

# Generative Adversarial Networks (Part 2)

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## Optimal Values in GANs

- ❖ GANs can be framed as a min-max problem where  $\hat{x}$  is the output of generator  $G$ :

$$\min_G \max_D V(G, D) = \mathbb{E}_{x \sim P_r(x)} [\log D(x)] + \mathbb{E}_{\hat{x} \sim P_r(\hat{x})} [\log(1 - D(\hat{x}))]$$

- ❖ The optimal discriminator  $D^*(x)$  is:

$$D^*(x) = \frac{P_r(x)}{P_r(x) + P_r(\hat{x})}$$

- ❖ For the optimal discriminator,  $V(G, D^*)$  becomes:

$$\begin{aligned} V(G, D^*) &= \mathbb{E}_{x \sim P_r(x)} \left[ \log \frac{P_r(x)}{P_r(x) + P_r(\hat{x})} \right] + \mathbb{E}_{\hat{x} \sim P_r(\hat{x})} \left[ \log \frac{P_r(\hat{x})}{P_r(x) + P_r(\hat{x})} \right] \\ &= \int P_r(x) \log \frac{P_r(x)}{\frac{P_r(x) + P_r(\hat{x})}{2}} dx + \int P_r(\hat{x}) \log \frac{P_r(\hat{x})}{\frac{P_r(x) + P_r(\hat{x})}{2}} d\hat{x} - \log 4 \\ &= D_{KL} \left[ P_r(x) \parallel \frac{P_r(x) + P_r(\hat{x})}{2} \right] + D_{KL} \left[ P_r(\hat{x}) \parallel \frac{P_r(x) + P_r(\hat{x})}{2} \right] - \log 4 \\ &= 2D_{JSD} [P_r(x) \parallel P_r(\hat{x})] - \log 4 \end{aligned}$$

## Jensen-Shannon Divergence

JSD is also known as symmetric KL divergence:

$$D_{JSD} [p \parallel q] = \frac{1}{2} \left( D_{KL} \left[ p \parallel \frac{p+q}{2} \right] + D_{KL} \left[ q \parallel \frac{p+q}{2} \right] \right)$$

Properties:

1.  $D_{JSD} [p \parallel q] \geq 0$
2.  $D_{JSD} [p \parallel q] = 0$  iff  $p = q$
3.  $D_{JSD} [p \parallel q] = D_{JSD} [q \parallel p]$

For GANs, when both  $G$  and  $D$  are at their optimal values:

$$P_{r(x)} = P_{r(\hat{x})}, \quad D^* = \frac{1}{2}, \quad V(G^*, D^*) = -\log 4$$