```
import numpy as np
import torch
import torch.nn as nn
import torch.optim as optim
import torch.nn.init as init
import torchvision.datasets as dset
import torchvision.transforms as transforms
from torch.utils.data import DataLoader
from torch.autograd import Variable
import matplotlib.pyplot as plt
from torch.optim import Ir_scheduler
batch_size=16
learning_rate=0.002
num_epoch=1
##cifar_train=dset.CIFAR10("CIFAR10/",train=True, transform=transforms.ToTensor(), target_transform
cifar_train=dset.CIFAR10("CIFAR10/",train=True, transform=transforms.Compose([transforms.ToTensor()
'''cifar_train=dset.CIFAR10("CIFAR10/",train=True, transform=transforms.Compose([transforms.Scale(@
                                                                               transforms.CenterCrop
                                                                                transforms.RandomHori
                                                                                transforms.Lambda(lam
                                                                                transforms.ToTensor()
                                                                                transforms.Normalize(
                                                                               ]))'''
##cifar_test=dset.CIFAR10("CIFAR10/",train=False, transform=transforms.ToTensor(), target_transform
cifar_test=dset.CIFAR10("CIFAR10/",train=False, transform=transforms.Compose([transforms.ToTensor()
     Downloading https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz to CIFAR10/cifar-10-python.tar.gz
                                                  170499072/? [00:14<00:00, 11510194.82it/s]
     Extracting CIFAR10/cifar-10-python.tar.gz to CIFAR10/
     Files already downloaded and verified
print("cifar_train 길이:", len(cifar_train))
print ("cifar_test 길이:", len(cifar_test))
image, label = cifar_train.__getitem__(1)
print ("image data 형태:", image.size())
print ("label:", label)
img = image.numpy()
```

 $r \circ h = ima[0 : :] ima[1 : :] ima[2 : :]$

```
note (1).ipynb - Colaboratory
img2 = np.zeros((img.shape[1], img.shape[2], img.shape[0]))
img2[:,:,0], img2[:,:,1], img2[:,:,2] = r,g,b
plt.title("label:%d" %label )
plt.imshow(img2,interpolation='bicubic')
plt.show()
     Clipping input data to the valid range for imshow with RGB data ([0..1] for float
     cifar_train 길이: 50000
     cifar_test 길이: 10000
     image data 형태: torch.Size([3, 32, 32])
     label: 9
                      label:9
      10
      15
      20
      25
def ComputeAccr(dloader, imodel):
 correct = 0
  total = 0
  for j, [imgs, labels] in enumerate(dloader):
    img = Variable(imgs,volatile=True).cuda()
    label = Variable(labels).cuda()
   output = imodel.forward(img)
```

```
nn.Conv2d(16,32,3,padding=1).
      nn.ReLU(),
      ##nn.Dropout2d(0.2),
      nn.BatchNorm2d(32),
      nn.MaxPool2d(2,2),
      nn.Conv2d(32,64,3,padding=1),
      nn.ReLU(),
      ##nn.Dropout2d(0.2),
      nn.BatchNorm2d(64),
      nn.MaxPool2d(2,2)
  )
  self.fc_layer=nn.Sequential(
      nn.Linear (64*8*8, 64*4*4),
      nn.PReLU(),
      ##nn.Dropout2d(0.2),
      nn.BatchNorm1d(64*4*4),
      nn.Linear(64*4*4, 64*4),
      nn.PReLU(),
      ##nn.Dropout2d(0.2),
      nn.BatchNorm1d(64*4),
      nn.Linear(64*4, 64*2),
      nn.PReLU().
      ##nn.Dropout2d(0.2),
      nn.BatchNorm1d(64*2),
      nn.Linear(64*2, 64),
      nn.PReLU(),
      ##nn.Dropout2d(0.2),
      nn.BatchNorm1d(64),
      nn.Linear(64, 32),
      nn.PReLU().
      ##nn.Dropout2d(0.2),
      nn.BatchNorm1d(32),
      nn.Linear (32, 10)
  for m in self.modules():
    if isinstance(m,nn.Conv2d):
      init.kaiming_normal(m.weight.data)
      m.bias.data.fill_(0)
      if isinstance(m,nn.Linear):
        init.kalming_normal(m.weight.data)
        m.blas.data.fill_(0)
def forward(self,x):
  out=self.layer(x)
  out=out.view(batch_size,-1)
  out=self.fc_layer(out)
  return out
```

```
mode I = CNN().cuda()
     /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:57: UserWarning: nn.init.kaimin
num_{epoch} = 100
loss_func=nn.CrossEntropyLoss()
##optimizer=torch.optim.SGD(model.parameters(), Ir=learning_rate)
optimizer=torch.optim.Adam(model.parameters(), Ir=learning_rate)
scheduler = Ir_scheduler.StepLR(optimizer, step_size=30, gamma=0.2)
model.train()
for i in range(num_epoch):
  for j,[image, label] in enumerate(train_loader):
   x=Variable(image).cuda()
   y_=Variable(label).cuda()
   optimizer.zero_grad()
    output=model.forward(x)
    loss=loss_func(output,y_)
    loss.backward()
   optimizer.step()
    if j%1000==0:
      print(j,loss)
     0 tensor(0.0005, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(0.0017, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(8.3147e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(3.6063e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     O tensor(1.4453e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(5.1557e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(0.0001, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(9.7824e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(1.9220e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(1.3865e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(0.0355, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(0.0587, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(5.5709e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(4.9485e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(7.6741e-07, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(3.0355e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(3.0480e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(0.0018, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(0.0002, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(9.5440e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(6.9290e-07, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(3.3963e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(0.0027, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(5.7815e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(9.1788e-06, device='cuda:0', grad_fn=<NIILossBackward>)
```

```
1000 tensor(4.3958e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(7.6491e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(3.2037e-07, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(0.0006, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(1.0594e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(0.0002, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(0.0004, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(0.0005, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(1.0766e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(7.2939e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(0.2410, device='cuda:0', grad_fn=<NIILossBackward>)
     O tensor(2.3026e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(4.7637e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(1.0207e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(0.0122, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(0.0023, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(1.8924e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(3.4273e-07, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(2.0121e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(2.4140e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(0.1180, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(2.3079e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(0.0002, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(0.0008, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(0.0001, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(0.0147, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(0.2050, device='cuda:0', grad_fn=<NIILossBackward>)
     0 tensor(3.0919e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(5.8116e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(6.1102e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     3000 tensor(3.7998e-07, device='cuda:0', grad_fn=<NIILossBackward>)
     O tensor(6.6977e-06, device='cuda:0', grad_fn=<NIILossBackward>)
     1000 tensor(1.0110e-05, device='cuda:0', grad_fn=<NIILossBackward>)
     2000 tensor(0.0002, device='cuda:0', grad_fn=<NIILossBackward>)
model.eval()
```

ComputeAccr(test_loader, model)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: UserWarning: volatile was rea

Accuracy of Test Data: 78.30999755859375

더블클릭 또는 Enter 키를 눌러 수정

```
netname = "./cifar10"
torch.save(model, netname,)
```

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