```
import torch
import torchvision.datasets
import torchvision.transforms as transforms
from torch.utils.data import DataLoader
import sys
sys.path.append("/content/drive/My Drive/Colab Notebooks/AdversarialExample/advGAN")
import ipynb importer
import models
from models import MNIST_target_net
use cuda=True
image nc=1
batch size = 128
gen input nc = image nc
# Define what device we are using
print("CUDA Available: ",torch.cuda.is available())
device = torch.device("cuda" if (use cuda and torch.cuda.is available()) else "cpu")
    CUDA Available: True
# load the pretrained model
pretrained model = "/content/drive/My Drive/Colab Notebooks/AdversarialExample/advGAN/MNIST target model.pth"
target model = MNIST target net().to(device)
target model.load state dict(torch.load(pretrained model))
target model.eval()
 С→
```

```
MNIST_target_net(
    (conv1): Conv2d(1, 32, kernel_size=(3, 3), stride=(1, 1))
    (conv2): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1))
    (conv3): Conv2d(32, 64, kernel_size=(3, 3), stride=(1, 1))
    (conv4): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1))
    (fcl): Linear(in_features=1024, out_features=200, bias=True)
    (fc2): Linear(in_features=200, out_features=200, bias=True)
    (logits): Linear(in_features=200, out_features=10, bias=True)
    )

# load the generator of adversarial examples
pretrained_generator_path = '/content/drive/My Drive/Colab Notebooks/AdversarialExample/advGAN/models/netG_epoch_60.pth'
pretrained_G = models.Generator(gen_input_nc, image_nc).to(device)
pretrained_G.load_state_dict(torch.load(pretrained_generator_path))
pretrained_G.eval()

[]
```

```
Generator(
  (encoder): Sequential(
    (0): Conv2d(1, 8, kernel size=(3, 3), stride=(1, 1))
   (1): InstanceNorm2d(8, eps=1e-05, momentum=0.1, affine=False, track running stats=False)
    (2): ReLU()
    (3): Conv2d(8, 16, kernel size=(3, 3), stride=(2, 2))
    (4): InstanceNorm2d(16, eps=1e-05, momentum=0.1, affine=False, track running stats=False)
   (5): ReLU()
   (6): Conv2d(16, 32, kernel size=(3, 3), stride=(2, 2))
   (7): InstanceNorm2d(32, eps=1e-05, momentum=0.1, affine=False, track running stats=False)
    (8): ReLU()
  (bottle neck): Sequential(
   (0): ResnetBlock(
      (conv block): Sequential(
        (0): ReflectionPad2d((1, 1, 1, 1))
        (1): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
        (2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
        (3): ReLU(inplace=True)
        (4): ReflectionPad2d((1, 1, 1, 1))
        (5): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
        (6): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
   (1): ResnetBlock(
      (conv block): Sequential(
        (0): ReflectionPad2d((1, 1, 1, 1))
        (1): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
        (2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
        (3): ReLU(inplace=True)
        (4): ReflectionPad2d((1, 1, 1, 1))
        (5): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
        (6): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
    (2): ResnetBlock(
      (conv block): Sequential(
        (0): ReflectionPad2d((1, 1, 1, 1))
        (1): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
        (2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
        (3): ReLU(inplace=True)
        (4): ReflectionPad2d((1. 1. 1. 1))
```

```
(5): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
             (6): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
        (3): ResnetBlock(
           (conv block): Sequential(
             (0): ReflectionPad2d((1, 1, 1, 1))
             (1): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
             (2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
             (3): ReLU(inplace=True)
             (4): ReflectionPad2d((1, 1, 1, 1))
             (5): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), bias=False)
             (6): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track running stats=True)
       (decoder): Sequential(
        (0): ConvTranspose2d(32, 16, kernel size=(3, 3), stride=(2, 2), bias=False)
        (1): InstanceNorm2d(16, eps=1e-05, momentum=0.1, affine=False, track running stats=False)
        (2): ReLU()
        (3): ConvTranspose2d(16, 8, kernel size=(3, 3), stride=(2, 2), bias=False)
        (4): InstanceNorm2d(8, eps=1e-05, momentum=0.1, affine=False, track running stats=False)
        (5): ReLU()
        (6): ConvTranspose2d(8, 1, kernel size=(6, 6), stride=(1, 1), bias=False)
        (7): Tanh()
# test adversarial examples in MNIST training dataset
mnist dataset = torchvision.datasets.MNIST('./dataset', train=True, transform=transforms.ToTensor(), download=False)
train dataloader = DataLoader(mnist dataset, batch size=batch size, shuffle=False, num workers=1)
num correct = 0
for i, data in enumerate(train dataloader, 0):
    test img, test label = data
    test_img, test_label = test_img.to(device), test_label.to(device)
    perturbation = pretrained G(test img)
    perturbation = torch.clamp(perturbation, -0.3, 0.3)
    adv img = perturbation + test img
    adv img = torch.clamp(adv img, 0, 1)
```

```
pred lab = torch.argmax(target model(adv img),1)
    num correct += torch.sum(pred lab==test label,0)
print('MNIST training dataset:')
print('num correct: ', num correct.item())
print('accuracy of adv imgs in training set: %f\n'%(num correct.item()/len(mnist dataset)))
   MNIST training dataset:
    num correct: 155
    accuracy of adv imgs in training set: 0.002583
# test adversarial examples in MNIST testing dataset
mnist dataset test = torchvision.datasets.MNIST('./dataset', train=False, transform=transforms.ToTensor(), download=False
test dataloader = DataLoader(mnist dataset test, batch size=batch_size, shuffle=False, num_workers=1)
num correct = 0
for i, data in enumerate(test dataloader, 0):
    test img, test label = data
    test imq, test label = test img.to(device), test_label.to(device)
    perturbation = pretrained G(test img)
    perturbation = torch.clamp(perturbation, -0.3, 0.3)
    adv img = perturbation + test img
    adv img = torch.clamp(adv img, 0, 1)
    pred lab = torch.argmax(target model(adv img),1)
    num correct += torch.sum(pred lab==test label,0)
print('num correct: ', num correct.item())
print('accuracy of adv imgs in testing set: %f\n'%(num correct.item()/len(mnist dataset test)))
    num correct: 54
    accuracy of adv imgs in testing set: 0.005400
```