

Quantitative Methods in Finance

Tutorial, Part 11:

Example 1: Generate two random walk series (e.g. with integer steps). See the code in the Stata do-file `randomwalk-commands.do`. Are such series always uncorrelated?

Computer printout of the results in Stata:

```
. set obs 1000
. set seed 100

. gen t=_n
. tsset t
      time variable:  t, 1 to 1000
                delta:  1 unit

. gen drift1=-1
. replace drift1=1 if runiform().>0.5
(519 real changes made)

. gen drift2=-1
. replace drift2=1 if runiform().>0.5
(482 real changes made)

. gen rw1=0
. gen rw2=0

. replace rw1=1.rw1+1.drift1 if t>1
(989 real changes made)

. replace rw2=1.rw2+1.drift2 if t>1
(968 real changes made)

. dfuller rw1
```

Dickey-Fuller test for unit root Number of obs = 999

		----- Interpolated Dickey-Fuller -----		
	Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value

Z (t)	-0.697	-3.430	-2.860	-2.570

MacKinnon approximate p-value for $Z(t) = 0.8476$

```
. dfuller rw2
```

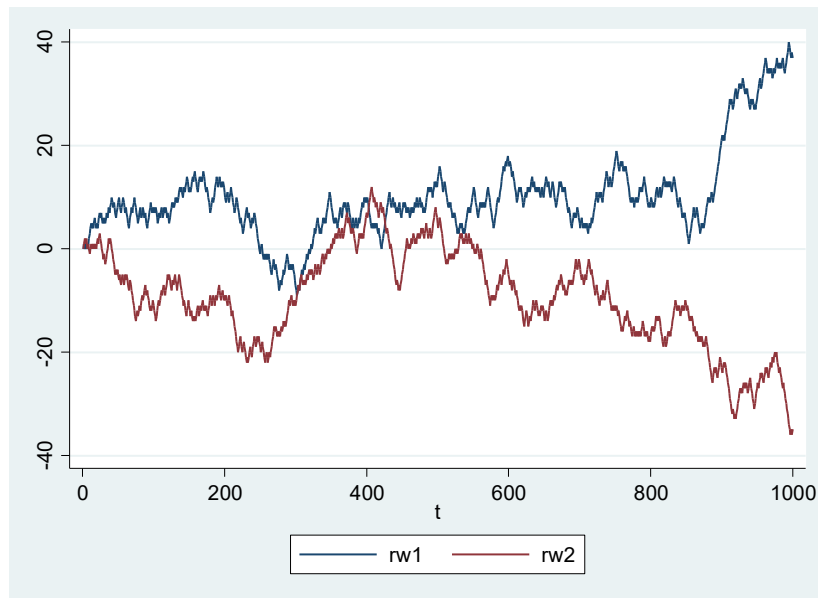
Dickey-Fuller test for unit root Number of obs = 999

		----- Interpolated Dickey-Fuller -----		
	Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value

Z (t)	-0.742	-3.430	-2.860	-2.570

MacKinnon approximate p-value for Z(t) = 0.8356

```
. tsline rw1 rw2
```



```
. regress rw1 rw2
```

Source	SS	df	MS	Number of obs	=	1,000
Model	22149.705	1	22149.705	F(1, 998)	=	422.52
Residual	52318.131	998	52.422977	Prob > F	=	0.0000
Total	74467.836	999	74.5423784	R-squared	=	0.2974
				Adj R-squared	=	0.2967
				Root MSE	=	7.2404

rw1	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
rw2	-.5043769	.0245376	-20.56	0.000	-.5525281 -.4562257
_cons	5.357481	.3264995	16.41	0.000	4.716777 5.998185

■

Example 2: Let us define the difference operator $\Delta = (1 - L)$, where L is the lag operator, such that $L^j Y_t = y_{t-j}$. In general, $\Delta_j^i = (1 - L^j)^i$, where i and j are typically omitted when they take the value of 1. Show the expressions in Y only when applying the difference operator to the following expressions, and provide an interpretation for the resulting expression, assuming that you are working with quarterly data:

- $\Delta_4 Y_t$;
- $\Delta^2 Y_t$;
- $\Delta_1 \Delta_4 Y_t$;
- $\Delta_4^2 Y_t$.

■

- Inspect the time series visually. Does it look stationary? Support your decision by using a graphical representation and an analytical test.
- How could the SBI Top time series be transformed to a stationary one? Apply the procedure(s) and check the stationarity of the transformed time series.
- Identify the SBI Top time series with an appropriate ARMA model.
- By using the identified ARMA model, provide an “in-sample” forecast for the last half a year (trading days, i.e. for the last 123 observations).

Computer printout of the results in Stata:

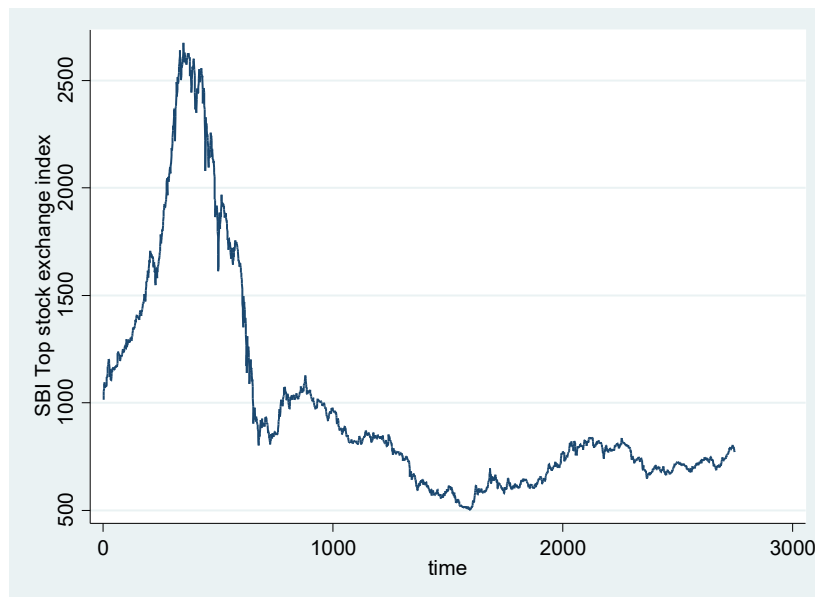
a) Checking for stationarity

```
. generate time=_n

. tsset time
      time variable:  time, 1 to 2749
             delta:  1 unit

. order time sbitop

. twoway line sbitop time
```



```
. dfuller sbitor
```

Dickey-Fuller test for unit root Number of obs = 2748

	Test Statistic	----- 1% Critical Value	Interpolated Dickey-Fuller 5% Critical Value	----- 10% Critical Value
Z(t)	-0.681	-3.430	-2.860	-2.570

MacKinnon approximate p-value for $Z(t) = 0.8515$

Dickey-Fuller test for unit root Number of obs = 2748

D.sbitop	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
sbitop						
l1	-.0003697	.0005425	-0.68	0.496	-.0014334	.0006941
_cons	.2813994	.6075337	0.46	0.643	-.9098699	1.472669

```
DF-GLS for sbitop                                Number of obs = 2721
Maxlag = 27 chosen by Schwert criterion
```

Opt Lag (Ng-Perron seq t) = 27 with RMSE 13.62633
 Min SC = 5.276277 at lag 2 with RMSE 13.92629
 Min MAIC = 5.245152 at lag 27 with RMSE 13.62633

. dfuller sbitop, lags(2)

Augmented Dickey-Fuller test for unit root Number of obs = 2746

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-0.770	-3.430	-2.860

MacKinnon approximate p-value for Z(t) = 0.8277

. dfuller sbitop, regress lags(2)

Augmented Dickey-Fuller test for unit root Number of obs = 2746

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-0.770	-3.430	-2.860

MacKinnon approximate p-value for Z(t) = 0.8277

D.sbitop	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
sbitop						
L1.	-.0004024	.0005223	-0.77	0.441	-.0014266	.0006218
LD.	.268612	.0189101	14.20	0.000	.2315326	.3056914
L2D.	-.1375565	.0189065	-7.28	0.000	-.174629	-.100484
_cons	.3135708	.5849874	0.54	0.592	-.8334898	1.460631

. dfuller sbitop, trend lags(2)

Augmented Dickey-Fuller test for unit root Number of obs = 2746

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.372	-3.960	-3.410

MacKinnon approximate p-value for Z(t) = 0.8689

. dfuller sbitop, trend regress lags(2)

Augmented Dickey-Fuller test for unit root Number of obs = 2746

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.372	-3.960	-3.410

MacKinnon approximate p-value for Z(t) = 0.8689

D.sbitop	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
sbitop						
L1.	-.0009808	.0007147	-1.37	0.170	-.0023823	.0004207
LD.	.2686385	.0189087	14.21	0.000	.2315618	.3057152

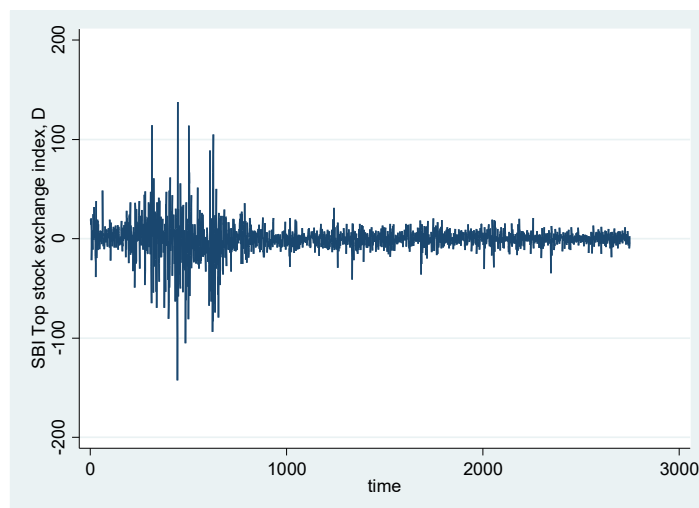
```

      L2D. | -.1373757   .0189058   -7.27   0.000   -.1744467   -.1003048
    _trend | -.0005445   .0004594   -1.19   0.236   -.0014453   .0003562
    _cons  |  1.639399   1.262146    1.30   0.194   -.8354547   4.114254
-----

```

b) Ensuring stationarity

```
. twoway line d.sbitop time // First differences //
```



```
. dfuller d.sbitop, lags(2)
```

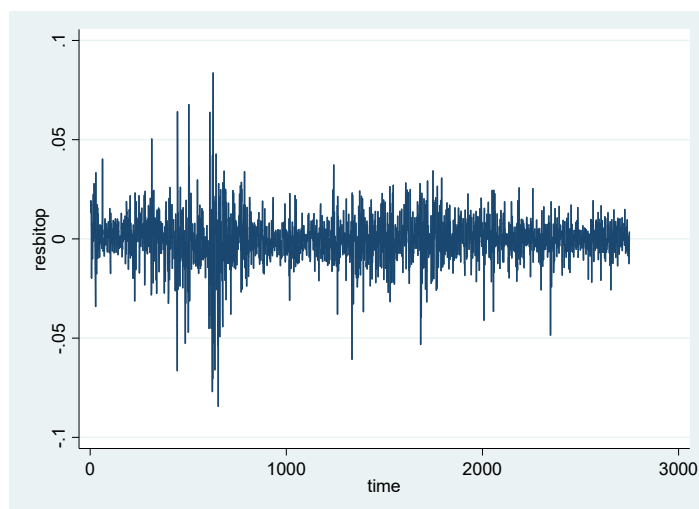
Augmented Dickey-Fuller test for unit root Number of obs = 2745

	Test	----- Interpolated Dickey-Fuller -----		
	Statistic	1% Critical	5% Critical	10% Critical
		Value	Value	Value
Z(t)	-31.012	-3.430	-2.860	-2.570

MacKinnon approximate p-value for Z(t) = 0.0000

```
. generate resbitop=ln(sbitop/L.sbitop) // Continuously compounded returns //
(1 missing value generated)
```

```
. twoway line resbitop time
```



```
. dfuller resbitop, lags(2)
```

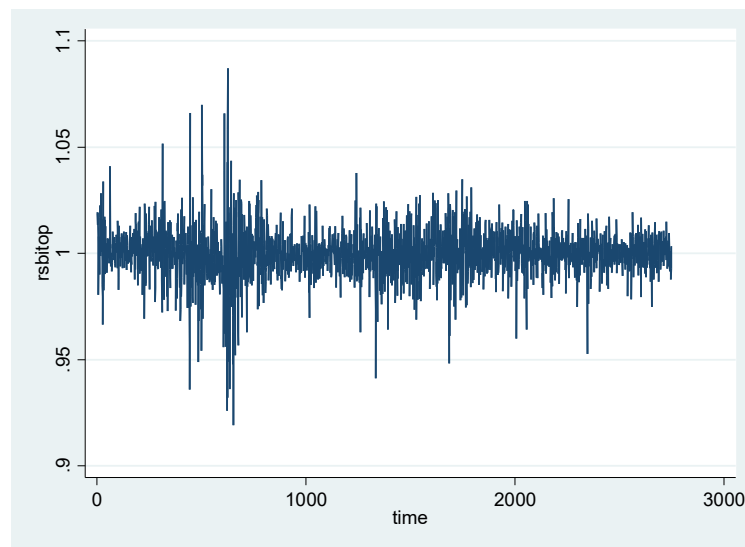
Augmented Dickey-Fuller test for unit root Number of obs = 2745

	Test Statistic	----- Interpolated Dickey-Fuller -----		
		1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-30.284	-3.430	-2.860	-2.570

MacKinnon approximate p-value for Z(t) = 0.0000

```
. generate rsbitop=sbitop/l.sbitop                      // Growth rates //
(1 missing value generated)
```

```
. twoway line rsbitop time
```



```
. dfuller rsbitop, lags(2)
```

Augmented Dickey-Fuller test for unit root Number of obs = 2745

	Test Statistic	----- Interpolated Dickey-Fuller -----		
		1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-30.363	-3.430	-2.860	-2.570

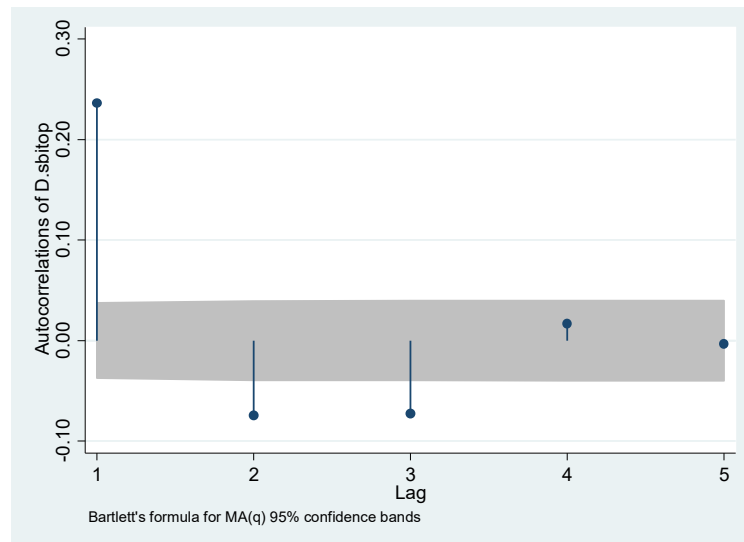
MacKinnon approximate p-value for Z(t) = 0.0000

c) Identification of the ARMA model

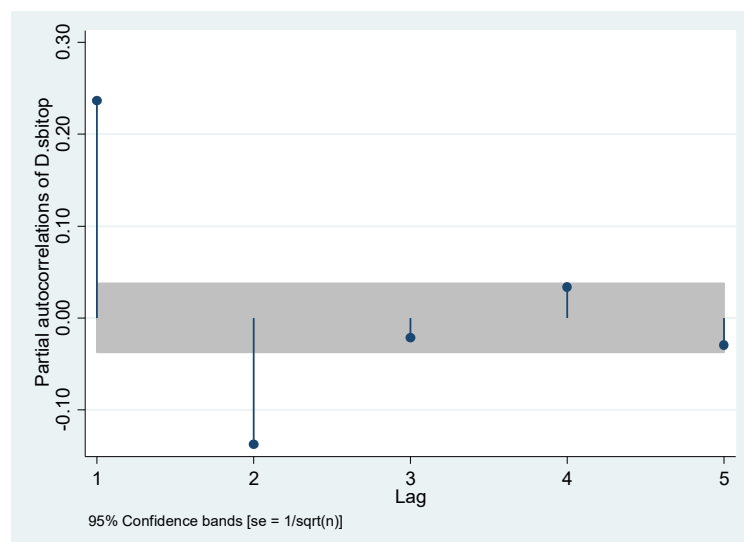
```
. corrgram d.sbitop, lags(5)
```

LAG	AC	PAC	Q	Prob>Q	-1	0	1	-1	0	1
					[Autocorrelation]			[Partial Autocor]		
1	0.2364	0.2364	153.69	0.0000		-			-	
2	-0.0744	-0.1378	168.91	0.0000					-	
3	-0.0726	-0.0216	183.41	0.0000						
4	0.0168	0.0339	184.19	0.0000						
5	-0.0031	-0.0294	184.21	0.0000						

```
. ac d.sbitop, lags(5)
```



```
. pac d.sbitop, lags(5)
```



```
. arima d.sbitop, arima(0,0,0)
```

```
(setting optimization to BHHH)
Iteration 0:  log likelihood = -11244.623
Iteration 1:  log likelihood = -11244.623
```

ARIMA regression

```
Sample: 2 - 2749                                Number of obs      =      2748
                                                Wald chi2(.)          =          .
Log likelihood = -11244.62                      Prob > chi2           =          .
```

		OPG		z	P> z	[95% Conf. Interval]	
D.sbitop	Coef.	Std. Err.					
sbitop							
_cons	-.0872525	.2774404	-0.31	0.753	-.6310257	.4565208	


```

-----
      /sigma |   14.48323   .0623727   232.20   0.000   14.36098   14.60548
-----
Note: The test of the variance against zero is one sided, and the two-sided
      confidence interval is truncated at zero.

```

```
. estat ic
```

Akaike's information criterion and Bayesian information criterion

```

-----
      Model |   Obs   ll(null)   ll(model)   df   AIC   BIC
-----+-----
      . |  2748         .  -11244.62     2   22493.25  22505.08
-----

```

Note: N=Obs used in calculating BIC; see [R] BIC note

```
. display e(ll), e(rank)
```

```
-11244.623 2
```

```

// We continue to estimate all possible ARMA models manually
// Alternatively, we can use the following program:

```

```
. generate variable=d.sbitop
```

```
(1 missing value generated)
```

```
. scalar pset=4
```

```
. scalar qset=4
```

```

. forvalues p=0/\`pset' {
2.     forvalues q=0/\`qset' {
3.         qui arima variable, arima(`p',0,`q')
4.         scalar aic_`p'`q'=-2*e(ll)+2*e(rank)
5.         scalar bic_`p'`q'=-2*e(ll)+e(rank)*ln(e(N))
6.     }
7. }

. forvalues p=0/\`pset' {
2.     forvalues q=0/\`qset' {
3.         display "AIC(`p',`q') = "aic_`p'`q'
4.     }
5. }

```

```

AIC(0,0) = 22493.246
AIC(0,1) = 22303.154
AIC(0,2) = 22302.321
AIC(0,3) = 22283.048
AIC(0,4) = 22284.417
AIC(1,0) = 22337.218
AIC(1,1) = 22303.481
AIC(1,2) = 22296.374
AIC(1,3) = 22280.056
AIC(1,4) = 22254.553
AIC(2,0) = 22286.494
AIC(2,1) = 22287.615
AIC(2,2) = 22285.706
AIC(2,3) = 22279.698
AIC(2,4) = 22254.738
AIC(3,0) = 22287.211
AIC(3,1) = 22279.644
AIC(3,2) = 22259.567
AIC(3,3) = 22281.604
AIC(3,4) = 22244.954
AIC(4,0) = 22286.043
AIC(4,1) = 22280.393
AIC(4,2) = 22253.788

