### Ensembling with Deep Generative Views Lucy Chai, Jun-Yan Zhu, Eli Shechtman, Phillip Isola, Richard Zhang



#### https://chail.github.io/gan-ensembling/















### GANs continuously approximate real images



#### Goodfellow et al. 2014; StyleGAN2. Karras et al. 2020

### Ensembling GAN views for Classification





Input Image

# Ensembling GAN views for Classification C C C C



Input Image

**GAN** Generated Variations

### Projecting images into GAN latent space



W

iGAN. Zhu et al. 2016; GAN Inversion: A Survey. Xia et al. 2021

### $w^* = \arg\min L_{img}(x, G(w)) + \lambda L_{latent}(w, E(x))$







**PCA** Directions  $\tilde{w} = w^* + \beta \tilde{v}_d$ 







**Isotropic**  $\tilde{w} \sim \mathcal{N}(w^*, \sigma I)$ 



PCA Directions  $\tilde{w} = w^* + \beta \tilde{v}_d$ 



### Style-mixing $\tilde{w} = \min(w^*, w_r)$



#### Reconstruction



#### Coarse Layers



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#### Reconstruction



#### **Coarse Layers**



**Fine Layers** 



# Investigating Ensemble Weight





#### Input Image



Test Images



Fine Layer Style-mix Variations

Input Image



# Investigating Ensemble Weight



Input Image

GAN Generated Variations





Single Image

Color Jitter



Single Image Color Crop Jitter Jitter



Single Image ColorCropStyle-mixJitterJitterJitter

![](_page_15_Figure_1.jpeg)

Single Image ColorCropStyle-mix CombinedJitterJitterJitter

#### Input

![](_page_16_Picture_2.jpeg)

#### Reconstruction

![](_page_16_Picture_4.jpeg)

![](_page_16_Figure_5.jpeg)

#### Input

![](_page_17_Picture_2.jpeg)

#### Reconstruction

![](_page_17_Picture_4.jpeg)

#### Style-mix Fine Layers

![](_page_17_Picture_6.jpeg)

![](_page_17_Figure_7.jpeg)

#### Input

![](_page_18_Picture_2.jpeg)

#### Reconstruction

![](_page_18_Picture_4.jpeg)

#### Style-mix Fine Layers

![](_page_18_Picture_6.jpeg)

![](_page_18_Figure_7.jpeg)

#### Input

![](_page_19_Picture_2.jpeg)

#### Reconstruction

![](_page_19_Picture_4.jpeg)

#### Style-mix Fine Layers

![](_page_19_Picture_6.jpeg)

![](_page_19_Figure_7.jpeg)

### 12-way Cats Domain

#### Input

![](_page_20_Picture_2.jpeg)

#### Reconstruction

![](_page_20_Picture_4.jpeg)

#### Style-mix Coarse Layers

![](_page_20_Picture_6.jpeg)

![](_page_20_Figure_7.jpeg)

### Limitations

- GAN reconstruction capability
- GAN inversion efficiency
- Classifier sensitivities to GAN output
- Currently limited to simple tasks with small, structured datasets...
- But generation and inversion technology is rapidly improving!

### Summary

- StyleGAN as a generator of image variations
- Project image into latent space and perturb
- Requires adjustments to mitigate classifier sensitivity to GAN output

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![](_page_23_Picture_4.jpeg)

Project Website + Code + Colab: https://chail.github.io/gan-ensembling/