

Package ‘copulaSim’

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Type Package

Title Virtual Patient Simulation by Copula Invariance Property

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Description To optimize clinical trial designs and data analysis methods consistently through trial simulation, it is important to simulate multivariate mixed-type virtual patient data independent of designs and analysis methods under evaluation. Using copula invariance property, we generate virtual patient data based on the dependence structure of observed patient-level data. The function `copula.sim` can generate virtual patient data with optional data validation methods that are based on energy distance and ball divergence measurement. The function `compare.copula.sim` can conduct comparison of marginal mean and covariance of simulated data. To simulate patient-level data from a hypothetical treatment arm that would perform differently from the observed data, the function `new.arm.copula.sim` can be used to generate new multivariate data with the same dependence structure of the original data but with a shifted mean vector.

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Depends R (>= 4.0.0)

Imports dplyr (>= 1.0.0),
magrittr (>= 1.5),
mvtnorm (>= 1.0-12),
rlang,
stats,
tibble,
utils

Suggests testthat (>= 3.1.1),
Ball (>= 1.3.0),
energy (>= 1.7-0),
TwoCop

URL <https://github.com/psyen0824/CopulaSim>

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copulaSim-package	<i>copulaSim: Virtual Patient Simulation by Copula Invariance Property</i>
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Description

To optimize clinical trial designs and data analysis methods consistently through trial simulation, it is important to simulate multivariate mixed-type virtual patient data independent of designs and analysis methods under evaluation. Using copula invariance property, we generate virtual patient data based on the dependence structure of observed patient-level data. The function `copula.sim` can generate virtual patient data with optional data validation methods that are based on energy distance and ball divergence measurement. The function `compare.copula.sim` can conduct comparison of marginal mean and covariance of simulated data. To simulate patient-level data from a hypothetical treatment arm that would perform differently from the observed data, the function `new.arm.copula.sim` can be used to generate new multivariate data with the same dependence structure of the original data but with a shifted mean vector.

Author(s)

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See Also

Useful links:

- <https://github.com/psyen0824/ColpulaSim>

compare.copula.sim	<i>Performing the comparison between empirical data and multiple simulated datasets.</i>
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Description

Performing the comparison between empirical data and multiple simulated datasets.

Usage

```
compare.copula.sim(object)
```

Arguments

object A copula.sim object for the comparison.

Value

Returned the comparison of marginal parameter and covariance.

1. mean.comparison: comparison between empirical mean and average value of simulated mean.
(1) simu.mean: average value of simulated mean (2) simu.sd: average value of simulated standard error (3) simu.mean.RB: relative bias for marginal mean (4) simu.mean.SB: standardized bias for marginal mean (5) simu.mean.RMSE: root mean square error for marginal mean
2. cov.comparison: comparison between empirical covariance and average value of simulated covariance

Author(s)

Pei-Shan Yen, Xuemin Gu

copula.sim	<i>To generate simulated datasets from empirical data by utilizing the copula invariance property.</i>
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Description

Based on the empirical data, generating simulated datasets through the copula invariance property.

Usage

```
copula.sim(
  data.input,
  id.vec,
  arm.vec,
  n.patient,
  n.simulation,
  seed = NULL,
  validation.type = "none",
  validation.sig.lvl = 0.05,
```

```

std.norm.lb = -3,
std.norm.ub = 3,
rmvnorm.matrix.decomp.method = "svd",
verbose = TRUE
)

```

Arguments

<code>data.input</code>	The empirical patient-level data to be used to simulate new virtual patient data.
<code>id.vec</code>	The ID for individual patient in the input data.
<code>arm.vec</code>	The column to identify the arm in clinical trial.
<code>n.patient</code>	The targeted number of patients in each simulated datasets.
<code>n.simulation</code>	The number of simulated datasets.
<code>seed</code>	The random seed. Default is NULL to use the current seed.
<code>validation.type</code>	A string to specify the hypothesis test used to detect the difference between input data and the simulated data. Default is "none". Possible methods are energy distance ("energy"), ball divergence ("ball") and the equability of dependence ("equalCopula"), "energy", "Ball, and "TwoCop" R packages are needed.
<code>validation.sig.lvl</code>	The significant level (alpha) value for the hypothesis test.
<code>std.norm.lb</code>	The minimum quantile value of standardized normal value when performing quantile transformation of empirical data.
<code>std.norm.ub</code>	The maximum quantile value of standardized normal value when performing quantile transformation of empirical data.
<code>rmvnorm.matrix.decomp.method</code>	The method to do the matrix decomposition used in the function <code>rmvnorm</code> . Default is "svd".
<code>verbose</code>	A Boolean value to specify whether to print message for simulation process or not.

Value

A `copula.sim` object with three elements.

1. `data.input`: empirical data
2. `data.transform`: quantile transformation of `data.input`
3. `data.simul`: simulated data

Author(s)

Pei-Shan Yen, Xuemin Gu

References

- Sklar, A. (1959). Functions de repartition an dimensionset leursmarges., Paris: PublInst Stat.
- Nelsen, R. B. (2007). An introduction to copulas. Springer Science & Business Media.
- Ross, S. M. (2013). Simulation. Academic Press.

Examples

```
library(copulaSim)

## Generate Empirical Data - assume the 2-arm empirical data follows multivariate normal data.
library(mvtnorm)
arm1 <- rmvnorm(n = 40, mean = rep(10, 5), sigma = diag(5))
arm2 <- rmvnorm(n = 40, mean = rep(12, 5), sigma = diag(5))
test_data <- as.data.frame(cbind(1:80, rep(1:2, each = 40), rbind(arm1, arm2)))
colnames(test_data) <- c("id", "arm", paste0("outcome_", 1:5))

## Generate 100 simulated datasets
copula.sim(data.input = test_data[, -c(1, 2)], id.vec = test_data[, 1], arm.vec = test_data[, 2],
n.patient = 100, n.simulation = 100, seed = 2022)
```

data.diff.test	<i>Performing the hypothesis test to compare the difference between the empirical data and the simulated data</i>
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Description

Performing the hypothesis test to compare the difference between the empirical data and the simulated data

Usage

```
data.diff.test(x, y, test.method)
```

Arguments

x	A numeric matrix.
y	A numeric matrix which is compared to x.
test.method	A string to specify the hypothesis test used to detect the difference between input data and the simulated data. Default is "none". Possible methods are energy distance ("energy"), ball divergence ("ball") and the equability of dependence ("equalCopula"), "energy", "Ball", and "TwoCop" R packages are needed.

Value

A list with two elements.

1. p.value: the p-value of the hypothesis test.
2. test.result: the returned object of the hypothesis test.

ecdf.inv	<i>Obtaining the inverse of marginal empirical cumulative distribution (ECDF)</i>
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Description

Obtaining the inverse of marginal empirical cumulative distribution (ECDF)

Usage

```
ecdf.inv(x, p, sort.flag = TRUE)
```

Arguments

x	A vector of numbers which is the marginal empirical data.
p	A vector of numbers which is the probability of the simulated data.
sort.flag	A boolean value to specify whether to sort the output data.

Value

The inverse values of p based on ECDF of x.

Examples

```
ecdf.inv(0:10, c(0.25, 0.75))
ecdf.inv(0:10, c(0.25, 0.75), FALSE)
```

extract.data.sim	<i>Converting data.simul in a copula.sim object into a list of wide-form matrices</i>
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Description

Converting data.simul in a copula.sim object into a list of wide-form matrices

Usage

```
extract.data.sim(object)
```

Arguments

object	A copula object.
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Value

A list of matrices for simulated data.

new.arm.copula.sim	<i>Simulating new multivariate datasets with shifted mean vector from existing empirical data</i>
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Description

Simulating new multivariate datasets with shifted mean vector from existing empirical data

Usage

```
new.arm.copula.sim(
  data.input,
  id.vec,
  arm.vec,
  shift.vec.list,
  n.patient,
  n.simulation,
  seed = NULL,
  validation.type = "none",
  validation.sig.lvl = 0.05,
  std.norm.lb = -3,
  std.norm.ub = 3,
  rmvnorm.matrix.decomp.method = "svd",
  verbose = TRUE
)
```

Arguments

`data.input`, `id.vec`, `arm.vec`, `n.patient`, `n.simulation`, `seed`, `std.norm.lb`, `std.norm.ub`
 Please refer to the function [copula.sim](#).

`shift.vec.list` A list of numeric vectors to specify the mean-shifted values for new arms.

`validation.type`, `validation.sig.lvl`, `rmvnorm.matrix.decomp.method`, `verbose`
 Please refer to the function [copula.sim](#).

Value

Please refer to the function [copula.sim](#).

Author(s)

Pei-Shan Yen, Xuemin Gu, Jenny Jiao, Jane Zhang

Examples

```
library(copulaSim)

## Generate Empirical Data - assume the 1-arm empirical data follows multivariate normal data.
library(mvtnorm)
arm1 <- rmvnorm(n = 80, mean = rep(10, 5), sigma = diag(5))
test_data <- as.data.frame(cbind(1:80, rep(1,80), arm1))
colnames(test_data) <- c("id", "arm", paste0("outcome_", 1:5))
```

```
## Generate 1 simulated datasets with one empirical arm and two new-arm.  
new.arm.copula.sim(data.input = test_data[, -c(1,2)], id.vec = test_data[, 1],  
  arm.vec = test_data[, 2], n.patient = 100 , n.simulation = 1, seed = 2022,  
  shift.vec.list = list(rep(1.5,5), rep(3,5) ))
```


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