

## Assignment 14

### ARIMA & GARCH Models

From the data set `assign_timeseries.dta`:

#### Part I

1. Estimate Autoregressive Integrated Moving Average (ARIMA(p,d,q)) model for spot and future – determine the most appropriated order for p, d, and q using SBIC given the maximum lag equals 4.

*According to the unit root test on the last assignment (Assignment13), it suggests that both spot and future are integrated series of order 1, i.e.  $d = 1$*

```
. tsset time
      time variable:  time, 1 to 795
              delta:  1 unit
      . qui arima spot, arima(1,1,1)
. est store arima111
. qui arima spot, arima(1,1,2)
. est store arima112
. qui arima spot, arima(1,1,3)
. est store arima113
. qui arima spot, arima(1,1,4)
. est store arima114
. qui arima spot, arima(2,1,1)
. est store arima211
. qui arima spot, arima(2,1,2)
. est store arima212
. qui arima spot, arima(2,1,3)
. est store arima213
. qui arima spot, arima(2,1,4)
. est store arima214
. qui arima spot, arima(3,1,1)
. est store arima311
. qui arima spot, arima(3,1,2)
. est store arima312
. qui arima spot, arima(3,1,3)
. est store arima313
. qui arima spot, arima(3,1,4)
```

```

. est store arima314
. qui arima spot, arima(4,1,1)
. est store arima411
. qui arima spot, arima(4,1,2)
. est store arima412
. qui arima spot, arima(4,1,3)
. est store arima413
. qui arima spot, arima(4,1,4)
. est store arima414
. est table arima11*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

```

Variable	arima111	arima112	arima113	arima114
-----+-----				
spot				
_cons	-.10471744	-.10468155	-.10474014	-.10717398
-----+-----				
ARMA				
ar				
L1.	.43536547*	.32884791	-.21609401	.59490084
ma				
L1.	-.35351081	-.25287121	.29330109	-.52209152
L2.		.02661686	.06275321	-.00167449
L3.			.07084457**	.03360908
L4.				-.07201064*
-----+-----				
sigma				
_cons	8.1399985***	8.1383248***	8.1283318***	8.1204174***
-----+-----				
Statistics				
N	794	794	794	794
ll	-2791.4965	-2791.3458	-2790.3816	-2789.5823
chi2	18.312657	16.97941	13.218602	28.319443
aic	5590.993	5592.6915	5592.7633	5593.1646
bic	5609.7013	5616.077	5620.8258	5625.9042

-----  
legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

. est table arima21\*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima211	arima212	arima213	arima214
-----+-----				
spot				
_cons	-.10453586	-.10767542	-.10609456	-.10747347
-----+-----				
ARMA				
ar				
L1.	.30210361	1.3247307***	.46523894***	1.3085163***
L2.	.02452891	-.92390998***	-.83397031***	-.92906275***
ma				
L1.	-.22494448	-1.3022249***	-.39216993***	-1.2433923***
L2.		.94779741***	.82697929***	.88679293***
L3.			.09595475***	.05677728
L4.				.00003472
-----+-----				
sigma				
_cons	8.1387104***	8.077039***	8.1149358***	8.0663508***
-----+-----				
Statistics				
N	794	794	794	794
ll	-2791.3835	-2785.5252	-2789.1057	-2784.5179
chi2	16.085901	3656.4671	241.72156	4791.7034
aic	5592.7671	5583.0504	5592.2113	5585.0358
bic	5616.1525	5611.1129	5624.9509	5622.4524

-----  
legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

. est table arima31\*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima311	arima312	arima313	arima314
spot				
_cons	-.10486338	-.10606433	-.10806841	-.10765191
ARMA				
ar				
L1.	-.56058829	.56814277***	.66749387**	.58444973
L2.	.08569307*	-.89762847***	-.06995188	.02671112
L3.	.06374492**	.09752317***	-.59846871*	-.67269639
ma				
L1.	.63856195	-.49161073***	-.61040004*	-.5189438
L2.		.88783708***	.0556601	-.03501278
L3.			.65798111**	.71899171
L4.				.02307553
sigma				
_cons	8.1302906***	8.1136284***	8.0655263***	8.0640793***
Statistics				
N	794	794	794	794
ll	-2790.5308	-2788.9397	-2784.4054	-2784.2793
chi2	21.334447	289.47751	1722.9017	1770.384
aic	5593.0617	5591.8793	5584.8108	5586.5587
bic	5621.1242	5624.6189	5622.2274	5628.6524

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

. est table arima41\*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima411	arima412	arima413	arima414
spot				

_cons		-.10781888	-.10752733	-.10764238	-.10766283
-----					
ARMA					
ar					
L1.		.6019371*	1.3677506***	.61524226	.8391139
L2.		-.00196743	-1.0093802***	.00960034	-.13065856
L3.		.02357655	.05753428	-.67663352	-.67731235
L4.		-.074501**	-.00157074	.02301592	.1703585
ma					
L1.		-.52565375	-1.3026494***	-.54889172	-.77179962
L2.			.964438***	-.01961375	.10834369
L3.				.72233737	.72448925
L4.					-.15754229
-----					
sigma					
_cons		8.1178063***	8.0662434***	8.0638126***	8.0634989***
-----					
Statistics					
N		794	794	794	794
ll		-2789.3297	-2784.5185	-2784.2686	-2784.2263
chi2		30.742143	5306.7984	1879.7454	2719.397
aic		5592.6594	5585.037	5586.5371	5588.4525
bic		5625.399	5622.4537	5628.6309	5635.2234
-----					

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

**From the result, the best ARIMA model (ARIMA(p,d,q)) for spot is ARIMA(1,1,1) since it gives the lowest value of BIC among any other combination of (p,1,q),  $1 \leq p, q \leq 4$**

```
. drop _est_arima111 _est_arima112 _est_arima113 _est_arima114 _est_arima211 _est_arima212
_est_arima213 _est_arima214 _est_arima311 _est_arima312 _est_arima313 _est_arima314 _est_arima411
_est_arima412 _est_arima413 _est_arima414

. qui arima future, arima(1,1,1)

. est store arima111

. qui arima future, arima(1,1,2)

. est store arima112

. qui arima future, arima(1,1,3)

. est store arima113

. qui arima future, arima(1,1,4)

. est store arima114

. qui arima future, arima(2,1,1)

. est store arima211

. qui arima future, arima(2,1,2)

. est store arima212

. qui arima future, arima(2,1,3)

. est store arima213

. qui arima future, arima(2,1,4)

. est store arima214

. qui arima future, arima(3,1,1)

. est store arima311

. qui arima future, arima(3,1,2)

. est store arima312

. qui arima future, arima(3,1,3)

. est store arima313

. qui arima future, arima(3,1,4)

. est store arima314

. qui arima future, arima(4,1,1)

. est store arima411

. qui arima future, arima(4,1,2)

. est store arima412

. qui arima future, arima(4,1,3)

. est store arima413

. qui arima future, arima(4,1,4)

. est store arima414
```

```
. est table arima11*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable	arima111	arima112	arima113	arima114
future				
_cons	-.11424575	-.11408115	-.1139183	-.11370744
ARMA				
ar				
L1.	-.36850497	.30224152	.07471074	-.26441554
ma				
L1.	.33119538	-.33885872	-.11128313	.23060131
L2.		.04456173	.0345267	.02935071
L3.			.03319847	.04074154
L4.				.03843721
sigma				
_cons	9.5427695***	9.5390832***	9.5360585***	9.532552***
Statistics				
N	794	794	794	794
ll	-2917.724	-2917.4236	-2917.166	-2916.8767
chi2	3.3320137	3.4714053	5.5726249	7.6343946
aic	5843.4479	5844.8473	5846.3321	5847.7535
bic	5862.1563	5868.2327	5874.3946	5880.493

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

```
. est table arima21*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable	arima211	arima212	arima213	arima214
future				

	1	2	3	4
_cons	-.11399264	-.11468686	-.11518755	-.114745
-----+-----				
ARMA				
ar				
L1.	.26006773	1.2835845***	.39680958***	.29779715***
L2.	.03927244	-.90741041***	-.83492246***	-.89274399***
ma				
L1.	-.29537225	-1.2846838***	-.43709178***	-.33588403***
L2.		.95026799***	.86844397***	.9442071***
L3.			.02125732	.00060118
L4.				.05136959*
-----+-----				
sigma				
_cons	9.5398841***	9.4618855***	9.4808276***	9.473342***
-----+-----				
Statistics				
N	794	794	794	794
ll	-2917.4896	-2911.1516	-2912.6217	-2912.0146
chi2	2.8388146	6764.7279	566.77521	690.44183
aic	5844.9792	5834.3032	5839.2433	5840.0292
bic	5868.3646	5862.3657	5871.9829	5877.4458

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

```
.
. est table arima31*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable	arima311	arima312	arima313	arima314
-----+-----				
future				
_cons	-.11395314	-.11434855	-.11482479	-.11489771
-----+-----				
ARMA				
ar				



L1.		.06416111	1.3204613***	-.56192851***	.01972589
L2.		.02920804	-.90163025***	-.42025281***	-.77499053***
L3.		.02294871	-.05791248**	-.8338931***	-.25249026
ma					
L1.		-.10003354	-1.3735497	.51793325	-.05826621
L2.			1.0000008	.40777617	.81652449***
L3.				.8898383	.26256083
L4.					.05469308*
-----+					
sigma					
_cons		9.5380895***	9.4185571	9.4577904	9.4722994***
-----+					
Statistics					
N		794	794	794	794
ll		-2917.3418	-2909.7635	-2911.5297	-2911.9403
chi2		3.8262796	32923.58	7392.6231	480.87615
aic		5846.6836	5833.5271	5839.0595	5841.8806
bic		5874.7461	5866.2666	5876.4761	5883.9743

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

```
. est table arima41*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable		arima411	arima412	arima413	arima414
-----+					
future					
_cons		-.11336618	-.11461152	-.11590868	-.11376705
-----+					
ARMA					
ar					
L1.		-.26142231	.3181852***	1.7676398***	-.41944632***
L2.		.01694481	-.85050807***	-1.50548*	.36718541***
L3.		.03124469	.00130972	.41843716	-.35522863***

L4.		.0307881	.04773204	.00058669	-.86090936***
ma					
L1.		.22521179	-.35750674***	-1.8167691***	.38743681***
L2.			.90090477***	1.6207448*	-.30501342***
L3.				-.4995254	.42560261***
L4.					.8658178***
-----+-----					
sigma					
_cons		9.5350812***	9.4739792***	9.4436415***	9.4058783***
-----+-----					
Statistics					
N		794	794	794	794
ll		-2917.0927	-2912.0811	-2909.6096	-2906.4848
chi2		5.4882963	661.06102	4207.3322	932.73412
aic		5848.1853	5840.1622	5837.2193	5832.9696
bic		5880.9249	5877.5789	5879.313	5879.7404
-----+-----					

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

***From the result, the best ARIMA model (ARIMA(p,d,q)) for future is ARIMA(1,1,1) since it gives the lowest value of BIC among any other combination of (p,1,q), 1 ≤ p,q ≤ 4***

## 2. Make dynamic forecast for period time = 796 to 800.

```
. arima spot, arima(1,1,1)
```

```
(setting optimization to BHHH)
```

```
Iteration 0: log likelihood = -2791.825
```

```
. . .
```

```
Iteration 7: log likelihood = -2791.4965
```

```
ARIMA regression
```

```
Sample: 2 - 795                                Number of obs   =           794
                                                Wald chi2(2)    =           18.31
Log likelihood = -2791.496                    Prob > chi2     =           0.0001
```

```
-----+-----
          |               OPG
          |   Coef.   Std. Err.   z   P>|z|   [95% Conf. Interval]
-----+-----
spot     |
   _cons |  -.1047174   .3525647   -0.30   0.766   - .7957316   .5862967
-----+-----
ARMA     |
   ar    |
   L1.   |   .4353655   .2343856    1.86   0.063   - .0240219   .8947528
          |
   ma    |
   L1.   |  -.3535108   .242659   -1.46   0.145   - .8291138   .1220922
-----+-----
   /sigma |   8.139999   .1507405   54.00   0.000    7.844553   8.435444
-----+-----
```

```
Note: The test of the variance against zero is one sided, and the two-sided
       confidence interval is truncated at zero.
```

```
. set obs 800
```

```
number of observations (_N) was 795, now 800
```

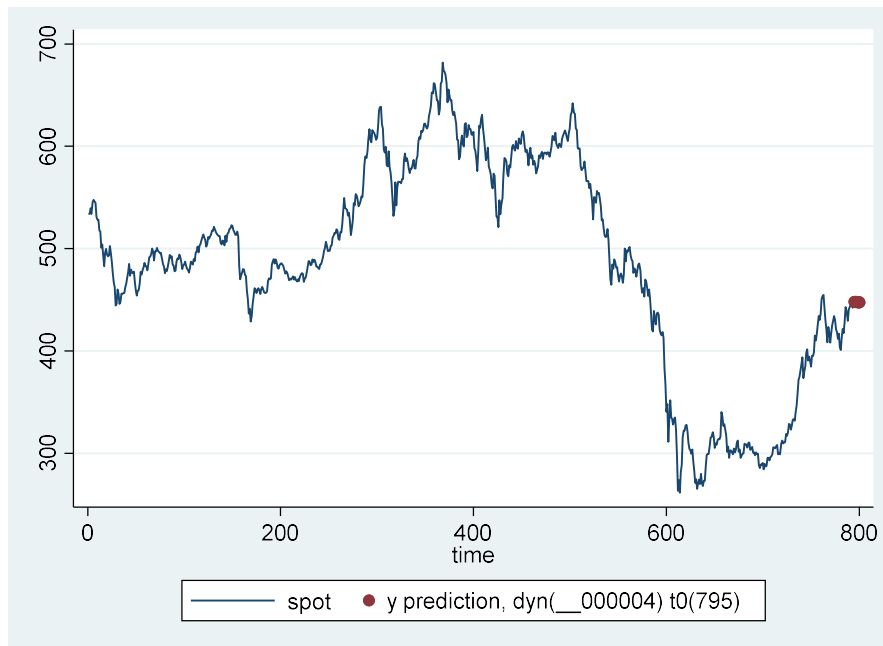
```

. replace time = _n
(5 real changes made)

. predict yhat, y dynamic(.) t0(795)
Note: beginning dynamic predictions in period      4.
(794 missing values generated)

. twoway (line spot time, sort) (scatter yhat time, sort)

```



```

. arima future, arima(1,1,1)

(setting optimization to BHHH)
Iteration 0:  log likelihood = -2918.0578
. . .
Iteration 6:  log likelihood = -2917.724

```

ARIMA regression

Sample: 2 - 795	Number of obs	=	794
	Wald chi2(2)	=	3.33
Log likelihood = -2917.724	Prob > chi2	=	0.1890

```

-----
                |
                |          OPG
D.future |      Coef.  Std. Err.    z    P>|z|    [95% Conf. Interval]
-----+-----
future    |
   _cons |  -.1142458   .3326737   -0.34   0.731   -1.7662743   .5377828
-----+-----
ARMA      |
   ar    |
  L1.    |  -.368505    .5847749   -0.63   0.529   -1.514643    .7776328
        |
   ma    |
  L1.    |   .3311954   .5949511    0.56   0.578   -.8348873    1.497278
-----+-----
   /sigma |   9.54277    .1687756   56.54   0.000    9.211975    9.873564
-----

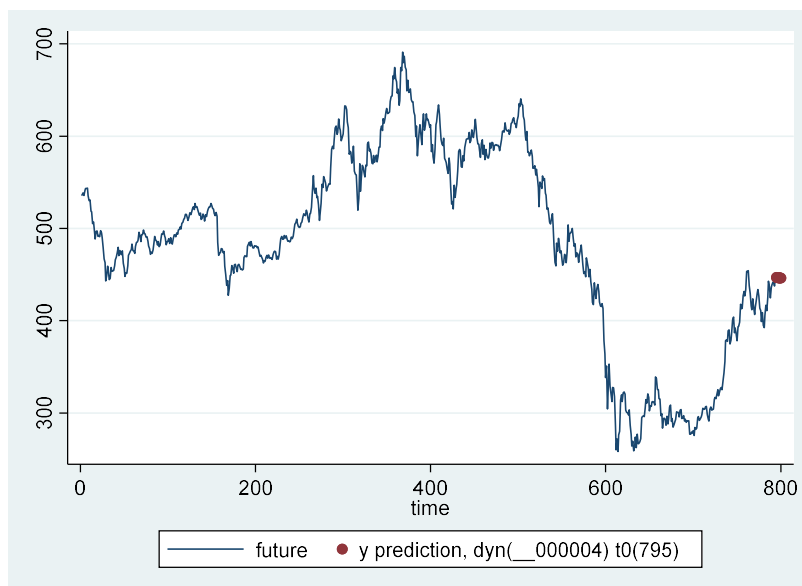
```

Note: The test of the variance against zero is one sided, and the two-sided confidence interval is truncated at zero.

```
. predict yhat2, y dynamic(.) t0(795)
```

Note: beginning dynamic predictions in period  
(794 missing values generated)

```
. twoway (line future time, sort) (scatter yhat2 time, sort)
```



Part II

The following GARCH model:

Mean Equation:  $rfuture_t = \alpha + \beta rspot_t + \varepsilon_t$  (1)

Variance Equation:  $\sigma_t^2 = \alpha_0 + \sum_{j=1}^p \delta_j \sigma_{t-j}^2 + \sum_{i=1}^q \alpha_i \varepsilon_{t-i}^2$  (2)

3. Estimate model (1) using OLS by employing future return (rfuture) as dependent variable and spot return (rspot) as explanatory variable, and determine whether ARCH-effect significantly occurs.

```
. gen rspot = (spot-l.spot)/l.spot
(6 missing values generated)
. gen rfuture = (future-l.future)/l.future
(6 missing values generated)
. reg rfuture rspot
```

Source	SS	df	MS	Number of obs	=	794
-----+-----				F(1, 792)	=	6189.46
Model	.338771077	1	.338771077	Prob > F	=	0.0000
Residual	.043348982	792	.000054734	R-squared	=	0.8866
-----+-----				Adj R-squared	=	0.8864
Total	.382120059	793	.000481866	Root MSE	=	.0074

rfuture	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----						
rspot	1.108778	.0140935	78.67	0.000	1.081113	1.136443
_cons	.0000525	.0002626	0.20	0.841	-.0004629	.0005679

```
. estat archlm
```

LM test for autoregressive conditional heteroskedasticity (ARCH)

lags(p)	chi2	df	Prob > chi2
-----+-----			
1	43.545	1	0.0000

H0: no ARCH effects vs. H1: ARCH(p) disturbance

**From the test, there exists ARCH effect since p-value of the test =0.00 < 0.05**

4. Estimate GARCH(p,q) for future return (rfuture) using spot return (rspot) as explanatory variable for mean equation (model (1) and (2)) – determine the most appropriated order p and q for variance equation using SBIC given the maximum lag equals to 2.

```
. qui arch rfuture rspot, garch(1/1) arch(1/1)
. est store garch11
. qui arch rfuture rspot, garch(1/1) arch(1/2)
. est store garch12
. qui arch rfuture rspot, garch(1/2) arch(1/1)
. est store garch21
. qui arch rfuture rspot, garch(1/2) arch(1/2)
. est store garch22
. estimates table garch11 garch12 garch21 garch22, star(0.1 0.05 0.01) stat(aic bic ll)
```

Variable	garch11	garch12	garch21	garch22
-----				
rfuture				
rspot	1.066779***	1.0686232***	1.0662306***	1.0681952***
_cons	.00003064	.00002254	.00002879	.00002246
-----				
ARCH				
arch				
L1.	.0846231***	.2121659***	.11369069***	.21614027***
L2.		-.14954562***		-.17303686***
garch				
L1.	.89700601***	.92239359***	.54514899	1.1227156***
L2.			.3148726	-.17547447
_cons	9.850e-07***	7.744e-07***	1.380e-06**	5.016e-07**
-----				
Statistics				
aic	-5702.2841	-5708.9307	-5702.54	-5707.938
bic	-5678.8987	-5680.8682	-5674.4775	-5675.1984
ll	2856.1421	2860.4654	2857.27	2860.969
-----				

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

*According to the result, the most appropriated order of GARCH is GARCH(1,2) since it gives the lowest BIC.*

5. From (6), predict the variance of future return (rfuture).

```
. predict sigma_hat, v  
. line sigma_hat time
```

