

- 쎄트렉아이 신기술연구팀 연구원
- 딥러닝과 컴퓨터 비전, 그리고 여러 프로그래밍과 관련된 지식에 흥미
- 끊임없이 하고 싶은걸 배우는 것이 목표





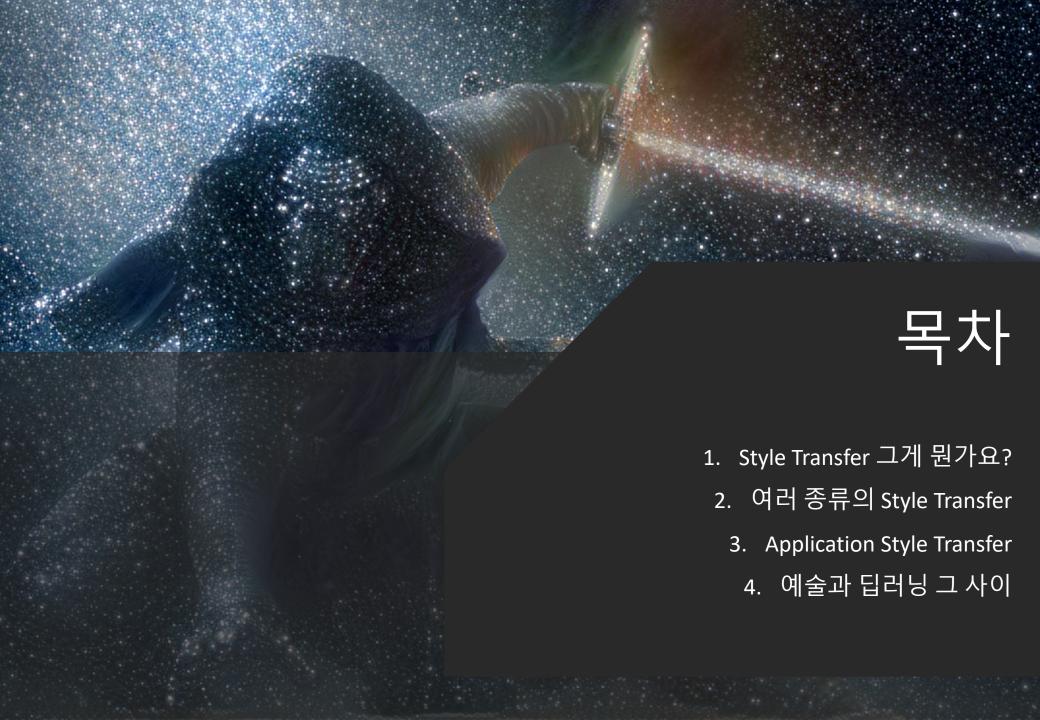












#### 오늘 시간을 통해

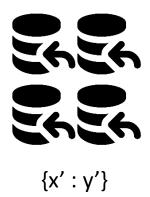
• Style Transfer 분야에 대한 흥미와 지식

• Style Transfer 활용 분야

• 집으로 돌아가서 공부해볼 거리

$$x * \Box = y$$

$$y - y'$$



Backpropagation!

$$x * \Box = y$$



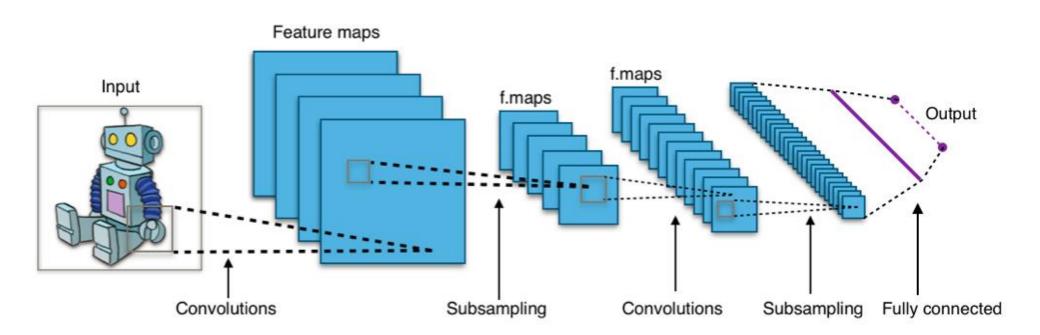
"magically make them work on your data"

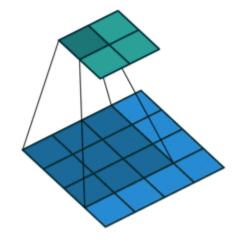
 $\{x':y'\}$ 

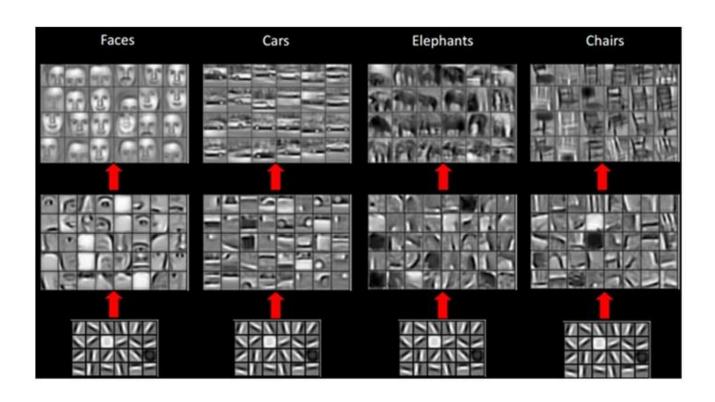
Backpropagation!



**Backpropagation!** 



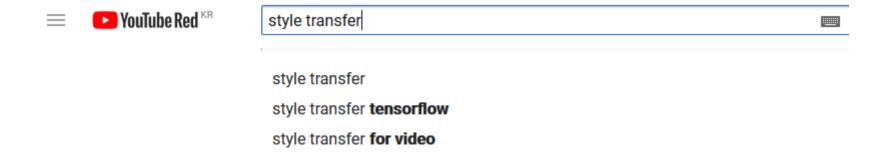


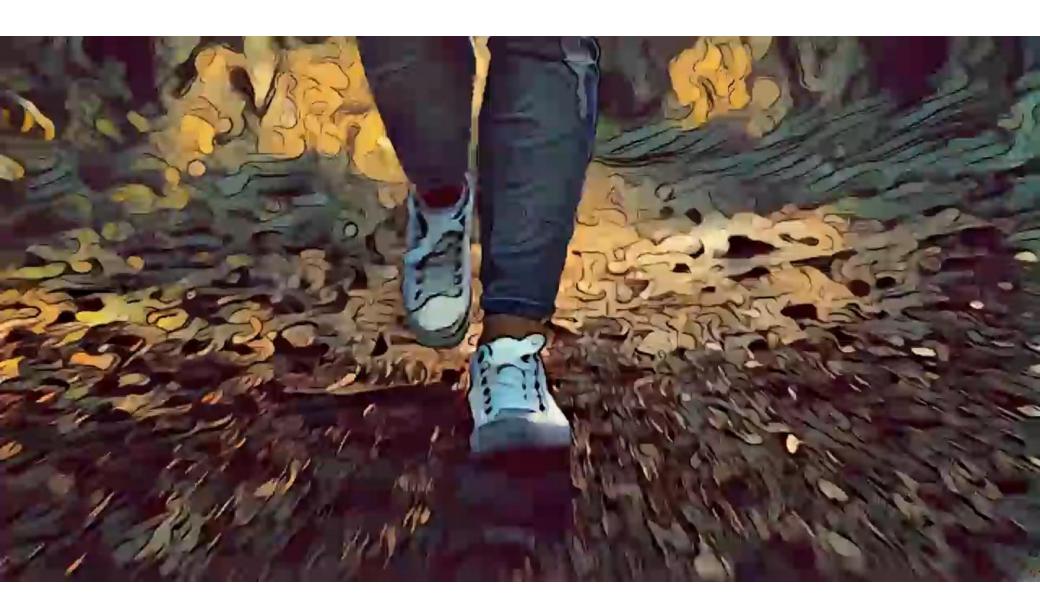


Style Transfer 그게 뭔가요?

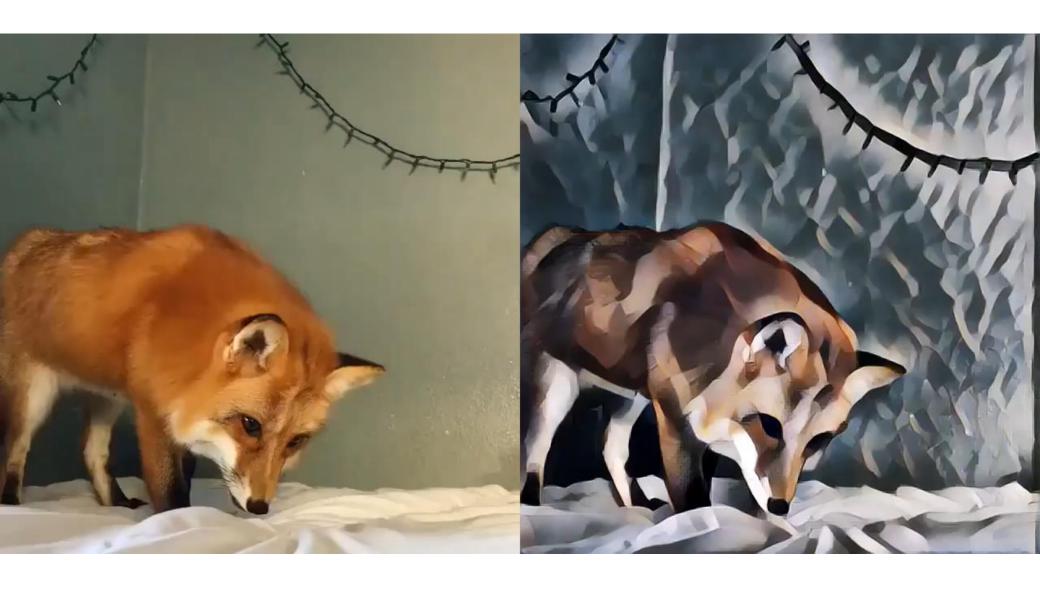


#### 몰라서 YouTube에 검색해보니..





https://youtu.be/ZIwy2pqIFG4







기존 이미지의 형상은 유지 스타일은 내가 원하는 스타일로



.l 형상은 유지 나 원하는 스타일로

#### Content + New Style = ?



Content Image

#### Content + New Style = ?



Content Image



Style Image

#### Content + New Style = ?

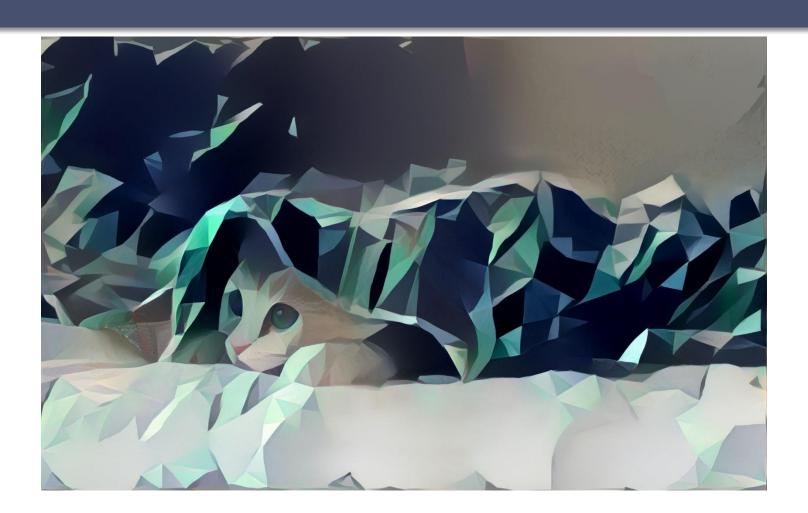


Content Image



Style Image

### Content + New Style = !



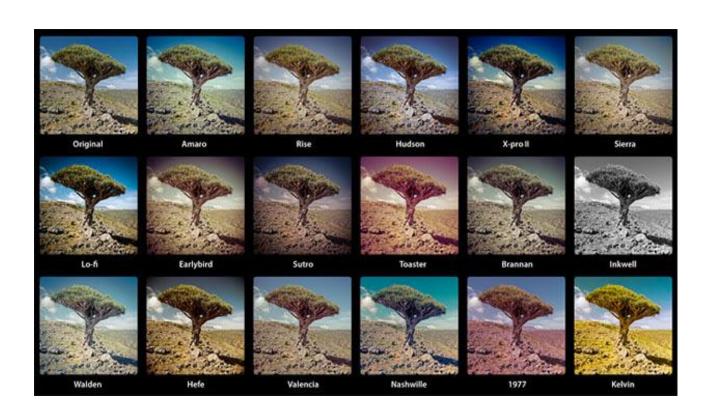
#### 인스타그램 필터랑 살짝 다른거 같은데요?





Style Image

#### 인스타그램 필터랑 살짝 다른거 같은데요?





• 기존 이미지의 형상은 유지

- 기존 이미지의 형상은 유지
- 스타일은 내가 원하는 스타일로

- 어떻게 이미지의 형상을 유지하지?
- 스타일은 내가 원하는 스타일로

- 어떻게 이미지의 형상을 유지하지?
- 이미지에서 스타일을 어떻게 뽑아내지?

#### A Taxonomy of Style Transfer Method

- Image Iteration
  - MMD-based
  - MRF-based
- Model Iteration

#### 어떻게 이미지의 형상을 유지하지?

#### **Understanding Deep Image Representations by Inverting Them**

Aravindh Mahendran University of Oxford

Andrea Vedaldi University of Oxford

### 이미지 컨텐츠 정보를 재구성

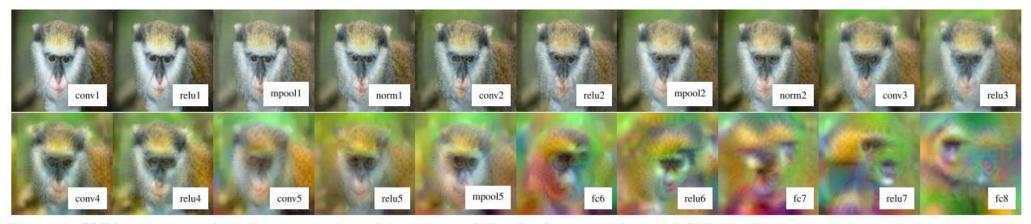


Figure 6. **CNN reconstruction.** Reconstruction of the image of Fig. 5.a from each layer of CNN-A. To generate these results, the regularization coefficient for each layer is chosen to match the highlighted rows in table 3. This figure is best viewed in color/screen.

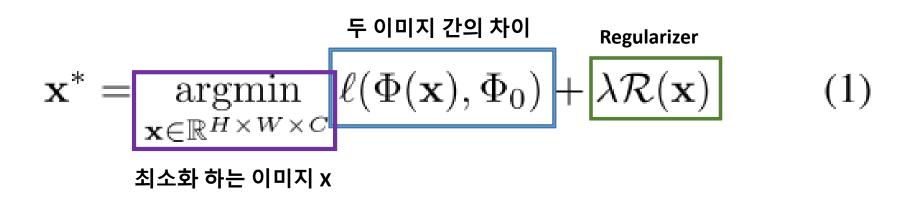
#### 이미지 컨텐츠 정보를 재구성

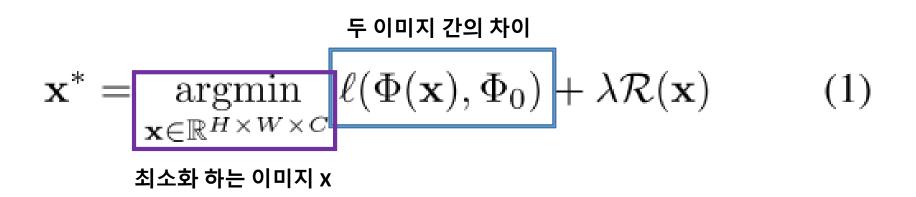
$$\mathbf{x}^* = \underset{\mathbf{x} \in \mathbb{R}^{H \times W \times C}}{\operatorname{argmin}} \ell(\Phi(\mathbf{x}), \Phi_0) + \lambda \mathcal{R}(\mathbf{x}) \tag{1}$$

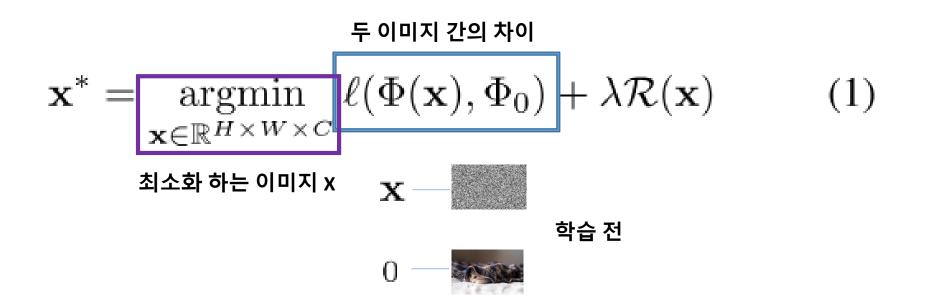
$$\mathbf{x}^* = \underset{\mathbf{x} \in \mathbb{R}^{H \times W \times C}}{\operatorname{argmin}} \ell(\Phi(\mathbf{x}), \Phi_0) + \lambda \mathcal{R}(\mathbf{x}) \tag{1}$$

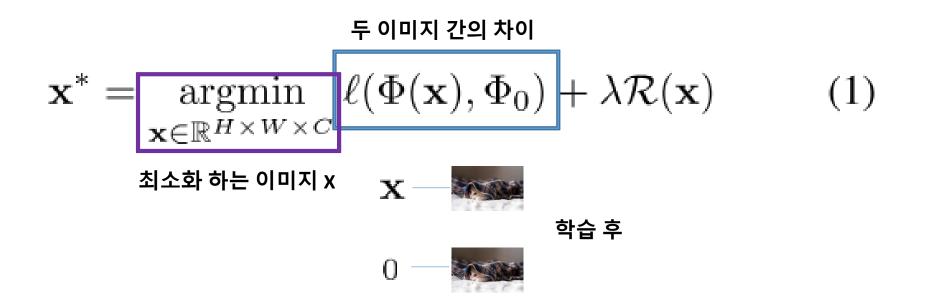
$$\mathbf{x} - \mathbf{x}$$

$$\mathbf{x}^* = \underset{\mathbf{x} \in \mathbb{R}^{H \times W \times C}}{\operatorname{argmin}} \ell(\Phi(\mathbf{x}), \Phi_0) + \lambda \mathcal{R}(\mathbf{x}) \tag{1}$$









### 어떻게 그런 것이 가능하죠?

- 어떻게 이미지의 형상을 유지하지? O
- 이미지에서 스타일을 어떻게 뽑아내지?

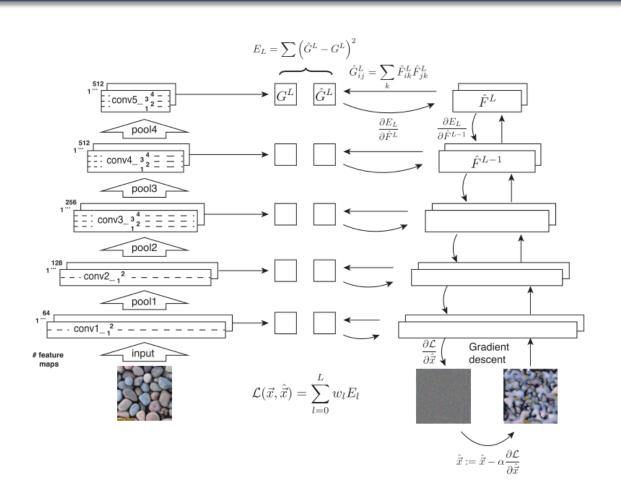
### 이미지에서 스타일을 어떻게 뽑아내지?

#### Texture Synthesis Using Convolutional Neural Networks

#### Leon A. Gatys

Centre for Integrative Neuroscience, University of Tübingen, Germany Bernstein Center for Computational Neuroscience, Tübingen, Germany Graduate School of Neural Information Processing, University of Tübingen, Germany leon.gatys@bethgelab.org

# 이미지에서 스타일을 어떻게 뽑아내지?



$$\hat{G}_{ij}^L = \sum_k \hat{F}_{ik}^L \hat{F}_{jk}^L$$

$$E_L = \sum \left( \hat{G}^L - G^L \right)^2$$

$$\mathcal{L}(\vec{x}, \hat{\vec{x}}) = \sum_{l=0}^{L} w_l E_l$$

https://arxiv.org/abs/1505.07376

# 이미지에서 스타일을 어떻게 뽑아내지?



### 어떻게 그런 것이 가능하죠?

- 어떻게 이미지의 형상을 유지하지? O
- 이미지에서 스타일을 어떻게 뽑아내지? O

#### A Neural Algorithm of Artistic Style

Leon A. Gatys, 1,2,3\* Alexander S. Ecker, 1,2,4,5 Matthias Bethge 1,2,4

<sup>1</sup>Werner Reichardt Centre for Integrative Neuroscience and Institute of Theoretical Physics, University of Tübingen, Germany <sup>2</sup>Bernstein Center for Computational Neuroscience, Tübingen, Germany

처음 Neural Style Transfer라는 개념을 사용한 논문









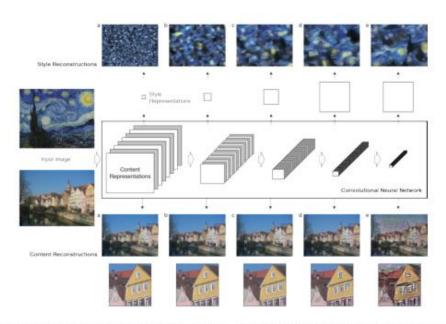


Figure 1: Convolutional Neural Network (CNN). A given input image is represented as a set of filtered images at each processing stage in the CNN. While the number of different filters increases along the processing hierarchy, the size of the filtered images is reduced by some downsampling mechanism (e.g. max-pooling) leading to a decrease in the total number of units per layer of the network. Content Reconstructions. We can visualise the information

처음 Neural Style Transfer라는 개념을 사용한 논문

$$\mathcal{L}_{total}(\vec{p}, \vec{a}, \vec{x}) = \alpha \mathcal{L}_{content}(\vec{p}, \vec{x}) + \beta \mathcal{L}_{style}(\vec{a}, \vec{x})$$

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$$\mathcal{L}_{total}(\vec{p}, \vec{a}, \vec{x}) = \alpha \mathcal{L}_{content}(\vec{p}, \vec{x}) + \beta \mathcal{L}_{stule}(\vec{a}, \vec{x})$$

이미지 형상을 유지 스타일 정보를 같게

$$\mathcal{L}_{total}(\vec{p}, \vec{a}, \vec{x}) = \alpha \mathcal{L}_{content}(\vec{p}, \vec{x}) + \beta \mathcal{L}_{style}(\vec{a}, \vec{x})$$

이미지 형상을 유지 스타일 정보를 같게 
$$\mathcal{L}_{total}(\vec{p}, \vec{a}, \vec{x}) = \alpha \mathcal{L}_{content}(\vec{p}, \vec{x}) + \beta \mathcal{L}_{style}(\vec{a}, \vec{x})$$
 어느 정도로? (비율)













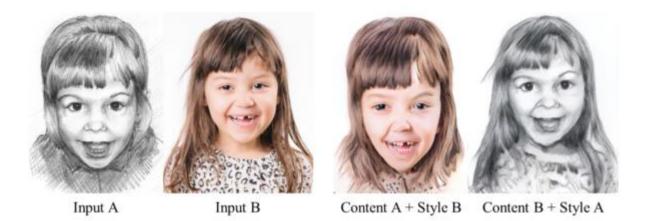
# 여러 종류의 Style Transfer

- Image Iteration
  - MMD-based
  - MRF-based
- Model Iteration

- Image Iteration
  - MMD-based
  - MRF-based
- Model Iteration

- Image Iteration
  - MMD-based
  - MRF-based
- Model Iteration

#### **Markov Random Fields based**



#### **Markov Random Fields based**

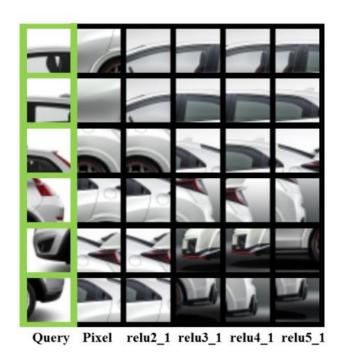
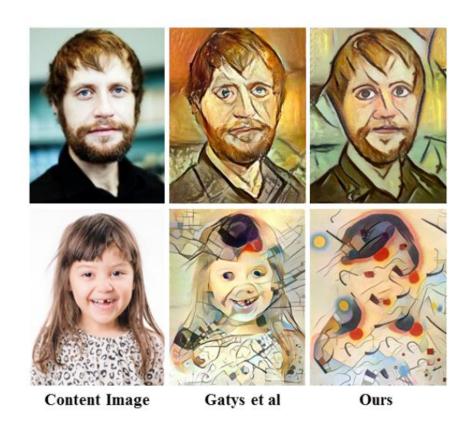


Figure 3: Comparison of patch matching at different layers of a VGG network.

#### **Markov Random Fields based**



### Can it be applied to video?



- Image Iteration
  - MMD-based
  - MRF-based
- Model Iteration

- Image Iteration
  - MMD-based
  - MRF-based
- Model Iteration

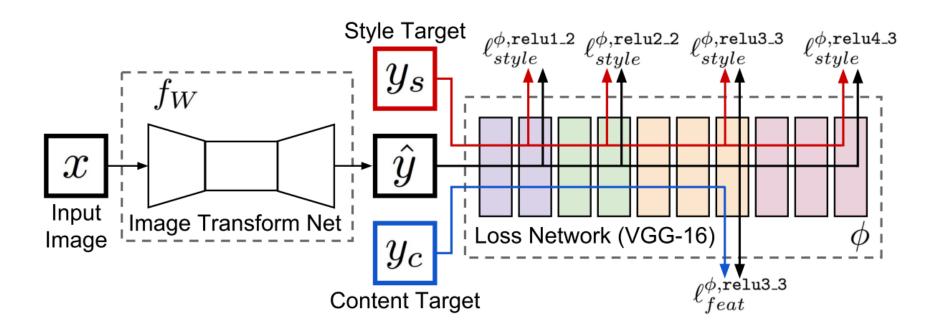
#### "Fast" Neural Style Transfer

#### Perceptual Losses for Real-Time Style Transfer and Super-Resolution

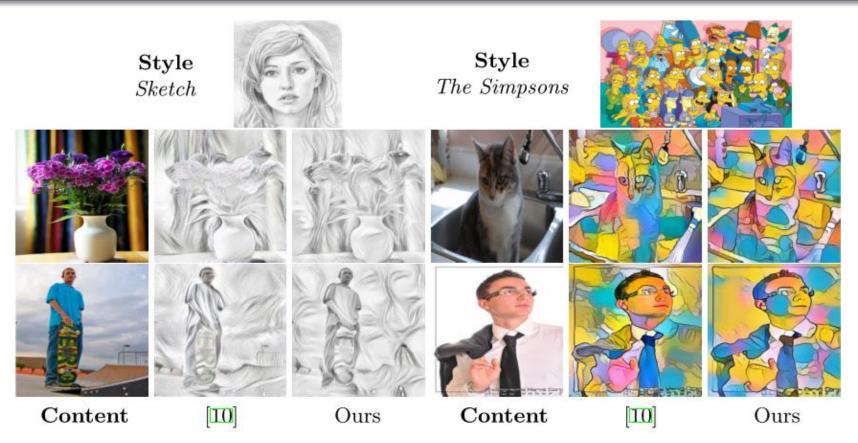
Justin Johnson, Alexandre Alahi, Li Fei-Fei {jcjohns, alahi, feifeili}@cs.stanford.edu

Department of Computer Science, Stanford University

### "Fast" Neural Style Transfer



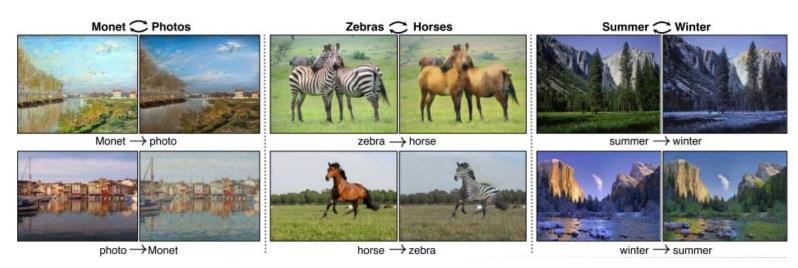
### "Fast" Neural Style Transfer





#### **Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks**

Jun-Yan Zhu\* Taesung Park\* Phillip Isola Alexei A. Efros Berkeley AI Research (BAIR) laboratory, UC Berkeley



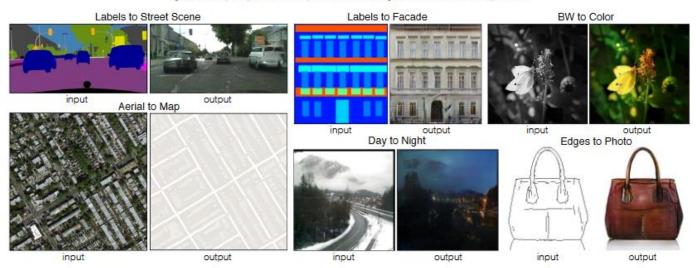
### Cycle GAN is a follow up to Pix2Pix

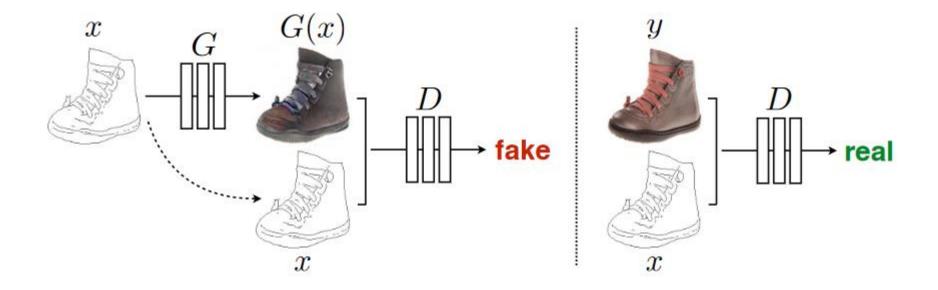
#### Image-to-Image Translation with Conditional Adversarial Networks

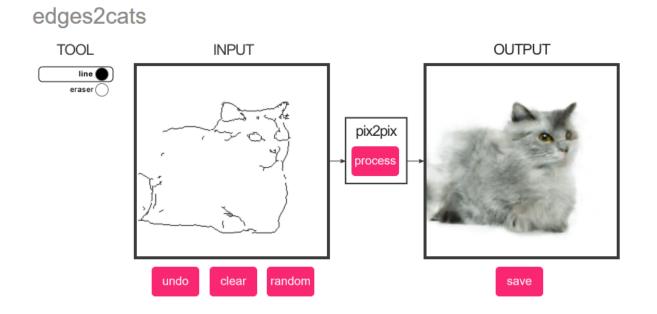
Phillip Isola Jun-Yan Zhu Tinghui Zhou Alexei A. Efros

Berkeley AI Research (BAIR) Laboratory, UC Berkeley

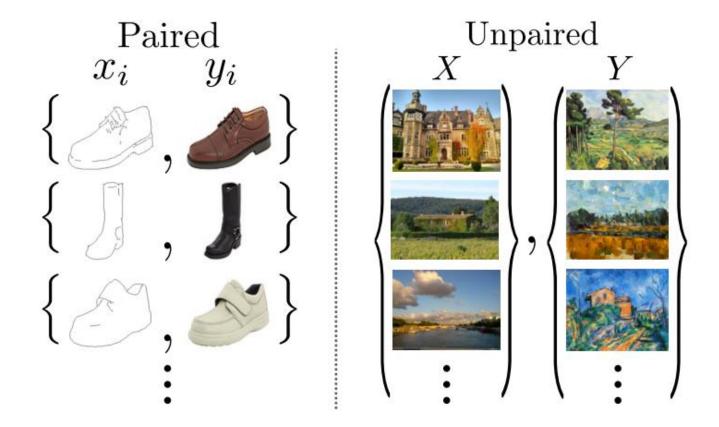
{isola, junyanz, tinghuiz, efros}@eecs.berkeley.edu



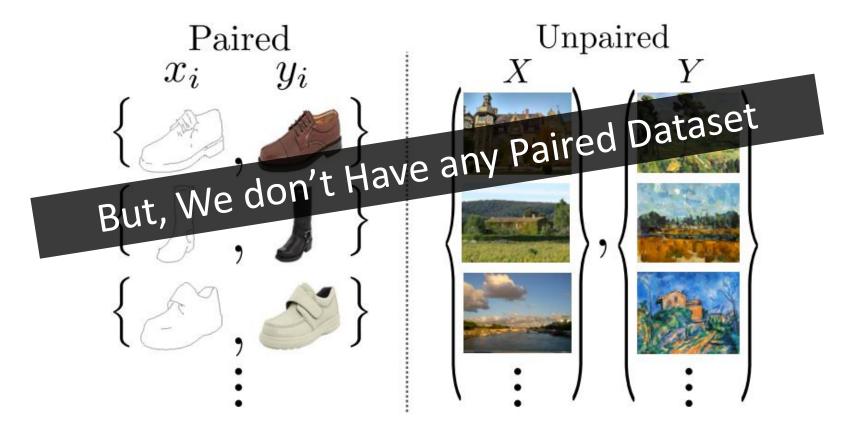




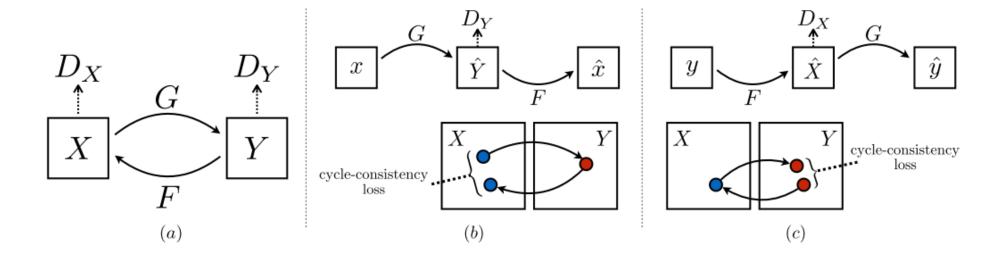
https://affinelayer.com/pixsrv/

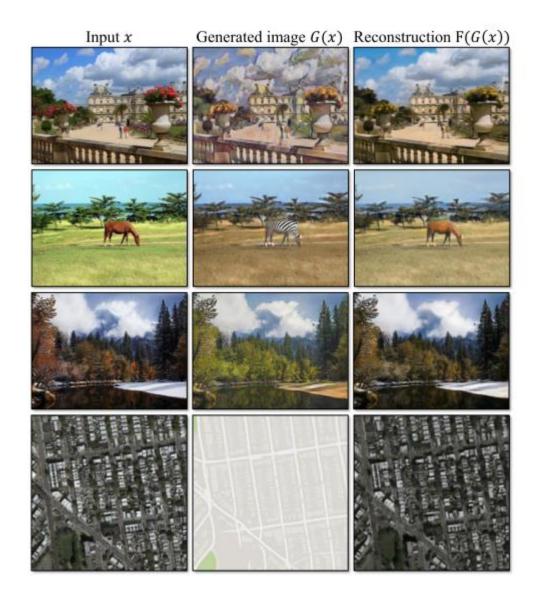


https://arxiv.org/abs/1703.10593



https://arxiv.org/abs/1703.10593





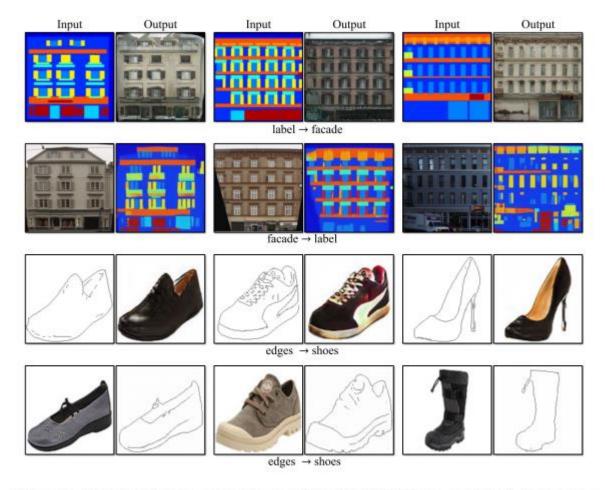


Figure 8: Example results of CycleGAN on paired datasets used in "pix2pix" [21] such as architectural labels↔photos and edges↔shoes.

### **Photorealistic Image Stylization**

#### A Closed-form Solution to Photorealistic Image Stylization

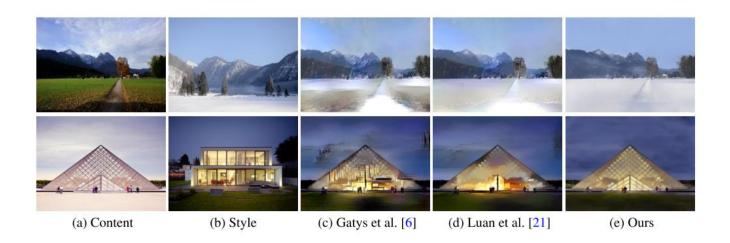
Yijun Li<sup>1</sup>, Ming-Yu Liu<sup>2</sup>, Xueting Li<sup>1</sup>, Ming-Hsuan Yang<sup>1,2</sup>, and Jan Kautz<sup>2</sup>

<sup>1</sup>University of California, Merced

<sup>2</sup>NVIDIA

{yli62,xli75,mhyang}@ucmerced.edu

{mingyul, jkautz}@nvidia.com



### **Photorealistic Image Stylization**

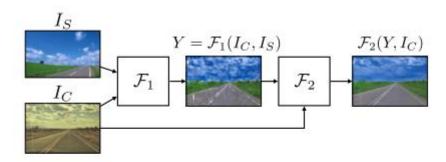
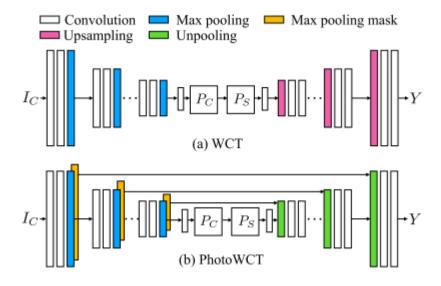


Figure 2: The proposed photorealistic image style transfer algorithm consists of two closed-form function mappings:  $\mathcal{F}_1$  and  $\mathcal{F}_2$ . While  $\mathcal{F}_1$  maps  $I_C$  to an intermediate image carrying the style of  $I_S$ ,  $\mathcal{F}_2$  removes noticeable artifacts and produces a photorealistic stylized result.



## **Photorealistic Image Stylization**



## **Application Style Transfer**

## **Instagram Filter**





shtowever

shtowever 호주 또가고싶다







sbeinone, pack\_won\_jin, jjgkpp, whitehorse\_tcmn님이 좋아합니다

2016년 10월 16일

댓글 달기...

...

### **Style Transfer in Movie**

#### Bringing Impressionism to Life with Neural Style Transfer in Come Swim

Bhautik J Joshi\* Research Engineer, Adobe Kristen Stewart Director, Come Swim David Shapiro Producer, Starlight Studios







### **Style Transfer in Movie**

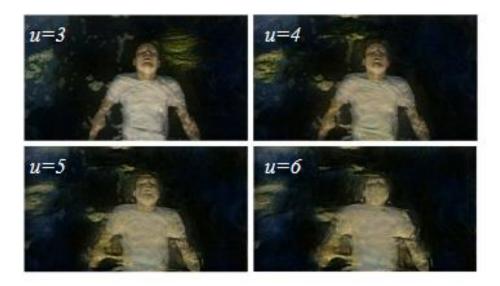
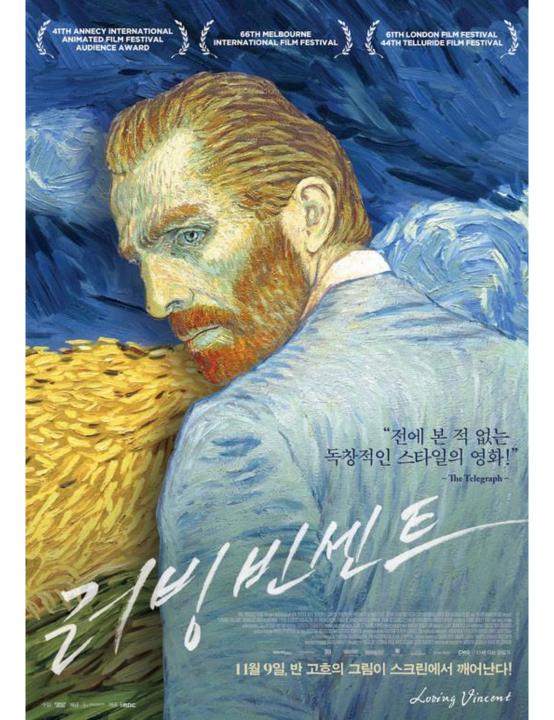
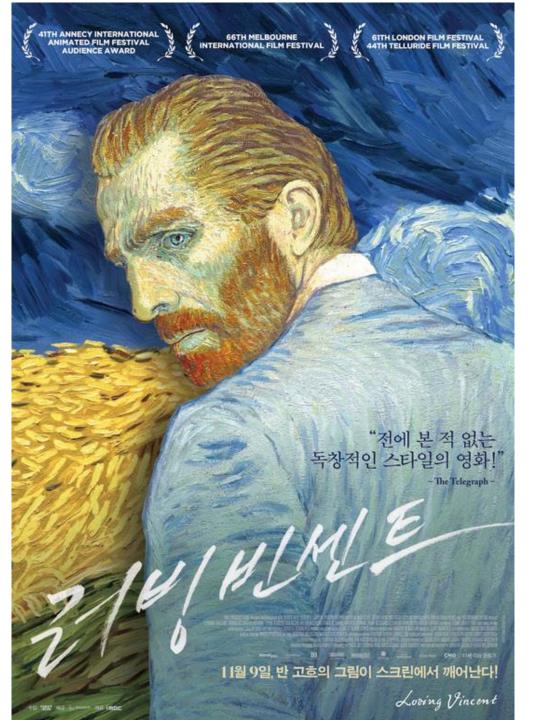


Figure 6: Increasing the value of u gave us a control to fine-tune the degree of subjectively-measured impressionism present in the style transfer

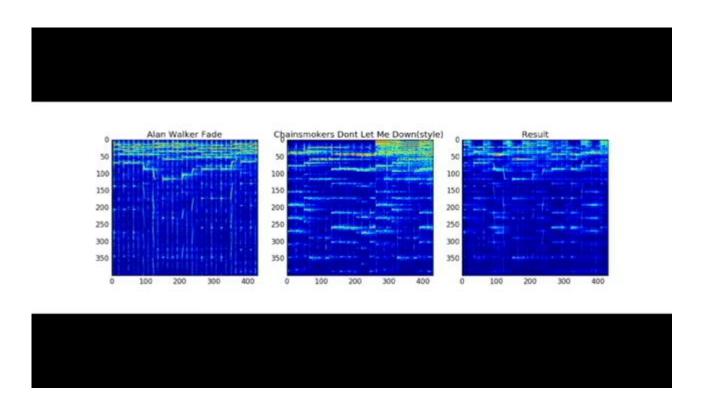




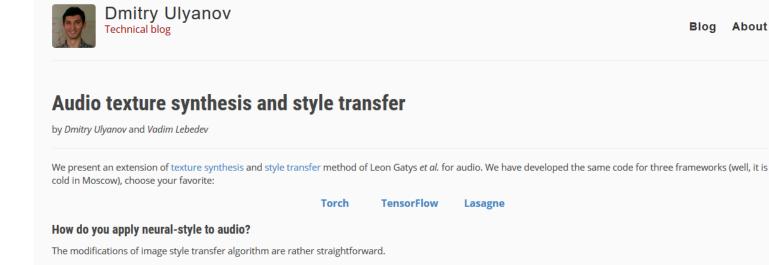


IF
They Used Style Transfer Methods..

## **Neural Style Transfer to Audio**



### **Neural Style Transfer to Audio**



are implemented for all three frameworks. Interestingly, the network we use has only one layer with 4096 filters.

And finally we need to reconstruct a signal from its spectrogram. The simplest way to do the inversion is to use Griffin-Lim algorithm.

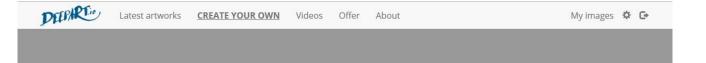
image. In fact is better to think of spectrogram as of 1xT image with F channels.

• The raw audio is converted to a spectrogram via Short Time Fourier Transform. Spectrogram is a 2D representation of a 1D signal so it can be treated (almost) as an

 Next we need a network. We cannot just use VGG-19, since 3x3 convolutions are not suited for our essentially 1D problem, for which we for sure want to use 1D convolutions. Then there are two options: use a pretrained network or use completely random weights. In Torch implementation I tried to train different kind of nets, but they seem to perform similarly. As [1,2,3] Vadim also found that quality of the network is not important for texture synthesis. Nets with random weights

https://dmitryulyanov.github.io/audio-texture-synthesis-and-style-transfer/

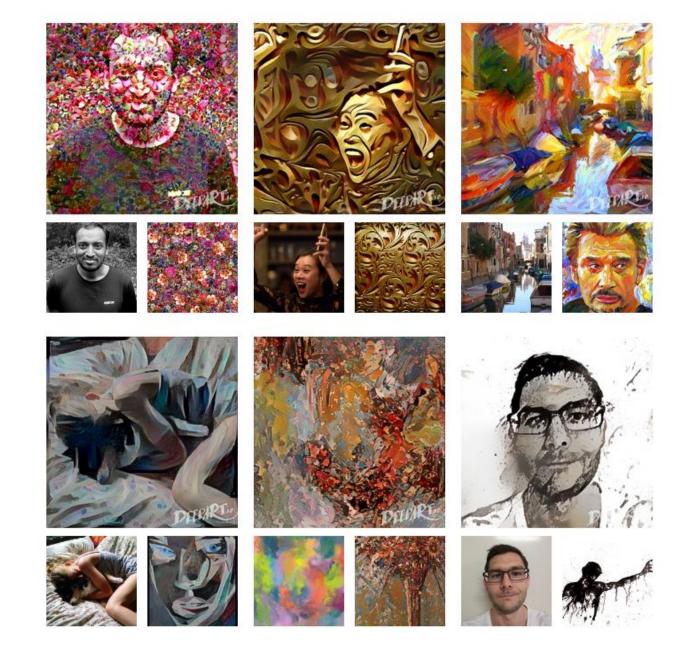
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#### **NIPS 2017 Poster Contest**

### MAKE NIPS PRETTY

Let's make the conference center pretty together! You create neural art – we will render the best 50 pieces in high resolution and print them on posters for the conference center. At the conference, we'll have a vote. The winner will receive an NVIDIA DGX Station, the fastest personal supercomputer for researchers and data scientists.

View submissions

Rules

Sponsors



By José ♥ 39



By Bluebird

♥ 0



By Rubeolo ♥ 10







By Prateep Bhattacharjee

♥ 110

## Fashion style transfer



Figure 1: (a) and (b) provide the shape & style respectively (c) Final Design

### Fashion style transfer



## Fashion style transfer







neural style transfer based on keras

## **QR Code With Style Transfer**



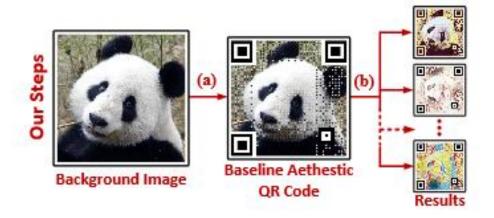


They used style transfer conv nets to customize QR Codes arxiv.org/pdf/1803.01146...

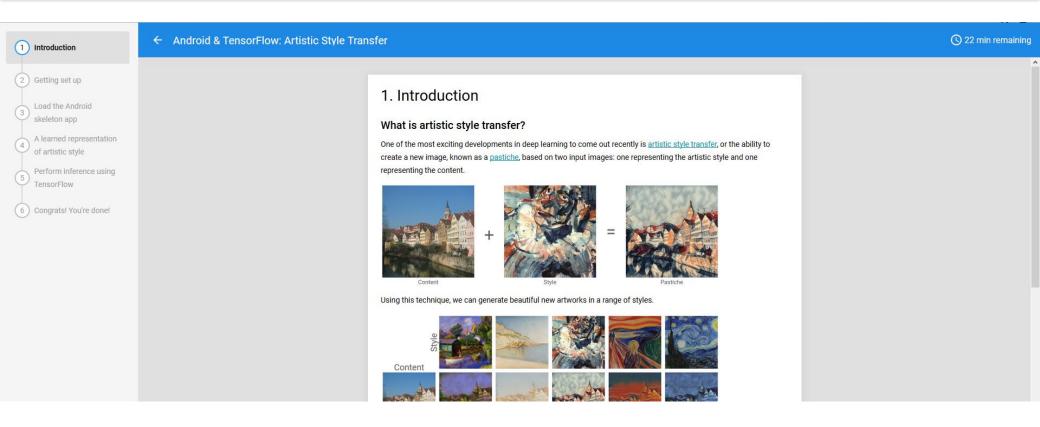




## **QR Code With Style Transfer**



#### Do It Yourself



https://codelabs.developers.google.com/codelabs/tensorflow-style-transfer-android/

#### Do It Yourself

#### 2. Getting set up

#### **Get the Code**

There are two ways to grab the source for this codelab: either download a ZIP file containing the code, or clone it from GitHub.

#### ZIP Download

Click the following button to download all the code for this codelab:



Unpack the downloaded zip file. This will unpack a root folder (tensorflow-style-transfer-android-codelab-start), which contains the base app we'll work on in this codelab, including all of the app resources.

#### Check out from GitHub

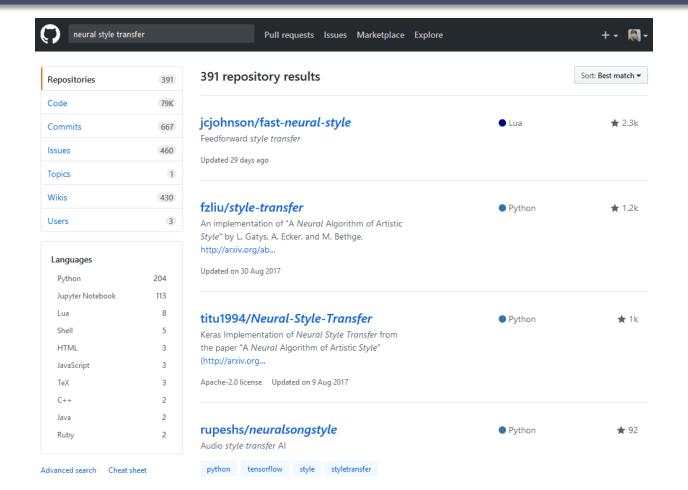
Check the code out from GitHub:

git clone https://github.com/googlecodelabs/tensorflow-style-transfer-android

This will create a directory containing everything you need. If you change into it you can use git checkout codelab-start and git checkout codelab-finish to switch between the start & end of the lab, respectively.

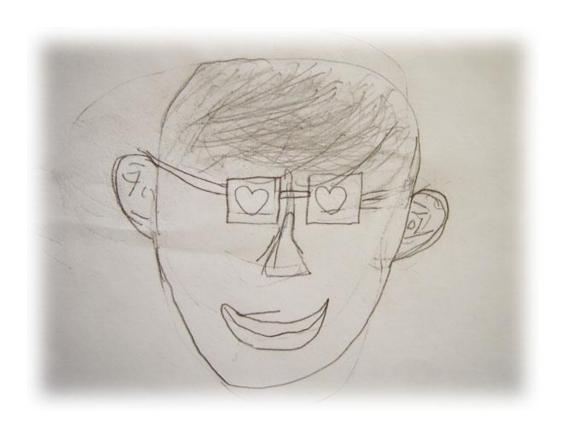
https://codelabs.developers.google.com/codelabs/tensorflow-style-transfer-android/

#### Do It Yourself

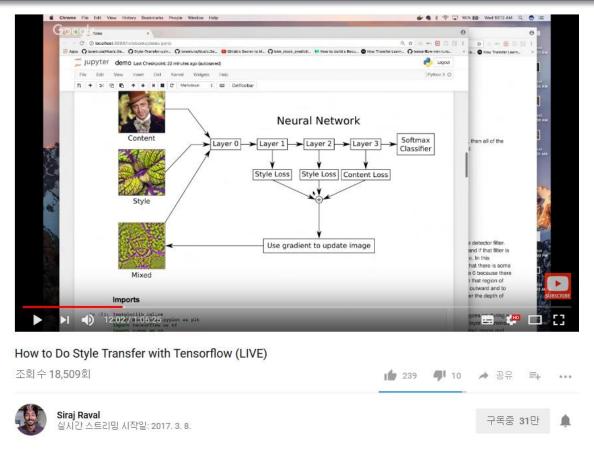


예술과 딥러닝 그 사이

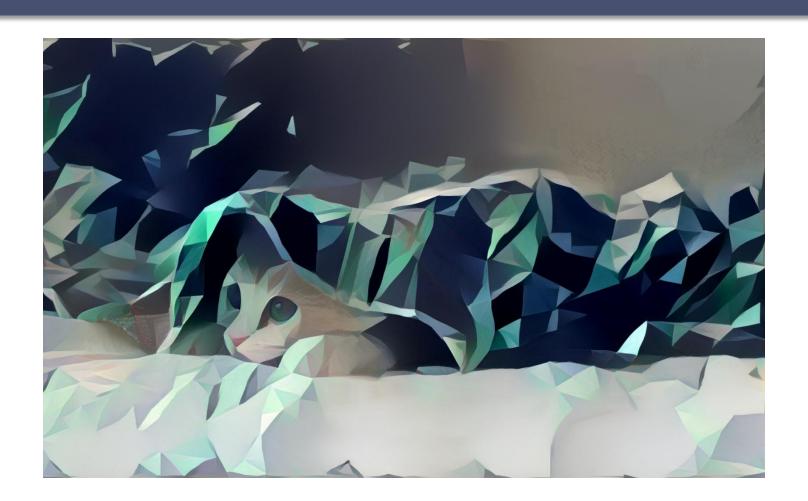
# 예술에 서투른 사람들에게



## 예술에 서투른 사람들에게



# 예술에 서투른 사람들에게



### Thanks You!