What is font impression??

Fonts have a huge number of variations.

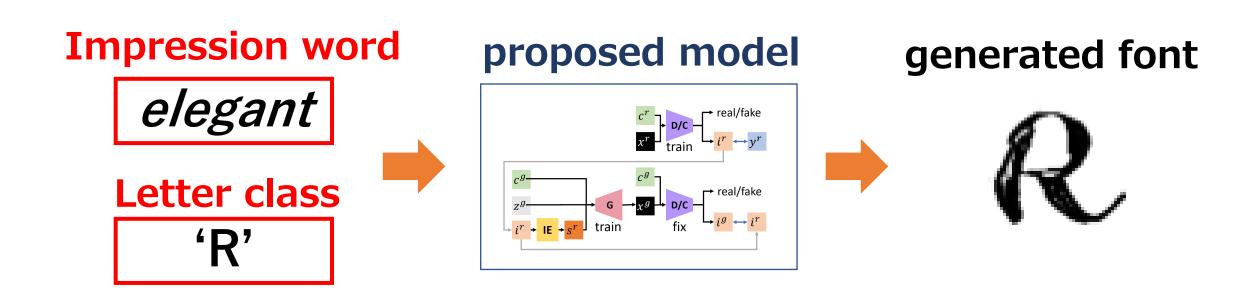


Myfonts dataset is released with impressions tagged to fonts.

1

Purpose

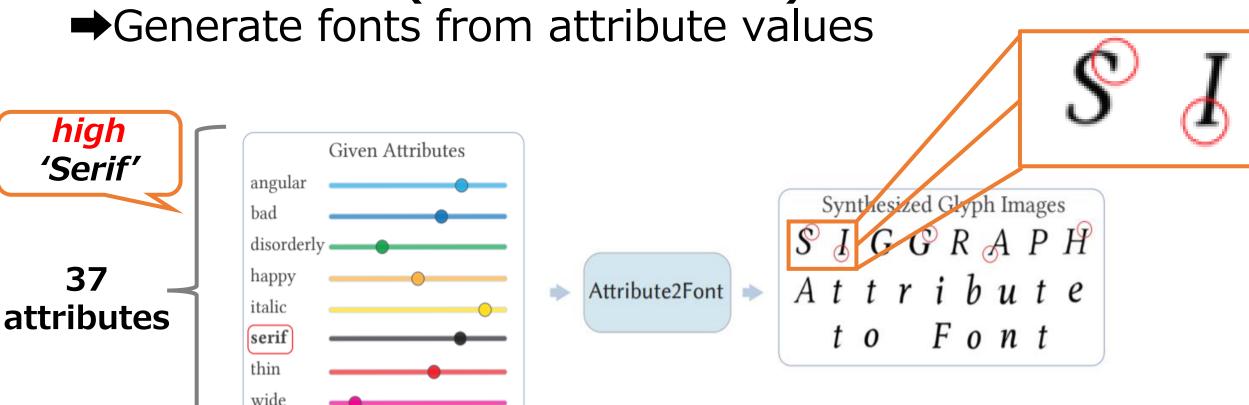
Generate Fonts with specific impressions.



We propose Impressions2Font

Related Work

Attribute2Font(SIGGRAPH2020)



Impressions that can be manipulated are limited.

First Problem

Impression word are inconsistent



Opposite meaning!!

