

dlnd_face_generation

March 3, 2019

1 Face Generation

In this project, you'll define and train a DCGAN on a dataset of faces. Your goal is to get a generator network to generate *new* images of faces that look as realistic as possible!

The project will be broken down into a series of tasks from **loading in data to defining and training adversarial networks**. At the end of the notebook, you'll be able to visualize the results of your trained Generator to see how it performs; your generated samples should look like fairly realistic faces with small amounts of noise.

1.0.1 Get the Data

You'll be using the [CelebFaces Attributes Dataset \(CelebA\)](#) to train your adversarial networks.

This dataset is more complex than the number datasets (like MNIST or SVHN) you've been working with, and so, you should prepare to define deeper networks and train them for a longer time to get good results. It is suggested that you utilize a GPU for training.

1.0.2 Pre-processed Data

Since the project's main focus is on building the GANs, we've done *some* of the pre-processing for you. Each of the CelebA images has been cropped to remove parts of the image that don't include a face, then resized down to 64x64x3 NumPy images. Some sample data is show below.

If you are working locally, you can download this data [by clicking here](#)

This is a zip file that you'll need to extract in the home directory of this notebook for further loading and processing. After extracting the data, you should be left with a directory of data `processed_celeba_small/`

```
In [32]: # can comment out after executing
        # !unzip processed_celeba_small.zip
```

```
In [33]: data_dir = 'processed_celeba_small/'
```

```
"""
DON'T MODIFY ANYTHING IN THIS CELL
"""

import pickle as pkl
import matplotlib.pyplot as plt
```

```
import numpy as np
import problem_unittests as tests
#import helper
```

```
%matplotlib inline
```

1.1 Visualize the CelebA Data

The [CelebA](#) dataset contains over 200,000 celebrity images with annotations. Since you're going to be generating faces, you won't need the annotations, you'll only need the images. Note that these are color images with [3 color channels \(RGB\)](#) each.

1.1.1 Pre-process and Load the Data

Since the project's main focus is on building the GANs, we've done *some* of the pre-processing for you. Each of the CelebA images has been cropped to remove parts of the image that don't include a face, then resized down to 64x64x3 NumPy images. This *pre-processed* dataset is a smaller subset of the very large CelebA data.

There are a few other steps that you'll need to **transform** this data and create a **DataLoader**.

Exercise: Complete the following `get_dataloader` function, such that it satisfies these requirements:

- Your images should be square, Tensor images of size `image_size x image_size` in the x and y dimension.
- Your function should return a `Dataloader` that shuffles and batches these Tensor images.

ImageFolder To create a dataset given a directory of images, it's recommended that you use PyTorch's [ImageFolder](#) wrapper, with a root directory `processed_celeba_small/` and data transformation passed in.

```
In [34]: # necessary imports
import torch
from torchvision import datasets
from torchvision import transforms

In [35]: def get_dataloader(batch_size, image_size, data_dir='processed_celeba_small/'):
    """
    Batch the neural network data using DataLoader
    :param batch_size: The size of each batch; the number of images in a batch
    :param img_size: The square size of the image data (x, y)
    :param data_dir: Directory where image data is located
    :return: DataLoader with batched data
    """

    # TODO: Implement function and return a dataloader
    img_transforms = transforms.Compose([transforms.RandomResizedCrop(image_size),
```

```

        transforms.ToTensor()])
    data = datasets.ImageFolder(data_dir, transform=img_transforms)
    dataloader = torch.utils.data.DataLoader(data, batch_size=batch_size, shuffle=True)

    return dataloader

```

1.2 Create a DataLoader

Exercise: Create a DataLoader `celeba_train_loader` **with appropriate hyperparameters**. Call the above function and create a dataloader to view images. * You can decide on any reasonable `batch_size` parameter * Your `image_size` **must be 32**. Resizing the data to a smaller size will make for faster training, while still creating convincing images of faces!

```

In [36]: # Define function hyperparameters
        batch_size = 32
        img_size = 32

        """
        DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
        """

        # Call your function and get a dataloader
        celeba_train_loader = get_dataloader(batch_size, img_size)

```

Next, you can view some images! You should see square images of somewhat-centered faces.

Note: You'll need to convert the Tensor images into a NumPy type and transpose the dimensions to correctly display an image, suggested `imshow` code is below, but it may not be perfect.

```

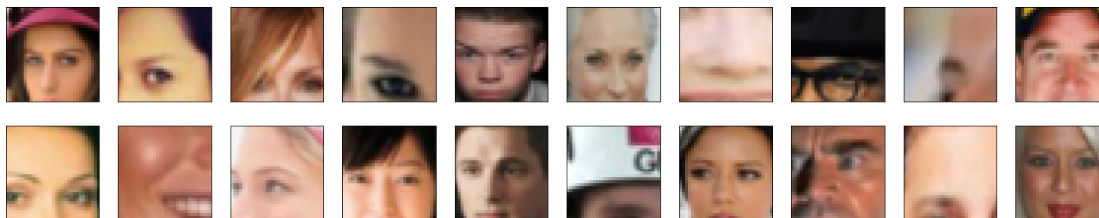
In [37]: # helper display function
        def imshow(img):
            npimg = img.numpy()
            plt.imshow(np.transpose(npimg, (1, 2, 0)))

            """
            DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
            """

            # obtain one batch of training images
            dataiter = iter(celeba_train_loader)
            images, _ = dataiter.next() # _ for no labels

            # plot the images in the batch, along with the corresponding labels
            fig = plt.figure(figsize=(20, 4))
            plot_size=20
            for idx in np.arange(plot_size):
                ax = fig.add_subplot(2, plot_size/2, idx+1, xticks=[], yticks=[])
                imshow(images[idx])

```



Exercise: Pre-process your image data and scale it to a pixel range of -1 to 1 You need to do a bit of pre-processing; you know that the output of a tanh activated generator will contain pixel values in a range from -1 to 1, and so, we need to rescale our training images to a range of -1 to 1. (Right now, they are in a range from 0-1.)

```
In [38]: # TODO: Complete the scale function
def scale(x, feature_range=(-1, 1)):
    ''' Scale takes in an image x and returns that image, scaled
        with a feature_range of pixel values from -1 to 1.
        This function assumes that the input x is already scaled from 0-1. '''
    # assume x is scaled to (0, 1)
    # scale to feature_range and return scaled x
    min, max = feature_range
    x = x * (max - min) + min

    return x
```

```
In [39]: """
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
"""

# check scaled range
# should be close to -1 to 1
img = images[0]
scaled_img = scale(img)

print('Min: ', scaled_img.min())
print('Max: ', scaled_img.max())
```

```
Min:  tensor(-1.)
Max:  tensor(0.9373)
```

2 Define the Model

A GAN is comprised of two adversarial networks, a discriminator and a generator.

2.1 Discriminator

Your first task will be to define the discriminator. This is a convolutional classifier like you've built before, only without any maxpooling layers. To deal with this complex data, it's suggested you use a deep network with **normalization**. You are also allowed to create any helper functions that may be useful.

Exercise: Complete the Discriminator class

- The inputs to the discriminator are 32x32x3 tensor images
- The output should be a single value that will indicate whether a given image is real or fake

```
In [40]: import torch.nn as nn
import torch.nn.functional as F

# helper conv function
def conv(in_channels, out_channels, kernel_size, stride=2, padding=1, batch_norm=True):
    """Creates a convolutional layer, with optional batch normalization.
    """
    layers = []
    conv_layer = nn.Conv2d(in_channels, out_channels,
                           kernel_size, stride, padding, bias=False)

    # append conv layer
    layers.append(conv_layer)

    if batch_norm:
        # append batchnorm layer
        layers.append(nn.BatchNorm2d(out_channels))

    # using Sequential container
    return nn.Sequential(*layers)

In [41]: class Discriminator(nn.Module):

    def __init__(self, conv_dim):
        """
        Initialize the Discriminator Module
        :param conv_dim: The depth of the first convolutional layer
        """
        super(Discriminator, self).__init__()

        # complete init function
        self.conv_dim = conv_dim

        # 32x32 input
        self.conv1 = conv(3, conv_dim, 4, batch_norm=False) # first layer, no batch_norm
        # 16x16 out
        self.conv2 = conv(conv_dim, conv_dim*2, 4)
```

```

# 8x8 out
self.conv3 = conv(conv_dim*2, conv_dim*4, 4)
# 4x4 out
self.conv4 = conv(conv_dim*4, conv_dim*8, 4)
# 2x2 out

# final, fully-connected layer
self.fc = nn.Linear(conv_dim*8*2*2, 1)

def forward(self, x):
    """
    Forward propagation of the neural network
    :param x: The input to the neural network
    :return: Discriminator logits; the output of the neural network
    """
    # define feedforward behavior
    # all hidden layers + leaky relu activation
    out = F.leaky_relu(self.conv1(x), 0.2)
    out = F.leaky_relu(self.conv2(out), 0.2)
    out = F.leaky_relu(self.conv3(out), 0.2)
    out = F.leaky_relu(self.conv4(out), 0.2)

    # flatten
    out = out.view(-1, self.conv_dim*8*2*2)

    # final output layer
    x = self.fc(out)

    return x

    """
    DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
    """

tests.test_discriminator(Discriminator)

```

Tests Passed

2.2 Generator

The generator should upsample an input and generate a *new* image of the same size as our training data 32x32x3. This should be mostly transpose convolutional layers with normalization applied to the outputs.

Exercise: Complete the Generator class

- The inputs to the generator are vectors of some length `z_size`

- The output should be a image of shape 32x32x3

```
In [42]: # helper deconv function
def deconv(in_channels, out_channels, kernel_size, stride=2, padding=1, batch_norm=True):
    """Creates a transposed-convolutional layer, with optional batch normalization.
    """
    # create a sequence of transpose + optional batch norm layers
    layers = []
    transpose_conv_layer = nn.ConvTranspose2d(in_channels, out_channels,
                                              kernel_size, stride, padding, bias=False)

    # append transpose convolutional layer
    layers.append(transpose_conv_layer)

    if batch_norm:
        # append batchnorm layer
        layers.append(nn.BatchNorm2d(out_channels))

    return nn.Sequential(*layers)
```

```
In [43]: class Generator(nn.Module):

    def __init__(self, z_size, conv_dim):
        """
        Initialize the Generator Module
        :param z_size: The length of the input latent vector, z
        :param conv_dim: The depth of the inputs to the *last* transpose convolutional
        """
        super(Generator, self).__init__()

        # complete init function

        self.conv_dim = conv_dim

        # first, fully-connected layer
        self.fc = nn.Linear(z_size, conv_dim*8*2*2)

        # transpose conv layers
        self.t_conv1 = deconv(conv_dim*8, conv_dim*4, 4)
        self.t_conv2 = deconv(conv_dim*4, conv_dim*2, 4)
        self.t_conv3 = deconv(conv_dim*2, conv_dim, 4)
        self.t_conv4 = deconv(conv_dim, 3, 4, batch_norm=False)

    def forward(self, x):
        """
        Forward propagation of the neural network
        :param x: The input to the neural network
        :return: A 32x32x3 Tensor image as output
```

```

"""
# define feedforward behavior
# fully-connected + reshape
out = self.fc(x)
out = out.view(-1, self.conv_dim*8, 2, 2) # (batch_size, depth, 2, 2)

# hidden transpose conv layers + relu
out = F.leaky_relu(self.t_conv1(out), 0.2)
out = F.leaky_relu(self.t_conv2(out), 0.2)
out = F.leaky_relu(self.t_conv3(out), 0.2)

# last layer + tanh activation
out = self.t_conv4(out)
out = F.tanh(out)

return out

"""
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
"""
tests.test_generator(Generator)

```

Tests Passed

2.3 Initialize the weights of your networks

To help your models converge, you should initialize the weights of the convolutional and linear layers in your model. From reading the [original DCGAN paper](#), they say: > All weights were initialized from a zero-centered Normal distribution with standard deviation 0.02.

So, your next task will be to define a weight initialization function that does just this!

You can refer back to the lesson on weight initialization or even consult existing model code, such as that from [the networks.py file in CycleGAN Github repository](#) to help you complete this function.

Exercise: Complete the weight initialization function

- This should initialize only **convolutional** and **linear** layers
- Initialize the weights to a normal distribution, centered around 0, with a standard deviation of 0.02.
- The bias terms, if they exist, may be left alone or set to 0.

```

In [44]: def weights_init_normal(m):
        """
        Applies initial weights to certain layers in a model .
        The weights are taken from a normal distribution
        with mean = 0, std dev = 0.02.
        :param m: A module or layer in a network

```



```

"""
# classname will be something like:
# `Conv`, `BatchNorm2d`, `Linear`, etc.
classname = m.__class__.__name__

# TODO: Apply initial weights to convolutional and linear layers
if classname == 'Conv' or classname == 'Linear':
    nn.init.normal_(m.weight.data, mean = 0, std = 0.02)

```

2.4 Build complete network

Define your models' hyperparameters and instantiate the discriminator and generator from the classes defined above. Make sure you've passed in the correct input arguments.

```

In [45]: """
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
"""

def build_network(d_conv_dim, g_conv_dim, z_size):
    # define discriminator and generator
    D = Discriminator(d_conv_dim)
    G = Generator(z_size=z_size, conv_dim=g_conv_dim)

    # initialize model weights
    D.apply(weights_init_normal)
    G.apply(weights_init_normal)

    print(D)
    print()
    print(G)

    return D, G

```

Exercise: Define model hyperparameters

```

In [46]: # Define model hyperparams
d_conv_dim = 32
g_conv_dim = 32
z_size = 100

"""
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
"""

D, G = build_network(d_conv_dim, g_conv_dim, z_size)

```

```

Discriminator(
  (conv1): Sequential(
    (0): Conv2d(3, 32, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)

```

```

)
(conv2): Sequential(
  (0): Conv2d(32, 64, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)
  (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
)
(conv3): Sequential(
  (0): Conv2d(64, 128, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)
  (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
)
(conv4): Sequential(
  (0): Conv2d(128, 256, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)
  (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
)
(fc): Linear(in_features=1024, out_features=1, bias=True)
)

Generator(
  (fc): Linear(in_features=100, out_features=1024, bias=True)
  (t_conv1): Sequential(
    (0): ConvTranspose2d(256, 128, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)
    (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
  )
  (t_conv2): Sequential(
    (0): ConvTranspose2d(128, 64, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)
    (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
  )
  (t_conv3): Sequential(
    (0): ConvTranspose2d(64, 32, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)
    (1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
  )
  (t_conv4): Sequential(
    (0): ConvTranspose2d(32, 3, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1), bias=False)
  )
)

```

2.4.1 Training on GPU

Check if you can train on GPU. Here, we'll set this as a boolean variable `train_on_gpu`. Later, you'll be responsible for making sure that `> * Models, * Model inputs, and * Loss function arguments`

Are moved to GPU, where appropriate.

```

In [47]: """
          DON'T MODIFY ANYTHING IN THIS CELL
          """
          import torch

```

```

# Check for a GPU
train_on_gpu = torch.cuda.is_available()
if not train_on_gpu:
    print('No GPU found. Please use a GPU to train your neural network.')
else:
    print('Training on GPU!')

```

Training on GPU!

2.5 Discriminator and Generator Losses

Now we need to calculate the losses for both types of adversarial networks.

2.5.1 Discriminator Losses

- For the discriminator, the total loss is the sum of the losses for real and fake images, $d_loss = d_real_loss + d_fake_loss$.
- Remember that we want the discriminator to output 1 for real images and 0 for fake images, so we need to set up the losses to reflect that.

2.5.2 Generator Loss

The generator loss will look similar only with flipped labels. The generator's goal is to get the discriminator to *think* its generated images are *real*.

Exercise: Complete real and fake loss functions You may choose to use either cross entropy or a least squares error loss to complete the following `real_loss` and `fake_loss` functions.

```

In [48]: def real_loss(D_out):
    '''Calculates how close discriminator outputs are to being real.
        param, D_out: discriminator logits
        return: real loss'''

    batch_size = D_out.size(0)
    labels = torch.ones(batch_size)*0.9 # real labels = 1, with smoothing
    # move labels to GPU if available
    if train_on_gpu:
        labels = labels.cuda()
    # binary cross entropy with logits loss
    criterion = nn.BCEWithLogitsLoss()
    # calculate loss
    loss = criterion(D_out.squeeze(), labels)
    return loss

def fake_loss(D_out):
    '''Calculates how close discriminator outputs are to being fake.

```

```

        param, D_out: discriminator logits
        return: fake loss'''
    batch_size = D_out.size(0)
    labels = torch.zeros(batch_size) # fake labels = 0
    if train_on_gpu:
        labels = labels.cuda()
    criterion = nn.BCEWithLogitsLoss()
    # calculate loss
    loss = criterion(D_out.squeeze(), labels)
    return loss

```

2.6 Optimizers

Exercise: Define optimizers for your Discriminator (D) and Generator (G) Define optimizers for your models with appropriate hyperparameters.

In [49]: `import torch.optim as optim`

```

# params
lr = 0.0001
beta1=0.1
beta2=0.999 # default value

# Create optimizers for the discriminator and generator
d_optimizer = optim.Adam(D.parameters(), lr, [beta1, beta2])
g_optimizer = optim.Adam(G.parameters(), lr, [beta1, beta2])

```

2.7 Training

Training will involve alternating between training the discriminator and the generator. You'll use your functions `real_loss` and `fake_loss` to help you calculate the discriminator losses.

- You should train the discriminator by alternating on real and fake images
- Then the generator, which tries to trick the discriminator and should have an opposing loss function

Saving Samples You've been given some code to print out some loss statistics and save some generated "fake" samples.

Exercise: Complete the training function Keep in mind that, if you've moved your models to GPU, you'll also have to move any model inputs to GPU.

```

In [50]: def train(D, G, n_epochs, print_every=50):
        '''Trains adversarial networks for some number of epochs
        param, D: the discriminator network
        param, G: the generator network

```

```

    param, n_epochs: number of epochs to train for
    param, print_every: when to print and record the models' losses
    return: D and G losses'''

# move models to GPU
if train_on_gpu:
    D.cuda()
    G.cuda()

# keep track of loss and generated, "fake" samples
samples = []
losses = []

# Get some fixed data for sampling. These are images that are held
# constant throughout training, and allow us to inspect the model's performance
sample_size=16
fixed_z = np.random.uniform(-1, 1, size=(sample_size, z_size))
fixed_z = torch.from_numpy(fixed_z).float()
# move z to GPU if available
if train_on_gpu:
    fixed_z = fixed_z.cuda()

# epoch training loop
for epoch in range(n_epochs):

    # batch training loop
    for batch_i, (real_images, _) in enumerate(celeba_train_loader):

        batch_size = real_images.size(0)

        # important rescaling step
        real_images = scale(real_images)

        # =====
        #          YOUR CODE HERE: TRAIN THE NETWORKS
        # =====

        # =====
        #          TRAIN THE DISCRIMINATOR
        # =====

        d_optimizer.zero_grad()

        # 1. Train with real images

        # Compute the discriminator losses on real images
        if train_on_gpu:
            real_images = real_images.cuda()

```

```

D_real = D(real_images)
d_real_loss = real_loss(D_real)

# 2. Train with fake images

# Generate fake images
z = np.random.uniform(-1, 1, size=(batch_size, z_size))
z = torch.from_numpy(z).float()
# move x to GPU, if available
if train_on_gpu:
    z = z.cuda()
fake_images = G(z)

# Compute the discriminator losses on fake images
D_fake = D(fake_images)
d_fake_loss = fake_loss(D_fake)

# add up loss and perform backprop
d_loss = d_real_loss + d_fake_loss
d_loss.backward()
d_optimizer.step()

# =====
#                               TRAIN THE GENERATOR
# =====
g_optimizer.zero_grad()

# 1. Train with fake images and flipped labels

# Generate fake images
z = np.random.uniform(-1, 1, size=(batch_size, z_size))
z = torch.from_numpy(z).float()
if train_on_gpu:
    z = z.cuda()
fake_images = G(z)

# Compute the discriminator losses on fake images
# using flipped labels!
D_fake = D(fake_images)
g_loss = real_loss(D_fake) # use real loss to flip labels

# perform backprop
g_loss.backward()
g_optimizer.step()

# =====
#                               END OF YOUR CODE

```

```

# =====

# Print some loss stats
if batch_i % print_every == 0:
    # append discriminator loss and generator loss
    losses.append((d_loss.item(), g_loss.item()))
    # print discriminator and generator loss
    print('Epoch [{:5d}/{:5d}] | d_loss: {:.4f} | g_loss: {:.4f}'.format(
        epoch+1, n_epochs, d_loss.item(), g_loss.item()))

## AFTER EACH EPOCH##
# this code assumes your generator is named G, feel free to change the name
# generate and save sample, fake images
G.eval() # for generating samples
samples_z = G(fixed_z)
samples.append(samples_z)
G.train() # back to training mode

# Save training generator samples
with open('train_samples.pkl', 'wb') as f:
    pkl.dump(samples, f)

# finally return losses
return losses

```

Set your number of training epochs and train your GAN!

```

In [51]: # set number of epochs
n_epochs = 50

### keep the cell running longer than 30 mins with GPU
import workspace_utils
from workspace_utils import active_session

with active_session():
    """
    DON'T MODIFY ANYTHING IN THIS CELL
    """
    # call training function
    losses = train(D, G, n_epochs=n_epochs)

```

Epoch [1/	50]	d_loss: 1.3803	g_loss: 0.8494
Epoch [1/	50]	d_loss: 0.6461	g_loss: 1.9046
Epoch [1/	50]	d_loss: 0.5009	g_loss: 2.8708
Epoch [1/	50]	d_loss: 0.5722	g_loss: 2.4141
Epoch [1/	50]	d_loss: 0.5629	g_loss: 2.5032
Epoch [1/	50]	d_loss: 0.6254	g_loss: 2.2364

Epoch [1/	50]	d_loss: 1.0231	g_loss: 1.0026
Epoch [1/	50]	d_loss: 1.2742	g_loss: 1.3416
Epoch [1/	50]	d_loss: 1.1663	g_loss: 1.7415
Epoch [1/	50]	d_loss: 1.2019	g_loss: 1.7604
Epoch [1/	50]	d_loss: 1.2846	g_loss: 1.4893
Epoch [1/	50]	d_loss: 1.0288	g_loss: 1.5160
Epoch [1/	50]	d_loss: 1.0549	g_loss: 1.5612
Epoch [1/	50]	d_loss: 1.1215	g_loss: 1.4031
Epoch [1/	50]	d_loss: 0.9655	g_loss: 1.0736
Epoch [1/	50]	d_loss: 1.1430	g_loss: 1.6721
Epoch [1/	50]	d_loss: 1.3802	g_loss: 1.8555
Epoch [1/	50]	d_loss: 1.0160	g_loss: 1.0736
Epoch [1/	50]	d_loss: 1.1436	g_loss: 0.8881
Epoch [1/	50]	d_loss: 1.1237	g_loss: 0.6998
Epoch [1/	50]	d_loss: 1.1498	g_loss: 1.3454
Epoch [1/	50]	d_loss: 1.1710	g_loss: 0.9878
Epoch [1/	50]	d_loss: 1.4403	g_loss: 0.9714
Epoch [1/	50]	d_loss: 1.2250	g_loss: 1.2557
Epoch [1/	50]	d_loss: 1.1430	g_loss: 1.2527
Epoch [1/	50]	d_loss: 1.1415	g_loss: 1.1005
Epoch [1/	50]	d_loss: 1.2322	g_loss: 0.9829
Epoch [1/	50]	d_loss: 1.3476	g_loss: 1.0419
Epoch [1/	50]	d_loss: 1.1861	g_loss: 1.0109
Epoch [1/	50]	d_loss: 1.1555	g_loss: 1.1664
Epoch [1/	50]	d_loss: 1.3512	g_loss: 0.9923
Epoch [1/	50]	d_loss: 1.4163	g_loss: 0.8937
Epoch [1/	50]	d_loss: 1.2439	g_loss: 1.0709
Epoch [1/	50]	d_loss: 1.1552	g_loss: 0.8325
Epoch [1/	50]	d_loss: 1.2124	g_loss: 0.8282
Epoch [1/	50]	d_loss: 1.1100	g_loss: 1.2178
Epoch [1/	50]	d_loss: 1.1792	g_loss: 0.8544
Epoch [1/	50]	d_loss: 1.2758	g_loss: 1.2120
Epoch [1/	50]	d_loss: 1.1353	g_loss: 1.1215
Epoch [1/	50]	d_loss: 1.1028	g_loss: 1.1491
Epoch [1/	50]	d_loss: 1.3218	g_loss: 0.7600
Epoch [1/	50]	d_loss: 1.1073	g_loss: 0.7604
Epoch [1/	50]	d_loss: 1.1200	g_loss: 1.0510
Epoch [1/	50]	d_loss: 1.1214	g_loss: 1.1632
Epoch [1/	50]	d_loss: 1.0761	g_loss: 0.8978
Epoch [1/	50]	d_loss: 1.0658	g_loss: 1.1011
Epoch [1/	50]	d_loss: 1.1099	g_loss: 1.1359
Epoch [1/	50]	d_loss: 1.2885	g_loss: 1.4494
Epoch [1/	50]	d_loss: 1.1682	g_loss: 1.4546
Epoch [1/	50]	d_loss: 1.2965	g_loss: 1.0339
Epoch [1/	50]	d_loss: 1.4040	g_loss: 1.7550
Epoch [1/	50]	d_loss: 1.2129	g_loss: 1.4314
Epoch [1/	50]	d_loss: 1.4347	g_loss: 1.4415
Epoch [1/	50]	d_loss: 1.4429	g_loss: 0.5565

Epoch [1/	50]	d_loss: 1.1246	g_loss: 1.1025
Epoch [1/	50]	d_loss: 1.2797	g_loss: 1.6608
Epoch [1/	50]	d_loss: 1.1278	g_loss: 1.1516
Epoch [2/	50]	d_loss: 1.0472	g_loss: 1.0834
Epoch [2/	50]	d_loss: 1.1879	g_loss: 1.0728
Epoch [2/	50]	d_loss: 1.1979	g_loss: 1.5817
Epoch [2/	50]	d_loss: 0.8940	g_loss: 1.0724
Epoch [2/	50]	d_loss: 1.1002	g_loss: 1.1548
Epoch [2/	50]	d_loss: 1.1943	g_loss: 1.1332
Epoch [2/	50]	d_loss: 1.0780	g_loss: 1.3673
Epoch [2/	50]	d_loss: 1.0312	g_loss: 1.0077
Epoch [2/	50]	d_loss: 1.1201	g_loss: 1.4476
Epoch [2/	50]	d_loss: 1.0284	g_loss: 0.8873
Epoch [2/	50]	d_loss: 1.1345	g_loss: 1.4173
Epoch [2/	50]	d_loss: 1.0544	g_loss: 1.4390
Epoch [2/	50]	d_loss: 1.2247	g_loss: 0.7394
Epoch [2/	50]	d_loss: 1.1676	g_loss: 0.9313
Epoch [2/	50]	d_loss: 1.3783	g_loss: 0.8390
Epoch [2/	50]	d_loss: 1.0494	g_loss: 0.9016
Epoch [2/	50]	d_loss: 0.9982	g_loss: 1.1686
Epoch [2/	50]	d_loss: 0.9565	g_loss: 1.0067
Epoch [2/	50]	d_loss: 1.0239	g_loss: 1.0553
Epoch [2/	50]	d_loss: 1.0469	g_loss: 1.7683
Epoch [2/	50]	d_loss: 1.1365	g_loss: 1.3954
Epoch [2/	50]	d_loss: 1.1178	g_loss: 1.7683
Epoch [2/	50]	d_loss: 0.8917	g_loss: 1.6792
Epoch [2/	50]	d_loss: 1.2121	g_loss: 0.8182
Epoch [2/	50]	d_loss: 1.0266	g_loss: 1.2618
Epoch [2/	50]	d_loss: 0.9855	g_loss: 1.4114
Epoch [2/	50]	d_loss: 1.0996	g_loss: 0.8922
Epoch [2/	50]	d_loss: 1.1200	g_loss: 1.3110
Epoch [2/	50]	d_loss: 1.1500	g_loss: 0.8072
Epoch [2/	50]	d_loss: 1.0811	g_loss: 1.0060
Epoch [2/	50]	d_loss: 1.3494	g_loss: 1.5080
Epoch [2/	50]	d_loss: 1.2277	g_loss: 0.9086
Epoch [2/	50]	d_loss: 1.0832	g_loss: 0.8291
Epoch [2/	50]	d_loss: 1.2186	g_loss: 1.2003
Epoch [2/	50]	d_loss: 0.9669	g_loss: 1.6667
Epoch [2/	50]	d_loss: 1.0819	g_loss: 1.2288
Epoch [2/	50]	d_loss: 0.9827	g_loss: 1.4863
Epoch [2/	50]	d_loss: 1.4366	g_loss: 0.7651
Epoch [2/	50]	d_loss: 0.9873	g_loss: 1.1224
Epoch [2/	50]	d_loss: 0.9341	g_loss: 1.7076
Epoch [2/	50]	d_loss: 0.9911	g_loss: 1.3171
Epoch [2/	50]	d_loss: 0.8090	g_loss: 1.7077
Epoch [2/	50]	d_loss: 1.2891	g_loss: 0.8069
Epoch [2/	50]	d_loss: 1.0575	g_loss: 2.3682
Epoch [2/	50]	d_loss: 1.0617	g_loss: 0.9958

Epoch [2/	50]	d_loss: 1.0067	g_loss: 1.4271
Epoch [2/	50]	d_loss: 1.0506	g_loss: 0.9332
Epoch [2/	50]	d_loss: 1.0361	g_loss: 1.2125
Epoch [2/	50]	d_loss: 1.1667	g_loss: 0.9192
Epoch [2/	50]	d_loss: 0.9146	g_loss: 1.0499
Epoch [2/	50]	d_loss: 1.0859	g_loss: 1.4667
Epoch [2/	50]	d_loss: 1.0657	g_loss: 1.8768
Epoch [2/	50]	d_loss: 0.9837	g_loss: 1.3804
Epoch [2/	50]	d_loss: 1.2979	g_loss: 1.2370
Epoch [2/	50]	d_loss: 1.1853	g_loss: 1.8718
Epoch [2/	50]	d_loss: 0.8646	g_loss: 1.3493
Epoch [2/	50]	d_loss: 1.2235	g_loss: 1.6517
Epoch [3/	50]	d_loss: 0.9083	g_loss: 1.2096
Epoch [3/	50]	d_loss: 0.9826	g_loss: 1.4169
Epoch [3/	50]	d_loss: 0.9747	g_loss: 0.8607
Epoch [3/	50]	d_loss: 0.8936	g_loss: 1.6162
Epoch [3/	50]	d_loss: 0.8566	g_loss: 1.6899
Epoch [3/	50]	d_loss: 1.0938	g_loss: 1.4469
Epoch [3/	50]	d_loss: 1.5314	g_loss: 0.5515
Epoch [3/	50]	d_loss: 1.1358	g_loss: 1.0849
Epoch [3/	50]	d_loss: 1.0672	g_loss: 1.7737
Epoch [3/	50]	d_loss: 0.9277	g_loss: 1.1751
Epoch [3/	50]	d_loss: 1.0568	g_loss: 1.1880
Epoch [3/	50]	d_loss: 1.3357	g_loss: 0.8824
Epoch [3/	50]	d_loss: 1.0849	g_loss: 0.9716
Epoch [3/	50]	d_loss: 1.1509	g_loss: 2.2696
Epoch [3/	50]	d_loss: 0.9296	g_loss: 0.9840
Epoch [3/	50]	d_loss: 0.8499	g_loss: 1.1294
Epoch [3/	50]	d_loss: 0.9250	g_loss: 1.5187
Epoch [3/	50]	d_loss: 0.9752	g_loss: 1.2153
Epoch [3/	50]	d_loss: 1.0897	g_loss: 1.5609
Epoch [3/	50]	d_loss: 1.2166	g_loss: 1.2485
Epoch [3/	50]	d_loss: 1.0585	g_loss: 1.2792
Epoch [3/	50]	d_loss: 1.1328	g_loss: 1.3780
Epoch [3/	50]	d_loss: 1.4914	g_loss: 0.7604
Epoch [3/	50]	d_loss: 1.2352	g_loss: 0.8078
Epoch [3/	50]	d_loss: 0.9753	g_loss: 1.2954
Epoch [3/	50]	d_loss: 0.8068	g_loss: 1.2589
Epoch [3/	50]	d_loss: 0.7578	g_loss: 1.0832
Epoch [3/	50]	d_loss: 1.1609	g_loss: 1.9294
Epoch [3/	50]	d_loss: 0.7721	g_loss: 1.3641
Epoch [3/	50]	d_loss: 0.8734	g_loss: 1.6767
Epoch [3/	50]	d_loss: 0.9926	g_loss: 0.8438
Epoch [3/	50]	d_loss: 0.8192	g_loss: 1.5730
Epoch [3/	50]	d_loss: 1.0796	g_loss: 2.4928
Epoch [3/	50]	d_loss: 0.8187	g_loss: 1.5718
Epoch [3/	50]	d_loss: 0.8999	g_loss: 0.9075
Epoch [3/	50]	d_loss: 0.8065	g_loss: 1.4747

Epoch [3/	50]	d_loss: 0.8861	g_loss: 1.0034
Epoch [3/	50]	d_loss: 1.0597	g_loss: 0.5692
Epoch [3/	50]	d_loss: 0.7915	g_loss: 1.6282
Epoch [3/	50]	d_loss: 1.0161	g_loss: 1.5536
Epoch [3/	50]	d_loss: 0.8700	g_loss: 1.4996
Epoch [3/	50]	d_loss: 0.8104	g_loss: 1.3608
Epoch [3/	50]	d_loss: 0.8978	g_loss: 1.1262
Epoch [3/	50]	d_loss: 1.0362	g_loss: 1.3126
Epoch [3/	50]	d_loss: 0.8241	g_loss: 1.8353
Epoch [3/	50]	d_loss: 1.0062	g_loss: 1.2063
Epoch [3/	50]	d_loss: 1.1117	g_loss: 2.1274
Epoch [3/	50]	d_loss: 1.1195	g_loss: 0.9693
Epoch [3/	50]	d_loss: 1.1691	g_loss: 1.1567
Epoch [3/	50]	d_loss: 1.1593	g_loss: 0.7216
Epoch [3/	50]	d_loss: 0.8675	g_loss: 1.7450
Epoch [3/	50]	d_loss: 0.6210	g_loss: 1.4213
Epoch [3/	50]	d_loss: 0.8246	g_loss: 1.6267
Epoch [3/	50]	d_loss: 1.0940	g_loss: 1.1765
Epoch [3/	50]	d_loss: 0.7398	g_loss: 0.9580
Epoch [3/	50]	d_loss: 0.7130	g_loss: 1.2839
Epoch [3/	50]	d_loss: 0.9981	g_loss: 1.0651
Epoch [4/	50]	d_loss: 0.9164	g_loss: 2.0580
Epoch [4/	50]	d_loss: 1.1504	g_loss: 2.5948
Epoch [4/	50]	d_loss: 0.7394	g_loss: 1.3091
Epoch [4/	50]	d_loss: 0.6011	g_loss: 1.7897
Epoch [4/	50]	d_loss: 1.0284	g_loss: 1.0963
Epoch [4/	50]	d_loss: 1.1719	g_loss: 1.0402
Epoch [4/	50]	d_loss: 0.7388	g_loss: 1.7034
Epoch [4/	50]	d_loss: 0.7896	g_loss: 1.8106
Epoch [4/	50]	d_loss: 0.7794	g_loss: 2.2269
Epoch [4/	50]	d_loss: 0.8555	g_loss: 1.9971
Epoch [4/	50]	d_loss: 0.6405	g_loss: 2.0575
Epoch [4/	50]	d_loss: 0.7616	g_loss: 1.0987
Epoch [4/	50]	d_loss: 0.6760	g_loss: 2.2106
Epoch [4/	50]	d_loss: 1.0032	g_loss: 3.1067
Epoch [4/	50]	d_loss: 1.4115	g_loss: 3.2817
Epoch [4/	50]	d_loss: 0.9967	g_loss: 1.1833
Epoch [4/	50]	d_loss: 1.0460	g_loss: 0.8590
Epoch [4/	50]	d_loss: 0.8039	g_loss: 1.4226
Epoch [4/	50]	d_loss: 0.7435	g_loss: 1.2388
Epoch [4/	50]	d_loss: 0.9552	g_loss: 0.9457
Epoch [4/	50]	d_loss: 0.7047	g_loss: 1.4866
Epoch [4/	50]	d_loss: 0.9016	g_loss: 1.4196
Epoch [4/	50]	d_loss: 0.9128	g_loss: 1.2094
Epoch [4/	50]	d_loss: 0.7330	g_loss: 1.7651
Epoch [4/	50]	d_loss: 1.0883	g_loss: 0.9388
Epoch [4/	50]	d_loss: 0.7991	g_loss: 1.1361
Epoch [4/	50]	d_loss: 0.9344	g_loss: 2.6567

Epoch [4/	50]	d_loss: 0.8538	g_loss: 1.3568
Epoch [4/	50]	d_loss: 0.7294	g_loss: 1.8912
Epoch [4/	50]	d_loss: 0.9863	g_loss: 0.9605
Epoch [4/	50]	d_loss: 1.0488	g_loss: 2.0667
Epoch [4/	50]	d_loss: 0.8384	g_loss: 1.1854
Epoch [4/	50]	d_loss: 0.9227	g_loss: 1.1929
Epoch [4/	50]	d_loss: 0.8344	g_loss: 2.5324
Epoch [4/	50]	d_loss: 1.0939	g_loss: 2.3384
Epoch [4/	50]	d_loss: 0.6929	g_loss: 1.8740
Epoch [4/	50]	d_loss: 0.9499	g_loss: 1.1974
Epoch [4/	50]	d_loss: 0.6738	g_loss: 1.4463
Epoch [4/	50]	d_loss: 0.7281	g_loss: 1.0042
Epoch [4/	50]	d_loss: 0.6881	g_loss: 1.1442
Epoch [4/	50]	d_loss: 0.6998	g_loss: 1.0781
Epoch [4/	50]	d_loss: 0.8606	g_loss: 1.2826
Epoch [4/	50]	d_loss: 0.7449	g_loss: 2.5864
Epoch [4/	50]	d_loss: 0.8292	g_loss: 1.9588
Epoch [4/	50]	d_loss: 0.7269	g_loss: 1.7999
Epoch [4/	50]	d_loss: 0.6890	g_loss: 1.9376
Epoch [4/	50]	d_loss: 0.6237	g_loss: 2.2460
Epoch [4/	50]	d_loss: 0.8038	g_loss: 1.1922
Epoch [4/	50]	d_loss: 0.9355	g_loss: 1.1957
Epoch [4/	50]	d_loss: 0.7814	g_loss: 2.0946
Epoch [4/	50]	d_loss: 0.7622	g_loss: 1.6294
Epoch [4/	50]	d_loss: 1.0034	g_loss: 1.1517
Epoch [4/	50]	d_loss: 1.0923	g_loss: 1.7276
Epoch [4/	50]	d_loss: 0.8538	g_loss: 1.1114
Epoch [4/	50]	d_loss: 0.8339	g_loss: 1.9034
Epoch [4/	50]	d_loss: 0.7168	g_loss: 2.0894
Epoch [4/	50]	d_loss: 0.8916	g_loss: 1.3755
Epoch [5/	50]	d_loss: 0.6233	g_loss: 1.8400
Epoch [5/	50]	d_loss: 0.6161	g_loss: 1.5990
Epoch [5/	50]	d_loss: 0.6776	g_loss: 1.9095
Epoch [5/	50]	d_loss: 0.9259	g_loss: 2.0895
Epoch [5/	50]	d_loss: 0.8576	g_loss: 2.0619
Epoch [5/	50]	d_loss: 0.8722	g_loss: 1.1776
Epoch [5/	50]	d_loss: 0.8395	g_loss: 1.1992
Epoch [5/	50]	d_loss: 0.5597	g_loss: 2.4227
Epoch [5/	50]	d_loss: 0.6401	g_loss: 1.9982
Epoch [5/	50]	d_loss: 0.6400	g_loss: 1.5235
Epoch [5/	50]	d_loss: 0.6962	g_loss: 1.5277
Epoch [5/	50]	d_loss: 0.6295	g_loss: 2.5910
Epoch [5/	50]	d_loss: 0.7231	g_loss: 1.4614
Epoch [5/	50]	d_loss: 0.7280	g_loss: 0.9132
Epoch [5/	50]	d_loss: 0.5556	g_loss: 1.9037
Epoch [5/	50]	d_loss: 0.8120	g_loss: 1.2821
Epoch [5/	50]	d_loss: 1.0020	g_loss: 2.7529
Epoch [5/	50]	d_loss: 0.6677	g_loss: 1.4735

Epoch [5/	50]	d_loss: 0.5398	g_loss: 1.8519
Epoch [5/	50]	d_loss: 0.8846	g_loss: 3.1437
Epoch [5/	50]	d_loss: 0.4779	g_loss: 1.9806
Epoch [5/	50]	d_loss: 0.5917	g_loss: 1.6043
Epoch [5/	50]	d_loss: 0.9504	g_loss: 2.2487
Epoch [5/	50]	d_loss: 0.8027	g_loss: 1.7037
Epoch [5/	50]	d_loss: 0.6707	g_loss: 1.5177
Epoch [5/	50]	d_loss: 0.6279	g_loss: 1.8522
Epoch [5/	50]	d_loss: 0.8409	g_loss: 1.5337
Epoch [5/	50]	d_loss: 1.0048	g_loss: 3.1626
Epoch [5/	50]	d_loss: 0.5336	g_loss: 1.8155
Epoch [5/	50]	d_loss: 0.6538	g_loss: 1.8662
Epoch [5/	50]	d_loss: 0.6123	g_loss: 1.8400
Epoch [5/	50]	d_loss: 0.7215	g_loss: 3.0555
Epoch [5/	50]	d_loss: 0.8175	g_loss: 2.1356
Epoch [5/	50]	d_loss: 0.4938	g_loss: 2.1341
Epoch [5/	50]	d_loss: 0.5271	g_loss: 2.0242
Epoch [5/	50]	d_loss: 0.6515	g_loss: 1.8113
Epoch [5/	50]	d_loss: 0.6048	g_loss: 2.1185
Epoch [5/	50]	d_loss: 0.7258	g_loss: 1.3787
Epoch [5/	50]	d_loss: 0.9164	g_loss: 2.8649
Epoch [5/	50]	d_loss: 0.5072	g_loss: 2.7120
Epoch [5/	50]	d_loss: 0.8076	g_loss: 2.7030
Epoch [5/	50]	d_loss: 0.8819	g_loss: 1.9872
Epoch [5/	50]	d_loss: 0.6792	g_loss: 1.6056
Epoch [5/	50]	d_loss: 0.8929	g_loss: 0.8780
Epoch [5/	50]	d_loss: 0.8643	g_loss: 1.8561
Epoch [5/	50]	d_loss: 0.7870	g_loss: 3.6213
Epoch [5/	50]	d_loss: 0.5751	g_loss: 2.2278
Epoch [5/	50]	d_loss: 0.5576	g_loss: 1.3898
Epoch [5/	50]	d_loss: 0.7590	g_loss: 3.2165
Epoch [5/	50]	d_loss: 1.7283	g_loss: 4.6210
Epoch [5/	50]	d_loss: 0.8764	g_loss: 0.8708
Epoch [5/	50]	d_loss: 0.5797	g_loss: 1.9620
Epoch [5/	50]	d_loss: 0.5569	g_loss: 2.1069
Epoch [5/	50]	d_loss: 0.8452	g_loss: 1.0773
Epoch [5/	50]	d_loss: 0.8446	g_loss: 1.6089
Epoch [5/	50]	d_loss: 0.7224	g_loss: 2.3796
Epoch [5/	50]	d_loss: 0.6513	g_loss: 1.3421
Epoch [6/	50]	d_loss: 0.9110	g_loss: 2.9294
Epoch [6/	50]	d_loss: 0.7608	g_loss: 1.9728
Epoch [6/	50]	d_loss: 0.5437	g_loss: 1.8852
Epoch [6/	50]	d_loss: 0.7466	g_loss: 1.5585
Epoch [6/	50]	d_loss: 0.6875	g_loss: 1.6308
Epoch [6/	50]	d_loss: 1.1197	g_loss: 2.7706
Epoch [6/	50]	d_loss: 0.5908	g_loss: 2.5048
Epoch [6/	50]	d_loss: 0.6297	g_loss: 2.5022
Epoch [6/	50]	d_loss: 0.5882	g_loss: 2.5725

Epoch [6/	50]	d_loss: 0.6749	g_loss: 1.7199
Epoch [6/	50]	d_loss: 0.9448	g_loss: 2.3090
Epoch [6/	50]	d_loss: 0.7130	g_loss: 2.7379
Epoch [6/	50]	d_loss: 0.9418	g_loss: 1.7283
Epoch [6/	50]	d_loss: 0.5122	g_loss: 1.5195
Epoch [6/	50]	d_loss: 0.7128	g_loss: 2.8077
Epoch [6/	50]	d_loss: 0.5878	g_loss: 2.0009
Epoch [6/	50]	d_loss: 0.7090	g_loss: 1.7630
Epoch [6/	50]	d_loss: 0.6782	g_loss: 1.7770
Epoch [6/	50]	d_loss: 0.7997	g_loss: 2.7615
Epoch [6/	50]	d_loss: 0.6943	g_loss: 2.1302
Epoch [6/	50]	d_loss: 0.8565	g_loss: 1.0594
Epoch [6/	50]	d_loss: 0.6697	g_loss: 2.0521
Epoch [6/	50]	d_loss: 0.6144	g_loss: 2.6046
Epoch [6/	50]	d_loss: 0.5203	g_loss: 1.9044
Epoch [6/	50]	d_loss: 0.6780	g_loss: 2.7696
Epoch [6/	50]	d_loss: 0.6719	g_loss: 1.2351
Epoch [6/	50]	d_loss: 0.7146	g_loss: 1.1112
Epoch [6/	50]	d_loss: 1.2527	g_loss: 0.8069
Epoch [6/	50]	d_loss: 0.6385	g_loss: 2.2227
Epoch [6/	50]	d_loss: 0.5478	g_loss: 2.0015
Epoch [6/	50]	d_loss: 0.7742	g_loss: 2.0845
Epoch [6/	50]	d_loss: 0.9472	g_loss: 2.0499
Epoch [6/	50]	d_loss: 0.8173	g_loss: 2.0825
Epoch [6/	50]	d_loss: 0.6050	g_loss: 2.1232
Epoch [6/	50]	d_loss: 0.6944	g_loss: 3.3430
Epoch [6/	50]	d_loss: 0.8545	g_loss: 1.0481
Epoch [6/	50]	d_loss: 0.6542	g_loss: 2.6027
Epoch [6/	50]	d_loss: 1.0478	g_loss: 4.0876
Epoch [6/	50]	d_loss: 0.5112	g_loss: 3.0513
Epoch [6/	50]	d_loss: 0.4711	g_loss: 2.6044
Epoch [6/	50]	d_loss: 0.6276	g_loss: 1.7705
Epoch [6/	50]	d_loss: 1.0895	g_loss: 4.3202
Epoch [6/	50]	d_loss: 0.7693	g_loss: 2.9916
Epoch [6/	50]	d_loss: 0.7173	g_loss: 2.9716
Epoch [6/	50]	d_loss: 0.8622	g_loss: 1.4394
Epoch [6/	50]	d_loss: 0.6029	g_loss: 1.9366
Epoch [6/	50]	d_loss: 0.8916	g_loss: 3.3963
Epoch [6/	50]	d_loss: 0.4321	g_loss: 3.0437
Epoch [6/	50]	d_loss: 0.7690	g_loss: 1.7620
Epoch [6/	50]	d_loss: 0.4802	g_loss: 2.3885
Epoch [6/	50]	d_loss: 0.4891	g_loss: 2.7759
Epoch [6/	50]	d_loss: 0.5461	g_loss: 1.9226
Epoch [6/	50]	d_loss: 0.5729	g_loss: 1.6236
Epoch [6/	50]	d_loss: 0.6122	g_loss: 2.3467
Epoch [6/	50]	d_loss: 0.8003	g_loss: 3.5105
Epoch [6/	50]	d_loss: 0.5331	g_loss: 2.3218
Epoch [6/	50]	d_loss: 0.6059	g_loss: 1.6776

Epoch [7/	50]	d_loss: 0.5644	g_loss: 2.6413
Epoch [7/	50]	d_loss: 0.6083	g_loss: 1.7861
Epoch [7/	50]	d_loss: 0.5533	g_loss: 1.5695
Epoch [7/	50]	d_loss: 0.4379	g_loss: 3.0312
Epoch [7/	50]	d_loss: 0.6359	g_loss: 1.5099
Epoch [7/	50]	d_loss: 0.8019	g_loss: 3.0775
Epoch [7/	50]	d_loss: 0.8135	g_loss: 2.5579
Epoch [7/	50]	d_loss: 0.8161	g_loss: 1.3459
Epoch [7/	50]	d_loss: 0.4396	g_loss: 2.4442
Epoch [7/	50]	d_loss: 0.8757	g_loss: 1.9786
Epoch [7/	50]	d_loss: 0.8182	g_loss: 1.5572
Epoch [7/	50]	d_loss: 0.7123	g_loss: 1.4954
Epoch [7/	50]	d_loss: 0.8278	g_loss: 2.6266
Epoch [7/	50]	d_loss: 0.6044	g_loss: 1.3669
Epoch [7/	50]	d_loss: 0.8487	g_loss: 1.0870
Epoch [7/	50]	d_loss: 0.5750	g_loss: 2.3321
Epoch [7/	50]	d_loss: 0.5169	g_loss: 2.2065
Epoch [7/	50]	d_loss: 0.5771	g_loss: 2.2225
Epoch [7/	50]	d_loss: 0.5516	g_loss: 2.4077
Epoch [7/	50]	d_loss: 0.6679	g_loss: 1.5447
Epoch [7/	50]	d_loss: 0.7650	g_loss: 3.0198
Epoch [7/	50]	d_loss: 0.4455	g_loss: 2.5789
Epoch [7/	50]	d_loss: 0.7029	g_loss: 2.5807
Epoch [7/	50]	d_loss: 0.6978	g_loss: 1.8931
Epoch [7/	50]	d_loss: 0.5304	g_loss: 1.6226
Epoch [7/	50]	d_loss: 0.6043	g_loss: 2.1585
Epoch [7/	50]	d_loss: 1.2838	g_loss: 0.9460
Epoch [7/	50]	d_loss: 1.3265	g_loss: 3.7481
Epoch [7/	50]	d_loss: 0.5365	g_loss: 2.5344
Epoch [7/	50]	d_loss: 0.7590	g_loss: 1.2479
Epoch [7/	50]	d_loss: 0.5113	g_loss: 2.6683
Epoch [7/	50]	d_loss: 0.5484	g_loss: 2.3578
Epoch [7/	50]	d_loss: 0.6097	g_loss: 1.7955
Epoch [7/	50]	d_loss: 0.6468	g_loss: 2.2385
Epoch [7/	50]	d_loss: 0.4965	g_loss: 1.8380
Epoch [7/	50]	d_loss: 0.7619	g_loss: 1.4332
Epoch [7/	50]	d_loss: 0.8191	g_loss: 2.5933
Epoch [7/	50]	d_loss: 0.7944	g_loss: 2.1346
Epoch [7/	50]	d_loss: 0.9841	g_loss: 0.9643
Epoch [7/	50]	d_loss: 0.5934	g_loss: 2.3913
Epoch [7/	50]	d_loss: 0.5484	g_loss: 2.3638
Epoch [7/	50]	d_loss: 0.6730	g_loss: 2.7081
Epoch [7/	50]	d_loss: 0.7134	g_loss: 2.9936
Epoch [7/	50]	d_loss: 0.7602	g_loss: 1.9537
Epoch [7/	50]	d_loss: 0.6102	g_loss: 1.8320
Epoch [7/	50]	d_loss: 0.5718	g_loss: 2.1587
Epoch [7/	50]	d_loss: 0.5126	g_loss: 2.3313
Epoch [7/	50]	d_loss: 0.5489	g_loss: 1.7204

Epoch [7/	50]	d_loss: 0.9546	g_loss: 3.1951
Epoch [7/	50]	d_loss: 0.6558	g_loss: 2.4966
Epoch [7/	50]	d_loss: 0.6551	g_loss: 2.5593
Epoch [7/	50]	d_loss: 0.4546	g_loss: 2.6728
Epoch [7/	50]	d_loss: 0.7782	g_loss: 3.7030
Epoch [7/	50]	d_loss: 0.6641	g_loss: 2.1965
Epoch [7/	50]	d_loss: 0.6069	g_loss: 2.7614
Epoch [7/	50]	d_loss: 0.6328	g_loss: 1.5428
Epoch [7/	50]	d_loss: 0.5867	g_loss: 2.7061
Epoch [8/	50]	d_loss: 0.8507	g_loss: 3.7454
Epoch [8/	50]	d_loss: 0.7846	g_loss: 1.6325
Epoch [8/	50]	d_loss: 0.5471	g_loss: 1.9898
Epoch [8/	50]	d_loss: 0.6091	g_loss: 2.2491
Epoch [8/	50]	d_loss: 0.7014	g_loss: 2.8579
Epoch [8/	50]	d_loss: 0.4859	g_loss: 2.7159
Epoch [8/	50]	d_loss: 0.5956	g_loss: 1.1499
Epoch [8/	50]	d_loss: 0.6137	g_loss: 2.3597
Epoch [8/	50]	d_loss: 0.7280	g_loss: 3.5860
Epoch [8/	50]	d_loss: 0.5305	g_loss: 3.2614
Epoch [8/	50]	d_loss: 0.7500	g_loss: 2.5677
Epoch [8/	50]	d_loss: 0.6426	g_loss: 1.8236
Epoch [8/	50]	d_loss: 0.6979	g_loss: 1.9607
Epoch [8/	50]	d_loss: 0.6064	g_loss: 2.0905
Epoch [8/	50]	d_loss: 0.5075	g_loss: 2.6335
Epoch [8/	50]	d_loss: 1.1182	g_loss: 1.4977
Epoch [8/	50]	d_loss: 0.5789	g_loss: 1.9810
Epoch [8/	50]	d_loss: 0.6585	g_loss: 2.5817
Epoch [8/	50]	d_loss: 0.6523	g_loss: 2.4826
Epoch [8/	50]	d_loss: 0.6528	g_loss: 2.7803
Epoch [8/	50]	d_loss: 0.6667	g_loss: 1.4969
Epoch [8/	50]	d_loss: 0.4680	g_loss: 2.5098
Epoch [8/	50]	d_loss: 0.5040	g_loss: 2.5176
Epoch [8/	50]	d_loss: 0.4436	g_loss: 2.9092
Epoch [8/	50]	d_loss: 0.7498	g_loss: 2.9059
Epoch [8/	50]	d_loss: 0.9553	g_loss: 2.7254
Epoch [8/	50]	d_loss: 1.0797	g_loss: 3.4984
Epoch [8/	50]	d_loss: 0.7815	g_loss: 1.1791
Epoch [8/	50]	d_loss: 0.6339	g_loss: 1.5629
Epoch [8/	50]	d_loss: 0.5724	g_loss: 1.7246
Epoch [8/	50]	d_loss: 0.5093	g_loss: 2.6177
Epoch [8/	50]	d_loss: 1.0954	g_loss: 1.0182
Epoch [8/	50]	d_loss: 0.7249	g_loss: 0.9441
Epoch [8/	50]	d_loss: 0.5797	g_loss: 3.4878
Epoch [8/	50]	d_loss: 0.5155	g_loss: 2.7396
Epoch [8/	50]	d_loss: 0.5308	g_loss: 1.7761
Epoch [8/	50]	d_loss: 0.5169	g_loss: 3.3755
Epoch [8/	50]	d_loss: 1.0719	g_loss: 1.4965
Epoch [8/	50]	d_loss: 0.5535	g_loss: 1.7103

Epoch [8/	50]	d_loss: 0.4958	g_loss: 2.4377
Epoch [8/	50]	d_loss: 0.6356	g_loss: 1.7474
Epoch [8/	50]	d_loss: 0.5563	g_loss: 2.4250
Epoch [8/	50]	d_loss: 0.8060	g_loss: 3.3052
Epoch [8/	50]	d_loss: 0.7451	g_loss: 1.6400
Epoch [8/	50]	d_loss: 0.4546	g_loss: 2.0622
Epoch [8/	50]	d_loss: 0.5928	g_loss: 1.7777
Epoch [8/	50]	d_loss: 0.5450	g_loss: 1.7723
Epoch [8/	50]	d_loss: 0.7057	g_loss: 3.9911
Epoch [8/	50]	d_loss: 0.5608	g_loss: 2.5366
Epoch [8/	50]	d_loss: 0.5156	g_loss: 2.6382
Epoch [8/	50]	d_loss: 0.8737	g_loss: 2.9253
Epoch [8/	50]	d_loss: 0.5647	g_loss: 2.5396
Epoch [8/	50]	d_loss: 0.5817	g_loss: 3.1395
Epoch [8/	50]	d_loss: 0.7831	g_loss: 1.2470
Epoch [8/	50]	d_loss: 1.2514	g_loss: 0.8158
Epoch [8/	50]	d_loss: 0.5466	g_loss: 2.5828
Epoch [8/	50]	d_loss: 0.5084	g_loss: 3.3185
Epoch [9/	50]	d_loss: 1.3100	g_loss: 1.4925
Epoch [9/	50]	d_loss: 0.5318	g_loss: 1.9152
Epoch [9/	50]	d_loss: 1.1260	g_loss: 3.6696
Epoch [9/	50]	d_loss: 0.4178	g_loss: 2.5518
Epoch [9/	50]	d_loss: 0.6151	g_loss: 2.6631
Epoch [9/	50]	d_loss: 0.4801	g_loss: 3.4363
Epoch [9/	50]	d_loss: 0.5663	g_loss: 1.1490
Epoch [9/	50]	d_loss: 0.5825	g_loss: 3.3954
Epoch [9/	50]	d_loss: 0.5967	g_loss: 1.3779
Epoch [9/	50]	d_loss: 0.5790	g_loss: 2.7743
Epoch [9/	50]	d_loss: 0.6579	g_loss: 1.7425
Epoch [9/	50]	d_loss: 0.8093	g_loss: 2.0835
Epoch [9/	50]	d_loss: 0.7044	g_loss: 2.3297
Epoch [9/	50]	d_loss: 0.6578	g_loss: 3.2518
Epoch [9/	50]	d_loss: 0.8955	g_loss: 2.7462
Epoch [9/	50]	d_loss: 0.4424	g_loss: 2.8498
Epoch [9/	50]	d_loss: 0.6277	g_loss: 3.6837
Epoch [9/	50]	d_loss: 0.6513	g_loss: 2.0988
Epoch [9/	50]	d_loss: 0.6131	g_loss: 2.5552
Epoch [9/	50]	d_loss: 0.5004	g_loss: 2.2577
Epoch [9/	50]	d_loss: 0.7753	g_loss: 1.1924
Epoch [9/	50]	d_loss: 0.5144	g_loss: 2.9285
Epoch [9/	50]	d_loss: 0.7719	g_loss: 2.8548
Epoch [9/	50]	d_loss: 0.5791	g_loss: 2.3177
Epoch [9/	50]	d_loss: 0.7939	g_loss: 3.3939
Epoch [9/	50]	d_loss: 0.7035	g_loss: 1.3985
Epoch [9/	50]	d_loss: 0.7614	g_loss: 0.9290
Epoch [9/	50]	d_loss: 0.3876	g_loss: 3.1873
Epoch [9/	50]	d_loss: 0.7018	g_loss: 2.3547
Epoch [9/	50]	d_loss: 0.4499	g_loss: 1.5429

Epoch [9/	50]	d_loss: 0.4138	g_loss: 2.6059
Epoch [9/	50]	d_loss: 1.2995	g_loss: 0.7807
Epoch [9/	50]	d_loss: 0.5273	g_loss: 3.2351
Epoch [9/	50]	d_loss: 0.9668	g_loss: 1.0292
Epoch [9/	50]	d_loss: 1.0973	g_loss: 1.1506
Epoch [9/	50]	d_loss: 0.8580	g_loss: 3.7180
Epoch [9/	50]	d_loss: 0.9081	g_loss: 1.3381
Epoch [9/	50]	d_loss: 0.5155	g_loss: 3.1670
Epoch [9/	50]	d_loss: 0.9564	g_loss: 2.0254
Epoch [9/	50]	d_loss: 0.7121	g_loss: 2.2369
Epoch [9/	50]	d_loss: 0.7674	g_loss: 1.8711
Epoch [9/	50]	d_loss: 0.5668	g_loss: 2.4261
Epoch [9/	50]	d_loss: 0.5796	g_loss: 2.0219
Epoch [9/	50]	d_loss: 0.7002	g_loss: 1.3625
Epoch [9/	50]	d_loss: 0.5558	g_loss: 2.5582
Epoch [9/	50]	d_loss: 0.6634	g_loss: 2.9330
Epoch [9/	50]	d_loss: 0.5218	g_loss: 2.7535
Epoch [9/	50]	d_loss: 0.4856	g_loss: 2.3360
Epoch [9/	50]	d_loss: 0.4786	g_loss: 2.7047
Epoch [9/	50]	d_loss: 0.6844	g_loss: 1.5752
Epoch [9/	50]	d_loss: 0.5129	g_loss: 2.2986
Epoch [9/	50]	d_loss: 0.5954	g_loss: 2.7821
Epoch [9/	50]	d_loss: 0.6080	g_loss: 2.5178
Epoch [9/	50]	d_loss: 0.6376	g_loss: 2.3617
Epoch [9/	50]	d_loss: 0.4514	g_loss: 2.0631
Epoch [9/	50]	d_loss: 0.6190	g_loss: 2.7747
Epoch [9/	50]	d_loss: 0.4483	g_loss: 3.0119
Epoch [10/	50]	d_loss: 0.6317	g_loss: 2.5171
Epoch [10/	50]	d_loss: 0.6235	g_loss: 2.7626
Epoch [10/	50]	d_loss: 0.5239	g_loss: 1.7267
Epoch [10/	50]	d_loss: 0.8833	g_loss: 3.0780
Epoch [10/	50]	d_loss: 0.5313	g_loss: 1.7076
Epoch [10/	50]	d_loss: 0.4602	g_loss: 2.4395
Epoch [10/	50]	d_loss: 0.5335	g_loss: 1.7684
Epoch [10/	50]	d_loss: 0.4541	g_loss: 1.6609
Epoch [10/	50]	d_loss: 0.5121	g_loss: 2.2903
Epoch [10/	50]	d_loss: 0.6644	g_loss: 2.8120
Epoch [10/	50]	d_loss: 0.7285	g_loss: 3.0036
Epoch [10/	50]	d_loss: 1.2740	g_loss: 0.9395
Epoch [10/	50]	d_loss: 0.5736	g_loss: 2.4536
Epoch [10/	50]	d_loss: 0.4040	g_loss: 3.0060
Epoch [10/	50]	d_loss: 0.5525	g_loss: 2.4807
Epoch [10/	50]	d_loss: 0.5321	g_loss: 2.9493
Epoch [10/	50]	d_loss: 0.6057	g_loss: 2.4587
Epoch [10/	50]	d_loss: 0.8719	g_loss: 3.1021
Epoch [10/	50]	d_loss: 0.6206	g_loss: 2.2740
Epoch [10/	50]	d_loss: 0.5112	g_loss: 1.8805
Epoch [10/	50]	d_loss: 0.5150	g_loss: 3.0069

Epoch [10/	50]	d_loss: 0.4576	g_loss: 2.3863
Epoch [10/	50]	d_loss: 0.4942	g_loss: 2.3007
Epoch [10/	50]	d_loss: 0.7155	g_loss: 2.8479
Epoch [10/	50]	d_loss: 0.6646	g_loss: 2.2103
Epoch [10/	50]	d_loss: 0.4636	g_loss: 2.1205
Epoch [10/	50]	d_loss: 0.8171	g_loss: 3.8356
Epoch [10/	50]	d_loss: 0.7244	g_loss: 3.4195
Epoch [10/	50]	d_loss: 0.8199	g_loss: 3.2825
Epoch [10/	50]	d_loss: 0.5527	g_loss: 2.0827
Epoch [10/	50]	d_loss: 0.4891	g_loss: 2.3847
Epoch [10/	50]	d_loss: 0.7332	g_loss: 1.3313
Epoch [10/	50]	d_loss: 0.5209	g_loss: 2.0621
Epoch [10/	50]	d_loss: 0.5532	g_loss: 2.5093
Epoch [10/	50]	d_loss: 0.6704	g_loss: 3.8682
Epoch [10/	50]	d_loss: 0.5617	g_loss: 2.0750
Epoch [10/	50]	d_loss: 0.6068	g_loss: 2.6753
Epoch [10/	50]	d_loss: 0.5023	g_loss: 3.4814
Epoch [10/	50]	d_loss: 0.7391	g_loss: 1.4852
Epoch [10/	50]	d_loss: 0.5584	g_loss: 3.1394
Epoch [10/	50]	d_loss: 0.6622	g_loss: 1.4034
Epoch [10/	50]	d_loss: 0.4932	g_loss: 2.9722
Epoch [10/	50]	d_loss: 0.6898	g_loss: 2.8307
Epoch [10/	50]	d_loss: 0.5513	g_loss: 3.3642
Epoch [10/	50]	d_loss: 0.8773	g_loss: 3.5188
Epoch [10/	50]	d_loss: 0.5459	g_loss: 1.9784
Epoch [10/	50]	d_loss: 0.8182	g_loss: 1.0961
Epoch [10/	50]	d_loss: 0.4044	g_loss: 2.1457
Epoch [10/	50]	d_loss: 0.4150	g_loss: 3.4066
Epoch [10/	50]	d_loss: 0.5515	g_loss: 1.7933
Epoch [10/	50]	d_loss: 0.8598	g_loss: 1.4247
Epoch [10/	50]	d_loss: 0.4994	g_loss: 2.9582
Epoch [10/	50]	d_loss: 0.5327	g_loss: 2.1400
Epoch [10/	50]	d_loss: 0.6245	g_loss: 3.0496
Epoch [10/	50]	d_loss: 0.6891	g_loss: 3.7075
Epoch [10/	50]	d_loss: 0.6736	g_loss: 2.6033
Epoch [10/	50]	d_loss: 0.5040	g_loss: 2.6582
Epoch [11/	50]	d_loss: 0.4737	g_loss: 2.0059
Epoch [11/	50]	d_loss: 1.0242	g_loss: 1.5014
Epoch [11/	50]	d_loss: 0.4662	g_loss: 3.2391
Epoch [11/	50]	d_loss: 0.6037	g_loss: 1.7252
Epoch [11/	50]	d_loss: 0.6978	g_loss: 1.9929
Epoch [11/	50]	d_loss: 0.5147	g_loss: 2.8882
Epoch [11/	50]	d_loss: 0.6834	g_loss: 1.6240
Epoch [11/	50]	d_loss: 0.4729	g_loss: 3.2160
Epoch [11/	50]	d_loss: 1.0059	g_loss: 1.1612
Epoch [11/	50]	d_loss: 0.7812	g_loss: 2.9945
Epoch [11/	50]	d_loss: 0.6731	g_loss: 1.5598
Epoch [11/	50]	d_loss: 0.5579	g_loss: 2.4015

Epoch [11/	50]	d_loss: 0.5483	g_loss: 2.4725
Epoch [11/	50]	d_loss: 0.6100	g_loss: 1.7631
Epoch [11/	50]	d_loss: 0.6966	g_loss: 2.6774
Epoch [11/	50]	d_loss: 0.6753	g_loss: 3.2199
Epoch [11/	50]	d_loss: 0.4690	g_loss: 1.8159
Epoch [11/	50]	d_loss: 0.5359	g_loss: 2.5417
Epoch [11/	50]	d_loss: 0.9531	g_loss: 1.0895
Epoch [11/	50]	d_loss: 0.8541	g_loss: 3.3040
Epoch [11/	50]	d_loss: 0.5411	g_loss: 3.3809
Epoch [11/	50]	d_loss: 0.3773	g_loss: 3.4931
Epoch [11/	50]	d_loss: 0.6149	g_loss: 1.4043
Epoch [11/	50]	d_loss: 0.6046	g_loss: 2.6144
Epoch [11/	50]	d_loss: 0.5541	g_loss: 3.2513
Epoch [11/	50]	d_loss: 0.4401	g_loss: 2.7574
Epoch [11/	50]	d_loss: 0.8263	g_loss: 3.4038
Epoch [11/	50]	d_loss: 0.4777	g_loss: 3.1469
Epoch [11/	50]	d_loss: 0.7276	g_loss: 3.0448
Epoch [11/	50]	d_loss: 0.4947	g_loss: 2.0166
Epoch [11/	50]	d_loss: 0.6994	g_loss: 2.0228
Epoch [11/	50]	d_loss: 0.5183	g_loss: 2.0654
Epoch [11/	50]	d_loss: 0.8715	g_loss: 1.6246
Epoch [11/	50]	d_loss: 0.6011	g_loss: 3.2947
Epoch [11/	50]	d_loss: 0.4724	g_loss: 4.0172
Epoch [11/	50]	d_loss: 1.0754	g_loss: 2.1116
Epoch [11/	50]	d_loss: 0.6702	g_loss: 1.1383
Epoch [11/	50]	d_loss: 0.9082	g_loss: 2.3503
Epoch [11/	50]	d_loss: 0.4334	g_loss: 1.8459
Epoch [11/	50]	d_loss: 0.5478	g_loss: 2.4287
Epoch [11/	50]	d_loss: 0.6186	g_loss: 3.0400
Epoch [11/	50]	d_loss: 0.8277	g_loss: 2.9390
Epoch [11/	50]	d_loss: 0.6437	g_loss: 1.6743
Epoch [11/	50]	d_loss: 0.5666	g_loss: 2.3734
Epoch [11/	50]	d_loss: 1.3880	g_loss: 1.7862
Epoch [11/	50]	d_loss: 0.6049	g_loss: 2.9422
Epoch [11/	50]	d_loss: 0.9160	g_loss: 4.9104
Epoch [11/	50]	d_loss: 0.6176	g_loss: 2.0675
Epoch [11/	50]	d_loss: 0.5961	g_loss: 3.1287
Epoch [11/	50]	d_loss: 0.5283	g_loss: 2.3393
Epoch [11/	50]	d_loss: 0.5196	g_loss: 3.3383
Epoch [11/	50]	d_loss: 0.5840	g_loss: 2.6839
Epoch [11/	50]	d_loss: 0.6559	g_loss: 2.1205
Epoch [11/	50]	d_loss: 0.6003	g_loss: 2.4759
Epoch [11/	50]	d_loss: 0.5606	g_loss: 1.7882
Epoch [11/	50]	d_loss: 0.5442	g_loss: 2.5026
Epoch [11/	50]	d_loss: 0.7160	g_loss: 3.3552
Epoch [12/	50]	d_loss: 0.6681	g_loss: 2.6059
Epoch [12/	50]	d_loss: 0.6123	g_loss: 2.4380
Epoch [12/	50]	d_loss: 0.5541	g_loss: 1.7714

Epoch [12/	50]	d_loss: 0.4432	g_loss: 2.0209
Epoch [12/	50]	d_loss: 0.5298	g_loss: 1.7090
Epoch [12/	50]	d_loss: 0.5408	g_loss: 2.3610
Epoch [12/	50]	d_loss: 0.4785	g_loss: 3.2908
Epoch [12/	50]	d_loss: 0.4910	g_loss: 1.9445
Epoch [12/	50]	d_loss: 0.5540	g_loss: 3.6708
Epoch [12/	50]	d_loss: 1.1576	g_loss: 1.2470
Epoch [12/	50]	d_loss: 0.4475	g_loss: 2.5519
Epoch [12/	50]	d_loss: 0.3770	g_loss: 2.3700
Epoch [12/	50]	d_loss: 0.5441	g_loss: 1.9263
Epoch [12/	50]	d_loss: 0.6511	g_loss: 1.8996
Epoch [12/	50]	d_loss: 0.5049	g_loss: 2.8299
Epoch [12/	50]	d_loss: 0.8610	g_loss: 3.1217
Epoch [12/	50]	d_loss: 0.4720	g_loss: 2.4660
Epoch [12/	50]	d_loss: 1.0160	g_loss: 0.7236
Epoch [12/	50]	d_loss: 0.8159	g_loss: 1.5462
Epoch [12/	50]	d_loss: 0.4191	g_loss: 3.2613
Epoch [12/	50]	d_loss: 0.9718	g_loss: 1.2587
Epoch [12/	50]	d_loss: 0.5980	g_loss: 3.3937
Epoch [12/	50]	d_loss: 0.5261	g_loss: 3.0807
Epoch [12/	50]	d_loss: 0.5835	g_loss: 3.3456
Epoch [12/	50]	d_loss: 0.4212	g_loss: 2.6927
Epoch [12/	50]	d_loss: 0.6281	g_loss: 1.3740
Epoch [12/	50]	d_loss: 0.6458	g_loss: 1.3834
Epoch [12/	50]	d_loss: 0.5616	g_loss: 1.7540
Epoch [12/	50]	d_loss: 0.4532	g_loss: 3.2407
Epoch [12/	50]	d_loss: 0.8907	g_loss: 4.3097
Epoch [12/	50]	d_loss: 0.4161	g_loss: 2.4626
Epoch [12/	50]	d_loss: 1.0138	g_loss: 4.8878
Epoch [12/	50]	d_loss: 0.4000	g_loss: 2.3372
Epoch [12/	50]	d_loss: 0.5187	g_loss: 2.2720
Epoch [12/	50]	d_loss: 0.5260	g_loss: 1.9057
Epoch [12/	50]	d_loss: 0.7998	g_loss: 1.5046
Epoch [12/	50]	d_loss: 0.5040	g_loss: 1.8660
Epoch [12/	50]	d_loss: 0.5514	g_loss: 3.8270
Epoch [12/	50]	d_loss: 0.7213	g_loss: 3.4320
Epoch [12/	50]	d_loss: 0.4505	g_loss: 2.4257
Epoch [12/	50]	d_loss: 0.4444	g_loss: 2.9148
Epoch [12/	50]	d_loss: 0.5294	g_loss: 1.5920
Epoch [12/	50]	d_loss: 0.5427	g_loss: 2.7082
Epoch [12/	50]	d_loss: 0.6459	g_loss: 2.9714
Epoch [12/	50]	d_loss: 0.4529	g_loss: 3.1382
Epoch [12/	50]	d_loss: 0.4832	g_loss: 3.6639
Epoch [12/	50]	d_loss: 0.8403	g_loss: 3.6905
Epoch [12/	50]	d_loss: 0.4009	g_loss: 2.8006
Epoch [12/	50]	d_loss: 0.4186	g_loss: 2.5255
Epoch [12/	50]	d_loss: 0.8785	g_loss: 1.3946
Epoch [12/	50]	d_loss: 0.6126	g_loss: 2.1330

Epoch [12/	50]	d_loss: 0.4646	g_loss: 4.2760
Epoch [12/	50]	d_loss: 0.4049	g_loss: 3.0247
Epoch [12/	50]	d_loss: 0.7299	g_loss: 2.5375
Epoch [12/	50]	d_loss: 0.4579	g_loss: 3.2705
Epoch [12/	50]	d_loss: 0.4901	g_loss: 2.8736
Epoch [12/	50]	d_loss: 0.7734	g_loss: 2.8265
Epoch [13/	50]	d_loss: 0.5162	g_loss: 3.1418
Epoch [13/	50]	d_loss: 0.4493	g_loss: 1.9765
Epoch [13/	50]	d_loss: 0.9676	g_loss: 4.2314
Epoch [13/	50]	d_loss: 0.5857	g_loss: 1.5374
Epoch [13/	50]	d_loss: 0.5903	g_loss: 3.1491
Epoch [13/	50]	d_loss: 0.7665	g_loss: 1.1099
Epoch [13/	50]	d_loss: 0.5702	g_loss: 1.5085
Epoch [13/	50]	d_loss: 0.4644	g_loss: 2.0237
Epoch [13/	50]	d_loss: 0.7080	g_loss: 3.8626
Epoch [13/	50]	d_loss: 0.6073	g_loss: 3.7736
Epoch [13/	50]	d_loss: 0.4229	g_loss: 2.3452
Epoch [13/	50]	d_loss: 0.6211	g_loss: 3.3994
Epoch [13/	50]	d_loss: 0.5779	g_loss: 2.6920
Epoch [13/	50]	d_loss: 0.5194	g_loss: 1.9875
Epoch [13/	50]	d_loss: 0.5137	g_loss: 2.9715
Epoch [13/	50]	d_loss: 0.4604	g_loss: 3.3435
Epoch [13/	50]	d_loss: 0.6002	g_loss: 2.6852
Epoch [13/	50]	d_loss: 0.4501	g_loss: 3.2895
Epoch [13/	50]	d_loss: 0.4570	g_loss: 2.8101
Epoch [13/	50]	d_loss: 0.6263	g_loss: 2.0342
Epoch [13/	50]	d_loss: 0.5744	g_loss: 3.0796
Epoch [13/	50]	d_loss: 0.5762	g_loss: 2.8360
Epoch [13/	50]	d_loss: 0.5998	g_loss: 1.5706
Epoch [13/	50]	d_loss: 0.5805	g_loss: 1.0944
Epoch [13/	50]	d_loss: 0.5255	g_loss: 2.3931
Epoch [13/	50]	d_loss: 0.5123	g_loss: 2.3546
Epoch [13/	50]	d_loss: 0.6174	g_loss: 2.3558
Epoch [13/	50]	d_loss: 0.4396	g_loss: 2.6656
Epoch [13/	50]	d_loss: 0.4125	g_loss: 2.5333
Epoch [13/	50]	d_loss: 0.7160	g_loss: 0.9482
Epoch [13/	50]	d_loss: 0.6996	g_loss: 3.9030
Epoch [13/	50]	d_loss: 0.4305	g_loss: 1.9897
Epoch [13/	50]	d_loss: 0.5825	g_loss: 2.4563
Epoch [13/	50]	d_loss: 0.4367	g_loss: 2.2664
Epoch [13/	50]	d_loss: 0.4187	g_loss: 2.8132
Epoch [13/	50]	d_loss: 0.4953	g_loss: 2.1551
Epoch [13/	50]	d_loss: 0.5709	g_loss: 2.1628
Epoch [13/	50]	d_loss: 0.4377	g_loss: 2.3118
Epoch [13/	50]	d_loss: 0.5530	g_loss: 2.7771
Epoch [13/	50]	d_loss: 0.4283	g_loss: 4.0285
Epoch [13/	50]	d_loss: 0.8306	g_loss: 1.1231
Epoch [13/	50]	d_loss: 0.4397	g_loss: 3.2930

Epoch [13/	50]	d_loss: 0.4770	g_loss: 2.8466
Epoch [13/	50]	d_loss: 0.9155	g_loss: 1.4477
Epoch [13/	50]	d_loss: 0.4404	g_loss: 2.8729
Epoch [13/	50]	d_loss: 0.8195	g_loss: 3.4963
Epoch [13/	50]	d_loss: 0.6061	g_loss: 1.5218
Epoch [13/	50]	d_loss: 0.4411	g_loss: 2.8413
Epoch [13/	50]	d_loss: 0.4711	g_loss: 2.8396
Epoch [13/	50]	d_loss: 0.4276	g_loss: 1.9531
Epoch [13/	50]	d_loss: 0.4743	g_loss: 2.0724
Epoch [13/	50]	d_loss: 0.4149	g_loss: 1.7638
Epoch [13/	50]	d_loss: 0.5222	g_loss: 2.5562
Epoch [13/	50]	d_loss: 0.4519	g_loss: 2.5538
Epoch [13/	50]	d_loss: 0.4093	g_loss: 2.6156
Epoch [13/	50]	d_loss: 0.4565	g_loss: 2.3271
Epoch [13/	50]	d_loss: 0.4868	g_loss: 2.6407
Epoch [14/	50]	d_loss: 0.4852	g_loss: 2.4838
Epoch [14/	50]	d_loss: 1.0944	g_loss: 1.1274
Epoch [14/	50]	d_loss: 0.6150	g_loss: 2.0135
Epoch [14/	50]	d_loss: 0.6685	g_loss: 2.4740
Epoch [14/	50]	d_loss: 0.4808	g_loss: 3.6810
Epoch [14/	50]	d_loss: 0.4523	g_loss: 2.8541
Epoch [14/	50]	d_loss: 0.4453	g_loss: 2.4451
Epoch [14/	50]	d_loss: 0.3942	g_loss: 2.5084
Epoch [14/	50]	d_loss: 0.4222	g_loss: 3.1715
Epoch [14/	50]	d_loss: 0.4562	g_loss: 2.7473
Epoch [14/	50]	d_loss: 0.4429	g_loss: 3.4587
Epoch [14/	50]	d_loss: 0.4675	g_loss: 2.1745
Epoch [14/	50]	d_loss: 0.5376	g_loss: 2.5988
Epoch [14/	50]	d_loss: 0.5449	g_loss: 3.7657
Epoch [14/	50]	d_loss: 0.4517	g_loss: 2.7851
Epoch [14/	50]	d_loss: 0.6408	g_loss: 4.0634
Epoch [14/	50]	d_loss: 0.4376	g_loss: 2.0314
Epoch [14/	50]	d_loss: 0.4311	g_loss: 2.1795
Epoch [14/	50]	d_loss: 0.8558	g_loss: 2.3307
Epoch [14/	50]	d_loss: 0.5479	g_loss: 2.5815
Epoch [14/	50]	d_loss: 0.5431	g_loss: 1.6003
Epoch [14/	50]	d_loss: 0.6484	g_loss: 2.7873
Epoch [14/	50]	d_loss: 0.5807	g_loss: 1.4374
Epoch [14/	50]	d_loss: 0.4243	g_loss: 2.7438
Epoch [14/	50]	d_loss: 0.4544	g_loss: 3.6510
Epoch [14/	50]	d_loss: 0.5783	g_loss: 2.3238
Epoch [14/	50]	d_loss: 0.4572	g_loss: 1.9197
Epoch [14/	50]	d_loss: 0.6739	g_loss: 1.7619
Epoch [14/	50]	d_loss: 0.4198	g_loss: 3.4115
Epoch [14/	50]	d_loss: 0.4109	g_loss: 3.0213
Epoch [14/	50]	d_loss: 0.4228	g_loss: 1.5622
Epoch [14/	50]	d_loss: 0.5812	g_loss: 1.6648
Epoch [14/	50]	d_loss: 0.4871	g_loss: 2.8452

Epoch [14/	50]	d_loss: 0.4742	g_loss: 3.9350
Epoch [14/	50]	d_loss: 0.5796	g_loss: 1.6898
Epoch [14/	50]	d_loss: 0.5447	g_loss: 2.9706
Epoch [14/	50]	d_loss: 0.4132	g_loss: 3.6875
Epoch [14/	50]	d_loss: 0.4882	g_loss: 2.7628
Epoch [14/	50]	d_loss: 0.5400	g_loss: 1.5287
Epoch [14/	50]	d_loss: 0.4497	g_loss: 2.6270
Epoch [14/	50]	d_loss: 0.4954	g_loss: 2.8521
Epoch [14/	50]	d_loss: 0.4588	g_loss: 2.7610
Epoch [14/	50]	d_loss: 0.4607	g_loss: 3.6279
Epoch [14/	50]	d_loss: 0.3856	g_loss: 3.0757
Epoch [14/	50]	d_loss: 1.3897	g_loss: 5.0074
Epoch [14/	50]	d_loss: 0.9487	g_loss: 4.3432
Epoch [14/	50]	d_loss: 0.6032	g_loss: 1.2316
Epoch [14/	50]	d_loss: 0.4464	g_loss: 2.8005
Epoch [14/	50]	d_loss: 0.6000	g_loss: 4.1629
Epoch [14/	50]	d_loss: 0.6509	g_loss: 1.6953
Epoch [14/	50]	d_loss: 0.4149	g_loss: 2.3874
Epoch [14/	50]	d_loss: 0.4345	g_loss: 2.2167
Epoch [14/	50]	d_loss: 0.5176	g_loss: 1.9020
Epoch [14/	50]	d_loss: 0.4145	g_loss: 2.6019
Epoch [14/	50]	d_loss: 0.6331	g_loss: 2.5884
Epoch [14/	50]	d_loss: 0.7951	g_loss: 3.4812
Epoch [14/	50]	d_loss: 0.5175	g_loss: 3.1254
Epoch [15/	50]	d_loss: 0.4672	g_loss: 3.6989
Epoch [15/	50]	d_loss: 0.4848	g_loss: 2.5031
Epoch [15/	50]	d_loss: 0.5045	g_loss: 3.0081
Epoch [15/	50]	d_loss: 0.4106	g_loss: 2.9986
Epoch [15/	50]	d_loss: 0.4521	g_loss: 2.1425
Epoch [15/	50]	d_loss: 0.9344	g_loss: 3.9875
Epoch [15/	50]	d_loss: 0.6778	g_loss: 2.7827
Epoch [15/	50]	d_loss: 0.7080	g_loss: 1.5625
Epoch [15/	50]	d_loss: 0.4772	g_loss: 2.8452
Epoch [15/	50]	d_loss: 0.6842	g_loss: 2.4431
Epoch [15/	50]	d_loss: 0.5933	g_loss: 3.8450
Epoch [15/	50]	d_loss: 0.7465	g_loss: 1.1253
Epoch [15/	50]	d_loss: 0.4279	g_loss: 2.1727
Epoch [15/	50]	d_loss: 0.8127	g_loss: 3.6298
Epoch [15/	50]	d_loss: 0.5607	g_loss: 3.1877
Epoch [15/	50]	d_loss: 0.3844	g_loss: 3.1589
Epoch [15/	50]	d_loss: 0.4225	g_loss: 2.9109
Epoch [15/	50]	d_loss: 0.6201	g_loss: 2.3302
Epoch [15/	50]	d_loss: 0.8441	g_loss: 2.0537
Epoch [15/	50]	d_loss: 0.4294	g_loss: 4.0929
Epoch [15/	50]	d_loss: 0.5397	g_loss: 2.7645
Epoch [15/	50]	d_loss: 0.5540	g_loss: 3.6958
Epoch [15/	50]	d_loss: 0.5550	g_loss: 2.5482
Epoch [15/	50]	d_loss: 0.5647	g_loss: 2.7956

Epoch [15/	50]	d_loss: 0.4430	g_loss: 3.0873
Epoch [15/	50]	d_loss: 0.6511	g_loss: 2.2481
Epoch [15/	50]	d_loss: 0.5451	g_loss: 2.0455
Epoch [15/	50]	d_loss: 0.4034	g_loss: 2.1772
Epoch [15/	50]	d_loss: 0.5690	g_loss: 1.7262
Epoch [15/	50]	d_loss: 0.4852	g_loss: 3.5816
Epoch [15/	50]	d_loss: 0.4223	g_loss: 2.9544
Epoch [15/	50]	d_loss: 0.8444	g_loss: 3.4761
Epoch [15/	50]	d_loss: 0.3918	g_loss: 3.9258
Epoch [15/	50]	d_loss: 0.5174	g_loss: 2.4788
Epoch [15/	50]	d_loss: 0.4615	g_loss: 2.2467
Epoch [15/	50]	d_loss: 0.5600	g_loss: 3.0106
Epoch [15/	50]	d_loss: 0.4497	g_loss: 2.9050
Epoch [15/	50]	d_loss: 0.5112	g_loss: 2.2691
Epoch [15/	50]	d_loss: 0.5072	g_loss: 3.6084
Epoch [15/	50]	d_loss: 0.5268	g_loss: 2.4398
Epoch [15/	50]	d_loss: 0.5371	g_loss: 3.0747
Epoch [15/	50]	d_loss: 0.5056	g_loss: 2.8067
Epoch [15/	50]	d_loss: 0.5484	g_loss: 2.6619
Epoch [15/	50]	d_loss: 0.4086	g_loss: 3.0125
Epoch [15/	50]	d_loss: 0.5111	g_loss: 1.5971
Epoch [15/	50]	d_loss: 0.4760	g_loss: 3.3663
Epoch [15/	50]	d_loss: 0.3978	g_loss: 3.0471
Epoch [15/	50]	d_loss: 0.5425	g_loss: 2.7232
Epoch [15/	50]	d_loss: 0.4834	g_loss: 3.8571
Epoch [15/	50]	d_loss: 0.4967	g_loss: 3.0087
Epoch [15/	50]	d_loss: 0.4424	g_loss: 2.9158
Epoch [15/	50]	d_loss: 0.3955	g_loss: 2.8507
Epoch [15/	50]	d_loss: 0.5794	g_loss: 2.7216
Epoch [15/	50]	d_loss: 0.5194	g_loss: 2.1910
Epoch [15/	50]	d_loss: 0.4334	g_loss: 2.6402
Epoch [15/	50]	d_loss: 0.4545	g_loss: 2.8027
Epoch [15/	50]	d_loss: 0.4342	g_loss: 2.9777
Epoch [16/	50]	d_loss: 0.6438	g_loss: 3.2258
Epoch [16/	50]	d_loss: 0.4412	g_loss: 1.9436
Epoch [16/	50]	d_loss: 0.5294	g_loss: 3.5052
Epoch [16/	50]	d_loss: 0.7020	g_loss: 1.3875
Epoch [16/	50]	d_loss: 0.4822	g_loss: 2.0660
Epoch [16/	50]	d_loss: 0.5086	g_loss: 2.5483
Epoch [16/	50]	d_loss: 0.4965	g_loss: 2.0730
Epoch [16/	50]	d_loss: 0.4837	g_loss: 3.3506
Epoch [16/	50]	d_loss: 0.6371	g_loss: 3.7091
Epoch [16/	50]	d_loss: 0.5333	g_loss: 2.7952
Epoch [16/	50]	d_loss: 0.5150	g_loss: 2.6887
Epoch [16/	50]	d_loss: 0.5170	g_loss: 3.2302
Epoch [16/	50]	d_loss: 0.9121	g_loss: 2.0929
Epoch [16/	50]	d_loss: 0.7474	g_loss: 4.1293
Epoch [16/	50]	d_loss: 0.5134	g_loss: 3.1731

Epoch [16/	50]	d_loss: 0.4838	g_loss: 3.3863
Epoch [16/	50]	d_loss: 0.5191	g_loss: 2.8519
Epoch [16/	50]	d_loss: 0.5059	g_loss: 2.5613
Epoch [16/	50]	d_loss: 0.5652	g_loss: 1.8993
Epoch [16/	50]	d_loss: 0.6696	g_loss: 3.7917
Epoch [16/	50]	d_loss: 0.3754	g_loss: 4.6636
Epoch [16/	50]	d_loss: 0.4295	g_loss: 2.3029
Epoch [16/	50]	d_loss: 0.5724	g_loss: 2.6887
Epoch [16/	50]	d_loss: 0.4316	g_loss: 3.2641
Epoch [16/	50]	d_loss: 0.5585	g_loss: 2.4629
Epoch [16/	50]	d_loss: 0.4977	g_loss: 2.3117
Epoch [16/	50]	d_loss: 0.4745	g_loss: 3.3025
Epoch [16/	50]	d_loss: 0.4884	g_loss: 2.4124
Epoch [16/	50]	d_loss: 0.4380	g_loss: 3.4614
Epoch [16/	50]	d_loss: 0.4249	g_loss: 2.6315
Epoch [16/	50]	d_loss: 0.4893	g_loss: 2.3848
Epoch [16/	50]	d_loss: 0.3749	g_loss: 2.3462
Epoch [16/	50]	d_loss: 0.4886	g_loss: 3.4539
Epoch [16/	50]	d_loss: 1.0437	g_loss: 1.6458
Epoch [16/	50]	d_loss: 0.4698	g_loss: 2.4702
Epoch [16/	50]	d_loss: 0.5316	g_loss: 2.3535
Epoch [16/	50]	d_loss: 0.4934	g_loss: 2.9501
Epoch [16/	50]	d_loss: 0.6440	g_loss: 2.5254
Epoch [16/	50]	d_loss: 0.3754	g_loss: 3.5156
Epoch [16/	50]	d_loss: 0.7533	g_loss: 1.8005
Epoch [16/	50]	d_loss: 0.5067	g_loss: 3.3669
Epoch [16/	50]	d_loss: 0.6208	g_loss: 3.3631
Epoch [16/	50]	d_loss: 0.4719	g_loss: 3.2108
Epoch [16/	50]	d_loss: 0.6713	g_loss: 2.9078
Epoch [16/	50]	d_loss: 0.4616	g_loss: 3.3230
Epoch [16/	50]	d_loss: 0.4841	g_loss: 4.1218
Epoch [16/	50]	d_loss: 0.4339	g_loss: 2.5142
Epoch [16/	50]	d_loss: 0.4988	g_loss: 2.4230
Epoch [16/	50]	d_loss: 0.4946	g_loss: 2.9312
Epoch [16/	50]	d_loss: 0.6839	g_loss: 1.2759
Epoch [16/	50]	d_loss: 0.5246	g_loss: 3.7556
Epoch [16/	50]	d_loss: 0.4449	g_loss: 3.0980
Epoch [16/	50]	d_loss: 0.3749	g_loss: 3.7731
Epoch [16/	50]	d_loss: 0.4961	g_loss: 3.3761
Epoch [16/	50]	d_loss: 0.4740	g_loss: 1.4245
Epoch [16/	50]	d_loss: 0.6447	g_loss: 4.5772
Epoch [16/	50]	d_loss: 0.4566	g_loss: 2.9244
Epoch [17/	50]	d_loss: 0.4509	g_loss: 2.7589
Epoch [17/	50]	d_loss: 0.9807	g_loss: 1.5151
Epoch [17/	50]	d_loss: 0.4238	g_loss: 3.1756
Epoch [17/	50]	d_loss: 0.5428	g_loss: 3.1493
Epoch [17/	50]	d_loss: 0.4109	g_loss: 3.6122
Epoch [17/	50]	d_loss: 0.8564	g_loss: 1.4799

Epoch [17/	50]	d_loss: 0.4815	g_loss: 4.0539
Epoch [17/	50]	d_loss: 0.6407	g_loss: 3.0030
Epoch [17/	50]	d_loss: 0.4162	g_loss: 2.7051
Epoch [17/	50]	d_loss: 0.4426	g_loss: 3.0569
Epoch [17/	50]	d_loss: 0.5145	g_loss: 2.5586
Epoch [17/	50]	d_loss: 0.5329	g_loss: 2.9028
Epoch [17/	50]	d_loss: 0.5866	g_loss: 3.6830
Epoch [17/	50]	d_loss: 0.5926	g_loss: 2.1067
Epoch [17/	50]	d_loss: 0.4662	g_loss: 4.0583
Epoch [17/	50]	d_loss: 0.3785	g_loss: 3.3872
Epoch [17/	50]	d_loss: 0.5754	g_loss: 3.0822
Epoch [17/	50]	d_loss: 0.5152	g_loss: 3.4258
Epoch [17/	50]	d_loss: 0.4393	g_loss: 2.7229
Epoch [17/	50]	d_loss: 0.7501	g_loss: 1.4166
Epoch [17/	50]	d_loss: 0.4509	g_loss: 2.7329
Epoch [17/	50]	d_loss: 0.7753	g_loss: 1.8111
Epoch [17/	50]	d_loss: 0.4100	g_loss: 3.1042
Epoch [17/	50]	d_loss: 0.8582	g_loss: 3.3922
Epoch [17/	50]	d_loss: 0.4946	g_loss: 3.1694
Epoch [17/	50]	d_loss: 0.4480	g_loss: 3.1270
Epoch [17/	50]	d_loss: 0.5121	g_loss: 2.5645
Epoch [17/	50]	d_loss: 0.4615	g_loss: 3.5710
Epoch [17/	50]	d_loss: 0.5277	g_loss: 2.3513
Epoch [17/	50]	d_loss: 0.5567	g_loss: 3.2975
Epoch [17/	50]	d_loss: 0.4622	g_loss: 3.0822
Epoch [17/	50]	d_loss: 0.6198	g_loss: 3.7857
Epoch [17/	50]	d_loss: 0.4756	g_loss: 4.3247
Epoch [17/	50]	d_loss: 0.9667	g_loss: 4.1574
Epoch [17/	50]	d_loss: 0.4081	g_loss: 2.4134
Epoch [17/	50]	d_loss: 0.4721	g_loss: 3.1012
Epoch [17/	50]	d_loss: 0.4303	g_loss: 3.4541
Epoch [17/	50]	d_loss: 0.4650	g_loss: 2.0629
Epoch [17/	50]	d_loss: 0.6317	g_loss: 4.9741
Epoch [17/	50]	d_loss: 0.4124	g_loss: 3.1380
Epoch [17/	50]	d_loss: 0.5280	g_loss: 2.0556
Epoch [17/	50]	d_loss: 0.5221	g_loss: 2.3949
Epoch [17/	50]	d_loss: 0.6834	g_loss: 3.2663
Epoch [17/	50]	d_loss: 0.6594	g_loss: 1.4751
Epoch [17/	50]	d_loss: 0.4019	g_loss: 3.9323
Epoch [17/	50]	d_loss: 0.4954	g_loss: 3.4759
Epoch [17/	50]	d_loss: 0.7725	g_loss: 4.9687
Epoch [17/	50]	d_loss: 0.5249	g_loss: 2.4079
Epoch [17/	50]	d_loss: 0.5708	g_loss: 2.6347
Epoch [17/	50]	d_loss: 0.6330	g_loss: 2.4381
Epoch [17/	50]	d_loss: 0.6265	g_loss: 2.4210
Epoch [17/	50]	d_loss: 0.4777	g_loss: 3.1836
Epoch [17/	50]	d_loss: 0.7283	g_loss: 2.6934
Epoch [17/	50]	d_loss: 0.6666	g_loss: 3.4357

Epoch [17/	50]	d_loss: 0.4992	g_loss: 2.8498
Epoch [17/	50]	d_loss: 0.4069	g_loss: 2.6037
Epoch [17/	50]	d_loss: 0.6689	g_loss: 1.8767
Epoch [18/	50]	d_loss: 0.4564	g_loss: 2.3230
Epoch [18/	50]	d_loss: 0.3747	g_loss: 3.3670
Epoch [18/	50]	d_loss: 0.4138	g_loss: 2.1282
Epoch [18/	50]	d_loss: 0.5148	g_loss: 2.5606
Epoch [18/	50]	d_loss: 0.7399	g_loss: 1.0264
Epoch [18/	50]	d_loss: 0.5002	g_loss: 1.7727
Epoch [18/	50]	d_loss: 0.4066	g_loss: 3.0570
Epoch [18/	50]	d_loss: 0.4074	g_loss: 3.7158
Epoch [18/	50]	d_loss: 0.4793	g_loss: 3.4838
Epoch [18/	50]	d_loss: 1.2057	g_loss: 1.0552
Epoch [18/	50]	d_loss: 0.3965	g_loss: 3.0083
Epoch [18/	50]	d_loss: 0.4985	g_loss: 3.0450
Epoch [18/	50]	d_loss: 0.4431	g_loss: 2.8660
Epoch [18/	50]	d_loss: 0.4230	g_loss: 4.1062
Epoch [18/	50]	d_loss: 0.5992	g_loss: 1.6557
Epoch [18/	50]	d_loss: 0.4821	g_loss: 2.1474
Epoch [18/	50]	d_loss: 0.4866	g_loss: 3.6424
Epoch [18/	50]	d_loss: 0.4130	g_loss: 2.7073
Epoch [18/	50]	d_loss: 0.5253	g_loss: 2.7608
Epoch [18/	50]	d_loss: 0.4348	g_loss: 3.3185
Epoch [18/	50]	d_loss: 0.4945	g_loss: 2.7809
Epoch [18/	50]	d_loss: 0.5074	g_loss: 2.7726
Epoch [18/	50]	d_loss: 0.6106	g_loss: 3.1866
Epoch [18/	50]	d_loss: 0.5818	g_loss: 2.3421
Epoch [18/	50]	d_loss: 1.0199	g_loss: 1.0290
Epoch [18/	50]	d_loss: 0.6298	g_loss: 2.2221
Epoch [18/	50]	d_loss: 0.4947	g_loss: 3.0508
Epoch [18/	50]	d_loss: 0.6319	g_loss: 1.5631
Epoch [18/	50]	d_loss: 0.7976	g_loss: 0.8382
Epoch [18/	50]	d_loss: 0.6643	g_loss: 4.4091
Epoch [18/	50]	d_loss: 0.3913	g_loss: 4.0758
Epoch [18/	50]	d_loss: 0.5606	g_loss: 1.7798
Epoch [18/	50]	d_loss: 0.5622	g_loss: 3.4482
Epoch [18/	50]	d_loss: 0.3910	g_loss: 2.6647
Epoch [18/	50]	d_loss: 0.5164	g_loss: 2.5506
Epoch [18/	50]	d_loss: 0.3832	g_loss: 3.7775
Epoch [18/	50]	d_loss: 0.4181	g_loss: 3.5270
Epoch [18/	50]	d_loss: 0.5605	g_loss: 3.3371
Epoch [18/	50]	d_loss: 0.4996	g_loss: 2.6894
Epoch [18/	50]	d_loss: 0.5062	g_loss: 3.3824
Epoch [18/	50]	d_loss: 0.4847	g_loss: 3.4581
Epoch [18/	50]	d_loss: 0.6089	g_loss: 1.2356
Epoch [18/	50]	d_loss: 0.4540	g_loss: 3.5271
Epoch [18/	50]	d_loss: 0.4283	g_loss: 2.3711
Epoch [18/	50]	d_loss: 0.4596	g_loss: 3.5609

Epoch [18/	50]	d_loss: 0.4036	g_loss: 3.1512
Epoch [18/	50]	d_loss: 0.4597	g_loss: 3.3989
Epoch [18/	50]	d_loss: 0.6983	g_loss: 1.5950
Epoch [18/	50]	d_loss: 0.5551	g_loss: 2.2639
Epoch [18/	50]	d_loss: 0.3668	g_loss: 3.8598
Epoch [18/	50]	d_loss: 0.5636	g_loss: 1.7091
Epoch [18/	50]	d_loss: 0.4622	g_loss: 3.7061
Epoch [18/	50]	d_loss: 0.4139	g_loss: 3.6018
Epoch [18/	50]	d_loss: 0.4120	g_loss: 3.4046
Epoch [18/	50]	d_loss: 0.6760	g_loss: 1.3189
Epoch [18/	50]	d_loss: 0.7859	g_loss: 1.4207
Epoch [18/	50]	d_loss: 0.6247	g_loss: 2.0825
Epoch [19/	50]	d_loss: 0.5743	g_loss: 3.9095
Epoch [19/	50]	d_loss: 0.5808	g_loss: 2.4896
Epoch [19/	50]	d_loss: 0.3931	g_loss: 2.7838
Epoch [19/	50]	d_loss: 0.4526	g_loss: 2.6477
Epoch [19/	50]	d_loss: 0.8005	g_loss: 1.1977
Epoch [19/	50]	d_loss: 0.5125	g_loss: 2.9524
Epoch [19/	50]	d_loss: 0.4941	g_loss: 3.9780
Epoch [19/	50]	d_loss: 0.4856	g_loss: 2.4288
Epoch [19/	50]	d_loss: 0.4171	g_loss: 3.0626
Epoch [19/	50]	d_loss: 0.6100	g_loss: 3.7280
Epoch [19/	50]	d_loss: 0.5503	g_loss: 2.7870
Epoch [19/	50]	d_loss: 0.3969	g_loss: 3.1124
Epoch [19/	50]	d_loss: 0.5881	g_loss: 2.4219
Epoch [19/	50]	d_loss: 0.4578	g_loss: 2.3962
Epoch [19/	50]	d_loss: 0.4368	g_loss: 1.9033
Epoch [19/	50]	d_loss: 0.4851	g_loss: 2.8952
Epoch [19/	50]	d_loss: 0.6433	g_loss: 3.7831
Epoch [19/	50]	d_loss: 0.6308	g_loss: 4.0917
Epoch [19/	50]	d_loss: 0.5574	g_loss: 2.6117
Epoch [19/	50]	d_loss: 1.0387	g_loss: 1.1159
Epoch [19/	50]	d_loss: 0.4109	g_loss: 3.1165
Epoch [19/	50]	d_loss: 0.5494	g_loss: 2.0592
Epoch [19/	50]	d_loss: 0.4963	g_loss: 2.1971
Epoch [19/	50]	d_loss: 0.4323	g_loss: 3.1367
Epoch [19/	50]	d_loss: 0.5261	g_loss: 2.9978
Epoch [19/	50]	d_loss: 0.5304	g_loss: 2.0337
Epoch [19/	50]	d_loss: 0.4293	g_loss: 3.3944
Epoch [19/	50]	d_loss: 0.6567	g_loss: 1.6053
Epoch [19/	50]	d_loss: 0.4121	g_loss: 3.4681
Epoch [19/	50]	d_loss: 0.3868	g_loss: 2.9874
Epoch [19/	50]	d_loss: 0.5644	g_loss: 3.1647
Epoch [19/	50]	d_loss: 0.4622	g_loss: 2.4341
Epoch [19/	50]	d_loss: 0.4893	g_loss: 2.7052
Epoch [19/	50]	d_loss: 0.7085	g_loss: 3.9177
Epoch [19/	50]	d_loss: 0.4243	g_loss: 2.5564
Epoch [19/	50]	d_loss: 0.3952	g_loss: 3.2360

Epoch [19/	50]	d_loss: 0.6313	g_loss: 2.0203
Epoch [19/	50]	d_loss: 0.7091	g_loss: 3.8586
Epoch [19/	50]	d_loss: 0.5638	g_loss: 2.3061
Epoch [19/	50]	d_loss: 0.4387	g_loss: 3.8212
Epoch [19/	50]	d_loss: 0.3822	g_loss: 2.8271
Epoch [19/	50]	d_loss: 0.3851	g_loss: 3.3450
Epoch [19/	50]	d_loss: 0.4708	g_loss: 3.4865
Epoch [19/	50]	d_loss: 0.5194	g_loss: 4.1179
Epoch [19/	50]	d_loss: 0.4708	g_loss: 2.1337
Epoch [19/	50]	d_loss: 0.4075	g_loss: 4.0590
Epoch [19/	50]	d_loss: 0.3763	g_loss: 3.3825
Epoch [19/	50]	d_loss: 0.5513	g_loss: 2.4719
Epoch [19/	50]	d_loss: 0.6817	g_loss: 1.6677
Epoch [19/	50]	d_loss: 0.7084	g_loss: 2.4836
Epoch [19/	50]	d_loss: 0.5846	g_loss: 1.3467
Epoch [19/	50]	d_loss: 0.5949	g_loss: 3.7689
Epoch [19/	50]	d_loss: 0.5190	g_loss: 1.8554
Epoch [19/	50]	d_loss: 0.5309	g_loss: 2.6834
Epoch [19/	50]	d_loss: 0.5067	g_loss: 4.0737
Epoch [19/	50]	d_loss: 0.4104	g_loss: 2.6887
Epoch [19/	50]	d_loss: 0.4615	g_loss: 3.7719
Epoch [20/	50]	d_loss: 2.2757	g_loss: 0.9093
Epoch [20/	50]	d_loss: 0.4020	g_loss: 2.8569
Epoch [20/	50]	d_loss: 0.3846	g_loss: 4.4978
Epoch [20/	50]	d_loss: 0.4092	g_loss: 3.0786
Epoch [20/	50]	d_loss: 0.4748	g_loss: 2.4125
Epoch [20/	50]	d_loss: 0.5721	g_loss: 2.9489
Epoch [20/	50]	d_loss: 0.5978	g_loss: 3.9846
Epoch [20/	50]	d_loss: 0.3915	g_loss: 3.2621
Epoch [20/	50]	d_loss: 0.4764	g_loss: 2.1724
Epoch [20/	50]	d_loss: 0.3870	g_loss: 3.5240
Epoch [20/	50]	d_loss: 0.5161	g_loss: 2.3199
Epoch [20/	50]	d_loss: 0.5051	g_loss: 3.2357
Epoch [20/	50]	d_loss: 0.3959	g_loss: 2.4396
Epoch [20/	50]	d_loss: 0.3977	g_loss: 2.6108
Epoch [20/	50]	d_loss: 0.4392	g_loss: 2.5590
Epoch [20/	50]	d_loss: 0.4591	g_loss: 3.0120
Epoch [20/	50]	d_loss: 0.6116	g_loss: 2.4265
Epoch [20/	50]	d_loss: 0.4650	g_loss: 2.1078
Epoch [20/	50]	d_loss: 0.5165	g_loss: 2.3364
Epoch [20/	50]	d_loss: 0.8128	g_loss: 1.3823
Epoch [20/	50]	d_loss: 0.6418	g_loss: 3.5996
Epoch [20/	50]	d_loss: 0.4017	g_loss: 3.1049
Epoch [20/	50]	d_loss: 0.5781	g_loss: 1.9725
Epoch [20/	50]	d_loss: 0.4352	g_loss: 2.4336
Epoch [20/	50]	d_loss: 0.4610	g_loss: 2.1323
Epoch [20/	50]	d_loss: 0.4815	g_loss: 2.6751
Epoch [20/	50]	d_loss: 0.8166	g_loss: 4.0355

Epoch [20/	50]	d_loss: 0.3907	g_loss: 3.3657
Epoch [20/	50]	d_loss: 0.4565	g_loss: 4.0036
Epoch [20/	50]	d_loss: 0.7208	g_loss: 5.4050
Epoch [20/	50]	d_loss: 0.4313	g_loss: 2.9051
Epoch [20/	50]	d_loss: 0.5344	g_loss: 3.4097
Epoch [20/	50]	d_loss: 0.4486	g_loss: 3.9357
Epoch [20/	50]	d_loss: 0.4176	g_loss: 2.5438
Epoch [20/	50]	d_loss: 0.6582	g_loss: 3.2608
Epoch [20/	50]	d_loss: 0.9535	g_loss: 4.9830
Epoch [20/	50]	d_loss: 0.5765	g_loss: 1.8520
Epoch [20/	50]	d_loss: 0.5039	g_loss: 2.4693
Epoch [20/	50]	d_loss: 0.8194	g_loss: 4.4372
Epoch [20/	50]	d_loss: 0.4322	g_loss: 2.1626
Epoch [20/	50]	d_loss: 0.6306	g_loss: 1.5399
Epoch [20/	50]	d_loss: 0.9547	g_loss: 1.3728
Epoch [20/	50]	d_loss: 0.5371	g_loss: 2.4749
Epoch [20/	50]	d_loss: 0.3884	g_loss: 4.3787
Epoch [20/	50]	d_loss: 0.6467	g_loss: 1.9382
Epoch [20/	50]	d_loss: 0.4780	g_loss: 2.5718
Epoch [20/	50]	d_loss: 0.4366	g_loss: 2.5854
Epoch [20/	50]	d_loss: 0.3985	g_loss: 2.4793
Epoch [20/	50]	d_loss: 0.4090	g_loss: 3.0293
Epoch [20/	50]	d_loss: 0.5160	g_loss: 2.9724
Epoch [20/	50]	d_loss: 0.4996	g_loss: 2.6260
Epoch [20/	50]	d_loss: 0.5998	g_loss: 2.5094
Epoch [20/	50]	d_loss: 0.5074	g_loss: 2.4143
Epoch [20/	50]	d_loss: 0.4893	g_loss: 3.4133
Epoch [20/	50]	d_loss: 0.4770	g_loss: 2.6505
Epoch [20/	50]	d_loss: 0.3741	g_loss: 4.3994
Epoch [20/	50]	d_loss: 0.6729	g_loss: 3.4171
Epoch [21/	50]	d_loss: 0.4945	g_loss: 2.2271
Epoch [21/	50]	d_loss: 0.4105	g_loss: 2.9992
Epoch [21/	50]	d_loss: 0.4663	g_loss: 2.9977
Epoch [21/	50]	d_loss: 0.4491	g_loss: 3.2100
Epoch [21/	50]	d_loss: 0.4737	g_loss: 3.9027
Epoch [21/	50]	d_loss: 0.6368	g_loss: 3.2427
Epoch [21/	50]	d_loss: 0.6342	g_loss: 1.9933
Epoch [21/	50]	d_loss: 0.4907	g_loss: 3.4996
Epoch [21/	50]	d_loss: 0.4739	g_loss: 2.8416
Epoch [21/	50]	d_loss: 0.4943	g_loss: 2.9070
Epoch [21/	50]	d_loss: 0.9202	g_loss: 3.9572
Epoch [21/	50]	d_loss: 0.4184	g_loss: 3.2161
Epoch [21/	50]	d_loss: 0.3815	g_loss: 3.6016
Epoch [21/	50]	d_loss: 0.4759	g_loss: 3.4828
Epoch [21/	50]	d_loss: 0.4690	g_loss: 3.6989
Epoch [21/	50]	d_loss: 0.4914	g_loss: 3.4074
Epoch [21/	50]	d_loss: 0.4042	g_loss: 3.1116
Epoch [21/	50]	d_loss: 0.4338	g_loss: 2.9105

Epoch [21/	50]	d_loss: 0.4936	g_loss: 4.1870
Epoch [21/	50]	d_loss: 0.5190	g_loss: 3.5250
Epoch [21/	50]	d_loss: 0.4744	g_loss: 3.4909
Epoch [21/	50]	d_loss: 0.4361	g_loss: 3.3833
Epoch [21/	50]	d_loss: 0.4948	g_loss: 1.8480
Epoch [21/	50]	d_loss: 0.4056	g_loss: 2.9490
Epoch [21/	50]	d_loss: 0.4786	g_loss: 2.0875
Epoch [21/	50]	d_loss: 0.4799	g_loss: 3.7971
Epoch [21/	50]	d_loss: 0.4393	g_loss: 2.7921
Epoch [21/	50]	d_loss: 0.5425	g_loss: 2.3076
Epoch [21/	50]	d_loss: 0.4232	g_loss: 3.4841
Epoch [21/	50]	d_loss: 0.4057	g_loss: 3.2277
Epoch [21/	50]	d_loss: 0.4173	g_loss: 1.9121
Epoch [21/	50]	d_loss: 0.9008	g_loss: 4.7008
Epoch [21/	50]	d_loss: 0.4692	g_loss: 3.8953
Epoch [21/	50]	d_loss: 0.3991	g_loss: 3.5857
Epoch [21/	50]	d_loss: 0.5450	g_loss: 2.2062
Epoch [21/	50]	d_loss: 0.4373	g_loss: 2.9993
Epoch [21/	50]	d_loss: 0.4044	g_loss: 4.1102
Epoch [21/	50]	d_loss: 0.3815	g_loss: 3.5675
Epoch [21/	50]	d_loss: 0.5016	g_loss: 2.6328
Epoch [21/	50]	d_loss: 0.5720	g_loss: 3.0960
Epoch [21/	50]	d_loss: 0.4260	g_loss: 3.0147
Epoch [21/	50]	d_loss: 0.5337	g_loss: 3.1843
Epoch [21/	50]	d_loss: 0.4016	g_loss: 3.7754
Epoch [21/	50]	d_loss: 0.4384	g_loss: 2.8570
Epoch [21/	50]	d_loss: 0.4561	g_loss: 2.6629
Epoch [21/	50]	d_loss: 0.4530	g_loss: 3.0550
Epoch [21/	50]	d_loss: 0.5272	g_loss: 2.0759
Epoch [21/	50]	d_loss: 0.3918	g_loss: 3.5115
Epoch [21/	50]	d_loss: 0.5068	g_loss: 2.0986
Epoch [21/	50]	d_loss: 0.3874	g_loss: 2.5886
Epoch [21/	50]	d_loss: 0.4088	g_loss: 3.7388
Epoch [21/	50]	d_loss: 0.4261	g_loss: 3.2827
Epoch [21/	50]	d_loss: 0.4403	g_loss: 3.8144
Epoch [21/	50]	d_loss: 0.7726	g_loss: 1.6069
Epoch [21/	50]	d_loss: 0.5687	g_loss: 3.5669
Epoch [21/	50]	d_loss: 0.4077	g_loss: 3.0326
Epoch [21/	50]	d_loss: 0.5247	g_loss: 2.2022
Epoch [22/	50]	d_loss: 0.5772	g_loss: 3.2754
Epoch [22/	50]	d_loss: 0.4011	g_loss: 3.6552
Epoch [22/	50]	d_loss: 0.5044	g_loss: 2.8314
Epoch [22/	50]	d_loss: 0.6977	g_loss: 4.2879
Epoch [22/	50]	d_loss: 0.4992	g_loss: 3.9077
Epoch [22/	50]	d_loss: 0.5200	g_loss: 2.1520
Epoch [22/	50]	d_loss: 0.4171	g_loss: 3.1931
Epoch [22/	50]	d_loss: 0.4150	g_loss: 3.1791
Epoch [22/	50]	d_loss: 0.5206	g_loss: 3.0977

Epoch [22/	50]	d_loss: 0.4191	g_loss: 4.5337
Epoch [22/	50]	d_loss: 0.4607	g_loss: 3.7121
Epoch [22/	50]	d_loss: 0.4954	g_loss: 3.5963
Epoch [22/	50]	d_loss: 1.1333	g_loss: 1.8939
Epoch [22/	50]	d_loss: 0.4221	g_loss: 2.2134
Epoch [22/	50]	d_loss: 0.5370	g_loss: 4.0734
Epoch [22/	50]	d_loss: 0.3793	g_loss: 3.5949
Epoch [22/	50]	d_loss: 0.5258	g_loss: 4.1880
Epoch [22/	50]	d_loss: 0.4416	g_loss: 3.8604
Epoch [22/	50]	d_loss: 0.8865	g_loss: 4.1736
Epoch [22/	50]	d_loss: 0.3988	g_loss: 3.8756
Epoch [22/	50]	d_loss: 0.4848	g_loss: 3.1951
Epoch [22/	50]	d_loss: 0.6469	g_loss: 2.4284
Epoch [22/	50]	d_loss: 0.6777	g_loss: 1.7672
Epoch [22/	50]	d_loss: 0.8575	g_loss: 3.7717
Epoch [22/	50]	d_loss: 0.4653	g_loss: 4.0292
Epoch [22/	50]	d_loss: 0.3834	g_loss: 2.8604
Epoch [22/	50]	d_loss: 0.3890	g_loss: 3.3602
Epoch [22/	50]	d_loss: 0.5034	g_loss: 2.8852
Epoch [22/	50]	d_loss: 0.3824	g_loss: 3.0502
Epoch [22/	50]	d_loss: 0.6025	g_loss: 3.2609
Epoch [22/	50]	d_loss: 0.6559	g_loss: 1.2739
Epoch [22/	50]	d_loss: 0.5064	g_loss: 2.7649
Epoch [22/	50]	d_loss: 0.4016	g_loss: 2.7914
Epoch [22/	50]	d_loss: 0.4287	g_loss: 2.4077
Epoch [22/	50]	d_loss: 0.4462	g_loss: 2.6557
Epoch [22/	50]	d_loss: 0.4014	g_loss: 4.0843
Epoch [22/	50]	d_loss: 0.4202	g_loss: 4.1510
Epoch [22/	50]	d_loss: 0.4657	g_loss: 2.3678
Epoch [22/	50]	d_loss: 0.4211	g_loss: 3.8332
Epoch [22/	50]	d_loss: 0.3775	g_loss: 3.9344
Epoch [22/	50]	d_loss: 0.4529	g_loss: 3.7440
Epoch [22/	50]	d_loss: 0.6286	g_loss: 3.6457
Epoch [22/	50]	d_loss: 0.4273	g_loss: 3.8618
Epoch [22/	50]	d_loss: 0.4891	g_loss: 4.1055
Epoch [22/	50]	d_loss: 0.4618	g_loss: 2.9411
Epoch [22/	50]	d_loss: 0.5649	g_loss: 4.0381
Epoch [22/	50]	d_loss: 0.4125	g_loss: 3.2902
Epoch [22/	50]	d_loss: 0.4162	g_loss: 3.2901
Epoch [22/	50]	d_loss: 0.3928	g_loss: 3.2743
Epoch [22/	50]	d_loss: 0.4136	g_loss: 2.5400
Epoch [22/	50]	d_loss: 0.4530	g_loss: 2.9324
Epoch [22/	50]	d_loss: 1.0746	g_loss: 1.4203
Epoch [22/	50]	d_loss: 0.4104	g_loss: 3.1080
Epoch [22/	50]	d_loss: 0.4615	g_loss: 2.1828
Epoch [22/	50]	d_loss: 0.4578	g_loss: 3.7214
Epoch [22/	50]	d_loss: 0.5090	g_loss: 2.6726
Epoch [22/	50]	d_loss: 0.3983	g_loss: 3.9871

Epoch [23/	50]	d_loss: 0.4239	g_loss: 3.1150
Epoch [23/	50]	d_loss: 0.5526	g_loss: 2.0581
Epoch [23/	50]	d_loss: 0.6001	g_loss: 2.8886
Epoch [23/	50]	d_loss: 1.4649	g_loss: 0.9862
Epoch [23/	50]	d_loss: 0.4386	g_loss: 2.7452
Epoch [23/	50]	d_loss: 0.3990	g_loss: 3.2990
Epoch [23/	50]	d_loss: 0.4706	g_loss: 3.3311
Epoch [23/	50]	d_loss: 0.5257	g_loss: 3.2963
Epoch [23/	50]	d_loss: 0.3762	g_loss: 3.1197
Epoch [23/	50]	d_loss: 0.4545	g_loss: 2.6729
Epoch [23/	50]	d_loss: 0.4791	g_loss: 2.3744
Epoch [23/	50]	d_loss: 0.4024	g_loss: 3.9652
Epoch [23/	50]	d_loss: 0.4070	g_loss: 3.7659
Epoch [23/	50]	d_loss: 0.4240	g_loss: 3.9249
Epoch [23/	50]	d_loss: 0.4631	g_loss: 4.3083
Epoch [23/	50]	d_loss: 0.4095	g_loss: 2.9374
Epoch [23/	50]	d_loss: 0.4032	g_loss: 3.5021
Epoch [23/	50]	d_loss: 0.3852	g_loss: 2.4734
Epoch [23/	50]	d_loss: 0.4423	g_loss: 3.5272
Epoch [23/	50]	d_loss: 0.3801	g_loss: 3.3453
Epoch [23/	50]	d_loss: 0.5738	g_loss: 1.4032
Epoch [23/	50]	d_loss: 0.4012	g_loss: 4.2090
Epoch [23/	50]	d_loss: 0.5545	g_loss: 2.7923
Epoch [23/	50]	d_loss: 0.5110	g_loss: 2.9718
Epoch [23/	50]	d_loss: 0.7552	g_loss: 3.8019
Epoch [23/	50]	d_loss: 0.4349	g_loss: 2.5500
Epoch [23/	50]	d_loss: 0.4420	g_loss: 2.9981
Epoch [23/	50]	d_loss: 0.4944	g_loss: 3.9316
Epoch [23/	50]	d_loss: 0.9311	g_loss: 4.4918
Epoch [23/	50]	d_loss: 0.4316	g_loss: 2.7377
Epoch [23/	50]	d_loss: 0.5181	g_loss: 3.4176
Epoch [23/	50]	d_loss: 0.5695	g_loss: 1.9960
Epoch [23/	50]	d_loss: 0.4766	g_loss: 2.7524
Epoch [23/	50]	d_loss: 0.4129	g_loss: 3.0337
Epoch [23/	50]	d_loss: 0.3884	g_loss: 3.4330
Epoch [23/	50]	d_loss: 0.7780	g_loss: 2.1280
Epoch [23/	50]	d_loss: 0.4831	g_loss: 3.1928
Epoch [23/	50]	d_loss: 0.5082	g_loss: 2.2829
Epoch [23/	50]	d_loss: 0.3858	g_loss: 3.4478
Epoch [23/	50]	d_loss: 0.4238	g_loss: 3.5053
Epoch [23/	50]	d_loss: 0.5143	g_loss: 2.4303
Epoch [23/	50]	d_loss: 0.3840	g_loss: 3.5297
Epoch [23/	50]	d_loss: 0.3815	g_loss: 3.1705
Epoch [23/	50]	d_loss: 0.4297	g_loss: 4.0401
Epoch [23/	50]	d_loss: 0.4368	g_loss: 3.1435
Epoch [23/	50]	d_loss: 0.4602	g_loss: 2.7310
Epoch [23/	50]	d_loss: 0.3911	g_loss: 4.1349
Epoch [23/	50]	d_loss: 0.4628	g_loss: 2.5262

Epoch [23/	50]	d_loss: 0.4163	g_loss: 4.1268
Epoch [23/	50]	d_loss: 0.4064	g_loss: 3.0527
Epoch [23/	50]	d_loss: 0.3987	g_loss: 3.1821
Epoch [23/	50]	d_loss: 0.9821	g_loss: 5.1579
Epoch [23/	50]	d_loss: 0.4028	g_loss: 2.6268
Epoch [23/	50]	d_loss: 0.6876	g_loss: 3.2865
Epoch [23/	50]	d_loss: 0.4733	g_loss: 3.8228
Epoch [23/	50]	d_loss: 0.4305	g_loss: 3.7699
Epoch [23/	50]	d_loss: 0.4571	g_loss: 2.8251
Epoch [24/	50]	d_loss: 0.4047	g_loss: 2.8594
Epoch [24/	50]	d_loss: 0.5081	g_loss: 3.3544
Epoch [24/	50]	d_loss: 0.4129	g_loss: 3.3218
Epoch [24/	50]	d_loss: 0.4286	g_loss: 2.9155
Epoch [24/	50]	d_loss: 0.3916	g_loss: 3.5341
Epoch [24/	50]	d_loss: 1.2473	g_loss: 4.5101
Epoch [24/	50]	d_loss: 0.4397	g_loss: 3.1014
Epoch [24/	50]	d_loss: 0.4057	g_loss: 2.5349
Epoch [24/	50]	d_loss: 0.5300	g_loss: 3.0150
Epoch [24/	50]	d_loss: 0.4688	g_loss: 2.4466
Epoch [24/	50]	d_loss: 0.5976	g_loss: 3.4001
Epoch [24/	50]	d_loss: 0.4409	g_loss: 3.3508
Epoch [24/	50]	d_loss: 0.4185	g_loss: 2.3902
Epoch [24/	50]	d_loss: 0.5028	g_loss: 3.7498
Epoch [24/	50]	d_loss: 0.3927	g_loss: 2.2495
Epoch [24/	50]	d_loss: 0.4410	g_loss: 2.6728
Epoch [24/	50]	d_loss: 0.4271	g_loss: 2.2928
Epoch [24/	50]	d_loss: 0.4201	g_loss: 3.7758
Epoch [24/	50]	d_loss: 0.4696	g_loss: 2.5798
Epoch [24/	50]	d_loss: 0.3934	g_loss: 2.8542
Epoch [24/	50]	d_loss: 0.3939	g_loss: 3.4748
Epoch [24/	50]	d_loss: 0.3729	g_loss: 3.9420
Epoch [24/	50]	d_loss: 0.4638	g_loss: 3.4074
Epoch [24/	50]	d_loss: 0.4601	g_loss: 2.9890
Epoch [24/	50]	d_loss: 0.4912	g_loss: 4.2130
Epoch [24/	50]	d_loss: 0.6658	g_loss: 2.4311
Epoch [24/	50]	d_loss: 0.6697	g_loss: 3.4939
Epoch [24/	50]	d_loss: 0.4693	g_loss: 2.0339
Epoch [24/	50]	d_loss: 0.3980	g_loss: 3.4272
Epoch [24/	50]	d_loss: 0.3757	g_loss: 4.5767
Epoch [24/	50]	d_loss: 0.4511	g_loss: 2.8659
Epoch [24/	50]	d_loss: 0.4573	g_loss: 3.2154
Epoch [24/	50]	d_loss: 0.4117	g_loss: 2.5442
Epoch [24/	50]	d_loss: 0.5276	g_loss: 3.4029
Epoch [24/	50]	d_loss: 0.4522	g_loss: 2.6275
Epoch [24/	50]	d_loss: 0.4989	g_loss: 3.0494
Epoch [24/	50]	d_loss: 0.4106	g_loss: 2.7949
Epoch [24/	50]	d_loss: 0.4291	g_loss: 4.5514
Epoch [24/	50]	d_loss: 0.3678	g_loss: 3.7423

Epoch [24/	50]	d_loss: 0.4762	g_loss: 3.5690
Epoch [24/	50]	d_loss: 0.4373	g_loss: 2.3682
Epoch [24/	50]	d_loss: 0.4540	g_loss: 2.8593
Epoch [24/	50]	d_loss: 0.4446	g_loss: 3.5463
Epoch [24/	50]	d_loss: 0.4147	g_loss: 2.6126
Epoch [24/	50]	d_loss: 0.3948	g_loss: 3.3802
Epoch [24/	50]	d_loss: 0.4257	g_loss: 3.7315
Epoch [24/	50]	d_loss: 0.3773	g_loss: 4.3033
Epoch [24/	50]	d_loss: 0.4216	g_loss: 4.4367
Epoch [24/	50]	d_loss: 0.4412	g_loss: 2.7673
Epoch [24/	50]	d_loss: 0.4001	g_loss: 3.5222
Epoch [24/	50]	d_loss: 0.3759	g_loss: 3.9344
Epoch [24/	50]	d_loss: 0.4325	g_loss: 3.4586
Epoch [24/	50]	d_loss: 0.4907	g_loss: 2.6566
Epoch [24/	50]	d_loss: 0.4818	g_loss: 4.0843
Epoch [24/	50]	d_loss: 0.4040	g_loss: 2.4783
Epoch [24/	50]	d_loss: 0.4010	g_loss: 3.1702
Epoch [24/	50]	d_loss: 0.4635	g_loss: 3.6211
Epoch [25/	50]	d_loss: 0.3800	g_loss: 3.7001
Epoch [25/	50]	d_loss: 0.8344	g_loss: 1.9292
Epoch [25/	50]	d_loss: 0.5103	g_loss: 3.9401
Epoch [25/	50]	d_loss: 0.3973	g_loss: 1.9876
Epoch [25/	50]	d_loss: 0.4264	g_loss: 2.8882
Epoch [25/	50]	d_loss: 0.5424	g_loss: 3.0453
Epoch [25/	50]	d_loss: 0.4611	g_loss: 2.5398
Epoch [25/	50]	d_loss: 1.0799	g_loss: 3.7050
Epoch [25/	50]	d_loss: 0.4420	g_loss: 4.3409
Epoch [25/	50]	d_loss: 0.4530	g_loss: 2.7161
Epoch [25/	50]	d_loss: 0.4488	g_loss: 3.3214
Epoch [25/	50]	d_loss: 0.5184	g_loss: 2.4294
Epoch [25/	50]	d_loss: 0.4203	g_loss: 2.4274
Epoch [25/	50]	d_loss: 0.3959	g_loss: 3.9863
Epoch [25/	50]	d_loss: 0.4240	g_loss: 2.6872
Epoch [25/	50]	d_loss: 0.4875	g_loss: 2.8003
Epoch [25/	50]	d_loss: 0.4545	g_loss: 2.4280
Epoch [25/	50]	d_loss: 0.4847	g_loss: 3.2150
Epoch [25/	50]	d_loss: 0.3877	g_loss: 3.3306
Epoch [25/	50]	d_loss: 0.4979	g_loss: 3.6899
Epoch [25/	50]	d_loss: 0.4858	g_loss: 3.8945
Epoch [25/	50]	d_loss: 0.8252	g_loss: 4.8907
Epoch [25/	50]	d_loss: 0.4001	g_loss: 4.9720
Epoch [25/	50]	d_loss: 0.4091	g_loss: 3.4951
Epoch [25/	50]	d_loss: 0.4283	g_loss: 3.2935
Epoch [25/	50]	d_loss: 0.3941	g_loss: 4.2880
Epoch [25/	50]	d_loss: 0.4233	g_loss: 3.4337
Epoch [25/	50]	d_loss: 0.3705	g_loss: 3.7227
Epoch [25/	50]	d_loss: 0.5335	g_loss: 3.3744
Epoch [25/	50]	d_loss: 0.5127	g_loss: 2.7684

Epoch [25/	50]	d_loss: 0.4352	g_loss: 3.4800
Epoch [25/	50]	d_loss: 0.3908	g_loss: 2.4494
Epoch [25/	50]	d_loss: 0.4876	g_loss: 2.1502
Epoch [25/	50]	d_loss: 0.5190	g_loss: 2.3374
Epoch [25/	50]	d_loss: 0.5896	g_loss: 3.1181
Epoch [25/	50]	d_loss: 0.4964	g_loss: 4.4938
Epoch [25/	50]	d_loss: 0.4450	g_loss: 2.5561
Epoch [25/	50]	d_loss: 0.8321	g_loss: 1.6922
Epoch [25/	50]	d_loss: 0.4605	g_loss: 3.1680
Epoch [25/	50]	d_loss: 0.4955	g_loss: 3.0459
Epoch [25/	50]	d_loss: 0.5241	g_loss: 3.2314
Epoch [25/	50]	d_loss: 0.6054	g_loss: 3.0592
Epoch [25/	50]	d_loss: 0.5907	g_loss: 3.3756
Epoch [25/	50]	d_loss: 0.5323	g_loss: 3.3788
Epoch [25/	50]	d_loss: 0.4931	g_loss: 3.6107
Epoch [25/	50]	d_loss: 0.4396	g_loss: 2.3135
Epoch [25/	50]	d_loss: 0.4087	g_loss: 3.0124
Epoch [25/	50]	d_loss: 0.5018	g_loss: 2.1803
Epoch [25/	50]	d_loss: 0.4064	g_loss: 3.4923
Epoch [25/	50]	d_loss: 0.4236	g_loss: 2.6348
Epoch [25/	50]	d_loss: 0.4479	g_loss: 3.5009
Epoch [25/	50]	d_loss: 0.5836	g_loss: 2.9757
Epoch [25/	50]	d_loss: 0.4241	g_loss: 2.4821
Epoch [25/	50]	d_loss: 0.4982	g_loss: 2.8353
Epoch [25/	50]	d_loss: 0.4456	g_loss: 2.0073
Epoch [25/	50]	d_loss: 0.5540	g_loss: 4.3538
Epoch [25/	50]	d_loss: 0.4013	g_loss: 2.6327
Epoch [26/	50]	d_loss: 0.6260	g_loss: 3.0710
Epoch [26/	50]	d_loss: 0.5238	g_loss: 3.6303
Epoch [26/	50]	d_loss: 0.4517	g_loss: 3.7714
Epoch [26/	50]	d_loss: 0.4109	g_loss: 3.6082
Epoch [26/	50]	d_loss: 0.4602	g_loss: 3.9566
Epoch [26/	50]	d_loss: 0.5145	g_loss: 2.8455
Epoch [26/	50]	d_loss: 0.3981	g_loss: 3.7005
Epoch [26/	50]	d_loss: 0.4636	g_loss: 2.9933
Epoch [26/	50]	d_loss: 0.4258	g_loss: 2.7060
Epoch [26/	50]	d_loss: 0.4480	g_loss: 2.7593
Epoch [26/	50]	d_loss: 0.4608	g_loss: 3.8161
Epoch [26/	50]	d_loss: 0.4812	g_loss: 3.8874
Epoch [26/	50]	d_loss: 0.4346	g_loss: 2.1181
Epoch [26/	50]	d_loss: 0.4214	g_loss: 2.9527
Epoch [26/	50]	d_loss: 0.5464	g_loss: 4.6195
Epoch [26/	50]	d_loss: 0.3986	g_loss: 3.9917
Epoch [26/	50]	d_loss: 0.4053	g_loss: 2.5972
Epoch [26/	50]	d_loss: 0.7278	g_loss: 4.9504
Epoch [26/	50]	d_loss: 0.4422	g_loss: 3.0457
Epoch [26/	50]	d_loss: 0.4444	g_loss: 3.3563
Epoch [26/	50]	d_loss: 0.7071	g_loss: 3.8811

Epoch [26/	50]	d_loss: 0.4264	g_loss: 2.5794
Epoch [26/	50]	d_loss: 0.5052	g_loss: 2.8207
Epoch [26/	50]	d_loss: 0.4493	g_loss: 3.3161
Epoch [26/	50]	d_loss: 0.5401	g_loss: 2.2305
Epoch [26/	50]	d_loss: 0.4519	g_loss: 3.3117
Epoch [26/	50]	d_loss: 0.3945	g_loss: 3.5806
Epoch [26/	50]	d_loss: 0.3882	g_loss: 4.0406
Epoch [26/	50]	d_loss: 0.4485	g_loss: 3.5273
Epoch [26/	50]	d_loss: 0.4328	g_loss: 4.6832
Epoch [26/	50]	d_loss: 1.0456	g_loss: 6.0910
Epoch [26/	50]	d_loss: 0.3741	g_loss: 3.7158
Epoch [26/	50]	d_loss: 0.3807	g_loss: 4.8513
Epoch [26/	50]	d_loss: 0.3748	g_loss: 3.6610
Epoch [26/	50]	d_loss: 0.3918	g_loss: 3.6748
Epoch [26/	50]	d_loss: 0.4881	g_loss: 3.0543
Epoch [26/	50]	d_loss: 0.4119	g_loss: 3.3164
Epoch [26/	50]	d_loss: 0.4848	g_loss: 2.7446
Epoch [26/	50]	d_loss: 0.4004	g_loss: 4.2554
Epoch [26/	50]	d_loss: 0.4792	g_loss: 3.9554
Epoch [26/	50]	d_loss: 0.4150	g_loss: 2.6549
Epoch [26/	50]	d_loss: 0.4104	g_loss: 3.7442
Epoch [26/	50]	d_loss: 0.5322	g_loss: 2.0821
Epoch [26/	50]	d_loss: 0.3811	g_loss: 3.9289
Epoch [26/	50]	d_loss: 0.6422	g_loss: 3.3948
Epoch [26/	50]	d_loss: 0.5188	g_loss: 3.4170
Epoch [26/	50]	d_loss: 0.4451	g_loss: 3.2149
Epoch [26/	50]	d_loss: 0.4776	g_loss: 2.0402
Epoch [26/	50]	d_loss: 0.6948	g_loss: 3.7867
Epoch [26/	50]	d_loss: 0.4475	g_loss: 4.2652
Epoch [26/	50]	d_loss: 0.3818	g_loss: 3.1894
Epoch [26/	50]	d_loss: 0.4099	g_loss: 3.7521
Epoch [26/	50]	d_loss: 0.4340	g_loss: 3.2798
Epoch [26/	50]	d_loss: 0.5066	g_loss: 2.9662
Epoch [26/	50]	d_loss: 0.4880	g_loss: 3.4176
Epoch [26/	50]	d_loss: 0.5655	g_loss: 1.8532
Epoch [26/	50]	d_loss: 0.3753	g_loss: 3.6246
Epoch [27/	50]	d_loss: 0.4295	g_loss: 3.6076
Epoch [27/	50]	d_loss: 0.5860	g_loss: 3.6830
Epoch [27/	50]	d_loss: 0.4822	g_loss: 2.3469
Epoch [27/	50]	d_loss: 0.4209	g_loss: 2.3174
Epoch [27/	50]	d_loss: 0.3840	g_loss: 3.4941
Epoch [27/	50]	d_loss: 0.5602	g_loss: 1.5161
Epoch [27/	50]	d_loss: 0.5298	g_loss: 5.0601
Epoch [27/	50]	d_loss: 0.4348	g_loss: 3.5959
Epoch [27/	50]	d_loss: 0.8320	g_loss: 4.8002
Epoch [27/	50]	d_loss: 0.5103	g_loss: 2.8375
Epoch [27/	50]	d_loss: 0.4096	g_loss: 4.6422
Epoch [27/	50]	d_loss: 0.3982	g_loss: 2.5864

Epoch [27/	50]	d_loss: 0.4338	g_loss: 3.0108
Epoch [27/	50]	d_loss: 0.3959	g_loss: 2.6254
Epoch [27/	50]	d_loss: 0.4512	g_loss: 3.7071
Epoch [27/	50]	d_loss: 0.4285	g_loss: 2.8094
Epoch [27/	50]	d_loss: 0.3803	g_loss: 3.4722
Epoch [27/	50]	d_loss: 0.5344	g_loss: 4.0713
Epoch [27/	50]	d_loss: 0.6232	g_loss: 3.5905
Epoch [27/	50]	d_loss: 0.4299	g_loss: 3.7777
Epoch [27/	50]	d_loss: 0.4205	g_loss: 4.1827
Epoch [27/	50]	d_loss: 0.5388	g_loss: 2.9476
Epoch [27/	50]	d_loss: 0.8396	g_loss: 0.4039
Epoch [27/	50]	d_loss: 0.4740	g_loss: 3.1274
Epoch [27/	50]	d_loss: 0.4394	g_loss: 3.4276
Epoch [27/	50]	d_loss: 0.3626	g_loss: 4.1580
Epoch [27/	50]	d_loss: 0.4162	g_loss: 4.1876
Epoch [27/	50]	d_loss: 0.5167	g_loss: 3.2676
Epoch [27/	50]	d_loss: 0.4210	g_loss: 2.7205
Epoch [27/	50]	d_loss: 0.4889	g_loss: 3.3214
Epoch [27/	50]	d_loss: 0.4165	g_loss: 2.3908
Epoch [27/	50]	d_loss: 0.3779	g_loss: 3.7241
Epoch [27/	50]	d_loss: 0.7070	g_loss: 4.2326
Epoch [27/	50]	d_loss: 0.5073	g_loss: 3.2872
Epoch [27/	50]	d_loss: 0.4699	g_loss: 3.8537
Epoch [27/	50]	d_loss: 0.3663	g_loss: 3.7528
Epoch [27/	50]	d_loss: 0.4082	g_loss: 3.2060
Epoch [27/	50]	d_loss: 0.4587	g_loss: 3.7663
Epoch [27/	50]	d_loss: 0.4326	g_loss: 4.8531
Epoch [27/	50]	d_loss: 0.4950	g_loss: 2.4517
Epoch [27/	50]	d_loss: 0.4946	g_loss: 2.4948
Epoch [27/	50]	d_loss: 0.3946	g_loss: 2.2525
Epoch [27/	50]	d_loss: 0.4133	g_loss: 2.8805
Epoch [27/	50]	d_loss: 0.6182	g_loss: 2.1832
Epoch [27/	50]	d_loss: 0.3879	g_loss: 4.2517
Epoch [27/	50]	d_loss: 0.4832	g_loss: 2.8732
Epoch [27/	50]	d_loss: 0.4755	g_loss: 2.5193
Epoch [27/	50]	d_loss: 0.3916	g_loss: 2.5514
Epoch [27/	50]	d_loss: 0.9503	g_loss: 4.7208
Epoch [27/	50]	d_loss: 0.5116	g_loss: 3.0071
Epoch [27/	50]	d_loss: 0.4147	g_loss: 3.8051
Epoch [27/	50]	d_loss: 0.4190	g_loss: 3.6525
Epoch [27/	50]	d_loss: 0.4131	g_loss: 4.0321
Epoch [27/	50]	d_loss: 0.3980	g_loss: 3.8618
Epoch [27/	50]	d_loss: 0.3929	g_loss: 3.2969
Epoch [27/	50]	d_loss: 0.5236	g_loss: 3.5766
Epoch [27/	50]	d_loss: 0.4004	g_loss: 4.0925
Epoch [28/	50]	d_loss: 0.3901	g_loss: 3.8073
Epoch [28/	50]	d_loss: 0.7075	g_loss: 1.4983
Epoch [28/	50]	d_loss: 0.4924	g_loss: 3.4128

Epoch [28/	50]	d_loss: 0.3814	g_loss: 3.9574
Epoch [28/	50]	d_loss: 0.3662	g_loss: 4.3140
Epoch [28/	50]	d_loss: 0.4576	g_loss: 3.7369
Epoch [28/	50]	d_loss: 0.4072	g_loss: 3.3721
Epoch [28/	50]	d_loss: 0.5642	g_loss: 3.2706
Epoch [28/	50]	d_loss: 0.5487	g_loss: 2.5991
Epoch [28/	50]	d_loss: 0.4233	g_loss: 3.3759
Epoch [28/	50]	d_loss: 0.4054	g_loss: 4.5697
Epoch [28/	50]	d_loss: 0.3860	g_loss: 3.2362
Epoch [28/	50]	d_loss: 0.3764	g_loss: 4.2794
Epoch [28/	50]	d_loss: 0.4685	g_loss: 3.3629
Epoch [28/	50]	d_loss: 0.4989	g_loss: 4.6278
Epoch [28/	50]	d_loss: 0.4325	g_loss: 2.7361
Epoch [28/	50]	d_loss: 0.3934	g_loss: 3.7793
Epoch [28/	50]	d_loss: 0.4042	g_loss: 4.1844
Epoch [28/	50]	d_loss: 0.6625	g_loss: 1.6046
Epoch [28/	50]	d_loss: 0.3940	g_loss: 2.7844
Epoch [28/	50]	d_loss: 0.7609	g_loss: 3.5853
Epoch [28/	50]	d_loss: 0.4189	g_loss: 3.0916
Epoch [28/	50]	d_loss: 0.4246	g_loss: 3.5525
Epoch [28/	50]	d_loss: 0.3764	g_loss: 3.0348
Epoch [28/	50]	d_loss: 0.4163	g_loss: 4.0987
Epoch [28/	50]	d_loss: 0.3880	g_loss: 2.7094
Epoch [28/	50]	d_loss: 0.4169	g_loss: 3.7196
Epoch [28/	50]	d_loss: 0.4453	g_loss: 2.8014
Epoch [28/	50]	d_loss: 0.4206	g_loss: 4.2985
Epoch [28/	50]	d_loss: 0.4323	g_loss: 2.2353
Epoch [28/	50]	d_loss: 0.3853	g_loss: 5.0874
Epoch [28/	50]	d_loss: 0.4834	g_loss: 2.9130
Epoch [28/	50]	d_loss: 0.4175	g_loss: 3.8218
Epoch [28/	50]	d_loss: 0.6508	g_loss: 1.4830
Epoch [28/	50]	d_loss: 0.5070	g_loss: 1.5568
Epoch [28/	50]	d_loss: 0.3971	g_loss: 2.4357
Epoch [28/	50]	d_loss: 0.6645	g_loss: 2.3920
Epoch [28/	50]	d_loss: 0.4300	g_loss: 3.2927
Epoch [28/	50]	d_loss: 0.3785	g_loss: 4.2086
Epoch [28/	50]	d_loss: 0.4826	g_loss: 2.4456
Epoch [28/	50]	d_loss: 0.4019	g_loss: 5.1995
Epoch [28/	50]	d_loss: 0.4261	g_loss: 4.7138
Epoch [28/	50]	d_loss: 0.4123	g_loss: 4.2711
Epoch [28/	50]	d_loss: 0.4165	g_loss: 5.1845
Epoch [28/	50]	d_loss: 0.3664	g_loss: 3.8675
Epoch [28/	50]	d_loss: 0.4646	g_loss: 3.9072
Epoch [28/	50]	d_loss: 0.4677	g_loss: 3.0862
Epoch [28/	50]	d_loss: 0.4408	g_loss: 3.0660
Epoch [28/	50]	d_loss: 0.5175	g_loss: 2.3312
Epoch [28/	50]	d_loss: 0.4665	g_loss: 2.9869
Epoch [28/	50]	d_loss: 0.5527	g_loss: 2.3529

Epoch [28/	50]	d_loss: 0.3966	g_loss: 3.6250
Epoch [28/	50]	d_loss: 0.4094	g_loss: 3.6971
Epoch [28/	50]	d_loss: 0.4584	g_loss: 4.2218
Epoch [28/	50]	d_loss: 0.5853	g_loss: 4.0139
Epoch [28/	50]	d_loss: 0.4340	g_loss: 3.2580
Epoch [28/	50]	d_loss: 0.5417	g_loss: 3.6269
Epoch [29/	50]	d_loss: 0.3995	g_loss: 2.9817
Epoch [29/	50]	d_loss: 0.4171	g_loss: 4.0199
Epoch [29/	50]	d_loss: 0.8042	g_loss: 3.1350
Epoch [29/	50]	d_loss: 0.4229	g_loss: 4.2037
Epoch [29/	50]	d_loss: 0.4319	g_loss: 3.3946
Epoch [29/	50]	d_loss: 0.3732	g_loss: 4.1714
Epoch [29/	50]	d_loss: 0.4022	g_loss: 3.2143
Epoch [29/	50]	d_loss: 0.5763	g_loss: 2.4545
Epoch [29/	50]	d_loss: 0.4245	g_loss: 3.7924
Epoch [29/	50]	d_loss: 0.4300	g_loss: 5.0679
Epoch [29/	50]	d_loss: 0.4405	g_loss: 3.0887
Epoch [29/	50]	d_loss: 0.3736	g_loss: 4.0735
Epoch [29/	50]	d_loss: 0.3733	g_loss: 4.6615
Epoch [29/	50]	d_loss: 1.5372	g_loss: 6.0065
Epoch [29/	50]	d_loss: 0.6409	g_loss: 2.3998
Epoch [29/	50]	d_loss: 0.4219	g_loss: 3.6690
Epoch [29/	50]	d_loss: 0.5379	g_loss: 2.6685
Epoch [29/	50]	d_loss: 0.5715	g_loss: 2.1390
Epoch [29/	50]	d_loss: 0.6481	g_loss: 1.5186
Epoch [29/	50]	d_loss: 0.4497	g_loss: 4.8000
Epoch [29/	50]	d_loss: 0.3739	g_loss: 3.8742
Epoch [29/	50]	d_loss: 0.4112	g_loss: 3.6795
Epoch [29/	50]	d_loss: 0.5186	g_loss: 2.8299
Epoch [29/	50]	d_loss: 0.5593	g_loss: 2.0207
Epoch [29/	50]	d_loss: 0.3819	g_loss: 3.9414
Epoch [29/	50]	d_loss: 0.3782	g_loss: 3.8511
Epoch [29/	50]	d_loss: 0.3938	g_loss: 3.0957
Epoch [29/	50]	d_loss: 0.4441	g_loss: 3.1295
Epoch [29/	50]	d_loss: 0.4142	g_loss: 2.6229
Epoch [29/	50]	d_loss: 0.4175	g_loss: 3.5910
Epoch [29/	50]	d_loss: 0.3825	g_loss: 3.2324
Epoch [29/	50]	d_loss: 0.5018	g_loss: 2.9448
Epoch [29/	50]	d_loss: 0.4506	g_loss: 4.2501
Epoch [29/	50]	d_loss: 0.4814	g_loss: 3.9492
Epoch [29/	50]	d_loss: 0.4364	g_loss: 3.3076
Epoch [29/	50]	d_loss: 0.4059	g_loss: 2.6614
Epoch [29/	50]	d_loss: 0.4186	g_loss: 2.7831
Epoch [29/	50]	d_loss: 0.4210	g_loss: 2.6836
Epoch [29/	50]	d_loss: 0.8072	g_loss: 0.9849
Epoch [29/	50]	d_loss: 0.6613	g_loss: 0.8063
Epoch [29/	50]	d_loss: 0.4951	g_loss: 4.7605
Epoch [29/	50]	d_loss: 0.4920	g_loss: 4.1684

Epoch [29/	50]	d_loss: 0.3829	g_loss: 4.3731
Epoch [29/	50]	d_loss: 0.7491	g_loss: 2.7498
Epoch [29/	50]	d_loss: 0.3773	g_loss: 4.0501
Epoch [29/	50]	d_loss: 0.4496	g_loss: 2.8889
Epoch [29/	50]	d_loss: 0.4513	g_loss: 3.4352
Epoch [29/	50]	d_loss: 0.5445	g_loss: 3.3580
Epoch [29/	50]	d_loss: 0.5807	g_loss: 3.9758
Epoch [29/	50]	d_loss: 0.4245	g_loss: 3.8260
Epoch [29/	50]	d_loss: 0.4059	g_loss: 4.4447
Epoch [29/	50]	d_loss: 0.5282	g_loss: 2.2712
Epoch [29/	50]	d_loss: 0.4109	g_loss: 2.6999
Epoch [29/	50]	d_loss: 0.3944	g_loss: 4.0890
Epoch [29/	50]	d_loss: 0.4330	g_loss: 3.9649
Epoch [29/	50]	d_loss: 0.3905	g_loss: 2.9998
Epoch [29/	50]	d_loss: 0.5915	g_loss: 3.6824
Epoch [30/	50]	d_loss: 0.4182	g_loss: 3.1080
Epoch [30/	50]	d_loss: 0.4580	g_loss: 3.4988
Epoch [30/	50]	d_loss: 0.8589	g_loss: 4.3181
Epoch [30/	50]	d_loss: 0.3953	g_loss: 3.8246
Epoch [30/	50]	d_loss: 0.8302	g_loss: 1.7999
Epoch [30/	50]	d_loss: 0.4545	g_loss: 3.4509
Epoch [30/	50]	d_loss: 0.3811	g_loss: 4.2059
Epoch [30/	50]	d_loss: 0.4252	g_loss: 2.3139
Epoch [30/	50]	d_loss: 0.4033	g_loss: 2.9094
Epoch [30/	50]	d_loss: 0.7116	g_loss: 3.6906
Epoch [30/	50]	d_loss: 0.4098	g_loss: 3.0806
Epoch [30/	50]	d_loss: 0.7521	g_loss: 3.8370
Epoch [30/	50]	d_loss: 0.4980	g_loss: 2.1310
Epoch [30/	50]	d_loss: 0.4413	g_loss: 4.5381
Epoch [30/	50]	d_loss: 0.4359	g_loss: 3.2420
Epoch [30/	50]	d_loss: 0.4036	g_loss: 2.4643
Epoch [30/	50]	d_loss: 0.4398	g_loss: 3.3755
Epoch [30/	50]	d_loss: 0.4069	g_loss: 3.9281
Epoch [30/	50]	d_loss: 0.4391	g_loss: 3.1716
Epoch [30/	50]	d_loss: 0.8846	g_loss: 5.0667
Epoch [30/	50]	d_loss: 0.5143	g_loss: 4.1995
Epoch [30/	50]	d_loss: 1.3183	g_loss: 4.4721
Epoch [30/	50]	d_loss: 0.5551	g_loss: 2.5054
Epoch [30/	50]	d_loss: 0.7399	g_loss: 0.7820
Epoch [30/	50]	d_loss: 0.3992	g_loss: 3.5360
Epoch [30/	50]	d_loss: 0.4440	g_loss: 2.4155
Epoch [30/	50]	d_loss: 0.4819	g_loss: 3.3977
Epoch [30/	50]	d_loss: 0.4294	g_loss: 4.3442
Epoch [30/	50]	d_loss: 0.3830	g_loss: 4.7086
Epoch [30/	50]	d_loss: 0.5547	g_loss: 2.7876
Epoch [30/	50]	d_loss: 0.5846	g_loss: 1.6515
Epoch [30/	50]	d_loss: 0.4711	g_loss: 3.6333
Epoch [30/	50]	d_loss: 0.5566	g_loss: 1.2391

Epoch [30/	50]	d_loss: 0.5008	g_loss: 3.2541
Epoch [30/	50]	d_loss: 0.4045	g_loss: 3.8528
Epoch [30/	50]	d_loss: 0.4386	g_loss: 3.5703
Epoch [30/	50]	d_loss: 0.3849	g_loss: 2.7754
Epoch [30/	50]	d_loss: 0.5309	g_loss: 2.2517
Epoch [30/	50]	d_loss: 0.5024	g_loss: 3.5361
Epoch [30/	50]	d_loss: 0.4493	g_loss: 3.6382
Epoch [30/	50]	d_loss: 0.3875	g_loss: 3.8824
Epoch [30/	50]	d_loss: 0.4909	g_loss: 3.3405
Epoch [30/	50]	d_loss: 0.3797	g_loss: 2.7172
Epoch [30/	50]	d_loss: 0.3774	g_loss: 3.4804
Epoch [30/	50]	d_loss: 0.3831	g_loss: 3.7699
Epoch [30/	50]	d_loss: 0.4676	g_loss: 3.2398
Epoch [30/	50]	d_loss: 0.4251	g_loss: 3.4390
Epoch [30/	50]	d_loss: 0.3777	g_loss: 3.0722
Epoch [30/	50]	d_loss: 0.5676	g_loss: 4.9662
Epoch [30/	50]	d_loss: 0.3793	g_loss: 3.2850
Epoch [30/	50]	d_loss: 0.4438	g_loss: 2.8254
Epoch [30/	50]	d_loss: 0.4139	g_loss: 3.1467
Epoch [30/	50]	d_loss: 0.7862	g_loss: 3.7980
Epoch [30/	50]	d_loss: 0.3996	g_loss: 3.7592
Epoch [30/	50]	d_loss: 0.4134	g_loss: 3.1436
Epoch [30/	50]	d_loss: 0.4168	g_loss: 3.4015
Epoch [30/	50]	d_loss: 0.4368	g_loss: 4.3811
Epoch [31/	50]	d_loss: 0.3821	g_loss: 3.4478
Epoch [31/	50]	d_loss: 0.4627	g_loss: 2.5123
Epoch [31/	50]	d_loss: 0.4056	g_loss: 3.4939
Epoch [31/	50]	d_loss: 0.3930	g_loss: 3.2470
Epoch [31/	50]	d_loss: 0.3989	g_loss: 4.0339
Epoch [31/	50]	d_loss: 0.4382	g_loss: 3.5868
Epoch [31/	50]	d_loss: 0.3688	g_loss: 3.9944
Epoch [31/	50]	d_loss: 0.4049	g_loss: 4.5628
Epoch [31/	50]	d_loss: 0.4480	g_loss: 2.6878
Epoch [31/	50]	d_loss: 0.3948	g_loss: 3.8986
Epoch [31/	50]	d_loss: 0.4572	g_loss: 3.0747
Epoch [31/	50]	d_loss: 0.6094	g_loss: 2.0685
Epoch [31/	50]	d_loss: 0.4630	g_loss: 3.2215
Epoch [31/	50]	d_loss: 0.3588	g_loss: 4.5987
Epoch [31/	50]	d_loss: 0.4605	g_loss: 2.0631
Epoch [31/	50]	d_loss: 0.6497	g_loss: 3.7392
Epoch [31/	50]	d_loss: 0.4295	g_loss: 2.8420
Epoch [31/	50]	d_loss: 0.4156	g_loss: 4.2472
Epoch [31/	50]	d_loss: 0.3567	g_loss: 4.4084
Epoch [31/	50]	d_loss: 0.3818	g_loss: 3.0760
Epoch [31/	50]	d_loss: 0.3985	g_loss: 4.1962
Epoch [31/	50]	d_loss: 0.4989	g_loss: 3.9208
Epoch [31/	50]	d_loss: 0.4221	g_loss: 3.5234
Epoch [31/	50]	d_loss: 0.4482	g_loss: 3.2094

Epoch [31/	50]	d_loss: 0.4895	g_loss: 2.2099
Epoch [31/	50]	d_loss: 0.5011	g_loss: 2.1871
Epoch [31/	50]	d_loss: 0.4392	g_loss: 2.4066
Epoch [31/	50]	d_loss: 0.5110	g_loss: 2.1706
Epoch [31/	50]	d_loss: 0.4035	g_loss: 2.8655
Epoch [31/	50]	d_loss: 0.3884	g_loss: 2.3646
Epoch [31/	50]	d_loss: 0.4203	g_loss: 4.1130
Epoch [31/	50]	d_loss: 0.4170	g_loss: 2.5840
Epoch [31/	50]	d_loss: 0.4724	g_loss: 3.5202
Epoch [31/	50]	d_loss: 0.4263	g_loss: 2.5201
Epoch [31/	50]	d_loss: 0.4367	g_loss: 2.5149
Epoch [31/	50]	d_loss: 0.3688	g_loss: 4.2787
Epoch [31/	50]	d_loss: 0.4730	g_loss: 2.7290
Epoch [31/	50]	d_loss: 0.8080	g_loss: 1.1452
Epoch [31/	50]	d_loss: 0.6915	g_loss: 5.0546
Epoch [31/	50]	d_loss: 0.3674	g_loss: 3.3933
Epoch [31/	50]	d_loss: 0.3805	g_loss: 3.8204
Epoch [31/	50]	d_loss: 0.4346	g_loss: 2.2888
Epoch [31/	50]	d_loss: 0.4186	g_loss: 3.8001
Epoch [31/	50]	d_loss: 0.3703	g_loss: 4.7047
Epoch [31/	50]	d_loss: 0.3960	g_loss: 3.3527
Epoch [31/	50]	d_loss: 0.4336	g_loss: 3.1319
Epoch [31/	50]	d_loss: 0.3983	g_loss: 2.4538
Epoch [31/	50]	d_loss: 0.4729	g_loss: 3.4449
Epoch [31/	50]	d_loss: 0.3882	g_loss: 3.0809
Epoch [31/	50]	d_loss: 0.3907	g_loss: 2.9428
Epoch [31/	50]	d_loss: 0.3937	g_loss: 3.1484
Epoch [31/	50]	d_loss: 0.4025	g_loss: 4.6281
Epoch [31/	50]	d_loss: 0.4215	g_loss: 2.6840
Epoch [31/	50]	d_loss: 0.3861	g_loss: 4.0307
Epoch [31/	50]	d_loss: 0.3877	g_loss: 4.4641
Epoch [31/	50]	d_loss: 0.4266	g_loss: 3.3351
Epoch [31/	50]	d_loss: 0.4311	g_loss: 3.6990
Epoch [32/	50]	d_loss: 0.4401	g_loss: 3.6023
Epoch [32/	50]	d_loss: 0.5134	g_loss: 3.2098
Epoch [32/	50]	d_loss: 0.4274	g_loss: 4.3807
Epoch [32/	50]	d_loss: 0.6495	g_loss: 3.0813
Epoch [32/	50]	d_loss: 0.4817	g_loss: 3.5825
Epoch [32/	50]	d_loss: 0.4349	g_loss: 2.0702
Epoch [32/	50]	d_loss: 0.5108	g_loss: 2.2682
Epoch [32/	50]	d_loss: 0.3750	g_loss: 3.4215
Epoch [32/	50]	d_loss: 0.5018	g_loss: 3.9039
Epoch [32/	50]	d_loss: 1.0379	g_loss: 4.8928
Epoch [32/	50]	d_loss: 0.4335	g_loss: 3.4654
Epoch [32/	50]	d_loss: 0.4319	g_loss: 4.0790
Epoch [32/	50]	d_loss: 0.4393	g_loss: 3.1707
Epoch [32/	50]	d_loss: 0.4760	g_loss: 3.0106
Epoch [32/	50]	d_loss: 0.4110	g_loss: 3.5093

Epoch [32/	50]	d_loss: 0.3891	g_loss: 4.1311
Epoch [32/	50]	d_loss: 0.5158	g_loss: 2.1262
Epoch [32/	50]	d_loss: 0.4233	g_loss: 3.5683
Epoch [32/	50]	d_loss: 0.4780	g_loss: 3.1372
Epoch [32/	50]	d_loss: 0.4227	g_loss: 2.9788
Epoch [32/	50]	d_loss: 0.3789	g_loss: 3.0343
Epoch [32/	50]	d_loss: 0.4521	g_loss: 3.3018
Epoch [32/	50]	d_loss: 0.4576	g_loss: 3.1611
Epoch [32/	50]	d_loss: 0.3802	g_loss: 4.2999
Epoch [32/	50]	d_loss: 0.4012	g_loss: 3.5795
Epoch [32/	50]	d_loss: 0.3793	g_loss: 3.0740
Epoch [32/	50]	d_loss: 0.3946	g_loss: 3.0893
Epoch [32/	50]	d_loss: 0.4402	g_loss: 3.1249
Epoch [32/	50]	d_loss: 0.4063	g_loss: 2.9929
Epoch [32/	50]	d_loss: 0.3550	g_loss: 3.9886
Epoch [32/	50]	d_loss: 0.4216	g_loss: 3.1888
Epoch [32/	50]	d_loss: 0.5837	g_loss: 2.9612
Epoch [32/	50]	d_loss: 0.3669	g_loss: 3.9975
Epoch [32/	50]	d_loss: 0.5157	g_loss: 3.6941
Epoch [32/	50]	d_loss: 0.4468	g_loss: 3.4280
Epoch [32/	50]	d_loss: 0.4760	g_loss: 2.8153
Epoch [32/	50]	d_loss: 0.4586	g_loss: 3.0298
Epoch [32/	50]	d_loss: 0.3782	g_loss: 3.7236
Epoch [32/	50]	d_loss: 0.5175	g_loss: 4.1659
Epoch [32/	50]	d_loss: 0.4642	g_loss: 2.4782
Epoch [32/	50]	d_loss: 0.3785	g_loss: 2.7978
Epoch [32/	50]	d_loss: 0.3824	g_loss: 3.4439
Epoch [32/	50]	d_loss: 0.3970	g_loss: 4.3150
Epoch [32/	50]	d_loss: 0.5128	g_loss: 2.8107
Epoch [32/	50]	d_loss: 0.4274	g_loss: 2.9092
Epoch [32/	50]	d_loss: 0.3708	g_loss: 4.0654
Epoch [32/	50]	d_loss: 0.3953	g_loss: 4.0236
Epoch [32/	50]	d_loss: 0.5125	g_loss: 2.5237
Epoch [32/	50]	d_loss: 0.4136	g_loss: 3.6277
Epoch [32/	50]	d_loss: 0.4011	g_loss: 2.4808
Epoch [32/	50]	d_loss: 0.6414	g_loss: 1.9547
Epoch [32/	50]	d_loss: 0.5433	g_loss: 3.6835
Epoch [32/	50]	d_loss: 0.4122	g_loss: 2.6676
Epoch [32/	50]	d_loss: 0.4366	g_loss: 4.7131
Epoch [32/	50]	d_loss: 0.4379	g_loss: 2.4710
Epoch [32/	50]	d_loss: 0.5922	g_loss: 1.7042
Epoch [32/	50]	d_loss: 0.5784	g_loss: 2.0278
Epoch [33/	50]	d_loss: 0.3761	g_loss: 3.4930
Epoch [33/	50]	d_loss: 0.3972	g_loss: 3.5619
Epoch [33/	50]	d_loss: 0.3881	g_loss: 5.1893
Epoch [33/	50]	d_loss: 0.4242	g_loss: 3.2547
Epoch [33/	50]	d_loss: 0.3740	g_loss: 4.7661
Epoch [33/	50]	d_loss: 0.4250	g_loss: 3.1613

Epoch [33/	50]	d_loss: 0.4148	g_loss: 4.0488
Epoch [33/	50]	d_loss: 0.4178	g_loss: 3.2238
Epoch [33/	50]	d_loss: 0.5386	g_loss: 2.1431
Epoch [33/	50]	d_loss: 0.4041	g_loss: 3.6665
Epoch [33/	50]	d_loss: 0.5380	g_loss: 2.8300
Epoch [33/	50]	d_loss: 0.8850	g_loss: 3.9507
Epoch [33/	50]	d_loss: 0.3761	g_loss: 2.3792
Epoch [33/	50]	d_loss: 0.5226	g_loss: 4.1268
Epoch [33/	50]	d_loss: 0.4814	g_loss: 3.8758
Epoch [33/	50]	d_loss: 0.3970	g_loss: 3.6931
Epoch [33/	50]	d_loss: 1.1093	g_loss: 4.6934
Epoch [33/	50]	d_loss: 0.4948	g_loss: 1.8920
Epoch [33/	50]	d_loss: 0.3742	g_loss: 3.1887
Epoch [33/	50]	d_loss: 0.4703	g_loss: 3.7451
Epoch [33/	50]	d_loss: 0.4217	g_loss: 1.9534
Epoch [33/	50]	d_loss: 0.4262	g_loss: 2.5629
Epoch [33/	50]	d_loss: 0.3747	g_loss: 2.6627
Epoch [33/	50]	d_loss: 0.4294	g_loss: 3.0132
Epoch [33/	50]	d_loss: 0.5221	g_loss: 3.0461
Epoch [33/	50]	d_loss: 0.3749	g_loss: 4.3272
Epoch [33/	50]	d_loss: 0.4107	g_loss: 3.4913
Epoch [33/	50]	d_loss: 0.3996	g_loss: 4.7493
Epoch [33/	50]	d_loss: 0.4712	g_loss: 2.5274
Epoch [33/	50]	d_loss: 0.4633	g_loss: 3.7758
Epoch [33/	50]	d_loss: 0.3840	g_loss: 3.1330
Epoch [33/	50]	d_loss: 0.5963	g_loss: 2.4385
Epoch [33/	50]	d_loss: 0.5096	g_loss: 4.1670
Epoch [33/	50]	d_loss: 0.4035	g_loss: 2.5218
Epoch [33/	50]	d_loss: 0.4193	g_loss: 4.5776
Epoch [33/	50]	d_loss: 0.4230	g_loss: 3.1794
Epoch [33/	50]	d_loss: 0.4022	g_loss: 3.2614
Epoch [33/	50]	d_loss: 0.5445	g_loss: 2.7247
Epoch [33/	50]	d_loss: 0.4112	g_loss: 5.3190
Epoch [33/	50]	d_loss: 0.4052	g_loss: 3.5977
Epoch [33/	50]	d_loss: 0.4999	g_loss: 3.6113
Epoch [33/	50]	d_loss: 0.3749	g_loss: 4.3014
Epoch [33/	50]	d_loss: 0.4625	g_loss: 5.1338
Epoch [33/	50]	d_loss: 0.5726	g_loss: 3.6053
Epoch [33/	50]	d_loss: 0.4386	g_loss: 4.3665
Epoch [33/	50]	d_loss: 0.4816	g_loss: 2.9828
Epoch [33/	50]	d_loss: 0.4127	g_loss: 4.3108
Epoch [33/	50]	d_loss: 0.5389	g_loss: 3.0642
Epoch [33/	50]	d_loss: 0.4000	g_loss: 4.5232
Epoch [33/	50]	d_loss: 0.4035	g_loss: 2.3341
Epoch [33/	50]	d_loss: 0.5386	g_loss: 2.6933
Epoch [33/	50]	d_loss: 0.4546	g_loss: 2.8949
Epoch [33/	50]	d_loss: 0.3835	g_loss: 3.9925
Epoch [33/	50]	d_loss: 0.4169	g_loss: 2.8696

Epoch [33/	50]	d_loss: 0.6079	g_loss: 2.7938
Epoch [33/	50]	d_loss: 0.4197	g_loss: 3.2886
Epoch [33/	50]	d_loss: 0.3663	g_loss: 3.1581
Epoch [34/	50]	d_loss: 0.4301	g_loss: 2.2773
Epoch [34/	50]	d_loss: 0.3931	g_loss: 2.7807
Epoch [34/	50]	d_loss: 0.5747	g_loss: 2.9063
Epoch [34/	50]	d_loss: 0.6891	g_loss: 1.5347
Epoch [34/	50]	d_loss: 0.5164	g_loss: 2.8767
Epoch [34/	50]	d_loss: 0.4057	g_loss: 4.5466
Epoch [34/	50]	d_loss: 0.6627	g_loss: 5.7179
Epoch [34/	50]	d_loss: 0.5841	g_loss: 4.8681
Epoch [34/	50]	d_loss: 0.3745	g_loss: 4.1192
Epoch [34/	50]	d_loss: 0.3986	g_loss: 5.0828
Epoch [34/	50]	d_loss: 0.3855	g_loss: 4.3373
Epoch [34/	50]	d_loss: 0.3998	g_loss: 3.9878
Epoch [34/	50]	d_loss: 0.4156	g_loss: 3.0333
Epoch [34/	50]	d_loss: 0.8118	g_loss: 3.4896
Epoch [34/	50]	d_loss: 0.3900	g_loss: 3.5509
Epoch [34/	50]	d_loss: 0.4104	g_loss: 2.9015
Epoch [34/	50]	d_loss: 0.3902	g_loss: 4.5444
Epoch [34/	50]	d_loss: 0.3902	g_loss: 4.4781
Epoch [34/	50]	d_loss: 0.4837	g_loss: 4.3775
Epoch [34/	50]	d_loss: 0.3808	g_loss: 2.7359
Epoch [34/	50]	d_loss: 0.5082	g_loss: 2.4807
Epoch [34/	50]	d_loss: 0.9995	g_loss: 1.3121
Epoch [34/	50]	d_loss: 0.5547	g_loss: 4.4170
Epoch [34/	50]	d_loss: 0.5924	g_loss: 4.4724
Epoch [34/	50]	d_loss: 0.3926	g_loss: 2.7326
Epoch [34/	50]	d_loss: 0.4241	g_loss: 2.0148
Epoch [34/	50]	d_loss: 0.4307	g_loss: 2.9852
Epoch [34/	50]	d_loss: 0.4494	g_loss: 4.2927
Epoch [34/	50]	d_loss: 0.4056	g_loss: 3.8274
Epoch [34/	50]	d_loss: 0.4999	g_loss: 3.4757
Epoch [34/	50]	d_loss: 0.4130	g_loss: 3.4154
Epoch [34/	50]	d_loss: 0.3647	g_loss: 3.2719
Epoch [34/	50]	d_loss: 0.6991	g_loss: 3.4709
Epoch [34/	50]	d_loss: 0.4187	g_loss: 3.4127
Epoch [34/	50]	d_loss: 0.5103	g_loss: 2.8782
Epoch [34/	50]	d_loss: 0.4766	g_loss: 2.7477
Epoch [34/	50]	d_loss: 0.7431	g_loss: 1.9369
Epoch [34/	50]	d_loss: 0.3612	g_loss: 3.7649
Epoch [34/	50]	d_loss: 0.4011	g_loss: 3.8240
Epoch [34/	50]	d_loss: 0.4000	g_loss: 3.8191
Epoch [34/	50]	d_loss: 0.4315	g_loss: 2.9870
Epoch [34/	50]	d_loss: 0.3938	g_loss: 3.3197
Epoch [34/	50]	d_loss: 0.4149	g_loss: 2.7134
Epoch [34/	50]	d_loss: 0.3899	g_loss: 4.1685
Epoch [34/	50]	d_loss: 0.3851	g_loss: 2.9961

Epoch [34/	50]	d_loss: 0.3982	g_loss: 3.0540
Epoch [34/	50]	d_loss: 1.1187	g_loss: 4.0078
Epoch [34/	50]	d_loss: 0.3813	g_loss: 4.2126
Epoch [34/	50]	d_loss: 0.3772	g_loss: 4.1495
Epoch [34/	50]	d_loss: 0.4418	g_loss: 3.1371
Epoch [34/	50]	d_loss: 0.4722	g_loss: 3.1142
Epoch [34/	50]	d_loss: 0.4289	g_loss: 3.1489
Epoch [34/	50]	d_loss: 0.3978	g_loss: 4.5207
Epoch [34/	50]	d_loss: 0.4757	g_loss: 2.8211
Epoch [34/	50]	d_loss: 0.8305	g_loss: 2.0902
Epoch [34/	50]	d_loss: 0.5024	g_loss: 2.5890
Epoch [34/	50]	d_loss: 0.5603	g_loss: 2.5999
Epoch [35/	50]	d_loss: 0.3590	g_loss: 4.1274
Epoch [35/	50]	d_loss: 0.4465	g_loss: 3.4230
Epoch [35/	50]	d_loss: 1.3559	g_loss: 5.8638
Epoch [35/	50]	d_loss: 0.4962	g_loss: 3.3036
Epoch [35/	50]	d_loss: 0.4061	g_loss: 2.7609
Epoch [35/	50]	d_loss: 0.8934	g_loss: 3.4635
Epoch [35/	50]	d_loss: 0.4057	g_loss: 3.3020
Epoch [35/	50]	d_loss: 0.4361	g_loss: 3.2855
Epoch [35/	50]	d_loss: 0.3900	g_loss: 4.1224
Epoch [35/	50]	d_loss: 0.4087	g_loss: 4.0760
Epoch [35/	50]	d_loss: 0.3973	g_loss: 3.0456
Epoch [35/	50]	d_loss: 0.5923	g_loss: 2.3981
Epoch [35/	50]	d_loss: 0.4004	g_loss: 3.7613
Epoch [35/	50]	d_loss: 0.4055	g_loss: 3.5358
Epoch [35/	50]	d_loss: 0.4124	g_loss: 2.9478
Epoch [35/	50]	d_loss: 0.5779	g_loss: 4.3119
Epoch [35/	50]	d_loss: 0.4874	g_loss: 3.8940
Epoch [35/	50]	d_loss: 0.6614	g_loss: 2.1924
Epoch [35/	50]	d_loss: 0.5601	g_loss: 3.3845
Epoch [35/	50]	d_loss: 0.6398	g_loss: 3.1579
Epoch [35/	50]	d_loss: 0.5911	g_loss: 1.9291
Epoch [35/	50]	d_loss: 0.4727	g_loss: 3.3866
Epoch [35/	50]	d_loss: 0.4479	g_loss: 3.2767
Epoch [35/	50]	d_loss: 0.4012	g_loss: 4.0132
Epoch [35/	50]	d_loss: 0.3661	g_loss: 4.1678
Epoch [35/	50]	d_loss: 0.6738	g_loss: 4.4672
Epoch [35/	50]	d_loss: 0.3769	g_loss: 4.6703
Epoch [35/	50]	d_loss: 0.4344	g_loss: 3.1200
Epoch [35/	50]	d_loss: 0.4138	g_loss: 3.1643
Epoch [35/	50]	d_loss: 0.3823	g_loss: 4.0756
Epoch [35/	50]	d_loss: 0.4335	g_loss: 2.1626
Epoch [35/	50]	d_loss: 0.4807	g_loss: 2.7795
Epoch [35/	50]	d_loss: 0.4693	g_loss: 3.0518
Epoch [35/	50]	d_loss: 0.3536	g_loss: 3.4110
Epoch [35/	50]	d_loss: 0.4100	g_loss: 3.8148
Epoch [35/	50]	d_loss: 0.5584	g_loss: 2.6735

Epoch [35/	50]	d_loss: 0.4685	g_loss: 3.8946
Epoch [35/	50]	d_loss: 0.5591	g_loss: 4.5504
Epoch [35/	50]	d_loss: 0.3837	g_loss: 3.2792
Epoch [35/	50]	d_loss: 0.4187	g_loss: 3.9780
Epoch [35/	50]	d_loss: 0.4792	g_loss: 2.6396
Epoch [35/	50]	d_loss: 0.4111	g_loss: 2.9881
Epoch [35/	50]	d_loss: 0.3997	g_loss: 3.3727
Epoch [35/	50]	d_loss: 0.4006	g_loss: 3.6323
Epoch [35/	50]	d_loss: 0.3798	g_loss: 3.6103
Epoch [35/	50]	d_loss: 0.3885	g_loss: 3.9339
Epoch [35/	50]	d_loss: 0.5346	g_loss: 2.4160
Epoch [35/	50]	d_loss: 0.3868	g_loss: 3.1399
Epoch [35/	50]	d_loss: 0.3913	g_loss: 2.0375
Epoch [35/	50]	d_loss: 0.5034	g_loss: 3.7023
Epoch [35/	50]	d_loss: 0.6850	g_loss: 1.0230
Epoch [35/	50]	d_loss: 0.5755	g_loss: 4.1791
Epoch [35/	50]	d_loss: 0.6458	g_loss: 2.1804
Epoch [35/	50]	d_loss: 0.3971	g_loss: 4.3707
Epoch [35/	50]	d_loss: 0.6382	g_loss: 5.3247
Epoch [35/	50]	d_loss: 0.4120	g_loss: 3.8589
Epoch [35/	50]	d_loss: 0.3795	g_loss: 3.2836
Epoch [36/	50]	d_loss: 0.5292	g_loss: 3.6657
Epoch [36/	50]	d_loss: 0.4039	g_loss: 3.1069
Epoch [36/	50]	d_loss: 0.8253	g_loss: 1.7204
Epoch [36/	50]	d_loss: 0.4181	g_loss: 3.2227
Epoch [36/	50]	d_loss: 0.3726	g_loss: 3.7593
Epoch [36/	50]	d_loss: 0.4591	g_loss: 2.6272
Epoch [36/	50]	d_loss: 0.4184	g_loss: 3.7421
Epoch [36/	50]	d_loss: 0.5865	g_loss: 2.0995
Epoch [36/	50]	d_loss: 0.5897	g_loss: 3.4115
Epoch [36/	50]	d_loss: 0.3672	g_loss: 4.4893
Epoch [36/	50]	d_loss: 0.4046	g_loss: 3.6308
Epoch [36/	50]	d_loss: 0.4906	g_loss: 2.8479
Epoch [36/	50]	d_loss: 0.3998	g_loss: 2.3808
Epoch [36/	50]	d_loss: 0.5376	g_loss: 3.5189
Epoch [36/	50]	d_loss: 0.4552	g_loss: 3.6080
Epoch [36/	50]	d_loss: 0.5899	g_loss: 3.5947
Epoch [36/	50]	d_loss: 0.4173	g_loss: 3.0882
Epoch [36/	50]	d_loss: 0.3781	g_loss: 3.5494
Epoch [36/	50]	d_loss: 0.3684	g_loss: 4.6990
Epoch [36/	50]	d_loss: 0.4098	g_loss: 5.2975
Epoch [36/	50]	d_loss: 0.4674	g_loss: 2.9766
Epoch [36/	50]	d_loss: 0.4442	g_loss: 2.8574
Epoch [36/	50]	d_loss: 0.3807	g_loss: 4.0237
Epoch [36/	50]	d_loss: 0.4967	g_loss: 2.9774
Epoch [36/	50]	d_loss: 0.4993	g_loss: 2.0482
Epoch [36/	50]	d_loss: 0.4302	g_loss: 3.3840
Epoch [36/	50]	d_loss: 0.4774	g_loss: 2.4759

Epoch [36/	50]	d_loss: 0.5961	g_loss: 2.3556
Epoch [36/	50]	d_loss: 0.4064	g_loss: 3.5057
Epoch [36/	50]	d_loss: 0.3938	g_loss: 3.5697
Epoch [36/	50]	d_loss: 0.5885	g_loss: 3.4509
Epoch [36/	50]	d_loss: 0.5086	g_loss: 2.4899
Epoch [36/	50]	d_loss: 0.3666	g_loss: 4.2874
Epoch [36/	50]	d_loss: 0.4283	g_loss: 2.7654
Epoch [36/	50]	d_loss: 0.4116	g_loss: 3.4024
Epoch [36/	50]	d_loss: 0.3877	g_loss: 3.3495
Epoch [36/	50]	d_loss: 0.4137	g_loss: 3.0947
Epoch [36/	50]	d_loss: 0.3700	g_loss: 4.3002
Epoch [36/	50]	d_loss: 0.3518	g_loss: 4.1822
Epoch [36/	50]	d_loss: 0.7171	g_loss: 3.4273
Epoch [36/	50]	d_loss: 0.4453	g_loss: 3.6642
Epoch [36/	50]	d_loss: 0.6743	g_loss: 2.2842
Epoch [36/	50]	d_loss: 0.3799	g_loss: 3.3002
Epoch [36/	50]	d_loss: 0.3585	g_loss: 3.6326
Epoch [36/	50]	d_loss: 0.6675	g_loss: 2.1843
Epoch [36/	50]	d_loss: 0.6182	g_loss: 2.0314
Epoch [36/	50]	d_loss: 0.4260	g_loss: 3.5343
Epoch [36/	50]	d_loss: 0.3700	g_loss: 2.7926
Epoch [36/	50]	d_loss: 0.4757	g_loss: 3.1916
Epoch [36/	50]	d_loss: 0.4219	g_loss: 2.6365
Epoch [36/	50]	d_loss: 0.4181	g_loss: 3.9579
Epoch [36/	50]	d_loss: 0.7248	g_loss: 4.8401
Epoch [36/	50]	d_loss: 0.3500	g_loss: 4.3541
Epoch [36/	50]	d_loss: 0.4753	g_loss: 3.2516
Epoch [36/	50]	d_loss: 0.4561	g_loss: 2.8320
Epoch [36/	50]	d_loss: 0.3691	g_loss: 4.5400
Epoch [36/	50]	d_loss: 0.4317	g_loss: 2.1383
Epoch [37/	50]	d_loss: 0.6521	g_loss: 1.4431
Epoch [37/	50]	d_loss: 0.4668	g_loss: 4.0120
Epoch [37/	50]	d_loss: 0.3869	g_loss: 3.5115
Epoch [37/	50]	d_loss: 0.4649	g_loss: 3.1343
Epoch [37/	50]	d_loss: 0.3509	g_loss: 3.9192
Epoch [37/	50]	d_loss: 0.6210	g_loss: 3.3892
Epoch [37/	50]	d_loss: 0.4585	g_loss: 4.5094
Epoch [37/	50]	d_loss: 0.4134	g_loss: 4.1186
Epoch [37/	50]	d_loss: 0.4278	g_loss: 3.0500
Epoch [37/	50]	d_loss: 0.3714	g_loss: 3.8500
Epoch [37/	50]	d_loss: 0.3836	g_loss: 4.7820
Epoch [37/	50]	d_loss: 0.4661	g_loss: 3.8734
Epoch [37/	50]	d_loss: 0.4608	g_loss: 2.7409
Epoch [37/	50]	d_loss: 0.4153	g_loss: 3.7467
Epoch [37/	50]	d_loss: 0.4390	g_loss: 2.9385
Epoch [37/	50]	d_loss: 0.4431	g_loss: 3.0778
Epoch [37/	50]	d_loss: 0.4040	g_loss: 3.1073
Epoch [37/	50]	d_loss: 0.6133	g_loss: 3.2599

Epoch [37/	50]	d_loss: 0.4063	g_loss: 3.1056
Epoch [37/	50]	d_loss: 0.3887	g_loss: 3.0204
Epoch [37/	50]	d_loss: 0.4036	g_loss: 3.1346
Epoch [37/	50]	d_loss: 0.5943	g_loss: 4.2494
Epoch [37/	50]	d_loss: 0.4100	g_loss: 4.0553
Epoch [37/	50]	d_loss: 0.4621	g_loss: 3.3108
Epoch [37/	50]	d_loss: 0.4534	g_loss: 2.8383
Epoch [37/	50]	d_loss: 0.3641	g_loss: 3.7350
Epoch [37/	50]	d_loss: 0.4653	g_loss: 3.4380
Epoch [37/	50]	d_loss: 0.8079	g_loss: 1.9407
Epoch [37/	50]	d_loss: 0.4922	g_loss: 4.5916
Epoch [37/	50]	d_loss: 0.5032	g_loss: 4.4551
Epoch [37/	50]	d_loss: 0.7338	g_loss: 1.2497
Epoch [37/	50]	d_loss: 0.3908	g_loss: 4.2496
Epoch [37/	50]	d_loss: 0.3916	g_loss: 3.1149
Epoch [37/	50]	d_loss: 0.4128	g_loss: 3.6273
Epoch [37/	50]	d_loss: 0.3804	g_loss: 2.4502
Epoch [37/	50]	d_loss: 0.3577	g_loss: 2.9263
Epoch [37/	50]	d_loss: 0.3692	g_loss: 4.3706
Epoch [37/	50]	d_loss: 0.3936	g_loss: 3.3378
Epoch [37/	50]	d_loss: 0.4249	g_loss: 2.4530
Epoch [37/	50]	d_loss: 0.3807	g_loss: 2.4129
Epoch [37/	50]	d_loss: 0.4169	g_loss: 3.5147
Epoch [37/	50]	d_loss: 0.4175	g_loss: 3.2324
Epoch [37/	50]	d_loss: 0.4562	g_loss: 3.4292
Epoch [37/	50]	d_loss: 0.3976	g_loss: 3.1576
Epoch [37/	50]	d_loss: 0.4033	g_loss: 3.8021
Epoch [37/	50]	d_loss: 0.3990	g_loss: 4.2608
Epoch [37/	50]	d_loss: 0.4846	g_loss: 3.1957
Epoch [37/	50]	d_loss: 0.3970	g_loss: 2.7149
Epoch [37/	50]	d_loss: 0.4035	g_loss: 3.0497
Epoch [37/	50]	d_loss: 0.3877	g_loss: 3.2022
Epoch [37/	50]	d_loss: 0.5055	g_loss: 1.6555
Epoch [37/	50]	d_loss: 0.4081	g_loss: 3.2544
Epoch [37/	50]	d_loss: 0.4451	g_loss: 3.6118
Epoch [37/	50]	d_loss: 0.5333	g_loss: 4.1296
Epoch [37/	50]	d_loss: 0.3909	g_loss: 3.5178
Epoch [37/	50]	d_loss: 0.3616	g_loss: 3.8700
Epoch [37/	50]	d_loss: 0.4352	g_loss: 3.1819
Epoch [38/	50]	d_loss: 0.5150	g_loss: 2.9556
Epoch [38/	50]	d_loss: 0.4285	g_loss: 4.1124
Epoch [38/	50]	d_loss: 0.4120	g_loss: 3.5878
Epoch [38/	50]	d_loss: 0.5205	g_loss: 2.3323
Epoch [38/	50]	d_loss: 0.6643	g_loss: 2.0882
Epoch [38/	50]	d_loss: 0.4118	g_loss: 3.9559
Epoch [38/	50]	d_loss: 0.3881	g_loss: 2.2702
Epoch [38/	50]	d_loss: 0.5189	g_loss: 4.1122
Epoch [38/	50]	d_loss: 0.3754	g_loss: 5.0968

Epoch [38/	50]	d_loss: 0.3930	g_loss: 3.6045
Epoch [38/	50]	d_loss: 0.4065	g_loss: 3.0088
Epoch [38/	50]	d_loss: 0.8607	g_loss: 4.3984
Epoch [38/	50]	d_loss: 0.3909	g_loss: 3.7486
Epoch [38/	50]	d_loss: 0.4730	g_loss: 3.0902
Epoch [38/	50]	d_loss: 0.5319	g_loss: 3.0717
Epoch [38/	50]	d_loss: 0.5245	g_loss: 2.2592
Epoch [38/	50]	d_loss: 0.3766	g_loss: 4.3195
Epoch [38/	50]	d_loss: 0.4903	g_loss: 4.0298
Epoch [38/	50]	d_loss: 0.4653	g_loss: 2.8577
Epoch [38/	50]	d_loss: 0.4907	g_loss: 3.8274
Epoch [38/	50]	d_loss: 0.4673	g_loss: 3.4156
Epoch [38/	50]	d_loss: 0.5812	g_loss: 4.0142
Epoch [38/	50]	d_loss: 0.4184	g_loss: 4.5081
Epoch [38/	50]	d_loss: 0.6167	g_loss: 3.0154
Epoch [38/	50]	d_loss: 0.4156	g_loss: 2.9813
Epoch [38/	50]	d_loss: 0.4483	g_loss: 3.3582
Epoch [38/	50]	d_loss: 0.4518	g_loss: 3.1937
Epoch [38/	50]	d_loss: 0.4199	g_loss: 5.1942
Epoch [38/	50]	d_loss: 0.5537	g_loss: 5.2024
Epoch [38/	50]	d_loss: 0.4043	g_loss: 3.6584
Epoch [38/	50]	d_loss: 0.4094	g_loss: 3.1442
Epoch [38/	50]	d_loss: 0.3813	g_loss: 3.6695
Epoch [38/	50]	d_loss: 0.4459	g_loss: 3.5334
Epoch [38/	50]	d_loss: 0.4385	g_loss: 3.8495
Epoch [38/	50]	d_loss: 0.3742	g_loss: 3.0204
Epoch [38/	50]	d_loss: 0.3822	g_loss: 4.4109
Epoch [38/	50]	d_loss: 0.3617	g_loss: 3.7607
Epoch [38/	50]	d_loss: 0.3989	g_loss: 3.1263
Epoch [38/	50]	d_loss: 0.4722	g_loss: 2.5973
Epoch [38/	50]	d_loss: 0.7947	g_loss: 4.6770
Epoch [38/	50]	d_loss: 0.4505	g_loss: 5.3810
Epoch [38/	50]	d_loss: 0.3704	g_loss: 4.7934
Epoch [38/	50]	d_loss: 0.3844	g_loss: 3.3426
Epoch [38/	50]	d_loss: 0.4705	g_loss: 5.2544
Epoch [38/	50]	d_loss: 0.3830	g_loss: 3.4363
Epoch [38/	50]	d_loss: 0.4003	g_loss: 2.7977
Epoch [38/	50]	d_loss: 0.4315	g_loss: 3.8200
Epoch [38/	50]	d_loss: 0.4563	g_loss: 2.6066
Epoch [38/	50]	d_loss: 0.3930	g_loss: 4.4483
Epoch [38/	50]	d_loss: 0.4153	g_loss: 3.7158
Epoch [38/	50]	d_loss: 0.4108	g_loss: 3.7661
Epoch [38/	50]	d_loss: 0.4568	g_loss: 3.7564
Epoch [38/	50]	d_loss: 0.4298	g_loss: 3.0811
Epoch [38/	50]	d_loss: 0.3868	g_loss: 4.3380
Epoch [38/	50]	d_loss: 0.3754	g_loss: 3.7809
Epoch [38/	50]	d_loss: 0.3734	g_loss: 4.4310
Epoch [38/	50]	d_loss: 0.4213	g_loss: 2.5907

Epoch [39/	50]	d_loss: 0.4264	g_loss: 3.8048
Epoch [39/	50]	d_loss: 0.5223	g_loss: 4.1906
Epoch [39/	50]	d_loss: 0.4310	g_loss: 2.8645
Epoch [39/	50]	d_loss: 0.3783	g_loss: 4.8988
Epoch [39/	50]	d_loss: 0.4983	g_loss: 4.2092
Epoch [39/	50]	d_loss: 0.3849	g_loss: 3.5719
Epoch [39/	50]	d_loss: 0.4360	g_loss: 3.8128
Epoch [39/	50]	d_loss: 0.4623	g_loss: 2.4115
Epoch [39/	50]	d_loss: 0.3948	g_loss: 3.4684
Epoch [39/	50]	d_loss: 0.4212	g_loss: 3.8373
Epoch [39/	50]	d_loss: 0.3924	g_loss: 3.1098
Epoch [39/	50]	d_loss: 0.4913	g_loss: 1.8380
Epoch [39/	50]	d_loss: 0.4454	g_loss: 4.0106
Epoch [39/	50]	d_loss: 0.4089	g_loss: 3.6008
Epoch [39/	50]	d_loss: 0.4178	g_loss: 4.4914
Epoch [39/	50]	d_loss: 0.4221	g_loss: 3.7289
Epoch [39/	50]	d_loss: 1.1703	g_loss: 1.0381
Epoch [39/	50]	d_loss: 0.4051	g_loss: 4.2222
Epoch [39/	50]	d_loss: 0.6060	g_loss: 3.9896
Epoch [39/	50]	d_loss: 0.5186	g_loss: 3.0619
Epoch [39/	50]	d_loss: 0.3823	g_loss: 4.0438
Epoch [39/	50]	d_loss: 0.3880	g_loss: 3.2053
Epoch [39/	50]	d_loss: 0.6976	g_loss: 4.2159
Epoch [39/	50]	d_loss: 0.4402	g_loss: 3.3123
Epoch [39/	50]	d_loss: 1.0329	g_loss: 3.4159
Epoch [39/	50]	d_loss: 0.7613	g_loss: 3.3894
Epoch [39/	50]	d_loss: 0.7347	g_loss: 3.4258
Epoch [39/	50]	d_loss: 0.4151	g_loss: 3.5096
Epoch [39/	50]	d_loss: 0.4158	g_loss: 3.3071
Epoch [39/	50]	d_loss: 0.4044	g_loss: 4.4274
Epoch [39/	50]	d_loss: 0.3870	g_loss: 3.4447
Epoch [39/	50]	d_loss: 0.3644	g_loss: 3.3638
Epoch [39/	50]	d_loss: 0.3888	g_loss: 3.7717
Epoch [39/	50]	d_loss: 0.9395	g_loss: 0.9776
Epoch [39/	50]	d_loss: 0.5105	g_loss: 3.5915
Epoch [39/	50]	d_loss: 0.4136	g_loss: 3.9798
Epoch [39/	50]	d_loss: 0.3865	g_loss: 4.0228
Epoch [39/	50]	d_loss: 0.5019	g_loss: 3.4221
Epoch [39/	50]	d_loss: 1.0299	g_loss: 1.2941
Epoch [39/	50]	d_loss: 0.4590	g_loss: 3.6900
Epoch [39/	50]	d_loss: 0.4344	g_loss: 2.8192
Epoch [39/	50]	d_loss: 0.8115	g_loss: 0.9596
Epoch [39/	50]	d_loss: 0.5633	g_loss: 1.4551
Epoch [39/	50]	d_loss: 0.4194	g_loss: 3.0585
Epoch [39/	50]	d_loss: 1.0312	g_loss: 4.8377
Epoch [39/	50]	d_loss: 0.4997	g_loss: 2.4844
Epoch [39/	50]	d_loss: 0.6108	g_loss: 3.2172
Epoch [39/	50]	d_loss: 0.3892	g_loss: 4.3599

Epoch [39/	50]	d_loss: 0.4002	g_loss: 4.1110
Epoch [39/	50]	d_loss: 0.4261	g_loss: 3.9447
Epoch [39/	50]	d_loss: 0.4065	g_loss: 2.8417
Epoch [39/	50]	d_loss: 0.3708	g_loss: 3.0347
Epoch [39/	50]	d_loss: 0.4457	g_loss: 2.5921
Epoch [39/	50]	d_loss: 0.5043	g_loss: 4.6878
Epoch [39/	50]	d_loss: 0.4994	g_loss: 2.8262
Epoch [39/	50]	d_loss: 0.3669	g_loss: 3.2954
Epoch [39/	50]	d_loss: 0.4846	g_loss: 2.7920
Epoch [40/	50]	d_loss: 0.4387	g_loss: 3.7476
Epoch [40/	50]	d_loss: 0.3952	g_loss: 3.3877
Epoch [40/	50]	d_loss: 0.3626	g_loss: 4.1278
Epoch [40/	50]	d_loss: 0.5451	g_loss: 2.7031
Epoch [40/	50]	d_loss: 0.4549	g_loss: 4.3791
Epoch [40/	50]	d_loss: 0.3916	g_loss: 3.3319
Epoch [40/	50]	d_loss: 0.4413	g_loss: 3.3672
Epoch [40/	50]	d_loss: 0.4375	g_loss: 3.5010
Epoch [40/	50]	d_loss: 0.3979	g_loss: 3.6044
Epoch [40/	50]	d_loss: 0.4773	g_loss: 2.2789
Epoch [40/	50]	d_loss: 0.5345	g_loss: 2.7584
Epoch [40/	50]	d_loss: 0.5233	g_loss: 2.3578
Epoch [40/	50]	d_loss: 0.3528	g_loss: 2.7331
Epoch [40/	50]	d_loss: 0.4380	g_loss: 3.8965
Epoch [40/	50]	d_loss: 0.3746	g_loss: 4.6666
Epoch [40/	50]	d_loss: 0.4010	g_loss: 4.0298
Epoch [40/	50]	d_loss: 0.5585	g_loss: 2.4450
Epoch [40/	50]	d_loss: 0.5220	g_loss: 4.3279
Epoch [40/	50]	d_loss: 0.4057	g_loss: 4.0290
Epoch [40/	50]	d_loss: 0.3710	g_loss: 4.2662
Epoch [40/	50]	d_loss: 0.4421	g_loss: 3.9075
Epoch [40/	50]	d_loss: 0.3789	g_loss: 3.8914
Epoch [40/	50]	d_loss: 0.3895	g_loss: 4.4120
Epoch [40/	50]	d_loss: 0.4006	g_loss: 3.6616
Epoch [40/	50]	d_loss: 0.3782	g_loss: 4.4928
Epoch [40/	50]	d_loss: 0.7514	g_loss: 0.8517
Epoch [40/	50]	d_loss: 0.3977	g_loss: 3.1826
Epoch [40/	50]	d_loss: 0.5282	g_loss: 2.0389
Epoch [40/	50]	d_loss: 0.3715	g_loss: 2.7773
Epoch [40/	50]	d_loss: 0.4805	g_loss: 3.8157
Epoch [40/	50]	d_loss: 0.7823	g_loss: 3.8155
Epoch [40/	50]	d_loss: 0.4097	g_loss: 4.0751
Epoch [40/	50]	d_loss: 0.5079	g_loss: 2.7071
Epoch [40/	50]	d_loss: 0.4099	g_loss: 4.3553
Epoch [40/	50]	d_loss: 0.3611	g_loss: 3.8060
Epoch [40/	50]	d_loss: 0.4651	g_loss: 3.0928
Epoch [40/	50]	d_loss: 0.6090	g_loss: 2.8465
Epoch [40/	50]	d_loss: 0.3858	g_loss: 3.3207
Epoch [40/	50]	d_loss: 0.5023	g_loss: 4.0441

Epoch [40/	50]	d_loss: 0.4455	g_loss: 3.8093
Epoch [40/	50]	d_loss: 0.3966	g_loss: 2.4130
Epoch [40/	50]	d_loss: 0.3562	g_loss: 4.2823
Epoch [40/	50]	d_loss: 0.3970	g_loss: 5.1500
Epoch [40/	50]	d_loss: 0.4009	g_loss: 4.2721
Epoch [40/	50]	d_loss: 0.3777	g_loss: 3.6258
Epoch [40/	50]	d_loss: 0.5835	g_loss: 4.8409
Epoch [40/	50]	d_loss: 0.3928	g_loss: 4.1185
Epoch [40/	50]	d_loss: 0.3873	g_loss: 3.4915
Epoch [40/	50]	d_loss: 0.3768	g_loss: 4.1380
Epoch [40/	50]	d_loss: 0.4280	g_loss: 3.6562
Epoch [40/	50]	d_loss: 0.3903	g_loss: 2.4816
Epoch [40/	50]	d_loss: 0.4256	g_loss: 3.0482
Epoch [40/	50]	d_loss: 0.4037	g_loss: 4.1762
Epoch [40/	50]	d_loss: 0.5241	g_loss: 3.4739
Epoch [40/	50]	d_loss: 0.3710	g_loss: 3.8720
Epoch [40/	50]	d_loss: 0.3821	g_loss: 2.8998
Epoch [40/	50]	d_loss: 0.4137	g_loss: 3.3562
Epoch [41/	50]	d_loss: 2.2830	g_loss: 5.4999
Epoch [41/	50]	d_loss: 0.3619	g_loss: 4.1905
Epoch [41/	50]	d_loss: 0.3763	g_loss: 3.8341
Epoch [41/	50]	d_loss: 0.3713	g_loss: 3.5024
Epoch [41/	50]	d_loss: 0.3789	g_loss: 2.8978
Epoch [41/	50]	d_loss: 0.7244	g_loss: 1.3538
Epoch [41/	50]	d_loss: 0.5451	g_loss: 3.3056
Epoch [41/	50]	d_loss: 0.4124	g_loss: 3.5244
Epoch [41/	50]	d_loss: 0.3800	g_loss: 3.1503
Epoch [41/	50]	d_loss: 0.4078	g_loss: 3.3693
Epoch [41/	50]	d_loss: 0.3944	g_loss: 3.0097
Epoch [41/	50]	d_loss: 0.4294	g_loss: 3.1330
Epoch [41/	50]	d_loss: 0.4110	g_loss: 3.2694
Epoch [41/	50]	d_loss: 0.3909	g_loss: 3.3973
Epoch [41/	50]	d_loss: 0.4515	g_loss: 4.4214
Epoch [41/	50]	d_loss: 0.3992	g_loss: 3.3127
Epoch [41/	50]	d_loss: 0.3609	g_loss: 4.3072
Epoch [41/	50]	d_loss: 0.4278	g_loss: 3.3192
Epoch [41/	50]	d_loss: 0.4479	g_loss: 3.4599
Epoch [41/	50]	d_loss: 0.4912	g_loss: 1.5474
Epoch [41/	50]	d_loss: 0.5988	g_loss: 2.5619
Epoch [41/	50]	d_loss: 0.3629	g_loss: 4.5092
Epoch [41/	50]	d_loss: 0.4727	g_loss: 2.5598
Epoch [41/	50]	d_loss: 0.4196	g_loss: 3.8352
Epoch [41/	50]	d_loss: 0.4363	g_loss: 3.6703
Epoch [41/	50]	d_loss: 0.4814	g_loss: 3.0014
Epoch [41/	50]	d_loss: 0.8624	g_loss: 1.8133
Epoch [41/	50]	d_loss: 0.4889	g_loss: 3.5024
Epoch [41/	50]	d_loss: 0.5755	g_loss: 2.9208
Epoch [41/	50]	d_loss: 0.4108	g_loss: 3.8591

Epoch [41/	50]	d_loss: 0.3973	g_loss: 4.4182
Epoch [41/	50]	d_loss: 0.3791	g_loss: 3.5541
Epoch [41/	50]	d_loss: 0.4073	g_loss: 3.9308
Epoch [41/	50]	d_loss: 0.4251	g_loss: 3.3102
Epoch [41/	50]	d_loss: 0.4251	g_loss: 2.0309
Epoch [41/	50]	d_loss: 0.4032	g_loss: 3.6964
Epoch [41/	50]	d_loss: 0.4179	g_loss: 4.0943
Epoch [41/	50]	d_loss: 0.5456	g_loss: 4.0625
Epoch [41/	50]	d_loss: 0.4621	g_loss: 2.6482
Epoch [41/	50]	d_loss: 0.4276	g_loss: 3.9690
Epoch [41/	50]	d_loss: 0.4296	g_loss: 4.3070
Epoch [41/	50]	d_loss: 0.4154	g_loss: 2.9877
Epoch [41/	50]	d_loss: 0.4009	g_loss: 3.9865
Epoch [41/	50]	d_loss: 0.4509	g_loss: 3.0879
Epoch [41/	50]	d_loss: 0.3801	g_loss: 3.9455
Epoch [41/	50]	d_loss: 0.4019	g_loss: 3.4397
Epoch [41/	50]	d_loss: 0.3902	g_loss: 3.2178
Epoch [41/	50]	d_loss: 0.5687	g_loss: 2.3774
Epoch [41/	50]	d_loss: 0.5614	g_loss: 2.8970
Epoch [41/	50]	d_loss: 0.4372	g_loss: 4.1727
Epoch [41/	50]	d_loss: 0.4366	g_loss: 3.9078
Epoch [41/	50]	d_loss: 0.3865	g_loss: 1.9234
Epoch [41/	50]	d_loss: 0.3622	g_loss: 3.4279
Epoch [41/	50]	d_loss: 0.5101	g_loss: 2.4482
Epoch [41/	50]	d_loss: 0.3732	g_loss: 4.0181
Epoch [41/	50]	d_loss: 0.4024	g_loss: 2.3107
Epoch [41/	50]	d_loss: 0.3980	g_loss: 3.3900
Epoch [42/	50]	d_loss: 0.4054	g_loss: 3.1419
Epoch [42/	50]	d_loss: 0.5566	g_loss: 5.2937
Epoch [42/	50]	d_loss: 0.3971	g_loss: 3.8449
Epoch [42/	50]	d_loss: 0.4137	g_loss: 3.5105
Epoch [42/	50]	d_loss: 0.3898	g_loss: 3.1762
Epoch [42/	50]	d_loss: 0.4505	g_loss: 4.8384
Epoch [42/	50]	d_loss: 0.6828	g_loss: 2.5451
Epoch [42/	50]	d_loss: 0.6027	g_loss: 3.8223
Epoch [42/	50]	d_loss: 0.6603	g_loss: 4.0764
Epoch [42/	50]	d_loss: 0.3815	g_loss: 4.3391
Epoch [42/	50]	d_loss: 0.8812	g_loss: 3.6694
Epoch [42/	50]	d_loss: 0.4749	g_loss: 3.4005
Epoch [42/	50]	d_loss: 0.4678	g_loss: 3.0871
Epoch [42/	50]	d_loss: 0.4476	g_loss: 3.9044
Epoch [42/	50]	d_loss: 0.4133	g_loss: 2.4040
Epoch [42/	50]	d_loss: 0.4087	g_loss: 5.1987
Epoch [42/	50]	d_loss: 0.8122	g_loss: 4.2840
Epoch [42/	50]	d_loss: 0.4829	g_loss: 3.3126
Epoch [42/	50]	d_loss: 0.4233	g_loss: 2.8970
Epoch [42/	50]	d_loss: 0.3659	g_loss: 4.3414
Epoch [42/	50]	d_loss: 0.4108	g_loss: 3.3164

Epoch [42/	50]	d_loss: 0.3951	g_loss: 3.2061
Epoch [42/	50]	d_loss: 0.4709	g_loss: 2.8804
Epoch [42/	50]	d_loss: 0.3608	g_loss: 3.7872
Epoch [42/	50]	d_loss: 0.4605	g_loss: 2.7076
Epoch [42/	50]	d_loss: 0.4527	g_loss: 2.4741
Epoch [42/	50]	d_loss: 0.4149	g_loss: 3.1104
Epoch [42/	50]	d_loss: 0.3811	g_loss: 3.7374
Epoch [42/	50]	d_loss: 0.4283	g_loss: 3.7170
Epoch [42/	50]	d_loss: 0.4344	g_loss: 3.8911
Epoch [42/	50]	d_loss: 0.7618	g_loss: 0.7497
Epoch [42/	50]	d_loss: 0.3685	g_loss: 5.0500
Epoch [42/	50]	d_loss: 0.4314	g_loss: 3.2328
Epoch [42/	50]	d_loss: 0.6218	g_loss: 4.2249
Epoch [42/	50]	d_loss: 0.4633	g_loss: 2.9143
Epoch [42/	50]	d_loss: 0.3676	g_loss: 3.7551
Epoch [42/	50]	d_loss: 0.5312	g_loss: 3.6361
Epoch [42/	50]	d_loss: 0.4293	g_loss: 2.3008
Epoch [42/	50]	d_loss: 0.3929	g_loss: 2.1225
Epoch [42/	50]	d_loss: 0.4363	g_loss: 3.9521
Epoch [42/	50]	d_loss: 0.4506	g_loss: 5.0615
Epoch [42/	50]	d_loss: 0.5811	g_loss: 3.7047
Epoch [42/	50]	d_loss: 0.3591	g_loss: 3.8255
Epoch [42/	50]	d_loss: 0.3787	g_loss: 3.6636
Epoch [42/	50]	d_loss: 0.4330	g_loss: 3.4707
Epoch [42/	50]	d_loss: 0.6313	g_loss: 2.7659
Epoch [42/	50]	d_loss: 0.4032	g_loss: 4.0682
Epoch [42/	50]	d_loss: 0.5450	g_loss: 2.7930
Epoch [42/	50]	d_loss: 0.5634	g_loss: 1.9915
Epoch [42/	50]	d_loss: 0.3786	g_loss: 4.2718
Epoch [42/	50]	d_loss: 0.4098	g_loss: 3.8668
Epoch [42/	50]	d_loss: 0.5051	g_loss: 4.3898
Epoch [42/	50]	d_loss: 0.3940	g_loss: 3.0758
Epoch [42/	50]	d_loss: 0.4542	g_loss: 3.5314
Epoch [42/	50]	d_loss: 0.5763	g_loss: 2.3832
Epoch [42/	50]	d_loss: 0.3809	g_loss: 4.3567
Epoch [42/	50]	d_loss: 0.3703	g_loss: 4.4926
Epoch [43/	50]	d_loss: 0.8999	g_loss: 4.3569
Epoch [43/	50]	d_loss: 0.3910	g_loss: 4.0867
Epoch [43/	50]	d_loss: 0.3992	g_loss: 3.6903
Epoch [43/	50]	d_loss: 0.4384	g_loss: 3.9263
Epoch [43/	50]	d_loss: 0.3723	g_loss: 3.0686
Epoch [43/	50]	d_loss: 0.4435	g_loss: 3.2850
Epoch [43/	50]	d_loss: 0.4012	g_loss: 4.2054
Epoch [43/	50]	d_loss: 0.5242	g_loss: 2.8219
Epoch [43/	50]	d_loss: 0.4292	g_loss: 3.0953
Epoch [43/	50]	d_loss: 0.3873	g_loss: 4.6031
Epoch [43/	50]	d_loss: 0.4376	g_loss: 3.6755
Epoch [43/	50]	d_loss: 0.4017	g_loss: 3.1504

Epoch [43/	50]		d_loss:	0.4109		g_loss:	3.1853
Epoch [43/	50]		d_loss:	0.4337		g_loss:	2.6498
Epoch [43/	50]		d_loss:	0.4009		g_loss:	3.5495
Epoch [43/	50]		d_loss:	0.3938		g_loss:	3.2280
Epoch [43/	50]		d_loss:	0.4045		g_loss:	3.7760
Epoch [43/	50]		d_loss:	0.4338		g_loss:	3.1251
Epoch [43/	50]		d_loss:	0.4722		g_loss:	4.0923
Epoch [43/	50]		d_loss:	0.4732		g_loss:	4.5467
Epoch [43/	50]		d_loss:	0.3815		g_loss:	4.8673
Epoch [43/	50]		d_loss:	0.3955		g_loss:	3.6952
Epoch [43/	50]		d_loss:	0.3621		g_loss:	4.2897
Epoch [43/	50]		d_loss:	0.3655		g_loss:	4.1517
Epoch [43/	50]		d_loss:	0.5580		g_loss:	2.5566
Epoch [43/	50]		d_loss:	0.3858		g_loss:	3.7632
Epoch [43/	50]		d_loss:	0.4347		g_loss:	4.1238
Epoch [43/	50]		d_loss:	0.4230		g_loss:	3.2652
Epoch [43/	50]		d_loss:	0.3765		g_loss:	4.2177
Epoch [43/	50]		d_loss:	0.3795		g_loss:	4.6824
Epoch [43/	50]		d_loss:	0.4351		g_loss:	3.4838
Epoch [43/	50]		d_loss:	0.4758		g_loss:	4.2504
Epoch [43/	50]		d_loss:	0.6568		g_loss:	4.0966
Epoch [43/	50]		d_loss:	0.5531		g_loss:	3.4432
Epoch [43/	50]		d_loss:	0.5220		g_loss:	2.7714
Epoch [43/	50]		d_loss:	0.7506		g_loss:	0.9798
Epoch [43/	50]		d_loss:	0.6471		g_loss:	1.3278
Epoch [43/	50]		d_loss:	0.4040		g_loss:	2.5705
Epoch [43/	50]		d_loss:	0.4530		g_loss:	4.6552
Epoch [43/	50]		d_loss:	0.4526		g_loss:	3.9917
Epoch [43/	50]		d_loss:	0.7641		g_loss:	4.3616
Epoch [43/	50]		d_loss:	0.3924		g_loss:	3.3674
Epoch [43/	50]		d_loss:	0.4155		g_loss:	2.8959
Epoch [43/	50]		d_loss:	0.4449		g_loss:	3.5317
Epoch [43/	50]		d_loss:	0.3569		g_loss:	4.9740
Epoch [43/	50]		d_loss:	0.4646		g_loss:	2.8600
Epoch [43/	50]		d_loss:	0.3792		g_loss:	4.7181
Epoch [43/	50]		d_loss:	0.4078		g_loss:	3.3304
Epoch [43/	50]		d_loss:	0.4418		g_loss:	4.2144
Epoch [43/	50]		d_loss:	0.3784		g_loss:	4.3646
Epoch [43/	50]		d_loss:	0.4360		g_loss:	3.1510
Epoch [43/	50]		d_loss:	0.5300		g_loss:	4.6496
Epoch [43/	50]		d_loss:	0.3799		g_loss:	3.1626
Epoch [43/	50]		d_loss:	0.3758		g_loss:	3.6684
Epoch [43/	50]		d_loss:	0.3854		g_loss:	3.1904
Epoch [43/	50]		d_loss:	0.3976		g_loss:	3.8432
Epoch [43/	50]		d_loss:	0.6406		g_loss:	3.8957
Epoch [44/	50]		d_loss:	0.4834		g_loss:	2.7464
Epoch [44/	50]		d_loss:	0.3881		g_loss:	5.6658
Epoch [44/	50]		d_loss:	0.4058		g_loss:	4.2573

Epoch [44/	50]	d_loss: 0.4506	g_loss: 3.7899
Epoch [44/	50]	d_loss: 0.4268	g_loss: 3.7862
Epoch [44/	50]	d_loss: 0.6542	g_loss: 3.4420
Epoch [44/	50]	d_loss: 0.4074	g_loss: 3.4060
Epoch [44/	50]	d_loss: 0.3615	g_loss: 3.0122
Epoch [44/	50]	d_loss: 0.4272	g_loss: 3.2433
Epoch [44/	50]	d_loss: 0.3757	g_loss: 3.7111
Epoch [44/	50]	d_loss: 0.4324	g_loss: 3.9011
Epoch [44/	50]	d_loss: 0.4344	g_loss: 1.9880
Epoch [44/	50]	d_loss: 0.3766	g_loss: 3.4149
Epoch [44/	50]	d_loss: 0.4317	g_loss: 3.4272
Epoch [44/	50]	d_loss: 0.4095	g_loss: 3.4260
Epoch [44/	50]	d_loss: 0.3813	g_loss: 3.7466
Epoch [44/	50]	d_loss: 0.3670	g_loss: 4.0550
Epoch [44/	50]	d_loss: 1.1683	g_loss: 0.4502
Epoch [44/	50]	d_loss: 0.4287	g_loss: 4.4166
Epoch [44/	50]	d_loss: 0.3908	g_loss: 3.2258
Epoch [44/	50]	d_loss: 0.8338	g_loss: 1.2491
Epoch [44/	50]	d_loss: 0.4855	g_loss: 4.7177
Epoch [44/	50]	d_loss: 0.4253	g_loss: 2.3892
Epoch [44/	50]	d_loss: 0.4047	g_loss: 3.3321
Epoch [44/	50]	d_loss: 0.3812	g_loss: 4.1635
Epoch [44/	50]	d_loss: 0.5460	g_loss: 2.2226
Epoch [44/	50]	d_loss: 0.3779	g_loss: 3.6535
Epoch [44/	50]	d_loss: 0.3704	g_loss: 2.8382
Epoch [44/	50]	d_loss: 0.5820	g_loss: 1.8879
Epoch [44/	50]	d_loss: 0.4420	g_loss: 3.5216
Epoch [44/	50]	d_loss: 0.4111	g_loss: 3.6006
Epoch [44/	50]	d_loss: 0.5242	g_loss: 3.7604
Epoch [44/	50]	d_loss: 0.3614	g_loss: 3.7694
Epoch [44/	50]	d_loss: 0.4599	g_loss: 3.8214
Epoch [44/	50]	d_loss: 0.3661	g_loss: 4.0040
Epoch [44/	50]	d_loss: 0.4307	g_loss: 4.4502
Epoch [44/	50]	d_loss: 0.5133	g_loss: 3.9308
Epoch [44/	50]	d_loss: 0.3840	g_loss: 3.0558
Epoch [44/	50]	d_loss: 0.4400	g_loss: 4.8167
Epoch [44/	50]	d_loss: 0.6134	g_loss: 3.3528
Epoch [44/	50]	d_loss: 0.3864	g_loss: 3.5526
Epoch [44/	50]	d_loss: 0.3575	g_loss: 4.3166
Epoch [44/	50]	d_loss: 0.4700	g_loss: 3.3339
Epoch [44/	50]	d_loss: 0.3791	g_loss: 3.9033
Epoch [44/	50]	d_loss: 0.4780	g_loss: 3.0512
Epoch [44/	50]	d_loss: 0.3910	g_loss: 3.9264
Epoch [44/	50]	d_loss: 0.3806	g_loss: 3.6292
Epoch [44/	50]	d_loss: 0.4164	g_loss: 3.6726
Epoch [44/	50]	d_loss: 0.4079	g_loss: 4.9334
Epoch [44/	50]	d_loss: 0.4378	g_loss: 3.6078
Epoch [44/	50]	d_loss: 0.3771	g_loss: 3.5151

Epoch [44/	50]	d_loss: 0.3983	g_loss: 3.5332
Epoch [44/	50]	d_loss: 0.4813	g_loss: 3.3898
Epoch [44/	50]	d_loss: 0.4727	g_loss: 4.1628
Epoch [44/	50]	d_loss: 0.4210	g_loss: 4.4999
Epoch [44/	50]	d_loss: 0.4061	g_loss: 3.4688
Epoch [44/	50]	d_loss: 0.4236	g_loss: 4.0167
Epoch [45/	50]	d_loss: 0.8329	g_loss: 5.5062
Epoch [45/	50]	d_loss: 0.4214	g_loss: 3.8043
Epoch [45/	50]	d_loss: 0.5093	g_loss: 4.1235
Epoch [45/	50]	d_loss: 0.4483	g_loss: 3.0087
Epoch [45/	50]	d_loss: 0.3751	g_loss: 3.5274
Epoch [45/	50]	d_loss: 0.4625	g_loss: 2.2575
Epoch [45/	50]	d_loss: 0.5222	g_loss: 4.3550
Epoch [45/	50]	d_loss: 0.4017	g_loss: 4.6507
Epoch [45/	50]	d_loss: 0.3648	g_loss: 4.3546
Epoch [45/	50]	d_loss: 0.4012	g_loss: 3.7300
Epoch [45/	50]	d_loss: 0.4111	g_loss: 3.7183
Epoch [45/	50]	d_loss: 0.4058	g_loss: 4.2742
Epoch [45/	50]	d_loss: 0.4039	g_loss: 3.9733
Epoch [45/	50]	d_loss: 0.3946	g_loss: 4.0588
Epoch [45/	50]	d_loss: 0.5748	g_loss: 3.7978
Epoch [45/	50]	d_loss: 0.3830	g_loss: 4.0004
Epoch [45/	50]	d_loss: 0.3727	g_loss: 4.5805
Epoch [45/	50]	d_loss: 0.3904	g_loss: 3.6442
Epoch [45/	50]	d_loss: 0.3792	g_loss: 5.1197
Epoch [45/	50]	d_loss: 0.4790	g_loss: 3.5172
Epoch [45/	50]	d_loss: 0.4008	g_loss: 3.8150
Epoch [45/	50]	d_loss: 0.3691	g_loss: 3.7914
Epoch [45/	50]	d_loss: 0.3751	g_loss: 5.3326
Epoch [45/	50]	d_loss: 0.5168	g_loss: 3.6068
Epoch [45/	50]	d_loss: 0.4077	g_loss: 3.6483
Epoch [45/	50]	d_loss: 0.3790	g_loss: 4.1718
Epoch [45/	50]	d_loss: 0.3926	g_loss: 2.9358
Epoch [45/	50]	d_loss: 0.4453	g_loss: 3.4053
Epoch [45/	50]	d_loss: 0.8318	g_loss: 1.3118
Epoch [45/	50]	d_loss: 0.3779	g_loss: 4.1845
Epoch [45/	50]	d_loss: 0.3803	g_loss: 2.7435
Epoch [45/	50]	d_loss: 0.3946	g_loss: 2.5807
Epoch [45/	50]	d_loss: 0.3761	g_loss: 4.4321
Epoch [45/	50]	d_loss: 0.3818	g_loss: 3.9240
Epoch [45/	50]	d_loss: 0.4000	g_loss: 3.2905
Epoch [45/	50]	d_loss: 0.4286	g_loss: 3.7943
Epoch [45/	50]	d_loss: 0.3674	g_loss: 3.8424
Epoch [45/	50]	d_loss: 0.3627	g_loss: 4.3745
Epoch [45/	50]	d_loss: 0.4715	g_loss: 3.3623
Epoch [45/	50]	d_loss: 0.4248	g_loss: 3.4693
Epoch [45/	50]	d_loss: 0.4023	g_loss: 3.9683
Epoch [45/	50]	d_loss: 0.3636	g_loss: 4.4827

Epoch [45/	50]	d_loss: 0.4327	g_loss: 4.6551
Epoch [45/	50]	d_loss: 0.6144	g_loss: 4.7521
Epoch [45/	50]	d_loss: 0.4128	g_loss: 3.0298
Epoch [45/	50]	d_loss: 0.4756	g_loss: 3.1005
Epoch [45/	50]	d_loss: 0.3576	g_loss: 3.8383
Epoch [45/	50]	d_loss: 0.5394	g_loss: 2.8906
Epoch [45/	50]	d_loss: 0.3997	g_loss: 4.0794
Epoch [45/	50]	d_loss: 0.4574	g_loss: 4.8792
Epoch [45/	50]	d_loss: 0.3861	g_loss: 4.4416
Epoch [45/	50]	d_loss: 0.4587	g_loss: 4.4048
Epoch [45/	50]	d_loss: 0.6919	g_loss: 3.5252
Epoch [45/	50]	d_loss: 0.4887	g_loss: 3.5708
Epoch [45/	50]	d_loss: 0.6953	g_loss: 1.2876
Epoch [45/	50]	d_loss: 0.4136	g_loss: 3.8081
Epoch [45/	50]	d_loss: 0.3667	g_loss: 3.5711
Epoch [46/	50]	d_loss: 0.3897	g_loss: 2.8188
Epoch [46/	50]	d_loss: 0.5577	g_loss: 3.6277
Epoch [46/	50]	d_loss: 0.3637	g_loss: 5.2512
Epoch [46/	50]	d_loss: 0.3812	g_loss: 5.6575
Epoch [46/	50]	d_loss: 0.5064	g_loss: 2.7167
Epoch [46/	50]	d_loss: 0.5104	g_loss: 2.9531
Epoch [46/	50]	d_loss: 0.5566	g_loss: 1.5941
Epoch [46/	50]	d_loss: 0.4368	g_loss: 4.6993
Epoch [46/	50]	d_loss: 0.4500	g_loss: 3.1389
Epoch [46/	50]	d_loss: 0.4459	g_loss: 3.5696
Epoch [46/	50]	d_loss: 0.4703	g_loss: 3.9906
Epoch [46/	50]	d_loss: 0.4136	g_loss: 1.9758
Epoch [46/	50]	d_loss: 0.3804	g_loss: 5.0103
Epoch [46/	50]	d_loss: 0.4152	g_loss: 4.3283
Epoch [46/	50]	d_loss: 0.3783	g_loss: 3.9030
Epoch [46/	50]	d_loss: 0.4189	g_loss: 4.2811
Epoch [46/	50]	d_loss: 0.3746	g_loss: 3.4295
Epoch [46/	50]	d_loss: 0.3754	g_loss: 4.9974
Epoch [46/	50]	d_loss: 0.4606	g_loss: 2.2529
Epoch [46/	50]	d_loss: 0.3675	g_loss: 4.2613
Epoch [46/	50]	d_loss: 0.4097	g_loss: 4.0087
Epoch [46/	50]	d_loss: 0.4746	g_loss: 3.1676
Epoch [46/	50]	d_loss: 0.3814	g_loss: 2.9870
Epoch [46/	50]	d_loss: 0.3709	g_loss: 3.2717
Epoch [46/	50]	d_loss: 0.4281	g_loss: 2.4536
Epoch [46/	50]	d_loss: 0.3615	g_loss: 4.5284
Epoch [46/	50]	d_loss: 0.5188	g_loss: 3.2060
Epoch [46/	50]	d_loss: 0.3763	g_loss: 3.8248
Epoch [46/	50]	d_loss: 0.3805	g_loss: 4.3371
Epoch [46/	50]	d_loss: 0.3949	g_loss: 4.2474
Epoch [46/	50]	d_loss: 0.4001	g_loss: 3.6114
Epoch [46/	50]	d_loss: 0.3405	g_loss: 4.5605
Epoch [46/	50]	d_loss: 0.3575	g_loss: 4.3858

Epoch [46/	50]	d_loss: 0.7132	g_loss: 1.7582
Epoch [46/	50]	d_loss: 0.5065	g_loss: 3.8851
Epoch [46/	50]	d_loss: 0.3630	g_loss: 2.6795
Epoch [46/	50]	d_loss: 0.4357	g_loss: 4.9385
Epoch [46/	50]	d_loss: 0.4248	g_loss: 2.8552
Epoch [46/	50]	d_loss: 0.3907	g_loss: 3.4159
Epoch [46/	50]	d_loss: 0.4651	g_loss: 3.7230
Epoch [46/	50]	d_loss: 0.4257	g_loss: 4.1842
Epoch [46/	50]	d_loss: 0.4164	g_loss: 3.9682
Epoch [46/	50]	d_loss: 0.4033	g_loss: 2.1413
Epoch [46/	50]	d_loss: 0.3648	g_loss: 3.0219
Epoch [46/	50]	d_loss: 0.3545	g_loss: 4.1386
Epoch [46/	50]	d_loss: 0.3906	g_loss: 3.7578
Epoch [46/	50]	d_loss: 0.3762	g_loss: 3.5962
Epoch [46/	50]	d_loss: 0.4107	g_loss: 2.7839
Epoch [46/	50]	d_loss: 0.4676	g_loss: 2.4923
Epoch [46/	50]	d_loss: 0.3637	g_loss: 3.6933
Epoch [46/	50]	d_loss: 0.4304	g_loss: 3.5403
Epoch [46/	50]	d_loss: 0.3676	g_loss: 3.9877
Epoch [46/	50]	d_loss: 0.3594	g_loss: 4.2124
Epoch [46/	50]	d_loss: 0.3745	g_loss: 4.0184
Epoch [46/	50]	d_loss: 0.3665	g_loss: 5.0173
Epoch [46/	50]	d_loss: 0.4019	g_loss: 3.0904
Epoch [46/	50]	d_loss: 0.3580	g_loss: 2.8407
Epoch [47/	50]	d_loss: 0.3748	g_loss: 3.1471
Epoch [47/	50]	d_loss: 0.5574	g_loss: 2.1718
Epoch [47/	50]	d_loss: 0.3676	g_loss: 2.4388
Epoch [47/	50]	d_loss: 0.3578	g_loss: 4.0435
Epoch [47/	50]	d_loss: 0.4705	g_loss: 4.1605
Epoch [47/	50]	d_loss: 0.3841	g_loss: 3.1982
Epoch [47/	50]	d_loss: 0.4389	g_loss: 2.4259
Epoch [47/	50]	d_loss: 0.4711	g_loss: 4.3030
Epoch [47/	50]	d_loss: 0.4204	g_loss: 4.0235
Epoch [47/	50]	d_loss: 0.3596	g_loss: 4.4793
Epoch [47/	50]	d_loss: 0.3731	g_loss: 4.9891
Epoch [47/	50]	d_loss: 0.4166	g_loss: 3.6954
Epoch [47/	50]	d_loss: 0.5440	g_loss: 3.3147
Epoch [47/	50]	d_loss: 0.4786	g_loss: 2.2043
Epoch [47/	50]	d_loss: 0.3719	g_loss: 2.9198
Epoch [47/	50]	d_loss: 0.3827	g_loss: 3.3703
Epoch [47/	50]	d_loss: 0.3879	g_loss: 3.5442
Epoch [47/	50]	d_loss: 0.5290	g_loss: 3.9737
Epoch [47/	50]	d_loss: 0.4715	g_loss: 1.8657
Epoch [47/	50]	d_loss: 0.4029	g_loss: 4.4253
Epoch [47/	50]	d_loss: 0.4104	g_loss: 2.8479
Epoch [47/	50]	d_loss: 0.3960	g_loss: 3.6117
Epoch [47/	50]	d_loss: 0.5077	g_loss: 2.6538
Epoch [47/	50]	d_loss: 0.3772	g_loss: 4.3497

Epoch [47/	50]	d_loss: 0.4425	g_loss: 2.8964
Epoch [47/	50]	d_loss: 0.5439	g_loss: 4.5224
Epoch [47/	50]	d_loss: 0.4706	g_loss: 4.4200
Epoch [47/	50]	d_loss: 0.3954	g_loss: 3.8002
Epoch [47/	50]	d_loss: 0.5079	g_loss: 2.3050
Epoch [47/	50]	d_loss: 0.4342	g_loss: 3.8363
Epoch [47/	50]	d_loss: 0.4347	g_loss: 3.2687
Epoch [47/	50]	d_loss: 0.4162	g_loss: 4.6539
Epoch [47/	50]	d_loss: 0.5262	g_loss: 2.8727
Epoch [47/	50]	d_loss: 0.3967	g_loss: 3.1077
Epoch [47/	50]	d_loss: 0.3862	g_loss: 3.5248
Epoch [47/	50]	d_loss: 0.5478	g_loss: 2.6390
Epoch [47/	50]	d_loss: 0.3932	g_loss: 3.6826
Epoch [47/	50]	d_loss: 0.6893	g_loss: 5.1383
Epoch [47/	50]	d_loss: 0.3970	g_loss: 4.2682
Epoch [47/	50]	d_loss: 0.3919	g_loss: 3.9011
Epoch [47/	50]	d_loss: 0.4043	g_loss: 5.3548
Epoch [47/	50]	d_loss: 0.3920	g_loss: 3.9281
Epoch [47/	50]	d_loss: 0.4504	g_loss: 2.6818
Epoch [47/	50]	d_loss: 0.3711	g_loss: 4.0160
Epoch [47/	50]	d_loss: 0.3836	g_loss: 3.7046
Epoch [47/	50]	d_loss: 0.3756	g_loss: 4.2062
Epoch [47/	50]	d_loss: 0.5147	g_loss: 2.7435
Epoch [47/	50]	d_loss: 0.4016	g_loss: 2.2204
Epoch [47/	50]	d_loss: 0.4676	g_loss: 3.8333
Epoch [47/	50]	d_loss: 0.4876	g_loss: 3.9792
Epoch [47/	50]	d_loss: 0.6492	g_loss: 3.1244
Epoch [47/	50]	d_loss: 0.3699	g_loss: 3.8539
Epoch [47/	50]	d_loss: 0.4158	g_loss: 2.7381
Epoch [47/	50]	d_loss: 0.4962	g_loss: 3.0174
Epoch [47/	50]	d_loss: 0.3588	g_loss: 3.8507
Epoch [47/	50]	d_loss: 0.3707	g_loss: 4.3305
Epoch [47/	50]	d_loss: 0.4915	g_loss: 2.4127
Epoch [48/	50]	d_loss: 0.5588	g_loss: 4.1975
Epoch [48/	50]	d_loss: 0.6023	g_loss: 2.2592
Epoch [48/	50]	d_loss: 0.4086	g_loss: 3.1284
Epoch [48/	50]	d_loss: 0.3953	g_loss: 3.7222
Epoch [48/	50]	d_loss: 0.5419	g_loss: 1.7715
Epoch [48/	50]	d_loss: 0.4459	g_loss: 4.1326
Epoch [48/	50]	d_loss: 0.4566	g_loss: 4.4313
Epoch [48/	50]	d_loss: 0.4312	g_loss: 4.0822
Epoch [48/	50]	d_loss: 0.3856	g_loss: 2.7540
Epoch [48/	50]	d_loss: 0.3625	g_loss: 3.9600
Epoch [48/	50]	d_loss: 0.5614	g_loss: 1.5821
Epoch [48/	50]	d_loss: 0.4700	g_loss: 1.8801
Epoch [48/	50]	d_loss: 0.3556	g_loss: 4.7826
Epoch [48/	50]	d_loss: 0.4384	g_loss: 3.3767
Epoch [48/	50]	d_loss: 0.3909	g_loss: 3.2668

Epoch [48/	50]	d_loss: 0.4718	g_loss: 3.2523
Epoch [48/	50]	d_loss: 0.3906	g_loss: 5.2058
Epoch [48/	50]	d_loss: 1.1771	g_loss: 1.0468
Epoch [48/	50]	d_loss: 0.4301	g_loss: 1.8661
Epoch [48/	50]	d_loss: 0.4670	g_loss: 3.5634
Epoch [48/	50]	d_loss: 0.5237	g_loss: 3.4964
Epoch [48/	50]	d_loss: 0.4048	g_loss: 4.2321
Epoch [48/	50]	d_loss: 0.8973	g_loss: 5.6130
Epoch [48/	50]	d_loss: 0.4116	g_loss: 5.0350
Epoch [48/	50]	d_loss: 0.3640	g_loss: 3.5820
Epoch [48/	50]	d_loss: 0.4062	g_loss: 4.5539
Epoch [48/	50]	d_loss: 0.6343	g_loss: 4.5518
Epoch [48/	50]	d_loss: 0.4186	g_loss: 2.8932
Epoch [48/	50]	d_loss: 0.4686	g_loss: 4.0102
Epoch [48/	50]	d_loss: 0.4362	g_loss: 3.4302
Epoch [48/	50]	d_loss: 0.4821	g_loss: 4.7183
Epoch [48/	50]	d_loss: 0.6429	g_loss: 3.1289
Epoch [48/	50]	d_loss: 0.4969	g_loss: 2.9570
Epoch [48/	50]	d_loss: 0.4844	g_loss: 4.6358
Epoch [48/	50]	d_loss: 0.3626	g_loss: 5.4284
Epoch [48/	50]	d_loss: 0.3664	g_loss: 4.6280
Epoch [48/	50]	d_loss: 0.5722	g_loss: 2.5476
Epoch [48/	50]	d_loss: 0.4138	g_loss: 5.7769
Epoch [48/	50]	d_loss: 0.5345	g_loss: 3.1547
Epoch [48/	50]	d_loss: 0.3876	g_loss: 4.2999
Epoch [48/	50]	d_loss: 0.4554	g_loss: 3.1879
Epoch [48/	50]	d_loss: 0.3744	g_loss: 4.1112
Epoch [48/	50]	d_loss: 0.4390	g_loss: 3.8734
Epoch [48/	50]	d_loss: 0.4740	g_loss: 2.2835
Epoch [48/	50]	d_loss: 0.4019	g_loss: 4.7516
Epoch [48/	50]	d_loss: 0.3718	g_loss: 4.2234
Epoch [48/	50]	d_loss: 0.4813	g_loss: 3.0307
Epoch [48/	50]	d_loss: 0.3643	g_loss: 4.0939
Epoch [48/	50]	d_loss: 0.3854	g_loss: 3.5400
Epoch [48/	50]	d_loss: 0.5300	g_loss: 3.3217
Epoch [48/	50]	d_loss: 0.5012	g_loss: 3.5414
Epoch [48/	50]	d_loss: 0.4201	g_loss: 2.1319
Epoch [48/	50]	d_loss: 0.4537	g_loss: 2.6192
Epoch [48/	50]	d_loss: 0.3688	g_loss: 4.9134
Epoch [48/	50]	d_loss: 0.3823	g_loss: 4.7895
Epoch [48/	50]	d_loss: 0.4184	g_loss: 4.3123
Epoch [48/	50]	d_loss: 0.3818	g_loss: 2.8741
Epoch [49/	50]	d_loss: 0.4039	g_loss: 3.4939
Epoch [49/	50]	d_loss: 0.4129	g_loss: 4.9324
Epoch [49/	50]	d_loss: 0.4086	g_loss: 3.3031
Epoch [49/	50]	d_loss: 0.3849	g_loss: 4.0775
Epoch [49/	50]	d_loss: 0.6657	g_loss: 1.7553
Epoch [49/	50]	d_loss: 0.3704	g_loss: 3.2991

Epoch [49/	50]	d_loss: 0.4465	g_loss: 3.6559
Epoch [49/	50]	d_loss: 0.6404	g_loss: 1.8003
Epoch [49/	50]	d_loss: 0.4056	g_loss: 1.8004
Epoch [49/	50]	d_loss: 0.3765	g_loss: 4.6802
Epoch [49/	50]	d_loss: 0.4271	g_loss: 2.9331
Epoch [49/	50]	d_loss: 0.4302	g_loss: 4.7585
Epoch [49/	50]	d_loss: 0.3573	g_loss: 4.3837
Epoch [49/	50]	d_loss: 0.3630	g_loss: 3.5916
Epoch [49/	50]	d_loss: 0.4722	g_loss: 3.4540
Epoch [49/	50]	d_loss: 0.4048	g_loss: 3.3909
Epoch [49/	50]	d_loss: 0.5120	g_loss: 3.6490
Epoch [49/	50]	d_loss: 0.4175	g_loss: 3.1535
Epoch [49/	50]	d_loss: 0.3761	g_loss: 4.0173
Epoch [49/	50]	d_loss: 0.4199	g_loss: 3.8663
Epoch [49/	50]	d_loss: 0.4021	g_loss: 3.8642
Epoch [49/	50]	d_loss: 0.7924	g_loss: 3.8661
Epoch [49/	50]	d_loss: 0.4384	g_loss: 3.4935
Epoch [49/	50]	d_loss: 0.5674	g_loss: 4.3948
Epoch [49/	50]	d_loss: 0.3982	g_loss: 4.0138
Epoch [49/	50]	d_loss: 0.4385	g_loss: 2.8667
Epoch [49/	50]	d_loss: 0.9655	g_loss: 5.8657
Epoch [49/	50]	d_loss: 0.4253	g_loss: 2.0647
Epoch [49/	50]	d_loss: 0.3629	g_loss: 3.5705
Epoch [49/	50]	d_loss: 0.3981	g_loss: 3.1945
Epoch [49/	50]	d_loss: 0.5181	g_loss: 2.9937
Epoch [49/	50]	d_loss: 0.4337	g_loss: 4.3040
Epoch [49/	50]	d_loss: 0.3807	g_loss: 4.3471
Epoch [49/	50]	d_loss: 0.3950	g_loss: 4.2013
Epoch [49/	50]	d_loss: 0.3920	g_loss: 3.8929
Epoch [49/	50]	d_loss: 0.3941	g_loss: 4.3034
Epoch [49/	50]	d_loss: 0.4484	g_loss: 3.8881
Epoch [49/	50]	d_loss: 0.5409	g_loss: 1.4781
Epoch [49/	50]	d_loss: 0.4224	g_loss: 4.0775
Epoch [49/	50]	d_loss: 0.3562	g_loss: 3.2354
Epoch [49/	50]	d_loss: 0.4750	g_loss: 2.4612
Epoch [49/	50]	d_loss: 0.4387	g_loss: 4.6616
Epoch [49/	50]	d_loss: 0.4260	g_loss: 3.9357
Epoch [49/	50]	d_loss: 0.5077	g_loss: 3.8975
Epoch [49/	50]	d_loss: 0.4156	g_loss: 4.3948
Epoch [49/	50]	d_loss: 0.4259	g_loss: 3.4931
Epoch [49/	50]	d_loss: 0.4042	g_loss: 3.4136
Epoch [49/	50]	d_loss: 0.4902	g_loss: 2.8921
Epoch [49/	50]	d_loss: 0.5072	g_loss: 3.6078
Epoch [49/	50]	d_loss: 0.3523	g_loss: 3.8635
Epoch [49/	50]	d_loss: 0.4764	g_loss: 2.5496
Epoch [49/	50]	d_loss: 0.4926	g_loss: 2.5454
Epoch [49/	50]	d_loss: 0.4413	g_loss: 4.1295
Epoch [49/	50]	d_loss: 0.3890	g_loss: 3.5132

Epoch [49/	50]	d_loss: 0.3933	g_loss: 4.3734
Epoch [49/	50]	d_loss: 0.3489	g_loss: 4.0165
Epoch [49/	50]	d_loss: 0.4961	g_loss: 3.4480
Epoch [50/	50]	d_loss: 1.0283	g_loss: 5.1819
Epoch [50/	50]	d_loss: 0.3695	g_loss: 5.0487
Epoch [50/	50]	d_loss: 0.4990	g_loss: 3.0497
Epoch [50/	50]	d_loss: 0.4017	g_loss: 4.1449
Epoch [50/	50]	d_loss: 0.3926	g_loss: 4.3916
Epoch [50/	50]	d_loss: 0.3881	g_loss: 4.2591
Epoch [50/	50]	d_loss: 0.4249	g_loss: 4.7112
Epoch [50/	50]	d_loss: 0.3663	g_loss: 4.9079
Epoch [50/	50]	d_loss: 0.3781	g_loss: 5.6560
Epoch [50/	50]	d_loss: 0.6450	g_loss: 1.2017
Epoch [50/	50]	d_loss: 0.3790	g_loss: 4.8968
Epoch [50/	50]	d_loss: 0.3963	g_loss: 3.9953
Epoch [50/	50]	d_loss: 0.4045	g_loss: 3.0367
Epoch [50/	50]	d_loss: 0.3600	g_loss: 4.6096
Epoch [50/	50]	d_loss: 0.6446	g_loss: 3.5718
Epoch [50/	50]	d_loss: 0.5641	g_loss: 3.1390
Epoch [50/	50]	d_loss: 0.4200	g_loss: 3.6371
Epoch [50/	50]	d_loss: 0.3715	g_loss: 5.9128
Epoch [50/	50]	d_loss: 0.4360	g_loss: 4.6959
Epoch [50/	50]	d_loss: 0.3710	g_loss: 4.7698
Epoch [50/	50]	d_loss: 0.3986	g_loss: 3.6502
Epoch [50/	50]	d_loss: 0.4343	g_loss: 3.7538
Epoch [50/	50]	d_loss: 0.3934	g_loss: 3.5658
Epoch [50/	50]	d_loss: 0.4767	g_loss: 2.1694
Epoch [50/	50]	d_loss: 0.9228	g_loss: 4.7315
Epoch [50/	50]	d_loss: 0.3666	g_loss: 3.7798
Epoch [50/	50]	d_loss: 0.3995	g_loss: 3.5330
Epoch [50/	50]	d_loss: 0.4143	g_loss: 2.8974
Epoch [50/	50]	d_loss: 0.3751	g_loss: 3.7622
Epoch [50/	50]	d_loss: 0.6938	g_loss: 0.6614
Epoch [50/	50]	d_loss: 0.4133	g_loss: 3.6102
Epoch [50/	50]	d_loss: 0.8395	g_loss: 1.1693
Epoch [50/	50]	d_loss: 0.3770	g_loss: 4.0759
Epoch [50/	50]	d_loss: 0.3931	g_loss: 3.1768
Epoch [50/	50]	d_loss: 0.3771	g_loss: 5.3877
Epoch [50/	50]	d_loss: 0.3739	g_loss: 3.0291
Epoch [50/	50]	d_loss: 0.5557	g_loss: 4.4647
Epoch [50/	50]	d_loss: 0.5678	g_loss: 3.5511
Epoch [50/	50]	d_loss: 0.6021	g_loss: 4.2385
Epoch [50/	50]	d_loss: 0.4375	g_loss: 4.7847
Epoch [50/	50]	d_loss: 0.3848	g_loss: 3.7333
Epoch [50/	50]	d_loss: 0.4286	g_loss: 3.6461
Epoch [50/	50]	d_loss: 0.3863	g_loss: 3.8938
Epoch [50/	50]	d_loss: 0.3854	g_loss: 4.2554
Epoch [50/	50]	d_loss: 0.4308	g_loss: 3.3593

```
Epoch [ 50/ 50] | d_loss: 0.3720 | g_loss: 3.6306
Epoch [ 50/ 50] | d_loss: 0.4486 | g_loss: 3.6524
Epoch [ 50/ 50] | d_loss: 0.3738 | g_loss: 4.4585
Epoch [ 50/ 50] | d_loss: 0.3864 | g_loss: 3.6021
Epoch [ 50/ 50] | d_loss: 0.4851 | g_loss: 2.3101
Epoch [ 50/ 50] | d_loss: 0.3679 | g_loss: 4.5696
Epoch [ 50/ 50] | d_loss: 0.3738 | g_loss: 3.3109
Epoch [ 50/ 50] | d_loss: 0.3801 | g_loss: 3.1624
Epoch [ 50/ 50] | d_loss: 0.4173 | g_loss: 2.4385
Epoch [ 50/ 50] | d_loss: 0.3816 | g_loss: 3.4204
Epoch [ 50/ 50] | d_loss: 0.3709 | g_loss: 4.8543
Epoch [ 50/ 50] | d_loss: 0.3558 | g_loss: 5.1553
```

2.8 Training loss

Plot the training losses for the generator and discriminator, recorded after each epoch.

```
In [52]: fig, ax = plt.subplots()
         losses = np.array(losses)
         plt.plot(losses.T[0], label='Discriminator', alpha=0.5)
         plt.plot(losses.T[1], label='Generator', alpha=0.5)
         plt.title("Training Losses")
         plt.legend()
```

```
Out[52]: <matplotlib.legend.Legend at 0x7f0a6f7914e0>
```



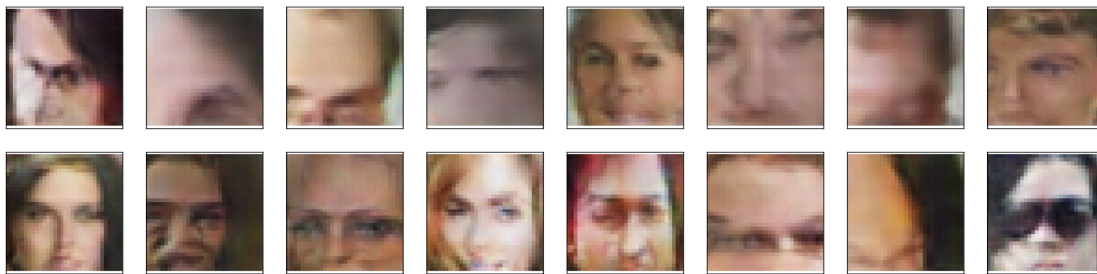
2.9 Generator samples from training

View samples of images from the generator, and answer a question about the strengths and weaknesses of your trained models.

```
In [53]: # helper function for viewing a list of passed in sample images
def view_samples(epoch, samples):
    fig, axes = plt.subplots(figsize=(16,4), nrows=2, ncols=8, sharey=True, sharex=True)
    for ax, img in zip(axes.flatten(), samples[epoch]):
        img = img.detach().cpu().numpy()
        img = np.transpose(img, (1, 2, 0))
        img = ((img + 1)*255 / (2)).astype(np.uint8)
        ax.xaxis.set_visible(False)
        ax.yaxis.set_visible(False)
        im = ax.imshow(img.reshape((32,32,3)))

In [54]: # Load samples from generator, taken while training
with open('train_samples.pkl', 'rb') as f:
    samples = pickle.load(f)

In [55]: _ = view_samples(-1, samples)
```



2.9.1 Question: What do you notice about your generated samples and how might you improve this model?

When you answer this question, consider the following factors: * The dataset is biased; it is made of "celebrity" faces that are mostly white * Model size; larger models have the opportunity to learn more features in a data feature space * Optimization strategy; optimizers and number of epochs affect your final result

Answer: The generated samples look like faces, but they don't look very realistic. I noticed that generator loss increases quite fast. I tried a lot of hyper parameter tuning: different learning rates, different beta values for Adam optimizer, more epochs, adding more layers, adding smoothing in real loss. The results can surely be improved by experimenting even more with hyper parameter settings or researching for GAN improvements in publications.

2.9.2 Submitting This Project

When submitting this project, make sure to run all the cells before saving the notebook. Save the notebook file as "dlnd_face_generation.ipynb" and save it as a HTML file under "File" -> "Download as". Include the "problem_unittests.py" files in your submission.

In []:

In []: