

## **Title**

rdsampsi — Sample size selection based on power calculations for Regression Discontinuity designs using robust bias-corrected local polynomial inference.

### Syntax

```
rdsampsi depvar runvar [if] [in] [, c(#) tau(#) alpha(#) beta(#) nsamples(# # # #)
    samph(# #) all plot graph_range(# #) graph_step(#) graph_options(graph_opt)
    bias(# #) variance(# #) nratio(#) init_cond(#) covs(covars)
    covs_drop(covsdropoption) deriv(#) p(#) h(# #) b(# #) rho(#)
    fuzzy(fuzzyvar [sharpbw]) kernel(kernelfn) bwselect(bwmethod) vce(vcetype
    [vceopt1 vceopt2]) weights(weightsvar) scalepar(#) scaleregul(#)
    masspoints(masspointsoption) bwcheck(#) bwrestrict(bwropt) stdvars(stdopt) ]
```

#### Description

rdsampsi provides sample size selection based on power calculations in Regression
 Discontinuity designs using conventional and robust bias-corrected local
 polynomial methods. Companion command is: rdpow for power calculations.

A detailed introduction to this command is given in  $\underline{\text{Cattaneo, Titiunik and }}$   $\underline{\text{Vazquez-Bare (2019)}}$ .

Companion R functions are also available <u>here</u>.

This command employs the Stata (and R) package <u>rdrobust</u> for underlying calculations. See <u>Calonico</u>, <u>Cattaneo and Titiunik (2014)</u> and <u>Calonico</u>, <u>Cattaneo</u>, <u>Farrell and Titiunik (2017)</u> for more details.

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

https://sites.google.com/site/rdpackages/

# Options

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rdsampsi options
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- c(#) specifies the RD cutoff for indepvar. Default is c(0).
- tau(#) specifies the treatment effect under the alternative at which the power function is evaluated. The default is half the standard deviation of the outcome for the untreated group.
- alpha(#) specifies the significance level for the power function. Default is alpha(.05).
- beta(#) specifies the desired power. Default is beta(.8).
- nsamples(# # # #) sets the total sample size to the left, sample size to the left
  inside the bandwidth, total sample size to the right and sample size to the
  right of the cutoff inside the bandwidth to calculate the variance when the
  running variable is not specified. When this option is not specified, the
  values are calculated using the running variable.
- samph(# #) sets the bandwidths at each side of the cutoff for power calculation.
   The first number is the bandwidth to the left of the cutoff and the second
   number is the bandwidth to the right. Default values are the bandwidths used
   by rdrobust.
- all displays the power using the conventional variance estimator, in addition to the robust bias corrected one.

- plot plots the power function using the robust bias corrected standard error from rdrobust. If all is specified, the conventional power function is also plotted.
- $graph\_range(# #)$  specifies the range of the plot when plot option is used. Default range is [-1.5\*tau ; 1.5\*tau].
- graph\_step(#) specifies the step increment of the plot when plot option is used.
   Default range is 0.2\*range.
- graph\_options(#) specifies the graph options (title, axes titles, etc) to be passed to the plot when plot option is used.
- bias(# #) allows the user to set bias to the left and right of the cutoff. If not specified, the biases are estimated using rdrobust.
- variance(# #) allows the user to set variance to the left and right of the cutoff.
  If not specified, the variances are estimated using rdrobust.
- nratio(#) specifies the proportion of treated units in the window. Default is the
   ratio of the standard deviation of the treated to the sum of the standard
   deviations for treated and controls.
- init\_cond(#) sets the initial condition for the Newton-Raphson algorithm that
   finds the sample size. Default is the number of observations in the sample
   with non-missing values of the outcome and running variable.

rdrobust options

The following options are passed directly to rdrobust:

- covs(covars) specifies additional covariates to be used for estimation and inference.
- covs\_drop(covsdropoption) specifies options to assess collinearity in covariates
   to be used for estimation and inference. Option on drops collinear additional
   covariates (default choice). Option off only checks collinear additional
   covariates but does not drop them.
- deriv(#) specifies the order of the derivative of the regression functions to be
   estimated. Default is deriv(0). Setting deriv(1) results in estimation of a
   Kink RD design (up to scale).
- p(#) specifies the order of the local polynomial used to construct the point estimator. Default is p(1) (local linear regression).
- ${\bf q}\,(\#)$  specifies the order of the local polynomial used to construct the bias correction. Default is  ${\bf q}\,(2)$  (local quadratic regression).
- $\mathbf{h}$  (# #) specifies the main bandwidth (h) used to construct the RD point estimator. If not specified, bandwidth h is computed by the companion command  $\underline{\mathbf{rdbwselect}}$ . 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.
- 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 equals b=h/rho. Default is **rho**(1) if h is specified but b is not.

- 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.
- kernel(kernelfn) specifies the kernel function used to construct the
   local-polynomial estimator(s). Options are: triangular, epanechnikov, and
   uniform. Default is kernel(triangular).
- **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.
  - 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.
  - ${\tt cerrd}$  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.
  - ${\tt cersum}$  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 <u>Calonico</u>, <u>Cattaneo and Titiunik (2014a)</u>, <u>Calonico</u>, <u>Cattaneo and Farrell (2016a)</u>, and <u>Calonico</u>, <u>Cattaneo</u>, <u>Farrell and Titiunik (2016)</u>, and the companion software articles.
- 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(hc1) for heteroskedasticity-robust plug-in residuals variance estimator
     with hc1 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).
- weights(weightsvar) is the variable used for optional weighting of the estimation
  procedure. The unit-specific weights multiply the kernel function.
- 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).
- 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).

```
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.
Example: Cattaneo, Frandsen and Titiunik (2015) Incumbency Data
        . use rdpow_senate.dta
    Sample size calculation against an alternative hypothesis of tau = 5
        . rdsampsi demvoteshfor2 demmv, tau(5)
    Sample size calculation with covariates
        . rdsampsi demvoteshfor2 demmv, tau(5) covs(population dopen dmidterm)
    Sample size calculation with user-specified bandwidths
        . rdsampsi demvoteshfor2 demmv, tau(5) h(16 18) b(18 20)
    Sample size calculation with user-specified options
        . rdsampsi demvoteshfor2 demmv, tau(5) beta(.9) all samph(18 19) nratio(.5)
   Power function plot with default options
        . rdsampsi demvoteshfor2 demmv, tau(5) plot
   Power function plot with user-specified range and step
        . rdsampsi demvoteshfor2 demmv, tau(5) plot graph_range(0 800) graph_step(200)
   Power function plot with user-specified options
        . rdsampsi demvoteshfor2 demmv, tau(5) plot graph_range(0 800) graph_step(200)
        graph_options(title(Power function) xtitle(sample size) ytitle(power)
        graphregion(fcolor(white)))
Saved results
    rdsampsi saves the following in r():
    Scalars
      r(alpha)
                          significance level
      r(beta)
                          desired power
                          desired effect
     r(tau)
      r(samph_1)
                          bandwidth to the left of the cutoff
                         bandwidth to the right of the cutoff
      r(samph_h)
      r(var_l)
                         robust bias corrected variance to the left of the cutoff
      r(var_r)
                         robust bias corrected variance to the right of the cutoff
                         bias to the left of the cutoff
      r(bias_l)
                         bias to the right of the cutoff
      r(bias_r)
      r(N_h_1)
                         sample size in bandwidth to the left of the cutoff for
                           variance calculation
      r(N_h_r)
                         sample size in bandwidth to the right of the cutoff for
                           variance calculation
                         sample size to the left of the cutoff for variance
      r(N_1)
                           calculation
      r(N_r)
                          sample size to the right of the cutoff for variance
```

calculation r(sampsi\_tot) implied total sample size using robust bias corrected s.e. r(sampsi\_h\_l) sample size to the left of the cutoff using robust bias corrected s.e. sample size to the right of the cutoff using robust bias r(sampsi\_h\_r) corrected s.e. sample size inside the window using robust bias corrected r(sampsi\_h\_tot) s.e. conventional variance to the left of the cutoff r(var\_l\_cl) conventional variance to the right of the cutoff r(var\_r\_cl) r(sampsi\_tot\_cl) implied total sample size using conventional s.e. r(sampsi\_h\_l\_cl) sample size to the left of the cutoff using conventional s.e. sample size to the right of the cutoff using conventional r(sampsi\_h\_r\_cl) s.e. r(sampsi\_h\_tot\_cl) sample size inside the window using conventional s.e. r(no\_iter) number of iterations until convergence of the Newton-Raphson algorithm r(init\_cond) initial condition of the Newton-Raphson algorithm

## References

- Calonico, S., M. D. Cattaneo, M. H. Farrell, and R. Titiunik. 2017. <a href="mailto:rdrobust: software for Regression Discontinuity Designs">rdrobust: software for Regression Discontinuity Designs</a>.

  Stata Journal 17(2): 372-404.
- Calonico, S., M. D. Cattaneo, and R. Titiunik. 2014. <u>Robust Data-Driven Inference in the Regression-Discontinuity Design</u>.

  Stata Journal 14(4): 909-946.
- Cattaneo, M. D., Frandsen, B., and R. Titiunik. 2015. <u>Randomization Inference in the Regression Discontinuity Design: An Application to Party Advantages in the U.S. Senate</u>.

  Journal of Causal Inference 3(1): 1-24.
- Cattaneo, M. D., R. Titiunik, and G. Vazquez-Bare. 2019. <u>Power Calculations for Regression Discontinuity Designs</u>.

  Stata Journal 19(1): 210-245.

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