

```
In [1]: # Importing Libraries
```

```
In [2]: import pandas as pd
import numpy as np
```

```
In [3]: # 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'])
```

Data

```
In [4]: # Data directory
DATADIR = 'UCI_HAR_Dataset'
```

```
In [5]: # 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"
]
```

```
In [6]: # 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'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))
```

```
In [7]: 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'UCI_HAR_Dataset/{subset}/y_{subset}.txt'
        y = _read_csv(filename)[0]

        return pd.get_dummies(y).as_matrix()
```

```
In [8]: 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
```

```
In [9]: # Importing tensorflow
np.random.seed(42)
import tensorflow as tf
tf.set_random_seed(42)
```

```
In [10]: # Configuring a session
session_conf = tf.ConfigProto(
    intra_op_parallelism_threads=1,
    inter_op_parallelism_threads=1
)
```

```
In [11]: # Import Keras
from keras import backend as K
sess = tf.Session(graph=tf.get_default_graph(), config=session_conf)
K.set_session(sess)
```

```
In [12]: # Importing libraries
from keras.models import Sequential
from keras.layers import LSTM
from keras.layers.core import Dense, Dropout
```

```
In [13]: # Initializing parameters
epochs = 30
batch_size = 16
n_hidden = 32
```

```
In [14]: # Utility function to count the number of classes
def _count_classes(y):
    return len(set([tuple(category) for category in y]))
```

```
In [15]: # Loading the train and test data
X_train, X_test, Y_train, Y_test = load_data()
```

C:\Users\LENOVO\Anaconda3\lib\site-packages\ipykernel_launcher.py:12: FutureWarning: Method .as_matrix will be removed in a future version. Use .values in

stead.

if sys.path[0] == '':

```
In [16]: len(X_train[0][0])
```

Out[16]: 9

```
In [17]: 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

- Defining the Architecture of LSTM

```
In [26]: # Initiliazing the sequential model
model = Sequential()
# Configuring the parameters
model.add(LSTM(n_hidden, input_shape=(timesteps, input_dim)))
# Adding a dropout Layer
model.add(Dropout(0.5))
# Adding a dense output Layer with sigmoid activation
model.add(Dense(n_classes, activation='sigmoid'))
model.summary()
```

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:541: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:4432: The name tf.random_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-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 C:\Users\LENOVO\Anaconda3\lib\site-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_2"

Layer (type)	Output Shape	Param #
lstm_1 (LSTM)	(None, 32)	5376
dropout_1 (Dropout)	(None, 32)	0
dense_1 (Dense)	(None, 6)	198
Total params: 5,574		
Trainable params: 5,574		
Non-trainable params: 0		

```
In [27]: # Compiling the model
model.compile(loss='categorical_crossentropy',
              optimizer='rmsprop',
              metrics=['accuracy'])
```

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:3576: The name tf.log is deprecated. Please use tf.math.log instead.

```
In [1]: # Training the model
model.fit(X_train,
          Y_train,
          batch_size=batch_size,
          validation_data=(X_test, Y_test),
          epochs=epochs)
```

```
In [24]: # Confusion Matrix
print(confusion_matrix(Y_test, model.predict(X_test)))
```

Pred \ True	LAYING	SITTING	STANDING	WALKING	WALKING_DOWNSTAIRS	WALKING_UPSTAIRS
LAYING	512	0	25	0	0	0
SITTING	3	410	75	0	0	0
STANDING	0	87	445	0	0	0
WALKING	0	0	0	481	2	2
WALKING_DOWNSTAIRS	0	0	0	0	382	0
WALKING_UPSTAIRS	0	0	0	2	18	451

Pred \ True	WALKING_UPSTAIRS
LAYING	0
SITTING	3
STANDING	0
WALKING	13
WALKING_DOWNSTAIRS	38
WALKING_UPSTAIRS	451

```
In [27]: score = model.evaluate(X_test, Y_test)
```

2947/2947 [=====] - 4s 2ms/step

```
In [28]: score
```

```
Out[28]: [0.3087582236972612, 0.9097387173396675]
```

- 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

Assignment:

Model-1:

```
In [17]: # Initiliazing the sequential model
model = Sequential()
# Configuring the parameters
model.add(LSTM(40, input_shape=(timesteps, input_dim)))
# Adding a dropout Layer
model.add(Dropout(0.5))
# Adding a dense output layer with sigmoid activation
model.add(Dense(n_classes, activation='sigmoid'))
model.summary()
```

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-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 C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:541: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:4432: The name tf.random_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-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 C:\Users\LENOVO\Anaconda3\lib\site-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, 40)	8000
dropout_1 (Dropout)	(None, 40)	0
dense_1 (Dense)	(None, 6)	246
Total params: 8,246		
Trainable params: 8,246		
Non-trainable params: 0		

```
In [18]: # Compiling the model
model.compile(loss='categorical_crossentropy',
              optimizer='rmsprop',
              metrics=['accuracy'])
```

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From C:\Users\LENOVO\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:3576: The name tf.log is deprecated. Please use tf.math.log instead.

```
In [19]: # Training the model
model.fit(X_train,
          Y_train,
          batch_size=batch_size,
          validation_data=(X_test, Y_test),
          epochs=epochs)
```


WARNING:tensorflow:From C:\Users\LENOVO\AppData\Roaming\Python\Python36\site-packages\tensorflow\python\ops\math_grad.py:1250: add_dispatch_support.<locals>.wrapper (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

Train on 7352 samples, validate on 2947 samples

Epoch 1/30

7352/7352 [=====] - 123s 17ms/step - loss: 1.2862 - acc: 0.4539 - val_loss: 1.1677 - val_acc: 0.5059

Epoch 2/30

7352/7352 [=====] - 121s 16ms/step - loss: 1.1326 - acc: 0.5135 - val_loss: 1.0953 - val_acc: 0.5161

Epoch 3/30

7352/7352 [=====] - 118s 16ms/step - loss: 0.9043 - acc: 0.6050 - val_loss: 0.9573 - val_acc: 0.6050

Epoch 4/30

7352/7352 [=====] - 124s 17ms/step - loss: 0.7925 - acc: 0.6255 - val_loss: 1.1005 - val_acc: 0.5249

Epoch 5/30

7352/7352 [=====] - 131s 18ms/step - loss: 0.7313 - acc: 0.6461 - val_loss: 0.8016 - val_acc: 0.6237

Epoch 6/30

7352/7352 [=====] - 132s 18ms/step - loss: 0.7000 - acc: 0.6665 - val_loss: 0.9460 - val_acc: 0.5989

Epoch 7/30

7352/7352 [=====] - 131s 18ms/step - loss: 0.7608 - acc: 0.6526 - val_loss: 0.7669 - val_acc: 0.6410

Epoch 8/30

7352/7352 [=====] - 131s 18ms/step - loss: 0.6546 - acc: 0.7021 - val_loss: 0.7351 - val_acc: 0.6919

Epoch 9/30

7352/7352 [=====] - 132s 18ms/step - loss: 0.6057 - acc: 0.7602 - val_loss: 0.7004 - val_acc: 0.7750

Epoch 10/30

7352/7352 [=====] - 144s 20ms/step - loss: 0.5372 - acc: 0.7988 - val_loss: 0.8634 - val_acc: 0.7027

Epoch 11/30

7352/7352 [=====] - 139s 19ms/step - loss: 0.4524 - acc: 0.8482 - val_loss: 0.6696 - val_acc: 0.8154

Epoch 12/30

7352/7352 [=====] - 138s 19ms/step - loss: 0.3766 - acc: 0.8821 - val_loss: 0.6002 - val_acc: 0.8402

Epoch 13/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.3165 - acc: 0.9036 - val_loss: 0.5824 - val_acc: 0.8422

Epoch 14/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.2830 - acc: 0.9097 - val_loss: 0.5283 - val_acc: 0.8514

Epoch 15/30

7352/7352 [=====] - 138s 19ms/step - loss: 0.2437 - acc: 0.9196 - val_loss: 0.5322 - val_acc: 0.8643

Epoch 16/30

7352/7352 [=====] - 144s 20ms/step - loss: 0.2347 - acc: 0.9263 - val_loss: 0.4037 - val_acc: 0.8870

Epoch 17/30

7352/7352 [=====] - 137s 19ms/step - loss: 0.2268 -

```
acc: 0.9280 - val_loss: 0.5046 - val_acc: 0.8819
Epoch 18/30
7352/7352 [=====] - 141s 19ms/step - loss: 0.1989 -
acc: 0.9358 - val_loss: 0.5478 - val_acc: 0.8897
Epoch 19/30
7352/7352 [=====] - 136s 19ms/step - loss: 0.1892 -
acc: 0.9361 - val_loss: 0.4753 - val_acc: 0.8850
Epoch 20/30
7352/7352 [=====] - 134s 18ms/step - loss: 0.1910 -
acc: 0.9372 - val_loss: 0.5522 - val_acc: 0.8812
Epoch 21/30
7352/7352 [=====] - 135s 18ms/step - loss: 0.2002 -
acc: 0.9354 - val_loss: 0.6198 - val_acc: 0.8772
Epoch 22/30
7352/7352 [=====] - 136s 19ms/step - loss: 0.1783 -
acc: 0.9387 - val_loss: 0.5255 - val_acc: 0.8839
Epoch 23/30
7352/7352 [=====] - 136s 19ms/step - loss: 0.1737 -
acc: 0.9361 - val_loss: 0.5526 - val_acc: 0.8877
Epoch 24/30
7352/7352 [=====] - 137s 19ms/step - loss: 0.1692 -
acc: 0.9410 - val_loss: 0.9415 - val_acc: 0.6393
Epoch 25/30
7352/7352 [=====] - 135s 18ms/step - loss: 0.1814 -
acc: 0.9361 - val_loss: 0.6644 - val_acc: 0.8731
Epoch 26/30
7352/7352 [=====] - 136s 19ms/step - loss: 0.1997 -
acc: 0.9344 - val_loss: 0.5646 - val_acc: 0.8890
Epoch 27/30
7352/7352 [=====] - 136s 18ms/step - loss: 0.1624 -
acc: 0.9459 - val_loss: 0.4422 - val_acc: 0.8931
Epoch 28/30
7352/7352 [=====] - 136s 18ms/step - loss: 0.1621 -
acc: 0.9449 - val_loss: 0.4937 - val_acc: 0.8850
Epoch 29/30
7352/7352 [=====] - 136s 18ms/step - loss: 0.1584 -
acc: 0.9470 - val_loss: 0.5160 - val_acc: 0.8992
Epoch 30/30
7352/7352 [=====] - 136s 18ms/step - loss: 0.1656 -
acc: 0.9440 - val_loss: 0.5032 - val_acc: 0.8894
```

Out[19]: <keras.callbacks.History at 0x1355507ed30>

