Package: TestGardener (via r-universe)

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Type Package

Title Information Analysis for Test and Rating Scale Data

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Depends R $(>= 3.5)$, fda, rgl, ggplot2, plotly

Description Develop, evaluate, and score multiple choice examinations, psychological scales, questionnaires, and similar types of data involving sequences of choices among one or more sets of answers. This version of the package should be considered as brand new. Almost all of the functions have been changed, including their argument list. See the file NEWS.Rd in the Inst folder for more information. Using the package does not require any formal statistical knowledge beyond what would be provided by a first course in statistics in a social science department. There the user would encounter the concept of probability and how it is used to model data and make decisions, and would become familiar with basic mathematical and statistical notation. Most of the output is in graphical form.

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Imports dplyr, ggpubr, stringr, tidyr, pracma, utf8, knitr, rmarkdown,

LazyData true

NeedsCompilation no

Repository https://jamesramsay5.r-universe.dev

RemoteUrl https://github.com/jamesramsay5/testgardener

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RemoteSha 321775349d843ec2a4ffbdd185d292640395915c

Contents

Description

The test or rating scale data have already been processed by function make_dataList or or other code to produce the list object dataList. The user defines a list vector ParameterList which stores results from a set of cycles of estimating surprisal curves followed by estimating optimal score index values for each examinee or respondent. These score index values are within the interval [0,100]. The number of analysis cycles is the length of the parmList list vector.

Usage

```
Analyze(index, indexQnt, dataList, NumDensBasis=7, ncycle=10, itdisp=FALSE,
        verbose=FALSE)
```
Arguments

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Details

The cycling process is described in detail in the references, and displayed in R code in the vignette SweSATQuantitativeAnalysis.

Value

The list vector parmList where each member is a named list object containing the results of an analysis cycle. These results are:

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Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[make_dataList,](#page-29-1) [TG_analysis,](#page-57-1) [index_distn,](#page-24-1) [index2info,](#page-22-1) [index_fun,](#page-25-1) [Sbinsmth](#page-40-1)

Examples

```
## Not run:
 # Example 1: Input choice data and key for the short version of the
 # SweSAT quantitative multiple choice test with 24 items and 1000 examinees
 # input the choice data as 1000 strings of length 24
 # setup the input data list object
 dataList <- Quant_13B_problem_dataList
 # define the initial examinee indices and bin locations
 index <- dataList$percntrnk
 indexQnt <- dataList$indexQnt
 # Set the number of cycles (default 10 but here 5)
 ncycle <- 5
 parmListvec <- Analyze(index, indexQnt, ncycle=ncycle, dataList,
                        verbose=TRUE)
 # two column matrix containing the mean fit and arclength values
 # for each cycle
 HALsave <- matrix(0,ncycle,2)
 for (icycle in 1:ncycle) {
   HALsave[icycle,1] <- parmListvec[[icycle]]$meanF
```

```
HALsave[icycle,2] <- parmListvec[[icycle]]$infoSurp
 }
 # plot the progress over the cycles of mean fit and arc length
 par(mfrow=c(2,1))
 plot(1:ncycle, HALsave[,1], type="b", lwd=2,
      xlab="Cycle Number",ylab="Mean H")
 plot(1:ncycle, HALsave[,2], type="b", lwd=2,
      xlab="Cycle Number", ylab="Arc Length")
## End(Not run)
```
chcemat_simulate *Simulate a test or scale data matrix.*

Description

Used in dataSimulation, this function sets up an N by n matrix of index values that specify the index of the option chosen by an examinee or respondent for a specific question.

Usage

chcemat_simulate(index.pop, SfdList)

Arguments

Details

For each question and each examinee a vector of random multinomial integer values is generated using the probability transforms of the surprisal curves and the examinee's score index value.

dataSimulation 7

Value

An N by n matrix of integer index values.

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315. s

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

Description

Estimate sum score,s score index values index and test information values bias and mean squared errors using simulated data.

Usage

```
dataSimulation(dataList, parmList, nsample = 1000)
```
Arguments

Value

A named list object containing objects produced from analyzing the simulations, one set for each simulation:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[scorePerformance](#page-45-1)

density_plot *Plot the probability density function for a set of test scores*

Description

Plots the probability density function of a set of score values that are not at the score boundaries as a smooth curve, and also plots the proportions of score values at both boundaries as points. The score values are typically either the values of the score index values index or the infoSurp or information score values.

Usage

```
density_plot(scrvec, scrrng, Qvec, xlabstr=NULL, titlestr=NULL,
                         scrnbasis=15, nfine=101)
```
Arguments

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Value

A plot of the density function and a list vector densfine containing:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[scoreDensity](#page-44-1)

Examples

```
# Example 1. Display probability density curve for the
# score index values for the short SweSAT multiple choice test with
# 24 items and 1000 examinees
index <- Quant_13B_problem_parmList$index
Qvec <- Quant_13B_problem_parmList$Qvec
# plot the density for the score indices within interval c(0,100)
oldpar <- par(no.readonly=TRUE)
on.exit(oldpar)
par(mfrow=c(2,1))
density_plot(index, c(0,100), Qvec, xlabstr="Score index",
             titlestr="SweSAT 13B Theta Density",
             scrnbasis=11, nfine=101)
# arc length or information values
scopevec <- Quant_13B_problem_infoList$scopevec
Qinfovec <- Quant_13B_problem_infoList$Qinfovec
infoSurp <- Quant_13B_problem_infoList$infoSurp
# plot the density for the score indices within interval c(0,infoSurp)
density_plot(scopevec, c(0,infoSurp), Qinfovec, xlabstr="Score index",
             titlestr="SweSAT 13B Theta Density",
             scrnbasis=11, nfine=101)
```


Description

DFfun computes the first and second derivatives of the negative log likelihoods for a set of examinees. Items can be either binary or multi-option. The analysis is within the closed interval [0,100].

Usage

DFfun(index, SfdList, chcemat)

Arguments

Value

A named list for results DF and D2F:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[make_dataList,](#page-29-1) [index_fun,](#page-25-1) [Ffun,](#page-16-1) [Ffuns_plot](#page-17-1)

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Examples

```
# Example 1:
# Compute the first and second derivative values of the objective function
# for locating each examinee for the 24-item short form of the
# SweSAT quantitative test on the percentile score index continuum.
# Use only the first five examinees.
chcemat <- Quant_13B_problem_dataList$chcemat
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
DFfunResult <- DFfun(index[1:5], SfdList, chcemat[1:5,])
DFval <- DFfunResult$DF
D2Fval <- DFfunResult$D2F
```
entropies *Entropy measures of inter-item dependency*

Description

Entropy I_1 is a scalar measure of how much information is required to predict the outcome of a choice number 1 exactly, and consequently is a measure of item effectiveness suitable for multiple choice tests and rating scales. Joint entropy $J_{1,2}$ is a scalar measure of the cross-product of multinomial vectors 1 and 2. Mutual entropy $I_{1,2} = I_1 + I_2 - J_{1,2}$ is a measure of the co-dependency of items 1 and 2, and thus the analogue of the negative log of a squared correlation R^2 . this function computes all four types of entropies for two specificed items.

Usage

entropies(index, m, n, chcemat, noption)

Arguments

Value

A named list object containing objects produced from analyzing the simulations, one set for each simulation:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[Entropy_plot](#page-11-1)

Examples

```
# Load needed objects
chcemat <- Quant_13B_problem_dataList$chcemat
index <- Quant_13B_problem_parmList$index
noption \leq matrix(5,24,1)
# compute mutual entropies for all pairs of the first 6 items
Mvec <-1:6Mlen <- length(Mvec)
Hmutual <- matrix(0, Mlen, Mlen)
for (i1 in 1:Mlen) {
  for (i2 in 1:i1) {
    Result <- entropies(index, Mvec[i1], Mvec[i2], chcemat, noption)
    Hmutual[i1,i2] = Result$Hmutual
   Hmutual[i2,i1] = Result$Hmutual
  }
}
print("Matrix of mutual entries (off-digagonal) and self-entropies (diagonal)")
print(round(Hmutual,3))
```
Entropy_plot *Plot item entropy curves for selected items or questions.*

Description

Item the value of the entropy curve at a point theta is the expected value of the surprisal curve values. Entropy is a measure of the randomness of the surprisal value, which is maximized when all the surprisal curves have the same value and has a minimum of zero if all but a single curve has probability zero. This is unattainable in the calculation, but can be arbitrarily close to this state.

Usage

```
Entropy_plot(scrfine, SfdList, Qvec, dataList, plotindex=1:n,
             plotrange=c(min(scrfine),max(scrfine)), height=1.0, value=0,
             ttlsz=NULL, axisttl=NULL, axistxt=NULL)
```
Entropy_plot 13

Arguments

Details

An entropy curve for each question indexed in the index argument. A request for a keystroke is made for each question. The answer to question strongly defines the optimal position of an estimated score index value where the curve is high value. Values of entropy curves typically range over [0,1].

Value

The plots of the entropy curves specified in plotindex are produced as a side effect. If saveplot is TRUE, the plots of item entropy curves specified in plotindex are bundled into a single postscript or .pdf file and the file name is defined by paste(dataList\$titlestr,i,'-entropy.pdf',sep=""). The file is then output as a returned value.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[Sensitivity_plot](#page-47-1), [Power_plot](#page-34-1), [Ffuns_plot](#page-17-1), [ICC_plot](#page-19-1)

Examples

```
# Example 1. Display the item entropy curves for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
# plot the entropy curve for the first item
dataList <- Quant_13B_problem_dataList
SfdList <- Quant_13B_problem_parmList$SfdList
Qvec <- Quant_13B_problem_parmList$Qvec
scrfine <- seq(0,100,len=101)
oldpar <- par(no.readonly=TRUE)
Entropy_plot(scrfine, SfdList, Qvec, dataList, plotindex=1)
par(oldpar)
```


eval.surp *Values of a Functional Data Object Defining Surprisal Curves.*

Description

A surprisal vector of length M is minus the log to a positive integer base M of a set of M multinomial probabilities. Surprisal curves are functions of a one-dimensional index set, such that at any value of the index set the values of the curves are a surprisal vector. See Details below for further explanations.

Usage

```
eval.surp(evalarg, Sfdobj, Zmat, nderiv = 0)
```
Arguments

Details

A surprisal M-vector is information measured in M-bits. Since a multinomial probability vector must sum to one, it follows that the surprisal vector S must satisfy the constraint $\log_{10}(M^(-S)) =$ 0. That is, surprisal vectors lie within a curved M-1-dimensional manifold.

Surprisal curves are defined by a set of unconstrained M-1 B-spline functional data objects defined over an index set that are transformed into surprisal curves defined over the index set.

Let C be a K by M-1 coefficient matrix defining the B-spline curves, where K is the number of Bspline basis functions.

Let a M by M-1 matrix Z have orthonormal columns. Matrices satisfying these constraints are generated by function zerobasis().

Fcurve $\qquad \qquad$ 15

Let N by K matrix be a matrix of B-spline basis values evaluated at N evaluation points using function eval.basis().

Let N by M matrix $X = B \times C \times t(Z)$.

Then the N by M matrix S of surprisal values is $S = -X + outer(log(rowSums(M^X))/log(M),rep(1,M))$.

Value

A N by M matrix S of surprisal values at points evalarg, or their first or second derivatives.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[smooth.surp](#page-51-1)

Examples

see example in man/smooth.surp.Rd

Fcurve *Construct grid of 101 values of the fitting function*

Description

A fast grid of values of the fitting function or one of its first two derivatives is constructed for use in function indexsearch.

Usage

Fcurve(SfdList, chcevec, nderiv=0)

Arguments

A vector of length 101 containing grid values of a derivative of the fitting function

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[index_search](#page-27-1)

Examples

```
# Compute a grid of values of the objective function for locating each
# examinee or respondent for the 24-item short form of the SweSAT
# quantitative test on the percentile score index continuum [0,100].
chcemat <- Quant_13B_problem_dataList$chcemat
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
n <- ncol(chcemat)
# Fitting function for the first examinee
j \leq -1chcevec <- as.numeric(chcemat[j,])
Fcurve1 <- Fcurve(SfdList, chcevec, 0)
# First derivative of the fitting function for the first examinee
DFcurve1 <- Fcurve(SfdList, chcevec, 1)
# Second derivative of the fitting function for the first examinee
D2Fcurve1 <- Fcurve(SfdList, chcevec, 2)
oldpar <- par(no.readonly=TRUE)
par(mfrow=c(3,1))
indfine \leq seq(0,100,len=101)
plot(indfine, Fcurve1, type="l", xlab="", ylab="Fitting curve",
     main="Examinee 1")
plot(indfine, DFcurve1, type="l", xlab="", ylab="First derivative")
points(index[1], 0, pch="o")
abline(0,0,lty=2)
plot(indfine, D2Fcurve1, type="l",
     xlab="Score index", ylab="Second derivative")
abline(0,0,lty=2)
points(index[1], 0, pch="o")
par(oldpar)
```


Description

Ffun computes the negative log likelihoods for a set of examinees, each at a single value index.

Usage

```
Ffun(index, SfdList, chcemat)
```
Arguments

Value

A vector of length N of negative log likelihood values.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[make_dataList,](#page-29-1) [index_fun,](#page-25-1) [Ffun,](#page-16-1) [Ffuns_plot](#page-17-1)

Examples

Example 1: Compute the values of the objective function for locating each # examinee or respondent for the 24-item short form of the SweSAT quantitative # test on the percentile score index continuum [0,100]. # Use only the first five examinees chcemat <- Quant_13B_problem_dataList\$chcemat SfdList <- Quant_13B_problem_parmList\$SfdList index <- Quant_13B_problem_parmList\$index Fval <- Ffun(index[1:5], SfdList, chcemat[1:5,])

Ffuns_plot *Plot a selection of fit criterion F functions and their first two derivatives.*

Description

These plots indicate whether an appropriate minimum of the fitting criterion was found. The value of index should be at the function minimum, the first derivative be close to zero there, and the second derivative should be positive. If these conditions are not met, it may be worthwhile to use function indexfun initialized with an approximate minimum value of score index index to re-estimate the value of index.

Usage

```
Ffuns_plot(evalarg, index, SfdList, chcemat, plotindex=1)
```
Arguments

Details

The curves are displayed in three vertically organized panels along with values of index and the values and first two derivative values of the fit criterion. If more than one index value is used, a press of the Enter or Return key moves to the next index value.

Value

A list vector is returned which is of the length of argument plotindex. Each member of the vector is a gg or ggplot object for the associated plotindex value. Each plot can be displayed using the print command. The plots of item power are produced as a side value even if no output object is specified in the call to the function.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

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See Also

[index_fun](#page-25-1), [Ffun](#page-16-1), [DFfun](#page-9-1)

Examples

```
# Example 1. Display fit criterion values and derivatives for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
chcemat <- Quant_13B_problem_dataList$chcemat
index <- Quant_13B_problem_parmList$index
SfdList <- Quant_13B_problem_parmList$SfdList
plotindex <- 1:3
indfine <- seq(0,100,len=101)
Ffuns_plot(indfine, index, SfdList, chcemat, plotindex)
```
ICC *Plotting probability and surprisal curves for an item*

Description

This is an S3 object that contains information essential plotting probability and surprisal curves for a single multiple choice or rating question. Bin probabilities and surprisal values can also be plotted.

Usage

ICC(x, M, Sfd, Zmat, Pbin, Sbin, Pmatfine, Smatfine, DSmatfine, D2Smatfine, PStdErr, SStdErr, ItemArcLen, itemStr=NULL, optStr=NULL)

Arguments

Details

The name ICC for this object is an acronym for the term "item characteristic curve" widely used in the psychometric commuunity.

Function ICC is set up after the initialization process in function make_dataList() has created the members of dataList. Within this list is object SfdList, which cintains a functional data object Sfd for each item. Both the intial coefficient matrices and the subsequent estimates of them are available from Sfd\$coefs, and therefore are available in the ICC object. These coefficient matrices are K by M-1 where K is the number of basis functions and M is the number of options for asn item.

Value

The values returned are simply those in the argument list. The S3 ICC object checks each of these and makes available the S3 commands or methods str, print and plot that apply the corresponding ICC versions of these opterations.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

ICC_plot *Plot probability and surprisal curves for test or scale items.*

Description

ICC_plot plots each item in argument plotindex in turn after functionSbinsmth() has used spline smoothing to estimate item and option characteristic curves.

ICC_plot 21

Usage

```
ICC_plot(scrfine, SfdList, dataList, Qvec,
        binctr=NULL, data_point = FALSE, ci = FALSE,
        plotType="S", Srng=c(0,5), DSrng=c(-0.2, 0.2), plotindex=1:n,
         titlestr = NULL, itemscopevec = rep(0, length(plotindex)),
        plotTitle = TRUE,autoplot = FALSE, plotMissing = TRUE,
        plotrange=c(min(scrfine),max(scrfine)), shaderange = NULL,
         ttlsz = NULL, axisttl = NULL, axistxt = NULL,
        lgdlab = NULL, lgdpos = "bottom")
```
Arguments

Value

A list vector is returned which is of the length of argument plotindex. Each member of the vector is a gg or ggplot object for the associated plotindex value. Each plot can be displayed using the print command. The plots of item power are produced as a side value even if no output object is specified in the call to the function.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[ICC](#page-18-1), [Sensitivity_plot](#page-47-1), [Power_plot](#page-34-1), [Entropy_plot](#page-11-1), [Sbinsmth](#page-40-1)

Examples

```
# Example 1. Display the item surprisal curves for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
dataList <- Quant_13B_problem_dataList
SfdList <- Quant_13B_problem_parmList$SfdList
Qvec <- Quant_13B_problem_parmList$Qvec
binctr <- Quant_13B_problem_parmList$binctr
infoSurpvec <- Quant_13B_problem_infoList$infoSurpvec
Qinfovec <- Quant_13B_problem_infoList$Qinfovec
bininfoctr <- Quant_13B_problem_infoList$bininfoctr
titlestr <- "Quant_13B_problem"
# plot the curves for the first question over the score index
oldpar <- par(no.readonly=TRUE)
indfine <- seq(0,100,len=101)
ICC_plot(indfine, SfdList, dataList, Qvec, binctr,
        data_point = TRUE, plotType = c("S", "P"),
        Srng=c(0,4), plotindex=1)
# plot the curves for the first question over test information
ICC_plot(infoSurpvec, SfdList, dataList, Qinfovec, bininfoctr,
         data_point = TRUE, plotType = c("S", "P"),
         Srng=c(0,4), plotindex=1)
par(oldpar)
```


Description

The one-dimensional psychometric model defines a space curve within the vector space defined by the total collection of option surprisal curves. This curve is a valuable resource since positions along the curve are defined in bits and positions on the curve are subject to the same strict properties that apply to physical measurements.

Function index2info is required to convert objects defined over the score index continuum c(0,100) to the same objects over the arc length continuum $c(\theta, \text{infoSurp})$, and also vice versa. Since the arc length or information continuum is along a space curve that is invariant under strictly monotone transformations of the score index index, and is also a metric, it is an ideal choice for the abscissa in all plots.

Usage

index2info(index, Qvec, SfdList, binctr, itemindex=1:n, plotrng=c(0,100), shortwrd)

Arguments

Value

A named list object containing these results of the analysis:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[Analyze](#page-2-1)

Examples

Description

Computes the cumulated density for distribution function, the probability density function, and the log probability density function as fd objects by spline smoothing of the score values indexdens using the basis object logdensbasis. The norming constant C is also output.

The score values may score index values index, expected test score values mu, or arc length locations on the test information or scale curve. The argument functional data object logdensfd should have a range that is appropriate for the score values being represented: For score indices, [0,100], for expected test scores, the range of observed or expected scores; and for test information curve locations in the interval [0,infoSurp].

Usage

```
index_distn(indexdens, logdensbasis,
              pvec=c(0.05, 0.25, 0.50, 0.75, 0.95), nfine = 101)
```
Arguments

Value

A named list containing:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[index_fun](#page-25-1), [index2info](#page-22-1), [mu](#page-32-1), [scoreDensity](#page-44-1)

Examples

```
# Example 1. Display the item power curves for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
# Assemble information for estimating index density
indfine <- seq(0,100,len=101)
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
N <- length(index)
# Define the density for only interior index values
inside \le index > 0 & index \le 100
indexdens <- index[inside]
logdensbasis <- fda::create.bspline.basis(c(0,100), 15)
index_distnList <- index_distn(index[inside], logdensbasis)
denscdf <- as.numeric(index_distnList$denscdf)
indcdf <- as.numeric(index_distnList$indcdf)
# adjusted marker score index values are computed by interpolation
markers <- c(.05, .25, .50, .75, .95)
Qvec <- pracma::interp1(denscdf, indcdf, markers)
result <- density_plot(indexdens, c(0,100), Qvec)
```
index_fun *Compute optimal scores*

Description

The percentile score index values are estimated for each person. The estimates minimize the negative log likelihoods, which are a type of surprisal. The main optimization method is a safe-guarded Newton-Raphson method.

For any iteration the method uses only those scores that are within the interior of the interval [0,100] or at a boundary with a first derivative that would take a step into the interior, and have second derivative values exceeding the value of argument crit. Consequently the number of values being optimized decrease on each iteration, and iterations cease when either all values meet the convergence criterion or are optimized on a boundary, or when the number of iterations reaches itermax.

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At that point, if there are any interior scores still associated with either non-positive second derivatives or values that exceed crit, the minimizing value along a fine mesh is used.

If itdisp is positive, the number of values to be estimated are printed for each iteration.

Usage

```
index_fun(index, SfdList, chcemat, itermax = 20, crit = 0.001,
           itdisp = FALSE)
```
Arguments

Value

A named list with these members:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[index_distn](#page-24-1), [Ffun](#page-16-1), [DFfun](#page-9-1), [index2info](#page-22-1), [scoreDensity](#page-44-1)

Examples

```
# Optimize the indices defining the data fits for the first five examinees
# input the choice indices in the 1000 by 24 choice index matrix
chcemat <- Quant_13B_problem_chcemat
# First set up the list object for surprisal curves computed from
# initial index estimates.
SfdList <- Quant_13B_problem_dataList$SfdList
# Their initial values are the percent rank values ranging over [0,100]
index_in <- Quant_13B_problem_dataList$percntrnk[1:5]
# set up choice indices for first five examinees
chemicalin < -</math> <math>chemical[1:5]</math># optimize the initial indices
indexfunList <- index_fun(index_in, SfdList, chcemat_in)
# optimal index values
index_out <- indexfunList$index_out
# The surprisal data fit values
Fval_out <- indexfunList$Fval
# The surprisal data fit first derivative values
DFval_out <- indexfunList$DFval
# The surprisal data fit second derivative values
D2Fval_out <- indexfunList$D2Fval
# The number of index values that have not reached the convergence criterion
active_out <- indexfunList$active
```


index_search *Ensure that estimated score index is global*

Description

Multiple minima are found quite often in the data fitting function that is minimized using function indexfun, and in roughly 10 percent of the estimates there is a minimum that it lower than that detected. The function searches a mesh of 101 points for minima, computes the fitting function at the minima, and assigns the location of the global minimum as the replacement index if the location differs by more than 0.5 from the value identified by index_fun. The function values and their first two derivatives are also replaced.

Usage

```
index_search(SfdList, chcemat, index, Fval, DFval, D2Fval, indexind=1:N)
```
Arguments

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Value

A named list object containing objects produced from analyzing the simulations, one set for each simulation:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[index_fun](#page-25-1)

Examples

```
# Search for values of index that are not at the global minimum of the
# fitting function and replace them as well as their function and
# derivative values associated with the fine grid value nearest the
# the global minimum.
chcemat <- Quant_13B_problem_chcemat
key <- Quant_13B_problem_key
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
Fval <- Quant_13B_problem_parmList$Fval
DFval <- Quant_13B_problem_parmList$DFval
D2Fval <- Quant_13B_problem_parmList$D2Fval
Result <- index_search(SfdList, chcemat, index, Fval, DFval, D2Fval)
changeindex <- Result$changeindex
print(paste("Number changed =",length(changeindex)))
change <- index[changeindex] - Result$index[changeindex]
```


Description

The list object dataList contains 22 objects that supply all of the information required to analyze the data. Initial values of the score indices in object theta and the bin boundaries and centres in object thetaQnt. The returned named list object contains 22 named members, which are described in the value section below.

Usage

```
make_dataList(chcemat, scoreList, noption, sumscr_rng=NULL,
              titlestr=NULL, itemlabvec=NULL, optlabList=NULL,
              nbin=nbinDefault(N), NumBasis=7, jitterwrd=TRUE,
              PcntMarkers=c( 5, 25, 50, 75, 95), verbose=FALSE)
```
Arguments

Details

The score range defined scrrng should contain all of the sum score values, but can go beyond their boundaries if desired. For example, it may be that no examinee gets a zero sum score, but for reporting and display purposes using zero as the lower limit seems desirable.

The number of bins is chosen so that a minimum of at least about 25 initial percentage ranks fall within a bin. For larger samples, the number per bin is also larger, making the proportions of choice more accurate. The number bins can be set by the user, or by a simple algorithm used to adjust the number of bins to the number N or examinees.

The number of spline basis functions used to represent a surprisal curve should be small for small sample sizes, but can be larger when larger samples are involved.

There must be at least two basis functions, corresponding to two straight lines. The norder of this simple spline would not exceed 1, corresponding to taking only a single derivative of the resulting spline. But this rule is bent here to allow higher higher derivatives, which will autmatically have values of zero, in order to allow these simple linear basis functions to be used. This permits direct comparisons of TestGardener models with the many classic item response models that use two or less parameters per item response curve.

Adding a small value to discrete values before computing ranks is considered a useful way of avoiding any biasses that might arise from the way the data are stored. The small values used leave the rounded jittered values fixed, but break up ties for sum scores.

It can be helpful to see in a plot where special marker percentages 5, 25, 50, 75 and 95 percent of the interval [0,100] are located. The median abscissa value is at 50 per cent for initial percent rank values, for example, but may not be located at the center of the interval after iterations of the analysis cycle.

Value

A named list with named members as follows:

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[TG_analysis,](#page-57-1) [Analyze,](#page-2-1) [index_distn,](#page-24-1) [index2info,](#page-22-1) [index_fun,](#page-25-1) [Sbinsmth](#page-40-1)

Examples

Example 1: Input choice data and key for the short version of the # SweSAT quantitative multiple choice test with 24 items and 1000 examinees # input the choice data as 1000 strings of length 24 # set up index and key data chcemat <- Quant_13B_problem_chcemat key <- Quant_13B_problem_key # number of examinees and of items N <- nrow(chcemat) n <- ncol(chcemat)

```
# number of options per item and option weights
noption \leq rep(0,n)
for (i in 1:n) noption[i] <-4scoreList <- list() # option scores
for (item in 1:n){
  scorei <- rep(0,noption[item])
  scorei[Quant_13B_problem_key[item]] <- 1
  scoreList[[item]] <- scorei
}
# Use the input information to define the
# big three list object containing info about the input data
dataList <- make_dataList(chcemat, scoreList, noption)
```
mu *Compute the expected test score by substituting probability of choices for indicator variable 0-1 values. Binary items assumed coded as two choice items.*

Description

Compute the expected test score by substituting probability of choices for indicator variable 0-1 values. Binary items assumed coded as two choice items.

Usage

```
mu(index, SfdList, scoreList)
```
Arguments

Value

A vector of test score values.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Siberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Siberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[scoreDensity](#page-44-1)

Examples

```
# Example 1. Compute expected sum score values for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
scoreList <- Quant_13B_problem_dataList$scoreList
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
muvec <- mu(index, SfdList, scoreList)
par(c(1,1))
hist(muvec,11)
```
mu_plot *Plot expected test score as a function of score index*

Description

The expected score mu(index) is a function of the score index index. A diagonal dashed line is displayed to show the linear relationship to the score range interval.

Usage

```
mu_plot(mufine, scrrng, titlestr)
```
Arguments

Value

A gg or ggplot object defining the plot of the expected test score mu as a function of the score index index. This is displayed by the print command. The plot is automatically displayed as a side value even if no return object is specified in the calling statement.

Author(s)

Juan Li and James Ramsay

Power_plot 35

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[scoreDensity,](#page-44-1) [mu](#page-32-1)

Power_plot *Plot item power curves for selected items or questions.*

Description

Item surprisal power curves are the square root of the sum over options of the squared surprisal sensitivity curves.

Usage

```
Power_plot(scrfine, SfdList, Qvec, dataList, plotindex=1:n,
           plotrange=c(min(scrfine),max(scrfine)), height=0.5,
           value=0, ttlsz=NULL, axisttl=NULL, axistxt=NULL)
```
Arguments

Details

A surprisal power curve for each question indexed in the index argument. A request for a keystroke is made for each question. The answer to question strongly defines the optimal position of an estimated score index value where the curve is high value. Values of power curves typically range over [0,0.5].

Value

The plots of the power curves specified in plotindex are produced as a side effect. If saveplot is TRUE, the plots of item power curves specified in plotindex are bundled into a single postscript or .pdf file and the file name is defined by paste(dataList\$titlestr,i,'-power.pdf',sep=""). The file is then outputas a returned value.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[Sensitivity_plot](#page-47-1), [Entropy_plot](#page-11-1), [Ffuns_plot](#page-17-1), [ICC_plot](#page-19-1)

Examples

```
# Example 1. Display the item power curves for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
# plot the power curve for the first item
dataList <- Quant_13B_problem_dataList
SfdList <- Quant_13B_problem_parmList$SfdList
Qvec <- Quant_13B_problem_parmList$Qvec
scrfine <- seq(0,100,len=101)
oldpar <- par(no.readonly=TRUE)
Power_plot(scrfine, SfdList, Qvec, dataList, plotindex=1)
par(oldpar)
```


Quant_13B_problem_chcemat

Test data for 24 math calculation questions from the SweSAT data.

Description

These data are for a randomly selected subset of 1000 examinees.

Usage

Quant_13B_problem_chcemat

Format

A matrix object with 1000 rows and 24 columns. The integers indicate which answer was chosen for each question by the examinee associated with the row.

Quant_13B_problem_dataList

List of objects essential for an analysis of the abbreviated SweSAT Quantitative multiple choice test.

Description

The data are for 1000 randomly selected examinees taking 24 math analysis multiple choice questions.

Usage

Quant_13B_problem_dataList

Format

A named list.

Details

A named list with 19 members:

- chcemat: A matrix of response data with N rows and n columns where N is the number of examinees or respondents and n is the number of items. Entries in the matrices are the indices of the options chosen. Column i of chcemat is expected to contain only the integers 1,...,noption.
- key: If the data are from a test of the multiple choices type where the right answer is scored 1 and the wrong answers 0, this is a numeric vector of length n containing the indices the right answers. Otherwise, it is NULL.

titlestr: A string containing a title for the analysis.

N: The number of persons tested

n: The number of questons or items

noption: A numeric vector of length n containing the numbers of options for each item.

Sdim: The total dimension of the surprisal scores.

- grbgvec: A vector of length indicating which option for each item contains missing or illegal choice values. If 0, there is no such option.
- **ScoreList:** A list vector or length n with each object a numeric vector of weights assigned to each option for each item.
- nbin: The number of bins for binning the data.

NumBasis: The number of spline basis functions.

- Sbasis: An basisfd object for the defining the surprisal curves.
- itemlabvec: A character vector with a title string for each item.
- optlabList: A list vector of length n with a character vector of labels for each object within each item.
- scrvec: A vector of length N containing the examinee or respondent sum scores.
- itmvec: A vector of length n containing the item sum scores.

scriti: A numeric vector of length N containing small jitters to each sum score to break up ties,

sumscr_rng: A vector of length 2 containing the limits of observed sum scores.

SfdList: A list vector containing essential objects for each item.

- scrfine: A fine mesh of test score values for plotting.
- **indexQnt:** A numeric vector of length $2*nbin + 1$ containing the bin boundaries alternating with the bin centers. These are initially defined as $seq(0,100,len=2*nbin+1)$.

percntrnk: A vector length N containing the sum score percentile ranks.

PcntMarkers: The marker percentages for plotting: 5, 25, 50, 75 and 95.

Quant_13B_problem_infoList

Arclength or information parameter list for 24 items from the quantitative SweSAT subtest.

Description

The data are for 1000 examinees randomly selected from those who took the 2013 quantitative subtest of the SweSAT university entrance exam. The questions are only the 24 math analysis questions, and each question has four options. The analysis results are after 10 cycles of alternating between estimating surprisal curves and estimating percentile score index values. The objects in list object Quant_13B_problem_infoList are required for plotting results over the arc length or information domain rather the score index domain. This domain is preferred because such plots are invariant with respect to changes in the score index domain. It also has a metric structure so that differences are comparable no matter where they fall within the information domain.

Usage

Quant_13B_problem_infoList

Format

A named list containing eight objects.

Value

The object Quant_13B_problem_parmList is a named list with these members:

- infoSurp: The total length of the information domain measured in M-bits, where M is the number of options for a question.
- Sfd: The log derivative functional data object defining a strictly increasing set of arc length values corresponding to set of score index values.
- infoSurpvec: A mesh of equally-spaced values of indefinite integrals of sum of norms of surprisal derivatives.
- scopevec The N arc length values corresponding to the N estimated score index values assigned to N examinees.
- Qinfovec: The arc length positions corresponding to the marker percentages 5, 25, 50, 75 and 95.
- index: A vector of score index values resulting from using function monfd with equally spaced arc length values and Sfd.info.
- Sdim: The dimension of the over space containing the surprisal pcurves.

Quant_13B_problem_key *Option information for the short form of the SweSAT Quantitative test.*

Description

A vector that contains the indices of the right answers among the options for the 24 questions

Usage

Quant_13B_problem_key

Quant_13B_problem_parmList

Parameter list for 24 items from the quantitative SweSAT subtest.

Description

The data are for 1000 examinees randomly selected from those who took the 2013 quantitative subtest of the SweSAT university entrance exam. The questions are only the 24 math analysis questions, and each question has four options. The analysis results are after 10 cycles of alternating between estimating surprisal curves and estimating percentile score index values.

Usage

Quant_13B_problem_parmList

Format

A named list.

Value

The object Quant_13B_problem_parmList is a named list with these members:

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Sbinsmth *Estimate the option probability and surprisal curves.*

Description

The surprisal curves for each item are fit to the surprisal transforms of choice probabilities for each of a set of bins of current performance values index. The error sums of squares are minimized by the surprisal optimization smooth.surp in the fda package. The output is a list vector of length n containing the functional data objects defining the curves.

Usage

```
Sbinsmth(index, dataList, indexQnt=seq(0,100, len=2*nbin+1),
        wtvec=matrix(1,n,1), iterlim=20, conv=1e-4, dbglev=0)
```
Arguments

Details

The function first bins the data in order to achieve rapid estimation of the option surprisal curves. The argument indexQnt contains the sequence of bin boundaries separated by the bin centers, so that it is of length 2*nbin + 1 where nbin is the number of bins. These bin values are distributed over the percentile interval [0,100] so that the lowest boundary is 0 and highest 100. Prior to the call to Sbinsmth these boundaries are computed so that the numbers of values of index falling in the bins are roughly equal. It is important that the number of bins be chosen so that the bins contain at least about 25 values.

After the values of index are binned, the proportions that the bins are chosen for each question and each option are computed. Proportions of zero are given NA values.

The positive proportions are then converted to surprisal values where surprisal $=$ -log M (proportion) where log_M is the logarithm with base M, the number of options associated with a question. Bins with zero proportions are assigned a surprisal that is appropriately large in the sense of being in the range of the larger surprisal values associated with small but positive proportions. This surprisal value is usually about 4.

The next step is to fit the surprisal values for each question by a functional data object that is smooth, passes as closely as possible to an option's surprisal values, and has values consistent with being a surprisal value. The function $smooth \cdot \text{supp}()$ is used for this purpose. The arc length of thme item information curve is also computed.

Finally the curves and other results for each question are saved in object SfdList, a list vector of length n, and the list vector is returned.

Value

The optimized numbered list object SfdList with length n that provides data on the probability and surprisal data and curves. The 12 objects for each item are as follows:

Sbinsmth_nom 43

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[ICC_plot](#page-19-1), [Sbinsmth](#page-40-1)

Examples

```
# Example 1. Display the initial probability and surprisal curves for the
# first item in the short SweSAT multiple choice test with 24 items and
# 1000 examinees.
# Note: The scope is 0 at this point because it is computed later
# in the analysis.
dataList <- Quant_13B_problem_dataList
index <- dataList$percntrnk
# Carry out the surprisal smoothing operation
SfdResult <- Sbinsmth(index, dataList)
 ## Not run:
 # Set up the list object for the estimated surprisal curves
 SfdList <- SfdResult$SfdList
 # The five marker percentage locations for (5, 25, 50, 75, 95)
 binctr <- dataList$binctr
 Qvec <- dataList$PcntMarkers
 # plot the curves for the first question
 scrfine <- seq(0,100,len=101)
 ICC_plot(scrfine, SfdList, dataList, Qvec, binctr,
           data_point = TRUE, plotType = c("S", "P"),
           Srng=c(0,3), plotindex=1)
```
End(Not run)

Sbinsmth_nom *List vector containing numbers of options and boundaries.*

Description

Set up objects needed for analyses of nominal data.

Usage

Sbinsmth_nom(bdry_nom, SfdList_nom)

Arguments

Details

Called twice.

Description

Arc length or scope is the distance along the space curved traced out as score index index increases from 0 to 100. It is measured in bits and is remains unchanged if the score index continuum is modified.

Usage

Scope_plot(infoSurp, infoSurpvec, titlestr=NULL)

Arguments

Value

A gg or ggplot object defining the plot of infoSurp along the test information curve as a function of the score index index. This is displayed by the print command. The plot is automatically displayed as a side value even if no return object is specified in the calling statement.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

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See Also

[index2info](#page-22-1)

Examples

```
#
# Example 1. Display the arc length curve for the
# SweSAT multiple choice test with 24 items and 1000 examinees
#
infoSurpvec <- Quant_13B_problem_infoList$infoSurpvec
infoSurp <- Quant_13B_problem_infoList$infoSurp
oldpar <- par(no.readonly=TRUE)
Scope_plot(infoSurp, infoSurpvec)
on.exit(oldpar)
```
scoreDensity *Compute and plot a score density histogram and and curve.*

Description

The tasks of function index.density() and plotting the density are combined. The score density is plotted both as a histogram and as a smooth curve. All the score types may be plotted: sum scores, expected test scores, percentile score index values, and locations on the test information or scale curve. The plot is output as a ggplot2 plot object, which is actually plotted using the print command.

Usage

```
scoreDensity(scrvec, scrrng=c(0,100), ndensbasis=15, ttlstr=NULL, pltmax=0)
```
Arguments

Value

A ggplot2 plot object dens.plot that can be displayed using command print(dens.plot).

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[index_fun](#page-25-1), [index2info](#page-22-1), [mu](#page-32-1), [index_distn](#page-24-1)

Examples

```
# Example 1. Display probability density curves for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
Qvec <- Quant_13B_problem_parmList$Qvec
# plot the density for the score indices within interval c(0,100)
index_int <- index[0 < index & index < 100]
oldpar <- par(no.readonly=TRUE)
scoreDensity(index_int)
par(oldpar)
```


Description

After the simulated data matrices have been analyzed, prepare the objects necessary for the performance plots produced by functions RMSEbias1.plot and RMSEbias2.plot.

Usage

```
scorePerformance(dataList, simList)
```
Arguments

- dataList A list that contains the objects needed to analyse the test or rating scale with the following fields:
	- chcemat: A matrix of response data with N rows and n columns where N is the number of examinees or respondents and n is the number of items. Entries in the matrices are the indices of the options chosen. Column i of chcemat is expected to contain only the integers 1,...,noption.

Value

A named list containing these objects:

- sumscr: A matrix with row dimension nchcemat, the number of population score index values and column dimension nsample, the number of simulated samples.
- chcemat: An nchcemat by nsample matrix of estimated score index values.

mu: An nchcemat by nsample matrix of estimated expected score values.

al: An nchcemat by nsample matrix of estimated test information curve values.

chcepop: A vector of population score index values.

mupop: A vector of expected scores computed from the population score index values.

infopop: A vector of test information values computed from the population score index values.

n: The number of questions.

Qvec: The five marker percentile values.

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[dataSimulation](#page-6-1)

Sensitivity_plot *Plots all the sensitivity curves for selected items or questions.*

Description

A sensitivity curve for an option is the first derivative of the corresponding surprisal curve. Its values can be positive or negative, and the size of the departure from zero at any point on the curve is the amount information contributed by that curve to locating the value of an examinee or respondent on the score index continuum.

Usage

```
Sensitivity_plot(scrfine, SfdList, Qvec, dataList, plotindex=1:n,
                 plotrange=c(min(scrfine),max(scrfine)),
                 key=NULL, titlestr=NULL, saveplot=FALSE, width=c(-0.2,0.2),
                 ttlsz=NULL, axisttl=NULL, axistxt=NULL, lgdlab=NULL)
```
Arguments

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Details

Sensitivity curves for each question indexed in the index argument. A request for a keystroke is made for each question.

Value

A list vector is returned which is of the length of argument plotindex. Each member of the vector is a gg or ggplot object for the associated plotindex value. Each plot can be displayed using the print command. The plots of item power are produced as a side value even if no output object is specified in the call to the function.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[Power_plot](#page-34-1), [Entropy_plot](#page-11-1), [ICC_plot](#page-19-1)

Examples

Example 1. Display the option sensitivity curves for the # short SweSAT multiple choice test with 24 items and 1000 examinees. dataList <- Quant_13B_problem_dataList SfdList <- Quant_13B_problem_parmList\$SfdList Qvec <- Quant_13B_problem_parmList\$Qvec scrfine <- seq(0,100,len=101) oldpar <- par(no.readonly=TRUE) Sensitivity_plot(scrfine, SfdList, Qvec, dataList, plotindex=1) par(oldpar)

Description

Simulation of data using a previous analysis requires only an ICC vector and two objects computed by function theta.distn along with a specification of the number of simulated the simulated persons.

Usage

SimulateData(nsim, indfine, denscdf, SfdList)

Arguments

Details

Arguments indfine and denscdf can be obtained from the original analysis, but also can be specified to describe a different distribution of score index values.

Value

An nsim by n matrix of integers including 1 and 2 that specify each person's option choice for each item.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[dataSimulation](#page-6-1), [chcemat_simulate](#page-5-1)

 $smooth.$ ICC 51

Examples

example code to be set up

smooth.ICC *Smooth binned probability and surprisal values to make an* ICC *object.*

Description

An N by n matrix of positive integer choice index values is transformed to an nbin by M matrix of probability values by iteravely minimizing the sum of squared errors for bin values.

Usage

smooth.ICC(x, item, index, dataList, indexQnt=seq(0,100, len=2*nbin+1), wtvec=matrix(1,n,1), iterlim=20, conv=1e-4, dbglev=0)

Arguments

Value

An S3 class ICC object for a single item.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

Examples

example code to be set up

smooth.surp *Fit data with surprisal smoothing.*

Description

Surprisal is -log(probability) where the logarithm is to the base being the dimension M of the multinomial observation vector. The surprisal curves for each question are estimated by fitting the surprisal values of binned data using curves whose values are within the M-1-dimensional surprisal subspace that is within the space of non-negative M-dimensional vectors.

Usage

smooth.surp(argvals, y, Bmat0, Sfd, Zmat, wtvec=NULL, conv=1e-4, iterlim=50, dbglev=0)

Arguments

Value

A named list of class surpFd with these members:

smooth.surp 53

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[eval.surp](#page-13-1), [ICC_plot](#page-19-1), [Sbinsmth](#page-40-1)

Examples

```
oldpar <- par(no.readonly=TRUE)
# evaluation points
x \leq -\text{seq}(-2, 2, \text{len}=11)# evaluate a standard normal distribution function
p \leq -pnorm(x)# combine with 1-p
mnormp \le - \text{cbind}(p, 1-p)M < - 2# convert to surprisal values
mnorms <- -log2(mnormp)
# plot the surprisal values
matplot(x, mnorms, type="l", lty=c(1,1), col=c(1,1),
        ylab="Surprisal (2-bits)")
# add some log-normal error
mnormdata <- exp(log(mnorms) + rnorm(11)*0.1)
# set up a b-spline basis object
nbasis <- 7
sbasis <- create.bspline.basis(c(-2,2),nbasis)
# define an initial coefficient matrix
cmat \leq matrix(0,7,1)
# set up a fd object for suprisal smoothing
Sfd <- fd(cmat, sbasis)
Zmat \leq matrix(c(1,-1),2,1)
# smooth the noisy data
result <- smooth.surp(x, mnormdata, cmat, Sfd, Zmat)
# plot the data and the fits of the two surprisal curves
```

```
xfine < -seq(-2, 2, len=51)sfine <- eval.surp(xfine, result$Sfd, Zmat)
matplot(xfine, sfine, type="l", lty=c(1,1), col=c(1,1))
points(x, mnormdata[,1])
points(x, mnormdata[,2])
# convert the surprisal fit values to probabilities
pfine <- 2^(-sfine)
# check that they sum to one
apply(pfine,1,sum)
par(oldpar)
```


Spca *Functional principal components analysis of information curve*

Description

A test or scale analysis produces a space curve that varies with in the space of possible option curves of dimension Sdim. Fortunately, it is usual that most of the shape variation in the curve is within only two or three dimensions, and these can be fixed by using functional principal components analysis.

Usage

```
Spca(SfdList, nharm=2, Sdim=NULL, rotate=TRUE)
```
Arguments

Value

A named list with these members:

Author(s)

Juan Li and James Ramsay

Spca_plot 55

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[Spca_plot](#page-54-1)

Examples

```
# Example 1. Display the test information curve for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
# plot a two-dimension version of manifold curve
Sdim <- Quant_13B_problem_dataList$Sdim
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
infoSurp <- Quant_13B_problem_parmList$infoSurp
# <- Quant_13B_problem_dataList$Sdim
oldpar <- par(no.readonly=TRUE)
on.exit(oldpar)
Results <- Spca(SfdList, nharm=2, rotate=FALSE)
varprop <- Results$varpropvarmx
print("Proportions of variance accounted for and their sum:")
print(round(100*c(varprop,sum(varprop)),1))
# plot a three-dimension version of manifold curve
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
infoSurp <- Quant_13B_problem_parmList$infoSurp
Results <- Spca(SfdList, nharm=3, rotate=FALSE)
varprop <- Results$varpropvarmx
print("Proportions of variance accounted for and their sum:")
print(round(100*c(varprop,sum(varprop)),1))
```
Spca_plot *Plot the test information or scale curve in either two or three dimensions.*

Description

A test or scale analysis produces a space curve that varies with in the space of possible option curves of dimension Sdim. Fortunately, it is usual that most of the shape variation in the curve is within only two or three dimensions, and these can be fixed by using functional principal components analysis.

Usage

```
Spca_plot(harmvarmxfd, nharm=2, titlestr=NULL)
```
Arguments

Value

Side effect is a two or three-dimensional plot of the principal component approximation of the information curve using the plotly package. Function plot_ly does not return a value, but does render the graphic.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[Spca](#page-53-1)

Examples

```
# Example 1. Display the test information curve for the
# short SweSAT multiple choice test with 24 items and 1000 examinees
# plot a two-dimension version of manifold curve
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
arclength <- Quant_13B_problem_parmList$arclength
Results <- Spca(SfdList, nharm=2, rotate=TRUE)
varprop <- Results$varpropvarmx
titlestr <- "SweSAT problem items"
oldpar <- par(no.readonly=TRUE)
on.exit(oldpar)
Spca_plot(Results$harmvarmxfd, nharm=2, titlestr)
print("Proportions of variance accounted for and their sum:")
print(round(100*c(varprop,sum(varprop)),1))
# plot a three-dimension version of manifold curve
SfdList <- Quant_13B_problem_parmList$SfdList
index <- Quant_13B_problem_parmList$index
arclength <- Quant_13B_problem_parmList$arclength
Results <- Spca(SfdList, nharm=3, rotate=TRUE)
varprop <- Results$varpropvarmx
Spca_plot(Results$harmvarmxfd, nharm=3, titlestr)
print("Proportions of variance accounted for and their sum:")
```
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print(round(100*c(varprop,sum(varprop)),1))

TestGardener *Analyses of Tests and Rating Scales using Information or Surprisal*

Description

TestGardener is designed to permit the analysis of choice data from multiple choice tests and rating scales using information as an alternative to the usual models based on probability of choice.

Probabability and information are related by the simple transformation "information = -log probability". Another term for information is "surprisal."

The advantage of information methodology, often used in the engineering and physical sciences, is that measurabe, and therefore is on what is called a "ratio scale" in the social sciences. That is, information or rurprisal has a lower limit of zero, is unbounded above, and can be added, subtracted and rescaled with a positive multiplier.

The disadvantage of probability as a basis for representing choice is that differences near its two boundaries are on very different scales than those near 0.5, and our visual and other sensory systems, which are adapted to mangitudes, have many problems in assessing the nonlinear probability continuum.

TestGardener uses highly adaptable and computationally efficient spline basis functions to represent item characteristic curves for both probability and surprisal. Splines bases permit as much flexibility as the task requires, and also can control the smoothness and the order of differentiation.

The higher variability revealed by information or surprisal curves reveals many more insights into choice behavior than the usual simple curve employed in standard probability-based item response theory.

The use of information as a measure also implies a measure of inter-item covariation called mutual entropy. Entropy a function whose value at any point is the average across surprisal curves produced by summing over curves for a given item of the product of probability and surprisal.

Graphical display is a large part of the TestGardener capacity, with extensive use of the ggplot2 and plotly packages.

TestInfo_svd *Image of the Test Tnformation Curve in 2 or 3 Dimensions*

Description

The test information curve is the trajectory of joint variation of all the surprisal curves within the ambient space of dimension the total number of curves. But usually a very high percent of the shape variation in the curve can be represented in either two or three dimensions using the singular value decomposition of a matrix of total curve values over a fine mesh. The resulting approximation is converted to a set of surprisal curve values.

```
TestInfo_svd(scrfine, SfdList, itemindex=1:n, nharm=2)
```
Arguments

Value

The approximation is returned as a surprisal functional data object, and so are the percentages of the total variation fit by each dimension in the approximation.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

TG_analysis *Statistics for Multiple choice Tests, Rating Scales and Other Choice Data)*

Description

Given an choice ingter-valued index matrix and a vector of numbers of item options, the function cycles through a set of iterations involving surprisal curve estimation followed by test taker index values.

Usage

```
TG_analysis(chcemat, scoreList, noption, sumscr_rng=NULL,
            titlestr=NULL, itemlabvec=NULL, optlabList=NULL,
           nbin=nbinDefault(N), NumBasis=7, NumDensBasis=7,
            jitterwrd=TRUE, PcntMarkers=c( 5, 25, 50, 75, 95),
            ncycle=10, itdisp=FALSE, verbose=FALSE)
```
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Arguments

Details

This function in package TestGardener processes at a mininum two objects: (1) A matrix chcemat that contains indices of choices made in a sequence of choice situations (its number columns n) by a set of persons making the choices (its number of rows N); and (2) A list vector scoreList of length n containing numerical weights or scores for each choice available with in each of n choice situations (referred to as items).

The function returns three large lists containing objects that can be used to assess: (1) the probability that a choice will be made, and (2) the quantity of information, called surprisal, that the choice made reveals about the performance or experience of the person making the choice.

Four list objects, each containing objects that are required for various displays, tables and other results:

parmList A list object containing objects useful for displaying results that involve the score index cotninuum:

- SfdList: A list object of length n, each containing objects for an item for displaying that item's surprisal curves as defined by the score index values after the analysis. See the help page for function Analyze for a description of these objects.
- Qvec: A vector containing the positions on the score index continuum of the marker percentages defined in the arguments of function make_dataList().
- binctr: A vector of length nbin containing the positions on the score index continuum of the bin centres.
- indexScore: A vector of length N containing the positions on the score index contiuum of each person.
- infoSurp: The length of the test or scale information continuum in M-bits.
- infoList A list object containing objects useful for displaying results that involve the scale information cotninuum:
	- infofine: A fine mesh of 101 values that is used to plot the scale information comntinuum.
	- scopevec: A vector of length N containing the positions on the scale information contiuum of each person.
	- Qinfovec: A vector containing the positions on the scale information continuum of the marker percentages defined in the arguments of function make_dataList().
	- infobinctr: A vector of length nbin containing the positions on the scale information continuum of the bin centres.

HALsave A ncycle by 2 matrix containing in the first column the data fit values and in the second column the infoSurp or arclength values.

analysisListvec

A list of length ncycle containing the parmList object for each cycle.

Author(s)

Juan Li and James Ramsay

References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[make_dataList,](#page-29-1) [Analyze,](#page-2-1) [index_distn,](#page-24-1) [index2info,](#page-22-1) [index_fun,](#page-25-1) [Sbinsmth](#page-40-1)

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Examples

```
## Not run:
```

```
# Example 1: Input choice data and key for the short version of the
 # SweSAT quantitative multiple choice test with 24 items and 1000 examinees
 # input the choice indices in the 1000 by 24 choice index matrix
 chcemat <- Quant_13B_problem_chcemat
 # set up the key data
 key <- Quant_13B_problem_key
 # number of examinees and of items
 N <- nrow(chcemat)
 n <- ncol(chcemat)
 # number of options per item and option weights
 noption \leq rep(0,n)
 for (i in 1:n) noption[i] \leq -4scoreList <- list() # option scores
 for (item in 1:n){
   scorei <- rep(0,noption[item])
   scorei[key[item]] <- 1
   scoreList[[item]] <- scorei
 }
 # Analyze the data and return the big three list objects:
 # dataList (input info), parmList (score index info) and
 # infoList (information or scope info)
 TGresult <- TG_analysis(chcemat, scoreList, noption,
                         NumBasis=4, ncycle=10, verbose=TRUE)
## End(Not run)
```
TG_density.fd *Compute a Probability Density Function*

Description

Like the regular S-PLUS function density, this function computes a probability density function for a sample of values of a random variable. However, in this case the density function is defined by a functional parameter object logdensfdPar along with a normalizing constant C.

The density function ϕ (indexdens) ϕ has the form p(indexdens) = C exp[W(indexdens)] where function \$W(indexdens)\$ is defined by the functional data object logdensfdPar.

Usage

```
## S3 method for class 'fd'
TG_density(indexdens, logdensfd, conv=0.0001, iterlim=20,
           active=1:nbasis, dbglev=0)
```
Arguments

indexdens a set observations, which may be one of two forms:

1. a vector of observations \$indexdens_i\$

Details

The goal of the function is provide a smooth density function estimate that approaches some target density by an amount that is controlled by the linear differential operator Lfdobj and the penalty parameter. For example, if the second derivative of $W(t)$ is penalized heavily, this will force the function to approach a straight line, which in turn will force the density function itself to be nearly normal or Gaussian. Similarly, to each textbook density function there corresponds a \$W(t)\$, and to each of these in turn their corresponds a linear differential operator that will, when apply to $W(t)\$, produce zero as a result. To plot the density function or to evaluate it, evaluate Wfdobj, exponentiate the resulting vector, and then divide by the normalizing constant C.

Value

a named list of length 4 containing:

Author(s)

Juan Li and James Ramsay

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References

Ramsay, J. O., Li J. and Wiberg, M. (2020) Full information optimal scoring. Journal of Educational and Behavioral Statistics, 45, 297-315.

Ramsay, J. O., Li J. and Wiberg, M. (2020) Better rating scale scores with information-based psychometrics. Psych, 2, 347-360.

See Also

[intensity.fd](#page-0-0)

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