

Lecture 1: Introduction

Jun-Yan Zhu 16-726, Spring 2024

Teaching Staff

Instructors



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Teaching Assistants



Jaskaran Singh Sodhi

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Hariharan Ravichandran

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- Computer Vision, Computer Graphics, Machine Learning, **Computational Photography**
- Love pets (cat & dog)
- Swimming, tennis, gaming.



Cat Paper Collection

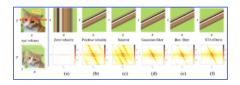
As reported by Cisco, 90% of net traffic will be visual, and indeed, most of the visual data are cat photos and videos. Thus, understanding, modeling, and synthesizing our feline friends becomes a more and more critical research problem these days, especially for our cat lovers.

Cat Paper Collection is an academic paper collection that includes computer graphics, computer vision, and machine learning papers that produce experimental results related to **cats**. If you would like to add/remove an article, please send an email to **Jun-Yan Zhu** (junyanz at cs dot cmu dot edu). We thank all the authors for their contribution and support.

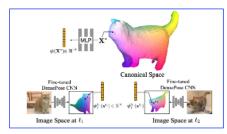
See also GitHub | CSV file



Efficient Halftoning via Deep Reinforcement Learning Haitian Jiang, Dongliang Xiong, Xiaowen Jiang, Li Ding, Liang Chen, Kai Huang In IEEE TIP 2023 [Paper]



Learning Spatio-Temporal Downsampling for Effective Video Upscaling Xiaoyu Xiang, Yapeng Tian, Vijay Rengarajan, Lucas Young, Bo Zhu, Rakesh Ranjan In ECCV 2022 [Paper]



BANMo: Building Animatable 3D Neural Models from Many Casual Videos Gengshan Yang, Minh Vo, Natalia Neverova, Deva Ramanan, Andrea Vedaldi, Hanbyul Joo In CVPR 2022 [Paper] [Project]



Jaskaran Singh Sodhi

- MRSD student at RI
- Working on Subcanopy Wildfire Monitoring with AirLab
- Interested in applications of deep learning to perception and motion planning



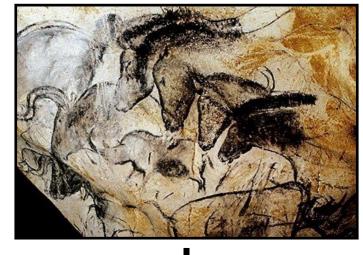
Hariharan Ravichandran

- 2nd year MRSD student in RI
- Project: "Autonomous Construction on Lunar-like Terrains" under Prof. Red Whittaker
- Currently working on learningbased planning methods for offroad driving



Visual Content Creation

Cave art



Time ______ 32,000 BC



Visual Content Creation

Painting

Cave art

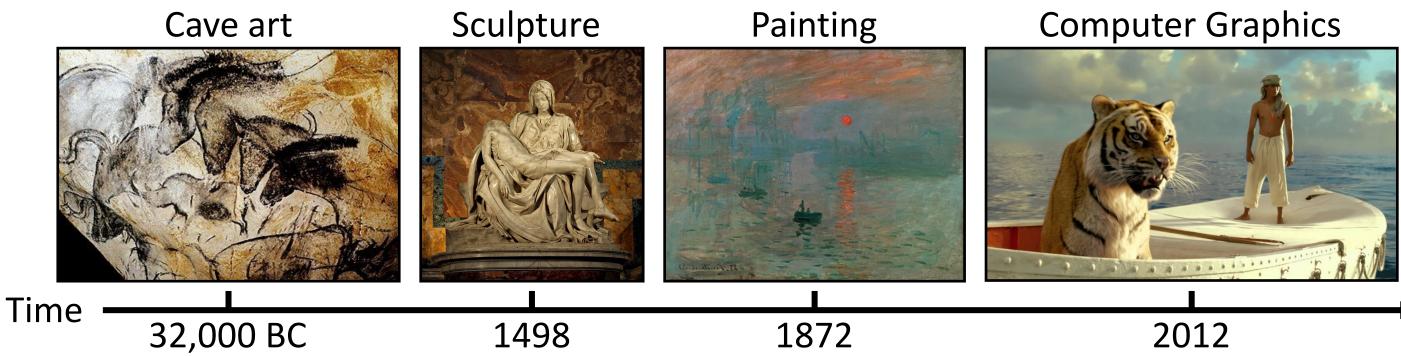
Sculpture



Time 32,000 BC 1498 1872



Visual Content Creation



Who is creating visual content?

Cave art

32,000 BC

Sculpture

1872

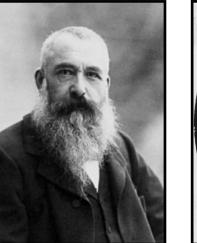
Painting



Time







Claude Monet Paul Cezanne



George Lucas

Michelangelo

Computer Graphics

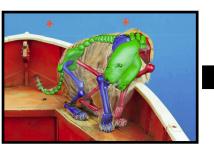
2012



Ang Lee

Who is creating visual content?



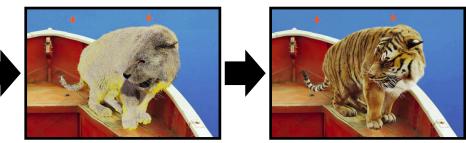


Skeleton



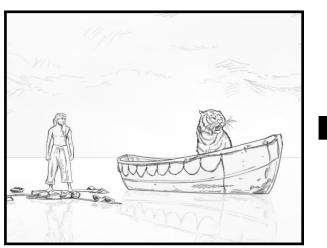
Geometry





Details

Ang Lee



Idea

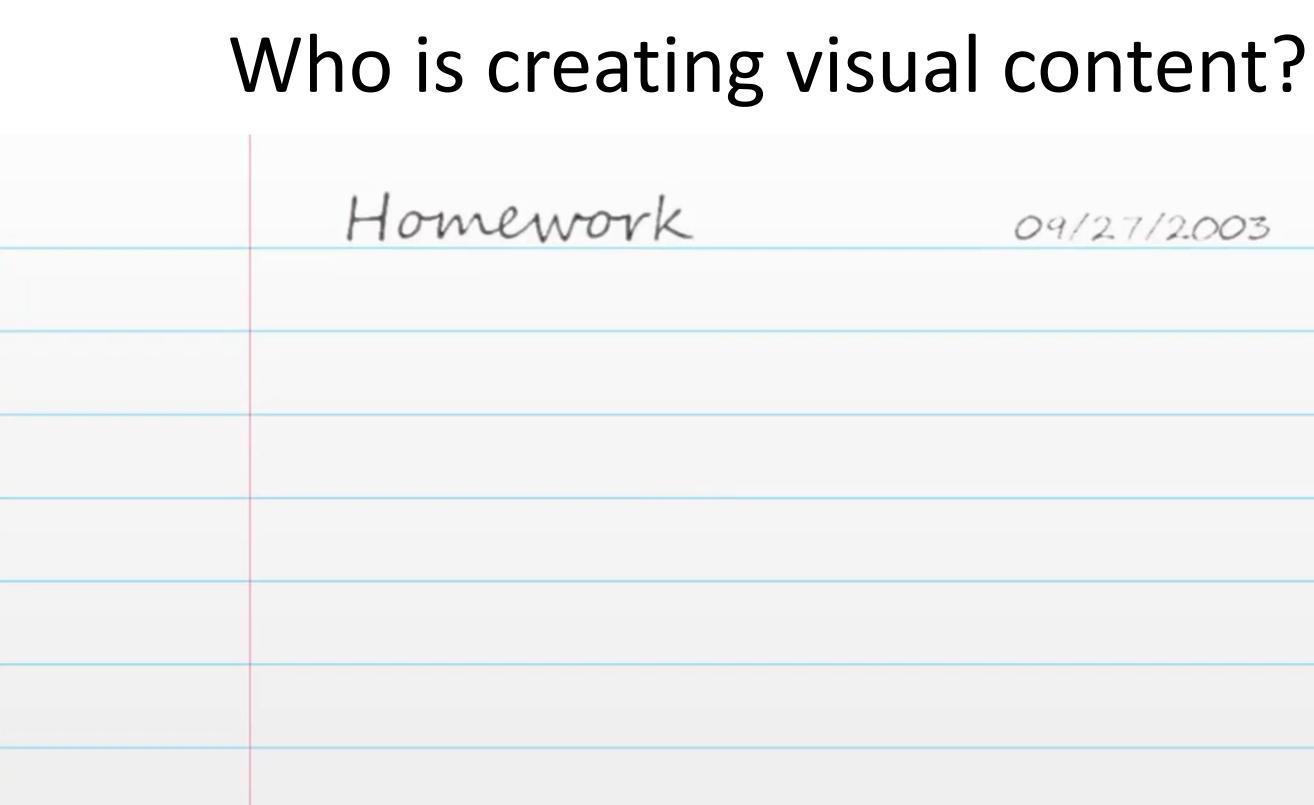




Image

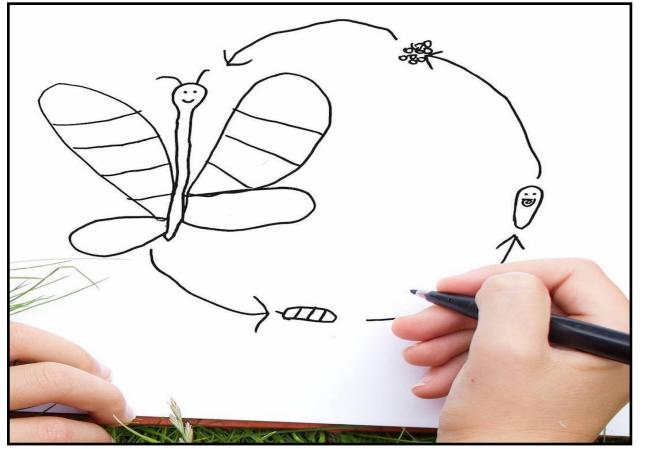


Visual Content 11





Who is creating visual content?



Kid's drawing



Photoshop result by his father

© Tom¹Curtis

Creating Visual Realism Manually



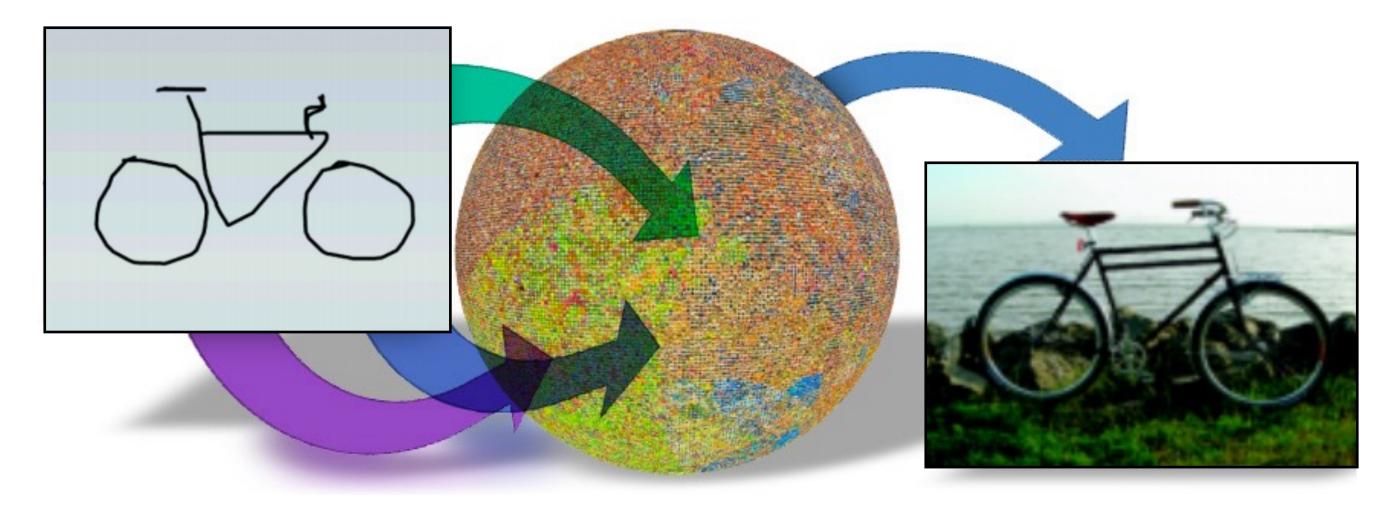
CG office (more details)



My advisor's office

Data-Driven Graphics (2000s)

Graphics \rightarrow Image Retrieval



Picture from James Hays

Data-Driven Graphics (2000s)

Compositing multiple parts



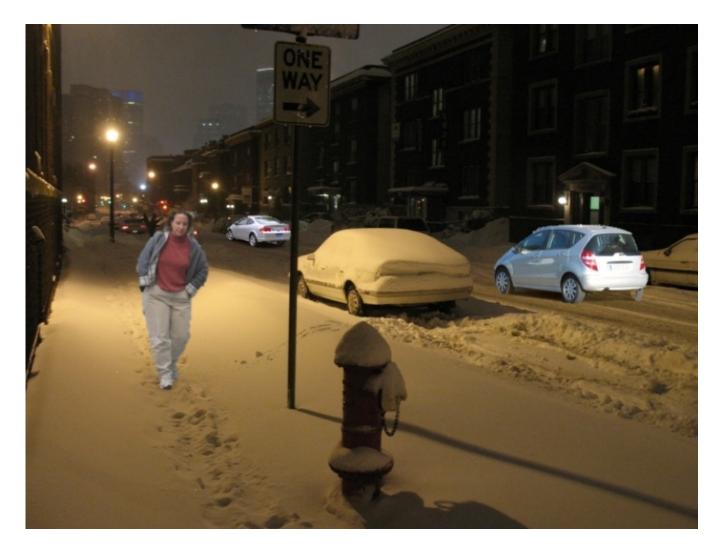
User Input

Database images

Sketch2Photo [Tao et al. SIGGRAPH Asia⁶2009]

Output

Data-Driven Graphics (2000s)





- Hard to combine pieces
- No understanding of visual realism

Help everyone easily create visual content

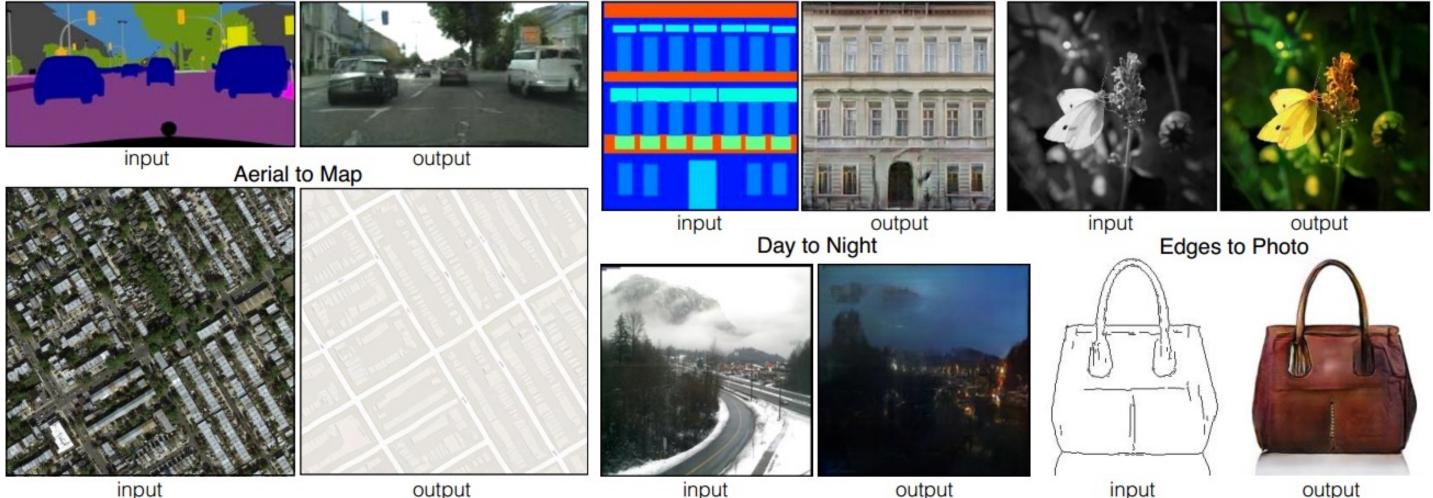
Teach machines how to create realistic content



Image-to-Image Translation with pix2pix

Labels to Facade

Labels to Street Scene

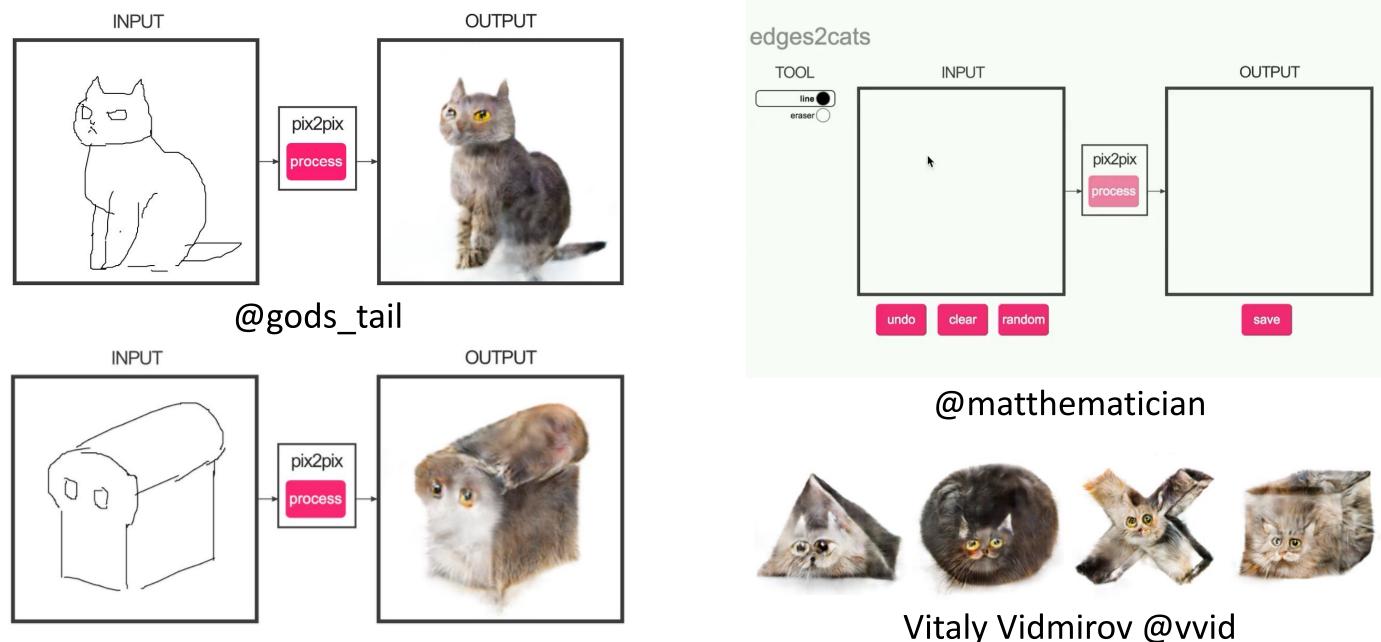


BW to Color

output

pix2pix [Isola, Zhu, Zhou, Efros. CVPR⁹2017]

#edges2cats with pix2pix

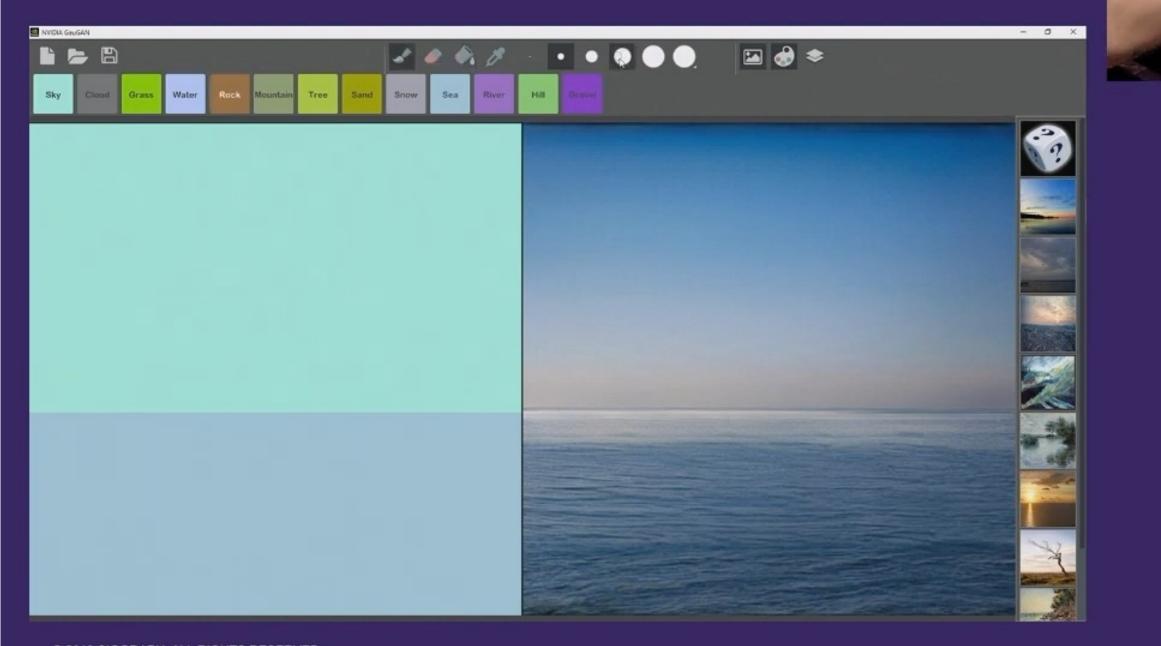


lvy Tasi @ivymyt

By Christopher Hesse https://affinelayer.com/pixsrv/

	OUTPUT	
]_		
	save	

GauGAN [Park, Liu, Wang, Zhu. 2019]



SIGGRAPH 2019 Real-time Live! "Best of Show Award" and "Audience Choice Award



E Award 28 JULY - 1 ADGUST

GauGAN input

and the second states

Giorno go

Compo o

GauGAN result

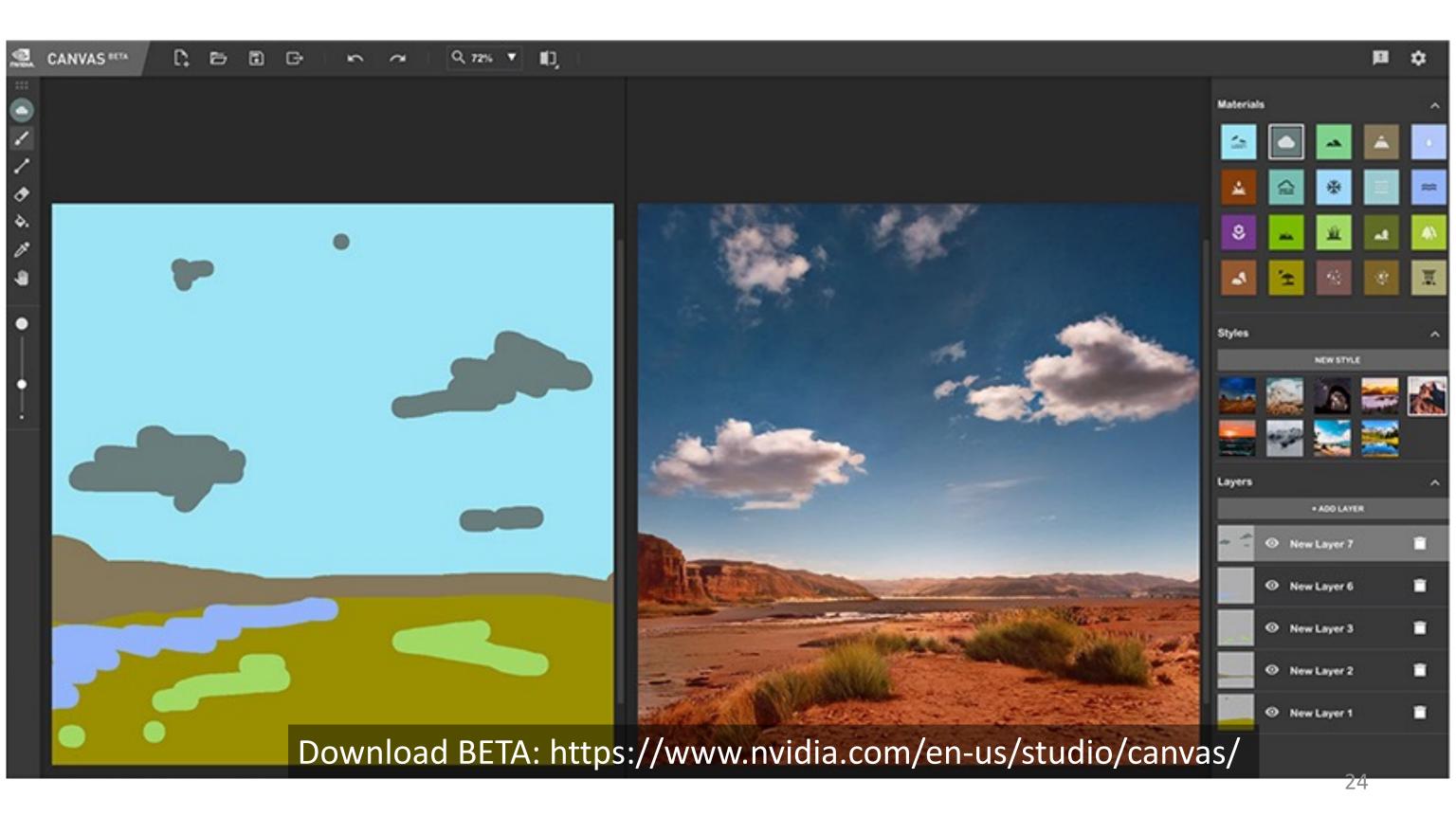
By Darek Zabrocki, Concept Designer and Illustrator

GauGAN input

GauGAN result

By Darek Zabrocki, Concept Designer and Illustrator





Collection Style Transfer



Photograph ©Alexei Efros



Monet







CycleGAN [Zhu, Park, Isola, Efros. 2017]

Cezanne

Van Gogh



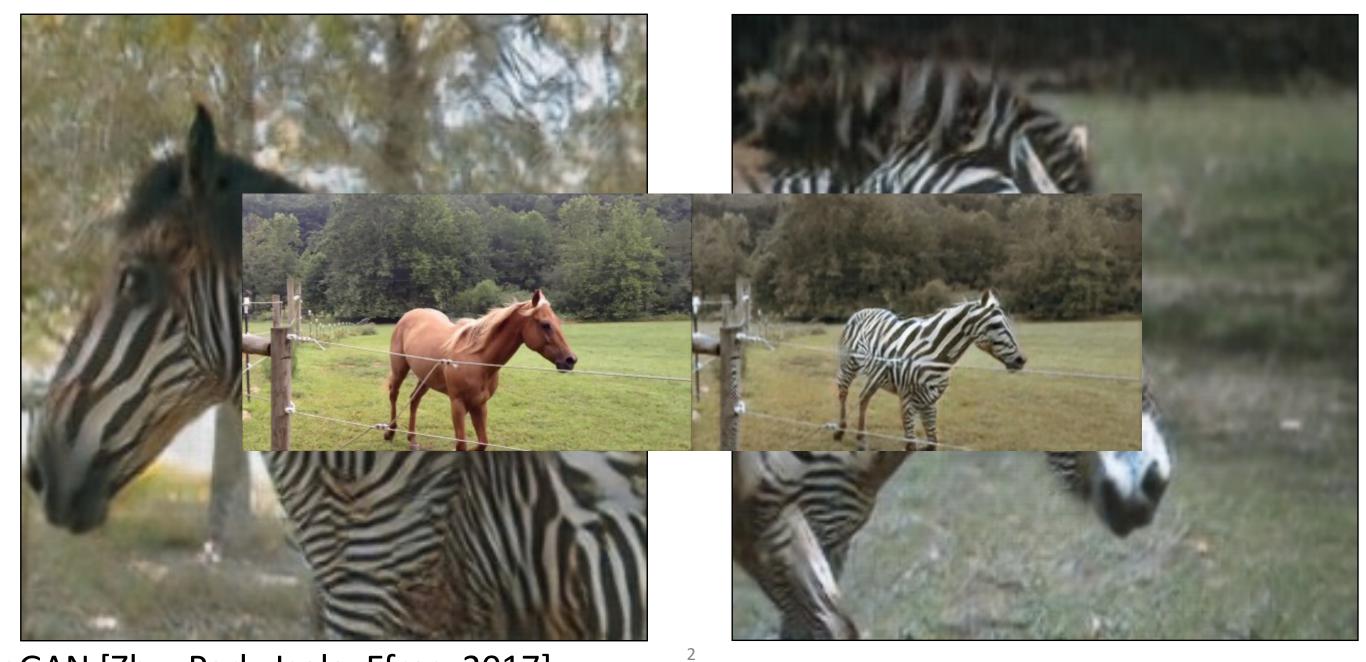


Monet's paintings \rightarrow photographic style



CycleGAN [Zhu, Park, Isola, Efros. 2017]

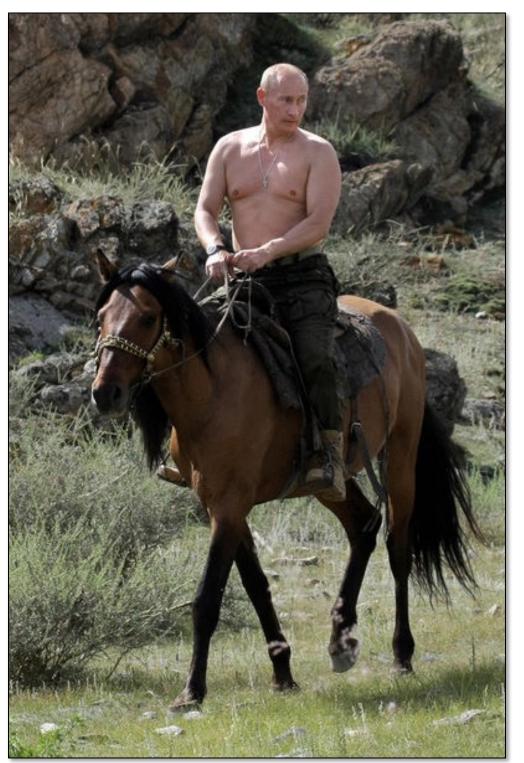
Horse \rightarrow Zebra



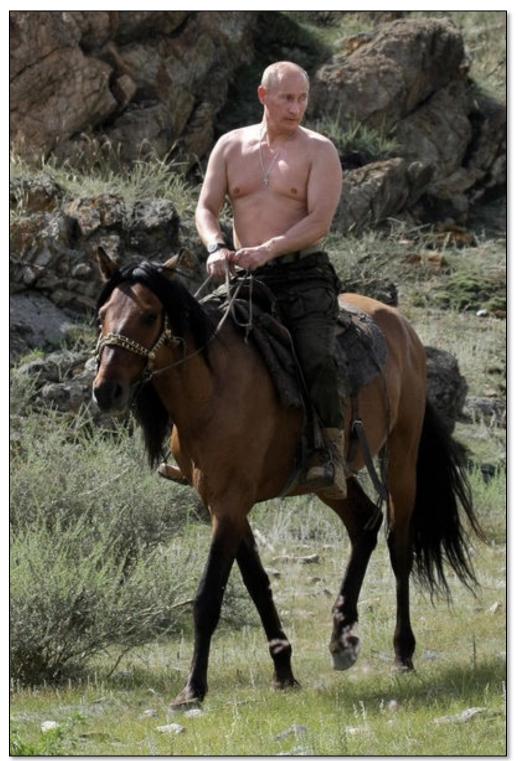
7

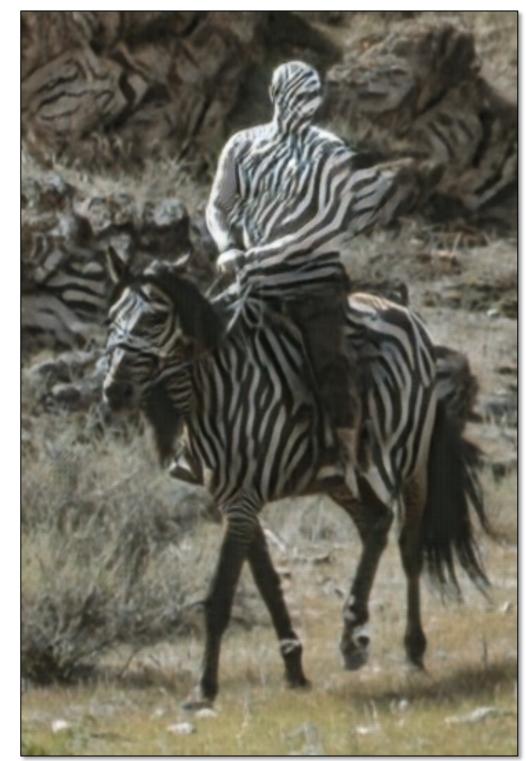
CycleGAN [Zhu, Park, Isola, Efros. 2017]

Failure case



Failure case





Swapping Autoencoder For Deep Image Manipulation Taesung Park¹, Jun-Yan Zhu², Oliver Wang², Jingwan Lu², Eli Shechtman², Alexei Efros¹, Richard Zhang² ¹UC Berkeley, ²Adobe Research

Swapping Autoencoder [Park et al. NeurIPS 2020]

<not ads>



PS



Photoshop 2021 Neural Filters



</not_ads>

Custom Stable Diffusion



A photo of a moongate



A moongate in the snowy ice



A squirrel in front of moongate



A digital illustration of a V* dog in front of a moongate



A photo of a V* dog

User input images



A V* dog in a swimming pool



A V* dog wearing sunglasses

Single-concept generation

custom-diffusion [Nupur Kumari et al.,³2022]

Multi-concept composition

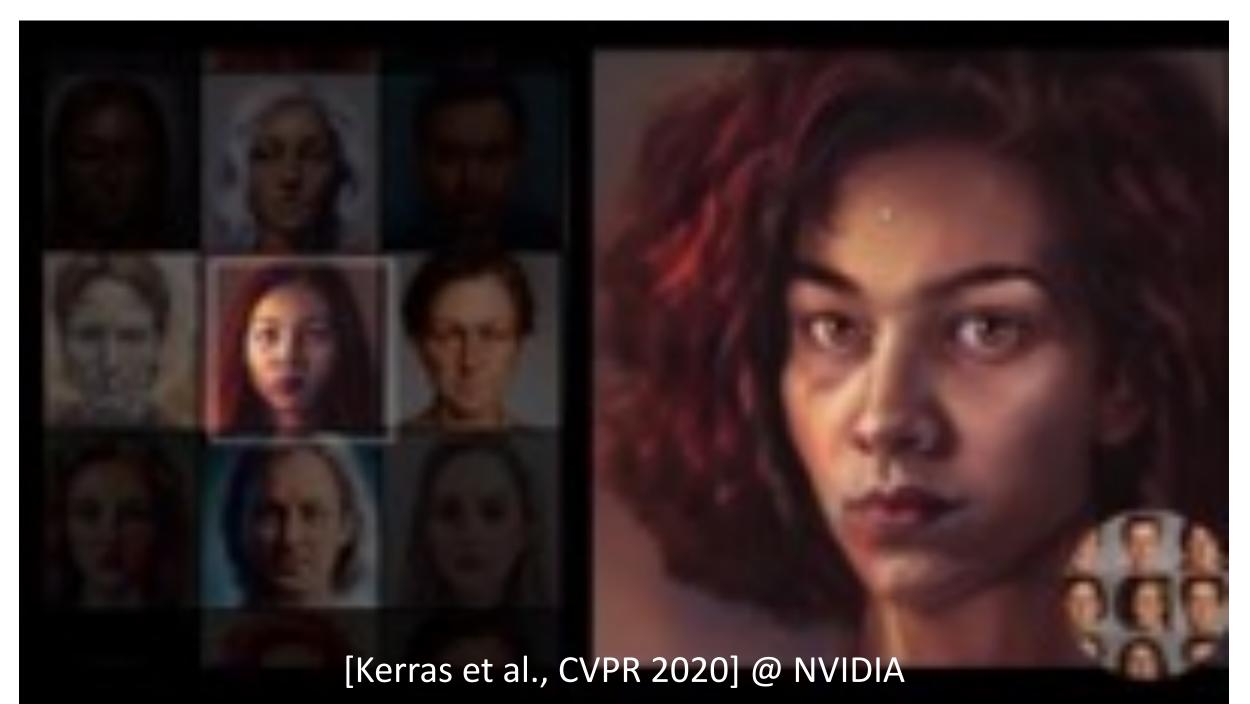
V* **dog** wearing sunglasses in front of a moongate





Research Highlights from other universities & industry labs

Synthesizing High-res Portraits



Everybody Dances Now



Caroline Chan, Shiry Ginosar, Tinghui Zhou, Alexei A. Efros, ICCV 2019. @ UC Berkeley

Neural Talking-Head Synthesis



Compressed videos at the same bit-rate

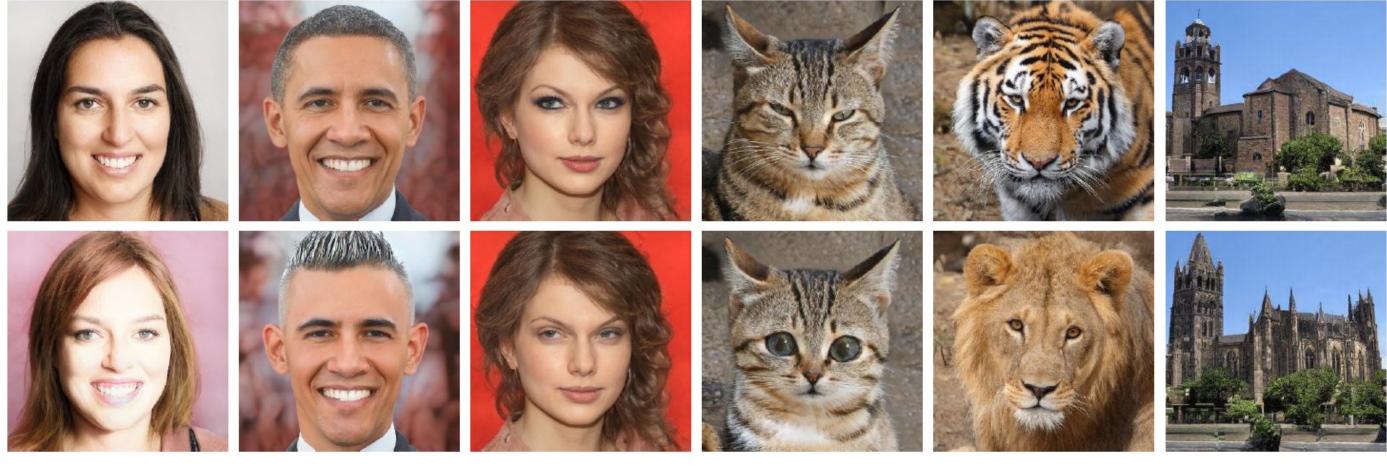
face-vid2vid: One-Shot Free-View Neural Talking-Head Synthesis for Video Conferencing Ting-Chun Wang, Arun Mallya, Ming-Yu Liu. CVPR 2021 @ NVIDIA

NeRF in the Wild

[Martin-Brualla et al., CVPR 2021] @ Google Research



Text-based Image Editing



"Emma Stone"

"Mohawk hairstyle"

"Without makeup"

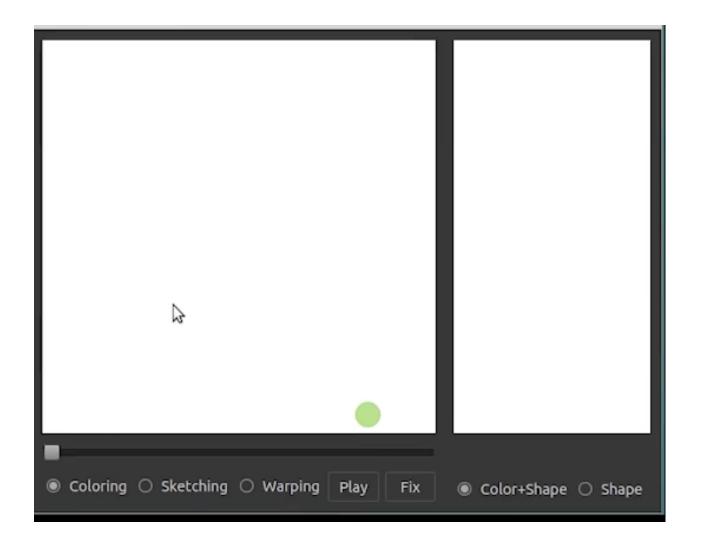
"Cute cat"

"Lion"

StyleCLIP [Or Patashnik*, Zongze Wu*, et al., ICCV 2021]



"Gothic church"



iGAN [Zhu et al., ECCV 2016]

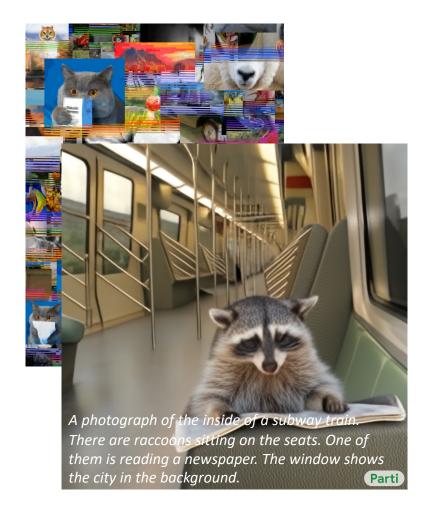


A toilet seat sits open in the grass field.

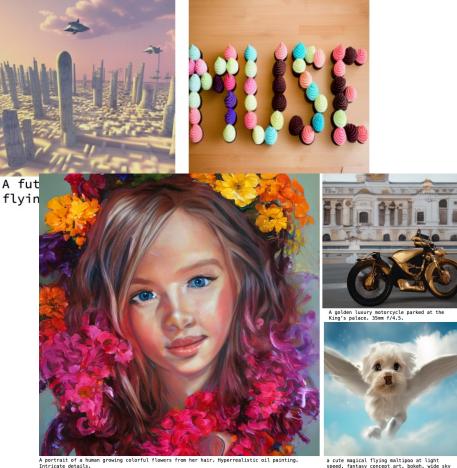
Text2image [Mansimov et al., ICLR 2016]



Diffusion models (DALL-E 2, Imagen, SD)



Autoregressive models (Image GPT, Parti)



GANs, Masked GIT (GigaGAN, MUSE)







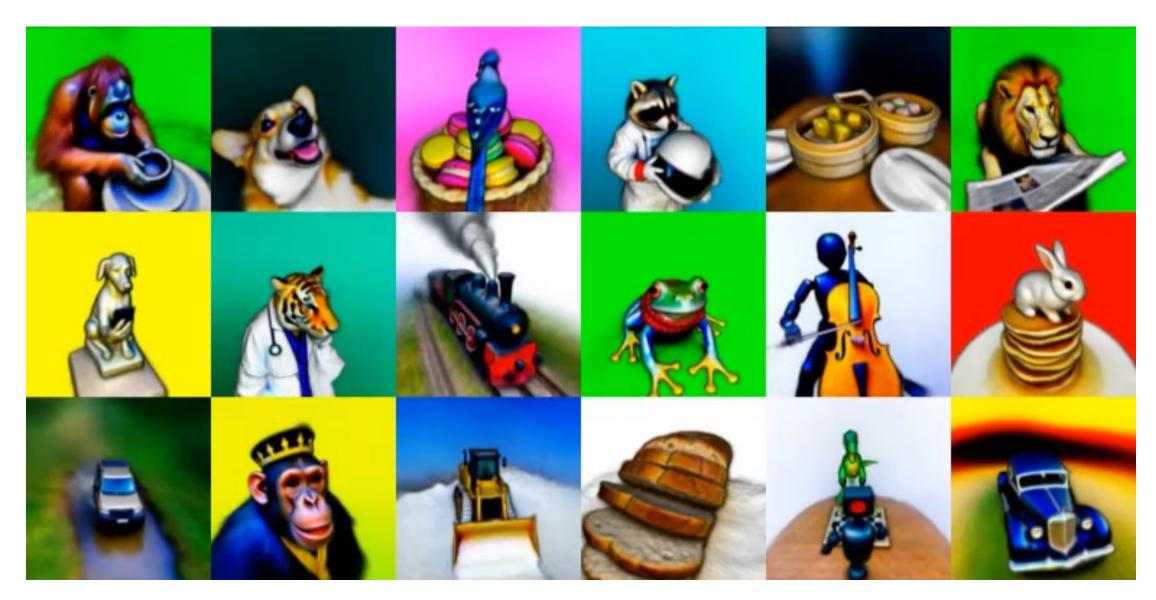


By DALL·E 3

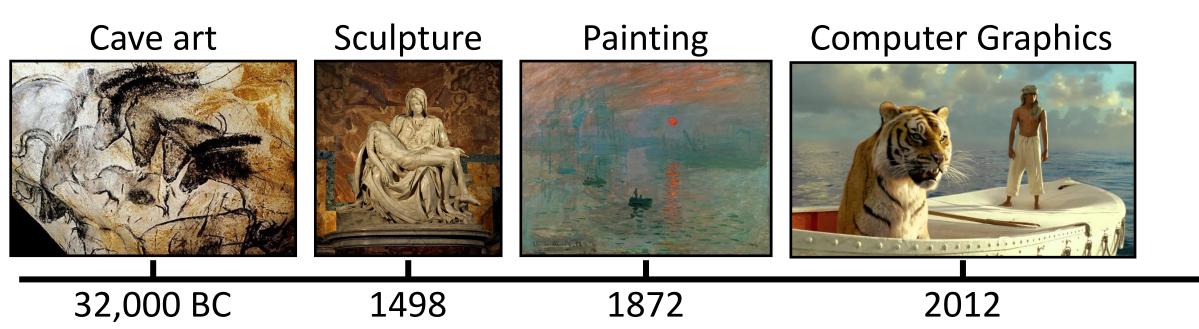
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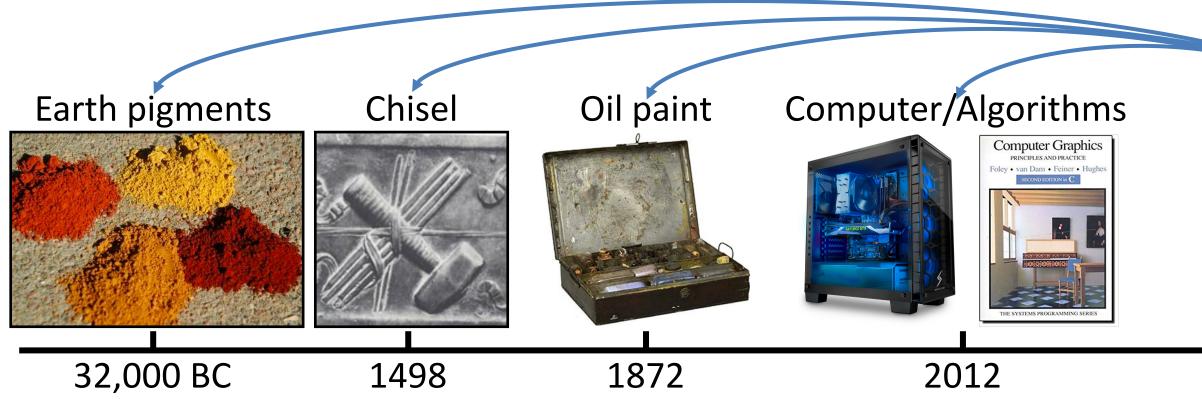
*The computation time is skipped in this video.

Text-Guided Synthesis of Eulerian Cinemagraphs [Mahapatra et al., SIGGRAPH Asia 2023]



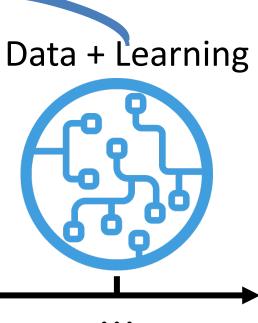
DreamFusion: Text-to-3D using 2D Diffusion [Poole et al., ICLR 2023] from UC Berkeley/Google





Course preview

- A modern machine learning perspective
- Widely-used learning algorithms
- Interactive content creation tools



Logistics

Course objectives

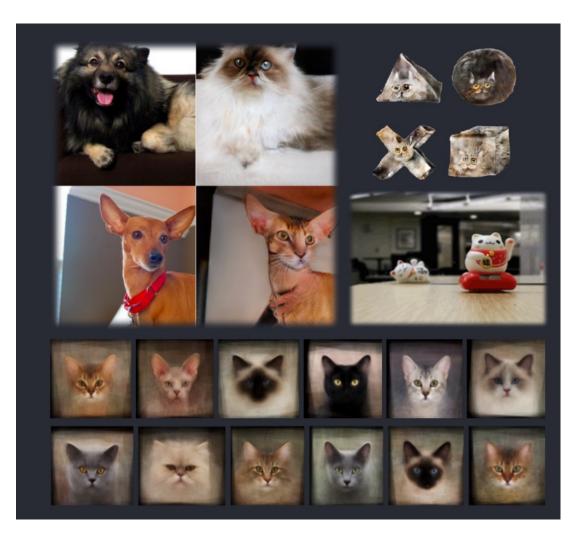
- 1. You will get a foundation in image editing and synthesis.
 - Texture synthesis and style transfer. Ο
 - Face modeling and synthesis. Ο
 - Image colorization and inpainting. Ο
 - Video generation and editing. Ο
 - Image-to-image translation. Ο
 - Image and video editing. (warping, morphing, compositing) Ο
 - Societal Implications: forensics, copyrights, biases. Ο

Course objectives

- 2. You will get a foundation of machine learning concepts
 - (fast) Nearest neighbor search. Ο
 - Principal component analysis, Gaussian Mixture model. Ο Markov Random Field (MRF)
 - Convolutional neural networks. \bigcirc
 - Deep generative models: Auto-encoder, Generative Ο Adversarial Networks, Flow-based models, Variational Auto-encoder, Autoregressive Models, Diffusion Models.
 - Conditional generative models.
 - Neural Radiance Fields (NeRF) Ο

Course objectives

3. You will have some cool results with your own photos



Textbook

TEXTS IN COMPUTER SCIENCE

Computer Vision

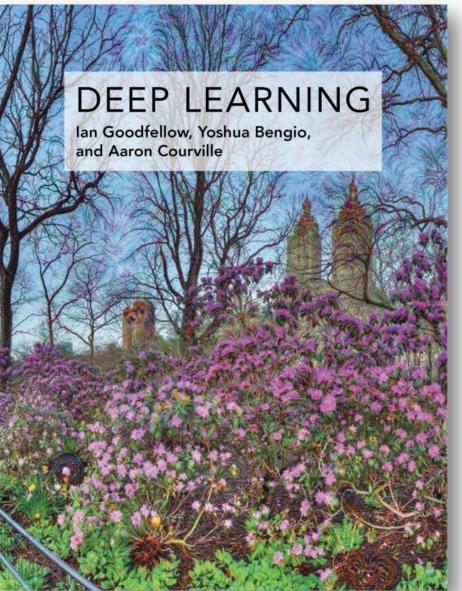
Algorithms and Applications



Richard Szeliski

Springer

https://szeliski.org/Book/



https://www.deeplearningbook.org/

Grading

- Emphasis on programming projects (60%).
 - Classic: 1. image alignment. 2. image blending
 - Deep learning: 3. neural style transfer. 4. GANs and conditional GANs.
 5. reconstructing and editing an image with GANs.
- Late Policy for programming assignments.
 - Five (5) emergency late days for semester, to be spent wisely
 - 10% of penalty per 24 hours afterwards
- Class attendance (5%)
- One paper presentation (10%):
 - 10-20 min, 1-2 people in a group.
 - Need to answer questions about this paper from now on.
- Final Project (25%)
 - A webpage-based report + a presentation.
 - No late day.
 - 2-3 people per group.

Assignments



Assignment #0 - How to submit assignments?



Assignment #1 - Colorizing the Prokudin-Gorskii Photo Collection 岌 🖺 Winner: [Konwoo Kim] Honorable Mentions: [Juyong Kim] [Zihang Lai] [Manuel Rodriguez]



Assignment #2 - Gradient Domain Fusion 👗 🏙 Winner: [Manuel Rodriguez] Honorable Mentions: [George Cazanavette]



Assignment #3 - When Cats meet GANs 🛛 🚺 Winner: [Jun Luo] Honorable Mentions: [George Cazanavette]



Assignment #4 - Neural Style Transfer 🔰 📓 🖺 Winner: [Zihang Lai] Honorable Mentions: [Zijie Li] [Tarang Shah]



Assignment #5 - GAN Photo Editing Winner: [George Cazenavette] Honorable Mentions: [Manuel Guevara] [Zijie Li] [Zhe Huang]



+ more Diffusion Model modules

For each assignment

- Derive the math, implement stuff from scratch (+ starter code), and apply it to your own photos
- Every person does their own project (except final)
- Reporting via web page (+ submit code to Canvas)
- Afterwards, vote for class **favorite**(s)! **Gift**!
- Programming Language:
 - Python and PyTorch
 - you can use other languages, but you are on your own

Academic Integrity

- Can discuss projects, but don't share code
- Don't look up code or copy from a friend
- If you're not sure if it's allowed, ask
- Acknowledge any inspirations
- If you get stuck, come talk to us

Getting help outside of class

- Course Web Page
 - <u>https://16726-image-synthesis.github.io/sp24/</u>
- Discussion board:
 - Piazza.com
- Assignment submission
 Canvas
 - Canvas
- Office hours
 - See piazza posts



16-726 Learning-Based Image Synthesis / Spring 2024

Time: Mondays, Wednesdays 9:30 am - 10:50 am ET

Location: NSH 3002



source

Course Description

This course introduces machine learning methods for image and video synthesis. The objectives of synthesis research vary from modeling statistical distributions of visual data, through realistic picture-perfect recreations of the world in graphics, and all the way to providing interactive tools for artistic expression. Key machine learning algorithms will be presented, ranging from classical learning methods (e.g., nearest neighbor, PCA, Markov Random Fields) to deep learning models (e.g., ConvNets, deep generative models, such as GANs, VAEs and Diffusion models). We will also introduce image and video forensics methods for detecting synthetic content. In this class, students will learn to build practical applications and create new visual effects using their own photos and videos.

Why you should NOT take this class

- Project-based class
 - No canned problem sets.
 - Not theory-heavy.
 - will read many research papers.
 - Open-ended by design.
- Need time to think, not just hack
 Creativity is a class requirement.
- Not worth it if you don't enjoy it.

Now... reasons TO take this class

- Not too many similar courses at other places.
- You get to create pictures and unleash your creative potential.
- Interested in grad school and research? ③
- Interested in industry jobs? ③



Become a friend with every pixel!

- Jitendra Malik, UC Berkeley

Thank You!



16-726, Spring 2024 https://learning-image-synthesis.github.io/sp24/

