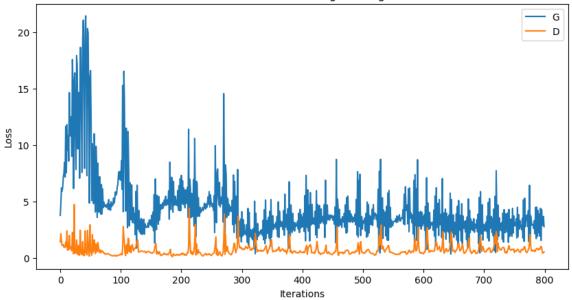
0. 수행 결과







1. 드라이브 마운트 및 파일 압축 해제

```
1 from google.colab import drive
2 drive.mount('/content/drive')
ive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
```

1 !tar -xvzf /content/drive/MyDrive/102flowers.tgz -C ./data

2. 필요 라이브러리 import 및 기초 설정

```
1 import numpy as np
2 import torch
3 import torch.nn as nn
4 import torchvision.utils as vutils
5 import torchvision.transforms as transforms
6 import torchvision.datasets as datasets
7 from torch.utils.data import DataLoader
8
9 import matplotlib.pyplot as plt

1 IMG_SIZE = 64
2 nc = 3
3 nz = 100
4 ngf = 64
5 ndf = 64
```

3. 데이터셋 로드 및 데이터로더 설정

4. 이미지 출력 과정을 간소화하는 함수 정의

```
1 def show_images(images, title):
       plt.figure(figsize=(8, 8))
 3
       plt.axis("off")
 4
       plt.title(title)
 5
       plt.imshow(
 6
           np.transpose(
 7
               vutils.make_grid(
 8
                   images[:64],
 9
                   padding=2,
10
                   normalize=True).cpu(),
11
               (1, 2, 0)
12
           ))
13
       plt.show()
14
15
16 sample_batch = next(iter(dataloader))
17 show_images(sample_batch[0], title="Training Images")
```

5. 생성자 정의

```
class Generator(nn.Module):
    def __init__(self):
        super().__init__()
        self.main = nn.Sequential(
            nn.ConvTranspose2d(nz, ngf*8, 4, 1, 0, bias=False),
            nn.BatchNorm2d(ngf*8),
            nn.ReLU(True),
            nn. ConvTranspose2d(ngf*8, ngf*4, 4, 2, 1, bias=False),\\
            nn.BatchNorm2d(ngf * 4),
            nn.ReLU(True),
            nn.ConvTranspose2d(ngf*4, ngf*2, 4, 2, 1, bias=False),
            nn.BatchNorm2d(ngf * 2),
            nn.ReLU(True),
            nn.ConvTranspose2d(ngf*2, ngf, 4, 2, 1, bias=False),
            nn.BatchNorm2d(ngf),
            nn.ReLU(True),
            nn.ConvTranspose2d(ngf, nc, 4, 2, 1, bias=False),
            nn.Tanh()
    def forward(self, x):
      return self.main(x)
```

6. 식별자 정의

```
class Discriminator(nn.Module):
    def __init__(self):
         super().__init__()
         self.main = nn.Sequential(
             nn. \texttt{Conv2d} (\texttt{nc}, \ \texttt{ndf}, \ \texttt{4}, \ \texttt{2}, \ \texttt{1}, \ \texttt{bias=False}) \,,
             nn.LeakyReLU(0.2, inplace=True),
             nn.Conv2d(ndf, ndf*2, 4, 2, 1, bias=False),
             nn.BatchNorm2d(ndf * 2),
             nn.LeakyReLU(0.2, inplace=True),
             nn.Conv2d(ndf*2, ndf*4, 4, 2, 1, bias=False),
             nn.BatchNorm2d(ndf * 4),
             nn.LeakyReLU(0.2, inplace=True),
             nn.Conv2d(ndf*4, ndf*8, 4, 2, 1, bias=False),
             nn.BatchNorm2d(ndf * 8),
             nn.LeakyReLU(0.2, inplace=True),
             nn.Conv2d(ndf*8, 1, 4, 1, 0, bias=False),
             nn.Sigmoid()
    def forward(self, x):
    return self.main(x)
```

7. 가중치 초기화 함수 정의 및 qpu 사용 설정

```
def weights_init(m):
    class_name = m.__class__.__name__
    if class_name.find("Conv") != -1:
        m.weight.data.normal_(0.0, 0.02)
    elif class_name.find("BatchNorm") != -1:
        m.weight.data.normal_(1.0, 0.02)
        m.bias.data.fill_(0)

device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
    print(f"Using device: {device}")
```

8. 모델 gpu로 이동 및 모델 확인

```
G = Generator().to(device)
G.apply(weights_init)
print(G)

D = Discriminator().to(device)
D.apply(weights_init)
print(D)
```

하이퍼 파라미터 설정 및 학습 기초 설정

```
criterion = nn.BCELoss()
d_optimizer = torch.optim.Adam(D.parameters(), lr=0.0002, betas=(0.5, 0.999))
g_optimizer = torch.optim.Adam(G.parameters(), lr=0.0002, betas=(0.5, 0.999))

G_losses = []
D_losses = []
fixed_noise = torch.randn(64, nz, 1, 1).to(device)
```

9. 모델 학습

```
NUM_EPOCHS = 50
for epoch in range(NUM_EPOCHS):
    d_{loss} = 0
    g_loss = 0
    for real_images, _ in dataloader:
       real_images = real_images.to(device)
        D.zero_grad()
        b_size = real_images.size(0)
        real_labels = torch.ones((b_size,)).to(device)
        output = D(real_images).view(-1)
        d_loss_real = criterion(output, real_labels)
        z = torch.randn(b_size, nz, 1, 1).to(device)
        fake_labels = torch.zeros((b_size,)).to(device)
        fake_images = G(z)
        output = D(fake_images.detach()).view(-1)
        d_loss_fake = criterion(output, fake_labels)
        d_loss = d_loss_real + d_loss_fake
        d_loss.backward()
        d_optimizer.step()
        G.zero_grad()
        output = D(fake_images).view(-1)
        g_loss = criterion(output, real_labels)
        g_loss.backward()
        g_optimizer.step()
        G_losses.append(g_loss.item())
        D_losses.append(d_loss.item())
    print(f"[{epoch + 1}/{NUM_EPOCHS}], d_loss: {d_loss.item():.4f}, g_loss: {g_loss.item():.4f}")
    if (epoch + 1) % 5 == 0 or epoch == 0:
        with torch.no_grad():
            generated_images = G(fixed_noise).detach().cpu()
        show_images(generated_images, title=f"202401833 - Epoch: {epoch + 1}")
```

10. 학습 과정에서의 손실 출력

```
1 plt.figure(figsize=(10, 5))
2 plt.title("G and D Loss During Training")
3 plt.plot(G_losses, label="G")
4 plt.plot(D_losses, label="D")
5 plt.xlabel("Iterations")
6 plt.ylabel("Loss")
7 plt.legend()
8 plt.show()
```