

# Package ‘stUPscales’

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**Type** Package

**Title** Spatio-Temporal Uncertainty Propagation Across Multiple Scales

**Version** 1.0.3.5

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**Description** Integrated environmental modelling requires coupling sub-models at different spatial and temporal scales, thus accounting for change of support procedures (aggregation and disaggregation). We contribute to state-of-the-art open source tools that support uncertainty propagation analysis in temporal and spatio-temporal domains. We implement the tool for uncertainty propagation in environmental modelling, with examples in the urban water domain. The functionalities of the class setup and the methods and functions MC.setup, MC.sim, MC.analysis, MC.analysis\_generic and Agg.t are contained, which are used for setting up, running and analysing Monte Carlo uncertainty propagation simulations, and for spatio-temporal aggregation. We also implement functionalities to model and predict variables that vary in space and time. stUPscales takes uncertainty characterisation and propagation a step further by including temporal and spatio-temporal auto- and cross-correlation, resulting in more realistic (spatio-)temporal series of environmental variables. Due to its modularity, the package allows the implementation of additional methods and functions for spatio-temporal disaggregation of model inputs and outputs, when linking models across multiple space-time scales.

**License** GPL (>=3) + file LICENSE

**Depends** R (>= 2.10),

methods,

stats,

graphics,

grDevices,

utils,

mAr,

lmom,

**Imports** parallel,  
doParallel,  
foreach,  
lattice,  
msm,  
ggplot2,  
moments,  
hydroGOF,  
zoo,  
data.table,  
xts,  
EmiStatR

**Suggests** sp, spacetime

**NeedsCompilation** no

## R topics documented:

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stUPscales-package

*Spatio-Temporal Uncertainty Propagation Across Multiple Scales***Description**

Integrated environmental modelling requires coupling sub-models at different spatial and temporal scales, thus accounting for change of support procedures (aggregation and disaggregation). We contribute to state-of-the-art open source tools that support uncertainty propagation analysis in temporal and spatio-temporal domains. We implement the tool for uncertainty propagation in environmental modelling, with examples in the urban water domain. The functionalities of the class setup and the methods and functions MC.setup, MC.sim, MC.analysis, MC.analysis\_generic and Agg.t are contained, which are used for setting up, running and analysing Monte Carlo uncertainty propagation simulations, and for spatio-temporal aggregation. We also implement functionalities to model and predict variables that vary in space and time. stUPscales takes uncertainty characterisation and propagation a step further by including temporal and spatio-temporal auto- and cross-correlation, resulting in more realistic (spatio-)temporal series of environmental variables. Due to its modularity, the package allows the implementation of additional methods and functions for spatio-temporal disaggregation of model inputs and outputs, when linking models across multiple space-time scales.

**Details**

The DESCRIPTION file:

```
Package:    stUPscales
Type:       Package
Version:    1.0.3.5
Date:       2019-05-19
License:    GPL (>= 3) + file LICENSE
Depends:    R (>= 2.10), methods, stats, graphics, grDevices, utils, mAr, lmom
Imports:    parallel, doParallel, foreach, lattice, msm, ggplot2, moments, hydroGOF, zoo, data.table, xts, EmiStatR
Suggests:   sp, spacetime
```

**Author(s)**

J.A. Torres-Matallana [aut, cre]; U. Leopold [ctb]; G.B.M. Heuvelink [ctb].

Maintainer: J.A. Torres-Matallana.

Agg.t

*Temporal aggregation of environmental variables***Description**

Function for temporal aggregation of environmental variables. Agg is a wrapper function of aggregate from stats package.

**Usage**

```
Agg.t(data, nameData, delta, func, namePlot)
```

**Arguments**

data	A data.frame that contains the time series of the environmental variable to be aggregated, e.g. precipitation. This data.frame should have at two columns: the first one, Time [y-m-d h:m:s]; the second one, a numeric value equal to the magnitude of the environmental variable. If the environmental variable is different than precipitation, then the column name of the values can be named as the name of the variable itself.
nameData	A character string that defines the name of the environmental variable to be aggregated.
delta	A numeric value that specifies the level of aggregation required in minutes.
func	The name of the function of aggregation e.g. mean, sum.
namePlot	A character string that defines the title of the plot generated.

**Value**

A data.frame with two columns:

time	the date-time time series of the aggregated variable
value	time series with the magnitude of the aggregated variable.

**Author(s)**

J.A. Torres-Matallana

**Examples**

```
## temporal aggregation
library(EmiStatR)
data(P1)
colnames(P1) <- c("time", "P1")
head(P1)

library(stUPscales)
P1.agg <- Agg.t(data = P1, nameData = "P1", delta = 120, func = sum,
               namePlot = "Temporal aggregation of precipitation P1")

head(P1.agg)
tail(P1.agg)
```

---

Germany\_precipitation\_201112*Sample precipitation time series in Germany*

---

**Description**

A 1-minute sample event for precipitation time series measured in 37 rain gauge stations distributed over the territory of Germany close to the frontier to the Grand-Duchy of Luxembourg.

**Usage**

```
data("Germany_precipitation_201112")
```

**Format**

The format is:

An 'xts' object on 2011-12-01/2011-12-31 23:59:00 containing:

Data: num [1:44640, 1:37] NA NA NA NA NA NA NA NA NA NA ...

- attr(\*, "dimnames")=List of 2

..\$ : NULL

..\$ : chr [1:37] "Station.1327" "Station.1964" "Station.200" "Station.2170" ...

Indexed by objects of class: [POSIXct,POSIXt] TZ:

xts Attributes:

NULL

**Source**

<https://www.dwd.de/>

**Examples**

```
library(stUPscales)
library(spacetime)

data(Germany_precipitation_201112)

summary(Germany_precipitation_201112)

library(zoo)
par(mfrow = c(5, 1))
for(i in 1:ncol(Germany_precipitation_201112)){
  plot(index(Germany_precipitation_201112), Germany_precipitation_201112[,i],
       typ="l", col="blue", xlab="Time", ylab='Precip. [mm]',
       main=colnames(Germany_precipitation_201112[,i]))
}
```

---

Germany_stations	<i>A SpatialPointsDataFrame with the location of 37 rain gauges in Germany</i>
------------------	--

---

### Description

A SpatialPointsDataFrame with the location of 37 rain gauges distributed over the territory of Germany close to frontier with the Grand-Duchy of Luxembourg. These 37 stations are the same related to the "Germany\_precipitation\_201112" dataset.

### Usage

```
data("Germany_stations")
```

### Format

The format is:

Formal class 'SpatialPointsDataFrame' [package "sp"] with 5 slots

..@ data :'data.frame': 37 obs. of 9 variables:

.. ..\$ Stations\_id : int [1:37] 200 450 460 523 603 723 902 942 953 1327 ...

.. ..\$ von\_datum : int [1:37] 20020924 20050920 19930930 20020807 20071024 20020717 20060618 20020925 19970730 20040707 ...

.. ..\$ bis\_datum : int [1:37] 20180820 20121204 20180820 20180604 20180820 20180820 20180820 20180820 20180820 20180820 ...

.. ..\$ Stationshoehe: int [1:37] 517 120 363 359 159 290 573 308 481 147 ...

.. ..\$ geoBreite : num [1:37] 50.1 49.9 49.3 50 50.7 ...

.. ..\$ geoLaenge : num [1:37] 6.32 7.07 6.69 6.53 7.19 ...

.. ..\$ Stationsname : Factor w/ 1109 levels "Aachen","Aachen-Orsbach",...: 535 90 92 104 118 456 172 179 180 1046 ...

.. ..\$ Bundesland : Factor w/ 16 levels "Baden-Wuerttemberg",...: 11 11 12 11 10 11 10 11 11 10 ...

.. ..\$ d : logi [1:37] NA NA NA NA NA NA ...

..@ coords.nrs : num(0)

..@ coords : num [1:37, 1:2] 90590 144511 117752 106135 152358 ...

.. .. attr(\*, "dimnames")=List of 2

.. .. ..\$ : NULL

.. .. ..\$ : chr [1:2] "coords.x1" "coords.x2"

..@ bbox : num [1:2, 1:2] 82780 31291 156554 200669

.. .. attr(\*, "dimnames")=List of 2

.. .. ..\$ : chr [1:2] "coords.x1" "coords.x2"

.. .. ..\$ : chr [1:2] "min" "max"

..@ proj4string:Formal class 'CRS' [package "sp"] with 1 slot

.. .. ..@ projargs: chr "+init=epsg:2169 +proj=tmerc +lat\_0=49.833333333333334 +lon\_0=6.166666666666667 +k=1 +x\_0=80000 +y\_0=100000 +ellp" | \_\_truncated\_\_

### Source

<https://www.dwd.de/>

**Examples**

```
library(stUPscales)
library(sp)

data(Germany_stations)

str(Germany_stations)

data(Lux_boundary)
plot(Germany_stations)
plot(boundary.Lux, add=TRUE) # Luxembourg boundary
```

GoF

*Wrapper function for the gof function from hydroGOF package***Description**

A wrapper function for the gof function from hydroGOF package

**Usage**

```
GoF(eval, col_sim, col_obs, name)
```

**Arguments**

eval	A matrix or data.frame with n observations of at least two variables: simulations and observations.
col_sim	A numeric value defining the column in eval data.frame that contains the simulated vector time series.
col_obs	A numeric value defining the column in eval data.frame that contains the observed vector time series.
name	A character string that defines the name of the files (.csv and .RData) created with the results. If missing then no files are created.

**Value**

A vector with 20 elements for each one of the following measures of godness-of-fit: 1) ME, mean error; 2) MAE, mean absolute error; 3) MSE, mean squared error; 4) RMSE, root mean square error; 5) NRMSE %, normalized root square error; 6) PBIAS %, percent bias; 7) RSR, Ratio of RMSE to the standard deviation of the observations; 8) rSD, Ratio of Standard Deviations; 9) NSE, Nash-Sutcliffe Efficiency; 10) mNSE, modified Nash-Sutcliffe efficiency; 11) rNSE, relative Nash-Sutcliffe efficiency; 12) d, Index of Agreement; 13) md, Modified index of agreement; 14) rd, Relative Index of Agreement; 15) cp, Coefficient of persistence; 16) r, Pearson product-moment correlation coefficient; 17) R2, Coefficient of Determination; 18) bR2, Coefficient of determination (r2) multiplied by the slope of the regression line between sim and obs; 19) KGE, Kling-Gupta Efficiency; 20) VE, Volumetric Efficiency.

**Author(s)**

J.A. Torres-Matallana

**References**

Mauricio Zambrano-Bigiarini, 2014. hydroGOF: Goodness-of-fit functions for comparison of simulated and observed hydrological time series. R package version 0.3-8.  
<https://CRAN.R-project.org/package=hydroGOF>.

**Examples**

```
library(stUPscales)

data_new <- rnorm(230, .25, .1)
data_new <- cbind(data_new, data_new*1.2)
colnames(data_new) <- c("sim", "obs")
head(data_new)

gof.new <- GoF(data_new, 1, 2)
gof.new

# writing files
gof.new <- GoF(data_new, 1, 2, "GoF_results")
```

---

HS_RW20111216_stfdf	<i>1-hour DWD precipitation radar imagery calibrated in STFDF format</i>
---------------------	--

---

**Description**

Calibrated hourly precipitation radar imagery from the German Weather Service (DWD from the initials in German) at one-kilometer spatial resolution over the Haute-Sure catchment of the Grand-Duchy of Luxembourg. The data was recorded at Neuheilenbach radar station located in territory of Germany, which covers the entire territory of the Grand-Duchy of Luxembourg and surroundings. This sample STFDF (spatio-temporal full data.frame) corresponds to 1-day sample event for precipitation recorded on 16th December 2011.

**Usage**

```
data("HS_RW20111216_stfdf")
```

**Format**

The format is: Formal class 'STFDF' [package "spacetime"] with 4 slots  
 ..@ data :'data.frame': 52900 obs. of 1 variable:  
 .. ..\$ raa01\_rw\_10000.1112160050.dwd...bin: num [1:52900] 1 1 1 1 1 1 1 1 1 ...  
 ..@ sp :Formal class 'SpatialPixels' [package "sp"] with 5 slots  
 .. ..@ grid :Formal class 'GridTopology' [package "sp"] with 3 slots  
 .. .. ..@ cellcentre.offset: Named num [1:2] -327 -4347



```

.. .. .. ..- attr(*, "names")= chr [1:2] "s1" "s2"
.. .. .. ..@ cellsize : num [1:2] 1 1
.. .. .. ..@ cells.dim : int [1:2] 46 46
.. .. ..@ grid.index : int [1:2116] 1 2 3 4 5 6 7 8 9 10 ...
.. .. ..@ coords : num [1:2116, 1:2] -327 -326 -325 -324 -323 ...
.. .. ..- attr(*, "dimnames")=List of 2
.. .. .. ..$ : NULL
.. .. .. ..$ : chr [1:2] "s1" "s2"
.. .. ..@ bbox : num [1:2, 1:2] -327 -4348 -281 -4302
.. .. ..- attr(*, "dimnames")=List of 2
.. .. .. ..$ : chr [1:2] "s1" "s2"
.. .. .. ..$ : chr [1:2] "min" "max"
.. .. ..@ proj4string:Formal class 'CRS' [package "sp"] with 1 slot
.. .. .. ..@ projargs: chr "+proj=stere +lat_0=90 +lat_ts=90 +lon_0=10 +k=0.93301270189
+x_0=0 +y_0=0 +a=6370040 +b=6370040 +units=km +no_defs"
..@ time :An 'xts' object on 2011-12-16 00:50:00/2011-12-17 00:50:00 containing:
Data: int [1:25, 1] 1 2 3 4 5 6 7 8 9 10 ...
- attr(*, "dimnames")=List of 2
..$ : NULL
..$ : chr "timeIndex"
Indexed by objects of class: [POSIXct,POSIXt] TZ:
xts Attributes:
NULL
..@ endTime: POSIXct[1:25], format: "2011-12-16 00:50:00" "2011-12-16 01:50:00" "2011-12-16
02:50:00" "2011-12-16 03:50:00" ...

```

## Source

<https://www.dwd.de/>

## Examples

```

library(stUPscales)
data(HS_RW20111216_stfdf)

library(spacetime)
stplot(HS_RW20111216_stfdf,
       scales=list(draw=TRUE),
       key.space="right", colorkey=TRUE,
       main="1-hour DWD sample precipitation radar imagery calibrated in STFDF format",
       cex=.74, par.strip.text=list(cex=.74))

```

---

HS_RY20111216_stfdf	<i>5-minute DWD precipitation radar imagery non-calibrated in STFDF format</i>
---------------------	--

---

## Description

Non-calibrated 5-minute precipitation radar imagery from the German Weather Service (DWD from the initials in German) at one-kilometer spatial resolution over the Haute-Sure catchment of the Grand-Duchy of Luxembourg. The data was recorded at Neuheilenbach radar station located in territory of Germany, which covers the entire territory of the Grand-Duchy of Luxembourg and surroundings. This sample STFDF (spatio-temporal full data.frame) corresponds to 1-day sample event for precipitation recorded on 16th December 2011.

*Please note that these are un-calibrated radar data.*

## Usage

```
data("HS_RY20111216_stfdf")
```

## Format

The format is:

Formal class 'STFDF' [package "spacetime"] with 4 slots

..@ data : 'data.frame': 609408 obs. of 1 variable:

.. ..\$ raa01.ry\_10000.1112160000.dwd...bin: num [1:609408] 1 2 1 1 1 1 1 1 1 1 ...

..@ sp : Formal class 'SpatialPixels' [package "sp"] with 5 slots

.. ..@ grid : Formal class 'GridTopology' [package "sp"] with 3 slots

.. .. ..@ cellcentre.offset: Named num [1:2] -327 -4347

.. .. .. ..- attr(\*, "names")= chr [1:2] "s1" "s2"

.. .. .. ..@ cellsize : num [1:2] 1 1

.. .. .. ..@ cells.dim : int [1:2] 46 46

.. .. ..@ grid.index : int [1:2116] 1 2 3 4 5 6 7 8 9 10 ...

.. .. ..@ coords : num [1:2116, 1:2] -327 -326 -325 -324 -323 ...

.. .. ..- attr(\*, "dimnames")=List of 2

.. .. .. ..\$ : NULL

.. .. .. ..\$ : chr [1:2] "s1" "s2"

.. .. ..@ bbox : num [1:2, 1:2] -327 -4348 -281 -4302

.. .. ..- attr(\*, "dimnames")=List of 2

.. .. .. ..\$ : chr [1:2] "s1" "s2"

.. .. .. ..\$ : chr [1:2] "min" "max"

.. .. ..@ proj4string: Formal class 'CRS' [package "sp"] with 1 slot

.. .. .. ..@ projargs: chr "+proj=stere +lat\_0=90 +lat\_ts=90 +lon\_0=10 +k=0.93301270189 +x\_0=0 +y\_0=0 +a=6370040 +b=6370040 +units=km +no\_defs"

..@ time : An 'xts' object on 2011-12-16/2011-12-16 23:55:00 containing:

Data: int [1:288, 1] 1 2 3 4 5 6 7 8 9 10 ...

- attr(\*, "dimnames")=List of 2

..\$ : NULL

..\$ : chr "timeIndex"

Indexed by objects of class: [POSIXct,POSIXt] TZ:

xts Attributes:

NULL

..@ endTime: POSIXct[1:288], format: "2011-12-16 00:00:00" "2011-12-16 00:05:00" "2011-12-16 00:10:00" "2011-12-16 00:15:00" ...

**Source**

<https://www.dwd.de/>

**Examples**

```
library(stUPscales)
library(spacetime)

data(HS_RY20111216_stfdf)

sample.idx <- seq.default(from = 1, to = 25, by = 1)
sample      <- HS_RY20111216_stfdf[, sample.idx]

stplot(sample,
        scales=list(draw=TRUE),
        key.space="right", colorkey=TRUE,
        main="5-minute DWD sample precipitation radar imagery non-calibrated in STFDF format",
        cex=.74, par.strip.text=list(cex=.74))
```

---

inputObs-class	<i>Class "inputObs"</i>
----------------	-------------------------

---

**Description**

The class provides a container for inputs required to invoke `Validation_Quantity` and `Validation_Quantity_Agg` methods.

**Objects from the Class**

Objects can be created by calls of the form `inputObs(...)`.

**Slots**

**id:** Object of class "numeric" to define an unique index for the object.

**plot:** Object of class "numeric". One of 0 (no plots are created) or 1 (to create plots).

**delta:** Object of class "list" to define the time step in minutes for temporal aggregation required e.g. `list(P1 = 10, wlt_obs = 10, vol_sim = 10)` for defining the time steps of 10 minutes for the three variables P1, wlt\_obs, vol\_sim.

**observations:** Object of class "list" to define the observed time series.

**lev2vol:** Object of class "list" to define the curve for the relationship level to volume.

**namePlot:** Object of class "character" to define the name of the plot to create.

**legendPosition:** Object of class "list" with three character objects, which define the position of the legend for the top, second and bottom insets of the plot.

**var:** Object of class "character" to define the name of the variable from which the time series simulated are defined.

**Methods**

**Validation\_Quantity\_Agg** signature(x = "input", y = "inputObs"): ...

**Validation\_Quantity** signature(x = "input", y = "inputObs"): ...

**Author(s)**

J.A. Torres-Matallana

**Examples**

```
showClass("inputObs")
```

```
inputObs()
```

---

IsReg.ts

*Wrapper function for function is.regular from zoo package for data.frame objects*

---

**Description**

"IsReg.ts" is a wrapping Function for Function "is.regular" from "zoo" package. Given a time series (ts) a "data.frame" object it is converted into a "xts" object, while the regularity of the object is checked. The first column of the "data.frame" should contain a character string vector to be converted via as.POSIXct accordingly with the date format (format) and time zone (tz).

**Usage**

```
IsReg.ts(data, format, tz)
```

**Arguments**

data	an object of class data.frame containing in its first column a character string vector to be converted via as.POSIXct into a date vector accordingly with the date format (format) and time zone (tz) defined
format	character string giving a date-time format as used by strptime.
tz	a time zone specification to be used for the conversion, if one is required. System-specific, but "" is the current time zone, and "GMT" is UTC (Universal Time, Coordinated). Invalid values are most commonly treated as UTC, on some platforms with a warning.

**Details**

"IsReg" calls the as.POSIXct function from base package to convert an object to one of the two classes used to represent date/times (calendar dates plus time to the nearest second). More details can be found in the "is.regular" function of the "zoo" package.

**Value**

Object of class "list". This object contains 2 elements, the first one contains a character string "\_TSregular" if the xts object created is strict regular, or "\_TSirregular" if it is strict irregular. More details can be found in the "is.regular" function of the "zoo" package.

**Author(s)**

J.A. Torres-Matallana

**Examples**

```
library(EmiStatR)
data("P1")

class(P1)
head(P1)

ts <- IsReg.ts(data = P1, format = "%Y-%m-%d %H:%M:%S", tz = "UTC")
str(ts)

ts[[1]]

head(ts[[2]]); tail(ts[[2]])

plot(ts[[2]], ylab = "Precipitation [mm]")
```

---

Lux\_boundary

*A shapefile for the boundary of the Grand-Duchy of Luxembourg*


---

**Description**

A shapefile for the country boundary of the Grand-Duchy of Luxembourg

**Usage**

```
data("Lux_boundary")
```

**Format**

The format is:  
 Formal class 'SpatialPolygonsDataFrame' [package "sp"] with 5 slots  
 ..@ data :'data.frame': 1 obs. of 3 variables:  
 .. ..\$ cat: int 1  
 .. ..\$ X\_ : Factor w/ 1 level " ": 1  
 .. ..\$ X\_1: Factor w/ 1 level " ": 1  
 ..@ polygons :List of 1  
 .. ..\$ :Formal class 'Polygons' [package "sp"] with 5 slots  
 .. .. ..@ Polygons :List of 1

```

.. .. .. ..$ :Formal class 'Polygon' [package "sp"] with 5 slots
.. .. .. .. ..@ labpt : num [1:2] 74692 93669
.. .. .. .. ..@ area : num 2.6e+09
.. .. .. .. ..@ hole : logi FALSE
.. .. .. .. ..@ ringDir: int 1
.. .. .. .. ..@ coords : num [1:1141, 1:2] 82274 82437 82449 82765 82782 ...
.. .. .. ..@ plotOrder: int 1
.. .. .. ..@ labpt : num [1:2] 74692 93669
.. .. .. ..@ ID : chr "0"
.. .. .. ..@ area : num 2.6e+09
..@ plotOrder : int 1
..@ bbox : num [1:2, 1:2] 49034 57132 106245 138879
.. .. attr(*, "dimnames")=List of 2
.. .. ..$ : chr [1:2] "x" "y"
.. .. ..$ : chr [1:2] "min" "max"
..@ proj4string:Formal class 'CRS' [package "sp"] with 1 slot
.. .. ..@ projargs: chr "+proj=tmerc +lat_0=49.83333333333334 +lon_0=6.166666666666667
+k=1 +x_0=80000 +y_0%%=100000 +ellps=intl +units=m +no_defs"

```

### Examples

```

library(stUPscales)

data(Lux_boundary)

str(boundary.Lux)

```

---

Lux_precipitation	<i>Sample precipitation time series in the Grand-Duchy of Luxembourg</i>
-------------------	--

---

### Description

A 10-hour sample event for precipitation time series measured in 25 rain gauge stations distributed over the territory of the Grand-Duchy of Luxembourg.

### Usage

```
data("Lux_precipitation")
```

### Format

The format is:  
 An 'xts' object on 2011-12-16/2011-12-16 10:00:00 containing:  
 Data: num [1:61, 1:25] 0 0 0 0 0.1 0.1 0.1 0.1 0.1 0.2 ...  
 - attr(\*, "dimnames")=List of 2  
 ..\$ : NULL  
 ..\$ : chr [1:25] "Dahl" "Echternach" "Esch-Sure" "Eschdorf" ...

Indexed by objects of class: [POSIXct,POSIXt] TZ:  
 xts Attributes:  
 NULL

### Source

<http://agrimeteo.lu/>

### Examples

```
data(Lux_precipitation)

library(xts)
head(event.subset.xts)
tail(event.subset.xts)

plot(event.subset.xts)
```

---

Lux\_precipitation\_2010\_2011

*Sample precipitation time series in the Grand-Duchy of Luxembourg  
 (2-year period)*

---

### Description

A 2-year period sample for precipitation time series measured at 10-minute time step in 25 rain gauge stations distributed over the territory of the Grand-Duchy of Luxembourg.

### Usage

```
data("Lux_precipitation_2010_2011")
```

### Format

The format is:  
 An 'xts' object on 2010-01-01/2011-12-31 23:50:00 containing:  
 Data: num [1:105120, 1:25] 0 0 0 0 0 0 0 0 0 ...  
 - attr(\*, "dimnames")=List of 2  
 ..\$ : NULL  
 ..\$ : chr [1:25] "Dahl" "Echternach" "Esch-Sure" "Eschdorf" ...  
 Indexed by objects of class: [POSIXct,POSIXt] TZ:  
 xts Attributes:  
 NULL

### Source

<http://agrimeteo.lu/>

## Examples

```
library(stUPscales)
data(Lux_precipitation_2010_2011)

library(xts)
head(Lux_precipitation_2010_2011)
tail(Lux_precipitation_2010_2011)

plot(Lux_precipitation_2010_2011)
```

---

Lux_stations	<i>A SpatialPointsDataFrame with the location of 25 rain gauges in Luxembourg</i>
--------------	---

---

## Description

A SpatialPointsDataFrame with the location of 25 rain gauges distributed over the territory of the Grand-Duchy of Luxembourg. These 25 stations are the same related to the "event.subset.xts" dataset.

## Usage

```
data("Lux_stations")
```

## Format

The format is:

Formal class 'SpatialPointsDataFrame' [package "sp"] with 5 slots

..@ data : 'data.frame': 25 obs. of 8 variables:

.. ..\$ id : Factor w/ 25 levels "1","11","12",...: 4 5 6 7 8 9 10 11 25 1 ...

.. ..\$ name : Factor w/ 25 levels "Arisdorf","Christnach",...: 4 5 7 6 8 9 10 11 22 1 ...

.. ..\$ north\_lure: Factor w/ 24 levels "101950","102913",...: 6 NA 4 3 1 5 22 20 21 2 ...

.. ..\$ east\_luref: Factor w/ 25 levels "56584","56990",...: 10 25 5 7 16 19 20 24 17 1 ...

.. ..\$ elev\_luref: Factor w/ 25 levels "190","202","207",...: 22 5 16 25 2 11 14 1 6 20 ...

.. ..\$ station\_ty: Factor w/ 1 level "1": 1 1 1 1 1 1 1 1 1 1 ...

.. ..\$ management: Factor w/ 1 level "ASTA": 1 1 1 1 1 1 1 1 1 1 ...

.. ..\$ telemetry : Factor w/ 0 levels: NA NA NA NA NA NA NA NA NA NA ...

..@ coords.nrs : num(0)

..@ coords : num [1:25, 1:2] 66562 99810 62258 63363 74929 ...

.. .. attr(\*, "dimnames")=List of 2

.. .. ..\$ : NULL

.. .. ..\$ : chr [1:2] "coords.x1" "coords.x2"

..@ bbox : num [1:2, 1:2] 56584 64215 99810 132012

.. .. attr(\*, "dimnames")=List of 2

.. .. ..\$ : chr [1:2] "coords.x1" "coords.x2"

.. .. ..\$ : chr [1:2] "min" "max"

..@ proj4string: Formal class 'CRS' [package "sp"] with 1 slot



```
.. .. @ projargs: chr "+proj=tmerc +lat_0=49.83333333333334 +lon_0=6.166666666666667
+k=1 +x_0=80000 +y_0=100000 +ellps=intl +units=m +no_defs"
```

### Source

<http://agrimeteo.lu/>

### Examples

```
library(stUPscales)
data(Lux_stations)

str(stations)

library(sp)
plot(stations)
```

---

MC.analysis

*Analysis of the Monte Carlo simulation*

---

### Description

Function for running the analysis of the Monte Carlo simulation.

### Usage

```
MC.analysis(x, delta, qUpper, p1.det, sim.det, event.ini, event.end,
ntick, summ.data = NULL)
```

### Arguments

x	A list .
delta	A numeric value that specifies the level of aggregation required in minutes.
qUpper	A character string that defines the upper percentile to plot the confidence band of results, several options are possible "q999" the 99.9th percentile, "q995" the 99.5th percentile, "q99" the 99th percentile, "q95" the 95th percentile, "q50" the 50th percentile. The lower boundary of the confidence band (showed in gray in the output plots) is the 5th percentile in all cases.
p1.det	A data.frame that contains the time series of the main driving force of the system to be simulated deterministically, e.g. precipitation. This data.frame should have only two columns: the first one, Time [y-m-d h:m:s]; the second one, a numeric value equal to the magnitude of the environmental variable.
sim.det	A list that contains the results of the deterministic simulation, here the output of EmiStatR given p1.det. See the method EmiStatR from the homonym package for details.

<code>event.ini</code>	A time-date string in POSIXct format that defines the initial time for event analysis.
<code>event.end</code>	A time-date string in POSIXct format that defines the final time for event analysis.
<code>ntick</code>	A numeric value to specify the number of ticks in the x-axis for the event time-window plots.
<code>summ.data</code>	A list by default NULL. If provided, the list should contain an output of the MC.analysis function, and the analysis is done again without the calculation of some of the internal variables, therefore the analysis is faster.

**Value**

A list of length 2:

<code>summ</code>	A list that contains the summary statistics of the Monte Carlo simulation per output variable. Each output variable is summarised by calculating the mean "Mean", standard deviation "sd", variance "Variance", 5th, 25th, 50th, 75th, 95th, 99.5th, 99.9th percentiles "q05", "q25", "q50", "q75", "q95", "q995", "q999", the max "Max", the sum "Sum", time "time", and the deterministic precipitation "p1", all variables as time series.
<code>variance</code>	A data.frame that contains the summary statistics of the variance of the Monte Carlo simulation per output variable.

**Author(s)**

J.A. Torres-Matallana

**See Also**

See also [setup-class](#), [MC.setup-methods](#), [MC.sim-methods](#).

**Examples**

```
## the Monte Carlo simulation: MC.sim
library(EmiStatR)

# library(xts)
# data(Esch_Sure2010)
# P <- IsReg(Esch_Sure2010, format = "%Y-%m-%d %H:%M:%S", tz = "CET")
# P1 <- P[[2]]
# P1 <- P1["2010-08",][1:55]
# P1 <- cbind.data.frame(time=index(P1), P1 = coredata(P1))

data(P1)
P1 <- P1[165:(110*2),]
plot(P1[,2], typ="l")

library(stUPscales)

setting_EmiStatR <- setup(id      = "MC_sim1",
```

```

nsim      = 3, # # use a larger number to have
           # a proper confidence band of simulations
seed      = 123,
mcCores   = 1,
ts.input  = P1,
rng       = rng <- list(
  qs      = 150, # [l/PE/d]
CODs = c(pdf = "nor", mu = 4.378, sigma = 0.751), # log[g/PE/d]
NH4s = c(pdf = "nor", mu = 1.473, sigma = 0.410), # log[g/PE/d]
  qf      = 0.04, # [l/s/ha]
  CODf    = 0, # [g/PE/d]
  NH4f    = 0, # [g/PE/d]
CODr = c(pdf = "nor", mu = 3.60, sigma = 1.45), # 71 log[mg/l]
  NH4r    = 1, # [mg/l]
  nameCSO = "E1", # [-]
  id      = 1, # [-]
  ns      = "FBH Goesdorf", # [-]
  nm      = "Goesdorf", # [-]
  nc      = "Obersauer", # [-]
  numc    = 1, # [-]
  use     = "R/I", # [-]
  Atotal  = 36, # [ha]
  Aimp    = c(pdf = "uni", min = 4.5, max = 25), # [ha]
  Cimp    = c(pdf = "uni", min = 0.25, max = 0.95), # [-]
  Cper    = c(pdf = "uni", min = 0.05, max = 0.60), # [-]
  tfS     = 1, # [time steps]
  pe      = 650, # [PE]
  Qd      = 5, # [l/s]
  Dd      = 0.150, # [m]
  Cd      = 0.18, # [-]
  V       = 190, # [m3]
  lev.ini = 0.10, # [m]
  lev2vol = list(lev = c(0.06, 1.10, 1.30, 3.30), # [m]
    vol = c(0, 31, 45, 190)) # [m3]
),
ar.model  = ar.model <- list(
  CODs    = 0.5,
  NH4s    = 0.5,
  CODr    = 0.7),
var.model = var.model <- list(
  inp     = c("", ""), # c("CODs", "NH4s"), # c("", ""),
  w       = c(0.04778205, 0.02079010),
  A       = matrix(c(9.916452e-01, -8.755558e-05,
    -0.003189094, 0.994553910), nrow=2, ncol=2),
  C       = matrix(c(0.009126591, 0.002237936,
    0.002237936, 0.001850941), nrow=2, ncol=2)))

MC_setup <- MC.setup(setting_EmiStatR)

sims <- MC.sim(x = MC_setup, EmiStatR.cores = 0)

## Monte Carlo simulation analysis: MC.analysis

```

```

# Deterministic simulation
# Definition of structure 1, E1:

E1 <- list(id = 1, ns = "FBH Goesdorf", nm = "Goesdorf", nc = "Obersauer", numc = 1,
          use = "R/I", Atotal = 36, Aimp = 25.2, Cimp = 0.80, Cper = 0.30,
          tfS = 0, pe = 650, Qd = 5,
          Dd = 0.150, Cd = 0.18, V = 190, lev.ini = 0.10,
          lev2vol = list(lev = c(.06, 1.10, 1.30, 3.30),
                        vol = c(0, 31, 45, 190))
          )

# Defining deterministic input:
library(EmiStatR)
# data(P1)

input.det <- input(spatial = 0, zero = 1e-5,
                  folder = system.file("shiny", package = "EmiStatR"),
                  cores = 0,
                  ww = list(qs = 150, CODs = 104, NH4s = 4.7),
                  inf = list(qf = 0.04, CODf = 0, NH4f = 0),
                  rw = list(CODr = 71, NH4r = 1, stat = "Dahl"),
                  P1 = P1, st = list(E1=E1), export = 0)

# Invoking `EmiStatR` with the deterministic input:
sim.det <- EmiStatR(input.det)

# further arguments
delta <- 10 # minutes
qUpper <- "q999"

event.ini <- as.POSIXct("2016-01-02 03:20:00")
event.end <- as.POSIXct("2016-01-02 12:30:00")

# uncomment to run:
# new_analysis <- MC.analysis(x = sims, delta = delta, qUpper = qUpper, p1.det = P1,
#                             sim.det = sim.det, event.ini = event.ini, event.end = event.end,
#                             ntick = 5, summ.data = NULL)

```

---

MC.analysis\_generic      *Analysis of the Monte Carlo simulation (general function)*

---

## Description

General function for running the analysis of the Monte Carlo simulation.

## Usage

```
MC.analysis_generic(x, delta, qUpper, data.det, sim.det, event.ini, event.end,
                  ntick, summ.data = NULL)
```

**Arguments**

<code>x</code>	A list of 1, which contains the output of the Monte Carlo simulation as a <code>data.frame</code> with <code>n</code> rows as time steps and the first column is time in format <code>POSIXct</code> and <code>m</code> columns named 1, 2, 3... <code>m</code> , where <code>m</code> is the number of Monte Carlo runs results.
<code>delta</code>	A numeric value that specifies the level of aggregation required in minutes.
<code>qUpper</code>	A character string that defines the upper percentile to plot the confidence band of results, several options are possible "q999" the 99.9th percentile, "q995" the 99.5th percentile, "q99" the 99th percentile, "q95" the 95th percentile, "q50" the 50th percentile. The lower boundary of the confidence band (showed in gray in the output plots) is the 5th percentile in all cases.
<code>data.det</code>	A <code>data.frame</code> that contains the time series of the main driving force of the system to be simulated deterministically, e.g. precipitation. This <code>data.frame</code> should have only two columns: the first one, Time [y-m-d h:m:s] in <code>POSIXct</code> format; the second one, a numeric value equal to the magnitude of the variable.
<code>sim.det</code>	A list of 1 that contains the results of the deterministic simulation, here the output given <code>data.det</code> . The format is the same as <code>data.det</code> .
<code>event.ini</code>	A time-date string in <code>POSIXct</code> format that defines the initial time for event analysis.
<code>event.end</code>	A time-date string in <code>POSIXct</code> format that defines the final time for event analysis.
<code>ntick</code>	A numeric value to specify the number of ticks in the x-axis for the event time-window plots.
<code>summ.data</code>	A list by default <code>NULL</code> . If provided, the list should contain an output of the <code>MC.analysis</code> function, and the analysis is done again without the calculation of some of the internal variables, therefore the analysis is faster.

**Value**

A list of length 2:

<code>summ</code>	A list that contains the summary statistics of the Monte Carlo simulation per output variable. Each output variable is summarised by calculating the mean "Mean", standard deviation "sd", variance "Variance", 5th, 25th, 50th, 75th, 95th, 99.5th, 99.9th percentiles "q05", "q25", "q50", "q75", "q95", "q995", "q999", the max "Max", the sum "Sum", time "time", and the deterministic data "p1", all variables as time series.
<code>variance</code>	A <code>data.frame</code> that contains the summary statistics of the variance of the Monte Carlo simulation per output variable.

**Author(s)**

J.A. Torres-Matallana

**See Also**

See also [setup-class](#), [MC.setup-methods](#), [MC.sim-methods](#).

**Examples**

```

## Creating meta-model
Model <- function(A, B, variable.1, variable.2){
  lum <- A*variable.1 + B*variable.2
}

## Model input and parameter set-up

time <- data.frame(time = seq.POSIXt(from = as.POSIXct("2019-01-01"),
                                     to = as.POSIXct("2019-01-02"), by = 60*60*6))

data <- cbind(time, data = 25)
data

new.setup <- setup(id = "MC_1",
                  nsim = 10,
                  seed = 123,
                  mcCores = 1,
                  ts.input = data,
                  rng = rng <- list(
                    A = 1.25,
                    B = 0.75,
                    variable.1 = c(pdf = "uni", min = 0, max = 4),
                    variable.2 = c(pdf = "uni", min = 2.2, max = 3.2)
                  )
)

str(new.setup)

## Monte Carlo simulation set-up
set.seed(slot(new.setup, "seed"))

new.mc.setup <- MC.setup(new.setup)
str(new.mc.setup)

## Monte Carlo simulation
output <- data.frame(time = new.mc.setup$ts.input[,1])
output[,2:(new.mc.setup$nsim + 1)] <- NA

for(i in 1:new.mc.setup$nsim){
  for(j in 1:nrow(new.mc.setup$ts.input)){

    ## model parameter definition
    A <- new.mc.setup$par$A
    B <- new.mc.setup$par$B

    ## model input definition
    variable.1 <- new.mc.setup$par$variable.1[i,j]
    variable.2 <- new.mc.setup$par$variable.2[i,j]

    ## model evaluation
    output[j,i+1] <- Model(A, B, variable.1, variable.2)
  }
}

```

```

}

output <- list(output1 = output)
output

## Deterministic simulation
# model parameter definition
A <- new.mc.setup$par$A
B <- new.mc.setup$par$B

# model input definition
variable.1.det <- apply(X = new.mc.setup$par$variable.1, MARGIN = 2, FUN = mean)
variable.2.det <- apply(X = new.mc.setup$par$variable.2, MARGIN = 2, FUN = mean)

output.det      <- Model(A, B, variable.1.det, variable.2.det)
output.det      <- cbind(time, output.det)
output.det      <- list(out1 = output.det)
str(output.det)

## Monte Carlo analysis
delta      <- 60*6 # minutes
qUpper     <- "q95"
event.ini  <- data$time[1]
event.end  <- data$time[nrow(data)]
ntick      <- 1

analysis <- MC.analysis_generic(x = output, delta = delta, qUpper = qUpper, data.det = data,
                                sim.det = output.det, event.ini = event.ini, event.end = event.end,
                                ntick = ntick)

```

---

MC.calibra-methods      *Methods for Function MC.calibra*

---

## Description

Given the arguments of the method a calibration routine takes place. Method used only for internal purpose.

## Methods

```
signature(x = "list", obs = "inputObs", EmiStatR.cores = "numeric")
```

**Description**

Given an object of class `setup`, the method can be invoked for setting-up the Monte Carlo simulation. The variables are sampled accordingly to their parameters specified in the slot `rng` of the `setup` object. If `ar.model` is defined in slot `ar.model`, then the specified variables are sampled from the pdf nor as an autorregresive (AR) model via the function `arma.sim` from base package `stats`. If `var.model` is defined in slot `var.model`, then the specified variables are sampled from the pdf nor as an vector autorregresive (VAR) model via the function `mAr.sim` from package `mAr` (see Barbosa, 2015, and Luetkepohl, 2005, for details). See `setup-class` for further details to define the AR and VAR models.

**Usage**

```
MC.setup(x)
```

**Arguments**

`x`                      an object of class `setup`.

**Methods**

```
signature(x = "setup")
```

**Author(s)**

J.A Torres-Matallana

**References**

S. M. Barbosa, Package "mAr": Multivariate AutoRegressive analysis, 1.1-2, The Comprehensive R Archive Network, CRAN, 2015.

H. Luetkepohl, New Introduction to Multiple Time Series Analysis, Springer, 2005.

**Examples**

```
# loading a precipitation time series as input for the setup class

library(EmiStatR)
data(P1)

# A setup with three variables to be considered in the Monte Carlo simulation:
# var1, a constant value variable; var2, a variable sampled from a uniform (uni)
# probability distribution function (pdf) with parameters min and max;
# var3, a variable sampled from a normal (nor) pdf with parameteres mu and sigma
```



```

ini <- setup(id = "MC_sim1", nsim = 500, seed = 123, mcCores = 1, ts.input = P1,
            rng = list(var1 = 150, var2 = c(pdf = "uni", min = 50, max = 110),
                      var3 = c(pdf = "nor", mu = 90, sigma = 2.25))
)

MC_setup <- MC.setup(ini)
str(MC_setup)

## definition of AR models for variables var2 and var3 with AR coefficients 0.995 and 0.460

library(EmiStatR)
data(P1)

ini_ar <- setup(id = "MC_sim1_ar", nsim = 500, seed = 123, mcCores = 1, ts.input = P1,
               rng = list(var1 = 150, var2 = c(pdf = "nor", mu = 150, sigma = 5),
                         var3 = c(pdf = "nor", mu = 90, sigma = 2.25)),
               ar.model = ar.model <- list(var2 = 0.995, var3 = 0.460)
)

MC_setup_ar <- MC.setup(ini_ar)
str(MC_setup_ar)

## definition of a bi-variate VAR model for variables var2 and var3

ini_var <- setup(id = "MC_sim1_ar", nsim = 500, seed = 123, mcCores = 1, ts.input = P1,
                rng = rng <- list(var1 = 150,
                                var2 = c(pdf = "nor", mu = 150, sigma = 5),
                                var3 = c(pdf = "nor", mu = 90, sigma = 2.25)),
                var.model = var.model <- list( inp = c("var2", "var3"),
                                              w = c(0.048, 0.021),
                                              A = matrix(c(0.992, -8.8e-05, -31e-4, 0.995),
                                                         nrow=2, ncol=2),
                                              C = matrix(c(0.0091, 0.0022, 0.0022, 0.0019),
                                                         nrow=2, ncol=2))
)

MC_setup_var <- MC.setup(ini_var)
str(MC_setup_var)

```

---

MC.sim-methods

---

*~~ Methods for Function MC.sim ~~*


---

## Description

Method to be invoked for running the Monte Carlo simulation. The simulator used is the method EmiStatR from the homonym package. This method should be rewritten for working with another simulator.



```

nc      = "Obersauer", # [-]
numc    = 1,           # [-]
use     = "R/I",      # [-]
Atotal  = 36,          # [ha]
Aimp    = c(pdf = "uni", min = 4.5, max = 25),      # [ha]
Cimp    = c(pdf = "uni", min = 0.25, max = 0.95),  # [-]
Cper    = c(pdf = "uni", min = 0.05, max = 0.60),  # [-]
tfS     = 1,           # [time steps]
pe      = 650,          # [PE]
Qd      = 5,            # [l/s]
Dd      = 0.150,        # [m]
Cd      = 0.18,         # [-]
V       = 190,          # [m3]
lev.ini = 0.10,         # [m]
lev2vol = list(lev = c(.06, 1.10, 1.30, 3.30),     # [m]
               vol = c(0, 31, 45, 190))           # [m3]
),
ar.model = ar.model <- list(
  CODs   = 0.5,
  NH4s   = 0.5,
  CODr   = 0.7),
var.model = var.model <- list(
  inp    = c("", ""), # c("CODs", "NH4s"), # c("", ""),
  w      = c(0.04778205, 0.02079010),
  A      = matrix(c(9.916452e-01, -8.755558e-05,
                    -0.003189094, 0.994553910), nrow=2, ncol=2),
  C      = matrix(c(0.009126591, 0.002237936,
                    0.002237936, 0.001850941), nrow=2, ncol=2)))

MC_setup <- MC.setup(setting_EmiStatR)

sims <- MC.sim(x = MC_setup, EmiStatR.cores = 0)
str(sims)

```

MC.summary

*Summary statistics computation of Monte Carlo simulation***Description**

A function that computes the summary statistics of a Monte Carlo simulation result.

**Usage**

```
MC.summary(p1, data)
```

**Arguments**

**p1** The independent variable. A dataframe object with two columns and number of rows equal to the number of rows of the Monte Carlo simulated data. The first column, named "time", contains the vector of time of the time series in format POSIXct. The second column contains the observations of the time series.

**data** A matrix or a dataframe that contains the results of a Monte Carlo simulation, with number of rows equal to the number of Monte Carlo realizations and number of columns equal to the number of observations i.e. equal to the number of rows of "p1".

### Details

This function is internally invoked by the MC.analysis function to compute the summary statistics of the Monte Carlo simulation under analysis.

### Value

A dataframe with n observations of 15 variables, where n is the number of columns of the "data" argument. The 15 variables are time series with the summary statistics of the Monte Carlo data: 1) idx: an index for the dataset equal to 1; 2) Mean: the mean; 3) Sd: the standard deviation; 4) Variance, the variance; 5) q05: the five percent quantile; 6) q25: the 25 percent quantile; 7) q50: the 50 percent quantile; 8) q75: the 75 percent quantile; 9) q95: the 95 percent quantile; 10) q995: the 99.5 percent quantile; 11) q999: the 99.9 percent quantile; 12) Max: the maximum; 13) Sum: the sum; 14) time: the time; 15) p1: the independent variable.

### Author(s)

J.A. Torres-Matallana

### Examples

```
library(stUPscales)
library(EmiStatR)

data(P1)
colnames(P1)

new_data <- t(matrix(data = rep(runif(nrow(P1), 10, 100), 5), nrow = nrow(P1), ncol = 5))
new_summary <- MC.summary(p1 = P1, data = new_data)
str(new_summary)
head(new_summary)
```

---

MC.summary.agg	<i>Summary statistics computation of aggregated Monte Carlo simulation</i>
----------------	--

---

### Description

A function that computes the summary statistics of aggregated Monte Carlo simulation result.

### Usage

```
MC.summary.agg(summ, det, delta, func.agg, func.agg.p)
```

**Arguments**

summ	A dataframe with n observations of 15 variables, where n is the number observations or time steps of the data. The 15 variables are time series with the summary statistics of the Monte Carlo data. This dataframe is in the format as is described in the MC.summary function value.
det	A dataframe that contains the deterministic simulation.
delta	A numeric value that represents the level of aggregation (required time stemp) in minutes.
func.agg	The aggregation function to be applied to the summ dataframe.
func.agg.p	The aggregation function to be applied to the independent variable p1 from summ dataframe.

**Value**

A dataframe containing the summ data aggregated to the level defined by delta

**Author(s)**

J.A. Torres-Matallana

**See Also**

See Also as [MC.summary](#)

**Examples**

```
library(stUPscales)
library(EmiStatR)

data(P1)
colnames(P1)

new_data <- t(matrix(data = rep(runif(nrow(P1), 10, 100), 5), nrow = nrow(P1), ncol = 5))
new_summary <- MC.summary(p1 = P1, data = new_data)
str(new_summary)
head(new_summary)

# deterministic simulation
det <- rnorm(nrow(P1), 45, .15)

# level of aggregation
delta <- 60*2 # 2 hours

new_summary_agg <- MC.summary.agg(summ = new_summary, det, delta, func.agg = mean, func.agg.p = sum)
str(new_summary_agg)
head(new_summary_agg)
```

PlotEval

*Function to execute evaluation plot***Description**

This function creates an evaluation plot for the Monte Carlo simulation result.

**Usage**

```
PlotEval(eval, ts, gof1, namePlot, pos1, pos2, pos3)
```

**Arguments**

eval	A data.frame with n observations of seven variables: 1) time: A POSIXct object with format "%Y-%m-%d %H:%M:%S" defining the time vector; 2) column 2: a numeric vector containing the values of the observed variable, which is the first variable of the Level2Volume relationship; 3) column 3: a numeric vector containing the values for the second variable of the Level2Volume relationship; 4) column 4: a numeric vector containing the corresponding simulated values for the second variable of the Level2Volume relationship; 5) column 5: a numeric vector containing the difference between the vectors volt_sim and volt_obs. 6) Rainfall: a numeric vector named "Rainfall" containing the values of the driving force variable used in the simulations, e.g. rainfall. 7) column 7: (Optional) a numeric vector containing the values of the driving force variable used in the simulations in other measurement units, e.g. rainfall in intensity units if rainfall is the driving force of the simulations.
ts	An xts object representing the eval data.frame indexed by the time vector of the eval argument: containing six data variables as it is defined by the eval argument: 1) column 2; 2) column 3; 3) column 4; 4) column 5; 5) Rainfall; 6) column 7.
gof1	A matrix with the output of GoF function.
namePlot	A character string defining the name of the plot to be created.
pos1	Location to place the legend on the inside of the first sub-plot frame. Can be one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right" and "center".
pos2	Location to place the legend on the inside of the second sub-plot frame. Can be one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right" and "center".
pos3	Location to place the legend on the inside of the third sub-plot frame. Can be one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right" and "center".

**Value**

The function creates a plot in the current working directory with the goodness-of-fit between simulations and observations. The plot is provided in pdf format.

**Author(s)**

J.A. Torres-Matallana

**Examples**

```

time <- seq(from = as.POSIXct("2017-11-09"), by = 60*60*24, length.out = 230) # the time vector
data <- cbind.data.frame(time, NA) # a NA vector
data[,3] <- rnorm(230, .25, .1) # random normal distributed data, obs
data[,4] <- data[,3]*1.2 # positive correlated data, sim
data[,5] <- data[,4] - data[,3] # difference sim and obs
data[,6] <- 0 # driving force
data[,7] <- NA # a NA vector

colnames(data) <- c("time", "var1", "obs", "sim", "difference", "Rainfall", "Rainfall2")
head(data)

ts <- IsReg.ts(data, "%Y-%m-%d", "ECT")
ts <- ts[[2]]

gof.new <- GoF(data, 4, 3, "")
gof.new

PlotEval(data, ts, gof.new, "ExamplePlot", "topright", "topright", "topright")

```

PlotMC.event

*A plot function for time series events***Description**

This is an internal function invoked by `MC.analysis` function to generate an event plot of the time series under analysis. A event means a time series with length lower to one month i.e. sub-monthly time series.

**Usage**

```
PlotMC.event(summ, summ1, obs, det.var, det.var1, namePlot, ylab, ylab1, ntick, qUpper)
```

**Arguments**

<code>summ</code>	A data.frame with n observations of m variables as is provided by the output of function <code>MC.summary.agg</code> for the first variable to be plotted.
<code>summ1</code>	A data.frame with n observations of m variables as is provided by the output of function <code>MC.summary.agg</code> for the second variable to be plotted.
<code>obs</code>	A numeric value equal to 0. used for internal use.
<code>det.var</code>	A character string defining the name of the first variable from <code>summ</code> object to be plotted.

det.var1	A character string defining the name of the second variable from summ object to be plotted.
namePlot	A character string defining the name of the plot. The file created with the plot has this name.
ylab	A character string to define the label of the axes y for the first variable sub-plot.
ylab1	A character string to define the label of the axes y for the second variable sub-plot.
ntick	A numeric value integer which defines the number of tick marks in the axis x of the sub-plots.
qUpper	A character string that defines the upper percentile to plot the confidence band of results, several options are possible "q999" the 99.9th percentile, "q995" the 99.5th percentile, "q99" the 99th percentile, "q95" the 95th percentile, "q50" the 50th percentile. The lower boundary of the confidence band (showed in gray in the output plots) is the 5th percentile in all cases.

### Value

The function creates the plot in the current working directory. The format of the plot is pdf.

### Author(s)

J.A. Torres-Matallana

### Examples

```
library(stUPscales)
library(EmiStatR)

# definition of the first summary.agg object
data("P1")
P1 <- P1[1:1100,]

new_data <- matrix(data = NA, nrow = nrow(P1), ncol = 55)
for(i in 1:55){
  new_data[,i] <- matrix(data = rnorm(nrow(P1), 45, 15),
                        nrow = nrow(P1), ncol = 1)
}
new_data <- t(new_data)

new_summary <- MC.summary(p1 = P1, data = new_data)

# deterministic simulation
det <- rnorm(nrow(P1), 45, 15)
det <- cbind(det, rnorm(nrow(P1), 55, 23))
colnames(det) <- c("det1", "det2")

# level of aggregation
delta <- 60*2 # 2 hours

new_summary_agg <- MC.summary.agg(summ = new_summary, det, delta,
```



```

func.agg = mean, func.agg.p = sum)

# definition of the second summary.agg object
new_data1 <- matrix(data = NA, nrow = nrow(P1), ncol = 55)
for(i in 1:55){
  new_data1[,i] <- matrix(data = rnorm(nrow(P1), 55, 23),
                          nrow = nrow(P1), ncol = 1)
}
new_data1 <- t(new_data1)

new_summary1 <- MC.summary(p1 = P1, data = new_data1)
new_summary_agg1 <- MC.summary.agg(summ = new_summary1, det, delta,
                                   func.agg = mean, func.agg.p = sum)

# creating the plot for the event
PlotMC.event(summ = new_summary_agg, summ1 = new_summary_agg1, obs = 0,
             det.var = "det1", det.var1 = "det2", namePlot = "ExamplePlot",
             ylab = "Variable 1 [units]", ylab1 = "Variable 2 [units]",
             ntick=10, qUpper= "q95")

```

---

PlotMC.season

*A plot function for time series seasons*


---

## Description

This is an internal function invoked by MC.analysis function to generate a season plot of the time series under analysis. A season means a time series with length greater to one month e.g. montly, yearly, decadal time series.

## Usage

```
PlotMC.season(summ1, namePlot, ylab, qUpper)
```

## Arguments

summ1	A data.frame with n observations of m variables as is provided by the output of function MC.summary.agg for the variable to be plotted, which the summary was computed.
namePlot	A character string defining the name of the plot. The file created with the plot has this name.
ylab	A character string to define the label of the axes y for the variable to plot.
qUpper	A character string that defines the upper percentile to plot the confidence band of results, several options are possible "q999" the 99.9th percentile, "q995" the 99.5th percentile, "q99" the 99th percentile, "q95" the 95th percentile, "q50" the 50th percentile. The lower boundary of the confidence band (showed in gray in the output plots) is the 5th percentile in all cases.

**Value**

The function creates the plot in the current working directory. The format of the plot is pdf.

**Author(s)**

J.A. Torres-Matallana

**Examples**

```
library(stUPscales)
library(EmiStatR)

data("P1")
P1 <- P1[1:550,]
new_data <- matrix(data = NA, nrow = nrow(P1), ncol = 55)
for(i in 1:55){
  new_data[,i] <- matrix(data = rnorm(nrow(P1), 22, 11),
                        nrow = nrow(P1), ncol = 1)
}
new_data <- t(new_data)

new_summary <- MC.summary(p1 = P1, data = new_data)
head(new_summary)
dim(new_summary)

new_summary$month <- strftime(new_summary[, "time"], format = "%Y-%m")

PlotMC.season(summ1 = new_summary, namePlot = "ExamplePlot",
              ylab = "Variable 1 [units]", qUpper = "q95")
```

---

 setup-class

---

 Class "setup"

---

**Description**

Class to create objects of signature setup. setup object should be passed to the method MC.setup.

**Objects from the Class**

Objects can be created by calls of the form setup().

**Slots**

id: Object of class "character" to identify the Monte Carlo simulation.

nsim: Object of class "numeric" to specify the number of Monte Carlo runs.

seed: Object of class "numeric" to specify the seed of the random numbers generator.

- mcCores:** Object of class "numeric" to specify the number of cores (CPUs) to be used in the Monte Carlo simulation.
- ts.input:** Object of class "data.frame" that contains the time series of the main driving force of the system to be simulated, e.g. precipitation. This data.frame should have at least two columns: the first one, Time [y-m-d h:m:s]; the second one, a numeric value equal to the magnitude of the environmental variable. This data.frame can also contain more than one column to allow several time series in several columns. If the data.frame has more than two columns, then the number of columns should be at least equal to nsim. If the number of columns is greater than nsim, the columns in excess are not recycled because the simulation will last nsim iterations.
- rng:** Object of class "list" that contains the names and values of the variables to be used in the Monte Carlo simulation. Five modes are available: 1) constant value, i.e. this variable will have a constant value along the Monte Carlo simulation; 2) a variable sampled from a uniform (uni) probability distribution function (pdf) with parameters for the lower boundary min and upper boundary max; 3) a variable sampled from a normal (nor) pdf with parameters mean mu and standard deviation sigma; 4) a variable sampled from an autoregressive (AR) model and normal (nor) pdf with parameters mean mu and standard deviation sigma, the coefficients of the AR model should be defined in the slot ar.model; 5) a variable sampled from a vector autoregressive (VAR) model and normal (nor) pdf with parameters mean mu and standard deviation sigma, this mode is enabled by defining the vector of intercept terms w, the matrix of AR coefficients A, and the noise covariance matrix C in the slot var.model. See examples for the definition of this slot.
- ar.model:** Object of class "list" containing the coefficients of the AR model as vectors which name is the variable to be modeled and length the order of the model as is required for function arima.sim from the base package stats. The named variables here should correspond to a pdf nor in the slot rng. See examples for the definition of this slot.
- var.model:** Object of class "list" containing the the vector of intercept terms w, the matrix of AR coefficients A, and the noise covariance matrix C of the VAR model which name is the variable to be modeled and length the order of the model as is required for function mAr.sim from the package mAr. The named variables in this slot should correspond to a pdf nor in the slot rng. The current implementation considers the bi-variate case. See examples for the definition of this slot. For mathematical details see Luetkepohl (2005).

## Methods

**MC.setup** signature(x = "setup"): execute MC.setup function

## Author(s)

J.A Torres-Matallana

## References

- S. M. Barbosa, Package "mAr": Multivariate AutoRegressive analysis, 1.1-2, The Comprehensive R Archive Network, CRAN, 2015.
- H.Luetkepohl, New Introduction to Multiple Time Series Analysis, Springer, 2005.

**Examples**

```

# loading a precipitation time series as input for the setup class

library(EmiStatR)
data(P1)

# A setup with three variables to be considered in the Monte Carlo simulation:
# var1, a constant value variable; var2, a variable sampled from a uniform (uni)
# probability distribution function (pdf) with parameters min and max;
# var3, a variable sampled from a normal (nor) pdf with parameters mu and sigma

ini <- setup(id = "MC_sim1", nsim = 500, seed = 123, mcCores = 1, ts.input = P1,
             rng = list(var1 = 150, var2 = c(pdf = "uni", min = 50, max = 110),
                       var3 = c(pdf = "nor", mu = 90, sigma = 2.25))
)

str(ini)

## definition of AR models for variables var2 and var3 with AR coefficients 0.995 and 0.460

library(EmiStatR)
data(P1)

ini_ar <- setup(id = "MC_sim1_ar", nsim = 500, seed = 123, mcCores = 1, ts.input = P1,
               rng = list(var1 = 150, var2 = c(pdf = "nor", mu = 150, sigma = 5),
                       var3 = c(pdf = "nor", mu = 90, sigma = 2.25)),
               ar.model = ar.model <- list(var2 = 0.995, var3 = 0.460)
)

str(ini_ar)

## definition of a bi-variate VAR model for variables var2 and var3

ini_var <- setup(id = "MC_sim1_ar", nsim = 500, seed = 123, mcCores = 1, ts.input = P1,
               rng = rng <- list(var1 = 150,
                               var2 = c(pdf = "nor", mu = 150, sigma = 5),
                               var3 = c(pdf = "nor", mu = 90, sigma = 2.25)),
               var.model = var.model <- list( inp = c("var2", "var3"),
                                             w = c(0.048, 0.021),
                                             A = matrix(c(0.992, -8.8e-05, -31e-4, 0.995),
                                                         nrow=2, ncol=2),
                                             C = matrix(c(0.0091, 0.0022, 0.0022, 0.0019),
                                                         nrow=2, ncol=2))
)

str(ini_var)

```

**Description**

Given the arguments of the method a validation routine takes place. Method used only for internal purpose.

**Methods**

```
signature(x = "input", y = "inputObs")
```

---

Validation\_Quantity\_Agg-methods

*Methods for Function* Validation\_Quantity\_Agg

---

**Description**

Given the arguments of the method a validation routine takes place. Method used only for internal purpose.

**Methods**

```
signature(x = "input", y = "inputObs")
```

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