

Solution to Chapter 6 Analytical Exercises

1. The hint is the answer.

2. (a) Let $\sigma_n \equiv \sum_{j=0}^n \psi_j^2$. Then

$$\begin{aligned} \mathbb{E}[(y_{t,m} - y_{t,n})^2] &= \mathbb{E}\left[\left(\sum_{j=n+1}^m \psi_j \varepsilon_{t-j}\right)^2\right] \\ &= \sigma^2 \sum_{j=n+1}^m \psi_j^2 \quad (\text{since } \{\varepsilon_t\} \text{ is white noise}) \\ &= \sigma^2 |\alpha_m - \alpha_n|. \end{aligned}$$

Since $\{\psi_j\}$ is absolutely summable (and hence square summable), $\{\alpha_n\}$ converges. So $|\alpha_m - \alpha_n| \rightarrow 0$ as $m, n \rightarrow \infty$. Therefore, $\mathbb{E}[(y_{t,m} - y_{t,n})^2] \rightarrow 0$ as $m, n \rightarrow \infty$, which means $\{y_{t,n}\}$ converges in mean square in n by (i).

(b) Since $y_{t,n} \rightarrow_{\text{m.s.}} y_t$ as shown in (a), $\mathbb{E}(y_t) = \lim_{n \rightarrow \infty} \mathbb{E}(y_{t,n})$ by (ii). But $\mathbb{E}(y_{t,n}) = 0$.

(c) Since $y_{t,n} - \mu \rightarrow_{\text{m.s.}} y_t - \mu$ and $y_{t-j,n} - \mu \rightarrow_{\text{m.s.}} y_{t-j} - \mu$ as $n \rightarrow \infty$,

$$\mathbb{E}[(y_t - \mu)(y_{t-j} - \mu)] = \lim_{n \rightarrow \infty} \mathbb{E}[(y_{t,n} - \mu)(y_{t-j,n} - \mu)].$$

(d) (reproducing the answer on pp. 441-442 of the book) Since $\{\psi_j\}$ is absolutely summable, $\psi_j \rightarrow 0$ as $j \rightarrow \infty$. So for any j , there exists an $A > 0$ such that $|\psi_{j+k}| \leq A$ for all j, k . So $|\psi_{j+k} \cdot \psi_k| \leq A|\psi_k|$. Since $\{\psi_k\}$ (and hence $\{A\psi_k\}$) is absolutely summable, so is $\{\psi_{j+k} \cdot \psi_k\}$ ($k = 0, 1, 2, \dots$) for any given j . Thus by (i),

$$|\gamma_j| = \sigma^2 \left| \sum_{k=0}^{\infty} \psi_{j+k} \psi_k \right| \leq \sigma^2 \sum_{k=0}^{\infty} |\psi_{j+k} \psi_k| = \sigma^2 \sum_{k=0}^{\infty} |\psi_{j+k}| |\psi_k| < \infty.$$

Now set a_{jk} in (ii) to $|\psi_{j+k}| \cdot |\psi_k|$. Then

$$\sum_{j=0}^{\infty} |a_{jk}| = \sum_{j=0}^{\infty} |\psi_k| |\psi_{j+k}| \leq |\psi_k| \sum_{j=0}^{\infty} |\psi_j| < \infty.$$

Let

$$M \equiv \sum_{j=0}^{\infty} |\psi_j| \quad \text{and} \quad s_k \equiv |\psi_k| \sum_{j=0}^{\infty} |\psi_{j+k}|.$$

Then $\{s_k\}$ is summable because $|s_k| \leq |\psi_k| \cdot M$ and $\{\psi_k\}$ is absolutely summable. Therefore, by (ii),

$$\sum_{j=0}^{\infty} \left(\sum_{k=0}^{\infty} |\psi_{j+k}| \cdot |\psi_k| \right) < \infty.$$

This and the first inequality above mean that $\{\gamma_j\}$ is absolutely summable.

3. (a)

$$\begin{aligned}
\gamma_j &= \text{Cov}(y_{t,n}, y_{t-j,n}) \\
&= \text{Cov}(h_0 x_t + h_1 x_{t-1} + \cdots + h_n x_{t-n}, h_0 x_{t-j} + h_1 x_{t-j-1} + \cdots + h_n x_{t-j-n}) \\
&= \sum_{k=0}^n \sum_{\ell=0}^n h_k h_\ell \text{Cov}(x_{t-k}, x_{t-j-\ell}) \\
&= \sum_{k=0}^n \sum_{\ell=0}^n h_k h_\ell \gamma_{j+\ell-k}^x.
\end{aligned}$$

(b) Since $\{h_j\}$ is absolutely summable, we have $y_{t,n} \rightarrow_{\text{m.s.}} y_t$ as $n \rightarrow \infty$ by Proposition 6.2(a). Then, using the facts (i) and (ii) displayed in Analytical Exercise 2, we can show:

$$\begin{aligned}
\sum_{k=0}^n \sum_{\ell=0}^n h_k h_\ell \gamma_{j+\ell-k}^x &= \text{Cov}(y_{t,n}, y_{t-j,n}) \\
&= \text{E}(y_{t,n} y_{t-j,n}) - \text{E}(y_{t,n}) \text{E}(y_{t-j,n}) \rightarrow \text{E}(y_t y_{t-j}) - \text{E}(y_t) \text{E}(y_{t-j}) = \text{Cov}(y_t, y_{t-j})
\end{aligned}$$

as $n \rightarrow \infty$. That is, $\sum_{k=0}^n \sum_{\ell=0}^n h_k h_\ell \gamma_{j+\ell-k}^x$ converges as $n \rightarrow \infty$, which is the desired result.

4. (a) (8) solves the difference equation $y_j - \phi_1 y_{j-1} - \phi_2 y_{j-2} = 0$ because

$$\begin{aligned}
&y_j - \phi_1 y_{j-1} - \phi_2 y_{j-2} \\
&= (c_{10} \lambda_1^{-j} + c_{20} \lambda_2^{-j}) - \phi_1 (c_{10} \lambda_1^{-j+1} + c_{20} \lambda_2^{-j+1}) - \phi_2 (c_{10} \lambda_1^{-j+2} + c_{20} \lambda_2^{-j+2}) \\
&= c_{10} \lambda_1^{-j} (1 - \phi_1 \lambda_1 - \phi_2 \lambda_1^2) + c_{20} \lambda_2^{-j} (1 - \phi_1 \lambda_2 - \phi_2 \lambda_2^2) \\
&= 0 \quad (\text{since } \lambda_1 \text{ and } \lambda_2 \text{ are the roots of } 1 - \phi_1 z - \phi_2 z^2 = 0).
\end{aligned}$$

Writing down (8) for $j = 0, 1$ gives

$$y_0 = c_{10} + c_{20}, \quad y_1 = c_{10} \lambda_1^{-1} + c_{20} \lambda_2^{-1}.$$

Solve this for (c_{10}, c_{20}) given $(y_0, y_1, \lambda_1, \lambda_2)$.

(b) This should be easy.

(c) For $j \geq J$, we have $j^n \xi^j < b^j$. Define B as

$$B \equiv \max \left\{ \frac{\xi}{b}, \frac{2^n \xi^j}{b^2}, \frac{3^n \xi^3}{b^3}, \dots, \frac{(J-1)^n \xi^{J-1}}{b^{J-1}} \right\}.$$

Then, by construction,

$$B \geq \frac{j^n \xi^j}{b^j} \quad \text{or} \quad j^n \xi^j \leq B b^j$$

for $j = 0, 1, \dots, J-1$. Choose A so that $A > 1$ and $A > B$. Then $j^n \xi^j < b^j < A b^j$ for $j \geq J$ and $j^n \xi^j \leq B b^j < A b^j$ for all $j = 0, 1, \dots, J-1$.

(d) The hint is the answer.

5. (a) Multiply both sides of (6.2.1') by $y_{t-j} - \mu$ and take the expectation of both sides to derive the desired result.

(b) The result follows immediately from the MA representation $y_{t-j} - \mu = \varepsilon_{t-j} + \phi \varepsilon_{t-j-1} + \phi^2 \varepsilon_{t-j-2} + \cdots$.

(c) Immediate from (a) and (b).

(d) Set $j = 1$ in (10) to obtain $\gamma_1 - \rho\gamma_0 = 0$. Combine this with (9) to solve for (γ_0, γ_1) :

$$\gamma_0 = \frac{\sigma^2}{1 - \phi^2}, \quad \gamma_1 = \frac{\sigma^2}{1 - \phi^2}\phi.$$

Then use (10) as the first-order difference equation for $j = 2, 3, \dots$ in γ_j with the initial condition $\gamma_1 = \frac{\sigma^2}{1 - \phi^2}\phi$. This gives: $\gamma_j = \frac{\sigma^2}{1 - \phi^2}\phi^j$, verifying (6.2.5).

6. (a) Should be obvious.

(b) By the definition of mean-square convergence, what needs to be shown is that $E[(x_t - x_{t,n})^2] \rightarrow 0$ as $n \rightarrow \infty$.

$$\begin{aligned} E[(x_t - x_{t,n})^2] &= E[(\phi^n x_{t-n})^2] \quad (\text{since } x_t = x_{t,n} + \phi^n x_{t-n}) \\ &= \phi^{2n} E(x_{t-n}^2) \\ &\rightarrow 0 \quad (\text{since } |\phi| < 1 \text{ and } E(x_{t-n}^2) < \infty). \end{aligned}$$

(c) Should be obvious.

7. (d) By the hint, what needs to be shown is that $(\mathbf{F})^n \boldsymbol{\xi}_{t-n} \rightarrow_{\text{m.s.}} \mathbf{0}$. Let $\mathbf{z}_n \equiv (\mathbf{F})^n \boldsymbol{\xi}_{t-n}$. Contrary to the suggestion of the hint, which is to show the mean-square convergence of the components of \mathbf{z}_n , here we show an equivalent claim (see Review Question 2 to Section 2.1) that $\lim_{n \rightarrow \infty} E(\mathbf{z}'_n \mathbf{z}_n) = 0$.

$$\mathbf{z}'_n \mathbf{z}_n = \text{trace}(\mathbf{z}'_n \mathbf{z}_n) = \text{trace}[\boldsymbol{\xi}'_{t-n} [(\mathbf{F})^n]' [(\mathbf{F})^n] \boldsymbol{\xi}_{t-n}] = \text{trace}\{\boldsymbol{\xi}_{t-n} \boldsymbol{\xi}'_{t-n} [(\mathbf{F})^n]' [(\mathbf{F})^n]\}$$

Since the trace and the expectations operator can be interchanged,

$$E(\mathbf{z}'_n \mathbf{z}_n) = \text{trace}\{E(\boldsymbol{\xi}_{t-n} \boldsymbol{\xi}'_{t-n}) [(\mathbf{F})^n]' [(\mathbf{F})^n]\}.$$

Since $\boldsymbol{\xi}_t$ is covariance-stationary, we have $E(\boldsymbol{\xi}_{t-n} \boldsymbol{\xi}'_{t-n}) = \mathbf{V}$ (the autocovariance matrix). Since all the roots of the characteristic equation are less than one in absolute value, $\mathbf{F}^n = \mathbf{T}(\boldsymbol{\Lambda})^n \mathbf{T}^{-1}$ converges to a zero matrix. We can therefore conclude that $E(\mathbf{z}'_n \mathbf{z}_n) \rightarrow 0$.

(e) ψ_n is the (1,1) element of $\mathbf{T}(\boldsymbol{\Lambda})^n \mathbf{T}^{-1}$.

8. (a)

$$\begin{aligned} E(y_t) &= \frac{1 - \phi^t}{1 - \phi} c + \phi^t E(y_0) \rightarrow \frac{c}{1 - \phi}, \\ \text{Var}(y_t) &= \frac{1 - \phi^{2t}}{1 - \phi^2} \sigma^2 + \phi^{2t} \text{Var}(y_0) \rightarrow \frac{\sigma^2}{1 - \phi^2}, \\ \text{Cov}(y_t, y_{t-j}) &= \phi^j \left[\frac{1 - \phi^{2(t-j)}}{1 - \phi^2} \sigma^2 + \phi^{2(t-j)} \text{Var}(y_0) \right] \rightarrow \phi^j \frac{\sigma^2}{1 - \phi^2}. \end{aligned}$$

(b) This should be easy to verify given the above formulas.

9. (a) The hint is the answer.

(b) Since $\gamma_j \rightarrow 0$, the result proved in (a) implies that $\frac{2}{n} \sum_{j=1}^n |\gamma_j| \rightarrow 0$. Also, $\gamma_0/n \rightarrow 0$. So by the inequality for $\text{Var}(\bar{y})$ shown in the question, $\text{Var}(\bar{y}) \rightarrow 0$.

10. (a) By the hint,

$$\sum_{j=1}^n j a_j \leq \sum_{j=1}^N \left| \sum_{k=j}^n a_k \right| + \sum_{j=N+1}^n \left| \sum_{k=j}^n a_k \right| < NM + (n - N) \frac{\varepsilon}{2}.$$

So

$$\frac{1}{n} \sum_{j=1}^n j a_j < \frac{NM}{n} + \frac{n - N}{n} \frac{\varepsilon}{2} < \frac{NM}{n} + \frac{\varepsilon}{2}.$$

By taking n large enough, NM/n can be made less than $\varepsilon/2$.

(b) From (6.5.2),

$$\text{Var}(\sqrt{n} \bar{y}) = \gamma_0 + 2 \sum_{j=1}^{n-1} \left(1 - \frac{j}{n}\right) \gamma_j = \left[\gamma_0 + 2 \sum_{j=1}^{n-1} \gamma_j \right] - \frac{2}{n} \sum_{j=1}^{n-1} j \gamma_j.$$

The term in brackets converges to $\sum_{j=-\infty}^{\infty} \gamma_j$ if $\{\gamma_j\}$ is summable. (a) has shown that the last term converges to zero if $\{\gamma_j\}$ is summable.