Why? →Impression are subjective







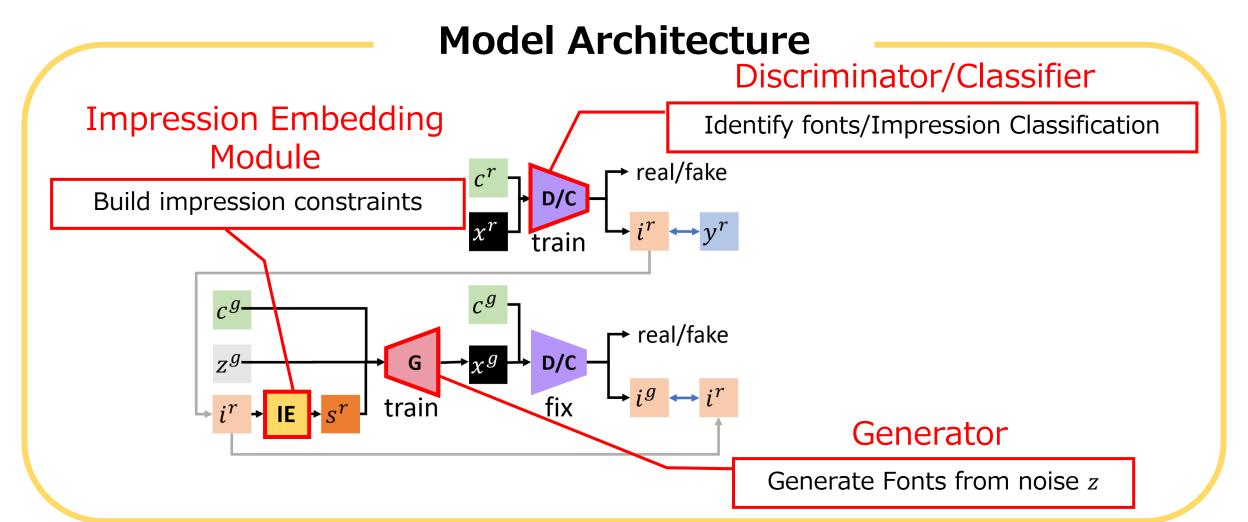




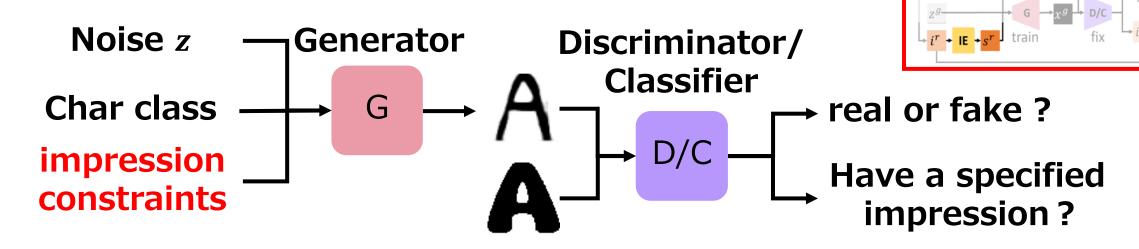
Impressions2Font

Our model is based on conditional GANs.

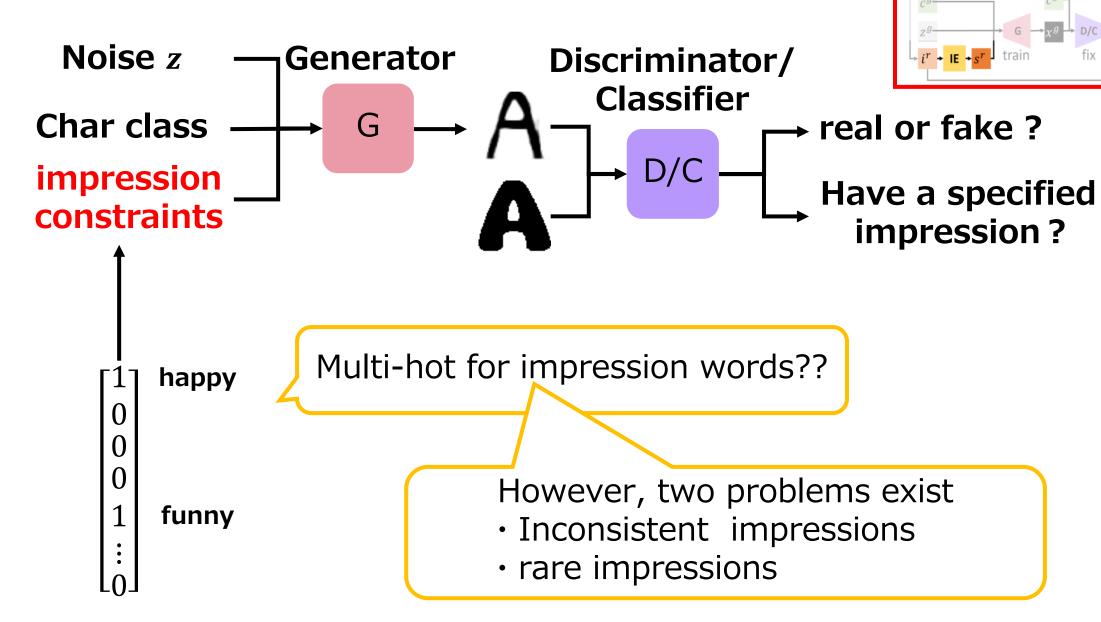
Composed of 3 modules



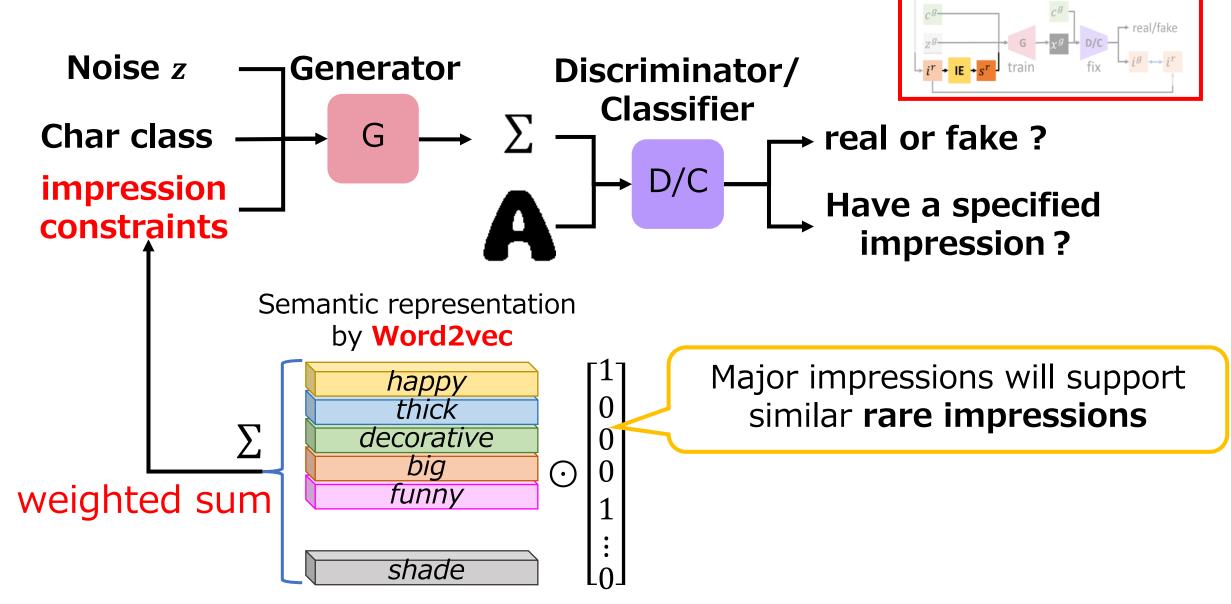
Why impression Embedding Module??



Why impression Embedding Module??

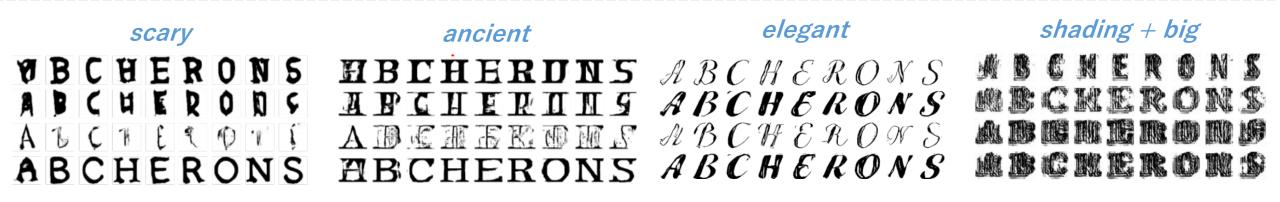


Why impression Embedding Module??



Why impression Embedding Module?? **Not subjective** Noise z Generator Discriminator/ Classifier Char class G real or fake? D/C impression Have a specified constraints impression? Semantic representation by Word2vec provability ⁻0.67 happy Optimize the Generator thick conditional on **soft constraints** 0.1 decorative \odot big 0.2 weighted sum funny 0.3 Robust to **inconsistency** shade

Generating from specific impressions



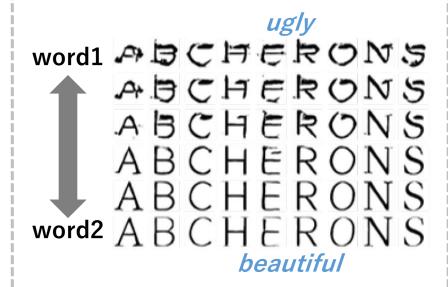
From unlearned impression

big(learned) ABCHERONS

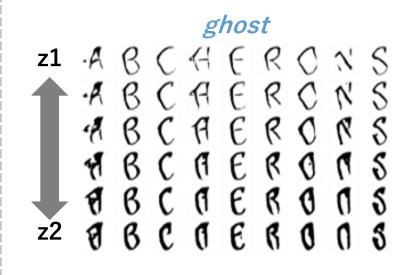
huge(unlearned)

