

HOMEWORK_11

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Q5.4

有 $B(t) \sim \mathcal{N}(0, t)$ 。所以

$$P(|B(t)| \leq s) = P(-s \leq B(t) \leq s) = \int_{-s}^s \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx, \quad \forall s \geq 0 \quad (1)$$

所以

$$f_{|B(t)|}(x) = \frac{2}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} 1_{\{x \geq 0\}} \quad (2)$$

另外

$$P(M(t) \leq s) = 1 - P(M(t) > s) = 1 - P(T_s < t) = 1 - 2(1 - \Phi(\frac{s}{\sqrt{t}})) = 2\Phi(\frac{s}{\sqrt{t}}) - 1 \quad (3)$$

$$= \int_{-s}^s \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx = P(|B(t)| \leq s) \quad (4)$$

且根据对称性

$$P(|\min_{0 \leq s \leq t} B(s)| \leq a) = P(|M(t)| \leq a) = P(M(t) \leq a) \quad (\because M(t) \geq B_0 = 0) \quad (5)$$

所以 $|\min_{0 \leq s \leq t} B(s)|$ 和 $M(t)$ 都与 $|B(t)|$ 同分布，二者的分布函数与 $|B(t)|$ 的相同。

最后

$$\delta(t) = M(t) - B(t) = \max_{0 \leq s \leq t} B(s) - B(t) \quad (6)$$

$$= \max_{0 \leq s \leq t} (B(s) - B(t)) \quad (7)$$

所以

$$P(\delta(t) \leq a) = P(\max_{0 \leq s \leq t} (B(s) - B(t)) \leq a) \quad (8)$$

$$= P(\max_{0 \leq s \leq t} B(s - t) \leq a) \quad (9)$$

$$= P(\max_{0 \leq s \leq t} B(-s) \leq a) = P(\max_{0 \leq s \leq t} B(s) \leq a) = P(M(t) \leq a) \quad (10)$$

因此 $\delta(t)$ 与 $M(t)$ 是同分布，它的分布函数也为

$$f_{\delta(t)}(x) = \frac{2}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} 1_{\{x \geq 0\}} \quad (11)$$

Q5.21

证明定理5.2.4a

$$\lim_{\lambda \rightarrow 0} \sum_{k=1}^n B(t_{k-1} + \theta(t_k - t_{k-1}))(B(t_k) - B(t_{k-1})) \stackrel{m.s}{=} \frac{1}{2} B^2(t) + \frac{1}{2} (2\theta - 1)t \quad (12)$$

其中 $0 \leq \theta \leq 1$ 。

证明：

令 $C_k = B(t_{k-1} + \theta(t_k - t_{k-1})) - B(t_{k-1})$, $D_k = B(t_k) - B(t_{k-1} + \theta(t_k - t_{k-1}))$, $1 \leq k \leq n$, 有

$$\sum_{k=1}^n B(t_{k-1} + \theta(t_k - t_{k-1}))(B(t_k) - B(t_{k-1})) \quad (13)$$

$$= \frac{1}{2} \sum_{k=1}^n [B^2(t_k) - B^2(t_{k-1}) + (B(t_{k-1} + \theta(t_k - t_{k-1})) - B(t_{k-1}))^2 \quad (14)$$

$$- (B(t_k) - B(t_{k-1} + \theta(t_k - t_{k-1})))^2] \quad (15)$$

$$= \frac{1}{2} \left[B^2(t) + \sum_{k=1}^n (C_k^2 - D_k^2) \right] \quad (16)$$

而

$$\sum_{k=1}^n (C_k^2 + D_k^2) \quad (17)$$

$$= \sum_{k=1}^n (B(t_{k-1} + \theta(t_k - t_{k-1})) - B(t_{k-1}))^2 + (B(t_k) - B(t_{k-1} + \theta(t_k - t_{k-1})))^2 \quad (18)$$

其中 $\forall 1 \leq k \leq n$,

$$\begin{aligned} (t_{k-1} + \theta(t_k - t_{k-1})) - t_{k-1} &= \theta(t_k - t_{k-1}) < t_k - t_{k-1} < \lambda \\ t_k - (t_{k-1} + \theta(t_k - t_{k-1})) &= (1 - \theta)(t_k - t_{k-1}) < t_k - t_{k-1} < \lambda \end{aligned} \quad (19)$$

因此根据定理5.2.3, 有

$$\lim_{\lambda \rightarrow 0} \sum_{k=1}^n (C_k^2 + D_k^2) \stackrel{m.s}{=} t \Leftrightarrow \lim_{\lambda \rightarrow 0} E \left[\sum_{k=1}^n (C_k^2 + D_k^2) - t \right]^2 = 0 \quad (20)$$

另一方面, $\forall 1 \leq k \leq n$, $C_k \sim B(\theta(t_k - t_{k-1}))$, $D_k \sim B((1 - \theta)(t_k - t_{k-1}))$ 。因此

$$Var D_k = \frac{1 - \theta}{\theta} Var C_k \quad (21)$$

而 $ED_k = EC_k = 0$, 因此

$$ED_k^2 = Var D_k = \frac{1 - \theta}{\theta} Var C_k = \frac{1 - \theta}{\theta} EC_k^2 \quad (22)$$

所以

$$E \sum_{k=1}^n D_k^2 = \frac{1 - \theta}{\theta} E \sum_{k=1}^n C_k^2 \quad (23)$$

及

$$E\left(\sum_{k=1}^n D_k^2\right)^2 = E\left(\sum_{k=1}^n D_k^4 + 2 \sum_{i < j} D_i^2 D_j^2\right) \quad (24)$$

$$= 3(\text{Var} D_k)^2 + 2 \sum_{i < j} E D_i^2 E D_j^2 \quad (25)$$

$$= \left(\frac{1-\theta}{\theta}\right)^2 \cdot [3(\text{Var} C_k)^2 + 2 \sum_{i < j} E C_i^2 E C_j^2] \quad (26)$$

$$= \left(\frac{1-\theta}{\theta}\right)^2 E\left(\sum_{k=1}^n C_k^2\right)^2 \quad (27)$$

所以

$$0 = \lim_{\lambda \rightarrow 0} E \left[\sum_{k=1}^n (C_k^2 + D_k^2) - t \right]^2 \quad (28)$$

$$= \lim_{\lambda \rightarrow 0} E \left[\left(\sum_{k=1}^n C_k^2\right)^2 + \left(\sum_{k=1}^n D_k^2\right)^2 + t^2 + 2\left(\sum_{k=1}^n C_k^2\right)\left(\sum_{k=1}^n D_k^2\right) - 2t\left(\sum_{k=1}^n C_k^2\right) - 2t\left(\sum_{k=1}^n D_k^2\right) \right] \quad (29)$$

$$= \lim_{\lambda \rightarrow 0} E \left[\left[1 + \left(\frac{1-\theta}{\theta}\right)^2\right] \left(\sum_{k=1}^n C_k^2\right)^2 + t^2 + 2\frac{1-\theta}{\theta} \left(\sum_{k=1}^n C_k^2\right)^2 - 2t\frac{1}{\theta} \left(\sum_{k=1}^n C_k^2\right) \right] \quad (30)$$

$$= \lim_{\lambda \rightarrow 0} E \left[\frac{1}{\theta^2} \left(\sum_{k=1}^n C_k^2\right)^2 - 2\frac{t}{\theta} \left(\sum_{k=1}^n C_k^2\right) + t^2 \right] \quad (31)$$

$$= \frac{1}{\theta^2} \lim_{\lambda \rightarrow 0} E \left[\sum_{k=1}^n C_k^2 - \theta t \right]^2 \quad (32)$$

所以

$$\lim_{\lambda \rightarrow 0} \sum_{k=1}^n C_k^2 \stackrel{m.s.}{=} \theta t \quad (33)$$

同理可知

$$\lim_{\lambda \rightarrow 0} \sum_{k=1}^n D_k^2 \stackrel{m.s.}{=} (1-\theta)t \quad (34)$$

而 $E \sum_{k=1}^n (C_k^2 - D_k^2) = \frac{2\theta-1}{\theta} E \sum_{k=1}^n C_k^2$, 进一步地, 我们有

$$\lim_{\lambda \rightarrow 0} \sum_{k=1}^n (C_k^2 - D_k^2) \stackrel{m.s.}{=} (2\theta-1)t \quad (35)$$

因此

$$\lim_{\lambda \rightarrow 0} \sum_{k=1}^n B(t_{k-1} + \theta(t_k - t_{k-1}))(B(t_k) - B(t_{k-1})) \quad (36)$$

$$= \lim_{\lambda \rightarrow 0} \frac{1}{2} \left[B^2(t) + \sum_{k=1}^n (C_k^2 - D_k^2) \right] \quad (37)$$

$$\stackrel{m.s.}{=} \frac{1}{2} B^2(t) + \frac{1}{2} (2\theta-1)t \quad (38)$$

得证。

Q5.22

(2)

记

$$X = \{X(t) = e^{\lambda B(t) - \frac{1}{2}\lambda^2 t}, t \geq 0\} \quad (39)$$

首先, $|EX(t)| \leq e^{\lambda|B(t)|} e^{-\frac{1}{2}\lambda^2 t} < +\infty$ 。其次, $\forall 0 \leq t_0 < t_1 < \cdots < t_n < t_{n+1}$, 有

$$E(X(t_{n+1})|X(t_n), \cdots, X(t_0)) \quad (40)$$

$$= E(X(t_{n+1})|X(t_n)) \quad (41)$$

$$= E(e^{\lambda B(t_{n+1}) - \frac{1}{2}\lambda^2 t_{n+1}} | X(t_n)) \quad (42)$$

$$= E(X(t_n) e^{\lambda(B(t_{n+1}) - B(t_n)) - \frac{1}{2}\lambda^2(t_{n+1} - t_n)} | X(t_n)) \quad (43)$$

$$= X(t_n) e^{-\frac{1}{2}\lambda^2(t_{n+1} - t_n)} E e^{\lambda(B(t_{n+1}) - B(t_n))} \quad (44)$$

而

$$E e^{\lambda(B(t_{n+1}) - B(t_n))} \quad (45)$$

$$= \int_{-\infty}^{\infty} e^{\lambda s} \frac{1}{\sqrt{2\pi(t_{n+1} - t_n)}} e^{-\frac{s^2}{2(t_{n+1} - t_n)}} ds \quad (46)$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi(t_{n+1} - t_n)}} e^{-\frac{(s - \lambda(t_{n+1} - t_n))^2}{2(t_{n+1} - t_n)} + \frac{\lambda^2(t_{n+1} - t_n)}{2}} ds \quad (47)$$

$$= e^{\frac{\lambda^2(t_{n+1} - t_n)}{2}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi(t_{n+1} - t_n)}} e^{-\frac{(s - \lambda(t_{n+1} - t_n))^2}{2(t_{n+1} - t_n)}} ds \quad (48)$$

$$= e^{\frac{1}{2}\lambda^2(t_{n+1} - t_n)} \quad (49)$$

所以

$$E(X(t_{n+1})|X(t_n), \cdots, X(t_0)) = X(t_n) \quad (50)$$

因此 X 是鞅。

(3)

记

$$Y = \{Y(t) = B^2(t) - t, t \geq 0\} \quad (51)$$

首先, $|EY(t)| \leq |B(t)|^2 + |t| < +\infty$ 。其次, $\forall 0 \leq t_0 < t_1 < \cdots < t_n < t_{n+1}$, 有

$$E(Y(t_{n+1})|Y(t_n), Y(t_{n-1}), \cdots, Y(t_0)) \quad (52)$$

$$= E(Y(t_{n+1})|Y(t_n)) \quad (53)$$

$$= E(B^2(t_{n+1}) - t_{n+1} | Y(t_n)) \quad (54)$$

$$= E(B^2(t_n) + 2B(t_n)(B(t_{n+1}) - B(t_n)) + (B(t_{n+1}) - B(t_n))^2 - t_{n+1} | Y(t_n)) \quad (55)$$

$$= Y(t_n) + 2B(t_n)E(B(t_{n+1}) - B(t_n)) + E(B(t_{n+1}) - B(t_n))^2 + t_n - t_{n+1} \quad (56)$$

$$= Y(t_n) + 0 + (t_{n+1} - t_n) + t_n - t_{n+1} \quad (57)$$

$$= Y(t_n) \quad (58)$$

所以 Y 也是鞅。

Q5.5

求出 (S_{t_1}, S_{t_2}) 的联合概率密度函数。设 $t_1 < t_2$,

首先, $\{S_t : t \geq 0\}$ 是一个正态过程, 有

$$\begin{aligned} ES_t &= E \int_0^t B_u du = \int_0^t EB_u du = 0 \\ \text{Cov}(S_{t_1}, S_{t_2}) &= \frac{t_1^2}{2} (t_2 - \frac{t_1}{3}) \\ ES_t^2 &= E \int_0^t \int_0^t B_u B_v dudv = \text{Cov}(S_t, S_t) = \frac{t^3}{3} \end{aligned} \quad (59)$$

因此 (S_{t_1}, S_{t_2}) 的联合概率密度函数就是二维高斯分布的联合概率密度函数, 其中 $\sigma_1 = \sqrt{\frac{t_1^3}{3}} = t_1 \sqrt{\frac{t_1}{3}}$, $\sigma_2 = \sqrt{\frac{t_2^3}{3}} = t_2 \sqrt{\frac{t_2}{3}}$, 因此相关系数 $\rho = \frac{\text{Cov}(S_{t_1}, S_{t_2})}{\sigma_1 \sigma_2} = \frac{t_1(3t_2 - t_1)}{2t_2 \sqrt{t_1 t_2}}$, 则

$$f_{S_{t_1}, S_{t_2}}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} e^{-\frac{1}{2(1-\rho^2)} \left[\frac{(x_1-\mu_1)^2}{\sigma_1^2} - \frac{2\rho(x_1-\mu_1)(x_2-\mu_2)}{\sigma_1\sigma_2} + \frac{(x_2-\mu_2)^2}{\sigma_2^2} \right]} \quad (60)$$

Q5.9

有 $\eta(t) = e^{B(t)}$,

有

$$E[\eta(t+h) - \eta(t) | \eta(t) = x] \quad (61)$$

$$= E[e^{B(t+h)} - e^{B(t)} | e^{B(t)} = x] \quad (62)$$

$$= x E[e^{B(t+h)-B(t)} - 1 | e^{B(t)} = x] \quad (63)$$

$$= x (E[e^{B(t+h)-B(t)}] - 1) \quad (64)$$

$$= x (e^{\frac{h}{2}} - 1) \quad (65)$$

所以

$$\alpha(x) = \lim_{h \downarrow 0} \frac{E[\eta(t+h) - \eta(t) | \eta(t) = x]}{h} \quad (66)$$

$$= \lim_{h \downarrow 0} \frac{x(e^{\frac{h}{2}} - 1)}{h} \quad (67)$$

$$= x \lim_{h \downarrow 0} \frac{e^{\frac{h}{2}} - 1}{h} \quad (68)$$

$$= x \lim_{h \downarrow 0} \frac{e^{\frac{h}{2}} \cdot \frac{1}{2}}{1} = \frac{x}{2} \quad (69)$$

另一方面,

$$E[(\eta(t+h) - \eta(t))^2 | \eta(t) = x] \quad (70)$$

$$= E[(e^{B(t+h)} - e^{B(t)})^2 | e^{B(t)} = x] \quad (71)$$

$$= E[e^{2B(t+h)} - 2e^{B(t+h)+B(t)} + e^{2B(t)} | e^{B(t)} = x] \quad (72)$$

$$= E[(e^{B(t)} e^{C(h)})^2 - 2(e^{B(t)} e^{C(h)}) e^{B(t)} + e^{2B(t)} | e^{B(t)} = x], \quad \text{其中 } C(h) = B(t+h) - B(t) \quad (73)$$

$$= x^2 E(e^{C(h)})^2 - 2x^2 E e^{C(h)} + x^2 \quad (74)$$

而 $C(h) = B(t+h) - B(t) \stackrel{d}{=} B(h)$, 因此

$$\begin{aligned} Ee^{C(h)} &= e^{\frac{h}{2}} \\ E(e^{C(h)})^2 &= e^{2h} \end{aligned} \quad (75)$$

所以

$$E[(\eta(t+h) - \eta(t))^2 | \eta(t) = x] \quad (76)$$

$$= x^2 e^{2h} - 2x^2 e^{\frac{h}{2}} + x^2 \quad (77)$$

$$= x^2 (e^{2h} - 2e^{\frac{h}{2}} + 1) \quad (78)$$

因此

$$\beta(x) = \lim_{h \downarrow 0} \frac{E[(\eta(t+h) - \eta(t))^2 | \eta(t) = x]}{h} \quad (79)$$

$$= x^2 \lim_{h \downarrow 0} \frac{e^{2h} - 2e^{\frac{h}{2}} + 1}{h} \quad (80)$$

$$= x^2 \lim_{h \downarrow 0} \frac{e^{2h} \cdot 2 - 2e^{\frac{h}{2}} \cdot \frac{1}{2}}{1} = x^2 \quad (81)$$

Q5.16

$$T_x = \inf \{t \geq 0 : B_t = x\} \quad (82)$$

则

$$P(T_1 < T_{-1} < T_2) = P(T_1 < T_{-1}, T_{-1} < T_2) \quad (83)$$

$$= P(T_{-1} < T_2 | T_1 < T_{-1}) P(T_1 < T_{-1}) \quad (84)$$

而根据布朗运动的对称性, $P(T_1 < T_{-1}) = P(T_{-1} < T_1) = \frac{1}{2}$ 。

令 $C = \{B_{t+T_1} : t \geq 0\}$, $U_x = \inf \{t \geq 0 : C_t = x\}$ 。则 $C_0 = B_{T_1} = 1$,

$$P(T_{-1} < T_2 | T_1 < T_{-1}) = P(U_{-1} < U_2 | U_1 < U_{-1}) \quad (85)$$

$$= P(U_{-1} < U_2) \quad (86)$$

$$= P(U_{-1} < U_2 | U_0 < U_2) \cdot P(U_0 < U_2) + P(U_{-1} < U_2 | U_0 \geq U_2) P(U_0 \geq U_2) \quad (87)$$

$$= P(U_{-1} < U_2 | U_0 < U_2) \cdot \frac{1}{2} + 0 \quad (88)$$

$$= \frac{1}{2} P(U_{-1} < U_2 | U_0 < U_2) \quad (89)$$

令 $D = \{C_{t+U_0} : t \geq 0\}$, $V_x = \inf \{t \geq 0 : D_t = x\}$, 则 $D_0 = C_{U_0} = 0$ 。

则

$$P(U_{-1} < U_2 | U_0 < U_2) = P(V_{-1} < V_2 | V_0 < V_2) \quad (90)$$

$$= P(V_{-1} < V_2) \quad (91)$$

$$= P(V_1 < V_{-2}) \quad (92)$$

$$= 1 - P(V_{-2} < V_1) \quad (93)$$

$$= 1 - P(V_{-1} < V_2 | V_1 < V_{-1}) \quad (94)$$

$$= 1 - P(T_{-1} < T_2 | T_1 < T_{-1}) \quad (95)$$

所以

$$P(T_{-1} < T_2 | T_1 < T_{-1}) = \frac{1}{2} (1 - P(T_{-1} < T_2 | T_1 < T_{-1})) \implies P(T_{-1} < T_2 | T_1 < T_{-1}) = \frac{1}{3} \quad (96)$$

所以

$$P(T_1 < T_{-1} < T_2) = P(T_{-1} < T_2 | T_1 < T_{-1})P(T_1 < T_{-1}) = \frac{1}{6} \quad (97)$$