01_linear_regression_using_tensorflow_homework

September 25, 2020

Import

```
[2]: #import tensorflow as tf
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()

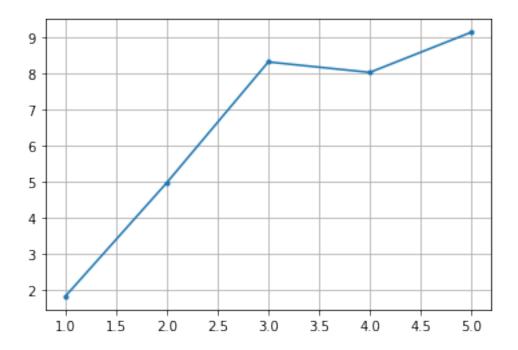
import numpy as np
import matplotlib.pyplot as plt
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/compat/v2_compat.py:96: disable_resource_variables (from tensorflow.python.ops.variable_scope) is deprecated and will be removed in a future version.

Instructions for updating:

non-resource variables are not supported in the long term

X and Y data



Initialization

```
[5]: useRandom = False
[6]: if useRandom:
    W = tf.Variable(tf.random_normal([1]), name='weight')
    b = tf.Variable(tf.random_normal([1]), name='bias')
else:
    w0 = 7.0;
    b0 = 5.0;

W = tf.Variable(w0*tf.ones([1]), name='weight')
    b = tf.Variable(b0*tf.ones([1]), name='bias')
```

Our hypothesis

$$H(x) = Wx + b$$

cost/loss function * loss of one training example :

$$loss = \mathcal{L}(\hat{y}, y) = (\hat{y}^{(i)} - y^{(i)})^2$$
(1)

[8]: loss = tf.reduce_mean(tf.square(hypothesis - y_train))

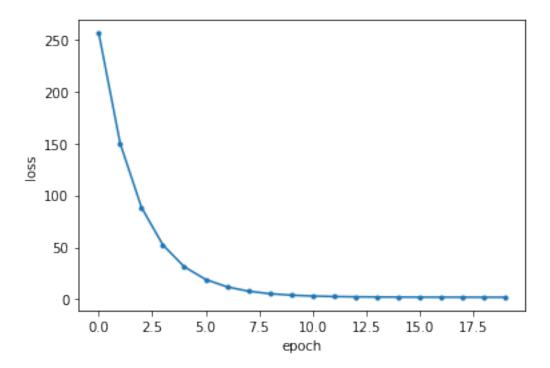
Optimizer

[9]: optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.01)
train = optimizer.minimize(loss)

Launch the graph in a session

```
[10]: sess = tf.Session()
       Initializes global variables in the graph.
[11]: sess.run(tf.global_variables_initializer())
[12]: nb epoch = 1001
     vloss = [] #empty list
     vb = [] #empty list
     vw = [] #empty list
     for step in range(nb_epoch):
         sess.run(train)
         loss1 = sess.run(loss)
         vloss.append(loss1)
         if step % 50 == 0: # 5
             w1 = sess.run(W)[0] #
             b1 = sess.run(b)[0] # bias
             print(step, '\t', loss1, '\t', w1, '\t', b1)
    0
              256.1197
                               5.61764
                                                4.609017
    50
              1.7335055
                               1.2467369
                                                3.0267956
              1.5506191
                               1.3274224
                                                2.735472
    100
    150
              1.4202728
                               1.3955445
                                                2.4895294
    200
              1.3273729
                               1.4530548
                                                2.2818992
    250
              1.2611611
                               1.5016066
                                                2.1066117
    300
              1.2139709
                               1.5425953
                                                1.9586297
    350
              1.180338
                               1.5771987
                                                1.8336997
    400
              1.1563662
                               1.6064122
                                                1.7282301
    450
              1.139282
                               1.6310749
                                                1.6391898
    500
              1.127105
                               1.6518958
                                                1.5640197
    550
              1.1184269
                               1.6694733
                                                1.5005594
    600
              1.1122416
                               1.6843129
                                                1.4469839
    650
              1.107833
                               1.6968408
                                                1.4017541
    700
              1.1046913
                               1.7074171
                                                1.3635705
    750
              1.102452
                               1.7163459
                                                1.3313347
    800
              1.1008556
                               1.7238839
                                                1.3041204
    850
              1.0997186
                               1.7302476
                                                1.2811451
    900
              1.0989077
                               1.7356201
                                                1.2617486
    950
                                                1.245374
              1.0983301
                               1.7401556
    1000
              1.0979183
                               1.7439846
                                                1.2315502
[13]: plt.plot(vloss[:20],'.-')
     plt.xlabel('epoch')
     plt.ylabel('loss')
```

[13]: Text(0, 0.5, 'loss')



```
[14]: w1 = sess.run(W)[0] #
b1 = sess.run(b)[0] # bias

[15]: print(w1, b1)

1.7439846 1.2315502

[16]: str1 = 'y = ' + str(w1) +'x + ' + str(b1)
print(str1)

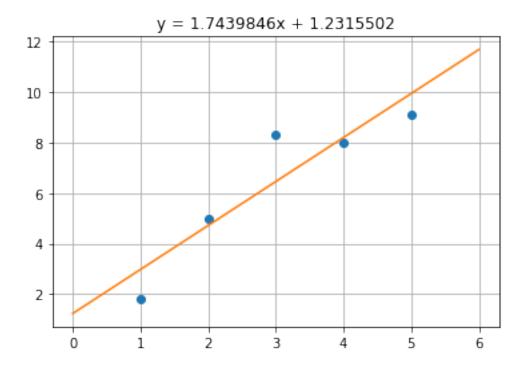
y = 1.7439846x + 1.2315502

[17]: print(w1, b1)
str1 = 'y = ' + str(w1) +'x + ' + str(b1)
print(str1)

1.7439846 1.2315502
y = 1.7439846x + 1.2315502

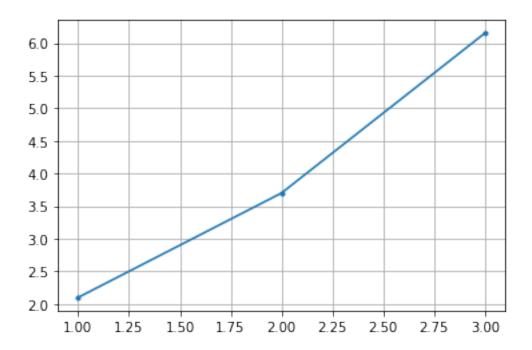
[18]: plt.figure(figsize=(6,4)) # figsize
plt.plot(x_train, y_train, 'o') #train data
```

[18]: Text(0.5, 1.0, 'y = 1.7439846x + 1.2315502')



**

```
[19]: x_train = [1, 2, 3]
y_train = [2+0.1, 4-0.3, 6+0.15] # noise
[20]: plt.plot(x_train, y_train,'.-')
plt.grid()
```

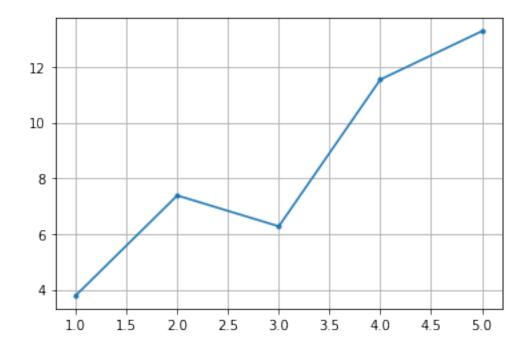


```
[21]: x_train = [1, 2, 3, 4, 5]
y_train = [2+2, 4+2, 6+2, 8+2, 10+2] #y=2x+2

signal_length = len(x_train)
y_noise = np.random.normal(0, 1, signal_length)

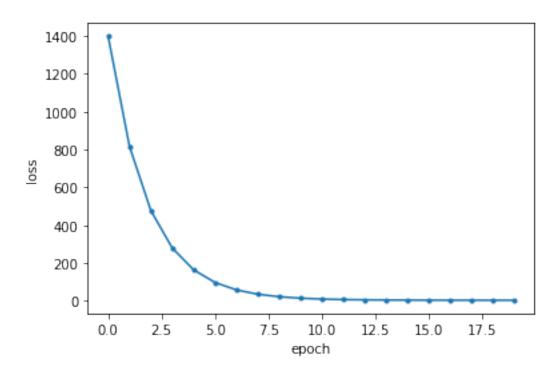
y_train = y_train + y_noise

[22]: plt.plot(x_train, y_train,'.-')
plt.grid()
```



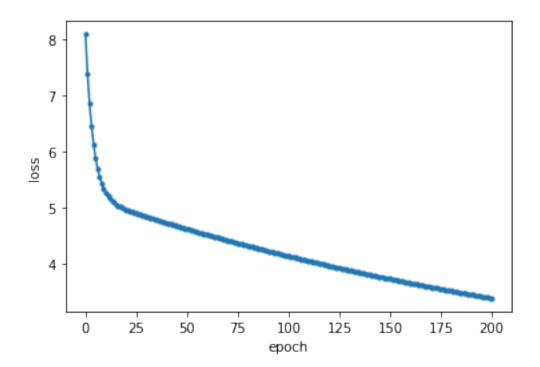
```
[23]: w0 = 15.0;
     b0 = 9.0;
     W = tf.Variable(w0*tf.ones([1]), name='weight')
     b = tf.Variable(b0*tf.ones([1]), name='bias')
[24]: hypothesis = x_{train} * W + b
     loss = tf.reduce_mean(tf.square(hypothesis - y_train))
     optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.01)
     train = optimizer.minimize(loss)
     sess = tf.Session()
     sess.run(tf.global_variables_initializer())
[25]: nb_epoch = 201
     vloss =[]
     vb = []
     vw = []
     for step in range(nb_epoch):
         sess.run(train)
         loss1 = sess.run(loss)
         vloss.append(loss1)
         w1 = sess.run(W)[0] #
         vw.append(w1)
         b1 = sess.run(b)[0] # bias
         vb.append(b1)
         if step % 10 == 0: # 200
```

```
print(step, '\t' ,loss1, '\t' ,w1, '\t', b1)
    0
              1398.9922
                               11.7596245
                                                8.08905
    10
              9.98214
                               2.0315993
                                                5.2628326
    20
              3.5587227
                               1.4082865
                                                4.9628606
    30
              3.3899102
                               1.3958577
                                                4.8363333
    40
                               1.4235034
                                                4.725004
              3.258503
    50
              3.1358213
                               1.4528939
                                                4.618121
    60
              3.0211744
                               1.4814848
                                                4.514847
    70
              2.9140356
                                                4.4150143
                               1.5091357
    80
              2.8139138
                               1.5358665
                                                4.3185077
    90
              2.7203498
                               1.5617073
                                                4.2252145
              2.6329129
    100
                               1.5866876
                                                4.1350274
    110
              2.5512025
                               1.6108363
                                                4.047844
    120
              2.4748433
                               1.6341804
                                                3.963564
    130
              2.4034848
                               1.6567472
                                                3.88209
              2.3367999
                               1.6785628
                                                3.8033292
    140
    150
              2.2744815
                               1.6996518
                                                3.727191
    160
              2.2162447
                               1.7200385
                                                3.6535883
    170
              2.1618223
                               1.7397466
                                                3.5824366
    180
              2.1109633
                               1.7587981
                                                3.5136542
    190
              2.0634356
                               1.7772154
                                                3.4471622
    200
              2.0190206
                               1.7950191
                                                3.3828845
[26]: plt.plot(vloss[:20],'.-')
     plt.xlabel('epoch')
     plt.ylabel('loss')
[26]: Text(0, 0.5, 'loss')
```



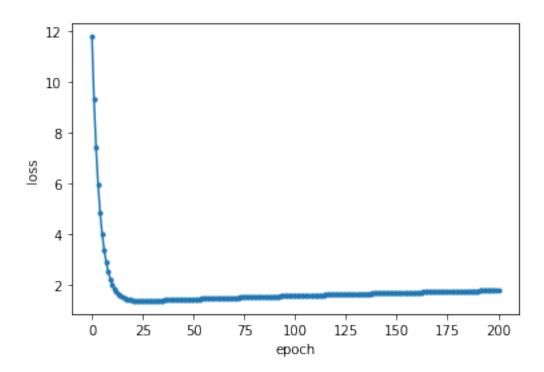
```
[33]: plt.plot(vb,'.-')
plt.xlabel('epoch')
plt.ylabel('loss')
```

[33]: Text(0, 0.5, 'loss')



```
[34]: plt.plot(vw,'.-')
plt.xlabel('epoch')
plt.ylabel('loss')
```

[34]: Text(0, 0.5, 'loss')



```
[29]: w1 = sess.run(W)[0] #
b1 = sess.run(b)[0] # bias

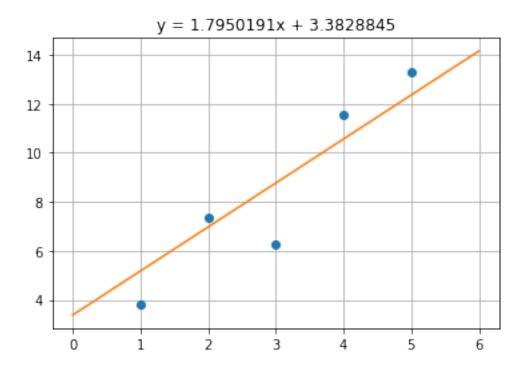
print(w1, b1)
str1 = 'y = ' + str(w1) +'x + ' + str(b1)
print(str1)
```

 $1.7950191 \ 3.3828845$ y = 1.7950191x + 3.3828845

```
[30]: plt.figure(figsize=(6,4)) # figsize
plt.plot(x_train, y_train,'o') #train data

#
# # x
x1 = np.linspace(np.min(x_train)-1, np.max(x_train)+1)
y1 = w1*x1 + b1
```

[30]: Text(0.5, 1.0, 'y = 1.7950191x + 3.3828845')



[30]: