

#### A.4. Proof of Theorem 3.1

*Proof.* According to Eq. (2), when  $\gamma = 1$  and  $\sigma_t = 0$ , we have the following formula:

$$\mathbf{x}_{t-1} = \sqrt{\frac{\alpha_{t-1}}{\alpha_t}} \mathbf{x}_t - \sqrt{\alpha_{t-1}} \psi(\alpha_t, \alpha_{t-1}, 0) \epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y)$$

We can then represent  $\mathbf{x}_t$  as:

$$\begin{aligned} \mathbf{x}_t &= \sqrt{\frac{\alpha_t}{\alpha_{t-1}}} \mathbf{x}_{t-1} + \sqrt{\alpha_t} \left( \sqrt{\frac{1}{\alpha_t} - 1} - \sqrt{\frac{1}{\alpha_{t-1}} - 1} \right) \epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y) \\ &= \sqrt{\frac{\alpha_t}{\alpha_{t-1}}} \left( \sqrt{\frac{\alpha_{t-1}}{\alpha_{t-2}}} \mathbf{x}_{t-2} + \sqrt{\alpha_{t-1}} \left( \sqrt{\frac{1}{\alpha_{t-1}} - 1} - \sqrt{\frac{1}{\alpha_{t-2}} - 1} \right) \epsilon_\theta(\mathbf{x}_{t-1}, t-1, \mathcal{C}_y) \right) \\ &\quad + \sqrt{\alpha_t} \left( \sqrt{\frac{1}{\alpha_t} - 1} - \sqrt{\frac{1}{\alpha_{t-1}} - 1} \right) \epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y) \\ &= \sqrt{\frac{\alpha_t}{\alpha_{t-1}}} \left( \sqrt{\frac{\alpha_{t-1}}{\alpha_{t-2}}} \left[ \sqrt{\frac{\alpha_{t-2}}{\alpha_{t-3}}} \mathbf{x}_{t-3} + \sqrt{\alpha_{t-2}} \left( \sqrt{\frac{1}{\alpha_{t-2}} - 1} - \sqrt{\frac{1}{\alpha_{t-3}} - 1} \right) \epsilon_\theta(\mathbf{x}_{t-2}, t-2, \mathcal{C}_y) \right] \right. \\ &\quad \left. + \sqrt{\alpha_{t-1}} \left( \sqrt{\frac{1}{\alpha_{t-1}} - 1} - \sqrt{\frac{1}{\alpha_{t-2}} - 1} \right) \right) \epsilon_\theta(\mathbf{x}_{t-1}, t-1, \mathcal{C}_y) \\ &\quad + \sqrt{\alpha_t} \left( \sqrt{\frac{1}{\alpha_t} - 1} - \sqrt{\frac{1}{\alpha_{t-1}} - 1} \right) \epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y) \\ &= \sqrt{\frac{\alpha_t \alpha_{t-1} \dots \alpha_1}{\alpha_{t-1} \alpha_{t-2} \dots \alpha_0}} \mathbf{x}_0 + \sum_{i=1}^t \sqrt{\alpha_t} \left( \sqrt{\frac{1}{\alpha_i} - 1} - \sqrt{\frac{1}{\alpha_{i-1}} - 1} \right) \epsilon_\theta(\mathbf{x}_i, i, \mathcal{C}_y). \end{aligned} \quad (16)$$

Let  $\epsilon_\theta(\mathbf{x}_i, i, \mathcal{C}_y) - \epsilon_\theta(\mathbf{x}_{i-1}, i, \mathcal{C}_y) = \delta_i$ . Then, we have

$$\epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y) = -\sqrt{1 - \bar{\alpha}_t} \nabla \log p_\theta(\mathbf{x}_{t-1} | y) + \delta_t,$$

which leads to:

$$\begin{aligned} \mathbf{x}_t &= \sqrt{\frac{\alpha_t \alpha_{t-1} \dots \alpha_1}{\alpha_{t-1} \alpha_{t-2} \dots \alpha_0}} \mathbf{x}_0 - \sum_{i=0}^{t-1} \sqrt{\alpha_t (1 - \bar{\alpha}_{i+1})} \left( \sqrt{\frac{1}{\alpha_{i+1}} - 1} - \sqrt{\frac{1}{\alpha_i} - 1} \right) \nabla_{\mathbf{x}_i} \log p_\theta(\mathbf{x}_i | y) \\ &\quad + \sum_{i=0}^{t-1} \sqrt{\alpha_t} \left( \sqrt{\frac{1}{\alpha_{i+1}} - 1} - \sqrt{\frac{1}{\alpha_i} - 1} \right) \delta_{i+1}. \end{aligned} \quad (17)$$

Let  $s_i = \sqrt{\alpha_t (1 - \bar{\alpha}_{i+1})} \left( \sqrt{\frac{1}{\alpha_{i+1}} - 1} - \sqrt{\frac{1}{\alpha_i} - 1} \right)$ , for  $0 \leq i \leq t-1$ . Then the above expression simplifies to:

$$\mathbf{x}_t = \sqrt{\alpha_t} \mathbf{x}_0 - \sum_{i=0}^{t-1} s_i \nabla_{\mathbf{x}_i} \log p_\theta(\mathbf{x}_i | y) + \sum_{i=0}^{t-1} \frac{s_i}{1 - \sqrt{\bar{\alpha}_{i+1}}} \delta_{i+1}. \quad (18)$$

Applying Bayes' theorem, the conditional term can be rewritten as:

$$\mathbf{x}_t = \sqrt{\alpha_t} \mathbf{x}_0 - \sum_{i=0}^{t-1} \nabla_{\mathbf{x}_i} \log \left( \frac{p_\theta(\mathbf{x}_i) p_\theta(y | \mathbf{x}_i)}{p_\theta(y)} \right)^{s_i} + \sum_{i=0}^{t-1} \frac{s_i}{1 - \sqrt{\bar{\alpha}_{i+1}}} \delta_{i+1}. \quad (19)$$

Since the gradient of  $\log p_\theta(y)$  with respect to  $\mathbf{x}_i$  is zero, we obtain:

$$\mathbf{x}_t = \sqrt{\alpha_t} \mathbf{x}_0 - \sum_{i=0}^{t-1} [\nabla_{\mathbf{x}_i} \log p_\theta(\mathbf{x}_i)^{s_i} + \nabla_{\mathbf{x}_i} \log p_\theta(y | \mathbf{x}_i)^{s_i}] + \sum_{i=0}^{t-1} \frac{s_i}{1 - \sqrt{\bar{\alpha}_{i+1}}} \delta_{i+1}. \quad (20)$$

The proof is complete.  $\square$

### A.5. Proof of the Forward Process in Negative Instance Generation

*Proof.* From Eq. (2), when  $\gamma = 1$  and  $\sigma_t = 0$ , the following equation holds:

$$\mathbf{x}_{t-1} = \sqrt{\frac{\alpha_{t-1}}{\alpha_t}} \mathbf{x}_t - \sqrt{\alpha_{t-1}} \psi(\alpha_t, \alpha_{t-1}, 0) \epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y).$$

Thus, we can represent  $\mathbf{x}_t$  as:

$$\mathbf{x}_t = \sqrt{\frac{\alpha_t}{\alpha_{t-1}}} \mathbf{x}_{t-1} + \sqrt{\alpha_t} \psi(\alpha_t, \alpha_{t-1}, 0) \epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y).$$

Since  $\epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y)$  is not directly accessible, we adopt a forward Euler approximation, replacing  $\epsilon_\theta(\mathbf{x}_t, t, \mathcal{C}_y)$  with  $\epsilon_\theta(\mathbf{x}_{t-1}, t, \mathcal{C}_y)$  following DDIM (Song et al., 2020a). As a result, we obtain:

$$\mathbf{x}_t \approx \sqrt{\frac{\alpha_t}{\alpha_{t-1}}} \mathbf{x}_{t-1} + \sqrt{\alpha_t} \psi(\alpha_t, \alpha_{t-1}, 0) \epsilon_\theta(\mathbf{x}_{t-1}, t, \mathcal{C}_y).$$

The proof is complete.  $\square$

### A.6. Proof of Theorem 3.2

*Proof.* We begin with the following equation:

$$\tilde{\mathbf{x}}_{t-1} = \sqrt{\frac{\alpha_{t-1}}{\alpha_t}} \tilde{\mathbf{x}}_t - \sqrt{\alpha_{t-1}} \left( \sqrt{\frac{1}{\alpha_t}} - 1 - \sqrt{\frac{1}{\alpha_{t-1}}} - 1 \right) \epsilon_\theta(\tilde{\mathbf{x}}_t, t) \quad (21)$$

Rearranging, we obtain:

$$\tilde{\mathbf{x}}_t = \sqrt{\frac{\alpha_t}{\alpha_{t-1}}} \tilde{\mathbf{x}}_{t-1} + \sqrt{\alpha_t} \left( \sqrt{\frac{1}{\alpha_t}} - 1 - \sqrt{\frac{1}{\alpha_{t-1}}} - 1 \right) \epsilon_\theta(\tilde{\mathbf{x}}_t, t)$$

Following a similar derivation process as in Eq. (16), we obtain Eq. (22):

$$\tilde{\mathbf{x}}_t = \sqrt{\frac{\alpha_t \alpha_{t-1} \dots \alpha_1}{\alpha_{t-1} \alpha_{t-2} \dots \alpha_0}} \tilde{\mathbf{x}}_0 + \sum_{i=1}^t \sqrt{\alpha_t} \left( \sqrt{\frac{1}{\alpha_i}} - 1 - \sqrt{\frac{1}{\alpha_{i-1}}} - 1 \right) \epsilon_\theta(\tilde{\mathbf{x}}_i, i). \quad (22)$$

Let  $\epsilon_\theta(\tilde{\mathbf{x}}_i, i) - \epsilon_\theta(\tilde{\mathbf{x}}_{i-1}, t) = \tilde{\delta}_i$ . Based on Eq. (3), we have:

$$\epsilon_\theta(\tilde{\mathbf{x}}_t, t) = -\sqrt{1 - \bar{\alpha}_t} \nabla_{\tilde{\mathbf{x}}_{t-1}} \log p_\theta(\tilde{\mathbf{x}}_{t-1}) + \tilde{\delta}_t.$$

Therefore, we obtain the following equation for  $\tilde{\mathbf{x}}_t$ :

$$\tilde{\mathbf{x}}_t = \sqrt{\alpha_t} \tilde{\mathbf{x}}_0 - \sum_{i=0}^{t-1} \nabla_{\tilde{\mathbf{x}}_i} \log p_\theta(\tilde{\mathbf{x}}_i)^{s_i} + \sum_{i=0}^{t-1} \frac{s_i}{\sqrt{1 - \bar{\alpha}_{i+1}}} \tilde{\delta}_{i+1}. \quad (23)$$

In the initial reversion step, where  $\tilde{\mathbf{x}}_t = \mathbf{x}_t$ , we have:

$$\nabla \log p_\theta(\tilde{\mathbf{x}}_t) = \nabla \log p_\theta(\mathbf{x}_t).$$

Since

$$\nabla \log p_\theta(\tilde{\mathbf{x}}_t) = \nabla \log p_\theta(\tilde{\mathbf{x}}_{t-1}) - \frac{1}{\sqrt{1 - \bar{\alpha}_t}} \tilde{\delta}_t, \quad \text{and} \quad \nabla \log p_\theta(\mathbf{x}_t) = \nabla \log p_\theta(\mathbf{x}_{t-1}) - \frac{1}{\sqrt{1 - \bar{\alpha}_t}} \delta_t,$$

we obtain:

$$\nabla \log p_\theta(\tilde{\mathbf{x}}_{t-1}) = \nabla \log p_\theta(\mathbf{x}_{t-1}) - \frac{1}{\sqrt{1 - \bar{\alpha}_t}} \delta_t + \frac{1}{\sqrt{1 - \bar{\alpha}_t}} \tilde{\delta}_t.$$

By induction, we get:

$$\nabla \log p_\theta(\tilde{\mathbf{x}}_{t-k}) = \nabla \log p_\theta(\mathbf{x}_{t-k}) + \sum_{j=t-k}^{t-1} \frac{1}{\sqrt{1 - \bar{\alpha}_{j+1}}} [\tilde{\delta}_{j+1} - \delta_{j+1}],$$

or more generally:

$$\nabla \log p_\theta(\tilde{\mathbf{x}}_i) = \nabla \log p_\theta(\mathbf{x}_i) + \sum_{j=i}^{t-1} \frac{1}{\sqrt{1 - \bar{\alpha}_{j+1}}} [\tilde{\delta}_{j+1} - \delta_{j+1}].$$

Substituting into Eq. (23), we get:

$$\tilde{\mathbf{x}}_t = \sqrt{\alpha_t} \tilde{\mathbf{x}}_0 - \sum_{i=0}^{t-1} \nabla_{\mathbf{x}_i} \log p_\theta(\mathbf{x}_i)^{s_i} - \sum_{i=0}^{t-1} \sum_{j=i}^{t-1} \frac{s_i}{\sqrt{1 - \bar{\alpha}_{j+1}}} [\tilde{\delta}_{j+1} - \delta_{j+1}] + \sum_{i=0}^{t-1} \frac{s_i}{\sqrt{1 - \bar{\alpha}_{i+1}}} \tilde{\delta}_{i+1}. \quad (24)$$

Meanwhile, the corresponding equation from Theorem 3.1 is:

$$\mathbf{x}_t = \sqrt{\alpha_t} \mathbf{x}_0 - \sum_{i=0}^{t-1} [\nabla_{\mathbf{x}_i} \log p_\theta(\mathbf{x}_i)^{s_i} + \nabla_{\mathbf{x}_i} \log p_\theta(y|\mathbf{x}_i)^{s_i}] + \sum_{i=0}^{t-1} \frac{s_i}{\sqrt{1 - \bar{\alpha}_{i+1}}} \delta_{i+1}. \quad (25)$$

Equating both sides, we obtain:

$$\begin{aligned} & \sqrt{\alpha_t} \tilde{\mathbf{x}}_0 - \sum_{i=0}^{t-1} [\nabla_{\mathbf{x}_i} \log p_\theta(\mathbf{x}_i)^{s_i} + \nabla_{\mathbf{x}_i} \log p_\theta(y|\mathbf{x}_i)^{s_i}] + \sum_{i=0}^{t-1} \frac{s_i}{\sqrt{1 - \bar{\alpha}_{i+1}}} \delta_{i+1} \\ &= \sqrt{\alpha_t} \tilde{\mathbf{x}}_0 - \sum_{i=0}^{t-1} \nabla_{\mathbf{x}_i} \log p_\theta(\mathbf{x}_i)^{s_i} - \sum_{i=0}^{t-1} \sum_{j=i}^{t-1} \frac{s_i}{\sqrt{1 - \bar{\alpha}_{j+1}}} [\tilde{\delta}_{j+1} - \delta_{j+1}] + \sum_{i=0}^{t-1} \frac{s_i}{\sqrt{1 - \bar{\alpha}_{i+1}}} \tilde{\delta}_{i+1}. \end{aligned} \quad (26)$$

Solving for  $\tilde{\mathbf{x}}_0$ , we arrive at:

$$\tilde{\mathbf{x}}_0 = \mathbf{x}_0 - \frac{1}{\sqrt{\alpha_t}} \sum_{i=0}^{t-1} \nabla_{\mathbf{x}_i} \log p_\theta(y|\mathbf{x}_i)^{s_i} + \sum_{i=1}^{t-1} \sum_{j=i}^{t-1} \frac{s_i}{\sqrt{\alpha_t(1 - \bar{\alpha}_{j+1})}} [\tilde{\delta}_{j+1} - \delta_{j+1}]. \quad (27)$$

The proof is complete.  $\square$