

# HAR\_LSTM

June 25, 2019

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

```
In [1]: from google.colab import drive
drive.mount('HAR_Dataset',force_remount=True)

rootpath="HAR_Dataset/My Drive"
```

Go to this URL in a browser: [https://accounts.google.com/o/oauth2/auth?client\\_id=947318989803-6b](https://accounts.google.com/o/oauth2/auth?client_id=947318989803-6b)

Enter your authorization code:

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Mounted at HAR\_Dataset

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

```
In [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

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

```

In [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"
        ]

In [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'HAR_Dataset/My Drive/UCI_HAR_Dataset/{subset}/Inertial Signals/{sig
                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 [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'HAR_Dataset/My Drive/UCI_HAR_Dataset/{subset}/y_{subset}.txt'
        y = _read_csv(filename)[0]

        return pd.get_dummies(y).as_matrix()

```

```

In [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

In [0]: # Importing tensorflow
np.random.seed(42)
import tensorflow as tf
tf.set_random_seed(42)

In [0]: # 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)

Using TensorFlow backend.

In [0]: # Importing libraries
from keras.models import Sequential
from keras.layers import LSTM
from keras.layers.core import Dense, Dropout

In [0]: # Initializing parameters

epochs = 30

batch_size = 16

n_hidden = 32

In [0]: # 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()

```

```
/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:11: FutureWarning: Method .as_matrix() is deprecated, use .to_matrix() instead.  
# This is added back by InteractiveShellApp.init_path()  
/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:12: FutureWarning: Method .as_matrix() is deprecated, use .to_matrix() instead.  
if sys.path[0] == '':
```

```
In [16]: 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
```

## 1 Assignment A) Hypertuning Num of Hidden Units

```
In [0]: hidden_Neurons_List=[8,16,32,64,128]
```

```
hidden_Neurons_Accuracy=[]
```

```
for n_hidden in hidden_Neurons_List:  
    model=Sequential()
```

```
    model.add(LSTM(n_hidden,input_shape=(timesteps,input_dim)))
```

```
    model.add(Dropout(0.5))
```

```
    model.add(Dense(n_classes,activation='sigmoid'))
```

```
    model.summary()
```

```
    model.compile(loss='categorical_crossentropy',optimizer='rmsprop',metrics=['accuracy'])
```

```
    model.fit(X_train,  
              Y_train,  
              batch_size=batch_size,  
              validation_data=(X_test, Y_test),  
              epochs=epochs)
```

```
    score = model.evaluate(X_test, Y_test)
```

```
hidden_Neurons_Accuracy.append(score)
```

Layer (type)	Output Shape	Param #
lstm_11 (LSTM)	(None, 8)	576
dropout_11 (Dropout)	(None, 8)	0
dense_11 (Dense)	(None, 6)	54

Total params: 630

Trainable params: 630

Non-trainable params: 0

Train on 7352 samples, validate on 2947 samples

Epoch 1/30

7352/7352 [=====] - 32s 4ms/step - loss: 1.5955 - acc: 0.3632 - val\_loss: 1.5955

Epoch 2/30

7352/7352 [=====] - 30s 4ms/step - loss: 1.3659 - acc: 0.4550 - val\_loss: 1.3659

Epoch 3/30

7352/7352 [=====] - 30s 4ms/step - loss: 1.2118 - acc: 0.4886 - val\_loss: 1.2118

Epoch 4/30

7352/7352 [=====] - 30s 4ms/step - loss: 1.0938 - acc: 0.5291 - val\_loss: 1.0938

Epoch 5/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.9912 - acc: 0.5642 - val\_loss: 0.9912

Epoch 6/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.9305 - acc: 0.5809 - val\_loss: 0.9305

Epoch 7/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.8942 - acc: 0.5838 - val\_loss: 0.8942

Epoch 8/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.8780 - acc: 0.5903 - val\_loss: 0.8780

Epoch 9/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.8793 - acc: 0.5926 - val\_loss: 0.8793

Epoch 10/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.8699 - acc: 0.6002 - val\_loss: 0.8699

Epoch 11/30

7352/7352 [=====] - 29s 4ms/step - loss: 0.8410 - acc: 0.6042 - val\_loss: 0.8410

Epoch 12/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.8768 - acc: 0.5898 - val\_loss: 0.8768

Epoch 13/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.9041 - acc: 0.5709 - val\_loss: 0.9041

Epoch 14/30

7352/7352 [=====] - 30s 4ms/step - loss: 0.8725 - acc: 0.5997 - val\_loss: 0.8725

```

Epoch 15/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8794 - acc: 0.5838 - val_loss: 0.8794
Epoch 16/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8170 - acc: 0.6055 - val_loss: 0.8170
Epoch 17/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8236 - acc: 0.6107 - val_loss: 0.8236
Epoch 18/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.7993 - acc: 0.6160 - val_loss: 0.7993
Epoch 19/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8132 - acc: 0.6137 - val_loss: 0.8132
Epoch 20/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.7903 - acc: 0.6182 - val_loss: 0.7903
Epoch 21/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8168 - acc: 0.6096 - val_loss: 0.8168
Epoch 22/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8677 - acc: 0.5887 - val_loss: 0.8677
Epoch 23/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.9627 - acc: 0.5664 - val_loss: 0.9627
Epoch 24/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8747 - acc: 0.5849 - val_loss: 0.8747
Epoch 25/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8360 - acc: 0.5979 - val_loss: 0.8360
Epoch 26/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8268 - acc: 0.6027 - val_loss: 0.8268
Epoch 27/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8024 - acc: 0.6113 - val_loss: 0.8024
Epoch 28/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.8014 - acc: 0.6119 - val_loss: 0.8014
Epoch 29/30
7352/7352 [=====] - 29s 4ms/step - loss: 0.7988 - acc: 0.6103 - val_loss: 0.7988
Epoch 30/30
7352/7352 [=====] - 30s 4ms/step - loss: 0.7801 - acc: 0.6163 - val_loss: 0.7801
2947/2947 [=====] - 1s 310us/step

```

Layer (type)	Output Shape	Param #
lstm_12 (LSTM)	(None, 16)	1664
dropout_12 (Dropout)	(None, 16)	0
dense_12 (Dense)	(None, 6)	102

```

Total params: 1,766
Trainable params: 1,766
Non-trainable params: 0

```

```

Train on 7352 samples, validate on 2947 samples
Epoch 1/30

```

7352/7352 [=====] - 33s 5ms/step - loss: 1.4487 - acc: 0.3957 - val\_loss: 1.4487  
 Epoch 2/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 1.1780 - acc: 0.4678 - val\_loss: 1.1780  
 Epoch 3/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 1.1275 - acc: 0.4739 - val\_loss: 1.1275  
 Epoch 4/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 1.0714 - acc: 0.4884 - val\_loss: 1.0714  
 Epoch 5/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 1.0013 - acc: 0.5394 - val\_loss: 1.0013  
 Epoch 6/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 0.8838 - acc: 0.6175 - val\_loss: 0.8838  
 Epoch 7/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 0.8260 - acc: 0.6488 - val\_loss: 0.8260  
 Epoch 8/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.7483 - acc: 0.6778 - val\_loss: 0.7483  
 Epoch 9/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.7194 - acc: 0.6812 - val\_loss: 0.7194  
 Epoch 10/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.6657 - acc: 0.7163 - val\_loss: 0.6657  
 Epoch 11/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.6537 - acc: 0.7145 - val\_loss: 0.6537  
 Epoch 12/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.6750 - acc: 0.7187 - val\_loss: 0.6750  
 Epoch 13/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.6330 - acc: 0.7295 - val\_loss: 0.6330  
 Epoch 14/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.6436 - acc: 0.7382 - val\_loss: 0.6436  
 Epoch 15/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 0.6338 - acc: 0.7533 - val\_loss: 0.6338  
 Epoch 16/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.5908 - acc: 0.7652 - val\_loss: 0.5908  
 Epoch 17/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.5812 - acc: 0.7595 - val\_loss: 0.5812  
 Epoch 18/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 0.5537 - acc: 0.7677 - val\_loss: 0.5537  
 Epoch 19/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.5465 - acc: 0.7777 - val\_loss: 0.5465  
 Epoch 20/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.5127 - acc: 0.7926 - val\_loss: 0.5127  
 Epoch 21/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 0.4995 - acc: 0.8292 - val\_loss: 0.4995  
 Epoch 22/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.5634 - acc: 0.8281 - val\_loss: 0.5634  
 Epoch 23/30  
 7352/7352 [=====] - 31s 4ms/step - loss: 0.4479 - acc: 0.8602 - val\_loss: 0.4479  
 Epoch 24/30  
 7352/7352 [=====] - 32s 4ms/step - loss: 0.4057 - acc: 0.8803 - val\_loss: 0.4057  
 Epoch 25/30

```

7352/7352 [=====] - 31s 4ms/step - loss: 0.3659 - acc: 0.8920 - val_loss: 0.3659
Epoch 26/30
7352/7352 [=====] - 32s 4ms/step - loss: 0.3439 - acc: 0.8984 - val_loss: 0.3439
Epoch 27/30
7352/7352 [=====] - 32s 4ms/step - loss: 0.3343 - acc: 0.8981 - val_loss: 0.3343
Epoch 28/30
7352/7352 [=====] - 31s 4ms/step - loss: 0.3665 - acc: 0.8915 - val_loss: 0.3665
Epoch 29/30
7352/7352 [=====] - 31s 4ms/step - loss: 0.3177 - acc: 0.9087 - val_loss: 0.3177
Epoch 30/30
7352/7352 [=====] - 31s 4ms/step - loss: 0.2947 - acc: 0.9087 - val_loss: 0.2947
2947/2947 [=====] - 1s 330us/step

```

Layer (type)	Output Shape	Param #
lstm_13 (LSTM)	(None, 32)	5376
dropout_13 (Dropout)	(None, 32)	0
dense_13 (Dense)	(None, 6)	198

```

Total params: 5,574
Trainable params: 5,574
Non-trainable params: 0

```

```

Train on 7352 samples, validate on 2947 samples
Epoch 1/30
7352/7352 [=====] - 35s 5ms/step - loss: 1.3670 - acc: 0.4244 - val_loss: 1.3670
Epoch 2/30
7352/7352 [=====] - 34s 5ms/step - loss: 1.0645 - acc: 0.5434 - val_loss: 1.0645
Epoch 3/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.8503 - acc: 0.6468 - val_loss: 0.8503
Epoch 4/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.7632 - acc: 0.6723 - val_loss: 0.7632
Epoch 5/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.6299 - acc: 0.7274 - val_loss: 0.6299
Epoch 6/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.5365 - acc: 0.7724 - val_loss: 0.5365
Epoch 7/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.4865 - acc: 0.7878 - val_loss: 0.4865
Epoch 8/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.4402 - acc: 0.7976 - val_loss: 0.4402
Epoch 9/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.4274 - acc: 0.8067 - val_loss: 0.4274
Epoch 10/30
7352/7352 [=====] - 34s 5ms/step - loss: 0.3714 - acc: 0.8312 - val_loss: 0.3714
Epoch 11/30
7352/7352 [=====] - 34s 5ms/step - loss: 0.3583 - acc: 0.8683 - val_loss: 0.3583

```



```

Epoch 12/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.3338 - acc: 0.8924 - val_loss: 0.3338
Epoch 13/30
7352/7352 [=====] - 34s 5ms/step - loss: 0.2621 - acc: 0.9230 - val_loss: 0.2621
Epoch 14/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.2387 - acc: 0.9263 - val_loss: 0.2387
Epoch 15/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.2641 - acc: 0.9165 - val_loss: 0.2641
Epoch 16/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.2405 - acc: 0.9189 - val_loss: 0.2405
Epoch 17/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.2195 - acc: 0.9309 - val_loss: 0.2195
Epoch 18/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.2089 - acc: 0.9295 - val_loss: 0.2089
Epoch 19/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1861 - acc: 0.9312 - val_loss: 0.1861
Epoch 20/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.1760 - acc: 0.9387 - val_loss: 0.1760
Epoch 21/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.1834 - acc: 0.9357 - val_loss: 0.1834
Epoch 22/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1873 - acc: 0.9357 - val_loss: 0.1873
Epoch 23/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.1811 - acc: 0.9376 - val_loss: 0.1811
Epoch 24/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1882 - acc: 0.9391 - val_loss: 0.1882
Epoch 25/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1621 - acc: 0.9395 - val_loss: 0.1621
Epoch 26/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1590 - acc: 0.9433 - val_loss: 0.1590
Epoch 27/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1739 - acc: 0.9410 - val_loss: 0.1739
Epoch 28/30
7352/7352 [=====] - 33s 4ms/step - loss: 0.1762 - acc: 0.9448 - val_loss: 0.1762
Epoch 29/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1490 - acc: 0.9431 - val_loss: 0.1490
Epoch 30/30
7352/7352 [=====] - 33s 5ms/step - loss: 0.1680 - acc: 0.9430 - val_loss: 0.1680
2947/2947 [=====] - 1s 377us/step

```

Layer (type)	Output Shape	Param #
lstm_14 (LSTM)	(None, 64)	18944
dropout_14 (Dropout)	(None, 64)	0
dense_14 (Dense)	(None, 6)	390

Total params: 19,334  
Trainable params: 19,334  
Non-trainable params: 0

-----  
Train on 7352 samples, validate on 2947 samples

Epoch 1/30

7352/7352 [=====] - 41s 6ms/step - loss: 1.2066 - acc: 0.4890 - val\_loss: 1.2066

Epoch 2/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.9584 - acc: 0.5694 - val\_loss: 0.9584

Epoch 3/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.8011 - acc: 0.6730 - val\_loss: 0.8011

Epoch 4/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.6206 - acc: 0.7609 - val\_loss: 0.6206

Epoch 5/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.5015 - acc: 0.8267 - val\_loss: 0.5015

Epoch 6/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.4565 - acc: 0.8516 - val\_loss: 0.4565

Epoch 7/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.2895 - acc: 0.9008 - val\_loss: 0.2895

Epoch 8/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.2505 - acc: 0.9195 - val\_loss: 0.2505

Epoch 9/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.2373 - acc: 0.9298 - val\_loss: 0.2373

Epoch 10/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.2064 - acc: 0.9317 - val\_loss: 0.2064

Epoch 11/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.2192 - acc: 0.9305 - val\_loss: 0.2192

Epoch 12/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.2107 - acc: 0.9334 - val\_loss: 0.2107

Epoch 13/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.1739 - acc: 0.9397 - val\_loss: 0.1739

Epoch 14/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.1773 - acc: 0.9422 - val\_loss: 0.1773

Epoch 15/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.1875 - acc: 0.9411 - val\_loss: 0.1875

Epoch 16/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.1709 - acc: 0.9395 - val\_loss: 0.1709

Epoch 17/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.1560 - acc: 0.9433 - val\_loss: 0.1560

Epoch 18/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.1454 - acc: 0.9441 - val\_loss: 0.1454

Epoch 19/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.1786 - acc: 0.9342 - val\_loss: 0.1786

Epoch 20/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.1506 - acc: 0.9471 - val\_loss: 0.1506

Epoch 21/30

7352/7352 [=====] - 39s 5ms/step - loss: 0.1713 - acc: 0.9436 - val\_loss: 0.1713

Epoch 22/30

