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Description Consensus analysis, model-based clustering, and cultural consensus theory applications to response data (e.g. questionnaires). The models are applied using hierarchical Bayesian inference. The current package version supports binary, ordinal, and continuous data formats.
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CCTpack-package

CCTpack: Consensus Analysis, Model-Based Clustering, and Cultural Consensus Theory Applications

Description

An R package for advanced model-based analyses of questionnaire data. These models can detect the consensus answers of the respondents, and perform a model-based clustering of the respondents. The methods can detect latent subgroups in the data, their differing consensuses, the expertise of each respondent, their response biases, and the difficulty of each question. The ability to parse the variance of the data, by these factors, results in sophisticated analyses of the consensus answers, the clusters, and the respondent knowledge/expertise. These are advanced models that are fit using hierarchical Bayesian inferential methods. Bayesian sampling routines are currently handled with JAGS (please install beforehand: mcmc-jags.sourceforge.net/). The package can currently handle data from questionnaires of binary responses, ordered categorical, or continuous responses. If not already in this format, your data might be appropriately transformed to one of these scales to be fit by these models. The models are based on mathematical publications in the domain of Cultural Consensus Theory (CCT): the General Condorcet Model (GCM), the Latent Truth Rater Model (LTRM), and the Continuous Response Model (CRM). See the relevant literature listed in the reference manual for more information about these CCT models and methods. Respectively, these

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models are applicable to dichotomous/binary (0,1), ordinal (1, 2, ...), and continuous data. In addition, there is functionality for component analyses/scree plots of the data, automatic plots of the model results, diagnostics for quality of fit, and exporting these plots/results.

Details

Package:	CCTpack
Type:	Package
Version:	1.5.2
Date:	2017-10-10
License:	GPL (>= 2)

1. Make sure you have JAGS installed

2. To install CCTpack use command: install.packages("CCTpack",dependencies=TRUE)

3. To load CCTpack use command: library(CCTpack)

4. Type ?cctapply to get a walkthrough with example data 5. For more information on the models and methods used, see the References section.

Author(s)

Royce Anders

Maintainer: Royce Anders <andersr@uci.edu>

References

Anders, R., Oravecz, Z., & Batchelder, W. H. (2014). Cultural consensus theory for continuous responses: A latent appraisal model for information pooling. Journal of Mathematical Psychology, 61, 1-13.

Anders, R., & Batchelder, W. H. (2015). Cultural consensus theory for the ordinal data case. Psychometrika, 1-31.

Anders, R., & Batchelder, W. H. (2012). Cultural consensus theory for multiple consensus truths. Journal for Mathematical Psychology, 56, 452-469.

Batchelder, W. H., & Anders, R. (2012). Cultural consensus theory: comparing different concepts of cultural truth. Journal of Mathematical Psychology, 56, 316-332.

Oravecz, Z., Anders, R. & Batchelder, W. H. (2015). Test theory without an answer key in a Bayesian hierarchical modeling framework. Psychometrika, 1-24.

See Also

To install JAGS, see: mcmc-jags.sourceforge.net/

cct-class

Description

Virtual class that contains "rjags" class, used to create plot, summary, and screeplot methods for the cctfit objects from package CCTpack.

Objects from the Class

Objects can be created by calls of the form new("cct", ...).

Slots

model: Object of class "jags" ~~
BUGSoutput: Object of class "bugs" ~~

Extends

Class "rjags", directly.

Methods

plot signature(x = "cct", y = "rjags"): ...

Author(s)

Royce Anders Maintainer: Royce Anders <andersr@uci.edu>

Examples

```
showClass("cct")
```

cctapply	Primary function to perform model-based consensus analysis: loads the data and fits the consensus model. Allows for model-based clus-
	tering based on the numbers of "clusters" specified. Options are also available to run diagnostics on the fit, and to export the results to saved files. Based on cultural consensus theory (CCT) models for data ma- trices of binary, ordered categorical, or continuous response data.

