Package 'RZigZag'

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Description Implements the Zig-Zag algorithm (Bierkens, Fearnhead, Roberts, 2016) <arxiv:1607.03188> applied and Bouncy Particle Sampler <arxiv:1510.02451> for a Gaussian target and Student distribution.</arxiv:1510.02451></arxiv:1607.03188>
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Description

Applies the BPS Sampler to a Gaussian target distribution, as detailed in Bouchard-Côté et al, 2017. Assume potential of the form

$$U(x) = (x - mu)^T V(x - mu)/2,$$

i.e. a Gaussian with mean vector mu and covariance matrix inv(V)

Usage

```
BPSGaussian(V, mu, n_iter = -1L, finalTime = -1, x0 = numeric(0), v0 = numeric(0), refresh_rate = 1, unit_velocity = FALSE)
```

Arguments

V	the inverse covariance matrix (or precision matrix) of the Gaussian target distribution.
mu	mean of the Gaussian target distribution
n_iter	Number of algorithm iterations; will result in the equivalent amount of skeleton points in Gaussian case because no rejections are needed.
finalTime	If provided and nonnegative, run the BPS sampler until a trajectory of continuous time length finalTime is obtained (ignoring the value of n_iterations)
x0	starting point (optional, if not specified taken to be the origin)
v0	starting direction (optional, if not specified taken to be a random vector)
refresh_rate	lambda_refresh
unit_velocity	TRUE indicates velocities uniform on unit sphere, FALSE (default) indicates standard normal velocities

Value

Returns a list with the following objects:

Times: Vector of switching times

Positions: Matrix whose columns are locations of switches. The number of columns is identical to the length of skeletonTimes. Be aware that the skeleton points themselves are NOT samples from the target distribution.

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Examples

```
V <- matrix(c(3,1,1,3),nrow=2)
mu <- c(2,2)
x0 <- c(0,0)
result <- BPSGaussian(V, mu, n_iter = 100, x0 = x0)
plot(result$Positions[1,], result$Positions[2,],type='l',asp=1)</pre>
```

BPSIIDGaussian

BPSIIDGaussian

Description

Applies the Bouncy Particle Sampler to a IID Gaussian distribution

Usage

```
BPSIIDGaussian(variance, dim = 1L, n_iter = -1L, finalTime = -1,
    x0 = numeric(0), v0 = numeric(0), refresh_rate = 1,
    unit_velocity = FALSE)
```

Arguments

variance	scalar indicating variance
dim	dimension
n_iter	Number of algorithm iterations; will result in the equivalent amount of skeleton points in Gaussian case because no rejections are needed.
finalTime	If provided and nonnegative, run the sampler until a trajectory of continuous time length final Time is obtained (ignoring the value of n_i)
x0	starting point (optional, if not specified taken to be the origin)
v0	starting direction (optional, if not specified taken to be a random vector)
refresh_rate	lambda_refresh
unit_velocity	TRUE indicates velocities uniform on unit sphere, FALSE (default) indicates standard normal velocities

Value

Returns a list with the following objects:

Times: Vector of switching times

Positions: Matrix whose columns are locations of switches. The number of columns is identical to the length of skeletonTimes. Be aware that the skeleton points themselves are NOT samples from the target distribution.

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Examples

```
result <- BPSIIDGaussian(1, 2, 1000)
plot(result$Positions[2,], result$Positions[1,],type='l',asp=1)
```

BPSStudentT

BPSStudentT

Description

Applies the Zig-Zag Sampler to a Student T distribution (IID or spherically symmetric)

Usage

```
BPSStudentT(dof, dim = 1L, n_iter = -1L, finalTime = -1,
  x0 = numeric(0), v0 = numeric(0), sphericallySymmetric = TRUE,
  refresh_rate = 1, unit_velocity = FALSE)
```

Arguments

dof scalar indicating degrees of freedom dim dimension n_iter Number of algorithm iterations; will result in the equivalent amount of skeleton points in Gaussian case because no rejections are needed. finalTime If provided and nonnegative, run the sampler until a trajectory of continuous time length finalTime is obtained (ignoring the value of n_iterations) starting point (optional, if not specified taken to be the origin) x0 v0 starting direction (optional, if not specified taken to be a random vector) sphericallySymmetric boolean. If false, sample iid Student T distribution, if true (default) sample spherically summetric Student t dsitribution.

refresh_rate lambda_refresh

unit_velocity TRUE indicates velocities uniform on unit sphere, FALSE (default) indicates

standard normal velocities

Value

Returns a list with the following objects:

Times: Vector of switching times

Positions: Matrix whose columns are locations of switches. The number of columns is identical to the length of skeletonTimes. Be aware that the skeleton points themselves are NOT samples from the target distribution.

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Examples

```
dim = 2
dof = 4
result <- BPSStudentT(dof, dim, n_iter=1000,sphericallySymmetric = TRUE)
plot(result$Positions[1,], result$Positions[2,],type='1',asp=1)</pre>
```

DiscreteSamples

DiscreteSamples

Description

Extract discrete samples from a skeleton

Usage

```
DiscreteSamples(skeletonList, n_samples, coordinate = -1L)
```

Arguments

skeletonList a piecewise deterministic skeleton (consisting of Times, Points and Velocities)

returned by a sampler

n_samples number of samples to obtain

coordinate if specified, only obtain samples of the specified coordinate, otherwise obtain

samples of all coordinates

Value

Returns a list containing the extracted samples and the times (on the continuous time scale) at which the samples are extracted

EstimateCovarianceMatrix

EstimateCovarianceMatrix

Description

Estimates the covariance matrix of a piecewise deterministic skeleton

Usage

```
EstimateCovarianceMatrix(skeletonList, coordinate = -1L,
  zeroMeans = FALSE)
```

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Arguments

skeletonList a piecewise deterministic skeleton (consisting of Times, Points and Velocities)

returned by a sampler

coordinate if specified, only estimate the variance of the specified coordinate, otherwise

estimate the covariance matrix of all coordinates

zeroMeans if TRUE do not estimate means but assume a centered distribution

Value

Returns a list containing the estimated moment

Description

Estimates the effective sample size (ESS) of a piecewise deterministic skeleton

Usage

```
EstimateESS(skeletonList, n_batches = 100L, coordinate = -1L,
  zeroMeans = FALSE)
```

Arguments

skeletonList	a piecewise de	eterministic skeleton	(consisting of	Times,	Points and	Velocities)
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returned by a sampler

n_batches optional argument indicating the number of batches to use in the batch means

estimation method

coordinate if specified, only estimate the ESS of the specified coordinate, otherwise esti-

mate the ESS of all coordinates

zeroMeans if TRUE do not estimate means but assume a centered distribution

Value

Returns a list containing the estimated asymptotic variance, ESS and estimated covariance matrix

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Description

Estimates the p-th moment of a piecewise deterministic skeleton

Usage

```
EstimateMoment(skeletonList, p, coordinate = −1L)
```

Arguments

skeletonList a piecewise deterministic skeleton (consisting of Times, Points and Velocities)

returned by a sampler

p moment to estimate

coordinate if specified, only estimate the ESS of the specified coordinate, otherwise esti-

mate the ESS of all coordinates

Value

Returns a list containing the estimated moment

Description

Implements various piecewise deterministic Monte Carlo methods, including the Zig-Zag Sampler (Bierkens, Fearnhead, Roberts, 2019, https://arxiv.org/abs/1607.03188) and the Bouncy Particle Sampler (BPS, Bouchard-Côté et al., 2017, https://arxiv.org/abs/1510.02451). Typical usage consists of first creating a "skeleton" consisting of "events", which can be used directly for plotting trajectories. The skeleton may be post-processed to extract information, such as as moment and covariance estimates, discrete samples at fixed time intervals along the trajectory, effective sample size and asymptotic variance.

Details

This package currently consists of the following functions for generating skeletons: ZigZagLogistic for logistic regression, ZigZagGaussian for multivariate Gaussian, ZigZagIIDGaussian for a IID Gaussian target using Zig-Zag, ZigZagStudentT for spherically symmetric or factorized Student-t distribution, BPSGaussian for multivariate Gaussian using BPS, BPSIIDGaussian for a IID Gaussian target using BPS, BPSStudentT for BPS applied to a spherically symmetric or factorized Student-t distribution. Furthermore the package contains the following functions for post-processing: EstimateESS (to estimate asymptotic variance and effective sample size for individual coordinates), EstimateMoment, EstimateCovarianceMatrix and DiscreteSamples.

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Author(s)

Joris Bierkens

With thanks to Matt Moores, https://mattstats.wordpress.com/, for his help in getting from C++ code to a CRAN-ready Rcpp based package.

ZigZagGaussian

ZigZagGaussian

Description

Applies the Zig-Zag Sampler to a Gaussian target distribution, as detailed in Bierkens, Fearnhead, Roberts, The Zig-Zag Process and Super-Efficient Sampling for Bayesian Analysis of Big Data, 2016. Assume potential of the form

$$U(x) = (x - mu)^T V(x - mu)/2,$$

i.e. a Gaussian with mean vector mu and covariance matrix inv(V)

Usage

```
ZigZagGaussian(V, mu, n_iter = -1L, finalTime = -1, x0 = numeric(0), v0 = numeric(0))
```

Arguments

V	the inverse covariance matrix (or precision matrix) of the Gaussian target distribution.
mu	mean of the Gaussian target distribution
n_iter	Number of algorithm iterations; will result in the equivalent amount of skeleton points in Gaussian case because no rejections are needed.
finalTime	If provided and nonnegative, run the sampler until a trajectory of continuous time length finalTime is obtained (ignoring the value of n_iterations)
x0	starting point (optional, if not specified taken to be the origin)
v0	starting direction (optional, if not specified taken to be +1 in every component)

Value

Returns a list with the following objects:

Times: Vector of switching times

Positions: Matrix whose columns are locations of switches. The number of columns is identical to the length of skeletonTimes. Be aware that the skeleton points themselves are NOT samples from the target distribution.

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Examples

```
V <- matrix(c(3,1,1,3),nrow=2)
mu <- c(2,2)
result <- ZigZagGaussian(V, mu, 100)
plot(result$Positions[1,], result$Positions[2,],type='l',asp=1)</pre>
```

ZigZagIIDGaussian

ZigZagIIDGaussian

Description

Applies the Zig-Zag Sampler to a IID Gaussian distribution

Usage

```
ZigZagIIDGaussian(variance, dim = 1L, n_iter = -1L, finalTime = -1,
    x0 = numeric(0), v0 = numeric(0))
```

Arguments

variance	scalar indicating variance
dim	dimension
n_iter	Number of algorithm iterations; will result in the equivalent amount of skeleton points in Gaussian case because no rejections are needed.
finalTime	If provided and nonnegative, run the sampler until a trajectory of continuous time length finalTime is obtained (ignoring the value of n_iterations)
x0	starting point (optional, if not specified taken to be the origin)
v0	starting direction (optional, if not specified taken to be +1 in every component)

Value

Returns a list with the following objects:

Times: Vector of switching times

Positions: Matrix whose columns are locations of switches. The number of columns is identical to the length of skeletonTimes. Be aware that the skeleton points themselves are NOT samples from the target distribution.

Velocities: Matrix whose columns are velocities just after switches. The number of columns is identical to the length of skeletonTimes.

Examples

```
result <- ZigZagIIDGaussian(1, 2, 1000)
plot(result$Positions[2,], result$Positions[1,],type='l',asp=1)</pre>
```

10 ZigZagLogistic

ZigZagLogistic

Description

Applies the Zig-Zag Sampler to logistic regression, as detailed in Bierkens, Fearnhead, Roberts, The Zig-Zag Process and Super-Efficient Sampling for Bayesian Analysis of Big Data, 2019.

Usage

```
ZigZagLogistic(dataX, dataY, n_iter = -1L, finalTime = -1,
    x0 = numeric(0), v0 = numeric(0), cv = FALSE)
```

Arguments

dataX	Design matrix containing observations of the independent variables x . The i-th row represents the i-th observation with components $x_i, 1,, x_i, d$.
dataY	Vector of length n containing 0, 1-valued observations of the dependent variable y.
n_iter	Number of algorithm iterations; will result in the equivalent amount of skeleton points in Gaussian case because no rejections are needed.
finalTime	If provided and nonnegative, run the sampler until a trajectory of continuous time length finalTime is obtained (ignoring the value of n_iterations)
x0	starting point (optional, if not specified taken to be the origin)
v0	starting direction (optional, if not specified taken to be +1 in every component)
CV	optional boolean to indicate the use of subsampling with control variates

Value

Returns a list with the following objects:

Times: Vector of switching times

Positions: Matrix whose columns are locations of switches. The number of columns is identical to the length of skeletonTimes. Be aware that the skeleton points themselves are NOT samples from the target distribution.

Velocities: Matrix whose columns are velocities just after switches. The number of columns is identical to the length of skeletonTimes.

Examples

```
require("RZigZag")
generate.logistic.data <- function(beta, n.obs) {
  dim <- length(beta)
  dataX <- cbind(rep(1.0,n.obs), matrix(rnorm((dim -1) * n.obs), ncol = dim -1));
  vals <- dataX %*% as.vector(beta)</pre>
```

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```
generateY <- function(p) { rbinom(1, 1, p)}
dataY <- sapply(1/(1 + exp(-vals)), generateY)
    return(list(dataX = dataX, dataY = dataY))
}
beta <- c(1,2)
data <- generate.logistic.data(beta, 1000)
result <- ZigZagLogistic(data$dataX, data$dataY, 1000)
plot(result$Positions[1,], result$Positions[2,],type='l',asp=1)</pre>
```

ZigZagStudentT

ZigZagStudentT

Description

Applies the Zig-Zag Sampler to a Student T distribution (IID or spherically symmetric)

Usage

```
ZigZagStudentT(dof, dim = 1L, n_iter = -1L, finalTime = -1,
    x0 = numeric(0), v0 = numeric(0), sphericallySymmetric = TRUE)
```

Arguments

dof	scalar indicating degrees of freedom
dim	dimension
n_iter	Number of algorithm iterations; will result in the equivalent amount of skeleton points in Gaussian case because no rejections are needed.
finalTime	If provided and nonnegative, run the sampler until a trajectory of continuous time length finalTime is obtained (ignoring the value of n_iterations)
x0	starting point (optional, if not specified taken to be the origin)
v0	starting direction (optional, if not specified taken to be +1 in every component)
sphericallySymmetric	
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boolean. If false, sample iid Student T distribution, if true (default) sample spherically summetric Student t distribution.

Value

Returns a list with the following objects:

Times: Vector of switching times

Positions: Matrix whose columns are locations of switches. The number of columns is identical to the length of skeletonTimes. Be aware that the skeleton points themselves are NOT samples from the target distribution.

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Examples

```
dim = 2
dof = 4
result <- ZigZagStudentT(dof, dim, n_iter=1000, sphericallySymmetric = TRUE)
plot(result$Positions[1,], result$Positions[2,],type='l',asp=1)</pre>
```

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