TensorFlow Tutorial

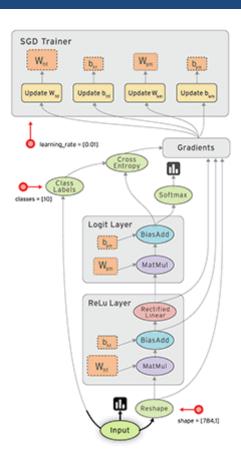
李理



- TensorFlow是什么
- 自动求梯度
- 自定义op
- 基本概念
- install和build
- 例子
- 参考资料

TensorFlow是什么

- TensorFlow就是 Tensor Flow
 - Scalar -> vector -> matrix -> tensor
- Flow in Graph

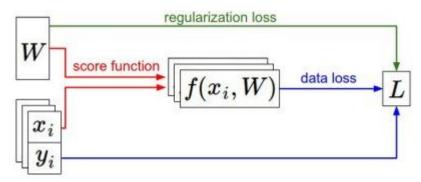


TensorFlow是什么

- Deep Learning -> Tensor Flowing
- Predict

•
$$y' = f(w;x)$$

- Train
 - Loss = L(y, y')
 - W* = argmaxw(Loss)



Optimization

- Stochastic Gradient Descent
 - W += -learning_rate * dw
- 关键问题
 - 求梯度

计算梯度/导数的4种方法

- 手工计算
 - d(x+y) = dx + dy
- 数值求导(Numeric Differentiation)
- 符号求导(Symbolic Differentiation)
- 自动求导(Automatic Differentiation)

数值求导

- 优点
 - 容易实现
- 缺点
 - 计算量大
 - 误差
- 可以用来gradient check

$$\frac{\partial f(\mathbf{x})}{\partial x_i} pprox \frac{f(\mathbf{x} + h\mathbf{e}_i) - f(\mathbf{x})}{h}$$
,

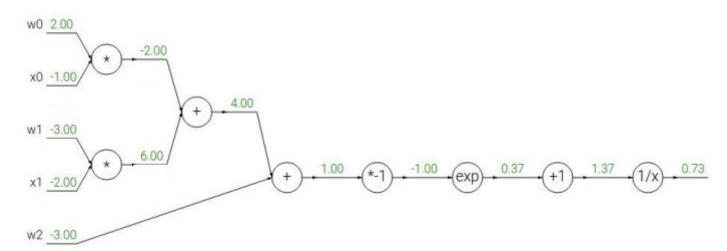
$$\frac{\partial f(\mathbf{x})}{\partial x_i} = \frac{f(\mathbf{x} + h\mathbf{e}_i) - f(\mathbf{x} - h\mathbf{e}_i)}{2h} + O(h^2) ,$$

```
def eval_numerical_gradient_array(f, x, df, h=1e-5):
  Evaluate a numeric gradient for a function that accepts a numpy
  array and returns a numpy array.
  grad = np.zeros_like(x)
  it = np.nditer(x, flags=['multi_index'], op_flags=['readwrite'])
  while not it.finished:
    ix = it.multi_index
    oldval = x[ix]
    x[ix] = oldval + h
    pos = f(x).copy()
    x[ix] = oldval - h
    neg = f(x).copy()
    x[ix] = oldval
    grad[ix] = np.sum((pos - neg) * df) / (2 * h)
    it.iternext()
  return grad
```

符号求导

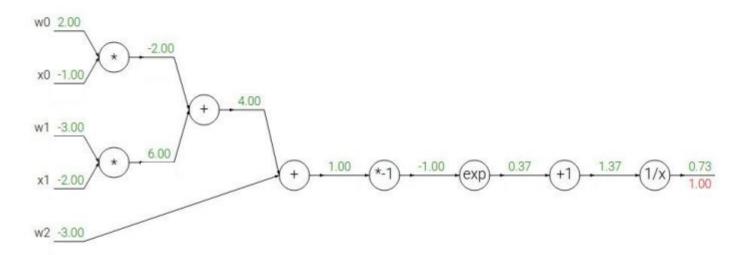
- 算法"自动"推导出Analytic Differentiation
 - matlab maple mathematica等
- 优点
 - 精确的导数,快(取决于符号运算库)
- 缺点
 - 不是所有函数都有解析的导数

$$f(w,x)=rac{1}{1+e^{-(w_0x_0+w_1x_1+w_2)}}$$





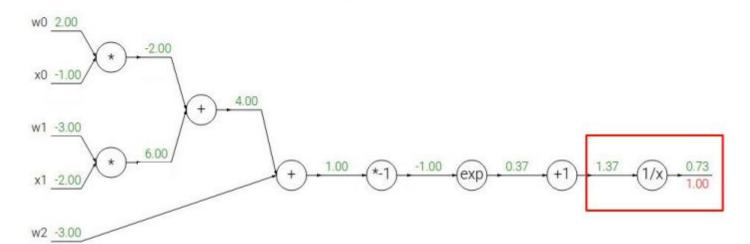
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$$f(x) = e^x \hspace{1cm} o \hspace{1cm} rac{df}{dx} = e^x \hspace{1cm} f(x) = rac{1}{x} \hspace{1cm} o \hspace{1cm} rac{df}{dx} = -1/x^2 \ f_a(x) = ax \hspace{1cm} o \hspace{1cm} rac{df}{dx} = a \hspace{1cm} f_c(x) = c + x \hspace{1cm} o \hspace{1cm} rac{df}{dx} = 1$$



$$f(w,x)=rac{1}{1+e^{-(w_0x_0+w_1x_1+w_2)}}$$

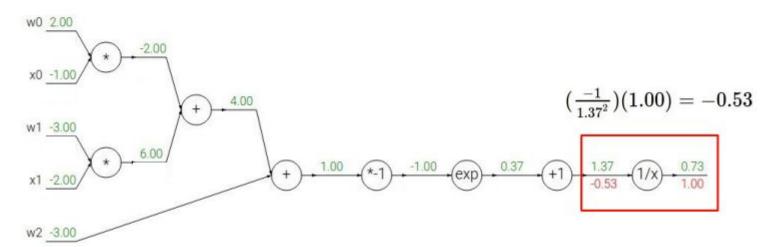


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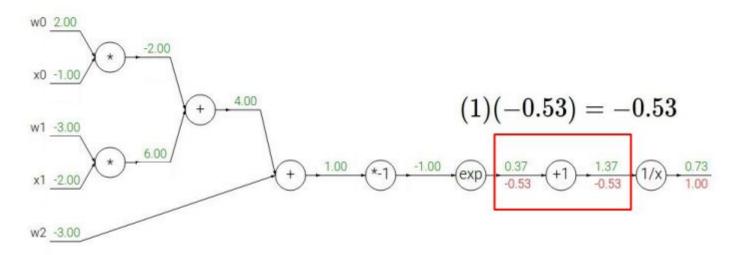


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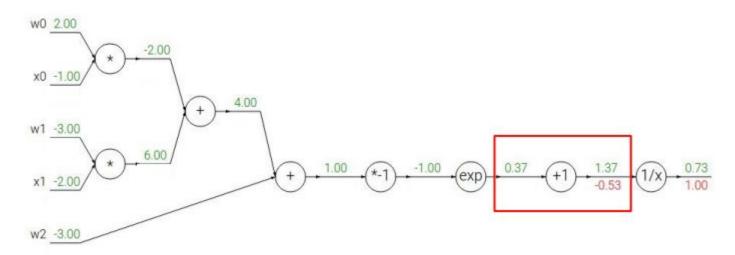
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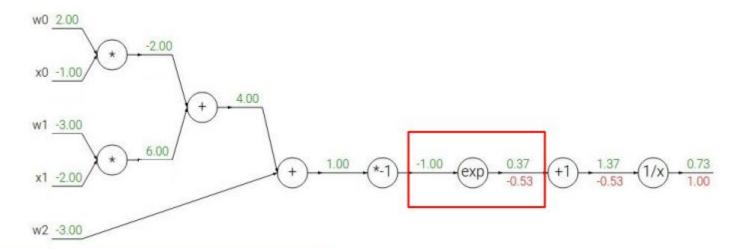
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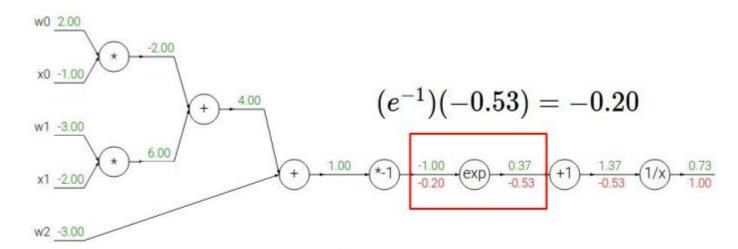


