

Diffusion Models for Visual Content Creation



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Part 2: Guidance and Conditioning



https://geometry.cs.ucl.ac.uk/courses/diffusion4ContentCreation_sigg24/

Presentation Schedule

Introduction to Diffusion Models

Guidance and Conditioning

Attention

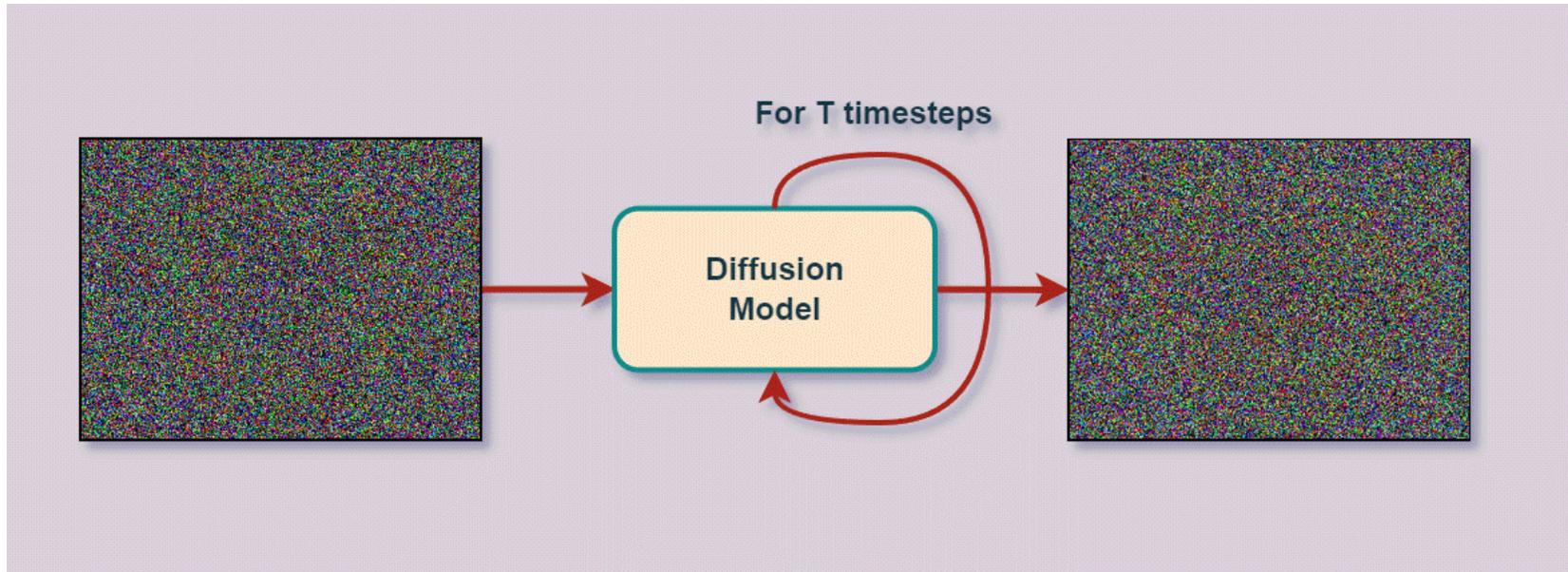
Break

Personalization and Editing

Beyond RGB Images

Diffusion Models for 3D Generation

Diffusion Models: Unconditional Sampling



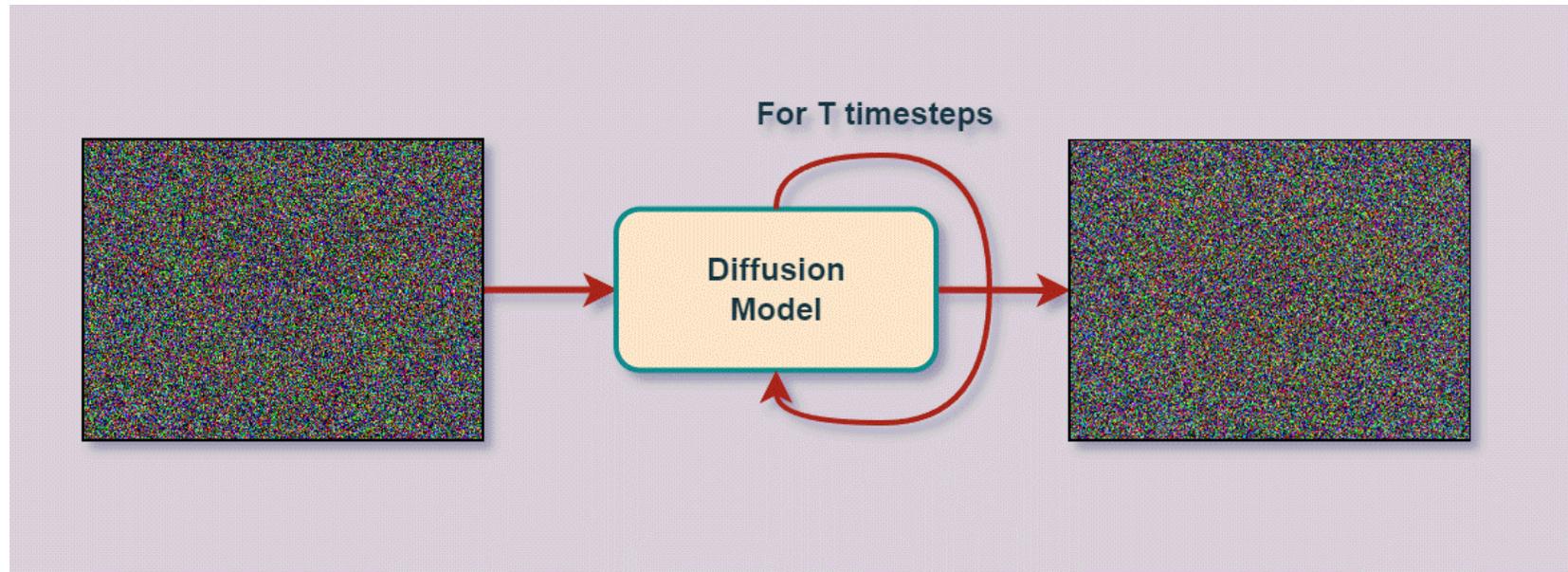
$$p(x) \approx \nabla_x \log p(x)$$

└─┬─> score function

<https://learnopencv.com/image-generation-using-diffusion-models/>

<https://sander.ai/2022/05/26/guidance.html>

Diffusion Models: Unconditional Sampling



$$p(x|y) \approx \nabla_x \log p(x|y) = \nabla_x \log p(y|x) + \nabla_x \log p(x)$$

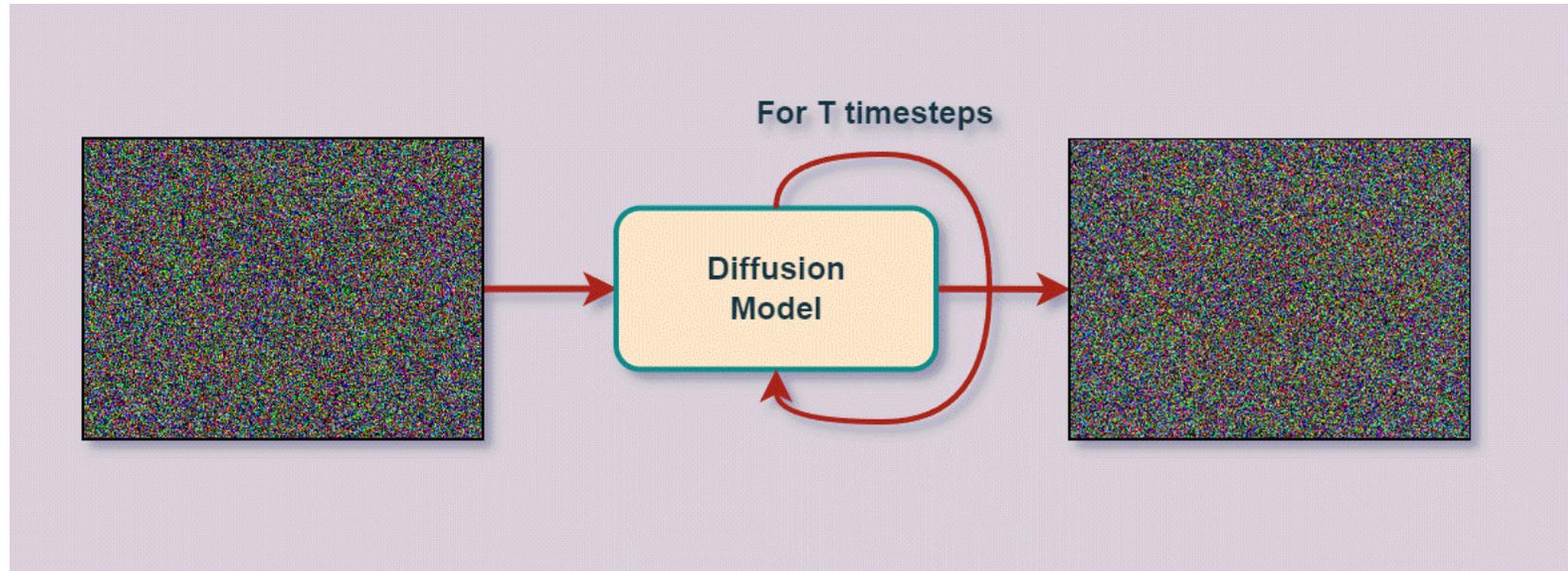
condition

unconditional score function

conditioning term

<https://learnopencv.com/image-generation-using-diffusion-models/>
<https://sander.ai/2022/05/26/guidance.html>

Diffusion Models: Conditional Sampling



$$p(x|y) \approx \nabla_x \log p(x|y) = \nabla_x \log p(y|x) + \nabla_x \log p(x)$$



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Classifier Guidance

$$\nabla_x \log p(x|y) = \gamma \nabla_x \log p(y|x) + \nabla_x \log p(x)$$

guidance scale



$\gamma = 1$

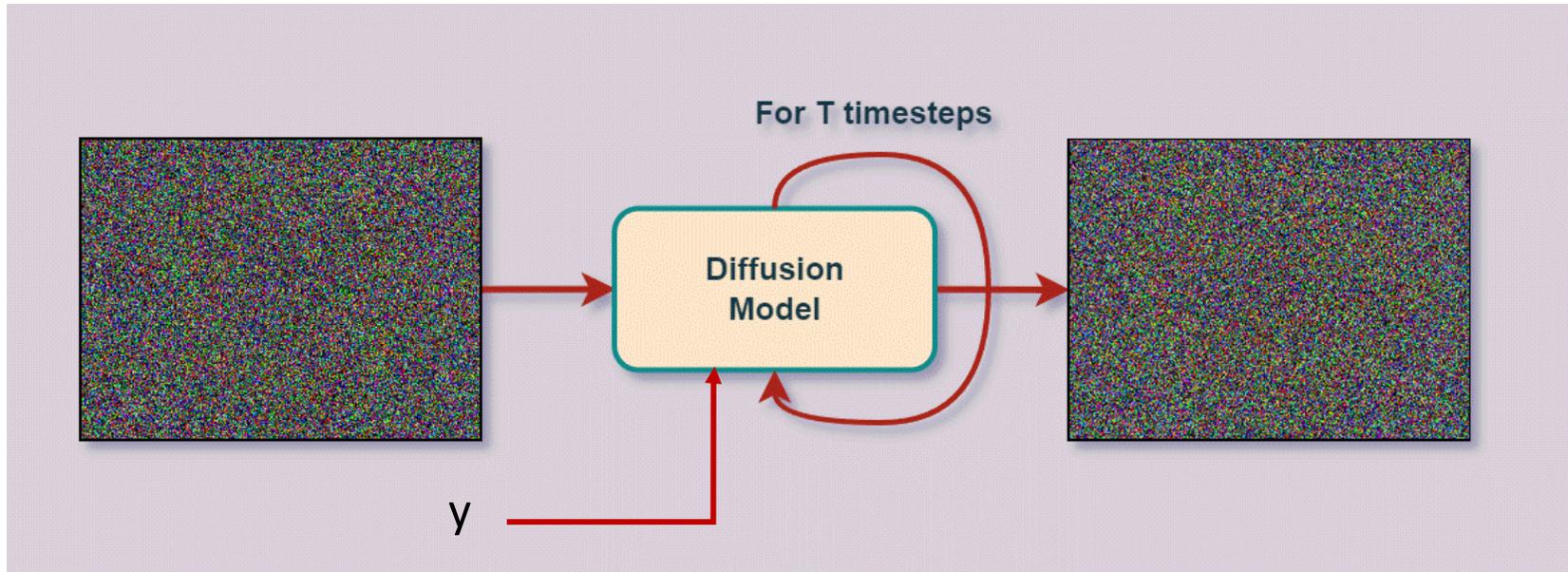


$\gamma = 10$

Classifier Guidance

- ✓ Effectively turns an unconditional diffusion model to a conditional one via a differentiable classifier
- ✗ Requires the classifier to cope with inputs with high noise levels (e.g., retraining the classifier)
- ✗ Most information in x may not be relevant to the classifier so the gradient can yield arbitrary directions

Classifier-free Guidance (CFG)



Dropout strategy: 10-20% of the time y is replaced with a special input (e.g., a null text in case of text conditioning)

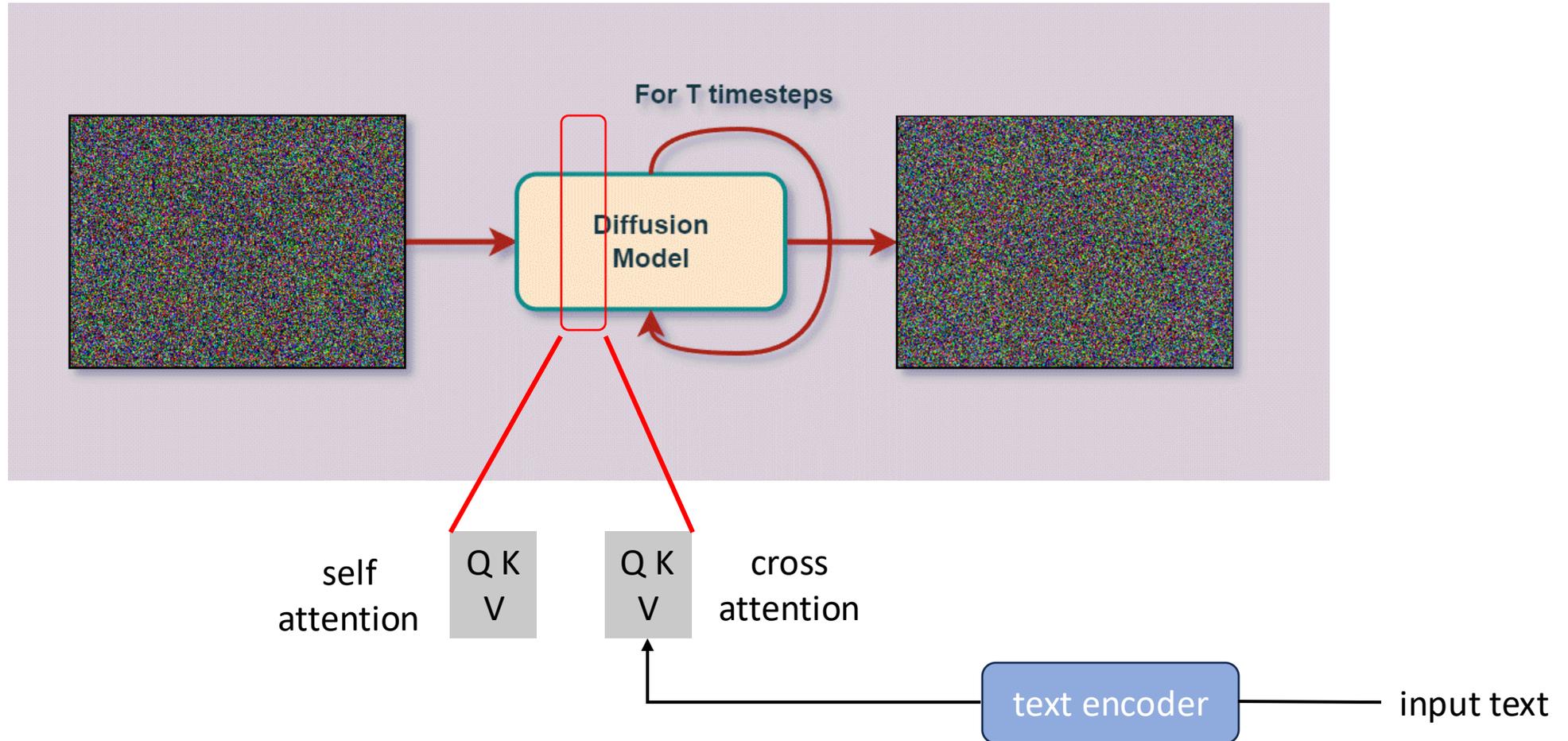
<https://learnopencv.com/image-generation-using-diffusion-models/>

<https://sander.ai/2022/05/26/guidance.html>

Classifier-free Guidance

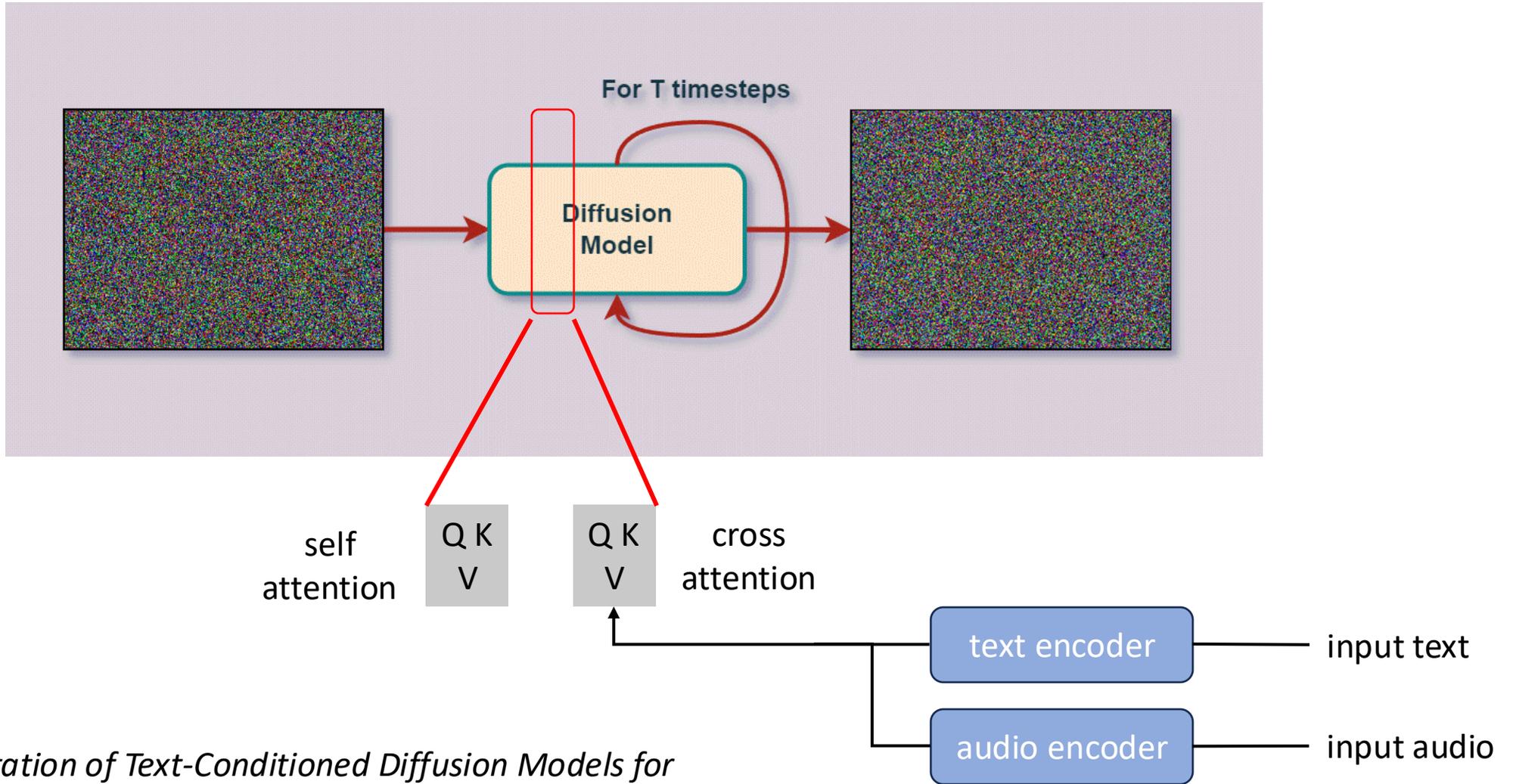
- ✓ Improves adherence to the conditioning signal
- ✗ Less diversity (may not be an issue for conditional generation)

How are Conditioning Signals Provided?