```

7352/7352 [=====] - 38s 5ms/step - loss: 0.1466 - acc: 0.9484 - val_loss: 0.1466
Epoch 23/30
7352/7352 [=====] - 38s 5ms/step - loss: 0.1567 - acc: 0.9470 - val_loss: 0.1567
Epoch 24/30
7352/7352 [=====] - 38s 5ms/step - loss: 0.1475 - acc: 0.9478 - val_loss: 0.1475
Epoch 25/30
7352/7352 [=====] - 38s 5ms/step - loss: 0.1833 - acc: 0.9436 - val_loss: 0.1833
Epoch 26/30
7352/7352 [=====] - 39s 5ms/step - loss: 0.1539 - acc: 0.9445 - val_loss: 0.1539
Epoch 27/30
7352/7352 [=====] - 38s 5ms/step - loss: 0.1573 - acc: 0.9472 - val_loss: 0.1573
Epoch 28/30
7352/7352 [=====] - 39s 5ms/step - loss: 0.2278 - acc: 0.9357 - val_loss: 0.2278
Epoch 29/30
7352/7352 [=====] - 38s 5ms/step - loss: 0.1552 - acc: 0.9474 - val_loss: 0.1552
Epoch 30/30
7352/7352 [=====] - 39s 5ms/step - loss: 0.1520 - acc: 0.9470 - val_loss: 0.1520
2947/2947 [=====] - 2s 526us/step

```

Layer (type)	Output Shape	Param #
lstm_15 (LSTM)	(None, 128)	70656
dropout_15 (Dropout)	(None, 128)	0
dense_15 (Dense)	(None, 6)	774

```

Total params: 71,430
Trainable params: 71,430
Non-trainable params: 0

```

```

Train on 7352 samples, validate on 2947 samples
Epoch 1/30
7352/7352 [=====] - 60s 8ms/step - loss: 1.2407 - acc: 0.4478 - val_loss: 1.2407
Epoch 2/30
7352/7352 [=====] - 58s 8ms/step - loss: 0.9093 - acc: 0.5819 - val_loss: 0.9093
Epoch 3/30
7352/7352 [=====] - 57s 8ms/step - loss: 0.7344 - acc: 0.6393 - val_loss: 0.7344
Epoch 4/30
7352/7352 [=====] - 58s 8ms/step - loss: 0.6823 - acc: 0.6927 - val_loss: 0.6823
Epoch 5/30
7352/7352 [=====] - 58s 8ms/step - loss: 0.4828 - acc: 0.8158 - val_loss: 0.4828
Epoch 6/30
7352/7352 [=====] - 58s 8ms/step - loss: 0.3518 - acc: 0.8799 - val_loss: 0.3518
Epoch 7/30
7352/7352 [=====] - 58s 8ms/step - loss: 0.2519 - acc: 0.9104 - val_loss: 0.2519
Epoch 8/30
7352/7352 [=====] - 58s 8ms/step - loss: 0.2390 - acc: 0.9249 - val_loss: 0.2390

```

Epoch 9/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1885 - acc: 0.9363 - val\_loss: 0.1885  
Epoch 10/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1728 - acc: 0.9361 - val\_loss: 0.1728  
Epoch 11/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1587 - acc: 0.9431 - val\_loss: 0.1587  
Epoch 12/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1562 - acc: 0.9445 - val\_loss: 0.1562  
Epoch 13/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1533 - acc: 0.9456 - val\_loss: 0.1533  
Epoch 14/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1491 - acc: 0.9486 - val\_loss: 0.1491  
Epoch 15/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1383 - acc: 0.9465 - val\_loss: 0.1383  
Epoch 16/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1292 - acc: 0.9517 - val\_loss: 0.1292  
Epoch 17/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1563 - acc: 0.9484 - val\_loss: 0.1563  
Epoch 18/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1503 - acc: 0.9484 - val\_loss: 0.1503  
Epoch 19/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1697 - acc: 0.9441 - val\_loss: 0.1697  
Epoch 20/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1383 - acc: 0.9486 - val\_loss: 0.1383  
Epoch 21/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1327 - acc: 0.9505 - val\_loss: 0.1327  
Epoch 22/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1564 - acc: 0.9450 - val\_loss: 0.1564  
Epoch 23/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1220 - acc: 0.9513 - val\_loss: 0.1220  
Epoch 24/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1352 - acc: 0.9495 - val\_loss: 0.1352  
Epoch 25/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1334 - acc: 0.9474 - val\_loss: 0.1334  
Epoch 26/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1211 - acc: 0.9527 - val\_loss: 0.1211  
Epoch 27/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1188 - acc: 0.9544 - val\_loss: 0.1188  
Epoch 28/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1287 - acc: 0.9506 - val\_loss: 0.1287  
Epoch 29/30  
7352/7352 [=====] - 57s 8ms/step - loss: 0.1709 - acc: 0.9436 - val\_loss: 0.1709  
Epoch 30/30  
7352/7352 [=====] - 58s 8ms/step - loss: 0.1315 - acc: 0.9517 - val\_loss: 0.1315  
2947/2947 [=====] - 3s 943us/step

-----  
TypeError Traceback (most recent call last)

```
<ipython-input-100-2acaa322236d> in <module>()
    30
    31 for i in range(len(hidden_Neurons_List)):
---> 32     print("Number of Hidden Neurons : %d, Accuracy : %d"%(hidden_Neurons_List[i],hidden_Neurons_Accuracy[i]))
    33
    34
```

TypeError: %d format: a number is required, not list

```
In [0]: print(hidden_Neurons_List)
        print(".....")
        print(hidden_Neurons_Accuracy)

[8, 16, 32, 64, 128]
...
[[0.7497640215505128, 0.6067186969799796], [0.5063654349498706, 0.8486596538853071], [0.44709858
```

```
In [0]: max_acc=0
        y=[]
        for i in range(len(hidden_Neurons_List)):
            if hidden_Neurons_Accuracy[i][1]>max_acc:
                max_acc=hidden_Neurons_Accuracy[i][1]

                best_n_hidden=hidden_Neurons_List[i]

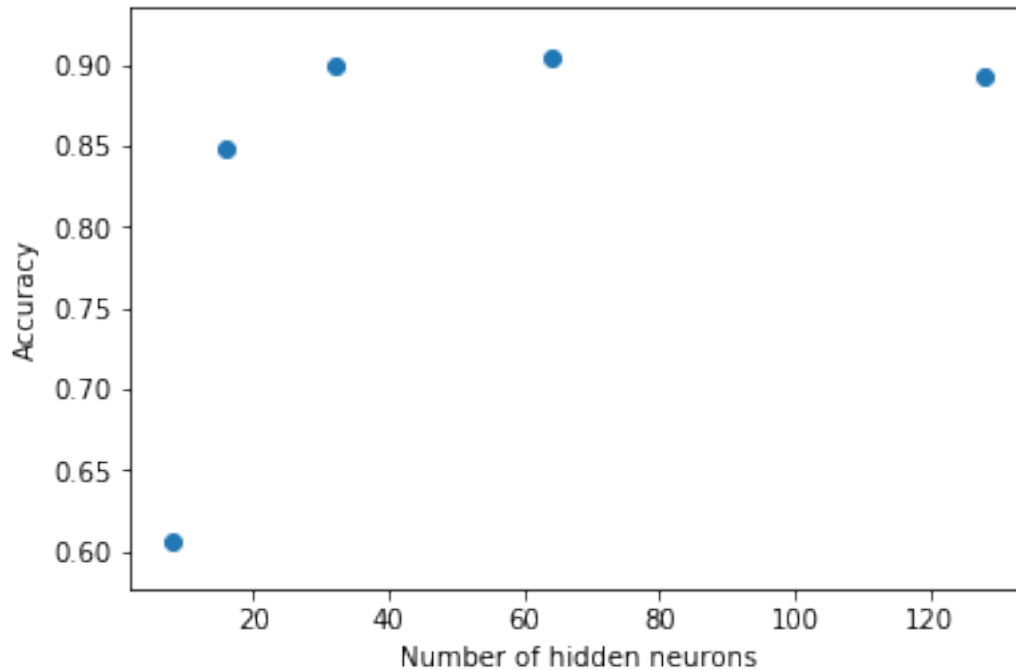
        y.append(hidden_Neurons_Accuracy[i][1])
        print("Number of Hidden Neurons : %d, Accuracy : %f"%(hidden_Neurons_List[i],hidden_Neurons_Accuracy[i][1]))
```

```
Number of Hidden Neurons : 8, Accuracy : 0.606719
Number of Hidden Neurons : 16, Accuracy : 0.848660
Number of Hidden Neurons : 32, Accuracy : 0.899220
Number of Hidden Neurons : 64, Accuracy : 0.904988
Number of Hidden Neurons : 128, Accuracy : 0.892094
```

```
In [0]: import numpy as np
        import matplotlib.pyplot as plt

        x = hidden_Neurons_List
```

```
plt.scatter(x, y)
plt.xlabel('Number of hidden neurons')
plt.ylabel('Accuracy')
plt.show()
```



```
In [0]: print("The best value of number of hidden neurons is %d"%best_n_hidden)
```

The best value of number of hidden neurons is 64

## 2 Assignment B) Hypertuning Dropout Rate

```
In [0]:
```

```
In [0]: dropout_List=[0.1,0.3,0.5,0.7,0.9]
```

```
dropout_Accuracy=[]
```

```
for dropout in dropout_List:
    model=Sequential()
```

```
    model.add(LSTM(best_n_hidden,input_shape=(timesteps,input_dim)))
```

```

model.add(Dropout(dropout))

model.add(Dense(n_classes,activation='sigmoid'))

model.summary()

model.compile(loss='categorical_crossentropy',optimizer='rmsprop',metrics=['accuracy'])

model.fit(X_train,
          Y_train,
          batch_size=batch_size,
          validation_data=(X_test, Y_test),
          epochs=epochs)

score = model.evaluate(X_test, Y_test)

dropout_Accuracy.append(score)

```

WARNING: Logging before flag parsing goes to stderr.

W0617 05:37:13.891773 140504591730560 deprecation\_wrapper.py:119] From /usr/local/lib/python3.6/

W0617 05:37:13.895210 140504591730560 deprecation\_wrapper.py:119] From /usr/local/lib/python3.6/

W0617 05:37:13.906264 140504591730560 deprecation\_wrapper.py:119] From /usr/local/lib/python3.6/

W0617 05:37:14.170121 140504591730560 deprecation\_wrapper.py:119] From /usr/local/lib/python3.6/

W0617 05:37:14.182163 140504591730560 deprecation.py:506] From /usr/local/lib/python3.6/dist-pac

Instructions for updating:

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

W0617 05:37:14.214296 140504591730560 deprecation\_wrapper.py:119] From /usr/local/lib/python3.6/

W0617 05:37:14.237277 140504591730560 deprecation\_wrapper.py:119] From /usr/local/lib/python3.6/

W0617 05:37:14.389814 140504591730560 deprecation.py:323] From /usr/local/lib/python3.6/dist-pac

Instructions for updating:

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

```

-----
Layer (type)                Output Shape                Param #
=====
lstm_1 (LSTM)                (None, 64)                  18944

```

-----  
dropout\_1 (Dropout) (None, 64) 0  
-----

dense\_1 (Dense) (None, 6) 390  
=====

Total params: 19,334

Trainable params: 19,334

Non-trainable params: 0

-----  
Train on 7352 samples, validate on 2947 samples

Epoch 1/30

7352/7352 [=====] - 37s 5ms/step - loss: 1.2624 - acc: 0.4302 - val\_loss: 1.2624

Epoch 2/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.9102 - acc: 0.5945 - val\_loss: 0.9102

Epoch 3/30

7352/7352 [=====] - 36s 5ms/step - loss: 0.6894 - acc: 0.7201 - val\_loss: 0.6894

Epoch 4/30

7352/7352 [=====] - 36s 5ms/step - loss: 0.6155 - acc: 0.7780 - val\_loss: 0.6155

Epoch 5/30

7352/7352 [=====] - 36s 5ms/step - loss: 0.4409 - acc: 0.8474 - val\_loss: 0.4409

Epoch 6/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.3038 - acc: 0.8988 - val\_loss: 0.3038

Epoch 7/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.2346 - acc: 0.9174 - val\_loss: 0.2346

Epoch 8/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1999 - acc: 0.9291 - val\_loss: 0.1999

Epoch 9/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1781 - acc: 0.9366 - val\_loss: 0.1781

Epoch 10/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1564 - acc: 0.9369 - val\_loss: 0.1564

Epoch 11/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1501 - acc: 0.9438 - val\_loss: 0.1501

Epoch 12/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1606 - acc: 0.9419 - val\_loss: 0.1606

Epoch 13/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1453 - acc: 0.9429 - val\_loss: 0.1453

Epoch 14/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1542 - acc: 0.9453 - val\_loss: 0.1542

Epoch 15/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1489 - acc: 0.9426 - val\_loss: 0.1489

Epoch 16/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1412 - acc: 0.9468 - val\_loss: 0.1412

Epoch 17/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1386 - acc: 0.9464 - val\_loss: 0.1386

Epoch 18/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1273 - acc: 0.9476 - val\_loss: 0.1273

Epoch 19/30

7352/7352 [=====] - 35s 5ms/step - loss: 0.1273 - acc: 0.9505 - val\_loss: 0.1273



```

Epoch 20/30
7352/7352 [=====] - 35s 5ms/step - loss: 0.1309 - acc: 0.9502 - val_loss: 0.1309
Epoch 21/30
7352/7352 [=====] - 35s 5ms/step - loss: 0.1280 - acc: 0.9502 - val_loss: 0.1280
Epoch 22/30
7352/7352 [=====] - 35s 5ms/step - loss: 0.1233 - acc: 0.9504 - val_loss: 0.1233
Epoch 23/30
7352/7352 [=====] - 35s 5ms/step - loss: 0.1221 - acc: 0.9508 - val_loss: 0.1221
Epoch 24/30
7352/7352 [=====] - 35s 5ms/step - loss: 0.1172 - acc: 0.9538 - val_loss: 0.1172
Epoch 25/30
7352/7352 [=====] - 38s 5ms/step - loss: 0.1187 - acc: 0.9531 - val_loss: 0.1187
Epoch 26/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1161 - acc: 0.9532 - val_loss: 0.1161
Epoch 27/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1402 - acc: 0.9521 - val_loss: 0.1402
Epoch 28/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1163 - acc: 0.9536 - val_loss: 0.1163
Epoch 29/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1366 - acc: 0.9486 - val_loss: 0.1366
Epoch 30/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1231 - acc: 0.9513 - val_loss: 0.1231
2947/2947 [=====] - 1s 482us/step

```

Layer (type)	Output Shape	Param #
=====		
lstm_2 (LSTM)	(None, 64)	18944
-----		
dropout_2 (Dropout)	(None, 64)	0
-----		
dense_2 (Dense)	(None, 6)	390
=====		

```

Total params: 19,334
Trainable params: 19,334
Non-trainable params: 0

```

```

-----
Train on 7352 samples, validate on 2947 samples
Epoch 1/30
7352/7352 [=====] - 38s 5ms/step - loss: 1.1909 - acc: 0.4815 - val_loss: 1.1909
Epoch 2/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.8578 - acc: 0.5934 - val_loss: 0.8578
Epoch 3/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.7158 - acc: 0.6548 - val_loss: 0.7158
Epoch 4/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.6563 - acc: 0.7359 - val_loss: 0.6563
Epoch 5/30
7352/7352 [=====] - 38s 5ms/step - loss: 0.5247 - acc: 0.8050 - val_loss: 0.5247
Epoch 6/30

```

7352/7352 [=====] - 38s 5ms/step - loss: 0.3771 - acc: 0.8769 - val\_loss: 0.3771  
Epoch 7/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2673 - acc: 0.9155 - val\_loss: 0.2673  
Epoch 8/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2216 - acc: 0.9222 - val\_loss: 0.2216  
Epoch 9/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2169 - acc: 0.9257 - val\_loss: 0.2169  
Epoch 10/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1829 - acc: 0.9355 - val\_loss: 0.1829  
Epoch 11/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1814 - acc: 0.9388 - val\_loss: 0.1814  
Epoch 12/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1636 - acc: 0.9419 - val\_loss: 0.1636  
Epoch 13/30  
7352/7352 [=====] - 38s 5ms/step - loss: 0.1737 - acc: 0.9400 - val\_loss: 0.1737  
Epoch 14/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1586 - acc: 0.9449 - val\_loss: 0.1586  
Epoch 15/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1422 - acc: 0.9459 - val\_loss: 0.1422  
Epoch 16/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1587 - acc: 0.9467 - val\_loss: 0.1587  
Epoch 17/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1336 - acc: 0.9504 - val\_loss: 0.1336  
Epoch 18/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1427 - acc: 0.9490 - val\_loss: 0.1427  
Epoch 19/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1423 - acc: 0.9489 - val\_loss: 0.1423  
Epoch 20/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1414 - acc: 0.9493 - val\_loss: 0.1414  
Epoch 21/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1332 - acc: 0.9474 - val\_loss: 0.1332  
Epoch 22/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1370 - acc: 0.9480 - val\_loss: 0.1370  
Epoch 23/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1298 - acc: 0.9514 - val\_loss: 0.1298  
Epoch 24/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1684 - acc: 0.9387 - val\_loss: 0.1684  
Epoch 25/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1460 - acc: 0.9450 - val\_loss: 0.1460  
Epoch 26/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1249 - acc: 0.9497 - val\_loss: 0.1249  
Epoch 27/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1340 - acc: 0.9501 - val\_loss: 0.1340  
Epoch 28/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1403 - acc: 0.9471 - val\_loss: 0.1403  
Epoch 29/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1313 - acc: 0.9532 - val\_loss: 0.1313  
Epoch 30/30

7352/7352 [=====] - 38s 5ms/step - loss: 0.1223 - acc: 0.9523 - val\_loss: 0.1223  
2947/2947 [=====] - 1s 490us/step

Layer (type)	Output Shape	Param #
lstm_3 (LSTM)	(None, 64)	18944
dropout_3 (Dropout)	(None, 64)	0
dense_3 (Dense)	(None, 6)	390

Total params: 19,334

Trainable params: 19,334

Non-trainable params: 0

Train on 7352 samples, validate on 2947 samples

Epoch 1/30

7352/7352 [=====] - 39s 5ms/step - loss: 1.2709 - acc: 0.4396 - val\_loss: 1.2709

Epoch 2/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.9985 - acc: 0.5336 - val\_loss: 0.9985

Epoch 3/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.8254 - acc: 0.6430 - val\_loss: 0.8254

Epoch 4/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.6177 - acc: 0.7556 - val\_loss: 0.6177

Epoch 5/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.4842 - acc: 0.8275 - val\_loss: 0.4842

Epoch 6/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.3198 - acc: 0.8998 - val\_loss: 0.3198

Epoch 7/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.2715 - acc: 0.9174 - val\_loss: 0.2715

Epoch 8/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.2208 - acc: 0.9251 - val\_loss: 0.2208

Epoch 9/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.2100 - acc: 0.9287 - val\_loss: 0.2100

Epoch 10/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.1847 - acc: 0.9397 - val\_loss: 0.1847

Epoch 11/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.1697 - acc: 0.9411 - val\_loss: 0.1697

Epoch 12/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.1787 - acc: 0.9380 - val\_loss: 0.1787

Epoch 13/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.1637 - acc: 0.9437 - val\_loss: 0.1637

Epoch 14/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.1527 - acc: 0.9416 - val\_loss: 0.1527

Epoch 15/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.1626 - acc: 0.9406 - val\_loss: 0.1626

Epoch 16/30

7352/7352 [=====] - 37s 5ms/step - loss: 0.1543 - acc: 0.9442 - val\_loss: 0.1543

