# 【关于 Bert 源码解析III 之 微调篇 】 那些你不知道的事



## 一、前言

本文 主要 解读 Bert 模型的 微调 模块代码:

• run\_classifier.py: 主要用于 文本分类 任务的微调

## 二、参数解析

```
flags = tf.flags
FLAGS = flags.FLAGS
 必要参数
1 1 1
# 数据地址
flags.DEFINE_string(
   "data_dir", None,
   "The input data dir. Should contain the .tsv files (or other data files) "
   "for the task.")
# Bert 配置文件地址
flags.DEFINE_string(
   "bert_config_file", None,
   "The config json file corresponding to the pre-trained BERT model."
   "This specifies the model architecture.")
# 训练任务
flags.DEFINE_string("task_name", None, "The name of the task to train.")
# Bert 词库
flags.DEFINE_string("vocab_file", None,
                   "The vocabulary file that the BERT model was trained on.")
# 训练输出 地址
flags.DEFINE_string(
```

```
"output_dir", None,
    "The output directory where the model checkpoints will be written.")
. . .
 其他参数
# 预训练 Bert 模型
flags.DEFINE_string(
   "init_checkpoint", None,
   "Initial checkpoint (usually from a pre-trained BERT model).")
# 是否小写
flags.DEFINE_bool(
   "do_lower_case", True,
    "Whether to lower case the input text. Should be True for uncased "
   "models and False for cased models.")
# 指定wordPiece tokenization 之后的sequence的最大长度,要求小于等于预训练模型的最大
sequence长度。当输入的数据长度小于max_seq_length时用0补齐,如果长度大于max_seq_length则
truncate处理;
flags.DEFINE_integer(
    "max_seq_length", 128,
   "The maximum total input sequence length after WordPiece tokenization."
   "Sequences longer than this will be truncated, and sequences shorter "
   "than this will be padded.")
# 训练
flags.DEFINE_bool("do_train", False, "Whether to run training.")
flags.DEFINE_bool("do_eval", False, "Whether to run eval on the dev set.")
# 预测
flags.DEFINE_bool(
    "do_predict", False,
   "Whether to run the model in inference mode on the test set.")
# 训练 Batch 大小
flags.DEFINE_integer("train_batch_size", 32, "Total batch size for training.")
# 评测 Batch 大小
flags.DEFINE_integer("eval_batch_size", 8, "Total batch size for eval.")
# 预测 Batch 大小
flags.DEFINE_integer("predict_batch_size", 8, "Total batch size for predict.")
# 学习率
flags.DEFINE_float("learning_rate", 5e-5, "The initial learning rate for Adam.")
# 训练 epochs
flags.DEFINE_float("num_train_epochs", 3.0,
                  "Total number of training epochs to perform.")
# 进行线性学习率预热的训练比例。
flags.DEFINE_float(
   "warmup_proportion", 0.1,
    "Proportion of training to perform linear learning rate warmup for. "
   "E.g., 0.1 = 10\% of training.")
# 保存模型 步长
flags.DEFINE_integer("save_checkpoints_steps", 1000,
                    "How often to save the model checkpoint.")
# 每个 estimator call 调用中要执行多少步
flags.DEFINE_integer("iterations_per_loop", 1000,
                    "How many steps to make in each estimator call.")
# 是否 使用 TPU
flags.DEFINE_bool("use_tpu", False, "Whether to use TPU or GPU/CPU.")
# TPU 名称
```

```
tf.flags.DEFINE_string(
    "tpu_name", None,
    "The Cloud TPU to use for training. This should be either the name " \,
    "used when creating the Cloud TPU, or a grpc://ip.address.of.tpu:8470 "
    "ur1.")
tf.flags.DEFINE_string(
    "tpu_zone", None,
    "[Optional] GCE zone where the Cloud TPU is located in. If not "
    "specified, we will attempt to automatically detect the GCE project from "
    "metadata.")
tf.flags.DEFINE_string(
    "gcp_project", None,
    "[Optional] Project name for the Cloud TPU-enabled project. If not "
    "specified, we will attempt to automatically detect the GCE project from "
    "metadata.")
tf.flags.DEFINE_string("master", None, "[Optional] TensorFlow master URL.")
flags.DEFINE_integer(
    "num_tpu_cores", 8,
    "Only used if `use_tpu` is True. Total number of TPU cores to use.")
```

## 三、输入数据实例

```
class InputExample(object):
    """A single training/test example for simple sequence classification."""

def __init__(self, guid, text_a, text_b=None, label=None):
    """Constructs a InputExample.

Args:
    guid: 实例 唯一 id
    text_a: string. 第一个序列的未标记文本。 对于单序列任务,仅必须指定此序列。
    text_b: (Optional) string. 第二个序列的未标记文本。 仅必须为序列对任务指定。
    label: (Optional) string. 实例的标签。 应该为train和dev实例指定此名称,但不为测试实例指定,如果是test数据集则label统一为0。
    """

self.guid = guid
    self.text_a = text_a
    self.text_b = text_b
    self.label = label
```

# 四、特定任务数据处理

#### 4.1 数据处理 接口

• 作用:数据预处理接口

```
class DataProcessor(object):
"""Base class for data converters for sequence classification data sets."""
```

```
def get_train_examples(self, data_dir):
  """Gets a collection of `InputExample`s for the train set."""
  raise NotImplementedError()
def get_dev_examples(self, data_dir):
  """Gets a collection of `InputExample`s for the dev set."""
  raise NotImplementedError()
def get_test_examples(self, data_dir):
  """Gets a collection of `InputExample`s for prediction."""
  raise NotImplementedError()
def get_labels(self):
  """Gets the list of labels for this data set."""
  raise NotImplementedError()
@classmethod
def _read_tsv(cls, input_file, quotechar=None):
  """Reads a tab separated value file."""
 with tf.gfile.Open(input_file, "r") as f:
    reader = csv.reader(f, delimiter="\t", quotechar=quotechar)
   lines = []
    for line in reader:
     lines.append(line)
    return lines
```

#### 4.2 推理任务 数据集处理

- 目标:读取两句话,并判定两者间的关系是否为"蕴含"(Entailment)、"矛盾"(Contradict)或"中性"(Neutral)
- 具体任务数据介绍: FAIR重磅发布大规模语料库XNLI: 支持15种语言, 解决跨语言理解难题

```
class XnliProcessor(DataProcessor):
  """Processor for the XNLI data set."""
  def __init__(self):
   self.language = "zh"
  def get_train_examples(self, data_dir):
    """See base class."""
    lines = self._read_tsv(
        os.path.join(data_dir, "multinli",
                     "multinli.train.%s.tsv" % self.language))
    examples = []
    for (i, line) in enumerate(lines):
     if i == 0:
        continue
      guid = "train-%d" % (i)
      text_a = tokenization.convert_to_unicode(line[0])
      text_b = tokenization.convert_to_unicode(line[1])
      label = tokenization.convert_to_unicode(line[2])
      if label == tokenization.convert_to_unicode("contradictory"):
        label = tokenization.convert_to_unicode("contradiction")
      examples.append(
```

```
InputExample(guid=guid, text_a=text_a, text_b=text_b, label=label))
  return examples
def get_dev_examples(self, data_dir):
  """See base class."""
  lines = self._read_tsv(os.path.join(data_dir, "xnli.dev.tsv"))
  examples = []
  for (i, line) in enumerate(lines):
   if i == 0:
      continue
    guid = "dev-%d" % (i)
    language = tokenization.convert_to_unicode(line[0])
   if language != tokenization.convert_to_unicode(self.language):
      continue
    text_a = tokenization.convert_to_unicode(line[6])
    text_b = tokenization.convert_to_unicode(line[7])
    label = tokenization.convert_to_unicode(line[1])
    examples.append(
        InputExample(guid=guid, text_a=text_a, text_b=text_b, label=label))
  return examples
def get_labels(self):
  """See base class."""
  return ["contradiction", "entailment", "neutral"]
```

#### 4.3 二分类任务 数据集处理

```
class ColaProcessor(DataProcessor):
  """Processor for the CoLA data set (GLUE version)."""
 def get_train_examples(self, data_dir):
   """See base class."""
   return self._create_examples(
       self._read_tsv(os.path.join(data_dir, "train.tsv")), "train")
 def get_dev_examples(self, data_dir):
   """See base class."""
   return self._create_examples(
       self._read_tsv(os.path.join(data_dir, "dev.tsv")), "dev")
 def get_test_examples(self, data_dir):
   """See base class."""
   return self._create_examples(
       self._read_tsv(os.path.join(data_dir, "test.tsv")), "test")
 def get_labels(self):
   """See base class."""
   return ["0", "1"]
 def _create_examples(self, lines, set_type):
   """Creates examples for the training and dev sets."""
   examples = []
   for (i, line) in enumerate(lines):
     # Only the test set has a header
```

```
if set_type == "test" and i == 0:
    continue
guid = "%s-%s" % (set_type, i)
if set_type == "test":
    text_a = tokenization.convert_to_unicode(line[1])
    label = "0"
else:
    text_a = tokenization.convert_to_unicode(line[3])
    label = tokenization.convert_to_unicode(line[1])
    examples.append(
        InputExample(guid=guid, text_a=text_a, text_b=None, label=label))
return examples
```

# 五、examples转换成features (file\_based\_convert\_examples\_to\_features)

#### 5.1 单例转化

- 作用:将单个InputExample转换为单个InputFeatures。
- 流程:

```
step 1: 判断 example 是否是 PaddingInputExample
step 2: 构建 label map
step 3: text_a 序列化
step 4: text_b 序列化
step 5: 训练 长度修改
step 6: 输入数据 转化未 Bert 所要求类型数据
step 7:输入数据 转化为 id 系列
step 8: Mask 数据
step 9: 利用 0 填充
step 10: 标签 处理
```

o step 11:构建 InputExample

def convert\_single\_example( ex\_index, example. label\_list, max\_seq\_length, tokenizer): """将单个 InputExample 转换为单个InputFeatures。""" # step 1: 判断 example 是否是 PaddingInputExample if isinstance(example, PaddingInputExample): return InputFeatures( input\_ids=[0] \* max\_seq\_length, input\_mask=[0] \* max\_seq\_length, segment\_ids=[0] \* max\_seq\_length, label\_id=0. is\_real\_example=False) # step 2: 构建 label map  $label_map = {}$ for (i, label) in enumerate(label\_list):  $label_map[label] = i$ 

```
# step 3: text_a 序列化
 tokens_a = tokenizer.tokenize(example.text_a)
 # step 4: text_b 序列化
 tokens_b = None
 if example.text_b:
   tokens_b = tokenizer.tokenize(example.text_b)
 # step 5: 训练 长度修改
 if tokens_b:
   # 在适当位置修改`tokens_a`和`tokens_b`,以使总长度小于指定长度。
   # Account for [CLS], [SEP], [SEP] with "- 3"
   _truncate_seq_pair(tokens_a, tokens_b, max_seq_length - 3)
 else:
   # Account for [CLS] and [SEP] with "- 2"
   if len(tokens_a) > max_seq_length - 2:
     tokens_a = tokens_a[0:(max_seq_length - 2)]
 # step 6: 输入数据 转化未 Bert 所要求类型数据
 # The convention in BERT is:
 # (a) For sequence pairs:
 # tokens: [CLS] is this jack ##son ##ville ? [SEP] no it is not . [SEP]
                   0 0
                                    0 00 11111
 # type_ids: 0
                         0 0
 # (b) For single sequences:
 # tokens: [CLS] the dog is hairy . [SEP]
                  0 0
                          0 0
                                   0 0
 # type_ids: 0
 # Where "type_ids" are used to indicate whether this is the first sequence or
the second sequence. The embedding vectors for `type=0` and `type=1` were learned
during pre-training and are added to the wordpiece embedding vector (and position
vector). This is not *strictly* necessary since the [SEP] token unambiguously
separates the sequences, but it makes it easier for the model to learn the
concept of sequences.
 # For classification tasks, the first vector (corresponding to [CLS]) is used
as the "sentence vector". Note that this only makes sense because the entire
model is fine-tuned.
 tokens = []
 segment_ids = []
 tokens.append("[CLS]")
 segment_ids.append(0)
 for token in tokens_a:
   tokens.append(token)
   segment_ids.append(0)
 tokens.append("[SEP]")
 segment_ids.append(0)
 if tokens b:
   for token in tokens_b:
     tokens.append(token)
     segment_ids.append(1)
   tokens.append("[SEP]")
   segment_ids.append(1)
 # step 7:输入数据 转化为 id 系列
 input_ids = tokenizer.convert_tokens_to_ids(tokens)
```

```
# step 8: Mask 数据
# The mask has 1 for real tokens and 0 for padding tokens. Only real
# tokens are attended to.
input_mask = [1] * len(input_ids)
# step 9: 利用 0 填充
# Zero-pad up to the sequence length.
while len(input_ids) < max_seq_length:</pre>
  input_ids.append(0)
  input_mask.append(0)
  segment_ids.append(0)
assert len(input_ids) == max_seq_length
assert len(input_mask) == max_seq_length
assert len(segment_ids) == max_seq_length
# step 10: 标签 处理
label_id = label_map[example.label]
if ex_index < 5:
 tf.logging.info("*** Example ***")
 tf.logging.info("guid: %s" % (example.guid))
  tf.logging.info("tokens: %s" % " ".join(
      [tokenization.printable_text(x) for x in tokens]))
  tf.logging.info("input_ids: %s" % " ".join([str(x) for x in input_ids]))
  tf.logging.info("input_mask: %s" % " ".join([str(x) for x in input_mask]))
  tf.logging.info("segment_ids: %s" % " ".join([str(x) for x in segment_ids]))
  tf.logging.info("label: %s (id = %d)" % (example.label, label_id))
# step 11:构建 InputFeatures 实例
feature = InputFeatures(
    input_ids=input_ids,
    input_mask=input_mask,
    segment_ids=segment_ids,
    label_id=label_id,
    is_real_example=True)
return feature
```

