

28/01/25

Computer Vision

Recap: • $\min_{\theta \in \Theta} \mathbb{E}_q [\| \tilde{\mu}(x^{(0)}, x^{(t)}) - f_{\mu}(x^{(t)}, t; \theta) \|_2^2] \quad - (1)$

↳ $q(x^{(t-1)} | x^{(t)}, x^{(0)}) \sim \mathcal{N}(x^{(t-1)}; \tilde{\mu}(x^{(t)}, x^{(0)}), \beta_t I)$

• $\tilde{\mu}(x^{(0)}, x^{(t)}) = \frac{\sqrt{\bar{\alpha}_t - 1}}{(1 - \bar{\alpha}_t)} \cdot (\beta_t \cdot x^{(0)}) + \frac{\sqrt{\alpha_t}(1 - \bar{\alpha}_{t-1})}{(1 - \bar{\alpha}_t)} x^{(t)} \quad - (2)$

• $p_{\theta}(x^{(t-1)} | x^{(t)}) \sim \mathcal{N}(x^{(t-1)}; f_{\mu}(x^{(t)}, t; \theta), \sigma_t^2 I) \quad - (3)$

$q(x^{(t)} | x^{(t+1)}) \sim \mathcal{N}(x^{(t)}; \sqrt{1 - \beta_t} \cdot x^{(t+1)}, \beta_t I) \quad - (4)$

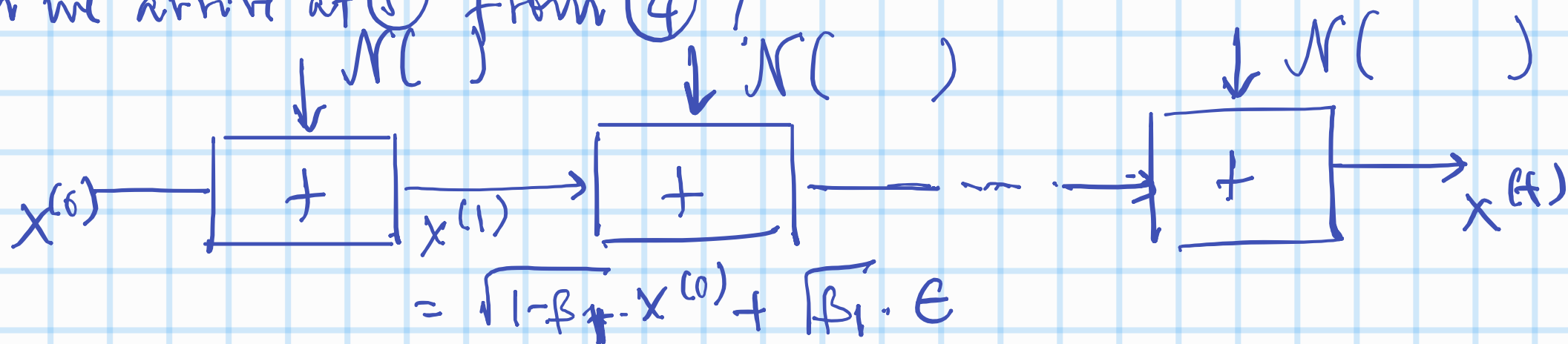
• Denoising model:

$q(x^{(t)} | x^{(0)}) \sim \mathcal{N}(x^{(t)}; \sqrt{\bar{\alpha}_t} \cdot x^{(0)}, (1 - \bar{\alpha}_t) I) \quad - (5)$

or $x^{(t)} = \sqrt{\bar{\alpha}_t} x^{(0)} + \sqrt{(1 - \bar{\alpha}_t)} \cdot \epsilon \quad \epsilon \sim \mathcal{N}(0, I), \quad - (6)$

where $\alpha_t = 1 - \beta_t$, $\bar{\alpha}_t = \prod_{s=1}^t \alpha_s \quad - (7)$

Can we arrive at (5) from (4)?



Q. Show how $x^{(t)}$ relates to $x^{(0)}$

From (6), $x^{(0)} = \frac{1}{\sqrt{\bar{\alpha}_t}} \cdot (x^{(t)} - \sqrt{(1 - \bar{\alpha}_t)} \cdot \epsilon) \quad - (8)$

Plug (8) into (2)

$\sqrt{\bar{\alpha}_t} = \sqrt{\alpha_1 \cdots \alpha_t} ; \alpha_t = 1 - \beta_t$

$\sqrt{\bar{\alpha}_t} = \sqrt{\alpha_t \cdot \bar{\alpha}_{t-1}}$

Writing (2) again,

$$\tilde{\mu}(x^{(t)}, x^{(0)}) = \frac{\sqrt{\bar{\alpha}_{t-1}}}{(1-\bar{\alpha}_t)} \beta_t \cdot x^{(0)} + \frac{\sqrt{\alpha_t}(1-\bar{\alpha}_{t-1})}{(1-\bar{\alpha}_t)} x^{(t)}$$

From (2),

$$\tilde{\mu}(x^{(t)}, x^{(0)}) = \frac{\sqrt{\bar{\alpha}_{t-1}}}{(1-\bar{\alpha}_t)} \beta_t \left(\frac{1}{\sqrt{\bar{\alpha}_t} \sqrt{\alpha_t}} (x^{(t)} - \sqrt{1-\bar{\alpha}_t} \epsilon) \right) +$$

$$\frac{\sqrt{\alpha_t}(1-\bar{\alpha}_{t-1})}{(1-\bar{\alpha}_t)} x^{(t)}$$

$$= \frac{1}{(1-\bar{\alpha}_t) \sqrt{\alpha_t}} \left(\beta_t + \alpha_t (1-\bar{\alpha}_{t-1}) \right) x^{(t)} - \frac{\beta_t}{\sqrt{\alpha_t} \sqrt{(1-\bar{\alpha}_t)}} \epsilon$$

$$= \frac{1}{\sqrt{\alpha_t} (1-\bar{\alpha}_t)} \left(\beta_t + \alpha_t - \underbrace{\alpha_t \bar{\alpha}_{t-1}}_{\alpha_t \cdot \alpha_{t-1} \dots \alpha_1} \right) x^{(t)}$$

$$= \frac{(1-\bar{\alpha}_t)}{\sqrt{\alpha_t} (1-\bar{\alpha}_t)} x^{(t)} - \frac{\beta_t}{\sqrt{\alpha_t} \sqrt{1-\bar{\alpha}_t}} \epsilon$$

$$\tilde{\mu}(x^{(t)}, x^{(0)}) = \frac{1}{\sqrt{\alpha_t}} \left(x^{(t)} - \frac{\beta_t}{\sqrt{1-\bar{\alpha}_t}} \epsilon \right) \quad - (9)$$

From (9), our $f_{\mu}(\quad)$ effectively learns to predict the noise term in (9). In other words we have $f_{\epsilon}(x^{(0)}, \epsilon, t)$

If $f_n(x^{(t)}, x^{(0)}) = \frac{1}{\sqrt{\alpha_t}} \left(x^{(t)} - \frac{\beta_t}{\sqrt{1-\bar{\alpha}_t}} \cdot f_{\epsilon}(\quad) \right)$, then

① becomes

$$E_n \left[\frac{\beta_t^2}{\alpha_t(1-\bar{\alpha}_t)} \| \epsilon - f_{\epsilon}(x^{(t)}, \epsilon_t) \|^2 \right] = \textcircled{10}$$