## gru\_13\_unique

## December 12, 2018

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In [1]: # GRU WITH UNIQUE DATASET IMPLEMENTATION 13
        # Depression Analysis in Bangla
        # copyright (c) ABDUL HASIB UDDIN <abdulhasibuddin@gmail.com>
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        # Courtesy: https://github.com/mchablani/deep-learning/blob/master/sentiment-rnn/Senti.
In [0]: import numpy as np
        import tensorflow as tf
        from timeit import default_timer as timer
        from collections import Counter
        from string import punctuation
        from google.colab import files
In [0]: # Build the graph::
        gru_size = 512
        gru_layers = 10
        batch_size = 5
        learning_rate = 0.0001
        epochs = 3
In [4]: fileName = "data_all_unique_dnd_stratified_13"
        checkpointName = "checkpoints/"+fileName+".ckpt"
        print(checkpointName)
        print(type(checkpointName))
checkpoints/data_all_unique_dnd_stratified_13.ckpt
<class 'str'>
In [5]: files.upload()
        files.upload()
        with open('data_all_unique_dnd_stratified_text.txt', 'r', encoding="utf8") as f:
            tweets = f.read()
        with open('data_all_unique_dnd_stratified_labels.txt', 'r', encoding="utf8") as f:
            labels_org = f.read()
        print('Done file uploading!')
```

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<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
Done file uploading!
In [0]: # Data preprocessing::
        #all_text = ''.join([c for c in tweets if c not in punctuation])
        all_text = ''.join([c for c in tweets])
        tweets = all_text.split('\n')
        all_text = ' '.join(tweets)
        words = all_text.split()
In [0]: counts = Counter(words)
        vocab = sorted(counts, key=counts.get, reverse=True)
        vocab_to_int = {word: ii for ii, word in enumerate(vocab, 1)}
        tweets_ints = []
        for each in tweets:
            tweets_ints.append([vocab_to_int[word] for word in each.split()])
In [8]: # Encoding the labels::
        list_labels = []
        for l in labels_org.split():
            if 1 == "depressive":
                list_labels.append(1)
            else:
                list_labels.append(0)
        labels = np.array(list_labels)
        print(len(labels))
1176
In [9]: tweets_lens = Counter([len(x) for x in tweets_ints])
        print("Zero-length tweets: {}".format(tweets_lens[0]))
        print("Maximum tweets length: {}".format(max(tweets_lens)))
Zero-length tweets: 1
Maximum tweets length: 63
In [0]: # Filter out that tweets with O length
        tweets_ints = [r[0:200] for r in tweets_ints if len(r) > 0]
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In [11]: from collections import Counter
         tweets_lens = Counter([len(x) for x in tweets_ints])
         print("Zero-length tweets: {}".format(tweets_lens[0]))
         print("Maximum tweet length: {}".format(max(tweets_lens)))
Zero-length tweets: 0
Maximum tweet length: 63
In [0]: seq_len = 200
        features = np.zeros((len(tweets_ints), seq_len), dtype=int)
        # print(features[:10,:100])
        for i, row in enumerate(tweets_ints):
            features[i, -len(row):] = np.array(row)[:seq_len]
        #features[:10,:100]
In [13]: split_frac = 0.8
         split_index = int(split_frac * len(features))
         train_x, val_x = features[:split_index], features[split_index:]
         train_y, val_y = labels[:split_index], labels[split_index:]
         split_frac = 0.5
         split_index = int(split_frac * len(val_x))
         val_x, test_x = val_x[:split_index], val_x[split_index:]
         val_y, test_y = val_y[:split_index], val_y[split_index:]
         print("\t\tFeature Shapes:")
         print("Train set: \t\t{}\".format(train_x.shape),
               "\nValidation set: \t{}".format(val_x.shape),
               "\nTest set: \t\t{}".format(test_x.shape))
         print("label set: \t\t{}".format(train_y.shape),
               "\nValidation label set: \t{}".format(val_y.shape),
               "\nTest label set: \t\t{}".format(test_y.shape))
                        Feature Shapes:
Train set:
                           (940, 200)
Validation set:
                        (118, 200)
                          (118, 200)
Test set:
label set:
                           (940,)
Validation label set:
                              (118,)
Test label set:
                                (118,)
In [0]: n_words = len(vocab_to_int) + 1 # Add 1 for 0 added to vocab
        # Create the graph object
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tf.reset_default_graph()
        with tf.name_scope('inputs'):
            inputs_ = tf.placeholder(tf.int32, [None, None], name="inputs")
            labels_ = tf.placeholder(tf.int32, [None, None], name="labels")
            keep_prob = tf.placeholder(tf.float32, name="keep_prob")
In [0]: # Size of the embedding vectors (number of units in the embedding layer)
        embed_size = 300
        with tf.name_scope("Embeddings"):
            embedding = tf.Variable(tf.random_uniform((n_words, embed_size), -1, 1))
            embed = tf.nn.embedding lookup(embedding, inputs )
In [0]: def gru_cell():
            # Basic GRU cell
            gru = tf.contrib.rnn.GRUCell(gru_size, reuse=tf.get_variable_scope().reuse)
            # Add dropout to the cell
            return tf.contrib.rnn.DropoutWrapper(gru, output_keep_prob=keep_prob)
       with tf.name_scope("RNN_layers"):
            # Stack up multiple GRU layers, for deep learning
            cell = tf.contrib.rnn.MultiRNNCell([gru_cell() for _ in range(gru_layers)])
            # Getting an initial state of all zeros
            initial_state = cell.zero_state(batch_size, tf.float32)
In [0]: with tf.name_scope("RNN_forward"):
            outputs, final_state = tf.nn.dynamic_rnn(cell, embed, initial_state=initial_state)
In [0]: # Output::
        with tf.name_scope('predictions'):
            predictions = tf.contrib.layers.fully_connected(outputs[:, -1], 1, activation_fn=t
            tf.summary.histogram('predictions', predictions)
        with tf.name_scope('cost'):
            cost = tf.losses.mean_squared_error(labels_, predictions)
            tf.summary.scalar('cost', cost)
        with tf.name_scope('train'):
            optimizer = tf.train.AdamOptimizer(learning_rate).minimize(cost)
       merged = tf.summary.merge_all()
In [0]: # Validation accuracy::
        with tf.name_scope('validation'):
            correct_pred = tf.equal(tf.cast(tf.round(predictions), tf.int32), labels_)
            accuracy = tf.reduce_mean(tf.cast(correct_pred, tf.float32))
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In [0]: # Batching::
        def get_batches(x, y, batch_size=100):
            n_batches = len(x)//batch_size
            x, y = x[:n_batches*batch_size], y[:n_batches*batch_size]
            for ii in range(0, len(x), batch_size):
                yield x[ii:ii+batch_size], y[ii:ii+batch_size]
In [21]: # Training::
         \#epochs = 5
         saver = tf.train.Saver()
         start = timer()
         with tf.Session() as sess:
             sess.run(tf.global_variables_initializer())
             train_writer = tf.summary.FileWriter('./logs/tb/train', sess.graph)
             test_writer = tf.summary.FileWriter('./logs/tb/test', sess.graph)
             iteration = 1
             for e in range(1, epochs+1):
                 state = sess.run(initial_state)
                 for ii, (x, y) in enumerate(get_batches(train_x, train_y, batch_size), 1):
                     feed = {inputs_: x,
                             labels_: y[:, None],
                             keep_prob: 1,
                             initial_state: state}
                     summary, loss, state, _ = sess.run([merged, cost, final_state, optimizer]
         #
                       loss, state, _ = sess.run([cost, final_state, optimizer], feed_dict=fee
                     train_writer.add_summary(summary, iteration)
                     if iteration%5==0:
                         print("Epoch: {}/{}".format(e, epochs),
                               "Iteration: {}".format(iteration),
                               "Train loss: {:.3f}".format(loss))
                     if iteration%25==0:
                         val acc = []
                         val_state = sess.run(cell.zero_state(batch_size, tf.float32))
                         for x, y in get_batches(val_x, val_y, batch_size):
                             feed = {inputs_: x,
                                     labels_: y[:, None],
                                     keep_prob: 1,
                                     initial_state: val_state}
                               batch_acc, val_state = sess.run([accuracy, final_state], feed_d
         #
                             summary, batch_acc, val_state = sess.run([merged, accuracy, final]
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val_acc.append(batch_acc)
                         print("Val acc: {:.3f}".format(np.mean(val_acc)))
                     iteration +=1
                     test_writer.add_summary(summary, iteration)
                     saver.save(sess, checkpointName)
                      tensorboard = TensorBoard(log\_dir="logs/tweet\_5000\_all\_sentiments\_six\_cl
             saver.save(sess, checkpointName)
         duration = timer() - start
         print('Time elasped =',duration,'sec(s)')
Epoch: 1/3 Iteration: 5 Train loss: 0.275
Epoch: 1/3 Iteration: 10 Train loss: 0.240
Epoch: 1/3 Iteration: 15 Train loss: 0.243
Epoch: 1/3 Iteration: 20 Train loss: 0.256
Epoch: 1/3 Iteration: 25 Train loss: 0.244
Val acc: 0.504
Epoch: 1/3 Iteration: 30 Train loss: 0.243
Epoch: 1/3 Iteration: 35 Train loss: 0.244
Epoch: 1/3 Iteration: 40 Train loss: 0.233
Epoch: 1/3 Iteration: 45 Train loss: 0.255
Epoch: 1/3 Iteration: 50 Train loss: 0.208
Val acc: 0.452
Epoch: 1/3 Iteration: 55 Train loss: 0.264
Epoch: 1/3 Iteration: 60 Train loss: 0.248
Epoch: 1/3 Iteration: 65 Train loss: 0.237
Epoch: 1/3 Iteration: 70 Train loss: 0.233
Epoch: 1/3 Iteration: 75 Train loss: 0.224
Val acc: 0.443
Epoch: 1/3 Iteration: 80 Train loss: 0.257
Epoch: 1/3 Iteration: 85 Train loss: 0.273
Epoch: 1/3 Iteration: 90 Train loss: 0.235
Epoch: 1/3 Iteration: 95 Train loss: 0.267
Epoch: 1/3 Iteration: 100 Train loss: 0.248
Val acc: 0.287
Epoch: 1/3 Iteration: 105 Train loss: 0.218
Epoch: 1/3 Iteration: 110 Train loss: 0.326
Epoch: 1/3 Iteration: 115 Train loss: 0.239
Epoch: 1/3 Iteration: 120 Train loss: 0.232
Epoch: 1/3 Iteration: 125 Train loss: 0.211
Val acc: 0.391
Epoch: 1/3 Iteration: 130 Train loss: 0.211
Epoch: 1/3 Iteration: 135 Train loss: 0.283
Epoch: 1/3 Iteration: 140 Train loss: 0.247
Epoch: 1/3 Iteration: 145 Train loss: 0.239
Epoch: 1/3 Iteration: 150 Train loss: 0.245
Val acc: 0.496
Epoch: 1/3 Iteration: 155 Train loss: 0.245
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Epoch: 1/3 Iteration: 160 Train loss: 0.244
Epoch: 1/3 Iteration: 165 Train loss: 0.229
Epoch: 1/3 Iteration: 170 Train loss: 0.196
Epoch: 1/3 Iteration: 175 Train loss: 0.092
Val acc: 0.670
Epoch: 1/3 Iteration: 180 Train loss: 0.282
Epoch: 1/3 Iteration: 185 Train loss: 0.308
Epoch: 2/3 Iteration: 190 Train loss: 0.298
Epoch: 2/3 Iteration: 195 Train loss: 0.264
Epoch: 2/3 Iteration: 200 Train loss: 0.265
Val acc: 0.504
Epoch: 2/3 Iteration: 205 Train loss: 0.276
Epoch: 2/3 Iteration: 210 Train loss: 0.259
Epoch: 2/3 Iteration: 215 Train loss: 0.251
Epoch: 2/3 Iteration: 220 Train loss: 0.248
Epoch: 2/3 Iteration: 225 Train loss: 0.243
Val acc: 0.504
Epoch: 2/3 Iteration: 230 Train loss: 0.219
Epoch: 2/3 Iteration: 235 Train loss: 0.262
Epoch: 2/3 Iteration: 240 Train loss: 0.234
Epoch: 2/3 Iteration: 245 Train loss: 0.253
