```
* Probit and Logit Models in Stata
* Copyright 2013 by Ani Katchova
clear all
set more off
use C:\Econometrics\Data\probit_insurance
global ylist ins
global xlist retire age hstatusg hhincome educyear married hisp
describe $ylist $xlist
summarize $ylist $xlist
tabulate $ylist
* Regression
reg $ylist $xlist
* Probit model
probit $ylist $xlist
* Logit model
logit $ylist $xlist
* Marginal effects (at the mean and average marginal effect)
quietly reg $ylist $xlist
margins, dydx(*) atmeans
margins, dydx(*)
quietly logit $ylist $xlist
margins, dydx(*) atmeans
margins, dydx(*)
quietly probit $ylist $xlist
margins, dydx(*) atmeans
margins, dydx(*)
*Logistic model gives odds ratio
logistic $ylist $xlist
* Predicted probabilities
quietly logit $ylist $xlist
predict plogit, pr
quietly probit $ylist $xlist
predict pprobit, pr
quietly regress $ylist $xlist
predict pols, xb
summarize $ylist plogit pprobit pols
```

* Percent correctly predicted values quietly logit \$ylist \$xlist estat classification

quietly probit \$ylist \$xlist
estat classification

- . * Probit and Logit Models in Stata
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. clear all

. set more off

.
. use C:\Econometrics\Data\probit_insurance

. global ylist ins

. global xlist retire age hstatusg hhincome educyear married hisp

. describe \$ylist \$xlist

. summarize \$ylist \$xlist

Variable	Obs	Mean	Std. Dev.	Min	Max
	+				
ins	3206	.3870867	.4871597	0	1
retire	3206	.6247661	.4842588	0	1
age	3206	66.91391	3.675794	52	86
hstatusg	3206	.7046163	.4562862	0	1
hhincome	3206	45.26391	64.33936	0	1312.124
	+				
educyear	3206	11.89863	3.304611	0	17
married	3206	.7330006	.442461	0	1
hisp	3206	.0726762	.2596448	0	1

. tabulate \$ylist

ins	Freq.	Percent	Cum.
0 1	1,965 1,241	61.29 38.71	61.29
Total	3,206	100.00	

. * Regression

. reg \$ylist \$xlist

	Source	SS	df		MS		Number of obs	=	3206
-	+						F(7, 3198)	=	41.14
	Model	62.8403396	7	8.977	719137		Prob > F	=	0.0000
	Residual	697.78505	3198	.21	181942		R-squared	=	0.0826
-	+						Adj R-squared	=	0.0806
	Total	760.62539	3205	.2373	324615		Root MSE	=	.46711
_	ins	Coef.	Std.	Err.	t	P> t	[95% Conf.	In	terval]
_	ins	Coef.	Std.	Err.	t 	P> t	[95% Conf.	In	terval]
	ins 	Coef.	Std. 		t 2.24	P> t 0.025	[95% Conf. 		terval]
				197					
	retire	.0408508	.0182	197 189	2.24	0.025	.0051273		0765743
	retire age	.0408508	.0182	197 189 531	2.24	0.025	.0051273		0765743 0018473

. * Probit model

. probit \$ylist \$xlist

Iteration 0: log likelihood = -2139.7712 Iteration 1: log likelihood = -1994.4552 Iteration 2: log likelihood = -1993.624 Iteration 3: log likelihood = -1993.6237 Iteration 4: log likelihood = -1993.6237

Number of obs = 3206 LR chi2(7) = 292.30 Prob > chi2 = 0.0000 Pseudo R2 = 0.0683 Probit regression

Log likelihood = -1993.6237

ins Coef. Std. Err. z P> z [95% Conf. Inter	val]
retire .1183567 .0512678 2.31 0.021 .0178736 .218	8397
age 0088696 .006899 -1.29 0.1990223914 .004	6521
hstatusg .1977357 .0554868 3.56 0.000 .0889835 .306	4878
hhincome .001233 .0003866 3.19 0.001 .0004754 .001	9907
educyear .0707477 .0084782 8.34 0.000 .0541308 .087	3647
married .362329 .0560031 6.47 0.000 .252565 .472	0931
hisp 4731099 .1104393 -4.28 0.000689567256	6529
_cons -1.069319 .4580794 -2.33 0.020 -1.967139171	5002

. * Logit model

. logit \$ylist \$xlist

Iteration 0: log likelihood = -2139.7712
Iteration 1: log likelihood = -1996.7434
Iteration 2: log likelihood = -1994.8864
Iteration 3: log likelihood = -1994.8784
Iteration 4: log likelihood = -1994.8784

Logistic regression Number of obs = 3206

LR chi2(7) = 289.79 Prob > chi2 = 0.0000 Pseudo R2 = 0.0677

Log likelihood = -1994.8784

ins	Coef.	Std. Err.	Z	P> z	[95% Conf.	Interval]
retire	.1969297	.0842067	2.34	0.019	.0318875	.3619718
age	0145955	.0112871	-1.29	0.196	0367178	.0075267
hstatusg	.3122654	.0916739	3.41	0.001	.1325878	.491943
hhincome	.0023036	.000762	3.02	0.003	.00081	.0037972
educyear	.1142626	.0142012	8.05	0.000	.0864288	.1420963
married	.578636	.0933198	6.20	0.000	.3957327	.7615394
hisp	8103059	.1957522	-4.14	0.000	-1.193973	4266387
_cons	-1.715578	.7486219	-2.29	0.022	-3.18285	2483064

.

. quietly reg \$ylist \$xlist

. margins, dydx(*) atmeans

Conditional marginal effects Number of obs = 3206

Model VCE : OLS

Expression : Linear prediction, predict()

 $\mbox{d} y / \mbox{d} x$ w.r.t. : retire age hstatusg hhincome educyear married hisp

at : retire = .6247661 (mean)
 age = 66.91391 (mean)
 hstatusg = .7046163 (mean)
 hhincome = 45.26391 (mean)
 educyear = 11.89863 (mean)
 married = .7330006 (mean)
 hisp = .0726762 (mean)

		Delta-method				
	dy/dx	Std. Err.	Z	P> z	[95% Conf.	Interval]
	+					
retire	.0408508	.0182197	2.24	0.025	.0051408	.0765608
age	0028955	.0024189	-1.20	0.231	0076366	.0018455
hstatusg	.0655583	.0194531	3.37	0.001	.027431	.1036856
hhincome	.0004921	.0001375	3.58	0.000	.0002226	.0007616
educyear	.0233686	.0028672	8.15	0.000	.0177491	.0289882
married	.1234699	.0193618	6.38	0.000	.0855215	.1614183

^{. *} Marginal effects (at the mean and average marginal effect)

hisp | -.1210059 .033666 -3.59 0.000 -.18699 -.0550218

. margins, dydx(*)

Number of obs = 3206 Average marginal effects

Model VCE : OLS

Expression : Linear prediction, predict()

dy/dx w.r.t. : retire age hstatusg hhincome educyear married hisp

	 dy/dx +	Delta-method	l z	P> z	[95% Conf.	Interval]
retire	.0408508	.0182197	2.24	0.025	.0051408	.0765608
age	0028955	.0024189	-1.20	0.231	0076366	.0018455
hstatusg	.0655583	.0194531	3.37	0.001	.027431	.1036856
hhincome	.0004921	.0001375	3.58	0.000	.0002226	.0007616
educyear	.0233686	.0028672	8.15	0.000	.0177491	.0289882
married	.1234699	.0193618	6.38	0.000	.0855215	.1614183
hisp	1210059	.033666	-3.59	0.000	18699	0550218

. quietly logit \$ylist \$xlist

. margins, dydx(*) atmeans

Conditional marginal effects Number of obs = 3206

Model VCE : OIM

Expression : Pr(ins), predict()

dy/dx w.r.t. : retire age hstatusg hhincome educyear married hisp

: retire = .6247661 (mean) age = 66.91391 (mean) hstatusq = .7046163 (mean) = 45.26391 (mean) hhincome = 11.89863 (mean) educyear married

= .7330006 (mean) hisp = .0726762 (mean)

	1	Delta-method				
[dy/dx	Std. Err.	Z	P> z	[95% Conf.	<pre>Interval]</pre>
+						
retire	.0460479	.0196856	2.34	0.019	.0074648	.084631
age	0034129	.0026389	-1.29	0.196	008585	.0017592
hstatusg	.0730168	.021412	3.41	0.001	.0310499	.1149836
hhincome	.0005386	.0001785	3.02	0.003	.0001888	.0008885
educyear	.0267179	.0033025	8.09	0.000	.0202452	.0331907
married	.135302	.0217469	6.22	0.000	.0926789	.1779251
hisp	1894732	.045563	-4.16	0.000	2787749	1001714

. margins, dydx(*)

Average marginal effects Number of obs = 3206

Model VCE : OIM

Expression : Pr(ins), predict()

dy/dx w.r.t. : retire age hstatusg hhincome educyear married hisp

	 dy/dx	Delta-method Std. Err.	d z	P> z	[95% Conf	. Interval]
	+					
retire	.0427616	.018228	2.35	0.019	.0070354	.0784878
age	0031693	.0024486	-1.29	0.196	0079686	.00163
hstatusg	.0678058	.0197778	3.43	0.001	.0290419	.1065696
hhincome	.0005002	.0001646	3.04	0.002	.0001777	.0008228
educyear	.0248111	.0029705	8.35	0.000	.0189891	.0306332
married	.1256459	.0198205	6.34	0.000	.0867985	.1644933
hisp	175951	.0421962	-4.17	0.000	258654	0932481

. quietly probit \$ylist \$xlist

. margins, dydx(*) atmeans

Conditional marginal effects Number of obs = 3206

Model VCE : OIM

Expression : Pr(ins), predict()

dy/dx w.r.t. : retire age hstatusg hhincome educyear married hisp

: retire = .6247661 (mean) = 66.91391 (mean) age = hstatusg .7046163 (mean) = 45.26391 (mean) hhincome educyear = 11.89863 (mean)

> married = .7330006 (mean) = .0726762 (mean) hisp

	[Delta-metho	d			
	dy/dx	Std. Err.	Z	P> z	[95% Conf	. Interval]
	+					
retire	.0448503	.0194248	2.31	0.021	.0067783	.0829222
age	0033611	.0026141	-1.29	0.199	0084846	.0017624
hstatusg	.0749303	.0210134	3.57	0.000	.0337448	.1161158
hhincome	.0004672	.0001466	3.19	0.001	.00018	.0007545
educyear	.0268093	.0032034	8.37	0.000	.0205306	.0330879
married	.1373016	.0211866	6.48	0.000	.0957766	.1788267
hisp	1792811	.0417552	-4.29	0.000	2611197	0974425

. margins, dydx(*)

Model VCE : OIM

Expression : Pr(ins), predict()

 $\mathrm{d}y/\mathrm{d}x$ w.r.t. : retire age hstatusg hhincome educyear married hisp

		Delta-method	i			
	dy/dx	Std. Err.	Z	P> z	[95% Conf	. Interval]
	+					
retire	.0419784	.0181404	2.31	0.021	.0064239	.0775329
age	0031459	.0024451	-1.29	0.198	0079382	.0016465
hstatusg	.0701324	.0195584	3.59	0.000	.0317987	.108466
hhincome	.0004373	.0001366	3.20	0.001	.0001696	.000705
educyear	.0250926	.0029084	8.63	0.000	.0193923	.0307929
married	.1285099	.0194623	6.60	0.000	.0903646	.1666552
hisp	1678014	.0388964	-4.31	0.000	244037	0915657

•

. logistic \$ylist \$xlist

Logistic regression Number of obs = 3206 LR chi2(7) = 289.79 Prob > chi2 = 0.0000 Log likelihood = -1994.8784 Pseudo R2 = 0.0677

ins	Odds Ratio	Std. Err.	Z	P> z	[95% Conf.	Interval]
retire	1.217658	.102535	2.34	0.019	1.032401	1.436159
age	.9855105	.0111235	-1.29	0.196	.9639481	1.007555
hstatusg	1.366517	.125274	3.41	0.001	1.141779	1.635491
hhincome	1.002306	.0007638	3.02	0.003	1.00081	1.003804
educyear	1.121046	.0159201	8.05	0.000	1.090274	1.152688
married	1.783604	.1664455	6.20	0.000	1.485472	2.14157
hisp	.444722	.0870553	-4.14	0.000	.3030149	.6526993
_cons	.1798597	.1346469	-2.29	0.022	.0414673	.7801209

•

^{. *}Logistic model gives odds ratio

^{. *} Predicted probabilities

[.] quietly logit \$ylist \$xlist

[.] predict plogit, pr

[.] quietly probit \$ylist \$xlist

[.] predict pprobit, pr

[.]

[.] quietly regress \$ylist \$xlist

. predict pols, xb

.

. summarize \$ylist plogit pprobit pols

Variable	Obs	Mean	Std. Dev.	Min	Max
ins	3206	.3870867	.4871597	0	1
plogit	3206	.3870867	.1418287	.0340215	.9649615
pprobit	3206	.3861139	.1421416	.0206445	.9647618
pols	3206	.3870867	.1400249	1557328	1.197223

•

- . * Percent correctly predicted values
- . quietly logit \$ylist \$xlist
- . estat classification

Logistic model for ins

		True		
Total	~D)	D .	Classified
653 2553	308 1657		345 896	+ -
3206	1965		1241	Total

Classified + if predicted Pr(D) >= .5True D defined as ins != 0

Sensitivity	Pr(+ D)	27.80%	
Specificity	Pr(- ~D)	84.33%	
Positive predictive value	Pr(D +)	52.83%	
Negative predictive value	Pr(~D -)	64.90%	
False + rate for true ~D	Pr(+ ~D)	15.67%	
False - rate for true D	Pr(- D)	72.20%	
False + rate for classified +	Pr(~D +)	47.17%	
False - rate for classified -	Pr(D -)	35.10%	
Correctly classified			

.

- . quietly probit \$ylist \$xlist
- . estat classification

Probit model for ins

------ True ------ Classified | D ~D | Total

		+			
+	335	305	640		
	906	1660			
	1241	1965			
Classified + if predicted $Pr(D) >= .5$ True D defined as ins != 0					
Sensitivity		Pr(+ D)	26.99%		
Specificity	Pr(- ~D)	84.48%			
Positive predi	Pr(D +)	52.34%			
Negative predi	ctive value	Pr(~D -)	64.69%		
False + rate f	or true ~D	Pr(+ ~D)	15.52%		
	or true D	:			
False + rate f	or classified +	Pr(~D +)	47.66%		
False - rate f	or classified -	Pr(D -)	35.31%		
Correctly clas	 sified 		62.23%		