

第七章 随机积分

§7.1 随机积分定义

(1) Lebesgue积分 (Ω, \mathcal{F}, P)

考虑 $A = \{A_t, t \geq 0\}$ 为有界变差过程 i.e. $\forall w \in \Omega, \{A_t(w), t \geq 0\}$ 为有界变差函数

函数 $f: [0, +\infty) \times \Omega \rightarrow \mathbb{R} \quad \mathcal{B}([0, +\infty)) \otimes \mathcal{F} \rightarrow \mathcal{B}_{\mathbb{R}}$ 可测且有界

由 Lebesgue 积分定义 $\int_0^t f(s) dA(s)$ 为

$$\text{对 } \forall w \in \Omega, (\int_0^t f(s) dA(s))(w) \triangleq \int_0^t f(s, w) dA(s, w)$$

(2) 定义关于布朗运动 $\{B(s), s \geq 0\}$ 的积分 $\int_0^t f(s) dB(s)$

这里考虑 f 是一个与 w 无关的确定的函数。

对 $\forall w \in \Omega, \{B(s, w), s \geq 0\}$ 不是有界变差函数，因此，(1) 中定义方式对 BM 不适用。

△ 若 f 与 w 无关，关于 s 连续的有界变差函数 (e.g. $s^2, \sin s$)

我们可以利用分部积分公式定义 $\int_0^t f(s) dB(s)$

$$(\int_0^t f(s) dB(s))(w) \triangleq f(t)B(t, w) - \int_0^t B(s, w) df(s) \xrightarrow[\text{IDEAL}]{} \text{缺点：限制太大}$$

△ 回忆 Riemann 积分 $\int_0^t \psi(s) ds$

△ $0 = t_0 < t_1 < \dots < t_n = t$ 取 $\theta_i \in [t_i, t_{i+1}]$

$$\int_0^t \psi(s) ds \triangleq \lim_{\Delta t \rightarrow 0} \sum_{i=0}^{n-1} \psi_{\theta_i}(t_{i+1} - t_i)$$

若 $g(s), 0 \leq s \leq t$ 有界变差函数，定义 $\int_0^t \psi(s) dg(s) \triangleq \lim_{\Delta t \rightarrow 0} \sum_{i=1}^{n-1} \psi_{\theta_i}(g(t_{i+1}) - g(t_i))$

→ 与 Lebesgue 积分思想不同，存在 Lebesgue 可积但 Riemann 不可积的情况

△ 简单情况：△ $0 = t_0 < t_1 < \dots < t_n = t$

若 f 满足 $f(s) = f_{j-1}, t_{j-1} < s \leq t_j, f(0) = f_0$ ，其中 $f_{j-1}, j \geq 1$ 为常数

称这样的函数为阶梯函数 (step function)

$$\text{则 } \int_0^t f(s) dB(s) \triangleq \sum_{j=1}^n f_{j-1} (B(t_j) - B(t_{j-1}))$$

① 记 $\mathcal{F}_t = \sigma(B(s), 0 \leq s \leq t)$ 则 $\int_0^t f(s) dB(s) \in \mathcal{F}_t$

$$② E[\int_0^t f(s) dB(s)] = \sum_{j=1}^n f_{j-1} E(B(t_j) - B(t_{j-1})) = 0$$

$$③ E[(\int_0^t f(s) dB(s))^2] = \sum_{i,j=1}^n f_{i-1} f_{j-1} E[(B(t_i) - B(t_{i-1}))(B(t_j) - B(t_{j-1}))]$$

$$= \sum_{j=1}^n f_{j-1}^2 (t_j - t_{j-1}) = \int_0^T f(s)^2 ds$$

④ 若 $g \sim$ 阶梯函数, 则 $\int_0^T f(s) dB(s) - \int_0^T g(s) dB(s) = \int_0^T (f(s) - g(s)) dB(s)$

$$\Rightarrow E[(\int_0^T f(s) dB(s) - \int_0^T g(s) dB(s))^2] = \int_0^T |f(s) - g(s)|^2 ds$$

⑤ 对 $\forall f \in L^2([0, T], \mathbb{R})$, i.e., $\int_0^T |f(s)|^2 ds < \infty$

存在 $f_n \sim$ 阶梯函数, s.t. $\int_0^T |f(s) - f_n(s)|^2 ds \rightarrow 0$ as $n \rightarrow \infty$

⑥ 考虑 $t \in [0, T]$, 由 ④ $E[(\int_0^T f_n(s) dB(s) - \int_0^T f_m(s) dB(s))^2] = \int_0^T |f_n(s) - f_m(s)|^2 ds \xrightarrow{n,m \rightarrow \infty} 0$

$\{\int_0^T f_n(s) dB(s), n \in \mathbb{N}\}$ 为 $L^2(\Omega, \mathcal{F}, P)$ 中柯西列

$$L^2(\Omega, \mathcal{F}_t, P) = \{h(\Omega, \mathcal{F}_t) \rightarrow \mathbb{R}, E|h(\omega)|^2 < \infty\}$$

$$\exists L^2(\Omega, \mathcal{F}_t, P) \text{ 中 r.v. } U \text{ s.t. } E[\int_0^T f_n(s) dB(s) - U]^2 \rightarrow 0 \text{ as } n \rightarrow \infty$$

$$\int_0^T f(s) dB(s) \triangleq U = \lim_{n \rightarrow \infty} \int_0^T f_n(s) dB(s) \quad (\text{在 } L^2(\Omega, \mathcal{F}_t, P) \text{ 中取极限})$$

上述定义包含 $f \in L^2([0, T], \mathbb{R})$, 但不包含 $\int_0^T B(s) dB(s)$

由上述思想, 可以推广关于布朗运动的积分

推广到 $L_{C,T}^2 \{ \phi = \{\phi(s)\}_{0 \leq s \leq T} : \phi \text{ 关于 } s \text{ 连续且 } \phi(s) \in \mathcal{F}_s \text{ 且 } E[\int_0^T |\phi(s)|^2 ds] < \infty \}$

此时 $B_s, B_s^2, \dots \in L_{C,T}^2$

(3) $L_{C,T}^2$ 上的随机积分

① $f = \{f_t, 0 \leq t \leq T\}$ 为 (Ω, \mathcal{F}, P) 上的随机过程

f 满足: $\exists 0 = t_0 < t_1 < \dots < t_n = T, \exists f_{j-1} \in \mathcal{F}_{t_{j-1}}, E|f_{j-1}|^2 < \infty$

$$\text{s.t. } f(s) = f_{j-1}, t_{j-1} < s \leq t_j, f(0) = f_0.$$

这样的 f 称为 random step process

② 对上述 f 定义 $\int_0^T f(s) dB(s) \triangleq \sum_{j=1}^n f_{j-1} (B_{t_j} - B_{t_{j-1}})$

③ $\int_0^T f(s) dB(s)$ 的性质:

1) $\int_0^T f(s) dB(s) \in \mathcal{F}_T$

$$2) E[\int_0^T f(s) dB(s)] = \sum_{j=1}^n E[f_{j-1} (B_{t_j} - B_{t_{j-1}})] = 0$$

$$- E[f_{j-1} (B_{t_j} - B_{t_{j-1}})] = E[E[f_{j-1} (B_{t_j} - B_{t_{j-1}}) | \mathcal{F}_{t_{j-1}}]]$$

$$= E[f_{j-1} E[B_{t_j} - B_{t_{j-1}} | \mathcal{F}_{t_{j-1}}]]$$

$$= E[f_{j-1} E(B_{t_j} - B_{t_{j-1}})] = 0$$

$$3) E\left[\int_0^T f(s) dB(s)\right]^2 = \sum_{i,j=1}^n E[f_{i-1}(B_{t_i} - B_{t_{i-1}}) \cdot f_{j-1}(B_{t_j} - B_{t_{j-1}})]$$

* i ≠ j 时期望为 0

$$\stackrel{\downarrow}{=} \sum_{i=1}^n E[f_{i-1}^2(B_{t_i} - B_{t_{i-1}})^2]$$

与 2) 一样取条件期望算

$$\stackrel{\downarrow}{=} \sum_{i=1}^n E[f_{i-1}^2](t_i - t_{i-1})$$

$$= E\left[\sum_{i=1}^n f_{i-1}^2(t_i - t_{i-1})\right]$$

$$= E\left(\int_0^T |f(s)|^2 ds\right)$$

$$\left(*: 不妨设 i < j, 则 E[f_{i-1}(B_{t_i} - B_{t_{i-1}}) \cdot f_{j-1}(B_{t_j} - B_{t_{j-1}})] \right.$$

$$\left. = E[E[\dots |f_{t_{j-1}}]] = E[\dots E[B_{t_j} - B_{t_{j-1}} | f_{t_{j-1}}]] = 0 \right)$$

④ 推广到 $L_{C,T}^2$.

对 $\forall \phi \in L_{C,T}^2, \exists \phi_n \sim \text{random step process}, \text{s.t. } E \int_0^T |\phi_n(t) - \phi(t)|^2 dt \xrightarrow{n \rightarrow \infty} 0$

由 $\int_0^T \phi_n(t) dB(t)$ 性质知 $\int_0^T \phi_n(t) dB(t) \in L^2(\Omega, \mathcal{F}_T, P)$ (证明见下节橙框)

$$E\left[\int_0^T \phi_n(t) dB(t) - \int_0^T \phi_m(t) dB(t)\right]^2 = E \int_0^T |\phi_m(t) - \phi_n(t)|^2 dt \rightarrow 0 \text{ as } m, n \rightarrow \infty$$

$\Rightarrow \{\int_0^T \phi_n(t) dB(t), n \in \mathbb{N}\}$ 为 $L^2(\Omega, \mathcal{F}_T, P)$ 中的柯西列

△ 记 $\lim_{n \rightarrow \infty} \int_0^T \phi_n(t) dB(t) = X$ in $L^2(\Omega, \mathcal{F}_T, P)$

则 $\int_0^T \phi(s) dB(s) \triangleq X$

(说明: $\int_0^T \phi(s) dB(s)$ 与 ϕ_n 的选取无关)

$\forall \phi \in L_{C,T}^2, \exists \phi_n \sim \text{random step process}, \text{s.t. } E \int_0^T |\phi_n(t) - \phi(t)|^2 dt \xrightarrow{n \rightarrow \infty} 0$

其中 $L_{C,T}^2 \{ \phi = \{\phi(s)\}_{0 \leq s \leq T} : \phi \text{ 关于 } s \text{ 连续且 } \phi(s) \in \mathcal{F}_s \text{ 且 } E \int_0^T |\phi(s)|^2 ds < \infty \}$

Pf. 定义 $\Psi^{(n)}(t) = (\phi(t) \wedge n) \vee (-n) \in L_{C,T}^2$

$$\text{则 } E \int_0^T |\Psi^{(n)}(t) - \phi(t)|^2 dt \leq E \int_0^T |\phi(t)|^2 \mathbf{1}_{(-\infty, -n) \cup (n, +\infty)}(\phi(t)) dt \xrightarrow{n \rightarrow \infty} 0 \text{ as } n \rightarrow \infty$$

(由 $E \int_0^T |\phi(t)|^2 dt < \infty \Rightarrow \int_0^T |\phi(t)|^2 dt < \infty$ P.a.s.)

$$\Rightarrow \int_0^T |\phi(t)|^2 \mathbf{1}_{(-\infty, -n) \cup (n, +\infty)}(\phi(t)) dt \xrightarrow{n \rightarrow \infty} 0 \text{ P.a.s.}$$

由控制收敛定理, $E \int_0^T |\phi(t)|^2 \mathbf{1}_{(-\infty, -n) \cup (n, +\infty)}(\phi(t)) dt \rightarrow 0$ as $n \rightarrow \infty$

· 又由于 $\Psi^{(n)} \in L_{C,T}^2$, 考察 $\phi = \Psi^{(n)}$ P.T.Q.

· 不妨设 $\sup_{w \in \Omega} \sup_{0 \leq s \leq T} |\phi(s, \omega)| \leq M$

$$\sum h_n : h_n(t) = \begin{cases} \phi(0), & 0 \leq t \leq \frac{T}{n} \\ \phi\left(\frac{jT}{n}\right), & \frac{jT}{n} < t \leq \frac{(j+1)T}{n} \end{cases}$$

则 h_n 为 random step process.

由 ϕ 关于 $s \in [0, T]$ 连续 $\Rightarrow \{\phi(s, \omega), 0 \leq s \leq T\}$ 为连续

$$\Rightarrow \text{fix } \omega, \int_0^T |\phi(s, \omega) - h_n(s, \omega)|^2 ds \rightarrow 0 \text{ as } n \rightarrow \infty$$

$$\text{且 } \int_0^T |\phi(s, \omega) - h_n(s, \omega)|^2 ds \leq 4M^2 T$$

由控制收敛定理, $E \int_0^T |h_n(s) - \phi(s)|^2 ds \rightarrow 0 \text{ as } n \rightarrow \infty$

对 $\forall f \in L^2([0, T], \mathbb{R})$, 记 $X(t) = \int_0^t f(s) dB(s)$, $\forall t \in [0, T]$. $X = \{X(t), 0 \leq t \leq T\}$ 为 r.v.

$$(1) E|X_t|^2 = \int_0^t f^2(s) ds$$

(2) $\{X_t, 0 \leq t \leq T\}$ 关于 t 连续

i.e. $\exists \{\tilde{X}(t), 0 \leq t \leq T\}$ s.t. fix ω , $\tilde{X}(t, \omega)$ 关于 t 连续 且 $P(X(t) = \tilde{X}(t)) = 1$

(3) $\{X_t, 0 \leq t \leq T\}$ 关于 $f_t = \int_0^t dB(s)$ 为连续平方可积鞅

$$EX_0 = EX_t = 0$$

$$(4) \text{ 对 } g \in L^2([0, T], \mathbb{R}), \text{ 则 } E \left[\int_0^t f(s) dB(s) \cdot \int_0^t g(s) dB(s) \right] = \int_0^t f(s) g(s) ds$$

特别地, $g = f$ 时, $EX_t^2 = E \int_0^t f^2(s) ds$ Itô 等距

$$(5) \forall a, b \in \mathbb{R}, \text{ 有 } \int_s^t (af(c) + bg(c)) dB(c) = a \int_s^t f(c) dB(c) + b \int_s^t g(c) dB(c)$$

$$\int_s^t f(c) dB(c) \triangleq \int_0^t f(c) dB(c) - \int_0^s f(c) dB(c)$$

(6) 若 $f \in L^2([0, T], \mathbb{R})$, 则 $\int_0^t f(s) dB(s) \sim ?$ 分布.

作业: (6), 2, 3, 15

$\forall f \in L^2([0, T], \mathbb{R}), \forall t \in [0, T], \text{ 记 } X(t) = \int_0^t f(s) dB(s)$ (1°)

$$(1) E|X(t)|^2 = \int_0^t f^2(s) ds$$

Pf. 由 $X(t)$ 定义, $\exists f_n$ 正阶梯函数, s.t. $\lim_{n \rightarrow \infty} \int_0^T |f_n(s) - f(s)|^2 ds = 0$

$$\Rightarrow \int_0^T |f_n(s)|^2 ds \rightarrow \int_0^T |f(s)|^2 ds \text{ as } n \rightarrow \infty \quad (1)$$

$X_n(t) \triangleq \int_0^t f_n(s) dB(s)$, 则 $X(t)$ 为 $X_n(t)$ 在 $L^2(\Omega, \mathcal{F}_t, P)$ 下的极限

i.e., $\lim_{n \rightarrow \infty} E|X(t) - X_n(t)|^2 = 0 \Rightarrow E|X(t)|^2 \rightarrow E|X(t)|^2 \text{ as } n \rightarrow \infty$ ②

$$E|X_n(t)|^2 = \int_0^t |f_n(s)|^2 ds \quad \& \text{①} \& \text{②} \Rightarrow E|X(t)|^2 = \int_0^t |f(s)|^2 ds.$$

(4) 对 $g \in L^2([0, T], \mathbb{R})$, 则 $E[\int_0^t f(s) dB(s) \cdot \int_0^t g(s) dB(s)] = \int_0^t f(s) g(s) ds$

Pf. $E(\int_0^t f(s) + g(s) dB(s))^2 = \int_0^t (f+g)^2 ds \quad \triangle$

$$E(\int_0^t f(s) - g(s) dB(s))^2 = \int_0^t (f-g)^2 ds \quad \triangle$$

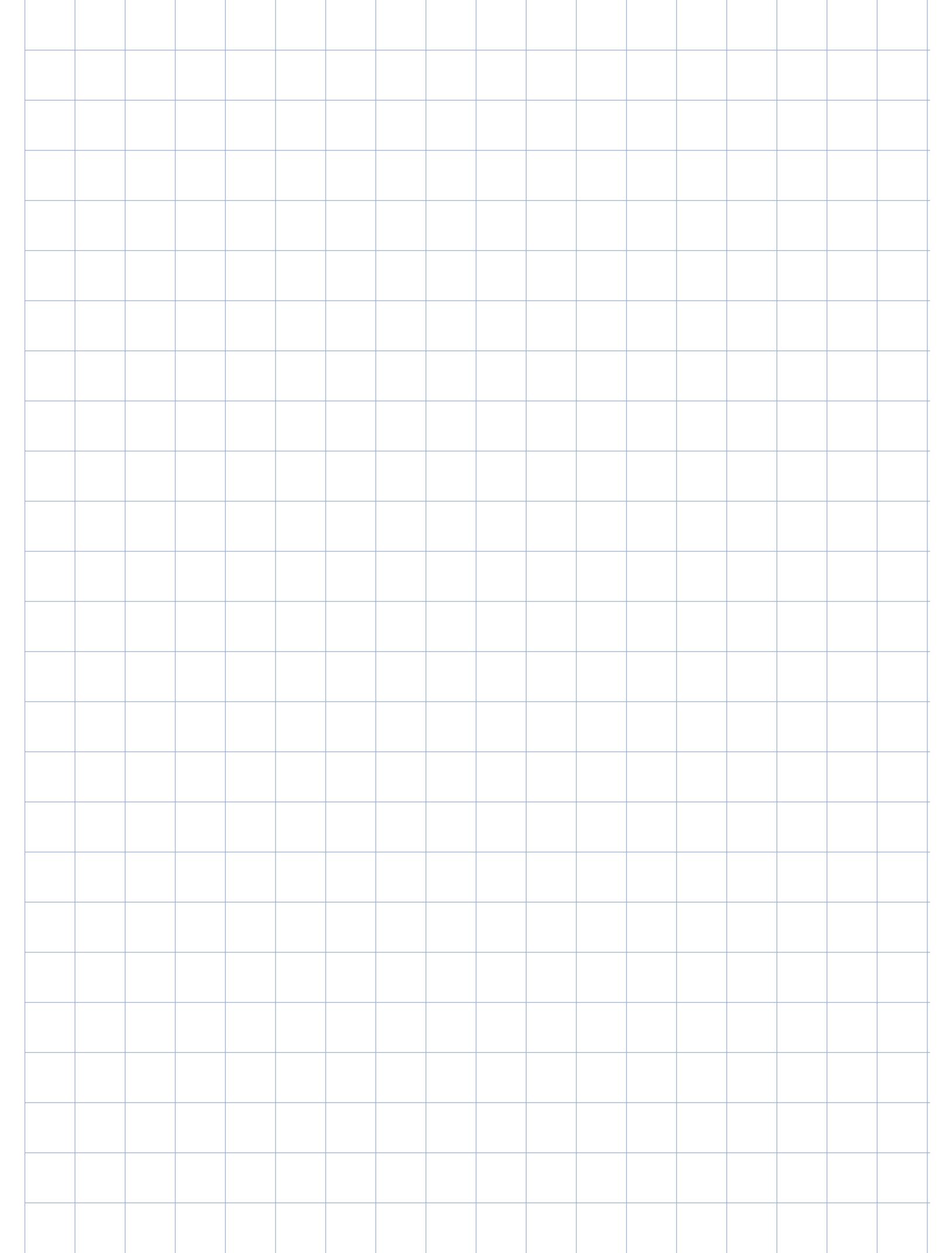
$$\triangle - \triangle \Rightarrow 4E[\int_0^t f(s) dB(s) \cdot \int_0^t g(s) dB(s)] = 4 \int_0^t f(s) g(s) ds$$

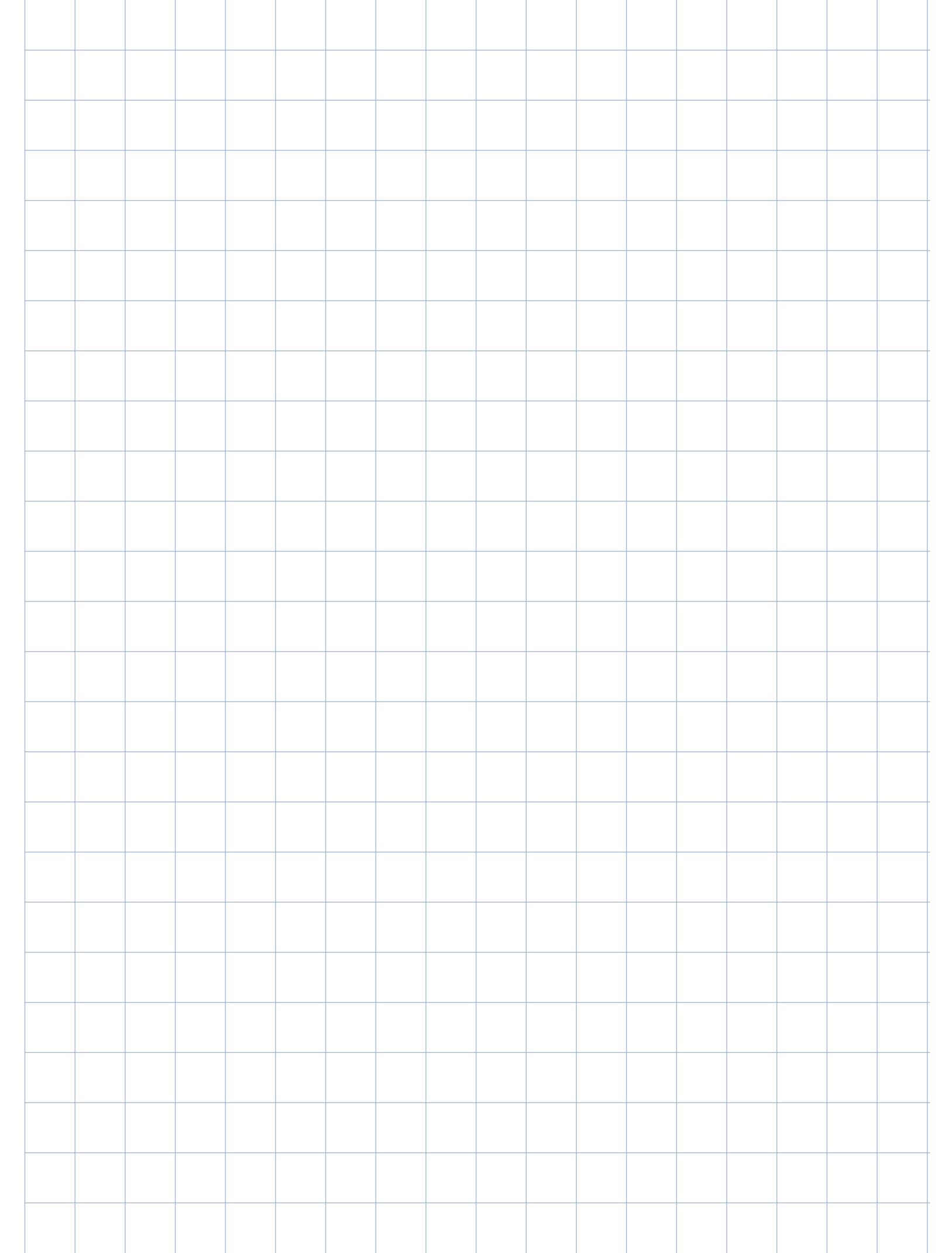
(6) 若 $f \in L^2([0, T], \mathbb{R})$, 则 $\int_0^t f(s) dB(s) \sim N(0, \int_0^t |f(s)|^2 ds)$ [f 是确定函数才 \checkmark]
随机过程 X

(1)(4) f 为随机过程也 \checkmark

Pf. 只证明 $f: [0, T] \rightarrow \mathbb{R}$ 为连续函数的情况

$$\text{定义 } f_n(s) \triangleq \sum_{i=0}^{n-1} f\left(\frac{i}{n}t\right) 1_{(i/n t, (i+1)/n t]}(s) + f(0) 1_{[0, 0]}(s)$$





§7.2 Itô公式

△随机过程中印第安法则

设 U 关于 $\mathcal{F}_t = \{\mathcal{F}_s, 0 \leq s \leq t\}$ 适应, $U_t \in \mathcal{F}_t, V \in L^2_{[0,T]}, X_0 \in \mathcal{F}_0$

$$\sum X_t = X_0 + \int_0^t U_s ds + \int_0^t V_s dB_s \quad \text{— 积分形式 (7.6)}$$

$$\begin{cases} dX_t = U_t dt + V_t dB_t \\ X(0) = X_0 \end{cases} \quad \text{— 微分形式 (7.7)}$$

定理 7.5. 设 X_t 形如 (7.6) 或 (7.7), 设函数 $f: [t, X] \rightarrow [0, T] \times \mathbb{R}$ 二阶连续可导

$$df(x, t) = \frac{\partial f}{\partial t}(t, X_t) dt + \frac{\partial f}{\partial x}(t, X_t) dX_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(t, X_t) (dX_t)^2$$

dt, dB_t 运算:

$$\begin{matrix} \text{乘积} & dt & dB_t \\ dt & 0 & 0 \\ dB_t & 0 & dt \end{matrix}$$

$$\begin{aligned} (dX_t)^2 &= U_t^2 dt dt + U_t V_t dt dB_t + U_t V_t dB_t dt + V_t^2 (dB_t)^2 \\ &= 0 + 0 + 0 + V_t^2 dt \end{aligned}$$

$$\begin{aligned} \text{从而 } df(t, X_t) &= \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x}(U_t dt + V_t dB_t) + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} V_t^2 dt \\ f(0, X(0)) &= f(0, X_0) \end{aligned}$$

例. $X_t = B_t, f(t, x) = \frac{x^2}{2}$

$$f(t, X_t) = \frac{B_t^2}{2}$$

$$d(\frac{1}{2} B_t^2) = 0 + B_t dB_t + \frac{1}{2} dt$$

$$\Rightarrow \frac{1}{2} B_t^2 = \int_0^t B_s dB_s + \frac{1}{2} t$$

Itô 公式证明:

设 $f, \frac{\partial f}{\partial t}, \frac{\partial f}{\partial x}, f$ 所有二阶导数、三阶导数都有界连续

U 关于 t 连续, V 关于 t 连续, $E \int_0^T |U_s|^2 ds < \infty$

• $\{X_t, t \in [0, T]\}$ 关于 t 连续, $X(t) = X(0) + \int_0^t U_s ds + \int_0^t V_s dB_s$

• 取 $0 = t_0 < t_1 < \dots < t_n = t$

$$f(t, X_t) = f(0, X_0) + \sum_{i=0}^{n-1} [f(t_{i+1}, X_{t_{i+1}}) - f(t_i, X_{t_i})] \leq A_i$$

由 Taylor 展开, $A_i = \frac{\partial f}{\partial t}(t_i, X_{t_i})(t_{i+1} - t_i) + \frac{\partial f}{\partial x}(t_i, X_{t_i})(X_{t_{i+1}} - X_{t_i})$

$$+ \frac{1}{2} \left[\frac{\partial^2 f}{\partial t^2}(t_{i+1} - t_i)^2 + 2 \frac{\partial^2 f}{\partial t \partial x}(t_{i+1} - t_i)(X_{t_{i+1}} - X_{t_i}) + \frac{\partial^2 f}{\partial x^2}(X_{t_{i+1}} - X_{t_i})^2 \right] + R_i$$

其中 $R_i = \frac{1}{3!} \left[\frac{\partial^3 f}{\partial t^3}(\theta_i, X_{\theta_i})(t_{i+1} - t_i)^3 + 3 \frac{\partial^3 f}{\partial t^2 \partial x}(\theta_i, X_{\theta_i})(t_{i+1} - t_i)^2 (X_{t_{i+1}} - X_{t_i}) \right.$

$$\left. + 3 \frac{\partial^3 f}{\partial t \partial x^2}(\theta_i, X_{\theta_i})(t_{i+1} - t_i)(X_{t_{i+1}} - X_{t_i})^2 + \frac{\partial^3 f}{\partial x^3}(\theta_i, X_{\theta_i})(X_{t_{i+1}} - X_{t_i})^3 \right]$$

$$\therefore f(t, X_t) = f(0, X_0) + \sum_{i=0}^{n-1} \frac{\partial f}{\partial t}(t_i, X_{t_i})(t_{i+1} - t_i) + \sum_{i=0}^{n-1} \frac{\partial f}{\partial x}(t_i, X_{t_i})(X_{t_{i+1}} - X_{t_i}) I_2^n(t)$$

$$+ \frac{1}{2} \sum_{i=0}^{n-1} \frac{\partial^2 f}{\partial t^2}(t_{i+1} - t_i)^2 + \sum_{i=0}^{n-1} \frac{\partial^2 f}{\partial t \partial x}(t_{i+1} - t_i)(X_{t_{i+1}} - X_{t_i}) I_3^n(t)$$

$$+ \frac{1}{2} \sum_{i=0}^{n-1} \frac{\partial^2 f}{\partial x^2}(X_{t_{i+1}} - X_{t_i})^2 + \sum_{i=0}^{n-1} R_i I_5^n(t)$$

$$\cdot I_1^n(t)(w) = \sum_{i=0}^{n-1} \frac{\partial f}{\partial t}(t_i, X_{t_i})(t_{i+1} - t_i) \xrightarrow[\Delta t \rightarrow 0]{} \int_0^t \frac{\partial f}{\partial t}(s, X_s(w)) ds \triangleq I_1(t)$$

($\frac{\partial f}{\partial t}$ 有界连续, X_t 关于 t 连续, $L^2([0, t], \mathbb{R})$ 之下)

由控制收敛定理 $\lim_{n \rightarrow \infty} E[I_i^n(t) - I_i(t)]^2 = 0$

$$\cdot I_3^n(t)(w) : |I_3^n(t)(w)| \leq \frac{1}{2} \sum_{i=0}^{n-1} \left| \frac{\partial^2 f}{\partial t^2}(t_i, X_{t_i}(w)) \right| (t_{i+1} - t_i)^2 \leq M |\Delta t| \sum_{i=0}^{n-1} (t_{i+1} - t_i)$$

$$|\Delta t| = \sup(t_{i+1} - t_i) = \frac{1}{n} M |\Delta t| \xrightarrow[\Delta t \rightarrow 0]{} 0 \quad dt dt = 0$$

$$\cdot I_4^n(t)(w) : |I_4^n(t)(w)| \leq \sum_{i=0}^{n-1} \left| \frac{\partial^2 f}{\partial x \partial t}(t_i, X_{t_i}(w)) (t_{i+1} - t_i) (X_{t_{i+1}}(w) - X_{t_i}(w)) \right|$$

$$(i.e. \Delta w, t = \sup_{j=i+1, \dots, n} |X_{t_{i+1}}(w) - X_{t_j}(w)|, \text{ 有 } \lim_{n \rightarrow \infty} \Delta w, t = 0)$$

$$\leq M \cdot \Delta w, t \sum_{i=0}^{n-1} (t_{i+1} - t_i) = M \cdot \Delta w, t \cdot t \xrightarrow[\Delta t \rightarrow 0]{} 0 \quad dt dBt = 0$$

$$\cdot X_{t_{i+1}} - X_{t_i} = \int_{t_i}^{t_{i+1}} Us ds + \int_{t_i}^{t_{i+1}} Us dB_s$$

$$\cdot 定义 |\tilde{R}_B^n(w)| \leq \sum_{i=0}^{2^n-1} \left| \frac{\partial^2 f}{\partial x^2}(\theta_i, X_{\theta_i}(w)) \right| \cdot |(B_{t_{i+1}}(w) - B_{t_i}(w))^3|$$

$$(考虑 t_{i+1} = \frac{i+1}{2^n} t)$$

$$\leq \sum_{i=0}^{2^n-1} M \cdot |B_{t_{i+1}}(w) - B_{t_i}(w)|^3$$

$$\leq M \cdot \Delta w, t \sum_{i=0}^{2^n-1} |B_{t_{i+1}}(w) - B_{t_i}(w)|^2 \xrightarrow[\Delta t \rightarrow 0]{} 0$$

$$\cdot I_2^n(t) : I_2^n(t) = \sum_{i=0}^{n-1} \frac{\partial f}{\partial x}(t_i, X_{t_i}) \left(\int_{t_i}^{t_{i+1}} Us ds + \int_{t_i}^{t_{i+1}} Us dB_s \right)$$

$$\xrightarrow[\Delta t \rightarrow 0]{} \int_0^t \frac{\partial f}{\partial x}(s, X_s) Us ds + \int_0^t \frac{\partial f}{\partial x}(s, X_s) Vs dB_s$$

$$\begin{aligned} \cdot I_5^n(t) : I_5^n(t) &= \sum_{i=0}^{n-1} \frac{\partial^2 f}{\partial X^2}(t_i, X_{t_i}) \left[\int_{t_i}^{t_{i+1}} V_s ds + \int_{t_i}^{t_{i+1}} V_s dB_s \right]^2 \\ &= \sum_{i=0}^{n-1} \frac{\partial^2 f}{\partial X^2}(t_i, X_{t_i}) \left[\left(\int_{t_i}^{t_{i+1}} V_s ds \right)^2 + 2 \int_{t_i}^{t_{i+1}} V_s ds \int_{t_i}^{t_{i+1}} V_s dB_s \right. \\ &\quad \left. + \left(\int_{t_i}^{t_{i+1}} V_s dB_s \right)^2 \right] \quad (\text{配对上前面的和}) \end{aligned}$$

$$\sum_{i=0}^{n-1} \frac{\partial^2 f}{\partial X^2}(t_i, X_{t_i}) \left[\int_{t_i}^{t_{i+1}} V_s dB_s \right]^2 \xrightarrow{n \rightarrow \infty} \sum_{i=0}^t \frac{\partial^2 f}{\partial X^2}(s, X_s) V_s^2 ds$$

例2. $Y_t = \int_0^t e^{a(t-s)} dB_s, a > 0, \sigma > 0$, Langerin 方程 (描述粒子在液体中的运动)

$$Y_t = e^{at} \int_0^t e^{-as} dB_s$$

$$\text{则 } ① Y_t \sim N(0, \int_0^t e^{2a(t-s)} \sigma^2 ds)$$

$$② X_t = \int_0^t e^{-as} \sigma dB_s \sim dX_t = e^{-at} \sigma dB_t$$

$$③ f(t, x) = e^{at} x, \text{ 则 } Y_t = f(t, X_t)$$

$$df(t, X_t) = \frac{\partial f}{\partial t}(t, X_t) dt + \frac{\partial f}{\partial X}(t, X_t) dX_t + \frac{1}{2} \frac{\partial^2 f}{\partial X^2}(t, X_t) (dX_t)^2$$

$$dY_t = ae^{at} X_t dt + e^{at} e^{-at} \sigma dB_t + 0$$

$$= aY_t dt + \sigma dB_t$$

$$\therefore Y_t = a \int_0^t Y_s ds + \sigma B_t$$

例3. 证明: $tB_t - \int_0^t B_s ds = \int_0^t s dB_s$

PF. 令 $f(t, x) = tx, X_t = B_t$, 则 $f(t, X_t) = tB_t$

由 Ito 公式: $d(tB_t) = B_t dt + t dB_t + 0$

$$\Rightarrow tB_t = \int_0^t B_s ds + \int_0^t s dB_s$$

$\Delta \{ \int_0^t s dB_s, t \geq 0 \}$ 关于 $f_t = \int_0^t B_s ds, 0 \leq s \leq t$ 为鞅

$\therefore \{ tB_t - \int_0^t B_s ds, t \geq 0 \}$ 关于 $f_t = \int_0^t B_s ds, 0 \leq s \leq t$ 为鞅

例4. 设某资产价格 S_t 满足

$$\begin{cases} dS_t = \sigma S_t dt + r S_t dB_t, t \geq 0, \sigma, r > 0 \\ S_0 = 1 \end{cases}$$

① 由 Ito 公式求解 S_t .

$$\Delta Y_t = \ln S_t \quad f(t, x) = \ln x$$

$$df(t, S_t) = \frac{\partial f}{\partial t}(t, S_t) dt + \frac{\partial f}{\partial X}(t, S_t) dS_t + \frac{1}{2} \frac{\partial^2 f}{\partial X^2}(t, S_t) (dS_t)^2$$

$$\begin{aligned}
 &= 0 + \frac{1}{S_t} dS_t - \frac{1}{2S_t^2} (dS_t)^2 \\
 &= \sigma dt + r dB_t - \frac{1}{2S_t^2} (r^2 S_t^2 dt) \\
 &= (\sigma - \frac{r^2}{2}) dt + r dB_t
 \end{aligned}$$

$$\Rightarrow \ln S_t = \ln S_0 + (\sigma - \frac{r^2}{2}) t + r B_t$$

$$\Rightarrow S_t = S_0 e^{(\sigma - \frac{r^2}{2}) t + r B_t}$$

$$② e^{-\sigma t} S_t = e^{-\frac{1}{2} r^2 t + r B_t} \text{ 关于 } \mathcal{F}_t = \sigma \{ B_s, 0 \leq s \leq t \} \text{ 为鞅}$$

问题: 求 μ, σ , s.t. $e^{-\mu t} S_t$ 为鞅.

P185 Ikeda N / Watanabe S

Stochastic differential equations and diffusion process

$$\begin{cases} dX_t = b(X_t) dt + dB_t & X_t \in \mathbb{R} \\ X(0) = x \end{cases} \quad b: \mathbb{R} \rightarrow \mathbb{R}$$

若 $b(x)$ 为有界, Borel 可测, 则上述方程存在唯一的解

i.e. $\exists X: (\Omega, \mathcal{F}, P) \rightarrow C([0, +\infty), \mathbb{R})$

$$X_t \in \mathcal{F}_t \text{ 且 } X(t) = x + \int_0^t b(X_s) ds + B_t$$

$$\begin{cases} dX_t = 2\sqrt{|X_t|} \wedge 1 dt + dB_t & b(x) = 2\sqrt{|x|} \wedge 1 \\ X(0) = 0 \end{cases}$$

$$\begin{cases} dX_t = 2\sqrt{|X_t|} \wedge 1 dt \\ X(0) = 0 \end{cases} \text{ 有无穷多个解 } X_t = 0, X_t = \begin{cases} t^2, & 0 \leq t \leq 1, \\ t, & t > 1 \end{cases}, X_t = \begin{cases} 0, & 0 \leq t \leq a \\ (t-a)^2, & a < t \leq a+1 \\ t-a, & t > a+1 \end{cases}$$

$$\begin{cases} dX_t = [b_1 1_0 + b_2 1_{0^c}] (X_t) + dB_t & (0 \text{ 为集合}) \\ X(0) = 0 \end{cases}$$