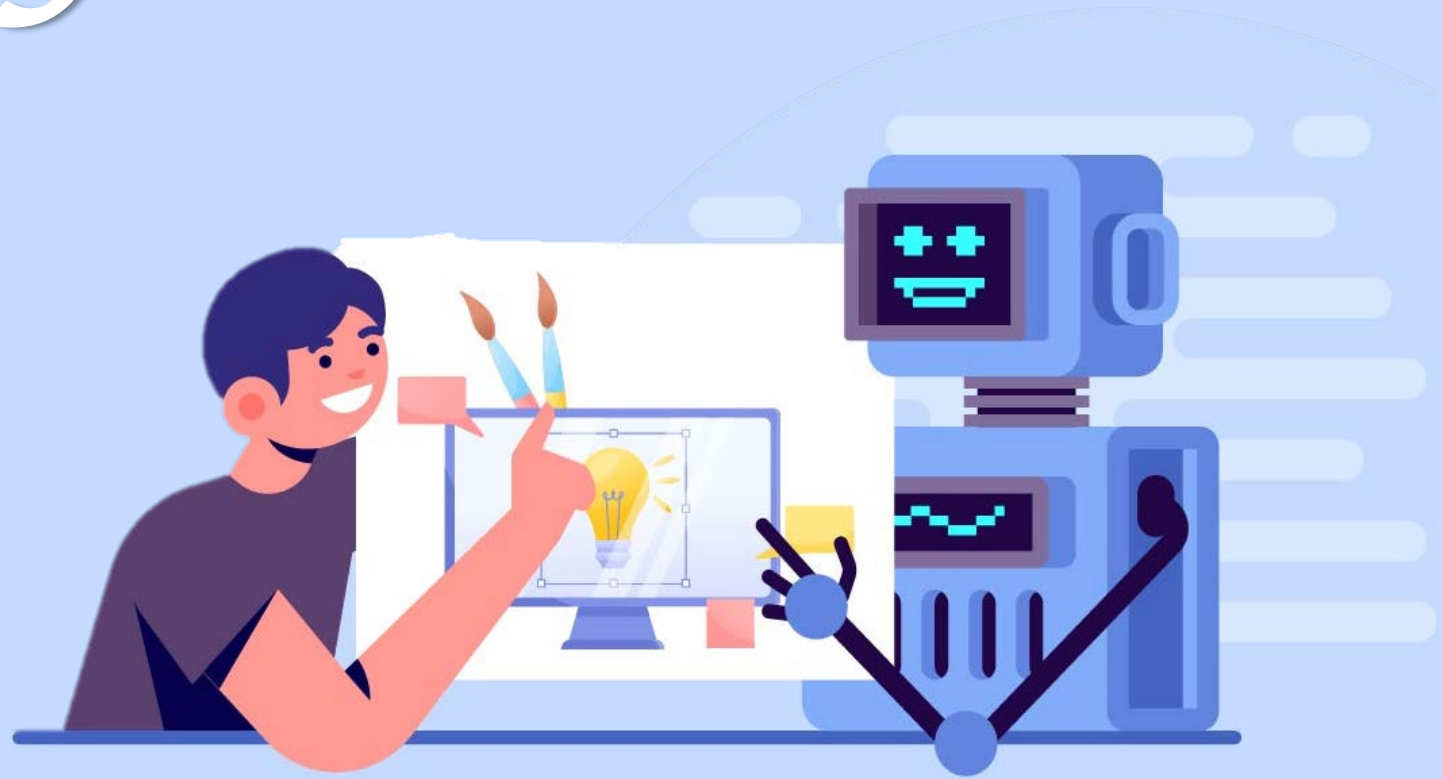


# Fancy Font



23-2 DSL 모델링 프로젝트 F조  
9기 김영현, 유선재, 이성균, 장현빈 & 10기 정성오



# *Contents*

---

Overview

U-Net

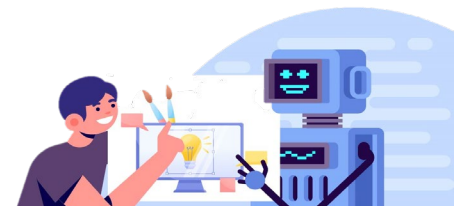
DCGAN + cGAN

Result

Improvement

# 1. INTRODUCTION

---



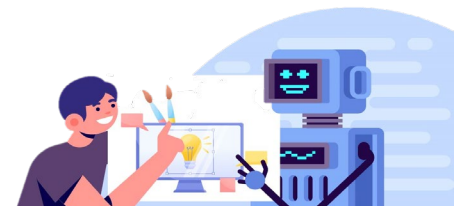
# 1. INTRODUCTION

---

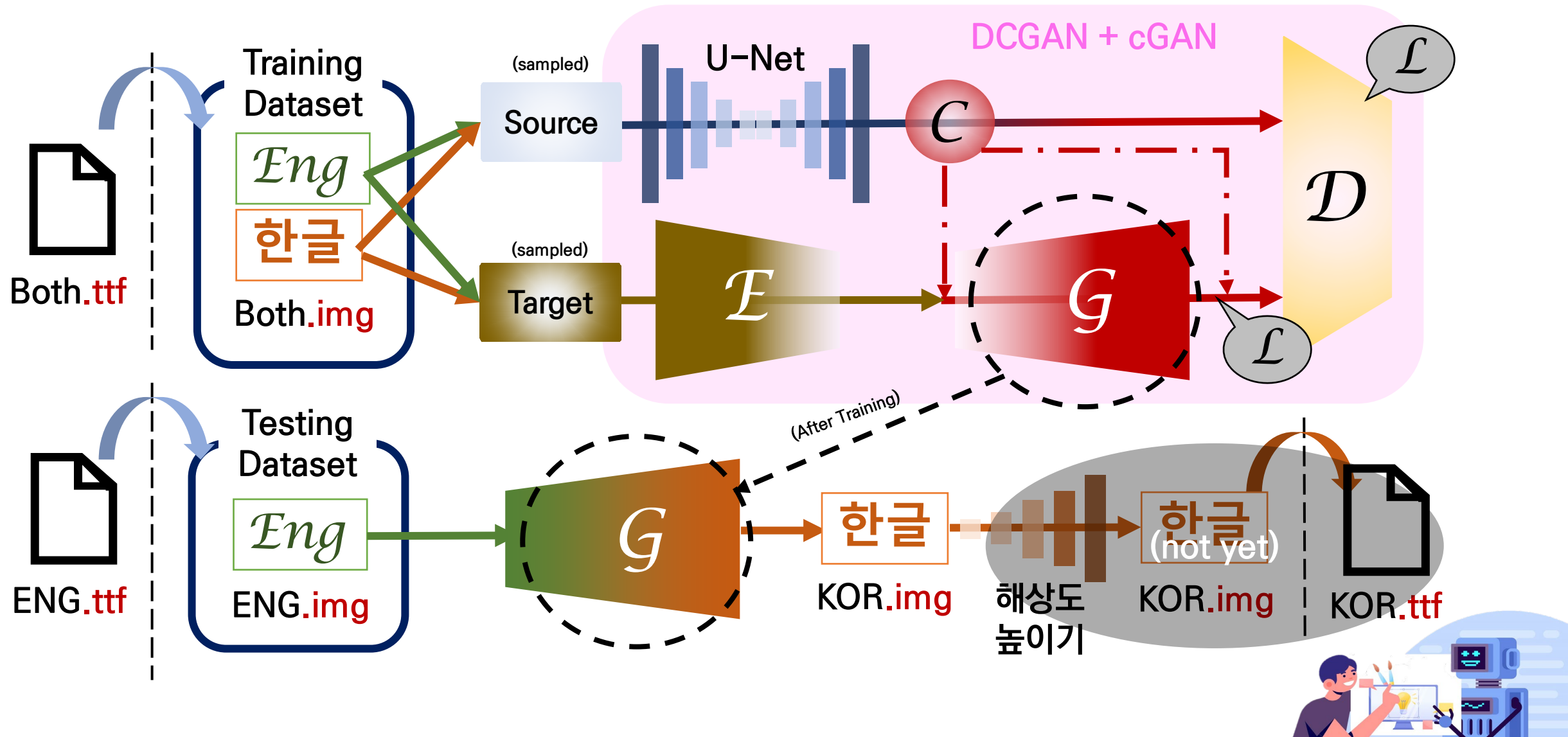
*Fancy  
Font*

: 한글이 호환되지 않는 영어 폰트의 한글 폰트를 생성하는 모델

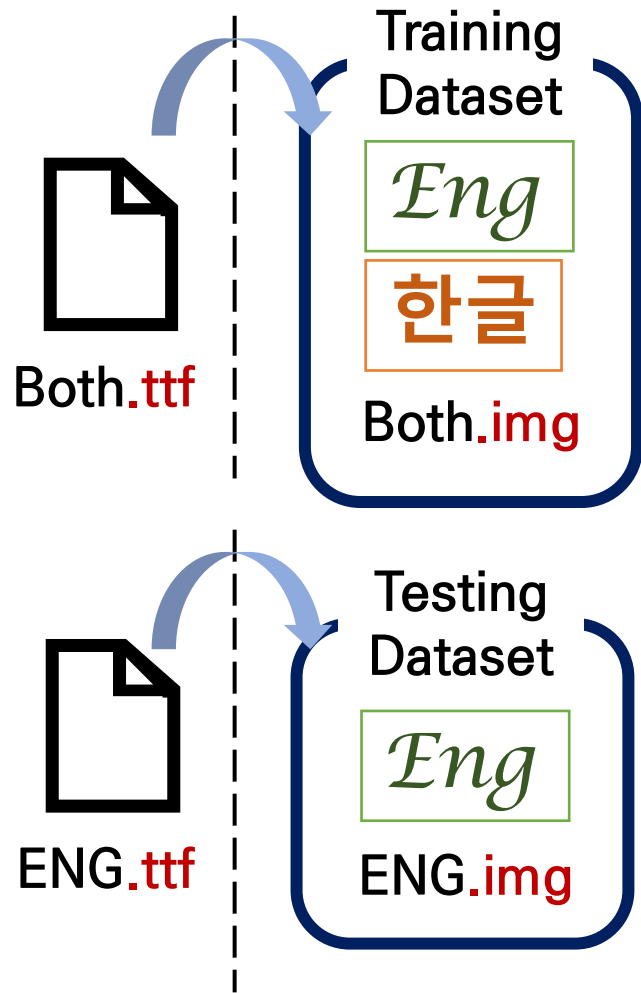
Backbone : DCGAN + conditional GAN



## 2. OVERVIEW



# 3. DATASET

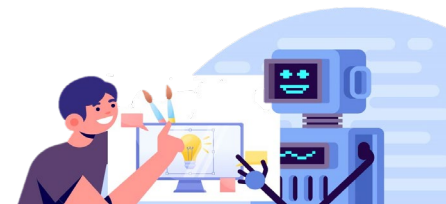


ASCII 코드에 맞춰서 embedding 부여

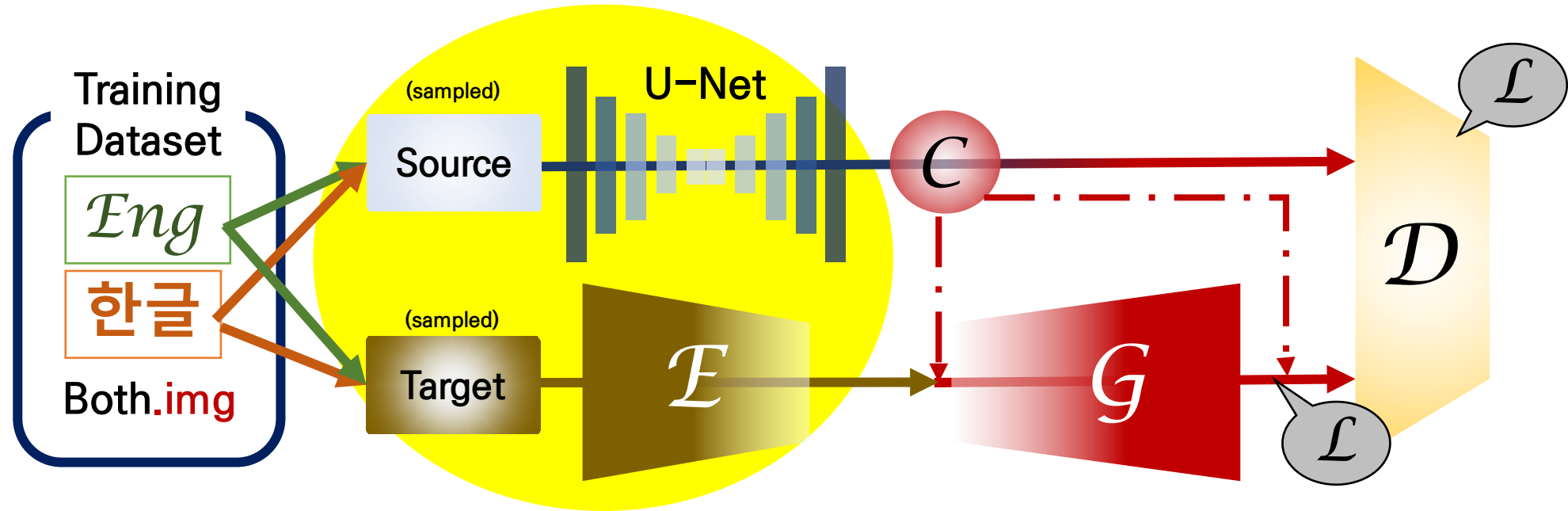
학습시킨 폰트 개수 : 22개(나눔손글씨 우선)

한글 글자 개수 : 11224 글자(초성, 중성, 종성의 조합)

영어 알파벳 개수(대,소문자 구분) : 52 글자



# 4.1. U-NET

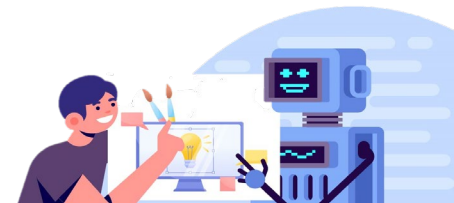


**Source** font image : 한/영 관계없이 무작위추출

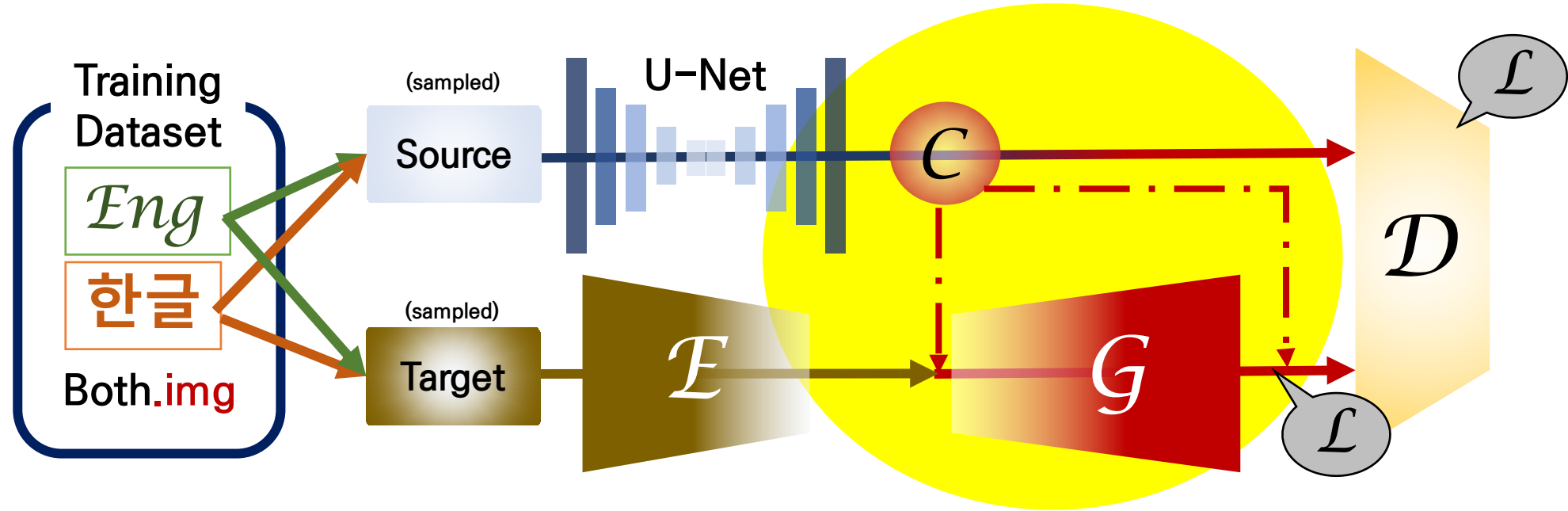
: 폰트의 'structure(글자 구조)'를 뽑아내야 하므로 U-Net 6개 층마다의 input을 condition으로 설정 & U-Net을 통과하여, condition에 structure feature가 잘 들어갔는지 확인

**Target** font image : 한/영 관계없이 무작위추출

: 'style(글자 모양)'을 뽑아낼 대상이며, Encoder로 style feature를 추출



## 4.2. CONDITIONAL GAN



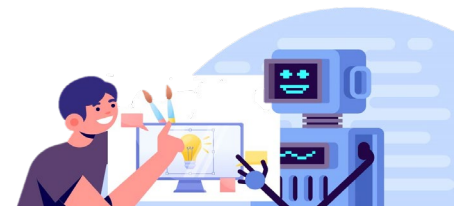
Encoder : 6단계의 Convolution Block 통과

: 1-32-64-128-256-1024

Generator(Decoder) : 6단계의 DeConvolution Block 통과, 이때 6개 층마다 Source의 condition을 삽입

