

▼ torch 사용 흐름

1. data preprocessing
 2. data load (pipe line) ? 제너레이터와 같은 방식인가
 3. define model
 4. loss, optimizer define
 5. train, test
- for i in epoch: for j in batch:

```
optimizer.zero_grad()
out = model(in)
loss = loss_func(out, label)
loss.backward()
optimizer.step()
```

```
1 import torch
2 import torch.nn as nn
3 import torch.optim as optim
4 import torchvision.datasets as dset
5 import torchvision.transforms as transforms
6 from torch.utils.data import DataLoader
7 from torch.autograd import Variable
8 import matplotlib.pyplot as plt
9 %matplotlib inline
10 import numpy as np
11 #device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
```

- 1.MNIST train, test dataset 가져오기

[illegible]

Downloading <http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz>
Downloading <http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz> to MNIST/raw/train-images-idx3-ubyte.gz
Failed to download (trying next):
HTTP Error 503: Service Unavailable

Downloading <https://ossci-datasets.s3.amazonaws.com/mnist/train-images-idx3-ubyte.gz>
Downloading <https://ossci-datasets.s3.amazonaws.com/mnist/train-images-idx3-ubyte.gz> to MNIST/raw/train-images-idx3-ubyte.gz
9913344/? [00:05<00:00, 1776131.00it/s]

Extracting MNIST/raw/train-images-idx3-ubyte.gz to MNIST/raw

Downloading <http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz>
Failed to download (trying next):
HTTP Error 503: Service Unavailable

Downloading <https://ossci-datasets.s3.amazonaws.com/mnist/train-labels-idx1-ubyte.gz>
Downloading <https://ossci-datasets.s3.amazonaws.com/mnist/train-labels-idx1-ubyte.gz> to MNIST/raw/train-labels-idx1-ubyte.gz
29696/? [00:01<00:00, 22711.95it/s]

Extracting MNIST/raw/train-labels-idx1-ubyte.gz to MNIST/raw

Downloading <http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz>
Downloading <http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz> to MNIST/raw/t10k-images-idx3-ubyte.gz
Failed to download (trying next):
HTTP Error 503: Service Unavailable

Downloading <https://ossci-datasets.s3.amazonaws.com/mnist/t10k-images-idx3-ubyte.gz>
Downloading <https://ossci-datasets.s3.amazonaws.com/mnist/t10k-images-idx3-ubyte.gz> to MNIST/raw/t10k-images-idx3-ubyte.gz
1649664/? [00:02<00:00, 804549.11it/s]

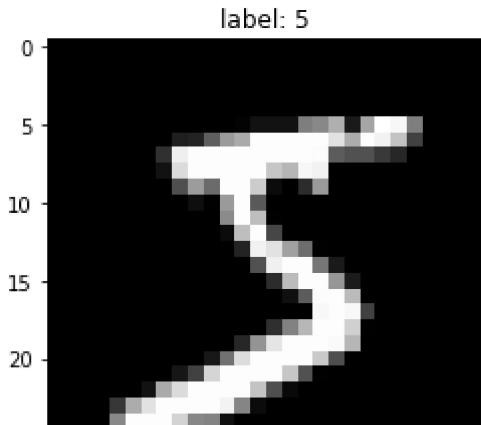
Extracting MNIST/raw/t10k-images-idx3-ubyte.gz to MNIST/raw

Downloading <http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz>
Failed to download (trying next):
HTTP Error 503: Service Unavailable

2. 대략적인 데이터 형태

```
1 print("mnist_train len:", len(mnist_train))
2 print("mnist_test len:", len(mnist_test))
3
4 Image, label = mnist_train.__getitem__(0)
5 print("Image data shape: ", Image.size())
6 print("label: ", label)
7
8 img = Image.numpy()
9 plt.title("label: %d"%label)
10 plt.imshow(img[0], cmap = 'gray')
11 plt.show()
```

```
mnist_train len: 60000
mnist_test len: 10000
Image data shape: torch.Size([1, 28, 28])
label: 5
```



▼ 3. 데이터 로드함수

학습시킬 때 batch_size 단위로 끊어서 로드하기 위함

```
1 batch_size = 1024
2 learning_rate = 0.01
3 num_epoch = 400
```

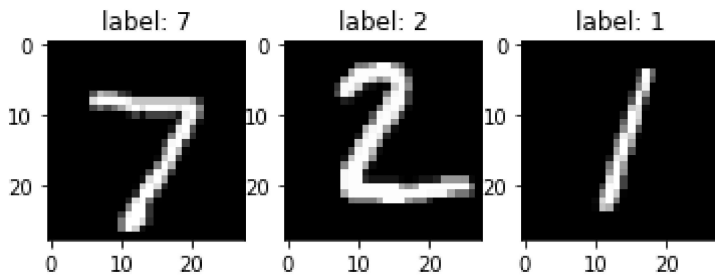
```
1 train_loader = torch.utils.data.DataLoader(mnist_train,
2                                             batch_size=batch_size,
3                                             shuffle=True, num_workers=2, # num_workers는 cpu 코어 개수
4                                             drop_last=True)
5 test_loader = torch.utils.data.DataLoader(mnist_test,
6                                           batch_size=batch_size,
7                                           shuffle=False, num_workers=2,
8                                           drop_last=True)
```

데이터 로드함수 이해하기

```
1 n = 3
2 for i, [imgs, labels] in enumerate(test_loader):
3     if i > 5:
4         break
5
6     print("[%d]" % i)
7     print("한 번에 로드되는 데이터 크기: ", len(imgs))
8
9     for j in range(n):
10         img = imgs[j].numpy()
11         img = img.reshape((img.shape[1], img.shape[2]))
12
13         plt.subplot(1, n, j+1)
14         plt.imshow(img, cmap='gray')
15         plt.title("label: %d" % labels[j])
16     plt.show()
```

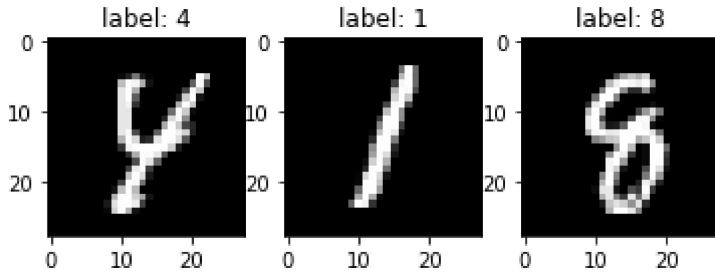
[0]

한 번에 로드되는 데이터 크기: 1024



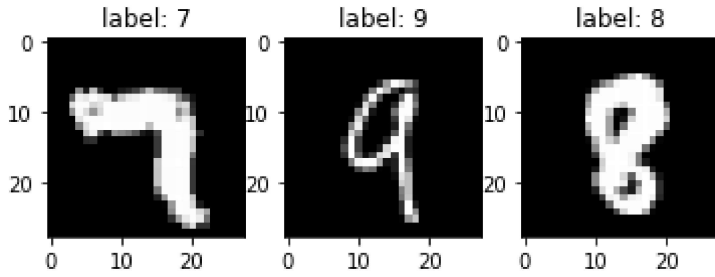
[1]

한 번에 로드되는 데이터 크기: 1024



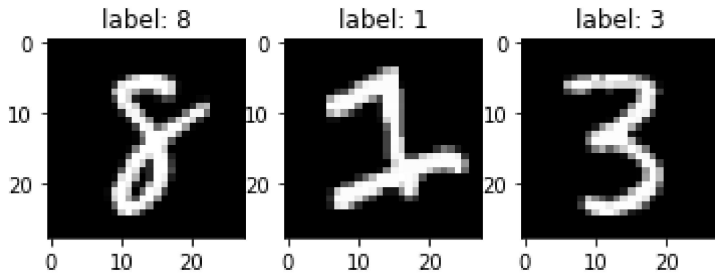
[2]

한 번에 로드되는 데이터 크기: 1024



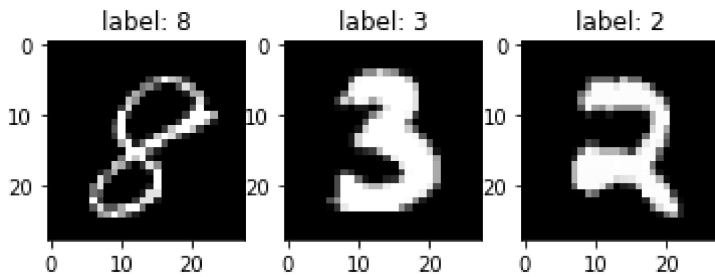
[3]

