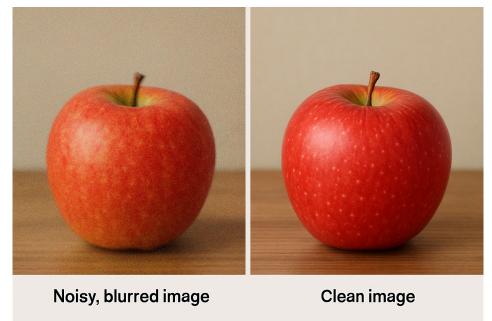
# Denoising Autoencoders (DAE)



## Core Idea of a DAE

Original image → Noisy version → Reconstructed original image



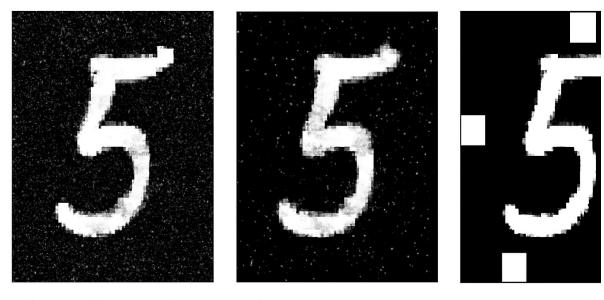
## **Analogy: DAE as a Photo Restorer**



Old, scratched photograph

**Digitally restored** 

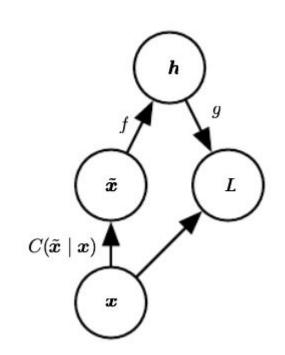
## **How Do We Introduce Corruption?**



Gaussian noise. Salt & pepper noise. Randon pixel

## Training a DAE – Step by Step

- **1.** Clean image x
- **2.** Corrupted image  $\tilde{x}$
- **3.** Latent code  $h=f( ilde{x})$
- **4.** Reconstruction  $\hat{x} = g(h)$



## **Applications of DAE**

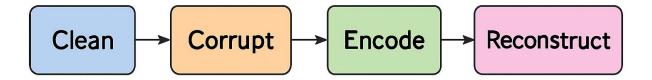
Image deblurring

Audio denoising

Anomaly detection

Pretraining for deep learning models

## **DAE** in a Nutshell



# Q1. What is the main difference between a standard autoencoder and a denoising autoencoder?

- A. DAEs reconstruct the input exactly as it is.
- B. DAEs add noise to the output to make it robust.
- C. DAEs are trained on corrupted inputs but reconstruct clean outputs.
- D. DAEs only work on image data.

#### Q2. In Denoising Autoencoders, what does the corruption process $C(\tilde{x}|x)$ represent?

- A. A function that compresses the input data.
- B. A distribution over possible denoised outputs.
- C. A mechanism to generate noisy versions of the input.
- D. A loss function used to measure reconstruction error.

# Q3. Which of the following is a valid example of a corruption technique used in DAEs?

- A. Max pooling
- B. Softmax activation
- C. Salt and pepper noise
- D. Feature scaling

# Q4. What benefit does a Denoising Autoencoder offer beyond just cleaning noisy data?

- A. It requires fewer training samples.
- B. It always achieves higher pixel accuracy.
- C. It learns robust and meaningful latent representations.
- D. It eliminates the need for a decoder.