# Package 'stuart'

August 22, 2022

August 22, 2022
Type Package
Title Subtests Using Algorithmic Rummaging Techniques
<b>Version</b> 0.10.1
<b>Date</b> 2022-08-16
<b>Description</b> Construct subtests from a pool of items by using ant-colony-optimization, genetic algorithms, brute force, or random sampling. Schultze (2017) <doi:10.17169 refubium-622="">.</doi:10.17169>
License GPL-3
<b>Depends</b> R (>= $4.1.0$ )
Imports stats, utils
<b>Suggests</b> parallel, lavaan (>= 0.5.18), MplusAutomation (>= 0.7-2), graphics, sn
RoxygenNote 7.2.1
Encoding UTF-8
LazyData true
NeedsCompilation no
Author Martin Schultze [aut, cre], Johanna Schüller [ctb]
Maintainer Martin Schultze <schultze@psych.uni-frankfurt.de></schultze@psych.uni-frankfurt.de>
Repository CRAN
<b>Date/Publication</b> 2022-08-22 16:50:02 UTC
R topics documented:
stuart-package as.stuartFixedObjective bruteforce combinations crossvalidate empiricalobjective  10

2 stuart-package

stuai	rt-package	STUART:	Subtest	s Using A	lgorithmic Run	ımaging Techniques	
Index							39
	sups						38
	_						
	mmas						25
	kfold						23
	holdout						22
	heuristics						20
	gene						13
	fixedobjective						12
	fairplayer						12

#### **Description**

The STUART-Package automates the generation of subtests from a given set of items within the confines of confirmatory factor analysis.

### **Functionality**

Using this package subtests can be generated in four different ways: using a pseudo-random approach rooted in Ant-Colony-Optimization via the mmas-function, using a simple genetic algorithm via the gene-function, using a brute-force approach via the aptly named bruteforce-function, or by random chance, using the randomsamples-function.

Additionally, there are some convenience functions which are more or less useful. The combinations-function can be used to determine the number of possible subtests to inform a decision on which selection approach to use. The crossvalidate-function can be used to evaluate the quality of a selection in a different (sub-)sample. To add to this functionality, the holdout-function randomly splits the data into a calibration and a validation sample. The entire process can be applied to k samples with kfold. The heuristics-function can be used to extract the formatting of heuristic matrices which can be provided to the mmas-function.

As of version 0.10.0 this package also includes some convenience functions for handling objective functions. These are mainly fixedobjective - to generate a fixed objective function containing any number of a variety of possible quality criteria - and empiricalobjective - to generate an adaptive objective function based on the quality determined for previous solution on any such criteria. In addition objectivematrices provides functionality to extract matrices for situations in which, for example, latent correlations or regressions predicting distal outcomes are included into the selection procedure.

The package also provides three datasets to try things out with: fairplayer, sia, and sups.

as.stuartFixedObjective 3

### Author(s)

Maintainer: Martin Schultze <schultze@psych.uni-frankfurt.de>

Other contributors:

• Johanna Schüller [contributor]

as.stuartFixedObjective

Convert empirical to fixed objective.

### Description

Convert an empirical objective to a fixed version to be used in item-selection. Sensible for extracting values from random selections and then using them in empirical but static objective functions.

### Usage

```
as.stuartFixedObjective(x)
```

### Arguments

x An object of class stuartEmpiricalObjective.

#### Value

Returns an object of class stuartFixedObjective

#### Author(s)

Martin Schultze

#### See Also

```
empiricalobjective, fixedobjective
```

4 bruteforce

bruteforce

Subtest construction using a brute-force approach

### **Description**

Construct subtests from a given pool of items using a brute-force approach (i.e. by estimating all possible combinations).

### Usage

```
bruteforce(
  data,
  factor.structure,
  capacity = NULL,
  item.invariance = "congeneric",
  repeated.measures = NULL,
  long.invariance = "strict",
 mtmm = NULL,
 mtmm.invariance = "configural",
  grouping = NULL,
  group.invariance = "strict",
  comparisons = NULL,
  auxiliary = NULL,
  use.order = FALSE,
  software = "lavaan",
  cores = NULL,
  objective = NULL,
  ignore.errors = FALSE,
  analysis.options = NULL,
  suppress.model = FALSE,
  request.override = 10000,
  filename = NULL
)
```

#### **Arguments**

data

A data.frame containing all relevant data.

factor.structure

A list linking factors to items. The names of the list elements correspond to the factor names. Each list element must contain a character-vector of item names that are indicators of this factor.

capacity

A list containing the number of items per subtest. This must be in the same order as the factor.structure provided. If a single number, it is applied to all subtests. If NULL all items are evenly distributed among the subtests.

item.invariance

A character vector of length 1 or the same length as factor. structure containing the desired invariance levels between items pertaining to the same subtest.

5 bruteforce

> Currently there are five options: 'congeneric', 'ess.equivalent', 'ess.parallel', 'equivalent', and 'parallel', the first being the default.

repeated.measures

A list linking factors that are repeated measures of each other. Repeated factors must be in one element of the list - other sets of factors in other elements of the list. When this is NULL (the default) a cross-sectional model is estimated.

long.invariance

A character vector of length 1 or the same length as repeated.measures containing the longitudinal invariance level of repeated items. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When repeated.measures=NULL this argument is ignored.

mtmm

A list linking factors that are measurements of the same construct with different methods. Measurements of the same construct must be in one element of the list - other sets of methods in other elements of the list. When this is NULL (the default) a single method model is estimated.

mtmm.invariance

A character vector of length 1 or the same length as mtmm containing the invariance level of MTMM items. Currently there are five options: 'none', 'configural', 'weak', 'strong', and 'strict'. Defaults to 'configural'. With 'none' differing items are allowed for different methods. When mtmm=NULL this argument is ignored.

The name of the grouping variable. The grouping variable must be part of data provided and must be a numeric variable.

group.invariance

A single value describing the assumed invariance of items across groups. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When grouping=NULL this argument is ignored.

A character vector containing any combination of 'item', 'long', 'mtmm', and 'group' indicating which invariance should be assessed via model comparisons. The order of the vector dictates the sequence in which model comparisons are performed. Defaults to NULL meaning that no model comparisons are performed.

The names of auxiliary variables in data. These can be used in additional modeling steps that may be provided in analysis.options\$model.

A logical indicating whether or not to take the selection order of the items into account. Defaults to FALSE.

The name of the estimation software. Can currently be 'lavaan' (the default), 'Mplus', or 'Mplus Demo'. Each option requires the software to be installed.

The number of cores to be used in parallel processing. If NULL (the default) the result of detectCores will be used. On Unix-y machines parallel processing is implemented via mclapply, on Windows machines it is realized via parLapply.

A function that converts the results of model estimation into a pheromone. See mmas for details.

A logical indicating whether or not to ignore estimation problems (such as non positive-definite latent covariance matrices). Defaults to FALSE.

grouping

comparisons

use.order

auxiliary

software

cores

objective

ignore.errors

6 bruteforce

analysis.options

A list additional arguments to be passed to the estimation software. The names of list elements must correspond to the arguments changed in the respective estimation software. E.g. analysis.options\$model can contain additional modeling commands - such as regressions on auxiliary variables.

suppress.model A logical indicating whether to suppress the default model generation. If TRUE a model must be provided in analysis.options\$model.

request.override

The maximum number of combinations for which the estimation is performed immediately, without an additional override request.

filename

The stem of the filenames used to save inputs, outputs, and data files when software='Mplus'. This may include the file path. When NULL (the default) files will be saved to the temporary directory, which is deleted when the R session is ended.

#### **Details**

The pheromone function provided via objective is used to assess the quality of the solutions. These functions can contain any combination of the fit indices provided by the estimation software. When using Mplus these fit indices are 'rmsea', 'srmr', 'cfi', 'tli', 'chisq' (with 'df' and 'pvalue'), 'aic', 'bic', and 'abic'. With lavaan any fit index provided by inspect can be used. Additionally 'crel' provides an aggregate of composite reliabilites, 'rel' provides a vector or a list of reliability coefficients for the latent variables, 'con' provides an aggregate consistency estimate for MTMM analyses, and 'lvcor' provides a list of the latent variable correlation matrices. For more detailed objective functions 'lambda', 'theta', 'psi', 'alpha', and 'nu' provide the model-implied matrices. Per default a pheromone function using 'crel', 'rmsea', and 'srmr' is used. Please be aware that the objective must be a function with the required fit indices as (correctly named) arguments.

Using model comparisons via the comparisons argument compares the target model to a model with one less degree of assumed invariance (e.g. if your target model contains strong invariance, the comparison model contain weak invariance). Adding comparisons will change the preset for the objective function to include model differences. With comparisons, a custom objective function (the recommended approach) can also include all model fit indices with a preceding delta. to indicate the difference in this index between the two models. If more than one type of comparison is used, the argument of the objective function should end in the type of comparison requested (e.g. delta.cfi.group to use the difference in CFI between the model comparison of invariance across groups).

