

Limited dependent variable estimation in Stata 11 (Spring, 2010)

```

. //
. //
. //
. //
. // Binomial probit
. //
. // womenwk from IMEUS
. summarize work age married children education

```

Variable	Obs	Mean	Std. Dev.	Min	Max
work	2000	.6715	.4697852	0	1
age	2000	36.208	8.28656	20	59
married	2000	.6705	.4701492	0	1
children	2000	1.6445	1.398963	0	5
education	2000	13.084	3.045912	10	20

```

. probit work age married children education, nolog
Probit regression
Number of obs = 2000
LR chi2(4) = 478.32
Prob > chi2 = 0.0000
Pseudo R2 = 0.1889
Log likelihood = -1027.0616

```

	work	age	married	children	education	_cons
Coef.		.0347211	.4308575	.4473249	.0583645	-2.467365
Std. Err.		.0042293	.074208	.0287417	.0109742	.1925635
z		8.21	5.81	15.56	5.32	-12.81
P> z		0.000	0.000	0.000	0.000	0.000
[95% Conf. Interval]		.0264318 .0430105	.2854125 .5763025	.3909922 .5036576	.0368555 .0798735	-2.844782 -2.089948

```

. //
. //
. //
. // Marginal effects (AMEs)
. //
. //
. margins, dydx(_all)
Average marginal effects
Model VCE : OIM
Expression : Pr(work), predict()
dy/dx w.r.t. : age married children education

```

	age	married	children	education
dy/dx	.0100768	.1250441	.1298233	.0169386
Delta-method Std. Err.	.0011647	.0210541	.0068418	.0031183
z	8.65	5.94	18.98	5.43
P> z	0.000	0.000	0.000	0.000
[95% Conf. Interval]	.0077941 .0123595	.0837788 .1663094	.1164137 .1432329	.0108269 .0230504

```

. // marginal effects for specific values of explanatory variables
. margins, dydx(_all) over(children)
Average marginal effects
Model VCE : OIM

```

Expression : Pr(work), predict()  
 dy/dx w.r.t. : age married children education  
 over : children

		Delta-method				
		dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]
age						
	children					
	0	.0122895	.0013949	8.81	0.000	.0095555 .0150236
	1	.0118794	.0013592	8.74	0.000	.0092154 .0145434
	2	.0103799	.0011888	8.73	0.000	.0080498 .0127099
	3	.0082797	.0010587	7.82	0.000	.0062047 .0103547
	4	.0038001	.00063	6.03	0.000	.0025652 .0050349
	5	.0009481	.0002622	3.62	0.000	.0004342 .001462
married						
	children					
	0	.1525019	.0259523	5.88	0.000	.1016363 .2033675
	1	.1474125	.0245563	6.00	0.000	.099283 .1955419
	2	.1288045	.0223317	5.77	0.000	.0850352 .1725737
	3	.1027434	.0178269	5.76	0.000	.0678032 .1376835
	4	.0471552	.0087329	5.40	0.000	.0300391 .0642714
	5	.0117653	.0029637	3.97	0.000	.0059567 .017574
children						
	children					
	0	.1583306	.0094474	16.76	0.000	.1398141 .1768471
	1	.1530466	.0099778	15.34	0.000	.1334904 .1726028
	2	.1337274	.007549	17.71	0.000	.1189316 .1485231
	3	.1066703	.0040684	26.22	0.000	.0986964 .1146441
	4	.0489575	.0030526	16.04	0.000	.0429744 .0549406
	5	.012215	.0024817	4.92	0.000	.007351 .017079
education						
	children					
	0	.0206581	.0037314	5.54	0.000	.0133447 .0279715
	1	.0199687	.0036937	5.41	0.000	.0127291 .0272083
	2	.017448	.0032004	5.45	0.000	.0111753 .0237207
	3	.0139178	.0026805	5.19	0.000	.008664 .0191715
	4	.0063877	.0013844	4.61	0.000	.0036743 .0091011
	5	.0015937	.0005207	3.06	0.002	.0005732 .0026143

. margins, dydx(\_all) over(married children)

Average marginal effects Number of obs = 2000

Model VCE : OIM

Expression : Pr(work), predict()

dy/dx w.r.t. : age married children education

over : married children

		Delta-method				
		dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]
age						
	married#					

children						
0 0	.0101772	.0012324	8.26	0.000	.0077618	.0125926
0 1	.0128782	.0014693	8.76	0.000	.0099984	.015758
0 2	.0125766	.0014342	8.77	0.000	.0097657	.0153876
0 3	.0111407	.0013874	8.03	0.000	.0084214	.01386
0 4	.0066499	.0009966	6.67	0.000	.0046965	.0086033
0 5	.0030729	.0006206	4.95	0.000	.0018567	.0042892
1 0	.0129026	.0014692	8.78	0.000	.0100231	.0157822
1 1	.011531	.0013277	8.68	0.000	.0089288	.0141333
1 2	.0080563	.0009752	8.26	0.000	.0061449	.0099678
1 3	.0053679	.0008026	6.69	0.000	.0037948	.006941
1 4	.0026145	.0005241	4.99	0.000	.0015873	.0036418
1 5	.0008761	.0002533	3.46	0.001	.0003797	.0013725
married						
married#						
children						
0 0	.1262901	.0185512	6.81	0.000	.0899304	.1626498
0 1	.1598069	.026885	5.94	0.000	.1071133	.2125004
0 2	.1560647	.0290645	5.37	0.000	.0990992	.2130301
0 3	.1382456	.0264296	5.23	0.000	.0864445	.1900466
0 4	.0825189	.0178597	4.62	0.000	.0475147	.1175232
0 5	.0381323	.0104974	3.63	0.000	.0175577	.0587069
1 0	.1601096	.0282537	5.67	0.000	.1047334	.2154858
1 1	.1430895	.0238027	6.01	0.000	.0964371	.1897419
1 2	.0999715	.0154146	6.49	0.000	.0697594	.1301837
1 3	.066611	.0095902	6.95	0.000	.0478146	.0854075
1 4	.0324439	.0054681	5.93	0.000	.0217267	.0431611
1 5	.0108718	.0027682	3.93	0.000	.0054462	.0162974
children						
married#						
children						
0 0	.1311169	.0061968	21.16	0.000	.1189715	.1432624
0 1	.1659147	.0101364	16.37	0.000	.1460477	.1857817
0 2	.1620295	.0104555	15.50	0.000	.1415371	.1825219
0 3	.1435293	.007698	18.64	0.000	.1284415	.1586172
0 4	.0856728	.0051047	16.78	0.000	.0756678	.0956779
0 5	.0395897	.0054331	7.29	0.000	.028941	.0502385
1 0	.166229	.0109681	15.16	0.000	.1447319	.1877261
1 1	.1485584	.0100644	14.76	0.000	.1288326	.1682843
1 2	.1037925	.0056113	18.50	0.000	.0927945	.1147904
1 3	.0691569	.0035835	19.30	0.000	.0621334	.0761803
1 4	.0336839	.0036321	9.27	0.000	.0265651	.0408028
1 5	.0112873	.0024416	4.62	0.000	.0065019	.0160728
education						
married#						
children						
0 0	.0171074	.0032533	5.26	0.000	.0107311	.0234838
0 1	.0216477	.0039896	5.43	0.000	.0138283	.029467
0 2	.0211407	.0038668	5.47	0.000	.0135619	.0287195
0 3	.0187269	.0035581	5.26	0.000	.0117532	.0257007
0 4	.0111781	.0022001	5.08	0.000	.006866	.0154902
0 5	.0051655	.0013474	3.83	0.000	.0025246	.0078063

1 0	.0216887	.0038982	5.56	0.000	.0140483	.029329
1 1	.0193831	.0035977	5.39	0.000	.0123317	.0264345
1 2	.0135423	.0025446	5.32	0.000	.0085549	.0185297
1 3	.0090232	.0018804	4.80	0.000	.0053378	.0127087
1 4	.0043949	.0011037	3.98	0.000	.0022318	.006558
1 5	.0014727	.0004977	2.96	0.003	.0004973	.0024481

```
. //
. //                               Binomial logit
. //
. logit work age married children education, nolog
Logistic regression                Number of obs   =       2000
                                   LR chi2(4)      =       476.62
                                   Prob > chi2     =       0.0000
Log likelihood = -1027.9144        Pseudo R2   =       0.1882
```

work	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
age	.0579303	.007221	8.02	0.000	.0437773 .0720833
married	.7417775	.1264705	5.87	0.000	.4938998 .9896552
children	.7644882	.0515289	14.84	0.000	.6634935 .865483
education	.0982513	.0186522	5.27	0.000	.0616936 .134809
_cons	-4.159247	.3320401	-12.53	0.000	-4.810034 -3.508461

```
. margins, dydx(_all)
Average marginal effects                Number of obs   =       2000
Model VCE      : OIM
Expression    : Pr(work), predict()
dy/dx w.r.t. : age married children education
```

	Delta-method				
	dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]
age	.0099674	.0011682	8.53	0.000	.0076778 .0122569
married	.127629	.021152	6.03	0.000	.0861717 .1690862
children	.1315365	.007073	18.60	0.000	.1176736 .1453994
education	.0169049	.0031243	5.41	0.000	.0107814 .0230285

```
. //
. //                               Ordered logit
. //
. // panel84extract from IMEUS
. // dia: change in Inc/Asset ratio, 1982-1983
. summarize rating83c ia83 dia
```

Variable	Obs	Mean	Std. Dev.	Min	Max
rating83c	98	3.479592	1.17736	2	5
ia83	98	10.11473	7.441946	-13.08016	30.74564
dia	98	.7075242	4.711211	-10.79014	20.05367

```
. tabulate rating83c
```

```
      Bond |
rating, |
```

1983	Freq.	Percent	Cum.
BA_B_C	26	26.53	26.53
BAA	28	28.57	55.10
AA_A	15	15.31	70.41
AAA	29	29.59	100.00
Total	98	100.00	

```
. ologit rating83c ia83 dia, nolog
```

```
Ordered logistic regression
```

```
Number of obs = 98
```

```
LR chi2(2) = 11.54
```

```
Prob > chi2 = 0.0031
```

```
Pseudo R2 = 0.0434
```

```
Log likelihood = -127.27146
```

rating83c	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
ia83	.0939166	.0296196	3.17	0.002	.0358633 .1519699
dia	-.0866925	.0449789	-1.93	0.054	-.1748496 .0014646
/cut1	-.1853053	.3571432			-.8852932 .5146826
/cut2	1.185726	.3882099			.4248488 1.946604
/cut3	1.908412	.4164896			1.092108 2.724717

```
. predict spBA_B_C spBAA spAA_A spAAA  
(option pr assumed; predicted probabilities)
```

```
. summarize spAAA,mean
```

```
. list sp* rating83c if spAAA==r(max)
```

	spBA_B_C	spBAA	spAA_A	spAAA	rati-83c
31.	.0388714	.0985567	.1096733	.7528986	AAA

```
. summarize spBA_B_C, mean
```

```
. list sp* rating83c if spBA_B_C==r(max)
```

	spBA_B_C	spBAA	spAA_A	spAAA	rati-83c
67.	.7158453	.1926148	.0449056	.0466343	AAA

```
. regress whrs k16 k618 wa we if whrs>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
Total	102120099	149	685369.794	Root MSE =	808.55

whrs	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
k16	-421.4822	167.9734	-2.51	0.013	-753.4748 -89.48953
k618	-104.4571	54.18616	-1.93	0.056	-211.5538 2.639668

wa	-4.784917	9.690502	-0.49	0.622	-23.9378	14.36797
we	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 whrs kl6 k618 wa we, ll(0) nolog
(note: 100 obs. truncated)
```

Truncated regression

```
Limit: lower = 0          Number of obs = 150
       upper = +inf       Wald chi2(4) = 10.05
Log likelihood = -1200.9157 Prob > chi2 = 0.0395
```

whrs	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
kl6	-803.0042	321.3614	-2.50	0.012	-1432.861	-173.1474
k618	-172.875	88.72898	-1.95	0.051	-346.7806	1.030578
wa	-8.821122	14.36848	-0.61	0.539	-36.98283	19.34059
we	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

```
. //
. //          Censored regression (tobit)
. //
. // womenwk from IMEUS
. use womenwk,clear
```

```
. regress lwf age married children education
```

Source	SS	df	MS	Number of obs = 2000		
Model	937.873188	4	234.468297	F( 4, 1995) =	134.21	
Residual	3485.34135	1995	1.74703827	Prob > F =	0.0000	
Total	4423.21454	1999	2.21271363	R-squared =	0.2120	
				Adj R-squared =	0.2105	
				Root MSE =	1.3218	

lwf	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	.0363624	.003862	9.42	0.000	.0287885	.0439362
married	.3188214	.0690834	4.62	0.000	.1833381	.4543046
children	.3305009	.0213143	15.51	0.000	.2887004	.3723015
education	.0843345	.0102295	8.24	0.000	.0642729	.1043961
_cons	-1.077738	.1703218	-6.33	0.000	-1.411765	-.7437105

```
. tobit lwf age married children education, ll(0)
```

Tobit regression

```
Number of obs = 2000
LR chi2(4) = 461.85
Prob > chi2 = 0.0000
Pseudo R2 = 0.0645
Log likelihood = -3349.9685
```

lwf	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	.052157	.0057457	9.08	0.000	.0408888	.0634252
married	.4841801	.1035188	4.68	0.000	.2811639	.6871964

children	.4860021	.0317054	15.33	0.000	.4238229	.5481812
education	.1149492	.0150913	7.62	0.000	.0853529	.1445454
_cons	-2.807696	.2632565	-10.67	0.000	-3.323982	-2.291409
<hr/>						
/sigma	1.872811	.040014			1.794337	1.951285

Obs. summary:            657 left-censored observations at lwf<=0  
                           1343 uncensored observations  
                              0 right-censored observations

. margins, predict(pr(0,)) dydx(\_all)

Average marginal effects    Number of obs =        2000

Model VCE    : OIM

Expression    : Pr(lwf>0), predict(pr(0,))

dy/dx w.r.t. : age married children education

	Delta-method				[95% Conf. Interval]	
	dy/dx	Std. Err.	z	P> z		
age	.0071483	.0007873	9.08	0.000	.0056052	.0086914
married	.0663585	.0142009	4.67	0.000	.0385254	.0941917
children	.0666082	.0044677	14.91	0.000	.0578516	.0753649
education	.0157542	.0020695	7.61	0.000	.0116981	.0198103

. //

. //

Regression with selection (heckman)

. //

. // note sign switch on children from tobit

. // twostep: note children now insignif in wage eqn

. heckman lw education age children, ///

> select(age married children education) nolog

Heckman selection model

(regression model with sample selection)

Number of obs =        2000

Censored obs =        657

Uncensored obs =      1343

Wald chi2(3) =        454.78

Prob > chi2 =        0.0000

Log likelihood = -1052.857

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
<hr/>						
lw						
education	.0397189	.0024525	16.20	0.000	.0349121	.0445256
age	.0075872	.0009748	7.78	0.000	.0056767	.0094977
children	-.0180477	.0064544	-2.80	0.005	-.0306981	-.0053973
_cons	2.305499	.0653024	35.30	0.000	2.177509	2.43349
<hr/>						
select						
age	.0350233	.0042344	8.27	0.000	.0267241	.0433225
married	.4547724	.0735876	6.18	0.000	.3105434	.5990014
children	.4538372	.0288398	15.74	0.000	.3973122	.5103621
education	.0565136	.0110025	5.14	0.000	.0349492	.0780781
_cons	-2.478055	.1927823	-12.85	0.000	-2.855901	-2.100208
<hr/>						
/athrho	.3377674	.1152251	2.93	0.003	.1119304	.5636045
/lnsigma	-1.375543	.0246873	-55.72	0.000	-1.423929	-1.327156

rho	.3254828	.1030183	.1114653	.5106469
sigma	.2527024	.0062385	.2407662	.2652304
lambda	.0822503	.0273475	.0286501	.1358505

LR test of indep. eqns. (rho = 0): chi2(1) = 5.53 Prob > chi2 = 0.0187

```
. heckman lw education age children, ///
> select(age married children education) twostep
Heckman selection model -- two-step estimates   Number of obs   =   2000
(regression model with sample selection)       Censored obs    =    657
                                                Uncensored obs  =   1343
                                                Wald chi2(3)    =   405.68
                                                Prob > chi2     =    0.0000
```

lw	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
lw					
education	.0427067	.003106	13.75	0.000	.0366191 .0487944
age	.009322	.0014343	6.50	0.000	.0065108 .0121333
children	-.0019549	.0115202	-0.17	0.865	-.0245341 .0206242
_cons	2.124787	.1249789	17.00	0.000	1.879833 2.369741
select					
age	.0347211	.0042293	8.21	0.000	.0264318 .0430105
married	.4308575	.074208	5.81	0.000	.2854125 .5763025
children	.4473249	.0287417	15.56	0.000	.3909922 .5036576
education	.0583645	.0109742	5.32	0.000	.0368555 .0798735
_cons	-2.467365	.1925635	-12.81	0.000	-2.844782 -2.089948
mills					
lambda	.1822815	.0638285	2.86	0.004	.05718 .307383
rho	0.66698				
sigma	.27329216				
lambda	.18228151	.0638285			

```
. //
. //                               Binomial probit with selection (heckprob)
. //
. // hmda from IMEUS
. summarize approve fanfred loanamt vacancy med_income aprr_value ///
> black appl_income debt_inc_r, sep(0)
```

Variable	Obs	Mean	Std. Dev.	Min	Max
approve	2380	.8802521	.3247347	0	1
fanfred	2095	.3331742	.4714608	0	1
loanamt	2380	139.1353	83.42097	2	980
vacancy	2380	.4365546	.4960626	0	1
med_income	2380	.8294118	.3762278	0	1
aprr_value	2380	198.5426	152.9863	25	4316
black	2380	.142437	.3495712	0	1
appl_income	2380	13.9406	116.9485	0	999.9994
debt_inc_r	2380	33.08136	10.72573	0	300



```

. heckprob fanfred loanamt vacancy med_income appr_value, ///
> sel(approve= black appl_income debt_inc_r) nolog
Probit model with sample selection      Number of obs   =    2380
                                         Censored obs   =     285
                                         Uncensored obs =    2095
                                         Wald chi2(4)   =     80.69
Log likelihood = -2063.066              Prob > chi2     =     0.0000

```

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
<b>fanfred</b>						
loanamt	-.0026434	.0008029	-3.29	0.001	-.0042169	-.0010698
vacancy	-.2163306	.0609798	-3.55	0.000	-.3358488	-.0968124
med_income	.2671338	.0893349	2.99	0.003	.0920407	.4422269
appr_value	-.0014358	.0005099	-2.82	0.005	-.0024351	-.0004364
_cons	.1684829	.1182054	1.43	0.154	-.0631954	.4001612
<b>approve</b>						
black	-.7343534	.081858	-8.97	0.000	-.8947921	-.5739147
appl_income	-.0006596	.000236	-2.80	0.005	-.0011221	-.0001971
debt_inc_r	-.0262367	.0036441	-7.20	0.000	-.033379	-.0190944
_cons	2.236424	.1319309	16.95	0.000	1.977844	2.495004
/athrho	-.6006626	.271254	-2.21	0.027	-1.132311	-.0690146
rho	-.5375209	.1928809			-.8118086	-.0689052

```

LR test of indep. eqns. (rho = 0):   chi2(1) =    4.99   Prob > chi2 = 0.0255

```