

BTVN (tuần 1 + 2)

Bài 1.1

$$a, E(Y - c)^2 = E(Y^2) - 2cY + c^2 \\ = E(Y^2) + (c - Y)^2 - Y^2 \geq E(Y^2) - Y^2$$

$$\text{Do đó } \mu = \arg \min_{c \in \mathbb{R}} E(Y - c)^2$$

$$b, E((Y - f(X))^2 | X) = E(Y^2 | X) - 2f(X) \cdot E(Y | X) + f(X)^2 \\ = E(Y^2 | X) + (f(X) - E(Y | X))^2 - E(Y | X)^2 \\ \geq E(Y^2 | X) - E(Y | X)^2$$

$$\text{Do đó } E(Y | X) = \arg \min_f E((Y - f(X))^2 | X)$$

$$c, E(Y - f(X))^2 = E(E((Y - f(X))^2 | X)) \\ \geq E(E(Y^2 | X) - E(Y | X)^2)$$

Đấu "=" xảy ra $\Leftrightarrow f(X) = E(Y | X)$ theo như ý (b).

Bài 1.2.

a, Tương tự 1.1b, ta có

$$E(X_{n+1} | X_1, \dots, X_n) = \arg \min E((X_{n+1} - f(X_1, \dots, X_n))^2 | X_1, \dots, X_n)$$

$$b, E(X_{n+1} - f(X_1, \dots, X_n))^2 = E_{X_1, \dots, X_n} \left[E((X_{n+1} - f(X_1, \dots, X_n))^2 | X_1, \dots, X_n) \right]$$

$$\text{đạt min khi } f(X_1, \dots, X_n) \\ = E(X_{n+1} | X_1, \dots, X_n)$$

$$c, \hat{X}_{n+1} = E_{X_1, \dots, X_n} (E(X_{n+1} | X_1, \dots, X_n)) = \mu$$

d, Ta tìm $c = (c_1, c_2, \dots, c_n)^T \in \mathbb{R}^n$ sao cho $\sum_{k=1}^n c_k X_k$ là ước lg tuyến tính o chích tốt nhất theo nghĩa phương sai cực tiểu.

Đặt $\underline{x} = (x_1, x_2, \dots, x_n)^T \in \mathbb{R}^n$. Khi đó $\sum_{k=1}^n c_k x_k = \underline{c}^T \underline{x}$.

$$\begin{aligned} \text{Var}(\underline{c}^T \underline{x}) &= E[(\underline{c}^T \underline{x} - E(\underline{c}^T \underline{x}))^2] = E[(\underline{c}^T (\underline{x} - E\underline{x}))^2] \\ &= E[\underline{c}^T (\underline{x} - E\underline{x}) \cdot (\underline{x} - E\underline{x})^T \underline{c}] \\ &= \underline{c}^T E[(\underline{x} - E\underline{x}) \cdot (\underline{x} - E\underline{x})^T] \underline{c} \\ &= \underline{c}^T \Sigma \underline{c} \end{aligned}$$

$$\begin{aligned} \text{Một khác: } \Sigma &= E[(\underline{x} - E\underline{x})(\underline{x} - E\underline{x})^T] \\ &= \text{diag}\{ \text{Var}(X_1), \text{Var}(X_2), \dots, \text{Var}(X_n) \} \\ &= \sigma^2 I_{n \times n} \quad (\text{do } \{X_i\} \text{ là iid}) \end{aligned}$$

Hơn nữa $\underline{c}^T \underline{x}$ là véc-tơ lg 0 chệch nên.

$$E[\underline{c}^T \underline{x}] = 0 \Rightarrow \underline{c}^T E[\underline{x}] = 0$$

$$\Rightarrow \underline{c}^T \underline{1} = 0 \quad (\underline{1} \text{ là vectơ cột gồm toàn } 1)$$

Có tìm \underline{c} để cực tiểu hoá hàm lagrange $L(\underline{c}; \lambda)$ xđ bđ.

$$\begin{aligned} L(\underline{c}; \lambda) &= \text{Var}(\underline{c}^T \underline{x}) + \lambda (\underline{c}^T \underline{1} - 1) \\ &= \sigma^2 \|\underline{c}\|_2^2 + \lambda (\underline{c}^T \underline{1} - 1) \end{aligned}$$

$$\frac{\partial L}{\partial \underline{c}}(\underline{c}; \lambda) = 2\sigma^2 \underline{c} + \lambda \underline{1} = \underline{0} \quad (\underline{0} \text{ là véc-tơ không})$$

$$\Rightarrow \underline{c} = \frac{-\lambda}{2\sigma^2} \underline{1}$$

$$\text{tại } \omega: \underline{c}^T \underline{1} = 1. \text{ Do đó } \frac{-\lambda}{2\sigma^2} \underline{1}^T \underline{1} = 1 \Rightarrow \lambda = \frac{-2\sigma^2}{n}$$

$$\text{Như vậy, } \underline{c} = \frac{-\lambda}{2\sigma^2} \underline{1} = \frac{1}{n} \underline{1} = \left(\frac{1}{n}, \dots, \frac{1}{n}\right)^T$$

hay $\underline{c}^T \underline{x} = \frac{1}{n} \sum_{k=1}^n x_k$ là véc-tơ lg π 0 chệch tốt nhất.



e, Suy ra trực tiếp từ b, và d)

g, Trước hết, ta tìm hàm g cực tiểu hóa

$$\begin{aligned} & E[(S_{n+1} - g(S_1, \dots, S_n))^2 | S_1, \dots, S_n] \\ &= E[S_{n+1}^2 | S_1, \dots, S_n] - 2g(S_1, \dots, S_n)E[S_{n+1} | S_1, \dots, S_n] \\ &\quad + g(S_1, \dots, S_n)^2 \\ &= \text{Var}(S_{n+1} | S_1, \dots, S_n) + (g(S_1, \dots, S_n) - E[S_{n+1} | S_1, \dots, S_n])^2 \\ &\geq \text{Var}(S_{n+1} | S_1, \dots, S_n) \end{aligned}$$

Như vậy $E[S_{n+1} | S_1, \dots, S_n] =$

$$= \underset{g}{\operatorname{argmin}} E[(S_{n+1} - g(S_1, \dots, S_n))^2 | S_1, \dots, S_n]$$

$$\text{Chọn } \hat{S}_{n+1} = E_{S_1, \dots, S_n}(E(S_{n+1} | S_1, \dots, S_n))$$

$$= E_{X_1, \dots, X_n}(E(X_1 + \dots + X_{n+1} | X_1, \dots, X_n))$$

$$= \mu_{n+1}$$

Bài 13.

Xét chuỗi Hgwan dừng chặt $\{X_t\}_{t \geq 0}$ là $\forall t \in \mathbb{R}$ và $\forall k > 0$.

$$\text{Tạo: } P(X_{t+k} \leq x_k) = P(X_{t+k} \leq x_k)$$

$\Rightarrow X_{t_k}$ và X_{t_k+h} có cùng pp $\forall t \in \mathbb{R}$ $\forall k > 0$

giả sử chúng có cùng pp vs biến X .

$$\text{Đặt } \mu_x = E(X), \text{ khi đó } E(X_t) = E(X) = \mu_x$$

không phụ thuộc vào t .

Do đó $\{X_t\}$ là chuỗi TG dừng yếu.

Bài 1.5.

a, ACVF:

$$\begin{aligned}\gamma_x(t+h; t) &= \text{Cov}(X_{t+h}; X_t) \\ &= \text{Cov}(Z_{t+h} + \theta Z_{t+h-2}; Z_t + \theta Z_{t-2}) \\ &= \text{Cov}(Z_{t+h}; Z_t) + \theta [\text{Cov}(Z_{t+h-2}; Z_t) + \text{Cov}(Z_{t+h}; Z_{t-2})] \\ &\quad + \theta^2 \text{Cov}(Z_{t+h-2}; Z_{t-2})\end{aligned}$$

$$= \gamma_{00}(h) + \theta \gamma_{02}(h) + \theta \gamma_{20}(h) + \theta^2 \gamma_{22}(h)$$

$$= \begin{cases} 1 + \theta^2; & \text{nếu } h=0 \\ \theta & ; \text{nếu } h = \pm 2 \\ 0 & ; \text{trái lại} \end{cases} \xrightarrow{\theta = 0,8} \begin{cases} 1,64; & \text{nếu } h=0 \\ 0,8; & \text{nếu } h = \pm 2 \\ 0; & \text{trái lại} \end{cases}$$

ACF:

$$\rho(h) = \frac{\gamma_x(h)}{\gamma_x(0)} = \begin{cases} 1 & \text{nếu } h=0 \\ 20/41 & \text{nếu } h = \pm 2 \end{cases}$$

$$b, \text{Var}\left(\frac{1}{4} \sum_{k=1}^4 X_k\right) = \frac{1}{16} \text{Var}\left(\sum_{k=1}^4 X_k\right)$$

$$= \frac{1}{16} \left(\sum_{k=1}^4 \text{Var}(X_k) + 2 \sum_{1 \leq k < l \leq 4} \text{Cov}(X_k; X_l) \right)$$

$$= \frac{1}{16} \left(\sum_{k=1}^4 \gamma_x(0) + 2 \text{Cov}(X_1; X_3) + 2 \text{Cov}(X_2; X_4) \right)$$

$$= \frac{1}{16} (4\gamma_x(0) + 4\gamma_x(2))$$

$$= \frac{1}{4} (\gamma_x(0) + \gamma_x(2)) = 0,61$$

c, Với $\theta = -0,8$ thì $\gamma_x(+2) = 0,8$ và $\gamma_x(0) = 1,64$. Khi đó:

$$\text{Var}\left(\frac{1}{4} \sum_{k=1}^4 X_k\right) = \frac{1}{4} (\gamma_x(0) + \gamma_x(2)) = 0,21 < 0,61.$$

Bài 1.7.

$$\text{Đặt } \mu_x = E(X_t) \text{ và } \mu_y = E(Y_t)$$

$$E(X_t + Y_t) = \mu_x + \mu_y \text{ không phụ thuộc } t.$$

$$\gamma_{x+y}(t+h; t) = \text{Cov}(X_{t+h} + Y_{t+h}; X_t + Y_t)$$

$$= \text{Cov}(X_{t+h}; X_t) + \text{Cov}(Y_{t+h}; X_t) + \text{Cov}(X_{t+h}; Y_t) + \text{Cov}(Y_{t+h}; Y_t)$$

$$= \gamma_x(t+h; t) + \gamma_y(t+h; t)$$

$$= \underbrace{\gamma_x(h) + \gamma_y(h)}_{\text{không phụ thuộc } t} \text{ do } \{X_t\}; \{Y_t\} \text{ là 2 chuỗi dừng.}$$

Do đó $\{X_t + Y_t\}$ là chuỗi dừng.