

DDPM Lower Bound

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1 Introduction

Derivation using Jensen's inequality ¹:

$$\begin{aligned}\log p(X) &= \log \int_Z p(X, Z) \\ &= \log \int_Z p(X, Z) \frac{p(Z|X)}{p(Z|X)} \\ &= \log \left(\mathbb{E}_q \left[\frac{p(X, Z)}{q(Z|X)} \right] \right) \\ &\geq \mathbb{E}_q \left[\log \frac{p(X, Z)}{q(Z|X)} \right] \\ &= \mathbb{E}_q [\log p(X, Z)] + H[Z|X]\end{aligned}$$

In the diffusion model, we replace by $X \sim x_0$ and $Z \sim x_{1:T}$, and since we have:

$$p(x_{0:T}) = p(x_T) \prod_{t=1}^T p(x_{t-1}|x_t) \quad (1)$$

and

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

¹<https://xyang35.github.io/2017/04/14/variational-lower-bound/>

we can derive the following:

$$\begin{aligned}
\log p(x_0) &= \log \int p(x_0, x_{1:T}) \frac{q(x_{1:T}|x_0)}{q(x_{1:T}|x_0)} \\
&= \log \mathbb{E} \left(\frac{p(x_{0:T})}{q(x_{1:T}|x_0)} \right) \\
&\geq \mathbb{E}_q \left[\log \frac{p(x_{0:T})}{q(x_{1:T}|x_0)} \right] \\
&= \mathbb{E}_q \left[\log \frac{p(x_T) \prod_{t=1}^T p(x_{t-1}|x_t)}{\prod_{t=1}^T q(x_t|x_{t-1})} \right] \text{ (using equation 1 and 2)} \\
&= \mathbb{E}_q [\log(p_T)] + \mathbb{E}_q \left[\frac{\prod_{t=1}^T p(x_{t-1}|x_t)}{\prod_{t=1}^T q(x_t|x_{t-1})} \right] \\
&= \mathbb{E}_q [\log(p_T)] + \mathbb{E}_q \left[\prod_{t=1}^T \frac{p(x_{t-1}|x_t)}{q(x_t|x_{t-1})} \right] \\
&= \mathbb{E}_q [\log(p_T)] + \sum_{t=1}^T \mathbb{E}_q \left[\frac{p(x_{t-1}|x_t)}{q(x_t|x_{t-1})} \right]
\end{aligned}$$