```
In [155]:
```

Code kindly shared by Juba Nait Saada
Imports
import numpy as np
import tensorflow as tf
from tensorflow.examples.tutorials.mnist import input_data
import matplotlib.pyplot as plt

In [156]:

```
# Import MNIST dataset
mnist = input_data.read_data_sets("MNIST_data/", one_hot=True)
```

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

In [157]:

```
print("Number of images in training dataset : ", len(mnist.train.images))
print("Number of labels in training dataset : ", len(mnist.train.labels))
print("Number of images in test dataset : ", len(mnist.test.images))
print("Number of labels in test dataset : ", len(mnist.test.labels))
```

```
Number of images in training dataset : 55000
Number of labels in training dataset : 55000
Number of images in test dataset : 10000
Number of labels in test dataset : 10000
```

In [158]:

```
#Some useful functions
def reshape(x):
    return x.reshape(28,28,1)
def visualize_image(x):
    return x.reshape(28,28)
```

In [159]:

```
images_train = np.apply_along_axis(reshape, 1, mnist.train.images)
images_train_visualize = np.apply_along_axis(visualize_image, 1, mnist.train.ima
ges)
images_test = np.apply_along_axis(reshape, 1, mnist.test.images)
images_test_visualize = np.apply_along_axis(visualize_image, 1, mnist.test.image
s)
```

In [166]:

#print("Example of one image : ", images_train[0])

```
In [161]:
```

```
# Visualizing some images
for i in range(10):
    plt.figure(figsize=(1,1))
    first_array=images_train_visualize[i]
    plt.imshow(first_array, cmap="gray")
    plt.show()
    print("label : ", np.argmax(mnist.train.labels[i]))
```



label : 7



label: 3



label : 4



label : 6



label : 1



label : 8



label : 1



label : 0



label : 9



label : 8

In [162]:

```
# Inputs
x = tf.placeholder(tf.float32, [None, 28, 28, 1])
y = tf.placeholder(tf.int32, [None, 10])
# Convolutional Layer #1
conv1 = tf.layers.conv2d(inputs= x,
                         filters=32,
                         kernel_size=[5, 5],
                         padding="same",
                         activation=tf.nn.relu)
# Pooling Layer #1
pool1 = tf.layers.max pooling2d(inputs=conv1, pool size=[2, 2], strides=2)
# Convolutional Layer #2
conv2 = tf.layers.conv2d(
      inputs=pool1,
      filters=64,
      kernel_size=[5, 5],
      padding="same",
      activation=tf.nn.relu)
# Pooling Layer #2
pool2 = tf.layers.max pooling2d(inputs=conv2, pool size=[2, 2], strides=2)
# Dense Layer
pool2_flat = tf.reshape(pool2, [-1, 7 * 7 * 64])
dense = tf.layers.dense(inputs=pool2_flat, units=1024, activation=tf.nn.relu)
# Logits Layer
logits = tf.layers.dense(inputs=dense, units=10)
predictions_probabilities = tf.nn.softmax(logits)
predictions class = tf.argmax(predictions probabilities, 1)
# Error + Training
error = tf.nn.softmax cross entropy with logits(labels=y, logits=predictions pro
babilities)
train = tf.train.GradientDescentOptimizer(learning_rate=0.001).minimize(error)
# Accuracy
correct prediction = tf.equal(tf.argmax(predictions_probabilities, 1), tf.argmax
(y, 1))
accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
```

In [164]:

```
with tf.Session() as sess:
   sess.run(tf.global_variables_initializer())
   epochs = 1000
   for e in range(epochs):
       p = np.random.permutation(len(mnist.train.labels))
       temp_x = images_train[p]
       temp_y = mnist.train.labels[p]
       batch_images = temp_x[:100]
       batch_labels = temp_y[:100]
       sess.run(train, feed_dict = {
           x : batch_images,
           y : batch_labels
       })
       print("Accuracy ", sess.run(accuracy, feed_dict = {
           x : batch_images,
           y : batch_labels
       }))
   print('-----')
   batch_images = images_test[:100]
   batch_labels = mnist.test.labels[:100]
   print("Accuracy on test dataset", sess.run(accuracy, feed_dict = {
           x : batch_images,
           y : batch_labels
       }))
   results predictions = sess.run(predictions class, feed dict = {
           x : batch_images,
           y : batch_labels
       })
```

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Accuracy	on test dataset 0.98	

```
In [171]:
```

```
# Visualizing some results
for i in range(20):
    plt.figure(figsize=(1,1))
    first_array=images_test_visualize[i]
    plt.imshow(first_array, cmap="gray")
    plt.show()
    print("Predicted label : ", results_predictions[i])
    print("True label : ", np.argmax(mnist.test.labels[i]))
```



Predicted label : 7 True label : 7



Predicted label : 2 True label : 2



Predicted label : 1 True label : 1



Predicted label : 0 True label : 0



Predicted label : 4 True label : 4



Predicted label : 1 True label : 1



Predicted label : 4 True label : 4



Predicted label : 9 True label : 9



Predicted label : 5 True label : 5



Predicted label : 9 True label : 9



Predicted label : 0 True label : 0



Predicted label : 6 True label : 6



Predicted label : 9 True label : 9



Predicted label : 0 True label : 0



Predicted label : 1 True label : 1



Predicted label : 5 True label : 5



Predicted label : 9 True label : 9



Predicted label : 7 True label : 7



Predicted label : 8 True label : 3



Predicted label : 4 True label : 4