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mnist.py
 Sep 26, 18 21:59
                                                                          Page 1/3
#!/bin/python3.6
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#CGML HW3
#Sept 26 2018
#Professor Curro
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
import logging
from tqdm import tqdm
import os
#os.environ['TF_CPP_MIN_LOG_LEVEL'] = '3'
# Define globals and import dataset
mnist = tf.keras.datasets.mnist
(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_{train}, x_{test} = x_{train}/255.0, x_{test}/255.0
                                                           # go from 0-1 instead of
0-255 8 bit greyscale
y_train = y_train.astype('int32')
                                                                            # fix du
e to tensorflow complaining
                                                                            # fix du
y_test = y_test.astype('int32')
e to tensorflow complaining
#import pdb; pdb.set_trace() # debug mode
''' def cutData(f, cut):
        1,w,h = f.shape
        s,e = cut,w-cut #assumes square
        fo = f[:,s:e,s:e]
        print (fo.shape)
        return fo
. . .
def cutData(f, cutx,cuty):
        l, w, h = f.shape
        sx,ex = cutx,w-cutx #assumes square
        sy, ey = cuty, w-cuty #assumes square
        fo = f[:,sv:ev,sx:ex]
        print (fo.shape)
        return fo
def genTrainAndVal(f,1): #split the features and labels of the training data 80:
20 train and validation
        valPercent=20
        lx, _, _= f.shape
        z = f.shape[0]
                                 # 60000 hopefully
        s = np.arange(z)
        np.random.shuffle(s)
                                          # features shuffled
        fs = f[s]
        ls = l[s]
                                          # labels shuffled
        lx = f.shape[0]
                                 # len of the features
                                 # num validation samp
        nv = int(lx *.2)
        print (fs[nv:].shape, ls[nv:].shape, fs[:nv].shape, ls[:nv].shape)
        return fs[nv:], ls[nv:], fs[:nv], ls[:nv]
#send this the output of only the first 2 returned val of genTrainAndVal
def getBatch(feat, lab):
        l,_,_= feat.shape
        choices = np.random.choice(1, size=BATCH_SIZE)
        return feat[choices],lab[choices]
def cnn(features, labels, mode):
        #layer 0
        h.w = 20.20
        L0 = tf.reshape(features["x"], [-1, h,w, 1]) #the -1 makes it guess how
big its supposed to be
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mnist.pv
 Sep 26, 18 21:59
                                                                         Page 2/3
                                                 #[numExaples, height, width, channe
1s]
        # hyperParams
        k = 4
        n = 5
        1 = 0.002
        fil = 1
        dr = .1
        # conv laver 1
        conv1 = tf.layers.conv2d(inputs=L0, filters=fil, kernel size=[k, k], paddi
ng="same", activation=tf.nn.elu)
        # pool layer 1
        pool1 = tf.layers.max_pooling2d(inputs=conv1, pool_size=[n, n], strides=
n)
        # flatten the layer for fully connected layer
        poolInt = int(w/n*h/n) * fil
        pFlat = tf.reshape(pool1, [-1, poolInt])
        # fully connected layer
        #dense = tf.layers.dense(inputs=pFlat, units=32, activation=tf.nn.elu)
        dropout = tf.layers.dropout(inputs=pFlat, rate=dr, training=mode == tf.e
stimator.ModeKeys.TRAIN)
        # logits layer layer that has the 10 outputs
        logits = tf.layers.dense(inputs=dropout, units=10)
        predictions = {
                "classes": tf.argmax(input=logits, axis=1),
                "probabilities": tf.nn.softmax(logits, name="softmax_tensor")}
        if mode == tf.estimator.ModeKeys.PREDICT:
                return tf.estimator.EstimatorSpec(mode=mode, predictions=predict
ions)
        num_params = np.sum([np.prod(v.get_shape().as_list()) for v in tf.traina
ble variables() 1)
        print("Num Params", num_params)
        for v in tf.trainable variables():
                print(v.get_shape())
        loss = tf.losses.sparse_softmax_cross_entropy(labels=labels, logits=logi
ts) + 1*tf.reduce_sum([tf.nn.12_loss(tV) for tV in tf.trainable_variables()])
        if mode == tf.estimator.ModeKeys.TRAIN:
                optimizer = tf.train.AdamOptimizer(learning_rate=0.001)
                train_op = optimizer.minimize(loss=loss,qlobal_step=tf.train.get
_global_step())
                return tf.estimator.EstimatorSpec(mode=mode, loss=loss, train_op
=train_op)
        eval_metric_ops = {
                                 "accuracy": tf.metrics.accuracy(labels=labels, pre
dictions=predictions["classes"])}
        return tf.estimator.EstimatorSpec(mode=mode, loss=loss, eval_metric_ops=
eval_metric_ops)
def main():
        mnist_classifier = tf.estimator.Estimator(model_fn=cnn) #, model_dir="./tm
p/modelCheckpoint"
        #tensors_to_log = {"probabilities": "softmax_tensor"}
        #logging_hook = tf.train.LoggingTensorHook(tensors=tensors_to_log, every
        tf.logging.set_verbosity(tf.logging.INFO)
        train_input_fn = tf.estimator.inputs.numpy_input_fn(x={"x": tx},y=ty,bat
ch_size=100,
                num_epochs=None, shuffle=True)
        mnist_classifier.train(input_fn=train_input_fn, steps=2000) #, hooks=[loggi
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mnist.py
 Sep 26, 18 21:59
                                                                          Page 3/3
ng hook])
        eval_input_fn = tf.estimator.inputs.numpy_input_fn(x={"x": vx},y=vy,
                num_epochs=1, shuffle=False)
        eval_results = mnist_classifier.evaluate(input_fn=eval_input_fn)
        print (eval_results)
        #I did not have this here when I was tuning hyper Parameters
        eval_input_fn = tf.estimator.inputs.numpy_input_fn(x={"x": x_test},y=y_t
est,
                num_epochs=1, shuffle=False)
        eval_results = mnist_classifier.evaluate(input_fn=eval_input_fn)
        print (eval_results)
cutx = 4
cuty = 4
tx, ty, vx, vy = genTrainAndVal(x_train, y_train)
tx=cutData(tx,cutx,cuty)
vx=cutData(vx,cutx,cuty)
x_test = cutData(x_test, cutx, cuty)
print (tx.shape, 'hi')
main()
#end of code
# just for shits
def plotVal():
        len_{,,,} = x_{train.shape}
        rn = np.random.randint(0,len-1)
        \#rn = 26563
        print(len,rn)
        test_val=x_train[rn]
        fig1= plt.figure(1)
        dr=4 #pixels top bottom left and right to drop
        xc,yc = np.linspace(0+dr,27-dr,28-2*dr), np.linspace(27-dr,0+dr,28-2*dr)
        xv, yv = np.meshgrid(xc, yc)
        print (type (y_train))
        #reduce dimentions of the test data
        f, 1 = 0 + dr, 28 - dr
        z = test_val[f:l, f:l]
        CS = plt.contourf(xv, yv, z, cmap='gray')
        w= plt.xlabel('w')
        h= plt.ylabel('h')
        h.set_rotation(0)
        plt.title("MNIST")
        plt.axis('equal')
        #plt.clabel(CS, fontsize=9, inline=1)
        plt.show()
```

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(4, 4, 1, 1)
(16, 10)
(10,)
INFO:tensorflow:Done calling model fn.
INFO:tensorflow:Starting evaluation at 2018-09-27-02:00:19
INFO:tensorflow:Graph was finalized.
2018-09-26 22:00:19.747814: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1484] Adding visible gpu devices: 0
2018-09-26 22:00:19.747856: I tensorflow/core/common runtime/qpu/qpu device.cc:965] Device interconnect StreamExecutor w
ith strength 1 edge matrix:
2018-09-26 22:00:19.747861: I tensorflow/core/common_runtime/gpu/gpu_device.cc:971]
2018-09-26 22:00:19.747865: I tensorflow/core/common runtime/qpu/qpu device.cc:984] 0:
2018-09-26 22:00:19.747952: I tensorflow/core/common runtime/gpu/gpu device.cc:1097] Created TensorFlow device (/job:loc
alhost/replica:0/task:0/device:GPU:0 with 2821 MB memory) -> physical GPU (device: 0, name: GeForce GTX 980, pci bus id:
0000:01:00.0, compute capability: 5.2)
INFO:tensorflow:Restoring parameters from /tmp/tmpf8hptawv/model.ckpt-2000
INFO:tensorflow:Running local init op.
INFO:tensorflow:Done running local init op.
INFO:tensorflow:Finished evaluation at 2018-09-27-02:00:19
INFO: tensorflow: Saving dict for global step 2000: <math>accuracy = 0.8451, global step = 2000, loss = 0.5841905
INFO:tensorflow:Saving 'checkpoint path' summary for global step 2000: /tmp/tmpf8hptawv/model.ckpt-2000
{'loss': 0.5841905, 'accuracy': 0.8451, 'qlobal step': 2000}
(venv) ostap@ostap-All-Series:~/Documents/Deep Learning/hw3curro$ |
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Num Params 187