

DDPM Training Objective Proof 2023

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Problem 1.

Please provide a detailed mathematical proof for the training objective of the Denoising Diffusion Probabilistic Model (DDPM), which aims to maximize the log-likelihood of data $\log p(x_0; \theta)$ where θ represents the model parameters. Your proof should demonstrate the derivation started from

$$\mathbb{E}_q \left[-\frac{\log p_\theta(x_{0:T})}{q(x_{1:T}|x_0)} \right]$$

to the training objective as following

$$\mathbb{E}_q \left[-\log p_\theta(x_0|x_1) + \sum_{t=2}^T D_{KL}(q(x_{t-1}|x_t, x_0) || p_\theta(x_{t-1}|x_t)) + D_{KL}(q(x_T|x_0) || p(x_T)) \right].$$

Hint: Given x_0, x_1, \dots, x_T , which form a first-order Markov chain. Then we have

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

Proof. For simplicity, we will first omit the expected value.

$$\begin{aligned} -\log \left(\frac{p_\theta(x_{0:T})}{q(x_{1:T}|x_0)} \right) &= -\log p(x_T) - \sum_{t=1}^T \log \left(\frac{p_\theta(x_{t-1}|x_t)}{q(x_t|x_{t-1})} \right) \\ &\text{since } q(x_{1:T}|x_0) = \prod_{t=1}^T q(x_t|x_{t-1}), \text{ and } p_\theta(x_{0:T}) = p(x_T) \prod_{t=1}^T p_\theta(x_{t-1}|x_t). \\ &= -\log p(x_T) - \sum_{t=2}^T \log \left(\frac{p_\theta(x_{t-1}|x_t)}{q(x_t|x_{t-1})} \right) - \log \left(\frac{p_\theta(x_0|x_1)}{q(x_1|x_0)} \right) \\ &= -\log p(x_T) - \sum_{t=2}^T \log \left(\frac{p_\theta(x_{t-1}|x_t)}{q(x_{t-1}|x_t, x_0)} \frac{q(x_{t-1}|x_0)}{q(x_t|x_0)} \right) - \log \left(\frac{p_\theta(x_0|x_1)}{q(x_1|x_0)} \right) \\ &= -\log \left(\frac{p(x_T)}{q(x_T|x_0)} \right) - \sum_{t=2}^T \log \left(\frac{p_\theta(x_{t-1}|x_t)}{q(x_{t-1}|x_t, x_0)} \right) - \log p_\theta(x_0|x_1) \\ &= D_{KL}(q(x_T|x_0) || p(x_T)) + \sum_{t=2}^T D_{KL}(q(x_{t-1}|x_t, x_0) || p_\theta(x_{t-1}|x_t)) - \log p_\theta(x_0|x_1). \end{aligned}$$

$$\begin{aligned} \mathbb{E}_q \left[-\log \left(\frac{p_\theta(x_{0:T})}{q(x_{1:T}|x_0)} \right) \right] &= \mathbb{E}_q [-\log p_\theta(x_0|x_1)] \\ &\quad + \sum_{t=2}^T \mathbb{E}_q [D_{KL}(q(x_{t-1}|x_t, x_0) || p_\theta(x_{t-1}|x_t)) + D_{KL}(q(x_T|x_0) || p(x_T))] \end{aligned}$$