Graph Plot for GRU Models

December 12, 2018

```
In [ ]: # Depression Analysis in Bangla with GRU-RNN RESULTS
                                        # copyright (c) ABDUL HASIB UDDIN <abdulhasibuddin@gmail.com>
                                        # LICENSE: GNU General Public License v3.0
In [1]: import matplotlib.pyplot as plt
                                      from scipy.interpolate import spline
                                       import numpy as np
In [2]: # GRU MODEL STATISTICS:
In [3]: # gru validation accuracies:
                                      gru\_unique\_1 = [0.427, 0.318, 0.300, 0.545, 0.500, 0.400, 0.509, 0.673, 0.464, 0.436, 0.545, 0.59]
                                      gru\_unique\_2 = [0.436, 0.391, 0.309, 0.709, 0.445, 0.427, 0.500, 0.709, 0.391, 0.409, 0.627, 0.730]
                                      gru_unique_3 = [0.355,0.355,0.309,0.700,0.364,0.473,0.509,0.618,0.427,0.436,0.718,0.518
                                      gru\_unique\_4 = [0.291, 0.273, 0.318, 0.582, 0.455, 0.418, 0.664, 0.736, 0.464, 0.445, 0.664, 0.485]
                                      gru_unique_5 = [0.336,0.318,0.336,0.318,0.500,0.455,0.682,0.509,0.464,0.500,0.745,0.518
                                      gru\_unique\_6 = [0.490, 0.410, 0.680]
                                      gru_unique_7 = [0.560,0.400,0.610,0.580,0.620,0.590,0.560]
                                      gru_unique 8 = [0.435, 0.296, 0.348, 0.470, 0.409, 0.409, 0.730, 0.539, 0.374, 0.426, 0.426, 0.513]
                                      gru_unique 9 = [0.432, 0.534, 0.551, 0.441, 0.398, 0.415, 0.458, 0.356, 0.398, 0.364, 0.356, 0.40]
                                      gru\_unique\_10 = [0.420, 0.340, 0.510, 0.390, 0.430, 0.530, 0.550, 0.620, 0.520, 0.480, 0.600, 0.570]
                                      gru_unique_11 = [0.420,0.360,0.650,0.390,0.460,0.760,0.510]
                                      gru\_unique\_12 = [0.470, 0.383, 0.365, 0.409, 0.461, 0.417, 0.713, 0.696, 0.522, 0.478, 0.452, 0.698, 0.452, 0.698, 0.452, 0.458, 0.452, 0.458, 0.452, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.458, 0.45
                                      gru\_unique\_13 = [0.504, 0.452, 0.443, 0.287, 0.391, 0.496, 0.670, 0.504, 0.504, 0.470, 0.348, 0.470, 0.348, 0.470, 0.348, 0.470, 0.470, 0.348, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.470, 0.47
                                      gru_unique_14 = [0.504,0.330,0.330,0.435,0.443,0.435,0.722,0.391,0.365,0.470,0.417,0.443,0.435,0.722,0.391,0.365,0.470,0.417,0.443,0.435,0.443,0.435,0.722,0.391,0.365,0.470,0.417,0.443,0.435,0.443,0.435,0.722,0.391,0.365,0.470,0.417,0.443,0.435,0.443,0.435,0.722,0.391,0.365,0.470,0.417,0.443,0.435,0.443,0.435,0.443,0.435,0.443,0.435,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.443,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0.444,0
                                      print(len(gru_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)

 $x_axis_2 = []$

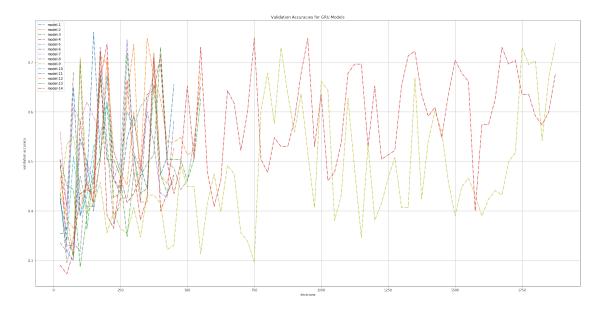
```
for iter_no in range(1,470+1):
    if iter_no%25 == 0:
        x_axis_2.append(iter_no)
x_axis_3 = []
for iter_no in range(1,470+1):
    if iter_no%25 == 0:
        x_axis_3.append(iter_no)
x_axis_4 = []
for iter_no in range(1,470+1):
    if iter_no%25 == 0:
        x_axis_4.append(iter_no)
x_axis_5 = []
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 = [gru_unique_1,gru_unique_2,gru_unique_3,gru_unique_4,gru_unique_5,gru_unique_5
        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 = [59.1,70.0,67.3,74.5,69.1,52.0,61.0,75.7,70.3,57.0,61.0,74.8,69.6,56.5]
       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)
14
[0.49994444444444444446, 0.482833333333333, 0.4788333333333344, 0.485777777]
In [7]: best_model_val_acc = gru_unique_8
        best_model_val_loss = [0.247,0.182,0.202,0.240,0.205,0.265,0.066,0.272,0.192,0.190,0.1
        print(len(best_model_val_acc))
        print(len(best_model_val_loss))
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(gru_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_leng'
len(x_axis_1) = 18 ; len(gru_unique_1) = 18 ; status = True
len(x_axis_2) = 18 ; len(gru_unique_2) = 18 ; status = True
len(x_axis_3) = 18 ; len(gru_unique_3) = 18 ; status = True
len(x_axis_4) = 18 ; len(gru_unique_4) = 18 ; status = True
len(x_axis_5) = 18 ; len(gru_unique_5) = 18 ; status = True
len(x_axis_6) = 3 ; len(gru_unique_6) = 3 ; status = True
len(x_axis_7) = 7 ; len(gru_unique_7) = 7 ; status = True
len(x_axis_8) = 22 ; len(gru_unique_8) = 22 ; status = True
len(x_axis_9) = 75 ; len(gru_unique_9) = 75 ; status = True
len(x_axis_10) = 14 ; len(gru_unique_10) = 14 ; status = True
len(x_axis_11) = 7 ; len(gru_unique_11) = 7 ; status = True
len(x_axis_12) = 22 ; len(gru_unique_12) = 22 ; status = True
len(x_axis_13) = 22 ; len(gru_unique_13) = 22 ; status = True
len(x_axis_14) = 75 ; len(gru_unique_14) = 75 ; status = True
In [9]: '''
       x_axis = x_axis_9
        y \ axis = []
        for i in range(10+1):
            y_axis.append(i/10)
        print(y_axis)
        print(len(x_axis))
        print(len(y_axis))
        IIII
Out[9]: '\nx_axis = x_axis_9\ny_axis = []\nfor i in range(10+1):\n
                                                                      y_{axis.append(i/10)\n}
In [10]: # gru plotting validation accuracies against iterations:
         linestyle='-.'
         linewidth = 2.5
         plt.figure(figsize=(30,15))
         plt.title('Validation Accuracies for GRU Models')
         plt.xlabel('iterations')
         plt.ylabel('validation accuracy')
         #plt.plot(x_axis, y_axis_dummy)
         #plt.plot(x_axis_dummy, y_axis)
         #plt.xticks(x_axis)
         #plt.yticks(y_axis)
         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.plot(x, y, marker='o', markersize=5, linestyle=linestyle, label=label)

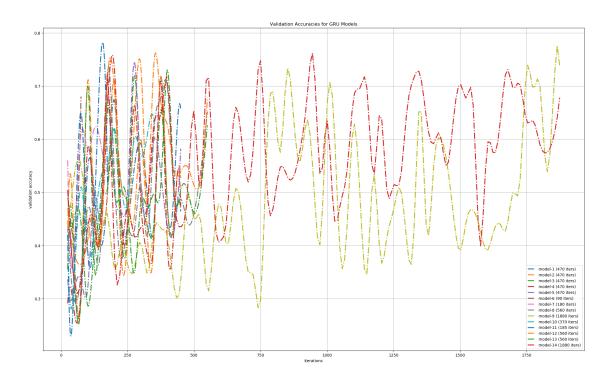
#plt.axis([min(gru_x_axis), max(gru_x_axis), 0, 1])
plt.grid(True)
plt.legend()
plt.show()
```



```
In [20]: # gru plotting validation accuracies against iterations:
         smoothing_factor = 200
         linestyle='-.'
         linewidth = 2.5
         legend_properties = {'weight':'bold'}
         #plt.savefig('gru_plotting _validation _accuracies _against _iterations.png')
         #plt.figure(figsize=(13,8))
         plt.figure(figsize=(25,15))
         plt.title('Validation Accuracies for GRU Models')
         plt.xlabel('iterations')
         plt.ylabel('validation accuracy')
         x = x_axis_1
         y = gru\_unique\_1
         x\_sm = np.array(x)
         y\_sm = np.array(y)
         x_{smooth} = np.linspace(x_{sm.min}(), x_{sm.max}(), smoothing_factor)
```

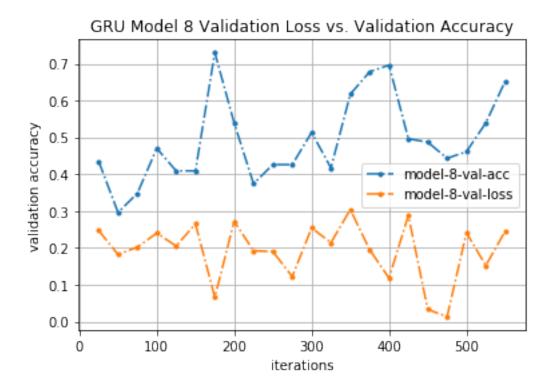
```
y_{smooth} = spline(x, y, x_{smooth})
plt.plot(x_smooth, y_smooth, marker='o', linestyle=linestyle, markersize=3, label="gr
#plt.plot(x_axis_1, gru_unique_1, marker='o', markersize=5, label="gru model 1")
\#x\_list = [x\_axis\_1, x\_axis\_2, x\_axis\_3, x\_axis\_4, x\_axis\_5, x\_axis\_6, x\_axis\_7, x\_axis\_8, x\_a
\#y\_list = [gru\_unique\_1, gru\_unique\_2, gru\_unique\_3, gru\_unique\_4, gru\_unique\_5, gru\_unique\_6, gru
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.plot(x, y, marker='o', markersize=5, linestyle=linestyle, label=label)
\#plt.axis([min(gru\_x\_axis), max(gru\_x\_axis), 0, 1])
plt.grid(True)
#plt.legend(prop=legend_properties)
plt.legend()
plt.savefig('images\gru_image_1_plotting _validation _accuracies _against _iterations
plt.show()
#plt.savefig('gru_plotting _validation _accuracies _against _iterations.png', bbox_in
```

c:\python36\lib\site-packages\ipykernel_launcher.py:33: DeprecationWarning: `spline` is deprecated in scipy 0.19.0, use Bspline class instead.



In []:

```
In [12]: # gru best model validation loss vs validation accuracy:
         #plt.figure(figsize=(30,15))
         plt.title('GRU Model 8 Validation Loss vs. Validation Accuracy')
         plt.xlabel('iterations')
         plt.ylabel('validation accuracy')
         \#plt.plot(x_axis, y_axis_dummy)
         #plt.plot(x_axis_dummy, y_axis)
         \#plt.xticks(x_axis)
         #plt.yticks(y_axis)
         plt.plot(x_axis_8, best_model_val_acc, marker='o', markersize=3, linestyle=linestyle,
         plt.plot(x_axis_8, best_model_val_loss, marker='o', markersize=3, linestyle=linestyle
         #plt.plot(x, y, marker='o', markersize=5, linestyle=linestyle, label=label)
         \#plt.axis([min(gru\_x\_axis), max(gru\_x\_axis), 0, 1])
         plt.grid(True)
         plt.legend()
         #plt.savefig('images\gru_plotting _validation _accuracies _against _iterations.png',
         plt.show()
```



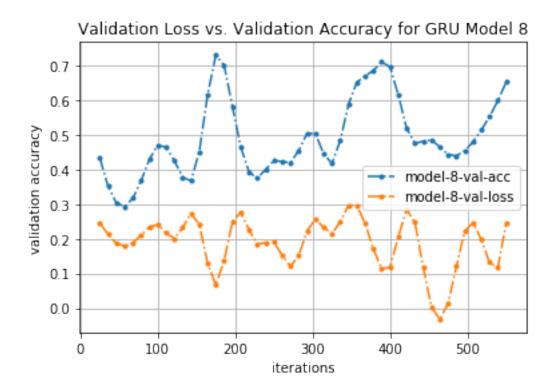
```
In [13]: smoothing_factor = 50
         linestyle='-.'
         \#plt.figure(figsize=(30,15))
         plt.title('Validation Loss vs. Validation Accuracy for GRU Model 8')
         plt.xlabel('iterations')
         plt.ylabel('validation accuracy')
         x = x_axis_8
         y = best_model_val_acc
         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')
         y = best_model_val_loss
         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.plot(x, y, marker='o', markersize=5, linestyle=linestyle, label=label)

