THE EVOLUTION OF SNP PETROM STOCK LIST - STUDY THROUGH AUTOREGRESSIVE MODELS

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Abstract

Stock exchange market is one of the most dynamic and unpredictable markets. In this context, this work intends to analyze the SNP Petrom shares on the REGS market, based on the chronological series.

The economic series are often not stationary, but they can be stationarized by different data transformations. The simplest method used for stationarizing a series is to apply differentiating operators of various classes on the series. After applying this operator, a stationary series that can be modified by an ARIMA (p,q) process is usually obtained.

Most time series with economic content include a seasonal component besides the trend and random component.

The purpose of this work is to estimate the parameters of an ARIMA (p,d,q) model for SNP Petrom shares, where p is the number of autoregressive terms, d is the integration level of the series (how many times the series must be differentiated in order to become stationary) and q is the number of moving average terms (MA).

Keywords: list, economic series, autoregressive models

Introduction

In literature the determination of the best ARIMA(p,d,q) sample in order to shape certain remarks for a series of time entails an assembly of techniques and methods, better known as the Box-Jenkins methodology.

A process $\{Y_t\}$, t belongs to Z, it admits a representation ARIMA(p,d,q) should this meet the subsequent equality: $\Phi(L)(1-L)^dY_t=\Theta(L)\epsilon_t$, whereas ϵ_t is a white noise, the two polinomes $\Phi(L)=1-\sum \phi_i L^i$, $\Theta(L)=1-\sum \theta_i L^i$ have roots larger than one, as the initial conditions $y_{-p-d},...,y_{-1},\epsilon_{-q},...,\epsilon_{-1}$ are not correlated with the random variables $\epsilon_0,\epsilon_1,...,\epsilon_t,...$

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Building the model with box-jenkins methodology

The Box-Jenkins methodology comprises three main aspects:

- ♣ identification:
- estimate:
- . checking.

Sample identification

Having available the sample of remarks on the evolution of SNP Petrom share quotation, a series of transformations must be brought to these so as to induce stationarity.

In case of time series describing the processes on the financial market, a scale transformation appears necessary, whereas most of the time the initial i series is being applied a logarithmic filter, in order to have a stationary series.

The next step is the elimination of the determinist component, after finding the possible oscillations present in the evolution of the series (Figure 1.).

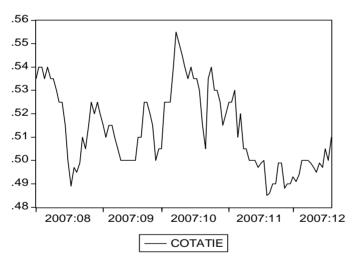


Figure 1 – Average price evolution of Petrom SA shares on the market

Currently we are able to determine for which values of the parameters p and q the ARMA(p,q) process shape to the best in the stationary series obtained. A criterion in this regard is the behaviour of the autocorrelation (ACF) and of the partial autocorrelation (PACF) functions.

Corelograma p_RRC

Included observations: 489

Autocorrelation	Partial Correlation		AC	PAC	Q-Stat	Prob
. ******	. ******	1	0.978	0.978	470.55	0.000
. ******	* .	2	0.951	-0.126	916.31	0.000
. ******	. .	3	0.924	0.001	1338.0	0.000
. ******	.].	4	0.900	0.048	1738.6	0.000
. * * * * * * *	. .	5	0.874	-0.054	2117.6	0.000
. * * * * * * *	. .	6	0.848	-0.020	2475.0	0.000
. *****	. .	7	0.824	0.045	2813.3	0.000
. *****	* .	8	0.798	-0.076	3131.5	0.000
. *****	. .	9	0.771	-0.031	3429.1	0.000
. *****	. .	10	0.745	0.011	3707.1	0.000
. *****	. .	11	0.720	0.018	3967.8	0.000
. ****	. .	12	0.697	0.005	4212.5	0.000
. ****	. .	13	0.676	0.025	4442.7	0.000
. ****	. .	14	0.653	-0.038	4658.3	0.000
. ****	. .	15	0.633	0.050	4861.6	0.000
. ****	. .	16	0.613	-0.046	5052.1	0.000
. *****	. .	17	0.592	0.005	5230.5	0.000
. ****	. .	18	0.572	-0.009	5397.3	0.000
. ****	. .	19	0.553	0.004	5553.5	0.000
. ****	. *	20	0.539	0.092	5702.0	0.000
. ****	. .	21	0.525	-0.018	5843.2	0.000
. ****	. *	22	0.514	0.079	5979.2	0.000
. ****	. .	23	0.504	-0.016	6110.1	0.000
. ****	. .	24	0.495	0.020	6236.8	0.000
. * * * *	. .	25	0.485	-0.030	6358.5	0.000
. * * * *	. .	26	0.476	0.014	6475.8	0.000
. ****	. .	27	0.466	-0.012	6588.7	0.000
. ***	* .	28	0.455	-0.058	6696.6	0.000
. ***	. .	29	0.443	-0.003	6799.1	0.000
. ***	. .	30	0.433	0.020	6897.1	0.000
. ***	. .	31	0.424	0.024	6991.2	0.000
. ***	.j.	32	0.417	0.051	7082.5	0.000
. ***	.j.	33	0.409	-0.029	7170.7	0.000
. ***	.j.	34	0.403	0.045	7256.4	0.000
. ***	.j.	35	0.397	-0.005	7339.9	0.000
. ***	. *	36	0.394	0.075	7422.3	0.000

We can see that ACF decreases very slowly (up to 36 lags are statistically significant), as PACF dramatically decreases after the first lag. ACF suggests that the series of prices is not stationary, and it must be differentiated before applying the Box-Jenkins methodology. The test for the unit-root Dickey Fuller set out below proves that our series is actually integrated of order 1 (and not more).

Null Hypothesis: P_RRC has a unit root

Exogenous: Constant

Lag Length: 1 (Automatic based on SIC, MAXLAG=17)

		t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic		-2.719685	0.0714
Test critical			
values:	1% level	-3.443551	
	5% level	-2.867255	
	10% level	-2.569876	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation

Dependent Variable: D(P_RRC)

Method: Least Squares

Sample (adjusted): 1/04/2006 11/15/2007 Included observations: 487 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
P_RRC(-1) D(P_RRC(-1))	-0.024636 0.130416	0.009059 0.044974	-2.719685 2.899791	0.0068 0.0039
C C	0.002333	0.000881	2.648450	0.0084
R-squared Adjusted R-	0.029038	Mean d	ependent var	-4.52E-05
squared S.E. of	0.025026	S.D. de	ependent var	0.002704
regression Sum squared	0.002670	Akaike	info criterion	-9.007141
resid	0.003451	Schwa	arz criterion	-8.981341
Log likelihood Durbin-	2196.239	F-	statistic	7.237291
Watson stat	2.000354	Prob((F-statistic)	0.000800

Null Hypothesis: D(P RRC) has a unit root

Exogenous: Constant

Lag Length: 0 (Automatic based on SIC, MAXLAG=17)

		t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic		-19.54109	0.0000
Test critical			
values:	1% level	-3.443551	
	5% level	-2.867255	
	10% level	-2.569876	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation

Dependent Variable: D(P_RRC,2)

Method: Least Squares

Sample (adjusted): 1/04/2006 11/15/2007

Variable	Coefficient	Std. Error	t-Statistic	Prob.
D(P_RRC(- 1))	-0.880858	0.045077	-19.54109	0.0000
C	-4.00E-05	0.000122	-0.328679	0.7425
			Mean dependent	
R-so	quared	0.440506	var	-2.05E-06
Adjusted	R-squared	0.439352	S.D. dependent var	0.003590
			Akaike info	
S.E. of	regression	0.002688	criterion	-8.996081
Sum squ	Sum squared resid		Schwarz criterion	-8.978881
Log likelihood		2192.546	F-statistic	381.8541
Durbin-Watson stat		1.996914	Prob(F-statistic)	0.000000

After having established that the series is integrated of order 1, we are interested in ACF and PACF for the first difference d(p RRC).

