#! /usr/bin/python Appendix E: TensorFlow code Sample import numpy as np import tensorflow as tf import time import dataset import utilities # Tensorflow convinience functions def weight_variable(shape, name): initial = tf.truncated normal(shape, stddev=0.1) return tf.Variable(initial, name=name) def bias variable(shape, name): initial = tf.constant(0.1, shape=shape) return tf.Variable(initial, name=name) def conv2d(x, W): return tf.nn.conv2d(x, W, strides=[1, 1, 1, 1], padding='SAME') def max_pool_3x3(x, name): return tf.nn.max pool(x, ksize=[1, 3, 3, 1], strides=[1, 2, 2, 1], padding='VALID', name=name) def max pool 2x2(x, name): return tf.nn.max_pool(x, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1], padding='SAME', name=name) def doDumbNet(trainDir, valDir, trainCsv, valCsv): f1 = open('log %d' % (time.time()), 'w+') f1.write('AAAAA\n') f1.write("Start %s\n" % time.time()) f1.flush() # Force CPU only mode because net overflows GPU mem with tf.device('/cpu:0'): sess = tf.Session() # Constants nClasses = 38imageSize = 256 batchSize = 10 learningRate = 1e-4 dropOutValue = 0.5 f1.write('nClasses: %d, imageSize: %d, batchSize: %d, learningRate: %e, dropOut: %f\n' % (nClasses, imageSize, batchSize, learningRate, dropOutValue)) f1.flush() x = tf.placeholder("float", shape=[None, imageSize, imageSize, 1], name="Input") y_ = tf.placeholder("float", shape=[None, nClasses], name="Output") # CONVOLUTIONAL NEURAL NET # The first two dimensions are the patch size, the next is the number of input channels, # and the last is the number of output channels. # We will also have a bias vector with a component for each output channel. d1 = 32d2 = 32d3 = 64d4 = 64d5 = 64fc = 300W_conv1 = weight_variable([5, 5, 1, d1], name="Weights_conv1") b_conv1 = bias_variable([d1], name="b_conv1") # We then convolve x_image with the weight tensor, # add the bias, apply the ReLU function, and finally max pool.

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h_conv1 = tf.nn.sigmoid(conv2d(x, W_conv1) + b_conv1)
h_pool1 = max_pool_2x2(h_conv1, name="pool1")
# SECOND CONV LAYER
# In order to build a deep network, we stack several layers of this type.
# The second layer will have 64 features for each 5x5 patch.
W_conv2 = weight_variable([5, 5, d1, d2], name="Weights_conv2")
b_conv2 = bias_variable([d2], name="biases_conv2")
h_conv2 = tf.nn.sigmoid(conv2d(h_pool1, W_conv2) + b_conv2)
h_pool2 = max_pool_2x2(h_conv2, name="pool2")
# THIRD CONV LAYER
W_conv3 = weight_variable([5, 5, d2, d3], name="Weights_conv3")
b conv3 = bias variable([d3], name="biases conv3")
h conv3 = tf.nn.sigmoid(conv2d(h pool2, W conv3) + b conv3)
h_pool3 = max_pool_2x2(h_conv3, name="pool3")
# FOURTH CONV LAYER
W_conv4 = weight_variable([3, 3, d3, d4], name="Weights_conv4")
b_conv4 = bias_variable([d4], name="biases_conv4")
h_conv4 = tf.nn.sigmoid(conv2d(h_pool3, W_conv4) + b conv4)
h pool4 = max pool 2x2(h conv4, name="pool4")
# FIFTH CONV LAYER
W_conv5 = weight_variable([3, 3, d4, d5], name="Weights_conv5")
b_conv5 = bias_variable([d5], name="biases_conv5")
h conv5 = tf.nn.sigmoid(conv2d(h pool4, W conv5) + b conv5)
h pool5 = max pool 2x2(h conv5, name="pool5")
# DENSELY CONNECTED LAYER
# Now that the image size has been reduced to 8x8,
# we add a fully-connected layer with 300 neurons to allow processing on the entire image.
# We reshape the tensor from the pooling layer into a batch of vectors, multiply by a weight
# matrix, add a bias, and apply a ReLU.
W_fc1 = weight_variable([8 * 8 * d5, fc], name="Weights_fc1")
b_fc1 = bias_variable([fc], name="biases_fc1")
h_conv5_flat = tf.reshape(h_pool5, [-1, 8 * 8 * d5])
h_fc1 = tf.nn.sigmoid(tf.matmul(h_conv5_flat, W_fc1) + b_fc1)
# DROPOUT
keep prob = tf.placeholder("float")
h fc1 drop = tf.nn.dropout(h fc1, keep prob)
# READOUT LAYER
W_fc2 = weight_variable([fc, nClasses], name="Weights_fc2")
b_fc2 = bias_variable([nClasses], name="biases_fc2")
y_conv = tf.nn.softmax(tf.matmul(h_fc1_drop, W_fc2) + b_fc2)
# Load the dataset
datasets = dataset.read data sets(trainDir, valDir, trainCsv, valCsv)
# Train and eval the model
cross_entropy = -tf.reduce_sum(y_ * tf.log(tf.clip_by_value(y_conv, 1e-10, 1.0)))
tf.scalar_summary('cross entropy', cross_entropy)
# train_step = tf.train.GradientDescentOptimizer(0.00001).minimize(cross_entropy)
train_step = tf.train.AdamOptimizer(learningRate).minimize(cross_entropy)
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correct prediction = tf.equal(tf.argmax(y_conv, 1), tf.argmax(y_, 1))
      accuracy = tf.reduce mean(tf.cast(correct prediction, "float"))
      sess.run(tf.initialize_all_variables())
      # saver = tf.train.Saver()
      # saver.restore(sess, 'my-model-batch1-10000')
      for i in xrange(5000):
         step_start = time.time()
         batch = datasets.train.get_sequential_batch(batchSize)
         train_step.run(feed_dict={x: batch[0], y_: batch[1], keep_prob: 1}, session=sess)
         if i % 25 == 0:
            # evaluate accuracy on random 100 samples from train set
            batch = datasets.train.get_random_batch(100)
            f1.write("step %d finished, time = %s\n" % (i, time.time() - step start))
            acc, cross_entropyD, yD = sess.run([accuracy, cross_entropy, y_conv],
                                           feed_dict={x: batch[0], y_: batch[1], keep_prob: 1})
            f1.write("Cross entropy = " + str(cross_entropyD) + "\n")
            f1.write("Accuracy = " + str(acc) + "\n")
            f1.write("Y = " + str(np.argmax(yD, axis=1)) + "\n")
            f1.write("\n--- %s seconds ---\n\n" % (time.time() - start time))
           f1.flush()
         f1.write("\nstep %d finished, %d seconds \n" % (i, time.time() - step_start))
        f1.flush()
   saver.save(sess, 'my-model-%d' % (time.time()), global_step=10000) # Evaluate the prediction
   test = datasets.validation.getAll()
   acc, y_convD, correct_predictionD = sess.run([accuracy, y_conv, correct_prediction],
                                                feed dict=\{x: test[0], y: test[1], keep prob: 1.0\})
  f1.write("Accuracy = " + str(acc) + "\n")
   f1.write("Correct prediction %d\n" % (sum(correct_predictionD)))
   f1.write("y %s\n" % str(test[1]))
   f1.write("y from net %s\n" % str(np.argmax(y_convD, axis=1)))
   f1.write("\n--- %s seconds ---\n\n" % (time.time() - start time))
   f1.flush()
   f1.close()
start time = time.time()
doDumbNet('train/train', 'train/validation', 'whales.csv', 'whales.csv')
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