

# Package ‘latentcor’

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**Type** Package

**Title** Fast Computation of Latent Correlations for Mixed Data

**Version** 1.2.0

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**Description** The first stand-alone R package for computation of latent correlation that takes into account all variable types (continuous/binary/ordinal/zero-inflated), comes with an optimized memory footprint, and is computationally efficient, essentially making latent correlation estimation almost as fast as rank-based correlation estimation. The estimation is based on latent copula Gaussian models.  
For continuous/binary types, see Fan, J., Liu, H., Ning, Y., and Zou, H. (2017) <[doi:10.1111/rssb.12168](https://doi.org/10.1111/rssb.12168)>. For ternary type, see Quan X., Booth J.G. and Wells M.T. (2018) <[arXiv:1809.06255](https://arxiv.org/abs/1809.06255)>. For truncated type or zero-inflated type, see Yoon G., Carroll R.J. and Gaynanova I. (2020) <[doi:10.1093/biomet/asaa007](https://doi.org/10.1093/biomet/asaa007)>. For approximation method of computation, see Yoon G., Müller C.L. and Gaynanova I. (2021) <[doi:10.1080/10618600.2021.1882468](https://doi.org/10.1080/10618600.2021.1882468)>.

**Depends** R (>= 3.0.0)

**Imports** stats, pcaPP, fMultivar, mnormt, chebpol, Matrix, MASS, heatmaply, ggplot2, plotly, graphics

**License** GPL-3

**Encoding** UTF-8

**RoxygenNote** 7.1.2

**Suggests** microbenchmark, rmarkdown, markdown, knitr, testthat (>= 3.0.0), covr

**Config/testthat/edition** 3

**VignetteBuilder** knitr

**NeedsCompilation** no

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gen_data	<i>Mixed type simulation data generator</i>
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### Description

Generates data of mixed types from the latent Gaussian copula model.

### Usage

```
gen_data(
  n = 100,
  types = c("ter", "con"),
  rhos = 0.5,
  copulas = "no",
  XP = NULL,
  showplot = FALSE
)
```

### Arguments

n	A positive integer indicating the sample size. The default value is 100.
types	A vector indicating the type of each variable, could be "con" (continuous), "bin" (binary), "tru" (truncated) or "ter" (ternary). The number of variables is determined based on the length of types, that is $p = \text{length}(\text{types})$ . The default value is <code>c("ter", "con")</code> which creates two variables: the first one is ternary, the second one is continuous.
rhos	A vector with lower-triangular elements of desired correlation matrix, e.g. <code>rhos = c(.3, .5, .7)</code> means the correlation matrix is <code>matrix(c(1, .3, .5, .3, 1, .7, .5, .7, 1), 3, 3)</code> . If only a scalar is supplied ( <code>length(rhos) = 1</code> ), then equi-correlation matrix is assumed with all pairwise correlations being equal to rhos. The default value is 0.5 which means correlations between any two variables are 0.5.
copulas	A vector indicating the copula transformation $f$ for each of the $p$ variables, e.g. $U = f(Z)$ . Each element can take value "no" ( $f$ is identity), "expo" (exponential transformation) or "cube" (cubic transformation). If the vector has length 1, then the same transformation is applied to all $p$ variables. The default value is "no": no copula transformation for any of the variables.

XP	A list of length p indicating proportion of zeros (for binary and truncated), and proportions of zeros and ones (for ternary) for each of the variables. For continuous variable, NA should be supplied. If NULL, the following values are automatically generated as elements of XP list for the corresponding data types: For continuous variable, the corresponding value is NA; for binary or truncated variable, the corresponding value is a number between 0 and 1 representing the proportion of zeros, the default value is 0.5; for ternary variable, the corresponding value is a pair of numbers between 0 and 1, the first number indicates the proportion of zeros, the second number indicates the proportion of ones. The sum of a pair of numbers should be between 0 and 1, the default value is $c(0.3, 0.5)$ .
showplot	Logical indicator. If TRUE, generates the plot of the data when number of variables p is no more than 3. The default value is FALSE.

### Value

gen\_data returns a list containing

- X: Generated data matrix (n by p) of observed variables.
- plotX: Visualization of the data matrix X. Histogram if p=1. 2D Scatter plot if p=2. 3D scatter plot if p=3. Returns NULL if showplot = FALSE.

### References

- Fan J., Liu H., Ning Y. and Zou H. (2017) "High dimensional semiparametric latent graphical model for mixed data" doi: [10.1111/rssb.12168](https://doi.org/10.1111/rssb.12168).
- Yoon G., Carroll R.J. and Gaynanova I. (2020) "Sparse semiparametric canonical correlation analysis for data of mixed types" doi: [10.1093/biomet/asaa007](https://doi.org/10.1093/biomet/asaa007).

### Examples

```
# Generate single continuous variable with exponential transformation (always greater than 0)
# and show histogram.
simdata = gen_data(n = 100, copulas = "expo", types = "con", showplot = FALSE)
X = simdata$X; plotX = simdata$plotX
# Generate a pair of variables (ternary and continuous) with default proportions
# and without copula transformation.
simdata = gen_data()
X = simdata$X
# Generate 3 variables (binary, ternary and truncated)
# corresponding copulas for each variables are "no" (no transformation),
# "cube" (cube transformation) and "cube" (cube transformation).
# binary variable has 30% of zeros, ternary variable has 20% of zeros
# and 40% of ones, truncated variable has 50% of zeros.
# Then show the 3D scatter plot (data points project on either 0 or 1 on Axis X1;
# on 0, 1 or 2 on Axis X2; on positive domain on Axis X3)
simdata = gen_data(n = 100, rhos = c(.3, .4, .5), copulas = c("no", "cube", "cube"),
  types = c("bin", "ter", "tru"), XP = list(.3, c(.2, .4), .5), showplot = TRUE)
X = simdata$X; plotX = simdata$plotX
# Check the proportion of zeros for the binary variable.
```

```

sum(simdata$X[ , 1] == 0)
# Check the proportion of zeros and ones for the ternary variable.
sum(simdata$X[ , 2] == 0); sum(simdata$X[ , 2] == 1)
# Check the proportion of zeros for the truncated variable.
sum(simdata$X[ , 3] == 0)

```

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get_types	<i>Automatically determine types of each variable (continuous/binary/ternary/truncated) in a data matrix.</i>
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### Description

Automatically determine types of each variable (continuous/binary/ternary/truncated) in a data matrix.

### Usage

```
get_types(X, tru_prop = 0.05)
```

### Arguments

X	A numeric data matrix (n by p), where n is number of samples, and p is number of variables. Missing values (NA) are allowed.
tru_prop	A scalar between 0 and 1 indicating the minimal proportion of zeros that should be present in a variable to be treated as "tru" (truncated type or zero-inflated) rather than as "con" (continuous type). The default value is 0.05 (any variable with more than 5% of zero values among n samples is treated as truncated or zero-inflated)

### Value

get\_types returns

- types: A vector of length p indicating the type of each of the p variables in X. Each element is one of "con" (continuous), "bin" (binary), "ter" (ternary) or "tru" (truncated).

### Examples

```

X = gen_data(types = c("ter", "con"))$X
get_types(X)

```

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latentcor	<i>Estimate latent correlation for mixed types.</i>
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### Description

Estimation of latent correlation matrix from observed data of (possibly) mixed types (continuous/binary/truncated/ternary) based on the latent Gaussian copula model. Missing values (NA) are allowed. The estimation is based on pairwise complete observations.

### Usage

```
latentcor(
  X,
  types = NULL,
  method = c("approx", "original"),
  use.nearPD = TRUE,
  nu = 0.001,
  tol = 1e-08,
  ratio = 0.9,
  showplot = FALSE
)
```

### Arguments

X	A numeric matrix or numeric data frame (n by p), where n is number of samples, and p is number of variables. Missing values (NA) are allowed, in which case the estimation is based on pairwise complete observations.
types	A vector of length p indicating the type of each of the p variables in X. Each element must be one of "con" (continuous), "bin" (binary), "ter" (ternary) or "tru" (truncated). If the vector has length 1, then all p variables are assumed to be of the same type that is supplied. The default value is NULL, and the variable types are determined automatically using function <code>get_types</code> . As automatic determination of variable types takes extra time, it is recommended to supply the types explicitly when they are known in advance.
method	The calculation method for latent correlations. Either "original" or "approx". If method = "approx", multilinear approximation method is used, which is much faster than the original method, see Yoon et al. (2021) for timing comparisons for various variable types. If method = "original", optimization of the bridge inverse function is used. The default is "approx".
use.nearPD	Logical indicator. use.nearPD = TRUE gets nearest positive definite matrix for the estimated latent correlation matrix with shrinkage adjustment by nu. Output R is the same as Rpointwise if use.nearPD = FALSE. Default value is TRUE.
nu	Shrinkage parameter for the correlation matrix, must be between 0 and 1. Guarantees that the minimal eigenvalue of returned correlation matrix is greater or equal to nu. When nu = 0, no shrinkage is performed, the returned correlation matrix will be semi-positive definite but not necessarily strictly positive definite.

	When $\nu = 1$ , the identity matrix is returned (not recommended). The default (recommended) value is 0.001.
tol	When method = "original", specifies the desired accuracy of the bridge function inversion via uniroot optimization and is passed to <code>optimize</code> . The default value is 1e-8. When method = "approx", this parameter is ignored.
ratio	When method = "approx", specifies the boundary value for multilinear interpolation, must be between 0 and 1. When ratio = 0, no linear interpolation is performed (the slowest execution) which is equivalent to method = "original". When ratio = 1, linear interpolation is always performed (the fastest execution) but may lead to high approximation errors. The default (recommended) value is 0.9 which controls the approximation error and has fast execution, see Yoon et al. (2021) for details. When method = "original", this parameter is ignored.
showplot	Logical indicator. showplot = TRUE generates a ggplot object plotR with the heatmap of latent correlation matrix R. plotR = NULL if showplot = FALSE. Default value is FALSE.

## Details

The function estimates latent correlation by calculating inverse bridge function (Fan et al., 2017) evaluated at the value of sample Kendall's tau ( $\hat{\tau}$ ). The bridge function F connects Kendall's tau to latent correlation r so that  $F(r) = E(\hat{\tau})$ . The form of function F depends on the variable types (continuous/binary/truncated/ternary), but is exact. The exact form of inverse is not available, so has to be evaluated numerically for each pair of variables leading to Rpointwise.

When method = "original", the inversion is done numerically by solving

$$\text{minimize}_r (F(r) - \hat{\tau})^2$$

using `optimize`. The parameter tol is used to control the accuracy of the solution.

When method = "approx", the inversion is done approximately by interpolating previously calculated and stored values of  $F^{-1}(\hat{\tau})$ . This is significantly faster than the original method (Yoon et al., 2021) for binary/ternary/truncated cases, however the approximation errors may be non-negligible on some regions of the space. The parameter ratio controls the region where the interpolation is performed with default recommended value of 0.9 giving a good balance of accuracy and computational speed. Increasing the value of ratio may improve speed (but possibly sacrifice the accuracy), whereas decreasing the value of ratio may improve accuracy (but possibly sacrifice the speed). See Yoon et al. 2021 and vignette for more details.

In case the pointwise estimator Rpointwise is has negative eigenvalues, it is projected onto the space of positive semi-definite matrices using `nearPD`. The parameter nu further allows to perform additional shrinkage towards identity matrix (desirable in cases where the number of variables p is very large) as

$$R = (1 - \nu)\tilde{R} + \nu I,$$

where  $\tilde{R}$  is Rpointwise after projection by `nearPD`.

## Value

latentcor returns

- `zratios`: A list of of length  $p$  corresponding to each variable. Returns NA for continuous variable; proportion of zeros for binary/truncated variables; the cumulative proportions of zeros and ones (e.g. first value is proportion of zeros, second value is proportion of zeros and ones) for ternary variable.
- `K`:  $(p \times p)$  Kendall Tau (Tau-a) Matrix for  $X$
- `R`:  $(p \times p)$  Estimated latent correlation matrix for  $X$
- `Rpointwise`:  $(p \times p)$  Point-wise estimates of latent correlations for  $X$ . This matrix is not guaranteed to be semi-positive definite. This is the original estimated latent correlation matrix without adjustment for positive-definiteness.
- `plotR`: Heatmap plot of latent correlation matrix  $R$ , NULL if `showplot = FALSE`

## References

Fan J., Liu H., Ning Y. and Zou H. (2017) "High dimensional semiparametric latent graphical model for mixed data" doi: [10.1111/rssb.12168](https://doi.org/10.1111/rssb.12168).

Yoon G., Carroll R.J. and Gaynanova I. (2020) "Sparse semiparametric canonical correlation analysis for data of mixed types" doi: [10.1093/biomet/asaa007](https://doi.org/10.1093/biomet/asaa007).

Yoon G., Müller C.L., Gaynanova I. (2021) "Fast computation of latent correlations" doi: [10.1080/10618600.2021.1882468](https://doi.org/10.1080/10618600.2021.1882468).

## Examples

```
# Example 1 - truncated data type, same type for all variables
# Generate data
X = gen_data(n = 300, types = rep("tru", 5))$X
# Estimate latent correlation matrix with original method and check the timing
start_time = proc.time()
R_nc_org = latentcor(X = X, types = "tru", method = "original")$R
proc.time() - start_time
# Estimate latent correlation matrix with approximation method and check the timing
start_time = proc.time()
R_nc_approx = latentcor(X = X, types = "tru", method = "approx")$R
proc.time() - start_time
# Heatmap for latent correlation matrix.
Heatmap_R_nc_approx = latentcor(X = X, types = "tru", method = "approx",
                                showplot = TRUE)$plotR

# Example 2 - ternary/continuous case
X = gen_data()$X
# Estimate latent correlation matrix with original method
R_nc_org = latentcor(X = X, types = c("ter", "con"), method = "original")$R
# Estimate latent correlation matrix with approximation method
R_nc_approx = latentcor(X = X, types = c("ter", "con"), method = "approx")$R
```

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