

Title

lpdensity — Local Polynomial Density Estimation and Inference.

Syntax

Description

lpdensity implements the local polynomial regression based density (and
 derivatives) estimator proposed in <u>Cattaneo</u>, <u>Jansson and Ma (2020)</u>. Robust
 bias-corrected inference, both pointwise (confidence intervals) and uniform
 (confidence bands) are also implemented following the results in <u>Cattaneo</u>,
 <u>Jansson and Ma (2020)</u> and <u>Cattaneo</u>, <u>Jansson and Ma (2023)</u>. See <u>Cattaneo</u>,
 <u>Jansson and Ma (2022)</u> for more implementation details and illustrations.

Companion command: <a href="https://linear.com/linear.c

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

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

https://nppackages.github.io/

Options

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Estimation
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- $\operatorname{\mathtt{grid}}(var)$ specifies the grid on which density is estimated. When set to default, grid points will be chosen as 0.05-0.95 percentiles of the data, with 0.05 step size.
- bw(var or #) specifies the bandwidth (either a variable containing bandwidth for each grid point or a single number) used for estimation. When omitted, bandwidth will be computed by method specified in bwselect(BwMethod).
- \mathbf{p} (#) specifies the local polynomial order for constructing point estimates. Default is \mathbf{p} (2) (local quadratic regression).
- ${\bf q}(\#)$ specifies the local polynomial order for constructing confidence intervals/bands (a.k.a. the bias correction order). Default is ${\bf p}(\#)+1$. When specified the same as ${\bf p}(\#)$, no bias correction will be performed. Otherwise it should be strictly larger than ${\bf p}(\#)$.
- \mathbf{v} (#) specifies the derivative of distribution function to be estimated. \mathbf{v} (0) for the distribution function, \mathbf{v} (1) (default) for the density funtion, etc.

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kernel(KernelFn) specifies the kernel function used to construct the local-polynomial estimator(s). 

triangular K(u) = (1 - |u|) * (|u| <= 1). This is the default option. 

epanechnikov K(u) = 0.75 * (1 - u^2) * (|u| <= 1). 

uniform K(u) = 0.5 * (|u| <= 1).
```

 $\underline{\text{nomass}}$ points will not adjust point estimates or standard errors even if there are mass points in the data.

Bandwidth Selection

mse-dpi mean squared error optimal bandwidth for each grid point. This is the
 default option.

 ${\tt imse-dpi}$ integrated mean squared error optimal bandwidth which is common for all grid points.

mse-rot rule-of-thumb bandwidth based on a Gaussian reference model.
imse-rot integrated rule-of-thumb bandwidth based on a Gaussian reference
model which is common for all grid points.

nlocalmin(#) specifies the minimum number of observations in each local
neighborhood. This option will be ignored if set to 0, or if noregularize is
used. The default value is 20+p(#)+1.

nuniquemin(#) specifies the minimum number of unique observations in each local neighborhood. This option will be ignored if set to 0, or if noregularize is used. The default value is 20+p(#)+1.

noregularize suppresses local sample size checking.

nostdvar will not standardize the data for bandwidth selection. Note that this may lead to unstable performance of the numerical optimization procedure.

☐ Weights

cweights (Var) specifies weights used for counterfactual distribution construction.

pweights (Var) specifies weights used in sampling. Should be nonnegative.

Storing and displaying results

genvars(NewVarName) specifies if new varaibles should be generated to store
 estimation results. If NewVarName is provided, the following new varaibles
 will be generated:

NewVarName_grid grid points,

NewVarName_bw bandwidth,

NewVarName_nh local/effective sample sizes,

 $NewVarName_f_p$ and $NewVarName_se_p$ point estimates with polynomial order p(#) and the corresponding standard errors,

NewVarName_f_q and NewVarName_se_q point estimates with polynomial order q(#)
and the corresponding standard errors, only available if different from
p(#).

 ${\it NewVarName_CI_1} \ {\it and} \ {\it NewVarName_CI_r} \ {\it confidence intervals/bands}.$

 $\underline{\mathbf{rgrid}}(var)$ specifies a set of grid points to display the results. When omitted, this will be the same as $\underline{\mathbf{grid}}(Var)$.

 $\underline{rindex}(var)$ specifies a set of indices to display the results. This option will be ignored if $\underline{rgrid}(Var)$ is provided.

level(#) controls the level of the confidence interval, and should be between 0
and 100. Default is level(95).

ciuiform computes a uniform confidence band instead of pointwise confidence
intervals.

- cisimul(#) specifies the number of simulations used to construct critical values.
 Default is cisimul(2000). This option will be ignored unless ciuniform is
 provided.
- separator(#) draw a seperation line after every # variables; default is separator(5).

```
Plotting
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plot if specified, point estimates and confidence intervals will be plotted.

estype(ESOpts) specifies the plotting style of point estimates.

line a curve. This is the default option.

points individual points.

both both of the above.

none will not plot point estimates.

- espoint_options(ESPointOpts) specifies additional twoway scatter options for plotting point estimates.
- citype(CIOpts) specifies the plotting style of confidence intervals/bands.

region shaded region. This is the default option.

line upper and lower bounds.

ebar error bars.

all of the above.

none will not plot confidence intervals/bands.

- ciregion_options(CIRegionOpts) specifies additional twoway rarea options for plotting confidence intervals/regions.

- histgram if specified, a histogram will be included in the background.
- hiplot_options(HistOpts) specifies additional twoway histogram options for the histogram.
- ${\tt \underline{graph_options}}\,({\it GraphOpts})$ specifies additional options for plotting, such as legends and labels.

Remarks

Bias correction is only used for the construction of confidence intervals/bands, but not for point estimation. The point estimates, denoted by f_p, are constructed using local polynomial estimates of order $\mathbf{p}(\#)$, while the centering of the confidence intervals/bands, denoted by f_q, are constructed using local polynomial estimates of order $\mathbf{q}(\#)$. The confidence intervals/bands take the form: $[f_q - cv * SE(f_q) , f_q + cv * SE(f_q)]$, where cv denotes the appropriate critical value and $SE(f_q)$ denotes an standard error estimate for the centering of the confidence interval/band. As a result, the confidence intervals/bands may not be centered at the point estimates because they have been bias-corrected. Setting $\mathbf{q}(\#)$ and $\mathbf{p}(\#)$ to be equal results on centered at the point estimate confidence intervals/bands, but requires undersmoothing for valid inference (i.e., (I)MSE-optimal bandwdith for the density point estimator cannot be used). Hence the bandwidth would need to be specified manually when $\mathbf{q}(\#) = \mathbf{p}(\#)$, and the point estimates will not be (I)MSE optimal. See Cattaneo, Jansson and Ma (2020, 2023) for details, and also Calonico, Cattaneo, and Farrell (2018, 2022) for robust bias correction methods.

Sometimes the density point estimates may lie outside of the confidence intervals/bands, which can happen if the underlying distribution exhibits high curvature at some evaluation point(s). One possible solution in this case is to increase the polynomial order p(#) or to employ a smaller bandwidth.

Examples

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Generate artifitial data:
        . set obs 2000
         . set seed 42
         . gen lpd_data = rnormal()
    Density estimation at empirical quantiles:
         . lpdensity lpd_data
    Density estimation at empirical quantiles with the IMSE-optimal bandwidth: . lpdensity lpd_data, bwselect(imse-dpi)
    Density estimation on a fixed grid (0.1, 0.2, ..., 1):
         . gen lpd_grid = _n / 10 if _n <= 10
. lpdensity lpd_data, grid(lpd_grid)</pre>
    Report uniform confidence bands (instead of pointwise confidence intervals):
         . lpdensity lpd_data, ciuniform
         . lpdensity lpd_data, ciuniform level(99)
    Save estimation results to new variables:
         . capture drop temp_*
         . lpdensity lpd_data, genvars(temp)
    Density plot:
        . Ipdensity lpd_data, plot
. lpdensity lpd_data, plot histogram
. lpdensity lpd_data, plot histogram ciuniform level(90)
Saved results
    lpdensity saves the following in e():
    Scalars
      e (N)
                             sample size
      e (p)
                             option p(#)
                             option q(#)
      e (q)
      e (v)
                             option v(#)
      e(scale)
                            option scale(#)
      e(level)
                            option level(#)
    Macros
      e(bwselect)
                         option bwselect(BwMethod)
      e(kernel)
                            option kernel (KernelFn)
    Matrices
      e(result)
                            estimation result
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.
    Calonico, S., M. D. Cattaneo, and M. H. Farrell. 2022. Coverage Error Optimal
         Confidence Intervals for Local Polynomial Regression.
         Bernoulli 28(4): 2998-3022.
    Cattaneo, M. D., Michael Jansson, and Xinwei Ma. 2020. Simple Local Polynomial
         Density Estimators.
         Journal of the American Statistical Association 115(531): 1449-1455.
    Cattaneo, M. D., Michael Jansson, and Xinwei Ma. 2022. <a href="light-red"><u>lpdensity: Local</u></a>
         Polynomial Density Estimation and Inference.
         Journal of Statistical Software 101(2): 1-25.
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Cattaneo, M. D., Michael Jansson, and Xinwei Ma. 2023. <u>Local Regression Distribution Estimators</u>. *Journal of Econometrics*, forthcoming.

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