

Package ‘gaussDiff’

February 19, 2015

Version 1.1

Date 2012-08-21

Title Difference measures for multivariate Gaussian probability density functions

Author Henning Rust <henning.rust@met.fu-berlin.de>

Maintainer Henning Rust <henning.rust@met.fu-berlin.de>

Depends R (>= 1.8.0)

Description A collection difference measures for multivariate Gaussian probability density functions, such as the Euclidean mean, the Mahalanobis distance, the Kullback-Leibler divergence, the J-Coefficient, the Minkowski L2-distance, the Chi-square divergence and the Hellinger Coefficient.

License GPL (>= 2)

URL www.geo.fu-berlin.de/met/ag/clidia/Mitarbeiter/HenningRust/

Repository CRAN

Date/Publication 2012-08-23 06:19:32

NeedsCompilation no

R topics documented:

normdiff 1

Index 4

normdiff *Difference measures for multivariate Gaussian pdfs*

Description

Various difference measures for Gaussian pdfs are implemented: Euclidean distance of the means, Mahalanobis distance, Kullback-Leibler divergence, J-Coefficient, Minkowski L2-distance, Chi-square divergence and the Hellinger coefficient which is a similarity measure.

Usage

```
normdiff(mu1, sigma1=NULL, mu2, sigma2=sigma1, inv=FALSE, s=0.5,
method=c("Mahalanobis", "KL", "J", "Chisq",
"Hellinger", "L2", "Euclidean"))
```

Arguments

mu1	mean value of pdf 1, a vector
sigma1	covariance matrix of pdf 1
mu2	mean value of pdf 2, a vector
sigma2	covariance matrix of pdf 2
method	difference measure to be used, see below
inv	if TRUE, 1-Hellinger is reported, default: inv=FALSE
s	exponent for Hellinger coefficient, default: s=0.5

Details

Equations can be found in H.-H. Bock, *Analysis of Symbolic Data*, Chapter *Dissimilarity Measures for Probability Distributions*

Value

A scalar object of class `normdiff` reporting the distance.

Author(s)

Henning Rust, <henning.rust@met.fu-berlin.de>

References

H.-H. Bock, *Analysis of Symbolic Data*, Chapter *Dissimilarity measures for Probabilistic Distributions*

Examples

```
library(gaussDiff)
mu1 <- c(0,0,0)
sig1 <- diag(c(1,1,1))
mu2 <- c(1,1,1)
sig2 <- diag(c(0.5,0.5,0.5))

## Euclidean distance
normdiff(mu1=mu1, mu2=mu2, method="Euclidean")

## Mahalanobis distance
normdiff(mu1=mu1, sigma1=sig1, mu2=mu2, method="Mahalanobis")

## Kullback-Leibler divergence
normdiff(mu1=mu1, sigma1=sig1, mu2=mu2, sigma2=sig2, method="KL")
```

```
## J-Coefficient
normdiff(mu1=mu1,sigma1=sig1,mu2=mu2,sigma2=sig2,method="J")

## Chi-sqr divergence
normdiff(mu1=mu1,sigma1=sig1,mu2=mu2,sigma2=sig2,method="Chisq")

## Minkowski L2 distance
normdiff(mu1=mu1,sigma1=sig1,mu2=mu2,sigma2=sig2,method="L2")

## Hellinger coefficient
normdiff(mu1=mu1,sigma1=sig1,mu2=mu2,sigma2=sig2,method="Hellinger")
```

Index

*Topic **cluster**

normdiff, 1

*Topic **distribution**

normdiff, 1

*Topic **multivariate**

normdiff, 1

maha (normdiff), 1

normdiff, 1

print.normdiff (normdiff), 1

tt (normdiff), 1