01_linear_regression_using_tensorflow

September 28, 2020

Linear regression, TensorFlow notebook.

Import

```
[1]: # 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 (given)

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

y_train = [3, 5, 7, 9, 11] # y = 2x + 1

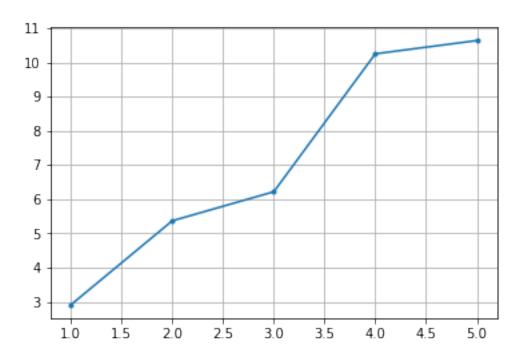
signal_length = len(x_train)
y_noise = np.random.normal(0,1, signal_length)

y_train = y_train + y_noise

# y_train = [2+0.1, 4-0.3, 6+0.15] # noise

# y_train = [2, 4, 6] # x_train 2
# y_train = [3, 5, 7]

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



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 XW+b

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

cost/loss function

• loss of one training example:

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

Optimizer

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

Launch the graph in a session

```
[10]: sess = tf.Session()
```

Initalizes global variables in the graph.

```
[11]: sess.run(tf.global_variables_initializer())
[13]: nb_epoch = 2001
vloss = [] # empty list
```

```
place = 2001
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)
    w1 = sess.run(W)
    vw.append(w1)
    b1 = sess.run(b)
    vb.append(b1)

if step % 50 == 0: # 100
    w1 = sess.run(W)[0] #
    b1 = sess.run(b)[0] # bias

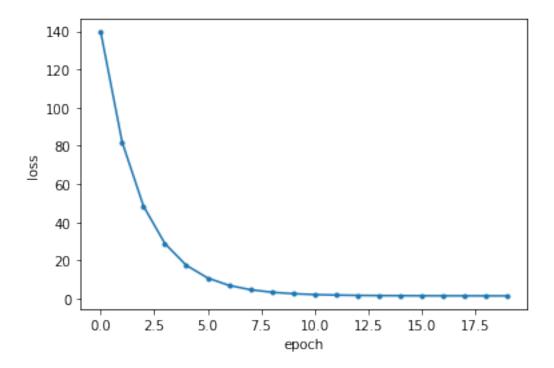
print(step, '\t', loss1, '\t', w1, '\t', b1)
```

0	139.53317	4.648285	4.3307257
50	1.2574391	1.4628112	3.0384212
100	1.0347639	1.5518444	2.7169657
150	0.8760597	1.6270123	2.445586
200	0.76294816	1.6904709	2.2164798
250	0.6823312	1.7440447	2.023062
300	0.6248741	1.7892729	1.8597734
350	0.58392346	1.8274558	1.721921
400	0.5547372	1.8596907	1.6055429
450	0.53393567	1.8869042	1.5072932
500	0.51910996	1.9098786	1.4243482
550	0.50854343	1.9292743	1.3543236
600	0.5010125	1.9456488	1.295207
650	0.4956451	1.9594723	1.2452993
700	0.49181947	1.9711426	1.2031658
750	0.489093	1.980995	1.1675959
800	0.48714995	1.9893126	1.1375667
850	0.4857649	1.9963346	1.1122152
900	0.48477775	2.0022626	1.0908133

