

# 3.基于LSTM+CTC实现不定长文本图片OCR



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上一篇实现了图片CNN多标签分类（4位定长验证码识别任务）

（地址：<https://www.jianshu.com/p/596db72a7e00>

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本文继续优化，实现不定长文本图片的识别任务

下一篇考虑玩一玩GAN网络

本文所用到的10w不定长验证码文本数据集百度网盘下载地址（也可使用下文代码自行生成）：<https://pan.baidu.com/s/11BzlvuT4pYw3B0aFCK0ndQ>

<https://pan.baidu.com/s/11BzlvuT4pYw3B0aFCK0ndQ>)

利用本文代码训练并生成的模型（对应项目中的my-model文件夹）：

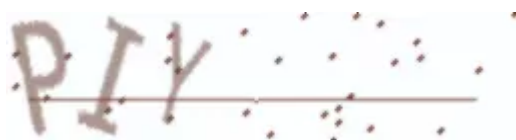
<https://pan.baidu.com/s/1AoKtZVyscWp3ZdOQU71qLA>

<https://pan.baidu.com/s/1AoKtZVyscWp3ZdOQU71qLA>)

项目简介：

需要预先安装pip install captcha==0.1.1,pip install opencv-python,pip install flask, pip install tensorflow/pip install tensorflow-gpu)

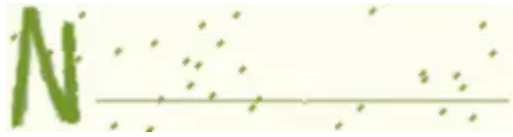
本文采用LSTM+CTC实现1-10位不定长验证码图片OCR（生成的验证码由随机的1-10位大写字母组成），本质上是一张图片多个标签的分类问题，且每个图片的标签数量不固定（数据如下图所示）



0\_PiY.png



1\_BCAVDPXT.png



2\_N.png

整体训练逻辑：

- 1, 将图像传入到LSTM中获得sequence, 和sequence的长度（大致的原理是：将图像的width看做LSTM中的time\_step, 将图像的height看做每个time\_step输入tensor的size)
- 2, 将真实的y\_label转为稀疏矩阵张量（此处的sparseTensor是个重点, 同学们可以把代码中的153行y\_train\_tmp打印出来观察一下)
- 3, 损失函数采用tf.nn.ctc\_loss, 然后对以上两步获得的数据进行训练, 最终使得损失函数尽可能的减小

关于ctc\_loss的原理可以百度科普一下, 它的主要作用可以大概理解为将上层网络预测出的AAABBBBCCDEE收敛成ABBCDE, 这里面牵涉到AAA到底收敛为几个A, BBBB又收敛为几个B, 这也是他的核心

整体预测逻辑：

- 1, 将图像传入到LSTM中获得sequence, 和sequence的长度
- 2, 将sequence, sequence的长度输入到tf.nn.ctc\_beam\_search\_decoder函数预测出稀疏矩阵张量
- 3, 将第二步得到的稀疏矩阵张量反向转化为sequence, 并最终解码成A~Z的大写字母并输出

后续优化逻辑：

- 1, 可以在LSTM之前先采用CNN对图像特征进行一次提取
- 2, TF自带的ctc\_loss可以换成百度开源的Warp\_CTC
- 3, 针对少量原始图片为AAA结果最终识别为AA, 丢掉了A的情况, 是否可以把原先的标签['A', 'A', 'A']扩充为['A-left', 'A-middle', 'A-right', 'A-left', 'A-middle', 'A-right', 'A-left', 'A-middle', 'A-right']将每个字由原先的1个标签扩充为三个标签, 此处抛砖引玉, 可以自行尝试优化

优缺点：

- 1, LSTM+CTC考虑了一行文本从左到右的序列关系, 这一点上比CNN更强, 同时可以轻松实现不定长的OCR

2，也正是由于RNN网络考虑了时序间的关系，所以运算量相对于CNN网络大幅增加，收敛比较慢，有条件的同学还是上一块好点的GPU吧，能提升很多效率

运行命令：

自行生成验证码训练集（本文生成了10w张，修改self.im\_total\_num变量）：

`python LstmCtcOcr.py create_dataset`

对数据集进行训练：`python LstmCtcOcr.py train`

对新的图片进行测试：`python LstmCtcOcr.py test`

启动成http服务：`python LstmCtcOcr.py start`

利用flask框架将整个项目启动成web服务，使得项目支持http方式调用

启动服务后调用以下地址测试

`http://127.0.0.1:5050/captchaOcr?img_path=./dataset/test/0_PIY.png`

(`http://127.0.0.1:5050/captchaOcr?img_path=./dataset/test/0_PIY.png`)

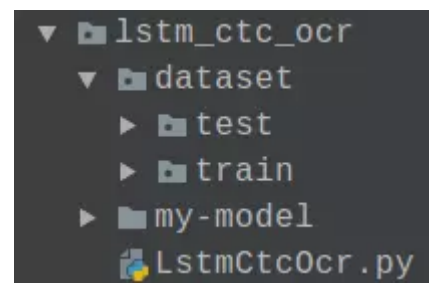
`http://127.0.0.1:5050/captchaOcr?img_path=./dataset/test/1_BCAVDPXT.png`

(`http://127.0.0.1:5050/captchaOcr?img_path=./dataset/test/1_BCAVDPXT.png`)

`http://127.0.0.1:5050/captchaOcr?img_path=./dataset/test/2_N.png`

(`http://127.0.0.1:5050/captchaOcr?img_path=./dataset/test/2_N.png`)

项目目录结构：



项目结构.png

训练200个epoch之后，可以看到model在val上的acc已经能达到84%了，后续大家可以自行修改学习率和增大epoch次数来提升精度（True表示预测正确，左边为预测值，右边为真实标签）：

```

epoch:199/200 batch:6120/6220 total_step:1243900 lr:0.0000004987
True ['P', 'P', 'Y', 'W'] <====> ['P', 'P', 'Y', 'W']
True ['F', 'P', 'D', 'H', 'I', 'D', 'L', 'V', 'M'] <====> ['F', 'P', 'D', 'H', 'I', 'D', 'L', 'V', 'M']
True ['W', 'F', 'W', 'H', 'F'] <====> ['W', 'F', 'W', 'H', 'F']
True ['X', 'E', 'B', 'S', 'F', 'R', 'I'] <====> ['X', 'E', 'B', 'S', 'F', 'R', 'I']
True ['M', 'U', 'A', 'H', 'Y', 'F'] <====> ['M', 'U', 'A', 'H', 'Y', 'F']
False ['E', 'O', 'M', 'F', 'J'] <====> ['F', 'E', 'O', 'M', 'F', 'J']
True ['R', 'D', 'J'] <====> ['R', 'D', 'J']
True ['P', 'C', 'Q', 'D', 'N', 'Q'] <====> ['P', 'C', 'Q', 'D', 'N', 'Q']
True ['E', 'A', 'X'] <====> ['E', 'A', 'X']
True ['A', 'S', 'E', 'P', 'V', 'J', 'E', 'L', 'N', 'Z'] <====> ['A', 'S', 'E', 'P', 'V', 'J', 'E', 'L', 'N', 'Z']
True ['Y', 'D', 'V'] <====> ['Y', 'D', 'V']
True ['Z', 'Y', 'P', 'E'] <====> ['Z', 'Y', 'P', 'E']
True ['X'] <====> ['X']

```

lstm-ctc-199-epoch.png

整体代码如下（LstmCtcOcr.py文件）：

```
# coding:utf-8
```

```
from captcha.image import ImageCaptcha
import numpy as np
import cv2
import tensorflow as tf
import random, os, sys
import operator
```

```
from flask import request
from flask import Flask
import json
app = Flask(__name__)
```

```
class LstmCtcOcr:
```

```
    def __init__(self):
        self.epoch_max = 200 # 最大迭代epoch次数
        self.batch_size = 16 # 训练时每个批次参与训练的图像数目，显存不足的可以调小
        self.lr = 5e-5 # 初始学习率
        self.save_epoch = 5 # 每相隔多少个epoch保存一次模型
        self.n_hidden = 256 # 隐藏神经元个数

        self.im_width = 256
        self.im_height = 64
        self.im_total_num = 100000 # 总共生成的验证码图片数量
        self.train_max_num = self.im_total_num # 训练时读取的最大图片数目
        self.val_num = 30 * self.batch_size # 不能大于self.train_max_num 做验
        self.words_max_num = 10 # 每张验证码图片上的最大字母个数
        self.words = 'ABCDEFGHIJKLMNOPQRSTUVWXYZ'
        self.n_classes = len(self.words) + 1 # 26个字母 + blank
        self.x = None
        self.y = None
```

```
    def captcha0cr(self, img_path):
```

```
        """
```

```
        验证码识别
```

```
        :param img_path:
```

```
        :return:
```

```
        """
```

```
        im = cv2.imread(img_path)
```

```
        im = cv2.resize(im, (self.im_width, self.im_height))
```

```
        im = np.array([im[:, :, 0]], dtype=np.float32)
```

```
        im -= 147
```

```
        pred = self.ssess.run(self.pred, feed_dict={self.x: im})
```

```
        sequence = self.sparseTensor2sequence(pred)
```

```
        return ''.join(sequence[0])
```

```
    def test(self, img_path):
```

```
        """
```

```
        测试接口
```

```
        :param img_path:
```

```
        :return:
```

```
        """
```

```
        self.batch_size = 1
```

```
        self.learning_rate = tf.placeholder(dtype=tf.float32) # 动态学习率
```

```
        self.weight = tf.Variable(tf.truncated_normal([self.n_hidden, self.n_
```

```
        self.bias = tf.Variable(tf.constant(0., shape=[self.n_classes]))
```

```
        self.x = tf.placeholder(tf.float32, [None, self.im_height, self.im_wi
```

```
        logits, seq_len = self.rnnNet(self.x, self.weight, self.bias)
```

```

        decoded, log_prob = tf.nn.ctc_beam_search_decoder(logits, seq_len,
self.pred = tf.cast(decoded[0], tf.int32)

saver = tf.train.Saver()
# tfconfig = tf.ConfigProto(allow_soft_placement=True)
# tfconfig.gpu_options.per_process_gpu_memory_fraction = 0.3 # 占用显
# self.ses = tf.Session(config=tfconfig)
self.ses = tf.Session()
self.ses.run(tf.global_variables_initializer()) # 全局tf变量初始化

# 加载w,b参数
saver.restore(self.ses, './my-model/LstmCtc0cr-200')
im = cv2.imread(img_path)
im = cv2.resize(im, (self.im_width, self.im_height))
im = np.array([im[:, :, 0]], dtype=np.float32)
im -= 147
pred = self.ses.run(self.pred, feed_dict={self.x: im})
sequence = self.sparseTensor2sequence(pred)
print(''.join(sequence[0]))

def train(self):
    """
    训练
    :return:
    """
    x_train_list, y_train_list, x_val_list, y_val_list = self.getTrainDat

    print('开始转换tensor队列')
    x_train_list_tensor = tf.convert_to_tensor(x_train_list, dtype=tf.str
    y_train_list_tensor = tf.convert_to_tensor(y_train_list, dtype=tf.int

    x_val_list_tensor = tf.convert_to_tensor(x_val_list, dtype=tf.string)
    y_val_list_tensor = tf.convert_to_tensor(y_val_list, dtype=tf.int32)

    x_train_queue = tf.train.slice_input_producer(tensor_list=[x_train_li
    y_train_queue = tf.train.slice_input_producer(tensor_list=[y_train_li

    x_val_queue = tf.train.slice_input_producer(tensor_list=[x_val_list_t
    y_val_queue = tf.train.slice_input_producer(tensor_list=[y_val_list_t

    train_im, train_label = self.dataset_opt(x_train_queue, y_train_queue
    train_batch = tf.train.batch(tensors=[train_im, train_label], batch_s

    val_im, val_label = self.dataset_opt(x_val_queue, y_val_queue)
    val_batch = tf.train.batch(tensors=[val_im, val_label], batch_size=se

    print('准备训练')
    self.learning_rate = tf.placeholder(dtype=tf.float32) # 动态学习率
    self.weight = tf.Variable(tf.truncated_normal([self.n_hidden, self.n_
    self.bias = tf.Variable(tf.constant(0., shape=[self.n_classes]))

    # self.global_step = tf.Variable(0, trainable=False) # 全局步骤计数

    # im_width看成LSTM的time_step , im_height看成是每个time_step输入tensor的s
    self.x = tf.placeholder(tf.float32, [None, self.im_height, self.im_wi
    # 定义ctc_loss需要的稀疏矩阵
    self.y = tf.sparse_placeholder(tf.int32)

    logits, seq_len = self.rnnNet(self.x, self.weight, self.bias)

    # loss

```

```

self.loss = tf.nn.ctc_loss(self.y, logits, seq_len)
# cost
self.cost = tf.reduce_mean(self.loss)
# optimizer
self.optimizer = tf.train.AdamOptimizer(learning_rate=self.learning_r

# 前面说的划分块之后找每块的类属概率分布, ctc_beam_search_decoder方法,是每次找
# 还有一种贪心策略是只找概率最大那个, 也就是K=1的情况ctc_ greedy_decoder
decoded, log_prob = tf.nn.ctc_beam_search_decoder(logits, seq_len, me
self.pred = tf.cast(decoded[0], tf.int32)
self.distance = tf.reduce_mean(tf.edit_distance(self.pred, self.y))

print('开始训练')
saver = tf.train.Saver() # 保存tf模型
with tf.Session() as self.sess:
    self.sess.run(tf.global_variables_initializer())
    coordinator = tf.train.Coordinator()
    threads = tf.train.start_queue_runners(sess=self.sess, coord=coord

    batch_max = len(x_train_list) // self.batch_size
    print('batch:', batch_max)
    total_step = 0
    for epoch_num in range(self.epoch_max):
        lr_tmp = self.lr * (1 - (epoch_num / self.epoch_max) ** 2) #
        print('lr:', lr_tmp)
        for batch_num in range(batch_max):
            # print(epoch_num, batch_num)
            x_train_tmp, y_train_tmp = self.sess.run(train_batch)
            y_train_tmp = self.sequence2sparseTensor(y_train_tmp) #
            self.sess.run(self.optimizer, feed_dict={self.x: x_train_

        if total_step % 100 == 0 or total_step == 0:
            print('epoch:%d/%d batch:%d/%d total_step:%d lr:%.10f
            # train部分
            train_loss, train_distance = self.sess.run([self.cost

            # val部分
            val_loss_list, val_distance_list, val_acc_list = [],
            for i in range(int(self.val_num / self.batch_size)):
                x_val_tmp, y_val_tmp_true = self.sess.run(val_bat
                y_val_tmp = self.sequence2sparseTensor(y_val_tmp_
                val_loss, val_distance, val_pred = self.sess.run(
                val_loss_list.append(val_loss)
                val_distance_list.append(val_distance)
                val_sequence = self.sparseTensor2sequence(val_pre
                ok = 0.
                for idx, val_seq in enumerate(val_sequence):
                    val_pred_tmp = [self.words.find(x) if self.wo
                    val_y_true_tmp = [x for x in y_val_tmp_true[i

                    is_eq = operator.eq(val_pred_tmp, val_y_true_

                    if idx == 0:
                        print(is_eq, [self.words[n] for n in val_

                    if is_eq:
                        ok += 1
                    val_acc_list.append(ok / len(val_sequence))
            val_acc_list = np.array(val_acc_list, dtype=np.float3

            print('train_loss:%.10f train_distance:%.10f' % (trai

```

```

        print('    val_loss:%.10f    val_distance:%.10f val_acc:%.10f' % (val_loss, val_distance, val_acc))
        print()
        print()

        total_step += 1

        # 保存模型
        if (epoch_num + 1) % self.save_epoch == 0:
            saver.save(self.sess, './my-model/LstmCtc0cr', global_step=total_step)

        coordinator.request_stop()
        coordinator.join(threads)

def rnnNet(self, inputs, weight, bias):
    """
    获取LSTM网络结构
    :param inputs:
    :param weight:
    :param bias:
    :return:
    """
    # 对于tf.nn.dynamic_rnn, 默认time_major=false, 此时inputs的shape=[batch_size, im_height, im_width]
    # (batch_size, im_height, im_width) ==> (batch_size, im_width, im_height)
    inputs = tf.transpose(inputs, [0, 2, 1])

    # 变长序列的最大值
    # seq_len = np.ones(self.batch_size) * self.im_width
    seq_len = np.ones(self.batch_size) * self.im_width

    cell = tf.nn.rnn_cell.LSTMCell(self.n_hidden, forget_bias=0.8, state_is_tuple=True)

    # 动态rnn实现输入变长
    outputs1, _ = tf.nn.dynamic_rnn(cell, inputs, seq_len, dtype=tf.float32)

    # (self.batch_size * self.im_width, self.n_hidden)
    outputs = tf.reshape(outputs1, [-1, self.n_hidden])

    logits = tf.matmul(outputs, weight) + bias # w * x + b
    logits = tf.reshape(logits, [self.batch_size, -1, self.n_classes])
    logits = tf.transpose(logits, (1, 0, 2)) # (im_width, batch_size, im_height)
    return logits, seq_len

def sequence2sparseTensor(self, sequences, dtype=np.int32):
    """
    序列 转化为 稀疏矩阵
    :param sequences:
    :param dtype:
    :return:
    """
    values, indices = [], []
    for n, seq in enumerate(sequences):
        indices.extend(zip([n] * len(seq), range(len(seq))))
        values.extend(seq)
    indices = np.asarray(indices, dtype=np.int64)
    values = np.asarray(values, dtype=dtype)
    shape = np.asarray([len(sequences), np.asarray(indices).max(0)[1] + 1])
    return indices, values, shape

def sparseTensor2sequence(self, sparse_tensor):

```



```

"""
稀疏矩阵 转化为 序列
:param sparse_tensor:
:return:
"""
decoded_indexes = list()
current_i = 0
current_seq = []
for offset, i_and_index in enumerate(sparse_tensor[0]):
    i = i_and_index[0]
    if i != current_i:
        decoded_indexes.append(current_seq)
        current_i = i
        current_seq = list()
    current_seq.append(offset)
decoded_indexes.append(current_seq)
result = []
for index in decoded_indexes:
    result.append(self.sequence2words(index, sparse_tensor))
return result

def sequence2words(self, indexes, spars_tensor):
"""
序列 转化为 文本
:param indexes:
:param spars_tensor:
:return:
"""
decoded = []
for m in indexes:
    str_tmp = self.words[spars_tensor[1][m]]
    decoded.append(str_tmp)
return decoded

def dataset_opt(self, x_train_queue, y_train_queue):
"""
处理图片和标签
:param queue:
:return:
"""
queue = x_train_queue[0]
contents = tf.read_file('./dataset/train/' + queue)
im = tf.image.decode_jpeg(contents)
tf.image.rgb_to_grayscale(im)
im = tf.image.resize_images(images=im, size=[self.im_height, self.im_width])
im = tf.reshape(im[:, :, 0], tf.stack([self.im_height, self.im_width]))
im -= 127 # 去均值化
return im, y_train_queue[0]

def getTrainDataset(self):
train_data_list = os.listdir('./dataset/train/')
print('共有%d张训练图片, 读取%d张: ' % (len(train_data_list), self.train_num))
random.shuffle(train_data_list) # 打乱顺序

y_val_list, y_train_list = [], []
x_val_list = train_data_list[:self.val_num]
for x_val in x_val_list:
    words_tmp = x_val.split('.')[0].split('_')[1]
    words_tmp = words_tmp + '?' * (self.words_max_num - len(words_tmp))

```

```

        y_val_list.append([self.words.find(x) if self.words.find(x) > -1

x_train_list = train_data_list[self.val_num:self.train_max_num]
for x_train in x_train_list:
    words_tmp = x_train.split('.')[0].split('_')[1]
    words_tmp = words_tmp + '?' * (self.words_max_num - len(words_tmp)
    y_train_list.append([self.words.find(x) if self.words.find(x) > -

return x_train_list, y_train_list, x_val_list, y_val_list

def createCaptchaDataset(self):
    """
    生成训练用图片数据集
    :return:
    """
    image = ImageCaptcha(width=self.im_width, height=self.im_height, font
    for i in range(self.im_total_num):
        words_tmp = ''
        for j in range(random.randint(1, self.words_max_num)):
            words_tmp = words_tmp + random.choice(self.words)
        print(words_tmp, type(words_tmp))
        im_path = './dataset/train/%d_%s.png' % (i, words_tmp)
        print(im_path)
        image.write(words_tmp, im_path)

if __name__ == '__main__':
    opt_type = sys.argv[1:][0]

    instance = LstmCtcOcr()

    if opt_type == 'create_dataset':
        instance.createCaptchaDataset()
    elif opt_type == 'train':
        instance.train()
    elif opt_type == 'test':
        instance.test('./dataset/test/0_P1Y.png')
    elif opt_type == 'start':
        # 将session持久化到内存中
        instance.test('./dataset/test/0_P1Y.png')

    # 启动web服务
    # http://127.0.0.1:5050/captcha0cr?img_path=./dataset/test/1_BCAVDPXT
    @app.route('/captcha0cr', methods=['GET'])
    def captcha0cr():
        img_path = request.args.to_dict().get('img_path')
        print(img_path)
        ret = instance.captcha0cr(img_path)
        print(ret)
        return json.dumps({'img_path': img_path, 'ocr_ret': ret})

    app.run(host='0.0.0.0', port=5050, debug=False)

```