
Title

binstest — Data-Driven Nonparametric Shape Restriction and Parametric
 Model Specification Testing using Binscatter.

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 $\emptyset \le s,v \le p$, 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 <u>Cattaneo</u>,
 <u>Crump</u>, <u>Farrell and Feng</u> (2024a) and <u>Cattaneo</u>, <u>Crump</u>, <u>Farrell and Feng</u>
 (2024b). If the binning scheme is not set by the user, the companion
 command <u>binsregselect</u> 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 <u>binsreg</u>, <u>binsqreg</u>, <u>binslogit</u> and binsprobit.

- A detailed introduction to this command is given in <u>Cattaneo, Crump,</u>
 <u>Farrell and Feng (2024c)</u>. Companion R and Python packages with the same capabilities are available (see website below).
- Companion commands: <u>binsreg</u> for binscatter regression with robust inference procedures and plots, <u>binsqreg</u> for binscatter quantile regression with robust inference procedures and plots, <u>binslogit</u> for binscatter logit estimation with robust inference procedures and plots, <u>binsprobit</u> for binscatter probit estimation with robust inference procedures and plots, and <u>binsregselect</u> 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().

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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/.

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	Parametric	Model	Specification	Testing	L

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 or smoothness s
 selection is requested via the option pselect() or sselect() (see more
 details in the explanation of pselect() and sselect()). 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.

testmodelpoly(*p*) specifies the degree of a global polynomial model to be tested against.

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	Nonparametric	Shape Rest	triction ⁱ	Testina	



- 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 l) is used unless the degree p or smoothness s
 selection is requested via the option pselect() or sselect() (see more
 details in the explanation of pselect() and sselect()). 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.</pre>
- **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: $inf \times mu(x) >= a$.
- **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 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 number q no less than 1. Note that lp(inf) ("sup norm") has
 to be used for testing one-sided shape restrictions.

Binning/Degree/Smoothness Selection

- 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 <u>binsregselect</u>. 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(5000, 0.01n) observations are used if the samples
 size n>5000.
- 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.

Simulation	
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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]. Setting at least
 nsims(2000) 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. Setting at least simsgrid(50) 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
 (2024c) 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:

masspoints(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(<u>vcetype</u>) specifies the vcetype for variance estimation used by the commands <u>regress</u>, <u>logit</u>, <u>probit</u>, <u>greg</u> 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
  e(N)
                   number of observations
                   number of distinct values
  e(Ndist)
  e(Nclust)
                   number of clusters
  e(nbins)
                   number of bins
  e(p)
                   degree of polynomial for bin selection
  e(s)
                   smoothness of polynomial for bin selection
  e(testshape_p)
                   degree of polynomial for testing shape restrictions
                   smoothness of polynomial for testing shape
  e(testshape_s)
                     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
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```
e(testvarlist)
                  varlist found in testmodel()
  e(testvalue2)
                  values in testshape2()
  e(testvalueR)
                  values in testshaper()
  e(testvalueL)
                  values in testshapel()
Matrices
  e(pval_model)
                  p values for testmodel()
  e(stat_model)
                  statistics for testmodel()
  e(pval_shape2)
                  p values for testshape2()
                  statistics for testshape2()
  e(stat shape2)
  e(pval_shapeR)
                  p values for testshaper()
  e(stat_shapeR)
                  statistics for testshaper()
  e(pval_shapeL)
                  p values for testshapel()
  e(stat_shapeL)
                  statistics for testshapel()
```

<u>References</u>

```
Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2024a. On Binscatter. American Economic Review 114(5): 1488–1514.
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Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2024b. <u>Nonlinear</u> Binscatter Methods. Working Paper.

Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2024c.

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