Vanilla GAN(Generative Adversarial Network)

```
import warnings
warnings.filterwarnings('ignore')
```

▼ Import Packages

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

import keras
from keras.layers import Dense, Dropout, Input
from keras.models import Model,Sequential
from keras.layers.advanced_activations import LeakyReLU
```

I. Load MNIST Dataset

• 'generator'의 'tanh' Activation 출력에 적합하도록 정규화

```
from keras.datasets import mnist

(X_train, y_train), (X_test, y_test) = mnist.load_data()

# Normalization
X_train = X_train.astype(np.float32) / 127.5 - 1

# Reshape
X_train = X_train.reshape(60000, 784)
```

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11493376/11490434 [==========] - Os Ous/step

⋆ II. 'adam' Optimizer

beta_1: 감쇠율 조정

```
from keras.optimizers import Adam

adam = Adam(Ir = 0.0002, beta_1 = 0.5)
```

→ III. 'generator' Model

- 랜덤 벡터(잠재공간의 랜덤 포인트)를 입력받아 이미지 생성
 - NOISE_DIM: 입력 랜덤 벡터 크기
- 'discriminator'를 속이도록 학습

```
NOISE_DIM = 10

generator = Sequential(name = 'generator')

generator.add(Dense(256, input_shape = (NOISE_DIM,)))

generator.add(LeakyReLU())

generator.add(Dense(512))

generator.add(LeakyReLU())

generator.add(Dense(1024))

generator.add(LeakyReLU())

generator.add(LeakyReLU())

generator.add(Dense(784, activation = 'tanh'))
```

Model Summary

generator.summary()

Model: "generator"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 256)	2816
leaky_re_lu (LeakyReLU)	(None, 256)	0
dense_1 (Dense)	(None, 512)	131584
leaky_re_lu_1 (LeakyReLU)	(None, 512)	0
dense_2 (Dense)	(None, 1024)	525312
leaky_re_lu_2 (LeakyReLU)	(None, 1024)	0
dense_3 (Dense)	(None, 784)	803600

Total params: 1,463,312 Trainable params: 1,463,312 Non-trainable params: 0

▼ IV. 'discriminator' Model

- 이미지를 입력받아 'Real Image'인지 'generator'가 생성한 'Fake Image' 인지 판별
 - 。 이진분류

Model Summary

discriminator.summary()

Model: "discriminator"

Layer (type)	Output Shape	Param #
dense_4 (Dense)	(None, 1024)	803840
leaky_re_lu_3 (LeakyReLU)	(None, 1024)	0
dropout (Dropout)	(None, 1024)	0
dense_5 (Dense)	(None, 512)	524800
leaky_re_lu_4 (LeakyReLU)	(None, 512)	0
dropout_1 (Dropout)	(None, 512)	0
dense_6 (Dense)	(None, 256)	131328
leaky_re_lu_5 (LeakyReLU)	(None, 256)	0
dropout_2 (Dropout)	(None, 256)	0
dense_7 (Dense)	(None, 1)	257

Total params: 1,460,225 Trainable params: 1,460,225 Non-trainable params: 0

→ 1) 'discriminator' Compile

• 학습 설정

→ V. 'gan' Model

▼ 1) 'generator', 'discriminator' 연결

- 'gan' 모델에서 'generator'만 학습하도록 설정
 - o discriminator.trainable = False

```
discriminator.trainable = False

gan_input = Input(shape = (NOISE_DIM,))
x = generator(gan_input)
output = discriminator(x)
```

→ 2) 'gan' Model

```
gan = Model(gan_input, output, name = 'gan')
```

→ 3) 'gan' Summary

gan.summary()

Model: "gan"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 10)]	0
generator (Sequential)	(None, 784)	1463312
discriminator (Sequential)	(None, 1)	1460225

Total params: 2,923,537 Trainable params: 1,463,312 Non-trainable params: 1,460,225

→ 4) 'gan' Compile

• 학습 설정

VI. Define 'get_batches()' Function

• MNIST image batch 생성

```
def get_batches(data, batch_size):
   batches = []

for i in range(data.shape[0] // batch_size):
   batch = data[i * batch_size : (i + 1) * batch_size]
   batches.append(batch)
   return np.asarray(batches)
```

VII. 'visualize_training()' Function

```
def visualize_training(epoch, d_losses, g_losses):
    # 오차 시각화
    # plt.figure(figsize=(8, 4))
    # plt.plot(d_losses, label='Discriminator Loss')
    # plt.plot(g_losses, label='Generatror Loss')
    # plt.xlabel('Epoch')
    # plt.ylabel('Loss')
    # plt.legend()
    # plt.show()
    # print('epoch: {}, Discriminator Loss: {}, Generator Loss: {}'.format(epoch, np.asarray(d_loss)
    # 이미지 생성 결과 시각화
    print('epoch :', epoch)
    noise = np.random.normal(0, 1, size = (24, NOISE_DIM))
    generated_images = generator.predict(noise)
    generated_images = generated_images.reshape(-1, 28, 28)
    plt.figure(figsize = (8, 4))
    for i in range(generated_images.shape[0]):
        plt.subplot(4, 6, i + 1)
        plt.imshow(generated_images[i], interpolation = 'nearest', cmap = 'Greys_r')
        plt.axis('off')
    plt.tight_layout()
    plt.show()
```

VIII. Model Training

- 약 35분
- .fit()
 - 'epoch', 'batch_size' 지정
- .train_on_batch()
 - 전달 받은 모든 데이터를 사용하여 학습 진행
- 'generator'가 매번 새로운 'Fake Image'를 생성하여 '.train_on_batch()' 사용

```
%%time
EPOCHS = 100
BATCH_SIZE = 128
# 'discriminator', 'gan' Loss 저장 List
d_{losses} = []
g_losses = []
for epoch in range(1, EPOCHS + 1):
    # batch 별 학습
    for real_images in get_batches(X_train, BATCH_SIZE):
       # Random Noise 생성
       input_noise = np.random.uniform(-1, 1, size = [BATCH_SIZE, NOISE_DIM])
       # Fake Image 데이터 생성
       generated_images = generator.predict(input_noise)
       # 'gan' 학습용 X 데이터 정의
       x_dis = np.concatenate([real_images, generated_images])
       # 'gan' 학습용 y 데이터 정의
       y_dis = np.zeros(2 * BATCH_SIZE)
       y_dis[:BATCH_SIZE] = 1
       # 'discriminator' 학습
       discriminator.trainable = True
       d_loss = discriminator.train_on_batch(x_dis, y_dis)
       # 'gan' 학습
       noise = np.random.uniform(-1, 1, size = [BATCH_SIZE, NOISE_DIM])
       y_gan = np.ones(BATCH_SIZE)
       # 'discriminator' 학습 정지
       discriminator.trainable = False
       g_loss = gan.train_on_batch(noise, y_gan)
    d_losses.append(d_loss)
    g_losses.append(g_loss)
    # 생성 결과 시각화
    if epoch == 1 or epoch % 5 == 0:
```

visualize_training(epoch, d_losses, g_losses)

#

#

#

The End

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#

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✓ 32분 12초 오전 9:33에 완료됨

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