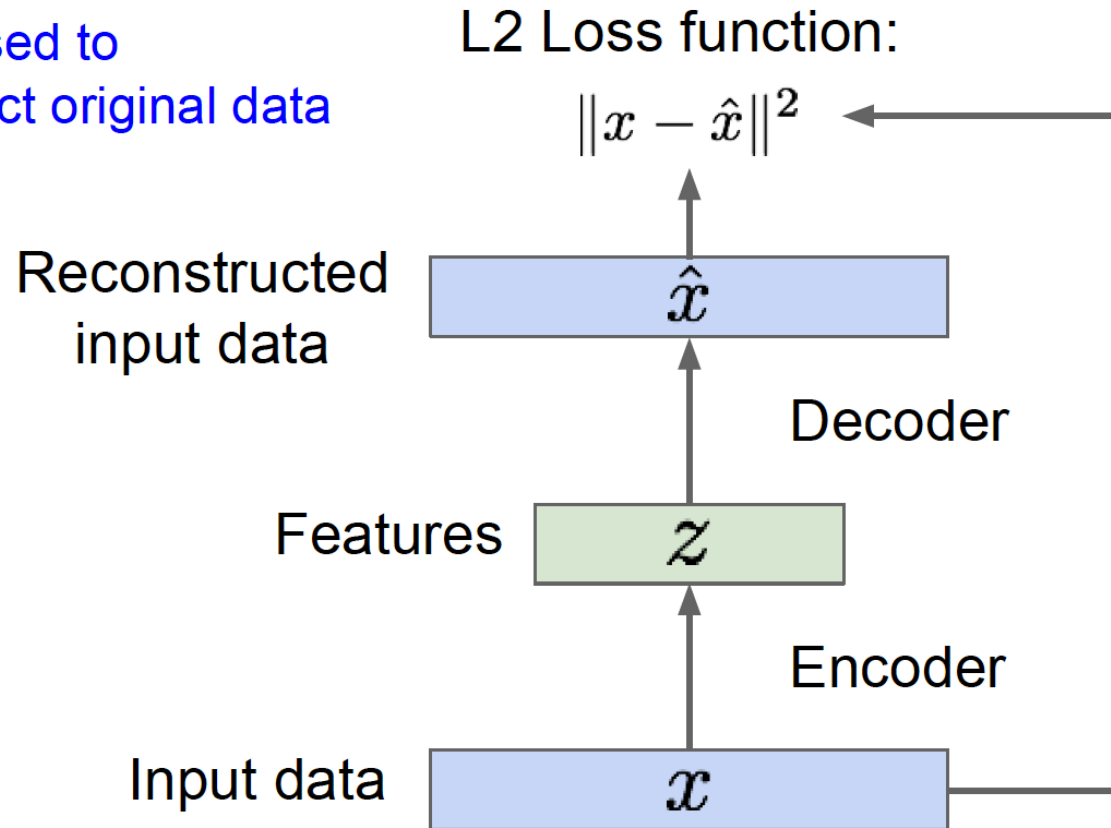


Generative Adversarial Networks And Their Applications

Bindita Chaudhuri

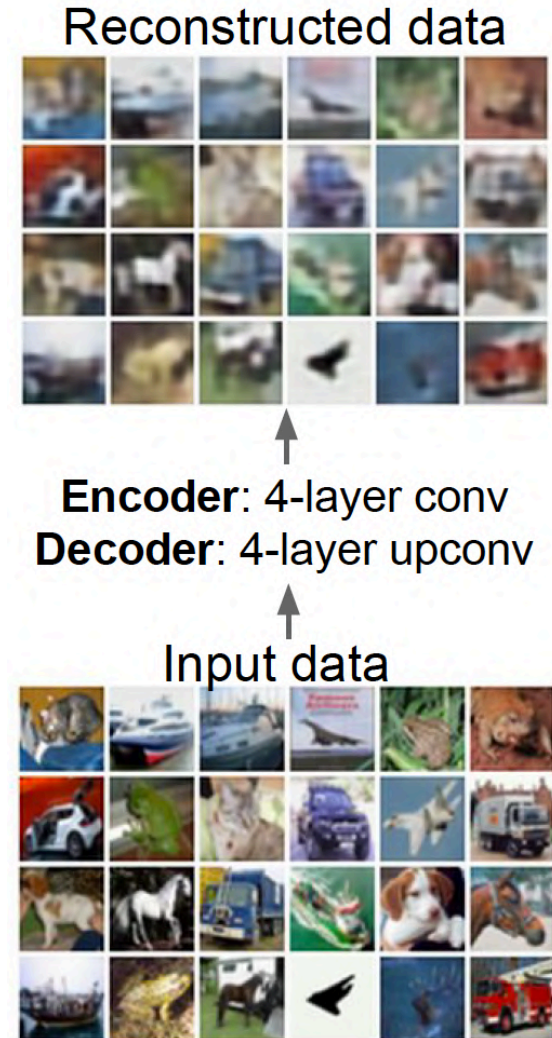
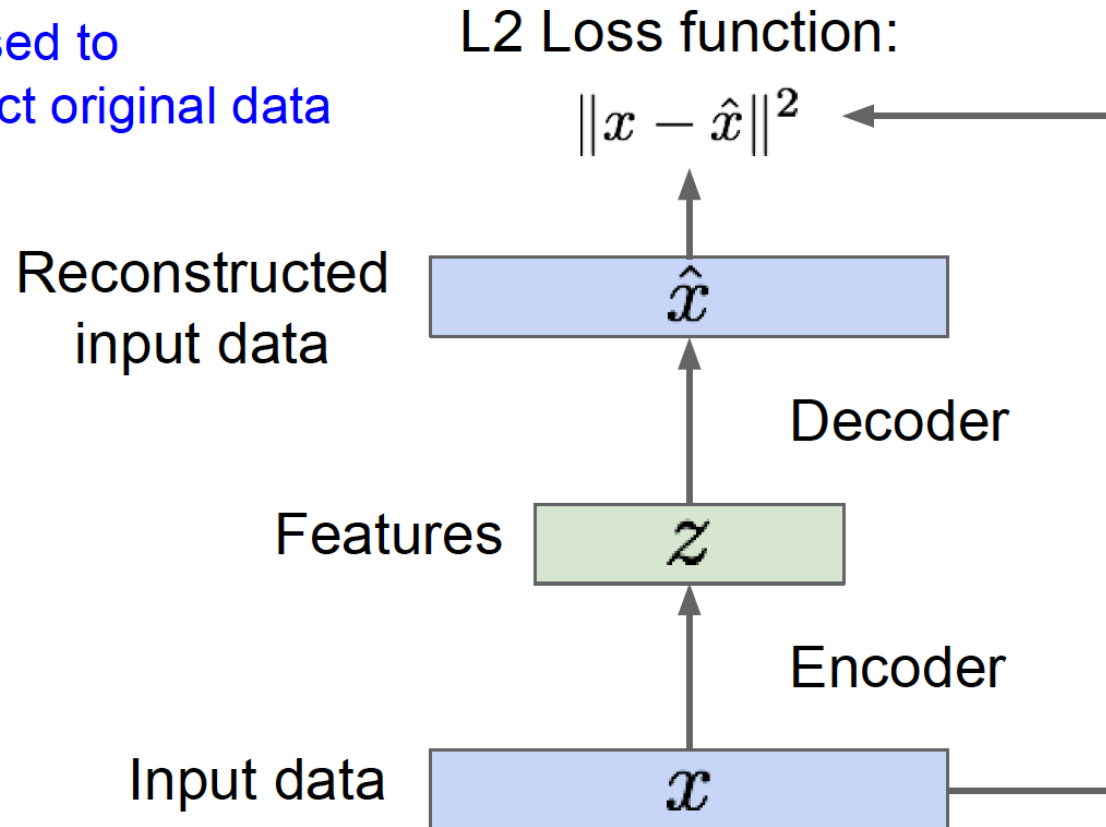
Unsupervised Learning: Autoencoders

Train such that features
can be used to
reconstruct original data

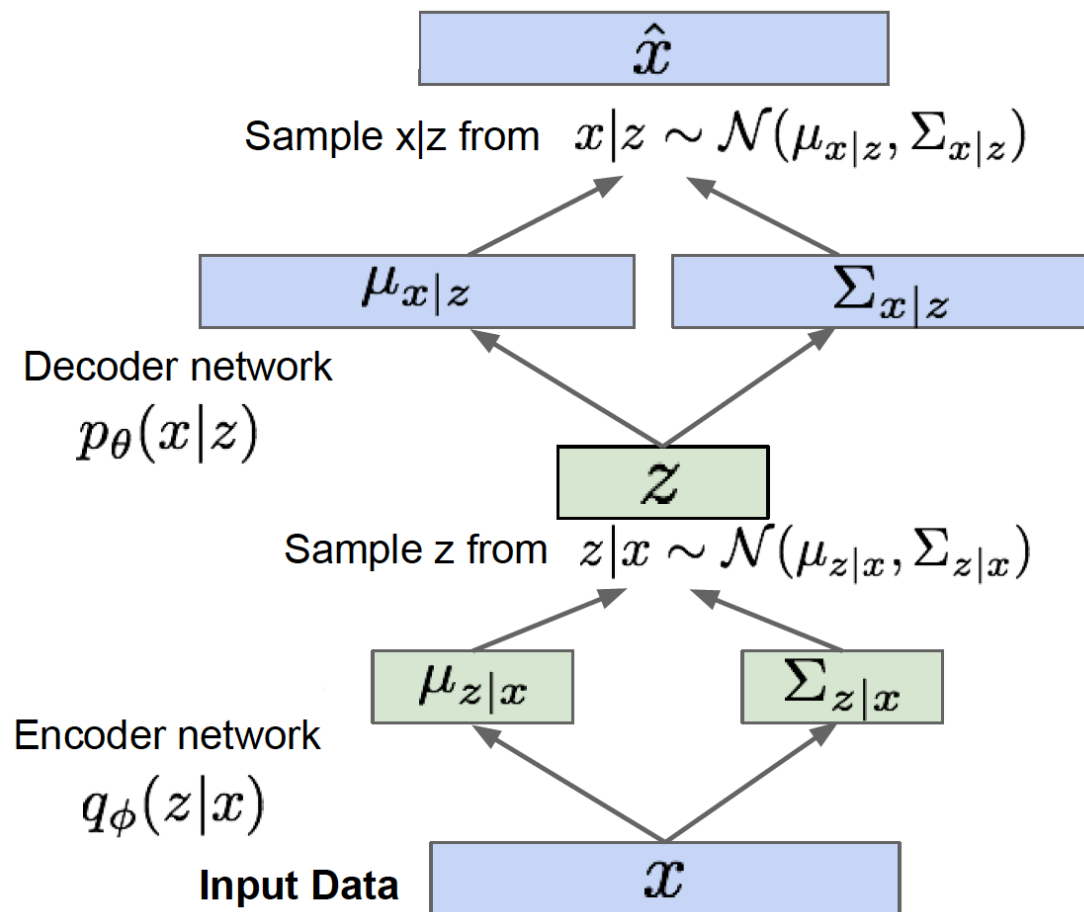


Unsupervised Learning: Autoencoders

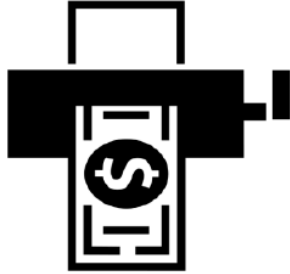
Train such that features
can be used to
reconstruct original data



