

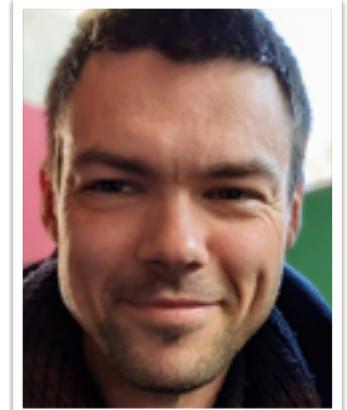
# Deep Avancé

(overview)

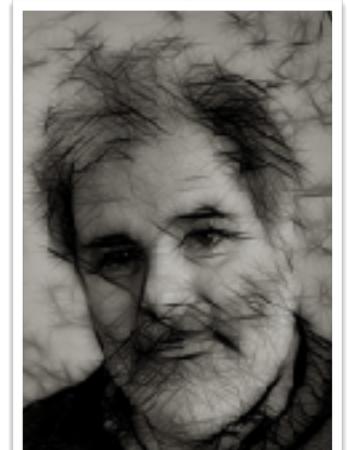
julien.Rabin / luc.brun (at) ensicaen.fr

# Organization

- **Volume:** 30 h (11 x 2h CM&TD + 4 x 2h TP)
- This course is composed of **Two main Topics:**
  - **Image Generative Models**  
(November - December) with **Julien Rabin**
  - Graph-Based Deep Learning  
(January) with **Luc Brun**



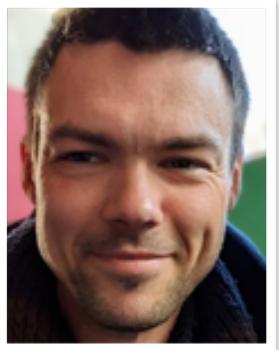
julien.rabin



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- **Evaluation:**
  - **Labs reports** (x4)
  - **Final exam** (1h30) *to be announced*

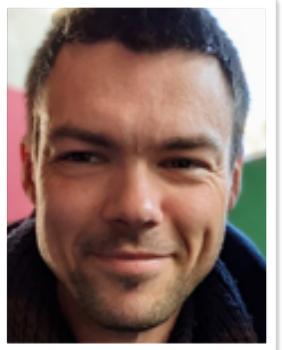


# About me



- Assistant Professor at **ENSICAEN** since 2023
- Member of the **Image Team** at **GREYC** Laboratory
- **Background:**
  - PhD in Object Recognition and Image Matching
  - Assistant Professor at IUT GON MP
- **Research:**
  - image processing & synthesis
  - optimal transportation
  - evaluation of generative models





# About me

- **contact:** [julien.rabin \[at\] ensicaen.fr](mailto:julien.rabin@ensicaen.fr)
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- **Feeback:**
  - mail
  - [online anonymous survey](#)

# About you

- **Sondage 1:**

« inclure des TD sous forme de notebook à faire sur portable ? »



<https://app.wooclap.com/NQOYXD>

# Image Generative Models

## *Overview*

- **Volume:** 16 h (6 x 2h CM&TD + 2 x 2h TP)
- **Prerequisite:** optimisation, deep learning, image processing
- **Tools:** python, pytorch, jupyter
- **Three main parts:**
  - **Patch-based** image synthesis
  - **Deep Neural Network** image generation
  - **Applications** and Evaluation of Generative Models

# Image Generative Models

## *Motivation*

- **Context:**
  - generative models in SOTA methods
  - at the core of Generative AI (e.g. text-to-image)
- **Why focusing on image ?**
  - signal processing and computer vision have many applications in industry
  - (text &) image have seen the first breakthrough for many deep learning techniques (over large datasets)

# Motivational Example

- **Text-to-Image Generation**

Example with Dall-E 3 (*via* Bing image creator)

**Prompt:**

*« A Koala riding a dolphin »*

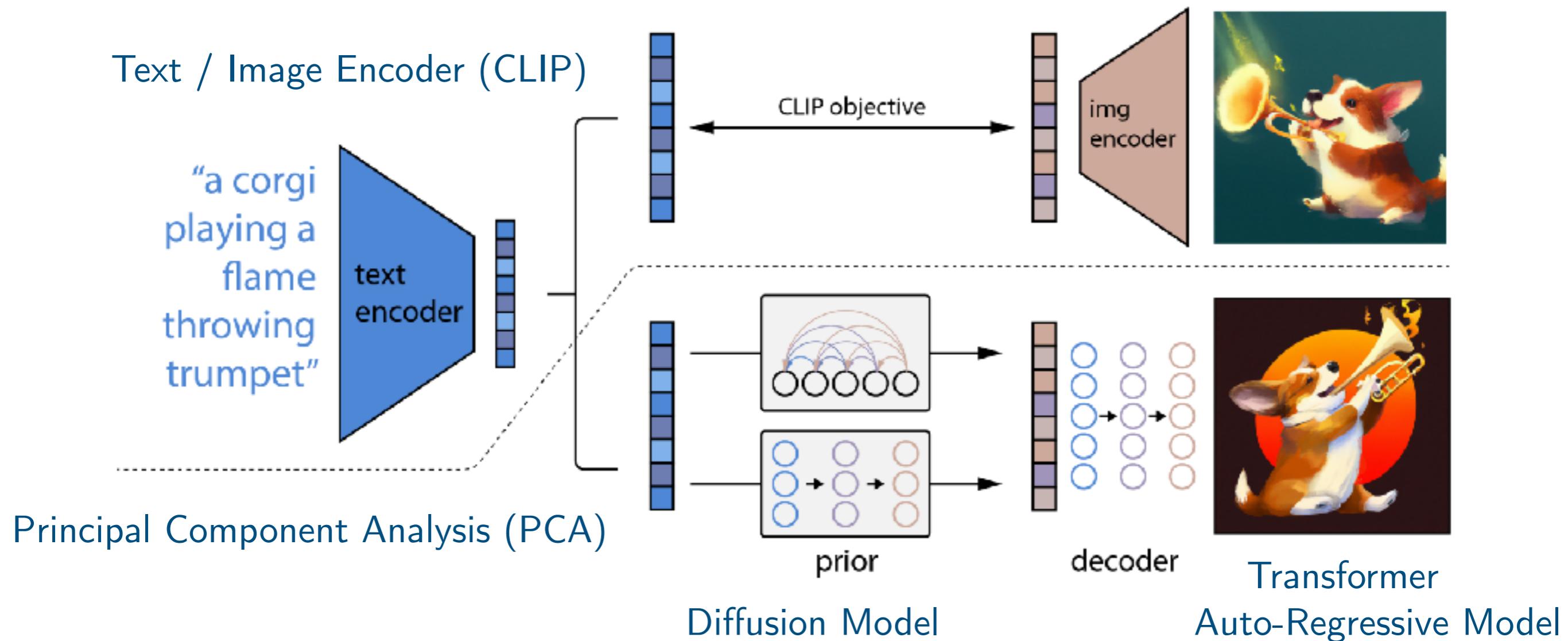
**Synthesis:**



# Motivational Example

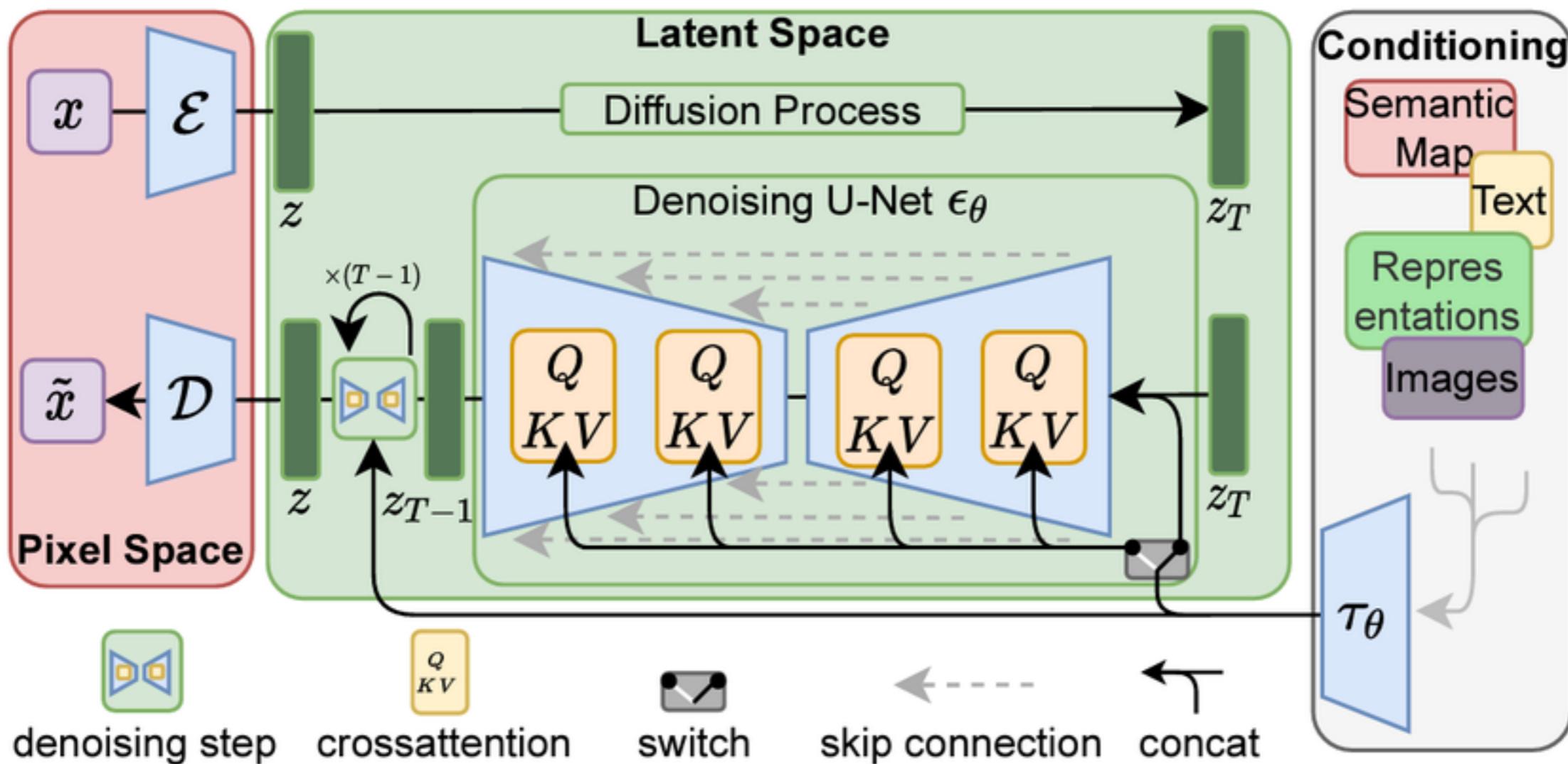
- **Text-to-Image Generation**

Architecture of Dall-E 2 (unCLIP)



# Motivational Example

- Another **architecture** based on a generative model
- Example of **latent diffusion** [Rombach *et al.*'22]



# Motivational Example

Latent representation

Vector Quantization (VQ)

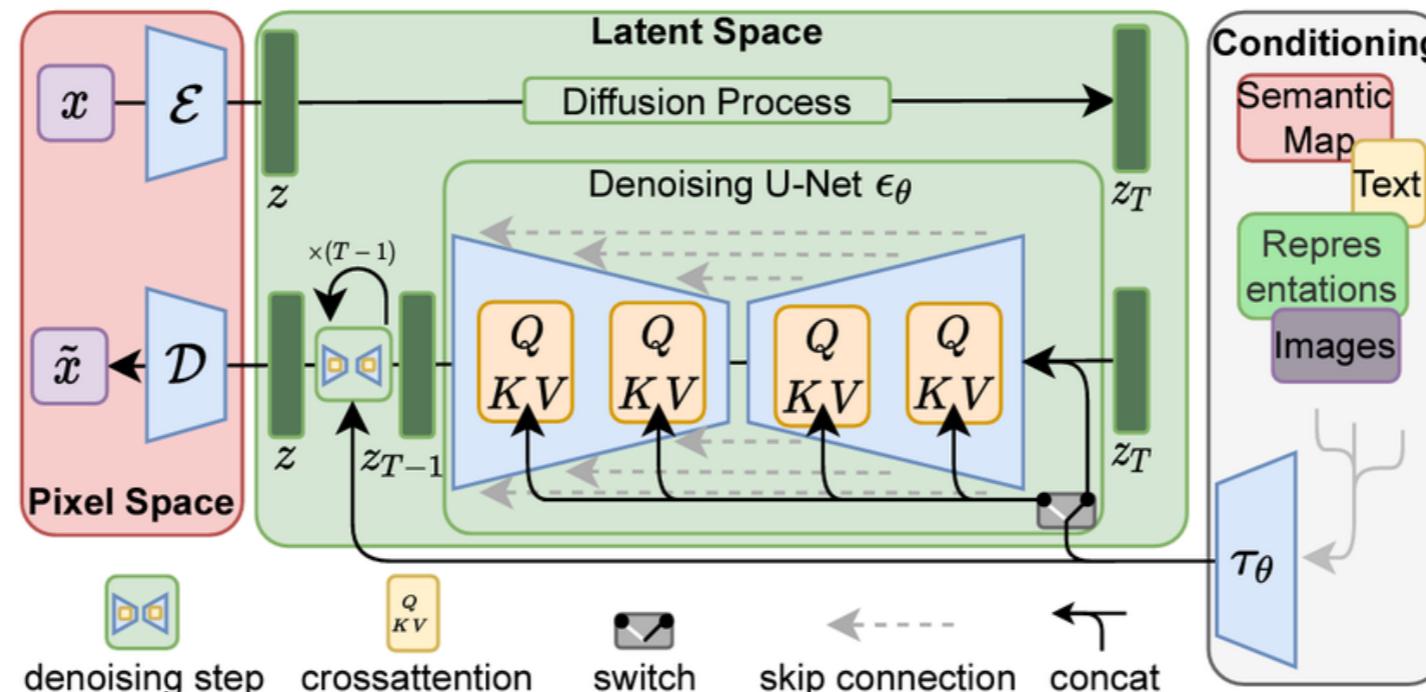
Variational Auto-Encoder (VAE)

Self-Attention

Adversarial  
Loss (GAN)

Cross-Attention

FID



Perceptual Loss

U-NET

Diffusion Model

Text Embedding

# Questions

- How to define a generative model for image ?
- What is the best embedding for image (and text) ?
- How to evaluate the performance of a model ?
- How to train a model with limited data ?
- How to use a generative model for image processing ?
- How to design a generative model for text (NLP) ?
- ...

# Part I: patch-based image synthesis & processing

*(Tentative Program)*

- **Objectives:**

- review some techniques (SOTA before deep learning) that are used in image generation
- practice on “small scale” problems

- **Outline:**

- quick review of neighbourhood-based image filtering
- application to image restauration
- acceleration of patch matching
- application to inpainting, texture synthesis & image stylisation

# Part II: **deep representation** for image synthesis & processing

*(Tentative Program)*

- **Objective:**

- review some deep learning techniques that are used in image processing
- again, practice on “small scale” problems

- **Outline:**

- unsupervised deep representation
- application to image restoration
- supervised deep representation for texture synthesis
- definition of the perceptual loss
- application to style transfert

# Part III: deep generative networks for images

- **Objective:** *(Tentative Program)*

- review SOTA methods for **unsupervised learning**
- again, practice on small datasets

- **Outline:**

- Auto-Encoder (AE)
- Variational Auto-Encoder (VAE)
- Generative Adversarial Network (GAN)
- Flow-based Models (realNVP, GLOW)
- Denoising Diffusion Probabilistic Model (DDPM), score energy-based models
- Auto-regressive Model (PixelCNN, Transformers)

# Part IV: Applications & Evaluation of generative models

*(Tentative Program)*

- **Objective:**

- review applications and limitations

- **Outline:**

- Semi-supervised training (conditional generation)
  - VQ-GAN, latent-diffusion
  - Applications to image processing and editing
  - Text & Image Embedding (CLIP)
  - FID & Precision Recall
  - Overfitting and Memorization
  - Membership Attacks, Data Poisoning