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# Project: Document Summarization
# H2020 Summa Project
# Comments: Jan 2017
# Improved for Reinforcement Learning
Document Summarization Final Model
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
import model_docsum
from my_flags import FLAGS
import model_utils
```

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class MY_Model:
 def init (self, sess, vocab size):
   dtype = tf.float16 if FLAGS.use fp16 else tf.float32
   ### Few variables that has been initianlised here
   # Word embedding variable
   self.vocab embed variable = model utils.get vocab embed variable(vocab size)
   ### Define Place Holders
   self.document_placeholder = tf.placeholder("int32", [None,
                                                   (FLAGS.max doc length + FLAGS.max title length + FLAGS.max image length),
                                                   FLAGS.max sent length], name='doc-ph')
   self.label placeholder = tf.placeholder(dtype, [None, FLAGS.max_doc_length, FLAGS.target_label_size], name='label-ph')
   self.weight_placeholder = tf.placeholder(dtype, [None, FLAGS.max_doc_length], name='weight-ph')
   # Reward related place holders: Pass both rewards as place holders to make them constant for rl optimizer
   self.actual reward multisample placeholder = tf.placeholder(dtype, [None, 1], name='actual-reward-multisample-ph') # [FLAGS.batch size, Single
Sample]
   # Self predicted label placeholder
   self.predicted multisample label placeholder = tf.placeholder(dtype, [None, 1, FLAGS.max doc length, FLAGS.target label size], name='pred-
multisample-label-ph')
   # Only used for test/validation corpus
   self.logits placeholder = tf.placeholder(dtype, [None, FLAGS.max doc length, FLAGS.target label size], name='logits-ph')
   ### Define Policy Core Network: Consists of Encoder, Decoder and Convolution.
   self.extractor output, self.logits = model docsum.policy network(self.vocab embed variable, self.document placeholder, self.label placeholder)
   ### Define Reward-Weighted Cross Entropy Loss
   self.rewardweighted_cross_entropy_loss_multisample = model_docsum.reward_weighted_cross_entropy loss multisample(self.logits,
self.predicted_multisample_label_placeholder,
                                                                                                          self.actual reward multisample placeholder,
self.weight placeholder)
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### Define training operators
   self.train op policynet expreward = model docsum.train neg expectedreward(self.rewardweighted cross entropy loss multisample)
   # accuracy operation : exact match
   self.accuracy = model docsum.accuracy(self.logits, self.label placeholder, self.weight placeholder)
   # final accuracy operation
   self.final_accuracy = model_docsum.accuracy(self.logits_placeholder, self.label_placeholder, self.weight_placeholder)
   # Create a saver.
   self.saver = tf.train.Saver(tf.all_variables(), max_to_keep=None)
   # Scalar Summary Operations
   self.rewardweighted ce multisample loss summary = tf.summary.scalar("rewardweighted-cross-entropy-multisample-loss",
self.rewardweighted cross entropy loss multisample)
   self.taccuracy_summary = tf.summary.scalar("training_accuracy", self.accuracy)
   self.vaccuracy_summary = tf.summary.scalar("validation_accuracy", self.final_accuracy)
   # # Build the summary operation based on the TF collection of Summaries.
   # # self.summary_op = tf.merge_all_summaries()
   # Build an initialization operation to run below.
   init = tf.initialize all variables()
   # Start running operations on the Graph.
   sess.run(init)
   # Create Summary Graph for Tensorboard
   self.summary_writer = tf.summary.FileWriter(FLAGS.train_dir, sess.graph)
```