Notebook

March 14, 2019

1 Import library

In [17]:
```
from __future__ import division, absolute_import, print_function
# import modules & set up logging
import gensim, logging
import smart_open, os
logging.basicConfig(format='%(asctime)s : %(levelname)s : %(message)s', level=logging
import datetime
import pandas as pd

# fichier incltu dans le projet
import save_notebook
import utils

import time
from six.moves import urllib, xrange
import tensorflow as tf
from tensorflow.contrib.tensorboard.plugins import projector
import numpy as np
import collections, math, os, random, zipfile
import glob
import pickle
import shutil
import sys
```

1.0.1 Check if tensorboard use GPU

In [2]:
```
from tensorflow.python.client import device_lib
print(device_lib.list_local_devices())
```

```
[name: "/device:CPU:0"
device_type: "CPU"
memory_limit: 268435456
locality {
}
incarnation: 2343263800986691833
, name: "/device:GPU:0"
device_type: "GPU"
```

```
memory_limit: 1461770649
locality {
  bus_id: 1
  links {
  }
}
incarnation: 6883058730262070735
physical_device_desc: "device: 0, name: GeForce GTX 960M, pci bus id: 0000:01:00.0, compute ca
]
```

2 Prepare input data

```
In [3]: word_embedding_model_name = "word2vec - SKIP GRAM"
        files =  os.listdir("../wikipedia/data")
        now = str(datetime.datetime.now()).replace(" ","")

In [4]: #Create a unique file big with one wikipedia page per line
        path="../wikipedia/data/"
        if not os.path.exists('./data/wikipedia_informatic.txt'):
            with open('./data/wikipedia_informatic.txt', 'w+',encoding="utf8" ) as out_file:
                for file in files:
                    if "ipynb_checkpoints" in file:
                        continue
                    try:
                        with open(path + file, encoding="utf8") as in_file:
                            out_file.write(in_file.read().replace("\n",""))
                    except:
                        continue
        else:
            # We tokenize the file
            with open('./data/wikipedia_informatic.txt', 'r', encoding="utf8") as f:
                wiki_vocab = f.readlines()
            wiki_vocab = [x.strip() for x in wiki_vocab]
            wiki_vocab_tokenized = []
            wiki_vocab_tokenized = gensim.utils.simple_preprocess(str(wiki_vocab))
```

3 Clean data

source https://github.com/udacity/deep-learning/blob/master/embeddings/Skip-Grams-Solution.ipynb

```
In [5]: vocabulary_size = 200000
        #https://medium.com/deep-math-machine-learning-ai/chapter-9-2-nlp-code-for-word2vec-ne

        def build_dataset(vocab_tokenized):
            dictionnary_count = {}
```

```
        dictionnary_count['UNK'] = -1
        dictionnary_count.update(collections.Counter(vocab_tokenized).most_common(vocabular

        dictionary = {}
        for word in dictionnary_count:
            dictionary[word] = len(dictionary)

    #     data_index = []

        unk_count = 0
        for word in vocab_tokenized:
            if word in dictionary:
                index = dictionary[word]
            else:
                index = 0  # dictionary['UNK']
                unk_count += 1

    #         data_index.append(index)

        dictionnary_count["UNK"] = unk_count
        reverse_dictionary = dict(zip(dictionary.values(), dictionary.keys()))

        return dictionnary_count, dictionary, reverse_dictionary


    dictionnary_count, dictionary, reverse_dictionary = build_dataset(wiki_vocab_tokenized)

    print('most common words (+UNK):', utils.take(10, dictionnary_count.items()))
    # print('sample data:', data_index[:10], [reverse_dictionary[i] for i in data_index[:1
```

```
most common words (+UNK): [('UNK', 0), ('de', 531786), ('la', 237117), ('le', 206228), ('et',
```

3.1 Subsampling

$$P(w_i) = 1 - \sqrt{\frac{t}{f(w_i)}}$$

where t is a threshold parameter and $f(w_i)$ is the frequency of word w_i in the total dataset.

```
In [6]: #subsampling data
        def subsampling_data(vocab_tokenized, reverse_dictionary, dictionary_count, threshold)
            np.random.seed(123)

        #     training_dictionary_subsample = dictionary_subsample
            training_dictionary_subsample = []
            size_of_corpus = len(vocab_tokenized)
```

```
            for index in reverse_dictionary:
                word = reverse_dictionary[index]
                if word == "UNK":
        #             dictionary_subsample[word] = -1
                    continue

                count_of_word = dictionnary_count[word]
                frequence_of_word = count_of_word / size_of_corpus
                probability = 1 - ( np.sqrt(threshold / frequence_of_word))

                if probability < np.random.random():
                    training_dictionary_subsample.append(index)

            return training_dictionary_subsample
```

```
In [7]: trainded_data = subsampling_data(wiki_vocab_tokenized, reverse_dictionary ,dictionnary
```

4 Definition of Model

```
In [8]: def get_target(words, index, window_size=5):
            R = np.random.randint(1, window_size+1)
            start = index - R if (index - R) > 0 else 0
            stop = index + R
            target_words = set(words[start:index] + words[index+1:stop+1])
            return list(target_words)
```

```
In [9]: def create_placeholders():
            x = tf.placeholder(tf.int64, [None], name="x")
            y = tf.placeholder(tf.int64, [None, None], name="y")
            return x, y
```

```
In [10]: def get_batches(words, batch_size, window_size=5):
             n_batches = len(words)//batch_size

             # only full batches
             words = words[:n_batches*batch_size]

             for index in range(0, len(words), batch_size):
                 x, y = [], []
                 step_batch = index+batch_size
                 batch = words[index:step_batch]
                 for i in range(len(batch)):
                     batch_x = batch[i]
                     batch_y = get_target(batch, i, window_size)
                     y.extend(batch_y)
                     x.extend([batch_x]*len(batch_y))
                 yield x, y
```

In [11]:
```
def get_embed(x, embedding_size, vocabulary_size):
    embedding_matrice = tf.Variable(tf.random_uniform([vocabulary_size,embedding_size]
    embed = tf.nn.embedding_lookup(embedding_matrice, x)

    return embedding_matrice, embed
```

In [12]:
```
def negative_sampling(vocab_size, embed, y, embedding_size, n_sampled = 100):
    # Number of negative y to sample

    softmax_w = tf.Variable(tf.truncated_normal((vocab_size, embedding_size), stddev=0
    softmax_b = tf.Variable(tf.zeros(vocab_size))

    # Calculate the loss using negative sampling

    loss = tf.nn.sampled_softmax_loss(softmax_w, softmax_b,
                                      y, embed,
                                      n_sampled, vocab_size)

    cost = tf.reduce_mean(loss)
    optimizer = tf.train.AdamOptimizer().minimize(cost)
    return loss, cost, optimizer
```

In [13]:
```
def get_result(embedding):
    norm = tf.sqrt(tf.reduce_sum(tf.square(embedding), 1, keepdims=True))
    normalized_embedding = embedding / norm
    return normalized_embedding
```

In [14]:
```
def print_result(normalized_embedding, id_word):

        ## From Thushan Ganegedara's implementation
    valid_size = 10 # Random set of words to evaluate similarity on.
    valid_window = 100
    # pick 8 samples from (0,100) and (1000,1100) each ranges. lower id implies more
    id_word = tf.constant(id_word)

#     np_valid_examples = np.array(id_word , valid_size//2)

#     np_valid_examples = np.append(np_valid_examples, random.sample(range(1000,1000+

    #valid_examples = tf.convert_to_tensor(np_valid_examples)
    #We use the cosine distance:
    valid_dataset = tf.Variable([id_word , valid_size//2])
#     valid_dataset = tf.constant(np_valid_examples)

    valid_embedding = tf.nn.embedding_lookup(normalized_embedding, valid_dataset)
    similarity = tf.matmul(valid_embedding, tf.transpose(normalized_embedding))
    return similarity, valid_dataset
```

In [15]:
```
def is_early_stopping(array_average_loss, max_steps_without_decrease):
    array_last_average_loss = array_average_loss[-max_steps_without_decrease:len(array
```

```
            for element in range(0, max_steps_without_decrease-1):
                # if one loss at step t is bigger than loss as t+1 so it is converging
                if array_last_average_loss[element] > array_last_average_loss[element +1]:
                    return False
            return True
```

In [16]:
```
def model_initializer(x, embedding_size, vocabulary_size, id_word):

    train_graph = tf.get_default_graph()

    #     on declare
    with train_graph.as_default():
        x, y = create_placeholders()
        embedding, x_embed = get_embed(x, embedding_size, vocabulary_size)
        loss, cost, optimizer = negative_sampling(vocabulary_size, x_embed, y, embeddi
        normalized_embedding = get_result(embedding)
        similarity, valid_examples = print_result(normalized_embedding, id_word)
        tf.summary.scalar("minibatch_cost", cost)
        merge = tf.summary.merge_all()

        saver = tf.train.Saver()
    return train_graph, x, y, x_embed, cost, optimizer, normalized_embedding, similari
```

In [29]:
```
def train(x_train, reverse_dictionary, epochs, batch_size, window_size, embedding_size
    # initialization of graph
    train_graph, x, y, x_embed, cost, optimizer, normalized_embedding, similarity, val

    #init var
    date_started = time.time()
    epoch_saver = tf.train.Saver(max_to_keep=3)  # keep 3 last iterations
    epoch_initializer = 0
    minibatch_cost_sum = 0

    #we store all result for early_stopping
    array_average_loss = []

    with tf.Session(graph=train_graph) as sess:

        iteration = 1
        sess.run(tf.global_variables_initializer())
        date_started = time.time()

         #If true we load the last checkpoint if exist
        if restore_last_session:
            try:
                    FIND_CHKP = False
                    LOG_DIR_SUBFOLDER = LOG_DIR.rsplit('/', 1)[0]+"/"
                    LOG_DIR_LIST = sorted(glob.glob(os.path.join(LOG_DIR_SUBFOLDER, '
```

```
                for folder in LOG_DIR_LIST:
                    if os.path.exists(folder + "save_model"):
                        ckpt = tf.train.get_checkpoint_state(folder +"/save_model/

                        #check if the checkpoint is the same architecture
                        if str(embedding_size) + "-epoch_saver" in str(ckpt):
                            LOG_DIR = folder
                            FIND_CHKP = True
                            break
                if FIND_CHKP is False:
                    sys.exit("no checkpoint to load")
                iteration = int(str(ckpt).rsplit("epoch_saver-i")[1].rsplit("i-")
                print("checkpoint load from directory " + LOG_DIR + " with archite

                epoch_saver.restore(sess, ckpt.model_checkpoint_path)
                epoch_initializer = int(ckpt.model_checkpoint_path.rsplit('i--',1)
                epochs = epochs + epoch_initializer

                with open (LOG_DIR +"/save_model/array_average_loss.save", 'rb') a
                    array_average_loss = pickle.load(file)

        except Exception as e: print(str(" can't load checkpoint => " + str(e)))


    #for tensorboard graph
    writer = tf.summary.FileWriter(LOG_DIR, sess.graph, filename_suffix=str(date_s

    for e in range(epoch_initializer, epochs+1):

        print("epoque ============================================================

        n_batches = len(x_train)//batch_size
        minibatch_data = get_batches(x_train, batch_size, window_size)

        minibatch_iteration = 0
        for minibatch_X, minibatch_Y in minibatch_data:

            minibatch_cost, _, merge_minibatch= sess.run([cost, optimizer, merge]

            minibatch_cost_sum = minibatch_cost_sum + minibatch_cost
            average_loss = minibatch_cost_sum / iteration
            array_average_loss.append(average_loss)

            average_loss = tf.Summary(value=[tf.Summary.Value(tag="average_loss",
            writer.add_summary(average_loss, iteration)


            #we stop learning it the loss does not decrease
```

```
                    if iteration > max_steps_without_decrease:
                        stop = is_early_stopping(array_average_loss, max_steps_without_dec
                        if stop:
                            print("stop because early stopping")
                            break

                    percent_in_epoch =  minibatch_iteration * 100 / utils.round_down(n_bat

                    if percent_in_epoch  % 25 == 0:
                        print("average loss  : "  + str(average_loss))
                        print("iteration => " + str(iteration) + " and " + str(percent_in_
                        sim, v_examples = sess.run([similarity,valid_examples], feed_dict


                        valid_word = reverse_dictionary[v_examples[0]]
                        top_k = 5 # number of nearest neighbors
                        nearest = (-sim[0, :]).argsort()[1:top_k+1]
                        log = 'Nearest to %s:' % valid_word
                        for k in range(top_k):
                            close_word = reverse_dictionary[nearest[k]]
                            log = '%s %s,' % (log, close_word)
                        print(log)

                        print("------------------------------------------------------------
                    iteration += 1
                    minibatch_iteration += + 1


                #we save model at each epoch
                if not os.path.exists(LOG_DIR +"/save_model/"):
                    os.makedirs(LOG_DIR +"/save_model")
                epoch_saver.save(sess,  LOG_DIR +"/save_model/" + str(embedding_size) + "-

                with open(LOG_DIR +"/save_model/array_average_loss.save", 'wb+') as file:
                    pickle.dump(array_average_loss, file)


            save_path = saver.save(sess,os.path.join(LOG_DIR ,'model'+ str(embedding_size)
            embed_mat = sess.run(normalized_embedding, feed_dict={x: np.array(trainded_dat

            return embed_mat, epochs, LOG_DIR
```

