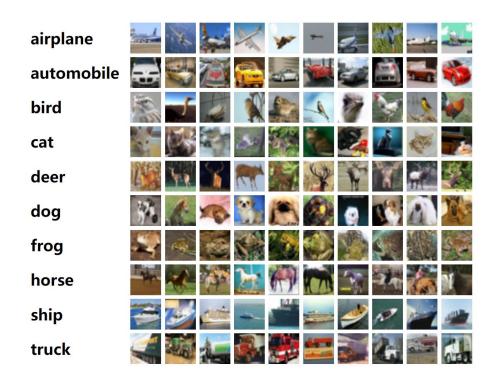
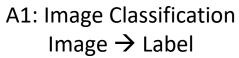
Assignment II: Image Generation GitHub page

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What you have learned in Assignment I







A2: Image Generation"Cat" - generated samples
Generative model

- What is a generative model
 - A model learns to capture the underlying data distribution to **generate new data samples** similar to the training data.
- Types of generative model
 - Generative Adversarial Networks (GANs)
 - Variational Autoencoders (VAEs)
 - Diffusion Models: a newer class of generative models

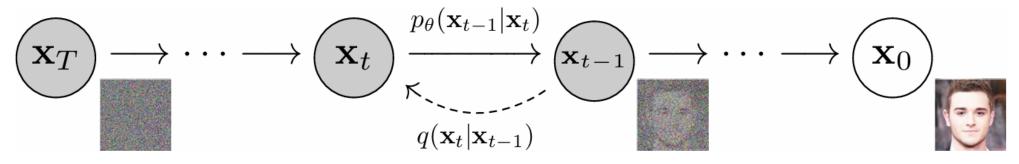


Stable diffusion: text to image

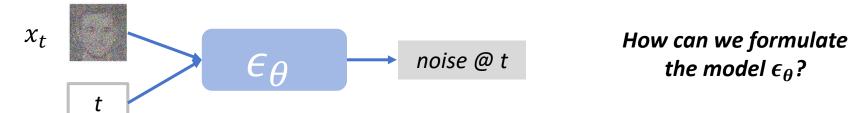


AIGC by Midjourney

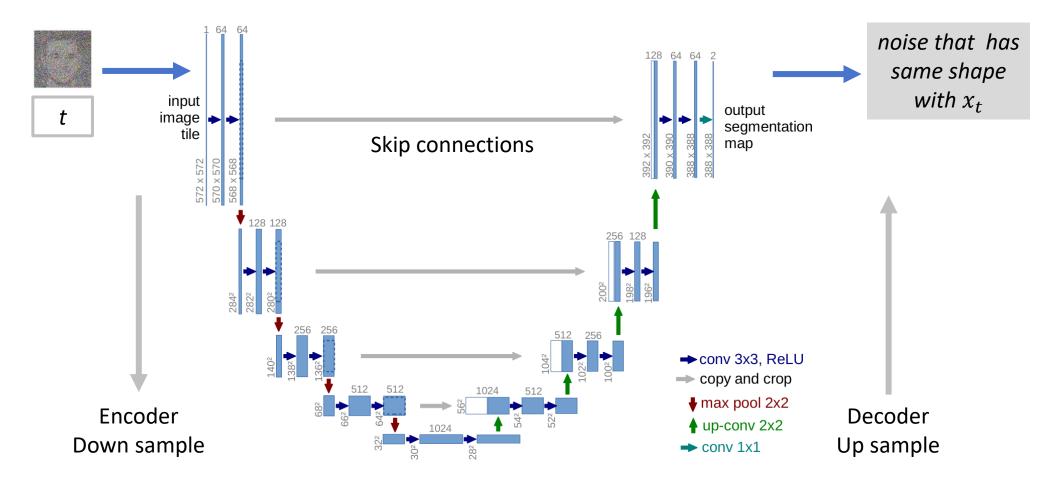
Denoising Diffusion Probabilistic Models (DDPM)



- Forward process: gradually adds noises to an image over several timesteps, converting a clean image into almost pure Gaussian noise.
- Reverse process: denoise the noisy samples step by step, reconstructing the original data. Specifically, to **predict the noise that needs to be subtracted** from the noisy input at each step



U-Net architecture for reverse diffusion modeling



2 Working on the Assignment

This assignment, in one word, is to train a DDPM with U-Net architecture on the MNIST handwritten digits dataset.

MNIST dataset:

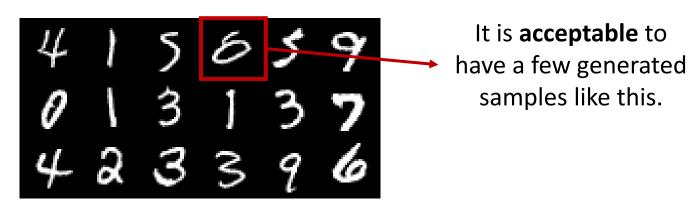


2.1 Task 1: Fill in the blank (70%)

We have provided you with the initial codebase, which is defined in three Python files:

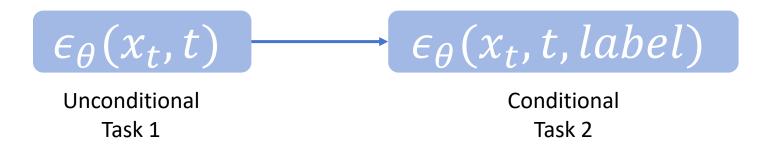
- model.py Contains the definition of the diffusion model.
- unet.py Contains the definition of the U-Net model that handles the denoising process.
- train_mnist.py The main training script. Run python train_mnist.py to start training and evaluating the model.
- There are **5** code blocks in <u>model.py</u> and <u>unet.py</u> that require you to complete them.
- Successfully implementing the code and showcasing the generated handwritten digits in the report will earn full marks.

An example



2.2 Task 2: Conditional generation (10%)

 This task involves extending the existing code to implement conditional generation, that is, to enable the generation of images corresponding to specific digits. The implementation should be detailed in your final report.



 You may look into the handing of time step input, and think about how to cope with the input of label in a similar way.

2.3 Task 3: Write a report (20%)

- No more than 2 pages
- Your report should be structured into three main sections: Introduction, Method, and Experiment
- It should highlight the effectiveness of the image generation, showcasing outputs for all digits from 0 to 9.
- Your analysis is important. [examples]

3 Files to submit

1. Final Report (PDF, up to 2 pages)

2. Codes

- a) 3 python files: model.py, unet.py, train_mnist.py
- b) README.txt if you added some python files.
- c) One .ipynb file is also acceptable if you find it more convenient to use Jupyter notebook (Pls copy and paste code into a notebook by yourself)

3. Model Weights

4 Important Dates

- > Oct. 6, 2024 (Sun.): The assignment release.
- > Nov. 10, 2024 (Sun.): Submission deadline (23:59 GMT+8).

Late submission policy:

- 10% for late assignments submitted within 1 day late.
- 20% for late assignments submitted within 2 days late.
- 50% for late assignments submitted within 7 days late.
- 100% for late assignments submitted after 7 days late.

5 Need More Support?

- For any questions about the assignment which potentially are common to all students, you shall first look for related resources as follows,
 - ✓ We encourage you to use <u>GitHub Issues</u> of this repository.
 - ✓ Or if you prefer online doc: <u>Discussion doc</u>.

For any other private questions, please contact Zhiqian Lan via email.