



PROBLEM 4

$$y_t = x_t + \epsilon_t$$

$$x_t = \phi x_{t-1} + \omega_t$$

$\epsilon_t, \omega_t \text{ iid } N(0, 1)$

(a) Assuming that $|\phi| < 1$, it follows that the ACF of $\{x_t\}$ is given by $\rho_0^x = 1, \rho_1^x = \frac{\gamma_1^x}{\gamma_0^x}, \rho_2^x = \frac{\gamma_2^x}{\gamma_0^x}, \dots$, where

$$\gamma_0^x = \frac{1}{1-\phi^2}, \quad \gamma_\ell^x = \frac{\phi^{2|\ell|}}{1-\phi^2} \quad \ell = \pm 1, \pm 2, \dots$$

Therefore

$$\gamma_0^y = V(y_t) = V(x_t) + 1 = \gamma_0^x + 1 = \frac{2-\phi^2}{1-\phi^2}$$

$$\gamma_\ell^y = E(y_t y_{t-\ell}) = E((x_t + \epsilon_t)(x_{t-\ell} + \epsilon_{t-\ell})) = \gamma_\ell^x \quad \ell = \pm 1, \pm 2, \dots$$

$$\rightarrow \rho_\ell^x = \frac{\phi^{2|\ell|}/(1-\phi^2)}{(2-\phi^2)/(1-\phi^2)} = \frac{\phi^{2|\ell|}}{2-\phi^2} \quad \ell = \pm 1, \pm 2, \dots \quad \#$$

(b) $E(y_{m+1} | y_{1:m}, \phi) = E(x_{m+1} + \epsilon_{m+1} | y_{1:m}, \phi)$
 $= E(\phi x_m + \epsilon_{m+1} + \omega_{m+1} | y_{1:m}, \phi)$
 $= \phi E(x_m | y_{1:m}, \phi) = \phi E(y_m - \epsilon_m | y_{1:m}, \phi)$
 $= \phi y_m$

$$E(y_{m+2} | y_{1:m}, \phi) = E(\phi x_{m+1} + \omega_{m+2} + \epsilon_{m+2} | y_{1:m}, \phi)$$

$$= \phi E(x_{m+1} | y_{1:m}, \phi) =$$

$$= \phi E(\phi x_m + \omega_{m+1} | y_{1:m}, \phi) =$$

$$= \phi^2 E(x_m | y_{1:m}, \phi) = \phi^2 y_m$$



(b) (continuing)

(2)

$$\begin{aligned} V(y_{n+1} | y_{1:n}, \phi) &= V(\phi x_n + w_{n+1} + \epsilon_{n+1} | y_{1:n}, \phi) \\ &= V(\phi(y_n - \epsilon_n) + w_{n+1} + \epsilon_{n+1} | y_{1:n}, \phi) \\ &= \phi V(\epsilon_n) + V(w_{n+1}) + V(\epsilon_{n+1}) \\ &= \phi^2 + 2 \end{aligned}$$

$$\begin{aligned} V(y_{n+2} | y_{1:n}, \phi) &= V(x_{n+2} + \epsilon_{n+2} | y_{1:n}, \phi) \\ &= V(\phi x_{n+1} + w_{n+2} + \epsilon_{n+2} | y_{1:n}, \phi) \\ &= V(\phi^2 x_n + \phi w_{n+1} + w_{n+2} + \epsilon_{n+2} | y_{1:n}, \phi) \\ &= V(\phi^2 y_n - \phi^2 \epsilon_n + \phi w_{n+1} + w_{n+2} + \epsilon_{n+2} | y_{1:n}, \phi) \\ &= \phi^4 + \phi^2 + 1 + 1 \\ &= \phi^4 + \phi^2 + 2 \end{aligned}$$

To sum up,

$$\hat{y}_{n+1} = \phi y_n \quad \epsilon_{n+1} = y_{n+1} - \hat{y}_{n+1} \quad V(\epsilon_{n+1}) = 2 + \phi^2$$

$$\hat{y}_{n+2} = \phi^2 y_n \quad \epsilon_{n+2} = y_{n+2} - \hat{y}_{n+2} \quad V(\epsilon_{n+2}) = 2 + \phi^2 + \phi^4$$

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(C) It is easy to see that

$$y_t = \epsilon_t + \sum_{j=0}^{t-1} \phi^j \omega_{t-j}$$

(recall that $x_0 = 0$)

for $t = 1, 2, \dots, n$. Therefore

$$E(y_t | \phi) = 0$$

$$V(y_t | \phi) = 1 + \sum_{j=0}^{t-1} \phi^{2j} = 1 + \frac{1 - \phi^{2t}}{1 - \phi^2}$$

How about $E(y_t y_{t+h} | \phi)$?

$$E(y_t y_{t+h}) = E \left\{ \left(\epsilon_t + \sum_{j=0}^{t-1} \phi^j \omega_{t-j} \right) \left(\epsilon_{t+h} + \sum_{j=0}^{t+h-1} \phi^j \omega_{t+h-j} \right) \right\}$$

$$\Rightarrow = \sum_{j=0}^{h-1} \phi^j \omega_{t+h-j} + \sum_{j=h}^{t+h-1} \phi^j \omega_{t+h-j} = \sum_{j=0}^{h-1} \phi^j \omega_{t+h-j} + \sum_{j=0}^{t-1} \phi^{j+h} \omega_{t-j}$$

$$\Rightarrow E(y_t y_{t+h}) = E \left\{ \left(\epsilon_t + \sum_{j=0}^{t-1} \phi^j \omega_{t-j} \right) \left(\epsilon_{t+h} + \sum_{j=0}^{h-1} \phi^j \omega_{t+h-j} + \sum_{j=0}^{t-1} \phi^{j+h} \omega_{t-j} \right) \right\}$$

$$= \sum_{j=0}^{t-1} \phi^{2j+h}$$



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For instance,

$$E(y_1 y_2) = \phi$$

$$E(y_2 y_3) = \phi + \phi^3$$

$$E(y_1 y_3) = \phi^2$$

$$E(y_2 y_4) = \phi^2 + \phi^4$$

$$\vdots$$
$$E(y_1 y_m) = \phi^{m-1}$$

$$\vdots$$
$$E(y_2 y_m) = \phi^{m-2} + \phi^m$$

Let $\Omega(\phi)$ be the $(n \times n)$ covariance matrix of y_1, \dots, y_m , i.e.

$$y = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{pmatrix} \sim N \left[\begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}; \Omega(\phi) \right].$$

Therefore, the likelihood function for ϕ is

$$L(\phi; y) = (2\pi)^{-\frac{n}{2}} |\Omega(\phi)|^{-\frac{1}{2}} \exp\left\{-\frac{1}{2} y^T \Omega^{-1}(\phi) y\right\}$$

$$y = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{pmatrix}$$

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Alternatively, one could notice that

$$\begin{aligned} x_1 &= \phi x_0 + \epsilon_1 \\ x_2 &= \phi x_1 + \epsilon_2 \\ &\vdots \\ x_n &= \phi x_{n-1} + \epsilon_n \end{aligned} \quad x_0 = 0$$

could be written as

$$\begin{bmatrix} 1 & 0 & 0 & \dots & 0 & 0 \\ -\phi & 1 & 0 & \dots & 0 & 0 \\ 0 & \phi & 1 & \dots & 0 & 0 \\ \vdots & & & & & \\ 0 & 0 & 0 & \dots & 1 & 0 \\ 0 & 0 & 0 & & -\phi & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} = \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \\ \vdots \\ \epsilon_{n-1} \\ \epsilon_n \end{bmatrix}$$

$$A(\phi) x = \epsilon \quad \epsilon \sim N(0, I_n)$$

$$\rightarrow x = A^{-1}(\phi) \epsilon \sim N(0, \underbrace{A^{-1}(\phi) A^{-1}(\phi)^T}_{B(\phi)})$$

Since $y = x + \epsilon$

$$\Rightarrow y \sim N(0, \underbrace{I_n + B(\phi)}_{\Omega(\phi)})$$



PROBLEM 2

⑥

$$a_t = \begin{cases} u_t & t \text{ even} \\ \frac{u_{t-1}^2 - 1}{\sqrt{2}} & t \text{ odd} \end{cases}$$

$\{u_t\}$ iid $N(0,1)$
 $u_t^2 \sim \chi_1^2$ $E(u_t^2) = 1$
 $V(u_t^2) = 2$

$$V(a_t) = \begin{cases} V(u_t) = 1 & t \text{ even} \\ V\left(\frac{u_{t-1}^2 - 1}{\sqrt{2}}\right) = \frac{1}{2} V(u_{t-1}^2) = \frac{2}{2} = 1 & t \text{ odd} \end{cases}$$

$$E(a_t) = \begin{cases} E(u_t) = 0 & t \text{ even} \\ \frac{E(u_{t-1}^2) - 1}{\sqrt{2}} = \frac{1 - 1}{\sqrt{2}} = 0 & t \text{ odd} \end{cases}$$

$$E(a_t a_{t+h}) = \begin{cases} E\left\{u_t \left(\frac{u_t^2 - 1}{\sqrt{2}}\right)\right\} = \frac{E(u_t^3) - E(u_t)}{\sqrt{2}} = \frac{0 - 0}{\sqrt{2}} = 0 & t \text{ even} \\ E\left(\left(\frac{u_{t-1}^2 - 1}{\sqrt{2}}\right) u_{t+1}\right) = 0 & t \text{ odd} \end{cases}$$

$$E(a_t a_{t+h}) = 0 \quad h = \pm 2, \pm 3, \dots$$

Not only $\{a_t\}$ is weakly stationary, $\{a_t\}$ is also a white noise.