

## 2.5 hr written exam

Answer ALL questions. Section A carries 40% of the total marks and Section B carries 60%. The relative weights attached to each question are as follows: A1 (8), A2 (9), A3 (9), A4 (6), A5 (8), B1 (30), B2 (30). The numbers in square brackets indicate the relative weight attached to each part of the question.

- Unless otherwise indicated, in all questions  $\{\epsilon_t\}$  denotes a sequence of uncorrelated zero-mean random variables with constant variance  $\sigma^2$ , i.e.  $\{\epsilon_t\} \sim WN(0, \sigma^2)$ , where  $WN(0, \sigma^2)$  denotes white noise.
- The 97.5<sup>th</sup> percentile point of the standard Normal distribution is 1.96.
- The transpose of a vector or matrix  $\mathbf{A}$  is denoted by  $\mathbf{A}^\top$ .
- Wold’s formula gives the autocovariance function of the stationary process  $X_t = \sum_{j=0}^{\infty} b_j \epsilon_{t-j}$  as  $\gamma(k) = \sigma^2 \sum_{j=0}^{\infty} b_j b_{j+k}$ .
- If  $|\lambda| < 1$ , then  $\sum_{j=0}^{\infty} \lambda^j = \frac{1}{1-\lambda}$ .

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## Section A

- A1** (a) Give an analytical mathematical expression for each of the following models using the backshift operator (for example, model AR(1) should be written as  $(1 - \phi B)X_t = \epsilon_t$ ):

i) ARMA(1, 2) [1]

ii) SARMA(0, 1)  $\times$  (1, 0)<sub>4</sub> [2]

iii) SARIMA(0, 1, 1)  $\times$  (1, 1, 1)<sub>6</sub> [2]

- (b) We are given the following models:

i)  $X_t = -\frac{1}{16}X_{t-4} + \epsilon_t + \epsilon_{t-2}$

ii)  $X_t = \frac{1}{2} + X_{t-1} + \epsilon_t - \frac{3}{2}\epsilon_{t-4}$

For each of i), ii), determine whether the model is stationary or not, and whether it is invertible or not. Explain your answers. [3]

(a)

- (b) i) Stationary; not invertible.  
ii) Not stationary; not invertible.

- A2** You are given the following stationary ARMA(1, 1) model:

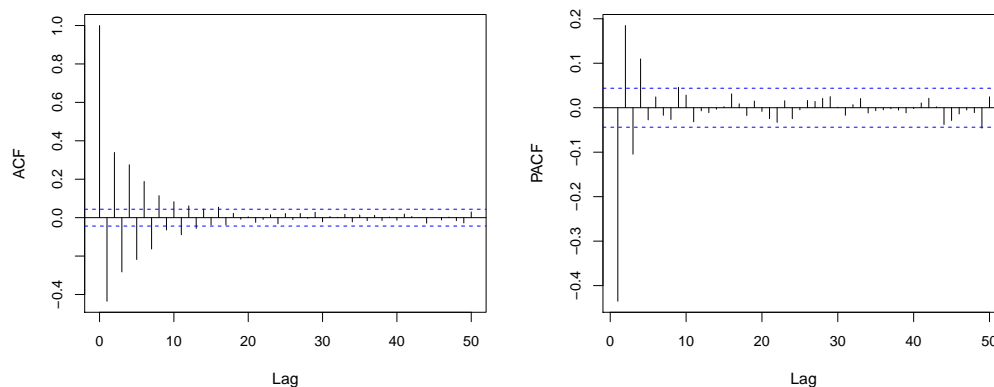
$$X_t = \phi X_{t-1} + \epsilon_t - \theta \epsilon_{t-1}, \quad |\phi| < 1,$$

- (a) Re-express this model as a MA( $\infty$ ) one. [3]

- (b) Using your answer in (a) and Wold's formula or otherwise, calculate  $\gamma(4)$ . What is the exact value of  $\gamma(4)$  when  $\theta = \phi$ ? [3]

- (c) You are given that  $\phi > 0$ . Are the following ACF and PACF plots consistent with this ARMA model? Explain your answer. [3]

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- (a)
- (b) When  $\phi = \theta$ , we have that  $\gamma(4) = 0$ .
- (c) No, they are not.

**A3** Consider the following MA(2)-model:

$$Y_t = \epsilon_t - \theta\epsilon_{t-1} + \frac{1}{2}\epsilon_{t-2}$$

Data from this model have provided the following estimates of the autocorrelation function:

Lag, $k$	1	2	3	4	5
$r(k)$	0.5	0.45	0.00	-0.04	0.01

and the estimate of the autocovariance  $\hat{\gamma}(1) = 0.75$ .

- (a) Use the above sample information to estimate  $\theta$  with the *method of moments*. Note that the model is required to be invertible. [3]
- (b) Find now an estimate of the variance of the white noise  $\sigma^2$ , again with the method of moments. [3]
- (c) Find the lag 1 and 2 sample partial autocorrelations. [3]

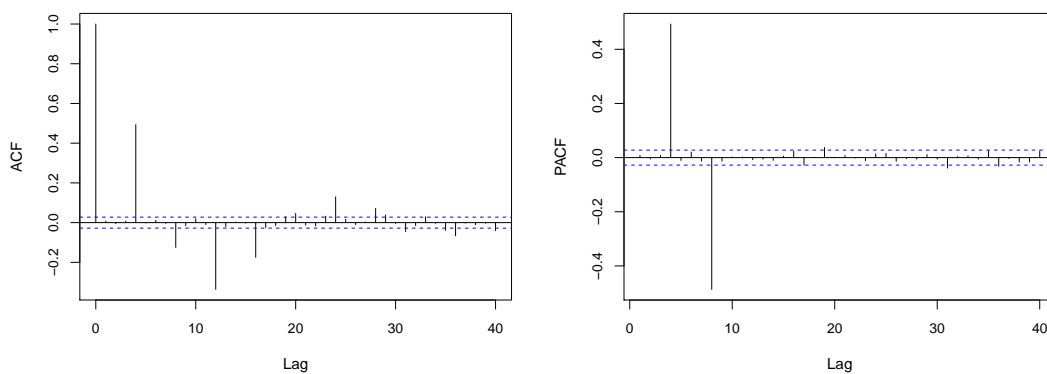
- (a)  $\hat{\theta} = -\frac{1}{2}$

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(b)  $\hat{\sigma}^2 = 1$

(c)  $\hat{\phi}_{11} = 0.5; \quad \hat{\phi}_{22} = 0.27$

**A4** A statistician wants to fit an ARMA model to a given time series. The empirical autocorrelation and partial autocorrelation functions of the time series are as shown in the following plots:



- a) Based on the above figures, suggest an ARMA model for this time series. Explain why both plots are consistent with the model you have suggested. [4]
- (b) Looking at these plots, is there any evidence for existence of complex roots in any of the characteristic polynomials of the model? [2]

(a)  $\text{SAR}(2)_4$

(b) Yes.

**A5** (a) Consider the following stationary AR(1) model:

$$Y_t = 1 + \phi Y_{t-1} + \epsilon_t, \quad |\phi| < 1$$

Find the variance of  $Y_t$ . [2]

- (b) Consider an arbitrary stationary process  $(X_t)$ , with mean  $\mu_X$ , variance  $\sigma_X^2$  and autocorrelation function  $\rho_X(k)$ ,  $k \geq 0$ . Given the observations  $X_1, X_2, \dots, X_T$ , the sample mean  $\bar{X} = \sum_{i=1}^T X_i / T$  is obtained as an estimate of  $\mu_X$ .

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- i) Show that the variance of the estimator  $\bar{X}$  is equal to:

$$\frac{\sigma_X^2}{T} \cdot \left( 1 + 2 \sum_{k=1}^{T-1} \left( 1 - \frac{k}{T} \right) \rho_X(k) \right) \quad [4]$$

- ii) How does the variance of the above formula contrast with the one corresponding to *independent* observations? [2]

(a)  $\sigma^2/(1 - \phi^2)$ .

(b) i)

ii) The difference is term  $\frac{\sigma_X^2}{T} \cdot \left( 2 \sum_{k=1}^{T-1} \left( 1 - \frac{k}{T} \right) \rho_X(k) \right)$

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- B1** In the UK, many sea birds rely on small fish called sandeels for their food. Let  $M_t$  and  $N_t$  denote respectively the population sizes of sea birds and sandeels (in tens of thousands), at the beginning of year  $t$ , and let  $Y_t$  denote a corresponding survey-based estimate of the sea bird population. Suppose that these quantities are related via the equations:

$$\begin{aligned} Y_t &= M_t + \epsilon_t \\ M_t &= -7 + 0.7M_{t-1} + 0.1N_{t-1} + h_t \\ N_t &= 70 - 2M_{t-1} + 0.5N_{t-1} + z_t, \end{aligned}$$

where  $(\epsilon_t)$ ,  $(h_t)$  and  $(z_t)$  are independent white noise sequences with variances  $\sigma_\epsilon^2$ ,  $\sigma_h^2$  and  $\sigma_z^2$  respectively.

- (a) You are given that both  $M_t$  and  $N_t$  are stationary processes. Calculate the long term average population sizes of the sea birds and the sandeels. [4]
- (b) Define  $\mathbf{S}_t = (1, M_t, N_t)^\top$ . Show that the system can be described by the two equations

$$\begin{aligned} Y_t &= \mathbf{B}^\top \mathbf{S}_t + \epsilon_t \\ \mathbf{S}_t &= \mathbf{C} \mathbf{S}_{t-1} + \mathbf{H}_t, \end{aligned}$$

where  $\mathbf{B}$  is a vector,  $\mathbf{C}$  is a matrix and  $(\mathbf{H}_t)$  is an uncorrelated sequence of random vectors with mean  $\mathbf{0}$  and variance-covariance matrix  $\mathbf{V}$ . Give the values of  $\mathbf{B}$ ,  $\mathbf{C}$  and  $\mathbf{V}$ . [5]

- (c) The Kalman Filter specifies that, if  $\hat{\mathbf{S}}_{t|t-\ell}$  is the best estimator of  $\mathbf{S}_t$  based on the observed  $Y$ 's up to time  $t - \ell$ , and  $\mathbf{P}_{t|t-\ell}$  is the covariance matrix of the associated forecast error, then we have that  $\hat{\mathbf{S}}_{t|t-1} = \mathbf{C} \hat{\mathbf{S}}_{t-1|t-1}$  and  $\mathbf{P}_{t|t-1} = \mathbf{C} \mathbf{P}_{t-1|t-1} \mathbf{C}^\top + \mathbf{V}$ .

Suppose that at the start of year zero we have  $\hat{\mathbf{S}}_{0|0} = (1, 8, 60)^\top$ , with  $\mathbf{P}_{0|0} = 2\mathbf{V}$ . Suppose also that  $\sigma_\epsilon^2 = 1$ ,  $\sigma_h^2 = 1$  and  $\sigma_z^2 = 25$ . Use the Kalman Filter to forecast the population sizes of both sea birds and sandeels one year later, and give the covariance matrix of the associated forecast errors. [5]

- (d) The Kalman Filter also gives that

$$\begin{aligned} \hat{\mathbf{S}}_{t|t} &= \hat{\mathbf{S}}_{t|t-1} + \mathbf{P}_{t|t-1} \mathbf{B} (Y_t - \mathbf{B}^\top \hat{\mathbf{S}}_{t|t-1}) / f_t \\ \text{and } \mathbf{P}_{t|t} &= \mathbf{P}_{t|t-1} - \mathbf{P}_{t|t-1} \mathbf{B} \mathbf{B}^\top \mathbf{P}_{t|t-1} / f_t, \\ \text{where } f_t &= \mathbf{B}^\top \mathbf{P}_{t|t-1} \mathbf{B} + \sigma_\epsilon^2. \end{aligned}$$

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Subsequently, the survey-based estimate of sea bird population becomes available:  $Y_1 = 5$ . You are also given that  $f_1 = 3.48$  and  $\mathbf{P}_{1|0} \mathbf{B} = (0, 2.48, -0.3)^\top$ .

- i) Use this information to calculate a new estimate of  $\mathbf{S}_1$ , along with its error covariance matrix. [5]
- ii) Give also an estimate of  $M_1 + 0.1N_1$  together with its variance. [3]

- (e) Notice that  $N_t$  can be written as  $(1 - 0.5B)^{-1} [70 - 2M_{t-1} + z_t]$ , where  $B$  is the backshift operator. *Without attempting any series expansions*, use this to show that  $M_t$  can be written as

$$M_t = 3.5 + 1.2M_{t-1} - 0.55M_{t-2} + \lambda_t + \xi_{t-1},$$

where  $(\lambda_t)$  and  $(\xi_t)$  are white noise sequences that can be constructed from  $(h_t)$  and  $(z_t)$ .

- i) Give the variances of  $(\lambda_t)$  and  $(\xi_t)$  in terms of  $\sigma_h^2$  and  $\sigma_z^2$ . [5]
- ii) Which ARMA model does the  $M_t$  process correspond to? Explain your answer. [3]

- (a)  $\mu_M = 10, \mu_N = 100$ .

(b)

$$\mathbf{B} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \quad \mathbf{C} = \begin{pmatrix} 1 & 0 & 0 \\ -7 & 0.7 & 0.1 \\ 70 & -2 & 0.5 \end{pmatrix}$$

and

$$\mathbf{V} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & \sigma_h^2 & 0 \\ 0 & 0 & \sigma_z^2 \end{pmatrix}$$

(c)

$$\hat{\mathbf{S}}_{1|0} = \begin{pmatrix} 1 \\ 4.6 \\ 84 \end{pmatrix}$$

$$\mathbf{P}_{1|0} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 2.48 & -0.3 \\ 0 & -0.3 & 45.5 \end{pmatrix}$$

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(d) i)

$$\hat{\mathbf{S}}_{1|1} = \begin{pmatrix} 1 \\ 4.89 \\ 83.97 \end{pmatrix}$$

$$\mathbf{P}_{1|1} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0.71 & -0.09 \\ 0 & -0.09 & 45.47 \end{pmatrix}$$

ii) Estimate of  $M_1 + 0.1N_1$  is 13.29; variance is 1.15.

(e) i)

ii) ARMA(2, 1)

**B2** The monthly sales of a retail shop in thousands of pounds are modeled according to a SARMA(0, 2)  $\times$  (1, 0)<sub>4</sub> process of the form

$$(1 - \frac{1}{4}B^4)X_t = \frac{150}{2} + (1 - \frac{1}{2}B - \frac{1}{4}B^2)\epsilon_t$$

where  $\epsilon_t \sim WN(0, \sigma^2)$  with  $\sigma^2 = 2$ .

- (a) Calculate the long term average sales  $\mathbb{E}[X_t]$  and the long term variance  $\text{Var}[X_t]$ . [4]
- (b) The table below shows the observations for the last 6 months of available data, and all corresponding estimated residuals except for the one for the last day:

Month	Sales	Residuals ( $e_t$ )
1	99	-1
2	103	4
3	97	-3
4	100	4
5	96	-2
6	104	

- i) Calculate the missing residual at time  $T = 6$  and show that it is equal to  $\frac{13}{4}$ . [3]
- ii) Produce forecasts for times  $T = 7$  and  $T = 8$ . [4]
- iii) Give a general expression for the forecasts at times  $t = 9 + 4k$ , for all  $k \geq 0$ . [3]

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- (c) Derive in detail the variance for the error of the estimate for time  $T = 8$  and calculate a 95% prediction interval. [5]
- (d) The practitioner carrying out this study used the BIC criterion to choose an appropriate model. For the  $\text{SARMA}(0, 2) \times (1, 0)_4$  model BIC was found to be 205, whereas for  $\text{SARMA}(0, 3) \times (1, 0)_4$  it was found to be 207. The size of the data was 240. You are reminded of the following expressions for BIC and AIC from the textbook:

$$\text{BIC} : T \log(s_e^2) + k \log T$$

$$\text{AIC} : T \log(s_e^2) + 2k$$

- i) Define all involved terms in the above given expressions for AIC and BIC. Did the practitioner make the correct decision to choose  $\text{SARMA}(0, 2) \times (1, 0)_4$  over  $\text{SARMA}(0, 3) \times (1, 0)_4$  according to the BIC criterion? [3]
- ii) Would the decision be the same if the AIC criterion was used instead? Explain your answer. [4]
- (e) Assume that in some other scenario, one claims that the  $\epsilon_t$  process appearing in the definition of the model is not white noise, but it is in fact such that  $\nabla \epsilon_t$  is a stationary  $\text{ARMA}(1, 1)$  process. Determine which model would the  $(X_t)$  process correspond to in such a case. [4]

(a)  $\mu = 100$ ;  $\text{var} = 42/15$ .

- (b) i)  
ii)

$$\hat{X}_{T+2} = 99.1875$$

- iii)

$$\hat{X}_{9+4k} = 100 - \frac{1}{4^k}$$

- (c)

$$\text{var} = \frac{5}{2}$$

- (d) i) Practitioner made a correct choice.  
ii) More complex model would be preferred.
- (e)  $X_t$  is now  $\text{SARIMA}(1, 1, 3) \times (1, 0, 0)_4$ .

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