

Title

rdrobust — Local Polynomial Regression Discontinuity Estimation with Robust Bias-Corrected Confidence Intervals and Inference Procedures.

Syntax

```
rdrobust depvar runvar [if] [in] [, c(#) fuzzy(fuzzyvar [sharpbw]) deriv(#)
    scalepar(#) p(#) q(#) h(# #) b(# #) rho(#) covs(covars)
    covs_drop(covsdropoption) kernel(kernelfn) weights(weightsvar)
    bwselect(bwmethod) scaleregul(#) masspoints(masspointsoption) bwcheck(#)
    bwrestrict(bwropt) stdvars(stdopt) vce(vcetype [vceopt1 vceopt2]) level(#) all
```

Description

rdrobust implements local polynomial Regression Discontinuity (RD) point
 estimators with robust bias-corrected confidence intervals and inference
 procedures developed in <u>Calonico</u>, <u>Cattaneo and Titiunik (2014a)</u>, <u>Calonico</u>,
 <u>Cattaneo and Farrell (2018)</u>, <u>Calonico</u>, <u>Cattaneo</u>, <u>Farrell and Titiunik (2019)</u>,
 and <u>Calonico</u>, <u>Cattaneo and Farrell (2020)</u>. It also computes alternative
 estimation and inference procedures available in the literature.

Companion commands are: rdbwselect for data-driven bandwidth selection, and rdbwselect for data-driven RD plots (see Calonico, Cattaneo and Titiunik (2015a) for details).

A detailed introduction to this command is given in <u>Calonico, Cattaneo and Titiunik (2014b)</u>, and <u>Calonico, Cattaneo, Farrell and Titiunik (2017)</u>. A companion <u>R</u> package is also described in <u>Calonico, Cattaneo and Titiunik (2015b)</u>.

Related Stata and R packages useful for inference in RD designs are described in the following website:

https://rdpackages.github.io/

Options |

Estimand

- c(#) specifies the RD cutoff for indepvar. Default is c(0).
- fuzzy(fuzzyvar [sharpbw]) specifies the treatment status variable used to
 implement fuzzy RD estimation (or Fuzzy Kink RD if deriv(1) is also
 specified). Default is Sharp RD design and hence this option is not used. If
 the option sharpbw is set, the fuzzy RD estimation is performed using a
 bandwidth selection procedure for the sharp RD model. This option is
 automatically selected if there is perfect compliance at either side of the
 threshold.
- deriv(#) specifies the order of the derivative of the regression functions to be
 estimated. Default is deriv(0) (for Sharp RD, or for Fuzzy RD if fuzzy(.) is
 also specified). Setting deriv(1) results in estimation of a Kink RD design
 (up to scale), or Fuzzy Kink RD if fuzzy(.) is also specified.
- scalepar(#) specifies scaling factor for RD parameter of interest. This option is
 useful when the estimator of interest requires a known multiplicative factor
 rescaling (e.g., Sharp Kink RD). Default is scalepar(1) (no rescaling).

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Local Polynomial Regression
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p(#) specifies the order of the local polynomial used to construct the point estimator. Default is p(1) (local linear regression).

- q(#) specifies the order of the local polynomial used to construct the bias correction. Default is q(2) (local quadratic regression).
- h(##) specifies the main bandwidth (h) used to construct the RD point estimator. If not specified, bandwidth h is computed by the companion command rdbwselect. If two bandwidths are specified, the first bandwidth is used for the data below the cutoff and the second bandwidth is used for the data above the cutoff.
- \mathbf{b} (# #) specifies the bias bandwidth (b) used to construct the bias-correction estimator. If not specified, bandwidth b is computed by the companion command rdbwselect. If two bandwidths are specified, the first bandwidth is used for the data below the cutoff and the second bandwidth is used for the data above the cutoff.
- **rho**(#) specifies the value of rho, so that the bias bandwidth b = h/rho. Default is rho(1) if h is specified but b is not.
- covs(covars) specifies additional covariates to be used for estimation and inference.
- covs_drop(covsdropoption) assess collinearity in additional covariates used for estimation and inference. Options pinv (default choice) and invsym drops collinear additional covariates, differing only in the type of inverse function used. Option **off** omits the check for collinear additional covariates.
- kernel (kernelfn) specifies the kernel function used to construct the local-polynomial estimator(s). Options are: triangular, epanechnikov, and uniform. Default is kernel (triangular).
- weights (weightsvar) is the variable used for optional weighting of the estimation procedure. The unit-specific weights multiply the kernel function.

Bandwidth Selection

- bwselect (bwmethod) specifies the bandwidth selection procedure to be used. By default it computes both h and b, unless rho is specified, in which case it only computes h and sets b=h/rho. Options are:
 - mserd one common MSE-optimal bandwidth selector for the RD treatment effect estimator.
 - msetwo two different MSE-optimal bandwidth selectors (below and above the cutoff) for the RD treatment effect estimator.
 - ${\tt msesum}$ one common MSE-optimal bandwidth selector for the sum of regression estimates (as opposed to difference thereof).
 - msecomb1 for min(mserd, msesum).
 - msecomb2 for median(msetwo, mserd, msesum), for each side of the cutoff separately.
 - cerrd one common CER-optimal bandwidth selector for the RD treatment effect estimator.
 - certwo two different CER-optimal bandwidth selectors (below and above the cutoff) for the RD treatment effect estimator.
 - cersum one common CER-optimal bandwidth selector for the sum of regression estimates (as opposed to difference thereof).
 - cercomb1 for min(cerrd, cersum).
 - cercomb2 for median(certwo,cerrd,cersum), for each side of the cutoff separately.
 - Note: MSE = Mean Square Error; CER = Coverage Error Rate.
 - Default is bwselect (mserd). For details on implementation see Calonico, Cattaneo and Titiunik (2014a), Calonico, Cattaneo and Farrell (2017), Calonico, Cattaneo and Farrell (2020), and Calonico, Cattaneo, Farrell and Titiunik (2019), and the companion software articles.
- scaleregul(#) specifies scaling factor for the regularization term added to the denominator of the bandwidth selectors. Setting scaleregul(0) removes the regularization term from the bandwidth selectors. Default is scaleregul(1).

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masspoints (masspointsoption) checks and controls for repeated observations in the
    running variable. Options are:
    off ignores the presence of mass points.
    check looks for and reports the number of unique observations at each side of
        the cutoff.
    adjust controls that the preliminary bandwidths used in the calculations
        contain a minimal number of unique observations. By default it uses 10
        observations, but it can be manually adjusted with the option bwcheck.
    Default option is masspoints (adjust).
bwcheck (bwcheck) if a positive integer is provided, the preliminary bandwidth used
    in the calculations is enlarged so that at least bwcheck unique observations
    are used.
bwrestrict(bwropt) if set on, computed bandwidths are restricted to lie within the
    range of runvar. Default is on.
stdvars(stdopt) if set on, depvar and runvar are standardized before computing the
    bandwidths. Default is off.
    Variance-Covariance Estimation
vce(vcetype [vceopt1 vceopt2]) specifies the procedure used to compute the
    variance-covariance matrix estimator. Options are:
    vce (nn [nnmatch]) for heteroskedasticity-robust nearest neighbor variance
        estimator with nnmatch indicating the minimum number of neighbors to be
        used.
    vce(hc0) for heteroskedasticity-robust plug-in residuals variance estimator
        without weights.
    vce (hcl) for heteroskedasticity-robust plug-in residuals variance estimator
        with hcl weights.
    vce(hc2) for heteroskedasticity-robust plug-in residuals variance estimator
        with hc2 weights.
    vce(hc3) for heteroskedasticity-robust plug-in residuals variance estimator
        with hc3 weights.
    vce(nncluster clustervar [nnmatch]) for cluster-robust nearest neighbor
        variance estimation using with clustervar indicating the cluster ID
        variable and nnmatch matches indicating the minimum number of neighbors to
        be used.
    vce(cluster clustervar) for cluster-robust plug-in residuals variance
        estimation with degrees-of-freedom weights and clustervar indicating the
        cluster ID variable.
    Default is vce(nn 3).
level (#) specifies confidence level for confidence intervals. Default is
    level(95).
```

Other Options

all if specified, rdrobust reports three different procedures:

- (i) conventional RD estimates with conventional variance estimator.
- (ii) bias-corrected RD estimates with conventional variance estimator.
- (iii) bias-corrected RD estimates with robust variance estimator.

Example: Cattaneo, Frandsen and Titiunik (2015) Incumbency Data

. use rdrobust senate.dta

Robust RD Estimation using MSE bandwidth selection procedure

. rdrobust vote margin

Robust RD Estimation with both bandwidths set to 15

. rdrobust vote margin, h(15)

e(vce_select)

e(bwselect)

e(kernel)

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Other generic examples (\mathbf{y} outcome variable, \mathbf{x} running variable, \mathbf{t} treatment
        take-up indicator):
        Estimation for Sharp RD designs
            . rdrobust y x, deriv(0)
        Estimation for Sharp Kink RD designs
            . rdrobust y x, deriv(1)
        Estimation for Fuzzy RD designs
            . rdrobust y x, fuzzy(t)
        Estimation for Fuzzy Kink RD designs
            . rdrobust y x, fuzzy(t) deriv(1)
Stored results
    rdrobust stores the following in e():
    Scalars
      e (N)
                          original number of observations
      e(N_1)
                          original number of observations to the left of the cutoff
      e(Nr)
                          original number of observations to the right of the cutoff
                          effective number of observations (given by the bandwidth
      e(N_h_1)
                            h_l) used to the left of the cutoff
      e (N_h_r)
                          effective number of observations (given by the bandwidth
                            h_r) used to the right of the cutoff
                          effective number of observations (given by the bandwidth
      e(N_b_1)
                            b_l) used to the left of the cutoff
      e(Nbr)
                          effective number of observations (given by the bandwidth
                            b_r) used to the right of the cutoff
      e(c)
                          cutoff value
                          order of the polynomial used for estimation of the
      e (p)
                            regression function
                          order of the polynomial used for estimation of the bias of
      e (q)
                            the regression function estimator
      e(h_1)
                          bandwidth used for estimation of the regression function
                            below the cutoff
                          bandwidth used for estimation of the regression function
      e(h_r)
                            above the cutoff
                          bandwidth used for estimation of the bias of the
      e(b_1)
                            regression function estimator below the cutoff
                          bandwidth used for estimation of the bias of the
      e(b_r)
                            regression function estimator above the cutoff
      e(tau cl)
                          conventional local-polynomial RD estimate
                          conventional local-polynomial left estimate
      e(tau_cl_1)
      e(tau_cl_r)
                          conventional local-polynomial right estimate
      e(tau_bc)
                          bias-corrected local-polynomial RD estimate
      e(tau_bc_1)
                          bias-corrected local-polynomial left estimate
      e(tau_bc_r)
                          bias-corrected local-polynomial right estimate
                          conventional standard error of the local-polynomial RD
      e(se_tau_cl)
                            estimator
                          robust standard error of the local-polynomial RD estimator
      e(se_tau_rb)
      e(bias_l)
                          estimated bias for the local-polynomial RD estimator below
                            the cutoff
                          estimated bias for the local-polynomial RD estimator above
      e(bias_r)
                            the cutoff
      e(runningvar)
                          name of running variable
      e(outcomevar)
                          name of outcome variable
      e(clustvar)
                          name of cluster variable
                          name of covariates
      e(covs)
```

vcetype specified in vce()

bandwidth selection choice

kernel choice

Matrices	
e(beta_p_r)	<pre>conventional p-order local-polynomial estimates to the right of the cutoff</pre>
e(beta_p_1)	<pre>conventional p-order local-polynomial estimates to the left of the cutoff</pre>
e(beta_covs)	<pre>coefficients of the additional covariates, only returned when covs() are used</pre>
e(V_cl_r)	<pre>conventional variance-covariance matrix to the right of the cutoff</pre>
e(V_cl_l)	<pre>conventional variance-covariance matrix to the left of the cutoff</pre>
e(V_rb_r)	robust variance-covariance matrix to the right of the cutoff
e(V_rb_1)	<pre>robust variance-covariance matrix to the left of the cutoff</pre>

References

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