

Name: \_\_\_\_\_

This exam has 5 questions, each of which is worth 12 points. Good luck!

[1] Consider the process

$$X_t = \phi X_{t-12} + Z_t,$$

where  $0 < |\phi| < 1$  and  $\{Z_t\} \sim \text{WN}(0, \sigma^2)$  (the above might serve as a model for a time series of monthly values for which the current month is related to the same month of the previous year, but not to any of the intervening months).

(a) Use a ‘telescoping’ argument to determine the weights  $\psi_j$  needed to represent the process as an infinite moving average process, i.e., as

$$X_t = \sum_{j=0}^{\infty} \psi_j Z_{t-j}.$$

- (b) Using the  $\psi_j$  weights derived in part (a), verify that  $X_t = \sum_{j=0}^{\infty} \psi_j Z_{t-j}$  is a solution to the equation  $X_t - \phi X_{t-12} = Z_t$ .

(c) Determine the ACVF for  $\{X_t\}$  (an item for your crib sheet:  $\sum_{j=0}^{\infty} x^j = \frac{1}{1-x}$ ).

(d) Determine the PACF for  $\{X_t\}$ . Hint: regard  $\{X_t\}$  as an AR( $p$ ) process.

- [2] The invertibility condition for the MA(1) process  $X_t = Z_t + \theta Z_{t-1}$  is  $|\theta| < 1$ , where, as usual,  $\{Z_t\} \sim \text{WN}(0, \sigma^2)$ . Determine the invertibility condition for  $X_t = Z_t + \theta Z_{t-q}$ , where  $q \geq 2$ . Hint: ‘telescoping’ might be helpful.

[3] Suppose that  $\{Y_t\}$  is a stationary process with mean 0 and ACVF  $\{\gamma_Y(h)\}$ . Define

$$X_t = \sum_{j=-J}^J \psi_j Y_{t-j},$$

where the  $2J + 1$  real-valued constants  $\psi_j$  are all finite.

- (a) Argue that  $\{X_t\}$  is a stationary process, and determine its ACVF  $\{\gamma_X(h)\}$  in terms of the ACVF for  $\{Y_t\}$ .

(b) Suppose  $\{Z_t\} \sim \text{WN}(0, \sigma^2)$ , where  $0 < \sigma^2 < \infty$ . Determine the ACVF for the causal and invertible ARMA(1,1) process  $X_t = \phi X_{t-1} + Z_t + \theta Z_{t-1}$  in terms of  $\phi$ ,  $\theta$  and  $\sigma^2$ . Do so by combining together

(1) your answer from part (a) and

(2) the fact that  $\gamma_Y(h) = \phi^{|h|}\sigma^2/(1 - \phi^2)$  is the ACVF for  $Y_t = \phi Y_{t-1} + Z_t$

(no credit will be given for just stating the ARMA(1,1) ACVF!).

- [4] Given a stationary process  $\{X_t\}$  with zero mean, let  $\hat{X}_{n+1}$  denote the best linear predictor of  $X_{n+1}$  given  $X_1, X_2, \dots, X_n$ . Argue that  $E\{X_{n+1}\hat{X}_{n+1}\} = \text{var}\{\hat{X}_{n+1}\}$ .

[5] Suppose we have a time series that is a portion  $X_1, X_2, \dots, X_n$  of an AR(1) process  $X_t = \phi X_{t-1} + Z_t$  with mean zero, where  $0 < |\phi| < 1$ , and  $\{Z_t\} \sim \text{WN}(0, \sigma^2)$  (we also assume that  $n \geq 2$  and that the series is not boring).

(a) By definition the Burg estimator  $\bar{\phi}_X$  of  $\phi$  is the value minimizing the sum of squares of forward (one-step-ahead) and backward (one-step-behind) prediction errors when we regard the sum of squares as a function of  $\phi$ . Derive an explicit expression for  $\bar{\phi}_X$  in terms  $X_1, X_2, \dots, X_n$ .

- (b) Define  $Y_1 = Y_{n+2} = 0$  and  $Y_t = X_{t-1}$  for  $t = 2, 3, \dots, n+1$ ; i.e., the time series  $\{Y_t\}$  is a series of length  $n+2$  formed by placing a zero at the beginning and at the end of the  $X_t$  series. Let  $\bar{\phi}_Y$  denote the Burg estimator for  $Y_1, Y_2, \dots, Y_{n+2}$ . Using part (a), what form does  $\bar{\phi}_Y$  take, and how is it related to the Yule–Walker estimator  $\hat{\phi}_X = \hat{\gamma}_X(1)/\hat{\gamma}_X(0)$  for the  $X_t$  series, where  $\hat{\gamma}_X(h)$  is the sample ACVF at lag  $h$ ?

- (c) Assume now that  $n = 2$ , i.e., the time series consists of just  $X_1$  and  $X_2$ . How is the Burg estimator  $\bar{\phi}_X$  related to the Yule–Walker estimator  $\hat{\phi}_X$  for this special case? Given that  $|\bar{\phi}_X| \leq 1$  for any sample size  $n \geq 2$  (a fact you need *not* prove), what range of values can  $\hat{\phi}_X$  assume when  $n = 2$ ?