samples.

- The dataset has 1,020 audio samples of 10 seconds each which are classified into 17 musical instruments.
- The dataset is balanced (60 samples for each instrument).

Reference link:

https://drive.google.com/drive/folders/1X6BuQLI-mAKJmjZNKBxvllfEGseh6a8x?usp=sharing

I. CNN:

FEATURE EXTRACTION METHOD:

CODE:

```
: #comparing CNN model for classification
  import os
  import librosa
  import numpy as np
  import pandas as pd
  from tqdm import tqdm
   !pip install python_speech_features
  from python_speech_features import mfcc
  from sklearn.preprocessing import LabelEncoder, OneHotEncoder
  from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
  from keras.layers import Conv2D, MaxPool2D, Flatten, Dense, Dropout from keras.models import Sequential
   import matplotlib.pyplot as plt
  import seaborn as sns
   #read csv
  df = pd.read_csv('C:/Users/GCS/testaudio/Instrumentstest.csv')
  df = df.drop(df.iloc[:,1:52],axis=1)
  #load audio
  path = 'C:/Users/GCS/testaudio/'
   audio_data = list()
   for i in tqdm(range(df.shape[0])):
      print(path+ df['Instrument'].iloc[i]+ '/'+ df['File Name'].iloc[i])
audio_data.append(librosa.load(path+ df['Instrument'].iloc[i]+ '/'+ df['File Name'].iloc[i]))
  audio_data = np.array(audio_data)
  df['audio_waves'] = audio_data[:,0]
df['samplerate'] = audio_data[:,1]
  bit lengths = list()
  for i in range(df.shape[0]):
      bit_lengths.append(len(df['audio_waves'].iloc[i]))
  bit_lengths = np.array(bit_lengths)
df['bit_lengths'] = bit_lengths
df['seconds_length'] = df['bit_lengths']/df['samplerate']
  df.head()
```

```
min_bits = np.min(df['bit_lengths'])
min_seconds = np.min(df['seconds_length'])
#identify audio wave and labels for each audio file for minimum of audio bitlength found
num_samples = 1020
generated_audio_waves = list()
generated_audio_labels = list()
 for i in tqdm(range(num_samples)):
     try:
           chosen_file =df['File Name'].iloc[i]
chosen_initial = 0
           generated_audio_waves.append(df[df['File Name']==chosen_file]['audio_waves'].values[0][chosen_initial:chosen_initial+min_generated_audio_labels.append(df[df['File Name']==chosen_file]['Instrument'].values)
      except ValueError:
           continue
generated_audio_waves = np.array(generated_audio_waves)
generated_audio_labels = np.array(generated_audio_labels)
#identify mfcc features
mfcc features = list()
for i in tqdm(range(len(generated_audio_waves))):
     mfcc_features.append(mfcc(generated_audio_waves[i]))
mfcc features = np.array(mfcc features)
label_encoder = LabelEncoder()
label_encoded = label_encoder.fit_transform(generated_audio_labels)
label_encoded = label_encoded[:, np.newaxis]
one_hot_encoder = OneHotEncoder(sparse=False)
one_hot_encoded = one_hot_encoder.fit_transform(label_encoded)
#scaling features
X = mfcc_features
y = one_hot_encoded
X = (X-X.min())/(X.max()-X.min())
```

OUTPUT:

```
46%
                                                        467/1020 [02:56<03:38, 2.53it/s]
C:/Users/GCS/testaudio/Sarod/Sarod18.wav
46%|
                                                        | 468/1020 [02:57<03:37, 2.54it/s]
C:/Users/GCS/testaudio/Sarod/Sarod19.wav
46%
                                                        | 469/1020 [02:57<03:34, 2.56it/s]
C:/Users/GCS/testaudio/Sarod/Sarod20.wav
46%
                                                        | 470/1020 [02:57<03:33, 2.58it/s]
C:/Users/GCS/testaudio/Sarod/Sarod21.wav
46%
                                                        | 471/1020 [02:58<03:36, 2.53it/s]
C:/Users/GCS/testaudio/Sarod/Sarod22.wav
46%|
                                                       | 472/1020 [02:58<03:35, 2.55it/s]
```

```
100%| | 1020/1020 [00:00<00:00, 1313.78it/s]
100%| | 1020/1020 [00:00<00:00, 1313.78it/s]
100%| | 1020/1020 [00:32<00:00, 31.59it/s]
C:\Users\GCS\anaconda3\lib\site-packages\sklearn\utils\validation.py:63: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
return f(*args, **kwargs)
```

METHODOLOGY

The input audio signal is loaded using the librosa package. The input audio signal's audio waves are clipped to minimum of the audio length found in the dataset. The audio labels are obtained from the name of the audio file. MFCC features are

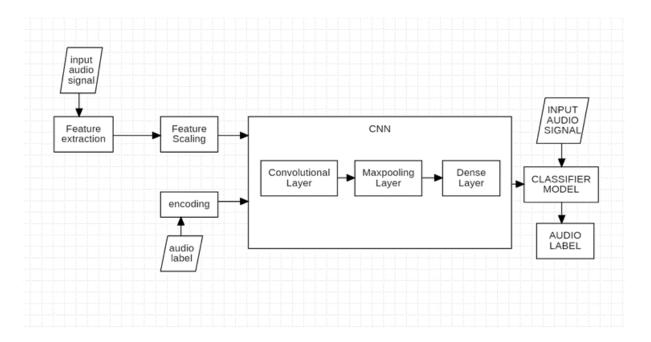
extracted from the audio waves and it is scaled. The audio labels are one hot encoded. The MFCC features and one hot encoded audio labels are X and Y respectively and they are split into train and test sets.

CNN model is built with different variations like 2,4,6 layers with different dropouts added after each output dense layer with 128 and 256 nodes in the output dense layer.

X train and X test are used to train the model.Y train and Y test are used to validate the CNN model.

The training of the CNN models happens for 20 epochs with accuracy, loss, validation accuracy and validation loss as metric,

BLOCK DIAGRAM



IMPLEMENTATION

```
: # CNN with 2 layers
  #test-train split and reshape to 4D
  X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
  X_train = X_train.reshape(X_train.shape[0], X_train.shape[1], X_train.shape[2], 1)
X_test = X_test.reshape(X_test.shape[0], X_test.shape[1], X_test.shape[2], 1)
  input_shape = (X_train.shape[1], X_train.shape[2], 1)
  model = Sequential()
  model.add(Conv2D(40, (3, 3), activation='relu', strides=(1, 1),
  padding='same', input_shape=input_shape))
model.add(Flatten())
  model.add(Dense(17, activation='softmax'))
  model.compile(loss='categorical_crossentropy',
        optimizer='adam',
metrics=['acc'])
  \label{eq:history = model.fit(X_train, y_train, epochs=20, validation_data=(X_test, y_test))} \\
  predictions = model.predict(X test)
  predictions = np.argmax(predictions, axis=1)
  y_test = np.argmax(y_test, axis=1)
        confusion_matrix(y_test, predictions)
  plt.figure(figsize=(8,8))
sns.heatmap(cm, annot=True, xticklabels=label_encoder.classes_, yticklabels=label_encoder.classes_, fmt='d', cmap=plt.cm.Blues, c
  plt.xlabel('Predicted Label')
  plt.ylabel('True Label')
  plt.show()
  \label{from:sklearn.metrics} \textbf{import} \ \ \text{accuracy\_score}
  print("accuracy for 2 layers : ",accuracy score(y test,predictions))
```

```
Epoch 1/20
26/26 [====
              =========] - 9s 307ms/step - loss: 28.0531 - acc: 0.0711 - val loss: 10.2035 - val acc: 0.2206
Epoch 2/20
26/26 [====
         Epoch 3/20
            26/26 [===:
Epoch 4/20
26/26 [====
         Epoch 5/20
26/26 [====
            ==========] - 5s 205ms/step - loss: 0.1762 - acc: 0.9877 - val loss: 0.1420 - val acc: 0.9951
Epoch 6/20
26/26 [====
           :==========] - 5s 207ms/step - loss: 0.1089 - acc: 0.9951 - val loss: 0.1500 - val acc: 0.9804
Epoch 7/20
            :=========] - 5s 201ms/step - loss: 0.1037 - acc: 0.9828 - val loss: 0.1127 - val acc: 0.9902
26/26 [====
Epoch 8/20
26/26 [====
            =========] - 5s 207ms/step - loss: 0.0549 - acc: 1.0000 - val loss: 0.0847 - val acc: 0.9902
Epoch 9/20
           26/26 [====
Epoch 10/20
26/26 [====
          :==========] - 5s 204ms/step - loss: 0.0326 - acc: 1.0000 - val_loss: 0.0643 - val_acc: 1.0000
Epoch 11/20
             26/26 [=====
Epoch 12/20
26/26 [====
           ========] - 6s 213ms/step - loss: 0.0191 - acc: 1.0000 - val_loss: 0.0464 - val_acc: 1.0000
Fnoch 13/20
             26/26 [====
Epoch 14/20
26/26 [====
         Fnoch 15/20
26/26 [====
              ========= 1 - 5s 208ms/step - loss: 0.0115 - acc: 1.0000 - val loss: 0.0363 - val acc: 1.0000
Epoch 16/20
26/26 [====
            =========] - 5s 207ms/step - loss: 0.0103 - acc: 1.0000 - val_loss: 0.0349 - val_acc: 1.0000
Fnoch 17/20
26/26 [====
             Epoch 18/20
26/26 [====
             ========] - 5s 205ms/step - loss: 0.0060 - acc: 1.0000 - val_loss: 0.0273 - val_acc: 1.0000
Epoch 19/20
            ==========] - 5s 200ms/step - loss: 0.0043 - acc: 1.0000 - val_loss: 0.0206 - val_acc: 1.0000
26/26 [====
Epoch 20/20
26/26 [============= - 6s 216ms/step - loss: 0.0032 - acc: 1.0000 - val_loss: 0.0219 - val_acc: 0.9951
  r_____1
```

RESULTS

Sno	No of Layers	No of Nodes in output dense layer	Dropout	Accuracy
1	2	17	-	0.9951
2	2	17	0.05	0.196
3	2	17	0.1	0.9951
4	2	17	0.2	0.9803
5	4	128	-	0.299
6	4	128	0.05	0.2205
7	4	128	0.1	0.2401
8	4	128	0.2	0.769
9	4	256	0.05	0.985
10	4	256	0.1	0.975
11	4	256	0.2	0.9705
12	6	128	-	0.9852
13	6	128	0.05	0.9950
14	6	128	0.1	0.9705
15	6	128	0.2	0.9705
16	6	256	0.05	0.9950
17	6	256	0.1	0.9509
18	6	256	0.2	0.9509

EPOCHS:

1)

```
Epoch 1/20
26/26 [======] - 23s 875ms/step - loss: 2.6034 - acc: 0.1777 -
val loss: 1.9584 - val acc: 0.3284
Epoch 2/20
val loss: 1.2063 - val acc: 0.5784
Epoch 3/20
val loss: 0.5226 - val acc: 0.8676
Epoch 4/20
26/26 [======] - 19s 732ms/step - loss: 0.3254 - acc: 0.9056 -
val_loss: 0.4464 - val_acc: 0.8676
Epoch 5/20
26/26 [====== 0.1990 - acc: 0.9620 -
val_loss: 0.2265 - val_acc: 0.9461
Epoch 6/20
val loss: 0.2880 - val acc: 0.8824
Epoch 7/20
val loss: 0.2015 - val acc: 0.9412
Epoch 8/20
26/26 [======] - 19s 727ms/step - loss: 0.0401 - acc: 0.9988 -
val loss: 0.1753 - val acc: 0.9265
Epoch 9/20
26/26 [======] - 19s 730ms/step - loss: 0.0186 - acc: 1.0000 -
val loss: 0.0979 - val acc: 0.9657
```

```
Epoch 10/20
26/26 [======
           val loss: 0.0777 - val acc: 0.9755
Epoch 11/20
26/26 [====== 0.0083 - acc: 1.0000 -
val loss: 0.0809 - val acc: 0.9804
Epoch 12/20
26/26 [======= 0.0065 - acc: 1.0000 -
val loss: 0.0693 - val acc: 0.9804
Epoch 13/20
val_loss: 0.0715 - val_acc: 0.9755
Epoch 14/20
val_loss: 0.0692 - val_acc: 0.9804
Epoch 15/20
26/26 [======= 0.0037 - acc: 1.0000 -
val loss: 0.0687 - val acc: 0.9755
Epoch 16/20
26/26 [=====
        val loss: 0.0609 - val acc: 0.9804
Epoch 17/20
26/26 [======] - 19s 736ms/step - loss: 0.0026 - acc: 1.0000 -
val_loss: 0.0668 - val_acc: 0.9804
Epoch 18/20
val_loss: 0.0654 - val_acc: 0.9755
Epoch 19/20
26/26 [======] - 19s 718ms/step - loss: 0.0022 - acc: 1.0000 -
```

val_loss: 0.0608 - val_acc: 0.9804

```
Epoch 20/20
26/26 [======
           val loss: 0.0600 - val acc: 0.9755
7/7 [=====] - 1s 84ms/step
2)
Epoch 1/20
26/26 [======] - 19s 708ms/step - loss: 2.8214 - acc: 0.0809 -
val loss: 2.7448 - val acc: 0.1029
Epoch 2/20
        26/26 [====
val loss: 2.1445 - val acc: 0.4608
Epoch 3/20
26/26 [======] - 18s 694ms/step - loss: 1.6607 - acc: 0.5404 -
val loss: 1.3396 - val acc: 0.7010
Epoch 4/20
26/26 [===========] - 18s 691ms/step - loss: 0.9209 - acc: 0.7806 -
val_loss: 1.0457 - val_acc: 0.5931
Epoch 5/20
26/26 [======= 0.6260 - acc: 0.8395 -
val loss: 0.6779 - val acc: 0.7794
Epoch 6/20
val loss: 0.3905 - val acc: 0.9216
Epoch 7/20
26/26 [======] - 18s 710ms/step - loss: 0.2521 - acc: 0.9350 -
val_loss: 0.3625 - val_acc: 0.8922
Epoch 8/20
```

```
26/26 [======] - 18s 703ms/step - loss: 0.2122 - acc: 0.9522 -
val_loss: 0.2730 - val_acc: 0.9265
Epoch 9/20
26/26 [====== 0.1615 - acc: 0.9681 -
val_loss: 0.2458 - val_acc: 0.9510
Epoch 10/20
26/26 [======= 0.0945 - acc: 0.9914 -
val loss: 0.2179 - val acc: 0.9461
Epoch 11/20
26/26 [======] - 18s 698ms/step - loss: 0.0595 - acc: 0.9963 -
val_loss: 0.1603 - val_acc: 0.9657
Epoch 12/20
val loss: 0.1514 - val acc: 0.9706
Epoch 13/20
val_loss: 0.1273 - val_acc: 0.9853
Epoch 14/20
26/26 [======= 0.0350 - acc: 0.9963 -
val loss: 0.1362 - val acc: 0.9559
Epoch 15/20
26/26 [======= - loss: 0.0230 - acc: 1.0000 -
val loss: 0.0958 - val acc: 0.9853
Epoch 16/20
val_loss: 0.0849 - val_acc: 0.9853
Epoch 17/20
val_loss: 0.0839 - val_acc: 0.9804
Epoch 18/20
```

```
26/26 [======] - 19s 726ms/step - loss: 0.0111 - acc: 1.0000 -
val_loss: 0.0707 - val_acc: 0.9853
Epoch 19/20
26/26 [======= - 20s 781ms/step - loss: 0.0089 - acc: 1.0000 -
val_loss: 0.0656 - val_acc: 0.9902
Epoch 20/20
26/26 [====== 0.0076 - acc: 1.0000 -
val loss: 0.0658 - val acc: 0.9902
7/7 [======] - 1s 87ms/step
3)
Epoch 1/20
26/26 [======] - 20s 750ms/step - loss: 2.7609 - acc: 0.1446 -
val loss: 2.3565 - val acc: 0.3088
Epoch 2/20
26/26 [======] - 19s 719ms/step - loss: 1.7009 - acc: 0.5196 -
val loss: 1.2368 - val acc: 0.6863
Epoch 3/20
26/26 [======= 0.8284 - 19s 730ms/step - loss: 0.7227 - acc: 0.8284 -
val_loss: 0.6919 - val_acc: 0.8824
Epoch 4/20
26/26 [========
                          ======] - 19s 725ms/step - loss: 0.4580 - acc: 0.8836 -
val loss: 0.5551 - val acc: 0.8824
Epoch 5/20
26/26 [======= 0.2566 - acc: 0.9400 -
val loss: 0.2723 - val acc: 0.9412
Epoch 6/20
val loss: 0.2276 - val acc: 0.9510
```

```
Epoch 7/20
26/26 [======
            val loss: 0.1700 - val acc: 0.9510
Epoch 8/20
26/26 [====== 0.0515 - acc: 0.9975 -
val loss: 0.1426 - val acc: 0.9706
Epoch 9/20
26/26 [======] - 19s 721ms/step - loss: 0.0379 - acc: 0.9988 -
val loss: 0.1420 - val acc: 0.9706
Epoch 10/20
val_loss: 0.1554 - val_acc: 0.9608
Epoch 11/20
val_loss: 0.0799 - val_acc: 0.9853
Epoch 12/20
26/26 [======= - 19s 720ms/step - loss: 0.0138 - acc: 1.0000 -
val loss: 0.0644 - val acc: 0.9902
Epoch 13/20
26/26 [======
          val loss: 0.0641 - val acc: 0.9902
Epoch 14/20
26/26 [======= - loss: 0.0093 - acc: 1.0000 -
val_loss: 0.0632 - val_acc: 0.9853
Epoch 15/20
val_loss: 0.0533 - val_acc: 0.9951
Epoch 16/20
26/26 [======] - 19s 718ms/step - loss: 0.0057 - acc: 1.0000 -
val_loss: 0.0548 - val_acc: 0.9804
```

```
Epoch 17/20
val loss: 0.0708 - val acc: 0.9804
Epoch 18/20
26/26 [======] - 19s 713ms/step - loss: 0.0036 - acc: 1.0000 -
val loss: 0.0491 - val acc: 0.9902
Epoch 19/20
val loss: 0.0559 - val acc: 0.9902
Epoch 20/20
26/26 [======] - 20s 760ms/step - loss: 0.0029 - acc: 1.0000 -
val_loss: 0.0568 - val_acc: 0.9853
7/7 [======] - 1s 80ms/step
4)
Epoch 1/20
26/26 [======] - 20s 752ms/step - loss: 2.7776 - acc: 0.1434 -
val_loss: 2.5612 - val_acc: 0.2255
Epoch 2/20
26/26 [======] - 19s 731ms/step - loss: 1.9435 - acc: 0.4620 -
val loss: 1.4274 - val acc: 0.7059
Epoch 3/20
26/26 [======= 0.7806 - 19s 726ms/step - loss: 0.8532 - acc: 0.7806 -
val loss: 0.6846 - val acc: 0.8088
Epoch 4/20
26/26 [======] - 19s 732ms/step - loss: 0.4381 - acc: 0.8971 -
val_loss: 0.3820 - val_acc: 0.9363
Epoch 5/20
```

```
26/26 [======] - 20s 752ms/step - loss: 0.1936 - acc: 0.9730 -
val_loss: 0.2048 - val_acc: 0.9853
Epoch 6/20
26/26 [====== 0.1036 - acc: 0.9828 -
val_loss: 0.1919 - val_acc: 0.9608
Epoch 7/20
26/26 [======= 0.0599 - acc: 0.9963 -
val loss: 0.1089 - val acc: 1.0000
Epoch 8/20
26/26 [======] - 18s 703ms/step - loss: 0.0311 - acc: 0.9988 -
val_loss: 0.0992 - val_acc: 0.9706
Epoch 9/20
val loss: 0.0709 - val acc: 1.0000
Epoch 10/20
val_loss: 0.0741 - val_acc: 0.9902
Epoch 11/20
26/26 [======= 0.0166 - acc: 1.0000 -
val loss: 0.0678 - val acc: 0.9902
Epoch 12/20
26/26 [======= - 20s 751ms/step - loss: 0.0098 - acc: 1.0000 -
val loss: 0.0543 - val acc: 0.9804
Epoch 13/20
26/26 [======= 0.0055 - acc: 1.0000 -
val_loss: 0.0473 - val_acc: 0.9951
Epoch 14/20
val_loss: 0.0484 - val_acc: 0.9853
Epoch 15/20
```

```
26/26 [======] - 21s 792ms/step - loss: 0.0037 - acc: 1.0000 -
val_loss: 0.0413 - val_acc: 0.9951
Epoch 16/20
26/26 [======= 0.0032 - acc: 1.0000 -
val_loss: 0.0385 - val_acc: 0.9951
Epoch 17/20
26/26 [======= - 20s 766ms/step - loss: 0.0029 - acc: 1.0000 -
val loss: 0.0409 - val acc: 0.9902
Epoch 18/20
26/26 [======= 0.0023 - acc: 1.0000 -
val_loss: 0.0404 - val_acc: 0.9902
Epoch 19/20
26/26 [=======0.0000 - 20s 775ms/step - loss: 0.0020 - acc: 1.0000 -
val loss: 0.0406 - val acc: 0.9853
Epoch 20/20
26/26 [=======0.0018 - acc: 1.0000 -
val loss: 0.0382 - val acc: 0.9853
7/7 [======] - 1s 87ms/step
5)
Epoch 1/20
26/26 [======
                        ======| - 21s 796ms/step - loss: 3.2300 - acc: 0.0637 -
val loss: 2.8229 - val acc: 0.1176
Epoch 2/20
26/26 [======] - 20s 771ms/step - loss: 2.6881 - acc: 0.1434 -
val loss: 2.5127 - val acc: 0.2647
Epoch 3/20
val loss: 1.7350 - val acc: 0.4951
```

```
Epoch 4/20
26/26 [======
             val loss: 1.1825 - val acc: 0.7402
Epoch 5/20
26/26 [====== 0.8746 - acc: 0.7610 -
val loss: 0.8826 - val acc: 0.7304
Epoch 6/20
val loss: 0.6829 - val acc: 0.8431
Epoch 7/20
val_loss: 0.5078 - val_acc: 0.8971
Epoch 8/20
val_loss: 0.4977 - val_acc: 0.8676
Epoch 9/20
26/26 [======= 0.2652 - acc: 0.9216 -
val loss: 0.3912 - val acc: 0.9069
Epoch 10/20
26/26 [======
                     =====] - 20s 789ms/step - loss: 0.1772 - acc: 0.9620 -
val loss: 0.2712 - val acc: 0.9657
Epoch 11/20
26/26 [============] - 20s 773ms/step - loss: 0.1142 - acc: 0.9877 -
val_loss: 0.2480 - val_acc: 0.9461
Epoch 12/20
26/26 [=======0.0886 - acc: 0.9877 -
val_loss: 0.2971 - val_acc: 0.9167
Epoch 13/20
26/26 [======] - 20s 778ms/step - loss: 0.0689 - acc: 0.9939 -
val_loss: 0.1689 - val_acc: 0.9510
```

```
Epoch 14/20
26/26 [======] - 20s 778ms/step - loss: 0.0426 - acc: 0.9975 -
val loss: 0.1272 - val acc: 0.9755
Epoch 15/20
26/26 [======] - 20s 780ms/step - loss: 0.0286 - acc: 0.9975 -
val loss: 0.1349 - val acc: 0.9657
Epoch 16/20
val loss: 0.1362 - val acc: 0.9657
Epoch 17/20
val_loss: 0.1010 - val_acc: 0.9706
Epoch 18/20
val_loss: 0.0944 - val_acc: 0.9755
Epoch 19/20
26/26 [======= - 20s 783ms/step - loss: 0.0141 - acc: 1.0000 -
val loss: 0.0853 - val acc: 0.9853
Epoch 20/20
26/26 [======] - 20s 777ms/step - loss: 0.0107 - acc: 1.0000 -
val loss: 0.0812 - val acc: 0.9804
7/7 [=====] - 1s 90ms/step
6)
Epoch 1/20
val_loss: 2.7455 - val_acc: 0.1225
Epoch 2/20
```

```
26/26 [======] - 12s 466ms/step - loss: 2.4404 - acc: 0.2843 -
val_loss: 1.9969 - val_acc: 0.4755
Epoch 3/20
val_loss: 1.0668 - val_acc: 0.6520
Epoch 4/20
val loss: 0.5910 - val acc: 0.7794
Epoch 5/20
26/26 [======] - 13s 489ms/step - loss: 0.4539 - acc: 0.8713 -
val_loss: 0.4263 - val_acc: 0.8676
Epoch 6/20
val loss: 0.2830 - val acc: 0.9216
Epoch 7/20
val_loss: 0.1845 - val_acc: 0.9559
Epoch 8/20
26/26 [====== 0.0804 - acc: 0.9939 -
val loss: 0.1485 - val acc: 0.9608
Epoch 9/20
val loss: 0.1536 - val acc: 0.9657
Epoch 10/20
val_loss: 0.1297 - val_acc: 0.9559
Epoch 11/20
val_loss: 0.1375 - val_acc: 0.9608
```

Epoch 12/20

```
26/26 [======] - 12s 475ms/step - loss: 0.0215 - acc: 1.0000 -
val_loss: 0.1063 - val_acc: 0.9657
Epoch 13/20
26/26 [====== loss: 0.0119 - acc: 1.0000 -
val_loss: 0.1012 - val_acc: 0.9706
Epoch 14/20
26/26 [====== 0.0083 - acc: 1.0000 -
val loss: 0.0839 - val acc: 0.9706
Epoch 15/20
26/26 [======] - 12s 458ms/step - loss: 0.0067 - acc: 1.0000 -
val_loss: 0.0904 - val_acc: 0.9706
Epoch 16/20
26/26 [====== 0.0058 - acc: 1.0000 -
val loss: 0.0873 - val acc: 0.9706
Epoch 17/20
val_loss: 0.0783 - val_acc: 0.9706
Epoch 18/20
26/26 [====== 0.0037 - acc: 1.0000 -
val loss: 0.0851 - val acc: 0.9755
Epoch 19/20
26/26 [======] - 12s 460ms/step - loss: 0.0028 - acc: 1.0000 -
val loss: 0.0855 - val acc: 0.9706
Epoch 20/20
26/26 [======] - 12s 458ms/step - loss: 0.0022 - acc: 1.0000 -
val_loss: 0.0872 - val_acc: 0.9755
7/7 [======] - 1s 88ms/step
```

```
Epoch 1/20
26/26 [=====
           val loss: 1.9584 - val acc: 0.3284
Epoch 2/20
26/26 [======] - 20s 757ms/step - loss: 1.4234 - acc: 0.5625 -
val loss: 1.2063 - val acc: 0.5784
Epoch 3/20
val loss: 0.5226 - val acc: 0.8676
Epoch 4/20
val_loss: 0.4464 - val_acc: 0.8676
Epoch 5/20
val_loss: 0.2265 - val_acc: 0.9461
Epoch 6/20
26/26 [======= 0.1003 - acc: 0.9804 -
val loss: 0.2880 - val acc: 0.8824
Epoch 7/20
26/26 [=====
           val loss: 0.2015 - val acc: 0.9412
Epoch 8/20
26/26 [======] - 19s 727ms/step - loss: 0.0401 - acc: 0.9988 -
val_loss: 0.1753 - val_acc: 0.9265
Epoch 9/20
val_loss: 0.0979 - val_acc: 0.9657
Epoch 10/20
26/26 [======] - 19s 727ms/step - loss: 0.0120 - acc: 1.0000 -
val_loss: 0.0777 - val_acc: 0.9755
```

```
Epoch 11/20
26/26 [======
           val loss: 0.0809 - val acc: 0.9804
Epoch 12/20
26/26 [====== 0.0065 - acc: 1.0000 -
val loss: 0.0693 - val acc: 0.9804
Epoch 13/20
26/26 [======= 0.0055 - acc: 1.0000 -
val loss: 0.0715 - val acc: 0.9755
Epoch 14/20
val_loss: 0.0692 - val_acc: 0.9804
Epoch 15/20
val_loss: 0.0687 - val_acc: 0.9755
Epoch 16/20
26/26 [======= 0.0030 - acc: 1.0000 -
val loss: 0.0609 - val acc: 0.9804
Epoch 17/20
26/26 [======= - loss: 0.0026 - acc: 1.0000 -
val loss: 0.0668 - val acc: 0.9804
Epoch 18/20
26/26 [======= - loss: 0.0025 - acc: 1.0000 -
val_loss: 0.0654 - val_acc: 0.9755
Epoch 19/20
val_loss: 0.0608 - val_acc: 0.9804
Epoch 20/20
26/26 [======] - 19s 712ms/step - loss: 0.0020 - acc: 1.0000 -
```

val_loss: 0.0600 - val_acc: 0.9755

```
7/7 [=====] - 1s 84ms/step
```

8)

Epoch 1/20

26/26 [======] - 19s 708ms/step - loss: 2.8214 - acc: 0.0809 -

val_loss: 2.7448 - val_acc: 0.1029

Epoch 2/20

26/26 [======] - 18s 691ms/step - loss: 2.5146 - acc: 0.2034 -

val_loss: 2.1445 - val_acc: 0.4608

Epoch 3/20

26/26 [======] - 18s 694ms/step - loss: 1.6607 - acc: 0.5404 -

val_loss: 1.3396 - val_acc: 0.7010

Epoch 4/20

26/26 [===========] - 18s 691ms/step - loss: 0.9209 - acc: 0.7806 -

val_loss: 1.0457 - val_acc: 0.5931

Epoch 5/20

26/26 [======= 0.6260 - acc: 0.8395 -

val_loss: 0.6779 - val_acc: 0.7794

Epoch 6/20

26/26 [======= 0.3775 - acc: 0.9105 -

val_loss: 0.3905 - val_acc: 0.9216

Epoch 7/20

val loss: 0.3625 - val acc: 0.8922

Epoch 8/20

26/26 [======] - 18s 703ms/step - loss: 0.2122 - acc: 0.9522 -

val_loss: 0.2730 - val_acc: 0.9265

Epoch 9/20

```
26/26 [======] - 18s 708ms/step - loss: 0.1615 - acc: 0.9681 -
val_loss: 0.2458 - val_acc: 0.9510
Epoch 10/20
26/26 [====== loss: 0.0945 - acc: 0.9914 -
val_loss: 0.2179 - val_acc: 0.9461
Epoch 11/20
26/26 [======= 0.0595 - acc: 0.9963 -
val loss: 0.1603 - val acc: 0.9657
Epoch 12/20
26/26 [======] - 18s 706ms/step - loss: 0.0592 - acc: 0.9926 -
val_loss: 0.1514 - val_acc: 0.9706
Epoch 13/20
val loss: 0.1273 - val acc: 0.9853
Epoch 14/20
val_loss: 0.1362 - val_acc: 0.9559
Epoch 15/20
val loss: 0.0958 - val acc: 0.9853
Epoch 16/20
val loss: 0.0849 - val acc: 0.9853
Epoch 17/20
26/26 [======= - 18s 707ms/step - loss: 0.0126 - acc: 1.0000 -
val_loss: 0.0839 - val_acc: 0.9804
Epoch 18/20
26/26 [======= | - 19s 726ms/step - loss: 0.0111 - acc: 1.0000 -
val_loss: 0.0707 - val_acc: 0.9853
Epoch 19/20
```

```
26/26 [======] - 20s 781ms/step - loss: 0.0089 - acc: 1.0000 -
val_loss: 0.0656 - val_acc: 0.9902
Epoch 20/20
26/26 [====== 0.0076 - acc: 1.0000 -
val loss: 0.0658 - val acc: 0.9902
7/7 [=====] - 1s 87ms/step
9)
Epoch 1/20
val loss: 2.3565 - val acc: 0.3088
Epoch 2/20
26/26 [======] - 19s 719ms/step - loss: 1.7009 - acc: 0.5196 -
val loss: 1.2368 - val acc: 0.6863
Epoch 3/20
26/26 [======] - 19s 730ms/step - loss: 0.7227 - acc: 0.8284 -
val loss: 0.6919 - val acc: 0.8824
Epoch 4/20
26/26 [======= 0.4580 - acc: 0.8836 -
val_loss: 0.5551 - val_acc: 0.8824
Epoch 5/20
26/26 [=======
                   val loss: 0.2723 - val acc: 0.9412
Epoch 6/20
26/26 [======= 0.1292 - acc: 0.9816 -
val loss: 0.2276 - val acc: 0.9510
Epoch 7/20
val loss: 0.1700 - val acc: 0.9510
```

```
Epoch 8/20
26/26 [======
            val loss: 0.1426 - val acc: 0.9706
Epoch 9/20
26/26 [======] - 19s 721ms/step - loss: 0.0379 - acc: 0.9988 -
val loss: 0.1420 - val acc: 0.9706
Epoch 10/20
26/26 [======] - 19s 715ms/step - loss: 0.0286 - acc: 0.9988 -
val loss: 0.1554 - val acc: 0.9608
Epoch 11/20
val_loss: 0.0799 - val_acc: 0.9853
Epoch 12/20
val_loss: 0.0644 - val_acc: 0.9902
Epoch 13/20
val loss: 0.0641 - val acc: 0.9902
Epoch 14/20
26/26 [======
            val loss: 0.0632 - val acc: 0.9853
Epoch 15/20
26/26 [===========] - 19s 715ms/step - loss: 0.0070 - acc: 1.0000 -
val_loss: 0.0533 - val_acc: 0.9951
Epoch 16/20
val_loss: 0.0548 - val_acc: 0.9804
Epoch 17/20
26/26 [======] - 19s 716ms/step - loss: 0.0045 - acc: 1.0000 -
```

val_loss: 0.0708 - val_acc: 0.9804

```
Epoch 18/20
26/26 [======
             ========= | - 19s 713ms/step - loss: 0.0036 - acc: 1.0000 -
val loss: 0.0491 - val acc: 0.9902
Epoch 19/20
26/26 [======] - 19s 728ms/step - loss: 0.0032 - acc: 1.0000 -
val loss: 0.0559 - val acc: 0.9902
Epoch 20/20
26/26 [======] - 20s 760ms/step - loss: 0.0029 - acc: 1.0000 -
val loss: 0.0568 - val acc: 0.9853
7/7 [======] - 1s 80ms/step
10)
Epoch 1/20
             26/26 [=====
val loss: 2.5612 - val acc: 0.2255
Epoch 2/20
26/26 [======] - 19s 731ms/step - loss: 1.9435 - acc: 0.4620 -
val_loss: 1.4274 - val_acc: 0.7059
Epoch 3/20
26/26 [======] - 19s 726ms/step - loss: 0.8532 - acc: 0.7806 -
val loss: 0.6846 - val acc: 0.8088
Epoch 4/20
val loss: 0.3820 - val acc: 0.9363
Epoch 5/20
26/26 [======] - 20s 752ms/step - loss: 0.1936 - acc: 0.9730 -
val_loss: 0.2048 - val_acc: 0.9853
Epoch 6/20
```

```
26/26 [======] - 19s 728ms/step - loss: 0.1036 - acc: 0.9828 -
val_loss: 0.1919 - val_acc: 0.9608
Epoch 7/20
26/26 [====== loss: 0.0599 - acc: 0.9963 -
val_loss: 0.1089 - val_acc: 1.0000
Epoch 8/20
26/26 [======] - 18s 703ms/step - loss: 0.0311 - acc: 0.9988 -
val loss: 0.0992 - val acc: 0.9706
Epoch 9/20
26/26 [======] - 18s 705ms/step - loss: 0.0214 - acc: 0.9988 -
val_loss: 0.0709 - val_acc: 1.0000
Epoch 10/20
val loss: 0.0741 - val acc: 0.9902
Epoch 11/20
26/26 [====== 0.0166 - acc: 1.0000 -
val_loss: 0.0678 - val_acc: 0.9902
Epoch 12/20
26/26 [======= 0.0098 - acc: 1.0000 -
val loss: 0.0543 - val acc: 0.9804
Epoch 13/20
val loss: 0.0473 - val acc: 0.9951
Epoch 14/20
26/26 [======= - 20s 769ms/step - loss: 0.0043 - acc: 1.0000 -
val_loss: 0.0484 - val_acc: 0.9853
Epoch 15/20
val_loss: 0.0413 - val_acc: 0.9951
```

Epoch 16/20

```
26/26 [======] - 20s 768ms/step - loss: 0.0032 - acc: 1.0000 -
val_loss: 0.0385 - val_acc: 0.9951
Epoch 17/20
26/26 [====== 0.0029 - acc: 1.0000 -
val_loss: 0.0409 - val_acc: 0.9902
Epoch 18/20
26/26 [======= 0.0023 - acc: 1.0000 -
val loss: 0.0404 - val acc: 0.9902
Epoch 19/20
26/26 [======] - 20s 775ms/step - loss: 0.0020 - acc: 1.0000 -
val_loss: 0.0406 - val_acc: 0.9853
Epoch 20/20
26/26 [====== 0.0018 - acc: 1.0000 -
val loss: 0.0382 - val acc: 0.9853
7/7 [======] - 1s 87ms/step
11)
Epoch 1/20
26/26 [======] - 21s 796ms/step - loss: 3.2300 - acc: 0.0637 -
val_loss: 2.8229 - val_acc: 0.1176
Epoch 2/20
26/26 [======
                         ======] - 20s 771ms/step - loss: 2.6881 - acc: 0.1434 -
val loss: 2.5127 - val acc: 0.2647
Epoch 3/20
26/26 [======] - 20s 766ms/step - loss: 2.1087 - acc: 0.3309 -
val loss: 1.7350 - val acc: 0.4951
Epoch 4/20
val loss: 1.1825 - val acc: 0.7402
```

```
Epoch 5/20
26/26 [======
              val loss: 0.8826 - val acc: 0.7304
Epoch 6/20
26/26 [======] - 20s 780ms/step - loss: 0.6816 - acc: 0.7978 -
val loss: 0.6829 - val acc: 0.8431
Epoch 7/20
26/26 [======] - 20s 772ms/step - loss: 0.4446 - acc: 0.8958 -
val loss: 0.5078 - val acc: 0.8971
Epoch 8/20
26/26 [=======0.3906 - acc: 0.8885 -
val_loss: 0.4977 - val_acc: 0.8676
Epoch 9/20
26/26 [==========
                      =======| - 20s 785ms/step - loss: 0.2652 - acc: 0.9216 -
val_loss: 0.3912 - val_acc: 0.9069
Epoch 10/20
26/26 [======= 0.1772 - acc: 0.9620 -
val loss: 0.2712 - val acc: 0.9657
Epoch 11/20
26/26 [======
              val loss: 0.2480 - val acc: 0.9461
Epoch 12/20
26/26 [============] - 20s 777ms/step - loss: 0.0886 - acc: 0.9877 -
val_loss: 0.2971 - val_acc: 0.9167
Epoch 13/20
val_loss: 0.1689 - val_acc: 0.9510
Epoch 14/20
26/26 [======] - 20s 778ms/step - loss: 0.0426 - acc: 0.9975 -
val_loss: 0.1272 - val_acc: 0.9755
```

```
Epoch 15/20
val loss: 0.1349 - val acc: 0.9657
Epoch 16/20
26/26 [======] - 20s 782ms/step - loss: 0.0254 - acc: 0.9988 -
val loss: 0.1362 - val acc: 0.9657
Epoch 17/20
val_loss: 0.1010 - val_acc: 0.9706
Epoch 18/20
26/26 [======] - 20s 777ms/step - loss: 0.0143 - acc: 1.0000 -
val_loss: 0.0944 - val_acc: 0.9755
Epoch 19/20
val_loss: 0.0853 - val_acc: 0.9853
Epoch 20/20
26/26 [======= - 20s 777ms/step - loss: 0.0107 - acc: 1.0000 -
val loss: 0.0812 - val acc: 0.9804
7/7 [=====] - 1s 90ms/step
12)
Epoch 1/20
val loss: 2.7455 - val acc: 0.1225
Epoch 2/20
val_loss: 1.9969 - val_acc: 0.4755
Epoch 3/20
```

```
val_loss: 1.0668 - val_acc: 0.6520
Epoch 4/20
26/26 [======= 0.7737 - acc: 0.7794 -
val_loss: 0.5910 - val_acc: 0.7794
Epoch 5/20
26/26 [======= 0.4539 - acc: 0.8713 -
val loss: 0.4263 - val acc: 0.8676
Epoch 6/20
26/26 [======] - 12s 476ms/step - loss: 0.2768 - acc: 0.9363 -
val_loss: 0.2830 - val_acc: 0.9216
Epoch 7/20
val loss: 0.1845 - val acc: 0.9559
Epoch 8/20
val_loss: 0.1485 - val_acc: 0.9608
Epoch 9/20
26/26 [======= 0.0831 - acc: 0.9779 -
val loss: 0.1536 - val acc: 0.9657
Epoch 10/20
val loss: 0.1297 - val acc: 0.9559
Epoch 11/20
26/26 [======= 0.0442 - acc: 0.9914 -
val_loss: 0.1375 - val_acc: 0.9608
Epoch 12/20
val_loss: 0.1063 - val_acc: 0.9657
Epoch 13/20
```

```
26/26 [======] - 12s 464ms/step - loss: 0.0119 - acc: 1.0000 -
val_loss: 0.1012 - val_acc: 0.9706
Epoch 14/20
26/26 [====== 0.0083 - acc: 1.0000 -
val_loss: 0.0839 - val_acc: 0.9706
Epoch 15/20
26/26 [======] - 12s 458ms/step - loss: 0.0067 - acc: 1.0000 -
val loss: 0.0904 - val acc: 0.9706
Epoch 16/20
26/26 [======] - 12s 476ms/step - loss: 0.0058 - acc: 1.0000 -
val_loss: 0.0873 - val_acc: 0.9706
Epoch 17/20
26/26 [======= 0.0047 - acc: 1.0000 -
val loss: 0.0783 - val acc: 0.9706
Epoch 18/20
val_loss: 0.0851 - val_acc: 0.9755
Epoch 19/20
26/26 [====== | - 12s 460ms/step - loss: 0.0028 - acc: 1.0000 -
val loss: 0.0855 - val acc: 0.9706
Epoch 20/20
26/26 [======] - 12s 458ms/step - loss: 0.0022 - acc: 1.0000 -
val loss: 0.0872 - val acc: 0.9755
7/7 [======] - 1s 88ms/step
13)
Epoch 1/20
val loss: 1.9584 - val acc: 0.3284
```

```
Epoch 2/20
26/26 [======
              val loss: 1.2063 - val acc: 0.5784
Epoch 3/20
26/26 [====== 0.6567 - acc: 0.8113 -
val loss: 0.5226 - val acc: 0.8676
Epoch 4/20
26/26 [=======] - 19s 732ms/step - loss: 0.3254 - acc: 0.9056 -
val loss: 0.4464 - val acc: 0.8676
Epoch 5/20
val_loss: 0.2265 - val_acc: 0.9461
Epoch 6/20
26/26 [=======0.0000 - 19s 741ms/step - loss: 0.1003 - acc: 0.9804 -
val_loss: 0.2880 - val_acc: 0.8824
Epoch 7/20
26/26 [======= 0.0950 - acc: 0.9853 -
val loss: 0.2015 - val acc: 0.9412
Epoch 8/20
26/26 [===
              ========| - 19s 727ms/step - loss: 0.0401 - acc: 0.9988 -
val loss: 0.1753 - val acc: 0.9265
Epoch 9/20
26/26 [======] - 19s 730ms/step - loss: 0.0186 - acc: 1.0000 -
val_loss: 0.0979 - val_acc: 0.9657
Epoch 10/20
val_loss: 0.0777 - val_acc: 0.9755
Epoch 11/20
26/26 [====== 0.0083 - acc: 1.0000 -
val_loss: 0.0809 - val_acc: 0.9804
```

```
Epoch 12/20
26/26 [======= 0.0065 - acc: 1.0000 -
val loss: 0.0693 - val acc: 0.9804
Epoch 13/20
26/26 [====== 0.0055 - acc: 1.0000 -
val loss: 0.0715 - val acc: 0.9755
Epoch 14/20
26/26 [======] - 19s 736ms/step - loss: 0.0041 - acc: 1.0000 -
val loss: 0.0692 - val acc: 0.9804
Epoch 15/20
val_loss: 0.0687 - val_acc: 0.9755
Epoch 16/20
val_loss: 0.0609 - val_acc: 0.9804
Epoch 17/20
26/26 [======= 0.0026 - acc: 1.0000 -
val loss: 0.0668 - val acc: 0.9804
Epoch 18/20
val loss: 0.0654 - val acc: 0.9755
Epoch 19/20
26/26 [======= - loss: 0.0022 - acc: 1.0000 -
val_loss: 0.0608 - val_acc: 0.9804
Epoch 20/20
26/26 [======= 0.0020 - acc: 1.0000 - acc: 1
val loss: 0.0600 - val acc: 0.9755
```

7/7 [=====] - 1s 84ms/step

```
Epoch 1/20
26/26 [======
             val loss: 2.7448 - val acc: 0.1029
Epoch 2/20
26/26 [======] - 18s 691ms/step - loss: 2.5146 - acc: 0.2034 -
val loss: 2.1445 - val acc: 0.4608
Epoch 3/20
val loss: 1.3396 - val acc: 0.7010
Epoch 4/20
val_loss: 1.0457 - val_acc: 0.5931
Epoch 5/20
val_loss: 0.6779 - val_acc: 0.7794
Epoch 6/20
26/26 [======= 0.3775 - acc: 0.9105 -
val_loss: 0.3905 - val_acc: 0.9216
Epoch 7/20
26/26 [====
            =======| - 18s 710ms/step - loss: 0.2521 - acc: 0.9350 -
val loss: 0.3625 - val acc: 0.8922
Epoch 8/20
26/26 [======] - 18s 703ms/step - loss: 0.2122 - acc: 0.9522 -
val_loss: 0.2730 - val_acc: 0.9265
Epoch 9/20
val_loss: 0.2458 - val_acc: 0.9510
Epoch 10/20
26/26 [======] - 18s 701ms/step - loss: 0.0945 - acc: 0.9914 -
val_loss: 0.2179 - val_acc: 0.9461
```

```
Epoch 11/20
26/26 [======
                                              val loss: 0.1603 - val acc: 0.9657
Epoch 12/20
26/26 [====== 0.0592 - acc: 0.9926 -
val loss: 0.1514 - val acc: 0.9706
Epoch 13/20
26/26 [======] - 18s 697ms/step - loss: 0.0367 - acc: 1.0000 -
val loss: 0.1273 - val acc: 0.9853
Epoch 14/20
val_loss: 0.1362 - val_acc: 0.9559
Epoch 15/20
26/26 [=====
                                    ======= | - 18s 699ms/step - loss: 0.0230 - acc: 1.0000 -
val_loss: 0.0958 - val_acc: 0.9853
Epoch 16/20
val loss: 0.0849 - val acc: 0.9853
Epoch 17/20
26/26 [=====
                                      val loss: 0.0839 - val acc: 0.9804
Epoch 18/20
26/26 [======= -19s 726ms/step - loss: 0.0111 - acc: 1.0000 -
val_loss: 0.0707 - val_acc: 0.9853
Epoch 19/20
26/26 [=======0.0009 - acc: 1.0000 - acc: 1.
val_loss: 0.0656 - val_acc: 0.9902
Epoch 20/20
26/26 [======] - 20s 753ms/step - loss: 0.0076 - acc: 1.0000 -
```

val_loss: 0.0658 - val_acc: 0.9902

```
7/7 [======] - 1s 87ms/step
```

15)

Epoch 1/20

26/26 [======] - 20s 750ms/step - loss: 2.7609 - acc: 0.1446 -

val_loss: 2.3565 - val_acc: 0.3088

Epoch 2/20

26/26 [======] - 19s 719ms/step - loss: 1.7009 - acc: 0.5196 -

val_loss: 1.2368 - val_acc: 0.6863

Epoch 3/20

26/26 [=====] - 19s 730ms/step - loss: 0.7227 - acc: 0.8284 -

val_loss: 0.6919 - val_acc: 0.8824

Epoch 4/20

26/26 [===========] - 19s 725ms/step - loss: 0.4580 - acc: 0.8836 -

val loss: 0.5551 - val acc: 0.8824

Epoch 5/20

26/26 [===========] - 19s 718ms/step - loss: 0.2566 - acc: 0.9400 -

val_loss: 0.2723 - val_acc: 0.9412

Epoch 6/20

26/26 [======] - 19s 718ms/step - loss: 0.1292 - acc: 0.9816 -

val loss: 0.2276 - val acc: 0.9510

Epoch 7/20

val_loss: 0.1700 - val_acc: 0.9510

Epoch 8/20

26/26 [======] - 19s 720ms/step - loss: 0.0515 - acc: 0.9975 -

val_loss: 0.1426 - val_acc: 0.9706

Epoch 9/20

```
26/26 [======] - 19s 721ms/step - loss: 0.0379 - acc: 0.9988 -
val_loss: 0.1420 - val_acc: 0.9706
Epoch 10/20
26/26 [====== loss: 0.0286 - acc: 0.9988 -
val_loss: 0.1554 - val_acc: 0.9608
Epoch 11/20
26/26 [======= 0.0257 - acc: 0.9963 -
val loss: 0.0799 - val acc: 0.9853
Epoch 12/20
26/26 [======] - 19s 720ms/step - loss: 0.0138 - acc: 1.0000 -
val_loss: 0.0644 - val_acc: 0.9902
Epoch 13/20
val loss: 0.0641 - val acc: 0.9902
Epoch 14/20
val_loss: 0.0632 - val_acc: 0.9853
Epoch 15/20
26/26 [======= 0.0070 - acc: 1.0000 -
val loss: 0.0533 - val acc: 0.9951
Epoch 16/20
val loss: 0.0548 - val acc: 0.9804
Epoch 17/20
26/26 [======= 0.0045 - acc: 1.0000 -
val_loss: 0.0708 - val_acc: 0.9804
Epoch 18/20
26/26 [=======0.000 - 19s 713ms/step - loss: 0.0036 - acc: 1.0000 -
val_loss: 0.0491 - val_acc: 0.9902
Epoch 19/20
```

```
26/26 [======] - 19s 728ms/step - loss: 0.0032 - acc: 1.0000 -
val_loss: 0.0559 - val_acc: 0.9902
Epoch 20/20
26/26 [====== 0.0029 - acc: 1.0000 -
val loss: 0.0568 - val acc: 0.9853
7/7 [=====] - 1s 80ms/step
16)
Epoch 1/20
val loss: 2.5612 - val acc: 0.2255
Epoch 2/20
26/26 [====== | - 19s 731ms/step - loss: 1.9435 - acc: 0.4620 -
val loss: 1.4274 - val acc: 0.7059
Epoch 3/20
26/26 [===========] - 19s 726ms/step - loss: 0.8532 - acc: 0.7806 -
val loss: 0.6846 - val acc: 0.8088
Epoch 4/20
26/26 [======= 0.4381 - acc: 0.8971 -
val_loss: 0.3820 - val_acc: 0.9363
Epoch 5/20
26/26 [========
                        ======| - 20s 752ms/step - loss: 0.1936 - acc: 0.9730 -
val loss: 0.2048 - val acc: 0.9853
Epoch 6/20
26/26 [======= 0.1036 - acc: 0.9828 -
val loss: 0.1919 - val acc: 0.9608
Epoch 7/20
val loss: 0.1089 - val acc: 1.0000
```

```
Epoch 8/20
26/26 [=====
              val loss: 0.0992 - val acc: 0.9706
Epoch 9/20
26/26 [======] - 18s 705ms/step - loss: 0.0214 - acc: 0.9988 -
val loss: 0.0709 - val acc: 1.0000
Epoch 10/20
26/26 [======] - 18s 702ms/step - loss: 0.0137 - acc: 1.0000 -
val loss: 0.0741 - val acc: 0.9902
Epoch 11/20
val_loss: 0.0678 - val_acc: 0.9902
Epoch 12/20
26/26 [=======0.0008 - acc: 1.0000 - 20s 751ms/step - loss: 0.0098 - acc: 1.0000 -
val_loss: 0.0543 - val_acc: 0.9804
Epoch 13/20
26/26 [======= 0.0055 - acc: 1.0000 -
val loss: 0.0473 - val acc: 0.9951
Epoch 14/20
26/26 [=====
              val loss: 0.0484 - val acc: 0.9853
Epoch 15/20
26/26 [============] - 21s 792ms/step - loss: 0.0037 - acc: 1.0000 -
val_loss: 0.0413 - val_acc: 0.9951
Epoch 16/20
26/26 [======= 0.0032 - acc: 1.0000 -
val_loss: 0.0385 - val_acc: 0.9951
Epoch 17/20
26/26 [======] - 20s 766ms/step - loss: 0.0029 - acc: 1.0000 -
```

val_loss: 0.0409 - val_acc: 0.9902

```
Epoch 18/20
val loss: 0.0404 - val acc: 0.9902
Epoch 19/20
26/26 [======] - 20s 775ms/step - loss: 0.0020 - acc: 1.0000 -
val loss: 0.0406 - val acc: 0.9853
Epoch 20/20
26/26 [======] - 20s 765ms/step - loss: 0.0018 - acc: 1.0000 -
val loss: 0.0382 - val acc: 0.9853
7/7 [======] - 1s 87ms/step
17)
Epoch 1/20
           26/26 [======
val loss: 2.8229 - val acc: 0.1176
Epoch 2/20
26/26 [======] - 20s 771ms/step - loss: 2.6881 - acc: 0.1434 -
val_loss: 2.5127 - val_acc: 0.2647
Epoch 3/20
26/26 [======] - 20s 766ms/step - loss: 2.1087 - acc: 0.3309 -
val loss: 1.7350 - val acc: 0.4951
Epoch 4/20
val loss: 1.1825 - val acc: 0.7402
Epoch 5/20
26/26 [====== 0.8746 - acc: 0.7610 -
val_loss: 0.8826 - val_acc: 0.7304
Epoch 6/20
```

```
26/26 [======] - 20s 780ms/step - loss: 0.6816 - acc: 0.7978 -
val_loss: 0.6829 - val_acc: 0.8431
Epoch 7/20
val_loss: 0.5078 - val_acc: 0.8971
Epoch 8/20
26/26 [======= 0.3906 - acc: 0.8885 -
val loss: 0.4977 - val acc: 0.8676
Epoch 9/20
26/26 [============] - 20s 785ms/step - loss: 0.2652 - acc: 0.9216 -
val_loss: 0.3912 - val_acc: 0.9069
Epoch 10/20
val loss: 0.2712 - val acc: 0.9657
Epoch 11/20
val_loss: 0.2480 - val_acc: 0.9461
Epoch 12/20
26/26 [======= 0.0886 - acc: 0.9877 -
val loss: 0.2971 - val acc: 0.9167
Epoch 13/20
26/26 [======= - 20s 778ms/step - loss: 0.0689 - acc: 0.9939 -
val loss: 0.1689 - val acc: 0.9510
Epoch 14/20
26/26 [============] - 20s 778ms/step - loss: 0.0426 - acc: 0.9975 -
val_loss: 0.1272 - val_acc: 0.9755
Epoch 15/20
val_loss: 0.1349 - val_acc: 0.9657
```

Epoch 16/20

```
val_loss: 0.1362 - val_acc: 0.9657
Epoch 17/20
26/26 [======= 0.0277 - acc: 0.9963 -
val_loss: 0.1010 - val_acc: 0.9706
Epoch 18/20
26/26 [======= - 20s 777ms/step - loss: 0.0143 - acc: 1.0000 -
val loss: 0.0944 - val acc: 0.9755
Epoch 19/20
26/26 [======] - 20s 783ms/step - loss: 0.0141 - acc: 1.0000 -
val_loss: 0.0853 - val_acc: 0.9853
Epoch 20/20
26/26 [======= 0.0107 - acc: 1.0000 -
val loss: 0.0812 - val acc: 0.9804
7/7 [======] - 1s 90ms/step
18)
Epoch 1/20
26/26 [=======] - 13s 473ms/step - loss: 2.8477 - acc: 0.0748 -
val_loss: 2.7455 - val_acc: 0.1225
Epoch 2/20
26/26 [======
                          ====] - 12s 466ms/step - loss: 2.4404 - acc: 0.2843 -
val loss: 1.9969 - val acc: 0.4755
Epoch 3/20
26/26 [======] - 12s 456ms/step - loss: 1.4274 - acc: 0.5895 -
val loss: 1.0668 - val acc: 0.6520
Epoch 4/20
val_loss: 0.5910 - val_acc: 0.7794
```

```
Epoch 5/20
26/26 [======
             val loss: 0.4263 - val acc: 0.8676
Epoch 6/20
26/26 [====== 0.9363 - 12s 476ms/step - loss: 0.2768 - acc: 0.9363 -
val loss: 0.2830 - val acc: 0.9216
Epoch 7/20
26/26 [======] - 12s 464ms/step - loss: 0.1367 - acc: 0.9890 -
val loss: 0.1845 - val acc: 0.9559
Epoch 8/20
val_loss: 0.1485 - val_acc: 0.9608
Epoch 9/20
26/26 [=========
                    =======| - 12s 459ms/step - loss: 0.0831 - acc: 0.9779 -
val_loss: 0.1536 - val_acc: 0.9657
Epoch 10/20
val loss: 0.1297 - val acc: 0.9559
Epoch 11/20
=====] - 12s 473ms/step - loss: 0.0442 - acc: 0.9914 -
val loss: 0.1375 - val acc: 0.9608
Epoch 12/20
val_loss: 0.1063 - val_acc: 0.9657
Epoch 13/20
val_loss: 0.1012 - val_acc: 0.9706
Epoch 14/20
26/26 [======] - 12s 462ms/step - loss: 0.0083 - acc: 1.0000 -
val_loss: 0.0839 - val_acc: 0.9706
```

Epoch 15/20

26/26 [======] - 12s 458ms/step - loss: 0.0067 - acc: 1.0000 - val_loss: 0.0904 - val_acc: 0.9706

Epoch 16/20

26/26 [==========] - 12s 476ms/step - loss: 0.0058 - acc: 1.0000 - val loss: 0.0873 - val acc: 0.9706

Epoch 17/20

26/26 [======] - 12s 463ms/step - loss: 0.0047 - acc: 1.0000 - val_loss: 0.0783 - val_acc: 0.9706

Epoch 18/20

26/26 [======] - 12s 458ms/step - loss: 0.0037 - acc: 1.0000 - val_loss: 0.0851 - val_acc: 0.9755

Epoch 19/20

26/26 [======] - 12s 460ms/step - loss: 0.0028 - acc: 1.0000 - val_loss: 0.0855 - val_acc: 0.9706

Epoch 20/20

26/26 [===========] - 12s 458ms/step - loss: 0.0022 - acc: 1.0000 - val loss: 0.0872 - val acc: 0.9755

7/7 [=====] - 1s 88ms/step

SNAPSHOTS:

1)

Flute -	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Ghatam -	0	13	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0
Harmonium -	0	0	8	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Kanjira -	0	0	0	11	0	0	0	0	0	0	0	0	0	0	0	0	0
Morsing -	0	0	0	0	13	0	0	0	0	0	0	0	0	0	0	0	0
Mrudangam -	0	0	0	0	0	14	0	0	0	0	0	0	0	0	0	0	0
ladaswaram -	0	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0	0
Parai -	0	0	0	0	0	0	0	15	0	0	0	0	0	0	0	0	0
Pungi -	0	0	0	0	0	0	0	0	16	0	0	0	0	0	0	0	0
Santoor -	0	0	0	0	0	0	0	0	0	10	0	0	0	0	0	0	0
Sarangi -	0	0	0	0	0	0	0	0	0	0	17	0	0	0	0	0	0
Sarod -	0	0	0	0	0	0	0	0	0	0	0	14	0	0	0	0	0
Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	8	0	0	0	0
Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	19	0	0	0
Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	8	0	0
Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	10	0
Veena -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	15
	Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	- jbund	Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	Thavil -	Udukkai -	Veena -
	Ghatam - Harmonium - Kanjira - Morsing - Mrudangam - Jadaswaram - Parai - Pungi - Santoor - Sarangi - Sarod - Tabla - Thavil - Udukkai -		Ghatam - 0 13 Harmonium - 0 0 Kanjira - 0 0 Morsing - 0 0 Mrudangam - 0 0 Parai - 0 0 Pungi - 0 0 Santoor - 0 0 Sarangi - 0 0 Sitar - 0 0 Tabla - 0 0 Udukkai - 0 0 Veena - 0 0	Ghatam - 0 13 0 Harmonium - 0 0 8 Kanjira - 0 0 0 Morsing - 0 0 0 Mrudangam - 0 0 0 Parai - 0 0 0 Pungi - 0 0 0 Santoor - 0 0 0 Sarangi - 0 0 0 Sitar - 0 0 0 Tabla - 0 0 0 Udukkai - 0 0 0 Veena - 0 0 0	Ghatam - 0 13 0 0 Harmonium - 0 0 8 0 Kanjira - 0 0 0 11 Morsing - 0 0 0 0 Mrudangam - 0 0 0 0 Parai - 0 0 0 0 Pungi - 0 0 0 0 Santoor - 0 0 0 0 Sarangi - 0 0 0 0 Sitar - 0 0 0 0 Tabla - 0 0 0 0 Udukkai - 0 0 0 0 Veena - 0 0 0 0	Ghatam - 0 13 0 0 0 Harmonium - 0 0 8 0 0 Kanjira - 0 0 0 11 0 Morsing - 0 0 0 0 13 Mrudangam - 0 0 0 0 0 0 Parai - 0 0 0 0 0 0 Pungi - 0 0 0 0 0 0 Santoor - 0 0 0 0 0 0 Sarangi - 0 0 0 0 0 0 Sitar - 0 0 0 0 0 0 Tabla - 0 0 0 0 0 0 Udukkai - 0 0 0 0 0 0 Veena - 0 0 0 0 0 0	Ghatam - 0 13 0 0 0 0 Harmonium - 0 0 8 0 0 0 Kanjira - 0 0 0 11 0 0 Morsing - 0 0 0 0 13 0 Mrudangam - 0 0 0 0 0 14 Madaswaram - 0 0 0 0 0 0 0 Parai - 0 0 0 0 0 0 0 Pungi - 0 0 0 0 0 0 0 Santoor - 0 0 0 0 0 0 0 Sarangi - 0 0 0 0 0 0 0 Sitar - 0 0 0 0 0 0 0 Tabla - 0 0 0 0 0 0 0 Udukkai - 0 0 0 0 0 0 0 Veena - 0 0 0 0 0 0 0	Ghatam - 0 13 0 0 0 0 0 0 Harmonium - 0 0 8 0 0 0 0 0 Kanjira - 0 0 0 11 0 0 0 Morsing - 0 0 0 0 13 0 0 Mrudangam - 0 0 0 0 0 14 0 Nadaswaram - 0 0 0 0 0 0 0 10 Parai - 0 0 0 0 0 0 0 0 0 Santoor - 0 0 0 0 0 0 0 0 0 Sarangi - 0 0 0 0 0 0 0 0 0 Sarangi - 0 0 0 0 0 0 0 0 0 Tabla - 0 0 0 0 0 0 0 0 0 Udukkai - 0 0 0 0 0 0 0 0 0 Veena - 0 0 0 0 0 0 0 0	Ghatam - 0 13 0 0 0 0 0 0 0 0 0 Harmonium - 0 0 8 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	Ghatam - 0 13 0 0 0 0 0 0 0 0 0	Ghatam - 0 13 0 0 0 0 0 0 0 0 0 0 0 Harmonium - 0 0 8 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	Ghatam - 0 13 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0					

accuracy for 2 layers : 0.9950980392156863

	Flute -	0	0	0	0	0	0	0	9	0	0	0	2	0	0	0	0	0
	Ghatam -	0	0	0	0	0	0	0	9	0	0	0	0	0	0	6	0	0
	Harmonium -	0	0	0	0	0	0	0	7	0	0	0	0	0	0	3	0	0
	Kanjira -	0	0	0	0	0	0	0	8	0	0	0	0	0	0	7	0	0
	Morsing -	0	0	0	0	0	0	0	0	0	0	0	3	0	0	9	0	0
	Mrudangam -	0	0	0	0	0	0	0	11	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	12	0	0
e e	Parai -	0	0	0	0	0	0	0	13	0	0	0	0	0	0	0	0	0
True Label	Pungi -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	10	0	0
골	Santoor -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	10	0	0
	Sarangi -	0	0	0	0	0	0	0	13	0	0	0	0	0	0	0	0	0
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	10	0	0	0	0	0
	Sitar -	0	0	0	0	0	0	0	12	0	0	0	0	0	0	2	0	0
	Tabla -	0	0	0	0	0	0	0	8	0	0	0	0	0	0	0	0	0
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	17	0	0
	Udukkai -	0	0	0	0	0	0	0	3	0	0	0	0	0	0	10	0	0
	Veena -		0	0	0	0	0	0	1	0	0	0	0	0	0	9	0	0
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam –	Nadaswaram -	Parai	- ibund	- Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -

	Flute -	6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Ghatam -	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Harmonium -	0	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0	1
	Kanjira -	0	0	0	15	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	0	0	0	0	11	0	0	0	0	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	14	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0	0	0
a	Parai -	0	0	0	0	0	0	0	14	0	0	0	0	0	0	0	0	0
True Label	Pungi -	0	0	0	0	0	0	0	0	14	0	0	0	0	0	0	0	0
2	Santoor -	0	0	0	0	0	0	0	0	0	16	0	0	0	0	0	0	0
	Sarangi -	0	0	0	0	0	0	0	0	0	0	16	0	0	0	0	0	0
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	12	0	0	0	0	0
	Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	10	0	0	0	0
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	12	0	0	0
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	13	0	0
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	9	0
	Veena -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	9
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai	- ibund cted	- Santoor -	Sarangi -	Sarod -	Sitar -	- Iabla -	- Thavil -	Udukkai -	Veena -

	_																	
	Flute -	13	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
	Ghatam -	0	8	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Harmonium -	0	0	10	0	0	0	0	1	0	0	0	0	0	0	0	0	0
	Kanjira -	0	0	0	12	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	0	0	0	0	9	0	0	0	0	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	12	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0	0	0
e e	Parai -	0	0	0	0	0	0	0	17	0	0	0	0	0	0	0	0	0
True Label	Pungi -	0	0	0	0	0	0	0	0	9	0	0	0	0	0	0	0	0
ř	Santoor -	0	0	0	0	0	0	0	0	0	14	0	0	0	0	0	0	0
	Sarangi -	0	0	0	0	0	0	0	0	0	0	7	0	0	0	0	0	0
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	10	0	0	0	0	0
	Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	11	0	0	0	0
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	13	0	0	0
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	13	0	0
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	17	0
	Veena -	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	1	13
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam –	Nadaswaram -	Parai -	- Janual	- Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	Thavil -	Udukkai -	Veena -

	Flute -	9	0	0	0	0	0	0	1	1	0	0	0	0	0	2	0	0
	Ghatam -	0	8	0	0	0	0	0	0	1	0	0	0	0	0	0	1	1
	Harmonium -	1	0	0	0	0	0	0	1	13	0	0	0	0	0	1	0	0
	Kanjira -	0	0	0	9	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	1	0	0	0	0	0	0	0	9	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	9	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	5	0	0	0	0	0	0	0	3	0	0	0	0	0	0	0	4
<u>e</u>	Parai -	4	0	0	0	0	0	1	0	3	0	0	0	0	0	0	0	8
True Label	Pungi -	1	0	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0
르	Santoor -	0	0	0	0	0	5	0	0	0	1	0	0	0	0	0	9	0
	Sarangi -	0	0	0	0	0	0	0	0	0	0	0	0	2	1	0	1	4
	Sarod -	0	0	0	2	0	0	0	0	0	0	0	0	0	0	0	1	12
	Sitar -	0	5	0	0	1	0	0	3	2	0	0	0	0	0	0	0	0
	Tabla -	0	0	0	0	0	13	0	0	0	0	0	0	0	0	0	0	0
	Thavil -	2	0	0	0	0	1	0	0	8	0	0	0	0	0	0	0	0
	Udukkai -	0	0	0	0	0	6	0	0	0	0	0	0	0	0	0	6	0
	Veena -	0	0	0	1	0	0	0	0	2	0	0	0	0	0	0	0	7
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	- Parai	- ibund cted I	- Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -

	Flute -	0	0	0	0	0	0	0	0	0	0	0	8	0	0	0	0	0
	Ghatam -	0	7	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0
	Harmonium -	0	0	0	0	3	0	1	0	5	0	0	0	0	0	0	1	0
	Kanjira -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	11	0
	Morsing -	0	0	0	0	0	0	0	0	12	0	0	1	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	5	5	0
	Nadaswaram -	0	2	0	0	0	0	9	0	0	0	0	0	0	0	0	0	0
e e	Parai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	12	0
Frue Label	Pungi -	0	0	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0
교	Santoor -	0	0	0	0	0	0	12	0	2	0	0	0	0	0	0	0	0
	Sarangi -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	7	1	0
	Sarod -	0	0	3	1	0	0	0	0	0	0	0	7	0	0	0	0	0
	Sitar -	0	3	0	0	0	0	12	0	0	0	0	0	0	0	0	0	0
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	4	13	0
	Thavil -	0	3	0	0	0	0	0	0	0	0	0	0	0	0	10	0	0
	Udukkai -	0	1	0	0	0	0	0	0	0	0	0	0	0	0	11	0	0
	Veena -	0	7	0	0	0	0	0	0	0	0	0	0	0	0	11	1	0
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	- jūnd cted l	- Santoor -	Sarangi -	Sarod -	Sitar -	Tabla -	- Thavil -	Udukkai -	Veena -

	Flute -	0	0	0	0	1	0	0	10	0	0	0	0	0	0	0	0	0
	Ghatam -	0	0	0	1	0	0	0	4	0	0	0	0	0	6	0	0	0
	Harmonium -	0	0	1	0	1	0	4	2	0	0	0	0	0	4	1	0	0
	Kanjira -	0	0	0	12	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	0	0	0	0	12	0	0	0	0	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	7	0	0	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0	1	0
<u>e</u>	Parai -	0	0	0	1	0	2	0	8	0	0	0	0	0	4	0	0	0
True Label	Pungi -	0	0	0	0	4	0	5	0	3	0	0	0	0	0	0	0	0
골	Santoor -	0	0	0	10	0	2	0	0	0	0	0	0	0	2	0	0	0
	Sarangi -	0	0	0	0	0	2	0	7	0	1	0	0	0	6	1	0	0
	Sarod -	0	0	0	0	0	0	0	9	0	0	0	0	0	3	0	0	0
	Sitar -	0	0	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0
	Tabla -	0	0	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0
	Thavil -	0	0	0	3	0	1	5	0	0	0	0	0	0	0	0	1	0
	Udukkai -	0	0	0	10	0	0	1	0	0	0	0	0	0	0	0	1	0
	Veena -	0	0	0	7	0	4	0	2	0	0	0	0	0	0	0	0	0
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Paraj	- iBund	- Santoor -	Sarangi -	Sarod -	Sitar -	Tabla -	- Thavil -	Udukkai -	Veena -

Flute - 5 0 4 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0	
Harmonium - 0 0 5 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0
Morsing - 0 0 0 0 7 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0
Morsing - 0 0 0 0 0 7 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0
Mrudangam - 0 0 0 0 0 0 0 16 0 0 0 0 0 0 0 0 0 0 0	0 0
Nadaswaram - 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0
Parai - 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	5 2
Pungi - 0 0 0 0 0 0 0 0 12 0 0 0 0 0 0 0 0 0 0	0 0
Sarangi - 0 0 3 0 0 0 0 0 0 0 0 15 0 0 0 0 0 0 0 0 0 0	0 0
Sarangi - 0 0 3 0 0 0 0 0 0 0 0 15 0 0 0 0 0 0 0 0 0 0	0 0
Sarod - 0 0 2 0 0 0 0 0 0 0 0 10 0 0 0 0 0 Sitar - 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0
Sitar - 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0
Tabla - 0 1 0 0 0 0 0 2 0 0 0 8 0 0 0 0 Thavil - 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 16 0 Udukkai - 0 3 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 Veena - 0 0 1 0 0 0 0 0 3 0 0 0 0 1 0	0 0
Thavil - 0 1 0 0 0 0 0 0 0 0 0 0 0 0 16 0 Udukkai - 0 3 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 Veena - 0 0 1 0 0 0 0 0 3 0 0 0 0 1 0	0 0
Udukkai - 0 3 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 Veena - 0 0 1 0 0 0 0 0 3 0 0 0 0 1 0	0 0
Veena - 0 0 1 0 0 0 0 0 3 0 0 0 0 1 0	0 0
	10 0
Predicted Label	Udukkai - Veena -

	Flute -	10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Ghatam -	0	9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Harmonium -	0	0	11	0	0	0	0	0	0	0	0	0	1	0	0	0	0
	Kanjira -	0	0	0	15	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	0	0	0	0	15	0	0	0	0	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	0	1	0	0	0	13	0	0	0	0	0	0	0	0	0	0
-	Parai -	0	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0	0
True Label	Pungi -	0	0	0	0	0	0	0	0	13	0	0	0	0	0	0	0	0
Tue	Santoor -	0	0	0	0	0	0	0	0	0	13	0	0	0	0	0	0	0
	Sarangi -	0	0	0	0	0	0	0	0	0	0	8	0	0	0	0	0	0
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	9	0	0	0	0	0
	Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	15	0	0	0	0
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	9	0	0	0
	Thavil -		0	0	0	0	0	0	0	0	0	0	0	0	0	12	0	0
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	11	0
	Veena -		0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	16
		1	_	_	_	_		_			Ĺ		-	L				_
		Flute	Ghatam	Harmonium	Kanjira	Morsing	Mrudangam	Nadaswaram	Parai	Pungi	Santoor	Sarangi	Sarod	Sitar	Tabla	Thavil	Udukkai	Veena
							_	z	Predi	cted l	Label							

	Flute -	11	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0
	Ghatam -	0	13	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Harmonium -	0	0	8	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Kanjira -	0	0	0	13	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	0	0	0	0	13	0	0	0	0	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	0	0	0	0	0		0	0	0	0	0	0	0	0	0	0
<u>e</u>	Parai -	0	0	0	0	0	0	0	13	0	0	0	0	0	0	0	0	0
Frue Label	Pungi -	0	0	0	0	0	0	0	0	13	0	0	0	0	0	0	0	0
Ī	Santoor -	0	0	0	0	0	0	0	0	0	11	0	0	0	0	0	0	0
	Sarangi -	0	0	0	0	0	0	0	1	0	0	10	0	0	0	0	0	0
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	11	0	0	0	0	0
	Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	13	0	0	0	0
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	14	0	0	0
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	17	0	0
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	12	0
	Veena -	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	2	8
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	- Pungi -	Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -

	Flute -	14	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	
	Ghatam -	0	8	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Harmonium -	0	1	15	0	0	0	0	1	0	0	0	0	0	0	0	0	0	
	Kanjira -	0	0	0	14	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Morsing -	0	0	0	0	7	0	0	0	0	0	0	0	0	0	0	0	0	
	Mrudangam -	0	0	0	0	0	8	0	0	0	0	0	0	0	0	0	0	0	
	Nadaswaram -	0	0	0	0	0	0	13	0	0	0	0	0	0	0	0	0	0	
<u>e</u>	Parai -	0	0	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0	
True Label	Pungi -	0	0	0	0	0	0	0	0	14	0	0	0	0	0	0	0	0	
교	Santoor -	0	0	0	0	0	0	0	0	0	10	0	0	0	0	0	0	0	
	Sarangi -	0	0	0	0	0	0	0	0	0	0	12	0	0	0	0	0	0	
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	10	0	0	0	0	0	
	Sitar -	0	0	0	0	0	0	0	1	0	0	0	0	10	0	0	0	0	
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	15	0	0	0	
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	11	0	0	
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	17	0	
	Veena -		0	0	0	0	0	1	0	0	0	0	0	0	0	1	0	10	
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam –	Nadaswaram -	Parai	- iBund cted I	- Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -	

	Flute -	16	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Ghatam -	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Harmonium -	0	0	7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Kanjira -	0	0	0	15	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Morsing -	1	0	2	0	16	0	0	0	0	0	0	0	0	0	0	0	0	
	Mrudangam -	0	0	0	0	0	11	0	0	0	0	0	0	0	0	0	0	0	
	Nadaswaram -	0	0	0	0	0	0	14	0	0	0	0	0	0	0	0	0	0	
-	Parai -	0	0	0	0	0	0	0	9	0	0	0	0	0	0	0	0	0	
Frue Label	Pungi -	0	0	0	0	0	0	0	0	10	0	0	0	0	0	0	0	0	
Ī	Santoor -	0	0	0	0	0	0	0	0	0	9	0	0	0	0	0	0	0	
	Sarangi -	0	0	0	0	0	0	0	0	0	0	12	0	0	0	0	0	0	
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	14	0	0	0	0	0	
	Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	10	0	0	0	0	
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	9	0	0	0	
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	13	0	0	
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	11	0	
	Veena -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	15	
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	- Pungi -	Santoor -	Sarangi -	Sarod -	Sitar -	Tabla -	Thavil -	Udukkai -	Veena -	

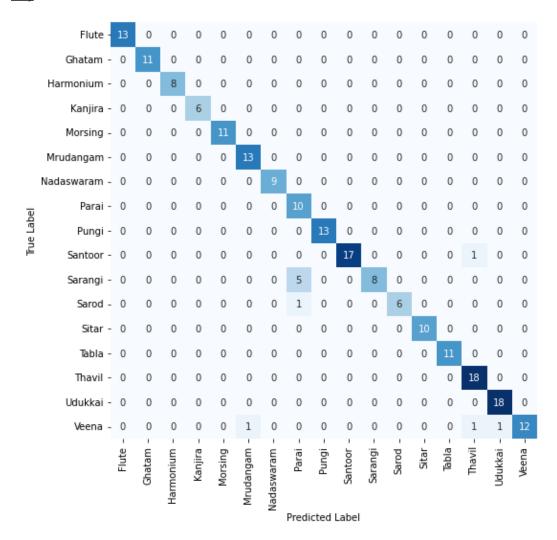
13	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	13	0	0	0	0	0	0	0	0	0	0	0	0	0	1
0	0	0	14	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	14	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	11	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	14	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	15	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	6	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	11	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	13	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	16	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	16	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	8	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	10	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	14	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	9
Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	- Ibund	- Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -
		0 6 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 6 0 0 0 13 0	0 6 0 0 0 13 0 0 0 14 0	0 6 0 0 0 0 0 13 0 0 0 0 0 14 0 0 0 0 0 14 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0<	0 6 0 0 0 0 0 0 13 0 0 0 0 0 0 14 0 0 0 0 0 0 14 0 0 0 0 0 0 11 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 6 0	Hammonium Hammonium - Horizon Balanta Balanta Hammonium - Horizon Balanta Hammonium - Horizon Balanta	Hammunium Hammun	0 6 0	Hammonium Hammon	Hammonium Hammon	O 6 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	Hote Hote	Linter Hammonium H	Hatten Hammunium Hammun

	Flute -	17	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Ghatam -	1	5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Harmonium -	0	0	14	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Kanjira -	0	0	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	0	0	0	0	13	0	0	0	0	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	11	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0	0
<u>e</u>	Parai -	0	0	0	0	0	0	0	14	0	0	0	0	0	0	0	0	0
True Label	Pungi -	0	0	0	0	0	0	0	0	19	0	0	0	0	0	0	0	0
Ĭ	Santoor -	0	0	0	0	0	0	0	0	0	11	0	0	0	0	0	0	0
	Sarangi -	0	0	0	0	0	0	0	2	0	0	7	0	0	0	0	0	0
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	11	0	0	0	0	0
	Sitar -	0	0	0	0	0	0	0	2	0	0	0	0	5	0	0	0	0
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	16	0	0	0
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	15	0	0
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	13	0
	Veena -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	7
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai	- jbund	- Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -

	Flute -	10	0	0	0	1	0	0	0	0	0	2	0	0	0	0	0	0	
	Ghatam -	0		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Harmonium -	0	0	11	0	0	0	0	0	0	0	0	0	1	0	0	0	0	
	Kanjira -	0	0	0		0	0	0	0	0	0	0	0	0	0	0	0	0	
	Morsing -	0	0	0	0	13	0	0	0	0	0	0	0	0	0	0	0	0	
	Mrudangam -	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0	0	0	
	Nadaswaram -	0	0	0	0	0	0	11	0	0	0	0	0	0	0	0	0	0	
<u>e</u>	Parai -	0	0	0	0	0	0	0	8	0	0	0	0	0	0	0	0	0	
True Label	Pungi -	0	0	0	0	0	0	0	0	15	0	0	0	0	0	0	0	0	
ī	Santoor -	0	0	0	0	0	0	0	0	0	14	0	0	0	0	0	0	0	
	Sarangi -	0	0	0	0	0	0	0	1	0	0	15	0	0	0	0	0	0	
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	13	0	0	0	0	0	
	Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	13	0	0	0	0	
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	1	9	0	0	0	
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	14	0	0	
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	10	0	
	Veena -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	14	
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	- Pungi -	Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -	
									Fredu	cred l	anei								

	Flute -	12	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Ghatam -	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Harmonium -	0	0	9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Kanjira -	0	0	0	9	0	0	0	0	0	0	0	0	0	0	0	0	0	
	Morsing -	0	0	0	0	11	0	0	0	0	0	0	0	0	0	0	0	0	
	Mrudangam -	0	0	0	0	0	15	0	0	0	0	0	0	0	0	0	0	0	
	Nadaswaram -	0	0	0	0	0	0	9	0	0	0	0	0	0	0	0	0	0	
υ	Parai -	0	0	0	0	0	0	0	10	0	0	0	0	0	0	0	0	0	
II NE LADEI	Pungi -	0	0	0	0	0	0	0	0	11	0	0	0	0	0	0	0	0	
1	Santoor -	0	0	0	0	0	0	0	0	0	16	0	0	0	0	0	0	0	
	Sarangi -	0	0	0	0	0	0	0	0	0	0	14	0	0	0	0	0	0	
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	18	0	0	0	0	0	
	Sitar -	0	0	0	0	0	0	0	0	0	0	0	0	15	0	0	0	0	
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	11	0	0	0	
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	12	0	0	
	Udukkai -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	9	0	
	Veena -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	12	
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	Pungi -	Santoor -	Sarangi -	Sarod -	Sitar -	Tabla -	Thavil -	Udukkai -	Veena -	

	Flute -	8	1	0	0	2	0	0	0	0	1	0	0	0	0	0	0	0
	Ghatam -	0	8	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Harmonium -	0	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	Kanjira -	0	0	0	16	0	0	0	0	0	0	0	0	0	0	0	0	0
	Morsing -	1	0	0	0	13	0	0	0	0	0	0	0	0	0	0	0	0
	Mrudangam -	0	0	0	0	0	11	0	0	0	0	0	0	0	0	0	0	0
	Nadaswaram -	0	1	0	0	0	0	8	0	0	0	0	0	0	0	0	0	0
-	Parai -	0	0	0	0	0	0	0	11	0	0	0	0	0	0	0	0	0
True Label	Pungi -	0	0	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0
ī	Santoor -	0	0	0	0	0	0	0	0	0	16	0	0	0	0	0	0	0
	Sarangi -	0	0	0	0	0	0	0	0	0	0	15	0	0	0	0	0	0
	Sarod -	0	0	0	0	0	0	0	0	0	0	0	7	0	0	0	0	0
	Sitar -	0	0	2	0	0	0	0	0	0	0	0	0	7	0	0	0	0
	Tabla -	0	0	0	0	0	0	0	0	0	0	0	0	0	14	0	0	0
	Thavil -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	8	0	0
	Udukkai -	0	0	0	0	0	2	0	0	0	0	0	0	0	0	0	15	0
	Veena -	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	15
		Flute -	Ghatam -	Harmonium -	Kanjira -	Morsing -	Mrudangam -	Nadaswaram -	Parai -	- Joundi	- Santoor -	Sarangi -	Sarod -	Sitar -	- Tabla -	- Thavil -	Udukkai -	Veena -



INFERENCES FOR CONVOLUTIONAL NEURAL NETWORK

The following inferences can be made from the above obtained results:

- Dropout in the network is inversely proportional to the accuracy. In most CNN models, accuracy increases when dropout value is decreased.
- Number of Nodes in the output layer in the network is directly proportional to the accuracy.

II. RNN-LSTM:

FEATURE EXTRACTION METHOD:

PROGRAM:

```
for i, (dirpath,dirnames,filenames) in enumerate(os.walk(dataset_path)):
    if dirpath != dataset path:
         #Adding all the labels
         label = str(dirpath).split('\\')[-1]
         data["mapping"].append(label)
         print("\nInside ",label)
         #Gping through each song within a label
                in filenames:
              file_path = dataset_path +"/" + str(label) + "/" + str(f)
              y, sr = librosa.load(file_path, sr = sample_rate)
              #Cutting each song into 5 segments
              for n in range(num_segment):
                   start = samples_per_segment * n
finish = start + samples_per_segment
                   #print(start,finish)
                  mfcc = librosa.feature.mfcc(y[start:finish], sample_rate, n_mfcc = num_mfcc, n_fft = n_fft, hop_length = hop_mfcc = mfcc.T #259 x 13
                   #Making sure if
if len(mfcc) == num_mfcc_vectors_per_segment:
    data["mfcc"].append(mfcc.tolist())
    data["labels"].append(i-1)
                       print("Track Name ", file_path, n+1)
```

```
with open(json_path, "w") as fp:
    json.dump(data, fp, indent = 4)

if __name__ == "__main__":
    preprocess(dataset_path,jsonpath,num_segment=5)
```

Output:

```
Çjupyter data_json✔ 05/26/2022
                                                                                                                                                                                                                        Logout
          Edit
                    View Language
                                                                                                                                                                                                                      Plain Text
                   "mapping": [
                         "Flute"
                         "Ghatam",
                          "Harmonium"
                         "Kanjira",
                         "Morsing",
"Mrudangam"
      8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
                         "Nadaswaram",
                         "Pungi",
"Santoor
                          "Sarangi",
                         "Sarod",
                         "Sitar",
"Tabla",
"Thavil"
                         "Veena'
```

```
∵ jupyter data_json🗸 05/26/2022
                                                                                                                                                                                         Logout
                           Language
   5000
5001
                ],
"mfcc": [
   5002
   5003
5004
                                -146.5601348876953,
   5005
                                110.26771545410156
   5005
5006
5007
                                -71.389404296875,
-4.81665563583374,
   5008
                                9.152339935302734
                                38.97675323486328,
   5010
                                1.018646001815796,
   5011
                                -12.832742691040039,
-10.293302536010742,
                                18.992544174194336.
   5013
                                15.533506393432617,
-1.5947884321212769,
   5014
   5016
                                -17.16718292236328
   5017
5018
                                -172.016357421875.
   5019
   5020
5021
                                104.93514251708984,
-47.28810119628906,
   5022
                                4.130656719207764.
   5023
5024
                                12.16766357421875,
46.16957092285156,
   5025
                                -6.451200485229492
   5027
                                -15.721405029296875,
   5028
                                22.899059295654297
```

METHODOLOGY

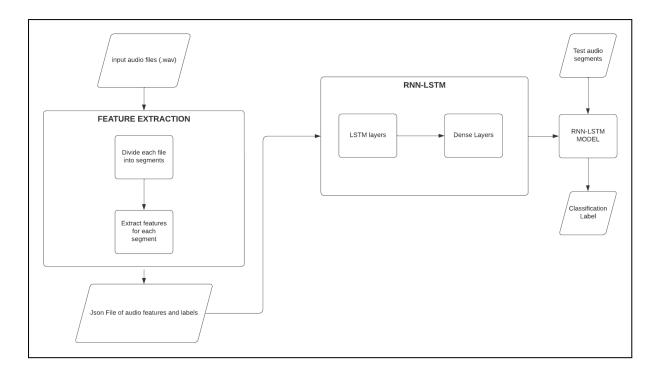
The LSTM (Long Short-Term Memory) network is a variety of recurrent neural networks (RNNs) that are capable of learning long-term dependencies, especially in sequence prediction problems. LSTM has feedback connections, i.e., it is capable of processing the entire sequence of data such as in audio files, apart from single data points such as images. Thus, its applications include musical information retrieval.

Though LSTM networks are primarily used for prediction, they provide high accuracy for classification problems also.

In the LSTM model proposed here, each audio file in the dataset described is divided into 5 segments each. For each segment the 13 MFCC features are extracted and written into a JavaScript Object Notation (json) file with the labels and mappings.

The MFCC vector is then read from the json file and the train-test split is done. An LSTM Network with 1 LSTM layer and 1 Dense layer (Base Model) is constructed and the model is compiled and trained with the train dataset. The model is tested with the test dataset and the output class probabilities for each segment are one hot encoded manually. The accuracy and confusion matrix is printed

BLOCK DIAGRAM



IMPLEMENTATION

```
import os
import json
import tensorflow as tf
from sklearn.model_selection import train_test_split
import numpy as np
{\bf import}\ {\tt matplotlib.pyplot}\ {\tt as}\ {\tt plt}
from sklearn.metrics import accuracy_score
from sklearn.metrics import confusion_matrix
data_path = "data_json"
def load_data(data_path):
    print("Data loading\n")
    with open(data_path, "r") as fp:
        data = json.load(fp)
    x = np.array(data["mfcc"])
    y = np.array(data["labels"])
    print(x.shape,y.shape)
    print("Loaded Data")
    return x, y
def prepare datasets(test size,val size):
    #Load the data
    x, y = load_data(data_path)
```

```
x_train, x_test, y_train, y_test = train_test_split(x,y,test_size = test_size)
x_train, x_val, y_train, y_val = train_test_split(x_train,y_train,test_size = val_size)
return x_train, x_val, x_test, y_train, y_val, y_test

def build_model(input_shape):

model = tf.keras.Sequential()
model.add(tf.keras.layers.LSTM(64, input_shape = input_shape, return_sequences = True))
model.add(tf.keras.layers.LSTM(64, input_shape = input_shape, return_sequences = True))
model.add(tf.keras.layers.LSTM(64, input_shape = input_shape, return_sequences = True))
model.add(tf.keras.layers.LSTM(64))

model.add(tf.keras.layers.Dense(128, activation="relu"))
model.add(tf.keras.layers.Dense(17,activation = "softmax"))
model.add(tf.keras.layers.Dropout(0.05))
return model

if __name__ == "__main__":
```

```
classes = ["Flute", "Ghatam", "Harmonium", "Kanjira", "Morsing", "Mrudangam", "Nadaswaram", "Parai", "Pungi", "Santoor", "Sarod", "Sita
class predictions = []
x_train, x_val, x_test, y_train, y_val, y_test = prepare_datasets(0.25, 0.2)
print(x_train.shape[0])
input_shape = (x_train.shape[1],x_train.shape[2])
model = build_model(input_shape)
# compile model
optimiser = tf.keras.optimizers.Adam(lr=0.001)
model.compile(optimizer=optimiser,
               loss='sparse_categorical_crossentropy',
               metrics=['accuracy'])
model.summary()
# train model
\label{eq:model.fit} \mbox{history = model.fit}(\mbox{x\_train, y\_train, validation\_data=}(\mbox{x\_val, y\_val}), \mbox{ batch\_size=32, epochs=20})
# plot accuracy/error for training and validation
#plot_history(history)
# evaluate model on test set
test_loss, test_acc = model.evaluate(x_test, y_test, verbose=2)
print('\nTest accuracy:', test_acc)
```

```
model.save("model_RNN_LSTM30.h5")
print("Saved model to disk")

model = tf.keras.models.load_model("model_RNN_LSTM30.h5")

array = model.predict(x_test)
print(array)
max_indices = array.argmax(axis=1)
count = 0
for i in range(array.shape[0]): #rows
    for j in range(array.shape[1]): #coLs
        if j == max_indices[i]:
            array[i][j] = 1
        else:
            array[i][j] = 0

print(array.shape,array) #BINARY 2D ARRAY WITH 1 IF CLASS IS PRESENT

print("Accuracy Score: ", accuracy_score(y_test, max_indices))
print("Confusion Matrix:\n", confusion_matrix(y_test, max_indices))
```

The above implementation is of a 6 layer LSTM model with 128 nodes in the 1st dense layer with a dropout value of 0.05 added after the last layer

RESULTS

The hyperparameters that were varied in RNN-LSTM:

- Number of layers
- Number of nodes in Dense layer
- Dropout value

The hyperparameters that were set constant in RNN-LSTM:

- Activation Function: relu, softmax
- Optimizer: Adam
- Number of Epochs: 20

*Dropout layer is added only after the last layer for all model variations

Case	Number of	No. of nodes	Dropout	Accuracy
	LSTM Layers	in 1st Dense		
		Layer		
1	2	17	-	0.99759
2	2	17	0.05	0.98955
3	2	17	0.1	0.98955
4	2	17	0.2	0.97349
5	4	128	-	0.99598
6	4	128	0.05	0.98634
7	4	128	0.1	0.98152
8	4	128	0.2	0.96706
9	4	256	0.05	0.96626
10	4	256	0.1	0.97429
11	4	256	0.2	0.97349
12	6	128	-	0.97028
13	6	128	0.05	0.97911
14	6	128	0.1	0.92771
15	6	128	0.2	0.93975
16	6	256	0.05	0.92289
17	6	256	0.1	0.95742
18	6	256	0.2	0.85783

RESULT SNAPSHOTS FOR RNN-LSTM

Case 1:

```
Epoch 1/20
0.3369 - val loss: 1.8296 - val accuracy: 0.6037
Epoch 2/20
- val_loss: 1.0310 - val_accuracy: 0.8728
Epoch 3/20
- val_loss: 0.6323 - val_accuracy: 0.9317
Epoch 4/20
- val loss: 0.4107 - val accuracy: 0.9612
Epoch 5/20
- val_loss: 0.2731 - val_accuracy: 0.9705
Epoch 6/20
- val loss: 0.2129 - val accuracy: 0.9759
Epoch 7/20
- val_loss: 0.1642 - val_accuracy: 0.9786
Epoch 8/20
- val_loss: 0.1236 - val_accuracy: 0.9826
Epoch 9/20
- val loss: 0.1087 - val accuracy: 0.9813
Epoch 10/20
- val_loss: 0.0941 - val_accuracy: 0.9893
Epoch 11/20
- val_loss: 0.0897 - val_accuracy: 0.9853
Epoch 12/20
- val loss: 0.0738 - val accuracy: 0.9893
```

```
Epoch 13/20
- val_loss: 0.0633 - val_accuracy: 0.9906
Epoch 14/20
- val loss: 0.0581 - val accuracy: 0.9906
Epoch 15/20
- val loss: 0.0507 - val accuracy: 0.9946
Epoch 16/20
- val_loss: 0.0451 - val_accuracy: 0.9946
Epoch 17/20
- val_loss: 0.0384 - val_accuracy: 0.9933
Epoch 18/20
- val_loss: 0.0361 - val_accuracy: 0.9946
Epoch 19/20
- val_loss: 0.0356 - val_accuracy: 0.9946
Epoch 20/20
- val loss: 0.0353 - val accuracy: 0.9933
```

Test accuracy: 0.9975903630256653

39/39 - 0s - loss: 0.0279 - accuracy: 0.9976 - 445ms/epoch - 11ms/step

Case 2:

```
Epoch 1/20
0.2796 - val loss: 1.8919 - val accuracy: 0.5408
Epoch 2/20
- val loss: 1.3227 - val accuracy: 0.7041
Epoch 3/20
- val_loss: 0.8886 - val_accuracy: 0.8648
Epoch 4/20
- val_loss: 0.5428 - val_accuracy: 0.9076
Epoch 5/20
94/94 [========================== - 4s 41ms/step - loss: 1.2092 - accuracy: 0.8881
- val_loss: 0.3668 - val_accuracy: 0.9478
Epoch 6/20
- val_loss: 0.2794 - val_accuracy: 0.9545
Epoch 7/20
94/94 [==========================] - 4s 43ms/step - loss: 0.9453 - accuracy: 0.9133
- val loss: 0.2039 - val accuracy: 0.9692
Epoch 8/20
- val_loss: 0.1766 - val_accuracy: 0.9679
Epoch 9/20
- val_loss: 0.1461 - val_accuracy: 0.9719
Epoch 10/20
- val loss: 0.1159 - val accuracy: 0.9786
Epoch 11/20
- val_loss: 0.1200 - val_accuracy: 0.9746
Epoch 12/20
- val_loss: 0.1028 - val_accuracy: 0.9813
Epoch 13/20
```

```
- val_loss: 0.0968 - val_accuracy: 0.9813
Epoch 14/20
- val_loss: 0.0721 - val_accuracy: 0.9880
Epoch 15/20
94/94 [============================ ] - 4s 44ms/step - loss: 0.6123 - accuracy: 0.9434
- val loss: 0.0820 - val accuracy: 0.9826
Epoch 16/20
- val_loss: 0.0734 - val_accuracy: 0.9826
Epoch 17/20
- val_loss: 0.0712 - val_accuracy: 0.9839
Epoch 18/20
- val_loss: 0.0554 - val_accuracy: 0.9839
Epoch 19/20
- val_loss: 0.0614 - val_accuracy: 0.9866
Epoch 20/20
- val_loss: 0.0577 - val_accuracy: 0.9853
39/39 - 1s - loss: 0.0530 - accuracy: 0.9896 - 619ms/epoch - 16ms/step
```

Case 3:

```
Epoch 1/20
- val_loss: 1.7806 - val_accuracy: 0.6024
Epoch 2/20
- val loss: 1.0658 - val accuracy: 0.8313
Epoch 3/20
- val_loss: 0.6704 - val_accuracy: 0.9063
Epoch 4/20
94/94 [=============================== - 5s 52ms/step - loss: 1.8311 - accuracy: 0.8473
- val_loss: 0.4324 - val_accuracy: 0.9451
Epoch 5/20
- val_loss: 0.2976 - val_accuracy: 0.9572
Epoch 6/20
94/94 [================================ - 5s 51ms/step - loss: 1.6115 - accuracy: 0.8707
- val_loss: 0.2171 - val_accuracy: 0.9719
Epoch 7/20
- val loss: 0.1650 - val accuracy: 0.9732
Epoch 8/20
94/94 [============================= ] - 4s 46ms/step - loss: 1.3784 - accuracy: 0.8791
- val_loss: 0.1430 - val_accuracy: 0.9772
Epoch 9/20
- val_loss: 0.1395 - val_accuracy: 0.9652
Epoch 10/20
- val loss: 0.1133 - val accuracy: 0.9759
Epoch 11/20
- val_loss: 0.0966 - val_accuracy: 0.9786
Epoch 12/20
- val_loss: 0.0877 - val_accuracy: 0.9759
Epoch 13/20
```

```
- val_loss: 0.0721 - val_accuracy: 0.9799
Epoch 14/20
- val_loss: 0.0792 - val_accuracy: 0.9746
Epoch 15/20
- val loss: 0.0891 - val accuracy: 0.9759
Epoch 16/20
- val_loss: 0.0563 - val_accuracy: 0.9826
Epoch 17/20
- val_loss: 0.0663 - val_accuracy: 0.9839
Epoch 18/20
- val_loss: 0.0528 - val_accuracy: 0.9893
Epoch 19/20
- val_loss: 0.0518 - val_accuracy: 0.9813
Epoch 20/20
- val_loss: 0.0471 - val_accuracy: 0.9906
39/39 - 1s - loss: 0.0427 - accuracy: 0.9896 - 593ms/epoch - 15ms/step
```

Case 4:

```
Epoch 1/20
0.2934 - val loss: 1.9145 - val accuracy: 0.5649
Epoch 2/20
- val loss: 1.2241 - val accuracy: 0.7617
Epoch 3/20
- val_loss: 0.8299 - val_accuracy: 0.8594
Epoch 4/20
- val_loss: 0.5403 - val_accuracy: 0.9183
Epoch 5/20
94/94 [================================ - 4s 42ms/step - loss: 3.1916 - accuracy: 0.7448
- val_loss: 0.3906 - val_accuracy: 0.9237
Epoch 6/20
- val_loss: 0.3006 - val_accuracy: 0.9451
Epoch 7/20
- val loss: 0.2398 - val accuracy: 0.9518
Epoch 8/20
- val_loss: 0.2126 - val_accuracy: 0.9585
Epoch 9/20
- val_loss: 0.2414 - val_accuracy: 0.9438
Epoch 10/20
- val loss: 0.1589 - val accuracy: 0.9585
Epoch 11/20
- val_loss: 0.1512 - val_accuracy: 0.9625
Epoch 12/20
- val_loss: 0.1604 - val_accuracy: 0.9612
Epoch 13/20
```

```
- val_loss: 0.1192 - val_accuracy: 0.9585
Epoch 14/20
- val_loss: 0.0974 - val_accuracy: 0.9692
Epoch 15/20
- val loss: 0.1220 - val accuracy: 0.9612
Epoch 16/20
- val loss: 0.1184 - val accuracy: 0.9692
Epoch 17/20
- val_loss: 0.1088 - val_accuracy: 0.9719
Epoch 18/20
- val_loss: 0.0804 - val_accuracy: 0.9719
Epoch 19/20
- val_loss: 0.0842 - val_accuracy: 0.9705
Epoch 20/20
- val_loss: 0.0919 - val_accuracy: 0.9719
39/39 - 1s - loss: 0.0921 - accuracy: 0.9735 - 558ms/epoch - 14ms/step
```

Case 5:

```
Epoch 1/20
94/94 [=============== ] - 18s 130ms/step - loss: 1.4212 - accuracy:
0.6407 - val loss: 0.5529 - val accuracy: 0.8581
Epoch 2/20
0.9106 - val loss: 0.2854 - val accuracy: 0.9224
Epoch 3/20
- val_loss: 0.2878 - val_accuracy: 0.9116
Epoch 4/20
- val loss: 0.1386 - val accuracy: 0.9545
Epoch 5/20
- val_loss: 0.1662 - val_accuracy: 0.9545
Epoch 6/20
- val_loss: 0.1336 - val_accuracy: 0.9625
Epoch 7/20
- val loss: 0.0749 - val accuracy: 0.9759
Epoch 8/20
- val_loss: 0.0692 - val_accuracy: 0.9786
Epoch 9/20
- val_loss: 0.0973 - val_accuracy: 0.9679
Epoch 10/20
- val loss: 0.0647 - val accuracy: 0.9786
Epoch 11/20
- val_loss: 0.0732 - val_accuracy: 0.9786
Epoch 12/20
- val_loss: 0.0622 - val_accuracy: 0.9746
Epoch 13/20
```

```
- val_loss: 0.0712 - val_accuracy: 0.9799
Epoch 14/20
- val_loss: 0.1683 - val_accuracy: 0.9465
Epoch 15/20
- val loss: 0.0437 - val accuracy: 0.9920
Epoch 16/20
- val_loss: 0.0365 - val_accuracy: 0.9920
Epoch 17/20
- val_loss: 0.0374 - val_accuracy: 0.9906
Epoch 18/20
- val_loss: 0.0369 - val_accuracy: 0.9866
Epoch 19/20
- val_loss: 0.0207 - val_accuracy: 0.9906
Epoch 20/20
- val_loss: 0.0189 - val_accuracy: 0.9946
39/39 - 1s - loss: 0.0152 - accuracy: 0.9960 - 1s/epoch - 33ms/step
```

Case 6:

```
Epoch 1/20
94/94 [==
                                             = ] - 17s 124ms/step - loss: 2.2464 - accuracy: 0.5429 -
val_loss: 0.7164 - val_accuracy: 0.7711
Epoch 2/20
94/94 [===
                                             = ] - 9s 97ms/step - loss: 1.2648 - accuracy: 0.7984 -
val loss: 0.2384 - val accuracy: 0.9277
Epoch 3/20
94/94 [=====
                                              =] - 7s 75ms/step - loss: 0.8421 - accuracy: 0.8768 -
val_loss: 0.7216 - val_accuracy: 0.8447
Epoch 4/20
94/94 [===
                                              =] - 8s 83ms/step - loss: 0.8615 - accuracy: 0.8731 -
val_loss: 0.1646 - val_accuracy: 0.9558
Epoch 5/20
94/94 [===
                                              =] - 8s 82ms/step - loss: 0.6599 - accuracy: 0.8935 -
val loss: 0.1406 - val accuracy: 0.9585
Epoch 6/20
94/94 [======
                                              =] - 7s 78ms/step - loss: 0.5960 - accuracy: 0.9099 -
val loss: 0.3037 - val accuracy: 0.9250
Epoch 7/20
94/94 [======
                                             =] - 8s 86ms/step - loss: 0.5047 - accuracy: 0.9143 -
val_loss: 0.1517 - val_accuracy: 0.9451
Epoch 8/20
94/94 [========
                                           ===] - 8s 84ms/step - loss: 0.4779 - accuracy: 0.9133 -
val_loss: 0.2536 - val_accuracy: 0.9384
Epoch 9/20
94/94 [======
                                              =] - 8s 80ms/step - loss: 0.5597 - accuracy: 0.8895 -
val loss: 0.3629 - val accuracy: 0.8929
Epoch 10/20
94/94 [==
                                              =] - 8s 85ms/step - loss: 0.5582 - accuracy: 0.9042 -
val_loss: 0.0981 - val_accuracy: 0.9692
Epoch 11/20
94/94 [=====
                                             = ] - 8s 83ms/step - loss: 0.4246 - accuracy: 0.9200 -
val loss: 0.1159 - val accuracy: 0.9665
Epoch 12/20
94/94 [==
                                              =] - 8s 85ms/step - loss: 0.3977 - accuracy: 0.9236 -
val_loss: 0.1307 - val_accuracy: 0.9639
Epoch 13/20
```

```
94/94 [======
                    val loss: 0.0659 - val accuracy: 0.9813
Epoch 14/20
94/94 [===
                                      ====] - 8s 85ms/step - loss: 0.3115 - accuracy: 0.9236 -
val loss: 0.1195 - val accuracy: 0.9759
Epoch 15/20
94/94 [======
                                     ====] - 8s 88ms/step - loss: 0.2789 - accuracy: 0.9364 -
val loss: 0.1442 - val accuracy: 0.9799
Epoch 16/20
94/94 [======
                             =======] - 8s 86ms/step - loss: 0.4075 - accuracy: 0.9173 -
val_loss: 0.3559 - val_accuracy: 0.9317
Epoch 17/20
94/94 [=====
                               ======] - 8s 86ms/step - loss: 0.3220 - accuracy: 0.9267 -
val loss: 0.0751 - val accuracy: 0.9866
Epoch 18/20
94/94 [==
                                        ==] - 8s 86ms/step - loss: 0.2957 - accuracy: 0.9257 -
val_loss: 0.1348 - val_accuracy: 0.9772
Epoch 19/20
94/94 [=====
                                        ==] - 8s 87ms/step - loss: 0.2924 - accuracy: 0.9357 -
val loss: 0.0638 - val accuracy: 0.9893
Epoch 20/20
94/94 [======
                              ======] - 8s 89ms/step - loss: 0.2119 - accuracy: 0.9441 -
val_loss: 0.0868 - val_accuracy: 0.9853
39/39 - 1s - loss: 0.0719 - accuracy: 0.9863 - 1s/epoch - 36ms/step
```

Case 7:

```
Epoch 1/20
94/94 [================ ] - 15s 111ms/step - loss: 2.8657 - accuracy:
0.5489 - val loss: 0.6104 - val accuracy: 0.8487
Epoch 2/20
94/94 [============== ] - 10s 101ms/step - loss: 1.6570 - accuracy:
0.7991 - val loss: 0.3395 - val accuracy: 0.9050
Epoch 3/20
- val_loss: 0.3594 - val_accuracy: 0.9050
Epoch 4/20
- val_loss: 0.2530 - val_accuracy: 0.9317
Epoch 5/20
- val_loss: 0.2682 - val_accuracy: 0.9210
Epoch 6/20
- val_loss: 0.1642 - val_accuracy: 0.9612
Epoch 7/20
- val loss: 0.2266 - val accuracy: 0.9545
Epoch 8/20
- val_loss: 0.1798 - val_accuracy: 0.9639
Epoch 9/20
- val_loss: 0.3575 - val_accuracy: 0.9264
Epoch 10/20
- val loss: 0.1330 - val accuracy: 0.9679
Epoch 11/20
- val_loss: 0.2646 - val_accuracy: 0.9505
Epoch 12/20
94/94 [=============================] - 10s 101ms/step - loss: 0.4660 - accuracy:
0.8878 - val_loss: 0.1050 - val_accuracy: 0.9799
Epoch 13/20
```

```
- val_loss: 0.1488 - val_accuracy: 0.9759
Epoch 14/20
- val_loss: 0.1144 - val_accuracy: 0.9732
Epoch 15/20
94/94 [================================ ] - 9s 90ms/step - loss: 0.5497 - accuracy: 0.8791
- val loss: 0.1182 - val accuracy: 0.9732
Epoch 16/20
- val_loss: 0.3126 - val_accuracy: 0.9491
Epoch 17/20
- val_loss: 0.1463 - val_accuracy: 0.9772
Epoch 18/20
- val_loss: 0.0764 - val_accuracy: 0.9866
Epoch 19/20
- val_loss: 0.0627 - val_accuracy: 0.9893
Epoch 20/20
- val_loss: 0.0774 - val_accuracy: 0.9799
39/39 - 1s - loss: 0.0842 - accuracy: 0.9815 - 1s/epoch - 30ms/step
```

```
Accuracy Score: 0.9815261044176706
Confusion Matrix:
 [[64 0 0 0 0 0 0 0 0
 1 0 59 0 0 0
    0 0 0 0 80 0
0 0 0 0 0 79
                    9 9
9 9
                            0
            0
               0 0 82
                      0 0
                            0
            0
               0 0
                    0 0 62
                            0
          0 0
               0 0
                    0
                       0 0 59
       0 0 0 0 0 0 0 0 0 68 0
1 0 0 0 0 0 0 1 0 1 48
       0 0 0 0 0 0 0 0 0
    0 0 0 0 0 0 0 0 0 0 0 0 0 0 83 0]
0 0 2 0 0 0 0 0 0 0 0 0 7 0 071]]
```

Case 8:

```
Epoch 1/20
0.5204 - val loss: 0.6361 - val accuracy: 0.8126
Epoch 2/20
94/94 [============= - 10s 103ms/step - loss: 2.6303 - accuracy:
0.7070 - val loss: 0.5215 - val accuracy: 0.8768
Epoch 3/20
- val_loss: 0.2861 - val_accuracy: 0.9411
Epoch 4/20
94/94 [================================ ] - 9s 95ms/step - loss: 2.1105 - accuracy: 0.6983
- val_loss: 0.3177 - val_accuracy: 0.9210
Epoch 5/20
94/94 [===============================] - 9s 96ms/step - loss: 1.5554 - accuracy: 0.7371
- val_loss: 0.4732 - val_accuracy: 0.9317
Epoch 6/20
- val_loss: 0.4730 - val_accuracy: 0.9290
Epoch 7/20
- val loss: 0.3438 - val accuracy: 0.9424
Epoch 8/20
0.7482 - val_loss: 0.3549 - val_accuracy: 0.9545
Epoch 9/20
94/94 [=============================== ] - 9s 96ms/step - loss: 1.4271 - accuracy: 0.7301
- val_loss: 0.4632 - val_accuracy: 0.9424
Epoch 10/20
- val loss: 0.2020 - val accuracy: 0.9505
Epoch 11/20
- val_loss: 0.5440 - val_accuracy: 0.9116
Epoch 12/20
- val_loss: 0.3777 - val_accuracy: 0.9545
Epoch 13/20
```

```
0.7746 - val_loss: 0.2717 - val_accuracy: 0.9679
Epoch 14/20
0.7739 - val loss: 0.1619 - val accuracy: 0.9786
Epoch 15/20
0.7843 - val loss: 0.1616 - val accuracy: 0.9759
Epoch 16/20
0.7833 - val loss: 0.1729 - val accuracy: 0.9772
Epoch 17/20
- val_loss: 0.2661 - val_accuracy: 0.9585
Epoch 18/20
94/94 [==============] - 9s 101ms/step - loss: 0.9820 - accuracy:
0.7706 - val_loss: 0.3867 - val_accuracy: 0.9438
Epoch 19/20
0.7230 - val_loss: 0.4042 - val_accuracy: 0.9264
Epoch 20/20
94/94 [=============== - 10s 103ms/step - loss: 0.9950 - accuracy:
0.7706 - val_loss: 0.3741 - val_accuracy: 0.9545
39/39 - 2s - loss: 0.2249 - accuracy: 0.9671 - 2s/epoch - 42ms/step
```

Case 9:

```
Epoch 1/20
0.5345 - val loss: 0.6465 - val accuracy: 0.7778
Epoch 2/20
- val loss: 0.4340 - val accuracy: 0.8594
Epoch 3/20
94/94 [=================================] - 10s 106ms/step - loss: 0.8240 - accuracy:
0.8774 - val_loss: 0.1972 - val_accuracy: 0.9531
Epoch 4/20
- val loss: 0.3000 - val accuracy: 0.9304
Epoch 5/20
0.8999 - val_loss: 0.2546 - val_accuracy: 0.9398
Epoch 6/20
- val_loss: 0.1322 - val_accuracy: 0.9679
Epoch 7/20
94/94 [==============] - 11s 116ms/step - loss: 0.5693 - accuracy:
0.8912 - val loss: 0.1901 - val accuracy: 0.9398
Epoch 8/20
0.9099 - val_loss: 0.1081 - val_accuracy: 0.9625
Epoch 9/20
- val_loss: 0.1843 - val_accuracy: 0.9545
Epoch 10/20
94/94 [==============] - 9s 101ms/step - loss: 0.4865 - accuracy:
0.9046 - val loss: 0.1165 - val accuracy: 0.9692
Epoch 11/20
- val_loss: 0.0727 - val_accuracy: 0.9826
Epoch 12/20
- val_loss: 0.1494 - val_accuracy: 0.9705
Epoch 13/20
```

```
- val_loss: 0.1577 - val_accuracy: 0.9625
Epoch 14/20
- val_loss: 0.1243 - val_accuracy: 0.9732
Epoch 15/20
- val loss: 0.1850 - val accuracy: 0.9746
Epoch 16/20
- val loss: 0.0894 - val accuracy: 0.9799
Epoch 17/20
- val_loss: 0.1452 - val_accuracy: 0.9746
Epoch 18/20
- val_loss: 0.0395 - val_accuracy: 0.9893
Epoch 19/20
94/94 [=========================== - 8s 90ms/step - loss: 0.1846 - accuracy: 0.9461
- val_loss: 0.0286 - val_accuracy: 0.9906
Epoch 20/20
- val_loss: 0.2080 - val_accuracy: 0.9639
39/39 - 1s - loss: 0.2153 - accuracy: 0.9663 - 1s/epoch - 37ms/step
```

Case 10:

```
Epoch 1/20
0.6309 - val loss: 0.4790 - val accuracy: 0.8474
Epoch 2/20
94/94 [============== ] - 10s 103ms/step - loss: 1.6748 - accuracy:
0.7914 - val loss: 0.3392 - val accuracy: 0.9157
Epoch 3/20
- val_loss: 0.2589 - val_accuracy: 0.9371
Epoch 4/20
0.8272 - val loss: 0.2916 - val accuracy: 0.9197
Epoch 5/20
- val_loss: 0.2076 - val_accuracy: 0.9465
Epoch 6/20
0.8583 - val_loss: 0.1926 - val_accuracy: 0.9585
Epoch 7/20
0.8617 - val loss: 0.2247 - val accuracy: 0.9545
Epoch 8/20
- val_loss: 0.1514 - val_accuracy: 0.9679
Epoch 9/20
- val_loss: 0.1523 - val_accuracy: 0.9772
Epoch 10/20
- val loss: 0.2405 - val accuracy: 0.9572
Epoch 11/20
0.8610 - val_loss: 0.1561 - val_accuracy: 0.9746
Epoch 12/20
94/94 [================ - 10s 103ms/step - loss: 0.4141 - accuracy:
0.8908 - val_loss: 0.1454 - val_accuracy: 0.9759
Epoch 13/20
```

```
94/94 [================================] - 10s 107ms/step - loss: 0.4602 - accuracy:
0.8875 - val loss: 0.3264 - val accuracy: 0.9545
Epoch 14/20
0.8804 - val loss: 0.0875 - val accuracy: 0.9853
Epoch 15/20
0.8808 - val loss: 0.1120 - val accuracy: 0.9866
Epoch 16/20
- val loss: 0.2309 - val accuracy: 0.9665
Epoch 17/20
0.8835 - val_loss: 0.2061 - val_accuracy: 0.9759
Epoch 18/20
- val_loss: 0.1349 - val_accuracy: 0.9746
Epoch 19/20
- val_loss: 0.3538 - val_accuracy: 0.9572
Epoch 20/20
94/94 [================================ ] - 9s 97ms/step - loss: 0.5812 - accuracy: 0.8691
- val_loss: 0.1297 - val_accuracy: 0.9759
39/39 - 2s - loss: 0.1513 - accuracy: 0.9743 - 2s/epoch - 44ms/step
```

Case 11:

```
Epoch 1/20
94/94 [=============== - - 13s 102ms/step - loss: 3.9236 - accuracy:
0.4876 - val loss: 0.5602 - val accuracy: 0.8273
Epoch 2/20
94/94 [============= - 10s 108ms/step - loss: 2.5140 - accuracy:
0.7050 - val loss: 0.4470 - val accuracy: 0.8715
Epoch 3/20
- val_loss: 0.4332 - val_accuracy: 0.9036
Epoch 4/20
94/94 [=============================== ] - 9s 90ms/step - loss: 1.6504 - accuracy: 0.7321
- val loss: 0.3099 - val accuracy: 0.9304
Epoch 5/20
0.7545 - val_loss: 0.5925 - val_accuracy: 0.8862
Epoch 6/20
0.7251 - val_loss: 0.3575 - val_accuracy: 0.9478
Epoch 7/20
0.7659 - val loss: 0.2892 - val accuracy: 0.9625
Epoch 8/20
0.7800 - val_loss: 0.6050 - val_accuracy: 0.9170
Epoch 9/20
0.7522 - val_loss: 1.1144 - val_accuracy: 0.8969
Epoch 10/20
0.7478 - val loss: 1.3051 - val accuracy: 0.8581
Epoch 11/20
0.7465 - val_loss: 0.3111 - val_accuracy: 0.9411
Epoch 12/20
- val_loss: 0.8020 - val_accuracy: 0.8728
Epoch 13/20
```

```
- val_loss: 0.3552 - val_accuracy: 0.9451
Epoch 14/20
- val_loss: 0.3207 - val_accuracy: 0.9518
Epoch 15/20
- val loss: 0.2978 - val accuracy: 0.9585
Epoch 16/20
- val_loss: 0.2409 - val_accuracy: 0.9679
Epoch 17/20
- val_loss: 0.3140 - val_accuracy: 0.9625
Epoch 18/20
- val_loss: 0.3320 - val_accuracy: 0.9652
Epoch 19/20
- val_loss: 0.3576 - val_accuracy: 0.9545
Epoch 20/20
- val_loss: 0.2285 - val_accuracy: 0.9786
39/39 - 2s - loss: 0.2418 - accuracy: 0.9735 - 2s/epoch - 43ms/step
```

Case 12:

```
Epoch 1/20
0.4900 - val loss: 0.9445 - val accuracy: 0.6867
Epoch 2/20
94/94 [============== - 16s 165ms/step - loss: 0.6540 - accuracy:
0.7847 - val loss: 0.6287 - val accuracy: 0.8005
Epoch 3/20
94/94 [============================] - 16s 174ms/step - loss: 0.4938 - accuracy:
0.8436 - val_loss: 0.4262 - val_accuracy: 0.8675
Epoch 4/20
94/94 [================ - 17s 185ms/step - loss: 0.4209 - accuracy:
0.8644 - val loss: 0.5141 - val accuracy: 0.8407
Epoch 5/20
94/94 [============== - 16s 169ms/step - loss: 0.2906 - accuracy:
0.9173 - val_loss: 0.2592 - val_accuracy: 0.9438
Epoch 6/20
0.9511 - val_loss: 0.2683 - val_accuracy: 0.9237
Epoch 7/20
0.9531 - val loss: 0.2072 - val accuracy: 0.9371
Epoch 8/20
94/94 [=========================== ] - 14s 152ms/step - loss: 0.1344 - accuracy:
0.9605 - val_loss: 0.3433 - val_accuracy: 0.8822
Epoch 9/20
0.9474 - val_loss: 0.2034 - val_accuracy: 0.9652
Epoch 10/20
0.9829 - val loss: 0.1433 - val accuracy: 0.9625
Epoch 11/20
0.9866 - val_loss: 0.1099 - val_accuracy: 0.9786
Epoch 12/20
0.9829 - val_loss: 0.1458 - val_accuracy: 0.9746
Epoch 13/20
```

```
0.9752 - val loss: 0.1593 - val accuracy: 0.9679
Epoch 14/20
0.9675 - val loss: 0.1717 - val accuracy: 0.9572
Epoch 15/20
94/94 [=============== - 17s 183ms/step - loss: 0.0785 - accuracy:
0.9769 - val loss: 0.1343 - val accuracy: 0.9705
Epoch 16/20
0.9943 - val loss: 0.1272 - val accuracy: 0.9786
Epoch 17/20
94/94 [============================] - 16s 170ms/step - loss: 0.1630 - accuracy:
0.9541 - val_loss: 0.2145 - val_accuracy: 0.9558
Epoch 18/20
0.9591 - val_loss: 0.1422 - val_accuracy: 0.9719
Epoch 19/20
94/94 [============== - 17s 178ms/step - loss: 0.1158 - accuracy:
0.9705 - val_loss: 0.1377 - val_accuracy: 0.9719
Epoch 20/20
94/94 [================ - 16s 173ms/step - loss: 0.0550 - accuracy:
0.9869 - val_loss: 0.2459 - val_accuracy: 0.9545
39/39 - 2s - loss: 0.1170 - accuracy: 0.9703 - 2s/epoch - 53ms/step
```

```
Accuracy Score: 0.970281124497992
Confusion Matrix:
[[58 1 0 0 0 0 0 0 0 0
9
       9
         0 87
0 0
             0
                   0
             0
                     9
         0 0
0 0
                   0
               0 65
                 0 70
     0 0
2 0
         0 0
0 0
             1
0
               9 9
9 9
                   1 5 62
0 3 0
     0 0 0 0 0 0 0 0 1 0 0
```

Case 13:

```
Epoch 1/20
0.5512 - val loss: 0.6799 - val accuracy: 0.8233
Epoch 2/20
94/94 [============== - 16s 172ms/step - loss: 1.1003 - accuracy:
0.8178 - val loss: 0.3162 - val accuracy: 0.9050
Epoch 3/20
94/94 [=========================== ] - 18s 187ms/step - loss: 1.0296 - accuracy:
0.8336 - val_loss: 0.5827 - val_accuracy: 0.8220
Epoch 4/20
0.7686 - val loss: 0.4491 - val accuracy: 0.8474
Epoch 5/20
0.8764 - val_loss: 0.2278 - val_accuracy: 0.9331
Epoch 6/20
0.8965 - val_loss: 0.1967 - val_accuracy: 0.9424
Epoch 7/20
0.9002 - val loss: 0.2162 - val accuracy: 0.9451
Epoch 8/20
0.9102 - val_loss: 0.1502 - val_accuracy: 0.9545
Epoch 9/20
0.8935 - val_loss: 0.3223 - val_accuracy: 0.9411
Epoch 10/20
0.8858 - val loss: 0.5750 - val accuracy: 0.8849
Epoch 11/20
0.8741 - val_loss: 0.3720 - val_accuracy: 0.9277
Epoch 12/20
0.9149 - val_loss: 0.1345 - val_accuracy: 0.9679
Epoch 13/20
```

```
0.9270 - val_loss: 0.1993 - val_accuracy: 0.9692
Epoch 14/20
0.8891 - val loss: 0.3870 - val accuracy: 0.8929
Epoch 15/20
0.9183 - val loss: 0.2131 - val accuracy: 0.9612
Epoch 16/20
0.9277 - val loss: 0.2364 - val accuracy: 0.9598
Epoch 17/20
94/94 [================================== ] - 17s 179ms/step - loss: 0.3605 - accuracy:
0.9243 - val_loss: 0.1148 - val_accuracy: 0.9799
Epoch 18/20
0.9290 - val_loss: 0.1892 - val_accuracy: 0.9679
Epoch 19/20
94/94 [=============== - 17s 180ms/step - loss: 0.4129 - accuracy:
0.9173 - val_loss: 0.1521 - val_accuracy: 0.9679
Epoch 20/20
0.9169 - val_loss: 0.1318 - val_accuracy: 0.9799
39/39 - 3s - loss: 0.1124 - accuracy: 0.9791 - 3s/epoch - 82ms/step
```

```
Accuracy Score: 0.9791164658634538
[[79 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 [ 0 75 0 0 4 0 0 0 0 0 0 0 0 0 0 0 0
 [00630000000
[0006800000
                0 0
77 0
0 64
          0
0
             0 77
                0 0 0 66
0 0 0 0
                              9
          0
0
             0
0
                           1
                        0 81
                  9
                           0 65 0
0 0 81
0 1 1
       0 0 0 0 0 0 0 0 0 0 0 0 0 87 0
0 0 0 0 0 0 0 0 0 0 0 0 3 60
```

Case 14:

```
Epoch 1/20
0.4722 - val loss: 1.1176 - val accuracy: 0.6466
Epoch 2/20
94/94 [============== - 19s 206ms/step - loss: 1.8697 - accuracy:
0.7083 - val loss: 0.5905 - val accuracy: 0.8072
Epoch 3/20
94/94 [===========================] - 20s 209ms/step - loss: 1.5781 - accuracy:
0.7482 - val_loss: 0.5606 - val_accuracy: 0.8220
Epoch 4/20
94/94 [=============== ] - 18s 189ms/step - loss: 1.4246 - accuracy:
0.7786 - val loss: 0.6436 - val accuracy: 0.8099
Epoch 5/20
94/94 [============== - 19s 200ms/step - loss: 1.2680 - accuracy:
0.8048 - val_loss: 0.3034 - val_accuracy: 0.9076
Epoch 6/20
94/94 [=================================] - 20s 209ms/step - loss: 1.3483 - accuracy:
0.7783 - val_loss: 0.4994 - val_accuracy: 0.8166
Epoch 7/20
0.7907 - val loss: 0.4002 - val accuracy: 0.8942
Epoch 8/20
0.8299 - val_loss: 0.4398 - val_accuracy: 0.8822
Epoch 9/20
94/94 [============================] - 20s 214ms/step - loss: 0.7568 - accuracy:
0.8640 - val_loss: 0.1401 - val_accuracy: 0.9665
Epoch 10/20
0.8630 - val loss: 0.7492 - val accuracy: 0.8849
Epoch 11/20
0.8228 - val_loss: 0.2029 - val_accuracy: 0.9491
Epoch 12/20
0.8540 - val_loss: 0.7371 - val_accuracy: 0.9130
Epoch 13/20
```

```
94/94 [============================] - 22s 230ms/step - loss: 0.8247 - accuracy:
0.8493 - val_loss: 0.2521 - val_accuracy: 0.9344
Epoch 14/20
0.8577 - val_loss: 0.2257 - val_accuracy: 0.9545
Epoch 15/20
0.8557 - val loss: 0.1984 - val accuracy: 0.9585
Epoch 16/20
0.8674 - val loss: 0.1923 - val accuracy: 0.9639
Epoch 17/20
94/94 [============================] - 21s 224ms/step - loss: 1.1598 - accuracy:
0.7991 - val_loss: 0.4531 - val_accuracy: 0.9076
Epoch 18/20
0.8382 - val_loss: 0.3499 - val_accuracy: 0.9371
Epoch 19/20
0.8634 - val_loss: 0.3168 - val_accuracy: 0.9237
Epoch 20/20
0.8500 - val_loss: 0.3464 - val_accuracy: 0.9424
39/39 - 3s - loss: 0.5245 - accuracy: 0.9277 - 3s/epoch - 69ms/step
```

```
Accuracy Score: 0.927710843373494
Confusion Matrix:
[[70 0 0 0
3 0 73 0
             1 73
             9
9
                0 0 76
0 0 0
                         0
0
                       0
                     0 82
                       0 51
0 0
0 8
        2 0 0
                0 0
0 0
0 0
                       0
                         0
                    0 0 0
0 0 0
    0 4 1 0 0 0 0 0 0 0 0 8 1 0 42]]
```

Case 15:

```
Epoch 1/20
0.4792 - val loss: 0.9651 - val accuracy: 0.6586
Epoch 2/20
94/94 [============== - 17s 183ms/step - loss: 3.0773 - accuracy:
0.6423 - val loss: 0.7017 - val accuracy: 0.8394
Epoch 3/20
94/94 [============================] - 15s 163ms/step - loss: 2.7119 - accuracy:
0.6661 - val_loss: 0.4339 - val_accuracy: 0.9130
Epoch 4/20
94/94 [================ - 16s 169ms/step - loss: 2.1041 - accuracy:
0.6959 - val loss: 2.3303 - val accuracy: 0.7015
Epoch 5/20
94/94 [============== - 17s 184ms/step - loss: 2.8441 - accuracy:
0.6303 - val_loss: 0.6534 - val_accuracy: 0.8621
Epoch 6/20
0.7184 - val_loss: 0.6408 - val_accuracy: 0.8996
Epoch 7/20
0.6986 - val loss: 0.5142 - val accuracy: 0.9143
Epoch 8/20
94/94 [=================================] - 20s 210ms/step - loss: 1.8107 - accuracy:
0.7247 - val_loss: 0.6097 - val_accuracy: 0.9143
Epoch 9/20
0.7579 - val_loss: 0.7243 - val_accuracy: 0.9116
Epoch 10/20
0.7398 - val loss: 0.4681 - val accuracy: 0.9451
Epoch 11/20
0.7696 - val_loss: 0.4376 - val_accuracy: 0.9531
Epoch 12/20
94/94 [============================] - 20s 217ms/step - loss: 1.1813 - accuracy:
0.7609 - val_loss: 1.0539 - val_accuracy: 0.9050
Epoch 13/20
```

```
0.7612 - val_loss: 0.3928 - val_accuracy: 0.9545
Epoch 14/20
0.7642 - val_loss: 0.4236 - val_accuracy: 0.9545
Epoch 15/20
94/94 [================ - - 26s 282ms/step - loss: 0.9086 - accuracy:
0.7770 - val loss: 0.8189 - val accuracy: 0.9183
Epoch 16/20
0.7723 - val loss: 0.4269 - val accuracy: 0.9558
Epoch 17/20
94/94 [================================= ] - 18s 195ms/step - loss: 2.1055 - accuracy:
0.6979 - val_loss: 0.6982 - val_accuracy: 0.8902
Epoch 18/20
0.7234 - val_loss: 1.3971 - val_accuracy: 0.8715
Epoch 19/20
0.7328 - val_loss: 0.9966 - val_accuracy: 0.8835
Epoch 20/20
0.7582 - val_loss: 0.4990 - val_accuracy: 0.9357
39/39 - 2s - loss: 0.5106 - accuracy: 0.9398 - 2s/epoch - 63ms/step
```

Case 16:

```
Epoch 1/20
0.4725 - val loss: 0.7464 - val accuracy: 0.7671
Epoch 2/20
94/94 [============== - - 20s 213ms/step - loss: 1.3101 - accuracy:
0.7522 - val loss: 0.4544 - val accuracy: 0.8541
Epoch 3/20
0.8182 - val_loss: 0.2709 - val_accuracy: 0.9090
Epoch 4/20
94/94 [============================] - 21s 221ms/step - loss: 0.8965 - accuracy:
0.8476 - val loss: 0.3504 - val accuracy: 0.8996
Epoch 5/20
94/94 [============== - 19s 205ms/step - loss: 0.6678 - accuracy:
0.8868 - val_loss: 0.1710 - val_accuracy: 0.9612
Epoch 6/20
0.8962 - val_loss: 0.2485 - val_accuracy: 0.9398
Epoch 7/20
0.9029 - val loss: 0.3076 - val accuracy: 0.9250
Epoch 8/20
0.8972 - val_loss: 0.1797 - val_accuracy: 0.9612
Epoch 9/20
0.9143 - val_loss: 0.4697 - val_accuracy: 0.9197
Epoch 10/20
0.8922 - val loss: 0.3845 - val accuracy: 0.8782
Epoch 11/20
0.8955 - val_loss: 0.2871 - val_accuracy: 0.9505
Epoch 12/20
94/94 [============================] - 21s 218ms/step - loss: 0.3527 - accuracy:
0.9273 - val_loss: 0.1891 - val_accuracy: 0.9679
Epoch 13/20
```

```
0.9283 - val_loss: 0.1226 - val_accuracy: 0.9719
Epoch 14/20
0.9270 - val_loss: 0.1128 - val_accuracy: 0.9732
Epoch 15/20
94/94 [================ ] - 22s 232ms/step - loss: 0.6204 - accuracy:
0.9086 - val loss: 0.2609 - val accuracy: 0.9491
Epoch 16/20
0.9203 - val loss: 0.2296 - val accuracy: 0.9639
Epoch 17/20
94/94 [============================] - 18s 193ms/step - loss: 0.4068 - accuracy:
0.9220 - val_loss: 0.3381 - val_accuracy: 0.9558
Epoch 18/20
94/94 [================ ] - 22s 233ms/step - loss: 0.3571 - accuracy:
0.9226 - val_loss: 0.1201 - val_accuracy: 0.9772
Epoch 19/20
0.9397 - val_loss: 0.2197 - val_accuracy: 0.9813
Epoch 20/20
94/94 [================ ] - 20s 210ms/step - loss: 0.4131 - accuracy:
0.9226 - val_loss: 0.7323 - val_accuracy: 0.9050
39/39 - 3s - loss: 0.6118 - accuracy: 0.9229 - 3s/epoch - 65ms/step
```

```
Accuracy Score: 0.9228915662650602
Confusion Matrix:
 [ 0 74 0 0
[ 0 0 69 0
                 0 0 0 0 0 0 2 0 1 0 0 0 0 1 0
       0 0 0 69 0
                 0 61
              9
5
                  9
2
                     0 0 83 0
0 0 0 48
                                    0
                                        9
                  0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
                                    70 0 0
0 72 0
                                 0 70
       0 2 0 0 0 0 0 0 0 2 2 58 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 84 0
0 0 1 0 0 0 0 0 0 0 0 0 0 0 2 73
```

Case 17:

```
Epoch 1/20
0.4933 - val loss: 0.7264 - val accuracy: 0.7497
Epoch 2/20
94/94 [============== - 19s 202ms/step - loss: 1.9316 - accuracy:
0.7197 - val loss: 0.5554 - val accuracy: 0.8367
Epoch 3/20
94/94 [============================] - 18s 188ms/step - loss: 1.6052 - accuracy:
0.7686 - val_loss: 0.9924 - val_accuracy: 0.7537
Epoch 4/20
0.7960 - val loss: 0.4751 - val accuracy: 0.8849
Epoch 5/20
94/94 [==============] - 22s 239ms/step - loss: 1.1329 - accuracy:
0.8319 - val_loss: 0.3000 - val_accuracy: 0.9384
Epoch 6/20
0.8302 - val_loss: 0.2649 - val_accuracy: 0.9478
Epoch 7/20
0.8255 - val loss: 0.4989 - val accuracy: 0.8956
Epoch 8/20
0.8449 - val_loss: 0.2491 - val_accuracy: 0.9518
Epoch 9/20
94/94 [============================] - 21s 229ms/step - loss: 0.8461 - accuracy:
0.8634 - val_loss: 0.5747 - val_accuracy: 0.9331
Epoch 10/20
0.8674 - val loss: 0.0897 - val accuracy: 0.9826
Epoch 11/20
0.8875 - val_loss: 0.4431 - val_accuracy: 0.9478
Epoch 12/20
0.8510 - val_loss: 0.2400 - val_accuracy: 0.9705
Epoch 13/20
```

```
0.8758 - val loss: 0.4795 - val accuracy: 0.9197
Epoch 14/20
0.8664 - val loss: 0.1619 - val accuracy: 0.9759
Epoch 15/20
0.8543 - val loss: 0.5561 - val accuracy: 0.9224
Epoch 16/20
0.8439 - val loss: 0.2515 - val accuracy: 0.9679
Epoch 17/20
94/94 [================================= ] - 17s 185ms/step - loss: 0.6167 - accuracy:
0.8731 - val_loss: 0.1577 - val_accuracy: 0.9759
Epoch 18/20
0.8912 - val_loss: 0.4649 - val_accuracy: 0.9518
Epoch 19/20
94/94 [=============== ] - 20s 213ms/step - loss: 0.5715 - accuracy:
0.8741 - val_loss: 0.3420 - val_accuracy: 0.9572
Epoch 20/20
0.8838 - val_loss: 0.5570 - val_accuracy: 0.9451
39/39 - 3s - loss: 0.4586 - accuracy: 0.9574 - 3s/epoch - 66ms/step
```

```
Accuracy Score: 0.957429718875502
[[83 0 0 0 0 0 0 0
                  9 74
9 9
9 9
            0
                            0
                                0
                         79
                            0 64 0
0 0 58
                   0
0
0
                      0
0
                                   2 68
0 1
                            0 0
                         0 0 1 0 1 53 0
0 0 0 4 0 0 52
0 1 0 0 0 0 0
                      0
                      0
         0
0
     0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 73 0]
```

Case 18:

```
Epoch 1/20
94/94 [=================================] - 31s 229ms/step - loss: 3.9505 - accuracy:
0.4699 - val loss: 1.0950 - val accuracy: 0.6252
Epoch 2/20
94/94 [============== ] - 20s 207ms/step - loss: 3.0074 - accuracy:
0.5968 - val loss: 0.7375 - val accuracy: 0.7537
Epoch 3/20
0.6460 - val_loss: 1.0579 - val_accuracy: 0.7483
Epoch 4/20
0.6621 - val loss: 1.0352 - val accuracy: 0.7510
Epoch 5/20
0.6842 - val_loss: 0.7042 - val_accuracy: 0.8501
Epoch 6/20
0.6919 - val_loss: 0.7237 - val_accuracy: 0.8461
Epoch 7/20
0.7063 - val loss: 0.8869 - val accuracy: 0.8246
Epoch 8/20
94/94 [=========================== ] - 18s 186ms/step - loss: 1.8136 - accuracy:
0.7083 - val_loss: 0.5086 - val_accuracy: 0.8876
Epoch 9/20
0.7157 - val_loss: 1.0551 - val_accuracy: 0.8179
Epoch 10/20
0.7093 - val loss: 0.6338 - val accuracy: 0.8942
Epoch 11/20
0.7415 - val_loss: 0.5929 - val_accuracy: 0.9116
Epoch 12/20
94/94 [=========================== ] - 18s 194ms/step - loss: 1.7513 - accuracy:
0.7204 - val_loss: 0.7671 - val_accuracy: 0.9063
Epoch 13/20
```

```
94/94 [============================] - 17s 183ms/step - loss: 1.7898 - accuracy:
0.7103 - val loss: 0.6848 - val accuracy: 0.8487
Epoch 14/20
0.7328 - val loss: 0.4634 - val accuracy: 0.9009
Epoch 15/20
94/94 [================ - 17s 181ms/step - loss: 1.2622 - accuracy:
0.7599 - val loss: 0.3702 - val accuracy: 0.9585
Epoch 16/20
0.7522 - val loss: 0.3770 - val accuracy: 0.9585
Epoch 17/20
94/94 [=================================] - 17s 182ms/step - loss: 1.2194 - accuracy:
0.7495 - val_loss: 0.4495 - val_accuracy: 0.9465
Epoch 18/20
94/94 [=============== ] - 18s 190ms/step - loss: 1.0718 - accuracy:
0.7713 - val_loss: 0.7666 - val_accuracy: 0.9197
Epoch 19/20
0.7646 - val_loss: 1.6133 - val_accuracy: 0.8621
Epoch 20/20
0.7612 - val_loss: 1.7371 - val_accuracy: 0.8541
39/39 - 2s - loss: 1.8306 - accuracy: 0.8578 - 2s/epoch - 62ms/step
```

```
Accuracy Score: 0.8578313253012049
Confusion Matrix:
 [[65 0 0 2 0
 [ 0 63 0 0 0 0 0 [ 0 0 69 0 0 0 0 0 [
                                      9 9
9 9
              3 0 70
2 0 0
0 0 0
                          0
                                      0
                          72
                  0 0
0 0
                         0
0
                             0 74 0
0 0 73
                                         0
0
                  0 0 0
0 0 0
0 0 0
             0
0
                              0 0
0 0
                                      0 0 63 4
0 1 0 37
          0 3 0 0 0
0 2 0 0 0
                              0 0
0 0
                                      0 0
0 0
                                             0 0
0 0
```

INFERENCES FOR RNN-LSTM:

The following inferences can be made from the above obtained results:

- Number of layers in the network is inversely proportional to the accuracy.
- The accuracy for the most model variations is highest when there is no dropout value

III. <u>DEEP BELIEF NETWORK</u>

FEATURE EXTRACTION METHOD

CODE:

```
import librosa
import numpy
import pandas
import csv
filename = "Instruments.csv"
    ins = 'Veena'#change for each instrument
fn = 'Veena{0}.wav'.format(i+1) #change for each instrument
audio_path = 'WAV AUDIO FILES CLIPPED/Veena{0}.wav'.format(i+1) #change for each instrument
    x , sr = librosa.load(audio_path)
    #spectral centroid
     spectral_centroids = librosa.feature.spectral_centroid(x, sr=sr)[0]
    #spectral rolloff
    spectral_rolloff = librosa.feature.spectral_rolloff(x, sr=sr)[0]
    S, phase = librosa.magphase(librosa.stft(x))
    rms = librosa.feature.rms(S=S)
    #chroma
    hop_length = 512
    chromagram = librosa.feature.chroma_stft(x, sr=sr, hop_length=hop_length)
    mfccs = librosa.feature.mfcc(x, sr=sr)
    mfccs_df = pandas.DataFrame(mfccs)
     #zero-crossing rate
    zcr = librosa.feature.zero_crossing_rate(x)
```

```
oenv = librosa.onset.onset_strength(y=x, sr=sr)
  tempo = librosa.beat.tempo(onset_envelope=oenv, sr=sr)[0]
  with open(filename, 'a', newline='') as csvfile:
                                                         rows = [fn, numpy.average(spectral_centroids), numpy.var(spectral_centroids),
                                                                                                                \verb|numpy.average(spectral_rolloff)|, \verb|numpy.var(spectral_rolloff)|, \verb|numpy.average(rms)|, \verb|numpy.var(rms)|, \verb|numpy.average(rms)|, \verb|numpy.var(rms)|, \verb|numpy.average(rms)|, \verb|numpy.average(rms)|, \verb|numpy.var(rms)|, \verb|numpy.average(rms)|, \verb|numpy.av
                                                                                                                   \verb|numpy.average(chromagram)|, \verb|numpy.var(chromagram)|, \verb|numpy.average(zcr)|, \verb|numpy.var(zcr)|, \verb|tempo|, \verb|numpy.average(zcr)|, a
                                                                                                                                                                       mfccs_df.mean(axis=1)[0],
                                                                                                                                                         mfccs\_df.var(axis=1)[0], \ mfccs\_df.mean(axis=1)[1], \ mfccs\_df.var(axis=1)[1], \ mfccs\_df.mean(axis=1)[2], \ mfccs\_df.var(axis=1)[2], \ mfccs\_df.var(axi
                                                                                                                                                                     mfccs\_df.var(axis=1)[2], \ mfccs\_df.mean(axis=1)[3], \ mfccs\_df.var(axis=1)[3], \ mfccs\_df.mean(axis=1)[4], \\ mfccs\_df.var(axis=1)[4], \ mfccs\_df.var(axis=1)[4], \\ mfccs\_df.var(axis=1)[4], \ mfccs\_df.var(axi
                                                                                                                                                           mfccs_df.var(axis=1)[4], mfccs_df.mean(axis=1)[5], mfccs_df.var(axis=1)[5], mfccs_df.mean(axis=1)[6],
                                                                                                                                                                       \label{eq:mfccs_df.var} \texttt{mfccs\_df.var(axis=1)[6], mfccs\_df.mean(axis=1)[7], mfccs\_df.var(axis=1)[7], mfccs\_df.mean(axis=1)[8], mfccs\_df.var(axis=1)[8], mfccs\_df.var(axi
                                                                                                                                                                       \verb|mfccs_df.var(axis=1)[8]|, \verb|mfccs_df.mean(axis=1)[9]|, \verb|mfccs_df.var(axis=1)[9]|, \verb|mfccs_df.mean(axis=1)[10]|, \verb|mfccs_df.var(axis=1)[10]|, \verb|mfccs_df.var(
                                                                                                                                                                       mfccs_df.var(axis=1)[10], mfccs_df.mean(axis=1)[11], mfccs_df.var(axis=1)[11], mfccs_df.var(axis=1)[12], mfccs_df.var(axis=1)[12], mfccs_df.var(axis=1)[12], mfccs_df.var(axis=1)[13], mfccs_df.var(axis=1)[14],
                                                                                                                                                                       mfccs_df.var(axis=1)[14], mfccs_df.mean(axis=1)[15], mfccs_df.var(axis=1)[15], mfccs_df.mean(axis=1)[16], mfccs_df.var(axis=1)[16], mfccs_df.var(axis=1)[17], mfccs_df.var(axis=1)[18], mfccs_df.var(axi
                                                                                                                                                                         \verb|mfccs_df.mean(axis=1)[18]|, \verb|mfccs_df.var(axis=1)[18]|, \verb|mfccs_df.mean(axis=1)[19]|, \verb|mfccs_df.var(axis=1)[19]|, \verb|mfccs_df.var(axis=1)[19]|, \verb|mfccs_df.mean(axis=1)[19]|, \verb|mfccs
                                                                                                                                                        ins]
                                                            csvwriter = csv.writer(csvfile)
                                                         csvwriter.writerow(rows)
```

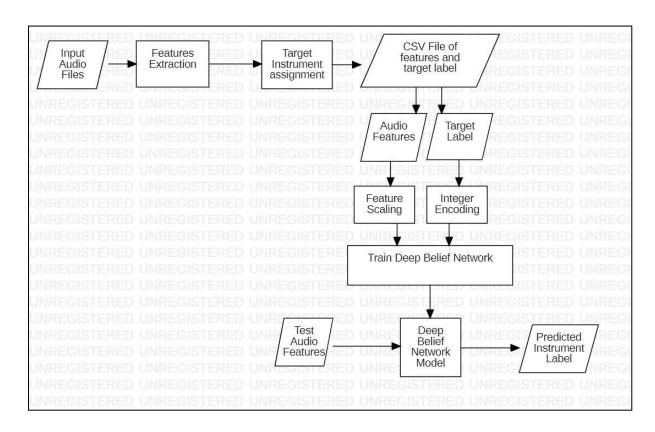
OUTPUT: <u>CSV file</u> is generated with 51 features, target label and audio sample filename.

METHODOLOGY

Deep Belief Network (DBN), a powerful generative model that uses a deep architecture of multiple stacks of Restricted Boltzmann machines(RBM) is used for musical instrument classification. 51 features such as Mel Frequency Cepstral Coefficients (MFCC), Tempo, Zero Crossing Rate, Spectral Centroid, Spectral Roll-off, RMS, Chroma and their corresponding means and variances are extracted from the audio samples and stored in a CSV file along with the file name of the audio sample. The target i.e the instrument name is annotated for each audio sample.

The obtained features are then scaled and the target, which is the name of the instrument, is integer coded. The dataset is split into train and test sets. Several Deep Belief Networks are built by tuning the hyperparameters such as number of layers, number of Restricted Boltzmann Machines (RBMs) and dropouts. The network is trained and tested over the dataset and accuracy and confusion matrix is obtained.

BLOCK DIAGRAM



IMPLEMENTATION CODE

```
#case4
import numpy as np
import pandas as pd
np.random.seed(3529)
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
from sklearn.metrics import confusion_matrix
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import StandardScaler
from dbn.tensorflow import SupervisedDBNClassification
dataset = pd.read_csv("Instruments.csv")
dataset = dataset.drop(["File Name"], axis=1)
dataset['Instrument'] = le.fit_transform(dataset['Instrument'])
X = np.array(dataset.drop(["Instrument"], axis=1))
Y = np.array(dataset.Instrument.values)
ss=StandardScaler()
X = ss.fit\_transform(X)
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state=0)
```

RESULTS

The hyperparameters that were varied in DBN:

- Number of layers (2, 4, 6)
- Number of Restricted Boltzmann Machines (RBM) in layers (128, 256)
- Dropout (0.05, 0.1, 0.2)

The hyperparameters that were set constant in DBN:

- Activation Function: Sigmoid
- Learning Rate: 0.4
- Learning Rate of RBM: 0.05

• Number of Epochs of RBM: 10

• Number of Iterations of Back Propagation: 100

Case	Number of	Number of	Dropout	Accuracy
	Layers	RBMs in		
		Each Layer		
1	2	128	0.05	92.6471
2	2	128	0.1	85.2941
3	2	128	0.2	81.3725
4	2	256	0.05	93.1373
5	2	256	0.1	92.6471
6	2	256	0.2	87.2549
7	4	128	0.05	85.2941
8	4	128	0.1	71.0784
9	4	128	0.2	75.4902
10	4	256	0.05	85.7843
11	4	256	0.1	89.7059
12	4	256	0.2	78.9216
13	6	128	0.05	61.7647
14	6	128	0.1	56.3725
15	6	128	0.2	49.5098
16	6	256	0.05	90.1961
17	6	256	0.1	85.7843
18	6	256	0.2	84.3137

RESULT SNAPSHOTS FOR DEEP BELIEF NETWORK

Case 1:

[START] Pre-training step:

>> Epoch 1 finished

RBM Reconstruction error 43.599544

>> Epoch 2 finished	RBM Reconstruction error 42.976945
>> Epoch 3 finished	RBM Reconstruction error 42.911898
>> Epoch 4 finished	RBM Reconstruction error 42.992147
>> Epoch 5 finished	RBM Reconstruction error 43.180143
>> Epoch 6 finished	RBM Reconstruction error 43.271735
>> Epoch 7 finished	RBM Reconstruction error 43.482608
>> Epoch 8 finished	RBM Reconstruction error 43.651730
>> Epoch 9 finished	RBM Reconstruction error 44.033955
>> Epoch 10 finished	RBM Reconstruction error 43.820545
>> Epoch 1 finished	RBM Reconstruction error 9.997884
>> Epoch 2 finished	RBM Reconstruction error 6.580758
>> Epoch 3 finished	RBM Reconstruction error 5.279427
>> Epoch 4 finished	RBM Reconstruction error 4.759158
>> Epoch 5 finished	RBM Reconstruction error 4.508030
>> Epoch 6 finished	RBM Reconstruction error 4.248742
>> Epoch 7 finished	RBM Reconstruction error 4.053245
>> Epoch 8 finished	RBM Reconstruction error 3.889613
>> Epoch 9 finished	RBM Reconstruction error 3.757097
>> Epoch 10 finished	RBM Reconstruction error 3.596914

>> Epoch 0 finished	ANN training loss 2.037231
>> Epoch 1 finished	ANN training loss 2.530553
>> Epoch 2 finished	ANN training loss 1.891335
>> Epoch 3 finished	ANN training loss 1.708371
>> Epoch 4 finished	ANN training loss 1.705091
>> Epoch 5 finished	ANN training loss 1.508305
>> Epoch 6 finished	ANN training loss 1.467426
>> Epoch 7 finished	ANN training loss 1.478177
>> Epoch 8 finished	ANN training loss 1.308451
>> Epoch 9 finished	ANN training loss 1.238117
>> Epoch 10 finished	ANN training loss 1.209970

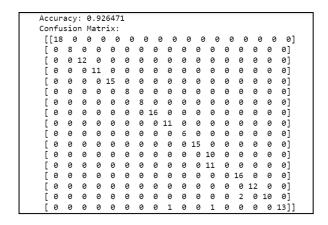
- >> Epoch 11 finished ANN training loss 1.123635
- >> Epoch 12 finished ANN training loss 1.187243
- >> Epoch 13 finished ANN training loss 1.089304
- >> Epoch 14 finished ANN training loss 1.323144
- >> Epoch 15 finished ANN training loss 1.095376
- >> Epoch 16 finished ANN training loss 1.378574
- >> Epoch 17 finished ANN training loss 1.214751
- >> Epoch 18 finished ANN training loss 1.263677
- >> Epoch 19 finished ANN training loss 0.960594
- >> Epoch 20 finished ANN training loss 0.875466
- >> Epoch 21 finished ANN training loss 0.955809
- >> Epoch 22 finished ANN training loss 0.862924
- >> Epoch 23 finished ANN training loss 0.911252
- >> Epoch 24 finished ANN training loss 0.805782
- >> Epoch 25 finished ANN training loss 0.961703
- >> Epoch 26 finished ANN training loss 0.942794
- >> Epoch 27 finished ANN training loss 0.728192
- >> Epoch 28 finished ANN training loss 0.667025
- >> Epoch 29 finished ANN training loss 0.763087
- >> Epoch 30 finished ANN training loss 0.751965
- >> Epoch 31 finished ANN training loss 0.928952
- >> Epoch 32 finished ANN training loss 0.774827
- >> Epoch 33 finished ANN training loss 0.587178
- >> Epoch 34 finished ANN training loss 0.665162
- >> Epoch 35 finished ANN training loss 0.776877
- >> Epoch 36 finished ANN training loss 0.565274
- >> Epoch 37 finished ANN training loss 0.569235
- >> Epoch 38 finished ANN training loss 0.596031
- >> Epoch 39 finished ANN training loss 0.577027
- >> Epoch 40 finished ANN training loss 0.553418
- >> Epoch 41 finished ANN training loss 0.620908
- >> Epoch 42 finished ANN training loss 0.571162

>> Epoch 43 finished ANN	training loss 0.526/46
--------------------------	------------------------

- >> Epoch 44 finished ANN training loss 0.530537
- >> Epoch 45 finished ANN training loss 0.608071
- >> Epoch 46 finished ANN training loss 0.593929
- >> Epoch 47 finished ANN training loss 0.494764
- >> Epoch 48 finished ANN training loss 0.487681
- >> Epoch 49 finished ANN training loss 0.564391
- >> Epoch 50 finished ANN training loss 0.523037
- >> Epoch 51 finished ANN training loss 0.474280
- >> Epoch 52 finished ANN training loss 0.510944
- >> Epoch 53 finished ANN training loss 0.475152
- >> Epoch 54 finished ANN training loss 0.486462
- >> Epoch 55 finished ANN training loss 0.456293
- >> Epoch 56 finished ANN training loss 0.475490
- >> Epoch 57 finished ANN training loss 0.393183
- >> Epoch 58 finished ANN training loss 0.413918
- >> Epoch 59 finished ANN training loss 0.437031
- >> Epoch 60 finished ANN training loss 0.432454
- >> Epoch 61 finished ANN training loss 0.679555
- >> Epoch 62 finished ANN training loss 0.398085
- >> Epoch 63 finished ANN training loss 0.353334
- >> Epoch 64 finished ANN training loss 0.425641
- >> Epoch 65 finished ANN training loss 0.476713
- 1
- >> Epoch 66 finished ANN training loss 0.380425
- >> Epoch 67 finished ANN training loss 0.410107
- >> Epoch 68 finished ANN training loss 0.411578
- >> Epoch 69 finished ANN training loss 0.321684
- >> Epoch 70 finished ANN training loss 0.305066
- >> Epoch 71 finished ANN training loss 0.314879
- >> Epoch 72 finished ANN training loss 0.280945
- >> Epoch 73 finished ANN training loss 0.254134
- >> Epoch 74 finished ANN training loss 0.246903

>> Epoch 75 finished	ANN training loss 0.234316
>> Epoch 76 finished	ANN training loss 0.282456
>> Epoch 77 finished	ANN training loss 0.230592
>> Epoch 78 finished	ANN training loss 0.209004
>> Epoch 79 finished	ANN training loss 0.334618
>> Epoch 80 finished	ANN training loss 0.222582
>> Epoch 81 finished	ANN training loss 0.204601
>> Epoch 82 finished	ANN training loss 0.203881
>> Epoch 83 finished	ANN training loss 0.235314
>> Epoch 84 finished	ANN training loss 0.269530
>> Epoch 85 finished	ANN training loss 0.205300
>> Epoch 86 finished	ANN training loss 0.197776
>> Epoch 87 finished	ANN training loss 0.294495
>> Epoch 88 finished	ANN training loss 0.181181
>> Epoch 89 finished	ANN training loss 0.174272
>> Epoch 90 finished	ANN training loss 0.204566
>> Epoch 91 finished	ANN training loss 0.211796
>> Epoch 92 finished	ANN training loss 0.196329
>> Epoch 93 finished	ANN training loss 0.162750
>> Epoch 94 finished	ANN training loss 0.172406
>> Epoch 95 finished	ANN training loss 0.191705
>> Epoch 96 finished	ANN training loss 0.195645
>> Epoch 97 finished	ANN training loss 0.226054
>> Epoch 98 finished	ANN training loss 0.160112
>> Epoch 99 finished	ANN training loss 0.182126

[END] Fine tuning step



Case 2:

[START] Pre-training step:

>> Epoch 1 finished	RBM Reconstruction error 43.782150
>> Epoch 2 finished	RBM Reconstruction error 42.928241
>> Epoch 3 finished	RBM Reconstruction error 42.655945
>> Epoch 4 finished	RBM Reconstruction error 42.693698
>> Epoch 5 finished	RBM Reconstruction error 42.927321
>> Epoch 6 finished	RBM Reconstruction error 43.209267
>> Epoch 7 finished	RBM Reconstruction error 43.481772
>> Epoch 8 finished	RBM Reconstruction error 43.672546
>> Epoch 9 finished	RBM Reconstruction error 44.062125
>> Epoch 10 finished	RBM Reconstruction error 43.818609
>> Epoch 1 finished	RBM Reconstruction error 10.392400
>> Epoch 2 finished	RBM Reconstruction error 6.700058
>> Epoch 3 finished	RBM Reconstruction error 5.298466
>> Epoch 4 finished	RBM Reconstruction error 4.796281
>> Epoch 5 finished	RBM Reconstruction error 4.568461
>> Epoch 6 finished	RBM Reconstruction error 4.397188
>> Epoch 7 finished	RBM Reconstruction error 4.223192
>> Epoch 8 finished	RBM Reconstruction error 4.061265
>> Epoch 9 finished	RBM Reconstruction error 4.007226
>> Epoch 10 finished	RBM Reconstruction error 3.796173
[END] Pre-training step	

[END] Pre-training step

>> Epoch 0 finished ANN to	raining loss 2.0484/3
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- >> Epoch 1 finished ANN training loss 2.555068
- >> Epoch 2 finished ANN training loss 1.943166
- >> Epoch 3 finished ANN training loss 1.845320
- >> Epoch 4 finished ANN training loss 1.839659
- >> Epoch 5 finished ANN training loss 1.595223
- >> Epoch 6 finished ANN training loss 1.597985
- >> Epoch 7 finished ANN training loss 1.599554
- >> Epoch 8 finished ANN training loss 1.423829
- >> Epoch 9 finished ANN training loss 1.343475
- >> Epoch 10 finished ANN training loss 1.404213
- >> Epoch 11 finished ANN training loss 1.280437
- >> Epoch 12 finished ANN training loss 1.263719
- >> Epoch 13 finished ANN training loss 1.295932
- >> Epoch 14 finished ANN training loss 1.376567
- >> Epoch 15 finished ANN training loss 1.238679
- >> Epoch 16 finished ANN training loss 1.426598
- >> Epoch 17 finished ANN training loss 1.251217
- >> Epoch 18 finished ANN training loss 1.269683
- >> Epoch 19 finished ANN training loss 1.157708
- >> Epoch 20 finished ANN training loss 1.030899
- >> Epoch 21 finished ANN training loss 1.006503
- >> Epoch 22 finished ANN training loss 1.036715
- >> Epoch 23 finished ANN training loss 1.050383
- >> Epoch 24 finished ANN training loss 0.966062

ANN training loss 1.019315

>> Epoch 25 finished

- >> Epoch 26 finished ANN training loss 0.969783
- >> Epoch 27 finished ANN training loss 0.830298
- >> Epoch 28 finished ANN training loss 0.992098
- >> Epoch 29 finished ANN training loss 0.902074
- >> Epoch 30 finished ANN training loss 0.933671
- >> Epoch 31 finished ANN training loss 1.063265

>> Epoch 32 finished ANN	N training loss 0.949229
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- >> Epoch 33 finished ANN training loss 0.699161
- >> Epoch 34 finished ANN training loss 0.749050
- >> Epoch 35 finished ANN training loss 0.831956
- >> Epoch 36 finished ANN training loss 0.697498
- >> Epoch 37 finished ANN training loss 0.649158
- >> Epoch 38 finished ANN training loss 0.676074
- >> Epoch 39 finished ANN training loss 0.642122
- >> Epoch 40 finished ANN training loss 0.669421
- >> Epoch 41 finished ANN training loss 0.657683
- >> Epoch 42 finished ANN training loss 0.625996
- >> Epoch 43 finished ANN training loss 0.676108
- >> Epoch 44 finished ANN training loss 0.635477
- >> Epoch 45 finished ANN training loss 0.625749
- >> Epoch 46 finished ANN training loss 0.692621
- >> Epoch 47 finished ANN training loss 0.539687
- >> Epoch 48 finished ANN training loss 0.588893
- >> Epoch 49 finished ANN training loss 0.591711
- >> Epoch 50 finished ANN training loss 0.604669
- >> Epoch 51 finished ANN training loss 0.519620
- >> Epoch 52 finished ANN training loss 0.542684
- >> Epoch 53 finished ANN training loss 0.644590
- >> Epoch 54 finished ANN training loss 0.529678
- >> Epoch 55 finished ANN training loss 0.498389
- >> Epoch 56 finished ANN training loss 0.540913
- >> Epoch 57 finished ANN training loss 0.478119
- >> Epoch 58 finished ANN training loss 0.516382
- >> Epoch 59 finished ANN training loss 0.503232
- >> Epoch 60 finished ANN training loss 0.565553
- >> Epoch 61 finished ANN training loss 0.475282
- >> Epoch 62 finished ANN training loss 0.508320
- >> Epoch 63 finished ANN training loss 0.459403

>> Epoch 64 finished ANN training loss 0.47	61	/0)
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- >> Epoch 65 finished ANN training loss 0.493728
- >> Epoch 66 finished ANN training loss 0.453150
- >> Epoch 67 finished ANN training loss 0.437306
- >> Epoch 68 finished ANN training loss 0.524561
- >> Epoch 69 finished ANN training loss 0.456132
- >> Epoch 70 finished ANN training loss 0.408970
- >> Epoch 71 finished ANN training loss 0.424216
- >> Epoch 72 finished ANN training loss 0.470458
- >> Epoch 73 finished ANN training loss 0.384628
- >> Epoch 74 finished ANN training loss 0.405904
- >> Epoch 75 finished ANN training loss 0.378710
- >> Epoch 76 finished ANN training loss 0.527515
- >> Epoch 77 finished ANN training loss 0.385493
- >> Epoch 78 finished ANN training loss 0.355130
- >> Epoch 79 finished ANN training loss 0.371594
- >> Epoch 80 finished ANN training loss 0.397235
- >> Epoch 81 finished ANN training loss 0.391941
- >> Epoch 82 finished ANN training loss 0.361381
- >> Epoch 83 finished ANN training loss 0.392323
- >> Epoch 84 finished

ANN training loss 0.374375

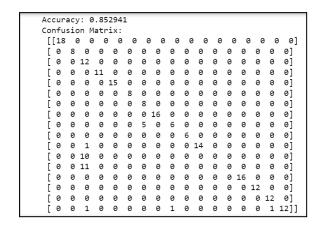
ANN training loss 0.363242

>> Epoch 85 finished

- >> Epoch 86 finished ANN training loss 0.374447
- >> Epoch 87 finished ANN training loss 0.403555
- >> Epoch 88 finished ANN training loss 0.329691
- >> Epoch 89 finished ANN training loss 0.334911
- >> Epoch 90 finished ANN training loss 0.342292
- ANN training loss 0.330106 >> Epoch 91 finished
- >> Epoch 92 finished ANN training loss 0.310393
- >> Epoch 93 finished ANN training loss 0.314418
- >> Epoch 94 finished ANN training loss 0.358456
- >> Epoch 95 finished ANN training loss 0.327619

>> Epoch 96 finished ANN training loss 0.293542
>> Epoch 97 finished ANN training loss 0.355162
>> Epoch 98 finished ANN training loss 0.440505
>> Epoch 99 finished ANN training loss 0.372699

[END] Fine tuning step



Case 3:

>> Epoch 1 finished	RBM Reconstruction error 43.579861
>> Epoch 2 finished	RBM Reconstruction error 42.648378
>> Epoch 3 finished	RBM Reconstruction error 42.461308
>> Epoch 4 finished	RBM Reconstruction error 42.441174
>> Epoch 5 finished	RBM Reconstruction error 42.601003
>> Epoch 6 finished	RBM Reconstruction error 42.680919
>> Epoch 7 finished	RBM Reconstruction error 42.801097
>> Epoch 8 finished	RBM Reconstruction error 42.832580
>> Epoch 9 finished	RBM Reconstruction error 43.133236
>> Epoch 10 finished	RBM Reconstruction error 42.917262
>> Epoch 1 finished	RBM Reconstruction error 10.114342
>> Epoch 2 finished	RBM Reconstruction error 6.303115
>> Epoch 3 finished	RBM Reconstruction error 5.110419
>> Epoch 4 finished	RBM Reconstruction error 4.639163
>> Epoch 5 finished	RBM Reconstruction error 4.457079
>> Epoch 6 finished	RBM Reconstruction error 4.256772

- >> Epoch 7 finished RBM Reconstruction error 4.115946
- >> Epoch 8 finished RBM Reconstruction error 3.963893
- >> Epoch 9 finished RBM Reconstruction error 3.893830
- >> Epoch 10 finished RBM Reconstruction error 3.757128

- >> Epoch 0 finished ANN training loss 2.084739
- >> Epoch 1 finished ANN training loss 2.693096
- >> Epoch 2 finished ANN training loss 2.023645
- >> Epoch 3 finished ANN training loss 1.936663
- >> Epoch 4 finished ANN training loss 1.938408
- >> Epoch 5 finished ANN training loss 1.711225
- >> Epoch 6 finished ANN training loss 1.646958
- >> Epoch 7 finished ANN training loss 1.562639
- >> Epoch 8 finished ANN training loss 1.439193
- >> Epoch 9 finished ANN training loss 1.368606
- >> Epoch 10 finished ANN training loss 1.421918
- >> Epoch 11 finished ANN training loss 1.213603
- >> Epoch 12 finished ANN training loss 1.258609
- >> Epoch 13 finished ANN training loss 1.207161
- >> Epoch 14 finished ANN training loss 1.266410
- >> Epoch 15 finished ANN training loss 1.160871
- >> Epoch 16 finished ANN training loss 1.171288
- >> Epoch 17 finished ANN training loss 1.218048
- >> Epoch 18 finished ANN training loss 1.098999
- >> Epoch 19 finished ANN training loss 1.062969
- >> Epoch 20 finished ANN training loss 0.980961
- >> Epoch 21 finished ANN training loss 1.051329
- >> Epoch 22 finished ANN training loss 0.966101
- >> Epoch 23 finished ANN training loss 0.999138
- >> Epoch 24 finished ANN training loss 0.893240
- >> Epoch 25 finished ANN training loss 0.988844

>> Epoch 26 finished	ANN training loss 0.925951
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- >> Epoch 27 finished ANN training loss 0.834280
- >> Epoch 28 finished ANN training loss 0.842872
- >> Epoch 29 finished ANN training loss 0.883498
- >> Epoch 30 finished ANN training loss 0.845186
- >> Epoch 31 finished ANN training loss 0.978131
- >> Epoch 32 finished ANN training loss 0.832040
- >> Epoch 33 finished ANN training loss 0.768827
- >> Epoch 34 finished ANN training loss 0.794108
- >> Epoch 35 finished ANN training loss 0.798608
- >> Epoch 36 finished ANN training loss 0.782939
- >> Epoch 37 finished ANN training loss 0.712071
- >> Epoch 38 finished ANN training loss 0.737106
- >> Epoch 39 finished ANN training loss 0.726010
- >> Epoch 40 finished ANN training loss 0.737025
- >> Epoch 41 finished ANN training loss 0.692586
- >> Epoch 42 finished ANN training loss 0.695108
- >> Epoch 43 finished ANN training loss 0.693935

ANN training loss 0.676004

ANN training loss 0.715704

ANN training loss 0.777306

>> Epoch 44 finished

>> Epoch 45 finished

>> Epoch 46 finished

- >> Epoch 47 finished ANN training loss 0.686728
- >> Epoch 48 finished ANN training loss 0.622795
- >> Epoch 49 finished ANN training loss 0.752038
- >> Epoch 50 finished ANN training loss 0.640022
- >> Epoch 51 finished ANN training loss 0.600058
- >> Epoch 52 finished ANN training loss 0.623110
- >> Epoch 53 finished ANN training loss 0.620080
- >> Epoch 54 finished ANN training loss 0.608240
- >> Epoch 55 finished ANN training loss 0.569297
- >> Epoch 56 finished ANN training loss 0.614504
- >> Epoch 57 finished ANN training loss 0.563499

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- >> Epoch 59 finished ANN training loss 0.547316
- >> Epoch 60 finished ANN training loss 0.652366
- >> Epoch 61 finished ANN training loss 0.645404
- >> Epoch 62 finished ANN training loss 0.556507
- >> Epoch 63 finished ANN training loss 0.508432
- >> Epoch 64 finished ANN training loss 0.480117
- >> Epoch 65 finished ANN training loss 0.549756
- >> Epoch 66 finished ANN training loss 0.559426
- >> Epoch 67 finished ANN training loss 0.482389
- >> Epoch 68 finished ANN training loss 0.614246
- >> Epoch 69 finished ANN training loss 0.505947
- >> Epoch 70 finished ANN training loss 0.450987
- >> Epoch 71 finished ANN training loss 0.441960
- >> Epoch 72 finished ANN training loss 0.459380
- >> Epoch 73 finished ANN training loss 0.435842
- >> Epoch 74 finished ANN training loss 0.411909
- >> Epoch 75 finished ANN training loss 0.453610
- >> Epoch 76 finished ANN training loss 0.553233
- >> Epoch 77 finished ANN training loss 0.444737
- >> Epoch 78 finished ANN training loss 0.420708
- >> Epoch 79 finished ANN training loss 0.425887
- >> Epoch 80 finished ANN training loss 0.473440
- >> Epoch 81 finished ANN training loss 0.386143
- >> Epoch 82 finished ANN training loss 0.395802
- >> Epoch 83 finished ANN training loss 0.395745
- >> Epoch 84 finished ANN training loss 0.383671
- >> Epoch 85 finished ANN training loss 0.395391
- >> Epoch 86 finished ANN training loss 0.452694
- >> Epoch 87 finished ANN training loss 0.375133
- >> Epoch 88 finished ANN training loss 0.394350
- >> Epoch 89 finished ANN training loss 0.354372

>> Epoch 90 finished	ANN training loss 0.359927
>> Epoch 91 finished	ANN training loss 0.397791
>> Epoch 92 finished	ANN training loss 0.375923
>> Epoch 93 finished	ANN training loss 0.343724
>> Epoch 94 finished	ANN training loss 0.371028
>> Epoch 95 finished	ANN training loss 0.330187
>> Epoch 96 finished	ANN training loss 0.350419
>> Epoch 97 finished	ANN training loss 0.356131
>> Epoch 98 finished	ANN training loss 0.326913
>> Epoch 99 finished	ANN training loss 0.425495

[END] Fine tuning step

Ac	cur	acy	/: (8.8	1372	5											
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Case 4:

>> Epoch 1 finished	RBM Reconstruction error 41.815052
>> Epoch 2 finished	RBM Reconstruction error 40.867323
>> Epoch 3 finished	RBM Reconstruction error 40.931374
>> Epoch 4 finished	RBM Reconstruction error 41.054906
>> Epoch 5 finished	RBM Reconstruction error 41.300267
>> Epoch 6 finished	RBM Reconstruction error 41.585167
>> Epoch 7 finished	RBM Reconstruction error 41.854288
>> Epoch 8 finished	RBM Reconstruction error 42.041261
>> Epoch 9 finished	RBM Reconstruction error 42.439653
>> Epoch 10 finished	RBM Reconstruction error 42.228696
>> Epoch 1 finished	RBM Reconstruction error 14.240275

>> Epoch 2 finished	RBM Reconstruction error	11.222238
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- >> Epoch 3 finished RBM Reconstruction error 10.207335
- >> Epoch 4 finished RBM Reconstruction error 9.561489
- >> Epoch 5 finished RBM Reconstruction error 8.943264
- >> Epoch 6 finished RBM Reconstruction error 8.386539
- >> Epoch 7 finished RBM Reconstruction error 8.092778
- >> Epoch 8 finished RBM Reconstruction error 7.484471
- >> Epoch 9 finished RBM Reconstruction error 7.351327
- >> Epoch 10 finished RBM Reconstruction error 6.951092

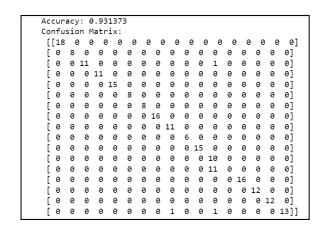
- >> Epoch 0 finished ANN training loss 3.202194
- >> Epoch 1 finished ANN training loss 2.924797
- >> Epoch 2 finished ANN training loss 1.900054
- >> Epoch 3 finished ANN training loss 1.650820
- >> Epoch 4 finished ANN training loss 1.939302
- >> Epoch 5 finished ANN training loss 1.507740
- >> Epoch 6 finished ANN training loss 1.211567
- >> Epoch 7 finished ANN training loss 1.176030
- >> Epoch 8 finished ANN training loss 1.068863
- >> Epoch 9 finished ANN training loss 1.129925
- >> Epoch 10 finished ANN training loss 0.909382
- >> Epoch 11 finished ANN training loss 0.868795
- >> Epoch 12 finished ANN training loss 0.739042
- >> Epoch 13 finished ANN training loss 0.740777
- >> Epoch 14 finished ANN training loss 0.865510
- >> Epoch 15 finished ANN training loss 0.884911
- >> Epoch 16 finished ANN training loss 0.799987
- >> Epoch 17 finished ANN training loss 0.885908
- >> Epoch 18 finished ANN training loss 0.910038
- >> Epoch 19 finished ANN training loss 0.683045
- >> Epoch 20 finished ANN training loss 0.647366

- >> Epoch 22 finished ANN training loss 0.817662
- >> Epoch 23 finished ANN training loss 1.131454
- >> Epoch 24 finished ANN training loss 0.628250
- >> Epoch 25 finished ANN training loss 0.929307
- >> Epoch 26 finished ANN training loss 0.680526
- >> Epoch 27 finished ANN training loss 0.494073
- >> Epoch 28 finished ANN training loss 0.480094
- >> Epoch 29 finished ANN training loss 0.503564
- >> Epoch 30 finished ANN training loss 0.712954
- >> Epoch 31 finished ANN training loss 1.083019
- >> Epoch 32 finished ANN training loss 1.127132
- >> Epoch 33 finished ANN training loss 0.403434
- >> Epoch 34 finished ANN training loss 0.413601
- >> Epoch 35 finished ANN training loss 0.906533
- >> Epoch 36 finished ANN training loss 0.415442
- >> Epoch 37 finished ANN training loss 0.370264
- >> Epoch 38 finished ANN training loss 0.485404
- >> Epoch 39 finished ANN training loss 0.346662
- >> Epoch 40 finished ANN training loss 0.347488
- >> Epoch 41 finished ANN training loss 0.437101
- >> Epoch 42 finished ANN training loss 0.361671
- >> Epoch 43 finished ANN training loss 0.405144
- >> Epoch 44 finished ANN training loss 0.286435
- >> Epoch 45 finished ANN training loss 0.376581
- >> Epoch 46 finished ANN training loss 0.357732
- >> Epoch 47 finished ANN training loss 0.263748
- >> Epoch 48 finished ANN training loss 0.259885
- >> Epoch 49 finished ANN training loss 0.597352
- >> Epoch 50 finished ANN training loss 0.336615
- >> Epoch 51 finished ANN training loss 0.414701
- >> Epoch 52 finished ANN training loss 0.357649

>> Epoch 53 finished A	ANN training loss 0.328534
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- >> Epoch 54 finished ANN training loss 0.307624
- >> Epoch 55 finished ANN training loss 0.565138
- >> Epoch 56 finished ANN training loss 0.333186
- >> Epoch 57 finished ANN training loss 0.229221
- >> Epoch 58 finished ANN training loss 0.283868
- >> Epoch 59 finished ANN training loss 0.274469
- >> Epoch 60 finished ANN training loss 0.193745
- >> Epoch 61 finished ANN training loss 0.343308
- >> Epoch 62 finished ANN training loss 0.250944
- >> Epoch 63 finished ANN training loss 0.221485
- >> Epoch 64 finished ANN training loss 0.287643
- >> Epoch 65 finished ANN training loss 0.261605
- >> Epoch 66 finished ANN training loss 0.232030
- >> Epoch 67 finished ANN training loss 0.212772
- >> Epoch 68 finished ANN training loss 0.341266
- >> Epoch 69 finished ANN training loss 0.228987
- >> Epoch 70 finished ANN training loss 0.247523
- >> Epoch 71 finished ANN training loss 0.185374
- >> Epoch 72 finished ANN training loss 0.239396
- >> Epoch 73 finished ANN training loss 0.171875
- >> Epoch 74 finished ANN training loss 0.168191
- >> Epoch 75 finished ANN training loss 0.168862
- >> Epoch 76 finished ANN training loss 0.319793
- >> Epoch 77 finished ANN training loss 0.205173
- >> Epoch 78 finished ANN training loss 0.174847
- >> Epoch 79 finished ANN training loss 0.159940
- >> Epoch 80 finished ANN training loss 0.169374
- >> Epoch 81 finished ANN training loss 0.161327
- >> Epoch 82 finished ANN training loss 0.162334
- >> Epoch 83 finished ANN training loss 0.354639
- >> Epoch 84 finished ANN training loss 0.189050

>> Epoch 85 finished	ANN training loss 0.203366
>> Epoch 86 finished	ANN training loss 0.143702
>> Epoch 87 finished	ANN training loss 0.228449
>> Epoch 88 finished	ANN training loss 0.123880
>> Epoch 89 finished	ANN training loss 0.156070
>> Epoch 90 finished	ANN training loss 0.139956
>> Epoch 91 finished	ANN training loss 0.231481
>> Epoch 92 finished	ANN training loss 0.118180
>> Epoch 93 finished	ANN training loss 0.113211
>> Epoch 94 finished	ANN training loss 0.141887
>> Epoch 95 finished	ANN training loss 0.177936
>> Epoch 96 finished	ANN training loss 0.131760
>> Epoch 97 finished	ANN training loss 0.125809
>> Epoch 98 finished	ANN training loss 0.121686
>> Epoch 99 finished	ANN training loss 0.165044
[END] Fine tuning step	



Case 5:

>> Epoch 1 finished	RBM Reconstruction error 41.452019
>> Epoch 2 finished	RBM Reconstruction error 40.543458
>> Epoch 3 finished	RBM Reconstruction error 40.711748
>> Epoch 4 finished	RBM Reconstruction error 41.021558
>> Epoch 5 finished	RBM Reconstruction error 41.274169

>> Epoch 6 finished	RBM Reconstruction error 41.495661
>> Epoch 7 finished	RBM Reconstruction error 41.789053
>> Epoch 8 finished	RBM Reconstruction error 41.998737
>> Epoch 9 finished	RBM Reconstruction error 42.236481
>> Epoch 10 finished	RBM Reconstruction error 42.178893
>> Epoch 1 finished	RBM Reconstruction error 14.061078
>> Epoch 2 finished	RBM Reconstruction error 10.807890
>> Epoch 3 finished	RBM Reconstruction error 9.859759
>> Epoch 4 finished	RBM Reconstruction error 9.207584
>> Epoch 5 finished	RBM Reconstruction error 8.653982
>> Epoch 6 finished	RBM Reconstruction error 8.070736
>> Epoch 7 finished	RBM Reconstruction error 7.613231
>> Epoch 8 finished	RBM Reconstruction error 7.054420
>> Epoch 9 finished	RBM Reconstruction error 6.930345

RBM Reconstruction error 6.447775

ANN training loss 3.099351

[END] Pre-training step

>> Epoch 10 finished

[START] Fine tuning step:

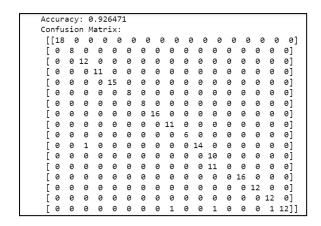
>> Epoch 0 finished

1	U
>> Epoch 1 finished	ANN training loss 3.690585
>> Epoch 2 finished	ANN training loss 1.846033
>> Epoch 3 finished	ANN training loss 1.780526
>> Epoch 4 finished	ANN training loss 1.993870
>> Epoch 5 finished	ANN training loss 1.328447
>> Epoch 6 finished	ANN training loss 1.570774
>> Epoch 7 finished	ANN training loss 1.445312
>> Epoch 8 finished	ANN training loss 1.099473
>> Epoch 9 finished	ANN training loss 0.944150
>> Epoch 10 finished	ANN training loss 1.163033
>> Epoch 11 finished	ANN training loss 0.915219
>> Epoch 12 finished	ANN training loss 1.020712
>> Epoch 13 finished	ANN training loss 0.825129
>> Epoch 14 finished	ANN training loss 1.211958

- >> Epoch 15 finished ANN training loss 0.939748
- >> Epoch 16 finished ANN training loss 0.948906
- >> Epoch 17 finished ANN training loss 1.122801
- >> Epoch 18 finished ANN training loss 1.027700
- >> Epoch 19 finished ANN training loss 0.868018
- >> Epoch 20 finished ANN training loss 0.639140
- >> Epoch 21 finished ANN training loss 0.752026
- >> Epoch 22 finished ANN training loss 0.920182
- >> Epoch 23 finished ANN training loss 0.917732
- >> Epoch 24 finished ANN training loss 0.755384
- >> Epoch 25 finished ANN training loss 1.310156
- >> Epoch 26 finished ANN training loss 0.614711
- >> Epoch 27 finished ANN training loss 0.537003
- >> Epoch 28 finished ANN training loss 0.515101
- >> Epoch 29 finished ANN training loss 0.602237
- >> Epoch 30 finished ANN training loss 0.611064
- >> Epoch 31 finished ANN training loss 1.283455
- >> Epoch 32 finished ANN training loss 1.082329
- >> Epoch 33 finished ANN training loss 0.446334
- >> Epoch 34 finished ANN training loss 0.477346
- >> Epoch 35 finished ANN training loss 0.863169
- >> Epoch 36 finished ANN training loss 0.427697
- >> Epoch 37 finished ANN training loss 0.509130
- >> Epoch 38 finished ANN training loss 0.493523
- >> Epoch 39 finished ANN training loss 0.443775
- >> Epoch 40 finished ANN training loss 0.432427
- >> Epoch 41 finished ANN training loss 0.499896
- >> Epoch 42 finished ANN training loss 0.456430
- >> Epoch 43 finished ANN training loss 0.612598
- >> Epoch 44 finished ANN training loss 0.378792
- >> Epoch 45 finished ANN training loss 0.443362
- >> Epoch 46 finished ANN training loss 0.432399

- >> Epoch 51 finished ANN training loss 0.365441
- >> Epoch 52 finished ANN training loss 0.490203
- >> Epoch 53 finished ANN training loss 0.379194
- >> Epoch 54 finished ANN training loss 0.282750
- >> Epoch 55 finished ANN training loss 0.321877
- >> Epoch 56 finished ANN training loss 0.385713
- >> Epoch 57 finished ANN training loss 0.254959
- >> Epoch 58 finished ANN training loss 0.310375
- >> Epoch 59 finished ANN training loss 0.284413
- >> Epoch 60 finished ANN training loss 0.293341
- >> Epoch 61 finished ANN training loss 0.292962
- >> Epoch 62 finished ANN training loss 0.238246
- >> Epoch 63 finished ANN training loss 0.258658
- >> Epoch 64 finished ANN training loss 0.283514
- >> Epoch 65 finished ANN training loss 0.284998
- >> Epoch 66 finished ANN training loss 0.289878
- >> Epoch 67 finished ANN training loss 0.262780
- >> Epoch 68 finished ANN training loss 0.232620
- >> Epoch 69 finished ANN training loss 0.239553
- >> Epoch 70 finished ANN training loss 0.270340
- >> Epoch 71 finished ANN training loss 0.260992
- >> Epoch 72 finished ANN training loss 0.194435
- >> Epoch 73 finished ANN training loss 0.212557
- >> Epoch 74 finished ANN training loss 0.242026
- >> Epoch 75 finished ANN training loss 0.260550
- >> Epoch 76 finished ANN training loss 0.334537
- >> Epoch 77 finished ANN training loss 0.614288
- >> Epoch 78 finished ANN training loss 0.166397

>> Epoch 79 finished	ANN training loss 0.179294
>> Epoch 80 finished	ANN training loss 0.202732
>> Epoch 81 finished	ANN training loss 0.167124
>> Epoch 82 finished	ANN training loss 0.145069
>> Epoch 83 finished	ANN training loss 0.178274
>> Epoch 84 finished	ANN training loss 0.219602
>> Epoch 85 finished	ANN training loss 0.208531
>> Epoch 86 finished	ANN training loss 0.143639
>> Epoch 87 finished	ANN training loss 0.204151
>> Epoch 88 finished	ANN training loss 0.152926
>> Epoch 89 finished	ANN training loss 0.178770
>> Epoch 90 finished	ANN training loss 0.191168
>> Epoch 91 finished	ANN training loss 0.197651
>> Epoch 92 finished	ANN training loss 0.146057
>> Epoch 93 finished	ANN training loss 0.131446
>> Epoch 94 finished	ANN training loss 0.146481
>> Epoch 95 finished	ANN training loss 0.159577
>> Epoch 96 finished	ANN training loss 0.127134
>> Epoch 97 finished	ANN training loss 0.153763
>> Epoch 98 finished	ANN training loss 0.131997
>> Epoch 99 finished	ANN training loss 0.161735
[END] Fine tuning step	



Case 6:

[START] Pre-training step:

>> Epoch 1 finished	RBM Reconstruction error 41.850058
>> Epoch 2 finished	RBM Reconstruction error 40.894614
>> Epoch 3 finished	RBM Reconstruction error 40.964123
>> Epoch 4 finished	RBM Reconstruction error 41.160024
>> Epoch 5 finished	RBM Reconstruction error 41.381316
>> Epoch 6 finished	RBM Reconstruction error 41.600540
>> Epoch 7 finished	RBM Reconstruction error 41.795998
>> Epoch 8 finished	RBM Reconstruction error 42.015733
>> Epoch 9 finished	RBM Reconstruction error 42.387931
>> Epoch 10 finished	RBM Reconstruction error 42.106841
>> Epoch 1 finished	RBM Reconstruction error 14.815408
>> Epoch 2 finished	RBM Reconstruction error 11.054120
>> Epoch 3 finished	RBM Reconstruction error 10.096250
>> Epoch 4 finished	RBM Reconstruction error 9.336926
>> Epoch 5 finished	RBM Reconstruction error 8.701862
>> Epoch 6 finished	RBM Reconstruction error 8.046473
>> Epoch 7 finished	RBM Reconstruction error 7.714260
>> Epoch 8 finished	RBM Reconstruction error 7.135638
>> Epoch 9 finished	RBM Reconstruction error 6.876895
>> Epoch 10 finished	RBM Reconstruction error 6.518500

[END] Pre-training step

>> Epoch 0 finished	ANN training loss 2.718808
>> Epoch 1 finished	ANN training loss 3.416688
>> Epoch 2 finished	ANN training loss 1.814120
>> Epoch 3 finished	ANN training loss 1.778116
>> Epoch 4 finished	ANN training loss 1.885764
>> Epoch 5 finished	ANN training loss 1.402628
>> Epoch 6 finished	ANN training loss 1.176439
>> Epoch 7 finished	ANN training loss 1.260576
>> Epoch 8 finished	ANN training loss 1.306911

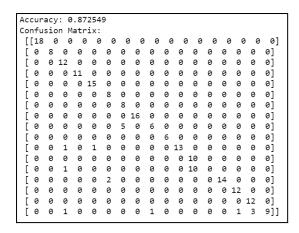
>> Epoch 9 finished	ANN training loss 1.18/582
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- >> Epoch 10 finished ANN training loss 0.969498
- >> Epoch 11 finished ANN training loss 0.976272
- >> Epoch 12 finished ANN training loss 1.066424
- >> Epoch 13 finished ANN training loss 0.802644
- >> Epoch 14 finished ANN training loss 1.021625
- >> Epoch 15 finished ANN training loss 1.114406
- >> Epoch 16 finished ANN training loss 1.012872
- >> Epoch 17 finished ANN training loss 1.134081
- >> Epoch 18 finished ANN training loss 0.989781
- >> Epoch 19 finished ANN training loss 0.777639
- >> Epoch 20 finished ANN training loss 0.725350
- >> Epoch 21 finished ANN training loss 0.752409
- >> Epoch 22 finished ANN training loss 0.768745
- >> Epoch 23 finished ANN training loss 0.967344
- >> Epoch 24 finished ANN training loss 0.754432
- >> Epoch 25 finished ANN training loss 0.860515
- >> Epoch 26 finished ANN training loss 0.645970
- >> Epoch 27 finished ANN training loss 0.638390
- >> Epoch 28 finished ANN training loss 0.506377
- >> Epoch 29 finished ANN training loss 0.621008
- >> Epoch 30 finished ANN training loss 0.624500
- >> Epoch 31 finished ANN training loss 1.045910
- >> Epoch 32 finished ANN training loss 0.916337
- >> Epoch 33 finished ANN training loss 0.512938
- >> Epoch 34 finished ANN training loss 0.589099
- >> Epoch 35 finished ANN training loss 0.860165
- >> Epoch 36 finished ANN training loss 0.592429
- >> Epoch 37 finished ANN training loss 0.591650
- >> Epoch 38 finished ANN training loss 0.455230
- >> Epoch 39 finished ANN training loss 0.557336
- >> Epoch 40 finished ANN training loss 0.518949

>> Epoch 41 finished	ANN training l	loss 0.581727
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- >> Epoch 42 finished ANN training loss 0.480343
- >> Epoch 43 finished ANN training loss 0.427032
- >> Epoch 44 finished ANN training loss 0.471221
- >> Epoch 45 finished ANN training loss 0.456733
- >> Epoch 46 finished ANN training loss 0.506937
- >> Epoch 47 finished ANN training loss 0.414737
- >> Epoch 48 finished ANN training loss 0.383229
- >> Epoch 49 finished ANN training loss 0.648670
- >> Epoch 50 finished ANN training loss 0.515359
- >> Epoch 51 finished ANN training loss 0.407835
- >> Epoch 52 finished ANN training loss 0.483763
- >> Epoch 53 finished ANN training loss 0.382627
- >> Epoch 54 finished ANN training loss 0.390435
- >> Epoch 55 finished ANN training loss 0.400153
- >> Epoch 56 finished ANN training loss 0.504246
- >> Epoch 57 finished ANN training loss 0.384212
- >> Epoch 58 finished ANN training loss 0.375902
- >> Epoch 59 finished ANN training loss 0.404311
- >> Epoch 60 finished ANN training loss 0.445458
- >> Epoch 61 finished ANN training loss 0.334489
- >> Epoch 62 finished ANN training loss 0.361197
- >> Epoch 63 finished ANN training loss 0.329897
- >> Epoch 64 finished ANN training loss 0.290024
- >> Epoch 65 finished ANN training loss 0.326399
- >> Epoch 66 finished ANN training loss 0.326028
- >> Epoch 67 finished ANN training loss 0.407774
- >> Epoch 68 finished ANN training loss 0.346832
- >> Epoch 69 finished ANN training loss 0.352717
- >> Epoch 70 finished ANN training loss 0.273628
- >> Epoch 71 finished ANN training loss 0.356377
- >> Epoch 72 finished ANN training loss 0.315419

>> Epoch 73 finished	ANN training loss 0.257262
>> Epoch 74 finished	ANN training loss 0.297363
>> Epoch 75 finished	ANN training loss 0.255015
>> Epoch 76 finished	ANN training loss 0.373465
>> Epoch 77 finished	ANN training loss 0.273223
>> Epoch 78 finished	ANN training loss 0.242477
>> Epoch 79 finished	ANN training loss 0.276181
>> Epoch 80 finished	ANN training loss 0.260728
>> Epoch 81 finished	ANN training loss 0.253995
>> Epoch 82 finished	ANN training loss 0.215218
>> Epoch 83 finished	ANN training loss 0.276128
>> Epoch 84 finished	ANN training loss 0.287112
>> Epoch 85 finished	ANN training loss 0.282366
>> Epoch 86 finished	ANN training loss 0.260891
>> Epoch 87 finished	ANN training loss 0.242337
>> Epoch 88 finished	ANN training loss 0.211894
>> Epoch 89 finished	ANN training loss 0.214325
>> Epoch 90 finished	ANN training loss 0.234565
>> Epoch 91 finished	ANN training loss 0.236237
>> Epoch 92 finished	ANN training loss 0.205068
>> Epoch 93 finished	ANN training loss 0.201132
>> Epoch 94 finished	ANN training loss 0.186718
>> Epoch 95 finished	ANN training loss 0.204233
>> Epoch 96 finished	ANN training loss 0.221619
>> Epoch 97 finished	ANN training loss 0.265261
>> Epoch 98 finished	ANN training loss 0.188180
>> Epoch 99 finished	ANN training loss 0.242758
[END] Fine tuning step	



Case 7:

>> Epoch 1 finished	RBM Reconstruction error 43.611068
>> Epoch 2 finished	RBM Reconstruction error 42.323009
>> Epoch 3 finished	RBM Reconstruction error 42.124878
>> Epoch 4 finished	RBM Reconstruction error 42.228901
>> Epoch 5 finished	RBM Reconstruction error 42.515722
>> Epoch 6 finished	RBM Reconstruction error 42.770853
>> Epoch 7 finished	RBM Reconstruction error 42.997237
>> Epoch 8 finished	RBM Reconstruction error 43.172563
>> Epoch 9 finished	RBM Reconstruction error 43.458511
>> Epoch 10 finished	RBM Reconstruction error 43.307241
>> Epoch 1 finished	RBM Reconstruction error 9.489080
>> Epoch 2 finished	RBM Reconstruction error 6.303519
>> Epoch 3 finished	RBM Reconstruction error 5.026020
>> Epoch 4 finished	RBM Reconstruction error 4.575114
>> Epoch 5 finished	RBM Reconstruction error 4.419034
>> Epoch 6 finished	RBM Reconstruction error 4.197204
>> Epoch 7 finished	RBM Reconstruction error 4.079415
>> Epoch 8 finished	RBM Reconstruction error 3.926174
>> Epoch 9 finished	RBM Reconstruction error 3.837740
>> Epoch 10 finished	RBM Reconstruction error 3.688907
>> Epoch 1 finished	RBM Reconstruction error 3.424402

>> Epoch 2 finished	RBM Reconstruction error 1.400390
>> Epoch 3 finished	RBM Reconstruction error 0.868024
>> Epoch 4 finished	RBM Reconstruction error 0.716623
>> Epoch 5 finished	RBM Reconstruction error 0.662347
>> Epoch 6 finished	RBM Reconstruction error 0.653877
>> Epoch 7 finished	RBM Reconstruction error 0.631490
>> Epoch 8 finished	RBM Reconstruction error 0.619303
>> Epoch 9 finished	RBM Reconstruction error 0.589857
>> Epoch 10 finished	RBM Reconstruction error 0.523940
>> Epoch 1 finished	RBM Reconstruction error 3.008495
>> Epoch 2 finished	RBM Reconstruction error 1.183653
>> Epoch 3 finished	RBM Reconstruction error 0.472936
>> Epoch 4 finished	RBM Reconstruction error 0.303096
>> Epoch 5 finished	RBM Reconstruction error 0.269574
>> Epoch 6 finished	RBM Reconstruction error 0.247869
>> Epoch 7 finished	RBM Reconstruction error 0.223870
>> Epoch 8 finished	RBM Reconstruction error 0.204703
>> Epoch 9 finished	RBM Reconstruction error 0.206227
>> Epoch 10 finished	RBM Reconstruction error 0.192435

>> Epoch 0 finished	ANN training loss 2.153566
>> Epoch 1 finished	ANN training loss 2.086350
>> Epoch 2 finished	ANN training loss 2.101676
>> Epoch 3 finished	ANN training loss 2.073774
>> Epoch 4 finished	ANN training loss 1.979826
>> Epoch 5 finished	ANN training loss 1.880534
>> Epoch 6 finished	ANN training loss 2.024677
>> Epoch 7 finished	ANN training loss 1.797245
>> Epoch 8 finished	ANN training loss 1.709345
>> Epoch 9 finished	ANN training loss 1.801885
>> Epoch 10 finished	ANN training loss 1.719839

- >> Epoch 11 finished ANN training loss 1.599324
- >> Epoch 12 finished ANN training loss 1.614488
- >> Epoch 13 finished ANN training loss 1.648429
- >> Epoch 14 finished ANN training loss 1.585097
- >> Epoch 15 finished ANN training loss 1.556749
- >> Epoch 16 finished ANN training loss 1.547412
- >> Epoch 17 finished ANN training loss 1.534549
- >> Epoch 18 finished ANN training loss 1.480174
- >> Epoch 19 finished ANN training loss 1.552689
- >> Epoch 20 finished ANN training loss 1.505747
- >> Epoch 21 finished ANN training loss 1.526027
- >> Epoch 22 finished ANN training loss 1.435349
- >> Epoch 23 finished ANN training loss 1.401416
- >> Epoch 24 finished ANN training loss 1.403844
- >> Epoch 25 finished ANN training loss 1.467097
- >> Epoch 26 finished ANN training loss 1.397644
- >> Epoch 27 finished ANN training loss 1.301118
- >> Epoch 28 finished ANN training loss 1.372970
- >> Epoch 29 finished ANN training loss 1.275185
- >> Epoch 30 finished ANN training loss 1.292008
- >> Epoch 31 finished ANN training loss 1.339373
- >> Epoch 32 finished ANN training loss 1.231508
- >> Epoch 33 finished ANN training loss 1.178891
- >> Epoch 34 finished ANN training loss 1.151788
- >> Epoch 35 finished ANN training loss 1.226502
- >> Epoch 36 finished ANN training loss 1.139428
- >> Epoch 37 finished ANN training loss 1.346881
- >> Epoch 38 finished ANN training loss 1.055764
- >> Epoch 39 finished ANN training loss 1.112608
- >> Epoch 40 finished ANN training loss 1.116235
- >> Epoch 41 finished ANN training loss 1.027043
- >> Epoch 42 finished ANN training loss 1.031240

>> Epoch 43 finished	ANN training loss	1.008655
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- >> Epoch 44 finished ANN training loss 1.042904
- >> Epoch 45 finished ANN training loss 0.982141
- >> Epoch 46 finished ANN training loss 0.954353
- >> Epoch 47 finished ANN training loss 1.037781
- >> Epoch 48 finished ANN training loss 0.953754
- >> Epoch 49 finished ANN training loss 0.947172
- >> Epoch 50 finished ANN training loss 0.929894
- >> Epoch 51 finished ANN training loss 1.004077
- >> Epoch 52 finished ANN training loss 0.848558
- >> Epoch 53 finished ANN training loss 0.849199
- >> Epoch 54 finished ANN training loss 0.855254
- >> Epoch 55 finished ANN training loss 0.957421
- >> Epoch 56 finished ANN training loss 0.786234
- >> Epoch 57 finished ANN training loss 0.935120
- >> Epoch 58 finished ANN training loss 0.969913
- >> Epoch 59 finished ANN training loss 0.824870
- >> Epoch 60 finished ANN training loss 0.860923
- >> Epoch 61 finished ANN training loss 0.798085
- >> Epoch 62 finished ANN training loss 0.722395
- >> Epoch 63 finished ANN training loss 0.687953
- >> Epoch 64 finished ANN training loss 0.759430
- >> Epoch 65 finished ANN training loss 0.700532

ANN training loss 0.688448

>> Epoch 66 finished

- >> Epoch 67 finished ANN training loss 0.768305
- >> Epoch 68 finished ANN training loss 0.655071
- >> Epoch 69 finished ANN training loss 0.704207
- >> Epoch 70 finished ANN training loss 0.663004
- >> Epoch 71 finished ANN training loss 0.600730
- >> Epoch 72 finished ANN training loss 0.656497
- >> Epoch 73 finished ANN training loss 0.683468
- >> Epoch 74 finished ANN training loss 0.586385

>> Epoch 75 finished	ANN training loss 0.581089
>> Epoch 76 finished	ANN training loss 0.685384
>> Epoch 77 finished	ANN training loss 0.589066
>> Epoch 78 finished	ANN training loss 0.619617
>> Epoch 79 finished	ANN training loss 0.665656
>> Epoch 80 finished	ANN training loss 0.551548
>> Epoch 81 finished	ANN training loss 0.682065
>> Epoch 82 finished	ANN training loss 0.564787
>> Epoch 83 finished	ANN training loss 0.716488
>> Epoch 84 finished	ANN training loss 0.550645
>> Epoch 85 finished	ANN training loss 0.571785
>> Epoch 86 finished	ANN training loss 0.583134
>> Epoch 87 finished	ANN training loss 0.549609
>> Epoch 88 finished	ANN training loss 0.629348
>> Epoch 89 finished	ANN training loss 0.481051
>> Epoch 90 finished	ANN training loss 0.507449
>> Epoch 91 finished	ANN training loss 0.484555
>> Epoch 92 finished	ANN training loss 0.526343
>> Epoch 93 finished	ANN training loss 0.696431
>> Epoch 94 finished	ANN training loss 0.688071
>> Epoch 95 finished	ANN training loss 0.697145
>> Epoch 96 finished	ANN training loss 0.672433
>> Epoch 97 finished	ANN training loss 0.628760
>> Epoch 98 finished	ANN training loss 0.511595
>> Epoch 99 finished	ANN training loss 0.402390

[END] Fine tuning step

Case 8:

>> Epoch 1 finished	RBM Reconstruction error 43.862700
>> Epoch 2 finished	RBM Reconstruction error 43.131203
>> Epoch 3 finished	RBM Reconstruction error 43.034387
>> Epoch 4 finished	RBM Reconstruction error 43.150786
>> Epoch 5 finished	RBM Reconstruction error 43.268527
>> Epoch 6 finished	RBM Reconstruction error 43.463622
>> Epoch 7 finished	RBM Reconstruction error 43.633679
>> Epoch 8 finished	RBM Reconstruction error 43.762163
>> Epoch 9 finished	RBM Reconstruction error 44.120498
>> Epoch 10 finished	RBM Reconstruction error 44.008004
>> Epoch 1 finished	RBM Reconstruction error 11.074246
>> Epoch 2 finished	RBM Reconstruction error 7.542301
>> Epoch 3 finished	RBM Reconstruction error 5.969889
>> Epoch 4 finished	RBM Reconstruction error 5.339882
>> Epoch 5 finished	RBM Reconstruction error 5.073346
>> Epoch 6 finished	RBM Reconstruction error 4.748718
>> Epoch 7 finished	RBM Reconstruction error 4.537777
>> Epoch 8 finished	RBM Reconstruction error 4.353485
>> Epoch 9 finished	RBM Reconstruction error 4.203320
>> Epoch 10 finished	RBM Reconstruction error 4.109626
>> Epoch 1 finished	RBM Reconstruction error 4.082773
>> Epoch 2 finished	RBM Reconstruction error 1.579649

>> Epoch 3 finished	RBM Reconstruction error 1.004919
>> Epoch 4 finished	RBM Reconstruction error 0.883139
>> Epoch 5 finished	RBM Reconstruction error 0.805344
>> Epoch 6 finished	RBM Reconstruction error 0.800753
>> Epoch 7 finished	RBM Reconstruction error 0.743595
>> Epoch 8 finished	RBM Reconstruction error 0.697562
>> Epoch 9 finished	RBM Reconstruction error 0.660530
>> Epoch 10 finished	RBM Reconstruction error 0.614470
>> Epoch 1 finished	RBM Reconstruction error 3.436915
>> Epoch 2 finished	RBM Reconstruction error 1.213127
>> Epoch 3 finished	RBM Reconstruction error 0.517840
>> Epoch 4 finished	RBM Reconstruction error 0.337148
>> Epoch 5 finished	RBM Reconstruction error 0.280519
>> Epoch 6 finished	RBM Reconstruction error 0.263852
>> Epoch 7 finished	RBM Reconstruction error 0.230781
>> Epoch 8 finished	RBM Reconstruction error 0.235931
>> Epoch 9 finished	RBM Reconstruction error 0.197512
>> Epoch 10 finished	RBM Reconstruction error 0.207421

ANN training loss 2.236196
ANN training loss 2.154101
ANN training loss 2.241579
ANN training loss 2.083358
ANN training loss 2.016831
ANN training loss 1.860466
ANN training loss 2.092218
ANN training loss 1.887535
ANN training loss 1.762124
ANN training loss 1.796763
ANN training loss 1.735878
ANN training loss 1.643505

- >> Epoch 12 finished ANN training loss 1.634183
- >> Epoch 13 finished ANN training loss 1.747173
- >> Epoch 14 finished ANN training loss 1.578380
- >> Epoch 15 finished ANN training loss 1.553794
- >> Epoch 16 finished ANN training loss 1.546279
- >> Epoch 17 finished ANN training loss 1.521776
- >> Epoch 18 finished ANN training loss 1.517237
- >> Epoch 19 finished ANN training loss 1.560154
- >> Epoch 20 finished ANN training loss 1.457675
- >> Epoch 21 finished ANN training loss 1.450898
- >> Epoch 22 finished ANN training loss 1.426255
- >> Epoch 23 finished ANN training loss 1.402175
- >> Epoch 24 finished ANN training loss 1.402005
- >> Epoch 25 finished ANN training loss 1.506710
- >> Epoch 26 finished ANN training loss 1.372117
- >> Epoch 27 finished ANN training loss 1.344714
- >> Epoch 28 finished ANN training loss 1.420532
- >> Epoch 29 finished ANN training loss 1.331015
- >> Epoch 30 finished ANN training loss 1.245345
- >> Epoch 31 finished ANN training loss 1.210212
- >> Epoch 32 finished ANN training loss 1.238708
- >> Epoch 33 finished ANN training loss 1.221147
- >> Epoch 34 finished ANN training loss 1.174080
- >> Epoch 35 finished ANN training loss 1.265576
- >> Epoch 36 finished ANN training loss 1.119487
- >> Epoch 37 finished ANN training loss 1.125924
- >> Epoch 38 finished ANN training loss 1.067990
- >> Epoch 39 finished ANN training loss 1.096529
- >> Epoch 40 finished ANN training loss 1.093302
- >> Epoch 41 finished ANN training loss 0.997662
- >> Epoch 42 finished ANN training loss 0.983302
- >> Epoch 43 finished ANN training loss 1.074824

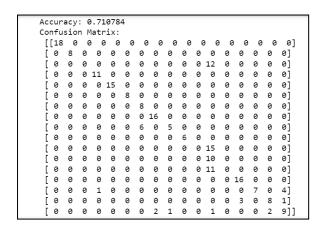
- >> Epoch 44 finished ANN training loss 1.075821
- >> Epoch 45 finished ANN training loss 0.908324
- >> Epoch 46 finished ANN training loss 0.883497
- >> Epoch 47 finished ANN training loss 0.919771
- >> Epoch 48 finished ANN training loss 1.043951
- >> Epoch 49 finished ANN training loss 0.892177
- >> Epoch 50 finished ANN training loss 0.837381
- >> Epoch 51 finished ANN training loss 0.832067
- >> Epoch 52 finished ANN training loss 0.830100
- >> Epoch 53 finished ANN training loss 0.807840
- >> Epoch 54 finished ANN training loss 0.802779
- >> Epoch 55 finished ANN training loss 0.745554
- >> Epoch 56 finished ANN training loss 0.906654
- >> Epoch 57 finished ANN training loss 0.842373
- >> Epoch 58 finished ANN training loss 0.805926
- >> Epoch 59 finished ANN training loss 0.855421
- >> Epoch 60 finished ANN training loss 0.779791
- >> Epoch 61 finished ANN training loss 0.769894
- >> Epoch 62 finished ANN training loss 0.826937
- >> Epoch 63 finished ANN training loss 0.729880
- >> Epoch 64 finished ANN training loss 0.687411
- >> Epoch 65 finished ANN training loss 0.702664
- >> Epoch 66 finished ANN training loss 0.717851
- >> Epoch 67 finished ANN training loss 0.728867
- >> Epoch 68 finished ANN training loss 0.713585
- >> Epoch 69 finished ANN training loss 0.697073
- >> Epoch 70 finished ANN training loss 0.717017

ANN training loss 0.660610

>> Epoch 71 finished

- >> Epoch 72 finished ANN training loss 0.662168
- >> Epoch 73 finished ANN training loss 0.659880
- >> Epoch 74 finished ANN training loss 0.642302
- >> Epoch 75 finished ANN training loss 0.632477

>> Epoch 76 finished	ANN training loss 0.637926
>> Epoch 77 finished	ANN training loss 0.706719
>> Epoch 78 finished	ANN training loss 0.643955
>> Epoch 79 finished	ANN training loss 0.604086
>> Epoch 80 finished	ANN training loss 0.698737
>> Epoch 81 finished	ANN training loss 0.647504
>> Epoch 82 finished	ANN training loss 0.617290
>> Epoch 83 finished	ANN training loss 0.609476
>> Epoch 84 finished	ANN training loss 0.602303
>> Epoch 85 finished	ANN training loss 0.622460
>> Epoch 86 finished	ANN training loss 0.611600
>> Epoch 87 finished	ANN training loss 0.584281
>> Epoch 88 finished	ANN training loss 0.607825
>> Epoch 89 finished	ANN training loss 0.603049
>> Epoch 90 finished	ANN training loss 0.639664
>> Epoch 91 finished	ANN training loss 0.618348
>> Epoch 92 finished	ANN training loss 0.580980
>> Epoch 93 finished	ANN training loss 0.616672
>> Epoch 94 finished	ANN training loss 0.580682
>> Epoch 95 finished	ANN training loss 0.572026
>> Epoch 96 finished	ANN training loss 0.559155
>> Epoch 97 finished	ANN training loss 0.568003
>> Epoch 98 finished	ANN training loss 0.541280
>> Epoch 99 finished	ANN training loss 0.580273



Case 9:

>> Epoch 1 finished	RBM Reconstruction error 43.231257
>> Epoch 2 finished	RBM Reconstruction error 42.375992
>> Epoch 3 finished	RBM Reconstruction error 42.243234
>> Epoch 4 finished	RBM Reconstruction error 42.214945
>> Epoch 5 finished	RBM Reconstruction error 42.350674
>> Epoch 6 finished	RBM Reconstruction error 42.420852
>> Epoch 7 finished	RBM Reconstruction error 42.563558
>> Epoch 8 finished	RBM Reconstruction error 42.631410
>> Epoch 9 finished	RBM Reconstruction error 42.861455
>> Epoch 10 finished	RBM Reconstruction error 42.754961
>> Epoch 1 finished	RBM Reconstruction error 9.733202
>> Epoch 2 finished	RBM Reconstruction error 6.367573
>> Epoch 3 finished	RBM Reconstruction error 5.268157
>> Epoch 4 finished	RBM Reconstruction error 4.822360
>> Epoch 5 finished	RBM Reconstruction error 4.627028
>> Epoch 6 finished	RBM Reconstruction error 4.419239
>> Epoch 7 finished	RBM Reconstruction error 4.263068
>> Epoch 8 finished	RBM Reconstruction error 4.109338
>> Epoch 9 finished	RBM Reconstruction error 3.995229
>> Epoch 10 finished	RBM Reconstruction error 3.835503
>> Epoch 1 finished	RBM Reconstruction error 3.560134
>> Epoch 2 finished	RBM Reconstruction error 1.531209

>> Epoch 3 finished	RBM Reconstruction error 0.942849
>> Epoch 4 finished	RBM Reconstruction error 0.857359
>> Epoch 5 finished	RBM Reconstruction error 0.817596
>> Epoch 6 finished	RBM Reconstruction error 0.791272
>> Epoch 7 finished	RBM Reconstruction error 0.732284
>> Epoch 8 finished	RBM Reconstruction error 0.695379
>> Epoch 9 finished	RBM Reconstruction error 0.723669
>> Epoch 10 finished	RBM Reconstruction error 0.674755
>> Epoch 1 finished	RBM Reconstruction error 2.774211
>> Epoch 2 finished	RBM Reconstruction error 1.092773
>> Epoch 3 finished	RBM Reconstruction error 0.459111
>> Epoch 4 finished	RBM Reconstruction error 0.312057
>> Epoch 5 finished	RBM Reconstruction error 0.263150
>> Epoch 6 finished	RBM Reconstruction error 0.256782
>> Epoch 7 finished	RBM Reconstruction error 0.229957
>> Epoch 8 finished	RBM Reconstruction error 0.204971
>> Epoch 9 finished	RBM Reconstruction error 0.212785
>> Epoch 10 finished	RBM Reconstruction error 0.183680

[END] Pre-training step

>> Epoch 0 finished	ANN training loss 2.161781
>> Epoch 1 finished	ANN training loss 2.162324
>> Epoch 2 finished	ANN training loss 2.231463
>> Epoch 3 finished	ANN training loss 2.037632
>> Epoch 4 finished	ANN training loss 2.005377
>> Epoch 5 finished	ANN training loss 1.926284
>> Epoch 6 finished	ANN training loss 2.027987
>> Epoch 7 finished	ANN training loss 1.953428
>> Epoch 8 finished	ANN training loss 1.832274
>> Epoch 9 finished	ANN training loss 1.863648
>> Epoch 10 finished	ANN training loss 1.835893
>> Epoch 11 finished	ANN training loss 1.673477

- >> Epoch 12 finished ANN training loss 1.691658
- >> Epoch 13 finished ANN training loss 1.692844
- >> Epoch 14 finished ANN training loss 1.663815
- >> Epoch 15 finished ANN training loss 1.615688
- >> Epoch 16 finished ANN training loss 1.618857
- >> Epoch 17 finished ANN training loss 1.618945
- >> Epoch 18 finished ANN training loss 1.594308
- >> Epoch 19 finished ANN training loss 1.616004
- >> Epoch 20 finished ANN training loss 1.566871
- >> Epoch 21 finished ANN training loss 1.586625
- >> Epoch 22 finished ANN training loss 1.526542
- >> Epoch 23 finished ANN training loss 1.529343
- >> Epoch 24 finished ANN training loss 1.497098
- >> Epoch 25 finished ANN training loss 1.551028
- >> Epoch 26 finished ANN training loss 1.448715
- >> Epoch 27 finished ANN training loss 1.457039
- >> Epoch 28 finished ANN training loss 1.531772
- >> Epoch 29 finished ANN training loss 1.426155
- >> Epoch 30 finished ANN training loss 1.395560
- >> Epoch 31 finished ANN training loss 1.373403
- >> Epoch 32 finished ANN training loss 1.393595
- >> Epoch 33 finished ANN training loss 1.356537
- >> Epoch 34 finished ANN training loss 1.338161
- >> Epoch 35 finished ANN training loss 1.357755
- >> Epoch 36 finished ANN training loss 1.334401
- >> Epoch 37 finished ANN training loss 1.326859
- >> Epoch 38 finished ANN training loss 1.294037
- >> Epoch 39 finished ANN training loss 1.310475
- >> Epoch 40 finished ANN training loss 1.285604
- >> Epoch 41 finished ANN training loss 1.251143
- >> Epoch 42 finished ANN training loss 1.238622
- >> Epoch 43 finished ANN training loss 1.244420

- >> Epoch 44 finished ANN training loss 1.227694
- >> Epoch 45 finished ANN training loss 1.234250
- >> Epoch 46 finished ANN training loss 1.192151
- >> Epoch 47 finished ANN training loss 1.282872
- >> Epoch 48 finished ANN training loss 1.202125
- >> Epoch 49 finished ANN training loss 1.195783
- >> Epoch 50 finished ANN training loss 1.173911
- >> Epoch 51 finished ANN training loss 1.180761
- >> Epoch 52 finished ANN training loss 1.126582
- >> Epoch 53 finished ANN training loss 1.107668
- >> Epoch 54 finished ANN training loss 1.080279
- >> Epoch 55 finished ANN training loss 1.054442
- >> Epoch 56 finished ANN training loss 1.064081
- >> Epoch 57 finished ANN training loss 1.106891
- >> Epoch 58 finished ANN training loss 1.017401
- >> Epoch 59 finished ANN training loss 1.091074
- >> Epoch 60 finished ANN training loss 1.101666
- >> Epoch 61 finished ANN training loss 0.970878
- >> Epoch 62 finished ANN training loss 0.990581
- >> Epoch 63 finished ANN training loss 0.924880
- >> Epoch 64 finished ANN training loss 0.933200
- >> Epoch 65 finished ANN training loss 0.888013
- >> Epoch 66 finished ANN training loss 0.914845

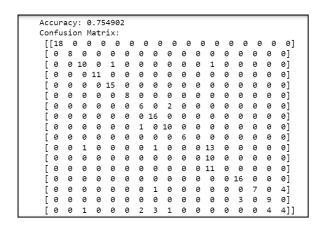
>> Epoch 67 finished

>> Epoch 68 finished ANN training loss 0.862002

ANN training loss 0.895561

- >> Epoch 69 finished ANN training loss 0.860283
- >> Epoch 70 finished ANN training loss 0.926866
- >> Epoch 71 finished ANN training loss 0.825624
- >> Epoch 72 finished ANN training loss 0.955979
- >> Epoch 73 finished ANN training loss 0.837455
- >> Epoch 74 finished ANN training loss 0.838362
- >> Epoch 75 finished ANN training loss 0.824579

>> Epoch 76 finished	ANN training loss 0.837506
>> Epoch 77 finished	ANN training loss 0.799540
>> Epoch 78 finished	ANN training loss 0.782582
>> Epoch 79 finished	ANN training loss 0.809945
>> Epoch 80 finished	ANN training loss 0.776058
>> Epoch 81 finished	ANN training loss 0.801965
>> Epoch 82 finished	ANN training loss 0.743689
>> Epoch 83 finished	ANN training loss 0.748219
>> Epoch 84 finished	ANN training loss 0.725424
>> Epoch 85 finished	ANN training loss 0.707048
>> Epoch 86 finished	ANN training loss 0.760290
>> Epoch 87 finished	ANN training loss 0.732380
>> Epoch 88 finished	ANN training loss 0.725342
>> Epoch 89 finished	ANN training loss 0.722347
>> Epoch 90 finished	ANN training loss 0.747846
>> Epoch 91 finished	ANN training loss 0.766685
>> Epoch 92 finished	ANN training loss 0.678031
>> Epoch 93 finished	ANN training loss 0.728679
>> Epoch 94 finished	ANN training loss 0.665033
>> Epoch 95 finished	ANN training loss 0.704792
>> Epoch 96 finished	ANN training loss 0.666352
>> Epoch 97 finished	ANN training loss 0.711128
>> Epoch 98 finished	ANN training loss 0.653001
>> Epoch 99 finished	ANN training loss 0.628350



Case 10:

>> Epoch 1 finished	RBM Reconstruction error 41.780271
>> Epoch 2 finished	RBM Reconstruction error 40.872977
>> Epoch 3 finished	RBM Reconstruction error 41.017986
>> Epoch 4 finished	RBM Reconstruction error 41.373823
>> Epoch 5 finished	RBM Reconstruction error 41.698571
>> Epoch 6 finished	RBM Reconstruction error 41.948013
>> Epoch 7 finished	RBM Reconstruction error 42.196067
>> Epoch 8 finished	RBM Reconstruction error 42.439778
>> Epoch 9 finished	RBM Reconstruction error 42.863548
>> Epoch 10 finished	RBM Reconstruction error 42.649242
>> Epoch 1 finished	RBM Reconstruction error 14.179907
>> Epoch 2 finished	RBM Reconstruction error 10.724631
>> Epoch 3 finished	RBM Reconstruction error 9.625305
>> Epoch 4 finished	RBM Reconstruction error 8.980324
>> Epoch 5 finished	RBM Reconstruction error 8.484160
>> Epoch 6 finished	RBM Reconstruction error 7.885944
>> Epoch 7 finished	RBM Reconstruction error 7.560121
>> Epoch 8 finished	RBM Reconstruction error 6.955149
>> Epoch 9 finished	RBM Reconstruction error 6.810843
>> Epoch 10 finished	RBM Reconstruction error 6.380751
>> Epoch 1 finished	RBM Reconstruction error 5.757024
>> Epoch 2 finished	RBM Reconstruction error 4.113161

>> Epoch 3 finished	RBM Reconstruction error 3.063645
>> Epoch 4 finished	RBM Reconstruction error 2.526045
>> Epoch 5 finished	RBM Reconstruction error 2.097313
>> Epoch 6 finished	RBM Reconstruction error 1.734041
>> Epoch 7 finished	RBM Reconstruction error 1.423209
>> Epoch 8 finished	RBM Reconstruction error 1.144207
>> Epoch 9 finished	RBM Reconstruction error 1.041854
>> Epoch 10 finished	RBM Reconstruction error 0.907573
>> Epoch 1 finished	RBM Reconstruction error 4.189174
>> Epoch 2 finished	RBM Reconstruction error 2.415335
>> Epoch 3 finished	RBM Reconstruction error 1.955680
>> Epoch 4 finished	RBM Reconstruction error 1.743029
>> Epoch 5 finished	RBM Reconstruction error 1.464093
>> Epoch 6 finished	RBM Reconstruction error 1.215696
>> Epoch 7 finished	RBM Reconstruction error 1.052109
>> Epoch 8 finished	RBM Reconstruction error 0.962401
>> Epoch 9 finished	RBM Reconstruction error 0.828475
>> Epoch 10 finished	RBM Reconstruction error 0.752980

[END] Pre-training step

>> Epoch 0 finished	ANN training loss 2.355754
>> Epoch 1 finished	ANN training loss 2.005809
>> Epoch 2 finished	ANN training loss 1.949715
>> Epoch 3 finished	ANN training loss 1.816342
>> Epoch 4 finished	ANN training loss 1.468088
>> Epoch 5 finished	ANN training loss 1.378142
>> Epoch 6 finished	ANN training loss 1.732937
>> Epoch 7 finished	ANN training loss 1.304914
>> Epoch 8 finished	ANN training loss 1.208001
>> Epoch 9 finished	ANN training loss 1.296047
>> Epoch 10 finished	ANN training loss 1.338111
>> Epoch 11 finished	ANN training loss 1.184829

- >> Epoch 12 finished ANN training loss 1.196126
- >> Epoch 13 finished ANN training loss 1.291828
- >> Epoch 14 finished ANN training loss 0.966330
- >> Epoch 15 finished ANN training loss 1.119303
- >> Epoch 16 finished ANN training loss 1.046144
- >> Epoch 17 finished ANN training loss 1.036193
- >> Epoch 18 finished ANN training loss 0.796963
- >> Epoch 19 finished ANN training loss 0.966526
- >> Epoch 20 finished ANN training loss 0.884428
- >> Epoch 21 finished ANN training loss 0.799144
- >> Epoch 22 finished ANN training loss 1.092183
- >> Epoch 23 finished ANN training loss 0.710476
- >> Epoch 24 finished ANN training loss 0.769813
- >> Epoch 25 finished ANN training loss 1.368138
- >> Epoch 26 finished ANN training loss 0.828462
- >> Epoch 27 finished ANN training loss 0.812755
- >> Epoch 28 finished ANN training loss 0.726576
- >> Epoch 29 finished ANN training loss 0.920830
- >> Epoch 30 finished ANN training loss 0.821545
- >> Epoch 31 finished ANN training loss 0.601399
- >> Epoch 32 finished ANN training loss 0.741971
- >> Epoch 33 finished ANN training loss 0.607226
- >> Epoch 34 finished ANN training loss 0.659265
- >> Epoch 35 finished ANN training loss 0.672722
- >> Epoch 36 finished ANN training loss 1.319239
- >> Epoch 37 finished ANN training loss 0.686871
- >> Epoch 38 finished ANN training loss 0.555630
- >> Epoch 39 finished ANN training loss 0.632449
- >> Epoch 40 finished ANN training loss 0.636977
- >> Epoch 41 finished ANN training loss 0.565965
- >> Epoch 42 finished ANN training loss 0.552924
- >> Epoch 43 finished ANN training loss 0.667722

>> Epoch 44 finished	ANN training loss 0.747259
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- >> Epoch 45 finished ANN training loss 0.532093
- >> Epoch 46 finished ANN training loss 0.589623
- >> Epoch 47 finished ANN training loss 0.713966
- >> Epoch 48 finished ANN training loss 0.498352
- >> Epoch 49 finished ANN training loss 0.561103
- >> Epoch 50 finished ANN training loss 0.566664
- >> Epoch 51 finished ANN training loss 0.807969
- >> Epoch 52 finished ANN training loss 0.521493
- >> Epoch 53 finished ANN training loss 0.504521
- >> Epoch 54 finished ANN training loss 0.541450
- >> Epoch 55 finished ANN training loss 0.443279
- >> Epoch 56 finished ANN training loss 0.606393
- >> Epoch 57 finished ANN training loss 0.653897
- >> Epoch 58 finished ANN training loss 0.602503
- >> Epoch 59 finished ANN training loss 0.574262
- >> Epoch 60 finished ANN training loss 0.541535
- >> Epoch 61 finished ANN training loss 0.525864
- >> Epoch 62 finished ANN training loss 0.444652
- >> Epoch 63 finished ANN training loss 0.430767
- >> Epoch 64 finished ANN training loss 0.644912
- >> Epoch 65 finished ANN training loss 0.474384

>> Epoch 66 finished

>> Epoch 67 finished ANN training loss 0.492417

ANN training loss 0.450690

- >> Epoch 68 finished ANN training loss 0.621144
- >> Epoch 69 finished ANN training loss 0.395571
- >> Epoch 70 finished ANN training loss 0.458535
- >> Epoch 71 finished ANN training loss 0.382509
- >> Epoch 72 finished ANN training loss 0.530205
- >> Epoch 73 finished ANN training loss 0.546226
- >> Epoch 74 finished ANN training loss 0.395878
- >> Epoch 75 finished ANN training loss 0.387898

>> Epoch 76 finished	ANN training loss 0.364658
>> Epoch 77 finished	ANN training loss 0.491180
>> Epoch 78 finished	ANN training loss 0.339123
>> Epoch 79 finished	ANN training loss 0.460687
>> Epoch 80 finished	ANN training loss 0.424633
>> Epoch 81 finished	ANN training loss 0.486973
>> Epoch 82 finished	ANN training loss 0.357474
>> Epoch 83 finished	ANN training loss 0.382459
>> Epoch 84 finished	ANN training loss 0.313187
>> Epoch 85 finished	ANN training loss 0.330671
>> Epoch 86 finished	ANN training loss 0.360356
>> Epoch 87 finished	ANN training loss 0.633519
>> Epoch 88 finished	ANN training loss 0.728780
>> Epoch 89 finished	ANN training loss 0.295521
>> Epoch 90 finished	ANN training loss 0.308737
>> Epoch 91 finished	ANN training loss 0.480726
>> Epoch 92 finished	ANN training loss 0.407640
>> Epoch 93 finished	ANN training loss 0.310695
>> Epoch 94 finished	ANN training loss 0.247827
>> Epoch 95 finished	ANN training loss 0.411460
>> Epoch 96 finished	ANN training loss 0.351295
>> Epoch 97 finished	ANN training loss 0.246896
>> Epoch 98 finished	ANN training loss 0.243083
>> Epoch 99 finished	ANN training loss 0.299315

Case 11:

-	-	Č	1
>>	Epoch	1 finished	RBM Reconstruction error 41.792308
>>	Epoch	2 finished	RBM Reconstruction error 40.697907
>>	Epoch	3 finished	RBM Reconstruction error 40.745301
>>	Epoch	4 finished	RBM Reconstruction error 40.877579
>>	Epoch	5 finished	RBM Reconstruction error 41.123398
>>	Epoch	6 finished	RBM Reconstruction error 41.327932
>>	Epoch	7 finished	RBM Reconstruction error 41.602009
>>	Epoch	8 finished	RBM Reconstruction error 41.763390
>>	Epoch	9 finished	RBM Reconstruction error 42.133638
>>	Epoch	10 finished	RBM Reconstruction error 41.999716
>>	Epoch	1 finished	RBM Reconstruction error 13.749287
>>	Epoch	2 finished	RBM Reconstruction error 10.896676
>>	Epoch	3 finished	RBM Reconstruction error 10.036325
>>	Epoch	4 finished	RBM Reconstruction error 9.319138
>>	Epoch	5 finished	RBM Reconstruction error 8.862726
>>	Epoch	6 finished	RBM Reconstruction error 8.321328
>>	Epoch	7 finished	RBM Reconstruction error 7.879983
>>	Epoch	8 finished	RBM Reconstruction error 7.386158
>>	Epoch	9 finished	RBM Reconstruction error 7.215208
>>	Epoch	10 finished	RBM Reconstruction error 6.832676
>>	Epoch	1 finished	RBM Reconstruction error 6.196856
>>	Epoch	2 finished	RBM Reconstruction error 4.699625

>> Epoch 3 finished	RBM Reconstruction error 3.261836
>> Epoch 4 finished	RBM Reconstruction error 2.646699
>> Epoch 5 finished	RBM Reconstruction error 2.082705
>> Epoch 6 finished	RBM Reconstruction error 1.679480
>> Epoch 7 finished	RBM Reconstruction error 1.360283
>> Epoch 8 finished	RBM Reconstruction error 1.247703
>> Epoch 9 finished	RBM Reconstruction error 1.059702
>> Epoch 10 finished	RBM Reconstruction error 0.925052
>> Epoch 1 finished	RBM Reconstruction error 3.949530
>> Epoch 2 finished	RBM Reconstruction error 2.221192
>> Epoch 3 finished	RBM Reconstruction error 1.842948
>> Epoch 4 finished	RBM Reconstruction error 1.674798
>> Epoch 5 finished	RBM Reconstruction error 1.392542
>> Epoch 6 finished	RBM Reconstruction error 1.244112
>> Epoch 7 finished	RBM Reconstruction error 1.069062
>> Epoch 8 finished	RBM Reconstruction error 0.922722
>> Epoch 9 finished	RBM Reconstruction error 0.873899
>> Epoch 10 finished	RBM Reconstruction error 0.741341

[END] Pre-training step

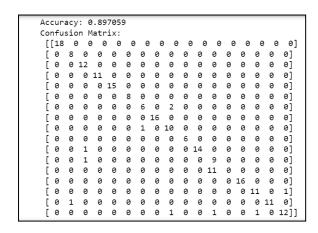
ANN training loss 2.303955
ANN training loss 1.943863
ANN training loss 2.000633
ANN training loss 1.531922
ANN training loss 1.559560
ANN training loss 1.439424
ANN training loss 1.793061
ANN training loss 1.361017
ANN training loss 1.248379
ANN training loss 1.297007
ANN training loss 1.270400
ANN training loss 1.072484

- >> Epoch 12 finished ANN training loss 1.065917
- >> Epoch 13 finished ANN training loss 1.206617
- >> Epoch 14 finished ANN training loss 0.970444
- >> Epoch 15 finished ANN training loss 1.100040
- >> Epoch 16 finished ANN training loss 0.990905
- >> Epoch 17 finished ANN training loss 0.926904
- >> Epoch 18 finished ANN training loss 0.989692
- >> Epoch 19 finished ANN training loss 0.954614
- >> Epoch 20 finished ANN training loss 0.921462
- >> Epoch 21 finished ANN training loss 0.916566
- >> Epoch 22 finished ANN training loss 0.789039
- >> Epoch 23 finished ANN training loss 0.955643
- >> Epoch 24 finished ANN training loss 0.683763
- >> Epoch 25 finished ANN training loss 1.111770
- >> Epoch 26 finished ANN training loss 0.810240
- >> Epoch 27 finished ANN training loss 0.802449
- >> Epoch 28 finished ANN training loss 0.831219
- >> Epoch 29 finished ANN training loss 0.703080
- >> Epoch 30 finished ANN training loss 0.751760
- >> Epoch 31 finished ANN training loss 0.987825
- >> Epoch 32 finished ANN training loss 0.728653
- >> Epoch 33 finished ANN training loss 0.605071
- >> Epoch 34 finished ANN training loss 0.819282
- >> Epoch 35 finished ANN training loss 0.716283
- >> Epoch 36 finished ANN training loss 0.839047
- >> Epoch 37 finished ANN training loss 0.659773
- >> Epoch 38 finished ANN training loss 0.572167
- >> Epoch 39 finished ANN training loss 0.534373
- >> Epoch 40 finished ANN training loss 0.544591
- >> Epoch 41 finished ANN training loss 0.573683
- >> Epoch 42 finished ANN training loss 0.520078
- >> Epoch 43 finished ANN training loss 0.724189

>> Epoch 44 finished AN	IN training loss 0.870827
-------------------------	---------------------------

- >> Epoch 45 finished ANN training loss 0.642290
- >> Epoch 46 finished ANN training loss 0.502335
- >> Epoch 47 finished ANN training loss 0.575561
- >> Epoch 48 finished ANN training loss 0.479302
- >> Epoch 49 finished ANN training loss 0.470576
- >> Epoch 50 finished ANN training loss 0.510612
- >> Epoch 51 finished ANN training loss 0.571177
- >> Epoch 52 finished ANN training loss 0.502410
- >> Epoch 53 finished ANN training loss 0.601242
- >> Epoch 54 finished ANN training loss 0.475851
- >> Epoch 55 finished ANN training loss 0.661030
- >> Epoch 56 finished ANN training loss 0.400963
- >> Epoch 57 finished ANN training loss 0.525064
- >> Epoch 58 finished ANN training loss 0.421402
- >> Epoch 59 finished ANN training loss 0.568796
- >> Epoch 60 finished ANN training loss 0.512244
- >> Epoch 61 finished ANN training loss 0.463851
- >> Epoch 62 finished ANN training loss 0.402834
- >> Epoch 63 finished ANN training loss 0.513126
- >> Epoch 64 finished ANN training loss 0.670645
- >> Epoch 65 finished ANN training loss 1.002682
- >> Epoch 66 finished ANN training loss 0.373528
- >> Epoch 67 finished ANN training loss 0.464029
- >> Epoch 68 finished ANN training loss 0.445517
- >> Epoch 69 finished ANN training loss 0.511435
- >> Epoch 70 finished ANN training loss 0.706556
- >> Epoch 71 finished ANN training loss 0.391917
- >> Epoch 72 finished ANN training loss 0.424611
- >> Epoch 73 finished ANN training loss 0.362159
- >> Epoch 74 finished ANN training loss 0.381115
- >> Epoch 75 finished ANN training loss 0.321061

>> Epoch 76 finished	ANN training loss 0.338619
>> Epoch 77 finished	ANN training loss 0.424649
>> Epoch 78 finished	ANN training loss 0.333186
>> Epoch 79 finished	ANN training loss 0.335362
>> Epoch 80 finished	ANN training loss 0.375474
>> Epoch 81 finished	ANN training loss 0.392937
>> Epoch 82 finished	ANN training loss 0.351414
>> Epoch 83 finished	ANN training loss 0.320896
>> Epoch 84 finished	ANN training loss 0.374062
>> Epoch 85 finished	ANN training loss 0.313674
>> Epoch 86 finished	ANN training loss 0.330555
>> Epoch 87 finished	ANN training loss 0.304709
>> Epoch 88 finished	ANN training loss 0.310597
>> Epoch 89 finished	ANN training loss 0.318290
>> Epoch 90 finished	ANN training loss 0.359511
>> Epoch 91 finished	ANN training loss 0.328263
>> Epoch 92 finished	ANN training loss 0.311251
>> Epoch 93 finished	ANN training loss 0.362394
>> Epoch 94 finished	ANN training loss 0.302975
>> Epoch 95 finished	ANN training loss 0.317551
>> Epoch 96 finished	ANN training loss 0.351703
>> Epoch 97 finished	ANN training loss 0.329673
>> Epoch 98 finished	ANN training loss 0.299616
>> Epoch 99 finished	ANN training loss 0.285132



Case 12:

>> Epoch 1 finished	RBM Reconstruction error 41.928273
>> Epoch 2 finished	RBM Reconstruction error 41.062688
>> Epoch 3 finished	RBM Reconstruction error 41.101849
>> Epoch 4 finished	RBM Reconstruction error 41.365208
>> Epoch 5 finished	RBM Reconstruction error 41.580365
>> Epoch 6 finished	RBM Reconstruction error 41.780221
>> Epoch 7 finished	RBM Reconstruction error 41.989691
>> Epoch 8 finished	RBM Reconstruction error 42.163667
>> Epoch 9 finished	RBM Reconstruction error 42.410279
>> Epoch 10 finished	RBM Reconstruction error 42.337359
>> Epoch 1 finished	RBM Reconstruction error 14.425925
>> Epoch 2 finished	RBM Reconstruction error 10.897882
>> Epoch 3 finished	RBM Reconstruction error 9.837923
>> Epoch 4 finished	RBM Reconstruction error 9.177402
>> Epoch 5 finished	RBM Reconstruction error 8.672123
>> Epoch 6 finished	RBM Reconstruction error 8.027297
>> Epoch 7 finished	RBM Reconstruction error 7.678715
>> Epoch 8 finished	RBM Reconstruction error 7.128404
>> Epoch 9 finished	RBM Reconstruction error 7.032723
>> Epoch 10 finished	RBM Reconstruction error 6.574803
>> Epoch 1 finished	RBM Reconstruction error 6.645475

>> Epoch 2 finished	RBM Reconstruction error 4.554062
>> Epoch 3 finished	RBM Reconstruction error 3.460196
>> Epoch 4 finished	RBM Reconstruction error 2.608315
>> Epoch 5 finished	RBM Reconstruction error 2.176655
>> Epoch 6 finished	RBM Reconstruction error 1.982105
>> Epoch 7 finished	RBM Reconstruction error 1.601931
>> Epoch 8 finished	RBM Reconstruction error 1.270529
>> Epoch 9 finished	RBM Reconstruction error 1.165354
>> Epoch 10 finished	RBM Reconstruction error 0.980679
>> Epoch 1 finished	RBM Reconstruction error 4.217915
>> Epoch 2 finished	RBM Reconstruction error 2.180746
>> Epoch 3 finished	RBM Reconstruction error 1.848991
>> Epoch 4 finished	RBM Reconstruction error 1.479886
>> Epoch 5 finished	RBM Reconstruction error 1.336478
>> Epoch 6 finished	RBM Reconstruction error 1.159839
>> Epoch 7 finished	RBM Reconstruction error 0.973260
>> Epoch 8 finished	RBM Reconstruction error 0.918541
>> Epoch 9 finished	RBM Reconstruction error 0.740847
>> Epoch 10 finished	RBM Reconstruction error 0.671451
[FND] Pre-training sten	

[END] Pre-training step

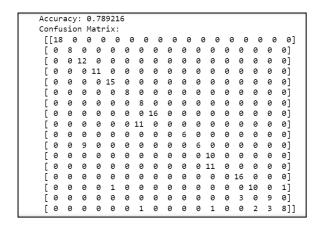
>> Epoch 0 finished	ANN training loss 2.299225
>> Epoch 1 finished	ANN training loss 2.094565
>> Epoch 2 finished	ANN training loss 2.055731
>> Epoch 3 finished	ANN training loss 2.129252
>> Epoch 4 finished	ANN training loss 1.651600
>> Epoch 5 finished	ANN training loss 1.491005
>> Epoch 6 finished	ANN training loss 1.797554
>> Epoch 7 finished	ANN training loss 1.563263
>> Epoch 8 finished	ANN training loss 1.392459
>> Epoch 9 finished	ANN training loss 1.443025
>> Epoch 10 finished	ANN training loss 1.463212

- >> Epoch 11 finished ANN training loss 1.301476
- >> Epoch 12 finished ANN training loss 1.207085
- >> Epoch 13 finished ANN training loss 1.442973
- >> Epoch 14 finished ANN training loss 1.239788
- >> Epoch 15 finished ANN training loss 1.191382
- >> Epoch 16 finished ANN training loss 1.172073
- >> Epoch 17 finished ANN training loss 1.092896
- >> Epoch 18 finished ANN training loss 1.051139
- >> Epoch 19 finished ANN training loss 1.121554
- >> Epoch 20 finished ANN training loss 1.149510
- >> Epoch 21 finished ANN training loss 1.085826
- >> Epoch 22 finished ANN training loss 0.965453
- >> Epoch 23 finished ANN training loss 0.944914
- >> Epoch 24 finished ANN training loss 0.970495
- >> Epoch 25 finished ANN training loss 0.902461
- >> Epoch 26 finished ANN training loss 0.856993
- >> Epoch 27 finished ANN training loss 0.955770
- >> Epoch 28 finished ANN training loss 1.027274
- >> Epoch 29 finished ANN training loss 0.857773
- >> Epoch 30 finished ANN training loss 0.852348
- >> Epoch 31 finished ANN training loss 0.889136
- >> Epoch 32 finished ANN training loss 0.859254
- >> Epoch 33 finished ANN training loss 0.787451
- >> Epoch 34 finished ANN training loss 0.823568
- >> Epoch 35 finished ANN training loss 0.880880
- >> Epoch 36 finished ANN training loss 0.783373
- >> Epoch 37 finished ANN training loss 0.878841
- >> Epoch 38 finished ANN training loss 0.762410
- >> Epoch 39 finished ANN training loss 0.925957
- >> Epoch 40 finished ANN training loss 0.836307
- >> Epoch 41 finished ANN training loss 0.707972
- >> Epoch 42 finished ANN training loss 0.768072

>> Epoch 43 finished	ANN training loss 0.729777

- >> Epoch 45 finished ANN training loss 0.731328
- >> Epoch 46 finished ANN training loss 0.702854
- >> Epoch 47 finished ANN training loss 0.747056
- >> Epoch 48 finished ANN training loss 0.682895
- >> Epoch 49 finished ANN training loss 0.818870
- >> Epoch 50 finished ANN training loss 0.665726
- >> Epoch 51 finished ANN training loss 0.716408
- >> Epoch 52 finished ANN training loss 0.625661
- >> Epoch 53 finished ANN training loss 0.659843
- >> Epoch 54 finished ANN training loss 0.663048
- >> Epoch 55 finished ANN training loss 0.585904
- >> Epoch 56 finished ANN training loss 0.637196
- >> Epoch 57 finished ANN training loss 0.667233
- >> Epoch 58 finished ANN training loss 0.584763
- >> Epoch 59 finished ANN training loss 0.752489
- >> Epoch 60 finished ANN training loss 0.657930
- >> Epoch 61 finished ANN training loss 0.616332
- >> Epoch 62 finished ANN training loss 0.654013
- >> Epoch 63 finished ANN training loss 0.616264
- >> Epoch 64 finished ANN training loss 0.688015
- >> Epoch 65 finished ANN training loss 0.565256
- >> Epoch 66 finished ANN training loss 0.541758
- >> Epoch 67 finished ANN training loss 0.529538
- >> Epoch 68 finished ANN training loss 0.518958
- >> Epoch 69 finished ANN training loss 0.608546
- >> Epoch 70 finished ANN training loss 0.854384
- >> Epoch 71 finished ANN training loss 0.542631
- >> Epoch 72 finished ANN training loss 0.629748
- >> Epoch 73 finished ANN training loss 0.549381
- >> Epoch 74 finished ANN training loss 0.554504

>> Epoch 75 finished	ANN training loss 0.643439
>> Epoch 76 finished	ANN training loss 0.545011
>> Epoch 77 finished	ANN training loss 0.509667
>> Epoch 78 finished	ANN training loss 0.625169
>> Epoch 79 finished	ANN training loss 0.559927
>> Epoch 80 finished	ANN training loss 0.629732
>> Epoch 81 finished	ANN training loss 0.535591
>> Epoch 82 finished	ANN training loss 0.512410
>> Epoch 83 finished	ANN training loss 0.521489
>> Epoch 84 finished	ANN training loss 0.478122
>> Epoch 85 finished	ANN training loss 0.445746
>> Epoch 86 finished	ANN training loss 0.481189
>> Epoch 87 finished	ANN training loss 0.467566
>> Epoch 88 finished	ANN training loss 0.516752
>> Epoch 89 finished	ANN training loss 0.448772
>> Epoch 90 finished	ANN training loss 0.580619
>> Epoch 91 finished	ANN training loss 0.514983
>> Epoch 92 finished	ANN training loss 0.485222
>> Epoch 93 finished	ANN training loss 0.448501
>> Epoch 94 finished	ANN training loss 0.578910
>> Epoch 95 finished	ANN training loss 0.402924
>> Epoch 96 finished	ANN training loss 0.465109
>> Epoch 97 finished	ANN training loss 0.477062
>> Epoch 98 finished	ANN training loss 0.492936
>> Epoch 99 finished	ANN training loss 0.441270



Case 13:

>> Epoch 1 finished	RBM Reconstruction error 43.996420
>> Epoch 2 finished	RBM Reconstruction error 42.872291
>> Epoch 3 finished	RBM Reconstruction error 42.578621
>> Epoch 4 finished	RBM Reconstruction error 42.563326
>> Epoch 5 finished	RBM Reconstruction error 42.745222
>> Epoch 6 finished	RBM Reconstruction error 42.918616
>> Epoch 7 finished	RBM Reconstruction error 43.089202
>> Epoch 8 finished	RBM Reconstruction error 43.303874
>> Epoch 9 finished	RBM Reconstruction error 43.557284
>> Epoch 10 finished	RBM Reconstruction error 43.464297
>> Epoch 1 finished	RBM Reconstruction error 9.802981
>> Epoch 2 finished	RBM Reconstruction error 6.206737
>> Epoch 3 finished	RBM Reconstruction error 4.927590
>> Epoch 4 finished	RBM Reconstruction error 4.479526
>> Epoch 5 finished	RBM Reconstruction error 4.279783
>> Epoch 6 finished	RBM Reconstruction error 4.175113
>> Epoch 7 finished	RBM Reconstruction error 3.988726
>> Epoch 8 finished	RBM Reconstruction error 3.847451
>> Epoch 9 finished	RBM Reconstruction error 3.718203
>> Epoch 10 finished	RBM Reconstruction error 3.574889
>> Epoch 1 finished	RBM Reconstruction error 3.725893
>> Epoch 2 finished	RBM Reconstruction error 1.769416

>> Epoch 3 finished	RBM Reconstruction error 0.991826
>> Epoch 4 finished	RBM Reconstruction error 0.793316
>> Epoch 5 finished	RBM Reconstruction error 0.805093
>> Epoch 6 finished	RBM Reconstruction error 0.699267
>> Epoch 7 finished	RBM Reconstruction error 0.685677
>> Epoch 8 finished	RBM Reconstruction error 0.630998
>> Epoch 9 finished	RBM Reconstruction error 0.595160
>> Epoch 10 finished	RBM Reconstruction error 0.570329
>> Epoch 1 finished	RBM Reconstruction error 2.900568
>> Epoch 2 finished	RBM Reconstruction error 1.137552
>> Epoch 3 finished	RBM Reconstruction error 0.497561
>> Epoch 4 finished	RBM Reconstruction error 0.333553
>> Epoch 5 finished	RBM Reconstruction error 0.268621
>> Epoch 6 finished	RBM Reconstruction error 0.253052
>> Epoch 7 finished	RBM Reconstruction error 0.232243
>> Epoch 8 finished	RBM Reconstruction error 0.229902
>> Epoch 9 finished	RBM Reconstruction error 0.215756
>> Epoch 10 finished	RBM Reconstruction error 0.214141
>> Epoch 1 finished	RBM Reconstruction error 2.834458
>> Epoch 2 finished	RBM Reconstruction error 0.821829
>> Epoch 3 finished	RBM Reconstruction error 0.406224
>> Epoch 4 finished	RBM Reconstruction error 0.247149
>> Epoch 5 finished	RBM Reconstruction error 0.241908
>> Epoch 6 finished	RBM Reconstruction error 0.182438
>> Epoch 7 finished	RBM Reconstruction error 0.159523
>> Epoch 8 finished	RBM Reconstruction error 0.169832
>> Epoch 9 finished	RBM Reconstruction error 0.142858
>> Epoch 10 finished	RBM Reconstruction error 0.134254
>> Epoch 1 finished	RBM Reconstruction error 2.700932
>> Epoch 2 finished	RBM Reconstruction error 0.854578
>> Epoch 3 finished	RBM Reconstruction error 0.315472
>> Epoch 4 finished	RBM Reconstruction error 0.200918

>> Epoch 5 finished RB	M Reconstruction error 0.143773
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>> Epoch 10 finished RBM Reconstruction error 0.091172

[END] Pre-training step

[START] Fine tuning step:

>> Epoch 0 finished	ANN training loss 2.248294
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>> Epoch 23 finished ANN training loss 1.431135

- >> Epoch 24 finished ANN training loss 1.594574
- >> Epoch 25 finished ANN training loss 1.630303
- >> Epoch 26 finished ANN training loss 1.389851
- >> Epoch 27 finished ANN training loss 1.451792
- >> Epoch 28 finished ANN training loss 1.393417
- >> Epoch 29 finished ANN training loss 1.366777
- >> Epoch 30 finished ANN training loss 1.457555
- >> Epoch 31 finished ANN training loss 1.434905
- >> Epoch 32 finished ANN training loss 1.512418
- >> Epoch 33 finished ANN training loss 1.322252
- >> Epoch 34 finished ANN training loss 1.457784
- >> Epoch 35 finished ANN training loss 1.296240
- >> Epoch 36 finished ANN training loss 1.328582
- >> Epoch 37 finished ANN training loss 1.401475
- >> Epoch 38 finished ANN training loss 1.319411
- >> Epoch 39 finished ANN training loss 1.239302
- >> Epoch 40 finished ANN training loss 1.657244
- >> Epoch 41 finished ANN training loss 1.344938
- >> Epoch 42 finished ANN training loss 1.279983
- >> Epoch 43 finished ANN training loss 1.812382
- >> Epoch 44 finished ANN training loss 1.229697
- >> Epoch 45 finished ANN training loss 1.176625
- >> Epoch 46 finished ANN training loss 1.331004
- >> Epoch 47 finished ANN training loss 1.203512
- >> Epoch 48 finished ANN training loss 1.235551
- >> Epoch 49 finished ANN training loss 1.219181
- >> Epoch 50 finished ANN training loss 1.153909
- >> Epoch 51 finished ANN training loss 1.154788
- >> Epoch 52 finished ANN training loss 1.156716
- >> Epoch 53 finished ANN training loss 1.271531
- >> Epoch 54 finished ANN training loss 1.212051
- >> Epoch 55 finished ANN training loss 1.099769

- >> Epoch 56 finished ANN training loss 1.109878
- >> Epoch 57 finished ANN training loss 1.106008
- >> Epoch 58 finished ANN training loss 1.098744
- >> Epoch 59 finished ANN training loss 1.164668
- >> Epoch 60 finished ANN training loss 1.085275
- >> Epoch 61 finished ANN training loss 1.223306
- >> Epoch 62 finished ANN training loss 1.091146
- >> Epoch 63 finished ANN training loss 1.078929
- >> Epoch 64 finished ANN training loss 1.088696
- >> Epoch 65 finished ANN training loss 1.043728
- >> Epoch 66 finished ANN training loss 1.021376
- >> Epoch 67 finished ANN training loss 1.420218
- >> Epoch 68 finished ANN training loss 1.066512
- >> Epoch 69 finished ANN training loss 1.283128
- >> Epoch 70 finished ANN training loss 0.975086
- >> Epoch 71 finished ANN training loss 1.003711
- >> Epoch 72 finished ANN training loss 0.966193
- >> Epoch 73 finished ANN training loss 1.474280
- >> Epoch 74 finished ANN training loss 1.036891
- >> Epoch 75 finished ANN training loss 1.074170
- >> Epoch 76 finished ANN training loss 0.990111
- >> Epoch 77 finished ANN training loss 0.979931
- >> Epoch 78 finished ANN training loss 1.038214
- >> Epoch 79 finished ANN training loss 0.936215
- >> Epoch 80 finished ANN training loss 0.914436
- >> Epoch 81 finished ANN training loss 0.909277
- >> Epoch 82 finished ANN training loss 0.932924
- >> Epoch 83 finished ANN training loss 0.943851
- >> Epoch 84 finished ANN training loss 1.005406
- >> Epoch 85 finished ANN training loss 0.935902
- >> Epoch 86 finished ANN training loss 0.969766
- >> Epoch 87 finished ANN training loss 0.866467

>> Epoch 88 finished	ANN training loss 0.918294
>> Epoch 89 finished	ANN training loss 0.896709
>> Epoch 90 finished	ANN training loss 0.962043
>> Epoch 91 finished	ANN training loss 0.983880
>> Epoch 92 finished	ANN training loss 0.892380
>> Epoch 93 finished	ANN training loss 0.881656
>> Epoch 94 finished	ANN training loss 0.856115
>> Epoch 95 finished	ANN training loss 0.951613
>> Epoch 96 finished	ANN training loss 0.970311
>> Epoch 97 finished	ANN training loss 0.841004
>> Epoch 98 finished	ANN training loss 0.827221
>> Epoch 99 finished	ANN training loss 0.894456
[END] Fine tuning step	

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]	0	0	0	0	0	8	0	0	0	0	0	0	0	0	0	0	6
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[0	0	0	0	0	0	0	0	0	0	0	0	0	16	0	0	6
[0	3	0	0	0	0	0	0	0	0	5	0	0	0	4	0	6
[0	0	0	0	0	0	0	0	0	0	0	0	0	3	0	7	2
[0	2	0	0	0	0	0	0	1	0	2	1	0	0	0	2	7

Case 14:

>> Epoch 1 finished	RBM Reconstruction error 43.609389
>> Epoch 2 finished	RBM Reconstruction error 42.852918
>> Epoch 3 finished	RBM Reconstruction error 42.818830
>> Epoch 4 finished	RBM Reconstruction error 42.948735
>> Epoch 5 finished	RBM Reconstruction error 43.098784
>> Epoch 6 finished	RBM Reconstruction error 43.210891
>> Epoch 7 finished	RBM Reconstruction error 43.344079

>> Epoch 8 finished	RBM Reconstruction error 43.456196
>> Epoch 9 finished	RBM Reconstruction error 43.823303
>> Epoch 10 finished	RBM Reconstruction error 43.657357
>> Epoch 1 finished	RBM Reconstruction error 10.006001
>> Epoch 2 finished	RBM Reconstruction error 6.766770
>> Epoch 3 finished	RBM Reconstruction error 5.402900
>> Epoch 4 finished	RBM Reconstruction error 4.843395
>> Epoch 5 finished	RBM Reconstruction error 4.601433
>> Epoch 6 finished	RBM Reconstruction error 4.367920
>> Epoch 7 finished	RBM Reconstruction error 4.198896
>> Epoch 8 finished	RBM Reconstruction error 4.049663
>> Epoch 9 finished	RBM Reconstruction error 3.886069
>> Epoch 10 finished	RBM Reconstruction error 3.772066
>> Epoch 1 finished	RBM Reconstruction error 3.487273
>> Epoch 2 finished	RBM Reconstruction error 1.489269
>> Epoch 3 finished	RBM Reconstruction error 0.965596
>> Epoch 4 finished	RBM Reconstruction error 0.761787
>> Epoch 5 finished	RBM Reconstruction error 0.755818
>> Epoch 6 finished	RBM Reconstruction error 0.702527
>> Epoch 7 finished	RBM Reconstruction error 0.681450
>> Epoch 8 finished	RBM Reconstruction error 0.645232
>> Epoch 9 finished	RBM Reconstruction error 0.620193
>> Epoch 10 finished	RBM Reconstruction error 0.605889
>> Epoch 1 finished	RBM Reconstruction error 2.788723
>> Epoch 2 finished	RBM Reconstruction error 1.098556
>> Epoch 3 finished	RBM Reconstruction error 0.495889
>> Epoch 4 finished	RBM Reconstruction error 0.348530
>> Epoch 5 finished	RBM Reconstruction error 0.273225
>> Epoch 6 finished	RBM Reconstruction error 0.247271
>> Epoch 7 finished	RBM Reconstruction error 0.262507
>> Epoch 8 finished	RBM Reconstruction error 0.231410
>> Epoch 9 finished	RBM Reconstruction error 0.226878

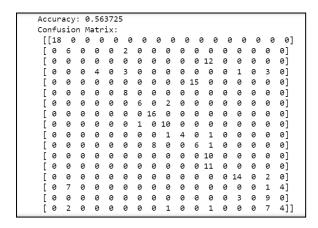
>> Epoch 10 finished	RBM Reconstruction error 0.212868
>> Epoch 1 finished	RBM Reconstruction error 3.120102
>> Epoch 2 finished	RBM Reconstruction error 1.230270
>> Epoch 3 finished	RBM Reconstruction error 0.503536
>> Epoch 4 finished	RBM Reconstruction error 0.382528
>> Epoch 5 finished	RBM Reconstruction error 0.297488
>> Epoch 6 finished	RBM Reconstruction error 0.287956
>> Epoch 7 finished	RBM Reconstruction error 0.242032
>> Epoch 8 finished	RBM Reconstruction error 0.209689
>> Epoch 9 finished	RBM Reconstruction error 0.192010
>> Epoch 10 finished	RBM Reconstruction error 0.166104
>> Epoch 1 finished	RBM Reconstruction error 2.667175
>> Epoch 2 finished	RBM Reconstruction error 1.017357
>> Epoch 3 finished	RBM Reconstruction error 0.393063
>> Epoch 4 finished	RBM Reconstruction error 0.240559
>> Epoch 5 finished	RBM Reconstruction error 0.184447
>> Epoch 6 finished	RBM Reconstruction error 0.168247
>> Epoch 7 finished	RBM Reconstruction error 0.141622
>> Epoch 8 finished	RBM Reconstruction error 0.118455
>> Epoch 9 finished	RBM Reconstruction error 0.120852
>> Epoch 10 finished	RBM Reconstruction error 0.107916
[END] Pre-training step	

>> Epoch 0 finished	ANN training loss 2.286046
>> Epoch 1 finished	ANN training loss 2.084279
>> Epoch 2 finished	ANN training loss 2.157613
>> Epoch 3 finished	ANN training loss 2.015646
>> Epoch 4 finished	ANN training loss 2.010698
>> Epoch 5 finished	ANN training loss 1.912616
>> Epoch 6 finished	ANN training loss 1.890512
>> Epoch 7 finished	ANN training loss 1.973773
>> Epoch 8 finished	ANN training loss 1.834459

- >> Epoch 9 finished ANN training loss 1.795783
- >> Epoch 10 finished ANN training loss 1.751677
- >> Epoch 11 finished ANN training loss 1.680129
- >> Epoch 12 finished ANN training loss 1.654729
- >> Epoch 13 finished ANN training loss 1.668286
- >> Epoch 14 finished ANN training loss 1.675589
- >> Epoch 15 finished ANN training loss 1.626807
- >> Epoch 16 finished ANN training loss 1.581257
- >> Epoch 17 finished ANN training loss 1.908555
- >> Epoch 18 finished ANN training loss 1.614903
- >> Epoch 19 finished ANN training loss 1.657209
- >> Epoch 20 finished ANN training loss 1.570308
- >> Epoch 21 finished ANN training loss 1.585315
- >> Epoch 22 finished ANN training loss 1.629024
- >> Epoch 23 finished ANN training loss 1.539175
- >> Epoch 24 finished ANN training loss 1.502236
- >> Epoch 25 finished ANN training loss 1.625121
- >> Epoch 26 finished ANN training loss 1.487644
- >> Epoch 27 finished ANN training loss 1.466213
- >> Epoch 28 finished ANN training loss 1.513992
- >> Epoch 29 finished ANN training loss 1.480664
- >> Epoch 30 finished ANN training loss 1.477660
- >> Epoch 31 finished ANN training loss 1.549744
- >> Epoch 32 finished ANN training loss 1.477133
- >> Epoch 33 finished ANN training loss 1.463639
- >> Epoch 34 finished ANN training loss 1.516390
- >> Epoch 35 finished ANN training loss 1.440364
- >> Epoch 36 finished ANN training loss 1.453493
- >> Epoch 37 finished ANN training loss 1.365522
- >> Epoch 38 finished ANN training loss 1.392066
- >> Epoch 39 finished ANN training loss 1.317458
- >> Epoch 40 finished ANN training loss 1.436495

- >> Epoch 41 finished ANN training loss 1.336759
- >> Epoch 42 finished ANN training loss 1.448644
- >> Epoch 43 finished ANN training loss 1.331624
- >> Epoch 44 finished ANN training loss 1.371016
- >> Epoch 45 finished ANN training loss 1.263318
- >> Epoch 46 finished ANN training loss 1.311395
- >> Epoch 47 finished ANN training loss 1.299449
- >> Epoch 48 finished ANN training loss 1.294016
- >> Epoch 49 finished ANN training loss 1.300636
- >> Epoch 50 finished ANN training loss 1.297210
- >> Epoch 51 finished ANN training loss 1.262365
- >> Epoch 52 finished ANN training loss 1.318717
- >> Epoch 53 finished ANN training loss 1.319631
- >> Epoch 54 finished ANN training loss 1.300252
- >> Epoch 55 finished ANN training loss 1.481310
- >> Epoch 56 finished ANN training loss 1.233731
- >> Epoch 57 finished ANN training loss 1.255174
- >> Epoch 58 finished ANN training loss 1.205952
- >> Epoch 59 finished ANN training loss 1.225810
- >> Epoch 60 finished ANN training loss 1.235041
- >> Epoch 61 finished ANN training loss 1.273451
- >> Epoch 62 finished ANN training loss 1.195080
- >> Epoch 63 finished ANN training loss 1.179816
- >> Epoch 64 finished ANN training loss 1.168583
- >> Epoch 65 finished ANN training loss 1.278595
- >> Epoch 66 finished ANN training loss 1.193485
- >> Epoch 67 finished ANN training loss 1.285697
- >> Epoch 68 finished ANN training loss 1.193301
- >> Epoch 69 finished ANN training loss 1.252163
- >> Epoch 70 finished ANN training loss 1.155391
- >> Epoch 71 finished ANN training loss 1.159760
- >> Epoch 72 finished ANN training loss 1.127603

>> Epoch 73 finished	ANN training loss 1.147693
>> Epoch 74 finished	ANN training loss 1.227342
>> Epoch 75 finished	ANN training loss 1.210005
>> Epoch 76 finished	ANN training loss 1.137557
>> Epoch 77 finished	ANN training loss 1.136181
>> Epoch 78 finished	ANN training loss 1.228256
>> Epoch 79 finished	ANN training loss 1.145252
>> Epoch 80 finished	ANN training loss 1.105463
>> Epoch 81 finished	ANN training loss 1.142132
>> Epoch 82 finished	ANN training loss 1.055376
>> Epoch 83 finished	ANN training loss 1.085218
>> Epoch 84 finished	ANN training loss 1.096447
>> Epoch 85 finished	ANN training loss 1.071598
>> Epoch 86 finished	ANN training loss 1.101200
>> Epoch 87 finished	ANN training loss 1.055888
>> Epoch 88 finished	ANN training loss 1.063743
>> Epoch 89 finished	ANN training loss 1.150140
>> Epoch 90 finished	ANN training loss 1.135835
>> Epoch 91 finished	ANN training loss 1.130028
>> Epoch 92 finished	ANN training loss 1.025904
>> Epoch 93 finished	ANN training loss 1.014976
>> Epoch 94 finished	ANN training loss 0.994553
>> Epoch 95 finished	ANN training loss 1.050592
>> Epoch 96 finished	ANN training loss 0.994025
>> Epoch 97 finished	ANN training loss 1.067141
>> Epoch 98 finished	ANN training loss 1.025416
>> Epoch 99 finished	ANN training loss 1.061215
[END] Fine tuning step	



Case 15:

>> Epoch 1 finished	RBM Reconstruction error 43.886352
>> Epoch 2 finished	RBM Reconstruction error 43.046046
>> Epoch 3 finished	RBM Reconstruction error 42.805803
>> Epoch 4 finished	RBM Reconstruction error 42.905842
>> Epoch 5 finished	RBM Reconstruction error 43.020510
>> Epoch 6 finished	RBM Reconstruction error 43.075038
>> Epoch 7 finished	RBM Reconstruction error 43.240156
>> Epoch 8 finished	RBM Reconstruction error 43.335682
>> Epoch 9 finished	RBM Reconstruction error 43.542940
>> Epoch 10 finished	RBM Reconstruction error 43.468387
>> Epoch 1 finished	RBM Reconstruction error 11.087082
>> Epoch 2 finished	RBM Reconstruction error 7.220903
>> Epoch 3 finished	RBM Reconstruction error 5.597691
>> Epoch 4 finished	RBM Reconstruction error 4.919397
>> Epoch 5 finished	RBM Reconstruction error 4.708293
>> Epoch 6 finished	RBM Reconstruction error 4.420802
>> Epoch 7 finished	RBM Reconstruction error 4.239472
>> Epoch 8 finished	RBM Reconstruction error 4.014754
>> Epoch 9 finished	RBM Reconstruction error 3.897891
>> Epoch 10 finished	RBM Reconstruction error 3.698362
>> Epoch 1 finished	RBM Reconstruction error 3.981275
>> Epoch 2 finished	RBM Reconstruction error 1.527598

>> Epoch 3 finished	RBM Reconstruction error 0.938647
>> Epoch 4 finished	RBM Reconstruction error 0.803073
>> Epoch 5 finished	RBM Reconstruction error 0.766450
>> Epoch 6 finished	RBM Reconstruction error 0.764909
>> Epoch 7 finished	RBM Reconstruction error 0.676009
>> Epoch 8 finished	RBM Reconstruction error 0.634078
>> Epoch 9 finished	RBM Reconstruction error 0.605052
>> Epoch 10 finished	RBM Reconstruction error 0.580835
>> Epoch 1 finished	RBM Reconstruction error 3.501202
>> Epoch 2 finished	RBM Reconstruction error 1.214631
>> Epoch 3 finished	RBM Reconstruction error 0.583485
>> Epoch 4 finished	RBM Reconstruction error 0.412185
>> Epoch 5 finished	RBM Reconstruction error 0.345668
>> Epoch 6 finished	RBM Reconstruction error 0.322766
>> Epoch 7 finished	RBM Reconstruction error 0.293792
>> Epoch 8 finished	RBM Reconstruction error 0.280508
>> Epoch 9 finished	RBM Reconstruction error 0.277080
>> Epoch 10 finished	RBM Reconstruction error 0.272022
>> Epoch 1 finished	RBM Reconstruction error 2.793688
>> Epoch 2 finished	RBM Reconstruction error 0.819042
>> Epoch 3 finished	RBM Reconstruction error 0.394557
>> Epoch 4 finished	RBM Reconstruction error 0.271191
>> Epoch 5 finished	RBM Reconstruction error 0.231584
>> Epoch 6 finished	RBM Reconstruction error 0.207396
>> Epoch 7 finished	RBM Reconstruction error 0.212356
>> Epoch 8 finished	RBM Reconstruction error 0.178923
>> Epoch 9 finished	RBM Reconstruction error 0.188060
>> Epoch 10 finished	RBM Reconstruction error 0.161279
>> Epoch 1 finished	RBM Reconstruction error 2.781979
>> Epoch 2 finished	RBM Reconstruction error 0.931701
>> Epoch 3 finished	RBM Reconstruction error 0.361872
>> Epoch 4 finished	RBM Reconstruction error 0.241981

>> Epoch 5 finished	RBM Reconstruction error 0.219268
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>> Epoch 10 finished RBM Reconstruction error 0.121886

[END] Pre-training step

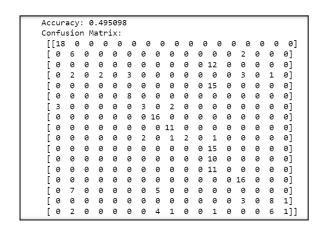
loss	2.251842
9	g loss

- >> Epoch 19 finished ANN training loss 1.576131
- >> Epoch 20 finished ANN training loss 1.572081
- >> Epoch 21 finished ANN training loss 1.550265
- >> Epoch 22 finished ANN training loss 1.630539
- >> Epoch 23 finished ANN training loss 1.517986

- >> Epoch 24 finished ANN training loss 1.480456
- >> Epoch 25 finished ANN training loss 1.493541
- >> Epoch 26 finished ANN training loss 1.467597
- >> Epoch 27 finished ANN training loss 1.464341
- >> Epoch 28 finished ANN training loss 1.458393
- >> Epoch 29 finished ANN training loss 1.442560
- >> Epoch 30 finished ANN training loss 1.494634
- >> Epoch 31 finished ANN training loss 1.434536
- >> Epoch 32 finished ANN training loss 1.433157
- >> Epoch 33 finished ANN training loss 1.609621
- >> Epoch 34 finished ANN training loss 1.469514
- >> Epoch 35 finished ANN training loss 1.380108
- >> Epoch 36 finished ANN training loss 1.633837
- >> Epoch 37 finished ANN training loss 1.370612
- >> Epoch 38 finished ANN training loss 1.365816
- >> Epoch 39 finished ANN training loss 1.344026
- >> Epoch 40 finished ANN training loss 1.375883
- >> Epoch 41 finished ANN training loss 1.355031
- >> Epoch 42 finished ANN training loss 1.492809
- >> Epoch 43 finished ANN training loss 1.329049
- >> Epoch 44 finished ANN training loss 1.351697
- >> Epoch 45 finished ANN training loss 1.316906
- >> Epoch 46 finished ANN training loss 1.345269
- >> Epoch 47 finished ANN training loss 1.360189
- >> Epoch 48 finished ANN training loss 1.317899
- >> Epoch 49 finished ANN training loss 1.409497
- >> Epoch 50 finished ANN training loss 1.341151
- >> Epoch 51 finished ANN training loss 1.331371
- >> Epoch 52 finished ANN training loss 1.301675
- >> Epoch 53 finished ANN training loss 1.343287
- >> Epoch 54 finished ANN training loss 1.298209
- >> Epoch 55 finished ANN training loss 1.341437

- >> Epoch 56 finished ANN training loss 1.310105
- >> Epoch 57 finished ANN training loss 1.331626
- >> Epoch 58 finished ANN training loss 1.313174
- >> Epoch 59 finished ANN training loss 1.327456
- >> Epoch 60 finished ANN training loss 1.289324
- >> Epoch 61 finished ANN training loss 1.295377
- >> Epoch 62 finished ANN training loss 1.290536
- >> Epoch 63 finished ANN training loss 1.249636
- >> Epoch 64 finished ANN training loss 1.244046
- >> Epoch 65 finished ANN training loss 1.317548
- >> Epoch 66 finished ANN training loss 1.246024
- >> Epoch 67 finished ANN training loss 1.270822
- >> Epoch 68 finished ANN training loss 1.260268
- >> Epoch 69 finished ANN training loss 1.262222
- >> Epoch 70 finished ANN training loss 1.244859
- >> Epoch 71 finished ANN training loss 1.258458
- >> Epoch 72 finished ANN training loss 1.256479
- >> Epoch 73 finished ANN training loss 1.226517
- >> Epoch 74 finished ANN training loss 1.266147
- >> Epoch 75 finished ANN training loss 1.279248
- >> Epoch 76 finished ANN training loss 1.301465
- >> Epoch 77 finished ANN training loss 1.341313
- >> Epoch 78 finished ANN training loss 1.226209
- >> Epoch 79 finished ANN training loss 1.211712
- >> Epoch 80 finished ANN training loss 1.219206
- >> Epoch 81 finished ANN training loss 1.246858
- >> Epoch 82 finished ANN training loss 1.205429
- >> Epoch 83 finished ANN training loss 1.220340
- >> Epoch 84 finished ANN training loss 1.229771
- >> Epoch 85 finished ANN training loss 1.239241
- >> Epoch 86 finished ANN training loss 1.228428
- >> Epoch 87 finished ANN training loss 1.298715

>> Epoch 88 finished	ANN training loss 1.234928
>> Epoch 89 finished	ANN training loss 1.221309
>> Epoch 90 finished	ANN training loss 1.199273
>> Epoch 91 finished	ANN training loss 1.204976
>> Epoch 92 finished	ANN training loss 1.188726
>> Epoch 93 finished	ANN training loss 1.275268
>> Epoch 94 finished	ANN training loss 1.214669
>> Epoch 95 finished	ANN training loss 1.225007
>> Epoch 96 finished	ANN training loss 1.195816
>> Epoch 97 finished	ANN training loss 1.239328
>> Epoch 98 finished	ANN training loss 1.193926
>> Epoch 99 finished	ANN training loss 1.222696
[END] Fine tuning step	



Case 16:

[START] Pre-training step:

>> Epoch 1 finished	RBM Reconstruction error 41.881581
>> Epoch 2 finished	RBM Reconstruction error 41.257958
>> Epoch 3 finished	RBM Reconstruction error 41.323386
>> Epoch 4 finished	RBM Reconstruction error 41.571337
>> Epoch 5 finished	RBM Reconstruction error 41.877749
>> Epoch 6 finished	RBM Reconstruction error 42.132660
>> Epoch 7 finished	RBM Reconstruction error 42.289599
>> Epoch 8 finished	RBM Reconstruction error 42.371411

>> Epoch 9 finished	RBM Reconstruction error 42.527196
>> Epoch 10 finished	RBM Reconstruction error 42.431652
>> Epoch 1 finished	RBM Reconstruction error 14.721472
>> Epoch 2 finished	RBM Reconstruction error 11.279943
>> Epoch 3 finished	RBM Reconstruction error 10.174998
>> Epoch 4 finished	RBM Reconstruction error 9.224299
>> Epoch 5 finished	RBM Reconstruction error 8.430844
>> Epoch 6 finished	RBM Reconstruction error 7.848715
>> Epoch 7 finished	RBM Reconstruction error 7.580234
>> Epoch 8 finished	RBM Reconstruction error 7.079891
>> Epoch 9 finished	RBM Reconstruction error 6.873018
>> Epoch 10 finished	RBM Reconstruction error 6.571279
>> Epoch 1 finished	RBM Reconstruction error 6.626338
>> Epoch 2 finished	RBM Reconstruction error 4.557794
>> Epoch 3 finished	RBM Reconstruction error 3.496681
>> Epoch 4 finished	RBM Reconstruction error 2.939007
>> Epoch 5 finished	RBM Reconstruction error 2.444906
>> Epoch 6 finished	RBM Reconstruction error 1.816902
>> Epoch 7 finished	RBM Reconstruction error 1.533974
>> Epoch 8 finished	RBM Reconstruction error 1.238105
>> Epoch 9 finished	RBM Reconstruction error 1.138116
>> Epoch 10 finished	RBM Reconstruction error 1.063996
>> Epoch 1 finished	RBM Reconstruction error 4.074636
>> Epoch 2 finished	RBM Reconstruction error 2.436015
>> Epoch 3 finished	RBM Reconstruction error 2.064846
>> Epoch 4 finished	RBM Reconstruction error 1.778906
>> Epoch 5 finished	RBM Reconstruction error 1.386577
>> Epoch 6 finished	RBM Reconstruction error 1.185548
>> Epoch 7 finished	RBM Reconstruction error 0.935022
>> Epoch 8 finished	RBM Reconstruction error 0.821247
>> Epoch 9 finished	RBM Reconstruction error 0.633960
>> Epoch 10 finished	RBM Reconstruction error 0.610639

>> Epoch 1 finished	RBM Reconstruction error 4.347401
>> Epoch 2 finished	RBM Reconstruction error 2.228135
>> Epoch 3 finished	RBM Reconstruction error 1.711066
>> Epoch 4 finished	RBM Reconstruction error 1.380954
>> Epoch 5 finished	RBM Reconstruction error 1.191205
>> Epoch 6 finished	RBM Reconstruction error 1.002162
>> Epoch 7 finished	RBM Reconstruction error 0.756898
>> Epoch 8 finished	RBM Reconstruction error 0.572574
>> Epoch 9 finished	RBM Reconstruction error 0.480395
>> Epoch 10 finished	RBM Reconstruction error 0.358597
>> Epoch 1 finished	RBM Reconstruction error 3.600170
>> Epoch 2 finished	RBM Reconstruction error 1.776046
>> Epoch 3 finished	RBM Reconstruction error 1.509164
>> Epoch 4 finished	RBM Reconstruction error 1.334722
>> Epoch 5 finished	RBM Reconstruction error 1.169990
>> Epoch 6 finished	RBM Reconstruction error 0.935144
>> Epoch 7 finished	RBM Reconstruction error 0.775750
>> Epoch 8 finished	RBM Reconstruction error 0.712667
>> Epoch 9 finished	RBM Reconstruction error 0.503954
>> Epoch 10 finished	RBM Reconstruction error 0.437597

[END] Pre-training step

[START] Fine tuning step:

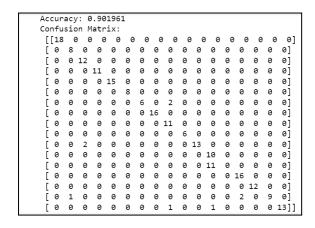
-	-	0 1	
>>	Epoch 0 finishe	d	ANN training loss 2.441036
>>	Epoch 1 finishe	d	ANN training loss 1.792657
>>	Epoch 2 finishe	d	ANN training loss 1.928725
>>	Epoch 3 finishe	d	ANN training loss 1.599205
>>	Epoch 4 finishe	d	ANN training loss 1.821750
>>	Epoch 5 finishe	d	ANN training loss 1.534096
>>	Epoch 6 finishe	d	ANN training loss 1.407854
>>	Epoch 7 finishe	d	ANN training loss 1.712880
>>	Epoch 8 finishe	d	ANN training loss 1.307491
>>	Epoch 9 finishe	d	ANN training loss 1.267714

- >> Epoch 10 finished ANN training loss 1.383784
- >> Epoch 11 finished ANN training loss 1.331258
- >> Epoch 12 finished ANN training loss 1.227321
- >> Epoch 13 finished ANN training loss 1.125127
- >> Epoch 14 finished ANN training loss 1.212347
- >> Epoch 15 finished ANN training loss 1.048481
- >> Epoch 16 finished ANN training loss 1.222216
- >> Epoch 17 finished ANN training loss 1.127892
- >> Epoch 18 finished ANN training loss 1.123433
- >> Epoch 19 finished ANN training loss 1.256029
- >> Epoch 20 finished ANN training loss 0.910242
- >> Epoch 21 finished ANN training loss 0.917687
- >> Epoch 22 finished ANN training loss 0.915742
- >> Epoch 23 finished ANN training loss 0.994854
- >> Epoch 24 finished ANN training loss 1.141139
- >> Epoch 25 finished ANN training loss 1.167651
- >> Epoch 26 finished ANN training loss 0.777356
- >> Epoch 27 finished ANN training loss 0.789659
- >> Epoch 28 finished ANN training loss 1.395347
- >> Epoch 29 finished ANN training loss 0.911472
- >> Epoch 30 finished ANN training loss 0.820440
- >> Epoch 31 finished ANN training loss 0.774344
- >> Epoch 32 finished ANN training loss 1.065869
- >> Epoch 33 finished ANN training loss 0.790273
- >> Epoch 34 finished ANN training loss 0.645244
- >> Epoch 35 finished ANN training loss 0.853284
- >> Epoch 36 finished ANN training loss 0.643309
- >> Epoch 37 finished ANN training loss 0.657322
- >> Epoch 38 finished ANN training loss 0.608522
- >> Epoch 39 finished ANN training loss 0.771529
- >> Epoch 40 finished ANN training loss 0.633999
- >> Epoch 41 finished ANN training loss 0.637887

>> Epoch 42 finished ANN tr	raining loss 0.621865
-----------------------------	-----------------------

- >> Epoch 43 finished ANN training loss 0.574590
- >> Epoch 44 finished ANN training loss 0.576875
- >> Epoch 45 finished ANN training loss 0.799400
- >> Epoch 46 finished ANN training loss 0.647421
- >> Epoch 47 finished ANN training loss 0.547465
- >> Epoch 48 finished ANN training loss 0.509092
- >> Epoch 49 finished ANN training loss 0.594612
- >> Epoch 50 finished ANN training loss 0.574606
- >> Epoch 51 finished ANN training loss 0.569939
- >> Epoch 52 finished ANN training loss 0.476568
- >> Epoch 53 finished ANN training loss 0.539329
- >> Epoch 54 finished ANN training loss 0.499898
- >> Epoch 55 finished ANN training loss 0.544789
- >> Epoch 56 finished ANN training loss 0.647368
- >> Epoch 57 finished ANN training loss 0.969598
- >> Epoch 58 finished ANN training loss 0.475547
- >> Epoch 59 finished ANN training loss 0.454910
- >> Epoch 60 finished ANN training loss 0.437104
- >> Epoch 61 finished ANN training loss 0.424194
- >> Epoch 62 finished ANN training loss 0.490134
- >> Epoch 63 finished ANN training loss 0.397937
- >> Epoch 64 finished ANN training loss 0.394233
- >> Epoch 65 finished ANN training loss 0.392327
- >> Epoch 66 finished ANN training loss 0.393688
- >> Epoch 67 finished ANN training loss 0.492650
- >> Epoch 68 finished ANN training loss 0.519179
- >> Epoch 69 finished ANN training loss 0.419632
- >> Epoch 70 finished ANN training loss 0.352470
- >> Epoch 71 finished ANN training loss 0.350642
- >> Epoch 72 finished ANN training loss 0.356442
- >> Epoch 73 finished ANN training loss 0.337871

>> Epoch 74 finished	ANN training loss 0.550652
>> Epoch 75 finished	ANN training loss 0.502454
>> Epoch 76 finished	ANN training loss 0.367960
>> Epoch 77 finished	ANN training loss 0.311995
>> Epoch 78 finished	ANN training loss 0.391720
>> Epoch 79 finished	ANN training loss 0.340036
>> Epoch 80 finished	ANN training loss 0.363924
>> Epoch 81 finished	ANN training loss 0.345140
>> Epoch 82 finished	ANN training loss 0.525064
>> Epoch 83 finished	ANN training loss 0.409130
>> Epoch 84 finished	ANN training loss 0.274633
>> Epoch 85 finished	ANN training loss 0.299966
>> Epoch 86 finished	ANN training loss 0.304363
>> Epoch 87 finished	ANN training loss 0.264791
>> Epoch 88 finished	ANN training loss 0.644932
>> Epoch 89 finished	ANN training loss 0.233666
>> Epoch 90 finished	ANN training loss 0.232599
>> Epoch 91 finished	ANN training loss 0.279451
>> Epoch 92 finished	ANN training loss 0.240310
>> Epoch 93 finished	ANN training loss 0.241849
>> Epoch 94 finished	ANN training loss 0.280076
>> Epoch 95 finished	ANN training loss 0.206140
>> Epoch 96 finished	ANN training loss 0.284892
>> Epoch 97 finished	ANN training loss 0.316268
>> Epoch 98 finished	ANN training loss 0.262847
>> Epoch 99 finished	ANN training loss 0.204403
[END] Fine tuning step	



Case 17:

[START] Pre-training step:

>> Epoch 1 finished	RBM Reconstruction error 41.792575
>> Epoch 2 finished	RBM Reconstruction error 40.771735
>> Epoch 3 finished	RBM Reconstruction error 40.879275
>> Epoch 4 finished	RBM Reconstruction error 41.065100
>> Epoch 5 finished	RBM Reconstruction error 41.151779
>> Epoch 6 finished	RBM Reconstruction error 41.341957
>> Epoch 7 finished	RBM Reconstruction error 41.493800
>> Epoch 8 finished	RBM Reconstruction error 41.654394
>> Epoch 9 finished	RBM Reconstruction error 41.835950
>> Epoch 10 finished	RBM Reconstruction error 41.851513
>> Epoch 1 finished	RBM Reconstruction error 14.581966
>> Epoch 2 finished	RBM Reconstruction error 11.285711
>> Epoch 3 finished	RBM Reconstruction error 10.320870
>> Epoch 4 finished	RBM Reconstruction error 9.670917
>> Epoch 5 finished	RBM Reconstruction error 9.090603
>> Epoch 6 finished	RBM Reconstruction error 8.423300
>> Epoch 7 finished	RBM Reconstruction error 8.043951
>> Epoch 8 finished	RBM Reconstruction error 7.458773
>> Epoch 9 finished	RBM Reconstruction error 7.287279
>> Epoch 10 finished	RBM Reconstruction error 6.902758
>> Epoch 1 finished	RBM Reconstruction error 6.242695
>> Epoch 2 finished	RBM Reconstruction error 4.327269

>> Epoch 3 finished	RBM Reconstruction error 3.471997
>> Epoch 4 finished	RBM Reconstruction error 2.576168
>> Epoch 5 finished	RBM Reconstruction error 2.113276
>> Epoch 6 finished	RBM Reconstruction error 1.711740
>> Epoch 7 finished	RBM Reconstruction error 1.451078
>> Epoch 8 finished	RBM Reconstruction error 1.233816
>> Epoch 9 finished	RBM Reconstruction error 1.042880
>> Epoch 10 finished	RBM Reconstruction error 0.993318
>> Epoch 1 finished	RBM Reconstruction error 4.023333
>> Epoch 2 finished	RBM Reconstruction error 2.284643
>> Epoch 3 finished	RBM Reconstruction error 1.906831
>> Epoch 4 finished	RBM Reconstruction error 1.619697
>> Epoch 5 finished	RBM Reconstruction error 1.306064
>> Epoch 6 finished	RBM Reconstruction error 1.113005
>> Epoch 7 finished	RBM Reconstruction error 1.012125
>> Epoch 8 finished	RBM Reconstruction error 0.831073
>> Epoch 9 finished	RBM Reconstruction error 0.770455
>> Epoch 10 finished	RBM Reconstruction error 0.684564
>> Epoch 1 finished	RBM Reconstruction error 4.326503
>> Epoch 2 finished	RBM Reconstruction error 2.405709
>> Epoch 3 finished	RBM Reconstruction error 1.889086
>> Epoch 4 finished	RBM Reconstruction error 1.370159
>> Epoch 5 finished	RBM Reconstruction error 1.126078
>> Epoch 6 finished	RBM Reconstruction error 0.913198
>> Epoch 7 finished	RBM Reconstruction error 0.778626
>> Epoch 8 finished	RBM Reconstruction error 0.592356
>> Epoch 9 finished	RBM Reconstruction error 0.489031
>> Epoch 10 finished	RBM Reconstruction error 0.438090
>> Epoch 1 finished	RBM Reconstruction error 3.242160
>> Epoch 2 finished	RBM Reconstruction error 1.408722
>> Epoch 3 finished	RBM Reconstruction error 1.224882
>> Epoch 4 finished	RBM Reconstruction error 0.925179

>> Epoch 5 finished RE	M Reconstruction error 0.891026
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- >> Epoch 7 finished RBM Reconstruction error 0.677794
- >> Epoch 8 finished RBM Reconstruction error 0.593075
- >> Epoch 9 finished RBM Reconstruction error 0.470849
- >> Epoch 10 finished RBM Reconstruction error 0.411954

[END] Pre-training step

[START] Fine tuning step:

- >> Epoch 0 finished ANN training loss 2.356841
- >> Epoch 1 finished ANN training loss 1.889281
- >> Epoch 2 finished ANN training loss 1.999467
- >> Epoch 3 finished ANN training loss 1.702392
- >> Epoch 4 finished ANN training loss 1.827894
- >> Epoch 5 finished ANN training loss 1.534763
- >> Epoch 6 finished ANN training loss 1.445119
- >> Epoch 7 finished ANN training loss 1.779731
- >> Epoch 8 finished ANN training loss 1.435609
- >> Epoch 9 finished ANN training loss 1.403093
- >> Epoch 10 finished ANN training loss 1.479311
- >> Epoch 11 finished ANN training loss 1.348117
- >> Epoch 12 finished ANN training loss 1.240289
- >> Epoch 13 finished ANN training loss 1.413820
- >> Epoch 14 finished ANN training loss 1.278798
- >> Epoch 15 finished ANN training loss 1.160457
- >> Epoch 16 finished ANN training loss 1.304313
- >> Epoch 17 finished ANN training loss 1.144585
- >> Epoch 18 finished ANN training loss 1.564302
- >> Epoch 19 finished ANN training loss 1.122434
- >> Epoch 20 finished ANN training loss 1.042869
- >> Epoch 21 finished ANN training loss 1.183307
- >> Epoch 22 finished ANN training loss 1.011542
- >> Epoch 23 finished ANN training loss 1.055059

>> Epoch 24 finished ANN	training loss 0.980324
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- >> Epoch 25 finished ANN training loss 0.842002
- >> Epoch 26 finished ANN training loss 0.889686
- >> Epoch 27 finished ANN training loss 0.988484
- >> Epoch 28 finished ANN training loss 0.934812
- >> Epoch 29 finished ANN training loss 0.891558
- >> Epoch 30 finished ANN training loss 0.910291
- >> Epoch 31 finished ANN training loss 0.785123
- >> Epoch 32 finished ANN training loss 0.937582
- >> Epoch 33 finished ANN training loss 0.728772
- >> Epoch 34 finished ANN training loss 0.936973
- >> Epoch 35 finished ANN training loss 0.762139
- >> Epoch 36 finished ANN training loss 1.126121
- >> Epoch 37 finished ANN training loss 0.665719
- >> Epoch 38 finished ANN training loss 0.764321
- >> Epoch 39 finished ANN training loss 0.691168
- >> Epoch 40 finished ANN training loss 0.750967
- >> Epoch 41 finished ANN training loss 0.808434
- >> Epoch 42 finished ANN training loss 0.637628
- >> Epoch 43 finished ANN training loss 0.754839
- >> Epoch 44 finished ANN training loss 0.626977
- >> Epoch 45 finished ANN training loss 0.561007
- >> Epoch 46 finished ANN training loss 0.571904
- >> Epoch 47 finished ANN training loss 0.574662
- >> Epoch 48 finished ANN training loss 0.675314
- >> Epoch 49 finished ANN training loss 0.540991
- >> Epoch 50 finished ANN training loss 0.591618
- >> Epoch 51 finished ANN training loss 0.667402
- >> Epoch 52 finished ANN training loss 0.539540
- >> Epoch 53 finished ANN training loss 0.634333
- >> Epoch 54 finished ANN training loss 0.537280
- >> Epoch 55 finished ANN training loss 0.536804

- >> Epoch 58 finished ANN training loss 0.477288
- >> Epoch 59 finished ANN training loss 0.541670
- >> Epoch 60 finished ANN training loss 0.413264
- >> Epoch 61 finished ANN training loss 0.478297
- >> Epoch 62 finished ANN training loss 0.581394
- >> Epoch 63 finished ANN training loss 0.444393
- >> Epoch 64 finished ANN training loss 0.449002
- >> Epoch 65 finished ANN training loss 0.554134
- >> Epoch 66 finished ANN training loss 0.442626
- >> Epoch 67 finished ANN training loss 0.521845
- >> Epoch 68 finished ANN training loss 0.569024
- >> Epoch 69 finished ANN training loss 0.399609
- >> Epoch 70 finished ANN training loss 0.587007
- >> Epoch 71 finished ANN training loss 0.453110
- >> Epoch 72 finished ANN training loss 0.401042
- >> Epoch 73 finished
- >> Epoch 74 finished ANN training loss 0.414249

ANN training loss 0.480804

ANN training loss 0.371734

ANN training loss 0.498519

ANN training loss 0.400970

>> Epoch 75 finished

>> Epoch 77 finished

- >> Epoch 76 finished

- >> Epoch 78 finished ANN training loss 0.369111
- >> Epoch 79 finished ANN training loss 0.349084
- >> Epoch 80 finished ANN training loss 0.396247
- >> Epoch 81 finished ANN training loss 0.417670
- >> Epoch 82 finished ANN training loss 0.709744
- ANN training loss 0.351304 >> Epoch 83 finished
- >> Epoch 84 finished ANN training loss 0.470447
- >> Epoch 85 finished ANN training loss 0.330241
- >> Epoch 86 finished ANN training loss 0.369712
- >> Epoch 87 finished ANN training loss 0.404151

>> Epoch 88 finished	ANN training loss 0.366452
>> Epoch 89 finished	ANN training loss 0.327789
>> Epoch 90 finished	ANN training loss 0.299704
>> Epoch 91 finished	ANN training loss 0.388818
>> Epoch 92 finished	ANN training loss 0.315336
>> Epoch 93 finished	ANN training loss 0.521408
>> Epoch 94 finished	ANN training loss 0.468917
>> Epoch 95 finished	ANN training loss 0.313262
>> Epoch 96 finished	ANN training loss 0.447776
>> Epoch 97 finished	ANN training loss 0.357050
>> Epoch 98 finished	ANN training loss 0.296343
>> Epoch 99 finished	ANN training loss 0.333221
[END] Fine tuning step	

Ac	cur	acy	/: E	3.8	5784	3											
Co	nfu:	sic	n N	lati	rix:												
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[0	0	1	0	0	0	0	0	1	0	0	0	0	0	0	1	12]]

Case 18:

[START] Pre-training step:

>> Epoch 1 finished	RBM Reconstruction error 41.372756
>> Epoch 2 finished	RBM Reconstruction error 40.476467
>> Epoch 3 finished	RBM Reconstruction error 40.702153
>> Epoch 4 finished	RBM Reconstruction error 41.071466
>> Epoch 5 finished	RBM Reconstruction error 41.296782
>> Epoch 6 finished	RBM Reconstruction error 41.597052
>> Epoch 7 finished	RBM Reconstruction error 41.837017

>> Epoch 8 finished	RBM Reconstruction error 42.108080
>> Epoch 9 finished	RBM Reconstruction error 42.290347
>> Epoch 10 finished	RBM Reconstruction error 42.388742
>> Epoch 1 finished	RBM Reconstruction error 14.298517
>> Epoch 2 finished	RBM Reconstruction error 11.045434
>> Epoch 3 finished	RBM Reconstruction error 9.978397
>> Epoch 4 finished	RBM Reconstruction error 9.239408
>> Epoch 5 finished	RBM Reconstruction error 8.596379
>> Epoch 6 finished	RBM Reconstruction error 7.899508
>> Epoch 7 finished	RBM Reconstruction error 7.500735
>> Epoch 8 finished	RBM Reconstruction error 6.909717
>> Epoch 9 finished	RBM Reconstruction error 6.811239
>> Epoch 10 finished	RBM Reconstruction error 6.425774
>> Epoch 1 finished	RBM Reconstruction error 6.294147
>> Epoch 2 finished	RBM Reconstruction error 3.999786
>> Epoch 3 finished	RBM Reconstruction error 3.082825
>> Epoch 4 finished	RBM Reconstruction error 2.855599
>> Epoch 5 finished	RBM Reconstruction error 2.173448
>> Epoch 6 finished	RBM Reconstruction error 1.733197
>> Epoch 7 finished	RBM Reconstruction error 1.338001
>> Epoch 8 finished	RBM Reconstruction error 1.153471
>> Epoch 9 finished	RBM Reconstruction error 0.995697
>> Epoch 10 finished	RBM Reconstruction error 0.872445
>> Epoch 1 finished	RBM Reconstruction error 4.564561
>> Epoch 2 finished	RBM Reconstruction error 2.602133
>> Epoch 3 finished	RBM Reconstruction error 2.189684
>> Epoch 4 finished	RBM Reconstruction error 1.826993
>> Epoch 5 finished	RBM Reconstruction error 1.576019
>> Epoch 6 finished	RBM Reconstruction error 1.354075
>> Epoch 7 finished	RBM Reconstruction error 1.128998
>> Epoch 8 finished	RBM Reconstruction error 0.916909
>> Epoch 9 finished	RBM Reconstruction error 0.799442

RBM Reconstruction error 0.650493
RBM Reconstruction error 3.881838
RBM Reconstruction error 2.328367
RBM Reconstruction error 1.997225
RBM Reconstruction error 1.697622
RBM Reconstruction error 1.325599
RBM Reconstruction error 1.037129
RBM Reconstruction error 0.861931
RBM Reconstruction error 0.751085
RBM Reconstruction error 0.636184
RBM Reconstruction error 0.554327
RBM Reconstruction error 3.110528
RBM Reconstruction error 1.473014
RBM Reconstruction error 1.174214
RBM Reconstruction error 0.989583
RBM Reconstruction error 0.941347
RBM Reconstruction error 0.749563
RBM Reconstruction error 0.567083
RBM Reconstruction error 0.489073
RBM Reconstruction error 0.457342
RBM Reconstruction error 0.397526
ep:

>> Epoch 0 finished	ANN training loss 2.533912
>> Epoch 1 finished	ANN training loss 1.912662
>> Epoch 2 finished	ANN training loss 2.138717
>> Epoch 3 finished	ANN training loss 1.712893
>> Epoch 4 finished	ANN training loss 1.925005
>> Epoch 5 finished	ANN training loss 1.608649
>> Epoch 6 finished	ANN training loss 1.566926
>> Epoch 7 finished	ANN training loss 1.772526
>> Epoch 8 finished	ANN training loss 1.534930

- >> Epoch 9 finished ANN training loss 1.583749
- >> Epoch 10 finished ANN training loss 1.472319
- >> Epoch 11 finished ANN training loss 1.528329
- >> Epoch 12 finished ANN training loss 1.366386
- >> Epoch 13 finished ANN training loss 1.360620
- >> Epoch 14 finished ANN training loss 1.489842
- >> Epoch 15 finished ANN training loss 1.419135
- >> Epoch 16 finished ANN training loss 1.276041
- >> Epoch 17 finished ANN training loss 1.331768
- >> Epoch 18 finished ANN training loss 1.273733
- >> Epoch 19 finished ANN training loss 1.292297
- >> Epoch 20 finished ANN training loss 1.261607
- >> Epoch 21 finished ANN training loss 1.203192
- >> Epoch 22 finished ANN training loss 1.264314
- >> Epoch 23 finished ANN training loss 1.126404
- >> Epoch 24 finished ANN training loss 1.070546
- >> Epoch 25 finished ANN training loss 1.009856
- >> Epoch 26 finished ANN training loss 1.154775
- >> Epoch 27 finished ANN training loss 1.033245
- >> Epoch 28 finished ANN training loss 1.028367
- >> Epoch 29 finished ANN training loss 1.117733

ANN training loss 1.101441

ANN training loss 1.043937

>> Epoch 30 finished

>> Epoch 32 finished

- >> Epoch 31 finished ANN training loss 0.978223
- >> Epoch 33 finished ANN training loss 0.928538
- >> Epoch 34 finished ANN training loss 0.919770
- >> Epoch 35 finished ANN training loss 0.917674
- >> Epoch 36 finished ANN training loss 0.898565
- >> Epoch 37 finished ANN training loss 0.925438
- >> Epoch 38 finished ANN training loss 0.904800
- >> Epoch 39 finished ANN training loss 0.904152
- >> Epoch 40 finished ANN training loss 0.818319

- >> Epoch 42 finished ANN training loss 0.835696
- >> Epoch 43 finished ANN training loss 0.792625
- >> Epoch 44 finished ANN training loss 0.844598
- >> Epoch 45 finished ANN training loss 0.833484
- >> Epoch 46 finished ANN training loss 0.889358
- >> Epoch 47 finished ANN training loss 0.768787
- >> Epoch 48 finished ANN training loss 0.791587
- >> Epoch 49 finished ANN training loss 0.765142
- >> Epoch 50 finished ANN training loss 0.845533
- >> Epoch 51 finished ANN training loss 0.788608
- >> Epoch 52 finished ANN training loss 0.728863
- >> Epoch 53 finished ANN training loss 0.947965
- >> Epoch 54 finished ANN training loss 0.817560
- >> Epoch 55 finished ANN training loss 0.887742
- >> Epoch 56 finished ANN training loss 0.787665
- >> Epoch 57 finished ANN training loss 0.929358
- >> Epoch 58 finished

ANN training loss 0.829298

- >> Epoch 59 finished ANN training loss 0.786177
- >> Epoch 60 finished ANN training loss 0.722860
- >> Epoch 61 finished ANN training loss 0.798352
- >> Epoch 62 finished ANN training loss 0.775610
- >> Epoch 63 finished ANN training loss 0.687135
- >> Epoch 64 finished ANN training loss 0.740055
- >> Epoch 65 finished ANN training loss 0.640857
- >> Epoch 66 finished ANN training loss 0.623001
- >> Epoch 67 finished ANN training loss 0.669290
- ANN training loss 0.640343 >> Epoch 68 finished
- >> Epoch 69 finished ANN training loss 0.657516
- >> Epoch 70 finished ANN training loss 0.752479
- >> Epoch 71 finished ANN training loss 0.640654
- >> Epoch 72 finished ANN training loss 0.590965

>> Epoch 73 finished	ANN training loss 0.652863
>> Epoch 74 finished	ANN training loss 0.683715
>> Epoch 75 finished	ANN training loss 0.626392
>> Epoch 76 finished	ANN training loss 0.625947
>> Epoch 77 finished	ANN training loss 0.628208
>> Epoch 78 finished	ANN training loss 0.580186
>> Epoch 79 finished	ANN training loss 0.620502
>> Epoch 80 finished	ANN training loss 0.785973
>> Epoch 81 finished	ANN training loss 0.571416
>> Epoch 82 finished	ANN training loss 0.705353
>> Epoch 83 finished	ANN training loss 0.579118
>> Epoch 84 finished	ANN training loss 0.630101
>> Epoch 85 finished	ANN training loss 0.548272
>> Epoch 86 finished	ANN training loss 0.559437
>> Epoch 87 finished	ANN training loss 0.541931
>> Epoch 88 finished	ANN training loss 0.663547
>> Epoch 89 finished	ANN training loss 0.570517
>> Epoch 90 finished	ANN training loss 0.529368
>> Epoch 91 finished	ANN training loss 0.550965
>> Epoch 92 finished	ANN training loss 0.508707
>> Epoch 93 finished	ANN training loss 0.563552
>> Epoch 94 finished	ANN training loss 0.550841
>> Epoch 95 finished	ANN training loss 0.530294
>> Epoch 96 finished	ANN training loss 0.535948
>> Epoch 97 finished	ANN training loss 0.535212
>> Epoch 98 finished	ANN training loss 0.518781
>> Epoch 99 finished	ANN training loss 0.551749
[END] Fine tuning step	

INFERENCES FOR DEEP BELIEF NETWORK

The following inferences can be made from the above obtained results:

- Number of layers in the network is inversely proportional to the accuracy.
- Dropout in the network is inversely proportional to the accuracy. In most DBN models, accuracy increases when dropout value is decreased.
- Number of RBMs in the layers in the network is directly proportional to the accuracy.

DISCUSSION:

The CNN is found to be more accurate when it comes to classifying audio files according to their instruments used than LSTM and DBN.

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