

# Package: nprobust (via r-universe)

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**Type** Package

**Title** Nonparametric Robust Estimation and Inference Methods using  
Local Polynomial Regression and Kernel Density Estimation

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**Description** Tools for data-driven statistical analysis using local polynomial regression and kernel density estimation methods as described in Calonico, Cattaneo and Farrell (2018, <[doi:10.1080/01621459.2017.1285776](https://doi.org/10.1080/01621459.2017.1285776)>): lprobust() for local polynomial point estimation and robust bias-corrected inference, lpbwselect() for local polynomial bandwidth selection, kdrobust() for kernel density point estimation and robust bias-corrected inference, kdbwselect() for kernel density bandwidth selection, and nprobust.plot() for plotting results. The main methodological and numerical features of this package are described in Calonico, Cattaneo and Farrell (2019, <[doi:10.18637/jss.v091.i08](https://doi.org/10.18637/jss.v091.i08)>).

**Depends** R (>= 3.1.1)

**License** GPL-2

**Imports** Rcpp, ggplot2

**LinkingTo** Rcpp, RcppArmadillo

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**Repository** CRAN

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## Contents

nprobust-package . . . . .	2
kdbwselect . . . . .	3
kdrobust . . . . .	5
lpbwselect . . . . .	7
lprobust . . . . .	9
nprobust.plot . . . . .	13

<b>Index</b>	<b>16</b>
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nprobust-package	<i>Nonparametric Robust Estimation and Inference Methods using Local Polynomial Regression and Kernel Density Estimation</i>
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## Description

This package provides tools for data-driven statistical analysis using local polynomial regression (LPR) and kernel density estimation (KDE) methods as described in Calonico, Cattaneo and Farrell (2018): [lprobust](#) for local polynomial point estimation and robust bias-corrected inference, [lpbwselect](#) for local polynomial bandwidth selection, [kdrobust](#) for kernel density point estimation and robust bias-corrected inference, [kdbwselect](#) for kernel density bandwidth selection, and [nprobust.plot](#) for plotting results. The main methodological and numerical features of this package are described in Calonico, Cattaneo and Farrell (2019).

## Details

Package: nprobust  
 Type: Package  
 Version: 0.4.0  
 Date: 2020-08-24  
 License: GPL-2

Function for LPR estimation and inference: [lprobust](#)  
 Function for LPR bandwidth selection: [lpbwselect](#)  
 Function for KDE estimation and inference: [kdrobust](#)  
 Function for KDE bandwidth selection: [kdbwselect](#)  
 Function for graphical analysis: [nprobust.plot](#)

## Author(s)

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 Max H. Farrell, University of Chicago, Chicago, IL. <[max.farrell@chicagobooth.edu](mailto:max.farrell@chicagobooth.edu)>.

## References

Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2018. [On the Effect of Bias Estimation on Coverage Accuracy in Nonparametric Inference](#). *Journal of the American Statistical Association*, 113(522): 767-779. doi: [10.1080/01621459.2017.1285776](https://doi.org/10.1080/01621459.2017.1285776).

Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2019. [nprobust: Nonparametric Kernel-Based Estimation and Robust Bias-Corrected Inference](#). *Journal of Statistical Software*, 91(8): 1-33. doi: [10.18637/jss.v091.i08](https://doi.org/10.18637/jss.v091.i08).

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kdbwselect	<i>Bandwidth Selection Procedures for Kernel Density Estimation and Inference</i>
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## Description

`kdbwselect` implements bandwidth selectors for kernel density point estimators and inference procedures developed in Calonico, Cattaneo and Farrell (2018). See also Calonico, Cattaneo and Farrell (2020) for related optimality results. It also implements other bandwidth selectors available in the literature. See Wand and Jones (1995) for background references.

Companion commands are: `kdrobust` for kernel density point estimation and inference procedures.

A detailed introduction to this command is given in Calonico, Cattaneo and Farrell (2019). For more details, and related Stata and R packages useful for empirical analysis, visit <https://nppackages.github.io/>.

## Usage

```
kdbwselect(x, eval = NULL, neval = NULL, kernel = "epa",
bwselect = "mse-dpi", bwcheck=21, imsegrid=30, subset = NULL)
```

## Arguments

<code>x</code>	independent variable.
<code>eval</code>	vector of evaluation point(s). By default it uses 30 equally spaced points over to support of <code>x</code> .
<code>neval</code>	number of quantile-spaced evaluation points on support of <code>x</code> . Default is <code>neval=30</code> .
<code>kernel</code>	kernel function used to construct the kernel estimators. Options are <code>epa</code> for the epanechnikov kernel, and <code>uni</code> for the uniform kernel. Default is <code>kernel = epa</code> .
<code>bwselect</code>	bandwidth selection procedure to be used. Options are: <code>mse-dpi</code> second-generation DPI implementation of MSE-optimal bandwidth. Default option. <code>imse-dpi</code> second-generation DPI implementation of IMSE-optimal bandwidth (computed using grid of evaluation points selected). <code>imse-rot</code> ROT implementation of IMSE-optimal bandwidth (computed using grid of evaluation points selected). <code>ce-dpi</code> second generation DPI implementation of CE-optimal bandwidth.

ce-rot ROT implementation of CE-optimal bandwidth.

all reports all available bandwidth selection procedures.

Note: MSE = Mean Square Error; IMSE = Integrated Mean Squared Error; CE = Coverage Error; DPI = Direct Plug-in; ROT = Rule-of-Thumb. For details on implementation see Calonico, Cattaneo and Farrell (2019).

bwcheck	if a positive integer is provided, then the selected bandwidth is enlarged so that at least bwcheck effective observations are available at each evaluation point. Default is bwcheck = 15.
imsegrid	number of evaluations points used to compute the IMSE bandwidth selector. Default is imsegrid = 30.
subset	optional rule specifying a subset of observations to be used.

### Value

Estimate	A matrix containing eval (grid points), h and b (bandwidths).
opt	A list containing options passed to the function.

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### References

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Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2019. [nprobust: Nonparametric Kernel-Based Estimation and Robust Bias-Corrected Inference](#). *Journal of Statistical Software*, 91(8). doi: [10.18637/jss.v091.i08](#).

Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2020. [Coverage Error Optimal Confidence Intervals for Local Polynomial Regression](#). Working Paper.

Fan, J., and Gijbels, I. 1996. *Local polynomial modelling and its applications*, London: Chapman and Hall.

Wand, M., and Jones, M. 1995. *Kernel Smoothing*, Florida: Chapman & Hall/CRC.

### See Also

[kdrobust](#)

### Examples

```
x <- rnorm(500)
est <- kdbwselect(x)
summary(est)
```

**Description**

`kdrobust` implements kernel density point estimators, with robust bias-corrected confidence intervals and inference procedures developed in Calonico, Cattaneo and Farrell (2018). See also Calonico, Cattaneo and Farrell (2020) for related optimality results. It also implements other estimation and inference procedures available in the literature. See Wand and Jones (1995) for background references.

Companion commands: `kdbwselect` for kernel density data-driven bandwidth selection, and `nprobust.plot` for plotting results.

A detailed introduction to this command is given in Calonico, Cattaneo and Farrell (2019). For more details, and related Stata and R packages useful for empirical analysis, visit <https://nppackages.github.io/>.

**Usage**

```
kdrobust(x, eval = NULL, neval = NULL, h = NULL, b = NULL, rho = 1,
kernel = "epa", bwselect = NULL, bwcheck = 21, imsegrid=30, level = 95, subset = NULL)
```

**Arguments**

<code>x</code>	independent variable.
<code>eval</code>	vector of evaluation point(s). By default it uses 30 equally spaced points over to support of <code>x</code> .
<code>neval</code>	number of quantile-spaced evaluation points on support of <code>x</code> . Default is <code>neval=30</code> .
<code>h</code>	main bandwidth used to construct the kernel density point estimator. Can be either scalar (same bandwidth for all evaluation points), or vector of same dimension as <code>eval</code> . If not specified, bandwidth <code>h</code> is computed by the companion command <code>kdbwselect</code> .
<code>b</code>	bias bandwidth used to construct the bias-correction estimator. Can be either scalar (same bandwidth for all evaluation points), or vector of same dimension as <code>eval</code> . By default it is set equal to <code>h</code> . If <code>rho</code> is set to zero, <code>b</code> is computed by the companion command <code>kdbwselect</code> .
<code>rho</code>	Sets $b=h/\rho$ . Default is <code>rho = 1</code> .
<code>kernel</code>	kernel function used to construct local polynomial estimators. Options are <code>epa</code> for the epanechnikov kernel, <code>tri</code> for the triangular kernel and <code>uni</code> for the uniform kernel. Default is <code>kernel = epa</code> .
<code>bwselect</code>	bandwidth selection procedure to be used via <code>lpbwselect</code> . By default it computes <code>h</code> and sets $b=h/\rho$ (with <code>rho=1</code> by default). It computes both <code>h</code> and <code>b</code> if <code>rho</code> is set equal to zero. Options are: <code>mse-dpi</code> second-generation DPI implementation of MSE-optimal bandwidth. Default option if only one evaluation point is chosen.

`imse-dpi` second-generation DPI implementation of IMSE-optimal bandwidth (computed using a grid of evaluation points). Default option if more than one evaluation point is chosen.

`imse-rot` ROT implementation of IMSE-optimal bandwidth (computed using a grid of evaluation points).

`ce-dpi` second generation DPI implementation of CE-optimal bandwidth.

`ce-rot` ROT implementation of CE-optimal bandwidth.

`all` reports all available bandwidth selection procedures.

Note: MSE = Mean Square Error; IMSE = Integrated Mean Squared Error; CE = Coverage Error; DPI = Direct Plug-in; ROT = Rule-of-Thumb. For details on implementation see Calonico, Cattaneo and Farrell (2019).

<code>bwcheck</code>	if a positive integer is provided, then the selected bandwidth is enlarged so that at least <code>bwcheck</code> effective observations are available at each evaluation point. Default is <code>bwcheck = 21</code> .
<code>imsegrid</code>	number of evaluations points used to compute the IMSE bandwidth selector. Default is <code>imsegrid = 30</code> .
<code>level</code>	confidence level used for confidence intervals; default is <code>level = 95</code> .
<code>subset</code>	optional rule specifying a subset of observations to be used.

#### Value

<code>Estimate</code>	A matrix containing <code>eval</code> (grid points), <code>h</code> , <code>b</code> (bandwidths), <code>N</code> (effective sample sizes), <code>tau.us</code> (point estimates with <code>p</code> -th order kernel function), <code>tau.bc</code> (bias corrected point estimates), <code>se.us</code> (standard error corresponding to <code>tau.us</code> ), and <code>se.rb</code> (robust standard error).
<code>opt</code>	A list containing options passed to the function.

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#### References

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Fan, J., and Gijbels, I. 1996. *Local polynomial modelling and its applications*, London: Chapman and Hall.

Wand, M., and Jones, M. 1995. *Kernel Smoothing*, Florida: Chapman & Hall/CRC.

**See Also**[kdbwselect](#)**Examples**

```
x <- rnorm(500)
est <- kdrobust(x)
summary(est)
```

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lpbwselect	<i>Bandwidth Selection Procedures for Local Polynomial Regression Estimation and Inference</i>
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**Description**

[lpbwselect](#) implements bandwidth selectors for local polynomial regression point estimators and inference procedures developed in Calonico, Cattaneo and Farrell (2018). See also Calonico, Cattaneo and Farrell (2020) for related optimality results. It also implements other bandwidth selectors available in the literature. See Wand and Jones (1995) and Fan and Gijbels (1996) for background references.

Companion commands: [lprobust](#) for local polynomial point estimation and inference procedures.

A detailed introduction to this command is given in Calonico, Cattaneo and Farrell (2019). For more details, and related Stata and R packages useful for empirical analysis, visit <https://nppackages.github.io/>.

**Usage**

```
lpbwselect(y, x, eval = NULL, neval = NULL, p = NULL, deriv = NULL,
kernel = "epa", bwselect = "mse-dpi", bwcheck = 21, bwregul = 1,
imsegrid = 30, vce = "nn", cluster = NULL,
nnmatch = 3, interior = FALSE, subset = NULL)
```

**Arguments**

y	dependent variable.
x	independent variable.
eval	vector of evaluation point(s). By default it uses 30 equally spaced points over to support of x.
neval	number of quantile-spaced evaluation points on support of x. Default is neval=30.
p	polynomial order used to construct point estimator; default is p = 1 (local linear regression).
deriv	derivative order of the regression function to be estimated. Default is deriv=0 (regression function).

kernel	kernel function used to construct local polynomial estimators. Options are <code>epa</code> for the epanechnikov kernel, <code>tri</code> for the triangular kernel, <code>uni</code> for the uniform kernel and <code>gau</code> for the gaussian kernel. Default is <code>kernel = epa</code> .
bwselect	bandwidth selection procedure to be used. Options are: <code>mse-dpi</code> second-generation DPI implementation of MSE-optimal bandwidth. Default option. <code>mse-rot</code> ROT implementation of MSE-optimal bandwidth. <code>imse-dpi</code> second-generation DPI implementation of IMSE-optimal bandwidth (computed using grid of evaluation points selected). <code>imse-rot</code> ROT implementation of IMSE-optimal bandwidth (computed using grid of evaluation points selected). <code>ce-dpi</code> second generation DPI implementation of CE-optimal bandwidth. <code>ce-rot</code> ROT implementation of CE-optimal bandwidth. <code>all</code> reports all available bandwidth selection procedures. Note: MSE = Mean Square Error; IMSE = Integrated Mean Squared Error; CE = Coverage Error; DPI = Direct Plug-in; ROT = Rule-of-Thumb. For details on implementation see Calonico, Cattaneo and Farrell (2019).
bwcheck	if a positive integer is provided, then the selected bandwidth is enlarged so that at least <code>bwcheck</code> effective observations are available at each evaluation point. Default is <code>bwcheck = 21</code> .
bwregul	specifies scaling factor for the regularization term added to the denominator of bandwidth selectors. Setting <code>bwregul = 0</code> removes the regularization term from the bandwidth selectors. Default is <code>bwregul = 1</code> .
imsegrid	number of evaluations points used to compute the IMSE bandwidth selector. Default is <code>imsegrid = 30</code> .
vce	procedure used to compute the variance-covariance matrix estimator. Options are: <code>nn</code> heteroskedasticity-robust nearest neighbor variance estimator with <code>nnmatch</code> the (minimum) number of neighbors to be used. Default choice. <code>hc0</code> heteroskedasticity-robust plug-in residuals variance estimator without weights. <code>hc1</code> heteroskedasticity-robust plug-in residuals variance estimator with <code>hc1</code> weights. <code>hc2</code> heteroskedasticity-robust plug-in residuals variance estimator with <code>hc2</code> weights. <code>hc3</code> heteroskedasticity-robust plug-in residuals variance estimator with <code>hc3</code> weights.
cluster	indicates the cluster ID variable used for cluster-robust variance estimation with degrees-of-freedom weights. By default it is combined with <code>vce=nn</code> for cluster-robust nearest neighbor variance estimation. Another option is plug-in residuals combined with <code>vce=hc1</code> .
nnmatch	to be combined with for <code>vce=nn</code> for heteroskedasticity-robust nearest neighbor variance estimator with <code>nnmatch</code> indicating the minimum number of neighbors to be used. Default is <code>nnmatch=3</code>
.	
interior	if TRUE, all evaluation points are assumed to be interior points. This option affects only data-driven bandwidth selection via <code>lpbwselect</code> . Default is <code>interior = FALSE</code> .
subset	optional rule specifying a subset of observations to be used.



**Value**

Estimate      A matrix containing grid (grid points), h and b (bandwidths), N (sample size)  
opt            A list containing options passed to the function.

**Author(s)**

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**References**

Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2018. [On the Effect of Bias Estimation on Coverage Accuracy in Nonparametric Inference](#). *Journal of the American Statistical Association*, 113(522): 767-779. doi: [10.1080/01621459.2017.1285776](#).

Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2019. [nprobust: Nonparametric Kernel-Based Estimation and Robust Bias-Corrected Inference](#). *Journal of Statistical Software*, 91(8): 1-33. doi: [10.18637/jss.v091.i08](#).

Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2020. [Coverage Error Optimal Confidence Intervals for Local Polynomial Regression](#). Working Paper.

Fan, J., and Gijbels, I. 1996. *Local polynomial modelling and its applications*, London: Chapman and Hall.

Wand, M., and Jones, M. 1995. *Kernel Smoothing*, Florida: Chapman & Hall/CRC.

**See Also**

[lprobust](#)

**Examples**

```
x <- runif(500)
y <- sin(4*x) + rnorm(500)
est <- lpbwselect(y,x)
summary(est)
```

**Description**

`lprobust` implements local polynomial regression point estimators, with robust bias-corrected confidence intervals and inference procedures developed in Calonico, Cattaneo and Farrell (2018). See also Calonico, Cattaneo and Farrell (2020) for related optimality results. It also implements other estimation and inference procedures available in the literature. See Wand and Jones (1995) and Fan and Gijbels (1996) for background references.

Companion commands: `lpbwselect` for local polynomial data-driven bandwidth selection, and `nprobust.plot` for plotting results.

A detailed introduction to this command is given in Calonico, Cattaneo and Farrell (2019). For more details, and related Stata and R packages useful for empirical analysis, visit <https://nppackages.github.io/>.

**Usage**

```
lprobust(y, x, eval = NULL, neval = NULL, p = NULL, deriv = NULL,
h = NULL, b = NULL, rho = 1, kernel = "epa", bwselect = NULL,
bwcheck = 21, bwregul = 1, imsegrid = 30, vce = "nn", covgrid = FALSE,
cluster = NULL, nmatch = 3, level = 95, interior = FALSE, subset = NULL)
```

**Arguments**

<code>y</code>	dependent variable.
<code>x</code>	independent variable.
<code>eval</code>	vector of evaluation point(s). By default it uses 30 equally spaced points over to support of <code>x</code> .
<code>neval</code>	number of quantile-spaced evaluation points on support of <code>x</code> . Default is <code>neval=30</code> .
<code>p</code>	polynomial order used to construct point estimator; default is <code>p = 1</code> (local linear regression).
<code>deriv</code>	derivative order of the regression function to be estimated. Default is <code>deriv=0</code> (regression function).
<code>h</code>	main bandwidth used to construct local polynomial point estimator. Can be either scalar (same bandwidth for all evaluation points), or vector of same dimension as <code>eval</code> . If not specified, bandwidth <code>h</code> is computed by the companion command <code>lpbwselect</code> .
<code>b</code>	bias bandwidth used to construct the bias-correction estimator. Can be either scalar (same bandwidth for all evaluation points), or vector of same dimension as <code>eval</code> . By default it is set equal to <code>h</code> . If <code>rho</code> is set to zero, <code>b</code> is computed by the companion command <code>lpbwselect</code> .
<code>rho</code>	Sets <code>b=h/rho</code> . Default is <code>rho = 1</code> .
<code>kernel</code>	kernel function used to construct local polynomial estimators. Options are <code>epa</code> for the epanechnikov kernel, <code>tri</code> for the triangular kernel, <code>uni</code> for the uniform kernel and <code>gau</code> for the gaussian kernel. Default is <code>kernel = epa</code> .
<code>bwselect</code>	bandwidth selection procedure to be used via <code>lpbwselect</code> . By default it computes <code>h</code> and sets <code>b=h/rho</code> (with <code>rho=1</code> by default). It computes both <code>h</code> and <code>b</code> if <code>rho</code> is set equal to zero. Options are:

	mse-dpi second-generation DPI implementation of MSE-optimal bandwidth. Default option if only one evaluation point is chosen.
	mse-rot ROT implementation of MSE-optimal bandwidth.
	imse-dpi second-generation DPI implementation of IMSE-optimal bandwidth (computed using a grid of evaluation points). Default option if more than one evaluation point is chosen.
	imse-rot ROT implementation of IMSE-optimal bandwidth (computed using a grid of evaluation points).
	ce-dpi second generation DPI implementation of CE-optimal bandwidth.
	ce-rot ROT implementation of CE-optimal bandwidth.
	all reports all available bandwidth selection procedures.
	Note: MSE = Mean Square Error; IMSE = Integrated Mean Squared Error; CE = Coverage Error; DPI = Direct Plug-in; ROT = Rule-of-Thumb. For details on implementation see Calonico, Cattaneo and Farrell (2019).
bwcheck	if a positive integer is provided, then the selected bandwidth is enlarged so that at least bwcheck effective observations are available at each evaluation point. Default is bwcheck = 21.
bwregul	specifies scaling factor for the regularization term added to the denominator of bandwidth selectors. Setting bwregul = 0 removes the regularization term from the bandwidth selectors. Default is bwregul = 1.
imsegrid	number of evaluations points used to compute the IMSE bandwidth selector. Default is imsegrid = 30.
vce	procedure used to compute the variance-covariance matrix estimator. Options are: nn heteroskedasticity-robust nearest neighbor variance estimator with nnmatch the (minimum) number of neighbors to be used. Default choice. hc0 heteroskedasticity-robust plug-in residuals variance estimator without weights. hc1 heteroskedasticity-robust plug-in residuals variance estimator with hc1 weights. hc2 heteroskedasticity-robust plug-in residuals variance estimator with hc2 weights. hc3 heteroskedasticity-robust plug-in residuals variance estimator with hc3 weights.
covgrid	if TRUE, it computes two covariance matrices (cov.us and cov.rb) for classical and robust covariances across point estimators over the grid of evaluation points.
cluster	indicates the cluster ID variable used for cluster-robust variance estimation with degrees-of-freedom weights. By default it is combined with vce=nn for cluster-robust nearest neighbor variance estimation. Another option is plug-in residuals combined with vce=hc1.
nnmatch	to be combined with for vce=nn for heteroskedasticity-robust nearest neighbor variance estimator with nnmatch indicating the minimum number of neighbors to be used. Default is nnmatch=3
.	
level	confidence level used for confidence intervals; default is level = 95.
interior	if TRUE, all evaluation points are assumed to be interior points. This option affects only data-driven bandwidth selection via <a href="#">lpbwselect</a> . Default is interior = FALSE.
subset	optional rule specifying a subset of observations to be used.

**Value**

Estimate      A matrix containing eval (grid points), h, b (bandwidths), N (effective sample sizes), m.us (point estimates with p-th order local polynomial), tau.bc (bias corrected point estimates with (p+1)-th order local polynomial, se.us (standard error corresponding to tau.us), and se.rb (robust standard error).

opt            A list containing options passed to the function.

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**References**

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Wand, M., and Jones, M. 1995. *Kernel Smoothing*, Florida: Chapman & Hall/CRC.

**See Also**

[lpbwselect](#)

**Examples**

```
x <- runif(500)
y <- sin(4*x) + rnorm(500)
est <- lprobust(y,x)
summary(est)
```

nprobust.plot

*Graphical Presentation of Results from nprobust Package.***Description**

nprobust.plot plots estimated density and regression function using the nprobust package. A detailed introduction to this command is given in Calonico, Cattaneo and Farrell (2019).

Companion commands: [lprobust](#) for local polynomial point estimation and inference procedures, and [kdrobust](#) for kernel density point estimation and inference procedures.

For more details, and related Stata and R packages useful for empirical analysis, visit <https://nppackages.github.io/>.

**Usage**

```
nprobust.plot(..., alpha = NULL, type = NULL, CItyp = NULL,
  title = "", xlabel = "", ylabel = "", lty = NULL, lwd = NULL,
  lcol = NULL, pty = NULL, pwd = NULL, pcol = NULL, Cishade = NULL,
  Cicol = NULL, legendTitle = NULL, legendGroups = NULL)
```

**Arguments**

...	Objects returned by <a href="#">kdrobust</a> or <a href="#">lprobust</a> .
alpha	Numeric scalar between 0 and 1, the significance level for plotting confidence regions. If more than one is provided, they will be applied to data series accordingly.
type	String, one of "line" (default), "points" or "both", how the point estimates are plotted. If more than one is provided, they will be applied to data series accordingly.
CItyp	String, one of "region" (shaded region, default), "line" (dashed lines), "ebar" (error bars), "all" (all of the previous) or "none" (no confidence region), how the confidence region should be plotted. If more than one is provided, they will be applied to data series accordingly.
title, xlabel, ylabel	Strings, title of the plot and labels for x- and y-axis.
lty	Line type for point estimates, only effective if type is "line" or "both". 1 for solid line, 2 for dashed line, 3 for dotted line. For other options, see the instructions for <a href="#">ggplot2</a> or <a href="#">par</a> . If more than one is provided, they will be applied to data series accordingly.
lwd	Line width for point estimates, only effective if type is "line" or "both". Should be strictly positive. For other options, see the instructions for <a href="#">ggplot2</a> or <a href="#">par</a> . If more than one is provided, they will be applied to data series accordingly.
lcol	Line color for point estimates, only effective if type is "line" or "both". 1 for black, 2 for red, 3 for green, 4 for blue. For other options, see the instructions for <a href="#">ggplot2</a> or <a href="#">par</a> . If more than one is provided, they will be applied to data series accordingly.

pty	Scatter plot type for point estimates, only effective if type is "points" or "both". For options, see the instructions for <a href="#">ggplot2</a> or <a href="#">par</a> . If more than one is provided, they will be applied to data series accordingly.
pwd	Scatter plot size for point estimates, only effective if type is "points" or "both". Should be strictly positive. If more than one is provided, they will be applied to data series accordingly.
pcol	Scatter plot color for point estimates, only effective if type is "points" or "both". 1 for black, 2 for red, 3 for green, 4 for blue. For other options, see the instructions for <a href="#">ggplot2</a> or <a href="#">par</a> . If more than one is provided, they will be applied to data series accordingly.
CIshade	Numeric, opaqueness of the confidence region, should be between 0 (transparent) and 1. Default is 0.2. If more than one is provided, they will be applied to data series accordingly.
CIcol	color for confidence region. 1 for black, 2 for red, 3 for green, 4 for blue. For other options, see the instructions for <a href="#">ggplot2</a> or <a href="#">par</a> . If more than one is provided, they will be applied to data series accordingly.
legendTitle	String, title of legend.
legendGroups	String Vector, group names used in legend.

### Details

Companion command: [lprobust](#) for local polynomial-based regression functions and derivatives estimation.

### Value

A standard [ggplot2](#) object is returned, hence can be used for further customization.

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### References

Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2019. [nprobust: Nonparametric Kernel-Based Estimation and Robust Bias-Corrected Inference](#). *Journal of Statistical Software*, 91(8): 1-33. doi: [10.18637/jss.v091.i08](https://doi.org/10.18637/jss.v091.i08).

### See Also

[lprobust](#), [kdrobust](#), [ggplot2](#)

**Examples**

```
x <- runif(500)
y <- sin(4*x) + rnorm(500)
est <- lprobust(y,x)
nprobust.plot(est)
```

# Index

## \* LPR

kdrobust, 5

lprobust, 9

## \* Robust Estimation

kdrobust, 5

lprobust, 9

ggplot2, 13, 14

kdbwselect, 2, 3, 3, 5, 7

kdrobust, 2–5, 5, 13, 14

lpbwselect, 2, 5, 7, 7, 8, 10–12

lprobust, 2, 7, 9, 9, 10, 13, 14

nprobust (nprobust-package), 2

nprobust-package, 2

nprobust.plot, 2, 5, 10, 13

par, 13, 14