

Tensorflow 개발 팁

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```
In [1]: %load_ext do_not_print_href
```

GPU 사용 방식 설정

tf.ConfigProto

- <http://devdocs.io/tensorflow~python/tf/configproto>
- <https://github.com/tensorflow/tensorflow/blob/r1.2/tensorflow/core/protobuf/config.proto>

```
config = tf.ConfigProto()
config.gpu_options.allow_growth = True
...
session = tf.Session(config=config)
```

CUDA_VISIBLE_DEVICES

- 여러개의 GPU가 장착된 시스템에서 사용시, 특정 GPU만 선택해서 사용하도록 설정

```
CUDA_VISIBLE_DEVICES=0,1
```

- GPU가 있는 시스템에서, CPU만 사용하도록 만들 때도 사용 가능

```
CUDA_VISIBLE_DEVICES=-
```

- 환경변수로 저장

```
export CUDA_VISIBLE_DEVICES=0,1  
python ...
```

- 필요시 ~/.bashrc 파일에 추가하면 로그인 할 때마다 적용
- 참고: <http://acceleware.com/blog/cudavisibledevices-masking-gpus>

주의사항

- GPU 번호가 `nvidia-smi`에서 보여주는 번호와 다를 수 있습니다.

모델의 저장과 복원

모델 파라미터의 저장과 복원

tf.train.Saver

```
tf.train.Saver(  
    var_list=None,  
    reshape=False,  
    sharded=False,  
    max_to_keep=5,  
    keep_checkpoint_every_n_hours=10000.0,  
    name=None,  
    restore_sequentially=False,  
    saver_def=None,  
    builder=None,  
    defer_build=False,  
    allow_empty=False,  
    write_version=tf.train.SaverDef.V2,  
    pad_step_number=False,  
    save_relative_paths=False  
)
```

모델 파라미터 저장

saver.save().

```
save(  
    sess,  
    save_path,  
    global_step=None,  
    latest_filename=None,  
    meta_graph_suffix='meta',  
    write_meta_graph=True,  
    write_state=True  
)
```

Returns: A string: path at which the variables were saved. If the saver is sharded, this string ends with: '-?????-of-nnnnn' where 'nnnnn' is the number of shards created. If the saver is empty, returns None.

모델 파라미터 복원

saver.restore().

```
restore(  
    sess,  
    save_path  
)
```

저장 복원 예시

MNIST 데이터 준비 (1주 3일차 데이터 활용)

```
In [2]: %%bash  
test -s ./mnist/train-images-idx3-ubyte || (  
    mkdir -p ./mnist  
    cd ./mnist  
    echo "$(pwd)"  
    wget -q \  
        http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz \  
        http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz \  
        http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz \  
        http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz  
    gzip -d *.gz  
)
```

MNIST 데이터 로딩

```
In [3]: from __future__ import print_function, division
import numpy as np
import tensorflow as tf
```

```
In [4]: data_dir = './mnist'
data_dir

images      = np.fromfile(data_dir +
                           '/train-images-idx3-ubyte',dtype=np.uint8)
images      = images[16:].reshape([-1,28,28]).astype(np.float32)
images      = images / 127.0 - 1.0

labels      = np.fromfile(data_dir +
                           '/train-labels-idx1-ubyte',dtype=np.uint8)
labels      = labels[8:].astype(np.int64)
```

```
In [5]: images.shape, images.dtype, labels.shape, labels.dtype
```

```
Out[5]: ((60000, 28, 28), dtype('float32'), (60000,), dtype('int64'))
```

MNIST 훈련 네트워크 정의

```

In [6]: class Model:
        '''
        simple 1-layer fully-connected network for MNIST problem
        '''
        def __init__(self):
            learning_rate = 0.05
            input_size     = 28 * 28
            output_size     = 10

            with tf.name_scope('data'):
                input_      = tf.placeholder(
                    shape=[None, input_size],
                    dtype=tf.float32, name="input")
                label       = tf.placeholder(
                    shape=[None],
                    dtype=tf.int64, name="label")
            with tf.name_scope('fcn'):
                weights      = tf.Variable(
                    tf.zeros([input_size, output_size]))
                biases      = tf.Variable(
                    tf.zeros([output_size]))
                output       = tf.matmul(input_, weights) + biases

            with tf.name_scope('optimize'):
                loss         = \
                    tf.losses.sparse_softmax_cross_entropy(
                        label,
                        output)
                optimize     = \
                    tf.train.AdamOptimizer(learning_rate) \
                        .minimize(loss)

            with tf.name_scope('prediction'):
                pred         = tf.argmax(output,
                                         axis=1,
                                         name='pred')

                accuracy     = \
                    1.0 - \
                    tf.cast(
                        tf.count_nonzero(pred-label),
                        tf.float32) / \
                    tf.cast(tf.size(label),tf.float32)

            self.input      = input_
            self.label      = label
            self.loss       = loss
            self.optimize   = optimize
            self.pred       = pred
            self.accuracy   = accuracy
            self.weights    = weights
            self.biases     = biases

```

```

In [7]: model = Model()

```

MNIST 훈련

```
In [8]: def train(model,max_epochs=20):

    batch_size      = 128
    batch_count     = 60000 // batch_size

    step            = 1

    config = tf.ConfigProto(gpu_options={'allow_growth': True})
    with tf.Session(config=config) as session:

        saver = tf.train.Saver()

        session.run(tf.global_variables_initializer())
        for ep in range(max_epochs):
            total_loss      = 0
            total_acc_v     = 0
            for i in range(batch_count):
                img = np.reshape(
                    images[i*batch_size:(i+1)*batch_size],
                    [batch_size, 28 * 28])
                lbl = (labels[i*batch_size:(i+1)*batch_size])
                _, loss_v, acc_v = \
                    session.run(
                        [model.optimize,
                         model.loss,
                         model.accuracy],
                        feed_dict= {
                            model.input: img,
                            model.label: lbl})
                step      += 1
                total_loss += loss_v
                total_acc_v += acc_v

            # <<<=== 주기적으로 save ===>>>
            #####
            # 2000 step 마다 save/example 에 저장
            #####
            if step % 2000 == 0:
                checkpoint = saver.save(session,
                                         'save/example',
                                         global_step=step)
                print('Saved: %s'%(checkpoint,))

            print('ep %d: loss: %.5f acc: %.3f%%' % (
                ep+1,
                total_loss / batch_count,
                total_acc_v / batch_count * 100))

            #####
            # save/graph 에 저장
            #####
```

```

tf.train.export_meta_graph(filename='save/graph')

#####
# save/example 에 저장
#####

checkpoint = saver.save(session,
                        'save/example',
                        global_step=step)
print('Saved: %s'%(checkpoint,))

writer = tf.summary.FileWriter(
    'save',
    tf.get_default_graph())

return step

```

```
In [9]: !rm -fr save
```

```
In [10]: train(model,20);
```

```

ep 1: loss: 2.14669 acc: 82.175%
ep 2: loss: 1.65718 acc: 85.445%
ep 3: loss: 1.94722 acc: 85.882%
ep 4: loss: 2.05991 acc: 86.485%
Saved: save/example-2000
ep 5: loss: 1.99479 acc: 86.988%
ep 6: loss: 2.04258 acc: 87.023%
ep 7: loss: 2.02497 acc: 87.336%
ep 8: loss: 2.11884 acc: 87.238%
Saved: save/example-4000
ep 9: loss: 2.04744 acc: 87.570%
ep 10: loss: 2.03975 acc: 87.690%
ep 11: loss: 1.98439 acc: 87.857%
ep 12: loss: 2.10790 acc: 87.630%
Saved: save/example-6000
ep 13: loss: 2.15829 acc: 87.759%
ep 14: loss: 2.23853 acc: 87.727%
ep 15: loss: 2.06537 acc: 88.418%
ep 16: loss: 2.12343 acc: 88.103%
ep 17: loss: 2.09931 acc: 88.303%
Saved: save/example-8000
ep 18: loss: 2.14287 acc: 88.134%
ep 19: loss: 2.19149 acc: 88.086%
ep 20: loss: 2.12249 acc: 88.361%
Saved: save/example-9361

```

```
In [11]: !ls -altr save
```

```
total 1808
-rw-r--r--  1 rhee  staff  52981 Sep  8 11:03 example-2000.meta
-rw-r--r--  1 rhee  staff   343 Sep  8 11:03 example-2000.index
-rw-r--r--  1 rhee  staff  94208 Sep  8 11:03 example-2000.data-0
0000-of-00001
drwxrwxr-x 29 rhee  staff   986 Sep  8 11:03 ..
-rw-r--r--  1 rhee  staff  52981 Sep  8 11:03 example-4000.meta
-rw-r--r--  1 rhee  staff   343 Sep  8 11:03 example-4000.index
-rw-r--r--  1 rhee  staff  94208 Sep  8 11:03 example-4000.data-0
0000-of-00001
-rw-r--r--  1 rhee  staff  52981 Sep  8 11:03 example-6000.meta
-rw-r--r--  1 rhee  staff   343 Sep  8 11:03 example-6000.index
-rw-r--r--  1 rhee  staff  94208 Sep  8 11:03 example-6000.data-0
0000-of-00001
-rw-r--r--  1 rhee  staff  52981 Sep  8 11:03 example-8000.meta
-rw-r--r--  1 rhee  staff   343 Sep  8 11:03 example-8000.index
-rw-r--r--  1 rhee  staff  94208 Sep  8 11:03 example-8000.data-0
0000-of-00001
-rw-r--r--  1 rhee  staff  52911 Sep  8 11:03 graph
-rw-r--r--  1 rhee  staff  52981 Sep  8 11:03 example-9361.meta
-rw-r--r--  1 rhee  staff   343 Sep  8 11:03 example-9361.index
-rw-r--r--  1 rhee  staff  94208 Sep  8 11:03 example-9361.data-0
0000-of-00001
-rw-r--r--  1 rhee  staff   253 Sep  8 11:03 checkpoint
drwxr-xr-x 20 rhee  staff   680 Sep  8 11:03 .
-rw-r--r--  1 rhee  staff  98509 Sep  8 11:03 events.out.tfevents
.1504836214.rhee-mbp.local
```

여기서 잠깐, 텐서보드 그래프 기능 한 번 확인 하겠습니다.

```
In [12]: !tensorboard --logdir save
```

```
Starting TensorBoard 54 at http://rhee-mbp.local:6006
(Press CTRL+C to quit)
^C
```

저장된 모델 파라미터의 복원

모델 파라미터 텐서는 문자열로 표현 가능

```
In [13]: model.pred.name, model.input.name
```

```
Out[13]: (u'prediction/pred:0', u'data/input:0')
```

```
In [14]: tf.reset_default_graph()
model = Model()
```



```
In [15]: model.pred.name, model.input.name
```

```
Out[15]: (u'prediction/pred:0', u'data/input:0')
```

```
In [16]: from __future__ import print_function
import numpy as np
import tensorflow as tf

def infer(image, label):

    config = tf.ConfigProto(gpu_options={'allow_growth': True})
    with tf.Session(config=config) as session:

        session.run(tf.global_variables_initializer())

        #####
        # save/example 에 저장한 파일로 부터 복원
        #####

        saver = tf.train.Saver()
        saver.restore(session, 'save/example-9361')

        pred = session.run(
            # model.pred
            'prediction/pred:0',
            # model.input
            {'data/input:0': [image.reshape([28*28])]}

            print('infer: label={}, pred={}'.format(label, pred[0]))
```

```
In [17]: infer(images[1234], labels[1234])
```

```
INFO:tensorflow:Restoring parameters from save/example-9361
infer: label=3, pred=3
```

그래프의 저장과 복원

그래프 저장

`tf.train.export_meta_graph('save/graph').`

```
tf.train.export_meta_graph(  
    filename=None,  
    meta_info_def=None,  
    graph_def=None,  
    saver_def=None,  
    collection_list=None,  
    as_text=False,  
    graph=None,  
    export_scope=None,  
    clear_devices=False,  
    **kwargs  
)
```

그래프 복원

`tf.train.import_meta_graph('save/graph').`

```
import_meta_graph(  
    meta_graph_or_file,  
    clear_devices=False,  
    import_scope=None,  
    **kwargs  
)
```

저장된 그래프와 모델 파라미터 복원

```
In [18]: tf.reset_default_graph()  
         # model = Model() # <<<=== 만들 필요 없음  
         tf.train.import_meta_graph('save/graph')
```

```
Out[18]: <tensorflow.python.training.saver.Saver at 0x11a7169d0>
```

```
In [19]: from __future__ import print_function
import numpy as np
import tensorflow as tf

def infer(image, label):

    config = tf.ConfigProto(gpu_options={'allow_growth': True})
    with tf.Session(config=config) as session:

        session.run(tf.global_variables_initializer())

        #####
        # save/example 에 저장한 파일로 부터 복원
        #####

        saver = tf.train.Saver()
        saver.restore(session, 'save/example-9361')

        model_input = 'data/input:0'
        model_pred = 'prediction/pred:0'

        pred = session.run(
            model_pred,
            {model_input:[image.reshape([28*28])]}})

        print('infer: label={}, pred={}'.format(label,pred[0]))

infer(images[1234],labels[1234])
```

```
INFO:tensorflow:Restoring parameters from save/example-9361
infer: label=3, pred=3
```

실제로 전혀 다른 파이썬 인터프리터로 확인

```
In [20]: %pycat restore_test.py
```

```
In [21]: !python restore_test.py
```

```
2017-09-08 11:03:53.163645: W tensorflow/core/platform/cpu_feature_guard.cc:45] The TensorFlow library wasn't compiled to use SSE4.2 instructions, but these are available on your machine and could speed up CPU computations.
2017-09-08 11:03:53.163666: W tensorflow/core/platform/cpu_feature_guard.cc:45] The TensorFlow library wasn't compiled to use AVX instructions, but these are available on your machine and could speed up CPU computations.
2017-09-08 11:03:53.163670: W tensorflow/core/platform/cpu_feature_guard.cc:45] The TensorFlow library wasn't compiled to use AVX2 instructions, but these are available on your machine and could speed up CPU computations.
2017-09-08 11:03:53.163674: W tensorflow/core/platform/cpu_feature_guard.cc:45] The TensorFlow library wasn't compiled to use FMA instructions, but these are available on your machine and could speed up CPU computations.
infer: label=3, pred=3
```

모델의 저장과 복원 정리

파라미터 저장

```
# 모든 파라미터 저장 (trainable/non-trainable 구별은 하지 않음)
saver = tf.train.Saver()
saver_checkpoint = saver.save(sess, 'save/my-model', global_step=step)

# 선택한 파라미터만 저장
selected_saver = tf.train.Saver([model.w1,model.w2])
selected_saver.save(sess, 'save2/weights', global_step=step)

# scope 속성을 활용한 파라미터 선택
export_saver = tf.train.Saver( \
    tf.get_collection(tf.GraphKeys.GLOBAL_VARIABLES, scope='my_block'))
export_saver.save(sess, 'export/my-block', global_step=step)
```

tf.get_collection 을 이용해서 원하는 텐서의 목록을 얻음

tf.GraphKeys.GLOBAL_VARIABLES 를 지정해서 모든 글로벌 파라미터의 목록을 얻을 수 있음

파라미터 복원

```
# checkpoint 경로를 미리 알고 있다면
saver = tf.train.Saver()
saver.restore(sess, saver_checkpoint)

# 지정 디렉토리 'save'에서 가장 최근의 체크포인트 경로를 읽어옴
checkpoint = tf.train.latest_checkpoint('save')
saver_for_restore = tf.train.Saver()
saver_for_restore.restore(sess, checkpoint)

# 모든 파라미터를 다 읽어오지 않고, 정해진 파라미터들만 읽어오고 싶다면
import_checkpoint = tf.train.latest_checkpoint('export')
import_saver = tf.train.Saver( \
    tf.get_collection(tf.GraphKeys.GLOBAL_VARIABLES, \
        scope='my_block'))
import_saver.restore(sess, import_checkpoint)
```

그래프 저장

```
tf.train.export_meta_graph(filename='save/graph')
```

그래프 복원

```
tf.train.import_meta_graph('save/graph')
```

저장된 모델을 텐서플로우 외부에서 확인

- inspect_checkpoint.py
- https://github.com/tensorflow/tensorflow/blob/master/tensorflow/python/tools/inspect_checkpoint.py

```
In [22]: !python -mtensorflow.python.tools.inspect_checkpoint
```

```
Usage: inspect_checkpoint --file_name=checkpoint_file_name [--tensor_name=tensor_to_print]
```

```
In [23]: %%bash
python -mtensorflow.python.tools.inspect_checkpoint \
    --file_name 'save/example-9361'
```

```
fcn/Variable (DT_FLOAT) [784,10]
fcn/Variable/Adam (DT_FLOAT) [784,10]
fcn/Variable/Adam_1 (DT_FLOAT) [784,10]
fcn/Variable_1 (DT_FLOAT) [10]
fcn/Variable_1/Adam (DT_FLOAT) [10]
fcn/Variable_1/Adam_1 (DT_FLOAT) [10]
optimize/beta1_power (DT_FLOAT) []
optimize/beta2_power (DT_FLOAT) []
```

그래프의 finalize()

tf.get_default_graph().finalize()

Finalizes this graph, making it read-only.

After calling `g.finalize()`, no new operations can be added to `g`. This method is used to ensure that no operations are added to a graph when it is shared between multiple threads, for example when using a `tf.train.QueueRunner`.

```
In [24]: tf.get_default_graph().finalize()
one_ = tf.constant(1.0)
```

```
-----
-----
RuntimeError                                Traceback (most recent c
all last)
<ipython-input-24-b1f25alfad0d> in <module>()
      1 tf.get_default_graph().finalize()
----> 2 one_ = tf.constant(1.0)

/opt/conda/envs/tensorflow/lib/python2.7/site-packages/tensorflow/
python/framework/constant_op.py in constant(value, dtype, shape,
name, verify_shape)
    104     const_tensor = g.create_op(
    105         "Const", [], [dtype_value.type],
--> 106         attrs={"value": tensor_value, "dtype": dtype_value},
name=name).outputs[0]
    107     return const_tensor
    108

/opt/conda/envs/tensorflow/lib/python2.7/site-packages/tensorflow/
python/framework/ops.py in create_op(self, op_type, inputs, dtype
s, input_types, name, attrs, op_def, compute_shapes, compute_devic
e)
    2456
    2457     """
-> 2458     self._check_not_finalized()
    2459     for idx, a in enumerate(inputs):
    2460         if not isinstance(a, Tensor):

/opt/conda/envs/tensorflow/lib/python2.7/site-packages/tensorflow/
python/framework/ops.py in _check_not_finalized(self)
    2179     """
    2180     if self._finalized:
-> 2181         raise RuntimeError("Graph is finalized and cannot be
modified.")
    2182
    2183     def _add_op(self, op):
```

RuntimeError: Graph is finalized and cannot be modified.

그래프 reset()

tf.reset_default_graph()

Clears the default graph stack and resets the global default graph.

NOTE: The default graph is a property of the current thread. This function applies only to the current thread. Calling this function while a `tf.Session` or `tf.InteractiveSession` is active will result in undefined behavior. Using any previously created `tf.Operation` or `tf.Tensor` objects after calling this function will result in undefined behavior.

```
In [25]: tf.reset_default_graph()  
         one_ = tf.constant(1.0)
```

텐서 수치오류 체크

tf.check_numerics(tensor, message)

```
tf.check_numerics(  
    tensor,  
    message,  
    name=None  
)
```

Checks a tensor for NaN and Inf values.

When run, reports an `InvalidArgument` error if tensor has any values that are not a number (NaN) or infinity (Inf). Otherwise, passes tensor as-is.

```
In [26]: config = tf.ConfigProto(gpu_options={'allow_growth': True})  
         sess = tf.InteractiveSession(config=config)  
         sess
```

```
Out[26]: <tensorflow.python.client.session.InteractiveSession at 0x11a727310>
```

```
In [27]: tf.get_default_session()
```

```
Out[27]: <tensorflow.python.client.session.InteractiveSession at 0x11a727310>
```



```
In [28]: zero_ = tf.constant(0.0)
        minus_one_ = tf.constant(-1.0)
        c_minus_one_ = tf.constant(-1.0, dtype=tf.complex64)
```

```
In [29]: sess.run(zero_)
```

```
Out[29]: 0.0
```

```
In [30]: minus_one_.eval(), c_minus_one_.eval()
```

```
Out[30]: (-1.0, (-1+0j))
```

0 으로 나눈다고 **exception** 이 바로 발생하지 않습니다

```
In [31]: div_by_zero_ = tf.div(tf.constant(1.0), zero_)
        div_by_zero_.eval()
```

```
Out[31]: inf
```

-1 제곱근을 구해도 **exception** 발생하지 않습니다

```
In [32]: sqrt_minus_one_ = tf.sqrt(tf.constant(-1.0))
        sqrt_minus_one_.eval()
```

```
Out[32]: nan
```

잠깐 주의: 텐서플로우는 복소수 타입도 지원합니다.

- 하지만 모델 작성에 이용하실 때는 복소수 타입이 지원 안되는 연산이 많다는 것에 유의
(<https://stackoverflow.com/questions/42284904/complex-gradients-on-gpu-in-tensorflow>)

```
In [33]: sqrt_c_minus_one_ = tf.sqrt(tf.constant(-1.0, dtype=tf.complex64))
        sqrt_c_minus_one_.eval()
```

```
Out[33]: (7.5497901e-08+1j)
```

0 의 로그 값을 계산해도, 음수의 로그값을 계산해도 ... (중략)

```
In [34]: log_zero_ = tf.log(zero_)
        log_zero_.eval()
```

```
Out[34]: -inf
```

```
In [35]: minus_one_ = tf.constant(-1.0)
log_minus_one_ = tf.log(minus_one_)
log_minus_one_.eval()
```

Out[35]: nan

모델 학습 중에 Inf 또는 NaN 이 발생했다면, 외관상 뭔가 계속 학습하는 것 같아도 결과적으로 쓸모없을 수 있음

- `tf.check_numerics(a_tensor)`
 - `a_tensor` 에 NaN 이나 Inf 값이 들어 있는지 확인. 들어있으면 예외발생.
- `tf.add_check_numerics_op()`
 - 모든 부동 소숫점 텐서에 `tf.check_numerics()` 계산을 추가

```
In [ ]: sess.run([ \
            tf.check_numerics(v, 'check_numerics {}'.format(v.name)) \
            for v in [
                zero_,
                minus_one_,
                div_by_zero_,
                sqrt_minus_one_,
                log_zero_,
                minus_one_]])
```

파이썬 디버거 **pdb** 사용법

- `python -m pdb myscript.py`

```
(Pdb) ?
Documented commands (type help <topic>):
=====
EOF      bt          cont      enable  0[jump]: pp
a         c          continue exit    l      q      s      until
alias    cl         d         h       In [19]: zero_ = tf.constant(0.0)
args     clear      debug    help    n      r      tbreak w
b         commands  disable  ignore In next]: restart(un(div_by_zer
break    condition down     j       Out[20]: return unalias where

Miscellaneous help topics:
=====
exec     pdb

Undocumented commands:
=====
retval  rv

(Pdb) 
```

```
Out[23]: <tensorflow.python.client.session
In [24]: tf.get_default_session()
Out[25]: tensorflow.python.client.session
In [19]: zero_ = tf.constant(0.0)
div_by_zero_ = tf.div(tf.const
Out[20]: return unalias where
In [21]: minus_one_ = tf.constant(-1.0)
log_ = tf.log(minus_one_)
log_.eval()
Out[21]: nan
## python pdb 사용법
### `python -m pdb mysc
In [ ]:
```

- w(here)
- d(own)
- u(p)
- l(ist) [first[, last]]

- s(tep)
- n(ext)
- unt(il)

- c(ontinue)
- r(eturn)

- `b(reak) [[filename:]lineno|function[,condition]]`

Without argument, list all breaks, including for each breakpoint, the number of times that breakpoint has been hit, the current ignore count, and the associated condition if any.

- `disable [bpnumber [bpnumber ...]]`
- `enable [bpnumber [bpnumber ...]]`

- `[!]statement`

Execute the (one-line) statement in the context of the current stack frame. The exclamation point can be omitted unless the first word of the statement resembles a debugger command. To set a global variable, you can prefix the assignment command with a global command on the same line, e.g.:

```
(Pdb) global list_options; list_options = ['-l']
(Pdb)
```

- `p expression`

Note `print` can also be used, but is not a debugger command — this executes the Python `print` statement.

- `pp expression`

- `a(rgs)`

Print the argument list of the current function.

- `run [args ...]`

Restart the debugged Python program. If an argument is supplied, it is split with “`shlex`” and the result is used as the new `sys.argv`. History, breakpoints, actions and debugger options are preserved. “`restart`” is an alias for “`run`”.

- `q(uit)`

주피터 노트북에서도 pdb 를 쓸 수 있습니다

- `debug`

```
In [37]: def function_1(x,y):
          return x / y

def function_2(x,y):
    return function_1(x,y) * x

def function_3(x,y):
    return function_2(x,y) + x

function_3(99,0)
```

```
-----
-----
ZeroDivisionError                                Traceback (most recent c
all last)
<ipython-input-37-161ffdb3c76d> in <module>()
      8     return function_2(x,y) + x
      9
----> 10 function_3(99,0)

<ipython-input-37-161ffdb3c76d> in function_3(x, y)
      6
      7 def function_3(x,y):
----> 8     return function_2(x,y) + x
      9
     10 function_3(99,0)

<ipython-input-37-161ffdb3c76d> in function_2(x, y)
      3
      4 def function_2(x,y):
----> 5     return function_1(x,y) * x
      6
      7 def function_3(x,y):

<ipython-input-37-161ffdb3c76d> in function_1(x, y)
      1 def function_1(x,y):
----> 2     return x / y
      3
      4 def function_2(x,y):
      5     return function_1(x,y) * x

ZeroDivisionError: division by zero
```

```
In [39]: debug
```

```
> <ipython-input-37-161ffdb3c76d>(2)function_1()
      1 def function_1(x,y):
----> 2     return x / y
      3
      4 def function_2(x,y):
      5     return function_1(x,y) * x

ipdb> q
```

텐서플로우 디버거 tfdbg

```
from tensorflow.python import debug as tf_debug
```

파일을 열어 보면 **training** 할 때 다음과 같이 **session** 을 새로 만들어 쓰는 것을 볼 수 있습니다

```
config = tf.ConfigProto(gpu_options={'allow_growth': True})
with tf.Session(config=config) as session:

    session = tf_debug.LocalCLIDebugWrapperSession(session)
    session.add_tensor_filter('has_inf_or_nan', tf_debug.has_inf_or_nan)

    session.run(tf.global_variables_initializer())
</code>
```

실행

```
python mnist_with_tf_debug.py --debug
```

```
--- run-start: run #1: 1 fetch (init); 0 feeds -----  
| list tensors | help |  
=====
```

Session.run() call #1:

Fetch(es):
 init

Feed dict(s):
 (Empty)

=====

Select one of the following commands to proceed ---->

run:
 Execute the run() call with debug tensor-watching

run -n:
 Execute the run() call without debug tensor-watching

run -t <T>:
 Execute run() calls (T - 1) times without debugging, then execute run() one more time and drop back to the CLI

```
--- Scroll (PgDn): 0.00% ----- Mouse: ON --  
tfdbg> |
```

tfdbg> run -t 10

```

--- run-end: run #10: 3 fetches; 2 feeds -----
| list_tensors | node_info | print_tensor | list_inputs | list_outputs | run\_info | help |
113 dumped tensor(s):

t (ms)  Size   Op type                               Tensor name
[0.000] 47    Size                                  Size:0
[0.011] 113   Const                                gradients/sparse\_softmax\_cross\_entropy\_loss/Sum\_grad/Reshape/shape:0
[0.033] 30.69k VariableV2                            Variable:0
[0.035] 93    VariableV2                          Variable\_1:0
[0.046] 49    Cast                                 Cast\_1:0
[0.087] 98    Identity                             Variable\_1/read:0
[0.089] 54    VariableV2                          beta1\_power:0
[0.122] 30.69k Identity                 Variable/read:0
[0.123] 124   Const                                gradients/sparse\_softmax\_cross\_entropy\_loss/xentropy/xentropy\_grad/ExpandDims/dim:0
[0.127] 59    Identity                             beta1\_power/read:0
[0.138] 54    VariableV2                          beta2\_power:0
[0.178] 59    Identity                             beta2\_power/read:0
[0.200] 30.69k VariableV2                Variable/Adam:0

--- Scroll (PgDn): 0.00% ----- Mouse: ON ---
tfdbg>

```

```
tfdbg> print_tensor Variable:0
```



```

--- run-end: run #10: 3 fetches; 2 feeds -----
| list_tensors | node_info | print_tensor | list_inputs | list_outputs | run_info | help |
Tensor "Variable:0:DebugIdentity":
  dtype: float32
  shape: (784, 10)

array([[ -0.16246668, -0.00396837, -0.13555561, ..., -0.07888098, -0.13244194, -0.03601011],
       [ -0.16246668, -0.00396837, -0.13555561, ..., -0.07888098, -0.13244194, -0.03601011],
       [ -0.16246668, -0.00396837, -0.13555561, ..., -0.07888098, -0.13244194, -0.03601011],
       ...,
       [ -0.16246668, -0.00396837, -0.13555561, ..., -0.07888098, -0.13244194, -0.03601011],
       [ -0.16246668, -0.00396837, -0.13555561, ..., -0.07888098, -0.13244194, -0.03601011],
       [ -0.16246668, -0.00396837, -0.13555561, ..., -0.07888098, -0.13244194, -0.03601011]], dtype=float32
)

--- Scroll (PgDn): 0.00% ----- Mouse: ON ---
tfdbg>

```

```
tfdbg> node_info Variable:0
```

```

--- run-end: run #10: 3 fetches; 2 feeds -----
| list tensors | node_info | print_tensor | list inputs | list outputs | run_info | help |
Node Variable

Op: VariableV2
Device: /job:localhost/replica:0/task:0/cpu:0

0 input(s) + 0 control input(s):
  0 input(s):

2 recipient(s) + 0 control recipient(s):
  2 recipient(s):
    [Identity] Variable/read
    [ApplyAdam] Adam/update Variable/ApplyAdam

Node attributes:
  shared_name:
    s: ""

  dtype:
--- Scroll (PgDn): 0.00% ----- Mouse: ON ---
tfdbg>

```

```
tfdbg> list_outputs Variable:0
```

```
--- run-end: run #10: 3 fetches; 2 feeds -----
| list tensors | node info | print tensor | list inputs | list outputs | run info | help |
Recipients of node "Variable" (Depth limit = 20, control recipients included):
|- (1) Variable/read
|   |- (2) MatMul
|   |   |- (3) gradients/add\_grad/Shape
|   |   |   |- (4) gradients/add\_grad/Reshape
|   |   |   |   |- (5) gradients/add\_grad/tuple/control\_dependency
|   |   |   |   |   |- (6) gradients/MatMul\_grad/MatMul
|   |   |   |   |   |   |- (7) (Ctrl) gradients/MatMul\_grad/tuple/group\_deps
|   |   |   |   |   |   |   |- (8) (Ctrl) gradients/MatMul\_grad/tuple/control\_dependency\_1
|   |   |   |   |   |   |   |   |- (9) Adam/update Variable/ApplyAdam
|   |   |   |   |   |   |   |   |   |- (10) (Ctrl) Adam
|   |   |   |   |   |   |   |   |   |   |- (10) (Ctrl) Adam/mul
|   |   |   |   |   |   |   |   |   |   |   |- (11) Adam/Assign
|   |   |   |   |   |   |   |   |   |   |   |   |- (12) (Ctrl) Adam
|   |   |   |   |   |   |   |   |   |   |   |   |   |- (10) (Ctrl) Adam/mul\_1
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |- (11) Adam/Assign\_1
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |- (12) (Ctrl) Adam
|   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |   |- (6) gradients/MatMul\_grad/MatMul\_1
--- Scroll (PgDn): 0.00% ----- Mouse: ON --
tfdbg>
```

tfdbg> run_info

--- run-end: run #10: 3 fetches; 2 feeds -----

| list tensors | help |

=====
Session.run() call #10:

Fetch(es):

Adam

sparse_softmax_cross_entropy_loss/value:0

sub_1:0

Feed dict(s):

input:0

label:0

=====
Select one of the following commands to proceed ---->

run:

Execute the run() call with debug tensor-watching

run -n:

Execute the run() call without debug tensor-watching

--- Scroll (PgDn): 0.00% -----

Mouse: ON --

tfdbg> |

기타 참고자료

- [A Practical Guide for Debugging TensorFlow Codes, Jongwook Choi](#)
- [Jupyter notebooks features](#)
- [IPython built-in magic commands](#)
- `%quickref`, `%magic`