linear_regression

January 29, 2019

1 Learn Tensorflow via a linear regression model

In this excercise, you will learn tensorflow in steps. Please use python 3 and tensorflow >=1.4.0.

In [5]:
```
# A bit of setup

import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt

from implementations.a_tensorflow import regression_graph
from plot_graph import show_graph


%matplotlib inline
plt.rcParams['figure.figsize'] = (10.0, 8.0) # set default size of plots
plt.rcParams['image.interpolation'] = 'nearest'
plt.rcParams['image.cmap'] = 'gray'

# for auto-reloading external modules
# see http://stackoverflow.com/questions/1907993/autoreload-of-modules-in-ipython
%load_ext autoreload
%autoreload 2
```

```
The autoreload extension is already loaded. To reload it, use:
  %reload_ext autoreload
```

1.1 Tensorflow graph

In the first step, we test a tensorflow graph with all constants. Please read `a_tensorflow.regression_graph()`. It calculates the squared loss for the linear regression model.

1.1.1 Question 1:

Please run the code in the cell below, comment out the first line, and then run the cell multiple times. Why does the right-most graph get numbers on it nodes with your runs?

```
In [153]: tf.reset_default_graph()
          loss = regression_graph() # the first function in a_tensorflow
          show_graph(loss.graph)
```

```
<IPython.core.display.HTML object>
```

Note 1: use `tf.Session()` to run the graph and evaluate the value of a graph node. Note 2: retrieve a graph node by its name.

1.1.2 Question 2:

Please uncomment the last two lines and evaluate the loss.

```
In [154]: tf.reset_default_graph()
          loss = regression_graph(print_info=True)
          session = tf.Session()

          # retrieve the variable from the graph
          f = tf.get_default_graph().get_tensor_by_name('score:0')
          print("We retrieve f from the graph and show its value: ", session.run(f))

          #print('The loss is ')
          #session.run(loss)
```

```
Msg from the function: we can evaluate any value in the graph with tf.Session
Msg from the function: the value of f = w * x + b is 2.18
We retrieve f from the graph and show its value:  2.18
```

1.2 Tensorflow graph with vectors

We can get a tensor's shape with its member function `get_shape()`. Note that this function is python function, not a tensorflow operation, so it runs when you build the graph.

We can also use the tensorflow operation `tf.shape()` to get the shape of a tensor. Since it is an operation, it is part of the graph. You need to run the operation to get the actual value.

1.2.1 Question 3:

Complete the code in the function `regression_graph_vectorized` so the following cell can run.

```
In [82]: from implementations.a_tensorflow import regression_graph_vectorized
         tf.reset_default_graph()
         loss = regression_graph_vectorized(print_info=True)

         print(tf.Session().run(loss))
```

```
By w.get_shape(), we get the shape of the tensor w: (2,)
With tf.shape(w), we get the shape of w as a one-element tensor: Tensor("Shape:0", shape=(1,),
```

```
The shape of tf.matmul(x, w) is: (3, 1)
The shape of tf.squeeze(tf.matmul(x, w)) is: (3,)
35.343597
```

1.3 Place holder in Tensorflow graph

A `placeholder` holds space for the data that will be fed into the graph in the future. It can has shape `None` in one or more dimensions. The size of that dimension will be decided by the data.

