# Package 'dexter'

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```
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### **Description**

Dexter provides a comprehensive solution for managing and analyzing educational test data.

### **Details**

The main features are:

- project databases providing a structure for storing data about persons, items, responses and booklets.
- methods to assess data quality using Classical test theory and plots.
- CML calibration of the extended nominal response model and interaction model.

To learn more about dexter, start with the vignettes: 'browseVignettes(package="dexter")'
Dexter uses the following global options

- 'dexter.use\_tibble' return tibbles instead of data.frames, defaults to FALSE
- 'dexter.progress' show progress bars, defaults to TRUE in interactive sessions
- 'dexter.max\_cores' set a maximum number of cores that dexter will use, defaults to the minimum of 'Sys.getenv("OMP\_THREAD\_LIMIT")' and 'getOption("Ncpus")', otherwise unlimited.

### Author(s)

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- · Ivailo Partchev

# See Also

Useful links:

- https://dexter-psychometrics.github.io/dexter/
- Report bugs at https://github.com/dexter-psychometrics/dexter/issues

4 ability

ability

Estimate abilities

# Description

Computes estimates of ability for persons or for booklet scores

# Usage

```
ability(
  dataSrc,
  parms,
  predicate = NULL,
  method = c("MLE", "EAP", "WLE"),
  prior = c("normal", "Jeffreys"),
  parms_draw = c("sample", "average"),
  mu = 0,
  sigma = 4,
  merge_within_persons = FALSE
ability_tables(
  parms,
  design = NULL,
  method = c("MLE", "EAP", "WLE"),
  prior = c("normal", "Jeffreys"),
  parms_draw = c("sample", "average"),
  mu = 0,
  sigma = 4
)
```

# Arguments

dataSrc	a connection to a dexter database, a matrix, or a data.frame with columns: person_id, item_id, item_score
parms	object produced by fit_enorm or a data.frame with columns item_id, item_score and, beta
predicate	An optional expression to subset data, if NULL all data is used
method	Maximum Likelihood (MLE), Expected A posteriori (EAP) or Weighted Likelihood (WLE)
prior	If an EAP estimate is produced one can choose a normal prior or Jeffreys prior; i.e., a prior proportional to the square root of test information.
parms_draw	When parms is Bayesian, parms_draw can be the index of the posterior sample of the item parameters that will be used for generating abilities. If parms_draw='sample' ability estimates are estimated over all draws and averaged. Rubin's rule is used

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to combine the imputation variance and sampling variance. If parms\_draw='average', the posterior mean of the item parameters is used.

mu Mean of the normal prior

sigma Standard deviation of the normal prior

merge\_within\_persons

for persons who were administered multiple booklets, whether to provide just

one ability value (TRUE) or one per booklet(FALSE)

design A data frame with columns item\_id and optionally booklet\_id. If the column

booklet\_id is not included, the score transformation table will be based on all items found in the design. If design is NULL and parms is an enorm fit object the score transformation table will be computed based on the test design that

was used to fit the items.

#### **Details**

MLE estimates of ability will produce -Inf and Inf estimates for the minimum (=0) and the maximum score on a booklet. If this is undesirable, we advise to use WLE. The WLE was proposed by Warm (1989) to reduce bias in the MLE and is also known as the Warm estimator.

#### Value

**ability** a data.frame with columns: booklet\_id, person\_id, booklet\_score, theta and optionally se (standard error)

ability\_tables a data.frame with columns: booklet\_id, booklet\_score, theta and optionally se (stan-dard error)

#### References

Warm, T. A. (1989). Weighted likelihood estimation of ability in item response theory. Psychometrika, 54(3), 427-450.

### **Examples**

6 add\_booklet

```
close_project(db)
```

add\_booklet

Add response data to a project

### **Description**

Add item response data in long or wide format.

### Usage

```
add_booklet(db, x, booklet_id, auto_add_unknown_rules = FALSE)
add_response_data(
  db,
  data,
  design = NULL,
 missing_value = "NA",
  auto_add_unknown_rules = FALSE
)
```

### **Arguments**

Х

db a connection to a dexter database, i.e. the output of start\_new\_project or open\_project

A data frame containing the responses and, optionally, person\_properties. The data.frame should have one row per respondent and the column names should correspond to the item\_id's in the rules or the names of the person\_properties. See details.

booklet\_id A (short) string identifying the test form (booklet)

auto\_add\_unknown\_rules

If FALSE (the default), an error will be generated if one or more responses do not appear in the scoring rules. If TRUE, unknown responses will be assumed

to have a score of 0 and will be added to your scoring rules

response data in normalized (long) format. Must contain columns person\_id, data

booklet\_id, item\_id and response and optionally item\_position (useful if

your data contains new booklets, see details)

design data.frame with columns booklet\_id, item\_id and optionally item\_position spec-

ifying the design of any \_new\_ booklets in your data.

missing\_value value to use for responses in missing rows in your data, see details add\_item\_properties 7

#### **Details**

It is a common practice to keep response data in tables where each row contains the responses from a single person. add\_booklet is provided to input data in that form, one booklet at a time.

If the dataframe x contains a variable named person\_id this variable will be used to identify unique persons. It is assumed that a single person will only make a single booklet once, otherwise an error will be generated.

If a person\_id is not supplied, dexter will generate unique person\_id's for each row of data.

Any column whose name has an exact match in the scoring rules inputted with function start\_new\_project will be treated as an item; any column whose name has an exact match in the person\_properties will be treated as a person property. If a name matches both a person\_property and an item\_id, the item takes precedence. Columns other than items, person properties and person\_id will be ignored.

add\_response\_data can be used to add data that is already normalized. This function takes a data.frame in long format with columns person\_id, booklet\_id, item\_id and response such as can usually be found in databases for example. For booklets that are not already known in your project, you need to specify the design via the design argument. Failure to do so will result in an error. Responses to items that should be there according to the design but which do not have a corresponding row in data will be added with missing\_value used for the response. If this missing value is not defined in your scoring rules and auto\_add\_unknown\_rules is set to FALSE, this will lead to an error message.

Note that responses are always treated as strings (in both functions), and NA values are transformed to the string "NA".

#### Value

A list with information about the recent import.

#### **Examples**

add\_item\_properties

Add item properties to a project

### **Description**

Add, change or define item properties in a dexter project

### Usage

```
add_item_properties(db, item_properties = NULL, default_values = NULL)
```

### **Arguments**

db a connection to a dexter database, e.g. the output of start\_new\_project or open\_project

item\_properties

A data frame containing a column item\_id (matching item\_id's already defined in the project) and 1 or more other columns with item properties (e.g. item\_type, subject)

default\_values a list where the names are item\_properties and the values are defaults. The defaults will be used wherever the item property is unknown.

### **Details**

When entering response data in the form of a rectangular person x item table, it is easy to provide person properties but practically impossible to provide item properties. This function provides a possibility to do so.

Note that is is not possible to add new items with this function, use touch\_rules if you want to add new items to your project.

#### Value

nothing

#### See Also

fit\_domains, profile\_plot for possible uses of item\_properties

### **Examples**

```
## Not run: \donttest{
db = start_new_project(verbAggrRules, "verbAggression.db")
head(verbAggrProperties)
add_item_properties(db, verbAggrProperties)
get_items(db)
close_project(db)
## End(Not run)
```

add\_person\_properties Add person properties to a project

### **Description**

Add, change or define person properties in a dexter project. Person properties defined here will also be automatically imported with add\_booklet

close\_project 9

### Usage

```
add_person_properties(db, person_properties = NULL, default_values = NULL)
```

### **Arguments**

db a connection to a dexter database, e.g. the output of start\_new\_project or open\_project

person\_properties

A data frame containing a column person\_id and 1 or more other columns with person properties (e.g. education\_type, birthdate)

default\_values a list where the names are person\_properties and the values are defaults. The defaults will be used wherever the person property is unknown.

#### Details

Due to limitations in the sqlite database backend that we use, the default values for a person property can only be defined once for each person\_property

#### Value

nothing

close\_project

Close a project

# Description

This is just an alias for DBI::dbDisconnect(db), included for completeness

# Usage

```
close_project(db)
```

# Arguments

db

connection to a dexter database

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coef.enorm

extract enorm item parameters

#### **Description**

extract enorm item parameters

# Usage

```
## S3 method for class 'enorm'
coef(object, hpd = 0.95, what = c("items", "var", "posterior"), ...)
```

further arguments to coef are ignored

### **Arguments**

an enorm parameters object, generated by the function fit\_enorm

hpd width of Bayesian highest posterior density interval around mean\_beta, value
must be between 0 and 1, default is 0.95

what which coefficients to return. Defaults to items (the item parameters). Can also
be var for the variance-covariance matrix (CML only) or posterior for all
draws of the item parameters (Bayes only)

#### **Details**

The parametrisation of IRT models is far from uniform and depends on the author. Dexter uses the following parametrisation for the extended Nominal Response Model (NRM):

$$P(X = a_j | \beta, \theta) = \frac{\exp(a_j \theta - \sum_{g=1}^j \beta_g (a_g - a_{g-1}))}{1 + \sum_h \exp(a_h \theta - \sum_{g=1}^h \beta_g (a_g - a_{g-1}))}$$

where  $a_i$  is a shorthand for the integer score belonging to the j-th category of an item.

For dichotomous items with  $a_1=1$  (i.e. the only possible scores are 0 and 1) this formula simplifies to the standard Rasch model:  $P(x=1|\beta,\theta)=\frac{\exp(\theta-\beta)}{1+\exp(\theta-\beta)}$ . For polytomous items, when all scores are equal to the categories (i.e.  $a_j=j$  for all j) the NRM is equal to the Partial Credit Model, although with a different parametrisation than is commonly used. For dichotomous items and for all polytomous items where  $a_j-a_{j-1}$  is constant, the formulation is equal to the OPLM.

#### Value

Depends on the calibration method and the value of 'what'. For what="items":

CML calibration a data.frame with columns: item\_id, item\_score, beta, SE\_beta
Bayesian calibration a data.frame with columns: item\_id, item\_score, mean\_beta, SD\_beta, <hpd\_b\_left>,
<hpd\_b\_right>

If what="var" or what="posterior" then a matrix is returned with the variance-covariance matrix or the posterior draws respectively.

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coef.inter

Extract interaction model parameters

### **Description**

Extract interaction model parameters

### Usage

```
## S3 method for class 'inter'
coef(object, what = c("items", "scoreprob"), ...)
```

# Arguments

object an object returned by the function fit\_inter

what which coefficients to return. Defaults to items (the item parameters), can also

be scoreprob for the probability of each item score per booklet score.

... further arguments to coef are ignored

coef.p2pass

extract equating information

### **Description**

extract equating information

# Usage

```
## S3 method for class 'p2pass'
coef(object, ...)
```

### **Arguments**

object an p2pass object, generated by probability\_to\_pass

... further arguments are currently ignored

# Value

A data.frame with columns:

```
booklet_id id of the target booklet
score_new score on the target booklet
probability_to_pass probability to pass on the reference test given score_new
true_positive proportion that correctly passes
```

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sensitivity The proportion of positives that are correctly identified as suchspecificity The proportion of negatives that are correctly identified as suchproportion proportion in sample with score\_new

design\_info

Information about the design

# Description

This function is useful to inspect incomplete designs

### Usage

```
design_info(dataSrc, predicate = NULL)
```

### **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

predicate An optional expression to subset data, if NULL all data is used

#### Value

a list with the following components

design a data.frame with columns booklet\_id, item\_id, item\_position, n\_persons

**connected\_booklets** a data.frame with columns booklet\_id, group; booklets with the same 'group' are connected to each other.

connected TRUE/FALSE indicating whether the design is connected or not

**testlets** a data.frame with columns item\_id and testlet; items within the same testlet always occur together in a booklet

adj\_matrix list of two adjacency matrices: \*weighted\_by\_items\* and \*weighted\_by\_persons\*;
These matrices can be useful in visually inspecting the design using a package like \*igraph\*

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DIF

Exploratory test for Differential Item Functioning

### **Description**

Exploratory test for Differential Item Functioning

### Usage

```
DIF(dataSrc, person_property, predicate = NULL)
```

### **Arguments**

dataSrc a connection to a dexter database or a data.frame with columns: person\_id, item\_id, item\_score

person\_property

Defines groups of persons to calculate DIF

predicate An optional expression to subset data, if NULL all data is used

#### Details

Tests for equality of relative item/category difficulties across groups. Supplements the confirmatory approach of the profile plot.

#### Value

An object of class DIF\_stats holding statistics for overall-DIF and a matrix of statistics for DIF in the relative position of item-category parameters in the beta-parameterization where they represent locations on the ability scale where adjacent categories are equally likely. If there is DIF, the function 'plot' can be used to produce an image of the pairwise DIF statistics.

#### References

Bechger, T. M. and Maris, G (2015); A Statistical Test for Differential Item Pair Functioning. Psychometrika. Vol. 80, no. 2, 317-340.

### See Also

A plot of the result is produced by the function plot.DIF\_stats

### **Examples**

```
db = start_new_project(verbAggrRules, ":memory:", person_properties=list(gender='unknown'))
add_booklet(db, verbAggrData, "agg")
dd = DIF(db,person_property="gender")
print(dd)
```

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```
plot(dd)
str(dd)
close_project(db)
```

distractor\_plot

Distractor plot

# Description

Produce a diagnostic distractor plot for an item

# Usage

```
distractor_plot(
  dataSrc,
  item_id,
  predicate = NULL,
  legend = TRUE,
  curtains = 10,
  adjust = 1,
  col = NULL,
  ...
)
```

# Arguments

dataSrc	a connection to a dexter database or a data.frame with columns: person_id, item_id, response, item_score and optionally booklet_id
item_id	The ID of the item to plot. A separate plot will be produced for each booklet that contains the item, or an error message if the item_id is not known. Each plot contains a non-parametric regression of each possible response on the total score.
predicate	An optional expression to subset data, if NULL all data is used
legend	logical, whether to include the legend. default is TRUE
curtains	100*the tail probability of the sum scores to be shaded. Default is 10. Set to 0 to have no curtains shown at all.
adjust	factor to adjust the smoothing bandwidth respective to the default value
col	vector of colors to use for plotting. The names of the vector can be responses. If the vector is not named, colors are assigned to the most frequent responses first.
	further arguments to plot.

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#### **Details**

Customization of title and subtitle can be done by using the arguments main and sub. These arguments can contain references to the variables item\_id, booklet\_id, item\_position(if available), pvalue, rit and rir. References are made by prefixing these variables with a dollar sign. Variable names may be postfixed with a sprintf style format string, e.g. distractor\_plot(db, main='item: \$item\_id', sub='Item rest correlation: \$rir:.2f')

#### Value

Silently, a data frame of response categories and colors used. Potentially useful if you want to customize the legend or print it separately

fit\_domains

Estimate the Rasch and the Interaction model per domain

### **Description**

Estimate the parameters of the Rasch model and the Interaction model

#### **Usage**

```
fit_domains(dataSrc, item_property, predicate = NULL)
```

### Arguments

dataSrc a connection to a dexter database or a data.frame with columns: person\_id,

item\_id, item\_score

item\_property The item property defining the domains (subtests)

predicate An optional expression to subset data, if NULL all data is used

# **Details**

We have generalised the interaction model for items having more than two (potentially, a largish number) of response categories. This function represents scores on subtests as super-items and analyses these as normal items.

#### Value

An object of class imp holding results for the Rasch model and the interaction model.

#### See Also

```
plot.inter, fit_inter, add_item_properties
```

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### **Examples**

```
db = start_new_project(verbAggrRules, ":memory:")
add_booklet(db, verbAggrData, "agg")
add_item_properties(db, verbAggrProperties)
mSit = fit_domains(db, item_property= "situation")
plot(mSit)
close_project(db)
```

fit\_enorm

Fit the extended nominal response model

# Description

Fits an Extended NOminal Response Model (ENORM) using conditional maximum likelihood (CML) or a Gibbs sampler for Bayesian estimation.

### Usage

```
fit_enorm(
  dataSrc,
  predicate = NULL,
  fixed_params = NULL,
  method = c("CML", "Bayes"),
  nDraws = 1000,
  merge_within_persons = FALSE
)
```

### **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

predicate An optional expression to subset data, if NULL all data is used

fixed\_params Optionally, an enorm object from a previous analysis or a data frame with pa-

rameters, see details.

method If CML, the estimation method will be Conditional Maximum Likelihood; oth-

erwise, a Gibbs sampler will be used to produce a sample from the posterior

nDraws Number of Gibbs samples when estimation method is Bayes.

merge\_within\_persons

whether to merge different booklets administered to the same person, enabling

linking over persons as well as booklets.

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#### **Details**

The eNRM is a generalization of the PCM and the OPLM. It reduces to the Rasch model for dichotomous items when all itemscores are 0 or 1, is equal to the PCM for polytomous items if all itemscores up to the maximum score occur. It is equal to the oplm if all itemscores have an equal common divisor larger than 1.

To support some flexibility in fixing parameters, fixed\_params can be a dexter enorm object or a data.frame. If it is a data.frame, it should contain the columns item\_id, item\_score and a difficulty parameter beta

### Value

An object of type enorm. The following methods are supported:

- coef
- plot
- logLik

In addition, many dexter functions accept an enorm object as input, e.g.

- ability
- plausible\_values
- plausible\_scores
- expected\_score

### References

Maris, G., Bechger, T.M. and San-Martin, E. (2015) A Gibbs sampler for the (extended) marginal Rasch model. Psychometrika. 80(4), 859-879.

Koops, J. and Bechger, T.M. and Maris, G. (2024); Bayesian inference for multistage and other incomplete designs. In Research for Practical Issues and Solutions in Computerized Multistage Testing. Routledge, London.

fit\_inter

Estimate the Interaction and the Rasch model

### Description

Estimate the parameters of the Interaction model and the Rasch model

# Usage

```
fit_inter(dataSrc, predicate = NULL)
```

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# Arguments

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

predicate An optional expression to subset data, if NULL all data is used

#### **Details**

Unlike the Rasch model, the interaction model cannot be computed concurrently for a whole design of test forms. This function therefore fits the Rasch model and the interaction model on complete data. This typically consist of responses to items in one booklet but can also consist of the intersection (common items) in two or more booklets. If the intersection is empty (no common items for all persons), the function will exit with an error message.

#### Value

An object of class inter holding results for the Rasch model and the interaction model.

#### See Also

```
plot.inter, fit_domains
```

### **Examples**

```
db = start_new_project(verbAggrRules, ":memory:")
add_booklet(db, verbAggrData, "agg")

m = fit_inter(db, booklet_id=='agg')
plot(m, "S1DoScold", show.observed=TRUE)

close_project(db)
```

get\_booklets

Booklets entered in a project

### **Description**

Retrieve information about the booklets entered in the db so far

### Usage

```
get_booklets(db)
```

get\_design 19

# Arguments

db a connection to a dexter database, i.e. the output of start\_new\_project or

open\_project

#### Value

A data frame with columns: booklet\_id, n\_persons, n\_items and booklet\_max\_score. booklet\_max\_score gives the maximum theoretically possible score according to the scoring rules

get\_design Test design

### **Description**

Retrieve all items that have been entered in the db so far by booklet and position in the booklet

#### Usage

```
get_design(
  dataSrc,
  format = c("long", "wide"),
  rows = c("booklet_id", "item_id", "item_position"),
  columns = c("item_id", "booklet_id", "item_position"),
  fill = NA
)
```

### Arguments

dataSrc a dexter database or any object form which a design can be inferred

format return format, see below

rows variable that defines the rows, ignored if format='long' columns variable that defines the columns, ignored if format='long'

fill If set, missing values will be replaced with this value, ignored if format='long'

#### Value

A data.frame with the design. The contents depend on the rows, columns and format parameters if format is 'long' a data.frame with columns: booklet\_id, item\_id, item\_position (if available) if format is 'wide' a data.frame with the rows defined by the rows parameter and the columns by the columns parameter, with the remaining variable (i.e. item\_id, booklet\_id or item\_position) making up the cells

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get\_items

Items in a project

# Description

Retrieve all items that have been entered in the db so far together with the item properties

# Usage

```
get_items(db)
```

# Arguments

db

a connection to a dexter database, e.g. the output of  $start_new_project$  or  $open_project$ 

### Value

A data frame with column item\_id and a column for each item property

get\_persons

Persons in a project

# **Description**

Retrieve all persons/respondents that have been entered in the db so far together with their properties

# Usage

```
get_persons(db)
```

### **Arguments**

db

a connection to a dexter database, e.g. the output of  $start_new_project$  or  $open_project$ 

#### Value

A data frame with columns person\_id and columns for each person\_property

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get\_responses

Selecting data

### **Description**

Extract data from a dexter database

### Usage

```
get_responses(
  dataSrc,
  predicate = NULL,
  columns = c("person_id", "item_id", "item_score")
)
```

# **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

predicate an expression to select data on

columns the columns you wish to select, can include any column in the project, see:

get\_variables

#### **Details**

Many functions in Dexter accept a data source and a predicate. Predicates are extremely flexible but they have a few limitations because they work on the individual response level. It is therefore not possible for example, to remove complete person cases from an analysis based on responses to a single item by using just a predicate expression.

For such cases, Dexter supports selecting the data and manipulating it before passing it back to a Dexter function or possibly doing something else with it. The following example will hopefully clarify this.

#### Value

a data.frame of responses

### **Examples**

```
## Not run:
# goal: fit the extended nominal response model using only persons
# without any missing responses
library(dplyr)

# the following would not work since it will omit only the missing
# responses, not the persons; which is not what we want in this case
wrong = fit_enorm(db, response != 'NA')
```

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get\_resp\_data

Functions for developers

### **Description**

These functions are meant for people who want to develop their own models based on the data management structure of dexter. The benefit is some extra speed and less memory usage compared to using get\_responses or get\_testscores. The return value of get\_resp\_data can be used as the 'dataSrc' argument in analysis functions.

### Usage

```
get_resp_data(
  dataSrc,
  qtpredicate = NULL,
  extra_columns = NULL,
  summarised = FALSE,
  env = NULL,
  protect_x = TRUE,
  retain_person_id = TRUE,
  merge_within_persons = FALSE,
  parms_check = NULL,
  raw = FALSE
)

get_resp_matrix(dataSrc, qtpredicate = NULL, env = NULL)
```

### **Arguments**

dataSrc data.frame, integer matrix, dexter database or 'dx\_resp\_data' object

qtpredicate quoted predicate, e.g. quote(booklet\_id=='bk01')

extra\_columns to be returned in addition to person\_id, booklet\_id, item\_score, item\_id

summarised if TRUE, no item scores are returned, just booklet scores

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environment for evaluation of qtpredicate, defaults to caller environment env

best set TRUE (default) protect\_x

retain\_person\_id

whether to retain the original person\_id levels or just use arbitrary integers

merge\_within\_persons

merge different booklets for the same person together

parms\_check data.frame of item\_id, item\_score to check for coverage of data

if raw is TRUE, no sum scores, booklets, or design is provided and arguments, raw

'parms\_check' and 'summarised' are ignored

#### **Details**

Regular users are advised not to use these functions as incorrect use can crash your R-session or lead to unexpected results.

### Value

get\_resp\_data returns a list with class 'dx\_resp\_data' with elements

x when summarised is FALSE, a tibble(person id, booklet id, item id, item score, booklet\_score [, extra\_columns]), sorted in such a way that all rows pertaining to the same person-booklet are together when summarised is TRUE, a tibble(person\_id, booklet\_id, booklet\_score [, extra\_columns])

design tibble(booklet\_id, item\_id), sorted

get\_resp\_matrix returns a matrix of item scores as commonly used in other IRT packages, facilitating easy connection of your own package to the data management capabilities of dexter

Get scoring rules

get\_rules

### **Description**

Retrieve the scoring rules currently present in the dexter project db

### Usage

```
get_rules(db)
```

# **Arguments**

db a connection to a Dexter database

#### Value

data.frame of scoring rules containing columns: item\_id, response, item\_score

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get\_testscores

Get test scores

### **Description**

Supplies the sum of item scores for each person selected.

### Usage

```
get_testscores(dataSrc, predicate = NULL)
```

### **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

predicate An optional expression to filter data, if NULL all data is used

#### Value

A tibble with columns person\_id, item\_id, booklet\_score

get\_variables

Variables that are defined in the project

### **Description**

Inspect the variables defined in your dexter project and their datatypes

### Usage

```
get_variables(db)
```

### **Arguments**

db

a dexter project database

### **Details**

The variables in Dexter consist of the item properties and person properties you specified and a number of reserved variables that are automatically defined like response and booklet\_id.

Variables in Dexter are most useful when used in predicate expressions. A number of functions can take a dataSrc argument and an optional predicate. Predicates are a concise and flexible way to filter data for the different psychometric functions in Dexter.

The variables can also be used to retrieve data in get\_responses

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#### Value

a data.frame with name and type of the variables defined in your dexter project

```
individual_differences
```

Test individual differences

### **Description**

Test individual differences

### Usage

```
individual_differences(dataSrc, predicate = NULL)
```

### **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

predicate An optional expression to subset data, if NULL all data are used.

#### **Details**

This function uses a score distribution to test whether there are individual differences in ability. First, it estimates ability based on the score distribution. Then, the observed distribution is compared to the one expected from the single estimated ability. The data are typically from one booklet but can also consist of the intersection (i.e., the common items) of two or more booklets. If the intersection is empty (i.e., no common items for all persons), the function will exit with an error message.

### Value

An object of type tind. Printing the object will show test results. Plotting it will produce a plot of expected and observed score frequencies. The former under the hypothesis that there are no individual differences.

### **Examples**

```
db = start_new_project(verbAggrRules, ":memory:")
add_booklet(db, verbAggrData, "agg")

dd = individual_differences(db)
print(dd)
plot(dd)

close_project(db)
```

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information

Functions of theta

### **Description**

returns information function, expected score function, score simulation function, or score distribution for a single item, an arbitrary group of items or all items

# Usage

```
information(
  parms,
  items = NULL,
 booklet_id = NULL,
 parms_draw = c("average", "sample")
expected_score(
  parms,
  items = NULL,
 booklet_id = NULL,
 parms_draw = c("average", "sample")
)
r_score(
  parms,
  items = NULL,
 booklet_id = NULL,
  parms_draw = c("average", "sample")
)
p_score(
 parms,
  items = NULL,
 booklet_id = NULL,
 parms_draw = c("average", "sample")
)
```

# Arguments

parms	object produced by fit_enorm or a data.frame with columns item_id, item_score and, depending on parametrization, a column named either beta/delta, eta or b
items	vector of one or more item_id's. If NULL and booklet_id is also NULL, all items in parms are used
booklet_id	id of a single booklet (e.g. the test information function), if items is not NULL this is ignored

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parms\_draw

when the item parameters are estimated with method "Bayes" (see: fit\_enorm), parms\_draw specifies whether to use a sample (a different item parameter draw for each output column) or the posterior mean of the item draws. Alternatively, it can be an integer specifying a specific draw. It is ignored when parms is not estimated Bayesianly.

#### Value

Each function returns a new function which accepts a vector of theta's. These return the following values:

information an equal length vector with the information estimate at each value of theta.

expected\_score an equal length vector with the expected score at each value of theta

**r\_score** a matrix with length(theta) rows and one column for each item containing simulated scores based on theta. To obtain test scores, use rowSums on this matrix

**p\_score** a matrix with length(theta) rows and one column for each possible sumscore containing the probability of the score given theta

# **Examples**

```
db = start_new_project(verbAggrRules,':memory:')
add_booklet(db,verbAggrData, "agg")
p = fit_enorm(db)
# plot information function for single item
ifun = information(p, "S1DoScold")
plot(ifun,from=-4,to=4)
# compare test information function to the population ability distribution
ifun = information(p, booklet="agg")
pv = plausible_values(db,p)
op = par(no.readonly=TRUE)
par(mar = c(5,4,2,4))
plot(ifun,from=-4,to=4, xlab='theta', ylab='test information')
par(new=TRUE)
plot(density(pv$PV1), col='green', axes=FALSE, xlab=NA, ylab=NA, main=NA)
axis(side=4)
mtext(side = 4, line = 2.5, 'population density (green)')
par(op)
close_project(db)
```

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keys\_to\_rules

Derive scoring rules from keys

### **Description**

For multiple choice items that will be scored as 0/1, derive the scoring rules from the keys to the correct responses

### Usage

```
keys_to_rules(keys, include_NA_rule = FALSE)
```

### **Arguments**

keys A data frame containing columns item\_id, noptions, and key See details. include\_NA\_rule

whether to add an option 'NA' (which is scored 0) to each item

#### **Details**

This function might be useful in setting up the scoring rules when all items are multiple-choice and scored as 0/1.

The input data frame must contain the exact id of each item, the number of options, and the key. If the keys are all integers, it will be assumed that responses are coded as 1 through noptions. If they are all letters, it is assumed that responses are coded as A,B,C,... All other cases result in an error.

### Value

A data frame that can be used as input to start\_new\_project

latent\_cor

Latent correlations

### Description

Estimates correlations between latent traits using plausible values as described in Marsman, et al. (2022). An item\_property is used to distinguish the different scales.

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### Usage

```
latent_cor(
  dataSrc,
  item_property,
  predicate = NULL,
  nDraws = 500,
  hpd = 0.95,
  use = "complete.obs"
)
```

### **Arguments**

dataSrc A connection to a dexter database or a data.frame with columns: person\_id,

item\_id, item\_score and the item\_property

item\_property The name of the item property used to define the domains. If dataSrc is a

dexter db then the item\_property must match a known item property. If dataSrc

is a data.frame, item\_property must be equal to one of its column names.

predicate An optional expression to subset data, if NULL all data is used

nDraws Number of draws for plausible values

hpd width of Bayesian highest posterior density interval around the correlations,

value must be between 0 and 1.

Use Only complete obs at this time. Respondents who don't have a score for one or

more scales are removed.

### **Details**

This function uses plausible values so results may differ slightly between calls.

#### Value

List containing a estimated correlation matrix, the corresponding standard deviations, and the lower and upper limits of the highest posterior density interval and the complete mcmc sample

#### References

Marsman, M., Bechger, T. M., & Maris, G. K. (2022). Composition algorithms for conditional distributions. In Essays on Contemporary Psychometrics (pp. 219-250). Cham: Springer International Publishing.

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open\_project

Open an existing project

# Description

Opens a database created by function start\_new\_project

### Usage

```
open_project(db_name = "dexter.db")
```

### **Arguments**

db\_name

The name of the database to be opened.

### Value

a database connection object

plausible\_scores

Draw plausible test scores

### **Description**

Draw plausible, i.e. posterior predictive sumscores on a set of items.

### Usage

```
plausible_scores(
  dataSrc,
  parms = NULL,
  predicate = NULL,
  items = NULL,
  parms_draw = c("sample", "average"),
  covariates = NULL,
  nPS = 1,
  prior_dist = c("normal", "mixture"),
  keep.observed = TRUE,
  by_item = FALSE,
  merge_within_persons = FALSE
)
```

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#### **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

parms An object returned by function fit\_enorm and containing parameter estimates.

If parms is given the function provides plausible scores conditional on the item parameters. These are considered known. If parms is NULL, Bayesian parameters

are calculated from the dataSrc

predicate an expression to filter data. If missing, the function will use all data in dataSrc

items vector of item\_id's, this specifies the itemset to generate the testscores for. If

items is NULL all items occurring in dataSrc are used.

parms\_draw when the item parameters are estimated Bayesianly (see: fit\_enorm), parms\_draw

specifies whether to use a sample(a different item parameter draw for each plausible values draw) or the posterior mean of the item draws. Alternatively, it can be an integer specifying a specific draw. Ignored when parms is not estimated

Bayesianly.

covariates name or a vector of names of the variables to group the population, used to

update the prior. A covariate must be a discrete person covariate that indicates nominal categories, e.g. gender or school If dataSrc is a data.frame, it must

contain the covariate.

nPS Number of plausible testscores to generate per person.

prior\_dist use a normal prior for the plausible values or a mixture of two normals. A

mixture is only possible when there are no covariates.

keep. observed If responses to one or more of the items have been observed, the user can choose

to keep these observations or generate new ones.

by\_item return scores per item instead of sumscores

merge\_within\_persons

If a person took multiple booklets, this indicates whether plausible scores are

generated per person (TRUE) or per booklet (FALSE)

#### **Details**

A typical use of this function is to generate plausible scores on a complete item bank when data is collected using an incomplete design

### Value

A data.frame with columns booklet\_id, person\_id, booklet\_score and nPS plausible scores named PS1...PSn.

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plausible\_values

Draw plausible values

### Description

Draws plausible values based on test scores

# Usage

```
plausible_values(
  dataSrc,
  parms = NULL,
  predicate = NULL,
  covariates = NULL,
  nPV = 1,
  parms_draw = c("sample", "average"),
  prior_dist = c("normal", "mixture"),
  merge_within_persons = FALSE
)
```

### **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

parms An object returned by function fit\_enorm containing parameter estimates or a

data.frame with columns item\_id, item\_score and, beta. If parms are provided, item parameters are considered known. If parms is NULL, they will be estimated

Bayesianly.

predicate an expression to filter data. If missing, the function will use all data in dataSrc

covariates name or a vector of names of the variables to group the populations used to

improve the prior. A covariate must be a discrete person property (e.g. not a float) that indicates nominal categories, e.g. gender or school. If dataSrc is a

data.frame, it must contain the covariate.

nPV Number of plausible values to draw per person.

parms\_draw when the item parameters are estimated with method "Bayes" (see: fit\_enorm),

parms\_draw specifies whether to use a sample (a different item parameter draw for each plausible values draw) or the posterior mean of the item draws. Alternatively, it can be an integer specifying a specific draw. It is ignored when parms

is not estimated Bayesianly.

prior\_dist use a normal prior for the plausible values or a mixture of two normals. A

mixture is only possible when there are no covariates.

merge\_within\_persons

If a person took multiple booklets, this indicates whether plausible values are

generated per person (TRUE) or per booklet (FALSE)

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#### **Details**

When the item parameters are estimated using fit\_enorm(..., method='Bayes') and parms\_draw = 'sample', the uncertainty of the item parameters estimates is taken into account when drawing multiple plausible values.

In there are covariates, the prior distribution is a hierarchical normal with equal variances across groups. When there is only one group this becomes a regular normal distribution. When there are no covariates and prior\_dist = "mixture", the prior is a mixture distribution of two normal distributions which gives a little more flexibility than a normal prior.

#### Value

A data.frame with columns booklet\_id, person\_id, booklet\_score, any covariate columns, and nPV plausible values named PV1...PVn.

#### References

Marsman, M., Maris, G., Bechger, T. M., and Glas, C.A.C. (2016) What can we learn from plausible values? Psychometrika. 2016; 81: 274-289. See also the vignette.

# Examples

```
db = start_new_project(verbAggrRules, ":memory:",
   person_properties=list(gender="<unknown>"))
add_booklet(db, verbAggrData, "agg")
add_item_properties(db, verbAggrProperties)
f=fit_enorm(db)
pv_M=plausible_values(db,f,(mode=="Do")&(gender=="Male"))
pv_F=plausible_values(db,f,(mode=="Do")&(gender=="Female"))
par(mfrow=c(1,2))
plot(ecdf(pv_M$PV1),
   main="Do: males versus females", xlab="Ability", col="red")
lines(ecdf(pv_F$PV1), col="green")
legend(-2.2,0.9, c("female", "male") ,
   lty=1, col=c('green', 'red'), bty='n', cex=.75)
pv_M=plausible_values(db,f,(mode=="Want")&(gender=="Male"))
pv_F=plausible_values(db,f,(mode=="Want")&(gender=="Female"))
plot(ecdf(pv_M$PV1),
   main="Want: males versus females", xlab=" Ability", col="red")
lines(ecdf(pv_F$PV1),col="green")
legend(-2.2,0.9, c("female", "male"),
   lty=1, col=c('green', 'red'), bty='n', cex=.75)
close_project(db)
```

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plot.DIF\_stats

plot method for pairwise DIF statistics

# Description

plot method for pairwise DIF statistics

# Usage

```
## S3 method for class 'DIF_stats'
plot(
    x,
    items = NULL,
    itemsX = items,
    itemsY = items,
    cluster = FALSE,
    alpha = 0.05,
    ...
)
```

# Arguments

X	object produced by DIF
items	character vector of item id's for a subset of the plot. Useful if you have many items. If NULL all items are plotted.
itemsX	character vector of item id's for the X axis
itemsY	character vector of item id's for the Y axis
cluster	arrange items by similarity.
alpha	significance level used to color the plot (two sided)
	further arguments to plot

### **Details**

Plotting produces an image of the matrix of pairwise DIF statistics. The statistics are standard normal deviates and colored to distinguish significant from non-significant values. If there is no DIF, a proportion alpha off the cells will be colored significant by chance alone.

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plot.enorm

Plot for the extended nominal Response model

# Description

The plot shows 'fit' by comparing the expected score based on the model (grey line) with the average scores based on the data (black line with dots) for groups of students with similar estimated ability.

# Usage

```
## S3 method for class 'enorm'
plot(
    x,
    item_id = NULL,
    dataSrc = NULL,
    predicate = NULL,
    nbins = 5,
    ci = 0.95,
    sort = c("none", "mse-desc", "mse-asc"),
    add = FALSE,
    col = "black",
    col.model = "grey80",
    ...
)
```

# Arguments

x	object produced by fit_enorm
item_id	which item to plot, if NULL, one plot for each item is made
dataSrc	data source, see details
predicate	an expression to subset data in dataSrc
nbins	number of ability groups
ci	confidence interval for the error bars, between 0 and 1. Use 0 to suppress the error bars. Default = $0.95$ for a $95\%$ confidence interval
sort	for multiple items, sort item_id by mean squared error (i.e. the mean squared distance between the data and the model prediction per plot), either ascending (best to worst) or descending (worst to best). If none (the default) the order of items is not changed
add	logical; if TRUE add to an already existing plot
col	color for the observed score average
col.model	color for the expected score based on the model
	further arguments to plot

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#### **Details**

The standard plot shows the fit against the sample on which the parameters were fitted. If dataSrc is provided, the fit is shown against the observed data in dataSrc. This may be useful for plotting the fit in different subgroups as a visual test for item level DIF. The confidence intervals denote the uncertainty about the predicted pvalues within the ability groups for the sample size in dataSrc (if not NULL) or the original data on which the model was fit.

#### Value

Silently, a data frame with observed and expected values possibly useful to create a numerical fit measure.

### **Examples**

```
db = start_new_project(verbAggrRules, ":memory:",
    person_properties=list(gender=""))

add_booklet(db, verbAggrData, "agg")

f = fit_enorm(db)

plot(f, item_id="S1DoShout")

# side by side for two different groups
# (it is also possible to show two lines in the same plot
# by specifying add=TRUE as an argument in the second plot)

par(mfrow=c(1,2))

plot(f,item_id="S1WantCurse",dataSrc=db, predicate = gender=='Male',
    main='men - $item_id')

plot(f,items="S1WantCurse",dataSrc=db, predicate = gender=='Female',
    main='women - $item_id')

close_project(db)
```

plot.inter

A plot method for the interaction model

# **Description**

Plot the item-total regressions fit by the interaction (or Rasch) model. Shows the Interaction model in a thicker (gray) line and the Rasch model in a thinner (black) line.

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# Usage

```
## S3 method for class 'inter'
plot(
    x,
    items = NULL,
    summate = TRUE,
    overlay = FALSE,
    curtains = 10,
    show.observed = TRUE,
    ...
)
```

# **Arguments**

x	An object produced by function fit_inter
items	The items to plot (item_id's). If NULL, all items will be plotted
summate	If FALSE, regressions for polytomous items will be shown for each response option separately; default is TRUE.
overlay	If TRUE and more than one item is specified, there will be two plots, one for the Rasch model and the other for the interaction model, with all items overlayed; otherwise, one plot for each item with the two models overlayed. Ignored if summate is FALSE. Default is FALSE
curtains	100*the tail probability of the sum scores to be shaded. Default is 10. Set to 0 to have no curtains shown at all.
show.observed	If TRUE, the observed proportion correct at each sum score will be shown as dots. Default is FALSE.
	Any additional plotting parameters.

# **Details**

Customization of title and subtitle can be done by using the arguments main and sub. These arguments can contain references to the variables item\_id (if overlay=FALSE) or model (if overlay=TRUE) by prefixing them with a dollar sign, e.g. plot(m, main='item: \$item\_id')

plot.p2pass	A plot method for probability_to_pass	

# Description

Plot equating information from probability\_to\_pass

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## Usage

```
## S3 method for class 'p2pass'
plot(
    x,
    what = c("all", "equating", "sens/spec", "roc"),
    booklet_id = NULL,
    ...
)
```

## **Arguments**

```
x An object produced by function probability_to_pass
what information to plot, 'equating', 'sens/spec', 'roc, or 'all'
booklet_id vector of booklet_id's to plot, if NULL all booklets are plotted
... Any additional plotting parameters; e.g., cex = 0.7.
```

probability\_to\_pass

The probability to pass on a reference test given a score on a new booklet

## **Description**

Given response data that form a connected design, compute the probability to pass on the reference set conditional on each score on one or more target tests.

## Usage

```
probability_to_pass(
  dataSrc,
  parms,
  ref_items,
  pass_fail,
  predicate = NULL,
  target_booklets = NULL,
  nDraws = 1000
)
```

# **Arguments**

dataSrc a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

parms object produced by fit\_enorm or a data.frame with columns item\_id, item\_score

and beta. If uncertainty about parameter estimation should be included in the computations, use a 'parms' object computed with 'method='Bayes' and nDraws

equal or larger than nDraws in probability\_to\_pass

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ref\_items vector with id's of items in the reference set, they must all occur in dataSrc pass\_fail pass-fail score on the reference set, the lowest score with which one passes predicate An optional expression to subset data in dataSrc, if NULL all data is used target\_booklets

The target test booklet(s). A data.frame with columns booklet\_id (if multiple booklets) and item\_id, if NULL (default) this will be derived from the dataSrc and the probability to pass will be computed for each test score for each booklet in your data.

in your dat

The function uses an Markov-Chain Monte-Carlo method to calculate the probability to pass and this is the number of Monte-Carlo samples used.

## **Details**

nDraws

Note that this function is computationally intensive and can take some time to run, especially when computing the probability to pass for multiple target booklets. Further technical details can be found in a vignette.

## Value

An object of type p2pass. Use coef() to extract the probability to pass for each booklet and score. Use plot() to plot the probabilities, sensitivity and specificity or a ROC-curve.

#### See Also

The function used to plot the results: plot.p2pass

profile\_plot

Profile plot

## **Description**

Profile plot

# Usage

```
profile_plot(
  dataSrc,
  item_property,
  covariate,
  predicate = NULL,
  model = c("IM", "RM"),
  x = NULL,
  col = NULL,
  col.diagonal = "lightgray",
  ...
)
```

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# Arguments

a connection to a dexter database or a data.frame with columns: person_id, item_id, item_score and the item_property and the covariate of interest.
The name of the item property defining the domains. The item property should have exactly two distinct values in your data
name of the person property used to create the groups. There will be one line for each distinct value.
An optional expression to filter data, if NULL all data is used
"IM" (default) or "RM" where "IM" is the interaction model and "RM" the Rasch model. The interaction model is the default as it fits the data better or at least as good as the Rasch model.
Which category of the item_property to draw on the x axis, if NULL, one is chosen automatically
vector of colors to use for plotting
color of the diagonal lines representing the testscores
further graphical arguments to plot. Graphical parameters for the legend can be postfixed with .legend $$

# **Details**

Profile plots can be used to investigate whether typically two, but possibly more, groups of respondents attain the same test score in the same way. The user must provide a meaningful classification of the items in two non-overlapping subsets such that the test score is the sum of the score on the subsets. The plot shows the expected scores on each subset of items given the test score, with diagonal lines indicating the same test score. The colored lines connect the most likely combination for each test score in each group. When applied to educational test data, the plots can be used to detect differences in the relative difficulty of (sets of) items for respondents that belong to different groups. This provides a content-driven way to investigate differential item functioning.

# **Examples**

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Profile analysis	
------------------	--

# Description

Expected and observed domain scores, conditional on the test score, per person or test score. Domains are specified as categories of items using item\_properties.

## Usage

```
profile_tables(parms, domains, item_property, design = NULL)
profiles(
  dataSrc,
  parms,
  item_property,
  predicate = NULL,
  merge_within_persons = FALSE
)
```

#### **Arguments**

parms	An object returned by fit_enorm or a data.frame of item parameters
domains	data.frame with column item_id and a column with name equal to $item\_property$
item_property	the name of the item property used to define the domains. If dataSrc is a dexter db then the item_property must match a known item property. If dataSrc is a data.frame, item_property must be equal to one of its column names. For profile_tables item_property must match a column name in domains.
design	data.frame with columns item_id and optionally booklet_id
dataSrc	a connection to a dexter database or a data.frame with columns: person_id, item_id, item_score, an arbitrarily named column containing an item property and optionally booklet_id
predicate	An optional expression to subset data in dataSrc, if NULL all data is used
merge_within_persons	
	whether to merge different booklets administered to the same person.

## **Details**

When using a unidimensional IRT Model like the extended nominal response model in dexter (see: fit\_enorm), the model is as a rule to simple to catch all the relevant dimensions in a test. Nevertheless, a simple model is quite useful in practice. Profile analysis can complement the model in this case by indicating how a test-taker, conditional on her/his test score, performs on a number of pre-specified domains, e.g. in case of a mathematics test the domains could be numbers, algebra and geometry or in case of a digital test the domains could be animated versus non-animated items. This can be done by comparing the achieved score on a domain with the expected score, given the test score.

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#### Value

**profiles** a data.frame with columns person\_id, booklet\_id, booklet\_score, <item\_property>, domain\_score, expected\_domain\_score

profile\_tables a data.frame with columns booklet\_id, booklet\_score, <item\_property>, expected\_domain\_score

## References

Verhelst, N. D. (2012). Profile analysis: a closer look at the PISA 2000 reading data. Scandinavian Journal of Educational Research, 56 (3), 315-332.

ratedData

Rated data

# **Description**

A data set with rated data. A number of student performances are rated twice on several aspects by independent judges. The ratings are binary and have been summed following the theory discussed by Maris and Bechger (2006, Handbook of Statistics). Data are a small subset of data collected on the State Exam Dutch as a second language for Speaking.

#### **Format**

A data set with 75 rows and 15 columns.

ratedDataProperties

Item properties in the rated data

## **Description**

A data set of item properties related to the rated data. These are the aspects: IH = content, WZ = word choice and phrasing, and WK = vocabulary.

## **Format**

A data set with 14 rows and 2 columns: item id and aspect

ratedDataRules

Scoring rules for the rated data

# Description

A set of (trivial) scoring rules for the rated data set

## **Format**

A data set with 42 rows and 3 columns (item\_id, response, item\_score).

read\_oplm\_par 43

read\_oplm\_par

Read item parameters from oplm PAR or CML files

# **Description**

Read item parameters from oplm PAR or CML files

# Usage

```
read_oplm_par(par_path)
```

# Arguments

par\_path

path to a file in the (binary) OPLM PAR format or the human readable CML format

#### **Details**

It is very occasionally useful to calibrate new items on an existing scale. This function offers the possibility to read parameters from the proprietary oplm format so that they can be used to fix a new calibration in Dexter on an existing scale of items that were calibrated in oplm.

# Value

depends on the input. For .PAR files a data.frame with columns: item\_id, item\_score, beta, nbr, for .CML files also several statistics columns that are outputted by OPLM as part of the calibration.

# **Examples**

```
## Not run:
\donttest{
par = read_oplm_par('/parameters.PAR')
f = fit_enorm(db, fixed_params=par)
}
## End(Not run)
```

r\_score\_IM

Simulation from the interaction model

## **Description**

Simulate item scores conditional on test scores using the interaction model

## Usage

```
r_score_IM(m, scores)
```

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# Arguments

m	an object produced by function fit_inter
scores	vector of test scores

#### Value

a matrix with item scores, one column per item and one row per test score. Row order equal to scores

standards\_3dc Standard setting

# **Description**

Set performance standards on one or more test forms using the data driven direct consensus (3DC) method

# Usage

```
standards_3dc(parms, design)
## S3 method for class 'sts_par'
coef(object, ...)
## S3 method for class 'sts_par'
plot(x, booklet_id = NULL, ...)
```

## **Arguments**

parms	parameters object returned from fit_enorm
design	a data.frame with columns 'cluster_id', 'item_id' and optionally 'booklet_id'
object	an object containing parameters for the 3DC standard setting procedure
	ignored Optionally you can include a column 'booklet_id' to specify multiple test forms for standard setting and/or columns 'cluster_nbr' and 'item_nbr' to specify ordering of clusters and items in the forms and application.
X	an object containing parameters for the 3DC standard setting procedure
booklet_id	which test form to plot

## **Details**

The data driven direct consensus (3DC) method of standard setting was invented by Gunter Maris and described in Keuning et. al. (2017). To easily apply this procedure, we advise to use the free digital 3DC application. This application can be downloaded from the Cito website, see the 3DC application download page. If you want to apply the 3DC method using paper forms instead, you can use the plot method to generate the forms from the sts\_par object.

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Although the 3DC method is used as explained in Keuning et. al., the method we use for computing the forms is a simple maximum likelihood scaling from an IRT model, described in Moe and Verhelst (2017)

#### Value

```
an object of type 'sts_par'
```

#### References

Keuning J., Straat J.H., Feskens R.C.W. (2017) The Data-Driven Direct Consensus (3DC) Procedure: A New Approach to Standard Setting. In: Blomeke S., Gustafsson JE. (eds) Standard Setting in Education. Methodology of Educational Measurement and Assessment. Springer, Cham

Moe E., Verhelst N. (2017) Setting Standards for Multistage Tests of Norwegian for Adult Immigrants In: Blomeke S., Gustafsson JE. (eds) Standard Setting in Education. Methodology of Educational Measurement and Assessment. Springer, Cham

#### See Also

how to make a database for the 3DC standard setting application: standards\_db

## **Examples**

```
library(dplyr)
db = start_new_project(verbAggrRules, ":memory:")
add_booklet(db, verbAggrData, "agg")
add_item_properties(db, verbAggrProperties)

design = get_items(db) |>
    rename(cluster_id='behavior')

f = fit_enorm(db)

sts_par = standards_3dc(f, design)

plot(sts_par)

# db_sts = standards_db(sts_par,'test.db',c('mildly aggressive','dangerously aggressive'))
```

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standards\_db

Export a standard setting database for use by the free 3DC application

# **Description**

This function creates an export (an sqlite database file) which can be used by the 3DC application. This is a free application with which a standard setting session can be facilitated through a LAN network using the Chrome browser. The 3DC application can be downloaded from 3DC application download page

# Usage

```
standards_db(
  par.sts,
  file_name,
  standards,
  population = NULL,
  group_leader = "admin"
)
```

## **Arguments**

par.sts	an object containing parameters for the 3DC standard setting procedure produced by standards_3dc
file_name	name of the exported database file
standards	vector of 1 or more standards. In case there are multiple test forms and they should use different performance standards, a list of such vectors. The names of this list should correspond to the names of the testforms
population	optional, a data.frame with three columns: 'booklet_id','booklet_score','n' (where n is a count)
group_leader	login name of the group leader. The login password will always be 'admin' but can be changed in the 3DC application

start\_new\_project Start a new project

# **Description**

Imports a complete set of scoring rules and starts a new project (database)

# Usage

```
start_new_project(rules, db_name = "dexter.db", person_properties = NULL)
```

#### **Arguments**

rules A data frame with columns item\_id, response, and item\_score. The order is

not important but spelling is. Any other columns will be ignored.

db\_name A string specifying a filename for a new sqlite database to be created. If this

name does not contain a path, the file will be created in the work directory. Any existing file with the same name will be overwritten. For an in-memory database you can use the string ":memory:". A connection object is also allowed.

person\_properties

An optional list of person properties. Names should correspond to person\_properties intended to be used in the project. Values are used as default (missing) values. The datatype will also be inferred from the values. Known person\_properties will be automatically imported when adding response data with add\_booklet.

## **Details**

This package only works with closed items (e.g. likert, MC or possibly short answer) it does not score any open items. The first step to creating a project is to import an exhaustive list of all items and all admissible responses, along with the score that any of the latter will be given. Responses may be integers or strings but they will always be treated as strings. Scores must be integers, and the minimum score for an item must be 0. When inputting data, all responses not specified in the rules can optionally be treated as missing and ultimately scored 0, but it is good style to include the missing responses in the list. NA values will be treated as the string "NA".

#### Value

a database connection object.

## **Examples**

```
start_new_project_from_oplm

Start a new project from oplm files
```

#### **Description**

Creates a dexter project database and fills it with response data based on a .dat and .scr file

## Usage

```
start_new_project_from_oplm(
  dbname,
  scr_path,
  dat_path,
  booklet_position = NULL,
  responses_start = NULL,
  response_length = 1,
  person_id = NULL,
  missing_character = c(" ", "9"),
  use_discrim = FALSE,
  format = "compressed"
)
```

#### **Arguments**

dbname filename/path of new dexter project database (will be overwritten if already ex-

ists)

scr\_path path to the .scr file dat\_path path to the .dat file

booklet\_position

vector of start and end of booklet position in the dat file, e.g. c(1,4), all positions are counted from 1, start and end are both inclusive. If NULL, this is read from

the scr file.

responses\_start

start position of responses in the .dat file. If NULL, this is read from the scr file.

response\_length

length of individual responses, default=1

person\_id optionally, a vector of start and end position of person\_id in the .dat file. If

NULL, person id's will be auto-generated.

missing\_character

vector of character(s) used to indicate missing in .dat file, default is to use both

a space and a 9 as missing characters.

use\_discrim if TRUE, the scores for the responses will be multiplied by the discrimination

parameters of the items

format not used, at the moment only the compressed format is supported.

#### **Details**

start\_new\_project\_from\_oplm builds a complete dexter database from a .dat and .scr file in the proprietary oplm format. Four custom variables are added to the database: booklet\_on\_off, oplm\_marginal, item\_local\_on\_off, item\_global\_on\_off. These are taken from the .scr file and can be used in predicates in the various dexter functions.

Booklet\_position and responses\_start are usually inferred from the scr file but since they are sometimes misspecified in the scr file they can be overridden. Response\_length is not inferred from the scr file since anything other than 1 is most often a mistake.

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## Value

a database connection object.

## **Examples**

```
## Not run: \donttest{
db = start_new_project_from_oplm('test.db',
    'path_to_scr_file', 'path_to_dat_file',
    booklet_position=c(1,3), responses_start=101,
    person_id=c(50,62))

prms = fit_enorm(db,
    item_global_on_off==1 & item_local_on_off==1 & booklet_on_off==1)
}
## End(Not run)
```

tia\_tables

Simple test-item analysis

# **Description**

Show simple Classical Test Analysis statistics at item and test level

# Usage

```
tia_tables(
  dataSrc,
  predicate = NULL,
  type = c("raw", "averaged", "compared"),
  max_scores = c("observed", "theoretical"),
  distractor = FALSE
)
```

## **Arguments**

dataSrc

a connection to a dexter database, a matrix, or a data.frame with columns: per-

son\_id, item\_id, item\_score

predicate

An optional expression to subset data, if NULL all data is used

type

How to present the item level statistics: raw for each test booklet separately, averaged booklets are ignored, with the exception of rit and rir which are averaged over the test booklets, with the number of persons as weights, or compared, in which case the pvalues, correlations with the sum score (rit), and correlations with the rest score (rit) are shown in separate tables and compared across book-

lets

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max\_scores use the observed maximum item score or the theoretical maximum item score

according to the scoring rules in the database to determine pvalues and maxi-

mum scores

distractor add a tia for distractors, only useful for selected response (MC) items

#### Value

## A list containing:

booklets a data.frame of statistics at booklet level

items a data.frame (or list if type='compared') of statistics at item level

distractors a data frame of statistics at the response level (if distractor==TRUE), i.e. rvalue

(pvalue for response) and rar (rest-alternative correlation)

touch\_rules Add or modify scoring rules

# **Description**

It is occasionally necessary to alter or add a scoring rule, e.g. in case of a key error. This function offers the possibility to do so and also allows you to add new items to your project

## Usage

touch\_rules(db, rules)

# **Arguments**

db a connection to a dexter project database

rules A data frame with columns item\_id, response, and item\_score. The order is

not important but spelling is. Any other columns will be ignored. See details

#### **Details**

The rules should contain all rules that you want to change or add. This means that in case of a key error in a single multiple choice question, you typically have to change two rules.

## Value

If the scoring rules pass a sanity check, a small summary of changes is printed and nothing is returned. Otherwise this function returns a data frame listing the problems found, with 4 columns:

item\_id id of the problematic item

less\_than\_two\_scores if TRUE, the item has only one distinct score

duplicated\_responses if TRUE, the item contains two or more identical response categories

min\_score\_not\_zero if TRUE, the minimum score of the item was not 0

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## **Examples**

```
## Not run: \donttest{
# given that in your dexter project there is an mc item with id 'itm_01',
# which currently has key 'A' but you want to change it to 'C'.

new_rules = data.frame(item_id='itm_01', response=c('A','C'), item_score=c(0,1))
touch_rules(db, new_rules)
}
## End(Not run)
```

verbAggrData

Verbal aggression data

## **Description**

A data set of self-reported verbal behaviour in different frustrating situations (Vansteelandt, 2000). The dataset also contains participants reported gender and scores on the 'anger' questionnaire.

#### **Format**

A data set with 316 rows and 26 columns.

verbAggrProperties

Item properties in the verbal aggression data

# **Description**

A data set of item properties related to the verbal aggression data

# **Format**

A data set with 24 rows and 5 columns.

verbAggrRules

Scoring rules for the verbal aggression data

# **Description**

A set of (trivial) scoring rules for the verbal aggression data set

## **Format**

A data set with 72 rows and 3 columns (item\_id, response, item\_score).

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