

$$Q1. \text{ Prove } LSSM = E_{x \sim p(x)} E_{v \sim p(v)} [\| V^T S(x; \theta) \|^2 + 2V^T J_x \\ [V^T S(x; \theta)]]$$

Step 1. From original def.

$$LSSM = E_{v \sim p(v)} E_{x \sim p(x)} [\| V^T S(x; \theta) - V^T J_x \log p(x) \|^2]$$

$$\text{Step 2. } \| A - B \|^2 = \| A \|^2 - 2A^T B + \| B \|^2$$

$$LSSM(\theta) = \bar{E}_{V, X} [\| V^T S(x; \theta) \|^2 - 2(V^T S(x; \theta)) (V^T J_x \log p(x)) \\ + \| V^T J_x \log p(x) \|^2]$$

$$\text{Step 3. } \bar{E}_{V, X} [\| V^T J_x \log p(x) \|^2] = \text{const}$$

$$\text{Step 4. } E_{p(x)} [g(x)^T J_x \log p(x)] = -E_{p(x)} [J_x p(x)]$$

$$\text{Step 5. } E_x [(V^T S(x; \theta)) (V^T J_x \log p(x))]$$

$\Leftarrow g(x) = V(V^T S(x; \theta))$ V for x is a const vector

$V^T S$ for x is a ~~func~~
scalar func. $\Rightarrow g(x)$ is a vec func

$$g(x)^T J_x \log p(x) = [V(V^T S)]^T J_x \log p(x) = (\cancel{V^T S}) \cancel{V^T J_x \log p(x)} \\ = (V^T S) V^T J_x \log p(x)$$

$$\text{From step 4, } E_x [(V^T S)(V^T J_x \log p(x))] = E_x [J_x (V^T \\ S(x; \theta))]$$

6.

 $\nabla_x(V(x))$, where $V \in \text{const vec}$ $c(x) = \sqrt{S}(x; \theta) \in \text{scalar func.}$

$$\nabla_x(V(x)) = \sum_i \frac{\partial}{\partial x_i} (V_i(x)) = \sum_i V_i \underbrace{\frac{\partial c(x)}{\partial x_i}}_{= \nabla_x(\sqrt{S}(x; \theta))}$$

$$\Rightarrow \nabla_x(V(\sqrt{S}(x; \theta))) = \sqrt{S} \nabla_x(\sqrt{S}(x; \theta))$$

7.

$$E_x[V^T S] (V^T \nabla_x \log p(x))] = -E_x[\sqrt{S} \nabla_x(\sqrt{S}(x; \theta))]$$

$$\Leftrightarrow L_{SSM} = \bar{E}_{V, X} [\| \sqrt{S}(x; \theta) \|^2 - 2 \cdot (-E_x[V^T \nabla_x(\sqrt{S}(x; \theta))])]$$

$$L_{SSM} = \bar{E}_{V, X} [\| \sqrt{S}(x; \theta) \|^2 + 2V^T \nabla_x(V^T S(x; \theta))]$$

Q2.

$$\text{SDE: } dx_t = \underbrace{f(x_t, t) dt}_{\text{Drift}} + \underbrace{G(x_t, t) dW_t}_{\text{diffusion}}$$

(deterministic) (stochastic)

$$W_t = W_0 = 0$$

$$W_{t+u} - W_t \sim N(0, uI)$$

dW_t each other

is indep for

W_t is continuous (but ~~nowhere~~ differentiable)

4 solve: $\int G dW_s$ is a Itô Integral

5 find solution: SDE 很難找解

Q3.

1. SDE 與材料中的原子擴散關係

$$\frac{\partial P}{\partial t} = \frac{\sigma^2}{2} \frac{\partial^2 P}{\partial x^2} \quad \text{vs} \quad \left(\frac{\partial C}{\partial t} = D \frac{\partial^2 C}{\partial x^2} \right) \text{ Fick's 2nd law}$$

2. SDE 與材料的相變能核化。

$f(x, t) dt$ with ~~表面能~~
表面能

$G(x, t) dW_t$ with 热波動隨机。

SDE is the 成核過程的 math tool?

3. 我們可以將一個材料的結構 (x_t) 視為 SDE 的解 x_t , train NN to learn 反向 SDE, 從而從 $N(0, I)$ 生成出一個全新物理上合理的 structure.