Chapter 1

ARMA and Nonstationary Times Series - Exercises

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1.1 ARMA

Exercise 14.15 A Gaussian AR model is an autoregression with i.i.d. $N\left(0,\sigma^2\right)$ errors. Consider the Gaussian AR(1) model

$$Y_t = \alpha_0 + \alpha_1 Y_{t-1} + e_t$$

$$e_t \sim N(0, \sigma^2)$$

with $|\alpha_1| < 1$. Show that the marginal distribution of Y_t is also normal:

$$Y_t \sim N\left(\frac{\alpha_0}{1-\alpha_1}, \frac{\sigma^2}{1-\alpha_1^2}\right).$$

Hint: Use the MA representation of Y_t .

Solution. 14.15

$$\begin{split} Y_t &= \alpha_0 + \alpha_1 Y_{t-1} + e_t \\ &= \alpha_0 + \alpha_1 \left(\alpha_0 + \alpha_1 Y_{t-2} + e_{t-1} \right) + e_t \\ &= \alpha_0 \left(1 + \alpha_1 \right) + \alpha_1^2 Y_{t-2} + \alpha_1 e_{t-1} + e_t \\ &= \alpha_0 \left(1 + \alpha_1 + \alpha_1^2 \right) + \alpha_1^3 Y_{t-3} + \alpha_1^2 e_{t-2} + \alpha_1 e_{t-1} + e_t \\ &= \cdots \\ &= \sum_{i=0}^{\infty} \alpha_1^i e_{t-i} + \frac{\alpha_0}{1 - \alpha_1} \end{split}$$

The mean of Y_t is:

$$E[Y_t] = E\left[\sum_{i=0}^{\infty} \alpha_1^i e_{t-i} + \frac{\alpha_0}{1 - \alpha_1}\right]$$
$$= \frac{\alpha_0}{1 - \alpha_1} + \sum_{i=0}^{\infty} \alpha_1^i E[e_{t-i}]$$
$$= \frac{\alpha_0}{1 - \alpha_1}$$

The variance of Y_t is:

$$var(Y_t) = var\left(\sum_{i=0}^{\infty} \alpha_1^i e_{t-i} + \frac{\alpha_0}{1 - \alpha_1}\right)$$

$$= var\left(\sum_{i=0}^{\infty} \alpha_1^i e_{t-i}\right)$$

$$= \sum_{i=0}^{\infty} (\alpha_1^i)^2 var(e_{t-i})$$

$$= \sigma^2 \sum_{i=0}^{\infty} (\alpha_1^i)^2$$

$$= \frac{\sigma^2}{1 - \alpha_1^2}$$

Thus we conclude

$$Y_t \sim N\left(\frac{\alpha_0}{1-\alpha_1}, \frac{\sigma^2}{1-\alpha_1^2}\right)$$

1.2 Nonstationary Times Series

Exercise 16.1 Take $S_t = S_{t-1} + e_t$ with $S_0 = 0$ and e_t i.i.d. $N(0, \sigma^2)$.

- (a) Calculate $E[S_t]$ and $var[S_t]$.
- (b) Set $Y_t = (S_t E[S_t]) / \sqrt{var[S_t]}$. By construction $E[Y_t] = 0$ and $var[Y_t] = 1$. Is Y_t stationary?
- (c) Find the asymptotic distribution of $Y_{\lfloor nr \rfloor}$ for $r \in [\delta, 1]$.

Solution. 16.1

(a)

$$E[S_t] = E[S_{t-1} + e_t]$$

$$= E[e_1 + e_2 + \dots + e_t]$$

$$= E\left[\sum_{i=1}^t e_i\right]$$

$$= \sum_{i=1}^t E[e_i] = 0$$

$$var(S_t) = var\left(\sum_{i=1}^t e_i\right)$$
$$= \sum_{i=1}^t var(e_i)$$
$$= t\sigma^2$$

(b) Notice that

$$Y_t = \frac{S_t - E\left[S_t\right]}{\sqrt{var\left[S_t\right]}} = \frac{S_t}{\sqrt{t}\sigma} = \frac{\sum_{i=1}^t e_i}{\sqrt{t}\sigma} = \frac{1}{\sigma} \frac{1}{\sqrt{t}} \sum_{i=1}^t e_i$$

If Y_t is stationary, its covariance must not depend on t. However,

$$\begin{aligned} Cov\left(Y_{t},Y_{t+k}\right) &= E\left[\left(Y_{t} - E\left[Y_{t}\right]\right)\left(Y_{t+k} - E\left[Y_{t+k}\right]\right)\right] \\ &= E\left[Y_{t}Y_{t+k}\right] \\ &= E\left[\frac{1}{\sigma}\frac{1}{\sqrt{t}}\sum_{i=1}^{t}e_{i}\frac{1}{\sigma}\frac{1}{\sqrt{t+k}}\sum_{i=1}^{t+k}e_{i}\right] \\ &= \frac{1}{\sigma^{2}\sqrt{t\left(t+k\right)}}E\left[\left(e_{1} + \cdots + e_{t}\right)\left(e_{1} + \cdots + e_{t} + \cdots + e_{t+k}\right)\right] \\ &= \frac{1}{\sigma^{2}\sqrt{t\left(t+k\right)}}\sum_{i=1}^{t}E\left[e_{i}^{2}\right] \\ &= \sqrt{\frac{t}{t+k}}. \end{aligned}$$

Therefore, Y_t is not stationary.

(c)

$$Y_{\lfloor nr \rfloor} = \frac{S_{\lfloor nr \rfloor} - E\left[S_{\lfloor nr \rfloor}\right]}{\sqrt{var\left[S_{\lfloor nr \rfloor}\right]}} = \frac{S_{\lfloor nr \rfloor}}{\sqrt{\lfloor nr \rfloor}\sigma} = \frac{\sum_{i=1}^{\lfloor nr \rfloor} e_i}{\sqrt{\lfloor nr \rfloor}\sigma} \sim N\left(0,1\right)$$