Package: EstCRM (via r-universe)

Version 1.6

Date 2022-9-16

Title Calibrating Parameters for the Samejima's Continuous IRT Model

Author Cengiz Zopluoglu

Maintainer Cengiz Zopluoglu <cen.zop@gmail.com>

Depends Hmisc, lattice

Description Estimates item and person parameters for the Continuous Response Model (CRM; Samejima, 1973, <doi:10.1007/BF02291114>), computes item fit residual statistics, draws empirical 3D item category response curves, draws theoretical 3D item category response curves, and generates data under the CRM for simulation studies.

License GPL (>= 2)

URL https://cengiz.me/

NeedsCompilation no

Date/Publication 2022-09-16 21:46:10 UTC

Repository https://czopluoglu.r-universe.dev

RemoteUrl https://github.com/cran/EstCRM

RemoteRef HEAD

RemoteSha c46090d13106c3840fdd076ae66274caa04bd8da

Contents

bootCRM	*Computing Standard Errors for Item Parameter Estimates using Bootstrap Sampling*

Description

Computes the standard errors of item parameter estimates using a non-parametric bootstrapping approach

Usage

```
bootCRM(data,max.item,min.item,max.EMCycle=500,converge=.01,
        type="Shojima",BFGS=TRUE,nsample=50)
```

Arguments

data	a data frame with N rows and m columns, with N denoting the number of subjects and m denoting the number of items.
max.item	a vector of length m indicating the maximum possible score for each item.
min.item	a vector of length m indicating the minimum possible score for each item.
max.EMCycle	a number of maximum EM Cycles used in the iteration. Default is 500.
converge	a criteria value indicating the difference between loglikelihoods of two consecutive EM cycles to stop the iteration. Default is .01
type	type of optimization. Default is the non-iterative EM developed by Shojima(2005).
BFGS	a valid argument when type is equal to "Wang&Zeng". If TRUE, the Broyden-Fletcher-Goldfarb-Shanno (BFGS) algorithm is used to update Hessian.
nsample	a number of bootstrap samples used in estimating the standard errors. Default is 50

Details

bootCRM computes the standard errors of item parameter estimates using a non-parametric bootstrap sampling approach.

Value

bootCRM returns a list with three elements. Each element of the list is an mx3 matrix where m is the number of items. The first column includes the item parameter estimates from the original sample, the second column includes the average item parameter estimates from R bootstrap samples (the mean of the empirical sampling distribution), and the third column includes the standard errors of the item parameter estimates obtained from bootstrap samples (standard deviation of the empirical sampling distribution).

Discrimination	Estimates for item discriminations
Difficulty	Estimates for item difficulties
Alpha	Estimates for alpha parameters

Note

It may make take too long to get the results if you request a large number of bootstrap samples. By default, the number of bootstrap samples is 50, and it takes a couple minutes for the sample datasets provided in the package. It would be a good idea to try 100 or 500 bootstrap samples if you have enough time.

Author(s)

Cengiz Zopluoglu

See Also

EstCRMitem for estimating item parameters, EstCRMperson for estimating person parameters, fitCRM for computing item-fit statistics and drawing empirical 3D item response curves, plotCRM for drawing theoretical 3D item category response curves,

Examples

```
## Not run:

##load the dataset EPIA
data(EPIA)

bootCRM(data=EPIA,
max.item=c(112,112,112,112,112),
min.item=c(0,0,0,0,0),
max.EMCycle=200,
converge=.01,
nsample=100)


##load the dataset SelfEff
data(SelfEff)

bootCRM(data=SelfEff,
max.item=c(11,11,11,11,11,11,11,11,11,11),
min.item=c(0,0,0,0,0,0,0,0,0,0),
max.EMCycle=200,
converge=.01,
nsample=100)

## End(Not run)
```

EPIA   *The Eysenck Personality Inventory Impulsivity Subscale*

Description

The data came from a published study and was kindly provided by Dr. Ferrando. A group of 1,033 undergraduate students were asked to check on a 112 mm line segment with two end points (almost never, almost always) using their own judgement for the five items taken from the Spanish version of the EPI-A impulsivity subscale. The direct item score was the distance in mm of the check mark from the left end point (Ferrando, 2002).

Usage

```
data(EPIA)
```

Format

A data frame with 1033 observations on the following 5 variables.

