

## 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.

```
tsset t
      time variable: time, 1 to 795
              delta: 1 unit

. g y = spot

. arima y, arima(1,1,1) nolog

ARIMA regression

Sample: 2 - 795
Log likelihood = -2791.496

Number of obs   =      794
Wald chi2(2)    =      18.31
Prob > chi2     =      0.0001
```

	D.y	Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]	
y	_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.

```
. est store arima111

. qui arima y, arima(1,1,2) nolog

. est store arima112

. qui arima y, arima(1,1,3) nolog

. est store arima113

. qui arima y, arima(1,1,4) nolog

. est store arima114

. qui arima y, arima(2,1,1) nolog

. est store arima211

. qui arima y, arima(2,1,2) nolog
```

```

. est store arima212

. qui arima y, arima(2,1,3) nolog

. est store arima213

. qui arima y, arima(2,1,4) nolog

. est store arima214

. qui arima y, arima(3,1,1) nolog

. est store arima311

. qui arima y, arima(3,1,2) nolog

. est store arima312

. qui arima y, arima(3,1,3) nolog

. est store arima313

. qui arima y, arima(3,1,4) nolog

. est store arima314

. est table arima1*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

```

Variable	arima111	arima112	arima113	arima114
-----				
y				
_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 arima2*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

```

Variable	arima211	arima212	arima213	arima214
-----				
Y				
_cons	-.10453586	-.10767543	-.10609456	-.10747347
-----				
ARMA				
ar				
L1.	.30210361	1.3247307***	.46523894***	1.3085163***
L2.	.02452891	-.92390999***	-.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	3656.4673	241.72156	4791.7037
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 arima3\*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima311	arima312	arima313	arima314
-----				
Y				
_cons	-.10486338	-.10606429	-.10806841	-.10765207
-----				
ARMA				
ar				
L1.	-.56058829	.56814277***	.66749387**	.58444954
L2.	.08569307*	-.89762851***	-.06995188	.02671133
L3.	.06374492**	.09752318***	-.59846871*	-.67269654
ma				
L1.	.63856195	-.49161072***	-.61040004*	-.51894361
L2.		.88783712***	.0556601	-.035013
L3.			.65798111**	.71899187
L4.				.02307553
-----				
sigma				
_cons	8.1302906***	8.1136284***	8.0655263***	8.0640794***
-----				
Statistics				
N	794	794	794	794
ll	-2790.5308	-2788.9397	-2784.4054	-2784.2793
chi2	21.334447	289.47777	1722.9017	1770.3818
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

```
. qui arima y, arima(4,1,1) nolog
. est store arima411
. qui arima y, arima(4,1,2) nolog
. est store arima412
. qui arima y, arima(4,1,3) nolog
. est store arima413
. qui arima y, arima(4,1,4) nolog
. est store arima414
. est table arima4*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable	arima411	arima412	arima413	arima414
-----				
Y				
_cons	-.10781888	-.10736741	-.10798992	-.10783286
-----				
ARMA				
ar				
L1.	.6019371*	1.3677375***	.61532081	.8379873
L2.	-.00196743	-1.0093777***	.00952804	-.12990838
L3.	.02357655	.05752757	-.6765738	-.67732522
L4.	-.074501**	-.00157964	.02303935	.1696285
ma				
L1.	-.52565375	-1.3026568***	-.54893042	-.77060667
L2.		.96445075***	-.01956286	.10759475
L3.			.72228244	.72452578
L4.				-.15673595
-----				
sigma				
_cons	8.1178063***	8.0663819***	8.0637584***	8.0632846***
-----				
Statistics				
N	794	794	794	794
ll	-2789.3297	-2784.5185	-2784.2686	-2784.2263
chi2	30.742143	5309.2623	1879.1703	2712.7298
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

```
. drop y
. g y = future
. arima y, arima(1,1,1) nolog
```

ARIMA regression

Sample: 2 - 795

Number of obs = 794

Wald chi2(2) = 3.33

Log likelihood = -2917.724

Prob > chi2 = 0.1890

	D.y	Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]	
Y							
	_cons	-.1142458	.3326737	-0.34	0.731	-.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.

```

.
. est store arima111

.
. qui arima y, arima(1,1,2) nolog

.
. est store arima112

.
. qui arima y, arima(1,1,3) nolog

.
. est store arima113

.
. qui arima y, arima(1,1,4) nolog

.
. est store arima114

.
. qui arima y, arima(2,1,1) nolog

.
. est store arima211

.
. qui arima y, arima(2,1,2) nolog

.
. est store arima212

```

```

. qui arima y, arima(2,1,3) nolog

.
. est store arima213

.
. qui arima y, arima(2,1,4) nolog

.
. est store arima214

.
. qui arima y, arima(3,1,1) nolog

.
. est store arima311

.
. qui arima y, arima(3,1,2) nolog

.
. est store arima312

.
. qui arima y, arima(3,1,3) nolog

.
. est store arima313

.
. qui arima y, arima(3,1,4) nolog

.
. est store arima314

.
. est table arima1*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

```

Variable	arima111	arima112	arima113	arima114
Y				
_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 arima2\*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima211	arima212	arima213	arima214
y				
_cons	-.11399264	-.11488781	-.11518755	-.114745
ARMA				
ar				
L1.	.26006773	1.2835738***	.39680958***	.29779715***
L2.	.03927244	-.90739116***	-.83492246***	-.89274399***
ma				
L1.	-.29537225	-1.2846762***	-.43709178***	-.33588403***
L2.		.95025686***	.86844397***	.9442071***
L3.			.02125732	.00060118
L4.				.05136959*
sigma				
_cons	9.5398841***	9.4620541***	9.4808276***	9.473342***
Statistics				
N	794	794	794	794
ll	-2917.4896	-2911.1516	-2912.6217	-2912.0146
chi2	2.8388146	3447.4967	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 arima3\*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima311	arima312	arima313	arima314
y				
_cons	-.11395314	-.11434855	-.11487997	-.11489771
ARMA				
ar				
L1.	.06416111	1.3204613***	-.56194538***	.01972589
L2.	.02920804	-.90163025***	-.42027467***	-.77499053***
L3.	.02294871	-.05791248**	-.83389917***	-.25249026
ma				
L1.	-.10003354	-1.3735497	.51794954	-.05826621

L2.			1.0000008	.40778767	.81652449***
L3.				.88983994	.26256083
L4.					.05469308*
-----					
sigma					
_cons		9.5380895***	9.4185571	9.4578166	9.4722994***
-----					
Statistics					
N		794	794	794	794
ll		-2917.3418	-2909.7635	-2911.5297	-2911.9403
chi2		3.8262796	32923.58	7410.3936	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

```

.
. qui arima y, arima(4,1,1) nolog
.
. est store arima411
.
. qui arima y, arima(4,1,2) nolog
.
. est store arima412
.
. qui arima y, arima(4,1,3) nolog
.
. est store arima413
.
. qui arima y, arima(4,1,4) nolog
.
. est store arima414
.
. est table arima4*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

```

Variable		arima411	arima412	arima413	arima414
-----					
y					
_cons		-.11336618	-.11461152	-.11590868	-.11376705
-----					
ARMA					
ar					
L1.		-.26142231	.3181852***	1.7676398***	-.41944632***
L2.		.01694481	-.85050807***	-1.50548*	.36718541***
L3.		.03124469	.00130972	.41843717	-.35522863***
L4.		.0307881	.04773204	.00058669	-.86090936***
ma					
L1.		.22521179	-.35750674***	-1.8167691***	.38743681***

L2.			.90090477***	1.6207448*	-.30501342***
L3.				-.49952541	.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	4207.3323	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 estimated result, ARIMA(1,1,1) is the most appropriated for both cases.

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

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

ARIMA regression

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

		OPG				
D.spot		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) (scatter yhat time, sort)
```



```
arima future, arima(1,1,1) nolog
```

```
ARIMA regression
```

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

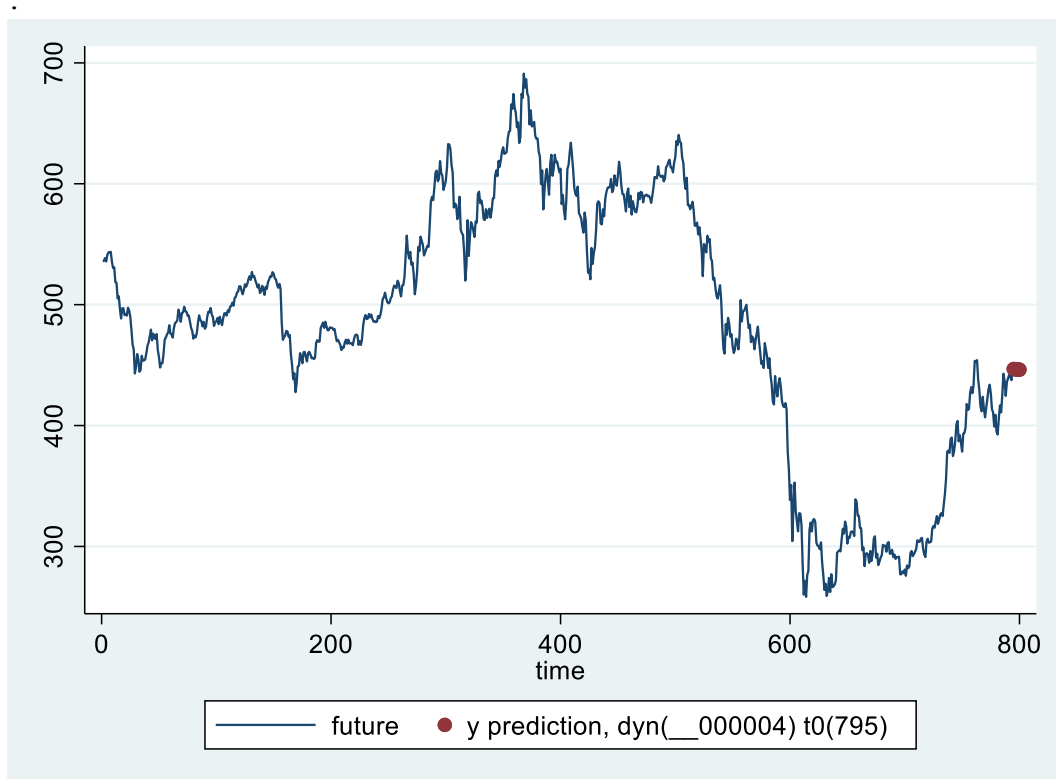
-----						
D.future	Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]	
-----						
future						
_cons	-.1142458	.3326737	-0.34	0.731	-.7662743	.5377828
-----						
ARMA						
ar						
L1.	-.368505	.5847749	-0.63	0.529	-1.514643	.7776328
ma						
L1.	.3311954	.5949511	0.56	0.578	-.8348873	1.497278
-----						
/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 4.  
(794 missing values generated)

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



## Part II

The following GARCH model:

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

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

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.

```
g rspot = (spot/l.spot)-1
(6 missing values generated)
```

```
. g rfuture = (future/l.future)-1
(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 tested result, there exists ARCH-effect.

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