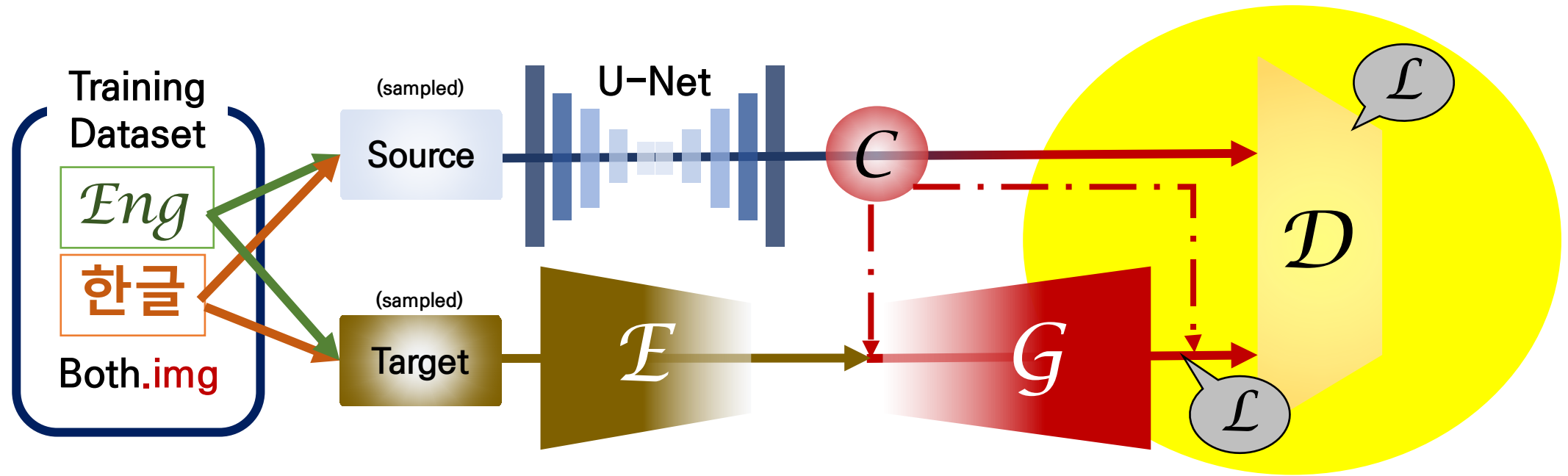
: 1024-512-256-128-64-32-1

Generator Loss : L1-norm 형태로 계산(절댓값)



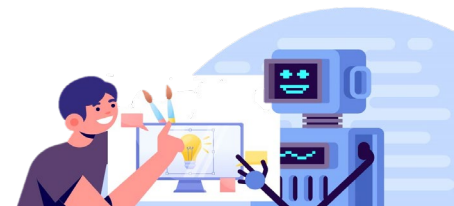


## 4.3. DISCRIMINATOR



Discriminator : 5단계의 Convolution Block 통과(with max-pooling)  
: 2-64-64-64-512-1

Discriminator Loss : 활성화함수(sigmoid)를 통과시킨 후, binary cross entropy 계산  
: adversarial loss



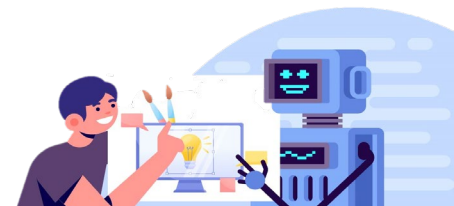
## 4.4. TRAINING

🎯 Generator Loss : L1 loss  $\mathcal{L}_{\mathcal{G}}(\mathbf{s}, \mathbf{s}') = \sum_{i=1}^N |s_i - s'_i|$

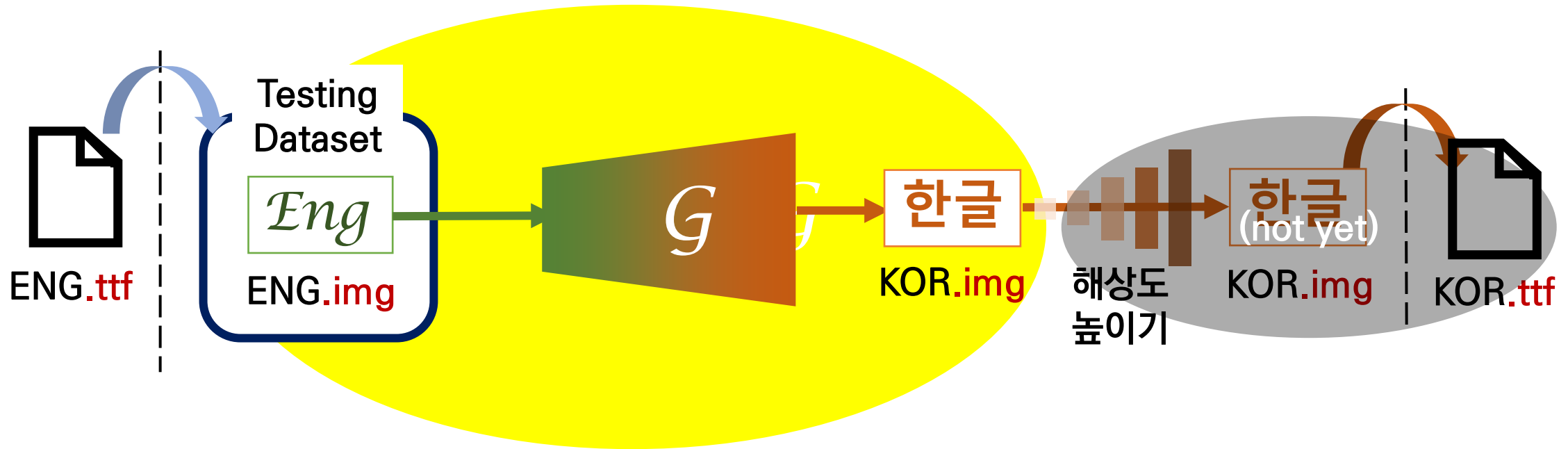
🎯 Discriminator Loss : **BCEwithlogits loss** (sigmoid layer + BCEloss)

$$\mathcal{L}_{\mathcal{D}}(\mathbf{s}', \mathbf{t}, \mathbf{w}) = \sum_{i=1}^N -w_i [t_i \cdot \log \sigma(s'_i) + (1 - t_i) \cdot \log (1 - \sigma(s'_i))]$$

→ Total Loss = Generator Loss + Discriminator Loss



## 4.5. INFERENCE



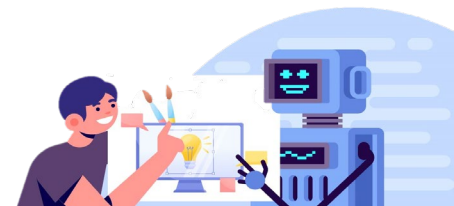
input

- 1) 영어만 지원되는 폰트(의 index)
- 2) 한글 글자 또는 문장(의 ASCII 코드)

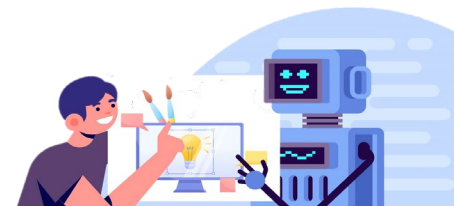
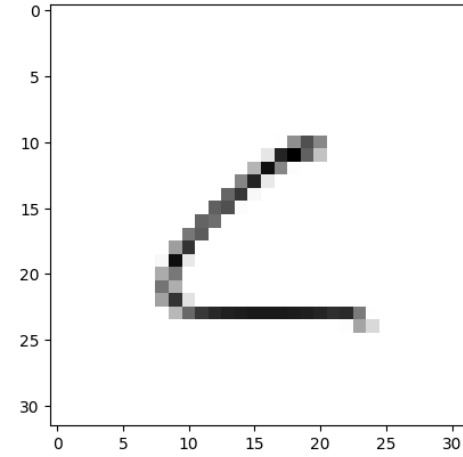
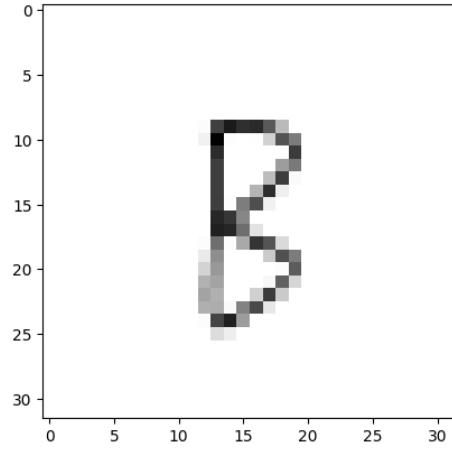
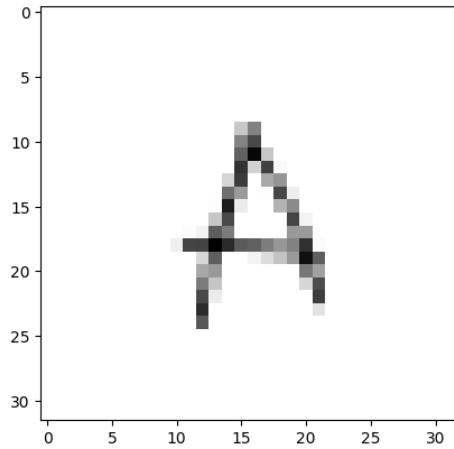


output

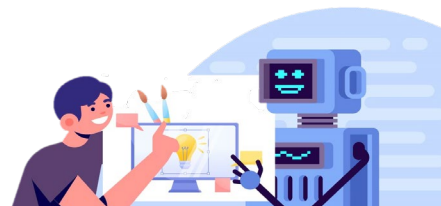
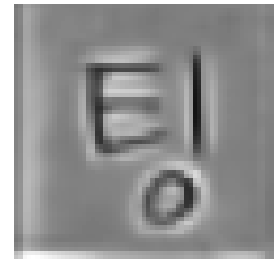
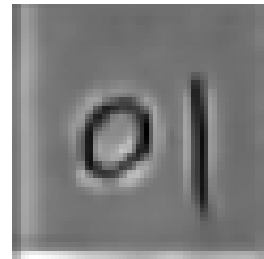
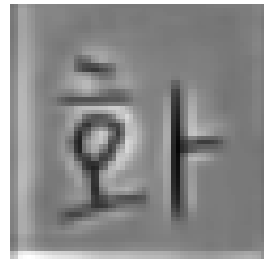
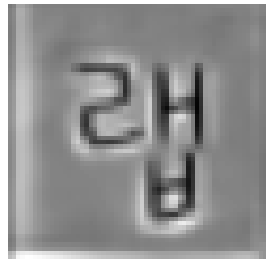
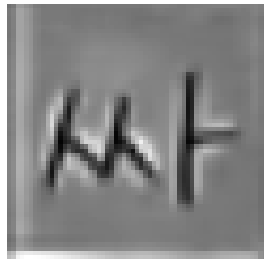
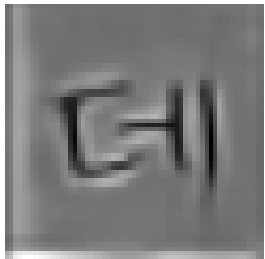
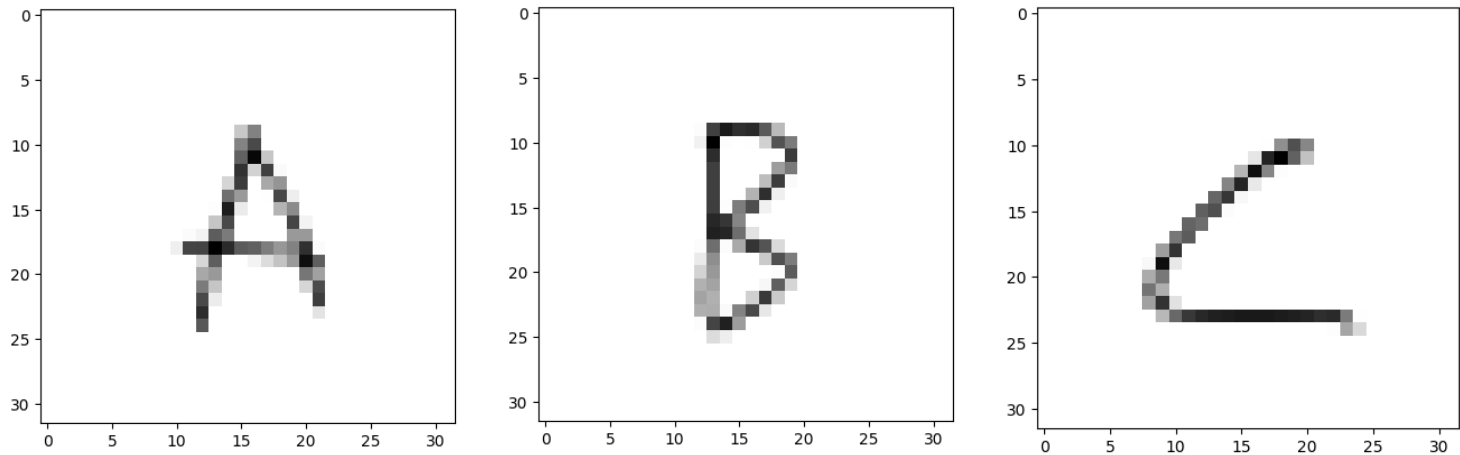
영어 폰트로 쓴  
한글 글자의 이미지



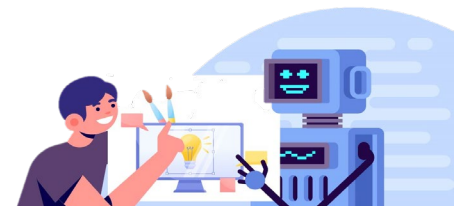
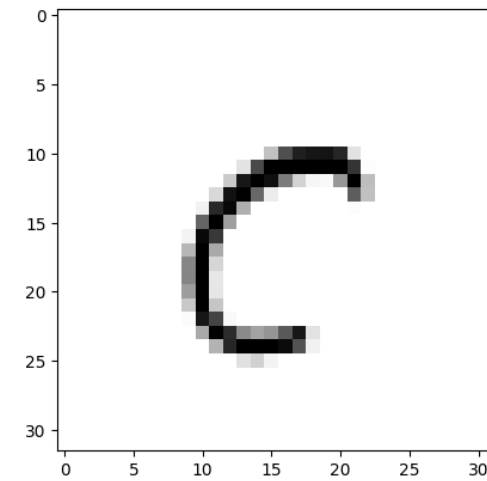
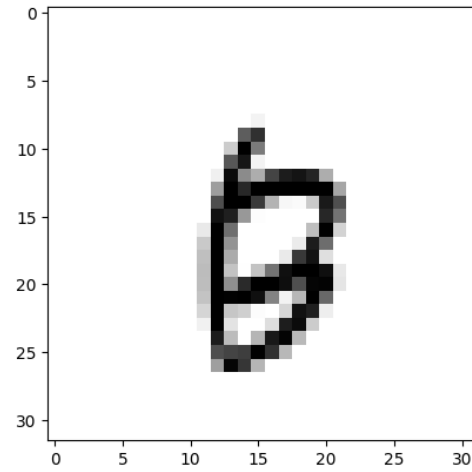
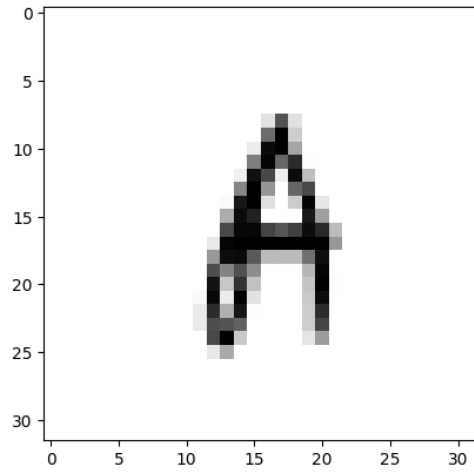
# 5. RESULT



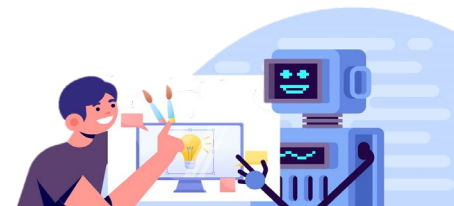
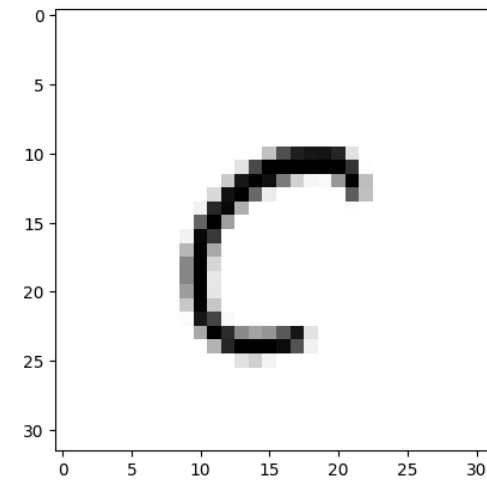
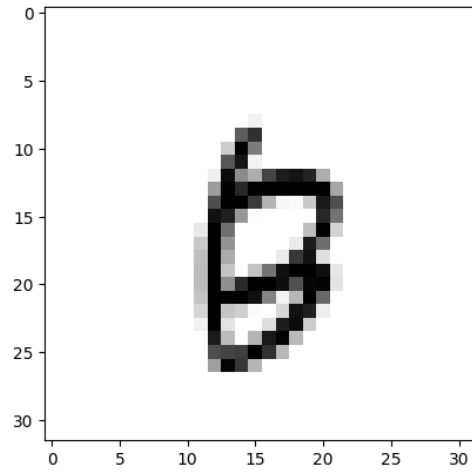
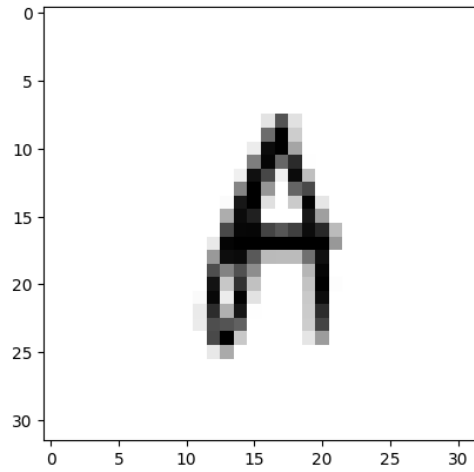
# 5. RESULT



# 5. RESULT



# 5. RESULT

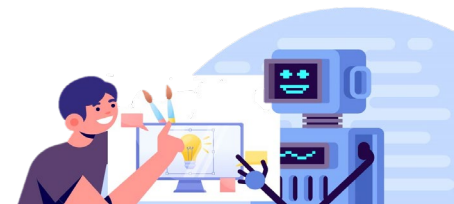


## 6.1. PITSFALL

- ④ 이미지 생성 속도가 다소 느림  
: 한글 글자 조합이 매우 많아서 발생하는 문제(11224개)
- ④ 생성된 이미지의 화소가 낮음( $32*32$ )  
: 모델의 속도를 다소 높이하고자 화소를 낮게 처리함

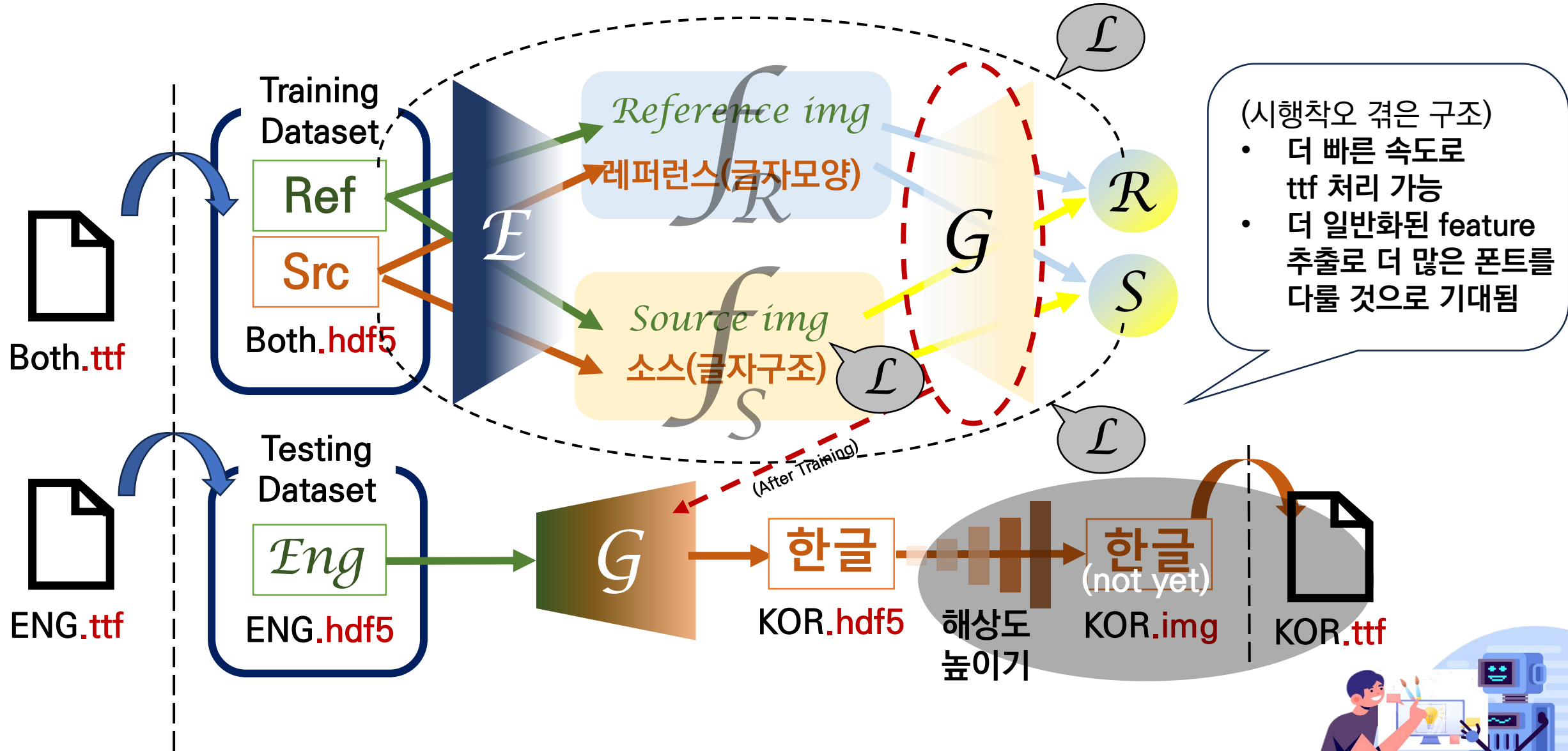
### 개선방안

- 데이터셋 처리 속도를 높이기 위해 **메타데이터 형식**으로 데이터셋을 구성하기
- Source feature(글자 구조에 대한 것) & target feature(글자 모양에 대한 것)을 **깔끔하게 추출하기**

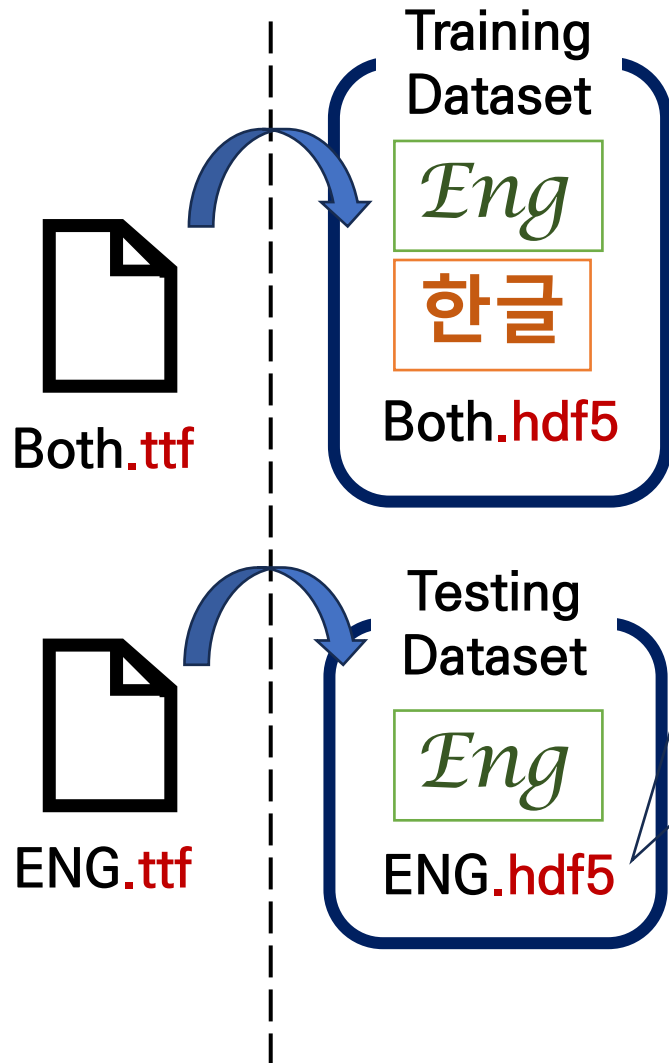




## 6.2. TRIAL AND ERROR

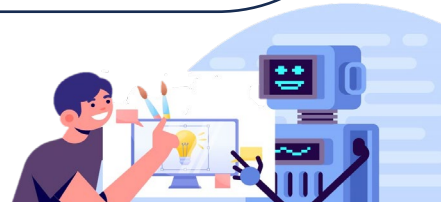
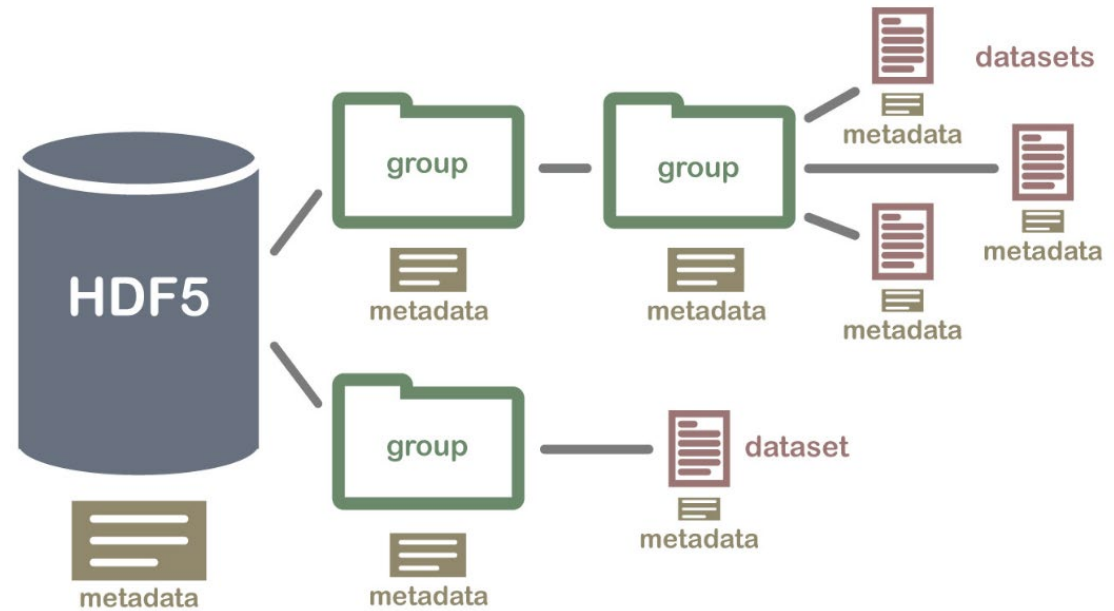


## 6.3. IMPROVEMENT



### hdf5 (Hierarchical Data Format)

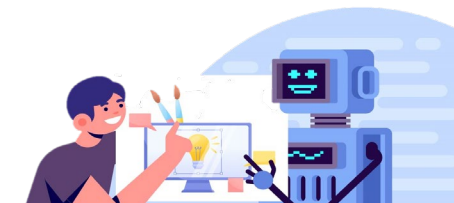
: 대용량 데이터를 빠르게 읽고 쓸 수 있는 메타데이터 확장자



## 6.3. IMPROVEMENT



생성된 이미지(32\*32)의 화소를 높여서 선명하게 해주는 부분  
: 그래야 궁극적으로 img를 ttf로 바꿀 수 있을 것으로 예상됨  
아직 구현하지 못함



*Thank you*

