

Generalized Autoregressive Conditional Heteroscedastic (GARCH) Models

Example Simulated Data

```

set obs 501
g t=_n
tsset t
g s2=1 in 1
g u=rnormal(0,s2) in 1
g u2=u^2 in 1
forvalue i=2(1)501 {
  replace s2=0.3+0.3*1.u2+0.3*1.s2 in `i'
  replace u=rnormal(0,s2) in `i'
  replace u2=u^2 in `i'
}
g x=rnormal(1,10)
g y=0.5+0.7*x+u
reg y x if t>1
estat archlm
arch y x if t>1, arch(1) garch(1) nolog
predict s2hat, v
twoway (line s2hat t) (scatter s2 t)

```

```

. set obs 501
number of observations (_N) was 0, now 501

. g t=_n

. tsset t
      time variable: t, 1 to 501
              delta: 1 unit

. g s2=1 in 1
(500 missing values generated)

. g u=rnormal(0,s2) in 1
(500 missing values generated)

. g u2=u^2 in 1
(500 missing values generated)

. forvalue i=2(1)501 {
2.  replace s2=0.3+0.3*1.u2+0.3*1.s2 in `i'
3.  replace u=rnormal(0,s2) in `i'
4.  replace u2=u^2 in `i'
5.  }

. g x=rnormal(1,10)

. g y=0.5+0.7*x+u

. reg y x if t>1

```

Source	SS	df	MS	Number of obs	=	500
-----+-----						
Model	23398.9523	1	23398.9523	F(1, 498)	=	63539.73
Residual	183.391992	498	.368257012	Prob > F	=	0.0000
-----+-----						
Total	23582.3443	499	47.259207	R-squared	=	0.9922
				Adj R-squared	=	0.9922
				Root MSE	=	.60684

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----						
x	.7013338	.0027823	252.07	0.000	.6958673	.7068002
_cons	.4743123	.0273053	17.37	0.000	.4206645	.5279601
-----+-----						

```

. estat archlm
LM test for autoregressive conditional heteroskedasticity (ARCH)
-----+-----
      lags(p) |             chi2             df             Prob > chi2
-----+-----
              1 |             35.821             1             0.0000
-----+-----
              H0: no ARCH effects       vs.   H1: ARCH(p) disturbance

. arch y x if t>1, arch(1) garch(1) nolog

ARCH family regression

Sample: 2 - 501                Number of obs   =           500
Distribution: Gaussian         Wald chi2(1)    =       84227.65
Log likelihood = -426.751      Prob > chi2     =           0.0000

-----+-----
              y |             Coef.             OPG             z             P>|z|             [95% Conf. Interval]
-----+-----
y
  x | .7028459 | .0024218 | 290.22 | 0.000 | .6980993 | .7075925
  _cons | .4758044 | .023265 | 20.45 | 0.000 | .4302058 | .5214029
-----+-----
ARCH
  arch
  L1. | .3970325 | .093721 | 4.24 | 0.000 | .2133428 | .5807223
  |
  garch
  L1. | .3355659 | .112419 | 2.98 | 0.003 | .1152287 | .5559032
  |
  _cons | .105968 | .0289994 | 3.65 | 0.000 | .0491302 | .1628059
-----+-----

. predict s2hat, v

. predict yhat, xb

. mat beta=e(b)

. mat list beta

beta[1,5]
      y:          y:      ARCH:      ARCH:      ARCH:
              x      _cons      L.      L.
y1 .70284588 .47580435 .39703251 .33556595 .10596802

. sca b0=e1(beta,1,2)

. sca b1=e1(beta,1,1)

. sca a0=e1(beta,1,5)

. sca d1=e1(beta,1,4)

. sca a1=e1(beta,1,3)

. g yhat_m=b0+b1*x

. g ehat_m=y-yhat_m

. predict uhat, r

. g uhat2=uhat^2

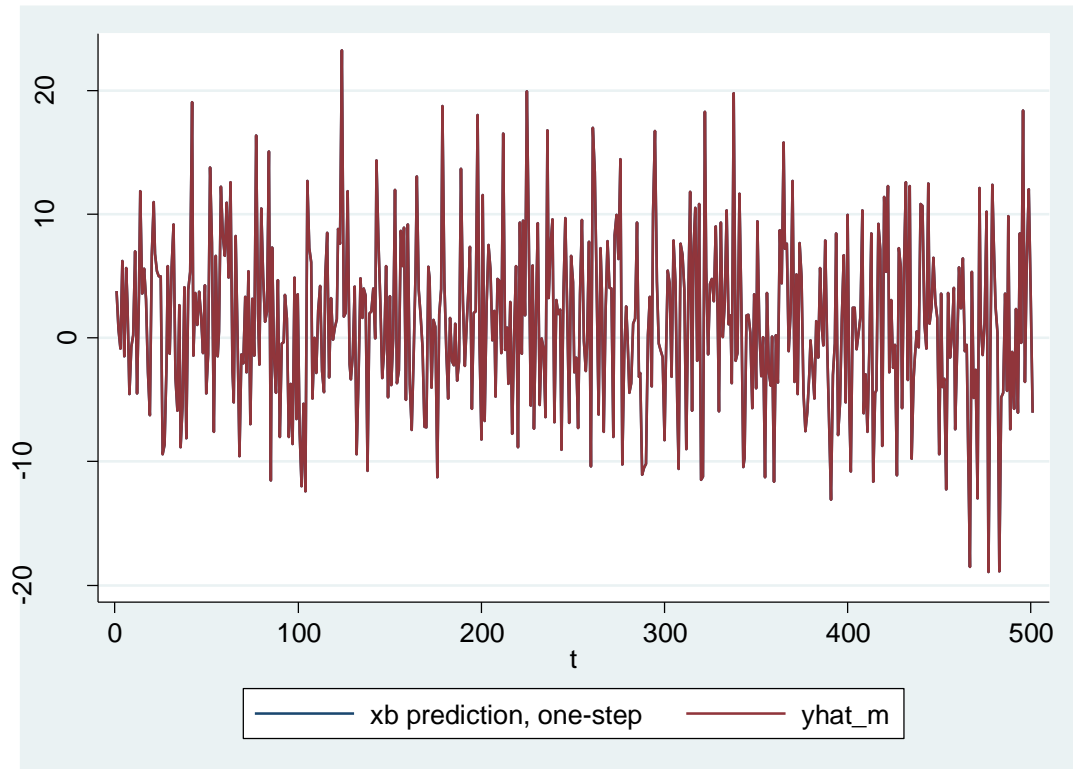
. g s2hat_m=a0+d1*1.uhat2+a1*1.uhat2 in 2

```

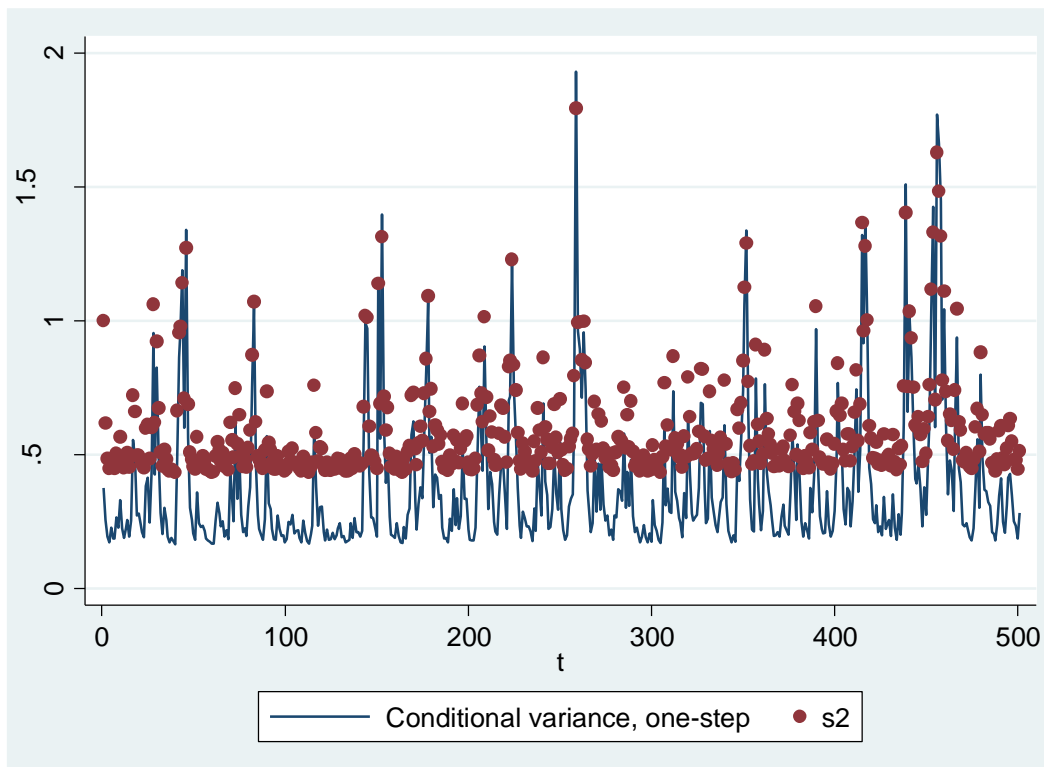
(500 missing values generated)

```
. replace s2hat_m=a0+d1*1.s2hat_m+a1*1.uhat2 if t>2  
(499 real changes made)
```

```
. line yhat yhat_m t
```



```
. twoway (line s2hat t) (scatter s2 t)
```



Example

$$SET_t = \beta_0 + \beta_1 IBR_t + \beta_2 GOLDB_t + \beta_3 USDS_t + \varepsilon_t$$

where: SET_t = Return on Stock Exchange of Thailand (SET).

IBR_t = Inter-bank rate.

$GOLDB_t$ = Gold price.

$USDS_t$ = Exchange rate (\$US/฿Baht).

ε_t = Residual which has GARCH(p,q) process:

$$\sigma_t^2 = \alpha_0 + \delta_1 \sigma_{t-1}^2 + \delta_2 \sigma_{t-2}^2 + \dots + \delta_p \sigma_{t-p}^2 + \alpha_1 \varepsilon_{t-1}^2 + \alpha_2 \varepsilon_{t-2}^2 + \dots + \alpha_q \varepsilon_{t-q}^2$$

Testing GARCH Effect

Firstly, estimate the model using OLS without ARCH(p) process.

```
. regress set ibr usds goldb
```

Source	SS	df	MS	Number of obs =	118
Model	215609.598	3	71869.8659	F(3, 114) =	103.43
Residual	79215.8193	114	694.875608	Prob > F =	0.0000
				R-squared =	0.7313
				Adj R-squared =	0.7242
Total	294825.417	117	2519.87536	Root MSE =	26.36

set	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ibr	-142.0644	68.73617	-2.07	0.041	-278.2302 -5.898551
usds	-78.06484	5.30484	-14.72	0.000	-88.57369 -67.556
goldb	.029042	.0192431	1.51	0.134	-.0090784 .0671624
_cons	3698.659	332.8835	11.11	0.000	3039.219 4358.099

```
. estat archlm
```

LM test for autoregressive conditional heteroskedasticity (ARCH)

lags(p)	chi2	df	Prob > chi2
1	63.052	1	0.0000

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

In this case, since p-value of the ARCH effect test (F-statistic or Chi-Square (Obs*R-squared = (117*0.538907) = 63.05211)) is less than level of significance 0.05, thus, null hypothesis that there is no ARCH effect is rejected, thus, there exists significant ARCH effect in this model with 0.05 significant level.

Identify Order (p,q) and Estimation

The next step is to identify order of GARCH(p,q) by estimating GARCH models in several orders and choose the model with the lowest AIC or SIC.

Estimate GARCH(p,q) using MLE

```
. arch set ibr usds goldb, arch(1) nolog
BFGS stepping has contracted, resetting BFGS Hessian (0)
...
BFGS stepping has contracted, resetting BFGS Hessian (4)
ARCH family regression
Sample: 02dec2003 to 28mar2004                Number of obs   =    118
Log likelihood = -517.3025                    wald chi2(3)    =   1122.15
                                                Prob > chi2     =    0.0000
```

set	Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]
set					
ibr	-64.13751	41.80188	-1.53	0.125	-146.0677 17.79268
usds	-72.48378	2.63951	-27.46	0.000	-77.65712 -67.31043
goldb	.0590191	.0138355	4.27	0.000	.031902 .0861362
_cons	3178.165	197.789	16.07	0.000	2790.506 3565.825
ARCH					
arch					
L1.	.9245129	.3233346	2.86	0.004	.2907888 1.558237
_cons	93.59632	38.20728	2.45	0.014	18.71142 168.4812

```
. estat ic
```

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	118	.	-517.3025	6	1046.605	1063.229

```
. arch set ibr usds goldb, arch(1/2) nolog
BFGS stepping has contracted, resetting BFGS Hessian (17)
ARCH family regression
Sample: 02dec2003 to 28mar2004                Number of obs   =    118
Log likelihood = -516.9879                    wald chi2(3)    =   1158.49
                                                Prob > chi2     =    0.0000
```

set	Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]
set					
ibr	-61.07753	33.81901	-1.81	0.071	-127.3616 5.206507
usds	-72.49558	2.831238	-25.61	0.000	-78.0447 -66.94645
goldb	.0599955	.0140527	4.27	0.000	.0324528 .0875382
_cons	3167.73	207.2899	15.28	0.000	2761.45 3574.011
ARCH					
arch					
L1.	.9952712	.3383216	2.94	0.003	.332173 1.658369
L2.	-.1255426	.1038731	-1.21	0.227	-.3291301 .078045
_cons	117.519	40.07261	2.93	0.003	38.97815 196.0599

```
. estat ic
```

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	118	.	-516.9879	7	1047.976	1067.371

Estimate GARCH(1,1)

```
. arch set usds, arch(1) garch(1) nolog
```

```
ARCH family regression
```

```
Sample: 02dec2003 to 28mar2004
```

```
Number of obs = 118
```

```
wald chi2(1) = 882.55
```

```
Log likelihood = -518.5556
```

```
Prob > chi2 = 0.0000
```

```
-----
```

set		Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]	
set							
	usds	-74.04275	2.492366	-29.71	0.000	-78.92769	-69.1578
	_cons	3607.771	97.79026	36.89	0.000	3416.106	3799.437
ARCH							
	arch						
	L1.	1.063916	.297246	3.58	0.000	.4813242	1.646507
	garch						
	L1.	-.1115611	.1054421	-1.06	0.290	-.3182239	.0951017
	_cons	112.9776	46.69068	2.42	0.016	21.46551	204.4896

```
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```