

时间序列 HW06

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3.11

$$\begin{aligned} (a) \quad \tilde{X}_{n+1} &= E[X_{n+1} | X_n, X_{n-1}, \dots, X_0, \dots] \\ &= P_{\mathcal{H}} X_{n+1} \quad \mathcal{H} = \text{span}\{X_n, X_{n-1}, \dots\}. \end{aligned}$$

$$\therefore X_t = W_t + \theta W_{t-1}$$

$$\begin{aligned} \therefore \tilde{X}_{n+1} &= \tilde{W}_{n+1} + \theta \tilde{W}_n = 0 + \theta W_n \\ &= \theta (X_{t-1} - \theta W_{t-2}) = \theta X_{t-1} - \theta^2 X_{t-2} + \theta^3 X_{t-3} - \dots \\ &= \sum_{i=1}^{\infty} -(-\theta)^i X_{n+1-i} \end{aligned}$$

$$P_{n+1}^n = E[(X_{n+1} - \tilde{X}_{n+1})^2] = E[W_t^2] = \sigma_w^2.$$

$$(b) \quad \tilde{X}_{n+1}^n = \tilde{W}_{n+1}^n + \theta \tilde{W}_n^n$$

$$\begin{cases} \tilde{W}_{n+1}^n = E[W_{n+1} | X_n, \dots, X_1] = 0. \\ \tilde{W}_n^n = E[W_n | X_n, \dots, X_1] = E[W_n | W_n, \dots, W_0] = W_n \end{cases}$$

$$\tilde{X}_{n+1}^n = \theta W_n = -\sum_{i=0}^{n-1} (-\theta)^{i+1} X_{n-i}$$

$$\begin{aligned} X_{n+1}^n - \tilde{X}_{n+1}^n &= W_{n+1} + \theta W_n - \sum_{i=0}^{n-1} (-\theta)^{i+1} (W_{n-i} + \theta W_{n-1-i}) \\ &= W_{n+1} + \theta W_n - \sum_{i=0}^{n-1} (-\theta)^{i+1} W_{n-i} - \sum_{i=1}^n (-\theta)^{i+1} W_{n-i} \\ &= W_{n+1} + \theta W_n - (-\theta) W_n - (-\theta)^{n+1} W_0 \\ &= W_{n+1} - (-\theta)^{n+1} W_0 \end{aligned}$$

$$\therefore W_{n+1} \perp W_0$$

$$\begin{aligned}\therefore E[(X_{n+1} - \tilde{X}_{n+1}^n)^2] &= E[W_{n+1}^2] + (-\theta)^{2n+2} E[W_0^2] \\ &= \sigma_w^2 (1 + \theta^{2+2n})\end{aligned}$$

当 MA invertible. i.e., $|\theta| < 1$.

$$E[(X_{n+1} - \tilde{X}_{n+1}^n)^2] \xrightarrow{n \uparrow \infty} E[(X_{n+1} - \tilde{X}_{n+1})^2] = \sigma_w^2$$

当 n 越大, 有限近似效果越好.

3.15 AR(1): $X_t = \phi X_{t-1} + W_t$

$$\begin{aligned}X_{t+m}^t &= E[X_{t+m} | X_t, \dots, X_1] \\ &= E[\phi^m X_t + \sum_{i=0}^{m-1} \phi^i W_{t+m-i} | X_t, \dots, X_1] \\ &= \phi^m X_t \quad (W_{t+m-i} \perp X_t, \dots, X_1 \quad \forall i \in [0, m-1])\end{aligned}$$

$$X_{t+m} - X_{t+m}^t = \sum_{i=0}^{m-1} \phi^i W_{t+m-i}$$

$$\therefore p_{t+m}^t = \sigma_w^2 \sum_{i=0}^{m-1} \phi^{2i} = \frac{1 - \phi^{2m}}{1 - \phi^2} \sigma_w^2$$

3.16 ARMA(1,1): $X_t = 0.9X_{t-1} + 0.5W_{t-1} + W_t$

$$\tilde{X}_{n+m}^n = -\sum_{j=1}^{m-1} \pi_j \tilde{X}_{n+m-j}^n - \sum_{j=m}^{n+m-1} \pi_j X_{n+m-j}, \quad \text{由 } W_{n+m} = \sum_{j=0}^{\infty} \pi_j X_{n+m-j} = \frac{(1-\phi B)}{(1+\theta B)} X_{n+m} \quad \text{得 } \pi_j \quad (3.91)$$

$$\tilde{X}_{n+m}^n = \phi \tilde{X}_{n+m-1}^n + \theta \tilde{W}_{n+m-1}^n, \quad \text{其中 } \tilde{X}_t^n = \begin{cases} 0 & t \leq 0 \\ X_t & 1 \leq t \leq n \end{cases}, \quad \tilde{W}_t^n = \begin{cases} 0 & t \leq 0 \\ W_t & 1 \leq t \leq n \\ \tilde{X}_t^n - \phi \tilde{X}_{t-1}^n - \theta \tilde{W}_{t-1}^n & t > n \end{cases} \quad (3.92)$$

$$(3.92) \Rightarrow \tilde{X}_{n+m}^n = \phi \tilde{X}_{n+m-1}^n + \theta \tilde{X}_{n+m-1}^n - \phi \theta \tilde{X}_{n+m-2}^n - \theta^2 \tilde{W}_{n-1}^n$$

$$= (\phi + \theta) \tilde{X}_{n+m-1}^n - \theta(\phi + \theta) \tilde{X}_{n+m-2}^n + \theta^2 \phi \tilde{X}_{n+m-3}^n + \theta^3 \tilde{W}_{n+m-3}^n$$

$$\vdots \\ = \sum_{j=1}^{m-1} (-\theta)^{j-1} (\phi + \theta) \tilde{X}_{n+m-j}^n + (-\theta)^{m-1} [\phi \tilde{X}_n^n + \theta \tilde{W}_n^n]$$

$$= \sum_{j=1}^{m-1} (-\theta)^{j-1} (\phi + \theta) \tilde{X}_{n+m-j}^n + (-\theta)^{m-1} [\phi X_n + \theta W_n] \quad \text{因为 } W_t = X_t - \phi X_{t-1} - \theta W_{t-1},$$

$$= \sum_{j=1}^{m-1} (-\theta)^{j-1} (\phi + \theta) \tilde{X}_{n+m-j}^n + \sum_{j=m}^{n+m-1} (-\theta)^{j-1} (\theta + \phi) X_{n+m-j}$$

$$(3.91) \Rightarrow 1 + \pi_1 B + \pi_2 B^2 + \dots = (1 - \phi B)(1 - \theta B + \theta B^2 - \theta B^3 + \dots)$$

$$\Rightarrow \pi_i = (-\theta)^{i-1} (\phi + \theta)$$

\therefore 在 ARMA(1,1) 下, 3.91 与 3.92 预测公式表达等价

$$3.46. \quad \text{令 } X_i^0 = 0, \quad P_i^0 = r(0).$$

为方便书写记 $e_i = X_i - X_i^{i-1}$. $\langle a, b \rangle =: E(ab)$.

$$1^0: \text{ 假设 } t=k \text{ 时: } X_{t+1}^t = \sum_{j=1}^t \theta_{t,j} e_j. \quad \text{i.e., } E[X_{t+1} | X_1, \dots, X_t] = E[X_{t+1} | e_1, \dots, e_t]$$

$$\begin{aligned} \text{则 } X_{k+1}^k &= E[X_{k+1} | X_1, \dots, X_k] \\ &= E[X_{k+1} | e_1, \dots, e_{k-1}, X_k] \end{aligned}$$

$$\alpha \langle X_k - X_k^{k-1}, e_i \rangle = 0, \quad i=1, 2, \dots, k-1$$

$$\Rightarrow \text{span}\{e_1, e_2, \dots, e_{k-1}, X_k\} = \text{span}\{e_1, \dots, e_k\}$$

$$\therefore X_{k+1}^k = E[X_{k+1} | e_1, \dots, e_k] = \sum_{j=1}^k \theta_{k,k-j} \cdot e_j$$

$$\therefore \text{归纳得 } X_{t+1}^t = \sum_{j=1}^t \theta_{t,t-j} e_j \quad j=1, 2, \dots \quad \text{也即 } X_t^{t-1} = \sum_{j=1}^{t-1} \theta_{t,j} e_{t+1-j} \quad \#$$

$$2^0: \quad P_t^{t-1} = E[e_t^2] = \langle e_t, e_t \rangle = \langle X_t - X_t^{t-1}, X_t - X_t^{t-1} \rangle$$

$$= r(0) + \langle X_t^{t-1}, X_t^{t-1} \rangle - 2 \langle X_t, X_t^{t-1} \rangle$$

$$= r(0) - \langle X_t^{t-1}, X_t^{t-1} \rangle - 2 \overbrace{\langle e_t, X_t^{t-1} \rangle}^{=0}$$

$$= r(0) - \sum_{j=1}^{t-1} \theta_{t,t-j}^2 \langle e_j, e_j \rangle$$

$$= r(0) - \sum_{j=1}^{t-1} \theta_{t,t-j}^2 P_j^{j-1} \quad \#$$

$$3^0: \quad E[e_t \cdot e_j] = \langle X_t, X_j \rangle - \langle X_t^{t-1}, X_j \rangle - \langle X_t, X_j^{j-1} \rangle + \langle X_t^{t-1}, X_j^{j-1} \rangle$$

$$= r(t-j) - \langle X_t^{t-1}, e_j + X_j^{j-1} \rangle - \langle e_t + X_t^{t-1}, X_j^{j-1} \rangle + \langle X_t^{t-1}, X_j^{j-1} \rangle$$

$$= r(t-j) - \langle X_t^{t-1}, e_j \rangle - \langle e_t, X_j^{j-1} \rangle - \langle X_t^{t-1}, X_j^{j-1} \rangle$$

$$= r(t-j) - \theta_{t,tj} \cdot \langle e_j, e_j \rangle - 0 - \langle X_t^{t+1}, X_j^{j+1} \rangle$$

$$= 0. \quad (\text{因 } e_i \perp e_j)$$

$$\Rightarrow \theta_{t,tj} \cdot p_j^{j+1} = r(t-j) - \langle X_t^{t+1}, X_j^{j+1} \rangle = r(t-j) - \sum_{k=0}^{j-1} \theta_{j,j-k} \theta_{t,t+k} p_{k+1}^k$$

$$\therefore \theta_{t,tj} = \left\{ r(t-j) - \sum_{k=0}^{j-1} \theta_{j,j-k} \theta_{t,t+k} p_{k+1}^k \right\} / p_j^{j+1} \quad \#.$$