

# HAR\_LSTM\_done

January 6, 2020

```
[0]: # Importing Libraries
```

```
[0]: %tensorflow_version 1.x
```

```
[0]: import pandas as pd  
import numpy as np
```

```
[4]: from google.colab import drive  
drive.mount('/content/drive')
```

Go to this URL in a browser: [https://accounts.google.com/o/oauth2/auth?client\\_id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect\\_uri=urn%3aietf%3awg%3aoauth%3a2.0%3aoob&response\\_type=code&scope=email%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdocs.test%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly%20https%3a%2f%2fwww.googleapis.com%2fauth%2fpeopleapi.readonly](https://accounts.google.com/o/oauth2/auth?client_id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect_uri=urn%3aietf%3awg%3aoauth%3a2.0%3aoob&response_type=code&scope=email%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdocs.test%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly%20https%3a%2f%2fwww.googleapis.com%2fauth%2fpeopleapi.readonly)

Enter your authorization code:

ûûûûûûûûûûû

Mounted at /content/drive

```
[0]: dir_path="/content/drive/My Drive/Colab Notebooks/AppliedAI/  
      ↳HumanActivityRecognition/HAR/"
```

```
[0]: # Activities are the class labels
```

```
# It is a 6 class classification
```

```
ACTIVITIES = {  
    0: 'WALKING',  
    1: 'WALKING_UPSTAIRS',  
    2: 'WALKING_DOWNSTAIRS',  
    3: 'SITTING',  
    4: 'STANDING',  
    5: 'LAYING',  
}
```

```
# Utility function to print the confusion matrix
```

```
def confusion_matrix(Y_true, Y_pred):  
    Y_true = pd.Series([ACTIVITIES[y] for y in np.argmax(Y_true, axis=1)])  
    Y_pred = pd.Series([ACTIVITIES[y] for y in np.argmax(Y_pred, axis=1)])
```

```
return pd.crosstab(Y_true, Y_pred, rownames=['True'], colnames=['Pred'])
```

## 0.0.1 Data

```
[0]: # Data directory
DATADIR = 'UCI_HAR_Dataset'
```

```
[0]: # Raw data signals
# Signals are from Accelerometer and Gyroscope
# The signals are in x,y,z directions
# Sensor signals are filtered to have only body acceleration
# excluding the acceleration due to gravity
# Triaxial acceleration from the accelerometer is total acceleration
SIGNALS = [
    "body_acc_x",
    "body_acc_y",
    "body_acc_z",
    "body_gyro_x",
    "body_gyro_y",
    "body_gyro_z",
    "total_acc_x",
    "total_acc_y",
    "total_acc_z"
]
```

```
[0]: # Utility function to read the data from csv file
def _read_csv(filename):
    return pd.read_csv(filename, delim_whitespace=True, header=None)

# Utility function to load the load
def load_signals(subset):
    signals_data = []

    for signal in SIGNALS:
        filename = f'{dir_path}UCI_HAR_Dataset/{subset}/Inertial Signals/
→{signal}_{subset}.txt'
        signals_data.append(
            _read_csv(filename).as_matrix()
        )

    # Transpose is used to change the dimensionality of the output,
    # aggregating the signals by combination of sample/timestep.
    # Resultant shape is (7352 train/2947 test samples, 128 timesteps, 9
→signals)
    return np.transpose(signals_data, (1, 2, 0))
```

```
[10]: print(dir_path)
```

/content/drive/My Drive/Colab Notebooks/AppliedAI/HumanActivityRecognition/HAR/

```
[0]: def load_y(subset):  
    """  
    The objective that we are trying to predict is a integer, from 1 to 6,  
    that represents a human activity. We return a binary representation of  
    every sample objective as a 6 bits vector using One Hot Encoding  
    (https://pandas.pydata.org/pandas-docs/stable/generated/pandas.get\_dummies.  
    →html)  
    """  
    filename = f'{dir_path}UCI_HAR_Dataset/{subset}/y_{subset}.txt'  
    y = _read_csv(filename)[0]  
  
    # return pd.get_dummies(y).as_matrix()  
    return pd.get_dummies(y).values
```

```
[0]: def load_data():  
    """  
    Obtain the dataset from multiple files.  
    Returns: X_train, X_test, y_train, y_test  
    """  
    X_train, X_test = load_signals('train'), load_signals('test')  
    y_train, y_test = load_y('train'), load_y('test')  
  
    return X_train, X_test, y_train, y_test
```

```
[13]: # Importing tensorflow  
np.random.seed(42)  
import tensorflow as tf  
from keras import backend as K  
tf.set_random_seed(42)
```

