

## Discussion 4

1. The difference operator  $\nabla$  is defined as  $\nabla X_t = X_t - X_{t-1}$ .
  - a) Suppose  $\{X_t\}$  has a linear trend,  $X_t = \beta_0 + \beta_1 t + Y_t$ , where  $\beta_i$  are constants, and  $\{Y_t\}$  is stationary. Show that  $\nabla X_t$  does not have a trend.
  - b) Suppose  $\{X_t\}$  has a quadratic trend,  $X_t = \beta_0 + \beta_1 t + \beta_2 t^2 + Y_t$ , where  $\beta_i$  are constants, and  $\{Y_t\}$  is stationary. Show that  $\nabla^2 X_t$  does not have a trend.
  - c) Suppose  $\{X_t\}$  is a random walk,  $X_t = X_{t-1} + \varepsilon_t$ , where  $\{\varepsilon_t\}$  is i.i.d. noise. Show that  $\nabla X_t$  is stationary.
  - d) Suppose  $\{X_t\}$  is a random walk with drift,  $X_t = \delta + X_{t-1} + \varepsilon_t$ , where  $\delta$  is a constant, and  $\{\varepsilon_t\}$  is i.i.d. noise. Show that  $\nabla X_t$  is stationary.
2. The backshift operator is defined as  $B(X_t) = X_{t-1}$ . Express the following using backshift notation.
  - a) An  $MA(3)$  process  $X_t = \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2} + \theta_3 \varepsilon_{t-3} + \varepsilon_t$ , where  $\theta_i$  are constants, and  $\{\varepsilon_t\}$  is i.i.d. noise.
  - b) An  $AR(3)$  process  $X_t = \phi_1 X_{t-1} + \phi_2 X_{t-2} + \phi_3 X_{t-3} + \varepsilon_t$ , where  $\phi_i$  are constants, and  $\{\varepsilon_t\}$  is i.i.d. noise.
  - c) An  $ARMA(2, 3)$  process  $X_t - \phi_1 X_{t-1} - \phi_2 X_{t-2} = \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2} + \theta_3 \varepsilon_{t-3} + \varepsilon_t$ , where  $\phi_i$  and  $\theta_i$  are constants, and  $\{\varepsilon_t\}$  is i.i.d. noise.
3. Let  $\{X_t\}$  be a sequence (not necessarily  $AR(p)$ ) with mean  $\mu$  and lag 1 autocovariance  $\rho(1) = \phi$ .
  - a) The best linear predictor of  $X_t$  using  $X_{t-1}$  is  $X_t^{(f)} = \mu + \phi(X_{t-1} - \mu)$ .  
Show that  $E(X_t - X_t^{(f)})^2 = (1 + \phi^2)\gamma(0) - 2\phi\gamma(1)$ .
  - b) The best linear predictor of  $X_{t-2}$  using  $X_{t-1}$  is  $X_{t-2}^{(b)} = \mu + \phi(X_{t-1} - \mu)$ .  
 $E(X_{t-2} - X_{t-2}^{(b)})^2 = (1 + \phi^2)\gamma(0) - 2\phi\gamma(1)$ .