LSTM Model

May 6, 2022

1 IE 8990 Course Project

1.1 Importing library

1.2 Custom Activation Function

```
[3]: ## Custom Activation Function

from tensorflow.keras import backend as K

def Swish(inputs):
    return inputs*K.sigmoid(inputs)
    #return K.maximum(inputs,0)
    #return K.minimum(K.maximum((K.sign(inputs)*K.pow(K.abs(inputs),1/3)),0),
    #inputs*K.sigmoid(inputs))
    #return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.

⇒sigmoid(inputs),
```

```
#0.05*x*K.siqmoid(inputs))
def CustFunc(inputs,p=0.00):
    return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.sigmoid(inputs),
                                                          p*inputs*K.
→sigmoid(inputs))
    #return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.
\rightarrow sigmoid(inputs), 0)
def CustFunc05(inputs,p=0.05):
    return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.sigmoid(inputs),
                                                          p*inputs*K.
→sigmoid(inputs))
    #return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.
\rightarrow sigmoid(inputs), 0)
def CustFunc15(inputs,p=0.15):
    return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.sigmoid(inputs),
                                                          p*inputs*K.
→sigmoid(inputs))
    #return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.
\rightarrow sigmoid(inputs), 0)
def CustFunc25(inputs,p=0.25):
    return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.sigmoid(inputs),
                                                          p*inputs*K.
→sigmoid(inputs))
    #return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.
\rightarrow sigmoid(inputs), 0)
def CustFunc35(inputs,p=0.35):
    return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.sigmoid(inputs),
                                                          p*inputs*K.
→sigmoid(inputs))
    #return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.
\rightarrow sigmoid(inputs), 0)
def CustFunc45(inputs,p=0.45):
    return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.sigmoid(inputs),
                                                          p*inputs*K.
→sigmoid(inputs))
    #return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.
\rightarrow sigmoid(inputs), 0)
def CustFunc1(inputs):
    \#return\ K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.
\rightarrow sigmoid(inputs),
    #0.05*inputs*K.sigmoid(inputs))
```

```
return K.maximum(K.softsign(inputs),0) + K.minimum(inputs*K.

→sigmoid(inputs), 0)
```

```
[4]: # Plotting the Custom Function

### Setting the Coordinate Space

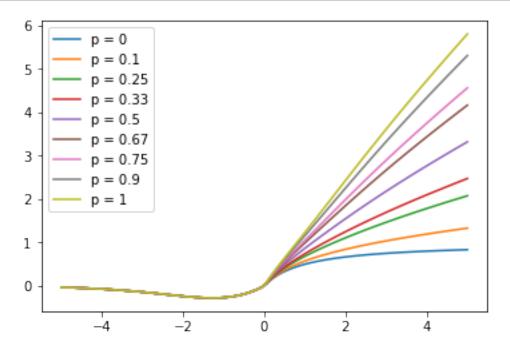
x = np.linspace(-5, 5, 100)
```

```
[5]: ### Several Combinations of Mixed Function

X = [0, .1, .25, .33, .5, .67, .75, .90, 1]

for i in X:
    plt.plot(x,CustFunc(x,i), label="p = {}".format(i))

plt.legend()
plt.show()
```



1.3 Importing Data

```
[7]: # Setting the training and test variables

## This loads the data row by row to make it a time series based dataset.

(x_train, y_train), (x_test, y_test) = mnist.load_data()
   x_train = x_train.astype("float32")/255.0
   x_test = x_test.astype("float32")/255.0
```

1.4 LSTM Model

1.4.1 Base Model

Model: "sequential"

Layer (type)	Output Shape	Param #
lstm (LSTM)	(None, None, 256)	291840
lstm_1 (LSTM)	(None, 256)	525312
dense (Dense)	(None, 10)	2570

Total params: 819,722 Trainable params: 819,722 Non-trainable params: 0

None

1.4.2 Reference for LSTM Model

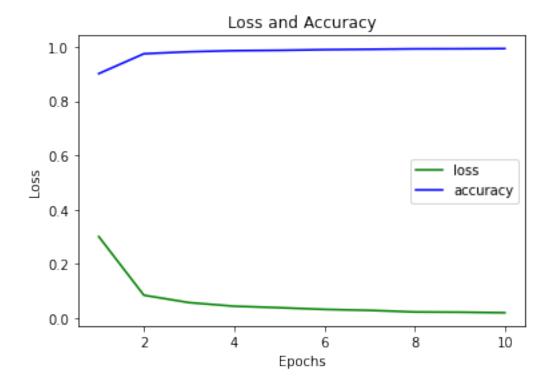
def init(self, activation='tanh', recurrent_activation='sigmoid', units, use bias=True, kernel initializer='glorot uniform', recurrent initializer='orthogonal', kernel regularizer=None, bias initializer='zeros', unit forget bias=True, rent regularizer=None, bias regularizer=None, kernel constraint=None, recurrent constraint=None, bias constraint=None, dropout=0.,recurrent dropout=0., super(LSTMCell, self).___init___(units, activation=activation, 'kwargs): rent_activation=recurrent_activation, use_bias=use_bias, kernel_initializer=kernel_initializer, recurrent initializer=recurrent initializer, bias initializer=bias initializer, unit forget bias=unit forget bias, kernel regularizer=kernel regularizer, recurrent regularizer=recurrent regularizer, bias regularizer=bias regularizer, kernel constraint=kernel constraint, recurrent constraint=recurrent constraint, bias constraint=bias constraint, dropout=dropout, recurrent dropout=recurrent dropout, implementation=kwargs.pop('implementation', 2), **kwargs)

```
[9]: # Compiling the Model

model.compile(
   loss = keras.losses.SparseCategoricalCrossentropy(from_logits=True),
   optimizer = keras.optimizers.Adam(learning_rate=0.001),
   metrics=['accuracy']
)
```

```
Epoch 1/10
938/938 - 11s - loss: 0.3006 - accuracy: 0.9016 - 11s/epoch - 11ms/step
Epoch 2/10
938/938 - 8s - loss: 0.0846 - accuracy: 0.9749 - 8s/epoch - 8ms/step
Epoch 3/10
938/938 - 8s - loss: 0.0573 - accuracy: 0.9822 - 8s/epoch - 8ms/step
Epoch 4/10
```

```
938/938 - 8s - loss: 0.0440 - accuracy: 0.9862 - 8s/epoch - 8ms/step
     Epoch 5/10
     938/938 - 8s - loss: 0.0385 - accuracy: 0.9875 - 8s/epoch - 8ms/step
     Epoch 6/10
     938/938 - 8s - loss: 0.0323 - accuracy: 0.9900 - 8s/epoch - 8ms/step
     Epoch 7/10
     938/938 - 8s - loss: 0.0287 - accuracy: 0.9913 - 8s/epoch - 8ms/step
     Epoch 8/10
     938/938 - 8s - loss: 0.0228 - accuracy: 0.9929 - 8s/epoch - 8ms/step
     Epoch 9/10
     938/938 - 8s - loss: 0.0220 - accuracy: 0.9931 - 8s/epoch - 8ms/step
     Epoch 10/10
     938/938 - 8s - loss: 0.0199 - accuracy: 0.9939 - 8s/epoch - 8ms/step
     157/157 - 1s - loss: 0.0341 - accuracy: 0.9911 - 1s/epoch - 7ms/step
[10]: [0.03406180813908577, 0.991100013256073]
[11]: loss_base = history.history['loss']
      acc_base = history.history['accuracy']
      epochs = range(1,11)
      plt.plot(epochs, loss_base, 'g', label='loss')
      plt.plot(epochs, acc_base, 'b', label='accuracy')
      plt.title('Loss and Accuracy')
      plt.xlabel('Epochs')
      plt.ylabel('Loss')
      plt.legend()
      plt.show()
```



1.4.3 Squish Function

WARNING:tensorflow:Layer lstm_2 will not use cuDNN kernels since it doesn't meet

the criteria. It will use a generic GPU kernel as fallback when running on GPU. WARNING:tensorflow:Layer lstm_3 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU. Model: "sequential_1"

Layer (type)	Output Shape	Param #
lstm_2 (LSTM)	(None, None, 256)	291840
lstm_3 (LSTM)	(None, 256)	525312
dense_1 (Dense)	(None, 10)	2570

Total params: 819,722 Trainable params: 819,722 Non-trainable params: 0

```
Epoch 1/10
938/938 - 274s - loss: 0.3854 - accuracy: 0.8730 - 274s/epoch - 292ms/step
Epoch 2/10
938/938 - 271s - loss: 0.1010 - accuracy: 0.9692 - 271s/epoch - 289ms/step
Epoch 3/10
938/938 - 272s - loss: 0.0674 - accuracy: 0.9796 - 272s/epoch - 290ms/step
Epoch 4/10
938/938 - 272s - loss: 0.0510 - accuracy: 0.9846 - 272s/epoch - 290ms/step
```

```
Epoch 5/10
     938/938 - 271s - loss: 0.0425 - accuracy: 0.9866 - 271s/epoch - 289ms/step
     Epoch 6/10
     938/938 - 272s - loss: 0.0340 - accuracy: 0.9896 - 272s/epoch - 290ms/step
     Epoch 7/10
     938/938 - 273s - loss: 0.0303 - accuracy: 0.9905 - 273s/epoch - 291ms/step
     Epoch 8/10
     938/938 - 271s - loss: 0.0246 - accuracy: 0.9919 - 271s/epoch - 289ms/step
     Epoch 9/10
     938/938 - 272s - loss: 0.0220 - accuracy: 0.9928 - 272s/epoch - 290ms/step
     Epoch 10/10
     938/938 - 276s - loss: 0.0212 - accuracy: 0.9932 - 276s/epoch - 294ms/step
     157/157 - 9s - loss: 0.0380 - accuracy: 0.9887 - 9s/epoch - 56ms/step
[13]: [0.0379953570663929, 0.9886999726295471]
[14]: loss_AF = history_AF.history['loss']
      acc_AF = history_AF.history['accuracy']
[15]: # Making the model
      # Input Layer
      modelRAF = keras.Sequential()
      modelRAF.add(keras.Input(shape = (None, 28))) #28 pixels per time step
      # Hidden Layers
      modelRAF.add(
          layers.LSTM(256, return_sequences = True, activation = 'tanh',
                      recurrent_activation= CustFunc))
      modelRAF.add(
          layers.LSTM(256, activation = 'tanh', recurrent_activation = CustFunc))
      # Output Layer
      modelRAF.add(layers.Dense(10))
      # Model Summary
      print(modelRAF.summary())
```

WARNING:tensorflow:Layer lstm_4 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU. WARNING:tensorflow:Layer lstm_5 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU. Model: "sequential_2"

Layer (type)	Output Shape	Param #
lstm_4 (LSTM)	(None, None, 256)	291840

```
lstm_5 (LSTM)
                                (None, 256)
                                                          525312
      dense_2 (Dense)
                                 (None, 10)
                                                          2570
     _____
     Total params: 819,722
     Trainable params: 819,722
     Non-trainable params: 0
     None
[16]: # Compiling the Model
     modelRAF.compile(
         loss = keras.losses.SparseCategoricalCrossentropy(from_logits=True),
         optimizer = keras.optimizers.Adam(learning_rate=0.001),
         metrics=['accuracy']
     # Fitting the Model
     history_RAF = modelRAF.fit(x_train, y_train, batch_size = 64, epochs = 10,__
      \rightarrowverbose = 2)
     # Evaluate the Model
     modelRAF.evaluate(x_test, y_test, batch_size = 64, verbose = 2)
     Epoch 1/10
     938/938 - 361s - loss: 2.3016 - accuracy: 0.1116 - 361s/epoch - 385ms/step
     Epoch 2/10
     938/938 - 356s - loss: 0.4030 - accuracy: 0.8627 - 356s/epoch - 380ms/step
     Epoch 3/10
     938/938 - 358s - loss: 0.0854 - accuracy: 0.9752 - 358s/epoch - 382ms/step
     Epoch 4/10
     938/938 - 354s - loss: 0.0565 - accuracy: 0.9833 - 354s/epoch - 378ms/step
     Epoch 5/10
     938/938 - 353s - loss: 0.0438 - accuracy: 0.9871 - 353s/epoch - 376ms/step
     Epoch 6/10
     938/938 - 356s - loss: 0.0368 - accuracy: 0.9888 - 356s/epoch - 379ms/step
     Epoch 7/10
     938/938 - 357s - loss: 0.0293 - accuracy: 0.9912 - 357s/epoch - 381ms/step
     Epoch 8/10
     938/938 - 360s - loss: 0.0285 - accuracy: 0.9912 - 360s/epoch - 384ms/step
     Epoch 9/10
```

```
938/938 - 360s - loss: 0.0220 - accuracy: 0.9934 - 360s/epoch - 384ms/step
Epoch 10/10
938/938 - 356s - loss: 0.0182 - accuracy: 0.9948 - 356s/epoch - 379ms/step
157/157 - 12s - loss: 0.0389 - accuracy: 0.9898 - 12s/epoch - 75ms/step

[16]: [0.03892815113067627, 0.989799976348877]

[17]: loss_RAF = history_RAF.history['loss']
acc_RAF = history_RAF.history['accuracy']
```

1.4.4 ReLU function

WARNING:tensorflow:Layer lstm_6 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU. WARNING:tensorflow:Layer lstm_7 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU. Model: "sequential_3"

Layer (type)	Output Shape	Param #
lstm_6 (LSTM)	(None, None, 256)	291840
lstm_7 (LSTM)	(None, 256)	525312
dense_3 (Dense)	(None, 10)	2570

Total params: 819,722

Trainable params: 819,722 Non-trainable params: 0

None

```
Epoch 1/10
938/938 - 136s - loss: 0.4601 - accuracy: 0.8533 - 136s/epoch - 145ms/step
Epoch 2/10
938/938 - 135s - loss: 0.1070 - accuracy: 0.9677 - 135s/epoch - 144ms/step
Epoch 3/10
938/938 - 135s - loss: 0.0706 - accuracy: 0.9786 - 135s/epoch - 144ms/step
Epoch 4/10
938/938 - 135s - loss: 0.0549 - accuracy: 0.9833 - 135s/epoch - 144ms/step
Epoch 5/10
938/938 - 134s - loss: 0.0472 - accuracy: 0.9854 - 134s/epoch - 143ms/step
Epoch 6/10
938/938 - 133s - loss: 0.0409 - accuracy: 0.9873 - 133s/epoch - 142ms/step
Epoch 7/10
938/938 - 134s - loss: 0.0343 - accuracy: 0.9894 - 134s/epoch - 143ms/step
Epoch 8/10
938/938 - 135s - loss: 0.0312 - accuracy: 0.9906 - 135s/epoch - 144ms/step
Epoch 9/10
938/938 - 134s - loss: 0.0257 - accuracy: 0.9920 - 134s/epoch - 142ms/step
Epoch 10/10
938/938 - 134s - loss: 0.0286 - accuracy: 0.9911 - 134s/epoch - 143ms/step
157/157 - 5s - loss: 0.0421 - accuracy: 0.9872 - 5s/epoch - 29ms/step
```

[19]: [0.042086515575647354, 0.9872000217437744]

```
[20]: loss_R = historyR.history['loss']
     acc_R = historyR.history['accuracy']
[21]: # Making the model
     # Input Layer
     modelRR = keras.Sequential()
     modelRR.add(keras.Input(shape = (None, 28))) #28 pixels per time step
     # Hidden Layers
     modelRR.add(
         layers.LSTM(256, return_sequences = True, activation = 'tanh',
                    recurrent_activation='relu'))
         layers.LSTM(256, activation = 'tanh', recurrent_activation='relu'))
     # Output Layer
     modelRR.add(layers.Dense(10))
     # Model Summary
     print(modelRR.summary())
     WARNING:tensorflow:Layer lstm_8 will not use cuDNN kernels since it doesn't meet
     the criteria. It will use a generic GPU kernel as fallback when running on GPU.
     WARNING:tensorflow:Layer lstm_9 will not use cuDNN kernels since it doesn't meet
     the criteria. It will use a generic GPU kernel as fallback when running on GPU.
     Model: "sequential_4"
     Layer (type)
                               Output Shape
     ______
                                (None, None, 256)
      lstm_8 (LSTM)
                                                         291840
      lstm_9 (LSTM)
                                 (None, 256)
                                                         525312
      dense_4 (Dense)
                                 (None, 10)
                                                         2570
     Total params: 819,722
     Trainable params: 819,722
     Non-trainable params: 0
     None
[22]: # Compiling the Model
```

```
modelRR.compile(
          loss = keras.losses.SparseCategoricalCrossentropy(from_logits=True),
          optimizer = keras.optimizers.Adam(learning_rate=0.001),
          metrics=['accuracy']
      # Fitting the Model
      historyRR = modelRR.fit(x_train, y_train, batch_size = 64, epochs = 10, verbose_
      # Evaluate the Model
     modelRR.evaluate(x_test, y_test, batch_size = 64, verbose = 2)
     Epoch 1/10
     938/938 - 136s - loss: 0.5441 - accuracy: 0.8208 - 136s/epoch - 145ms/step
     Epoch 2/10
     938/938 - 134s - loss: 0.1337 - accuracy: 0.9606 - 134s/epoch - 143ms/step
     Epoch 3/10
     938/938 - 135s - loss: 0.1182 - accuracy: 0.9665 - 135s/epoch - 143ms/step
     Epoch 4/10
     938/938 - 133s - loss: 0.0878 - accuracy: 0.9745 - 133s/epoch - 142ms/step
     Epoch 5/10
     938/938 - 135s - loss: 0.0829 - accuracy: 0.9765 - 135s/epoch - 143ms/step
     Epoch 6/10
     938/938 - 133s - loss: 0.0689 - accuracy: 0.9804 - 133s/epoch - 141ms/step
     Epoch 7/10
     938/938 - 133s - loss: 0.0645 - accuracy: 0.9811 - 133s/epoch - 142ms/step
     Epoch 8/10
     938/938 - 134s - loss: 0.0570 - accuracy: 0.9837 - 134s/epoch - 143ms/step
     Epoch 9/10
     938/938 - 134s - loss: 0.0711 - accuracy: 0.9797 - 134s/epoch - 143ms/step
     Epoch 10/10
     938/938 - 135s - loss: 0.0584 - accuracy: 0.9828 - 135s/epoch - 144ms/step
     157/157 - 5s - loss: 0.1710 - accuracy: 0.9425 - 5s/epoch - 29ms/step
[22]: [0.17101824283599854, 0.9424999952316284]
[23]: loss_RR = historyRR.history['loss']
      acc_RR = historyRR.history['accuracy']
```

1.4.5 Swish Function

WARNING:tensorflow:Layer lstm_10 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

WARNING:tensorflow:Layer lstm_11 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

Model: "sequential_5"

Layer (type)	Output Shape	Param #
lstm_10 (LSTM)	(None, None, 256)	291840
lstm_11 (LSTM)	(None, 256)	525312
dense_5 (Dense)	(None, 10)	2570

Total params: 819,722 Trainable params: 819,722 Non-trainable params: 0

```
[26]: # Compiling the Model
```

```
modelS.compile(
          loss = keras.losses.SparseCategoricalCrossentropy(from_logits=True),
          optimizer = keras.optimizers.Adam(learning_rate=0.001),
          metrics=['accuracy']
      # Fitting the Model
      historyS = modelS.fit(x_train, y_train, batch_size = 64, epochs = 10, verbose = 0
      # Evaluate the Model
     modelS.evaluate(x_test, y_test, batch_size = 64, verbose = 2)
     Epoch 1/10
     938/938 - 170s - loss: 0.6607 - accuracy: 0.7753 - 170s/epoch - 182ms/step
     Epoch 2/10
     938/938 - 171s - loss: 0.1354 - accuracy: 0.9587 - 171s/epoch - 182ms/step
     Epoch 3/10
     938/938 - 171s - loss: 0.0820 - accuracy: 0.9744 - 171s/epoch - 183ms/step
     Epoch 4/10
     938/938 - 172s - loss: 0.0588 - accuracy: 0.9823 - 172s/epoch - 183ms/step
     Epoch 5/10
     938/938 - 170s - loss: 0.0455 - accuracy: 0.9854 - 170s/epoch - 181ms/step
     Epoch 6/10
     938/938 - 169s - loss: 0.0370 - accuracy: 0.9887 - 169s/epoch - 180ms/step
     Epoch 7/10
     938/938 - 173s - loss: 0.0334 - accuracy: 0.9897 - 173s/epoch - 184ms/step
     Epoch 8/10
     938/938 - 172s - loss: 0.0306 - accuracy: 0.9900 - 172s/epoch - 183ms/step
     Epoch 9/10
     938/938 - 174s - loss: 0.0238 - accuracy: 0.9926 - 174s/epoch - 186ms/step
     Epoch 10/10
     938/938 - 174s - loss: 0.0241 - accuracy: 0.9925 - 174s/epoch - 185ms/step
     157/157 - 5s - loss: 0.0365 - accuracy: 0.9882 - 5s/epoch - 33ms/step
[26]: [0.03652001544833183, 0.9882000088691711]
[27]: loss_S = historyS.history['loss']
      acc_S = historyS.history['accuracy']
[28]: # Making the model
      # Input Layer
```

WARNING:tensorflow:Layer lstm_12 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

WARNING:tensorflow:Layer lstm_13 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

Model: "sequential_6"

Layer (type)	Output Shape	Param #
lstm_12 (LSTM)	(None, None, 256)	291840
lstm_13 (LSTM)	(None, 256)	525312
dense_6 (Dense)	(None, 10)	2570

Total params: 819,722 Trainable params: 819,722 Non-trainable params: 0

```
[29]: # Compiling the Model

modelSR.compile(
   loss = keras.losses.SparseCategoricalCrossentropy(from_logits=True),
   optimizer = keras.optimizers.Adam(learning_rate=0.001),
   metrics=['accuracy']
)
```

```
# Fitting the Model
historySR = modelSR.fit(x_train, y_train, batch_size = 64, epochs = 10, verbose_
 \Rightarrow= 2)
# Evaluate the Model
modelSR.evaluate(x_test, y_test, batch_size = 64, verbose = 2)
Epoch 1/10
938/938 - 193s - loss: 2.1998 - accuracy: 0.1673 - 193s/epoch - 206ms/step
Epoch 2/10
938/938 - 191s - loss: 0.6255 - accuracy: 0.7884 - 191s/epoch - 204ms/step
Epoch 3/10
938/938 - 191s - loss: 0.2406 - accuracy: 0.9261 - 191s/epoch - 203ms/step
Epoch 4/10
938/938 - 193s - loss: 2.7624 - accuracy: 0.4051 - 193s/epoch - 206ms/step
Epoch 5/10
938/938 - 191s - loss: 1.9513 - accuracy: 0.3921 - 191s/epoch - 204ms/step
Epoch 6/10
938/938 - 189s - loss: 1.8397 - accuracy: 0.4491 - 189s/epoch - 201ms/step
Epoch 7/10
938/938 - 187s - loss: 1.7553 - accuracy: 0.4752 - 187s/epoch - 199ms/step
Epoch 8/10
938/938 - 187s - loss: 1.6859 - accuracy: 0.4886 - 187s/epoch - 199ms/step
Epoch 9/10
938/938 - 185s - loss: 1.8491 - accuracy: 0.4029 - 185s/epoch - 197ms/step
Epoch 10/10
938/938 - 189s - loss: 1.6941 - accuracy: 0.4744 - 189s/epoch - 201ms/step
157/157 - 6s - loss: 1.5770 - accuracy: 0.4990 - 6s/epoch - 37ms/step
```

[29]: [1.5770384073257446, 0.49900001287460327]

```
[30]: loss_SR = historySR.history['loss'] acc_SR = historySR.history['accuracy']
```

1.4.6 SoftMax Function

```
[31]: # Making the model

# Input Layer
modelSM = keras.Sequential()
modelSM.add(keras.Input(shape = (None, 28))) #28 pixels per time step
```

WARNING:tensorflow:Layer lstm_14 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

WARNING:tensorflow:Layer lstm_15 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

Model: "sequential_7"

Layer (type)	Output Shape	Param #
lstm_14 (LSTM)	(None, None, 256)	291840
lstm_15 (LSTM)	(None, 256)	525312
dense_7 (Dense)	(None, 10)	2570

Total params: 819,722 Trainable params: 819,722 Non-trainable params: 0

```
[32]: # Compiling the Model

modelSM.compile(
   loss = keras.losses.SparseCategoricalCrossentropy(from_logits=True),
   optimizer = keras.optimizers.Adam(learning_rate=0.001),
   metrics=['accuracy']
)

# Fitting the Model
```

```
historySM = modelSM.fit(x_train, y_train, batch_size = 64, epochs = 10, verbose_
       ⇒= 2)
      # Evaluate the Model
      modelSM.evaluate(x_test, y_test, batch_size = 64, verbose = 2)
     Epoch 1/10
     938/938 - 150s - loss: 2.3016 - accuracy: 0.1120 - 150s/epoch - 160ms/step
     Epoch 2/10
     938/938 - 145s - loss: 2.3013 - accuracy: 0.1124 - 145s/epoch - 155ms/step
     Epoch 3/10
     938/938 - 147s - loss: 2.3013 - accuracy: 0.1124 - 147s/epoch - 156ms/step
     Epoch 4/10
     938/938 - 145s - loss: 2.3013 - accuracy: 0.1124 - 145s/epoch - 154ms/step
     Epoch 5/10
     938/938 - 145s - loss: 2.3014 - accuracy: 0.1124 - 145s/epoch - 155ms/step
     Epoch 6/10
     938/938 - 145s - loss: 2.3014 - accuracy: 0.1124 - 145s/epoch - 155ms/step
     Epoch 7/10
     938/938 - 146s - loss: 2.3013 - accuracy: 0.1124 - 146s/epoch - 155ms/step
     Epoch 8/10
     938/938 - 146s - loss: 2.3014 - accuracy: 0.1124 - 146s/epoch - 156ms/step
     Epoch 9/10
     938/938 - 146s - loss: 2.3014 - accuracy: 0.1124 - 146s/epoch - 155ms/step
     Epoch 10/10
     938/938 - 148s - loss: 2.3014 - accuracy: 0.1124 - 148s/epoch - 157ms/step
     157/157 - 5s - loss: 2.3010 - accuracy: 0.1135 - 5s/epoch - 31ms/step
[32]: [2.301018714904785, 0.11349999904632568]
[33]: loss_SM = historySM.history['loss']
      acc_SM = historySM.history['accuracy']
[34]: # Making the model
      # Input Layer
      modelSMR = keras.Sequential()
      modelSMR.add(keras.Input(shape = (None, 28))) #28 pixels per time step
      # Hidden Layers
      modelSMR.add(
          layers.LSTM(256, return_sequences = True, activation = 'tanh',
                      recurrent activation='softmax'))
      modelSMR.add(
```

```
layers.LSTM(256, activation = 'tanh',recurrent_activation='softmax'))

# Output Layer
modelSMR.add(layers.Dense(10))

# Model Summary
print(modelSM.summary())
```

WARNING:tensorflow:Layer lstm_16 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

WARNING:tensorflow:Layer lstm_17 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

Model: "sequential_7"

Layer (type)	Output Shape	Param #
lstm_14 (LSTM)	(None, None, 256)	291840
lstm_15 (LSTM)	(None, 256)	525312
dense_7 (Dense)	(None, 10)	2570

Total params: 819,722 Trainable params: 819,722 Non-trainable params: 0

```
[35]: # Compiling the Model

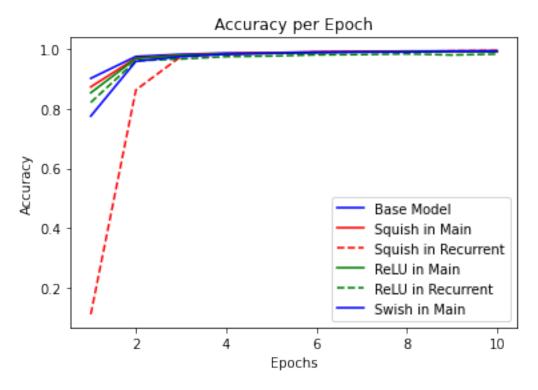
modelSMR.compile(
    loss = keras.losses.SparseCategoricalCrossentropy(from_logits=True),
    optimizer = keras.optimizers.Adam(learning_rate=0.001),
    metrics=['accuracy']
)

# Fitting the Model

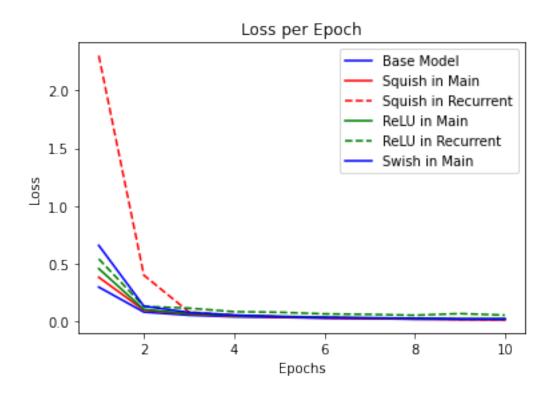
historySMR = modelSMR.fit(x_train, y_train, batch_size = 64, epochs = 10, \( \text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\
```

```
modelSMR.evaluate(x_test, y_test, batch_size = 64, verbose = 2)
     Epoch 1/10
     938/938 - 156s - loss: 2.3015 - accuracy: 0.1116 - 156s/epoch - 166ms/step
     Epoch 2/10
     938/938 - 153s - loss: 2.3013 - accuracy: 0.1124 - 153s/epoch - 163ms/step
     Epoch 3/10
     938/938 - 152s - loss: 2.3013 - accuracy: 0.1124 - 152s/epoch - 162ms/step
     Epoch 4/10
     938/938 - 156s - loss: 2.3013 - accuracy: 0.1124 - 156s/epoch - 166ms/step
     Epoch 5/10
     938/938 - 152s - loss: 2.3013 - accuracy: 0.1124 - 152s/epoch - 163ms/step
     Epoch 6/10
     938/938 - 154s - loss: 2.3013 - accuracy: 0.1124 - 154s/epoch - 164ms/step
     Epoch 7/10
     938/938 - 152s - loss: 2.3013 - accuracy: 0.1124 - 152s/epoch - 162ms/step
     Epoch 8/10
     938/938 - 150s - loss: 2.3013 - accuracy: 0.1124 - 150s/epoch - 160ms/step
     Epoch 9/10
     938/938 - 154s - loss: 2.3013 - accuracy: 0.1124 - 154s/epoch - 164ms/step
     Epoch 10/10
     938/938 - 157s - loss: 2.3013 - accuracy: 0.1124 - 157s/epoch - 168ms/step
     157/157 - 6s - loss: 2.3010 - accuracy: 0.1135 - 6s/epoch - 35ms/step
[35]: [2.3009729385375977, 0.11349999904632568]
[36]: loss_SMR = historySMR.history['loss']
      acc_SMR = historySMR.history['accuracy']
[37]: ## Results Analysis
[41]: # Plotting Accuracy
      epochs = range(1,11)
      plt.plot(epochs, acc_base, 'b', label='Base Model')
      plt.plot(epochs, acc_AF, 'r-', label='Squish in Main')
      plt.plot(epochs, acc_RAF, 'r--', label='Squish in Recurrent')
      plt.plot(epochs, acc_R, 'g-', label='ReLU in Main')
      plt.plot(epochs, acc_RR, 'g--', label='ReLU in Recurrent')
      plt.plot(epochs, acc_S, 'b-', label='Swish in Main')
      #plt.plot(epochs, acc_SR, 'b--', label='Swish in Recurrent')
      #plt.plot(epochs, acc_SM, 'm-', label='SoftMax in Main')
      #plt.plot(epochs, acc_SM, 'm--', label='SoftMax in Recurrent')
      plt.title('Accuracy per Epoch')
      plt.xlabel('Epochs')
```

```
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```



```
[40]: # Plotting Loss
      epochs = range(1,11)
      plt.plot(epochs, loss_base, 'b', label='Base Model')
      plt.plot(epochs, loss_AF, 'r-', label='Squish in Main')
      plt.plot(epochs, loss_RAF, 'r--', label='Squish in Recurrent')
      plt.plot(epochs, loss_R, 'g-', label='ReLU in Main')
      plt.plot(epochs, loss_RR, 'g--', label='ReLU in Recurrent')
      plt.plot(epochs, loss_S, 'b-', label='Swish in Main')
      #plt.plot(epochs, loss_SR, 'b--', label='Swish in Recurrent')
      #plt.plot(epochs, loss_SM, 'm-', label='SoftMax in Main')
      #plt.plot(epochs, loss_SM, 'm--', label='SoftMax in Recurrent')
      plt.title('Loss per Epoch')
      plt.xlabel('Epochs')
      plt.ylabel('Loss')
      plt.legend()
      plt.show()
```



[]:	### Loss Landscapes
[]:	
[]:	
[]:	
[]:	