

MC

n

\rightarrow

$n-1$

\rightarrow

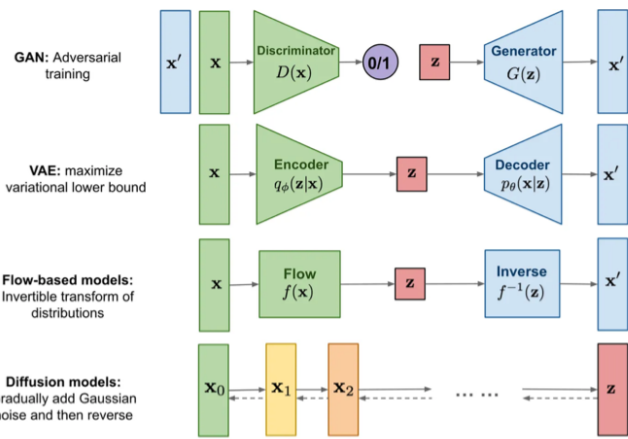
$n-2$

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

x_0

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

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \\ x_2 \end{pmatrix}$$



$0 \leftarrow 0$

$$P(x_t | x_{t-1})$$

FC: $F = AX + B$

$$\frac{\partial F}{\partial x} = A$$

$$\frac{\partial F}{\partial A} = \text{✓}$$

$$\frac{\partial F}{\partial B} = 1$$

MLP

FCs



MC

MC



Random sampling

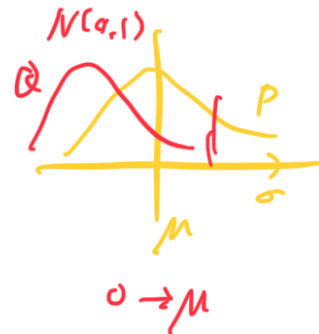
$$z \sim N(\mu, \sigma)$$

Reparameterization trick

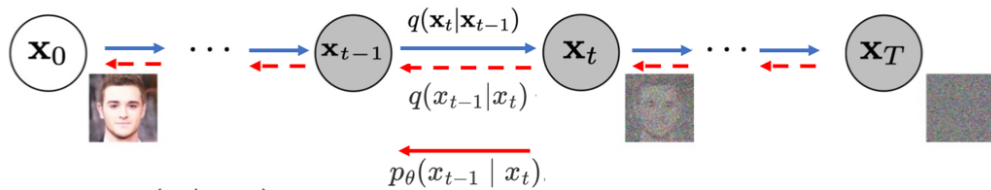
$$z \sim N(\mu, \sigma)$$

$$z := N(z; \mu, \sigma I)$$

$$z = \mu + \sigma \odot \varepsilon, \quad \varepsilon \sim N(0, 1)$$



Key concept / definitions



Diffuse : $q(x_t | x_{t-1})$

add noise (dim dim dim + mask mask mask)

Reverse-Diffuse : $q(x_{t-1} | x_t)$

The reverse process of diffuse. (can't be calculated but can be estimated)

Denoise : $p_\theta(x_{t-1} | x_t)$

The estimated process projecting denoised samples.

$$q: x_t \sim N(\underbrace{\sqrt{1-\beta_t}}_{\mu} x_{t-1}, \underbrace{\beta_t}_{\sigma} I) \quad 0 < \beta_t < 1$$

$$\Rightarrow x_t = N(x_t; \sqrt{1-\beta_t} x_{t-1}, \beta_t \cdot I)$$

$$q: \text{Diffuse } 0^{t-1} \xrightarrow{q} 0^t \quad \underbrace{\sqrt{1-\beta_t}}_{\text{Dim Dim Dim}} x_{t-1} + \underbrace{\sqrt{\beta_t} \odot \varepsilon}_{\text{mask mask mask}}, \quad \varepsilon \sim N(0, 1)$$

$$\left\{ \begin{array}{l} \text{Diffuse: } q(x_t | x_{t-1}) = N(x_t; \sqrt{1-\beta_t} x_{t-1}, \beta_t I) \\ \text{prob. } q(x_{1:T} | x_0) = \prod_{t=1}^T q(x_t | x_{t-1}) \end{array} \right.$$

$z \rightarrow v \rightarrow x \rightarrow v$

$$X_0 \sim \dots \sim X_T$$

$$\star N(0, \sigma_1^2 I) + N(0, \sigma_2^2 I) \sim N(0, (\sigma_1^2 + \sigma_2^2) I)$$

$\varepsilon_1 \quad \varepsilon_2 \quad \tilde{\varepsilon}$

$$X_0 \xrightarrow{\beta_1} X_1 \xrightarrow{\beta_2} \dots \xrightarrow{\beta_t} X_t \quad X_t \rightarrow X_0, \beta = (\beta_1, \dots, \beta_t)$$

$$\star \text{ let } \alpha_t = 1 - \beta_t, \quad \bar{\alpha}_t = \prod_{i=1}^t \alpha_i$$

$$\hookrightarrow X_t = \sqrt{\alpha_t} X_{t-1} + \sqrt{1 - \alpha_t} \varepsilon_{t-1} \quad \star$$

$$\Rightarrow X_{t-1} = \sqrt{\alpha_{t-1}} X_{t-2} + \sqrt{1 - \alpha_{t-1}} \varepsilon_{t-2} \quad \star$$

$$\begin{aligned} \Rightarrow X_t &= \sqrt{\alpha_t} (\sqrt{\alpha_{t-1}} X_{t-2}) + \sqrt{\alpha_t} \sqrt{1 - \alpha_{t-1}} \varepsilon_{t-2} \\ &\quad + \sqrt{1 - \alpha_t} \varepsilon_{t-1} \quad \varepsilon \sim N(0, 1) \\ &= \sqrt{\alpha_t \cdot \alpha_{t-1}} X_{t-2} + \sqrt{\alpha_t \cdot (1 - \alpha_{t-1}) + 1 - \alpha_t} \tilde{\varepsilon}_{t-2} \\ &= \sqrt{\alpha_t \cdot \alpha_{t-1}} X_{t-2} + \sqrt{1 - \alpha_t \cdot \alpha_{t-1}} \tilde{\varepsilon}_{t-2} \end{aligned}$$

$$\Rightarrow X_t = \sqrt{\alpha_t \cdot \alpha_{t-1} \dots \alpha_1} X_0 + \sqrt{1 - \alpha_t \cdot \alpha_{t-1} \dots \alpha_1} \tilde{\varepsilon}_0$$

$$\Rightarrow X_t = \sqrt{\bar{\alpha}_t} X_0 + \sqrt{1 - \bar{\alpha}_t} \varepsilon, \quad \varepsilon \sim N(0, 1)$$

$$\beta \in (0,1), \boxed{T}$$

$$\lim_{T \rightarrow \infty} X_T = \sqrt{\bar{\alpha}_T} X_0 + \sqrt{1 - \bar{\alpha}_T} \varepsilon$$

$$= 0 + 1 \cdot \varepsilon$$

$$X_T \sim N(0,1)$$

$$X_0 \rightarrow 0 \rightarrow 0 \rightarrow 0 \rightarrow 0 \rightarrow X_T \sim \varepsilon \sim N(0,1)$$

$\beta \downarrow$ $\beta \uparrow$ (reverse q)

$$\beta = (0.0001, 0.02)$$

$$q(X_t | X_{t-1}) \rightarrow N$$

$$\tilde{q}(X_{t-1} | X_t)$$

reverse diffuse
 $\tilde{q} \rightarrow N$

$$P_0(X_{t-1} | X_t)$$

$$\otimes \underline{P_0(X_{t-1} | (X_t, X_0))}$$

$$= N(X_{t-1}; \tilde{\mu}(X_t, X_0), \tilde{\beta}_t I)$$

Bayesian

$$P(A|B) = \frac{P(A)P(B|A)}{P(B)} \quad (b-0)$$

$$P(AB) = P(A)P(B|A) \quad (b-1)$$

$$P(ABC) = P(A) P(B|A) P(C|AB) \quad (b-2)$$

$$(b-0) \leftarrow (b-1) \Rightarrow P(A|B) = \frac{P(AB)}{P(B)} \quad (b-3)$$

transmission matrix

$$\Rightarrow q(x_{t-1} | (x_t, x_0)) = \frac{q(x_t, x_0, x_{t-1})}{q(x_t, x_0)} \quad (b-3)$$

$$= \frac{q(x_0) q(x_{t-1} | x_0) q(x_t | (x_{t-1}, x_0))}{q(x_0) q(x_t | x_0)} \quad (b-2)$$

$$= \underbrace{q(x_t | (x_{t-1}, x_0))}_{0 \rightarrow 0} \cdot \frac{q(x_{t-1} | x_0)}{q(x_t | x_0)} \quad (b-1)$$

$$= q(x_t | x_{t-1}) \frac{q(x_{t-1} | x_0)}{q(x_t | x_0)}$$

$x \sim N(\mu, \sigma)$ Gaussian exp

$$\propto \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

$$= \exp\left(-\frac{1}{2} \left(\underbrace{\frac{1}{\sigma^2}}_{\uparrow_2} x^2 - \frac{2\mu}{\sigma^2} x + \underbrace{\frac{\mu^2}{\sigma^2}}_{\uparrow_0} \right)\right)$$

$$\propto \exp\left(-\frac{1}{2} \left(\frac{(x_t - \sqrt{\alpha_t} x_{t-1})^2}{\beta_t} + \frac{(x_{t-1} - \sqrt{\alpha_{t-1}} x_0)^2}{1 - \alpha_{t-1}} - \frac{(x_t - \sqrt{\alpha_t} x_0)^2}{1 - \alpha_t} \right)\right)$$

$$= \exp\left(-\frac{1}{2} \left(\frac{\alpha_t}{\beta_t} + \frac{1}{1-\bar{\alpha}_{t-1}} \right) \chi_{t-1}^2 - \left(\frac{2\sqrt{\alpha_t}}{\beta_t} \chi_t + \frac{2\sqrt{\bar{\alpha}_{t-1}}}{1-\bar{\alpha}_{t-1}} \chi_0 \right) \chi_{t-1} + \underbrace{(1\chi_t, \chi_0)}_0 \right)$$

\uparrow χ_{t-1} \uparrow χ_t \uparrow χ_0

⇒ Based on matching the component:

$$\tilde{\beta}_t = \frac{1}{\frac{\alpha_t}{\beta_t} + \frac{1}{1-\bar{\alpha}_{t-1}}} = 1 / \left(\frac{\alpha_t - \bar{\alpha}_t + \beta_t}{\beta_t (1-\bar{\alpha}_{t-1})} \right) = \frac{1-\bar{\alpha}_{t-1}}{1-\bar{\alpha}_t} \cdot \beta_t$$

$$\begin{aligned} \tilde{\mu}_t(\chi_t, \chi_0) &= \left(\frac{\sqrt{\alpha_t}}{\beta_t} \chi_t + \frac{\sqrt{\bar{\alpha}_{t-1}}}{1-\bar{\alpha}_{t-1}} \chi_0 \right) / \left(\frac{\alpha_t}{\beta_t} + \frac{1}{1-\bar{\alpha}_{t-1}} \right) \\ &= \left(\frac{\sqrt{\alpha_t}}{\beta_t} \chi_t + \frac{\sqrt{\bar{\alpha}_{t-1}}}{1-\bar{\alpha}_{t-1}} \chi_0 \right) \cdot \frac{1-\bar{\alpha}_{t-1}}{1-\bar{\alpha}_t} \cdot \beta_t \\ &= \frac{\sqrt{\alpha_t} (1-\bar{\alpha}_{t-1})}{1-\bar{\alpha}_t} \chi_t + \frac{\sqrt{\bar{\alpha}_{t-1}} \beta_t}{1-\bar{\alpha}_t} \chi_0 \end{aligned}$$

and because of

$$X_t = \sqrt{\alpha_t} X_0 + \sqrt{1 - \alpha_t} \epsilon \leftarrow \epsilon \sim N(0, 1)$$

$$\Rightarrow X_0 = \frac{1}{\sqrt{\alpha_t}} (X_t - \sqrt{1 - \alpha_t} \epsilon_t)$$

put it \downarrow

$$\Rightarrow \tilde{\mu}_t = \frac{1}{\sqrt{\alpha_t}} \left(X_t - \frac{1 - \alpha_t}{\sqrt{1 - \alpha_t}} \epsilon_t \right), \quad \epsilon_t \sim N(0, 1)$$

therefore then $q(X_{t-1} | (X_t, X_0))$ can be calculated

$$q: X_{t-1} = N(X_{t-1}; \tilde{\mu}_t, \tilde{\sigma}_t)$$

$$N\left(X_{t-1}; \frac{1}{\sqrt{\alpha_t}} \left(X_t - \frac{1 - \alpha_t}{\sqrt{1 - \alpha_t}} \epsilon_t \right), \underbrace{\frac{1 - \alpha_{t-1}}{1 - \alpha_t} \cdot \beta_t}\right)$$