


```

(240 real changes made)
(240 real changes made)
Iteration 2:   log likelihood = -617.63957
(240 real changes made)
(240 real changes made)
(240 real changes made)
(240 real changes made)
(240 real changes made)
(240 real changes made)
(240 real changes made)
Iteration 3:   log likelihood = -617.44101
(240 real changes made)
(240 real changes made)
(240 real changes made)
(240 real changes made)
(240 real changes made)
Iteration 4:   log likelihood = -617.44095

```

```

Log likelihood = -617.44095
Number of obs   =      240
Wald chi2(1)   =    1455.04
Prob > chi2    =      0.0000

```

y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
x	.9145708	.0239762	38.14	0.000	.8675783 .9615632
_cons	.1363996	.1215726	1.12	0.262	-.1018784 .3746776

The default algorithm in estimating ML is Newton-Raphson (nr), if we would like to use BHHH algorithm, we can use the following command.

```
. ml model lf ml_cauchy (y=x), tech(bhhh)
```

```
. ml maximize
```

```

(240 real changes made)
initial:      log likelihood = -847.98921
...
Iteration 5:   log likelihood = -617.44095

```

```

Log likelihood = -617.44095
Number of obs   =      240
Wald chi2(1)   =    1969.66
Prob > chi2    =      0.0000

```

y	Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]
x	.914568	.0206073	44.38	0.000	.8741785 .9549575
_cons	.1364103	.0914454	1.49	0.136	-.0428195 .31564

2. Normal Distribution

If the disturbances term u_t are assumed to be normal distribution. To estimate the model using maximum likelihood, we use the following command:

```

program ml_norm
  version 10
  args todo b lnf
  tempvar thetal sigma
  mlevel `thetal' = `b', eq(1)
  mlevel `sigma' = `b', eq(2)
  tempvar res
  quietly gen double `res' = $ML_y1-`thetal'
  mlsum `lnf' = -0.5*ln(2*_pi)-ln(`sigma')-0.5*((`res'/ `sigma')^ 2)
end

```

```

. ml model d0 ml_norm (y=x) /sigma

. ml maximize
initial:      log likelihood =      -<inf> (could not be evaluated)
feasible:     log likelihood = -13484.369
rescale:      log likelihood = -831.27985
rescale eq:   log likelihood = -761.59207
Iteration 0:  log likelihood = -761.59207 (not concave)
Iteration 1:  log likelihood = -606.96261
Iteration 2:  log likelihood = -593.59437
Iteration 3:  log likelihood = -593.47885
Iteration 4:  log likelihood = -593.47858
Iteration 5:  log likelihood = -593.47858

Log likelihood = -593.47858
Number of obs   =      240
Wald chi2(1)    =      569.99
Prob > chi2     =      0.0000

-----+-----
      Y |          Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
eq1    x |          .9315372   .039018     23.87   0.000   .8550633   1.008011
      _cons |          .1978772   .1878496     1.05   0.292   -.1703013   .5660558
-----+-----
sigma  _cons |          2.868786   .1309416    21.91   0.000   2.612145   3.125427
-----+-----

. test x=0

( 1) [eq1]x = 0
      chi2( 1) =      569.99
      Prob > chi2 =      0.0000

. est store unres

```

In case of assuming normal distribution, the estimated results are the same as estimating the model using OLS, except the estimated result of standard error of regression (standard deviation of the error term (σ)). In OLS, Root MSE = 2.8808 while in MLE, $\sigma_{\text{cons}} = 2.868786$

```

. regress y x

-----+-----
      Source |          SS          df           MS              Number of obs =      240
-----+-----+-----+-----+-----
      Model | 4691.00669           1   4691.00669          F( 1, 238) = 565.24
      Residual | 1975.18377         238   8.29909148          Prob > F      = 0.0000
-----+-----+-----+-----
      Total | 6666.19046         239   27.8920103          R-squared     = 0.7037
                                          Adj R-squared = 0.7025
                                          Root MSE     = 2.8808

```

	y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
	x	.9315372	.0391816	23.77	0.000	.8543501 1.008724
	_cons	.1978772	.1886373	1.05	0.295	-.1737347 .5694892

. g one=1

. ml model d0 ml_norm (y=one, nocons) /sigma

. ml maximize

initial: log likelihood = -<inf> (could not be evaluated)
feasible: log likelihood = -13484.369
rescale: log likelihood = -831.27985
rescale eq: log likelihood = -761.59207
Iteration 0: log likelihood = -761.59207
Iteration 1: log likelihood = -739.46007
Iteration 2: log likelihood = -739.44504
Iteration 3: log likelihood = -739.44504

Log likelihood = -739.44504

Number of obs	=	240
Wald chi2(1)	=	7.82
Prob > chi2	=	0.0052

	y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
eq1	one	.9513828	.3401948	2.80	0.005	.2846133 1.618152
sigma	_cons	5.270275	.240554	21.91	0.000	4.798798 5.741752

. est store res

. lrtest unres res

Likelihood-ratio test
(Assumption: res nested in unres)

LR chi2(1)	=	291.93
Prob > chi2	=	0.0000