Sample: 1/02/2006 1/18/2008 Included observations: 488

Autocorrelation	Partial Correlation		AC	PAC	Q-Stat	Prob
. * . . * .	.j. j	2	0.002	0.119 -0.012 -0.062	6.9716	0.031

	. .	1	. .	1	4	0.021	0.036	9.1067	0.058
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* . * . 14 -0.077 -0.081 18.311 0.193 15 0.009 0.021 18.352 0.245 16 -0.009 -0.013 18.392 0.301 17 0.007 0.004 18.419 0.363 18 -0.035 -0.033 19.053 0.389 * . 19 -0.107 -0.093 24.885 0.164 . . 20 -0.024 -0.010 25.182 0.195 * . 21 -0.091 -0.089 29.418 0.104 . . 22 -0.011 -0.005 29.480 0.132 . . 24 0.012 -0.006 29.638 0.197		<u> </u>		i					
				i.	14				
	. .	İ		Ė	15	0.009	0.021	18.352	0.245
		İ		İ	16	-0.009	-0.013	18.392	0.301
		İ		İ	17	0.007	0.004	18.419	0.363
* . * . 19 -0.107 -0.093 24.885 0.164 . . 20 -0.024 -0.010 25.182 0.195 * . 21 -0.091 -0.089 29.418 0.104 . . 22 -0.011 -0.005 29.480 0.132 . . 23 -0.013 -0.008 29.568 0.162 . . 24 0.012 -0.006 29.638 0.197 . . 25 0.004 -0.002 29.648 0.238 . . 26 0.011 0.016 29.716 0.280 . . 27 0.027 0.027 30.096 0.310 . . 28 0.009 -0.001 30.143 0.356 . . 29 -0.010 -0.012 30.192 0.404 . . 30 -0.041 -0.049 31.081 0.411 . . 32 0.033		ĺ		ĺ	18	-0.035	-0.033	19.053	0.389
* . * . 21 -0.091 -0.089 29.418 0.104 . . . 22 -0.011 -0.005 29.480 0.132 . . . 23 -0.013 -0.008 29.568 0.162 . . . 24 0.012 -0.006 29.638 0.197 . . . 25 0.004 -0.002 29.648 0.238 . . . 26 0.011 0.016 29.716 0.280 . . . 27 0.027 0.027 30.096 0.310 . . . 28 0.009 -0.001 30.143 0.356 . . . 29 -0.010 -0.012 30.192 0.404 . . . 30 -0.041 -0.049 31.081 0.411 . . 31 -0.056 -0.066 32.702 0.383 . .				Ì	19	-0.107	-0.093	24.885	0.164
	. .	Ī	. .	ĺ	20	-0.024	-0.010	25.182	0.195
	* .		* .		21	-0.091	-0.089	29.418	0.104
	. .				22	-0.011	-0.005	29.480	0.132
		23	-0.013	-0.008	29.568	0.162
		24	0.012	-0.006	29.638	0.197
		25	0.004	-0.002	29.648	0.238
		26	0.011	0.016	29.716	0.280
		27	0.027	0.027	30.096	0.310
		28	0.009	-0.001	30.143	0.356
		29	-0.010	-0.012	30.192	0.404
		30	-0.041	-0.049	31.081	0.411
	. .		* .		31	-0.056	-0.066	32.702	0.383
		32	0.033	0.040	33.260	0.406
* . * . 35 -0.078 -0.102 37.098 0.372		33	0.000	-0.039	33.260	0.455
. . . . 36 -0.015 0.007 37.211 0.413	* .		* .						
		36	-0.015	0.007	37.211	0.413

The new correlogram has by far less statistically significant terms, therefore we should search for a sample of ARIMA (3,1,3) type, and even if we take into account how separate are the significant terms, it is possible that this sample be actually ARIMA (1,1,1).

2.2 Sample estimation

The stage of sample estimation includes the effective use of data to do parameter inferences according to the soundness of the sample. In order to estimate parameters the method of maximum probability also known as the method of maximum likelihood or the method of the least squares can be used.

By using least squares, we have estimated the following model in Eviews: d(p rrc) c ar(1) ar(2) ar(3) ma(1) ma(2) ma(3)

Dependent Variable: D(P RRC)

Method: Least Squares

Sample (adjusted): 1/06/2006 11/15/2007 Included observations: 485 after adjustments Convergence achieved after 78 iterations

Backcast: 1/03/2006 1/05/2006

Variable	Coefficient	Std. Error	t-Statistic	Prob.
С	4.78E-05	3.50E-05	1.364149	0.1732
AR(1)	0.410036	0.217294	1.887007	0.0598
AR(2)	-0.030056	0.259614	-0.115771	0.9079
AR(3)	0.547195	0.171668	3.187527	0.0015
MA(1)	-0.309835	0.203756	-1.520619	0.1290
MA(2)	-0.022812	0.224538	-0.101597	0.9191
MA(3)	-0.656745	0.155067	-4.235223	0.0000
R-squared	0.040818	Mean deper	ndent var	-4.33E-05
Adjusted R-squared	0.028778	S.D. depen		0.002710
S.E. of regression	0.002670	Akaike info	criterion	-8.998945
Sum squared resid	0.003408	Schwarz cr	iterion	-8.938555
Log likelihood	2189.244	F-statistic		3.390230
Durbin-Watson stat	1.979233	Prob(F-stat	istic)	0.002766
Inverted AR Roots	.97	2870i		28+.70i
Inverted MA Roots	1.00	34+.74i		3474i

Taking into account that the terms AR (2) and MA (2) are statistically non-significant, we re-estimate the sample without these:

Dependent Variable: D(P_RRC)

Method: Least Squares

Sample (adjusted): 1/06/2006 11/15/2007 Included observations: 485 after adjustments Convergence achieved after 56 iterations

Backcast: 1/03/2006 1/05/2006

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	4.84E-05	3.79E-05	1.277387	0.2021
AR(1)	0.364003	0.130325	2.793049	0.0054
AR(3)	0.566562	0.125066	4.530114	0.0000
MA(1)	-0.291995	0.113841	-2.564937	0.0106
MA(3)	-0.696925	0.114552	-6.083924	0.0000
R-squared	0.037552	Mean dependent v	ar	-4.33E-05
Adjusted R-squared	0.029532	S.D. dependent va	r	0.002710
S.E. of regression	0.002669	Akaike info criteri	on	-9.003793
Sum squared resid	0.003420	Schwarz criterion		-8.960658
Log likelihood	2188.420	F-statistic		4.682116
Durbin-Watson stat	1.927134	Prob(F-statistic)		0.001026
Inverted AR Roots	97	30+.70i		3070i
Inverted MA Roots	1.00	3576i		35+.76i

In this sample, all coefficients except the constant are statistically significant.

2.3 Sample Checking

This last stage of the Box-Jenkins methodology is at least equally important as identification or estimate stage. The purpose is seeing in what extent the sample built complies with the available observations dealing with the stochastic process studied.

The stage implies testing the sample adjusted in its relation with data in order to discover the inadequacies of the sample and to obtain its improvement.

Taking into account that we have estimated an ARIMA(3,1,3) sample, we are in the first instance interested in knowing if we have eliminated autocorrelation of residuals. The correlogram of residuals (in the object equation -> view -> residual tests -> correlogram Q statistic) proves that there are no more autoregressive statistically significant terms. For verify this assumption we can used the Breusch-Godfrey test.

Breusch-Godfrey Serial Correlation LM Test:

F-statistic	0.796514	Prob. F(2,478)	0.451496
Obs*R-squared	1.413281	Prob. Chi-Square(2)	0.493299

Test Equation:

Dependent Variable: RESID Method: Least Squares

Sample: 1/06/2006 11/15/2007 Included observations: 485

Presample missing value lagged residuals set to zero.

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	1.26E-07	3.79E-05	0.003333	0.9973
AR(1)	-0.082179	0.181701	-0.452276	0.6513
AR(3)	0.070815	0.168008	0.421497	0.6736
MA(1)	0.039016	0.136551	0.285722	0.7752
MA(3)	-0.039871	0.137752	-0.289442	0.7724
RESID(-1)	0.080796	0.078024	1.035525	0.3009
RESID(-2)	-0.014196	0.054266	-0.261598	0.7937
R-squared	0.002914	Mean dependent var		5.37E-05
Adjusted R-squared	-0.009602	S.D. dependent var		0.002658
S.E. of regression	0.002670	Akaike info criterion		-8.998873
Sum squared resid	0.003409	Schwarz criterion		-8.938483
Log likelihood	2189.227	F-statistic		0.232826
Durbin-Watson stat	1.999891	Prob(F-statistic)		0.965813

The assumption can be accepted. Nevertheless, residuals are relatively far from normality, with both excess kurtosis and skewness positive (figure 2).

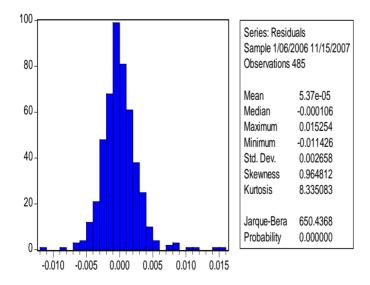


Figure 2 – The residual distribution

The test of double residual autocorrelation (squared residuals) also suggests that the heteroskedasticity hypothesis is not verified, and the ARIMA (3,1,3) sample should be estimated with a ARCH sample for variant, not at all simple least squares.

If we estimate the ARIMA (3,1,3) sample by means of a GARCH (1,1) sample for a variant, results are more encouraging:

Dependent Variable: D(P_RRC)

Method: ML - ARCH (Marquardt) - Normal distribution

Sample (adjusted): 1/06/2006 11/15/2007 Included observations: 485 after adjustments Convergence achieved after 72 iterations

MA backcast: OFF (Roots of MA process too large), Variance backcast: ON

 $GARCH = C(6) + C(7)*RESID(-1)^2 + C(8)*GARCH(-1)$

	Coefficient	Std. Error	z-Statistic	Prob.
C	-1.18E-06	1.42E-05	-0.083107	0.9338
AR(1)	-0.334103	4.52E-05	-7383.989	0.0000
AR(3)	0.805044	0.000128	6272.457	0.0000
MA(1)	0.377929	0.000473	798.9506	0.0000
MA(3)	-0.883694	0.000149	-5921.544	0.0000
	Variance Ed	quation		
C	5.56E-07	1.75E-07	3.179919	0.0015
RESID(-1)^2	0.256283	0.046331	5.531570	0.0000
GARCH(-1)	0.680985	0.056834	11.98200	0.0000
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.076118	Mean dependent var		-4.33E-05
	0.062560	S.D. dependent var		0.002710
	0.002623	Akaike info criterion		-9.276988
	0.003283	Schwarz criterion		-9.207971
	2257.670	F-statistic		5.614269
	1.745316	Prob(F-statistic)		0.000003
Inverted AR Roots Inverted MA Roots	.83 .85 Estimated M	58+.79i 61+.82i AA process is 1	5879i 6182i noninvertible	

Now, the residuals distribution is presented in figure 3

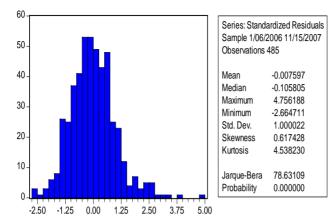


Figure 3 – The residual distribution

Conclusions

ARIMA(3,1,3) sample, possibly with a GARCH (1,1) sample for the variant of residuals, adequately describes the structure of autocorrelation in the field of Rompetrol share prices.

References

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