

1 Maths

The *noising process* $q(\mathbf{z}_t \mid \mathbf{z}_{t-1}, \mathbf{x})$ is a Markov process which iteratively adds Gaussian noise to data \mathbf{x} until the signal is destroyed, $\mathbf{z}_T \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$.

We learn a Gaussian *generative denoising process* p_θ that aims to approximate the *true denoising process* $q(\mathbf{z}_{t-1} \mid \mathbf{z}_t, \mathbf{x})$ without knowledge of the data \mathbf{x} . Cross-entropy loss (often called NLL in the literature) is intractable, so a variational bound is used:

$$\mathbb{E}(-\log p_\theta(\mathbf{x})) \leq \mathbb{E}(-\log p_\theta(\mathbf{x} \mid \mathbf{z}_0)) + \sum_{t=1}^T \mathbb{E}[\text{KL}(q(\mathbf{z}_{t-1} \mid \mathbf{z}_t, \mathbf{x}) \parallel p_\theta(\mathbf{z}_{t-1} \mid \mathbf{z}_t))]$$

Parameterisation The noising process has a marginal distribution $\mathbf{z}_t \stackrel{\text{dist.}}{=} \alpha_t \mathbf{x} + \beta_t \boldsymbol{\varepsilon}$ where $\boldsymbol{\varepsilon} \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$.

$$p_\theta(\mathbf{z}_{t-1} \mid \mathbf{z}_t) := q(\mathbf{z}_{t-1} \mid \mathbf{z}_t, \hat{\mathbf{x}}_\theta(\mathbf{z}_t, t))$$

$$\min \mathbb{E}_{t, \mathbf{x}, \mathbf{z}_t} \left(\|\boldsymbol{\varepsilon} - \hat{\boldsymbol{\varepsilon}}_\theta(\mathbf{z}_t, t)\|^2 \right)$$