

# Package ‘mvtBinaryEP’

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**Title** Generates Correlated Binary Data

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**Description** Generates correlated binary data based on the method of Emrich and Piedmonte (1991)

**Depends** mvtnorm

**License** GPL (>= 2)

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## R topics documented:

mvtBinaryEP-package . . . . .	1
ep . . . . .	2
ranMVN . . . . .	4
ranMVN2 . . . . .	5
ranMvnXch . . . . .	6

<b>Index</b>	<b>7</b>
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mvtBinaryEP-package    *A Package For Generating Correlated Binary Data*

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## Description

Uses the algorithm of Emrich and Piedmonte (1991) to generate correlated binary data for a given correlation matrix

## Details

Package: mvtBinaryEP  
 Type: Package  
 Version: 1.0  
 Date: 2009-02-16  
 License: GPL(>=2)

The most important function is `ep`. Other functions include `ranMVN`, `ranMVN2`, and `ranMvnXch`. These are used in the `ep` function to generate correlated binary responses. The user may use these functions if multivariate normal data is desired. We note that the `mvtnorm` package also avails the user of functions to generate multivariate data which may be more efficient than our mutivariate normal functions.

### Author(s)

These routines were written by Dr. Bahjat Qaqish in SAS and translated to R by Kunthel By. Questions may be relayed to <kby@bios.unc.edu>

### References

Emrich, L.J. and Piedmonte, M.R., A method for generating high-dimensional multivariate binary variates, *The American Statistician*, 45:302-304, 1991.

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ep *Correlated Binary Data*

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### Description

Generates correlated binary data based on the method of Emrich and Piedmonte (1991)

### Usage

```
ep(mu, R, rho, n, isd = NULL, nRep = 1, seed = NULL, crit = 1e-06,
maxiter = 20)
```

### Arguments

mu	Vector of means. If rho is specified then mu must be of length 1.
R	Correlation matrix. If rho is specified then R is ignored. Only the upper part of R is used.
rho	If common mean and exchangeable correlation is desired, then this correlation parameter must be specified.
n	Cluster size. If rho is specified, then this must be specified as well.

<code>isd</code>	Internal Simulation Descriptor. This is useful for generating more responses based on the parameters used in the prior call to <code>ep</code> . This increases efficiency since the intermediate quantities need not be recomputed. <code>isd</code> is a <b>list</b> containing some of the input parameters as well as some intermediate quantities. If this is provided then <code>R</code> and <code>rho</code> are ignored.
<code>nRep</code>	Number of clusters (replications).
<code>seed</code>	Sets the seed
<code>crit</code>	Level of precision used in solving for the tetra-choric correlations.
<code>maxiter</code>	Maximum number of iterations used in solving for the tetra-choric correlations.

### Details

The method relies on simulating multivariate normal vectors and then dichotomizing each coordinate. The cutpoints are determined by  $\mu$ . The correlation matrix  $S$  (which are the tetra-choric correlations) of the multivariate normal vectors is computed in such a way that the resulting binary vectors have correlation matrix  $R$ . One possible complication is that this process is not always possible. Further, when all tetra-choric correlations are computed, the resulting matrix,  $S$ , may not be positive definite. These are two possible failure points in the algorithm; both are detected and reported back by the code.

### Value

Returns a list with the following two components:

<code>y</code>	Multivariate response of dimension <code>nRep</code> by <code>length(mu)</code>
<code>isd</code>	Internal Simulation Descriptor

### Note

The returned object **isd** is also a list with the following fields:

- `mu` input parameter
- `rho` input parameter
- `n` input parameter
- `R` input parameter
- `rootS` The Cholesky root of the tetrachoric correlation matrix  $S$  (if positive definite)
- `S` The tetrachoric correlation matrix  $S$  (if NOT positive definite)
- `pd` Flag, TRUE if  $S$  is positive definite, FALSE otherwise
- `sp` Flag, TRUE if successful in solving for tetrachoric correlations
- `i` row where solving for the tetrachoric correlation failed, if it did fail
- `j` column where solving for the tetrachoric correlation failed, if it did fail

### See Also

See Also [ranMVN](#), [ranMVN2](#), [ranMvnXch](#)

**Examples**

```

# Create mean vector
mu=c(0.5, 0.3, 0.20, 0.1)

# Create correlation matrix
R = c(
  1 , 0.2 , 0.15, -0.05,
  0.2 , 1 , 0.25, 0.2 ,
  0.15 , 0.25, 1 , 0.25 ,
  -0.05, 0.2 , 0.25, 1
)
R = matrix(R, ncol=4)

ep0 = ep(mu=mu, R=R, nRep=1000, seed=NULL)
apply(ep0$y, 2, mean); cor(ep0$y)

#Generates more responses based on the parameters provided above.
ep1 = ep(isd = ep0$isd, nRep=1000, seed=NULL)
apply(ep1$y, 2, mean); cor(ep1$y)

# 5-variate based on common mean and exchangeable correlation.
ep2 = ep(mu=0.3, rho=0.2, n=5, nRep=10000)
apply(ep2$y, 2, mean); cor(ep2$y)

```

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ranMVN

*Multivariate Normal Data*


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**Description**

Generates multivariate normal data based on a specified covariance matrix.

**Usage**

```
ranMVN(nRep = 1, Sigma, seed = NULL)
```

**Arguments**

nRep	Numer of clusters (replications)
Sigma	Specified covariance matrix
seed	Initialize random number generator

**Details**

The returned matrix of responses, call it Y, has mean 0.

**Value**

Returns a matrix of response variates of dimension nRep by nrow(Sigma)

**See Also**

See Also [ranMVN2](#), [ranMvnXch](#), [ep](#)

**Examples**

```
mu = c(25.3, 28.4, 35.7, 50.2)
c1 = c(17, 11, 8, 10)
c2 = c(11, 17, 9, 8)
c3 = c(8, 9, 17, 9)
c4 = c(10, 8, 9, 17)
S = rbind(c1, c2, c3, c4)
```

```
y = mu + ranMVN(nRep = 25, S, seed = NULL)
```

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ranMVN2

*Multivariate Normal Data*

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**Description**

Generates multivariate normal data based on the Cholesky root of the covariance matrix.

**Usage**

```
ranMVN2(nRep = 1, rootS, seed = NULL)
```

**Arguments**

nRep	Numer of clusters (replications)
rootS	Cholesky root of the desired covariance matrix
seed	Initialize random number generator

**Details**

The matrix of response, call it Y, has mean 0.

**Value**

Returns a matrix of response variates of dimension nRep by nrow(rootS)

**See Also**

See Also [ranMVN](#), [ranMvnXch](#), [ep](#)

**Examples**

```
mu = c(25.3, 28.4, 35.7, 50.2)
c1 = c(17, 11, 8, 10)
c2 = c(11, 17, 9, 8)
c3 = c(8, 9, 17, 9)
c4 = c(10, 8, 9, 17)
S = rbind(c1, c2, c3, c4)

rootS=chol(S)

y = mu + ranMVN2(nRep = 25, rootS, seed = NULL)
```

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ranMvnXch

*Multivariate Normal Data*

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**Description**

Generates multivariate normal data based on an EXCHANGEABLE correlation matrix.

**Usage**

```
ranMvnXch(rho, n, nRep = 1, seed = NULL)
```

**Arguments**

rho	correlation
n	cluster size
nRep	number of clusters (replications)s
seed	initializes the random number generator

**Details**

This is more efficient than ranMVN and ranMVN2 for exchangeable correlation. The returned matrix of responses, call it Y, has mean 0. We require that rho is greater than zero.

**Value**

Returns a matrix of response variates of dimension nRep by n

**See Also**

See Also [ranMVN](#), [ranMVN2](#), [ep](#)

# Index

\*Topic **datagen**

ep, [2](#)

ranMVN, [4](#)

ranMVN2, [5](#)

ranMvnXch, [6](#)

\*Topic **distribution**

ep, [2](#)

ranMVN, [4](#)

ranMVN2, [5](#)

ranMvnXch, [6](#)

\*Topic **multivariate**

ep, [2](#)

ranMVN, [4](#)

ranMVN2, [5](#)

ranMvnXch, [6](#)

ep, [2](#), [5](#), [6](#)

mvtBinaryEP-package, [1](#)

ranMVN, [3](#), [4](#), [5](#), [6](#)

ranMVN2, [3](#), [5](#), [5](#), [6](#)

ranMvnXch, [3](#), [5](#), [6](#)