# Package 'MultiStatM'

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Hermite polynomials; estimation and derivation of theoretical vector moments and vector

**Description** Algorithms to build set partitions and commutator matrices and their use in the

Type Package

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construction of multivariate d-

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Contents
CommutatorIndx

2 CommutatorIndx

	utatorIndx	
Index		40
	VarianceSkew	38
	VarianceKurt	
	UnivMomCum	
	SymMatr	
	SymIndx	35
	SampleVarianceSkewKurt	35
	SampleSkew	
	SampleMomCum	32
	SampleKurt	31
	SampleHermiteN	31
	SampleGC	30
	SampleEVSK	29
	rUniS	29
	rSkewNorm	28
	rCFUSSD	27
	rCFUSN	
	QplicMatr	25
	QplicIndx	24
	PermutationInv	23
	PartitionTypeAll	
	Partitions	21
	MVStandardize	20
	MomCumZabs	19
	MomCumUniS	18
	MomCumSkewNorm	17
	MomCumCFUSN	16
	Mom2Cum	15
	HermiteN2X	14
	HermiteN	13
	HermiteCov12	12
	HermiteCoeff	11
	EVSKUniS	10
	EVSKSkewNorm	Ģ
	EliminMatr	8
	EliminIndx	
	Cum2Mom	$\epsilon$
	CommutatorMatr	_

# Description

This function calculates the commutator index based on the specified type. The available types are "Kmn", "Kperm", "Mixing", and "Moment". Depending on the selected type, the corresponding specific function is called.

CommutatorIndx 3

# Usage

```
CommutatorIndx(Type, ...)
```

# **Arguments**

Type a string specifying the type of commutator index to be calculated. Must be one of "Kmn", "Kperm", "Mixing", or "Moment".

. . . additional arguments passed to the specific commutator function.

#### **Details**

The function 'CommutatorIndx' acts as a wrapper to call specific commutator functions based on the input 'Type'.

Type "Kmn":

**Parameters:** • m - Row-dimension.

• n - Col-dimension.

**Return:** A vector of indexes to provide the commutation, transforming vec A to vec of the transposed A.

Type "Kperm":

Parameters: • perm - Vector indicating the permutation of the order in the Kronecker product.

• dims - Vector indicating the dimensions of the vectors.

**Return:** An index vector to produce the permutation of the Kronecker products of vectors of any length.

Type "Mixing":

**Parameters:** • x - A vector of dimension prod(d1)\*prod(d2).

- d1 Dimension of the first group of vectors.
- d2 Dimension of the second group of vectors.

**Return:** A vector Kx representing the product of the moment commutator and the vector x.

Type "Moment":

**Parameters:** • x - A vector of length d^n where n is the length of el\_rm.

- el\_rm Type of a partition.
- d Dimensionality of the underlying multivariate distribution.

**Return:** A vector Kx representing the product of the moment commutator and the vector x.

#### Value

A vector representing the commutator index.

# See Also

Other Commutators: CommutatorMatr()

4 CommutatorMatr

# **Examples**

```
# Kmn example
A <- 1:6
CommutatorIndx(Type = "Kmn", m = 3, n = 2)
# Kperm example
a1 <- c(1, 2)
a2 <- c(2, 3, 4)
a3 <- c(1, 3)
p1 <- a1 %x% a2 %x% a3
CommutatorIndx(Type = "Kperm", perm = c(3, 1, 2), dims = c(2, 3, 2))
# Mixing example
d1 <- c(2, 3, 2)
d2 < -c(3, 2, 2)
x <- 1:(prod(d1) * prod(d2))
CommutatorIndx(Type = "Mixing", x = x, d1 = d1, d2 = d2)
# Moment example
n <- 4
r < -2
m < -1
d <- 2
PTA <- PartitionTypeAll(n)
el_r \leftarrow PTA\$eL_r[[r]][m, ]
x <- 1:d^n
CommutatorIndx(Type = "Moment", x = x, el_rm = el_r, d = d)
```

CommutatorMatr

Commutator Matrix

# **Description**

This function generates various types of commutator matrices.

# Usage

```
CommutatorMatr(Type, ...)
```

# Arguments

Type A string specifying the type of commutator matrix. Choices are "Kmn", "Kperm", "Mixing", or "Moment".

.. Additional arguments specific to the type of commutator matrix (see Details).

CommutatorMatr 5

#### **Details**

The function CommutatorMatr supports the following types of commutator matrices:

**Kmn Description:** Transforms vec(A) to vec(A^T), where A^T is the transpose of matrix A. An option for sparse matrix is provided. By default, a non-sparse matrix is produced. Using sparse matrices increases computation times but requires far less memory. **Arguments:** 

m (integer) Number of rows of the first matrix.

n (integer) Number of columns of the first matrix.

useSparse (logical, optional) If TRUE, returns a sparse matrix. Default is FALSE.

**Kperm Description:** Generates a commutation matrix for a specified permutation of matrix dimensions. An option for sparse matrix is provided. By default, a non-sparse matrix is produced. Using sparse matrices increases computation times but requires far less memory. **Arguments:** 

perm (integer vector) The permutation vector.

dims (integer vector) The dimensions of the matrices involved.

useSparse (logical, optional) If TRUE, returns a sparse matrix. Default is FALSE.

**Mixing Description:** Generates the Mixing commutation matrix used in linear algebra transformations involving tensor products. An option for sparse matrix is provided. By default, a non-sparse matrix is produced. Using sparse matrices increases computation times but requires far less memory. **Arguments:** 

d1 (integer vector) Dimensions of the first set.

d2 (integer vector) Dimensions of the second set.

useSparse (logical, optional) If TRUE, returns a sparse matrix. Default is FALSE.

**Moment Description:** Generates the Moment commutation matrix based on partitioning of moments. An option for sparse matrix is provided. By default, a non-sparse matrix is produced. Using sparse matrices increases computation times but requires far less memory. **Arguments:** 

el\_rm (integer vector) Elements of the partition.

d (integer) Dimension of the partition.

useSparse (logical, optional) If TRUE, returns a sparse matrix. Default is FALSE.

#### Value

Depending on the type:

**Kmn** A commutation matrix of dimension  $mn \times mn$ . If useSparse=TRUE, an object of class "dgCMatrix" is produced.

**Kperm** A square permutation matrix of size prod(dims). If useSparse=TRUE, an object of class "dgCMatrix" is produced.

**Mixing** A square matrix of dimension prod(d1) \* prod(d2). If useSparse=TRUE, an object of class "dgCMatrix" is produced.

**Moment** A commutator matrix for moment formulae.

# See Also

Other Commutators: CommutatorIndx()

6 Cum2Mom

# **Examples**

```
# Example for Kmn
CommutatorMatr("Kmn", m = 3, n = 2)
# Example for Kperm
dims < -c(2, 3, 2)
perm <- c(1, 3, 2)
CommutatorMatr("Kperm", perm = perm, dims = dims)
# Example for Mixing
d1 \leftarrow c(2, 3, 2)
d2 < -c(3, 2, 2)
CommutatorMatr("Mixing", d1 = d1, d2 = d2)
# Example for Moment
n <- 4
r <- 2
m <- 1
d <- 2
PTA <- PartitionTypeAll(n)
el_r \leftarrow PTA\$eL_r[[r]][m,]
CommutatorMatr("Moment", el_r = el_r, d = d)
```

Cum2Mom

Convert cumulants to moments (univariate and multivariate)

# **Description**

Obtains a vector of moments from a vector of cumulants for either univariate or multivariate data.

# Usage

```
Cum2Mom(cumulants, Type = c("Univariate", "Multivariate"))
```

# **Arguments**

cumulants Either a vector of univariate cumulants or a list of vectors of multivariate cumu-

lants.

Type A character string specifying the type of cumulants provided. Use "Univariate"

for univariate cumulants and "Multivariate" for multivariate cumulants.

#### Value

The vector of moments if Type is "Univariate" or the list of vectors of moments if Type is "Multivariate".

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Section 3.4.

EliminIndx 7

# See Also

```
Other Moments and cumulants: EVSKSkewNorm(), EVSKUniS(), Mom2Cum(), MomCumCFUSN(), MomCumSkewNorm(), MomCumUniS(), MomCumZabs()
```

# **Examples**

```
# Univariate example
cum_x <- c(1, 2, 3, 4)
Cum2Mom(cum_x, Type = "Univariate")

# Multivariate example
cum <- list(c(0,0), c(1,0,0,1), c(rep(0,8)), c(rep(0,16)), c(rep(0,32)))
Cum2Mom(cum, Type = "Multivariate")</pre>
```

EliminIndx

Distinct values selection vector

# **Description**

Eliminates the duplicated/q-plicated elements in a T-vector of multivariate moments and cumulants. Produces the same results as EliminMatr. Note EliminIndx does not provide the same results as unique()

# Usage

```
EliminIndx(d, q)
```

# **Arguments**

d dimension of a vector x

q power of the Kronecker product

#### Value

A vector of indexes of the distinct elements in the T-vector

#### References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Section 1.3.2 Multi-Indexing, Elimination, and Duplication, p.21,(1.32)

# See Also

```
Other Matrices and commutators: EliminMatr(), QplicIndx(), QplicMatr(), SymIndx(), SymMatr(), UnivMomCum()
```

8 EliminMatr

# **Examples**

```
x<-c(1,0,3)
y<-kronecker(x,kronecker(x,x))
y[EliminIndx(3,3)]
## Not the same results as
unique(y)</pre>
```

EliminMatr

Elimination Matrix

# **Description**

Eliminates the duplicated/q-plicated elements in a T-vector of multivariate moments and cumulants.

# Usage

```
EliminMatr(d, q, useSparse = FALSE)
```

# **Arguments**

d dimension of a vector x

q power of the Kronecker product

useSparse TRUE or FALSE.

# Value

Elimination matrix of order  $\eta_{d,q} \times d^q = \binom{d+q-1}{q}$ . If useSparse=TRUE an object of the class "dgCMatrix" is produced.

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Section 1.3.2 Multi-Indexing, Elimination, and Duplication, p.21,(1.32)

#### See Also

```
Other Matrices and commutators: EliminIndx(), QplicIndx(), QplicMatr(), SymIndx(), SymMatr(), UnivMomCum()
```

```
x<-c(1,2,3)
y<-kronecker(kronecker(x,x),x)
## Distinct elements of y
z<-as.matrix(EliminMatr(3,3))%*%y
## Restore eliminated elements in z
as.vector(QplicMatr(3,3)%*%z)</pre>
```

EVSKSkewNorm 9

EVSKSkewNorm	EVSK multivariate Skew Normal

# **Description**

Computes the theoretical values of the mean vector, covariance, skewness vector, total skenwness, kurtosis vector and total kurtosis for the multivariate Skew Normal distribution

# Usage

```
EVSKSkewNorm(omega, alpha)
```

# Arguments

omega A  $d \times d$  correlation matrix alpha shape parameter d-vector

# Value

A list of theoretical values for the mean vector, covariance, skewness vector, total skenwness, kurtosis vector and total kurtosis

### References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021 (5.5) p.247 S. R. Jammalamadaka, E. Taufer, Gy. Terdik. On multivariate skewness and kurtosis. Sankhya A, 83(2), 607-644.

# See Also

```
Other \ Moments \ and \ cumulants: \ Cum2Mom(), EVSKUniS(), Mom2Cum(), MomCumCFUSN(), MomCumSkewNorm(), MomCumUniS(), MomCumZabs()
```

```
alpha<-c(10,5,0)
omega<-diag(3)
EVSKSkewNorm(omega,alpha)</pre>
```

10 EVSKUniS

F١	/SK	Un	i	S

EVSK of the Uniform distribution on the sphere or its modulus

# Description

Cumulants (up to the 4th order), skewness, and kurtosis of the d-variate Uniform distribution on the sphere or the modulus of the d-variate Uniform distribution on the sphere.

# Usage

```
EVSKUniS(d, nCum = TRUE, Type = c("Standard", "Modulus"))
```

# **Arguments**

d dimensions

nCum if it is FALSE then moments (up to the 4th order) are calculated.

Type specify the type of distribution: "Standard" for the Uniform distribution on the

sphere, or "Modulus" for the modulus of the Uniform distribution on the sphere.

#### Value

A list of computed moments and cumulants.

When Type is "Standard":

EU1 Mean vector

varU Covariance matrix

Skew. U Skewness vector (always zero)

Skew.tot Total skewness (always zero)

Kurt.U Kurtosis vector
Kurt.tot Total kurtosis

When Type is "Modulus":

EU1 Mean vector

varU Covariance matrix

EU.k List of moments up to 4th order cumU.k List of cumulants up to 4th order

skew.U Skewness vector kurt.U Kurtosis vector HermiteCoeff 11

# References

Gy. Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021 Proposition 5.3 p.297

S. R. Jammalamadaka, E. Taufer, Gy. Terdik. On multivariate skewness and kurtosis. Sankhya A, 83(2), 607-644.

Gy. Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021, Lemma 5.12 p.298

#### See Also

```
Other Moments and cumulants: Cum2Mom(), EVSKSkewNorm(), Mom2Cum(), MomCumCFUSN(), MomCumSkewNorm(), MomCumUniS(), MomCumZabs()
```

# **Examples**

```
# Example for Standard type
EVSKUniS(d=3, Type="Standard")
# Example for Modulus type
EVSKUniS(d=3, Type="Modulus")
```

HermiteCoeff

Coefficients of Hermite polynomials

# **Description**

Provides the coefficients of Hermite polynomials, either univariate or multivariate.

#### Usage

```
HermiteCoeff(Type, N, d = NULL)
```

#### Arguments

Туре	A character string specifying the type of Hermite polynomial. Must be either "Univariate" or "Multivariate".
N	The order of polynomial. Required for both types.
d	The dimension of the d-variate X. Required only for multivariate type.

# Value

For 'Type = "Univariate"', returns a vector of coefficients of  $x^N$ ,  $x^{N-2}$ , etc. For 'Type = "Multivariate"', returns a list of matrices of coefficients for the d-variate polynomials from 1 to N.

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Sections 4.4 (4.24) and 4.6.2, p. 223, Remark 4.8

12 HermiteCov12

# See Also

Other Hermite Polynomials: HermiteCov12(), HermiteN(), HermiteN2X()

# **Examples**

```
# Univariate example
H_uni <- HermiteCoeff(Type = "Univariate", N = 5)

# Multivariate example
N <- 5; d <- 3
H_multi <- HermiteCoeff(Type = "Multivariate", N = N, d = d)
X <- c(1:3)
X3 <- kronecker(X, kronecker(X, X))
X5 <- kronecker(X3, kronecker(X, X))
Idv <- as.vector(diag(d)) # vector of variance matrix
# value of H5 at X is
vH5 <- H_multi[[1]] %*% X5 + H_multi[[2]] %*% kronecker(Idv, X3) +
H_multi[[3]] %*% kronecker(kronecker(Idv, Idv), X)</pre>
```

HermiteCov12

Covariance matrix for multivariate T-Hermite polynomials

# **Description**

Computation of the covariance matrix between d-variate T-Hermite polynomials  $H_N(X_1)$  and  $H_N(X_2)$ .

# Usage

```
HermiteCov12(SigX12, N)
```

# **Arguments**

SigX12 Covariance matrix of the Gaussian vectors X1 and X2 respectively of dimen-

sions d1 and d2

N Common degree of the multivariate Hermite polynomials

# Value

```
Covariance matrix of H_N(X_1) and H_N(X_2)
```

#### References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. (4.59), (4.66),

# See Also

```
Other Hermite Polynomials: HermiteCoeff(), HermiteN(), HermiteN2X()
```

HermiteN 13

# **Examples**

```
Covmat<-matrix(c(1,0.8,0.8,1),2,2)
Cov_X1_X2 <- HermiteCov12(Covmat,3)</pre>
```

HermiteN	Hermite Polynomials (Univariate and Multivariate)	

# Description

Computes either univariate or multivariate Hermite polynomials up to a specified order.

# Usage

```
HermiteN(x, N, Type, sigma2 = 1, Sig2 = diag(length(x)))
```

# Arguments

Х	A scalar (for univariate) or a vector (for multivariate) at which to evaluate the Hermite polynomials.
N	The maximum order of the polynomials.
Туре	A character string specifying the type of Hermite polynomials to compute. Can be either "Univariate" or "Multivariate".
sigma2	The variance for univariate Hermite polynomials. Default is 1. (Only used if Type is "Univariate").
Sig2	The covariance matrix for multivariate Hermite polynomials. Default is the unit matrix diag(length(x)). (Only used if Type is "Multivariate").

# **Details**

Depending on the value of the 'Type' parameter, this function computes either the univariate or the multivariate Hermite polynomials.

# Value

Depending on the type, the function returns:

- Univariate: A vector of univariate Hermite polynomials with degrees from 1 to N evaluated at x.
- Multivariate: A list of multivariate polynomials of order from 1 to N evaluated at vector x.

### References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Section 4.1 (for univariate), Section 4.6.2, (4.73), p.223 (for multivariate).

14 HermiteN2X

# See Also

Other Hermite Polynomials: HermiteCoeff(), HermiteCov12(), HermiteN2X()

# **Examples**

```
# Univariate example
HermiteN(x = 1, N = 3, Type = "Univariate")

# Multivariate example
HermiteN(x = c(1, 3), N = 3, Type = "Multivariate", Sig2 = diag(2))
```

HermiteN2X

Inverse Hermite Polynomial

# **Description**

Compute the inverse of univariate or multivariate Hermite polynomials.

# Usage

```
HermiteN2X(Type, H_N, N, Sig2 = NULL)
```

# **Arguments**

Туре	A string specifying the type of Hermite polynomial inversion. Must be either "Univariate" or "Multivariate".
H_N	Input Hermite polynomials. For univariate, it is a vector. For multivariate, it is a list.
N	The highest polynomial order.
Sig2	The variance matrix of x for multivariate, or variance for univariate. Defaults to identity matrix for multivariate and 1 for univariate.

# **Details**

This function computes the powers of x when Hermite polynomials are given. Depending on the type specified, it handles either univariate or multivariate Hermite polynomials.

#### Value

A list of x powers:  $x, x^{\otimes 2}, \dots, x^{\otimes N}$  for multivariate, or a vector of x powers:  $x^n, n = 1:N$  for univariate.

### References

Gy.Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Section 4.6.2, (4.72), p.223 and Section 4.4, (4.23), p.198.

Mom2Cum 15

# See Also

Other Hermite Polynomials: HermiteCoeff(), HermiteCov12(), HermiteN()

# **Examples**

```
# Univariate example
H_N_x <- c(1, 2, 3, 4)
x_powers <- HermiteN2X(Type = "Univariate", H_N = H_N_x, N = 4, Sig2 = 1)

# Multivariate example
x <- c(1, 3)
Sig2 <- diag(length(x))
N <- 4
H_N_X <- HermiteN(x, N, Type="Multivariate")
x_ad_n <- HermiteN2X(Type = "Multivariate", H_N = H_N_X, N = N, Sig2 = Sig2)</pre>
```

Mom2Cum

Convert moments to cumulants (univariate and multivariate)

# **Description**

Obtains a vector of cumulants from a vector of moments for either univariate or multivariate data.

# Usage

```
Mom2Cum(moments, Type = c("Univariate", "Multivariate"))
```

#### **Arguments**

moments Either a vector of univariate moments or a list of vectors of multivariate mo-

ments.

Type A character string specifying the type of moments provided. Use "Univariate"

for univariate moments and "Multivariate" for multivariate moments.

### Value

The vector of cumulants if Type is "Univariate" or the list of vectors of cumulants if Type is "Multivariate".

#### References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Section 3.4.

#### See Also

```
Other Moments and cumulants: Cum2Mom(), EVSKSkewNorm(), EVSKUniS(), MomCumCFUSN(), MomCumSkewNorm(), MomCumUniS(), MomCumZabs()
```

16 MomCumCFUSN

# **Examples**

```
# Univariate example
mu_x <- c(1, 2, 3, 4)
Mom2Cum(mu_x, Type = "Univariate")

# Multivariate example
mu <- list(c(0,0), c(1,0,0,1), c(0,0,0,0,0,0,0), c(3,0,0,1,0,1,1,0,0,1,1,0,1,0,0,3), c(rep(0,32)))
Mom2Cum(mu, Type = "Multivariate")</pre>
```

MomCumCFUSN

Moments and cumulants CFUSN

# **Description**

Provides the theoretical cumulants of the multivariate Canonical Fundamental Skew Normal distribution

# Usage

```
MomCumCFUSN(r, d, p, Delta, nMu = FALSE)
```

# **Arguments**

r	The highest cumulant order
d	The multivariate dimension and number of rows of the skewness matrix Delta
р	The number of cols of the skewness matrix Delta
Delta	The skewness matrix
nMu	If set to TRUE, the list of the first r d-variate moments is provided

#### Value

The list of theoretical cumulants in vector form

# References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021, Lemma 5.3 p.251

# See Also

```
Other Moments and cumulants: Cum2Mom(), EVSKSkewNorm(), EVSKUniS(), Mom2Cum(), MomCumSkewNorm(), MomCumUniS(), MomCumZabs()
```

MomCumSkewNorm 17

# **Examples**

```
r <- 4; d <- 2; p <- 3
Lamd <- matrix(sample(1:50-25, d*p), nrow=d)
ieg<- eigen(diag(p)+t(Lamd)%*%Lamd)
V <- ieg$vectors
Delta <-Lamd %*% V %*% diag(1/sqrt(ieg$values)) %*% t(V)
MomCum <- MomCumCFUSN(r,d,p,Delta)</pre>
```

MomCumSkewNorm

Moments and cumulants d-variate Skew Normal

# **Description**

Computes the theoretical values of moments and cumulants up to the r-th order. Warning: if nMu = TRUE it can be very slow

# Usage

```
MomCumSkewNorm(r = 4, omega, alpha, nMu = FALSE)
```

# **Arguments**

r the highest moment and cumulant order

omega A  $d \times d$  correlation matrix alpha shape parameter d-vector

nMu if it is TRUE then moments are calculated as well

#### Value

A list of theoretical moments and cumulants

# References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021 (5.5) p.247, Lemma 5.1 p. 246

S. R. Jammalamadaka, E. Taufer, Gy. Terdik. On multivariate skewness and kurtosis. Sankhya A, 83(2), 607-644.

# See Also

```
Other Moments and cumulants: Cum2Mom(), EVSKSkewNorm(), EVSKUniS(), Mom2Cum(), MomCumCFUSN(), MomCumUniS(), MomCumZabs()
```

```
alpha<-c(10,5,0)
omega<-diag(3)
MomCumSkewNorm(r=4,omega,alpha)</pre>
```

18 MomCumUniS

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Moments and cumulants Uniform Distribution on the Sphere

# Description

By default, only moments are provided

# Usage

```
MomCumUniS(r, d, nCum = FALSE)
```

# **Arguments**

r highest order of moments and cumulants

d dimension

nCum if it is TRUE then cumulants are calculated

# Value

The list of moments and cumulants in vector form

# References

Gy. Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021 Proposition  $5.3 \, \mathrm{p.}297$ 

#### See Also

```
Other Moments and cumulants: Cum2Mom(), EVSKSkewNorm(), EVSKUniS(), Mom2Cum(), MomCumCFUSN(), MomCumSkewNorm(), MomCumZabs()
```

```
# The first four moments for d=3
MomCumUniS(4,3,nCum=0)
# The first four moments and cumulants for d=3
MomCumUniS(4,3,nCum=4)
```

MomCumZabs 19

MomCumZabs	Moments and Cumulants of the Central Folded Normal Distribution

# Description

Provides the theoretical moments and cumulants of the Central Folded Normal distribution. Depending on the choice of 'Type', either the univariate or d-variate distribution is used.

#### **Usage**

```
MomCumZabs(r, d, Type, nCum = FALSE)
```

# Arguments

r	The highest moment (cumulant) order.
d	Integer; the dimension of the distribution. Must be 1 when 'Type' is "Univariate" and greater than 1 when 'Type' is "Multivariate".
Type	Character; specifies the type of distribution. Must be either "Univariate" or "Multivariate".
nCum	Logical; if TRUE, then cumulants are calculated.

# Value

A list containing moments and optionally cumulants.

- For "Univariate" type:
  - MuZ: The moments of the univariate Central Folded Normal distribution.
  - CumZ: The cumulants of the univariate Central Folded Normal distribution.
- For "Multivariate" type:
  - MuZ: The moments of the d-variate Central Folded Normal distribution.
  - CumZ: The cumulants of the d-variate Central Folded Normal distribution.

# References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021, Proposition 5.1 p.242 and formula: p. 301

# See Also

```
Other Moments and cumulants: Cum2Mom(), EVSKSkewNorm(), EVSKUniS(), Mom2Cum(), MomCumCFUSN(), MomCumSkewNorm(), MomCumUniS()
```

20 MVStandardize

# **Examples**

```
# Univariate case: The first three moments
MomCumZabs(3, 1, Type = "Univariate")
# Univariate case: The first three moments and cumulants
MomCumZabs(3, 1, Type = "Univariate", nCum = TRUE)
# d-variate case: The first three moments
MomCumZabs(3, 2, Type = "Multivariate")
# d-variate case: The first three moments and cumulants
MomCumZabs(3, d=2, Type = "Multivariate", nCum = TRUE)
```

MVStandardize

Standardize multivariate data

# Description

For data formed by d-variate vectors x with sample covariance S and sample mean M, it computes the values  $z=S^{-1/2}(x-M)$ 

# Usage

MVStandardize(x)

# Arguments

Χ

a multivariate data matrix, sample size is the number of rows

# Value

a matrix of multivariate data with null mean vector and identity sample covariance matrix

```
x<-MASS::mvrnorm(1000,c(0,0,1,3),diag(4))
z<-MVStandardize(x)
mu_z<- apply(z,2,mean)
cov_z<- cov(z)</pre>
```

Partitions 21

Partitions

General Partition Function

# **Description**

A unified function to compute different types of partitions. Depending on the partition type specified, it calls the appropriate function: Partition\_2Perm, Partition\_DiagramsClosedNoLoops, Partition\_Indecomposable, or Partition\_Pairs.

# Usage

```
Partitions(Type, ...)
```

### **Arguments**

A character string specifying the type of partion to compute. Choose from "2Perm", "Diagram", "Indecomp", "Pairs".
 ... Additional arguments passed to the specific partition function:
 For "2Perm", "Diagram" and "Indecomp": • L: A partition matrix.

 For "Pairs": • N: An integer specifying the number of elements to be partitioned.

#### Value

Depending on the commutator type:

**2Perm** A vector with the elements 1 to N permuted according to L.

**Diagram** The list of partition matrices indecomposable with respect to L, representing diagrams without loops.

**Indecomp** A list of partition matrices indecomposable with respect to L and a vector indicating the number of indecomposable partitions by sizes.

**Pairs** The list of partition matrices with blocks containing two elements. The list is empty if N is odd.

#### See Also

```
Other Partitions: PartitionTypeAll(), PermutationInv()
```

```
# Example for 2Perm
PA <- PartitionTypeAll(4)
Partitions("2Perm", L = PA$Part.class[[3]])
# Example for Diagram
L <- matrix(c(1,1,0,0,0,0,1,1),2,4,byrow=TRUE)
Partitions("Diagram", L = L)</pre>
```

22 PartitionTypeAll

```
# Example for Indecomp
L <- matrix(c(1,1,0,0,0,0,1,1),2,4,byrow=TRUE)
Partitions("Indecomp", L = L)
# Example for Pairs
Partitions("Pairs", N = 4)</pre>
```

PartitionTypeAll

Partitions, type and number of partitions

# Description

Generates all partitions of N numbers and classify them by type

# Usage

```
PartitionTypeAll(N)
```

# **Arguments**

Ν

The (integer) number of elements to be partitioned

# Value

Part. class The list of all possible partitions given as partition matrices

S\_N\_r A vector with the number of partitions of size r=1, r=2, etc. (Stirling numbers of second kind )

eL\_r A list of partition types with respect to partitions of size r=1, r=2, etc.

S\_r\_j Vectors of number of partitions with given types grouped by partitions of size r=1, r=2, etc.

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Case 1.4, p.31 and Example 1.18, p.32.

# See Also

```
Other Partitions: Partitions(), PermutationInv()
```

PermutationInv 23

# **Examples**

```
# See Example 1.18, p. 32, reference below
PTA<-PartitionTypeAll(4)
# Partitions generated
PTA$Part.class
# Partitions of size 2 includes two types
PTA$eL_r[[2]]
# Number of partitions with r=1 blocks, r=2 blocks, etc-
PTA$S_N_r
# Number of different types collected by partitions of size r=1, r=2, etc.
PTA$S_r_j
# Partitions with size r=2, includes two types (above) each with number
PTA$S_r_j[[2]]</pre>
```

PermutationInv

Inverse of a Permutation

# **Description**

Inverse of a Permutation

# Usage

PermutationInv(permutation0)

# Arguments

permutation 0 A permutation of numbers 1:n

# Value

A vector containing the inverse permutation of permutation0

### References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021, Remark 1.1, p.2

# See Also

```
Other Partitions: PartitionTypeAll(), Partitions()
```

24 *QplicIndx* 

QplicIndx

**Qplication** vector

# **Description**

Restores the duplicated/q-plicated elements which are eliminated by EliminMatr or EliminIndx in a T-product of vectors of dimension d. It produces the same results as QplicMatr.

# Usage

```
QplicIndx(d, q)
```

# **Arguments**

- d dimension of the vectors in the T-product
- q power of the Kronecker product

#### Value

A vector (T-vector) with all elements previously eliminated by EliminIndx

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021, p.21, (1.31)

# See Also

```
Other Matrices and commutators: EliminIndx(), EliminMatr(), QplicMatr(), SymIndx(), SymMatr(), UnivMomCum()
```

```
x<-c(1,2,3)
y<-kronecker(kronecker(x,x),x)
## Distinct elements of y
z<-y[EliminIndx(3,3)]
## Restore eliminated elements in z
z[QplicIndx(3,3)]</pre>
```

QplicMatr 25

|--|

# **Description**

Restores the duplicated/q-plicated elements which are eliminated by EliminMatr in a T-product of vectors of dimension d.

# Usage

```
QplicMatr(d, q, useSparse = FALSE)
```

# **Arguments**

d dimension of a vector x

q power of the Kronecker product

useSparse TRUE or FALSE.

#### **Details**

Note: since the algorithm of elimination is not unique, q-plication works together with the function EliminMatr only.

# Value

Qplication matrix of order  $d^q \times \eta_{d,q}$ , see (1.30), p.15. If useSparse=TRUE an object of the class "dgCMatrix" is produced.

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021, p.21, (1.31)

# See Also

```
Other Matrices and commutators: EliminIndx(), EliminMatr(), QplicIndx(), SymIndx(), SymMatr(), UnivMomCum()
```

```
x<-c(1,2,3)
y<-kronecker(kronecker(x,x),x)
## Distinct elements of y
z<-as.matrix(EliminMatr(3,3))%*%y
## Restore eliminated elements in z
as.vector(QplicMatr(3,3)%*%z)</pre>
```

26 rCFUSN

rCFUSN

Random multivariate CFUSN

# **Description**

Generate random d-vectors from the multivariate Canonical Fundamental Skew-Normal (CFUSN) distribution

# Usage

```
rCFUSN(n, Delta)
```

# **Arguments**

n The number of variates to be generated

Delta Correlation matrix, the skewness matrix Delta

#### Value

A random matrix  $n \times d$ 

# References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021 (5.5) p.247

S. R. Jammalamadaka, E. Taufer, Gy. Terdik. On multivariate skewness and kurtosis. Sankhya A, 83(2), 607-644.

# See Also

```
Other Random generation: rCFUSSD(), rSkewNorm(), rUniS()
```

```
d <- 2; p <- 3
Lamd <- matrix(sample(1:50-25, d*p), nrow=d)
ieg<- eigen(diag(p)+t(Lamd)%*%Lamd)
V <- ieg$vectors
Delta <-Lamd %*% V %*% diag(1/sqrt(ieg$values)) %*% t(V)
x<-rCFUSN(20,Delta)</pre>
```

rCFUSSD 27

|--|

# Description

Generate random d-vectors from the multivariate Canonical Fundamental Skew-Spherical distribution (CFUSSD) with Gamma generator

# Usage

```
rCFUSSD(n, d, p, a, b, Delta)
```

# **Arguments**

n	sample size
d	dimension
p	dimension of the first term of (5.5)
a	shape parameter of the Gamma generator
b	scale parameter of the Gamma generator
Delta	skewness matrix

# Value

A matrix of  $n \times d$  random numbers

# References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021, (5.36) p. 266, (see p.247 for Delta)

# See Also

```
Other Random generation: rCFUSN(), rSkewNorm(), rUniS()
```

```
n <- 10^3; d <- 2; p <- 3; a <- 1; b <- 1
Lamd <- matrix(sample(1:50-25, d*p), nrow=d)
ieg<- eigen(diag(p)+t(Lamd)%*%Lamd)
V <- ieg$vectors
Delta <-Lamd %*% V %*% diag(1/sqrt(ieg$values)) %*% t(V)
rCFUSSD(20,d,p,1,1,Delta)</pre>
```

28 rSkewNorm

rSkewNorm

Random Multivariate Skew Normal

# **Description**

Generate random d-vectors from the multivariate Skew Normal distribution

# Usage

```
rSkewNorm(n, omega, alpha)
```

# **Arguments**

n sample size

omega correlation matrix with d dimension
alpha shape parameter vector of dimension d

# Value

A random matrix  $n \times d$ 

# References

Azzalini, A. with the collaboration of Capitanio, A. (2014). The Skew-Normal and Related Families. Cambridge University Press, IMS Monographs series.

Gy.H.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021, Section 5.1.2

# See Also

```
Other Random generation: rCFUSN(), rCFUSSD(), rUniS()
```

```
alpha<-c(10,5,0)
omega<-diag(3)
x<-rSkewNorm(20,omega,alpha)</pre>
```

rUniS 29

rUniS

Random Uniform on the sphere

# **Description**

Generate random d-vectors from the Uniform distribution on the sphere

# Usage

```
rUniS(n, d)
```

# **Arguments**

n sample size d dimension

# Value

A random matrix  $n \times d$ 

#### References

Gy. Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021

S. R. Jammalamadaka, E. Taufer, Gy. Terdik. On multivariate skewness and kurtosis. Sankhya A, 83(2), 607-644.

# See Also

Other Random generation: rCFUSN(), rCFUSSD(), rSkewNorm()

SampleEVSK Estimation of multivariate Mean, Variance, T-Skewness and T-Kurtosis vectors

# Description

Provides estimates of mean, variance, skewness and kurtosis vectors for d-variate data

# Usage

```
SampleEVSK(X)
```

# **Arguments**

X d-variate data vector

30 SampleGC

# Value

The list of the estimated mean, variance, skewness and kurtosis vectors

#### References

Gy.Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021, Sections 6.4.1 and 6.5.1

# See Also

```
Other Estimation: SampleKurt(), SampleMomCum(), SampleSkew(), VarianceKurt(), VarianceSkew()
```

# **Examples**

```
x<- MASS::mvrnorm(100,rep(0,3), 3*diag(rep(1,3)))
EVSK<-SampleEVSK(x)
names(EVSK)
EVSK$estSkew</pre>
```

SampleGC

Gram-Charlier approximation to a multivariate density

# **Description**

Provides the truncated Gram-Charlier approximation to a multivariate density. Approximation can be up to the first k=8 cumulants.

# Usage

```
SampleGC(X, k = 4, cum = NULL)
```

# **Arguments**

X A matrix of d-variate data

k the order of the approximation, by default set to 4; (k must not be smaller than

3 or greater than 8)

cum if NULL (default) the cumulant vector is estimated from X. If cum is provided

no estimation of cumulants is performed.

# Value

The vector of the Gram-Charlier density evaluated at X

# References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021. Section 4.7.

SampleHermiteN 31

# **Examples**

```
# Gram-Charlier density approximation (k=4) of data generated from
# a bivariate skew-gaussian distribution
n<-50
alpha<-c(10,0)
omega<-diag(2)
X<-rSkewNorm(n,omega,alpha)
EC<-SampleEVSK(X)
fy4<-SampleGC(X[1:5,],cum=EC)</pre>
```

SampleHermiteN

Estimate the N-th d-variate Hermite polynomial

# Description

The vector x is standardized and the N-th d-variate polynomial is computed

### **Usage**

```
SampleHermiteN(x, N)
```

# **Arguments**

x a d-variate data vector

N the order of the d-variate Hermite polynomial

# Value

The vector of the N-th d-variate polynomial

# **Examples**

```
x<-MASS::mvrnorm(100,rep(0,3),diag(3))
H3<-SampleHermiteN(x,3)</pre>
```

SampleKurt

Estimation of Sample Kurtosis (Mardia, MRSz, Total)

# **Description**

Estimates the sample kurtosis index based on the specified method: Mardia, MRSz, or Total.

# Usage

```
SampleKurt(x, Type = c("Mardia", "MRSz", "Total"))
```

32 SampleMomCum

# **Arguments**

x A matrix of multivariate data.

Type A character string specifying the type of kurtosis index to estimate. Use "Mar-

dia" for Mardia's kurtosis index, "MRSz" for the Mori-Rohatgi-Szekely kurtosis

matrix, or "Total" for the total kurtosis index.

# Value

A list containing the estimated kurtosis index or matrix and the associated p-value under the Gaussian hypothesis.

Mardia.Kurtosis

The kurtosis index when Type is "Mardia".

MRSz.Kurtosis The kurtosis matrix when Type is "MRSz".

Total.Kurtosis The total kurtosis index when Type is "Total".

p. value The p-value under the Gaussian hypothesis for the estimated kurtosis.

#### References

Gy.Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Example 6.1 and 6.9.

# See Also

Other Estimation: SampleEVSK(), SampleMomCum(), SampleSkew(), VarianceKurt(), VarianceSkew()

# **Examples**

```
# Mardia's kurtosis example
x <- matrix(rnorm(100*5), ncol=5)
SampleKurt(x, Type = "Mardia")

# MRSz's kurtosis example
SampleKurt(x, Type = "MRSz")

# Total kurtosis example
SampleKurt(x, Type = "Total")</pre>
```

SampleMomCum

Estimation of multivariate T-Moments and T-Cumulants

# **Description**

Provides estimates of univariate and multivariate moments and cumulants up to order r. By default data are standardized; using only demeaned or raw data is also possible.

SampleSkew 33

# Usage

```
SampleMomCum(X, r, centering = FALSE, scaling = TRUE)
```

#### **Arguments**

X d-vector data

r The highest moment order (r > 2)

centering set to T (and scaling = F) if only centering is needed

scaling set to T (and centering=F) if standardization of multivariate data is needed

#### Value

```
estMu.r: the list of the multivariate moments up to order r estCum.r: the list of the multivariate cumulants up to order r
```

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021.

#### See Also

```
Other Estimation: SampleEVSK(), SampleKurt(), SampleSkew(), VarianceKurt(), VarianceSkew()
```

# **Examples**

```
## generate random data from a 3-variate skew normal distribution
alpha<-c(10,5,0)
omega<-diag(3)
x<-rSkewNorm(50,omega,alpha)
## estimate the first three moments and cumulants from raw (uncentered and unstandardized) data
SampleMomCum(x,3,centering=FALSE,scaling=FALSE)
## estimate the first three moments and cumulants from standardized data
SampleMomCum(x,3,centering=FALSE,scaling=TRUE)</pre>
```

SampleSkew

Estimation of Sample Skewness (Mardia, MRSz)

# **Description**

Estimates the sample skewness index based on the specified method: Mardia or MRSz.

# Usage

```
SampleSkew(x, Type = c("Mardia", "MRSz"))
```

34 SampleSkew

# **Arguments**

x A matrix of multivariate data.

Type A character string specifying the type of skewness index to estimate. Use "Mar-

dia" for Mardia's skewness index or "MRSz" for the Mori-Rohatgi-Szekely

skewness vector and index.

#### Value

A list containing the estimated skewness index or vector and the associated p-value under the Gaussian hypothesis.

Mardia.Skewness

The skewness index when Type is "Mardia".

MRSz.Skewness.Vector

The skewness vector when Type is "MRSz".

MRSz.Skewness.Index

The skewness index when Type is "MRSz".

p. value The p-value under the Gaussian hypothesis for the estimated skewness.

#### References

Gy.Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Example 6.1 and 6.2.

S. R. Jammalamadaka, E. Taufer, Gy. Terdik. On multivariate skewness and kurtosis. Sankhya A, 83(2), 607-644.

# See Also

```
Other Estimation: SampleEVSK(), SampleKurt(), SampleMomCum(), VarianceKurt(), VarianceSkew()
```

```
# Mardia's skewness example
x <- matrix(rnorm(100*5), ncol=5)
SampleSkew(x, Type = "Mardia")
# MRSz's skewness example
SampleSkew(x, Type = "MRSz")</pre>
```

SampleVarianceSkewKurt

Estimated Variance of skewness and kurtosis vectors

# **Description**

Provides the estimated covariance matrices of the data-estimated skewness and kurtosis vectors.

# Usage

SampleVarianceSkewKurt(X)

# **Arguments**

Χ

A matrix of d-variate data

#### Value

The list of covariance matrices of the skewness and kurtosis vectors

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021.

SymIndx

Symmetrizing vector

# **Description**

Vector symmetrizing a T-product of vectors of the same dimension d. Produces the same results as SymMatr

# Usage

```
SymIndx(x, d, n)
```

# Arguments

x the vector to be symmetrized of dimension d^n

d size of the single vectors in the product

n power of the T-product

# Value

A vector with the symmetrized version of x of dimension d^n

36 SymMatr

# References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021.Section 1.3.1 Symmetrization, p.14. (1.29)

#### See Also

```
Other Matrices and commutators: EliminIndx(), EliminMatr(), QplicIndx(), QplicMatr(), SymMatr(), UnivMomCum()
```

# **Examples**

```
a<-c(1,2)
b<-c(2,3)
c<-kronecker(kronecker(a,a),b)
## The symmetrized version of c is
SymIndx(c,2,3)</pre>
```

SymMatr

Symmetrizer Matrix

# **Description**

Based on Chacon and Duong (2015) efficient recursive algorithms for functionals based on higher order derivatives. An option for sparse matrix is provided. By using sparse matrices far less memory is required and faster computation times are obtained

# Usage

```
SymMatr(d, n, useSparse = FALSE)
```

# **Arguments**

d dimension of a vector x

n power of the Kronecker product

useSparse TRUE or FALSE. If TRUE an object of the class "dgCMatrix" is produced.

# Value

A Symmetrizer matrix with order  $d^n \times d^n$ . If useSparse=TRUE an object of the class "dgCMatrix" is produced.

#### References

Chacon, J. E., and Duong, T. (2015). Efficient recursive algorithms for functionals based on higher order derivatives of the multivariate Gaussian density. Statistics and Computing, 25(5), 959-974.

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021.Section 1.3.1 Symmetrization, p.14. (1.29)

UnivMomCum 37

# See Also

```
Other Matrices and commutators: EliminIndx(), EliminMatr(), QplicIndx(), QplicMatr(), SymIndx(), UnivMomCum()
```

# **Examples**

```
a<-c(1,2)
b<-c(2,3)
c<-kronecker(kronecker(a,a),b)
## The symmetrized version of c is
as.vector(SymMatr(2,3)%*%c)</pre>
```

UnivMomCum

Univariate moments and cumulants from T-vectors

# Description

A vector of indexes to select the moments and cumulants of the single components of the random vector X for which a T-vector of moments and cumulants is available

# Usage

```
UnivMomCum(d, q)
```

# **Arguments**

d dimension of a vector X q power of the Kronecker product

#### Value

A vector of indexes

#### See Also

```
Other Matrices and commutators: EliminIndx(), EliminMatr(), QplicIndx(), QplicMatr(), SymIndx(), SymMatr()
```

```
## For a 3-variate skewness and kurtosis vectors estimated from data, extract
## the skewness and kurtosis estimates for each of the single components of the vector
alpha<-c(10,5,0)
omega<-diag(rep(1,3))
X<-rSkewNorm(200, omega, alpha)
EVSK<-SampleEVSK(X)
## Get the univariate skewness and kurtosis for X1,X2,X3
EVSK$estSkew[UnivMomCum(3,3)]
EVSK$estKurt[UnivMomCum(3,4)]</pre>
```

38 VarianceSkew

VarianceKurt

Asymptotic Variance of the estimated kurtosis vector

# Description

Warning: the function requires 8! computations, for d>3, the timing required maybe large.

# Usage

VarianceKurt(cum)

# **Arguments**

cum

The theoretical/estimated cumulants up to the 8th order in vector form

# Value

The matrix of theoretical/estimated variance

#### References

Gy. Terdik, Multivariate statistical methods - going beyond the linear, Springer 2021. Ch. 6, formula (6.26)

# See Also

Other Estimation: SampleEVSK(), SampleKurt(), SampleMomCum(), SampleSkew(), VarianceSkew()

VarianceSkew

Asymptotic Variance of the estimated skewness vector

# **Description**

Asymptotic Variance of the estimated skewness vector

# Usage

VarianceSkew(cum)

# **Arguments**

cum

The theoretical/estimated cumulants up to order 6 in vector form

# Value

The matrix of theoretical/estimated variance

VarianceSkew 39

# References

Gy.Terdik, Multivariate statistical methods - Going beyond the linear, Springer 2021. Ch.6, formula (6.13)

# See Also

```
Other Estimation: SampleEVSK(), SampleKurt(), SampleMomCum(), SampleSkew(), VarianceKurt()
```

```
alpha<-c(10,5)
omega<-diag(rep(1,2))
MC <- MomCumSkewNorm(r = 6,omega,alpha)
cum <- MC$CumX
VS <- VarianceSkew(cum)</pre>
```

# **Index**

* Commutators	rSkewNorm, 28
CommutatorIndx, 2	rUniS,29
CommutatorMatr, 4	
* Estimation	CommutatorIndx, $2, 5$
SampleEVSK, 29	CommutatorMatr, 3, 4
SampleKurt, 31	Cum2Mom, 6, 9, 11, 15–19
SampleMomCum, 32	51: : 7   5 0 0 0 0 5 0 0 0 0
SampleSkew, 33	EliminIndx, 7, 8, 24, 25, 36, 37
VarianceKurt, 38	EliminMatr, 7, 8, 24, 25, 36, 37
VarianceSkew, 38	EVSKSkewNorm, 7, 9, 11, 15–19
* Hermite Polynomials	EVSKUniS, 7, 9, 10, 15–19
HermiteCoeff, 11	HermiteCoeff, 11, <i>12</i> , <i>14</i> , <i>15</i>
HermiteCov12, 12	HermiteCov12, 12, 12, 14, 15
HermiteN, 13	HermiteN, 12, 13, 15
HermiteN2X, 14	HermiteN2X, 12, 14, 14
* Matrices and commutators	11CT III CCN2A, 12, 14, 14
EliminIndx, 7	Mom2Cum, 7, 9, 11, 15, 16–19
EliminMatr, 8	MomCumCFUSN, 7, 9, 11, 15, 16, 17–19
QplicIndx, 24	MomCumSkewNorm, 7, 9, 11, 15, 16, 17, 18, 19
QplicMatr, 25	MomCumUniS, 7, 9, 11, 15–17, 18, 19
SymIndx, 35	MomCumZabs, 7, 9, 11, 15-18, 19
SymMatr, 36	MVStandardize, 20
UnivMomCum, 37	
* Moments and cumulants	Partitions, 21, 22, 23
Cum2Mom, 6	PartitionTypeAll, <i>21</i> , 22, <i>23</i>
EVSKSkewNorm, 9	PermutationInv, <i>21</i> , <i>22</i> , 23
EVSKUniS, 10	0.11.7.1.7.0.04.05.36.37
Mom2Cum, 15	QplicIndx, 7, 8, 24, 25, 36, 37
MomCumCFUSN, 16	QplicMatr, 7, 8, 24, 25, 36, 37
MomCumSkewNorm, 17	rCFUSN, 26, 27–29
MomCumUniS, 18	rCFUSSD, 26, 27, 28, 29
MomCumZabs, 19	rSkewNorm, 26, 27, 28, 29
* Partitions	rUnis, 26–28, 29
Partitions, 21	101113, 20 20, 25
PartitionTypeAll, 22	SampleEVSK, 29, 32-34, 38, 39
PermutationInv, 23	SampleGC, 30
* Random generation	SampleHermiteN, 31
rCFUSN, 26	SampleKurt, 30, 31, 33, 34, 38, 39
rCFUSSD, 27	SampleMomCum, 30, 32, 32, 34, 38, 39

INDEX 41

```
SampleSkew, 30, 32, 33, 33, 38, 39
SampleVarianceSkewKurt, 35
SymIndx, 7, 8, 24, 25, 35, 37
SymMatr, 7, 8, 24, 25, 36, 36, 37
UnivMomCum, 7, 8, 24, 25, 36, 37, 37
VarianceKurt, 30, 32–34, 38, 39
VarianceSkew, 30, 32–34, 38, 38
```