# Package 'seqimpute'

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Type Package

Title Imputation of Missing Data in Sequence Analysis

Version 2.2.0

**Description** Multiple imputation of missing data in a dataset using MICT or MICT-timing methods. The core idea of the algorithms is to fill gaps of missing data, which is the typical form of missing data in a longitudinal setting, recursively from their edges. Prediction is based on either a multinomial or random forest regression model. Covariates and time-dependent covariates can be included in the model.

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2 addcluster

# **Contents**

addcluster		Function with the s			ring	resu	t to a s	seqimp (	object obta	iined
Index										19
	summary.seqimp .		 • • •	 						1
	seqwithmiss									
	seqTrans		 	 						16
	seqQuickLook		 	 						15
	seqmissIplot		 	 						14
	seqmissimplic									
	seqmissfplot									
	sequente									
	sequently sequen									
	print.seqimp seqaddNA									
	plot.seqimp									
	gameadd									
	fromseqimp									
	addcluster									

## Description

Function that adds the clustering result to a seqimp object obtained with the seqimpute function

## Usage

```
addcluster(impdata, clustering)
```

## **Arguments**

impdata An object of class seqimp as created by the seqimpute function

clustering made on the multiple imputed dataset. Can either be a dataframe or

a matrix, where each row correspond to an observation and each column to a

multiple imputed dataset

## Value

Returns a seqimp object containing the cluster to which each sequence in each imputed dataset belongs. Specifically, a column named cluster is added to the imputed datasets.

fromseqimp 3

fromseqimp	Transform an object of class seqimp into a dataframe or a mids object

## Description

The function converts a seqimp object into a specified format.

## Usage

```
fromseqimp(data, format = "long", include = FALSE)
```

#### **Arguments**

data An object of class seqimp as created by the function seqimpute

format The format in which the seqimp object should be returned. It could be: "long",

"stacked" and "mids". See the Details section for the interpretation.

include logical that indicates if the original dataset with missing value should be in-

cluded or not. This parameter does not apply if format="mids".

#### **Details**

The argument format specifies the object that should be returned by the function. It can take the following values

"long" produces a data set in which imputed data sets are stacked vertically. The following columns are added: 1) .imp referring to the imputation number, and 2) .id the row names of the original dataset

"stacked" the same as "long", but without the inclusion of the two columns .imp and .id

"mids" produces an object of class mids, which is the format used by the mice package.

#### Value

Transform a seqimp object into the desired format.

## Author(s)

Kevin Emery

```
## Not run:
# Imputation with the MICT algorithm
imp <- seqimpute(data = gameadd, var = 1:4)
# The object imp is transformed to a dataframe, where completed datasets are
# stacked vertically
imp.stacked <- fromseqimp(</pre>
```

plot.seqimp

```
data = imp,
  format = "stacked", include = FALSE
)
## End(Not run)
```

gameadd

Example data set: Game addiction

#### Description

Dataset containing variables on the gaming addiction of young people. The data consists of gaming addiction, coded as either 'no' or 'yes', measured over four consecutive years for 500 individuals, three covariates and one time-dependent covariate. The yearly states are recorded in columns 1 (T1\_abuse) to 4 (T4\_abuse).

The three covariates are

- Gender (female or male),
- Age (measured at time 1),
- Track (school or apprenticeship).

The time-varying covariate consists of the individual's relationship to gambling at each of the four time points, appearing in columns T1\_gambling, T2\_gambling, T3\_gambling, and T4\_gambling. The states are either no, gambler or problematic gambler

#### Usage

```
data(gameadd)
```

#### **Format**

A data frame containing 500 rows, 4 states variable, 3 covariates and a time-dependent covariate.

plot.seqimp

Plot a seqimp object

## Description

Plot a seqimp object. The state distribution plot of the first m completed datasets is shown, possibly alongside the original dataset with missing data

#### Usage

```
## S3 method for class 'seqimp'
plot(x, m = 5, include = TRUE, ...)
```

print.seqimp 5

#### **Arguments**

x Object of class seqimp

m Number of completed datasets to show

include logical that indicates if the original dataset with missing value should be plotted

or not

. . . Arguments to be passed to the seqdplot function

## Author(s)

Kevin Emery

print.seqimp

Print a seqimp object

## Description

Print a seqimp object

#### Usage

```
## S3 method for class 'seqimp' print(x, ...)
```

## **Arguments**

x Object of class seqimp

... additional arguments passed to other functions

## Author(s)

Kevin Emery

seqaddNA

Generation of missing on longitudinal categorical data.

## Description

Generation of missing data in sequence based on a Markovian approach.

6 seqaddNA

#### Usage

```
seqaddNA(
  data,
  var = NULL,
  states.high = NULL,
  propdata = 1,
  pstart.high = 0.1,
  pstart.low = 0.005,
  pcont = 0.66,
  maxgap = 3,
  maxprop = 0.75,
  only.traj = FALSE
)
```

## **Arguments**

data	A data frame containing sequences of a categorical (multinomial) variable, where missing data are coded as NA.
var	A vector specifying the columns of the dataset that contain the trajectories. Default is NULL, meaning all columns are used.
states.high	A list of states with a higher probability of initiating a subsequent missing data gap.
propdata	Proportion of trajectories for which missing data is simulated, as a decimal between 0 and 1.
pstart.high	Probability of starting a missing data gap for the states specified in the states.high argument.
pstart.low	Probability of starting a missing data gap for all other states.
pcont	Probability of a missing data gap to continue.
maxgap	Maximum length of a missing data gap.
maxprop	Maximum proportion of missing data allowed in a sequence, as a decimal between 0 and 1.
only.traj	Logical, if TRUE, only the trajectories (specified in var) are returned. If FALSE, the entire data frame is returned.

## **Details**

The first time point of a trajectory has a pstart.low probability to be missing. For the next time points, the probability to be missing depends on the previous time point. There are four cases:

- 1. If the previous time point is missing and the maximum length of a missing gap, which is specified by the argument maxgap, is reached, the time point is set as observed.
- 2. If the previous time point is missing, but the maximum length of a gap is not reached, there is a point probability that this time point is missing.
- 3. If the previous time point is observed and the previous time point belongs to the list of states specified by pstart.high, the probability to be missing is pstart.high.

seqcomplete 7

4. If the previous time point is observed but the previous time point does not belong to the list of states specified by pstart.high, the probability to be missing is pstart.low.

If the proportion of missing data in a given trajectory exceeds the proportion specified by maxprop, the missing data simulation is repeated for the sequence.

#### Value

A data frame with simulated missing data.

## Author(s)

Kevin Emery

#### **Examples**

```
# Generate MCAR missing data on the mvad dataset
# from the TraMineR package
## Not run:
data(mvad, package = "TraMineR")
mvad.miss <- seqaddNA(mvad, var = 17:86)

# Generate missing data on mvad where joblessness is more likely to trigger
# a missing data gap
mvad.miss2 <- seqaddNA(mvad, var = 17:86, states.high = "joblessness")
## End(Not run)</pre>
```

segcomplete

Extract all the trajectories without missing value.

## **Description**

Extract all the trajectories without missing value.

## Usage

```
seqcomplete(data, var = NULL)
```

## **Arguments**

data either a data frame containing sequences of a multinomial variable with missing

data (coded as NA) or a state sequence object built with the TraMineR package

var the list of columns containing the trajectories. Default is NULL, i.e. all the

columns.

#### Value

Returns either a data frame or a state sequence object, depending the type of data that was provided to the function

#### Author(s)

Kevin Emery

## **Examples**

```
# Game addiction dataset
data(gameadd)
# Extract the trajectories without any missing data
gameadd.complete <- seqcomplete(gameadd, var = 1:4)</pre>
```

seqimpute

seqimpute: Imputation of missing data in longitudinal categorical data

## **Description**

The seqimpute package implements the MICT and MICT-timing methods. These are multiple imputation methods for longitudinal data. The core idea of the algorithms is to fills gaps of missing data, which is the typical form of missing data in a longitudinal setting, recursively from their edges. The prediction is based on either a multinomial or a random forest regression model. Covariates and time-dependent covariates can be included in the model.

The MICT-timing algorithm is an extension of the MICT algorithm designed to address a key limitation of the latter: its assumption that position in the trajectory is irrelevant.

#### Usage

```
seqimpute(
  data,
  var = NULL,
  np = 1,
  nf = 1,
  m = 5,
  timing = FALSE,
  frame.radius = 0,
  covariates = NULL,
  time.covariates = NULL,
  regr = "multinom",
  npt = 1,
  nfi = 1,
  ParExec = FALSE,
  ncores = NULL,
```

```
SetRNGSeed = FALSE,
end.impute = TRUE,
verbose = TRUE,
available = TRUE,
pastDistrib = FALSE,
futureDistrib = FALSE,
...
)
```

## Arguments

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	data	Either a data frame containing sequences of a categorical variable, where missing data are coded as NA, or a state sequence object created using the seqdef function. If using a state sequence object, any "void" elements will also be treated as missing. See the end.impute argument if you wish to skip imputing values at the end of the sequences.
	var	A specifying the columns of the dataset that contain the trajectories. Default is NULL, meaning all columns are used.
	np	Number of prior states to include in the imputation model for internal gaps.
	nf	Number of subsequent states to include in the imputation model for internal gaps.
	m	Number of multiple imputations to perform (default: 5).
	timing	Logical, specifies the imputation algorithm to use. If FALSE, the MICT algorithm is applied; if TRUE, the MICT-timing algorithm is used.
	frame.radius	Integer, relevant only for the MICT-timing algorithm, specifying the radius of the timeframe.
	covariates	List of the columns of the dataset containing covariates to be included in the imputation model.
	time.covariates	
		List of the columns of the dataset with time-varying covariates to include in the imputation model.
	regr	Character specifying the imputation method. Options include "multinom" for multinomial models and "rf" for random forest models.
	npt	Number of prior observations in the imputation model for terminal gaps (i.e., gaps at the end of sequences).
	nfi	Number of future observations in the imputation model for initial gaps (i.e., gaps at the beginning of sequences).
	ParExec	Logical, indicating whether to run multiple imputations in parallel. Setting to TRUE can improve computation time depending on available cores.
	ncores	Integer, specifying the number of cores to use for parallel computation. If unset, defaults to the maximum number of CPU cores minus one.
	SetRNGSeed	Integer, to set the random seed for reproducibility in parallel computations. Note that setting set.seed() alone does not ensure reproducibility in parallel mode.
	end.impute	Logical. If FALSE, missing data at the end of sequences will not be imputed.

verbose Logical, if TRUE, displays progress and warnings in the console. Use FALSE for

silent computation.

available Logical, specifies whether to consider already imputed data in the predictive

model. If TRUE, previous imputations are used; if FALSE, only original data are

considered.

pastDistrib Logical, if TRUE, includes the past distribution as a predictor in the imputation

model.

futureDistrib Logical, if TRUE, includes the future distribution as a predictor in the imputation

model.

... Named arguments that are passed down to the imputation functions.

#### **Details**

The imputation process is divided into several steps, depending on the type of gaps of missing data. The order of imputation of the gaps are:

Internal gap: there is at least np observations before an internal gap and nf after the gap

Initial gap: gaps situated at the very beginning of a trajectory

Terminal gap: gaps situated at the very end of a trajectory

Left-hand side specifically located gap (SLG): gaps that have at least nf observations after the gap, but less than np observation before it

Right-hand side SLG: gaps that have at least np observations before the gap, but less than nf observation after it

Both-hand side SLG: gaps that have less than np observations before the gap, and less than nf observations after it

The primary difference between the MICT and MICT-timing algorithms lies in their approach to selecting patterns from other sequences for fitting the multinomial model. While the MICT algorithm considers all similar patterns regardless of their temporal placement, MICT-timing restricts pattern selection to those that are temporally closest to the missing value. This refinement ensures that the imputation process adequately accounts for temporal dynamics, imping in more accurate imputed values.

#### Value

An object of class seqimp, which is a list with the following elements:

data A data. frame containing the original (incomplete) data.

imp A list of m data. frame corresponding to the imputed datasets.

m The number of imputations.

method A character vector specifying whether MICT or MICT-timing was used.

np Number of prior states included in the imputation model.

nf Number of subsequent states included in the imputation model.

regr A character vector specifying whether multinomial or random forest imputation models were applied.

call The call that created the object.

#### Author(s)

Kevin Emery <a href="mailto:kevin.emery@unige.ch">kevin.emery@unige.ch</a>, Andre Berchtold, Anthony Guinchard, and Kamyar Taher

#### References

Halpin, B. (2012). Multiple imputation for life-course sequence data. Working Paper WP2012-01, Department of Sociology, University of Limerick. http://hdl.handle.net/10344/3639.

Halpin, B. (2013). Imputing sequence data: Extensions to initial and terminal gaps, Stata's. Working Paper WP2013-01, Department of Sociology, University of Limerick. http://hdl.handle.net/10344/3620

Emery, K., Studer, M., & Berchtold, A. (2024). Comparison of imputation methods for univariate categorical longitudinal data. Quality & Quantity, 1-25. https://link.springer.com/article/10.1007/s11135-024-02028-z

```
# Default multiple imputation of the trajectories of game addiction with the
# MICT algorithm
## Not run:
set.seed(5)
imp1 <- seqimpute(data = gameadd, var = 1:4)</pre>
# Default multiple imputation with the MICT-timing algorithm
imp2 <- seqimpute(data = gameadd, var = 1:4, timing = TRUE)</pre>
# Inclusion in the MICt-timing imputation process of the three background
# characteristics (Gender, Age and Track), and the time-varying covariate
# about gambling
set.seed(4)
imp3 <- seqimpute(</pre>
 data = gameadd, var = 1:4, covariates = 5:7,
 time.covariates = 8:11
# Parallel computation
imp4 <- segimpute(</pre>
 data = gameadd, var = 1:4, covariates = 5:7,
 time.covariates = 8:11, ParExec = TRUE, ncores = 5, SetRNGSeed = 2
)
## End(Not run)
```

12 seqmissfplot

seqmissfplot Plot the	most common patterns of missing data.
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## **Description**

This function plots the most frequent patterns of missing data, based on the seqfplot function.

#### Usage

```
seqmissfplot(data, var = NULL, with.complete = TRUE, void.miss = TRUE, ...)
```

## **Arguments**

data	Either a data frame containing sequences of a categorical variable, where missing data are coded as NA, or a state sequence object created using the seqdef function.
var	A vector specifying the columns of the dataset that contain the trajectories. Default is NULL, meaning all columns are used.
with.complete	Logical, if TRUE, complete trajectories will be included in the plot.
void.miss	Logical, if TRUE, treats void elements as missing values. Applies only to state sequence objects created with seqdef. Note that the default behavior of seqdef is to treat missing data at the end of sequences as void elements.
	Additional parameters passed to the seqfplot function.

## **Details**

This plot function is based on the seqfplot function, allowing users to visualize patterns of missing data within sequences. For details on additional customizable arguments, see the seqfplot documentation.

By default, this function plots the 10 most frequent patterns. The number of patterns to be plotted can be adjusted using the idxs argument in seqfplot.

## Author(s)

Kevin Emery

```
# Plot the 10 most common patterns of missing data
seqmissfplot(gameadd, var = 1:4)
# Plot the 10 most common patterns of missing data discarding
# complete trajectories
seqmissfplot(gameadd, var = 1:4, with.missing = FALSE)
```

seqmissimplic 13

```
# Plot only the 5 most common patterns of missing data discarding
# complete trajectories
seqmissfplot(gameadd, var = 1:4, with.missing = FALSE, idxs = 1:5)
```

 ${\tt seqmissimplic}$ 

Identification and visualization of states that best characterize sequences with missing data

## **Description**

This function identifies and visualizes states that best characterize sequences with missing data at each position (time point), comparing them to sequences without missing data at each position (time point). It is based on the seqimplic function. For more information on the methodology, see the seqimplic documentation.

## Usage

```
seqmissimplic(data, var = NULL, void.miss = TRUE, ...)
```

## **Arguments**

data	Either a data frame containing sequences of a categorical variable, where missing data are coded as NA, or a state sequence object created using the seqdef function.
var	A vector specifying the columns of the dataset that contain the trajectories. Default is NULL, meaning all columns are used.
void.miss	Logical, if TRUE, treats void elements as missing values. This argument applies only to state sequence objects created with seqdef. Note that the default behavior of seqdef is to treat missing data at the end of sequences as void elements.
	parameters to be passed to the seqimplic function

#### Value

returns a seqimplic object that can be plotted and printed.

## Author(s)

Kevin Emery

14 seqmissIplot

#### **Examples**

```
# For illustration purpose, we simulate missing data on the mvad dataset,
# available in the TraMineR package. The state "joblessness" state has a
# higher probability of triggering a missing gap

## Not run:
data(mvad, package = "TraMineR")
mvad.miss <- seqaddNA(mvad, var = 17:86, states.high = "joblessness")

# The states that best characterize sequences with missing data
implic <- seqmissimplic(mvad.miss, var = 17:86)

# Visualization of the results
plot(implic)

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

seqmissIplot

Plot all the patterns of missing data.

## **Description**

This function plots all patterns of missing data within sequences, based on the seqIplot function.

#### Usage

```
seqmissIplot(data, var = NULL, with.complete = TRUE, void.miss = TRUE, ...)
```

## Arguments

data	Either a data frame containing sequences of a categorical variable, where missing data are coded as NA, or a state sequence object created using the seqdef function.
var	A vector specifying the columns of the dataset that contain the trajectories. Default is NULL, meaning all columns are used.
with.complete	Logical, if TRUE, complete trajectories will be included in the plot.
void.miss	Logical, if TRUE, treats void elements as missing values. Applies only to state sequence objects created with seqdef. Note that the default behavior of seqdef is to treat missing data at the end of sequences as void elements.
	Additional parameters passed to the seqIplot function.

#### **Details**

This function uses seqIplot to visualize all patterns of missing data within sequences. For further customization options, refer to the seqIplot documentation.

seqQuickLook 15

#### Author(s)

Kevin Emery

#### **Examples**

```
# Plot all the patterns of missing data
seqmissIplot(gameadd, var = 1:4)
# Plot all the patterns of missing data discarding
# complete trajectories
seqmissIplot(gameadd, var = 1:4, with.missing = FALSE)
```

seqQuickLook

Summary of the types of gaps among a dataset

## **Description**

The seqQuickLook() function aimed at providing an overview of the number and size of the different types of gaps spread in the original dataset.

#### Usage

```
seqQuickLook(data, var = NULL, np = 1, nf = 1)
```

## **Arguments**

data	a data.frame where missing data are coded as NA or a state sequence object built with seqdef function
var	the list of columns containing the trajectories. Default is NULL, i.e. all the columns.
np	number of previous observations in the imputation model of the internal gaps.
nf	number of future observations in the imputation model of the internal gaps.

#### **Details**

The distinction between internal and SLG gaps depends on the number of previous (np) and future (nf) observations that are set for the MICT and MICT-timing algorithms.

#### Value

Returns a data.frame object that summarizes, for each type of gaps (Internal Gaps, Initial Gaps, Terminal Gaps, LEFT-hand side SLG, RIGHT-hand side SLG, Both-hand side SLG), the minimum length, the maximum length, the total number of gaps and the total number of missing they contain.

seqTrans

#### Author(s)

Andre Berchtold and Kevin Emery

## **Examples**

```
data(gameadd)
seqQuickLook(data = gameadd, var = 1:4, np = 1, nf = 1)
```

seqTrans

Spotting impossible transitions in longitudinal categorical data

## Description

The purpose of seqTrans is to spot impossible transitions in longitudinal categorical data.

## Usage

```
seqTrans(data, var = NULL, trans)
```

## Arguments

data	a data frame containing sequences of a multinomial variable with missing data (coded as NA)
var	the list of columns containing the trajectories. Default is NULL, i.e. all the columns.
trans	character vector gathering the impossible transitions. For example: trans $<$ c("1->3","1->4","2->1","4->1","4->3")

## Value

It returns a matrix where each row is the position of an impossible transition.

## Author(s)

Andre Berchtold and Kevin Emery

```
data(gameadd)
seqTransList <- seqTrans(data = gameadd, var = 1:4, trans = c("yes->no"))
```

seqwithmiss 17

seqwithmiss	Extract all the trajectories with at least one missing value	

## **Description**

Extract all the trajectories with at least one missing value

## Usage

```
seqwithmiss(data, var = NULL)
```

## **Arguments**

data either a data frame containing sequences of a multinomial variable with missing

data (coded as NA) or a state sequence object built with the TraMineR package

var the list of columns containing the trajectories. Default is NULL, i.e. all the

columns.

## Value

Returns either a data frame or a state sequence object, depending the type of data that was provided to the function

#### Author(s)

Kevin Emery

## **Examples**

```
# Game addiction dataset
data(gameadd)
# Extract the trajectories without any missing data
gameadd.withmiss <- seqwithmiss(gameadd, var = 1:4)</pre>
```

summary.seqimp

Summary of a seqimp object

## **Description**

Summary of a seqimp object

#### Usage

```
## S3 method for class 'seqimp'
summary(object, ...)
```

18 summary.seqimp

# Arguments

object Object of class seqimp

... additional arguments passed to other functions

# Author(s)

Kevin Emery

# **Index**

```
\ast datasets
    gameadd, 4
{\it addcluster}, {\color{red}2}
fromseqimp, 3
{\it gameadd}, {\it 4}
plot.seqimp, 4
print.seqimp, 5
seqaddNA, 5
seqcomplete, 7
seqdef, 9, 12–15
seqfplot, 12
seqimplic, 13
seqimpute, 3, 8
seqIplot, 14
seqmissfplot, 12
{\it seqmissimplic}, 13
seqmissIplot, 14
seqQuickLook, 15
seqTrans, 16
seqwithmiss, 17
summary.seqimp, 17
```