Item 1 Longs for excitement

Item 2 Does not stop and think things over before doing anything

Item 3 Often shouts back when shouted at

Item 4 Likes doing things in which he/she has to act quickly

Item 5 Tends to do many things at the same time

Source

Ferrando, P.J.(2002). Theoretical and Empirical Comparison between Two Models for Continuous Item Responses. *Multivariate Behavioral Research*, 37(4), 521-542.

References

Ferrando, P.J.(2002). Theoretical and Empirical Comparison between Two Models for Continuous Item Responses. *Multivariate Behavioral Research*, 37(4), 521-542.

EstCRM   *Calibrating Parameters for the Samejima's Continuous Response Model*

Description

This package includes the tools to estimate model parameters for the Samejima's Continuous Response Model (CRM) via marginal maximum likelihood estimation and EM algorithm, to compute item fit residual statistics, to draw empirical 3D item category response curves, to draw theoretical 3D item category response curves, and to generate data under the CRM for simulation studies.

Details

Package:	EstCRM
Type:	Package
Version:	1.5
Date:	2022-06-13
License:	GPL-2
LazyLoad:	yes

The package can be used to estimate item parameters by using `EstCRMitem`, to estimate person ability parameters by using `EstCRMperson`, to compute the item fit residual statistics after parameter calibration by using `fitCRM`, to draw empirical 3D item category response curves by using `fitCRM`, to draw theoretical 3D item category response curves by using `plotCRM`, and to generate data under the Continuous Response Model for the simulation studies by using `simCRM`.

Author(s)

Cengiz Zopluoglu

Maintainer: Cengiz Zopluoglu <cen.zop@gmail.com>

References

Kan, A.(2009). Effect of scale response format on psychometric properties in teaching self-efficacy. *Euroasian Journal of Educational Research*, *34*, 215-228.

Ferrando, P.J.(2002). Theoretical and Empirical Comparison between Two Models for Continuous Item Responses. *Multivariate Behavioral Research*, 37(4), 521-542.

Samejima, F.(1973). Homogeneous Case of the Continuous Response Model. *Psychometrika*, 38(2), 203-219.

Shojima, K.(2005). A Noniterative Item Parameter Solution in Each EM Cycyle of the Continuous Response Model. *Educational Technology Research*, 28, 11-22.

Wang, T. & Zeng, L.(1998). Item Parameter Estimation for a Continuous Response Model Using an EM Algorithm. *Applied Psychological Measurement*, 22(4), 333-343.

See Also

`EstCRMitem` for estimating item parameters, `EstCRMperson` for estimating person parameters, `fitCRM` for computing item-fit residual statistics and drawing empirical 3D item category response curves, `plotCRM` for drawing theoretical 3D item category response curves, `simCRM` for simulating data under CRM.

EstCRMitem   *Estimating Item Parameters for the Continuous Response Model*

Description

Estimate item parameters for the Continuous Response Model (Samejima,1973)

Usage

```
EstCRMitem(data, max.item, min.item, max.EMCycle = 500, converge = 0.01,
           type="Shojima",BFGS=TRUE)
```

Arguments

data	a data frame with *N* rows and *m* columns, with *N* denoting the number of subjects and *m* denoting the number of items.
max.item	a vector of length *m* indicating the maximum possible score for each item.
min.item	a vector of length *m* indicating the minimum possible score for each item.
max.EMCycle	a number of maximum EM Cycles used in the iteration. Default 500.
converge	a criteria value indicating the difference between loglikelihoods of two consecutive EM cycles to stop the iteration. Default .01
type	type of optimization. Takes two values, either "Shojima" or "Wang&Zeng". Default is the non-iterative EM developed by Shojima(2005). See details.
BFGS	a valid argument when type is equal to "Wang&Zeng". If TRUE, the Broyden-Fletcher-Goldfarb-Shanno (BFGS) algorithm is used to update Hessian.

Details

Samejima (1973) proposed an IRT model for continuous item scores as a limiting form of the graded response model. Even though the Continuous Response Model (CRM) is as old as the well-known and popular binary and polytomous IRT models, it is not commonly used in practice. This may be due to the lack of accessible computer software to estimate the parameters for the CRM. Another reason might be that the continuous outcome is not a type of response format commonly observed in the field of education and psychology. There are few published studies that used the CRM (Ferrando, 2002; Wang & Zeng, 1998). In the field of education, the model may have useful applications for estimating a single reading ability from a set of reading passages in the Curriculum Based Measurement context. Also, this type of continuous response format may be more frequently observed in the future as the use of computerized testing increases. For instance, the examinees or raters may check anywhere on the line between extremely positive and extremely negative in a computerized testing environment rather than responding to a likert type item.

Wang & Zeng (1998) proposed a re-parameterized version of the CRM. In this re-parameterization, the probability of an examinee *i* with a spesific θ obtaining a score of *x* or higher on a particular item *j* with a continuous measurement scale ranging from 0 to *k* and with the parameters *a*, *b*, and

α is defined as the following:

$$P(X_{ij} \geq x | \theta_i, a_j, b_j, \alpha_j) = \frac{1}{\sqrt{2\pi}} \int\limits_{-\infty}^{v} e^{\frac{-t^2}{2}} dt$$

$v = a_j(\theta_i - b_j - \frac{1}{\alpha_j} ln \frac{x_{ij}}{k_j - x_{ij}})$

where *a* is a discrimination parameter, *b* is a difficulty parameter, and α represents a scaling parameter that defines some scale transformation linking the original observed score scale to the θ scale (Wang & Zeng, 1998). *k* is the maximum possible score for the item. *a* and *b* in this model have practical meaning and are interpreted same as in the binary and polytomous IRT models. α is a scaling parameter and does not have a practical meaning.

In the model fitting process, the observed *X* scores are first transformed to a random variable *Z* by using the following equation:

$$Z_{ij} = ln(\frac{X_{ij}}{k_j - X_{ij}})$$

Then, the conditional probability density function of the random variable *Z* is equal to:

$$f(z_{ij} | \theta_i, a_j, b_j, \alpha_j) = \frac{a_j}{\sqrt{2\pi}\alpha_j} exp^{-\frac{[a_j(\theta_i - b_j - \frac{z_{ij}}{\alpha_j})]^2}{2}}$$

The conditional pdf of *Z* is a normal density function with a mean of $\alpha(\theta\text{-}b)$ and a variance of α^2 / a^2.

Wang & Zeng (1998) proposed an algorithm to estimate the CRM parameters via marginal maximum likelihood and Expectation-Maximization (EM) algorithm. In the Expectation step, the expected log-likelihood function is obtained based on the integration over the posterior θ distribution by using the Gaussian quadrature points. In the Maximization step, the parameters are estimated by solving the first and second derivatives of the expected log-likelihood function with rescpect to *a*, *b*, and α parameters via Newton-Raphson procedure. A sequence of E-step and M-step repeats until the difference between the two consecutive loglikelihoods is smaller than a convergence criteria. This procedure is available through `type="Wang&Zeng"` argument. If type is equal to "Wang&Zeng", then user can specify `BFGS` argument as either TRUE or FALSE. If `BFGS` argument is TRUE, then the Broyden-Fletcher-Goldfarb-Shanno (BFGS) algorithm is used to approximate Hessian. If `BFGS` argument is FALSE, then the Hessian is directly computed.

Shojima (2005) simplified the EM algorithm proposed by Wang & Zeng(1998). He derived the closed formulas for computing the loglikelihood in the E-step and estimating the parameters in the M-step (please see the reference paper below for equations). He showed that the equations of the first derivatives in the M-step can be solved algebraically by assuming flat (non-informative) priors for the item parameters. This procedure is available through `type="Shojima"` argument.

Value

`EstCRMitem()` returns an object of class `"CRM"`. An object of class `"CRM"` is a list containing the following components:

`data`	original data
`descriptive`	descriptive statistics for the original data and the transformed z scores
`param`	estimated item parameters in the last EM Cycle
`iterations`	a list that reports the information for each EM cycle
`dif`	The difference of loglikelihoods between the last two EM cycles

Note

* The ID variable ,if included, has to be excluded from the data before the analysis. If the format of the data is not a "data frame", the function does not work. Please use "as.data.frame()" to change the format of the input data.

* A previous published simulation study (Shojima, 2005) was replicated as an example to check the performance of `EstCRMitem()`.Please see the examples of `simCRM` in this package.

*The example below is a reproduction of the results from another published study (Ferrando, 2002). Dr. Ferrando kindly provided the dataset used in the published study. The dataset was previously analyzed by using EM2 program and the estimated CRM item parameters were reported in Table 2 in the paper. The estimates from `EstCRMitem()` is comparable to the item parameter estimates reported in Table 2.

Author(s)

Cengiz Zopluoglu

References

Ferrando, P.J.(2002). Theoretical and Empirical Comparison between Two Models for Continuous Item Responses. *Multivariate Behavioral Research*, 37(4), 521-542.

Samejima, F.(1973). Homogeneous Case of the Continuous Response Model. *Psychometrika*, 38(2), 203-219.

Shojima, K.(2005). A Noniterative Item Parameter Solution in Each EM Cycle of the Continuous Response Model. *Educational Technology Research*, 28, 11-22.

Wang, T. & Zeng, L.(1998). Item Parameter Estimation for a Continuous Response Model Using an EM Algorithm. *Applied Psychological Measurement*, 22(4), 333-343.

See Also

`EstCRMperson` for estimating person parameters, `fitCRM` for computing item-fit residual statistics and drawing empirical 3D item category response curves, `plotCRM` for drawing theoretical 3D item category response curves, `simCRM` for generating data under CRM.

Examples

```
## Not run:

##load the dataset EPIA
data(EPIA)

##Check the class. "data.frame" is required.
class(EPIA)

##Define the vectors "max.item" and "min.item". The maximum possible
##score was 112 and the minimum possible score was 0 for all items

max.item <- c(112,112,112,112,112)
min.item <- c(0,0,0,0,0)

##The maximum number of EM Cycle and the convergence criteria can be
##specified

max.EMCycle=200
converge=.01

##Estimate the item parameters
CRM <- EstCRMitem(EPIA, max.item, min.item, max.EMCycle, converge)
CRM

##Other details

CRM$descriptive
CRM$param
CRM$iterations
CRM$dif

##load the dataset SelfEff

data(SelfEff)

##Check the class. "data.frame" is required.

class(SelfEff)

##Define the vectors "max.item" and "min.item". The maximum possible
##score was 11 and the minimum possible score was 0 for all items

max.item <- c(11,11,11,11,11,11,11,11,11,11)
min.item <- c(0,0,0,0,0,0,0,0,0,0)

##Estimate the item parameters
CRM2 <- EstCRMitem(SelfEff, max.item, min.item, max.EMCycle=200, converge=.01)
CRM2

##Other details
```

```
    CRM2$descriptive
    CRM2$param
    CRM2$iterations
    CRM2$dif

  ## End(Not run)
```

EstCRMperson   *Estimating Person Parameters for the Continuous Response Model*

Description

Estimate person parameters for the Continuous Response Model(Samejima,1973)

Usage

```
EstCRMperson(data,ipar,min.item,max.item)
```

Arguments

data	a data frame with *N* rows and *m* columns, with *N* denoting the number of subjects and *m* denoting the number of items.
ipar	a matrix with *m* rows and three columns, with *m* denoting the number of items. The first column is the *a* parameters, the second column is the *b* parameters, and the third column is the *alpha* parameters
min.item	a vector of length *m* indicating the minimum possible score for each item.
max.item	a vector of length *m* indicating the maximum possible score for each item.

Details

Samejima(1973) derived the closed formula for the maximum likelihood estimate (MLE) of the person parameters (ability levels) under Continuous Response Model. Given that the item parameters are known, the parameter for person *i* (ability level) can be estimated from the following equation:

$$MLE(\theta_i) = \frac{\sum_{j=1}^{m}[a_j^2(b_j + \frac{z_{ij}}{\alpha_j})]}{\sum_{j=1}^{m}\alpha_j^2}$$

As suggested in Ferrando (2002), the ability level estimates are re-scaled once they are obtained from the equation above. So, the mean and the variance of the sample estimates equal those of the latent distribution on which the estimation is based.

Samejima (1973) also derived that the item information is equal to:

$$I(\theta_i) = a^2$$

So, the item information is constant along the theta continuum in the Continuous Response Model.

The asymptotic standard error of the MLE ability level estimate is the square root of the reciprocal of the total test information:

$$SE(\hat{\theta}_i) = \sqrt{\frac{1}{\sum_{j=1}^{m} a_j^2}}$$

In the Continuous Response Model, the standard error of the ability estimate is same for all examinees unless an examinee has missing data. If the examinee has missing data, the ability level estimate and its standard error is obtained from the available item scores.

Value

EstCRMperson() returns an object of class "CRMtheta". An object of class "CRMtheta" contains the following component:

thetas   a three column matrix, the first column is the examinee ID, the second column is the re-scaled theta estimate, the third column is the standard error of the theta estimate.

Note

See the examples of simCRM for an R code that runs a simulation to examine the recovery of person parameter estimates by EstCRMperson.

Author(s)

Cengiz Zopluoglu

References

Ferrando, P.J.(2002). Theoretical and Empirical Comparison between Two Models for Continuous Item Responses. *Multivariate Behavioral Research*, 37(4), 521-542.

Samejima, F.(1973). Homogeneous Case of the Continuous Response Model. *Psychometrika*, 38(2), 203-219.

Wang, T. & Zeng, L.(1998). Item Parameter Estimation for a Continuous Response Model Using an EM Algorithm. *Applied Psychological Measurement*, 22(4), 333-343.

See Also

EstCRMitem for estimating item parameters, fitCRM for computing item-fit residual statistics and drawing empirical 3D item category response curves, plotCRM for drawing theoretical 3D item category response curves, simCRM for generating data under CRM.

Examples

```
## Not run:
  #Load the dataset EPIA

  data(EPIA)

  ##Define the vectors "max.item" and "min.item". The maximum possible
  ##score was 112 and the minimum possible score was 0 for all items
  max.item <- c(112,112,112,112,112)
  min.item <- c(0,0,0,0,0)

  ##Estimate item parameeters
  CRM <- EstCRMitem(EPIA, max.item, min.item, max.EMCycle = 500, converge = 0.01)
  par <- CRM$param

  ##Estimate the person parameters

  CRMthetas <- EstCRMperson(EPIA,par,min.item,max.item)
  theta.par <- CRMthetas$thetas
  theta.par

  #Load the dataset SelfEff
  data(SelfEff)

  ##Define the vectors "max.item" and "min.item". The maximum possible
  ##score was 11 and the minimum possible score was 0 for all items

  max.item <- c(11,11,11,11,11,11,11,11,11,11)
  min.item <- c(0,0,0,0,0,0,0,0,0,0)

  ##Estimate the item parameters
  CRM2 <- EstCRMitem(SelfEff, max.item, min.item, max.EMCycle=200, converge=.01)
  par2 <- CRM2$param

  ##Estimate the person parameters

  CRMthetas2 <- EstCRMperson(SelfEff,par2,min.item,max.item)
  theta.par2 <- CRMthetas2$thetas
  theta.par2

## End(Not run)
```

---

fitCRM   *Compute item fit residual statistics for the Continuous Response Model*

Description

Compute item fit residual statistics for the Continuous Response Model as described in Ferrando (2002)

Usage

```
fitCRM(data, ipar, est.thetas, max.item,group)
```

Arguments

data	a data frame with *N* rows and *m* columns, with *N* denoting the number of subjects and *m* denoting the number of items.
ipar	a matrix with *m* rows and three columns, with *m* denoting the number of items. The first column is the *a* parameters, the second column is the *b* parameters, and the third column is the *alpha* parameters
est.thetas	object of class "CRMtheta" obtained by using EstCRMperson()
max.item	a vector of length *m* indicating the maximum possible score for each item.
group	an integer, number of ability groups to compute item fit residual statistics. Default 20.

Details

The function computes the item fit residual statistics as decribed in Ferrando (2002). The steps in the procedure are as the following:

1- Re-scaled θ estimates are obtained.

2- θ estimates are sorted and assigned to *k* intervals on the θ continuum.

3- The mean item score $\bar{X}_{mk}$ is computed in each interval for each of the items.

4- The expected item score and the conditional variance in each interval are obtained with the item parameter estimates and taking the median theta estimate for the interval.

5- An approximate standardized residual for item *m* at ability interval *k* is obtained as:

$$z_{mk} = \frac{\bar{X}_{mk} - E(X_m|\theta_k)}{\sqrt{\frac{\sigma^2(X_m|\theta_k)}{N_k}}}$$

Value

fit.stat	a data frame with *k* rows and *m+1* columns with *k* denoting the number of ability intervals and *m* denoting the number of items. The first column is the ability interval. Other elements are the standardized residuals of item *m* in ability interval *k*.
emp.irf	a list of length *m* with *m* denoting the number of items. Each element is a 3D plot representing the item category response curve based on the empirical probabilities. See examples below.

Author(s)

Cengiz Zopluoglu

References

Ferrando, P.J.(2002). Theoretical and Empirical Comparison between Two Models for Continuous Item Responses. *Multivariate Behavioral Research*, 37(4), 521-542.

See Also

EstCRMperson for estimating person parameters, EstCRMitem for estimating item parameters plotCRM for drawing theoretical 3D item category response curves, simCRM for generating data under CRM.

Examples

```
## Not run:

  ##load the dataset EPIA

  data(EPIA)

  ##Define the vectors "max.item" and "min.item". The maximum possible
  ##score was 112 and the minimum possible score was 0 for all items

  max.item <- c(112,112,112,112,112)
  min.item <- c(0,0,0,0,0)

  ##Estimate item parameters

  CRM <- EstCRMitem(EPIA, max.item, min.item, max.EMCycle = 500, converge = 0.01)
  par <- CRM$param

  ##Estimate the person parameters

  CRMthetas <- EstCRMperson(EPIA,par,min.item,max.item)

  ##Compute the item fit residual statistics and empirical item category
  ##response curves

  fit <- fitCRM(EPIA, par, CRMthetas, max.item,group=10)

  ##Item-fit residual statistics

  fit$fit.stat

  ##Empirical item category response curves
  fit$emp.irf[[1]]   #Item 1

## End(Not run)
```

---

plotCRM   *Draw Three-Dimensional Item Category Response Curves*

Description

This function draws the 3D item category response curves given the item parameter matrix and the maximum and minimum possible values that examinees can get as a score for the item. This 3D representation is an extension of 2D item category response curves in binary and polytomous IRT models. Due to the fact that there are many possible response categories, another dimension is drawn for the response categories in the 3D representation.

Usage

```
plotCRM(ipar, item, min.item, max.item)
```

Arguments

ipar	a matrix with *m* rows and three columns, with *m* denoting the number of items. The first column is the *a* parameters, the second column is the *b* parameters, and the third column is the *alpha* parameters
item	item number, an integer between 1 and *m* with *m* denoting the number of items
min.item	a vector of length *m* indicating the minimum possible score for each item.
max.item	a vector of length *m* indicating the maximum possible score for each item.

Value

irf	a 3D theoretical item category response curves

Author(s)

Cengiz Zopluoglu

See Also

[EstCRMitem](#) for estimating item parameters, [EstCRMperson](#) for estimating person parameters, [fitCRM](#) for computing item-fit residual statistics and drawing empirical 3D item category response curves, [simCRM](#) for generating data under CRM.

Examples

```
## Not run:

  ##load the dataset EPIA

  data(EPIA)

  ##Define the vectors "max.item" and "min.item". The maximum possible
```

```
    ##score was 112 and the minimum possible score was 0 for all items

    min.item <- c(0,0,0,0,0)
    max.item <- c(112,112,112,112,112)

    ##Estimate item parameters

    CRM <- EstCRMitem(EPIA, max.item, min.item, max.EMCycle = 500, converge = 0.01)
    par <- CRM$param

    ##Draw theoretical item category response curves for Item 2
    plotCRM(par,2,min.item, max.item)

  ## End(Not run)
```

---

print.CRM   *Summary output for the object of class* CRM

Description

Print the output for the object of class CRM

Usage

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

Arguments

x   an object of class CRM obtained from EstCRMitem
...   More options to pass to summary and print

Value

The summary output includes the information from the last EM Cycle, final parameter estimates, and the descriptive statistics of the original data and the transformed data.

SelfEff   *Teaching Self Efficacy Scale*

Description

The data came from a published study and was kindly provided by Dr. Adnan Kan. A group of 307 pre-service teachers graduated from various departments in the college of education were asked to check on a 11 cm line segment with two end points (can not do at all, highly certain can do) using their own judgement for the 10 items that measure teacher self-efficacy on different activities. The direct item score was the distance in cm of the check mark from the left end point (Kan, 2009).

Usage

```
data(SelfEff)
```

Format

A data frame with 307 observations on the following 10 items.

Item1 Implementing variety of teaching strategies

Item2 Creating learning environments for the effective participation of students

Item3 Providing the acquisition of learning experiences according to the students' physical, emotional, and psychomotor characteristics

Item4 Creating learning environments in which students can effectively express themselves

Item5 Implementing a variety of assessment strategies (formal, informal)

Item6 Understanding differences in students' learning styles and providing learning opportunities in compliance with these differences

Item7 Using teaching methods and techniques within the framework of educational plans and aims

Item8 Discovering students' interests and skills and guiding them accordingly

Item9 Using educational technology effectively in classrooms to support teaching

Item10 Designing and implementing educational variables such as motivation, reinforcement, feedback, etc.

Source

Kan, A.(2009). Effect of scale response format on psychometric properties in teaching self-efficacy. *Euroasian Journal of Educational Research*, *34*, 215-228.

References

Kan, A.(2009). Effect of scale response format on psychometric properties in teaching self-efficacy. *Euroasian Journal of Educational Research*, *34*, 215-228.

---

simCRM   *Generating Data under the Continuous Response Model*

Description

Generating data under the Continuous Response Model

Usage

```
simCRM(thetas, true.param, max.item)
```

Arguments

thetas	a vector of length *N* with *N* denoting the number of examinees. Each element of the vector is the true ability level for an examinee
true.param	a matrix of true item parameters with *m* rows and three columns, with *m* denoting the number of items. The first column is the *a* parameters, the second column is the *b* parameters, and the third column is the *alpha* parameters
max.item	a vector of length *m* indicating the maximum possible score for each hypothetical item.

Details

The simCRM generates data under Continuous Response Model as described in Shojima(2005). Given the true ability level for person *i* and the true item parameters for item *j*, the transformed response z_{ij} of person *i* for item *j* follows a normal distribution with a mean of $\alpha_j(\theta_i - \beta_j)$ and a standard deviation of $\frac{\alpha_j^2}{a_j^2}$.

Value

a data frame with *N* rows and *m* columns with *N* denoting the number of observations and *m* denoting the number of items.

Author(s)

Cengiz Zopluoglu

References

Shojima, K.(2005). A Noniterative Item Parameter Solution in Each EM Cycle of the Continuous Response Model. *Educational Technology Research*, 28, 11-22.

See Also

[EstCRMitem](#) for estimating item parameters, [EstCRMperson](#) for estimating person parameters, [fitCRM](#) for computing item-fit statistics and drawing empirical 3D item response curves, [plotCRM](#) for drawing theoretical 3D item category response curves,

Examples

```
## Not run:

#####################################################
#                      Example 1:                   #
#   Basic data generation and parameter recovery    #
#####################################################

#Generate true person ability parameters for 1000 examinees from
#a standard normal distribution

true.thetas <- rnorm(1000,0,1)
```

```
#Generate the true item parameter matrix for the hypothetical items

true.par <- matrix(c(.5,1,1.5,2,2.5,
-1,-.5,0,.5,1,1,.8,1.5,.9,1.2),
nrow=5,ncol=3)
true.par

#Generate the vector maximum possible scores that students can
#get for the items

max.item <- c(30,30,30,30,30)

#Generate the response matrix

simulated.data <- simCRM(true.thetas,true.par,max.item)

#Let's examine the simulated data

head(simulated.data)
summary(simulated.data)

#Let's try to recover the item parameters

min.item <- c(0,0,0,0,0)

CRM <- EstCRMitem(simulated.data,max.item, min.item,
max.EMCycle=500,converge=0.01)

#Compare the true item parameters with the estimated item parameters.
#The first three column is the true item parameters, and the second
#three column is the estimated item parameters

cbind(true.par,CRM$param)

#Let's recover the person parameters

par <- CRM$param
CRMthetas <- EstCRMperson(simulated.data,par,min.item,max.item)
theta.par <- CRMthetas$thetas

#Compare the true person ability parameters to the estimated person
##ability parameters.The first column is the true parameters and the
##second column is the estimated parameters

thetas <- cbind(true.thetas,theta.par[,2])
head(thetas)
cor(thetas)
plot(thetas[,1],thetas[,2])

#RMSE for the estimated person parameters

sqrt(sum((thetas[,1]-thetas[,2])^2)/nrow(thetas))
```

```
#RMSE is comparable and similar to the standard error of the
#theta estimates. Standard error of the theta estimate is the square
#root of the reciprocal of the total test information which is the sum
#of square of the "a" parameters

sqrt(1/sum(CRM$param[,1]^2))

######################################################
#                      Example 2:                    #
#    Item fit Residuals, Empirical and Theoretical   #
#   Item Category Response Curves for the Simulated  #
#                      Data Above                    #
######################################################

#Because of the run time issues during the package development,
#I run the fit analysis for a subset of simulated data above.
#The simulated data has 1000 examinees, but I run the fit analysis
#for the first 100 subjects of the simulated data. Please ignore the
#following line and run the analysis for whole data

simulated.data <- simulated.data[1:100,] #Ignore this line

par <- CRM$param
max.item <- c(30,30,30,30,30)
min.item <- c(0,0,0,0,0)
CRMthetas <- EstCRMperson(simulated.data,par,min.item,max.item)
theta.par <- CRMthetas$thetas
mean(theta.par[,2])
sd(theta.par[,2])
hist(theta.par[,2])

fit <- fitCRM(simulated.data,par, CRMthetas, max.item, group=10)

#Item-Fit Residuals

fit$fit.stat

#Empirical Item Category Response Curves
fit$emp.irf[[1]]   #Item 1
fit$emp.irf[[5]]   #Item 5

#Theoretical Item Category Response Curves
plotCRM(par,1,min.item, max.item)    #Item 1


######################################################
#                      Example 3:                    #
#    The replication of Shojima's simulation study   #
#                        2005                        #
######################################################
```

```
#In Shojima's simulation study published in 2005

 #true person parameters were generated from a standard normal distribution.
 #The natural logarithm of the true "a" parameters were generated from a N(0,0.09)
 #The true "b" parameters were generated from a N(0,1)
 #The natural logarithm of the true "alpha" parameters were generated from a N(0,0.09)

#The independent variables were the number of items and sample size
#in the simulation study

#There were 9 different conditions and 100 replications for each condition.

#In Table 1 (Shojima,2005), the RMSD statistics were reported for each condition.

#The code below replicates the same study. The results are comparable to the Table 1.
#The user
#should only specify the sample size and the number of items. Then, the user should
#run the rest of the code.
#At the end, RMSEa, RMSEb, RMSEalp are the item parameter recovery statistics which is
#comparable to Table 1


#Set the conditions for the simulation study.
#It takes longer to run for big number of replications

N=500             #sample size
n=10              #number of items
replication=1     #number if replications for each condition

###########################################################
#       Run the rest of the code from START to END        #
###########################################################

#START

true.person <- vector("list",replication)
true.item <- vector("list",replication)
est.person <- vector("list",replication)
est.item <- vector("list",replication)
simulated.datas <- vector("list",replication)

for(i in 1:replication) {
true.person[[i]] <- rnorm(N,0,1)
true.item[[i]] <- cbind(exp(rnorm(n,0,.09)),rnorm(n,0,1),1/exp(rnorm(n,0,.09)))
}

max.item <- rep(50,n)
min.item <- rep(0,n)

for(i in 1:replication) {
simulated.datas[[i]] <- simCRM(true.person[[i]],true.item[[i]],max.item)
}
```

```
      for(i in 1:replication) {
      CRM<-EstCRMitem(simulated.datas[[i]],max.item,min.item,max.EMCycle= 500,converge=0.01)
      est.item[[i]]=CRM$par
      }
      for(i in 1:replication) {
      persontheta <- EstCRMperson(simulated.datas[[i]],est.item[[i]],min.item,max.item)
      est.person[[i]]<- persontheta$thetas[,2]
      }

      #END
      ##############################################################

      #RMSE for parameter "a"

      RMSEa <- c()
      for(i in 1:replication) {
      RMSEa[i]=sqrt(sum((true.item[[i]][,1]-est.item[[i]][,1])^2)/n)
      }

      mean(RMSEa)


      #RMSE for parameter "b"

      RMSEb <- c()
      for(i in 1:replication) {
      RMSEb[i]=sqrt(sum((true.item[[i]][,2]-est.item[[i]][,2])^2)/n)
      }

      mean(RMSEb)

      #RMSE for parameter "alpha"
      RMSEalp <- c()
      for(i in 1:replication) {
      RMSEalp[i]=sqrt(sum((true.item[[i]][,3]-est.item[[i]][,3])^2)/n)
      }

      mean(RMSEalp)

      #RMSE for person parameter

      RMSEtheta <- c()
      for(i in 1:replication) {
      RMSEtheta[i]=sqrt(sum((true.person[[i]]-est.person[[i]])^2)/N)
      }

      mean(RMSEtheta)

  ## End(Not run)
```

Index