# Generative Adversarial Networks (GANs)

Ian Goodfellow, OpenAI Research Scientist Presentation at AI With the Best, 2016-09-24



### Generative Modeling

• Density estimation



• Sample generation



(Goodfellow 2016)

### Adversarial Nets Framework



(Goodfellow 2016)

### DCGAN Architecture



(Radford et al 2015)

#### DCGANs for LSUN Bedrooms



#### (Radford et al 2015)

## Vector Space Arithmetic





Man



Man with glasses Woman



Woman with Glasses

## Mode Collapse

- Fully optimizing the discriminator with the generator held constant is safe
- Fully optimizing the generator with the discriminator held constant results in mapping all points to the argmax of the discriminator
- Can partially fix this by adding nearest-neighbor features constructed from the current minibatch to the discriminator ("minibatch GAN") (Salimans et al 2016)

### Minibatch GAN on CIFAR



Training Data

Samples

(Salimans et al 2016)

#### Minibatch GAN on ImageNet



(Salimans et al 2016)

### Cherry-Picked Results















(Goodfellow 2016)

### Text to Image with GANs

this small bird has a pink breast and crown, and black primaries and secondaries.



the flower has petals that are bright pinkish purple with white stigma

this magnificent fellow is almost all black with a red crest, and white cheek patch.



this white and yellow flower have thin white petals and a round yellow stamen





(Reed et al 2016)

### Generating Pokémon



youtube

(Yota Ishida)

#### Single Image Super-Resolution

original



bicubic (21.59dB/0.6423)







SRGAN (20.34dB/0.6562)



#### (Ledig et al 2016)

### iGAN



youtube

(Zhu et al 2016)

#### Introspective Adversarial Networks



youtube

### Conclusion

- GANs are generative models based on supervised learning and game theory
- GANs learn to generate realistic samples
- Like other generative models, GANs still need a lot of improvement