

2019 秋季本科时间序列  
第 4 次作业参考答案

2019 年 11 月 12 日

1. (a)

由题目可知  $Y_t = \frac{1}{1-\lambda} X_t$ , 且  $|\lambda| > 1$ , 故可得  $Y_t = \sum_{i=0}^{\infty} \left(\frac{1}{\lambda^i}\right) X_{t-i}$   
故  $E(Y_t) = \sum_{i=0}^{\infty} \left(\frac{1}{\lambda^i}\right) E(X_{t-i}) = \frac{1}{1-\lambda} E(X_{t-i})$ , 又知  $X_t$  为平稳时间序列, 则  $E(Y_t)$  与  $t$  无关

$$\begin{aligned} cov(Y_t, X_{t-j}) &= cov\left(\sum_{i=0}^{\infty} \left(\frac{1}{\lambda^i}\right) X_{t-i}, X_{t-j}\right) \\ &= \sum_{i=0}^{\infty} \left(\frac{1}{\lambda^i}\right) cov(X_{t-i}, X_{t-j}) \\ &= \sum_{i=0}^{\infty} \left(\frac{1}{\lambda^i}\right) \sigma_X^2(j-i) \\ cov(Y_t, Y_{t-k}) &= cov\left(Y_t, \sum_{m=0}^{\infty} \left(\frac{1}{\lambda^m}\right) X_{t-k-m}\right) \\ &= \sum_{m=0}^{\infty} \left(\frac{1}{\lambda^m}\right) cov(Y_t, X_{t-k-m}) \\ &= \sum_{m=0}^{\infty} \left(\frac{1}{\lambda^m}\right) \sum_{i=0}^{\infty} \left(\frac{1}{\lambda^i}\right) \sigma_X^2(k+m-i) \end{aligned}$$

由上式易知  $\sigma_Y^2(k) = cov(Y_t, Y_{t-k})$  与  $t$  无关  
即可证  $Y_t$  为平稳时间序列

(b)

对  $AR(p)$  过程有

$$\begin{aligned} X_t &= \sum_{i=1}^p \phi_i X_{t-i} + \varepsilon_t \\ \left(1 - \sum_{i=1}^p \phi_i \mathcal{L}^i\right) X_t &= \varepsilon_t \end{aligned}$$

已知  $z_1 \dots z_p$  为特征多项式  $A(z)$  的根

$$A(z) = 1 - \phi_1 z - \dots - \phi_p z^p = \left(1 - \frac{z}{z_1}\right) \left(1 - \frac{z}{z_2}\right) \dots \left(1 - \frac{z}{z_p}\right)$$

把  $\mathcal{L}$  带入  $A(z)$ , 则有  $A(\mathcal{L}) = (1 - \frac{\mathcal{L}}{z_1})(1 - \frac{\mathcal{L}}{z_2}) \dots (1 - \frac{\mathcal{L}}{z_p})$ , 则有如下

$$\begin{aligned} X_t &= \frac{\varepsilon_t}{A(\mathcal{L})} \\ &= \frac{1}{(1 - \frac{\mathcal{L}}{z_1})(1 - \frac{\mathcal{L}}{z_2}) \dots (1 - \frac{\mathcal{L}}{z_p})} \varepsilon_t \\ &= \frac{1}{(1 - \frac{\mathcal{L}}{z_1})} \frac{1}{(1 - \frac{\mathcal{L}}{z_2})} \dots \frac{1}{(1 - \frac{\mathcal{L}}{z_p})} \varepsilon_t \end{aligned}$$

其中  $\varepsilon_t$  为白噪声, 故其为平稳时间序列。由 (a) 中结论可知, 当  $|z_p| > 1$  时,  $\frac{1}{(1 - \frac{\mathcal{L}}{z_p})} \varepsilon_t$  为平稳时间序列, 依次类推可知当  $z_1 \dots z_p$  的模长均大于 1 时,  $X_t$  为平稳时间序列, 即  $AR(p)$  过程平稳

## 2. (a)

已知特征多项式  $A(z) = 1 - \phi_1 z - \phi_2 z^2$ ,  $A(0) = 1 > 0$  恒成立

当  $\lambda$  和  $\eta$  为实数时, 若两实根模长大于 1, 则必有

$$\begin{cases} |\lambda\eta| = |-\frac{1}{\phi_2}| > 1 \Rightarrow |\phi_2| < 1 \\ \Delta = \phi_1^2 + 4\phi_2 > 0 \end{cases}$$

分情况讨论 (1) 当  $-1 < \phi_2 < 0$  时, 函数开口向上, 若使两实根模长大于 1, 则必有

$$\begin{cases} -1 < \phi_2 < 0 \\ \Delta = \phi_1^2 + 4\phi_2 > 0 \\ A(1) > 0 \\ A(-1) > 0 \end{cases}$$

(2) 当  $0 < \phi_2 < 1$  时, 函数开口向下, 若使两实根模长大于 1, 则必有

$$\begin{cases} 0 < \phi_2 < 1 \\ \Delta = \phi_1^2 + 4\phi_2 > 0 \\ A(1) > 0 \\ A(-1) > 0 \end{cases}$$

解得此时  $\phi_1$  和  $\phi_2$  的取值满足

$$\begin{cases} \phi_2 < 1 \pm \phi_1 \\ \phi_1^2 > -4\phi_2 \end{cases}$$

当  $\lambda$  和  $\eta$  为复数时, 有

$$\begin{aligned} \lambda &= \frac{\phi_1 + i\sqrt{-\phi_1^2 - 4\phi_2}}{-2\phi_2} \\ \eta &= \frac{\phi_1 - i\sqrt{-\phi_1^2 - 4\phi_2}}{-2\phi_2} \end{aligned}$$

两复根共轭, 故模长相等,  $|\lambda\eta| > 1$  即可说明两根模长均大于 1, 于是, 若存在两个模长大于 1 的复根, 则以下条件成立

$$\begin{cases} |\lambda\eta| = \left| \frac{\phi_1^2 + (-\phi_1^2 - 4\phi_2)}{4\phi_2^2} \right| = \left| \frac{1}{\phi_2} \right| > 1 \\ \Delta = \phi_1^2 + 4\phi_2 < 0 \end{cases}$$

解得此时  $\phi_1$  和  $\phi_2$  的取值满足

$$\begin{cases} |\phi_2| < 1 \\ \phi_1^2 < -4\phi_2 \end{cases}$$

(b)

$$\text{系数矩阵 } \Phi = \begin{bmatrix} \phi_1 & \phi_2 - 1 & 0 \\ \phi_2 & \phi_1 & -1 \\ 1 & -\phi_1 & -\phi_2 \end{bmatrix}$$

$$|\Phi| = (\phi_2 + 1)(\phi_1 + \phi_2 - 1)(\phi_2 - \phi_1 - 1)$$

结合 (a) 可知, 当满足平稳性条件时,  $|\Phi| \neq 0$ , 系数矩阵可逆

$$\begin{aligned} \Phi^{-1} &= \frac{\Phi^*}{|\Phi|} \\ \Phi^* &= \begin{bmatrix} A_{11} & A_{21} & A_{31} \\ A_{12} & A_{22} & A_{32} \\ A_{13} & A_{23} & A_{33} \end{bmatrix} \\ \Phi \begin{bmatrix} \sigma_X^2(0) \\ \sigma_X^2(1) \\ \sigma_X^2(2) \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \\ \sigma_\varepsilon^2 \end{bmatrix} \end{aligned}$$

故有

$$\begin{bmatrix} \sigma_X^2(0) \\ \sigma_X^2(1) \\ \sigma_X^2(2) \end{bmatrix} = \Phi^{-1} \begin{bmatrix} 0 \\ 0 \\ \sigma_\varepsilon^2 \end{bmatrix} = \frac{\Phi^*}{|\Phi|} \begin{bmatrix} 0 \\ 0 \\ \sigma_\varepsilon^2 \end{bmatrix} = \frac{\sigma_\varepsilon^2}{(\phi_2 + 1)(\phi_1 + \phi_2 - 1)(\phi_2 - \phi_1 - 1)} \begin{bmatrix} 1 - \phi_2 \\ \phi_1 \\ \phi_1^2 - \phi_2^2 + \phi_2 \end{bmatrix}$$

(c)

令

$$\begin{aligned} |\mu I - A| &= \begin{vmatrix} \mu - \phi_1 & -\phi_2 \\ -1 & \mu \end{vmatrix} \\ &= \mu^2 - \phi_1\mu - \phi_2 = 0 \end{aligned}$$

即  $1 - \phi_1 \frac{1}{\mu} - \phi_2 \frac{1}{\mu^2} = 0$ , 又  $\lambda, \eta \neq 0$ , 故得到特征值为  $\mu_1 = \frac{1}{\lambda}, \mu_2 = \frac{1}{\eta}$   
当  $\mu = \mu_1 = \frac{1}{\lambda}$  时

$$\begin{aligned} (\mu I - A)x &= \begin{bmatrix} \frac{1}{\lambda} - \phi_1 & -\phi_2 \\ -1 & \frac{1}{\lambda} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ &= 0 \end{aligned}$$

又  $(\frac{1}{\lambda} - \phi_1)\frac{1}{\lambda} - \phi_2 = \frac{1}{\lambda^2} - \phi_1\frac{1}{\lambda} - \phi_2 = 0$   
故取

$$x = \begin{bmatrix} \frac{1}{\lambda} \\ 1 \end{bmatrix}$$

单位化得特征向量

$$\xi_1 = \begin{bmatrix} \frac{1}{\sqrt{1+\lambda^2}} \\ \frac{\lambda}{\sqrt{1+\lambda^2}} \end{bmatrix}$$

同理特征向量  $\xi_2 = \begin{bmatrix} \frac{1}{\sqrt{1+\eta^2}} \\ \frac{\eta}{\sqrt{1+\eta^2}} \end{bmatrix}$ , 特征向量矩阵为  $S = \begin{bmatrix} \frac{1}{\sqrt{1+\lambda^2}} & \frac{1}{\sqrt{1+\eta^2}} \\ \frac{\lambda}{\sqrt{1+\lambda^2}} & \frac{\eta}{\sqrt{1+\eta^2}} \end{bmatrix}$

特征向量矩阵的逆矩阵为

$$\begin{aligned} S^{-1} &= \frac{1}{|S|} \begin{bmatrix} \frac{\eta}{\sqrt{1+\eta^2}} & -\frac{1}{\sqrt{1+\eta^2}} \\ -\frac{\lambda}{\sqrt{1+\lambda^2}} & \frac{1}{\sqrt{1+\lambda^2}} \end{bmatrix} \\ &= \frac{1}{\eta - \lambda} \begin{bmatrix} \eta\sqrt{1+\lambda^2} & -\sqrt{1+\lambda^2} \\ -\lambda\sqrt{1+\eta^2} & \sqrt{1+\eta^2} \end{bmatrix} \end{aligned}$$

则

$$\begin{aligned} A^k &= S\Lambda^k S^{-1} \\ &= \frac{1}{\eta - \lambda} \begin{bmatrix} \frac{1}{\sqrt{1+\lambda^2}} & \frac{1}{\sqrt{1+\eta^2}} \\ \frac{\lambda}{\sqrt{1+\lambda^2}} & \frac{\eta}{\sqrt{1+\eta^2}} \end{bmatrix} \begin{bmatrix} \frac{1}{\lambda^k} & 0 \\ 0 & \frac{1}{\eta^k} \end{bmatrix} \begin{bmatrix} \eta\sqrt{1+\lambda^2} & -\sqrt{1+\lambda^2} \\ -\lambda\sqrt{1+\eta^2} & \sqrt{1+\eta^2} \end{bmatrix} \\ &= \frac{1}{\eta - \lambda} \begin{bmatrix} \frac{1}{\lambda^k\sqrt{1+\lambda^2}} & \frac{1}{\eta^k\sqrt{1+\eta^2}} \\ \frac{1}{\lambda^{k+1}\sqrt{1+\lambda^2}} & \frac{1}{\eta^{k+1}\sqrt{1+\eta^2}} \end{bmatrix} \begin{bmatrix} \eta\sqrt{1+\lambda^2} & -\sqrt{1+\lambda^2} \\ -\lambda\sqrt{1+\eta^2} & \sqrt{1+\eta^2} \end{bmatrix} \\ &= \frac{1}{(\lambda\eta)^k(\eta - \lambda)} \begin{bmatrix} \eta^{k+1} - \lambda^{k+1} & \lambda^k - \eta^k \\ \lambda\eta(\eta^k - \lambda^k) & \lambda\eta(\lambda^{k-1} - \eta^{k-1}) \end{bmatrix} \end{aligned}$$

(d)

当  $\lambda = \eta$  时,  $A$  只有一个特征值  $\mu = \frac{\phi_1}{2}$  (二重代数重数), 有  $\phi_1^2 + 4\phi_2 = 0$

$$\begin{aligned} (\mu I - A) &= \begin{bmatrix} -\frac{\phi_1}{2} & -\phi_2 \\ -1 & \frac{\phi_1}{2} \end{bmatrix} \\ &\rightarrow \begin{bmatrix} -\frac{\phi_1}{2} & -\phi_2 \\ \frac{\phi_1}{2} & -\frac{\phi_1^2}{4} \end{bmatrix} \\ &\rightarrow \begin{bmatrix} -\frac{\phi_1}{2} & -\phi_2 \\ 0 & 0 \end{bmatrix} \end{aligned}$$

得  $\text{rank}(\lambda I - A) = 1$ ,  $A$  无两个线性无关的特征向量。

故其 Jordan 标准型有 1 一个 Jordan 块, 其 Jordan 标准型为

$$J = \begin{bmatrix} \frac{\phi_1}{2} & 1 \\ 0 & \frac{\phi_1}{2} \end{bmatrix}$$

由  $(\mu I - A)z = 0$  可得  $t_1 = \begin{bmatrix} \frac{\phi_1}{2} \\ 1 \end{bmatrix}$  为链首

记  $T = (t_1, t_2), t_2 = (t_{21}, t_{22})^T$  有

$$(At_1, At_2) = (t_1, t_2) \begin{bmatrix} \frac{\phi_1}{2} & 1 \\ 0 & \frac{\phi_1}{2} \end{bmatrix}$$

有

$$\begin{cases} \frac{\phi_1}{2}t_{21} + \phi_2 t_{22} = \frac{\phi_1}{2} \\ t_{21} = 1 + \frac{\phi_1}{2}t_{22} \end{cases}$$

故取  $t_2 = (1, 0)^T$ , 变换矩阵则为  $T = \begin{bmatrix} \frac{\phi_1}{2} & 1 \\ 1 & 0 \end{bmatrix}$

其逆矩阵为

$$T^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & -\frac{\phi_1}{2} \end{bmatrix}$$

故

$$\begin{aligned} A^k &= T J^k T^{-1} \\ &= \begin{bmatrix} \frac{\phi_1}{2} & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{\phi_1^k}{2^k} & \frac{k\phi_1^{k-1}}{2^{k-1}} \\ 0 & \frac{\phi_1^k}{2^k} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -\frac{\phi_1}{2} \end{bmatrix} \\ &= \frac{\phi_1^k}{2^k} \begin{bmatrix} 1+k & -k \\ \frac{\phi_1}{2}k & 1-k \end{bmatrix} \end{aligned}$$