

Style transfer using cycleGAN

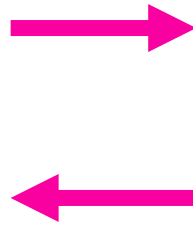
Hyukhun koh, Junbeom Kwon



Objective : Image Translation



Style A : without glasses



Style B : with glasses

Why we need this Application?

- Recommend the best glasses for the user
- Applied to AI applications (included in camera application)
- Used to another domain that needs style transfer

Datasets

1. Kaggle Datasets (unpaired & unclassified datasets)

- baseline model
- glass(2144)/ no-glasses(2144)
- resolution : 1024 * 1024
- filtered out confusing data
- Train(2139 each)/Test(5 each)

2. AI Hub

3. Google Image Search (by crawling)

4. celebA



Kaggle



AI Hub

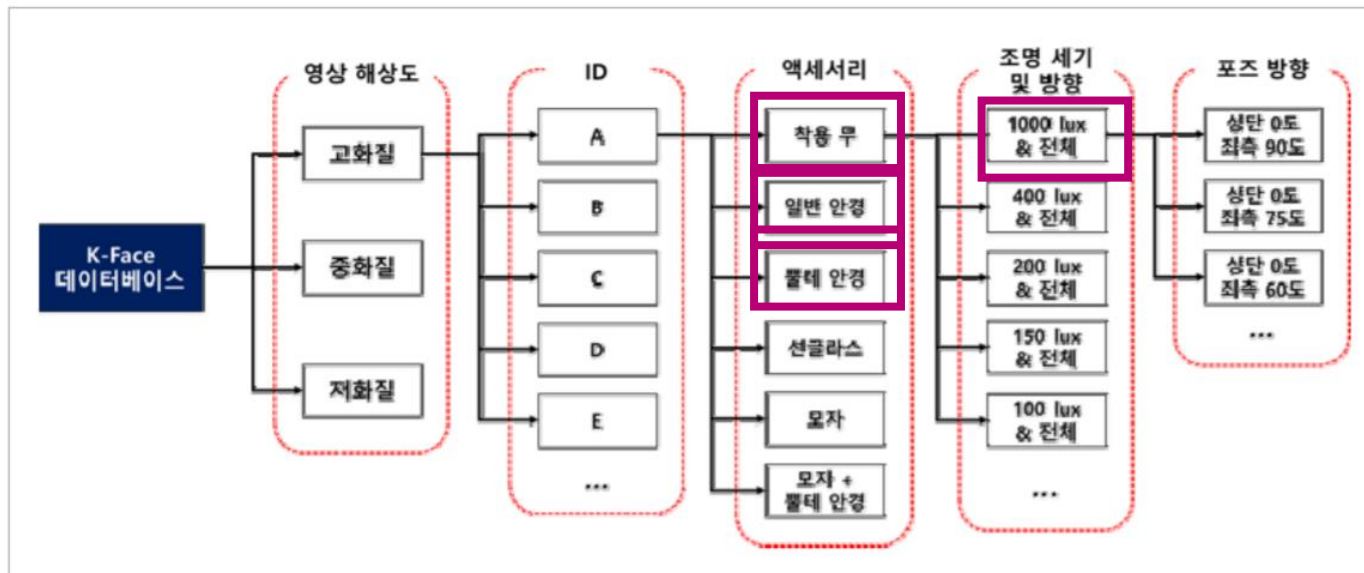


그림6 | 한국인 안면 이미지 데이터셋 구조



Google Image Search

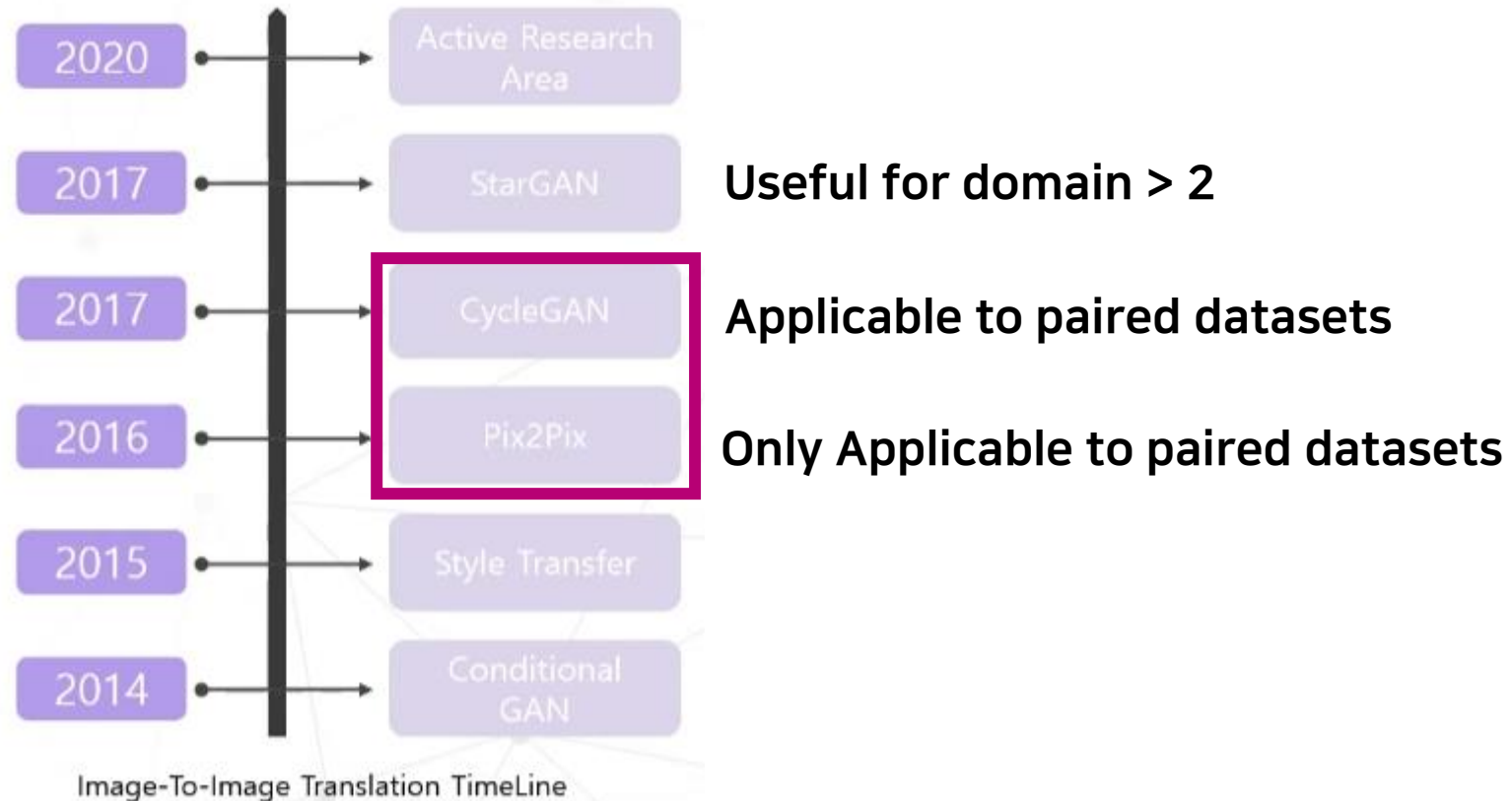
```
1 from selenium import webdriver
2 from selenium.webdriver.common.keys import Keys
3 import time
4 import urllib.request
5
6 driver = webdriver.Chrome()
7 driver.get("https://www.google.co.kr/imghp?hl=ko&ogbl")
8 elem = driver.find_element_by_name("q")
9 elem.send_keys("여예미 완결체")
10 elem.send_keys(Keys.RETURN)
11
12 SCROLL_PAUSE_TIME = 1
13
14 last_height = driver.execute_script("return document.body.scrollHeight")
15
16 while True:
17     driver.execute_script("window.scrollTo(0, document.body.scrollHeight);")
18     time.sleep(SCROLL_PAUSE_TIME)
19
20     new_height = driver.execute_script("return document.body.scrollHeight")
21     if new_height == last_height:
22         try:
23             driver.find_element_by_css_selector(".mye4qd").click()
24         except:
25             break
26     last_height = new_height
27     print(new_height)
28 images = driver.find_elements_by_css_selector(".rg_i.04LuWd")
```



