

五、

5.1 具体实现过程

MNIST 手写识别的训练图片是28*28寸的图片。

- 第一层隐藏层

第一步是将图片矩阵打平处理为 $[image]_{1 \times 784}$

第二步是我们将第一层隐藏层将要设置的参数数量，这些数量会影响我们的训练精度，将设我们设置第一层的隐藏层权重和偏置参数分别为500个，则可以设置第一个隐藏层的维数为 $[\omega_1]_{784 \times 500}$ ，同样偏置的维数为 $[b_1]_{1 \times 500}$

- 第二层隐藏层

第三步是设置第二层隐藏层的权重和偏置维数，因为我们只需要三层神经网络结构实现，所以我们这一层应该做池化实现onehot，输出应该为1*10的独热编码。所以 ω_2 应该是500行和10列。即

$[\omega_2]_{500 \times 10}$ ， $[b_2]_{1 \times 10}$

- 整个流程

$\{[image]_{1 \times 784}\} \rightarrow \{[\omega_1]_{784 \times 500}, [b_1]_{1 \times 500}\} \rightarrow \{[\omega_2]_{500 \times 10}, [b_2]_{1 \times 10}\} \rightarrow \{[onehot]_{1 \times 10}\}$

5.2 前向传播

训练的过程是，首先输入一批数据，然后对每批数据进行上个流程的计算过程，比如我一批次输入200张图片进行训练。

$\{[image]_{200 \times 784}\} \rightarrow \{[\omega_1]_{784 \times 500}, [b_1]_{1 \times 500}\} = \{[data]_{200 \times 500}\} \rightarrow \{[\omega_2]_{500 \times 10}, [b_2]_{1 \times 10}\} = \{[onehot_{output}]_{200 \times 10}\}$

前向传播算法的代码如下部分代码所示

```
import tensorflow as tf
INPUT_NODE = 784
OUTPUT_NODE = 10
LAYER1_NODE = 500

def get_weight(shape, regularizer):
    w = tf.Variable(tf.truncated_normal(shape, stddev=0.1))
    if regularizer != None: tf.add_to_collection('losses', tf.contrib.layers.l2_regularizer(0.001)(w))
    return w

def get_bias(shape):
    b = tf.Variable(tf.zeros(shape))
    return b

def forward(x, regularizer):
    # [ ]_{784*500}
    w1 = get_weight([INPUT_NODE, LAYER1_NODE], regularizer)
    b1 = get_bias([LAYER1_NODE])
    # relu 非线性函数的，修正线性单元
    y1 = tf.nn.relu(tf.matmul(x, w1) + b1)

    # 这里w= get_weight([500, 784] 因为x=[None, 784]
    w2 = get_weight([LAYER1_NODE, OUTPUT_NODE], regularizer)
    b2 = get_bias([OUTPUT_NODE])
    y = tf.matmul(y1, w2) + b2
    return y
```

反向传播过程就是利用梯度下降算法进行最佳的各隐藏层的权值和偏置的优化过程。

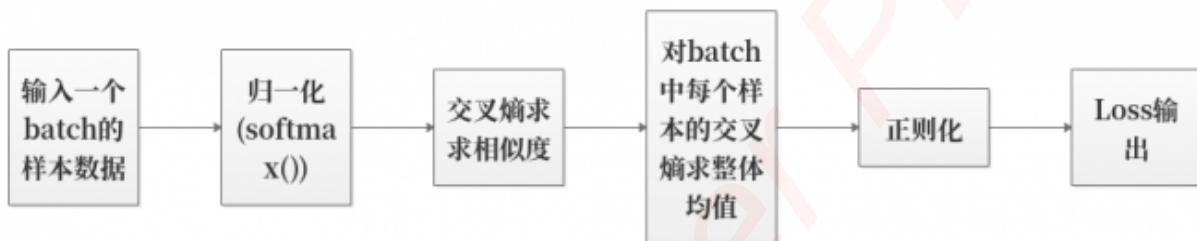
一般选择的梯度优化算法有：

- 批量梯度优化算法
- 随机梯度优化
- 自适应梯度优化
- 等等

```
train_step = tf.train.GradientDescentOptimizer(learning_rate).minimize(loss, global_step=global_step)
```

整个流程

前向传播输出结果--> 归一化(softmax()) --> 求取每回batch中样本的输出相似度（交叉熵求相似度 sparse_softmax_cross_entropy_with_logits） --> 得到每个batch中每个样本中的交叉熵均值 (tf.reduce_mean()) --> 正则化输出--> loss(loss = cem + tf.add_n(tf.get_collection('losses')))



相关函数解释：

- tf.add_n(p1, p2, p3...)

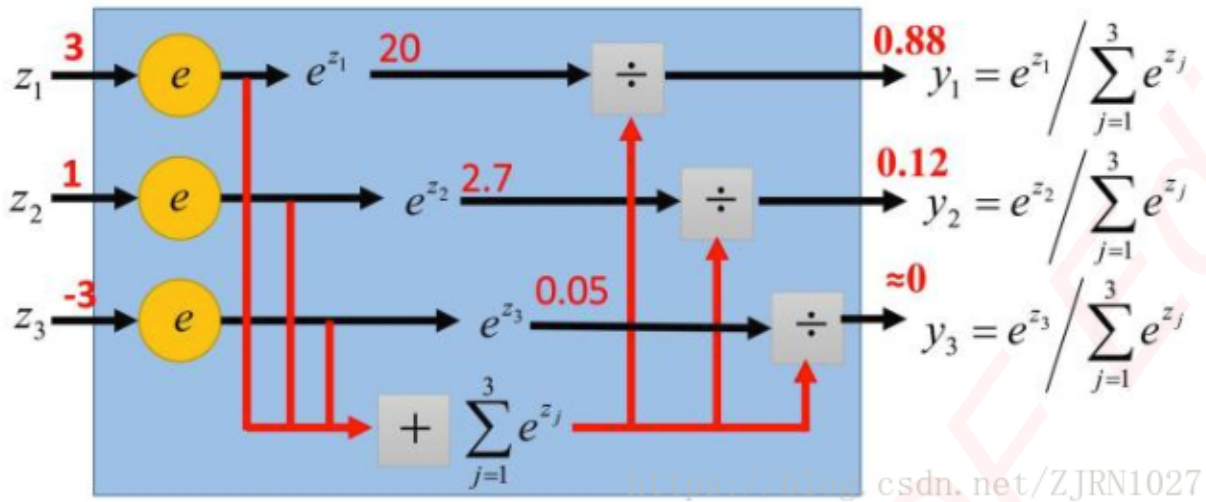
实现列表中的元素相加

```
input1 = tf.constant([1.0, 2.0, 3.0])
input2 = tf.Variable(tf.random_uniform([3]))
output = tf.add_n([input1, input2]) # = input1 + input2
```

5.3.1 Softmax(归一化前向传播输出结果)

$$loss = \frac{(y - y_j^2)}{n}$$

$$y_j = \frac{e^{z_j}}{\sum_i e^{z_i}}$$

Softmax Layer

如上图所示，softmax函数的作用是归一化，将所有的输出以概率和为1的方式，将输出结果用概率表示。

5.3.2 交叉熵

$$H_{y_{(y)}} = - \sum_i y_i \log(y_i)$$

$y_{(y)}$ y

$$y = [0.00589975 \ 0.8756006 \ 0.11849965]$$

$$\begin{aligned} H_{y_{(y)}} &= -0 \times \log(0.00589975) - 1 \times \log(0.8756006) - 0 \times \log(0.11849965) \\ &= 0.6355716 - 0 \times \log(0.00589975) - 1 \times \log(0.8756006) - 0 \times \log(0.11849965) \\ &= 0.6355716 - 0 \times \log(0.00589975) - 1 \times \log(0.8756006) - 0 \times \log(0.11849965) \\ &= 0.6355716 \end{aligned}$$

$y_{[200 \times 10]}$

axis=0时比较每一列的元素，将每一列最大元素所在的索引记录下来，最后输出每一列最大元素所在的索引数组。

```
test[0] = array([1, 2, 3])
test[1] = array([2, 3, 4])
test[2] = array([5, 4, 3])
test[3] = array([8, 7, 2])
# output : [3, 3, 1]
```

axis = 1:

axis=1的时候，将每一行最大元素所在的索引记录下来，最后返回每一行最大元素所在的索引数组。

```
test[0] = array([1, 2, 3]) #2
test[1] = array([2, 3, 4]) #2
test[2] = array([5, 4, 3]) #0
test[3] = array([8, 7, 2]) #0
```

5.3.3 batch个样本的交叉熵后整体均值计算

L^2

ω

$$\Omega(\theta) = \frac{1}{2} \|\omega\|_2^2$$

$$\begin{aligned}\tilde{J}(\omega; X, y) &= J(\omega; X, y) + \alpha \Omega(\theta) \\ &= J(\omega; X, y) + \frac{\alpha}{2} \omega^\top \omega\end{aligned}$$

与之对应的梯度为：

$$\nabla_{\omega} \tilde{J}(\omega; X, y) = \alpha \omega + \nabla_{\omega} J(\omega; X, y)$$

得到 L^2 参数正则化的参数 ω 参数更新的表达式为

$$\omega \leftarrow \omega - \epsilon \nabla_{\omega} \tilde{J}(\omega; X, y)$$

正则化后的代价函数 $J(\omega; X, y)$ 实现代码为

```
loss = cem + tf.add_n(tf.get_collection('losses'))
```

```
import tensorflow as tf
from tensorflow.examples.tutorials.mnist import input_data
# import mnist_forward
import os

BATCH_SIZE = 200
LEARNING_RATE_BASE = 0.1
LEARNING_RATE_DECAY = 0.99
REGULARIZER = 0.0001
STEPS = 50000
MOVING_AVERAGE_DECAY = 0.99
MODEL_SAVE_PATH = "./model/"
MODEL_NAME = "mnist_model"

def backward(mnist):
    x = tf.placeholder(tf.float32, [None, mnist_forward.INPUT_NODE])
    y_ = tf.placeholder(tf.float32, [None, mnist_forward.OUTPUT_NODE])
    y = mnist_forward.forward(x, REGULARIZER)
    global_step = tf.Variable(0, trainable=False)

    ce = tf.nn.sparse_softmax_cross_entropy_with_logits(logits=y, labels=tf.argmax(y_, 1))
    cem = tf.reduce_mean(ce)
    loss = cem + tf.add_n(tf.get_collection('losses'))

    learning_rate = tf.train.exponential_decay(
        LEARNING_RATE_BASE,
        global_step,
        mnist.train.num_examples / BATCH_SIZE,
        LEARNING_RATE_DECAY,
        staircase=True)

    train_step = tf.train.GradientDescentOptimizer(learning_rate).minimize(loss,
        global_step=global_step)

    ema = tf.train.ExponentialMovingAverage(MOVING_AVERAGE_DECAY, global_step)
    ema_op = ema.apply(tf.trainable_variables())
    with tf.control_dependencies([train_step, ema_op]):
        train_op = tf.no_op(name='train')

    saver = tf.train.Saver()

    with tf.Session() as sess:
        init_op = tf.global_variables_initializer()
        sess.run(init_op)

        ckpt = tf.train.get_checkpoint_state(MODEL_SAVE_PATH)
        if ckpt and ckpt.model_checkpoint_path:
            saver.restore(sess, ckpt.model_checkpoint_path)

        for i in range(STEPS):
            xs, ys = mnist.train.next_batch(BATCH_SIZE)
            _, loss_value, step = sess.run([train_op, loss, global_step], feed_dict={x: xs, y_: ys})
            if i % 1000 == 0:
                print("After %d training step(s), loss on training batch is %g."
```

```
saver.save(sess, os.path.join(MODEL_SAVE_PATH, MODEL_NAME), glob
```

```
def main():  
    mnist = input_data.read_data_sets("./data/", one_hot=True)  
    backward(mnist)  
  
if __name__ == '__main__':  
    main()
```

5.4.1 指数衰减学习率

$$dlr = lr * dr^{\frac{globals}{decays}}$$

```
learning_rate = tf.train.exponential_decay(learning_rate =  
LEARNING_RATE_BASE,  
    global_step = global_step,  
    decay_steps = mnist.train.num_examples / BATCH_SIZE,  
    decay_rate = LEARNING_RATE_DECAY,  
    staircase=True)
```

5.4.2 梯度下降并开始迭代训练

滑动平均值的参数更新, 比如对权重 ω_1 更新, 开始轮数为num_updates= 0 次时, ω_1 初值设置0, 第1轮更新为variable=1($\omega_1 = 1$); decay=0.99 ;此时有:

$$\omega_1 =$$

$$\min(\text{ma_decay}, \frac{1+0}{10+0}) * 0 + (1 - \min(1 - \min(\text{decay}, \frac{1+0}{10+0}) * 1)) = 0.1 * 0 + (1 - 0.9) * 1 =$$

训练过程的代码如下:

```
#coding:utf-8
import time
import tensorflow as tf
from tensorflow.examples.tutorials.mnist import input_data
import mnist_forward
import mnist_backward
TEST_INTERVAL_SECS = 5

def test(mnist):
    with tf.Graph().as_default() as g:
        x = tf.placeholder(tf.float32, [None, mnist_forward.INPUT_NODE])
        y_ = tf.placeholder(tf.float32, [None, mnist_forward.OUTPUT_NODE])
        y = mnist_forward.forward(x, None)

        ema = tf.train.ExponentialMovingAverage(mnist_backward.MOVING_AVERAGE_DECAY)
        ema_restore = ema.variables_to_restore()
        saver = tf.train.Saver(ema_restore)

        correct_prediction = tf.equal(tf.argmax(y, 1), tf.argmax(y_, 1))
        accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))

        while True:
            with tf.Session() as sess:
                ckpt = tf.train.get_checkpoint_state(mnist_backward.MODEL_SAVE_PATH)
                if ckpt and ckpt.model_checkpoint_path:
                    saver.restore(sess, ckpt.model_checkpoint_path)
                    global_step = ckpt.model_checkpoint_path.split('/')[-1].split('.')[0]
                    accuracy_score = sess.run(accuracy, feed_dict={x: mnist.test_images, y_: mnist.test_labels})
                    print("After %s training step(s), test accuracy = %g" % (global_step, accuracy_score))
                else:
                    print('No checkpoint file found')
                    return
            time.sleep(TEST_INTERVAL_SECS)

def main():
    mnist = input_data.read_data_sets("../MNIST_data/", one_hot=True)
    test(mnist)

if __name__ == '__main__':
    main()
```

5.5.1 保存模型


```

MODEL_SAVE_PATH = "./model/"
MODEL_NAME = "mnist_model"
saver = tf.train.Server()# 实例化Saver对象

with tf.Session() as sess:
    for i in range(STEPS):
        if i%轮数 == 0:
            saver.save(sess, os.path.join(MODEL_SAVE_PATH,
                                            MODEL_NAME), global_step =
global_step)

```

os.path.join() 是将括号中的参数都变成路径的字符形式。

比如，当前的global_step=100，则save后的文件名是"./model/mnist_model-100.meta、 ./model/mnist_model-100.index"等

5.5.2 恢复模型

5.6 验证阶段神经网络模型准确性的评估方法

```

correct_prediction = tf.equal(tf.argmax(y,1),tf.argmax(y_, 1))# 返回
的都是bool型变量
accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))#
将bool型转化为float型，然后求均值

```

对训练后的模型进行准确度评判

神经网络的测试验证阶段，只需要前向传播的模型，不需要对模型参数进行梯度下降的迭代优化，具体的过程入下

```

with tf.Graph().as_default() as g:
    x = tf.placeholder(tf.float32, [None,
mnist_forward.INPUT_NODE])
    y_ = tf.placeholder(tf.float32, [None,
mnist_forward.OUTPUT_NODE])
    y = mnist_forward.forward(x, None)

    ema =
tf.train.ExponentialMovingAverage(mnist_backward.MOVING_AVERAGE_DEC

```

AY)

```

ema_restore = ema.variables_to_restore()
saver = tf.train.Saver(ema_restore)

correct_prediction = tf.equal(tf.argmax(y, 1),
tf.argmax(y_, 1))
accuracy = tf.reduce_mean(tf.cast(correct_prediction,
tf.float32))

# 开始读取保存后的神经网络模型参数，恢复到我们复制Graph()中，然后开始准确性判断
with tf.Session() as sess:
    ckpt =
tf.train.get_checkpoint_state(mnist_backward.MODEL_SAVE_PATH)
    if ckpt and ckpt.model_checkpoint_path:
        saver.restore(sess, ckpt.model_checkpoint_path)
        global_step = ckpt.model_checkpoint_path.split('/')[-1].split('-')[-1]
        accuracy_score = sess.run(accuracy, feed_dict=
{x:mnist.test.images, y_:mnist.test.labels})
        print("After %s training step(s), test accuracy =
%g" % (global_step, accuracy_score))
    else:
        print('No checkpoint file found')
        return

```

▼ 5.7 训练后的模型使用

本部分的过程和测试部分的过程稍微不同的是，需要引入输入待识别的图像的图像预处理工作。

预处理工作，应该是如下步骤

- 图片的灰度化（0-255取值范围）
- 将灰度化处理后的图片像素取值进行缩放（0-1的取值范围`np.multiply(nm_arr, 1.0/255.0)`）
- 图像的reshape

复制神经网络计算图

```
tf.Graph().as_default()
```

然后重保存的神经网络训练参数文件中恢复模型参数

（和测试程序一样，这部分也不需要反向传播优化模型参数）

然后就是用前向传播计算的流程一样

```

x = tf.placeholder(tf.float32, [None,
mnist_forward.INPUT_NODE])
y = mnist_forward.forward(x, None)
preValue = tf.argmax(y, 1)

```

具体的整体代码如下部分所示

#coding:utf-8

```

import tensorflow as tf
import numpy as np
from PIL import Image
import mnist_backward
import mnist_forward

def restore_model(testPicArr):
    with tf.Graph().as_default() as tg:
        x = tf.placeholder(tf.float32, [None, mnist_forward.INPUT_NODE])
        y = mnist_forward.forward(x, None)
        preValue = tf.argmax(y, 1)

        variable_averages = tf.train.ExponentialMovingAverage(mnist_backward.MOV)
        variables_to_restore = variable_averages.variables_to_restore()
        saver = tf.train.Saver(variables_to_restore)

        with tf.Session() as sess:
            ckpt = tf.train.get_checkpoint_state(mnist_backward.MODEL_SAVE_PATH)
            if ckpt and ckpt.model_checkpoint_path:
                saver.restore(sess, ckpt.model_checkpoint_path)

                preValue = sess.run(preValue, feed_dict={x:testPicArr})
                return preValue
            else:
                print("No checkpoint file found")
                return -1

def pre_pic(picName):
    img = Image.open(picName)
    reIm = img.resize((28,28), Image.ANTIALIAS)
    im_arr = np.array(reIm.convert('L'))
    threshold = 50
    for i in range(28):
        for j in range(28):
            im_arr[i][j] = 255 - im_arr[i][j]
            if (im_arr[i][j] < threshold):
                im_arr[i][j] = 0
            else: im_arr[i][j] = 255

    nm_arr = im_arr.reshape([1, 784])
    nm_arr = nm_arr.astype(np.float32)
    img_ready = np.multiply(nm_arr, 1.0/255.0)

    return img_ready

def application():
    testNum = input("input the number of test pictures:")
    for i in range(testNum):
        testPic = raw_input("the path of test picture:")
        testPicArr = pre_pic(testPic)
        preValue = restore_model(testPicArr)
        print("The prediction number is:", preValue)

def main():
    application()

if __name__ == '__main__':
    main()

```