Description

loads the data, fits the appropriate model, runs the posterior predictive checks, and optionally exports results

cctapply

Usage

```
cctapply(data, clusters = 1, itemdiff = FALSE, biases = TRUE, samples = 10000, chains = 3,
    burnin = 2000, thinning = 1, runchecks = FALSE, exportfilename = "",
    polych = FALSE, parallel = FALSE, seed = NULL,plotr=FALSE)
```

Arguments

data	a 2-dimensional matrix or array, missing values should be input as NA.
clusters	The number of possible clusters (cultures) to use
itemdiff	Whether item difficulty should be measured and accounted for
biases	Whether response biases should be measured and accounted for
samples	The number of samples for the inference
chains	The number of chains for the inference
burnin	The number of burn-in for the inference
thinning	The amount of thinning in the inference
runchecks	If the posterior predictive checks should be calculated after the inference
exportfilename	If you'd like to export: specify a filename and optionally its location. Ex: exportfilename = "C:/CCTpack/CCTpackdata.Rdata"
polych	used for ordinal data only, if the polychoric correlations, rather than Pearson correlations, should be used (for the posterior predictive checks) – these take a long time to calculate but are more precise in the ordinal data case.
parallel	Whether the inference should be computed in parallel (1 chain per logical processor)
seed	Set the random number seed here (to reproduce results as before). If not speci- fied, the seed is randomly generated.
plotr	Whether to plot the posterior mean results for each parameter. Note: runchecks = TRUE will plot the posterior predictive checks after the posterior mean results. The posterior mean results plot can be later called via cctresults().

Details

This is the main function to fit the consensus models. The function fits the model using hierarchical Bayesian inference. The Bayesian sampling is performed using JAGS.

Value

cctfit is returned, which has the structure of a 'jagsfit' object as in Rjags, but has additional data included.

Examples

Load Data (here binary responses, 1/0 for yes/no)
data(hotcold)

Calculate scree plot to decide how many clusters to run, looks like 2 clusters here

```
dat <- cctscree(hotcold)</pre>
# Retrieve factors from the scree plot
cctfac(dat)
               # dat$factors
# Fit the Model
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,</pre>
                      chains = 3, burnin = 2000, seed = 1, runchecks = FALSE)
# Calculate Fit Diagnostics (Posterior Predictive Checks)
# cctfit <- cctppc(cctfit)</pre>
# Plot Parameter Results
# cctresults(cctfit)
# Tables of Subject and Item Parameter Values and Credible Intervals
# cctsubj(cctfit)
                      # cctfit$subj
# cctsubjhdi(cctfit) # cctfit$subjhdi
# cctitem(cctfit)
                       # cctfit$item
# cctitemhdi(cctfit) # cctfit$itemhdi
# Show Missing Value Model Estimates if there was missing data
# cctmvest(cctfit)
# Export Results (saves data and plots)
# cctexport(cctfit,filename="CCTpackdata.Rdata")
# Load and Fit Example Data for ordered categorical or continuous responses
# data(raterdata)
# cctfit <- cctapply(data = raterdata, clusters = 1, itemdiff = TRUE, samples = 10000,</pre>
#
                      chains = 3, burnin = 2000, seed = 1, runchecks = FALSE)
# data(continuousdata)
# cctfit <- cctapply(data = continuousdata, clusters = 1, itemdiff = TRUE, samples = 10000,</pre>
                      chains = 3, burnin = 2000, seed = 1, runchecks = FALSE)
#
##Note: if an insufficient memory message occurs, you can increase the
##memory allocation by the command 'memory.limit(25000)' (or as high as needed)
```

cctcat

Accessor function for the model-estimated category boundaries (obtained from the model applied to the data). Applicable for the LTRM model only.

Description

Outputs a table read out of the category boundary parameters of the model inference, as well as their credible intervals (posterior highest density intervals, HDIs), for each cluster.

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cctexport

Usage

cctcat(cctfit)

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

Examples

```
data(raterdata)
# cctfit <- cctapply(data = raterdata, clusters = 1, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctcat(cctfit)</pre>
```

cctexport

Export the results (the cctfit object)

Description

Exports the cctfit object as an .Rdata file, as well as .jpeg and .eps files of the relevant plots

Usage

cctexport(cctfit, filename = "CCTpackdata.Rdata")

Arguments

cctfit	The cctfit object as obtained from the cctapply() function.
filename	The filename and location you would like to use. If no location is specified, it is saved to the current R working directory (see getwd()).

Details

Saves the cctfit object as an .Rdata file, as well as .jpeg and .eps files of the relevant plots, which include: the scree plot, results plot, and posterior predictive check plots.

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = TRUE)
# cctexport(cctfit, filename = "C:/CCTpack/CCTpackdata.Rdata")</pre>
```

cctfac

Accessor function for the factor coefficients after dat <- cctscree(dat) has been run.

Description

Outputs a vector of the factor coefficients obtained from a principal components analysis of the respondent by respondent correlation matrix of the data.

Usage

cctfac(dat)

Arguments

dat

The object as obtained from the dat <- cctscree(dat) function.

Examples

```
data(hotcold)
dat <- cctscree(hotcold)
cctfac(dat)</pre>
```

cctgui

The CCT GUI

Description

There is a graphical user interfance available, though it is deprecated / being phased out – it is not compatible with all operating systems.

Instructions how to use the GUI: 1. Type 'cctgui()' in the R command prompt to start the CCT GUI 2. Click "Load data" (data must be in csv format as a .csv or .txt file)

data format: respondents (rows) by items (columns), csv format, no row/column names necessary the models can handle missing data, missing data should be input as NA

- 3. Verify that the number of respondents, items, and data type the GUI detects is correct.
- 4. Check the number of significant factors in the scree plot
- 5. Use this as "the number of cultures to assume," and select an item difficulty option
- 6. Click "Apply CCT Model"
- 7. Check inference results (if the number of Rhats is appropriate);
- if not, one could try running more samples, or a different number of cultures
- 8. Click "Run Checks" to run the posterior predictive checks, and verify if they are satisfied.

Note: the CCT method suggests that the fewest number of cultures to pass the checks should be the model used.

9. Click "Plot Results" to see the posterior results

10. Click "Export Results" to save the model fit and plots.

cctgui

Usage

cctgui()

Value

All information will be saved in the 'cctfit' object. This is in the same structure as the 'jagsfit' object returned by the function jags(), which utilizes the 'rjags' and 'R2jags' packages.

Author(s)

Royce Anders

Examples

#First visit 'mcmc-jags.sourceforge.net/' to install JAGS (Plummer, 2003) if it's not installed

#1) Convert your data into a .csv file
Or use our example data such as with the following commands:

```
data(hotcold);
write.csv(x=hotcold, file="hotcold.csv",row.names=FALSE)
```

#2) Invoke the GUI

cctgui()

```
#Click "Load Data" then find and select "hotcold.csv"
#Note that 23 respondents, 27 items are detected,
#
      and that it is Dichotomous (binary) data, which the GCM is applicable for
      the GUI detects that there are 14 missing data values in the matrix
#
#Click "Scree Plot" Note that there are 2 apparent significant factors,
      thus we assume 2 cultures in the data, click "yes" to estimate item difficulty
#
#Click "Apply CCT Model" and wait for the inference to finish
#Take note of the Number of Rhats above 1.1 (if too many, perhaps run more samples)
#Click "Run Checks" to run the posterior predictive checks, wait for the checks to complete
#Note that the model satisfies both checks in the plots shown
#Proceed to inspect the inference results by clicking "Plot Results"
#type 'cctfit$MVest' to view the model estimates of the 14 missing data values
#Click "Export Results" to save the plot and the inference results
#The Inference results are included in the object 'cctfit'
#Type 'cctfit' in the R prompt to see the summary,
   type 'str(cctfit)' to see what the object contains
#
```


#The corollary to the GUI instructions above, for the command line, is below

```
#for an ordinal data example, use data(raterdata)
data(hotcold);
#Loads data and provides the Scree Plot
# cctscree(data = hotcold)
#Loads data and Runs the Inference
# cctfit <- cctapply(data = hotcold,clusters=2,itemdiff=TRUE,samples=10000,</pre>
#
                      chains=3,burnin=2000,runchecks=FALSE)
#Calculates and Plots Posterior Predictive Checks
# cctfit <- cctppc(cctfit)</pre>
#Show Missing Value Model Estimates if there was missing data
# cctfit$MVest
#Plots Posterior Results
# cctresults(cctfit)
#Exports Results
# cctexport(cctfit,filename="CCTpackdata.Rdata")
##Note: if an insufficient memory message occurs, you can increase the
##memory allocation by the command 'memory.limit(25000)' (or as high as needed)
```

cctitem	Accessor function for the item parameters (obtained from the model
	applied to the data)

Description

Outputs a table read out of the item parameters of the model inference (for example: item #, estimated consensus answer for each cluster, and item difficulty for each cluster)

Usage

```
cctitem(cctfit)
```

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctitem(cctfit)</pre>
```

cctitemhdi

Accessor function for the item parameter credible intervals (obtained from the model applied to the data)

Description

Outputs a table read out of the credible intervals (posterior highest density intervals, HDIs) of the item parameters of the model inference (for example, the lower and upper bounds of estimated consensus answers and item difficulties for each cluster)

Usage

cctitemhdi(cctfit)

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

Examples

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctitemhdi(cctfit)</pre>
```

Accessor function for the cluster memberships of the respondents.

Description

Outputs the cluster (cultural) assignment of each respondent, as determined by the CCT model fit to the data .

Usage

```
cctmemb(cctfit)
```

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctmemb(cctfit)</pre>
```

cctmvest

Description

Outputs an N by 3 matrix, where N is the number of missing values estimated by the model, column 1 is the person index, column 2 the item index, column 3 is the value estimate.

Usage

cctmvest(cctfit)

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

Examples

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctmvest(cctfit)</pre>
```

cctppc

Calculate or Plot the Posterior Predictive Model Checks

Description

Plots (and calculates if not calculated already), the posterior predictive model checks for the cctfit object

Usage

```
cctppc(cctfit, polych = FALSE, doplot = TRUE)
```

Arguments

cctfit	The cctfit object as obtained from the cctapply() function.
polych	used for ordinal data only, if the polychoric correlations, rather than Pearson correlations, should be used (for the posterior predictive checks) – these take a long time to calculate but are more precise in the ordinal data case.
doplot	If the diagnostics should be plotted.

cctresults

Details

Generates 500 posterior predictive data sets that are randomly sampled from the posterior predictive data; it uses these to calculate 2 posterior predictive checks that respectively pertain to fitting the consensus structure of the data (the number of latent cultures), and if heterogeneous item difficulty should be used.

Value

returns the cctfit object with the posterior predictive data and checks saved.

Examples

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctfit <- cctppc(cctfit)</pre>
```

cctresults

```
Plot the posterior results from the model inference
```

Description

Plot the posterior results from the model inference, a specialized display is produced depending on the model that is applied.

Usage

```
cctresults(cctfit)
```

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctresults(cctfit)</pre>
```

cctscree

Description

Creates a screeplot of the data, providing the first 8 factors and their magnitudes.

Usage

```
cctscree(data, polych = FALSE)
```

Arguments

data	a 2-dimensional matrix or array, missing values should be input as NA.
polych	used for ordinal data only, if the polychoric correlations, rather than Pearson correlations, should be used

Details

The scree plot is generated from the respondent by respondent correlation (Pearson) matrix.

Examples

```
data(hotcold)
cctscree(data = hotcold)
## With this example data, a scree plot with 2 substantial factors
## (suggesting two cultures) is produced.
```

cctsubj	Accessor function for the subject parameters (obtained from the model
	applied to the data)

Description

Outputs a table read out of the subject parameters of the model inference (for example: participant #, cluster membership, competency, response biases)

Usage

```
cctsubj(cctfit)
```

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

cctsubjhdi

Examples

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctsubj(cctfit)</pre>
```

cctsubjhdi	Accessor function for the subject parameter credible intervals (ob-
	tained from the model applied to the data)

Description

Outputs a table read out of the credible intervals (posterior highest density intervals, HDIs) of the subject parameters of the model inference (for example, the lower and upper bounds of estimated participant knowledge competencies and response biases)

Usage

cctsubjhdi(cctfit)

Arguments

cctfit The cctfit object as obtained from the cctapply() function.

Examples

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# cctsubjhdi(cctfit)</pre>
```

continuousdata.csv continuousdata

Description

Example 2 culture continuous data with item difficulty for cctgui() This data should load as 40 respondents by 40 items, and as continuous data It is an example of 2 culture data, with 16 missing data values

Note

csv or text data files need not use header or row names Though respondents should be by the rows, and items by the columns dtraceplot

Description

Produces all of the traceplots the discrete parameters of a cctfit object, in a 3x3 design and multiple plot windows, via traceplot from **R2jags**.

Usage

dtraceplot(cctfit,ask = FALSE)

Arguments

cctfit	The cctfit object as obtained from the cctapply() function.
ask	logical; if TRUE, the user is asked before each plot, to proceed to the next. See par(ask=.)

Examples

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# dtraceplot(cctfit)</pre>
```

hotcold.csv hotcold

Description

Example 2 culture binary data with item difficulty for cctgui() This data should load as 23 respondents by 27 items, and as binary data It is an example of 2 culture data, with 14 missing data values

Note

csv or text data files need not use header or row names Though respondents should be by the rows, and items by the columns plot

Description

Plots the posterior results from the model inference, a specialized display is produced depending on the model that is applied.

Usage

plot(x,y,...)

Arguments

x	is a cctfit object as obtained from the cctapply() function, which has class 'cct'.
У	NULL
	Additional arguments

Examples

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# plot(cctfit)</pre>
```

plot-methods ~~ *Methods for Function* plot *in Package* graphics ~~

Description

~~ Methods for function plot in package graphics ~~

Methods:

signature(x = "ANY") Default plot method.

signature(x = "cct") CCTpack plot method for cctfit objects from cctapply.

raterdata.csv raterdata

Description

Example 1 culture ordinal data with item difficulty for cctgui() This data should load as 25 respondents by 40 items, and as ordinal data It is an example of 1 culture data, with 9 missing data values

Note

csv or text data files need not use header or row names Though respondents should be by the rows, and items by the columns

screeplot

Generate a scree plot

Description

Screeplot method for CCTpack data, or a cctfit object of class 'cct'; equivalent to function cctscree().

Usage

```
screeplot(x, ...)
```

Arguments

х	a 2-dimensional matrix or dataframe, missing values should be input as NA, or a cctfit object of class 'cct'.
	Additional arguments: polych=T (for ordinal data only), to use the polychoric correlations rather than Pearson correlations.

Details

The scree plot is generated from the respondent by respondent correlation (Pearson) matrix.

```
data(hotcold)
screeplot(hotcold)
## With this example data, a scree plot with 2 substantial factors
## (suggesting two cultures) is produced.
```

screeplot-methods ~~ *Methods for Function* screeplot *in Package* stats ~~

Description

~~ Methods for function screeplot in package stats ~~

Methods:

```
signature(x = "ANY") A matrix or data.frame.
signature(x = "cct") An cctfit object from cctapply.
signature(x = "data.frame")
signature(x = "matrix")
```

summary

Summary of the cctfit object of class 'cct'.

Description

Summary method for a cctfit object of class 'cct'. Information about the data, the fit, posterior predictive checks.

Usage

summary(object, ...)

Arguments

object	The cctfit object as obtained from the cctapply() function, which has class 'cct'.
	Additional arguments

```
data(hotcold)
# cctfit <- cctapply(data = hotcold, clusters = 2, itemdiff = TRUE, samples = 10000,
# chains = 3, burnin = 2000, runchecks = FALSE)
# summary(cctfit)</pre>
```

summary-methods

Description

~~ Methods for function summary in package base ~~

Methods:

```
signature(object = "ANY") Default method.
signature(object = "cct") CCTpack plot method for cctfit objects from cctapply.
```

testdat.csv testdat

Description

Example test data for cctgui() This data should load as 20 respondents by 25 items, and as binary data It is an example of 1 culture data

Note

csv or text data files need not use header or row names Though respondents should be by the rows, and items by the columns

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