

Machine Learning with TensorFlow

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Training Neural Networks: Deep Learning Libraries

TensorFlow

Platform: Linux, Mac OS, Windows

– Written in: C++, Python

Interface: Python, C/C++, Java, Go, R



Keras



PyTorch

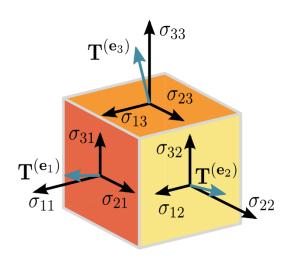




TensorFlow

- Developed by Google and it is one of the most popular Machine Learning libraries on GitHub.
- It is a framework to perform computation very efficiently, and it can tap into the GPU in order to speed it up even further.
- TensorFlow is one of the widely used libraries for implementing machine learning and deep learning involving large number of mathematical operations.

- Tensor and Flow
 - TensorFlow gets its name from tensors, which are arrays of arbitrary dimensionality.
 - The "flow" part of the name refers to computation flowing through a graph.





Computational Graph

- TensorFlow is an open-source software library for deep learning
 - tf.constant
 - tf.Variable
 - tf.placeholder

```
import tensorflow as tf

a = tf.constant([1,2,3])
b = tf.constant(4, shape=[1,3])

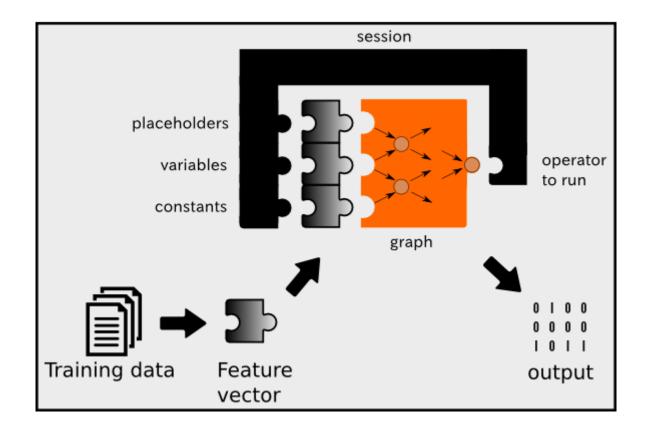
A = a + b
B = a*b

print(A)
```

Tensor("add_1:0", shape=(1, 3), dtype=int32)

TensorFlow: Session

• To run any of the three defined operations, we need to create a session for that graph. The session will also allocate memory to store the current value of the variable.



TensorFlow

```
import tensorflow as tf

a = tf.constant([1,2,3])
b = tf.constant(4, shape=[1,3])

A = a + b
B = a*b

print(A)
```

```
a = tf.constant([1,2,3])
b = tf.constant([4,5,6])

result = tf.multiply(a, b)

with tf.Session() as sess:
    output = sess.run(result)
```

print(output)

← Interactive Session: run the result and close the Session automatically

```
sess = tf.Session()
sess.run(A)

array([[5, 6, 7]])

sess.run(B)

array([[ 4,  8, 12]])
```

TensorFlow: tf.Variable

- tf. Variable is regarded as the decision variable in optimization.
- We should initialize variables.

```
x1 = tf.Variable([1, 1], dtype = tf.float32)
x2 = tf.Variable([2, 2], dtype = tf.float32)
y = x1 + x2
print(y)
<tf.Tensor 'add 8:0' shape=(2,) dtype=float32>
```

```
sess = tf.Session()
init = tf.global_variables_initializer()
sess.run(init)
sess.run(y)
```

```
array([ 3., 3.], dtype=float32)
```

TensorFlow: Placeholder

The value of tf.placeholder must be fed using the feed_dict optional argument to Session.run()

```
a = tf.placeholder(tf.float32, shape = [2])
b = tf.placeholder(tf.float32, shape = [2])
sum = a + b
sess.run(sum, feed_dict = {a : [1,2], b : [3,4]})
array([ 4., 6.], dtype=float32)
```

Tensor Manipulation: Adding

```
x1 = tf.constant(1, shape = [3])
x2 = tf.constant(2, shape = [3])
output = tf.add(x1, x2)

with tf.Session() as sess:
    result = sess.run(output)
    print(result)
```

[3 3 3]

```
x1 = tf.constant(1, shape = [2, 3])
x2 = tf.constant(2, shape = [2, 3])
output = tf.add(x1, x2)

with tf.Session() as sess:
    result = sess.run(output)
    print(result)
```

```
[[3 3 3]
[3 3 3]]
```

Tensor Manipulation: Multiplying

```
output1 = tf.matmul(x1, x2)
with tf.Session() as sess:
   result = sess.run(output1)
   print(result)

[[ 8]
   [18]]
```

```
output2 = x1*x2
with tf.Session() as sess:
    result = sess.run(output2)
    print(result)

[[ 2   4]
    [ 9 12]]
```

Tensor Manipulation: Reshape

```
x = [1, 2, 3, 4, 5, 6, 7, 8]
```

TensorFlow as an Optimization Solver

```
w = tf.Variable(0, dtype = tf.float32)
cost = w*w - 8*w + 16

LR = 0.05
optm = tf.train.GradientDescentOptimizer(LR).minimize(cost)

init = tf.global_variables_initializer()

sess = tf.Session()
sess.run(init)

print(sess.run(w))
```

```
\min_{\omega} \ (\omega - 4)^2
```

```
# runs one step of gradient descent
sess.run(optm)
print(sess.run(w))

# runs two step of gradient descent
sess.run(optm)
print(sess.run(w))
```

```
for _ in range(100):
    sess.run(optm)

print(sess.run(w))
```

0.0

0.4

0.76

3.99991

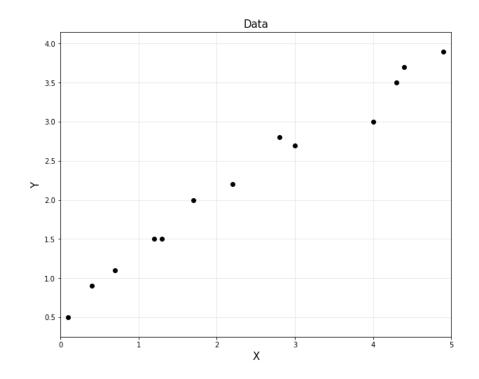
sess.close()

Machine Learning with TensorFlow

Regression

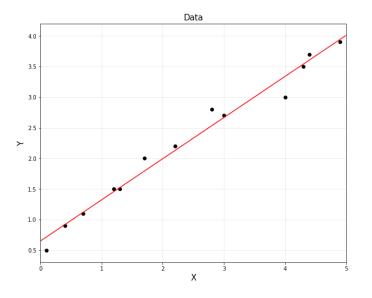
- Given (x_i, y_i) for $i = 1, \dots, m$,
- Want to estimate

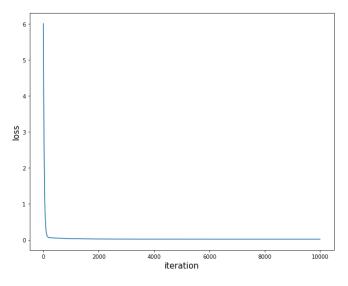
$${\hat y}_i = \omega x_i + b \quad ext{ such that } \quad \min_{\omega,b} \sum_{i=1}^m ({\hat y}_i - y_i)^2$$



Regression with TensorFlow

```
LR = 0.001
n iter = 10000
x = tf.placeholder(tf.float32, [m, 1])
y = tf.placeholder(tf.float32, [m, 1])
w = tf.Variable([[0]], dtype = tf.float32)
b = tf.Variable([[0]], dtype = tf.float32)
\#y\_pred = tf.matmul(x, w) + b
y_pred = tf.add(tf.matmul(x, w), b)
loss = tf.square(y_pred - y)
loss = tf.reduce_mean(loss)
optm = tf.train.GradientDescentOptimizer(LR).minimize(loss)
sess = tf.Session()
sess.run(tf.global_variables_initializer())
loss record = []
for epoch in range(n iter):
    _, c = sess.run([optm, loss], feed_dict = {x: train_x, y: train_y})
    loss record.append(c)
w val = sess.run(w)
b val = sess.run(b)
sess.close()
```



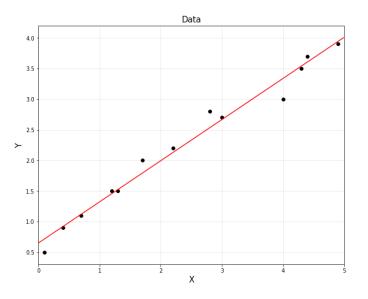


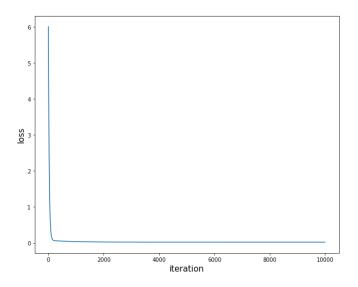


Regression with TensorFlow

• with tf.Session() as sess:

```
LR = 0.001
n iter = 10000
x = tf.placeholder(tf.float32, [m, 1])
y = tf.placeholder(tf.float32, [m, 1])
w = tf.Variable([[0]], dtype = tf.float32)
b = tf.Variable([[0]], dtype = tf.float32)
\#y\_pred = tf.matmul(x, w) + b
y_pred = tf.add(tf.matmul(x, w), b)
loss = tf.square(y pred - y)
loss = tf.reduce_mean(loss)
optm = tf.train.GradientDescentOptimizer(LR).minimize(loss)
with tf.Session() as sess:
    sess.run(tf.global_variables_initializer())
   for epoch in range(n_iter):
        sess.run(optm, feed_dict = {x: train_x, y: train_y})
    w val = sess.run(w)
    b_val = sess.run(b)
```



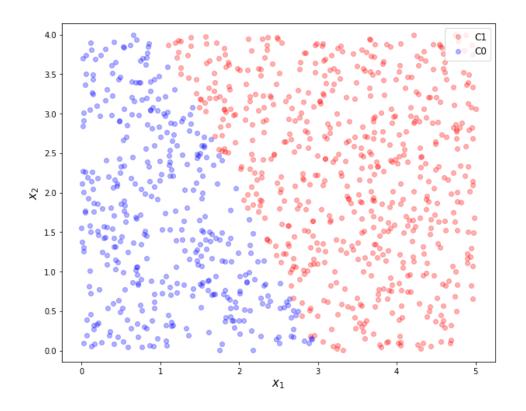




Logistic Regression

$$\omega = egin{bmatrix} \omega_0 \ \omega_1 \ \omega_2 \end{bmatrix}, \qquad x = egin{bmatrix} 1 \ x_1 \ x_2 \end{bmatrix}$$

$$X = egin{bmatrix} \left(x^{(1)}
ight)^T \ \left(x^{(2)}
ight)^T \ \left(x^{(3)}
ight)^T \ dots \end{bmatrix} = egin{bmatrix} 1 & x_1^{(1)} & x_2^{(1)} \ 1 & x_1^{(2)} & x_2^{(2)} \ 1 & x_1^{(3)} & x_2^{(3)} \ dots & dots \end{bmatrix}, \quad y = egin{bmatrix} y^{(1)} \ y^{(2)} \ y^{(3)} \ dots \end{bmatrix}$$

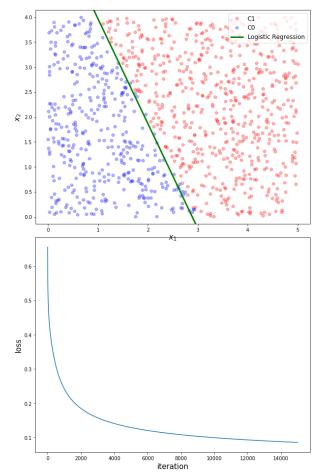




Logistic Regression with TensorFlow

$$egin{aligned} \ell(\omega) &= \log \mathscr{L}(\omega) = \sum_{i=1}^m y^{(i)} \log h_\omega \left(x^{(i)}
ight) + \left(1 - y^{(i)}
ight) \log \left(1 - h_\omega \left(x^{(i)}
ight)
ight) \ &\Rightarrow rac{1}{m} \sum_{i=1}^m y^{(i)} \log h_\omega \left(x^{(i)}
ight) + \left(1 - y^{(i)}
ight) \log \left(1 - h_\omega \left(x^{(i)}
ight)
ight) \end{aligned}$$

```
LR = 0.05
n iter = 15000
X = tf.placeholder(tf.float32, [m, 3])
y = tf.placeholder(tf.float32, [m, 1])
W = tf.Variable([[0],[0],[0]], dtype = tf.float32)
y pred = tf.sigmoid(tf.matmul(X,w))
loss = - y*tf.log(y pred) - (1-y)*tf.log(1-y pred)
loss = tf.reduce mean(loss)
optm = tf.train.GradientDescentOptimizer(LR).minimize(loss)
loss record = []
with tf.Session() as sess:
    sess.run(tf.global variables initializer())
    for epoch in range(n iter):
        _, c = sess.run([optm, loss], feed_dict = {X: train_X, y: train_y})
        loss record.append(c)
    w hat = sess.run(w)
```



Logistic Regression with TensorFlow

- TensorFlow embedded functions
 - tf.nn.sigmoid_cross_entropy_with_logits for binary classification
 - tf.nn.softmax cross entropy with logits for multiclass classification

```
LR = 0.05
n iter = 30000
X = tf.placeholder(tf.float32, [m, 3])
y = tf.placeholder(tf.float32, [m, 1])
w = tf.Variable(tf.random normal([3,1]), dtype = tf.float32)
y pred = tf.matmul(X,w)
loss = tf.nn.sigmoid cross entropy with logits(logits = y pred, labels = y)
loss = tf.reduce mean(loss)
optm = tf.train.GradientDescentOptimizer(LR).minimize(loss)
with tf.Session() as sess:
    sess.run(tf.global_variables_initializer())
   for epoch in range(n iter):
        sess.run(optm, feed dict = {X: train X, y: train y})
   w hat = sess.run(w)
```

