Graph Plot for LSTM Models

December 12, 2018

```
In [ ]: # Depression Analysis in Bangla with LSTM-RNN RESULTS
                                       # copyright (c) ABDUL HASIB UDDIN <abdulhasibuddin@gmail.com>
In [1]: import matplotlib.pyplot as plt
                                      from scipy.interpolate import spline
                                       import numpy as np
In [2]: # LSTM MODEL STATISTICS:
In [3]: # lstm validation accuracies:
                                      lstm_with_unique_1 = [0.445, 0.409, 0.327, 0.500, 0.464, 0.445, 0.309, 0.500, 0.427, 0.455, 0.500]
                                      lstm_with_unique_2 = [0.436,0.345,0.273,0.555,0.455,0.436,0.500,0.500,0.436,0.436,0.500]
                                      lstm_with_unique_3 = [0.445,0.318,0.309,0.491,0.400,0.418,0.500,0.491,0.391,0.400,0.65
                                      lstm_with_unique_4 = [0.345,0.336,0.355,0.745,0.336,0.364,0.555,0.609,0.336,0.418,0.736]
                                      lstm_with_unique_5 = [0.291,0.382,0.473,0.500,0.382,0.355,0.618,0.445,0.482,0.318,0.68
                                      lstm_with_unique_6 = [0.500, 0.430, 0.390]
                                      lstm_with_unique_7 = [0.450, 0.420, 0.440, 0.440, 0.560, 0.530, 0.550]
                                      lstm_with_unique_8 = [0.435,0.496,0.435,0.400,0.426,0.270,0.522,0.504,0.548,0.461,0.426]
                                      lstm_with_unique_9 = [0.475,0.475,0.466,0.517,0.449,0.466,0.466,0.466,0.458,0.398,0.376
                                      lstm_with_unique_10 = [0.410, 0.500, 0.500, 0.470, 0.510, 0.410, 0.450, 0.620, 0.660, 0.480, 0.500, 0.480, 0.580]
                                      lstm_with_unique_11 = [0.430,0.400,0.500,0.480,0.510,0.500,0.430]
                                      lstm_with_unique_12 = [0.391,0.435,0.365,0.357,0.426,0.322,0.504,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0.487,0
                                      lstm_with_unique_13 = [0.478, 0.470, 0.496, 0.452, 0.417, 0.365, 0.504, 0.530, 0.617, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478, 0.478
                                      lstm_with_unique_14 = [0.357, 0.426, 0.365, 0.357, 0.391, 0.270, 0.539, 0.496, 0.478, 0.443, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498, 0.498
                                      print(len(lstm_with_unique_9))
75
In [4]: # iterations (x-axis):
                                     x_axis_1 = []
                                      for iter_no in range(1,470+1):
                                                          if iter_no%25 == 0:
                                                                             x_axis_1.append(iter_no)
```

