

tnbreg postestimation — Postestimation tools for tnbreg

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Description

The following postestimation commands are available after `tnbreg`:

Command	Description
<code>contrast</code>	contrasts and ANOVA-style joint tests of estimates
<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> ¹	dynamic forecasts and simulations
<code>lincom</code>	point estimates, standard errors, testing, and inference for linear combinations of coefficients
<code>lrtest</code> ²	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>	predictions, residuals, influence statistics, and other diagnostic measures
<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

¹ `forecast` is not appropriate with `svy` estimation results.

² `lrtest` is not appropriate with `svy` estimation results.

Syntax for predict

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

```
predict [type] { stub* | newvarreg newvardisp } [if] [in] , scores
```

<i>statistic</i>	Description
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Main

<code>n</code>	number of events; the default
<code>ir</code>	incidence rate
<code>cm</code>	conditional mean, $E(y_j y_j > \tau_j)$
<code>pr(n)</code>	probability $\Pr(y_j = n)$
<code>pr(a,b)</code>	probability $\Pr(a \leq y_j \leq b)$
<code>cpr(n)</code>	conditional probability $\Pr(y_j = n y_j > \tau_j)$
<code>cpr(a,b)</code>	conditional probability $\Pr(a \leq y_j \leq b y_j > \tau_j)$
<code>xb</code>	linear prediction
<code>stdp</code>	standard error of the linear prediction

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

Menu for predict

Statistics > Postestimation > Predictions, residuals, etc.

Options for predict

Main

`n`, the default, calculates the predicted number of events, which is $\exp(\mathbf{x}_j\beta)$ if neither `offset()` nor `exposure()` was specified when the model was fit; $\exp(\mathbf{x}_j\beta + \text{offset}_j)$ if `offset()` was specified; or $\exp(\mathbf{x}_j\beta) \times \text{exposure}_j$ if `exposure()` was specified.

`ir` calculates the incidence rate $\exp(\mathbf{x}_j\beta)$, which is the predicted number of events when exposure is 1. This is equivalent to specifying both the `n` and the `nooffset` options.

`cm` calculates the conditional mean,

$$E(y_j | y_j > \tau_j) = \frac{E(y_j)}{\Pr(y_j > \tau_j)}$$

where τ_j is the truncation point found in `e(11opt)`.

`pr(n)` calculates the probability $\Pr(y_j = n)$, where n is a nonnegative integer that may be specified as a number or a variable.

`pr(a,b)` calculates the probability $\Pr(a \leq y_j \leq b)$, where a and b are nonnegative integers that may be specified as numbers or variables;

b missing ($b \geq .$) means $+\infty$;

`pr(20,.)` calculates $\Pr(y_j \geq 20)$;

`pr(20,b)` calculates $\Pr(y_j \geq 20)$ in observations for which $b \geq .$ and calculates $\Pr(20 \leq y_j \leq b)$ elsewhere.

`pr(., b)` produces a syntax error. A missing value in an observation of the variable a causes a missing value in that observation for `pr(a, b)`.

`cpr(n)` calculates the conditional probability $\Pr(y_j = n \mid y_j > \tau_j)$, where τ_j is the truncation point found in `e(1lopt)`. n is an integer greater than the truncation point that may be specified as a number or a variable.

`cpr(a, b)` calculates the conditional probability $\Pr(a \leq y_j \leq b \mid y_j > \tau_j)$, where τ_j is the truncation point found in `e(1lopt)`. The syntax for this option is analogous to that used for `pr(a, b)` except that a must be greater than the truncation point.

`xb` calculates the linear prediction, which is $\mathbf{x}_j\beta$ if neither `offset()` nor `exposure()` was specified when the model was fit; $\mathbf{x}_j\beta + \text{offset}_j$ if `offset()` was specified; or $\mathbf{x}_j\beta + \ln(\text{exposure}_j)$ if `exposure()` was specified; see `nooffset` below.

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

`nooffset` is relevant only if you specified `offset()` or `exposure()` when you fit the model. It modifies the calculations made by `predict` so that they ignore the offset or exposure variable; the linear prediction is treated as $\mathbf{x}_j\beta$ rather than as $\mathbf{x}_j\beta + \text{offset}_j$ or $\mathbf{x}_j\beta + \ln(\text{exposure}_j)$. Specifying `predict ..., nooffset` is equivalent to specifying `predict ..., ir`.

`scores` calculates equation-level score variables.

The first new variable will contain $\partial \ln L / \partial (\mathbf{x}_j\beta)$.

The second new variable will contain $\partial \ln L / \partial (\ln \alpha)$ for `dispersion(mean)`.

The second new variable will contain $\partial \ln L / \partial (\ln \delta)$ for `dispersion(constant)`.

Methods and formulas

In the following formulas, we use the same notation as in [R] [tmbreg](#).

Methods and formulas are presented under the following headings:

Mean-dispersion model

Constant-dispersion model

Mean-dispersion model

The equation-level scores are given by

$$\begin{aligned} \text{score}(\mathbf{x}\beta)_j &= p_j(y_j - \mu_j) - \frac{p_j^{(m+1)} \mu_j}{\Pr(Y > \tau_j \mid p_j, m)} \\ \text{score}(\omega)_j &= -m \left\{ \frac{\alpha(\mu_j - y_j)}{1 + \alpha\mu_j} - \ln(1 + \alpha\mu_j) + \psi(y_j + m) - \psi(m) \right\} \\ &\quad - \frac{p_j^m}{\Pr(Y > \tau_j \mid p_j, m)} \{m \ln(p_j) + \mu_j p_j\} \end{aligned}$$

where $\omega_j = \ln \alpha_j$, $\psi(z)$ is the digamma function, and τ_j is the truncation point found in `e(1lopt)`.

Constant-dispersion model

The equation-level scores are given by

$$\begin{aligned}\text{score}(\mathbf{x}\beta)_j &= m_j \left\{ \psi(y_j + m_j) - \psi(m_j) + \ln(p) + \frac{p^{m_j} \ln(p)}{\Pr(Y > \tau_j | p, m_j)} \right\} \\ \text{score}(\omega)_j &= y_j - (y_j + m_j)(1 - p) - \text{score}(\mathbf{x}\beta)_j - \frac{\mu_j p}{\Pr(Y > \tau_j | p, m_j)}\end{aligned}$$

where $\omega_j = \ln \delta_j$ and τ_j is the truncation point found in `e(11opt)`.

Also see

[R] **tnbreg** — Truncated negative binomial regression

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