

## ECOM073: Topics in Financial Econometrics

### Exercise 7.

**Problem 7.1.** Assume that  $X_t$  is a random walk with a drift:

$$X_t = \mu + X_{t-1} + \varepsilon_t, \quad t = 1, 2, \dots$$

with the initial value  $X_0 = 0$ , where  $\varepsilon_t$  is a white noises with zero mean and variance  $\sigma_\varepsilon^2$ .

Show that

- (i)  $E[X_t] = \mu t$ .
- (ii)  $\text{Var}(X_t) = t\sigma_\varepsilon^2$ .
- (iii)  $\text{Cov}(X_t, X_s) = \min(t, s)\sigma_\varepsilon^2$ .

**Solution 1.** (i) We can write

$$\begin{aligned} X_t &= \mu + X_{t-1} + \varepsilon_t = \mu + (\mu + X_{t-2} + \varepsilon_{t-1}) + \varepsilon_t \\ &= 2\mu + X_{t-2} + \varepsilon_{t-1} + \varepsilon_t = 3\mu + X_{t-3} + \varepsilon_{t-2} + \varepsilon_{t-1} + \varepsilon_t \\ &= \dots = t\mu + X_0 + \varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \dots + \varepsilon_1 \\ &= t\mu + \varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \dots + \varepsilon_1, \end{aligned}$$

since  $X_0 = 0$ . Then

$$\begin{aligned} E[X_t] &= E[t\mu + \varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_1] \\ &= t\mu + E[\varepsilon_t] + E[\varepsilon_{t-1}] + \dots + E[\varepsilon_1] = t\mu + 0 + 0 + \dots + 0 = t\mu. \end{aligned}$$

(ii)

$$\begin{aligned} \text{Var}(X_t) &= E[(X_t - E[X_t])^2] = E[X_t^2] \\ &= E[(\varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_1)^2] = E[\varepsilon_t^2] + E[\varepsilon_{t-1}^2] + \dots + E[\varepsilon_1^2] \\ &= \sigma_\varepsilon^2 + \sigma_\varepsilon^2 + \dots + \sigma_\varepsilon^2 = t\sigma_\varepsilon^2. \end{aligned}$$

because  $\varepsilon_t$  is a white noise and therefore  $E[\varepsilon_i\varepsilon_j] = 0$ , if  $i \neq j$ ; and  $E[\varepsilon_i\varepsilon_j] = \sigma_\varepsilon^2$ , if  $i = j$ .

(iii) By definition,

$$\text{Cov}(X_t, X_s) = E[(X_t - E[X_t])(X_s - E[X_s])].$$

Since  $E[X_t] = t\mu$  and

$$X_t = t\mu + \varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \dots + \varepsilon_1,$$

then for  $t \geq s$ ,

$$\begin{aligned} \text{Cov}(X_t, X_s) &= E[X_t X_s] = E[(\varepsilon_t + \varepsilon_{s-1} + \dots + \varepsilon_1)(\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1)] \\ &= E[(\{\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1\} + \{\varepsilon_{s+1} + \dots + \varepsilon_t\})(\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1)] \\ &= E[(\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1)^2] + E[(\varepsilon_{s+1} + \dots + \varepsilon_t)(\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1)] \\ &= s\sigma_\varepsilon^2 + 0 = s\sigma_\varepsilon^2. \end{aligned}$$

Therefore

$$\text{Cov}(X_t, X_s) = s\sigma_\varepsilon^2 = \min(t, s)\sigma_\varepsilon^2.$$

### Problem 7.2.

(a) Suppose that  $X_t$  follows a unit root model

$$X_t = X_{t-1} + \varepsilon_t$$

with initial value  $X_0 = 0$  where  $\varepsilon_t$  as a white noise process with zero mean and variance  $\sigma_\varepsilon^2$

- (i) Obtain the 1-step ahead prediction  $X_t(1)$  at the origin  $t$ .
- (ii) Derive 2-step ahead prediction  $X_t(2)$
- (iii) Is the forecast mean reverting?

**Solution.** (i) We have that

$$\hat{X}_t(1) = E[X_{t+1}|F_t] = [X_{t+1}] = [X_t + \varepsilon_{t+1}] = [X_t] + [\varepsilon_{t+1}] = X_t + 0 = X_t.$$

(ii) Similarly,

$$\hat{X}_t(2) = E[X_{t+2}|F_t] = [X_{t+2}] = [X_{t+1} + \varepsilon_{t+2}] = [X_{t+1}] + [\varepsilon_{t+2}] = \hat{X}_t(1) = X_t.$$

(iii) We find that

$$\begin{aligned}\hat{X}_t(k) &= [X_{t+k}] = [X_{t+k-1} + \varepsilon_{t+k}] = [X_{t+k-1}] \\ &= \text{continuing as above} = [X_{t+k-2}] = \dots = [X_t] = X_t.\end{aligned}$$

We showed in the first exercise that  $EX_0 = 0$ , which is different from  $X_t$ . So the forecast is not mean reverting.

**Problem 7.3.**

(a) Suppose that  $X_t$  follows a unit root model and

$$X_t = X_{t-1} + z_t$$

where  $z_t$  is an AR(2) time series:

$$z_t = 3 + 0.5z_{t-1} + 0.1z_{t-2} + \varepsilon_t,$$

where  $\varepsilon_t$  is a white noise with the mean 0 and variance  $\sigma_\varepsilon^2$ .

Find the 1-step ahead forecast  $X_t(1)$ .

**Solution.** First we find the 1-step ahead forecast  $z_t(1)$  of  $z_{t+1}$ :

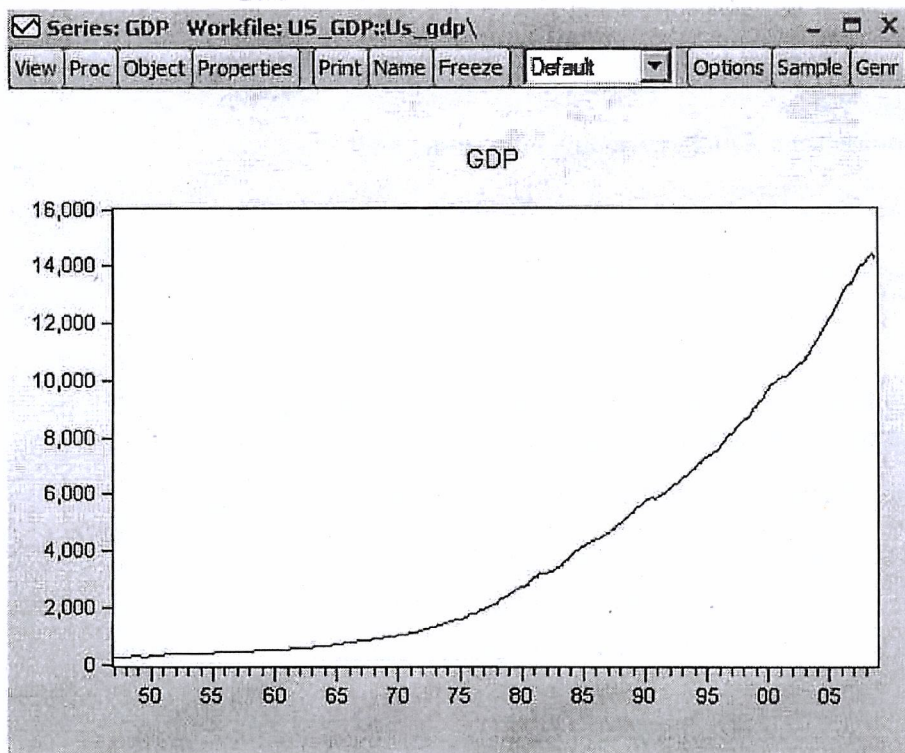
$$\begin{aligned}\hat{z}_t(1) &= E[z_{t+1}|F_t] = [z_{t+1}] \\ &= [3 + 0.5z_t + 0.1z_{t-1} + \varepsilon_{t+1}] \\ &= 3 + 0.5[z_t] + 0.1[z_{t-1}] + [\varepsilon_{t+1}] \\ &= 3 + 0.5z_t + 0.1z_{t-1} \\ &= 3 + 0.5(X_t - X_{t-1}) + 0.1(X_{t-1} - X_{t-2}).\end{aligned}$$

Then

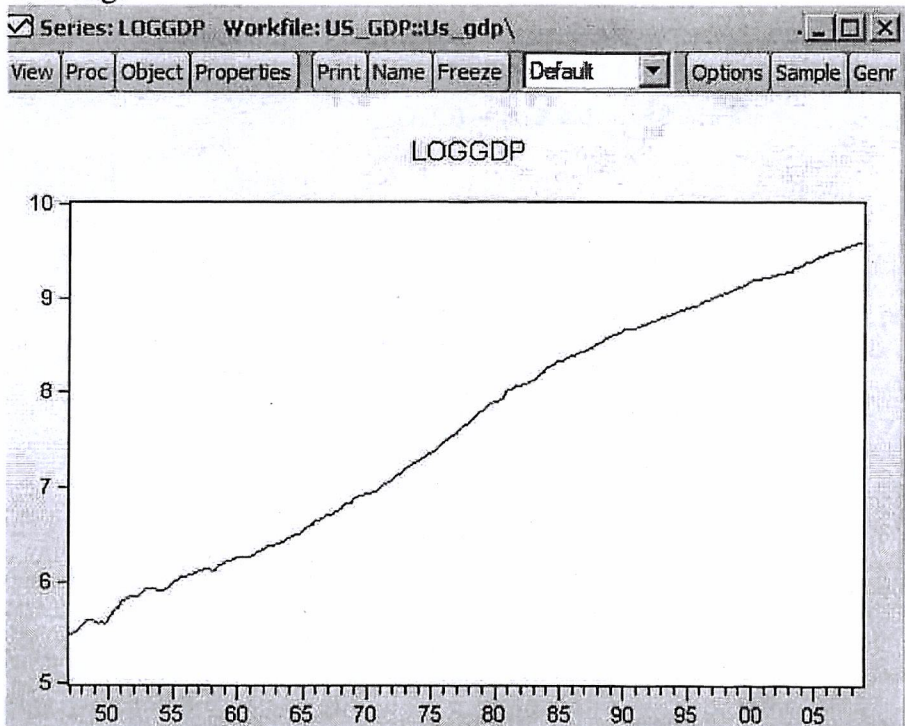
$$\begin{aligned}\hat{X}_t(1) &= [X_{t+1}] = [X_t + z_{t+1}] = [X_t] + [z_{t+1}] = X_t + \hat{z}_t(1) \\ &= X_t + 3 + 0.5(X_t - X_{t-1}) + 0.1(X_{t-1} - X_{t-2}) \\ &= 3 + 1.5X_t - 0.4X_{t-1} - 0.1X_{t-2}.\end{aligned}$$

**Problem 7.4.** (EViews exercise) Test for unit root in the U.S. quarterly GDP data. Fit a model to that data.

**Solution.** 1 To test for unit root, use the log  $X_t$ ,  $t = 1, \dots, 243$  observations shown in the graph LOGGDP.



Take logs



2. To test for unit root use the Augmented D-F test with automatic selection of the number of lags  $p$  (by AIC criterion). It gives  $p = 4$ .

$p = 0.9956$  value shows that we do not reject unit root.

Output shows that to the data  $Y_t = \log X_t$  we can fit the model (with  $\Delta Y_t = Y_t - Y_{t-1}$ ):

$$\Delta Y_t = c + \text{"trend"}t + \beta_c Y_{t-1} + \delta_1 \Delta Y_{t-1} + \dots + \delta_4 \Delta Y_{t-4} + \varepsilon_t.$$

Estimation shows that  $\beta_c = -0.000214$ ,  $c = 0.011618$  and "trend" =  $-6.83E-05$  are not significant, as well as the coefficients  $\delta_3 = 0.0704$  and  $\delta_4 = 0.0658$ . The fitted model is

$$\Delta Y_t = 0.4221 \Delta Y_{t-1} + 0.1941 \Delta Y_{t-2} + \varepsilon_t.$$

## USE LGDP (log of GDP)

### Estimation and ADF test on log(GDP)

Views - [Series: LGDP, Workfile: US GDP=US_gdp^]				
File Edit Object View Proc Quick Options Add-ins Window Help				
View	Proc	Object	Properties	Print Name Freeze Sample Genr Sheet Graph Stats Ident
<b>Augmented Dickey-Fuller Unit Root Test on LGDP</b>				
Null Hypothesis: LGDP has a unit root				
Exogenous: Constant, Linear Trend				
Lag Length: 4 (Fixed)				
			t-Statistic	Prob.*
<b>Augmented Dickey-Fuller test statistic</b>			<b>-0.039245</b>	<b>0.9956</b>
Test critical values:	1% level		-3.996271	
	5% level		-3.428426	
	10% level		-3.137619	
*MacKinnon (1996) one-sided p-values.				
Augmented Dickey-Fuller Test Equation				
Dependent Variable: D(LGDP)				
Method: Least Squares				
Date: 03/06/12 Time: 18:18				
Sample (adjusted): 1948Q2 2008Q4				
Included observations: 243 after adjustments				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
LGDP(-1)	-0.000214	0.005462	-0.039245	0.9687
D(LGDP(-1))	0.422150	0.066154	6.381327	0.0000
D(LGDP(-2))	0.194117	0.070375	2.758317	0.0063
D(LGDP(-3))	-0.132396	0.070432	-1.879774	0.0614
D(LGDP(-4))	-0.042067	0.065829	-0.638041	0.5234
C	0.011618	0.028132	0.398790	0.6904
@TREND(1947Q1)	-6.93E-06	9.84E-05	-0.069416	0.9447
R-squared	0.253101	Mean dependent var		0.016456
Adjusted R-squared	0.234113	S.D. dependent var		0.011168
S.E. of regression	0.009773	Akaike Info criterion		-6.389901
Sum squared resid	0.022543	Schwarz criterion		-6.289278
Log likelihood	793.3730	Hannan-Quinn criter.		-6.349371
F-statistic	13.32888	Durbin-Watson stat		1.977094
Prob(F-statistic)	0.000000			

We cannot reject the null of the presence of a unit root.

3. The above results suggest that  $z_t = Y_t - Y_{t-1}$  follows AR(2) model. If we test for the unit root in  $z_t$ , we find no unit root.

Take differences and estimate (this is a process with a constant and a trend).

genr diflgdp = d(loggdp)

EViews - [Series: DIFLGDP Workfile: US GDP:US gdp\]				
File Edit Object View Proc Quick Options Add-ins Window Help				
View	Proc	Object	Properties	Print Name Freeze Sample Genr Sheet Graph Stats Ident
<b>Augmented Dickey-Fuller Unit Root Test on DIFLGDP</b>				
Null Hypothesis: DIFLGDP has a unit root				
Exogenous: Constant, Linear Trend				
Lag Length: 4 (Fixed)				
			t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic			-6.948482	0.0000
Test critical values:				
	1% level		-3.986431	
	5% level		-3.428503	
	10% level		-3.137665	
*MacKinnon (1996) one-sided p-values.				
Augmented Dickey-Fuller Test Equation				
Dependent Variable: D(DIFLGDP)				
Method: Least Squares				
Date: 03/06/12 Time: 18:23				
Sample (adjusted): 1948Q3 2008Q4				
Included observations: 242 after adjustments				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
DIFLGDP(-1)	-0.625886	0.090075	-6.948482	0.0000
D(DIFLGDP(-1))	0.042208	0.086496	0.487980	0.6260
D(DIFLGDP(-2))	0.219443	0.079212	2.770327	0.0060
D(DIFLGDP(-3))	0.109443	0.075679	1.446146	0.1495
D(DIFLGDP(-4))	0.114535	0.065271	1.754753	0.0806
C	0.011686	0.002112	5.531876	0.0000
@TREND(1947Q1)	-1.14E-05	9.05E-06	-1.256593	0.2101
R-squared	0.297334	Mean dependent var		-0.000169
Adjusted R-squared	0.279393	S.D. dependent var		0.011463
S.E. of regression	0.009730	Akaike info criterion		-6.398606
Sum squared resid	0.022250	Schwarz criterion		-6.297687
Log likelihood	781.2314	Hannan-Quinn criter.		-6.357952
F-statistic	16.57341	Durbin-Watson stat		1.946147
Prob(F-statistic)	0.000000			

No unit root found, as expected.