

时间序列 HW 06

18300290007 力口兴华

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反证：假设 Γ_n 正定且 Γ_{n+1} 半正定。 $\{x_i\}$ 为经中心化的平稳序列。 $\bar{x}_n \triangleq (x_1, \dots, x_n)'$

$$b' \Gamma_n b = (b_1, \dots, b_n)' \Gamma_n (b_1, \dots, b_n) \neq 0. \quad \text{s.t. } E(b' \bar{x}_n)^2 = b' \Gamma_n b \neq 0. \Rightarrow b' \bar{x}_n \neq 0.$$

$$\text{且 } \exists a = (a_1, \dots, a_n, a_{n+1}) \neq 0. \quad \text{s.t. } E(a' \bar{x}_{n+1})^2 = a' \Gamma_{n+1} a = 0. \Rightarrow a' \bar{x}_{n+1} = 0.$$

$\Rightarrow x_{n+1}$ 不能被 \bar{x}_n 线性表示。而 $\{x_i\}$ 平稳 $\Rightarrow x_{n+k+1}$ 不能被 \bar{x}_{n+k} 线性表示。

$\Rightarrow x_{n+k}$ 不能被 \bar{x}_n 线性表示。 $\forall k \geq 1$.

不妨设 $x_{n+k} = \alpha' \bar{x}_n$. $\alpha' = (\alpha_1, \dots, \alpha_n), \alpha \neq 0$.

$$\textcircled{1}: r^{(0)} = \text{Cov}(x_{n+k}, x_{n+k}) = E(\alpha' \bar{x}_n)^2 = \alpha' \Gamma_n \alpha$$

$$\geq \lambda_1 \cdot \|\alpha\|_2^2 > 0 \quad \lambda_1 \text{ 为 } \Gamma_n \text{ 最小特征值.}$$

$$\textcircled{2}: r^{(0)} = \text{Cov}(\alpha' \bar{x}_n, x_{n+k}) = E(\alpha' \bar{x}_n x_{n+k}) = \alpha' \begin{bmatrix} \vdots \\ r^{(n+k-1)} \\ \vdots \\ r^{(k-1)} \end{bmatrix}$$

$\rightarrow \lambda_1 r^{(k)} \xrightarrow{k \nearrow \infty} 0$ 条件。s.t. $r^{(0)} \rightarrow 0$ 与 $\textcircled{1} r^{(0)} > 0$ 矛盾。原假设错误。

\therefore 若 Γ_n 正定 $\Rightarrow \Gamma_{n+1}$ 正定

又 $\Gamma_1 = r^{(0)} > 0$ 正定 $\Rightarrow \Gamma_n$ 正定 $\forall n \in \mathbb{N}^*$

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$$\frac{\Gamma_n \varphi_n}{r^{(0)}} = \frac{\varphi_n}{r^{(0)}} \Rightarrow \begin{bmatrix} R_{n+1} & \tilde{P}_{n+1} \\ \tilde{P}_{n+1}' & p^{(0)} \end{bmatrix} \begin{bmatrix} \alpha_n \\ d_{nn} \end{bmatrix} = \begin{bmatrix} p_{n+1} \\ p^{(0)} \end{bmatrix}$$

$$\left\{ \begin{array}{l} R_{h-1} \cdot \alpha + \tilde{P}_{h-1} \cdot \alpha_{nh} = p_{h-1} \quad ① \\ \tilde{P}_{h-1}' \cdot \alpha + p^{(0)} \cdot \alpha_{nh} = p^{(h)} \quad ② \end{array} \right.$$

$$① : \alpha = R_{h-1}^{-1} (p_{h-1} - \tilde{P}_{h-1} \cdot \alpha_{nh})$$

$$② : (\tilde{P}_{h-1}' R_{h-1}^{-1} p_{h-1} - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1} \alpha_{nh} + p^{(0)} \alpha_{nh} = p^{(h)})$$

$$\Rightarrow \alpha_{nh} = \frac{p^{(h)} - \tilde{P}_{h-1}' R_{h-1}^{-1} p_{h-1}}{p^{(0)} - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1}} \quad \#$$

$$\left. \begin{array}{l} \hat{X}_t = \sum_{i=1}^{h-1} a_i X_{t-i} \\ \hat{X}_{t-h} = \sum_{i=1}^{h-1} b_i X_{t-i} \end{array} \right\} \Rightarrow a = \tilde{b} \quad \# \quad \left. \begin{array}{l} \min E[\varepsilon_t^2] \\ \min E[\delta_{t-h}^2] \end{array} \right\} \Rightarrow \tilde{T}_{h-1} a = \tilde{Y}_{h-1} \\ \tilde{T}_{h-1} b = \tilde{Y}_{h-1}$$

$$E(\varepsilon_t^2) = E[(X_t - \sum_{i=1}^{h-1} a_i X_{t-i})^2] = E[(1, -a) \tilde{X} \cdot \tilde{X}' \cdot \begin{bmatrix} 1 \\ -a \end{bmatrix}] = (1, -a) \tilde{T}_h (1, -a)'$$

$$= (1, -a) \cdot \begin{bmatrix} Y_{(0)} & Y_{h-1}' \\ Y_{h-1} & \tilde{T}_{h-1} \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -a \end{bmatrix}$$

$$= (Y_{(0)} - a' \tilde{Y}_{h-1}, Y_{h-1}' - a' \tilde{T}_{h-1}) \cdot \begin{bmatrix} 1 \\ -a \end{bmatrix}$$

$$= Y_{(0)} - a' \tilde{Y}_{h-1} - Y_{h-1}' a + a' \tilde{T}_{h-1} a \Leftarrow (a = \tilde{T}_{h-1}' \tilde{Y}_{h-1})$$

$$= Y_{(0)} - \tilde{Y}_{h-1}' \tilde{T}_{h-1}' \tilde{Y}_{h-1} = E(\delta_{t-h}^2)$$

$$\varepsilon_t = (1, -a) \cdot \begin{bmatrix} X_t \\ 1 \\ X_{t-h+1} \end{bmatrix} \quad \delta_{t-h} = (\tilde{a}', 1) \cdot \begin{bmatrix} X_{t-1} \\ \vdots \\ X_{t-h} \end{bmatrix}$$

$$\Rightarrow \text{Cov}(\varepsilon_t, \delta_{t-h}) = (1, -a) \cdot E\left\{ \begin{bmatrix} X_t \\ X_{t-h+1} \end{bmatrix} \cdot [X_{t-1}, -X_{t-h}] \right\} \cdot \begin{bmatrix} \tilde{a}' \\ 1 \end{bmatrix}$$

$$= (1, -a) \cdot E\left\{ \begin{bmatrix} X_t \\ \tilde{X}_{h-1} \end{bmatrix} \cdot [\tilde{X}_{h-1}', X_{t-h}] \right\} \cdot \begin{bmatrix} \tilde{a}' \\ 1 \end{bmatrix}$$

$$= (1, -a) \cdot \begin{bmatrix} \tilde{\alpha}_{h-1}' & \gamma^{(h)} \\ \tilde{T}_{h-1} & Y_{h-1} \end{bmatrix} \begin{bmatrix} \tilde{a}' \\ 1 \end{bmatrix} \Leftarrow a = \tilde{T}_{h-1}' Y_{h-1}, \tilde{a} = \tilde{T}_{h-1}' \tilde{Y}_{h-1}$$

$$= (1, \tilde{Y}_{h-1}' \tilde{T}_{h-1}') \cdot \begin{bmatrix} \tilde{\alpha}_{h-1}' & \gamma^{(h)} \\ \tilde{T}_{h-1} & Y_{h-1} \end{bmatrix} \cdot \begin{bmatrix} \tilde{T}_{h-1}' & \tilde{Y}_{h-1} \\ 1 & 1 \end{bmatrix}$$

$$= Y_{(h)} - \tilde{Y}_{h-1}' \tilde{T}_{h-1}' \tilde{Y}_{h-1}$$

$$\therefore \varphi_{nh} = \frac{\text{cov}(\varepsilon_t, \delta_{t-h})}{N E(\varepsilon_t^2) \cdot E(\delta_{t-h}^2)} = \frac{Y_{(h)} - \tilde{Y}_{h-1}' \tilde{T}_{h-1}' \tilde{Y}_{h-1}}{Y_{(0)} - \tilde{Y}_{h-1}' \tilde{T}_{h-1}' \tilde{Y}_{h-1}} = \frac{Y_{(h)} - \tilde{Y}_{h-1}' \tilde{T}_{h-1}' \tilde{Y}_{h-1}}{Y_{(0)} - \tilde{Y}_{h-1}' \tilde{T}_{h-1}' \tilde{Y}_{h-1}} \Leftarrow \begin{cases} \tilde{Y}_{h-1}' \alpha_{nh} = Y_{h-1} \tilde{\alpha}_{h-1} \\ \tilde{Y}_{h-1}' \tilde{\alpha}_{h-1} = Y_{h-1} \alpha_{nh} \end{cases}$$

$$= \frac{p^{(h)} - \tilde{P}_{h-1}' \tilde{T}_{h-1}' P_{h-1}}{p^{(0)} - \tilde{P}_{h-1}' \tilde{T}_{h-1}' \tilde{P}_{h-1}} \quad \#$$

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$$\begin{aligned}
 \text{(a). } \text{MSE} &= E[(y - g(x))^2] \\
 &= E[(y - E(y|x) + E(y|x) - g(x))^2] \\
 &= E[(y - E(y|x))^2] + E[(E(y|x) - g(x))^2] + 2E[(y - E(y|x))(E(y|x) - g(x))]
 \end{aligned}$$

$$\begin{aligned}
 \text{第3项} &= E_x E_{Y|X} [(y - E(y|x))(E(y|x) - g(x))|x] \\
 &= E_x \{ E_{Y|X} [(y|x - E(y|x))] \cdot (E(y|x) - g(x)) \} \\
 &= E_x \{ 0 \cdot (E(y|x) - g(x)) \} = 0.
 \end{aligned}$$

第1项和第2项无关

$$\therefore \underset{g}{\operatorname{argmin}} \text{MSE} = \underset{g}{\operatorname{argmin}} E[(E(y|x) - g(x))^2] = E(y|x).$$

$$\text{i.e. } g(x) = E(y|x) \quad \#$$

$$(b) \quad X, Z \stackrel{\text{i.i.d.}}{\sim} N(0, 1)$$

$$g(x) = E(y|x) = E(x^2 + z|x) = E(x^2|x) + E(z|x) = x^2$$

$$\text{MSE} = E[(y - g(x))^2] = E[z^2] = \text{Var } Z = 1$$

$$\text{(c). } \text{MSE} = E[(y - g(x))^2] = E[(y - a - bx)^2]$$

$$\left\{
 \begin{array}{l}
 \frac{\partial \text{MSE}}{\partial a} = E[2(y - a - bx)] = -2[Ey - a - bEx] = 0 \quad ① \\
 \frac{\partial \text{MSE}}{\partial b} = E[-2x(y - a - bx)] = -2[E(xy) - aEx - bEx^2] = 0 \quad ②
 \end{array}
 \right.$$

$$①: a = Ey - bEx \text{ 代入 } ②$$

$$②: bEx^2 - E(x^2) + (Ey - bEx)Ex = 0.$$

$$\Rightarrow b = \frac{E(XY) - EXEY}{E(X^2) - (EX)^2} = \frac{\text{cov}(X, Y)}{\text{Var}(X)}$$

若 $y = x^2 + z$ $\text{Cov}(x, y) = E(x^3) - 0 \cdot \text{Var}x = 1 \Rightarrow b = 0$ (如果没有(b)中y的形式感觉做不了)

$$\Rightarrow a = E(y) - bE(x) = E(x^2 + z) - 0 = E(x^2) = 1$$

$$\Rightarrow \text{MSE} = E[(y - 1)^2] = E[y^2] - 2E[y] + 1 = E[(x^2 + z)^2] - 1$$

$$= E[x^4 + 2x^2z + z^2] - 1$$

$$= E(x^4) + 2E(x^2)E(z) + E(z^2) - 1$$

$$\text{由 } M_x(\theta) = e^{\frac{\theta^2}{2}} = (1 + \frac{\theta^2}{2} + \frac{1}{2!} \cdot \frac{\theta^4}{4} + \dots)$$

$$\therefore E(x^4) = M_x^{(4)}(0) = 3$$

$$\therefore \text{MSE} = 3 + 0 + 1 - 1 = 3.$$