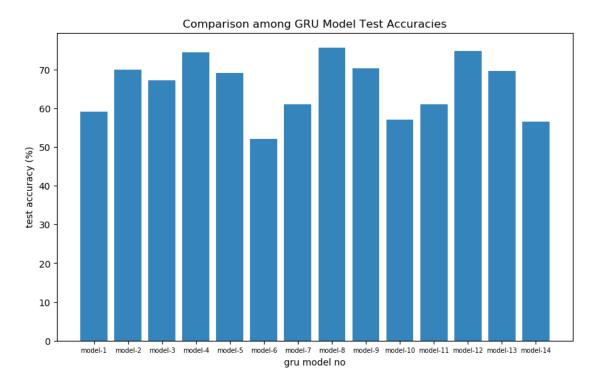
```
plt.grid(True)
plt.legend()
plt.savefig('images\gru_image_2_best_model_validation_loss_vs_validation _accuracy', l
plt.show()
```

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.

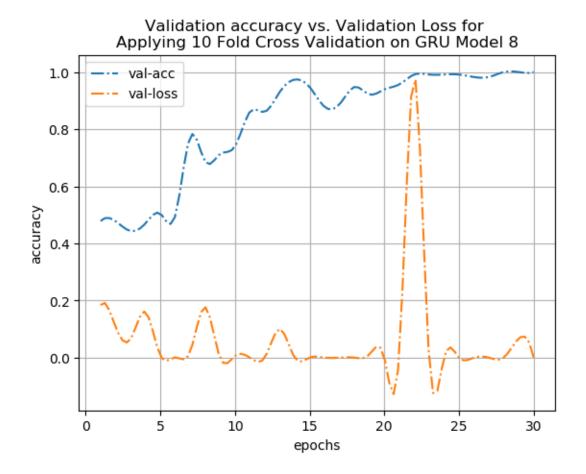


```
111
plt.figure(figsize=(10,6))
objects = []
for i in range(0,len(model_list)):
    object_name = "model-"+str(i+1)
    objects.append(object_name)
y_pos = np.arange(len(objects))
performance = test_acc_list ###
plt.bar(y_pos, performance, align='center', alpha=0.9)
plt.xticks(y_pos, objects)
#plt.tick_params(axis='both', which='major', labelsize=10)
plt.tick_params(axis='x', which='major', labelsize=7)
plt.xlabel('gru model no')
plt.ylabel('test accuracy (%)')
plt.title('Comparison among GRU Model Test Accuracies')
plt.savefig('images\gru_image_3_comparing_gru_model_test_accuracies_bar_chart.png', b
plt.show()
```



In []:

```
print(len(gru_10_fold_cross_val_acc_list))
                                                print(len(gru_10_fold_cross_val_loss_list))
30
30
In [16]: 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(gru_10_fold_cross_val_acc_list)+1)]
                                                y = gru_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=3, linestyle=linestyle, label='val
                                                y = gru_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=3, linestyle=linestyle, label='val
                                                plt.grid(True)
                                                plt.legend()
                                                plt.savefig('images\gru_image_4_accuracy_vs_loss_for_10_fold_cross_validation.png', b
                                                plt.show()
c:\python36\lib\site-packages\ipykernel_launcher.py:14: DeprecationWarning: `spline` is deprec
spline is deprecated in scipy 0.19.0, use Bspline class instead.
c:\python36\lib\site-packages\ipykernel_launcher.py:21: DeprecationWarning: `spline` is deprec
spline is deprecated in scipy 0.19.0, use Bspline class instead.
```



GRY 10 FOLD CROSS VALIDATION MODEL ACCURACY = 0.833909999999998

In []: