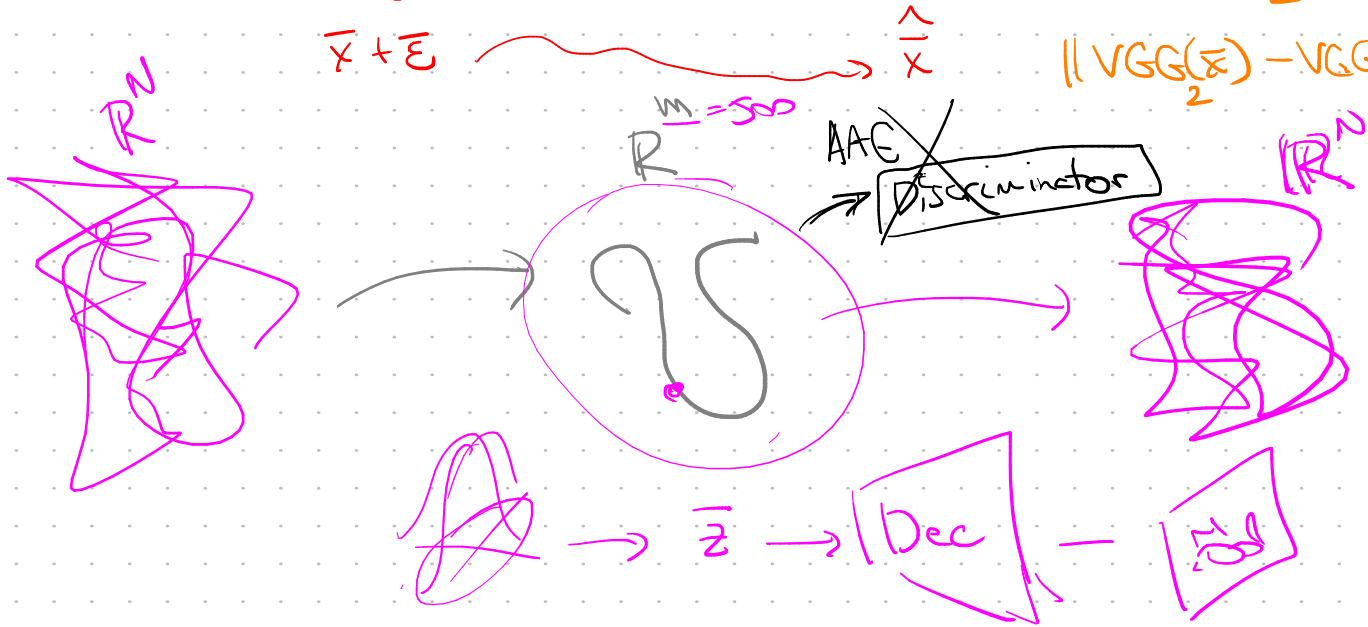
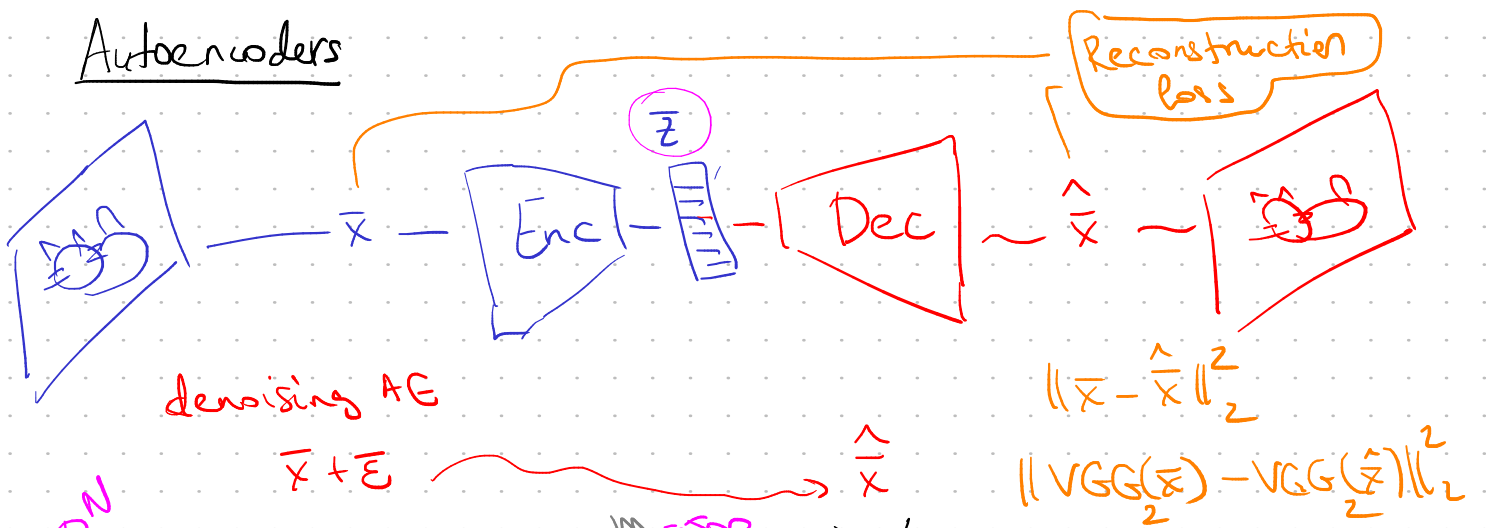
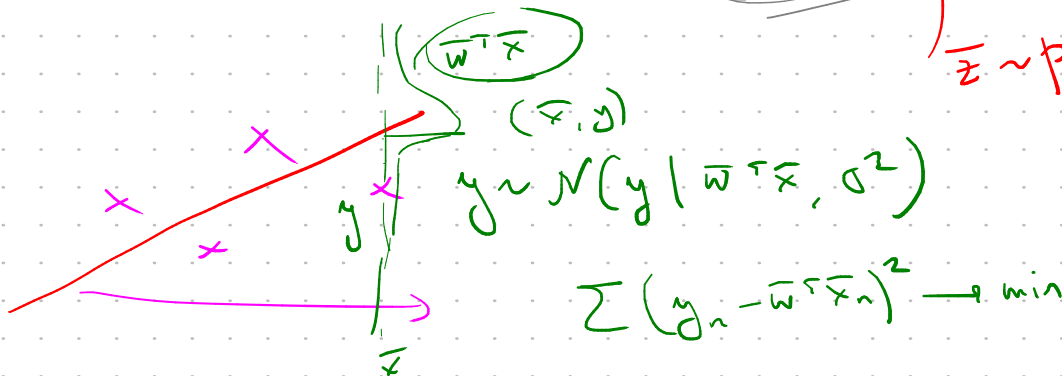
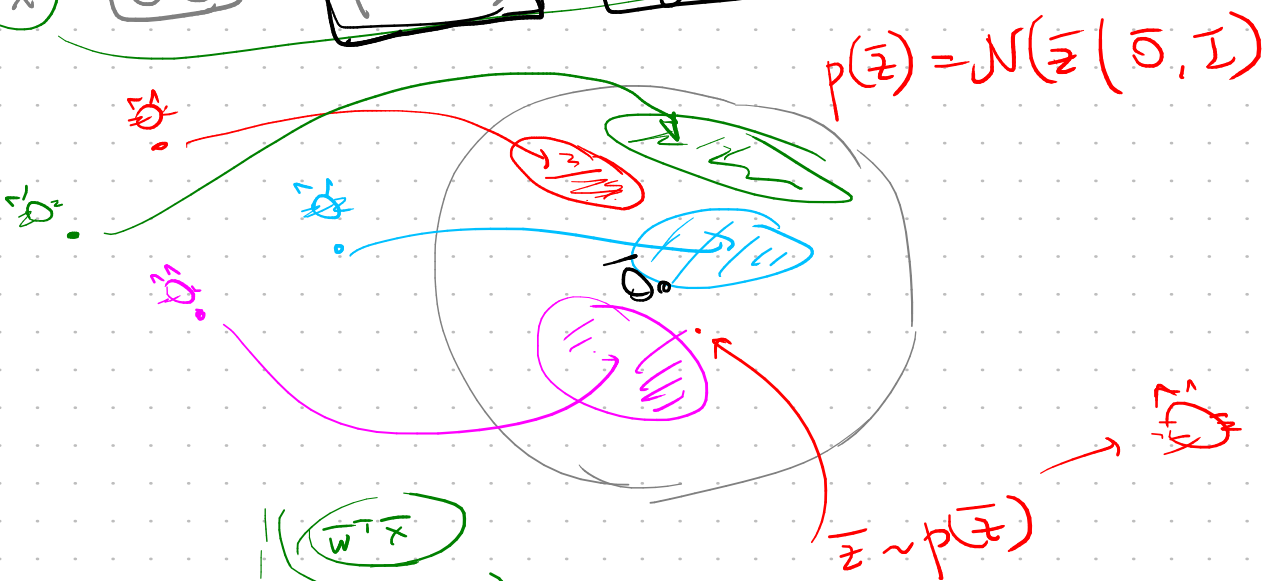
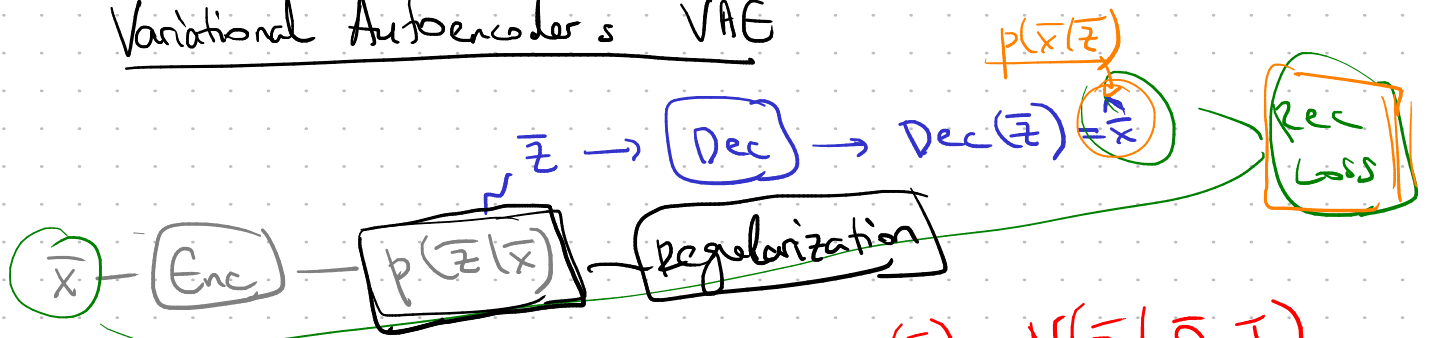


Autoencoders



Variational Autoencoders VAE



$$y \sim \mathcal{N}(y | \bar{w}^T x, \sigma^2)$$

$$\sum (y_n - \bar{w}^T x_n)^2 \rightarrow \min$$

$$\underbrace{p(\bar{z})}_{\mathcal{N}(\bar{z}|\mathbf{0}, \mathbf{I})} \underbrace{p(\bar{x}|\bar{z})}_{\text{Decoder}} = \underbrace{p(\bar{x}, \bar{z})}_{\mathcal{N}} = \underbrace{p(\bar{x})}_{\mathbb{R}^N} \underbrace{p(\bar{z}|\bar{x})}_{\text{Encoder}} \approx q(\bar{z})$$

↑
Variational approximation

$$p(\bar{x}|\bar{z}) = \mathcal{N}(\bar{x} | f(\bar{z}), c\mathbf{I})$$

" Decoder(\bar{z})

$$\log p(\bar{x}, \bar{z}) = \log p(\bar{x}) \log p(\bar{z}|\bar{x})$$

$$\log p(\bar{x}) = \log p(\bar{x}, \bar{z}) - \log p(\bar{z}|\bar{x}) \quad \left. \vphantom{\log p(\bar{x})} \right\} \mathbb{E}_{q(\bar{z})}$$

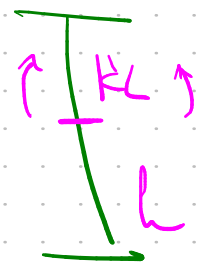
$$\log p(\bar{x}) = \mathbb{E}_q[\log p(\bar{x}, \bar{z})] - \mathbb{E}_q[\log p(\bar{z}|\bar{x})] + \mathbb{E}_q[\log q(\bar{z})]$$

$$\log p(\bar{x}) = \mathbb{E}_{q(\bar{z})} \left[\log \frac{p(\bar{x}, \bar{z})}{q(\bar{z})} \right] = \mathbb{E}_{q(\bar{z})} \left[\log \frac{p(\bar{z}|\bar{x})}{q(\bar{z})} \right]$$

$\mathcal{L}(q)$
 Variational lower bound
 ELBO

$$\int q(\bar{z}) \log \frac{q(\bar{z})}{p(\bar{z}|\bar{x})} d\bar{z} = \text{KL}(q || p(\bar{z}|\bar{x}))$$

$$\text{Cost} = \underbrace{\mathcal{L}(q)}_{\rightarrow \max} + \underbrace{\text{KL}(q(\bar{z}) || p(\bar{z}|\bar{x}))}_{\rightarrow \min}$$



$$\mathcal{L}(q) = \int q(\bar{z}) \log \frac{p(\bar{x}, \bar{z})}{q(\bar{z})} d\bar{z} =$$

" $p(\bar{z})p(\bar{x}|\bar{z})$ "

$$= \underbrace{\int q(\bar{z}) \log p(\bar{x}(\bar{z})) d\bar{z}}_{\text{"}} - \underbrace{\int q(\bar{z}) \log \frac{q(\bar{z})}{p(\bar{z})} d\bar{z}}_{\text{"}}$$

$$E_{q(\bar{z})} [\log N(\bar{x} | f(\bar{z}), \sigma^2 I)]$$

$$\text{KL}(q(\bar{z}) || p(\bar{z}))$$

$$E_q \left[\text{Const} - \frac{1}{2c} \cdot \|\bar{x} - f(\bar{z})\|^2 \right]$$

$$\text{Const} - \frac{1}{2c} \cdot \sum_n \|\bar{x}_n - \underbrace{f}_{\text{Dec}}(\text{Enc}(\bar{x}_n))\|^2$$

$$N(\bar{z} | \bar{0}, I)$$

$$h(q) = \text{Const} - \underbrace{\left(\frac{1}{2c} \cdot E_q [\|\bar{x} - f(\bar{z})\|^2] \right)}_{\text{Reconstruction loss}} + \underbrace{\text{KL}(q(\bar{z}) || p(\bar{z}))}_{\text{Regulariser}}$$

$$q(\bar{z}) = N(\bar{z} | \bar{\mu}_x, \bar{\Sigma}_x)$$

$$p(\bar{z}) = N(\bar{z} | \bar{0}, I)$$

$$\begin{pmatrix} \sigma_{x_1} & 0 \\ 0 & \sigma_{x_d} \end{pmatrix}$$

$$\text{KL}(q || p) = \text{KL} \left(\prod_i q_i(z_i) || \prod_i p_i(z_i) \right) =$$

$$= \sum_{i=1}^d \text{KL}(q_i(z_i) || p_i(z_i))$$

$$\text{KL} \left(N(z_i | \mu_{x_i}, \sigma_{x_i}^2) || N(z_i | 0, 1) \right) =$$

$$= \int \frac{1}{\sqrt{2\pi\sigma_{x_i}^2}} e^{-\frac{1}{2\sigma_{x_i}^2}(z_i - \mu_{x_i})^2} \cdot \log \frac{N(-)}{N(-)} dz_i =$$

$$= E_{p(z_i | \mu_{x_i}, \sigma_{x_i}^2)} \left[-\frac{1}{2} \log \sigma_{x_i}^2 - \frac{1}{2\sigma_{x_i}^2} (z_i - \mu_{x_i})^2 + \frac{1}{2} z_i^2 \right] =$$

$$= E_{N(z_i | \mu_{x_i}, \sigma_{x_i}^2)} \left[\left(-\frac{1}{2\sigma_{x_i}^2} + \frac{1}{2} \right) z_i + \frac{\mu_{x_i}}{\sigma_{x_i}^2} z_i \right] - \frac{1}{2} \log \sigma_{x_i}^2 - \frac{\mu_{x_i}^2}{2\sigma_{x_i}^2}$$

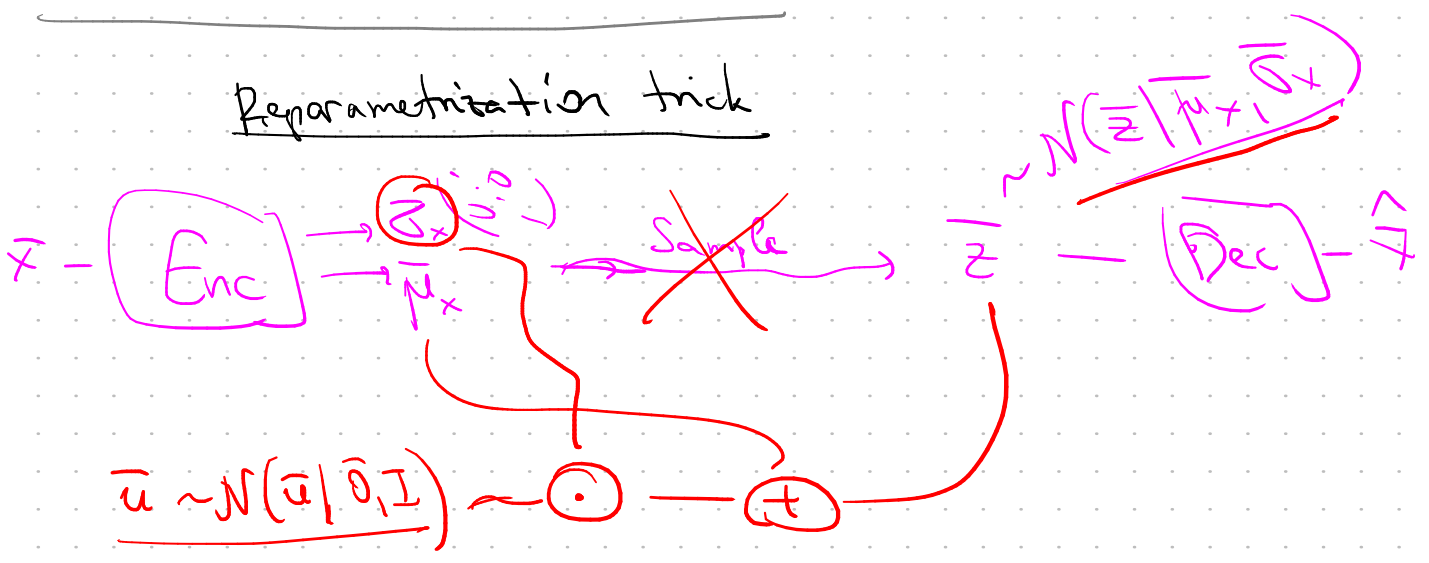
$$E[z_i^2] = E[z_i]^2 + \text{Var}[z_i]$$

$$= \left(\frac{1}{2} - \frac{1}{2\sigma_{x_i}^2} \right) (\mu_{x_i}^2 + \sigma_{x_i}^2) + \frac{\mu_{x_i}^2}{2\sigma_{x_i}^2} - \frac{1}{2} \log \sigma_{x_i}^2 - \frac{\mu_{x_i}^2}{2\sigma_{x_i}^2} =$$

$$= \frac{1}{2} \mu_{x_i}^2 + \frac{1}{2} \sigma_{x_i}^2 - \frac{1}{2} \log \sigma_{x_i}^2 - \frac{1}{2}$$

$$KL(q || p) = \sum_{z_i} \frac{d}{2} (\mu_{x_i}^2 + \sigma_{x_i}^2 - \log \sigma_{x_i}^2 - 1)$$

Reparametrization trick



$$\Rightarrow \bar{u} \odot \bar{\sigma}_x + \bar{\mu}_x \sim N(\cdot | \bar{\mu}_x, \bar{\sigma}_x)$$

