

Package ‘cmvnorm’

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Type Package

Title The Complex Multivariate Gaussian Distribution

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Imports elliptic

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cmvnorm-package

The complex multivariate Gaussian distribution

Description

Various utilities for investigating the complex generalization of the multivariate Gaussian distribution

Details

Package: cmvnorm
Type: Package
Version: 1.0
Date: 2014-01-10
License: GPL-2

Author(s)

Robin K. S.Hankin; much of the code is inspired by the mvtnorm package

Maintainer: <hankin.robin@gmail.com>

References

N. R. Goodman 1963. "Statistical analysis based on a certain multivariate complex Gaussian distribution". *The Annals of Mathematical Statistics*. 34(1): 152–177

Examples

```
Covmat <- matrix(c(1,0.1i,-0.1i,1),2,2)
rcmvnorm(10,sigma=Covmat)
dcmvnorm(emulator::latin.hypercube(5,2,complex=TRUE),sigma=Covmat)
```

corr_complex

Various utilities for complex emulation

Description

Various utilities for investigating complex Gaussian processes

Usage

```
corr_complex(z1, z2 = NULL, distance.function = complex_CF, means =
NULL, scales = NULL, pos.def.matrix = NULL)
complex_CF(z1,z2, means, pos.def.matrix)
scales.likelihood.complex(pos.def.matrix, scales, means, zold, z,
give_log = TRUE, func = regressor.basis)
interpolant.quick.complex(x, d, zold, Ainv, scales = NULL, pos.def.matrix = NULL,
means=NULL, func = regressor.basis, give.Z = FALSE,
distance.function = corr_complex, ...)
```

Arguments

`z, z1, z2` Points in C^n

`distance.function` Function giving the (complex) covariance between two points in C^n

`means, pos.def.matrix, scales` In function `complex_CF()`, the mean and covariance matrix of the distribution whose characteristic function is used as to give the covariance matrix; `scales` is used to specify the diagonal of `pos.def.matrix` if the off-diagonal elements are zero

`zold, d, give_log, func, x, Ainv, give.Z, ...` Direct analogues of the arguments in `interpolant()` and `scales.likelihood()` in the **emulator** package

Details

- Function `complex_CF()` returns a (slightly reparameterized) characteristic function of a complex Gaussian distribution. The covariance is given by

$$c(\mathbf{t}) = \exp(i\text{Re}(\mathbf{t}^* \mu) - \mathbf{t}^* B \mathbf{t})$$

where $\mathbf{t} = \mathbf{x} - \mathbf{x}'$ is interpreted as the distance between two observations, μ is the mean of the distribution (which is in general a complex vector), and B a positive-definite matrix.

- Function `corr_complex()` is the complex analogue of `corr.matrix()`. It returns a matrix with entry (i, j) equal to the covariance of the process at observation i and observation j , or $\text{cov}(\eta(\mathbf{x}_i), \eta(\mathbf{x}_j))$. The elements are calculated by `complex_CF()`.

This function includes only a single method, that of nested calls to `apply()`. I could not figure out how to generalize method 1 of `corr.matrix()` to the complex case.

- Function `scales.likelihood.complex()` is a complex version of `scales.likelihood()` which takes a positive definite matrix and a mean. The formula used is

$$(\sigma^2)^{-(n-q)} |A|^{-1} |H^T A^{-1} H|^{-1}$$

- Function `interpolant.quick.complex()` is a complex version of `interpolant.quick()`.

$$\mathbf{h}(\mathbf{x})^* \hat{\beta} + \mathbf{t}(\mathbf{x})^* A^{-1} (\mathbf{y} - H \hat{\beta})$$

This is the complex version of Oakley's equation 2.30 or Hankin's equation 5.

More details are given in the package vignette.

Author(s)

Robin K. S. Hankin

References

- Hankin, R. K. S. 2005. “Introducing BACCO, an R bundle for Bayesian Analysis of Computer Code Output”, *Journal of Statistical Software*, 14(15)
- J. Oakley 1999. *Bayesian uncertainty analysis for complex computer codes*, PhD thesis, University of Sheffield.

Examples

```
complex_CF(c(1,1i),c(1,-1i),means=c(1i,1i),pos.def.matrix=diag(2))

V <- latin.hypercube(7,2,complex=TRUE)

cm <- c(1,1+1i)           # "complex mean"
cs <- matrix(c(2,1i,-1i,1),2,2) # "complex scales"
tb <- c(1,1i,1-1i)       # "true beta"

A <- corr_complex(V,means=cm,pos.def.matrix=cs)
Ainv <- solve(A)
z <- drop(rcmvnorm(n=1,mean=regressor.multi(V) %*% tb, sigma=A))

betahat.fun(V,Ainv,z)    # should be close to 'tb'

#scales.likelihood.complex(cs,cm,V,z) # log-likelihood evaluated true parameters

interpolant.quick.complex(x=0.1i+V[1:3,],d=z,zold=V,Ainv=Ainv,pos.def.matrix=cs,means=cm)
```

isHermitian

Is a Matrix Hermitian?

Description

Returns TRUE if a matrix is Hermitian or Hermitian positive-definite

Usage

```
isHermitian(x, tol = 100 * .Machine$double.eps)
ishpd(x,tol= 100 * .Machine$double.eps)
zapim(x,tol= 100 * .Machine$double.eps)
```

Arguments

x A square matrix
tol Tolerance for numerical scruff

Details

Functions `isHermitian()` and `ishpd()` return a Boolean. Function `zapim()` zaps small imaginary parts of components vector, returning real if all elements are so zapped.

Author(s)

Robin K. S. Hankin

Examples

```
v <- 2^(1:30)
zapim(v+1i*exp(-v))
```

```
ishpd(matrix(c(1,0.1i,-0.1i,1),2,2)) # should be TRUE
isHermitian(matrix(c(1,3i,-3i,1),2,2)) # should be TRUE
```

Mvnorm

Multivariate complex Gaussian density and random deviates

Description

Density function and a random number generator for the multivariate complex Gaussian distribution.

Usage

```
rcnorm(n)
dcmvnorm(z, mean, sigma, log = FALSE)
rcmvnorm(n, mean = rep(0, nrow(sigma)), sigma = diag(length(mean)),
         method = c("svd", "eigen", "chol"),
         tol= 100 * .Machine$double.eps)
```

Arguments

z Vector or matrix of quantiles. If x is a matrix, each row is taken to be a quantile
n Number of observations
mean Mean vector
sigma Covariance matrix, Hermitian positive-definite

tol	numerical tolerance term for verifying positive definiteness
log	Boolean with default FALSE meaning that the (natural) log of densities is returned
method	Matrix decomposition used to determine the matrix root of sigma, possible methods are eigenvalue decomposition ("eigen", default), and singular value decomposition

Details

Function `dcmvnorm()` is the density function of the complex multivariate normal (Gaussian) distribution:

$$p(\mathbf{z}) = \frac{\exp(-\mathbf{z}^* \Gamma \mathbf{z})}{|\pi \Gamma|}$$

Function `rcnorm()` is a low-level function designed to generate observations drawn from a standard complex Gaussian. Function `rcmvnorm()` is a user-friendly wrapper for this.

Author(s)

Robin K. S. Hankin

References

N. R. Goodman 1963. "Statistical analysis based on a certain multivariate complex Gaussian distribution". *The Annals of Mathematical Statistics*. 34(1): 152–177

Examples

```
S <- emulator::cprod(rcmvnorm(3,mean=c(1,1i),sigma=diag(2)))

rcmvnorm(10,sigma=S)
rcmvnorm(10,mean=c(0,1+10i),sigma=S)

# Now try and estimate the mean (viz 1,1i) and variance (S) from a
# random sample:

n <- 101
z <- rcmvnorm(n,mean=c(0,1+10i),sigma=S)
xbar <- colMeans(z)
Sbar <- cprod(sweep(z,2,xbar))/n
```

setreal *Manipulate real or imaginary components of an object*

Description

Manipulate real or imaginary components of an object

Usage

```
Im(x) <- value  
Re(x) <- value
```

Arguments

x	Complex-valued object
value	Real-valued object

Author(s)

Robin K. S. Hankin

Examples

```
A <- matrix(c(1,0.1i,-0.1i,1),2,2)  
Im(A) <- Im(A)*3  
Re(A) <- matrix(c(5,2,2,5),2,2)
```

var *Variance and standard deviation of complex vectors*

Description

Complex generalizations of `stats::sd()` and `stats::var()`

Usage

```
var(x, y=NULL, na.rm=FALSE, use)  
sd(x, na.rm=FALSE)
```

Arguments

x, y	Complex vector or matrix
na.rm	Boolean with default FALSE meaning to leave NA values present and TRUE meaning to remove them
use	Ignored

Details

Intended to be broadly compatible with `stats::sd()` and `stats::var()`.

If given real values, `var()` and `sd()` return the variance and standard deviation as per ordinary real analysis. If given complex values, returns the complex generalization in which Hermitian transposes are used.

If `z` is a complex matrix, `var(z)` returns the variance of the rows.

These functions use $n - 1$ on the denominator purely for consistency with `stats::var()` (for the record, I disagree with the rationale for $n - 1$).

Author(s)

Robin K. S. Hankin

Examples

```
sd(rcnorm(10)) # imaginary component suppressed by zapim()
var(rcmvnorm(1e5, mean=c(0,0)))
```


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