Package 'InSilicoVA'

May 5, 2015

Type Package	
Title Java implementation of InSilicoVA	
Version 1.0	
Date 2015-02-06	
Author Zehang Li, Tyler McCormick, Sam Clark	
Maintainer Zehang Li <1izehang@uw.edu>	
Depends R (>= 2.15.0), rJava, coda, ggplot2	
Description Computes individual cause of death and population cause-specific mortality fractions using the InSilicoVA algorithm.	
License GPL-2	
causetext	1 2 2
csmf.diag	3 5 6 1
Index 1	4
InsilicoVA-package What the package does (short line) ~~ package title ~~	_

Description

Computes individual cause of death and population cause-specific mortality fractions using the InSilicoVA algorithm. Provides a simple graphical representation of the result.

2 condprob

Details

Package: InsilicoVA
Type: Package
Version: 1.0

Date: 2015-02-06 License: GPL-2

Author(s)

Zehang Li, Tyler McCormick, Sam Clark Maintainer: Zehang Li lizehang@uw.edu>

References

Tyler H. McCormick, Zehang R. Li, Clara Calvert, Amelia C. Crampin, Kathleen Kahn and Samuel J. Clark Probabilistic cause-of-death assignment using verbal autopsies, *arXiv* preprint arXiv:1411.3042 http://arxiv.org/abs/1411.3042 (2014)

causetext

Translation list of COD codes

Description

This is the translation of COD abbreviation codes into their corresponding full names.

Format

A data frame with the translation of COD codes to their names on 68 CODs (both the version of COD only and COD with group code).

Examples

data(causetext)

condprob

Conditional probability table used by InterVA-4

Description

This is a conditional probability matrix used by InterVA-4.2. There are 60 causes and 245 symptoms. The orders of the rows and columns must not be changed.

Format

A data frame with 245 observations on 60 variables. Each observation is the conditional probability.

csmf.diag 3

Examples

data(condprob)

csmf.diag

Convergence test for fitted InSilico model

Description

Produce convergence test for CSMFs from fitted "insilico" objects.

Usage

Arguments

csmf	It could be either fitted "insilico" object, a list of fitted "insilico" object from different chains of the same length but different starting values, i.e., different seed. Or it could be the matrix of CSMF obtained by insilico, or the list of matrices of CSMF. All CSMF could contain more than one subpopulations, but should be in the same format and order.
conv.csmf	The minimum mean CSMF to be checked. Default to be 0.02, which means any causes with mean CSMF lower than 0.02 will not be tested.
test	Type of test. Currently supporting Gelman and Rubin's test (test = "gelman") for multi-chain test, and Heidelberger and Welch's test (test = "heidel") for single-chain test.
verbose	Logical indicator to return the test detail instead of one logical outcome for Heidelberger and Welch's test. Default to be TRUE.
autoburnin	Logical indicator of whether to omit the first half of the chain as burn in. Default to be FALSE since insilico return only the iterations after burnin by default.
	Arguments to be passed to heidel.diag or gelman.diag

Details

The tests are performed using heidel.diag and gelman.diag functions in coda package. The function takes either one or a list of output from insilico function, or only the iteration by CSMF matrix. Usually in practice, many causes with very tiny CSMF are hard to converge based on standard tests, thus it is suggested to check convergence for only causes with mean CSMF over certain threshold by setting proper conv.csmf.

Note for Gelman and Rubin's test, all chains should have the same length. If the chains are sampled with automatically length determination, they might not be comparable by this test.

Author(s)

Zehang Li, Tyler McCormick, Sam Clark Maintainer: Zehang Li lizehang@uw.edu> 4 csmf.diag

References

Tyler H. McCormick, Zehang R. Li, Clara Calvert, Amelia C. Crampin, Kathleen Kahn and Samuel J. Clark Probabilistic cause-of-death assignment using verbal autopsies, *arXiv* preprint arXiv:1411.3042 http://arxiv.org/abs/1411.3042 (2014)

Gelman, Andrew, and Donald B. Rubin. Inference from iterative simulation using multiple sequences. *Statistical science* (1992): 457-472.

Brooks, Stephen P., and Andrew Gelman. General methods for monitoring convergence of iterative simulations. *Journal of computational and graphical statistics* 7.4 (1998): 434-455.

Heidelberger, Philip, and Peter D. Welch. A spectral method for confidence interval generation and run length control in simulations. *Communications of the ACM* 24.4 (1981): 233-245.

Heidelberger, Philip, and Peter D. Welch. Simulation run length control in the presence of an initial transient. *Operations Research* 31.6 (1983): 1109-1144.

Schruben, Lee W. Detecting initialization bias in simulation output. *Operations Research* 30.3 (1982): 569-590.

See Also

```
insilico, summary.insilico
```

Examples

```
# load sample data together with sub-population list
data(SampleInput_insilico)
## Not run:
# extract INterVA style input data
data <- SampleInput_insilico$data</pre>
# extract sub-population information.
# The groups are "HIV Positive", "HIV Negative" and "HIV status unknown".
subpop <- SampleInput_insilico$subpop</pre>
# run without sub-population
fit1a<- insilico( data, subpop = NULL,</pre>
              length.sim = 400, burnin = 200, thin = 10, seed = 1,
              auto.length = FALSE)
fit1b<- insilico( data, subpop = NULL,</pre>
              length.sim = 400, burnin = 200, thin = 10 , seed = 2,
              auto.length = FALSE)
fit1c<- insilico( data, subpop = NULL,</pre>
              length.sim = 400, burnin = 200, thin = 10, seed = 3,
              auto.length = FALSE)
# single chain check
csmf.diag(fit1a)
# equivalent way of check one chain
csmf.diag(fit1a$csmf)
# multiple chains check
csmf.diag(list(fit1a, fit1b, fit1c), test = "gelman")
# equivalent way of check one chain
csmf.diag(list(fit1a$csmf, fit1b$csmf, fit1c$csmf), test = "gelman")
# with sub-populations
fit2a<- insilico( data, subpop = subpop,</pre>
```

insilico 5

```
length.sim = 400, burnin = 200, thin = 10 , seed = 1,
              auto.length = FALSE)
fit2b<- insilico( data, subpop = subpop,</pre>
              length.sim = 400, burnin = 200, thin = 10 , seed = 2,
              auto.length = FALSE)
fit2c<- insilico( data, subpop = subpop,</pre>
              length.sim = 400, burnin = 200, thin = 10, seed = 3,
              auto.length = FALSE)
# single chain check
csmf.diag(fit2a)
# equivalent way of check one chain
csmf.diag(fit2a$csmf)
# multiple chains check
csmf.diag(list(fit2a, fit2b, fit2c), test = "gelman")
# equivalent way of check one chain
csmf.diag(list(fit2a$csmf, fit2b$csmf, fit2c$csmf), test = "gelman")
## End(Not run)
```

insilico

Implement InSilicoVA methods

Description

This function implements InSilicoVA model. The InSilicoVA model is fitted with MCMC implemented in Java. For more detail, see the paper on http://arxiv.org/abs/1411.3042.

For Windows user, this function will produce a popup window showing the progress. For Mac and Unix user, this function will print progress messages on the console. Special notice for users using default R GUI for mac, the output will not be printed on console while the function is running, and will only be printed out after it is completed. Thus if you use a Mac, we suggest using either RStudio for mac, or running R from terminal.

Usage

Arguments

data

The original data to be used. It is suggested to use similar input as InterVA4, with the first column being death IDs. The only difference in input is InsilicoVA takes three levels: "present", "absent", and "missing (no data)". Similar to InterVA software, "present" symptoms takes value "Y"; "absent" symptoms take take value "NA" or "". For missing symptoms, e.g., questions not asked or answered in the original interview, corrupted data, etc., the input should be coded by "." to distinguish from "absent" category.

6 insilico

isNumeric Indicator if the input is already in numeric form. If the input is coded numeri-

cally such that 1 for "present", 0 for "absent", and -1 for "missing", this indicator

could be set to True to avoid conversion to standard InterVA format.

useProbbase Logical indicator. If TRUE, then use InterVA conditional probability table with-

out re-estimating.

keepProbbase.level

Logical indicator when useProbbase is FALSE. If TRUE, then only estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate the InterVA's conditional probability table interpretation table; if FALSE, estimate table ta

mate the whole conditional probability matrix. Default to TRUE.

cond.prob.touse

Customized conditional probability table to use. Currently only accepting the same configuration as InterVA-4 software. It should be a matrix of 245 rows of symptoms and 60 columns of causes, arranged in the same order as in InterVA-4 specification. The elements in the matrix should be the conditional probability of corresponding symptom given the corresponding cause, represented in alpha-

betic form indicating levels. For example input, see condprob

datacheck Logical indicator for whether to check the data satisfying InterVA rules. Default

set to be TRUE. If warning.write is set to true, the inconsistent input will be logged in file warnings.txt. It's strongly suggested to be set to TRUE.

warning.write Logical indicator for whether to save the changes made to data input by datacheck.

If set to TRUE, the changes will be logged in file warnings.txt in current working

directory.

external.sep Logical indicator for whether to separate out external causes first. Default set to

be TRUE. If set to TRUE, the algorithm will estimate external causes, e.g., traffic accident, accidental fall, suicide, etc., by checking the corresponding indicator only without considering other medical symptoms. It is strongly suggested to

set to be TRUE.

length.sim Number of iterations to run. Default to be 4000.

thin Proportion of thinning for storing parameters. For example, if thin = k, the

output parameters will only be saved every k iterations. Default to be 10

burnin Number of iterations as burn-in period. Parameters sampled in burn-in period

will not be saved.

auto.length Logical indicator of whether to automatically increase chain length if conver-

gence not reached.

conv.csmf Minimum CSMF value to check for convergence if auto.length is set to TRUE.

For example, under the default value 0.02, all causes with mean CSMF at least

0.02 will be checked for convergence.

jump. scale The scale of Metropolis proposal in the Normal model. Default to be 0.1.

levels.prior Vector of prior expectation of conditional probability levels. They do not have

to be scaled. The algorithm internally calibrate the scale to the working scale through levels.strength. If NULL the algorithm will use InterVA table as

prior.

levels.strength

Scaling factor for the strength of prior beliefs in the conditional probability levels. Larger value constrain the posterior estimates to be closer to prior expectation. Defult value 1 scales levels.prior to a suggested scale that works

empirically.

trunc.min Minimum possible value for estimated conditional probability table. Default to

be 0.0001

insilico 7

trunc.max Maximum possible value for estimated conditional probability table. Default to

be 0.9999

subpop A vector of sub-population assignments of the same length of death records. It

could be numerical indicators or character vectors of names.

java_option Option to initialize java JVM. Default to "-Xmx1g", which sets the maximum

heap size to be 1GB. If R produces "java.lang.OutOfMemoryError: Java heap space" error message, consider increasing heap size using this option, or one of the following: (1) decreasing length.sim, (2) increasing thin, or (3) disabling

auto.length.

seed Seed used for initializing sampler. The algorithm will produce the same outcome

with the same seed in each machine.

Details

The chains could be set to run automatically longer. If set auto.length to be TRUE, the chain will assess convergence after finishing the length K chain input by user using Heidelberger and Welch's convergence diagnostic. If convergence is not reached, the chain will run another K iterations and use the first K iterations as burn-in. If the chain is still not converged after 2K iterations, it will proceed to another 2K iterations and again use the first 2K iterations as burn-in. If convergence is still not reached by the end, it will not double the length again to avoid heavy memory use. A warning will be given in that case. The extended chains will be thinned in the same way.

For more detail of model specification, see the paper on http://arxiv.org/abs/1411.3042.

Value

id A vector of death ID. Note the order of the ID is in general different from the

input file. See report for organizing the report.

indiv.prob Matrix of individual mean cause of death distribution. Each row corresponds to

one death with the corresponding ID.

csmf Matrix of CSMF vector at each iterations after burn-in and thinning. Each col-

umn corresponds to one cause.

conditional.probs

If the model is estimated with keepProbbase.level = TRUE, this value gives a matrix of each conditional probability at each level at each iterations. Each column corresponds to one level of probability. If keepProbbase.level = FALSE, this value gives a three-dimensional array. If UseProbbase = TRUE, the value

will be set to NULL. See report for more analysis.

missing.symptoms

Vector of symptoms missing from all input data.

external Logical indicator of whether the model is fitted with external causes separated

calculated.

Author(s)

Zehang Li, Tyler McCormick, Sam Clark

Maintainer: Zehang Li lizehang@uw.edu>

8 plot.insilico

References

Tyler H. McCormick, Zehang R. Li, Clara Calvert, Amelia C. Crampin, Kathleen Kahn and Samuel J. Clark(2014) *Probabilistic cause-of-death assignment using verbal autopsies*, http://arxiv.org/abs/1411.3042

Working paper no. 147, Center for Statistics and the Social Sciences, University of Washington

See Also

```
plot.insilico, summary.insilico
```

Examples

plot.insilico

plot CSMF from a "insilico" object

Description

Produce a bar plot of the CSMFs for a fitted "insilico" object.

Usage

```
## S3 method for class 'insilico'
plot(x, type = c("errorbar", "bar", "compare")[1],
  top = 10, causelist = NULL, which.sub = NULL,
  xlab = "Causes", ylab = "CSMF", title = "Top CSMF Distribution",
  horiz = TRUE, angle = 60, fill = "lightblue",
  err_width = .4, err_size = .6, point_size = 2,
  border = "black", bw = FALSE, ...)
```

plot.insilico 9

Arguments

x fitted "insilico" object

type An indicator of the type of chart to plot. "bar" for bar chart on single population;

"compare" for bar charts on multiple sub-populations.

The number of top causes to plot. If multiple sub-populations are to be plotted,

it will plot the union of the top causes in all sub-populations.

causelist The list of causes to plot. It could be a numeric vector indicating the position

of the causes in the InterVA cause list (see causetext), or a vector of character string of the cause names. The argument supports partial matching of the cause names. e.g., "HIV/AIDS related death" could be abbreviated into "HIV"; "Other and unspecified infect dis" could be abbreviated into "Other and unspecified

infect".

which.sub Specification of which sub-population to plot if there are multiple and type is

set to "bar".

xlab Labels for the causes.

ylab Labels for the CSMF values.

title Title of the plot.

horiz Logical indicator indicating if the bars are plotted horizontally.

angle Angle of rotation for the texts on x axis when horiz is set to FALSE

fill The color to fill the bars when type is set to "bar".

border The color to color the borders of bars when type is set to "bar".

err_width Size of the error bars.

err_size Thickness of the error bar lines.

point_size Size of the points.

bw Logical indicator for setting the theme of the plots to be black and white.

... Not used.

Details

To-do

Author(s)

Zehang Li, Tyler McCormick, Sam Clark Maintainer: Zehang Li lizehang@uw.edu>

References

Tyler H. McCormick, Zehang R. Li, Clara Calvert, Amelia C. Crampin, Kathleen Kahn and Samuel J. Clark Probabilistic cause-of-death assignment using verbal autopsies, *arXiv* preprint *arXiv*:1411.3042 http://arxiv.org/abs/1411.3042 (2014)

See Also

insilico, summary.insilico

10 probbase

Examples

```
# load sample data together with sub-population list
data(SampleInput_insilico)
# extract INterVA style input data
data <- SampleInput_insilico$data</pre>
# extract sub-population information.
# The groups are "HIV Positive", "HIV Negative" and "HIV status unknown".
subpop <- SampleInput_insilico$subpop</pre>
# run without sub-population
fit1<- insilico( data, subpop = NULL,</pre>
              length.sim = 400, burnin = 200, thin = 10 , seed = 1,
              external.sep = TRUE, keepProbbase.level = TRUE)
# default plot
plot(fit1)
# customized line plot
plot(fit1, top = 15, horiz = FALSE, fill = "gold",
   bw = TRUE, title = "Top 15 CSMFs", angle = 70,
       err_width = .2, err_size = .6, point_size = 2)
# customized bar plot
plot(fit1, type = "bar", top = 15, horiz = TRUE,
       bw = TRUE, title = "Top 15 CSMFs", angle = 70,
       err_width = .5, err_size = .6)
# run with sub-populations
fit2<- insilico( data, subpop = subpop,</pre>
              length.sim = 400, burnin = 200, thin = 10, seed = 1,
              external.sep = TRUE, keepProbbase.level = TRUE)
# default plot
plot(fit2, type = "compare", top = 5, title = "Top 5 causes comparison")
# customized single sub-population plot
plot(fit2, which.sub = "Unknown",
title = "Top 10 causes in HIV status unknown population")
# customized plot with specified causes, with abbreviation here.
some_causes <- c("HIV", "Pulmonary",</pre>
"Other and unspecified infect dis")
plot(fit2, type = "compare", horiz = FALSE, causelist = some_causes,
   title = "Infectious diseases in three sub-populations",
   angle = 20)
```

probbase

Conditional probability of InterVA4

Description

This is the table of conditional probabilities of symptoms given CODs, together with the data check rules. The values are from InterVA-4.2.

Format

A data frame with 246 observations on 81 variables.

SampleInput_insilico 11

Examples

```
data(probbase)
```

SampleInput_insilico 100 records of Sample Input

Description

This is a dataset consisting of 10 arbitrary sample input deaths in the acceptable format of InSilicoVA and InterVA4. Any data that needs to be analyzed by this package should be in the same format. The orders of the input fields must not be changed.

Format

100 arbitrary input records.

Examples

```
data(SampleInput_insilico)
```

summary.insilico

Summarizing InSilicoVA Model Fits

Description

This function is the summary method for class insilico.

Usage

```
## S3 method for class 'insilico'
summary(object, CI.csmf = 0.95, CI.cond = 0.95,
    file = NULL, top = 10, ...)
```

Arguments

object	Fitted "insilico" object.
CI.csmf	Confidence interval for CSMF estimates.
CI.cond	Confidence interval for conditional probability estimates
file	Optional .csv file to write to. If it is specified, individual cause of death distribution will be saved to the file.
top	Number of top causes to display on screen.
	Not used.

Details

summary.insilico formats some basic information about the InSilicoVA fitted object on screen and show the several top CSMFs of user's choice. See below for more detail.

12 summary.insilico

Value

id the ID of the deaths.

indiv individual Cause of Death distribution matrix.

csmf CSMF distribution and confidence interval for each cause.

csmf.ordered CSMF distribution and confidence interval for each cause, ordered by mean.

condprob Conditional probability matrix and confidence intervals.

useProbbase Component of "insilico" object.

keepProbbase.level

Component of "insilico" object.

datacheck Component of "insilico" object.
length.sim Component of "insilico" object.
thin Component of "insilico" object.
burnin Component of "insilico" object.

jump.scale Component of "insilico" object. levels.prior Component of "insilico" object.

levels.strength

Component of "insilico" object.

trunc.min Component of "insilico" object.
trunc.max Component of "insilico" object.
subpop_counts Component of "insilico" object.
showTop Component of "insilico" object.

Author(s)

Zehang Li, Tyler McCormick, Sam Clark Maintainer: Zehang Li lizehang@uw.edu>

References

Tyler H. McCormick, Zehang R. Li, Clara Calvert, Amelia C. Crampin, Kathleen Kahn and Samuel J. Clark Probabilistic cause-of-death assignment using verbal autopsies, *arXiv* preprint *arXiv*:1411.3042 http://arxiv.org/abs/1411.3042 (2014)

See Also

```
insilico, plot.insilico
```

Examples

```
# load sample data together with sub-population list
data(SampleInput_insilico)
# extract INterVA style input data
data <- SampleInput_insilico$data
# extract sub-population information.
# The groups are "HIV Positive", "HIV Negative" and "HIV status unknown".
subpop <- SampleInput_insilico$subpop</pre>
# run without subpopulation
```

summary.insilico 13

Index

```
*Topic InSilicoVA
    csmf.diag, 3
    insilico, 5
    InsilicoVA-package, 1
    \verb"plot.insilico", 8"
*Topic datasets
    causetext, 2
    condprob, 2
    probbase, 10
    {\tt SampleInput\_insilico}, 11
causetext, 2, 9
{\tt condprob}, {\tt 2}, {\tt 6}
csmf.diag, 3
gelman.diag, 3
heidel.diag, 3
insilico, 4, 5, 9, 12
InsilicoVA (InsilicoVA-package), 1
InsilicoVA-package, 1
plot.insilico, 8, 8, 12
print.insilico(insilico), 5
print.insilico_summary
         (summary.insilico), 11
probbase, 10
SampleInput_insilico, 11
summary.insilico, 4, 8, 9, 11
```