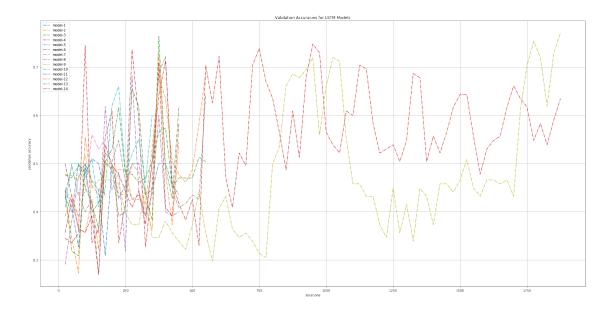
 $x_axis_2 = []$

for iter_no in range(1,470+1):

```
if iter_no%25 == 0:
        x_axis_2.append(iter_no)
x_axis_3 = []
for iter_no in range(1,470+1):
    if iter no%25 == 0:
        x_axis_3.append(iter_no)
x_axis_4 = []
for iter_no in range(1,470+1):
    if iter_no\%25 == 0:
        x_axis_4.append(iter_no)
x_axis_5 = []
for iter_no in range(1,470+1):
    if iter_no%25 == 0:
        x_axis_5.append(iter_no)
x_axis_6 = []
for iter_no in range(1,90+1):
    if iter_no%25 == 0:
        x_axis_6.append(iter_no)
x_axis_7 = []
for iter_no in range(1,180+1):
    if iter_no%25 == 0:
        x_axis_7.append(iter_no)
x_axis_8 = []
for iter_no in range(1,560+1):
    if iter_no\%25 == 0:
        x_axis_8.append(iter_no)
x_axis_9 = []
for iter_no in range(1,1880+1):
    if iter_no%25 == 0:
        x_axis_9.append(iter_no)
x_axis_10 = []
for iter_no in range(1,370+1):
    if iter_no%25 == 0:
        x_axis_10.append(iter_no)
x axis 11 = []
for iter_no in range(1,185+1):
    if iter no%25 == 0:
        x_axis_11.append(iter_no)
x_axis_12 = []
for iter_no in range(1,560+1):
    if iter_no%25 == 0:
        x_axis_12.append(iter_no)
x_axis_13 = []
for iter_no in range(1,560+1):
    if iter_no\%25 == 0:
        x_axis_13.append(iter_no)
x_axis_14 = []
for iter_no in range(1,1880+1):
```

```
if iter_no%25 == 0:
               x_axis_14.append(iter_no)
In [5]: x_list = [x_axis_1,x_axis_2,x_axis_3,x_axis_4,x_axis_5,x_axis_6,x_axis_7,x_axis_8,x_ax
       model_list = [lstm_with_unique_1,lstm_with_unique_2,lstm_with_unique_3,lstm_with_unique
       required_iteration_list = [470,470,470,470,470,90,180,560,1880,370,185,560,560,1880]
       required_epoch_list = [5,5,5,5,5,5,10,3,2,10,5,3,3,10]
       test_acc_list = [50.0,73.6,72.7,69.1,70.9,36.0,59.0,73.9,77.1,57.0,40.0,73.9,42.6,62.6]
       print(len(x_list))
       print(len(model_list))
       print(len(required_iteration_list))
       print(len(required_epoch_list))
       print(len(test_acc_list))
14
14
14
14
14
In [6]: # average validation accuracies:
       avg_val_acc_list = []
       for model in model_list:
           avg_val_acc_list.append(sum(model)/len(model))
       print(len(avg_val_acc_list))
       print(avg_val_acc_list)
In [7]: best_model_val_acc_1 = lstm_with_unique_9 # 77.1%
       best_model_val_loss_1 = [0.282,0.251,0.283,0.250,0.253,0.202,0.254,0.254,0.358,0.189,0
       best_model_val_acc_2 = lstm_with_unique_8 # 73.9%
       best_model_val_loss_2 = [0.245,0.198,0.221,0.251,0.183,0.268,0.205,0.225,0.257,0.238,0
       print(len(best_model_val_acc_1))
       print(len(best_model_val_loss_1))
       print(len(best_model_val_acc_2))
       print(len(best_model_val_loss_2))
75
75
22
22
In []:
```

```
In [8]: for i in range(0,len(x_list)):
            x_axis_name = 'len(x_axis_'+str(i+1)+') ='
           model_name = '; len(lstm_with_unique_'+str(i+1)+') ='
            x_axis_length = len(x_list[i])
           model length = len(model list[i])
            print(x_axis_name,x_axis_length, model_name,model_length, '; status =',x_axis_length)
len(x_axis_1) = 18 ; len(lstm_with_unique_1) = 18 ; status = True
len(x_axis_2) = 18 ; len(lstm_with_unique_2) = 18 ; status = True
len(x_axis_3) = 18 ; len(lstm_with_unique_3) = 18 ; status = True
len(x_axis_4) = 18 ; len(lstm_with_unique_4) = 18 ; status = True
len(x_axis_5) = 18 ; len(lstm_with_unique_5) = 18 ; status = True
len(x_axis_6) = 3 ; len(lstm_with_unique_6) = 3 ; status = True
len(x_axis_7) = 7 ; len(lstm_with_unique_7) = 7 ; status = True
len(x_axis_8) = 22; len(lstm_with_unique_8) = 22; status = True
len(x_axis_9) = 75 ; len(lstm_with_unique_9) = 75 ; status = True
len(x_axis_10) = 14 ; len(lstm_with_unique_10) = 14 ; status = True
len(x_axis_11) = 7 ; len(lstm_with_unique_11) = 7 ; status = True
len(x_axis_12) = 22 ; len(lstm_with_unique_12) = 22 ; status = True
len(x_axis_13) = 22 ; len(lstm_with_unique_13) = 22 ; status = True
len(x_axis_14) = 75 ; len(lstm_with_unique_14) = 75 ; status = True
In [9]: # lstm plotting validation accuracies against iterations:
        linestyle='-.'
       linewidth = 2.5
       plt.figure(figsize=(30,15))
       plt.title('Validation Accuracies for LSTM Models')
       plt.xlabel('iterations')
       plt.ylabel('validation accuracy')
        for i in range(0,len(x_list)):
            label = "model-"+str(i+1)
            x = x_list[i]
            y = model_list[i]
           plt.plot(x, y, linestyle=linestyle, label=label)
       plt.grid(True)
       plt.legend()
       plt.show()
```

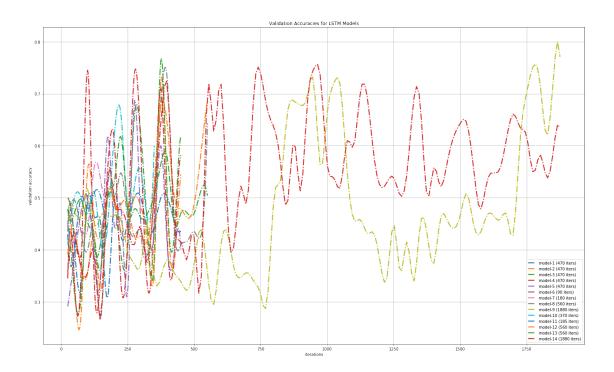


```
In [10]: # lstm plotting validation accuracies against iterations:
         smoothing_factor = 200
         linestyle='-.'
         linewidth = 2.5
         #plt.figure(figsize=(13,8))
         plt.figure(figsize=(25,15))
         plt.title('Validation Accuracies for LSTM Models')
         plt.xlabel('iterations')
         plt.ylabel('validation accuracy')
         for i in range(0,len(x_list)):
             label = "model-"+str(i+1)+" ("+str(required_iteration_list[i])+" iters)"
             x = x_list[i]
             y = model_list[i]
             x_sm = np.array(x)
             y_{sm} = np.array(y)
             x_smooth = np.linspace(x_sm.min(), x_sm.max(), smoothing_factor)
             y_smooth = spline(x, y, x_smooth)
             plt.plot(x_smooth, y_smooth, linestyle=linestyle, linewidth=linewidth, label=labe
        plt.grid(True)
         plt.legend()
         plt.savefig('images\lstm_image_1_plotting _validation _accuracies _against _iteration
         plt.show()
```

c:\python36\lib\site-packages\ipykernel_launcher.py:19: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.

c:\python36\lib\site-packages\scipy\interpolate\interpolate.py:2752: LinAlgWarning: scipy.lina Ill-conditioned matrix detected. Result is not guaranteed to be accurate. Reciprocal condition number5.244205e-17

p = scipy.linalg.solve(Q, tmp)

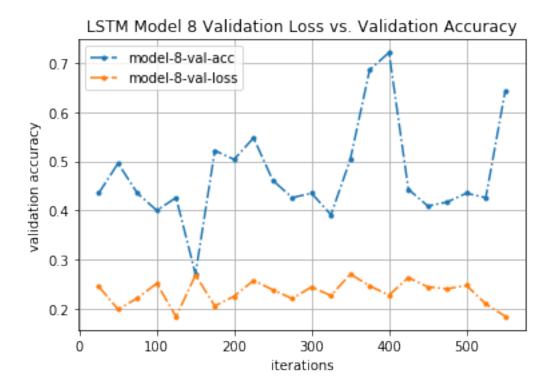


In []:

In [11]: # lstm best model validation loss vs validation accuracy:

```
#plt.figure(figsize=(30,15))
plt.title('LSTM Model 8 Validation Loss vs. Validation Accuracy')
plt.xlabel('iterations')
plt.ylabel('validation accuracy')

plt.plot(x_axis_8, best_model_val_acc_2, marker='o', markersize=3, linestyle=linestyle
plt.plot(x_axis_8, best_model_val_loss_2, marker='o', markersize=3, linestyle=linestyle
plt.grid(True)
plt.legend()
plt.show()
```

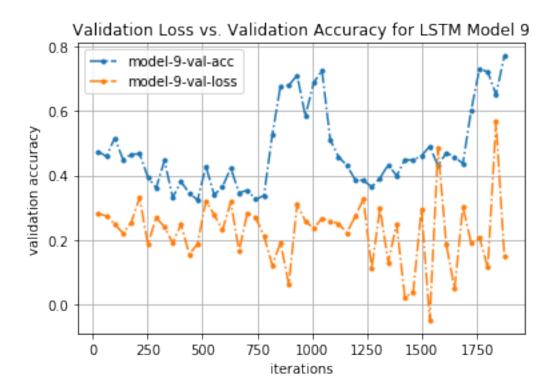


```
In [12]: smoothing_factor = 50
         linestyle='-.'
         \#plt.figure(figsize=(30,15))
         plt.title('Validation Loss vs. Validation Accuracy for LSTM Model 9')
         plt.xlabel('iterations')
         plt.ylabel('validation accuracy')
         x = x_axis_9
         y = best_model_val_acc_1
         x_sm = np.array(x)
         y_{sm} = np.array(y)
         x_smooth = np.linspace(x_sm.min(), x_sm.max(), smoothing_factor)
         y_smooth = spline(x, y, x_smooth)
         plt.plot(x_smooth, y_smooth, marker='o', markersize=3, linestyle=linestyle, label='mo'
         y = best_model_val_loss_1
         x_sm = np.array(x)
         y_{sm} = np.array(y)
         x_smooth = np.linspace(x_sm.min(), x_sm.max(), smoothing_factor)
         y_smooth = spline(x, y, x_smooth)
         plt.plot(x_smooth, y_smooth, marker='o', markersize=3, linestyle=linestyle, label='mo'
         plt.grid(True)
```

```
plt.legend()
plt.savefig('images\lstm_image_2_1_best_model_9_validation_loss_vs_validation _accurace
plt.show()
```

c:\python36\lib\site-packages\ipykernel_launcher.py:14: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.

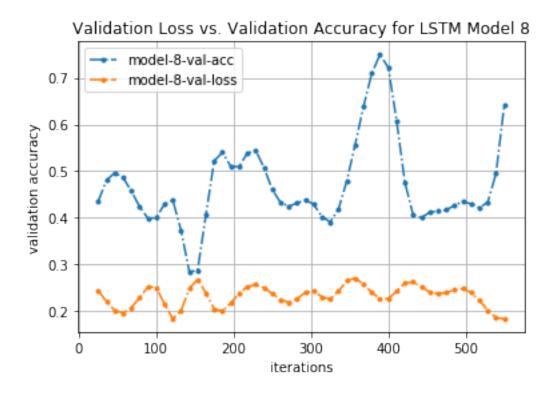
c:\python36\lib\site-packages\ipykernel_launcher.py:21: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.



```
y_smooth = spline(x, y, x_smooth)
plt.plot(x_smooth, y_smooth, marker='o', markersize=3, linestyle=linestyle, label='mooty
y = best_model_val_loss_2
x_sm = np.array(x)
y_sm = np.array(y)
x_smooth = np.linspace(x_sm.min(), x_sm.max(), smoothing_factor)
y_smooth = spline(x, y, x_smooth)
plt.plot(x_smooth, y_smooth, marker='o', markersize=3, linestyle=linestyle, label='mooth.plt.grid(True)
plt.grid(True)
plt.legend()
plt.savefig('images\lstm_image_2_2_best_model_8_validation_loss_vs_validation_accurately.show()
```

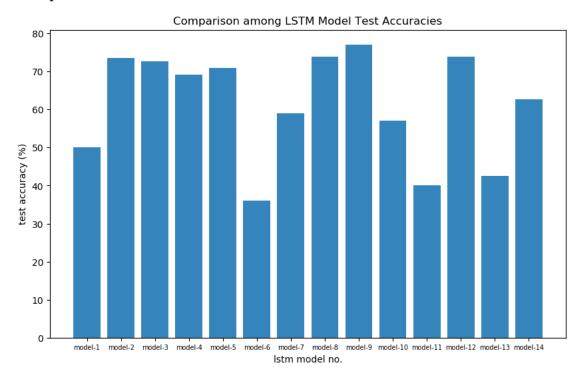
c:\python36\lib\site-packages\ipykernel_launcher.py:14: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.

c:\python36\lib\site-packages\ipykernel_launcher.py:21: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.



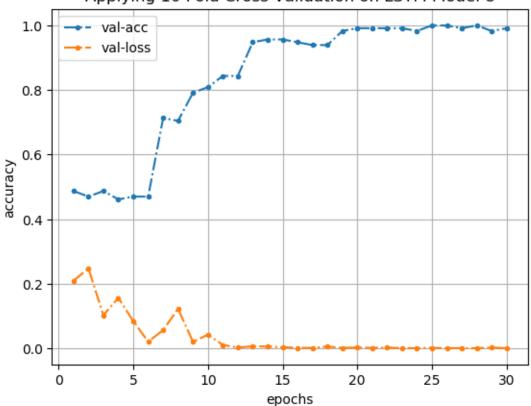
In []:

```
In [14]: # comparing lstm model test accuracies (bar chart):
         plt.rcdefaults()
         plt.figure(figsize=(10,6))
         objects = []
         for i in range(0,len(model_list)):
             object_name = "model-"+str(i+1)
             objects.append(object_name)
         y_pos = np.arange(len(objects))
         performance = test_acc_list ###
         plt.bar(y_pos, performance, align='center', alpha=0.9)
         plt.xticks(y_pos, objects)
         #plt.tick_params(axis='both', which='major', labelsize=10)
         plt.tick_params(axis='x', which='major', labelsize=7)
         plt.xlabel('lstm model no.')
         plt.ylabel('test accuracy (%)')
         plt.title('Comparison among LSTM Model Test Accuracies')
         plt.savefig('images\lstm_image_3_comparing_lstm_model_test_accuracies_bar_chart.png',
         plt.show()
```



```
lstm_10_fold_cross_val_loss_list = [0.209,0.247,0.102,0.156,0.084,0.020,0.056,0.122,0
                               print(len(lstm_10_fold_cross_val_acc_list))
                               print(len(lstm_10_fold_cross_val_loss_list))
30
30
In [16]: smoothing_factor = 200
                               linestyle='-.'
                               #plt.figure(figsize=(30,15))
                               plt.title('Validation accuracy vs. Validation Loss for \nApplying 10 Fold Cross Validation Loss for \napplying 
                               plt.xlabel('epochs')
                              plt.ylabel('accuracy')
                               x = [i for i in range (1,len(lstm_10_fold_cross_val_acc_list)+1)]
                               y = lstm_10_fold_cross_val_acc_list
                              plt.plot(x, y, marker='o', markersize=3, linestyle=linestyle, label='val-acc')
                               y = lstm_10_fold_cross_val_loss_list
                              plt.plot(x, y, marker='o', markersize=3, linestyle=linestyle, label='val-loss')
                              plt.grid(True)
                              plt.legend()
                               \#plt.savefig('images \setminus lstm\_image\_4\_accuracy\_vs\_loss\_for\_10\_fold\_cross\_validation.png',
                              plt.show()
```

Validation accuracy vs. Validation Loss for Applying 10 Fold Cross Validation on LSTM Model 8

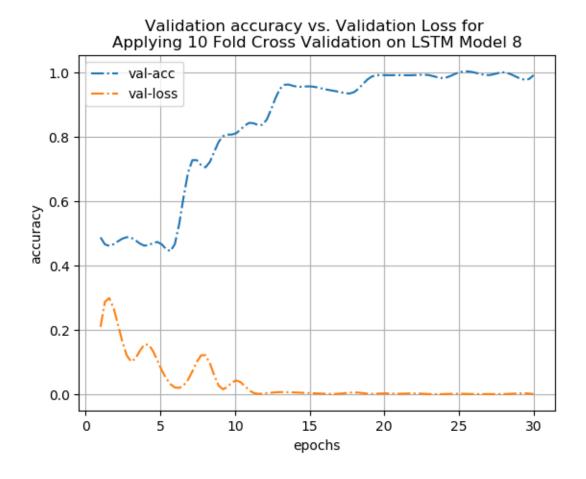


```
In [17]: smoothing_factor = 100
                                      linestyle='-.'
                                      #plt.figure(figsize=(30,15))
                                      plt.title('Validation accuracy vs. Validation Loss for \nApplying 10 Fold Cross Validation Loss for \napplying 
                                      plt.xlabel('epochs')
                                      plt.ylabel('accuracy')
                                      x = [i for i in range (1,len(lstm_10_fold_cross_val_acc_list)+1)]
                                      y = lstm_10_fold_cross_val_acc_list
                                      x_sm = np.array(x)
                                      y_sm = np.array(y)
                                      x_smooth = np.linspace(x_sm.min(), x_sm.max(), smoothing_factor)
                                      y_smooth = spline(x, y, x_smooth)
                                      plt.plot(x_smooth, y_smooth, marker='', markersize=1, linestyle=linestyle, label='val
                                      y = lstm_10_fold_cross_val_loss_list
                                      x_sm = np.array(x)
                                      y_sm = np.array(y)
```

```
x_smooth = np.linspace(x_sm.min(), x_sm.max(), smoothing_factor)
y_smooth = spline(x, y, x_smooth)
plt.plot(x_smooth, y_smooth, marker='', markersize=1, linestyle=linestyle, label='val-
plt.grid(True)
plt.legend()
plt.savefig('images\lstm_image_4_accuracy_vs_loss_for_10_fold_cross_validation.png', legend()
```

c:\python36\lib\site-packages\ipykernel_launcher.py:14: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.

c:\python36\lib\site-packages\ipykernel_launcher.py:21: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.



In [18]: # LSTM 10 FOLD CROSS VALIDATION MODEL ACCURACY::
 lstm_10_fold_cross_val_folds_acc_list = [0.4870,0.4696,0.7913,0.8435,0.9565,0.9391,0.9391]
 lstm_10_fold_cross_val_model_acc = sum(lstm_10_fold_cross_val_folds_acc_list)/len(lstm_print('LSTM 10 FOLD CROSS VALIDATION MODEL ACCURACY = ',lstm_10_fold_cross_val_model_acc_list)

In []: