

help binstest

<u>Title</u>

Syntax

where <u>depvar</u> is the dependent variable, <u>indvar</u> is the independent variable for binning, and <u>othercovs</u> are other covariates to be controlled for.

The degree of the piecewise polynomial p, the number of smoothness constraints s, and the derivative order v are integers satisfying $0 \le s,v \le p$, which can take different values in each case.

At least one test has to be specified via testmodelparfit(), testmodelpoly(), testshapel(), testshaper() and/or testshape2().

fweights, aweights and pweights are allowed; see weight.

Description

binstest implements binscatter-based hypothesis testing procedures for parametric
functional forms of and nonparametric shape restrictions on the regression
function estimators, following the results in Cattaneo, Crump, Farrell and
Feng (2022a). If the binning scheme is not set by the user, the companion
command binsregselect is used to implement binscatter in a data-driven
(optimal) way and inference procedures are based on robust bias correction.
Binned scatter plots based on different models can be constructed using the
companion commands binsreg, binsqreg, binslogit and binsprobit.

A detailed introduction to this command is given in <u>Cattaneo, Crump, Farrell and Feng (2022b)</u>. Companion R and Python packages with the same capabilities are available (see website below).

Companion commands: binsreg for binscatter regression with robust inference procedures and plots, binsqreg for binscatter quantile regression with robust inference procedures and plots, binslogit for binscatter logit estimation with robust inference procedures and plots, binsprobit for binscatter probit estimation with robust inference procedures and plots, and binscatter probit for data-driven (optimal) binning selection.

Related Stata, R and Python packages are available in the following website:

https://nppackages.github.io/

Options

- estmethod(cmdname) specifies the binscatter model. The default is estmethod(reg),
 which corresponds to the binscatter least squares regression. Other options
 are: estmethod(qreg #) for binscatter quantile regression where # is the
 quantile to be estimated, estmethod(logit) for binscatter logistic regression
 and estmethod(probit) for binscatter probit regression.
- deriv(v) specifies the derivative order of the regression function for estimation,
 testing and plotting. The default is deriv(0), which corresponds to the
 function itself.
- at(position) specifies the values of othercovs at which the estimated function is
 evaluated for plotting. The default is at(mean), which corresponds to the
 mean of othercovs. Other options are: at(median) for the median of othercovs,
 at(0) for zeros, and at(filename) for particular values of othercovs saved in
 another file.
- Note: When **at(mean)** or **at(median)** is specified, all factor variables in *othercovs* (if specified) are excluded from the evaluation (set as zero).
- nolink specifies that the function within the inverse link (logistic) function be reported instead of the conditional probability function. This option is used only if logit or probit model is specified in estmethod().

Regulate -	

- absorb(absvars) specifies categorical variables (or interactions) representing the
 fixed effects to be absorbed. This is equivalent to including an
 indicator/dummy variable for each category of each absvar. When absorb() is
 specified, the community-contributed command reghdfe instead of the command
 regress is used.
- reghdfeopt(reghdfe_option) options to be passed on to the command reghdfe.
 Important: absorb() and vce() should not be specified within this option.
- For more information about the community-contributed command **reghdfe**, please see http://scorreia.com/software/reghdfe/.

Parametric Model	Cnocification	Tooting	
Parametric Moder	specification	resting	

- testmodel(testmodelopt) sets the degree of polynomial and the number of smoothness
 constraints for parametric model specification testing. If testmodel(p s) is
 specified, a piecewise polynomial of degree p with s smoothness constraints is
 used. If testmodel(T) or testmodel() is specified, testmodel(1 1) is used
 unless the degree p and smoothness s selection is requested via the option
 pselect() (see more details in the explanation of pselect()). The default is
 testmodel().
- testmodelparfit(filename) specifies a dataset which contains the evaluation grid
 and fitted values of the model(s) to be tested against. The file must have a
 variable with the same name as indvar, which contains a series of evaluation
 points at which the binscatter model and the parametric model of interest are
 compared with each other. Each parametric model is represented by a variable
 named as binsreg_fit*, which must contain the fitted values at the
 corresponding evaluation points.
- ${\tt testmodelpoly(p)}$ specifies the degree of a global polynomial model to be tested against.

onparametric Shape Restrictio	on Testing

testshape(testshapeopt) sets the degree of polynomial and the number of smoothness
 constraints for nonparametric shape restriction testing. If testshape(p s) is
 specified, a piecewise polynomial of degree p with s smoothness constraints is
 used. If testshape(T) or testshape() is specified, testshape(1 1) is used
 unless the degree p and smoothness s selection is requested via the option
 pselect() (see more details in the explanation of pselect()). The default is
 testshape().

- **testshapel**(numlist) specifies a numlist of null boundary values for hypothesis testing. Each number a in the numlist corresponds to one boundary of a one-sided hypothesis test to the left of the form H0: $sup_x mu(x) <= a$.
- **testshaper**(numlist) specifies a numlist of null boundary values for hypothesis testing. Each number a in the numlist corresponds to one boundary of a one-sided hypothesis test to the right of the form H0: $inf_x mu(x) >= a$.
- **testshape2** (numlist) specifies a <u>numlist</u> of null boundary values for hypothesis testing. Each number a in the numlist corresponds to one boundary of a two-sided hypothesis test of the form H0: $\sup_{x} |mu(x) a| = 0$.

Metric for Hypothesis Testing

lp(metric) specifies an Lp metric used for (two-sided) parametric model
 specification testing and/or shape restriction testing. The default is
 lp(inf), which corresponds to the sup-norm. Other options are lp(q) for a
 positive integer q.

 $^{flue{}}$ Binning/Degree/Smoothness Selection $^{flue{}}$

- **bins** $(p \ s)$ sets a piecewise polynomial of degree p with s smoothness constraints for data-driven (IMSE-optimal) selection of the partitioning/binning scheme. The default is **bins** $(0 \ 0)$, which corresponds to the piecewise constant.
- nbins(nbinsopt) sets the number of bins for partitioning/binning of indvar. If
 nbins(T) or nbins() (default) is specified, the number of bins is selected via
 the companion command binsregselect in a data-driven, optimal way whenever
 possible. If a numlist with more than one number is specified, the number of
 bins is selected within this list via the companion command binsregselect.
- binspos(position) specifies the position of binning knots. The default is
 binspos(qs), which corresponds to quantile-spaced binning (canonical
 binscatter). Other options are: es for evenly-spaced binning, or a numlist
 for manual specification of the positions of inner knots (which must be within
 the range of indvar).
- binsmethod(method) specifies the method for data-driven selection of the number of bins via the companion command binsregselect. The default is binsmethod(dpi), which corresponds to the IMSE-optimal direct plug-in rule. The other option is: rot for rule of thumb implementation.
- nbinsrot(#) specifies an initial number of bins value used to construct the DPI number of bins selector. If not specified, the data-driven ROT selector is used instead.
- randcut(#) specifies the upper bound on a uniformly distributed variable used to
 draw a subsample for bins/degree/smoothness selection. Observations for which
 runiform() <= # are used. # must be between 0 and 1. By default, max(5,000,
 0.01n) observations are used if the samples size n>5,000.
- **pselect**(numlist) specifies a list of numbers within which the degree of polynomial p for point estimation is selected. If the selected optimal degree is p, then piecewise polynomials of degree p+1 are used to conduct testing for nonparametric shape restrictions or parametric model specifications.
- sselect(numlist) specifies a list of numbers within which the number of smoothness
 constraints s for point estimation. If the selected optimal smoothness is s,
 then piecewise polynomials with s+1 smoothness constraints are used to conduct
 testing for nonparametric shape restrictions or parametric model
 specifications. If not specified, for each value p supplied in the option
 pselect(), only the piecewise polynomial with the maximum smoothness is
 considered, i.e., s=p.
- Note: To implement the degree or smoothness selection, in addition to **pselect()** or **sselect()**, **nbins(#)** must be specified.

J Simulation

- nsims(#) specifies the number of random draws for hypothesis testing. The default is nsims(500), which corresponds to 500 draws from a standard Gaussian random vector of size [(p+1)*J (J-1)*s]. A large number of random draws is recommended to obtain the final results.
- simsgrid(#) specifies the number of evaluation points of an evenly-spaced grid
 within each bin used for evaluation of the supremum (infimum or Lp metric)
 operation needed for hypothesis testing procedures. The default is
 simsgrid(20), which corresponds to 20 evenly-spaced evaluation points within
 each bin for approximating the supremum (infimum or Lp metric) operator. A
 large number of evaluation points is recommended to obtain the final results.

simsseed(#) sets the seed for simulations.

Mass Points and Degrees of Freedom

dfcheck(n1 n2) sets cutoff values for minimum effective sample size checks, which
 take into account the number of unique values of indvar (i.e., adjusting for
 the number of mass points), number of clusters, and degrees of freedom of the
 different statistical models considered. The default is dfcheck(20 30). See
 Cattaneo, Crump, Farrell and Feng (2022b) for more details.

masspoints (masspointsoption) specifies how mass points in indvar are handled. By
 default, all mass point and degrees of freedom checks are implemented.
 Available options:

 ${\it masspoints}$ (${\it noadjust}$) omits mass point checks and the corresponding effective sample size adjustments.

masspoints(nolocalcheck) omits within-bin mass point and degrees of freedom
checks.

masspoints(off) sets masspoints(noadjust) and masspoints(nolocalcheck)
simultaneously.

masspoints(veryfew) forces the command to proceed as if indvar has only a few number of mass points (i.e., distinct values). In other words, forces the command to proceed as if the mass point and degrees of freedom checks were failed.

」 Other Options ↓

vce(vcetype) specifies the vcetype for variance estimation used by the commands regress, logit, probit, greg or reghdfe. The default is vce(robust).

asyvar(on/off) specifies the method used to compute standard errors. If
asyvar(on) is specified, the standard error of the nonparametric component is
used and the uncertainty related to other control variables othercovs is
omitted. Default is asyvar(off), that is, the uncertainty related to othercovs
is taken into account.

estmethodopt(cmd_option) options to be passed on to the estimation command specified in estmethod(). For example, options that control for the optimization process can be added here.

usegtools(on/off) forces the use of several commands in the community-distributed
 Stata package gtools to speed the computation up, if on is specified. Default
 is usegtools(off).

For more information about the package **gtools**, please see https://gtools.readthedocs.io/en/latest/index.html.

Examples

Setup

. sysuse auto

Test for linearity

. binstest mpg weight foreign, testmodelpoly(1)

Test for monotonicity

. binstest mpg weight foreign, deriv(1) bins(1 1) testshapel(0)

Stored results

```
Scalars
                  number of observations
 e (N)
  e(Ndist)
                 number of distinct values
 e(Nclust)
                  number of clusters
 e(nbins)
                 number of bins
 e(p)
                  degree of polynomial for bin selection
                  smoothness of polynomial for bin selection
 e(s)
 e(testshape_p)
                   degree of polynomial for testing shape restrictions
 e(testshape_s)
                   smoothness of polynomial for testing shape restrictions
 e(testmodel_p)
                  degree of polynomial for testing model specifications
 e(testmodel_s)
                   smoothness of polynomial for testing model specifications
 e(testpolyp)
                   degree of polynomial regression model
 e(stat_poly)
                  statistic for testing global polynomial model
 e (pval_poly)
                  p value for testing global polynomial model
 e(imse_var_rot) variance constant in IMSE, ROT selection
 e(imse_bsq_rot) bias constant in IMSE, ROT selection
 e(imse_var_dpi)
                  variance constant in IMSE, DPI selection
  e(imse_bsq_dpi) bias constant in IMSE, DPI selection
Macros
 e(testvarlist) varlist found in testmodel()
 e(testvalue2)
e(testvalueR)
                  values in testshape2()
                  values in testshaper()
  e(testvalueL) values in testshapel()
Matrices
                p values for testmodel()
 e(pval_model)
 e(stat_model)
                  statistics for testmodel()
 e(pval_shape2)
                  p values for testshape2()
 e(stat_shape2)
                  statistics for testshape2()
 e (pval_shapeR)
                  p values for testshaper()
 e(stat_shapeR)
                  statistics for testshaper()
                  p values for testshapel()
 e(pval_shapeL)
 e(stat_shapeL)
                   statistics for testshapel()
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References

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Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2022a. <u>On Binscatter</u>. arXiv:1902.09608.
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Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2022b. <u>Binscatter Regressions</u>. arXiv:1902.09615.
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