Your Projects Applications Theory

Generative Adversarial Networks Deep Learning — Unit 9

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Slides available at jonkrohn.com/talks

December 1st, 2018



- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 "Quick, Draw!" Implementation





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Progress Check Your Deep Learning Project V







Progress Check Your Deep Learning Project V

- Splitting your data
 - training set (80% for optimizing parameters)
 - validation set (10% for hyperparameters)
 - test set (10% don't touch yet!)
- 2 Building and assessing architecture
 - get above chance (simplifying problem, if necessary)
 - do existing performance benchmarks exist?
 - if not, use a simple architecture as benchmark





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- build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - one of dozens of "image" data sets from [CrowdFlower]
 - one of the Computer Vision data sets from [Luke de Oliveira's post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.
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GANs

Goodfellow et al. (2014)

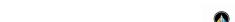








d)





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DCGANs

Radford et al. (2016)



(a) Generated by LSGANs.



(b) Generated by DCGANs (Reported in [13]).

Figure 5: Generated images on LSUN-bedroom.



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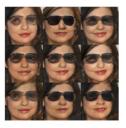
In Practic

DCGANs Radford et al. (2016)











man with glasses



man without glasses



woman without glasses

woman with glasses















Results of doing the same arithmetic in pixel space



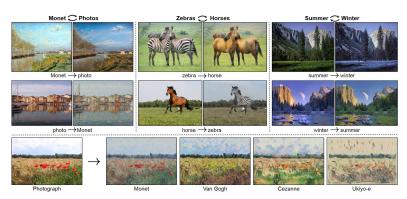
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CycleGANs Zhu et al. (2017)

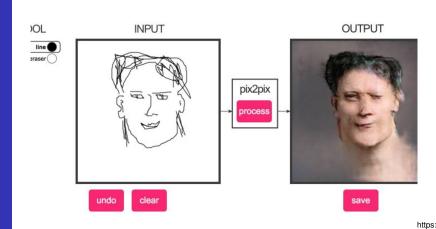


https://junyanz.github.io/CycleGAN



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pix2pix Isola et al. (2017)





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StackGAN Zhang et al. (2017)



Figure 3. Example results by our proposed StackGAN, GAWWN [20], and GAN-INT-CLS [22] conditioned on text descriptions from CUB test set. GAWWN and GAN-INT-CLS generate 16 images for each text description, respectively. We select the best one for each of them to compare with our StackGAN.



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["celebrity" latent-space interpolation]



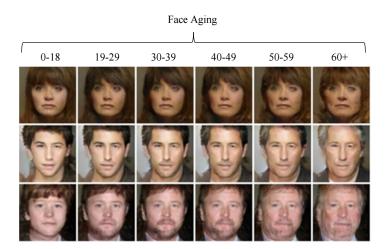
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Latent-Space Interpolation



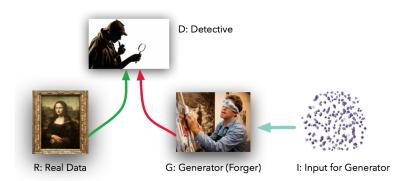


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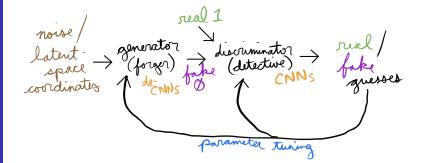
Goodfellow et al. (2014)





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Goodfellow et al. (2014)





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1-D Gaussian

Approximating a Toy Distribution

[video]



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[Quick, Draw!]



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GANimation

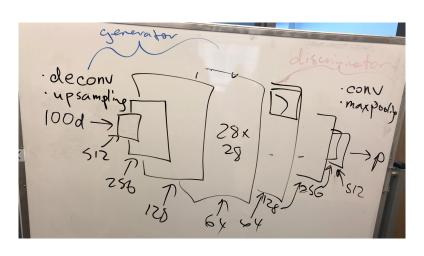
(Requires Adobe Acrobat Reader)



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GAN Code



[notebook]

