

Prove DDPM Q1

problem:

$$\text{Given } q(x_{1:T} | x_0) = \prod_{t=1}^T q(x_t | x_{t-1})$$

$$\text{Show } q(x_{1:T} | x_0) = q(x_T | x_0) \prod_{t=2}^T q(x_{t-1} | x_t, x_0)$$

Because  $x_1, x_2, \dots, x_T$  form a Markov chain

when conditioned on  $x_0$ , we can rewrite  $q(x_t | x_{t-1}) = q(x_t | x_{t-1}, x_0)$ , where  $x_0$  term is superfluous due to the Markov property.

$$\begin{aligned} q(x_{1:T} | x_0) &= \prod_{t=1}^T q(x_t | x_{t-1}) \\ &= q(x_1 | x_0) \prod_{t=2}^T q(x_{t-1} | x_t, x_0) \rightarrow \textcircled{1} \end{aligned}$$

Then, according to Bayes rule, we have the following derivation.

$$\begin{aligned}
p(x_t | x_{t-1}, x_0) &= \frac{p(x_t, x_{t-1}, x_0)}{p(x_{t-1}, x_0)} \\
&= \frac{p(x_{t-1}, x_t, x_0)}{p(x_{t-1}, x_0)} \\
&= \frac{p(x_{t-1} | x_t, x_0) p(x_t, x_0)}{p(x_{t-1}, x_0)} \\
&= \frac{p(x_{t-1} | x_t, x_0) p(x_t | x_0) \cancel{p(x_0)}}{p(x_{t-1} | x_0) \cancel{p(x_0)}} \\
&= \frac{p(x_{t-1} | x_t, x_0) p(x_t | x_0)}{p(x_{t-1} | x_0)} \rightarrow \textcircled{2}
\end{aligned}$$

Now we can substitute  $\textcircled{2}$  into  $\textcircled{1}$ .

$$\begin{aligned}
p(x_{1:T} | x_0) &= p(x_1 | x_0) \prod_{t=2}^T p(x_t | x_{t-1}, x_0) \\
&= p(x_1 | x_0) \prod_{t=2}^T \frac{p(x_{t-1} | x_t, x_0) p(x_t | x_0)}{p(x_{t-1} | x_0)}
\end{aligned}$$

$$= \cancel{p(x_1|x_0)} \frac{p(x_1|x_2, x_0) \cancel{p(x_2|0)}}{p(x_1|x_0)} \frac{p(x_2|x_3, x_0) \cancel{p(x_3|0)}}{\cancel{p(x_2|x_0)}} \dots \frac{p(x_{T-1}|x_T, x_0) \cancel{p(x_T|0)}}{\cancel{p(x_{T-1}|x_0)}}$$

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

$$= p(x_T|x_0) \prod_{t=2}^T p(x_{t-1}|x_t, x_0) \quad \#$$

## DDPM Q2

1. Proof of Eq (4) in DDPM paper

$$\text{Given } q(x_t | x_{t-1}) = \mathcal{N}(x_t; \sqrt{1-\beta_t} x_{t-1}, \beta_t I)$$

$$\text{let } \alpha_t = 1 - \beta_t$$

$$q(x_t | x_{t-1}) = \mathcal{N}(x_t; \sqrt{\alpha_t} x_{t-1}, (1-\alpha_t)I)$$

With the reparameterization trick,  $x_t \sim q(x_t | x_{t-1})$  can be rewritten as:

$$x_t = \sqrt{\alpha_t} x_{t-1} + \sqrt{1-\alpha_t} \epsilon \text{ with } \epsilon \sim \mathcal{N}(\epsilon; 0, I)$$

Then, we can do the following derivation:

$$\begin{aligned} x_t &= \sqrt{\alpha_t} x_{t-1} + \sqrt{1-\alpha_t} \epsilon_{t-1}^* \\ &= \sqrt{\alpha_t} (\sqrt{\alpha_{t-1}} x_{t-2} + \sqrt{1-\alpha_{t-1}} \epsilon_{t-2}^*) + \sqrt{1-\alpha_t} \epsilon_{t-1}^* \\ &= \sqrt{\alpha_t \alpha_{t-1}} x_{t-2} + \sqrt{\alpha_t - \alpha_t \alpha_{t-1}} \epsilon_{t-2}^* + \sqrt{1-\alpha_t} \epsilon_{t-1}^* \\ &= \sqrt{\alpha_t \alpha_{t-1}} x_{t-2} + \sqrt{\alpha_t - \alpha_t \alpha_{t-1} + \alpha_t - \alpha_t} \epsilon_{t-2}^* \\ &= \sqrt{\alpha_t \alpha_{t-1}} x_{t-2} + \sqrt{\alpha_t - \alpha_t \alpha_{t-1} + 1 - \alpha_t} \epsilon_{t-2}^* \\ &= \sqrt{\alpha_t \alpha_{t-1}} x_{t-2} + \sqrt{1 - \alpha_t \alpha_{t-1}} \epsilon_{t-2}^* \end{aligned}$$

$$= \dots$$

$$= \sqrt{\prod_{i=1}^t \alpha_i} x_0 + \sqrt{1 - \prod_{i=1}^t \alpha_i} \epsilon_0$$

$$\text{let } \bar{\alpha}_t = \prod_{i=1}^t \alpha_i, \text{ we have}$$

$$x_t = \sqrt{\bar{\alpha}_t} x_0 + \sqrt{1-\bar{\alpha}_t} \epsilon_0 \sim \mathcal{N}(x_t; \sqrt{\bar{\alpha}_t} x_0, (1-\bar{\alpha}_t)I)$$

$$q(x_t | x_0) = \mathcal{N}(x_t; \sqrt{\bar{\alpha}_t} x_0, (1-\bar{\alpha}_t)I) \quad \#$$

## 2. Proof of $E_g(b)$ in DDPM paper

In 1., we have derived the Gaussian form of  $g(x_t | x_0)$ , this also means that with some modification we are able to obtain the Gaussian form of  $g(x_{t-1} | x_0)$ . With these prerequisites, we can start calculate the form of  $g(x_{t-1} | x_t, x_0)$  by substituting into the Bayes rule expansion:

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

$$= \exp \left\{ -\frac{1}{2} \left[ \left( \frac{\alpha_t}{1-\alpha_t} + \frac{1}{1-\bar{\alpha}_{t-1}} \right) x_{t-1}^2 - 2 \left( \frac{\sqrt{\alpha_t} x_t}{1-\alpha_t} + \frac{\sqrt{\bar{\alpha}_{t-1}} x_0}{1-\bar{\alpha}_{t-1}} \right) x_{t-1} + C(x_t, x_0) \right] \right\}$$

$$= \exp \left\{ -\frac{1}{2} \left[ \frac{\alpha_t(1-\bar{\alpha}_{t-1}) + 1 - \alpha_t}{(1-\alpha_t)(1-\bar{\alpha}_{t-1})} x_{t-1}^2 - 2 \left( \frac{\sqrt{\alpha_t} x_t}{1-\alpha_t} + \frac{\sqrt{\bar{\alpha}_{t-1}} x_0}{1-\bar{\alpha}_{t-1}} \right) x_{t-1} + C(x_t, x_0) \right] \right\}$$

$$= \exp \left\{ -\frac{1}{2} \left[ \frac{\alpha_t - \bar{\alpha}_t + 1 - \alpha_t}{(1-\alpha_t)(1-\bar{\alpha}_{t-1})} x_{t-1}^2 - 2 \left( \frac{\sqrt{\alpha_t} x_t}{1-\alpha_t} + \frac{\sqrt{\bar{\alpha}_{t-1}} x_0}{1-\bar{\alpha}_{t-1}} \right) x_{t-1} + C(x_t, x_0) \right] \right\}$$

$$= \exp \left\{ -\frac{1}{2} \left[ \frac{1 - \bar{\alpha}_t}{(1-\alpha_t)(1-\bar{\alpha}_{t-1})} x_{t-1}^2 - 2 \left( \frac{\sqrt{\alpha_t} x_t}{1-\alpha_t} + \frac{\sqrt{\bar{\alpha}_{t-1}} x_0}{1-\bar{\alpha}_{t-1}} \right) x_{t-1} + C(x_t, x_0) \right] \right\}$$

$$= \exp \left\{ -\frac{1}{2} \left( \frac{1 - \bar{\alpha}_t}{(1-\alpha_t)(1-\bar{\alpha}_{t-1})} \right) \left[ x_{t-1}^2 - 2 \frac{\left( \frac{\sqrt{\alpha_t} x_t}{1-\alpha_t} + \frac{\sqrt{\bar{\alpha}_{t-1}} x_0}{1-\bar{\alpha}_{t-1}} \right)}{\frac{1 - \bar{\alpha}_t}{(1-\alpha_t)(1-\bar{\alpha}_{t-1})}} x_{t-1} + C'(x_t, x_0) \right] \right\}$$

$$= \exp \left\{ -\frac{1}{2} \left( \frac{1 - \bar{\alpha}_t}{(1-\alpha_t)(1-\bar{\alpha}_{t-1})} \right) \left[ x_{t-1}^2 - 2 \frac{\left( \frac{\sqrt{\alpha_t} x_t}{1-\alpha_t} + \frac{\sqrt{\bar{\alpha}_{t-1}} x_0}{1-\bar{\alpha}_{t-1}} \right) (1-\alpha_t)(1-\bar{\alpha}_{t-1})}{1 - \bar{\alpha}_t} x_{t-1} + C'(x_t, x_0) \right] \right\}$$

$$= \exp \left\{ -\frac{1}{2} \left( \frac{1}{\frac{(1-\alpha_t)(1-\bar{\alpha}_{t-1})}{1-\bar{\alpha}_t}} \right) \left[ x_{t-1}^2 - 2 \frac{\sqrt{\alpha_t}(1-\bar{\alpha}_{t-1})x_t + \sqrt{\bar{\alpha}_{t-1}}(1-\alpha_t)x_0}{1-\bar{\alpha}_t} x_{t-1} + C'(x_t, x_0) \right] \right\}$$

$$x_{t-1} \sim \mathcal{N} \left( x_{t-1}; \frac{\sqrt{\alpha_t}(1-\bar{\alpha}_{t-1})x_t + \sqrt{\bar{\alpha}_{t-1}}(1-\alpha_t)x_0}{1-\bar{\alpha}_t}, \frac{(1-\alpha_t)(1-\bar{\alpha}_{t-1})}{1-\bar{\alpha}_t} \mathbf{I} \right)$$

Where  $C(X_t, X_0)$  and  $C'(X_t, X_0)$  are functions that composed only of  $X_t, X_0$  and  $\alpha$ , in which  $C'(X_t, X_0)$  can be shown to complete the square.

$$X_{t-1}^2 - 2 \frac{\sqrt{\alpha_t}(1-\bar{\alpha}_{t-1})X_t + \sqrt{\bar{\alpha}_{t-1}}(1-\alpha_t)X_0}{1-\bar{\alpha}_t} X_{t-1} + C'(X_t, X_0) = \left( X_{t-1} - \frac{\sqrt{\alpha_t}(1-\bar{\alpha}_{t-1})X_t + \sqrt{\bar{\alpha}_{t-1}}(1-\alpha_t)X_0}{1-\bar{\alpha}_t} \right)^2$$

Using the notations  $\tilde{\mu}_t(X_t, X_0) = \frac{\sqrt{\alpha_{t-1}}\beta_t}{1-\bar{\alpha}_t} X_0 + \frac{\sqrt{\alpha_t}(1-\bar{\alpha}_{t-1})}{1-\bar{\alpha}_t} X_t$  and

$\tilde{\beta}_t = \frac{1-\bar{\alpha}_{t-1}}{1-\alpha_t} \beta_t$ , we have

$$q(X_{t-1} | X_t, X_0) = \mathcal{N}(X_{t-1}; \tilde{\mu}_t(X_t, X_0), \tilde{\beta}_t \mathbf{I}) \quad \text{##}$$

### 3. Proof of $E_8(8)$ in DDPM paper

Given the denoising matching term

$$L_{t-1} = E_{q(x_t|x_0)} [D_{KL}(q(x_{t-1}|x_t, x_0) \| p_\theta(x_{t-1}|x_t))]$$

We already know the gaussian form of  $q(x_{t-1}|x_t, x_0)$ , thus we can set the variance of  $p_\theta(x_{t-1}|x_t)$  to match exactly, optimizing the KL term reduces to minimizing the difference between the means of the two distributions.

$$\text{let } \sigma_t^2 = \tilde{\beta}_t = \frac{1-\bar{\alpha}_t}{1-\bar{\alpha}_t} \beta_t, \quad \Sigma(t) = \sigma_t^2 \mathbf{I}$$

$$\begin{aligned} L_{t-1} &= E_{q(x_t|x_0)} [D_{KL}(\mathcal{N}(x_{t-1}|\tilde{\mu}_t, \Sigma(t)) \| \mathcal{N}(x_{t-1}|\mu_\theta, \Sigma(t)))] \\ &= E_{q(x_t|x_0)} \left[ \frac{1}{2} \left[ \log \frac{\det(\Sigma(t))}{\det(\Sigma(t))} - d + \text{tr}(\Sigma(t)^{-1} \Sigma(t)) + (\mu_\theta - \tilde{\mu}_t)^T \Sigma(t)^{-1} (\mu_\theta - \tilde{\mu}_t) \right] \right] \\ &= E_{q(x_t|x_0)} \left[ \frac{1}{2} \left[ -d + d + (\mu_\theta - \tilde{\mu}_t)^T \Sigma(t)^{-1} (\mu_\theta - \tilde{\mu}_t) \right] \right] \\ &= E_{q(x_t|x_0)} \left[ \frac{1}{2} \left[ (\mu_\theta - \tilde{\mu}_t)^T \Sigma(t)^{-1} (\mu_\theta - \tilde{\mu}_t) \right] \right] \\ &= E_{q(x_t|x_0)} \left[ \frac{1}{2} \left[ (\mu_\theta - \tilde{\mu}_t)^T (\sigma_t^{-2} \mathbf{I})^{-1} (\mu_\theta - \tilde{\mu}_t) \right] \right] \\ &= E_{q(x_t|x_0)} \left[ \frac{1}{2\sigma_t^2} \|\mu_\theta - \tilde{\mu}_t\|^2 \right] \# \end{aligned}$$