

# Package ‘DrBats’

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**Type** Package

**Title** Data Representation: Bayesian Approach That's Sparse

**Version** 0.1.6

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**Description** Feed longitudinal data into a Bayesian Latent Factor Model to obtain a low-rank representation. Parameters are estimated using a Hamiltonian Monte Carlo algorithm with STAN. See G. Weinrott, B. Fontez, N. Hilgert and S. Holmes, ``Bayesian Latent Factor Model for Functional Data Analysis'', Actes des JdS 2016.

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**Imports** ade4, coda, MASS, Matrix, sde

**License** GPL-3

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**NeedsCompilation** no

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coda.obj	<i>Convert a STAN objet to MCMC list</i>
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### Description

Convert a STAN objet to MCMC list

### Usage

```
coda.obj(stanfit)
```

### Arguments

```
stanfit      a STAN object
```

### Value

codafit an mcmc.list

### Author(s)

Gabrielle Weinrott

### Examples

```
data(stanfit) # output of modelFit or main.modelFit
coda.fit <- coda.obj(stanfit)
head(coda.fit)
```

---

coinertia.drbats	<i>Perform Coinertia Analysis on the PCA of the Weighted PCA and Deville's PCA</i>
------------------	--

---

## Description

Perform Coinertia Analysis on the PCA of the Weighted PCA and Deville's PCA

## Usage

```
coinertia.drbats(  
  X.histo = NULL,  
  Qp = NULL,  
  X = NULL,  
  t = NULL,  
  t.range = c(0, 1000),  
  breaks  
)
```

## Arguments

X.histo	the data matrix projected onto the histogram basis
Qp	a matrix of weights, if Qp = NULL the function specifies a diagonal weight matrix
X	a data matrix, if X.histo is NULL and needs to be built
t	a matrix of observation times, if X.histo is NULL and needs to be built
t.range	the range of observation times in vector form, if X.histo is NULL and needs to be built (default: t.range = c(0, 1000))
breaks	integer number of histogram windows

## Value

co\_weight the co-inertia object

## Author(s)

Gabrielle Weinrott

## Examples

```
res <- drbats.simul(N = 5, P = 100, t.range = c(5, 100), breaks = 8)  
res.coinertia <- coinertia.drbats(X = res$X, t = res$t.simul, t.range = c(5, 100), breaks = 8)  
res.coinertia
```

drbats.simul

*Main simulation function***Description**

Main simulation function

**Usage**

```
drbats.simul(
  N = 10,
  P = 150,
  t.range = c(0, 1000),
  b.range = c(0.2, 0.4),
  c.range = c(0.6, 0.8),
  b.sd = 2,
  c.sd = 2,
  a.range = c(-0.4, 0.4),
  y.range = c(0, 10),
  amp = 10,
  per = 12,
  data.type = "sparse",
  breaks = 15,
  sigma2 = 0.2,
  seed = NULL
)
```

**Arguments**

N	integer number of functions to simulate (default = 10)
P	a number of observation times (default = 150)
t.range	a range of times in which to place the P observations (default = c(1, 1000))
b.range	a vector giving the range of values for the mean of the first mode (default b.range = c(0.2, 0.4))
c.range	a vector giving the range of values for the mean of the second mode (default c.range = c(0.6, 0.8))
b.sd	the standard deviation for the first mode (default b.sd = 2)
c.sd	the standard deviation for the second mode (default c.sd = 2)
a.range	a vector giving the range of values for the slope (default a.range = c(-0.4, 0.4))
y.range	a vector giving the range of values for the intercept (default y.range = c(0, 10))
amp	the amplitude of the cosine function (default = 10)
per	the periodicity of the cosine function (default = 12)
data.type	string indicating type of functions (options :sparse, sparse.tend, sparse.tend.cos)

breaks            number of breaks in the histogram basis  
 sigma2           the precision of the error terms (default = 0.2)  
 seed             integer specification of a seed (default = NULL)

**Value**

Y.simul a list containing a matrix Y, a matrix beta, and a matrix epsilon  
 t.simul a matrix of simulated observation times  
 X the underlying signal to build the data, see DataSimulationandProjection vignette  
 proj.pca the outputs of the function pca.proj.Xt  
 wlu the outputs of the function W.QR

**Author(s)**

Gabrielle Weinrott

**Examples**

```
res <- drbats.simul(N = 5, P = 100, t.range = c(5, 100), breaks = 8)
X <- res$X
t <- res$t.simul
# To plot the observations, ie the rows
matplot(t(t), t(X), type = 'l', xlab = "Time", ylab = "X")
```

---

histoProj                            *Project a set of curves onto a histogram basis*

---

**Description**

Project a set of curves onto a histogram basis

**Usage**

```
histoProj(X, t, t.range, breaks)
```

**Arguments**

X                    a matrix  
 t                    a matrix of observation times  
 t.range            a range of times in which to place the P projections (default = c(0, 1000))  
 breaks             the number of intervals in the histogram basis

**Value**

X.proj the matrix X after projection

X.count a matrix containing the number of observations used to build the projection onto the histogram basis

windows a vector containing the first time of each window of the histogram intervals

X.max the matrix of minimum values in each window

X.min the matrix of maximum values in each window

**Author(s)**

Gabrielle Weinrott

**Examples**

```
res <- drbats.simul(N = 5, P = 100, t.range = c(5, 100), breaks = 8)
res.proj <- histoProj(res$X, res$t.simul, t.range = c(5, 100), breaks = 8)
res.proj
```

---

modelFit

*Fit a Bayesian Latent Factor to a data set using STAN*

---

**Description**

Fit a Bayesian Latent Factor to a data set using STAN

**Usage**

```
modelFit(
  model = "PLT",
  var.prior = "IG",
  prog = "stan",
  parallel = TRUE,
  Xhisto = NULL,
  nchains = 4,
  nthin = 10,
  niter = 10000,
  R = NULL
)
```

**Arguments**

model	a string indicating the type of model ("PLT", or "sparse", default = "PLT")
var.prior	the family of priors to use for the variance parameters ("IG" for inverse gamma, or "cauchy")

prog	a string indicating the MCMC program to use (default = "stan")
parallel	true or false, whether or not to parallelize (done using the package "parallel")
Xhisto	matrix of simulated data (projected onto the histogram basis)
nchains	number of chains (default = 2)
nthin	the number of thinned iterations (default = 1)
niter	number of iterations (default = 1e4)
R	rotation matrix of the same dimension as the number of desired latent factors

**Value**

stanfit, a STAN object

**Author(s)**

Gabrielle Weinrott

**References**

The Stan Development Team Stan Modeling Language User's Guide and Reference Manual. <http://mc-stan.org/>

---

pca.Deville

*Perform a PCA using Deville's method*

---

**Description**

Perform a PCA using Deville's method

**Usage**

pca.Deville(X, t, t.range, breaks)

**Arguments**

X	a data matrix
t	a matrix of observation times corresponding to X
t.range	the range of observation times in vector form (ex. t.range = c(0, 1000))
breaks	integer number of histogram windows

**Value**

X.histo the matrix projected onto the histogram basis

U.histo a matrix of eigenvectors in the histogram basis

Cp a matrix of principal components

lambda a vector of eigenvalues

perc.lambda a vector of the percentage of total inertia explained by each principal component

**Author(s)**

Gabrielle Weinrott

**References**

JC Deville, "Methodes statistiques et numeriques de l'analyse harmonique", Annales de l'INSEE, 1974.

**Examples**

```
res <- drbats.simul(N = 5, P = 100, t.range = c(5, 100), breaks = 8)
res.pca <- pca.Deville(res$X, res$t.simul, t.range = c(5, 100), breaks = 8)
res.pca
```

---

pca.proj.Xt

*PCA data projected onto a histogram basis*

---

**Description**

PCA data projected onto a histogram basis

**Usage**

```
pca.proj.Xt(X, t, t.range = c(0, 1000), breaks = 15)
```

**Arguments**

X	the data matrix
t	the matrix of observation times
t.range	a vector specifying the observation time range (default : c(0, 1000))
breaks	the number of breaks in the histogram basis (default : breaks = 15)

**Value**

Xt.proj a matrix of projected observations  
U a matrix of eigenvectors  
lambda a vector of eigenvalues  
lambda.perc the percentage of inertia captured by each axis

**Author(s)**

Gabrielle Weinrott



**Examples**

```
res <- drbats.simul(N = 5, P = 100, t.range = c(5, 100), breaks = 8)
pca.proj.Xt(res$X, res$t.simul, t.range = c(0, 100), breaks = 8)
```

---

postdens

*Calculate the unnormalized posterior density of the model*

---

**Description**

Calculate the unnormalized posterior density of the model

**Usage**

```
postdens(mcmc.output, Y, D, chain = 1)
```

**Arguments**

mcmc.output	an mcmc list as produced by clean.mcmc
Y	the data matrix
D	the number of latent factors
chain	the chain to plot (default = 1)

**Value**

post a vector containing the posterior density at each iteration##' @examples

**Author(s)**

Gabrielle Weinrott

**Examples**

```
data("toydata")
data("stanfit")
dens <- postdens(coda.obj(stanfit), Y = toydata$Y.simul$Y, D = 2, chain = 1)
hist(dens)
```

---

stanfit	<i>A stanfit object fitted to the toydata</i>
---------	---

---

**Description**

A stanfit object fitted to the toydata

**Usage**

```
stanfit
```

**Format**

A large stanfit object

---

toydata	<i>A toy longitudinal data set</i>
---------	------------------------------------

---

**Description**

A toy longitudinal data set

**Usage**

```
toydata
```

**Format**

A list with 5 elements :

**Y.simul** a list of simulated data with 3 elements

**t.simul** a matrix with 5 rows and 150 columns giving the observation times of the original data

**X** the original data matrix with 5 rows and 150 columns

**proj.pca** a list with 4 elements : results of the function `histoProj(X, t, t.range = c(0, 1000), breaks = 8)`

**wlu** a list with 4 elements : results of the function `W.QR(U, lambda)` where U and lambda are the results of the PCA of X

---

visbeta	<i>Format scores output for visualization</i>
---------	---

---

**Description**

Format scores output for visualization

**Usage**

```
visbeta(mcmc.output, Y, D, chain = 1, axes = c(1, 2), quant = NULL)
```

**Arguments**

mcmc.output	an mcmc list as produced by clean.mcmc
Y	the matrix of data
D	the number of latent factors
chain	the chain to use (default = 1)
axes	the axes to use (default = c(1, 2))
quant	a vector of quantiles to retain (default = NULL)

**Value**

mean.df are the MCMC estimates for the parameters

points.df contains all of the estimates of the chain

contour.df contains the exterior points of the convex hull of the cloud of estimates

**Author(s)**

Gabrielle Weinrott

**Examples**

```
data("toydata")
data("stanfit")
codafit <- coda.obj(stanfit) ## convert to mcmc.list
beta.res <- visbeta(codafit, Y = toydata$Y.simul$Y, D = toydata$wlu$D, chain = 1,
axes = c(1, 2), quant = c(0.05, 0.95))

ggplot2::ggplot() +
  ggplot2::geom_path(data = beta.res$contour.df, ggplot2::aes(x = x, y = y, colour = ind)) +
  ggplot2::geom_point(data = beta.res$mean.df, ggplot2::aes(x = x, y = y, colour = ind))
```

visW

*Plot the estimates for the latent factors***Description**

Plot the estimates for the latent factors

**Usage**

```
visW(mcmc.output, Y, D, chain = 1, factors = c(1, 2))
```

**Arguments**

mcmc.output	an mcmc list as produced by clean.mcmc
Y	the matrix of data
D	the number of latent factors
chain	the chain to plot (default = 1)
factors	a vector indicating the factors to plot (default = c(1, 2))

**Value**

res.W a data frame containing the estimates for the factors, and their lower and upper bounds

Inertia the percentage of total inertia captured by each of the factors

**Author(s)**

Gabrielle Weinrott

**Examples**

```
data("toydata")
data("stanfit")
codafit <- coda.obj(stanfit) ## convert to mcmc.list
W.res <- visW(codafit, Y = toydata$Y.simul$Y, D = toydata$wlu$D,
chain = 1, factors = c(1, 2))

## plot the results

data <- data.frame(time = rep(1:9, 2), W.res$res.W)
ggplot2::ggplot() +
  ggplot2::geom_step(data = data, ggplot2::aes(x = time, y = Estimation, colour = Factor)) +
  ggplot2::geom_step(data = data, ggplot2::aes(x = time, y = Lower.est, colour = Factor),
  linetype = "longdash") +
  ggplot2::geom_step(data = data, ggplot2::aes(x = time, y = Upper.est, colour = Factor),
  linetype = "longdash")
```

---

`W.QR`*Build and decompose a low-rank matrix W*

---

**Description**

Build and decompose a low-rank matrix from a matrix of eigenvectors and eigenvalues from principal component analysis

**Usage**

```
W.QR(U, lambda)
```

**Arguments**

U	a matrix of eigenvectors
lambda	a vector of corresponding eigenvalues

**Value**

W a low-rank matrix  
D the number of latent factors  
Q the orthogonal matrix of the  $W = QR$  matrix decomposition  
R the upper triangular matrix of the  $W = QR$  matrix decomposition

**Author(s)**

Gabrielle Weinrott

**Examples**

```
res <- drbats.simul(N = 5, P = 100, t.range = c(5, 100), breaks = 8)
res.pca <- pca.Deville(res$X, res$t.simul, t.range = c(5, 100), breaks = 8)
Wres.pca <- W.QR(res.pca$U, res.pca$lambda)
Wres.pca
```

---

weighted.Deville	<i>Perform a weighted PCA using Deville's method on a data matrix X that we project onto a histogram basis and weighted</i>
------------------	---

---

**Description**

Perform a weighted PCA using Deville's method on a data matrix  $X$  that we project onto a histogram basis and weighted

**Usage**

```
weighted.Deville(X, t, t.range, breaks, Qp = NULL)
```

**Arguments**

$X$	a data matrix
$t$	a matrix of observation times corresponding to $X$
$t.range$	the range of observation times in vector form (ex. $t.range = c(a, b)$ )
$breaks$	integer number of histogram windows
$Qp$	a matrix of weights, if $Qp = NULL$ the function specifies a diagonal weight matrix

**Value**

$X.histo$  the matrix projected onto the histogram basis  
 $U.histo$  a matrix of eigenvectors in the histogram basis  
 $Cp$  a matrix of principal components  
 $lambda$  a vector of eigenvalues  
 $perc.lambda$  a vector of the percentage of total inertia explained by each principal component

**Author(s)**

Gabrielle Weinrott

**Examples**

```
res <- drbats.simul(N = 5, P = 100, t.range = c(5, 100), breaks = 8)
res.weighted <- weighted.Deville(res$X, res$t.simul, t.range = c(5, 100), breaks = 8, Qp = NULL)
res.weighted
```

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