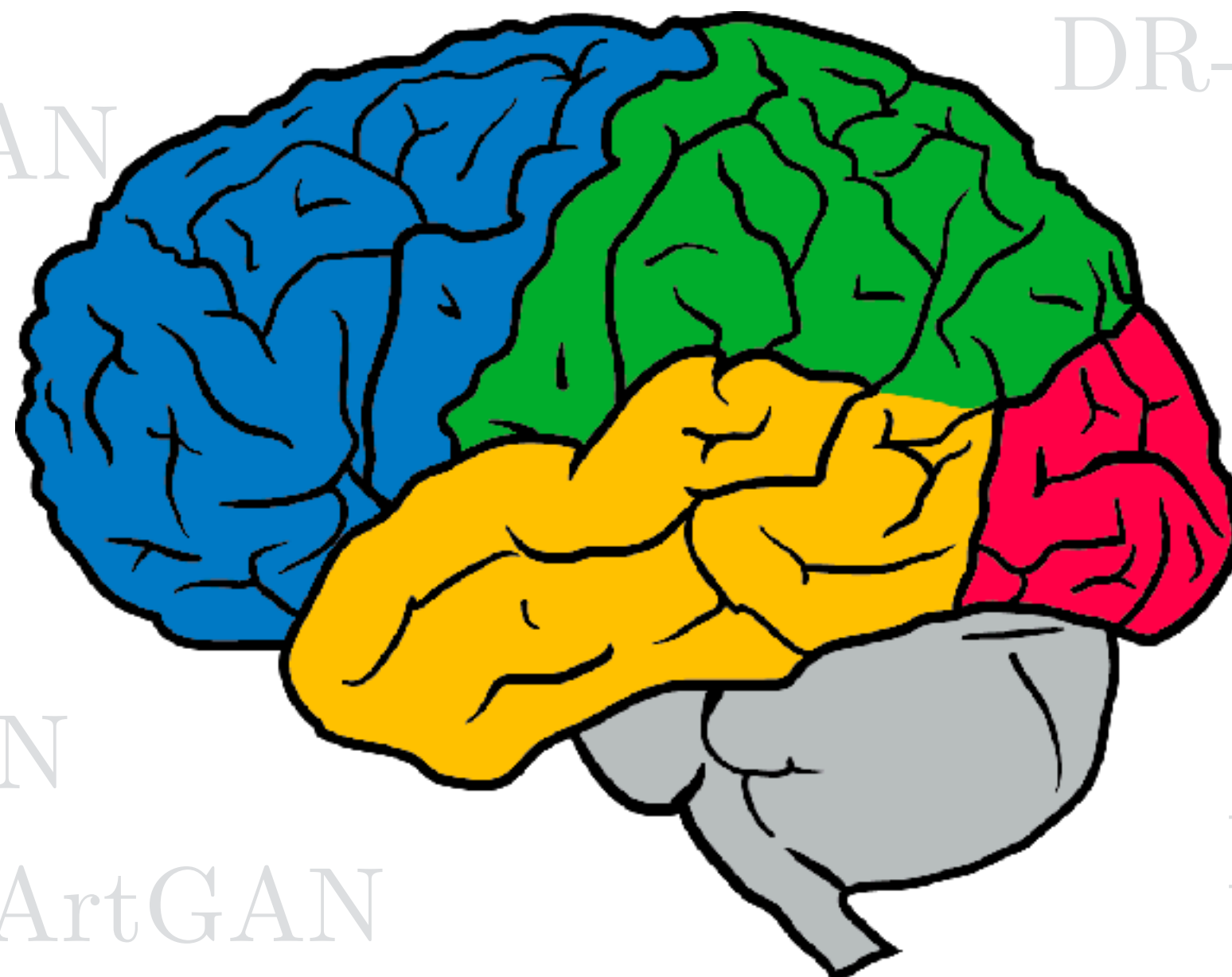


Introduction to GANs

Ian Goodfellow, Staff Research Scientist, Google Brain

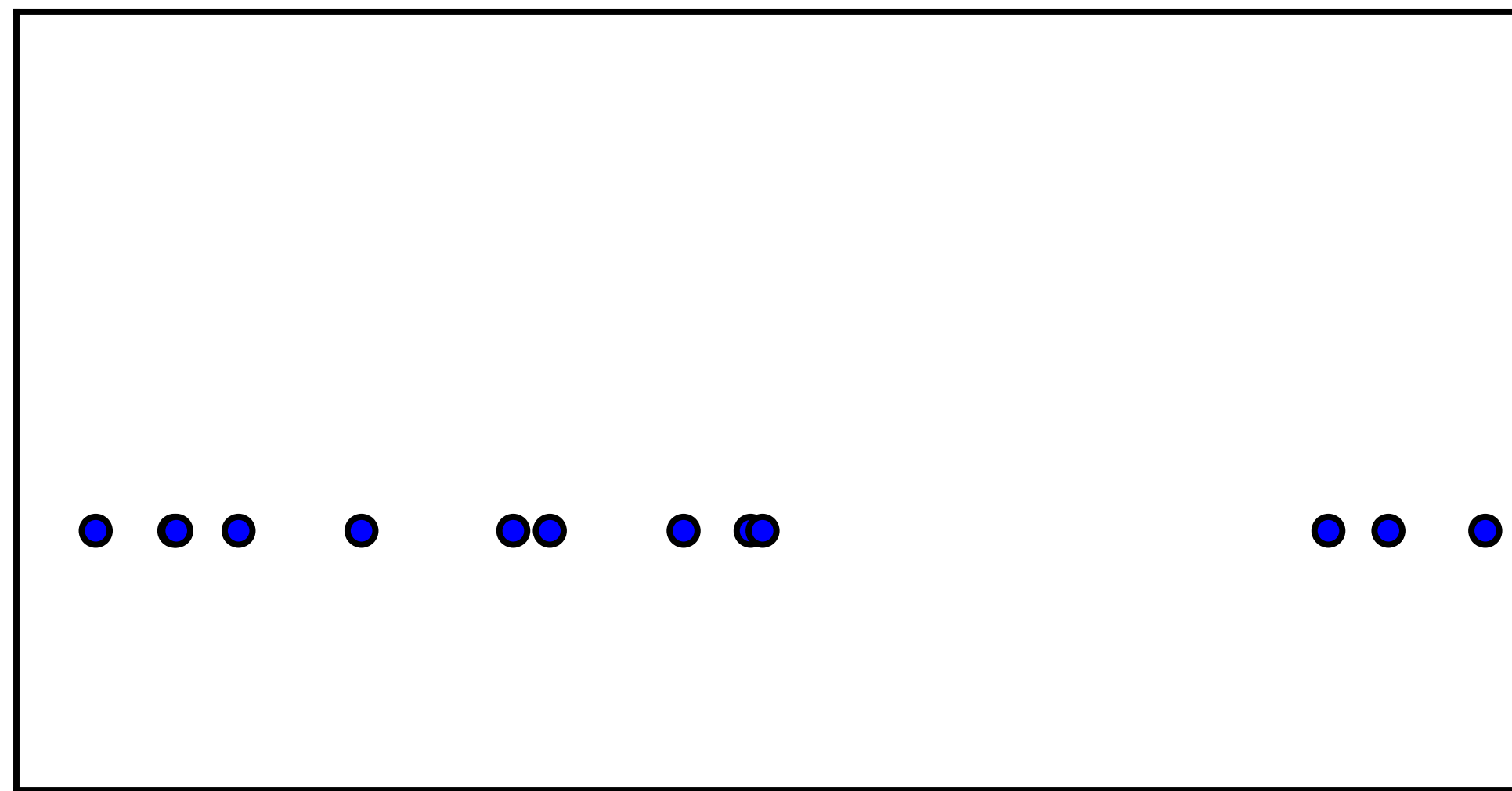
CVPR Tutorial on GANs

Salt Lake City, 2018-06-22

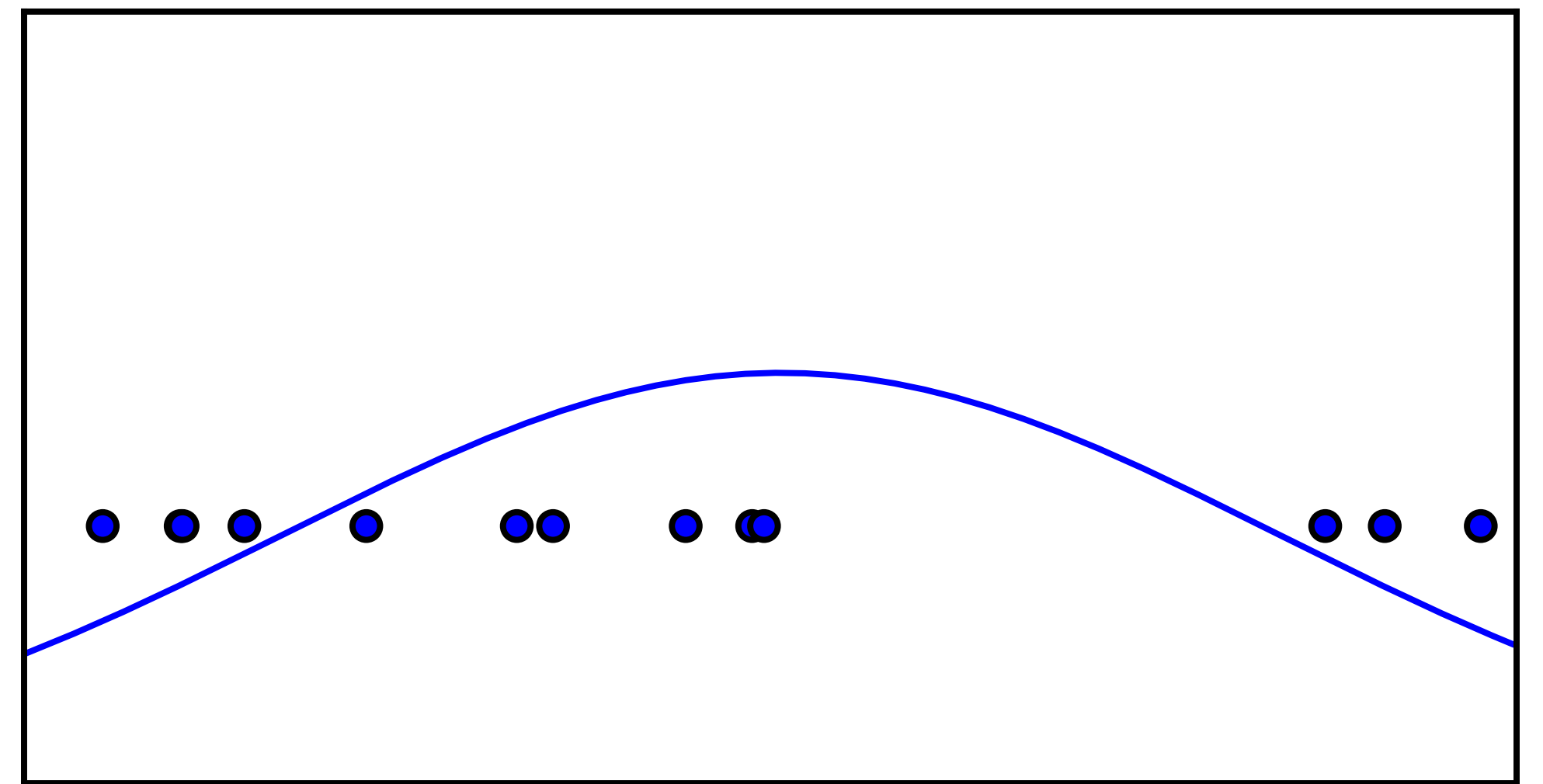


MedGAN ID-CGAN CoGAN LR-GAN CGAN IcGAN
b-GAN LS-GAN LAPGAN DiscoGAN MPM-GAN AdaGAN
InfoGAN CatGAN AMGAN iGAN IAN
LSGAN SAGAN
McGAN
MGAN BS-GAN
FF-GAN
C-VAE-GAN C-RNN-GAN GoGAN
MAGAN 3D-GAN CCGAN DR-GAN DCGAN
GAWWN DualGAN BiGAN
Bayesian GAN GP-GAN
EBGAN AnoGAN DTN
ALI Context-RNN-GAN MAD-GAN
MARTA-GAN f-GAN ArtGAN BEGAN AL-CGAN
MalGAN

Generative Modeling: Density Estimation

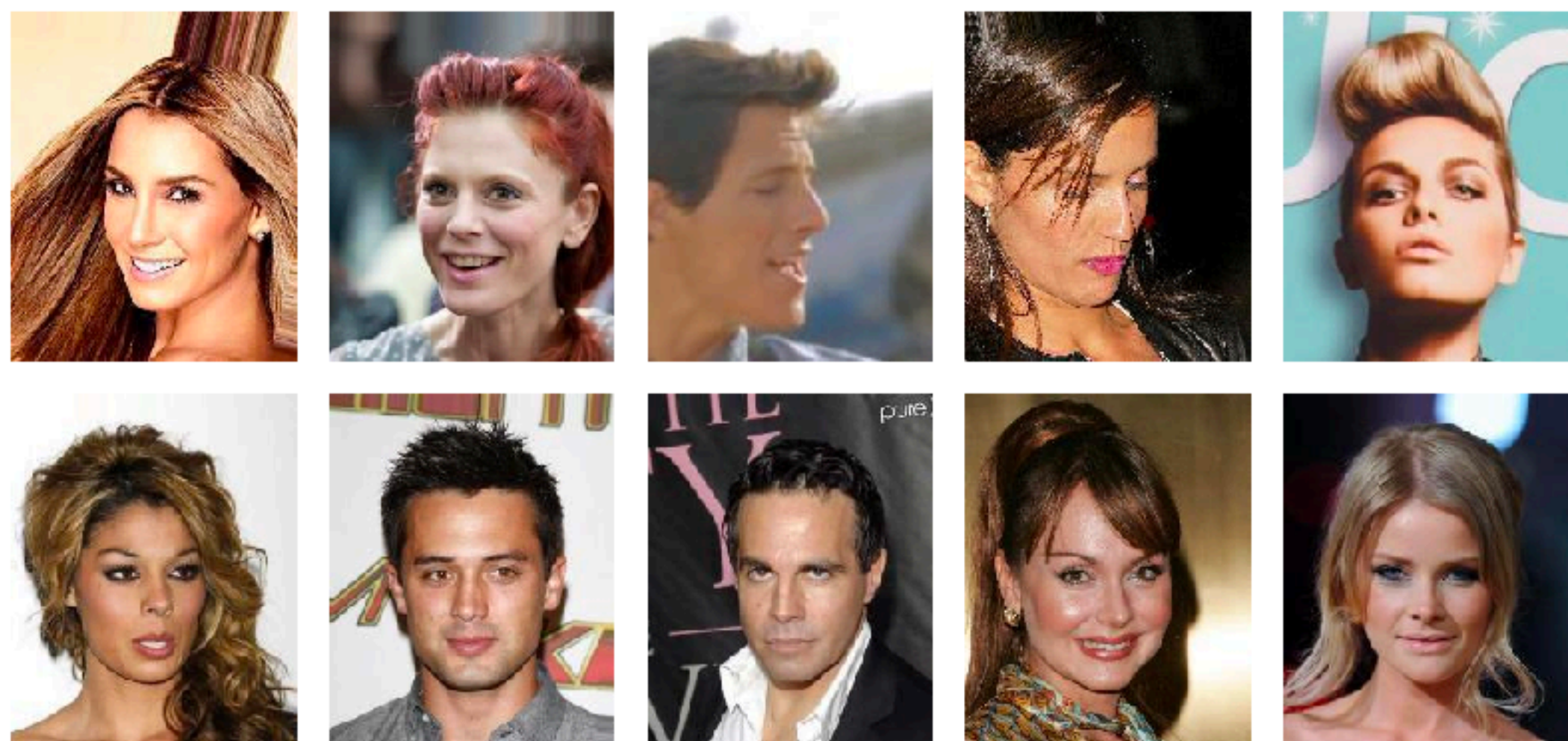


Training Data



Density Function

Generative Modeling: Sample Generation

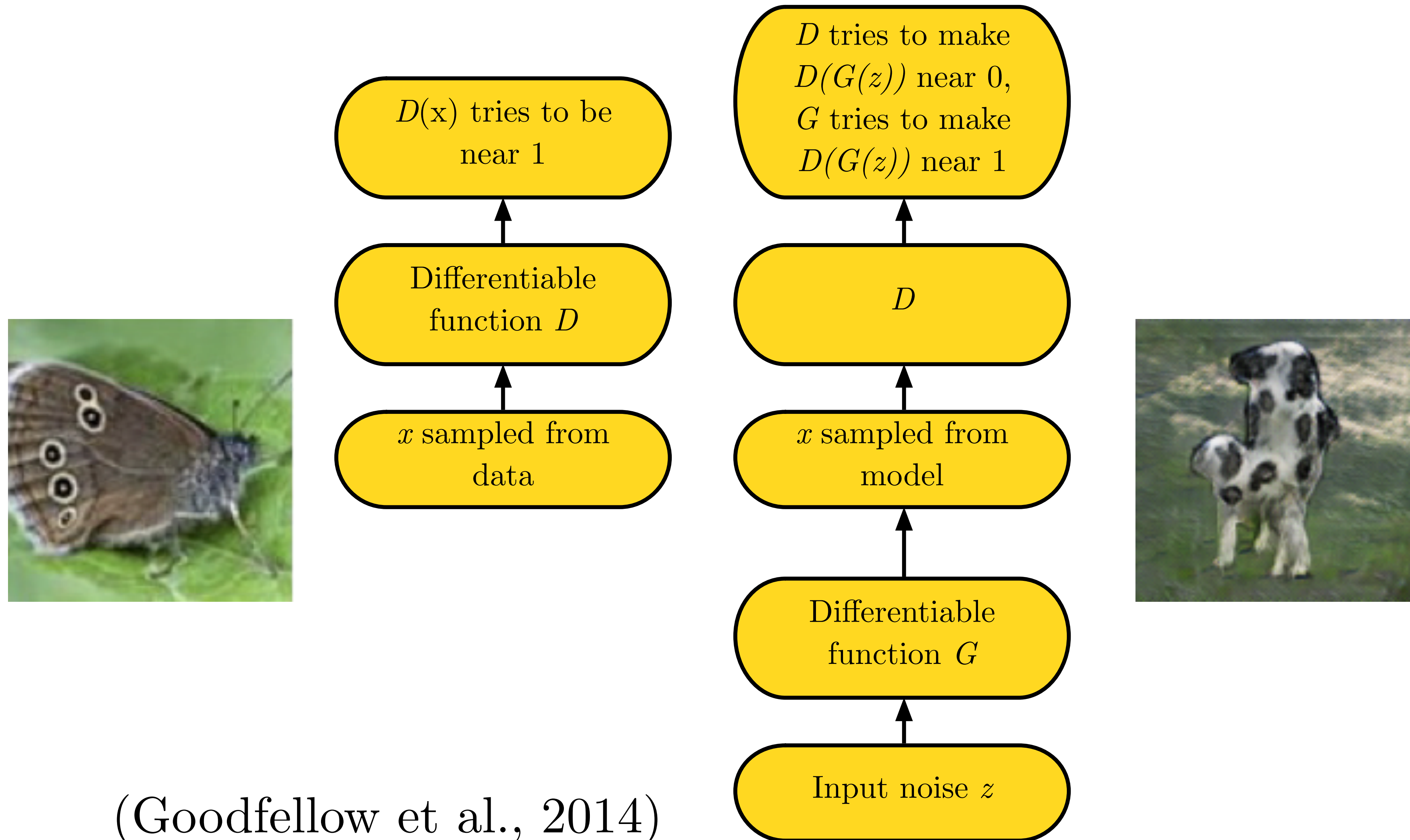


Training Data
(CelebA)



Sample Generator
(Karras et al, 2017)

Adversarial Nets Framework



Self-Play

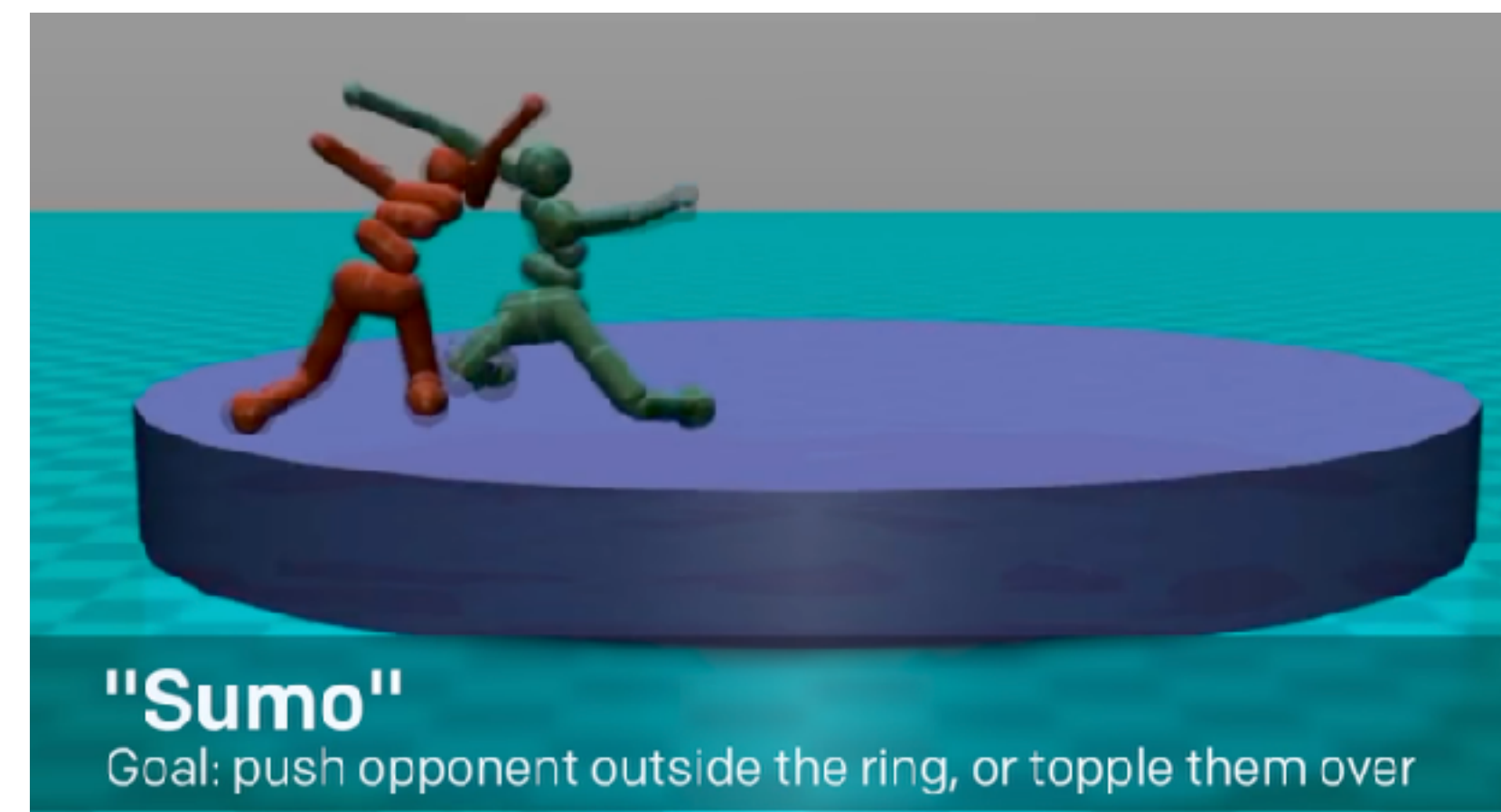
1959: Arthur Samuel's checkers agent



(Silver et al, 2017)



(OpenAI, 2017)



(Bansal et al, 2017)

3.5 Years of Progress on Faces



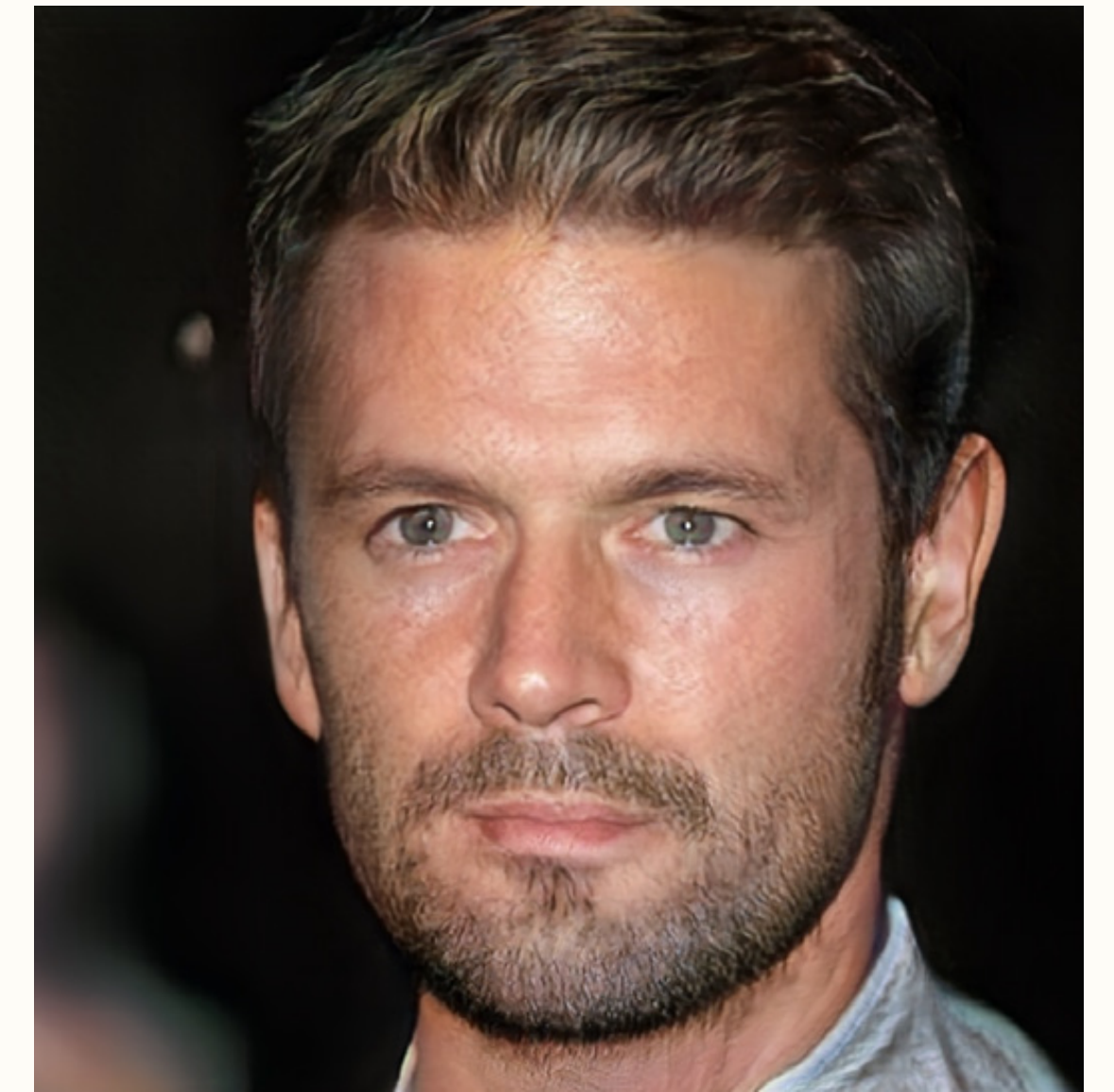
2014



2015



2016

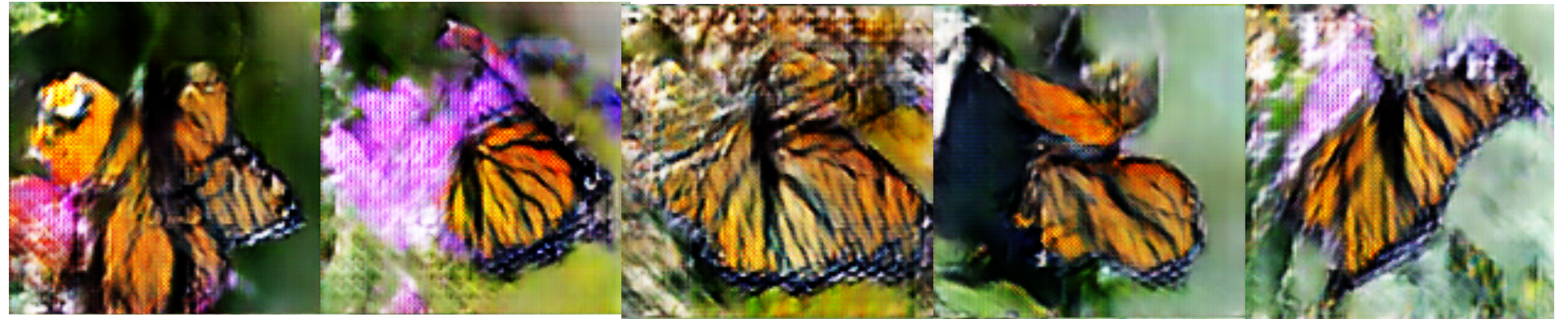


2017

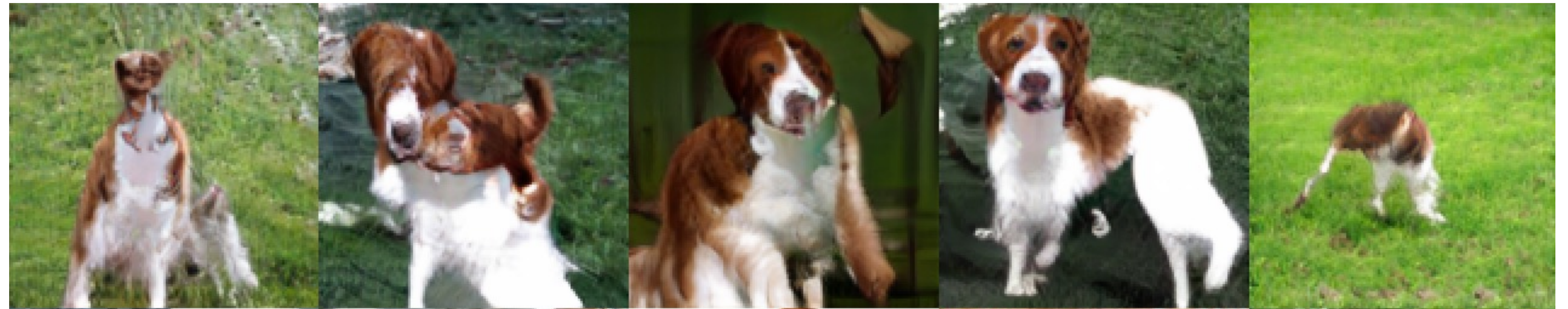
(Brundage et al, 2018)

<2 Years of Progress on ImageNet

Odena et al
2016



Miyato et al
2017



Zhang et al
2018



Self-Attention GAN

State of the art FID on ImageNet: 1000 categories, 128x128 pixels



Goldfish



Redshank



Broccoli



Tiger Cat



Geyser



Indigo Bunting



Stone Wall



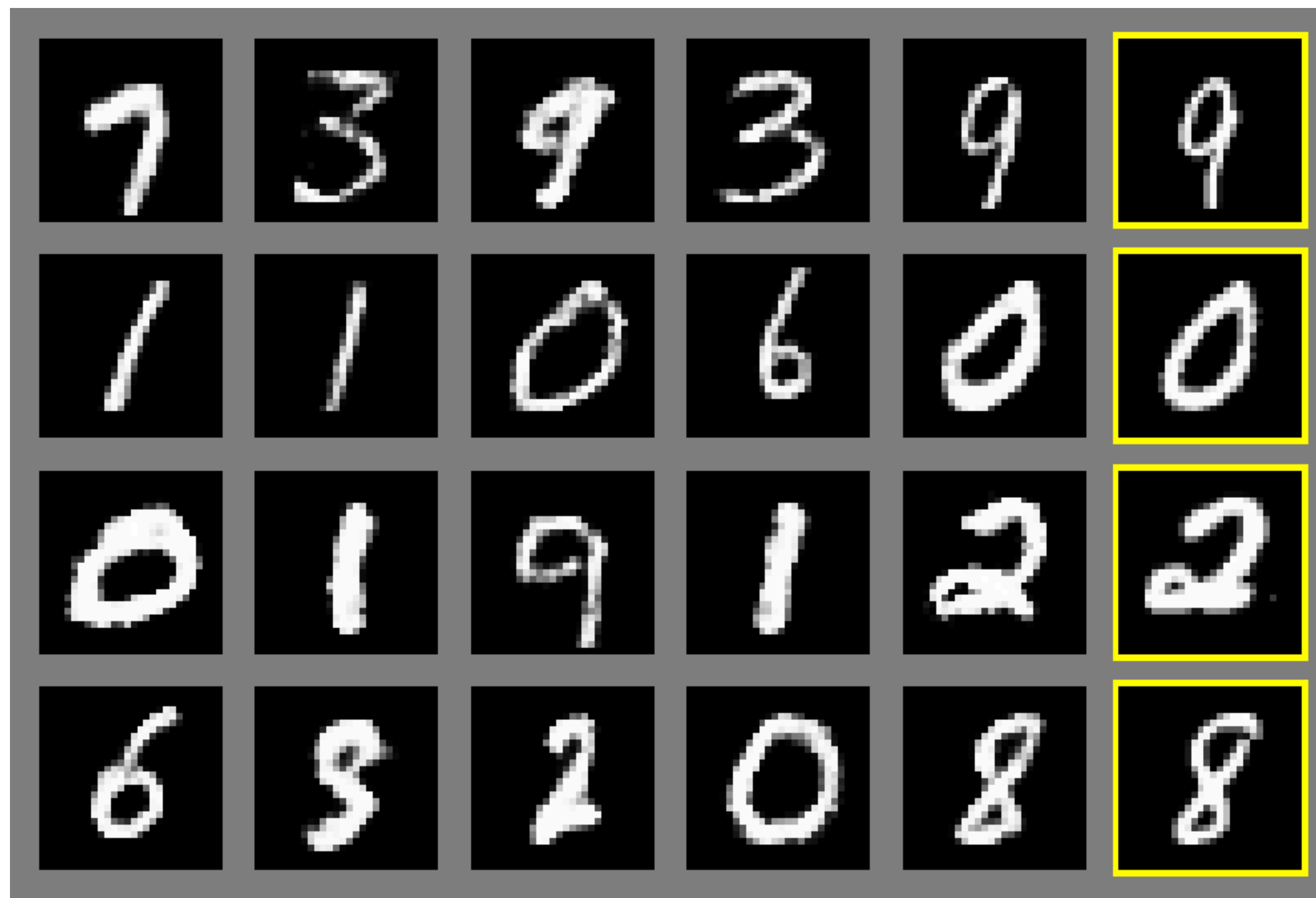
Saint Bernard

(Zhang et al., 2018)

From GAN to SAGAN

- Depth and Convolution
- Class-conditional generation
- Spectral Normalization
- Hinge loss
- Two-timescale update rule
- Self-attention

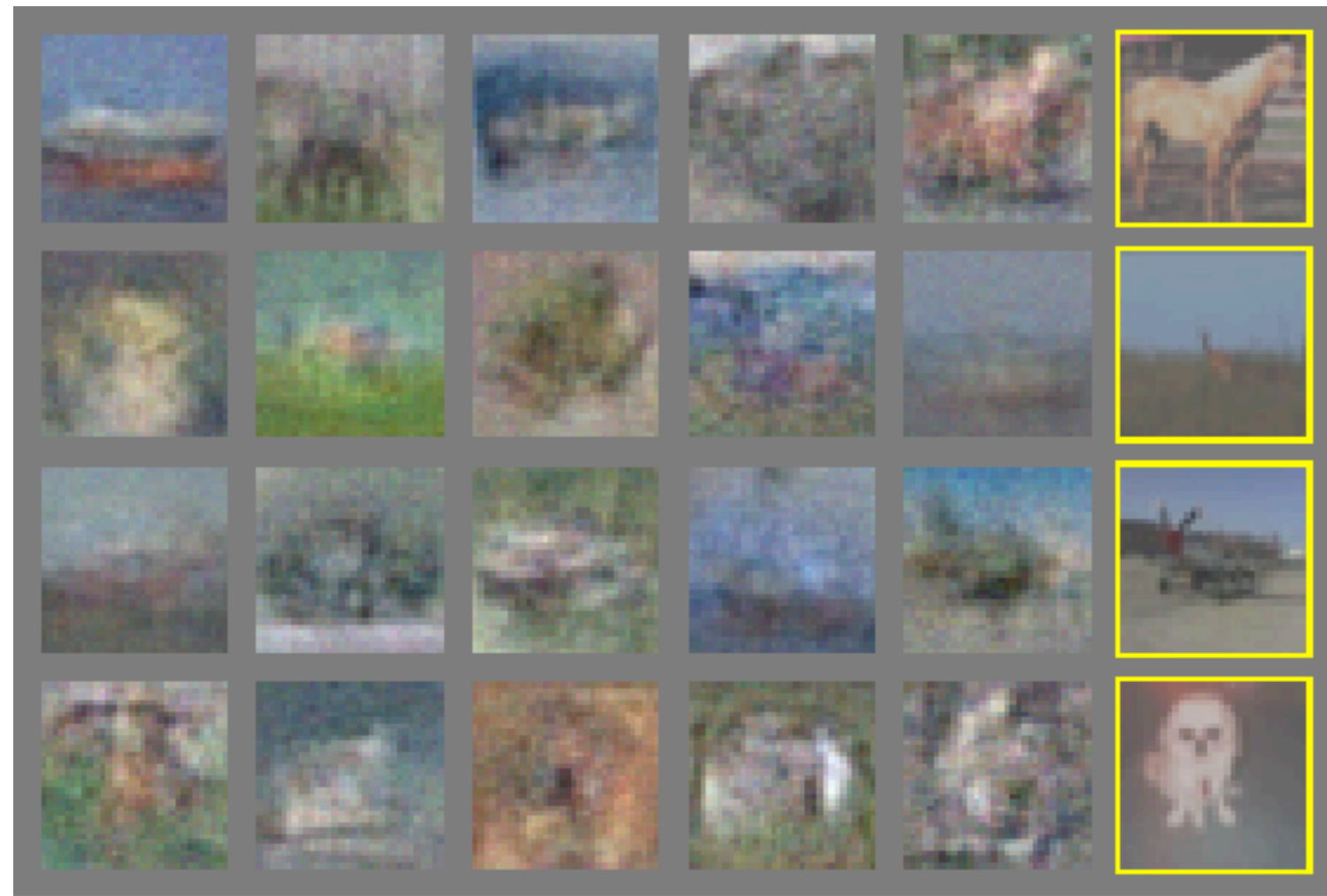
No Convolution Needed to Solve Simple Tasks



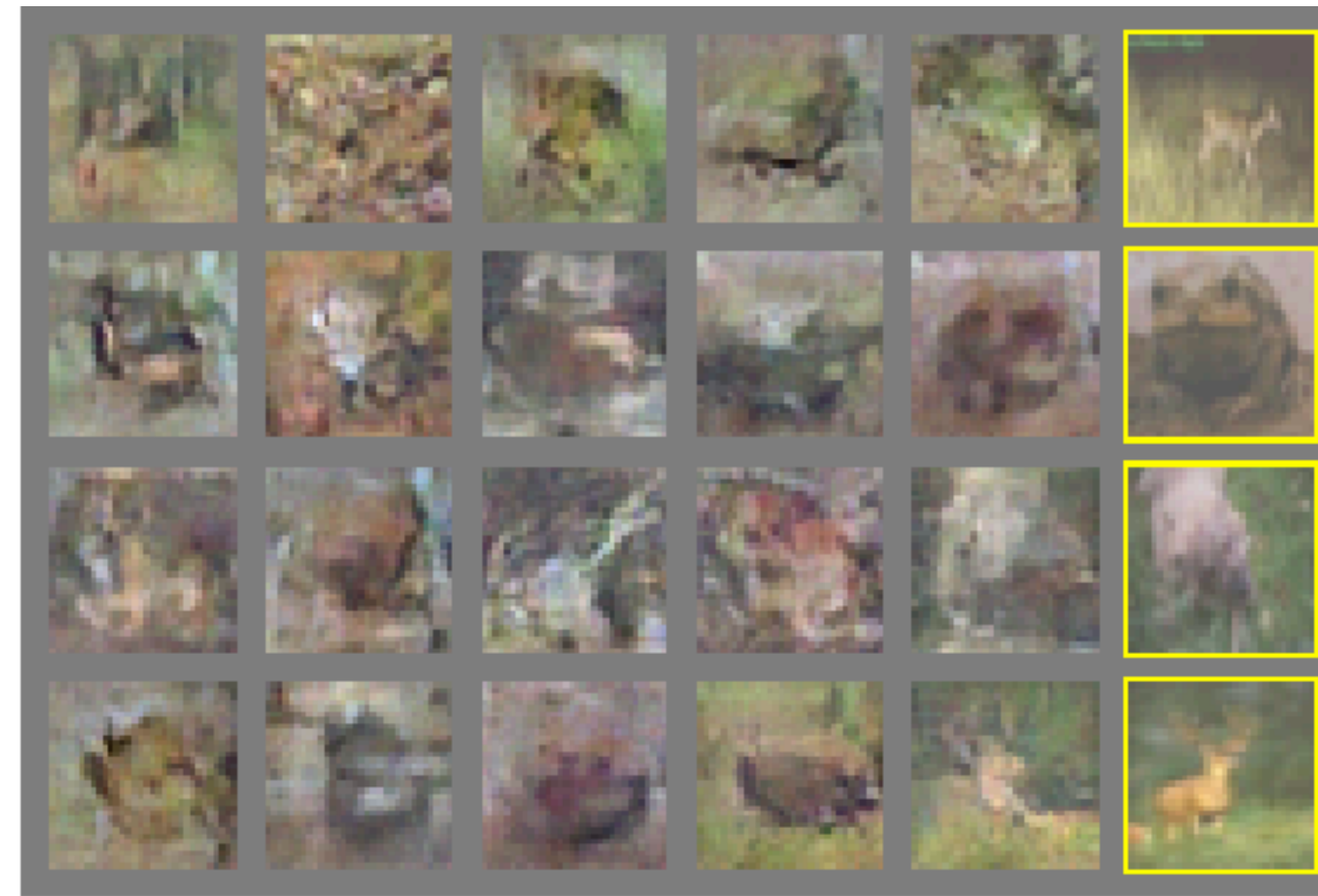
Original GAN, 2014

Depth and Convolution for Harder Tasks

Original GAN (CIFAR-10)



No convolution



One convolutional layer

DCGAN (ImageNet)

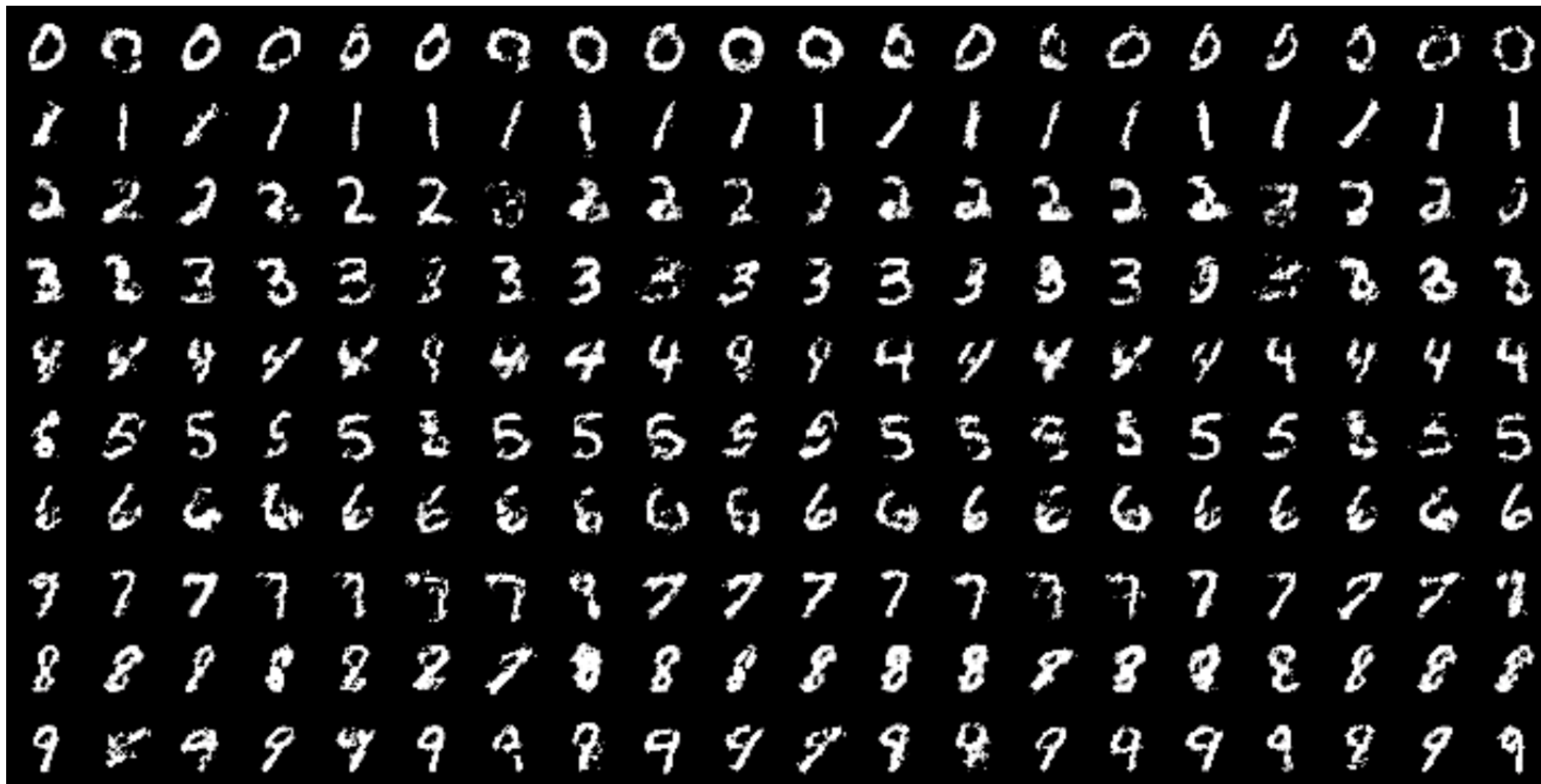


Many convolutional layers
(Radford et al, 2015)

From GAN to SAGAN

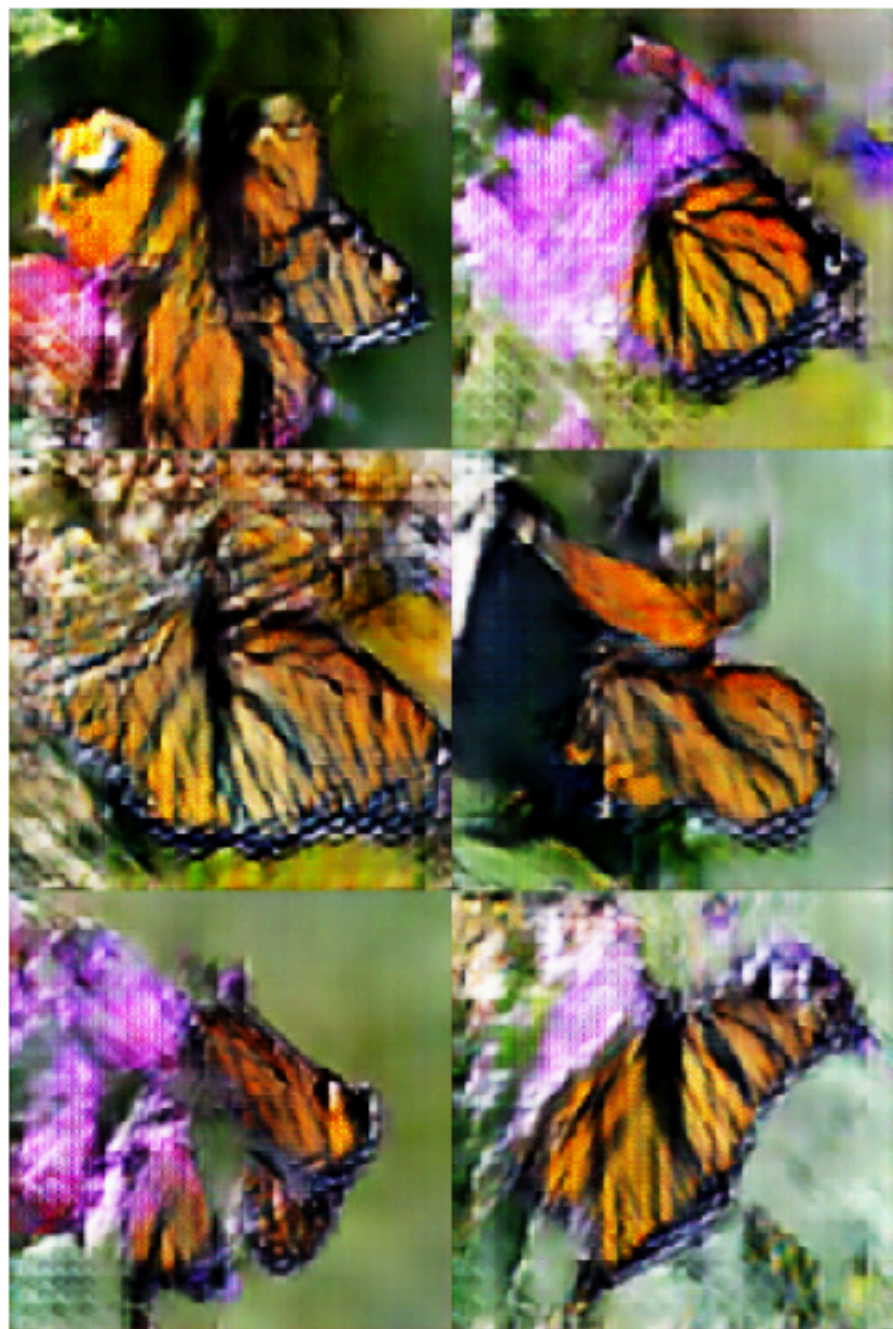
- Depth and Convolution
- Class-conditional generation
- Spectral Normalization
- Hinge loss
- Two-timescale update rule
- Self-attention

Class-Conditional GANs



(Mirza and Osindero, 2014)

AC-GAN: Specialist Generators



monarch butterfly



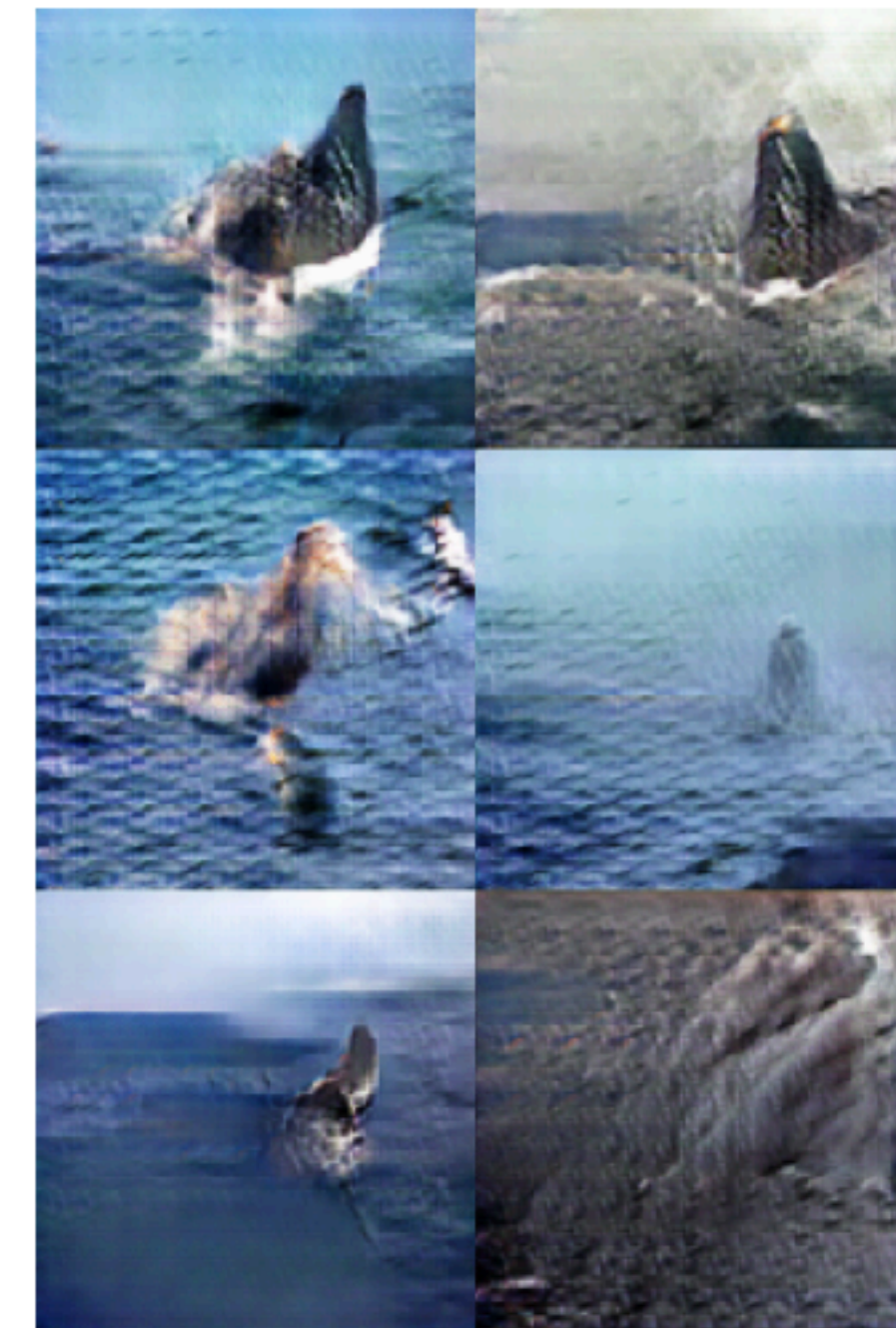
goldfinch



daisy



redshank

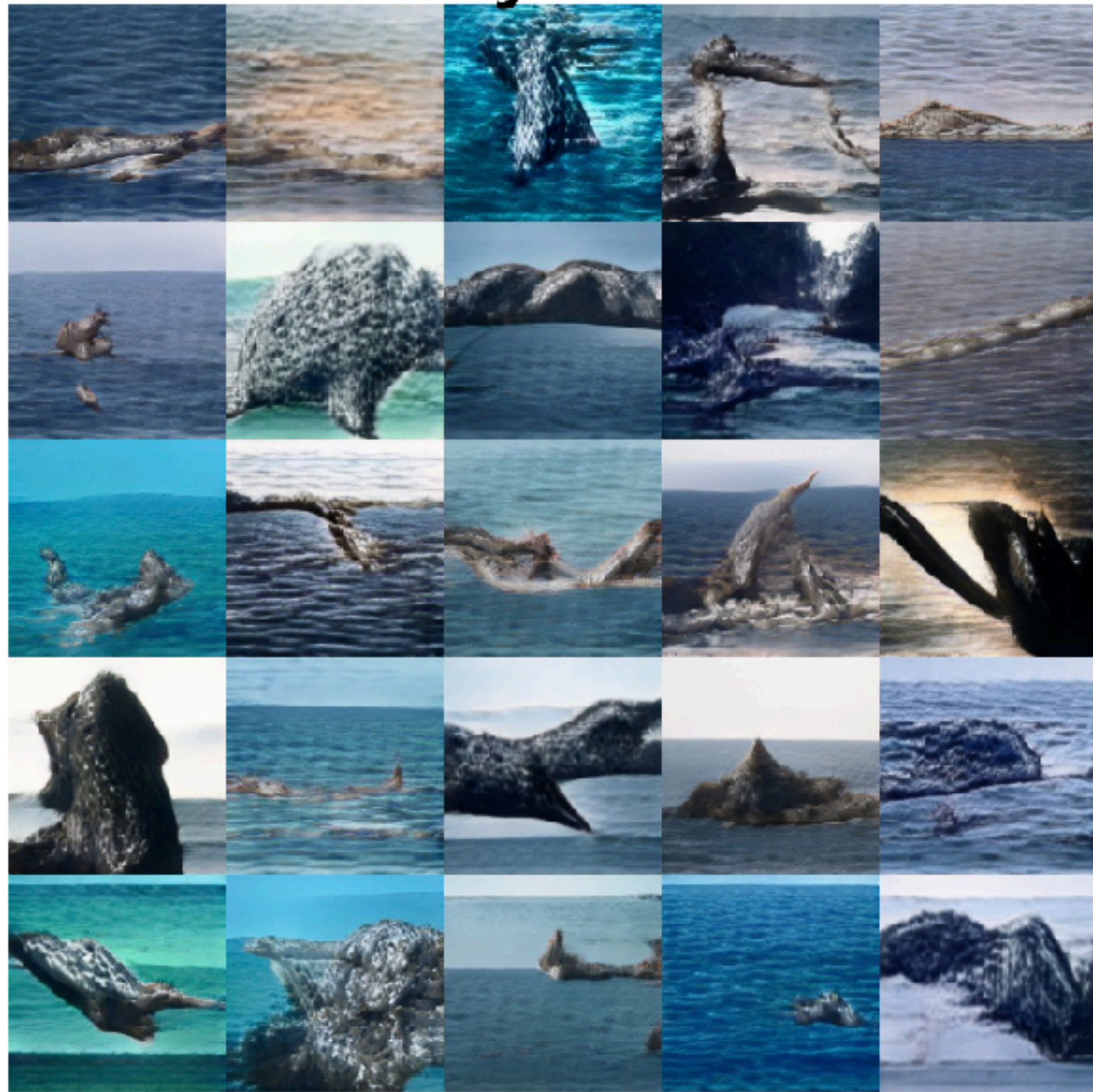


grey whale

(Odena et al, 2016)

SN-GAN: Shared Generator

Gray whale



Welsh springer spaniel



Persian cat



(Miyato et al, 2017)

From GAN to SAGAN

- Depth and Convolution
- Class-conditional generation
- Spectral Normalization
- Hinge loss
- Two-timescale update rule
- Self-attention

Spectral Normalization

$$\sigma(A) := \max_{\mathbf{h}:\mathbf{h}\neq\mathbf{0}} \frac{\|A\mathbf{h}\|_2}{\|\mathbf{h}\|_2} = \max_{\|\mathbf{h}\|_2\leq 1} \|A\mathbf{h}\|_2$$

$$\|f\|_{\text{Lip}} \leq \prod_{l=1}^{L+1} \sigma(W^l)$$

$$\bar{W}_{\text{SN}}(W) := W/\sigma(W)$$

(Miyato et al, 2017)

From GAN to SAGAN

- Depth and Convolution
- Class-conditional generation
- Spectral Normalization
- Hinge loss
- Two-timescale update rule
- Self-attention

Hinge Loss

$$V_D(\hat{G}, D) = \mathbb{E}_{\mathbf{x} \sim q_{\text{data}}(\mathbf{x})} [\min(0, -1 + D(\mathbf{x}))] + \mathbb{E}_{\mathbf{z} \sim p(\mathbf{z})} \left[\min\left(0, -1 - D\left(\hat{G}(\mathbf{z})\right)\right) \right] \quad (16)$$

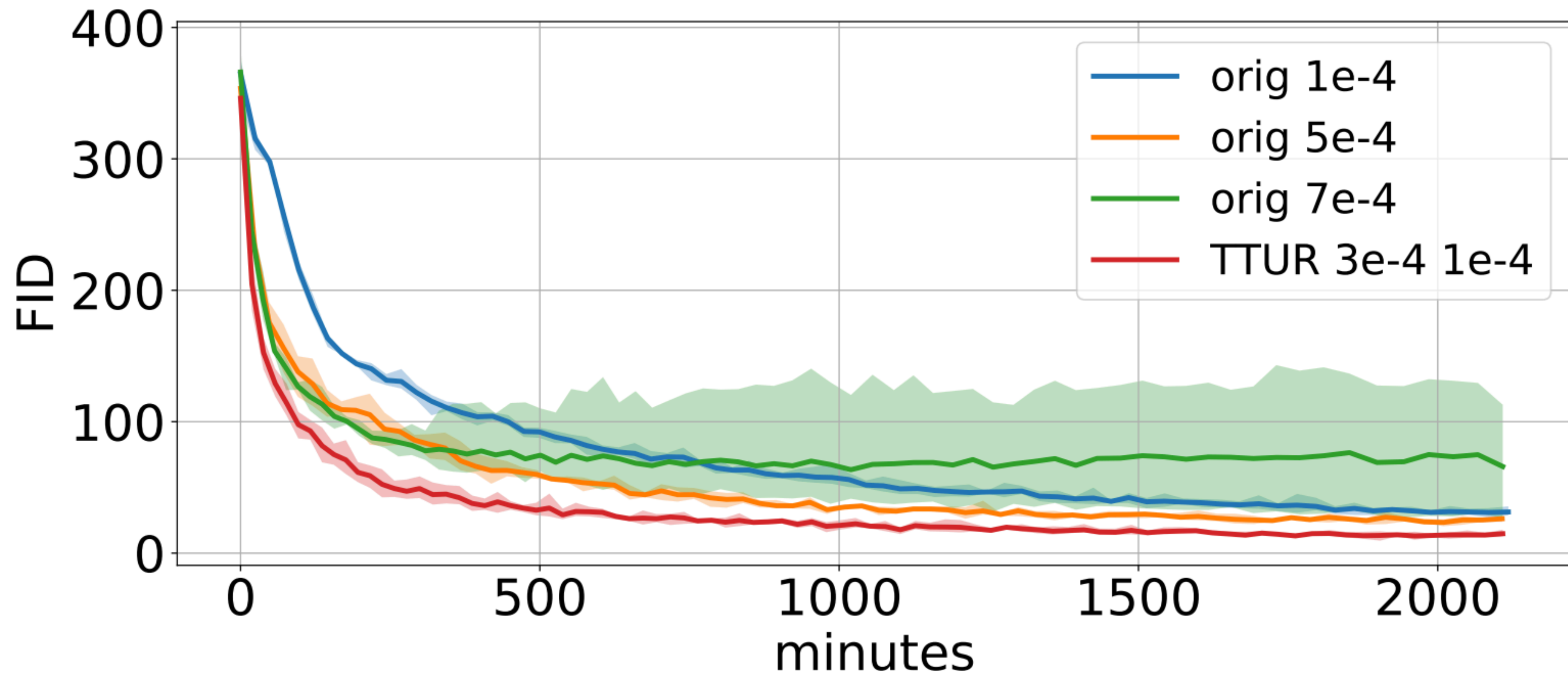
$$V_G(G, \hat{D}) = - \mathbb{E}_{\mathbf{z} \sim p(\mathbf{z})} \left[\hat{D}(G(\mathbf{z})) \right], \quad (17)$$

(Miyato et al 2017, Lim and Ye 2017, Tran et al 2017)

From GAN to SAGAN

- Depth and Convolution
- Class-conditional generation
- Spectral Normalization
- Hinge loss
- Two-timescale update rule
- Self-attention

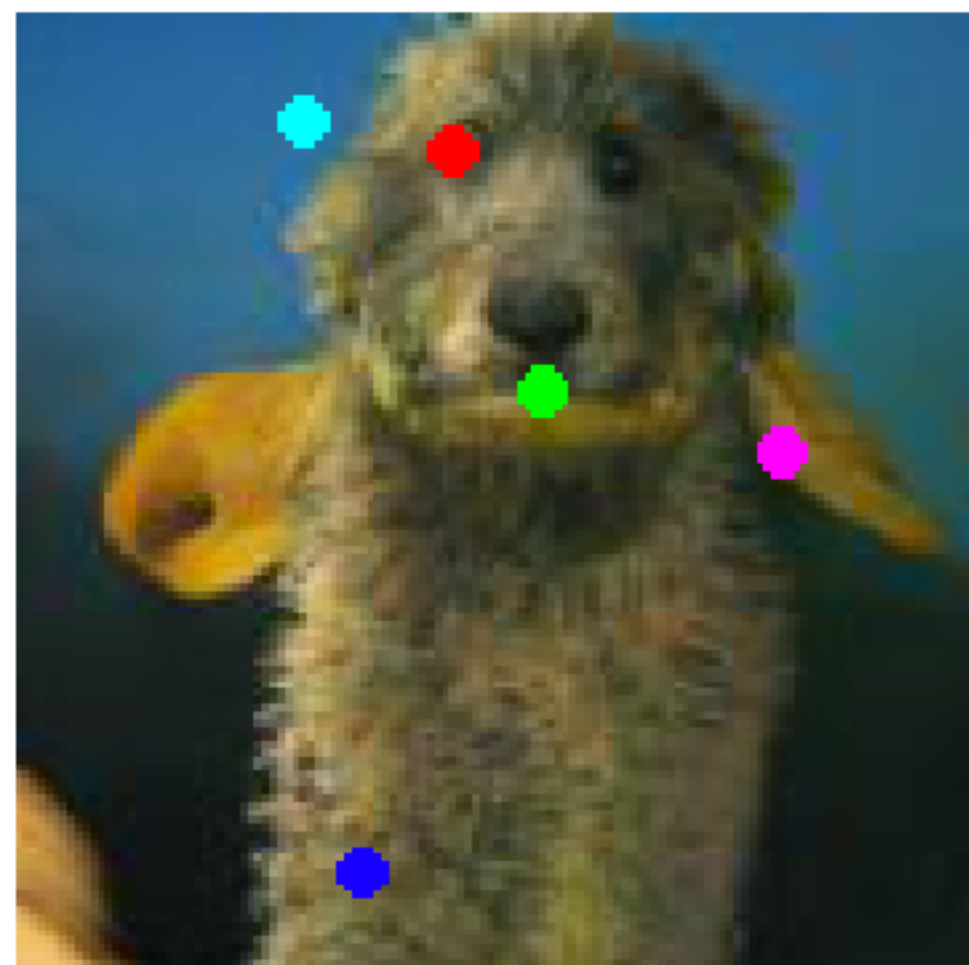
Two-Timescale Update Rule



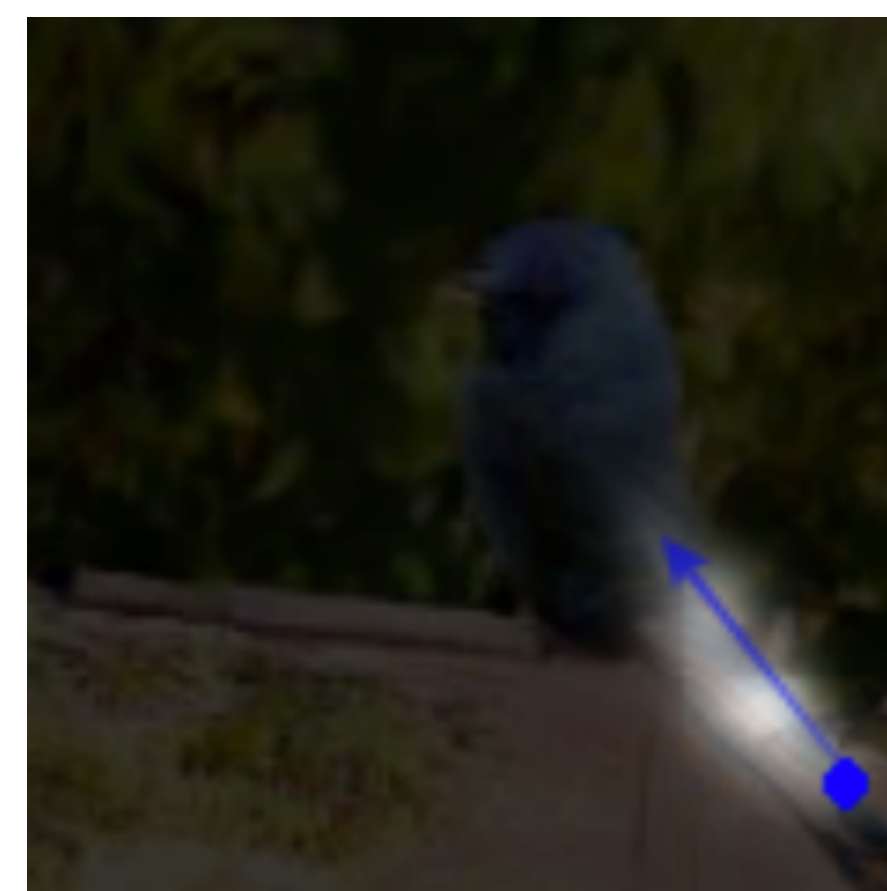
From GAN to SAGAN

- Depth and Convolution
- Class-conditional generation
- Spectral Normalization
- Hinge loss
- Two-timescale update rule
- Self-attention

Self-Attention



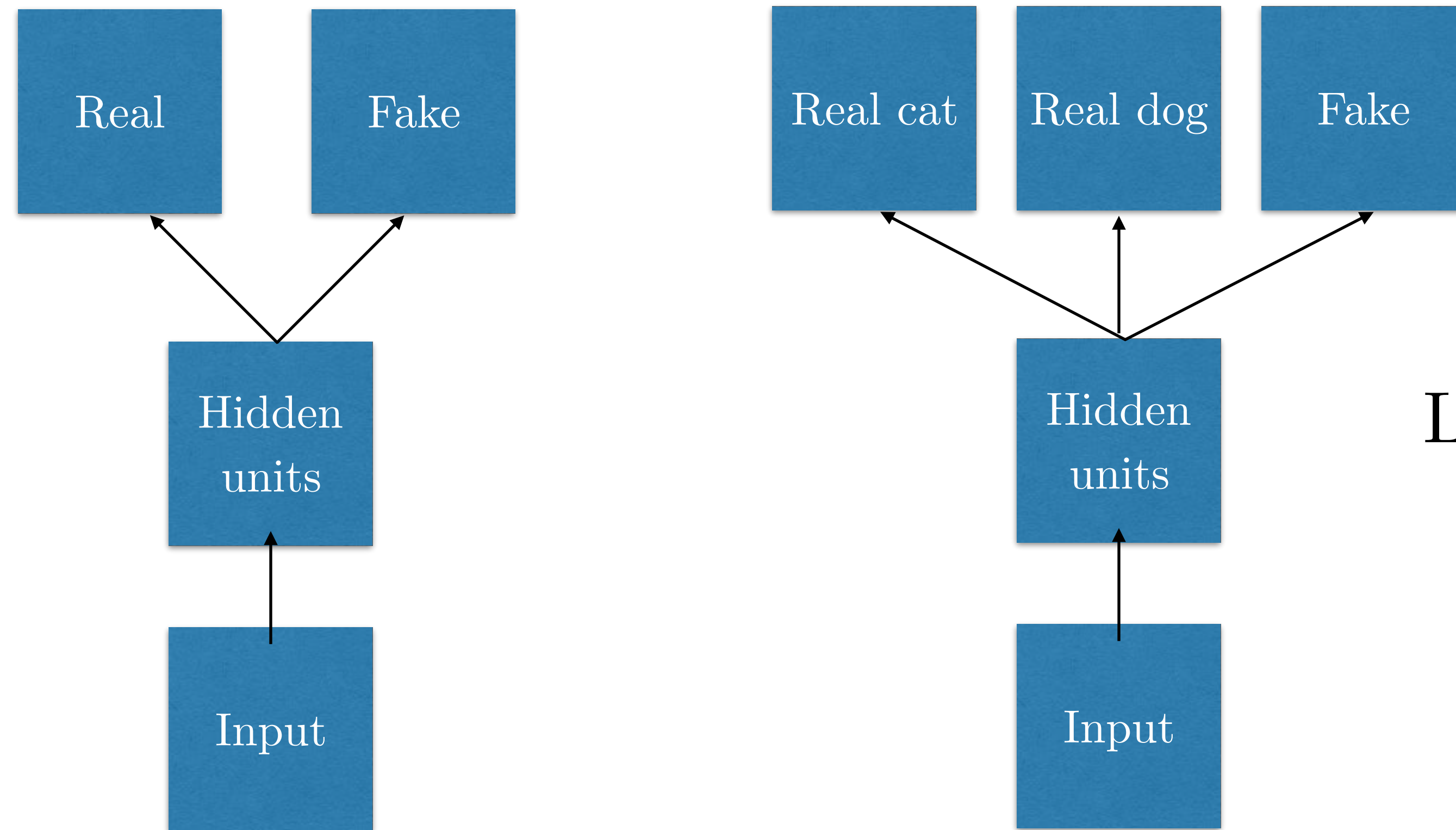
Use layers from
Wang et al 2018



Applying GANs

- Semi-supervised Learning
- Model-based optimization
- Extreme personalization
- Program synthesis

Supervised Discriminator for Semi-Supervised Learning



Learn to read with
100 labels rather
than 60,000

(Odena 2016, Salimans et al 2016)

Semi-Supervised Classification

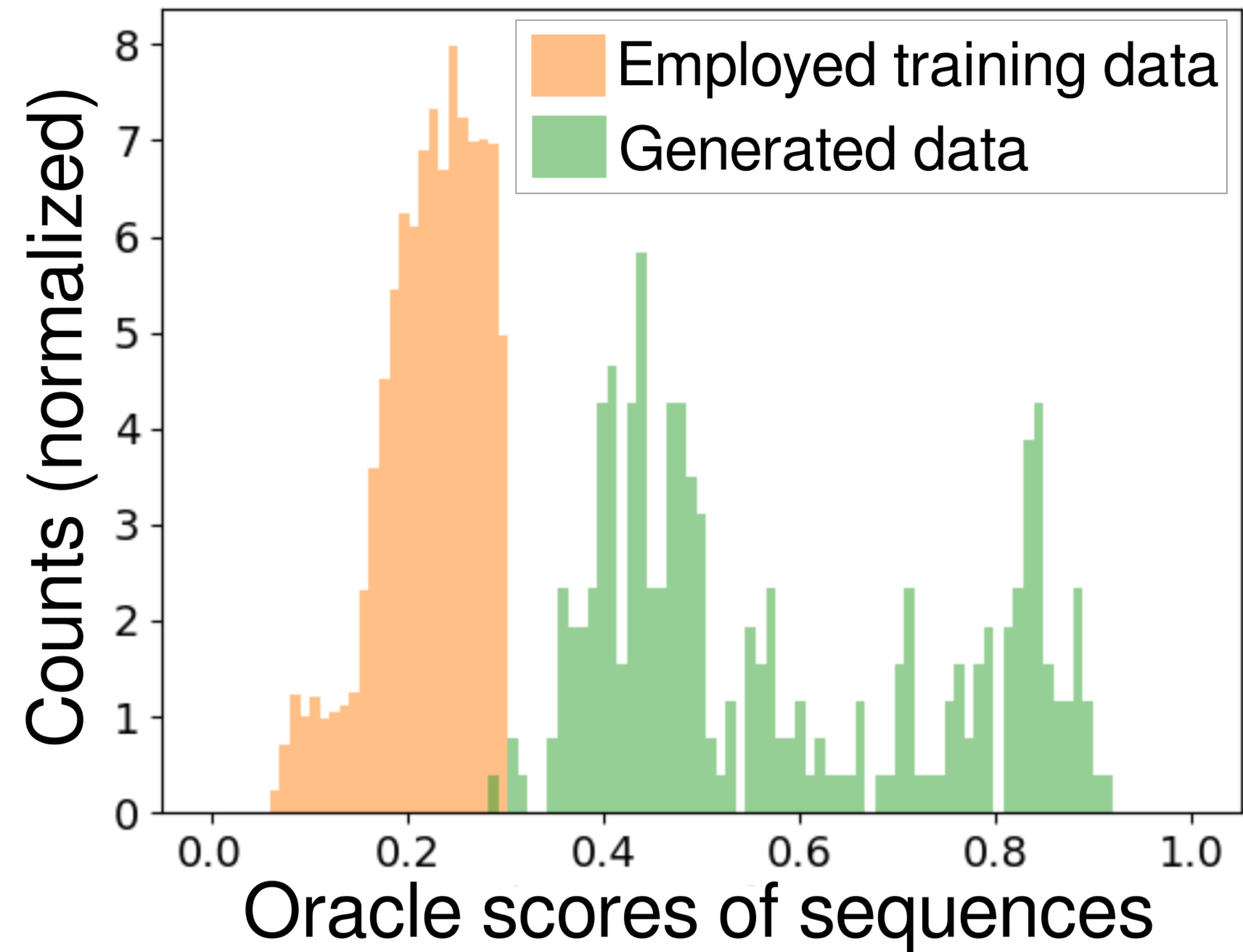
MNIST: 100 training labels \rightarrow 80 test mistakes

SVHN: 1,000 training labels \rightarrow 4.3% test error

CIFAR-10: 4,000 labels \rightarrow 14.4% test error

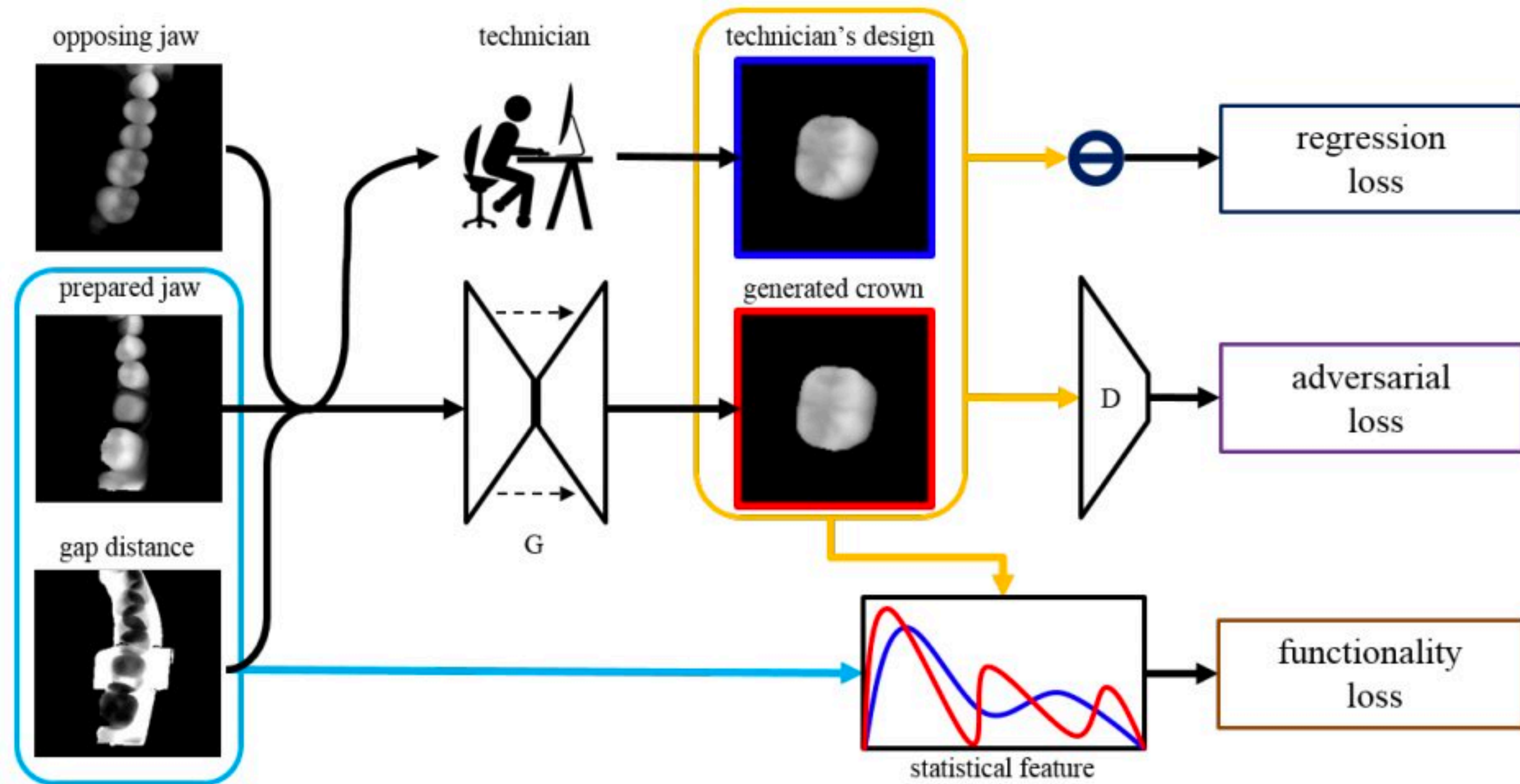
(Dai et al 2017)

Designing DNA to optimize protein binding



(Killoran et al, 2017)

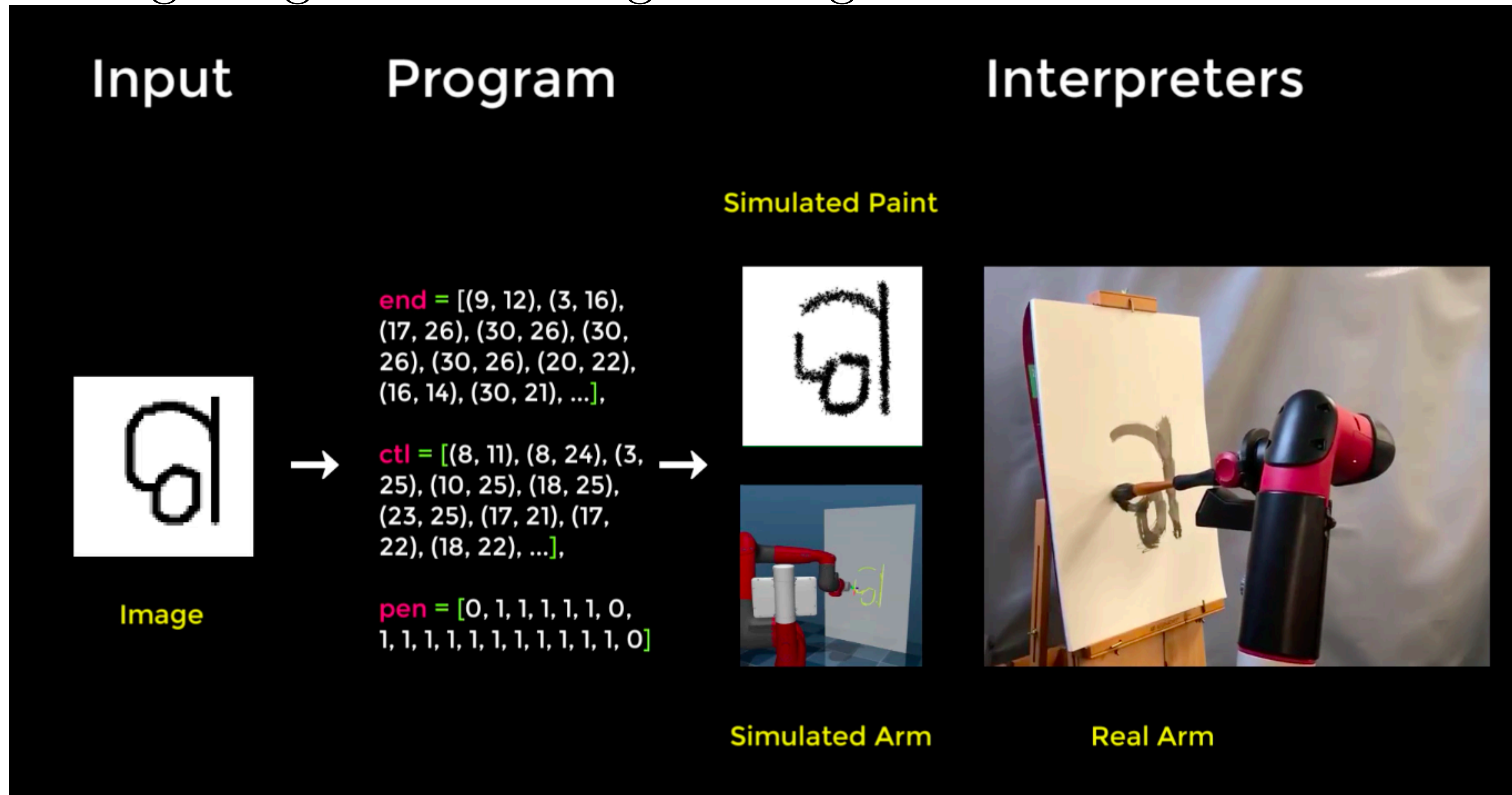
Personalized GANufacturing



(Hwang et al 2018)

SPIRAL

Synthesizing Programs for Images Using Reinforced Adversarial Learning

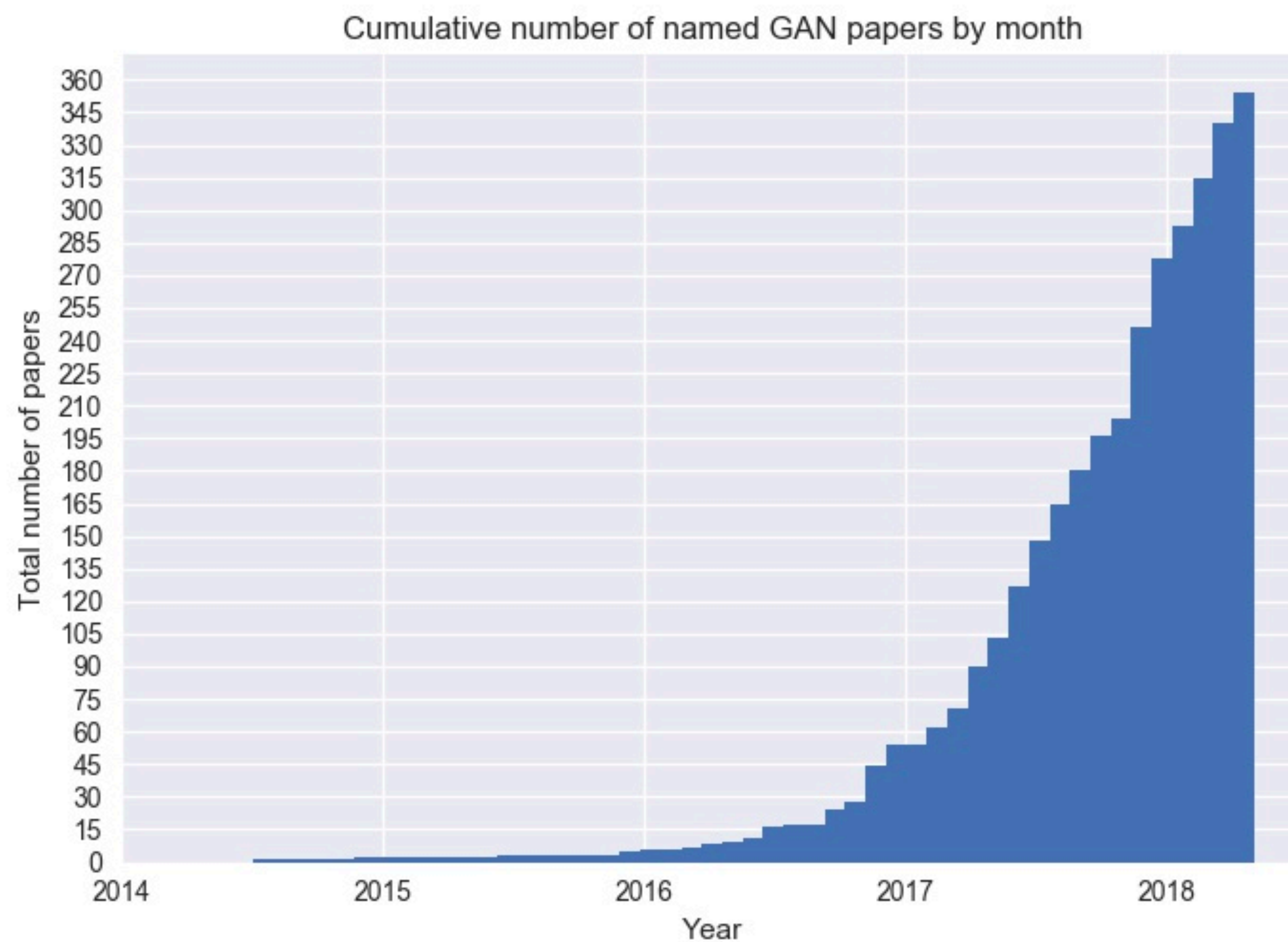


(Ganin et al, 2018)

Other applications

- Planning
- World Models for RL agents
- Fairness and Privacy
- Missing data
- Topics covered at workshop:
 - Training data for other agents (Philip Isola, Taesung Park, Jun-Yan Zhu)
 - Inference in other probabilistic models (Mihaela Rosca)
 - Domain adaptation (Judy Hoffman)
 - Imitation Learning (Stefano Ermon)

Track updates at the GAN Zoo



<https://github.com/hindupuravinash/the-gan-zoo>

Questions