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Abstract

This project is a PyTorch implementation of Variational AutoEncoder (VAE) and Generative Adversarial Network (GAN)

We train the model with a CustomDataset and analyze the differences and see the results.

Introduction

VAE and GAN are both Generative models.

But they have different approaches to generate new data.

However, the difference is that GAN is generally harder to learn than VAE because of the properties below.

- 1. GAN is a zero-sum game with adversarial process. Especially, if the discriminator is too strong, the generator can't learn anything.
- 2. Hard to find Nash equilibrium. The Nash equilibrium is the point where the discriminator can't distinguish the real data and the fake data. But it is hard to find the point because the discriminator and the generator are trained alternately.

To solve the problems above

- 1. Pretrain GAN with VAE and give noise to the discriminator.
- Use Least Squares error (LSGAN) loss function instead of Binary Cross Entropy (BCE) loss function.
- 3. Use label flipping and label noise.
- 4. Use unbalanced layer.

5. Use huge epoch.

Dataset

The model is trained on the CelebA dataset. with using CustomDataset class.

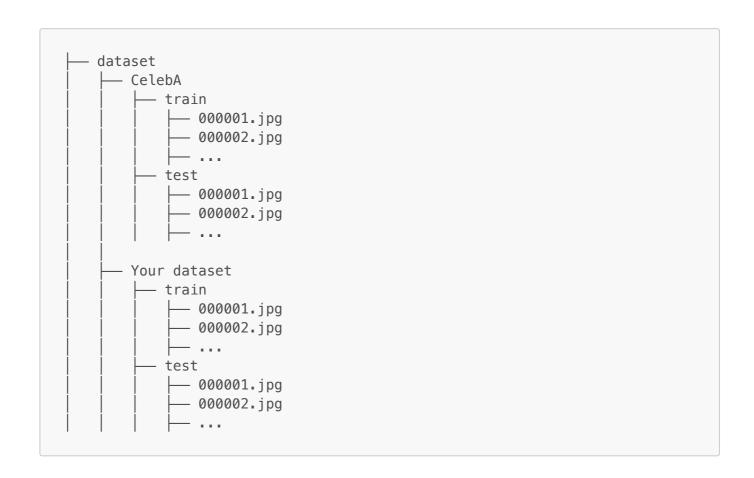
The dataset contains 202,599 face images of various celebrities.

The images are cropped and resized to 64 square images.

There is a CustomDataset class to load your own dataset.

You can set your own dataset by setting the root variable.

The dataset should be organized as below.



VAE

Introduction VAE

This is a PyTorch implementation of Variational AutoEncoder (VAE) based on the paper Auto-Encoding Variational Bayes.

Used hyperparameters VAE

• Batch size: 128

Learning rate: 0.005Optimizer: Adam

• Number of epochs: 5

KL divergence weight: 0.00025Latent space dimension: 128

• Loss function: Mean Squared Error (MSE)

• Activation function: Leaky ReLU

• Image size: 64 square

Model architecture VAE

• Unit test with torchsummary

 Param #	Output Shapa	Layor (type)
# Paralli =======	Output Shape 	Layer (type)
896	[-1, 32, 32, 32]	Conv2d-1
64	[-1, 32, 32, 32]	BatchNorm2d-2
0	[-1, 32, 32, 32]	LeakyReLU-3
18,496	[-1, 64, 16, 16]	Conv2d-4
128	[-1, 64, 16, 16]	BatchNorm2d-5
0	[-1, 64, 16, 16]	LeakyReLU-6
73,856	[-1, 128, 8, 8]	Conv2d-7
256	[-1, 128, 8, 8]	BatchNorm2d-8
0	[-1, 128, 8, 8]	LeakyReLU-9
295,168	[-1, 256, 4, 4]	Conv2d-10
512	[-1, 256, 4, 4]	BatchNorm2d-11
0	[-1, 256, 4, 4]	LeakyReLU-12
1,180,160	[-1, 512, 2, 2]	Conv2d-13
1,024	[-1, 512, 2, 2]	BatchNorm2d-14
0	[-1, 512, 2, 2]	LeakyReLU-15
262,272	[-1, 128]	Linear-16
262,272	[-1, 128]	Linear-17
264,192	[-1, 2048]	Linear-18
1,179,904	[-1, 256, 4, 4]	ConvTranspose2d-19
512	[-1, 256, 4, 4]	BatchNorm2d-20
0	[-1, 256, 4, 4]	LeakyReLU-21
295,040	[-1, 128, 8, 8]	ConvTranspose2d-22
256	[-1, 128, 8, 8]	BatchNorm2d-23
0	[-1, 128, 8, 8]	LeakyReLU-24
73,792	[-1, 64, 16, 16]	ConvTranspose2d-25
128	[-1, 64, 16, 16]	BatchNorm2d-26
0	[-1, 64, 16, 16]	LeakyReLU-27
18,464	[-1, 32, 32, 32]	ConvTranspose2d-28
64	[-1, 32, 32, 32]	BatchNorm2d-29
0	[-1, 32, 32, 32]	LeakyReLU-30
867	[-1, 3, 64, 64]	ConvTranspose2d-31
6	[-1, 3, 64, 64]	BatchNorm2d-32
0	[-1, 3, 64, 64]	Tanh-33

Total params: 3,928,329 Trainable params: 3,928,329 Non-trainable params: 0

Input size (MB): 0.05

Forward/backward pass size (MB): 3.16

Params size (MB): 14.99

Estimated Total Size (MB): 18.19

GAN

Introduction GAN

This is a PyTorch implementation of Generative Adversarial Network (GAN) based on the paper Generative Adversarial Networks.

It is Deep Convolutional GAN (DCGAN) with Least Squares error (LSGAN) loss function based on the paper Least Squares Generative Adversarial Networks.

It has unbanlanced layer and is pre-trained with VAE based on the paper Unbalanced GANs.

Used hyperparameters GAN

• Batch size: 128

• Learning rate: 0.005

• Optimizer: Adam

• Number of epochs: 5

• KL divergence weight in pre-training: 0.00025

• Latent space dimension: 128

• Loss function: Mean Squared Error (MSE)

Activation function: Leaky ReLU

• Image size: 64 square

• Label flipping step: 16

• Label noise with Gaussian distribution: 0.1

Model architecture GAN

- Unit test with torchsummary
- Generator

Layer (type)	Output	Shape	Param #
 Linear–1	[-1,	2048]	264 , 192
ConvTranspose2d-2	[-1, 256,	4, 4]	1,179,904
BatchNorm2d-3	[-1, 256,	4, 4]	512
LeakyReLU-4	[-1, 256,	4, 4]	0
ConvTranspose2d-5	[-1, 128,	8, 8]	295,040
BatchNorm2d-6	[-1, 128,	8, 8]	256
LeakyReLU-7	[-1, 128,	8, 8]	0

ConvTranspose2d-8	[-1, 64, 16, 16] 73,792
BatchNorm2d-9	[-1, 64, 16, 16] 128
LeakyReLU-10	[-1, 64, 16, 16]
ConvTranspose2d-11	[-1, 32, 32, 32] 18,464
BatchNorm2d-12	[-1, 32, 32, 32] 64
LeakyReLU-13	[-1, 32, 32, 32] 0
ConvTranspose2d-14	[-1, 32, 64, 64] 9,248
BatchNorm2d-15	[-1, 32, 64, 64] 64
LeakyReLU-16	[-1, 32, 64, 64]
Conv2d-17	[-1, 3, 64, 64] 867
Tanh-18	[-1, 3, 64, 64] 0

Total params: 1,842,531 Trainable params: 1,842,531 Non-trainable params: 0

Input size (MB): 0.00

Forward/backward pass size (MB): 4.61

Params size (MB): 7.03

Estimated Total Size (MB): 11.64

Discriminator

Param #	Output Shape	Layer (type)
=========== 896	[-1, 32, 32, 32]	======================================
64	[-1, 32, 32, 32]	BatchNorm2d-2
0	[-1, 32, 32, 32]	LeakyReLU-3
18,496	[-1, 64, 16, 16]	Conv2d-4
128	[-1, 64, 16, 16]	BatchNorm2d-5
0	[-1, 64, 16, 16]	LeakyReLU-6
73,856	[-1, 128, 8, 8]	Conv2d-7
256	[-1, 128, 8, 8]	BatchNorm2d-8
0	[-1, 128, 8, 8]	LeakyReLU-9
295,168	[-1, 256, 4, 4]	Conv2d-10
512	[-1, 256, 4, 4]	BatchNorm2d-11
0	[-1, 256, 4, 4]	LeakyReLU-12
524,416	[-1, 128]	Linear-13
0	[-1, 128]	LeakyReLU-14
129	[-1, 1]	Linear-15

Total params: 913,921 Trainable params: 913,921 Non-trainable params: 0

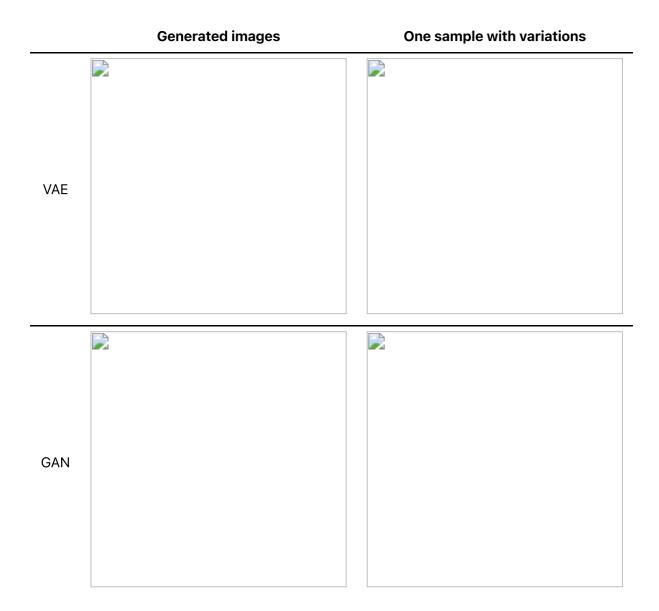
Input size (MB): 0.05

Forward/backward pass size (MB): 1.41

Params size (MB): 3.49

Estimated Total Size (MB): 4.94

Results



Contact

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