```

Epoch 17/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1458 - acc: 0.9467 - val_loss: 0.1458
Epoch 18/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1391 - acc: 0.9505 - val_loss: 0.1391
Epoch 19/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1359 - acc: 0.9465 - val_loss: 0.1359
Epoch 20/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1804 - acc: 0.9445 - val_loss: 0.1804
Epoch 21/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.2048 - acc: 0.9441 - val_loss: 0.2048
Epoch 22/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.2478 - acc: 0.9264 - val_loss: 0.2478
Epoch 23/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1511 - acc: 0.9460 - val_loss: 0.1511
Epoch 24/30
7352/7352 [=====] - 36s 5ms/step - loss: 0.1383 - acc: 0.9516 - val_loss: 0.1383
Epoch 25/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1342 - acc: 0.9508 - val_loss: 0.1342
Epoch 26/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1315 - acc: 0.9509 - val_loss: 0.1315
Epoch 27/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1282 - acc: 0.9483 - val_loss: 0.1282
Epoch 28/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1411 - acc: 0.9489 - val_loss: 0.1411
Epoch 29/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1463 - acc: 0.9463 - val_loss: 0.1463
Epoch 30/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.1289 - acc: 0.9513 - val_loss: 0.1289
2947/2947 [=====] - 1s 494us/step

```

W0617 06:32:15.475296 140504591730560 nn\_ops.py:4224] Large dropout rate: 0.7 (>0.5). In TensorFlow

Layer (type)	Output Shape	Param #
lstm_4 (LSTM)	(None, 64)	18944
dropout_4 (Dropout)	(None, 64)	0
dense_4 (Dense)	(None, 6)	390

Total params: 19,334

Trainable params: 19,334

Non-trainable params: 0

Train on 7352 samples, validate on 2947 samples

Epoch 1/30  
7352/7352 [=====] - 39s 5ms/step - loss: 1.3130 - acc: 0.4402 - val\_loss: 1.3130  
Epoch 2/30  
7352/7352 [=====] - 37s 5ms/step - loss: 1.0343 - acc: 0.5465 - val\_loss: 1.0343  
Epoch 3/30  
7352/7352 [=====] - 38s 5ms/step - loss: 0.8394 - acc: 0.6143 - val\_loss: 0.8394  
Epoch 4/30  
7352/7352 [=====] - 38s 5ms/step - loss: 0.7459 - acc: 0.6659 - val\_loss: 0.7459  
Epoch 5/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.7040 - acc: 0.7108 - val\_loss: 0.7040  
Epoch 6/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.5503 - acc: 0.8098 - val\_loss: 0.5503  
Epoch 7/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.4508 - acc: 0.8615 - val\_loss: 0.4508  
Epoch 8/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.3289 - acc: 0.9038 - val\_loss: 0.3289  
Epoch 9/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2942 - acc: 0.9064 - val\_loss: 0.2942  
Epoch 10/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2701 - acc: 0.9168 - val\_loss: 0.2701  
Epoch 11/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2548 - acc: 0.9246 - val\_loss: 0.2548  
Epoch 12/30  
7352/7352 [=====] - 38s 5ms/step - loss: 0.2221 - acc: 0.9291 - val\_loss: 0.2221  
Epoch 13/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2188 - acc: 0.9334 - val\_loss: 0.2188  
Epoch 14/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1972 - acc: 0.9369 - val\_loss: 0.1972  
Epoch 15/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1874 - acc: 0.9365 - val\_loss: 0.1874  
Epoch 16/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1929 - acc: 0.9374 - val\_loss: 0.1929  
Epoch 17/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.2103 - acc: 0.9339 - val\_loss: 0.2103  
Epoch 18/30  
7352/7352 [=====] - 37s 5ms/step - loss: 0.1873 - acc: 0.9422 - val\_loss: 0.1873  
Epoch 19/30  
7352/7352 [=====] - 38s 5ms/step - loss: 0.2091 - acc: 0.9384 - val\_loss: 0.2091  
Epoch 20/30  
7352/7352 [=====] - 38s 5ms/step - loss: 0.1869 - acc: 0.9422 - val\_loss: 0.1869  
Epoch 21/30  
7352/7352 [=====] - 38s 5ms/step - loss: nan - acc: 0.9393 - val\_loss: 0.9393  
Epoch 22/30  
7352/7352 [=====] - 37s 5ms/step - loss: nan - acc: 0.9463 - val\_loss: 0.9463  
Epoch 23/30  
7352/7352 [=====] - 37s 5ms/step - loss: nan - acc: 0.9404 - val\_loss: 0.9404  
Epoch 24/30  
7352/7352 [=====] - 36s 5ms/step - loss: nan - acc: 0.9433 - val\_loss: 0.9433

```

Epoch 25/30
7352/7352 [=====] - 36s 5ms/step - loss: nan - acc: 0.9444 - val_loss:
Epoch 26/30
7352/7352 [=====] - 43s 6ms/step - loss: nan - acc: 0.9412 - val_loss:
Epoch 27/30
7352/7352 [=====] - 37s 5ms/step - loss: nan - acc: 0.9444 - val_loss:
Epoch 28/30
7352/7352 [=====] - 38s 5ms/step - loss: nan - acc: 0.9412 - val_loss:
Epoch 29/30
7352/7352 [=====] - 38s 5ms/step - loss: nan - acc: 0.9459 - val_loss:
Epoch 30/30
7352/7352 [=====] - 37s 5ms/step - loss: nan - acc: 0.9425 - val_loss:
2947/2947 [=====] - 1s 487us/step

```

W0617 06:51:05.217503 140504591730560 nn\_ops.py:4224] Large dropout rate: 0.9 (>0.5). In TensorFlow

Layer (type)	Output Shape	Param #
lstm_5 (LSTM)	(None, 64)	18944
dropout_5 (Dropout)	(None, 64)	0
dense_5 (Dense)	(None, 6)	390

```

Total params: 19,334
Trainable params: 19,334
Non-trainable params: 0

```

Train on 7352 samples, validate on 2947 samples

```

Epoch 1/30
7352/7352 [=====] - 39s 5ms/step - loss: 1.4367 - acc: 0.3902 - val_loss:
Epoch 2/30
7352/7352 [=====] - 37s 5ms/step - loss: 1.2442 - acc: 0.4649 - val_loss:
Epoch 3/30
7352/7352 [=====] - 37s 5ms/step - loss: 1.1148 - acc: 0.5132 - val_loss:
Epoch 4/30
7352/7352 [=====] - 37s 5ms/step - loss: 1.0714 - acc: 0.5355 - val_loss:
Epoch 5/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.9871 - acc: 0.5768 - val_loss:
Epoch 6/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.9340 - acc: 0.5944 - val_loss:
Epoch 7/30
7352/7352 [=====] - 37s 5ms/step - loss: 0.8972 - acc: 0.6145 - val_loss:
Epoch 8/30
7352/7352 [=====] - 37s 5ms/step - loss: 1.0225 - acc: 0.5970 - val_loss:

```

Epoch 9/30

4624/7352 [=====>...] - ETA: 12s - loss: 0.8600 - acc: 0.6122Buffered data was truncated

