

# Machine Learning Course

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# Unsupervised Learning

- ▶ **Dimensionality Reduction**
- ▶ **Clustering**
- ▶ **Autoencoding**
- ▶ **Anomaly Detection**
- ▶ **...**

# Autoencoder

- ▶ Autoencoders are a type of **unsupervised neural network** (i.e., no class labels or labeled data) that seek to:
  - ▶ Accept an input set of data (i.e., the *input*).
  - ▶ Internally *compress* the input data into a **latent-space representation** (i.e., a single vector that *compresses* and *quantifies* the input).
  - ▶ **Reconstruct the input data** from this latent representation (i.e., the output).

# Autoencoder

- ▶ **Encoder:** Accepts the input data and compresses it into the latent-space.
- ▶ **Decoder:** The decoder is responsible for accepting the latent-space representation and then reconstructing the original input.

$X$  = input image

$E \rightarrow$  encoder

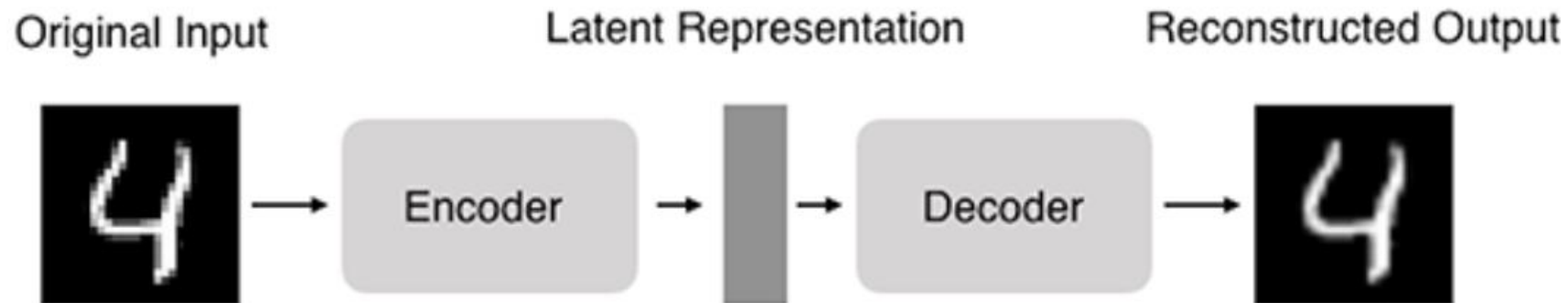
$D \rightarrow$  decoder

$\hat{X}$  = reconstructed image

$\hat{X} = D(E(X))$

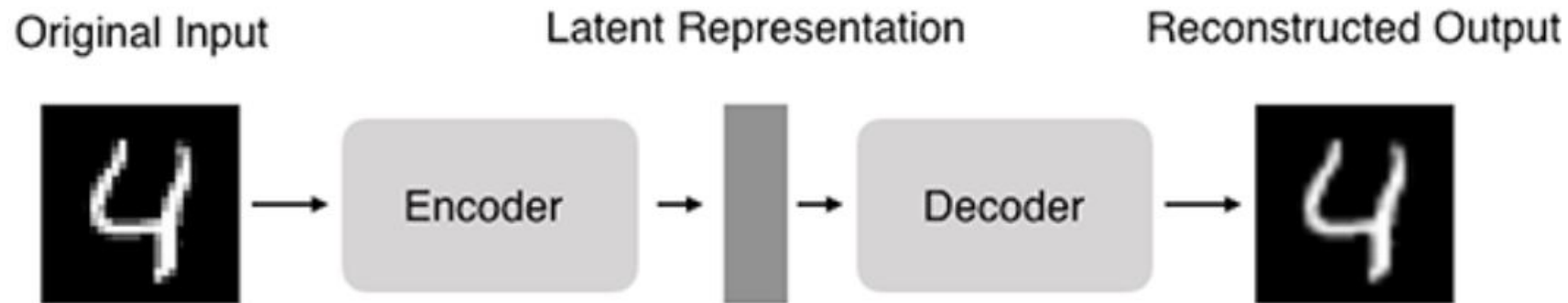
# Autoencoder

- ▶ We input a digit to the autoencoder.
- ▶ The encoder subnetwork creates a latent representation of the digit. This latent representation is ***substantially smaller*** (in terms of dimensionality) than the input.
- ▶ The decoder subnetwork then reconstructs the original digit from the latent representation.



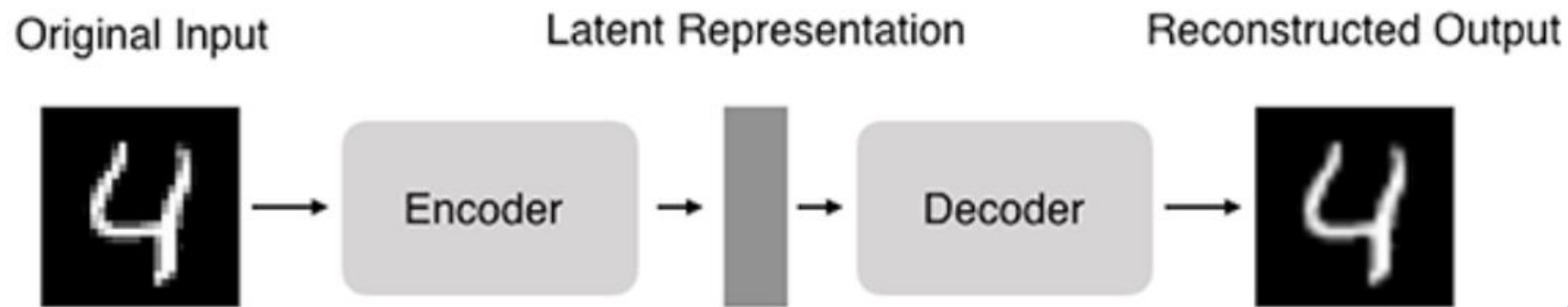
# Autoencoder

- ▶ An autoencoder is a network that *reconstructs its input?!*



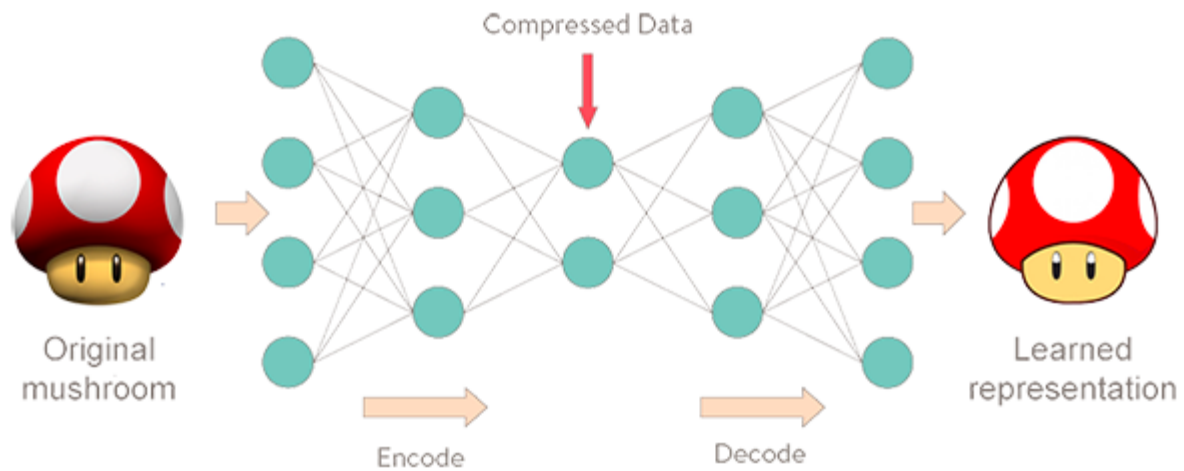
# Autoencoder

- ▶ During the training process, our goal is to train a network that can learn how to reconstruct our input data. So what's the point?
- ▶ **The true value of the autoencoder lives inside that latent-space representation.**
- ▶ Keep in mind that autoencoders *compress* our input data and, more to the point, when we train autoencoders, what we *really* care about is the **encoder**.



# Autoencoder

- ▶ **Loss is difference between the reconstructed and the original image**

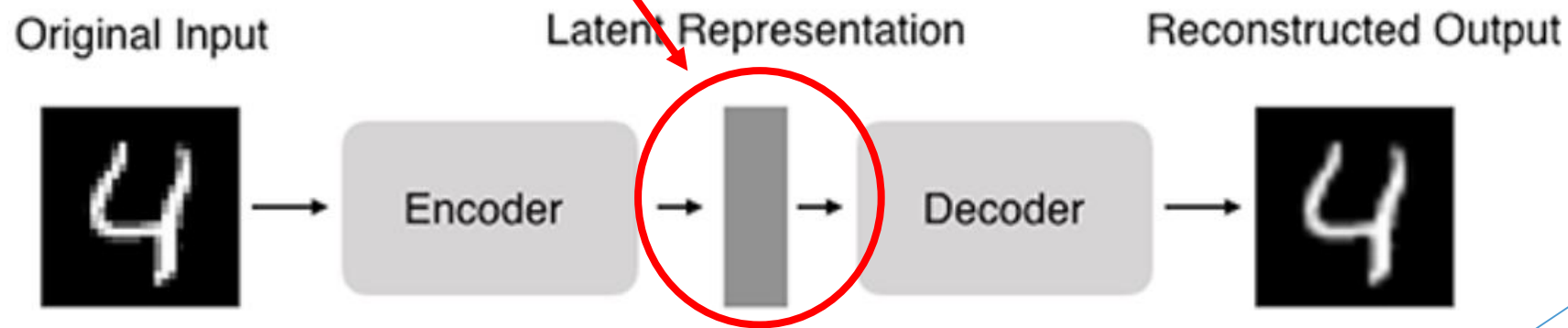




# Autoencoder

- ▶ Applications of autoencoders

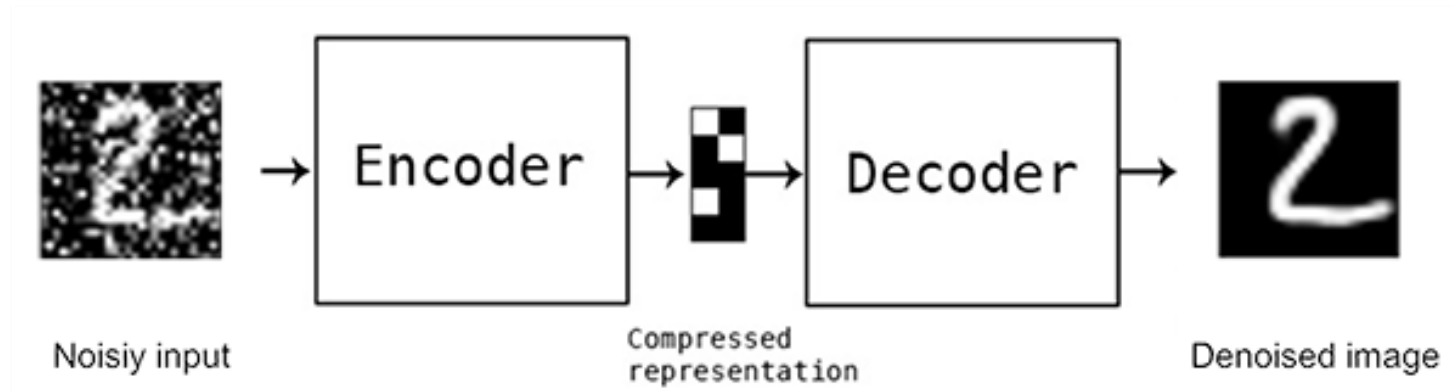
- ▶ **Dimensionality reduction** (i.e., think PCA but more powerful/intelligent).



# Autoencoder

- ▶ Applications of autoencoders

- ▶ **Denoising** (ex., removing noise and preprocessing images to improve OCR accuracy).



# Autoencoder

- ▶ **Applications of autoencoders**

- ▶ **Anomaly/outlier detection** (ex., detecting mislabeled data points in a dataset or detecting when an input data point falls well outside our typical data distribution).