#### Value

Returns an object of the class stuartOutput for which specific summary and plot methods are available. The results are a list.

call The called function.

software used to fit the CFA models.

parameters A list of the ACO parameters used.

analysis.options

A list of the additional arguments passed to the estimation software.

combinations 7

timer An object of the class proc\_time which contains the time used for the analysis.

log A data. frame containing the estimation history.

log\_mat A list of matrices (e.g. lvcor) relevant to the estimation history, if any.

solution NULL pheromones NULL

subtests A list containing the names of the selected items and their respective subtests.

final The results of the estimation of the global-best solution.

### Author(s)

Martin Schultze

#### See Also

```
mmas, gene, randomsamples, combinations
```

### **Examples**

```
# Bruteforce selection in a minimal example
# selecting 3 of 5 items
# requires lavaan
data(fairplayer)
fs <- list(ra = names(fairplayer)[53:57])
sel <- bruteforce(fairplayer, fs, 3,
   cores = 1) # number of cores set to 1
summary(sel) # Fit is perfect because of just-identified model</pre>
```

combinations

Compute the number of possible subtest combinations

#### **Description**

Used to compute the number of possible subtest constellations prior to performing item selection.

### Usage

```
combinations(
  data,
  factor.structure,
  capacity = NULL,
  repeated.measures = NULL,
  mtmm = NULL,
  use.order = FALSE,
  ...
)
```

8 combinations

### Arguments

data A data.frame containing all relevant data.

factor.structure

A list linking factors to items. The names of the list elements correspond to the factor names. Each list element must contain a character-vector of item names that are indicators of this factor.

capacity A list containing the num

A list containing the number of items per subtest. This must be in the same order as the factor.structure provided. If a single number, it is applied to all

subtests. If NULL all items are evenly distributed among the subtests.

repeated.measures

A list linking factors that are repeated measures of each other. Repeated factors must be in one element of the list - other sets of factors in other elements of the list. When this is NULL (the default) a cross-sectional model is estimated.

mtmm A list linking factors that are measurements of the same construct with different

methods. Measurements of the same construct must be in one element of the list - other sets of methods in other elements of the list. When this is NULL (the

default) a single method model is estimated.

use.order A logical indicating whether or not to take the selection order of the items into

account. Defaults to FALSE.

. . . Other arguments normally provided to mmas, which will be ignored.

#### Value

Returns the number of possible subtest constellations.

#### Author(s)

Martin Schultze

### See Also

```
bruteforce, mmas, gene
```

#### **Examples**

```
# Determine number of combinations in a simple situation
data(fairplayer)
fs <- list(si = names(fairplayer)[83:92])
combinations(fairplayer, fs, 4)

# Number of combinations with repeated measures
data(fairplayer)
fs <- list(si1 = names(fairplayer)[83:92],
    si2 = names(fairplayer)[93:102],
    si3 = names(fairplayer)[103:112])
repe <- list(si = c('si1', 'si2', 'si3'))
combinations(fairplayer, fs, 4, repeated.measures = repe)</pre>
```

crossvalidate 9

crossvalidate	Cross-Validate a Measurement Model
CIOSSVATIANCE	Cross randate a measurement model

#### **Description**

Cross-validate a measurement model obtained from STUART.

### Usage

```
crossvalidate(
  selection,
  old.data,
  new.data,
  max.invariance = "strict",
  filename = NULL
)
```

### Arguments

selection An object of class stuartOutput.

old.data A data.frame of the calibration sample.

new.data A data.frame of the validation sample.

max.invariance The maximum measurement invariance level which will be tested. Currently

there are four options: 'configural', 'weak', 'strong', and 'strict' (the default).

All levels below max.invariance are also tested.

filename The stem of the filenames used to save inputs, outputs, and data files when

software='Mplus'. This may include the file path. When NULL (the default) files will be saved to the temporary directory, which is deleted when the R ses-

sion is ended.

#### Value

Returns a list containing the data.frame comparison and an object containing the model results of the four different invariance assumptions.

comparison A data.frame with 4 observations, each observation representing a level of

measurement invariance. The number of columns depends on the arguments of the objective used in the original selection. In addition to those columns, three additional columns with the (corrected) Likelihood-Ratio-Tests are reported.

models A list of the four model results either of class lavaan or mplus.model, depend-

ing on the software-setting of the original selection.

#### Author(s)

Martin Schultze

10 empiricalobjective

#### See Also

holdout, mmas, bruteforce

#### **Examples**

```
# Split data into two halves
data(fairplayer)
half1 <- fairplayer[1:72,]
half2 <- fairplayer[73:143,]</pre>
# Simple example from bruteforce
fs <- list(ra = names(fairplayer)[53:57])</pre>
sel <- bruteforce(half1, fs, 3,
  cores = 1) # number of cores set to 1
# Validation
crossvalidate(sel, half1, half2)
# Using the 'holdout' function for data split
data(fairplayer)
split <- holdout(fairplayer, seed = 55635)</pre>
# Simple example from bruteforce
fs <- list(ra = names(fairplayer)[53:57])</pre>
sel <- bruteforce(split, fs, 3,
  cores = 1) # number of cores set to 1
# Validation
crossvalidate(sel, split)
```

empiricalobjective

Generate an empirical objective function for item selection.

### **Description**

Generate an empirical objective function from default or empirical values for use in an item selection using STUART.

### Usage

```
empiricalobjective(
  criteria = c("rmsea", "srmr", "crel"),
  add = c("chisq", "df", "pvalue"),
  x = NULL,
  n = 50,
  side = NULL,
  skew = FALSE,
```

empiricalobjective 11

```
scale = 1,
matrices = NULL,
fixed = NULL,
comparisons = NULL,
...
)
```

### Arguments

criteria	A vector of names of criteria included in the objective function. Defaults to $c(\text{'rmsea'}, \text{'srmr'}, \text{'crel'}).$
add	A vector of names of criteria not used in the objective function, but added in order to be included in the log of solutions.
х	Either a vector of values or an object of class stuartOutput from which to determine values in the objective function. If NULL (the default) values are generated from criteria-specific presets.
n	Number of values to use in function determining. Defaults to 50, meaning if side = 'top' the 50 largest values are used to determine discrimination and difficulty parameters for each criterion.
side	Which side good values are located at. 'top' means large values are good (e.g. Reliability), 'bottom' means small values are good (e.g. RMSEA), and 'middle' means average values are good (e.g. factor correlations).
skew	Whether to account for skew in the distribution using the psn() function from the sn-Package. Defaults to FALSE, meaning a normal distribution is used.
scale	A numeric scale to use in weighting the objective component. Defaults to 1.
matrices	An object of class stuartObjectiveMatrices to include matrices (e.g. latent correlations) into the objective function.
fixed	An object of class stuartFixedObjective to include already previously defined fixed objectives.
comparisons	A vector of names of criteria included in the objective function which are related to model comparisons (e.g. when determining measurement invariance).
	Additional arguments.

### Value

Returns an object of class stuartFixedObjective

### Author(s)

Martin Schultze

### See Also

fixedobjective, extractobjective, objectivematrices

12 fixedobjective

fairplayer

MTMM fairplayer Intervention Data (2009)

### **Description**

Self- and teacher-reported empathy (8 item scale), relational aggression (5 item scale), and social intelligence (10 item scale) at three different occasions.

#### **Format**

A data frame with 143 observations on 142 variables. The variable names consist of an initial letter indicating the source (s: self-report, t: teacher-report), two letters indicating the construct (EM: empathy, RA: relational aggression, SI: social intelligence), a number indicating the item number on the scale, and a "t" followed by a number indicating the measurement occasion.

#### Source

Bull, H., Schultze, M., Scheithauer, H. (2009) School-based prevention of bullying and relational aggression: The fairplayer.manual. European Journal of Developmental Science, 3:313-317.

Schultze, M. (2012). Evaluating What The Crowd Says. A longitudinal structural equation model for exchangeable and structurally different methods for evaluating interventions. Unpublished Diploma Thesis.

fixedobjective

Generate a fixed objective function for item selection.

### **Description**

Generate an objective function from default values for use in an item selection using STUART.

### Usage

```
fixedobjective(
  criteria = c("rmsea", "srmr", "crel"),
  add = c("chisq", "df", "pvalue"),
  side = NULL,
  scale = 1,
  matrices = NULL,
  fixed = NULL,
  comparisons = NULL,
  ...
)
```

### Arguments

criteria	A vector of names of criteria included in the objective function. Defaults to $c('rmsea', 'srmr', 'crel')$ .
add	A vector of names of criteria not used in the objective function, but added in order to be included in the log of solutions.
side	Which side good values are located at. 'top' means large values are good (e.g. Reliability), 'bottom' means small values are good (e.g. RMSEA), and 'middle' means average values are good (e.g. factor correlations).
scale	A numeric scale to use in weighting the objective component. Defaults to 1.
matrices	An object of class $stuartObjectiveMatrices$ to include matrices (e.g. latent correlations) into the objective function.
fixed	An object of class stuartFixedObjective to include already previously defined fixed objectives.
comparisons	A vector of names of criteria included in the objective function which are related to model comparisons (e.g. when determining measurement invariance).
	Additional arguments.

### Value

Returns an object of class stuartFixedObjective

### Author(s)

Martin Schultze

### See Also

```
empirical objective, extract objective, objective matrices
```

g	ene	Subtest construction using a simple genetic algorithm

### Description

Construct subtests from a given pool of items using a simple genetic algorithm. Allows for multiple constructs, occasions, and groups.

### Usage

```
gene(
  data,
  factor.structure,
  capacity = NULL,
  item.weights = NULL,
  item.invariance = "congeneric",
```

```
repeated.measures = NULL,
long.invariance = "strict",
mtmm = NULL,
mtmm.invariance = "configural",
grouping = NULL,
group.invariance = "strict",
comparisons = NULL,
auxiliary = NULL,
use.order = FALSE,
software = "lavaan",
cores = NULL,
objective = NULL,
ignore.errors = FALSE,
burnin = 5,
generations = 256,
individuals = 64,
selection = "tournament",
selection.pressure = NULL,
elitism = NULL,
reproduction = 0.5,
mutation = 0.05,
mating.index = 0,
mating.size = 0.25,
mating.criterion = "similarity",
immigration = 0,
convergence.criterion = "geno.between",
tolerance = NULL,
reinit.n = 1,
reinit.criterion = convergence.criterion,
reinit.tolerance = NULL,
reinit.prop = 0.75,
schedule = "run",
analysis.options = NULL,
suppress.model = FALSE,
seed = NULL,
filename = NULL
```

### Arguments

)

data A data.frame containing all relevant data.

factor.structure

A list linking factors to items. The names of the list elements correspond to the factor names. Each list element must contain a character-vector of item names that are indicators of this factor.

capacity

A list containing the number of items per subtest. This must be in the same order as the factor.structure provided. If a single number, it is applied to all subtests. If NULL all items are evenly distributed among the subtests.

item.weights A placeholder. Currently all weights are assumed to be one. item.invariance

A character vector of length 1 or the same length as factor.structure containing the desired invariance levels between items pertaining to the same subtest. Currently there are five options: 'congeneric', 'ess.equivalent', 'ess.parallel', 'equivalent', and 'parallel', the first being the default.

repeated.measures

A list linking factors that are repeated measures of each other. Repeated factors must be in one element of the list - other sets of factors in other elements of the list. When this is NULL (the default) a cross-sectional model is estimated.

long.invariance

A character vector of length 1 or the same length as repeated.measures containing the longitudinal invariance level of repeated items. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When repeated.measures=NULL this argument is ignored.

A list linking factors that are measurements of the same construct with different methods. Measurements of the same construct must be in one element of the list - other sets of methods in other elements of the list. When this is NULL (the default) a single method model is estimated.

A character vector of length 1 or the same length as mtmm containing the invariance level of MTMM items. Currently there are five options: 'none', 'configural', 'weak', 'strong', and 'strict'. Defaults to 'configural'. With 'none' differing items are allowed for different methods. When mtmm=NULL this argument is ignored.

The name of the grouping variable. The grouping variable must be part of data provided and must be a numeric variable.

A single value describing the assumed invariance of items across groups. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When grouping=NULL this argument is ignored.

A character vector containing any combination of 'item', 'long', 'mtmm', and 'group' indicating which invariance should be assessed via model comparisons. The order of the vector dictates the sequence in which model comparisons are performed. Defaults to NULL meaning that no model comparisons are performed.

The names of auxiliary variables in data. These can be used in additional modeling steps that may be provided in analysis.options\$model.

A logical indicating whether or not to take the selection order of the items into account. Defaults to FALSE.

The name of the estimation software. Can currently be 'lavaan' (the default) or 'Mplus'. Each option requires the software to be installed.

The number of cores to be used in parallel processing. If NULL (the default) the result of detectCores will be used. On Unix-y machines parallel processing is implemented via mclapply, on Windows machines it is realized via parLapply.

A function that converts the results of model estimation into a pheromone. See 'details' for... details.

mtmm

mtmm.invariance

grouping

group.invariance

comparisons

auxiliary

use.order

software

cores

objective

ignore.errors A logical indicating whether or not to ignore estimation problems (such as non

positive-definite latent covariance matrices). Defaults to FALSE.

burnin Number of generations for which to use fixed objective function before switch-

 $ing \ to \ empirical \ objective. \ Ignored \ if \ objective \ is \ not \ of \ class \ stuart \ Empirical \ Objetive.$ 

Defaults to 5.

generations Maximum number of generations to run. Defaults to 256.

individuals The number of individuals per generation. Defaults to 64.

selection The method used for selecting possible parents. Can be either 'proportional'

for fitness proportional random selection or 'tournament' (the default) for a

semi-deterministic selection.

selection.pressure

The pressure exerted during the selection process, depending on the selection: if selection = 'proportional' the non-linearity coefficient of the pheromone

when determining selection probability (the default is 1); if selection = 'proportional'

the number of randomly selected individuals from which to choose the best (the

default is 5).

elitism The proportion of individuals from the last generation to carry over to the next

generation. Defaults to 1/individuals, meaning that the best individual is retained

into the next generation.

reproduction The proportion of individuals that are allowed to sire offspring. These individu-

als are selected using fitness proportionate selection. Defaults to .5.

mutation The mutation probability. Defaults to .05. See 'details'.

mating.index The relative rank of the selected mate within the mating pool. A number bewteen

0 (the default) and 1. The meaning depends on the setting of mating.criterion.

See 'details'.

mating.size The proportion of potential mates sampled from the pool of reproducers for each

selected individual. Defaults to .25. See 'details'.

mating.criterion

The criterion by which to select mates. Can be either 'similarity' (the default)

or 'fitness'. See 'details'.

immigration The proportion of individuals per generation that are randomly generated immi-

grants. Defaults to 0.

convergence.criterion

The criterion by which convergence is determined. Can be one of four criteria

'variance', 'median', 'geno.within', and 'geno.between' (the default).

See 'details'.

tolerance The tolerance for determining convergence. The default depends on the setting

used for convergence.criterion. See 'details'.

reinit.n The maximum number of reinitilizations to be performed. Defaults to 1. See

'details'.

reinit.criterion

The convergence criterion used to determine whether the population should be reinitialized. Can be one of four criteria 'variance', 'median', 'geno.within',

and 'geno.between'. Per default, the same criterion provided to convergence.criterion

is used. See 'details'.

reinit.tolerance

The tolerance for determining the necessity of reinitialization. The default depends on the setting used for convergence.criterion. See 'details'.

reinit.prop The proportion of the population to be discarded and replaced by random indi-

viduals when reinitializing. Defaults to .75. See 'details'.

schedule The counter which the scheduling of parameters pertains to. Can be either

'run' (the default), for a continuous schedule, 'generation', for a schedule that is

restarted every time the population is reinitialized.

analysis.options

A list additional arguments to be passed to the estimation software. The names of list elements must correspond to the arguments changed in the respective estimation software. E.g. analysis.options\$model can contain additional modeling commands - such as regressions on auxiliary variables.

suppress.model A logical indicating whether to suppress the default model generation. If TRUE

a model must be provided in analysis.options\$model.

seed A random seed for the generation of random samples. See Random for more

details.

filename The stem of the filenames used to save inputs, outputs, and data files when

software='Mplus'. This may include the file path. When NULL (the default) files will be saved to the temporary directory, which is deleted when the R ses-

sion is ended.

#### **Details**

The pheromone function provided via objective is used to assess the quality of the solutions. These functions can contain any combination of the fit indices provided by the estimation software. When using Mplus these fit indices are 'rmsea', 'srmr', 'cfi', 'tli', 'chisq' (with 'df' and 'pvalue'), 'aic', 'bic', and 'abic'. With lavaan any fit index provided by inspect can be used. Additionally 'crel' provides an aggregate of composite reliabilites, 'rel' provides a vector or a list of reliability coefficients for the latent variables, 'con' provides an aggregate consistency estimate for MTMM analyses, and 'lvcor' provides a list of the latent variable correlation matrices. For more detailed objective functions 'lambda', 'theta', 'psi', and 'alpha' provide the model-implied matrices. Per default a pheromone function using 'crel', 'rmsea', and 'srmr' is used. Please be aware that the objective must be a function with the required fit indices as (correctly named) arguments.

Using model comparisons via the comparisons argument compares the target model to a model with one less degree of assumed invariance (e.g. if your target model contains strong invariance, the comparison model contain weak invariance). Adding comparisons will change the preset for the objective function to include model differences. With comparisons, a custom objective function (the recommended approach) can also include all model fit indices with a preceding delta. to indicate the difference in this index between the two models. If more than one type of comparison is used, the argument of the objective function should end in the type of comparison requested (e.g. delta.cfi.group to use the difference in CFI between the model comparison of invariance across groups).

The genetic algorithm implemented selects parents in a two-step procedure. First, either a tournament or a fitness proportionate selection is performed to select inviduals times reproduction viable parents. Then, the non-self-adaptive version of mating proposed by Galán, Mengshoel, and Pinter (2013) is used to perform mating. In contrast to the original article, the mating index and

mating.size are handled as proportions, not integers. Similarity-based mating is based on the Jaccard Similarity. Mutation is currently always handled as an exchange of the selection state between two items. This results in mutation selecting one item that was not selected prior to mutation and dropping one item selected prior to mutation.

Per default (convergence.criterion = 'geno.between'), convergence is checked by tracking the changes between selection probabilities over three subsequent generations. If the difference between these selections probabilities falls below tolerance (.01 by default) in three consecutive generations, the algorithm is deemed to have converged. To avoid false convergence in the early search, the lower of either 10% of the generations or 10 generations must be completed, before convergence is checked. When using reinitialization the default for reinit. tolerance is .05 to initiate a full reinitialization of the population. An alternative convergence criterion is the variance of the global-best values on the objective function, as proposed by Bhandari, Murthy, and Pal (2012). For generalizability over different functions provided to objective, variances are scaled to the first global-best found. In this case the setting for tolerance pertains to the pure variance estimate and defaults to .0005 (or .005 when regarding the reinitialization process discussed below). Alternatively, the setting 'median' checks for the relative difference between the objective function value of the generation-best and the median value of a generation (scaled by the former). Here, the default is .05 (or .10 when regarding the reinitialization process). The setting 'geno.within' checks for the variability of genotypes in a generation, by determining the relative frequency, with which each item is selected. Convergence is reached if this relative frequency is either tolerance (.8, by default - or .7 for the reinitialization process) or 1 - tolerance for all items within a generation.

A reinitialization procedure can be used to avoid premature convergence. The behavior is controlled via the arguments starting in reinit. The argument reinit.n determines the maximum number of possible reinitializations. After each reinitialization, the generation counter is reset, allowing for the maximum number of generations before the search is aborted. The reinit.criterion and reinit.tolerance relate to convergence criteria outlined above. It is recommended to use a higher tolerance on reinitialization than on final convergence to avoid long periods of stagnant search. The reinit.prop determines the proportion of the population to be replaced by random individuals when reinitializing. Note that even when reinit.prop = 1, the number of individuals kept due to elitism is not discarded.

#### Value

Returns an object of the class stuartOutput for which specific summary and plot methods are available. The results are a list.

call The called function.

software used to fit the CFA models.

parameters A list of the parameters used.

analysis.options

A list of the additional arguments passed to the estimation software.

timer An object of the class proc\_time which contains the time used for the analysis.

log A data. frame containing the optimization history.

log\_mat A list of matrices (e.g. lvcor) relevant to the estimation history, if any.

A list of matrices with the choices made in the global-best solution.

pheromones A list of matrices with the relative selection frequency of items in the final gen-

eration.

subtests A list containing the names of the selected items and their respective subtests.

final The results of the estimation of the global-best solution.

#### Author(s)

Martin Schultze

#### References

Bhandari, D., Murthy, C.A., & Pal, S.K. (2012). Variance as a Stopping Criterion for Genetic Algorithms with Elitist Model. Fundamenta Informaticae, 120, 145-164. doi:10.3233/FI-2012-754 Galán, S.F., Mengshoel, O.J., & Pinter, R. (2013). A novel mating approach for genetic algorithms. Evolutionary Computation, 21(2), 197-229. doi:10.1162/EVCO\_a\_00067