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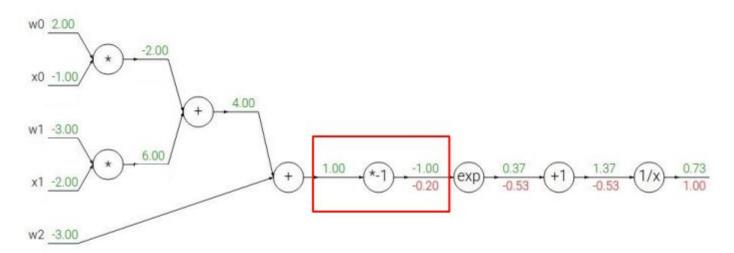


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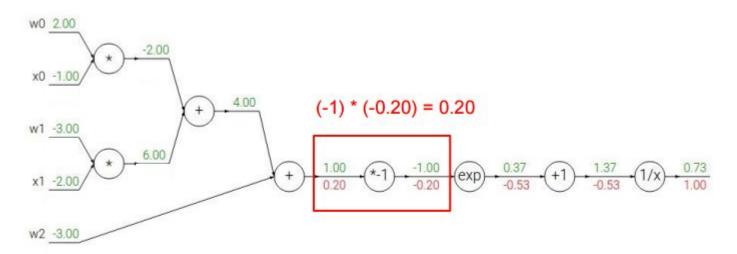


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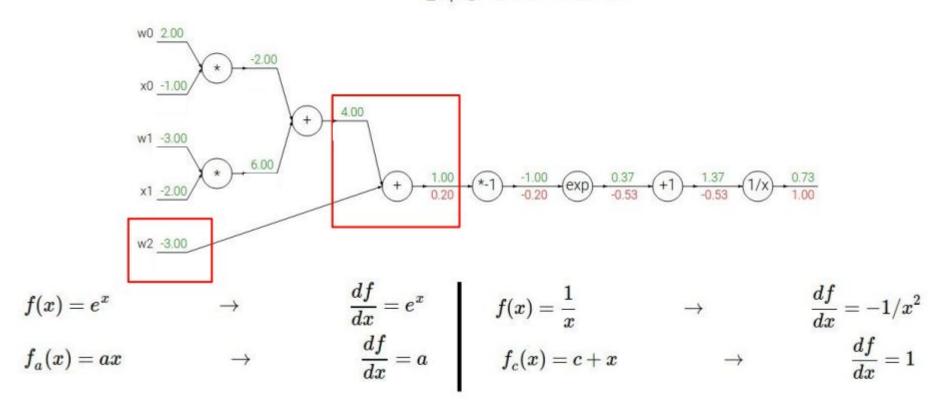


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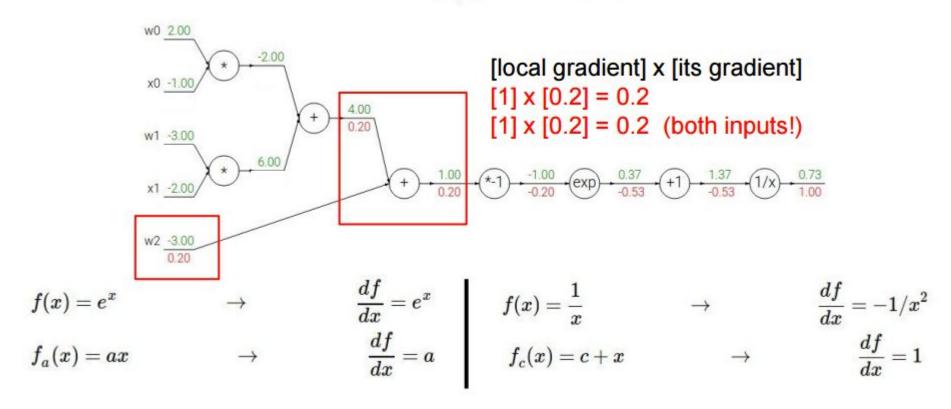


$$f(w,x) = rac{1}{1 + e^{-(w_0 x_0 + w_1 x_1 + w_2)}}$$



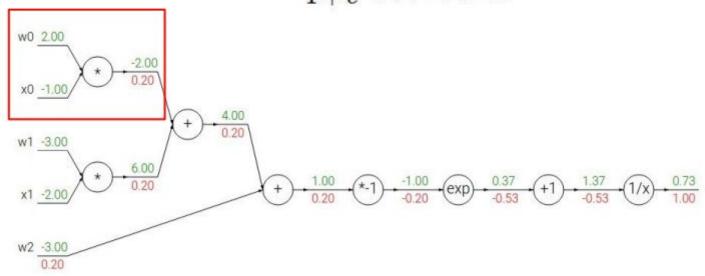


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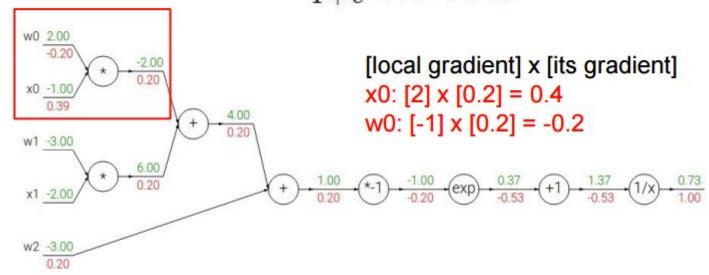


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$$f(w,x) = rac{1}{1 + e^{-(w_0 x_0 + w_1 x_1 + w_2)}}$$



$$f(x) = e^x$$

$$f_a(x) = ax$$

$$\frac{df}{dx} = e^x$$

$$\frac{df}{dx} = a$$

$$f(x) = \frac{1}{x}$$

$$f_c(x) = c + x$$

$$\rightarrow$$

$$egin{aligned} rac{df}{dx} = e^x & f(x) = rac{1}{x} &
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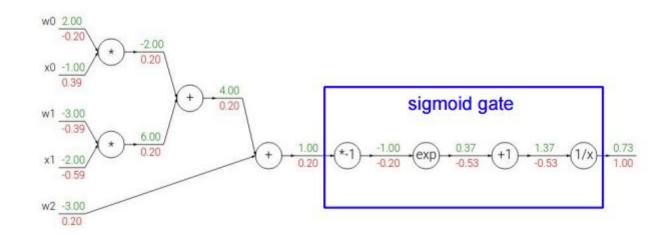
自动求导

$$f(w,x)=rac{1}{1+e^{-(w_0x_0+w_1x_1+w_2)}}$$

$$\sigma(x) = \frac{1}{1+e^{-x}}$$

sigmoid function

$$rac{d\sigma(x)}{dx} = rac{e^{-x}}{(1+e^{-x})^2} = \left(rac{1+e^{-x}-1}{1+e^{-x}}
ight) \left(rac{1}{1+e^{-x}}
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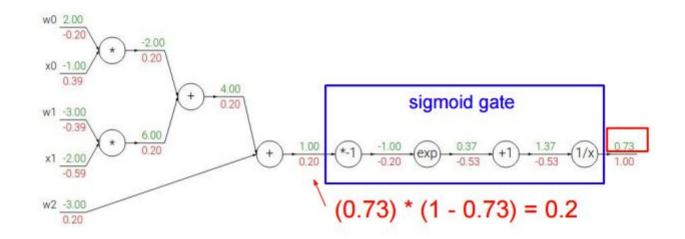
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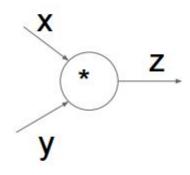


Graph (or Net) object. (Rough psuedo code)

```
class ComputationalGraph(object):
    #...

def forward(inputs):
    # 1. [pass inputs to input gates...]
    # 2. forward the computational graph:
    for gate in self.graph.nodes_topologically_sorted():
        gate.forward()
    return loss # the final gate in the graph outputs the loss

def backward():
    for gate in reversed(self.graph.nodes_topologically_sorted()):
        gate.backward() # little piece of backprop (chain rule applied)
    return inputs_gradients
```



```
class MultiplyGate(object):
    def forward(x,y):
        z = x*y
        self.x = x # must keep these around!
        self.y = y
        return z

def backward(dz):
    dx = self.y * dz # [dz/dx * dL/dz]
    dy = self.x * dz # [dz/dy * dL/dz]
    return [dx, dy]
```

TensorFlow的基本概念

- TensorFlow是什么?
 - 一个图计算平台
 - 它提供一些基本的操作(op)
 - 复杂的计算由这些基本的操作来表示
 - 自动梯度工具
- 基本概念
 - Graph Session op Variable ...

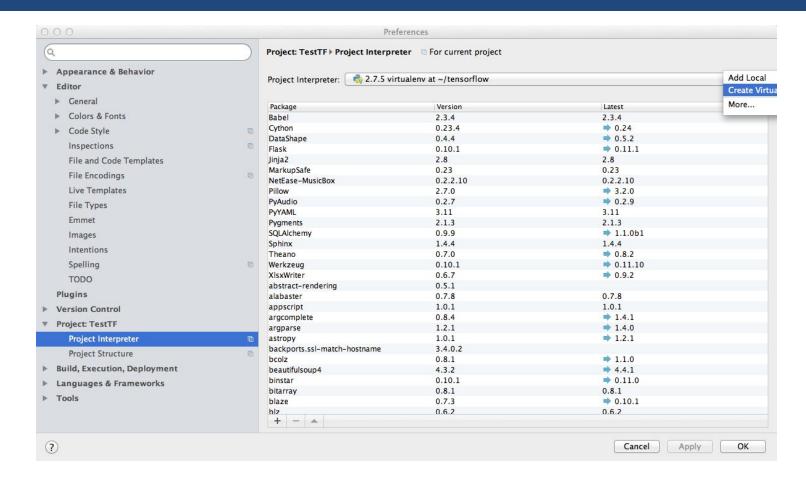


参考

- https://www.tensorflow.org/versions/r0.8/how_tos/a
 dding_an_op/index.html
- https://www.evernote.com/shard/s17/sh/f3938d7a-74c5-4d74-8ad8-d2effb6c49be/796d30bf0e06d086
- 现场示例
 - Mac os x & Ubuntu 14.04(not work in 12.04)



Using virtualenv installation in PyCharm

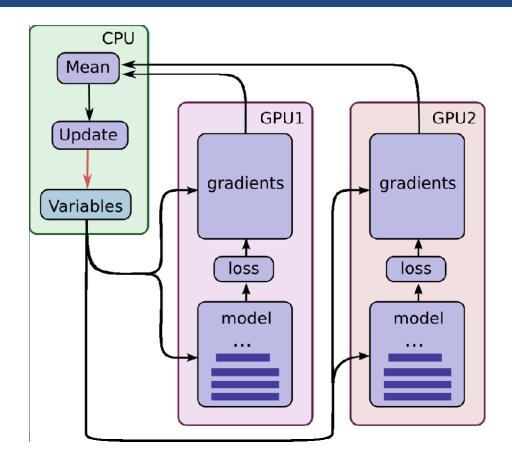


- 基本概念
- 自动求导

```
# Create 100 phony x, y data points in NumPy, y = x * 0.1 + 0.3
x_data = np.random.rand(100).astype(np.float32)
v data = x data * 0.1 + 0.3
# Try to find values for W and b that compute y data = W * x data + b
# (We know that W should be 0.1 and b 0.3, but Tensorflow will
# figure that out for us.)
W = tf.Variable(tf.random_uniform([1], -1.0, 1.0))
b = tf.Variable(tf.zeros([1]))
y = W * x data + b
# Minimize the mean squared errors.
loss = tf.reduce_mean(tf.square(y - y_data))
optimizer = tf.train.GradientDescentOptimizer(0.5)
train = optimizer.minimize(loss)
# Before starting, initialize the variables. We will 'run' this first.
init = tf.initialize all variables()
# Launch the graph.
sess = tf.Session()
sess.run(init)
# Fit the line.
for step in xrange(201):
    sess.run(train)
    if step % 20 == 0:
        print(step, sess.run(W), sess.run(b))
```

- InteractiveSession
- Computation Graph
- Placeholders
- Variables

- CPU vs GPU
- TensorBoard
- Prefetching Queue
- AlexNet Components



Define the Op's interface

Implement the kernel for the Op

Building the Op library

Using the Op in Python

GPU Support

```
REGISTER_OP("ZeroOut")
     .Attr("preserve_index: int = 0")
     .Input("to_zero: int32")
     .Output("zeroed: int32");
 using namespace tensorflow;
⇒ class ZeroOutOp : public OpKernel {
  public:
explicit ZeroOutOp(OpKernelConstruction* context) : OpKernel(context) {
   void Compute(OpKernelContext* context) override {
     // Grab the input tensor
     const Tensor& input_tensor = context->input(0);
     auto input = input_tensor.flat<int32>();
     // Create an output tensor
     Tensor* output_tensor = NULL;
     OP_REQUIRES_OK(context, context->allocate_output(0, input_tensor.shape(),
                                                      &output_tensor));
     auto output = output_tensor->template flat<int32>();
     // Set all but the first element of the output tensor to 0.
     const int N = input.size();
     for (int i = 0; i < N; i++) {
       output(i) = 0;
     output_flat(preserve_index_) = input(preserve_index_);
```

为什么用Python而不是C++? 为什么只能用基本的op来构造?

```
@ops.RegisterGradient("ZeroOut")
def _zero_out_grad(op, grad):
  """The gradients for `zero_out`.
  Args:
   op: The `zero_out` `Operation` that we are differentiating, which we can u
     to find the inputs and outputs of the original op.
    grad: Gradient with respect to the output of the 'zero_out' op.
  Returns:
    Gradients with respect to the input of 'zero_out'.
  11 11 11
 to_zero = op.inputs[0]
 shape = array_ops.shape(to_zero)
 index = array_ops.zeros_like(shape)
 first_grad = array_ops.reshape(grad, [-1])[0]
 to_zero_grad = sparse_ops.sparse_to_dense(index, shape, first_grad, 0)
 return [to_zero_grad] # List of one Tensor, since we have one input
```

示例——Tensor Serving

参考资料

- 1.1 Tensorflow的论文 【tensorflow的官方论文】
 http://download.tensorflow.org/paper/whitepaper2015.pdf
- 1.2 Overview of Caffe/Torch/Theano/TensorFlow【from cs231n,对深度学习感兴趣的同学推荐这个课程,作业设计的非常好,python notebook,每一步都能看到结果】 http://cs231n.stanford.edu/slides/winter1516_lecture12.pdf
- 1.3 Automatic differentiation in machine learning: a survey 【向机器学习社区的人介绍 Automatic differentiation】

http://arxiv.org/abs/1502.05767

1.4 Calculus on Computational Graphs: Backpropagation 【非常通俗易懂,如果觉得1.3 太长,看这个也差不多了,另外这个博客其它的文章也很好】 http://colah.github.io/posts/2015-08-Backprop/

1.5 Michael Nielsen的在线书的第二章 【非常详细的backprop推导,即使大学数学忘光了,也能看懂,初学的同学建议把这本书的每个公式都走一遍,python的代码跑一跑】

http://neuralnetworksanddeeplearning.com/chap2.html

参考资料

2.1 安装文档【目前最新的已经是0.9了,不过我之前测试的都是0.8,而且官方提供的whl也是0.8的】

https://www.tensorflow.org/versions/r0.8/get_started/os_setup.html

2.2 自己从源码build【0.8的whl安装了后跑gpu的例子会coredump,需要自己从源码编译】

https://www.evernote.com/shard/s17/sh/f3938d7a-74c5-4d74-8ad8-d2effb6c49be/796d30bf0e06d086

- 3. tutorial
 - 3.1 官方的tutorial,可以装了后在自己的机器上跑一跑 https://www.tensorflow.org/versions/r0.8/tutorials/index.html
- 3.2 CNN做情感分类,有tensorflow代码,第一篇是原理,第二篇是代码 http://www.wildml.com/2015/11/understanding-convolutional-neural-networksfor-nlp/

http://www.wildml.com/2015/12/implementing-a-cnn-for-text-classification-intensorflow/



- 3.3 一个不错的tensorflow的tutorial,包括常见的CNN,RNN等 https://github.com/aymericdamien/TensorFlow-Examples
- 3.5 CS224d的第七章介绍tensorflow 【这个课程也强烈推荐】 http://cs224d.stanford.edu/lectures/CS224d-Lecture7.pdf
- 3.6 Awesome TensorFlow 【名字很对,资源很丰富】 https://github.com/jtoy/awesome-tensorflow/
- 3.7 Awesome Tensorflow Implementations 【论文和tensorflow的实现,非常酷,之前的大部分是非常成熟的东西,比如CNN/LSTM,这里有最新的进展,如Sequence to Sequence -- chatbot,Show and Tell: A Neural Image Caption Generator,Using Deep Q-Network to Learn How To Play Flappy Bird,…】

 https://github.com/TensorFlowKR/awesome_tensorflow_implementations
- 3.8 google 的 deep learning course 【当然是用tensor flow了,很基础的课程】 https://cn.udacity.com/course/deep-learning--ud730/

谢谢!

出门问问 | 李理 lli@mobvoi.com