

Q1. Given $g(x_{1:T}|x_0) = \prod_{t=1}^T g(x_t|x_{t+1})$

Show $g(x_{1:T}|x_0) = g(x_T|x_0) \prod_{t=1}^T g(x_{t+1}|x_t, x_0)$

[Hint: $g(x_{1:T}|x_0) = \prod_{t=1}^T g(x_t|x_{t+1}) = \prod_{t=1}^T g(x_t|x_{t+1}, x_0)$]

x_1, x_2, \dots, x_T form a Markov chain when conditioned on x_0 .]

$$g(x_{1:T}|x_0) = \prod_{t=1}^T g(x_t|x_{t+1})$$

$$= \prod_{t=1}^T g(x_t|x_{t+1}, x_0)$$

$$= g(x_1|x_0, x_0) \prod_{t=2}^T g(x_t|x_{t+1}, x_0)$$

$$= g(x_1|x_0) \prod_{t=2}^T \frac{g(x_t) g(x_t|x_0) g(x_{t+1}|x_t, x_0)}{g(x_0) g(x_{t+1}|x_0)}$$

$$= g(x_1|x_0) \prod_{t=2}^T \frac{g(x_t|x_0)}{g(x_{t+1}|x_0)} \prod_{t=2}^T g(x_{t+1}|x_t, x_0)$$

$$= \cancel{g(x_1|x_0)} \frac{g(x_T|x_0)}{\cancel{g(x_1|x_0)}} \prod_{t=2}^T g(x_{t+1}|x_t, x_0)$$

$$= g(x_T|x_0) \prod_{t=2}^T g(x_{t+1}|x_t, x_0)$$

Q2. Prove Eqs. (4), (6), (8) in DDPM paper

$$(4) q(x_t | x_0) = N(x_t; \sqrt{a_t} x_0, (1 - \bar{a}_t) I)$$

we only need to prove $x_t = \sqrt{a_t} x_0 + \sqrt{1 - \bar{a}_t} z$,
where $t \in [1, T]$ and $z \sim N(0, I)$

$$a_t = 1 - \beta_t$$

$$\bar{a}_t = \frac{t}{T} a_t$$

① when $t=1$, it satisfied

$$x_1 = \sqrt{a_1} x_0 + \sqrt{1 - a_1} z_1 = \sqrt{a_1} x_0 + \sqrt{1 - a_1} z_1$$

② Let we assume that x_{t+1} can also satisfy the above assumption, we can get

$$\begin{aligned} x_t &= \sqrt{a_t} x_{t+1} + \sqrt{1 - a_t} z_t \\ &= \sqrt{a_t} (\sqrt{\bar{a}_{t+1}} x_0 + \sqrt{1 - \bar{a}_{t+1}} z_{t+1}) + \sqrt{1 - a_t} z_t \\ &= \underbrace{\sqrt{a_t \bar{a}_{t+1}} x_0}_{\text{value}} + \underbrace{\sqrt{a_t - a_t \bar{a}_{t+1}} z_{t+1}}_{\sim N(0, a_t - \bar{a}_t)} + \underbrace{\sqrt{1 - a_t} z_t}_{\sim N(0, 1 - a_t)} \end{aligned}$$

Based on the property of Gaussian function, we can know

$$x_t \sim N(\sqrt{a_t} x_0, (1 - a_t + a_t - \bar{a}_t) I) \sim N(\sqrt{a_t} x_0, (1 - \bar{a}_t) I)$$

$$\Rightarrow x_t = \sqrt{a_t} x_0 + \sqrt{1 - \bar{a}_t} \hat{z}_t$$

i.e. this satisfy the initial assumption

$$(b) q(x_{t+1} | x_t, x_0) = N(x_{t+1}; \tilde{m}_t(x_t, x_0), \tilde{\beta}_t I)$$

As we can know, $\tilde{m}_t(x_t, x_0) := \frac{\sqrt{\bar{\alpha}_t} \beta_t}{1 - \bar{\alpha}_t} x_0 + \frac{\sqrt{\bar{\alpha}_t} (1 - \bar{\alpha}_t)}{1 - \bar{\alpha}_t} x_t$

$$\text{and } \tilde{\beta}_t := \frac{1 - \bar{\alpha}_{t-1}}{1 - \bar{\alpha}_t} \beta_t$$

$$\begin{aligned}
q(x_{t+1} | x_t, x_0) &= \frac{q(x_t | x_{t+1}, x_0) q(x_{t+1} | x_0)}{q(x_t | x_0)} \\
&= \frac{q(x_t | x_{t+1}) q(x_{t+1} | x_0)}{q(x_t | x_0)} \\
&= \frac{N(x_t; \sqrt{\bar{\alpha}_t} x_{t+1}, (1 - \bar{\alpha}_t) I) N(x_{t+1}; \sqrt{\bar{\alpha}_t} x_0, (1 - \bar{\alpha}_t) I)}{N(x_t; \sqrt{\bar{\alpha}_t} x_0, (1 - \bar{\alpha}_t) I)} \\
&\propto \exp \left(- \left(\frac{(x_t - \sqrt{\bar{\alpha}_t} x_{t+1})^2}{2(1 - \bar{\alpha}_t)} + \frac{(x_{t+1} - \sqrt{\bar{\alpha}_t} x_0)^2}{2(1 - \bar{\alpha}_{t+1})} - \frac{(x_t - \sqrt{\bar{\alpha}_t} x_0)^2}{2(1 - \bar{\alpha}_t)} \right) \right) \\
&= \exp \left(- \frac{1}{2} \left(\frac{(x_t - \sqrt{\bar{\alpha}_t} x_{t+1})^2}{1 - \bar{\alpha}_t} + \frac{(x_{t+1} - \sqrt{\bar{\alpha}_{t+1}} x_0)^2}{1 - \bar{\alpha}_{t+1}} - \frac{(x_t - \sqrt{\bar{\alpha}_t} x_0)^2}{1 - \bar{\alpha}_t} \right) \right) \\
&= \exp \left(- \frac{1}{2} \left(\frac{-2\sqrt{\bar{\alpha}_t} x_t x_{t+1} + \bar{\alpha}_t x_{t+1}^2}{1 - \bar{\alpha}_t} + \frac{x_{t+1}^2 - 2\sqrt{\bar{\alpha}_{t+1}} x_{t+1} x_0}{1 - \bar{\alpha}_{t+1}} + C(x_t, x_0) \right) \right) \\
&= \exp \left(- \frac{1}{2} \left(\left(\frac{\bar{\alpha}_t}{1 - \bar{\alpha}_t} + \frac{1}{1 - \bar{\alpha}_{t+1}} \right) x_{t+1}^2 - 2 \left(\frac{\sqrt{\bar{\alpha}_t} x_t}{1 - \bar{\alpha}_t} + \frac{\sqrt{\bar{\alpha}_{t+1}} x_0}{1 - \bar{\alpha}_{t+1}} \right) x_{t+1} \right. \right. \\
&\quad \left. \left. + C(x_t, x_0) \right) \right) \\
&= \exp \left(- \frac{1}{2} \left(\frac{\bar{\alpha}_t - \bar{\alpha}_{t+1} + 1 - \bar{\alpha}_t}{(1 - \bar{\alpha}_t)(1 - \bar{\alpha}_{t+1})} x_{t+1}^2 - \left(\frac{\sqrt{\bar{\alpha}_t} x_t}{1 - \bar{\alpha}_t} + \frac{\sqrt{\bar{\alpha}_{t+1}} x_0}{1 - \bar{\alpha}_{t+1}} \right) x_{t+1} \right. \right. \\
&\quad \left. \left. - \frac{1}{2} C(x_t, x_0) \right) \right) \\
&= \exp \left(- \frac{1}{2} \left(\frac{1 - \bar{\alpha}_t}{(1 - \bar{\alpha}_t)(1 - \bar{\alpha}_{t+1})} \right) \left(x_{t+1}^2 - 2 \frac{\frac{\sqrt{\bar{\alpha}_t} x_t}{1 - \bar{\alpha}_t} + \frac{\sqrt{\bar{\alpha}_{t+1}} x_0}{1 - \bar{\alpha}_{t+1}}}{\frac{1 - \bar{\alpha}_t}{(1 - \bar{\alpha}_t)(1 - \bar{\alpha}_{t+1})}} x_{t+1} \right) \right. \\
&\quad \left. - \frac{1}{2} C(x_t, x_0) \right)
\end{aligned}$$

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

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

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

$$\propto N\left(\chi_{t+1}; \underbrace{\frac{\sqrt{\alpha_t}(1-\bar{\alpha}_t)\chi_t + \sqrt{\bar{\alpha}_{t+1}}(1-\alpha_t)\chi_0}{1-\bar{\alpha}_t}}_{\tilde{\mu}_0(\chi_t, \chi_0)}, \underbrace{\frac{(1-\alpha_t)(1-\bar{\alpha}_{t+1})}{1-\bar{\alpha}_t} I}_{\tilde{\beta}_0}\right)$$

$$(8) L_{t+1} = E_g \left[\frac{1}{2\sigma_0^2} \| \tilde{m}_g(x_e, x_0) - \tilde{m}_g(x_{t+1}, t) \|^2 \right] + c$$

Based on (b) $q(x_{t+1} | x_t, x_0) = N(x_{t+1}; \tilde{m}_g(x_t, x_0), \tilde{\sigma}_g^2 I)$

and $p_\theta(x_{t+1} | x_t) = N(x_{t+1}; \tilde{m}_\theta(x_t, t), \sigma_\theta^2 I)$

$$\begin{aligned} \therefore \frac{q(x_{t+1} | x_t, x_0)}{q(x_t, x_0)} &= \frac{q(x_{t+1}, x_t, x_0)}{q(x_t, x_0)} = \frac{q(x_{t+1}, x_t | x_0) q(x_0)}{q(x_t | x_0) q(x_0)} \\ &= \frac{q(x_{t+1}, x_t | x_0)}{q(x_t | x_0)} \end{aligned}$$

$$\Leftrightarrow q(x_{t+1}, x_t | x_0) = q(x_{t+1} | x_t, x_0) q(x_t | x_0)$$

② if p, q are Gaussian distribution, KL divergence is

$$\begin{aligned} KL(p, q) &= KL(N(\mu_1, \sigma_1^2) || N(\mu_2, \sigma_2^2)) \\ &= \int_X \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}} \log \frac{\frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}}}{\frac{1}{\sqrt{2\pi\sigma_2^2}} e^{-\frac{(x-\mu_2)^2}{2\sigma_2^2}}} dx \\ &= \int_X \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}} \left[\log \frac{\sigma_2}{\sigma_1} - \frac{(x-\mu_1)^2}{2\sigma_1^2} + \frac{(x-\mu_2)^2}{2\sigma_2^2} \right] dx \\ &= \log \frac{\sigma_2}{\sigma_1} \int_X \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}} dx \\ &\quad + (\mu_1 - \mu_2)^2 \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}} dx + (\mu_2 - \mu_1)^2 \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}} dx \\ &= \log \frac{\sigma_2}{\sigma_1} - \frac{1}{2} + \frac{\sigma_1^2 + (\mu_1 - \mu_2)^2}{2\sigma_2^2} \end{aligned}$$

$$\therefore L_{t+1} = E_g(x_t, x_{t+1} | x_0) \left[\sum_{t=2}^T \log \left(\frac{q(x_{t+1} | x_0, x_t)}{p_\theta(x_{t+1} | x_t)} \right) \right]$$

$$= E_g(x_t | x_0) q(x_{t+1} | x_t, x_0) \left[\sum_{t=2}^T \log \left(\frac{q(x_{t+1} | x_0, x_t)}{p_\theta(x_{t+1} | x_t)} \right) \right]$$

$$= E_g(x_t | x_0) \underbrace{\left[KL(q(x_t | x_0, x_e) || p_\theta(x_t | x_e)) \right]}_{\downarrow}$$

$$KL(N(\mu, \sigma^2) || N(\mu_1, \sigma_1^2)) = \frac{1}{2} \left(\log \frac{\sigma^2}{\sigma_1^2} + \log \frac{\sigma_1^2}{\sigma^2} - 1 + \frac{\mu^2}{\sigma_1^2} + \frac{(\mu - \mu_1)^2}{\sigma_1^2} \right)$$

$$\Leftrightarrow KL(q(x_{t+1} | x_0, x_e) || p_\theta(x_{t+1} | x_e)) = \frac{1}{2} \left(c + \frac{\| \tilde{m}_g(x_e, x_0) - m_\theta(x_e, t) \|^2}{\sigma_\theta^2} \right)$$