

help binspwc

<u>Title</u>

binspwc — Data-Driven Nonparametric Pairwise Group Comparison using Binscatter.

Syntax

```
binspwc depvar indvar [othercovs] [if] [in] [weight] , by(varname) [
    estmethod(cmdname) deriv(v) at (position) nolink
    absorb(absvars) reghdfeopt(reghdfe_option)
    pwc(pwcopt) testtype(type) lp(metric)
    bins(p s) bynbins(bynbinsopt) binspos(position) binsmethod(method)
    nbinsrot(#) samebinsby randcut(#)
    pselect(numlist) sselect(numlist)
    nsims(#) simsgrid(#) simsseed(seed)
    dfcheck(n1 n2) masspoints(masspointsoption)
    vce(vcetype) asyvar(on/off) estmethodopt(cmd_option) usegtools(on/off) ]
```

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 <= s,v <= p, which can take different values in each case.

fweights, aweights and pweights are allowed; see weight.

Description

binspwc implements binscatter-based hypothesis testing procedures for pairwise group comparison of binscatter estimators, following the results in 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, 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 least squares 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, binscatter probit estimation with robust inference procedures and plots, and binsregselect for data-driven (optimal) binning selection.

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

https://nppackages.github.io/

<u>Options</u>

_____ Estimand

by (varname) specifies the variable containing the group indicator to perform subgroup analysis; both numeric and string variables are supported. When by (varname) is specified, binspwc implements estimation for each subgroup separately and then conduct all pairwise comparison tests. By default, the binning structure is selected for each subgroup separately, but see the option samebinsby below for imposing a common binning structure across subgroups. This option is required.

- 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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	Reghdfe	

- 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/.

Pairwise Group Comparison Testing
Pairwise Group Comparison Testing

- pwc(pwcopt) sets the degree of polynomial and the number of smoothness constraints
 for pairwise group comparison. If pwc(p s) is specified, a piecewise
 polynomial of degree p with s smoothness constraints is used. If pwc(T) or
 pwc() is specified, pwc(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 pwc().
- **testtype**(*type*) specifies the type of pairwise comparison test. The default is **testtype**(2), which corresponds to a two-sided test of the form H0: $mu_1(x) = mu_2(x)$. Other options are: **testtype**(1) for the one-sided test of the form H0: $mu_1(x) < = mu_2(x)$ and **testtype**(r) for the one-sided test of the form H0: $mu_1(x) > = mu_2(x)$.
- lp(metric) specifies an Lp metric used for a (two-sided) test for the difference
 between two groups. The default is lp(inf), which corresponds to the sup-norm.
 Other options are lp(q) for a positive integer q.

Binning/Degree/Smoothness	Selection	
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 $\begin{array}{ll} \textbf{bins}\,(p\ s) \text{ sets a piecewise polynomial of degree } p \text{ with } s \text{ smoothness constraints} \\ \text{for data-driven (IMSE-optimal) selection of the partitioning/binning scheme.} \\ \text{The default is } \textbf{bins}\,(0\ 0), \text{ which corresponds to the piecewise constant.} \end{array}$

- bynbins(bynbinsopt) sets the number of bins for partitioning/binning of indvar.
 If bynbins(numlist) is specified, the number in the number is applied to the binscatter estimation for each group. The ordering of the group follows the result of tabulate. If a single number of bins is specified, it applies to the estimation for all groups. If bynbins(T) or bynbins() (default) is specified, the number of bins is selected via the companion command binsregselect in a data-driven, optimal way whenever possible.
- Note: If a *numlist* with more than one number is supplied within **bynbins()**, it is understood as the number of bins applied to binscatter estimation for each subgroup rather than the range for selecting the number of bins.
- 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.
- samebinsby forces a common partitioning/binning structure across all subgroups
 specified by the option by(). The knots positions are selected according to
 the option binspos() and using the full sample. If nbins() is not specified,
 then the number of bins is selected via the companion command binsregselect
 and using the full sample.
- 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 pairwise group
 comparison.
- **sselect**(numlist) specifies a list of numbers within which the number of smoothness constraints s for point estimation is selected. If the selected optimal smoothness is s, then piecewise polynomials with s+1 smoothness constraints are used to conduct pairwise group comparison. 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()**, **bynbins(**numlist**)** must be specified.

\longrightarrow Simulation ${}^{igstyle L}$

- 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 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
 (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 <code>indvar</code> (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:

masspoints (noadjust) omits mass point checks and the corresponding effective sample size adjustments.

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

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, logit, 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 **qtools** 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

Generate two groups

. gen group=price>5000

Test for the difference between two groups

. binspwc mpg weight foreign, by (group)

Stored results

```
Scalars
 e (N)
                   number of observations
  e (p)
                   degree of polynomial for bin selection
                   smoothness of polynomial for bin selection
  e(s)
  e (pwc_p)
                   degree of polynomial for testing
                   smoothness of polynomial for testing
  e (pwc_s)
Macros
  e(byvalue)
                   name of groups found in by ()
Matrices
 e (N_by)
                   number of observations for each group
  e(Ndist_by)
                   number of distinct values for each group
                   number of clusters for each group
  e(Nclust_by)
```

```
e(nbins_by)
e(stat)
e(pval)
e(imse_bsq_rot)
e(imse_bsq_dpi)
e(imse_bsq_dpi)
e(imse_bsq_dpi)
e(imse_bsq_dpi)
number of bins for each group
test statistics for all pairwise comparisons
p values for all pairwise comparisons
variance constant in IMSE, ROT selection
variance constant in IMSE, DPI selection
bias constant in IMSE, DPI selection
```

References

Cattaneo, M. D., R. K. Crump, M. H. Farrell, and Y. Feng. 2022a. On Binscatter. arXiv:1902.09608.

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

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