

Deep Learning: Theory and Practice

Advanced Topics in Deep Learning

14-5-2020

Supervised



$\downarrow L(y, t)$

Unsupervised



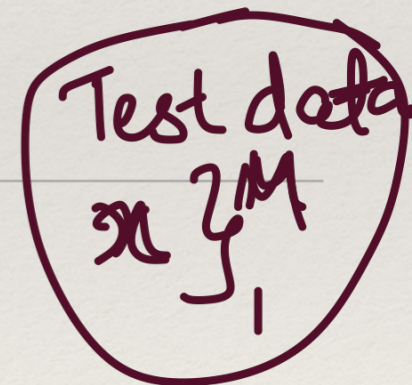
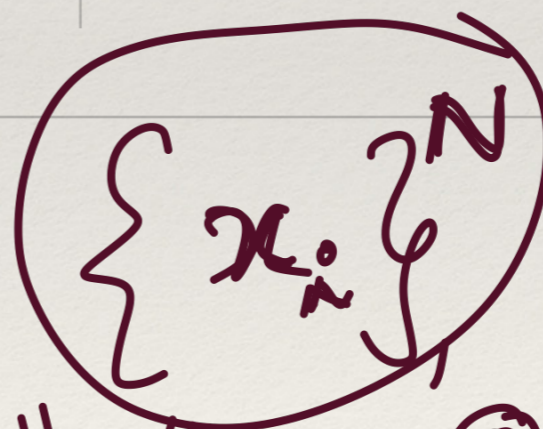
→ Generate new data

