Save and Restore TensorFlow Models

Training a model can take hours. But once you close your TensorFlow session, you lose all the trained weights and biases. If you were to reuse the model in the future, you would have to train it all over again!

Fortunately, TensorFlow gives you the ability to save your progress using a class called **tf.train.Saver**. This class provides the functionality to save any **tf.Variable** to your file system.

Saving Variables

Let's start with a simple example of saving weights and bias Tensors. For the first example you'll just save two variables. Later examples will save all the weights in a practical model.

```
import tensorflow as tf
# The file path to save the data
save_file = './model.ckpt'
# Two Tensor Variables: weights and bias
weights = tf.Variable(tf.truncated_normal([2, 3]))
bias = tf.Variable(tf.truncated_normal([3]))
# Class used to save and/or restore Tensor Variables
saver = tf.train.Saver()
with tf.Session() as sess:
    # Initialize all the Variables
    sess.run(tf.global_variables_initializer())
    # Show the values of weights and bias
    print('Weights:')
    print(sess.run(weights))
    print('Bias:')
    print(sess.run(bias))
    # Save the model
    saver.save(sess, save_file)
```

```
[[-0.97990924 1.03016174 0.74119264]
```

```
[-0.82581609 -0.07361362 -0.86653847]]
```

Bias:

```
[ 1.62978125 -0.37812829 0.64723819]
```

```
The Tensors weights and bias are set to random values using the tf.truncated_normal() function. The values are then saved to the save_file location, "model.ckpt", using the tf.train.Saver.save() function. (The ".ckpt" extension stands for "checkpoint".)
```

If you're using TensorFlow 0.11.0RC1 or newer, a file called "model.ckpt.meta" will also be created. This file contains the TensorFlow graph.

Loading Variables

Now that the Tensor Variables are saved, let's load them back into a new model.

```
# Remove the previous weights and bias
tf.reset_default_graph()

# Two Variables: weights and bias
weights = tf.Variable(tf.truncated_normal([2, 3]))
bias = tf.Variable(tf.truncated_normal([3]))

# Class used to save and/or restore Tensor Variables
saver = tf.train.Saver()

with tf.Session() as sess:
    # Load the weights and bias
    saver.restore(sess, save_file)

# Show the values of weights and bias
print('Weight:')
print(sess.run(weights))
print('Bias:')
print('Bias:')
print(sess.run(bias))
```

[[-0.97990924 1.03016174 0.74119264]

[-0.82581609 -0.07361362 -0.86653847]]

Bias:

[1.62978125 -0.37812829 0.64723819]

You'll notice you still need to create the weights and bias Tensors in Python. The tf.train.Saver.restore() function loads the saved data into weights and bias.

Since tf.train.Saver.restore() sets all the TensorFlow Variables, you don't need to call tf.global_variables_initializer().

Save a Trained Model

Let's see how to train a model and save its weights.

First start with a model:

```
# Remove previous Tensors and Operations
tf.reset_default_graph()
from tensorflow.examples.tutorials.mnist import input data
import numpy as np
learning_rate = 0.001
n_input = 784 # MNIST data input (img shape: 28*28)
n_classes = 10 # MNIST total classes (0-9 digits)
# Import MNIST data
mnist = input_data.read_data_sets('.', one_hot=True)
# Features and Labels
features = tf.placeholder(tf.float32, [None, n input])
labels = tf.placeholder(tf.float32, [None, n_classes])
# Weights & bias
weights = tf.Variable(tf.random_normal([n_input, n_classes]))
bias = tf.Variable(tf.random_normal([n_classes]))
# Logits - xW + b
logits = tf.add(tf.matmul(features, weights), bias)
# Define loss and optimizer
cost = tf.reduce mean(\
    tf.nn.softmax_cross_entropy_with_logits(logits=logits, labels=labels))
optimizer = tf.train.GradientDescentOptimizer(learning_rate=learning_rate)\
    .minimize(cost)
# Calculate accuracy
correct_prediction = tf.equal(tf.argmax(logits, 1), tf.argmax(labels, 1))
accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
```

Let's train that model, then save the weights:

```
import math
save_file = './train_model.ckpt'
batch size = 128
n_{epochs} = 100
saver = tf.train.Saver()
# Launch the graph
with tf.Session() as sess:
    sess.run(tf.global_variables_initializer())
    # Training cycle
    for epoch in range(n_epochs):
        total batch = math.ceil(mnist.train.num examples / batch size)
        # Loop over all batches
        for i in range(total_batch):
            batch_features, batch_labels = mnist.train.next_batch(batch_size)
            sess.run(
                optimizer,
                feed_dict={features: batch_features, labels: batch_labels})
        # Print status for every 10 epochs
        if epoch % 10 == 0:
            valid_accuracy = sess.run(
                accuracy,
                feed_dict={
                    features: mnist.validation.images,
                    labels: mnist.validation.labels})
            print('Epoch {:<3} - Validation Accuracy: {}'.format(</pre>
                epoch,
                valid_accuracy))
    # Save the model
    saver.save(sess, save_file)
    print('Trained Model Saved.')
```

Epoch 0 - Validation Accuracy: 0.06859999895095825

Epoch 10 - Validation Accuracy: 0.20239999890327454

Epoch 20 - Validation Accuracy: 0.36980000138282776

```
Epoch 40 - Validation Accuracy: 0.5601999759674072
```

```
Epoch 50 - Validation Accuracy: 0.6097999811172485
```

```
Epoch 60 - Validation Accuracy: 0.6425999999046326
```

```
Epoch 70 - Validation Accuracy: 0.6733999848365784
```

```
Epoch 80 - Validation Accuracy: 0.6916000247001648
```

```
Epoch 90 - Validation Accuracy: 0.7113999724388123
```

Trained Model Saved.

Load a Trained Model

Let's load the weights and bias from memory, then check the test accuracy.

```
saver = tf.train.Saver()

# Launch the graph
with tf.Session() as sess:
    saver.restore(sess, save_file)

test_accuracy = sess.run(
    accuracy,
    feed_dict={features: mnist.test.images, labels: mnist.test.labels})

print('Test Accuracy: {}'.format(test_accuracy))
```

Test Accuracy: 0.7229999899864197

That's it! You now know how to save and load a trained model in TensorFlow. Let's look at loading weights and biases into modified models in the next section.