Generative Adversarial Networks

Stephan Halbritter

June 13, 2017

Hamburg University of Applied Science

Outline

- 1. Basic idea
- 2. Generative Adversarial Nets (Goodfellow et al. 2014)
- 3. Deep Convolutional GAN (Radford et al. 2015)
- 4. Examples of current research
- 5. Future work and resources

Basic idea

Goal

Generate "realistic-looking" data

Intuition

Interpret images as samples from a high-dimensional probability distribution





Source: kvfrans.com/generative-adversial-networks-explained/

Other approaches

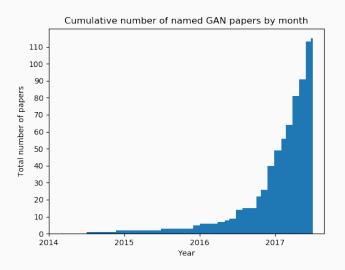
- Restricted Bolzmann Machines (RBM)
- Fully visible belief networks (FVBN)
- Variational autoencoders (VAE)

Generative Adversarial Nets (Goodfellow et al. 2014)

Generative Adversarial Networks (Goodfellow et al. 2014)

- presented at the Neural Information Processing Systems (NIPS) conference 2014
- proposes the framework
- · ground-breaking paper

Papers naming GAN variants



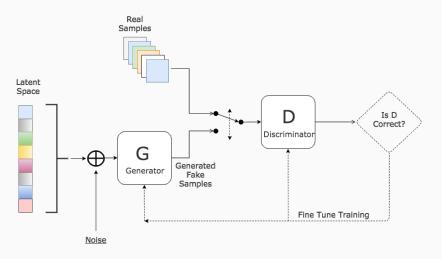
Source: github.com/hindupuravinash/the-gan-zoo as at June 10, 2017

Basic approach

Set up a game between two neural nets

- the generator creates samples
- the discriminator classifies these samples as real or fake
- both train each other

Overview



Source:

www.kdnuggets.com/2017/01/generative-adversarial-networks-hot-topic-machine-learning.html

Deep Convolutional GAN (Radford

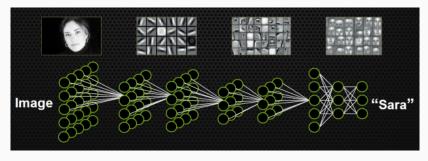
et al. 2015)

Deep Convolutional GAN (Radford et al. 2015)

Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks

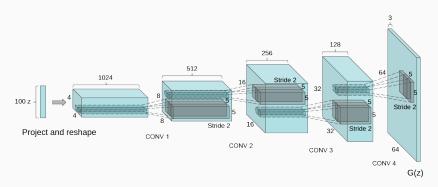
- · presented at ICLR 2016
- stabilizes GAN with architectural constraints
- · combines CNNs and GANs

Discriminator



Source: https://devblogs.nvidia.com/parallelforall/accelerate-machine-learning-cudnn-deep-neural-network-library/

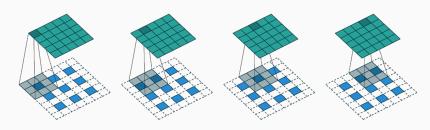
Generator



Source: Radford et al. (2015)

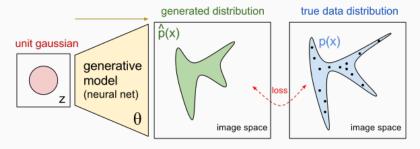
Upsampling with fractionally-strided convolution

Convolving a 3 \times 3 kernel over a 3 \times 3 input (with 1 zero inserted between inputs) padded with a 1 \times 1 border of zeros using unit strides.



Source: Dumoulin and Visin (2016)

Training approach



Source: blog.openai.com/generative-models/

Minimax game

$$\min_{G} \max_{D} V(D, G) = \mathbb{E}_{x \sim p_{\text{data}}}[\log D(x)] + \mathbb{E}_{z \sim p_{z}}[\log(1 - D(G(z)))]$$

$$z \text{ is a real image}$$

Training

for number of training iterations do

for k steps do

- Sample minibatch $\{z^{(1)}, \ldots, z^{(m)}\}$ from $p_g(z)$.
- Sample minibatch $\{x^{(1)}, \ldots, x^{(m)}\}$ from $p_{\text{data}}(x)$.
- Update D by ascending its stochastic gradient:

$$\nabla_{\theta_d} \frac{1}{m} \sum_{i=1}^{m} \left[\log D\left(x^{(i)} \right) + \log \left(1 - D\left(G\left(z^{(i)} \right) \right) \right) \right].$$

end for

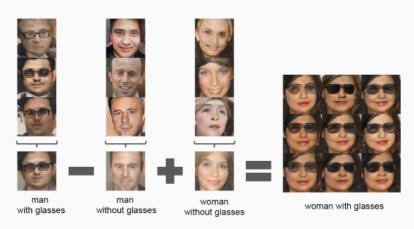
- Sample minibatch $\{z^{(1)}, \ldots, z^{(m)}\}$ from $p_q(z)$.
- Update G by descending its stochastic gradient:

$$\nabla_{\theta_g} \frac{1}{m} \sum_{i=1}^m \log \left(1 - D\left(G\left(\mathbf{z}^{(i)}\right)\right)\right).$$

end for

Examples of current research

DCGAN (Radford et al. 2015)



Source: Radford et al. (2015)

StackGAN (H. Zhang et al. 2016)

A small yellow bird with a black crown and a short black pointed beak



Source: H. Zhang et al. (2016)

Beyond Face Rotation (R. Huang et al. 2017)





Source: R. Huang et al. (2017)

Future work and resources

Future work

- · learn more about deep learning
- better understand mathematical background
- · learn a framework and create something

Helpful stuff

- arxiv-sanity.com
- · blog.openai.com
- distill.pub
- offconvex.org

Conferences 2017

International Conference on Learning Representations (ICLR) April 24 - 26 in Toulon, France.

International Conference on Machine Learning (ICML) August 6 - 11 in Sidney, Australia.

Neural Information Processing System (NIPS) December 4 - 9 in Long Beach, USA.

References

Dumoulin, Vincent, and **Francesco Visin**. 2016. "A Guide to Convolution Arithmetic for Deep Learning." arXiv:1603.07285.

Goodfellow, Ian, Jean Pouget-Abadie, et al. 2014. "Generative Adversarial Networks." arXiv:1406.2661.

Huang, Rui, Shu Zhang, et al. 2017. "Beyond Face Rotation: Global and Local Perception GAN for Photorealistic and Identity Preserving Frontal View Synthesis" arXiv:1704.04086.

Radford, Alec, Luke Metz, and Soumith Chintala. 2015. "Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks." CORR abs/1511.06434.

Zhang, Han, Tao Xu, et al. 2016. "StackGAN: Text to Photo-Realistic Image Synthesis with Stacked Generative Adversarial Networks." arXiv:1612.03242.

Any questions?