한 번에 로드되는 데이터 크기: 1024



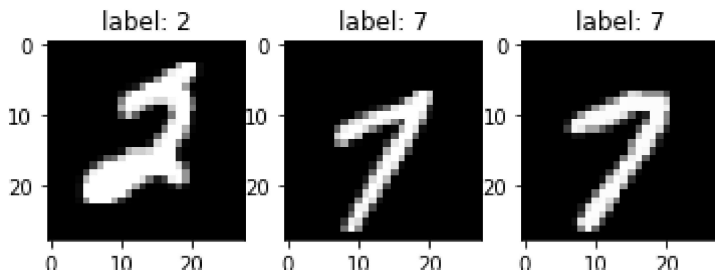
[4]

한 번에 로드되는 데이터 크기: 1024



[5]

한 번에 로드되는 데이터 크기: 1024



▼ 4.모델 선언

```
1 model = nn.Sequential(  
2     nn.Linear(28*28,256),  
3     nn.Sigmoid(),  
4     nn.Linear(256,128),  
5     nn.Linear(128,10),  
6 )  
  
1 def ComputeAccr(dloader, imodel):  
2     correct = 0  
3     total = 0  
4  
5     for j, [imgs, labels] in enumerate(dloader):  
6         img = imgs  
7         label = Variable(labels)  
8  
9         img = img.reshape((img.shape[0],img.shape[2], img.shape[3]))  
10  
11        img = img.reshape((img.shape[0],img.shape[1]*img.shape[2]))  
12        img = Variable(img, requires_grad=False)  
13        output = imodel(img)  
14  
15        _, output_index = torch.max(output, 1)  
16  
17        total += label.size(0)  
18        correct += (output_index == label).sum().float()  
19    print("Accuracy of Test Data: {}".format(100*correct/total))
```

```
1 ComputeAccr(test_loader,model)
```

Accuracy of Test Data: 9.765625

▼ 5.loss, optimizer

```
1 loss_func = nn.CrossEntropyLoss()  
2 optimizer = optim.SGD(model.parameters(), lr=learning_rate)
```

▼ 6.학습

```
1 %time  
2 num_epoch = 400  
3 for j in range(num_epoch):  
4     for i ,[imgs, labels] in enumerate(train_loader):  
5         img = imgs  
6         label = Variable(labels)  
7
```

```

8         img = img.reshape((img.shape[0],img.shape[2], img.shape[3]))
9
10        img = img.reshape((img.shape[0],img.shape[1]*img.shape[2]))
11        img = Variable(img, requires_grad=True)
12
13        optimizer.zero_grad()
14        output = model(img)
15        loss = loss_func(output, label)
16
17        loss.backward()
18        optimizer.step()
19
20    if j%50 == 0:
21        print("%d.."%j)
22        ComputeAccr(test_loader, model)
23        print(loss)

```

CPU times: user 4 μ s, sys: 0 ns, total: 4 μ s

Wall time: 8.34 μ s

0..

Accuracy of Test Data: 11.295573234558105

tensor(2.2978, grad_fn=<NLLossBackward>)

50..

Accuracy of Test Data: 80.67491149902344

tensor(0.7203, grad_fn=<NLLossBackward>)

100..

Accuracy of Test Data: 88.37890625

tensor(0.4582, grad_fn=<NLLossBackward>)

150..

Accuracy of Test Data: 89.96310424804688

tensor(0.3168, grad_fn=<NLLossBackward>)

200..

Accuracy of Test Data: 90.69010162353516

tensor(0.3112, grad_fn=<NLLossBackward>)

250..

Accuracy of Test Data: 91.17838287353516

tensor(0.3241, grad_fn=<NLLossBackward>)

300..

Accuracy of Test Data: 91.61241149902344

tensor(0.3024, grad_fn=<NLLossBackward>)

350..

Accuracy of Test Data: 91.82942962646484

tensor(0.3025, grad_fn=<NLLossBackward>)

▼ 7.테스트

```
1 ComputeAccr(test_loader, model)
```

Accuracy of Test Data: 91.90538024902344

▼ 8.학습된 파라미터 저장

```
1 # netname=''
```

```
2 # torch.save(model, netname)
```

```
2 # torch.save(model, netname, /
3 # #model = torch.load(netname)
```

9.(Optional) 실습1에 쓰인 .npz만드려면?

```
1 # np.savez_compressed('',
2 #                      W1=W1, b1=b1,
3 #                      W2=W2, b2=b2,
4 #                      W3=W3, b3=b3,
5 #                      )
```

✓ 0초 오후 11:17에 완료됨



reCAPTCHA 서비스에 연결할 수 없습니다. 인터넷 연결을 확인한 후 페이지를 새로고침하여 reCAPTCHA 보안문자를 다시 로드하세요.