

2021 秋季本科时间序列

第 9 次作业答案

12 月 26 日

1. (a) 若 Φ 的特征值模长小于 1, 则 X_t 平稳。

$$|\Phi - \lambda I| = \begin{vmatrix} 0.1 - \lambda & 0.5 \\ -0.4 & 1 - \lambda \end{vmatrix} = \lambda^2 - 1.1\lambda + 0.3 = 0$$

解得 $\lambda_1 = 0.6, \lambda_2 = 0.5$, 故满足平稳性要求。 $\mathbb{E}X_t = c + \Phi \mathbb{E}X_{t-1}$, 则有

$$\mathbb{E}X_t = (I - \Phi)^{-1}c = \begin{bmatrix} 0 & 2.5 \\ -2 & 4.5 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2.5 \\ 2.5 \end{bmatrix}$$

(b) 把 $\lambda = 0.6$ 代入得,

$$(\Phi - \lambda I)x_1 = \begin{bmatrix} -0.5 & 0.5 \\ -0.4 & 0.4 \end{bmatrix} x_1 = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

同理, 当 $\lambda = 0.5$ 时, 解得 $x_2 = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$, 故 $A = [x_1 \ x_2] = \begin{bmatrix} 1 & 5 \\ 1 & 4 \end{bmatrix}, A^{-1} = \begin{bmatrix} -4 & 5 \\ 1 & -1 \end{bmatrix}$

$$\Phi^i = A\Lambda^i A^{-1} = \begin{bmatrix} 1 & 5 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 0.6^i & 0 \\ 0 & 0.5^i \end{bmatrix} \begin{bmatrix} -4 & 5 \\ 1 & -1 \end{bmatrix} = \begin{bmatrix} -4 \times 0.6^i + 5 \times 0.5^i & 5 \times 0.6^i - 5 \times 0.5^i \\ -4 \times 0.6^i + 4 \times 0.5^i & 5 \times 0.6^i - 4 \times 0.5^i \end{bmatrix}$$

X_t 的 $MA(\infty)$ 展开为 $X_t = (I + \Phi + \dots + \Phi^n)c + \sum_{i=0}^n \Phi^i \epsilon_{t-i}, n \rightarrow \infty$, 则

$$\begin{aligned} \text{var}X_t &= \sum_{i=0}^{\infty} \Phi^i \Omega \Phi^i \\ &= \sum_{i=0}^{\infty} A\Lambda^i A^{-1} \Omega (A^{-1})^T \Lambda^{Ti} A^T \\ &= \begin{bmatrix} \frac{475}{84} & \frac{515}{84} \\ \frac{515}{84} & \frac{1003}{84} \end{bmatrix} \end{aligned}$$

(c) 设 $\Omega = ADA^T$, 其中 $A = \begin{bmatrix} 1 & 0 \\ a & 1 \end{bmatrix}$, $D = \begin{bmatrix} b & 0 \\ 0 & c \end{bmatrix}$, 解得 $a = \frac{1}{2}, b = 2, c = \frac{7}{2}$, 令

$P = AD^{\frac{1}{2}}$, 则 $P^T = D^{\frac{1}{2}}A^T$, 则 $\Omega = PP^T$

$$P = AD^{\frac{1}{2}} = \begin{bmatrix} 1 & 0 \\ \frac{1}{2} & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{\frac{7}{2}} \end{bmatrix} = \begin{bmatrix} \sqrt{2} & 0 \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{14}}{2} \end{bmatrix}, \Omega = \begin{bmatrix} \sqrt{2} & 0 \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{14}}{2} \end{bmatrix} \begin{bmatrix} \sqrt{2} & \frac{\sqrt{2}}{2} \\ 0 & \frac{\sqrt{14}}{2} \end{bmatrix}$$

$$X_{t+s} = (I - \Phi)^{-1}c + \sum_{i=0}^{\infty} \Phi^i P u_{t+s-i}, \frac{\partial X_{t+s}}{\partial u_{kt}} = \Phi^s p_k$$

则

$$\frac{\partial X_{1t+s}}{\partial u_{1t}} = \sqrt{2}(-4 \times 0.6^s + 5 \times 0.5^s) + \frac{\sqrt{2}}{2}(5 \times 0.6^s - 5 \times 0.5^s)$$

$$\frac{\partial X_{2t+s}}{\partial u_{1t}} = \sqrt{2}(-4 \times 0.6^s + 4 \times 0.5^s) + \frac{\sqrt{2}}{2}(5 \times 0.6^s - 4 \times 0.5^s)$$

$$\frac{\partial X_{1t+s}}{\partial u_{2t}} = \frac{\sqrt{14}}{2}(5 \times 0.6^s - 5 \times 0.5^s)$$

$$\frac{\partial X_{2t+s}}{\partial u_{2t}} = \frac{\sqrt{14}}{2}(5 \times 0.6^s - 4 \times 0.5^s)$$

2. (a) 令 $Y = \begin{bmatrix} X_2 \\ \vdots \\ X_t \end{bmatrix}$, $X = \begin{bmatrix} X_1 \\ \vdots \\ X_{t-1} \end{bmatrix}$, $u = \begin{bmatrix} u_2 \\ \vdots \\ u_t \end{bmatrix}$, 有 $Y = X\rho + u$, 则 OLS 估计量

$$\begin{aligned} \hat{\rho} &= (X^T X)^{-1} X^T Y \\ &= (X^T X)^{-1} X^T (X\rho + u) \\ &= \rho + (X^T X)^{-1} X^T u \\ &= \rho + \left(\frac{1}{T} X^T X\right)^{-1} \frac{1}{T} X^T u \\ &= \rho + \left(\frac{1}{T} \sum_{t=1}^T X_t^2\right)^{-1} \frac{1}{T} \sum_{t=1}^T X_{t-1} u_t \end{aligned}$$

又因为 $\mathbb{E}[u_t] = \mathbb{E}[\mathbb{E}[u_t | X_{t-1}, X_{t-2}, \dots, X_1]] = \sqrt{\alpha_0 + \alpha_1 u_{t-1}^2} \mathbb{E}[\epsilon_t | \Omega_{t-1}] = 0$, 所以 $\frac{1}{T} \sum_{t=1}^T X_{t-1} u_t \xrightarrow{a.s.} 0$, OLS 估计依然具有一致性, 即 $\hat{\rho} \xrightarrow{a.s.} \rho$

(b)

$$\sqrt{T}(\hat{\rho} - \rho) = \left(\frac{1}{T} \sum_{t=1}^T X_t^2\right)^{-1} \frac{1}{T} \sum_{t=1}^T X_t u_t$$

其中 $\frac{1}{T} \sum_{t=1}^T X_t u_t \xrightarrow{d} N(0, \Sigma)$, $\Sigma = \lim_T \frac{1}{T} \sum_t \mathbb{E}[u_t^2 X_t^2]$ 。设 $\lim_{T \rightarrow \infty} \frac{1}{T} \sum_t X_t^2 = M$, 则 $\sqrt{T}(\hat{\rho} - \rho) \xrightarrow{d} N(0, M^{-1} \Sigma M^{-1})$

(c) 不需要知道 α_0, α_1 的参数取值。

(d) 首先利用 $\hat{\rho}$ 得到 $\hat{u}_t, \hat{u}_t = Y_t - \hat{\rho}X_t$, 进而计算 $\hat{\Sigma} = \frac{1}{T} \sum_{t=1}^T \hat{u}_t^2 X_t^2$, 则 $\hat{\rho}$ 的异方差渐进标准误是样本渐进协方差 $\frac{1}{T} \hat{M}^{-1} \hat{\Sigma} \hat{M}^{-1}$ 的平方根。

若已知 α_0, α_1 的取值, 则

$$\lim_{T \rightarrow \infty} \frac{1}{T} \sum_t \mathbb{E}[u_t^2 X_t^2] = \lim_{T \rightarrow \infty} \frac{1}{T} \sum_t \mathbb{E}[(\alpha_0 + \alpha_1 u_{t-1}^2) \epsilon_t^2 X_t^2]$$

因为 u_t 为一个平稳 ARCH 过程, 故 $\mathbb{E}[u_t^2] = \frac{\alpha_0}{1-\alpha_1}$, 故上式为

$$\left(\alpha_0 + \frac{\alpha_0 \alpha_1}{1 - \alpha_1}\right) \lim_{T \rightarrow \infty} \frac{1}{T} \sum_t \mathbb{E}[\epsilon_t^2 X_t^2]$$

设 $\frac{1}{T} \sum_{t=1}^T X_t \epsilon_t \xrightarrow{d} N(0, \Sigma')$, 有 $\sqrt{T}(\hat{\rho} - \rho) \xrightarrow{d} N(0, \frac{\alpha_0}{1-\alpha_1} M^{-1} \Sigma' M^{-1})$

3. (a)

$$T(\hat{\rho} - 1) = \frac{\frac{1}{T} \sum_{t=1}^T \Delta X_t X_{t-1}}{\frac{1}{T^2} \sum_{t=1}^T X_{t-1}^2} \xrightarrow{d} \frac{\frac{\sigma_\epsilon^2}{2} B(1)^2 - \frac{\sigma_\epsilon^2}{2}}{\sigma_\epsilon^2 \int_0^1 B(s)^2 ds} = \frac{0.5(B(1)^2 - 1)}{\int_0^1 B(s)^2 ds}$$

可得

$$T^{1-\delta}(\hat{\rho} - 1) \xrightarrow{d} T^{-\delta} \frac{0.5(B(1)^2 - 1)}{\int_0^1 B(s)^2 ds}$$

当 $T \rightarrow \infty, T^{-\delta} \xrightarrow{\text{a.s.}} 0$, 故对任意的 $\delta > 0, T^{1-\delta}(\hat{\rho} - 1) \xrightarrow{\text{a.s.}} 0$

(b) $T(\hat{\rho} - 1) \xrightarrow{\text{a.s.}} T(\rho - 1)$, 由 $|\rho| < 1$ 有 $\rho - 1 < 0$, 故 $T \rightarrow \infty$ 时有 $T(\rho - 1) \rightarrow -\infty$, 即 $T(\hat{\rho} - 1) \xrightarrow{\text{a.s.}} -\infty$ 。