```
In [20]: # Confusion Matrix
print(confusion_matrix(Y_test, model.predict(X_test)))
```

Pred \ True	LAYING	SITTING	STANDING	WALKING	WALKING_DOWNSTAIRS	WALKING_UPSTAIRS
LAYING	510	0	0	0	0	0
SITTING	0	405	62	0	0	5
STANDING	0	106	424	1	0	0
WALKING	0	0	7	406	0	27
WALKING_DOWNSTAIRS	0	0	0	0	417	0
WALKING_UPSTAIRS	0	0	0	1	0	11

Pred \ True	WALKING_UPSTAIRS
LAYING	27
SITTING	19
STANDING	1
WALKING	56
WALKING_DOWNSTAIRS	3
WALKING_UPSTAIRS	459

```
In [21]: score = model.evaluate(X_test, Y_test)
```

2947/2947 [=====] - 8s 3ms/step

```
In [22]: score
```

```
Out[22]: [0.5031543293462738, 0.8893790295215473]
```

```
In [ ]:
```

Model-2:

```
In [47]: # code from https://keras.io/regularizers/
from keras.regularizers import L1L2
from keras.models import load_model
from keras.callbacks import ModelCheckpoint
from keras.layers import LSTM, BatchNormalization
reg = L1L2(0.01, 0.01)
```

```
In [42]: model = Sequential()
model.add(LSTM(100, input_shape=(timesteps, input_dim), kernel_initializer='glorot_normal', return_sequences=True, bias_regularizer=reg))
model.add(BatchNormalization())
model.add(Dropout(0.70))
model.add(LSTM(50))
model.add(Dropout(0.70))
model.add(Dense(n_classes, activation='sigmoid'))
print("Model Summary: ")
model.summary()
```

Model Summary:

Model: "sequential_5"

Layer (type)	Output Shape	Param #
lstm_8 (LSTM)	(None, 128, 100)	44000
batch_normalization_4 (Batch Normalization)	(None, 128, 100)	400
dropout_7 (Dropout)	(None, 128, 100)	0
lstm_9 (LSTM)	(None, 50)	30200
dropout_8 (Dropout)	(None, 50)	0
dense_4 (Dense)	(None, 6)	306
Total params: 74,906		
Trainable params: 74,706		
Non-trainable params: 200		

```
In [43]: # Compiling the model
model.compile(loss='categorical_crossentropy',
              optimizer='rmsprop',
              metrics=['accuracy'])
checkpoint_3 = ModelCheckpoint("model_6.h5", monitor="val_acc", mode="max", save_best_only = True, verbose=1)
```

```
In [44]: # Training the model
model.fit(X_train,
          Y_train,
          batch_size=batch_size,
          validation_data=(X_test, Y_test),
          epochs=20, callbacks=[checkpoint_3])
```

Train on 7352 samples, validate on 2947 samples

Epoch 1/20

7352/7352 [=====] - 248s 34ms/step - loss: 2.3953 -
acc: 0.5909 - val_loss: 1.8404 - val_acc: 0.5999

Epoch 00001: val_acc improved from -inf to 0.59993, saving model to model_6.h5

Epoch 2/20

7352/7352 [=====] - 235s 32ms/step - loss: 1.1531 -
acc: 0.6884 - val_loss: 1.0318 - val_acc: 0.5999

Epoch 00002: val_acc did not improve from 0.59993

Epoch 3/20

7352/7352 [=====] - 232s 32ms/step - loss: 0.5914 -
acc: 0.7877 - val_loss: 0.4268 - val_acc: 0.8660

Epoch 00003: val_acc improved from 0.59993 to 0.86597, saving model to model_6.h5

Epoch 4/20

7352/7352 [=====] - 232s 32ms/step - loss: 0.4100 -
acc: 0.8750 - val_loss: 0.3189 - val_acc: 0.8968

Epoch 00004: val_acc improved from 0.86597 to 0.89684, saving model to model_6.h5

Epoch 5/20

7352/7352 [=====] - 234s 32ms/step - loss: 0.3255 -
acc: 0.9000 - val_loss: 0.3291 - val_acc: 0.8931

Epoch 00005: val_acc did not improve from 0.89684

Epoch 6/20

7352/7352 [=====] - 237s 32ms/step - loss: 0.2785 -
acc: 0.9135 - val_loss: 0.3620 - val_acc: 0.8850

Epoch 00006: val_acc did not improve from 0.89684

Epoch 7/20

7352/7352 [=====] - 245s 33ms/step - loss: 0.2566 -
acc: 0.9221 - val_loss: 0.3146 - val_acc: 0.8921

Epoch 00007: val_acc did not improve from 0.89684

Epoch 8/20

7352/7352 [=====] - 237s 32ms/step - loss: 0.2339 -
acc: 0.9280 - val_loss: 0.5996 - val_acc: 0.8728

Epoch 00008: val_acc did not improve from 0.89684

Epoch 9/20

7352/7352 [=====] - 236s 32ms/step - loss: 0.2438 -
acc: 0.9244 - val_loss: 0.2507 - val_acc: 0.9209

Epoch 00009: val_acc improved from 0.89684 to 0.92094, saving model to model_6.h5

Epoch 10/20

7352/7352 [=====] - 236s 32ms/step - loss: 0.2304 -
acc: 0.9256 - val_loss: 0.3254 - val_acc: 0.8979

Epoch 00010: val_acc did not improve from 0.92094

Epoch 11/20

7352/7352 [=====] - 235s 32ms/step - loss: 0.2392 -

acc: 0.9244 - val_loss: 0.4724 - val_acc: 0.8558

Epoch 00011: val_acc did not improve from 0.92094

Epoch 12/20

7352/7352 [=====] - 242s 33ms/step - loss: 0.2030 -
acc: 0.9314 - val_loss: 0.3208 - val_acc: 0.9186

Epoch 00012: val_acc did not improve from 0.92094

Epoch 13/20

7352/7352 [=====] - 246s 33ms/step - loss: 0.2041 -
acc: 0.9369 - val_loss: 0.2538 - val_acc: 0.9233

Epoch 00013: val_acc improved from 0.92094 to 0.92331, saving model to model_6.h5

Epoch 14/20

7352/7352 [=====] - 236s 32ms/step - loss: 0.2126 -
acc: 0.9310 - val_loss: 0.2945 - val_acc: 0.8951

Epoch 00014: val_acc did not improve from 0.92331

Epoch 15/20

7352/7352 [=====] - 237s 32ms/step - loss: 0.2042 -
acc: 0.9306 - val_loss: 0.2545 - val_acc: 0.9338

Epoch 00015: val_acc improved from 0.92331 to 0.93383, saving model to model_6.h5

Epoch 16/20

7352/7352 [=====] - 235s 32ms/step - loss: 0.2000 -
acc: 0.9339 - val_loss: 0.2633 - val_acc: 0.9179

Epoch 00016: val_acc did not improve from 0.93383

Epoch 17/20

7352/7352 [=====] - 235s 32ms/step - loss: 0.1971 -
acc: 0.9334 - val_loss: 0.3233 - val_acc: 0.9203

Epoch 00017: val_acc did not improve from 0.93383

Epoch 18/20

7352/7352 [=====] - 235s 32ms/step - loss: nan - ac
c: 0.3274 - val_loss: nan - val_acc: 0.1683

Epoch 00018: val_acc did not improve from 0.93383

Epoch 19/20

7352/7352 [=====] - 235s 32ms/step - loss: nan - ac
c: 0.1668 - val_loss: nan - val_acc: 0.1683

Epoch 00019: val_acc did not improve from 0.93383

Epoch 20/20

7352/7352 [=====] - 235s 32ms/step - loss: nan - ac
c: 0.1668 - val_loss: nan - val_acc: 0.1683

Epoch 00020: val_acc did not improve from 0.93383

Out[44]: <keras.callbacks.History at 0x15ca245cf28>

In [48]: model = load_model('model_6.h5')

