```
In [1]:
import numpy as np
import pandas as pd
import bcolz as bz
import pickle
import torch
import torch.nn as nn
import torch.nn.functional as F
from torch.nn.utils.rnn import pack_padded_sequence, pad_packed_sequence
from torch.autograd import Variable
import time
from sklearn import metrics
from sklearn.metrics import roc_curve
from sklearn.metrics import auc
In [2]:
def divide_string(sentence):
   s l = sentence.split()
   r 1 = []
   punc = [ '?', '!', '*']
   for i in range(len(s l)):
      tmp = 0
       flag = True;
       for j in range(len(s_l[i])):
           if s l[i][j] in punc:
              if (flag):
                 r l.append(s l[i][tmp : j])
              tmp = j
              flag = True
           if s l[i][j] in delete:
              if(flag):
                 r_l.append(s_l[i][tmp : j])
              tmp = j
              flag = False
           else:
              if(not flag):
                 tmp = tmp + 1;
              flag = True
       if(flag):
           r_1.append(s_1[i][tmp:len(s_1[i])])
   str list = [x for x in r l if x != '']
   return str list
In [3]:
b = divide string(a)
print(len(b))
print(b)
['THIS', 'is', 'a', 'good', 'day', '!', "don't", 'know']
In [4]:
vectors = bz.open(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final Project\glove\27B.100.dat")[:]
words = pickle.load(open(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final
Project\glove\27B.100 words.pkl", 'rb'))
word2idx = pickle.load(open(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final
Project\glove\27B.100 idx.pkl", 'rb'))
glove = {w: vectors[word2idx[w]] for w in words}
print(len(glove))
```

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```
def remove infrequent words(sents):
   word counts = {}
   divide sentence = []
   #divide each sentence first
    for s in sents:
        tmp = divide string(s)
        divide sentence.append(tmp)
    # count the word
    for s in divide sentence:
       for w in s:
           if w in word counts:
               word_counts[w] += 1
            else:
                word_counts[w] = 1
    threshold = 2
    filtered sents = []
    for s in divide sentence:
       new s = []
        for w in s:
            if word counts[w] < threshold:</pre>
               new s.append('<UNKOWN>')
            else:
               new_s.append(w)
        filtered_sents.append(new_s)
    return filtered sents
```

### In [6]:

```
s = ["This is a good time!!", "You are pretty@%&^$@ happy", "test", "test*%& test"]
def word embedding glove total(sentence):
    #create the total size of weight matrix
    weights matrix = np.zeros((100,len(sentence), 103))
    processed sentence = remove infrequent words(sentence)
    for k in range(len(processed_sentence)):
        target vocab = processed sentence[k]
        emb dim = 100
        # enumerate the word, break at position 100
        for i, word in enumerate(target vocab):
            if i == 100:
                break;
            try:
                tmp vec = glove[word]
            except KeyError:
               tmp_vec = np.random.normal(scale=0.6, size=(emb_dim, ))
                tmp dic = {word:tmp vec}
                glove.update(tmp_dic)
            #write something to change it to 103
            ind = 0
            if word.isupper():
               tmp_vec = np.append(tmp_vec, [1])
                tmp_vec = np.append(tmp_vec, [0])
                ind = ind + 1
            if word.islower():
               tmp vec = np.append(tmp vec, [1])
            else:
               tmp_vec = np.append(tmp_vec, [0])
                ind = ind + 1
            if ind == 2:
                tmp_vec = np.append(tmp_vec, [1])
            else:
                tmp vec = np.append(tmp vec, [0])
            weights_matrix[i][k] = tmp_vec;
    return weights matrix
```

```
test = word_embedding_glove_total(s)
print(test.shape)
sentences = remove infrequent words(s)
print(sentences)
(100, 4, 103)
[['<UNKOWN>', '<UNKOWN>', '<UNKOWN>', '<UNKOWN>', '!', '!'], ['<UNKOWN>', '<UNKOWN>',
'<UNKOWN>', '<UNKOWN>'], ['test'], ['test', '<UNKOWN>', 'test']]
In [7]:
s = ["This is a good time!!", "You are pretty@$&^$@ happy", "test", "test*%& test"]
def base glove total(sentence):
    #create the total size of weight matrix
    weights_matrix = np.zeros((100,len(sentence), 103))
    for k in range(len(sentence)):
        target vocab = sentence[k].split()
        emb_dim = 100
        # enumerate the word, break at position 100
        for i in range(len(target_vocab)):
            if i == 100:
                break;
            try:
                word = target vocab[i]
               tmp_vec = glove[target_vocab[i]]
            except KeyError:
                tmp vec = np.random.normal(scale=0.6, size=(emb dim, ))
            #write something to change it to 103
            ind = 0
            if word.isupper():
               tmp_vec = np.append(tmp_vec, [1])
               tmp_vec = np.append(tmp_vec, [0])
                ind = ind + 1
            if word.islower():
               tmp vec = np.append(tmp vec, [1])
            else:
               tmp vec = np.append(tmp vec, [0])
               ind = ind + 1
            if ind == 2:
                tmp vec = np.append(tmp vec, [1])
            else:
                tmp vec = np.append(tmp vec, [0])
            weights_matrix[i][k] = tmp_vec;
    return weights matrix
test = base glove total(s)
print(test.shape)
(100, 4, 103)
In [8]:
def create_emb_layer(weights_matrix, non_trainable=False):
    num_embeddings, embedding_dim = weights_matrix.shape
    emb layer = nn.Embedding(num embeddings, embedding dim)
    emb layer.load state dict({'weight': weights matrix})
    if non trainable:
        emb layer.weight.requires grad = False
    return emb layer, num embeddings, embedding dim
class tcLSTM(nn.Module):
    def __init__(self, input_dim, hidden_size, target_size, batch_size):
```

```
super(tcLSTM, self).__init__()
        self.embedding dim = input dim
        self.hidden size = hidden size
       self.batch size = batch size
       self.lstm = nn.LSTM(self.embedding dim, hidden size)
        # The linear layer that maps from hidden state space to target space
        self.hidden2tar = nn.Linear(self.hidden size, target size)
        self.hidden = self.init hidden()
    def forward(self, weights matrix):
        #didn't use batch, may need to add batch size in the middle
        lstm out, self.hidden = self.lstm(weights matrix)
        y pred = self.hidden2tar(lstm out[-1])
        return torch.sigmoid(y pred)
    def init hidden(self):
        return (Variable(torch.zeros(1, self.batch_size, self.hidden_size)),
               Variable(torch.zeros(1, self.batch_size, self.hidden_size)))
In [9]:
data = pd.read csv(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final Project\data\train.csv",
encoding = "ISO-8859-1")
#data.head()
In [10]:
x train = data['comment text']
y_train = data.iloc[:,2:8]
print(x train.shape)
print(y_train.shape)
(159571,)
(159571, 6)
In [11]:
data = pd.read csv(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final Project\data\test.csv", encoding
= "ISO-8859-1")
x_test = data['comment_text']
print(x_test.shape)
data = pd.read_csv(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final Project\data\test labels.csv", e
ncoding = "ISO-8859-1")
y test = data.iloc[:,1:7]
print(y_test.shape)
(153164,)
(153164, 6)
In [12]:
#the embedding
train set = word embedding glove total(x train)
#train_set = base_glove_total(x_train)
In [13]:
print(train set.shape)
w = train set[:,512:1024,:]
print(w.shape)
c = y_train.values
c = c.reshape((1,159571,6))
print(c.shape)
(100, 159571, 103)
(100, 512, 103)
```

(1, 159571, 6)

#### In [14]:

```
data = pd.read csv(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final Project\data\test.csv", encoding
= "ISO-8859-1")
data2 = pd.read_csv(r"C:\Users\mul02\Desktop\Course\LIGN 167\Final Project\data\test labels.csv",
encoding = "ISO-8859-1")
df = pd.concat([data.reset index(drop=True), data2.reset index(drop=True)], axis=1)
df = df.drop(['id', 'id'],axis=1)
df = df[df.toxic != -1]
print(data.shape)
print(data2.shape)
print(df.shape)
x test = df.reset index()['comment text']
y test = df.reset index().iloc[:, 2:8]
print(x test.shape)
print(y_test.shape)
(153164, 2)
(153164, 7)
(63978, 7)
(63978,)
(63978, 6)
```

### In [15]:

```
batch_size = 512
epoch = 100
lr = 0.005
hidden_state = 24
output_dim = 6
input_dim = 103
train_size = len(x_train)
train_output = y_train.values

model = tcLSTM(input_dim=input_dim, hidden_size=hidden_state, target_size=output_dim, batch_size = batch_size)
model = model.cuda()
loss_fn = nn.BCELoss()
optimizer = torch.optim.Adam(model.parameters(), lr=lr)
```

# In [16]:

```
start = time.time()
#Training
for i in range(epoch): # again, normally you would NOT do 300 epochs, it is toy data
   j = 0
   total loss = 0
    while(j < train size ):</pre>
       k = j + 512;
        if k > train size:
           break
        #prepare input and output
        x_t = train_set[:,j:k,:]
        x t = torch.cuda.FloatTensor(x t)
       y_t = train_output[j:k,:]
        \#y_t = y_t.sum(axis=1)
       y t = torch.cuda.FloatTensor(y t)
        # Step 1. Remember that Pytorch accumulates gradients.
        # We need to clear them out before each instance
       model.zero grad()
        # Also, we need to clear out the hidden state of the LSTM,
        # detaching it from its history on the last instance.
       model.hidden = model.init hidden()
        # Step 2. Run our forward pass.
        y_pred = model(x_t)
        v pred = v pred.reshape((512.6))
```

```
# Step 4. Compute the loss, gradients, and update the parameters by
        # calling optimizer.step()
        loss = loss_fn(y_pred, y_t)
        total loss += loss
        loss.backward()
        optimizer.step()
        j = k
    print(total loss)
end = time.time()
print(end - start)
tensor(50.9755, device='cuda:0', grad fn=<ThAddBackward>)
tensor(26.5789, device='cuda:0', grad fn=<ThAddBackward>)
tensor(20.6462, device='cuda:0', grad fn=<ThAddBackward>)
tensor(18.2622, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(17.1125, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(16.5756, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(16.0052, device='cuda:0', grad fn=<ThAddBackward>)
tensor(15.6461, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(15.2870, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(15.0594, device='cuda:0', grad fn=<ThAddBackward>)
tensor(14.8109, device='cuda:0', grad fn=<ThAddBackward>)
tensor(14.5929, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(14.4486, device='cuda:0', grad fn=<ThAddBackward>)
tensor(14.2640, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(14.1223, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(14.0939, device='cuda:0', grad fn=<ThAddBackward>)
tensor(13.9003, device='cuda:0', grad fn=<ThAddBackward>)
tensor(13.8095, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(13.8289, device='cuda:0', grad fn=<ThAddBackward>)
tensor(13.5723, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(13.6184, device='cuda:0', grad fn=<ThAddBackward>)
tensor(13.6492, device='cuda:0', grad fn=<ThAddBackward>)
tensor(13.7871, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(13.3744, device='cuda:0', grad fn=<ThAddBackward>)
tensor(13.1718, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(13.0925, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(13.0016, device='cuda:0', grad fn=<ThAddBackward>)
tensor(12.8202, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.6674, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.7638, device='cuda:0', grad fn=<ThAddBackward>)
tensor(12.8251, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.6669, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.6374, device='cuda:0', grad fn=<ThAddBackward>)
tensor(12.5004, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.4183, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.3117, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.4911, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.3462, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.2893, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.2929, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.3215, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.1323, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.1822, device='cuda:0', grad fn=<ThAddBackward>)
tensor(12.0597, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.9271, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.9397, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.8751, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.9765, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.9381, device='cuda:0', grad fn=<ThAddBackward>)
```

tensor(12.0447, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.8232, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.7424, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.7341, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.8392, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.7028, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.7153, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.7565, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.6475, device='cuda:0', grad\_fn=<ThAddBackward>) tensor(11.6694, device='cuda:0', grad\_fn=<ThAddBackward>)

```
tensor(11.5199, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.5948, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.5233, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.5300, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.5061, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.3988, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.3948, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.4354, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.2962, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.4013, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.4855, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.3449, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.2600, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.2647, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.2138, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.4501, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.1831, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.0584, device='cuda:0', grad fn=<ThAddBackward>)
tensor(13.1655, device='cuda:0', grad fn=<ThAddBackward>)
tensor(12.6704, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.8622, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.6644, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.3593, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.2854, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.3863, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.3569, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.2513, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.0933, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.2250, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.1404, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.1831, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.0189, device='cuda:0', grad fn=<ThAddBackward>)
tensor(10.9212, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.0028, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(10.9860, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(10.8652, device='cuda:0', grad fn=<ThAddBackward>)
tensor(10.9190, device='cuda:0', grad fn=<ThAddBackward>)
tensor(11.0704, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(12.0151, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.7395, device='cuda:0', grad_fn=<ThAddBackward>)
tensor(11.3191, device='cuda:0', grad_fn=<ThAddBackward>)
797.7614636421204
```

# In [17]:

```
#test set
total = 10000
test_set_d = x_test[0:total]
test_output = y_test.values[0:total, :]
test_set = word_embedding_glove_total(test_set_d)
#test_set = base_glove_total(test_set_d)
print(len(glove))
```

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## In [18]:

[n [ ]:	3669		
[n [ ]:	[n [ ]:		
[n [ ]:			
	[n [ ]:		