

17. $X \sim N(\mu, \Sigma)$

$\therefore \Sigma$ 对称

$\therefore \exists$ 正交 Q , 对角 Λ , 使得 $\Sigma = Q\Lambda Q^T$
令 $A = Q^T$

$\therefore Y \sim N(0, Q^T \Sigma Q)$

而 $Q^T \Sigma Q = \Lambda$, 故 $Y \sim N(0, \Lambda)$

对每个 Y_i , $Y_i \sim N(0, \Lambda_{ii})$

18 (a) $A = \begin{pmatrix} 3 & 2 \\ 2 & 3 \end{pmatrix}$

$Y = AX$, 故 $Y \sim (AM, A\Sigma A^T)$

$\therefore \begin{pmatrix} \frac{11}{5} & \frac{12}{5} \\ \frac{12}{5} & \frac{11}{5} \end{pmatrix}$

(b) $Y \sim \left(\begin{pmatrix} 7 \\ 8 \end{pmatrix}, \begin{pmatrix} \frac{11}{5} & \frac{12}{5} \\ \frac{12}{5} & \frac{11}{5} \end{pmatrix} \right)$

而 $(Y_1, Y_2) \sim N(\mu_1, \mu_2, \rho, \sigma_1^2, \sigma_2^2)$

其中 $\sigma_1 = \sqrt{\frac{11}{5}}$ $\sigma_2 = \sqrt{\frac{11}{5}}$ $\rho = \frac{\text{Cov}(Y_1, Y_2)}{\sigma_1 \sigma_2}$

$\therefore (Y_1, Y_2) \sim N(\mu_1, \mu_2, \rho, \sigma_1^2, \sigma_2^2)$

$Y_1 + Y_2 = (1, 1) \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix} \therefore E(Y_1 + Y_2) = 15$

而 $(Y_2 | Y_1) \sim N(\mu_2 + \rho \frac{\sigma_2}{\sigma_1} (Y_1 - \mu_1), (1 - \rho^2) \sigma_2^2)$

$\therefore E(Y_2 | Y_1) = 8 + \frac{12}{11} (Y_1 - 7)$

(c) $\rho = \frac{4}{5}$ $\sigma_1 = 1$ $\sigma_2 = 1$

则 $X_2 | X_1 \sim N(2 + \frac{4}{5}(X_1 - 1), \frac{9}{25})$

$E(X_2 | X_1) = \frac{4}{5} X_1 + \frac{6}{5}$

$D(X_2 | X_1) = \frac{9}{25}$

(d) $X_2 - E(X_2 | X_1) = X_2 - \frac{4}{5} X_1 - \frac{6}{5}$ 为高斯分布

而 $\text{Cov}(X_2 - E(X_2 | X_1), X_1) = \text{Cov}(X_2, X_1) - \frac{4}{5} \text{Cov}(X_1, X_1) = 0$

\therefore 相互独立

20. (a) 必要性:

\therefore 相同联合分布 则 $U_t = U_0$ 恒成立

而 $\text{Cov}(X_{t_i}, X_{t_j}) = \text{Cov}(X_{t_i+a}, X_{t_j+a})$

$\therefore \text{Cov}(0, X_{t-s}) = \text{Cov}(X_{t-s}, X_t)$

充分性易证

$$(b) \quad U_t = e^{-\frac{at}{2}} B e^{at}$$

$$E U_t = e^{-\frac{at}{2}} E B e^{at}$$

若 $t \geq s$:

$$\begin{aligned} \text{Cov}(U_s, U_t) &= \text{Cov}(e^{-\frac{as}{2}} B e^{as}, e^{-\frac{at}{2}} B e^{at}) \\ &= e^{-\frac{a(t-s)}{2}} \text{只与 } t-s \text{ 有关} \end{aligned}$$

\therefore 由前得, 为平稳的 Gauss 过程

21. (a)

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

$\therefore \sum_{i=1}^n B_i$ 服从 Gauss 分布

$$\therefore E\left(\sum_{i=1}^n B_i\right) = 0$$

$$\begin{aligned} D\left(\sum_{i=1}^n B_i\right) &= \sum_{i=1}^n D B_i + 2 \sum_{0 \leq i < j \leq n} \text{Cov}(B_i, B_j) \\ &= \sum_{i=1}^n i^2 = \frac{n(n+1)(2n+1)}{6} \end{aligned}$$

$$\therefore N\left(0, \frac{n(n+1)(2n+1)}{6}\right)$$

(b) 设 $t \geq s$

$$s+1 \leq t$$

$$\text{则 } \text{Cov}(Y_s, Y_t) = 0$$

$$s = t < s+1$$

$$\text{则 } \text{Cov}(Y_s, Y_t) = s - t + 1$$

故 $\text{Cov}(Y_s, Y_t)$ 只依赖于 $t-s$

$$\text{而 } E Y_t = 0$$

\therefore 为平稳过程

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

$s < t$ 时

$$\begin{pmatrix} B_s \\ B_t \end{pmatrix} \sim N\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} ss & st \\ st & tt \end{pmatrix}\right)$$

$$\therefore E(B_s | B_t = x) = \frac{xs}{t}$$

$$12. \quad f(x) = \int_{-\infty}^x f(x) dx = 1 - e^{-(x-a)}$$

$$P(M_1 - a \geq \varepsilon) = 1 - F(a + \varepsilon) = e^{-\varepsilon}$$

$$P(|Y_n - a| \geq \varepsilon) = e^{-n\varepsilon}$$

$$\lim_{n \rightarrow \infty} P(|Y_n - a| \geq \varepsilon) = \lim_{n \rightarrow \infty} e^{-n\varepsilon} = 0$$

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

补充题

$$1. \begin{pmatrix} Y \\ Z \end{pmatrix} \sim N \left(\begin{pmatrix} A\mu + B \\ C\mu + D \end{pmatrix}, \begin{pmatrix} A \Sigma A^T + \Sigma_{YZ} \\ \Sigma_{ZY} & C \Sigma C^T \end{pmatrix} \right)$$

$$\Sigma_{YZ} = \text{cov}(Y, Z)$$

$$= \begin{pmatrix} a_1^T \Sigma c_1 & \dots & a_1^T \Sigma c_n \\ \vdots & & \vdots \\ a_n^T \Sigma c_1 & \dots & a_n^T \Sigma c_n \end{pmatrix} = A \Sigma C^T$$

$$\text{同理 } \Sigma_{ZY} = C \Sigma A^T = (A \Sigma C^T)^T$$

$$Y, Z \text{ 独立} \Leftrightarrow \begin{cases} A \Sigma C^T = 0 \\ (A \Sigma C^T)^T = 0 \end{cases} \Rightarrow A \Sigma C^T = 0$$

2. (1) $X = 3B_1 + 2B_2 - B_3$ 仍为 Gauss 分布

$$DX = 9DB_1 + 4DB_2 + DB_3 + 12\text{cov}(B_1, B_2) - 6\text{cov}(B_1, B_3) - 4\text{cov}(B_2, B_3)$$

$$= 18$$

$$\therefore X \sim N(0, 18) \quad \varphi_X(\theta) = e^{-\theta^2/2}$$

$$(2) \text{cov}(X, Y) = 4C + J$$

$$\begin{pmatrix} X \\ Y \end{pmatrix} \sim \left[N \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} DX & \text{cov}(X, Y) \\ \text{cov}(Y, X) & DY \end{pmatrix} \right]$$

$$X, Y \text{ 独立} \Leftrightarrow \text{cov}(X, Y) = 0$$

$$\text{即 } C = -\frac{J}{4}$$

$$(3) E(X|B_2=x) = 3E(B_1|B_2=x) + 2E(B_2|B_2=x) \\ - E(B_3|B_2=x)$$

$$\text{则} E(X|B_2=1) = \frac{5}{2}$$

三.

$$\text{由 } (X, Y) \sim N(0, 0, \rho, 1, 1)$$

$$\text{则} X \sim N(0, 1)$$

$$Y \sim N(0, 1)$$

则 $X+Y, X-Y$ 也是 Gauss 的

$$\text{而 } \text{Cov}(X+Y, X-Y) = 0$$

$\therefore X+Y$ 和 $(X-Y)^{2022}$ 独立

$$\text{故 } E((X+Y)(X-Y)^{2022}) = E(X-Y)^{2022} E(X+Y) = 0$$

$$D(X+Y) = E(X+Y)^2 = 2 + 2\rho$$

$$D(X-Y) = 2 - 2\rho$$

$$\therefore (X+Y) \sim (0, 2+2\rho)$$

$$(X-Y) \sim (0, 2-2\rho)$$

$$\therefore P(X^2 > Y^2) = P(X+Y > 0) P(X-Y) > 0 + P(X+Y < 0) P(X-Y < 0) = \frac{1}{2}$$