Epoch: 2/3 Iteration: 250 Train loss: 0.222
Val acc: 0.470
Epoch: 2/3 Iteration: 255 Train loss: 0.203
Epoch: 2/3 Iteration: 260 Train loss: 0.252
Epoch: 2/3 Iteration: 265 Train loss: 0.288
Epoch: 2/3 Iteration: 270 Train loss: 0.208
Epoch: 2/3 Iteration: 275 Train loss: 0.206
Val acc: 0.348
Epoch: 2/3 Iteration: 280 Train loss: 0.213
Epoch: 2/3 Iteration: 285 Train loss: 0.144
Epoch: 2/3 Iteration: 290 Train loss: 0.264
Epoch: 2/3 Iteration: 295 Train loss: 0.263
Epoch: 2/3 Iteration: 300 Train loss: 0.370
Val acc: 0.478
Epoch: 2/3 Iteration: 305 Train loss: 0.203
Epoch: 2/3 Iteration: 310 Train loss: 0.242
Epoch: 2/3 Iteration: 315 Train loss: 0.256
Epoch: 2/3 Iteration: 320 Train loss: 0.246
Epoch: 2/3 Iteration: 325 Train loss: 0.213
Val acc: 0.470
Epoch: 2/3 Iteration: 330 Train loss: 0.303
Epoch: 2/3 Iteration: 335 Train loss: 0.281
Epoch: 2/3 Iteration: 340 Train loss: 0.258
Epoch: 2/3 Iteration: 345 Train loss: 0.252
Epoch: 2/3 Iteration: 350 Train loss: 0.363
Val acc: 0.496
Epoch: 2/3 Iteration: 355 Train loss: 0.245
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Epoch: 2/3 Iteration: 360 Train loss: 0.245
Epoch: 2/3 Iteration: 365 Train loss: 0.247
Epoch: 2/3 Iteration: 370 Train loss: 0.250
Epoch: 2/3 Iteration: 375 Train loss: 0.251
Val acc: 0.513
Epoch: 3/3 Iteration: 380 Train loss: 0.257
Epoch: 3/3 Iteration: 385 Train loss: 0.252
Epoch: 3/3 Iteration: 390 Train loss: 0.251
Epoch: 3/3 Iteration: 395 Train loss: 0.256
Epoch: 3/3 Iteration: 400 Train loss: 0.249
Val acc: 0.730
Epoch: 3/3 Iteration: 405 Train loss: 0.246
Epoch: 3/3 Iteration: 410 Train loss: 0.249
Epoch: 3/3 Iteration: 415 Train loss: 0.252
Epoch: 3/3 Iteration: 420 Train loss: 0.249
Epoch: 3/3 Iteration: 425 Train loss: 0.253
Val acc: 0.504
Epoch: 3/3 Iteration: 430 Train loss: 0.254
Epoch: 3/3 Iteration: 435 Train loss: 0.251
Epoch: 3/3 Iteration: 440 Train loss: 0.249
Epoch: 3/3 Iteration: 445 Train loss: 0.249
Epoch: 3/3 Iteration: 450 Train loss: 0.243
Val acc: 0.504
Epoch: 3/3 Iteration: 455 Train loss: 0.243
Epoch: 3/3 Iteration: 460 Train loss: 0.190
Epoch: 3/3 Iteration: 465 Train loss: 0.167
Epoch: 3/3 Iteration: 470 Train loss: 0.170
Epoch: 3/3 Iteration: 475 Train loss: 0.259
Val acc: 0.504
Epoch: 3/3 Iteration: 480 Train loss: 0.118
Epoch: 3/3 Iteration: 485 Train loss: 0.309
Epoch: 3/3 Iteration: 490 Train loss: 0.157
Epoch: 3/3 Iteration: 495 Train loss: 0.125
Epoch: 3/3 Iteration: 500 Train loss: 0.407
Val acc: 0.461
Epoch: 3/3 Iteration: 505 Train loss: 0.327
Epoch: 3/3 Iteration: 510 Train loss: 0.321
Epoch: 3/3 Iteration: 515 Train loss: 0.301
Epoch: 3/3 Iteration: 520 Train loss: 0.248
Epoch: 3/3 Iteration: 525 Train loss: 0.236
Val acc: 0.504
Epoch: 3/3 Iteration: 530 Train loss: 0.248
Epoch: 3/3 Iteration: 535 Train loss: 0.251
Epoch: 3/3 Iteration: 540 Train loss: 0.251
Epoch: 3/3 Iteration: 545 Train loss: 0.252
Epoch: 3/3 Iteration: 550 Train loss: 0.245
Val acc: 0.626
Epoch: 3/3 Iteration: 555 Train loss: 0.243
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Epoch: 3/3 Iteration: 560 Train loss: 0.245
Time elasped = 5795.448450488 \text{ sec(s)}
In [22]: # Testing::
         test_acc = []
         with tf.Session() as sess:
             saver.restore(sess, checkpointName)
             test_state = sess.run(cell.zero_state(batch_size, tf.float32))
             for ii, (x, y) in enumerate(get_batches(test_x, test_y, batch_size), 1):
                 feed = {inputs_: x,
                         labels_: y[:, None],
                         keep_prob: 1,
                         initial_state: test_state}
                 batch_acc, test_state = sess.run([accuracy, final_state], feed_dict=feed)
                 test_acc.append(batch_acc)
             print("Test accuracy: {:.3f}".format(np.mean(test_acc)))
INFO:tensorflow:Restoring parameters from checkpoints/data_all_unique_dnd_stratified_13.ckpt
Test accuracy: 0.696
In [0]:
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