

# Answers to selected problems in

Shumway and Stoffer(2017): Time Series Analysis and Its Applications, 4th ed.

1.6. (a)  $x_t$  is not stationary (the mean is not constant)

1.7.

$$\gamma(h) = \begin{cases} 6\sigma_w^2 & \text{if } h = 0, \\ 4\sigma_w^2 & \text{if } |h| = 1, \\ \sigma_w^2 & \text{if } |h| = 2, \\ 0 & \text{if } |h| \geq 3. \end{cases}$$

$$\rho(h) = \begin{cases} 1 & \text{if } h = 0, \\ 2/3 & \text{if } |h| = 1, \\ 1/6 & \text{if } |h| = 2, \\ 0 & \text{if } |h| \geq 3. \end{cases}$$

1.8. (b)  $\mu_t = t\delta$ ,  $\gamma(s, t) = \sigma_w^2 \min(s, t)$

(c) For  $h \geq 0$ ,  $\gamma(t + h, t) = \sigma_w^2 t$ , which is a function of  $t$  (for stationarity, it should be a function only of  $h$ ).

(d) For large  $t$ ,  $x_t$  is approximately a linear function of  $x_{t-1}$ . (In fact,  $x_t \approx \delta + x_{t-1}$ .)

(e)  $y_t = x_t - x_{t-1} = \delta + w_t$  is stationary

1.13. (a)

$$\rho_y(h) = \begin{cases} 1 & \text{if } h = 0, \\ -\frac{\theta\sigma_w^2}{(1+\theta^2)\sigma_w^2 + \sigma_u^2} & \text{if } |h| = 1, \\ 0 & \text{if } |h| \geq 2. \end{cases}$$

(b)

$$\rho_{xy}(h) = \begin{cases} \frac{\sigma_w}{\sqrt{(1+\theta^2)\sigma_w^2 + \sigma_u^2}} & \text{if } h = 0, \\ -\frac{\theta\sigma_w}{\sqrt{(1+\theta^2)\sigma_w^2 + \sigma_u^2}} & \text{if } h = -1, \\ 0 & \text{otherwise.} \end{cases}$$

(c)  $\rho_{xy}(h)$  is a function of  $h$  only.

1.14. (a)  $E(y_t) = \exp\{\mu_x + \frac{1}{2}\gamma(0)\}$

(b)  $\gamma(h) = e^{2\mu_x + \gamma(0)}\{e^{\gamma(h)} - 1\}$

1.15.  $\mu_t = 0$ ,  $\rho(h) = \begin{cases} \sigma_w^4 & \text{if } h = 0, \\ 0 & \text{otherwise.} \end{cases}$  Stationary.

1.17. (a)  $\phi_w(-\theta\lambda_1)\phi_w(\lambda_1 - \theta\lambda_2) \cdots \phi_w(\lambda_{n-1} - \theta\lambda_n)\phi_w(\lambda_n)$

- 2.6. (a)  $\mu_t = \beta_0 + \beta_1 t$  not constant in  $t$   
 (b)  $\mu_t = \beta_1$ ,  $\gamma(t, t) = 2\sigma_w^2$  finite and constant.  

$$\gamma(t+h, t) = \begin{cases} 2\sigma_w^2 & \text{if } h = 0, \\ -\sigma_w^2 & \text{if } |h| = 1, \\ 0 & \text{if } |h| \geq 2, \end{cases}$$
 function only of  $h$ .  
 (c)  $\mu_t = \beta_1$ ,  $\gamma(t, t) = 2\gamma_y(0) - 2\gamma_y(1)$  finite and constant.  
 $\gamma(t+h, t) = 2\gamma_y(h) - \gamma_y(h+1) - \gamma_y(h-1)$ , function only of  $h$ .
- 2.7.  $\mu_t = \delta$  constant in  $t$ .  
 $\gamma(t, t) = \sigma_w^2 + 2\gamma_y(0) - 2\gamma_y(1)$ , finite and constant.  
 $\gamma(t+h, t) = \sigma_w^2 I\{h = 0\} + 2\gamma_y(h) - \gamma_y(h+1) - \gamma_y(h-1)$ , function only of  $h$ .  
 ( $I\{h = 0\} = 1$  if  $h = 0$  and 0 otherwise)
- 3.1.  $\rho_x(1) = \frac{\theta}{1+\theta^2}$  is maximized at  $\theta = 1$  (maximum value  $1/2$ ) and minimized at  $\theta = -1$ .
- 3.2. (a)  $\mu_t = 0$ ,  $\text{var}(x_t) = \sigma_w^2 \frac{1-\phi^{2t}}{1-\phi^2}$ . Not stationary since the variance is a function of  $t$ .  
 (c) As  $t \rightarrow \infty$ ,  $\text{var}(x_t) \rightarrow \sigma_w^2 \frac{1}{1-\phi^2}$ , and so,  $\text{corr}(x_t, x_{t-h}) \rightarrow \phi^h$ .  
 (d) Use a “burn-in” sample.  
 (e) Yes, because here  $\text{var}(x_t) = \sigma_w^2 \frac{1}{1-\phi^2}$  and  $\text{corr}(x_t, x_{t-h}) = \phi^h$ .
- 3.4. (a) AR(1) with  $\phi = 0.5$ . Causal and invertible.  
 (b) ARMA(2,1), causal but not invertible.
- 3.6. Roots  $\pm i \frac{1}{\sqrt{0.9}}$
- 3.11. (a)  $\tilde{x}_{n+1} = \sum_{j=1}^{\infty} (-\theta)^j x_{n+1-j}$ , MSE  $\sigma_w^2$ .
- 3.15.  $x_{t+m}^t = \phi^m x_t$
- 3.24. (a)  $\mu_t = \frac{\alpha}{1-\phi}$ ,

$$\gamma(h) = \begin{cases} \sigma_w^2 \frac{1+2\theta\phi+\theta^2}{1-\phi^2} & \text{if } h = 0, \\ \sigma^2 \phi^{h-1} \left( \phi \frac{1+2\theta\phi+\theta^2}{1-\phi^2} + \theta \right) & \text{if } |h| \geq 1. \end{cases}$$

$$\rho(h) = \begin{cases} 1 & \text{if } h = 0, \\ \phi^{h-1} \frac{(\theta+\phi)(1+\theta\phi)}{1+2\theta\phi+\theta^2} & \text{if } |h| \geq 1. \end{cases}$$

Weakly stationary. Also strictly stationary if the  $w_t$  are normal.

- (b) Asymptotically normal with mean  $\frac{\alpha}{1-\phi}$  and variance  $n^{-1} \sigma_w^2 \left( \frac{1+\theta}{1-\phi} \right)^2$ .
- 3.29. (c)  $P_{n+m}^n = \sigma_w^2 \left\{ 1 + \frac{1}{(1-\phi)^2} \left( m-1 - 2 \frac{\phi^2 - \phi^{m+1}}{1-\phi} + \frac{\phi^4 - \phi^{2m+2}}{1-\phi^2} \right) \right\}$
- 3.38. (a) ARIMA(0, 0, 0)  $\times$  (0, 0, 1)<sub>2</sub>  
 (c)  $\tilde{x}_{n+m} = - \sum_{j=1}^{\lfloor m/2 \rfloor} (-\Theta)^j \tilde{x}_{n+m-2j} - \sum_{\lfloor m/2 \rfloor + 1}^{\infty} (-\Theta)^j x_{n+m-2j}$ ,  
 where  $\lfloor a \rfloor$  is the integer part of  $a$ .  

$$P_{n+m}^n = \begin{cases} \sigma_w^2 & \text{if } m = 1, 2, \\ \sigma_w^2 (1 + \Theta^2) & \text{if } m \geq 3. \end{cases}$$

- 4.5. (a) Both have mean zero.  $\gamma_w(h) = 1$  if  $h = 0$  and 0 otherwise.

$$\gamma_x(h) = \begin{cases} 1 + \theta^2 & \text{if } h = 0, \\ -\theta & \text{if } |h| = 1, \\ 0 & \text{if } |h| \geq 2. \end{cases}$$

Both are weakly stationary, because the means and variances are constant, the variances are finite and the autocovariance functions only depend on  $h$ . In fact, they are also stationary since they are normal processes.

(b)  $f_x(\omega) = 1 + \theta^2 - 2\theta \cos(2\pi\omega)$

- 4.18. (a)

$$\gamma_{xy}(h) = \begin{cases} 0 & \text{if } h = 0, \\ 1/2 & \text{if } h = 1, \\ -1/2 & \text{if } h = -1, \\ 0 & \text{if } |h| \geq 2. \end{cases}$$

The cross autocovariance is a function only of  $h$ , hence we have joint stationarity.

- (b)  $f_x(\omega) = 2 - 2 \cos(2\pi\omega)$ ,  $f_y(\omega) = \frac{1}{2} + \frac{1}{2} \cos(2\pi\omega)$ . When one is high, the other one is low.

(c)  $a = 0.2465$ ,  $b = 1.898$ .

4.25. (a)  $\phi/(1 + \phi^2)$

4.28.  $f_y(\omega) = \{6 + 8 \cos(2\pi\omega) + 2 \cos(4\pi\omega)\}^2 \sigma_w^2$

4.30. (b) The resulting spectrum is  $4\{1 - \cos(2\pi\omega)\}\{1 - \cos(24\pi\omega)\}f_x(\omega)$ .

- (c) For  $0 \leq \omega \leq 1/2$ , the filter gives highest weights to frequencies which are odd multiples of  $1/24$ , increasing in  $\omega$ .

4.31. (a)  $f_y(\omega) = \{1 + a^2 - 2a \cos(2\pi\omega)\}^{-1} f_x(\omega)$

6.1. (b)  $\sigma_0^2 = \sigma_1^2 = \sigma_w^2/0.19$

6.2.  $\phi^{|s-t|}(P_t^{t-1} + \sigma_v^2)$