

# Epoch-Example

October 10, 2018

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In [5]: from tensorflow.examples.tutorials.mnist import input_data
import tensorflow as tf
import numpy as np
from batchhelper import batches # Helper function created in Mini-batching section

def print_epoch_stats(epoch_i, sess, last_features, last_labels):
    """
    Print cost and validation accuracy of an epoch
    """
    current_cost = sess.run(
        cost,
        feed_dict={features: last_features, labels: last_labels})
    valid_accuracy = sess.run(
        accuracy,
        feed_dict={features: valid_features, labels: valid_labels})
    print('Epoch: {:<4} - Cost: {:<8.3} Valid Accuracy: {:<5.3}'.format(
        epoch_i,
        current_cost,
        valid_accuracy))

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('MNIST data', one_hot=True)

# The features are already scaled and the data is shuffled
train_features = mnist.train.images
valid_features = mnist.validation.images
test_features = mnist.test.images

train_labels = mnist.train.labels.astype(np.float32)
valid_labels = mnist.validation.labels.astype(np.float32)
test_labels = mnist.test.labels.astype(np.float32)

# Features and Labels
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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
learning_rate = tf.placeholder(tf.float32)
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))

init = tf.global_variables_initializer()

batch_size = 128
epochs = 80
learn_rate = 0.1

train_batches = batches(batch_size, train_features, train_labels)

with tf.Session() as sess:
    sess.run(init)

    # Training cycle
    for epoch_i in range(epochs):

        # Loop over all batches
        for batch_features, batch_labels in train_batches:
            train_feed_dict = {
                features: batch_features,
                labels: batch_labels,
                learning_rate: learn_rate}
            sess.run(optimizer, feed_dict=train_feed_dict)

        # Print cost and validation accuracy of an epoch
        print_epoch_stats(epoch_i, sess, batch_features, batch_labels)

    # Calculate accuracy for test dataset
    test_accuracy = sess.run(
        accuracy,
        feed_dict={features: test_features, labels: test_labels})

