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# Date: September 2016
# Project: Document Summarization
# H2020 Summa Project
# Comments: Jan 2017
# Improved for Reinforcement Learning
Document Summarization System
from __future__ import absolute_import
from __future__ import division
from __future__ import print_function
import math
import os
import random
import sys
import time
import numpy as np
import tensorflow as tf
from reward_utils import Reward_Generator
from data_utils import DataProcessor
from my_flags import FLAGS
from my_model import MY_Model
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########################## Batch Testing a model on some dataset ###############
def batch_predict_with_a_model(data, model, session=None):
  data_logits = []
  data labels = []
  data_weights = []
  step = 1
  while (step * FLAGS.batch size) <= len(data.fileindices):</pre>
   # Get batch data as Numpy Arrays : Without shuffling
   batch_docnames, batch_docs, batch_label, batch_weight, batch_oracle_multiple, batch_reward_multiple = data.get_batch(((step-1)*FLAGS.batch_size),
(step * FLAGS.batch size))
   batch logits = session.run(model.logits, feed dict={model.document placeholder: batch docs})
   data_logits.append(batch_logits)
   data labels.append(batch label)
   data_weights.append(batch_weight)
   # Increase step
   step += 1
  # Check if any data left
 if (len(data.fileindices) > ((step-1)*FLAGS.batch_size)):
   # Get last batch as Numpy Arrays
   batch docnames, batch docs, batch label, batch weight, batch oracle multiple, batch reward multiple = data.get batch(((step-1)*FLAGS.batch size),
len(data.fileindices))
   batch_logits = session.run(model.logits, feed_dict={model.document_placeholder: batch_docs})
   data logits.append(batch logits)
   data_labels.append(batch_label)
   data_weights.append(batch_weight)
   # print(data logits)
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data_logits = tf.concat(0, data_logits)
 data_lables = tf.concat(0, data_labels)
 data_weights = tf.concat(0, data_weights)
 # print(data_logits,data_lables,data_weights)
 return data_logits, data_lables, data_weights
def train():
 Training Mode: Create a new model and train the network
 # Training: use the tf default graph
 with tf.Graph().as_default() and tf.device(FLAGS.use_gpu):
   config = tf.ConfigProto(allow_soft_placement = True)
   # Start a session
   with tf.Session(config = config) as sess:
     ### Prepare data for training
     print("Prepare vocab dict and read pretrained word embeddings ...")
     vocab_dict, word_embedding_array = DataProcessor().prepare_vocab_embeddingdict()
     # vocab_dict contains _PAD and _UNK but not word_embedding_array
     print("Prepare training data ...")
     train_data = DataProcessor().prepare_news_data(vocab_dict, data_type="training")
     print("Prepare validation data ...")
     validation_data = DataProcessor().prepare_news_data(vocab_dict, data_type="validation")
     print("Prepare ROUGE reward generator ...")
     rouge_generator = Reward_Generator()
     # Create Model with various operations
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model = MY_Model(sess, len(vocab_dict)-2)
# Start training with some pretrained model
start epoch = 1
# selected modelpath = FLAGS.train dir+"/model.ckpt.epoch-"+str(start epoch-1)
# if not (os.path.isfile(selected_modelpath)):
# print("Model not found in checkpoint folder.")
# exit(0)
# # Reload saved model and test
# print("Reading model parameters from %s" % selected modelpath)
# model.saver.restore(sess, selected_modelpath)
# print("Model loaded.")
# Initialize word embedding before training
print("Initialize word embedding vocabulary with pretrained embeddings ...")
sess.run(model.vocab_embed_variable.assign(word_embedding_array))
# Reward aware training as part of Reward weighted CE ,
# No Curriculam learning: No annealing, consider full document like in MRT
# Multiple Samples (include gold sample), No future reward, Similar to MRT
# During training does not use PYROUGE to avoid multiple file rewritings
# Approximate MRT with multiple pre-estimated oracle samples
# June 2017: Use Single sample from multiple oracles
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for epoch in range(start epoch, FLAGS.train epoch wce + 1):
       print("MRT: Epoch "+str(epoch))
       print("MRT: Epoch "+str(epoch)+" : Reshuffle training document indices")
       train_data.shuffle_fileindices()
       print("MRT: Epoch "+str(epoch)+" : Restore Rouge Dict")
       rouge generator.restore rouge dict()
       # Start Batch Training
       step = 1
       while (step * FLAGS.batch size) <= len(train data.fileindices):</pre>
         # Get batch data as Numpy Arrays
         batch_docnames, batch_docs, batch_label, batch_weight, batch_oracle_multiple, batch_reward_multiple = train_data.get_batch(((step-
1)*FLAGS.batch_size),
                                                                                                                          (step * FLAGS.batch size))
         # print(batch docnames)
         # print(batch label[0])
         # print(batch_weight[0])
         # print(batch_oracle_multiple[0])
         # print(batch reward multiple[0])
         # exit(0)
         # Print the progress
         if (step % FLAGS.training checkpoint) == 0:
           ce_loss_val, ce_loss_sum, acc_val, acc_sum = sess.run([model.rewardweighted_cross_entropy_loss_multisample,
model.rewardweighted_ce_multisample_loss_summary,
                                                            model.accuracy, model.taccuracy summary],
                                                            feed dict={model.document placeholder: batch docs,
                                                                       model.predicted_multisample_label_placeholder: batch_oracle_multiple,
                                                                       model.actual_reward_multisample_placeholder: batch_reward_multiple,
                                                                       model.label_placeholder: batch_label,
                                                                       model.weight placeholder: batch weight})
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model.summary_writer.add_summary(ce_loss_sum, ((epoch-1)*len(train_data.fileindices)+ step*FLAGS.batch_size))
          model.summary_writer.add_summary(acc_sum, ((epoch-1)*len(train_data.fileindices)+step*FLAGS.batch_size))
          print("MRT: Epoch "+str(epoch)+" : Covered " + str(step*FLAGS.batch size)+"/"+str(len(train data.fileindices)) +
                ": Minibatch Reward Weighted Multisample CE Loss= {:.6f}".format(ce loss val) + ": Minibatch training accuracy=
{:.6f}".format(acc val))
        # Run optimizer: optimize policy network
         sess.run([model.train op policynet expreward], feed dict={model.document placeholder: batch docs,
                                                             model.predicted_multisample_label_placeholder: batch_oracle_multiple,
                                                              model.actual_reward_multisample_placeholder: batch_reward_multiple,
                                                             model.weight_placeholder: batch_weight})
        # Increase step
         step += 1
        # if step == 20:
        # break
       # Save Model
       print("MRT: Epoch "+str(epoch)+" : Saving model after epoch completion")
       checkpoint path = os.path.join(FLAGS.train dir, "model.ckpt.epoch-"+str(epoch))
       model.saver.save(sess, checkpoint_path)
       # Backup Rouge Dict
       print("MRT: Epoch "+str(epoch)+" : Saving rouge dictionary")
       rouge_generator.save_rouge_dict()
       # Performance on the validation set
       print("MRT: Epoch "+str(epoch)+" : Performance on the validation data")
       # Get Predictions: Prohibit the use of gold labels
       validation_logits, validation_labels, validation_weights = batch_predict_with_a_model(validation_data, model, session=sess)
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# Validation Accuracy and Prediction
       validation acc, validation sum = sess.run([model.final accuracy, model.vaccuracy summary], feed dict={model.logits placeholder:
validation_logits.eval(session=sess),
                                                                                                    model.label placeholder:
validation labels.eval(session=sess),
                                                                                                    model.weight placeholder:
validation weights.eval(session=sess)})
       # Print Validation Summary
       model.summary writer.add summary(validation sum, (epoch*len(train data.fileindices)))
       print("MRT: Epoch "+str(epoch)+" : Validation ("+str(len(validation data.fileindices))+") accuracy= {:.6f}".format(validation acc))
       # Writing validation predictions and final summaries
       print("MRT: Epoch "+str(epoch)+" : Writing final validation summaries")
       validation data.write prediction summaries(validation logits, "model.ckpt.epoch-"+str(epoch), session=sess)
       # Extimate Rouge Scores
       rouge_score = rouge_generator.get_full_rouge(FLAGS.train_dir+"/model.ckpt.epoch-"+str(epoch)+".validation-summary-topranked", "validation")
       print("MRT: Epoch "+str(epoch)+" : Validation ("+str(len(validation data.fileindices))+") rouge= {:.6f}".format(rouge score))
       # break
     print("Optimization Finished!")
```

```
def test():
 Test Mode: Loads an existing model and test it on the test set
 # Training: use the tf default graph
 with tf.Graph().as default() and tf.device(FLAGS.use gpu):
   config = tf.ConfigProto(allow_soft_placement = True)
   # Start a session
   with tf.Session(config = config) as sess:
     ### Prepare data for training
     print("Prepare vocab dict and read pretrained word embeddings ...")
     vocab_dict, word_embedding_array = DataProcessor().prepare_vocab_embeddingdict()
     # vocab_dict contains _PAD and _UNK but not word_embedding_array
     print("Prepare test data ...")
     test_data = DataProcessor().prepare_news_data(vocab_dict, data_type="test")
     # Create Model with various operations
     model = MY_Model(sess, len(vocab_dict)-2)
     # # Initialize word embedding before training
     \# print("Initialize word embedding vocabulary with pretrained embeddings ...")
     # sess.run(model.vocab_embed_variable.assign(word_embedding_array))
     # Select the model
     if (os.path.isfile(FLAGS.train_dir+"/model.ckpt.epoch-"+str(FLAGS.model_to_load))):
       selected_modelpath = FLAGS.train_dir+"/model.ckpt.epoch-"+str(FLAGS.model_to_load)
     else:
       print("Model not found in checkpoint folder.")
       exit(0)
     # Reload saved model and test
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```
print("Reading model parameters from %s" % selected modelpath)
     model.saver.restore(sess, selected_modelpath)
     print("Model loaded.")
     # Initialize word embedding before training
     print("Initialize word embedding vocabulary with pretrained embeddings ...")
     sess.run(model.vocab_embed_variable.assign(word_embedding_array))
     # Test Accuracy and Prediction
     print("Performance on the test data:")
     FLAGS.authorise_gold_label = False
     test_logits, test_labels, test_weights = batch_predict_with_a_model(test_data, model, session=sess)
     test_acc = sess.run(model.final_accuracy, feed_dict={model.logits_placeholder: test_logits.eval(session=sess),
                                                    model.label_placeholder: test_labels.eval(session=sess),
                                                    model.weight_placeholder: test_weights.eval(session=sess)})
     # Print Test Summary
     print("Test ("+str(len(test data.fileindices))+") accuracy= {:.6f}".format(test acc))
     # Writing test predictions and final summaries
     test_data.write_prediction_summaries(test_logits, "model.ckpt.epoch-"+str(FLAGS.model_to_load), session=sess)
def main( ):
 if FLAGS.exp_mode == "train":
   train()
  else:
   test()
if name__ == "__main__":
 tf.app.run()
```