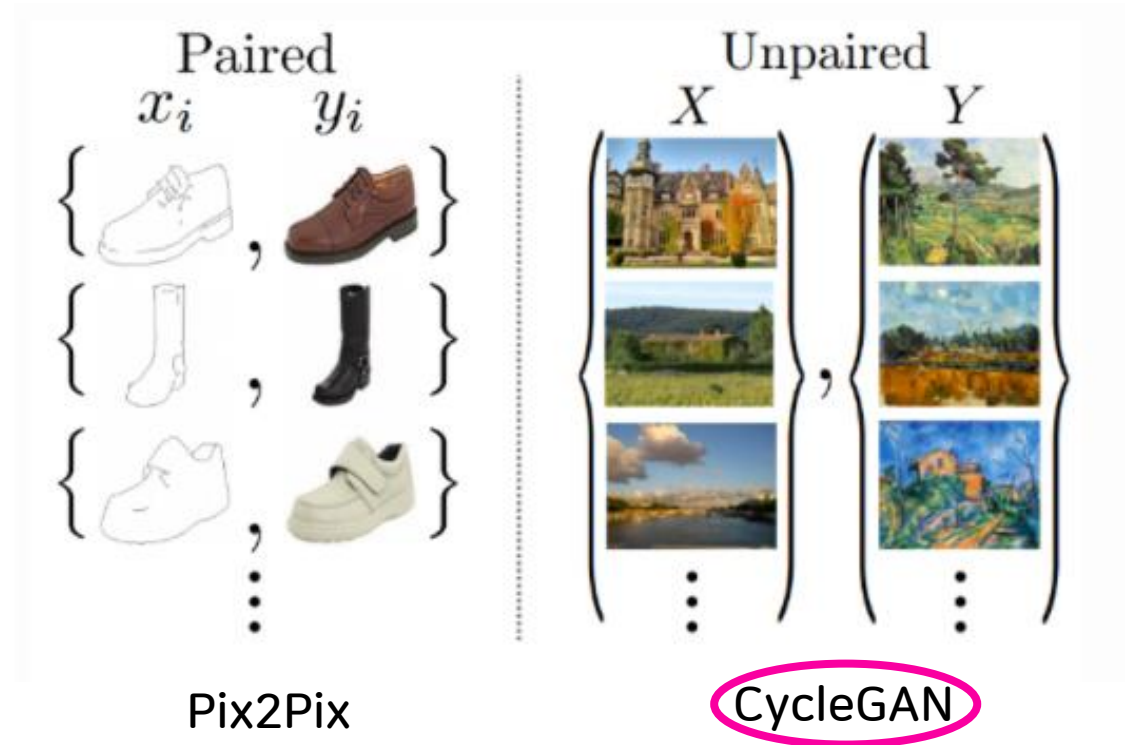
CelebA



Model

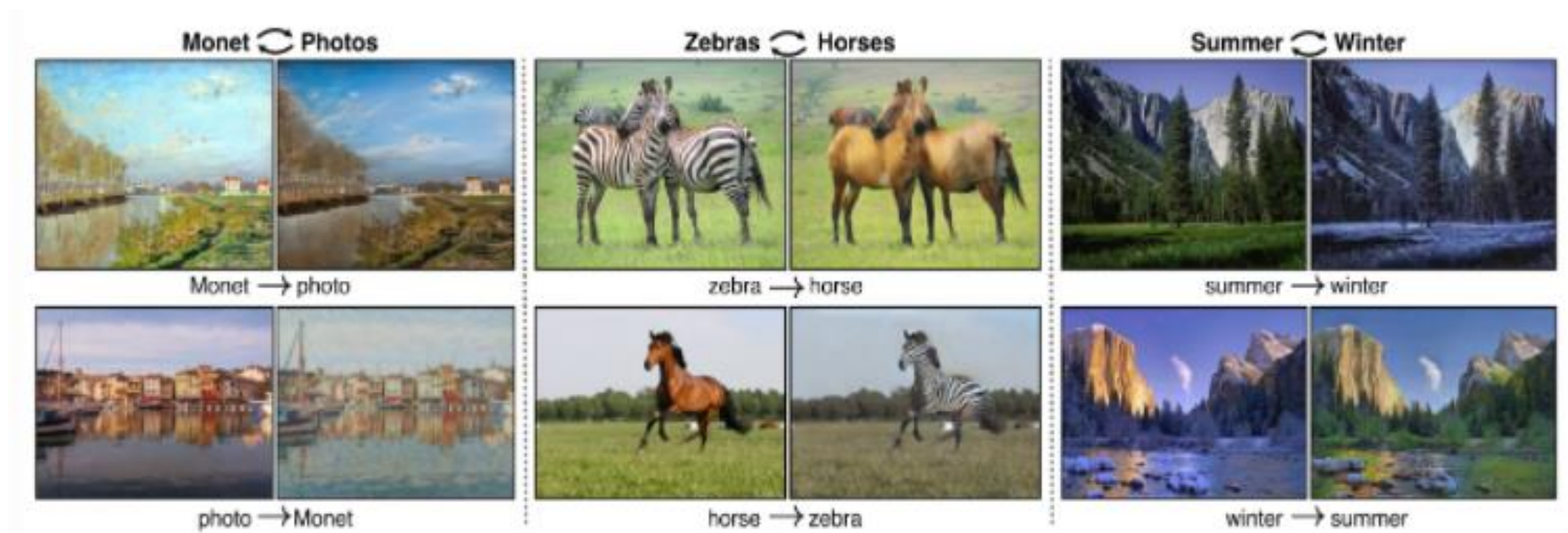


Model



→ harder to collect paired datasets than unpaired datasets

CycleGAN



Architecture

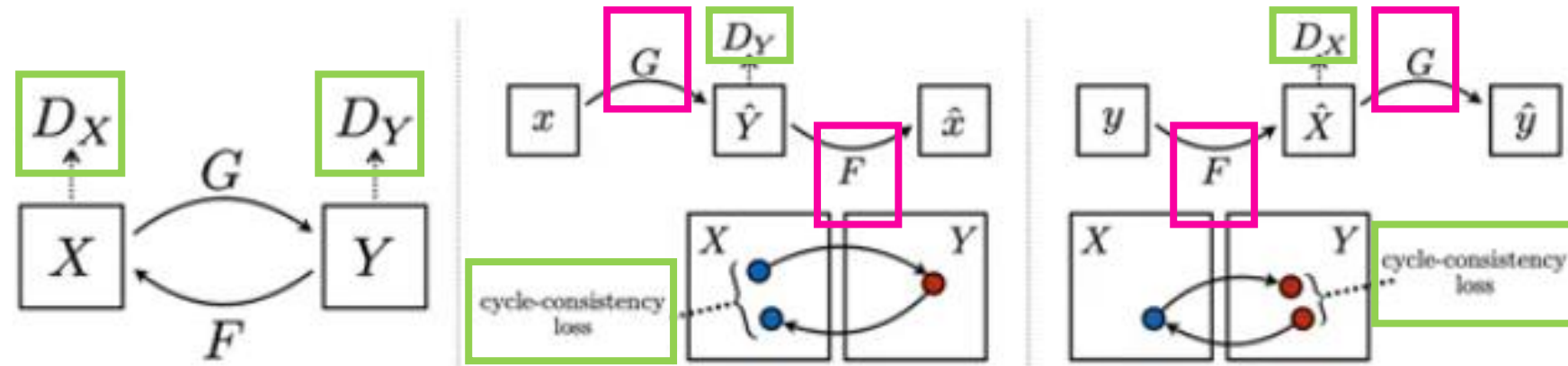
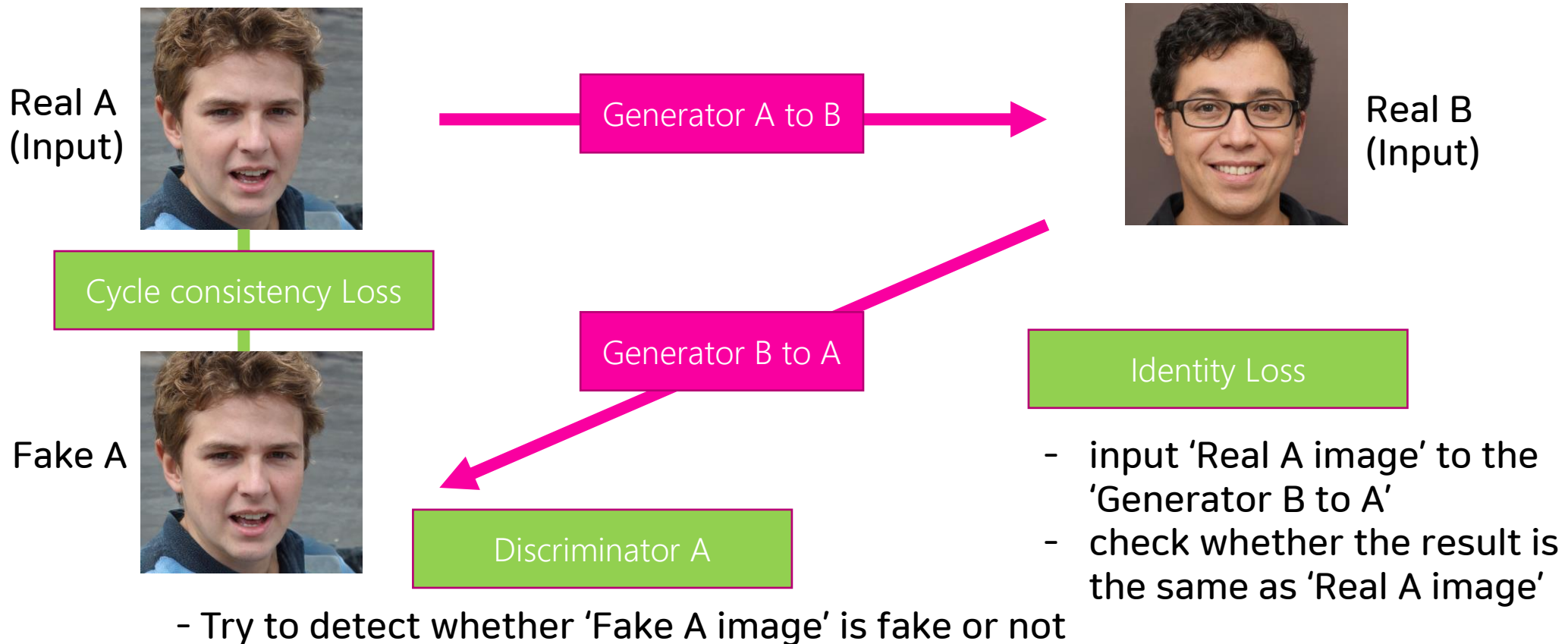


Fig5: The training procedure for CycleGAN.

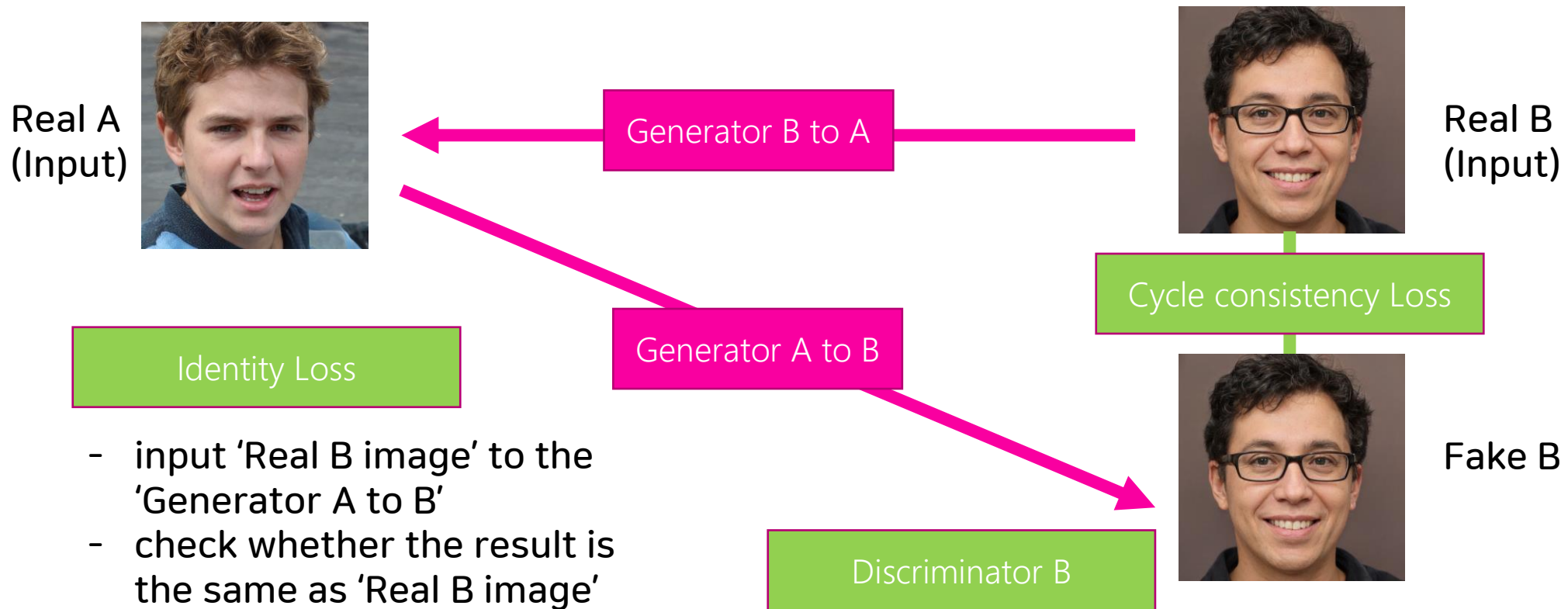
Model

- For training with unpaired dataset(from original A to fake A)



Model

- For training with unpaired dataset(from original B to fake B)



- input 'Real B image' to the 'Generator A to B'
- check whether the result is the same as 'Real B image'

- Try to detect whether 'Fake B image' is fake or not

Model

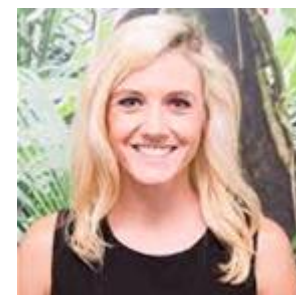
- For test



Generator A to B



Generator B to A



Preprocessing

Data Filtering

- Having full shape of glasses and two eyes
- One person per picture
- Face accounts for one third of the picture
- No mask / Included sunglasses

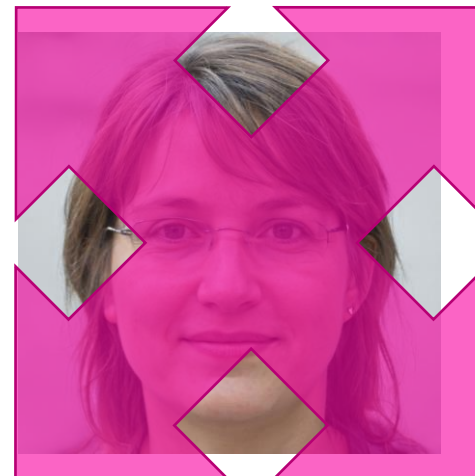
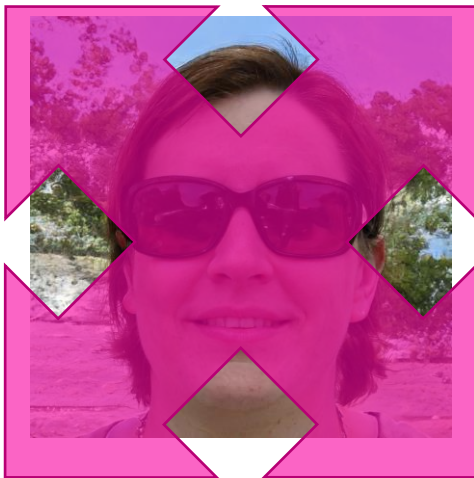


no



yes

Preprocessing



Code – Model

Generator

```
netG_a2b = CycleGAN(in_channels=nch, out_channels=nch, nker=nker, norm=norm, nblk=9).to(device)
netG_b2a = CycleGAN(in_channels=nch, out_channels=nch, nker=nker, norm=norm, nblk=9).to(device)

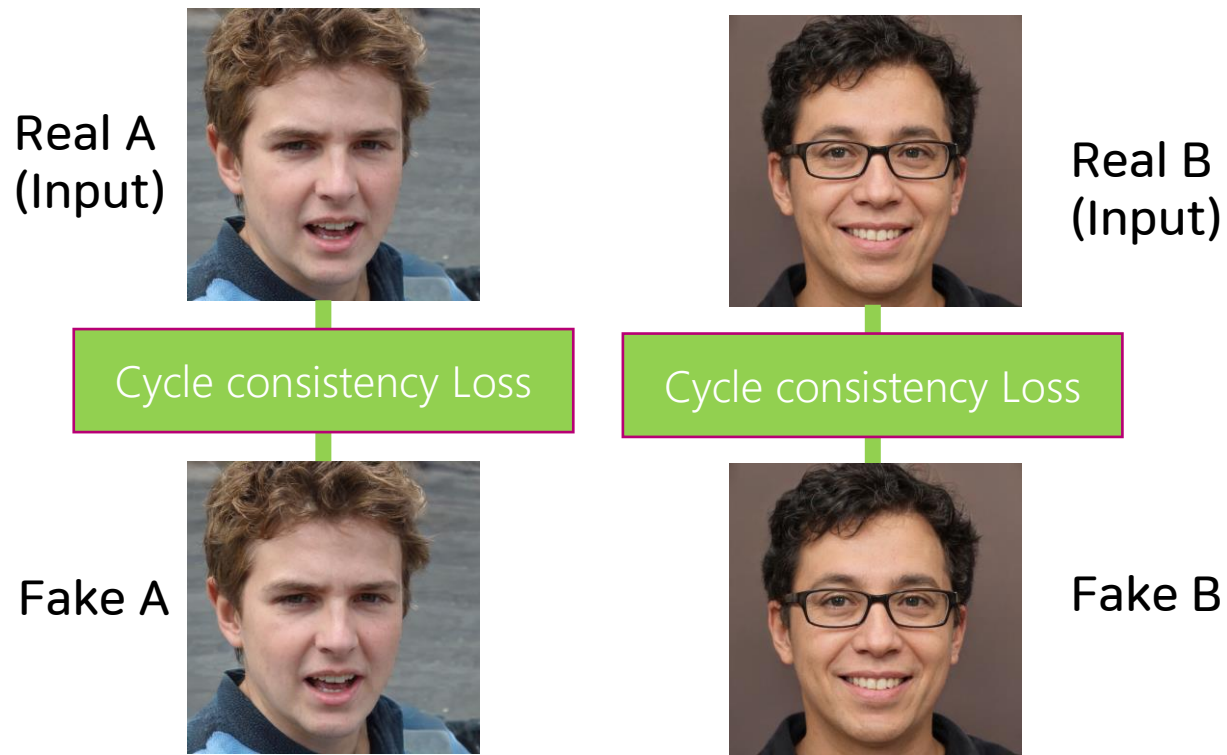
init_weights(netG_a2b, init_type='normal', init_gain=0.02)
init_weights(netG_b2a, init_type='normal', init_gain=0.02)
```

Discriminator

```
netD_a = Discriminator(in_channels=nch, out_channels=1, nker=nker, norm=norm).to(device)
netD_b = Discriminator(in_channels=nch, out_channels=1, nker=nker, norm=norm).to(device)

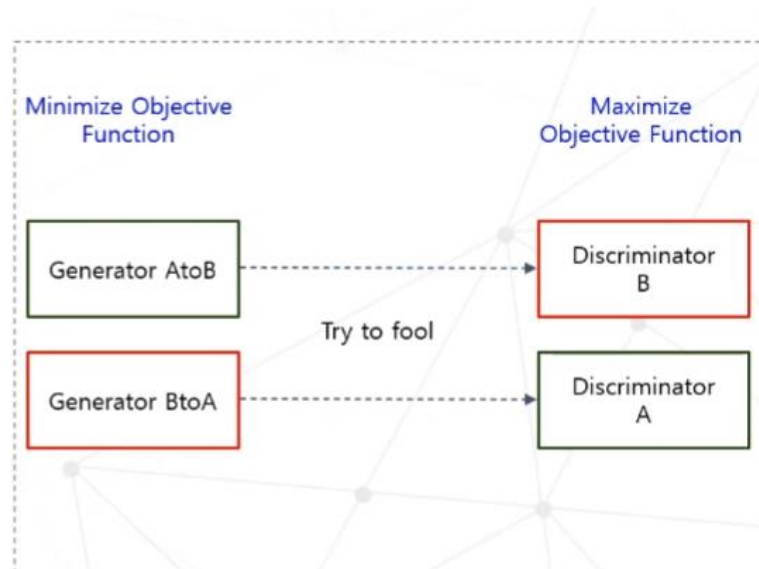
init_weights(netD_a, init_type='normal', init_gain=0.02)
init_weights(netD_b, init_type='normal', init_gain=0.02)
```

code – Cycle loss



```
fn_cycle = nn.L1Loss().to(device)
loss_cycle_a = fn_cycle(input_a, recon_a)
loss_cycle_b = fn_cycle(input_b, recon_b)
```

code – Gan loss



```
fn_gan = nn.BCELoss().to(device)
```

```
# backward netD_a
```

```
pred_real_a = netD_a(input_a)
```

```
pred_fake_a = netD_a(output_a.detach())
```

```
loss_D_a_real = fn_gan(pred_real_a, torch.ones_like(pred_real_a))
```

```
loss_D_a_fake = fn_gan(pred_fake_a, torch.zeros_like(pred_fake_a))
```

```
loss_D_a = 0.5 * (loss_D_a_real + loss_D_a_fake)
```

```
# backward netD_b
```

```
pred_real_b = netD_b(input_b)
```

```
pred_fake_b = netD_b(output_b.detach())
```

```
loss_D_b_real = fn_gan(pred_real_b, torch.ones_like(pred_real_b))
```

```
loss_D_b_fake = fn_gan(pred_fake_b, torch.zeros_like(pred_fake_b))
```

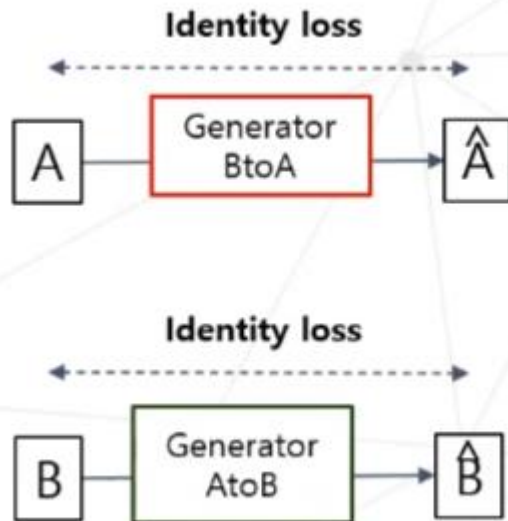
```
loss_D_b = 0.5 * (loss_D_b_real + loss_D_b_fake)
```

```
loss_D = loss_D_a + loss_D_b
```

```
loss_G_a2b = fn_gan(pred_fake_a, torch.ones_like(pred_fake_a))
```

```
loss_G_b2a = fn_gan(pred_fake_b, torch.ones_like(pred_fake_b))
```


code – Identity loss



```
fn_ident = nn.L1Loss().to(device)
```

```
loss_ident_a = fn_ident(input_a, ident_a)
```

```
loss_ident_b = fn_ident(input_b, ident_b)
```

Results- From A to B



Results- From B to A



Applications



Discussion & Limitations

1. Impossible to make various kinds of glasses for each person.
 - Because the weight of generators are fixed once the training is over
2. Low resolution & changes other than glasses
 - Problems in scaling images
3. changes other than glasses
 - May reflect actual defects of glasses including refraction.
4. Changes in hues
 - Instance normalization
 - Background becomes brighter by faces
 - Faces becomes darker by background

Reference

https://openaccess.thecvf.com/content_ICCV_2017/papers/Zhu_Unpaired_Image-To-Image_Translation_ICCV_2017_paper.pdf

<https://velog.io/@tobigs-gm1/Style-GAN>

<https://blog.lunit.io/2017/04/27/style-transfer/>

<https://arxiv.org/abs/1703.10593>

[**https://www.kaggle.com/jeffheaton/glasses-or-no-glasses**](https://www.kaggle.com/jeffheaton/glasses-or-no-glasses)

<http://mmlab.ie.cuhk.edu.hk/projects/CelebA.html>

<http://www.kwangsiklee.com/2018/03/cyclegan%E9D%B4-%EB%AC%B4%E97%87%E9D%B8%E%A7%80-%EC%95%8C%E95%84%EB%B3%B4%E9E%90/>