< Deep Learning - PART1 TF2 Basics >

Ch 1. Workshop - TensorFlow 2.0 Installation & Testing

2021/10/01

<< Installation of TF 2.0 with Anaconda3 >>

- First, install Anaconda 3 for Windows/macOS/Linux from https://www.anaconda.com/distribution/ (https://www.anaconda.com/distribution/)
- Next, run TensorFlow 2 (for CPU) Setup on Anaconda Prompt :

```
conda install tensorflow
```

[Reference]:

- TensorFlow.org, "Install TensorFlow 2" https://www.tensorflow.org/install)
- 海萨, "Anaconda 安装tensorflow 2.0 报错解决办法"
 https://zhuanlan.zhihu.com/p/62031082 (https://zhuanlan.zhihu.com/p/62031082)
- TensorFlow.org, "Get Started with TensorFlow" https://www.tensorflow.org/tutorials/#get-started-with-tensorflow)

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- <u>1. Testing TF 2.0</u>
- 2. How to run TensorFlow 1.x code on TF 2.0

1. Testing TF 2.0

```
In [1]:

1 import tensorflow as tf
2 print(tf.__version__)
```

In [2]: ▶

```
2 # The following code is adopted from
3 # Tutorial document of TensorFlow.org
4 # for testing TensorFlow 2.0 setup:
6 # "Get Started with TensorFlow"
7 # https://www.tensorflow.org/tutorials/#get-started-with-tensorflow
8 # -----
9
10 mnist = tf.keras.datasets.mnist
11
12 (x_train, y_train),(x_test, y_test) = mnist.load_data()
13 x_train, x_test = x_train / 255.0, x_test / 255.0
14
15 model = tf.keras.models.Sequential([
     tf.keras.layers.Flatten(input_shape=(28, 28)),
16
     tf.keras.layers.Dense(512, activation=tf.nn.relu),
17
18
    tf.keras.layers.Dropout(0.2),
19
    tf.keras.layers.Dense(10, activation=tf.nn.softmax)
20 ])
21 model.compile(optimizer='adam',
22
                loss='sparse_categorical_crossentropy',
23
                metrics=['accuracy'])
24
25 model.fit(x_train, y_train, epochs=5)
26 model.evaluate(x_test, y_test, verbose=0) # verbose: Verbosity mode. 0=sile
```

```
Epoch 1/5
s: 0.3670 - accuracy: 0.8910
Epoch 2/5
s: 0.0985 - accuracy: 0.9696
Epoch 3/5
1875/1875 [============ ] - 15s 8ms/step - los
s: 0.0697 - accuracy: 0.9785
Epoch 4/5
s: 0.0506 - accuracy: 0.9837
Epoch 5/5
s: 0.0406 - accuracy: 0.9865
Out[2]:
```

[0.07084144651889801, 0.9793000221252441]

2. How to run TensorFlow 1.x code on TF 2.0

 It is still possible to run 1.X code, unmodified (except for contrib), in TensorFlow 2.0:

```
import tensorflow.compat.v1 as tf

tf.disable_v2_behavior()

[NOTE]:
```

 More detailed information regarding "Migrate your TensorFlow 1 code to TensorFlow 2 "can be found here: https://www.tensorflow.org/guide/migrate)

```
In [7]:

1  import tensorflow.compat.v1 as tf
2  tf.disable_v2_behavior()
3  4  print(tf.__version__)
```

WARNING:tensorflow:From C:\Users\USER\Anaconda3\lib\site-package s\tensorflow_core\python\compat\v2_compat.py:65: disable_resourc e_variables (from tensorflow.python.ops.variable_scope) is depre cated and will be removed in a future version.

Instructions for updating:
non-resource variables are not supported in the long term 2.0.0

The following code is adopted for testing TensorFlow 2.0 setup from the reference below:

- Tom Hope, Yehezkel S. Resheff, and Itay Lieder, "Learning TensorFlow: A Guide to Building Deep Learning Systems," Chapter 2 & 4, O'Reilly, 2017. https://goo.gl/iEmehh) (https://goo.gl/iEmehh)
- Download the code from GitHub : https://github.com/gigwegbe/Learning-TensorFlow
 (https://github.com/gigwegbe/Learning-TensorFlow)

Loading the MNIST dataset (from TensorFlow 2.0)

```
In [8]:

1  mnist = tf.keras.datasets.mnist
2  
3  (x_train, y_train),(x_test, y_test) = mnist.load_data()
4  x_train, x_test = x_train / 255.0, x_test / 255.0
```

```
In [9]:
                                                                             M
 1 import numpy as np
 3 x_train = np.array([x_train[i].flatten() for i in range(len(x_train))])
 4 x_train.shape
Out[9]:
(60000, 784)
                                                                             H
In [10]:
 1 x_test = np.array([x_test[i].flatten() for i in range(len(x_test))])
 2 x_test.shape
Out[10]:
(10000, 784)
In [11]:
                                                                             M
 1 y_train[0], y_test[0]
Out[11]:
(5, 7)
                                                                             M
In [12]:
 1 def one_hot(vec, vals=10):
 2
        n = len(vec)
        out = np.zeros((n, vals))
 3
 4
        out[range(n), vec] = 1
        return out
In [13]:
                                                                             M
 1 y_train = one_hot(y_train)
 2 y_train[0]
Out[13]:
array([0., 0., 0., 0., 1., 0., 0., 0., 0.])
```

```
H
In [14]:
 1 y_test = one_hot(y_test)
 2 y_test[0]
Out[14]:
array([0., 0., 0., 0., 0., 0., 1., 0., 0.])
Building a Computation Graph on TF 1.x
In [15]:
                                                                            M
 1 # Each Input Image, X, with 28*28 (= 784) pixels
 2 X = tf.placeholder(tf.float32, [None, 784])
 4 # y_true : the training labeled dataset
 5 y_true = tf.placeholder(tf.float32,[None, 10])
In [16]:
                                                                            M
 1 # Initializing Weights & Biases for Nodes in All Hidden Layers
 2 def weight_variable(shape):
        initial = tf.truncated_normal(shape, stddev=0.1)
 3
        return tf.Variable(initial)
 4
 6 def bias_variable(shape):
        initial = tf.constant(0.1, shape=shape)
 7
        return tf.Variable(initial)
 8
                                                                            M
In [17]:
 1 # Building a Fully-Connected Deep Network
   def full_layer(inputs, size):
 3
        in_size = int(inputs.get_shape()[1])
 4
        W = weight_variable([in_size, size])
 5
        b = bias_variable([size])
        return tf.add(tf.matmul(inputs, W), b)
```

```
H
In [18]:
   keep prob = tf.placeholder(tf.float32)
 3 # < Hidden Layer 1 >
 4 layer_1_drop = tf.nn.dropout(X, keep_prob=keep_prob)
        Activation Function: ReLU
 6 | layer_1_Outputs = tf.nn.relu(full_layer(layer_1_drop, 256))
 8 # < Hidden Layer 2 >
 9 layer 2 drop = tf.nn.dropout(layer 1 Outputs, keep prob=keep prob)
        Activation Function: ReLU
10 #
11 layer_2_Outputs = tf.nn.relu(full_layer(layer_2_drop, 128))
12
13 # < Output Layer >
14 output_drop = tf.nn.dropout(layer_2_Outputs, keep_prob=keep_prob)
15 # Without Activation Function
16 y_pred = full_layer(output_drop, 10)
WARNING:tensorflow:From <ipython-input-18-931f684597d4>:4: calli
ng dropout (from tensorflow.python.ops.nn_ops) with keep_prob is
deprecated and will be removed in a future version.
Instructions for updating:
Please use `rate` instead of `keep_prob`. Rate should be set to
`rate = 1 - keep_prob`.
In [19]:
                                                                             M
 1 cross_entropy = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits(logits)
 2 gd_step = tf.train.GradientDescentOptimizer(0.5).minimize(cross_entropy)
 4 correct_mask = tf.equal(tf.argmax(y_pred, 1), tf.argmax(y_true, 1))
 5 accuracy = tf.reduce mean(tf.cast(correct mask, tf.float32))
WARNING:tensorflow:From <ipython-input-19-315e39f82a5d>:1: softm
ax_cross_entropy_with_logits (from tensorflow.python.ops.nn_ops)
is deprecated and will be removed in a future version.
Instructions for updating:
Future major versions of TensorFlow will allow gradients to flow
```

See `tf.nn.softmax_cross_entropy_with_logits_v2`.

into the labels input on backprop by default.

Launching the Computation Graph on TF 1.x

In [20]: ▶

```
def next_batch(i, images, labels, batch_size):
    i_start = (i * batch_size) % len(images)
    x, y = images[i_start : i_start+batch_size], labels[i_start : i_start+batch_size], return x, y
```

In [21]:

```
NUM STEPS = 8000
   MINIBATCH SIZE = 100
 3
  Display_Step = 1000
4
 5
   with tf.Session() as sess:
 6
       sess.run(tf.global_variables_initializer())
7
8
       for i in range(NUM_STEPS):
            batch_xs, batch_ys = next_batch(i, x_train, y_train, MINIBATCH_SIZE
9
            sess.run(gd_step, feed_dict ={X: batch_xs,
10
11
                                           y_true: batch_ys,
12
                                           keep_prob: 0.5})
13
            if (i+1) % Display_Step == 0:
14
                # Calculate batch loss and accuracy
15
                loss_temp, accu_temp = sess.run([cross_entropy, accuracy],
16
17
                                                 feed_dict={X: batch_xs,
18
                                                             y_true: batch_ys,
19
                                                             keep_prob: 1.0})
                print("Step " + str(i+1).rjust(4) + \
20
21
                       " : Loss = " + "{:.4f}".format(loss_temp) + \
                      ", Accuracy = " + "{:.3f}".format(accu temp))
22
23
24
       print("\n Computing the test accuracy ... ", end = " ")
25
26
27
       ## Split the test procedure into 10 blocks of 1,000 images each.
28
        ## Doing this is important mostly for much larger datasets.
29
        ##
30
       ## mnist.test.images.shape : (10000, 784)
       X_{\text{test}} = x_{\text{test.reshape}}(10, 1000, 784)
31
32
       ## mnist.test.labels.shape : (10000, 10)
33
       Y_test = y_test.reshape(10, 1000, 10)
34
35
       test_loss = np.mean([sess.run(cross_entropy,
36
                                       feed_dict={X: X_test[i],
37
                                                  y_true: Y_test[i],
                                                  keep prob: 1.0})
38
39
                                       for i in range(10)])
40
       test_accu = np.mean([sess.run(accuracy,
41
                                       feed_dict={X: X_test[i],
42
                                                  y_true: Y_test[i],
43
                                                  keep_prob: 1.0})
44
                                       for i in range(10)])
45
       print("\n [ Test Accuracy ] : {}".format(test_accu) +
          "\n [ Test Loss Score ] : {}".format(test_loss))
46
```

```
Step 1000 : Loss = 0.2206, Accuracy = 0.960

Step 2000 : Loss = 0.1248, Accuracy = 0.970

Step 3000 : Loss = 0.1113, Accuracy = 0.990

Step 4000 : Loss = 0.0721, Accuracy = 0.980

Step 5000 : Loss = 0.0949, Accuracy = 0.960

Step 6000 : Loss = 0.1253, Accuracy = 0.990

Step 7000 : Loss = 0.0690, Accuracy = 0.990

Step 8000 : Loss = 0.0743, Accuracy = 0.980
```

Computing the test accuracy ...
[Test Accuracy] : 0.9604999423027039
[Test Loss Score] : 0.12836746871471405