Package 'rpf'

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Type Package

Title Response Probability Functions

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Description The purpose of this package is to factor out logic and math common to Item Factor Analysis fitting, diagnostics, and analysis. It is envisioned as core support code suitable for more specialized IRT packages to build upon. Complete access to optimized C functions are made available with R_RegisterCCallable().

LazyData yes

LazyDataCompression bzip2

License GPL (>= 3)

LinkingTo RcppEigen

Suggests testthat, roxygen2, ggplot2, reshape2, gridExtra, numDeriv, knitr, mirt

Imports mvtnorm

Depends methods, parallel, R (>= 2.14.0)

VignetteBuilder knitr

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R topics documented:

`An introduction` *rpf - Response Probability Functions*

Description

Factor out logic and math common to Item Factor Analysis fitting, diagnostics, and analysis. It is envisioned as core support code suitable for more specialized IFA packages to build upon.

Details

This package provides optimized, low-level functions to map parameters to response probabilities for dichotomous (1PL, 2PL and 3PL) `rpf.drm` and polytomous (graded response `rpf.grm`, partial credit/generalized partial credit (via the nominal model), and nominal `rpf.nrm` items.

Item model parameters are passed around as a numeric vector. A 1D matrix is also acceptable. Regardless of model, parameters are always ordered as follows: discrimination/slope ("a"), difficulty/intercept ("b"), and pseudo guessing/upper-bound ("g"/"u"). If person ability ranges from negative to positive then probabilities are output from incorrect to correct. That is, a low ability person (e.g., ability = -2) will be more likely to get an item incorrect than correct. For example, a dichotomous model that returns [.25, .75] indicates a probability of .25 for incorrect and .75 for correct. A polytomous model will have the most incorrect probability at index 1 and the most correct probability at the maximum index.

All models are always in the logistic metric. To obtain normal ogive discrimination parameters, divide slope parameters by `rpf.ogive`. Item models are estimated in slope-intercept form. Input/output matrices arranged in the way most convenient for low-level processing in C. The maximum absolute logit is 35 because f(x) := 1-exp(x) loses accuracy around f(-35) and equals 1 at f(-38) due to the limited accuracy of double precision floating point.

This package could also accrete functions to support plotting (but not the actual plot functions).

References

Pritikin, J. N., Hunter, M. D., & Boker, S. M. (2015). Modular open-source software for Item Factor Analysis. Educational and Psychological Measurement, 75(3), 458-474

Thissen, D. and Steinberg, L. (1986). A taxonomy of item response models. *Psychometrika 51*(4), 567-577.

See Also

See `rpf.rparam` to create item parameters.

`as.IFAgroup` *Convert an OpenMx MxModel object into an IFA group*

Description

When "minItemsPerScore" is passed, EAP scores will be computed from the data and stored. Scores are required for some diagnostic tests. See discussion of "minItemsPerScore" in EAPscores.

Usage

```
as.IFAgroup(mxModel, data = NULL, container = NULL, ...,
  minItemsPerScore = NULL)
```

Arguments

`mxModel`	MxModel object
`data`	observed data (otherwise the data will be taken from the mxModel)
`container`	an MxModel in which to search for the latent distribution matrices
`...`	Not used. Forces remaining arguments to be specified by name.
`minItemsPerScore`	minimum number of items required to compute a score (also see description)

Value

a groups with item parameters and latent distribution

See Also

ifaTools

`bestToOmit` *Identify the columns with most missing data*

Description

If a reference column is given then only rows that are not missing on the reference column are considered. Otherwise all rows are considered.

Usage

```
bestToOmit(grp, omit, ref = NULL)
```

Arguments

`grp`	an IFA group
`omit`	the maximum number of items to omit
`ref`	the reference column (optional)

`ChenThissen1997` *Computes local dependence indices for all pairs of items*

Description

Item Factor Analysis makes two assumptions: (1) that the latent distribution is reasonably approximated by the multivariate Normal and (2) that items are conditionally independent. This test examines the second assumption. The presence of locally dependent items can inflate the precision of estimates causing a test to seem more accurate than it really is.

Usage

```
ChenThissen1997(grp, ..., data = NULL, inames = NULL, qwidth = 6,
  qpoints = 49, method = "pearson", .twotier = TRUE,
  .parallel = TRUE)
```

Arguments

grp	a list with the spec, param, mean, and cov describing the group
...	Not used. Forces remaining arguments to be specified by name.
data	data
inames	a subset of items to examine
qwidth	quadrature width
qpoints	number of equally spaced quadrature points
method	method to use to calculate P values. The default is the Pearson X^2 statistic. Use "lr" for the similar likelihood ratio statistic.
.twotier	whether to enable the two-tier optimization
.parallel	whether to take advantage of multiple CPUs (default TRUE)

Details

Statically significant entries suggest that the item pair has local dependence. Since log(.01)=-4.6, an absolute magitude of 5 is a reasonable cut-off. Positive entries indicate that the two item residuals are more correlated than expected. These items may share an unaccounted for latent dimension. Consider a redesign of the items or the use of testlets for scoring. Negative entries indicate that the two item residuals are less correlated than expected.

Value

a list with raw, pval and detail. The pval matrix is a lower triangular matrix of log P values with the sign determined by relative association between the observed and expected tables (see ordinal.gamma)

References

Chen, W.-H. & Thissen, D. (1997). Local dependence indexes for item pairs using Item Response Theory. *Journal of Educational and Behavioral Statistics, 22*(3), 265-289.

Thissen, D., Steinberg, L., & Mooney, J. A. (1989). Trace lines for testlets: A use of multiple-categorical-response models. *Journal of Educational Measurement, 26* (3), 247–260.

Wainer, H. & Kiely, G. L. (1987). Item clusters and computerized adaptive testing: A case for testlets. *Journal of Educational measurement, 24*(3), 185–201.

See Also

ifaTools

Class rpf.1dim *The base class for 1 dimensional response probability functions.*

Description

The base class for 1 dimensional response probability functions.

Class rpf.1dim.drm *Unidimensional dichotomous item models (1PL, 2PL, and 3PL).*

Description

Unidimensional dichotomous item models (1PL, 2PL, and 3PL).

Class rpf.1dim.graded *The base class for 1 dimensional graded response probability functions.*

Description

This class contains methods common to both the generalized partial credit model and the graded response model.

Class rpf.1dim.grm *The unidimensional graded response item model.*

Description

The unidimensional graded response item model.

Class rpf.1dim.lmp *Unidimensional logistic function of a monotonic polynomial.*

Description

Unidimensional logistic function of a monotonic polynomial.

Class rpf.base *The base class for response probability functions.*

Description

Item specifications should not be modified after creation.

Class rpf.mdim *The base class for multi-dimensional response probability functions.*

Description

The base class for multi-dimensional response probability functions.

Class rpf.mdim.drm *Multidimensional dichotomous item models (M1PL, M2PL, and M3PL).*

Description

Multidimensional dichotomous item models (M1PL, M2PL, and M3PL).

Class rpf.mdim.graded *The base class for multi-dimensional graded response probability functions.*

Description

This class contains methods common to both the generalized partial credit model and the graded response model.

Class rpf.mdim.grm *The multidimensional graded response item model.*

Description

The multidimensional graded response item model.

Class rpf.mdim.mcm *The multiple-choice response item model (both unidimensional and multidimensional models have the same parameterization).*

Description

The multiple-choice response item model (both unidimensional and multidimensional models have the same parameterization).

Class rpf.mdim.nrm — *The nominal response item model (both unidimensional and multidimensional models have the same parameterization).*

Description

The nominal response item model (both unidimensional and multidimensional models have the same parameterization).

compressDataFrame — *Compress a data frame into unique rows and frequencies*

Description

Compress a data frame into unique rows and frequency counts.

Usage

```
compressDataFrame(tabdata, freqColName = "freq", .asNumeric = FALSE)
```

Arguments

tabdata	An object of class data.frame
freqColName	Column name to contain the frequencies
.asNumeric	logical. Whether to cast the frequencies to the numeric type

Value

Returns a compressed data frame

Examples

```
df <- as.data.frame(matrix(c(sample.int(2, 30, replace=TRUE)), 10, 3))
compressDataFrame(df)
```

crosstabTest *Monte-Carlo test for cross-tabulation tables*

Description

This is for developers.

Usage

```
crosstabTest(ob, ex, trials)
```

Arguments

ob	observed table
ex	expected table
trials	number of Monte-Carlo trials

EAPscores *Compute EAP scores*

Description

If you have missing data then you must specify `minItemsPerScore`. This option will set scores to NA when there are too few items to make an accurate score estimate. If you are using the scores as point estimates without considering the standard error then you should set `minItemsPerScore` as high as you can tolerate. This will increase the amount of missing data but scores will be more accurate. If you are carefully considering the standard errors of the scores then you can set `minItemsPerScore` to 1. This will mimic the behavior of most other IFA software wherein scores are estimated if there is at least 1 non-NA item for the score. However, it may make more sense to set `minItemsPerScore` to 0. When set to 0, all NA rows are scored to the prior distribution.

Usage

```
EAPscores(grp, ..., naAction = NULL, compressed = FALSE)
```

Arguments

grp	a list with spec, param, data, and minItemsPerScore
...	Not used. Forces remaining arguments to be specified by name.
naAction	deprecated, will be removed in the next release
compressed	output one score per observed data row even when weightColumn is set (default FALSE)

Examples

```
spec <- list()
spec[1:3] <- list(rpf.grm(outcomes=3))
param <- sapply(spec, rpf.rparam)
data <- rpf.sample(5, spec, param)
colnames(param) <- colnames(data)
grp <- list(spec=spec, param=param, data=data, minItemsPerScore=1L)
EAPscores(grp)
```

---

expandDataFrame *Expand summary table of patterns and frequencies*

Description

Expand a summary table of unique response patterns to a full sized data-set.

Usage

```
expandDataFrame(tabdata, freqName = NULL)
```

Arguments

tabdata	An object of class data.frame with the unique response patterns and the number of frequencies
freqName	Column name containing the frequencies

Value

Returns a data frame with all the response patterns

Author(s)

Based on code by Phil Chalmers <rphilip.chalmers@gmail.com>

Examples

```
data(LSAT7)
expandDataFrame(LSAT7, freqName="freq")
```

fromFactorLoading *Convert factor loadings to response function slopes*

Description

Convert factor loadings to response function slopes

Usage

```
fromFactorLoading(loading, ogive = rpf.ogive)
```

Arguments

loading	a matrix with items in the rows and factors in the columns
ogive	the ogive constant (default rpf.ogive)

Value

a slope matrix with items in the columns and factors in the rows

fromFactorThreshold *Convert factor thresholds to response function intercepts*

Description

Convert factor thresholds to response function intercepts

Usage

```
fromFactorThreshold(threshold, loading, ogive = rpf.ogive)
```

Arguments

threshold	a matrix with items in the columns and thresholds in the rows
loading	a matrix with items in the rows and factors in the columns
ogive	the ogive constant (default rpf.ogive)

Value

an item intercept matrix with items in the columns and intercepts in the rows

itemOutcomeBySumScore *Produce an item outcome by observed sum-score table*

Description

Produce an item outcome by observed sum-score table

Usage

```
itemOutcomeBySumScore(grp, mask, interest)
```

Arguments

grp	a list with spec, param, and data
mask	a vector of logicals indicating which items to include
interest	index or name of the item of interest

Examples

```
set.seed(1)
spec <- list()
spec[1:3] <- rpf.grm(outcomes=3)
param <- sapply(spec, rpf.rparam)
data <- rpf.sample(5, spec, param)
colnames(param) <- colnames(data)
grp <- list(spec=spec, param=param, data=data)
itemOutcomeBySumScore(grp, c(FALSE,TRUE,TRUE), 1L)
```

kct *Knox Cube Test dataset*

Description

These data from Wright & Stone (1979, p. 31) were fit with Winsteps 3.73 using a 1PL model (slope fixed to 1).

References

Wright, B. D. & Stone, M. H. (1979). *Best Test Design: Rasch Measurement.* Univ of Chicago Social Research.

Examples

```
data(kct)
```

logit *Transform from [0,1] to the reals*

Description

The logit function is a standard transformation from [0,1] (such as a probability) to the real number line. This function is exactly the same as qlogis.

Usage

```
logit(p, location = 0, scale = 1, lower.tail = TRUE, log.p = FALSE)
```

Arguments

p	a number between 0 and 1
location	see qlogis
scale	see qlogis
lower.tail	see qlogis
log.p	see qlogis

See Also

qlogis, plogis

Examples

```
logit(.5)  # 0
logit(.25) # -1.098
logit(0)   # -Inf
```

LSAT6 *Description of LSAT6 data*

Description

Data from Thissen (1982); contains 5 dichotomously scored items obtained from the Law School Admissions Test, section 6.

Author(s)

Phil Chalmers <rphilip.chalmers@gmail.com>

References

Thissen, D. (1982). Marginal maximum likelihood estimation for the one-parameter logistic model. *Psychometrika, 47*, 175-186.

Examples

```
data(LSAT6)
```

LSAT7 *Description of LSAT7 data*

Description

Data from Bock & Lieberman (1970); contains 5 dichotomously scored items obtained from the Law School Admissions Test, section 7.

Author(s)

Phil Chalmers <rphilip.chalmers@gmail.com>

References

Bock, R. D., & Lieberman, M. (1970). Fitting a response model for *n* dichotomously scored items. *Psychometrika, 35*(2), 179-197.

Examples

```
data(LSAT7)
```

---

multinomialFit *Multinomial fit test*

Description

For degrees of freedom, we use the number of observed statistics (incorrect) instead of the number of possible response patterns (correct) (see Bock, Giibons, & Muraki, 1998, p. 265). This is not a huge problem because this test is becomes poorly calibrated when the multinomial table is sparse. For more accurate p-values, you can conduct a Monte-Carlo simulation study (see examples).

Usage

```
multinomialFit(grp, independenceGrp, ..., method = "lr", log = TRUE,
  .twotier = TRUE)
```

Arguments

grp	a list with the spec, param, mean, and cov describing the group
independenceGrp	a list with the spec, param, mean, and cov describing the independence group
...	Not used. Forces remaining arguments to be specified by name.
method	lr (default) or pearson
log	whether to report p-value in log units
.twotier	whether to use the two-tier optimization (default TRUE)

Details

Rows with missing data are ignored.

The full information test is described in Bartholomew & Tzamourani (1999, Section 3).

For CFI and TLI, you must provide an independence model.

References

Bartholomew, D. J., & Tzamourani, P. (1999). The goodness-of-fit of latent trait models in attitude measurement. *Sociological Methods and Research, 27*(4), 525-546.

Bock, R. D., Gibbons, R., & Muraki, E. (1988). Full-information item factor analysis. *Applied Psychological Measurement, 12*(3), 261-280.

Examples

```
# Create an example IFA group
grp <- list(spec=list())
grp$spec[1:10] <- rpf.grm()
grp$param <- sapply(grp$spec, rpf.rparam)
colnames(grp$param) <- paste("i", 1:10, sep="")
grp$mean <- 0
grp$cov <- diag(1)
grp$uniqueFree <- sum(grp$param != 0)
grp$data <- rpf.sample(1000, grp=grp)

# Monte-Carlo simulation study
mcReps <- 3    # increase this to 10,000 or so
stat <- rep(NA, mcReps)
for (rx in 1:mcReps) {
   t1 <- grp
   t1$data <- rpf.sample(grp=grp)
   stat[rx] <- multinomialFit(t1)$statistic
}
sum(multinomialFit(grp)$statistic > stat)/mcReps   # better p-value
```

---

observedSumScore *Compute the observed sum-score*

Description

Compute the observed sum-score

Usage

```
observedSumScore(grp, ..., mask, summary = TRUE)
```

Arguments

grp	a list with spec, param, and data
...	Not used. Forces remaining arguments to be specified by name.
mask	a vector of logicals indicating which items to include
summary	whether to return a summary (default) or per-row scores

Examples

```
spec <- list()
spec[1:3] <- rpf.grm(outcomes=3)
param <- sapply(spec, rpf.rparam)
data <- rpf.sample(5, spec, param)
colnames(param) <- colnames(data)
grp <- list(spec=spec, param=param, data=data)
observedSumScore(grp)
```

---

omitItems *Omit the given items*

Description

Omit the given items

Usage

```
omitItems(grp, excol)
```

Arguments

grp	an IFA group
excol	vector of column names to omit

---

omitMostMissing *Omit items with the most missing data*

Description

Items with no missing data are never omitted, regardless of the number of items requested.

Usage

```
omitMostMissing(grp, omit)
```

Arguments

grp	an IFA group
omit	the maximum number of items to omit

orderCompletely *Order a data.frame by missingness and all columns*

Description

Completely order all rows in a data.frame.

Usage

```
orderCompletely(observed)
```

Arguments

observed	a data.frame holding ordered factors in every column

Value

the sorted order of the rows

Examples

```
df <- as.data.frame(matrix(c(sample.int(2, 30, replace=TRUE)), 10, 3))
mask <- matrix(c(sample.int(3, 30, replace=TRUE)), 10, 3) == 1
df[mask] <- NA
df[orderCompletely(df),]
```

ordinal.gamma *Compute the ordinal gamma association statistic*

Description

Compute the ordinal gamma association statistic

Usage

```
ordinal.gamma(mat)
```

Arguments

mat	a cross tabulation matrix

References

Agresti, A. (1990). Categorical data analysis. New York: Wiley.

Examples

```
# Example data from Agresti (1990, p. 21)
jobsat <- matrix(c(20,22,13,7,24,38,28,18,80,104,81,54,82,125,113,92), nrow=4, ncol=4)
ordinal.gamma(jobsat)
```

`ptw2011.gof.test` *Compute the P value that the observed and expected tables come from the same distribution*

Description

This test is an alternative to Pearson's X^2 goodness-of-fit test. In contrast to Pearson's X^2, no ad hoc cell collapsing is needed to avoid an inflated false positive rate in situations of sparse cell frequences. The statistic rapidly converges to the Monte-Carlo estimate as the number of draws increases.

Usage

```
ptw2011.gof.test(observed, expected)
```

Arguments

`observed`	observed matrix
`expected`	expected matrix

Value

The P value indicating whether the two tables come from the same distribution. For example, a significant result (P < alpha level) rejects the hypothesis that the two matrices are from the same distribution.

References

Perkins, W., Tygert, M., & Ward, R. (2011). Computing the confidence levels for a root-mean-square test of goodness-of-fit. *Applied Mathematics and Computations, 217*(22), 9072-9084.

Examples

```
draws <- 17
observed <- matrix(c(.294, .176, .118, .411), nrow=2) * draws
expected <- matrix(c(.235, .235, .176, .353), nrow=2) * draws
ptw2011.gof.test(observed, expected)  # not signficiant
```

read.flexmirt *Read a flexMIRT PRM file*

Description

Load the item parameters from a flexMIRT PRM file.

Usage

```
read.flexmirt(fname)
```

Arguments

fname	file name

Value

a list of groups each with item parameters and the latent distribution

rpf.1dim.fit *Calculate item and person Rasch fit statistics*

Description

Note: These statistics are only appropriate if all discrimination parameters are fixed equal and items are conditionally independent (see ChenThissen1997). A best effort is made to cope with missing data.

Usage

```
rpf.1dim.fit(spec, params, responses, scores, margin, group = NULL,
  wh.exact = TRUE)
```

Arguments

spec	list of item models
params	matrix of item parameters, 1 per column
responses	persons in rows and items in columns
scores	model derived person scores
margin	for people 1, for items 2
group	spec, params, data, and scores can be provided in a list instead of as arguments
wh.exact	whether to use the exact Wilson-Hilferty transformation

Details

Exact distributional properties of these statistics are unknown (Masters & Wright, 1997, p. 112). For details on the calculation, refer to Wright & Masters (1982, p. 100).

The Wilson-Hilferty transformation is biased for less than 25 items. Consider wh.exact=FALSE for less than 25 items.

References

Masters, G. N. & Wright, B. D. (1997). The Partial Credit Model. In W. van der Linden & R. K. Kambleton (Eds.), *Handbook of modern item response theory* (pp. 101-121). Springer.

Wilson, E. B., & Hilferty, M. M. (1931). The distribution of chi-square. *Proceedings of the National Academy of Sciences of the United States of America,* 17, 684-688.

Wright, B. D. & Masters, G. N. (1982). *Rating Scale Analysis.* Chicago: Mesa Press.

Examples

```
data(kct)
responses <- kct.people[,paste("V",2:19, sep="")]
rownames(responses) <- kct.people$NAME
colnames(responses) <- kct.items$NAME
scores <- kct.people$MEASURE
params <- cbind(1, kct.items$MEASURE, logit(0), logit(1))
rownames(params) <- kct.items$NAME
items<-list()
items[1:18] <- rpf.drm()
params[,2] <- -params[,2]
rpf.1dim.fit(items, t(params), responses, scores, 2, wh.exact=TRUE)
```

---

rpf.1dim.moment *Calculate cell central moments*

Description

Popular central moments include 2 (variance) and 4 (kurtosis).

Usage

```
rpf.1dim.moment(spec, params, scores, m)
```

Arguments

spec	list of item models
params	data frame of item parameters, 1 per row
scores	model derived person scores
m	which moment

Value

moment matrix

---

rpf.1dim.residual *Calculate residuals*

Description

Calculate residuals

Usage

```
rpf.1dim.residual(spec, params, responses, scores)
```

Arguments

spec	list of item models
params	data frame of item parameters, 1 per row
responses	persons in rows and items in columns
scores	model derived person scores

Value

residuals

---

rpf.1dim.stdresidual *Calculate standardized residuals*

Description

Calculate standardized residuals

Usage

```
rpf.1dim.stdresidual(spec, params, responses, scores)
```

Arguments

spec	list of item models
params	data frame of item parameters, 1 per row
responses	persons in rows and items in columns
scores	model derived person scores

Value

standardized residuals

rpf.dLL *Item parameter derivatives*

Description

Evaluate the partial derivatives of the log likelihood with respect to each parameter at `where` with `weight`.

Usage

```
rpf.dLL(m, param, where, weight)
```

Arguments

`m`	item model
`param`	item parameters
`where`	location in the latent space
`weight`	per outcome weights (typically derived by observation)

Details

It is not easy to write an example for this function. To evaluate the derivative, you need to sum the derivatives across a quadrature. You also need response outcome weights at each quadrature point. It is not anticipated that this function will be often used in R code. It's mainly to expose a C-level function for occasional debugging.

Value

first and second order partial derivatives of the log likelihood evaluated at `where`. For p parameters, the first p values are the first derivative and the next p(p+1)/2 columns are the lower triangle of the second derivative.

See Also

The numDeriv package.

`rpf.drm` *Create a dichotomous response model*

Description

For slope vector a, intercept c, pseudo-guessing parameter g, upper bound u, and latent ability vector theta, the response probability function is

$$\mathrm{P}(\mathrm{pick} = 0|a, c, g, u, \theta) = 1 - \mathrm{P}(\mathrm{pick} = 1|a, c, g, u, \theta)$$

$$\mathrm{P}(\mathrm{pick} = 1|a, c, g, u, \theta) = g + (u - g)\frac{1}{1 + \exp(-(a\theta + c))}$$

Usage

```
rpf.drm(factors = 1, multidimensional = TRUE, poor = FALSE)
```

Arguments

`factors`	the number of factors
`multidimensional`	whether to use a multidimensional model. Defaults to `TRUE`.
`poor`	if TRUE, use the traditional parameterization of the 1d model instead of the slope-intercept parameterization

Details

The pseudo-guessing and upper bound parameter are specified in logit units (see `logit`).

For discussion on the choice of priors see Cai, Yang, and Hansen (2011, p. 246).

Value

an item model

References

Cai, L., Yang, J. S., & Hansen, M. (2011). Generalized Full-Information Item Bifactor Analysis. *Psychological Methods, 16*(3), 221-248.

Examples

```
spec <- rpf.drm()
rpf.prob(spec, rpf.rparam(spec), 0)
```

rpf.dTheta *Item derivatives with respect to the location in the latent space*

Description

Evaluate the partial derivatives of the response probability with respect to ability. See rpf.info for an application.

Usage

```
rpf.dTheta(m, param, where, dir)
```

Arguments

m	item model
param	item parameters
where	location in the latent distribution
dir	if more than 1 factor, a basis vector]

rpf.grm *Create a graded response model*

Description

For outcomes k in 0 to K, slope vector a, intercept vector c, and latent ability vector theta, the response probability function is

$$\mathrm{P}(\mathrm{pick}=0|a,c,\theta) = 1 - \mathrm{P}(\mathrm{pick}=1|a,c_1,\theta)$$

$$\mathrm{P}(\mathrm{pick}=k|a,c,\theta) = \frac{1}{1+\exp(-(a\theta+c_k))} - \frac{1}{1+\exp(-(a\theta+c_{k+1}))}$$

$$\mathrm{P}(\mathrm{pick}=K|a,c,\theta) = \frac{1}{1+\exp(-(a\theta+c_K))}$$

Usage

```
rpf.grm(outcomes = 2, factors = 1, multidimensional = TRUE)
```

Arguments

outcomes	The number of choices available
factors	the number of factors
multidimensional	whether to use a multidimensional model. Defaults to TRUE.

Details

The graded response model was designed for a item with a series of dependent parts where a higher score implies that easier parts of the item were surmounted. If there is any chance your polytomous item has independent parts then consider `rpf.nrm`. If your categories cannot cross then the graded response model provides a little more information than the nominal model. Stronger a priori assumptions offer provide more power at the cost of flexibility.

Value

an item model

Examples

```
spec <- rpf.grm()
rpf.prob(spec, rpf.rparam(spec), 0)
```

`rpf.id_of` *Convert an rpf item model name to an ID*

Description

This is an internal function and should not be used.

Usage

```
rpf.id_of(name)
```

Arguments

`name`	name of the item model (string)

Value

the integer ID assigned to the given model

rpf.info — *Map an item model, item parameters, and person trait score into a information vector*

Description

Map an item model, item parameters, and person trait score into a information vector

Usage

```
rpf.info(ii, ii.p, where, basis = 1)
```

Arguments

ii	an item model
ii.p	item parameters
where	the location in the latent distribution
basis	if more than 1 factor, a positive basis vector

Value

Fisher information

References

Dodd, B. G., De Ayala, R. J. & Koch, W. R. (1995). Computerized adaptive testing with polytomous items. *Applied psychological measurement 19*(1), 5-22.

Examples

```
i1 <- rpf.drm()
i1.p <- c(.6,1,.1,.95)
theta <- seq(0,3,.05)
plot(theta, rpf.info(i1, i1.p, t(theta)), type="l")
```

rpf.lmp — *Create logistic function of a monotonic polynomial (LMP) model*

Description

This model is a dichotomous response model originally proposed by Liang (2007) and is implemented using the parameterization by Falk & Cai (in press).

Usage

```
rpf.lmp(k = 0, multidimensional = FALSE)
```

Arguments

k	a non-negative integer that controls the order of the polynomial (2k+1) with a default of k=0 (1st order polynomial = 2PL).
multidimensional	whether to use a multidimensional model. Defaults to FALSE. The multidimensional version is not yet available.

Details

The LMP model replaces the linear predictor part of the two-parameter logistic function with a monotonic polynomial, $m(\theta, \omega, \xi, \alpha, \tau)$,

$$\mathrm{P}(\mathrm{pick} = 1 | \omega, \xi, \alpha, \tau, \theta) = \frac{1}{1 + \exp(-(\xi + m(\theta, \omega, \xi, \alpha, \tau)))}$$

where α and τ are vectors of length k.

The order of the polynomial is always odd and is controlled by the user specified non-negative integer, k. The model contains 2+2*k parameters and are used as input to the rpf.prob function in the following order: ω - the natural log of the slope of the item model when k=0, ξ - the intercept, α and τ - two parameters that control bends in the polynomial. These latter parameters are repeated in the same order for models with k>1. For example, a k=2 polynomial with have an item parameter vector of: $\omega, \xi, \alpha_1, \tau_1, \alpha_2, \tau_2$.

See Falk & Cai (in press) for more details as to how the polynomial is constructed. In general, the polynomial looks like the following, but coefficients, b, are not directly estimated, but are a function of the item parameters.

$$\mathrm{m}(\theta) = \xi + b_1\theta + b_2\theta^2 + \ldots + b_{2k+1}\theta^{2k+1}$$

At the lowest order polynomial (k=0) the model reduces to the two-parameter logistic (2PL) model. However, parameterization of the slope parameter, ω, is currently different than the 2PL (i.e., slope = exp(ω)). This parameterization ensures that the response function is always monotonically increasing without requiring constrained optimization.

Please note that the functions implementing this item model may eventually be replaced or subsumed by an alternative item model. That is, backwards compatability will not necessarily be guaranteed and this item model should be considered experimental until further notice.

For example, Falk & Cai present a polytomous item model derived from the generalized partial credit model that also uses a monotonic polynomial as the linear predictor, referred to as a GPC-MP item model. Since the GPC-MP reduces to the LMP when the number of categories is 2, this is a potential candidate for replacing the LMP item model. An alternative may include the retention of a dichotomous response model, but with a lower (and upper) asymptote that further reduces to the three-parameter logistic (or four-parameter logistic) item model when k=0. Finally, future versions may reparameterize ω, or allow the option to release constraints on monotonicity. For instance, releasing constraints on ω may be desirable in cases where the user wishes to have the option of a monotonically decreasing response function. Further releasing constraints on τ would allow nonmontonicity and would be equivalent to replacing the linear predictor with a polynomial.

Value

an item model

References

Falk, C. F., & Cai, L. (in press). Maximum marginal likelihood estimation of a monotonic polynomial generalized partial credit model with applications to multiple group analysis. *Psychometrika*. http://dx.doi.org/10.1007/s11336-014-9428-7

Liang (2007). *A semi-parametric approach to estimating item response functions*. Unpublished doctoral dissertation, Department of Psychology, The Ohio State University.

Examples

```
spec <- rpf.lmp(1) # 3rd order polynomial
theta<-seq(-3,3,.1)
p<-rpf.prob(spec, c(-.11,.37,.24,-.21),theta)

spec <- rpf.lmp(2) # 5th order polynomial
p<-rpf.prob(spec, c(.69,.71,-.5,-8.48,.52,-3.32),theta)
```

`rpf.logprob` *Map an item model, item parameters, and person trait score into a probability vector*

Description

Note that in general, exp(rpf.logprob(..)) != rpf.prob(..) because the range of logits is much wider than the range of probabilities due to limitations of floating point numerical precision.

Usage

```
rpf.logprob(m, param, theta)
```

Arguments

`m`	an item model
`param`	item parameters
`theta`	the trait score(s)

Value

a vector of probabilities. For dichotomous items, probabilities are returned in the order incorrect, correct. Although redundent, both incorrect and correct probabilities are returned in the dichotomous case for API consistency with polytomous item models.

Examples

```
i1 <- rpf.drm()
i1.p <- rpf.rparam(i1)
rpf.logprob(i1, c(i1.p), -1)   # low trait score
rpf.logprob(i1, c(i1.p), c(0,1))    # average and high trait score
```

`rpf.mcm` *Create a multiple-choice response model*

Description

WARNING: This model is mostly not implemented.

Usage

```
rpf.mcm(outcomes = 2, numChoices = 5, factors = 1)
```

Arguments

`outcomes`	the number of possible outcomes
`numChoices`	the number of choices available
`factors`	the number of factors

Details

This function instantiates a multiple-choice response model.

Value

an item model

Author(s)

Jonathan Weeks <weeksjp@gmail.com>

`rpf.mean.info` *Find the point where an item provides mean maximum information*

Description

This is a point estimate of the mean difficulty of items that do not offer easily interpretable parameters such as the Generalized PCM. Since the information curve may not be unimodal, this function integrates across the latent space.

Usage

```
rpf.mean.info(spec, param, grain = 0.1)
```

Arguments

`spec`	list of item specs
`param`	list or matrix of item parameters
`grain`	the step size for numerical integration (optional)

Details

WARNING: This function is experimental and may disappear.

`rpf.mean.info1` *Find the point where an item provides mean maximum information*

Description

WARNING: This function is experimental and may disappear.

Usage

```
rpf.mean.info1(spec, iparam, grain = 0.1)
```

Arguments

`spec`	an item spec
`iparam`	an item parameter vector
`grain`	the step size for numerical integration (optional)

rpf.modify *Create a similar item specification with the given number of factors*

Description

Create a similar item specification with the given number of factors

Usage

```
rpf.modify(m, factors)
```

Arguments

m	item model
factors	the number of factors/dimensions

Examples

```
s1 <- rpf.grm(factors=3)
rpf.rparam(s1)
s2 <- rpf.modify(s1, 1)
rpf.rparam(s2)
```

rpf.nrm *Create a nominal response model*

Description

This function instantiates a nominal response model.

Usage

```
rpf.nrm(outcomes = 3, factors = 1, T.a = "trend", T.c = "trend")
```

Arguments

outcomes	The number of choices available
factors	the number of factors
T.a	the T matrix for slope parameters
T.c	the T matrix for intercept parameters

Details

The transformation matrices T.a and T.c are chosen by the analyst and not estimated. The T matrices must be invertible square matrices of size outcomes-1. As a shortcut, either T matrix can be specified as "trend" for a Fourier basis or as "id" for an identity basis. The response probability function is

$$a = T_a \alpha$$

$$c = T_c \gamma$$

$$\mathrm{P}(\mathrm{pick} = k | s, a_k, c_k, \theta) = C \, \frac{1}{1 + \exp(-(s\theta a_k + c_k))}$$

where a_k and c_k are the result of multiplying two vectors of free parameters α and γ by fixed matrices T_a and T_c, respectively; a_0 and c_0 are fixed to 0 for identification; and C is a normalizing factor to ensure that $\sum_k \mathrm{P}(\mathrm{pick} = k) = 1$.

Value

an item model

References

Thissen, D., Cai, L., & Bock, R. D. (2010). The Nominal Categories Item Response Model. In M. L. Nering & R. Ostini (Eds.), *Handbook of Polytomous Item Response Theory Models* (pp. 43–75). Routledge.

Examples

```
spec <- rpf.nrm()
rpf.prob(spec, rpf.rparam(spec), 0)
# typical parameterization for the Generalized Partial Credit Model
gpcm <- function(outcomes) rpf.nrm(outcomes, T.c=lower.tri(diag(outcomes-1),TRUE) * -1)
spec <- gpcm(4)
rpf.prob(spec, rpf.rparam(spec), 0)
```

---

rpf.numParam *Length of the item parameter vector*

Description

Length of the item parameter vector

Usage

```
rpf.numParam(m)
```

Arguments

m	item model

Examples

```
rpf.numParam(rpf.grm(outcomes=3))
rpf.numParam(rpf.nrm(outcomes=3))
```

`rpf.numSpec` *Length of the item model vector*

Description

Length of the item model vector

Usage

```
rpf.numSpec(m)
```

Arguments

`m`	item model

Examples

```
rpf.numSpec(rpf.grm(outcomes=3))
rpf.numSpec(rpf.nrm(outcomes=3))
```

`rpf.ogive` *The ogive constant*

Description

The ogive constant can be multiplied by the discrimination parameter to obtain a response curve very similar to the Normal cumulative distribution function (Haley, 1952; Molenaar, 1974). Recently, Savalei (2006) proposed a new constant of 1.749 based on Kullback-Leibler information.

Usage

```
rpf.ogive
```

Format

An object of class `numeric` of length 1.

Details

In recent years, the logistic has grown in favor, and therefore, this package does not offer any special support for this transformation (Baker & Kim, 2004, pp. 14-18).

References

Camilli, G. (1994). Teacher's corner: Origin of the scaling constant d=1.7 in Item Response Theory. *Journal of Educational and Behavioral Statistics, 19*(3), 293-295.

Baker & Kim (2004). *Item Response Theory: Parameter Estimation Techniques.* Marcel Dekker, Inc.

Haley, D. C. (1952). *Estimation of the dosage mortality relationship when the dose is subject to error* (Technical Report No. 15). Stanford University Applied Mathematics and Statistics Laboratory, Stanford, CA.

Molenaar, W. (1974). De logistische en de normale kromme [The logistic and the normal curve]. *Nederlands Tijdschrift voor de Psychologie* 29, 415-420.

Savalei, V. (2006). Logistic approximation to the normal: The KL rationale. *Psychometrika, 71*(4), 763–767.

`rpf.paramInfo` *Retrieve a description of the given parameter*

Description

Retrieve a description of the given parameter

Usage

```
rpf.paramInfo(m, num = NULL)
```

Arguments

`m`	item model
`num`	vector of parameters (defaults to all)

Value

a list containing the type, upper bound, and lower bound

Examples

```
rpf.paramInfo(rpf.drm())
```

rpf.prob *Map an item model, item parameters, and person trait score into a probability vector*

Description

Map an item model, item parameters, and person trait score into a probability vector

Usage

```
rpf.prob(m, param, theta)
```

Arguments

m	an item model
param	item parameters
theta	the trait score(s)

Value

a vector of probabilities. For dichotomous items, probabilities are returned in the order incorrect, correct. Although redundant, both incorrect and correct probabilities are returned in the dichotomous case for API consistency with polytomous item models.

Examples

```
i1 <- rpf.drm()
i1.p <- rpf.rparam(i1)
rpf.prob(i1, c(i1.p), -1)   # low trait score
rpf.prob(i1, c(i1.p), c(0,1))    # average and high trait score
```

rpf.rescale *Rescale item parameters*

Description

Adjust item parameters for changes in mean and covariance of the latent distribution.

Usage

```
rpf.rescale(m, param, mean, cov)
```

Arguments

m	item model
param	item parameters
mean	vector of means
cov	covariance matrix

Examples

```
spec <- rpf.grm()
p1 <- rpf.rparam(spec)
testPoint <- rnorm(1)
move <- rnorm(1)
cov <- as.matrix(rlnorm(1))
Icov <- solve(cov)
padj <- rpf.rescale(spec, p1, move, cov)
pr1 <- rpf.prob(spec, padj, (testPoint-move) %*% Icov)
pr2 <- rpf.prob(spec, p1, testPoint)
abs(pr1 - pr2) < 1e9
```

---

rpf.rparam *Generates item parameters*

Description

This function generates random item parameters. The version argument is available if you are writing a test that depends on reproducable random parameters (using `set.seed`).

Usage

```
rpf.rparam(m, version = 2L)
```

Arguments

m	an item model
version	the version of random parameters

Value

item parameters

Examples

```
i1 <- rpf.drm()
rpf.rparam(i1)
```

`rpf.sample` *Randomly sample response patterns given a list of items*

Description

Returns a random sample of response patterns given a list of item models and parameters. If `grp` is given then theta, items, params, mean, and cov can be omitted.

Usage

```
rpf.sample(theta, items, params, ..., prefix = "i", mean = NULL,
  cov = NULL, mcar = 0, grp = NULL)
```

Arguments

`theta`	either a vector (for 1 dimension) or a matrix (for >1 dimension) of person abilities or the number of response patterns to generate randomly
`items`	a list of item models
`params`	a list or matrix of item parameters. If omitted, random item parameters are generated for each item model.
`...`	Not used. Forces remaining arguments to be specified by name.
`prefix`	Column names are taken from param or items. If no column names are available, some will be generated using the given prefix.
`mean`	mean vector of latent distribution (optional)
`cov`	covariance matrix of latent distribution (optional)
`mcar`	proportion of generated data to set to NA (missing completely at random)
`grp`	a list with spec, param, mean, and cov

Value

Returns a data frame of response patterns

See Also

`sample`

Examples

```
# 1 dimensional items
i1 <- rpf.drm()
i1.p <- rpf.rparam(i1)
i2 <- rpf.nrm(outcomes=3)
i2.p <- rpf.rparam(i2)
rpf.sample(5, list(i1,i2), list(i1.p, i2.p))
```

science *Liking for Science dataset*

Description

These data are from Wright & Masters (1982, p. 18).

Details

All items were fit to a 3 category Partial Credit Model (PCM) using Ministep 3.75.0.

References

Wright, B. D. & Masters, G. N. (1982). *Rating Scale Analysis.* Chicago: Mesa Press.

Examples

```
data(science)
```

SitemFit *Compute the S fit statistic for a set of items*

Description

Runs SitemFit1 for every item and accumulates the results.

Usage

```
SitemFit(grp, ..., method = "pearson", log = TRUE, qwidth = 6,
  qpoints = 49L, alt = FALSE, omit = 0L, .twotier = TRUE,
  .parallel = TRUE)
```

Arguments

grp	a list with spec, param, mean, cov, data, and the free variable pattern
...	Not used. Forces remaining arguments to be specified by name.
method	whether to use a pearson or rms test
log	whether to return pvalues in log units
qwidth	the positive width of the quadrature in Z units
qpoints	the number of quadrature points
alt	whether to include the item of interest in the denominator
omit	number of items to omit (a single number) or a list of the length the number of items
.twotier	whether to enable the two-tier optimization
.parallel	whether to take advantage of multiple CPUs (default TRUE)

Value

a list of output from SitemFit1

Examples

```
grp <- list(spec=list())
grp$spec[1:20] <- rpf.grm()
grp$param <- sapply(grp$spec, rpf.rparam)
colnames(grp$param) <- paste("i", 1:20, sep="")
grp$mean <- 0
grp$cov <- diag(1)
grp$free <- grp$param != 0
grp$data <- rpf.sample(500, grp=grp)
SitemFit(grp)
```

SitemFit1 *Compute the S fit statistic for 1 item*

Description

Implements the Kang & Chen (2007) polytomous extension to S statistic of Orlando & Thissen (2000). Rows with missing data are ignored, but see the omit option.

Usage

```
SitemFit1(grp, item, free = 0, ..., method = "pearson", log = TRUE,
  qwidth = 6, qpoints = 49L, alt = FALSE, omit = 0L,
  .twotier = TRUE)
```

Arguments

grp	a list with spec, param, mean, cov, and data
item	the item of interest
free	the number of free parameters involved in estimating the item (to adjust the df)
...	Not used. Forces remaining arguments to be specified by name.
method	whether to use a pearson or rms test
log	whether to return pvalues in log units
qwidth	the positive width of the quadrature in Z units
qpoints	the number of quadrature points
alt	whether to include the item of interest in the denominator
omit	number of items to omit or a character vector with the names of the items to omit when calculating the observed and expected sum-score tables
.twotier	whether to enable the two-tier optimization

Details

This statistic is good at finding a small number of misfitting items among a large number of well fitting items. However, be aware that misfitting items can cause other items to misfit.

Observed tables cannot be computed when data is missing. Therefore, you can optionally omit items with the greatest number of responses missing relative to the item of interest.

Pearson is slightly more powerful than RMS in most cases I examined.

Setting `alt` to `TRUE` causes the tables to match published articles. However, the default setting of `FALSE` probably provides slightly more power when there are less than 10 items.

The name of the test, "S", probably stands for sum-score.

References

Kang, T. and Chen, T. T. (2007). An investigation of the performance of the generalized S-Chisq item-fit index for polytomous IRT models. ACT Research Report Series.

Orlando, M. and Thissen, D. (2000). Likelihood-Based Item-Fit Indices for Dichotomous Item Response Theory Models. *Applied Psychological Measurement, 24*(1), 50-64.

`stripData` *Strip data and scores from an IFA group*

Description

In addition, the weightColumn is reset to NULL.

Usage

```
stripData(grp)
```

Arguments

`grp`	an IFA group

`sumScoreEAP` *Compute the sum-score EAP table*

Description

Observed tables cannot be computed when data is missing. Therefore, you can optionally omit items with the greatest number of responses missing when conducting the distribution test.

Usage

```
sumScoreEAP(grp, ..., qwidth = 6, qpoints = 49L, .twotier = TRUE)
```

Arguments

grp	a list with spec, param, mean, and cov
...	Not used. Forces remaining arguments to be specified by name.
qwidth	positive width of quadrature in Z units
qpoints	number of quadrature points
.twotier	whether to enable the two-tier optimization

Details

When two-tier covariance structure is detected, EAP scores are only reported for primary factors. It is possible to compute EAP scores for specific factors, but it is not clear why this would be useful because they are conditional on the specific factor sum scores. Moveover, the algorithm to compute them efficiently has not been published yet (as of Jun 2014).

Examples

```
# see Thissen, Pommerich, Billeaud, & Williams (1995, Table 2)
 spec <- list()
 spec[1:3] <- rpf.grm(outcomes=4)

 param <- matrix(c(1.87, .65, 1.97, 3.14,
                   2.66, .12, 1.57, 2.69,
                   1.24, .08, 2.03, 4.3), nrow=4)
 # fix parameterization
 param <- apply(param, 2, function(p) c(p[1], p[2:4] * -p[1]))

 grp <- list(spec=spec, mean=0, cov=matrix(1,1,1), param=param)
 sumScoreEAP(grp)
```

---

sumScoreEAPTest *Conduct the sum-score EAP distribution test*

Description

Conduct the sum-score EAP distribution test

Usage

```
sumScoreEAPTest(grp, ..., qwidth = 6, qpoints = 49L, .twotier = TRUE)
```

Arguments

grp	a list with spec, param, mean, and cov
...	Not used. Forces remaining arguments to be specified by name.
qwidth	positive width of quadrature in Z units
qpoints	number of quadrature points
.twotier	whether to enable the two-tier optimization

References

Li, Z., & Cai, L. (2012, July). Summed score likelihood based indices for testing latent variable distribution fit in Item Response Theory. Paper presented at the annual International Meeting of the Psychometric Society, Lincoln, NE. Retrieved from http://www.cse.ucla.edu/downloads/files/SD2-final-4.pdf

---

`tabulateRows` *Tabulate data.frame rows*

Description

Like `tabulate` but entire rows are the unit of tabulation. The data.frame is not sorted, but must be sorted already.

Usage

```
tabulateRows(observed)
```

Arguments

`observed`	a sorted data.frame holding ordered factors in every column

See Also

`orderCompletely`

Examples

```
df <- as.data.frame(matrix(c(sample.int(2, 30, replace=TRUE)), 10, 3))
df <- df[orderCompletely(df),]
tabulateRows(df)
```

`toFactorLoading` *Convert response function slopes to factor loadings*

Description

All slopes are divided by the ogive constant. Then the following transformation is applied to the slope matrix,

Usage

```
toFactorLoading(slope, ogive = rpf.ogive)
```

Arguments

slope	a matrix with items in the columns and slopes in the rows
ogive	the ogive constant (default rpf.ogive)

Details

$$\frac{\text{slope}}{\left[1 + \text{rowSums}(\text{slope}^2)\right]^{\frac{1}{2}}}$$

Value

a factor loading matrix with items in the rows and factors in the columns

See Also

rpf.ogive

---

toFactorThreshold *Convert response function intercepts to factor thresholds*

Description

Convert response function intercepts to factor thresholds

Usage

```
toFactorThreshold(intercept, slope, ogive = rpf.ogive)
```

Arguments

intercept	a matrix with items in the columns and intercepts in the rows
slope	a matrix with items in the columns and slopes in the rows
ogive	the ogive constant (default rpf.ogive)

Value

a factor threshold matrix with items in the columns and factor thresholds in the rows

`write.flexmirt` *Write a flexMIRT PRM file*

Description

Formats item parameters in the way that flexMIRT expects to read them. Use `read.flexmirt` to see what shape the groups parameter of this function should take.

Usage

```
write.flexmirt(groups, file = NULL, fileEncoding = "")
```

Arguments

`groups`	a list of groups each with items and latent parameters
`file`	the destination file name
`fileEncoding`	how to encode the text file (optional)

Details

NOTE: Support for the graded response model may not be complete.

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