

gru_4_unique

December 12, 2018

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In [1]: # GRU WITH UNIQUE DATASET IMPLEMENTATION 4
        # 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 = 5
        batch_size = 10
        learning_rate = 0.0001
        epochs = 5

In [4]: fileName = "data_all_unique_dnd_stratified_4"
        checkpointName = "checkpoints/"+fileName+".ckpt"
        print(checkpointName)
        print(type(checkpointName))

checkpoints/data_all_unique_dnd_stratified_4.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!')
```

<IPython.core.display.HTML object>

<IPython.core.display.HTML object>

Done file uploading!

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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 l == "depressive":
                list_labels.append(1)
            else:
                list_labels.append(0)

        labels = np.array(list_labels)
        print(len(labels))
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1176

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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)))
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Zero-length tweets: 1

Maximum tweets length: 63

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In [0]: # Filter out that tweets with 0 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)))
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Zero-length tweets: 0
Maximum tweet length: 63
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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]
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In [13]: split_frac = 0.8
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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\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))
```

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                                Feature Shapes:
Train set:                      (940, 200)
Validation set:                  (118, 200)
Test set:                       (118, 200)
label set:                      (940,)
Validation label set:            (118,)
Test label set:                 (118,)
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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=tanh)
    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::
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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]
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In [21]: # Training::
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#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=feed)  
  
            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 elapsed =',duration,'sec(s)')

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Epoch: 1/5 Iteration: 5 Train loss: 0.259
Epoch: 1/5 Iteration: 10 Train loss: 0.283
Epoch: 1/5 Iteration: 15 Train loss: 0.271
Epoch: 1/5 Iteration: 20 Train loss: 0.262
Epoch: 1/5 Iteration: 25 Train loss: 0.200
Val acc: 0.291
Epoch: 1/5 Iteration: 30 Train loss: 0.228
Epoch: 1/5 Iteration: 35 Train loss: 0.240
Epoch: 1/5 Iteration: 40 Train loss: 0.247
Epoch: 1/5 Iteration: 45 Train loss: 0.228
Epoch: 1/5 Iteration: 50 Train loss: 0.252
Val acc: 0.273
Epoch: 1/5 Iteration: 55 Train loss: 0.253
Epoch: 1/5 Iteration: 60 Train loss: 0.175
Epoch: 1/5 Iteration: 65 Train loss: 0.267
Epoch: 1/5 Iteration: 70 Train loss: 0.257
Epoch: 1/5 Iteration: 75 Train loss: 0.252
Val acc: 0.318
Epoch: 1/5 Iteration: 80 Train loss: 0.269
Epoch: 1/5 Iteration: 85 Train loss: 0.172
Epoch: 1/5 Iteration: 90 Train loss: 0.236
Epoch: 2/5 Iteration: 95 Train loss: 0.290
Epoch: 2/5 Iteration: 100 Train loss: 0.202
Val acc: 0.582
Epoch: 2/5 Iteration: 105 Train loss: 0.265
Epoch: 2/5 Iteration: 110 Train loss: 0.261
Epoch: 2/5 Iteration: 115 Train loss: 0.230
Epoch: 2/5 Iteration: 120 Train loss: 0.261
Epoch: 2/5 Iteration: 125 Train loss: 0.219
Val acc: 0.455
Epoch: 2/5 Iteration: 130 Train loss: 0.258
Epoch: 2/5 Iteration: 135 Train loss: 0.266
Epoch: 2/5 Iteration: 140 Train loss: 0.235
Epoch: 2/5 Iteration: 145 Train loss: 0.194
Epoch: 2/5 Iteration: 150 Train loss: 0.287
Val acc: 0.418
Epoch: 2/5 Iteration: 155 Train loss: 0.237

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Epoch: 2/5 Iteration: 160 Train loss: 0.223
Epoch: 2/5 Iteration: 165 Train loss: 0.253
Epoch: 2/5 Iteration: 170 Train loss: 0.232
Epoch: 2/5 Iteration: 175 Train loss: 0.207
Val acc: 0.664
Epoch: 2/5 Iteration: 180 Train loss: 0.044
Epoch: 2/5 Iteration: 185 Train loss: 0.189
Epoch: 3/5 Iteration: 190 Train loss: 0.492
Epoch: 3/5 Iteration: 195 Train loss: 0.217
Epoch: 3/5 Iteration: 200 Train loss: 0.226
Val acc: 0.736
Epoch: 3/5 Iteration: 205 Train loss: 0.267
Epoch: 3/5 Iteration: 210 Train loss: 0.226
Epoch: 3/5 Iteration: 215 Train loss: 0.195
Epoch: 3/5 Iteration: 220 Train loss: 0.175
Epoch: 3/5 Iteration: 225 Train loss: 0.209
Val acc: 0.464
Epoch: 3/5 Iteration: 230 Train loss: 0.111
Epoch: 3/5 Iteration: 235 Train loss: 0.164
Epoch: 3/5 Iteration: 240 Train loss: 0.182
Epoch: 3/5 Iteration: 245 Train loss: 0.244
Epoch: 3/5 Iteration: 250 Train loss: 0.236
Val acc: 0.445
Epoch: 3/5 Iteration: 255 Train loss: 0.251
Epoch: 3/5 Iteration: 260 Train loss: 0.186
Epoch: 3/5 Iteration: 265 Train loss: 0.218
Epoch: 3/5 Iteration: 270 Train loss: 0.033
Epoch: 3/5 Iteration: 275 Train loss: 0.226
Val acc: 0.664
Epoch: 3/5 Iteration: 280 Train loss: 0.229
Epoch: 4/5 Iteration: 285 Train loss: 0.251
Epoch: 4/5 Iteration: 290 Train loss: 0.196
Epoch: 4/5 Iteration: 295 Train loss: 0.281
Epoch: 4/5 Iteration: 300 Train loss: 0.162
Val acc: 0.482
Epoch: 4/5 Iteration: 305 Train loss: 0.098
Epoch: 4/5 Iteration: 310 Train loss: 0.074
Epoch: 4/5 Iteration: 315 Train loss: 0.151
Epoch: 4/5 Iteration: 320 Train loss: 0.123
Epoch: 4/5 Iteration: 325 Train loss: 0.074
Val acc: 0.382
Epoch: 4/5 Iteration: 330 Train loss: 0.140
Epoch: 4/5 Iteration: 335 Train loss: 0.104
Epoch: 4/5 Iteration: 340 Train loss: 0.084
Epoch: 4/5 Iteration: 345 Train loss: 0.103
Epoch: 4/5 Iteration: 350 Train loss: 0.198
Val acc: 0.427
Epoch: 4/5 Iteration: 355 Train loss: 0.243

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Epoch: 4/5 Iteration: 360 Train loss: 0.198
Epoch: 4/5 Iteration: 365 Train loss: 0.201
Epoch: 4/5 Iteration: 370 Train loss: 0.106
Epoch: 4/5 Iteration: 375 Train loss: 0.148
Val acc: 0.709
Epoch: 5/5 Iteration: 380 Train loss: 0.216
Epoch: 5/5 Iteration: 385 Train loss: 0.221
Epoch: 5/5 Iteration: 390 Train loss: 0.234
Epoch: 5/5 Iteration: 395 Train loss: 0.194
Epoch: 5/5 Iteration: 400 Train loss: 0.233
Val acc: 0.400
Epoch: 5/5 Iteration: 405 Train loss: 0.090
Epoch: 5/5 Iteration: 410 Train loss: 0.132
Epoch: 5/5 Iteration: 415 Train loss: 0.342
Epoch: 5/5 Iteration: 420 Train loss: 0.121
Epoch: 5/5 Iteration: 425 Train loss: 0.128
Val acc: 0.436
Epoch: 5/5 Iteration: 430 Train loss: 0.122
Epoch: 5/5 Iteration: 435 Train loss: 0.077
Epoch: 5/5 Iteration: 440 Train loss: 0.263
Epoch: 5/5 Iteration: 445 Train loss: 0.067
Epoch: 5/5 Iteration: 450 Train loss: 0.179
Val acc: 0.473
Epoch: 5/5 Iteration: 455 Train loss: 0.139
Epoch: 5/5 Iteration: 460 Train loss: 0.151
Epoch: 5/5 Iteration: 465 Train loss: 0.041
Epoch: 5/5 Iteration: 470 Train loss: 0.138
Time elapsed = 4025.368504245 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)))

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INFO:tensorflow:Restoring parameters from checkpoints/data_all_unique_dnd_stratified_4.ckpt
Test accuracy: 0.745

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In [0]: