

help binstest

Title

othercovs

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.

p, s and v are integers satisfying 0 <= s,v <= 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().

 $fweights,\ aweights\ and\ pweights\ are\ allowed;\ see\ \underline{weight}.$

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 <u>Cattaneo</u>, <u>Crump</u>, <u>Farrell and
 Feng (2021a)</u>. 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 <u>binsprobit</u>.

A detailed introduction to this command is given in $\underline{\text{Cattaneo, Crump, Farrell and }}$ $\underline{\text{Feng (2021b)}}$. 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(greg #) 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.
- $\textbf{reghdfeopt} (\textit{reghdfe_option}) \text{ options to be passed on to the command } \textbf{reghdfe}.$ 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/.

- testmodel(p s) sets a piecewise polynomial of degree p with s smoothness constraints for parametric model specification testing. The default is testmodel(3 3), which corresponds to a cubic B-spline estimate of the regression function of interest for testing against the fitting from a parametric model specification.
- 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.

- 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 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: $sup_x mu(x) \le a$.

- 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: $\sup_{x \to a} x /mu(x) - a/=0$ or .

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.

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
- nbins(#) sets the number of bins for partitioning/binning of indvar. If not specified, the number of bins is selected via the companion command binsregselect 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 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 selection. Observations for which runiform() <= # are used. # must be between 0 and 1.

 \lrcorner Simulation \sqcup

- 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].
- 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.
- simsseed(#) sets the seed for simulations.

 igsqc Mass Points and Degrees of Freedom igll

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 (2021b) for more details.

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

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Other Options
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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.

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

Stored results

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Scalars
 e (N)
                    number of observations
  e(Ndist)
                    number of distinct values
 e(Nclust)
                    number of clusters
                    number of bins
 e(nbins)
                    degree of polynomial for bin selection
  e (p)
  e (s)
                    smoothness of polynomial for bin selection
  e(testshape_p)
                    degree of polynomial for testing shape restrictions
                    smoothness of polynomial for testing shape restrictions degree of polynomial for testing model specifications
  e(testshape s)
  e(testmodel_p)
  e(testmodel_s)
                    smoothness of polynomial for testing model specifications
  e(testpolyp)
                    degree of polynomial regression model
                    statistic for testing global polynomial model
  e(stat_poly)
  e(pval_poly)
                    p value for testing global polynomial model
Locals
  e(testvalueL)
                    values in testshapel()
  e(testvalueR)
                    values in testshaper()
  e(testvalue2)
                    values in testshape2()
  e(testvarlist)
                    varlist found in testmodel()
Matrices
  e(stat_shapeL)
                    statistics for testshapel()
  e (pval_shapeL)
                    p values for testshapel()
  e(stat_shapeR)
                    statistics for testshaper()
  e (pval_shapeR)
                    p values for testshaper()
                    statistics for testshape2()
  e(stat_shape2)
```

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.

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