

probit postestimation — Postestimation tools for probit

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Postestimation commands

The following postestimation commands are of special interest after `probit`:

| Command | Description |
|-----------------------------------|---|
| <code>estat classification</code> | report various summary statistics, including the classification table |
| <code>estat gof</code> | Pearson or Hosmer–Lemeshow goodness-of-fit test |
| <code>lroc</code> | compute area under ROC curve and graph the curve |
| <code>lsens</code> | graph sensitivity and specificity versus probability cutoff |
| <code>lassogof</code> | calculate goodness-of-fit predictions |

These commands are not appropriate with `svy` estimation results.

The following standard postestimation commands are also available:

| Command | Description |
|------------------------------|---|
| <code>contrast</code> | contrasts and ANOVA-style joint tests of estimates |
| <code>estat ic</code> | Akaike's, consistent Akaike's, corrected Akaike's, and Schwarz's Bayesian information criteria (AIC, CAIC, AICc, 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>etable</code> | table of 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>linktest</code> | link test for model specification |
| * <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 margins (profile plots, interaction plots, etc.) |
| <code>nlcom</code> | point estimates, standard errors, testing, and inference for nonlinear combinations of coefficients |
| <code>predict</code> | probabilities, linear predictions and their SEs, etc. |
| <code>predictnl</code> | point estimates, standard errors, testing, and inference for generalized predictions |

| | |
|------------------------|--|
| <code>pwcompare</code> | pairwise comparisons of estimates |
| <code>suest</code> | seemingly unrelated estimation |
| <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. `forecast` is also not appropriate with `mi` estimation results.

predict

Description for predict

`predict` creates a new variable containing predictions such as probabilities, linear predictions, standard errors, deviance residuals, and the equation-level score.

Menu for predict

Statistics > Postestimation

Syntax for predict

```
predict [type] newvar [if] [in] [, statistic nooffset rules asif]
```

| <i>statistic</i> | Description |
|------------------|-------------|
|------------------|-------------|

Main

| | |
|-------------------------|--|
| <code>pr</code> | probability of a positive outcome; the default |
| <code>xb</code> | linear prediction |
| <code>stdp</code> | standard error of the linear prediction |
| * <code>deviance</code> | deviance residual |
| <code>score</code> | first derivative of the log likelihood with respect to $\mathbf{x}_j\beta$ |

Unstarred statistics are available both in and out of sample; type `predict ... if e(sample) ...` if wanted only for the estimation sample. Starred statistics are calculated only for the estimation sample, even when `if e(sample)` is not specified.

Options for predict

Main

`pr`, the default, calculates the probability of a positive outcome.

`xb` calculates the linear prediction.

`stdp` calculates the standard error of the linear prediction.

`deviance` calculates the deviance residual.

`score` calculates the equation-level score, $\partial \ln L / \partial (\mathbf{x}_j\beta)$.

`nooffset` is relevant only if you specified `offset` (*varname*) for `probit`. It modifies the calculations made by `predict` so that they ignore the offset variable; the linear prediction is treated as $\mathbf{x}_j\mathbf{b}$ rather than as $\mathbf{x}_j\mathbf{b} + \text{offset}_j$.

`rules` requests that Stata use any rules that were used to identify the model when making the prediction. By default, Stata calculates missing for excluded observations.

`asif` requests that Stata ignore the rules and exclusion criteria and calculate predictions for all observations possible using the estimated parameter from the model.

margins

Description for margins

`margins` estimates margins of response for probabilities and linear predictions.

Menu for margins

Statistics > Postestimation

Syntax for margins

```
margins [marginlist] [, options]
```

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

| <i>statistic</i> | Description |
|-----------------------|--|
| <code>pr</code> | probability of a positive outcome; the default |
| <code>xb</code> | linear prediction |
| <code>stdp</code> | not allowed with <code>margins</code> |
| <code>deviance</code> | not allowed with <code>margins</code> |
| <code>score</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

[stata.com](http://www.stata.com)

Remarks are presented under the following headings:

Obtaining predicted values
Performing hypothesis tests

Obtaining predicted values

Once you have fit a probit model, you can obtain the predicted probabilities by using the `predict` command for both the estimation sample and other samples; see [U] 20 **Estimation and postestimation commands** and [R] **predict**. Here we will make only a few additional comments.

`predict` without arguments calculates the predicted probability of a positive outcome. With the `xb` option, `predict` calculates the linear combination $\mathbf{x}_j\mathbf{b}$, where \mathbf{x}_j are the independent variables in the j th observation and \mathbf{b} is the estimated parameter vector. This is known as the index function because the cumulative density indexed at this value is the probability of a positive outcome.

In both cases, Stata remembers any rules used to identify the model and calculates missing for excluded observations unless `rules` or `asif` is specified. This is covered in the following example.

With the `stdp` option, `predict` calculates the standard error of the prediction, which is *not* adjusted for replicated covariate patterns in the data.

You can calculate the unadjusted-for-replicated-covariate-patterns diagonal elements of the hat matrix, or leverage, by typing

```
. predict pred
. predict stdp, stdp
. generate hat = stdp^2*pred*(1-pred)
```

► Example 1

In [example 4](#) of [R] **probit**, we fit the probit model `probit foreign b3.repair`. To obtain predicted probabilities, we type

```
. predict p
(option pr assumed; Pr(foreign))
(10 missing values generated)
. summarize foreign p
```

| Variable | Obs | Mean | Std. dev. | Min | Max |
|----------|-----|----------|-----------|-----|-----|
| foreign | 58 | .2068966 | .4086186 | 0 | 1 |
| p | 48 | .25 | .1956984 | .1 | .5 |

Stata remembers any rules used to identify the model and sets predictions to missing for any excluded observations. In [example 4](#) of [R] **probit**, `probit` omitted the variable `1.repair` from our model and excluded 10 observations. When we typed `predict p`, those same 10 observations were again excluded and their predictions set to missing.

`predict`'s `rules` option uses the rules in the prediction. During estimation, we were told, “`1.repair != 0` predicts failure perfectly”, so the rule is that when `1.repair` is not zero, we should predict 0 probability of success or a positive outcome:

```
. predict p2, rules
(option pr assumed; Pr(foreign))
. summarize foreign p p2
```

| Variable | Obs | Mean | Std. dev. | Min | Max |
|----------|-----|----------|-----------|-----|-----|
| foreign | 58 | .2068966 | .4086186 | 0 | 1 |
| p | 48 | .25 | .1956984 | .1 | .5 |
| p2 | 58 | .2068966 | .2016268 | 0 | .5 |

`predict`'s `asif` option ignores the rules and the exclusion criteria and calculates predictions for all observations possible using the estimated parameters from the model:

```
. predict p3, asif
(option pr assumed; Pr(foreign))
. summarize for p p2 p3
```

| Variable | Obs | Mean | Std. dev. | Min | Max |
|----------|-----|----------|-----------|-----|-----|
| foreign | 58 | .2068966 | .4086186 | 0 | 1 |
| p | 48 | .25 | .1956984 | .1 | .5 |
| p2 | 58 | .2068966 | .2016268 | 0 | .5 |
| p3 | 58 | .2931034 | .2016268 | .1 | .5 |

Which is right? By default, `predict` uses the most conservative approach. If many observations had been excluded due to a simple rule, we could be reasonably certain that the `rules` prediction is correct. The `asif` prediction is correct only if the exclusion is a fluke and we would be willing to exclude the variable from the analysis, anyway. Then, however, we should refit the model to include the excluded observations. ◀

Performing hypothesis tests

After estimation with `probit`, you can perform hypothesis tests by using the `test` or `testnl` command; see [U] 20 [Estimation and postestimation commands](#).

Methods and formulas

Let index j be used to index observations. Define M_j for each observation as the total number of observations sharing j 's covariate pattern. Define Y_j as the total number of positive responses among observations sharing j 's covariate pattern. Define p_j as the predicted probability of a positive outcome for observation j .

For $M_j > 1$, the deviance residual d_j is defined as

$$d_j = \pm \left(2 \left[Y_j \ln \left(\frac{Y_j}{M_j p_j} \right) + (M_j - Y_j) \ln \left\{ \frac{M_j - Y_j}{M_j (1 - p_j)} \right\} \right] \right)^{1/2}$$

where the sign is the same as the sign of $(Y_j - M_j p_j)$. In the limiting cases, the deviance residual is given by

$$d_j = \begin{cases} -\sqrt{2M_j |\ln(1 - p_j)|} & \text{if } Y_j = 0 \\ \sqrt{2M_j |\ln p_j|} & \text{if } Y_j = M_j \end{cases}$$

Also see

[R] [probit](#) — Probit regression

[R] [estat classification](#) — Classification statistics and table

[R] [estat gof](#) — Pearson or Hosmer–Lemeshow goodness-of-fit test

[R] [iroc](#) — Compute area under ROC curve and graph the curve

[R] [lsens](#) — Graph sensitivity and specificity versus probability cutoff

[LASSO] [lassogof](#) — Goodness of fit after lasso for prediction

[U] **20 Estimation and postestimation commands**

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