## CS 534 Machine Learning

Project 3, due Monday, April 16 Chenxi Cai Yifei Ren Qingfeng (Kee) Wang CONTENTS

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# 1 This is first section

I said nothing.

## 1.1 This is a subsection section.

Table 1: This is a table.							
Iteration	3	5	10	20			
Error rate(%)	16.67	10.0	10.0	13.3			

#### 2 Source codes

Finally, codes are enclosed here.

```
import tensorflow as tf
import numpy as np
#This is to list all INFO
tf.logging.set_verbosity(tf.logging.INFO)
def cnn_model_fn(features, labels, mode):
  """ Model function for CNN. This function is used to be called
     later in main function.
    features :: the feature in array, size nnumber-by-784.
       Here 784 = 28*28 is the flattened representation of a hand
        written figure.
                     Features is a dict structure, { 'x ': < tf.
                        Tensor 'fifo_queue_DequeueUpTo:1' shape=(
                       nnumber, 784) dtype = float32 > 
                     features['x'] corresponds to the acutal
                        tensor object
    labels
                :: values from 0 to 9. Size <math>nnumber-by-1. A
       tensor object
                :: One of three modes: TRAIN, EVAL, PREDICT
    mode
  22 22 22
  # It is a function that used to train data.
  # Input Layer
  input_layer = tf.reshape(features["x"], [-1, 28, 28, 1])
  # Convolutional Layer #1
  conv1 = tf.layers.conv2d(
      inputs=input_layer,
      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 and Pooling Layer #2
conv2 = tf.layers.conv2d(
    inputs=pool1,
    filters=64,
    kernel_size=[5, 5],
    padding="same",
    activation=tf.nn.relu)
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)
dropout = tf.layers.dropout(
    inputs=dense, rate=0.4, training=mode == tf.estimator.
      ModeKeys.TRAIN)
# Logits Layer
logits = tf.layers.dense(inputs=dropout, units=10)
predictions = {
    # Generate predictions (for PREDICT and EVAL mode)
    "classes": tf.argmax(input=logits, axis=1),#Calculate class
        on the fly
    \# Add 'softmax_tensor' to the graph. It is used for PREDICT
        and by the
    \# 'logging_hook'.
    "probabilities": tf.nn.softmax(logits, name="softmax_tensor
       ")
}
if mode == tf.estimator.ModeKeys.PREDICT:
  return tf.estimator.EstimatorSpec(mode=mode, predictions=
    predictions)
# Calculate Loss (for both TRAIN and EVAL modes)
loss = tf.losses.sparse_softmax_cross_entropy(labels=labels,
  logits=logits)
```

```
# Configure the Training Op (for TRAIN mode)
  if mode == tf.estimator.ModeKeys.TRAIN:
    optimizer = tf.train.GradientDescentOptimizer(learning_rate
       =0.001) #potimized and learning rate can change
    train_op = optimizer.minimize(
        loss=loss,
        global_step=tf.train.get_global_step())
    return tf.estimator.EstimatorSpec(mode=mode, loss=loss,
      train_op=train_op)
  # Add evaluation metrics (for EVAL mode)
  eval_metric_ops = {"accuracy": tf.metrics.accuracy(labels=
    labels, predictions["classes"])}\#predictions is
     a dict and calculate classes on the fly
  return tf.estimator.EstimatorSpec(mode=mode, loss=loss,
     eval_metric_ops=eval_metric_ops)
def main(aa):
    # Load training and eval data
    mnist = tf.contrib.learn.datasets.load_dataset("mnist")
    p = int(len(mnist.train.images)*0.8) #Probably need to
       randomize
    train_data = mnist.train.images[0:p] # Returns np.array
    train_labels = np.asarray(mnist.train.labels, dtype=np.int32)
       [q:0]
    valid_data = mnist.train.images[p:] # Returns np.array
    valid_labels = np.asarray(mnist.train.labels, dtype=np.int32)
       [p:]
    eval_data = mnist.test.images # Returns np.array
    eval_labels = np.asarray(mnist.test.labels, dtype=np.int32)
```

```
# Create the Estimator
mnist_classifier = tf.estimator.Estimator(model_fn=
   cnn_model_fn, model_dir="/tmp/mnist_convnet_model")
# Set up logging for predictions
tensors_to_log = {"probabilities": "softmax_tensor"}
logging_hook = tf.train.LoggingTensorHook(tensors=
   tensors_to_log, every_n_iter=1)
# Train the model
train_input_fn = tf.estimator.inputs.numpy_input_fn(
    x = \{"x": train_data\}, \#First input, convert the numpy array
        data into a dict structure
    y=train_labels,
    batch_size=200,
    num_epochs=None,
    shuffle=True)
valid_input_fn = tf.estimator.inputs.numpy_input_fn(
    x={"x": valid_data},
    y=valid_labels,
    num_epochs=1,
    shuffle=False)
experiment = tf.contrib.learn.Experiment(
    mnist_classifier,
    train_input_fn,
    valid_input_fn,
    train_steps = 5000,
    eval_steps = None,
    train_steps_per_iteration = 500)
\#experiment.continuous\_train\_and\_eval()
#The rest come from tutorial
\# mnist\_classifier.train (
      input_-fn = train_-input_-fn,
#
#
      s t e p s = 1,
      hooks = [logging_hook])
#
#
#
## Evaluate the model and print results
\# eval_{-}input_{-}fn = tf.estimator.inputs.numpy_input_{-}fn
```

```
x = \{ "x ": eval_- data \},
    #
    #
           y = e v a l_{-} l a b e l s,
    #
           num_{-}epochs=1,
            shuffle = False)
    \# eval\_results = mnist\_classifier.evaluate(input\_fn =
        e v a l_{-} i n p u t_{-} f n)
    \# print(eval_results)
if __name__ == '__main__':
    ""Runs\ whole\ fitting\ program\ automatically"""
    import time
    start_time = time.time()
    tf.app.run()
    print("--- %s seconds ---" % (time.time() - start_time))
```