

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>covars</u> are other covariates to be controlled for.

p, s and v are integers satisfying 0 <= s,v <= p, which can take different values in each case.

fweights, aweights and pweights are allowed; see weight.

Description

binstest implements binscatter-based hypothesis testing procedures for parametric
 functional forms and nonparametric shape restrictions on of the regression
 function estimators, following the results in Cattaneo, Crump, Farrell and
 Feng (2021a). 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 (2021b)</u>. A companion R package with the same capabilities is 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 binsregselect data-driven (optimal) binning selection.

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

https://nppackages.github.io/

Options

Estimand

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.

- 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.

☐ Nonparametric Shape Restriction Testing

- **testshape** (p s) sets a piecewise polynomial of degree p with s smoothness constraints for nonparametric shape restriction testing. The default is **testmodel(3 3)**, which corresponds to a cubic B-spline estimate of the regression function of interest for one-sided or two-sided testing.
- **testshapel** (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 one-sided hypothesis test to the left of the form H0: $\sup_{x \in A} \sup_{x \in A} \sup_{$
- testshaper(numlist) specifies a <u>numlist</u> of null boundary values for hypothesis
 testing. Each number a in the <u>numlist</u> corresponds to one boundary of a
 one-sided hypothesis test to the right of the form HO: 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 a 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.

☐ Partitioning/Binning 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 piecewise constant (canonical binscatter).
- nbins(#) sets the number of bins for partitioning/binning of indvar. If not specified, the number of bins is selected via the companion command <u>binsregselect</u> in a data-driven, optimal way whenever possible.
- 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.

```
☐ Simulation
```

- **nsims(#)** specifies the number of random draws for constructing confidence bands and 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].
- simsgrid(#) specifies the number of evaluation points of an evenly-spaced grid
 within each bin used for evaluation of the supremum (or infimum) operation
 needed to construct confidence bands and hypothesis testing procedures. The
 default is simsgrid(20), which corresponds to 20 evenly-spaced evaluation
 points within each bin for approximating the supremum (or infimum) operator.
- 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 (2019b) 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(vcetype) specifies the vcetype for variance estimation used by the commands regress, logit or qreg. The default is vce(robust).

Examples

Test linear model

. binstest y x w, testmodelpoly(1)

Stored results

```
Scalars
  e (N)
                   number of observations
  e(Ndist)
                   number of distince values
 e(Nclust)
                   number of clusters
 e(nbins)
                   number of bins
                   degree of polynomial for bin selection
  e (p)
                   smoothness of polynomial for bin selection
  e(s)
  e(testshape_p)
                   degree of polynomial for testing shape
  e(testshape s)
                   smoothnes of polynomial for testing shape
                   degree of polynomial for testing models
  e(testmodel_p)
  e(testmodel_s)
                   smoothness of polynomial for testing models
  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
Locals
```

```
e(testvalueL)
e(testvalueR)
e(testvalue2)
e(testvarlist)

Matrices
e(stat_shapeL)
e(stat_shapeR)
e(stat_shapeR)
e(pval_shapeR)
e(pval_shapeR)
e(stat_shape2)
e(stat_shape2)
e(stat_shape2)
e(stat_shape2)
e(stat_shape2)
e(stat_model)
e(pval_model)
values in testshaper()
values in testshape2()
values in testshape2()
values in testshape2()
values for testshape2()
values for testshape1()
statistics for testshape2()
e(stat_model)
values for testshape2()
values for testshape2()
values for testshape2()
values for testshape2()
values for testshape1()
values in testshape2()
values in testshape2()
values in testshape2()
values in testshape2()
values for testshape1()
values in testshape2()
values in testshape1()
values in testshape2()
values in testshape2()
values in testshape2()
values in testshape2()
values in testshape1()
values in tests
```

References

```
Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2021a. <u>On Binscatter</u>. arXiv:1902.09608.
```

Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2021b. <u>Binscatter Regressions</u>. arXiv:1902.09615.

<u>Authors</u>

Matias D. Cattaneo, Princeton University, Princeton, NJ. cattaneo@princeton.edu.

Richard K. Crump, Federal Reserve Band of New York, New York, NY. richard.crump@ny.frb.org.

Max H. Farrell, University of Chicago, Chicago, IL. max.farrell@chicagobooth.edu.

Yingjie Feng, Princeton University, Princeton, NJ. vingjief@princeton.edu.