

nl postestimation — Postestimation tools for nl

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Postestimation commands

The following postestimation commands are available after `nl`:

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

* `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.

Menu for predict

Statistics > Postestimation

Syntax for predict

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

```
predict [type] { stub* | newvar1 ... newvark } [if] [in] , scores
```

where k is the number of parameters in the model.

<i>statistic</i>	Description
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Main

<code>yhat</code>	fitted values; the default
<code>residuals</code>	residuals
<code>pr(a,b)</code>	$\Pr(y_j \mid a < y_j < b)$
<code>e(a,b)</code>	$E(y_j \mid a < y_j < b)$
<code>ystar(a,b)</code>	$E(y_j^*), y_j^* = \max\{a, \min(y_j, b)\}$

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 $\Pr(a < \mathbf{x}_j \mathbf{b} + u_j < b)$, 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 $\Pr(20 < \mathbf{x}_j \mathbf{b} + u_j < 30)$;

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

`pr(20,ub)` calculates $\Pr(20 < \mathbf{x}_j \mathbf{b} + u_j < ub)$.

a missing ($a \geq .$) means $-\infty$; `pr(.,30)` calculates $\Pr(-\infty < \mathbf{x}_j \mathbf{b} + u_j < 30)$;

`pr(lb,30)` calculates $\Pr(-\infty < \mathbf{x}_j \mathbf{b} + u_j < 30)$ in observations for which $lb \geq .$ and calculates $\Pr(lb < \mathbf{x}_j \mathbf{b} + u_j < 30)$ elsewhere.

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

`e(a, b)` calculates $E(\mathbf{x}_j \mathbf{b} + u_j \mid a < \mathbf{x}_j \mathbf{b} + u_j < b)$, the expected value of $y_j \mid \mathbf{x}_j$ conditional on $y_j \mid \mathbf{x}_j$ being in the interval (a, b) , meaning that $y_j \mid \mathbf{x}_j$ is truncated. a and b are specified as they are for `pr()`.

`ystar(a, b)` calculates $E(y_j^*)$, where $y_j^* = a$ if $\mathbf{x}_j \mathbf{b} + u_j \leq a$, $y_j^* = b$ if $\mathbf{x}_j \mathbf{b} + u_j \geq b$, and $y_j^* = \mathbf{x}_j \mathbf{b} + u_j$ 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)`.

margins

Description for margins

`margins` estimates margins of response for fitted values.

Menu for margins

Statistics > Postestimation

Syntax for margins

```
margins [marginlist] [, options]
margins [marginlist] , predict(statistic ...) [options]
```

<i>statistic</i>	Description
<code>yhat</code>	fitted values; the default
<code>pr(a, b)</code>	not allowed with <code>margins</code>
<code>e(a, b)</code>	not allowed with <code>margins</code>
<code>ystar(a, b)</code>	not allowed with <code>margins</code>
<code>residuals</code>	not allowed with <code>margins</code>

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

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 http://www.stata-press.com/data/r15/auto
(1978 Automobile Data)
. nl (mpg = {b0} + {b1}*weight^{gamma=-.5}), variables(weight) nolog
(obs = 74)
```

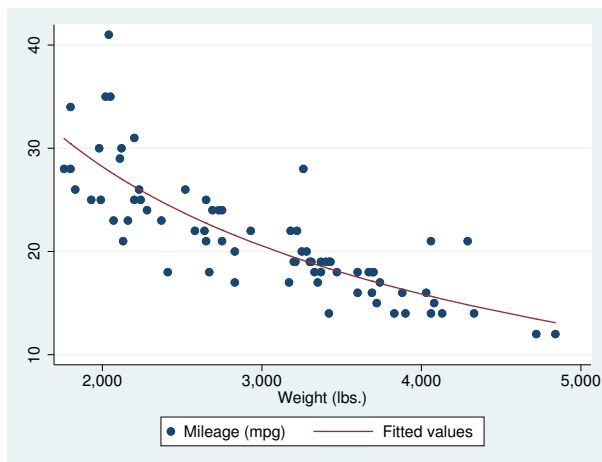
Source	SS	df	MS			
Model	1646.4376	2	823.218806	Number of obs =	74	
Residual	797.02185	71	11.2256598	R-squared =	0.6738	
				Adj R-squared =	0.6646	
				Root MSE =	3.350472	
Total	2443.4595	73	33.4720474	Res. dev. =	385.8874	

mpg	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
/b0	-18.17583	60.61762	-0.30	0.765	-139.0439	102.6923
/b1	1377.267	5292.379	0.26	0.795	-9175.436	11929.97
/gamma	-.4460916	.6763641	-0.66	0.512	-1.794723	.9025401

Parameter `b0` taken as constant term in model & ANOVA table

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 w.r.t.  : weight
at            : weight                =           3000
```

	Delta-method				[95% Conf. Interval]	
	ey/ex	Std. Err.	z	P> z		
weight	-.8408119	.0804339	-10.45	0.000	-.9984594	-.6831644

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.

□

Also see

[R] [nl](#) — Nonlinear least-squares estimation

[U] [20 Estimation and postestimation commands](#)