Unsupervised Learning: Variational Autoencoders



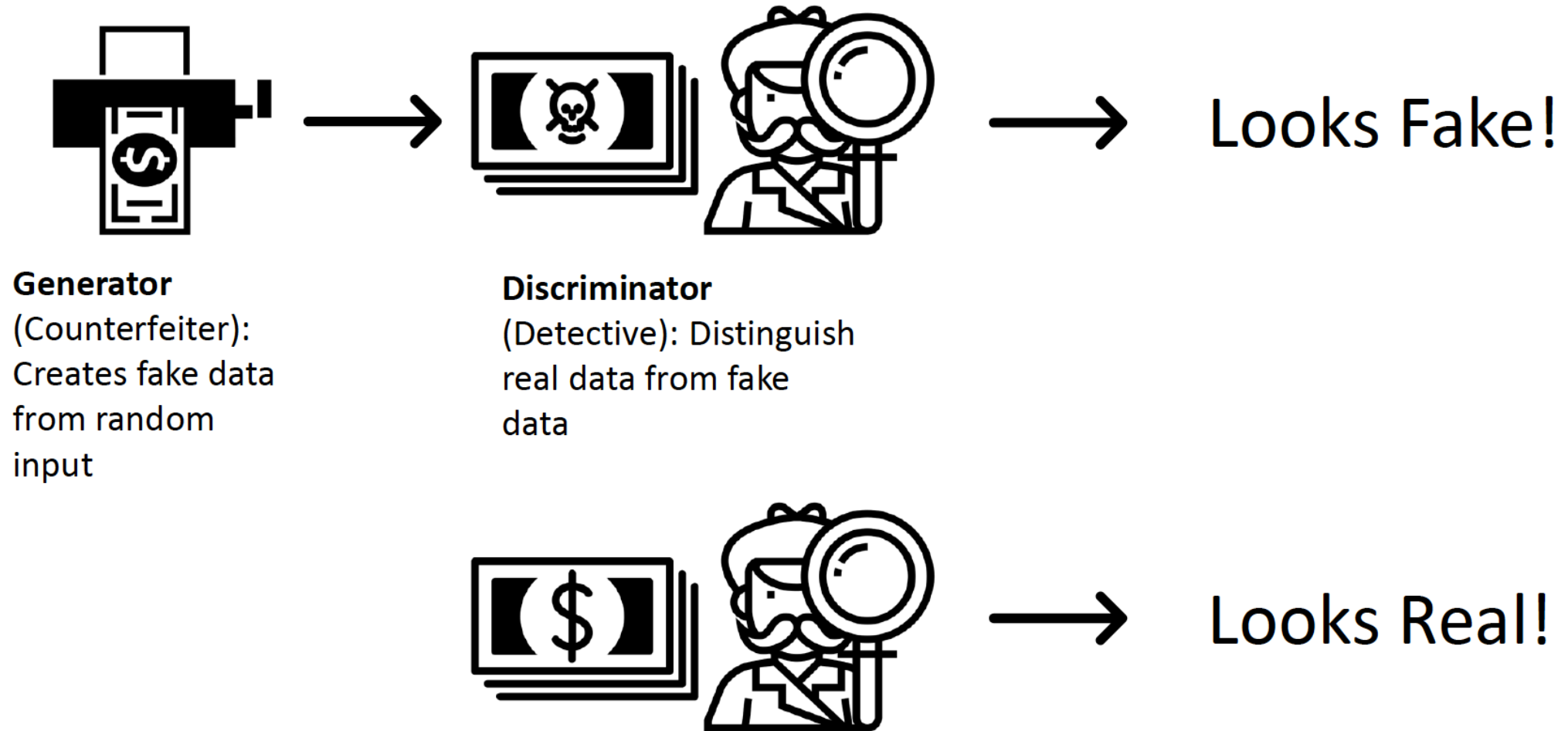
Generative Adversarial Networks: Idea



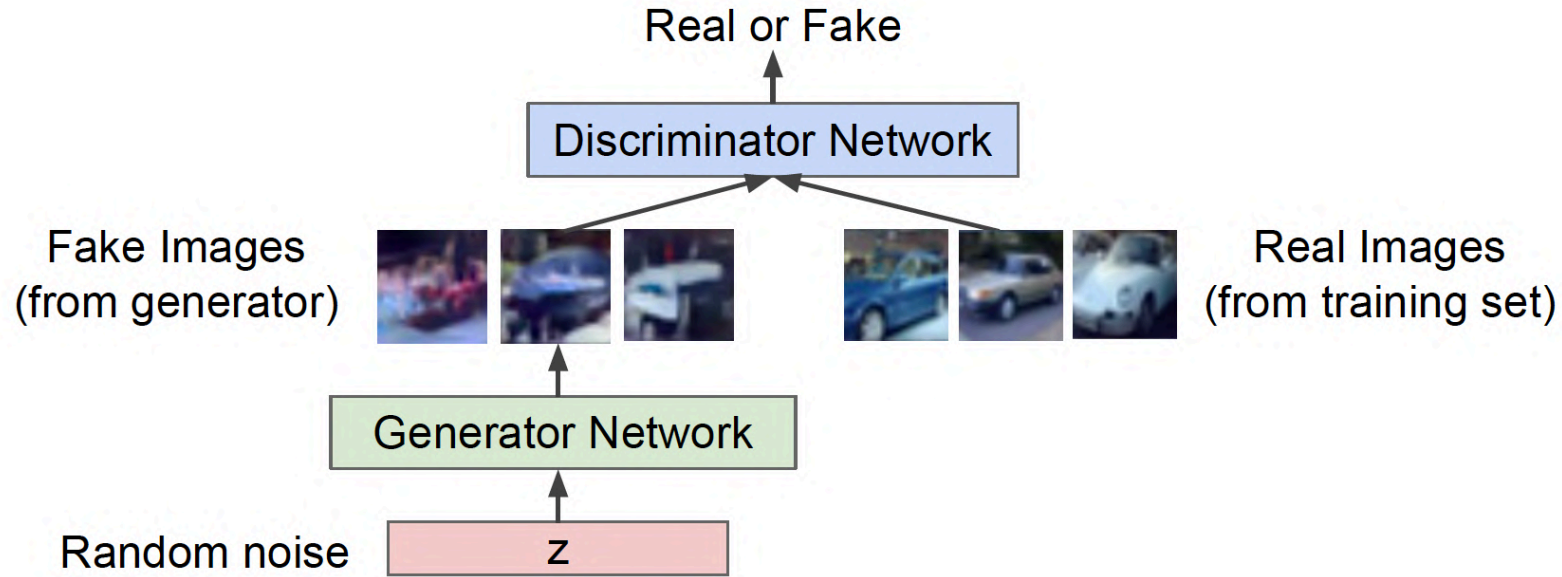
Generator

(Counterfeiter):
Creates fake data
from random
input

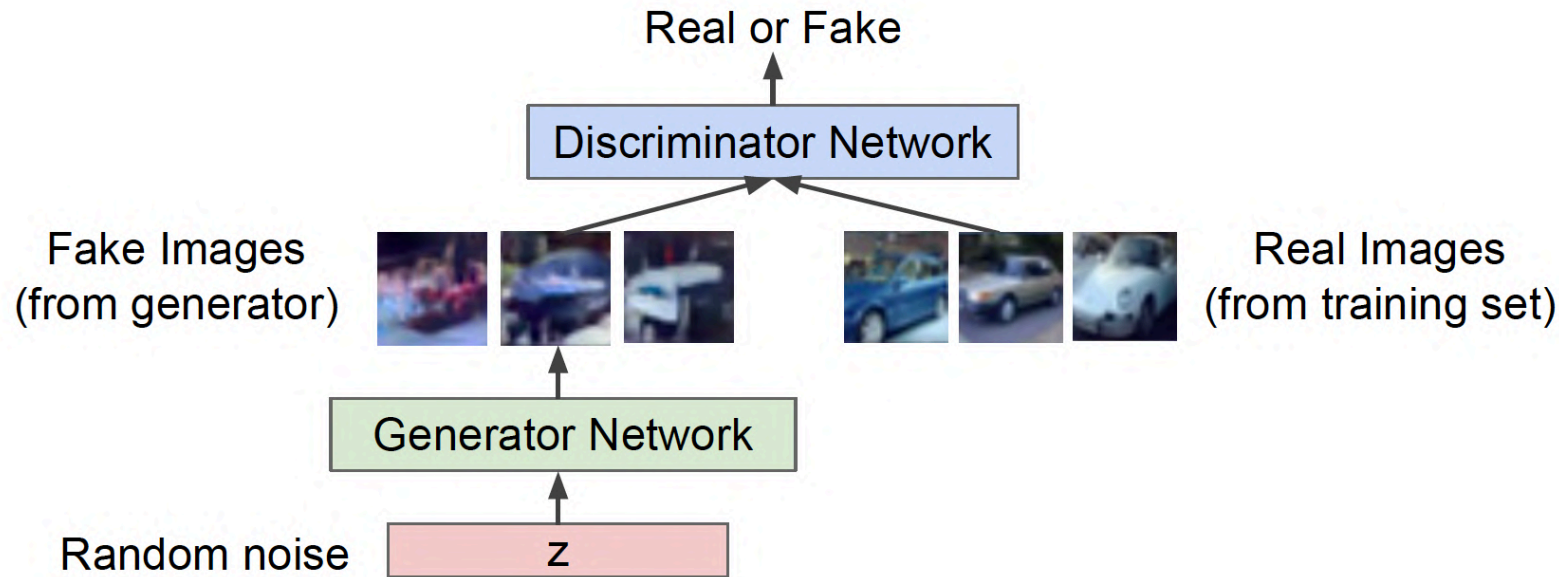
Generative Adversarial Networks: Idea



Generative Adversarial Networks



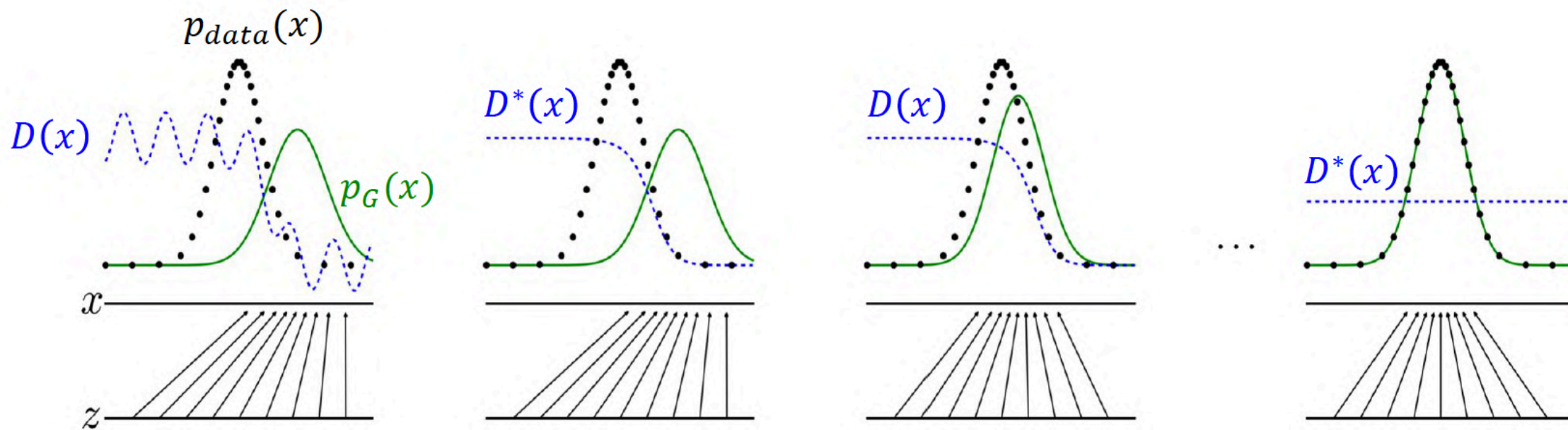
Generative Adversarial Networks



Minimax objective function:

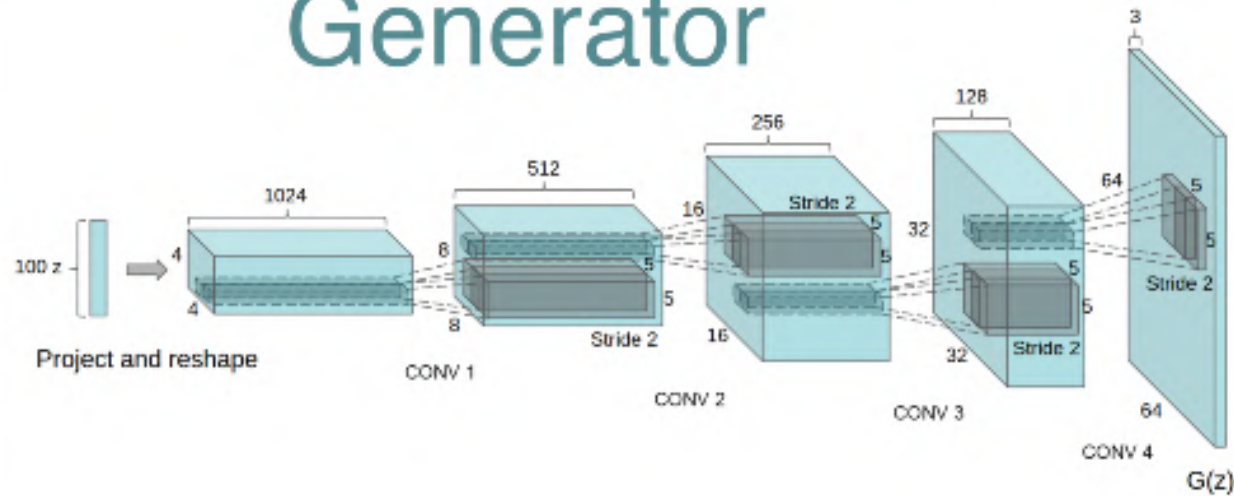
$$\min_{\theta_g} \max_{\theta_d} \left[\mathbb{E}_{x \sim p_{data}} \log \underbrace{D_{\theta_d}(x)}_{\text{Discriminator output for real data } x} + \mathbb{E}_{z \sim p(z)} \log(1 - \underbrace{D_{\theta_d}(G_{\theta_g}(z))}_{\text{Discriminator output for generated fake data } G(z)}) \right]$$

Distributions during training

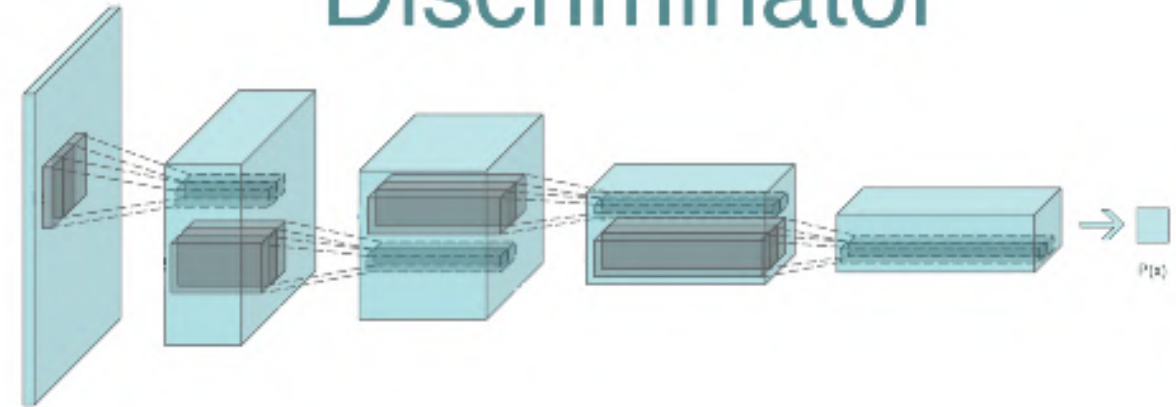


GAN: Sample Architecture (DC-GAN)

Generator

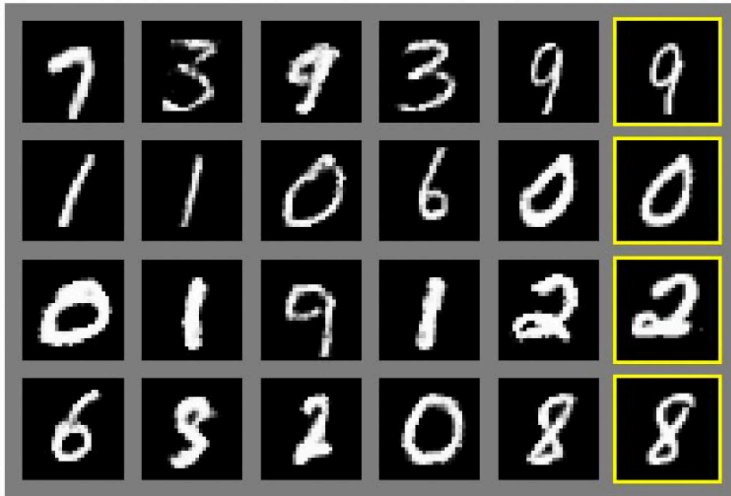


Discriminator



Generated Samples

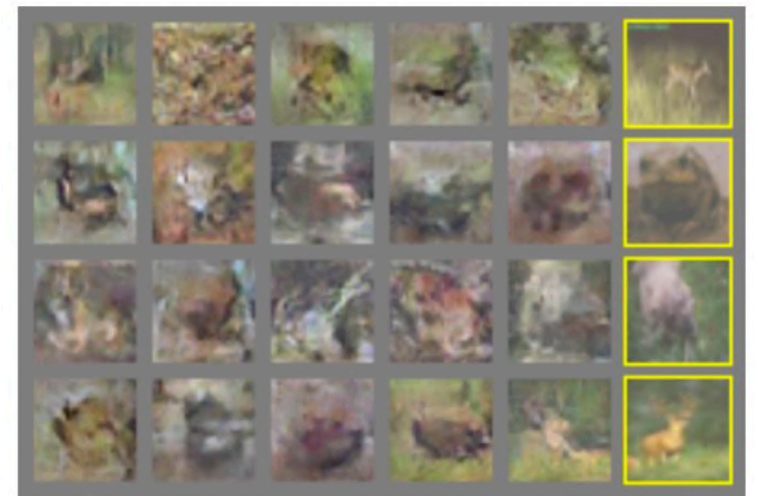
MNIST



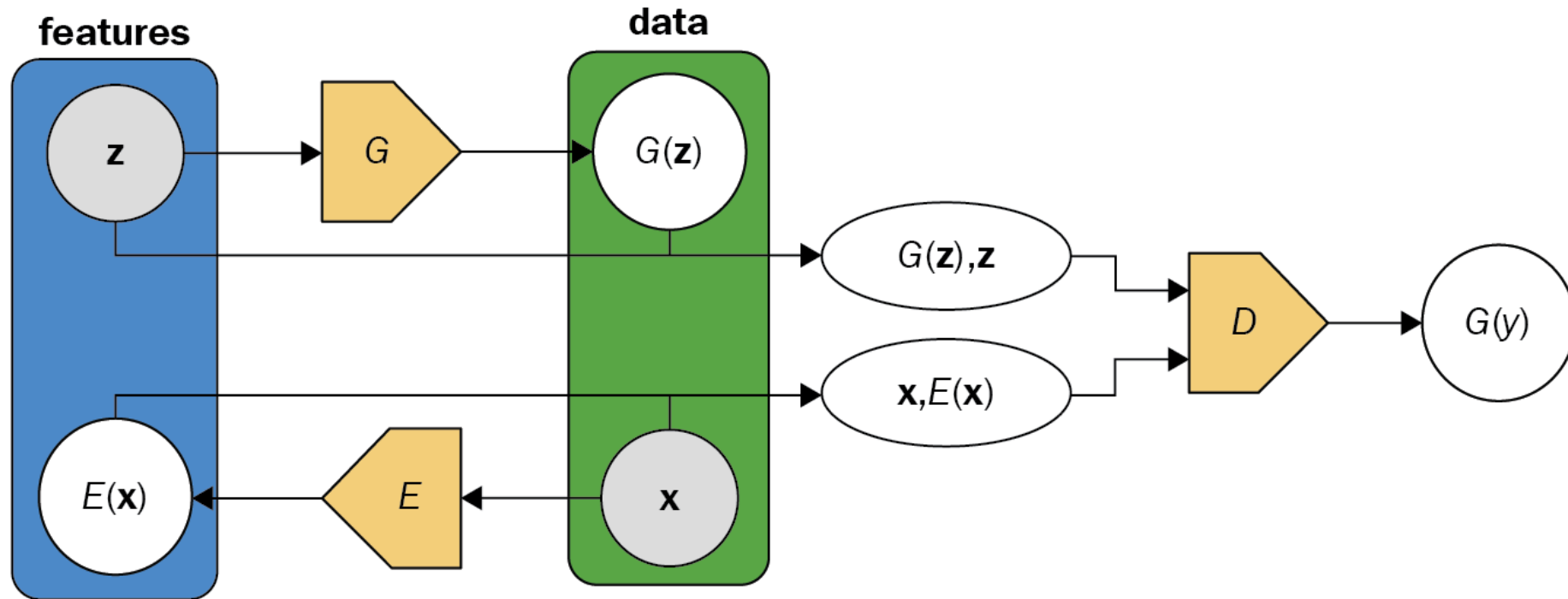
Faces



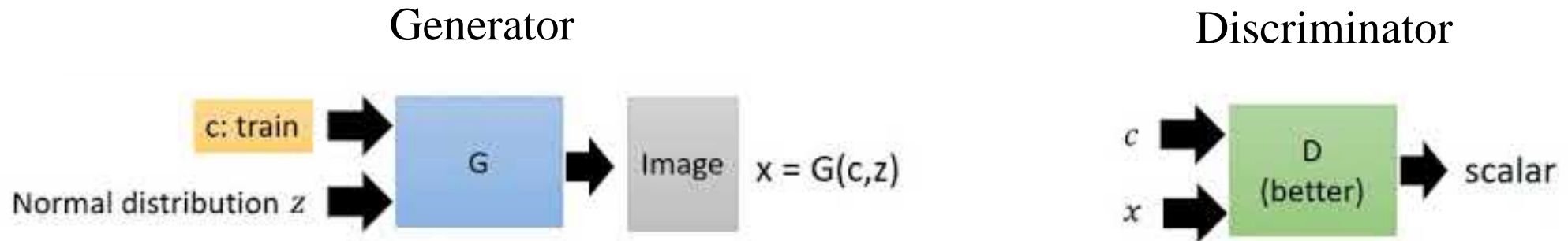
CIFAR 10



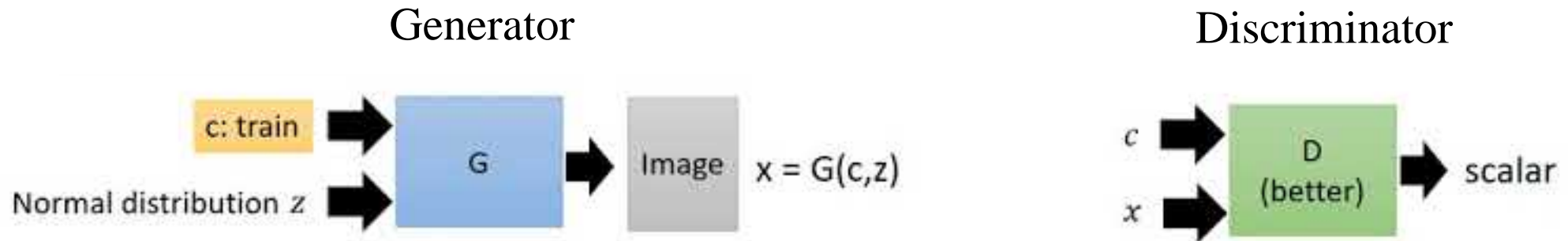
Bidirectional GAN (BiGAN)






Conditional GAN (cGAN)



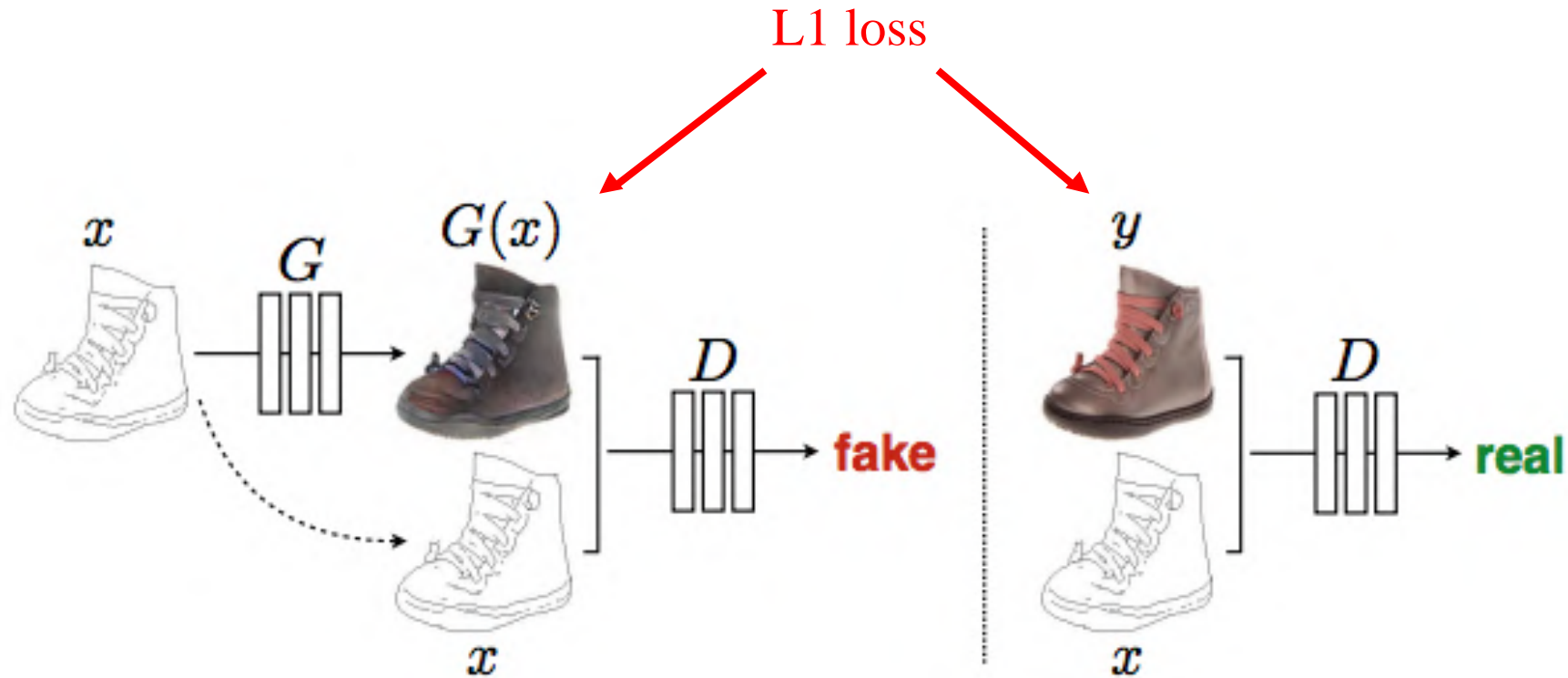
Conditional GAN (cGAN)



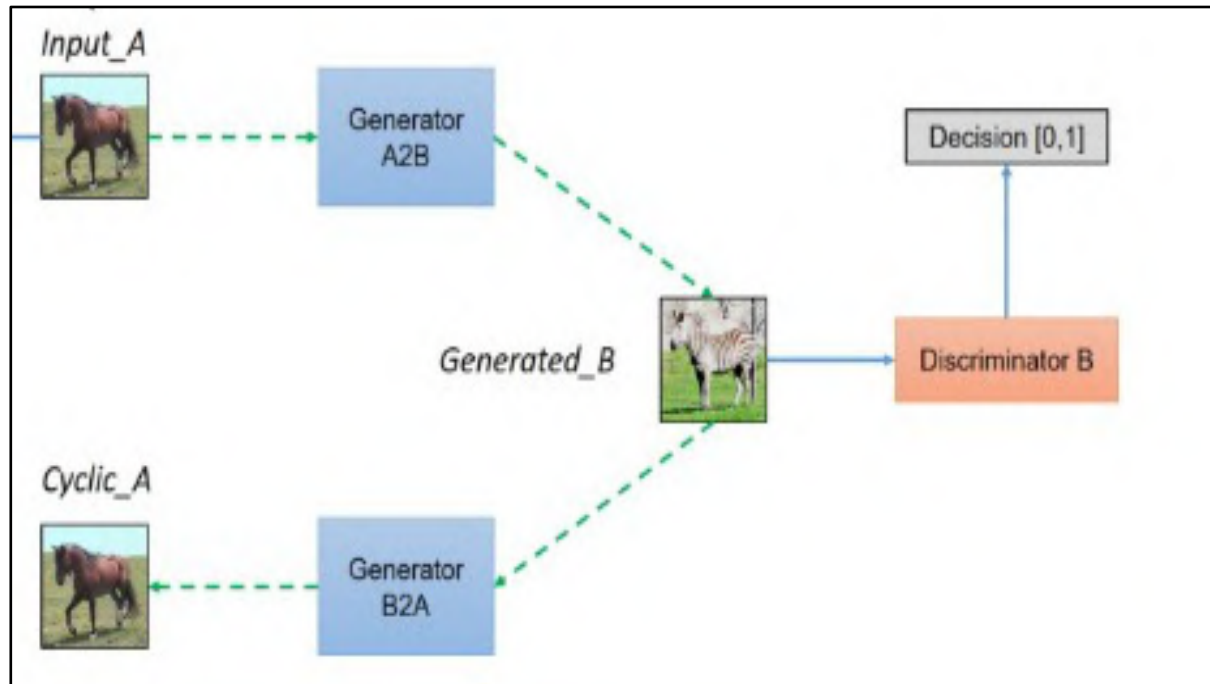
True text-image pairs: (train, ) 1

(cat, ) 0 (train, ) 0

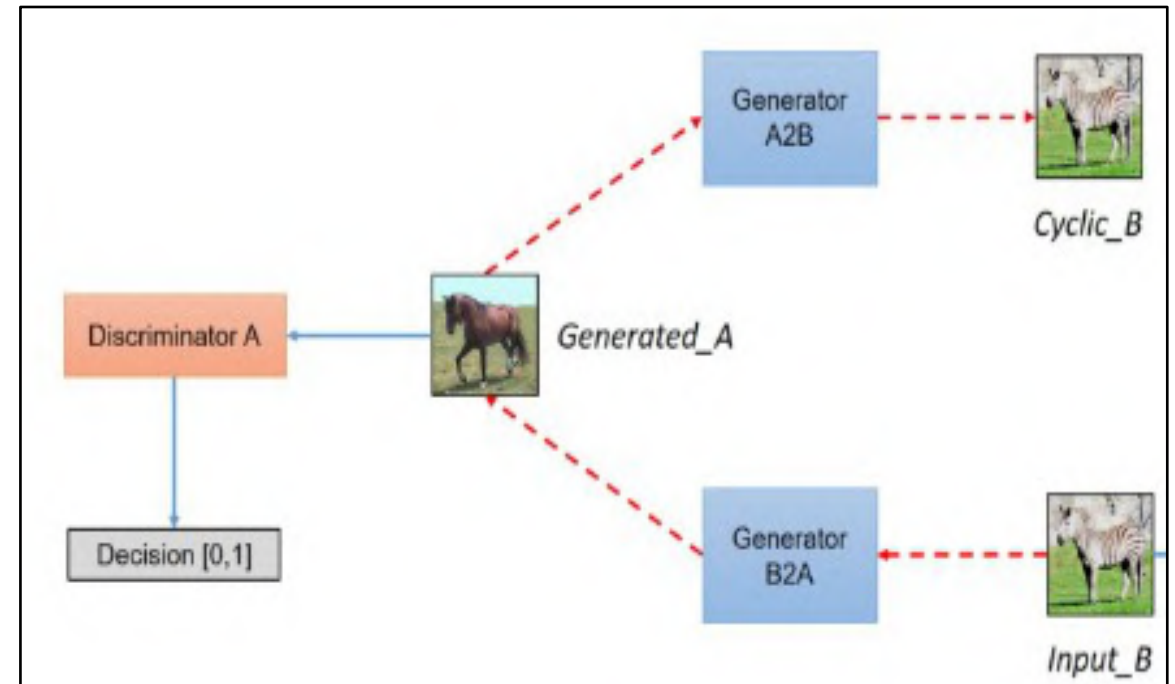
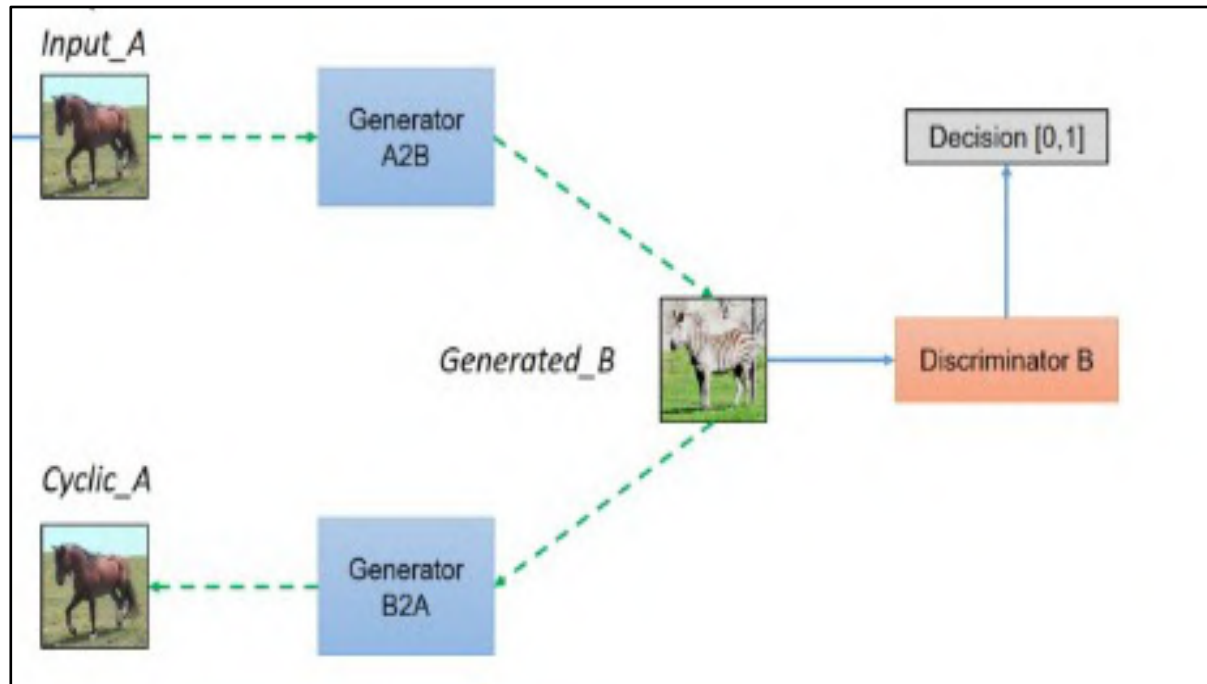
Pix2Pix: Type of cGAN



CycleGAN: Unsupervised Pix2Pix



CycleGAN: Unsupervised Pix2Pix



CycleGAN Results

Monet \leftrightarrow Photos



Monet \rightarrow photo



photo \rightarrow Monet

Zebras \leftrightarrow Horses



zebra \rightarrow horse



horse \rightarrow zebra

Summer \leftrightarrow Winter

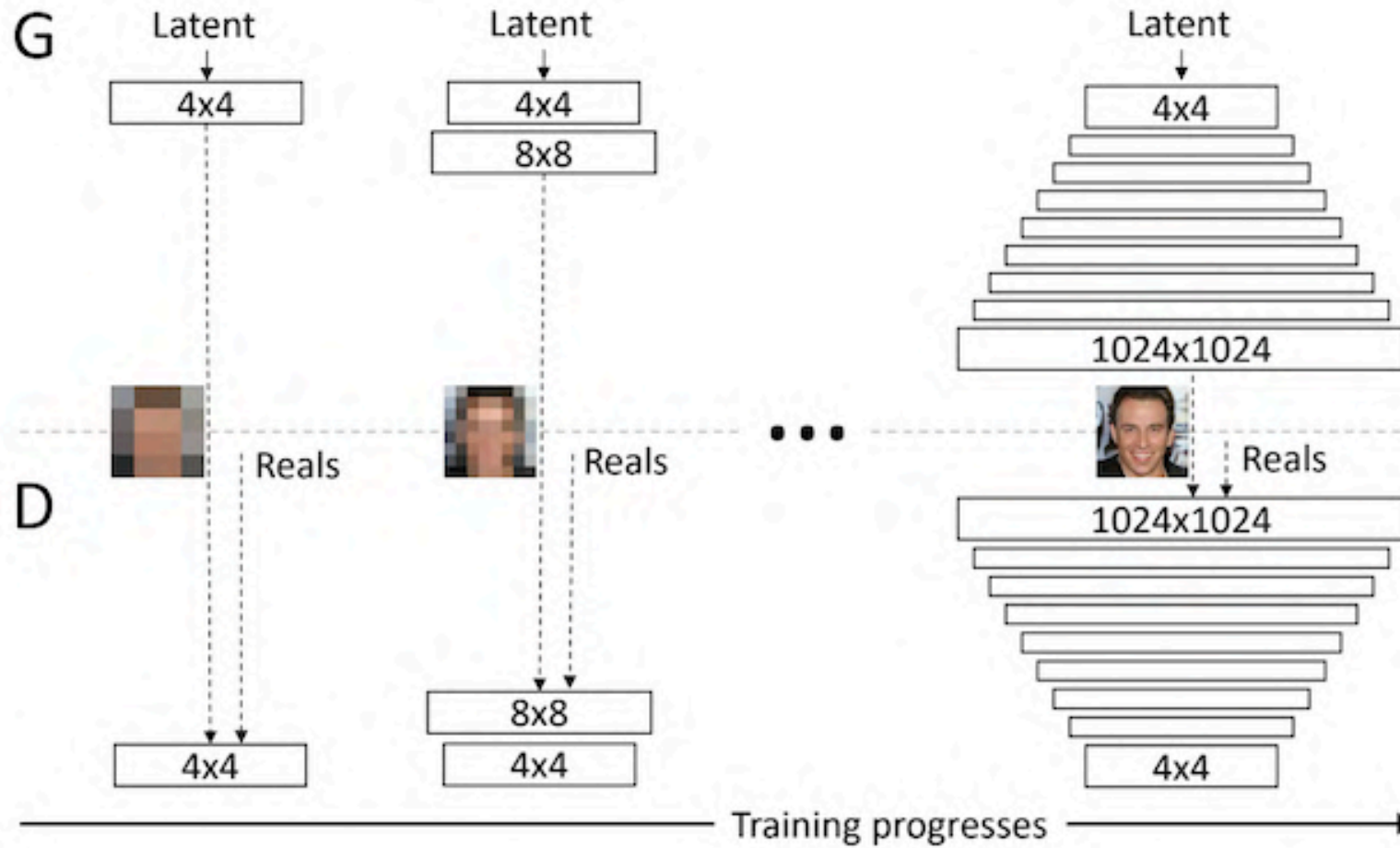


summer \rightarrow winter



winter \rightarrow summer

Progressive Growing of GANs



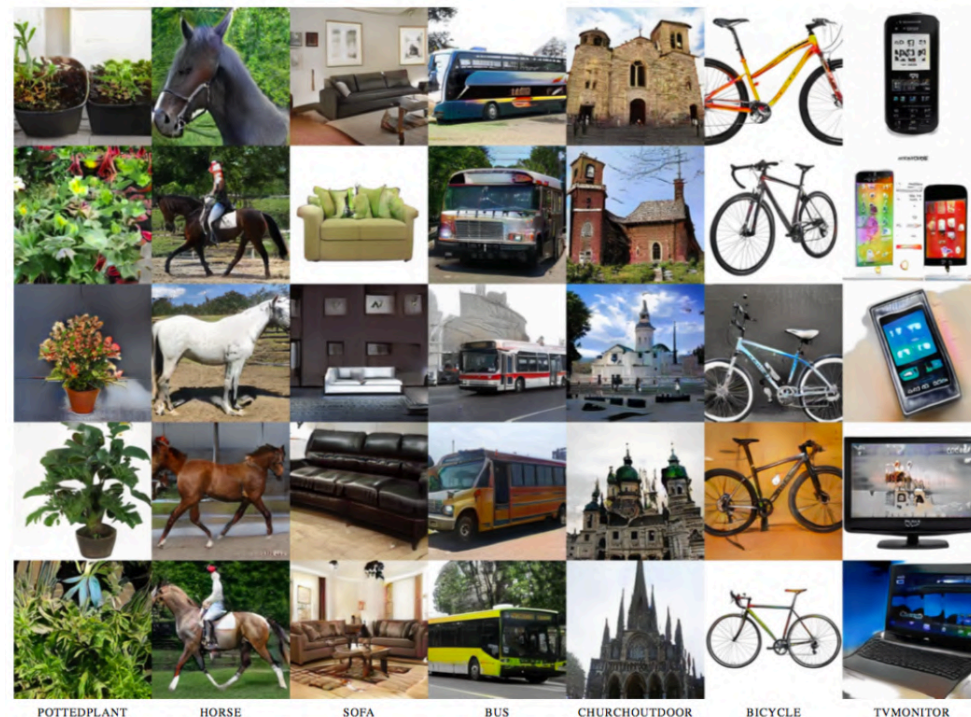
Progressive GAN Results



Celebrities

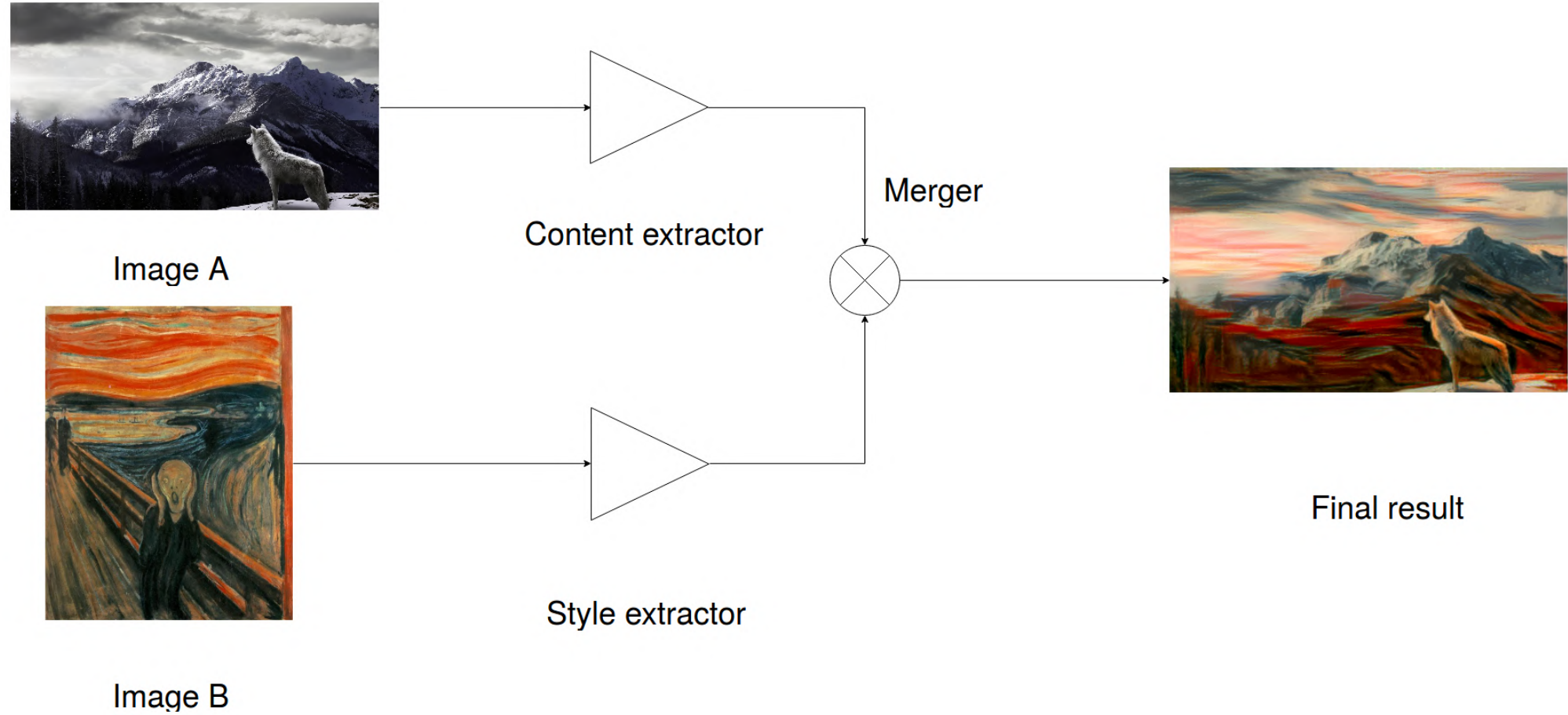


Bedrooms

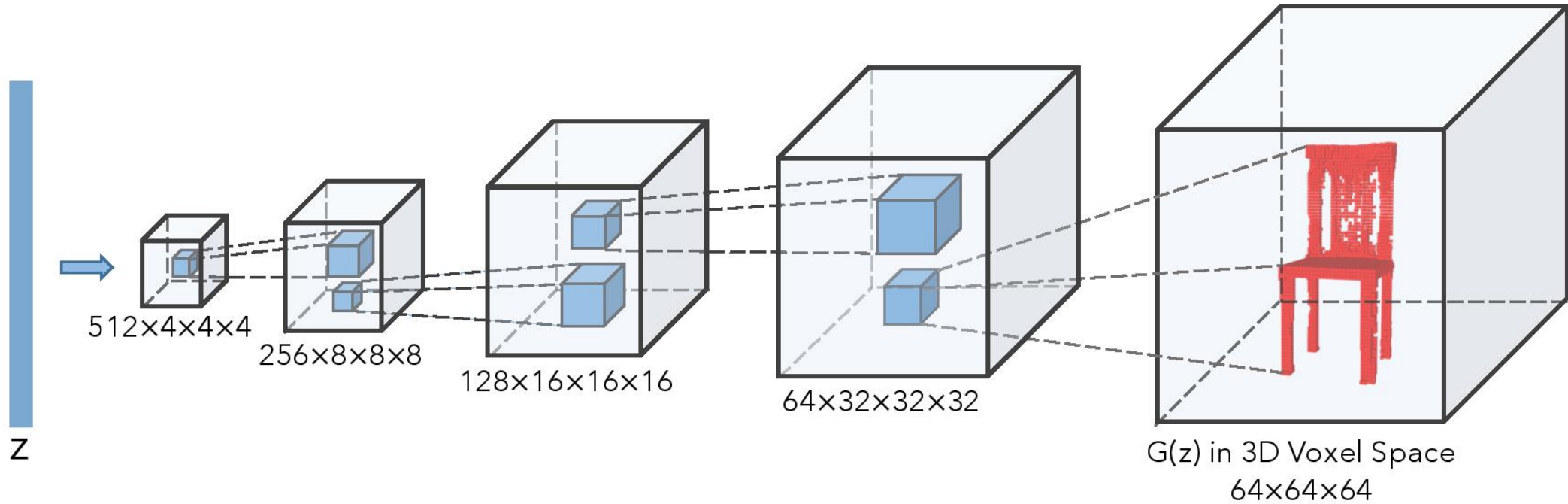


Objects

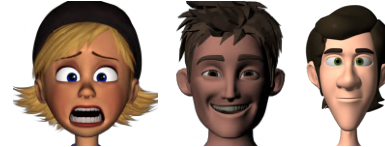
Application: Neural Style transfer



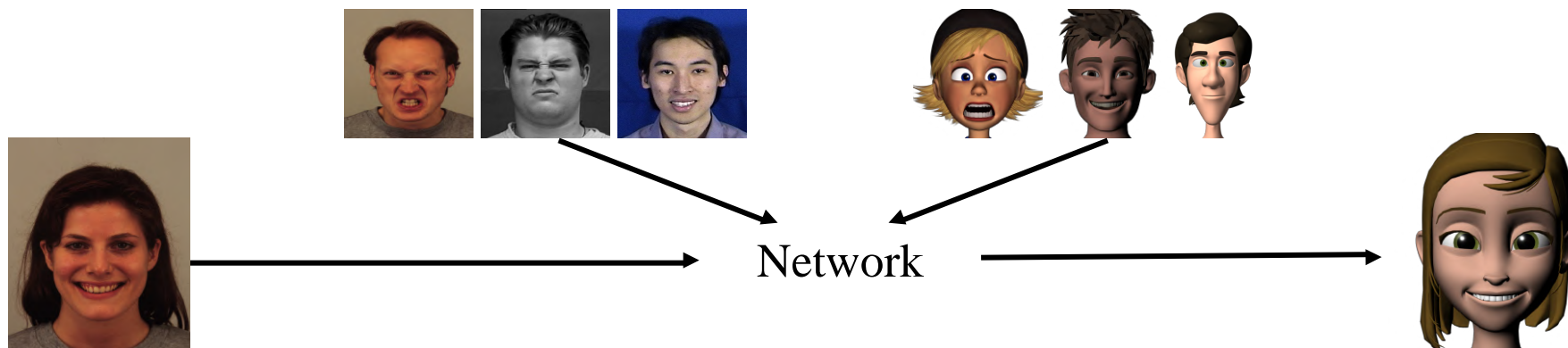
Application: 3D GAN



My Project: Facial Motion Retargeting

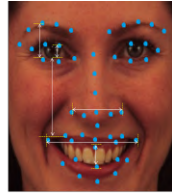


My Project: Facial Motion Retargeting



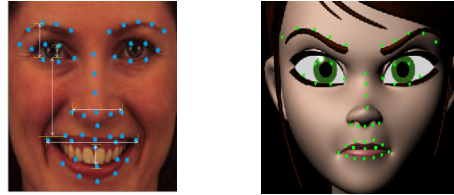
2D-to-3D CycleGAN

- Compute facial landmarks:



2D-to-3D CycleGAN

- Compute facial landmarks:

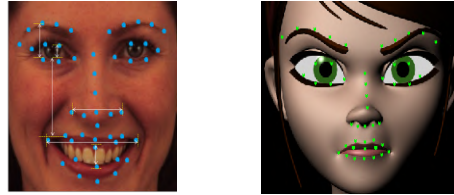


- Convert 3D model to 2D position map:



2D-to-3D CycleGAN

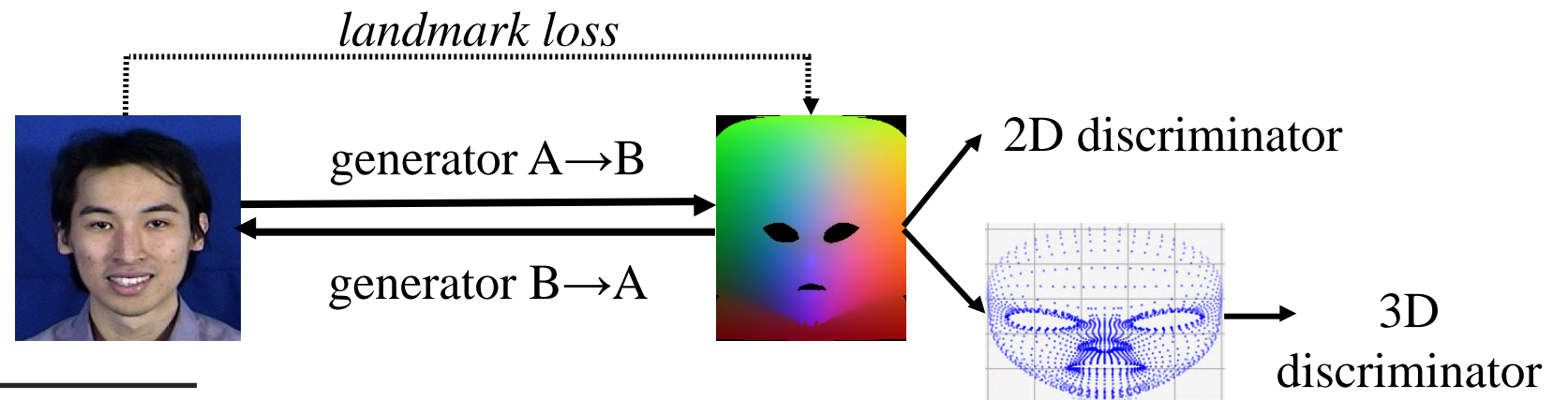
- Compute facial landmarks:



- Convert 3D model to 2D position map:

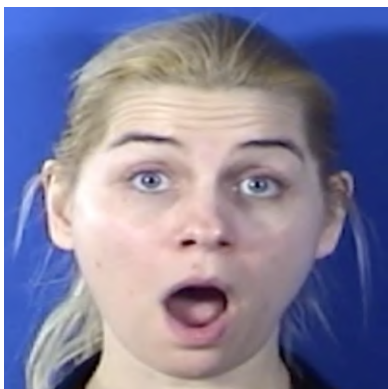
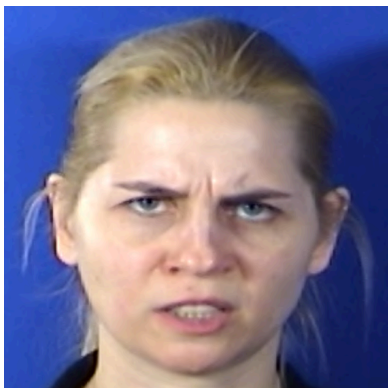


- Train CycleGAN:



Results

Input



Landmark only

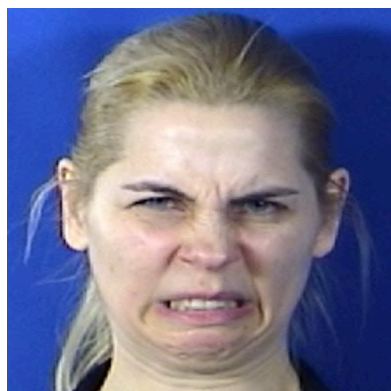


CycleGAN



Results

Input



Landmark only



CycleGAN



Useful links

- GAN Zoo: <https://github.com/hindupuravinash/the-gan-zoo>
- GAN hacks: <https://github.com/soumith/ganhacks>
- **Code Bases:**
 - Tensorflow: <https://www.tensorflow.org/tutorials/generative/dcgan>
 - Keras: <https://github.com/eriklindernoren/Keras-GAN>
 - Pytorch: <https://github.com/pytorch/examples/tree/master/dcgan>
- **References:**
 - http://cs231n.stanford.edu/slides/2017/cs231n_2017_lecture13.pdf
 - https://www.cs.toronto.edu/~rgrosse/courses/csc321_2018/slides/lec19.pdf
 - <https://www.cs.cmu.edu/~bhiksha/courses/deeplearning/Fall.2015/slides/lec13.GAN.pdf>

Questions

