

# Homework #3

## Image Generation



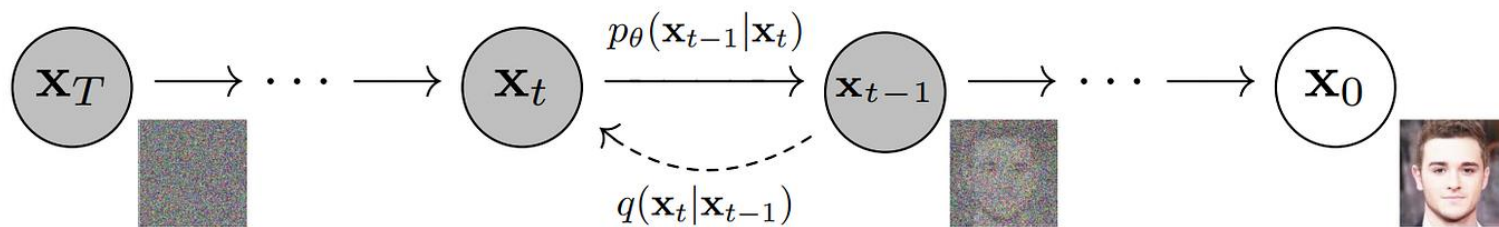
Wen-Huang Cheng (鄭文皇)

National Taiwan University

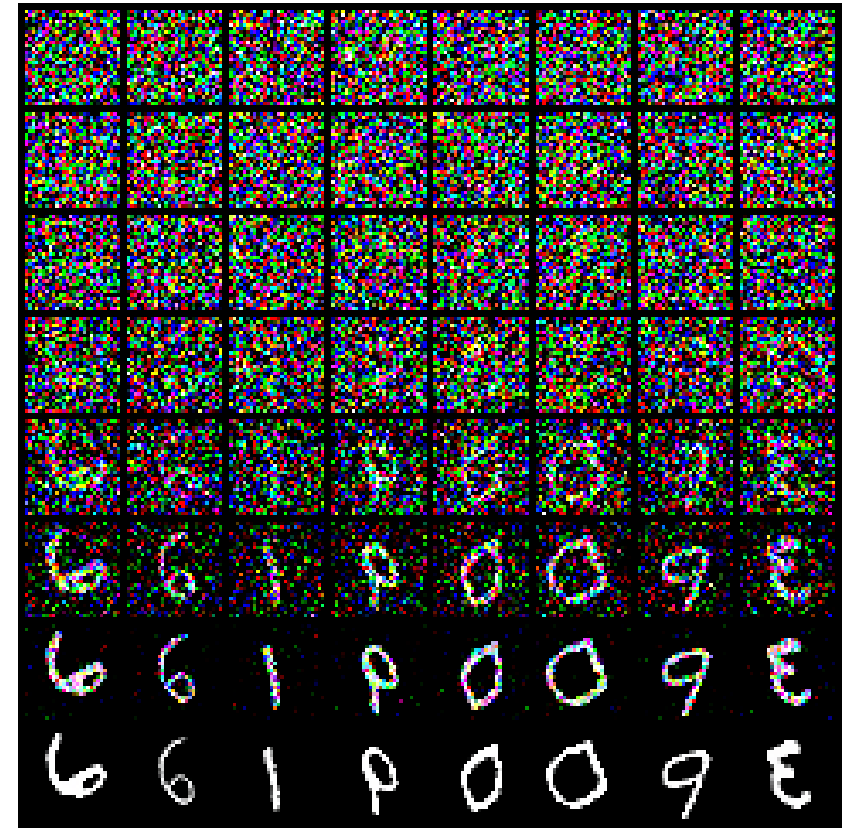
[wenhuang@csie.ntu.edu.tw](mailto:wenhuang@csie.ntu.edu.tw)

## TOPIC: Image Generation for Handwritten Digits

- Implementing diffusion generative models.
- Evaluating generative models in terms of FID.



DDPM (NeurIPS 2020, [link](#))





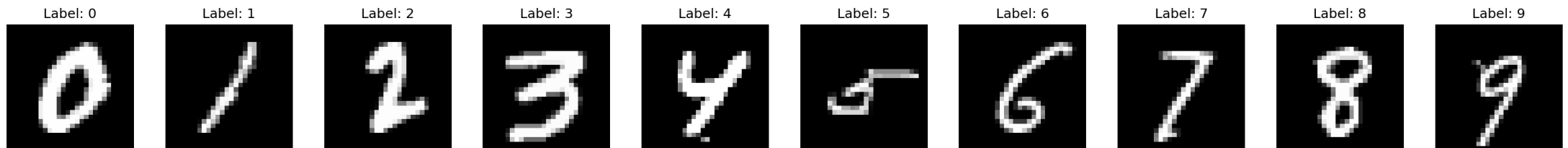
# Submission Deadline

**2025/11/28 23:59**

**NTU COOL: Late Submission Policy is Applicable  
(See TA01-Assignments\_Overview.pdf)**

- **The grading policy for TAICA students should be determined by your institution's teaching assistants or professors. Any questions, please contact them.**

- Download link: [link](#)
- Training set: 60,000 handwritten digits
- Image size: 28x28
- The images have been converted to RGB for simpler implementation





# Evaluation Metric – FID

- The Fréchet Inception Distance (FID) is a widely used metric to measure the divergence between two distributions of images, specifically to evaluate the quality of images generated by models.
- A lower FID score indicates that the two distributions are closer, meaning the generated images are more similar to the real ones in terms of quality and diversity.
- For more information, you can refer to [link](#).



# FID Implementation

- We use pytorch-fid ([GitHub Repositories](#)) to calculate FID:

```
pip install pytorch-fid
```

- Calculate FID (Two ways):

- With the training dataset

```
python -m pytorch_fid path/to/images path/to/mnist
```

- With precalculated mean and covariance

```
python -m pytorch_fid path/to/images path/to/mnist.npz
```

- path/to/images: the folder of the generated images.
- path/to/mnist: the folder of the training data.
- path/to/mnist.npz: the precalculated mean and covariance of training data, which can be downloaded from [link](#).

➤ FID (90%) (evaluated between 10000 generated images and training dataset)

Your FID	Points
< 30	90
30 ~ 70	Linear between 60 ~ 90
70 ~ 100	Linear between 0 ~ 60
> 100	0

➤ Report (10%)





# Programming Spec (1)

- Implement a **diffusion-based** generative model **on your own**
- Use Python  $\geq 3.10$  (for consistency and reproducibility).
- **Pretrained weights and external datasets are forbidden.**
- Train your model from scratch.
- You may consult online resources (e.g., [link](#)) but you must understand it and **write the code yourself.**
- No plagiarism.
  - **Violating the above rules on this page will result in a score of zero.**
  - **If you are uncertain about the legitimacy of the usage, email the TAs for clarification**



# Programming Spec (2)

## ➤ Python Package White List:

- Deep learning packages: Tensorflow (including Keras), PyTorch.
  - Deep learning related packages: torchvision.
  - Machine learning packages: sklearn, xgboost.
  - Others: pandas, tqdm, tensorboardX, tensorboard, opencv-python, pillow, scipy, numpy.
- Violating the above rules on this page will result in a score of zero.
  - If you are uncertain about the legitimacy of the usage, email the TAs for clarification



# Submission File – Source Code

- Zip your source code and submit to NTU COOL:

|----- code\_< student-id>.zip

|----- src/ (Your source code)

|----- readme.md (Show how to install the environment, dataset, run training & generation)

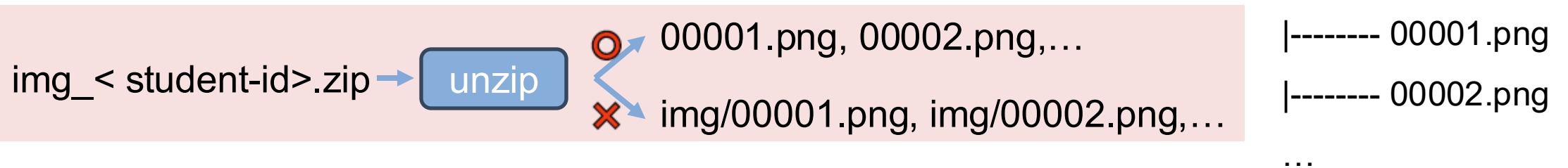
|----- requirements.txt (The list of necessary packages)

- **DO NOT** include dataset, add the description about dataset path in readme.
- **DO NOT** include model weights.

- An incorrect format will result in a deduction of a -5 score.

# A Submission File – Generated Images

- The number of generated images: **10000**.
- The dimension of generated images: **28x28 RGB**.
- The format/filename of generated images: **png** (**00001.png ~ 10000.png**).
- Zip these images into a .zip file **without any subdirectory** |----- img\_< student-id>.zip



- **DO NOT** include any files other than images (e.g., for MacOS users, **DO NOT** include `.DS_Store` and `__MACOSX` in your zip file).
- Violating the above rules on this page will result in 0 score of FID evaluation.

## 1. Model Description

- Introduce your model (must include an **architecture illustration**)

## 2. Implementation Details

- Preprocessing, augmentation, hyperparameters, loss functions, training strategies, etc.

## 3. Result Analysis

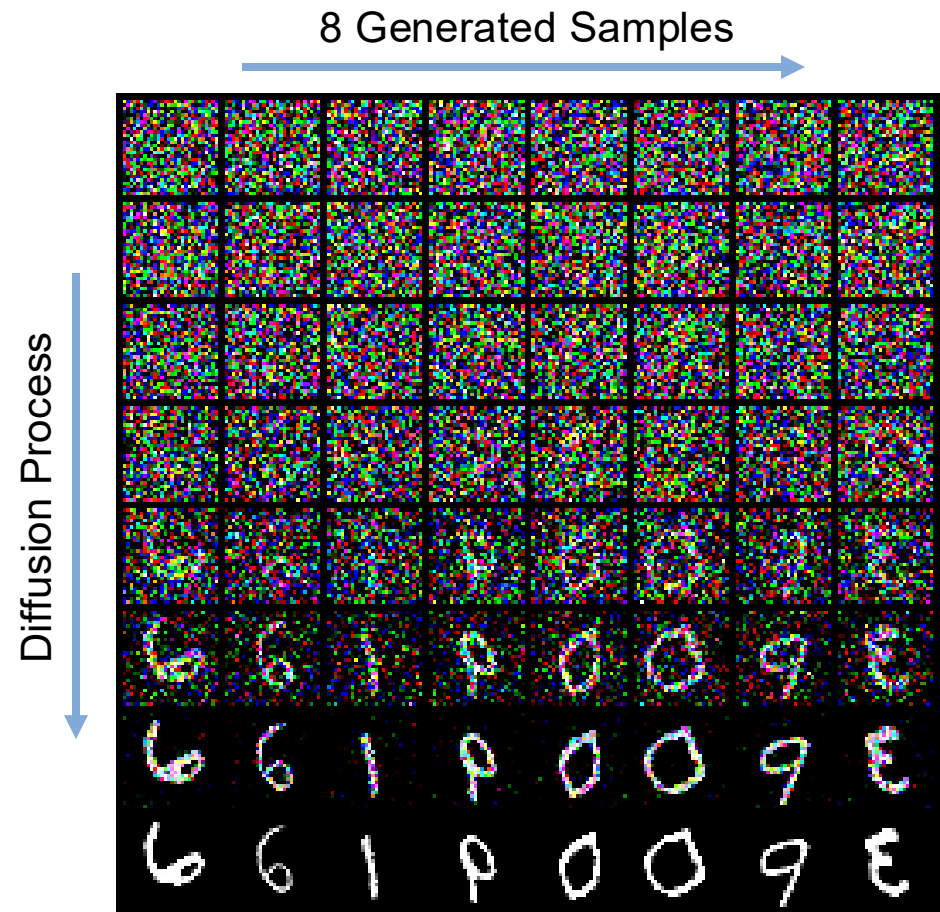
- Quantitative improvements (tables, metrics, discussion)
- Diffusion Process Visualizations (see next page) (5%)

## 4. Short conclusion

- 3-5 pages (excluding references), exceeding the limit will result in a -5 score

## ➤ Diffusion Process

- Start the diffusion process from noise
- Generate 8 different results.
- Divide the time steps into 7 equal parts.  
For example, if the total number of steps is 1000, record the results every 142 steps.
- Arrange these 64 images into a form similar to the image on the right and paste it on the report.





# NTU COOL Submission Rules

- Your submission should be a zipped file with the following structure:
  - hw3\_<student-id>.zip (e.g., hw3\_R12345678.zip)
    - |-- hw3\_<student-id> (Should contain this folder, not separate files)
      - |----- report\_<student-id>.pdf (Your report)
      - |----- code\_< student-id>.zip (Your source code)
      - |----- img\_< student-id>.zip (Your generated images)
- An incorrect format will result in a deduction of a -5 score.
- Failure to re-implement similar performance will result in a 60% discount of the total score.
- Plagiarism in the report or code will result in 0%.

# Any Question

**Ask peers first**

(Join with name: <school\_student-id>)



**Then ask TAs**

(only for NTU students)

[cvpdl.ta.fall.2025@gmail.com](mailto:cvpdl.ta.fall.2025@gmail.com)