



GAN **with** **Pokemon**

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Classification vs GAN

일반적인 classification

- : 대상을 판별하는 판별망 (Discriminative Network)
- > 입력데이터 x 에 대하여 y 가 될 조건부 확률을 구하는 것

(ex) 입력 데이터 : 고양이사진
-> 예측 : "고양이"

GAN (Generative Adversarial Networks)

- : 특정 확률분포를 갖는 데이터로 학습을 시키면,
이것과 유사한 분포를 갖는 데이터를 생성하는 것
- > label이 없더라도 스스로 데이터 속 **중요한 정보**를 찾아
새로운 데이터를 만들어냄

GAN 이란?

생성망 (Generator)

: **Discriminator**를 속일 수 있을 정도로 진짜 데이터와 비슷한 가짜 데이터를 만들어 냄

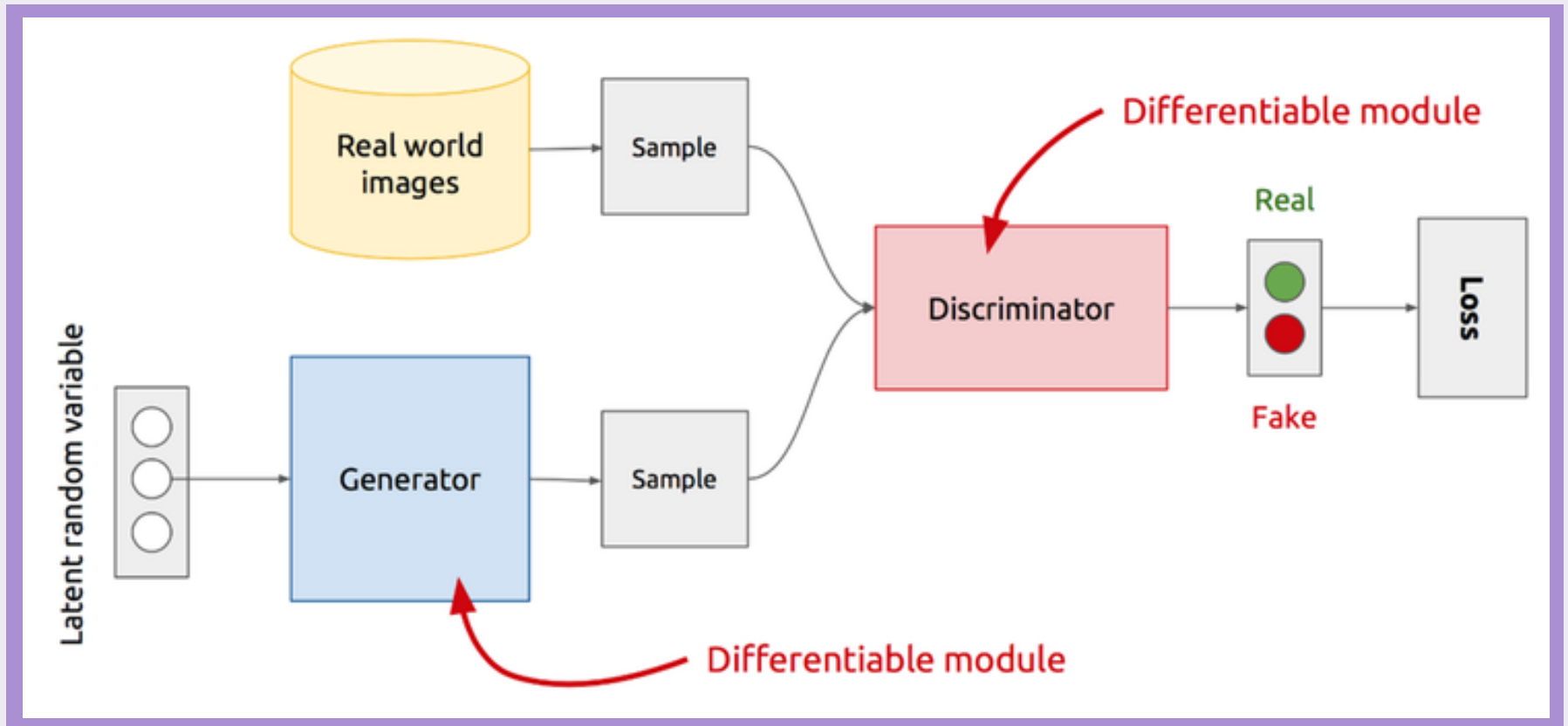
판별망 (Discriminator)

: 실제 학습 데이터와 **Generator**를 거쳐 만들어진 가짜 데이터를 이용하여 학습
-> 데이터가 진짜인지 가짜인지 구별하는 역할



Discriminator는 판별을 잘하는 방향으로,
Generator는 **Discriminator**를 잘 속이는 방향으로 학습됨

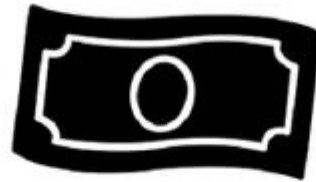
GAN 이란?



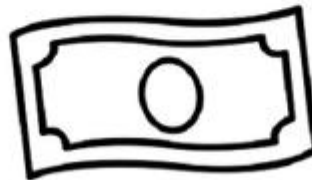
GAN 이란?



Generator



Fake data



Real data



Discriminator

위조 지폐를 만들어 내는 사기꾼과
위조 지폐 여부를 판별하는 형사 ?

GAN 학습

1. Generator 고정 -> Discriminator 학습
 2. Discriminator 고정 -> Generator 학습
- 위 과정을 반복하면 평형 상태에 도달

상호 협조적이지 않는 상대가
서로 최적에 도달하려고 노력 -> **내쉬 평형** 상태

어떤 기준으로 최적에 도달했는지
명확하게 판단할 근거가 부족
-> **사람의 개입**이 필요

Data Exploration

Import Packages

```
import numpy as np
import matplotlib.pyplot as plt
import os
from PIL import Image
import imageio

import tensorflow as tf
from keras import layers
from keras.datasets import mnist
from keras.models import Sequential, Model, load_model
from keras.optimizers import Adam
```

Data Exploration

Make Dataset

```
def load_and_preprocessing(dir):  
    data = []  
    img_list = os.listdir(dir)  
    for name in img_list :  
  
        png = imageio.imread(dir + name)  
        png = Image.fromarray(png)  
        png.load() # for splitting  
  
        # convert RGBA to RGB -> alpha channel  
        if(len(png.split()) == 4):  
            img = Image.new('RGB', png.size, (255, 255, 255)) # white  
            img.paste(png, mask = png.split()[3])  
        else:  
            img = png  
  
        images = tf.keras.preprocessing.image.img_to_array(img)  
        images /= 255. # preprocessing  
  
        data.append(images)  
  
    return np.stack(data)
```


Data Exploration

Make Dataset

```
dir = "images/"
img_list = os.listdir(dir)
img_len = len(os.listdir(dir))

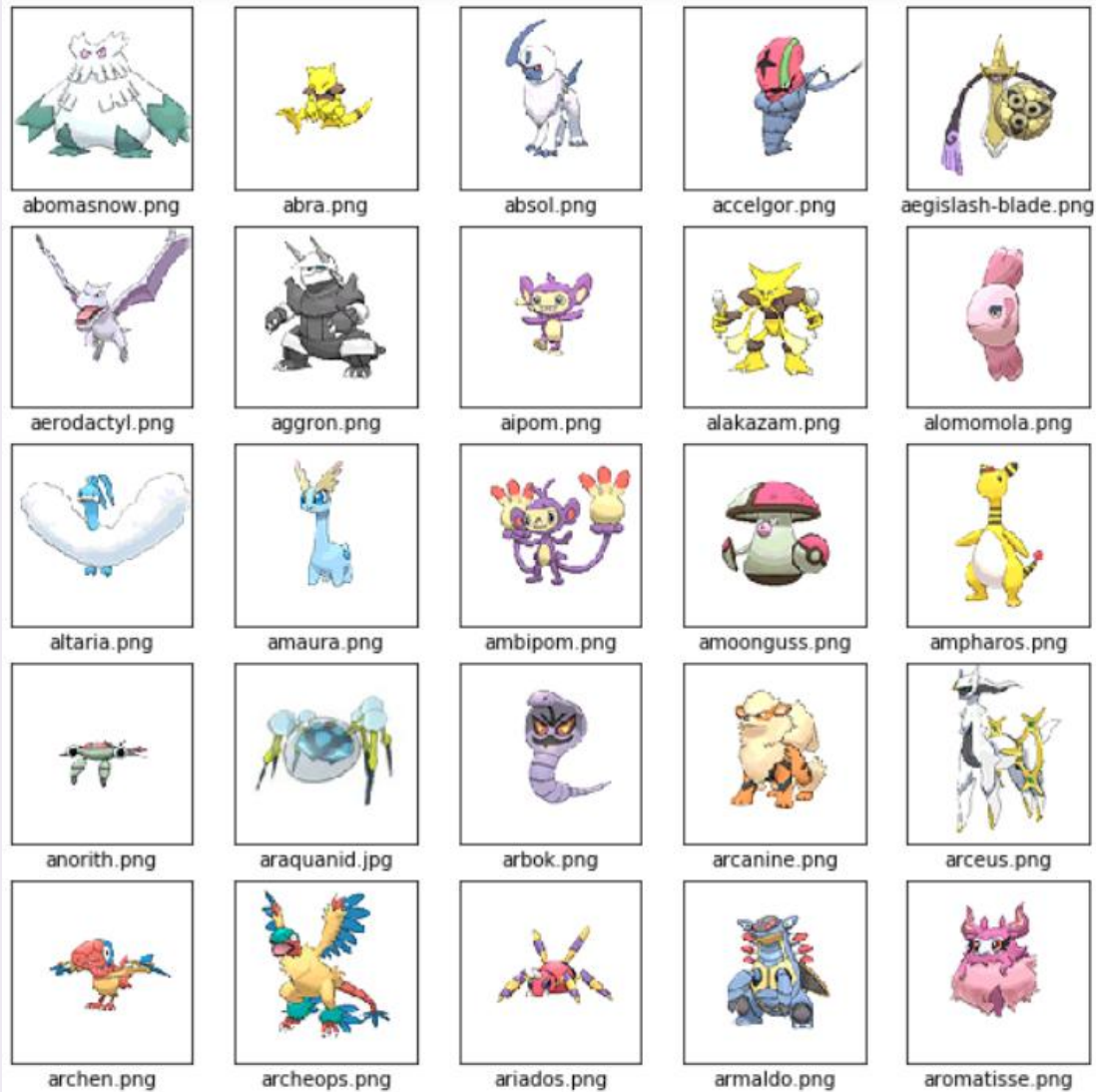
print("The number of images :",img_len)
print(img_list[0:10])
```

```
The number of images : 809
['abomasnow.png', 'abra.png', 'absol.png', 'accelgor.png',
m.png', 'alomomola.png']
```

```
# Make dataset
dataset = load_and_preprocessing(dir)
print("Shape of dataset :", dataset.shape)
```

```
Shape of dataset : (809, 120, 120, 3)
```

Data Exploration



Create Generator

Generator

```
# params
latent_dim = 100
height = 120
width = 120
channels = 3

generator_input = layers.Input(shape=(latent_dim,))
g = layers.Dense(128 * 15 * 15)(generator_input)
g = layers.Reshape((15, 15, 128))(g)

g = layers.Conv2DTranspose(128, 3, strides=2, padding='same')(g)
g = layers.BatchNormalization(momentum=0.8)(g)
g = layers.ReLU()(g)

g = layers.Conv2DTranspose(128, 3, strides=2, padding='same')(g)
g = layers.BatchNormalization(momentum=0.8)(g)
g = layers.ReLU()(g)

g = layers.Conv2DTranspose(64, 3, strides=2, padding='same')(g)
g = layers.BatchNormalization(momentum=0.8)(g)
g = layers.ReLU()(g)

g = layers.Conv2D(channels, 3, activation='tanh', padding='same')(g)

generator = Model(generator_input, g)
generator.summary()
```

Create Generator

Model: "model_21"

| Layer (type) | Output Shape | Param # |
|------------------------------|----------------------|---------|
| input_26 (InputLayer) | (None, 100) | 0 |
| dense_18 (Dense) | (None, 28800) | 2908800 |
| reshape_12 (Reshape) | (None, 15, 15, 128) | 0 |
| conv2d_transpose_24 (Conv2DT | (None, 30, 30, 128) | 147584 |
| batch_normalization_39 (Batc | (None, 30, 30, 128) | 512 |
| re_lu_21 (ReLU) | (None, 30, 30, 128) | 0 |
| conv2d_transpose_25 (Conv2DT | (None, 60, 60, 128) | 147584 |
| batch_normalization_40 (Batc | (None, 60, 60, 128) | 512 |
| re_lu_22 (ReLU) | (None, 60, 60, 128) | 0 |
| conv2d_transpose_26 (Conv2DT | (None, 120, 120, 64) | 73792 |
| batch_normalization_41 (Batc | (None, 120, 120, 64) | 256 |
| re_lu_23 (ReLU) | (None, 120, 120, 64) | 0 |
| conv2d_50 (Conv2D) | (None, 120, 120, 3) | 1731 |

Total params: 3,280,771

Trainable params: 3,280,131

Non-trainable params: 640

Create Discriminator

Discriminator

```
discriminator_input = layers.Input(shape=(height, width, channels))

d = layers.Conv2D(128, 3, strides=2, padding='same')(discriminator_input)
d = layers.LeakyReLU(alpha=0.2)(d)

d = layers.Conv2D(128, 3, strides=2, padding='same')(d)
d = layers.BatchNormalization(momentum=0.8)(d)
d = layers.LeakyReLU(alpha=0.2)(d)

d = layers.Conv2D(64, 3, strides=2, padding='same')(d)
d = layers.BatchNormalization(momentum=0.8)(d)
d = layers.LeakyReLU()(d)

d = layers.Conv2D(64, 3, strides=2, padding='same')(d)
d = layers.BatchNormalization(momentum=0.8)(d)
d = layers.LeakyReLU()(d)

d = layers.Flatten()(d)
d = layers.Dense(1, activation='sigmoid')(d)

discriminator = Model(discriminator_input, d)
discriminator_optimizer = Adam(lr=0.0002, beta_1=0.5, clipvalue=1.0)
discriminator.compile(optimizer=discriminator_optimizer, loss='binary_crossentropy', metrics=['accuracy'])
discriminator.summary()
```

Create Discriminator

Model: "model_26"

| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| ===== | | |
| input_31 (InputLayer) | (None, 120, 120, 3) | 0 |
| conv2d_59 (Conv2D) | (None, 60, 60, 128) | 3584 |
| leaky_re_lu_33 (LeakyReLU) | (None, 60, 60, 128) | 0 |
| conv2d_60 (Conv2D) | (None, 30, 30, 128) | 147584 |
| batch_normalization_48 (Batch Normalization) | (None, 30, 30, 128) | 512 |
| leaky_re_lu_34 (LeakyReLU) | (None, 30, 30, 128) | 0 |
| conv2d_61 (Conv2D) | (None, 15, 15, 64) | 73792 |
| batch_normalization_49 (Batch Normalization) | (None, 15, 15, 64) | 256 |
| leaky_re_lu_35 (LeakyReLU) | (None, 15, 15, 64) | 0 |
| conv2d_62 (Conv2D) | (None, 8, 8, 64) | 36928 |
| batch_normalization_50 (Batch Normalization) | (None, 8, 8, 64) | 256 |
| leaky_re_lu_36 (LeakyReLU) | (None, 8, 8, 64) | 0 |
| flatten_9 (Flatten) | (None, 4096) | 0 |
| dense_21 (Dense) | (None, 1) | 4097 |
| ===== | | |

Total params: 267,009

Trainable params: 266,497

Non-trainable params: 512

GAN - Training

GAN

```
gan_input = layers.Input(shape=(latent_dim,))
discriminator.trainable = False

gan_output = discriminator(generator(gan_input))
gan = Model(gan_input, gan_output)

gan_optimizer = Adam(lr=0.0002, beta_1=0.5, clipvalue=1.0)
gan.compile(optimizer = gan_optimizer, loss = 'binary_crossentropy', metrics=['accuracy'])
```

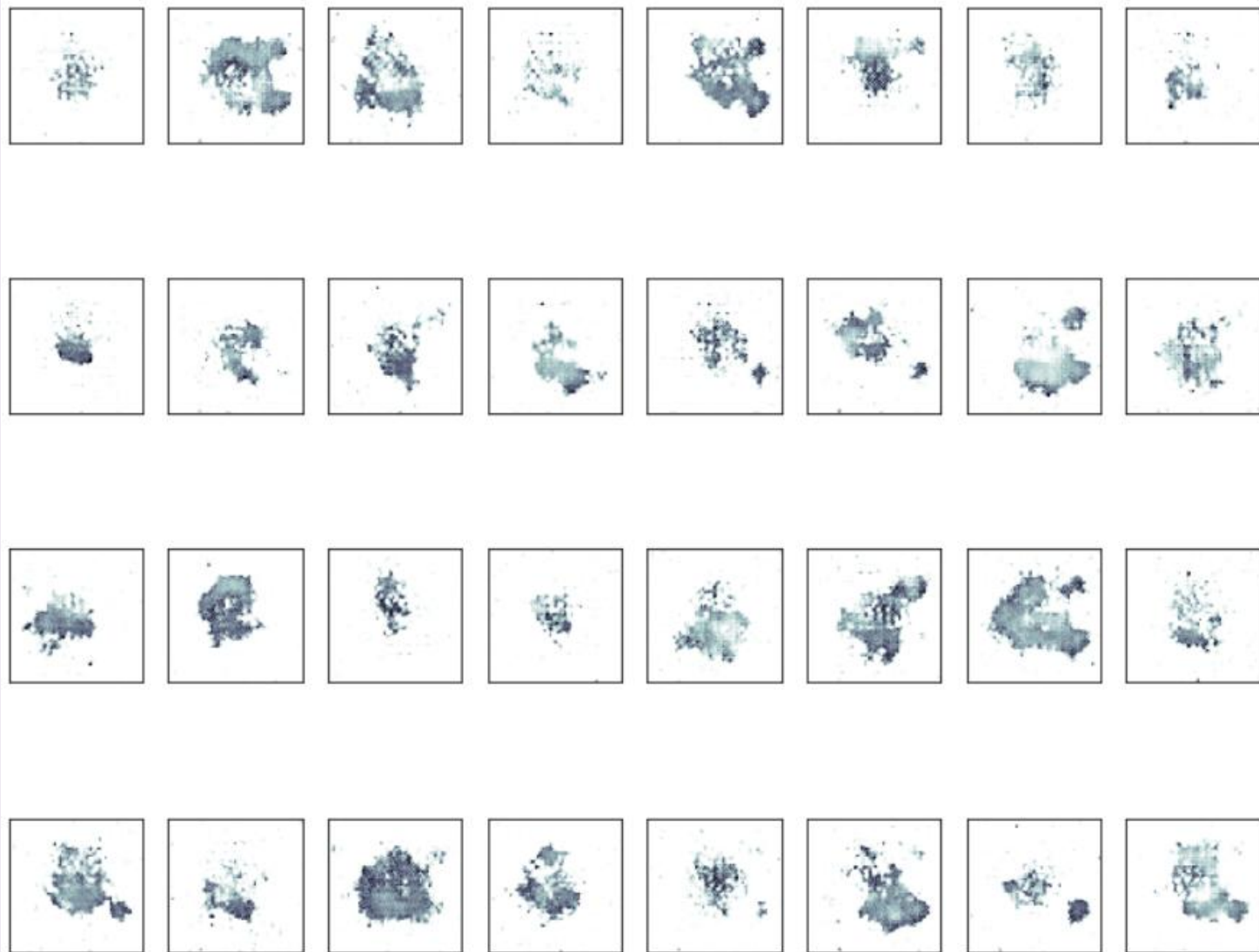
```
%%time
hist_1000 = train(1000, 1)
```

| | | | | |
|----|---------------------------------|--------|-----------------|--------|
| 55 | iteration - discriminator loss: | 6.724, | generator loss: | 10.687 |
| 56 | iteration - discriminator loss: | 5.479, | generator loss: | 4.935 |
| 57 | iteration - discriminator loss: | 1.801, | generator loss: | 1.393 |
| 58 | iteration - discriminator loss: | 0.806, | generator loss: | 2.114 |
| 59 | iteration - discriminator loss: | 0.177, | generator loss: | 3.102 |
| 60 | iteration - discriminator loss: | 0.162, | generator loss: | 2.834 |
| 61 | iteration - discriminator loss: | 0.224, | generator loss: | 2.583 |
| 62 | iteration - discriminator loss: | 0.167, | generator loss: | 2.694 |
| 63 | iteration - discriminator loss: | 0.148, | generator loss: | 2.271 |
| 64 | iteration - discriminator loss: | 0.294, | generator loss: | 4.602 |
| 65 | iteration - discriminator loss: | 1.343, | generator loss: | 6.806 |
| 66 | iteration - discriminator loss: | 1.140, | generator loss: | 1.308 |
| 67 | iteration - discriminator loss: | 0.031, | generator loss: | 2.293 |
| 68 | iteration - discriminator loss: | 0.117, | generator loss: | 3.423 |

Visualization – 100 iterations



Visualization – 750 iterations



Visualization – 1500 iterations



Dataset

[1] Pokemon Image Dataset,
<https://www.kaggle.com/vishalsubbiah/pokemon-images-and-types>

Reference

[1] 라온피플 ML Academy,
<https://laonple.blog.me/221190581073>