# Package 'RMAWGEN'

July 21, 2025

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```
License GPL (>= 2)
Title Multi-Site Auto-Regressive Weather GENerator
Type Package
Description S3 and S4 functions are implemented for spatial multi-site
      stochastic generation of daily time series of temperature and
      precipitation. These tools make use of Vector AutoRegressive models (VARs).
      The weather generator model is then saved as an object and is calibrated by
      daily instrumental ``Gaussianized" time series through the 'vars' package
      tools. Once obtained this model, it can it can be used for weather
      generations and be adapted to work with several climatic monthly time
      series.
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```

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## **Description**

RMAWGEN-package

Multi-site autoregressive Models for Daily Weather Generation. The modeling in climate change applications for agricultural or hydrological purposes often requires daily time-series of precipitation and temperature. This is the case of downscaled series from monthly or seasonal predictions of Global Climate Models (GCMs). The R package RMAWGEN (R Multi-Sites Auto regressive Weather GENerator) is built to generate daily temperature and precipitation time series in several sites by using the theory of vectorial autoregressive models (VAR). The VAR model is used because it is able to maintain the temporal and spatial correlations among the several series. In particular, observed time series of daily maximum and minimum temperature and precipitation are used to calibrate the parameters of a VAR model (saved as "GPCAvarest2" or "varest2" classes, which inherit the "varest" S3 class defined in the package vars [Pfaff, 2008]). Therefore the VAR model, coupled with monthly mean weather variables downscaled by GCM predictions, allows to generate several stochastic daily scenarios. The structure of the package consists in functions that transform precipitation and temperature time series into Gaussian-distributed random variables through deseasonalization and Principal Component Analysis. Then a VAR model is calibrated on transformed time series. The time series generated by VAR are then inversely re transformed into precipitation and/or temperature series. An application dateset is included in the RMAWGEN package as an example; it is presented by using a dataset with daily weather time series recorded in 59 different sites of Trentino (Italy) and its neighborhoods for the period 1958-2007. The software is distributed as a Free Software with General Public License (GPL) and is available on CRAN and Github. A presentation of the package is available on https://docs.google.com/file/d/

R - Multi-site Autoregressive WEather Generator

0B66otCUk3Bv6V3RPbm1mUG4zVHc/edit. Example script files about package usage are available on https://github.com/ecor/RMAWGENCodeCorner.

# **Details**

 Package:
 RMAWGEN

 Type:
 Package

 Version:
 1.3.6

 Date:
 2019-11-13

 License:
 GPL (>= 2)

 LazyLoad:
 yes

Depends: R(>=2.12),time,chron,vars

#### Note

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## Author(s)

Emanuele Cordano <emanuele.cordano@gmail.org>, Emanuele Eccel <emanuele.eccel@fmach.it>

# References

Cordano E. and Eccel E. (2016), Tools for stochastic weather series generation in R environment, Italian Journal of Agrometeorology doi:10.19199/2016.3.20385625.031

Pfaff B. (2008). VAR, SVAR and SVEC Models: Implementation Within R Package vars. Journal of Statistical Software 27(4). https://www.jstatsoft.org/v27/i04/(doi:10.18637/jss.v027.i04)

acvWGEN 5

acvWGEN	Plots the auto- and cross- covariance functions between measured and simulated data for several stations

# Description

Plots the auto- and cross- covariance functions between measured and simulated data for several stations

# Usage

```
acvWGEN(measured, simulated, titles = c("Sim.", "Mes."), station = NULL)
```

# Arguments

measured matrix containing measured time series simulated matrix containing simulated time series

titles title suffixes for the simulated and measured data respectively c("Sim.","Mes.") station string vector containing the IDs of the meteorological stations where the auto-

covariance is calculated. If it is NULL (default) all stations (corresponding to the

columns of "simulated" and "measured") are applied

#### Value

0 in case of success

## Note

It uses acf function

adddate	Inserts three columns (year,month,day) passing dates to a matrix or to a dataframe

# **Description**

Inserts three columns (year,month,day) passing dates to a matrix or to a dataframe

## Usage

```
adddate(data, origin = "1961-1-1")
```

# **Arguments**

data matrix of daily data

origin character string containing the date of the first row of data as YYYY-MM-DD

6 addsuffixes

#### Value

a data frame with dates and data values

#### See Also

findDate

addsuffixes

Adds suffixes for daily maximum and minimum temperature to the names of a column data frame

# Description

Adds suffixes for daily maximum and minimum temperature to the names of a column data frame

# Usage

```
addsuffixes(
  names = c("T0001", "T0099", "T0001", "T0099"),
  suffix = c("_Tx", "_Tn"),
  sep = ""
)
```

# **Arguments**

names a character string vector with column names

suffix suffixes to add to the first and second groups of column names respectively

sep separation element

#### **Details**

This function is used for data frames with duplicated field names

#### Value

the vector of names with suffixes added

# See Also

```
getVARmodel
```

# **Examples**

```
names <- addsuffixes()</pre>
```

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arch_test	arch.test function for varest2 object

# **Description**

```
arch.test function for varest2 object
```

#### Usage

```
arch_test(object, interval = NULL, overlap = 20, list.output = FALSE, ...)
```

# **Arguments**

object	a varest2 object
interval	string or subset interval of time (e.g. days) or length of this subset interval to which the ARCH test is applied (see Note). Default is NULL.
overlap	number of time instants (e.g. days) which are overlapped on two different subsequent intervals. Default is 20. It is used only if interval has length 1.
list.output	logical value. If TRUE the function returns a list of the test results of each interval. It is used if interval is not NULL. Default is FALSE.
	further arguments for arch.test

#### **Details**

This function is a wrapper of arch.test. It can compute the test also for some subsets (intervals) of the time-series or for all the time-series divided in overlapping intervals. The intervals considered for the ARCH test are defined with the argument interval. If interval is an integer number instead of a vector, it indicates the length of the intervals in which the time-series is split. If interval is set to NULL, the test is done on the comprehensive residual time-series without splitting.

#### Value

One object or a list of objects with class attribute varcheck as reported in arch.test

#### See Also

```
arch.test
```

collinear\_dataset

Collinear Dataset

# Description

It is an artificial example dataset containing 16 variables with collinearity among some of them.

# Usage

```
data(collinear_dataset)
```

#### **Format**

Data frame

#### **Details**

The user can easily use the package with his/her own data after replacing the values of such variables.

#### Source

This dataset is intended for research purposes only, being distributed in the hope that it will be useful, but WITHOUT ANY WARRANTY.

 ${\tt Comprehensive Precipitation Generator}$ 

The comprehensive Precipitation Generator

# Description

The comprehensive Precipitation Generator

# Usage

```
ComprehensivePrecipitationGenerator(
   station = c("T0001", "T0010", "T0099"),
   prec_all,
   mean_climate_prec = NULL,
   year_max = 1990,
   year_min = 1961,
   leap = TRUE,
   nmonth = 12,
   cpf = NULL,
   verbose = TRUE,
   p = 1,
```

```
type = "none",
  lag.max = NULL,
  ic = "AIC",
  activateVARselect = FALSE,
  exogen = NULL,
  exogen_sim = NULL,
  is_exogen_gaussian = FALSE,
  year_max_sim = year_max,
  year_min_sim = year_min,
 mean_climate_prec_sim = NULL,
  onlygeneration = FALSE,
  varmodel = NULL,
  type_quantile = 3,
  qnull = NULL,
  valmin = 0.5,
  step = 0,
  n_{GPCA_iteration} = 0,
  n_GPCA_iteration_residuals = n_GPCA_iteration,
  sample = NULL,
  extremes = TRUE,
  exogen_all = NULL,
  exogen_all_col = station,
  no_spline = FALSE,
  nscenario = 1,
  seed = NULL,
 noise = NULL.
  nearPD = FALSE
)
```

#### **Arguments**

station

```
prec_all
                  data frame containing daily precipitation of all meteorological stations. See
                  PRECIPITATION defined in the trentino dataset for formatting.
mean_climate_prec
                  a matrix containing monthly mean daily precipitation for the considered station.
                  If it is NULL, it is calculated. See input of is.monthly.climate
                  start year of the recorded (calibration) period
year_max
                  end year of the recorded (calibration) period
year_min
                  logical variables. If it is TRUE (default)(recommended), leap years are consid-
leap
                  ered, otherwise all years have 365 days
nmonth
                  number of months in one year (default is 12)
cpf
                  see normalizeGaussian_severalstations
verbose
                  logical variable
p, type, lag.max, ic, activateVARselect
                  see respective input parameter on getVARmodel
```

character vector of the IDs of the considered meteorological stations

exogen data frame or matrix containing the (normalized or not) exogenous variables

(predictors) for the recorded (calibration) period.

exogen\_sim data frame or matrix containing the (normalized or not) exogenous variables

(predictors) for the simulation period. Default is NULL. If it is NULL, it is replaced

with exogen within the function.

is\_exogen\_gaussian

logical value. If TRUE, exogen\_sim and exogen are given as already normalized

variables, otherwhise they are not normalized. Default is FALSE

year\_max\_sim last year of the simulation period. Default is equal to year\_max

year\_min\_sim first year of the simulation period. Default is equal to year\_min

mean\_climate\_prec\_sim

a matrix containing monthly mean daily precipitation for the simulation period.

If is NULL (Default), it is set equal to mean\_climate\_prec.

onlygeneration logical value. If TRUE the VAR model varmodel is given as input and only

random generation is done, otherwise (default) is calculated from measured data

varmodel the comprehensinve VAR model as a varest2 S4 object or a NULL object. If

NULL (default), the comprehensinve VAR is estimated from measured data within the function, otherwise it is given as input and only random generation is done.

type\_quantile see type on quantile

step see normalizeGaussian\_severalstations. Default is 0.

n\_GPCA\_iteration

number of iterations of Gaussianization process for data. Default is 0 (no Gaus-

sianization)

n\_GPCA\_iteration\_residuals

number of iterations of Gaussianization process for VAR residuals. Default is 0

(no Gaussianization)

sample, extremes, qnull, valmin

see normalizeGaussian\_severalstations

exogen\_all data frame containing exogenous variable formatted like prec\_all. Default

is NULL. It is alternative to exogen and if it not NULL, is\_exogen\_gaussian is

automatically set FALSE

exogen\_all\_col vector of considered columns of exogen\_all. Default is station.

no\_spline logical value. See splineInterpolateMonthlytoDailyforSeveralYears. De-

fault is TRUE.

nscenario number of generated scenarios for daily maximum and minimum temperature

seed seed for stochastic random generation see set. seed.

noise stochastic noise to add for variabile generation. Default is NULL. See newVARmultieventRealization.

Not used in case that nscenario>1.

nearPD logical. Default is FALSE. See getVARmodel.

#### Value

A list of the following variables:

prec\_mes matrix containing measured daily precipitation (the data is copied by the measured data given as input for the period and the station considered for varmodel estimation)

prec\_spline matrix containing climatic "spline-interpolated" daily preciptation from mean\_climate\_prec

data\_prec matrix containing normalized measured precipitation variable

prec\_gen matrix containing generated daily precipitation [mm]

prec\_spline\_sim matrix containing climatic "spline-interpolated" daily precipitation from mean\_climate\_prec\_sim data\_prec\_gen matrix containing normalized generated precipitation variable

mean\_climate\_prec matrix containing monthly means of daily precipitation (historical scenario)

mean\_climate\_prec\_sim matrix containing monthly means of daily precipitation (predicted/simulated scenario)

var a varest object containing the used VAR model

#### Note

It pre-processes and generates a multi-site precipitation fields. It uses getVARmodel. Detailed examples can be viewed of this function in this presentation. Unfortunately, using this approach, the spatial correlations are underestimated. This is due to the persinstence of zeros in the precipitation records. This problem is known in literature and can be solved in the future versions of RMAW-GEN. See the R code for further details

## Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

 ${\tt splineInterpolateMonthlytoDailyforSeveralYears}$ 

# **Examples**

```
data(trentino)
set.seed(1222) # set the seed for random generations!
year_max <- 1990
year_min <- 1961
year_max_sim <- 1982
year_min_sim <- 1981

n_GPCA_iter <- 2
p <- 1
nscenario=1
station <- c("T0090","T0083")

## Not Run: the call to ComprehensivePrecipitationGenerator may elapse too
## long time (more than 5 eseconds) and is not executed by default CRAN check.
## Please uncomment the following line to run the example on your own PC.</pre>
```

```
generation00 <- ComprehensivePrecipitationGenerator(station=station,
prec_all=PRECIPITATION,year_min=year_min,year_max=year_max,
year_min_sim=year_min_sim,year_max_sim=year_max_sim,p=p,
n_GPCA_iteration=n_GPCA_iter,n_GPCA_iteration_residuals=0,
sample="monthly",nscenario=nscenario,no_spline=TRUE)</pre>
```

ComprehensiveTemperatureGenerator

The Comprehensive Temperature Generator

# **Description**

The Comprehensive Temperature Generator

# Usage

```
ComprehensiveTemperatureGenerator(
  station = c("T0001", "T0010", "T0099"),
 Tx_all,
  Tn_all,
 mean_climate_Tn = NULL,
 mean_climate_Tx = NULL,
 Tx_spline = NULL,
 Tn_spline = NULL,
 year_max = 1990,
 year_min = 1961,
  leap = TRUE,
  nmonth = 12,
  verbose = TRUE,
 p = 1,
  type = "none",
  lag.max = NULL,
  ic = "AIC",
  activateVARselect = FALSE,
  year_max_sim = year_max,
 year_min_sim = year_min,
 mean_climate_Tn_sim = NULL,
 mean_climate_Tx_sim = NULL,
 Tn_spline_sim = NULL,
  Tx_spline_sim = NULL,
  onlygeneration = FALSE,
  varmodel = NULL,
  normalize = TRUE,
  type_quantile = 3,
```

```
sample = NULL,
  extremes = TRUE,
 option = 2,
 yearly = FALSE,
 yearly_sim = yearly,
 n_GPCA_iteration = 0,
  n_GPCA_iteration_residuals = n_GPCA_iteration,
  exogen = NULL,
 exogen_sim = exogen,
  is_exogen_gaussian = FALSE,
  exogen_all = NULL,
  exogen_all_col = station,
  nscenario = 1,
 seed = NULL,
 noise = NULL,
 nearPD = FALSE
)
```

#### Arguments

```
station
                  see respective input parameter on setComprehensiveTemperatureGeneratorParameters
Tx_all, Tn_all, mean_climate_Tn, mean_climate_Tx, Tx_spline, Tn_spline
                  see respective input parameter on setComprehensiveTemperatureGeneratorParameters
year_max, year_min, leap, nmonth, verbose
                  see respective input parameter on setComprehensiveTemperatureGeneratorParameters
p, type, lag.max, ic, activateVARselect
                  see respective input parameter on getVARmodel
                 last year of the simulation period. Default is equal to year_max
year_max_sim
                  first year of the simulation period. Default is equal to year_min
year_min_sim
mean_climate_Tn_sim
                  monthly averaged daily minimum temperatures for the simulated scenario and
                  used by the random generator . Default is mean_climate_Tn
mean_climate_Tx_sim
                  monthly averaged daily maximum temperatures for the simulated scenario and
                  used by the random generator. Default is mean_climate_Tx
Tn_spline_sim
                 daily timeseries (from the first day of year_min_sim to the last day of year_max_sim)
                  of averaged minimum temperature which can be obtained by a spline interpola-
                  tion of monthly mean values (for the generation period). Default is Tn_spline.
                  See for spline interpolation utilized splineInterpolateMonthlytoDailyforSeveralYears.
Tx_spline_sim
                 daily timeseries (from the first day of year_min_sim to the last day of year_max_sim)
                  of averaged maximum temperature which can be obtained by a spline interpola-
                  tion of monthly mean values (for the generation period). Default is Tx_spline.
                  See for spline interpolation utilized spline Interpolate {\tt MonthlytoDaily} for {\tt Several Years}.
onlygeneration logical variable. If TRUE the VAR model varmodel is given as input and only
                  random generation is done, otherwise (default) is calculated from measured data
```

varmodel the comprehensinve VAR model as a varest2 or GPCAvarest2 S4 object or a

NULL object. If NULL (default), the comprehensinve VAR is estimated from measured data within the function, otherwise it is given as input and only random

generation is done.

normalize, sample, extremes

type\_quantile see type on quantile

option integer value. If 1, the generator works with minimun and maximum tempera-

ture, if 2 (default) it works with the average value between maximum and mini-

mum temparature and the respective daily thermal range.

yearly logical value. If TRUE the monthly mean values are calculated for each year from

year\_min to year\_max separately. Default is FALSE.

yearly\_sim logical value. If TRUE the monthly mean values are calculated for each year from

year\_min\_sim to year\_max\_sim separately. Default is yearly.

n\_GPCA\_iteration

number of iterations of Gaussianization process for data. Default is 0 (no Gaus-

sianization)

n\_GPCA\_iteration\_residuals

number of iterations of Gaussianization process for VAR residuals. Default is 0

(no Gaussianization)

exogen data frame or matrix containing the (normalized or not) exogenous variables

(predictors) for the recorded (calibration) period. Default is NULL.

exogen\_sim data frame or matrix containing the (normalized or not) exogenous variables

(predictors) for the simulation period. Default is NULL. If it is NULL, exogen\_sim

is set equal to exogen within the function.

is\_exogen\_gaussian

logical value, If TRUE, exogen\_sim and exogen are given as already normalized

variables, otherwhise they are not normalized. Default is FALSE

exogen\_all data frame containing exogenous variable formatted like Tx\_all and Tn\_all.

Default is NULL. It is alternative to exogen and if it not NULL, is\_exogen\_gaussian

is automatically set to FALSE

exogen\_all\_col vector of considered columns of exogen\_all. Default is station.

nscenario number of generated scenarios for daily maximum and minimum temperature

seed seed for stochastic random generation see set.seed

noise stochastic noise to add for variabile generation. Default is NULL. See newVARmultieventRealization.

Not used in case that nscenario>1.

nearPD logical. Default is FALSE. See getVARmodel.

# Value

A list of the following variables:

input list of variables returned by setComprehensiveTemperatureGeneratorParameters var varest object containing the used VAR model (if useVAR is true), NULL (otherwise)

output list variables returned by generateTemperatureTimeseries (i.e. generated timeseries)

continuity\_ratio 15

#### Note

It pre-processes series and generates multi-site temperature fields by using setComprehensiveTemperatureGeneratorParamand generateTemperatureTimeseries. Detailed examples can be viewed of this function in this presentation.

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

setComprehensiveTemperatureGeneratorParameters, generateTemperatureTimeseries, generateTemperatureTim

# **Examples**

```
data(trentino)
set.seed(1222) # set the seed for random generations!
year_min <- 1961
year_max <- 1990
year_min_sim <- 1982
year_max_sim <- 1983
n_GPCA_iter <- 5
n_GPCA_iteration_residuals <- 5
p < -1
vstation <- c("B2440", "B6130", "B8570", "B9100", "LAVIO", "POLSA", "SMICH", "T0001",
"T0010", "T0014", "T0018", "T0032", "T0064", "T0083", "T0090", "T0092",
"T0094", "T0099", "T0102", "T0110", "T0129", "T0139", "T0147", "T0149",
"T0152", "T0157", "T0168", "T0179", "T0189", "T0193", "T0204", "T0210",
"T0211", "T0327", "T0367", "T0373")
## Not Run: the call to ComprehensiveTemperatureGenerator may elapse
## too long time (more than 5 eseconds) and is not executed by CRAN check.
## Please uncomment the following line to run the example on your own PC.
# generation00 <-ComprehensiveTemperatureGenerator(station=vstation[16],</pre>
# Tx_all=TEMPERATURE_MAX,Tn_all=TEMPERATURE_MIN,year_min=year_min,year_max=year_max,
# p=p,n_GPCA_iteration=n_GPCA_iter,n_GPCA_iteration_residuals=n_GPCA_iteration_residuals,
# sample="monthly",year_min_sim=year_min_sim,year_max_sim=year_max_sim)
```

continuity\_ratio

Calculates the continuity ratio of a set of precipitation measured or generated data in several sites as defined by Wilks, 1998 (see reference link) 16 continuity\_ratio

## **Description**

Calculates the continuity ratio of a set of precipitation measured or generated data in several sites as defined by Wilks, 1998 (see reference link)

## Usage

```
continuity_ratio(data, lag = 0, valmin = 0.5)
```

## **Arguments**

data containing daily precipitation time series for several gauges (one gauge time

series per column)

lag numeric lag (expressed as number of days) used for computation for "cross"

continuity ratio and joint probability of prercipitation (no)occurrence.

valmin threshold precipitation value [mm] for wet/dry day indicator. If precipitation is

lower than valmin, day is considered dry. Default is 0.5 mm.

#### Value

A list containing the following matrices:

continuity\_ratio: lag-day lagged continuity ratio,

occurrence: joint probability of lag-day lagged precipitation occurrence

nooccurrence: joint probability of lag-day lagged no precipitation occurrence.

nooccurrence\_occurrence : joint probability of lag-day lagged no precipitation and precipitation occurrence respectively.

occurrence\_nooccurrence: joint probability of lag-day lagged precipitation and no precipitation occurrence respectively.

probability\_continuity\_ratio: lag-day lagged ratio about precipitation probability contitioned to no precipitation/preciitation occurrence in the other site

#### Note

If lag==0 the function returns the continuity ratio and joint probability as described by Wilks, 1998. Otherwise the precipitation values for each couple of rain gauges are taken with lag-day lag.

#### References

Mhanna, M. and Bauwens, W. (2012), A stochastic space-time model for the generation of daily rainfall in the Gaza Strip. Int. J. Climatol., 32: 1098-1112. doi:10.1002/joc.2305

D.S. Wilks (1998), Multisite generalization of a daily stochastic precipitation generation model, Journal of Hydrology, doi:10.1016/S00221694(98)001863

countNAs 17

#### **Examples**

```
data(trentino)
year_min <- 1961
year_max <- 1990
origin <- paste(year_min,1,1,sep="-")</pre>
period <- PRECIPITATION$year>=year_min & PRECIPITATION$year<=year_max</pre>
station <- names(PRECIPITATION)[!(names(PRECIPITATION) %in% c("day", "month", "year"))]</pre>
prec_mes <- PRECIPITATION[period, station]</pre>
## removing nonworking stations (e.g. time series with NA)
accepted <- array(TRUE,length(names(prec_mes)))</pre>
names(accepted) <- names(prec_mes)</pre>
for (it in names(prec_mes)) {
accepted[it] <- (length(which(!is.na(prec_mes[,it])))==length(prec_mes[,it]))</pre>
prec_mes <- prec_mes[,accepted]</pre>
## the dateset is reduced!!!
prec_mes <- prec_mes[,1:2]</pre>
continuity_ratio <-continuity_ratio(data=prec_mes,lag=0,valmin=0.5)</pre>
continuity_ratio1 <-continuity_ratio(data=prec_mes,lag=-1,valmin=0.5)</pre>
```

countNAs

counts NAs in each row of data

#### **Description**

counts NAs in each row of data

## Usage

```
countNAs(data)
```

# **Arguments**

data a data input matrix
@export

#### Value

the vector with numbers of NA values for each data column

18 covariance

covariance Calculates the covariance matrix of the normally standardized variables obtained from the columns of x

# Description

Calculates the covariance matrix of the normally standardized variables obtained from the columns of x

# Usage

```
covariance(
    x,
    data = x,
    cpf = NULL,
    mean = 0,
    sd = 1,
    step = NULL,
    prec = 10^-4,
    use = "pairwise.complete.obs",
    type = 3,
    extremes = TRUE,
    sample = NULL,
    origin_x = NULL,
    origin_data = origin_x
)
```

# Arguments

X	variable
data	a sample of data on which a non-parametric pghjjrobability distribution is estimated
cpf	cumulative probability distribution. If NULL (default) is calculated as ecdf (data)
mean	mean (expected value) of the normalized random variable. Default is 0.
sd	standard deviation of the normalized random variable. Default is 1.
step	vector of values in which step discontinuities of the cumulative probability function occur. Default is NULL
prec	amplitude of the neighbourhood of the step discontinuities where cumulative probability function is treated as non continuous.
use	see cov
type	see quantile
extremes	logical variable. If TRUE (default) the probability or frequency is multiplied by

$$\frac{N}{N+1}$$

where N is the length of data

ElevationOf 19

sample information about sample or probability distribution. Default is NULL

origin\_x date corresponding to the first row of x origin\_data date corresponding to the first row of data

#### Value

a matrix with the normalized variable or its inverse

# Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

normalizeGaussian\_severalstations,normalizeGaussian

@note It applies normalizeGaussian\_severalstations to x and data and then calculates the covariances among the column. See the R code for further details

ElevationOf Extracts the elevation of a meteorological station expressed in meters

above a reference (sea level)

## Description

Extracts the elevation of a meteorological station expressed in meters above a reference (sea level)

#### **Usage**

ElevationOf(name, station\_names, elevation)

## **Arguments**

name character ID of the station

station\_names vector of the IDs (characters) of the considered meteorological stations. An

example is STATION\_NAMES, which is defined in the trentino dataset.

elevation vector of the elevation of the considered meteorological stations. An example is

ELEVATION, which is defined in the trentino dataset.

#### Value

the elevation given the vectors of station IDs and the respective elevations

## **Examples**

```
data(trentino)
ElevationOf("T0099", station_names=STATION_NAMES, elevation=ELEVATION)
```

20 extractdays

extractdays	Extracts the rows of a matrix corresponding to the requested days (ex-
	pressed as dates YYYY-MM-DD) given the date (origin) of the first row

# Description

Extracts the rows of a matrix corresponding to the requested days (expressed as dates YYYY-MM-DD) given the date (origin) of the first row

# Usage

```
extractdays(
  data = array(1:ndim_max, dim = c(ndim_max, 1)),
  ndim_max = 1e+05,
  when = "1990-1-1",
  origin = "1961-1-1",
  nday = 1
)
```

# **Arguments**

data an input data matrix where each row corresponds to a daily record ndim\_max maximum (integer) number of rows in data where to find when. Default is

100000 and works if data is missing.

when desired dates for which the data are requested origin date corresponding to the first row of data

nday (optional) number of days since when to extract the data

## Value

a matrix containing the requested rows

## Note

```
It uses julian
```

# **Examples**

```
extractdays()
```

extractmonths 21

extractmonths	Extracts the rows of a matrix corresponding to requested months of a year given the date (origin) of the first row

# **Description**

Extracts the rows of a matrix corresponding to requested months of a year given the date (origin) of the first row

# Usage

```
extractmonths(
  data = array(1:ndim_max, dim = c(ndim_max, 1)),
  ndim_max = 1e+05,
  when = c("Dec", "Jan", "Feb"),
  year = NULL,
  origin = "1961-1-1"
)
```

# Arguments

data	an input data matrix where each row corresponds to a daily record
ndim_max	maximum (integer) number of rows in data where to find when. Default is 100000 and works if data is missing.
when	character vactor of months for which the data are required. It must be a subset of c("Jan", "Feb", "Mar", "Apr", "May", "Jun", "Jul", "Aug", "Sep", "Oct", "Nov", "Dec")
year	year(s) when data must be extracted
origin	date corresponding to the first row of data

# Value

a matrix containing the requested rows

## Note

```
It uses months and julian
```

# Author(s)

Emanuele Cordano, Emanuele Eccel

### See Also

```
extractdays
```

22 extractTnFromAnomalies

#### **Examples**

```
extractmonths()

data(trentino)

dates <- sprintf("%02d-%02d",TEMPERATURE_MAX$year,TEMPERATURE_MAX$month,TEMPERATURE_MAX$day)
origin <- dates[1]
out <- extractmonths(data=TEMPERATURE_MAX,origin=origin)</pre>
```

extractTnFromAnomalies

Extracts generated time series of Daily Minimum Temperature from a random multi-realization obtained by generateTemperatureTimeseries function

# **Description**

Extracts generated time series of Daily Minimum Temperature from a random multi-realization obtained by generateTemperatureTimeseries function

# Usage

```
extractTnFromAnomalies(res_multigen, std, SplineAdv)
```

# Arguments

res\_multigen matrix containing standardized values of daily temperature as returned by generateTemperatureTimeser

(first item)

std vector containing standard deviation for each minimun temperature anomalies

SplineAdv matrix containing the averaged daily values of minimum temperature obtained

by a spline interpolation of the monthly climate

#### Value

a matrix with generated minimum temperature

#### Author(s)

Emanuele Cordano, Emanuele Eccel

extractTxFromAnomalies 23

extractTxFromAnomalies

Extracts generated time series of Daily Maximum Temperature from a random multi-realization obtained by generateTemperatureTimeseries function

# Description

Extracts generated time series of Daily Maximum Temperature from a random multi-realization obtained by generateTemperatureTimeseries function

#### Usage

```
extractTxFromAnomalies(res_multigen, std, SplineAdv)
```

# **Arguments**

res_multigen	matrix containing standardized values of daily temperature as returned by generateTemperatureTimese
	(first item)

vector containing standard deviation for each maximum temperature anomalies

SplineAdv matrix containing the averaged values of maximum temperature obtained by a

spline interpolation of monthly climate

#### Value

a matrix with generated maximum temperature

## Author(s)

Emanuele Cordano, Emanuele Eccel

extractyears	Extracts the elements of a data frame corresponding to a period be-
	tween year_min and year_max for the stations listed in station

# **Description**

Extracts the elements of a data frame corresponding to a period between year\_min and year\_max for the stations listed in station

24 findDate

#### Usage

```
extractyears(
  data,
  year_min = 1961,
  year_max = 1990,
  station = c("T0001", "T0014", "T0129")
)
```

# **Arguments**

data a dataframe containing daily data.

year\_min start year

year\_max end year

station character vector of the IDs of the station where the data are required

#### Value

a matrix containing the requested daily data where each day corresponds to a row and each station corresponds to a column

#### Note

The input data frame data must have the following fields: year, month, day, variables\_ID1, variables\_ID2,... where the fields, variables\_ID1, variables\_ID2,... contain the daily variables referred to the respective stations and the field names are replaced with the respective station ID.

findDate Finds the date corresponding a row index of a matrix given the date (origin) of the first row

# **Description**

Finds the date corresponding a row index of a matrix given the date (origin) of the first row

#### Usage

```
findDate(
   k,
   origin = "1961-1-1",
   data.frame = TRUE,
   decimal = FALSE,
   character = FALSE
)
```

forecastEV 25

# **Arguments**

k	integer or decimal value corresponding to number of days since origin
origin	origin date. See also extractdays
data.frame	logical variable. If TRUE (default) the date is returned as data frame (like data in extractyears), otherwise it is returned as character or POSIXct.
decimal	logical variable. If FALSE (default) k is integer and starts from 1, otherwise is consider as the decimal julian day since origin (deprecated)
character	logical variable. It is used if data.frame is FALSE, if it is FALSE, the date is returned as POSIXct, otherwise it is a character in the following form: YYYY-MM-DD

#### Value

the date(s) corresponding to k under different formats

#### Note

It uses functions of time package. It works like an inverse functions of extractdays. If k is a vector, the function returns several dates for each element of k

# See Also

```
date.mdy,extractdays
```

# **Examples**

```
findDate <- findDate(100,origin="1961-1-1",data.frame=FALSE,character=TRUE)</pre>
```

forecastEV	Forecasts the expected value of a VAR realization given the prievious
	one

# Description

Forecasts the expected value of a VAR realization given the prievious one

# Usage

```
forecastEV(var, xprev = NULL, exogen = NULL)
```

# Arguments

var	A VAR model represented by a varest object as returned by ${\tt getVAR model}$ or ${\tt VAR}$
xprev	previous status of the random variable
exogen	vector containing the values of the "exogen" variables (predictor) for the generation

26 forecastResidual

#### Value

a vector of values

#### See Also

forecastResidual

@export

forecastResidual Forecasts the residual value of a VAR realization given the white noise

covariance matrix

# **Description**

Forecasts the residual value of a VAR realization given the white noise covariance matrix

# Usage

```
forecastResidual(var, xprev = NULL, B = NULL)
```

# **Arguments**

var A VAR model represented by a varest object as returned by getVARmodel or

VAR

xprev previous status of the random variable, in this case the "current instant" white-

noise". Default is NULL and then randomly generated.

B matrix of coefficients for the vectorial white-noise component

# Value

a vector of values

#### Author(s)

Emanuele Cordano, Emanuele Eccel

# See Also

forecastEV,NewVAReventRealization

generateTemperatureTimeseries

Returns time series of Daily Maximum and Minimum with a random multi-realization obtained by using newVARmultieventRealization. This function is called by ComprehensiveTemperatureGenerator.

# **Description**

Returns time series of Daily Maximum and Minimum with a random multi-realization obtained by using newVARmultieventRealization. This function is called by ComprehensiveTemperatureGenerator.

# Usage

```
generateTemperatureTimeseries(
  std_tn,
  std_tx,
  SplineTx,
  SplineTn,
  SplineTm,
  SplineDeltaT,
  std_tm,
  var = NULL,
  exogen = NULL,
 normalize = TRUE,
  type = 3,
  extremes = TRUE,
  sample = NULL,
  option = 1,
 original_data,
 origin_x = NULL,
 origin_data = NULL,
  noise = NULL
)
```

# Arguments

 $std\_tn$ 

	stdTn is default, see setComprehensiveTemperatureGeneratorParameters.
std_tx	vector containing standard deviation of daily maximum temperature anomalies. stdTx is default, see setComprehensiveTemperatureGeneratorParameters.
SplineTx	matrix containing the averaged daily maximum temperature obtained by a spline interpolation of monthly means. SplineAdvTx is default, see setComprehensiveTemperatureGenerato
SplineTn	matrix containing the averaged daily minimum temperature obtained by a spline interpolation of monthly means. SplineAdvTn is default, see setComprehensiveTemperatureGenerato
SplineTm	matrix containing the averaged daily "mean" temperature obtained by a spline interpolation of monthly means. SplineAdvTm is default, see setComprehensiveTemperatureGenerato

vector containing standard deviation of daily minimum temperature anomalies.

SplineDeltaT matrix containing the rescaled averaged daily temperature range obtained by a

spline interpolation of monthly means. SplineAdvDelta\_T\_sim/SplineAdvDelta\_T

is default, see setComprehensiveTemperatureGeneratorParameters.

std\_tm vector containing standard deviation of daily "mean" temperature anomalies.

stdTn is default, see setComprehensiveTemperatureGeneratorParameters.

var A VAR model represented by a varest object as returned by getVARmodel or

**VAR** 

exogen see VAR

normalize logical variable If TRUE normalizeGaussian\_severalstations is used, other-

wise not. If option is 2, it is always TRUE.

type see quantile

sample, origin\_x, origin\_data, extremes

see normalizeGaussian\_severalstations

option integer value. If 1, the generator works with minimum and maximum tem-

perature, if 2 (Default) it works with th average value between maximum and

minimum temparature and the respective daily Thermal Range.

original\_data matrix containing the measured standardized temperature anomalies

noise stochastic noise to add for variabile generation. Default is NULL. See newVARmultieventRealization.

#### Value

This function returns a list of the following variables:

res\_multigen matrix containing standardized values of daily maximum and minimum temperature anomalies

Tx\_spline matrix containing climatic "spline-interpolated" daily maximum temperature

Tn\_spine matrix containing climatic "spline-interpolated" daily minimum temperature

Tx\_gen matrix containing generated daily maximum daily temperature  $(Tx_{qen})$ 

Tn\_gen matrix containing generated daily minimum daily temperature  $(Tn_{gen})$ 

Tm\_gen matrix containing generated "mean" daily temperature defined as  $\frac{Tx_{gen}+Tn_{gen}}{2}$ 

DeltaT\_gen matrix containing generated daily thermal range defined as  $Tx_{qen} - Tn_{qen}$ 

See the R code for further details

# Author(s)

Emanuele Cordano, Emanuele Eccel

# See Also

newVARmultieventRealization,normalizeGaussian\_severalstations

getDailyMean 29

getDailyMean	Calculates the daily means of a range of days around each date of a data frame corresponding to a period between year_min and
	year_max for stations listed in station

# **Description**

Calculates the daily means of a range of days around each date of a data frame corresponding to a period between year\_min and year\_max for stations listed in station

# Usage

```
getDailyMean(
  data,
  year_min = 1961,
  year_max = 1990,
  station = c("T0001", "T0010"),
  origin = "1961-1-1",
  lag = 5
)
```

# Arguments

data	a data frame containing daily data.
year_min	start year
year_max	end year
station	character vector of the IDs of the station where the data are requested
origin	origin date of time-series
lag	lag (number of days) on which daily mean is calculated. The mean is calculated considering lag days before and after each day.

# Value

a matrix containing the requested daily mean data where each day corresponds to a row and each station corresponds to a column

# Note

The input data frame data must have the following fields: year, month, day, variables\_ID1, variables\_ID2,... where the fields ,variables\_ID1,variables\_ID2,... contain the daily variables referred to the respective stations and the field names are replaced with the respective station ID.

# Author(s)

Emanuele Cordano, Emanuele Eccel

30 getMonthlyMean

#### See Also

#### extractyears

getMonthlyMean	Calculates the monthly means of a data frame corresponding to a pe-
	riod between year_min and year_max for stations listed in station

## **Description**

@author Emanuele Cordano, Emanuele Eccel

# Usage

```
getMonthlyMean(
  data,
  year_min = 1961,
  year_max = 1990,
  station = names(data),
  no_date = FALSE,
  origin = "1961-1-1",
  yearly = FALSE
)
```

## **Arguments**

data	a dataframe containing daily data.
year_min	start year
year_max	end year
station	character vector of the IDs of the station where the data are requested
no_date	logical value if TRUE the function extractmonths is used. Default is FALSE. It is recommended if data does not contain columns for the dates.
origin	date corresponding to the first row
yearly	logical value. If TRUE the monthly mean values are calculated for each year from year_min to year_max separately. Default is FALSE.

#### Value

a matrix containing the requested monthly means where each month corresponds to a row and each station corresponds to a column or a list of such matrices in case the monthly mean values are calculated separately for each year (if yearly is TRUE)

#### Note

The input data frame data must have the following fields: year, month, day, variables\_ID1, variables\_ID2,... where the fields, variables\_ID1, variables\_ID2,... contain the daily variables referred to the respective stations and the field names are replaced with the respective station ID. In case yearly is TRUE the returned output is a list of matrices whose names are the corresponding year.

getVARmodel 31

## See Also

extractyears

getVARmodel

Either creates a VAR model or chooses a VAR model by using VAR or VARselect commands of vars package

# Description

Either creates a VAR model or chooses a VAR model by using VAR or VARselect commands of vars package

# Usage

```
getVARmodel(
  data,
  suffix = c("_Tx", "_Tn"),
  sep = "",
  p = 1,
  type = "none",
  season = NULL,
  exogen = NULL,
  lag.max = NULL,
  ic = "AIC",
  activateVARselect = FALSE,
  na.rm = TRUE,
  n_{GPCA_{iteration}} = 0,
  n\_GPCA\_iteration\_residuals = n\_GPCA\_iteration,
  extremes = TRUE,
  nearPD = FALSE
)
```

# Arguments

```
data
                 see VAR and addsuffixes
suffix
                 see addsuffixes
                 separator element. See addsuffixes).
sep
                 lag considered for the auto-regression see VAR
                 see VAR
type
                 see VAR
season
exogen
                 see VAR
lag.max
                 see VARselect
ic
                 see VAR
```

32 getVARmodel

activateVARselect

logical variables. If TRUE, the function VARselect is run. Default and recom-

mended use is FALSE.

na.rm logical variables. If TRUE (default), it takes into account NA values

n\_GPCA\_iteration

number of iterations of Gaussianization process for data. Default is 0 (no Gaus-

sianization)

n\_GPCA\_iteration\_residuals

number of iterations of Gaussianization process for data. Default is 0 (no Gaus-

sianization)

extremes see normalizeGaussian\_severalstations and GPCA

nearPD logical (experimental) and passed to GPCA. Default is FALSE. If TRUE covari-

ance matrix is corrected through Nearest Positive Definite Matrix procedure,

i.e. nearPD

#### Value

a varest2 or GPCAvarest2 object representing a VAR model or a GPCA-varest object which also contains the GPCA transformation parameters

#### Note

It inherits input parameters of VAR, VARselect and addsuffixes. The variable data contains the measured data on which the vector auto-regressive models is estimated. It is a matrix where each row is a realization of the vector random variable. In some application of this package, the random variables may be the daily maximum and minimum temperature anomalies for different stations. Often the columns of data are called with the IDs of the stations whithout specifying the type of variable (e.g. minimum or maximum temperature anomalies). This means that two or more columns may have the same name. Therefore the function addsuffixes, which is called from this function, adds suitable suffixes to the column names.

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### **Examples**

GPCA 33

**GPCA** 

This function makes a Gaussianization procedure based on PCA iteration (see GPCA\_iteration)

# Description

This function makes a Gaussianization procedure based on PCA iteration (see GPCA\_iteration)

#### Usage

```
GPCA(x_prev, n = 30, extremes = TRUE, nearPD = FALSE)
```

## **Arguments**

x\_prev previous set of the random variable x. If it is a varest object, the residuals are

taken into account.

n number of reiterations

extremes see normalizeGaussian\_severalstations

nearPD logical. Default is FALSE. If TRUE covariance matrix is corrected through Nearest

Positive Definite Matrix procedure, i.e. nearPD

#### Value

A GPCA-class S3 object returned by GPCA\_iteration at each iteration and the final results of the G-PCA procedure (matrix final\_results)

## Note

This function re-iterates the equation (1) of "PCA Gaussianization for One-Class Remote Sensing Image" by V. Laparra et al., https://www.uv.es/lapeva/papers/SPIE09\_one\_class.pdf,https://www.uv.es/vista/vistavalencia/papers/SPIE\_09\_Gaussianization\_presentation.pdf

# Author(s)

Emanuele Cordano

34 GPCA-class

#### See Also

GPCA,GPCA\_iteration,inv\_GPCA\_iteration,inv\_GPCA,GPCA-class for 'GPCA' S3 class

# **Examples**

```
library(RMAWGEN)
set.seed(1222)
nIterations <- 30
N <- 20
x <- rexp(N)
y <- x+rnorm(N)
df <- data.frame(x=x,y=y)

GPCA <- GPCA(df,n=nIterations,extremes=TRUE)

x <- rnorm(N)
y <- x+rnorm(N)
dfn <- data.frame(x=x,y=y)</pre>
GPCAn <- GPCA(dfn,n=nIterations,extremes=TRUE)
```

GPCA-class

GPCA-class

# Description

GPCA S3 class returned by GPCA

# **Details**

```
list of GPCA_iteration subsequent GPCA iterations
final_results data.frame or matrix of the "gaussianized" data
```

# Note

Formal definition with setOldClass for the S3 class GPCA

#### Author(s)

Emanuele Cordano

# **Examples**

```
showClass("GPCA")
```

GPCAiteration-class 35

GPCAiteration-class GPCAiteration-class

#### **Description**

GPCAiteration S3 class returned by GPCA\_iteration

#### **Details**

```
x_prev Previous set of random variable, x_prev input variable of GPCA_iteration
```

 $\verb|x_gauss_prev| Marginal Gaussianization of \verb|x_prev| obtained through normalize Gaussian_several stations| \\$ 

B\_prev rotation matrix (i. e. eigenvector matrix of the covariance matrix of x\_gauss\_prev)

x\_next results obtained by multiplying B\_prev by x\_gauss\_prev (see equation 1 of the reference in GPCA\_iteration)

## Note

Formal definition with setOldClass for the S3 class GPCAiteration

#### Author(s)

Emanuele Cordano

## **Examples**

showClass("GPCAiteration")

GPCAvarest2-class

GPCAvarest2-class

## **Description**

This class inherits varest2 and contains all information about GPCA (GPCA transformation.

#### **Details**

GPCA\_data: A "GPCA" S3 object containing the parameters of the Multi-variate Gaussianization of the time series, it is the result of GPCA function applied to the input data of getVARmodel

GPCA\_residuals: A "GPCA" S3 object containing the parameters of the Multi-variate Gaussianization of the residuals of the VAR model contained in the VAR slot; it is NULL if no Gaussiatization of residuals is applied. Object of class "list"

VAR: S3 Object of class "varest"

#' @note A GPCAvarest2 object can be created by new("GPCAvarest2", ...) or returned by the function getVARmodel

36 GPCA\_iteration

#### Author(s)

Emanuele Cordano

#### **Examples**

```
showClass("GPCAvarest2")
```

GPCA\_iteration

This function makes an iteration of PCA-Gaussianization process

## Description

This function makes an iteration of PCA-Gaussianization process

# Usage

```
GPCA_iteration(x_prev, extremes = TRUE, nearPD = FALSE)
```

## **Arguments**

x\_prev previous set of random variable x

extremes see normalizeGaussian\_severalstations

nearPD logical. Default is FALSE. If TRUE covariance matrix is corrected through Nearest

Positive Definite Matrix procedure, i.e. nearPD

#### Value

A GPCA\_iteration S3 object which contains the following objects:

x\_prev Previous set of random variable, x\_prev input variable

x\_gauss\_prev Marginal Gaussianization of x\_prev obtained through normalizeGaussian\_severalstations

B\_prev rotation matrix (i. e. eigenvector matrix of the covariance matrix of x\_gauss\_prev

x\_next results obtained by multiplying B\_prev by x\_gauss\_prev (see equation 1 of the reference)

#### Note

This function is based on equation (1) of "PCA Gaussianization for One-Class Remote Sensing Image" by V. Laparra et al., https://www.uv.es/lapeva/papers/SPIE09\_one\_class.pdf and https://ieeexplore.ieee.org/document/5413808/

#### Author(s)

Emanuele Cordano

### See Also

```
GPCA,GPCA_iteration,inv_GPCA_iteration,inv_GPCA
```

inv\_GPCA 37

#### **Examples**

```
library(RMAWGEN)
set.seed(1222)
N <- 20
x <- rexp(N)
y <- x+rnorm(N)
df <- data.frame(x=x,y=y)

GPCA <- GPCA_iteration(df,extremes=TRUE)

x <- rnorm(N)
y <- x+rnorm(N)
dfn <- data.frame(x=x,y=y)</pre>
GPCAn <- GPCA_iteration(dfn,extremes=TRUE)
```

inv\_GPCA

This function makes an inverse Gaussianization procedure besad on PCA iteration (see inv\_GPCA\_iteration

#### **Description**

This function makes an inverse Gaussianization procedure besad on PCA iteration (see inv\_GPCA\_iteration

#### Usage

```
inv_GPCA(x = NULL, GPCA_param, type = 3, extremes = TRUE)
```

#### **Arguments**

x gaussian random variable to transform

GPCA\_param GPCA-class S3 object returned by the function GPCA

#### Value

the non-Gaussian random variable

#### Note

This function re-iterates the inverse of equation (1) of "PCA Gaussianization for One-Class Remote Sensing Image" by V. Laparra et al., https://ieeexplore.ieee.org/document/5413808/

38 inv\_GPCA\_iteration

#### Author(s)

Emanuele Cordano

# See Also

```
GPCA,GPCA_iteration,inv_GPCA_iteration,inv_GPCA
```

# **Examples**

```
library(RMAWGEN)
set.seed(1222)
nIterations <- 30
N <- 20
x <- rexp(N)
y <- x+rnorm(N)
df <- data.frame(x=x,y=y)

GPCA <- GPCA(df,n=nIterations,extremes=TRUE)

x <- rnorm(N)
y <- x+rnorm(N)
dfn <- data.frame(x=x,y=y)

GPCAn <- GPCA(dfn,n=nIterations,extremes=TRUE)

df_out <- inv_GPCA(GPCA_param=GPCA,extremes=TRUE)

dfn_out <- inv_GPCA(GPCA_param=GPCAn,extremes=TRUE)</pre>
```

inv\_GPCA\_iteration

This function makes an inverse iteration of PCA-Gaussianization process

# **Description**

This function makes an inverse iteration of PCA-Gaussianization process

```
inv_GPCA_iteration(
   x = GPCA_iter_param$x_next,
   GPCA_iter_param,
   type = 3,
   extremes = TRUE
)
```

inv\_GPCA\_iteration 39

# **Arguments**

```
x matrix of gaussian random variale to transform

GPCA_iter_param

GPCAiteration S3 object returned by the function GPCA_iteration corresponding the related direct iteration

type see normalizeGaussian_severalstations

extremes see normalizeGaussian_severalstations
```

# Value

the non-Gaussian random variable

#### Note

This function is based on the inverse of the equation (1) of "PCA Gaussianization for One-Class Remote Sensing Image" by V. Laparra et al., https://ieeexplore.ieee.org/document/5413808/

#### See Also

```
GPCA,GPCA_iteration,inv_GPCA_iteration,inv_GPCA,GPCA-class for 'GPCA' S3 class
```

# **Examples**

```
library(RMAWGEN)
set.seed(1222)
N <- 20
x <- rexp(N)
y <- x+rnorm(N)
df <- data.frame(x=x,y=y)

GPCA <- GPCA_iteration(df,extremes=TRUE)

x <- rnorm(N)
y <- x+rnorm(N)
dfn <- data.frame(x=x,y=y)

GPCAn <- GPCA_iteration(dfn,extremes=TRUE)

df_out <- inv_GPCA_iteration(GPCA_iter_param=GPCA,extremes=TRUE)

dfn_out <- inv_GPCA_iteration(GPCA_iter_param=GPCAn,extremes=TRUE)</pre>
```

is.monthly.climate

is.monthly.climate	Verifies if 'climate' represents the monthly climatology in one year, i.e 'climate' is monthly.climate type matrix whose rows represent months and each column represents a station. It is also used in
	setComprehensiveTemperatureGeneratorParameters.

# Description

Verifies if 'climate' represents the monthly climatology in one year, i.e 'climate' is monthly.climate type matrix whose rows represent months and each column represents a station. It is also used in setComprehensiveTemperatureGeneratorParameters.

# Usage

```
is.monthly.climate(climate, nstation = 3, nmonth = 12, verbose = TRUE)
```

# Arguments

climate	matrix containing the 'monthly climatology' data
nstation	number of variable measurement stations (columns of the matrix 'climate')
nmonth	number of months in one year (it can be different if climate is represented by seasonal avarages or others), Default is 12 (recommended). (it can be different if climate is represented by seasonal averages, in this case 4)
verbose	Prints output and warining messagrs only if is TRUE.

# Value

A logical variable if the matrix 'climate' is monthly.climate type

# Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

 ${\tt setComprehensiveTemperatureGeneratorParameters}$ 

months\_f 41

months_f months REPLACEMAN	months_f	months REPLACEMAN
----------------------------	----------	-------------------

# Description

months REPLACEMANT

# Usage

```
months_f(x, ...)
```

# **Arguments**

x an object. See months
... arguments

# Description

Generates a new realization of a VAR model

# Usage

```
NewVAReventRealization(var, xprev, noise, exogen = NULL, B = NULL)
```

# Arguments

var	A VAR model represented by a varest object as returned by ${\tt getVAR model}$ or ${\tt VAR}$
xprev	previous status of the random variable
noise	<pre>uncorrelated or white noise (residual). Default is rnorm(length(xprev)) (or rnorm(ncol(B))</pre>
exogen	vector containing the values of the "exogen" variables (predictor) for the generation
В	matrix of coefficients for the vectorial white-noise component

#### Value

a vector of values

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

forecastEV.forecastResidual

newVARmultieventRealization

Generates several realizations of a VAR model

#### **Description**

Generates several realizations of a VAR model

# Usage

```
newVARmultieventRealization(
  var,
  xprev = rnorm(var@VAR$K * var@VAR$p),
  exogen = NULL,
  nrealization = 10,
  B = t(chol(cov(residuals(var)))),
  extremes = TRUE,
  type = 3,
  noise = NULL
)
```

#### **Arguments**

var A VAR model represented by a varest2 object as returned by getVARmodel

xprev previous status of the random variable

exogen matrix containing the values of the "exogen" variables (predictor) for the gener-

ation

nrealization number of realization (e.g. days to simulate). If exogen is not NULL and it is a

matrix, it must be lower or equal to the number of rows of exogen

B matrix of coefficients for the vector white-noise component

extremes, type see inv\_GPCA

noise stochastic noise to add for variabile generation. Default is NULL and it is au-

tomatically randomly genereted accordind to matrix B. If the VAR model (var argument) does not fit well the residuals (e.g. non-normality, non-serialty or heteroskesticity) and the white noise is manually inserted, in this case argument

B is not taken into account.

normality\_test 43

# Value

a matrix of values

# Author(s)

Emanuele Cordano, Emanuele Eccel

normality\_test

normality.test method for varest2 object

# Description

```
normality.test method for varest2 object
```

# Usage

```
normality_test(object, ...)
```

# Arguments

```
object a varest2 object ... passed arguments
```

#### See Also

```
normality.test
```

normalizeGaussian

Converts a random variable x extracted by a population represented by the sample data or sample to a normally-distributed variable with assigned mean and standard deviation or vice versa in case inverse is TRUE

# Description

Converts a random variable x extracted by a population represented by the sample data or sample to a normally-distributed variable with assigned mean and standard deviation or vice versa in case inverse is TRUE

44 normalizeGaussian

# Usage

```
normalizeGaussian(
  x = 0,
  data = x,
  cpf = NULL,
  mean = 0,
  sd = 1,
  inverse = FALSE,
  step = NULL,
  prec = 10^-4,
  type = 3,
  extremes = TRUE,
  sample = NULL
)
```

# Arguments

X	value or vector of values to be converted	
data	a sample of data on which a non-parametric probability distribution is estimated	
cpf	cumulative probability distribution. If NULL (default) is calculated as ecdf (data)	
mean	mean (expected value) of the normalized random variable. Default is 0.	
sd	standard deviation of the normalized random variable. Default is 1.	
inverse	logical value. If TRUE the function works inversely (the opposite way). Default is FALSE.	
step	vector of values in which step discontinuities of the cumulative probability function occur. Default is NULL	
prec	amplitude of the neighbourhood of the step discontinuities where cumulative probability function is treated as non-continuous.	
type	see quantile	
extremes	logical variable. If TRUE (default) the probability or frequency is multiplied by	
	$rac{N}{N+1}$	
	where $N$ is the length of data	
sample	a character string or NULL containing sample or probability distribution information. Default is NULL	

# Value

the normalized variable or its inverse

@note This function makes a Marginal Gaussianization. See the R code for further details

# Author(s)

Emanuele Cordano, Emanuele Eccel

normalizeGaussian\_prec

Converts precipitation values to "Gaussinized" normally-distributed values taking into account the probability of no precipitation occurrences, values or vice versa in case inverse is TRUE

# Description

Converts precipitation values to "Gaussinized" normally-distributed values taking into account the probability of no precipitation occurrences. values or vice versa in case inverse is TRUE

# Usage

```
normalizeGaussian_prec(
  x = 0,
  data = x,
  cpf = NULL,
  mean = 0,
  sd = 1,
  inverse = FALSE,
  type = 3,
  extremes = TRUE,
  sample = NULL,
  qnull = 0,
  valmin = 1
)
```

# **Arguments**

X	value or vector of values to be converted
data	a sample of data on which a non-parametric probability distribution is estimated
cpf	cumulative probability distribution. If NULL (default) is calculated as ecdf (data)
mean	mean (expected value) of the normalized random variable. Default is 0.
sd	standard deviation of the normalized random variable. Default is 1.
inverse	logical value. If TRUE the function works inversely (the opposite way). Default is FALSE.
type	see quantile
extremes	logical variable. If TRUE (default) the probability or frequency is multiplied by
	N
	$\overline{N+1}$
	where $N$ is the length of data
sample	a character string or NULL containing sample or probability distribution information. Default is NULL
qnull	probability of no precipitation occurrence
valmin	minimum value of precipitation to consider a wet day

#### Value

the normalized variable or its inverse

#### Note

In the version 1.2.5 of RMAWGEN This function is deprecated and not used.

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

normalizeGaussian

# **Examples**

```
library(RMAWGEN)
NDATA <- 1000
occurrence <- as.logical(runif(NDATA)>0.5)
prec <- rexp(NDATA,rate=1/3)
prec[!occurrence] <- 0
valmin <- 0.5 #0.01
x <- normalizeGaussian_prec(x=prec,valmin=valmin)
prec2 <- normalizeGaussian_prec(x=x,data=prec,valmin=valmin,inverse=TRUE)
qqplot(prec,prec2)
occurrence3 <- as.logical(runif(NDATA)>0.5)
prec3 <- rexp(NDATA,rate=1/3)
prec3[!occurrence3] <- 0
x3 <- normalizeGaussian_prec(x=prec3,valmin=valmin)
qqplot(x,x3)
abline(0,1)</pre>
```

normalizeGaussian\_severalstations

Converts several samples x random variable extracted by populations represented by the columns of data respectively or sample to a normally-distributed samples with assinged mean and standard deviation or vice versa in case inverse is TRUE

# Description

Converts several samples x random variable extracted by populations represented by the columns of data respectively or sample to a normally-distributed samples with assinged mean and standard deviation or vice versa in case inverse is TRUE

# Usage

```
normalizeGaussian_severalstations(
    x,
    data = x,
    cpf = NULL,
    mean = 0,
    sd = 1,
    inverse = FALSE,
    step = NULL,
    prec = 10^-4,
    type = 3,
    extremes = TRUE,
    sample = NULL,
    origin_x = NULL,
    origin_data = origin_x
)
```

value to be converted

# **Arguments** ×

* *	
data	a sample of data on which a non-parametric probability distribution is estimated
cpf	cumulative probability distribution. If NULL (default) is calculated as ecdf (data)
mean	mean (expected value) of the normalized random variable. Default is 0.
sd	standard deviation of the normalized random variable. Default is 1.
inverse	logical value. If TRUE the function works inversely (the opposite way). Default is FALSE.
step	vector of values in which step discontinuities of the cumulative probability function occur. Default is NULL
prec	amplitude of the neighbourhood of the step discontinuities where cumulative probability function is treated as non-continuous.
type	see quantile
extremes	logical variable. If TRUE (default) the probability or frequency is multiplied by
	N
	$\frac{N}{N+1}$
	where $N$ is the length of data
sample	information on how to sample $x$ and data. Default is NULL, this means that the values of each column of $x$ and data belong to the same sample. If $x$ and data are sampled for each month seperately, it is set to monthly.
origin_x	date corresponding to the first row of x
origin_data	date corresponding to the first row of data

# Value

a matrix with the normalized variable or its inverse

#### Note

It applies normalizeGaussian for each column of x and data. See the R code for further details

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

normalizeGaussian

# **Examples**

```
## Not run:
library(RMAWGEN)
set.seed(1234)
N <- 30
x \leftarrow rexp(N)
y <- x+rnorm(N)
df <- data.frame(x=x,y=y)</pre>
dfg <- normalizeGaussian_severalstations(df,data=df,extremes=TRUE,inverse=FALSE)</pre>
dfi <- normalizeGaussian_severalstations(dfg,data=df,extremes=TRUE,inverse=TRUE)</pre>
N <- 365*2
origin <- "1981-01-01"
x \leftarrow rexp(N)
y <- x+rnorm(N)</pre>
df <- data.frame(x=x,y=y)</pre>
dfgm <- normalizeGaussian_severalstations(df,data=df,extremes=TRUE,</pre>
inverse=FALSE,origin_x=origin,origin_data=origin,sample="monthly")
dfim <- normalizeGaussian_severalstations(dfg,data=df,extremes=TRUE,</pre>
inverse=TRUE,origin_x=origin,origin_data=origin,sample="monthly")
## Compatibility with 'lubridate' package
library(lubridate)
N <- 30
x < - rexp(N)
y <- x+rnorm(N)
df <- data.frame(x=x,y=y)</pre>
dfg <- normalizeGaussian_severalstations(df,data=df,extremes=TRUE,inverse=FALSE)</pre>
```

```
dfi <- normalizeGaussian_severalstations(dfg,data=df,extremes=TRUE,inverse=TRUE)
N <- 365*2
origin <- "1981-01-01"
x <- rexp(N)
y <- x+rnorm(N)
df <- data.frame(x=x,y=y)

dfgm <- normalizeGaussian_severalstations(df,data=df,extremes=TRUE,inverse=FALSE,origin_x=origin,origin_data=origin,sample="monthly")

dfim <- normalizeGaussian_severalstations(dfg,data=df,extremes=TRUE,inverse=TRUE,origin_x=origin,origin_data=origin,sample="monthly")
## End(Not run)</pre>
```

normalizeGaussian\_severalstations\_prec

DEPRECATED Converts several samples x random variable (daily precipitation values) extracted by populations represented by the columns of data respectively or sample to a normally-distributed samples with assinged mean and standard deviation or vice versa in case inverse is TRUE using the function normalizeGaussian\_prec

# Description

DEPRECATED Converts several samples x random variable (daily precipitation values) extracted by populations represented by the columns of data respectively or sample to a normally-distributed samples with assinged mean and standard deviation or vice versa in case inverse is TRUE using the function normalizeGaussian\_prec

```
normalizeGaussian_severalstations_prec(
    x,
    data = x,
    cpf = NULL,
    mean = 0,
    sd = 1,
    inverse = FALSE,
    qnull = NULL,
    valmin = 0.5,
    type = 3,
    extremes = TRUE,
```

```
sample = NULL,
origin_x = NULL,
origin_data = NULL)
```

# Arguments

X	value to be converted
data	a sample of data on which a non-parametric probability distribution is estimated
cpf	cumulative probability distribution. If NULL (default) is calculated as ecdf(data)
mean	mean (expected value) of the normalized random variable. Default is 0.
sd	standard deviation of the normalized random variable. Default is 1.
inverse	logical value. If TRUE the function works inversely (the opposite way). Default is FALSE.
qnull	probability of no precipitation occurrence. (It can be a matrix in case sample="monthly"
valmin	minimum value of precipitation to consider a wet day
type	see quantile
extremes	logical variable. If TRUE (default) the probability or frequency is multiplied by
	$\frac{N}{N+1}$
	where $N$ is the length of data
sample	information about sample or probability distribution. Default is NULL
origin_x	date corresponding to the first row of x
origin_data	date corresponding to the first row of data

# Value

a matrix or a data.frame with the normalized variable or its inverse

# Note

In the version 1.2.5 of **RMAWGEN** This function is deprecated and not used.

# Author(s)

Emanuele Cordano, Emanuele Eccel

# See Also

normalizeGaussian\_prec

plotDailyClimate 51

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Plots daily climatology through one year

# Description

Plots daily climatology through one year

#### Usage

```
plotDailyClimate(
   data,
   title = "Daily_Avereged_Temperture_in_one_year",
   origin = "1961-1-1",
   when = "1979-1-1",
   ylab = "Temperature [degC]",
   xlab = "Time [days]",
   nday = 365,
   bicolor = FALSE,
   col = "black",
   lwd = 1
)
```

# **Arguments**

data

matrix whose columns contain daily-averaged climatic series of variables (e.g. maximum or minum daily averaged temperature obtained by spline interpolation

of monthly climatology)

title, xlab, ylab, col, lwd

see plot.default

origin origin date corresponding to the first row of data

when start day for daily climatology plot

nday number of days in one year. Default is 365.

bicolor logical variable. If TRUE and data represents climatologies of minimun and

maximum daily temperature, the lines are plotted with blue and red colors re-

spectively.

## Value

a matrix containing the plotted variables

# Author(s)

Emanuele Cordano, Emanuele Eccel

52 plot\_sample

plot\_sample

It makes a plot by sampling (e.g. monthly) the variables x and y

# **Description**

It makes a plot by sampling (e.g. monthly) the variables x and y

#### Usage

```
plot_sample(
 Х,
 y = normalizeGaussian_severalstations(x = as.data.frame(x), data = as.data.frame(data),
  origin_x = origin_x, origin_data = origin_data, sample = sample, step = step, prec =
   prec)[, 1],
 xlim = range(x, na.rm = TRUE),
  legend_position = "topleft",
 ylim = range(y, na.rm = TRUE),
 pch = 1,
  col = 1,
  col_max = 0.9,
  col_min = 0.1,
 origin,
  sample = NULL,
  xhist = hist(x, breaks = breaks, plot = FALSE),
 yhist = hist(y, breaks = breaks, plot = FALSE),
  axes = FALSE,
  step = NULL,
 prec = 1e-04,
 breaks = 50,
 origin_x = origin,
 origin_data = origin,
  data = x,
  xlab = ""
 ylab = "",
  color = FALSE,
  gray = TRUE,
  sort = FALSE,
  valmin_x = valmin,
  valmin_y = valmin,
  valmin = -9999,
  abline = c(0, 1),
)
```

#### **Arguments**

x vector of input data

plot\_sample 53

vector of second input data. Default is normalizeGaussian\_severalstations(x=as.data.frame(x),

xlim, ylim, xlab, ylab see plot.default (Graphic) legend\_position legend position. Default is "topleft". See legend. pch integer single or multi values for pch (see plot.default). Default is 1. col integer single or multi values for col (see plot.default). Default is 1. maximum value for color scale to apply to rainbow or rainbow. Utilized if col col\_max is not a vector and both gray or color are TRUE. Default is 0.9. col\_min minimum value for color scale to apply to rainbow or rainbow. Utilized if col is not a vector and both gray or color are TRUE. Default is 0.1. origin date of the first row of x. See normalizeGaussian\_severalstations. string character containg informatio how to sample x and y. Default is NULL. If sample NULL no sampling is done.see normalizeGaussian\_severalstations. Only NULL or "monthly" options are implemented. frequency histogram for x. Default is hist(x,breaks=breaks,plot=FALSE). xhist If it is NULL, no marginal histograms appear. yhist frequency histogram for y. Default is hist(y,breaks=breaks,plot=FALSE). If it is NULL, no marginal histograms appear. =hist(y,breaks=breaks,plot=FALSE), see barplot axes see normalizeGaussian\_severalstations step, prec breaks origin\_x see normalizeGaussian\_severalstations. Default value is set equal to origin. origin\_data normalizeGaussian\_severalstations. Default value is set equal to origin. data normalizeGaussian\_severalstations. Default value is set equal to x. color logical value. If TRUE and if col is unspecified, a color scale is applied according to col\_min and col\_max (see rainbow). Default is FALSE. logical value. If TRUE and if col is unspecified, a color scale is applied according gray to col\_min and col\_max (see gray). Default is TRUE. logical value. If TRUE, x and y are sorted and a Q-Q plot is presented. Deafault sort is FALSE. numerical threshold value over which the variable x is plotted. It is enabled only valmin\_x if sort is set TRUE. numerical threshold value over which the variable y is plotted. It is enabled only valmin\_y if sort is set TRUE. valmin numerical threshold value for valmin\_y and valmin\_x if there are not specified. abline arguments for abline function. Default is c(0,1). If it is NULL, abline is disabled and not called. see graphical parametes on plot.default @usage plot\_sample(x, y = normalizeGaussian\_severalstations(x = as.data.frame(x), data = as.data.frame(data), origin\_x = origin\_x, origin\_data = origin\_data, sample = sample, step = step, prec = prec)[, 1], xlim = range(x, na.rm = TRUE),

54 PrecipitationEndDay

legend\_position = "topleft", ylim = range(y, na.rm = TRUE), pch = 1, col = 1, col\_max = 0.9, col\_min = 0.1, origin, sample = NULL, xhist = hist(x, breaks = breaks, plot = FALSE), yhist = hist(y, breaks = breaks, plot = FALSE), axes = FALSE, step = NULL, prec = 1e-04, breaks = 50, origin\_x = origin, origin\_data = origin, data = x, xlab = "", ylab = "", color = FALSE, gray = TRUE, sort = FALSE, valmin\_x = valmin, valmin\_y = valmin, valmin = -9999, abline = c(0, 1), ...)

#### Value

0 in case of success

#### Note

It makes a plot betwee x and y and shows thair respective probibilty histograms. If y is missing, it is automatically calculated as one-dimensional Gaussianization of x through the function normalizeGaussian\_severalstations.

#### See Also

plot.default,extractmonths, see normalizeGaussian\_severalstations

#### **Examples**

```
## Not run:
library(lubridate)
data(trentino)
plot_sample(x=TEMPERATURE_MIN$T0090,sample="monthly",
    origin="1958-1-1",axes=FALSE,xlab="Tn [ degC]",
    ylab="x")

set.seed(123456)
z <- rexp(10000,rate=0.5)
x <- normalizeGaussian(x=z,data=z)
plot_sample(x=z,xlab="z",ylab="x")

## End(Not run)</pre>
```

PrecipitationEndDay

Gets the last day in a precipitation time series, expressed in decimal julian days since 1970-1-1 00:00 UTC

# **Description**

@author Emanuele Cordano, Emanuele Eccel

55 PrecipitationStartDay

#### Usage

PrecipitationEndDay(name, station\_names, end\_day)

#### **Arguments**

charcacter ID of the station name

station\_names vector containing the IDs (characters) of the considered meteorological stations.

An example is STATION\_NAMES defined in trentino.

vector containing the measurement end day. An example is TEMPERATURE\_MEASUREMENT\_END\_DAY end\_day

defined in trentino.

#### Value

the precipitation measurement end day given the vectors of station IDs and the precipitation measurement end days

#### **Examples**

data(trentino)

PrecipitationEndDay("T0099", station\_names=STATION\_NAMES, end\_day=PRECIPITATION\_MEASUREMENT\_END\_DAY)

PrecipitationStartDay Gets the first day in a precipitation time series, expressed in decimal

julian days since 1970-1-1 00:00 UTC

#### **Description**

@author Emanuele Cordano

#### Usage

PrecipitationStartDay(name, station\_names, start\_day)

#### **Arguments**

character ID of the station name

vector containing the IDs (characters) of the considered meteorological stations. station\_names

An example is STATION\_NAMES defined in the trentino dataset.

start\_day vector containing the precipitation measurement start day. An example is TEMPERATURE\_MEASUREMENT\_ST

defined in the trentino dataset.

#### Value

the precipitation measurement start day given the vectors of station IDs and the respective precipitation measurement start days

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#### **Examples**

```
data(trentino)
PrecipitationStartDay("T0099",
    station_names=STATION_NAMES,
    start_day=PRECIPITATION_MEASUREMENT_START_DAY)
```

print.GPCA

print S3 method for GPCA or GPCA\_iteration object

#### **Description**

```
print S3 method for GPCA or GPCA_iteration object
```

# Usage

```
## S3 method for class 'GPCA'
print(x, rmin = 1, rmax = 4, cmin = rmin, cmax = rmax, ...)
## S3 method for class 'GPCAiteration'
print(x, rmin = 1, rmax = 4, cmin = rmin, cmax = rmax, ...)
```

#### **Arguments**

```
x a GPCA or GPCAiteration object
rmin, rmax, cmin, cmax
maximum and minimum rows and columns to be printed
... passed arguments
```

#### See Also

```
GPCA,GPCA_iteration
GPCA_iteration
```

qqplot.lagged

This function creates a Q-Q plot of the lag-lag moving cumulative addition of the values in the samples x,y,z

# Description

This function creates a Q-Q plot of the lag-lag moving cumulative addition of the values in the samples x,y,z

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# Usage

```
qqplot.lagged(
    x = rnorm(1000),
    y = rnorm(1000),
    z = NULL,
    when = 1:length(x),
    lag = 1,
    pch = 1,
    ...
)
```

# **Arguments**

```
x, y samples. If x is a data frame, y and z can be omitted.
z further samples organized as a list
when (integer) inidices of x and y on which the Q-Q plot is made.
lag (current index included) on whose value the addition is made.
pch a vector of plotting characters or symbols: see points
... further arguments for qqplot
```

#### Value

the Q-Q plot

#### See Also

qqplot

qqplotprecWGEN

Makes a applot of measured and simulated data for several stations.

# Description

Makes a qqplot of measured and simulated data for several stations.

```
qqplotprecWGEN(
  measured,
  simulated,
  xlab = "simulated[mm]",
  ylab = "measured[mm]",
  title = "daily precipitation",
  station = NULL,
  diff = FALSE,
  quantile = 0
)
```

#### Arguments

measured matrix containing measured data (each station corresponds to a column)

matrix containing respective generated data (each station corresponds to a column)

xlab, ylab see plot.default,qqplotWGEN

title title

station character vector containing IDs of analyzed stations. If NULL (default) all stations (columns of simulated and measured) are considered

diff, quantile see qqplotWGEN

#### Value

0 in case of success

#### Note

It uses qqplotWGEN and makes a figure for each pair of columns from measured and simulated. See the R code for further details.

#### Author(s)

Emanuele Cordano, Emanuele Eccel

```
qqplotprecWGEN_seasonal
```

Makes four seasonal qqplots (winter, spring, summer and autumn) of measured and simulated data for several stations.

# Description

Makes four seasonal qqplots (winter, spring, summer and autumn) of measured and simulated data for several stations.

```
qqplotprecWGEN_seasonal(
  measured,
  simulated,
  origin = "1961-1-1",
  xlab = "simulated[mm]",
  ylab = "measured[mm]",
  title = "daily_precipitation",
  directorypdf,
  station = names(simulated)
)
```

qqplotTnTxWGEN 59

# **Arguments**

measured matrix containing measured data (each station corresponds to a column) simulated matrix containing respective generated data (each station corresponds to a column) origin first day of data, see extractmenths for format and other information xlab, ylab see plot.default,qqplotWGEN title title directorypdf name of the directory (path included) where to seva the outputs station character vector containing IDs of analyzed stations. If NULL (default) all stations (columns of simulated and measured) are considered

#### Value

0 in case of success

#### Note

Uses qqplotprecWGEN for each season of collected data and saves the output on pdf files. See the R code for further details.

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

qqplotprecWGEN,extractmonths

qqplotTnTxWGEN

Makes a applot of measured and simulated data for several stations.

# Description

Makes a qqplot of measured and simulated data for several stations.

```
qqplotTnTxWGEN(
  measured,
  simulated,
  xlab = "simulated[degC]",
  ylab = "measured[degC]",
  titles = c("Q-Qplot_An._Tx", "Q-Qplot_An._Tn"),
  station = NULL,
  diff = FALSE,
  quantile = 0
)
```

# **Arguments**

measured	matrix containing measured data (each station corresponds to a column)
simulated	matrix containing respective generated data (each station corresponds to a column)
xlab, ylab	see plot.default,qqplotWGEN
titles	titles that will be added to main argument of plot.default
station	character vector containing IDs of analyzed station. If NULL (default) all station (columns of simulated and measured) are considered
diff, quantile	see qqplotWGEN

# Value

0 in case of success

#### Note

It uses qqplotWGEN and makes a figure for each pair of columns from measured and simulated. See the R code for further details.

#### Author(s)

Emanuele Cordano, Emanuele Eccel

```
qqplotTnTxWGEN_seasonal
```

Makes four seasonal qqplots (winter, spring, summer and autumn) of measured and simulated data for several stations.

# Description

Makes four seasonal qqplots (winter, spring, summer and autumn) of measured and simulated data for several stations.

```
qqplotTnTxWGEN_seasonal(
  measured,
  simulated,
  origin = "1961-1-1",
  xlab = "simulated[degC]",
  ylab = "measured[degC]",
  titles = c("Q-Qplot_An._Tx", "Q-Qplot_An._Tn"),
  directorypdf,
  station = NULL
)
```

qqplotWGEN 61

# **Arguments**

measured	matrix containing measured data (each station corresponds to a column)
simulated	matrix containing respective generated data (each station corresponds to a column)
origin	first day of data, see extractmonths for format and other information
xlab, ylab	see plot.default,qqplotWGEN
titles	titles that will be added
directorypdf	name of the directory (path included) where to seva the outputs
station	character vector containing IDs of analyzed station. If NULL (default) all station (columns of simulated and measured) are considered

#### Value

0 in case of success

#### Note

Uses qqplotTnTxWGEN for each seasons of collected data and saves the output on pdf files. See the R code for further details.

# Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

qqplotTnTxWGEN, extractmonths

 ${\tt qqplotWGEN}$ 

Makes a qqplot and Wilcoxon test between the two columns of val

# Description

Makes a qqplot and Wilcoxon test between the two columns of val

```
qqplotWGEN(
  val,
  xlab = "simulated",
  ylab = "measured",
  main = "title",
  ylim = c(min(val), max(val)),
  xlim = c(min(val), max(val)),
  diff = FALSE,
  quantile = 0
)
```

#### **Arguments**

```
a matrix with two columns containing the two samples to be compared xlab, ylab, main see plot.default xlim, ylim see plot.default logical variable, if TRUE the function is applied to diff(val) instead of val. See diff quantile quantile value on which data samples in val are considered. Default is 0.
```

#### Value

Wilcoxon test between the two columns of 'val'

#### Author(s)

Emanuele Cordano, Emanuele Eccel

qqplot\_RMAWGEN\_Tx

It makes the Q-Q plots observed vs generated time series of daily maximum, minimum temperature and daily thermal range for a list of collected stochastic generations

#### **Description**

It makes the Q-Q plots observed vs generated time series of daily maximum, minimum temperature and daily thermal range for a list of collected stochastic generations

```
qqplot_RMAWGEN_Tx(
  Tx_mes,
 Tx_gen,
 Tn_gen,
 Tn_mes,
 Tx_spline = NULL,
  Tn_spline = NULL,
  xlab = "observed",
 ylab = "simulated",
 when = 1:nrow(Tx_mes),
 main = names(Tx\_gen),
  station,
  pdf = NULL,
 xlim = range(Tx_mes),
 ylim = xlim,
  cex = 0.4,
  cex.main = 1,
  cex.lab = 1,
```

```
cex.axis = 1
qqplot_RMAWGEN_Tn(
  Tx_mes,
  Tx_gen,
  Tn_gen,
  Tn_mes,
  Tx_spline = NULL,
  Tn_spline = NULL,
 xlab = "observed",
 ylab = "simulated",
 when = 1:nrow(Tn_mes),
 main = names(Tn_gen),
  station,
  pdf = NULL,
  xlim = range(Tn_mes),
 ylim = xlim,
 cex = 0.4,
  cex.main = 1,
  cex.lab = 1,
  cex.axis = 1
)
qqplot_RMAWGEN_deltaT(
 Tx_mes,
 Tx_gen,
 Tn_gen,
  Tn_mes,
  xlab = "observed",
  ylab = "simulated",
 when = 1:nrow(Tx_mes),
 main = names(Tx_gen),
  station,
  pdf = NULL,
  xlim = range(Tx_mes - Tn_mes),
 ylim = xlim,
  cex = 0.4,
  cex.main = 1,
  cex.lab = 1,
  cex.axis = 1
)
qqplot_RMAWGEN_prec(
 prec_mes,
  prec_gen,
  xlab = "observed",
 ylab = "simulated",
```

```
when = 1:nrow(prec_mes),
main = names(prec_gen),
station,
pdf = NULL,
xlim = range(prec_mes),
ylim = xlim,
cex = 0.4,
cex.main = 1,
cex.lab = 1,
cex.axis = 1,
lag = 1
)
```

# Arguments

Tx_mes	data frame containing measured daily maximum temperature	
Tx_gen	data frame containing generated daily maximum temperature	
Tn_gen	data frame containing generated daily minimum temperature	
Tn_mes	data frame containing measured daily minimum temperature	
Tx_spline	data frame containing spline-interpolated daily maximum temperature. Default is NULL and not considered for Q-Q plot.	
Tn_spline	data frame containing spline-interpolated daily minimum temperature Default is NULL and not considered for Q-Q plot.	
xlab, ylab	lables of x and y axes. See qqplot.	
when	day indices on which the data frame are extracted for Q-Q plot. Default is 1:nrow(Tn_mes) (in qqplot_RMAWGEN_Tn) or 1:nrow(Tx_mes) (otherwise)	
main	main titles for each plot. Default is names(Tn_gen) (in qqplot_RMAWGEN_Tn) or names(Tx_gen) (otherwise)	
station	identification name (ID) of the station used for the Q-Q plot	
pdf	name of pdf file if output is written in a pdf file	
xlim	<pre>see qqplot. Default is range(Tn_mes) (in qqplot_RMAWGEN_Tn) or range(Tx_mes) (in qqplot_RMAWGEN_Tx).or range(Tx_mes-Tn_mes) (in qqplot_RMAWGEN_deltaT)</pre>	
ylim, cex, cex.main, cex.lab, cex.axis		
	see qqplot and plot	
prec_mes	data frame containing measured daily precipitation (in millimeters)	
prec_gen	data frame containing generated daily precipitation (in millimeters)	
lag	lag (current index included) on whose value the precipitation addition is made. See qqplot.lagged.	

#### Note

 $Tx\_gen, Tn\_gen$  and main must have an even number of elements.

# Author(s)

Emanuele Cordano

removeNAs 65

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Replaces each entry of the rows containing NA values with NA

# Description

Replaces each entry of the rows containing NA values with NA

#### **Usage**

```
removeNAs(data)
```

#### **Arguments**

data a matrix

@author Emanuele Cordano, Emanuele Eccel

#### Value

the matrix data with the modified rows of NA values

#### Note

In getVARmodel, when using VAR or VARselect, all NAs will be removed

#### See Also

getVARmodel

rescaling_n	nonthly
-------------	---------

This function adjusts the monthly mean to a daily weather dataset (e. g. spline-interpolated temperature)

# **Description**

This function adjusts the monthly mean to a daily weather dataset (e. g. spline-interpolated temperature)

# Usage

```
rescaling_monthly(data, val, origin = "1961-1-1")
```

# Arguments

data frame of wheather variables)

val monthly means returned by getMonthlyMean

origin character string containing the gregorian date of the first day of data

66 residuals.varest2

# Value

A data frame with data of data rescaled with val for each month

#### Note

```
It uses months and julian
```

#### Author(s)

Emanuele Cordano

@export

#### See Also

```
extractdays
```

residuals.varest2

residuals S3 method for varest2 object

# **Description**

residuals S3 method for varest2 object

# Usage

```
## S3 method for class 'varest2'
residuals(object, squared = FALSE, ...)
```

# **Arguments**

object a blockmatrix object

squared logical value. Default is FALSE. If TRUE the method returns the squared residuals.

... passed arguments

#### Value

residuals of object as a data frame. In case squared=TRUE, the squared residuals are returned, otherwise simple residuals are returned. The squared residuals can be useful in case of ARCH analysis.

# Author(s)

Emanuele Cordano

serial\_test 67

serial\_test

serial.test function for varest2 object

# **Description**

```
serial.test function for varest2 object
```

# Usage

```
serial_test(object, ...)
```

# **Arguments**

```
object a varest2 object ... passed arguments
```

#### See Also

```
serial.test
```

 $\verb|setComprehensiveTemperatureGeneratorParameters|\\$ 

Computes climatic and correlation information useful for creating an auto-regeressive random generation of maximum and minimun daily temparature. This function is called by ComprehensiveTemperatureGenerator.

# Description

Computes climatic and correlation information useful for creating an auto-regeressive random generation of maximum and minimum daily temparature. This function is called by ComprehensiveTemperatureGenerator.

```
setComprehensiveTemperatureGeneratorParameters(
   station,
   Tx_all,
   Tn_all,
   mean_climate_Tn = NULL,
   mean_climate_Tx = NULL,
   Tx_spline = NULL,
   Tn_spline = NULL,
   year_max = 1990,
   year_min = 1961,
   leap = TRUE,
```

```
nmonth = 12,
verbose = FALSE,
cpf = NULL,
normalize = TRUE,
sample = NULL,
option = 2,
yearly = FALSE
```

## **Arguments**

station character vector of the IDs of the considered meteorological stations

Tx\_all data frame containing daily maximum temperature of all meteorological station.

See TEMPERATURE\_MAX for formatting.

Tn\_all data frame containing daily minimum temperature of all meteorological station.

See TEMPERATURE\_MIN for formatting.

mean\_climate\_Tn

a matrix containing monthly mean minimum daily temperature for the considered station or an object as returned by getMonthlyMean. If NULL, it is calcu-

lated. See input of is.monthly.climate

mean\_climate\_Tx

a matrix containing monthly mean maximum daily temperature for the considered station or an object as returned by getMonthlyMean. If NULL, it is calcu-

lated. See input of is.monthly.climate

Tx\_spline daily timeseries (from the first day of year\_min to the last day of year\_max) of

averaged maximum temperature which can be obtained by a spline interpolation of monthly mean values. Default is NULL and returned as output. See for spline interpolation utilized: splineInterpolateMonthlytoDailyforSeveralYears.

Tn\_spline daily timeseries (from the first day of year\_min to the last day of year\_max) of

averaged minimum temperature which can be obtained by a spline interpolation of monthly mean values. Default is NULL and returned as output. See for spline interpolation utilized: splineInterpolateMonthlytoDailyforSeveralYears.

year\_max start year of the recorded (calibration) period year\_min end year of the recorded (calibration) period

leap logical variables. It is TRUE (Default) if leap years are considered

nmonth number of months in one year. Default is 12.

verbose logical variable

cpf see normalizeGaussian\_severalstations

normalize logical variable If TRUE normalizeGaussian\_severalstations is used, other-

wise it is not. If option is 2, it is always TRUE.

sample see normalizeGaussian\_severalstations

option integer value. If 1, the generator works with minimum and maximum tem-

perature, if 2 (default) it works with the average value between maximum and

minimum temperature and the respective daily thermal range.

yearly logical value. If TRUE the monthly mean values are calculated for each year from

year\_min to year\_max separately. Default is FALSE.

#### Value

This function creates and returns the following gloabal variables:

data\_original matrix containing normalized and standardized data (i.e. data\_original)

data\_for\_var matrix returned from normalizeGaussian\_severalstations by processing data\_original if normalize is TRUE), otherwise it is equal to data\_original.

Tn\_mes matrix containing measured minimum daily temperature in the analyzed time period ( $Tn_{mes}$ )

Tx\_mes matrix containing measured maximum daily temperature in the analyzed time period (  $Tx_{mes}$ )

Tm\_mes matrix calculated as to

$$\frac{Tx_{mes} + Tn_{mes}}{2}$$

DeltaT\_mes matrix corresponding to  $Tx_{mes} - Tn_{mes}$ 

monthly\_mean\_Tn matrix containing monthly means of minimum daily temperature for the considered station. It is calculated according to the input format is.monthly.climate if saveMonthlyClimate is TRUE.

monthly\_mean\_Tx matrix containing monthly means of maximum daily temperature for the considered station. It is calculated according to the input format is.monthly.climate if saveMonthlyClimate is TRUE.

Tx\_spline matrix containing the averaged daily values of maximimum temperature obtained by a spline interpolation of the monthly climate monthly\_mean\_Tx or mean\_climate\_Tx using splineInterpolateMonthlytoDa ( $Tx_s$ )

Tn\_spline matrix containing the averaged daily values of minimum temperature obtained by a spline interpolation of the monthly climate monthly\_mean\_Tn or mean\_climate\_Tn using splineInterpolateMonthlytoDa( $Tn_s$ )

SplineAdvTm matrix calculated as  $\frac{Tx_s+Tn_s}{2}$ 

SplineAdvDeltaT, matrix corresponding to  $Tx_s - Tn_s$ 

stdTn vector containing the standard deviation of minimum temperature anomalies  $Tn_{mes} - Tn_s$   $(\sigma_{Tn})$ 

stdTx vector containing the standard deviation of maximum temperature anomalies  $Tx_{mes} - Tx_s$   $(\sigma_{Tx})$ 

stdTm vector containing the standard deviation of "mean" temperature anomalies  $Tm_{mes} - Tm_s$   $(\sigma_{Tm})$ 

Tn\_mes\_res standard core (standardization) of  $Tn_mes$  obtained by solving column by column the expression

$$\frac{Tn_{mes} - Tn_s}{\sigma_{Tn}}$$

Tx\_mes\_res standard core (standardization) of  $Tx_mes$  obtained by solving column-by-column the expression

$$\frac{Tx_{mes} - Tn_s}{sd_{Tm}}$$

Tm\_mes\_res standard core (standardization) of  $Tm_mes$  obtained by solving column-by-column the expression

 $\frac{Tm_{mes} - Tn_s}{sd_{Tm}}$ 

DeltaT\_mes\_res equal to DeltaT\_mes

data\_original matrix obtained as cbind( $Tx_mes_res$ ,  $Tn_mes_res$ ) if option==1, or cbind( $Tm_mes_res$ ,  $DeltaT_mes_res$  if option==2

See the R code for further details.

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

spline Interpolate Monthly to Daily for Several Years, Comprehensive Temperature Generator Generator Temperature Generator Temperature Generator G

```
splineInterpolateMonthlytoDaily
```

Interpolates monthly data to daily data using spline and preserving monthly mean values

# Description

Interpolates monthly data to daily data using spline and preserving monthly mean values

# Usage

```
splineInterpolateMonthlytoDaily(
  nday = 365,
  val = as.matrix(cbind(1 * (0.5:11.5) * nday/12, 2 * (0.5:11.5) * nday/12)),
  origin = "1961-1-1",
  first_row = 1,
  last_row = nday,
  no_spline = FALSE,
  no_mean = FALSE
)
```

#### **Arguments**

nday number of days on which the daily data is requested, e.g. number of days in one

year

val matrix containing monthly mean data

origin date corresponding to the first row of the returned matrix

first\_row row corresponding the first day of time interval where montlhy mean conserva-

tion is applied

- ·	1. 4 1 1 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
last_row	corresponding the last day of time interval where monthly mean conservation is
	torresponding the rast day of time miter and where monthly mean conservation is

applied

no\_spline logical value. If TRUE no spline interpolation is calculated and the daily value

corresponds to the monthly average value. Default is FALSE.

no\_mean logical value. Default is FALSE. If TRUE the function output is not rescaled in

order to maintain observed mean monthly values.

#### Value

a matrix or data frame with interpolated daily data

#### Author(s)

Emanuele Cordano, Emanuele Eccel

# See Also

spline,splineInterpolateMonthlytoDailyforSeveralYears

```
splineInterpolateMonthlytoDailyforSeveralYears

**Interpolates monthly data to daily data using splineInterpolateMonthlytoDaily for several years**
```

#### **Description**

Interpolates monthly data to daily data using splineInterpolateMonthlytoDaily for several years

# Usage

```
splineInterpolateMonthlytoDailyforSeveralYears(
  val,
  start_year = 2010,
  nyear = 1,
  leap = TRUE,
  offset = 2,
  no_spline = FALSE,
  yearly = FALSE
)
```

# Arguments

```
val matrix containing monthly mean data for one year
```

start\_year first year

nyear number of years since start\_year

72 TemperatureEndDay

leap logical variable If TRUE (default) leap years are considered, otherwise they are

not

offset integer values. Default is 2. Number of years considered beyond the extremes

in order to avoid edge errors

no\_spline logical value. If TRUE no spline interpolation is calculated and the daily value

corresponds to the monthly average value. Default is FALSE.

yearly logical value. If TRUE the result with men value per each month per each year.

Default is FALSE.

@return a matrix or data frame with interpolated daily data

#### Author(s)

Emanuele Cordano, Emanuele Eccel

#### See Also

spline, splineInterpolateMonthlytoDaily

TemperatureEndDay Gets the last day in a temperature time series, expressed as decimal

julian days since 1970-1-1 00:00 UTC

#### **Description**

Gets the last day in a temperature time series, expressed as decimal julian days since 1970-1-1 00:00 UTC

# Usage

TemperatureEndDay(name, station\_names, end\_day)

#### **Arguments**

name character ID of the station

station\_names vector containing the IDs (characters) of the considered meteorological stations.

An example is STATION\_NAMES defined in the trentino dataset.

end\_day vector containing the measurement end day. An example is TEMPERATURE\_MEASUREMENT\_END\_DAY

defined in the trentino dataset.

#### Value

the temperature measurement end day given the vectors of station IDs and the temperature measurement end days

# Author(s)

Emanuele Cordano, Emanuele Eccel

TemperatureStartDay 73

#### **Examples**

data(trentino)

TemperatureEndDay("T0099", station\_names=STATION\_NAMES, end\_day=TEMPERATURE\_MEASUREMENT\_END\_DAY)

TemperatureStartDay

Gets the first day in a temperature time series, expressed as decimal

julian days since 1970-1-1 00:00 UTC

# **Description**

@author Emanuele Cordano, Emanuele Eccel

# Usage

TemperatureStartDay(name, station\_names, start\_day)

#### **Arguments**

name character ID of the station

station\_names vector containing the IDs (characters) of the considered meteorological stations.

An example is STATION\_NAMES defined in the trentino dataset.

start\_day vector containing the temperature measurement start day. Default is TEMPERATURE\_MEASUREMENT\_START\_

@examples data(trentino) TemperatureStartDay("T0099", station\_names=STATION\_NAMES, start\_day=TEMPERATURE\_

defined in the trentino dataset.

@export

#### Value

the temperature measurement start day given the vectors of station IDs and the respective temperature measurement start days

trentino Trentino Dataset

# Description

It contains the following variables:

TEMPERATURE\_MIN Data frame containing year, month, day and daily minimum temperature in 59 stations in Trentino region

TEMPERATURE\_MAX Data frame containing year, month, day and daily maximum temperature in 59 stations in Trentino region

74 trentino

PRECIPITATION Data frame containing year, month, day and daily precipitation in 59 stations in Trentino region

STATION\_NAMES Vector containing the names of the meteorological stations

ELEVATION Vector containing the elevations of the meteorological stations respectively

STATION\_LATLON Matrix containing the latitude and longitude coordinates, respectively, of the meteorological stations

LOCATION Vector containing the names of the location of each meteorological station

TEMPERATURE\_MEASUREMENT\_START\_DAY Vector containing the first days referred to midday (expressed as decimal julian day since 1970-1-1 00:00 UTC) of temperature measurement of each meteorological station

TEMPERATURE\_MEASUREMENT\_END\_DAY Vector containing the last days referred to midday (expressed as decimal julian day since 1-1-1970 00:00 UTC) of temperature measurement of each meteorological station

PRECIPITATION\_MEASUREMENT\_START\_DAY Vector containing the first days referred to midday (expressed as decimal julian day since 1-1-1970 00:00 UTC) of precipitation measurement of each meteorological station

PRECIPITATION\_MEASUREMENT\_END\_DAY Vector containing the last days referred to midday (expressed as decimal julian day since 1-1-1970) of precipitation measurement of each meteorological station

#### Usage

data(trentino)

#### Format

Data frames and vectors

#### **Details**

This dataset stores all information about meteorological stations and instrumental timeseries. The user can easily use the package with his/her own data after replacing the values of such variables.

#### Source

Original data are provided by Provincia Autonoma di Trento (https://www.meteotrentino.it/), Fondazione Edmund Mach (https://www.fmach.it), Provincia Autonama di Bolzano/Autome Provinz Bozen, ARPA Lombardia, ARPA Veneto (Italy).

This dataset is intended for research purposes only, being distributed in the hope that it will be useful, but WITHOUT ANY WARRANTY.

varest-class 75

varest-class

varest-class

# Description

varest S3 class (formal definition) see VAR

# **Details**

The details of the class are reported on VAR documentation in "vars" package

#### Note

Formal definition with setOldClass for the S3 class varest

# Author(s)

Bernhard Pfaff

# **Examples**

showClass("varest")

varest2-class

varest2-class

# Description

This class derives from a varest S3 class which is a list of objects describing a Vectorial AutoRegressive Model (see VAR)

# **Details**

VAR: a varest S3 object created by VAR

#### Note

A varest 2 object can be created by new ("varest 2", ...) or returned by the function getVAR model

# Author(s)

Emanuele Cordano

76 WhereIs

#### **Examples**

```
showClass("varest2")
```

VAR\_mod

Modified version of VAR function allowing to describe white-noise as VAR-(0) model (i. e. varest objects)

# Description

Modified version of VAR function allowing to describe white-noise as VAR-(0) model (i. e. varest objects)

# Usage

```
VAR_mod(
    y,
    p = 1,
    type = c("const", "trend", "both", "none"),
    season = NULL,
    exogen = NULL,
    lag.max = NULL,
    ic = c("AIC", "HQ", "SC", "FPE")
)
```

# **Arguments**

```
y, p, type, season, exogen, lag.max, ic see VAR function
```

#### Value

a Vector Auto-Regeressive model (VAR) as varest object

WhereIs

Gets the toponym where a meteorological station is located

# **Description**

Gets the toponym where a meteorological station is located

```
WhereIs(name, station_names, location)
```

WhereIs 77

# **Arguments**

name character ID of the station

station\_names vector containing the IDs (characters) of the considered meteorological stations.

An example is STATION\_NAMES defined in the trentino dataset.

location vector containing the toponyms. An example is LOCATION defined in the trentino

dataset.

#### Value

the location toponym given the vectors of station IDs and the respective location toponyms

# Author(s)

Emanuele Cordano, Emanuele Eccel

# Examples

```
data(trentino)
WhereIs("T0099",station_names=STATION_NAMES,location=LOCATION)
```

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