In [31]:
```
model_name = utils.get_model_name(word_embedding_model_name)
# LOG_DIR  = './tensorboard-graph/'+ model_name
tf.reset_default_graph()
date_before_learning = datetime.datetime.now()
```

In [32]:
```
id_word = dictionary["redhat"]
word_embedding, epochs, LOG_DIR = train(trainded_data, reverse_dictionary, 15, 16, 5,
```

```
        time_training = datetime.datetime.now() - date_before_learning

2019-03-14 18:51:10,698 : INFO : Restoring parameters from ./tensorboard-graph/word2vec - SKIP


checkpoint load from directory ./tensorboard-graph/word2vec - SKIP GRAM\03-14-18-36\ with archi
epoque ==========================================================================================
average loss  : value {
  tag: "average_loss"
  simple_value: 0.0003204015374649316
}

iteration => 33928 and 0.0 % for this epoch at time 2019-03-1418:51:11.149501
Nearest to redhat: simenon, afe, librestheodor, culturellela, concession,
--------------------------------------------------------------------------------
average loss  : value {
  tag: "average_loss"
  simple_value: 0.35378092527389526
}

iteration => 36753 and 25.0 % for this epoch at time 2019-03-1418:51:31.142871
Nearest to redhat: afe, simenon, rhl, peint, culturellela,
--------------------------------------------------------------------------------
average loss  : value {
  tag: "average_loss"
  simple_value: 0.5552752017974854
}

iteration => 39578 and 50.0 % for this epoch at time 2019-03-1418:51:52.430693
Nearest to redhat: afe, simenon, rhl, peint, culturellela,
--------------------------------------------------------------------------------
average loss  : value {
  tag: "average_loss"
  simple_value: 0.6949954032897949
}

iteration => 42403 and 75.0 % for this epoch at time 2019-03-1418:52:11.848346
Nearest to redhat: culturellela, afe, rhl, peint, professoral,
--------------------------------------------------------------------------------
average loss  : value {
  tag: "average_loss"
  simple_value: 0.797066867351532
}

iteration => 45228 and 100.0 % for this epoch at time 2019-03-1418:52:33.616246
Nearest to redhat: afe, culturellela, librestheodor, rhl, peint,
--------------------------------------------------------------------------------
```

```
epoque ================================================================================
average loss  : value {
  tag: "average_loss"
  simple_value: 0.7975736260414124
}

iteration => 45237 and 0.0 % for this epoch at time 2019-03-1418:52:34.407815
Nearest to redhat: afe, culturellela, librestheodor, rhl, peint,
--------------------------------------------------------------------------------
average loss  : value {
  tag: "average_loss"
  simple_value: 1.0240849256515503
}

iteration => 48062 and 25.0 % for this epoch at time 2019-03-1418:53:12.365607
Nearest to redhat: afe, culturellela, peint, rhl, professoral,
--------------------------------------------------------------------------------
stop because early stopping
epoque ================================================================================
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```

4.0.1 Create index label for tensorboard

```
In [ ]: f = open(os.path.join(LOG_DIR, "metadata.tsv"), 'w+', encoding="utf8")
        i = 0
        for idx in trainded_data:
            f.write(reverse_dictionary[idx] + '\n')
            if i ==  99999:
                break
            i = i + 1

        f.close()
```

4.1 Visu

```
In [ ]: train_graph = tf.get_default_graph()

        with tf.Session(graph=train_graph) as sess:

            #prepare for embedding on tensorboard
            embedding_writer = tf.summary.FileWriter(LOG_DIR +'/projector', sess.graph)
            config = projector.ProjectorConfig()
            embedding_conf = config.embeddings.add()
            embedding_conf.tensor_name = 'embedding'
            embedding_conf.metadata_path = os.path.join(LOG_DIR+'/projector' ,  'metadata.tsv')
            projector.visualize_embeddings(embedding_writer, config)
```

5 Save

```
In [ ]: name_notebook_exported = save_notebook.save_notebook("word2vec_with_tf.ipynb")
        utils.write_result(word_embedding_model_name, time_training, name_notebook_exported, ep

In [ ]:
```