

Generative Art

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AGENDA

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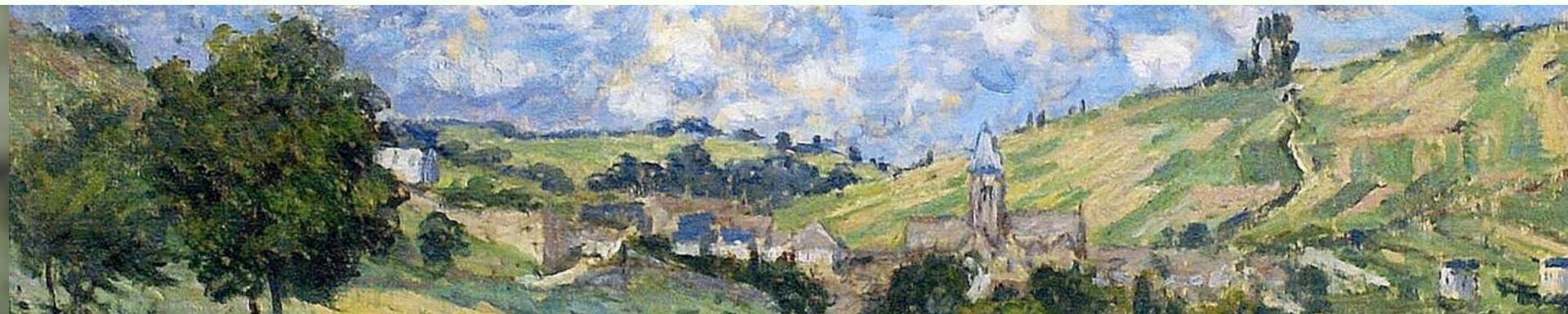
INTRODUCTION

What is Generative Art, Backstory
and Problem statement



What is Generative Art?

Any work of art created with
the assistance of artificial
intelligence.



Art doesn't have a concrete definition.

Some people might see it as a pure human activity
and it is a crime to automate art.

Others might see computers as tools to make art just
like brushes and colors.

We see it as an interesting opportunity to explore
how far computers can reach by understanding
human culture and communications.





Problem Statement

Generate aesthetic images with
the aid of Neural Networks.

02

DATA and Algorithm

More about the data





The data consists of 1,193 Images of Monet paintings, it can be found [in Kaggle](#).

-**Claude Monet** was a famous French painter whose work gave a name to the art movement Impressionism with capturing light and natural forms.

Workflow

1. Presproseseng

- Resize images
- CenterCrop
- Change image to Tensor
- Normalizing



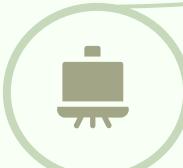
2. Build model

GAN discriminator & Generator layers



3. Train model

Finetune the model to find best result.



4. Present the results

- Make a short movie
- Monet Gallery VR experience



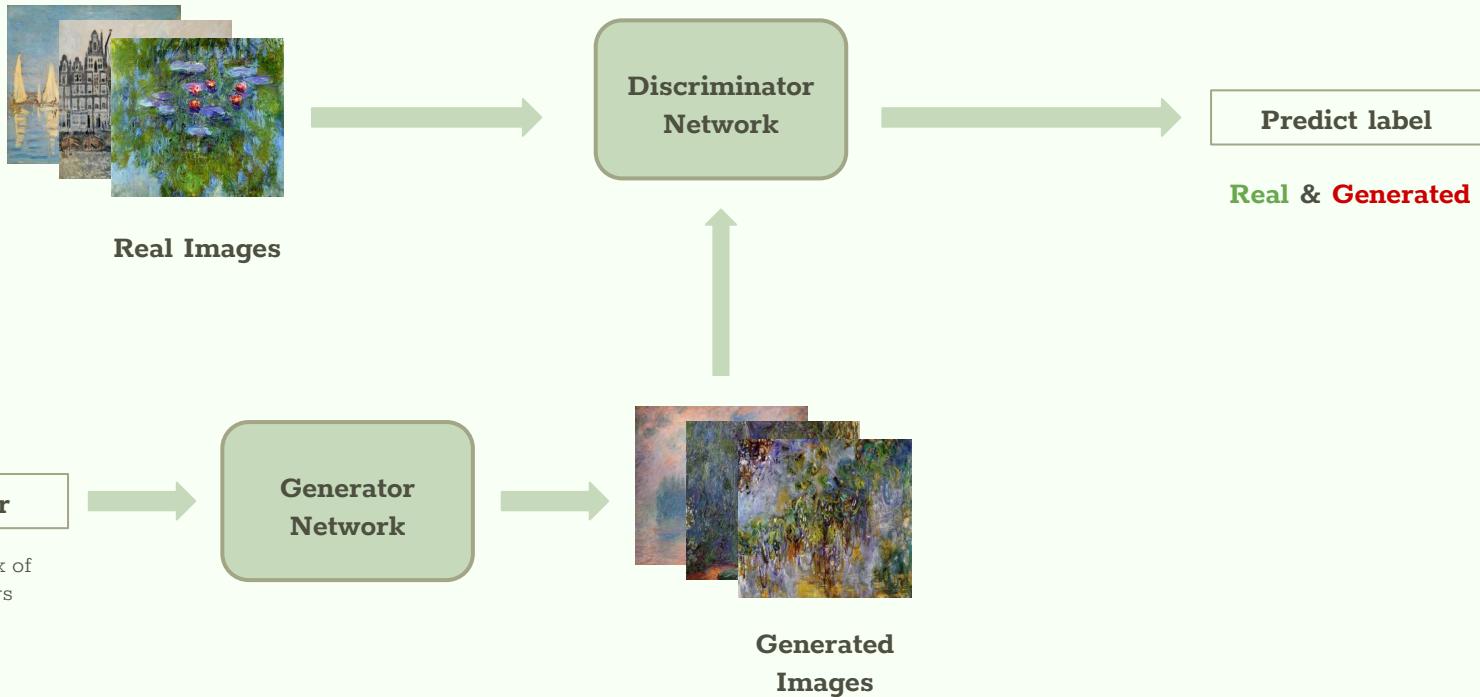
Generative adversarial network (GAN)

