

## 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
-----+-----						

```

. est store linear

```

```

. sca logLr=e(l1)

. sca list logLr
    logLr = -458.72372

. 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

. predict ehat, r

. g ehat2=ehat^2

. reg ehat2 l.ehat2 if t>2
-----+-----
      Source |             SS             df             MS             Number of obs =             499
-----+-----+-----+-----+-----+-----
      Model |    11.1919499             1    11.1919499             F(1, 497) =             38.44
      Residual |   144.716148            497     .291179372             Prob > F =             0.0000
-----+-----+-----+-----+-----+-----
      Total |   155.908098            498     .313068469             R-squared =             0.0718
                                           Adj R-squared =             0.0699
                                           Root MSE =             .53961

-----+-----
      ehat2 |             Coef.      Std. Err.      t      P>|t|      [95% Conf. Interval]
-----+-----+-----+-----+-----+-----
      ehat2 |
      L1. |     .2679035     .0432121      6.20   0.000     .1830025     .3528045
      |
      _cons |     .2690816     .0289072      9.31   0.000     .2122862     .3258769
-----+-----

. sca r2=e(r2)

. sca list r2
    r2 = .07178556

. sca N=e(N)

. sca list N
    N =          499

. sca archlm=r2*N

. sca list archlm
    archlm = 35.820994

. 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

-----+-----
      |             OPG
      y |             Coef.      Std. Err.      z      P>|z|      [95% Conf. Interval]
-----+-----+-----+-----+-----+-----
      |
      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

```

-----
. est store garch

. lrtest linear garch, force

Likelihood-ratio test                                LR chi2(3) =      63.95
(Assumption: linear nested in garch)                Prob > chi2 =      0.0000

. test [ARCH]l1.arch [ARCH]l1.garch

( 1) [ARCH]L.arch = 0
( 2) [ARCH]L.garch = 0

           chi2( 2) =    67.73
           Prob > chi2 =    0.0000

. 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.
           arch      garch      _cons
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 uhat_m=y-yhat_m

. g uhat2_m=uhat_m^2

. predict uhat, r

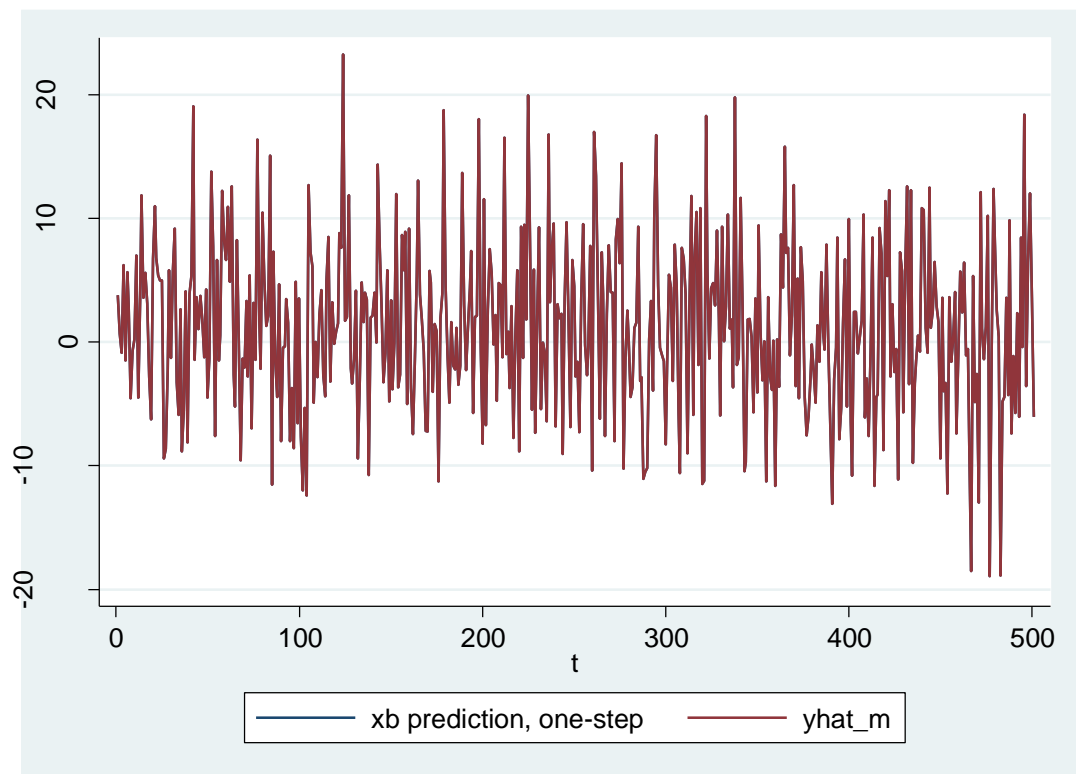
. g uhat2=uhat^2

. g s2hat_m=a0+d1*1.uhat2_m+a1*1.uhat2_m in 2
(500 missing values generated)

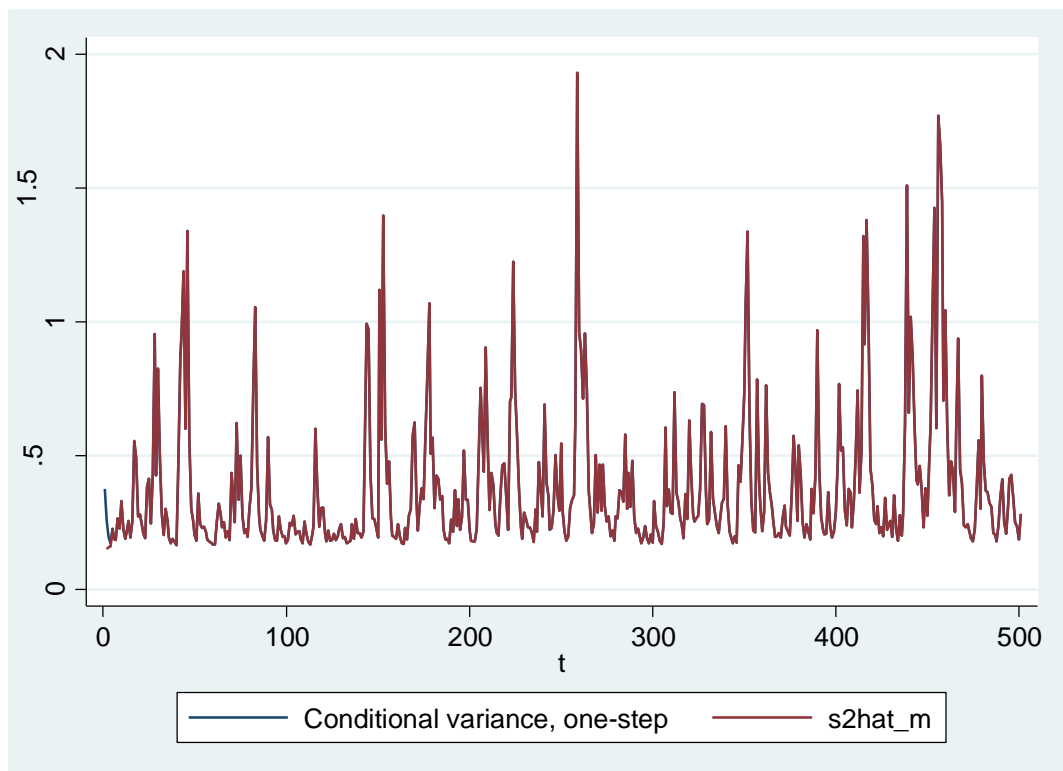
. replace s2hat_m=a0+d1*1.s2hat_m+a1*1.uhat2_m if t>2
(499 real changes made)

```

```
. line yhat yhat_m t
```



```
. twoway (line s2hat t) (line s2hat_m 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).

$\varepsilon_t$  = 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			
Model	215609.598	3	71869.8659	Number of obs =	118	
Residual	79215.8193	114	694.875608	F( 3, 114) =	103.43	
Total	294825.417	117	2519.87536	Prob > F =	0.0000	
				R-squared =	0.7313	
				Adj R-squared =	0.7242	
				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) no1og
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) no1og
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