```
In [49]: # Confusion Matrix
print(confusion_matrix(Y_test, model.predict(X_test)))
```

Pred \ True	LAYING	SITTING	STANDING	WALKING	WALKING_DOWNSTAIRS	WALKING_UPSTAIRS
LAYING	537	0	0	0	0	0
SITTING	0	389	98	0	0	0
STANDING	0	47	482	3	0	0
WALKING	0	0	0	475	20	0
WALKING_DOWNSTAIRS	0	0	0	0	420	0
WALKING_UPSTAIRS	0	0	0	8	0	14

Pred \ True	WALKING_UPSTAIRS
LAYING	0
SITTING	4
STANDING	0
WALKING	1
WALKING_DOWNSTAIRS	0
WALKING_UPSTAIRS	449

```
In [50]: score = model.evaluate(X_test, Y_test)
```

2947/2947 [=====] - 17s 6ms/step

```
In [51]: score
```

```
Out[51]: [0.2544564464753497, 0.9338310145911096]
```

Model-3:


```
In [32]: # Initiliazing the sequential model
model = Sequential()
# Configuring the parameters
model.add(LSTM(70, input_shape=(timesteps, input_dim)))
# Adding a dropout Layer
model.add(Dropout(0.7))
# Adding a dense output layer with sigmoid activation
model.add(Dense(n_classes, activation='sigmoid'))
model.summary()
```

WARNING:tensorflow:Large dropout rate: 0.7 (>0.5). In TensorFlow 2.x, dropout () uses dropout rate instead of keep_prob. Please ensure that this is intended.

Model: "sequential_4"

Layer (type)	Output Shape	Param #
=====		
lstm_4 (LSTM)	(None, 70)	22400

dropout_4 (Dropout)	(None, 70)	0

dense_4 (Dense)	(None, 6)	426
=====		
Total params: 22,826		
Trainable params: 22,826		
Non-trainable params: 0		

```
In [33]: # Compiling the model
model.compile(loss='categorical_crossentropy',
              optimizer='rmsprop',
              metrics=['accuracy'])
```