```
In [0]: print(dropout_List)
        print(".....")
        print(dropout_Accuracy)
```

[0.1, 0.3, 0.5, 0.7, 0.9]

...

[[0.26970847920445745, 0.9165252799457075], [0.5416957346667051, 0.8961655921275874], [0.4099368

```
In [0]: print(dropout_List)
        print(".....")
        print(dropout_Accuracy)
```

```
max_acc=0
best_dropout=0
y=[]
for i in range(len(dropout_List)):
    if dropout_Accuracy[i][1]>max_acc:
        max_acc=dropout_Accuracy[i][1]
        best_dropout=dropout_List[i]
    y.append(dropout_Accuracy[i][1])
    print("Number of Hidden Neurons : %d, Accuracy : %f"%(dropout_List[i],dropout_Accuracy[i][1]))

print(".....\n")
print("The best dropout value is : %f"%best_dropout)
```

[0.1, 0.3, 0.5, 0.7, 0.9]

...

[[0.26970847920445745, 0.9165252799457075], [0.5416957346667051, 0.8961655921275874], [0.4099368

Number of Hidden Neurons : 0, Accuracy : 0.916525

Number of Hidden Neurons : 0, Accuracy : 0.896166

Number of Hidden Neurons : 0, Accuracy : 0.908381

Number of Hidden Neurons : 0, Accuracy : 0.876485

Number of Hidden Neurons : 0, Accuracy : 0.487275

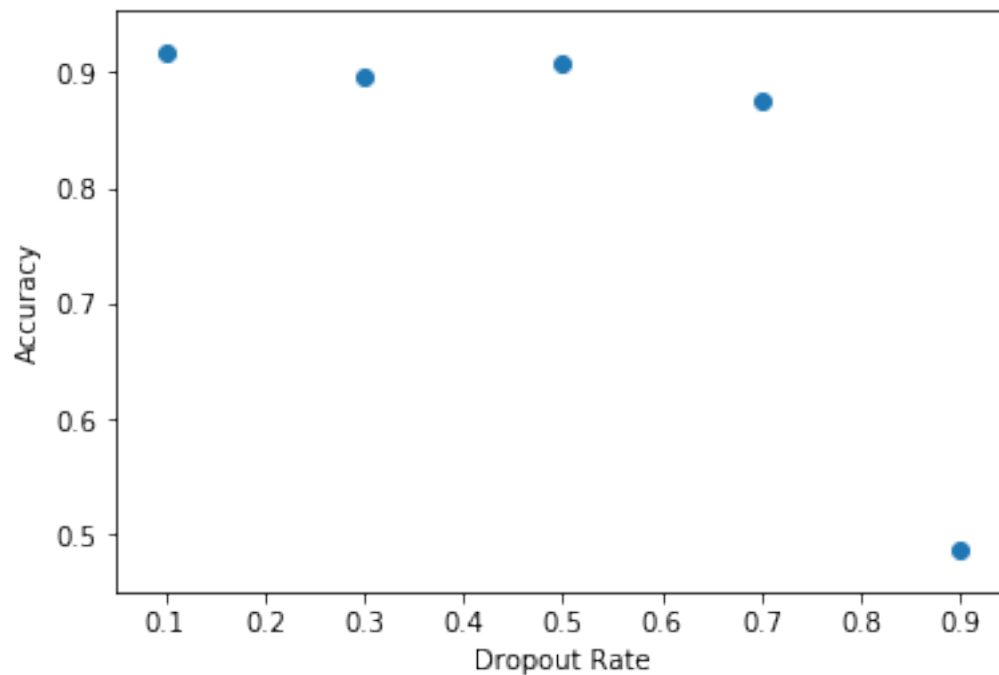
...

The best dropout value is : 0.100000

```
In [0]: import numpy as np
        import matplotlib.pyplot as plt
```

```
x = dropout_List
```

```
plt.scatter(x, y)
plt.xlabel('Dropout Rate')
plt.ylabel('Accuracy')
plt.show()
```



### 3 Assignment C) 2 Layer LSTM and Higher dropout

```
In [0]: from keras.layers import TimeDistributed
        from keras.layers.convolutional import Conv1D
        from keras.layers.convolutional import MaxPooling1D
        from keras.layers import Flatten
```

```
        from keras.layers import BatchNormalization
```

```
In [0]: epochs = 75
        kernel_size = 1
        pool_size = 2
```

```
        dropout_rate = 0.1
```

```
        f_act = 'relu'
```



```

In [23]: model = Sequential()
          model.add(Conv1D(512, (kernel_size), input_shape=(X_train.shape[1],X_train.shape[2]), a
          model.add(BatchNormalization())
          model.add(MaxPooling1D(pool_size=(pool_size)))
          model.add(Dropout(dropout_rate))

          model.add(Conv1D(64, (kernel_size), activation=f_act, padding='same'))
          model.add(BatchNormalization())
          model.add(MaxPooling1D(pool_size=(pool_size)))
          model.add(Dropout(dropout_rate))

          model.add(Conv1D(32, (kernel_size), activation=f_act, padding='same'))
          model.add(BatchNormalization())
          model.add(MaxPooling1D(pool_size=(pool_size)))
          model.add(Dropout(dropout_rate))

          model.add(LSTM(256, return_sequences=True))
          model.add(LSTM(128, return_sequences=True))
          model.add(LSTM(64))

          model.add(Dense(n_classes,activation='sigmoid'))

          #model.summary()

          model.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])

          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 1/75
7352/7352 [=====] - 76s 10ms/step - loss: 0.6044 - acc: 0.7349 - val_lo
Epoch 2/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.2458 - acc: 0.9142 - val_lo
Epoch 3/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.1767 - acc: 0.9347 - val_lo
Epoch 4/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.1652 - acc: 0.9395 - val_lo
Epoch 5/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.1463 - acc: 0.9402 - val_lo
Epoch 6/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.1405 - acc: 0.9475 - val_lo
Epoch 7/75

```

7352/7352 [=====] - 72s 10ms/step - loss: 0.1327 - acc: 0.9470 - val\_lo  
Epoch 8/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1376 - acc: 0.9440 - val\_lo  
Epoch 9/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1231 - acc: 0.9508 - val\_lo  
Epoch 10/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1273 - acc: 0.9482 - val\_lo  
Epoch 11/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1260 - acc: 0.9490 - val\_lo  
Epoch 12/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1253 - acc: 0.9478 - val\_lo  
Epoch 13/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1237 - acc: 0.9468 - val\_lo  
Epoch 14/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1246 - acc: 0.9487 - val\_lo  
Epoch 15/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1303 - acc: 0.9474 - val\_lo  
Epoch 16/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1215 - acc: 0.9495 - val\_lo  
Epoch 17/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1140 - acc: 0.9543 - val\_lo  
Epoch 18/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1081 - acc: 0.9540 - val\_lo  
Epoch 19/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1114 - acc: 0.9559 - val\_lo  
Epoch 20/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1125 - acc: 0.9533 - val\_lo  
Epoch 21/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1108 - acc: 0.9540 - val\_lo  
Epoch 22/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1160 - acc: 0.9520 - val\_lo  
Epoch 23/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1189 - acc: 0.9494 - val\_lo  
Epoch 24/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1029 - acc: 0.9581 - val\_lo  
Epoch 25/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1085 - acc: 0.9561 - val\_lo  
Epoch 26/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.1045 - acc: 0.9584 - val\_lo  
Epoch 27/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.1038 - acc: 0.9584 - val\_lo  
Epoch 28/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.0988 - acc: 0.9607 - val\_lo  
Epoch 29/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.0920 - acc: 0.9634 - val\_lo  
Epoch 30/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0936 - acc: 0.9627 - val\_lo  
Epoch 31/75

7352/7352 [=====] - 71s 10ms/step - loss: 0.0985 - acc: 0.9601 - val\_lo  
Epoch 32/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0959 - acc: 0.9619 - val\_lo  
Epoch 33/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0933 - acc: 0.9642 - val\_lo  
Epoch 34/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0889 - acc: 0.9637 - val\_lo  
Epoch 35/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0952 - acc: 0.9635 - val\_lo  
Epoch 36/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0887 - acc: 0.9642 - val\_lo  
Epoch 37/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0908 - acc: 0.9645 - val\_lo  
Epoch 38/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0866 - acc: 0.9664 - val\_lo  
Epoch 39/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0799 - acc: 0.9682 - val\_lo  
Epoch 40/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0926 - acc: 0.9656 - val\_lo  
Epoch 41/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0804 - acc: 0.9695 - val\_lo  
Epoch 42/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0847 - acc: 0.9679 - val\_lo  
Epoch 43/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0749 - acc: 0.9721 - val\_lo  
Epoch 44/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0834 - acc: 0.9668 - val\_lo  
Epoch 45/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.0744 - acc: 0.9694 - val\_lo  
Epoch 46/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0689 - acc: 0.9708 - val\_lo  
Epoch 47/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.0731 - acc: 0.9723 - val\_lo  
Epoch 48/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0761 - acc: 0.9690 - val\_lo  
Epoch 49/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0659 - acc: 0.9737 - val\_lo  
Epoch 50/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0748 - acc: 0.9733 - val\_lo  
Epoch 51/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0634 - acc: 0.9739 - val\_lo  
Epoch 52/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.0671 - acc: 0.9728 - val\_lo  
Epoch 53/75  
7352/7352 [=====] - 72s 10ms/step - loss: 0.0681 - acc: 0.9736 - val\_lo  
Epoch 54/75  
7352/7352 [=====] - 71s 10ms/step - loss: 0.0602 - acc: 0.9765 - val\_lo  
Epoch 55/75

```

7352/7352 [=====] - 71s 10ms/step - loss: 0.0579 - acc: 0.9770 - val_lo
Epoch 56/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0590 - acc: 0.9755 - val_lo
Epoch 57/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0600 - acc: 0.9765 - val_lo
Epoch 58/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0619 - acc: 0.9758 - val_lo
Epoch 59/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0594 - acc: 0.9777 - val_lo
Epoch 60/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0530 - acc: 0.9782 - val_lo
Epoch 61/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0599 - acc: 0.9765 - val_lo
Epoch 62/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0532 - acc: 0.9793 - val_lo
Epoch 63/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0473 - acc: 0.9803 - val_lo
Epoch 64/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0585 - acc: 0.9762 - val_lo
Epoch 65/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0506 - acc: 0.9804 - val_lo
Epoch 66/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0502 - acc: 0.9803 - val_lo
Epoch 67/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0483 - acc: 0.9814 - val_lo
Epoch 68/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0486 - acc: 0.9801 - val_lo
Epoch 69/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0506 - acc: 0.9804 - val_lo
Epoch 70/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0427 - acc: 0.9835 - val_lo
Epoch 71/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0418 - acc: 0.9823 - val_lo
Epoch 72/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0465 - acc: 0.9805 - val_lo
Epoch 73/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0388 - acc: 0.9835 - val_lo
Epoch 74/75
7352/7352 [=====] - 71s 10ms/step - loss: 0.0427 - acc: 0.9833 - val_lo
Epoch 75/75
7352/7352 [=====] - 72s 10ms/step - loss: 0.0455 - acc: 0.9811 - val_lo
2947/2947 [=====] - 4s 1ms/step

```

```
In [0]: #model=Sequential()
```

```
#model.add(LSTM(200,return_sequences=True,input_shape=(timesteps,input_dim)))
```

```

#model.add(LSTM(100))

#model.add(Dropout(0.8))

#model.add(Dense(n_classes,activation='sigmoid'))

#model.summary()

#model.compile(loss='categorical_crossentropy',optimizer='rmsprop',metrics=['accuracy'])

#model.fit(X_train,
#          Y_train,
#          batch_size=batch_size,
#          validation_data=(X_test, Y_test),
#          epochs=epochs)

#score = model.evaluate(X_test, Y_test)

In [0]: #print("Best: %f using %s" % (grid_result.best_score_, grid_result.best_params_))

In [0]:

In [28]: # Confusion Matrix
         print(confusion_matrix(Y_test, model.predict(X_test)))

```

Pred \ True	LAYING	SITTING	...	WALKING_DOWNSTAIRS	WALKING_UPSTAIRS
LAYING	537	0	...	0	0
SITTING	5	433	...	0	3
STANDING	0	27	...	0	0
WALKING	0	0	...	3	0
WALKING_DOWNSTAIRS	0	0	...	415	2
WALKING_UPSTAIRS	0	0	...	0	442

[6 rows x 6 columns]

```

In [26]: score = model.evaluate(X_test, Y_test)

2947/2947 [=====] - 4s 1ms/step

In [27]: print(score)

[0.1823121076857176, 0.9637048147261827]

```

## 4 Conclusions

```
In [29]: from prettytable import PrettyTable
```

```
x = PrettyTable()
```

```
x.field_names = ["Excercise", "Task", "Remarks", "Accuracy"]
```

```
x.add_row(["A", "Hypertuning Hidden Neurons", "Best value found at 64", 0.904988])
```

```
x.add_row(["B", "Hypertuning Dropout value", "Best value found at 0.1", 0.916525])
```

```
x.add_row(["C", "To improve accuracy to beyond 96%", "CNN + LSTM used", 0.9637048147261827])
```

```
print(x)
```

Excercise	Task	Remarks	Accuracy
A	Hypertuning Hidden Neurons	Best value found at 64	0.904988
B	Hypertuning Dropout value	Best value found at 0.1	0.916525
C	To improve accuracy to beyond 96%	CNN + LSTM used	0.9637048147261827

```
In [0]:
```