nl postestimation — Postestimation tools for nl

Postestimation commands predict margins Remarks and examples Also see

Postestimation commands

Command	Description
estat ic	Akaike's, consistent Akaike's, corrected Akaike's, and Schwarz's Bayesian infor- mation criteria (AIC, CAIC, AICc, and BIC, respectively)
estat summarize	summary statistics for the estimation sample
estat vce	variance-covariance matrix of the estimators (VCE)
estat (svy)	postestimation statistics for survey data
estimates	cataloging estimation results
etable	table of estimation results
forecast	dynamic forecasts and simulations
'hausman	Hausman's specification test
lincom	point estimates, standard errors, testing, and inference for linear combinations of parameters
'lrtest	likelihood-ratio test
margins	marginal means, predictive margins, marginal effects, and average marginal effects
marginsplot	graph the results from margins (profile plots, interaction plots, etc.)
nlcom	point estimates, standard errors, testing, and inference for nonlinear combinations of parameters
predict	fitted values, residuals, etc.
predictnl	point estimates, standard errors, testing, and inference for generalized predictions
test	Wald tests of simple and composite linear hypotheses
testnl	Wald tests of nonlinear hypotheses

The following postestimation commands are available after nl:

*forecast, hausman, and lrtest are not appropriate with svy estimation results.

[†]You must specify the variables() option with nl.

predict

Description for predict

predict creates a new variable containing predictions such as fitted values, residuals, probabilities, and expected values. It can also create multiple new variables containing predicted, named substitutable expressions.

Menu for predict

Statistics > Postestimation

Syntax for predict

Syntax for predicting without named substitutable expressions (parameters)

```
predict [type] newvar [if] [in] [, statistic]
```

```
predict [type] stub* [if ] [in], scores
```

Syntax for predicting with named substitutable expressions (parameters)

```
Predict all parameters
```

```
predict [type] { stub* | newvarlist } [if ] [in ], parameters
```

Predict specific parameters

```
predict [type] (newvar = \{param:\}) [(newvar = \{param:\}) [...]] [if] [in]
```

```
predict [type] { stub* | newvarlist } [if ] [in], parameters(paramnames)
```

statistic	Description			
Main				
yhat	fitted values; the default			
	residuals			
pr(<i>a</i> , <i>b</i>)	$\Pr(y_j \mid a < y_j < b)$			
e(<i>a</i> , <i>b</i>)	$E(y_j \mid a < y_j < b)$			
$\underline{ys}tar(a,b)$	$E(y_j^*), y_j^* = \max\{a, \min(y_j, b)\}$			

These statistics are available both in and out of sample; type predict ... if e(sample) ... if wanted only for the estimation sample.

Options for predict

Main

yhat, the default, calculates the fitted values.

residuals calculates the residuals.

pr(*a*, *b*) calculates $Pr(a < f(\mathbf{x}_j, \mathbf{b}) + u_j < b)$, the probability that $y_j | \mathbf{x}_j$ would be observed in the interval (a, b).

a and b may be specified as numbers or variable names; lb and ub are variable names; pr (20,30) calculates $Pr(20 < f(\mathbf{x}_i, \mathbf{b}) + u_i < 30)$;

pr(*lb*, *ub*) calculates $Pr(lb < f(\mathbf{x}_j, \mathbf{b}) + u_j < ub)$; and pr(20, *ub*) calculates $Pr(20 < f(\mathbf{x}_j, \mathbf{b}) + u_j < ub)$.

a missing $(a \ge .)$ means $-\infty$; pr(., 30) calculates $Pr(-\infty < f(\mathbf{x}_j, \mathbf{b}) + u_j < 30)$; pr(*lb*, 30) calculates $Pr(-\infty < f(\mathbf{x}_j, \mathbf{b}) + u_j < 30)$ in observations for which $lb \ge .$ and calculates $Pr(lb < f(\mathbf{x}_j, \mathbf{b}) + u_j < 30)$ elsewhere.

b missing $(b \ge .)$ means $+\infty$; pr(20,.) calculates Pr($+\infty > f(\mathbf{x}_j, \mathbf{b}) + u_j > 20$); pr(20,*ub*) calculates Pr($+\infty > f(\mathbf{x}_j, \mathbf{b}) + u_j > 20$) in observations for which $ub \ge .$ and calculates Pr($20 < f(\mathbf{x}_j, \mathbf{b}) + u_j < ub$) elsewhere.

- e(a, b) calculates $E(f(\mathbf{x}_j, \mathbf{b}) + u_j | a < f(\mathbf{x}_j, \mathbf{b}) + u_j < b)$, the expected value of $y_j | \mathbf{x}_j$ conditional on $y_j | \mathbf{x}_j$ being in the interval (a, b), meaning that $y_j | \mathbf{x}_j$ is truncated. a and b are specified as they are for pr().
- ystar(*a*, *b*) calculates $E(y_j^*)$, where $y_j^* = a$ if $f(\mathbf{x}_j, \mathbf{b}) + u_j \le a$, $y_j^* = b$ if $f(\mathbf{x}_j, \mathbf{b}) + u_j \ge b$, and $y_j^* = f(\mathbf{x}_j, \mathbf{b}) + u_j$ otherwise, meaning that y_j^* is censored. *a* and *b* are specified as they are for pr().
- scores calculates the scores. The *j*th new variable created will contain the score for the *j*th parameter in e(b).
- parameters and parameters (*paramnames*) calculate predictions for all or a subset of the named substitutable expressions in the model. parameters () does not appear in the dialog box.
 - *paramnames* is *param* [*param* [...]], and *param* is a name of a substitutable expression as specified in one of nl's define() options.

margins

Description for margins

margins estimates margins of response for fitted values.

Menu for margins

Statistics > Postestimation

Syntax for margins

<pre>margins [marginlist][, options]</pre>		
<pre>margins [marginlist], predict(statistic</pre>) [options]

statistic	Description				
yhat	fitted values; the default				
$\operatorname{pr}(a,b)$	not allowed with margins				
e(<i>a</i> , <i>b</i>)	not allowed with margins				
ystar(a,b)	not allowed with margins				
<u>r</u> esiduals	not allowed with margins				
parameters	predicted parameters				
<pre>parameters(param)</pre>	predicted, named substitutable expression param				

Statistics not allowed with margins are functions of stochastic quantities other than e(b).

For the full syntax, see [R] margins.

Remarks and examples

> Example 1: Basic usage of predict and margins

Obtaining predictions after fitting a nonlinear regression model with nl is no more difficult than obtaining predictions after fitting a linear regression model with regress. Here we fit a model of mpg on weight, allowing for a nonlinear relationship:

. use https:// (1978 automobi	'www.stata-pre lle data)	ss.com/dat	a/r19/auto				
. nl (mpg = {k	00} + {b1}*wei	ght^{gamma	=5}), va	riables((weight)	nolog	
Source	SS	df	MS				
				– Num	ber of o	obs =	74
Model	1646.4376	2	823.21880	5 R-s	squared	=	0.6738
Residual	797.02185	71	11.225659	9 Adj	j R-squar	ed =	0.6646
				- Roc	ot MSE	=	3.350472
Total	2443.4595	73	33.472047	4 Res	s. dev.	=	385.8874
mpg	Coefficient	Std. err.	t	P> t	[95%	conf.	interval]
/ъ0	-18.17687	60.62086	-0.30	0.765	-139.0)514	102.6977
/b1	1377.163	5291,956	0.26	0.795	-9174.	697	11929.02
/gamma	4460793	.6763682	-0.66	0.512	-1.794	1719	.9025606

Note: Parameter **b0** is used as a constant term during estimation.

Now, we obtain the predicted values of mpg and plot them in a graph along with the observed values:

```
. predict mpghat
(option yhat assumed; fitted values)
```

. scatter mpg weight || line mpghat weight, sort



Suppose we wanted to know how sensitive mpg is to changes in weight for cars that weigh 3,000 pounds. We can use margins to find out:

```
. margins, eyex(weight) at(weight = 3000)
Conditional marginal effects
                                                              Number of obs = 74
Model VCE: GNR
Expression: Fitted values, predict()
ey/ex wrt: weight
At: weight = 3000
                           Delta-method
                                                  P>|z|
                                                            [95% conf. interval]
                    ey/ex
                             std. err.
                                            z
      weight
                -.8408143
                             .0804327
                                        -10.45
                                                  0.000
                                                           -.9984596
                                                                         -.683169
```

With the eyex() option, margins reports elasticities. These results show that if we increase weight by 1%, then mpg decreases by about 0.84%.

Technical note

Observant readers will notice that margins issued a warning message stating that it could not perform its usual check for estimable functions. In the case of nl, as long as you do not specify the predict() option of margins or specify the default predict(yhat), you can safely ignore that message. The predicted values that nl produces are suitable for use with margins. However, if you specify any predict() options other than yhat, then the output from margins after using nl will not be correct.

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Example 2: Predictions from named expressions

Continuing with example 1, consider now a logistic regression that explains the origin of the cars as a function of their price, their fuel efficiency, and their length.

. logit foreig	gn price mpg l	ength				
Iteration 0: Iteration 1: Iteration 2: Iteration 3: Iteration 4: Iteration 5: Iteration 6:	Log likelihoo Log likelihoo Log likelihoo Log likelihoo Log likelihoo Log likelihoo Log likelihoo	d = -45.03 $d = -27.130$ $d = -23.885$ $d = -23.460$ $d = -23.460$ $d = -23.460$ $d = -23.460$	321 212 718 167 272 253 253			
Logistic regre	ession 1 = -23.460253	i			Number of ob LR chi2(3) Prob > chi2 Pseudo R2	s = 74 = 43.15 = 0.0000 = 0.4790
foreign	Coefficient	Std. err.	Z	P> z	[95% conf.	interval]
price mpg length _cons	.0005602 0606663 1718435 27.7666	.0001889 .0854276 .0538215 9.860407	2.97 -0.71 -3.19 2.82	0.003 0.478 0.001 0.005	.0001899 2281012 2773316 8.440557	.0009306 .1067687 0663553 47.09264

After fitting this model, the predict command computes predicted probabilities by default or the linear prediction if we use the xb option:

```
. predict p_logit
(option pr assumed; Pr(foreign))
. predict xb_logit, xb
```

Next recall that logistic regression fits the logistic cumulative density function to our data via maximum likelihood. We could instead fit this (nonlinear) function via nonlinear least squares using the nl command:

. nl (foreign	$= 1/(1 + \exp(1))$	(-{lp: price	mpg lengt	h})))			
Iteration 0:	Residual SS =	■ 18.5					
Iteration 1:	Residual SS =	7.7385558					
Iteration 2:	Residual SS =	7.6199595					
Iteration 3:	Residual SS =	7.6149448					
Iteration 4:	Residual SS =	7.6143961					
Iteration 5:	Residual SS =	7.6143036					
Iteration 6:	Residual SS =	7.6142876					
Iteration 7:	Residual SS =	7.6142848					
Iteration 8:	Residual SS =	7.6142843					
Source	SS	df	MS				
				· Numb	er of ob	s =	74
Model	7.8451752	3	2.61505839	R-sq	uared	=	0.5075
Residual	7.6142843	70	.10877549	Adj	R-square	d =	0.4864
				Root	MSE	=	.3298113
Total	15.459459	73	.211773417	Res.	dev.	=	41.72401
C	Q., ((), (), (), (), (), (), (), (), (), (G • 1		Do 1 to 1	F05%		
foreign	Coefficient	Std. err.	t	P> t	[95% c	onf.	interval]
price	.0006768	.0002826	2.39	0.019	.00011	31	.0012405
mpg	0703169	.0704894	-1.00	0.322	21090	36	.0702698
length	2241772	.0912304	-2.46	0.016	40613	05	042224
cons	36,6067	15.28359	2.40	0.019	6.1245	31	67.08886
_00110							

Note: Parameter _cons is used as a constant term during estimation.

Computing predicted probabilities from this model is still the default for predict because these are simply the values our model predicts for the dependent variable:

```
. predict p_nl
(option yhat assumed; fitted values)
```

To obtain the linear prediction, we need the predicted values of the named linear combination {lp:} we created. We can compute those using the parameters() option:

. predict xb_nl, parameters(lp)

Also see

[R] **nl** — Nonlinear least-squares estimation

[U] 20 Estimation and postestimation commands

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