

Truncated & Censored Data

Simulated Data

STATA command

```
set obs 200
g e=rnormal(0,1000)
g lnw=rnormal(2.75,0.6)
g ystar=-2500+1000*lnw+e
g y_truncate=ystar if ystar>0
g y_censor=ystar
replace y_censor=0 if ystar<=0
g y_out=ystar
g out=runiform(0,100000)
replace y_out=out if ystar>2000
```

```
. set obs 200
number of observations (_N) was 0, now 200
```

```
. g e=rnormal(0,1000)
```

```
. g lnw=rnormal(2.75,0.6)
```

```
. g ystar=-2500+1000*lnw+e
```

```
. g y_truncate=ystar if ystar>0
(85 missing values generated)
```

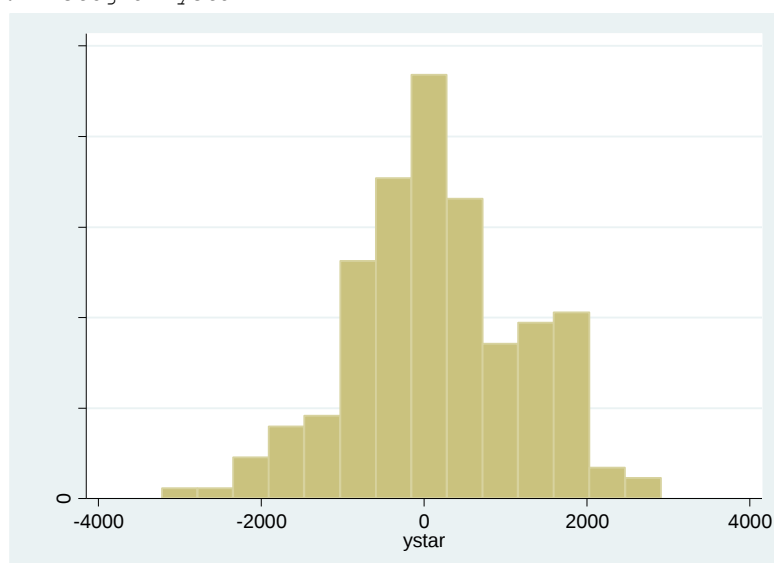
```
. g y_censor=ystar
```

```
. replace y_censor=0 if ystar<=0
(85 real changes made)
```

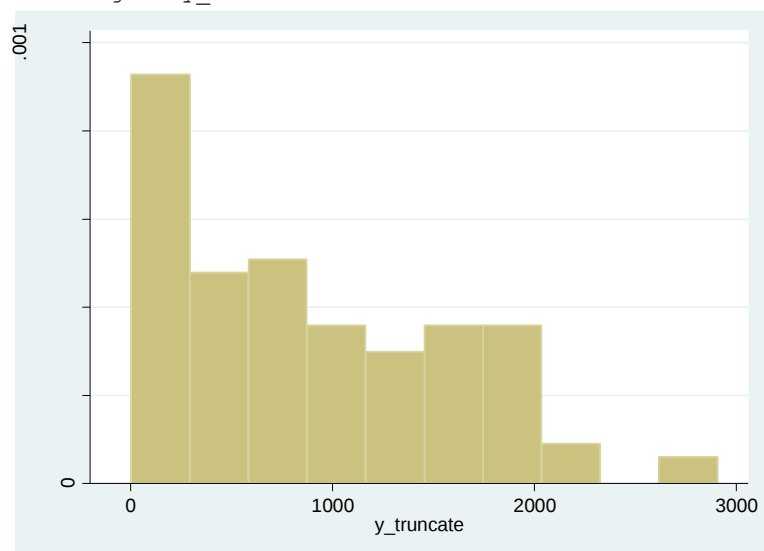
```
. sum ystar y_truncate y_censor lnw
```

Variable	Obs	Mean	Std. Dev.	Min	Max
ystar	200	193.6635	1079.189	-3219.43	2908.469
y_truncate	115	913.6604	693.9303	2.456205	2908.469
y_censor	200	525.3547	693.4551	0	2908.469
lnw	200	2.731851	.6024763	1.138471	4.317463

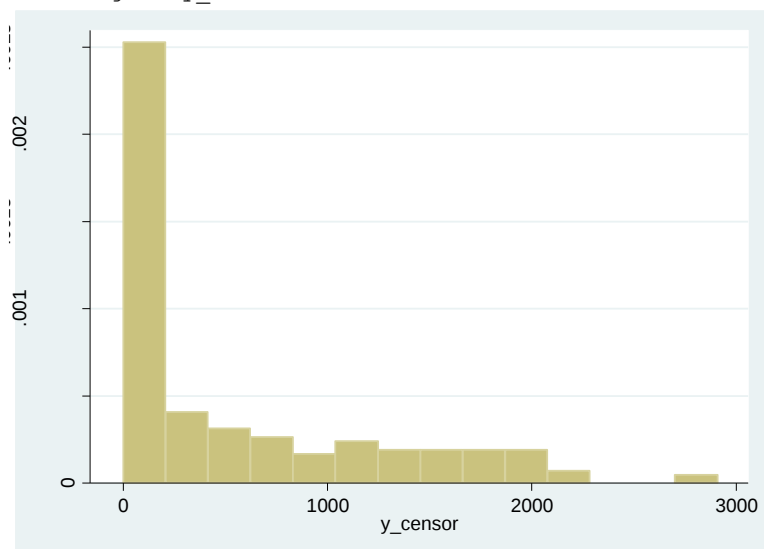
```
. histogram ystar
```



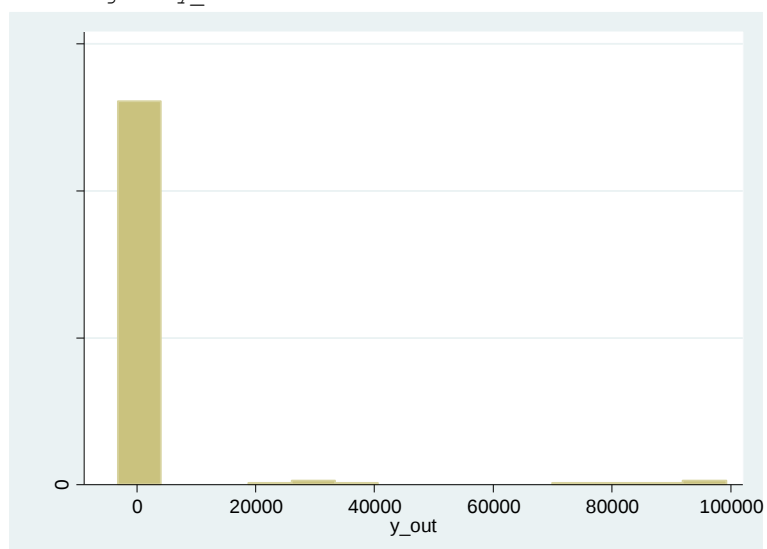
```
. histogram y_truncate
```



```
. histogram y_censor
```



```
. histogram y_out
```



```
. reg ystar lnw
```

Source	SS	df	MS	Number of obs	=	200
Model	46936405.2	1	46936405.2	F(1, 198)	=	50.28
Residual	184828723	198	933478.401	Prob > F	=	0.0000
Total	231765129	199	1164648.89	R-squared	=	0.2025
				Adj R-squared	=	0.1985
				Root MSE	=	966.17

ystar	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnw	806.0991	113.6804	7.09	0.000	581.9193	1030.279
_cons	-2008.479	317.9838	-6.32	0.000	-2635.549	-1381.41

```
. predict uncensored
(option xb assumed; fitted values)
```

```
. est store m_ystar
```

```
. reg y_truncate lnw
```

Source	SS	df	MS	Number of obs	=	115
Model	3091312.84	1	3091312.84	F(1, 113)	=	6.74
Residual	51804162.2	113	458443.913	Prob > F	=	0.0107
Total	54895475	114	481539.255	R-squared	=	0.0563
				Adj R-squared	=	0.0480
				Root MSE	=	677.08

y_truncate	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnw	275.7176	106.1784	2.60	0.011	65.35908	486.0762
_cons	109.8138	315.9333	0.35	0.729	-516.1072	735.7347

```
. predict y_that
(option xb assumed; fitted values)
```

```
. est store m_ythat
```

```
. truncreg y_truncate lnw, ll(0) nolog
(note: 0 obs. truncated)
```

Truncated regression

Limit: lower =	0	Number of obs	=	115
upper =	+inf	Wald chi2(1)	=	5.19
Log likelihood =	-890.08705	Prob > chi2	=	0.0227

y_truncate	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
lnw	710.0844	311.6458	2.28	0.023	99.26996	1320.899
_cons	-1947.443	1174.974	-1.66	0.097	-4250.349	355.4633
/sigma	1063.349	180.9337	5.88	0.000	708.726	1417.973

```
. predict truncated, e(0,.)
```

```
. est store m_ytrunc
```

```
. reg y_censor lnw
```

Source	SS	df	MS	Number of obs	=	200
				F(1, 198)	=	30.80

Model		12882696.1	1	12882696.1	Prob > F	=	0.0000
Residual		82812426.1	198	418244.576	R-squared	=	0.1346
-----+-----							
Total		95695122.1	199	480880.011	Adj R-squared	=	0.1303
					Root MSE	=	646.72

y_censor	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnw	422.3154	76.0937	5.55	0.000	272.2573	572.3735
_cons	-628.3481	212.8472	-2.95	0.004	-1048.087	-208.6097

```
. predict y_hat
(option xb assumed; fitted values)
```

```
. est store m_ychat
```

```
. tobit y_censor lnw, ll(0)
```

Tobit regression	Number of obs	=	200
	LR chi2(1)	=	33.26
	Prob > chi2	=	0.0000
Log likelihood = -1013.3248	Pseudo R2	=	0.0161

y_censor	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnw	739.4631	129.0678	5.73	0.000	484.947	993.9792
_cons	-1829.827	375.1629	-4.88	0.000	-2569.632	-1090.022
/sigma	970.6334	69.3574			833.8636	1107.403

```
85 left-censored observations at y_censor <= 0
115 uncensored observations
0 right-censored observations
```

```
. predict censored, ystar(0,.)
```

```
. est store m_ycensor
```

```
. reg y_out lnw
```

Source		SS	df	MS	Number of obs	=	200
-----+-----							
Model		3.2119e+09	1	3.2119e+09	F(1, 198)	=	17.06
Residual		3.7272e+10	198	188242784	Prob > F	=	0.0001
-----+-----							
Total		4.0484e+10	199	203436844	R-squared	=	0.0793
					Adj R-squared	=	0.0747
					Root MSE	=	13720

y_out	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnw	6668.25	1614.331	4.13	0.000	3484.76	9851.739
_cons	-15364.3	4515.564	-3.40	0.001	-24269.07	-6459.533

```
. predict y_ohat
(option xb assumed; fitted values)
```

```
. est store m_yout
```

```
. reg y_out lnw if y_out<=2000
```

Source		SS	df	MS	Number of obs	=	191
-----+-----							
Model		30189512	1	30189512	F(1, 189)	=	35.66
Residual		159997207	189	846546.067	Prob > F	=	0.0000
-----+-----							
					R-squared	=	0.1587
					Adj R-squared	=	0.1543

Total | 190186719 190 1000982.73 Root MSE = 920.08

y_out	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnw	679.4609	113.7789	5.97	0.000	455.0212	903.9007
_cons	-1736.592	314.0032	-5.53	0.000	-2355.994	-1117.191

. predict y_ohat2000
(option xb assumed; fitted values)

. est store m_yout2000

. tobit y_out lnw, ul(2000)

Tobit regression	Number of obs	=	200
	LR chi2(1)	=	47.11
	Prob > chi2	=	0.0000
Log likelihood = -1595.7889	Pseudo R2	=	0.0145

y_out	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnw	837.9424	115.8093	7.24	0.000	609.5716	1066.313
_cons	-2085.912	323.0429	-6.46	0.000	-2722.939	-1448.885
/sigma	974.443	50.38531			875.0854	1073.801

0 left-censored observations
191 uncensored observations
9 right-censored observations at y_out >= 2000

. predict y_outcensor
(option xb assumed; fitted values)

. est store m_youtcensor

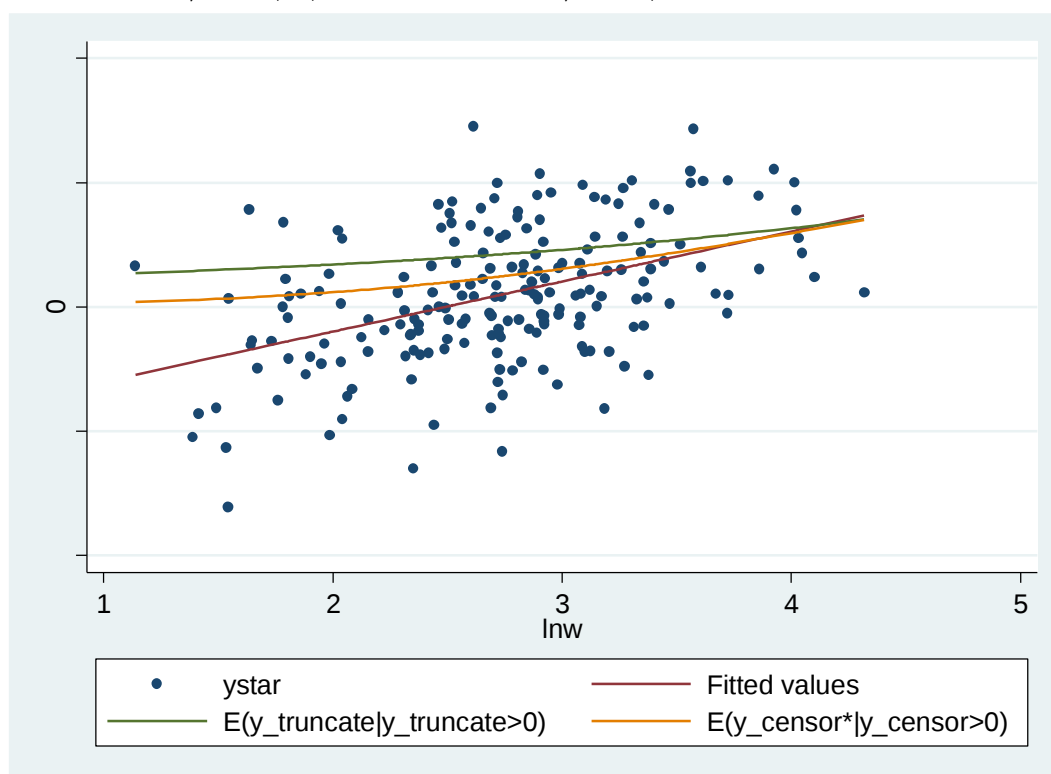
. est table m_ystar m_ythat m_ytrunc m_ychat m_ycensor m_yout m_yout2000
m_youtcensor, star(.1 .05 .01) stat(N rss ll rmse F chi2 r2 r2_a)

Variable	m_ystar	m_ythat	m_ytrunc	m_ychat	m_ycensor	m_yout	m_yout2000	m_youtcensor
lnw	806.09908***	275.71764**		422.31538***		6668.2498***	679.46094***	
_cons	-2008.4794***	109.81375		-628.34814***		-15364.304***	-1736.5923***	
eq1								
lnw			710.08442**					
_cons			-1947.4428*					
sigma								
_cons			1063.3495***		970.63337***			974.44302***
model								
lnw					739.46306***			837.94243***
_cons					-1829.8268***			-2085.9121***
Statistics								
N	200	115	115	200	200	200	191	200
rss	1.848e+08	51804162		82812426		3.727e+10	1.600e+08	
ll	-1657.45	-911.71574	-890.08705	-1577.1648	-1013.3248	-2188.107	-1573.4838	-1595.7889
rmse	966.16686	677.08486		646.71831		13720.16	920.07938	
F	50.28119	6.7430557		30.801825		17.062331	35.661984	
chi2			5.1915558		33.259306			47.114991
r2	.20251712	.05631271		.13462229		.07933668	.15873617	
r2_a	.19848942	.04796149		.13025169		.07468686	.15428504	

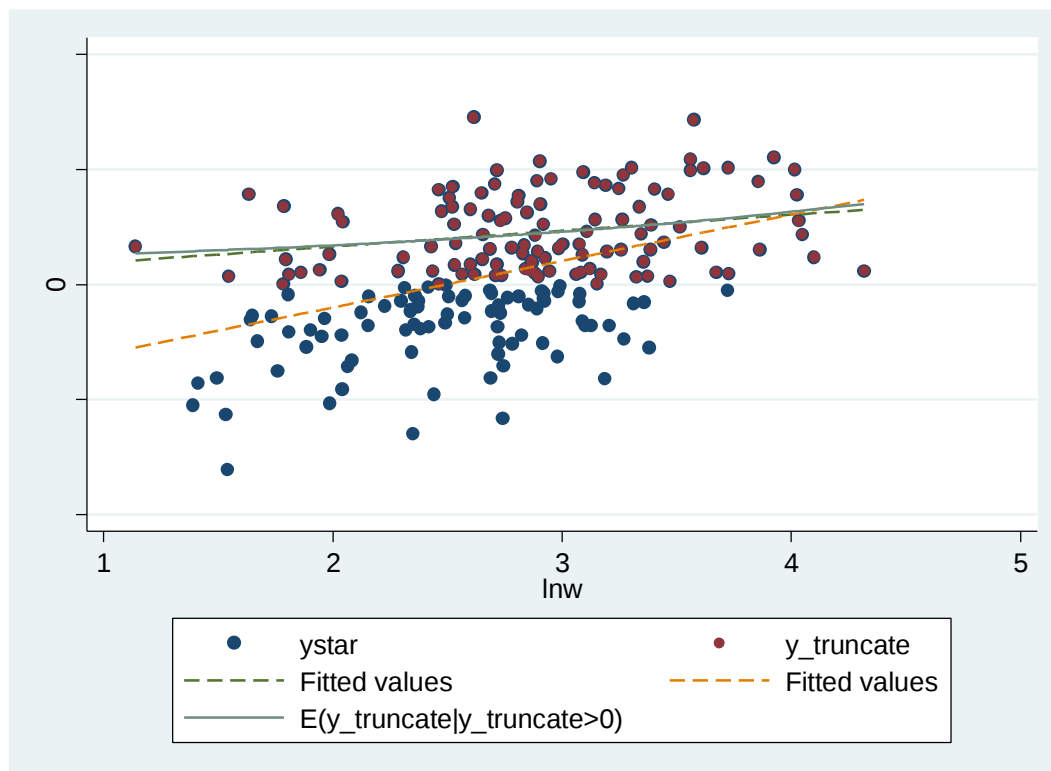
legend: * p<.1; ** p<.05; *** p<.01

	m_ystar	m_ythat	m_ytrunc	m_ychat	m_ycensor
lnw	806.099 ***	275.718 **	710.084 **	422.315 ***	739.463 ***
Constant	-2008.479 ***	109.814	-1947.443 *	-628.348 ***	-1829.827 ***
sigma			1063.350 ***		970.633 ***
N	200	115	115	200	200
RSS	184800000	51804162		82812426	
logL	-1657.45	-911.72	-890.09	-1577.16	-1013.32
RMSE	966.1669	677.0849		646.7183	
F-test	50.2812	6.7431		30.8018	
Chi-square			5.1916		33.2593
R-square	0.2025	0.0563		0.1346	
Adj. R2	0.1985	0.0480		0.1303	

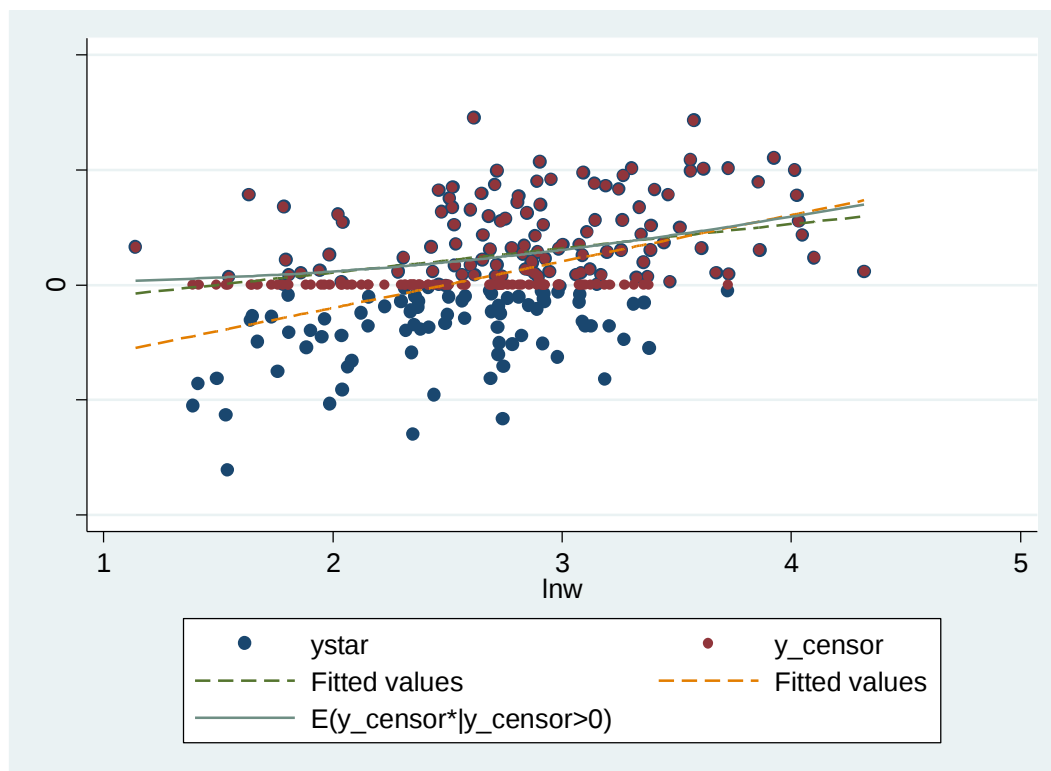
```
. graph twoway (scatter ystar lnw, ms(o)) (line uncensored lnw, sort) (line truncated lnw, sort) (line censored lnw, sort)
```



```
. graph twoway (scatter ystar lnw) (scatter y_truncate lnw, ms(o)) (line y_that
lnw, sort lpattern(dash)) (line uncensored lnw, sort lpattern(dash)) (line
truncated lnw, sort)
```



```
. graph twoway (scatter ystar lnw) (scatter y_censor lnw, ms(o)) (line y_chat
lnw, sort lpattern(dash)) (line uncensored lnw, sort lpattern(dash)) (line
censored lnw, sort)
```



Truncated Regression Model

The model:

$$y_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \beta_4 x_{4i} + \varepsilon_i$$

where

- y_i = wife's hours of work
- x_{1i} = number of children younger than 6
- x_{2i} = number of children between 6 and 18
- x_{3i} = wife's age
- x_{4i} = wife's educational attainment
- ε_i = Error term

```
. regress y x1 x2 x3 x4
```

Source	SS	df	MS	Number of obs = 250		
Model	16526046.1	4	4131511.52	F(4, 245) = 5.27		
Residual	192218058	245	784563.5	Prob > F = 0.0004		
				R-squared = 0.0792		
				Adj R-squared = 0.0641		
				Root MSE = 885.76		
Total	208744104	249	838329.733			

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x1	-462.1233	124.6768	-3.71	0.000	-707.6985	-216.5481
x2	-91.141	45.85001	-1.99	0.048	-181.4515	-.8305151
x3	-13.1577	8.334958	-1.58	0.116	-29.57502	3.259612
x4	53.26156	26.09369	2.04	0.042	1.864986	104.6581
_cons	940.0593	530.7197	1.77	0.078	-105.296	1985.415

```
. regress y x1 x2 x3 x4 if y>0
```

Source	SS	df	MS	Number of obs = 150		
Model	7326995.15	4	1831748.79	F(4, 145) = 2.80		
Residual	94793104.2	145	653745.546	Prob > F = 0.0281		
				R-squared = 0.0717		
				Adj R-squared = 0.0461		
				Root MSE = 808.55		
Total	102120099	149	685369.794			

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x1	-421.4822	167.9734	-2.51	0.013	-753.4748	-89.48953
x2	-104.4571	54.18616	-1.93	0.056	-211.5538	2.639668
x3	-4.784917	9.690502	-0.49	0.622	-23.9378	14.36797
x4	9.353195	31.23793	0.30	0.765	-52.38731	71.0937
_cons	1629.817	615.1301	2.65	0.009	414.0371	2845.597

```
. truncreg y x1 x2 x3 x4, ll(0) nolog
(note: 100 obs. truncated)
```

Truncated regression

Limit: lower = 0
upper = +inf
Log likelihood = -1200.9157

Number of obs = 150
wald chi2(4) = 10.05
Prob > chi2 = 0.0395

y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
x1	-803.0042	321.3614	-2.50	0.012	-1432.861	-173.1474
x2	-172.875	88.72898	-1.95	0.051	-346.7806	1.030578
x3	-8.821122	14.36848	-0.61	0.539	-36.98283	19.34059
x4	16.52873	46.50375	0.36	0.722	-74.61695	107.6744
_cons	1586.26	912.355	1.74	0.082	-201.9233	3374.442
/sigma	983.7262	94.44303	10.42	0.000	798.6213	1168.831

Marginal Effect on Truncated Expected Value

```
. mfx compute, predict(e(0,.))
```

```
Marginal effects after truncreg
      y = E(y|y>0) (predict, e(0,.))
      = 1310.0294
```

variable	dy/dx	Std. Err.	z	P> z	[95% C.I.]		x
x1	-521.9979	199.4	-2.62	0.009	-912.818	-131.178	.173333
x2	-112.3785	56.487	-1.99	0.047	-223.091	-1.66638	1.31333
x3	-5.734226	9.31985	-0.62	0.538	-24.0008	12.5323	42.7867
x4	10.7446	30.21	0.36	0.722	-48.4655	69.9547	12.64

Marginal Effect on Censored Expected Value

```
. mfx compute, predict(ystar(0,.))
```

```
Marginal effects after truncreg
      y = E(y*|y>0) (predict, ystar(0,.))
      = 1123.2815
```

variable	dy/dx	Std. Err.	z	P> z	[95% C.I.]		x
x1	-688.534	266.12	-2.59	0.010	-1210.11	-166.957	.173333
x2	-148.2313	74.802	-1.98	0.048	-294.841	-1.62154	1.31333
x3	-7.56365	12.299	-0.61	0.539	-31.6686	16.5413	42.7867
x4	14.17252	39.853	0.36	0.722	-63.9373	92.2823	12.64

Tobit Model: Factors Determining Research and Development Expenditures of the Firm in Thai Electronic Industry

Tobit (Censored regression) Model

Metakunnavat (1999) studied factors determined research and development (R&D) expenditures of the firm in Thai electronic industry. The study applied Tobit model since the data set was censored distributed. The model is as follows:

$$\begin{aligned} Z_i &= \alpha + \beta_1 MS_i + \beta_2 PCM_i + \beta_4 EXI_i + \beta_5 TOR_i + \beta_6 TOA_i + \beta_7 GOV_i + u_i & \text{if } Z_{ik} > 0 \\ Z_i &= 0 & \text{Otherwise} \end{aligned}$$

where:

- Z_i = Research and development expenditures of firm i
- MS_i = Market share of firm i
- PCM_i = Profit contribution margin of firm i
- CR_i = Concentration ratio of the industry
- EXI_i = Export concentration of the industry
- TOR_i = Dummy Variable = 1 if technology change within or more than 1 year
= 0 otherwise
- TOA_i = Dummy Variable = 1 if technology transfers from mother company
= 0 otherwise
- GOV_i = Dummy Variable = 1 if the industry receives government incentive for R&D
= 0 otherwise
- u_i = Error term

α and β_k are coefficients, where $\beta_1, \beta_2, \beta_3, \beta_7$ can be positive or negative, β_4 should be positive, β_5 and β_6 should be negative.

Since the data is censored distributed, the model should be estimated by using Maximum Likelihood Estimation (MLE) method. The log-likelihood function is as follows:

$$\ln L = \sum_0 \ln \left[\frac{1}{\sqrt{(2\pi\sigma^2)}} \right] - \sum_1 \frac{1}{2\sigma^2} (Z_i = \alpha + \beta_1 MS_i + \beta_2 PCM_i + \beta_4 EXI_i + \beta_5 TOR_i + \beta_6 TOA_i + \beta_7 GOV_i + u_i)^2$$

- where $\sum_0$ is sum of observation N_0 for $Z_i = 0$
- $\sum_1$ is sum of observation N_1 for $Z_i > 0$

Estimate Tobit Model

```
. reg z ms pcm exi tor toa gov
```

Source	SS	df	MS	Number of obs =	39
Model	216.685223	6	36.1142038	F(6, 32) =	32.63
Residual	35.4154445	32	1.10673264	Prob > F =	0.0000
				R-squared =	0.8595
				Adj R-squared =	0.8332
				Root MSE =	1.052
Total	252.100667	38	6.63422808		

z	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ms	-.0640993	.0351223	-1.83	0.077	-.1356411 .0074426
pcm	.0309549	.0168661	1.84	0.076	-.0034001 .06531
exi	-.0400559	.0277357	-1.44	0.158	-.0965517 .0164399
tor	-.2361991	.7166092	-0.33	0.744	-1.695884 1.223486
toa	-.3944311	.8892493	-0.44	0.660	-2.205773 1.41691
gov	4.507958	.5964943	7.56	0.000	3.292939 5.722977
_cons	2.127431	.639876	3.32	0.002	.8240464 3.430816

```
. reg z ms pcm exi tor toa gov if z>0
```

Source	SS	df	MS	Number of obs =	32
Model	175.885314	6	29.314219	F(6, 25) =	28.21
Residual	25.9772868	25	1.03909147	Prob > F =	0.0000
				R-squared =	0.8713
				Adj R-squared =	0.8404
				Root MSE =	1.0194
Total	201.862601	31	6.5116968		

z	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ms	-.1203294	.039932	-3.01	0.006	-.2025708 -.038088
pcm	-3.27e-06	.0195447	-0.00	1.000	-.0402564 .0402498
exi	-.004589	.0555295	-0.08	0.935	-.1189541 .109776
tor	-.7234769	.7208991	-1.00	0.325	-2.208196 .7612425
toa	-.0729069	.8929615	-0.08	0.936	-1.911996 1.766182
gov	4.227732	.5889693	7.18	0.000	3.014727 5.440737
_cons	3.710537	.8150497	4.55	0.000	2.03191 5.389163

```
. tobit z ms pcm exi tor toa gov, ll
```

Tobit regression

Log likelihood = -50.51938

Number of obs = 39
 LR chi2(6) = 72.99
 Prob > chi2 = 0.0000
 Pseudo R2 = 0.4194

z	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ms	-.0477107	.0371136	-1.29	0.208	-.1232189 .0277974
pcm	.0348347	.0176433	1.97	0.057	-.0010609 .0707303
exi	-.0868502	.037064	-2.34	0.025	-.1622576 -.0114429
tor	-.0636297	.7400413	-0.09	0.932	-1.569255 1.441996
toa	-.3227235	.9279608	-0.35	0.730	-2.210674 1.565227
gov	4.603485	.6137318	7.50	0.000	3.354838 5.852132
_cons	1.857899	.6802613	2.73	0.010	.4738968 3.241901
/sigma	1.078782	.1373532			.7993348 1.358229

obs. summary: 7 left-censored observations at z<=0
 32 uncensored observations
 0 right-censored observations

Marginal Effect on Censored Expected Value

```
. mfx compute, predict(ystar(0,.))
```

```
Marginal effects after tobit
      y = E(z*|z>0) (predict, ystar(0,.))
      = 2.1685388
```

variable	dy/dx	Std. Err.	z	P> z	[	95% C.I.	]	x
ms	-.0466297	.03637	-1.28	0.200	-.117921	.024661		8.87769
pcm	.0340454	.01723	1.98	0.048	.000284	.067807		25.741
exi	-.0848824	.03577	-2.37	0.018	-.154982	-.014783		7.34615
tor*	-.0621636	.72276	-0.09	0.931	-1.47874	1.35441		.384615
toa*	-.3142121	.89959	-0.35	0.727	-2.07738	1.44895		.307692
gov*	4.568452	.60762	7.52	0.000	3.37755	5.75936		.128205

(*) dy/dx is for discrete change of dummy variable from 0 to 1

Marginal Effect on Truncated Expected Value

```
. mfx compute, predict(e(0,.))
```

```
Marginal effects after tobit
      y = E(z|z>0) (predict, e(0,.))
      = 2.2188111
```

variable	dy/dx	Std. Err.	z	P> z	[	95% C.I.	]	x
ms	-.0423084	.03336	-1.27	0.205	-.107687	.023071		8.87769
pcm	.0308903	.0157	1.97	0.049	.000118	.061662		25.741
exi	-.0770161	.03163	-2.44	0.015	-.139005	-.015027		7.34615
tor*	-.0563431	.65456	-0.09	0.931	-1.33925	1.22656		.384615
toa*	-.2824412	.8012	-0.35	0.724	-1.85277	1.28789		.307692
gov*	4.442333	.59864	7.42	0.000	3.26902	5.61564		.128205

(*) dy/dx is for discrete change of dummy variable from 0 to 1

Tobit Model using Panel Data

The model:

```
. reg illiq cg3 cg4 cg5 priceinverse volatility age lnta lntv
```

Source	SS	df	MS	Number of obs	=	2,977
				F(8, 2968)	=	97.15
Model	56363870.5	8	7045483.82	Prob > F	=	0.0000
Residual	215247984	2,968	72522.9055	R-squared	=	0.2075
				Adj R-squared	=	0.2054
Total	271611854	2,976	91267.4241	Root MSE	=	269.3

illiq	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
cg3	24.80985	13.18405	1.88	0.060	-1.040957	50.66066
cg4	-7.020101	13.5703	-0.52	0.605	-33.62825	19.58805
cg5	-.3996451	19.35628	-0.02	0.984	-38.35273	37.55344
priceinverse	17.58497	4.064562	4.33	0.000	9.61532	25.55461
volatility	49.96149	12.54445	3.98	0.000	25.36479	74.55819
age	-.7652942	.6221881	-1.23	0.219	-1.985258	.4546697
lnta	2.136847	3.549215	0.60	0.547	-4.822325	9.096019
lntv	-30.41288	1.190944	-25.54	0.000	-32.74804	-28.07772
_cons	410.1833	51.51606	7.96	0.000	309.1725	511.1941

```
. xtreg illiq cg3 cg4 cg5 priceinverse volatility age lnta lntv
```

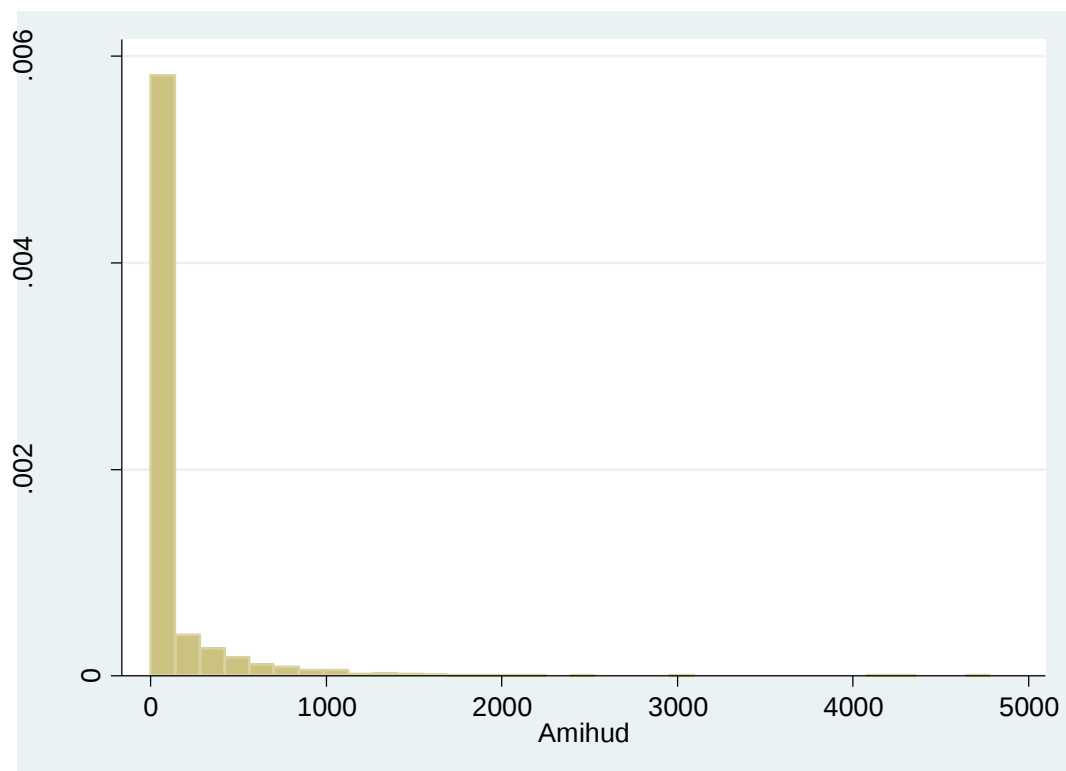
```
Random-effects GLS regression           Number of obs   =       2,977
Group variable: id                     Number of groups =       364
```

```
R-sq:                                Obs per group:
    within = 0.2147                      min =           1
    between = 0.1891                     avg  =          8.2
    overall = 0.2075                     max  =         11
```

```
corr(u_i, X)   = 0 (assumed)           Wald chi2(8)     =       777.19
                                           Prob > chi2      =       0.0000
```

illiq	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
cg3	24.80985	13.18405	1.88	0.060	-1.030415	50.65012
cg4	-7.020101	13.5703	-0.52	0.605	-33.6174	19.5772
cg5	-.3996451	19.35628	-0.02	0.984	-38.33726	37.53797
priceinverse	17.58497	4.064562	4.33	0.000	9.61857	25.55136
volatility	49.96149	12.54445	3.98	0.000	25.37482	74.54816
age	-.7652942	.6221881	-1.23	0.219	-1.98476	.4541722
lnta	2.136847	3.549215	0.60	0.547	-4.819487	9.093181
lntv	-30.41288	1.190944	-25.54	0.000	-32.74708	-28.07867
_cons	410.1833	51.51606	7.96	0.000	309.2137	511.1529
sigma_u	0					
sigma_e	263.35265					
rho	0	(fraction of variance due to u_i)				

```
. histogram illiq
(bin=34, start=0, width=140.68379)
```



```
. reg illiq cg3 cg4 cg5 priceinverse volatility age lnta lntv if illiq<=10
```

Source	SS	df	MS	Number of obs	=	1,955
Model	4107.06559	8	513.383199	F(8, 1946)	=	119.48
Residual	8361.3469	1,946	4.29668392	Prob > F	=	0.0000
				R-squared	=	0.3294
				Adj R-squared	=	0.3266
Total	12468.4125	1,954	6.38096852	Root MSE	=	2.0728

illiq	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
cg3	.1123272	.1307872	0.86	0.391	-.1441704	.3688249
cg4	.0250911	.1295811	0.19	0.846	-.2290412	.2792234
cg5	-.2842084	.1745142	-1.63	0.104	-.6264628	.0580461
priceinverse	.0490283	.0361919	1.35	0.176	-.0219506	.1200072
volatility	.4704564	.3073261	1.53	0.126	-.1322667	1.073179
age	-.0057336	.0061178	-0.94	0.349	-.0177317	.0062646
lnta	-.0873353	.0322229	-2.71	0.007	-.1505304	-.0241403
lntv	-.3887633	.0141307	-27.51	0.000	-.4164762	-.3610503
_cons	8.613544	.4742816	18.16	0.000	7.683391	9.543697

```
. est store m_ols
```

```
. xtreg illiq cg3 cg4 cg5 priceinverse volatility age lnta lntv if illiq<=10
```

Random-effects GLS regression	Number of obs	=	1,955
Group variable: id	Number of groups	=	363
R-sq:	Obs per group:		
within = 0.2137	min =		1
between = 0.3917	avg =		5.4
overall = 0.3265	max =		11

```
corr(u_i, X) = 0 (assumed)      Wald chi2(8) = 675.63
                                Prob > chi2 = 0.0000
```

illiq	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
cg3	.1798028	.138313	1.30	0.194	-.0912858	.4508913
cg4	.0675304	.1518841	0.44	0.657	-.2301568	.3652177
cg5	-.3543256	.206638	-1.71	0.086	-.7593285	.0506774
priceinverse	.0458346	.0449247	1.02	0.308	-.0422162	.1338855
volatility	.4216365	.2782431	1.52	0.130	-.1237101	.966983
age	-.0055548	.0101237	-0.55	0.583	-.0253969	.0142873
lnta	-.1492212	.0498566	-2.99	0.003	-.2469383	-.0515042
lntv	-.3708759	.0156255	-23.74	0.000	-.4015014	-.3402505
_cons	9.305451	.753601	12.35	0.000	7.82842	10.78248
sigma_u	1.237113					
sigma_e	1.7903483					
rho	.32316608	(fraction of variance due to u_i)				

```
. est store m_re
```

```
. tobit illiq cg3 cg4 cg5 priceinverse volatility age lnta lntv, ul(10)
```

```
Tobit regression      Number of obs = 2,977
                      LR chi2(8) = 2533.12
                      Prob > chi2 = 0.0000
Log likelihood = -6029.448      Pseudo R2 = 0.1736
```

illiq	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
cg3	.2172985	.2003655	1.08	0.278	-.1755708	.6101678
cg4	-.3826641	.202276	-1.89	0.059	-.7792795	.0139513
cg5	-1.071681	.2817109	-3.80	0.000	-1.624049	-.5193121
priceinverse	.4909429	.0600655	8.17	0.000	.3731686	.6087172
volatility	1.280187	.409213	3.13	0.002	.4778173	2.082557
age	-.0608728	.0093941	-6.48	0.000	-.0792925	-.0424532
lnta	.1099787	.0518977	2.12	0.034	.0082196	.2117379
lntv	-1.026462	.0198271	-51.77	0.000	-1.065339	-.9875858
_cons	16.63525	.7591861	21.91	0.000	15.14666	18.12383
/sigma	3.728514	.0639989			3.603027	3.854001

```
0 left-censored observations
1,955 uncensored observations
1,022 right-censored observations at illiq >= 10
```

```
. est store m_tobit
```

```
. xtlogit illiq cg3 cg4 cg5 priceinverse volatility age lnta lntv, ul(10) tobit
```

```
Fitting comparison model:
```

```
Fitting constant-only model:
```

```
Iteration 0: log likelihood = -7610.0147
```

```
...
```

```
Iteration 4: log likelihood = -7296.0086
```

```
Fitting full model:
```

```
Iteration 0: log likelihood = -6428.0774
```

```
...
```

```
Iteration 4: log likelihood = -6029.448
```

Obtaining starting values for full model:

Iteration 0: log likelihood = -20869.134
 ...
 Iteration 3: log likelihood = -20868.303

Fitting full model:

Iteration 0: log likelihood = -13142.023 (not concave)
 ...
 Iteration 24: log likelihood = -5963.092

Random-effects tobit regression Number of obs = 2,977
 Group variable: id Number of groups = 364

Random effects u_i ~ Gaussian Obs per group:
 min = 1
 avg = 8.2
 max = 11

Integration method: mvaghermite Integration pts. = 12

Log likelihood = -5963.092 Wald chi2(8) = 2733.60
 Prob > chi2 = 0.0000

	illiq	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
	cg3	.2228948	.2097675	1.06	0.288	-.1882418	.6340315
	cg4	-.4571934	.2307403	-1.98	0.048	-.909436	-.0049508
	cg5	-1.323098	.3258371	-4.06	0.000	-1.961727	-.6844687
priceinverse		.4500789	.0707748	6.36	0.000	.3113629	.588795
volatility		.9726182	.3663227	2.66	0.008	.254639	1.690597
age		-.0727522	.015117	-4.81	0.000	-.1023811	-.0431234
lnta		.07527	.0776228	0.97	0.332	-.076868	.227408
lntv		-1.110499	.0219535	-50.58	0.000	-1.153527	-1.067471
_cons		18.19677	1.177496	15.45	0.000	15.88892	20.50462
/sigma_u		1.787151	.1292638	13.83	0.000	1.533799	2.040503
/sigma_e		3.371697	.0621459	54.25	0.000	3.249893	3.493501
rho		.219328	.0264152			.1711513	.274485

LR test of sigma_u=0: chibar2(01) = 132.71 Prob >= chibar2 = 0.000

0 left-censored observations
 1,955 uncensored observations
 1,022 right-censored observations

. est store m_tobitRE

```
. est table m_ols m_re m_tobit m_tobitRE, star(.1 .05 .01) stat(N rss ll rmse F chi2 r2)
```

Variable	m_ols	m_re	m_tobit	m_tobitRE
cg3	.11232723	.17980277		
cg4	.02509113	.06753045		
cg5	-.28420835	-.35432556*		
priceinverse	.04902826	.04583464		
volatility	.47045635	.42163649		
age	-.00573357	-.00555479		
lnta	-.08733535***	-.14922121***		
lntv	-.38876328***	-.37087594***		
_cons	8.6135439***	9.3054512***		
model				
cg3			.21729851	
cg4			-.38266409*	
cg5			-1.0716805***	
priceinverse			.4909429***	
volatility			1.2801871***	
age			-.06087281***	
lnta			.10997874**	
lntv			-1.0264622***	
_cons			16.635248***	
sigma				
_cons			3.7285138***	
illiq				
cg3				.2228948
cg4				-.45719339**
cg5				-1.3230978***
priceinverse				.45007892***
volatility				.97261821***
age				-.07275223***
lnta				.07527003
lntv				-1.1104994***
_cons				18.196771***
sigma_u				
_cons				1.787151***
sigma_e				
_cons				3.3716973***
Statistics				
N	1955	1955	2977	2977
rss	8361.3469			
ll	-4194.5565		-6029.448	-5963.092
rmse	2.0728444	1.7955137		
F	119.48359			
chi2		675.63485	2533.1212	2733.6048
r2	.32939764			

legend: * p<.1; ** p<.05; *** p<.01

Variable	m_ols	m_re	m_tobit	m_tobitRE
cg3	0.1123	0.1798	0.2173	0.2229
cg4	0.0251	0.0675	-0.3827 *	-0.4572 **
cg5	-0.2842	-0.3543 *	-1.0717 ***	-1.3231 ***
priceinverse	0.0490	0.0458	0.4909 ***	0.4501 ***
volatility	0.4705	0.4216	1.2802 ***	0.9726 ***
age	-0.0057	-0.0056	-0.0609 ***	-0.0728 ***
lnta	-0.0873 ***	-0.1492 ***	0.1100 **	0.0753
lntv	-0.3888 ***	-0.3709 ***	-1.0265 ***	-1.1105 ***
Constant	8.6135 ***	9.3055 ***	16.6352 ***	18.1968 ***
sigma_u				1.7872 ***
sigma_e			3.7285 ***	3.3717 ***
N	1955	1955	2977	2977
RSS	8361.35			
Log-likelihood	-4194.6		-6029.4	-5963.09
RMSE	2.07284	1.79551		
F-test	119.484 ***			
Chi-square Test		675.635 ***	2533.12 ***	2733.605 ***
Chi2 Comparison				132.712 ***
R-square	0.3294			
Overall R2		0.32654		
Pseudo R2			0.1736	