1.3.1 Question 4:

Please implement the function `regression_graph_with_placeholder` and get the cell below run.

```
In [69]: from implementations.a_tensorflow import regression_graph_with_placeholder
         tf.reset_default_graph()

         # build a graph with place holders
         x = tf.placeholder(shape=[None, 2], dtype=tf.float32, name='feature')
         y = tf.placeholder(shape=[None], dtype=tf.float32, name='label')

         print('The place holder x is:', x) # take a look of the place holder

         loss = regression_graph_with_placeholder(x=x, y=y)

         # need to feed in actual values to x and y to evaluate the variable loss

         # numpy values
         np.random.seed(seed=2019)
         npx = np.random.random_sample([10, 2])
         npy = np.squeeze(npx.dot([[0.3], [0.6]])) + 1.0 + 0.05 * np.random.random_sample([10])

         # this line cannot run because you need to feed in values for place holders. Can you
         tf.Session().run(loss)

         #tf.Session().run(loss, feed_dict={x: npx, y: npy})


         # feed in a different set of data
         npx = np.random.random_sample([10, 2])
         npy = np.squeeze(npx.dot([[0.7], [0.9]])) + 1.0 + 0.05 * np.random.random_sample([10])

         # it get a different loss value
         #tf.Session().run(loss, feed_dict={x: npx, y: npy})
```

```
The place holder x is: Tensor("feature:0", shape=(?, 2), dtype=float32)
```

```
Out[69]: 14.469966
```

1.4 Tensorflow graph with variables

We need to use `tf.Variable` to declare variables that we can optimize later. We often use `Variables` for model parameters, which are mutable. We seldomly use `Variables` for data.

1.4.1 Question 5:

Please implement the function `regression_graph_with_variable` and get the cell below run.

```
In [114]: from implementations.a_tensorflow import regression_graph_with_variable
          tf.reset_default_graph()

          w = tf.Variable(np.full(shape=[2], fill_value=1.6, dtype=np.float32), name='weight')
          b = tf.Variable(tf.constant(0.9), name='bias')

          loss = regression_graph_with_variable(w=w, b=b)


          session = tf.Session()

          #init = tf.global_variables_initializer()
          #session.run(init)

          session.run(loss) # you cannot directly run loss because you need to initialize vari

Out[114]: 35.343597
```

1.5 Gradient calculation

Tensorflow can calculate gradient with respect to a `Variable` (but not a `constant` or a `placeholder`). Let's use an optimizer to optimize a gradient.

1.5.1 Question 6:

Please read the documentation of `compute_gradients` and `apply_gradients`. Can you update w and b ten times and make the loss smaller?

```
In [115]: # initialize an optimizer
          opt = tf.train.GradientDescentOptimizer(learning_rate=0.01)
          grads_vars = opt.compute_gradients(loss, var_list=[w, b])
          update = opt.apply_gradients(grads_vars)


          print('The current loss is: ', session.run(loss))
          print('The current variable gradients and values are')
          print(session.run(grads_vars))

          #for iter in range(10):  # can you run multiple steps to minimize the loss?
```

```
            session.run(update)

            print('The current loss is: ', session.run(loss))
            print('The current variable gradients and values are')
            print(session.run(grads_vars))
```

```
The current loss is:  35.343597
The current variable gradients and values are
[(array([ 9.504, 19.008], dtype=float32), array([1.6, 1.6], dtype=float32)), (35.64, 0.9)]
The current loss is:  20.22711
The current variable gradients and values are
[(array([ 7.2116356, 14.343437 ], dtype=float32), array([1.5049601, 1.40992  ], dtype=float32)
```

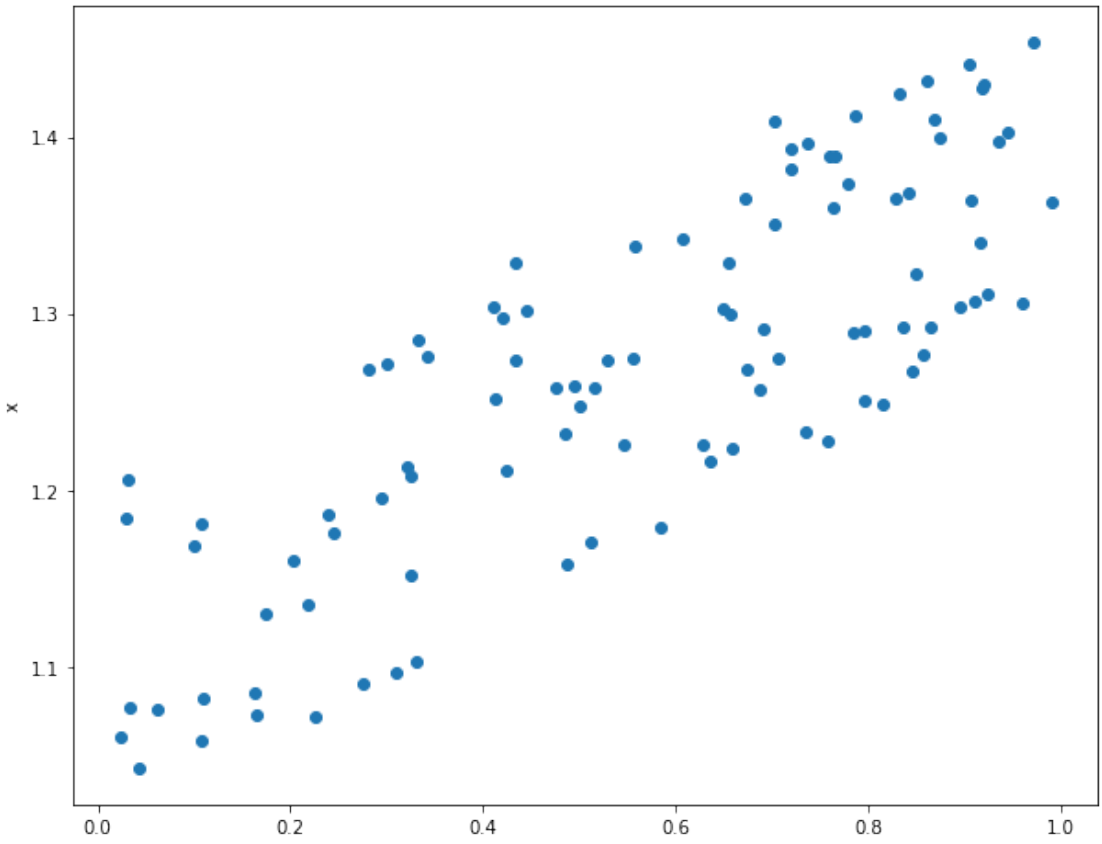
1.6 Compose the full version of linear regression

1.6.1 Question 7:

Can you use all techniques you have learned above to compose a full version of linear regression?

We use data with one feature only so we can plot the data easily. Assume the feature matrix is `npx` and the label is `npy`.

```
In [121]: # plot the data before fitting the model.
          # numpy values
          N = 100
          npx = np.random.random_sample([N, 1])
          npy = np.squeeze(npx.dot([[0.3]])) + 1.0 + 0.2 * np.random.random_sample([N])

          plt.plot(npx, npy, 'o')
          plt.ylabel('y')
          plt.ylabel('x')
          plt.show()
```



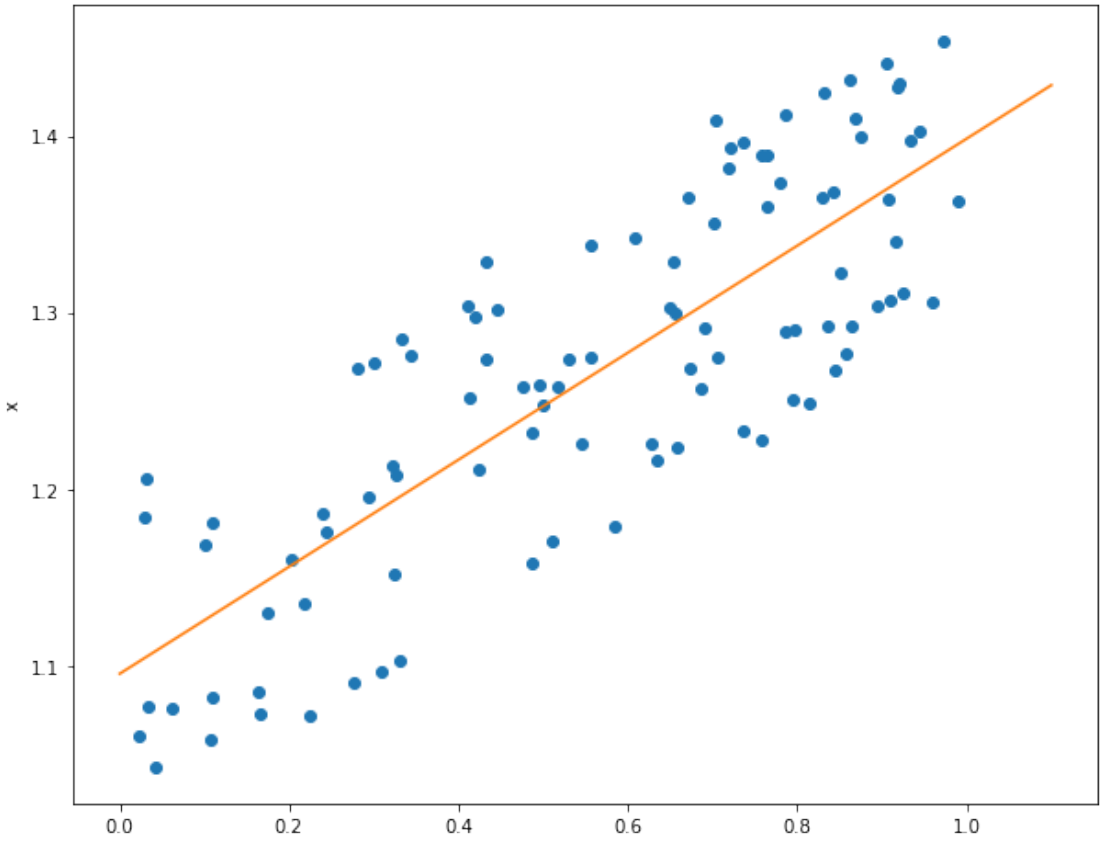
In [142]:
```
from implementations.a_tensorflow import regression_graph_full
tf.reset_default_graph()

#x = tf.placeholder(shape=??, dtype=tf.float32, name='feature')
#y = tf.placeholder(shape=??, dtype=tf.float32, name='label')

#w = tf.Variable(np.full(shape=?, fill_value=1.6, dtype=np.float32), name='weight')
#b = tf.Variable(tf.constant(0.9), name='bias')

loss = regression_graph_full(x=x, y=y, w=w, b=b) # you need to implement this functi

# get an optimizer
opt = tf.train.GradientDescentOptimizer(learning_rate=0.001)

# calculate the gradient with compute_gradients
#grads_vars = ??

# get an update operation with apply_gradients
#update = ??
```

```
        # get a session
        session = tf.Session()

        # initialize your variables, namely w and b
        #session.run(??)

        # get feeding dictionary
        feed_dict={x: npx, y: npy}


        print('The current loss is: ', session.run(loss, feed_dict=feed_dict))
        print('The current variable gradients and values are')
        print(session.run(grads_vars)) # something is missing here?

        for iter in range(1000):  # Run 1000 steps to minimize the loss

            # perform gradient steps by running the update operation
            #session.run(??)

            if iter % 100 == 1:
                print('Loss values become: ', session.run(loss, feed_dict=feed_dict))


        # obtain the trained model
        npw = session.run(w)
        npb = session.run(b)
```

```
The current loss is:  42.28681
The current variable gradients and values are
[(array([80.82121], dtype=float32), array([1.6], dtype=float32)), (107.27521, 0.9)]
Loss values become:  18.15213
Loss values become:  1.1406087
Loss values become:  0.43377265
Loss values become:  0.366474
Loss values become:  0.36006635
Loss values become:  0.35945624
Loss values become:  0.35939822
Loss values become:  0.35939267
Loss values become:  0.35939217
Loss values become:  0.35939205
```

In [147]:
```
# Plot the result

        x_line = np.arange(12) / 10.0
        y_line = x_line * np.squeeze(npw) + npb
```

```
plt.plot(npx, npy, 'o')
plt.plot(x_line, y_line)
plt.ylabel('y')
plt.ylabel('x')
plt.show()
```



In [ ]: