```
import tensorflow as tf
tf.test.gpu device name()
    Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.moun
!ln -sf /opt/bin/nvidia-smi /usr/bin/nvidia-smi
!pip install gputil
!pip install psutil
!pip install humanize
import psutil
import humanize
import os
import GPUtil as GPU
GPUs = GPU.getGPUs()
# XXX: only one GPU on Colab and isn't guaranteed
gpu = GPUs[0]
def printm():
process = psutil.Process(os.getpid())
print("Gen RAM Free: " + humanize.naturalsize( psutil.virtual_memory().available ), " | Proc
print("GPU RAM Free: {0:.0f}MB | Used: {1:.0f}MB | Util {2:3.0f}% | Total {3:.0f}MB".format(
printm()
from google.colab import drive
drive.mount('/content/drive')
from future import print function
import numpy as np
import glob
np.random.seed(1337)
from tensorflow.python.keras.models import Model
from tensorflow.python.keras.layers import Input, LSTM
from tensorflow.python.keras.utils import np_utils
from tensorflow.python.keras.callbacks import EarlyStopping
import tensorflow as tf
import matplotlib.pyplot as plt
%matplotlib inline
import datetime
now = datetime.datetime.now
batch size = 10
# length of data sequences
n \text{ timesteps} = 240
# dimension of data sequences
```

 $n \dim = 117$

→ 1. Loading Data

```
def reading_files():
   all files corrected = glob.glob("/content/drive/My Drive/DeepLearningProject/Segmented Mo
    all files incorrected = glob.glob("/content/drive/My Drive/DeepLearningProject/Segmented
   return all files corrected, all files incorrected
all files corrected , all files incorrected =reading files()
all files corrected = sorted(all files corrected)
all files incorrected = sorted(all files incorrected)
import csv
import numpy as np
def load data(corrected file, incorrected file):
   f = open( corrected_file )
   csv f = csv.reader(f)
   X Corr = list(csv f)
   # Convert the input sequences into numpy arrays
   train_input1 = np.asarray(X_Corr)
   n \dim = 117
   data correct = np.zeros((90,240,n dim))
   for i in range(len(train input1)//n dim):
          data correct[i,:,:] = np.transpose(train input1[n dim*i:n dim*(i+1),:])
   f = open( incorrected file )
   csv_f = csv.reader(f)
   X Incor = list(csv f)
   # Convert the input sequences into numpy arrays
   train input2 = np.asarray(X Incor)
   n \dim = 117
   data incorrect = np.zeros((90,240,n dim))
   for i in range(len(train input2)//n dim):
          data_incorrect[i,:,:] = np.transpose(train_input2[n_dim*i:n_dim*(i+1),:])
   return data_correct, data_incorrect
exercises names = [ "deep squat",
                   "hurdle step",
                   "inline lunge",
                   "side lunge",
                   "sit to stand",
```

2. Adding Frames

```
def adding frames(X correct, X incorrect, exercise num ):
   # Add 50 time frames at the beginning and end of sequences
 # The autoencoder has difficulties with the beginning and ending frames
 data correct = np.zeros((X correct.shape[0],n timesteps+100,n dim))
 for i in range(X_correct.shape[0]):
      data correct[i,:,:] = np.concatenate((np.concatenate((np.tile(X correct[i,0,:],[50, 1])
 data_incorrect = np.zeros((X_incorrect.shape[0],n_timesteps+100,n_dim))
 for i in range(X_incorrect.shape[0]):
      data_incorrect[i,:,:] = np.concatenate((np.concatenate((np.tile(X_incorrect[i,0,:],[50,
 # Plot the first sequences of correct and incorrect data
 plt.figure(figsize = (12,6))
 # plt.title()
 ax = plt.subplot(1,2,1)
 ax.set title(' Sequence of correct data for exercise \n'+exercises names[exercise num] ,fo
 plt.plot(data_correct[0])
 plt.ylim([-1,1])
 ax2 = plt.subplot(1,2,2)
 ax2.set_title(' Sequence of incorrect data for exercise \n'+exercises_names[exercise_num] ,
 plt.plot(data incorrect[0])
 plt.ylim([-1,1])
 plt.tight layout()
 plt.savefig("/content/drive/My Drive/DeepLearningProject/Figures/pos_m0"+str(exercise_num+1
 plt.show()
```

3. Ploting segences of correct and incorrect

```
def sequence of the exercises(files corrected, files incorrected, exercise num):
 X_correct, X_incorrect = load_data( files_corrected , files_incorrected)
 print(X correct.shape, 'correct sequences')
 print(X_incorrect.shape, 'incorrect sequences')
 # Plot the first sequences of correct and incorrect data
 plt.figure(figsize = (12,6))
 plt.subplot(1,2,1)
 plt.plot(X_correct[0])
 plt.ylim([-1,1])
 plt.subplot(1,2,2)
 plt.plot(X_incorrect[0])
 plt.ylim([-1,1])
 plt.tight_layout()
 plt.show()
 # adding frames(X correct, X incorrect , exercise num )
```

4. Model Building

```
# Encoder layers
input seq = Input(shape=(n timesteps+100, n dim))
encoded1 = LSTM(30,return_sequences = True)(input_seq)
encoded2 = LSTM(10,return sequences = True)(encoded1)
# Encoded representation of the input, 340x4 vector
encoded = LSTM(4,return_sequences = True)(encoded2)
# Decoder layers
decoded1 = LSTM(10, return_sequences = True)(encoded)
decoded2 = LSTM(30,return sequences = True)(decoded1)
decoded = LSTM(n dim, return sequences = True)(decoded2)
# The model maps an input to its reconstruction
autoencoder = Model(inputs=input_seq, outputs=decoded)
autoencoder.compile(optimizer='adam', loss='mse')
autoencoder.summary()
def running_models(index):
 data correct , data incorrect = data setting(index)
 import random
 trainidx = random.sample(range(0,data correct.shape[0]),63)
 valididx = np.setdiff1d(np.arange(0,90,1),trainidx)
 train_data = data_correct[trainidx,:,:]
 valid data = data correct[valididx,:,:]
  import os
```

```
TIIIPOI C 03
# Directory where the checkpoints will be saved
# checkpoint dir = '/content/drive/My Drive/DeepLearningProject/trainingcheckpoints/m0'+str
# # Name of the checkpoint files
# checkpoint prefix = os.path.join(checkpoint dir, "ckpt {epoch}")
# checkpoint callback=tf.keras.callbacks.ModelCheckpoint(
#
      filepath=checkpoint prefix,
#
      save weights only=True)
# Train an autoencoder on the correct data sequences
# Measure the training time
t = now()
# Request to stop before reaching the number of epochs if the validation loss does not decr
early stopping = EarlyStopping(monitor='val loss', patience = 1000)
history = autoencoder.fit(train data, train data, epochs = 10000, batch size = batch size,
                validation data=(valid data, valid data), verbose = 0, callbacks = [early s
print('Training time: %s' % (now() - t))
loss(history,index)
# Encode and decode sequences to check the model performance
decoded seqs = autoencoder.predict(data correct)
decode Sequence Plot(decoded seqs,index,data correct)
  # Create an encoder model, that maps an input to its encoded representation
encoder = Model(inputs=input seq, outputs=encoded)
# Test the encoder model
encoded seqs = encoder.predict(data correct)
encoded sequence plot(encoded seqs,index,data correct)
time frame autencoder plot(encoded seqs,index,data correct)
# Remove the added first and last 50 frames
encoded_seqs = encoded_seqs[:,50:-50,:]
print(encoded seqs.shape, 'encoded sequences shape')
# Reshape the encoded sequences, because savetxt saves two dimensional data
seqs = encoded seqs.reshape(encoded seqs.shape[0],encoded seqs.shape[1]*encoded seqs.shape[
print(seqs.shape, 'encoded sequences shape for saving')
# Save the data in the file 'Autoencoder Output Correct.csv'
np.savetxt('/content/drive/My Drive/DeepLearningProject/Segmented Movements/Data_Proccessed
#files.download('/content/drive/My Drive/DeepLearningProject/Segmented Movements/Data Procc
# Reduce the dimensionality of the incorrect sequences
encoded seqs incorrect = encoder.predict(data incorrect)
# Remove the added first and last 50 frames
encoded seqs incorrect = encoded seqs incorrect[:,50:-50,:]
```

```
print(encoded_seqs_incorrect.shape, 'encoded incorrect sequences shape')
# Reshape the encoded sequences, because savetxt saves only tow dimensional data
seqs_incorrect = encoded_seqs_incorrect.reshape(encoded_seqs_incorrect.shape[0],encoded_seq
print(seqs_incorrect.shape, 'encoded incorrect sequences shape for saving')
# Save the incorrect data in the file 'Autoencoder_Output_Incorrect.csv'
np.savetxt('/content/drive/My Drive/DeepLearningProject/Segmented Movements/Data_Proccessed
#files.download('/content/drive/My Drive/DeepLearningProject/Segmented Movements/Data_Procc
```

→ 5 . Ploting loss , encoders output, decoder output and autoence

```
def loss(history , index):
 from google.colab import files
 plt.figure()
 plt.subplot(121)
 plt.plot(history.history['loss'])
 plt.title('Loss for \n'+str(exercises_names[index]))
 plt.subplot(122)
 plt.plot(history.history['val loss'])
 plt.title('Validation Loss for \n'+str(exercises_names[index]))
 plt.tight layout()
 plt.savefig("/content/drive/My Drive/DeepLearningProject/Figures/loss_pos_m0"+str(index+1)+
 plt.show()
 #files.download('/content/drive/My Drive/DeepLearningProject/Figures/loss m0'+str(index+1)+
 # Print the resulting training and validation loss values
 print(history.history['loss'][-1])
 print(history.history['val_loss'][-1])
def decode_Sequence_Plot(decoded_seqs,index,data_correct):
 # Plot the results
 n = 2 # how many sequences we will display
 plt.figure(figsize = (12,6))
 for i in range(n):
      # display original sequences
      plt.subplot(n, 2, 2*i+1)
      plt.plot(data_correct[i])
      plt.title('original sequences \n'+str(exercises_names[index]))
      # display reconstruction
      plt.subplot(n, 2, 2*i+2)
      plt.plot(decoded seqs[i])
      plt.title('decoded sequences \n'+str(i+1)+" : "+str(exercises_names[index]))
  plt.tight layout()
 plt.savefig("/content/drive/My Drive/DeepLearningProject/Figures/original seq vs decoded se
 #files.download("/content/drive/My Drive/DeepLearningProject/Figures/original_seq_vs_decode
 plt.show()
```

```
def encoded sequence plot(encoded seqs,index,data correct):
 # Plot the results
 n = 2 # how many sequences we will display
 plt.figure(figsize = (12,6))
 for i in range(n):
      # display original sequences
      plt.subplot(n, 2, 2*i+1)
      plt.plot(data correct[i])
      plt.title('original sequences \n'+str(exercises_names[index]))
      # display reconstruction
      plt.subplot(n, 2, 2*i+2)
      plt.plot(encoded seqs[i])
      plt.title('encoded sequences \n'+str(i+1)+" :"+str(exercises_names[index]))
 plt.tight_layout()
 plt.savefig("/content/drive/My Drive/DeepLearningProject/Figures/original seq vs encoded se
 #files.download("/content/drive/My Drive/DeepLearningProject/Figures/original_seq_vs_encode
 plt.show()
def time_frame_autencoder_plot(encoded_seqs,index,data_correct):
 plt.figure(figsize = (14,14))
 for i in range(data correct.shape[0]):
      plt.subplot(4,1,1)
      plt.plot(encoded_seqs[i,50:-50,0])
      plt.xlabel('Time Frame', fontsize=12)
      plt.ylabel('Angle (Degrees)',fontsize=12)
      plt.subplot(4,1,2)
      plt.plot(encoded seqs[i,50:-50,1])
      plt.xlabel('Time Frame', fontsize=12)
      plt.ylabel('Angle (Degrees)',fontsize=12)
      plt.subplot(4,1,3)
      plt.plot(encoded seqs[i,50:-50,2])
      plt.xlabel('Time Frame', fontsize=12)
      plt.ylabel('Angle (Degrees)',fontsize=12)
      plt.subplot(4,1,4)
      plt.plot(encoded seqs[i,50:-50,3])
      plt.xlabel('Time Frame', fontsize=12)
      plt.ylabel('Angle (Degrees)',fontsize=12)
  plt.tight layout()
 plt.savefig("/content/drive/My Drive/DeepLearningProject/Figures/autoencoder output pos m0"
 #files.download("/content/drive/My Drive/DeepLearningProject/Figures/autoencoder output m0"
 plt.show()
```

→ . 6 Running Models

```
for i in range(0,10):
    running_models(i)
```