```

```
AIC(4,3) = 22255.754
AIC(4,4) = 22246.645
```

```
. forvalues p=0(1) `=pset' {
2.     forvalues q=0(1) `=qset' {
3.         display "SBIC(`p',`q') = "bic_`p'`q'
4.     }
5. }
```

```
SBIC(0,0) = 22505.083
SBIC(0,1) = 22320.91
SBIC(0,2) = 22325.995
SBIC(0,3) = 22312.641
SBIC(0,4) = 22319.929
SBIC(1,0) = 22354.973
SBIC(1,1) = 22327.155
SBIC(1,2) = 22325.967
SBIC(1,3) = 22315.568
SBIC(1,4) = 22295.984
SBIC(2,0) = 22310.168
SBIC(2,1) = 22317.208
SBIC(2,2) = 22321.217
SBIC(2,3) = 22321.128
SBIC(2,4) = 22302.087
SBIC(3,0) = 22316.805
SBIC(3,1) = 22315.156
SBIC(3,2) = 22300.998
SBIC(3,3) = 22328.953
SBIC(3,4) = 22298.221
SBIC(4,0) = 22321.555
SBIC(4,1) = 22321.823
SBIC(4,2) = 22301.137
SBIC(4,3) = 22309.022
SBIC(4,4) = 22305.832
```

```
. arima d.sbitop, ar(1) ma(1/4)
```

```
(setting optimization to BHHH)
Iteration 0:   log likelihood = -11136.971
Iteration 1:   log likelihood = -11136.345
Iteration 2:   log likelihood = -11130.593
Iteration 3:   log likelihood = -11127.363
Iteration 4:   log likelihood = -11124.873
(switching optimization to BFGS)
Iteration 5:   log likelihood = -11123.942
Iteration 6:   log likelihood = -11121.744
Iteration 7:   log likelihood = -11120.387
Iteration 8:   log likelihood = -11120.335
Iteration 9:   log likelihood = -11120.305
Iteration 10:  log likelihood = -11120.288
Iteration 11:  log likelihood = -11120.284
Iteration 12:  log likelihood = -11120.278
Iteration 13:  log likelihood = -11120.277
Iteration 14:  log likelihood = -11120.277
(switching optimization to BHHH)
Iteration 15:  log likelihood = -11120.277
Iteration 16:  log likelihood = -11120.277
Iteration 17:  log likelihood = -11120.277
Iteration 18:  log likelihood = -11120.277
Iteration 19:  log likelihood = -11120.277
(switching optimization to BFGS)
Iteration 20:  log likelihood = -11120.277
Iteration 21:  log likelihood = -11120.277
Iteration 22:  log likelihood = -11120.277
Iteration 23:  log likelihood = -11120.277
```

ARIMA regression

Sample: 2 - 2749

Log likelihood = -11120.28

Number of obs = 2748

Wald chi2(5) = 173101.70

Prob > chi2 = 0.0000

		OPG					
D.sbitop		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
sbitop							
	_cons	.0239208	.9110657	0.03	0.979	-1.761735	1.809577
ARMA							
	ar						
	L1.	.9919309	.0034908	284.15	0.000	.985089	.9987728
	ma						
	L1.	-.7376104	.0068678	-107.40	0.000	-.751071	-.7241497
	L2.	-.3306478	.0088123	-37.52	0.000	-.3479196	-.3133759
	L3.	-.0198507	.0097767	-2.03	0.042	-.0390127	-.0006887
	L4.	.1121987	.0086706	12.94	0.000	.0952047	.1291927
	/sigma	13.8415	.0581916	237.86	0.000	13.72744	13.95555

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

. arima d.sbitop, arima(1,0,4)

(setting optimization to BHHH)

Iteration 0: log likelihood = -11136.971

Iteration 1: log likelihood = -11136.345

Iteration 2: log likelihood = -11130.593

Iteration 3: log likelihood = -11127.363

Iteration 4: log likelihood = -11124.873

(switching optimization to BFGS)

Iteration 5: log likelihood = -11123.942

Iteration 6: log likelihood = -11121.744

Iteration 7: log likelihood = -11120.387

Iteration 8: log likelihood = -11120.335

Iteration 9: log likelihood = -11120.305

Iteration 10: log likelihood = -11120.288

Iteration 11: log likelihood = -11120.284

Iteration 12: log likelihood = -11120.278

Iteration 13: log likelihood = -11120.277

Iteration 14: log likelihood = -11120.277

(switching optimization to BHHH)

Iteration 15: log likelihood = -11120.277

Iteration 16: log likelihood = -11120.277

Iteration 17: log likelihood = -11120.277

Iteration 18: log likelihood = -11120.277

Iteration 19: log likelihood = -11120.277

(switching optimization to BFGS)

Iteration 20: log likelihood = -11120.277

Iteration 21: log likelihood = -11120.277

Iteration 22: log likelihood = -11120.277

Iteration 23: log likelihood = -11120.277

ARIMA regression

Sample: 2 - 2749

Log likelihood = -11120.28

Number of obs = 2748

Wald chi2(5) = 173101.70

Prob > chi2 = 0.0000

		OPG					
D.sbitop		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
sbitop							
	_cons	.0239208	.9110657	0.03	0.979	-1.761735	1.809577
ARMA							
	ar						
	L1.	.9919309	.0034908	284.15	0.000	.985089	.9987728
	ma						
	L1.	-.7376104	.0068678	-107.40	0.000	-.751071	-.7241497
	L2.	-.3306478	.0088123	-37.52	0.000	-.3479196	-.3133759
	L3.	-.0198507	.0097767	-2.03	0.042	-.0390127	-.0006887
	L4.	.1121987	.0086706	12.94	0.000	.0952047	.1291927
	/sigma	13.8415	.0581916	237.86	0.000	13.72744	13.95555

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

. arima sbitop, arima(1,1,4)

```
(setting optimization to BHHH)
Iteration 0:  log likelihood = -11136.971
Iteration 1:  log likelihood = -11136.345
Iteration 2:  log likelihood = -11130.593
Iteration 3:  log likelihood = -11127.363
Iteration 4:  log likelihood = -11124.873
(switching optimization to BFGS)
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Iteration 11: log likelihood = -11120.284
Iteration 12: log likelihood = -11120.278
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(switching optimization to BHHH)
Iteration 15: log likelihood = -11120.277
Iteration 16: log likelihood = -11120.277
Iteration 17: log likelihood = -11120.277
Iteration 18: log likelihood = -11120.277
Iteration 19: log likelihood = -11120.277
(switching optimization to BFGS)
Iteration 20: log likelihood = -11120.277
Iteration 21: log likelihood = -11120.277
Iteration 22: log likelihood = -11120.277
Iteration 23: log likelihood = -11120.277
```

ARIMA regression

```
Sample: 2 - 2749
Log likelihood = -11120.28

Number of obs      =      2748
Wald chi2(5)       =   173101.70
Prob > chi2        =      0.0000
```

		OPG					
D.sbitop		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
sbitop							
	_cons	.0239208	.9110657	0.03	0.979	-1.761735	1.809577

ARMA							
+-----							
ar							
L1.		.9919309	.0034908	284.15	0.000	.985089	.9987728
+-----							
ma							
L1.		-.7376104	.0068678	-107.40	0.000	-.751071	-.7241497
L2.		-.3306478	.0088123	-37.52	0.000	-.3479196	-.3133759
L3.		-.0198507	.0097767	-2.03	0.042	-.0390127	-.0006887
L4.		.1121987	.0086706	12.94	0.000	.0952047	.1291927
+-----							
/sigma		13.8415	.0581916	237.86	0.000	13.72744	13.95555

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

. estat aroots, dlabel

Eigenvalue stability condition

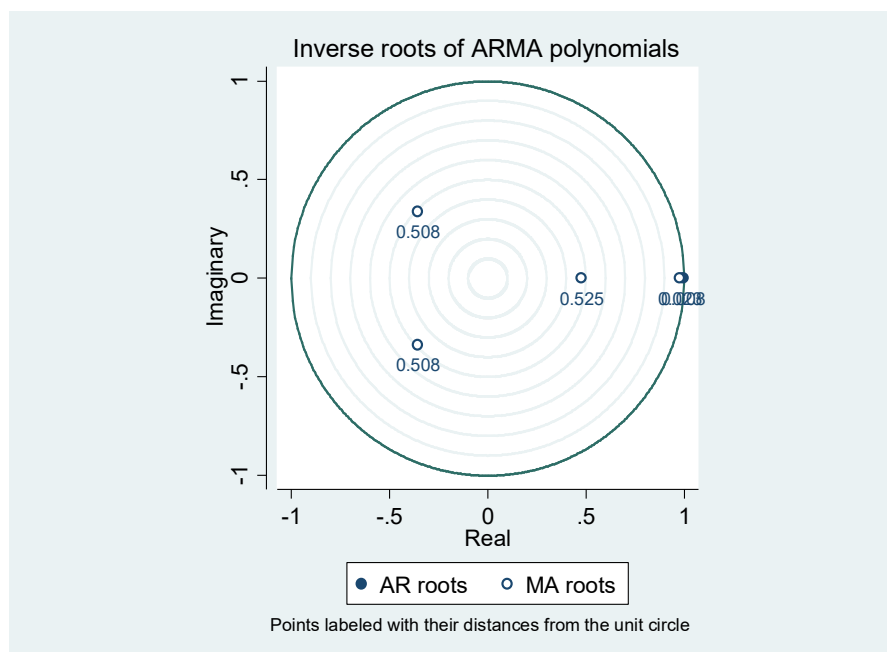
+-----	
Eigenvalue	Modulus
+-----	
.9919309	.991931
+-----	

All the eigenvalues lie inside the unit circle.
AR parameters satisfy stability condition.

Eigenvalue stability condition

+-----	
Eigenvalue	Modulus
+-----	
.9765367	.976537
-.3569913 + .338248i	.491787
-.3569913 - .338248i	.491787
.4750564	.475056
+-----	

All the eigenvalues lie inside the unit circle.
MA parameters satisfy invertibility condition.



```
. irf create ARMA, set(myirf, replace)
(file myirf.irf created)
(file myirf.irf now active)
(file myirf.irf updated)
```

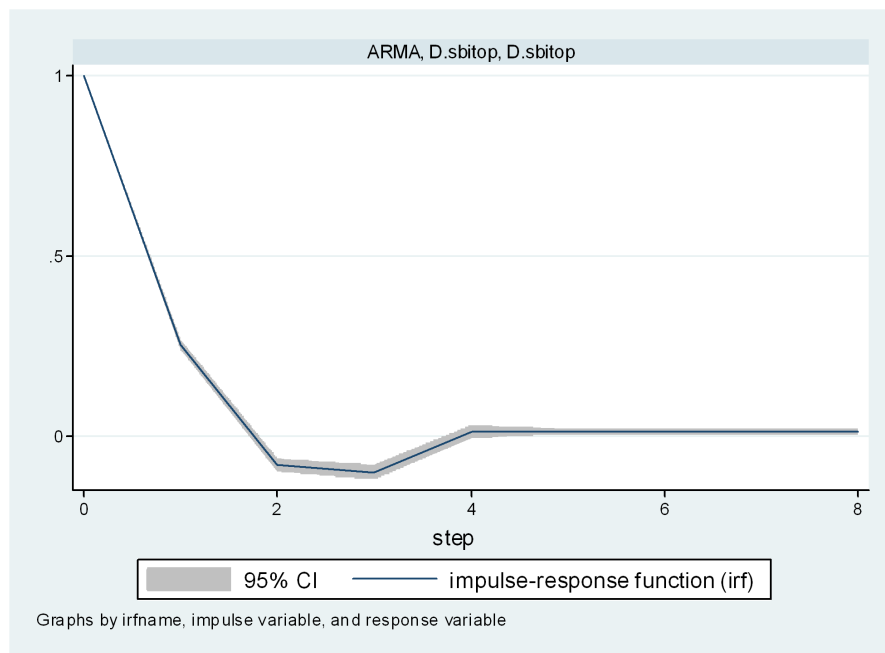
``` . irf describe ```

Contains irf results from myirf.irf

irfname	model	endogenous variables and order (*)
ARMA	arma	D.sbitop

(*) order is relevant only when model is var

``` . irf graph irf ```



``` . irf table irf ```

Results from ARMA

step	(1) irf	(1) Lower	(1) Upper
0	1	1	1
1	.254321	.241383	.267258
2	-.078379	-.095964	-.060794
3	-.097598	-.118049	-.077146
4	.015389	-.002311	.033088
5	.015264	.008027	.022502
6	.015141	.007858	.022425
7	.015019	.007691	.022347
8	.014898	.007526	.02227

95% lower and upper bounds reported

(1) irfname = ARMA, impulse = D.sbitop, and response = D.sbitop

d) Forecasting using the ARMA model

```
. arima d.sbitop if time <= 2626, arima(1,0,4)
```

```
(setting optimization to BHHH)
Iteration 0: log likelihood = -10690.67
Iteration 1: log likelihood = -10690.058
Iteration 2: log likelihood = -10684.201
Iteration 3: log likelihood = -10681.319
Iteration 4: log likelihood = -10678.887
(switching optimization to BFGS)
Iteration 5: log likelihood = -10678.421
Iteration 6: log likelihood = -10676.075
Iteration 7: log likelihood = -10674.765
Iteration 8: log likelihood = -10674.635
Iteration 9: log likelihood = -10674.593
Iteration 10: log likelihood = -10674.582
Iteration 11: log likelihood = -10674.572
Iteration 12: log likelihood = -10674.569
Iteration 13: log likelihood = -10674.565
Iteration 14: log likelihood = -10674.564
(switching optimization to BHHH)
Iteration 15: log likelihood = -10674.564
Iteration 16: log likelihood = -10674.563
Iteration 17: log likelihood = -10674.563
Iteration 18: log likelihood = -10674.563
Iteration 19: log likelihood = -10674.563
(switching optimization to BFGS)
Iteration 20: log likelihood = -10674.563
Iteration 21: log likelihood = -10674.563
Iteration 22: log likelihood = -10674.563
Iteration 23: log likelihood = -10674.563
Iteration 24: log likelihood = -10674.563
Iteration 25: log likelihood = -10674.563
Iteration 26: log likelihood = -10674.563
Iteration 27: log likelihood = -10674.563
```

ARIMA regression

```
Sample: 2 - 2626
Log likelihood = -10674.56

Number of obs      =      2625
Wald chi2(5)       =    160118.49
Prob > chi2        =      0.0000
```

		OPG				
D.sbitop		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
sbitop						
_cons		.015196	.9543819	0.02	0.987	-1.855358 1.88575
ARMA						
ar						
L1.		.9919454	.0036254	273.61	0.000	.9848398 .999051
ma						
L1.		-.7364219	.0071441	-103.08	0.000	-.7504241 -.7224196
L2.		-.3325661	.0091552	-36.33	0.000	-.35051 -.3146222
L3.		-.020701	.0101662	-2.04	0.042	-.0406264 -.0007755
L4.		.1138036	.0090247	12.61	0.000	.0961156 .1314916
/sigma		14.11854	.0618202	228.38	0.000	13.99737 14.2397

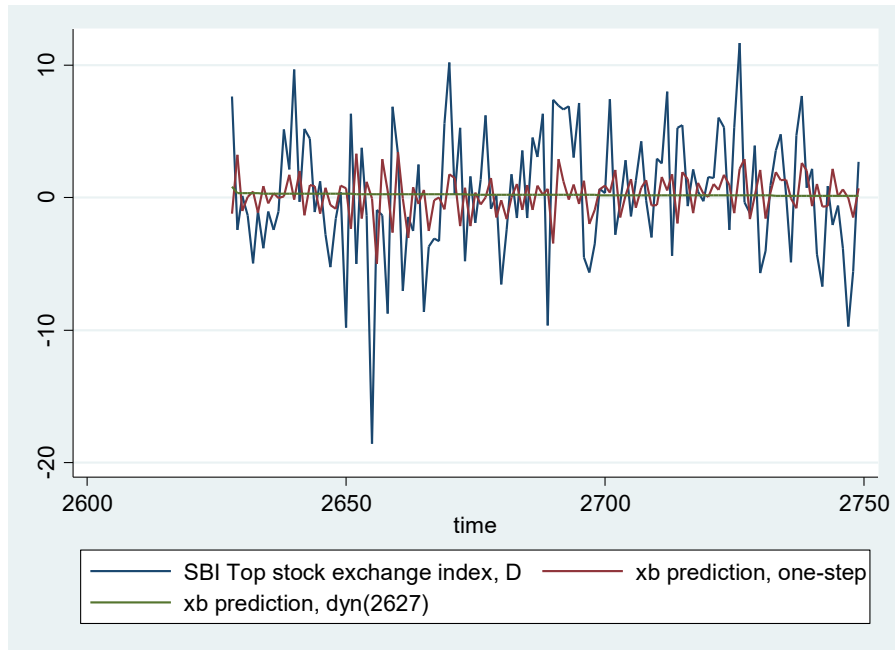
Note: The test of the variance against zero is one sided, and the two-sided confidence interval is truncated at zero.

```

. predict sbitop_fstatic, xb
. predict sbitop_fdynamic, xb dynamic(2627)

. twoway (tsline d.sbitop) (tsline sbitop_fstatic) (tsline sbitop_fdynamic) if time>2627

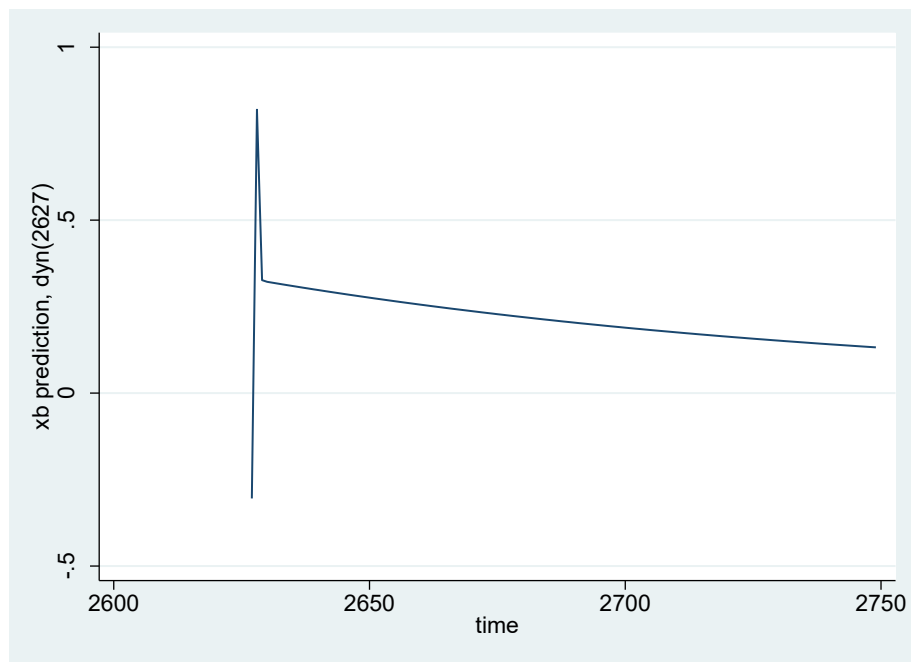
```



```

. twoway (tsline sbitop_fdynamic) if time>2626

```



■