

# Financial Forecasting

## M.Sc. in Finance

### List of Exercises

(based on the Lecture Notes of Prof. António Costa (ISEG) and Prof. Nuno Sobreira (ISEG) and on the textbook Gloria Gonzalez-Rivera, *Forecasting for Economics and Business*, Pearson, 2013)

Exercises 5, 6a) (chapter 3 Textbook)

Exercises 7 a) b), 8a) and 9 (chapter 3 Textbook)

1. Consider the following stochastic processes where  $\varepsilon_t \sim \text{WN}(0, \sigma_\varepsilon^2)$ ,  $\beta_1, \beta_2 \neq 0$  :
  - i.  $X_t = \alpha + \varepsilon_t$
  - ii.  $X_t = \beta_0 + \beta_1 t + \beta_2 t^2 + \varepsilon_t$
  - iii.  $X_t = \alpha + X_{t-1} + \varepsilon_t$  with  $X_0$  fixed
- a. Identify the processes that are stationary.
- b. For the stationary processes verify that  $\rho_k = \text{Corr}(X_t, X_{t-k}) \rightarrow 0$  with  $k \rightarrow \infty$
- c. For the nonstationary processes, propose a transformation that makes the process stationary.

Exercise 3 (chapter 6 Textbook)

2. Consider the MA(1) process  $y_t = \varepsilon_t - 0.12 \varepsilon_{t-1}$  where  $\varepsilon_t \sim \text{WN}(0, \sigma_\varepsilon^2)$ 
  - a. Find the ACF of the process.
  - b. Is the model stationary? Justify your answer.
  - c. Is the model invertible? Justify your answer.
  - d. Characterize the behavior of the Partial Autocorrelation Function of the process.
3. Consider the MA(2) process  $y_t = 14 + \varepsilon_t - 0.1 \varepsilon_{t-1} + 0.21 \varepsilon_{t-2}$  where  $\varepsilon_t \sim \text{WN}(0, \sigma_\varepsilon^2)$ 
  - a. Find the ACF of the process.
  - b. Is the model stationary? Justify your answer.
  - c. Is the model invertible? Justify your answer.
  - d. Characterize the behavior of the Partial Autocorrelation Function of the process.

4. Suppose that you have time series data of a given country's inflation denoted as  $y_t$ . With these data the following model was estimated:

Dependent Variable: INFL  
Method: Least Squares  
Date: 11/08/12 Time: 16:08  
Sample: 1971M01 2011M12  
Included observations: 492  
Convergence achieved after 35 iterations  
MA Backcast: 1970M10 1970M12

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	1.000127	0.007137	140.1296	0.0000
MA(1)	0.372642	0.034864	10.68835	0.0000
MA(2)	-0.287517	0.036356	-7.908265	0.0000
MA(3)	-0.642579	0.035015	-18.35160	0.0000
R-squared	0.568424	Mean dependent var	1.002714	
Adjusted R-squared	0.565771	S.D. dependent var	0.536875	
S.E. of regression	0.353779	Akaike info criterion	0.767809	
Sum squared resid	61.07791	Schwarz criterion	0.801943	
Log likelihood	-184.8810	Hannan-Quinn criter.	0.781212	
F-statistic	214.2467	Durbin-Watson stat	1.389038	
Prob(F-statistic)	0.000000			
Inverted MA Roots	.85	-.61+.62i	-.61-.62i	

- Obtain the general theoretical expression for  $E[y_t]$  and  $Var(y_t)$
- Using the Eviews output provide an estimate for  $E[y_t]$  and  $Var(y_t)$

*Exercises 1, 5 and 6 (chapter 7 Textbook)*

- Consider the AR(2) process  $y_t = 2 + 0.8y_{t-1} - 0.1y_{t-2} + \varepsilon_t$  where  $\varepsilon_t \sim WN(0, \sigma_\varepsilon^2)$ 
  - Is the process stationary?
  - Compute the unconditional mean of the process.
  - Determine the PACF and describe the ACF of the process.
- Suppose that you have time series data of a given country's inflation denoted as  $y_t$ . With these data the following model was estimated:

Dependent Variable: INFL  
 Method: Least Squares  
 Date: 11/07/12 Time: 14:02  
 Sample (adjusted): 1971M03 2011M12  
 Included observations: 490 after adjustments  
 Convergence achieved after 3 iterations

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.999994	0.016895	59.18885	0.0000
AR(1)	1.006304	0.029335	34.30388	0.0000
AR(2)	-0.769362	0.029310	-26.24904	0.0000
R-squared	0.718467	Mean dependent var		1.000375
Adjusted R-squared	0.717311	S.D. dependent var		0.536716
S.E. of regression	0.285364	Akaike info criterion		0.336001
Sum squared resid	39.65768	Schwarz criterion		0.361681
Log likelihood	-79.32032	Hannan-Quinn criter.		0.346087
F-statistic	621.4079	Durbin-Watson stat		2.779603
Prob(F-statistic)	0.000000			
Inverted AR Roots	.50+.72i	.50-.72i		

- a. Obtain the general theoretical expression for  $E[y_t]$ .
  - b. Using the Eviews output provide an estimate for  $E[y_t]$ .
  - c. Suppose that the inflation rate at November 2011 and December 2011 were 1 and 1.2 respectively. Obtain the optimal forecast estimate for the inflation rate according to this model for:
    - i. January 2012
    - ii. February 2012
  - d. Provide estimates for the forecast uncertainty for:
    - i. January 2012
    - ii. February 2012
11. Write the equation that defines a process  $ARMA(0,1)(0,1)_{12}$  with parameters  $\theta_{12} = 0.8$  and  $\theta_1 = 0.6$  and find the ACF of the process.
  12. Write the equation that defines a process  $ARMA(1,0)(1,0)_4$  with parameters  $\phi_4 = 0.8$  and  $\phi_1 = 0.6$  and find the PACF of the process.
  13. Suppose that you want to analyze a given time series data with the correlogram of Figure 1 . According to this information, what is the best model for this time series? Justify your answer.

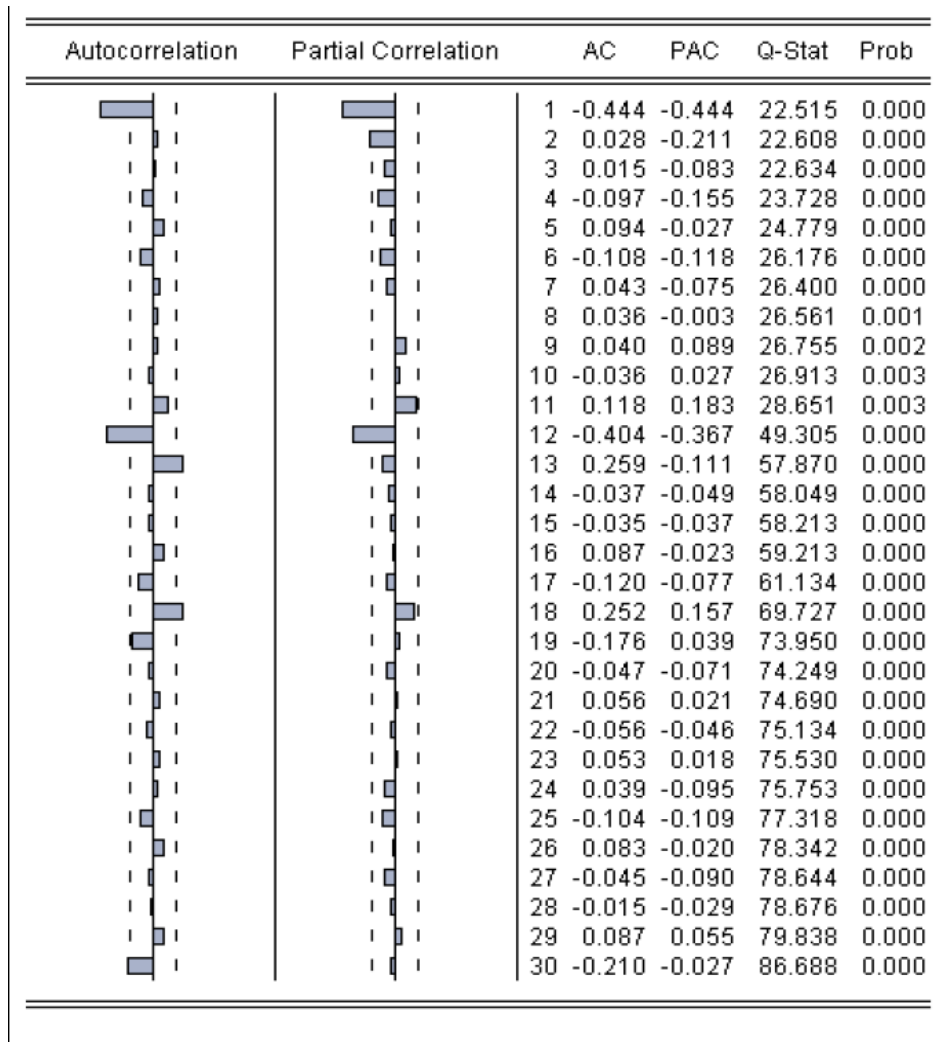


Figure 1

14. Are the following processes stationary/causal? Are the following processes invertible? Justify your answers. Consider that  $\varepsilon_t \sim WN(0, \sigma_\varepsilon^2)$ .
- $y_t = \varepsilon_t + 0.8 \varepsilon_{t-1} - \varepsilon_{t-2}$
  - $y_t = 0.6 y_{t-1} + 0.4 y_{t-2} + \varepsilon_t$
  - $y_t = (1 - 0.7L + 0.3L^2) \varepsilon_t$
15. Consider the process ARMA(1,1) with  $\phi = 0.8$  and  $\theta = 0.5$  and with mean equal to 10.
- Formulate the equation that defines the process
  - Find the ACF of the process.

16. For the following processes identify the orders of the autoregressive and moving average part and write the ARMA representation without the lag operator:

i.  $Y_t = (1 - 0.5L)\varepsilon_t$

ii.  $(1 + 0.8L)Y_t = (1 - 1.2L)\varepsilon_t$

iii.  $(1 - 0.7L + 0.4L^2)Y_t = (1 - 1.2L)\varepsilon_t$

iv.  $(1 + 0.8L)Y_t = (1 - 0.7L + 0.4L^2 + L^3)\varepsilon_t$

17. Consider the following models where  $\varepsilon_t \sim \text{WN}(0, \sigma_\varepsilon^2)$ :

v.  $Y_t = Y_{t-1} + \varepsilon_t - 1.5\varepsilon_{t-1}$

vi.  $Y_t = 0.8Y_{t-1} + \varepsilon_t - 0.5\varepsilon_{t-1}$

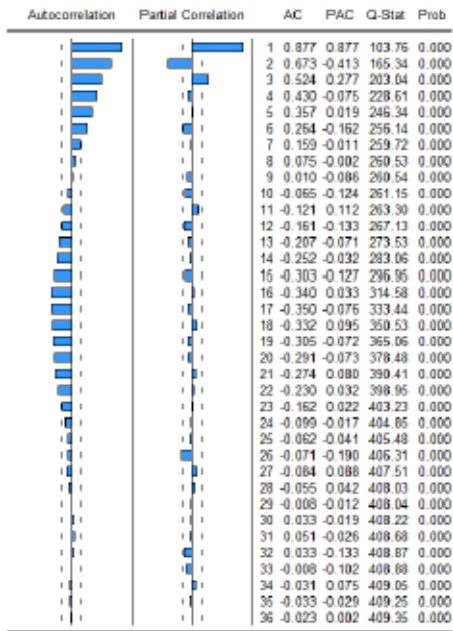
vii.  $Y_t = 1.1Y_{t-1} + 0.8Y_{t-2} + \varepsilon_t - 1.7\varepsilon_{t-1} + 0.72\varepsilon_{t-2}$

viii.  $Y_t = 0.6Y_{t-1} + \varepsilon_t - 1.2\varepsilon_{t-1} + 0.2\varepsilon_{t-2}$

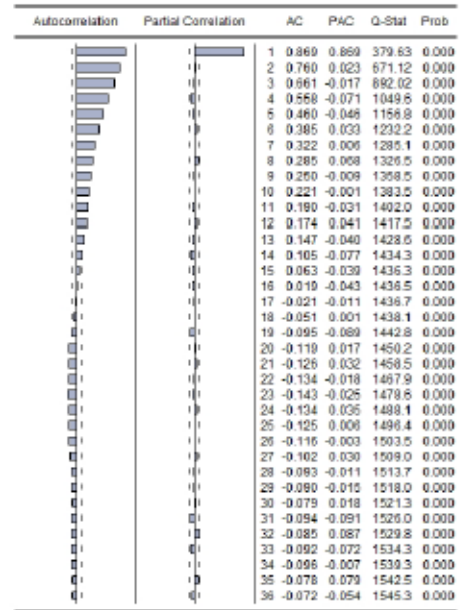
a) Verify if  $Y_t$  is stationary and invertible.

b) Characterize the behavior of the ACF and PACF.

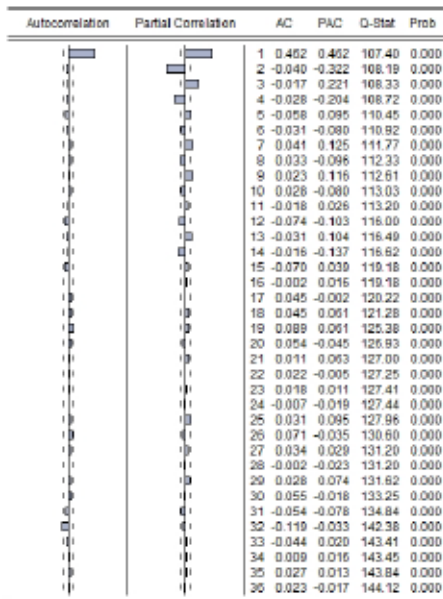
18. In the following figure you may find the ACF and PACF of four time series.



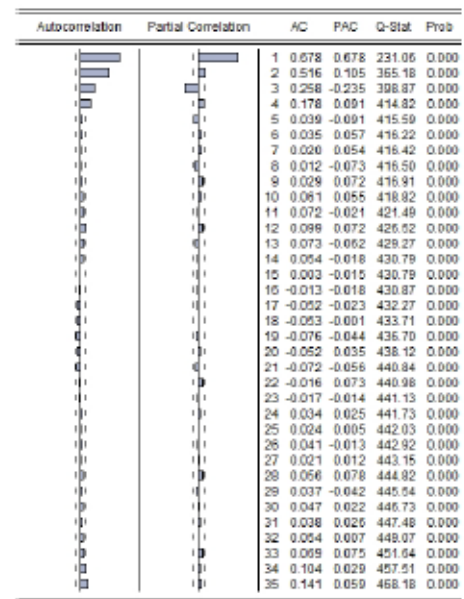
(a) Time series  $x_t$



(b) Time series  $y_t$



(c) Time series  $w_t$



(d) Time series  $z_t$

According to the previous figures identify an appropriate ARMA model for each series. Justify.

19. Consider the following estimation outputs for a fitted model on a price index (IPI).

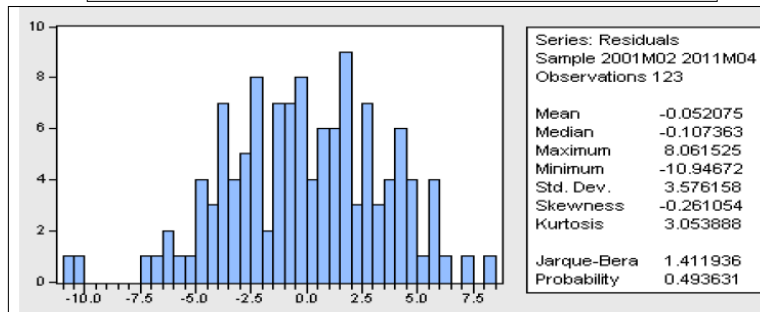
- b. Write the estimated model in equation form.
- c. Comment on the residuals distribution.
- d. Is the proposed model acceptable?

Dependent Variable: D(IPI,1,12)  
 Sample (adjusted): 2001M02 2011M04  
 Included observations: 123 after adjustments  
 MA Backcast: 2000M01 2001M01

Variable	Coefficient	Std. Error	t-Statistic	Prob.
MA(1)	-0.740267	0.060517	-12.23246	0.0000
SMA(12)	-0.887638	0.023475	-37.81162	0.0000

R-squared	0.557799	Mean dependent var	-0.071479
Adjusted R-squared	0.554144	S.D. dependent var	5.378399
S.E. of regression	3.591289	Akaike info criterion	5.411026
Sum squared resid	1560.580	Schwarz criterion	5.456753
Log likelihood	-330.7781	Hannan-Quinn crter.	5.429600
Durbin-Watson stat	2.249783		



Correlogram of Residuals

Sample: 2001M02 2011M04  
 Included observations: 123  
 Q-statistic probabilities adjusted for 2 ARMA term(s)

Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob
1		-0.128	-0.128	2.0665	
2		0.012	-0.005	2.0842	
3		0.232	0.237	8.9770	0.003
4		-0.162	-0.111	12.376	0.002
5		0.013	-0.029	12.399	0.006
6		0.156	0.119	15.601	0.004
7		-0.139	-0.055	18.168	0.003
8		0.032	-0.015	18.308	0.006
9		0.194	0.162	23.366	0.001
10		-0.147	-0.048	26.323	0.001
11		0.084	0.023	27.287	0.001
12		0.221	0.187	34.072	0.000
13		-0.153	-0.030	37.332	0.000
14		0.024	-0.083	37.411	0.000
15		-0.051	-0.152	37.788	0.000
16		-0.073	0.044	38.544	0.000
17		-0.001	-0.051	38.544	0.001
18		-0.033	-0.079	38.706	0.001
19		-0.139	-0.100	41.557	0.001
20		0.103	0.077	43.153	0.001
21		0.033	0.040	43.312	0.001
22		-0.193	-0.160	48.972	0.000
23		0.218	0.158	56.288	0.000
24		-0.204	-0.188	62.736	0.000
25		-0.091	-0.069	64.034	0.000
26		0.087	0.023	65.235	0.000
27		-0.231	-0.078	73.750	0.000
28		0.057	0.049	74.276	0.000
29		0.004	-0.099	74.278	0.000
30		-0.134	0.017	77.241	0.000

20. Consider the following process:  $y_t = 2.5 + 0.75y_{t-1} + \varepsilon_t + 0.6\varepsilon_{t-1} - 0.3\varepsilon_{t-2}$  where  $\varepsilon_t \sim WN(0, \sigma_\varepsilon^2)$ :
- Given  $y_n = 12, \hat{\varepsilon}_n = 1.5$  and  $\hat{\varepsilon}_{n-1} = 1$  obtain point forecasts for the next 3 periods.
  - Characterize the forecasting function,  $f_{t,n}$  and the prediction error variance,  $\sigma^2_{n+h|n}$  for the long run (when  $h \rightarrow \infty$ )
21. Given the following estimation outputs what model you think is best to describe and forecast the AIRPASS time series?

M1	M2																																																																																																																																																																
<p>Dependent Variable: D(LOG(AIRPASS),1,12) Method: Least Squares Sample (adjusted): 1951M02 1961M06 Included observations: 125 after adjustments Convergence achieved after 8 iterations MA Backcast: 1950M01 1951M01</p> <table border="1"> <thead> <tr> <th>Variable</th> <th>Coefficient</th> <th>Std. Error</th> <th>t-Statistic</th> <th>Prob.</th> </tr> </thead> <tbody> <tr> <td>MA(1)</td> <td>-0.394813</td> <td>0.082393</td> <td>-4.792405</td> <td>0.0000</td> </tr> <tr> <td>SMA(12)</td> <td>-0.640659</td> <td>0.072003</td> <td>-8.897624</td> <td>0.0000</td> </tr> </tbody> </table> <p>R-squared 0.369398 Mean dependent var 0.000791 Adjusted R-squared 0.364271 S.D. dependent var 0.046431 S.E. of regression 0.037021 Akaike info criterion -3.739810 Sum squared resid 0.168675 Schwarz criterion -3.693557 Log likelihood 235.6756 Hannan-Quinn criter. -3.720426 Durbin-Watson stat 1.934731</p>	Variable	Coefficient	Std. Error	t-Statistic	Prob.	MA(1)	-0.394813	0.082393	-4.792405	0.0000	SMA(12)	-0.640659	0.072003	-8.897624	0.0000	<p>Dependent Variable: D(LOG(AIRPASS),1,12) Method: Least Squares Sample (adjusted): 1951M02 1961M06 Included observations: 125 after adjustments Convergence achieved after 10 iterations MA Backcast: 1949M11 1951M01</p> <table border="1"> <thead> <tr> <th>Variable</th> <th>Coefficient</th> <th>Std. Error</th> <th>t-Statistic</th> <th>Prob.</th> </tr> </thead> <tbody> <tr> <td>MA(1)</td> <td>-0.388933</td> <td>0.082464</td> <td>-4.716423</td> <td>0.0000</td> </tr> <tr> <td>MA(3)</td> <td>-0.190940</td> <td>0.083017</td> <td>-2.300027</td> <td>0.0231</td> </tr> <tr> <td>SMA(12)</td> <td>-0.675069</td> <td>0.069970</td> <td>-9.661783</td> <td>0.0000</td> </tr> </tbody> </table> <p>R-squared 0.392082 Mean dependent var 0.000791 Adjusted R-squared 0.382116 S.D. dependent var 0.046431 S.E. of regression 0.036497 Akaike info criterion -3.759444 Sum squared resid 0.162511 Schwarz criterion -3.691564 Log likelihood 237.9652 Hannan-Quinn criter. -3.731868 Durbin-Watson stat 1.959840</p>	Variable	Coefficient	Std. Error	t-Statistic	Prob.	MA(1)	-0.388933	0.082464	-4.716423	0.0000	MA(3)	-0.190940	0.083017	-2.300027	0.0231	SMA(12)	-0.675069	0.069970	-9.661783	0.0000																																																																																																																													
Variable	Coefficient	Std. Error	t-Statistic	Prob.																																																																																																																																																													
MA(1)	-0.394813	0.082393	-4.792405	0.0000																																																																																																																																																													
SMA(12)	-0.640659	0.072003	-8.897624	0.0000																																																																																																																																																													
Variable	Coefficient	Std. Error	t-Statistic	Prob.																																																																																																																																																													
MA(1)	-0.388933	0.082464	-4.716423	0.0000																																																																																																																																																													
MA(3)	-0.190940	0.083017	-2.300027	0.0231																																																																																																																																																													
SMA(12)	-0.675069	0.069970	-9.661783	0.0000																																																																																																																																																													
<p>Sample: 1951M02 1961M06 Included observations: 125</p> <table border="1"> <thead> <tr> <th>Autocorrelation</th> <th>Partial Correlation</th> <th>AC</th> <th>PAC</th> <th>Q-Stat</th> </tr> </thead> <tbody> <tr><td>   </td><td>   </td><td>1 0.030 0.030 0.1118</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>2 0.021 0.020 0.1679</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>3 -0.144 -0.146 2.8815</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>4 -0.124 -0.118 4.8944</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>5 0.036 0.050 5.0683</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>6 0.038 0.022 5.2572</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>7 -0.051 -0.093 5.6111</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>8 -0.050 -0.054 5.9560</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>9 0.090 0.123 7.0687</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>10 -0.081 -0.102 7.9664</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>11 0.027 -0.016 8.0647</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>12 0.018 0.053 8.1093</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>13 0.055 0.069 8.5337</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>14 0.034 -0.010 8.6955</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>15 0.046 0.049 9.0060</td><td></td><td></td></tr> </tbody> </table>	Autocorrelation	Partial Correlation	AC	PAC	Q-Stat			1 0.030 0.030 0.1118					2 0.021 0.020 0.1679					3 -0.144 -0.146 2.8815					4 -0.124 -0.118 4.8944					5 0.036 0.050 5.0683					6 0.038 0.022 5.2572					7 -0.051 -0.093 5.6111					8 -0.050 -0.054 5.9560					9 0.090 0.123 7.0687					10 -0.081 -0.102 7.9664					11 0.027 -0.016 8.0647					12 0.018 0.053 8.1093					13 0.055 0.069 8.5337					14 0.034 -0.010 8.6955					15 0.046 0.049 9.0060			<p>Sample: 1951M02 1961M06 Included observations: 125</p> <table border="1"> <thead> <tr> <th>Autocorrelation</th> <th>Partial Correlation</th> <th>AC</th> <th>PAC</th> <th>Q-Stat</th> </tr> </thead> <tbody> <tr><td>   </td><td>   </td><td>1 0.015 0.015 0.0287</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>2 0.025 0.025 0.1121</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>3 0.033 0.032 0.2537</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>4 -0.083 -0.084 1.1514</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>5 0.064 0.065 1.6894</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>6 0.067 0.069 2.2913</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>7 -0.081 -0.083 3.1658</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>8 -0.050 -0.063 3.4986</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>9 0.103 0.120 4.9397</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>10 -0.087 -0.079 5.9877</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>11 0.024 -0.002 6.0648</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>12 0.049 0.050 6.4049</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>13 0.017 0.058 6.4473</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>14 0.037 -0.001 6.6433</td><td></td><td></td></tr> <tr><td>   </td><td>   </td><td>15 0.039 0.025 6.8612</td><td></td><td></td></tr> </tbody> </table>	Autocorrelation	Partial Correlation	AC	PAC	Q-Stat			1 0.015 0.015 0.0287					2 0.025 0.025 0.1121					3 0.033 0.032 0.2537					4 -0.083 -0.084 1.1514					5 0.064 0.065 1.6894					6 0.067 0.069 2.2913					7 -0.081 -0.083 3.1658					8 -0.050 -0.063 3.4986					9 0.103 0.120 4.9397					10 -0.087 -0.079 5.9877					11 0.024 -0.002 6.0648					12 0.049 0.050 6.4049					13 0.017 0.058 6.4473					14 0.037 -0.001 6.6433					15 0.039 0.025 6.8612		
Autocorrelation	Partial Correlation	AC	PAC	Q-Stat																																																																																																																																																													
		1 0.030 0.030 0.1118																																																																																																																																																															
		2 0.021 0.020 0.1679																																																																																																																																																															
		3 -0.144 -0.146 2.8815																																																																																																																																																															
		4 -0.124 -0.118 4.8944																																																																																																																																																															
		5 0.036 0.050 5.0683																																																																																																																																																															
		6 0.038 0.022 5.2572																																																																																																																																																															
		7 -0.051 -0.093 5.6111																																																																																																																																																															
		8 -0.050 -0.054 5.9560																																																																																																																																																															
		9 0.090 0.123 7.0687																																																																																																																																																															
		10 -0.081 -0.102 7.9664																																																																																																																																																															
		11 0.027 -0.016 8.0647																																																																																																																																																															
		12 0.018 0.053 8.1093																																																																																																																																																															
		13 0.055 0.069 8.5337																																																																																																																																																															
		14 0.034 -0.010 8.6955																																																																																																																																																															
		15 0.046 0.049 9.0060																																																																																																																																																															
Autocorrelation	Partial Correlation	AC	PAC	Q-Stat																																																																																																																																																													
		1 0.015 0.015 0.0287																																																																																																																																																															
		2 0.025 0.025 0.1121																																																																																																																																																															
		3 0.033 0.032 0.2537																																																																																																																																																															
		4 -0.083 -0.084 1.1514																																																																																																																																																															
		5 0.064 0.065 1.6894																																																																																																																																																															
		6 0.067 0.069 2.2913																																																																																																																																																															
		7 -0.081 -0.083 3.1658																																																																																																																																																															
		8 -0.050 -0.063 3.4986																																																																																																																																																															
		9 0.103 0.120 4.9397																																																																																																																																																															
		10 -0.087 -0.079 5.9877																																																																																																																																																															
		11 0.024 -0.002 6.0648																																																																																																																																																															
		12 0.049 0.050 6.4049																																																																																																																																																															
		13 0.017 0.058 6.4473																																																																																																																																																															
		14 0.037 -0.001 6.6433																																																																																																																																																															
		15 0.039 0.025 6.8612																																																																																																																																																															
<p>Forecast: AIRPASSF Actual: AIRPASS Forecast sample: 1961M07 1961M12 Included observations: 6 Root Mean Squared Error 10.43481 Mean Absolute Error 10.19416 Mean Abs. Percent Error 2.071642</p>	<p>Forecast: AIRPASSF Actual: AIRPASS Forecast sample: 1961M07 1961M12 Included observations: 6 Root Mean Squared Error 10.57776 Mean Absolute Error 8.880508 Mean Abs. Percent Error 1.783386</p>																																																																																																																																																																



22. Suppose that the last five observations of a given time series are:

$$x_{96} = 60.4, x_{97} = 58.9, x_{98} = 64.7, x_{99} = 70.4 \text{ and } x_{100} = 62.6.$$

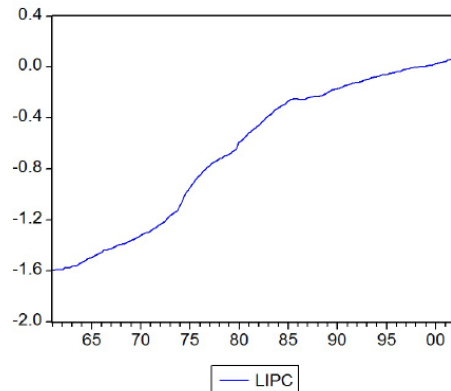
Obtain the forecasts of the next four observations for the following estimated models where

$$\varepsilon_t \sim \text{WN}(0, \sigma_\varepsilon^2):$$

- i.  $(1 - 0.43L)(1 - L)x_t = \varepsilon_t$
- ii.  $x_t = 0.2 + 1.8x_{t-1} - 0.81x_{t-2} + \varepsilon_t$
- iii.  $(1 - 1.4L + 0.8L^2)(1 - L)x_t = \varepsilon_t$

*Exercise 4 (chapter 10 Textbook)*

23. This exercise makes use of quarterly data from the Belgium Consumer Price Index. The data has been seasonally adjusted and covers the period 1961Q01-2002Q02. In the next figure you will find the time series plot, in logs. The time series will be denoted mathematically as  $\log(IPC_t)$  and in EViews output by LIPC.



- a. Before proposing a model, the practitioner needs to study the stationary properties of the data. A possibility is to apply the Augmented Dickey- Fuller (ADF) test to the log of the time series of interest. What made the practitioner apply the log transformation to the time series? And what are the issues that the practitioner needs to be worry to apply the ADF correctly? Justify your answers
- b. According to all the figures given below apply ADF tests to the time series  $\log(IPC_t)$ . In particular, indicate the null and alternative hypotheses, estimated equations, test statistics, critical regions.

Exogenous: Constant, Linear Trend  
Lag Length: 2 (Automatic based on SIC, MAXLAG=13)

	t-Statistic
Augmented Dickey-Fuller test statistic	-0.171180
Test critical values:	
1% level	-4.015341
5% level	-3.437629
10% level	-3.143037

Augmented Dickey-Fuller Test Equation  
Dependent Variable: D(LIPC)  
Method: Least Squares  
Date: 03/26/13 Time: 10:12  
Sample(adjusted): 1961:4 2002:2  
Included observations: 153 after adjusting endpoints

Variable	Coefficient	Std. Error	t-Statistic	Prob.
LIPC(-1)	-0.000698	0.004079	-0.171180	0.8643
D(LIPC(-1))	0.389586	0.074688	5.216175	0.0000
D(LIPC(-2))	0.340018	0.075088	4.528247	0.0000
C	0.003237	0.007059	0.458514	0.6472
@TREND(1961:1)	-1.11E-05	5.02E-05	-0.221045	0.8253
R-squared	0.486964	Mean dependent var	0.010180	
Adjusted R-squared	0.473976	S.D. dependent var	0.008596	
S.E. of regression	0.006235	Akaike info criterion	-7.287211	
Sum squared resid	0.006141	Schwarz criterion	-7.192311	
Log likelihood	598.9077	F-statistic	37.49267	
Durbin-Watson stat	2.148955	Prob(F-statistic)	0.000000	

### (a) ADF test with constant and trend

Exercises 8 and 9 (chapter 10 Textbook)

- Consider the process ARIMA(1,1,0) with  $\phi = 0.9$ . Given  $y_n = 100$  and  $y_{n-1} = 120$  obtain point forecasts for the next two periods. Comment on the behavior of the forecasting function  $f_{t,n}$  and the prediction error variance,  $\sigma^2_{n+h|n}$  for the long run (when  $h \rightarrow \infty$ ).
- Given that  $X_t$  follows a process ARMA(0,2,1) with  $\theta = 0.9$  and that the last observed data is  $X_t = 500, X_{t-1} = 490, \hat{\varepsilon}_t = -10$  and  $\hat{\varepsilon}_{t-1} = 2$  obtain point forecasts for the next 4 periods.
- According to the simple Keynesian model, the relation between consumption,  $C_t$ , and disposable income,  $Y_t$ , can be represented by a linear function:

$$\log(C_t) = \beta_0 + \beta_1 \log(Y_t) + \varepsilon_t \quad (1)$$

Where  $\beta_1$  is the marginal propensity to consume, a quantity of substantial interest and  $\varepsilon_t \sim WN(0, \sigma_\varepsilon^2)$ :

In the next figures you will find the graphical representation of the series of Consumption Expenditure (left) and Disposable Income (right) in the US, at constant prices and in logs and the estimation EViews output of equation (1).

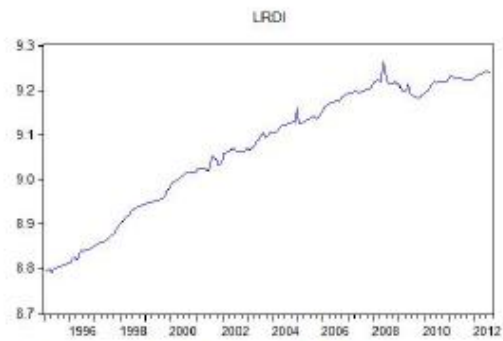
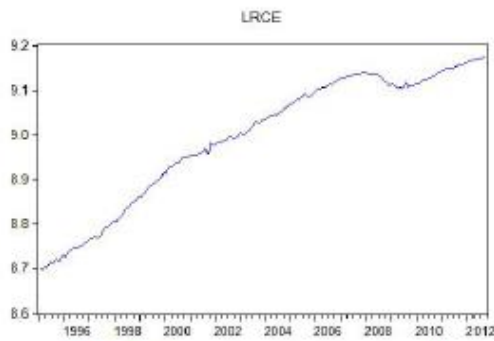
Exogenous: Constant  
Lag Length: 2 (Automatic based on SIC, MAXLAG=13)

	t-Statistic
Augmented Dickey-Fuller test statistic	-1.783204
Test critical values:	
1% level	-3.470679
5% level	-2.879155
10% level	-2.576241

Augmented Dickey-Fuller Test Equation  
Dependent Variable: D(LIPC)  
Method: Least Squares  
Date: 03/26/13 Time: 10:09  
Sample(adjusted): 1961:4 2002:2  
Included observations: 163 after adjusting endpoints

Variable	Coefficient	Std. Error	t-Statistic	Prob.
LIPC(-1)	-0.001578	0.000885	-1.783204	0.0765
D(LIPC(-1))	0.391527	0.073948	5.294619	0.0000
D(LIPC(-2))	0.343006	0.073640	4.657877	0.0000
C	0.001689	0.000900	1.877070	0.0623
R-squared	0.486805	Mean dependent var	0.010180	
Adjusted R-squared	0.477123	S.D. dependent var	0.008596	
S.E. of regression	0.006216	Akaike info criterion	-7.299172	
Sum squared resid	0.006143	Schwarz criterion	-7.223251	
Log likelihood	598.8825	F-statistic	50.27468	
Durbin-Watson stat	2.151675	Prob(F-statistic)	0.000000	

### (b) ADF test with constant



Dependent Variable: LRCE  
 Method: Least Squares  
 Date: 11/07/12 Time: 14:50  
 Sample: 1995M01 2012M09  
 Included observations: 213

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-0.341302	0.056125	-6.081114	0.0000
LRDI	1.029216	0.006187	166.3432	0.0000
R-squared	0.992432	Mean dependent var	8.993582	
Adjusted R-squared	0.992396	S.D. dependent var	0.144663	
S.E. of regression	0.012615	Akaike info criterion	-5.898585	
Sum squared resid	0.033576	Schwarz criterion	-5.867024	
Log likelihood	630.1994	Hannan-Quinn criter.	-5.885830	
F-statistic	27670.05	Durbin-Watson stat	0.353176	
Prob(F-statistic)	0.000000			

- Using the available information, what can you conclude from the marginal propensity to consume? Motivate your answer.
- The following figure depicts the result of the ADF test applied to the time series  $\log(C_t)$ . For this time series should you apply the ADF test with a constant and a trend or only the constant term? Justify your answer.
- According to the EViews output, is it possible to conclude that  $\log(C_t)$  has a unit root? Indicate the null and alternative hypotheses, test statistic, significance level, critical value and the test conclusion.

Null Hypothesis: LRCE has a unit root  
 Exogenous: Constant, Linear Trend  
 Lag Length: 1 (Automatic - based on SIC, maxlag=14)

	t-Statistic
Augmented Dickey-Fuller test statistic	-0.616883
Test critical values:	
1% level	-4.002142
5% level	-3.431265
10% level	-3.139292

\*Mackinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation  
 Dependent Variable: D(LRCE)  
 Method: Least Squares  
 Date: 11/07/12 Time: 16:05  
 Sample (adjusted): 1995M03 2012M09  
 Included observations: 211 after adjustments

Variable	Coefficient	Std. Error	t-Statistic
LRCE(-1)	-0.003970	0.006436	-0.616883
D(LRCE(-1))	-0.228968	0.067366	-3.398880
C	0.039599	0.056317	0.703143
@TREND(1995M01)	-1.04E-05	1.52E-05	-0.683498
R-squared	0.113574	Mean dependent var	
Adjusted R-squared	0.100727	S.D. dependent var	
S.E. of regression	0.003699	Akaike info criterion	
Sum squared resid	0.002832	Schwarz criterion	
Log likelihood	884.1650	Hannan-Quinn criter.	
F-statistic	8.840660	Durbin-Watson stat	
Prob(F-statistic)	0.000015		

*Exercises 5 and 6 (chapter 13 Textbook)*

*Exercises 5 and 6 (chapter 14 Textbook)*

27. For each of the following ARCH/GARCH models decide if they are stationary, compute the unconditional variance and obtain the prediction of the conditional variance for the next three periods:

a.  $\sigma_t^2 = \alpha_0 + 0.65e_{t-1}^2 + 0.25e_{t-2}^2 + 0.10e_{t-3}^2$

b.  $\sigma_t^2 = \alpha_0 + 0.20e_{t-1}^2 + 0.20e_{t-2}^2 + 0.50\sigma_{t-1}^2$

c.  $\sigma_t^2 = \alpha_0 + 0.10e_{t-1}^2 + 0.20e_{t-2}^2 + 0.60e_{t-3}^2$

d.  $\sigma_t^2 = \alpha_0 + 0.10e_{t-1}^2 + 0.90\sigma_{t-1}^2$

28. Given the correlograms, presented above, of the log returns of a financial series what stylized characteristics can be observed? Define the order of an ARCH model to fit the conditional variance of the series of returns.

Correlogram of $(1 - B)\ln(y_t)$						Correlogram of squared $(1 - B)\ln(y_t)$					
Included observations: 952						Included observations: 952					
Autocorrelation	Partial Correlation	AC	PAC	Q-Stat		Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	
		1	-0.001	-0.001	0.0015			1	0.011	0.011	0.1093
		2	0.007	0.007	0.0436			2	0.126	0.126	15.376
		3	0.037	0.037	1.3636			3	0.124	0.124	30.202
		4	0.042	0.042	3.0663			4	0.097	0.083	39.244
		5	0.059	0.059	6.4413			5	0.162	0.138	64.519
		6	-0.041	-0.042	8.0183			6	0.057	0.028	67.642
		7	0.027	0.023	8.7098			7	0.060	0.008	71.051
		8	-0.018	-0.024	9.0378			8	0.081	0.033	77.313
		9	-0.022	-0.025	9.5155			9	0.029	-0.011	78.122
		10	-0.018	-0.020	9.8171			10	0.062	0.016	81.842
		11	0.041	0.046	11.401			11	0.048	0.021	84.098
		12	-0.084	-0.085	18.152			12	0.067	0.042	88.450
		13	-0.002	0.005	18.157			13	0.076	0.049	94.102
		14	0.027	0.027	18.843			14	0.046	0.022	96.122
		15	0.052	0.057	21.416			15	0.091	0.056	104.08
		16	-0.030	-0.030	22.262			16	0.002	-0.036	104.08
		17	-0.010	0.001	22.369			17	0.073	0.026	109.27
		18	-0.008	-0.025	22.424			18	0.003	-0.037	109.28
		19	-0.058	-0.059	25.674			19	0.087	0.055	116.63
		20	0.042	0.040	27.426			20	0.077	0.053	122.48

29. True or False? Correct the sentence and justify when appropriate.

- Volatility clustering is one of the most prominent features of financial returns. Time series analysis reproduces this stylized fact using the ARMA model with white noise errors.
- An ARCH(2) model is equivalent to an AR(2) model for the squared returns.
- A GARCH(1,1) model is equivalent to a MA(2) model for the squared returns.
- The ARMA-GARCH model only generates forecasts for the variance.
- Usually the final ARMA-GARCH model uses the same ARMA model that was fitted before modelling the volatility.
- The GARCH model is able to describe adequately the dynamic properties of volatility of standard financial time series with less parameters than the ARCH model.

30. Consider the estimation output presented below:

Dependent Variable: D(LOG(PB))				
Method: ML - ARCH (Marquardt) - Normal distribution				
Sample (adjusted): 1/04/2000 4/04/2011				
Included observations: 2935 after adjustments				
Convergence achieved after 10 iterations				
Presample variance: backcast (parameter = 0.7)				
GARCH = C(2) + C(3)*RESID(-1)^2 + C(4)*RESID(-2)^2 + C(5)*RESID(-3)^2 + C(6)*RESID(-4)^2 + C(7)*RESID(-5)^2 + C(8)*RESID(-6)^2				
Variable	Coefficient	Std. Error	z-Statistic	Prob.
C	0.001321	0.000396	3.337769	0.0008
Variance Equation				
C	0.000280	1.41E-05	19.87422	0.0000
RESID(-1)^2	0.089043	0.014307	6.223588	0.0000
RESID(-2)^2	0.068808	0.017353	3.965242	0.0001
RESID(-3)^2	0.082096	0.017453	4.703768	0.0000
RESID(-4)^2	0.100291	0.016906	5.932440	0.0000
RESID(-5)^2	0.092199	0.019404	4.751628	0.0000
RESID(-6)^2	0.054665	0.016496	3.313866	0.0009
R-squared	-0.001094	Mean dependent var		0.000543
Adjusted R-squared	-0.003488	S.D. dependent var		0.023535
S.E. of regression	0.023576	Akaike info criterion		-4.754119
Sum squared resid	1.626930	Schwarz criterion		-4.737807
Log likelihood	6984.670	Hannan-Quinn criter.		-4.748245
Durbin-Watson stat	1.952501			

- Write explicitly the estimated equation.
- Obtain the estimate for the unconditional variance of the error of the series.
- Comment on the correlogram of the standardized squared residuals presented below.

Correlogram of Standardized Residuals Squared						
Sample: 1/04/2000 4/04/2011						
Included observations: 2935						
Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob	
		1	0.010	0.010	0.3221	0.570
		2	-0.010	-0.010	0.6151	0.735
		3	-0.018	-0.018	1.5714	0.666
		4	-0.030	-0.030	4.2216	0.377
		5	-0.014	-0.014	4.8152	0.439
		6	-0.026	-0.026	6.7396	0.346
		7	0.026	0.025	8.7758	0.269
		8	0.040	0.037	13.388	0.099
		9	0.077	0.076	30.951	0.000
		10	0.031	0.030	33.817	0.000
		11	0.015	0.018	34.440	0.000
		12	0.056	0.062	43.776	0.000
		13	0.014	0.022	44.325	0.000
		14	0.042	0.050	49.509	0.000
		15	0.031	0.038	52.422	0.000
		16	0.049	0.052	59.550	0.000
		17	0.027	0.027	61.649	0.000
		18	-0.012	-0.011	62.080	0.000
		19	0.009	0.008	62.325	0.000
		20	0.009	0.007	62.549	0.000
		21	0.042	0.034	67.804	0.000
		22	0.044	0.037	73.525	0.000
		23	0.013	0.002	74.047	0.000
		24	0.036	0.023	77.784	0.000
		25	0.015	0.006	78.462	0.000
		26	0.045	0.039	84.482	0.000
		27	0.029	0.029	87.061	0.000
		28	0.040	0.036	91.834	0.000
		29	0.030	0.024	94.528	0.000

32. Suppose that the return series of a given stock,  $r_t$ , is well described by the following model:

$$r_t = \varepsilon_t = \sigma_t z_t, \quad z_t \stackrel{iid}{\rightarrow} D(0,1)$$

$$\sigma_t^2 = 1 + 0.4\varepsilon_t^2 + 0.2\varepsilon_{t-1}^2 + 0.3\varepsilon_{t-2}^2$$

- Derive the forecast equations that are used to obtain the forecasts,  $\sigma_{T+s|T}^2$  with origin at time T.
- Suppose that the last two observations for the returns are  $r_{T-1} = 0.03$  and  $r_T = 0.06$ . Using  $\sigma_T^2 = 1$  obtain the forecasts for  $\sigma_{T+1}^2$ ,  $\sigma_{T+2}^2$  and  $\sigma_{T+3}^2$  with origin at time T.