Problem Set #5 Spring 2014

- 1. Rewrite the following expressions without using the lag operator
 - (a) $(1 \rho L)y_t = \varepsilon_t$
 - (b) $(1 + .2L .8L^2)y_t = \varepsilon_t$
 - (c) $y_t = (1 \theta L)\varepsilon_t$
 - (d) $y_t = (1 .3L + .5L^2)\varepsilon_t$
- 2. Rewrite the following expressions in lag operator form
 - (a) $y_t = .8y_{t-1} + \varepsilon_t$
 - (b) $y_t = .2y_{t-1} + .3y_{t-2} .7y_{t-5} + \varepsilon_t$
 - (c) $y_t = \varepsilon_t .7\varepsilon_{t-1}$
 - (d) $y_t = \varepsilon_t + .3\varepsilon_{t-4}$
- 3. Consider the following MA(1) process

$$y_t = 2.3 - 0.95\varepsilon_{t-1} + \varepsilon_t$$

with ε_t iid N(0,1).

- (a) What is the optimal forecast for time periods T + h, h = 1, 2, 3. Write your answer as a function of $y_1, ..., y_T$ and/or $\varepsilon_1, ..., \varepsilon_T$
- (b) Now suppose that $\varepsilon_T = 0.4$ and $\varepsilon_{T-1} = -1.2$. Re-answer part (a)
- 4. The stata file "gdp2013.dta" contains quarterly observations on the real growth rates of the components of U.S. gdp.

Pick two of the series. Graph their autocorrelation functions. Comment and discuss.

- 5. The file "housingstarts.dta" has data on housing starts. Graph the autocorrelation function for the "south" and "midwest" regions. Comment and discuss.
- 6. The file "s&p.dta" has stock return dat, include transaction volume. In an earlier exercise you estimated a trend model for the natural log of volume. Re-estimate that model, but this time on the full sample. That is, take the log of the volume and regress it on a time trend. Take the residuals of that regression. (The command is **predict e, residuals** after estimating the regression). Graph a time-series plot of the residuals. Graph the autocorrelation function of the residuals. Comment and discuss.
- 7. For this problem you will create simulated data. See the description of simulation in STATA at the end of this problem set.

Simulate each of processes with T = 100 observations, with ε_t iid N(0, 1). For each series, graph the autocorrelation function, and estimate a MA(2) model. Commend and discuss. For estimation, use the command (where y is the name of the series you created)

.arima y arima(0,0,2)

- (a) $y_t = \varepsilon_t$ (white noise)
- (b) $y_t = \varepsilon_t + 0.8\varepsilon_{t-1}$
- (c) $y_t = \varepsilon_t 0.6\varepsilon_{t-1} + 0.4\varepsilon_{t-2}$
- (d) $y_t = .5y_{t-1} + \varepsilon_t$

Simulation in STATA

Starting with a blank STATA session, you can set the number of observations and declare the file to be time-series as follows. If you want to create a file with 100 observations, use the commands

. set obs 100

.gen t=_n

.ts set t

To simulate a Gaussian white noise process use the command

.gen e=rnormal()

This generates (creates) a variable "e" (or any name you decide) filled with random normal N(0,1) innovations

To create a moving average process $y_t = \varepsilon_t + 0.3\varepsilon_{t-1}$, use the command

.gen y=e+0.3*L.e

This takes the variable "e" previously defined, creates the lagged value, and creates the moving average with the coefficient 0.3

To create an autoregression, you need a bit more creativity because of the recursion. To create the process $x_t = 0.5x_{t-1} + \varepsilon_t$ with $x_1 = 0$

.gen x=0

.replace x=0.5*L.x+e if t>1

The first command sets the variable $x_t = 0$. We only use this for the first observation, thus $x_1 = 0$. (You can relace 0 with whatever starting value is appropriate.) Then the second command replaces the x_t by the recursion $x_t = 0.5x_{t-1} + e_t$, for observations t > 1, where the e variable is previously defined, and the time index t was previously defined. The command is **replace** rather than **gen** since the variable x already exists, and **gen** will not write over an existing variable.