

# Submission for Deep Learning Exercise

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## Pen Paper Task: Diffusion Models

If we have

$$x_t = \sqrt{\alpha_t}x_{t-1} + \sqrt{1 - \alpha_t}\epsilon_{t-1} \quad (1)$$

then

$$x_t = \left(\sqrt{\prod_{\tau=1}^t \alpha_\tau}\right)x_0 + \sum_{\tau=0}^{t-2} \left(\sqrt{\prod_{i=\tau+2}^t \alpha_i - \prod_{i=\tau+1}^t \alpha_i}\right)\epsilon_\tau + (\sqrt{1 - \alpha_t})\epsilon_{t-1} \quad (2)$$

When  $t = 1$ , it is easy to see equations (1) and (2) are the same. For an arbitrary  $t$

$$\begin{aligned} x_{t-1} &= \left(\sqrt{\prod_{\tau=1}^{t-1} \alpha_\tau}\right)x_0 + \sum_{\tau=0}^{t-3} \left(\sqrt{\prod_{i=\tau+2}^{t-1} \alpha_i - \prod_{i=\tau+1}^{t-1} \alpha_i}\right)\epsilon_\tau + (\sqrt{1 - \alpha_{t-1}})\epsilon_{t-2} \\ (\sqrt{\alpha_t})x_{t-1} &= \left(\sqrt{\prod_{\tau=1}^t \alpha_\tau}\right)x_0 + \sum_{\tau=0}^{t-3} \left(\sqrt{\prod_{i=\tau+2}^t \alpha_i - \prod_{i=\tau+1}^t \alpha_i}\right)\epsilon_\tau + (\sqrt{\alpha_t - \alpha_t \alpha_{t-1}})\epsilon_{t-2} \\ &= \left(\sqrt{\prod_{\tau=1}^t \alpha_\tau}\right)x_0 + \sum_{\tau=0}^{t-2} \left(\sqrt{\prod_{i=\tau+2}^t \alpha_i - \prod_{i=\tau+1}^t \alpha_i}\right)\epsilon_\tau \\ (\sqrt{\alpha_t})x_{t-1} + (\sqrt{1 - \alpha_t})\epsilon_{t-1} &= \left(\sqrt{\prod_{\tau=1}^t \alpha_\tau}\right)x_0 + \sum_{\tau=0}^{t-2} \left(\sqrt{\prod_{i=\tau+2}^t \alpha_i - \prod_{i=\tau+1}^t \alpha_i}\right)\epsilon_\tau + (\sqrt{1 - \alpha_t})\epsilon_{t-1} = x_t \end{aligned}$$

Therefore, by induction we have proven that the equation (2) is true.

We should have an equation of the form

$$x_t = \mu + \sigma \cdot \epsilon \quad (3)$$

We found that  $\mu = (\sqrt{\prod_{\tau=1}^t \alpha_\tau})x_0 = (\sqrt{\tilde{a}})x_0$ . By the property of *sum of normally distributed random variables* if

$$X_i \sim N(\mu_i, \sigma_i^2) \quad (4)$$

and

$$Y = \sum_{i=1}^n c_i X_i \quad (5)$$

then

$$Y \sim N\left(\sum_{i=1}^n c_i \mu_i, \sum_{i=1}^n c_i^2 \sigma_i^2\right) \quad (6)$$

We know  $\epsilon \sim N(0, I)$ . Substituting  $X_i = \epsilon_i$ ,

$$\begin{aligned} \sum_{\tau=0}^{t-2} \left( \prod_{i=\tau+2}^t \alpha_i - \prod_{i=\tau+1}^t \alpha_i \right) + (1 - \alpha_t) &= \prod_{i=2}^t \alpha_i - \prod_{i=1}^t \alpha_i + \prod_{i=3}^t \alpha_i - \prod_{i=2}^t \alpha_i + \dots + \prod_{i=t-1}^t \alpha_i - \prod_{i=t-2}^t \alpha_i + \prod_{i=t}^t \alpha_i - \prod_{i=t-1}^t \alpha_i + (1 - \alpha_t) \\ &= \prod_{i=2}^t \alpha_i - \prod_{i=1}^t \alpha_i + \prod_{i=3}^t \alpha_i - \prod_{i=2}^t \alpha_i + \dots + \prod_{i=t-1}^t \alpha_i - \prod_{i=t-2}^t \alpha_i + \alpha_t - \prod_{i=t-1}^t \alpha_i + 1 - \alpha_t \end{aligned}$$

Note that similar elements cancel each other and we find  $\sigma^2 = 1 - \prod_{i=1}^t \alpha_i = 1 - \tilde{a}$ .