



GAN

• 이미지 생성 AI

• GAN

DCGAN

Generative Adversarial Nets

Ian J. Goodfellow, Jean Pouget-Abadie, Mehdi Mirza, Bing Xu, David Warde-Farley, Sherjil Ozair, Aaron Courville, Yoshua Bengio;

Département d'informatique et de recherche opérationnelle Université de Montréal Montréal, QC H3C 3J7





• OIDIN 생성 AI GAN DCGAN

• GAN

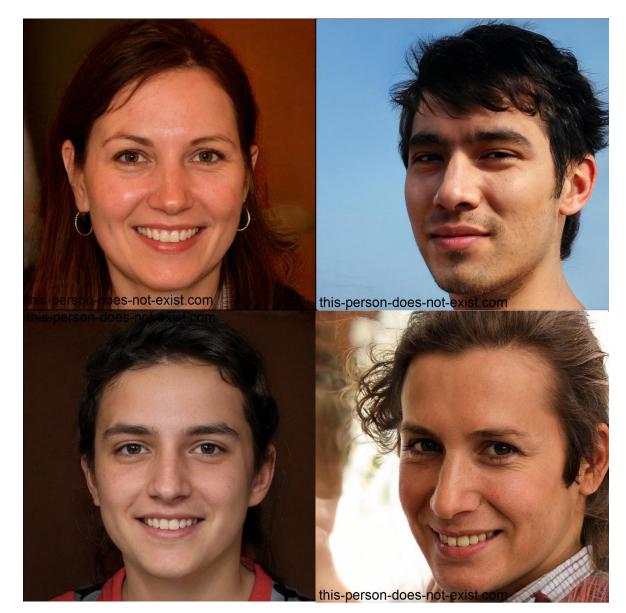






• OIDN 생성 AI GAN DCGAN

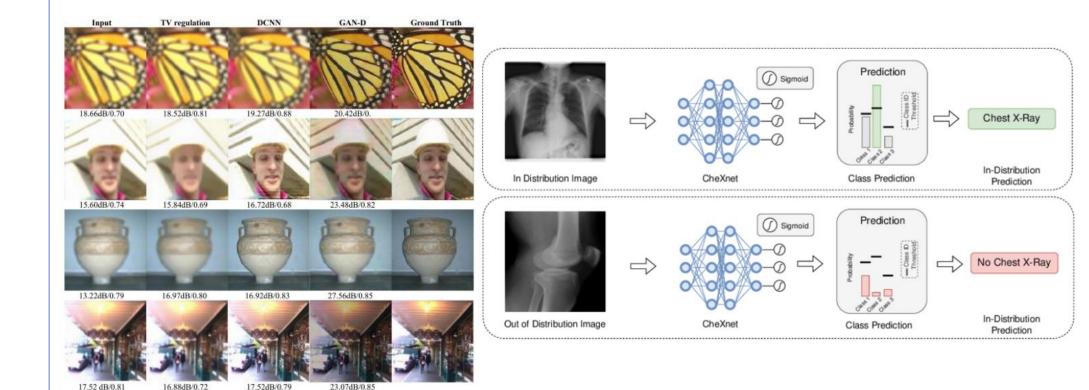
• GAN





• OIDN 생성 AI GAN DCGAN

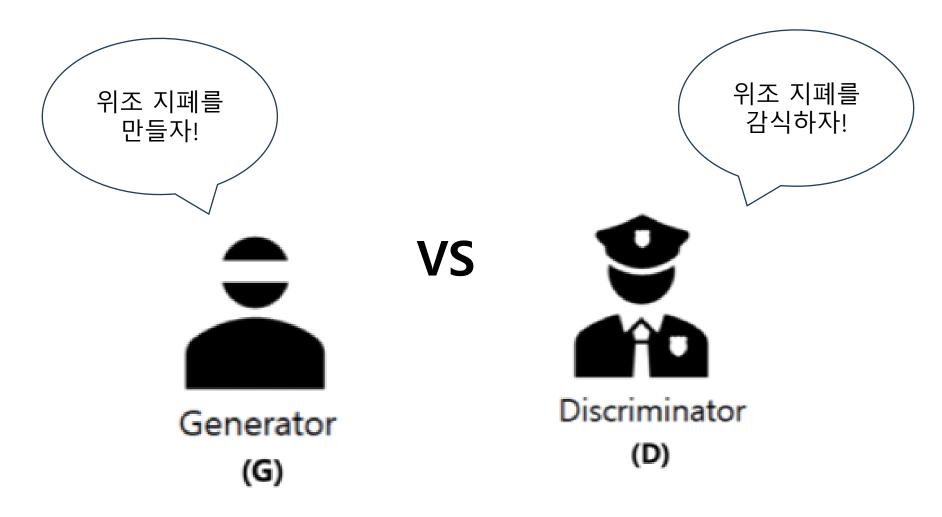
• GAN의 활용







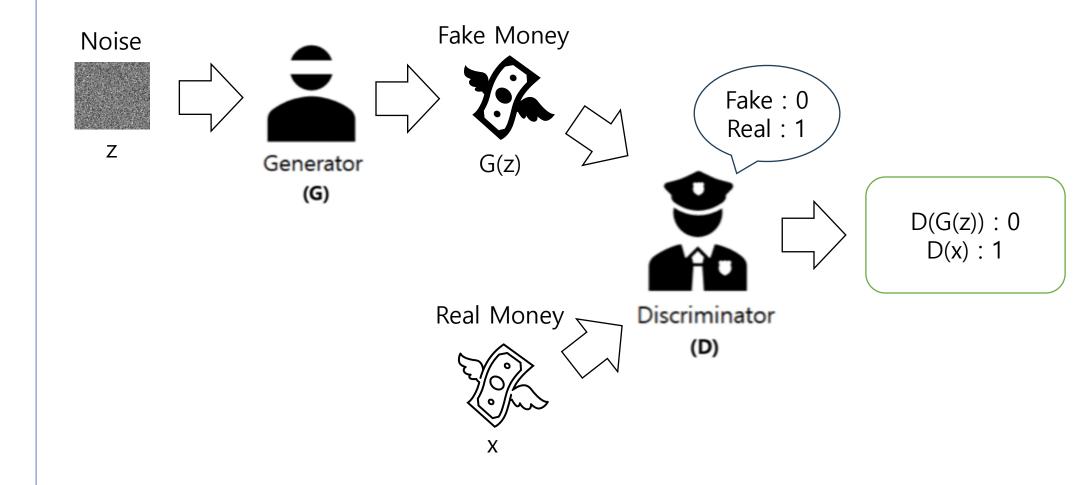
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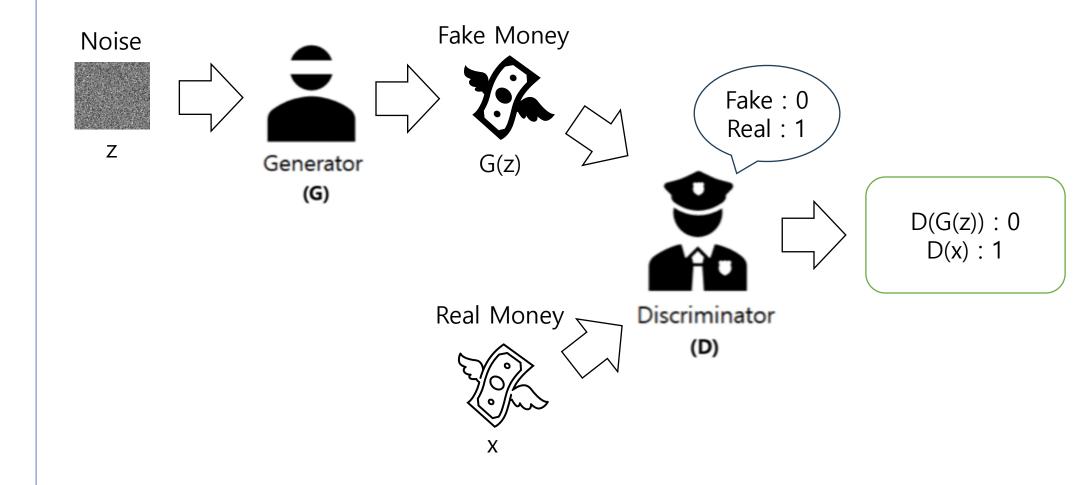
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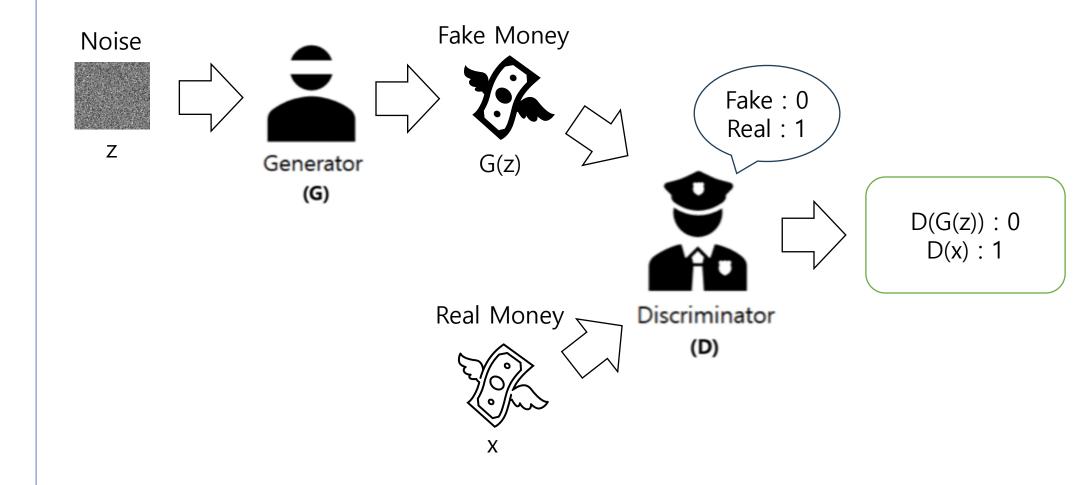
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$$\min_{G} \max_{D} V(D,G) = \mathbb{E}_{\boldsymbol{x} \sim p_{\text{data}}(\boldsymbol{x})}[\log D(\boldsymbol{x})] + \mathbb{E}_{\boldsymbol{z} \sim p_{\boldsymbol{z}}(\boldsymbol{z})}[\log (1 - D(G(\boldsymbol{z})))].$$



• GAN DCGAN

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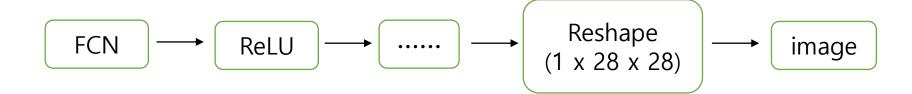
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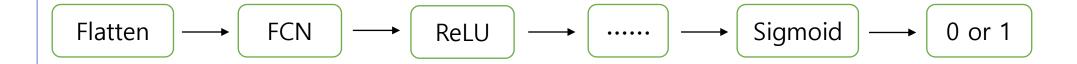
- Generator





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- Discriminator





• GAN
DCGAN

- 조금 더 자세하게…

Discriminator: Supervised Learning

Generator: Unsupervised Learning



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DCGAN

- 조금 더 자세하게…

Discriminator: input(위조지폐)가 0인지 1인지 분류

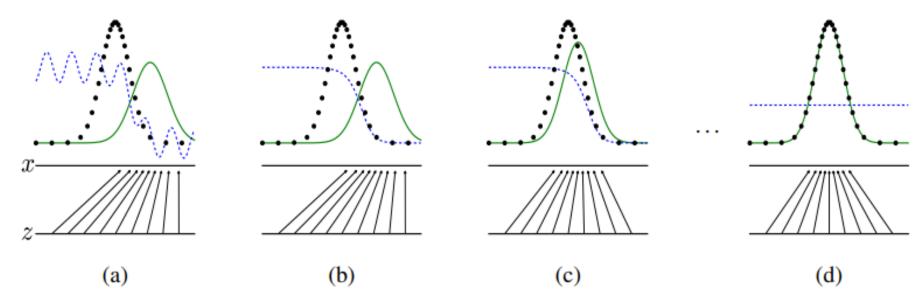
Generator : input(Noise=latent vector)을 training 데이터가 되도록 학습



• GAN DCGAN

- Generator

: 데이터의 분포를 보고, Noise를 training 데이터(Real data)의 분포가 되도록 변화 시킨다!



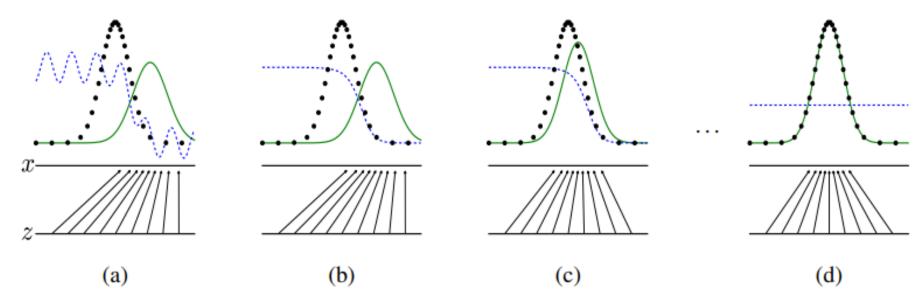
- blue dashed line : distribution of discriminator
- ullet black dotted line : distribution of data generating (real) $o p_{data}$
- ullet green solid line : distribution of **generator's generative distribution (fake)** $o p_q$
- ullet lower horizontal line : domain of ${f z}$ (sampled) ightarrow in this case, uniform distribution
- ullet upper horiziontal line : domain of ${f x}$
- ullet upward arrows : mapping ${f x}=G({f z})$



• GAN DCGAN

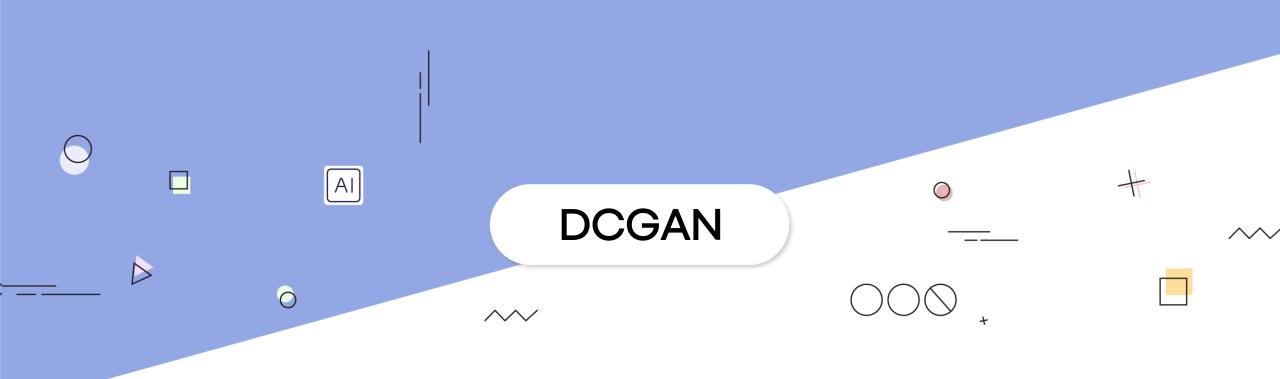
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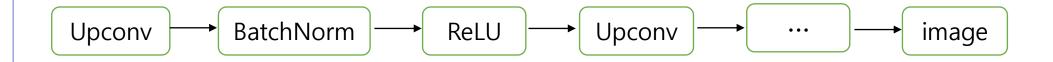
- DCGAN (Deep Convolutional GAN)
- (1) G와 D에서 FCN이 아닌 Convolution Layer 을 사용한다. (모든 G와 D에서 FCN을 사용하지 않는다.)
- (2) Pooling layer 을 사용하지 않는다. (대신 strided convolutions를 사용)
- (3) Batch normalization을 사용하여 모델의 학습을 안정화 (GAN이 가지고 있던 안정성 문제를 어느정도 해결함)
- (4) CNN의 특성으로 인해 공간 정보를 더 잘 학습하여, DCGAN으로 만들어 진 latent vector는 이미지의 연속성을 더 잘 표현한다.



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- Generator

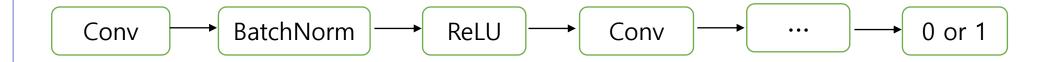




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- Discriminator









참고문헌

- 1. Goodfellow, lan, et al. "Generative adversarial nets." Advances in neural information processing systems 27 (2014).
- 2. Teoh, Teik Toe, and Zheng Rong. "Deep Convolutional Generative Adversarial Network." Artificial Intelligence with Python. Singapore: Springer Singapore, 2022