GAN is a machine learning (**ML**) model in which two neural networks compete with each other to become more accurate in their predictions.

The two neural networks are **generator** and **discriminator**.



Generative adversarial network (GAN)



Discriminator and Generator

Generator Network

The generator takes latent tensor (vector or a matrix of random numbers) as input, and generate images.

We used a convolutional neural networks (CNN)

- Sequential with ReLU and Tanh as an activation function
- Used PyTorch for creating grid, save images.

Discriminator Network

The discriminator takes an image as input, and tries to classify it as "real" or "generated".

We used a convolutional neural networks (CNN)

- Sequential with LeakyReLU and Sigmoid as an activation function

Binary cross entropy

$$\text{LogLoss} = -\frac{1}{n} \sum_{i=1}^n [y_i \log(\hat{y}_i) + (1 - y_i) \log(1 - \hat{y}_i)]$$

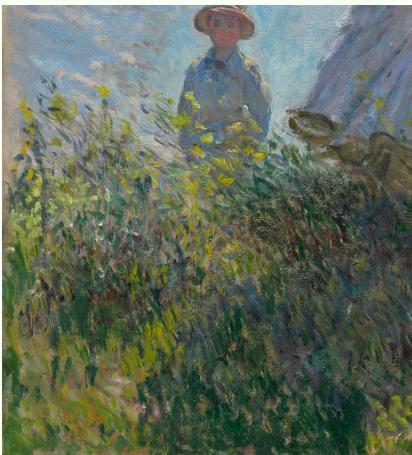
- Creates a criterion that measures the Binary Cross Entropy between the generated and the real image probabilities.
- Treat it as a binary classification problem with labels 0 and 1 for generated and real images respectively.
- Best Log Loss value is zero.

[Source](#)

Challenges

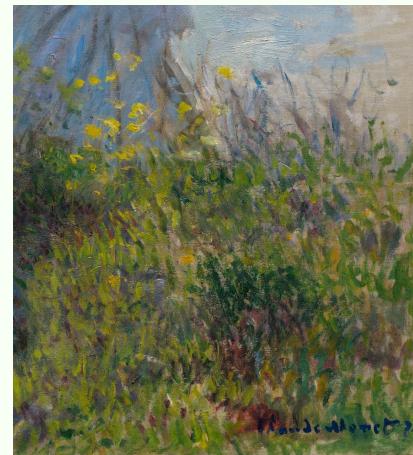
1. Resources

GPU and its memory was small for training the model with high pixels



3. Training Time

The model took a lot of time to train and experiment.



2. Model Complexity

Understanding the model and its best practices was challenging and took a lot of time.

03

MODEL and RESULTS

What are the used technique? Results?



PRE-PROCESSING STEPS



Resizing Images

Convert images from 256x256 pixels to 64x64 pixels to match our GPU Capacity

Center Crop

Crop the given image at the center.

ToTensor

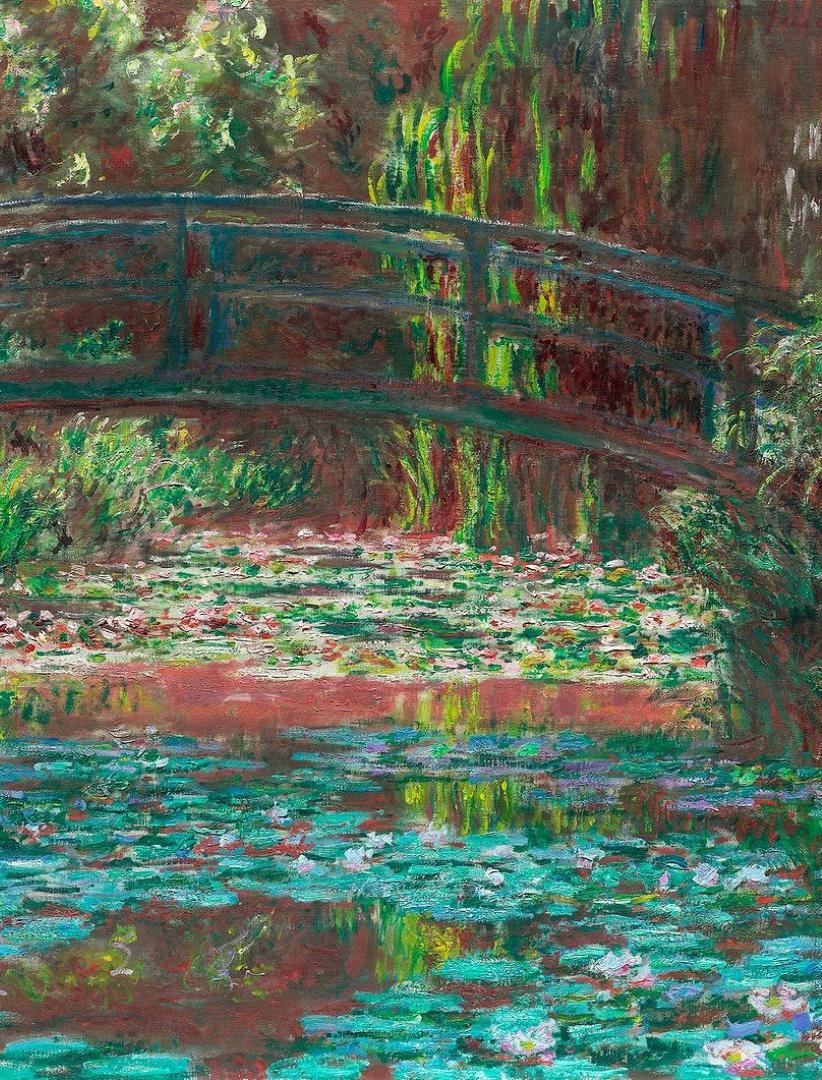
Convert the image to `torch.FloatTensor`

Normalization

Normalize a float tensor with mean and standard deviation

GPU Set-up

CUDA is a parallel computing platform and application programming interface allowing using GPU.



Device Specification

RAM: 16.0 GB

GPU: Nvidia GeForce RTX 2070 / 930

VRAM: 8.0 GB

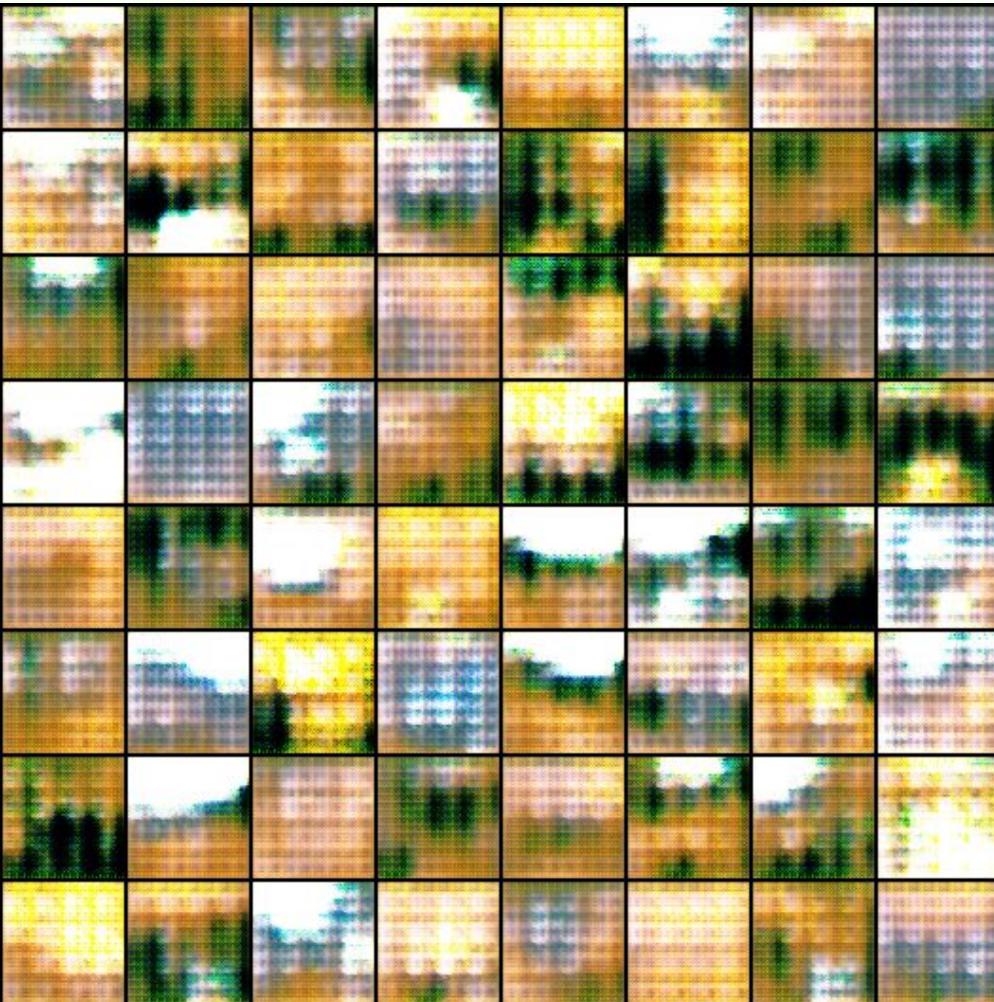
Training Time: Around 8 hours



Baseline Model

- Epochs=100
- Learning rate=0.01
- Pixels = 64
- Batch size = 128
- Latent size = 128



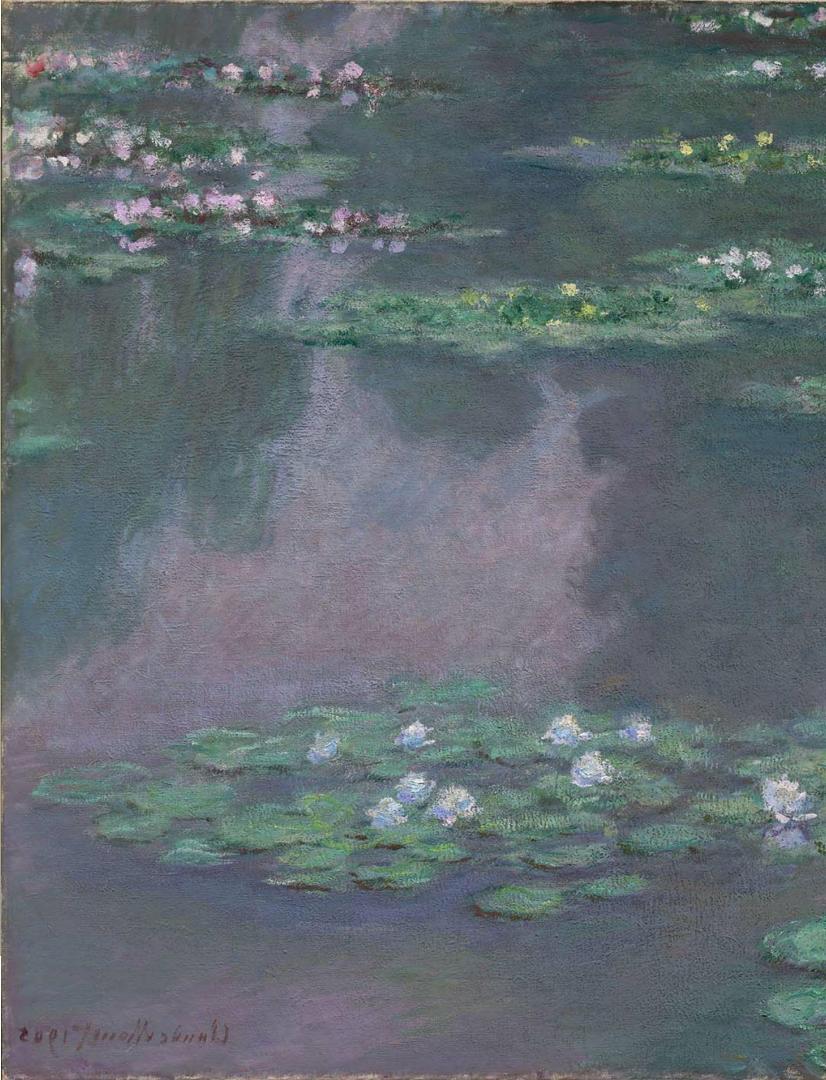


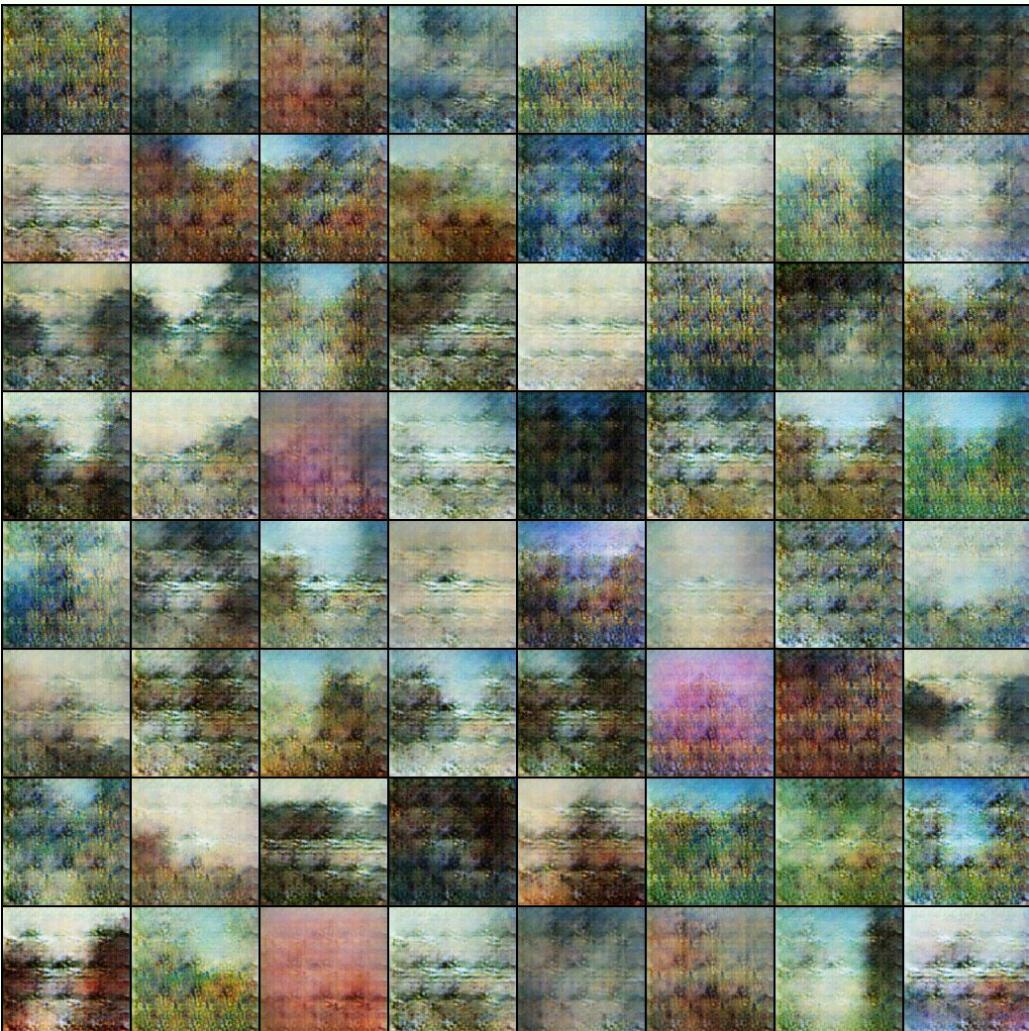
Baseline Model

- 64x64 pixels
- Epoch: 100
- High Loss

More Pixels

- 128 pixels failed to give a good results, epochs stopped at 365 with high loss.
- 256 pixels wouldn't work on our GPU.



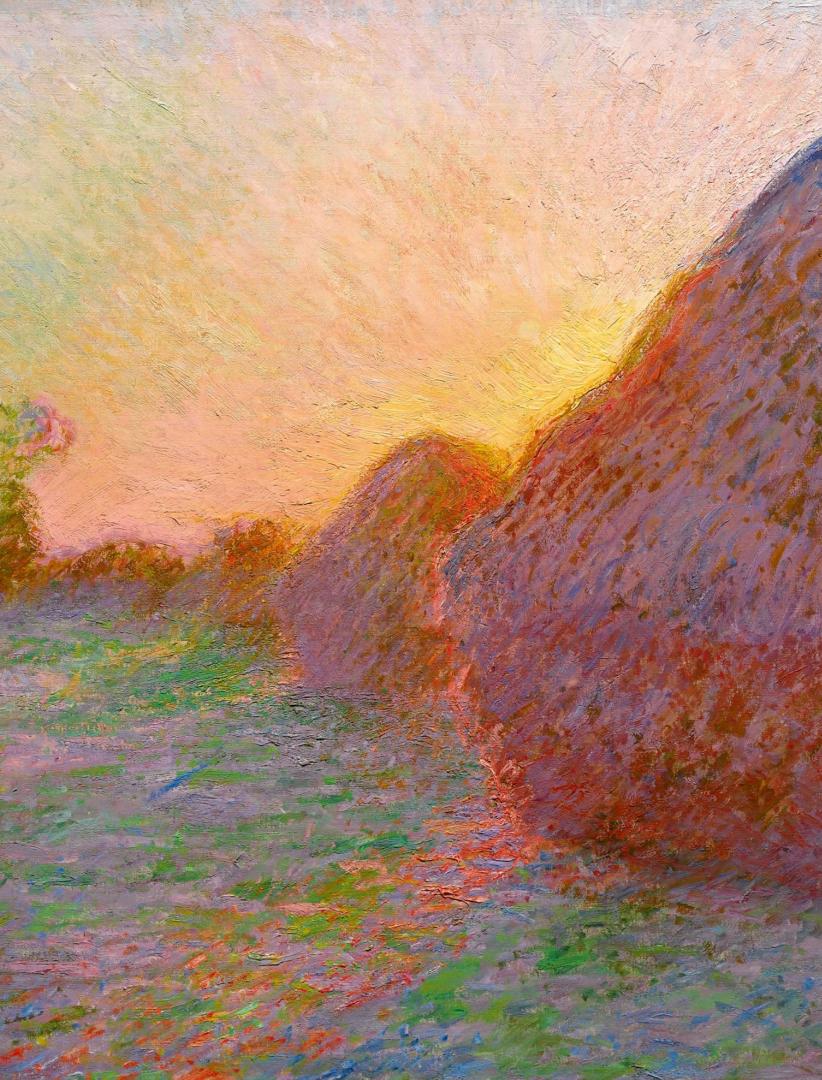


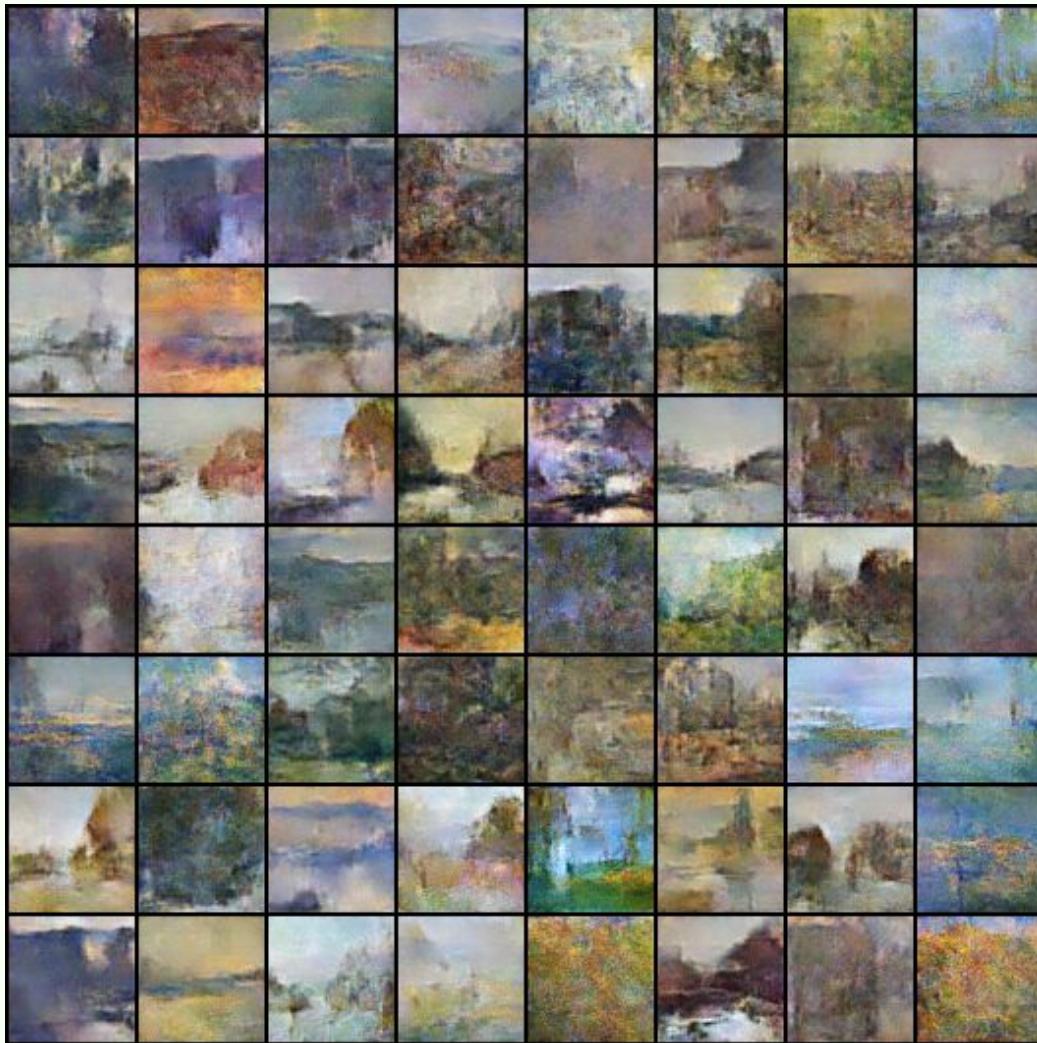
More Pixels

- 128x128 pixels
- Epoch: 365,
loss_g: 5.2822,
loss_d: 0.0240

Best Model

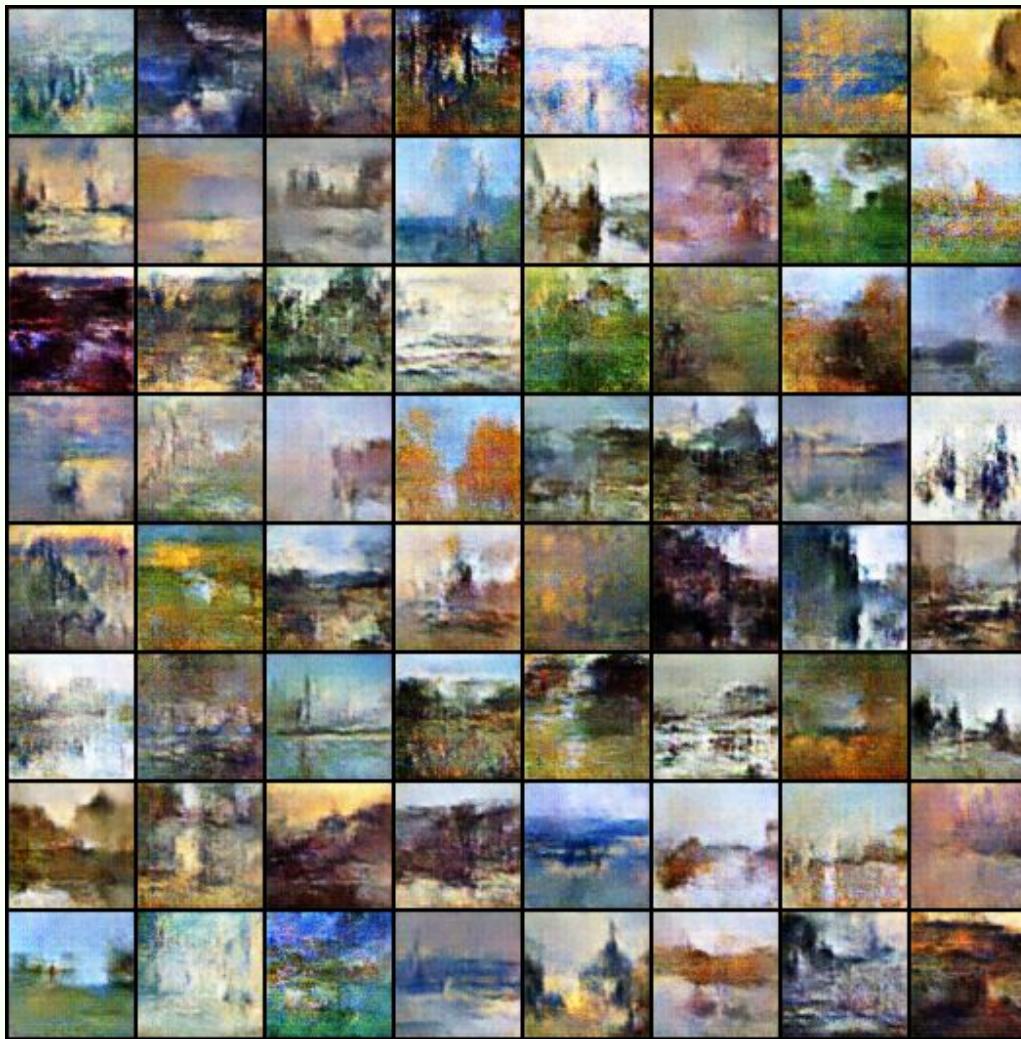
- epochs=500
- learning rate=0.0001
- Pixels = 64
- batch size = 64
- latent size = 128

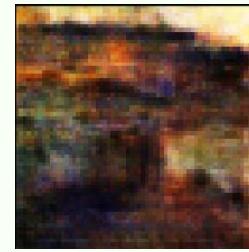
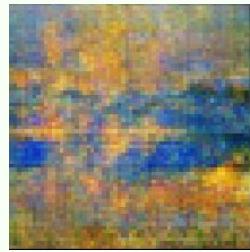
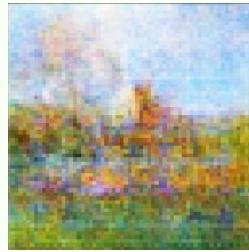




More Pixels

- 128x128 pixels
- Epoch: 500
 - loss_g: 5.2822,
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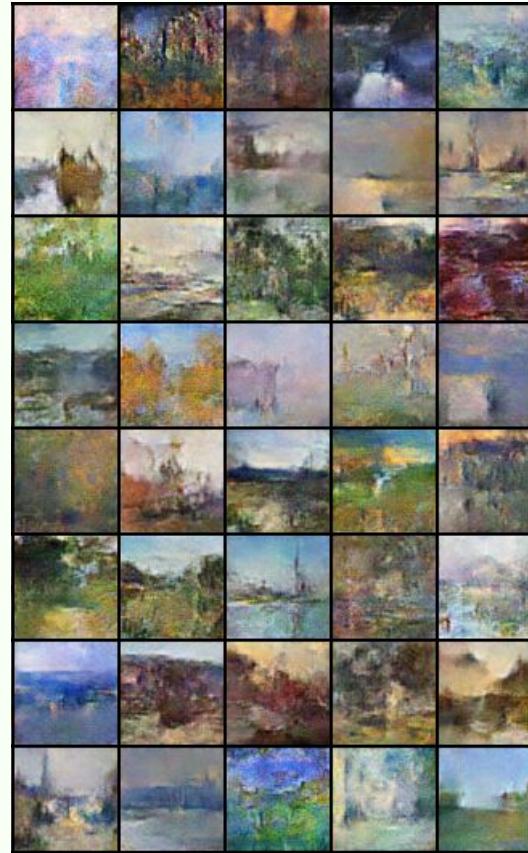


Model Score

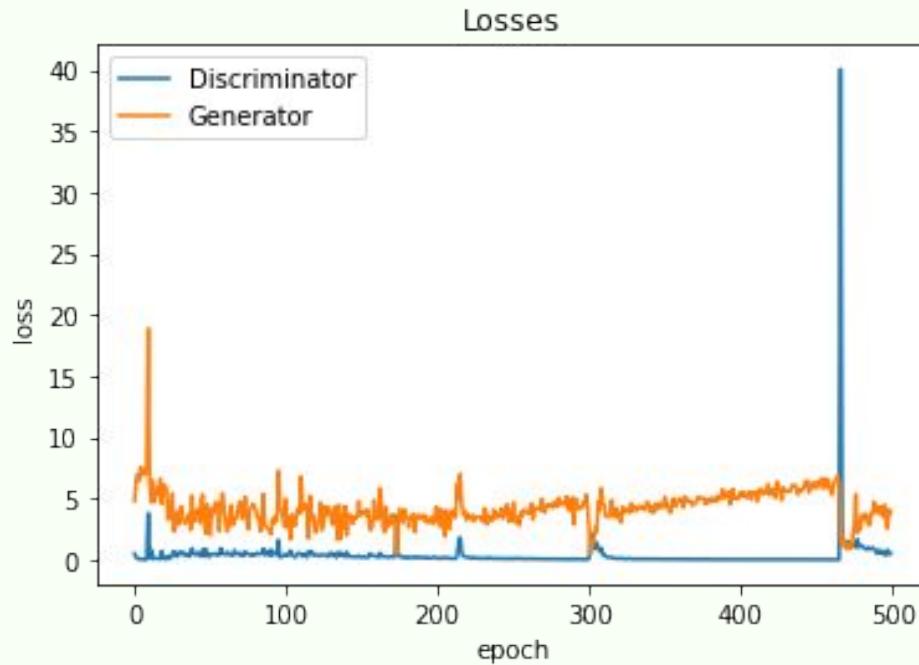
Early Stage



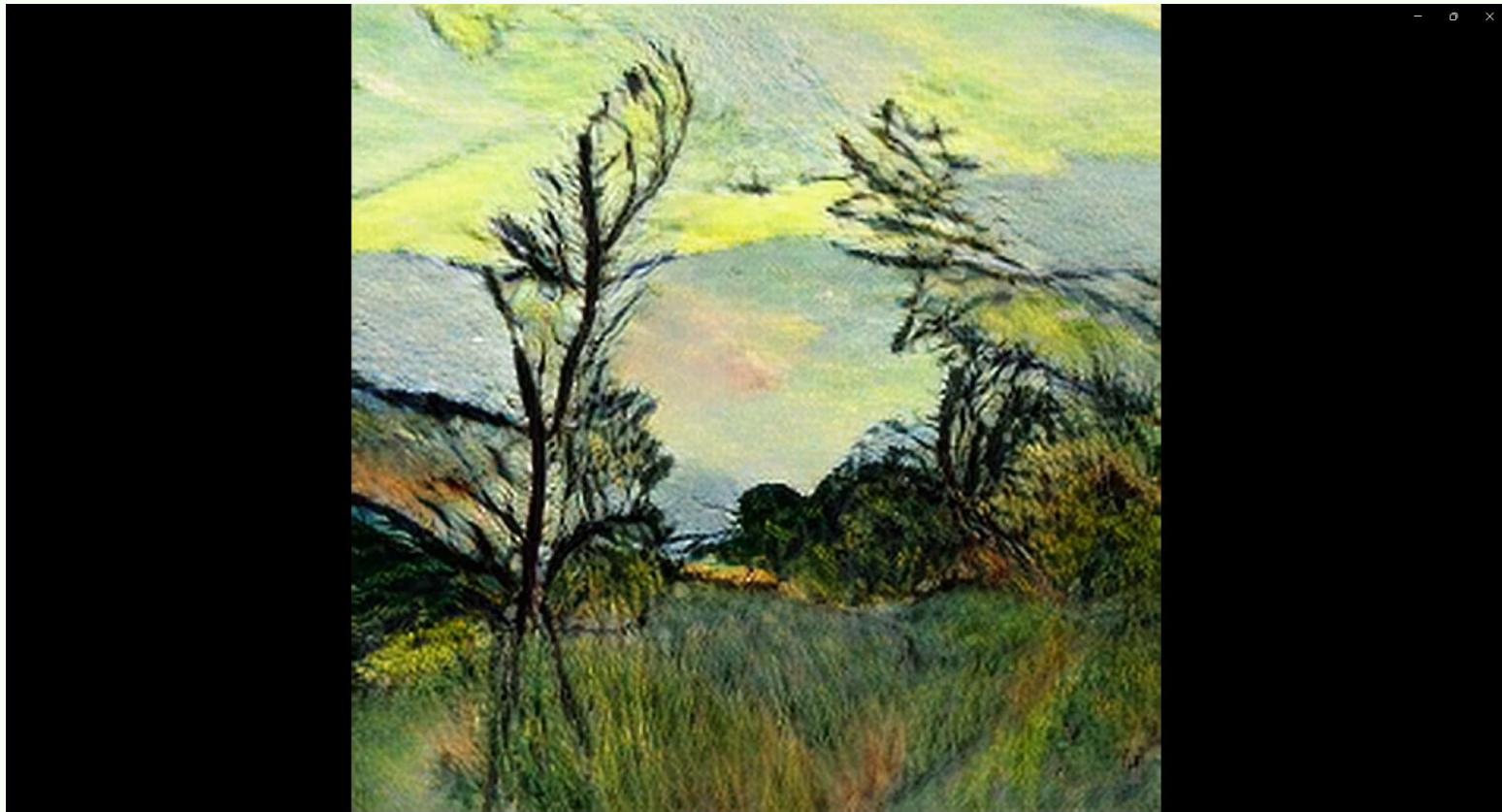
Final Model



Loss



Short Film



VR Monet Gallery Experience



04

CONCLUSION

Deliverables, Recommendations



CONCLUSION

Deliverables

1. Generated images.
2. Machines have a good chance to destroy humanity because we made them smart enough to make art.

Recommendations

Train the model on more powerful GPU and with higher pixels.