```
950
         0.48407435
                          2.0072672
                                           1.0727447
1000
         0.4835728
                          2.0114923
                                           1.0574908
1050
         0.48321557
                          2.0150595
                                           1.0446132
1100
         0.48296076
                          2.0180707
                                           1.0337412
         0.48277935
1150
                          2.0206127
                                           1.0245632
1200
         0.48264986
                          2.022759
                                           1.0168151
1250
         0.48255762
                          2.024571
                                           1.0102735
1300
         0.48249203
                          2.0261004
                                           1.0047512
1350
         0.48244524
                          2.0273917
                                           1.0000889
1400
                                           0.996153
         0.48241192
                          2.028482
         0.48238793
                          2.0294023
                                           0.9928301
1450
         0.4823711
                          2.0301793
                                           0.99002516
1500
         0.482359
1550
                          2.0308352
                                           0.98765707
         0.4823505
                                           0.985658
1600
                          2.0313888
1650
         0.48234433
                          2.031856
                                           0.9839705
1700
         0.48233995
                          2.032251
                                           0.9825457
1750
         0.4823367
                          2.032584
                                           0.9813428
1800
         0.48233452
                          2.0328653
                                           0.98032737
1850
         0.4823329
                          2.0331028
                                           0.9794701
1900
         0.4823318
                          2.0333033
                                           0.9787463
1950
         0.48233098
                          2.0334723
                                           0.97813535
2000
         0.4823304
                                           0.9776195
                          2.033615
```

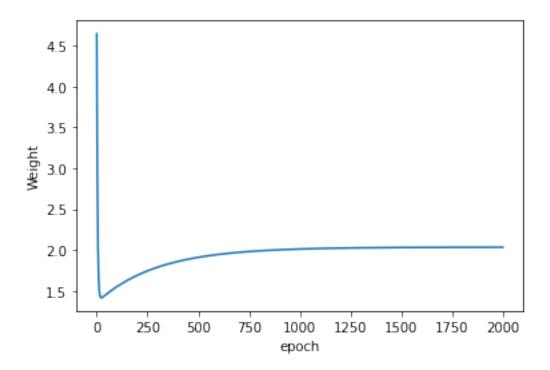
```
[14]: plt.plot(vloss[:20], '.-')
plt.xlabel('epoch')
plt.ylabel('loss')
```

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



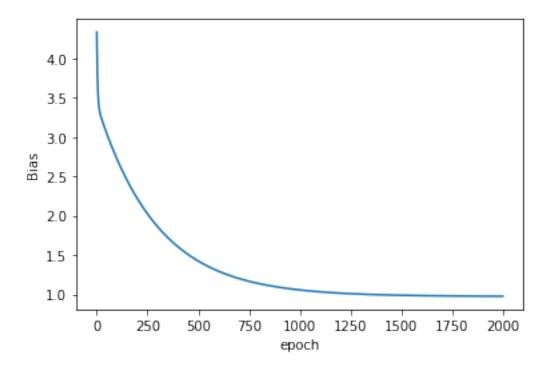
```
[15]: # W epoch
plt.plot(vw[:], '-')
plt.xlabel('epoch')
plt.ylabel('Weight')
```

[15]: Text(0, 0.5, 'Weight')



```
[18]: # b epoch
plt.plot(vb[:], '-')
plt.xlabel('epoch')
plt.ylabel('Bias')
```

[18]: Text(0, 0.5, 'Bias')



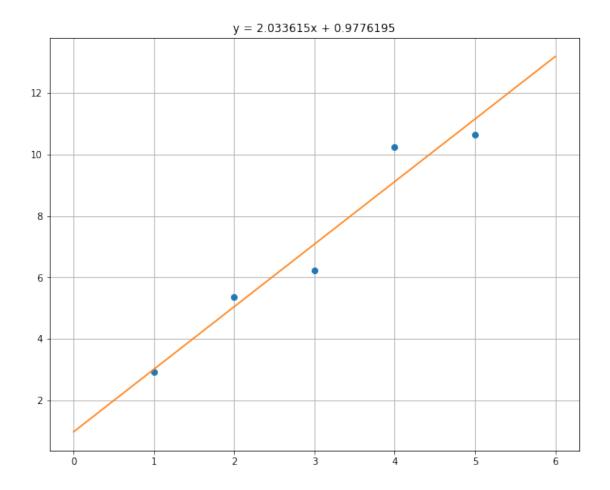
```
[19]: w1 = sess.run(W)[0] # b1 = sess.run(b)[0] # bias
```

```
[20]: print(w1, b1)

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

 $2.033615 \ 0.9776195$ y = 2.033615x + 0.9776195

[21]: Text(0.5, 1.0, 'y = 2.033615x + 0.9776195')



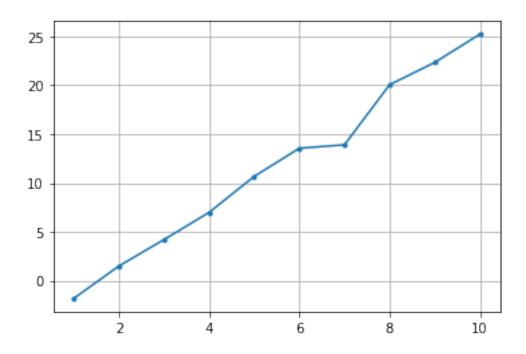
```
1
[22]: # X and Y data
x_train = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]

y_train = [-2, 1, 4, 7, 10, 13, 16, 19, 22, 25] # y = 3x - 5

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

y_train = y_train + y_noise

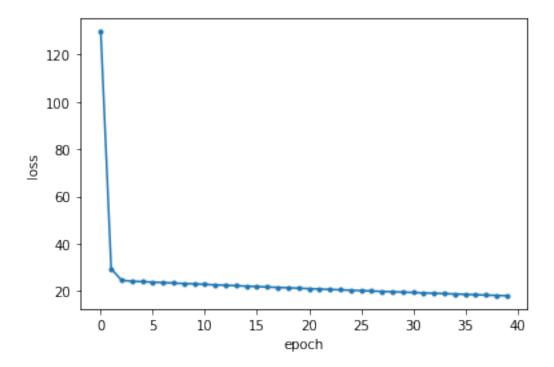
plt.plot(x_train, y_train,'.-')
plt.grid()
```



```
[23]: # initialization
     useRandom = False
     if useRandom:
        W = tf.Variable(tf.random_normal([1]), name='weight')
        b = tf.Variable(tf.random_normal([1]), name = 'bias')
     else:
        w0 = 9.0;
        b0 = 7.0;
        W = tf.Variable(w0*tf.ones([1]), name='weight')
        b = tf.Variable(b0*tf.ones([1]), name='bias')
[24]: # Our hypothesis XW+b
     hypothesis = x_train * W + b
     # loss function
     loss = tf.reduce_mean(tf.square(hypothesis - y_train))
     # Optimizer
     optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.01)
     train = optimizer.minimize(loss)
[25]: # Launch the graph in a session
     sess = tf.Session()
     # Initializes global variables in the graph
```

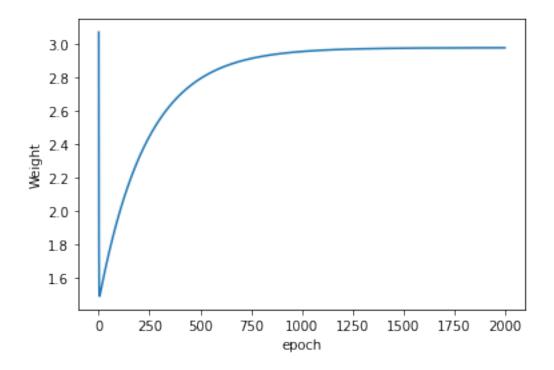
```
sess.run(tf.global_variables_initializer())
[26]: nb_epoch = 2001
     vloss = [] # empty list
     vb = [] # empty list
     vw = [] # empty list
     for step in range(nb_epoch):
         sess.run(train)
         loss1 = sess.run(loss)
         vloss.append(loss1)
         w1 = sess.run(W)
         vw.append(w1)
         b1 = sess.run(b)
         vb.append(b1)
         if step % 100 == 0: # 100
           w1 = sess.run(W)[0] #
           b1 = sess.run(b)[0] # bias
           print(step, '\t', loss1, '\t', w1, '\t', b1)
    0
             129.6079
                              3.0743654
                                               6.1031837
    100
             11.036634
                              1.9807149
                                               2.2274022
             5.123662
    200
                              2.3242965
                                               -0.16455165
    300
             2.575247
                              2.5498564
                                               -1.7348604
    400
             1.4769113
                              2.6979358
                                               -2.7657626
    500
             1.0035429
                              2.7951493
                                               -3.4425457
    600
             0.79952586
                              2.8589697
                                               -3.8868515
    700
                                               -4.1785364
             0.71159804
                              2.9008675
    800
             0.673702
                                               -4.370027
                              2.9283733
    900
             0.65736955
                              2.946431
                                               -4.4957404
    1000
             0.65032965
                              2.9582858
                                               -4.5782723
    1100
             0.6472961
                              2.966068
                                               -4.632451
                              2.971177
                                               -4.66802
    1200
             0.64598924
    1300
             0.6454252
                              2.9745312
                                               -4.6913695
    1400
             0.6451827
                              2.976733
                                               -4.706699
    1500
             0.64507794
                                               -4.7167616
                              2.9781785
    1600
             0.6450325
                              2.9791276
                                               -4.72337
    1700
             0.64501286
                              2.9797509
                                               -4.7277083
    1800
             0.64500475
                              2.9801598
                                               -4.7305555
    1900
             0.64500105
                              2.9804285
                                               -4.7324266
    2000
                                               -4.7336535
             0.64499986
                              2.9806046
[28]: plt.plot(vloss[:40], '.-')
     plt.xlabel('epoch')
     plt.ylabel('loss')
```

```
[28]: Text(0, 0.5, 'loss')
```



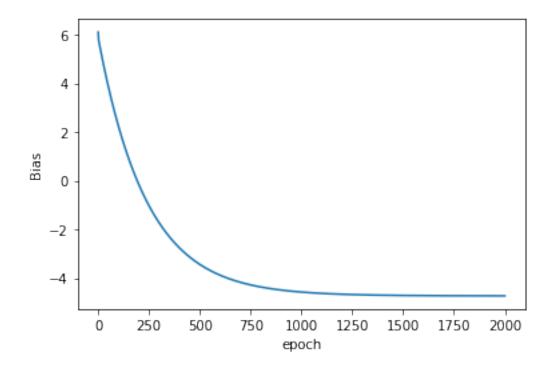
```
[29]: # W epoch
plt.plot(vw[:], '-')
plt.xlabel('epoch')
plt.ylabel('Weight')
```

[29]: Text(0, 0.5, 'Weight')

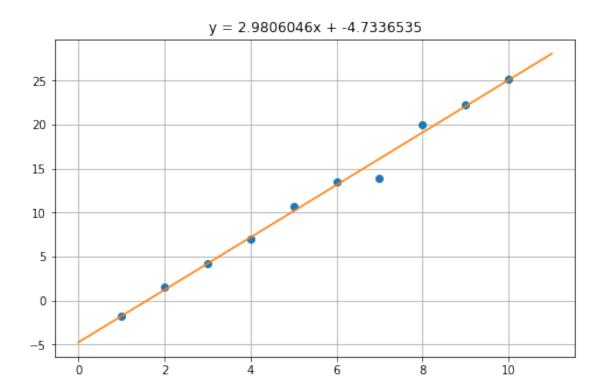


```
[30]: # b epoch
plt.plot(vb[:], '-')
plt.xlabel('epoch')
plt.ylabel('Bias')
```

[30]: Text(0, 0.5, 'Bias')



```
[31]: #
     w1 = sess.run(W)[0] #
    b1 = sess.run(b)[0] # bias
[32]: #
     print(w1, b1)
     str1 = 'y = ' + str(w1) + 'x + ' + str(b1)
    print(str1)
    2.9806046 -4.7336535
    y = 2.9806046x + -4.7336535
[33]: plt.figure(figsize=(8,5))
     plt.plot(x_train, y_train, 'o') # train_data
     #
     x1 = np.linspace(np.min(x_train)-1, np.max(x_train)+1)
     y1 = w1*x1 + b1
    plt.plot(x1, y1)
     plt.grid() #
     \# plt.axis((np.min(x_train) - 1, np.max(x_train) + 1, np.min(y_train) - 1,
     \# np.max(y\_train) + 1))
     plt.title(str1)
[33]: Text(0.5, 1.0, 'y = 2.9806046x + -4.7336535')
```



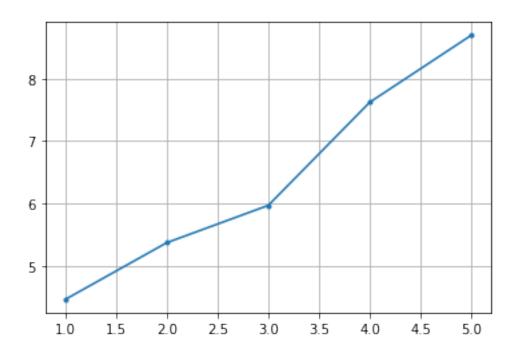
```
[40]: # X and Y data
x_train = [1, 2, 3, 4, 5]

y_train = [4.2, 5.4, 6.6, 7.8, 9.0] # y = 1.2x + 3

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

y_train = y_train + y_noise

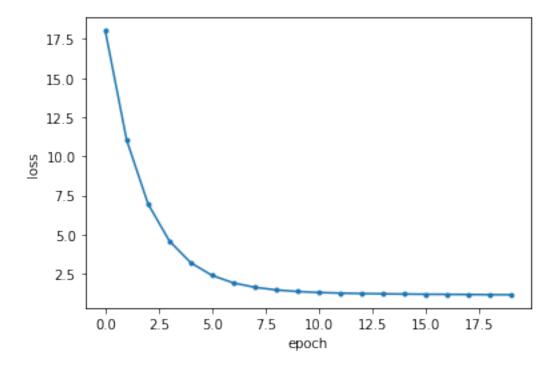
plt.plot(x_train, y_train,'.-')
plt.grid()
```



```
[41]: # initialization
     useRandom = True
     if useRandom:
        W = tf.Variable(tf.random_normal([1]), name='weight')
        b = tf.Variable(tf.random_normal([1]), name = 'bias')
     else:
        w0 = 5.0;
        b0 = 3.0;
        W = tf.Variable(w0*tf.ones([1]), name='weight')
        b = tf.Variable(b0*tf.ones([1]), name='bias')
[42]: # Our hypothesis XW+b
     hypothesis = x_train * W + b
     # loss function
     loss = tf.reduce_mean(tf.square(hypothesis - y_train))
     # Optimizer
     optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.01)
     train = optimizer.minimize(loss)
[43]: # Launch the graph in a session
     sess = tf.Session()
     # Initializes global variables in the graph
```

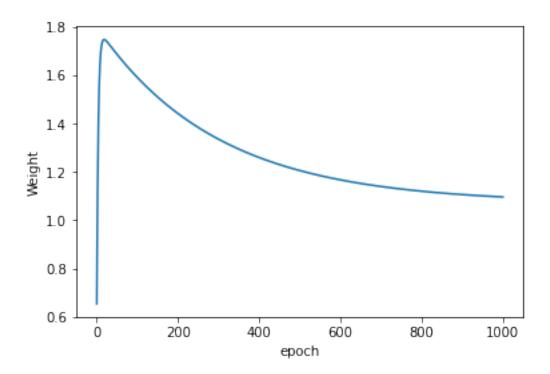
```
sess.run(tf.global_variables_initializer())
[44]: nb_epoch = 1001
     vloss = [] # empty list
     vb = [] # empty list
     vw = [] # empty list
     for step in range(nb_epoch):
         sess.run(train)
         loss1 = sess.run(loss)
         vloss.append(loss1)
         w1 = sess.run(W)
         vw.append(w1)
         b1 = sess.run(b)
         vb.append(b1)
         if step % 50 == 0: # 50
           w1 = sess.run(W)[0] #
           b1 = sess.run(b)[0] # bias
           print(step, '\t', loss1, '\t', w1, '\t', b1)
    0
             18.025177
                              0.6545198
                                               0.2635625
    50
             0.9510582
                              1.6865096
                                               0.99031955
                                               1.3360273
    100
             0.69351435
                              1.5907558
    150
             0.50995827
                              1.5099164
                                               1.6278827
             0.3791346
                                               1.8742745
    200
                              1.44167
    250
             0.28589422
                              1.3840544
                                               2.082285
    300
             0.21944025
                              1.3354139
                                               2.2578924
    350
             0.17207697
                              1.29435
                                               2.4061456
    400
             0.13832036
                              1.259683
                                               2.5313048
    450
             0.11426141
                              1.2304162
                                               2.6369677
    500
             0.09711422
                              1.2057081
                                               2.726171
    550
             0.08489314
                              1.1848494
                                               2.8014784
             0.076182835
                                               2.865055
    600
                              1.1672395
    650
             0.06997497
                              1.1523731
                                               2.9187279
    700
             0.065550424
                              1.1398224
                                               2.96404
    750
             0.062397044
                              1.1292267
                                               3.0022936
    800
             0.060149513
                              1.1202816
                                               3.034588
    850
             0.058547687
                              1.1127298
                                               3.0618525
    900
             0.05740603
                              1.1063545
                                               3.0848694
    950
             0.056592345
                              1.1009722
                                               3.1043012
    1000
             0.056012414
                              1.0964284
                                               3.1207058
[45]: plt.plot(vloss[:20], '.-')
     plt.xlabel('epoch')
     plt.ylabel('loss')
```

```
[45]: Text(0, 0.5, 'loss')
```

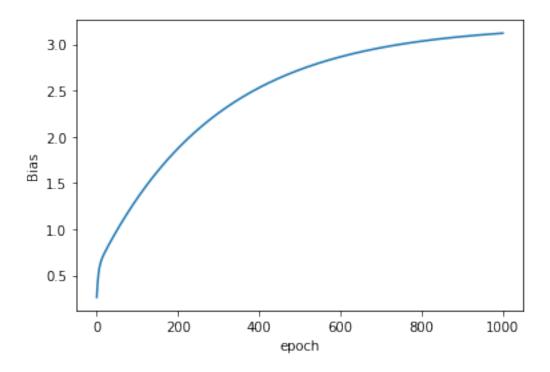


```
[46]: # W epoch
plt.plot(vw[:], '-')
plt.xlabel('epoch')
plt.ylabel('Weight')
```

[46]: Text(0, 0.5, 'Weight')

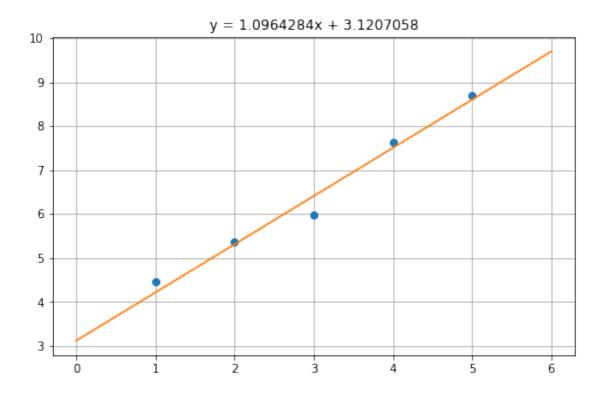


```
[47]: # b epoch
plt.plot(vb[:], '-')
plt.xlabel('epoch')
plt.ylabel('Bias')
[47]: Text(0, 0.5, 'Bias')
```



```
[48]: #
     w1 = sess.run(W)[0] #
     b1 = sess.run(b)[0] # bias
[49]: #
     print(w1, b1)
     str1 = 'y = ' + str(w1) + 'x + ' + str(b1)
     print(str1)
    1.0964284 3.1207058
    y = 1.0964284x + 3.1207058
[50]: plt.figure(figsize=(8,5))
     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
     plt.plot(x1, y1)
     plt.grid() #
     \# plt.axis((np.min(x_train) - 1, np.max(x_train) + 1, np.min(y_train) - 1,
     \# np.max(y\_train) + 1))
     plt.title(str1)
```

[50]: Text(0.5, 1.0, 'y = 1.0964284x + 3.1207058')



[]: