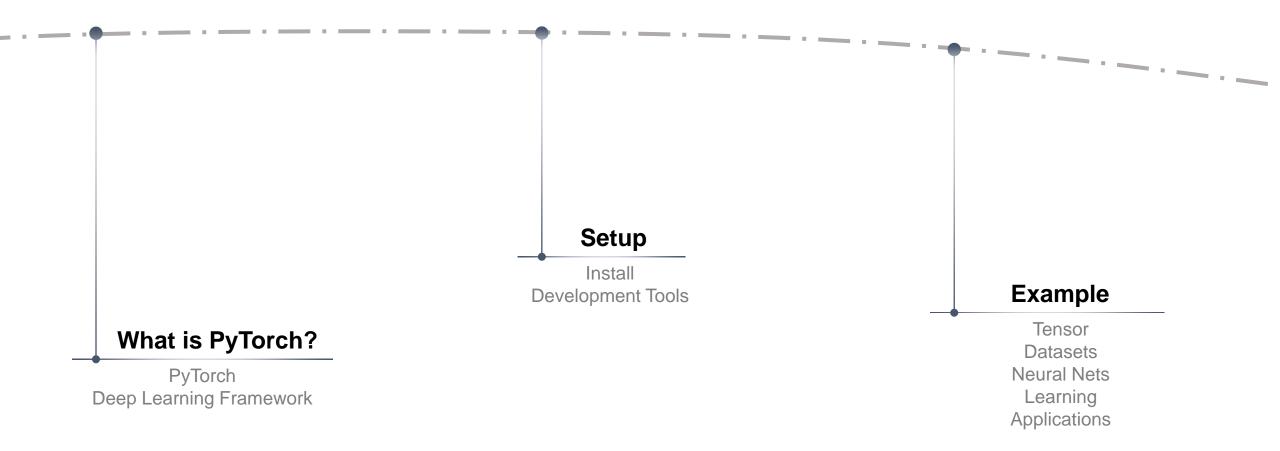
Touch to PyTorch

: From basic to vanilla GAN

ISL Lab Seminar Hansol Kang

Contents





What is PyTorch?

PyTorch







Vs.







What is PyTorch?

Deep Learning Framework



Keras

Nov. 2010

Written in: Python Interface: Python



Written in: C++

Dec. 2013

Interface: Python, MATLAB, C++



Jul. 2014

PYTÖRCH

Mar. 2015

Written in: Python Interface: Python, R **TensorFlow**

Written in: C++, Python, CUDA Nov. 2015

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

Oct. 2016

Written in : Python, C, CUDA

Recommend to choose these framework

Interface: Python

Written in:

Interface: Python, C++ Apr. 2017

DL4J(Java) Chainer(Python) MXNet(C++, Python, Julia, MATLAB, JavaScript, Go, R, Scala, Perl) CNTK(Python, C++), TF Learn(Python)

TF-Slim(Python)

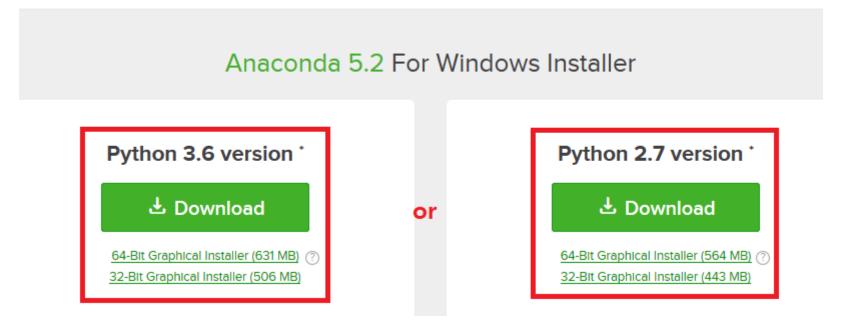
Etc.

"What is this? Gum? It's GAN.", pp. 21-22



Install

https://www.anaconda.com/download/





Install

https://pytorch.org/

Select your preferences and run the install command. Please ensure that you are on the latest pip and numpy packages. Anaconda is our recommended package manager. You can also install previous versions of PyTorch. Note that LibTorch is only available for C++. PyTorch Build Stable Preview Linux Mac Windows Your OS LibTorch Conda Pip Source Package Python 3.6 Python 3.7 Python 2.7 Python 3.5 C++ Language 8.0 9.0 9.2 None CUDA conda install pytorch -c pytorch Run this Command: pip3 install torchvision

C++ : Preview, Linux only



Install

conda create -n PyTorch python=3.6 activate PyTorch conda install pytorch cuda90 -c pytorch pip install torchvision

- 。 **conda create**는 환경 생성하는 명령어. PyTorch 뿐만 아니라 Tensorflow 같은 다른 딥러닝 프레임워크를 사용한다거나 하위 파이썬 버전도 사용해야하는 경우 환경마다 설정해주면, 디펜던시가 꼬이지 않음.
- 。 -n 환경명, python=파이썬버전 입력. 환경설정 리스트는 conda env list를 입력하면 확인 가능.
- 。 activate는 해당 환경을 활성화 시키는 명령어. 반대로 환경을 빠져나오는 명령어는 deactivate.
- 。 PyTorch를 설치하는 명령어는 conda install pytorch cuda90 -c pytorch..
- 。 torchvision은 딥러닝 학습에 많이 사용되는 데이터셋, 네트워크 구조, 이미지 변환과 같은 기능을 제공하므로 설치하는 것을 권장.



Development Tools



: Intellisense

: Cell based execution

: Intellisense

: Cell based execution

: Management of python env.



: Intellisense

: Management of python env.

: GitHub

: Al tool package

: Intellisense

: Environment setting



: Insane extension program

: Startup file



Image System Laboratory

Development Tools



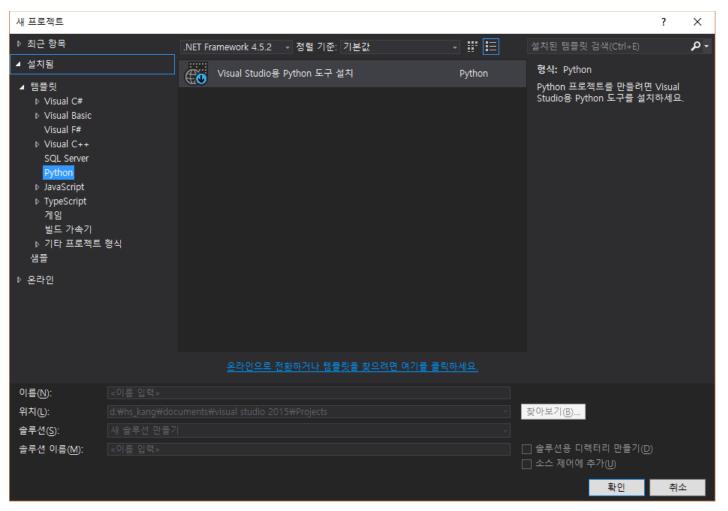






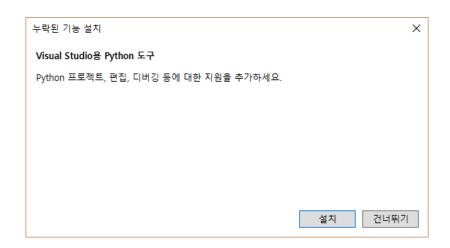
Development Tools – Visual Studio

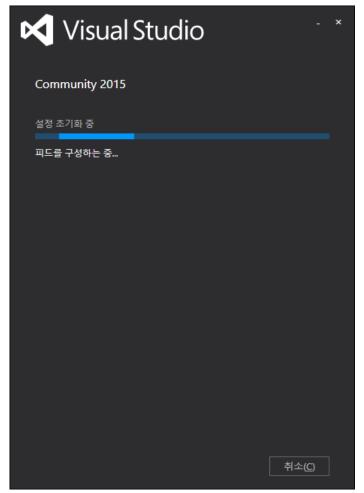
File-New-Project



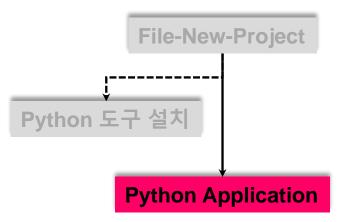


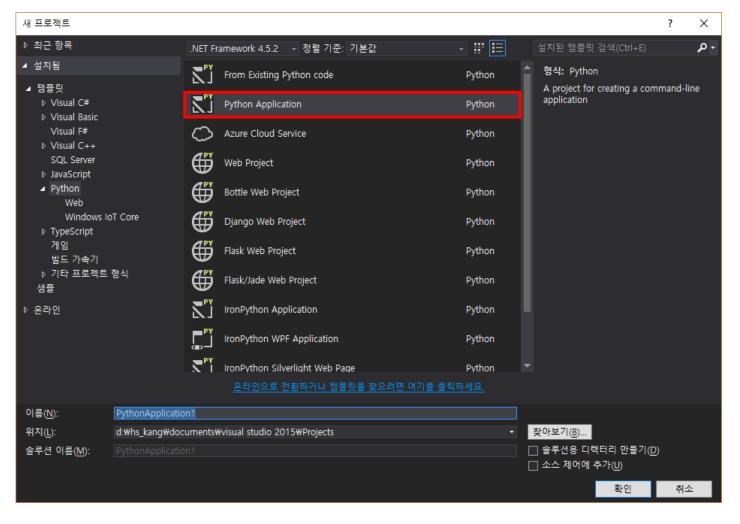




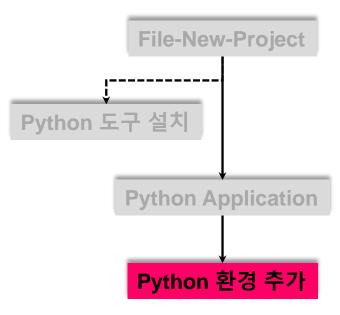












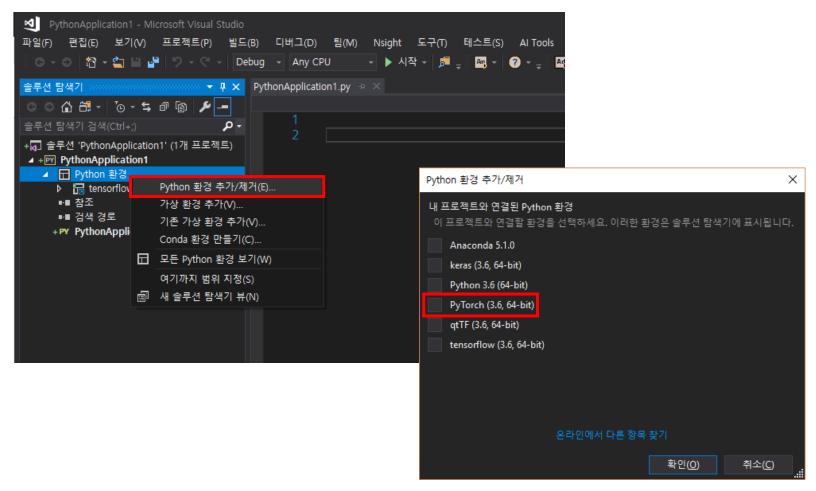
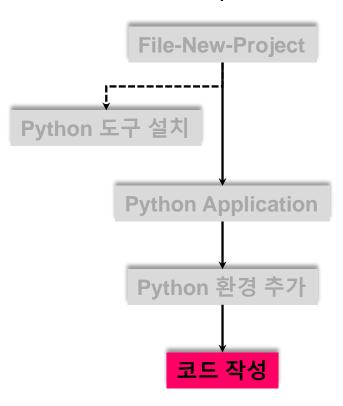
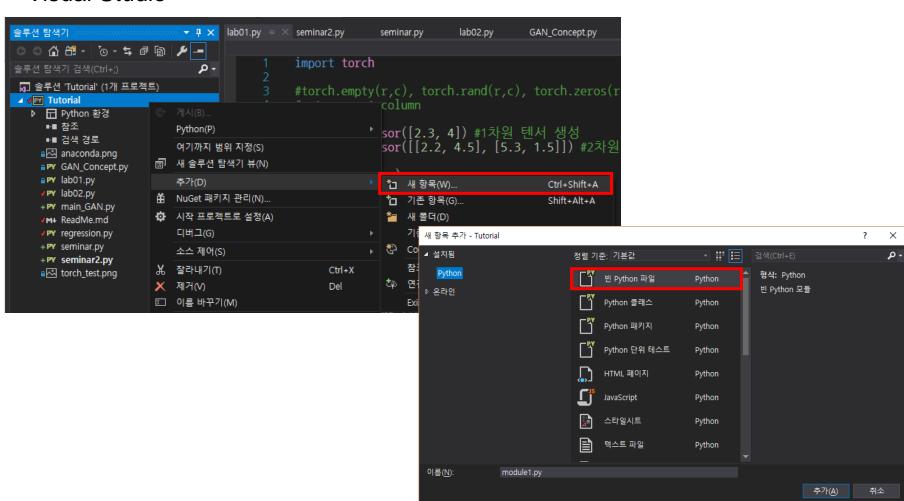


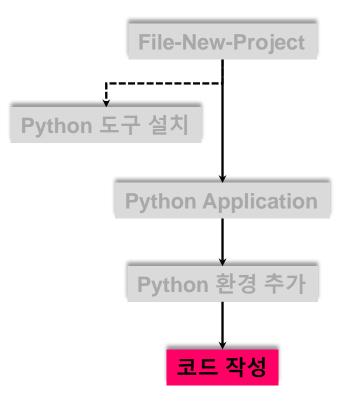


Image System Laboratory

Setup



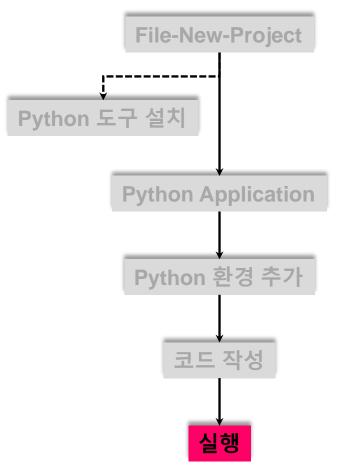




```
lab01.py + X seminar2.py
                                     lab02.py
                                                GAN_Concept.py
                        seminar.py
                                                                 regression.py
          import torch
          #torch.empty(r,c), torch.rand(r,c), torch.zeros(r,c,dtype)
          #r : row, c: column
          a = torch.tensor([2.3, 4]) #1차원 텐서 생성
b = torch.tensor([[2.2, 4.5], [5.3, 1.5]]) #2차원 텐서 생성
          print("a : ", a)
          print("size of a : ", a.size(), "\n")
          print("b : ", b)
          print("size of b : ", b.size(), "\n")
          x = a.new_ones(5,3,dtype=torch.double) #텐서 재사용
          print("x : ", x)
          #텐서를 넘겨 받음. dtype override
          x = torch.randn_like(x, dtype=torch.float)
          print("print : ", x)
          #y.add_(x), 언더바 붙이면 in-place 자기 자신
          c = torch.rand(2,2)
          c.add_(b)
          print(c)
```

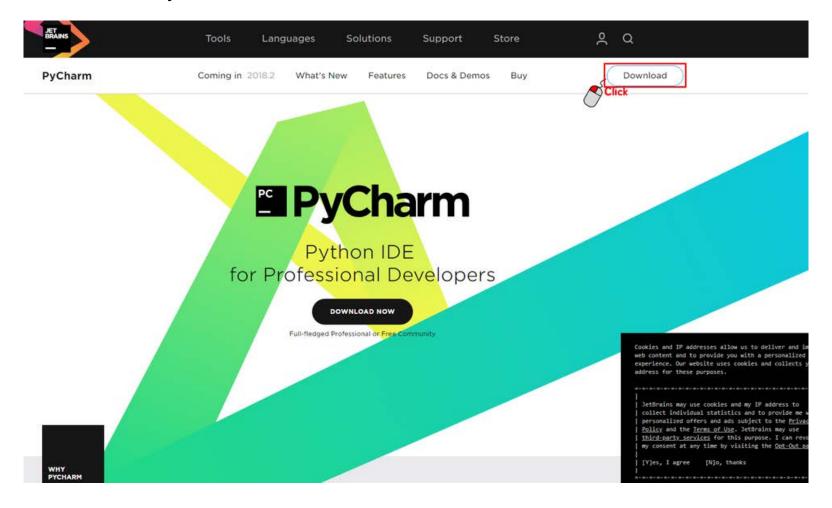


Development Tools – Visual Studio

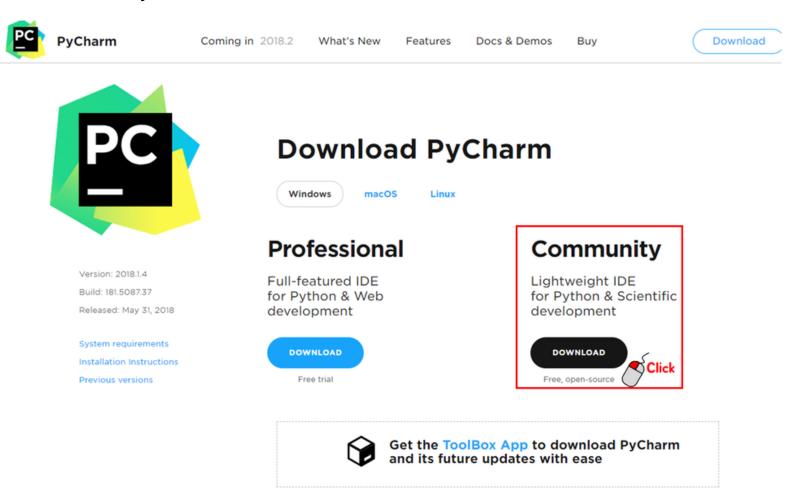


: 시작 파일로 설정-> Ctrl+F5



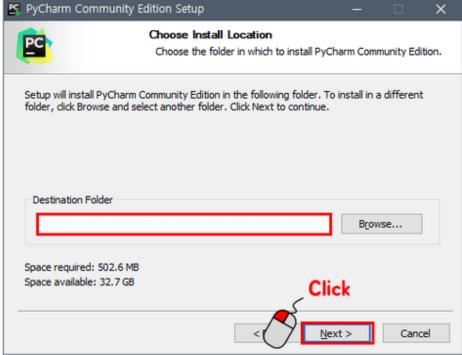




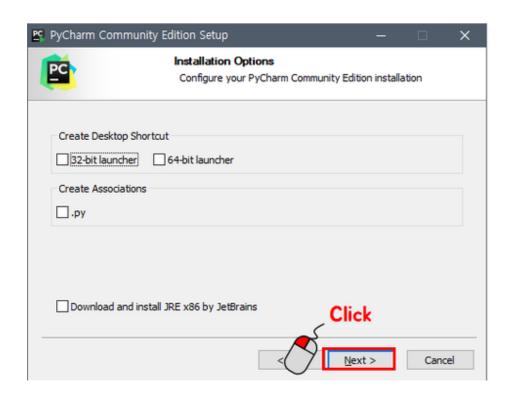


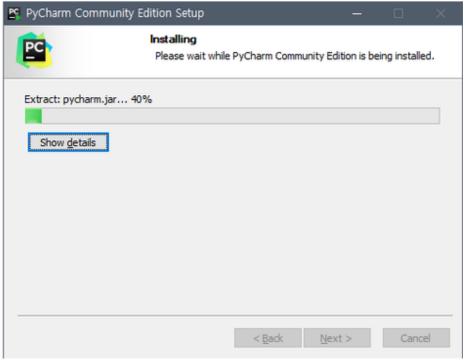




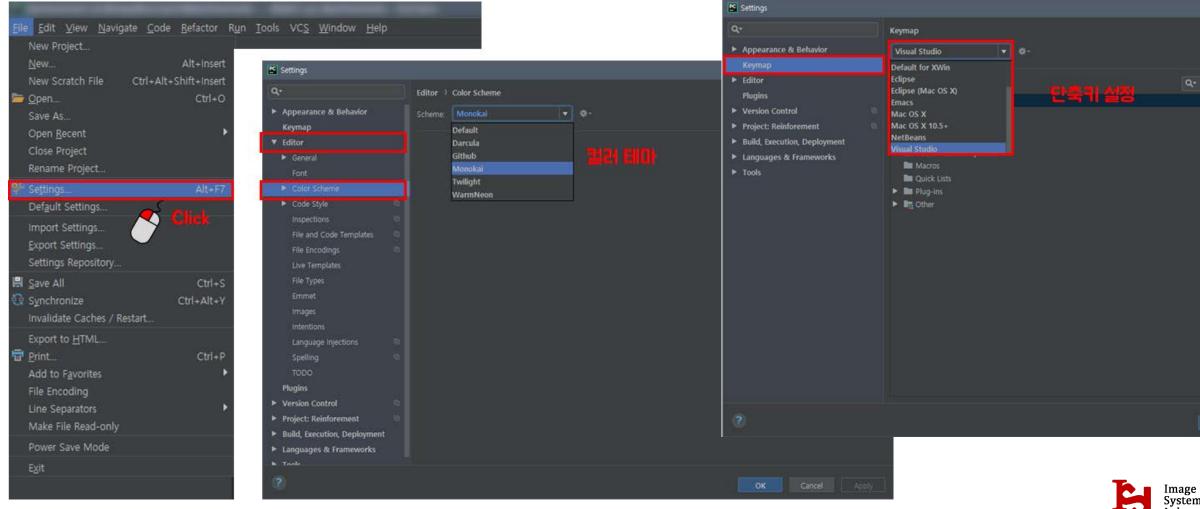






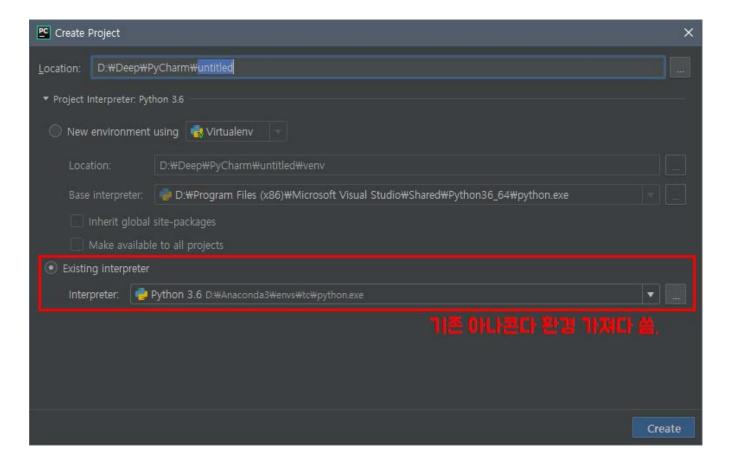






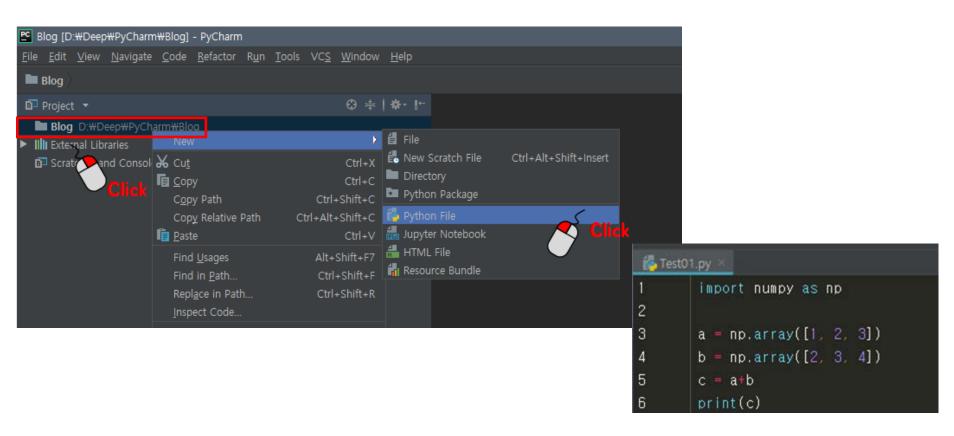
Development Tools – PyCharm

File-New Project









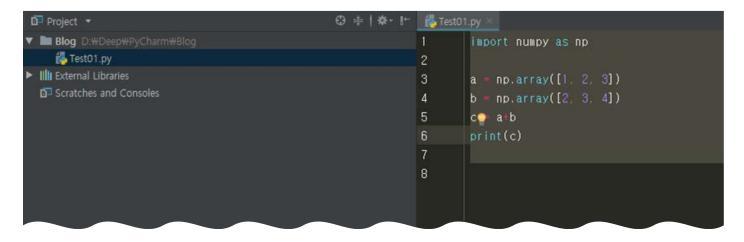


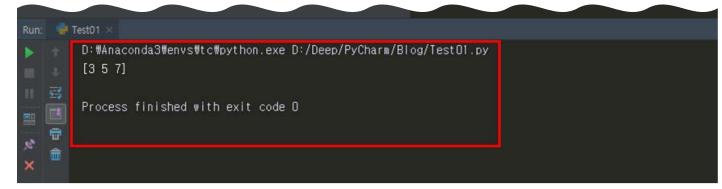
Development Tools – PyCharm



: Ctrl+F5

"Too easy, but important things", pp. 42-43







Tensor



random.randn, random.rand



reshape

linspace

ones, zeros

Tensor

view(reshape)

linspace

ones, zeros

randn, rand

to, cuda



Tensor

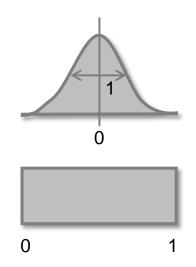
```
Tensor a:
import torch as to
                                               tensor([ 2.5000, 4.0000])
                                               Size of a: torch.Size([2])
#Tensor
                                               Tensor b:
                                               tensor([[ 1.0000, 2.5000],
a = tc.Tensor([2.5, 4])
                                                   [ 2.5000, 6.0000]])
b = tc.Tensor([[1, 2.5], [2.5, 6]])
                                               Size of b: torch.Size([2, 2])
                                               Tensor c:
                                               tensor([[ 1.0000, 2.5000, 2.5000, 6.0000]])
#reshape
                                               Size of c: torch.Size([1, 4])
c = b.reshape(1, 4)
                                               Tensor d:
d = b.reshape(1, -1)
                                                tensor([[ 1.0000, 2.5000, 2.5000, 6.0000]])
                                               Size of d: torch.Size([1, 4])
```



Tensor Tensor x: tensor([-1.0000, -0.7500, -0.5000, #linspace -0.2500, 0.0000, 0.2500, 0.5000, x = tc.linspace(-1, 1, 9)0.7500, 1.0000]) Start Size of x : torch.Size([9]) Tensor real: tensor([[1., 1.], [1., 1.]]) #ones, zeros Size of real: torch.Size([2, 2]) real = tc.ones(2, 2)Tensor fake: fake = tc.zeros(2, 2)tensor([[0., 0.], [0., 0.]]) Size of fake: torch.Size([2, 2])



Tensor



```
#to, cuda 

z_cuda = z1.cuda() 

device = tc.device('cuda' if tc.cuda.is_available() else 'cpu') 

z_device = z1.to(device)

Tensor z_cuda: 

tensor([[-0.0750, -1.0555, -0.0706], 

[-0.1946, -0.8593, -0.2238]], device='cuda:0')
```

Image System Laboratory

Tensor

```
#from_numpy, numpy import numpy as np

a = np.array([3.5, 4])
b = tc.from_numpy(a)
c = b.numpy()

array a :

[3.5 4. ]

Tensor b :

tensor([ 3.5000, 4.0000], dtype=torch.float64)

array c :

[3.5 4. ]
```



Datasets

The following datasets are available:

✓ : 바로 사용 가능.

✓ : 추가적인 과정 필요.

✓ : Custom dataset에 활용.

- **✓** MNIST
- ▼ Fashion-MNIST

EMNIST

COCO

- ✓ LSUN
- ✓ ImageFolder

DatasetFolder

Imagenet-12

V CIFAR

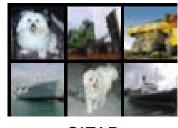
STL10

SVHN

PhotoTour







MNIST

Fashion-MNIST

CIFAR



LSUN



CelebA



K-pop(Custom)



Datasets – MNIST

```
import torch as to
import torchvision as tv
import torchvision.transforms as transforms
```

trans = transforms.Compose([transforms.ToTensor(), transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))])
datasets = tv.datasets.MNIST(root='./MNIST', train=True, download=True, transform=trans)
dataloader = tc.utils.data.DataLoader(datasets=datasets, batch_size=100, shuffle=True)

trans

- Compose() is used when there are multiple transform options. Here, ToTensor() and Normalize(mean, std) are used.
- ToTensor () changes the PIL Image to a tensor. torchvision dataset The default type is PIL Image.
- Normalize (mean, std) transforms the range of the image. Here, the value of [0, 1] is adjusted to [-1, 1]. ((value-mean) / std)



Datasets – MNIST

```
import torch as to
import torchvision as tv
import torchvision.transforms as transforms
```

```
trans = transforms.Compose([transforms.ToTensor(), transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))]) datasets = tv.datasets.MNIST(root='./MNIST', train=True, download=True, transform=trans) dataloader = tc.utils.data.DataLoader(datasets=datasets, batch_size=100, shuffle=True)
```

dataset

- root: This is the path to store (MNIST data). Folders are automatically created with the specified name.
- train: Set the data to be used for the train.
- transform: Transform the data according to the transform option set previously.
- download : Download (MINST data). (If you downloaded it once, it will not do it again.)



Datasets – MNIST

```
import torch as to
import torchvision as tv
import torchvision.transforms as transforms
```

```
trans = transforms.Compose([transforms.ToTensor(), transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))]) datasets = tv.datasets.MNIST(root='./MNIST', train=True, download=True, transform=trans) dataloader = tc.utils.data.DataLoader(datasets=datasets, batch_size=100, shuffle=True)
```

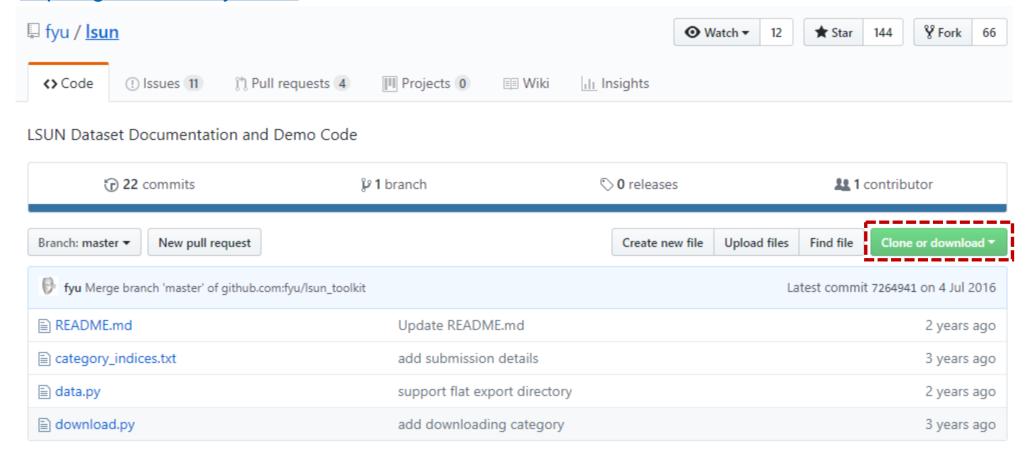
dataloader

- dataset : Set the dataset to load.
- batch_size : Set the batch size.
- shuffle: Shuffle the data and load it.



Datasets – LSUN

https://github.com/fyu/lsun





Datasets – LSUN

Run download.py download.py -c bedroom

bedroom 0
bridge 1
church_outdoor 2
classroom 3
conference_room 4
dining_room 5
kitchen 6
living_room 7
restaurant 8
tower 9

If you are using Python 3.0 or later, modify the code from urllib2.urlopen (url) to urlopen (url).

```
def list_categories(tag):
    url = 'http://lsun.cs.princeton.edu/htbin/list.cgi?tag=' + tag
    f = urlopen(url)
    return json.loads(f.read())
```



Datasets – LSUN

```
trans = transforms. Compose([transforms.Resize((64,64)) , transforms. To Tensor(), transforms. Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))]) \\ datasets = tv.datasets. LSUN('.', classes=['bedroom_train'], transform=trans) \\ dataloader = tc.utils.data.DataLoader(datasets=datasets, batch_size=100, shuffle=True)
```

trans

Resize() is used to resize the image.

datasets

- root: Root directory for the database files.
- classes: One of {'train', 'val', 'test'} or a list of categories to load. e,g. ['bedroom_train', 'church_train'].



Datasets – ImageFolder(CelebA)

http://mmlab.ie.cuhk.edu.hk/projects/CelebA.html

News

2016-07-29 If DropBox are not accessable, please download the dataset using Google Drive or Baid

Details

CelebFaces Attributes Dataset (CelebA) is a large-scale face attributes dataset with more than 200K celebrity images, each with 40 attribute annotations. The images in this dataset cover large pose variations and background clutter. CelebA has large diversities, large quantities, and rich annotations, including

- 10,177 number of identities.
- 202,599 number of face images, and
- 5 landmark locations, 40 binary attributes annotations per image.

The dataset can be employed as the training and test sets for the following computer vision tasks: face attribute recognition, face detection, and landmark (or facial part) localization.

Sample Images







Wearing





Datasets – ImageFolder(CelebA)

http://mmlab.ie.cuhk.edu.hk/projects/CelebA.html



System Laboratory

Datasets – ImageFolder(CelebA)

```
trans = transforms.Compose([transforms.Resize((64,64)) ,transforms.ToTensor(), transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))])
datasets = tv.datasets.ImageFolder('./img_align_celeba', trans)
dataloader = tc.utils.data.DataLoader(datasets=datasets, batch_size=100, shuffle=True)
```

datasets

- root: Root directory for the database files.
- transform: Transform the data according to the transform option set previously.



Neural Nets

Class base

```
class Model(tc.nn.Module):
    def __init__(self):
        super(Model, self).__init__()
        self.linear1 = tc.nn.Linear(D_in, H),
        self.linear2 = tc.nn.Linear(H, D_out)

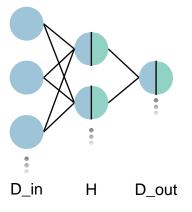
def forward(self, input):
    x = tc.nn.functional.relu(self.linear1(input))
    x = tc.nn.functional.sigmoid(self.linear2(x))

    return x
```

Data manipulation

Sequence base

```
model = tc.nn.Sequential(
   tc.nn.Linear(D_in, H),
   tc.nn.ReLU(),
   tc.nn.Linear(H, D_out),
   tc.nn.Sigmoid()
)
```



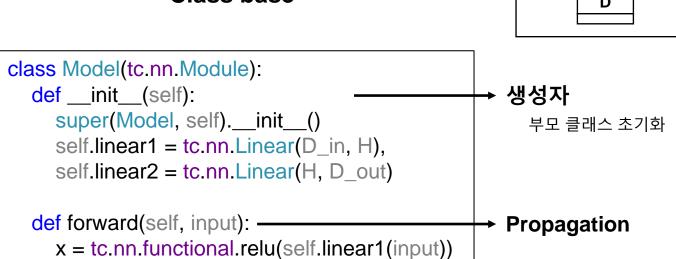


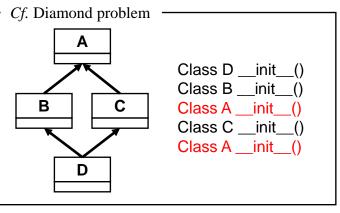
Neural Nets

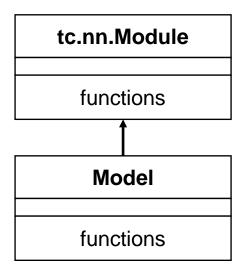
return x

Class base

x = tc.nn.functional.sigmoid(self.linear2(x))

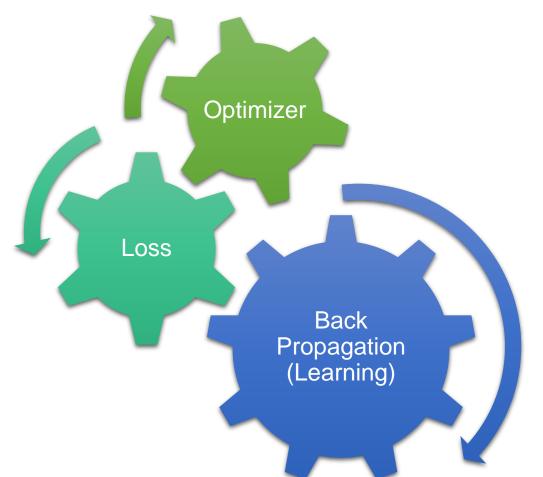








Learning



```
loss_func = tc.nn.MSELoss()
opt = tc.optim.Adam(model.parameters(), Ir=0.01)
for ep in range(epoch_sz):
  for step, (images, labels) in enumerate(dataloader):
    opt.zero_grad()
    images = images.to(device)
    labels = labels.to(device)
    output = model(images)
    loss = loss_func(output)
    loss.backward()
    opt.step()
```



Learning

cf.

trans = transforms.Compose([transforms.ToTensor(), transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5))])
datasets = tv.datasets.MNIST(root='./MNIST', train=True, download=True, transform=trans)
dataloader = tc.utils.data.DataLoader(datasets=datasets, batch_size=100, shuffle=True)

```
→ Loss 선언
loss_func = tc.nn.MSELoss()
opt = tc.optim.Adam(model.parameters(), Ir=0.01)-
                                                     → Optimizer 선언 (업데이트 하려는 parameter)
for ep in range(epoch_sz):-
                                                    ──▶ 정해진 epoch 만큼 수행
  for step, (images, labels) in enumerate(dataloader): -
                                                     → 앞서 설정한 dataloader에 따라 image(data)와 label(_) 불러옴
    opt.zero grad() -
                                                     → Gradient 초기화.
    images = images.to(device)
    labels = labels.to(device)
    output = model(images)
    loss = loss_func(output) -
                                                      → Loss 구함
    loss.backward() —
                                                     → Back propagation(Gradient 구함)
    opt.step() —
                                                     → 정해진 optimizer에 따라 parameter 업데이트
```



Learning

Loss

- L1Loss
- MSELoss
- CrossEntropyLoss
- NLLLoss
- PoissonNLLLoss
- KLDivLoss
- BCELoss
- BCEWithLogitsLoss
- MarginRankingLoss
- HingeEmbeddingLoss

- SmoothL1Loss
- SoftMarginLoss
- MultiLabelSoftMarginLoss
- CosineEmbeddingLoss
- MultiMarginLoss
- TripleMarginLoss

Optimizer

- Adadelta
- Adagrad
- Adam
- SparseAdam
- Adamax
- ASGD
- LBFGS
- RMSprop
- Rprop
- SGD

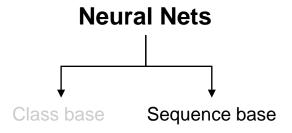


^{*} Bold체는 주로 사용하는 함수들.

Applications –Vanilla GAN

"What is this? Gum? It's GAN."







```
loss_func = tc.nn.BCELoss()
d_opt = tc.optim.Adam(D.parameters(), lr=lr)
g_opt = tc.optim.Adam(G.parameters(), lr=lr)
```



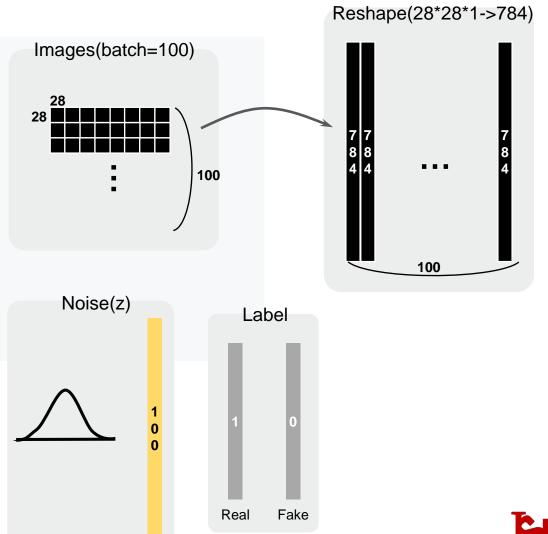
```
for ep in range(nEpoch):
    for step, (images, ) in enumerate(dataloader):
        images = images.reshape(batch sz, -1).to(device)
        z - tc.randn(batch_sz, noise_sz).to(device)

        real_label = tc.ones(batch_sz, 1).to(device)
        fake_label = tc.zeros(batch_sz, 1).to(device)

        loss_real = loss_func(D(images), real_label)
        loss_fake = loss_func(D(G(z)), fake_label)

        d_loss = loss_real + loss_fake

        d_opt.zero_grad()
        d_loss.backward()
        d_opt.step()
```





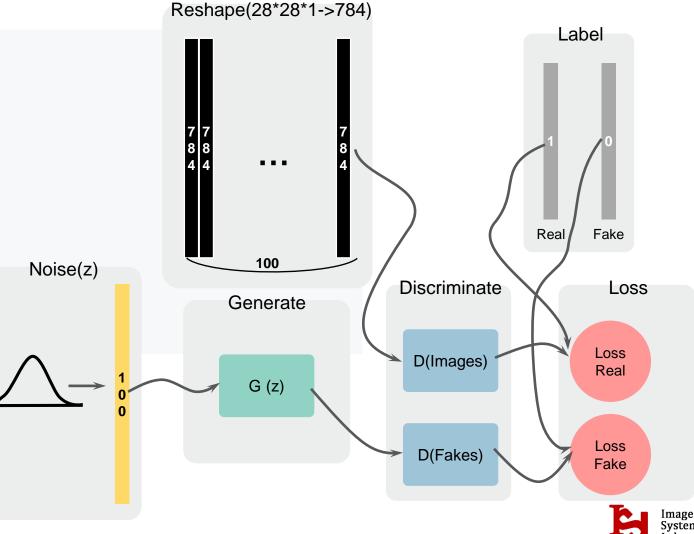
```
for ep in range(nEpoch):
    for step, (images, _) in enumerate(dataloader):
        images = images.reshape(batch_sz, -1).to(device)
        z = tc.randn(batch_sz, noise_sz).to(device)

        real_label = tc.ones(batch_sz, 1).to(device)
        fake_label = tc.zeros(batch_sz, 1).to(device)

        loss_real = loss_func(D(images), real_label)
        loss_fake = loss_func(D(G(z)), fake_label)

        d_loss = loss_real + loss_fake

        d_opt.zero_grad()
        d_loss.backward()
        d_opt.step()
```

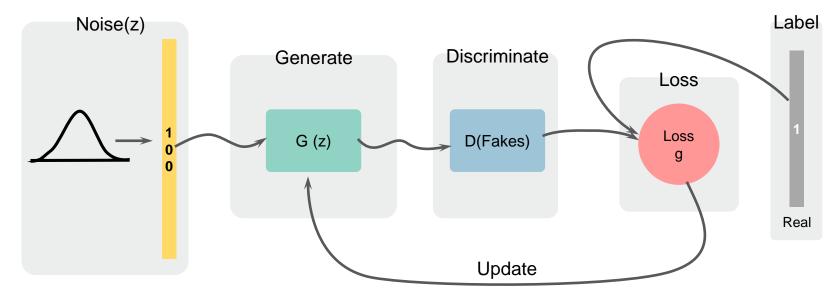


```
for ep in range(nEpoch):
   for step, (images, ) in enumerate(dataloader):
       images = images.reshape(batch_sz, -1).to(device)
        z = tc.randn(batch_sz, noise_sz).to(device)
                                                                        Discriminate
                                                                                                           Loss
       real_label = tc.ones(batch_sz, 1).to(device)
       fake_label = tc.zeros(batch_sz, 1).to(device)
                                                                                                  Loss
       loss_real = loss_func(D(images), real_label)
                                                                         D(Images)
                                                                                                  Real
       loss_fake = loss_func(D(G(z)), fake_label)
                                                                                                                     Total
       d_loss = loss_real + loss_fake
                                                                                                                     Loss
                                                                                                 Loss
        d_opt.zero_grad()
                                                                          D(Fakes)
                                                                                                 Fake
        d_loss.backward()
        d_opt.step()
                                                                                                  Update
```



```
fake_images = G(z)
g_loss = loss_func(D(fake_images), real_label)

g_opt.zero_grad()
g_loss.backward()
g_opt.step()
```





Future Work