ABCHERONS ABCHERONS

Impression interpolation

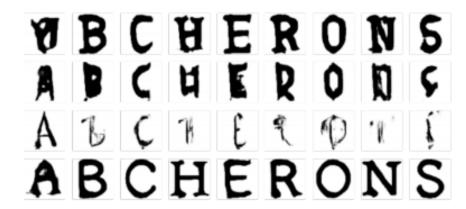


Impression interpolation



ghost

fake image



real image



elegant

fake image

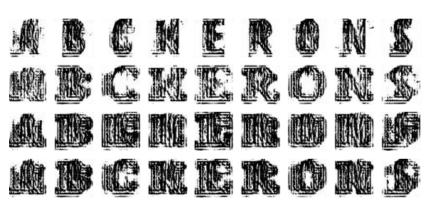


real image

ABCHERONS HILLERONG ABCHERONG LBCHERONS

shading + big

fake image



real image



Icd + ancient

fake image

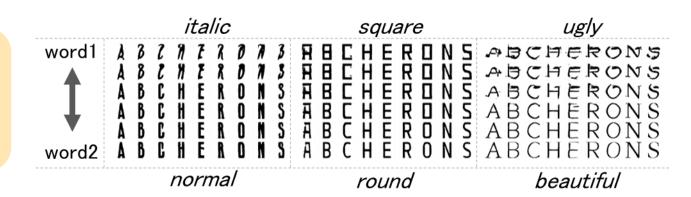
real image



interpolate between different fonts

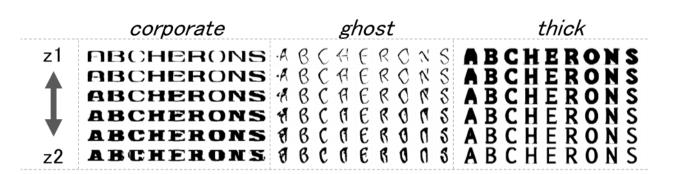
Interpolation impression

$$\alpha \phi(w_1) + (1 - \alpha)\phi(w_2) \ (0 \le \alpha \le 1)$$

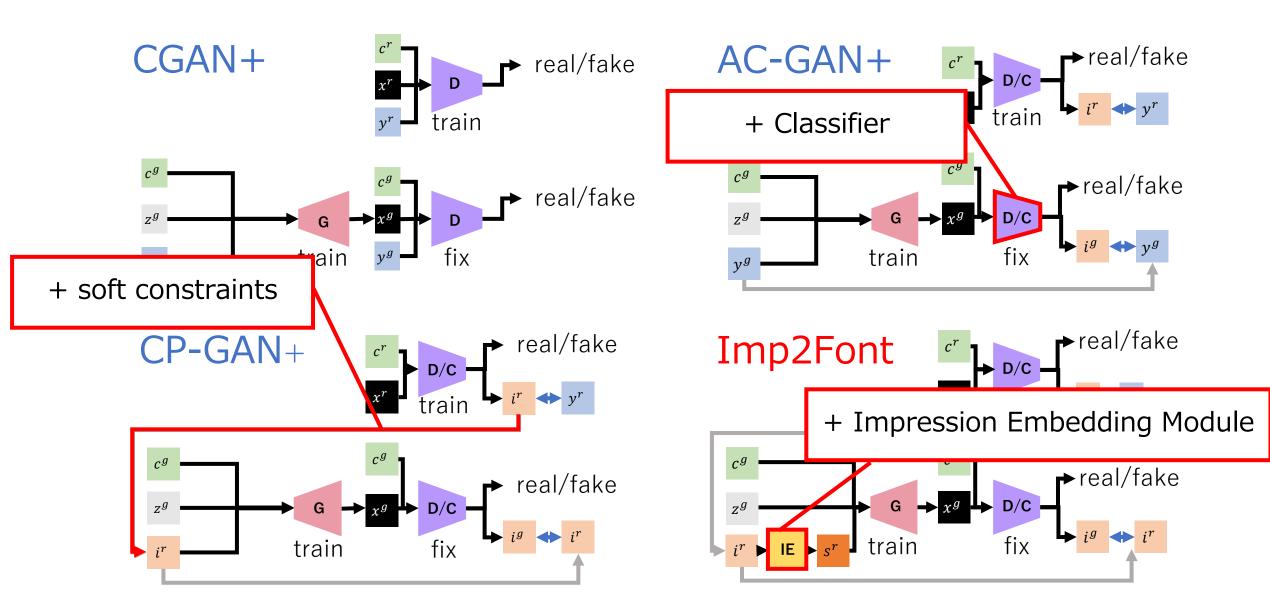


Interpolation noise

$$\alpha\phi(z_1) + (1-\alpha)\phi(z_2) \ (0 \le \alpha \le 1)$$



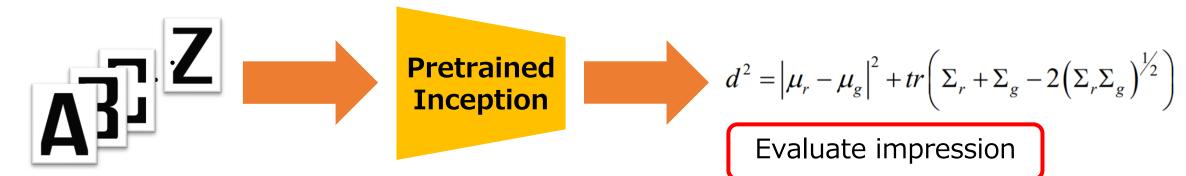
Quantitative evaluations



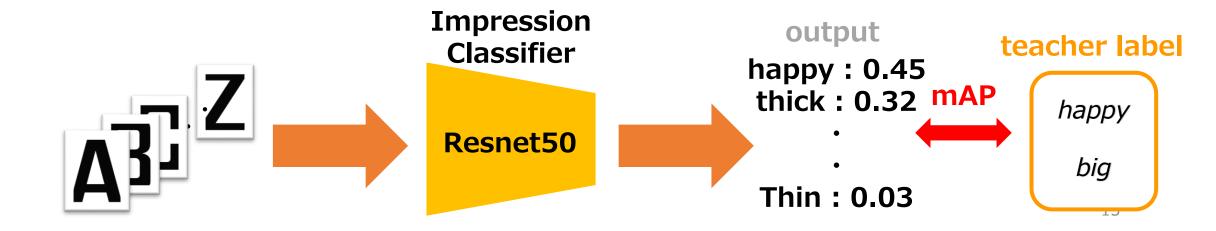
Quantitative evaluations

Evaluate **Diversity** & **Quality**

1. FID · · Compare the distribution of the generated image with the distribution of the actual image



- 2. 1mAP-train · · train: fake data/test: real data
 - 2mAP-test · · train: real data/test: fake data



Quantitative evaluations

C-GAN+ · · General Conditional GANs

AC-GAN+ · · + Classifier

CP-GAN+ · · + soft constraints

Imp2Font • • + Impression Embedding Module

	C-GAN+	AC-GAN+	CP-GAN+	Imp2Font
↓FID	39.634	39.302	33.667	24.903
↑ mAP-train	1.524	1.157	1.823	1.765
↑ mAP-test	1.155	1.158	1.600	1.708

- ✓ Imp2Font greatly improves FID
- → High quality and diversity compared to other methods
- ✓ Imp2Font and CP-GANs+ greatly improves mAP-train/test
- → Impression Embedding Module is very effective.