# TensorFlow and Keras

CMPT 498/820 Machine Learning Tutorial 8

## Najeeb Khan

November 30, 2016

#### 1 TensorFlow

TensorFlow is an interface for expressing machine learning algorithms, and an implementation for executing such algorithms (Abadi et al., 2015).

- Represents computations as graphs
- Nodes in the graph are called **ops**
- Edges are tensors
- Represents data as tensors
- A tensor is an n-dimensional array with a rank, shape and type.
- For example [batch, height, width, channels]
- Executes graphs in the context of **sessions**
- A session places the graph ops onto devices such as CPUs/GPUs etc.
- Maintains state with variables
- Typically represent the parameters of a statistical model as a set of variables
- Uses **feeds** and **fetches** to get data into and out of arbitrary operations.

#### 1.1 TensorFlow Example I

```
In [1]: %matplotlib inline
    import tensorflow as tf

# Define two constants
    matrix1 = tf.constant([[3., 3.]])
    matrix2 = tf.constant([[2.],[2.]])

# Define a matmul operation
    product = tf.matmul(matrix1, matrix2)

# Launch the default graph.
    sess = tf.Session()

# Run the matmul operation
    result = sess.run(product)
    print(result)
```

```
# Close the Session.
sess.close()
[[ 12.]]
```

## 1.2 TensorFlow Example II

```
In [2]: state = tf.Variable(0.0, name="counter")
        inc = tf.placeholder(tf.float32)
        new_value = tf.add(state, inc)
        update = tf.assign(state, new_value)
        init_op = tf.initialize_all_variables()
        # Launch the graph
        with tf.Session() as sess:
            with tf.device("/gpu:0"):
                # Run the init op
                sess.run(init_op)
                # Run the op that updates state
                for _{\rm in} range(3):
                    sess.run([update], feed_dict={inc:0.5})
                    print(sess.run(state))
0.5
1.0
1.5
```

#### 1.3 Autoencoder in TensorFlow

```
In [3]: import tensorflow as tf
    import numpy as np
    import matplotlib.pyplot as plt
```

#### Import data

## Hyperparameters

```
In [5]: # Parameters
        learning_rate = 0.05
        training\_epochs = 30
        batch\_size = 64
        display_step = 10
        examples_to_show = 10
        # Network Parameters
        n_hidden_1 = 64 # 1st layer num features
        n_input = 784 # MNIST data input (img shape: 28*28)
Construct the graph
In [6]: # tf Graph input (only pictures)
        X = tf.placeholder("float", [None, n_input])
        weights = {
            'encoder_h1': tf.Variable(tf.random_normal([n_input, n_hidden_1])),
            'decoder_h1': tf.Variable(tf.random_normal([n_hidden_1, n_input])),
        }
        biases = {
            'encoder_b1': tf.Variable(tf.random_normal([n_hidden_1])),
            'decoder_b1': tf.Variable(tf.random_normal([n_input])),
        }
        # Building the encoder
        def encoder(x):
            # Encoder Hidden layer with sigmoid activation #1
            layer_1 = tf.nn.sigmoid(tf.add(tf.matmul(x, weights['encoder_h1']),
                                            biases['encoder_b1']))
            return layer_1
        # Building the decoder
        def decoder(x):
            # Encoder Hidden layer with sigmoid activation #1
            layer_1 = tf.nn.sigmoid(tf.add(tf.matmul(x, weights['decoder_h1']),
                                            biases['decoder_b1']))
            return layer_1
        # Construct model
        encoder_op = encoder(X)
        decoder_op = decoder(encoder_op)
```

```
y true = X
        # Define loss and optimizer, minimize the squared error
        cost = tf.reduce_mean(tf.pow(y_true - y_pred, 2))
        # Create an optimizer with the desired parameters.
        optimizer = tf.train.RMSPropOptimizer(learning_rate).minimize(cost)
        # Initializing the variables
        init = tf.initialize_all_variables()
Execute the graph
In [7]: # Launch the graph
        with tf.Session() as sess:
            sess.run(init)
            total_batch = int(mnist.train.num_examples/batch_size)
            # Training cycle
            for epoch in range(training_epochs):
                # Loop over all batches
                for i in range(total batch):
                    batch_xs, batch_ys = mnist.train.next_batch(batch_size)
                    # Run optimization op (backprop) and cost op (to get loss value
                    _, c = sess.run([optimizer, cost], feed_dict={X: batch_xs})
                # Display logs per epoch step
                if epoch % display_step == 0:
                    print("Epoch:", '%04d' % (epoch+1),
                          "cost=", "{:.9f}".format(c))
            print("Optimization Finished!")
            # Applying encode and decode over test set
            encode_decode = sess.run(
                y_pred, feed_dict={X: mnist.test.images[:examples_to_show]})
            # Compare original images with their reconstructions
            f, a = plt.subplots(2, 10, frameon=False, figsize=(10, 2))
            f.suptitle('Top row: original images, bottom row: predicted images')
            for i in range(examples_to_show):
                a[0][i].imshow(np.reshape(mnist.test.images[i], (28, 28)))
                a[1][i].imshow(np.reshape(encode_decode[i], (28, 28)))
                a[0][i].axis('off')
                a[1][i].axis('off')
```

