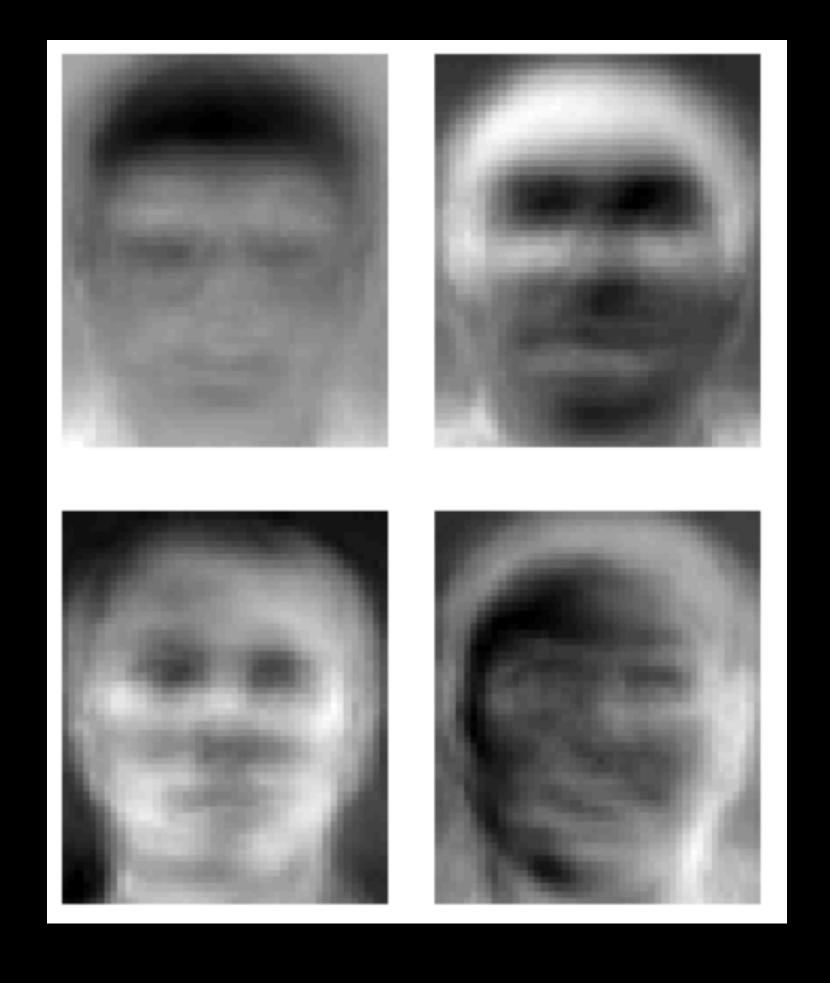


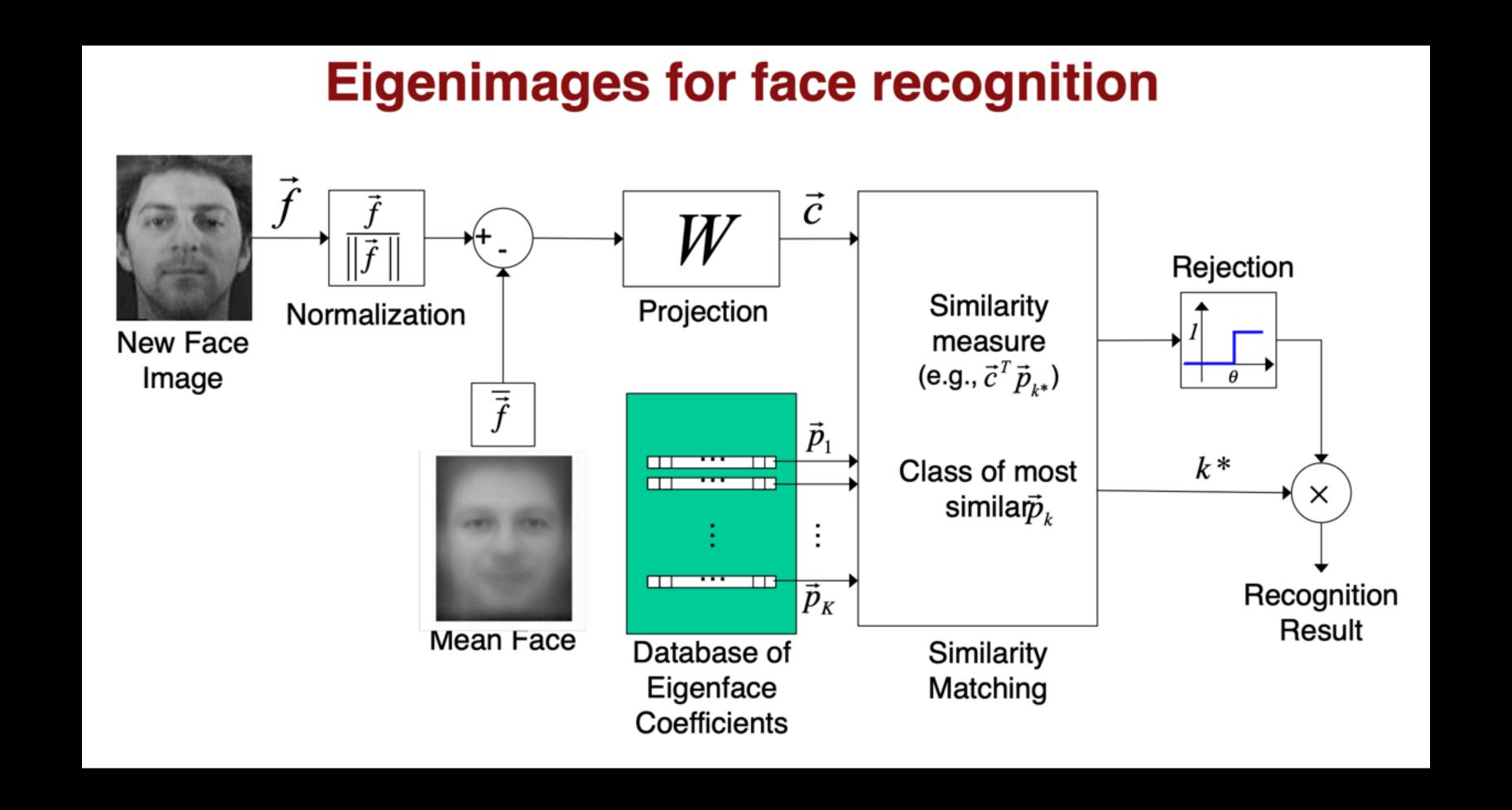
Parag K. Mital UCLA DMA

CULTURAL CONTEXT + HISTORY OF GENERATIVE IMAGES

EIGENIMAGES (1980 - 2010)



https://en.wikipedia.org/wiki/Eigenface



https://web.stanford.edu/class/ee368/Handouts/Lectures/2019 Winter/10-EigenImages.pdf

The first 8 eigenfaces obtained from a training set of 100 male and 100 female

training images



Mean Face



Eigenface 1

Eigenface 5



Eigenface 2



Eigenface 3



Eigenface 4









Can be used to generate faces by adjusting 8 coefficients.

Can be used for face recognition by nearest-neighbor search in 8-d "face space.

IMAGE ANALOGIES

Image Analogies

Aaron Hertzmann^{1,2} Charles E. Jacobs² Nuria Oliver² Brian Curless³ David H. Salesin^{2,3}

¹New York University ²Microsoft Research ³University of Washington

Figure 1 An image analogy. Our problem is to compute a new "analogous" image B' that relates to B in "the same way" as A' relates to A. Here, A, A', and B are inputs to our algorithm, and B' is the output. The full-size images are shown in Figures 10 and 11.

Abstract

This paper describes a new framework for processing images by example, called "image analogies." The framework involves two stages: a design phase, in which a pair of images, with one image purported to be a "filtered" version of the other, is presented as "training data"; and an application phase, in which the learned filter is applied to some new target image in order to create an "analogous" filtered result. Image analogies are based on a simple multiscale autoregression, inspired primarily by recent results in texture synthesis. By choosing different types of source image pairs as input, the framework supports a wide variety of "image filter" effects, including traditional image filters, such as blurring or embossing; improved texture synthesis, in which some textures are synthesized with higher quality than by previous approaches; super-resolution, in which a higher-resolution image is inferred from a low-resolution source; texture transfer, in which images are "texturized" with some arbitrary source texture: artistic filters, in which various drawing

Introduction

a·nal·o·gy n. A systematic comparison between structures that uses properties of and relations between objects of a source structure to infer properties of and relations between objects of a target structure. [14]

A native talent for perceiving analogies is . . . the leading fact in genius of every order.

-William James, 1890 [28]

Analogy is a basic reasoning process, one that we as humans employ quite commonly, and often unconsciously, to solve problems, provide explanations, and make predictions [44]. In this paper, we explore the use of analogy as a means for creating complex image filters (Figure 1). In particular, we attempt to solve the following problem:



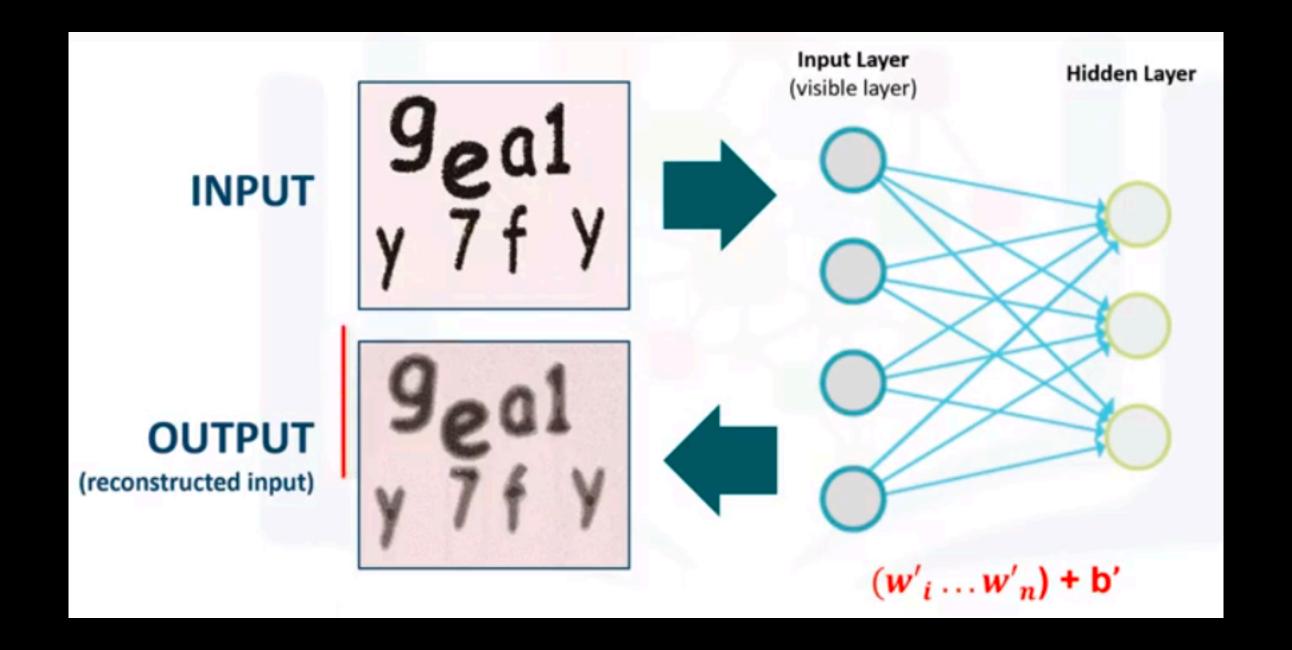
https://pkmital.com/home/works/early-visual-synthesis-works/

ACTIVE APPEARANCE MODELS (1998-2010)



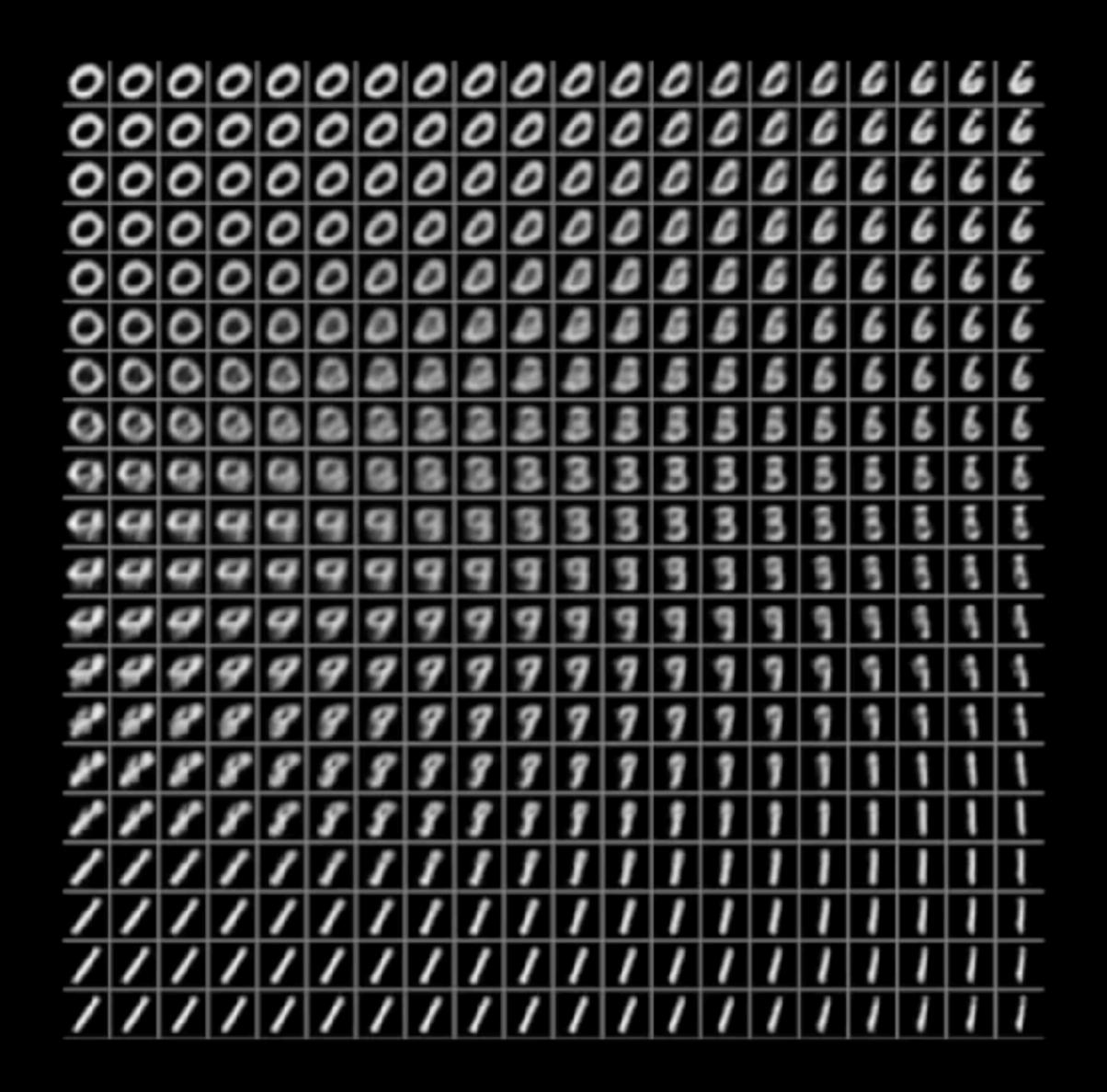
https://vimeo.com/24409802

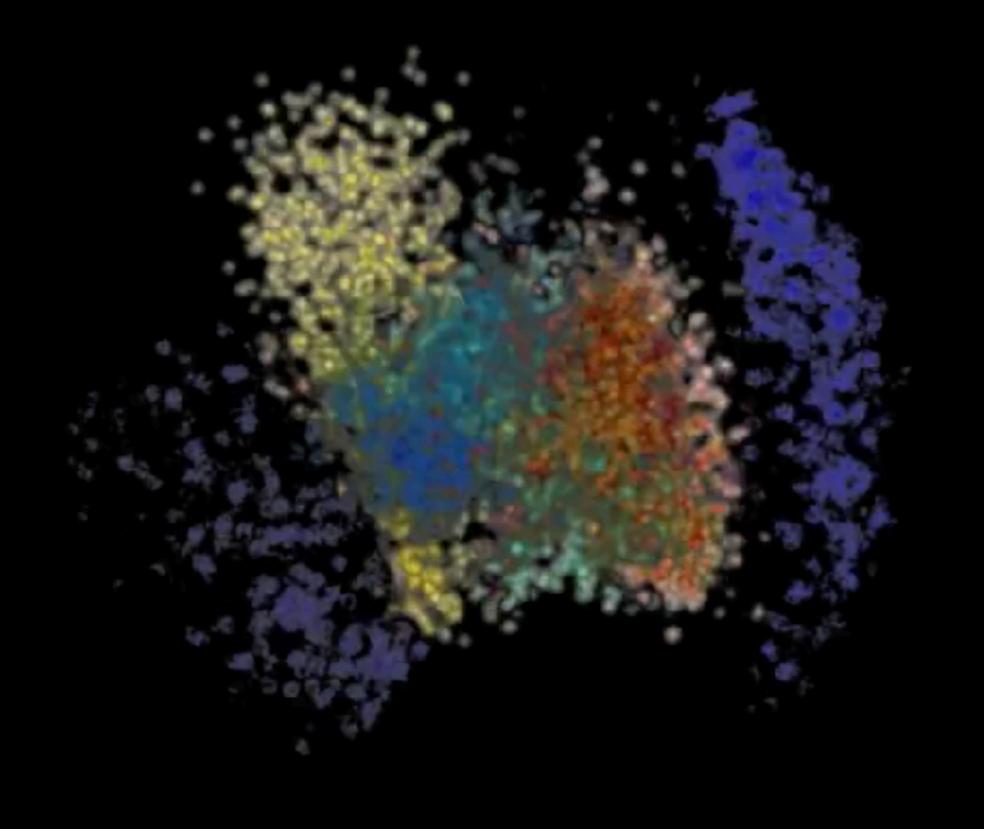
RBM (1986, 2006-2013)



https://medium.com/datadriveninvestor/dimensionality-reduction-and-feature-extraction-with-rbm-f499965979e9

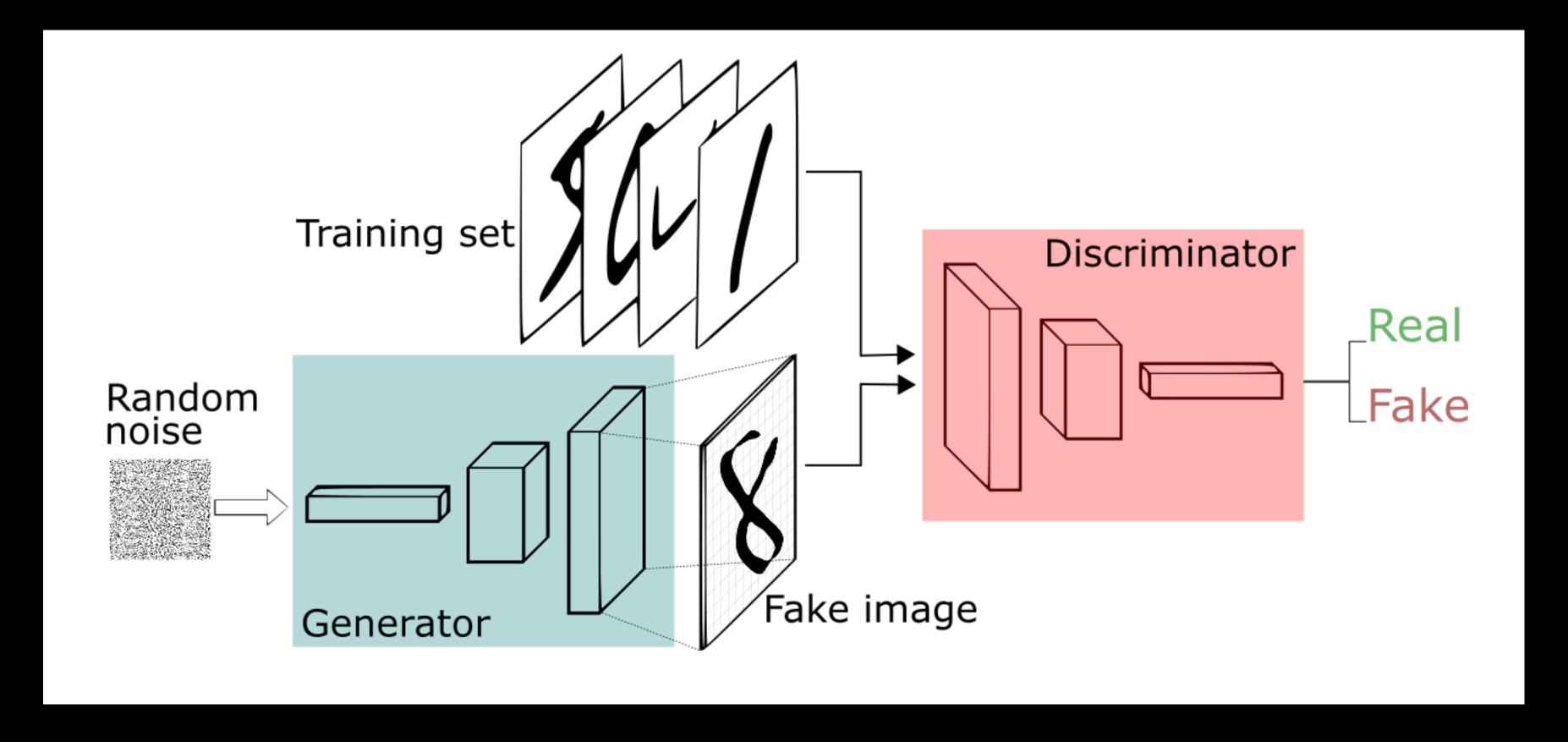
(V)AE (2010 - 2020)



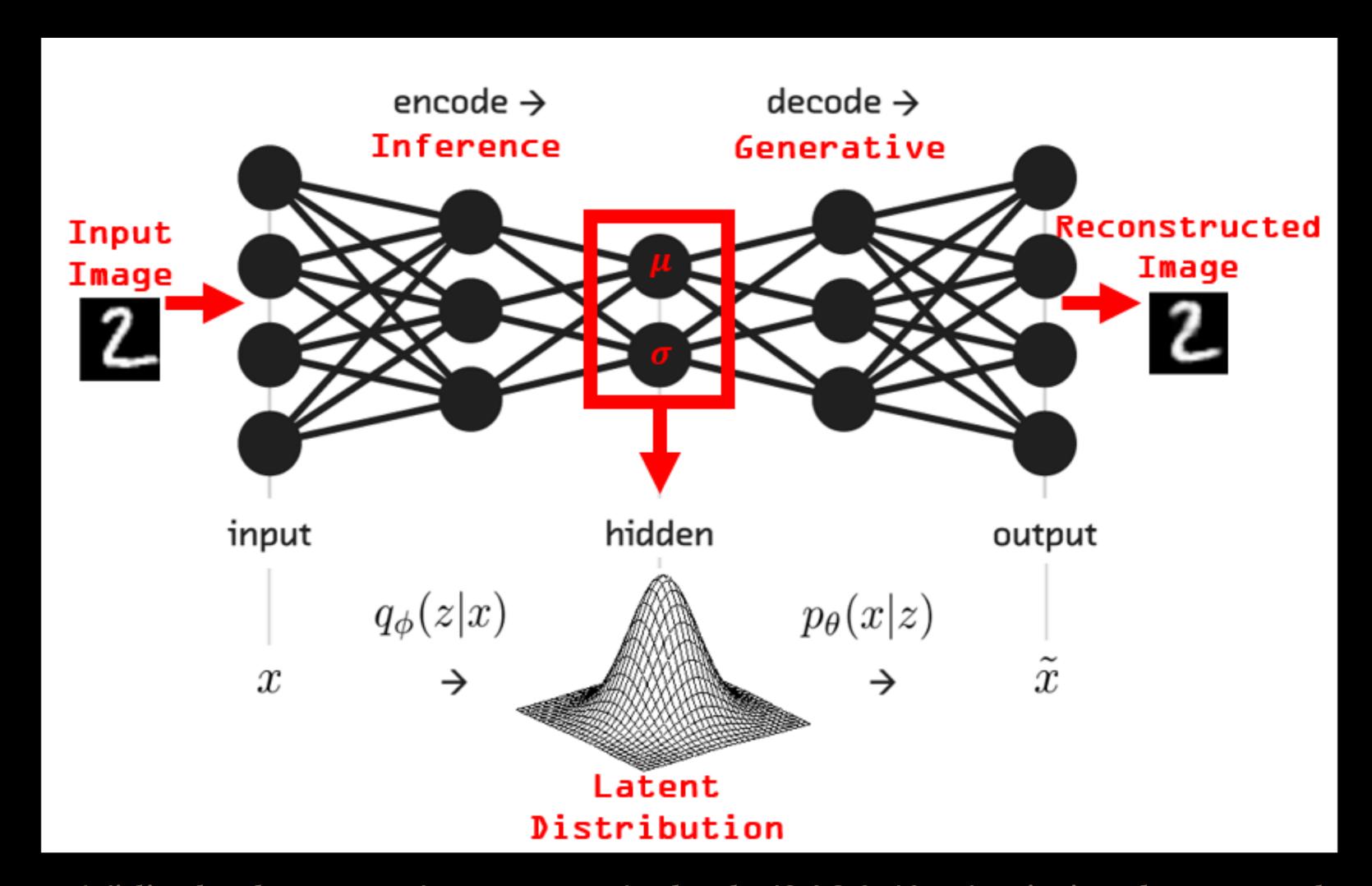


https://vimeo.com/155061675

GANS (2014 - CURRENT)



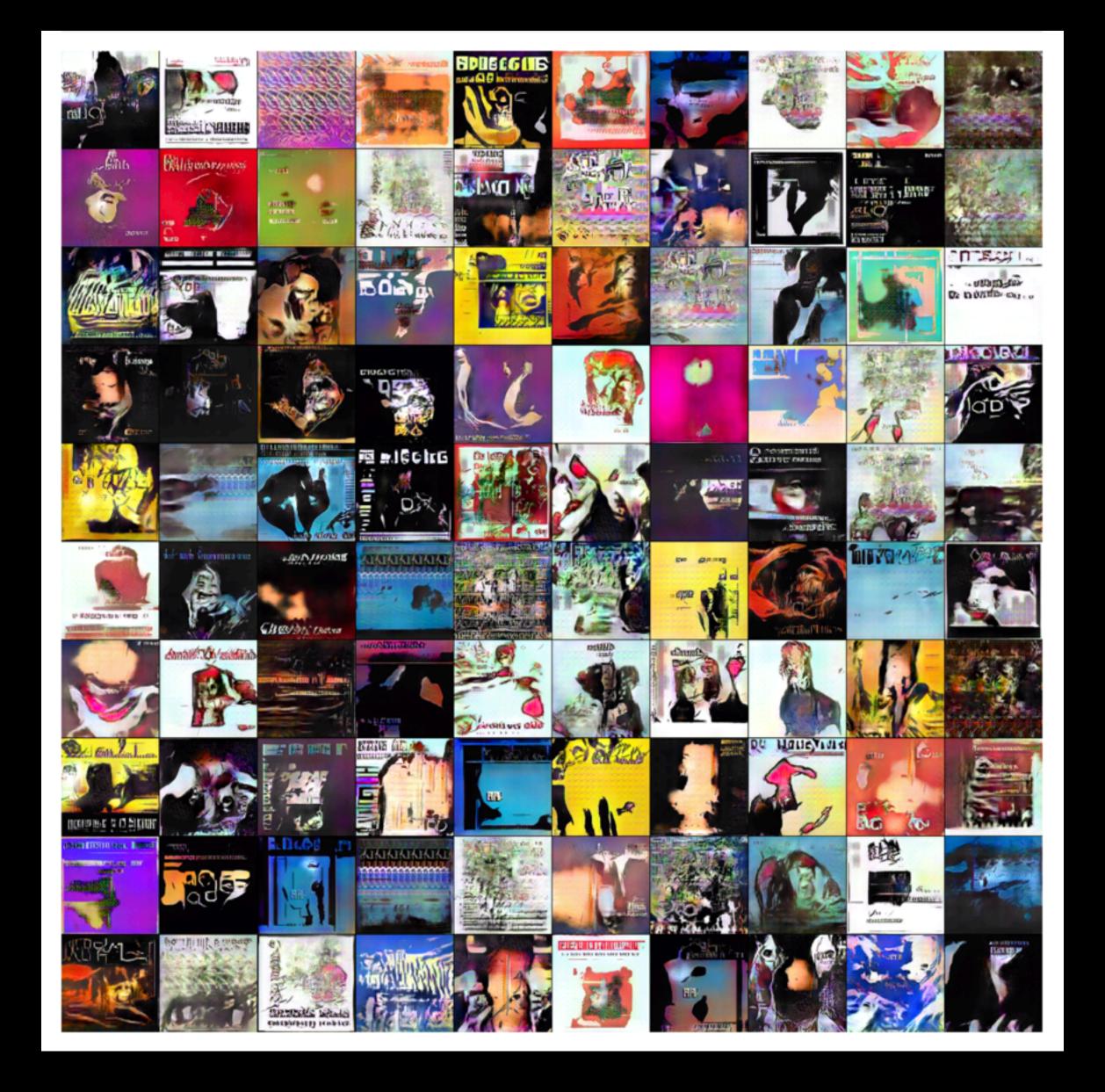
https://www.freecodecamp.org/news/an-intuitive-introduction-to-generative-adversarial-networks-gans-7a2264a81394/



https://idiotdeveloper.com/wp-content/uploads/2020/05/variational-autoencoder.png



https://github.com/Newmu/dcgan_code



https://github.com/Newmu/dcgan_code

Arithmetic on faces smiling neutral neutral smiling man woman woman man

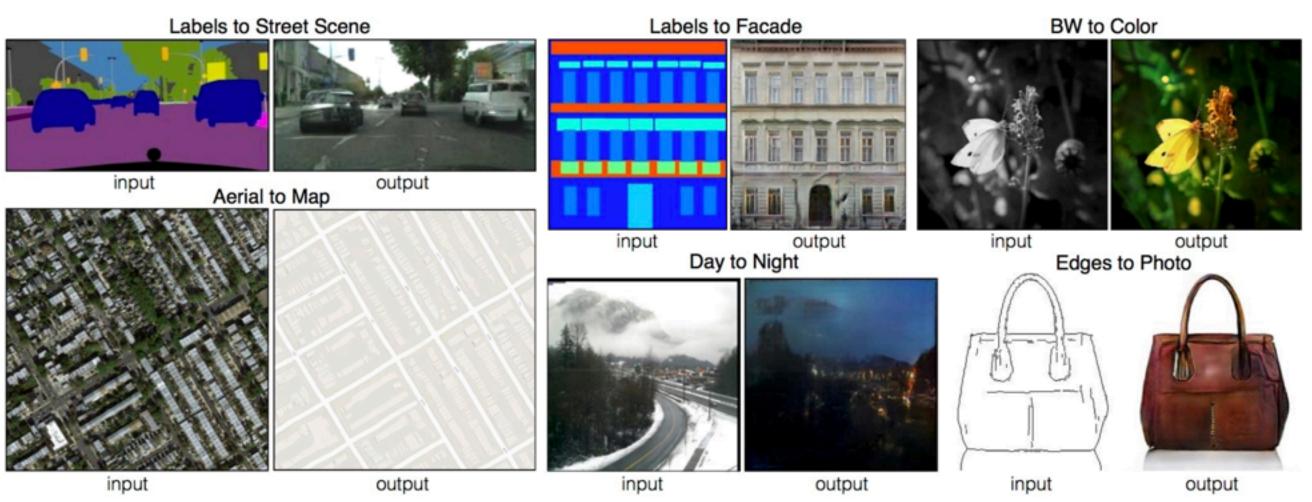
https://github.com/Newmu/dcgan_code

Image-to-Image Translation with Conditional Adversarial Nets

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

University of California, Berkeley In CVPR 2017

[Paper] [GitHub]



Example results on several image-to-image translation problems. In each case we use the same architecture and objective, simply training on different data.

Abstract

We investigate conditional adversarial networks as a general-purpose solution to image-to-image translation problems. These networks not only learn the mapping from input image to output image, but also learn a loss function to train this mapping. This makes it possible to apply the same generic approach to problems that traditionally would require very different loss formulations. We demonstrate that this approach is effective at synthesizing photos from label maps, reconstructing objects from edge maps, and colorizing images, among other tasks. As a community, we no longer hand-engineer our mapping functions, and this work suggests we can achieve reasonable results without hand-engineering our loss functions either.

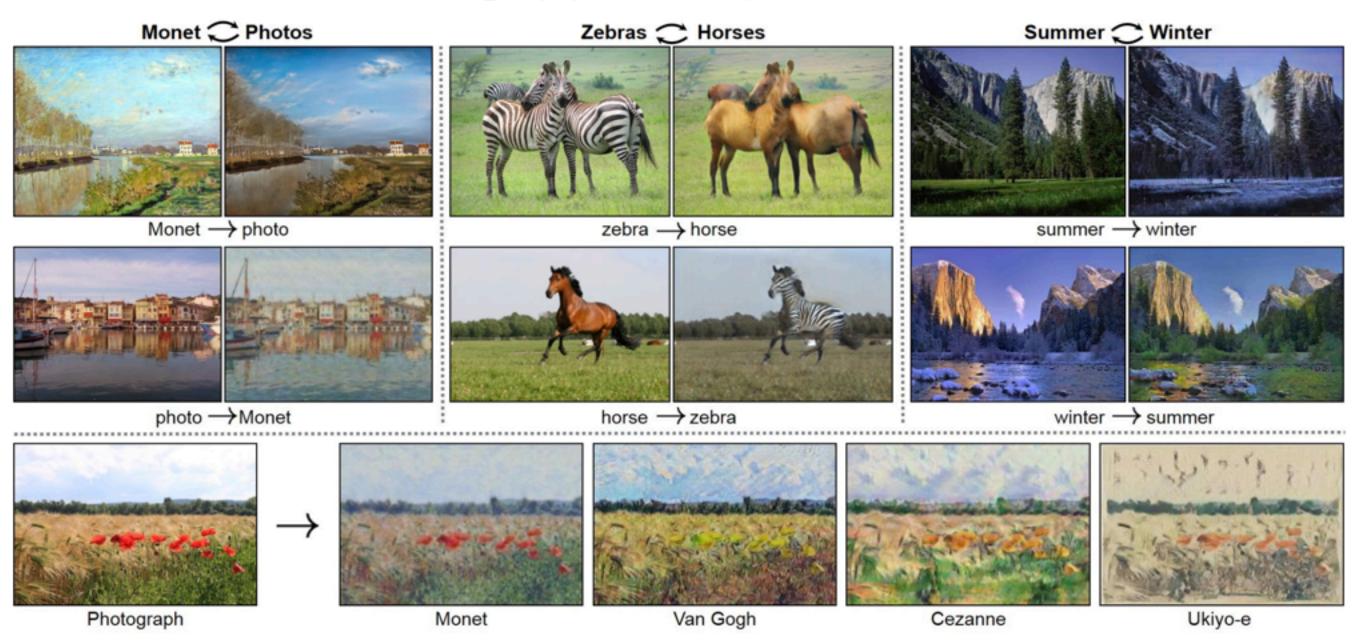
Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks

Jun-Yan Zhu* Taesung Park* Phillip Isola Alexei A. Efros

UC Berkeley

In ICCV 2017

Paper | PyTorch code | Torch code



Abstract

Image-to-image translation is a class of vision and graphics problems where the goal is to learn the mapping between an input image and an output image using a training set of aligned image pairs. However, for many tasks, paired training data will not be available. We present an approach for learning to translate an image from a source domain X to a target domain Y in the absence of paired examples. Our goal is to learn a mapping $G: X \rightarrow Y$, such that the distribution of images from G(X) is indistinguishable from the distribution Y using an adversarial loss. Because this mapping is highly under-constrained, we couple it with an inverse mapping $F: Y \rightarrow X$ and introduce a cycle consistency loss to push $F(G(X)) \approx X$ (and vice versa). Qualitative results are presented on several tasks where paired training data does not exist, including collection style transfer, object transfiguration, season transfer, photo enhancement, etc. Quantitative comparisons against several prior methods demonstrate the superiority of our approach.

Flickr-Faces-HQ Dataset (FFHQ)

python 3.6 license CC format PNG resolution 1024×1024 images 70,000



Flickr-Faces-HQ (FFHQ) is a high-quality image dataset of human faces, originally created as a benchmark for generative adversarial networks (GAN):

A Style-Based Generator Architecture for Generative Adversarial Networks
Tero Karras (NVIDIA), Samuli Laine (NVIDIA), Timo Aila (NVIDIA)
https://arxiv.org/abs/1812.04948

The dataset consists of 70,000 high-quality PNG images at 1024×1024 resolution and contains considerable variation in terms of age, ethnicity and image background. It also has good coverage of accessories such as eyeglasses, sunglasses, hats, etc. The images were crawled from Flickr, thus inheriting all the biases of that website, and automatically aligned and cropped using dlib. Only images under permissive licenses were collected. Various automatic filters were used to prune the set, and finally Amazon Mechanical Turk was used to remove the occasional statues, paintings, or photos of photos.

For business inquiries, please contact researchinquiries@nvidia.com

For press and other inquiries, please contact Hector Marinez at hmarinez@nvidia.com

StyleGAN — Official TensorFlow Implementation

python 3.6 tensorflow 1.10 cudnn 7.3.1 license CC BY-NC

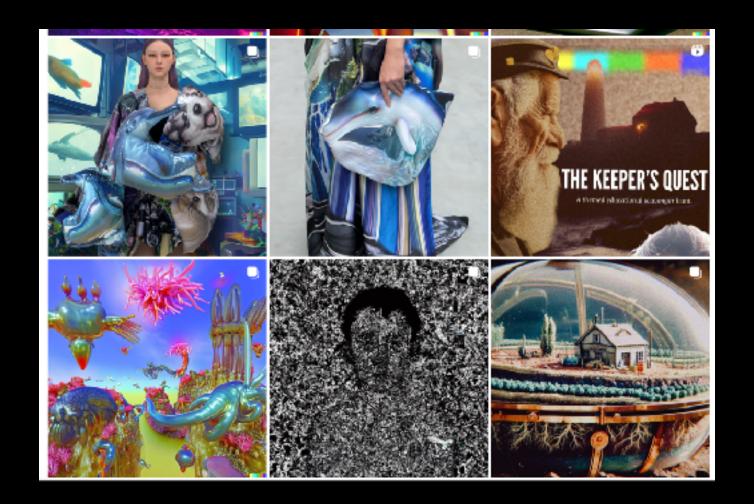


Picture: These people are not real – they were produced by our generator that allows control over different aspects of the image.

This repository contains the official TensorFlow implementation of the following paper:

A Style-Based Generator Architecture for Generative Adversarial Networks
Tero Karras (NVIDIA), Samuli Laine (NVIDIA), Timo Aila (NVIDIA)
https://arxiv.org/abs/1812.04948

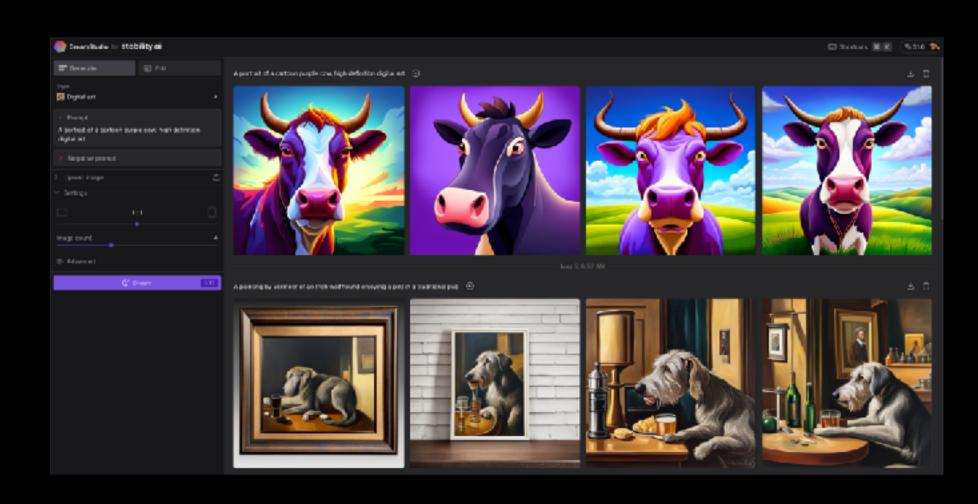
DIFFUSION MODELS



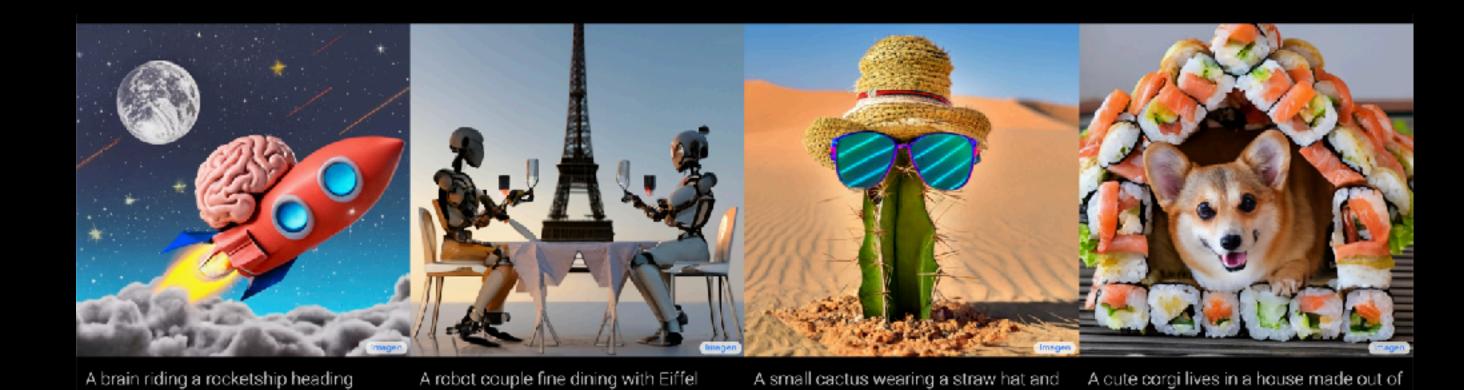
Dall-E2

https://openai.com/dall-e-2

https://www.instagram.com/openaidalle/



Stable Diffusion
https://stablediffusionweb.com/

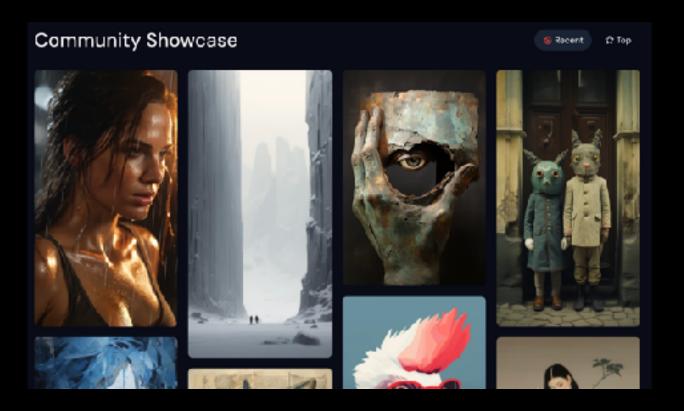


neon sunglasses in the Sahara desert. sushi.

Imagen (by Google)
https://imagen.research.google/

Tower in the background.

towards the moon.



Midjourney

https://www.midjourney.com/home/?

callbackUrl=%2Fapp%2F



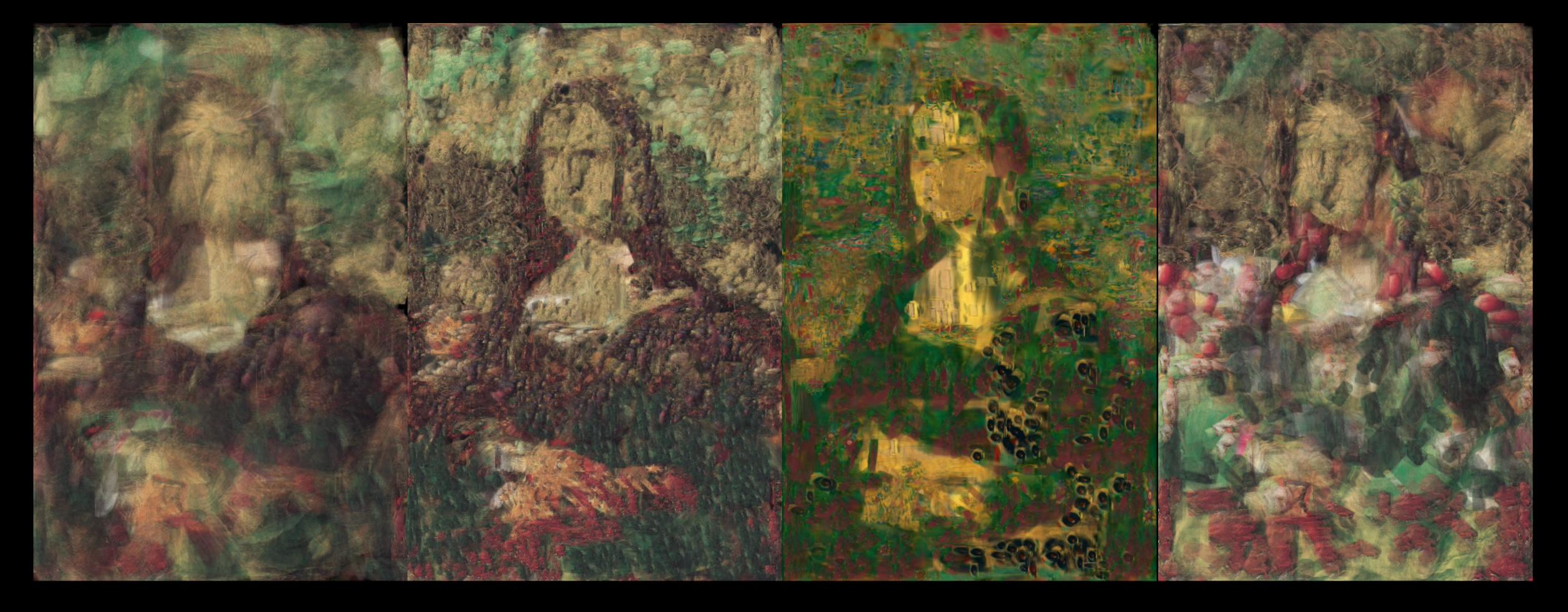
https://analyticsindiamag.com/ youll-never-have-to-see-anotherboring-qr-code-again/



https://huggingface.co/spaces/AP123/IllusionDiffusion

ARTISTS EXPLORING GANS

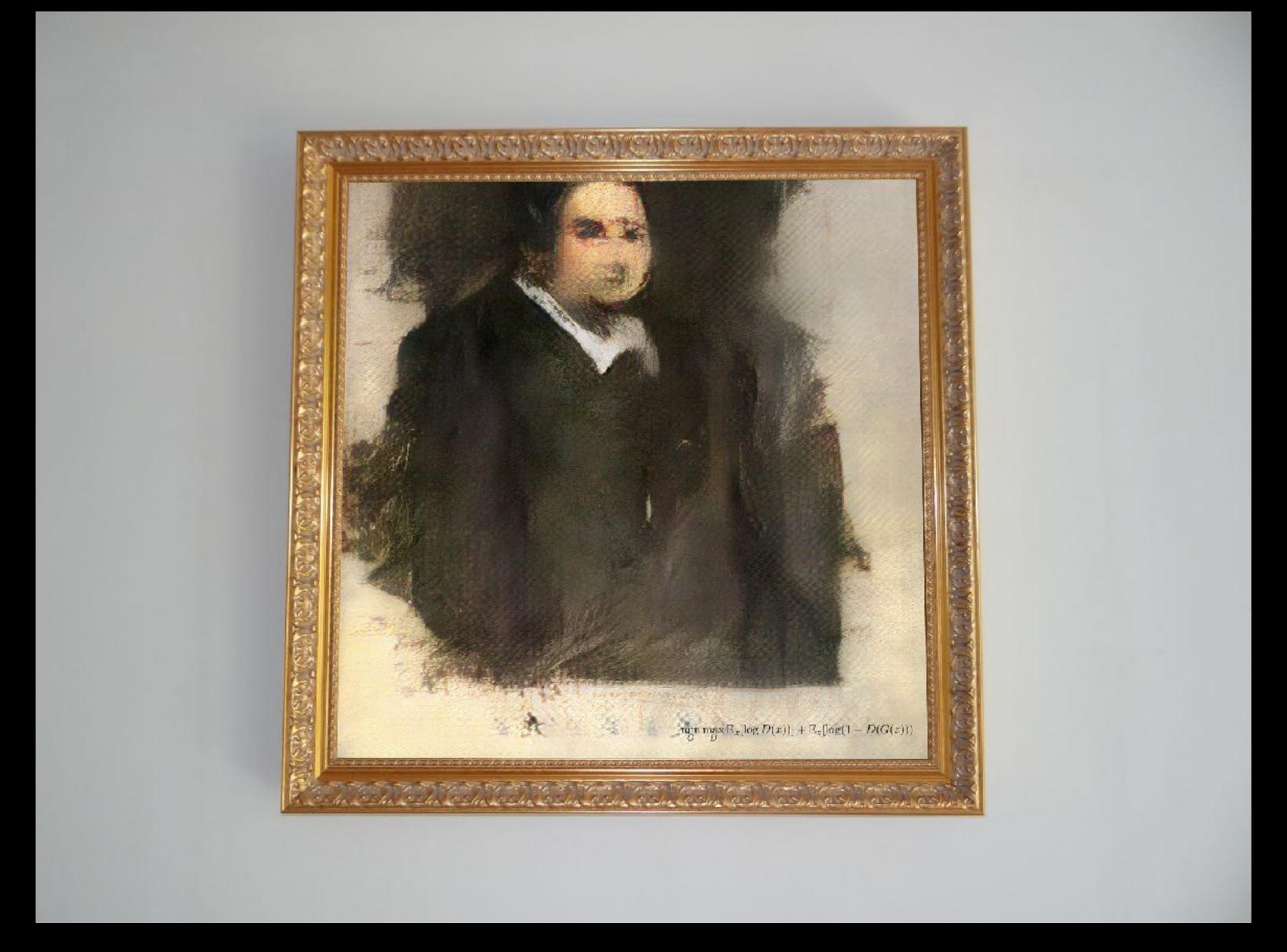
https://mlart.co/



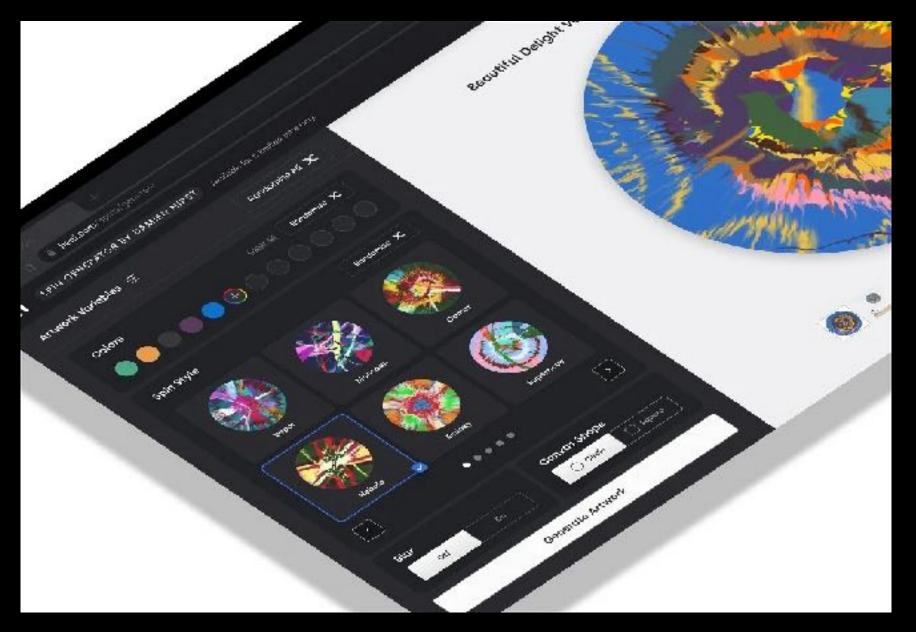
https://pkmital.com/home/works/early-visual-synthesis-works/



https://underdestruction.com/2018/12/29/memories-of-passersby-i/



"Edmond de Bellamy", 2018





https://heni.com/spins



https://spruethmagers.com/artists/jon-rafman/



https://www.jakeelwes.com/project-zizimotion.html



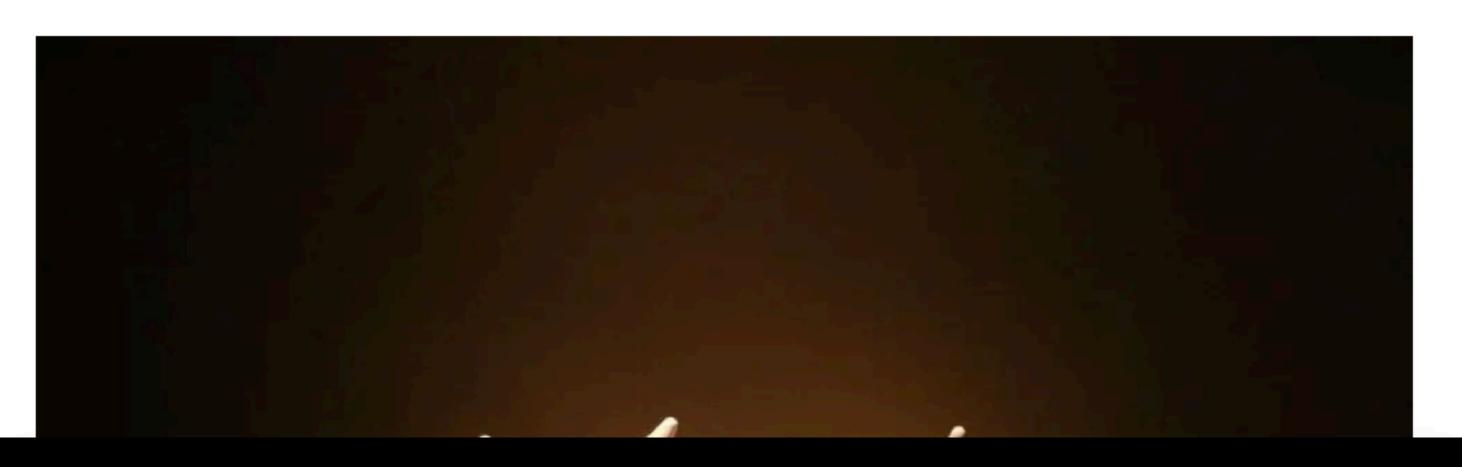
https://vimeo.com/689052977





https://breannabrowning.com/

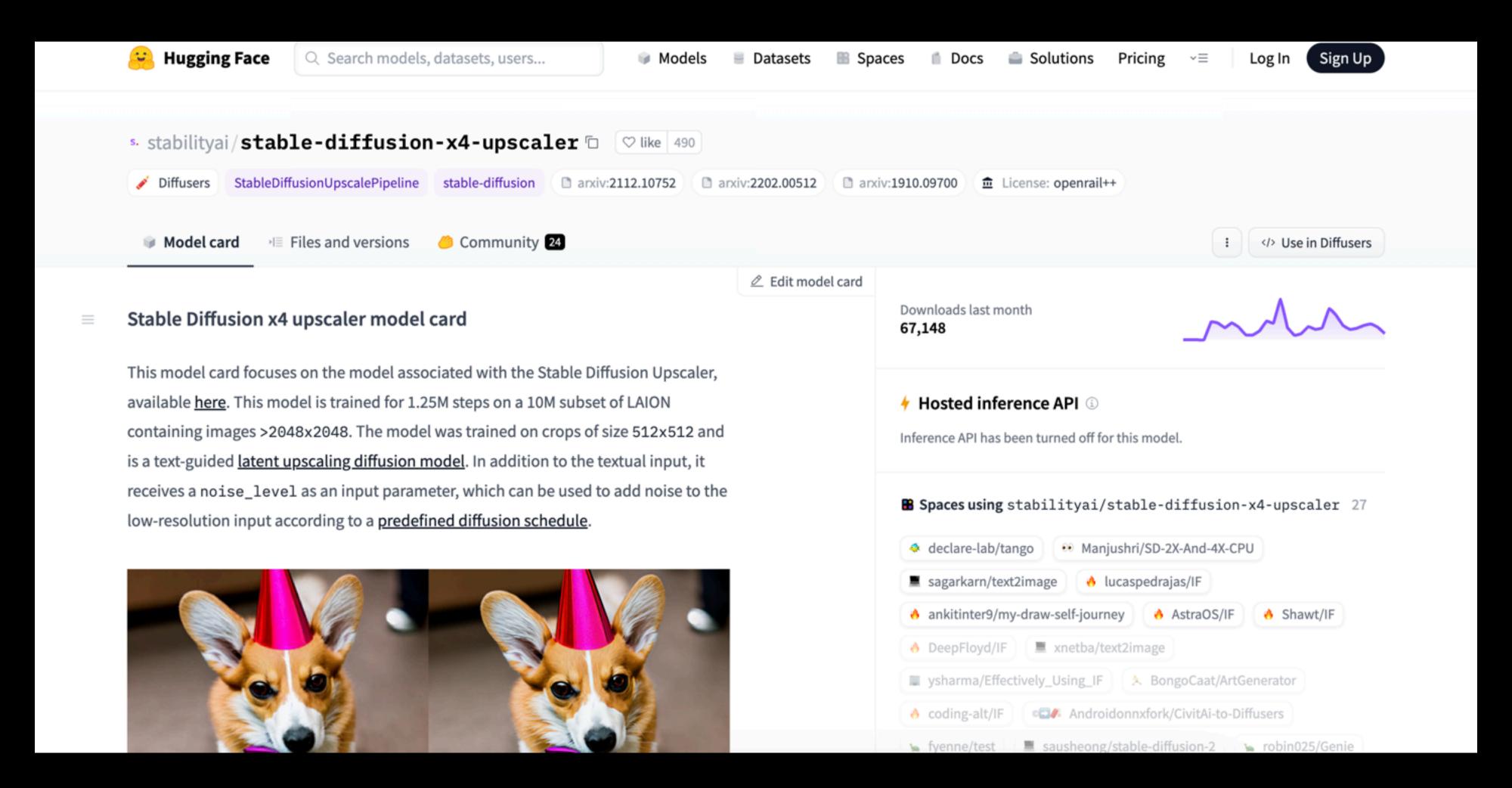
We are documenting the first decades of artists who create work with software and for the internet. From 16mm films made with code to CD-ROMs to websites, from the 1960s to the present, artists have explored and have defined a new medium for the visual arts. We are collecting information to be shared with artists, curators, and scholars, and we hope this resource will be useful in the present and far into the future.



https://mirrorarchive.net/about

LAB

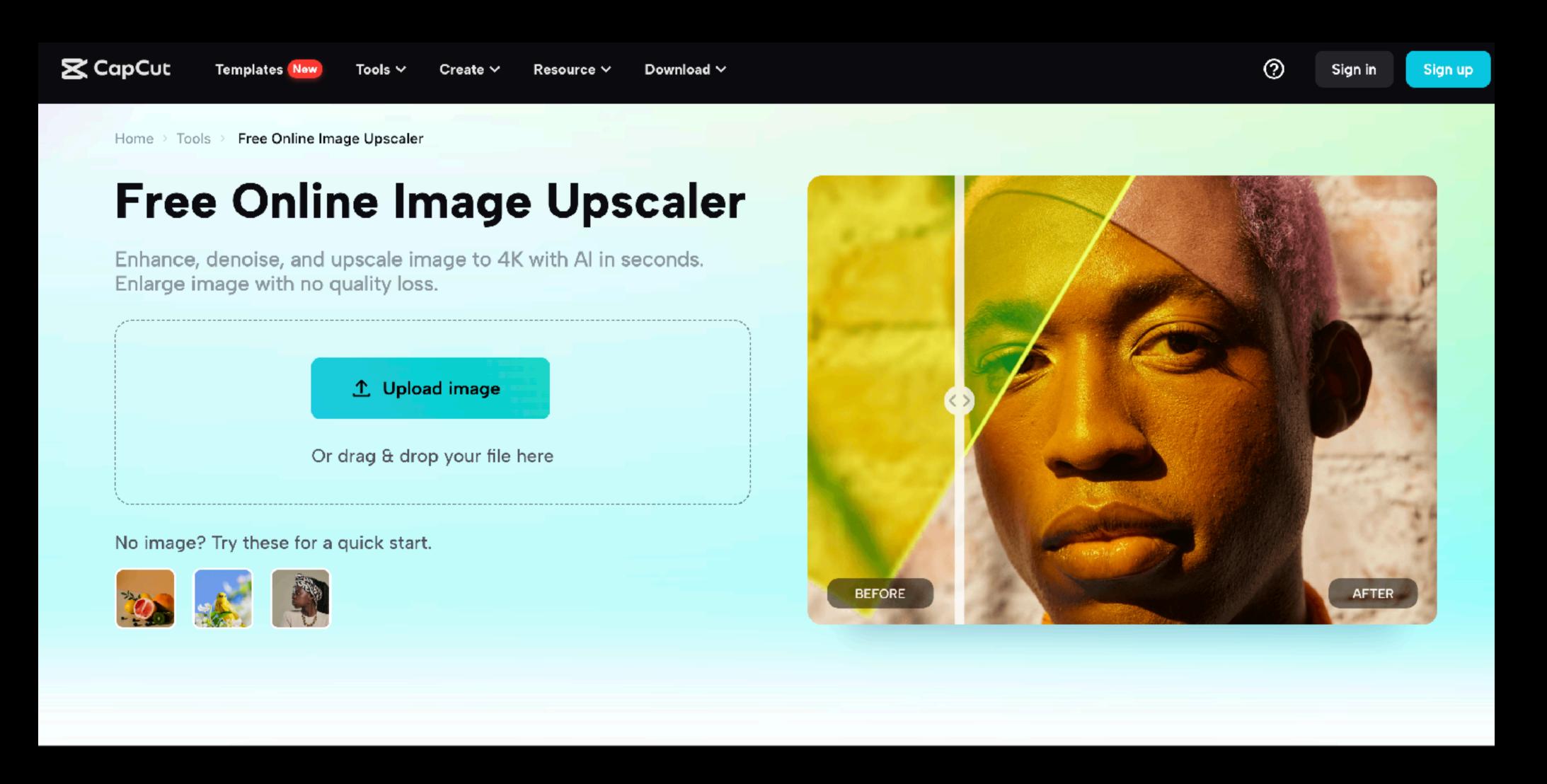
UPRES AI



https://huggingface.co/stabilityai/stable-diffusion-x4-upscaler

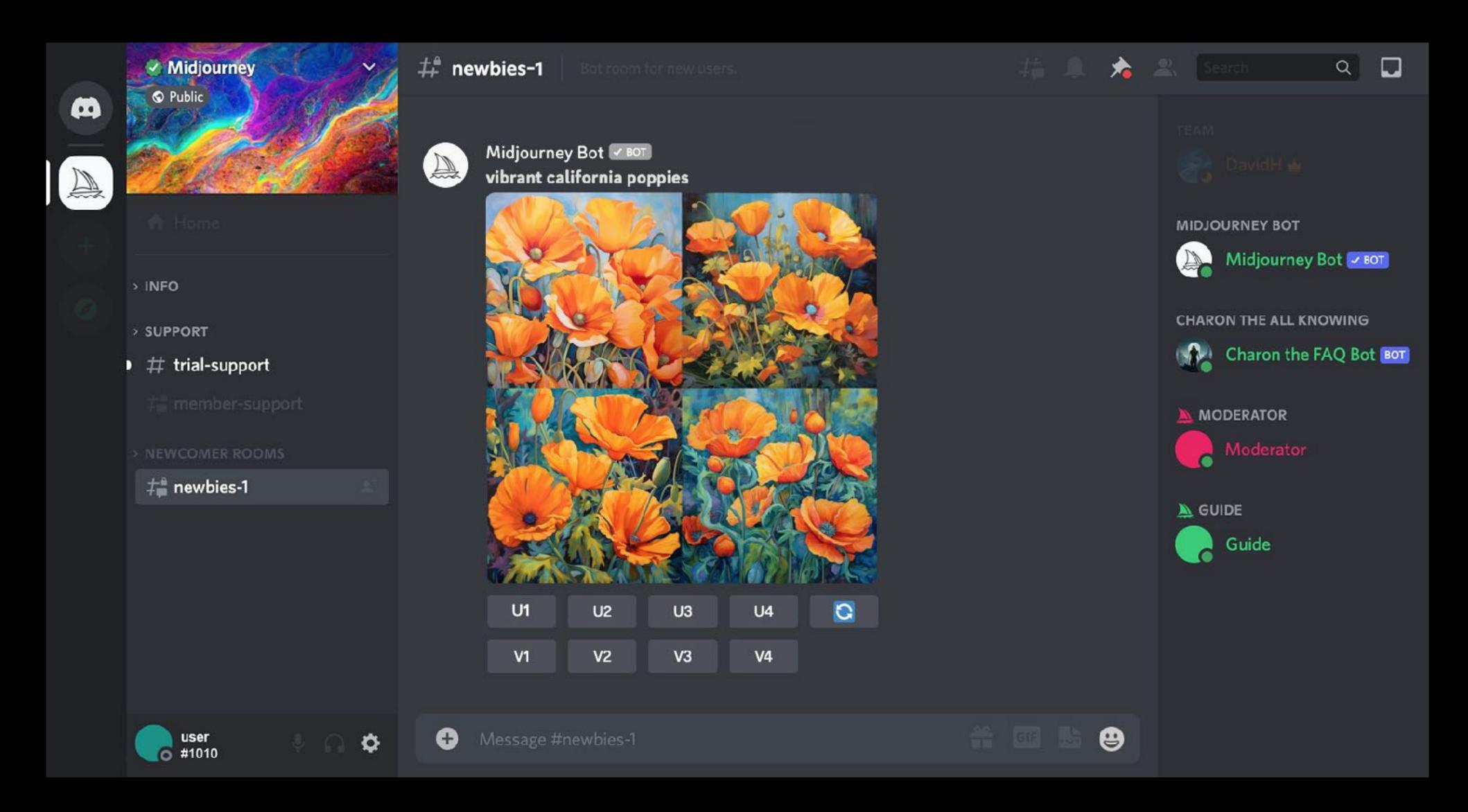


https://www.topazlabs.com/

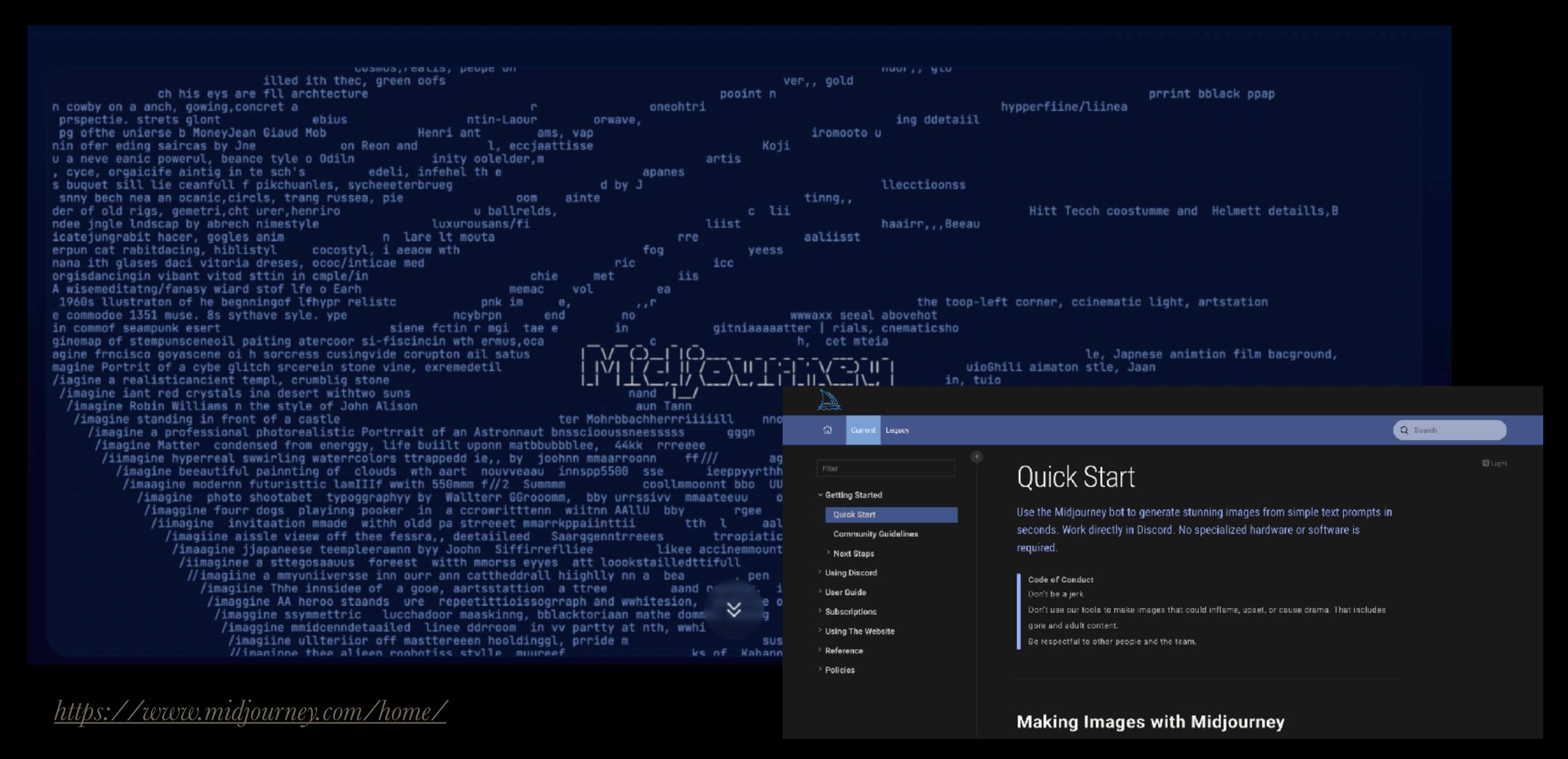


https://www.capcut.com/tools/ai-image-upscaler

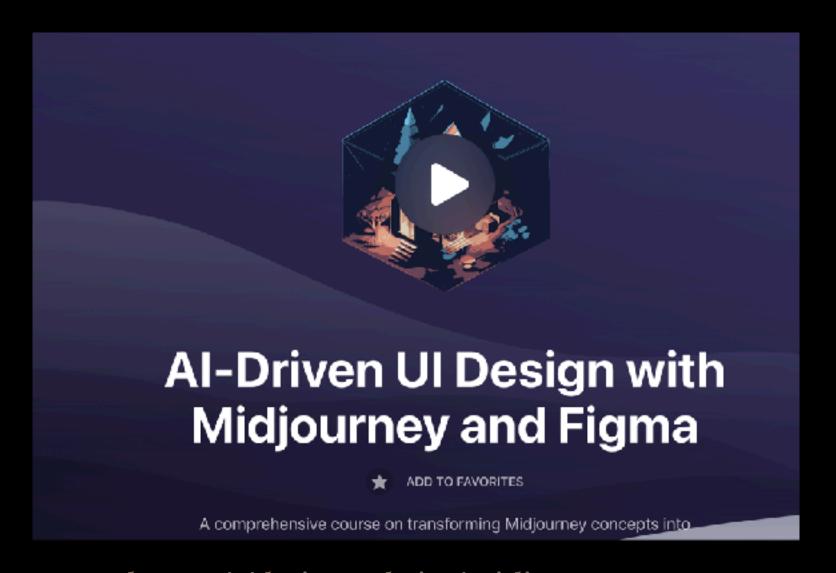
TEXT TO IMAGE IMAGE TO IMAGE



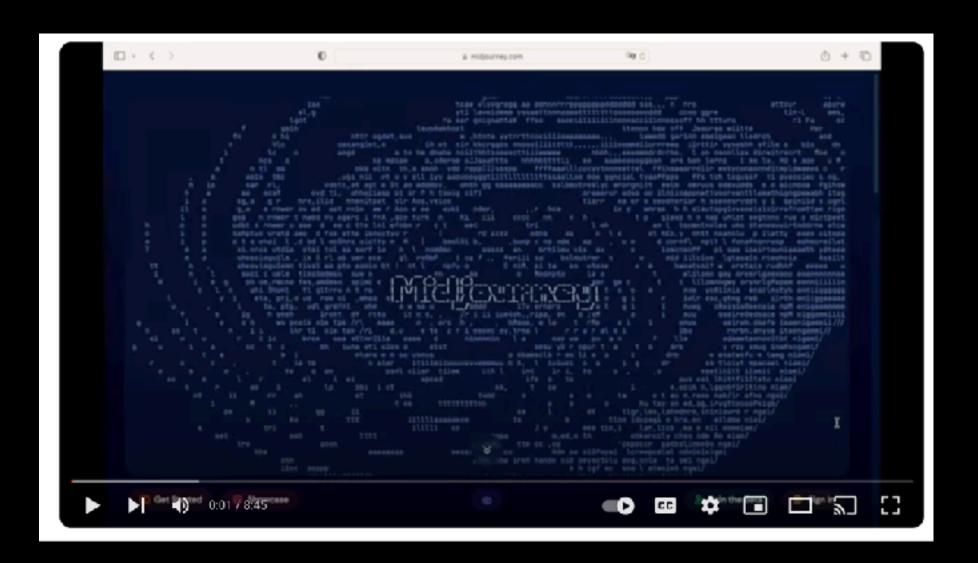
https://www.midjourney.com/home/



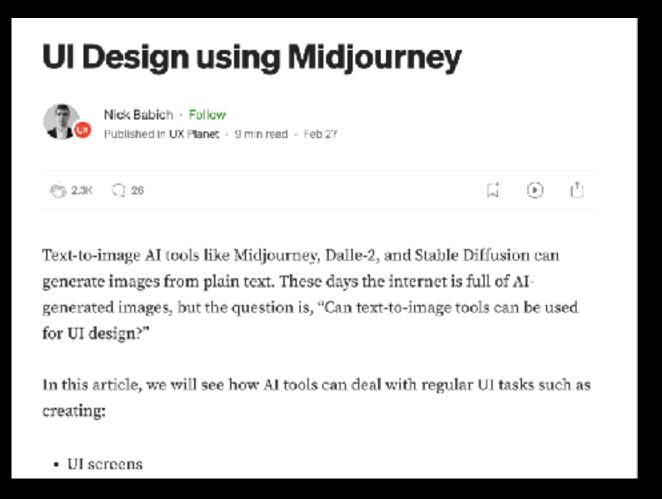
https://docs.midjourney.com/docs/quick-start



https://designcode.io/midjourney-apps



https://www.youtube.com/watch?v=0-4eVgEB26M



https://uxplanet.org/ui-design-using-midjourney-713e8d1b6a6b

- Providing keywords 'style'
- stylize
- chaos
- Resolution
- Aspect ratio
- Passing an image as a prompt as
 URL
- Applying weights to the image prompts
- Weights to the word prompts
- Filtering out words

https://medium.com/mlearning-ai/an-advanced-guideto-writing-prompts-for-midjourney-text-to-imageaa12a1e33b6



10 Cinematic Midjourney Prompts VOL2.

Bookmark for future use. 🎤

Here is a list of cinematic Midjourney prompts. Feel free to copy & paste, and change or remove parts of the topic as needed.

Happy testing!

Prompt: Cinematic Crane Shot, The sun setting over the scenic coastline of Malibu, a luxury convertible cruising along the Pacific Coast Highway, capturing the carefree and glamourous LA lifestyle, as if through the lens of a vintage Panavision camera −-style raw −-ar 21:9 −-c 10

Prompt: A ransom drop-off in dirty LA backstreets, gritty surroundings, neon lights clash with the stark, moonless night, mimicking the suspense of Tarantino's plots. Captured on lustrous Kodak Vision3 Color Negative Film 500T 5219 --style raw --ar 21:9

Prompt: Ultra Panavision 70, Action scene in 1970 Los Angeles, muscle car chase, at day time, shot on warm-toned Eastman Kodak film

https://twitter.com/bubblez_jazzy/status/ 1698255033991565655? s=46&t=MHjSXuuPGHmWLD4It3ZkFA



Seems like the X algorithm was weird lately, and few people saw my previous post.

It's better now, so I'm updating the post and sharing it again.

Previously I showed you guys how to make vector illustrations w Midjourney.

Prompt:

png white background, [subject], in the style of animated illustra [environment], full body, text-based --style raw --stylize 100

Here's a neat trick to control the detail level of the images:

Change the "stylize" value at the end of the prompt.

The default value is 100.

If you prefer flatter images with less detail, use a low value like 0

If you like more artistic images, use a high value like 500.

It's an easy way to customize the images for your project or presentation.





--stylize 250 --stylize 500



https://twitter.com/chaseleantj/status/
1697946877772394775?
s=46&t=MHjSXuuPGHmWLD4It3ZkFA

/Imagine png white background, a little family of kittens, in the style of animated illustrations, in a forest, full body, text-based --style raw --stylize (x)

stylize 0



stylize 100



stylize 500



/imagine Cinematic Crane Shot, The sun setting over the scenic coastline of Malibu, a group of 3 motorcycles driving down the Pacific Coast Highway, as if through the lens of a vintage Panavision camera --style raw --ar 21:9 --c 10



/imagine Close Up shot, cinematic scene of a trendy female profile, at an old New York Hotel, as if shot on vintage Agfa Vista 200 film. --style raw --ar 21:9 --c 10



High-Resolution Image Synthesis with Latent Diffusion Models

Robin Rombach, Andreas Blattmann, Dominik Lorenz, Patrick Esser, Björn Ommer

By decomposing the image formation process into a sequential application of denoising autoencoders, diffusion models (DMs) achieve state-of-the-art synthesis results on image data and beyond. Additionally, their formulation allows for a guiding mechanism to control the image generation process without retraining. However, since these models typically operate directly in pixel space, optimization of powerful DMs often consumes hundreds of GPU days and inference is expensive due to sequential evaluations. To enable DM training on limited computational resources while retaining their quality and flexibility, we apply them in the latent space of powerful pretrained autoencoders. In contrast to previous work, training diffusion models on such a representation allows for the first time to reach a near-optimal point between complexity reduction and detail preservation, greatly boosting visual fidelity. By introducing cross-attention layers into the model architecture, we turn diffusion models into powerful and flexible generators for general conditioning inputs such as text or bounding boxes and high-resolution synthesis becomes possible in a convolutional manner. Our latent diffusion models (LDMs) achieve a new state of the art for image inpainting and highly competitive performance on various tasks, including unconditional image generation, semantic scene synthesis, and super-resolution, while significantly reducing computational requirements compared to pixel-based DMs. Code is available at this https: URL .

Comments: CVPR 2022

Subjects: Computer Vision and Pattern Recognition (cs.CV)

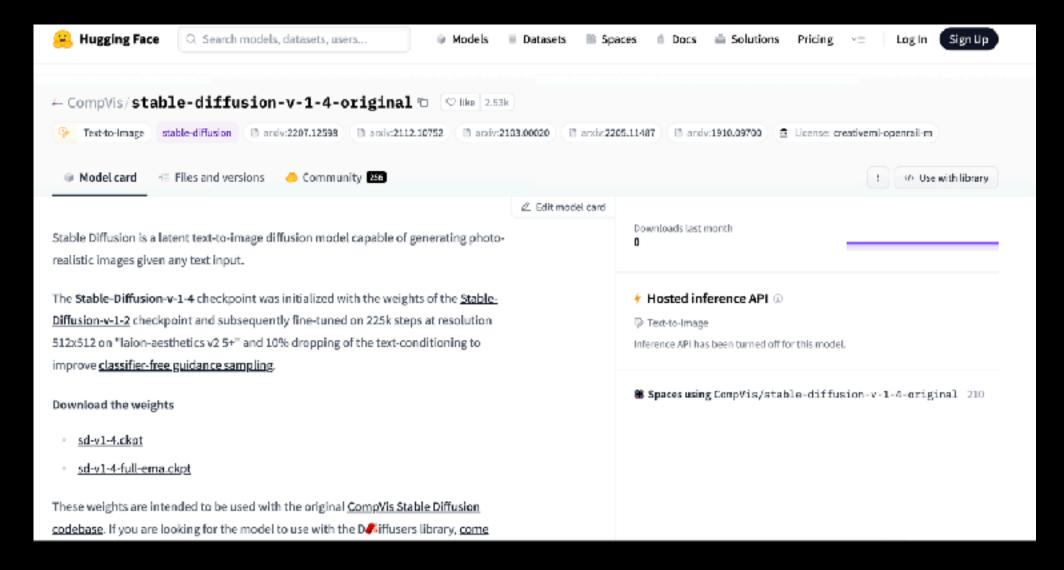
Cite as: arXiv:2112.10752 [cs.CV]

(or arXiv:2112.10752v2 [cs.CV] for this version) https://doi.org/10.48550/arXiv.2112.10752

Submission history

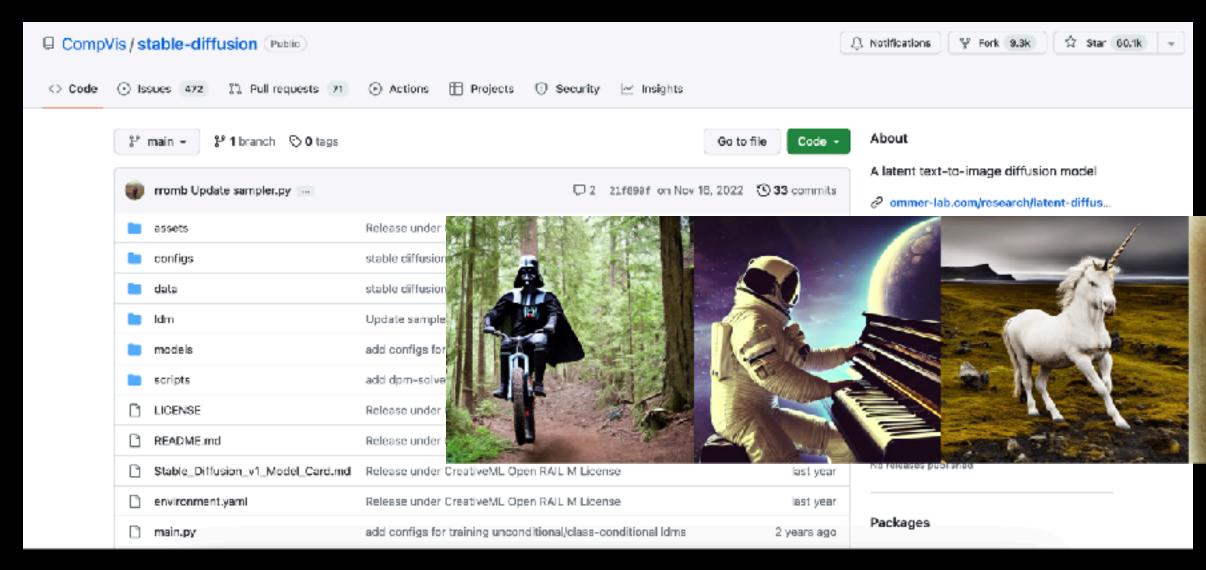
From: Robin Rombach [view email]

[v1] Mon, 20 Dec 2021 18:55:25 UTC (46,150 K8) [v2] Wed, 13 Apr 2022 11:38:44 UTC (38,971 KB)



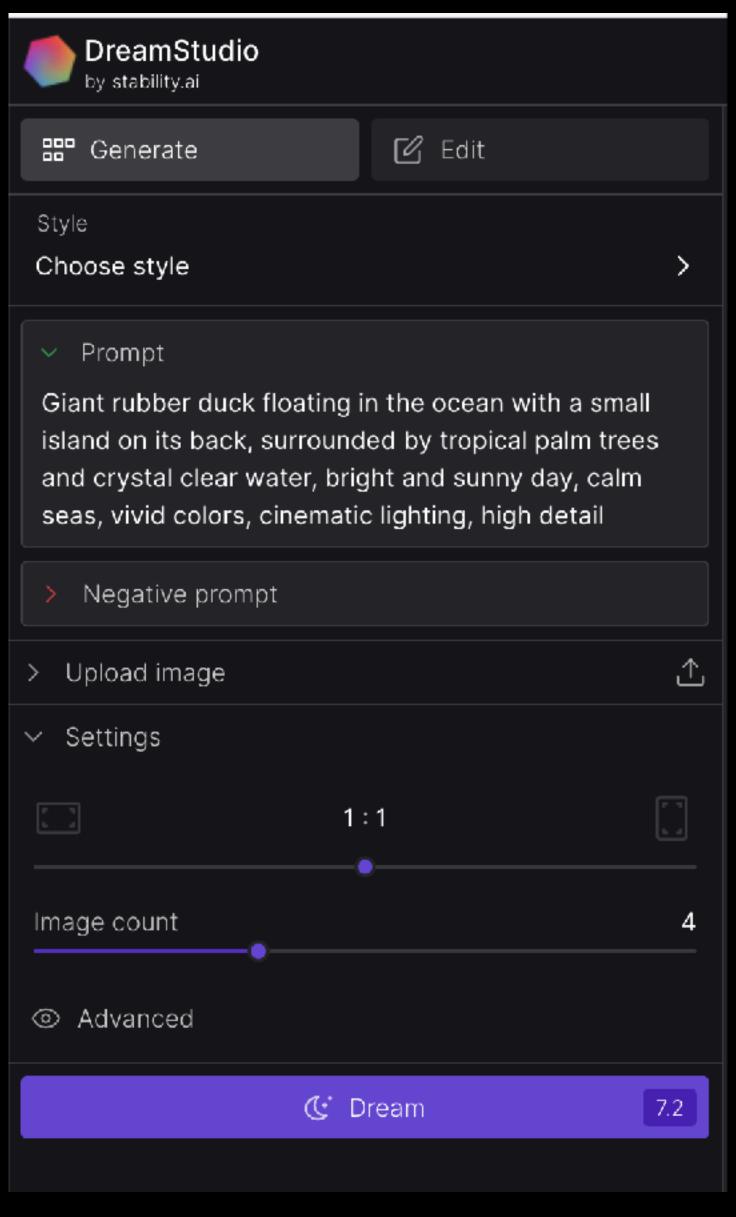
https://huggingface.co/CompVis/stable-diffusion-v-1-4-original

https://arxiv.org/abs/2112.10752





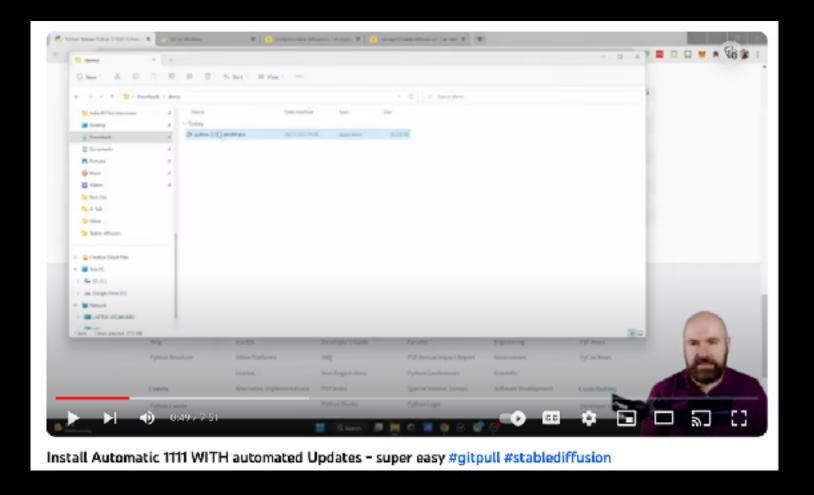
<u>https://github.com/CompVis/stable-diffusion</u>



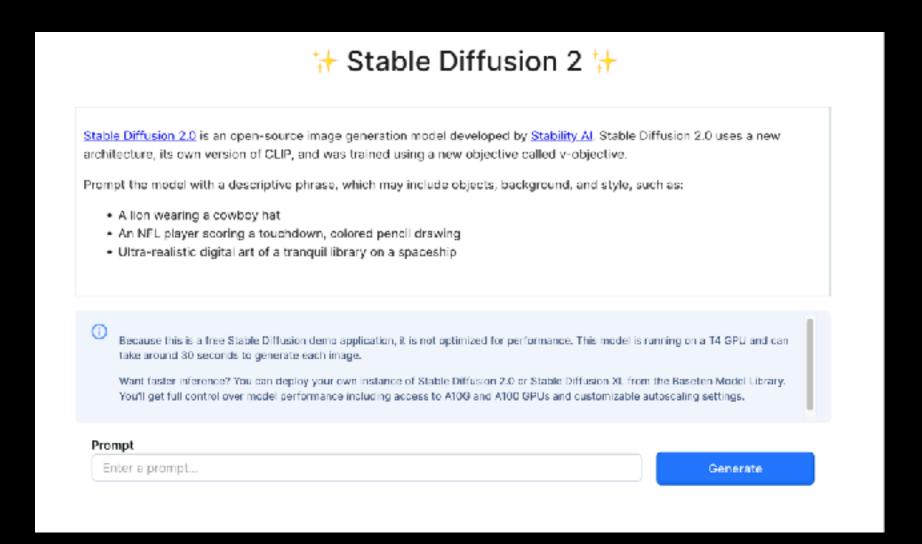
https://beta.dreamstudio.ai/generate

2img	img2img	Extras	PNG Ir	nfa C	heckpoint M	erger	Trai	in T	okenizer	Settings	Exte	ensions				
green sapling rowing out of ground, mud, dirt, grass, high quality, photorealistic, sharp focus, depth									of fleld		li		Generate			
Negative prompt (press Ctrl+Enter or Alt+Enter to generate)											li.	26/75	Style 1 None	Style 2		
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Euler a			~	_								10.5		-		
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1441/8/	169				0			Extra						3		
None								v			Save	Zip	Send to img2img	Send to inpaint	Send to	
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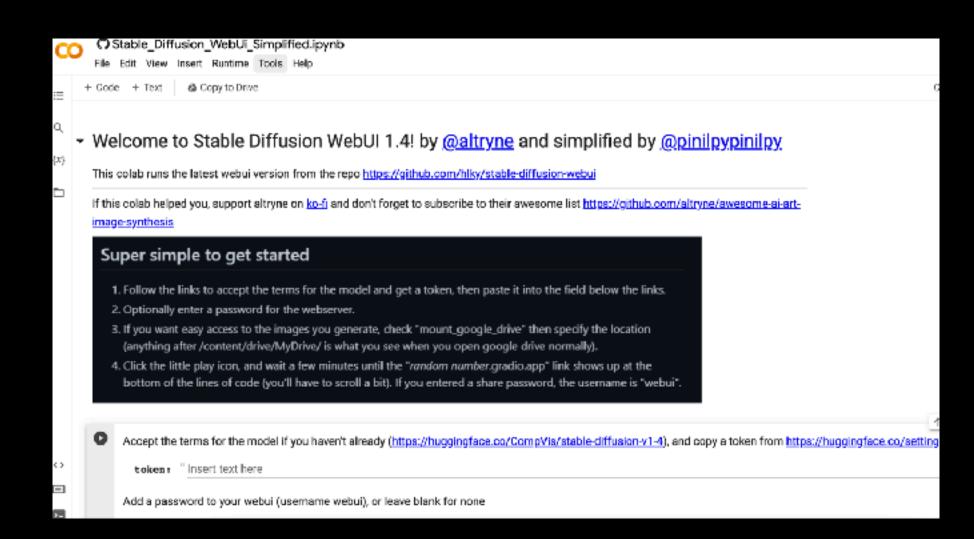
https://github.com/AUTOMATIC1111/stable-diffusion-webui



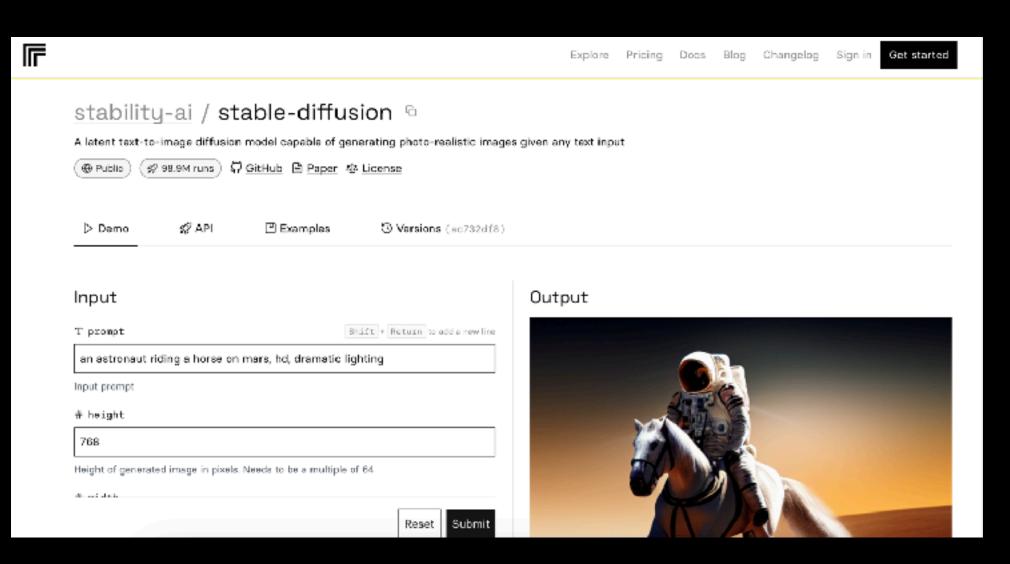
https://www.youtube.com/watch?v=3cvP7yJotUM



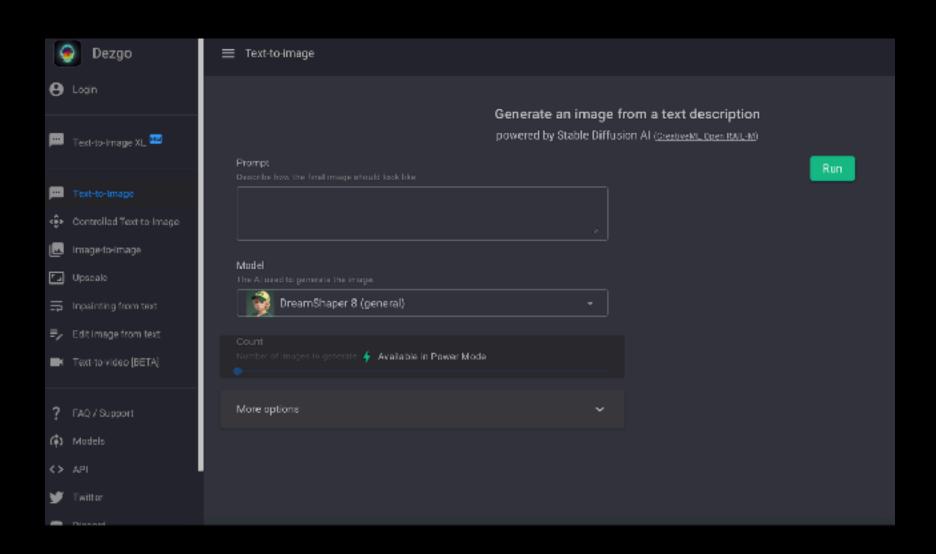
https://app.baseten.co/apps/VBlnMVP/operator_views/nBrd8zP



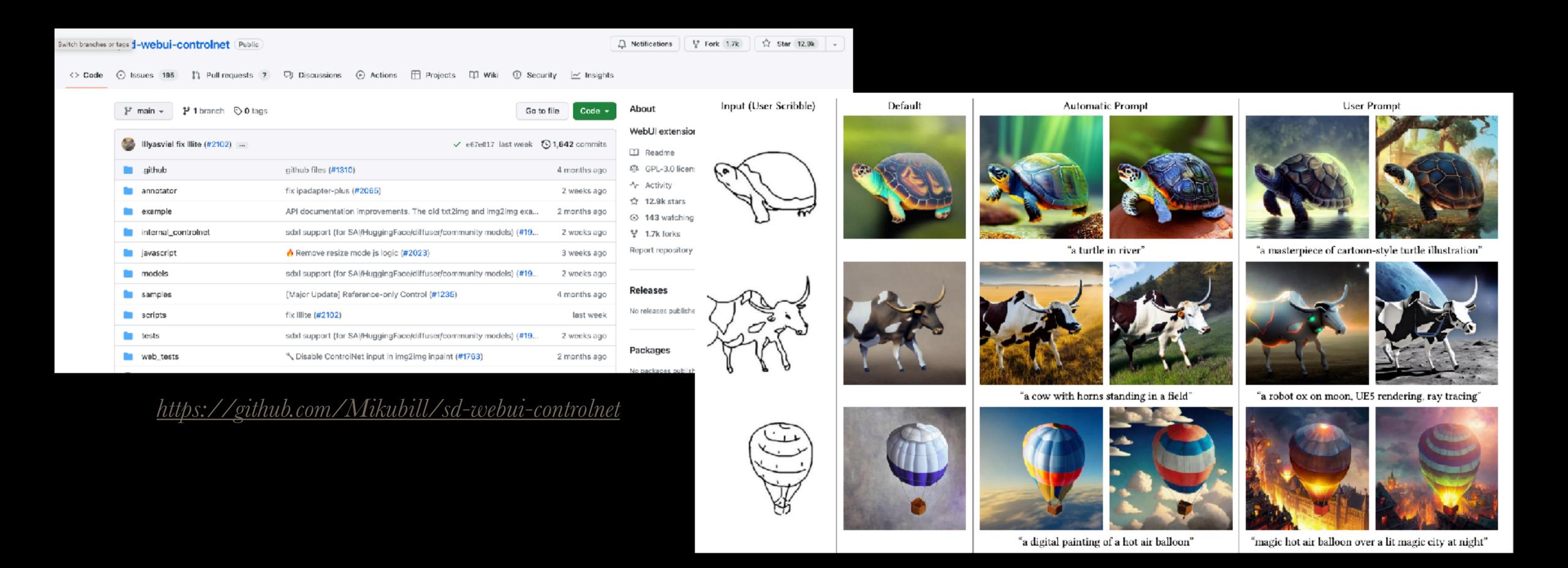
https://colab.research.google.com/github/pinilpypinilpy/sdwebui-colab-simplified/blob/main/ Stable Diffusion WebUi Simplified.ipynb#scrollTo=gk1T yBA0Arxt

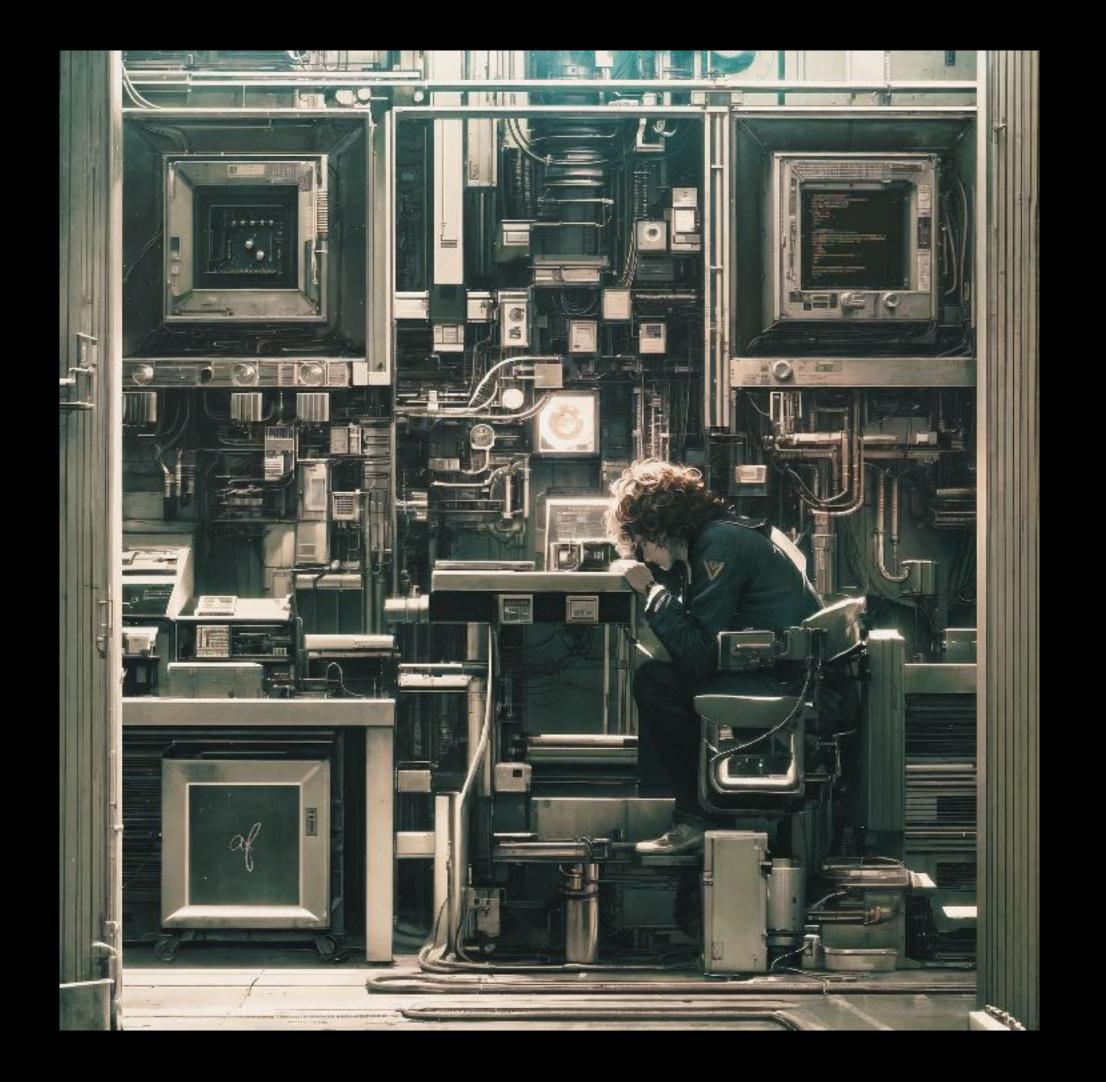


https://replicate.com/stability-ai/stable-diffusion



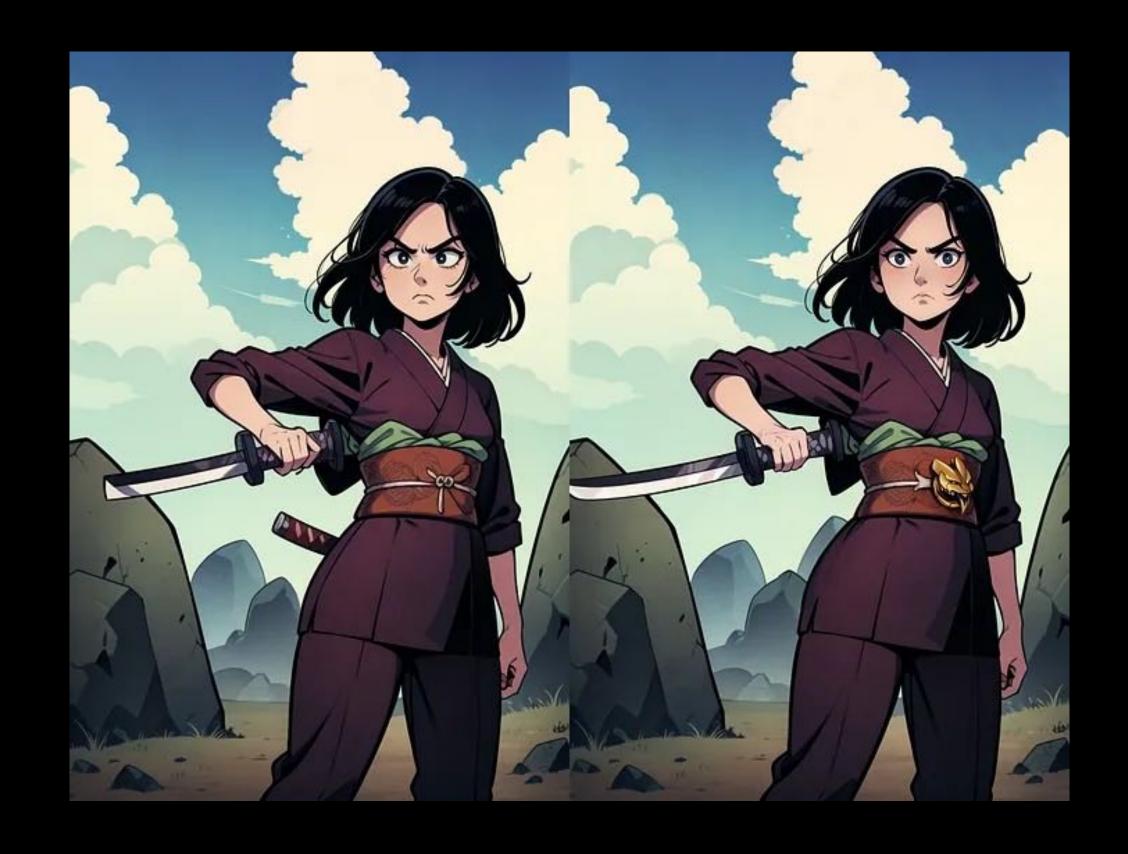
https://dezgo.com/text2image





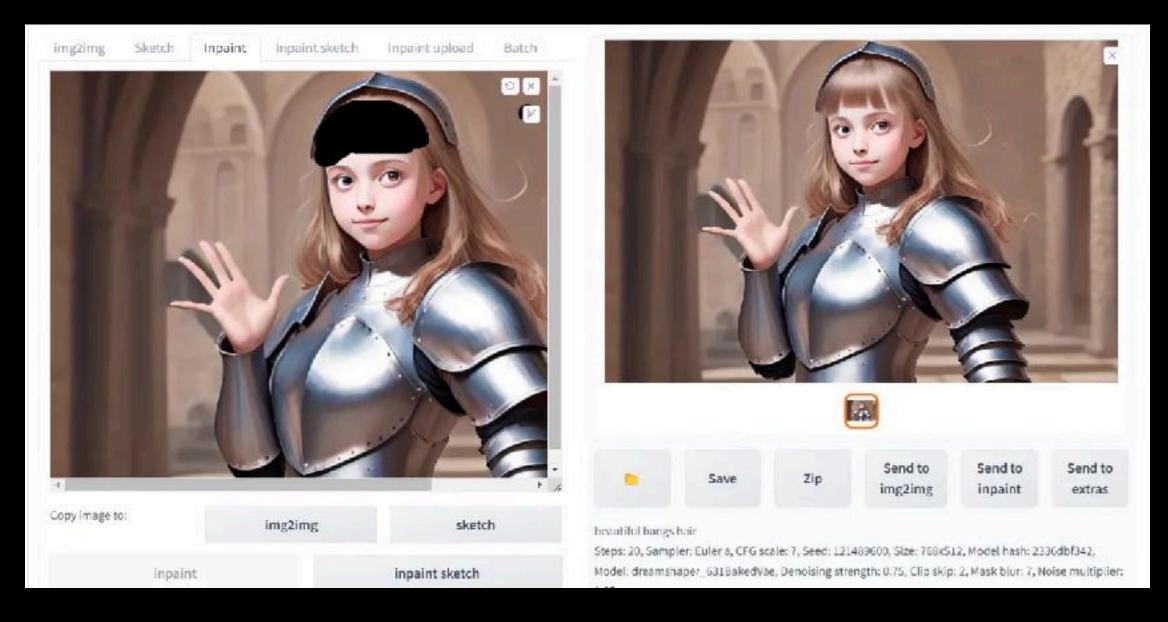


https://antfu.me/posts/ai-grcode

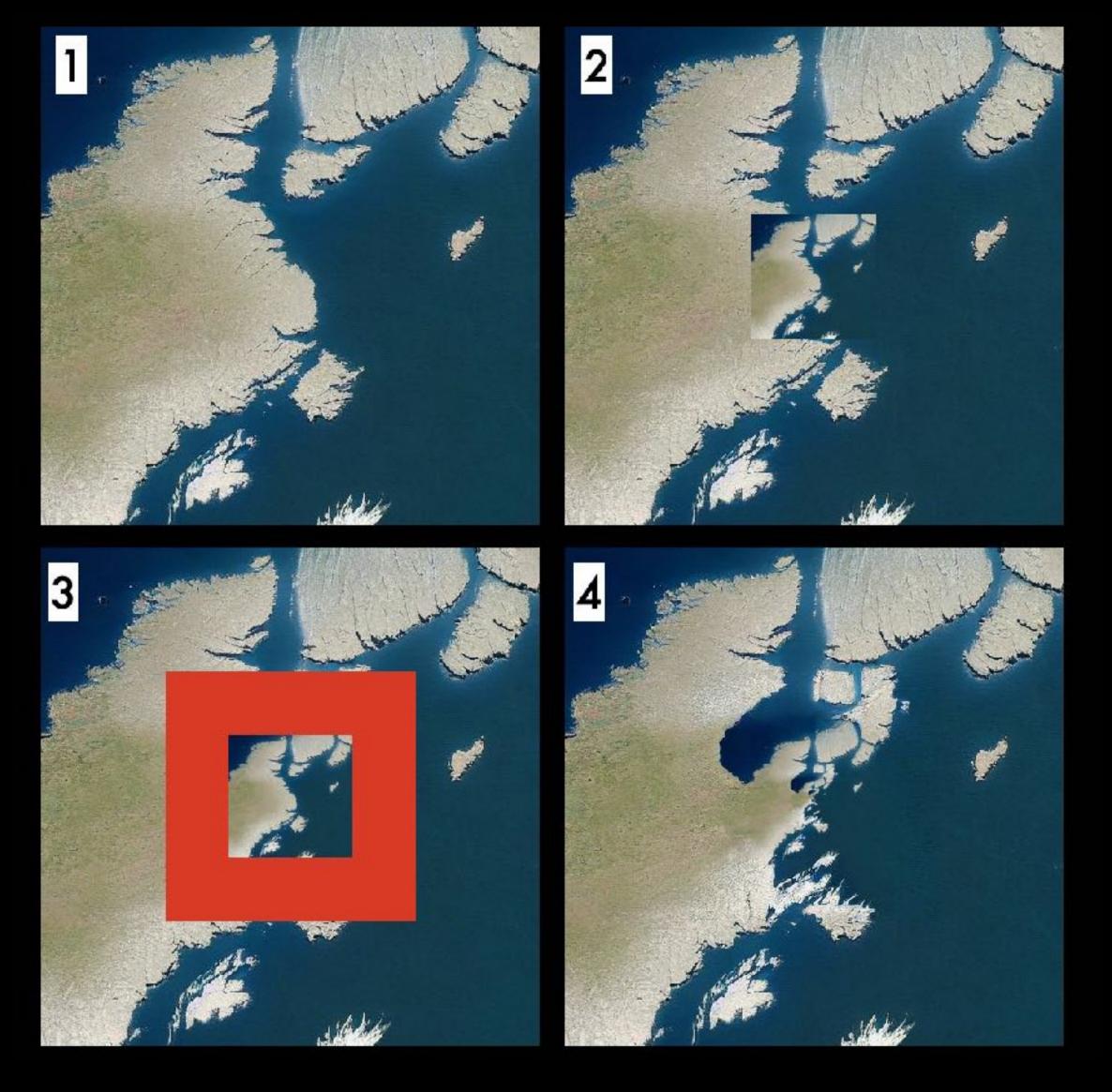


https://medium.com/@inzaniak/stable-diffusion-ultimateguide-pt-4-inpainting-772ea69472c9

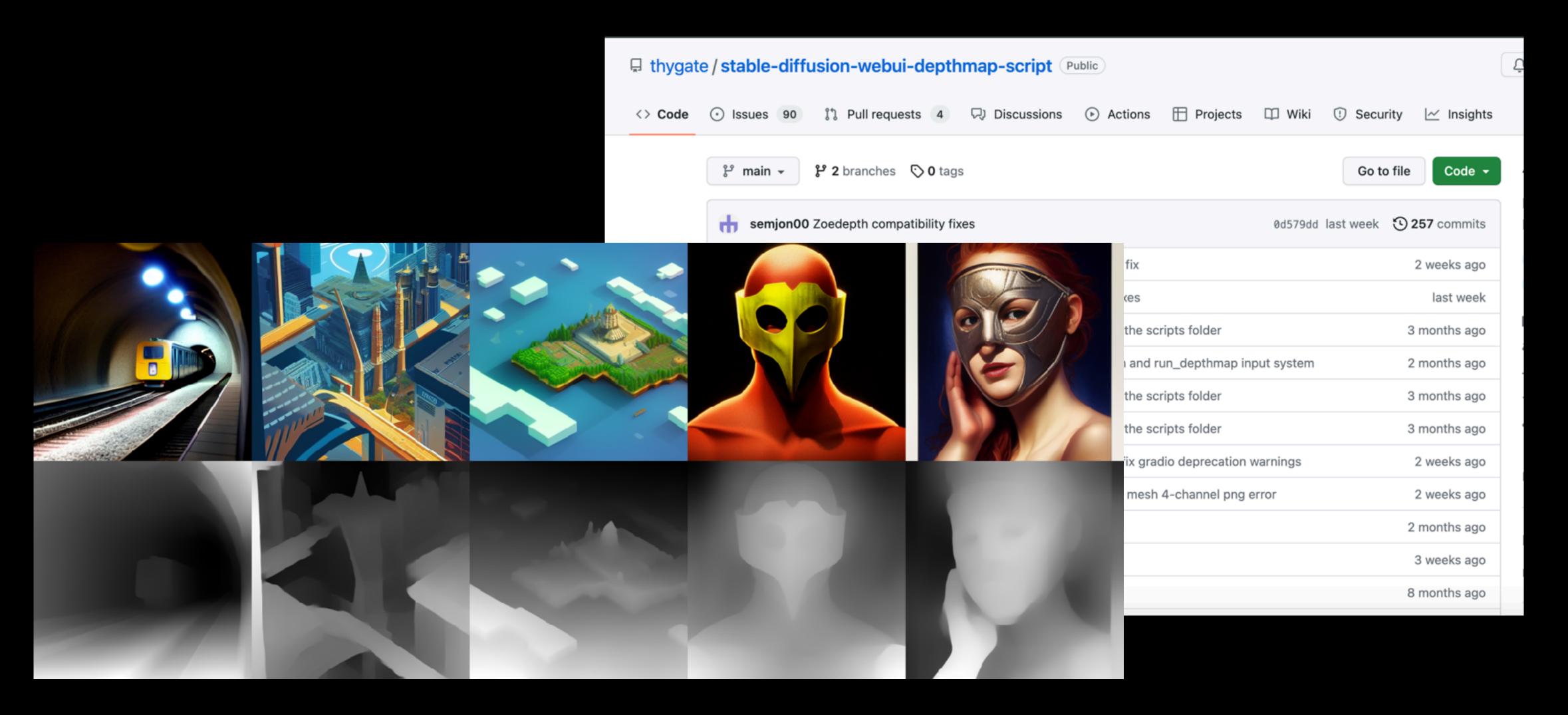




https://aituts.com/inpainting/



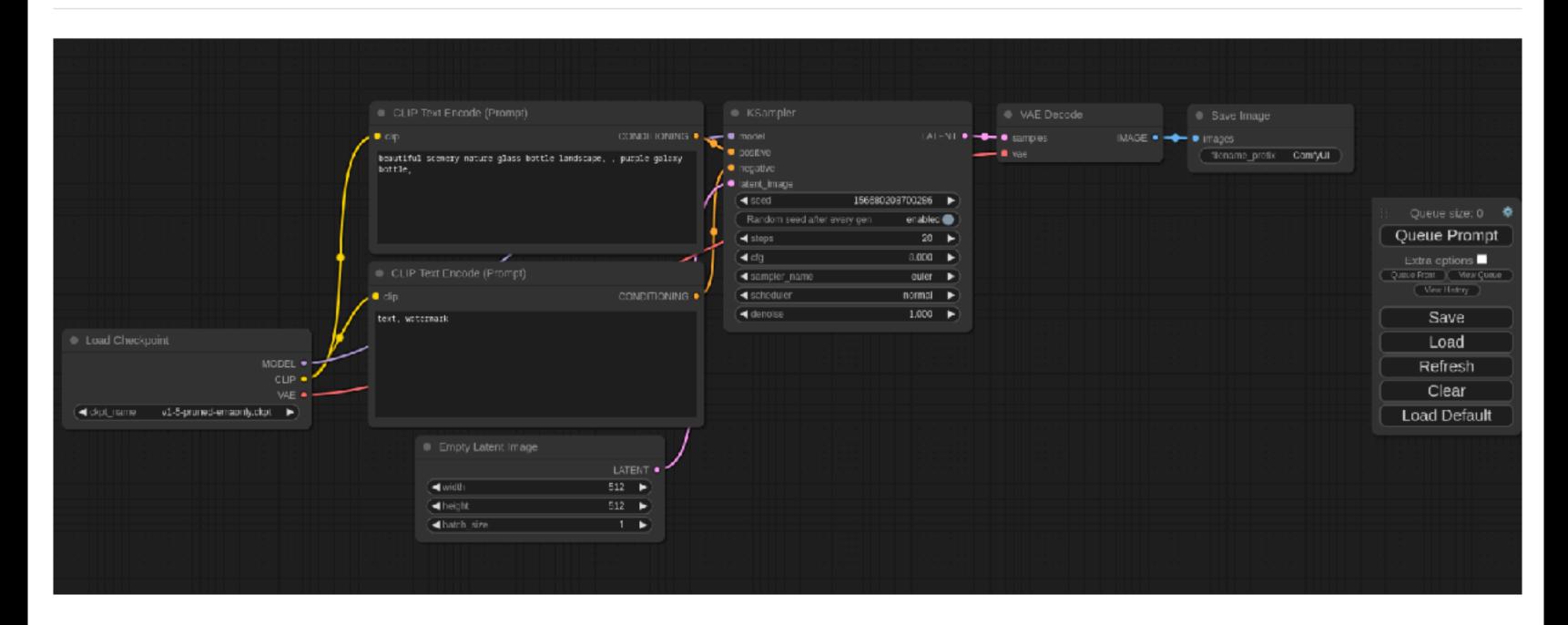
https://twitter.com/matthen2/status/1564608773485895692



https://github.com/thygate/stable-diffusion-webui-depthmap-script

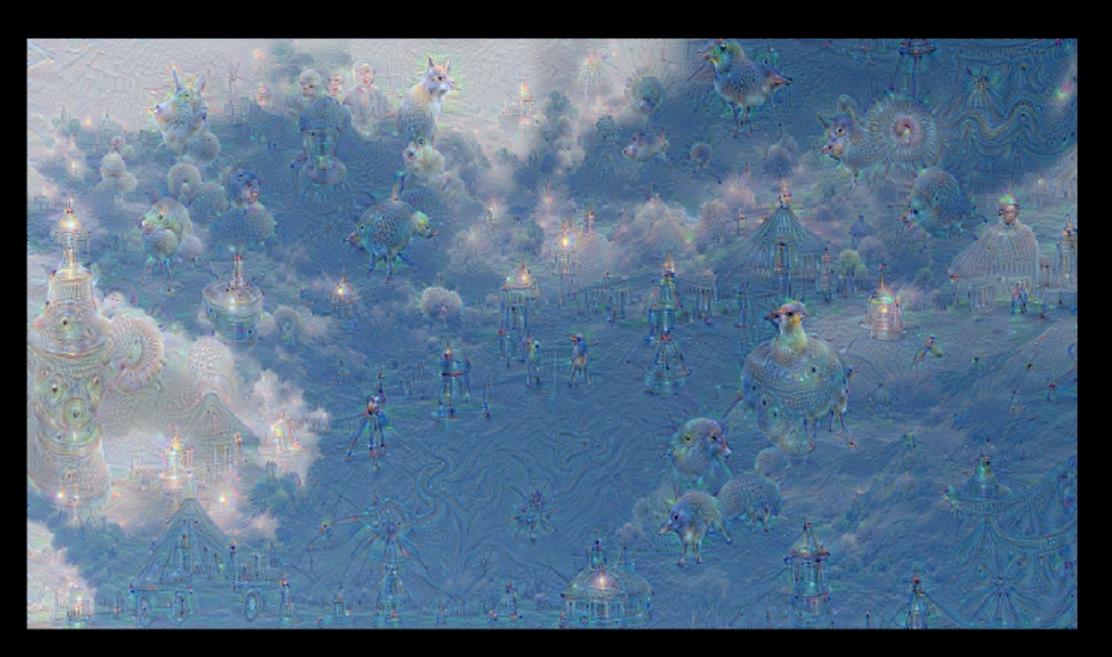
ComfyUl @

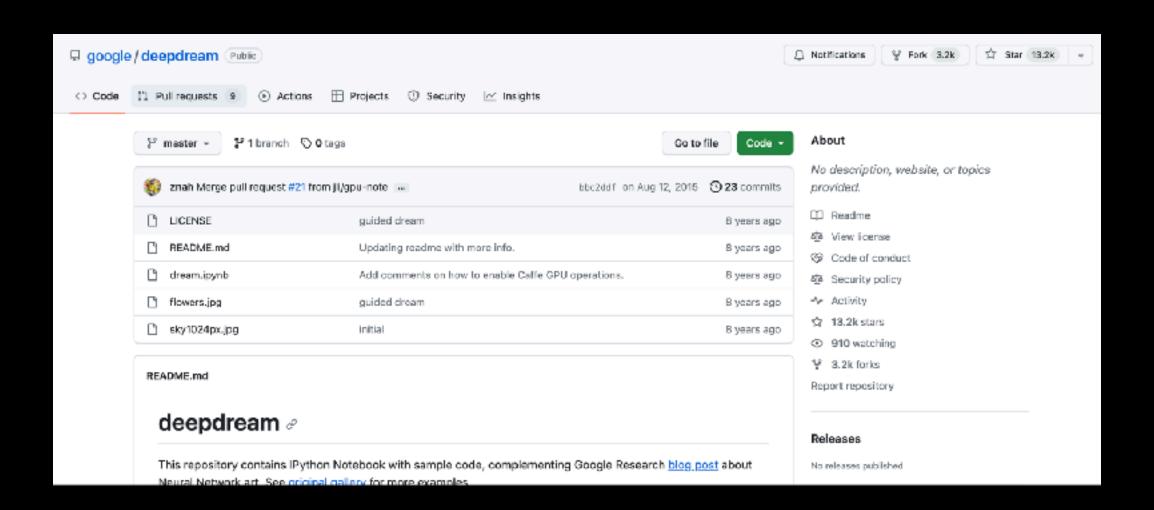
The most powerful and modular stable diffusion GUI and backend. @



This ui will let you design and execute advanced stable diffusion pipelines using a graph/nodes/flowchart based interface. For some workflow examples and see what ComfyUl can do you can check out:







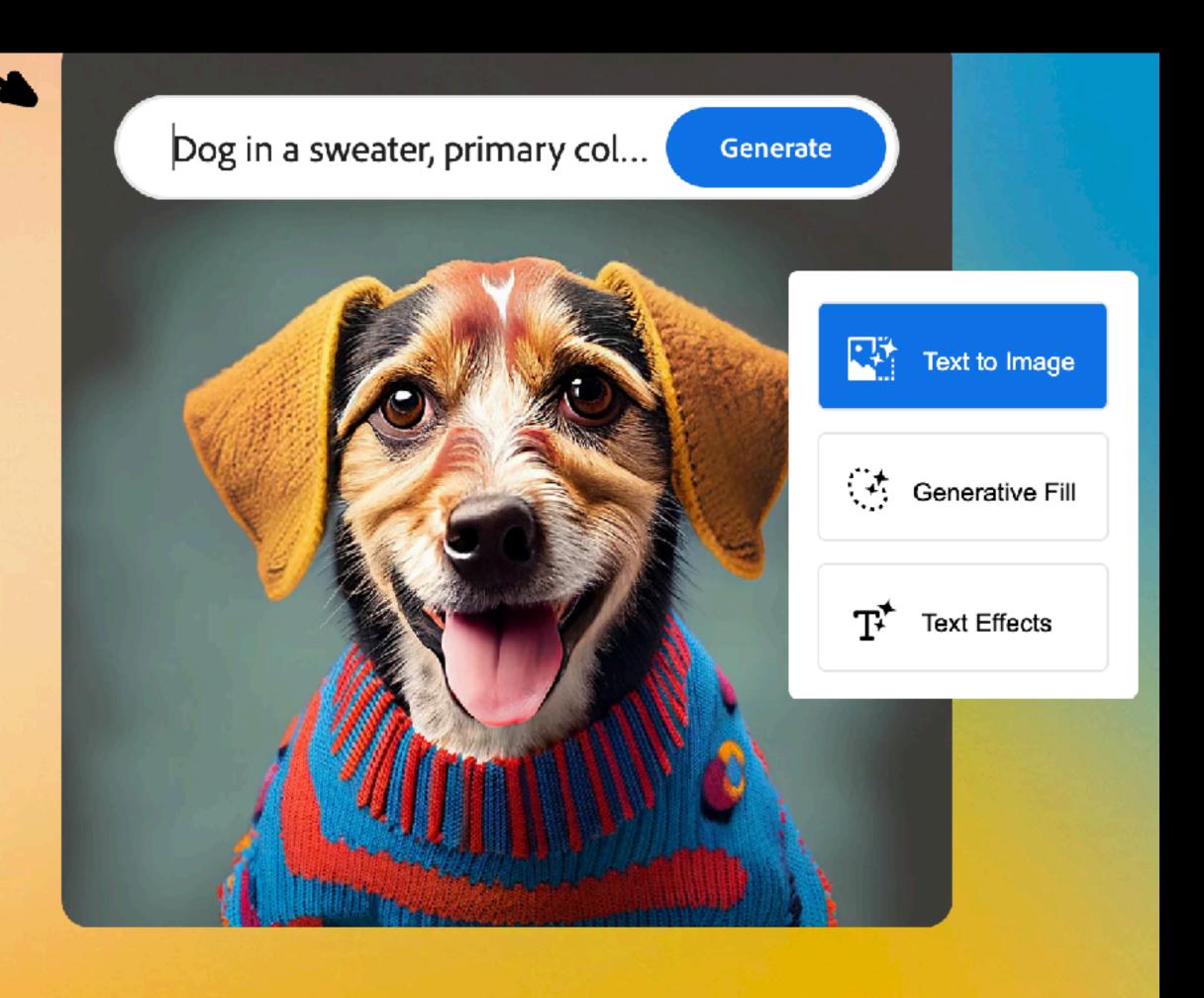
https://github.com/google/deepdream

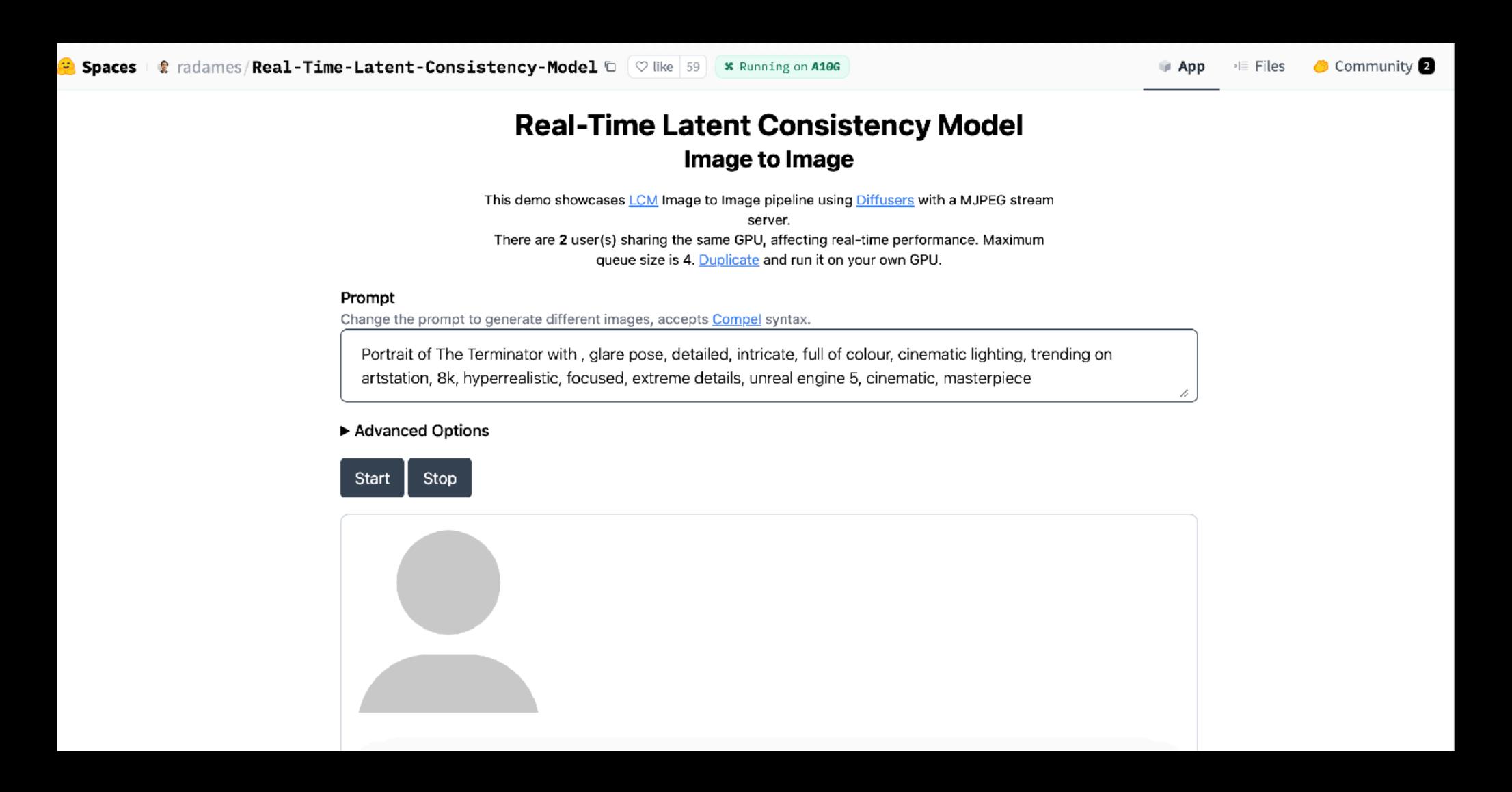
Adobe Firefly

Your imagination's new best friend.

Use generative AI and simple text prompts to create the highestquality output — beautiful images, text effects, and fresh color palettes. Make all-new content from reference images and explore more possibilities, more quickly.

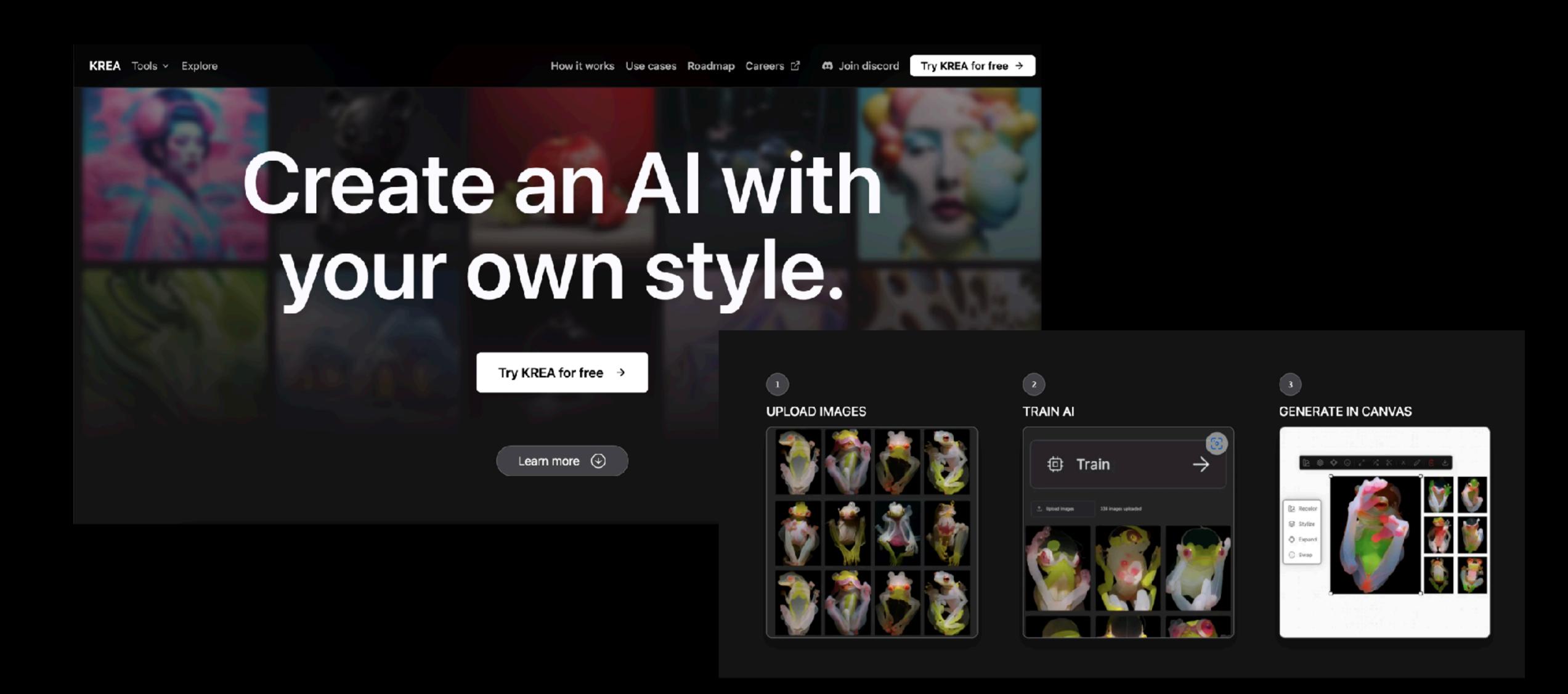
Get Firefly free



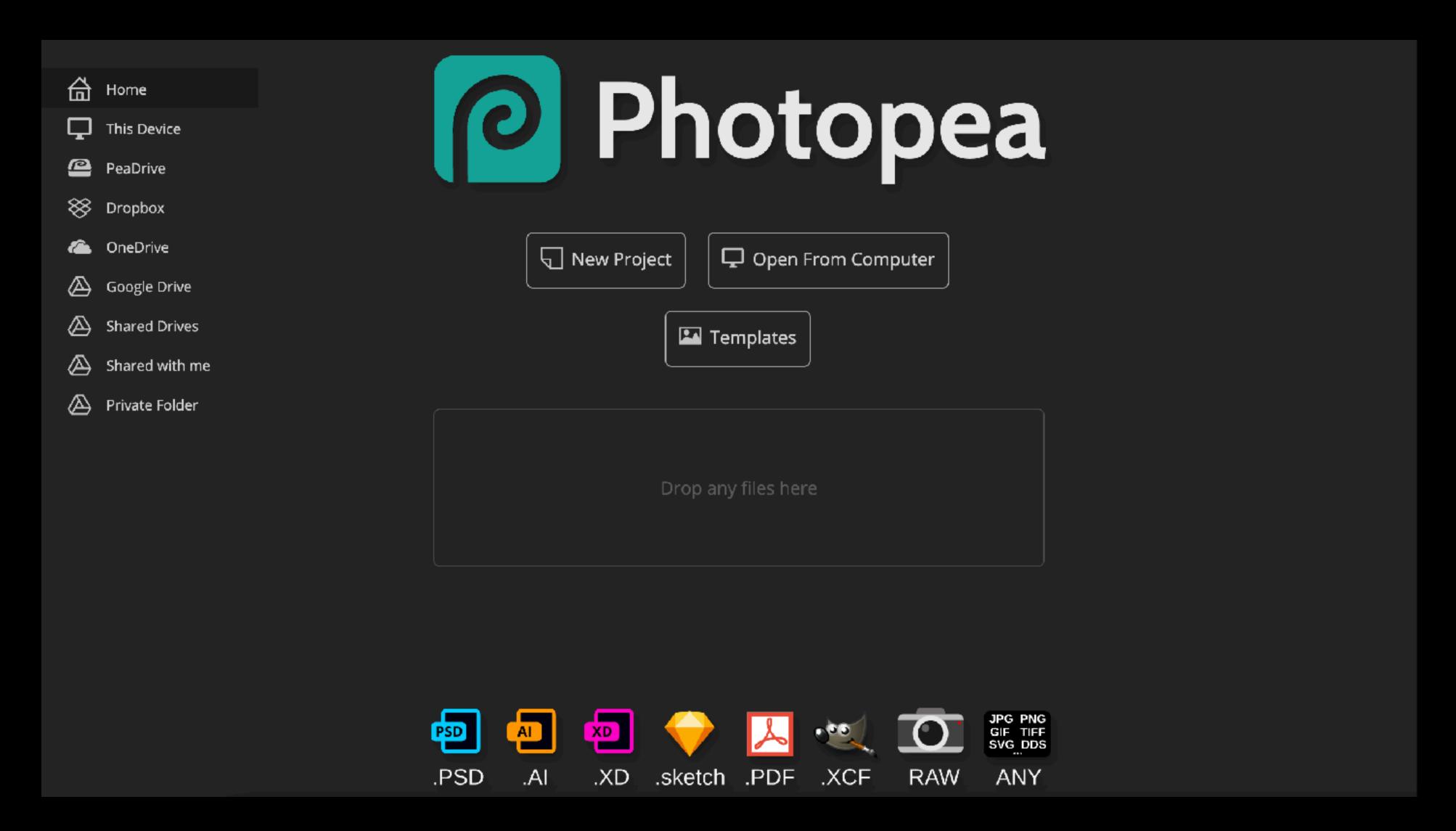


https://huggingface.co/spaces/radames/Real-Time-Latent-Consistency-Model

IMAGE STYLIZATION + EDITING

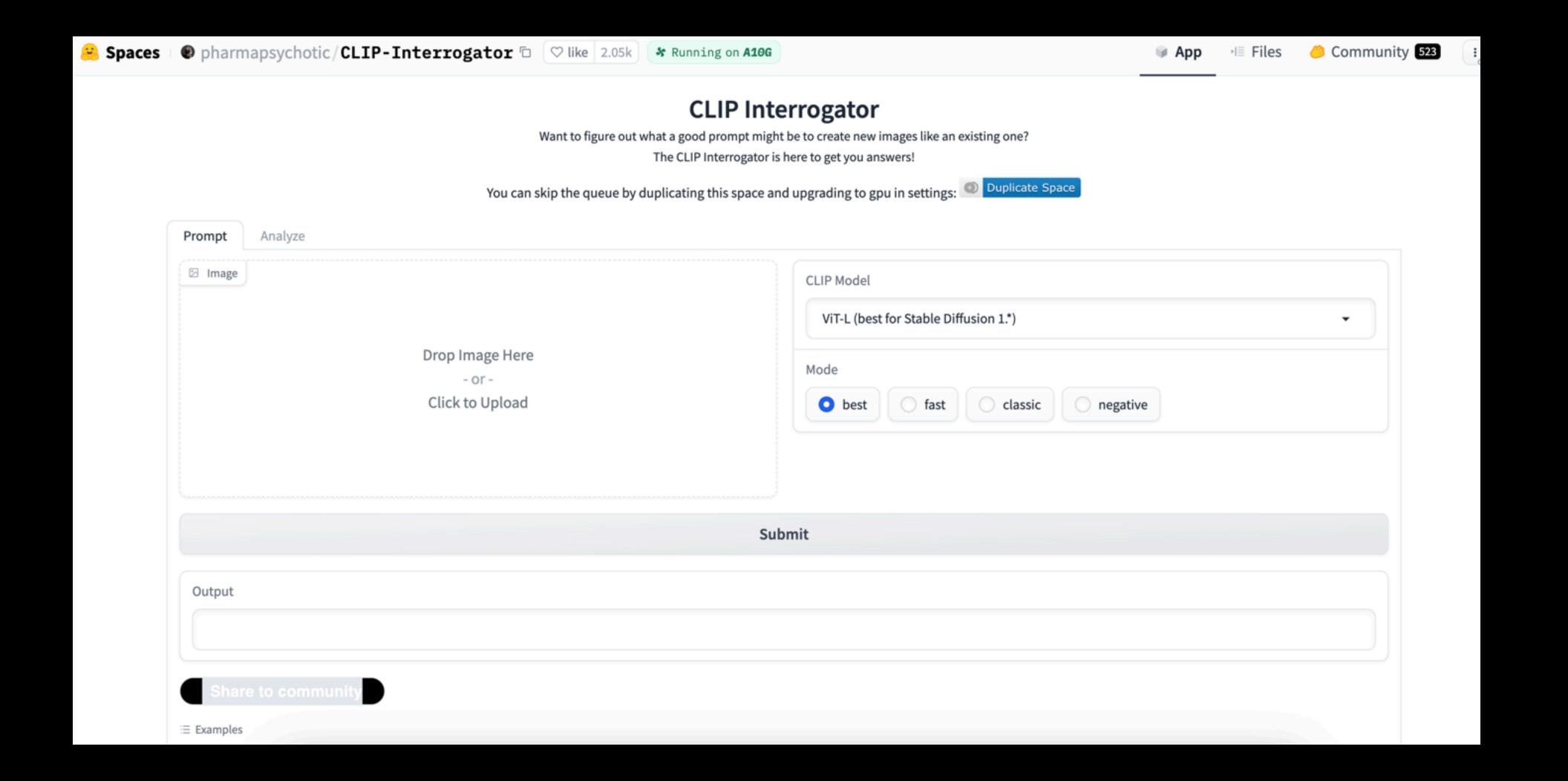


https://www.krea.ai/#how-it-works



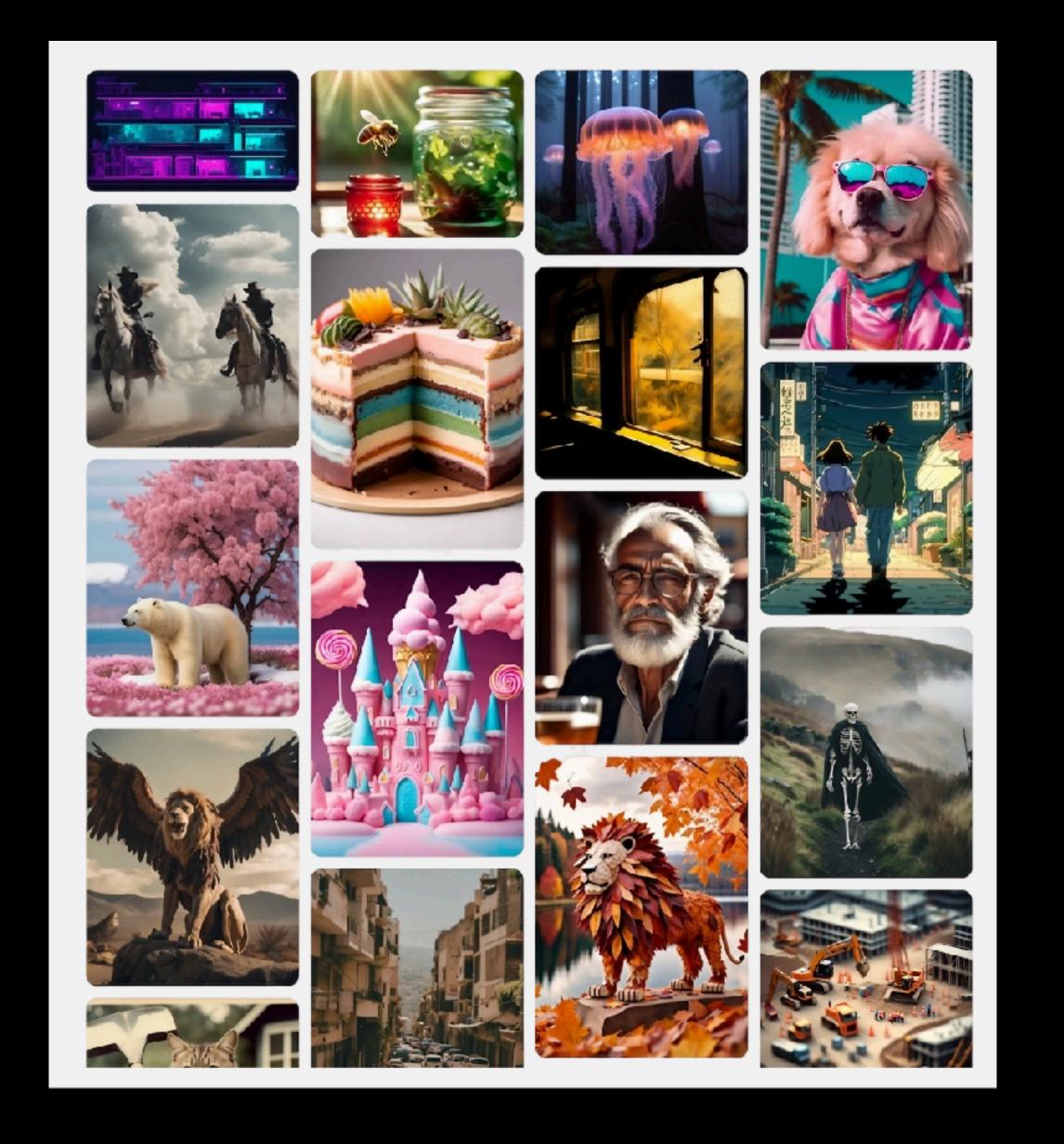
https://www.photopea.com/

IMAGE TO TEXT



What are some creative ways you could imagine using Stable

Diffusion or other generative image techniques?



https://huggingface.co/spaces/pharmapsychotic/CLIP-Interrogator

ARCHIVAL RESOURCES

Further approaches

- Image generation: These AI-powered tools leverage generative models to create new images from scratch. Some of the popular tools that use GANs (Generative Adversarial Networks) or VAE (Variational Auto-Encoders) include DALL-E, Artbreeder, and GANBreeder.
- Image enhancement: These tools use AI to enhance the quality of existing images by improving their resolution, removing noise, and adjusting the color balance. Some of the popular tools that offer these features include Let's Enhance, Enhance.AI, and Waifu2x.
- Image recognition: These tools use computer vision and deep learning algorithms to detect and classify objects within images. Popular examples include Clarifai, IBM Watson Visual Recognition, and Google Cloud Vision.
- Remove.bg: An Al-powered tool that removes background from images
- Let's Enhance: An image upscaling tool that uses AI to enhance and improve image quality
- DeepArt: A tool that uses neural networks to transform images into artistic styles, similar to famous paintings, etc.
- Artisto: An Al-based app that transforms pictures and videos into works of art.
- Enhance.ai: A solution that uses AI to improve the clarity, quality, resolution and color of images.

HOMEWORK

Homework

- In groups of two, explore the use of 3+ of the image generation tools presented in class
- Create images, artworks, etc.. related to your interests / topic for the course
- This work should be able to be presented as a standalone project, but should ideally contribute towards the development of your final project
- Presentation format is open to student preference, but should be limited to 3-5 minutes