

Package ‘BayesDecon’

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Type Package

Title Density Deconvolution Using Bayesian Semiparametric Methods

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Description Estimates the density of a variable in a measurement error setup, potentially with an excess of zero values. For more details see Sarkar (2022) <[doi:10.1080/01621459.2020.1782220](https://doi.org/10.1080/01621459.2020.1782220)>.

License GPL (>= 2)

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NeedsCompilation yes

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autoplot.bdeconv_result
Visualization of bdeconv_result object using ggplot2

Description

Visualization of bdeconv_result object using ggplot2

Usage

```
autoplot.bdeconv_result(object)
```

Arguments

object A bdeconv_result object.

Value

A collection of plots illustrating various densities and related functions made using ggplot2.

BayesDecon_data_simulated
Simulated dataset for BayesDecon

Description

A small example dataset used to demonstrate bayesdecon functions. There are 1000 observations each with 3 replicates for 3 variables V1, V2, V3. The first column represents the observations.

Usage

```
BayesDecon_data_simulated
```

Format

A data frame with 3000 rows and 4 variables.

Source

Generated for package examples.

bdeconv

*Obtaining MCMC samples from deconvoluted density***Description**

Obtaining MCMC samples from deconvoluted density

Usage

```
bdeconv(
  obj,
  lwr = 0,
  upr = NULL,
  wgts = NULL,
  regular_var = NULL,
  res_type = c("final", "mean", "all"),
  err_type = c("flex", "norm", "lapl"),
  var_type = c("heteroscedastic", "homoscedastic"),
  na_rm = FALSE,
  control = list(),
  progress = FALSE,
  core_num = 1,
  parallel = FALSE
)
```

Arguments

obj	A data frame where the first column is subject labels.
lwr	Unused.
upr	The ratio of an upper bound for the deconvoluted estimates to the upper bound of the surrogates. It can be either a scalar or a vector with a length that matches the number of numeric columns in obj.
wgts	Unused.
regular_var	Specifies which variables to consider as regular; all others are treated as zero inflated.
res_type	The type of result returned from the MCMC sampler.
err_type	The shape of the error distribution used for deconvolution.
var_type	whether to use homogeneous or heterogeneous error distribution.
na_rm	A logical value indicating whether to ignore rows which contain NA values in numeric columns.
control	A list of arguments specifying prior hyper-parameters and parameters of the MCMC sampler.
progress	Whether a progress bar for MCMC will be shown or not. Default is FALSE.
core_num	If parallelized, number of cores to be used.
parallel	For multivariate data. Whether the pre-processing of each variable will be parallelized or not. Default is FALSE.

Details

All hyper-parameters of the model and variables necessary for the MCMC sampler can be specified through a list via the control argument; some of the most important ones are listed.

- `nsamp`: Number of finally retained MCMC samples after burn-in and thinning. Default is 1000.
- `nburn`: Burn-in period. Default is 2000.
- `nthin`: Thinning interval. Default is 5.
- `niter_uni`: For multivariate data, number of iteration for pre-processing of each variable. Default is 500.
- `K_t`: Number of knot points for bspline. Default is 10.
- `K_x`: Number of mixture components in the model for the density of X when the surrogates are strictly continuous. Not applicable when the surrogates are zero-inflated in which case the density is modeled using normalized bsplines. Default is 10.
- `K_epsilon`: Number of mixture components in the model for the density of scaled errors when applicable. Default is 10.
- `sigmasq_epsilon`: variance parameter for the normal prior of the scaled error mixture components' location parameter. Default is 4.
- `a_epsilon`: shape parameter for the inverse gamma prior of the variances of scaled error mixture components. Default is 3.
- `b_epsilon`: rate parameter for the inverse gamma prior of the variances of scaled error mixture components. Default is 2.
- `a_vartheta`: shape parameter for the inverse gamma prior of the variances of bspline coefficients corresponding to the variance function. Default is 100.
- `b_vartheta`: rate parameter for the inverse gamma prior of the variances of bspline coefficients corresponding to the variance function. Default is 1.
- `a_xi`: shape parameter for the inverse gamma prior of the variances of bspline coefficients corresponding to the densities of episodic components. Default is 100.
- `b_xi`: rate parameter for the inverse gamma prior of the variances of bspline coefficients corresponding to the densities of episodic components. Default is 10.
- `a_beta`: shape parameter for the inverse gamma prior of the variances of bspline coefficients corresponding to the probability of consumption function. Default is 100.
- `b_beta`: rate parameter for the inverse gamma prior of the variances of bspline coefficients corresponding to the probability of consumption function. Default is 1.

Value

A `bdeconv_result` object, containing all the posterior samples, suitable for further analysis and visualization.

Examples

```

data(BayesDecon_data_simulated)
set.seed(123)
### Selecting first 300 rows for demonstration.
### For best results, using the whole data is recommended.
result_uni <- bdeconv(BayesDecon_data_simulated[1:300,c(1,2)],
  upr = 1, res_type = "all",
  control = list(nsamp = 500, nburn = 0, nthin = 1),
  progress = TRUE)
plot(result_uni)

set.seed(123)
result_mult <- bdeconv(BayesDecon_data_simulated, upr = 1, res_type = "all", progress = TRUE)
plot(result_mult)
plot(result_mult[,c(1,2)])
value = rbind(c(1,1,1),c(2,2,2),c(3,3,3),c(4,4,4))
colMeans(dist_x(result_mult, vals = value))
colMeans(dens_x(result_mult, vals = value))
print(result_mult)

```

dens_x

*Estimated density function for the true variable(s)***Description**

Estimated density function for the true variable(s)

Usage

```
dens_x(obj, vals, expand = FALSE)
```

Arguments

obj	A <code>bdeconv_result</code> object.
vals	A numeric vector (for univariate) or matrix (for multivariate) on which to evaluate the density.
expand	Used only for <code>mv_res</code> objects. A logical value indicating whether or not to expand the matrix of values.

Value

A numeric matrix or array of density values for which each row represents the estimated density corresponding to each MCMC iteration.

dist_x	<i>Estimated cumulative distribution function for the true variable(s)</i>
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Description

Estimated cumulative distribution function for the true variable(s)

Usage

```
dist_x(obj, vals, expand = FALSE)
```

Arguments

obj	A bdeconv_result object.
vals	A numeric vector (for univariate) or matrix (for multivariate) on which to evaluate the CDF.
expand	Used only for mv_res objects. A logical value indicating whether or not to expand the matrix of values.

Value

A numeric matrix or array of cumulative distribution values for which each row represents the estimated CDF corresponding to each MCMC iteration.

mean.bdeconv_result	<i>Computing mean of MCMC samples</i>
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Description

Computing mean of MCMC samples

Usage

```
## S3 method for class 'bdeconv_result'
mean(x, ...)
```

Arguments

x	A bdeconv_result object.
...	Unused.

Details

Where a direct point-wise mean is not appropriate (i.e. for mixture density results), mixtures are expanded to included every component from every sampled density with weights divided by n, the number of samples from which the mean is being taken.

Value

A bdeconv_result object with one sample constituting the mean

plot.bdeconv_result *Visualization of bdeconv_result object*

Description

Visualization of bdeconv_result object

Usage

```
## S3 method for class 'bdeconv_result'
plot(x, ..., use_ggplot = TRUE)
```

Arguments

x	A bdeconv_result object.
...	Unused.
use_ggplot	A logical value indicating whether the plots will be generated using ggplot2 or using base plot functions.

Details

By default it uses ggplot2 package if it is available in the system library, otherwise it falls back to the base R plot function.

Value

A collection of plots illustrating various densities and related functions.

print.bdeconv_result *Printing summary of MCMC samples for bdeconv_result object*

Description

Printing summary of MCMC samples for bdeconv_result object

Usage

```
## S3 method for class 'bdeconv_result'
print(x, ...)
```

Arguments

x A bdeconv_result object.
 ... Unused.

Details

For episodic components, this gives five number summary for the observed surrogates (y), the latent surrogates corresponding to the consumption days (w), the true intake patterns (x) and the errors (u). For regular components, this gives five number summary for y, x and u only.

Value

Prints the five number summary for the relevant variables.

subset_result *Subset Methods for bdeconv_result object*

Description

Subset Methods for bdeconv_result object

Usage

```
## S3 method for class 'zi_res'
obj[i]

## S3 method for class 'nc_res'
obj[i]

## S3 method for class 'mv_res'
obj[i, j]
```

Arguments

obj A bdeconv_result object.
 i An integer row index (or indices) to select some specific MCMC iterations.
 j An integer column index (or indices) to select some specific set of variables
 (For multivariate case only).

Value

A bdeconv_result object containing the selected iterations of posterior samples for the specified variables (in the multivariate case), suitable for further analysis and visualization.

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