

Introduction to Deep Learning (CS474)

Lecture 27

Outline

Module 4

- **Introduction to Generative Adversarial Networks (GANs)**

Recap

To sum up, Discriminator (D) and Generator (G) are playing a “minimax” game with the comprehensive objective function:

$$\min_D \max_G \{ -E_{x \sim \text{Data}} \log D(\mathbf{x}) - E_{z \sim \text{Noise}} \log(1 - D(G(\mathbf{z}))) \}.$$

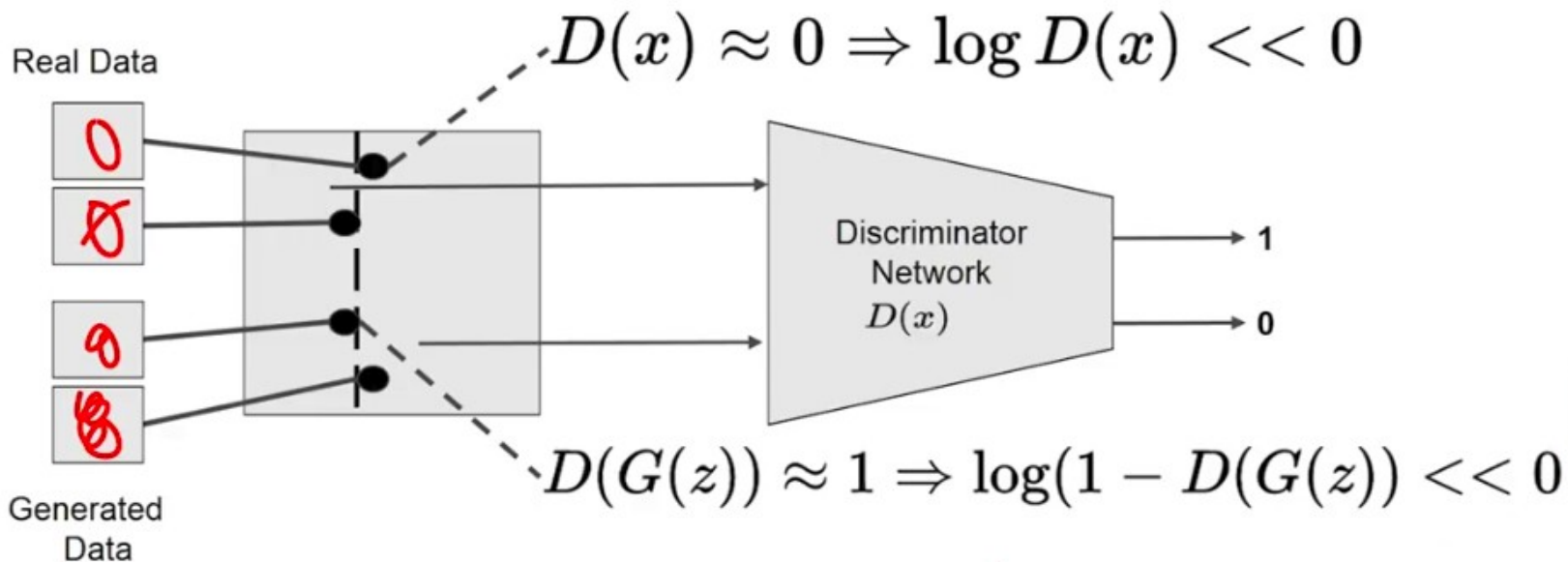
Training Discriminator Network

Before Training

$$\max_D [\mathbb{E}_{x \sim P_{data}(x)} \log(D(x)) + \mathbb{E}_{z \sim P_z(z)} \log(1 - D(G(z)))]$$

$D(x)$ should be 1

$D(G(z))$ should be 0



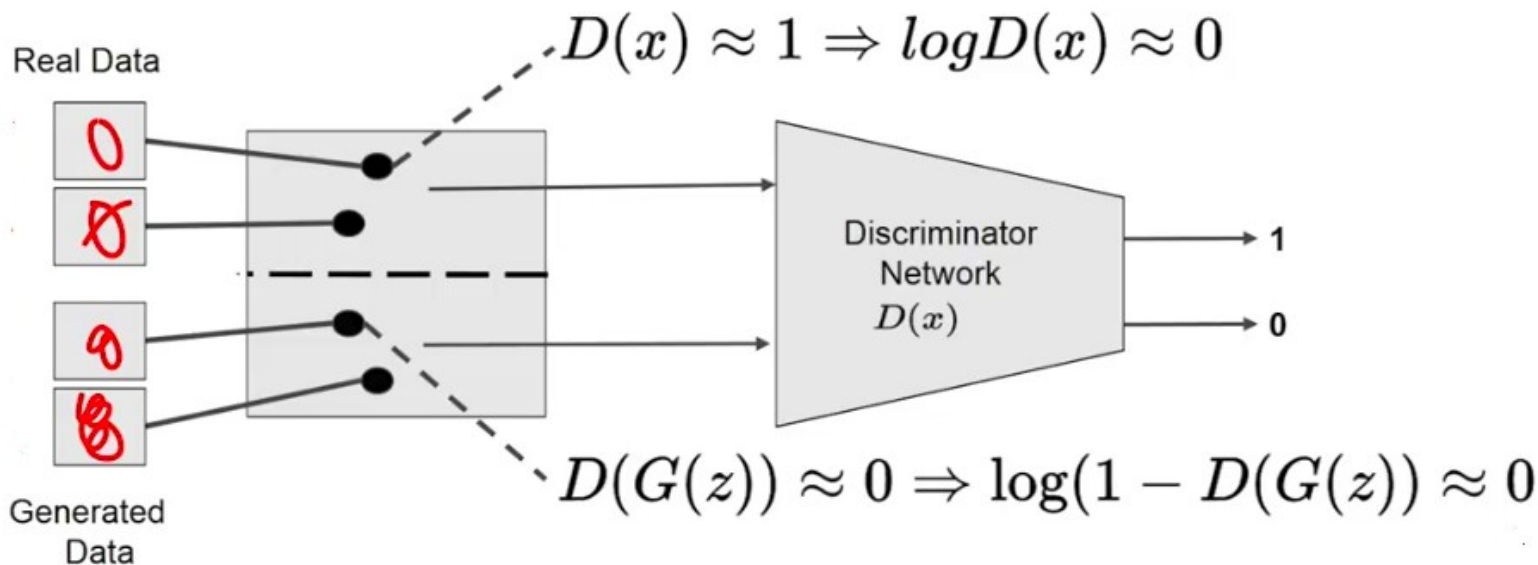
Training Discriminator Network

After Training

$$\max_D [\mathbb{E}_{x \sim P_{data}(x)} \boxed{\log(D(x))} + \mathbb{E}_{z \sim P_z(z)} \boxed{\log(1 - D(G(z)))}]$$

$D(x)$ should be 1

$D(G(z))$ should be 0



Training Generator Network

$$\min_G [\mathbb{E}_{x \sim P_{data}} (\log D(x)) + \mathbb{E}_{z \sim P(z)} (\log(1 - D(G(z))))]$$

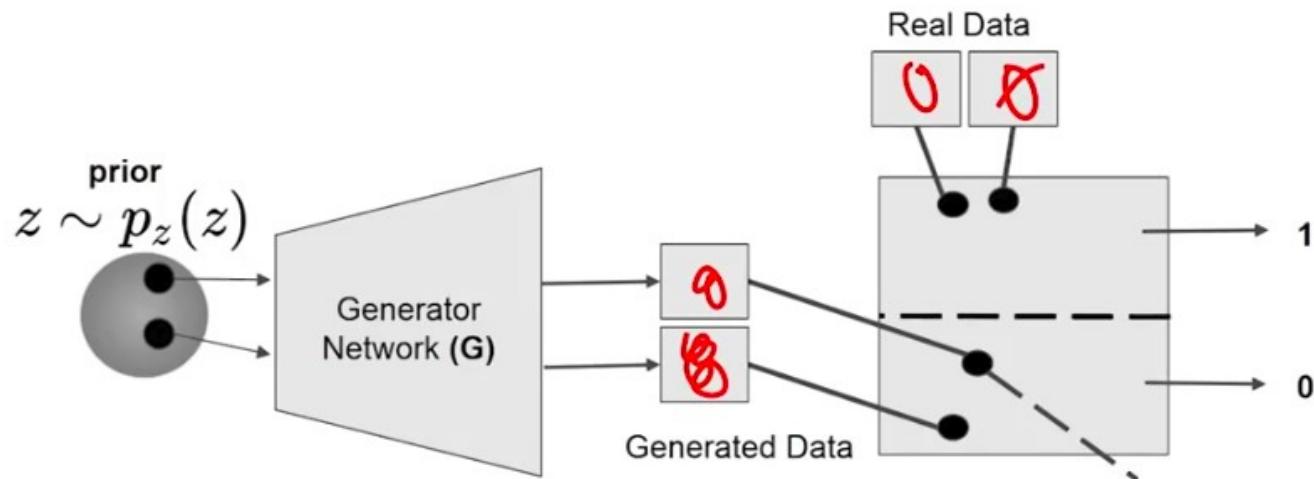
$$\Rightarrow \min_G [\mathbb{E}_{z \sim P(z)} (\log(1 - D(G(z))))]$$

Training Generator Network

Before Training

$$\max_G [\mathbb{E}_{z \sim P_z(z)} \log(D(G(z)))]$$

$D(G(z))$ should be 1



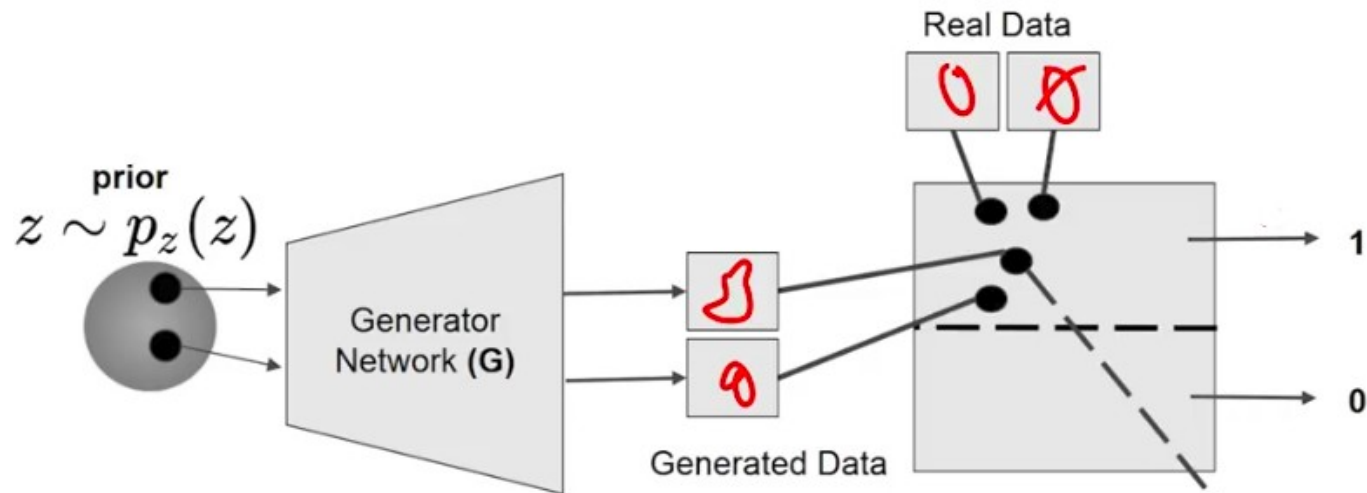
$$D(G(z)) \approx 0 \Rightarrow \log(D(G(z))) \ll 0$$

Training Generator Network

After Training

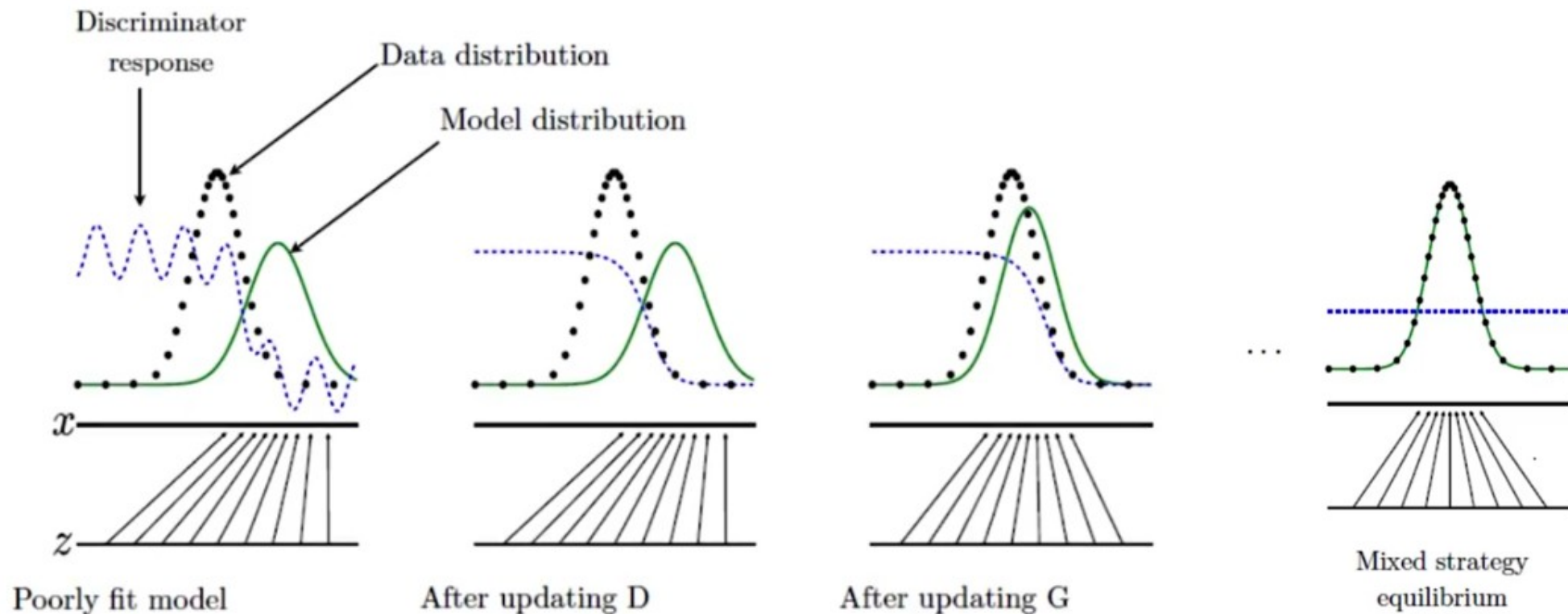
$$\max_G [\mathbb{E}_{z \sim P_z(z)} \log(D(G(z)))]$$

$D(G(z))$ should be 1



$$D(G(z)) \approx 1 \Rightarrow \log(D(G(z)))$$

Learning Process



Slide credit: Goodfellow, Ian, et al. "Generative adversarial nets." *Advances in neural information processing systems*. 2014.

References

- All the contents present in the slides are taken from various online resources. Due credit is given in the respective slides. These slides are used for *academic* purposes only.