### In [ ]:

# This ipython notebook shows how to train a GRU-based RNN classifier on IMDB dataset # using global pre-trained word vectors and achieve 92.075% test accuracy

#### In [ ]:

# Titan X Pascal (Driver Version: 375.26) # Keras Version: 1.2.2 # Theano Version: 0.9.0rc1 (cuda backend)

#### In [ ]:

000 TO GET DATA EXECUTE THE FOLLOWING SHELL SCRIPT # this script is based on https://github.com/mesnilgr/iclr15/blob/master/scripts/data.sh wget http://ai.stanford.edu/~amaas/data/sentiment/acllmdb\_v1.tar.gz tar -xvf aclImdb v1.tar.gz for j in train/pos train/neg test/pos test/neg train/unsup; do for i in `ls aclImdb/\$j'; do cat aclImdb/\$j/\$i >> temp; awk 'BEGIN{print;}' >> temp; done mv temp aclImdb/\$j/norm.txt done mkdir data mv aclImdb/train/pos/norm.txt data/full-train-pos.txt mv aclImdb/train/neg/norm.txt data/full-train-neg.txt mv aclImdb/test/pos/norm.txt data/test-pos.txt mv aclImdb/test/neg/norm.txt data/test-neg.txt cd data head -n 10000 full-train-pos.txt > small-train-pos.txt head -n 10000 full-train-neg.txt > small-train-neg.txt tail -n 2500 full-train-pos.txt > valid-pos.txt tail -n 2500 full-train-neg.txt > valid-neg.txt cd ... rm -r aclImdb

# In [ ]:

## In [1]:

%matplotlib inline import matplotlib.pyplot as plt

import os

 $os. environ ["THEANO\_FLAGS"] = "mode=FAST\_RUN, device=gpu0, floatX=float32" \\$ 

import numpy as np

rm acllmdb v1.tar.gz

np.random.seed(2525) # for reproducibility

import re import time

from gensim.models import Word2Vec

from keras.models import Sequential

from keras.layers.core import Dense, Dropout, Activation

from keras.layers.embeddings import Embedding

from keras.layers.recurrent import GRU

from keras.utils.generic\_utils import Progbar

from sklearn.metrics import precision\_recall\_fscore\_support as score

Using Theano backend.

WARNING (theano.sandbox.cuda): The cuda backend is deprecated and will be removed in the next release (v0.10). Pleas e switch to the gpuarray backend. You can get more information about how to switch at this URL: https://github.com/Theano/Theano/wiki/Converting-to-the-new-gpu-back-end%28gpuarray%29

Using gpu device 0: TITAN X (Pascal) (CNMeM is enabled with initial size: 95.0% of memory, cuDNN 5005)

```
# ******* 1. LOADING WORD VECTORS AND PREPARING EMBEDDING MATRIX **********************************
*****
starttime = time.time()
# to get GoogleNews vectors visit this page https://code.google.com/archive/p/word2vec/
w2vmodel = Word2Vec.load word2vec format("../vectors/GoogleNews-vectors-negative300.bin", binary=True)
embedding matrix = w2vmodel.syn0
# adding padding vector
embedding_matrix = np.insert(embedding_matrix, 0, np.zeros(300), axis=0) # pad index = 0
# adding unknown vector as averaged sum of 100 most rear word vectors based on from word2vec vocad
unk_vec = w2vmodel.syn0[-100:].sum(axis=0)/100
embedding_matrix = np.insert(embedding_matrix, 1, unk_vec, axis=0) # unk index = 1
print("embedding_matrix: "+str(embedding_matrix.shape)+" "+str(embedding_matrix.dtype))
print ("Took %ds" % (time.time() - starttime))
embedding_matrix: (3000002, 300) float32
Took 26s
In [3]:
starttime = time.time()
maxlen = 500 #max num of words in sample
def preproc(line):
  # 1. removing stopwords
 line = " ".join([word for word in line.split(" ") if word.lower() not in ["a","to","of","and"]])
  # 2. removing HTML tags
 line = re.sub(r'<[^>]+>', '', line)
  # 3. removing URLs
 line = re.sub(r'(https::\/\/(?:www.)[^\s\.]+\.[^\s]{2,}|www\.[^\s]+\.[^\s]{2,})', ' ', line)
  # 4. removing all non-alphabetic/numerical chars except \s and -
 line = re.sub(r'[^a-zA-Z0-9-\s]', ' ', line)
  # 5. replacing numbers with #
 line = re.sub(r'\d', '#', line)
  # 6. converting words into word ids
 def word2id(word):
   try: return w2vmodel.vocab[word].index+2;
    except: return 1 # unkown word id
 line = [word2id(word) for word in line.split()[-maxlen:]]
  # 7. adding pad vector (pad word id = 0)
 line += [0]*(maxlen-len(line))
  return line
def load imdb(filename):
 x,y = [],[]
  # loading negative and positive samples
 for yi,fnsuffix in enumerate(["-neg.txt","-pos.txt"]):
    with open("data/"+filename+fnsuffix, "r") as txtfile:
     for line in txtfile:
       x.append(preproc(line))
       y.append(yi)
 return np.array(x,dtype="int32"),np.array(y,dtype="int32")
#x_train,y_train = load_imdb("small-train")
\#x\_val,y\_val = load\_imdb("valid")
x_train,y_train = load_imdb("full-train")
x_test,y_test = load_imdb("test")
print("x train:"+str(x train.shape))
print("x_test:"+str(x_test.shape))
print ("Took %ds" % (time.time() - starttime))
x_train:(25000, 500)
x_test:(25000, 500)
Took 103s
In [4]:
```

```
starttime = time.time()
model = Sequential()
model.add(Embedding(embedding_matrix.shape[0],
            embedding_matrix.shape[1],
            weights=[embedding_matrix],
            input length=maxlen,
            trainable=False,
            mask zero=True,
            dropout=0.2))
model.add(GRU(output_dim=64,name="layer01",return_sequences=True,dropout_W=0.22,dropout_U=0.22))
model.add (GRU (output\_dim=64, name="layer02", return\_sequences=True, dropout\_W=0.22, dropout\_U=0.22))
model.add(GRU(output_dim=64,name="layer03",return sequences=True,dropout_W=0.22,dropout_U=0.22))
model.add(GRU(output_dim=64,name="layer04",return_sequences=True,dropout_W=0.22,dropout_U=0.22))
model.add(GRU(output\_dim=64,name="layer05",return\_sequences=False,dropout\_W=0.22,dropout\_U=0.22)) \\ model.add(Dense(output\_dim=1,name="layer06",activation="sigmoid")) \\
#model.load weights("imdb rnnw2v 140epoch",by name=True)
model.compile(loss="binary_crossentropy", optimizer="adam",metrics=['accuracy'])
model.summary()
print ("Took %ds" % (time.time() - starttime))
```

Layer (type)	Output Shape	Param #	Connected to	<del>_</del> 
embedding 1 (Embed	======================================		900000600 embedding input 1[0][0]	
layer01 (GRU)	(None, 500, 64)	70080	embedding_1[0][0]	_
layer02 (GRU)	(None, 500, 64)	24768	layer01[0][0]	
layer03 (GRU)	(None, 500, 64)	24768	layer02[0][0]	
layer04 (GRU)	(None, 500, 64)	24768	layer03[0][0]	_
layer05 (GRU)	(None, 64)	24768	layer04[0][0]	
layer06 (Dense)	(None, 1)	65 ======	layer05[0][0] =================================	
Total params: 900,16 Trainable params: 16 Non-trainable params	9,217	=		

Took 12s

In [5]:

```
# to free memory before training

del embedding_matrix

del w2vmodel
```

In [6]:

```
starttime = time.time()
def train(x_train,y_train,x_val=np.array([]),y_val=np.array([]),
      nb_epoch=100,save=True,plotMetrics=True):
 history = {"loss":[],"acc":[],"val_loss":[],"val_acc":[]}
 x slice size = 500 # batch size
 nb_batch = len(x_train)/x_slice_size + (0 if len(x_train)%x_slice_size == 0 else 1)
 print('Train on %d samples, validate on %d samples' % (x_train.shape[0], x_val.shape[0]))
 for epoch in range(nb_epoch):
    print("Epoch "+str(epoch+1)+"/"+str(nb_epoch))
   pbar = Progbar(len(x_train))
    shuffledIndices = np.arange(len(x train))
    np.random.shuffle(shuffledIndices)
    x_train = np.array([x_train[i] for i in shuffledIndices])
   y_train = np.array([y_train[i] for i in shuffledIndices])
    for hatch in range(nh hatch).
```

```
IVI DULCH III TUNGC(IID_DULCH).
     start = batch*x_slice_size
     stop = batch*x_slice_size+x_slice_size
     train = model.train\_on\_batch(x\_train[start:stop], y\_train[start:stop])
      pbar.update(stop,[("loss",train[0]),("acc",train[1])], force=True)
    history \verb|["loss"]|.append (pbar.sum_values["loss"][0]/float (pbar.sum_values["loss"][1])) \\
    history["acc"].append(pbar.sum_values["acc"][0]/float(pbar.sum_values["acc"][1]))
    if len(x val):
      val = model.evaluate(x_val, y_val, verbose=1,batch_size=1000)
     history["val_loss"].append(val[0])
history["val_acc"].append(val[1])
      print("validation: " + str(val))
  name = "imdb_rnnw2v_"+str(nb_epoch)+"epoch"
  if save:
    tmpmodel = Sequential()
    for layer in model.layers[1:]:
     tmpmodel.add(layer)
    tmpmodel.save_weights(name,overwrite=False)
    print("Saved as "+name)
  if plotMetrics:
    plt.figure(figsize=(12, 3))
    plt.plot(history['loss'],"b",label='Trainig loss')
    plt.plot(history['acc'],"y",label='Trainig accuracy')
    if len(x_val):
      plt.plot(history['val_loss'],"g",label='Validation loss')
      plt.plot(history['val_acc'],"r",label='Validation accuracy')
    plt.ylabel('values')
    plt.xlabel('epoch')
    plt.title(name)
    plt.grid(True)
    leg = plt.legend(loc=1,fancybox=True)
    leg.get_frame().set_alpha(0.0)
    plt.show()
  print("Done!")
train(x_train,y_train,nb_epoch=140)
print ("Took %ds" % (time.time() - starttime))
Train on 25000 samples, validate on 0 samples
Epoch 1/140
Epoch 140/140
Saved as imdb_rnnw2v_140epoch
                                 imdb_rnnw2v_140epoch

    Trainig loss

    Trainig accuracy

  0.8
  0.6
  0.4
                                                                 120
                                                                          140
Done!
Took 15964s
In [7]:
print(model.evaluate(x_test,y_test,batch_size=1000))
# F1 score
```

predictions = model.predict(x\_test,batch\_size=1000, verbose=1)

p = (predictions[:,0] > = 0.5).astype(int)

precision, recall, fscore, support = score(y\_test, p)
precision = np.around(precision\*100.decimals=2)

```
recall = np.around(recall*100,decimals=2)
fscore = np.around(fscore*100,decimals=2)
print("labels :[ neg pos]")
print("precision :"+str(precision))
print("recall :"+str(recall))
print("fscore :"+str(fscore))
print("-"*25)
print("fscore.mean: "+str(fscore.mean()))
print("test distribution :"+str(support))
25000/25000 [=========] - 15s
[0.20878124713897706, 0.92075999736785885]
25000/25000 [=========] - 15s
labels :[ neg pos]
precision: [92.11 92.04]
recall :[ 92.03 92.12]
fscore :[ 92.07 92.08]
fscore.mean: 92.075
test distribution :[12500 12500]
In [ ]:
In [ ]:
In [ ]:
In [7]:
# if you want to tune parameters yourself use "small-train" and "valid" datasets
# to avoid testset overfitting
# x_train,y_train = load_imdb("small-train")
\# x_{val,y_{val}} = load_{imdb("valid")}
starttime = time.time()
train(x_train,y_train,x_val,y_val,nb_epoch=200)
print ("Took %ds" % (time.time() - starttime))
Train on 20000 samples, validate on 5000 samples
Epoch 1/200
5000/5000 [========] - 3s
validation: [0.58315455913543701, 0.70420000553131101]
Epoch 200/200
5000/5000 [========] - 3s
validation: [0.27798883616924286, 0.90319999456405642]
Saved as imdb rnnw2v 200epoch
                                  imdb_rnnw2v_200epoch
                                                                    Trainig loss
                                                                    Trainig accuracy
  0.8
                                                                    Validation loss
                                                                   Validation accuracy
  0.6
  0.4
  0.2
                         50
                                         100
                                                  125
                                                          150
                                                                   175
                                                                            200
                                         epoch
Done!
Took 19260s
```

In [ ]: