

Econometrics Homework #3, Spring 2007  
Max Stinchcombe

1. [Comparing linear prediction and best prediction] Let  $X_1$  and  $X_2$  be independent random variables with  $P(X_i = -1) = P(X_i = +1) = \frac{1}{2}$ ,  $i = 1, 2$  and define  $Y = X_1 \cdot X_2$ .
  - (a) Give the distribution of  $Y$ .
  - (b) Give  $E(Y | X_1)$ ,  $E(Y | X_2)$ , and  $E(Y | X_1, X_2)$ .
  - (c) Solve the problem  $\min_{\beta_0, \beta_1, \beta_2} E(Y - [\beta_0 + \beta_1 X_1 + \beta_2 X_2])^2$ .

2. Let  $X_t$  follow an invertible MA(1) process:  $X_t = \alpha X_{t-1} + u_t$ , where  $|\alpha| < 1$  and  $u_t \sim N(0, \sigma_u^2)$ . Let  $v_t$  be another white noise process such that for all  $s, t$ ,  $v_t$  is uncorrelated with  $u_s$ , and  $Var(v_t) = \sigma_v^2$ .

We will now see that  $Z_t := X_t + v_t$  can also be represented as an invertible MA(1) process. To do this, it is sufficient to show that there exists  $|\theta| < 1$  such that the process

$$\epsilon_t := Z_t - \theta Z_{t-1} + \theta^2 Z_{t-2} - \theta^3 Z_{t-3} + \dots \quad (1)$$

is a white noise process.

- (a) Why does the claim follow from eqn. (1)?
  - (b) Find the variance and autocovariances of  $Z_t$ . Are the autocovariances consistent with the MA(1) pattern in general? Give the spectral density of  $Z_t$ .
  - (c) Assuming  $Z_t$  indeed has an MA(1) representation, set up a system of equations that  $\theta$  would have to satisfy by equating variances and autocovariances from this representation with those under (b).
  - (d) Show that this system of equations has a solution  $0 < |\theta^*| < 1$ .
  - (e) Using the  $\theta^*$  just described under (d), define  $Z_t$  as in eqn. (1). Show that this is a white noise process. [Think spectral density.]
3. This exercise will demonstrate through a simple example how to use maximum likelihood or method of moments to estimate moving average models. Consider the MA(1) model  $Y_t = \epsilon_t + \theta \epsilon_{t-1}$  where  $\epsilon_t$  is iid  $N(0, \sigma^2)$ .
  - (a) Show that the conditional distribution of  $Y_t$  given  $Y_{t-1}, \dots, Y_1, \epsilon_0$  is the same as the conditional distribution of  $Y_t$  given  $\epsilon_{t-1}, \dots, \epsilon_0$ .
  - (b) Show that the conditional distribution of  $Y_t$  given  $Y_{t-1}, \dots, Y_1, \epsilon_0$  is the same as the conditional distribution of  $Y_t$  given  $\epsilon_{t-1}$ .
  - (c) Suppose we have three observations  $Y_3, Y_2$  and  $Y_1$ . Using the previous, Express the conditional density of  $(Y_3, Y_2, Y_1)$  given  $\epsilon_0$  in terms of the conditional densities  $Y_t | \epsilon_{t-1}$ ,  $t = 3, 2, 1$ . Using the normality assumption, write down an explicit expression for this conditional density.

- (d) Suppose  $\epsilon_0 = 0$ . Substitute out  $\epsilon_2$  and  $\epsilon_1$  in terms of  $Y_2, Y_1$ . Find the conditional MLE of  $\theta$  if  $Y_1 = -0.5, Y_2 = 0, Y_3 = -0.5$ .
- (e) Using the same sample, calculate the method of moments estimator of  $\theta$  by equating the population autocorrelations of the process with their sample counterparts.
4. [For both this problem and the next one, you can also simulate to get your answers. If you do this, be explicit about your simulation process.] Suppose that  $Y_t = c + \varphi_1 Y_{t-1} + \varphi_2 Y_{t-2} + \epsilon_t$  where  $\epsilon_t$  is a white noise process,  $(\varphi_1, \varphi_2) = (0.5, 0.24)$ , and  $t \in \{\dots, -2, -1, 0, 1, 2, \dots\}$ .
- (a) Find the eigen-values of the associated  $F$  matrix and show that they are inside the unit circle. [From here onwards, you can assume that  $Y_t$  is a weakly stationary sequence with  $\sum_j |\gamma_j| < \infty$ .]
- (b) Give  $E Y_t$ , and the  $\gamma_j, j \in \{\dots, -2, -1, 0, 1, 2, \dots\}$ .
- (c) Based on the sample  $Y_1, \dots, Y_T$ , let  $\hat{\mu}_T$  and  $\hat{\gamma}_{j,T}$  be the population moments,  $j = 0, 1, 2, 3$ . Find  $E(\hat{\mu}_T), Var(\hat{\mu}_T), E\hat{\gamma}_{j,T}$ , and  $Var\hat{\gamma}_{j,T}$ .
5. One expects problems when the eigen-values get close to the edge of the unit circle. Repeat the previous problem for  $(\varphi_1, \varphi_2) = (1.85, -0.9)$  and explain the source(s) of the difference(s).
6. Suppose that  $Y_t = a + b \cdot t + \varphi_1 Y_{t-1} + \dots + \varphi_p Y_{t-p} + \epsilon_t, b \neq 0, \epsilon_t$  a white noise process.
- (a) Is  $Y_t$  any kind of stationary? Explain.
- (b) Define  $X_t = \Delta Y_t := Y_t - Y_{t-1}$ . What kind of process is  $X_t$ ? Is it weakly stationary? If  $\epsilon_t$  is strictly stationary, is  $X_t$  also strictly stationary?
- (c) Define  $Z_t = \Delta X_t = X_t - X_{t-1}$ . What kind of process is  $Z_t$ ? Is it weakly stationary? If  $\epsilon_t$  is strictly stationary, is  $Z_t$  also strictly stationary?
7. Suppose that  $Y_t = a + b \cdot t + c \cdot t^2 + \varphi_1 Y_{t-1} + \dots + \varphi_p Y_{t-p} + \epsilon_t, c \neq 0, \epsilon_t$  a white noise process.
- (a) Is  $Y_t$  any kind of stationary? Explain.
- (b)  $X_t = \Delta Y_t := Y_t - Y_{t-1}$ . Is  $X_t$  any kind of stationary? Explain.
- (c) Define  $Z_t = \Delta X_t = X_t - X_{t-1}$ . What kind of process is  $Z_t$ ? Is it weakly stationary? If  $\epsilon_t$  is strictly stationary, is  $Z_t$  also strictly stationary?
8. Volatile time series are often “smoothed out” by some sort of averaging. In particular, let  $X_t$  be a covariance stationary time series and let  $\tilde{X}_t$  denote its smoothed version defined by an  $m$ -period centered moving average

$$\tilde{X}_t = (X_{t-m} + \dots + X_{t-1} + X_t + X_{t+1} + \dots + X_{t+m}) / (2m + 1).$$

- (a) Let  $m = 1$ . Using lag operator notation, write down the linear filter that transforms  $X_t$  into  $\tilde{X}_t$ .
- (b) Find the filter function (i.e. the function by which the spectral density of  $X_t$  has to be multiplied to obtain the spectral density of  $\tilde{X}_t$ ).
- (c) Compare the spectral density of  $X_t$  with the spectral density of  $\tilde{X}_t$ . Which frequencies are missing from the spectrum of  $\tilde{X}_t$ ? Which frequencies are dampened down? Which are amplified?
- (d) Let  $m = 2$ . Write down and graph the filter function (most easily done with a computer program). By averaging over more observations in the time domain, we should get a smoother series than before. Justify this claim in the frequency domain by comparing the graphs of the filter functions obtained for  $m = 1$  and  $m = 2$ .