lstm_with_unique_10

December 12, 2018

In [1]:
```
# LSTM WITH UNIQUE DATASET IMPLEMENTATION 10
# Depression Analysis in Bangla with LSTM-RNN
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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::

lstm_size = 128
lstm_layers = 5
batch_size = 25
learning_rate = 0.0001
epochs = 10
```

In [4]:
```
fileName = "lstm_with_unique_10"
checkpointName = "checkpoints/"+fileName+".ckpt"
print(checkpointName)
print(type(checkpointName))
```

```
checkpoints/lstm_with_unique_10.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()
```

```
<IPython.core.display.HTML object>
```

```
<IPython.core.display.HTML object>
```

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))
```

```
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 0 length
tweets_ints = [r[0:200] for r in tweets_ints if len(r) > 0]
```

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\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)
Test set:                   (118, 200)
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
        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 [16]:
```
def lstm_cell():
    # Your basic LSTM cell
    lstm = tf.contrib.rnn.BasicLSTMCell(lstm_size, reuse=tf.get_variable_scope().reuse
    # Add dropout to the cell
    return tf.contrib.rnn.DropoutWrapper(lstm, output_keep_prob=keep_prob)

with tf.name_scope("RNN_layers"):
    # Stack up multiple LSTM layers, for deep learning
    cell = tf.contrib.rnn.MultiRNNCell([lstm_cell() for _ in range(lstm_layers)])

    # Getting an initial state of all zeros
    initial_state = cell.zero_state(batch_size, tf.float32)
```

```
WARNING:tensorflow:From <ipython-input-16-678741cf60df>:3: BasicLSTMCell.__init__ (from tensori
Instructions for updating:
This class is deprecated, please use tf.nn.rnn_cell.LSTMCell, which supports all the feature tl
```

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=ti
    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))
```

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
                            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: 1/10 Iteration: 5 Train loss: 0.246
Epoch: 1/10 Iteration: 10 Train loss: 0.245
Epoch: 1/10 Iteration: 15 Train loss: 0.245
Epoch: 1/10 Iteration: 20 Train loss: 0.247
Epoch: 1/10 Iteration: 25 Train loss: 0.240
Val acc: 0.410
Epoch: 1/10 Iteration: 30 Train loss: 0.257
Epoch: 1/10 Iteration: 35 Train loss: 0.278
Epoch: 2/10 Iteration: 40 Train loss: 0.251
Epoch: 2/10 Iteration: 45 Train loss: 0.256
Epoch: 2/10 Iteration: 50 Train loss: 0.248
Val acc: 0.500
Epoch: 2/10 Iteration: 55 Train loss: 0.247
Epoch: 2/10 Iteration: 60 Train loss: 0.251
Epoch: 2/10 Iteration: 65 Train loss: 0.242
Epoch: 2/10 Iteration: 70 Train loss: 0.278
Epoch: 3/10 Iteration: 75 Train loss: 0.246
Val acc: 0.500
Epoch: 3/10 Iteration: 80 Train loss: 0.252
Epoch: 3/10 Iteration: 85 Train loss: 0.249
Epoch: 3/10 Iteration: 90 Train loss: 0.249
Epoch: 3/10 Iteration: 95 Train loss: 0.244
Epoch: 3/10 Iteration: 100 Train loss: 0.251
Val acc: 0.470
Epoch: 3/10 Iteration: 105 Train loss: 0.251
Epoch: 3/10 Iteration: 110 Train loss: 0.253
Epoch: 4/10 Iteration: 115 Train loss: 0.249
Epoch: 4/10 Iteration: 120 Train loss: 0.247
Epoch: 4/10 Iteration: 125 Train loss: 0.250
Val acc: 0.510
Epoch: 4/10 Iteration: 130 Train loss: 0.246
Epoch: 4/10 Iteration: 135 Train loss: 0.238
Epoch: 4/10 Iteration: 140 Train loss: 0.250
Epoch: 4/10 Iteration: 145 Train loss: 0.260
Epoch: 5/10 Iteration: 150 Train loss: 0.244
Val acc: 0.410
Epoch: 5/10 Iteration: 155 Train loss: 0.245
Epoch: 5/10 Iteration: 160 Train loss: 0.246
Epoch: 5/10 Iteration: 165 Train loss: 0.245
```

```
Epoch: 5/10 Iteration: 170 Train loss: 0.244
Epoch: 5/10 Iteration: 175 Train loss: 0.237
Val acc: 0.450
Epoch: 5/10 Iteration: 180 Train loss: 0.260
Epoch: 5/10 Iteration: 185 Train loss: 0.246
Epoch: 6/10 Iteration: 190 Train loss: 0.247
Epoch: 6/10 Iteration: 195 Train loss: 0.240
Epoch: 6/10 Iteration: 200 Train loss: 0.236
Val acc: 0.620
Epoch: 6/10 Iteration: 205 Train loss: 0.219
Epoch: 6/10 Iteration: 210 Train loss: 0.208
Epoch: 6/10 Iteration: 215 Train loss: 0.240
Epoch: 6/10 Iteration: 220 Train loss: 0.253
Epoch: 7/10 Iteration: 225 Train loss: 0.153
Val acc: 0.660
Epoch: 7/10 Iteration: 230 Train loss: 0.197
Epoch: 7/10 Iteration: 235 Train loss: 0.159
Epoch: 7/10 Iteration: 240 Train loss: 0.215
Epoch: 7/10 Iteration: 245 Train loss: 0.184
Epoch: 7/10 Iteration: 250 Train loss: 0.145
Val acc: 0.480
Epoch: 7/10 Iteration: 255 Train loss: 0.257
Epoch: 8/10 Iteration: 260 Train loss: 0.196
Epoch: 8/10 Iteration: 265 Train loss: 0.208
Epoch: 8/10 Iteration: 270 Train loss: 0.165
Epoch: 8/10 Iteration: 275 Train loss: 0.150
Val acc: 0.520
Epoch: 8/10 Iteration: 280 Train loss: 0.135
Epoch: 8/10 Iteration: 285 Train loss: 0.161
Epoch: 8/10 Iteration: 290 Train loss: 0.166
Epoch: 8/10 Iteration: 295 Train loss: 0.140
Epoch: 9/10 Iteration: 300 Train loss: 0.141
Val acc: 0.550
Epoch: 9/10 Iteration: 305 Train loss: 0.121
Epoch: 9/10 Iteration: 310 Train loss: 0.150
Epoch: 9/10 Iteration: 315 Train loss: 0.095
Epoch: 9/10 Iteration: 320 Train loss: 0.064
Epoch: 9/10 Iteration: 325 Train loss: 0.123
Val acc: 0.460
Epoch: 9/10 Iteration: 330 Train loss: 0.118
Epoch: 10/10 Iteration: 335 Train loss: 0.049
Epoch: 10/10 Iteration: 340 Train loss: 0.089
Epoch: 10/10 Iteration: 345 Train loss: 0.059
Epoch: 10/10 Iteration: 350 Train loss: 0.131
Val acc: 0.600
Epoch: 10/10 Iteration: 355 Train loss: 0.109
Epoch: 10/10 Iteration: 360 Train loss: 0.072
Epoch: 10/10 Iteration: 365 Train loss: 0.095
```

```
Epoch: 10/10 Iteration: 370 Train loss: 0.140
Time elasped = 2457.6001075900003 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/lstm_with_unique_10.ckpt
Test accuracy: 0.570
```

In [0]: