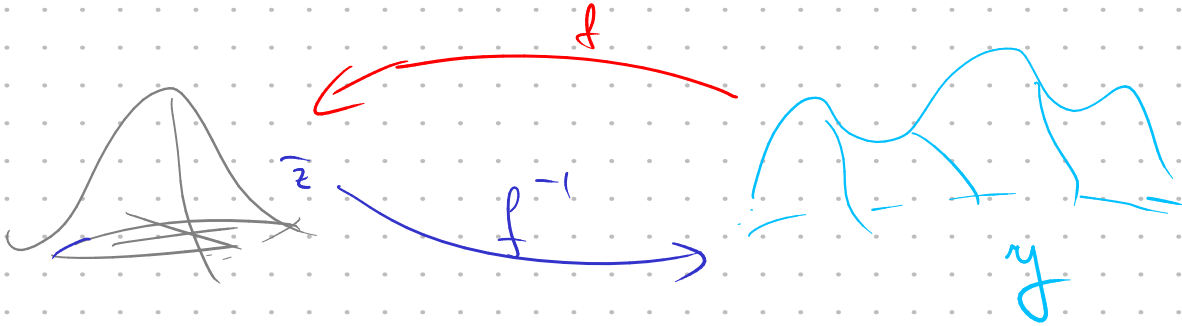
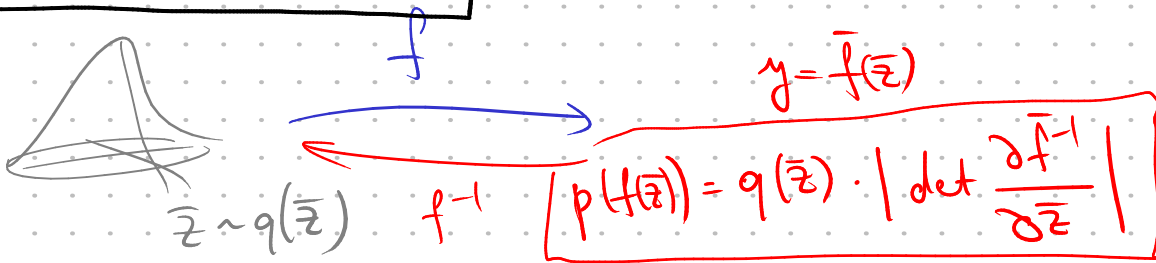


Flow-based models



$$\bar{z}_k = f_k \circ f_{k-1} \circ \dots \circ f_1(\bar{z}_0) \quad \bar{z}_0 \sim q_0(\bar{z}_0)$$

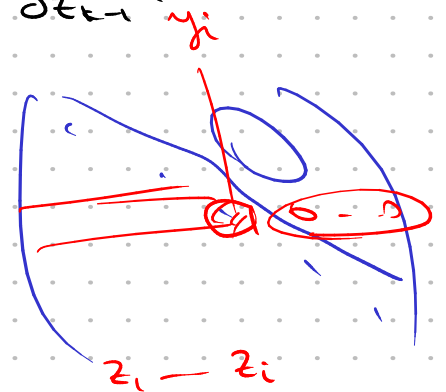
$$\bar{z}_k \sim \underbrace{q_k(\bar{z})}_{\text{update}(z)} = q_0(\bar{z}_0) \prod_{k=1}^k \left| \det \frac{\partial f_k}{\partial \bar{z}_{k-1}} \right|^{-1}$$

$$\log q_k = \log q_0 - \sum_k \log \left| \det \frac{\partial f_k}{\partial \bar{z}_{k-1}} \right|$$

$f: \det \frac{\partial f}{\partial \bar{z}}$

$$\bar{y} = f(\bar{z})$$

$$(y_1, \dots, y_d) = f(z_1, \dots, z_d)$$



$$y_i = f(z_1, \dots, z_i)$$

$$\det \frac{\partial f}{\partial \bar{z}} = \prod_{i=1}^d \frac{\partial y_i}{\partial z_i}$$

— autoregressive flows

R-NVP real non-volume-preserving flows

$$\bar{y}_{1:k} \rightarrow \bar{z}_{1:k}$$

$$\bar{y}_{k+1:d} = \bar{z}_{k+1:d} \odot \sigma(\bar{z}_{1:k}) + \mu(\bar{z}_{1:k})$$



$$\det \frac{\partial \bar{y}}{\partial \bar{z}} = \prod_{i=k+1}^d \sigma_i(\bar{z}_{1:k})$$

$$\bar{y}_{k+1:d} = \bar{z}_{k+1:d} \circ \sigma(\bar{z}_{1:k}) + \bar{\mu}(\bar{z}_{1:k})$$

$$\bar{z}_{k+1:d} = \left(\bar{y}_{k+1:d} - \bar{\mu}(\bar{z}_{1:k}) \right) \circ \frac{1}{\sigma(\bar{z}_{1:k})}$$

$\frac{1}{\sigma(\bar{z}_{1:k})} = \frac{1}{\prod_{i=k+1}^d \sigma_i(\bar{z}_{1:k})}$

Masked Autoregressive Flows (MAF)

$$p(\bar{x}) = \prod_{i=1}^D p(x_i | \bar{x}_{1:i-1})$$

unpaired functions

$$p(x_i | \bar{x}_{1:i-1}) = z_i \cdot \sigma_i(\bar{x}_{1:i-1}) + \mu_i(\bar{x}_{1:i-1})$$

$$\bar{z} \sim \pi(\bar{z}) \xrightarrow{f} \bar{x}$$

$$z_i = \frac{x_i - \mu_i(\bar{x}_{1:i-1})}{\sigma_i(\bar{x}_{1:i-1})}$$

$$\bar{z} = \dots(\bar{x})$$

$$x_1 = z_1 \cdot \sigma_1 + \mu_1$$

$$x_2 = z_2 \cdot \sigma_2(x_1) + \mu_2(x_1)$$

$$x_d = z_d \cdot \sigma_d(\bar{x}_{1:d-1}) + \mu_d(\bar{x}_{1:d-1})$$

Inverse Autoregressive Flows (IAF)

$$x_i \sim p(x_i | \bar{z}_{1:i}) = z_i \cdot \sigma_i(\bar{z}_{1:i-1}) + \mu_i(\bar{z}_{1:i-1})$$

$$z_i = \frac{x_i - \mu_i(\bar{z}_{1:i-1})}{\sigma_i(\bar{z}_{1:i-1})}$$

$$\bar{z} \sim \pi(\bar{z}) \xrightarrow{f} \bar{x} = \bar{z} \circ \sigma(\bar{z}) + \bar{\mu}(\bar{z})$$

$$z_1 = \dots$$

$$z_2 = \dots(z_1)$$

$$z_d = \dots(z_1, \dots, z_{d-1})$$

Diffusion-based models

$$\bar{x}_0 \sim q(\bar{x})$$

$$\hookrightarrow \bar{x}_1 \rightarrow \bar{x}_2 \rightarrow \dots \rightarrow \bar{x}_T$$

$$q(\bar{x}_t | \bar{x}_{t-1}) = \mathcal{N}(\bar{x}_t | \sqrt{1-\beta_t} \bar{x}_{t-1}, \beta_t \mathbf{I})$$

$$q(\bar{x}_1, \dots, \bar{x}_T | \bar{x}_0) = \prod_t q(\bar{x}_t | \bar{x}_{t-1})$$

Reparametrization trick

$$\bar{x}_t = \sqrt{1-\beta_t} \bar{x}_{t-1} + \sqrt{\beta_t} \bar{\epsilon}_{t-1}$$

$$\bar{\epsilon}_{t-1} \sim \mathcal{N}(\bar{0}, \mathbf{I})$$

$$= \sqrt{\alpha_t} \bar{x}_{t-1} + \sqrt{1-\alpha_t} \bar{\epsilon}_{t-1}$$

$$= \sqrt{\alpha_t \alpha_{t-1}} \bar{x}_{t-2} + \sqrt{1-\alpha_t} \bar{\epsilon}_{t-1} + \sqrt{\alpha_t (1-\alpha_{t-1})} \bar{\epsilon}_{t-2}$$

$$\sim \mathcal{N}(\bar{0}, (1-\alpha_t)\mathbf{I}) + \mathcal{N}(\bar{0}, \alpha_t(1-\alpha_{t-1})\mathbf{I})$$

$$= \sqrt{\alpha_t \alpha_{t-1}} \bar{x}_{t-2} + \sqrt{1-\alpha_t \alpha_{t-1}} \bar{\epsilon}'_{t-2} \sim \mathcal{N}(\bar{0}, (1-\alpha_t + \alpha_t - \alpha_t \alpha_{t-1})\mathbf{I})$$

$$\dots = \sqrt{\alpha_t \alpha_{t-1} \dots \alpha_1} \bar{x}_0 + \sqrt{1-\alpha_t \alpha_{t-1} \dots \alpha_1} \bar{\epsilon}$$

$$A_t = \alpha_1 \alpha_2 \dots \alpha_t$$

$$q(\bar{x}_t | \bar{x}_0) = \mathcal{N}(\bar{x}_t | \sqrt{A_t} \bar{x}_0, (1-A_t)\mathbf{I})$$

$$p_{\theta}(\bar{x}_0, \bar{x}_1, \dots, \bar{x}_T) = p_{\theta}(\bar{x}_T) \cdot \prod_{t=1}^T p_{\theta}(\bar{x}_{t-1} | \bar{x}_t)$$

~~$$q(\bar{x}_t | \bar{x}_t)$$~~

$$p_{\theta}(\bar{x}_{t-1} | \bar{x}_t) = \mathcal{N}(\bar{x}_{t-1} | \mu_{\theta}(\bar{x}_t, t), \Sigma_{\theta}(\bar{x}_t, t))$$

$$q(\bar{x}_{t-1} | \bar{x}_t, \bar{x}_0) = \frac{q(\bar{x}_{t-1} | \bar{x}_0)}{q(\bar{x}_t | \bar{x}_0)} =$$

$$= \text{Const.} \cdot e^{-\frac{1}{2} \left(\frac{(\bar{X}_t - \sqrt{A_t} \bar{X}_{t-1})^2}{\beta_t} + \frac{(\bar{X}_{t-1} - \sqrt{A_{t-1}} \bar{X}_0)^2}{1 - A_{t-1}} - \frac{(\bar{X}_t - \sqrt{A_t} \bar{X}_0)^2}{1 - A_t} \right)}$$

$$= \text{Const.} \cdot e^{-\frac{1}{2} \left(\bar{X}_{t-1}^2 \cdot \left(\frac{\alpha_t}{\beta_t} + \frac{1}{1 - A_{t-1}} \right) - 2 \bar{X}_{t-1} \left(\frac{\sqrt{A_t} \cdot \bar{X}_t}{\beta_t} + \frac{\sqrt{A_{t-1}} \cdot \bar{X}_0}{1 - A_{t-1}} \right) + \frac{(\bar{X}_t - \sqrt{A_t} \bar{X}_0)^2}{1 - A_t} \right)}$$

$$q(\bar{X}_{t-1} | \bar{X}_t, \bar{X}_0) = \mathcal{N}(\bar{X}_{t-1} | \hat{\mu}(\bar{X}_t, \bar{X}_0), \hat{\beta}_t I)$$

use $\hat{\beta}_t = \frac{1 - A_{t-1}}{1 - A_t} \beta_t$, $\bar{X}_t = \sqrt{A_t} \bar{X}_0 + \sqrt{1 - A_t} \cdot \bar{\varepsilon}$

$$\hat{\mu}(\bar{X}_t, \bar{X}_0) = \frac{\sqrt{A_t} (1 - A_{t-1})}{1 - A_t} \bar{X}_t + \frac{\sqrt{A_{t-1}} \beta_t}{1 - A_t} \bar{X}_0 =$$

$$\bar{X}_1, \dots, \bar{X}_T \sim q = \frac{1}{\sqrt{A_t}} \left(\bar{X}_t - \frac{1 - A_t}{\sqrt{1 - A_t}} \cdot \bar{\varepsilon}_t \right) \quad \frac{p_\theta(\bar{X}_0 - \bar{X}_T)}{p_\theta(\bar{X}_0)}$$

$$\log p_\theta(\bar{X}_0) \geq \log p_\theta(\bar{X}_0) - \text{KL}(q(\bar{X}_1, \dots, \bar{X}_T | \bar{X}_0) \| p_\theta(\bar{X}_1, \dots, \bar{X}_T | \bar{X}_0)) =$$

$$= \log p_\theta(\bar{X}_0) - \mathbb{E}_q \left[\log \frac{q(\bar{X}_1, \dots, \bar{X}_T | \bar{X}_0)}{p_\theta(\bar{X}_0, \bar{X}_T) / p_\theta(\bar{X}_0)} \right] =$$

$$= - \mathbb{E}_q \left[\log \frac{q(\bar{X}_1, \dots, \bar{X}_T | \bar{X}_0)}{p_\theta(\bar{X}_0, \bar{X}_T)} \right]$$

$$- \mathbb{E}_{q(\bar{X}_0)} [\log p_\theta(\bar{X}_0)] \leq \mathbb{E}_{q(\bar{X}_0, \bar{X}_1, \dots, \bar{X}_T)} \left[\log \frac{q(\bar{X}_1, \dots, \bar{X}_T | \bar{X}_0)}{p_\theta(\bar{X}_0, \bar{X}_T)} \right]$$

$\rightarrow \min$

$$E_q \left[\log \frac{\prod_{t=1}^T q(\bar{x}_t | \bar{x}_{t-1}, \bar{x}_0)}{p_\theta(\bar{x}_T) \cdot \prod_{t=1}^T p_\theta(\bar{x}_{t-1} | \bar{x}_t)} \right] = q(\bar{x}_{t-1} | \bar{x}_t, \bar{x}_0) \cdot \frac{q(\bar{x}_t | \bar{x}_0)}{q(\bar{x}_{t+1} | \bar{x}_0)}$$

$$= E_q \left[-\log p_\theta(\bar{x}_T) + \sum_{t=2}^T \log \frac{q(\bar{x}_t | \bar{x}_{t-1}, \bar{x}_0)}{p_\theta(\bar{x}_{t-1} | \bar{x}_t)} + \log \frac{q(\bar{x}_1 | \bar{x}_0)}{p_\theta(\bar{x}_0 | \bar{x}_1)} \right]$$

$$= E_q \left[-\log p_\theta(\bar{x}_T) + \sum_{t=2}^T \log \frac{q(\bar{x}_{t-1} | \bar{x}_t, \bar{x}_0)}{p_\theta(\bar{x}_{t-1} | \bar{x}_t)} + \sum_{t=2}^T \log \frac{q(\bar{x}_t | \bar{x}_0)}{q(\bar{x}_{t-1} | \bar{x}_0)} \right]$$

$\log \dots - \log \dots - \log \dots - \log \dots$
 $\log q(\bar{x}_t | \bar{x}_0) - \log q(\bar{x}_{t-1} | \bar{x}_0)$

$$= E_q \left[\log \frac{q(\bar{x}_T | \bar{x}_0)}{p_\theta(\bar{x}_T)} + \sum_{t=2}^T \log \frac{q(\bar{x}_{t-1} | \bar{x}_t, \bar{x}_0)}{p_\theta(\bar{x}_{t-1} | \bar{x}_t)} - \log p_\theta(\bar{x}_0 | \bar{x}_1) \right] \xrightarrow{\text{min}}$$

$L_T + L_{T-1} + L_{T-2} + \dots + L_1 + L_0$
 $\text{KL}(q(\bar{x}_T | \bar{x}_0) \parallel p_\theta(\bar{x}_T)) \quad \text{KL}(q(\bar{x}_{t-1} | \bar{x}_t, \bar{x}_0) \parallel p_\theta(\bar{x}_{t-1} | \bar{x}_t))$