# Prediction

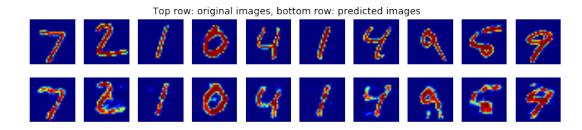
y\_pred = decoder\_op

# Targets (Labels) are the input data.

#### plt.draw()

```
Epoch: 0001 cost= 0.064531557
Epoch: 0011 cost= 0.028448787
Epoch: 0021 cost= 0.023096683
```

Optimization Finished!



#### 2 Keras

Keras is a high-level library that can utilize TensorFlow or Theano as its backend. Here we train a simple deep NN on the MNIST dataset.

### **Imports**

Using TensorFlow backend.

## Hyperparameters

#### Preprocessing

```
In [10]: # the data, shuffled and split between train and test sets
         (X_train, y_train), (X_test, y_test) = mnist.load_data()
         X \text{ train} = X \text{ train.reshape}(60000, 784)
         X_{\text{test}} = X_{\text{test.reshape}} (10000, 784)
         X_train = X_train.astype('float32')
         X_test = X_test.astype('float32')
         X_{train} /= 255
         X_test /= 255
         print(X_train.shape[0], 'train samples')
         print(X_test.shape[0], 'test samples')
         # convert class vectors to binary class matrices
         Y_train = np_utils.to_categorical(y_train, nb_classes)
         Y_test = np_utils.to_categorical(y_test, nb_classes)
60000 train samples
10000 test samples
Model Creation
In [11]: model = Sequential()
         model.add(Dense(512, input_shape=(784,)))
         model.add(Activation('relu'))
         model.add(Dropout(0.2))
         model.add(Dense(512))
         model.add(Activation('relu'))
         model.add(Dropout(0.2))
         model.add(Dense(10))
         model.add(Activation('softmax'))
In [12]: model.compile(loss='categorical_crossentropy',
                        optimizer=RMSprop(),
                        metrics=['accuracy'])
Train and Evaluate
In [13]: history = model.fit(X_train, Y_train,
                              batch_size=batch_size, nb_epoch=nb_epoch,
                              verbose=0, validation_data=(X_test, Y_test),
                              callbacks=[TensorBoard(log_dir='./logs',
                                                      histogram_freq=0,
                                                      write_graph=True,
                                                      write_images=False)])
         score = model.evaluate(X_test, Y_test, verbose=0)
         print('Test score:', score[0])
         print('Test accuracy:', score[1])
```

Test score: 0.109998893419

Test accuracy: 0.9841

## Go to a terminal and type

tensorboard --logdir=./logs

then open the address given by the above command (usually localhost:6006) using Google Chrome.

# val\_acc