Using TensorFlow backend.

```
[0]: # Configuring a session  
session_conf = tf.ConfigProto(  
    intra_op_parallelism_threads=1,  
    inter_op_parallelism_threads=1  
)
```

```
[0]: # Import Keras  
from keras import backend as K  
sess = tf.Session(graph=tf.get_default_graph(), config=session_conf)  
K.set_session(sess)
```

```
[0]: # Importing libraries  
from keras.models import Sequential  
from keras.layers import LSTM, Conv1D, MaxPooling1D, InputLayer, TimeDistributed, \
```

```

from keras.utils import to_categorical
from keras.optimizers import Adam
from keras.layers.core import Dense, Dropout, Flatten
from keras.layers.normalization import BatchNormalization
from keras.callbacks import ModelCheckpoint

```

```
[0]: # Initializing parameters
```

```

epochs = 30
batch_size = 16
n_hidden = 32

```

```
[0]: # Utility function to count the number of classes
```

```

def _count_classes(y):
    return len(set([tuple(category) for category in y]))

```

```
[40]: # Loading the train and test data
```

```
X_train, X_test, Y_train, Y_test = load_data()
```

/usr/local/lib/python3.6/dist-packages/ipykernel\_launcher.py:11: FutureWarning:  
Method .as\_matrix will be removed in a future version. Use .values instead.

# This is added back by InteractiveShellApp.init\_path()

```
[41]: print(X_train.shape)
```

```
print(X_test.shape)
```

```
print(Y_train.shape)
```

```
print(Y_test.shape)
```

```
(7352, 128, 9)
```

```
(2947, 128, 9)
```

```
(7352, 6)
```

```
(2947, 6)
```

```
[42]: timesteps = len(X_train[0])
```

```
input_dim = len(X_train[0][0])
```

```
n_classes = _count_classes(Y_train)
```

```
print(timesteps)
```

```
print(input_dim)
```

```
print(len(X_train))
```

```
128
```

```
9
```

```
7352
```

```
[0]: #container to hold all the parameters and results
```

```

d = {
    "layers" : [],

```

```

        "n_hidden" : [],
        "dropout" : [],
        "parameters_trained" : [],
        "train_acc" : [],
        "test_acc" : []
    }

```

```

[0]: def get_trainable_parameters(model):
    trainable_count = int(np.sum([K.count_params(p) for p in set(model.
    ↪ trainable_weights)]))
    return trainable_count

```

## 0.0.2 Defining the Architecture of LSTM

```

[0]: def define_compile_model(n_hidden=[32], layer=1, dropout=[0.5]):
    assert(len(n_hidden)==layer)
    global timesteps, input_dim, n_classes, d
    ret_seq = True if layer > 1 else False
    d["layers"].append(layer)
    d["n_hidden"].append(n_hidden)
    d["dropout"].append(dropout)
    model = Sequential()
    model.add(LSTM(n_hidden[0], input_shape=(timesteps, input_dim),
                  return_sequences=ret_seq))
    model.add(BatchNormalization())
    model.add(Dropout(dropout[0]))
    for i in range(1, layer):
        if i==layer-1:
            ret_seq = False
            model.add(LSTM(n_hidden[i], return_sequences=ret_seq))
            model.add(BatchNormalization())
            model.add(Dropout(dropout[i]))
    model.add(Dense(n_classes, activation='sigmoid'))
    model.summary()
    # Compiling the model
    model.compile(loss='categorical_crossentropy', optimizer='rmsprop',
                  metrics=['accuracy']) # also tried adam
    return model

```

```

[25]: # https://machinelearningmastery.com/
    ↪ how-to-develop-rnn-models-for-human-activity-recognition-time-series-classification/
    ↪
    # Trying out different custom models

    # Model 1 : LSTM

    model = Sequential()

```

```

model.add(LSTM(128, input_shape=(timesteps,input_dim),return_sequences=True))
model.add(Dropout(0.5))
model.add(LSTM(128))
model.add(Dropout(0.5))
model.add(Dense(100, activation='relu'))
model.add(Dense(n_classes, activation='softmax'))
model.compile(loss='categorical_crossentropy', optimizer='adam',
              metrics=['accuracy'])

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:541: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:4432: The name tf.random\_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:148: The name tf.placeholder\_with\_default is deprecated. Please use tf.compat.v1.placeholder\_with\_default instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3733: calling dropout (from tensorflow.python.ops.nn\_ops) with keep\_prob is deprecated and will be removed in a future version.

Instructions for updating:

Please use `rate` instead of `keep\_prob`. Rate should be set to `rate = 1 - keep\_prob`.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3576: The name tf.log is deprecated. Please use tf.math.log instead.

[0]: *# Model 2: CNN-LSTM*

```

n_steps, n_length = 4, 32
X_train = X_train.reshape((X_train.shape[0], n_steps, n_length, input_dim))
X_test = X_test.reshape((X_test.shape[0], n_steps, n_length, input_dim))

```

[0]:

```

model = Sequential()
model.add(TimeDistributed(Conv1D(filters=64, kernel_size=3,
                                activation='relu'), input_shape=(None,n_length,input_dim)))
model.add(TimeDistributed(Conv1D(filters=32, kernel_size=3,

```

```

        activation='relu'))))
model.add(TimeDistributed(Dropout(0.5)))
model.add(TimeDistributed(MaxPooling1D(pool_size=2)))
model.add(TimeDistributed(Flatten()))
model.add(LSTM(100))
model.add(Dropout(0.5))
model.add(Dense(100, activation='relu'))
model.add(Dense(n_classes, activation='softmax'))
model.compile(loss='categorical_crossentropy', optimizer='adam',
              metrics=['accuracy'])

```

[0]: *# Model 3: ConvLSTM2D*

```

timesteps, input_dim, n_classes = X_train.shape[1], X_train.shape[2], \
                                   Y_train.shape[1]
# reshape into subsequences (samples, time steps, rows, cols, channels)
n_steps, n_length = 4, 32
X_train = X_train.reshape((X_train.shape[0], n_steps, 1, n_length, input_dim))
X_test = X_test.reshape((X_test.shape[0], n_steps, 1, n_length, input_dim))

```

[0]: *# define model*

```

model = Sequential()
model.add(ConvLSTM2D(filters=64, kernel_size=(1,3),
                    activation='relu',
                    input_shape=(n_steps, 1, n_length, input_dim)))
model.add(Dropout(0.5))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dense(n_classes, activation='softmax'))
model.compile(loss='categorical_crossentropy', optimizer='adam',
              metrics=['accuracy'])

```

[46]: model.summary()

Model: "sequential\_5"

Layer (type)	Output Shape	Param #
time_distributed_11 (TimeDis	(None, None, 30, 64)	1792
time_distributed_12 (TimeDis	(None, None, 28, 32)	6176
time_distributed_13 (TimeDis	(None, None, 28, 32)	0
time_distributed_14 (TimeDis	(None, None, 14, 32)	0
time_distributed_15 (TimeDis	(None, None, 448)	0
lstm_5 (LSTM)	(None, 100)	219600

```

-----
dropout_9 (Dropout)                (None, 100)                0
-----
dense_9 (Dense)                    (None, 100)                10100
-----
dense_10 (Dense)                   (None, 6)                  606
=====
Total params: 238,274
Trainable params: 238,274
Non-trainable params: 0
-----

```

### 0.0.3 fit and predict model ..

```

[0]: def fit_model(X_train,y_train,batch_size,X_test,y_test,epochs,model):
      global d
      # Training the modela
      model.fit(X_train,Y_train,batch_size=batch_size,
                validation_data=(X_test, Y_test),epochs=epochs)
      d["parameters_trained"].append(get_trainable_parameters(model))
      d["train_acc"].append(model.history.history['acc'][-1])
      # Confusion Matrix
      print(confusion_matrix(Y_test, model.predict(X_test)))
      return model

      def predict_model(X_test,y_test):
          #evaluate the model :: socre[1] contains the accuracy for the given data ...
          score = model.evaluate(X_test,Y_test)
          d["test_acc"].append(score[1])
          return model

```

### 0.0.4 LSTM with layer = 1 , n\_hidden = 32 and dropout = [0.5]

```

[0]: model = define_compile_model(n_hidden=[32,32,32],layer=3,
                                   dropout=[0.2,0.2,0.2])

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:541: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:4432: The name tf.random\_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:148: The name tf.placeholder\_with\_default is deprecated. Please use



tf.compat.v1.placeholder\_with\_default instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3733: calling dropout (from tensorflow.python.ops.nn\_ops) with keep\_prob is deprecated and will be removed in a future version.

Instructions for updating:

Please use `rate` instead of `keep\_prob`. Rate should be set to `rate = 1 - keep\_prob`.

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
lstm_1 (LSTM)	(None, 128, 32)	5376
batch_normalization_1 (Batch Normalization)	(None, 128, 32)	128
dropout_1 (Dropout)	(None, 128, 32)	0
lstm_2 (LSTM)	(None, 128, 32)	8320
batch_normalization_2 (Batch Normalization)	(None, 128, 32)	128
dropout_2 (Dropout)	(None, 128, 32)	0
lstm_3 (LSTM)	(None, 32)	8320
batch_normalization_3 (Batch Normalization)	(None, 32)	128
dropout_3 (Dropout)	(None, 32)	0
dense_1 (Dense)	(None, 6)	198

Total params: 22,598

Trainable params: 22,406

Non-trainable params: 192

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3576: The name tf.log is deprecated. Please use tf.math.log instead.

```
[47]: model = fit_model(X_train,Y_train,batch_size,X_test,Y_test,15,model)
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow\_core/python/ops/math\_grad.py:1424: where (from tensorflow.python.ops.array\_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.where in 2.0, which has the same broadcast rule as np.where

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:1033: The name tf.assign\_add is deprecated. Please use tf.compat.v1.assign\_add instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:1020: The name tf.assign is deprecated. Please use tf.compat.v1.assign instead.

Train on 7352 samples, validate on 2947 samples

Epoch 1/15

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:190: The name tf.get\_default\_session is deprecated. Please use tf.compat.v1.get\_default\_session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:207: The name tf.global\_variables is deprecated. Please use tf.compat.v1.global\_variables instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:216: The name tf.is\_variable\_initialized is deprecated. Please use tf.compat.v1.is\_variable\_initialized instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:223: The name tf.variables\_initializer is deprecated. Please use tf.compat.v1.variables\_initializer instead.

7352/7352 [=====] - 18s 2ms/step - loss: 0.4835 - acc: 0.8032 - val\_loss: 0.5491 - val\_acc: 0.8310

Epoch 2/15

7352/7352 [=====] - 5s 626us/step - loss: 0.1771 - acc: 0.9338 - val\_loss: 0.4791 - val\_acc: 0.8500

Epoch 3/15

7352/7352 [=====] - 5s 629us/step - loss: 0.1387 - acc: 0.9441 - val\_loss: 0.3689 - val\_acc: 0.8945

Epoch 4/15

7352/7352 [=====] - 5s 643us/step - loss: 0.1294 - acc: 0.9472 - val\_loss: 0.6042 - val\_acc: 0.8551

Epoch 5/15

7352/7352 [=====] - 5s 638us/step - loss: 0.1340 - acc:

```

0.9506 - val_loss: 0.4288 - val_acc: 0.8948
Epoch 6/15
7352/7352 [=====] - 5s 633us/step - loss: 0.1163 - acc:
0.9508 - val_loss: 0.4495 - val_acc: 0.8877
Epoch 7/15
7352/7352 [=====] - 5s 635us/step - loss: 0.1135 - acc:
0.9525 - val_loss: 0.4045 - val_acc: 0.8924
Epoch 8/15
7352/7352 [=====] - 5s 628us/step - loss: 0.1151 - acc:
0.9514 - val_loss: 0.4384 - val_acc: 0.8928
Epoch 9/15
7352/7352 [=====] - 5s 640us/step - loss: 0.1090 - acc:
0.9520 - val_loss: 0.4849 - val_acc: 0.8985
Epoch 10/15
7352/7352 [=====] - 5s 649us/step - loss: 0.0986 - acc:
0.9563 - val_loss: 0.4449 - val_acc: 0.9046
Epoch 11/15
7352/7352 [=====] - 5s 633us/step - loss: 0.1029 - acc:
0.9570 - val_loss: 0.5560 - val_acc: 0.8768
Epoch 12/15
7352/7352 [=====] - 5s 640us/step - loss: 0.1087 - acc:
0.9548 - val_loss: 0.3356 - val_acc: 0.9080
Epoch 13/15
7352/7352 [=====] - 5s 645us/step - loss: 0.0956 - acc:
0.9567 - val_loss: 0.5069 - val_acc: 0.9013
Epoch 14/15
7352/7352 [=====] - 5s 646us/step - loss: 0.1087 - acc:
0.9548 - val_loss: 0.5025 - val_acc: 0.8873
Epoch 15/15
7352/7352 [=====] - 5s 643us/step - loss: 0.1041 - acc:
0.9550 - val_loss: 0.4978 - val_acc: 0.9013
Pred          LAYING  SITTING  ...  WALKING_DOWNSTAIRS  WALKING_UPSTAIRS
True
LAYING          510         0  ...                0                1
SITTING          0        412  ...                0                7
STANDING         0         89  ...                0                0
WALKING          0         0  ...               35                7
WALKING_DOWNSTAIRS  0         0  ...             401                7
WALKING_UPSTAIRS   0         0  ...             28               436

```

[6 rows x 6 columns]

```
[49]: model = predict_model(X_test,Y_test)
```

```
2947/2947 [=====] - 0s 87us/step
```

## 0.0.5 printing d value

```
[0]: print(d)
```

```
{'layers': [], 'n_hidden': [], 'dropout': [], 'parameters_trained': [302982, 302982], 'train_acc': [0.9606909684439608, 0.9639553862894451], 'test_acc': [0.9104173736002714, 0.8730912792670512]}
```

```
[0]: print(d)
```

```
{'layers': [], 'n_hidden': [], 'dropout': [], 'parameters_trained': [249026, 238274, 238274], 'train_acc': [0.9560663764961915, 0.9575625680087051, 0.9672198041349293], 'test_acc': [0.9121140142517815, 0.9243298269426535, 0.8598574821852731]}
```

- With a simple 2 layer architecture we got 90.09% accuracy and a loss of 0.30
- We can further improve the performance with Hyperparameter tuning

```
[0]: #####
# hyper parameter tuning
# hidden units , dropout, lstm layers (but be careful that it will not overfit)
→(use different dropout rate .. )
# use pretty table to conclude epochs can also be changed ..
# pretty table columns - layers , n_hidden, dropout used for each layer,
→train_Acc, test_Acc
#####

from prettytable import PrettyTable
x = PrettyTable()
x.field_names = ["layers", "n_hidden", "dropout",
→"parameters_trained", "train_acc", "test_acc"]
for i in range(len(d["layers"])):
    x.
→add_row([d["layers"][i], d["n_hidden"][i], d["dropout"][i], d["parameters_trained"][i], d["train_
print(x)
```

```
+-----+-----+-----+-----+-----+
+-----+
| layers | n_hidden | dropout | parameters_trained | train_acc |
| test_acc |
+-----+-----+-----+-----+-----+
+-----+
| 1 | 32 | [0.5] | 5574 | 0.9434167573449401 |
| 0.8842891075670173 |
| 2 | 32 | [0.5, 0.5] | 13894 | 0.9473612622415669 |
| 0.8934509670851714 |
| 3 | 64 | [0.3, 0.3, 0.3] | 85382 | 0.9533460282916213 |
```

```
| 0.9131319986426875 |
```

```
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
```

```
[0]: from prettytable import PrettyTable
x = PrettyTable()
x.field_names = ["layers", "n_hidden", "dropout",
    ↳ "parameters_trained", "train_acc", "test_acc"]
for i in range(len(d["layers"])):
    x.
    ↳ add_row([d["layers"][i], d["n_hidden"][i], d["dropout"][i], d["parameters_trained"][i], d["train_acc"][i], d["test_acc"][i]])
print(x)
```

```
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
| layers |      n_hidden      | dropout | parameters_trained |
train_acc |      test_acc      |         |                    |
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
| 1 | [32] | [0.5] | 5574 |
0.9434167573449401 | 0.8842891075670173 |
| 1 | [64] | [0.5] | 19462 |
0.9480413492927094 | 0.8842891075670173 |
| 1 | [64] | [0.3] | 19462 |
0.9500816104461371 | 0.9267051238547676 |
| 1 | [128] | [0.5] | 71686 |
0.6228237214363439 | 0.168306752629793 |
| 1 | [128] | [0.5] | 71686 |
0.9465451577801959 | 0.9280624363759755 |
| 2 | [32, 32] | [0.4, 0.3] | 14022 |
0.9506256800870512 | 0.9287410926365796 |
| 1 | [256] | [0.5] | 274438 |
0.9396082698585418 | 0.8992195453003053 |
| 2 | [64, 32] | [0.5, 0.4] | 31750 |
0.9476332970620239 | 0.9124533423820834 |
| 2 | [64, 64] | [0.5, 0.5] | 52614 |
0.9426006528835691 | 0.9267051238547676 |
| 3 | [32, 32, 32] | [0.5, 0.4, 0.3] | 22406 |
0.9472252448313384 | 0.9178825924669155 |
| 3 | [64, 32, 16] | [0.5, 0.3, 0.1] | 34822 |
0.9430087051142546 | 0.9243298269426535 |
| 4 | [32, 32, 32, 32] | [0.4, 0.4, 0.4, 0.4] | 30790 |
0.9367519042437432 | 0.8948082796063793 |
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
```

```
[0]: from prettytable import PrettyTable
x = PrettyTable()
x.field_names = ["layers", "n_hidden", "dropout", "parameters_trained", "train_acc", "test_acc"]
for i in range(len(d["layers"])):
    x.add_row([d["layers"][i], d["n_hidden"][i], d["dropout"][i], d["parameters_trained"][i], d["train_acc"][i], d["test_acc"][i]])
print(x)
```

```
+-----+-----+-----+-----+-----+
| layers | n_hidden | dropout | parameters_trained | train_acc | test_acc |
+-----+-----+-----+-----+-----+
| 3      | [32, 32, 32] | [0.2, 0.2, 0.2] | 22406 | 0.9413764961915125 | 0.8988802171700034 |
+-----+-----+-----+-----+-----+
```

```
[0]: table = PrettyTable(['layers', 'n_hidden', 'dropout', 'parameters_trained', 'train_acc', 'test_acc'])
```

```
[0]: table.add_row([4, [100, 100, 6], [0.5, 0.5], 54706, 0.9163492927094669, 0.9066847641669494])
table.add_row([3, [64, 100, 6], [0.5], 249026, 0.9560663764961915, 0.9121140142517815])
table.add_row([7, [64, 32, 100, 100, 6], [0.5, 0.5], 238274, 0.9572198041349293, 0.9298574821852731])
table.add_row([7, [64, 32, 100, 100, 6], [0.5, 0.5], 238274, 0.9675625680087051, 0.9443298269426535])
```

```
[0]: print(table)
```

```
+-----+-----+-----+-----+-----+
| layers | n_hidden | dropout | parameters_trained | train_acc | test_acc |
+-----+-----+-----+-----+-----+
| 4      | [100, 100, 6] | [0.5, 0.5] | 54706 | 0.9163492927094669 | 0.9066847641669494 |
| 3      | [64, 100, 6] | [0.5] | 249026 | 0.9560663764961915 | 0.9121140142517815 |
| 7      | [64, 32, 100, 100, 6] | [0.5, 0.5] | 238274 | 0.9572198041349294 | 0.9298574821852731 |
| 7      | [64, 32, 100, 100, 6] | [0.5, 0.5] | 238274 | 0.9675625680087051 | 0.9443298269426535 |
```

+-----+-----+-----+-----+-----+  
-----+-----+-----+-----+-----+

```
[0]: # Conclusions

# We clearly see that the model performs fairly well on shallow LSTM layers
# but as we increase the depth of the model architecture it suffers overfitting

# Also, we experimented by adding large values of dropouts, and reducing the
# number of training epochs but still the deep networks are not getting any
# better substantially

# The best model having two LSTM layers each having 32 hidden units
```

```
[0]: # At last we experimented with 2 other kinds of custom models instead of
# vanilla LSTMs's. They were CNN-LSTM and ConvLSTM. We were able to achieve our
# target accuracy of >94% using the CNN-LSTM model after playing around with
# the hyperparams and number of epochs to avoid overfitting
```

```
[0]:
```

```
[0]: # Trying out a completely new Github implementation
```

```
[0]: # Using code directly from :
# https://github.com/heeryoncho/sensors2018cnnhar/
```

```
[0]: from scipy import ndimage
```

```
[0]: dir_path = dir_path+'UCI_HAR_Dataset/'
```

```
[11]: print(dir_path)
```

/content/drive/My Drive/Colab  
Notebooks/AppliedAI/HumanActivityRecognition/HAR/UCI\_HAR\_Dataset/

```
[0]: def load_x(train_or_test):
    global dir_path
    if train_or_test is "train":
        x_path = dir_path + 'train/X_train.txt'
    elif train_or_test is "test":
        x_path = dir_path + 'test/X_test.txt'

    with open(x_path) as f:
        container = f.readlines()

    result = []
    for line in container:
        tmp1 = line.strip()
        tmp2 = tmp1.replace(' ', ' ') # removes inconsistent blank spaces
        tmp_ary = list(map(float, tmp2.split(' ')))
        result.append(tmp_ary)
```

```
return np.array(result)
```

```
[0]: def load_y(train_or_test):  
    global dir_path  
    if train_or_test is "train":  
        y_path = dir_path + 'train/y_train.txt'  
    elif train_or_test is "test":  
        y_path = dir_path + 'test/y_test.txt'  
  
    with open(y_path) as f:  
        container = f.readlines()  
  
    result = []  
    for line in container:  
        num_str = line.strip()  
        result.append(int(num_str))  
    return np.array(result)
```

```
[0]: import os  
os.environ['TF_CPP_MIN_LOG_LEVEL']='2'  
import numpy as np  
import random  
from numpy.random import seed  
from sklearn.metrics import accuracy_score, confusion_matrix  
from keras.optimizers import Adam  
from keras.models import Sequential, load_model  
from keras.layers import Conv1D, MaxPooling1D, Dense, Flatten, Dropout  
from keras.callbacks import ModelCheckpoint  
from keras.utils import plot_model  
import keras.backend as K
```

```
[0]: X_train_all = load_x("train")  
y_train_all = load_y("train")  
  
X_test_all = load_x("test")  
y_test_all = load_y("test")
```

```
[31]: # -----  
# Only dynamic HAR data are selected  
# -----  
  
# Select dynamic HAR train data  
  
dynamic_1 = np.where(y_train_all == 1)[0]  
dynamic_2 = np.where(y_train_all == 2)[0]  
dynamic_3 = np.where(y_train_all == 3)[0]  
dynamic = np.concatenate([dynamic_1, dynamic_2, dynamic_3])  
  
X_train = X_train_all[dynamic]
```



```

y_train = y_train_all[dynamic]

# Convert (1, 2, 3) labels to (0, 1, 2)
y_train = y_train - 1

print("\n+++ DATA STATISTICS +++\n")
print("train_dynamic shape: ", X_train.shape)

# Select dynamic HAR test data

dynamic_1 = np.where(y_test_all == 1)[0]
dynamic_2 = np.where(y_test_all == 2)[0]
dynamic_3 = np.where(y_test_all == 3)[0]
dynamic = np.concatenate([dynamic_1, dynamic_2, dynamic_3])

X_test = X_test_all[dynamic]
y_test = y_test_all[dynamic]

# Convert (1, 2, 3) labels to (0, 1, 2)
y_test = y_test - 1

print("test_dynamic shape: ", X_test.shape)

```

+++ DATA STATISTICS +++

```

train_dynamic shape: (3285, 561)
test_dynamic shape: (1387, 561)

```

[32]: dir\_path

[32]: '/content/drive/My Drive/Colab  
Notebooks/AppliedAI/HumanActivityRecognition/HAR/UCI\_HAR\_Dataset/'

[35]: # Display dynamic model accuracy

```

print("\n+++ DYNAMIC MODEL ACCURACY (See Table 8 in paper) +++\n")

model_path = dir_path+"models/dynamic.hdf5"
model = load_model(model_path)

pred_train = model.predict(np.expand_dims(X_train, axis=2), batch_size=32)
print("----- TRAIN ACCURACY -----")
print(accuracy_score(y_train, np.argmax(pred_train, axis=1)))
print(confusion_matrix(y_train, np.argmax(pred_train, axis=1)))

pred_test = model.predict(np.expand_dims(X_test, axis=2), batch_size=32)
print("----- TEST ACCURACY -----")

```

```
print(accuracy_score(y_test, np.argmax(pred_test, axis=1)))
print(confusion_matrix(y_test, np.argmax(pred_test, axis=1)))
```

+++ DYNAMIC MODEL ACCURACY (See Table 8 in paper) +++

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:66: The name tf.get\_default\_graph is deprecated. Please use tf.compat.v1.get\_default\_graph instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:541: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:4432: The name tf.random\_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:4267: The name tf.nn.max\_pool is deprecated. Please use tf.nn.max\_pool2d instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:148: The name tf.placeholder\_with\_default is deprecated. Please use tf.compat.v1.placeholder\_with\_default instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3733: calling dropout (from tensorflow.python.ops.nn\_ops) with keep\_prob is deprecated and will be removed in a future version.

Instructions for updating:

Please use `rate` instead of `keep\_prob`. Rate should be set to `rate = 1 - keep\_prob`.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:190: The name tf.get\_default\_session is deprecated. Please use tf.compat.v1.get\_default\_session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:197: The name tf.ConfigProto is deprecated. Please use tf.compat.v1.ConfigProto instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:203: The name tf.Session is deprecated. Please use tf.compat.v1.Session instead.

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:207: The name tf.global_variables is deprecated. Please use tf.compat.v1.global_variables instead.
```

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:216: The name tf.is_variable_initialized is deprecated. Please use tf.compat.v1.is_variable_initialized instead.
```

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:223: The name tf.variables_initializer is deprecated. Please use tf.compat.v1.variables_initializer instead.
```

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.
```

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1033: The name tf.assign_add is deprecated. Please use tf.compat.v1.assign_add instead.
```

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1020: The name tf.assign is deprecated. Please use tf.compat.v1.assign instead.
```

```
/usr/local/lib/python3.6/dist-packages/keras/engine/saving.py:350: UserWarning: Error in loading the saved optimizer state. As a result, your model is starting with a freshly initialized optimizer.
```

```
warnings.warn('Error in loading the saved optimizer ')
```

```
----- TRAIN ACCURACY -----
```

```
0.9863013698630136
```

```
[[1223   3   0]
 [   4 1038  31]
 [   0   7 979]]
```

```
----- TEST ACCURACY -----
```

```
0.9798125450612833
```

```
[[495   0   1]
 [   2 469   0]
 [   2  23 395]]
```

```
[38]: static_1 = np.where(y_train_all == 4)[0]
static_2 = np.where(y_train_all == 5)[0]
static_3 = np.where(y_train_all == 6)[0]
static = np.concatenate([static_1, static_2, static_3])
```

```

X_train = X_train_all[static]
y_train = y_train_all[static]

# Convert (4, 5, 6) labels to (0, 1, 2)
y_train = y_train - 4

print("\n+++ DATA STATISTICS +++\n")
print("train_static shape: ", X_train.shape)

# Select static HAR test data

static_1 = np.where(y_test_all == 4)[0]
static_2 = np.where(y_test_all == 5)[0]
static_3 = np.where(y_test_all == 6)[0]
static = np.concatenate([static_1, static_2, static_3])

X_test = X_test_all[static]
y_test = y_test_all[static]

# Convert (4, 5, 6) labels to (0, 1, 2)
y_test = y_test - 4

print("test_static shape: ", X_test.shape)

# Display static model accuracy

print("\n+++ STATIC MODEL ACCURACY (See Table 8 in paper) +++\n")

model_path = dir_path+"models/static.hdf5"
model = load_model(model_path)

pred_train = model.predict(np.expand_dims(X_train, axis=2), batch_size=32)
print("----- TRAIN ACCURACY -----")
print(accuracy_score(y_train, np.argmax(pred_train, axis=1)))
print(confusion_matrix(y_train, np.argmax(pred_train, axis=1)))

pred_test = model.predict(np.expand_dims(X_test, axis=2), batch_size=32)
print("----- TEST ACCURACY -----")
print(accuracy_score(y_test, np.argmax(pred_test, axis=1)))
print(confusion_matrix(y_test, np.argmax(pred_test, axis=1)))

```

+++ DATA STATISTICS +++

train\_static shape: (4067, 561)  
test\_static shape: (1560, 561)

```
+++ STATIC MODEL ACCURACY (See Table 8 in paper) +++
```

```
/usr/local/lib/python3.6/dist-packages/keras/engine/saving.py:350: UserWarning:  
Error in loading the saved optimizer state. As a result, your model is starting  
with a freshly initialized optimizer.  
warnings.warn('Error in loading the saved optimizer ')
```

```
----- TRAIN ACCURACY -----
```

```
0.9923776739611507
```

```
[[1275  11   0]
```

```
 [ 20 1354   0]
```

```
 [  0   0 1407]]
```

```
----- TEST ACCURACY -----
```

```
0.966025641025641
```

```
[[453  38   0]
```

```
 [ 14 518   0]
```

```
 [  1   0 536]]
```

[0]: