

Package: lifecourse (via r-universe)

August 26, 2024

Type Package

Title Quantification of Lifecourse Fluidity

Version 2.0

Date 2016-09-09

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Description Provides in built datasets and three functions. These functions are `mobility_index`, `nonStanTest` and `linkedLives`. The `mobility_index` function facilitates the calculation of lifecourse fluidity, whilst the `nonStanTest` and the `linkedLives` functions allow the user to determine the probability that the observed sequence data was due to chance. The `linkedLives` function acknowledges the fact that some individuals may have identical sequences. The datasets available provide sequence data on marital status (`maritalData`) and mobility (`mydata`) for a selected group of individuals from the British Household Panel Study (BHPS). In addition, personal and house ID's for 100 individuals are provided in a third dataset (`myHouseID`) from the BHPS.

License GPL

Depends TraMineR,graphics

NeedsCompilation no

Date/Publication 2016-09-13 14:40:35

Repository <https://marmalade-icecream.r-universe.dev>

RemoteUrl <https://github.com/cran/lifecourse>

RemoteRef HEAD

RemoteSha e90ae1691b8cbc1b11f686a677545e44851a0015

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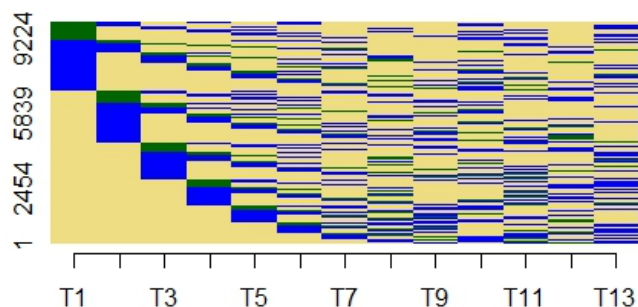
lifecourse-package *lifecourse*

Description

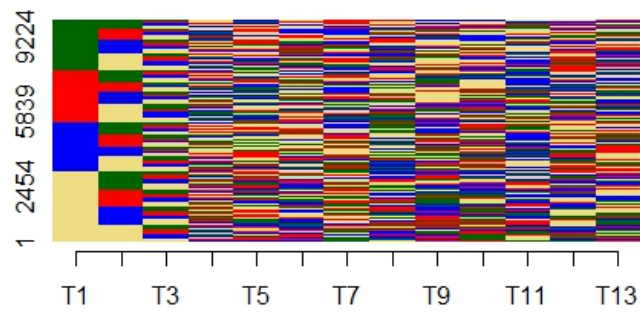
Quantifying lifecourse fluidity.

The mobility index function is provided together with in-built datasets. The mobility index function calculates the mobility index (number of distinct state episodes in a given sequence). The mobility index can be applied to different types of lifecourse channels apart from the migration channel. The output from this function gives an indication of the number of distinct state episodes within the lifecourse and hence the 'mobility' between state episodes. A state episode by definition would contain at least one state.

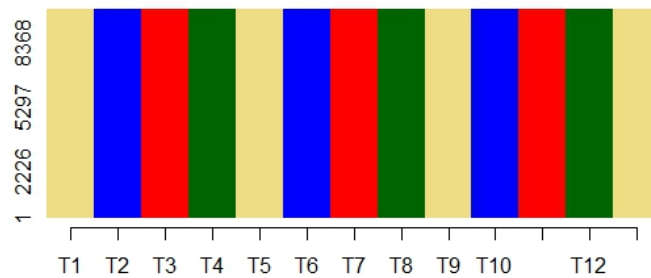
The difference in the average mobility index for three datasets is illustrated in the figures below. In the figures the lifecourse sequences for 10000 individuals (with their ID's on the y axis) over 13 years (denoted by T1 to T13 on the x axis) are shown. Each state can be either of four states (represented by yellow, blue, red and green blocks). In the first figure we note that there is a preponderance of yellow and blue blocks (which represent particular states). This dataset exhibits a high frequency of state episodes represented by either blue and yellow blocks. There are relatively few states represented by the green blocks and none represented by the red blocks. The average mobility index (over the 10000 sequences) is 6.52. This indicates that this dataset exhibits state episodes of high membership (recall that the lowest membership for a state episode is 1). The maximum value for the mobility index is 13 since the maximum number of years in the dataset is 13.



In the figure below, the states have been assigned randomly, and the average membership of the state episodes is much less than that in the previous dataset. Also all four of the possible states are evident in this dataset. For this dataset, the average mobility index is 10.54.



Finally, in the figure below each sequence contains states such that each successive state is different from that preceding it. In this case the mobility index for each sequence would be 13 and the average mobility index is 13. This indicates that this dataset exhibits high mobility between state episodes and the membership per episode is 1.



The in-built datasets are derived from the British Household Panel Survey data (BHPS). The data are derived from the BHPS indresp data files.

Details

Package: lifecourse
 Type: Package
 Version: 1.0
 Date: 2016-03-18

License: GPL

Author(s)

Glenna Nightingale <glenna.evans@gmail.com>

Examples

```
#-----
# obtaining the in-built data derived from BHPS data.
#-----
#data(mydata)
#bbc = mydata
#lbbc = length(mydata[,1])
#-----
# plotting the sequences with the sequence states
# colour coded.
#-----
#balphabet =c("non-mover", "mover within gb")
# specifying the sequence alphabet. For the BHPS mobility data we
# specify "non-mover" and "mover within gb"
#blabels = c("non-mover", "mover within gb")
#bcodes = c("non-mover", "mover within gb")
#bseq = seqdef(bbc[,2:14], alphabet=balphabet, states = bcodes, labels = blabels)
# forming the dataset of sequences
#seqIplot(bseq, sortv = "from.start", cex.legend=1.6, cex.main=3, cex.lab=1.5, cex = 2,
#         cpal=c("lightgoldenrod", "blue"))
# plotting the sequences

#-----
# calculating the mobility index
# for the observed sequences
#-----
#bbcseq = bbc[,2:14]
# removing the first column which contains the ID for the persons involved.
#totalScore = score = 0
# calculating the mobility index for the observed sequences
#for(i in 1:length(bbcseq[,1])){
#   myseq = bbcseq[i,1:13]
#   score = (mobility_index(myseq, c("non-mover", "mover within gb"), 2))
#   totalScore = totalScore + score
#}
#totalScore
#testStatistic = totalScore/(13*lbbc)
# the length of the lifecourse = 13,
#the number of sequences = lbbc
#testStatistic

#-----
```

```

# Running the non-standardization test
#-----
#nll = nonStanTest(bbcseq,balphabet) # obtaining the null distribution
#hist(nll[[1]],main="Null distribution",xlab="Test statistics",ylab="Frequency",col="grey")
# plotting the null distribution to compare with the observed test statistic.
#par(new=T)
#plot(density(nll[[1]],adjust=3),col="blue",axes=F,lwd=4,xlab="",ylab="",main="")
#abline(v=testStatistic,col="orange",lwd=4)
#nll[[2]]# Left Tailed test - pval
#nll[[3]]# Right Tailed test - pval

```

linkedLives

non-standardization test for lifecourses of linked lives

Description

The non-standardization test for lifecourses with linked lives. The simulations carried out for this account for the fact that some individuals are more likely to have identical lifecourse trajectories. Acknowledgements: CPC St Andrews/Edinburgh.

Usage

```
linkedLives(currentData, sequenceUnits, lifecourseSequences)
```

Arguments

`currentData` Dataset containing a sequence of linking ID's (such as house ID's).
`sequenceUnits` Vector containing the sequence alphabet.
`lifecourseSequences` Dataset containing the sequences under investigation.

Author(s)

Glenna Nightingale

Examples

```

data(myHouseID)
data(mydata)
#-----Exploring data-----
length(unique(myHouseID[,2]))
# number of unique individuals represented = 64
length((myHouseID[,2]))
# number of individuals in dataset = 100
length(myHouseID[,2])
# number of individuals represented = 100
hist(myHouseID[1:10,2],col="blue")
# plotting the first 10 house ID's
# Frequencies greater than 1 indicate individuals

```

```

# with the same house ID.
#-----

#---- Running Function-----
#B = linkedLives(myHouseID[,2:14],c("non-mover", "mover within gb"),mydata[,2:14])
# the first column of the myHouseID and mydata
# datasets contain personal ID's
#-----

#---- Plotting-----
#hist(B[[1]],col="red",xlab="Simulated test statistics"
#,ylab="Frequency",main="Results of non-standardization test",
#cex.lab=1.5,xlim=c(range(B[[1]][1],B[[4]]))
#abline(v=B[[4]],lwd=3,col="orange")
#B[[2]]# pval (Left tailed)
#B[[3]]# pval (Right tailed)
#B[[4]]# Empirical test statistic
#-----

```

maritalData

Marital status sequences

Description

Each sequence contains the marital status for a given individual across 12 years. The sequence dataset is derived from data from the British Household Panel Survey.

Usage

```
data("maritalData")
```

Format

A data frame with 4728 observations across 12 BHPS census waves.

Details

Each sequence represents a unique individual. The sequence alphabet is of length seven and the possible states are: "divorced", "have a dissolved civil partnership", "in a civil partnership", "married", "never married", "separated", and "widowed".

Examples

```

library(TraMineR)
data(maritalData)

#-----
# Converting the data into a sequence object
#-----

```

```

#balphabet = c( "divorced" , "have a dissolved civil partnership" ,
#"in a civil partnership", "married" ,
#"never married" ,"separated" ,"widowed" )

#blabels = c( "divorced" , "have a dissolved civil partnership" ,
#"in a civil partnership", "married" ,
#"never married" ,"separated" ,"widowed" )

#bcodes = c( "divorced" , "have a dissolved civil partnership",
#"in a civil partnership", "married" ,
#"never married" ,"separated" ,"widowed" )

#bseq = seqdef(maritalData, alphabet=balphabet,
#states = bcodes, labels = blabels)
# forming a sequence object

#-----
# Calculating the mobility index
# per sequence and storing
# the indices in an array
#-----
#lseq = length(bseq[,1])
#wseq = length(bseq[1,])
#sequence_summary = array(0,c(lseq,1))
#for(i in 1:lseq){
#sequence_summary[i] = mobility_index(bseq[i,],balphabet,7)
#}
#plot(hist(sequence_summary),xlim=c(1,7),col="red")

#-----
# Generating subsets of the sequence data
#-----
#seqIplot(bseq, sortv = "from.start",cex.legend=1)
#mseq = bseq[which(bseq[,1]=="married"),] # sequences which start with the married state
#seqIplot(mseq, sortv = "from.start",cex.legend=1)
#sseq = bseq[which(bseq[,1]=="never married"),] # sequences which start with the never married state
#seqIplot(sseq, sortv = "from.start",cex.legend=1)

#-----
# Lifecourse destandardization
#-----
#object = summary(as.factor(sequence_summary))
#summarize = chisq.test(object,p = rep(1/length(object),9))
# assuming the "standard" probabilities is p = rep(1/length(object),9)
# alternative standard probabilities can be used.
#summary(maritalData)

```

Description

Calculates the mobility index for a given sequence of states.

Arguments

sequence	A given sequence of states
alphabet	Vector with sequence alphabet
nalpha	Length of sequence

Details

The mobility index facilitates the calculation of each individual's mobility over the life course. This index is defined as the number of separate state clusters. A state cluster is defined as a bout of identical states ($n \geq 1$). Let A, B, and C denote the sequence alphabet for a given trajectory $T = \text{BABBA AAC}$. The mobility index assigned for T_1 is 5 since there are 5 distinct state clusters: B, A, BB, AAA, and C. For a second trajectory, $T_2 = \text{BBBBBBBBB}$, the mobility index is 1. The mobility index can be applied to different types of lifecourse data. The output from this function gives an indication of the number of state clusters within the lifecourse and hence the 'mobility' between states. The in-built datasets are derived from the British Household Panel Survey data (BHPS). The data are derived from the BHPS indresp data files.

Value

mobility index

Author(s)

Glenna Nightingale

Examples

```
#-----
# Constructing 10000 sequences and calculating
#a test statistic (using the mobility index) from the resulting dataset.
#-----

score = totalScore = 0
P = matrix("",nrow=10000,ncol=13)
myseq = sample( LETTERS[c(1,4,5,7)], 13, prob=c(.25,.25,.25,.25), replace=TRUE )
# Each sequence contains four states.
#Examples of states are distinct geographical locations or marital status categories.
for(i in 1:10000){
myseq1 = sample( myseq )
P[i,1:13] = myseq1
score = (mobility_index(myseq1,LETTERS[c(1,4,5,7)],13))
totalScore = totalScore + score
}
dataset_one_score = totalScore/(10000*13)
# test statistic =
#(sum of mobility index)/(total number of sequences * total number of years)
```



```
dataset_one_score
```

```
mydata
```

```
mydata
```

Description

A hundred mobility sequences representing 100 unique individuals. These data are derived from the British Household Panel Study (BHPS)dataset.

Usage

```
data("mydata")
```

Format

A data frame with 100 observations on the following 14 variables.

id a numeric vector

X1996 a factor with levels mover within gb non-mover

X1997 a factor with levels mover within gb non-mover

X1998 a factor with levels mover within gb non-mover

X1999 a factor with levels mover within gb non-mover

X2000 a factor with levels mover within gb non-mover

X2001 a factor with levels mover within gb non-mover

X2002 a factor with levels mover within gb non-mover

X2003 a factor with levels mover within gb non-mover

X2004 a factor with levels mover within gb non-mover

X2005 a factor with levels mover within gb non-mover

X2006 a factor with levels mover within gb non-mover

X2007 a factor with levels mover within gb non-mover

X2008 a factor with levels mover within gb non-mover

Examples

```
data(mydata)
## maybe str(mydata) ; plot(mydata) ...
```

myHouseID	<i>my house ID's</i>
-----------	----------------------

Description

House ID's for 100 individuals over 13 BHPS census waves. Each individual is labelled with a personal ID in the first column of the dataset.

Usage

```
data("myHouseID")
```

Format

A data frame with 100 observations on the following 14 variables.

id a numeric vector
X1996 a numeric vector
X1997 a numeric vector
X1998 a numeric vector
X1999 a numeric vector
X2000 a numeric vector
X2001 a numeric vector
X2002 a numeric vector
X2003 a numeric vector
X2004 a numeric vector
X2005 a numeric vector
X2006 a numeric vector
X2007 a numeric vector
X2008 a numeric vector

Details

The individuals represented in the dataset "mydata" are also represented in this dataset ("myHouseID").

Examples

```
data(myHouseID)  
## maybe str(myHouseID) ; plot(myHouseID) ...
```

nonStanTest	<i>non standardization test</i>
-------------	---------------------------------

Description

Allows the user to test a dataset of observed sequences to determine the probability that these sequences could have arisen due to chance. Acknowledgements: CPC St Andrews/Edinburgh.

Usage

```
nonStanTest(lifecourseSequences, sequenceUnits)
```

Arguments

lifecourseSequences
A dataset (matrix form) consisting of sequences.

sequenceUnits A vector containing the possible sequence states.

Details

This is an approach to answering the question of whether or not the observed sequences for a given set of lifecourses could have arisen due to chance. This is achieved by comparing the observed sequences to a set of randomly simulated sequences. For each observed sequence A, there is a corresponding simulated sequence, B. B is constructed by assigning at random, an order to the observed states in A. The simulated sequences serve as a null distribution for the test.

Value

A p value (LeftTailed_pval) representing the probability that the observed test statistic (corresponding to the observed data) is greater than those obtained under the null hypothesis. A p value (RightTailed_pval) representing the probability that the observed test statistic (corresponding to the observed data) is less than those obtained under the null hypothesis.

The test statistics obtained under the null hypothesis are also provided.

Author(s)

Glenna Nightingale

Examples

```
data(mydata)
# obtaining the in-built data derived from BHPS data.
bbc = mydata
bbcseq = bbc[,2:14]
# removing the first column which contains the ID for the persons involved.
balphabet =c("non-mover", "mover within gb")
#nll = nonStanTest(bbcseq, balphabet)
# obtaining the null distribution and p values
```

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