

Prof. Dr. L. Paditz, 19.03.2019

HTW Dresden,

paditz@htw-dresden.de

Skript1:

**A linear regression learning algorithm example
using TensorFlow library.**

Skript2: Wiederholung

**Basic Operations example using TensorFlow
library.**



Examples:

<https://databricks.com/tensorflow/examples>

Placeholders:

<https://databricks.com/tensorflow/placeholders>

Variables:

<https://databricks.com/tensorflow/variables>


TensorFlow Randomness:

<https://databricks.com/tensorflow/tensorflow-randomness>

Visualisation with TensorBoard:

<https://databricks.com/tensorflow/visualisation>

Dimensionality and Broadcasting:

<https://databricks.com/tensorflow/dimensionality-and-broadcasting> 

Broadcasting semantics:

<https://www.tensorflow.org/xla/broadcasting>

Skript1:

mit ClassPad:

```
train_X:={3.3, 4.4, 5.5, 6.71, 6.93, 4.168, 9.779, 6.182, 7.5}
train_Y:={1.7, 2.76, 2.09, 3.19, 1.694, 1.573, 3.366, 2.596}
dim(ans)
```

17

LinearReg train_X, train_Y, 1, y1, On

done

DispStat

done

=====
Lineare Regression(Train)

$y=a \cdot x+b$

a = 0.2516349

b = 0.7988012

r = 0.8323918

$r^2 = 0.692876$

MSe = 0.174372
=====

Tensorflow: verschiedene **training_epochs** und **learning_rate**

Training cost= 0.084093794 W= 0.2968933 b= 0.4611957

Training cost= 0.081328355 W= 0.23506787 b= 0.8165899

Training cost= 0.07699074 W= 0.24960834 b=


0.80136055

MSerr=0.07699074*(2*17)/15


MSerr=0.174512344

stop

STAT-Menü 

test_X:={6.83, 4.668, 8.9, 7.91, 5.7, 8.7, 3.1} 

{6.83, 4.668, 8.9, 7.91, 5.7, 8.7, 3.1, 2.1}

test_Y:={1.84, 2.273, 3.2, 2.831, 2.92, 3.24, 

{1.84, 2.273, 3.2, 2.831, 2.92, 3.24, 1.35, 1.03}

dim(ans)

8

LinearReg test_X, test_Y, 1, y1, On

done

DispStat

done

=====

Lineare Regression(Test)

$y=a \cdot x+b$

a = 0.293093

b = 0.5803128

r = 0.8775618

$r^2 = 0.7701147$

MSe = 0.1945445

=====

STAT-Menü 

stop

Tensorflow:

Skript1 mit zufälligen Startwerten für W und b:

```

python3
'''
A linear regression learning algorithm example using
TensorFlow library.
Author: Aymeric Damien
Project:
https://github.com/aymericdamien/TensorFlow-Examples/
'''

import tensorflow as tf
import numpy
import matplotlib.pyplot as plt
rng = numpy.random

# Parameters
# learning_rate = 0.01 bzw. 0.1
learning_rate = 0.01
# training_epochs = 1000
training_epochs = 10000
# display_step = 50
display_step = 500

# Training Data
train_X =
numpy.asarray([3.3, 4.4, 5.5, 6.71, 6.93, 4.168, 9.7▶
train_Y =
numpy.asarray([1.7, 2.76, 2.09, 3.19, 1.694, 1.573,▶
n_samples = train_X.shape[0]

```

```

# Testing example, as requested (Issue #2)
test_X = numpy.asarray([6.83, 4.668, 8.9, 7.91,
5.7, 8.7, 3.1, 2.1])
test_Y = numpy.asarray([1.84, 2.273, 3.2, 2.831,
2.92, 3.24, 1.35, 1.03])

# tf Graph Input
X = tf.placeholder("float")
Y = tf.placeholder("float")

# Set model weights
W = tf.Variable(rng.randn(), name="weight")
b = tf.Variable(rng.randn(), name="bias")

# Construct a linear model
pred = tf.add(tf.multiply(X, W), b)

# Mean squared error
cost = tf.reduce_sum(tf.pow(pred-Y,
2))/(2*n_samples)
# Gradient descent
# Note, minimize() knows to modify W and b
because Variable objects are trainable=True by default
optimizer =
tf.train.GradientDescentOptimizer(learning_rate).minimize

# Initialize the variables (i.e. assign their default
value)
init = tf.global_variables_initializer()

```

```

# Start training
with tf.Session() as sess:
    # Run the initializer
    sess.run(init)
    # Fit all training data
    for epoch in range(training_epochs):
        for (x, y) in zip(train_X, train_Y):
            sess.run(optimizer, feed_dict={X: x, Y:
y})

            # Display logs per epoch step
            if (epoch+1) % display_step == 0:
                c = sess.run(cost, feed_dict={X:
train_X, Y:train_Y})
                print("Epoch:", '%04d' % (epoch+1),
"cost=", "{:.9f}".format(c), \
                    "W=", sess.run(W), "b=",
sess.run(b))
                print("Optimization Finished!")
                training_cost = sess.run(cost, feed_dict={X:
train_X, Y: train_Y})
                print("Training cost=", training_cost, "W=",
sess.run(W), "b=", sess.run(b), '\n')
                # Testing example, as requested (Issue #2)
                print("Testing... (Mean square loss
Comparison)")
                testing_cost = sess.run(
                    tf.reduce_sum(tf.pow(pred - Y, 2)) / (2
* test_X.shape[0]),

```

```

        feed_dict={X: test_X, Y: test_Y}) # same
function as cost above
    print("Testing cost=", testing_cost)
    print("Absolute mean square loss difference:",
abs(
        training_cost - testing_cost))
# Graphic display
plt.plot(train_X, train_Y, 'ro', label='Original
data')
plt.plot(test_X, test_Y, 'bo', label='Testing data')
plt.plot(train_X, sess.run(W) * train_X +
sess.run(b), label='Fitted line')
plt.legend()
plt.show()

```

Rechnerprotokoll:

```

parallels@parallels-Parallels-Virtual-Platform:~$ python3
Python 3.6.7 (default, Oct 22 2018, 11:32:17)
[GCC 8.2.0] on linux
Type "help", "copyright", "credits" or "license" for
more information.
>>> """
... A linear regression learning algorithm example using
TensorFlow library.
... Author
... Project:
https://github.com/aymericdamien/TensorFlow-Examples/
... """
'\nA linear regression learning algorithm example using

```


TensorFlow library. \nAuthor\nProject:

<https://github.com/aymericdamien/TensorFlow-Examples>

```
>>> import tensorflow as tf
>>> import numpy
>>> import matplotlib.pyplot as plt
>>> rng = numpy.random
>>>
>>> # Parameters
... # learning_rate = 0.01 bzw. 0.1
... learning_rate = 0.01
>>> # training_epochs = 1000
... training_epochs = 10000
>>> # display_step = 50
... display_step = 500
>>>
>>> # Training Data
... train_X =
numpy.asarray([3.3, 4.4, 5.5, 6.71, 6.93, 4.168, 9.57])
>>> train_Y =
numpy.asarray([1.7, 2.76, 2.09, 3.19, 1.694, 1.573, 2.364])
>>> n_samples = train_X.shape[0]
>>>
>>> # Testing example, as requested (Issue #2)
... test_X = numpy.asarray([6.83, 4.668, 8.9,
7.91, 5.7, 8.7, 3.1, 2.1])
>>> test_Y = numpy.asarray([1.84, 2.273, 3.2,
2.831, 2.92, 3.24, 1.35, 1.03])
>>>
>>> # tf Graph Input
```

```

... X = tf.placeholder("float")
>>> Y = tf.placeholder("float")
>>>
>>> # Set model weights
... W = tf.Variable(rng.randn(), name="weight")
WARNING:tensorflow:From
/home/parallels/.local/lib/python3.6/site-packag
colocate_with (from
tensorflow.python.framework.ops) is deprecated
and will be removed in a future version.
Instructions for updating:
Colocations handled automatically by placer.
>>> b = tf.Variable(rng.randn(), name="bias")
>>>
>>> # Construct a linear model
... pred = tf.add(tf.multiply(X, W), b)
>>>
>>> # Mean squared error
... cost = tf.reduce_sum(tf.pow(pred-Y,
2))/(2*n_samples)
>>> # Gradient descent
... # Note, minimize() knows to modify W and b
because Variable objects are trainable=True by default
... optimizer =
tf.train.GradientDescentOptimizer(learning_rate).minimi
WARNING:tensorflow:From
/home/parallels/.local/lib/python3.6/site-packag
to_int32 (from tensorflow.python.ops.math_ops)
is deprecated and will be removed in a future

```


version.

Instructions for updating:

Use `tf.cast` instead.

```
>>>
>>> # Initialize the variables (i.e. assign their default
value)
... init = tf.global_variables_initializer()
>>>
>>> # Start training
... with tf.Session() as sess:
...     # Run the initializer
...     sess.run(init)
...     # Fit all training data
...     for epoch in range(training_epochs):
...         for (x, y) in zip(train_X, train_Y):
...             sess.run(optimizer, feed_dict={X:
x, Y: y})
...         # Display logs per epoch step
...         if (epoch+1) % display_step == 0:
...             c = sess.run(cost, feed_dict={X:
train_X, Y:train_Y})
...             print("Epoch:", '%04d' %
(epoch+1), "cost=", "{:.9f}".format(c), \
...                   "W=", sess.run(W), "b=",
sess.run(b))
...         print("Optimization Finished!")
...         training_cost = sess.run(cost, feed_dict={X:
train_X, Y: train_Y})
...         print("Training cost=", training_cost, "W=",
```

```

sess.run(W), "b=", sess.run(b), '\n')
...     # Testing example, as requested (Issue #2)
...     print("Testing... (Mean square loss
Comparison)")
...     testing_cost = sess.run(
...         tf.reduce_sum(tf.pow(pred - Y, 2)) /
(2 * test_X.shape[0]),
...         feed_dict={X: test_X, Y: test_Y}) #
same function as cost above
...     print("Testing cost=", testing_cost)
...     print("Absolute mean square loss difference:",
abs(
...         training_cost - testing_cost))
...     # Graphic display
...     plt.plot(train_X, train_Y, 'ro', label='Original
data')
...     plt.plot(test_X, test_Y, 'bo', label='Testing
data')
...     plt.plot(train_X, sess.run(W) * train_X +
sess.run(b), label='Fitted line')
...     plt.legend()
...     plt.show()
...
2019-03-26 13:14:53.558391: I
tensorflow/core/platform/profile_utils/cpu_utils.c 
CPU Frequency: 2494315000 Hz
2019-03-26 13:14:53.558647: I
tensorflow/compiler/xla/service/service.cc:150]
XLA service 0x20630a0 executing computations on

```

platform Host. Devices:

2019-03-26 13:14:53.558773: I

tensorflow/compiler/xla/service/service.cc:158]

StreamExecutor device (0): <undefined>,

<undefined>

Epoch: 0500 cost= 0.085635014 W= 0.19790356 b=
1.1733209

Epoch: 1000 cost= 0.079545885 W= 0.22159624 b=
1.0028771

Epoch: 1500 cost= 0.077750765 W= 0.23442905 b=
0.91055906

Epoch: 2000 cost= 0.077219352 W= 0.24137971 b=
0.86055654

Epoch: 2500 cost= 0.077060856 W= 0.24514504 b=
0.83346885

Epoch: 3000 cost= 0.077012964 W= 0.24718496 b=
0.8187941

Epoch: 3500 cost= 0.076998182 W= 0.24829005 b=
0.8108439

Epoch: 4000 cost= 0.076993421 W= 0.24888843 b=
0.8065385

Epoch: 4500 cost= 0.076991804 W= 0.24921261 b=
0.8042072

Epoch: 5000 cost= 0.076991208 W= 0.2493885 b=
0.80294234

Epoch: 5500 cost= 0.076990969 W= 0.2494826 b=
0.8022646

Epoch: 6000 cost= 0.076990865 W= 0.24953318 b=
0.8019009

Epoch: 6500 cost= 0.076990828 W= 0.24955891 b=
0.8017156
Epoch: 7000 cost= 0.076990791 W= 0.24957383 b=
0.80160946
Epoch: 7500 cost= 0.076990768 W= 0.24958722 b=
0.80151314
Epoch: 8000 cost= 0.076990783 W= 0.24959211 b=
0.8014765
Epoch: 8500 cost= 0.076990761 W= 0.24959634 b=
0.8014467
Epoch: 9000 cost= 0.076990761 W= 0.24960251 b=
0.8014019
Epoch: 9500 cost= 0.076990746 W= 0.24960734 b=
0.8013679
Epoch: 10000 cost= 0.076990739 W= 0.24960737 b=
0.8013675
Optimization Finished!
Training cost= 0.07699074 W= 0.24960737 b=
0.8013675

Testing... (Mean square loss Comparison)

Testing cost= 0.079108685

Absolute mean square loss difference: 0.0021179467

[<matplotlib.lines.Line2D object at 0x7fa9400fb940>]

[<matplotlib.lines.Line2D object at 0x7fa9400fbb8>]

[<matplotlib.lines.Line2D object at 0x7fa9400fbe80>]

<matplotlib.legend.Legend object at 0x7fa940090438>

=====

Skript2:

mit ClassPad:

Basic constant operations:

a:=2 2

b:=3 3

add=a+b add=5

mul=a*b mul=6

Matrix Multiplication:

matrix1:=[[3., 3.]] [3 3]

matrix2:=[[2.], [2.]]

$\begin{bmatrix} 2 \\ 2 \end{bmatrix}$

product:=matrix1*matrix2 [12]

Skript2 mit Tensorflow:

python3
'''

Basic Operations example using TensorFlow library.

Author: Aymeric Damien

Project:

```
https://github.com/aymericdamien/TensorFlow-Examples/  
'''
```

```
import tensorflow as tf
```

```
# Basic constant operations
```

```
# The value returned by the constructor represents the  
output
```

```
# of the Constant op.
```

```
a = tf.constant(2)
```

```
b = tf.constant(3)
```

```
# Launch the default graph.
```

```
with tf.Session() as sess:
```

```
    print("a=2, b=3")
```

```
    print("Addition with constants: %i" %
```

```
sess.run(a+b))
```

```
    print("Multiplication with constants: %i" %
```

```
sess.run(a*b))
```

```
# Basic Operations with variable as graph input
```

```
# The value returned by the constructor represents the  
output
```

```
# of the Variable op. (define as input when running  
session)
```

```
# tf Graph input
```

```
a = tf.placeholder(tf.int16)
```

```
b = tf.placeholder(tf.int16)
```



```

# Define some operations
add = tf.add(a, b)
mul = tf.multiply(a, b)

# Launch the default graph.
with tf.Session() as sess:
    # Run every operation with variable input
    print("Addition with variables: %i" %
sess.run(add, feed_dict={a: 2, b: 3}))
    print("Multiplication with variables: %i" %
sess.run(mul, feed_dict={a: 2, b: 3}))

# -----
# More in details:
# Matrix Multiplication from TensorFlow official
tutorial

# Create a Constant op that produces a 1x2 matrix.
The op is
# added as a node to the default graph.
#
# The value returned by the constructor represents the
output
# of the Constant op.
matrix1 = tf.constant([[3., 3.]])

# Create another Constant that produces a 2x1 matrix.
matrix2 = tf.constant([[2.], [2.]])

```

```

# Create a Matmul op that takes 'matrix1' and
'matrix2' as inputs.
# The returned value, 'product', represents the result
of the matrix
# multiplication.
product = tf.matmul(matrix1, matrix2)

# To run the matmul op we call the session 'run()'
method, passing 'product'
# which represents the output of the matmul op. This
indicates to the call
# that we want to get the output of the matmul op
back.
#
# All inputs needed by the op are run automatically by
the session. They
# typically are run in parallel.
#
# The call 'run(product)' thus causes the execution of
three ops in the
# graph: the two constants and matmul.
#
# The output of the op is returned in 'result' as a
numpy `ndarray` object.
with tf.Session() as sess:
    result = sess.run(product)
    print(result)
# ==> [[ 12.]]

```

Rechnerprotokoll:

```
parallels@parallels-Parallels-Virtual-Platform:~$ python3
Python 3.6.7 (default, Oct 22 2018, 11:32:17)
[GCC 8.2.0] on linux
Type "help", "copyright", "credits" or "license" for
more information.
>>> """
... Basic Operations example using TensorFlow library.
... Author
... Project:
https://github.com/aymericdamien/TensorFlow-Examples/
... """
'\nBasic Operations example using TensorFlow
library.\nAuthor\nProject:
https://github.com/aymericdamien/TensorFlow-Examples▶
>>>
>>> import tensorflow as tf
>>>
>>> # Basic constant operations
... # The value returned by the constructor represents
the output
... # of the Constant op.
... a = tf.constant(2)
>>> b = tf.constant(3)
>>>
>>> # Launch the default graph.
... with tf.Session() as sess:
...     print("a=2, b=3")
```

```

...     print("Addition with constants: %i" %
sess.run(a+b))
...     print("Multiplication with constants: %i" %
sess.run(a*b))
...
a=2, b=3
Addition with constants: 5
Multiplication with constants: 6
>>> # Basic Operations with variable as graph input
... # The value returned by the constructor represents
the output
... # of the Variable op. (define as input when
running session)
... # tf Graph input
... a = tf.placeholder(tf.int16)
>>> b = tf.placeholder(tf.int16)
>>>
>>> # Define some operations
... add = tf.add(a, b)
>>> mul = tf.multiply(a, b)
>>>
>>> # Launch the default graph.
... with tf.Session() as sess:
...     # Run every operation with variable input
...     print("Addition with variables: %i" %
sess.run(add, feed_dict={a: 2, b: 3}))
...     print("Multiplication with variables: %i" %
sess.run(mul, feed_dict={a: 2, b: 3}))
...

```

Addition with variables: 5

Multiplication with variables: 6

```
>>>
>>> # -----
... # More in details:
... # Matrix Multiplication from TensorFlow official
tutorial
...
>>> # Create a Constant op that produces a 1x2
matrix. The op is
... # added as a node to the default graph.
... #
... # The value returned by the constructor represents
the output
... # of the Constant op.
... matrix1 = tf.constant([[3., 3.]])
>>>
>>> # Create another Constant that produces a 2x1
matrix.
... matrix2 = tf.constant([[2.],[2.]])
>>>
>>> # Create a Matmul op that takes 'matrix1' and
'matrix2' as inputs.
... # The returned value, 'product', represents the
result of the matrix
... # multiplication.
... product = tf.matmul(matrix1, matrix2)
>>>
>>> # To run the matmul op we call the session
```

```

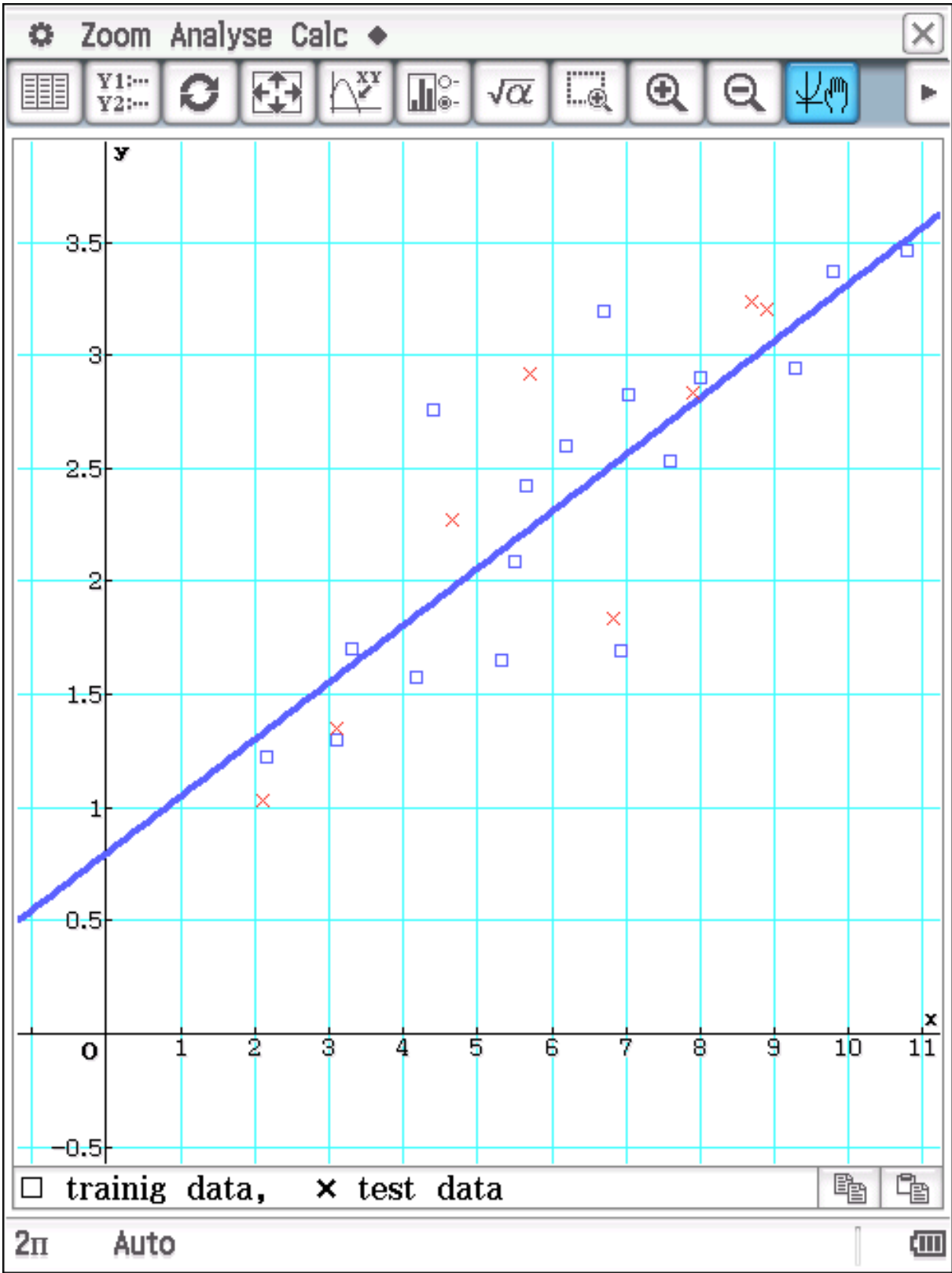
'run()' method, passing 'product'
... # which represents the output of the matmul op.
This indicates to the call
... # that we want to get the output of the matmul
op back.
... #
... # All inputs needed by the op are run
automatically by the session. They
... # typically are run in parallel.
... #
... # The call 'run(product)' thus causes the
execution of three ops in the
... # graph: the two constants and matmul.
... #
... # The output of the op is returned in 'result' as
a numpy `ndarray` object.
... with tf.Session() as sess:
...     result = sess.run(product)
...     print(result)
... # ==> [[ 12.]]
...
[[12.]]
>>>

```

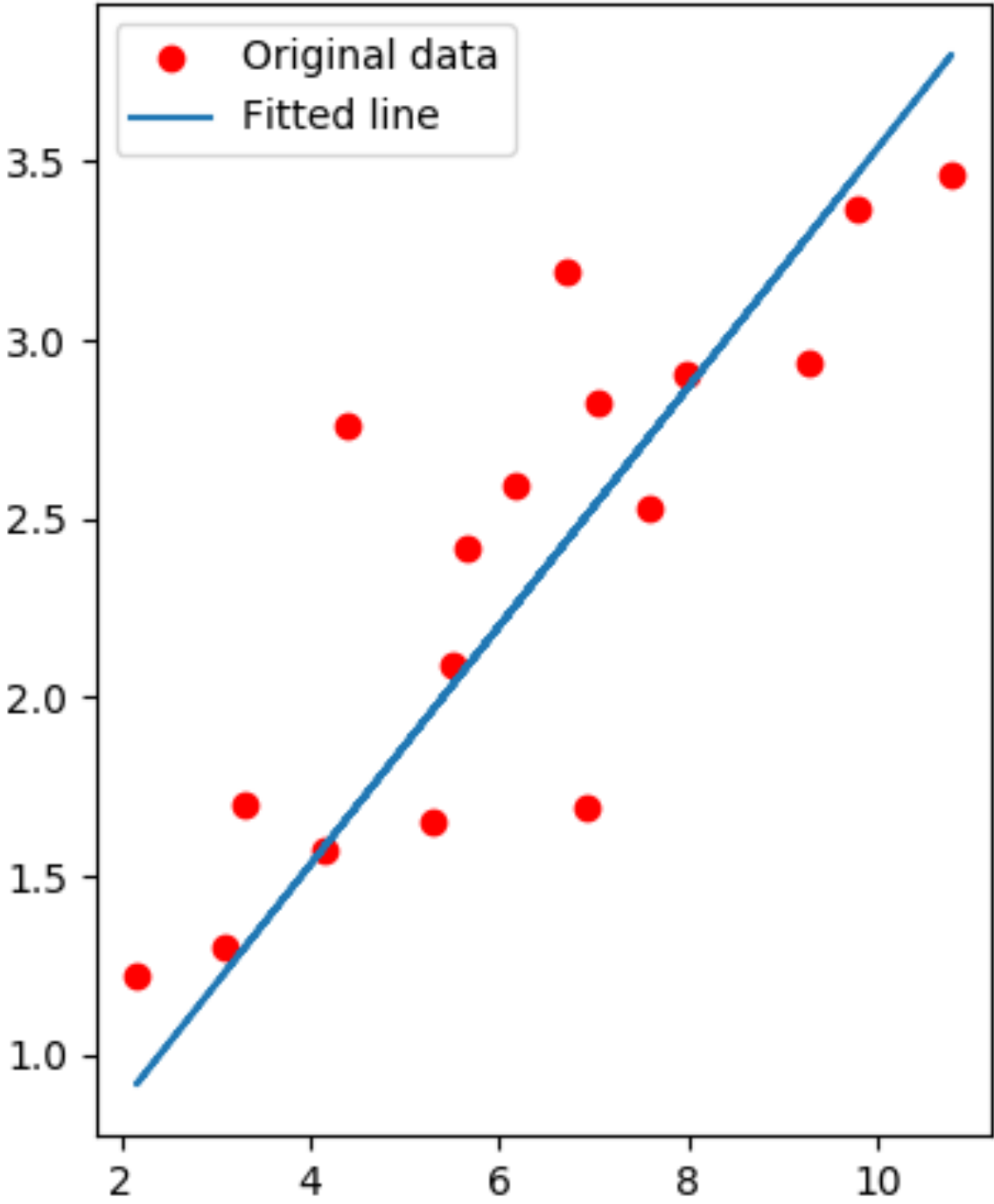
Download für dieses Dokument:

[www.informatik.htw-dresden.de/
~paditz/Tensorflow-Ue17.pdf](http://www.informatik.htw-dresden.de/~paditz/Tensorflow-Ue17.pdf)

Lineare Regression: mit ClassPad



Lineare Regression: mit Tensorflow



Lineare Regression: mit Tensorflow

