

Tensorflow Tutorial

In this tutorial we will learn:

- XOR Example
- Examples using MNIST



XOR

x_1	x_2	x_1 XOR x_2
0	0	0
0	1	1
1	0	1
1	1	0

Functions

$$\hat{y} = \sigma(w^T \max\{0, W^T x + c\} + b)$$

x is the input matrix with size 4 x 2.

W is the weight matrix of 2 x 2.

c is the bias matrix of 4 x 2.

b is the bias vector of 4 x 1

w is the weight matrix of 2 x 1.

y is output matrix of 4 x 1.

Placeholders

- placeholder – a value that we'll input when we ask TensorFlow to run a computation.
- `X = tf.placeholder(tf.float32, shape=[4,2], name = 'X')`
- `Y = tf.placeholder(tf.float32, shape=[4,1], name = 'Y')`

Variables

- A TensorFlow variable is the best way to represent shared, persistent state manipulated by your program.
- We will fill the weight matrices with random values distributed normally.
- `W = tf.Variable(tf.truncated_normal([2,2]), name = "W")`
- `w = tf.Variable(tf.truncated_normal([2,1]), name = "w")`

Variables

- We will fill the bias matrices with zeroes.
- `c = tf.Variable(tf.zeros([4,2]), name = "c")`
- `b = tf.Variable(tf.zeros([4,1]), name = "b")`

Name Scopes

- Name scopes behave similarly to functions.
- We will define the hidden layer $h = \text{relu}(W.X + c)$. ReLU is the activation function with $\text{relu}(x) = \max\{0, x\}$
- with `tf.name_scope("hidden_layer")` as scope:
- `h = tf.nn.relu(tf.add(tf.matmul(X, W), c))`

Name Scopes

- `y = sigmoid(h.w + b)`
- `with tf.name_scope("output") as scope:`
- `y_estimated = tf.sigmoid(tf.add(tf.matmul(h,w),b))`

Loss Function

- Loss function will be the mean of the squared difference.
- with `tf.name_scope("loss")` as scope:
- `loss = tf.reduce_mean(tf.squared_difference(y_estimated, Y))`

Training Function

- Make the function of training here.
- We are using gradient descent with a step size of 0.01.
- with `tf.name_scope("train")` as scope:
- `train_step = tf.train.GradientDescentOptimizer(0.01).minimize(loss)`

Initialize the variables

- Graph creation is done. Let's start initializing the variables.
- `INPUT_XOR = [[0,0],[0,1],[1,0],[1,1]]`
- `OUTPUT_XOR = [[0],[1],[1],[0]]`

Initialize the variables

- `init = tf.global_variables_initializer()`
- Whatever global variables that you have initialized will be returned here.
- `sess = tf.Session()`
- A session allows to execute graphs or part of graphs. It allocates resources for that and holds the actual values of intermediate results and variables.

Initialize the variables

- **`writer = tf.summary.FileWriter("./logs/xor_logs", sess.graph)`**
- Store your logs
- **`sess.run(init)`**
- Run your sessions by passing initial global variables.
- **`t_start = time.clock()`**
- Start the clock.

WE ARE IN THE FOR LOOP NOW

- Start the for loop.
- for epoch in range(100001):
 - `sess.run(train_step, feed_dict={X: INPUT_XOR, Y: OUTPUT_XOR})`
- We have passed into the `sess.run()` function's `feed_dict` parameter to provide the input examples for this step of training.

WE ARE IN THE FOR LOOP NOW

- We will print out the parameters at every 10000th step.
- if epoch % 10000 == 0:
 - for element in sess.run(y_estimated, feed_dict={X: INPUT_XOR, Y: OUTPUT_XOR}):
 - print(' ', element)
- This will print out values of output at every 10000th step.

WE ARE IN THE FOR LOOP NOW

- Let's print W!
- **for element in sess.run(W):**
 - **print(' ',element)**
- Let's print c!
- **for element in sess.run(c):**
 - **print(' ',element)**
- Let's print loss!
- **print(' loss: ', sess.run(loss, feed_dict={X: INPUT_XOR, Y: OUTPUT_XOR}))**

It's done.

- End your for loops.
- End your clocks.
- You are done. Run your program now.
- Let's have a small break and we will see demos on how to use them for real datasets.