Package 'MIRT4FC'

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Type Package
Title Fit FC-Model by iStEM Algorithm
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Authors Peiyi Xu [aut, cre],
Chanjin Zheng [aut]

Maintainer Peiyi Xu <xupy.ecnu@foxmail.com>

Description The objective of MIRT4FC is to efficiently implement various forced-choice models using the *i*stem algorithm.Currently, it includes Thurstone's Item Response Theory (TIRT, Brown et al., 2011) Model, Multi-Unidimensional Pairwise Preference Two Parameter Logistic Model (MUPP-2PLM, Morillo et al., 2016), Multi-Unidimensional Pairwise Preference Generalized Graded Unfolding Model (MUPP-GGUM, Stark et al., 2005) and Generalized Graded Unfolding-RANK Model (GGUM-RANK, Lee et al., 2018), and we plan to continue updating and adding new models in the future. In addition to item parameter estimation capabilities, our R package also offers the ability to estimate ability parameters using MAP, EAP, and MLE methods. It can generate simulated response matrices, calculate standard errors (SE) for both ability and item parameters, and include a set of empirical data.

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Depends parallel

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StEM

Improved Stochastic EM algorithm for solving FC: MUPP-2PL model

Description

Improved Stochastic EM algorithm for solving Force Choice: Multi-Unidimensional Pairwise-Preference Two-Parameter Logistic (FC: MUPP-2PL) model.

Usage

```
StEM (Y, BID, positive = rep (TRUE, nrow (BID)), blocksize = 3, res = "rank", M = 10, B = 20, a = NULL, d = NULL, item.par = NULL, sigma = NULL, theta = NULL, fix.sigma = FALSE, burnin. maxitr = 40, maxitr = 500, eps1 = 1.5, eps2 = 0.4, frac1 = 0.2, frac2 = 0.5, cores = 1)
```

Arguments

Y A # of subjects * # of blocks matrix; item responses.

BID A # of statements * 3 matrix; item information, columns are "Block",

"Item" and "Dimensions".

positive A logical vector; indicating whether each statement is positive directional

or not.

blocksize A number; block size of FC (2/3/4).

res A string; response format('pick'/'rank'/'mole'), pick-2/rank-2/mole-2 are

equivalent, rank-3/mole-3 are equivalent.

M A number; # of batch.

B A number; # of iterations in each batch.

a A vector; length = # of statements, initial alpha parameters.

d A vector; length = # of statements, initial beta parameters.

item.par A data frame; initial parameters for a and d.

sigma A # of dimensions * # of dimensions matrix; initial sigma parameters.

theta A # of subjects * # of dimensions matrix; initial theta parameters.

fix.sigma Logical; TRUE if sigma is estimated.

burnin.maxitr A number; max burn-in allowed.

maxitr A number; max iterations allowed.

eps1 A number; stability criteria.

eps2 A number; convergence criterion.

frac1 A number; cutoffs for calculating Geweke z.

frac2 A number; cutoffs for calculating Geweke z.

cores A number; number of parallel cores.

Value

The function returns a list with the following components:

a A vector; length = # of statements, alpha parameters
 d A vector; length = # of statements, beta parameters.

sigm A # of dimensions x # of dimensions matrix; sigma parameters.

total batch number A number; total batch number.

final chain size A number; final chain size.

burn-in size A number; burn-in size.

time A number; time.

Examples

```
library (MIRT4FC)
D <- 6
                                              # Dimension
nitem.per.dim <- 10
                                              # Iems number per dimension
nblock <- D * nitem.per.dim / 3
                                              # Blocks number
set.seed(123456)
                                              # Set random seed
# Simulate block-item-demension correspondence table
BID <- data.frame (Block = rep (1:nblock,each=3),
                       Item=rep (1:3, nblock),
                       Dim=c(combn(D,3) [, sample(choose(D,3), nblock,replace = TRUE)]))
# Simulate item parameter truth value
item.par <- data.frame (a = seq len (D * nitem.per.dim))
item.par <- within (item.par, {
  a <- runif (D*nitem.per.dim,0.7,3)
  b <- rnorm (D*nitem.per.dim)
  d <- a*b
})
item.par$d <- c (t (aggregate (item.par$d, by=list (BID$Block), function(x)x-mean(x)) [, -1]))
N <- 1000
                                              # Sample number
v \le matrix (0.5, D, D)
                                              # Intertrait correlation
diag(v) < -1
# Simulate latent trait parameter truth value
theta \leftarrow mvnfast::rmvn (N, seq(-1, 1, length.out = D),sigma = v)
####### Generate a simulated dataset
Y <- data.sim (item.par, theta, BID, blocksize = 3, res = 'rank')
```

####### Item parameter estimation

```
fit <- StEM (Y, BID, maxitr = 100, blocksize = 3, res = 'rank', fix.sigma = TRUE, cores = 1)
```

data.sim

Simulated dataset for multivariate FC item response theory model.

Description

Simulated dataset for multivariate FC item response theory model.

Usage

```
data.sim (item.par, theta, BID, blocksize = 3, res = 'rank')
```

Arguments

BID A # of statements * 3 matrix; item information, columns are "Block", "Item"

and "Dimensions".

blocksize A number; block size of FC (2/3/4).

res A string; response format('pick'/'rank'/'mole'), pick-2 (blocksize = 2)/rank-

2/mole-2 are equivalent, rank-3/mole-3 are equivalent.

item.par A data frame; parameters for a and d.

theta A # of subjects * # of dimensions matrix; theta parameters.

Value

A # of subjects x # of block number matrix.

Examples

```
library (MIRT4FC)
```

```
D <- 6 # Dimension
```

nitem.per.dim <- 10 # Iems number per dimension

nblock <- D * nitem.per.dim / 3 # Blocks number set.seed(123456) # Set random seed

Simulate block-item-demension correspondence table

BID <- data.frame (Block = rep (1:nblock,each=3),

Item=rep (1:3, nblock),

Dim=c(combn(D,3) [, sample(choose(D,3), nblock,replace = TRUE)]))

Simulate item parameter truth value

```
item.par <- data.frame (a = seq_len (D * nitem.per.dim))
```

item.par <- within (item.par, {

a <- runif (D*nitem.per.dim,0.7,3)

b <- rnorm (D*nitem.per.dim)

```
d <- a*b \\ \}) item.par$d <- c (t (aggregate (item.par$d, by=list (BID$Block), function(x)x-mean(x)) [, -1]))  
N <- 1000 \qquad \qquad \# Sample number \\ v <- matrix (0.5, D, D) \qquad \qquad \# Intertrait correlation \\ diag (v) <- 1 \\ \# Simulate latent trait parameter truth value \\ theta <- mvnfast::rmvn (N, seq(-1, 1, length.out = D), sigma = v) \\ \# \# \# \# \# \# \# Generate a simulated dataset \\ Y <- data.sim (item.par, theta, BID, blocksize = 3, res = 'rank')
```

 MAP_data

A Triple Empirical Data for Dominance Modell.

Description

Simulated dataset for multivariate FC item response theory model. This data set contains a real data set of 1391 participants on 88 triplets. In each triplet, participants had to rank the three alternative items according to their preference.

Usage

MAP data

Format

A large matrix of 1391 observations containing information on 6 variables. In each block, 3 items (e.g., A/B/C) measure 3 different traits.

```
A>B>C=1; A>C>B=2; B>A>C=3; B>C>A=4; C>A>B=5; C>B>A=6
```

Examples

library (MIRT4FC)

Y <- data("MAP data")

####### Item parameter estimation

fit <- StEM (Y, BID, maxitr = 150, blocksize = 3, res = 'rank', fix.sigma = TRUE, cores = 1)