Generative Models Generative Adversarial Network GAN

IPRODAM3D

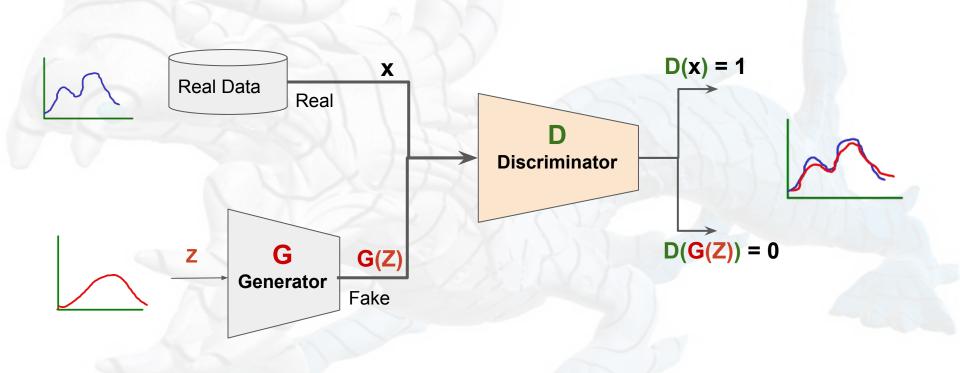
Prof. Cristian López Del Alamo

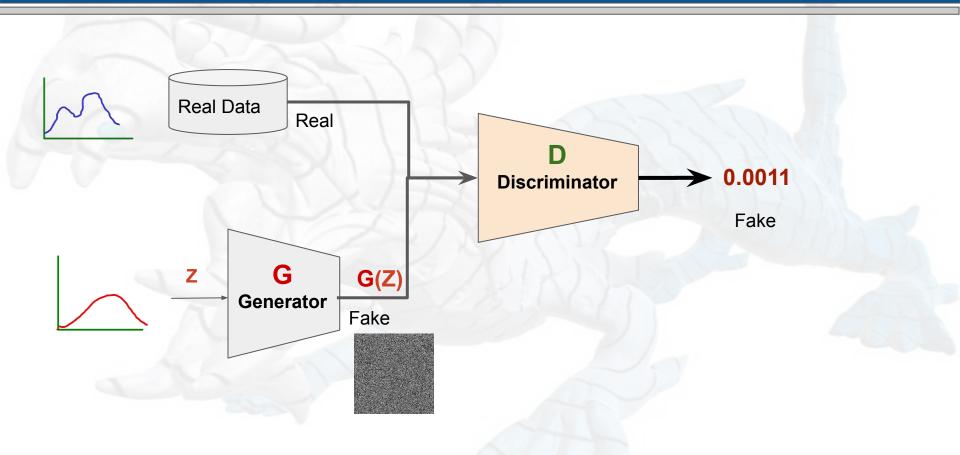


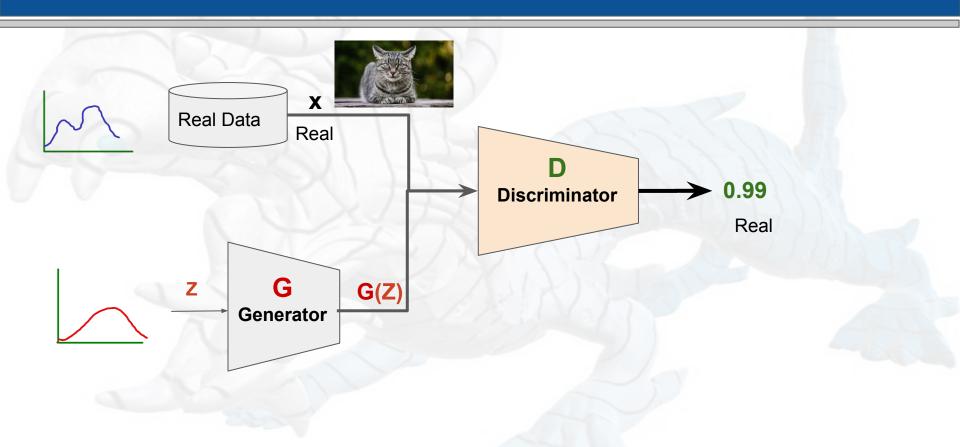
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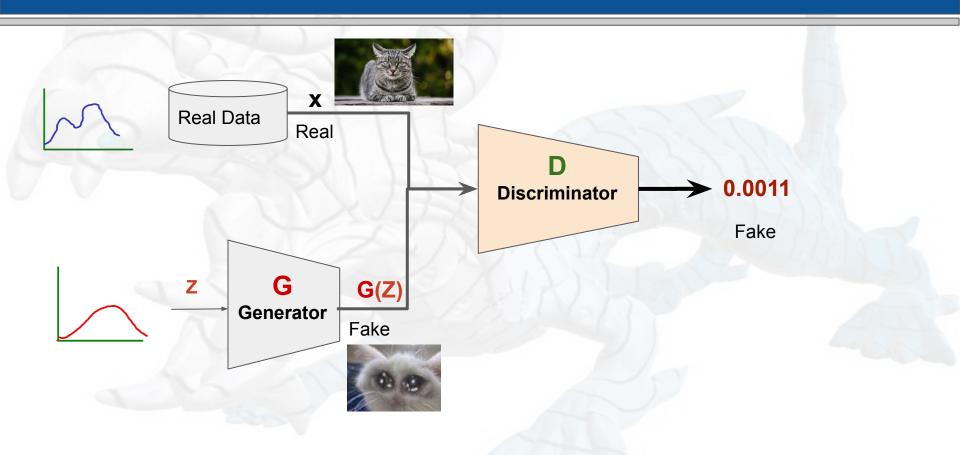


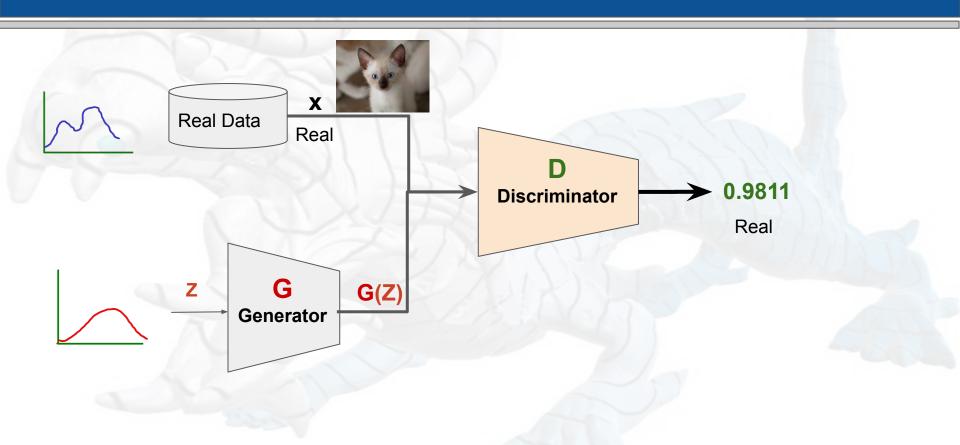
Face generation with generative modelling. Source: David Foster. "Generative Deep Learning."

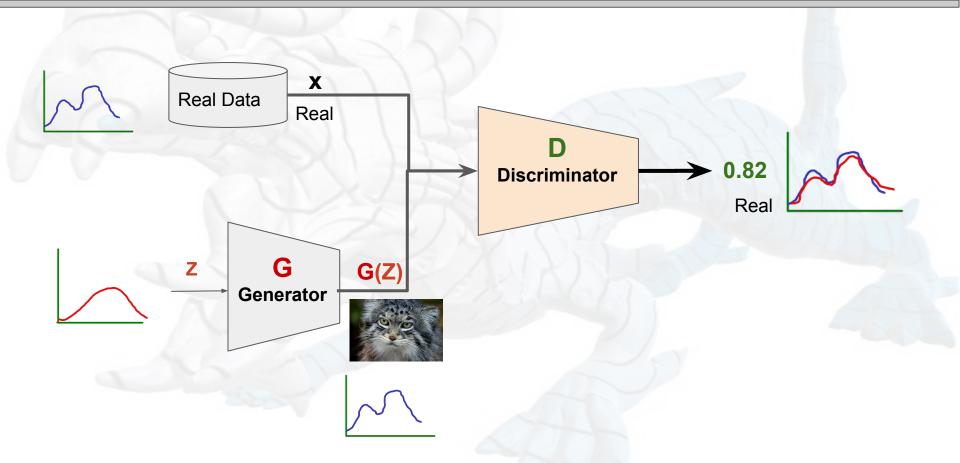












GAN: Loss Function

Loss Function Discriminator

$$\mathcal{L}_D(\theta_D, \theta_G) = -\frac{1}{2} E_{x \sim p_{data}(x)} [log(D(x))] - \frac{1}{2} E_{z \sim p_z(x)} [log(1 - D(G(z)))]$$

Positive classification

Negative classification

Loss Function Generator

$$\mathcal{L}_G(\theta_D, \theta_G) = \frac{1}{2} E_{x \sim p_{data}(x)} [log(D(x))] + \frac{1}{2} E_{z \sim p_z(x)} [log(1 - D(G(z)))]$$

Positive classification

Negative classification

GAN: Loss Function

$$\mathcal{L}_G(\theta_D, \theta_G) = -\mathcal{L}_D(\theta_D, \theta_G)$$

$$\min_{\mathsf{G}} \max_{\mathsf{D}} = -\mathcal{L}_D(\theta_D, \theta_G)$$

GAN: Loss Function Discriminator

$$\mathcal{L}_{D}(\theta_{D}, \theta_{G}) = -\frac{1}{2} E_{x \sim p_{data}(x)} [log(D(x))] - \frac{1}{2} E_{z \sim p_{z}(x)} [log(1 - D(G(z)))]$$

$$\mathcal{L}_D(\theta_D, \theta_G) = -\frac{1}{2} \left[\int_x p_{data}(x) log(D(x)) dx + \int_z p_z(z) log(1 - D(G(z))) dz \right]$$

GAN: Loss Function Discriminator

$$\mathcal{L}_D(\theta_D, \theta_G) = -\frac{1}{2} \left[\int_x p_{data}(x) log(D(x)) dx + \int_x p_G(x) log(1 - D(x)) dx \right]$$

$$\mathcal{L}_D(\theta_D, \theta_G) = -\frac{1}{2} \left[\int_x p_{data}(x) log(D(x)) dx + p_G(x) log(1 - D(x)) dx \right]$$

$$\frac{\partial \mathcal{L}_D(\theta_D, \theta_G)}{\partial D(x)} = \frac{p_{data}(x)}{D(x)} - \frac{p_G(x)}{1 - D(x)} \qquad \qquad \frac{p_{data}(x)}{D(x)} - \frac{p_G(x)}{1 - D(x)} = 0$$

$$D^*(x) = \frac{p_{data}(x)}{p_{data}(x) + p_G(x)}$$

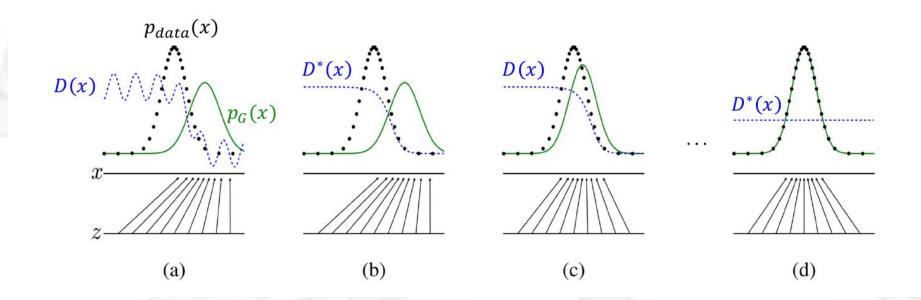
GAN: Loss Function Discriminator

Replacing D(x) with optimo $D^*(x)$

$$\mathcal{L}_D(\theta_D, \theta_G) = -\frac{1}{2} \left[\int_x p_{data}(x) log(\frac{p_{data}(x)}{p_{data}(x) + p_g(x)}) dx + p_G(x) log(\frac{p_{data}(x) + p_g(x)}{p_{data}(x)}) dx \right]$$

$$\mathcal{L}_D(\theta_D, \theta_G) = -\frac{1}{2} \left[\int_x p_{data}(x) log(\frac{p_{data}(x)}{\frac{p_{data}(x) + p_g(x)}{2}}) dx + p_G(x) log(\frac{p_{data}(x) + p_g(x)}{\frac{p_{data}(x)}{2}}) dx - log(4) \right]$$

$$\mathcal{L}_{D}(\theta_{D}, \theta_{G}) = -\frac{1}{2} \left[D_{KL} \left(p_{data}(x) || \frac{p_{data}(x) + p_{g}(x)}{2} \right) + D_{KL} \left(p_{G}(x) || \frac{p_{data}(x) + p_{g}(x)}{2} \right) - \log(4) \right]$$



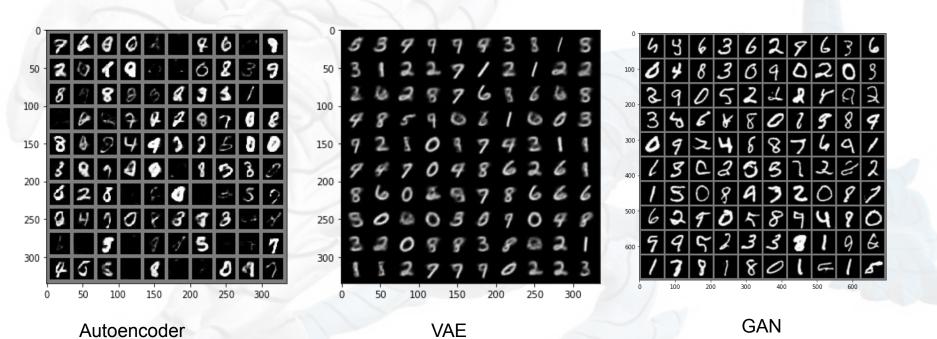
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- Initialize $\theta^{(G)}$, $\theta^{(D)}$
- For t = 1: b: T
 - Initialize $\Delta \theta^{(D)} = 0$
 - For i = t: t + b 1
 - Sample $z_i \sim p(z_i)$
 - Compute $D(G(z_i)), D(x_i)$
 - $\Delta\theta_i^{(D)} \leftarrow \text{Compute gradient of Discriminator loss}, J^{(D)}(\theta^{(G)}, \theta^{(D)})$
 - $\Delta \theta^{(D)} \leftarrow \Delta \theta^{(D)} + \Delta \theta_i^{(D)}$
 - Update $\theta^{(D)}$

Source click

GAN

- Initialize $\Delta \theta^{(G)} = 0$
- For $j = t \cdot t + h 1$
 - Sample $z_j \sim p(z_j)$
 - Compute $D\left(G\left(z_{j}\right)\right)$, $D\left(x_{j}\right)$
 - $\Delta \theta_j^{(G)} \leftarrow \text{Compute gradient of Generator loss, } J^{(G)}(\theta^{(G)}, \theta^{(D)})$
 - $\bullet \quad \Delta \theta^{(G)} \leftarrow \Delta \theta^{(G)} + \Delta \theta_j^{(G)}$
- Update $\theta^{(G)}$

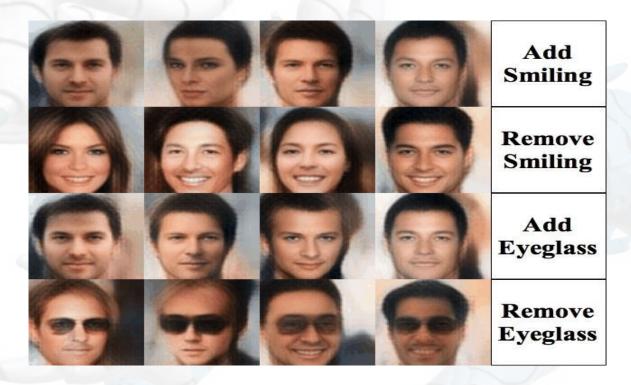


Pytorch example

GAN pytorch code example: souce click

GAN in action : source click

Applications



Fuente: https://medium.com/analytics-vidhya/an-introduction-to-generative-deep-learning-792e93d1c6d4

Applications



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IS THIS FACE REAL OR FACE?



Thank alot for your participation

Modelos Generativos Autoencoders GAN

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