

# Tensorflow

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- TensorFlow enables researchers to build machine learning model
- Combination of strategy and mechanism

- 1、 Using the **graph** to represent the computational task
- 2、 Launch the graph in **session**
- 3、 Using **tensor** to represent data
- 4、 Using **variable** maintenance state
- 5、 Using **feed** and **fetch** to assign for and get data from any operation

# Process

- 1、 Build a graph
- 2、 Launch the graph in a session
- 3、 close the session

<https://www.tensorflow.org/>

```
# Build a graph.  
a = tf.constant(5.0)  
b = tf.constant(6.0)  
c = a * b  
# Launch the graph in a session.  
sess = tf.Session()  
print sess.run(c)  
#close session  
Sess.close()
```

# Constant Variable placeholder

## **Constant:**

```
tf.constant(value, dtype=None, shape=None,  
name='Const')
```

## **Variable:**

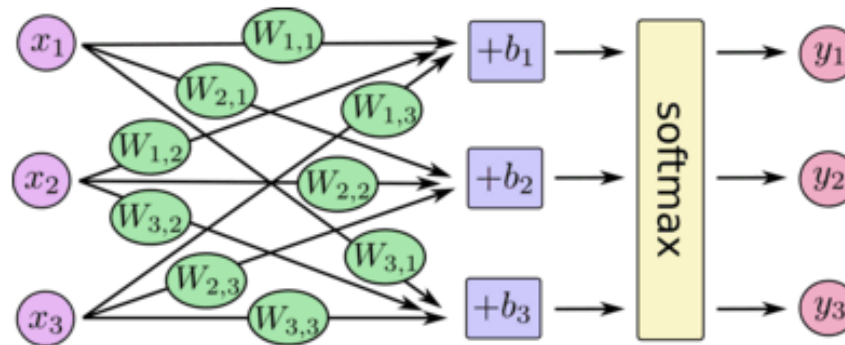
```
vs.get_variable(name, shape=None, dtype=tf.float32,  
initializer=None, regularizer=None, trainable=True,  
collections=None)
```

## **Placeholder**

```
placeholder(dtype, shape=None, name=None)
```

## \$ Example 1

- MNIST & softmax



$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \text{softmax} \begin{pmatrix} W_{1,1}x_1 + W_{1,2}x_2 + W_{1,3}x_3 + b_1 \\ W_{2,1}x_1 + W_{2,2}x_2 + W_{2,3}x_3 + b_2 \\ W_{3,1}x_1 + W_{3,2}x_2 + W_{3,3}x_3 + b_3 \end{pmatrix}$$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \text{softmax} \left( \begin{bmatrix} W_{1,1} & W_{1,2} & W_{1,3} \\ W_{2,1} & W_{2,2} & W_{2,3} \\ W_{3,1} & W_{3,2} & W_{3,3} \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right)$$

# Softmax Train

- **load data**

pass

- **Set Parameters**

Personal habits: Storing parameters in a dictionary

```
Params['W'] = tf.Variable(tf.zeros([784,10]))
```

```
Params['b'] = tf.Variable(tf.zeros([10]))
```



## Softmax Train

- **set Placeholder**

```
x = tf.placeholder(tf.float32, [None, 784])
```

```
y_ = tf.placeholder("float", [None, 10])
```

- **Build graph**

```
y = tf.nn.softmax(tf.matmul(x, params['W']) + params['b'])
```

```
cross_entropy = -tf.reduce_sum(y_*tf.log(y))
```

```
train_step =
```

```
tf.train.GradientDescentOptimizer(0.01).minimize(cross_entropy)
```

# Softmax Train

- **initialize variable op**

Before Variables can be used within a session, they must be initialized using that session.

```
init = tf.initialize_all_variables()
```

- **Launch the graph in a session**

with `tf.Session()` as `sess`:

```
sess.run(init)
```

```
for i in range(1000):
```

```
    batch_xs, batch_ys = mnist.train.next_batch(100)
```

```
    sess.run(train_step, feed_dict={x: batch_xs, y_: batch_ys})
```

```
saver = tf.train.Saver(tf.all_variables())
```

```
saver.save(sess, output_file_name)
```

## Softmax Train

- **If we want to observe the loss:**

```
with tf.Session() as sess:
```

```
    sess.run(init)
```

```
    for i in range(1000):
```

```
        batch_xs, batch_ys = mnist.train.next_batch(100)
```

```
        _, loss = sess.run([train_step, cross_entropy], feed_dict={x: batch_xs,  
                                                                    y_: batch_ys})
```

```
    saver = tf.train.Saver(tf.all_variables())
```

```
    saver.save(sess, output_file_name)
```

# Softmax Train

- **Evaluation**

- **Build graph**

```
y = tf.nn.softmax(tf.matmul(x, params['W']) + params['b'])
```

```
cross_entropy = -tf.reduce_sum(y_*tf.log(y))
```

```
train_step = tf.train.GradientDescentOptimizer(0.01).minimize(cross_entropy)
```

```
correct_prediction = tf.equal(tf.argmax(y,1), tf.argmax(y_,1))
```

```
accuracy = tf.reduce_mean(tf.cast(correct_prediction, "float"))
```

# Softmax Train

- **Launch the graph in a session**

with `tf.Session()` as `sess`:

```
sess.run(init)
```

```
for i in range(1000):
```

```
    batch_xs, batch_ys = mnist.train.next_batch(100)
```

```
    _, loss = sess.run([train_step, cross_entropy], feed_dict={x: batch_xs,  
                                                                y_: batch_ys})
```

```
    if i%100==0:
```

```
        print sess.run(accuracy, feed_dict={x: mnist.test.images,  
                                          y_: mnist.test.labels})
```

```
saver = tf.train.Saver(tf.all_variables())
```

```
saver.save(sess, output_file_name)
```

# Softmax Test

- **test**
  - Load data
  - Set parameters
  - Set Placeholder
  - **Build graph**
  - **Load parameters**
  - **Launch the graph in a session**

# Softmax Test

- **Build graph**

```
y = tf.nn.softmax(tf.matmul(x,params['W']) + params['b'])  
class = tf.argmax(y,1)
```

- **Load parameters**

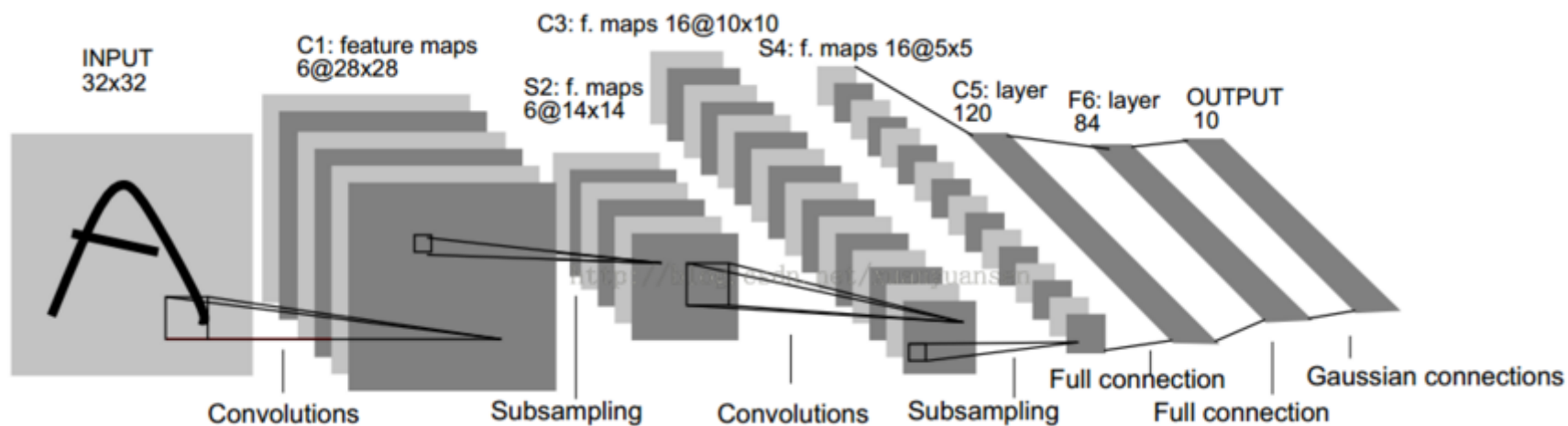
```
saver = tf.train.Saver(tf.all_variables())
```

- **Launch the graph in a session**

```
with tf.Session() as sess:  
    saver.restore(sess, model_path)  
    c = sess.run(class,feed_dict={x:test_x})  
    print c
```

## \$ Example 2

- MNIST & CNN



<http://blog.csdn.net/qiaofangjie/article/details/16826849>



## conv2d

- **conv2d(input, filter, strides, padding, use\_cudnn\_on\_gpu=None, data\_format=None, name=None)**

Computes a 2-D convolution given 4-D `input` and `filter` tensors.

### Args:

**input:** A `Tensor`. Must be one of the following types: `float32`, `float64`.

**filter:** A `Tensor`. Must have the same type as `input`.

**strides:** A list of `ints`. 1-D of length 4. The stride of the sliding window for each dimension of `input`. Must be in the same order as the dimension specified with format.

**padding:** A `string` from: `"SAME", "VALID"`. The type of padding algorithm to use.

**use\_cudnn\_on\_gpu:** An optional `bool`. Defaults to `True`.

## conv2d

**data\_format:** An optional `string` from: `"NHWC", "NCHW"`. Defaults to `"NHWC"`. Specify the data format of the input and output data. With the default format "NHWC", the data is stored in the order of: [batch, in\_height, in\_width, in\_channels]. Alternatively, the format could be "NCHW", the data storage order of: [batch, in\_channels, in\_height, in\_width].

### Returns:

A `Tensor`. Has the same type as `input`.

## conv2d

Given an input tensor of shape `[batch, in\_height, in\_width, in\_channels]` and a filter / kernel tensor of shape `[filter\_height, filter\_width, in\_channels, out\_channels]`, this op performs the following:

1. Flattens the filter to a 2-D matrix with shape `[filter\_height \* filter\_width \* in\_channels, output\_channels]`.
2. Extracts image patches from the input tensor to form a \*virtual\* tensor of shape `[batch, out\_height, out\_width, filter\_height \* filter\_width \* in\_channels]`.
3. For each patch, right-multiplies the filter matrix and the image patch vector.

## conv2d

**In detail, with the default NHWC format,**

$$\text{output}[b, i, j, k] = \sum_{\{d_i, d_j, q\}} \text{input}[b, \text{strides}[1] * i + d_i, \text{strides}[2] * j + d_j, q] * \text{filter}[d_i, d_j, q, k]$$

Must have `strides[0] = strides[3] = 1`. For the most common case of the same horizontal and vertical strides, `strides = [1, stride, stride, 1]`.

## dropout

- **dropout(x, keep\_prob, noise\_shape=None, seed=None, name=None)**

Computes dropout.

With probability `keep_prob`, outputs the input element scaled up by `1 / keep_prob`, otherwise outputs `0`. The scaling is so that the expected sum is unchanged.

By default, each element is kept or dropped independently. If `noise_shape` is specified, it must be

[broadcastable](<http://docs.scipy.org/doc/numpy/user/basics.broadcasting.html>)

to the shape of `x`, and only dimensions with `noise_shape[i] == shape(x)[i]` will make independent decisions. For example, if `shape(x) = [k, l, m, n]` and `noise_shape = [k, 1, 1, n]`, each batch and channel component will be kept independently and each row and column will be kept or not kept together.

# dropout

## Args:

**x:** A tensor.

**keep\_prob:** A scalar `Tensor` with the same type as x. The probability that each element is kept.

**noise\_shape:** A 1-D `Tensor` of type `int32`, representing the shape for randomly generated keep/drop flags.

**seed:** A Python integer. Used to create random seeds. See [`set_random_seed`](../..//api\_docs/python/constant\_op.md#set\_random\_seed) for behavior.

## Returns:

A Tensor of the same shape of `x`.

## Raises:

`ValueError`: If `keep_prob` is not in `(0, 1]`.

## Max\_pool

- **max\_pool(value, ksize, strides, padding, data\_format='NHWC', name=None)**

Performs the max pooling on the input.

### Args:

**value:** A 4-D `Tensor` with shape `[batch, height, width, channels]` and type `tf.float32`.

**ksize:** A list of ints that has length  $\geq 4$ . The size of the window for each dimension of the input tensor.

**strides:** A list of ints that has length  $\geq 4$ . The stride of the sliding window for each dimension of the input tensor.

**padding:** A string, either `VALID` or `SAME`. The padding algorithm.

**data\_format:** A string. 'NHWC' and 'NCHW' are supported.

### Returns:

A `Tensor` with type `tf.float32`. The max pooled output tensor.

# CNN Train

- **Set Parameters**

```
def weight_variable(shape):  
    initial = tf.truncated_normal(shape, stddev=0.1)  
    return tf.Variable(initial)
```

```
def bias_variable(shape):  
    initial = tf.constant(0.1, shape=shape)  
    return tf.Variable(initial)
```

```
params['W_conv1'] = weight_variable([5, 5, 1, 32])  
params['b_conv1'] = bias_variable([32])
```

```
params['W_conv2'] = weight_variable([5, 5, 32, 64])  
params['b_conv2'] = bias_variable([64])
```

```
params['W_fc1'] = weight_variable([7 * 7 * 64, 1024])  
params['b_fc1'] = bias_variable([1024])
```

```
params['W_fc2'] = weight_variable([1024, 10])  
params['b_fc2'] = bias_variable([10])
```



## CNN Train

- **set placeholder**

```
x = tf.placeholder("float", shape=[None, 784])
```

```
x_image = tf.reshape(x, [-1,28,28,1])
```

```
y_ = tf.placeholder("float", shape=[None, 10])
```

```
keep_prob = tf.placeholder("float")
```

# CNN Train

- **build graph**

```
def conv2d(x, W):
```

```
    return tf.nn.conv2d(x, W, strides=[1, 1, 1, 1], padding='SAME')
```

```
def max_pool_2x2(x):
```

```
    return tf.nn.max_pool(x, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1], padding='SAME')
```

```
h_conv1 = tf.nn.relu(conv2d(x_image, params['W_conv1']) + params['b_conv1'])
```

```
h_pool1 = max_pool_2x2(h_conv1)
```

```
h_conv2 = tf.nn.relu(conv2d(h_pool1, params['W_conv2']) + params['b_conv2'])
```

```
h_pool2 = max_pool_2x2(h_conv2)
```

```
h_pool2_flat = tf.reshape(h_pool2, [-1, 7*7*64])
```

```
h_fc1 = tf.nn.relu(tf.matmul(h_pool2_flat, params['W_fc1']) + params['b_fc1'])
```

```
h_fc1_drop = tf.nn.dropout(h_fc1, keep_prob)
```

```
y_conv = tf.nn.softmax(tf.matmul(h_fc1_drop, params['W_fc2']) + params['b_fc2'])
```

## CNN Train

```
cross_entropy = -tf.reduce_sum(y_*tf.log(y_conv))  
train_step = tf.train.AdamOptimizer(1e-4).minimize(cross_entropy)  
correct_prediction = tf.equal(tf.argmax(y_conv,1), tf.argmax(y_,1))  
accuracy = tf.reduce_mean(tf.cast(correct_prediction, "float"))  
saver = tf.train.Saver(tf.all_variables())
```

**tf.cast:** convert the Boolean value into float

## \$ Example 3

Based DNN TTS

## \$ Example 4

Based-RNN TTS