

Package ‘dsample’

May 11, 2013

Type Package

Title Discretization-based Direct Random Sample Generation

Version 0.91.1

Date 2013-05-11

Author Liqun Wang and Chel Hee Lee

Maintainer Chel Hee Lee <gnustats@gmail.com>

Description The package provides two discretization-based Monte Carlo algorithms, namely the Fu-Wang algorithm and the Wang-Lee algorithm, for random sample generation from a high dimensional distribution of complex structure. The normalizing constant of the target distribution needs not to be known.

License GPL-3

R topics documented:

dsample-package	1
sample.fw	4
sample.wl	5
summary.samplefwl	6
Index	8

dsample-package	<i>Discretization-based Direct Random Sample Generation</i>
-----------------	---

Description

The package provides two discretization-based Monte Carlo algorithms, namely the Fu-Wang algorithm and the Wang-Lee algorithm, for random sample generation from a high dimensional distribution of complex structure. The normalizing constant of the target distribution needs not to be known.

Details

Package: dsample
Type: Package
Version: 0.91.1
Date: 2013-05-11
License: GPL-3

Author(s)

Liqun Wang and Chel Hee Lee

Maintainer: Chel Hee Lee <gnustats@gmail.com>

References

Fu, J. C. and Wang, L. (2002). A random-discretization based Monte Carlo sampling method and its application. *Methodology and Computing in Applied Probability*, 4, 5-25.

Wang, L. and Fu, J. (2007). A practical sampling approach for a bayesian mixture model with unknown number of components. *Statistical Papers*, 48, 631-653.

Wang, L. and Lee, C.H. (2008). A discretization-based efficient random sampling algorithm. Working Paper, Department of Statistics, University of Manitoba.

Examples

```
## The following example is taken from West (1993, page 414).
## West, M. (1993). Approximating posterior distributions by mixture. Journal of the
## Royal Statistical Society - B, 55, 409-422.

## More accurate results can be achieved by increasing the number of dicretization points
## and the number of contours. The default value for the number of discretization points
## is 1e7 and for the number of contours is 1e5

x1 <- runif(1e5)
x2 <- runif(1e5)
val <- (x1*(1-x2))^5 * (x2*(1-x1))^3 * (1-x1*(1-x2)-x2*(1-x1))^37
support <- as.data.frame(cbind(val, x1, x2))

## Applying the Fu-Wang algorithm
out1 <- sample.fw(X=support, nc=1e4, n=1e3)
summary(out1)

## Applying the Wang-Lee algorithm
out2 <- sample.wl(X=support, nc=1e4, n=1e3)
summary(out2)
```

sample.fw

Random Sample Generation Through The Fu-Wang Algorithm

Description

sample.fw generates a sample of size n from the target density function (up to a normalizing constant) using the Fu-Wang algorithm

Usage

```
sample.fw(X, nc = 10000, n = 1000)
```

Arguments

X	a data.frame or a matrix. See ‘Details’.
nc	a positive integer, the number of contours. See ‘Details’.
n	a non-negative integer giving the user specified sample size.

Details

X has the number of rows equals to the number of discrete base points. In each row, the first element contains the functional value of the target density and the rest elements are the coordinates at which the density is evaluated.

Value

sample.fw gives the drawn sample as a data.frame with the number of rows equals to the specified size n and number of columns equals to ncol(X)-1.

Author(s)

Liqun Wang and Chel Hee Lee

References

Fu, J. C. and Wang, L. (2002). A random-discretization based Monte Carlo sampling method and its application. Methodology and Computing in Applied Probability, 4, 5-25.

Wang, L. and Fu, J. (2007). A practical sampling approach for a bayesian mixture model with unknown number of components. Statistical Papers, 48(4):631-653.

Wang, L. and Lee, C.H. (2008). A discretization-based efficient random sampling algorithm (Working paper), Department of Statistics, University of Manitoba, Canada

See Also

[sample.wl](#)

Examples

```
## The following example is taken from West (1993, page 414).
## West, M. (1993). Approximating posterior distributions by mixture. Journal of the Royal
## Statistical Society - B, 55, 409-422.

## More accurate results can be achieved by increasing the number of discretization points
## and the number of contours. The default value for the number of discretization points
## is 1e7 and for the number of contours is 1e5.

x1 <- runif(1e5)
x2 <- runif(1e5)
val <- (x1*(1-x2))^5 * (x2*(1-x1))^3 * (1-x1*(1-x2)-x2*(1-x1))^37
support <- as.data.frame(cbind(val, x1, x2))

summary(sample.fw(X=support, nc=1e4, n=1e3))
```

sample.wl

Random Samples Generation Through The Wang-Lee Algorithm

Description

sample.wl generates a sample of specified size n from the target density function (up to a normalizing constant) based on the Wang-Lee algorithm

Usage

```
sample.wl(X, nc = 10000, n = 1000, wconst = NULL)
```

Arguments

X	a data.frame or a matrix. See ‘Details’.
nc	a positive integer, the number of contours. See ‘Details’.
n	a non-negative integer, the desired sample size.
wconst	a real number between 0 and 1. See ‘Details’.

Details

X has the number of rows equals to the number of discrete base points. In each row, the first element contains the functional value of the target density and the rest elements are the coordinates at which the density is evaluated.

wconst is a constant for adjusting the volume of the last contour.

Value

sample.wl gives the drawn sample as a data.frame with number of rows equals the specified size n and number of columns equals ncol(x)-1.

Note

[sample.fw](#)

Author(s)

Liqun Wang and Chel Hee Lee

References

Fu, J. C. and Wang, L. (2002). A random-discretization based Monte Carlo sampling method and its application. *Methodology and Computing in Applied Probability*, 4, 5-25.

Wang, L. and Fu, J. (2007). A practical sampling approach for a bayesian mixture model with unknown number of components. *Statistical Papers*, 48(4):631-653.

Wang, L. and Lee, C.H. (2008). A discretization-based efficient random sampling algorithm (Working paper), Department of Statistics, University of Manitoba, Canada

Examples

```
## The following example is taken from West (1993, page 414).
## West, M. (1993). Approximating posterior distributions by mixture. Journal of the Royal
## Statistical Society - B, 55, 409-422.
```

```
## More accurate results can be achieved by increasing the number of discretization points
## and the number of contours. The default value of the number of discretization points
## is 1e7 and for the number of contours is 1e5
```

```
x1 <- runif(1e5)
x2 <- runif(1e5)
val <- (x1*(1-x2))^5 * (x2*(1-x1))^3 * (1-x1*(1-x2)-x2*(1-x1))^37
support <- as.data.frame(cbind(val, x1, x2))

summary(sample.fw(X=support, nc=1e4, n=1e3))
```

summary.samplefwl

Generating Basic Summary Statistics of Marginal Distributions

Description

Producing basic summary statistics (the mean, the standard deviation and the first five modes) from the sample drawn via either the Fu-Wang algorithm or the Wang-Lee algorithm, for all marginal distributions of the target distribution.

Usage

```
## S3 method for class 'samplefwl'
summary(object, ...)
```

Arguments

object	a data.frame, contains the sample drawn via either the Fu-Wang algorithm or the Wang-Lee algorithm
...	dot-dot-dot arguments

Author(s)

Liqun Wang and Chel Hee Lee

Examples

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (object, ...)
{
  cat("\nMeans: \n")
  print(round(sapply(object, mean), 4))
  cat("\nStandard deviations: \n")
  print(round(sapply(object, mean), 4))
  cat("\nFirst five modes: \n")
  print(round(object[1:5, ], 4))
}
```

Index

*Topic **package**

dsample-package, [1](#)

dsample (dsample-package), [1](#)

dsample-package, [1](#)

sample.fw, [4](#), [6](#)

sample.wl, [4](#), [5](#)

summary.samplefwl, [6](#)