# TensorFlow - MNIST example -

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Ando Ki, Ph.D. adki@future-ds.com

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# Get TensorFlow packages

- This package is required to utilize MNIST dataset.
- Go to the project directory

\$ cd \$(PROJECT)/codes

Get a copy of TensorFlow package

\$ git clone https://github.com/tensorflow/tensorflow.git

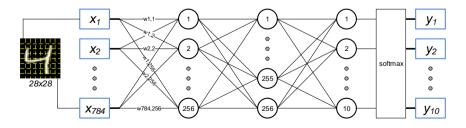
- or visit following site and get a copy of it
  - check directory hierarchy and its name; modify if necessary

https://github.com/tensorflow/tensorflow

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## MNIST using MLP

- MLP (Multi-Layer Perceptron) applies to MNIST
  - ▶ 784 (28x28) inputs of black and white → converted to floating number 0.0 ~ 1.0
  - ▶ 10 outputs representing digit 0 to 9
  - ► Two hidden layer
  - ▶ 256 features for each hidden layer



\$(PROJECT)/codes/tensorflow-projects/mnist-projects/mlp

# MNIST using MLP

- Total testing patterns
  - ► 60K images (train patterns: 55K images)
- Batch number
  - ► 100 → group 600 batches (train batches: 550)
- Epoch number
  - ► The number of whole trainings (forward calculation and back propagation)

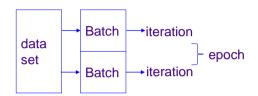
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# MNIST using MLP Weight for layer 1 bias for layer 2 bias for output This operation makes y, which consists of 10 elements

# **Terminology**

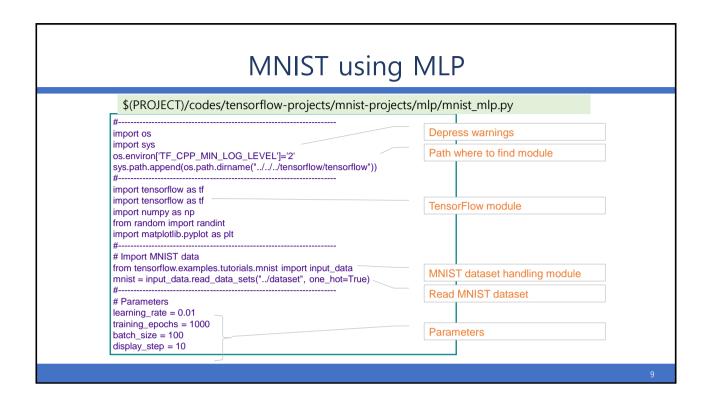
- Iteration
  - Forward and backward for a number of inputs (i.e., batch or minibatch)
- Batch or minibatch
  - A number of inputs (e.g., testing images) to complete an iteration
  - Batch size: the number of training examples in one forward/backward pass
    - The bigger the batch size, the more memory space needed
  - Minibatch
    - Take a small number of examples at a time, ranging from 1 to a few hundred, during one iteration
- Epoch
  - one epoch: one forward pass and one backward pass of all the training examples
  - it contains a number of iterations

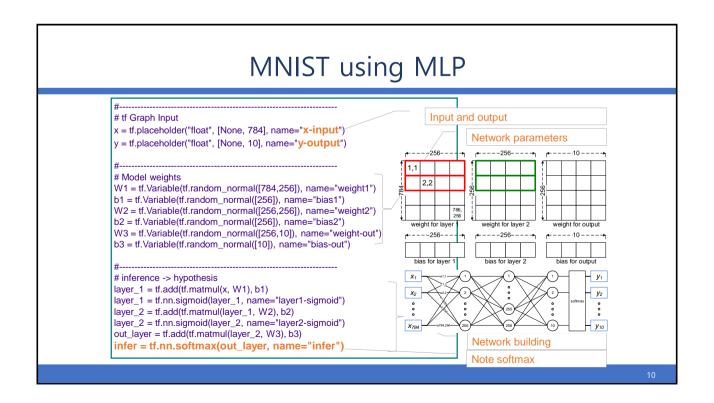
- 1 epoch = (number of iterations)
- = (total training examples) /
- (batch size or minibatch size)



TensorFlow oprators

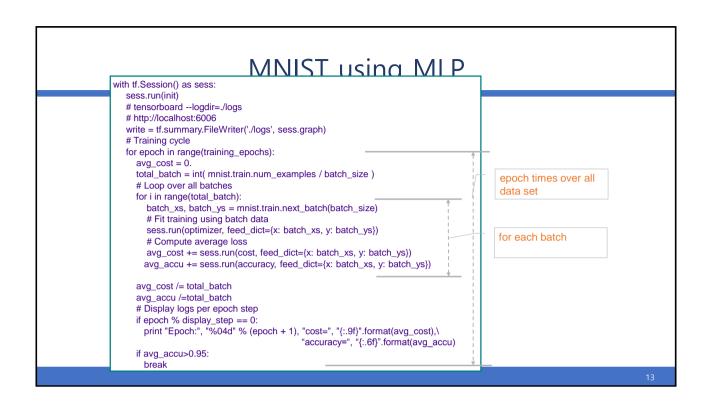
- tf.placeholder()
- tf.Variable()
- tf.global\_variables\_initilizer()
- tf.add()
- tf.matmul()
- tf.reduce\_mean()
- tf.nn.relu()
- tf.train.AdamOptimizer()
- tf.train.GradientDescentOptimizer()
- tf.argmax()

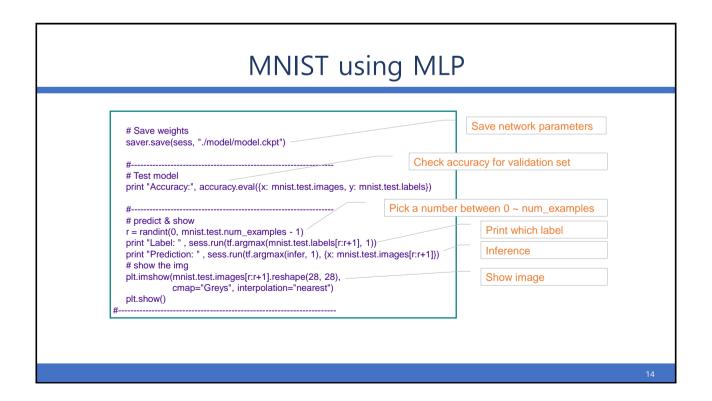


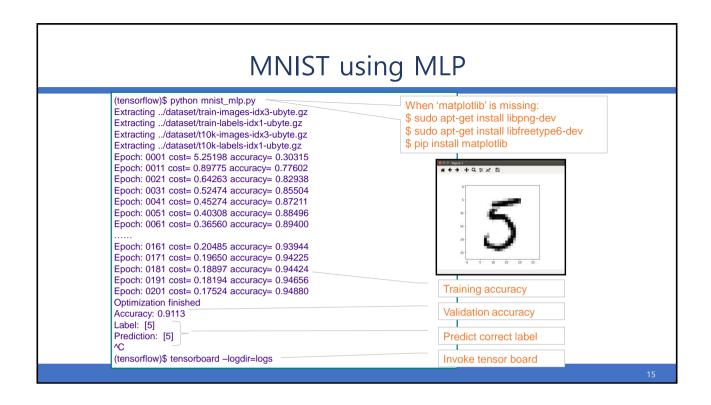


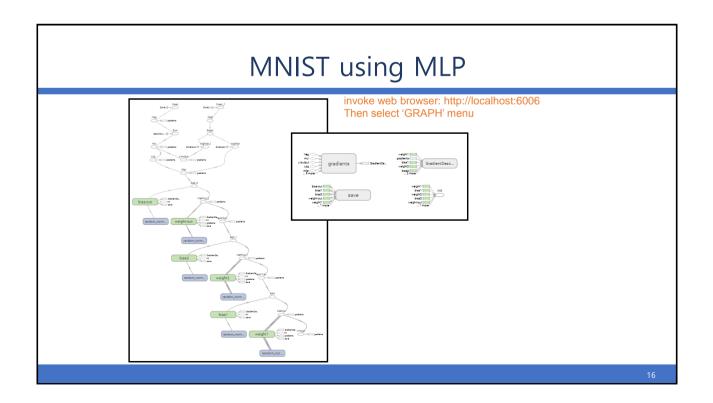
#### MNIST using MLP cost and optimizer # Minimize error using cross entropy cost = tf.reduce\_mean(-tf.reduce\_sum(y \* tf.log(infer), reduction\_indices=1)) y => (100, 10)optimizer = tf.train.GradientDescentOptimizer(learning\_rate).minimize(cost) cost will be scalar # Evaluate -- 0~1 accuracy Accuracy correct = tf.equal(tf.argmax(infer, 1), tf.argmax(y,1)) accuracy = tf.reduce\_mean(tf.cast(correct, "float")) # Initializing the variables Initialize init = tf.global\_variables\_initializer() # Add ops to save and restore all the variables saver = tf.train.Saver() Prepare for save/restore

MNIST using MIP with tf.Session() as sess: training epochs = 100 sess.run(init) # tensorboard --logdir=./logs total batch = 550# http://localhost:6006 write = tf.summary.FileWriter('./logs', sess.graph) # Training cycle  $batch_xs = (100, 784)$ for epoch in range(training\_epochs): batch\_ys = (100, 10) $avg_cost = 0.$ total\_batch = int( mnist.train.num\_examples / batch\_size ) # Loop over all batches for i in range(total\_batch): batch\_xs, batch\_ys = mnist.train.next\_batch(batch\_size) # Fit training using batch data sess.run(optimizer, feed\_dict={x: batch\_xs, y: batch\_ys}) # Compute average loss avg\_cost += sess.run(cost, feed\_dict={x: batch\_xs, y: batch\_ys}) avg\_accu += sess.run(accuracy, feed\_dict={x: batch\_xs, y: batch\_ys}) avg\_cost /= total\_batch avg accu /=total batch Stop when sufficient accuracy has been reached # Distplay logs per epoch step if epoch % display\_step == 0: print "Epoch:", "%04d" % (epoch + 1), "cost=", "{:.9f}".format(avg\_cost),\ "accuracy=", "{:.6f}".format(avg\_accu) if avg\_accu>0.95: break

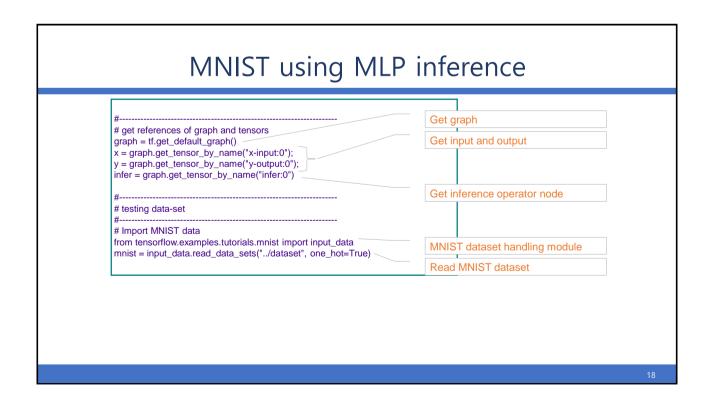


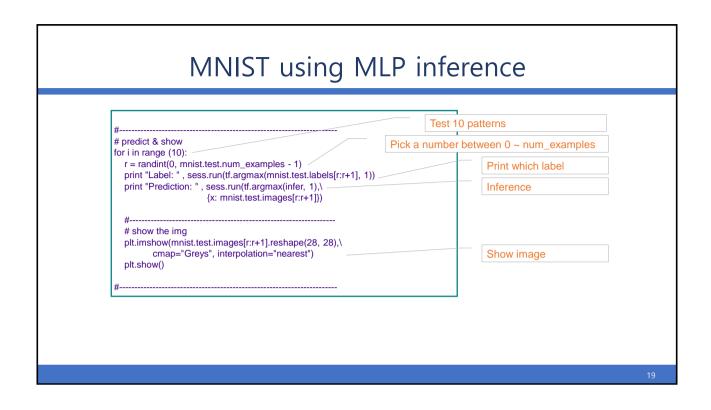


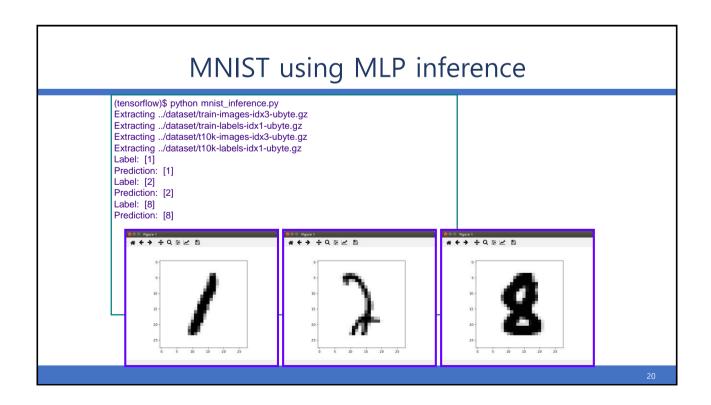




MNIST using MLP inference	
\$(PROJECT)/codes/tensorflow-projects/mnist-projects/mlp/mnist_inference.py  #	
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# MNIST using MLP

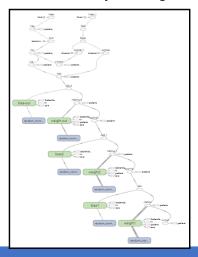
- This example shows how to use MLP to test MNIST
  - ► Step 1: go to your project directory
    - [user@host] cd \$(PROJECT)/codes/tensorflow-project/mnist-project/mlp
  - ► Step 2: see the codes
  - ► Step 3: run Python under virtual environment
    - ⇒ (do not forget to run '\$ source ~/tensorflow/bin/activate')
    - [user@host] python mnist\_mlp.py
    - [user@host] python mnist\_inference.py

[user@host] cd \$(PROJECT)/codes/tensorflow-project/mnist-project/mlp [user@host] python mnist\_mlp.py [user@host] python mnist\_inference.py

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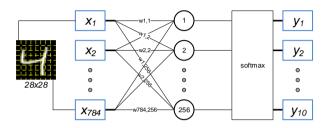
# MNIST using MLP

■ Make more structured style using "with tf.name.scope("LAYER1"):"



# MNIST using single hidden layer network

- Logistic regression for MNIST
  - ▶ 784 (28x28) inputs of black and white → converted to floating number 0.0 ~ 1.0
  - ▶ 10 outputs representing digit 0 to 9
  - one hidden layer
  - > 256 features for each hidden layer



\$(PROJECT)/codes/tensorflow-projects/mnist-projects/single

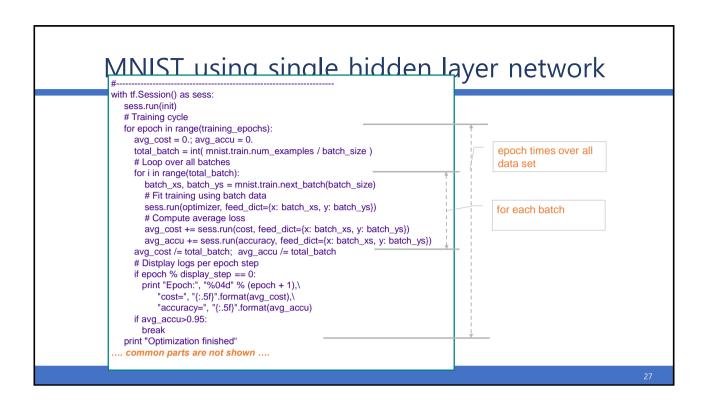
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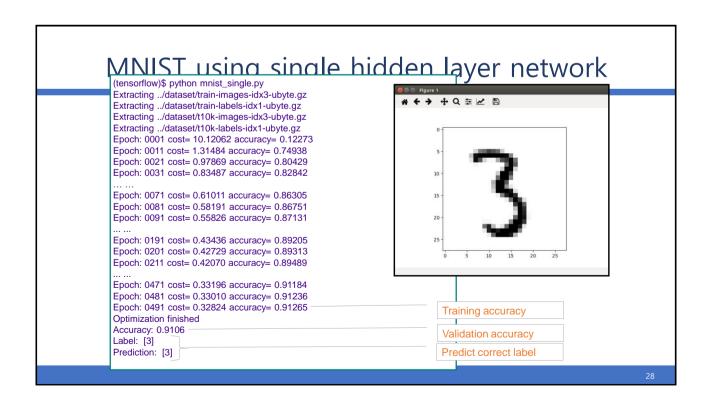
# MNIST using single hidden layer network

- Total testing patterns
  - ▶ 60K images
- Batch number
  - ► 100 → group 600 batches
- Epoch number
  - ► The number of whole trainings (forward calculation and back propagation)

#### MNIST using single hidden layer network 0.5 0.7 0.1 0.0 1,1 2,2 84 786. 256 weight for layer 1 weight for output --256----10output (y) of each batch size bias for layer 1 bias for output This operation makes y, which consists of 10 elements

#### MNIST using single hidden layer network # tf Graph Input x = tf.placeholder("float", [None, 784], name="x-input") # mnist data image of shape 28\*28 y = tf.placeholder("float", [None, 10], name="y-output") # 0-9 digits recognition => 10 classes # Model weights W = tf.Variable(tf.zeros([784, 10]), name="weight") b = tf.Variable(tf.zeros([10]), name="bias") # inference -> hypothesis infer = tf.nn.softmax(tf.matmul(x, W) + b, name="infer") # Minimize error using cross entropy cost = tf.reduce\_mean(-tf.reduce\_sum(y \* tf.log(infer), reduction\_indices=1)) optimizer = tf.train.GradientDescentOptimizer(learning\_rate).minimize(cost) # Evaluate -- 0~1 accuracy correct = tf.equal(tf.argmax(infer, 1), tf.argmax(y,1)) accuracy = tf.reduce\_mean(tf.cast(correct, "float")) # Initializing the variables init = tf.global\_variables\_initializer()





# MNIST using single hidden layer network

- This example shows how to use single hidden layer to test MNIST
  - Step 1: go to your project directory
    - [user@host] cd \$(PROJECT)/codes/tensorflow-project/mnist-project/single
  - Step 2: see the codes
  - ► Step 3: run Python under virtual environment
    - (do not forget to run '\$ source ~/tensorflow/bin/activate')
    - [user@host] python mnist\_single.py
    - [user@host] python mnist\_inference.py

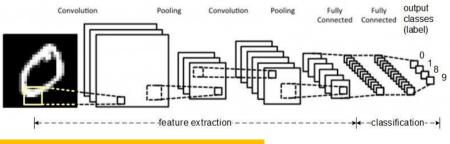
Prepare your own code in \$(PROJECT)/codes/tensorflow-projects/mnist-projects/single

[user@host] cd \$(PROJECT)/codes/tensorflow-project/mnist-project/single [user@host] python mnist\_single.py [user@host] python mnist\_inference.py

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## **MNIST using CNN**

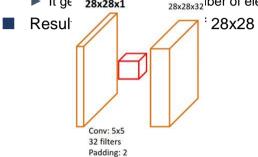
- Logistic regression for MNIST
  - ▶ 784 (28x28) inputs of black and white → converted to floating number 0.0 ~ 1.0
  - ▶ 10 outputs representing digit 0 to 9
  - one hidden layer
  - 256 features for each hidden layer



\$(PROJECT)/codes/tensorflow-projects/mnist-projects/cnn

# MNIST using CNN: 1st convolutional layer

- Input: 28x28 pixels
- Convolution filter: 32 kernels with 5x5
- Convolution: stride 1
  - It ge 28x28x1 28x28x32 ber of elements



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# MNIST using CNN: 1st pooling

- Input: 32 features with 28x28
- Max pooling filter: 5x5
- Convolution: stride 2

Results

Pool: 2x2
stride: 2

with tf.name\_scope("CONV-POOL-1"):

# Convolution Layer

conv1 = conv2d(x, weights['wc1'], biases['bc1'])

# Max Pooling (down-sampling)

conv1 = maxpool2d(conv1, k=2)

# MNIST using CNN: 2nd convolutional layer

- Input: 32 features with 14x14 pixels
- Convolution filter: 64 kernels with 5x5
- Convolution: stride 1
  - ► It gener 14x14x32 14x14x64 r of elements
- Results i

Conv: 5x5 64 filters Padding: 2

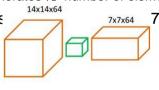
```
with tf.name scope("WEIGHT-BIAS"):
  weights = {
       #5x5 conv, 1 input, 32 outputs
       'wc1': tf.Variable(tf.random_normal([5, 5, 1, 32]),
       # 5x5 conv, 32 inputs, 64 outputs
       'wc2': tf.Variable(tf.random_normal([5, 5, 32, 64]),
name="wc2"),
with tf.name_scope("CONV-POOL-2"):
     # Convolution Layer
     conv2 = conv2d(conv1, weights['wc2'], biases['bc2'])
     # Max Pooling (down-sampling)
     conv2 = maxpool2d(conv2, k=2)
def conv2d(x, W, b, strides=1):
  with tf.name_scope("CONV"):
     # Conv2D wrapper, with bias and relu activation
     x = tf.nn.conv2d(x, W, strides=[1, strides, strides, 1],
        padding='SAME')
     x = tf.nn.bias\_add(x, b)
     return tf.nn.relu(x)
```

3.

# MNIST using CNN: 2nd pooling

- Input: 64 features with 14x14
- Max pooling filter: 2x2
- Convolution: stride 2
  - ▶ It generates ½ number of elements

Results



Pool: 2x2 stride: 2 with tf.name\_scope("CONV-POOL-2"):

# Convolution Layer

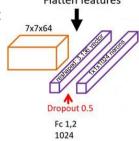
conv2 = conv2d(conv1, weights['wc2'], biases['bc2'])

# Max Pooling (down-sampling)

conv2 = maxpool2d(conv2, k=2)

# MNIST using CNN: fully connected layer

- Input: 64 features with 7x7
- Reshaping: 3-D array to 1-D vector
  - ► 64x7x7 → 3 136
    Flatten features
- Neurons:



with tf.name\_scope("FC"):
# Fully connected layer
# Reshape conv2 output to fit fully connected layer
input
fc1 = tf.reshape(conv2, [-1,
weights['wd1'].get\_shape().as\_list()[0]])
fc1 = tf.add(tf.matmul(fc1, weights['wd1']),
biases['bd1'])
fc1 = tf.nn.relu(fc1)
# Apply Dropout
fc1 = tf.nn.dropout(fc1, dropout)

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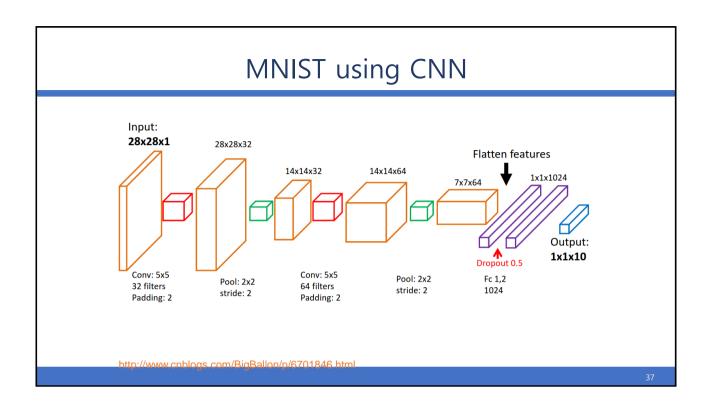
# MNIST using CNN: read-out layer

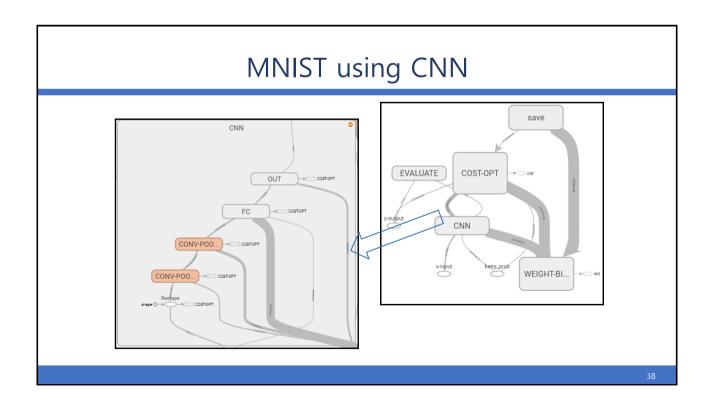
- Input: 1024 neurons
- Output: 10 classes



with tf.name\_scope("OUT"):

# Output, class prediction
out = tf.add(tf.matmul(fc1, weights['out']), biases['out'],
name="add-output")





# **MNIST using CNN**

- This example shows how to use CNN to test MNIST
  - ► Step 1: go to your project directory
    - [user@host] cd \$(PROJECT)/codes/tensorflow-project/mnist-project/cnn
  - ► Step 2: see the codes
  - ► Step 3: run Python under virtual environment
    - (do not forget to run '\$ source ~/tensorflow/bin/activate')
    - [user@host] python mnist\_cnn.py
    - [user@host] python mnist\_inference.py

[user@host] cd \$(PROJECT)/codes/tensorflow-project/mnist-project/cnn [user@host] python mnist\_cnn.py [user@host] python mnist\_inference.py

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㈜퓨쳐디자인시스템 34051 대전광역시 유성구 문지로 193, KAIST 문지캠퍼스, F723호 (042) 864-0211~0212 / contact@future-ds.com / www.future-ds.com

Future Design Systems, Inc.
Faculty Wing F723, KAIST Munji Campus, 193 Munji-ro, Yuseong-gu, Daejeon 34051, Korea +82-042-864-0211~0212 / contact@future-ds.com / www.future-ds.com



