```
In [1]:
         2
               Author:
                              TFT
            # Written:
                              05/02/2018
              Last updated: 05/02/2018
         6
            # TFT Machine Learning
            # Assignment Week 3-1
         9
In [2]:
         1 import numpy as np
         2 import tensorflow as tf
         3 from tensorflow.examples.tutorials.mnist import input data
            import time
         5
           # download the dataset
         7  # each image is 28 x 28
         8 mnist = input data.read data sets('data', one hot=True)
        Extracting data/train-images-idx3-ubyte.gz
```

Extracting data/train-images-idx3-ubyte.gz Extracting data/train-labels-idx1-ubyte.gz Extracting data/t10k-images-idx3-ubyte.gz Extracting data/t10k-labels-idx1-ubyte.gz

```
In [3]:
         1 print(mnist.train.num examples)
         2 print(mnist.validation.num examples)
         3 print(mnist.test.num examples)
            print(mnist.train.images.shape)
         5 print(mnist.train.labels.shape)
           img = mnist.train.images[1]
         7 label = mnist.train.labels[1]
         8 print(img.shape)
            print(label.shape)
        10 print(label)
        11 print(np.argmax(label))
        12
        13 import matplotlib.pyplot as plt
        14 %matplotlib inline
        15 plt.imshow(img.reshape([28,28]), cmap = "gray")
        16 plt.show()
```

```
55000

5000

10000

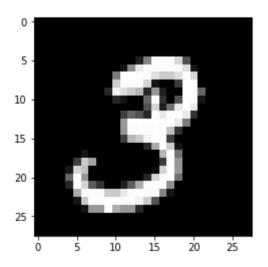
(55000, 784)

(55000, 10)

(784,)

(10,)

[0. 0. 0. 1. 0. 0. 0. 0. 0. 0.]
```





Ground Truth: 7



Ground Truth: 3



Ground Truth: 4



Ground Truth: 6



Ground Truth: 1



Ground Truth: 8



Ground Truth: 1



Ground Truth: 0



Ground Truth: 9



Ground Truth: 8

```
1 # 1. Create the model (build the computation graph)
In [5]:
         2  ## Hyperparameters
         3 batch size = 256
          4 n input = 784 # 784 = 28 x 28
         5 n hidden = 256
          6 n classes = 10
         7 | learning rate = 0.725
          9 ## Model input
        10 x = tf.placeholder(tf.float32, [None, n input])
        11
        12 | ## Hidden layer
        13 W1 = tf.Variable(tf.random normal([n input, n hidden], stddev=0.1))
        14 b1 = tf. Variable(tf.zeros([n hidden]))
        15 h1 = tf.matmul(x, W1) + b1
        16 \mid h1 = tf.nn.relu(h1)
        17
        18 ## Hidden layer
        19 #W2 = tf.Variable(tf.random normal([n hidden, n hidden], stddev=0.1))
        20 | #b2 = tf.Variable(tf.zeros([n hidden]))
        21 \mid \#h2 = tf.matmul(h1, W2) + b2
        22 \mid \#h2 = tf.nn.relu(h2)
         23
         24 ## Output layer
        25 W out = tf.Variable(tf.random normal([n hidden, n classes], stddev=0.1))
         26 | b out = tf. Variable(tf.zeros([n classes]))
        27 | y pred = tf.matmul(h1, W out) + b out
        28
        29 | #W1 = tf.Variable(tf.random normal([n input, n classes], stddev=0.1))
        30 | #b1 = tf.Variable(tf.zeros([n classes]))
         31 \#y \text{ pred} = tf.matmul(x, W1) + b1
         32
         33
        34 ## Define loss and optimizer
        35 y gt = tf.placeholder(tf.float32, [None, n classes])
        36 loss = tf.reduce mean(tf.nn.softmax cross entropy with logits v2(logits=y pred, labels=y gt, name='loss'))
         37
         38 | ## Train (update model parameters)
        39 optimizer = tf.train.GradientDescentOptimizer(learning rate)
        40 #optimizer = tf.train.MomentumOptimizer(learning rate, 0.9)
         41 #optimizer = tf.train.AdamOptimizer()
```

```
train_step = optimizer.minimize(loss)

## Compute Accuracy

cls_pred = tf.argmax(y_pred, axis = 1)

cls_gt = tf.argmax(y_gt, axis = 1)

correct_prediction = tf.equal(cls_pred, cls_gt)

accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
```

```
In [6]:
         1 # 2. Train
         2 ## Initialize
         3 sess = tf.InteractiveSession()
           tf.global variables initializer().run()
         5 max iter = 2000
            for iter in range(max iter):
                batch x, batch y = mnist.train.next batch(batch size)
         8
                sess.run(train step, feed dict = {x: batch x, y qt: batch y})
         9
                if iter % 100 == 0:
        10
                    train loss = sess.run(loss, feed dict = {x: batch x, y gt: batch y})
        11
                    train accuracy = sess.run(accuracy, feed dict = {x: batch x, y qt: batch y})
        12
        13
                    validation x = mnist.validation.images
        14
                    validation y = mnist.validation.labels
        15
                    validation accuracy = sess.run(accuracy, {x: validation x, y qt: validation y})
        16
        17
                    print("iter step %d, loss %f, training accuracy %f, validation accuracy %f" %
        18
                           (iter, train loss, train accuracy, validation accuracy))
```

```
iter step 0, loss 2.573666, training accuracy 0.398438, validation accuracy 0.288800
iter step 100, loss 0.234085, training accuracy 0.929688, validation accuracy 0.934200
iter step 200, loss 0.130955, training accuracy 0.957031, validation accuracy 0.949600
iter step 300, loss 0.062175, training accuracy 0.984375, validation accuracy 0.958000
iter step 400, loss 0.095586, training accuracy 0.976562, validation accuracy 0.964200
iter step 500, loss 0.090505, training accuracy 0.984375, validation accuracy 0.966800
iter step 600, loss 0.063141, training accuracy 0.984375, validation accuracy 0.970000
iter step 700, loss 0.072876, training accuracy 0.984375, validation accuracy 0.969400
iter step 800, loss 0.037901, training accuracy 0.992188, validation accuracy 0.973600
iter step 900, loss 0.031467, training accuracy 1.000000, validation accuracy 0.976000
iter step 1000, loss 0.030806, training accuracy 0.996094, validation accuracy 0.976200
iter step 1100, loss 0.022884, training accuracy 1.000000, validation accuracy 0.976600
iter step 1200, loss 0.020247, training accuracy 1.000000, validation accuracy 0.976400
iter step 1300, loss 0.021146, training accuracy 0.996094, validation accuracy 0.975200
iter step 1400, loss 0.023413, training accuracy 0.996094, validation accuracy 0.977200
iter step 1500, loss 0.016684, training accuracy 1.000000, validation accuracy 0.980200
iter step 1600, loss 0.012238, training accuracy 1.000000, validation accuracy 0.979400
iter step 1700, loss 0.018151, training accuracy 0.996094, validation accuracy 0.977400
iter step 1800, loss 0.014926, training accuracy 1.000000, validation accuracy 0.978000
iter step 1900, loss 0.018532, training accuracy 0.996094, validation accuracy 0.978400
```

```
1 # 3. Test the trained model
In [7]:
         2 train x = mnist.train.images
         3 train_y = mnist.train.labels
          train accuracy = sess.run(accuracy, {x: train x, y gt: train y})
         6 validation x = mnist.validation.images
         7 validation y = mnist.validation.labels
         8 validation accuracy = sess.run(accuracy, {x: validation_x, y_gt: validation_y})
         9
        10 test x = mnist.test.images
        11 test y = mnist.test.labels
        12 test accuracy = sess.run(accuracy, {x: test_x, y_gt: test_y})
        13
        14 print("train accuray: %f" % train accuracy)
        15 print("validation accuray: %f" % validation accuracy)
        16 print("test accuray: %f" % test accuracy)
```

train accuray: 0.993527 validation accuray: 0.979600 test accuray: 0.977800

```
In [8]:
            for i in range(10):
                img = mnist.test.images[i]
          2
          3
                label = mnist.test.labels[i]
                plt.figure(figsize=(1,1))
          5
                plt.imshow(img.reshape([28, 28]), cmap = "gray")
                plt.show()
                cls = np.argmax(label)
                print("Ground Truth: %d" % cls)
          9
                pred_label = sess.run(y_pred, feed_dict = {x: img.reshape([1, -1])})
        10
                pred cls = np.argmax(pred label)
        11
        12
                print("Model prediction: %s" % pred cls)
```



Ground Truth: 7
Model prediction: 7



Ground Truth: 2
Model prediction: 2



Ground Truth: 1
Model prediction: 1



Ground Truth: 0
Model prediction: 0



Ground Truth: 4
Model prediction: 4



Ground Truth: 1
Model prediction: 1



Ground Truth: 4
Model prediction: 4



Ground Truth: 9

Model prediction: 9



Ground Truth: 5
Model prediction: 5



Ground Truth: 9
Model prediction: 9