→ Can I combine unsupervised with supervised small data

Deep Unsupervised Learning

Active learning

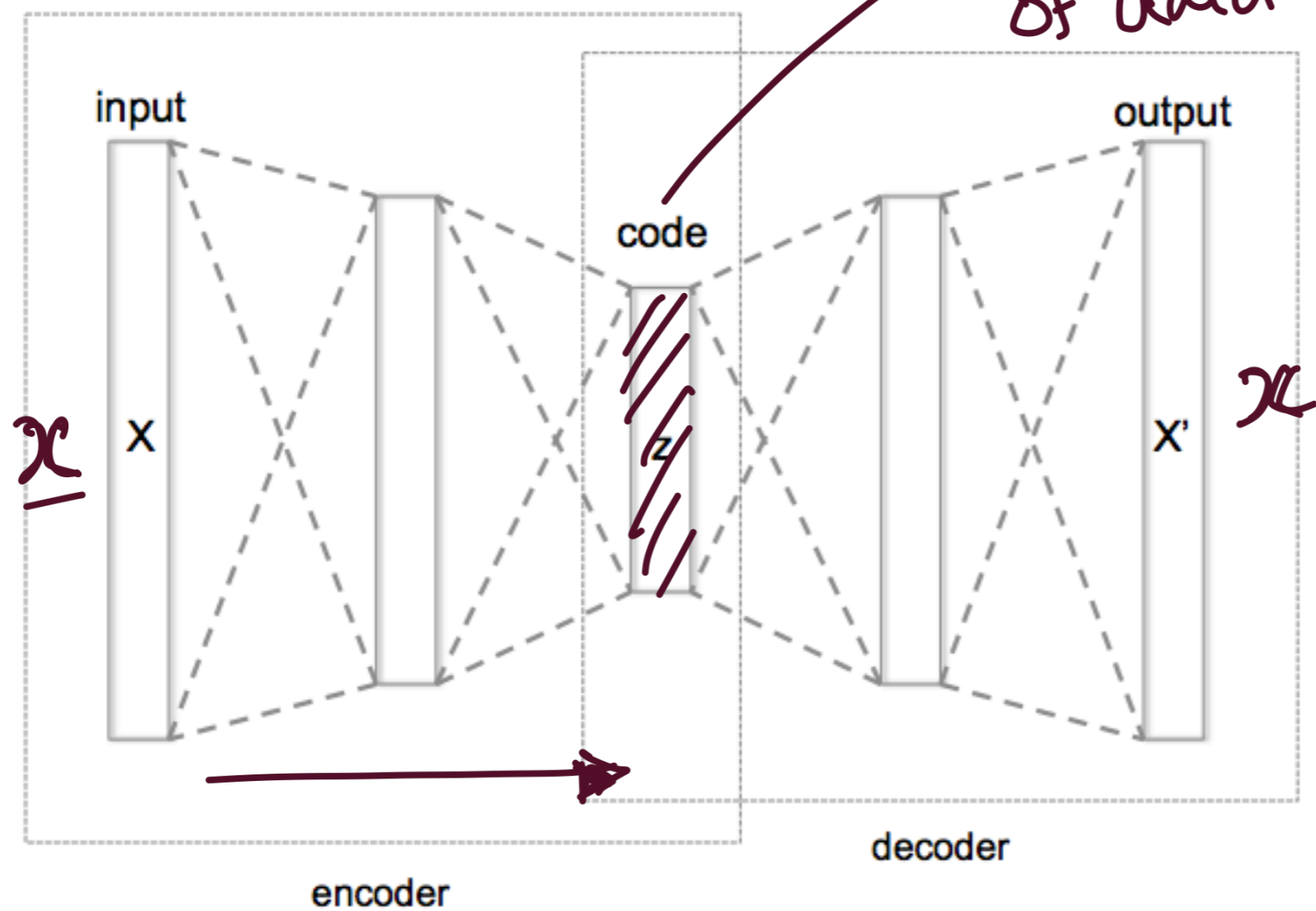
Resources to label "n" $n \ll N$



$N = 100,000$
 $n = 1,000$

Autoencoders

Rich representation of data

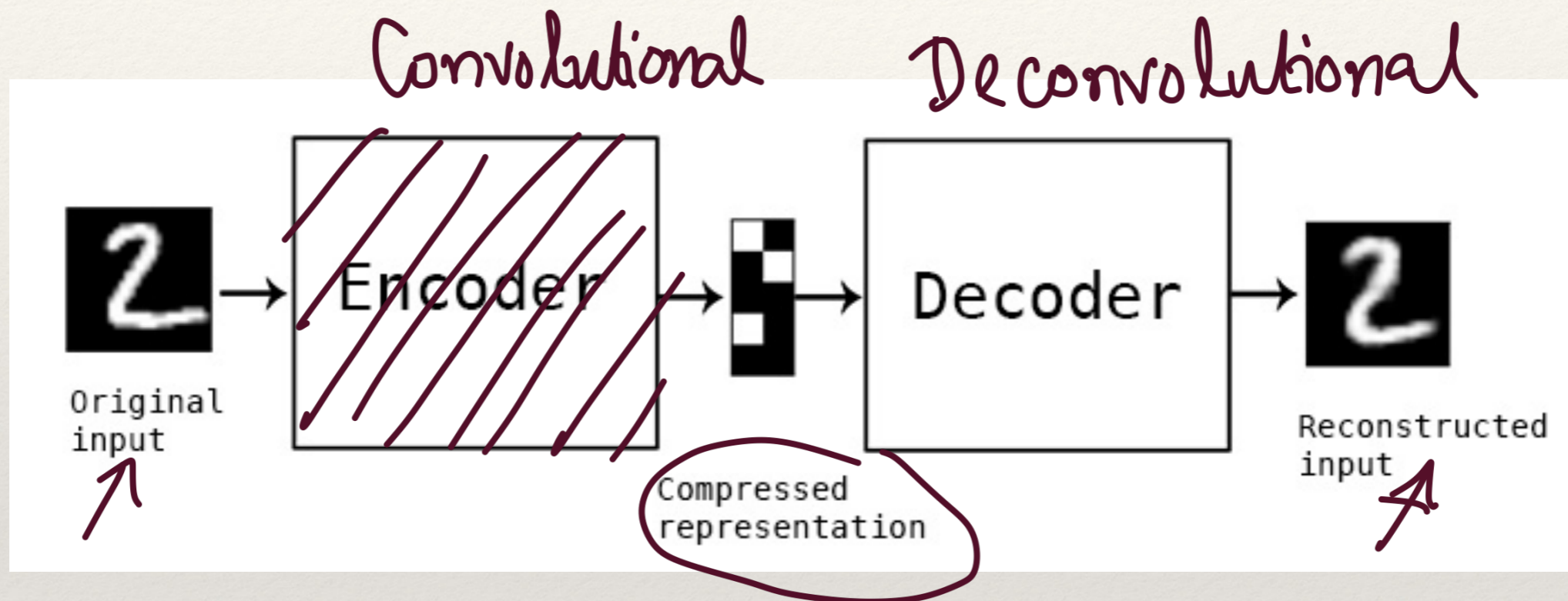


$$\mathcal{X} = \frac{1}{N} \mathcal{X}$$

Training loss

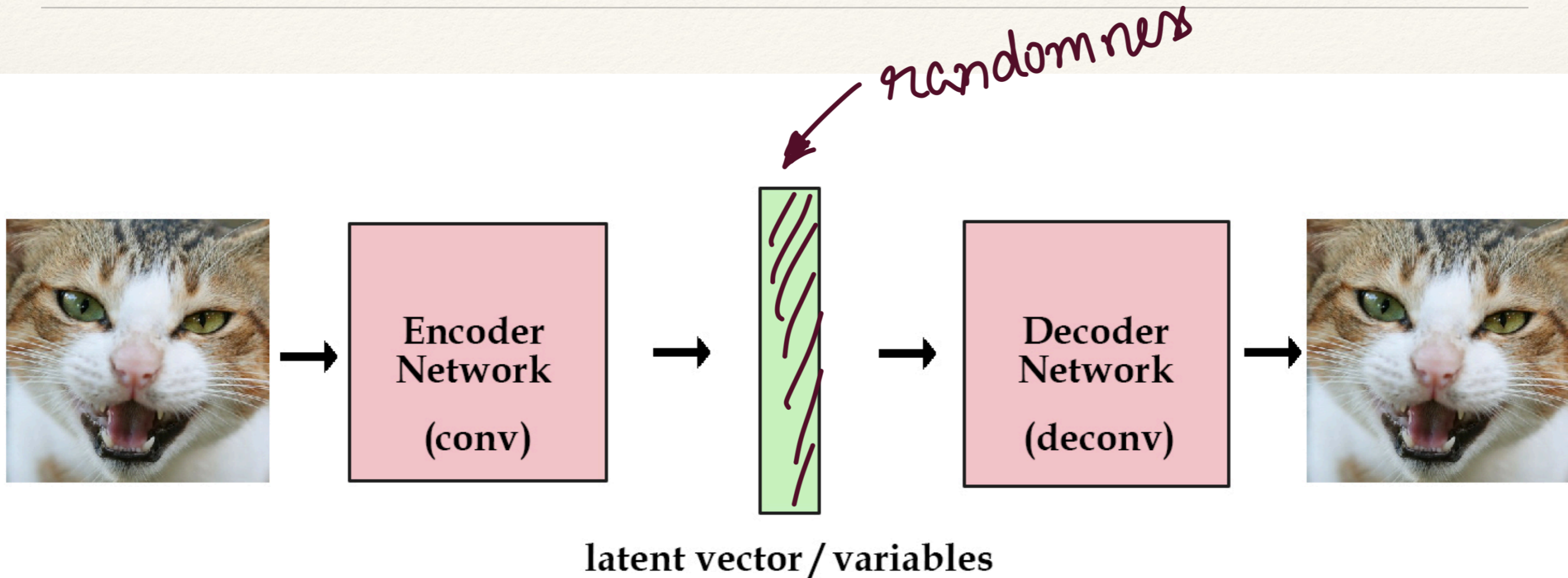
Avoid Identity Mapping

Autoencoders



Encoder \rightarrow LSTM, sequence to vector
Decoder \rightarrow LSTMs. vector to sequence

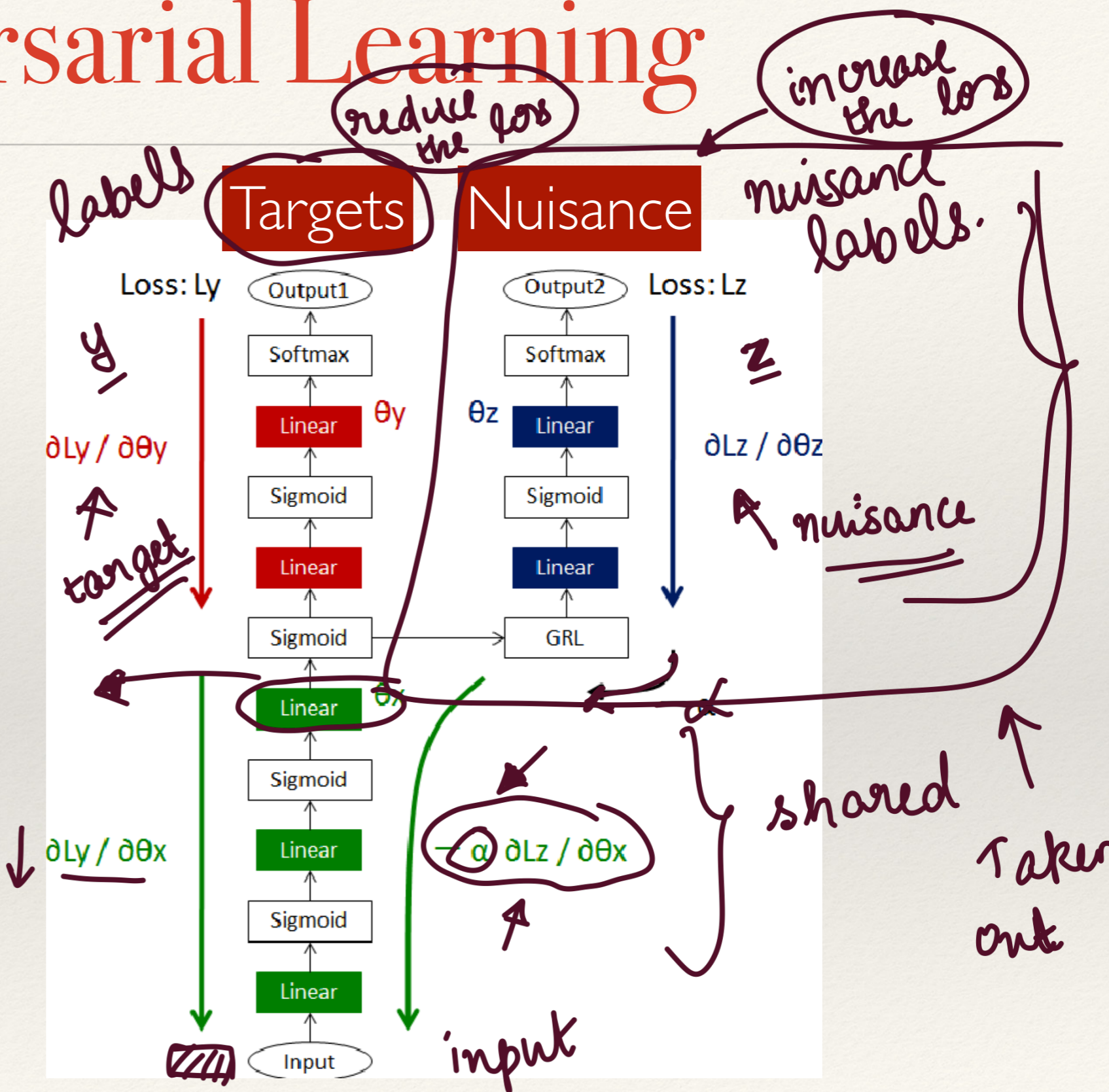
Convolutional Autoencoders



The latent vectors can form deep features for other supervised tasks.

Supervised Adversarial Learning

- The target can be learned
 - using original gradient.
 - domain adversarial gradient.
- Model will learn to be domain invariant

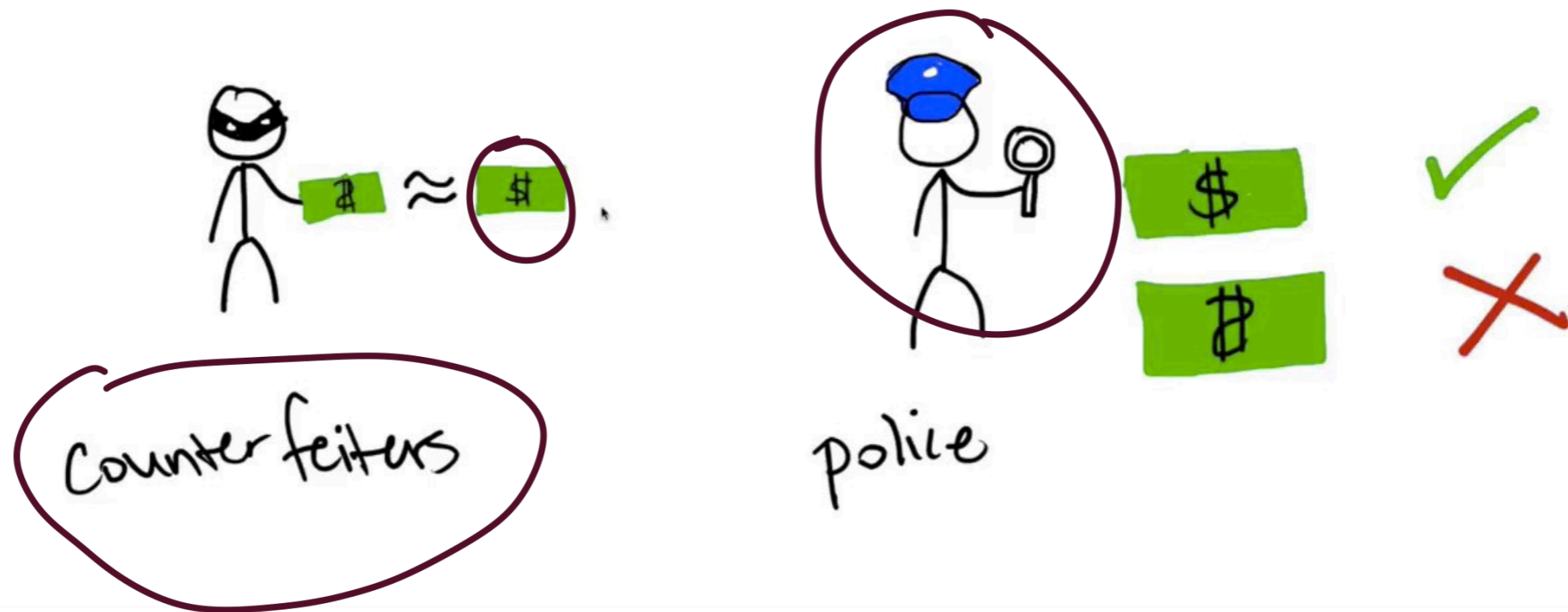


[Shinohara et al. 2017]

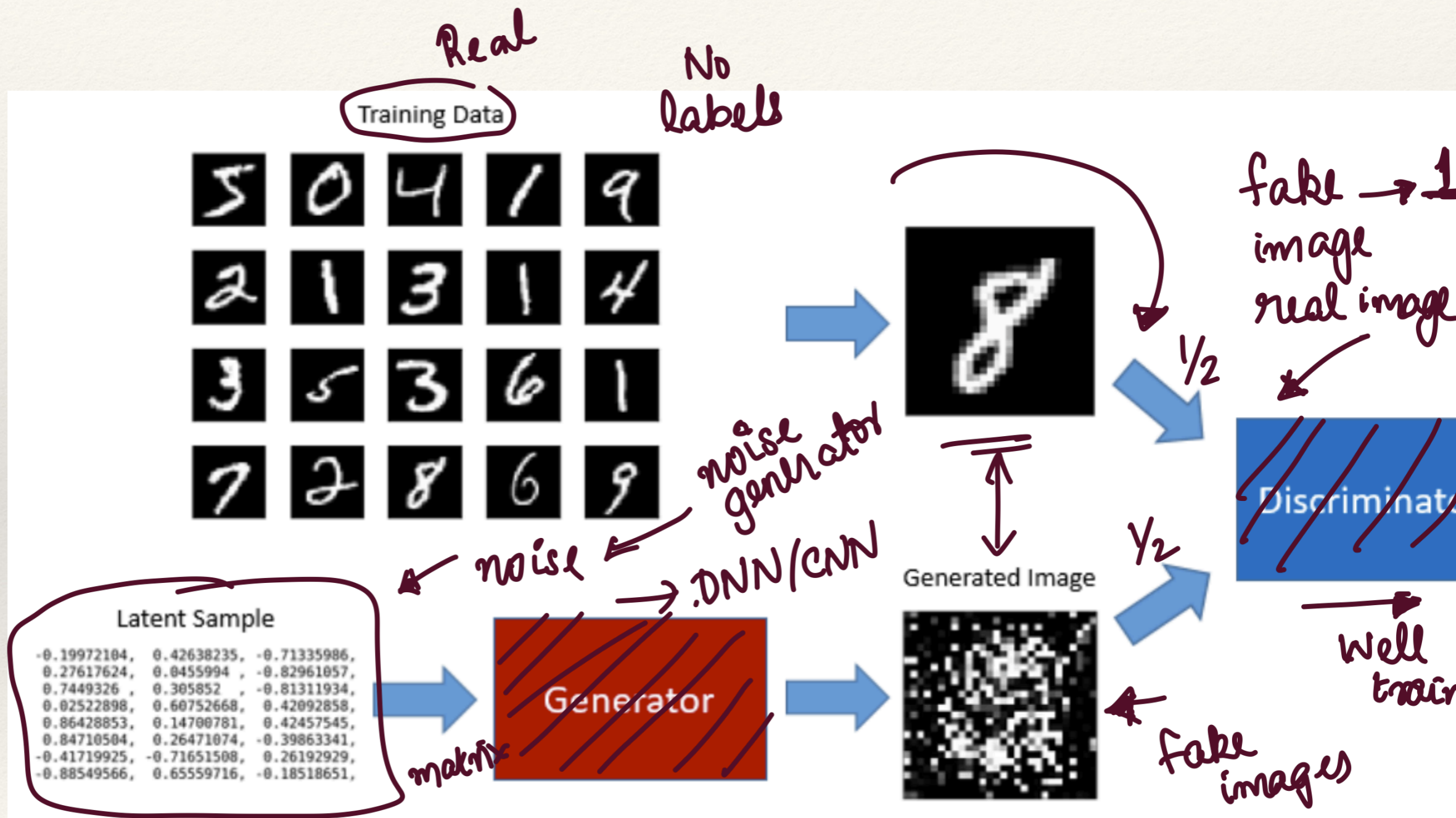
Generative Adversarial Networks

600 x 400

Generative Adversarial Networks (GANs)



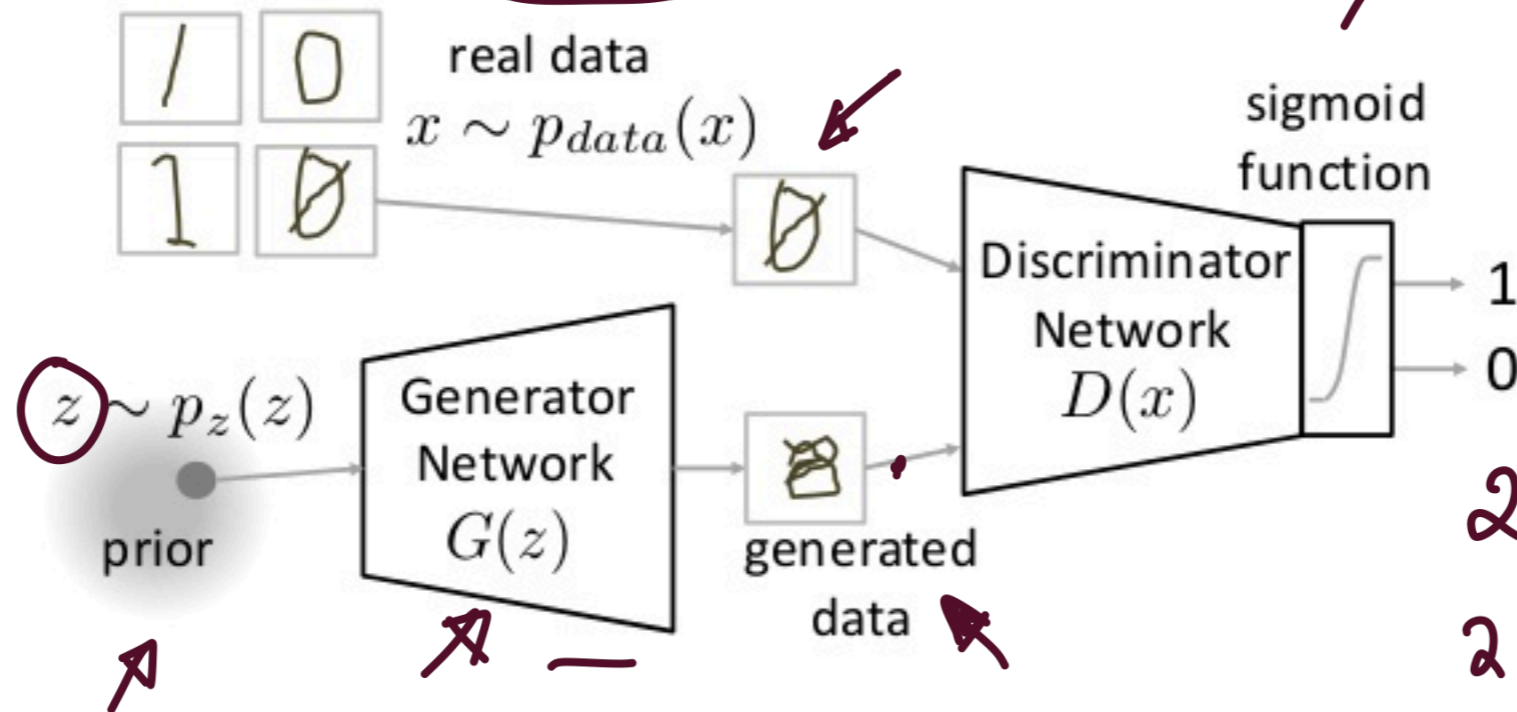
Generative Adversarial Networks



Generative Adversarial Networks

Generative Adversarial Networks

$$V(D, G) = \mathbb{E}_{x \sim p_{data}(x)} [\log D(x)] + \mathbb{E}_{z \sim p_z(z)} [\log(1 - D(G(z)))]$$



standard CE loss for Discrimination

Adversarial loss

image mapping

$G(z)$

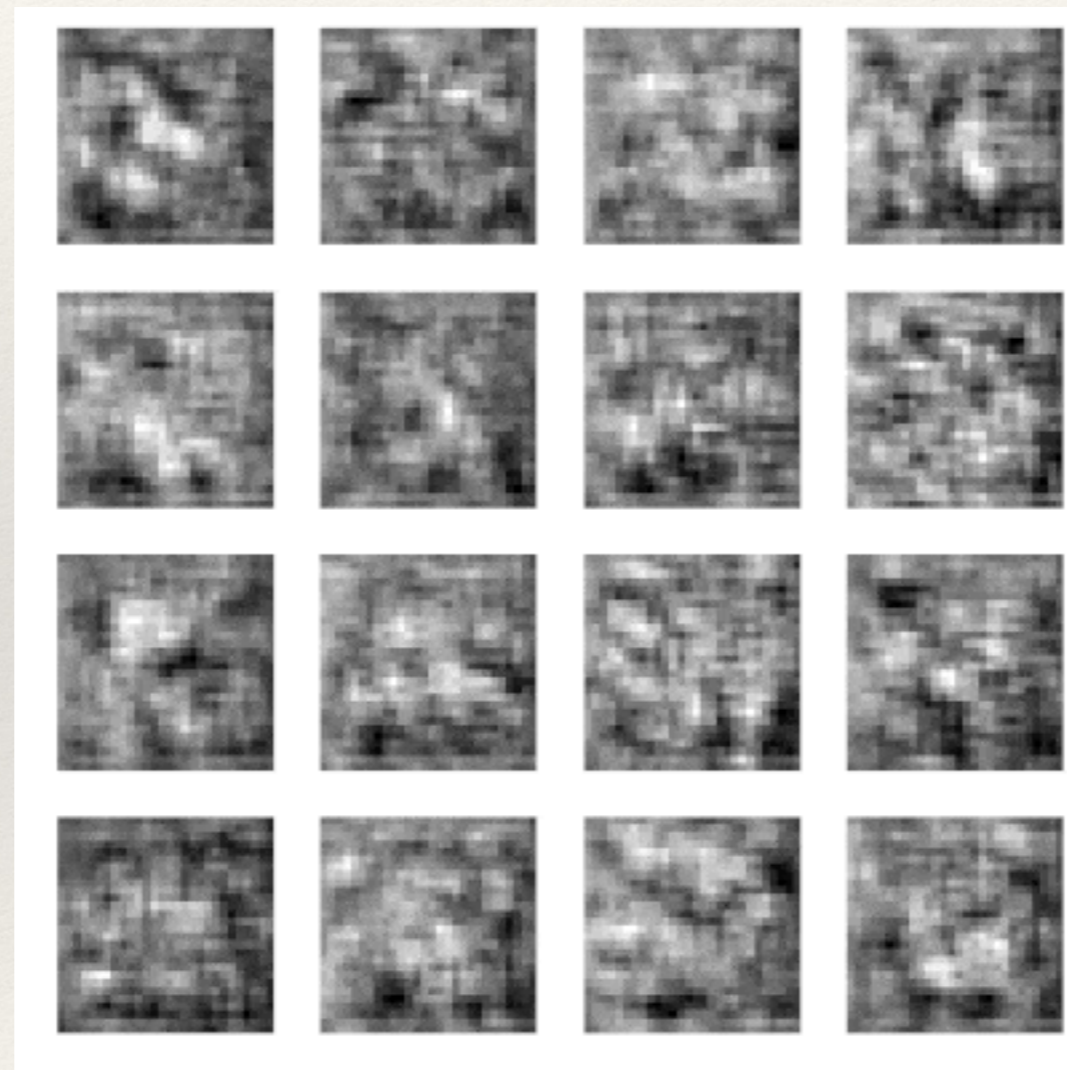
June 4th



usual
class
time

6:00pm

Generative Adversarial Networks



DCGANs