## Time Series Exercises and Problems, Sheet 2

## Hilary Term 2002

Dr. Gesine Reinert

1. A time series with periodic component can be constructed from

$$Y_t = U_1 sin(2\pi\nu t) + U_2 cos(2\pi\nu t),$$

where  $U_1$  and  $U_2$  are independent random variables with zero means and variances  $\sigma^2$ . The constant  $\nu$  determines the period or the time that it takes the process to make one complete cycle. Show that this series is weakly stationary with acf

$$\rho_h = \cos(2\pi\nu h).$$

2. Consider a process consisting of a linear trend with an additive noise term consisting of independent random variables  $\epsilon_t$  with zero means and variances  $\sigma^2$ , that is,

$$Y_t = \beta_0 + \beta_1 t + \epsilon_t,$$

where  $\beta_0$  and  $\beta_1$  are fixed constants. Prove that  $Y_t$  is nonstationary, but that the first difference series  $\nabla Y_t = Y_t - Y_{t-1}$  is second-order stationary.

**3.** A vector time series  $Y_{\mathbf{t}} = (Y_{t1}, \dots, Y_{tp})^T$  contains as its components p univariate time series. For the stationary case, the  $p \times 1$  mean vector is

$$\mu = E(Y_{\mathbf{t}}) = (\mu_{t1}, \dots, \mu_{tp})^T$$

and the autocovariance function can be defined as a function of the multidimensional lag vector  $\mathbf{h} = (h_1, \dots, h_p)^T$  as

$$\gamma_{\mathbf{h}} = E(Y_{\mathbf{s}+\mathbf{h}} - \mu)(Y_{\mathbf{s}} - \mu).$$

For a two-dimensional process, for example, we have

$$\gamma_{(h_1,h_2)} = E(Y_{s_1+h_1,s_2+h_2} - \mu)(Y_{s_1,s_2} - \mu).$$

A concept used in geostatistics is that of the *variogram*, defined for a spatial process  $Y_{\mathbf{s}}, \mathbf{s} = (s_1, s_2), s_1, s_2 = 0, \pm 1, \pm 2, \dots$  as

$$V_Y(\mathbf{h}) = \frac{1}{2} E\left( (Y_{\mathbf{s}+\mathbf{h}} - Y_{\mathbf{s}})^2 \right).$$

Show that, for a stationary process, the variogram and the autocovariance function can be related through

$$V_Y(\mathbf{h}) = \gamma_0 - \gamma_h,$$

where  $\mathbf{0} = (0, 0)$ .

- 4. Determine which of the following processes are causal and/or invertible. Here,  $\epsilon$  denotes white noise.
- (a)  $Y_t + .2Y_{t-1} .48Y_{t-2} = \epsilon_t$
- (b)  $Y_t + 1.9Y_{t-1} + .88Y_{t-2} = \epsilon_t + .2\epsilon_{t-1} + .7\epsilon_{t-2}$
- (c)  $Y_t + .6Y_{t-2} = \epsilon_t + 1.2\epsilon_{t-1}$
- (d)  $Y_t + 1.8Y_{t-1} + .81Y_{t-2} = \epsilon_t$
- (e)  $Y_t + 1.6Y_{t-1} = \epsilon_t .4\epsilon_{t-1} + .04\epsilon_{t-2}$ .
  - **5.** For an ARMA(1,1) process with  $|\theta| < 1$ , verify that

$$\rho_h = \frac{(1 + \theta\phi)(\theta + \phi)}{1 + 2\theta\phi + \theta^2} \phi^{h-1}, \quad h \ge 1.$$