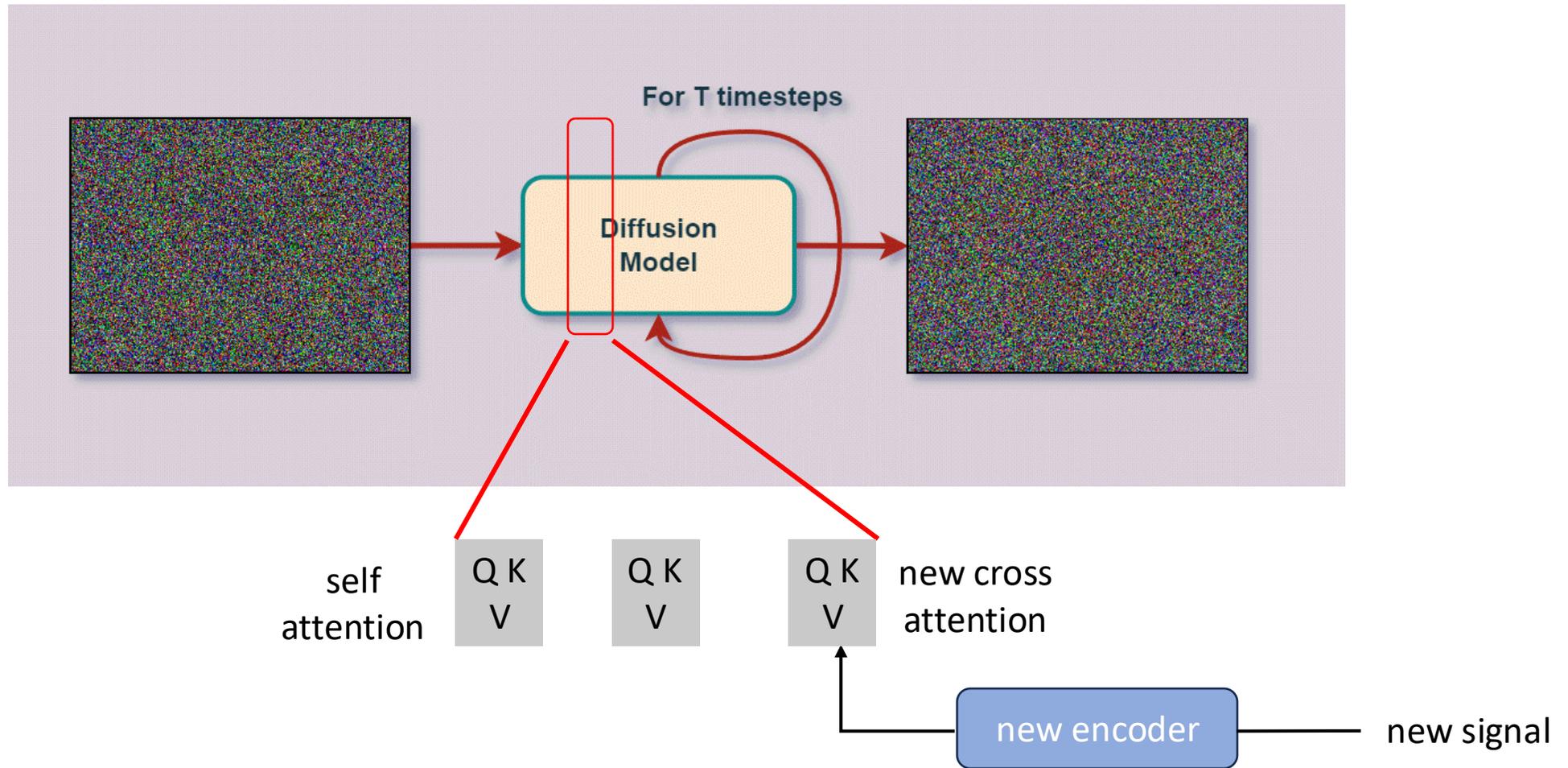
<https://learnopencv.com/image-generation-using-diffusion-models/>

Different Conditioning Signals

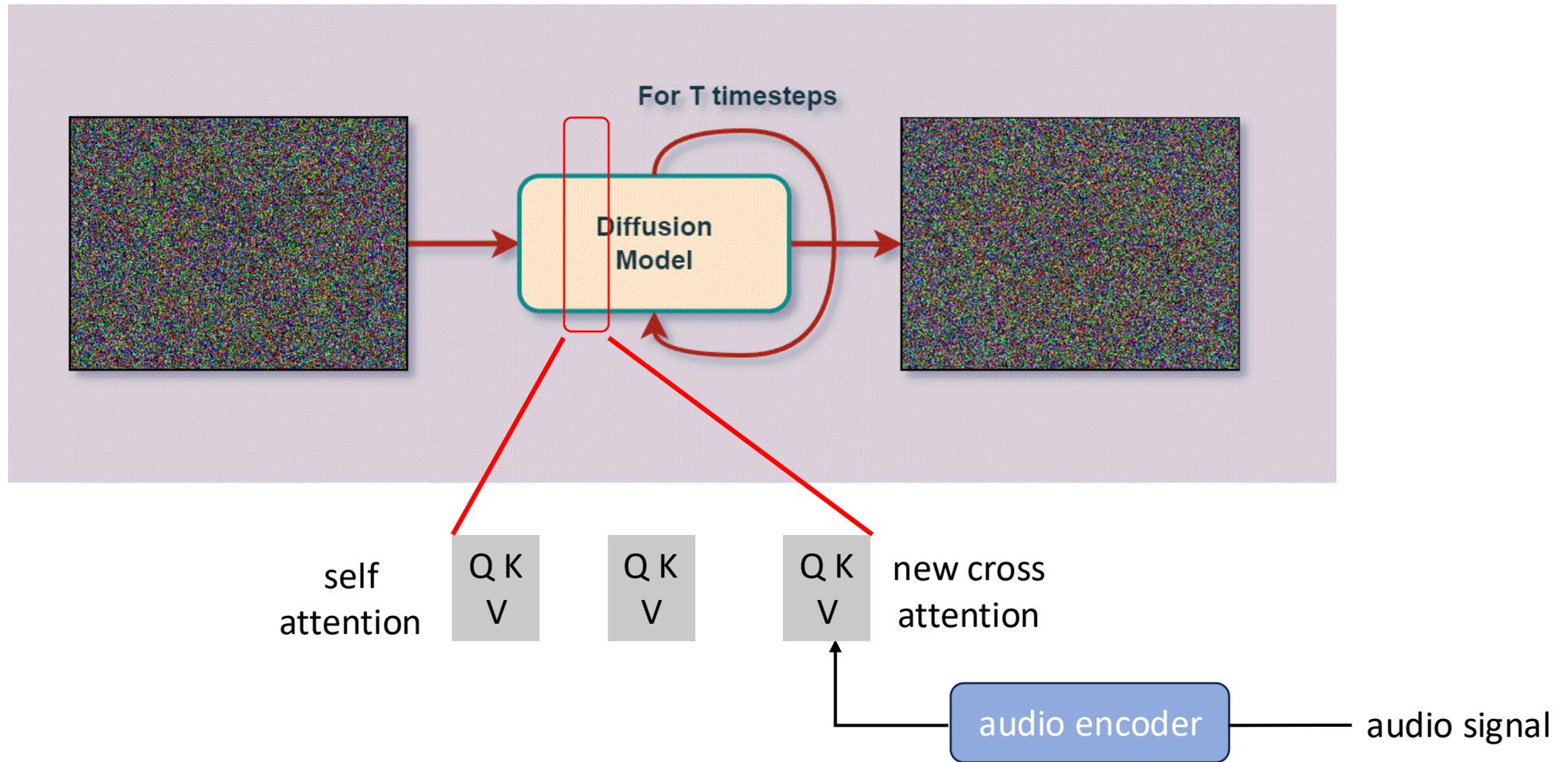


AudioToken: Adaptation of Text-Conditioned Diffusion Models for Audio-to-Image Generation, Yariv et al. 2023

Different Conditioning Signals



Different Conditioning Signals



under review

Audio-conditioned Generation



under review

SIGGRAPH 2024 Course



Guidance and Conditioning

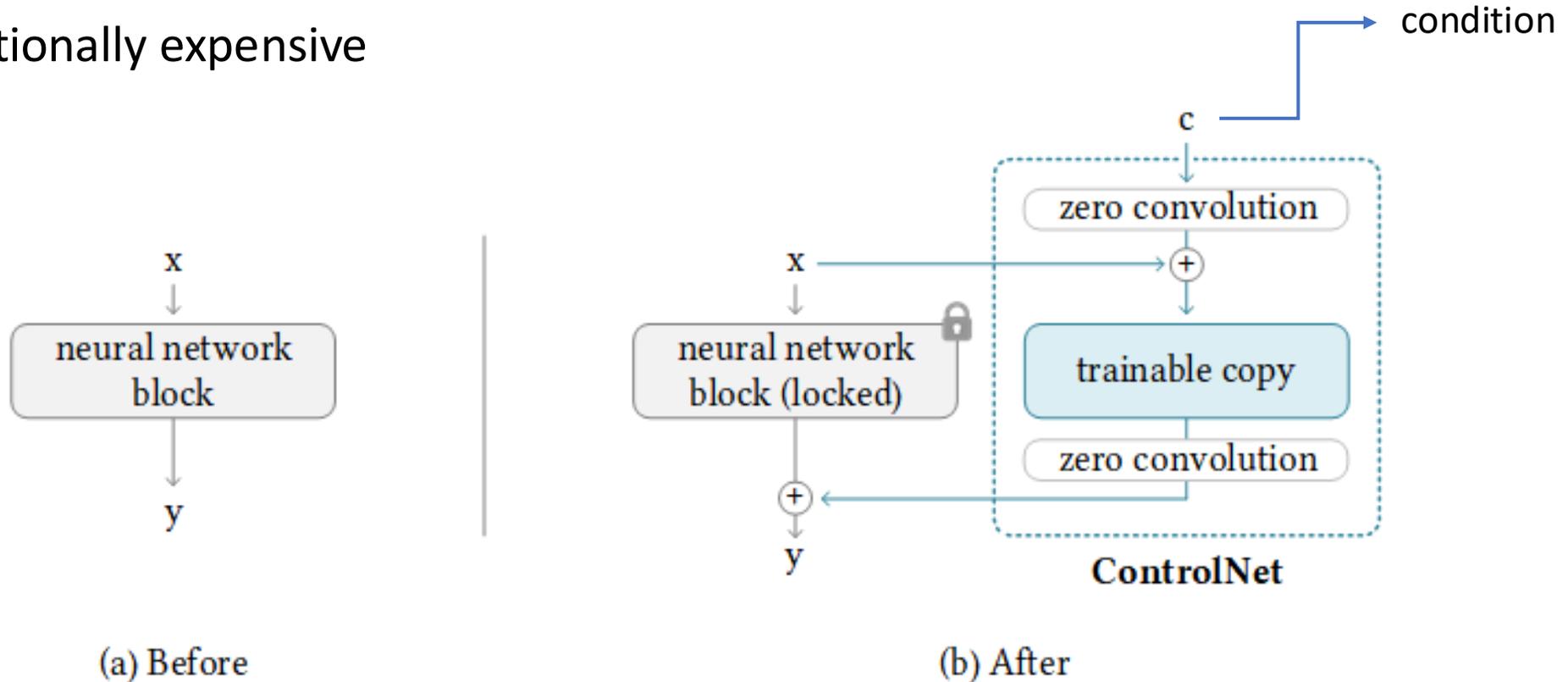


Diffusion Models in Visual Computing

Other Conditioning Methods: ControlNet

✓ Can be trained on smaller amount of data.

✗ Computationally expensive



Adding Conditional Control to Text-to-Image Diffusion Models, Zhang et al. 2023

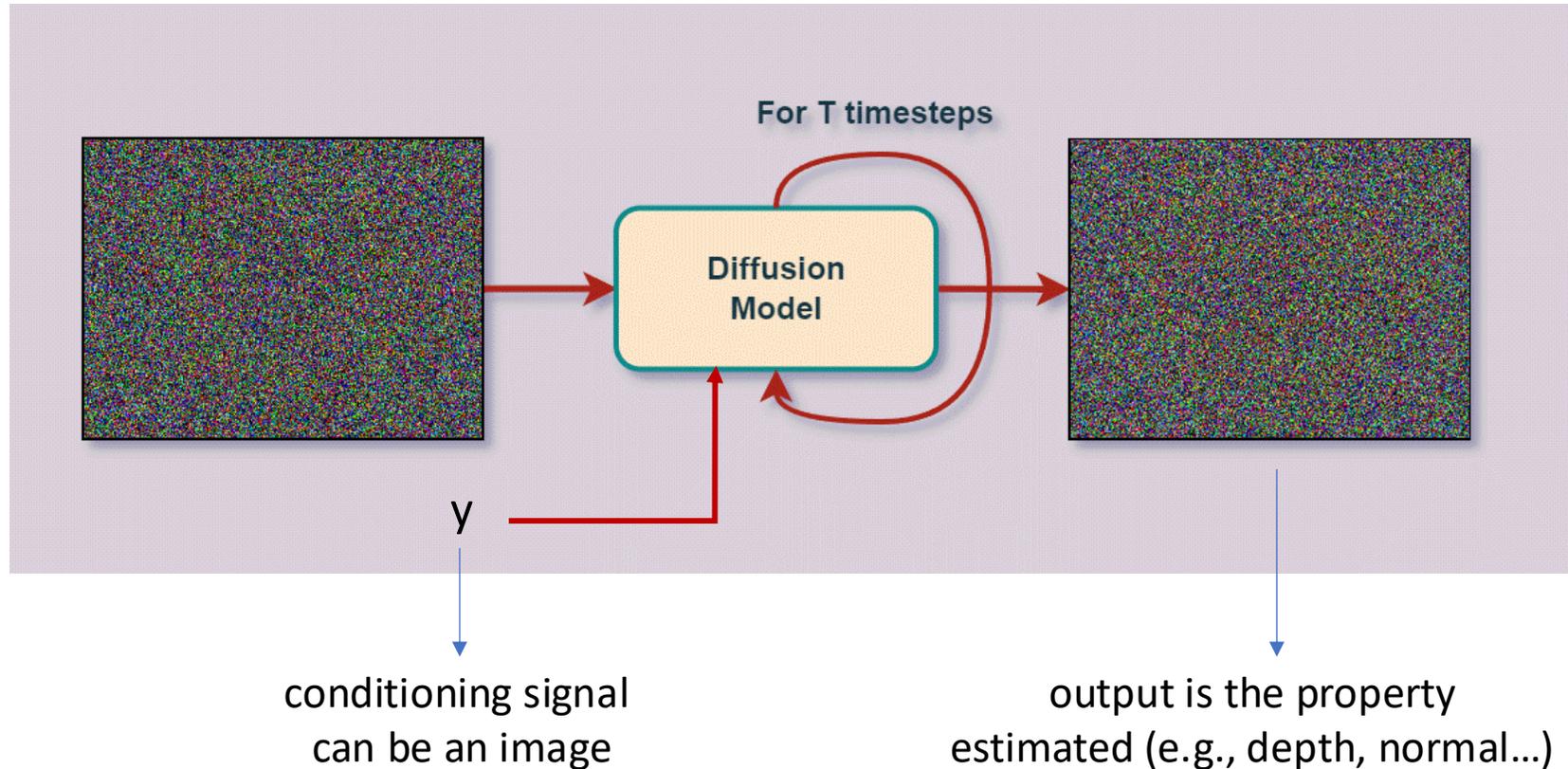
Other Conditioning Methods

- Concatenate spatial conditioning signal with the input noise

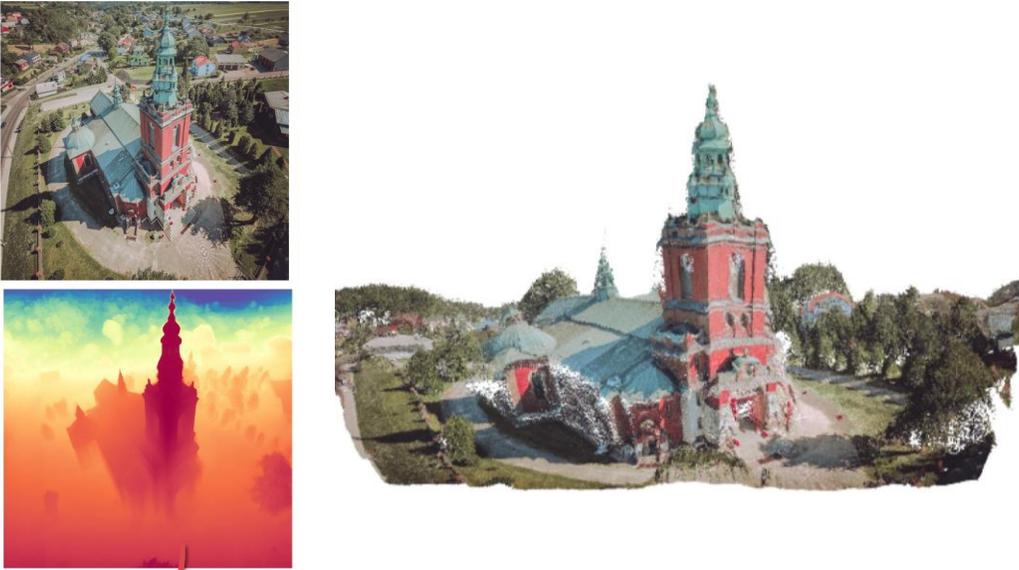


Stable Diffusion v2 Depth-to-Image model

Conditioning to Convert Generators to Regressors



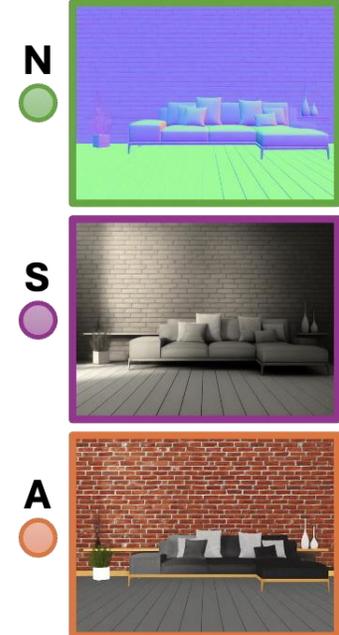
Generation -> Analysis



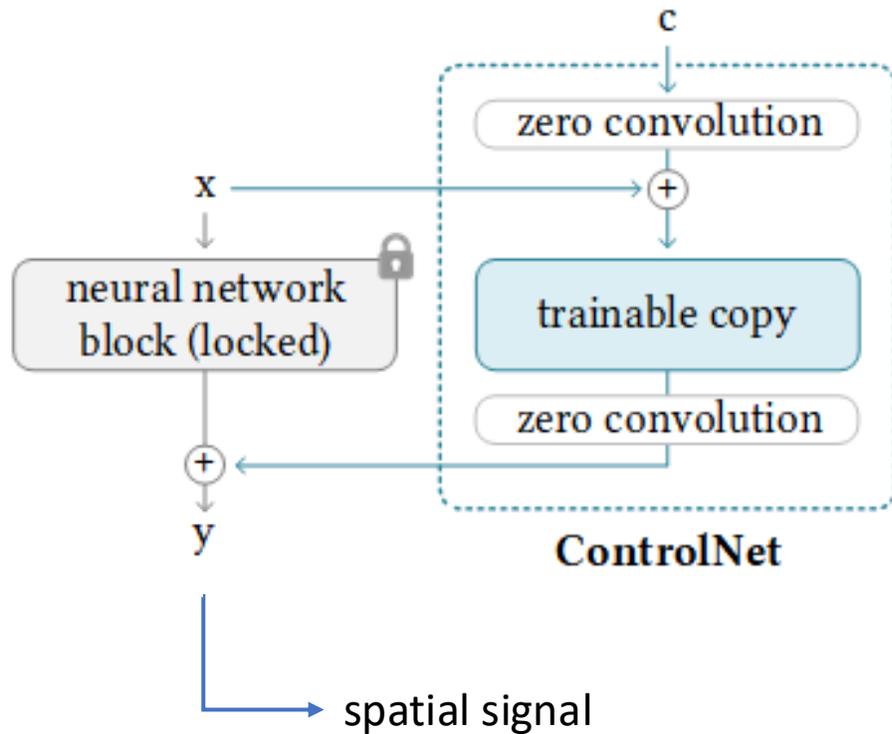
Marigold Depth, CVPR 2024
(Best paper award candidate)



RGB-X, Intrinsic Diffusion
SIGGRAPH 2024

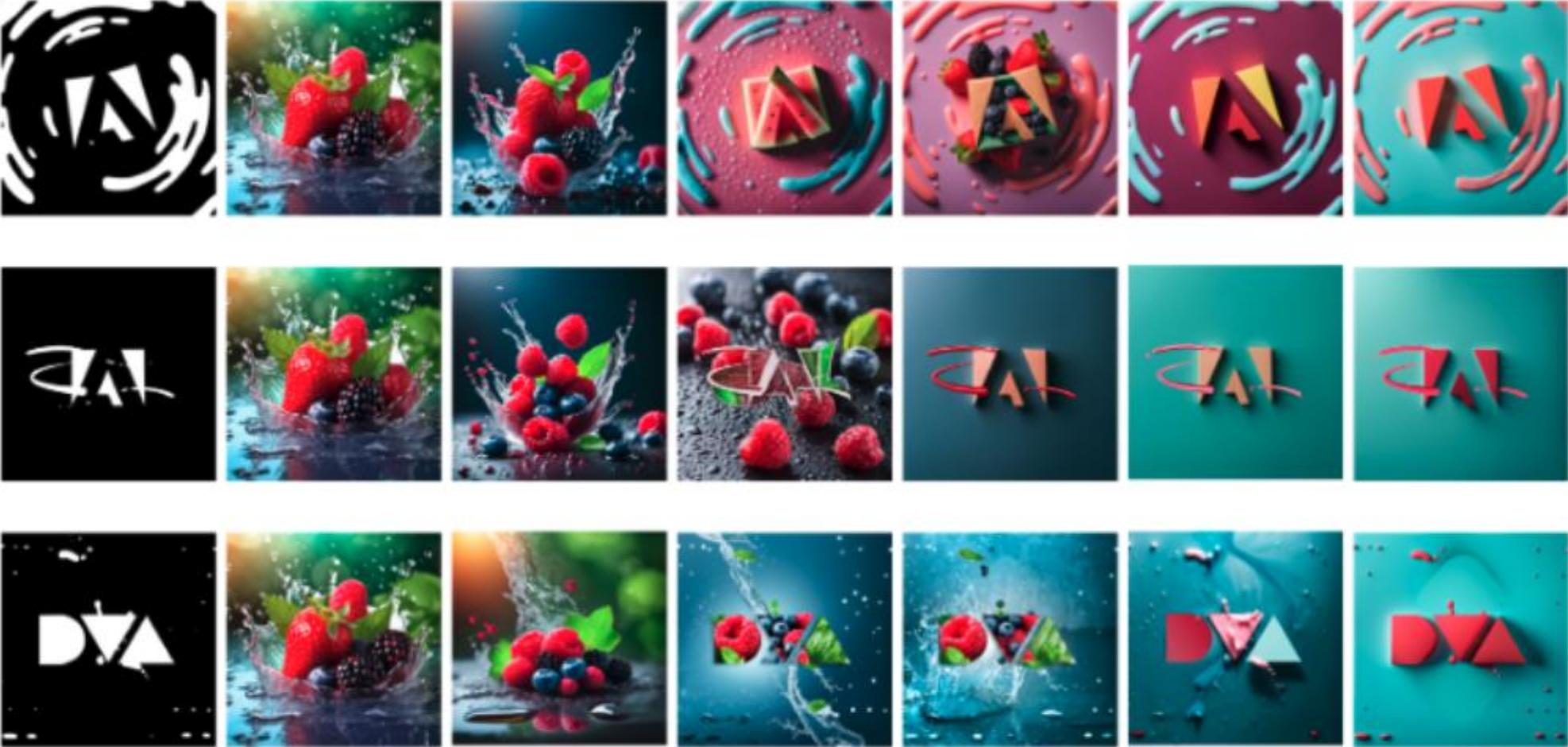


Adherence to the Conditioning Signal



Weighting can be applied!

Adherence to the Conditioning Signal



control

Weight=0.0

0.1

0.2

0.5

1.0

2.0

Adherence to the Conditioning Signal



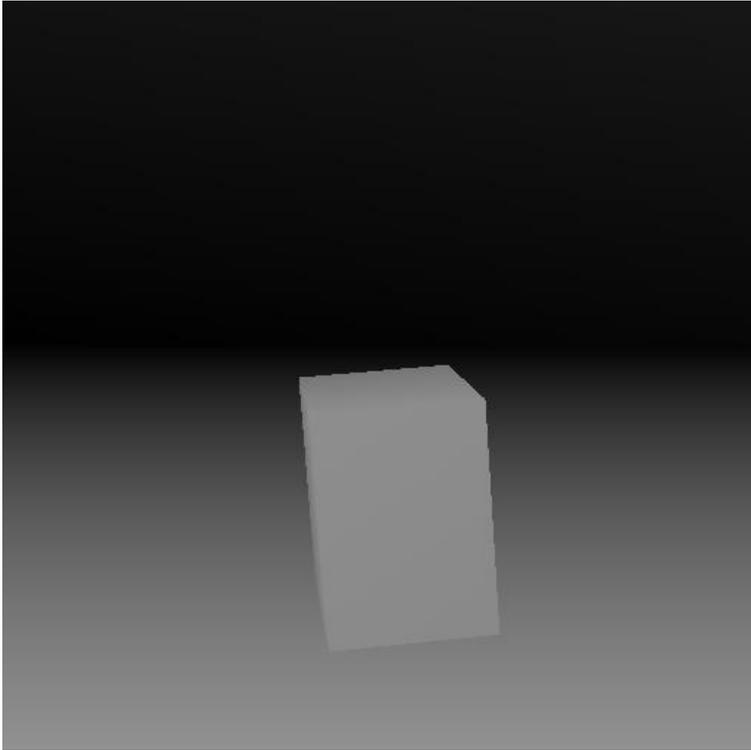
control

spatial weight

w/ spatial weight

w/ global weight

Adherence to the Conditioning Signal



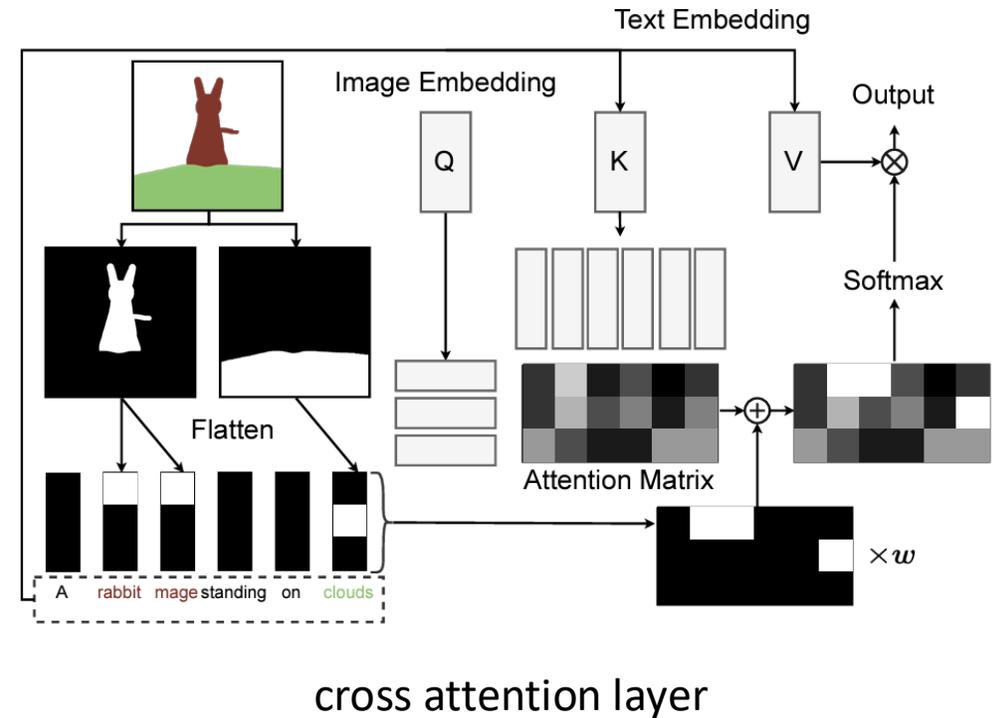
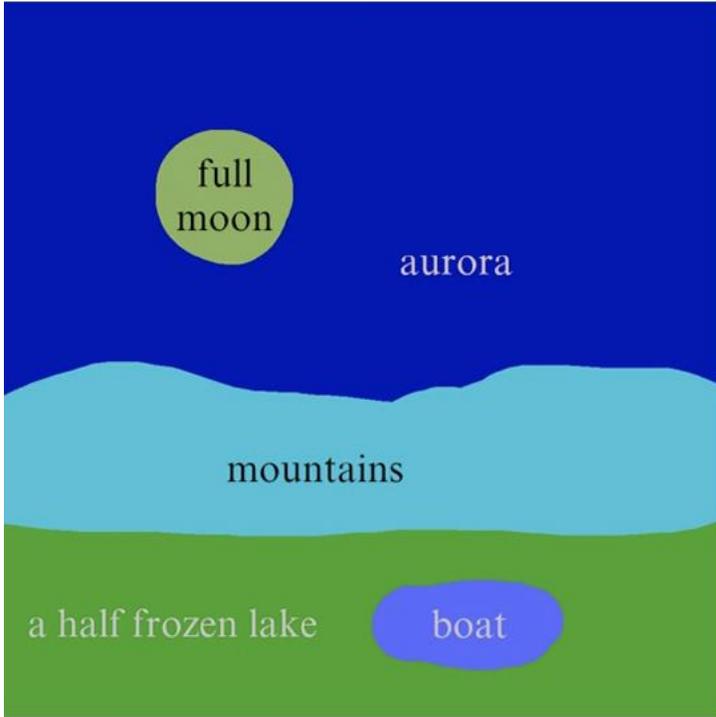
original ControlNet



LooseControl

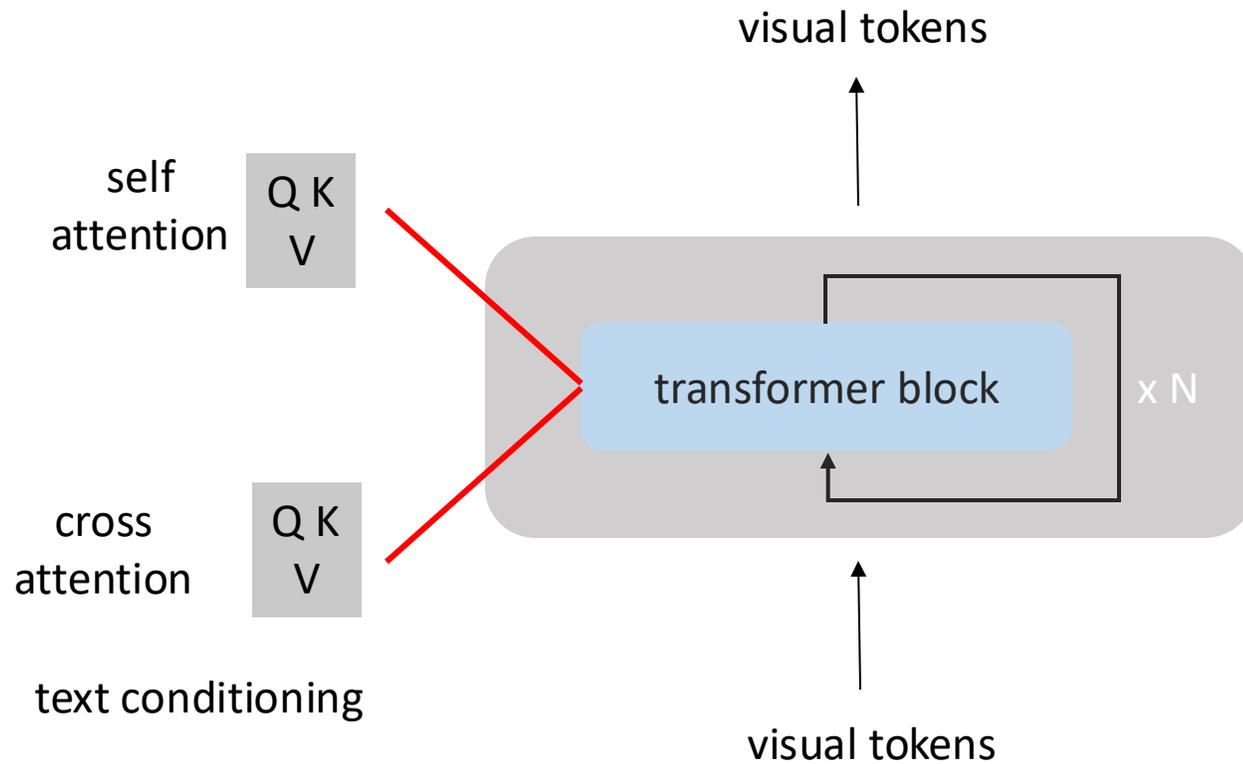
LooseControl: Lifting ControlNet for Generalized Depth Conditioning, Bhat et al. 2023

Regional Control of Conditioning Signal

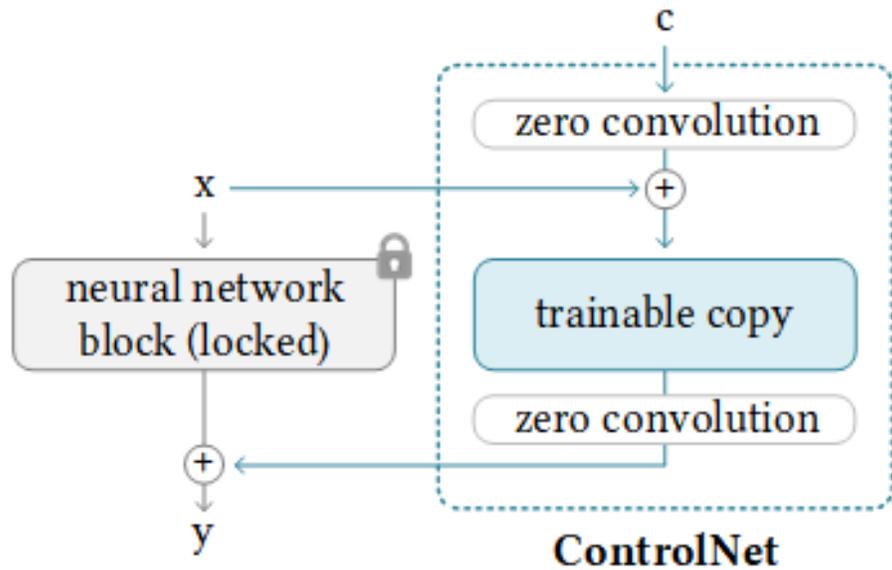


eDiff-I: Text-to-Image Diffusion Models with an Ensemble of Expert Denoisers, Balaji et al. 2023

How About Diffusion Transformers?



ControlNet for DiT



farm workers in the style of photorealistic detailing

PIXART- δ : Fast and Controllable Image Generation with Latent Consistency Models

Conclusion

- Various options for providing conditioning signals
- Interesting research questions:
 - Fine-grained control on adherence to conditioning signals
 - Regional control over conditioning signals
 - Multi-modal generation

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