Title

xtpoisson postestimation - Postestimation tools for xtpoisson

Description

The following postestimation commands are available for xtpoisson:

command	description
*estat	AIC, BIC, VCE, and estimation sample summary
estimates	cataloging estimation results
hausman	Hausman's specification test
lincom	point estimates, standard errors, testing, and inference for linear combinations of coefficients
lrtest	likelihood-ratio test
margins	marginal means, predictive margins, marginal effects, and average marginal effects
nlcom	point estimates, standard errors, testing, and inference for nonlinear combinations of coefficients
predict	predictions, residuals, influence statistics, and other diagnostic measures
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

*estat ic is not appropriate after xtpoisson, pa.

See the corresponding entries in the Base Reference Manual for details.

Syntax for predict

Random-effects (RE) and fixed-effects (FE) models

predict [type] newvar [if] [in] [, RE/FE_statistic nooffset]

Population-averaged (PA) model

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

RE/FE_statistic description

Main	
xb	linear prediction; the default
stdp	standard error of the linear prediction
nu0	predicted number of events; assumes fixed or random effect is zero
iru0	predicted incidence rate; assumes fixed or random effect is zero
pr0(<i>n</i>)	probability $Pr(y_j = n)$ assuming the random effect is zero;
	only allowed after xtpoisson, re
pr0(<i>a</i> , <i>b</i>)	probability $Pr(a \le y_j \le b)$ assuming the random effect is zero;
	only allowed after xtpoisson, re

PA_statistic	description
Main	
mu	predicted number of events; considers the offset(); the default
rate	predicted number of events
xb	linear prediction
pr(<i>n</i>)	probability $\Pr(y_j = n)$
pr(a,b)	probability $\Pr(a \le y_j \le b)$
stdp	standard error of the linear prediction
<u>sc</u> ore	first derivative of the log likelihood with respect to $\mathbf{x}_j \boldsymbol{\beta}$

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

Menu

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Statistics > Postestimation > Predictions, residuals, etc.

Options for predict

Main

xb calculates the linear prediction. This is the default for the random-effects and fixed-effects models.

mu and rate both calculate the predicted number of events. mu takes into account the offset(), and rate ignores those adjustments. mu and rate are equivalent if you did not specify offset(). mu is the default for the population-averaged model.

stdp calculates the standard error of the linear prediction.

nu0 calculates the predicted number of events, assuming a zero random or fixed effect.

iru0 calculates the predicted incidence rate, assuming a zero random or fixed effect.

- pr0(n) calculates the probability $Pr(y_j = n)$ assuming the random effect is zero, where n is a nonnegative integer that may be specified as a number or a variable (only allowed after xtpoisson, re).
- pr0(*a*,*b*) calculates the probability $Pr(a \le y_j \le b)$ assuming the random effect is zero, where *a* and *b* are nonnegative integers that may be specified as numbers or variables (only allowed after xtpoisson, re);

b missing $(b \ge .)$ means $+\infty$; pr0(20,.) calculates $\Pr(y_j \ge 20)$; pr0(20,*b*) calculates $\Pr(y_j \ge 20)$ in observations for which $b \ge .$ and calculates $\Pr(20 \le y_j \le b)$ elsewhere.

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

- 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 (only allowed after xtpoisson, pa).
- pr(a,b) calculates the probability $Pr(a \le y_j \le b)$ (only allowed after xtpoisson, pa). The syntax for this option is analogous to that used with pr0(a,b).

score calculates the equation-level score, $u_j = \partial \ln L_j(\mathbf{x}_j \boldsymbol{\beta}) / \partial(\mathbf{x}_j \boldsymbol{\beta})$.

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

Remarks

Example 1

In example 1 of [XT] **xtpoisson**, we fit a random-effects model of the number of accidents experienced by five different types of ships on the basis of when the ships were constructed and operated. Here we obtain the predicted number of accidents for each observation, assuming that the random effect for each panel is zero:

```
. use http://www.stata-press.com/data/r11/ships
```

```
. xtpoisson accident op_75_79 co_65_69 co_70_74 co_75_79, exposure(service) irr
(output omitted)
. predict n_acc, nu0
```

```
(6 missing values generated)
```

```
. summarize n_acc
```

Variable	Obs	Mean	Std. Dev.	Min	Max
n_acc	34	13.52307	23.15885	.0617592	83.31905

From these results, you may be tempted to conclude that some types of ships are safe, with a predicted number of accidents close to zero, whereas others are dangerous, because 1 observation is predicted to have more than 83 accidents.

However, when we fit the model, we specified the exposure(service) option. The variable service records the total number of months of operation for each type of ship constructed in and operated during particular years. Because ships experienced different utilization rates and thus were exposed to different levels of accident risk, we included service as our exposure variable. When comparing different types of ships, we must therefore predict the number of accidents, assuming that all ships faced the same exposure to risk. To do that, we use the iru0 option with predict:

	. predict acc_rate, iru0				
	. summarize acc_rate				
Max	Min	Std. Dev.	Mean	Obs	Variable
.0047429	.0013724	.0010497	.002975	40	acc_rate

These results show that if each ship were used for 1 month, the expected number of accidents is 0.002975. Depending on the type of ship and years of construction and operation, the *incidence rate* of accidents ranges from 0.00137 to 0.00474.

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Methods and formulas

All postestimation commands listed above are implemented as ado-files.

The probabilities calculated using the pr0(n) option are the probability $Pr(y_i = n)$ for a RE model assuming the random effect is zero. These are calculated using

$$Pr(0|\mathbf{x}_i) = \omega_i + (1 - \omega_i)p_2(0|\mathbf{x}_i)$$
$$Pr(n|\mathbf{x}_i) = (1 - \omega_i)p_2(n|\mathbf{x}_i) \quad \text{for } n = 1, 2, \dots$$

where ω_i is the probability of obtaining an observation from the degenerate distribution whose mass is concentrated at zero, and $p_2(n|\mathbf{x}_i)$ is the probability of $y_i = n$ from the nondegenerate, Poisson, RE model. ω_i can be obtained from the pr0() option.

Similarly, the probabilities calculated using the pr(n) option are the probability $Pr(y_i = n)$ for a PA model assuming averaged effects are zero. These are calculated using

$$Pr(0|\mathbf{x}_i) = \omega_i + (1 - \omega_i)p_2(0|\mathbf{x}_i)$$
$$Pr(n|\mathbf{x}_i) = (1 - \omega_i)p_2(n|\mathbf{x}_i) \quad \text{for } n = 1, 2, \dots$$

where ω_i is the probability of obtaining an observation from the degenerate distribution whose mass is concentrated at zero, and $p_2(n|\mathbf{x}_i)$ is the probability of $y_i = n$ from the nondegenerate, Poisson, PA model. ω_i can be obtained from the pr() option.

See Cameron and Trivedi (1998, sec. 4.7) for further details.

Reference

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Cameron, A. C., and P. K. Trivedi. 1998. Regression Analysis of Count Data. Cambridge: Cambridge University Press.

Also see

[XT] **xtpoisson** — Fixed-effects, random-effects, and population-averaged Poisson models

[U] 20 Estimation and postestimation commands