


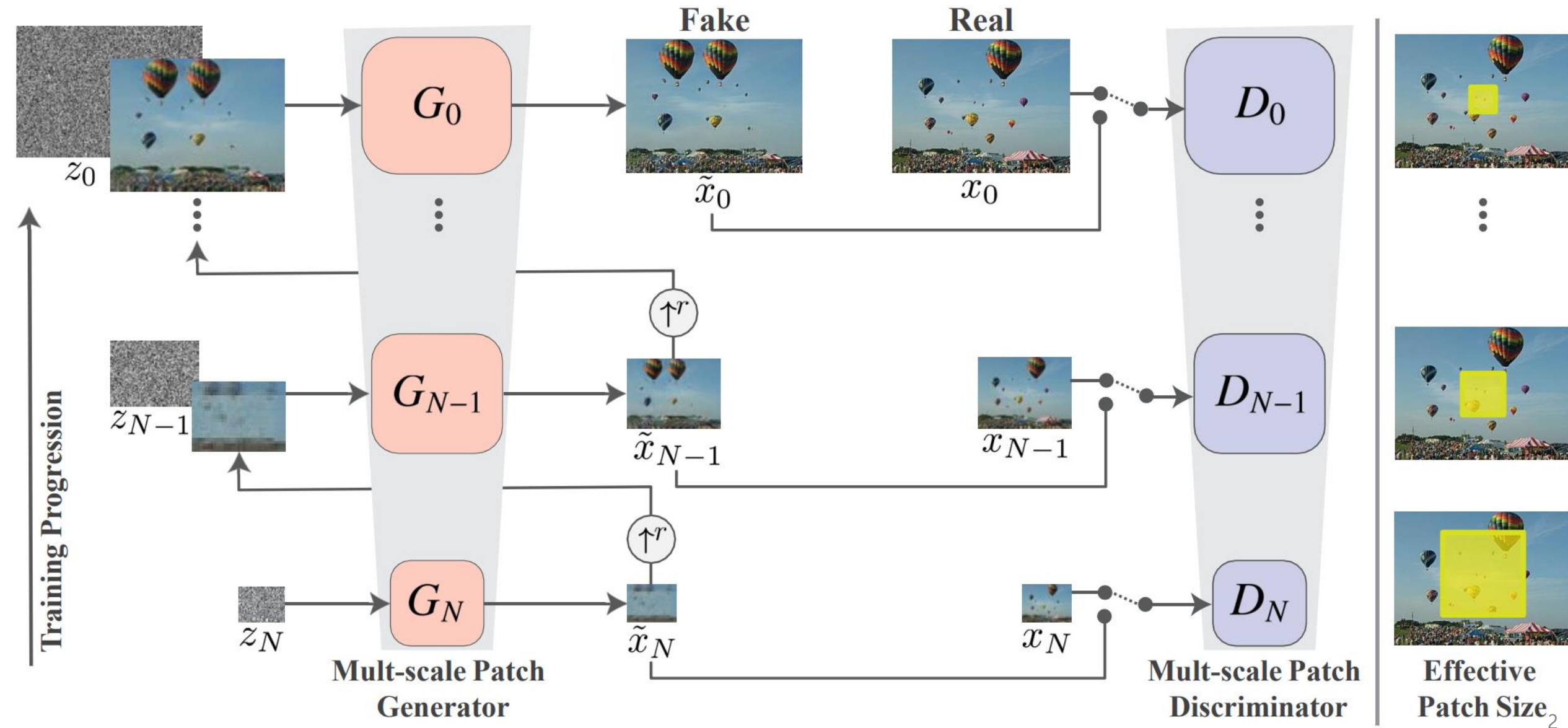


SinGAN: Learning a Generative Model from a Single Natural Image (CVPR 2019)

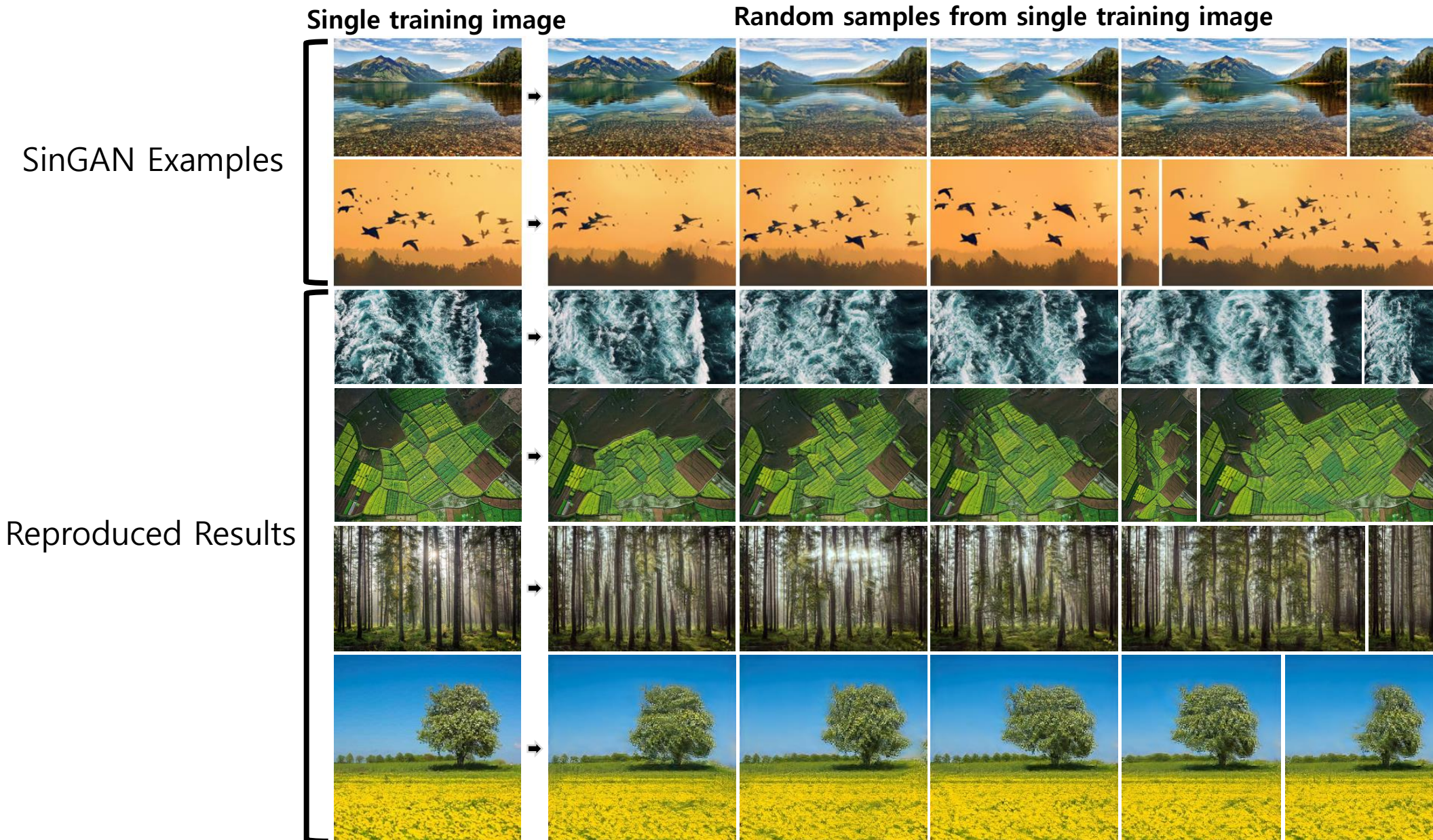
Sungwoo Son
Rehman Abdur
Rene Solzbacher



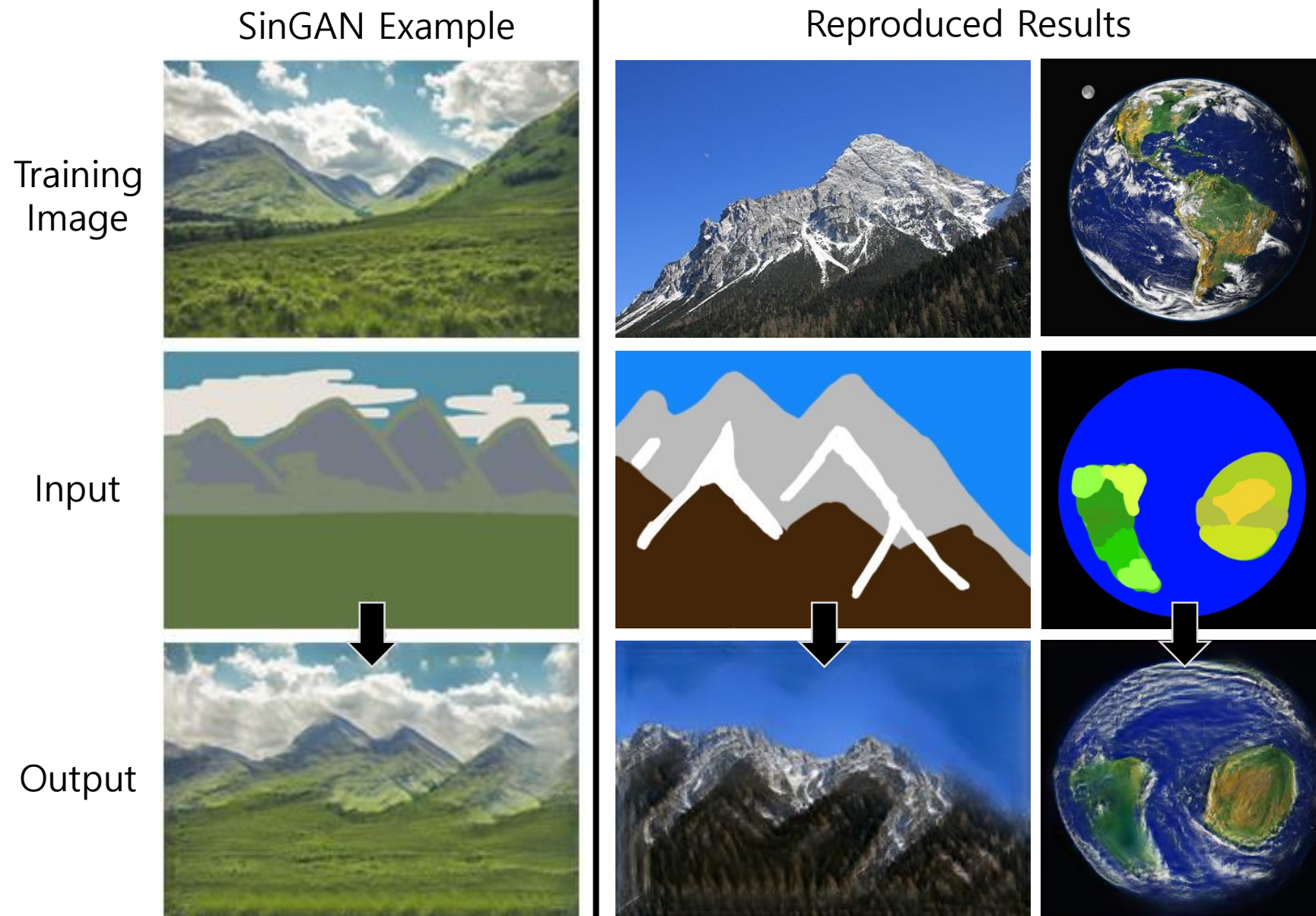
Overview of SinGAN



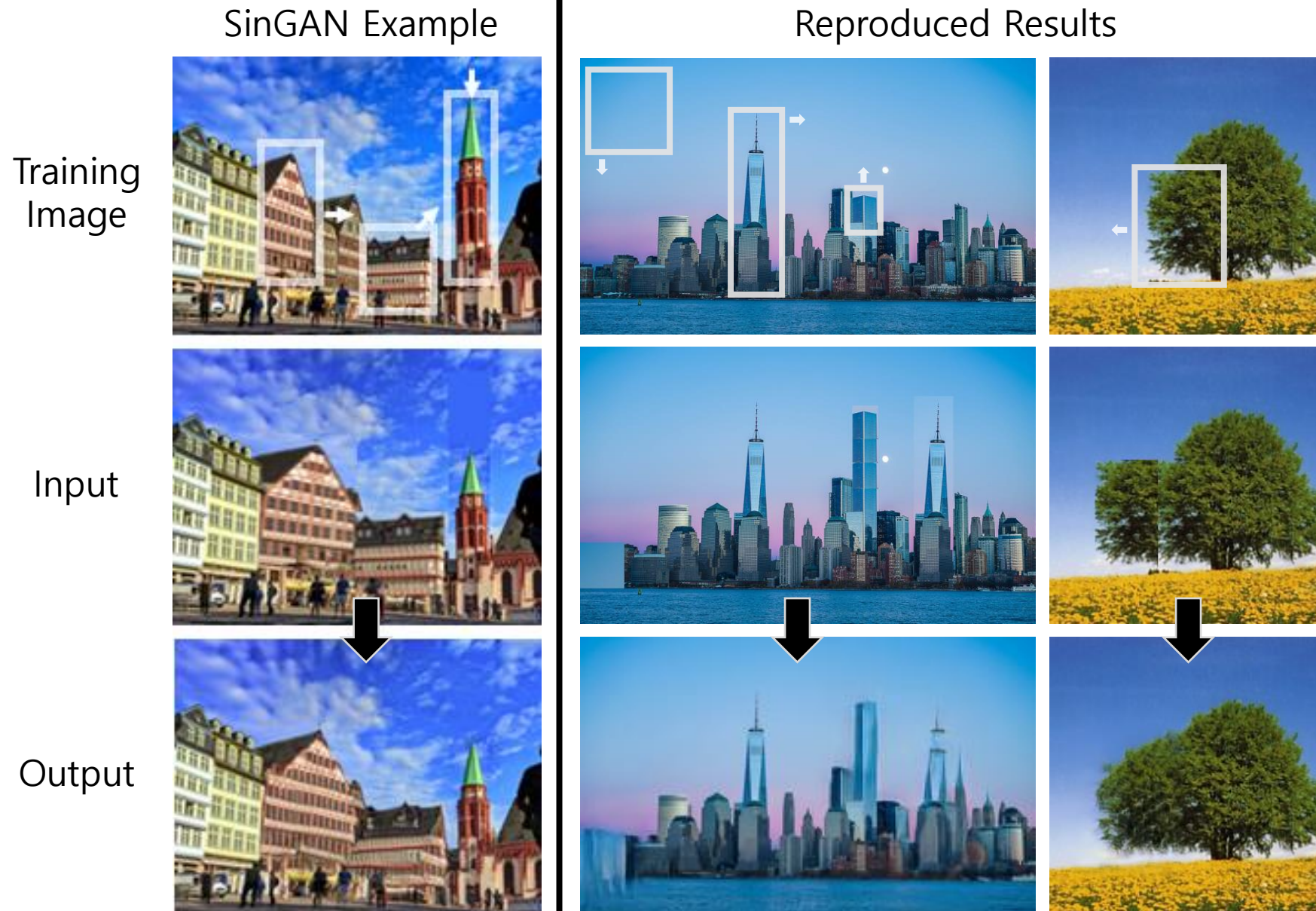
Reproduced Work (Random Samples)



Reproduced Work (Paint to Image)



Reproduced Work (Editing)



Reproduced Work (Super Resolution)

SinGAN Example

Training Image



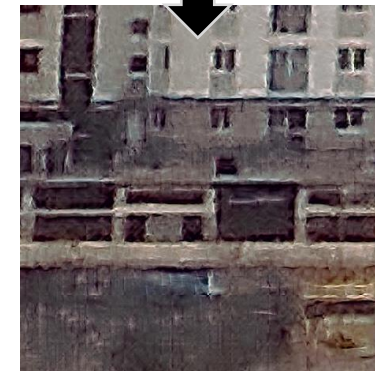
Input



Output



Reproduced Results



Reproduced Work (Animation)

SinGAN Example



Reproduced Results

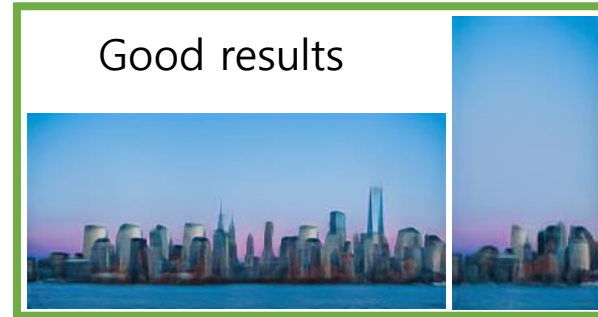


- Quality depends on subject
- Not all subjects are suitable

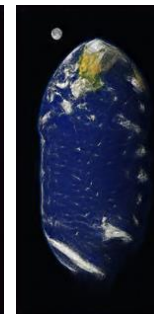
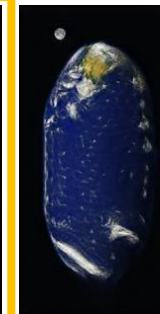
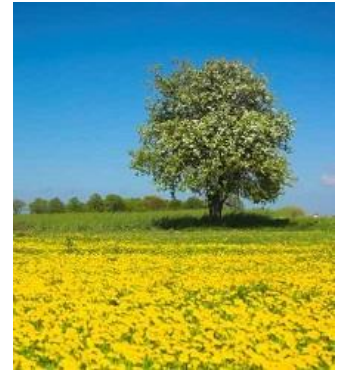


Limitations (Aspect Ratio)

- **Only some** aspect ratio changes work
- Quality is dependent on subject and image composition
- Some changes may look good but are illogical



Training Images



Limitations (Paint to Image)

Training Image



Output



Scale 1

Output

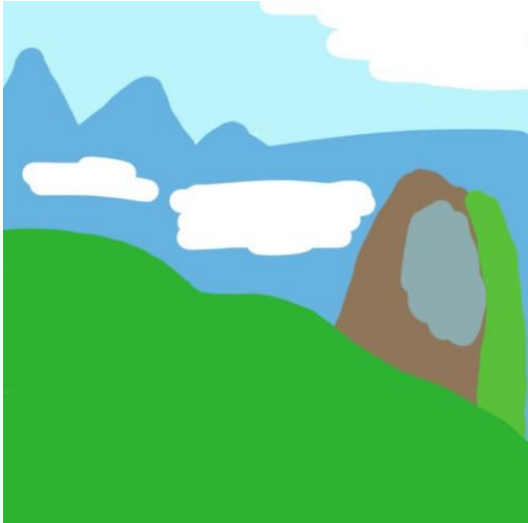


Scale 2

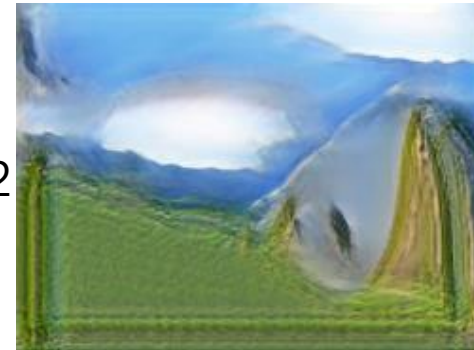
Training Image



Input



Scale 3

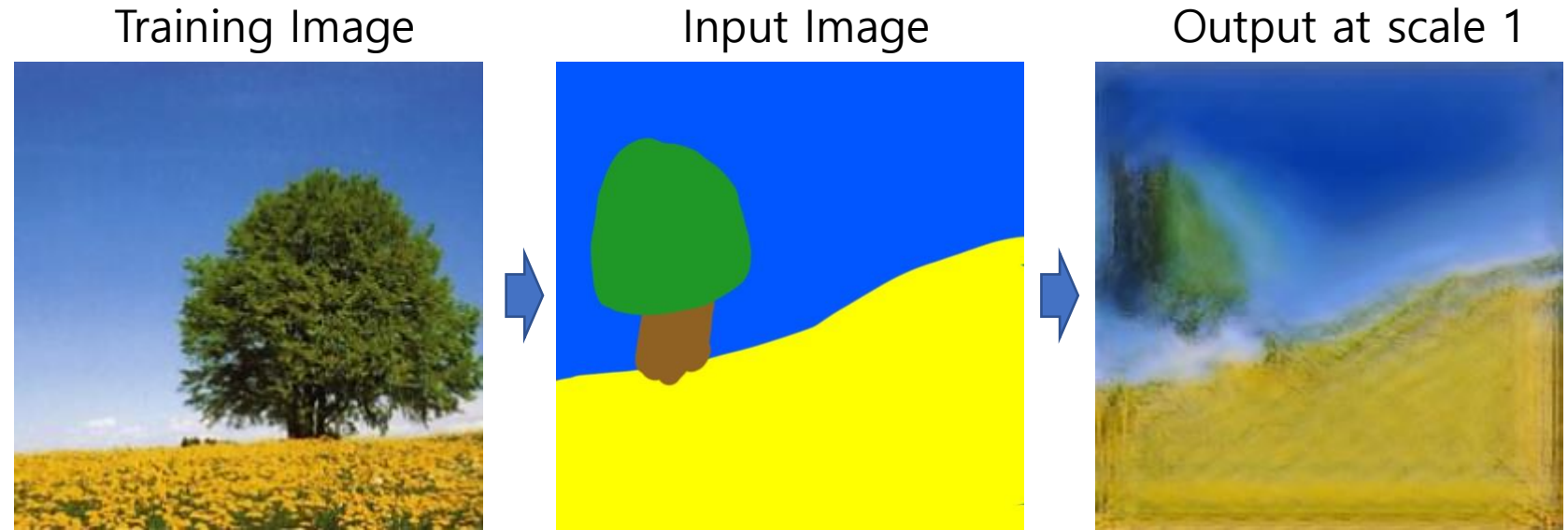


Input



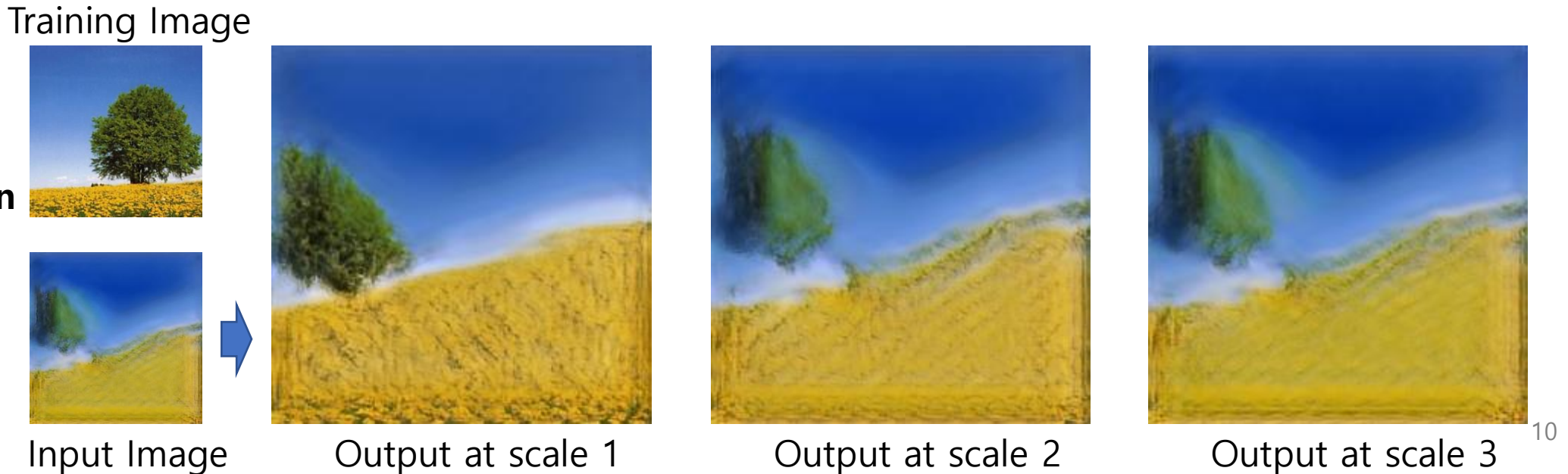
Paint to Image Improvement

First Generation



Second Generation

(Using output
as input image)



- x Limited to 250 pixels (longest side)
- x Small images are less useful

Generate high resolution output images with SinGAN

- Increase image dimensions
 - Keep comparable image quality
 - Keep acceptable training times (currently 40-60 min)
-
- Change scale distribution
 - Try image splicing
 - Change hyperparameter
 - Modify inner structure of generator

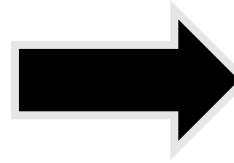
SinGAN Image Size

- SinGAN resizes training image to 250 pixels (longest side)
- Scales are based on resized image dimensions

```
> # 0 = {Tensor: (1, 3, 25, 40)} tensor
> # 1 = {Tensor: (1, 3, 32, 51)} tensor
> # 2 = {Tensor: (1, 3, 40, 64)} tensor
> # 3 = {Tensor: (1, 3, 50, 80)} tensor
> # 4 = {Tensor: (1, 3, 63, 100)} tensor
> # 5 = {Tensor: (1, 3, 79, 126)} tensor
> # 6 = {Tensor: (1, 3, 100, 158)} tensor
> # 7 = {Tensor: (1, 3, 125, 199)} tensor
> # 8 = {Tensor: (1, 3, 157, 250)} tensor
```

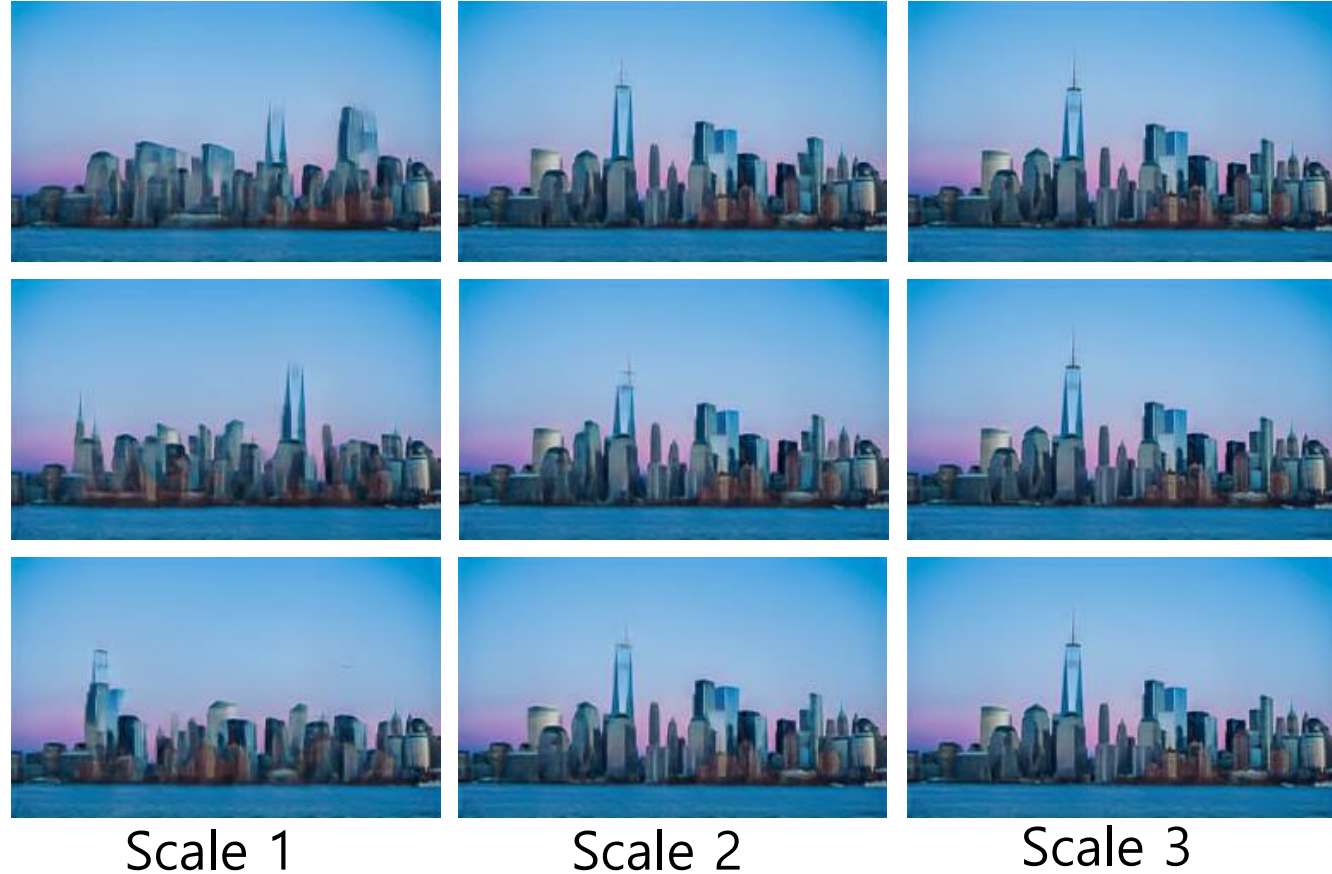


Input Image (585 x 366)



Resized Image (250 x 157)

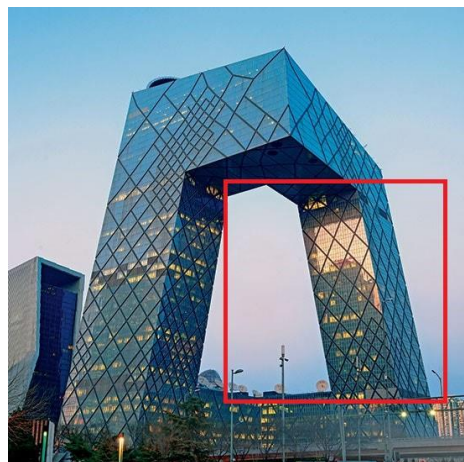
- Lower scales (0,1) produce **major structural changes**
- Higher scales produce **fine detail**
- Scales are linearly implemented
- At each scale, the image is upscaled the same amount
- Anything beyond scale 2/3 produces too little change



Investigate different rescaling strategy that focuses on coarse scales

Author's High-Resolution Proposal

1. Train model to generate random samples
 2. Train super resolution model to up-sample selected image
- ✓ SinGAN super resolution restores much detail
- x Super resolution becomes problematic for large upscaling



Original crop



Down sampled (250px)



SR from 250px crop

High-Resolution (Image Splicing)

Structures largely remain the same towards the image edges



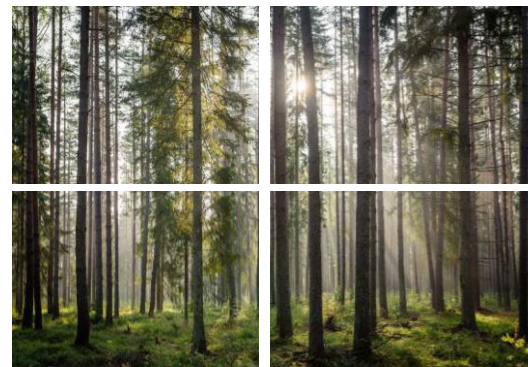
1. Split image into sections
2. Train model for each section
3. Generate random sample for each section
4. Train high resolution model
5. Splice sections together



Single
generation



≠



Individual
generation



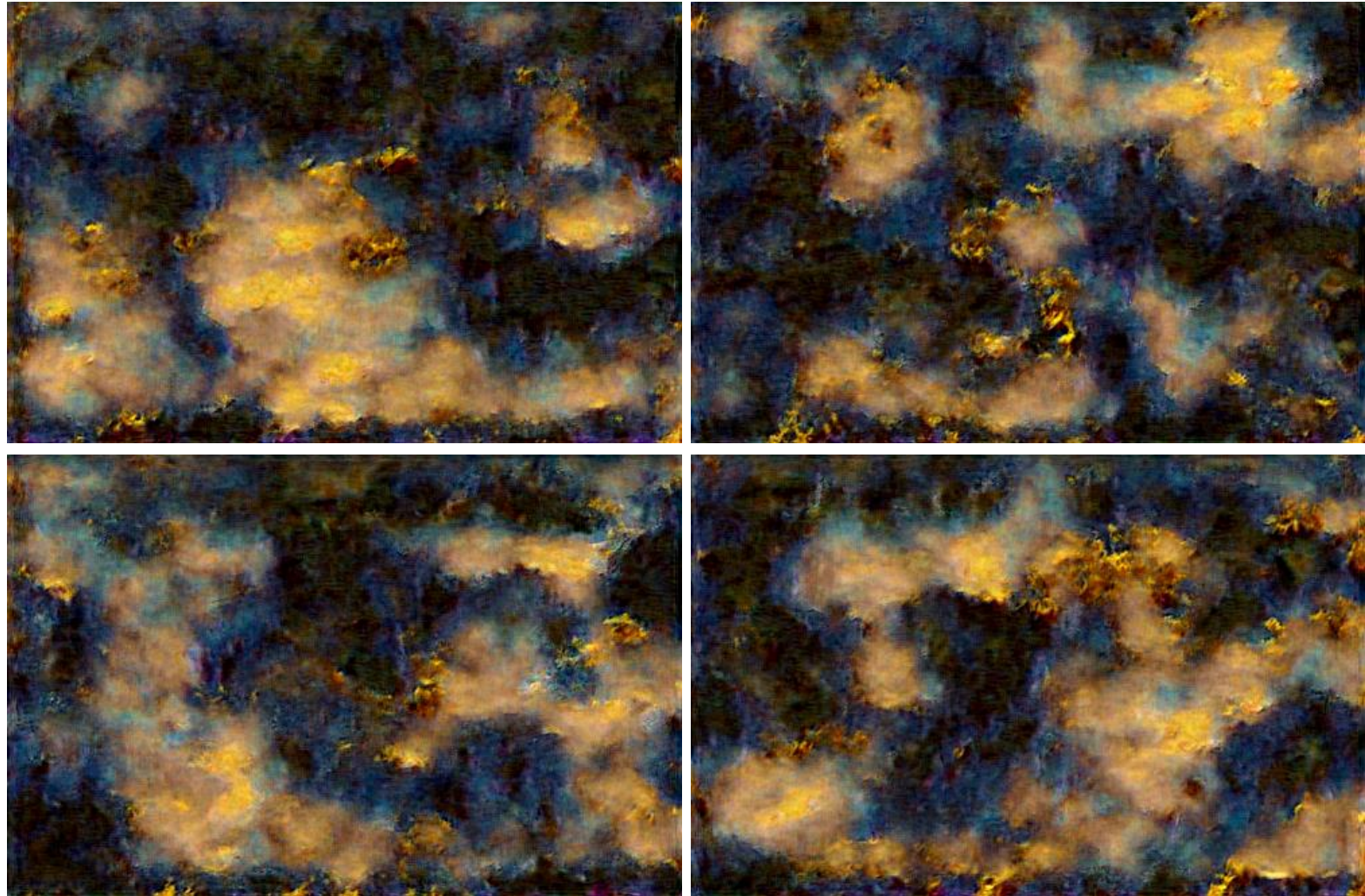
?

- Values to fine-tune SinGAN
 - Receptive field size default: 11x11px
 - Rescaling resolution default: 250px
 - Conv kernels default: 32
 - Learning rate default: 0.0005
 - Epochs per scale default: 2000
- Option to change the behaviour and performance of SinGAN
- SinGAN is optimized for images of **250px size**
- We want to optimize it for images of **500px size**

High-Resolution (Hyperparameter Changes)



741 x 486

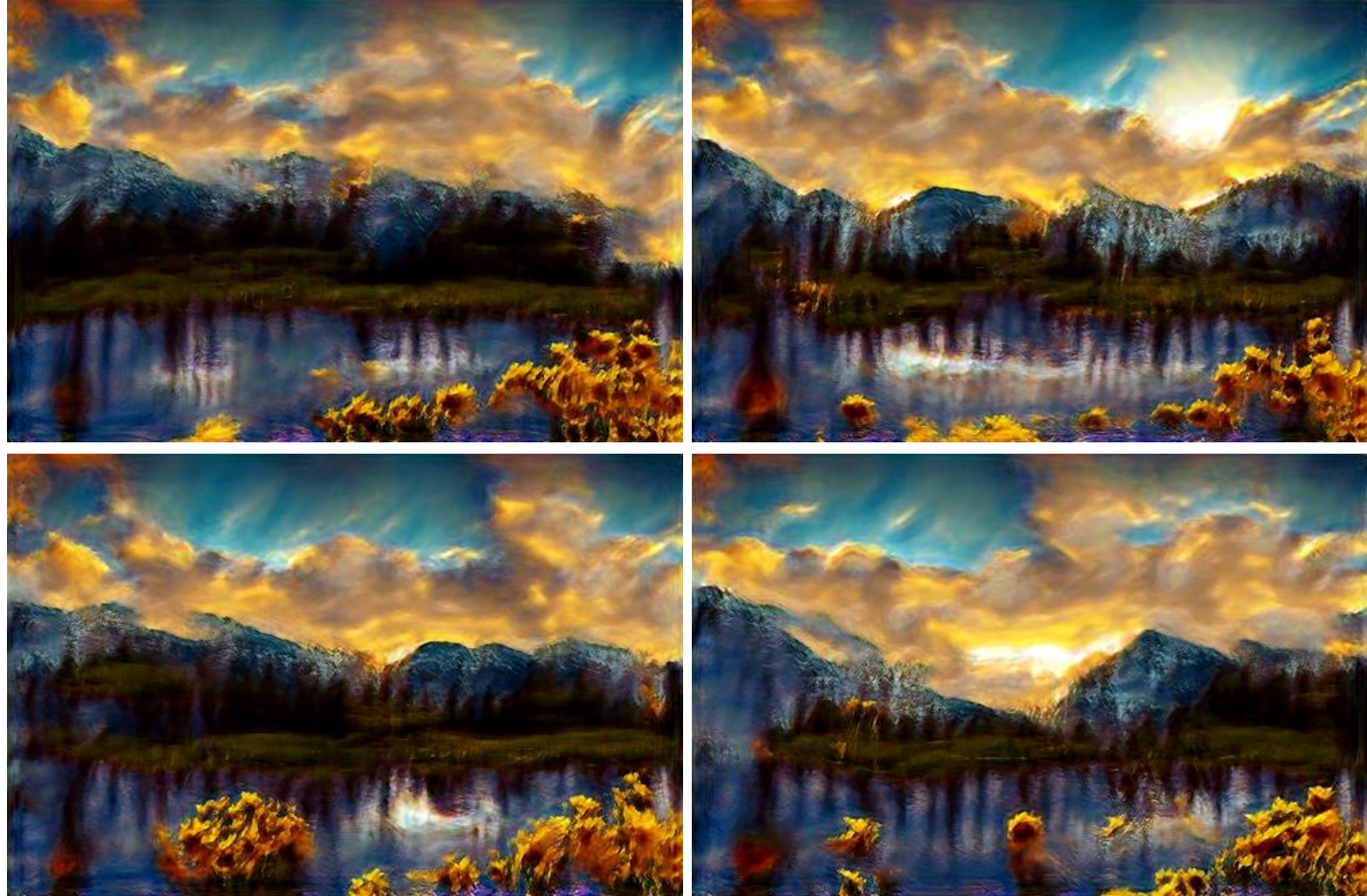


400 x 263

High-Resolution (Hyperparameter Changes)



741 x 486



400 x 263

High-Resolution (Hyperparameter Changes)



741 x 486



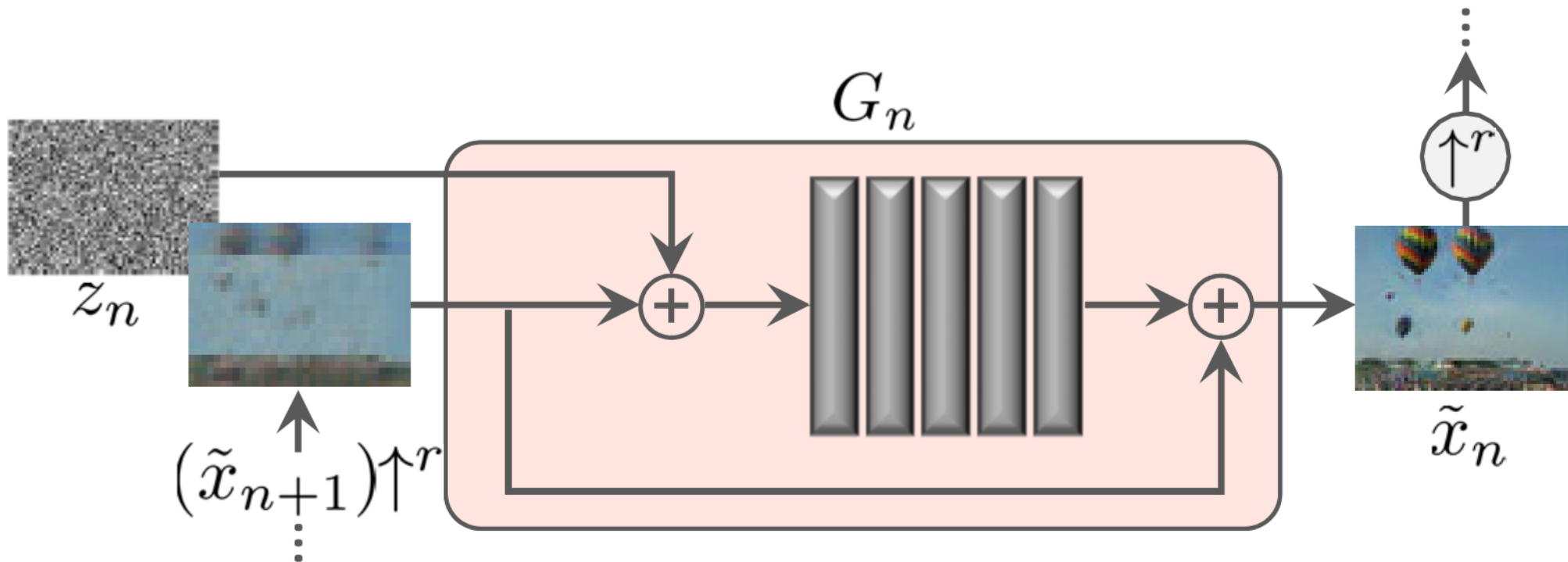
400 x 263

- Hyperparameters that can increase quality:
 - Padding size
 - Scaling factor for up/down sampling images
 - Learning rate
 - Reconstruction loss weight (α)

Future Work

- Automate hyperparameters to adjust to image size automatically
 - SinGAN is only optimized for 250px images
 - Only basic starting parameters need to be set by user, the rest is automatically scaled for the desired image resolution

Modify Generator Structure



- Up-sampling in the structure and outside the structure
- Use Resnet for conformity to the original training image



Thank you