## 5.2 单例转化

```
features = collections.OrderedDict()
features["input_ids"] = create_int_feature(feature.input_ids)
features["input_mask"] = create_int_feature(feature.input_mask)
features["segment_ids"] = create_int_feature(feature.segment_ids)
features["label_ids"] = create_int_feature([feature.label_id])
features["is_real_example"] = create_int_feature(
        [int(feature.is_real_example)])

tf_example = tf.train.Example(features=tf.train.Features(feature=features))
writer.write(tf_example.SerializeToString())
writer.close()
```

## 六、创建模型

#### 6.1 create\_model 创建 分类模型

```
def create_model(bert_config, is_training, input_ids, input_mask, segment_ids,
                 labels, num_labels, use_one_hot_embeddings):
  """创建 分类模型"""
  model = modeling.BertModel(
     config=bert_config,
     is_training=is_training,
      input_ids=input_ids,
      input_mask=input_mask,
      token_type_ids=segment_ids,
      use_one_hot_embeddings=use_one_hot_embeddings)
  # In the demo, we are doing a simple classification task on the entire
segment.
  # If you want to use the token-level output, use model.get_sequence_output()
instead.
  output_layer = model.get_pooled_output()
  hidden_size = output_layer.shape[-1].value
  output_weights = tf.get_variable(
      "output_weights", [num_labels, hidden_size],
      initializer=tf.truncated_normal_initializer(stddev=0.02))
  output_bias = tf.get_variable(
      "output_bias", [num_labels], initializer=tf.zeros_initializer())
  # 计算损失函数
  with tf.variable_scope("loss"):
   if is_training:
      # I.e., 0.1 dropout
      output_layer = tf.nn.dropout(output_layer, keep_prob=0.9)
    logits = tf.matmul(output_layer, output_weights, transpose_b=True)
    logits = tf.nn.bias_add(logits, output_bias)
    probabilities = tf.nn.softmax(logits, axis=-1)
    log_probs = tf.nn.log_softmax(logits, axis=-1)
```

```
one_hot_labels = tf.one_hot(labels, depth=num_labels, dtype=tf.float32)

per_example_loss = -tf.reduce_sum(one_hot_labels * log_probs, axis=-1)
loss = tf.reduce_mean(per_example_loss)

return (loss, per_example_loss, logits, probabilities)
```

#### 6.2 model\_fn\_builder

• 作用:

```
def model_fn_builder(bert_config, num_labels, init_checkpoint, learning_rate,
                    num_train_steps, num_warmup_steps, use_tpu,
                    use_one_hot_embeddings):
  """Returns `model_fn` closure for TPUEstimator."""
 def model_fn(features, labels, mode, params): # pylint: disable=unused-
argument
   """The `model_fn` for TPUEstimator."""
   tf.logging.info("*** Features ***")
   for name in sorted(features.keys()):
     tf.logging.info(" name = %s, shape = %s" % (name, features[name].shape))
   input_ids = features["input_ids"]
   input_mask = features["input_mask"]
   segment_ids = features["segment_ids"]
   label_ids = features["label_ids"]
   is_real_example = None
   if "is_real_example" in features:
     is_real_example = tf.cast(features["is_real_example"], dtype=tf.float32)
   else:
     is_real_example = tf.ones(tf.shape(label_ids), dtype=tf.float32)
   is_training = (mode == tf.estimator.ModeKeys.TRAIN)
   # 总的损失定义为两者之和
    (total_loss, per_example_loss, logits, probabilities) = create_model(
       bert_config, is_training, input_ids, input_mask, segment_ids, label_ids,
       num_labels, use_one_hot_embeddings)
   # 获取所有变量
   tvars = tf.trainable_variables()
   initialized_variable_names = {}
   scaffold_fn = None
   # 如果有之前保存的模型,则进行恢复
   if init_checkpoint:
      (assignment_map, initialized_variable_names
     ) = modeling.get_assignment_map_from_checkpoint(tvars, init_checkpoint)
     if use_tpu:
       def tpu_scaffold():
         tf.train.init_from_checkpoint(init_checkpoint, assignment_map)
          return tf.train.Scaffold()
       scaffold_fn = tpu_scaffold
```

```
else:
      tf.train.init_from_checkpoint(init_checkpoint, assignment_map)
 tf.logging.info("**** Trainable Variables ****")
 for var in tvars:
   init_string = ""
   if var.name in initialized_variable_names:
      init_string = ", *INIT_FROM_CKPT*"
    tf.logging.info(" name = %s, shape = %s%s", var.name, var.shape,
                   init_string)
  # 训练过程,获得spec
  output_spec = None
  if mode == tf.estimator.ModeKeys.TRAIN:
    train_op = optimization.create_optimizer(
        total_loss, learning_rate, num_train_steps, num_warmup_steps, use_tpu)
    output_spec = tf.contrib.tpu.TPUEstimatorSpec(
       mode=mode,
       loss=total_loss,
       train_op=train_op,
        scaffold_fn=scaffold_fn)
  # 验证过程spec
  elif mode == tf.estimator.ModeKeys.EVAL:
    def metric_fn(per_example_loss, label_ids, logits, is_real_example):
      predictions = tf.argmax(logits, axis=-1, output_type=tf.int32)
      accuracy = tf.metrics.accuracy(
          labels=label_ids, predictions=predictions, weights=is_real_example)
      loss = tf.metrics.mean(values=per_example_loss, weights=is_real_example)
      return {
          "eval_accuracy": accuracy,
          "eval_loss": loss,
      }
    eval_metrics = (metric_fn,
                    [per_example_loss, label_ids, logits, is_real_example])
    output_spec = tf.contrib.tpu.TPUEstimatorSpec(
       mode=mode,
       loss=total_loss,
       eval_metrics=eval_metrics,
        scaffold_fn=scaffold_fn)
 # 预测过程spec
  else:
    output_spec = tf.contrib.tpu.TPUEstimatorSpec(
        predictions={"probabilities": probabilities},
        scaffold_fn=scaffold_fn)
  return output_spec
return model_fn
```