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

```
Extracting MNIST data/train-images-idx3-ubyte.gz
Extracting MNIST data/train-labels-idx1-ubyte.gz
Extracting MNIST data/t10k-images-idx3-ubyte.gz
Extracting MNIST data/t10k-labels-idx1-ubyte.gz
Epoch: 0    - Cost: 2.06      Valid Accuracy: 0.71
Epoch: 1    - Cost: 1.49      Valid Accuracy: 0.782
Epoch: 2    - Cost: 1.28      Valid Accuracy: 0.815
Epoch: 3    - Cost: 1.14      Valid Accuracy: 0.836
Epoch: 4    - Cost: 1.03      Valid Accuracy: 0.849
Epoch: 5    - Cost: 0.956     Valid Accuracy: 0.857
Epoch: 6    - Cost: 0.895     Valid Accuracy: 0.863
Epoch: 7    - Cost: 0.845     Valid Accuracy: 0.868
Epoch: 8    - Cost: 0.804     Valid Accuracy: 0.871
Epoch: 9    - Cost: 0.769     Valid Accuracy: 0.873
Epoch: 10   - Cost: 0.739     Valid Accuracy: 0.876
Epoch: 11   - Cost: 0.713     Valid Accuracy: 0.879
Epoch: 12   - Cost: 0.689     Valid Accuracy: 0.88
Epoch: 13   - Cost: 0.668     Valid Accuracy: 0.881
Epoch: 14   - Cost: 0.649     Valid Accuracy: 0.883
Epoch: 15   - Cost: 0.631     Valid Accuracy: 0.885
Epoch: 16   - Cost: 0.614     Valid Accuracy: 0.885
Epoch: 17   - Cost: 0.599     Valid Accuracy: 0.886
Epoch: 18   - Cost: 0.585     Valid Accuracy: 0.889
Epoch: 19   - Cost: 0.571     Valid Accuracy: 0.89
Epoch: 20   - Cost: 0.559     Valid Accuracy: 0.891
Epoch: 21   - Cost: 0.547     Valid Accuracy: 0.893
Epoch: 22   - Cost: 0.536     Valid Accuracy: 0.893
Epoch: 23   - Cost: 0.525     Valid Accuracy: 0.892
Epoch: 24   - Cost: 0.515     Valid Accuracy: 0.892
Epoch: 25   - Cost: 0.505     Valid Accuracy: 0.892
Epoch: 26   - Cost: 0.496     Valid Accuracy: 0.893
Epoch: 27   - Cost: 0.488     Valid Accuracy: 0.893
Epoch: 28   - Cost: 0.48      Valid Accuracy: 0.893
Epoch: 29   - Cost: 0.472     Valid Accuracy: 0.894
Epoch: 30   - Cost: 0.464     Valid Accuracy: 0.895
Epoch: 31   - Cost: 0.457     Valid Accuracy: 0.895
Epoch: 32   - Cost: 0.45      Valid Accuracy: 0.895
Epoch: 33   - Cost: 0.444     Valid Accuracy: 0.896
Epoch: 34   - Cost: 0.438     Valid Accuracy: 0.896
Epoch: 35   - Cost: 0.432     Valid Accuracy: 0.896
Epoch: 36   - Cost: 0.426     Valid Accuracy: 0.898
Epoch: 37   - Cost: 0.42      Valid Accuracy: 0.898
Epoch: 38   - Cost: 0.415     Valid Accuracy: 0.899
Epoch: 39   - Cost: 0.41      Valid Accuracy: 0.899
Epoch: 40   - Cost: 0.405     Valid Accuracy: 0.9
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| Epoch: 41 | - Cost: 0.4   | Valid Accuracy: 0.901 |
| Epoch: 42 | - Cost: 0.396 | Valid Accuracy: 0.901 |
| Epoch: 43 | - Cost: 0.392 | Valid Accuracy: 0.901 |
| Epoch: 44 | - Cost: 0.387 | Valid Accuracy: 0.901 |
| Epoch: 45 | - Cost: 0.383 | Valid Accuracy: 0.902 |
| Epoch: 46 | - Cost: 0.379 | Valid Accuracy: 0.902 |
| Epoch: 47 | - Cost: 0.376 | Valid Accuracy: 0.903 |
| Epoch: 48 | - Cost: 0.372 | Valid Accuracy: 0.903 |
| Epoch: 49 | - Cost: 0.368 | Valid Accuracy: 0.904 |
| Epoch: 50 | - Cost: 0.365 | Valid Accuracy: 0.904 |
| Epoch: 51 | - Cost: 0.361 | Valid Accuracy: 0.904 |
| Epoch: 52 | - Cost: 0.358 | Valid Accuracy: 0.905 |
| Epoch: 53 | - Cost: 0.355 | Valid Accuracy: 0.905 |
| Epoch: 54 | - Cost: 0.352 | Valid Accuracy: 0.906 |
| Epoch: 55 | - Cost: 0.349 | Valid Accuracy: 0.907 |
| Epoch: 56 | - Cost: 0.346 | Valid Accuracy: 0.907 |
| Epoch: 57 | - Cost: 0.343 | Valid Accuracy: 0.907 |
| Epoch: 58 | - Cost: 0.341 | Valid Accuracy: 0.908 |
| Epoch: 59 | - Cost: 0.338 | Valid Accuracy: 0.908 |
| Epoch: 60 | - Cost: 0.336 | Valid Accuracy: 0.908 |
| Epoch: 61 | - Cost: 0.333 | Valid Accuracy: 0.908 |
| Epoch: 62 | - Cost: 0.331 | Valid Accuracy: 0.908 |
| Epoch: 63 | - Cost: 0.328 | Valid Accuracy: 0.909 |
| Epoch: 64 | - Cost: 0.326 | Valid Accuracy: 0.909 |
| Epoch: 65 | - Cost: 0.324 | Valid Accuracy: 0.909 |
| Epoch: 66 | - Cost: 0.322 | Valid Accuracy: 0.909 |
| Epoch: 67 | - Cost: 0.32  | Valid Accuracy: 0.91  |
| Epoch: 68 | - Cost: 0.317 | Valid Accuracy: 0.91  |
| Epoch: 69 | - Cost: 0.315 | Valid Accuracy: 0.91  |
| Epoch: 70 | - Cost: 0.314 | Valid Accuracy: 0.91  |
| Epoch: 71 | - Cost: 0.312 | Valid Accuracy: 0.911 |
| Epoch: 72 | - Cost: 0.31  | Valid Accuracy: 0.911 |
| Epoch: 73 | - Cost: 0.308 | Valid Accuracy: 0.911 |
| Epoch: 74 | - Cost: 0.306 | Valid Accuracy: 0.911 |
| Epoch: 75 | - Cost: 0.304 | Valid Accuracy: 0.911 |
| Epoch: 76 | - Cost: 0.303 | Valid Accuracy: 0.911 |
| Epoch: 77 | - Cost: 0.301 | Valid Accuracy: 0.911 |
| Epoch: 78 | - Cost: 0.299 | Valid Accuracy: 0.911 |
| Epoch: 79 | - Cost: 0.298 | Valid Accuracy: 0.912 |

Test Accuracy: 0.909200012684

In [ ]: