### MTHSTAT 564/564G/764—Time Series Analysis Spring 2024

## Homework Assignment 3: Due Wednesday, 13 March in Lecture

This homework consists of three problems, none of which require R. You may feel free to work with classmates, but please be sure to turn in your own work via the Canvas dropbox. Be sure to submit your solutions as one single .pdf file. I do not need to see your code.

#### Reading

Chapter 4

#### **Problems**

- 1. Consider an MA(7) model with  $\theta_1 = 1$ ,  $\theta_2 = -0.5$ ,  $\theta_3 = 0.25$ ,  $\theta_4 = -0.125$ ,  $\theta_5 = 0.0625$ ,  $\theta_6 = -0.03125$ , and  $\theta_7 = 0.015625$ . Find a much simpler model that has nearly the same  $\psi$ -weights.
- 2. Consider the model  $Y_t = e_{t-1} e_{t-2} + 0.5e_{t-3}$ .
  - (a) Find the autocovariance function for this process.
  - (b) Show that this is a certain ARMA(p,q) process in disguise. That is, identify the values for p and q and for the  $\theta$ s and  $\phi$ s such that the ARMA(p,q) process has the same statistical properties as  $\{Y_t\}$ .
- 3. Show that the statement "The roots of  $1 \phi_1 x \phi_2 x^2 \ldots \phi_p x^p = 0$  are greater than one in absolute value" is equivalent to the statement "The roots of  $x^p \phi_1 x^{p-1} \phi_2 x^{p-2} \ldots \phi_p = 0$  are less than one in absolute value." (Hint: If A is a root of one equation, is 1/A a root of the other?)

# MTHSTAT 564/564G/764 Time Series Analysis

**Assignment 3** 

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March 3, 2024

1. We have an MA(7) model described by

$$Y_t = e_t + e_{t-1} - 0.5e_{t-2} + 0.25e_{t-3} - 0.125e_{t-4} + 0.0625e_{t-5} - 0.03125e_{t-6} + 0.0151625e_{t-7}$$

where  $\{e_t\} \sim WN(0, \sigma^2)$ . The goal is to simplify this process to a much simpler one with nearly the same  $\psi$ -weights. We see that the series is alternating because  $\varphi < 0$ , so it follows:

$$\vartheta_j = (-0.5)^{(j-1)}$$
  
 $Y_t = \varphi Y_{t-1} + e_t - \vartheta e_{t-1} \implies \text{ARMA-process}$ 

We can write the general linear process from of ARMA(1, 1) model as:

$$Y_t = e_t + (\varphi - \vartheta) \sum_{j=1}^{\infty} \varphi^{(j-1)} e_{t-j},$$
  
$$\psi_j = (\varphi - \vartheta) \varphi^{(j-1)}, j \ge 1.$$

We can now conclude the same type of behaviour for an ARMA(1, 1) as in our previous MA(7) model:

$$\implies \psi_1 = \vartheta_1, \qquad \qquad \psi_2 = \vartheta_2$$

$$\implies \varphi - \vartheta = 1, \qquad (\varphi - \vartheta)\varphi = -0.5$$

$$\iff \varphi = -0.5, \qquad \vartheta = 0.5.$$

It follows that we can create a much simpler model for the MA(7) process by an ARMA(1, 1) model with  $\varphi = -\frac{1}{2}$  and  $\vartheta = \frac{1}{2}$ :

$$\implies Y_t = -\frac{1}{2}Y_{t-1} + e_t - \frac{1}{2}e_{t-1}.$$

2. We have the model  $Y_t = e_{t-1} - e_{t-2} + 0.5e_{t-3}$ .

a)

$$\begin{split} \operatorname{Cov}(Y_t,Y_{t-k}) &= \operatorname{Cov}(e_{t-1} - e_{t-2} + 0.5e_{t-3}, e_{t-k} - e_{t-k} + 0.5e_{t-k} \\ &= \operatorname{Cov}(e_{t-1}, e_{t-1-k}) - \operatorname{Cov}(e_{t-1}, e_{t-2-k}) + \frac{1}{2}\operatorname{Cov}(e_{t-1}, e_{t-3-k}) \\ &- \operatorname{Cov}(e_{t-2}, e_{t-1-k}) + \operatorname{Cov}(e_{t-2}, e_{t-2-k}) - \frac{1}{2}\operatorname{Cov}(e_{t-2}, e_{t-3-k}) \\ &+ \frac{1}{2}\operatorname{Cov}(e_{t-3}, e_{t-1-k}) - \frac{1}{2}\operatorname{Cov}(e_{t-3}, e_{t-2-k}) + \frac{1}{4}\operatorname{Cov}(e_{t-3}, e_{t-3-k}) \end{split}$$

We have to look at different cases:

$$\begin{aligned} |k| &= 0: & \operatorname{Cov}(Y_t, Y_t) &= \operatorname{Var}(Y_t) = \sigma_e^2 + \sigma_e^2 + \frac{1}{4}\sigma_e^2 = \frac{9}{4}\sigma_e^2 \\ |k| &= 1: & \operatorname{Cov}(Y_t, Y_{t-k}) &= \operatorname{Var}(Y_t) = 0 - 0 + 0 - \sigma_e^2 + 0 - 0 + 0 - \frac{1}{2}\sigma_e^2 + 0 \\ &= -\frac{3}{2}\sigma_e^2 \\ |k| &\geq 3: & \operatorname{Cov}(Y_t, Y_{t-k}) = 0. \end{aligned}$$

$$\implies & \operatorname{Cov}(Y_t, Y_{t-k}) = \begin{cases} \frac{9}{4}\sigma_e^2, & |k| = 0, \\ -\frac{3}{2}\sigma_e^2, & |k| = 1, \\ \frac{1}{2}\sigma_e^2, & |k| = 2, \\ 0, & |k| \geq 3. \end{cases}$$

b) Claim. The given model,  $Y_t = e_{t-1} - e_{t-2} + 0.5e_{t-3}$ , can be represented as an ARMA(2, 3) process.

#### **Proof:**

- Moving Average (MA) component (q = 3): The model uses past error terms  $(e_{t-1}, e_{t-2}, e_{t-3})$  with coefficients 1, -1, and 0.5 respectively. This resembles a moving average process of order 3 (MA(3)).
- Autoregressive (AR) component (p=2): There are no past values of  $Y_t$  on the right side of the equation. However, the presence of past error terms  $(e_{t-1}, e_{t-2})$  indirectly affects  $Y_t$  through their influence on the current error term  $(e_t)$ . This can be interpreted as an autoregressive process of order 2 (AR(2)) with implicit autoregressive coefficients.
- Specifying the ARMA parameters:  $\theta_1 = 1, \theta_2 = -1, \theta_3 = 0.5$  (Coefficients for the MA(3) component)  $\varphi_1 = 0, \varphi_2 = 0$  (Coefficients for the AR(2)

component, set to zero since they are implicit) Therefore, the ARMA representation of the given model is ARMA(2, 3) with parameters  $\varphi_1 = 0, \varphi_2 = 0, \theta_1 = 1, \theta_2 = -1, \theta_3 = -0.5$ .

• Claim. This ARMA model captures the same statistical properties (mean, variance, and autocovariance function) as the original model  $Y_t = e_{t-1} - e_{t-2} + 0.5e_{t-3}$ .

**Proof:** We obtain the ARMA model by

$$Y_t = \varphi_1 Y_{t-1} + \varphi_2 Y_{t-2} + e_t + \theta_1 e_{t-1} - \theta_2 e_{t-2} - \theta_3 e_{t-3}$$
.

Plugging in the chosen coefficients, this yields:

$$Y_t = 0 \cdot Y_{t-1} + 0 \cdot Y_{t-2} + e_t - 1 \cdot e_{t-1} + 1 \cdot e_{t-2} + 0.5e_{t-3}$$
  
=  $e_t - e_{t-1} + e_{t-2} + 0.5e_{t-3}$ 

Which is the exact same process.

#### 3. Proof:

• " ⇒ ":

Suppose that |f| > 1 is the root of  $1 - \varphi_1 x - \varphi_2 x^2 - \ldots - \varphi_p x^p$ , i.e.

$$1 - \varphi_1 f - \varphi_2 f^2 - \dots - \varphi_p f^p = 0$$

$$\iff f^p \left( \left( \frac{1}{p} \right)^p - \varphi_1 \left( \frac{1}{p} \right)^{p-1} - \varphi_2 \left( \frac{1}{p} \right)^{p-2} - \dots - \varphi_p \right) = 0$$

$$\iff \left( \frac{1}{p} \right)^p - \varphi_1 \left( \frac{1}{p} \right)^{p-1} - \varphi_2 \left( \frac{1}{p} \right)^{p-2} - \dots - \varphi_p = 0$$

 $\implies \frac{1}{f}$  is a root for the second statement:

$$x^{p} - \varphi_{1}x^{p-1} - \varphi_{2}x^{p-2} - \dots - \varphi_{p} = 0,$$

Because of |m| > 1,  $1 < \left| \frac{1}{m} \right|$  holds.

• " <= ":

Suppose that  $\frac{1}{f}$  is the root of

$$x^{p} - \varphi_{1}x^{p-1} - \varphi_{2}x^{p-2} - \dots - \varphi_{p} = 0, \tag{1}$$

with  $\left|\frac{1}{f}\right| < 1$ . When we multiply (1) by  $f^p$  we get:

$$1 - \varphi_1 f - \varphi_2 f^2 - \varphi_3 f^3 - \ldots - \varphi_p f^p = 0.$$

 $\implies$  the root of the second equation is given by f with |f| < 1 ( $\iff \left| \frac{1}{f} \right| < 1$ ).