# 七、主入口

```
def main(_):
 tf.logging.set_verbosity(tf.logging.INFO)
 # 任务处理器 映射表
 processors = {
     "cola": ColaProcessor,
     "mnli": MnliProcessor,
     "mrpc": MrpcProcessor,
     "xnli": XnliProcessor,
 }
 tokenization.validate_case_matches_checkpoint(FLAGS.do_lower_case,
                                                FLAGS.init_checkpoint)
 if not FLAGS.do_train and not FLAGS.do_eval and not FLAGS.do_predict:
   raise ValueError(
       "At least one of `do_train`, `do_eval` or `do_predict' must be True.")
 # 加载 Bert 配置
 bert_config = modeling.BertConfig.from_json_file(FLAGS.bert_config_file)
 if FLAGS.max_seq_length > bert_config.max_position_embeddings:
   raise ValueError(
        "Cannot use sequence length %d because the BERT model "
        "was only trained up to sequence length %d" %
        (FLAGS.max_seq_length, bert_config.max_position_embeddings))
 tf.gfile.MakeDirs(FLAGS.output_dir)
 task_name = FLAGS.task_name.lower()
 if task_name not in processors:
   raise ValueError("Task not found: %s" % (task_name))
 # 定义任务处理器
 processor = processors[task_name]()
 # 获取标签项
 label_list = processor.get_labels()
 # 数据预处理
 tokenizer = tokenization.FullTokenizer(
     vocab_file=FLAGS.vocab_file, do_lower_case=FLAGS.do_lower_case)
 tpu_cluster_resolver = None
 if FLAGS.use_tpu and FLAGS.tpu_name:
   tpu_cluster_resolver = tf.contrib.cluster_resolver.TPUClusterResolver(
        FLAGS.tpu_name, zone=FLAGS.tpu_zone, project=FLAGS.gcp_project)
 is_per_host = tf.contrib.tpu.InputPipelineConfig.PER_HOST_V2
  run_config = tf.contrib.tpu.RunConfig(
     cluster=tpu_cluster_resolver,
     master=FLAGS.master,
     model_dir=FLAGS.output_dir,
     save_checkpoints_steps=FLAGS.save_checkpoints_steps,
     tpu_config=tf.contrib.tpu.TPUConfig(
          iterations_per_loop=FLAGS.iterations_per_loop,
         num_shards=FLAGS.num_tpu_cores,
          per_host_input_for_training=is_per_host))
```

```
train_examples = None
num_train_steps = None
num_warmup_steps = None
# 模型训练 数据加载
if FLAGS.do_train:
 # 加载训练数据
 train_examples = processor.get_train_examples(FLAGS.data_dir)
  num_train_steps = int(
      len(train_examples) / FLAGS.train_batch_size * FLAGS.num_train_epochs)
  num_warmup_steps = int(num_train_steps * FLAGS.warmup_proportion)
# 自定义模型用于estimator训练
model_fn = model_fn_builder(
    bert_config=bert_config,
    num_labels=len(label_list),
    init_checkpoint=FLAGS.init_checkpoint,
    learning_rate=FLAGS.learning_rate,
    num_train_steps=num_train_steps,
    num_warmup_steps=num_warmup_steps,
    use_tpu=FLAGS.use_tpu,
    use_one_hot_embeddings=FLAGS.use_tpu)
# 如果没有TPU,会自动转为CPU/GPU的Estimator
estimator = tf.contrib.tpu.TPUEstimator(
    use_tpu=FLAGS.use_tpu,
   model_fn=model_fn,
    config=run_config,
    train_batch_size=FLAGS.train_batch_size,
    eval_batch_size=FLAGS.eval_batch_size,
    predict_batch_size=FLAGS.predict_batch_size)
# 模型 训练
if FLAGS.do train:
 train_file = os.path.join(FLAGS.output_dir, "train.tf_record")
  file_based_convert_examples_to_features(
      train_examples, label_list, FLAGS.max_seq_length, tokenizer, train_file)
  tf.logging.info("***** Running training *****")
  tf.logging.info(" Num examples = %d", len(train_examples))
 tf.logging.info(" Batch size = %d", FLAGS.train_batch_size)
 tf.logging.info(" Num steps = %d", num_train_steps)
  train_input_fn = file_based_input_fn_builder(
      input_file=train_file,
      seq_length=FLAGS.max_seq_length,
      is_training=True,
      drop_remainder=True)
  estimator.train(input_fn=train_input_fn, max_steps=num_train_steps)
# 模型 验证 数据加载
if FLAGS.do_eval:
  eval_examples = processor.get_dev_examples(FLAGS.data_dir)
  num_actual_eval_examples = len(eval_examples)
  if FLAGS.use_tpu:
    # TPU requires a fixed batch size for all batches, therefore the number
    # of examples must be a multiple of the batch size, or else examples
    # will get dropped. So we pad with fake examples which are ignored
    # later on. These do NOT count towards the metric (all tf.metrics
    # support a per-instance weight, and these get a weight of 0.0).
   while len(eval_examples) % FLAGS.eval_batch_size != 0:
```

```
eval_examples.append(PaddingInputExample())
 eval_file = os.path.join(FLAGS.output_dir, "eval.tf_record")
 file_based_convert_examples_to_features(
      eval_examples, label_list, FLAGS.max_seq_length, tokenizer, eval_file)
 tf.logging.info("***** Running evaluation *****")
 tf.logging.info(" Num examples = %d (%d actual, %d padding)",
                 len(eval_examples), num_actual_eval_examples,
                  len(eval_examples) - num_actual_eval_examples)
 tf.logging.info(" Batch size = %d", FLAGS.eval_batch_size)
 # This tells the estimator to run through the entire set.
 eval_steps = None
 # However, if running eval on the TPU, you will need to specify the
 # number of steps.
 if FLAGS.use_tpu:
   assert len(eval_examples) % FLAGS.eval_batch_size == 0
   eval_steps = int(len(eval_examples) // FLAGS.eval_batch_size)
 eval_drop_remainder = True if FLAGS.use_tpu else False
 eval_input_fn = file_based_input_fn_builder(
     input_file=eval_file,
     seq_length=FLAGS.max_seq_length,
     is_training=False,
     drop_remainder=eval_drop_remainder)
 result = estimator.evaluate(input_fn=eval_input_fn, steps=eval_steps)
 output_eval_file = os.path.join(FLAGS.output_dir, "eval_results.txt")
 with tf.gfile.GFile(output_eval_file, "w") as writer:
   tf.logging.info("***** Eval results *****")
   for key in sorted(result.keys()):
     tf.logging.info(" %s = %s", key, str(result[key]))
     writer.write("%s = %s\n" % (key, str(result[key])))
# 模型预测
if FLAGS.do_predict:
 predict_examples = processor.get_test_examples(FLAGS.data_dir)
 num_actual_predict_examples = len(predict_examples)
 if FLAGS.use_tpu:
   # TPU requires a fixed batch size for all batches, therefore the number
   # of examples must be a multiple of the batch size, or else examples
   # will get dropped. So we pad with fake examples which are ignored
   # later on.
   while len(predict_examples) % FLAGS.predict_batch_size != 0:
     predict_examples.append(PaddingInputExample())
 predict_file = os.path.join(FLAGS.output_dir, "predict.tf_record")
 file_based_convert_examples_to_features(predict_examples, label_list,
                                          FLAGS.max_seq_length, tokenizer,
                                          predict_file)
 tf.logging.info("***** Running prediction*****")
 tf.logging.info(" Num examples = %d (%d actual, %d padding)",
                 len(predict_examples), num_actual_predict_examples,
```

```
len(predict_examples) - num_actual_predict_examples)
tf.logging.info(" Batch size = %d", FLAGS.predict_batch_size)
predict_drop_remainder = True if FLAGS.use_tpu else False
predict_input_fn = file_based_input_fn_builder(
    input_file=predict_file,
    seq_length=FLAGS.max_seq_length,
    is_training=False,
    drop_remainder=predict_drop_remainder)
result = estimator.predict(input_fn=predict_input_fn)
output_predict_file = os.path.join(FLAGS.output_dir, "test_results.tsv")
with tf.gfile.GFile(output_predict_file, "w") as writer:
  num_written_lines = 0
  tf.logging.info("***** Predict results *****")
  for (i, prediction) in enumerate(result):
    probabilities = prediction["probabilities"]
    if i >= num_actual_predict_examples:
      break
    output_line = "\t".join(
        str(class_probability)
        for class_probability in probabilities) + "\n"
    writer.write(output_line)
    num_written_lines += 1
assert num_written_lines == num_actual_predict_examples
```

## 八、总结

本章 主要介绍了利用 Bert fineturn, 代码比较简单。

## 参考文档

- 1. Bert系列(四)——源码解读之Fine-tune
- 2. BERT源码分析PART III