```
In [34]: # Training the model  
model.fit(X_train,  
          Y_train,  
          batch_size=batch_size,  
          validation_data=(X_test, Y_test),  
          epochs=epochs)
```

Train on 7352 samples, validate on 2947 samples

Epoch 1/30

7352/7352 [=====] - 126s 17ms/step - loss: 1.2871 - acc: 0.4425 - val_loss: 1.2365 - val_acc: 0.4669

Epoch 2/30

7352/7352 [=====] - 137s 19ms/step - loss: 1.0508 - acc: 0.5423 - val_loss: 1.0076 - val_acc: 0.6088

Epoch 3/30

7352/7352 [=====] - 136s 18ms/step - loss: 0.8036 - acc: 0.6270 - val_loss: 0.8368 - val_acc: 0.5979

Epoch 4/30

7352/7352 [=====] - 134s 18ms/step - loss: 0.7237 - acc: 0.6608 - val_loss: 0.7729 - val_acc: 0.6159

Epoch 5/30

7352/7352 [=====] - 136s 18ms/step - loss: 0.9305 - acc: 0.6065 - val_loss: 0.8180 - val_acc: 0.5826

Epoch 6/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.6588 - acc: 0.7073 - val_loss: 0.6039 - val_acc: 0.7238

Epoch 7/30

7352/7352 [=====] - 136s 18ms/step - loss: 0.5586 - acc: 0.7560 - val_loss: 0.5590 - val_acc: 0.7631

Epoch 8/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.5349 - acc: 0.7851 - val_loss: 0.4933 - val_acc: 0.8368

Epoch 9/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.5117 - acc: 0.8266 - val_loss: 0.4738 - val_acc: 0.8269

Epoch 10/30

7352/7352 [=====] - 136s 18ms/step - loss: 0.3452 - acc: 0.8925 - val_loss: 0.3833 - val_acc: 0.8663

Epoch 11/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.2779 - acc: 0.9172 - val_loss: 0.3797 - val_acc: 0.8856

Epoch 12/30

7352/7352 [=====] - 141s 19ms/step - loss: 0.2701 - acc: 0.9221 - val_loss: 0.3169 - val_acc: 0.8711

Epoch 13/30

7352/7352 [=====] - 138s 19ms/step - loss: 0.2534 - acc: 0.9286 - val_loss: 0.2987 - val_acc: 0.9009

Epoch 14/30

7352/7352 [=====] - 132s 18ms/step - loss: 0.2325 - acc: 0.9309 - val_loss: 0.3455 - val_acc: 0.8951

Epoch 15/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.2166 - acc: 0.9340 - val_loss: 0.2628 - val_acc: 0.9091

Epoch 16/30

7352/7352 [=====] - 129s 18ms/step - loss: 0.2118 - acc: 0.9374 - val_loss: 0.4613 - val_acc: 0.8860

Epoch 17/30

7352/7352 [=====] - 129s 18ms/step - loss: 0.2248 - acc: 0.9343 - val_loss: 0.5067 - val_acc: 0.8755

Epoch 18/30

7352/7352 [=====] - 129s 17ms/step - loss: 0.2107 - acc: 0.9347 - val_loss: 0.2522 - val_acc: 0.9230

Epoch 19/30

7352/7352 [=====] - 128s 17ms/step - loss: 0.2013 -

```
acc: 0.9433 - val_loss: 0.2784 - val_acc: 0.9175
Epoch 20/30
7352/7352 [=====] - 151s 21ms/step - loss: 0.1917 -
acc: 0.9402 - val_loss: 0.2704 - val_acc: 0.9230
Epoch 21/30
7352/7352 [=====] - 137s 19ms/step - loss: 0.1846 -
acc: 0.9450 - val_loss: 0.2450 - val_acc: 0.9186
Epoch 22/30
7352/7352 [=====] - 130s 18ms/step - loss: 0.1932 -
acc: 0.9450 - val_loss: 0.2845 - val_acc: 0.9060
Epoch 23/30
7352/7352 [=====] - 129s 18ms/step - loss: 0.2236 -
acc: 0.9358 - val_loss: 0.3241 - val_acc: 0.9158
Epoch 24/30
7352/7352 [=====] - 130s 18ms/step - loss: 0.1734 -
acc: 0.9429 - val_loss: 0.2799 - val_acc: 0.9162
Epoch 25/30
7352/7352 [=====] - 128s 17ms/step - loss: 0.1593 -
acc: 0.9456 - val_loss: 0.3036 - val_acc: 0.9186
Epoch 26/30
7352/7352 [=====] - 128s 17ms/step - loss: 0.3904 -
acc: 0.9135 - val_loss: 0.3578 - val_acc: 0.9063
Epoch 27/30
7352/7352 [=====] - 128s 17ms/step - loss: 0.1722 -
acc: 0.9430 - val_loss: 0.6189 - val_acc: 0.8880
Epoch 28/30
7352/7352 [=====] - 132s 18ms/step - loss: 0.2481 -
acc: 0.9327 - val_loss: 0.4034 - val_acc: 0.9030
Epoch 29/30
7352/7352 [=====] - 134s 18ms/step - loss: 0.1677 -
acc: 0.9479 - val_loss: 0.4361 - val_acc: 0.9131
Epoch 30/30
7352/7352 [=====] - 130s 18ms/step - loss: 0.3103 -
acc: 0.9259 - val_loss: 0.4625 - val_acc: 0.8989
```

Out[34]: <keras.callbacks.History at 0x1361bad9630>

```
In [35]: # Confusion Matrix
print(confusion_matrix(Y_test, model.predict(X_test)))
```

Pred \ True	LAYING	SITTING	STANDING	WALKING	WALKING_DOWNSTAIRS	WALKING_UPSTAIRS
LAYING	485	13	27	1	0	9
SITTING	0	373	111	4	0	0
STANDING	0	81	450	0	0	0
WALKING	0	0	0	462	19	0
WALKING_DOWNSTAIRS	0	0	0	1	416	0
WALKING_UPSTAIRS	0	0	0	2	0	6

Pred \ True	WALKING_UPSTAIRS
LAYING	2
SITTING	3
STANDING	1
WALKING	15
WALKING_DOWNSTAIRS	3
WALKING_UPSTAIRS	463

```
In [36]: score = model.evaluate(X_test, Y_test)
```

2947/2947 [=====] - 8s 3ms/step

```
In [37]: score
```

```
Out[37]: [0.46246355840847214, 0.8988802171700034]
```

```
In [ ]:
```

```
In [ ]:
```

Model-4:

```
In [23]: # Initiliazing the sequential model
model = Sequential()
# Configuring the parameters
model.add(LSTM(50, input_shape=(timesteps, input_dim)))
# Adding a dropout Layer
model.add(Dropout(0.5))
# Adding a dense output Layer with sigmoid activation
model.add(Dense(n_classes, activation='sigmoid'))
model.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
=====		
lstm_2 (LSTM)	(None, 50)	12000

dropout_2 (Dropout)	(None, 50)	0

dense_2 (Dense)	(None, 6)	306
=====		
Total params: 12,306		
Trainable params: 12,306		
Non-trainable params: 0		

```
In [24]: # Compiling the model
model.compile(loss='categorical_crossentropy',
              optimizer='rmsprop',
              metrics=['accuracy'])
```

```
In [25]: # Training the model  
model.fit(X_train,  
          Y_train,  
          batch_size=batch_size,  
          validation_data=(X_test, Y_test),  
          epochs=epochs)
```

Train on 7352 samples, validate on 2947 samples

Epoch 1/30

7352/7352 [=====] - 125s 17ms/step - loss: 1.1955 - acc: 0.4834 - val_loss: 0.8875 - val_acc: 0.6159

Epoch 2/30

7352/7352 [=====] - 125s 17ms/step - loss: 0.8413 - acc: 0.6178 - val_loss: 0.7971 - val_acc: 0.6057

Epoch 3/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.7409 - acc: 0.6605 - val_loss: 0.7410 - val_acc: 0.6810

Epoch 4/30

7352/7352 [=====] - 134s 18ms/step - loss: 0.7369 - acc: 0.6862 - val_loss: 1.2686 - val_acc: 0.4058

Epoch 5/30

7352/7352 [=====] - 134s 18ms/step - loss: 0.5986 - acc: 0.7520 - val_loss: 0.5492 - val_acc: 0.7652

Epoch 6/30

7352/7352 [=====] - 134s 18ms/step - loss: 0.4807 - acc: 0.8075 - val_loss: 0.5612 - val_acc: 0.7727

Epoch 7/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.4437 - acc: 0.8478 - val_loss: 0.5127 - val_acc: 0.8324

Epoch 8/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.3448 - acc: 0.8921 - val_loss: 0.4319 - val_acc: 0.8504

Epoch 9/30

7352/7352 [=====] - 134s 18ms/step - loss: 0.3270 - acc: 0.9007 - val_loss: 0.4478 - val_acc: 0.8646

Epoch 10/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.2576 - acc: 0.9146 - val_loss: 0.2975 - val_acc: 0.9006

Epoch 11/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.2145 - acc: 0.9268 - val_loss: 0.3361 - val_acc: 0.9002

Epoch 12/30

7352/7352 [=====] - 138s 19ms/step - loss: 0.1917 - acc: 0.9332 - val_loss: 0.3147 - val_acc: 0.9009

Epoch 13/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.1934 - acc: 0.9340 - val_loss: 0.3482 - val_acc: 0.8890

Epoch 14/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.1755 - acc: 0.9402 - val_loss: 0.4045 - val_acc: 0.8921

Epoch 15/30

7352/7352 [=====] - 134s 18ms/step - loss: 0.1868 - acc: 0.9362 - val_loss: 0.3341 - val_acc: 0.9128

Epoch 16/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.1909 - acc: 0.9391 - val_loss: 0.3687 - val_acc: 0.9145

Epoch 17/30

7352/7352 [=====] - 135s 18ms/step - loss: 0.1932 - acc: 0.9373 - val_loss: 0.3218 - val_acc: 0.9070

Epoch 18/30

7352/7352 [=====] - 134s 18ms/step - loss: 0.1643 - acc: 0.9425 - val_loss: 0.3123 - val_acc: 0.9080

Epoch 19/30

7352/7352 [=====] - 137s 19ms/step - loss: 0.1666 -


```
acc: 0.9438 - val_loss: 0.3939 - val_acc: 0.9182
Epoch 20/30
7352/7352 [=====] - 135s 18ms/step - loss: 0.1633 -
acc: 0.9423 - val_loss: 0.3127 - val_acc: 0.9155
Epoch 21/30
7352/7352 [=====] - 136s 19ms/step - loss: 0.1758 -
acc: 0.9436 - val_loss: 0.4191 - val_acc: 0.8999
Epoch 22/30
7352/7352 [=====] - 134s 18ms/step - loss: 0.1579 -
acc: 0.9433 - val_loss: 0.4217 - val_acc: 0.8982
Epoch 23/30
7352/7352 [=====] - 137s 19ms/step - loss: 0.1589 -
acc: 0.9441 - val_loss: 0.2486 - val_acc: 0.9006
Epoch 24/30
7352/7352 [=====] - 135s 18ms/step - loss: 0.1497 -
acc: 0.9445 - val_loss: 0.3153 - val_acc: 0.9023
Epoch 25/30
7352/7352 [=====] - 136s 18ms/step - loss: 0.1422 -
acc: 0.9479 - val_loss: 0.3781 - val_acc: 0.9189
Epoch 26/30
7352/7352 [=====] - 136s 18ms/step - loss: 0.1503 -
acc: 0.9499 - val_loss: 0.4217 - val_acc: 0.9030
Epoch 27/30
7352/7352 [=====] - 135s 18ms/step - loss: 0.1594 -
acc: 0.9431 - val_loss: 0.4035 - val_acc: 0.9104
Epoch 28/30
7352/7352 [=====] - 135s 18ms/step - loss: 0.1597 -
acc: 0.9410 - val_loss: 0.3851 - val_acc: 0.9087
Epoch 29/30
7352/7352 [=====] - 135s 18ms/step - loss: 0.1853 -
acc: 0.9434 - val_loss: 0.4077 - val_acc: 0.8853
Epoch 30/30
7352/7352 [=====] - 134s 18ms/step - loss: 0.1434 -
acc: 0.9461 - val_loss: 0.4164 - val_acc: 0.9046
```

Out[25]: <keras.callbacks.History at 0x135711a3da0>

```
In [26]: # Confusion Matrix
print(confusion_matrix(Y_test, model.predict(X_test)))
```

Pred \ True	LAYING	SITTING	STANDING	WALKING	WALKING_DOWNSTAIRS	WALKING_UPSTAIRS
LAYING	510	0	27	0	0	0
SITTING	0	379	112	0	0	0
STANDING	0	70	461	1	0	0
WALKING	0	0	0	472	0	0
WALKING_DOWNSTAIRS	0	0	0	5	378	0
WALKING_UPSTAIRS	0	1	0	4	0	0

Pred \ True	WALKING_UPSTAIRS
LAYING	0
SITTING	0
STANDING	0
WALKING	24
WALKING_DOWNSTAIRS	37
WALKING_UPSTAIRS	466

```
In [27]: score = model.evaluate(X_test, Y_test)
```

2947/2947 [=====] - 8s 3ms/step

```
In [28]: score
```

```
Out[28]: [0.4163782881195881, 0.9046487953851374]
```

Model-5:

```
In [52]:
```

```
In [53]: model = Sequential()
model.add(LSTM(100, input_shape=(timesteps, input_dim), kernel_initializer='glorot_normal', return_sequences=True, bias_regularizer=reg))
model.add(BatchNormalization())
model.add(Dropout(0.70))
model.add(LSTM(50))
model.add(Dropout(0.70))
model.add(Dense(n_classes, activation='sigmoid'))
print("Model Summary: ")
model.summary()
```

Model Summary:

Model: "sequential_6"

Layer (type)	Output Shape	Param #
lstm_10 (LSTM)	(None, 128, 100)	44000
batch_normalization_5 (Batch Normalization)	(None, 128, 100)	400
dropout_9 (Dropout)	(None, 128, 100)	0
lstm_11 (LSTM)	(None, 50)	30200
dropout_10 (Dropout)	(None, 50)	0
dense_5 (Dense)	(None, 6)	306
Total params: 74,906		
Trainable params: 74,706		
Non-trainable params: 200		

```
In [56]: # Compiling the model
model.compile(loss='binary_crossentropy',
              optimizer='rmsprop',
              metrics=['accuracy'])
checkpoint_3 = ModelCheckpoint("model_7.h5", monitor="val_acc", mode="max", save_best_only = True, verbose=1)
```

```
In [57]: # Training the model  
model.fit(X_train,  
          Y_train,  
          batch_size=batch_size,  
          validation_data=(X_test, Y_test),  
          epochs=20, callbacks=[checkpoint_3])
```

Train on 7352 samples, validate on 2947 samples

Epoch 1/20

7352/7352 [=====] - 244s 33ms/step - loss: 1.7097 -
acc: 0.8483 - val_loss: 1.0533 - val_acc: 0.8947

Epoch 00001: val_acc improved from -inf to 0.89470, saving model to model_7.h5

Epoch 2/20

7352/7352 [=====] - 249s 34ms/step - loss: 0.6515 -
acc: 0.9053 - val_loss: 0.2413 - val_acc: 0.9536

Epoch 00002: val_acc improved from 0.89470 to 0.95357, saving model to model_7.h5

Epoch 3/20

7352/7352 [=====] - 240s 33ms/step - loss: 0.1730 -
acc: 0.9407 - val_loss: 0.1105 - val_acc: 0.9600

Epoch 00003: val_acc improved from 0.95357 to 0.96002, saving model to model_7.h5

Epoch 4/20

7352/7352 [=====] - 243s 33ms/step - loss: 0.1270 -
acc: 0.9586 - val_loss: 0.1410 - val_acc: 0.9484

Epoch 00004: val_acc did not improve from 0.96002

Epoch 5/20

7352/7352 [=====] - 253s 34ms/step - loss: 0.1128 -
acc: 0.9617 - val_loss: 0.1013 - val_acc: 0.9673

Epoch 00005: val_acc improved from 0.96002 to 0.96725, saving model to model_7.h5

Epoch 6/20

7352/7352 [=====] - 247s 34ms/step - loss: 0.1016 -
acc: 0.9652 - val_loss: 0.0821 - val_acc: 0.9706

Epoch 00006: val_acc improved from 0.96725 to 0.97059, saving model to model_7.h5

Epoch 7/20

7352/7352 [=====] - 248s 34ms/step - loss: 0.0906 -
acc: 0.9697 - val_loss: 0.1346 - val_acc: 0.9522

Epoch 00007: val_acc did not improve from 0.97059

Epoch 8/20

7352/7352 [=====] - 257s 35ms/step - loss: 0.0874 -
acc: 0.9704 - val_loss: 0.1169 - val_acc: 0.9579

Epoch 00008: val_acc did not improve from 0.97059

Epoch 9/20

7352/7352 [=====] - 265s 36ms/step - loss: 0.0859 -
acc: 0.9700 - val_loss: 0.1319 - val_acc: 0.9574

Epoch 00009: val_acc did not improve from 0.97059

Epoch 10/20

7352/7352 [=====] - 264s 36ms/step - loss: 0.0807 -
acc: 0.9726 - val_loss: 0.0992 - val_acc: 0.9663

Epoch 00010: val_acc did not improve from 0.97059

Epoch 11/20

```
7352/7352 [=====] - 249s 34ms/step - loss: 0.0816 -  
acc: 0.9730 - val_loss: 0.0948 - val_acc: 0.9680
```

Epoch 00011: val_acc did not improve from 0.97059

Epoch 12/20

```
7352/7352 [=====] - 255s 35ms/step - loss: 0.0767 -  
acc: 0.9729 - val_loss: 0.1639 - val_acc: 0.9501
```

Epoch 00012: val_acc did not improve from 0.97059

Epoch 13/20

```
7352/7352 [=====] - 263s 36ms/step - loss: 0.0767 -  
acc: 0.9724 - val_loss: 0.1193 - val_acc: 0.9649
```

Epoch 00013: val_acc did not improve from 0.97059

Epoch 14/20

```
7352/7352 [=====] - 270s 37ms/step - loss: 0.0733 -  
acc: 0.9738 - val_loss: 0.1060 - val_acc: 0.9657
```

Epoch 00014: val_acc did not improve from 0.97059

Epoch 15/20

```
7352/7352 [=====] - 289s 39ms/step - loss: 0.0737 -  
acc: 0.9738 - val_loss: 0.0743 - val_acc: 0.9751
```

Epoch 00015: val_acc improved from 0.97059 to 0.97512, saving model to model_7.h5

Epoch 16/20

```
7352/7352 [=====] - 300s 41ms/step - loss: 0.0722 -  
acc: 0.9745 - val_loss: 0.1129 - val_acc: 0.9722
```

Epoch 00016: val_acc did not improve from 0.97512

Epoch 17/20

```
7352/7352 [=====] - 295s 40ms/step - loss: 0.0723 -  
acc: 0.9751 - val_loss: 0.0909 - val_acc: 0.9710
```

Epoch 00017: val_acc did not improve from 0.97512

Epoch 18/20

```
7352/7352 [=====] - 358s 49ms/step - loss: 0.0701 -  
acc: 0.9752 - val_loss: 0.1355 - val_acc: 0.9623
```

Epoch 00018: val_acc did not improve from 0.97512

Epoch 19/20

```
7352/7352 [=====] - 329s 45ms/step - loss: 0.0715 -  
acc: 0.9743 - val_loss: 0.0846 - val_acc: 0.9731
```

Epoch 00019: val_acc did not improve from 0.97512

Epoch 20/20

```
7352/7352 [=====] - 275s 37ms/step - loss: 0.0684 -  
acc: 0.9748 - val_loss: 0.0841 - val_acc: 0.9748
```

Epoch 00020: val_acc did not improve from 0.97512

Out[57]: <keras.callbacks.History at 0x15c9fe36f28>

In [58]: `model = load_model('model_7.h5')`

```
In [59]: # Confusion Matrix
print(confusion_matrix(Y_test, model.predict(X_test)))
```

Pred \ True	LAYING	SITTING	STANDING	WALKING	WALKING_DOWNSTAIRS	WALKING_UPSTAIRS
LAYING	537	0	0	0	0	0
SITTING	20	419	51	0	0	0
STANDING	0	116	416	0	0	0
WALKING	0	0	2	490	0	3
WALKING_DOWNSTAIRS	0	0	0	0	420	0
WALKING_UPSTAIRS	0	1	0	11	0	10

Pred \ True	WALKING_UPSTAIRS
LAYING	0
SITTING	1
STANDING	0
WALKING	1
WALKING_DOWNSTAIRS	0
WALKING_UPSTAIRS	449

```
In [60]: score = model.evaluate(X_test, Y_test)
```

2947/2947 [=====] - 18s 6ms/step

```
In [61]: score
```

```
Out[61]: [0.07431104327187013, 0.9751159407787497]
```

```
In [ ]:
```

Results(PrettyTable):

```
In [3]: from prettytable import PrettyTable
x = PrettyTable()
x.field_names = ["Model", "Description", "Test loss", "Test Accuracy"]
x.add_row(["1", "1 Layer of LSTM(40)", "0.5031", "0.8893"])
x.add_row(["2", "2 layers of LSTM, BN, Categoricalcross_entropy", "0.2544", "0.9338"])
x.add_row(["3", "1 Layer of LSTM(70)", "0.4624", "0.8988"])
x.add_row(["4", "1 Layer of LSTM(50)", "0.4163", "0.9046"])
x.add_row(["5", "2 layers of LSTM, BN, Binarycross_entropy", "0.0743", "0.9751"])
print(x)
```

```
+-----+-----+-----+-----+
+-----+
| Model |           Description           | Test loss | Test A
ccuracy |
+-----+-----+-----+-----+
+-----+
|  1   |           1 Layer of LSTM(40)           |    0.5031 |    0.
8893   |
|  2   | 2 layers of LSTM, BN, Categoricalcross_entropy |    0.2544 |    0.
9338   |
|  3   |           1 Layer of LSTM(70)           |    0.4624 |    0.
8988   |
|  4   |           1 Layer of LSTM(50)           |    0.4163 |    0.
9046   |
|  5   | 2 layers of LSTM, BN, Binarycross_entropy |    0.0743 |    0.
9751   |
+-----+-----+-----+-----+
+-----+
```

Conclusions:

1. First i've used a single LSTM layer with 40 units, and got 0.88
2. Then i used 2 layers of LSTM follwed by Batch normalizationn with categorical cross entropy and got 0.93
3. For the 3rd and 4th model ,i used 1 Layer of LSTM with 70,50 units respectively and got a accuracy of 0.90
4. Now For 5th model i've used 2 Layers of LSTM with Binary cross entropy along with Batch normalization,with some dropout and got accuracy of 0.97

In []: