

扩散模型

$$dX_t = \underline{f(t)X_t dt + g(t)dW_t}$$

$$d\left(\frac{X_t}{\lambda_t}\right) = \frac{1}{\lambda_t} dX_t - \frac{d\lambda_t}{\lambda_t^2} X_t$$

需要 $\frac{1}{\lambda_t} f(t) dt = \frac{d\lambda_t}{\lambda_t^2}$ $f(t) = \frac{d \log \lambda_t}{dt} (\lambda_0=1)$

$$L(t, X_t) = \frac{\lambda_t}{\lambda_t}, \text{ 使用 Ito 公式}$$

$$dL(t, X_t) = \frac{\partial L}{\partial t} dt + \frac{\partial L}{\partial X} dX_t + \frac{1}{2} \frac{\partial^2 L}{\partial X^2} g(t)^2 dt$$

$$\begin{aligned} &= \frac{\partial L}{\partial t} dt + \frac{\partial L}{\partial X} f(t) X_t dt + \frac{\partial L}{\partial X} g(t) dW_t \\ &\quad + \frac{1}{2} \frac{\partial^2 L}{\partial X^2} g(t)^2 dt \end{aligned}$$

$$= \frac{-\dot{\lambda}_t}{\lambda_t^2} X_t dt + \frac{f(t) X_t}{\lambda_t} dt + \frac{g(t)}{\lambda_t} dW_t$$

$$= \frac{g(t)}{\lambda_t} dW_t$$

$$\frac{X_t}{\lambda_t} = \frac{X_0}{\lambda_0} + \underbrace{\int_0^t \frac{g(\tau)}{\lambda_\tau} dW_\tau}_{\text{高斯}}$$

方差为

$$\int_0^t \frac{g(\tau)^2}{\lambda_\tau^2} d\tau = \frac{\beta_t^2}{\lambda_t^2} \Rightarrow \frac{X_t}{\lambda_t} = \frac{X_0}{\lambda_0} + \frac{\beta_t}{\lambda_t} W$$

$$X_t = \lambda + X_0 + \beta_t W$$

$$\begin{aligned} \frac{d}{dt} \frac{\beta_t^2}{\lambda_t^2} &= \frac{g(t)^2}{\lambda_t^2} \\ &= \frac{1}{\lambda_t^2} \frac{d}{dt} \beta_t^2 + \beta_t^2 \frac{-\dot{\lambda}_t}{\lambda_t^4} \end{aligned}$$

$$g(t)^2 = \frac{d}{dt} \beta_t^2 - 2 \beta_t^2 \frac{d \log \lambda_t}{dt}$$

正向方程

$$\partial_t q_t(x) = -\nabla \cdot (f(t)x q_t(x)) + \frac{1}{2} \Delta (g(t)^2 q_t(x))$$

$$\partial_t \rho_t(x) = \nabla (f(t-x) x \rho_t(x)) - \frac{1}{2} \Delta (g(t-x)^2 \rho_t(x))$$

$$dY_t = \hat{f}(t, Y_t) dt + \hat{g}(t) dW_t$$

\Rightarrow

$$\partial_t \rho_t(x) = -\nabla (\hat{f}(t, x) \rho_t(x)) + \frac{1}{2} \Delta (\hat{g}(t)^2 \rho_t(x))$$

OU 过程

$$f(t) = -1 \quad g(t) = \sqrt{2}$$

$$\log \lambda_t = -t \Rightarrow \lambda_t = e^{-t}$$

$$2 = \frac{d\beta_t^2}{dt} + 2\beta_t^2 \quad d(1-x) = -2(1-x)dt \\ \Rightarrow 1 - \beta_t^2 = e^{-2t}, \quad \beta_t^2 = 1 - e^{-2t}$$

分数匹配

$S_\theta(t, x)$ 口鼻音 \rightarrow 图片

$$\begin{aligned} & E_{q_{0,T-t}} \|S_\theta(t, x) - \nabla \log q_{T-t}(x)\|^2 \\ &= \iint q_{0,T-t}(x|x_0) q_0(x_0) \|S_\theta(t, x) - \nabla \log q_{T-t}(x)\|^2 dx dx_0 \\ &= \iint \underbrace{q_{0,T-t}(x|x_0)}_{\text{分子}} \underbrace{q_0(x_0)}_{\text{分子}} \left[S_\theta(t, x)^2 - 2S_\theta(t, x) \cdot \nabla \log q_{T-t}(x) \right] dx dx_0 \\ & \int S_\theta(t, x) \nabla q_{T-t}(x) dx \\ &= \int S_\theta(t, x) \int \nabla_x q_{0,T-t}(x|x_0) q_0(x_0) dx_0 dx \\ &= \iint S_\theta(t, x) \int \nabla_x \log q_{0,T-t}(x|x_0) \underbrace{q_{0,T-t}(x|x_0) q_0(x_0)}_{\text{分子}} dx_0 dx \\ &\text{分子} \\ &= \iint q_{0,T-t}(x|x_0) q_0(x_0) \left[S_\theta(t, x) - \nabla_x \log q_{0,T-t}(x|x_0) \right]^2 \end{aligned}$$

数值实现

$$\int_0^{T-t} w_{T-t} E_{q_0} E_{q_{0,T-t}(x|x_0)} \| S_0(t, x) - \nabla \log q_{0,T-t}(x|x_0) \|^2 \quad T-t \rightarrow t$$

$$= \int_T^T w_t E_{q_0} E_{q_{0,t}(x|x_0)} \| S_0(T-t, x) - \nabla \log q_{0,t}(x|x_0) \|^2$$

权重: $w_t E_{q_0} E_{q_{0,t}(x|x_0)} \| \nabla \log q_{0,t}(x|x_0) \|^2$

$$= w_t E_{q_0} E_{q_{0,t}(x|x_0)} \frac{x-m_t}{\delta_t^2} \cdot \frac{x-m_t}{\delta_t^2}$$

$$= w_t E_{q_0} \frac{d \delta_t^2}{\delta_t^4} \approx \text{常数}$$

$$w_t \sim \delta_t^2 = 1 - e^{-2t}$$

$$= \int_T^T w_t E_{q_0} E_{q_{0,t}(x|x_0)} \| S_0(T-t, x) + \frac{x - e^{-t} x_0}{\delta_t^2} \|^2$$

定义 $\varepsilon_0(T-t, x) = -S_0(T-t, x) - \delta_t$

$$= \int_T^T w_t E_{q_0} E_{q_{0,t}(x|x_0)} \| -\frac{\varepsilon_0(T-t, x)}{\delta_t} + \frac{x - e^{-t} x_0}{\delta_t^2} \|^2$$

$$= \int_T^T \frac{w_t}{\delta_t^2} E_{q_0} E_{q_{0,t}(x|x_0)} \| -\varepsilon_0(T-t, x) + \frac{x - e^{-t} x_0}{\delta_t} \|^2$$

$$\text{由 } x = \lambda_t x_0 + \delta_t w$$

$$E_{q_{0:t}(x|x_0)} \left[\left| -\varepsilon_0(T-t, x) + \frac{x - e^{t} x_0}{\delta_t} \right|^2 \right]$$

$$\approx \frac{1}{N} \sum_{i=1}^N \left\| \varepsilon_0(T-t_i, \lambda_t x_0 + \delta_t w_i) - w_i \right\|^2$$