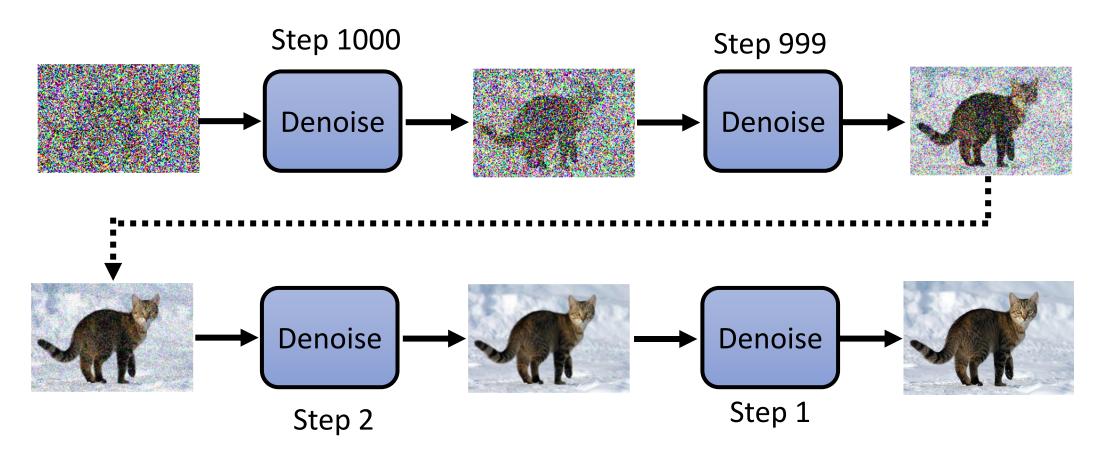
# Diffusion Model

Denoising Diffusion Probabilistic Models (DDPM) https://arxiv.org/abs/2006.11239

### Diffusion Model 是如何運作的?



**Reverse Process** 

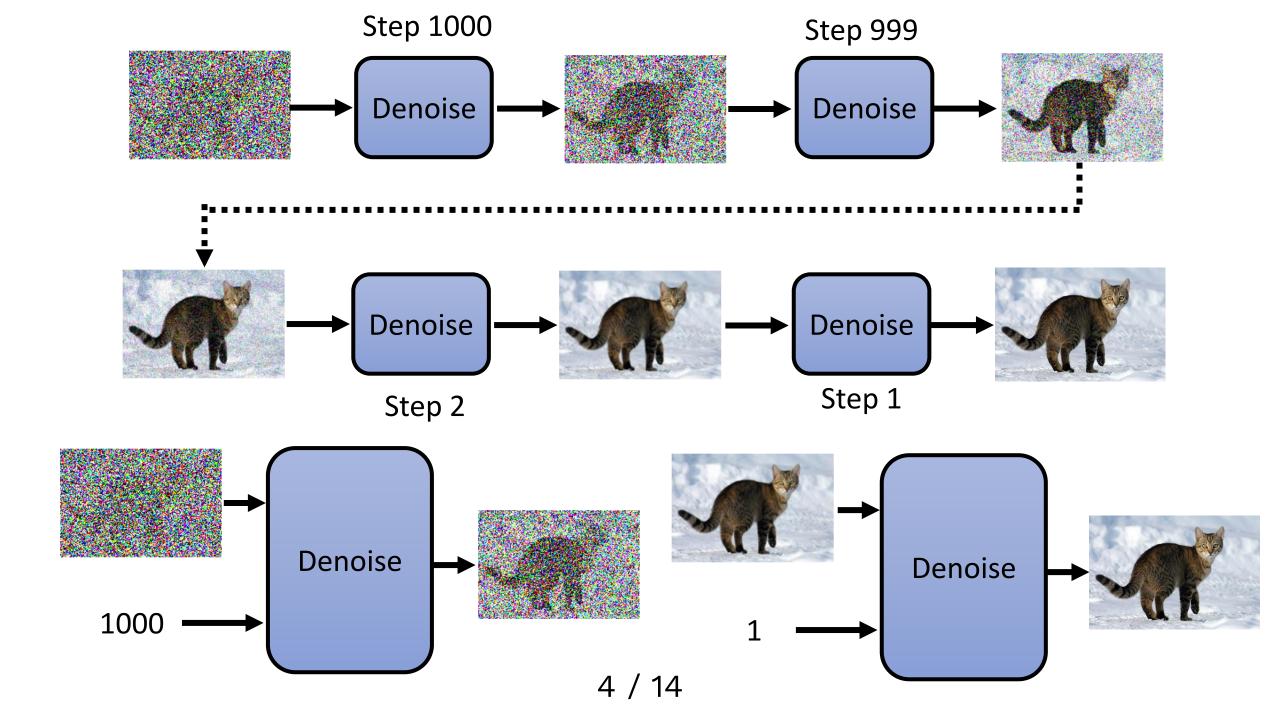


The sculpture is already complete within the marble block, before I start my work. It is already there, I just have to chisel away the superfluous material. - Michelangelo

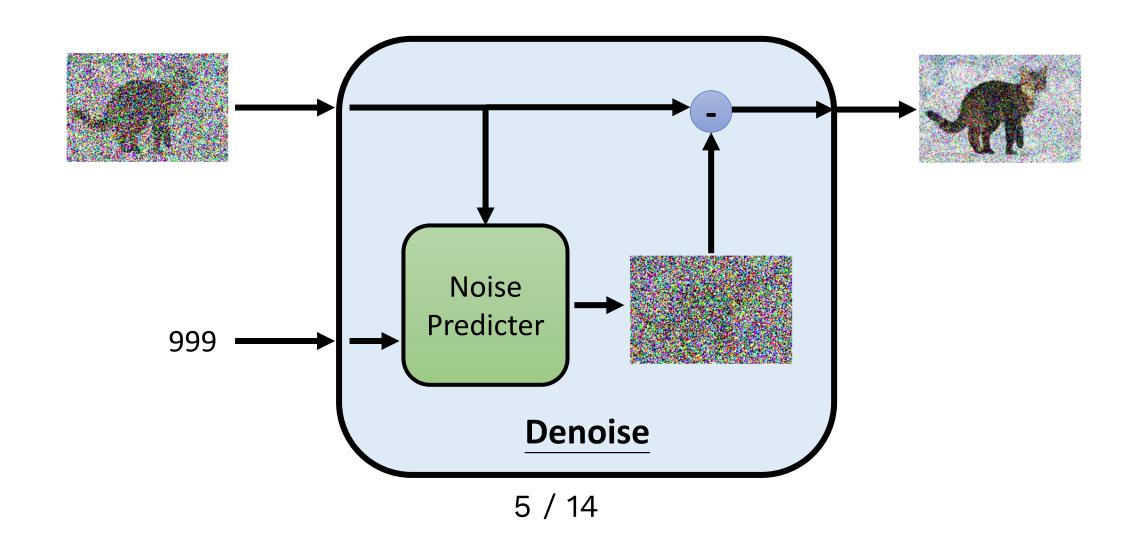




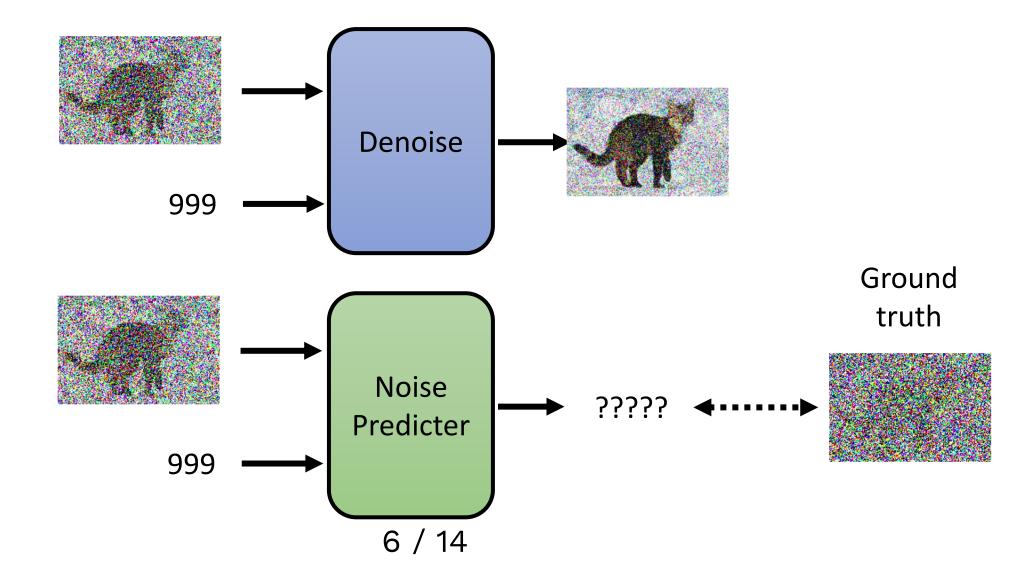


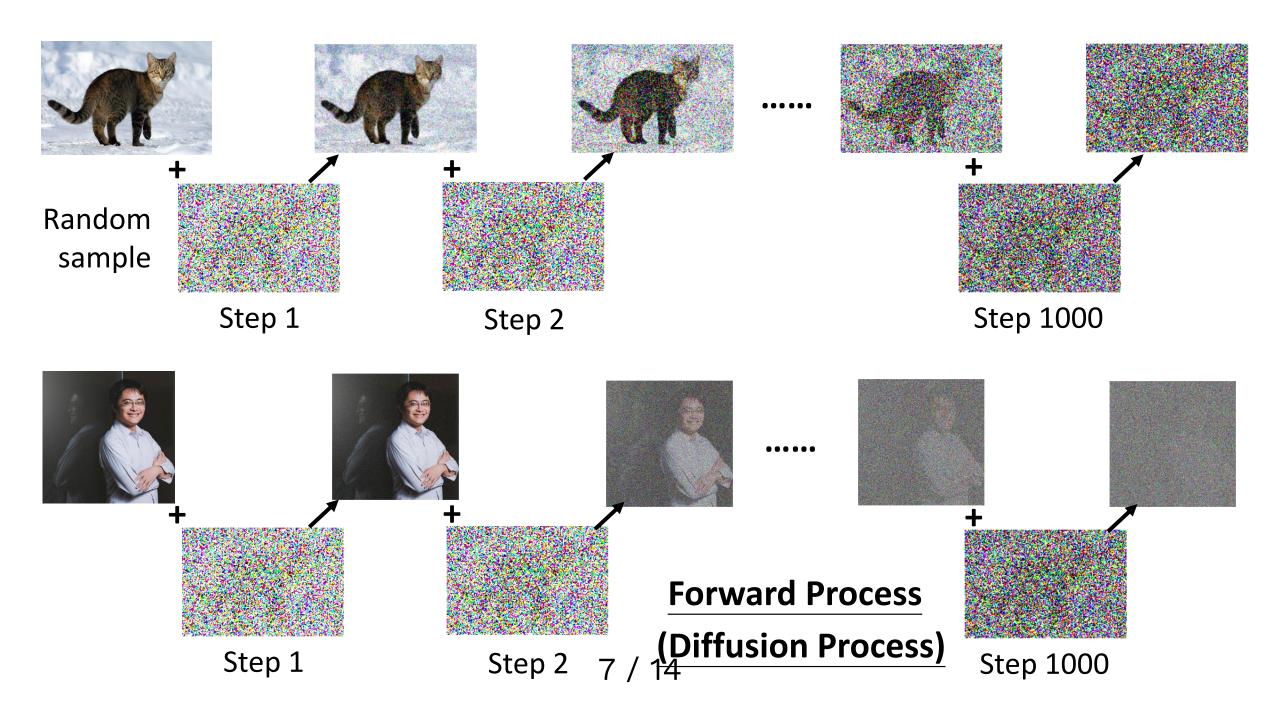


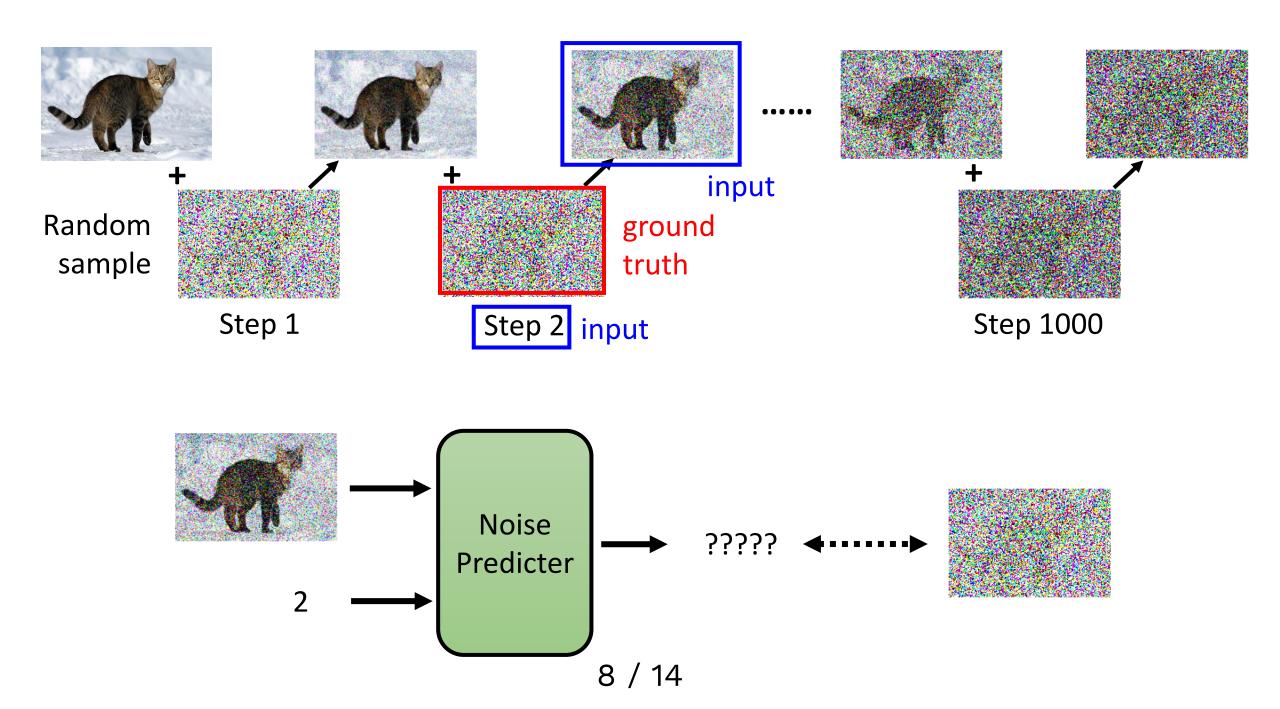
## Denoise 模組內部實際做的事情



### 如何訓練 Noise Predictor







#### Text-to-Image

https://laion.ai/blog/laion-5b/

HW6 ImageNet LAION 70k 1M 5.85B

A cat in the snow

Text-to-image Generator



Backend url:

https://knn5.laior

Index:

laion\_5B

french cat

Clip retrieval works by converting the text query to a CLIP embedding, then using that embedding to query a knn index of clip image embedddings

Display captions Display full captions Display similarities

Safe mode ✓ Hide duplicate urls

Hide (near)
duplicate images

Search over

image V

Serch with multilingual clip



french cat



french cat



How to tell if your feline is french. He wears a b...

イケ

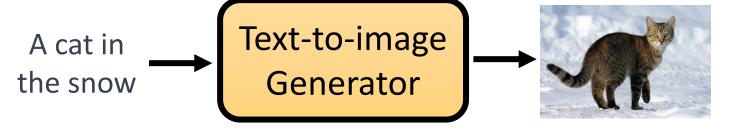
NAVI

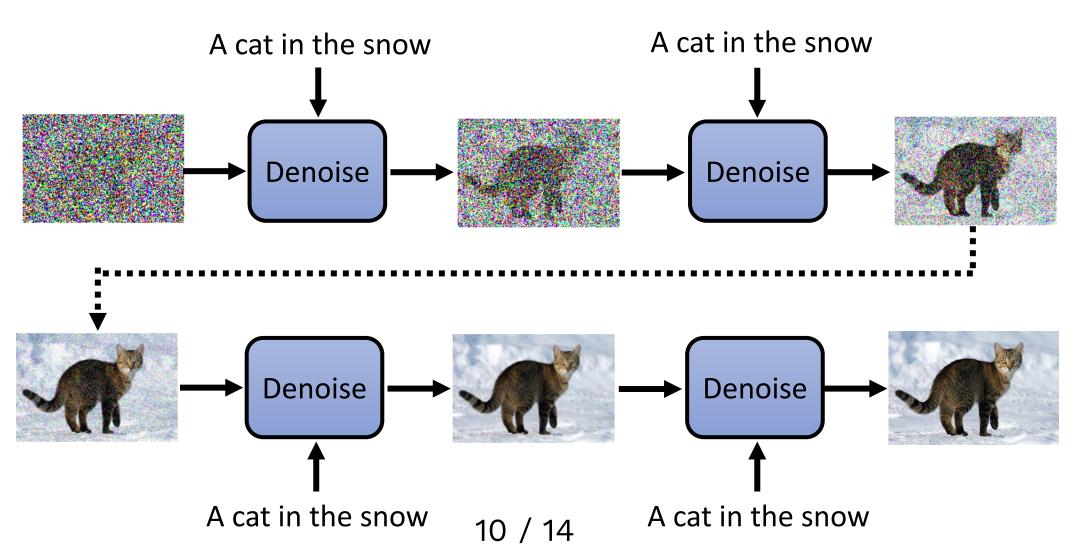
cat in a suit Georgian sells tomatoes



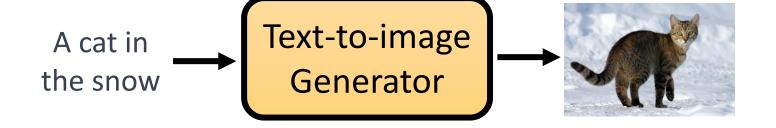
網友挑戰「加幾筆畫 出最創意貓咪圖片」,

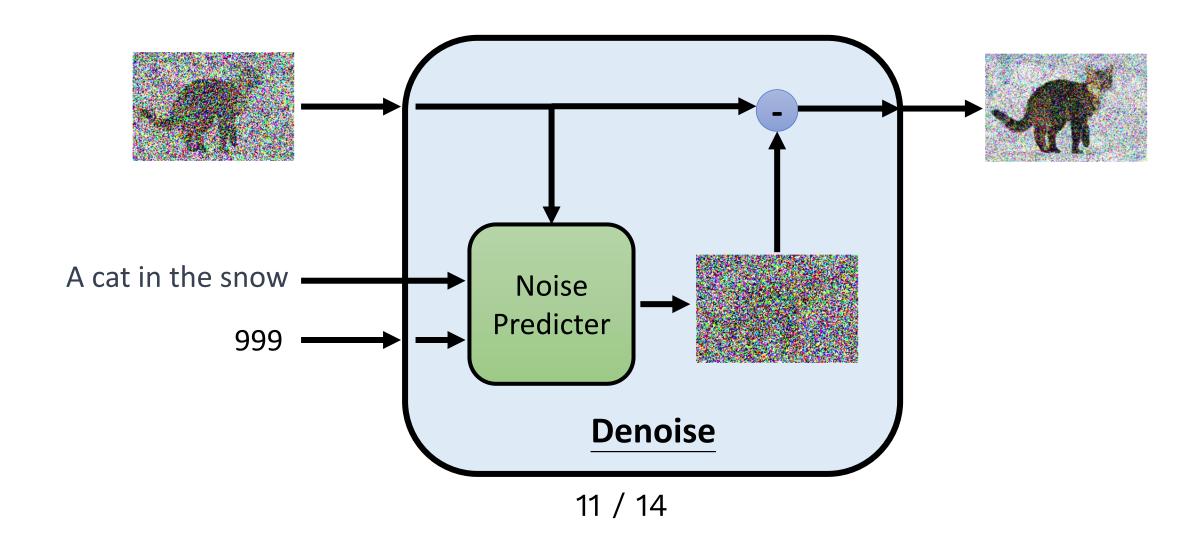
#### Text-to-Image

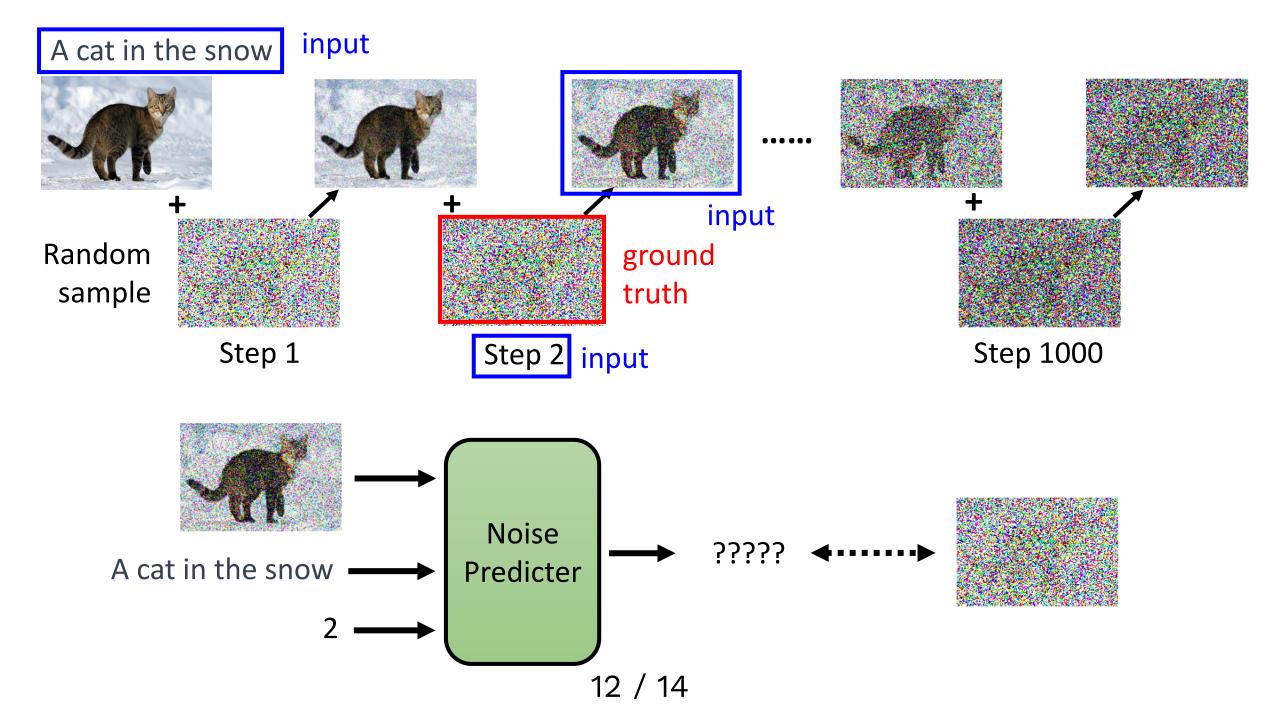




#### Text-to-Image







## Denoising Diffusion Probabilistic Models

#### **Algorithm 1** Training

#### 1: repeat

- 2:  $\mathbf{x}_0 \sim q(\mathbf{x}_0)$
- 3:  $t \sim \text{Uniform}(\{1, \dots, T\})$
- 4:  $\epsilon \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$
- Take gradient descent step on

$$\nabla_{\theta} \| \boldsymbol{\epsilon} - \boldsymbol{\epsilon}_{\theta} (\sqrt{\bar{\alpha}_t} \mathbf{x}_0 + \sqrt{1 - \bar{\alpha}_t} \boldsymbol{\epsilon}, t) \|^2$$

6: until converged

#### **Algorithm 2** Sampling

- 1:  $\mathbf{x}_T \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$
- 2: **for** t = T, ..., 1 **do**
- 3:  $\mathbf{z} \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$  if t > 1, else  $\mathbf{z} = \mathbf{0}$
- 4:  $\mathbf{x}_{t-1} = \frac{1}{\sqrt{\alpha_t}} \left( \mathbf{x}_t \frac{1-\alpha_t}{\sqrt{1-\bar{\alpha}_t}} \boldsymbol{\epsilon}_{\theta}(\mathbf{x}_t, t) \right) + \sigma_t \mathbf{z}$
- 5: end for
- 6: **return**  $\mathbf{x}_0$

# Diffusion Model

Denoising Diffusion Probabilistic Models (DDPM) https://arxiv.org/abs/2006.11239