

# 数学作业纸

(科目: 随机数学与统计)

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20. (a)

必要性:  $s \leq t$  时,

$$\text{Cov}(X_s, X_t) = \text{Cov}(X_0, X_{t-s}) \text{ 依赖于 } t-s$$

令  $n=1$ , 令  $t_1=0, a=t, u$

$X_0$  与  $X_t$  有相同分布

$$\therefore EX_t = EX_0 = \text{const}$$

充分性: 设  $\text{Cov}(X_s, X_t) = f(t-s), EX_t = EX_0$

$$\begin{pmatrix} X_{t_1+a} \\ X_{t_2+a} \\ \vdots \\ X_{t_n+a} \end{pmatrix} \sim N \left( \begin{pmatrix} EX_0 \\ EX_0 \\ \vdots \\ EX_0 \end{pmatrix}, \begin{pmatrix} f(0) & f(t_1-t_1) & \dots & f(t_n-t_1) \\ f(t_1-t_1) & f(0) & & \vdots \\ \vdots & & \ddots & \vdots \\ f(t_n-t_1) & \dots & \dots & f(0) \end{pmatrix} \right)$$

$$\text{即 } \text{Cov}(X_{t_1+a}, X_{t_2+a}) = \text{Cov} f(t_2-t_1) \because \sum_{i=1}^n B_i \sim N(0, \frac{n(n+1)(2n+1)}{6})$$

$$= \text{Cov}(X_{t_1}, X_{t_2}) \text{ 且 } EX_t = EX_0$$

21.

$$(a) B_t \sim N(0, t)$$

由高斯分布的线性组合仍是高斯分布

可得  $B_1 + B_2 + \dots + B_n$  服从高斯分布.

$$E(\sum_{i=1}^n B_i) = \sum_{i=1}^n EB_i = 0$$

$$D(\sum_{i=1}^n B_i) = \sum_{i=1}^n DB_i + 2 \sum_{0 \leq i < j \leq n} \text{Cov}(B_i, B_j)$$

$$= \sum_{i=1}^n i + 2 \sum_{0 \leq i < j \leq n} i$$

$$= \sum_{i=1}^n i + 2 \sum_{j=1}^n \sum_{i=1}^{j-1} i$$

$$= \sum_{j=1}^n j + \sum_{j=1}^n j(j-1)$$

$$= \sum_{j=1}^n j^2 = \frac{n(n+1)(2n+1)}{6}$$

(b) 由于  $Y_t = B_{t+1} - B_t$

$\therefore Y_t$  的任意有限维分布均为高斯分布

由高斯分布的线性组合, 即使任意有限维分布均为高斯分布

$\therefore Y_t: t \geq 0$  是高斯过程

由 20. (a) 的结论, 只需证:  $\begin{cases} \text{Cov}(X_s, X_t) \text{ 依赖于 } t-s \\ EX_t \text{ 为常数} \end{cases}$

$$(1) E Y_t = EB_{t+1} - EB_t = 0 = \text{const}$$

$$(2) \forall s \leq t, \text{Cov}(X_s, X_t) = \text{Cov}(B_{s+1} - B_s, B_{t+1} - B_t)$$

$$= \text{Cov}(B_{s+1}, B_{t+1}) + \text{Cov}(B_s, B_t) - \text{Cov}(B_{s+1}, B_t) - \text{Cov}(B_s, B_{t+1})$$

$$= s+1 + s - \min(s+1, t) - s = s+1 - \min(s+1, t) = \begin{cases} 1 & t-s \geq 1 \\ t-s & t-s < 1 \end{cases} = \max\{0, t-s\}$$

$\therefore Y_n, a, \forall t_1, t_2, \dots, t_n, X_{t_1}, X_{t_2}, \dots, X_{t_n} \sim X_{t_1+a}, X_{t_2+a}, \dots, X_{t_n+a}$  分布相同

$\therefore$  Gauss 过程为平稳过程

(b):  $B_t: t \geq 0$  是 Brown 运动

$\therefore B_t: t \geq 0$  是高斯过程

$$U_t = e^{-\frac{\alpha t}{2}} B_{e^{\alpha t}}$$

$U_t$  的任意有限维分布都是高斯分布

$\therefore U_t$  是高斯过程

$$E U_t = e^{-\frac{\alpha t}{2}} E B_{e^{\alpha t}} = 0 = \text{const}$$

$$\text{Cov}(U_t, U_s) = \text{Cov}(e^{-\frac{\alpha t}{2}} B_{e^{\alpha t}}, e^{-\frac{\alpha s}{2}} B_{e^{\alpha s}})$$

$$= e^{-\frac{\alpha(t+s)}{2}} \text{Cov}(B_{e^{\alpha t}}, B_{e^{\alpha s}}) = e^{-\frac{\alpha(t+s)}{2}} (e^{\alpha t} \wedge e^{\alpha s})$$

$$= e^{-\frac{\alpha}{2}(t+s)} = e^{-\frac{\alpha}{2}|t-s|} \text{ 只与 } t-s \text{ 有关}$$

由 (1)  $\{U_t: t \geq 0\}$  是平稳高斯过程



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$\therefore \text{Cov}(X_s, X_t)$  只跟  $t-s$

由 (1)  $|Y_t: t \geq 0|$  为平稳过程

22. (1)

平移不变性:  $|B_{t+a} - B_a, t \geq 0|$   $a \geq 0$  为常数为 Brown 运动

证:  $\therefore |B_t: t \geq 0|$  是 Brown 运动

$\therefore |B_t, t \geq 0|$  是 Gauss 过程

令  $Y_t = B_{t+a} - B_a$

$\therefore Y_t$  与任意有限维分布为 Gauss 分布 & 线性变换, 即也是 Gauss 分布

$\therefore |Y_t: t \geq 0|$  为 Gauss 过程

$\therefore Y_0 = B_{0+a} - B_a = 0$

$EY_t = E(B_{t+a} - B_a) = EB_{t+a} - EB_a = 0$

$E(Y_t | Y_s) = E((B_{t+a} - B_a)(B_{s+a} - B_a))$

$= EB_{t+a}B_{s+a} - EB_{t+a}B_a - EB_{s+a}B_a + EB_aB_a$

$= (t+a) \wedge (s+a) - a - a + a$

$= (t+a) \wedge (s+a) - a$

$= t \wedge s$

$\therefore |B_{t+a} - B_a, t \geq 0|$   $a \geq 0$  为常数为 Brown 运动

(2)

尺度不变性:  $|\frac{B_{ct}}{\sqrt{c}}: t \geq 0|$   $c > 0$  为常数为 Brown 运动

证: 令  $Y_t = \frac{B_{ct}}{\sqrt{c}}$

$\therefore B_t, t \geq 0$  为 Brown 运动

$\therefore |B_t, t \geq 0|$  为 Gauss 过程

$\therefore Y_t \sim \Gamma$  任意有限维分布为 Gauss 分布的线性变换, 故也是 Gauss 分布

$\therefore |Y_t: t \geq 0|$  为 Gauss 过程

$\therefore Y_0 = \frac{B_0}{\sqrt{c}} = 0$

$EY_t = \frac{EB_{ct}}{\sqrt{c}} = 0$

$E(Y_t | Y_s) = \frac{1}{c} E(B_{ct} | B_{cs})$

$= \frac{1}{c} (t \wedge cs)$

$= t \wedge s$

$\therefore |\frac{B_{ct}}{\sqrt{c}}, t \geq 0|$   $c > 0$  为常数

仍为 Brown 运动

23.

$B_t \sim N(0, t)$

$\begin{pmatrix} B_s \\ B_t \end{pmatrix} \sim N\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} s & s \\ s & t \end{pmatrix}\right)$

$E(B_s | B_t = x) = \mu_s + \rho \frac{\sigma_s}{\sigma_t} (x - \mu_t)$

$= \rho \frac{\sigma_s}{\sigma_t} x$

$= \frac{\text{Cov}(B_s, B_t)}{\sqrt{\text{Var}(B_s)} \sqrt{\text{Var}(B_t)}} \frac{\sigma_s}{\sigma_t} x$

$= \frac{s}{\sqrt{s} \sqrt{t}} \cdot \frac{\sqrt{s}}{\sqrt{t}} x$

$= \frac{s x}{t}$



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4. 令  $Y_i = (X_i - \mu)^2 \quad i=1, 2, \dots$

则  $Y_i$  独立同分布且  $E Y_i = E(X_i - \mu)^2$

$$= EX_i^2 - 2E\mu X_i + E\mu^2$$

$$= DX_i + (EX_i)^2 - 2\mu EX_i + \mu^2$$

$$= \sigma^2 + \mu^2 - 2\mu^2 + \mu^2$$

$$= \sigma^2$$

由 Khintchine 大数定律

$$\frac{1}{n} \sum_{k=1}^n (X_k - \mu)^2 \xrightarrow{P} \sigma^2$$

5.  $\therefore \lim_{|i-j| \rightarrow \infty} r(X_i, X_j) = 0$

$\therefore \forall \varepsilon > 0, \exists N > 0, \text{ s.t. } \forall |i-j| > N,$

$$r(X_i, X_j) < \varepsilon$$

$$\therefore \text{Cov}(X_i, X_j) = r(X_i, X_j) \sqrt{DX_i} \sqrt{DX_j}$$

$$DX_n < C (n \geq 1)$$

$$\therefore \text{Cov}(X_i, X_j) < C\varepsilon_0$$

$$\therefore D\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n \sum_{j=1}^n \text{Cov}(X_i, X_j)$$

$$= \sum_{\substack{i,j=1 \\ |i-j| \leq N}} \text{Cov}(X_i, X_j) + \sum_{\substack{i,j=1 \\ |i-j| > N}} \text{Cov}(X_i, X_j) + \sum_{i=1}^n DX_i$$

$$\leq 2N \cdot n \cdot \max_{0 \leq |i-j| \leq N} \text{Cov}(X_i, X_j) + n^2 C\varepsilon_0 + nC$$

由  $X_i, X_j$  不相关  $|\text{Cov}(X_i, X_j)| = |EX_i X_j - EX_i EX_j| \leq |EX_i X_j| + |EX_i EX_j|$

$$\leq \sqrt{EX_i^2 EX_j^2} + \sqrt{EX_i^2 EX_j^2} \leq 2\sqrt{DX_i DX_j} \leq 2C$$

$$\therefore \forall \varepsilon > 0, \exists N =$$

$$\text{s.t. } P\left(\left|\frac{1}{n} \sum_{i=1}^n X_i - E\left(\frac{1}{n} \sum_{i=1}^n X_i\right)\right| \geq \varepsilon\right)$$

$$\leq \frac{D\left(\sum_{i=1}^n X_i\right)}{n^2 \varepsilon^2}$$

$$\leq \frac{2N \cdot n \cdot \max_{0 \leq |i-j| \leq N} \text{Cov}(X_i, X_j) + n^2 C\varepsilon_0 + nC}{n^2 \varepsilon^2}$$

$$= \frac{C + 2N \cdot \max_{0 \leq |i-j| \leq N} \text{Cov}(X_i, X_j)}{n \varepsilon^2} +$$

$$\therefore \text{s.t. } \leq 2N \cdot n \cdot 2C + n^2 C\varepsilon_0 + nC$$

$$= 4NnC + n^2 C\varepsilon_0 + nC$$

$$\therefore \forall \varepsilon > 0, \exists N = \left\lceil \frac{2C(4N+1)}{\varepsilon} \right\rceil + 1 > 0$$

$$\text{s.t. } P\left(\left|\frac{1}{n} \sum_{i=1}^n X_i - E\left(\frac{1}{n} \sum_{i=1}^n X_i\right)\right| \geq \varepsilon\right)$$

$$\leq \frac{D\left(\sum_{i=1}^n X_i\right)}{n^2 \varepsilon^2}$$

$$\leq \frac{4NnC + n^2 C\varepsilon_0 + nC}{n^2 \varepsilon^2}$$

$$= \frac{C+nC}{n\varepsilon^2} + \frac{C\varepsilon_0}{\varepsilon^2} \xrightarrow{\text{取 } \varepsilon_0 = \frac{\varepsilon}{2C}} \frac{C+nC}{n\varepsilon^2} + \frac{1}{2}\varepsilon$$

$$< \varepsilon$$



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9. 记  $Y_k = \cos kX$   
 $i \neq j$  且  $i, j \in \mathbb{N}^+$   
 $\text{Cov}(Y_i, Y_j)$

$$= \text{Cov}(\cos iX, \cos jX)$$

$$= E(\cos iX \cos jX) - E \cos iX E \cos jX$$

$$E \cos iX = \int_{-\pi}^{\pi} \cos iX \cdot \frac{1}{2\pi} dX = 0$$

$$E \cos jX = \int_{-\pi}^{\pi} \cos jX \cdot \frac{1}{2\pi} dX = 0$$

$$E(\cos iX \cos jX)$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} \cos iX \cos jX dX$$

$$= \frac{1}{2\pi} \times \frac{1}{2} \int_{-\pi}^{\pi} (\cos(i+j)X + \cos(i-j)X) dX$$

$$= 0 \quad (i \neq j)$$

$$\therefore \text{Cov}(Y_i, Y_j) = 0 \quad \forall i \neq j$$

$$DY_k = E Y_k^2 - (E Y_k)^2 = E Y_k^2$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} \cos^2 kX dX$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{\cos 2kX + 1}{2} dX$$

$$= \frac{1}{2}$$

$\therefore \{Y_n\}$  满足两两不相关且方差一致有界  
 由 Chebyshev 大数定律:

$$\frac{1}{n} S_n = \frac{1}{n} \sum_{i=1}^n Y_i - \frac{1}{n} \sum_{i=1}^n E Y_i \xrightarrow{P} 0$$

12.

$$P(|M_n - a| \geq \varepsilon)$$

$$= P(|\min\{X_1, X_2, \dots, X_n\} - a| \geq \varepsilon)$$

$$= P(\min\{X_1, X_2, \dots, X_n\} - a \geq \varepsilon)$$

$$= P(\min\{X_1, X_2, \dots, X_n\} \geq a + \varepsilon)$$

$$= P(X_1 \geq a + \varepsilon, X_2 \geq a + \varepsilon, \dots, X_n \geq a + \varepsilon)$$

$$= \prod_{i=1}^n P(X_i \geq a + \varepsilon)$$

$$= P(X_1 \geq a + \varepsilon)^n$$

$$P(X_1 < a + \varepsilon) = \int_a^{a+\varepsilon} e^{-(x-a)} dx$$

$$= 1 - e^{-\varepsilon}$$

$$\therefore P(X_1 \geq a + \varepsilon) = e^{-\varepsilon}$$

$$\therefore P(|M_n - a| \geq \varepsilon) = e^{-n\varepsilon} = \frac{1}{e^{n\varepsilon}}$$

$$\forall 0 < \varepsilon < 1, \exists N = \left\lceil \frac{1}{\varepsilon} \ln \frac{1}{\varepsilon} \right\rceil + 1$$

$$\text{s.t. } \forall n > N.$$

$$P(|M_n - a| \geq \varepsilon) < \varepsilon$$

$$\therefore \lim_{n \rightarrow \infty} P(|M_n - a| \geq \varepsilon) = 0$$

$$\therefore M_n \xrightarrow{P} a.$$



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$$13. \ln T_n = \frac{\sum_{k=1}^n \ln X_k}{n}$$

$$\text{令 } V_i = \ln X_i \quad i=1, 2, 3, \dots$$

$$\because X_i \text{ 独立 } U[0, 1]$$

$\therefore V_i$  独立同分布

$$\text{且 } EV_i = \int_0^1 \ln x \, dx \quad (\text{瑕积分收敛}).$$

$$= x \ln x \Big|_0^1 - \int_0^1 x d(\ln x)$$

$$= - \int_0^1 dx = -1$$

$\therefore$  由 Khintchine 大数定律

$$\ln T_n \xrightarrow{P} -1$$

$\because e^x$  是连续函数

$$\therefore e^{\ln T_n} \xrightarrow{P} e^{-1}$$

$$\therefore T_n \xrightarrow{P} e^{-1}$$

$$\therefore c = \frac{1}{e}$$



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前面少写了4数, 补在这:

$$B_t \sim N(0, t)$$

$$f_{B_t}(x) = \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}}$$

$$EY_t = \int_{-\infty}^{+\infty} e^x \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx$$

$$= \int_{-\infty}^{+\infty} \frac{e^{x - \frac{x^2}{2t}}}{\sqrt{2\pi t}} dx$$

$$= \frac{e^{\frac{t}{2}}}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2t}(x-t)^2} dx$$

$$= \frac{e^{\frac{t}{2}}}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2t}x^2} dx$$

$$= \frac{e^{\frac{t}{2}}}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{x^2}{2t}} d\left(\frac{x}{\sqrt{t}}\right)$$

$$= \frac{e^{\frac{t}{2}}}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2}x^2} dx$$

$$= \frac{e^{\frac{t}{2}}}{\sqrt{2\pi t}} \cdot \sqrt{2\pi}$$

$$= e^{\frac{t}{2}}$$

$$EY_t^2 = \int_{-\infty}^{+\infty} e^{2x} \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi t}} e^{-\frac{1}{2t}(x^2 - 4xt)} dx$$

$$= \frac{1}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2t}(x^2 - 4xt + 4t^2) + 2t} dx$$

$$= \frac{e^{2t}}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{x^2}{2t}} dx = \frac{e^{2t}}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{x^2}{2t}} d\left(\frac{x}{\sqrt{t}}\right)$$

$$= \frac{e^{2t}}{\sqrt{2\pi t}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2}x^2} dx = \frac{e^{2t}}{\sqrt{2\pi t}} \cdot \sqrt{2\pi} = e^{2t}$$

$$\therefore DY_t = EY_t^2 - (EY_t)^2$$

$$= e^{2t} - e^{\frac{1}{2} \cdot 2}$$

$$= e^{2t} - e^t$$

