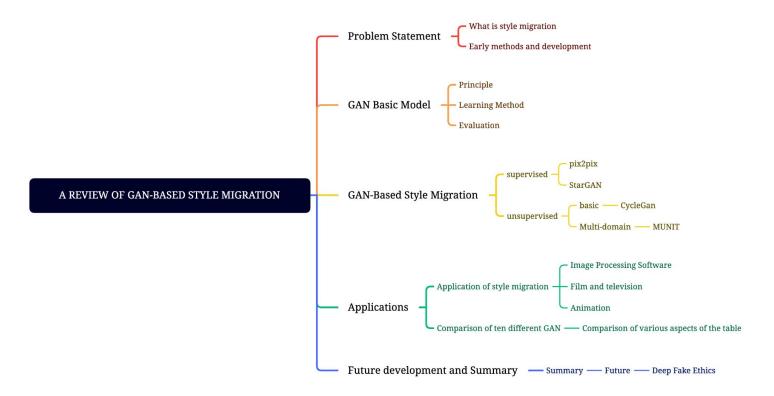
# A Review of Style Migration Based on GAN Networks

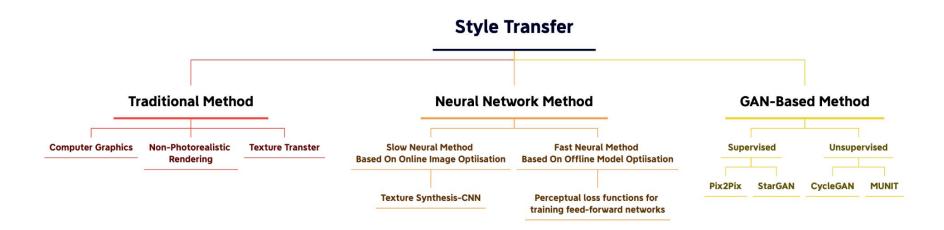
**EECE7370 Final Project** 

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## Content



# Style Migration Development History



#### Control the brush stroke size in NST

Since the emergence of NST algorithms, there are also some researches devoted to improving current NST algorithms by controlling perceptual factors (e.g., stroke size control, spatial style control, and colour control)

(a)Content (b)Style (c)Small Stroke (d)Large Stroke



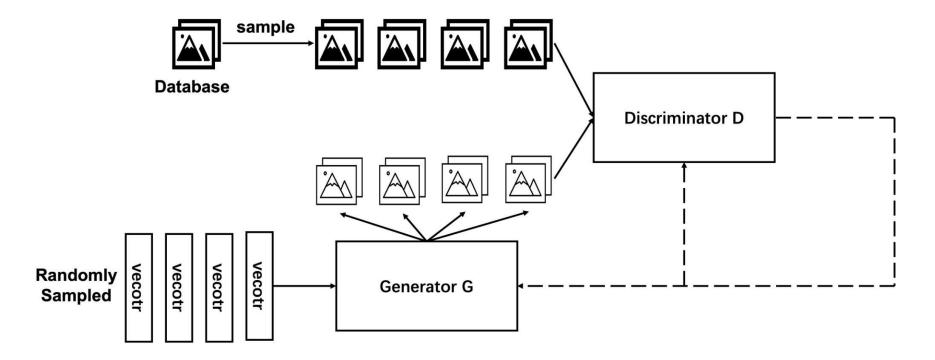






Y. Jing, Y. Yang, Z. Feng, J. Ye, Y. Yu and M. Song, "Neural Style Transfer: A Review," in IEEE Transactions on Visualization and Computer Graphics, vol. 26, no. 11, pp. 3365-3385, 1 Nov. 2020, doi: 10.1109/TVCG.2019.2921336.

# GAN algorithm



## GAN algorithm

The generator captures the potential distribution of real data samples and generates new data samples; the discriminator is a binary classifier that discriminates whether the input is real data or generated samples.

Both the generator and the discriminator can be used in deep neural networks, which are currently a hot research activity.

#### Algorithm 1 GAN algorithm [9]

for number of training iterations do

#### for k steps do

- •Sample minibatch of m noise samples  $z^{(1)}, ..., z^{(m)}$  from noise prior  $p_a(z)$ .
- •Sample minibatch of m examples  $x^{(1)}, ..., x^{(m)}$  from data generating distribution  $p_{data}(x)$ .
- •Update the discriminator by ascending its stochastic gradient:

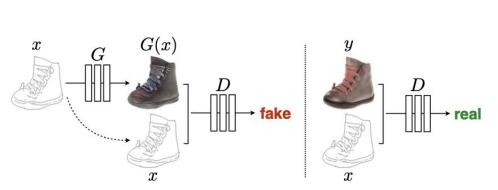
$$\nabla_{\theta_d} \frac{1}{m} \sum_{i=1}^m [\log D(x^{(i)}) + \log (1 - D(G(z^{(i)})))]$$

#### end for

- •Sample minibatch of m noise samples  $z^{(1)}, ..., z^{(m)}$  from noise prior  $p_q(z)$ .
- •Update the generator by descending its stochastic gradient:

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

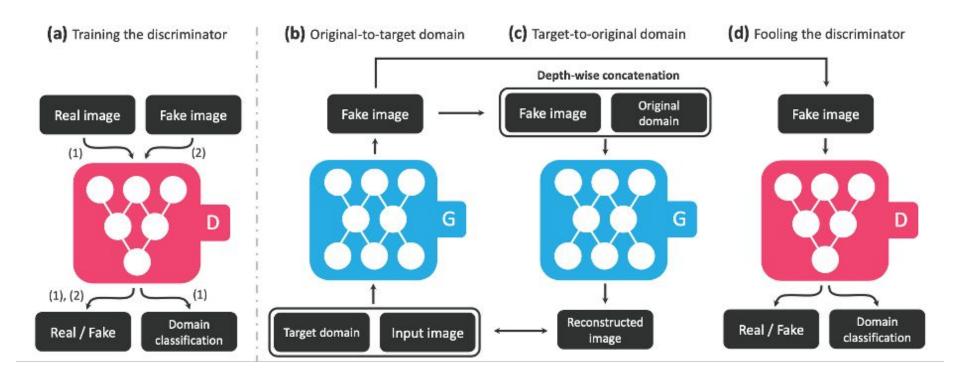
## Pix2Pix algorithm



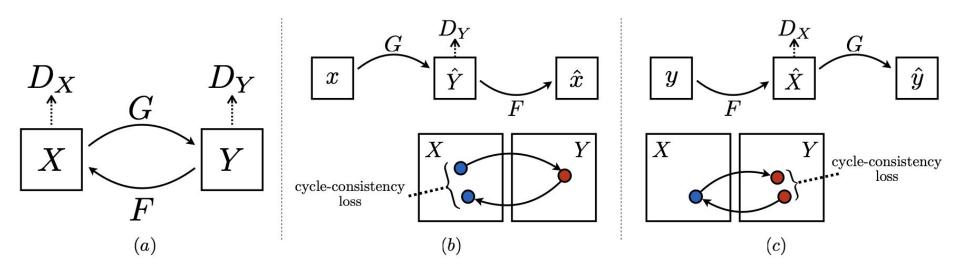


Phillip Isola, Jun-Yan Zhu, Tinghui Zhou, and Alexei A Efros. Image-to-image translation with conditional adver- sarial networks. In Proceedings of the IEEE conference on computer vision and pattern recognition, pages 1125–1134, 2017.

# StarGAN algorithm

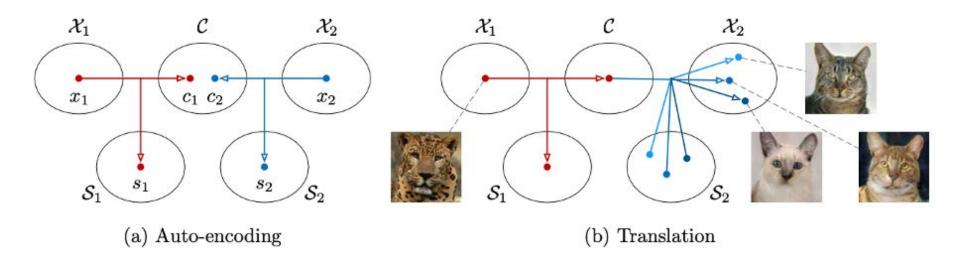


## CycleGAN algorithm



Jun-Yan Zhu, Taesung Park, Phillip Isola, and Alexei AEfros. Unpaired image-to-image translation using cycle- consistent adversarial networks. In Proceedings of the IEEE international conference on computer vision, pages 2223–2232, 2017.

## MUNIT algorithm



Huang, Xun, et al. "Multimodal unsupervised image-to-image translation." *Proceedings of the European conference on computer vision (ECCV)*. 2018.

## Application

- Photo and Video Editors
  - Photo-to-Animation Translation
  - Photo-to-Oil Paintings
  - Al Generate Faces
- Text Style Transfer
- Video Gaming
- Virtual Reality

#### Photo and Video Editors

#### **Photo-to-Animation Translation**





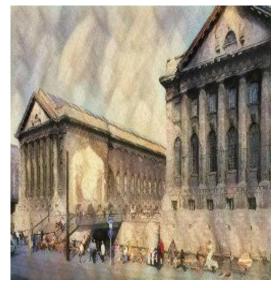
Chen J, Liu G, Chen X. AnimeGAN: a novel lightweight GAN for photo animation[C]//International symposium on intelligence computation and applications. Springer, Singapore, 2020: 242-256.

### Photo and Video Editors

#### **Photo-to-Oil Paintings**







Xu, Wenju, et al. "Drb-gan: A dynamic resblock generative adversarial network for artistic style transfer." *Proceedings of the IEEE/CVF International Conference on Computer Vision*. 2021.

#### Photo and Video Editors

#### **Al Generate Faces**



# Reference(part)

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- [2] Yunjey Choi, Minje Choi, Munyoung Kim, Jung-Woo Ha, Sunghun Kim, and Jaegul Choo. Stargan: Unified genera- tive adversarial networks for multi-domain image-to-image translation. In Proceedings of the IEEE conference on computer vision and pattern recognition, pages 8789–8797, 2018.
- [3]Goodfellow lan, Jean Pouget-Abadie, Mehdi Mirza, Bing Xu, and David Warde-Farley. Generative adversarial nets." in advances in neural information processing systems. 2014.
- [4]Huang, Xun, et al. "Multimodal unsupervised image-to-image translation." *Proceedings of the European conference on computer vision (ECCV)*. 2018.
- [5] Phillip Isola, Jun-Yan Zhu, Tinghui Zhou, and Alexei A Efros. Image-to-image translation with conditional adver- sarial networks. In Proceedings of the IEEE conference on computer vision and pattern recognition, pages 1125–1134, 2017.
- [6]Choi, Yunjey, et al. "Stargan: Unified generative adversarial networks for multi-domain image-to-image translation." *Proceedings of the IEEE conference on computer vision and pattern recognition*. 2018.
- [7]Y. Jing, Y. Yang, Z. Feng, J. Ye, Y. Yu and M. Song, "Neural Style Transfer: A Review," in IEEE Transactions on Visualization and Computer Graphics, vol. 26, no. 11, pp. 3365-3385, 1 Nov. 2020, doi: 10.1109/TVCG.2019.2921336.
- [8] Xu, Wenju, et al. "Drb-gan: A dynamic resblock generative adversarial network for artistic style transfer." *Proceedings of the IEEE/CVF International Conference on Computer Vision*. 2021.
- [9]Chen J, Liu G, Chen X. AnimeGAN: a novel lightweight GAN for photo animation[C]//International symposium on intelligence computation and applications. Springer, Singapore, 2020: 242-256.