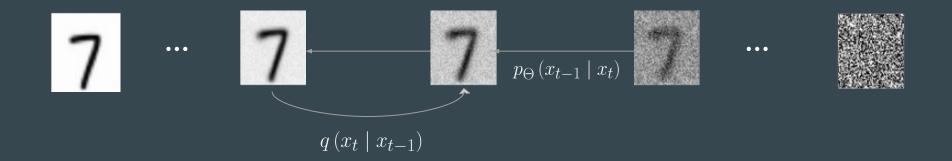
# Diffusion Model

•••

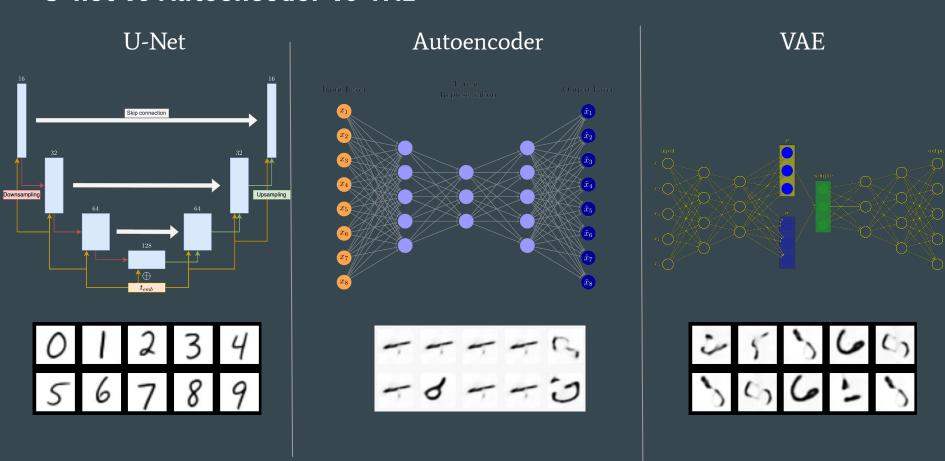
Kayo Christian, Vallaeys Théophane, Wemaere Maximilien

# Working principle of the diffusion model

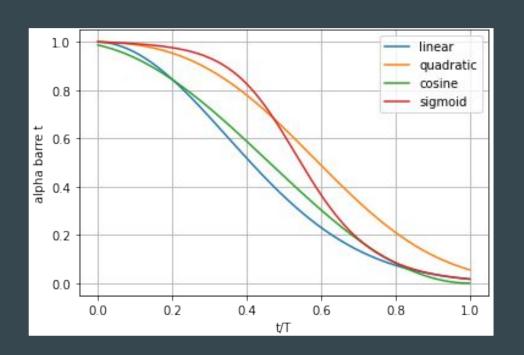
- Inspired by thermodynamics
- Based on learning how to noise to be able to denoise
- The concept is also borrowed from Markov Chains
- Originally proposed by Solh-Dickstein in 2015



## **U-net vs Autoencoder vs VAE**



### **Different schedulers**



#### FID:

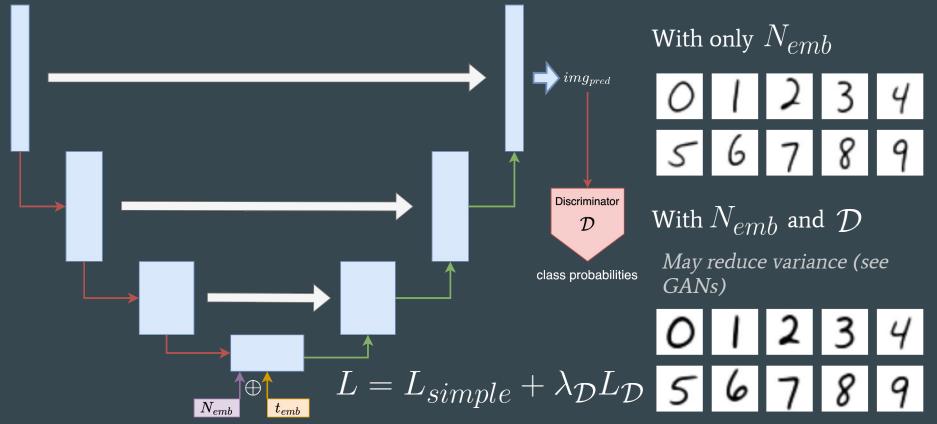
Linear: 41.43

Quadratic: 85.88

Cosine: 215.80

Sigmoid: 95.74

# **Embedding of the number to generate / Discriminator**

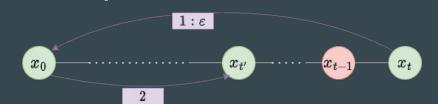


# Faster image sampling

#### **DDIM**

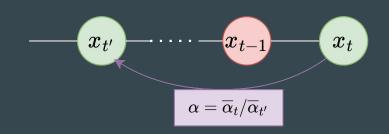
Idea : instead of sampling using  $q(x_{t-1}|x_t)$ 

we will predict  $x_0$  and arepsilon ( $q(x_0|x_t)$ ) and then  $q(x_{t'}|x_0,arepsilon)$ 

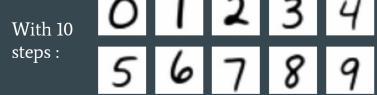


#### **Reducing T**

Training with  $\overline{T_{large}}$ Only predicting  $T_{small}$  steps







## Adding variance to the sampling process

$$\mu_{t-1} = \mu_{pred}(\mu_t) + \sigma_t \varepsilon_t'$$

$$\sigma_t = \sqrt{ ilde{eta}_t}$$
 |  $O$  | 1 | 2 | 3 | 4 |  $O$  | FID:17.46

### **Conclusion and further research**

Using the FID with a pre-trained InceptionV3 network was not reflective on sample quality on MNIST: **should be fine-tuned**.



Variance was not taken into account: may be learned.

Unstable results between two experiments, with large differences in the results: **should be investigated further**.

Denoising result was basically settled from the first step. Should use a better noise scheduling, or another noising method (gaussian blur, or noise inside a VAE latent space).