#### See Also

bruteforce, mmas, randomsamples

### **Examples**

```
# Genetic selection in a simple situation
# requires lavaan
# number of cores set to 1 in all examples
data(fairplayer)
fs <- list(si = names(fairplayer)[83:92])</pre>
# minimal example
sel <- gene(fairplayer, fs, 4,</pre>
  generations = 1, individuals = 10, # minimal runtime, remove for application
  seed = 55635, cores = 1)
summary(sel)
# longitudinal example
data(fairplayer)
fs <- list(si1 = names(fairplayer)[83:92],</pre>
  si2 = names(fairplayer)[93:102],
  si3 = names(fairplayer)[103:112])
repe <- list(si = c('si1', 'si2', 'si3'))
# run to convergence
# switching to best-last mating and 50\% mating size
sel <- gene(fairplayer, fs, 4,</pre>
  repeated.measures = repe, long.invariance = 'strong',
  mating.criterion = 'fitness', mating.index = 0,
  mating.size = .5,
  seed = 55635, cores = 1)
# forcing a run through all generations
```

20 heuristics

```
# by disabling the convergence rule
sel <- gene(fairplayer, fs, 4,
  repeated.measures = repe, long.invariance = 'strong',
  tolerance = 0, seed = 55635,
  cores = 1)</pre>
```

heuristics

Generating heuristics for the use in STUART subtest construction

### **Description**

Creates uninformative heuristic matrices for the use in mmas.

### Usage

```
heuristics(
  data,
  factor.structure,
  capacity = NULL,
  repeated.measures = NULL,
  mtmm = NULL,
  grouping = NULL,
  localization = "nodes",
  ...
)
```

#### **Arguments**

data

A data.frame containing all relevant data.

factor.structure

A list linking factors to items. The names of the list elements correspond to the factor names. Each list element must contain a character-vector of item names that are indicators of this factor.

capacity

A list containing the number of items per subtest. This must be in the same order as the factor.structure provided. If a single number, it is applied to all subtests. If NULL all items are evenly distributed among the subtests.

repeated.measures

A list linking factors that are repeated measures of each other. Repeated factors must be in one element of the list - other sets of factors in other elements of the list. When this is NULL (the default) a cross-sectional model is estimated.

mtmm

A list linking factors that are measurements of the same construct with different methods. Measurements of the same construct must be in one element of the list - other sets of methods in other elements of the list. When this is NULL (the default) a single method model is estimated.

heuristics 21

grouping The name of the grouping variable. The grouping variable must be part of data provided and must be a numeric variable.

Which parameterization to use when depositing pheromones. Can be either localization

'nodes' (the default) for depositing pheromones on selected nodes or 'arcs' for

depositing on selection arcs.

Other arguments normally provided to mmas, which will be ignored.

#### **Details**

This function generates a list of matrices which can be used as heuristics for all STUART constructions. This is mainly intended to write the structure of the heuristic matrices to an object, change components in line with theoretically derived heuristics and feed them back into mmas via the heuristics argument. The generated heuristics will contain only 1s and 0s, making it no heuristic information. Selection probabilities can be altered by manipulating the contents of the object created by heuristics. Setting a value to 0 will result in prohibiting a certain choice to be made. Please note, that it will lead to unpredictable behavior if the diagonal elements of the matrices produced in the arcs parameterization are set to values other than 0.

#### Value

Returns a list of the same length as the factor. structure argument provided.

#### Author(s)

Martin Schultze

### See Also

mmas

#### **Examples**

```
# heuristics for node localization
data(fairplayer)
fs <- list(si = names(fairplayer)[83:92])</pre>
(heu <- heuristics(fairplayer, fs, 4))
# Define anchor-item
heu$si[1] <- 10000
# heuristics for arc localization
data(fairplayer)
fs <- list(si = names(fairplayer)[83:92])</pre>
(heu <- heuristics(fairplayer, fs, 4, localization = 'arcs'))
# Define equal selection of odd and even items
heusi[1:10,] < c(rep(c(0, 1), 5), rep(c(1, 0), 5))
```

22 holdout

heu

t Data selection for holdout validation.
- ····································

### Description

Split a data. frame into two subsets for holdout validation.

### Usage

```
holdout(data, prop = 0.5, grouping = NULL, seed = NULL, determined = NULL)
```

### Arguments

data A data.frame.

prop A single value or vector of proportions of data in calibration sample. Defaults

to .5, for an even split.

grouping Name of the grouping variable. Providing a grouping variable ensures that the

provided proportion is selected within each group.

seed A random seed. See Random for more details.

determined Name of a variable indicating the pre-determined assignment to the calibration

or the validation sample. This variable must be a factor containing only NA (no determined assingment), "calibrate", or "validate". If no variable is

provided (the default) all cases are assigned randomly.

### Value

Returns a list containing two data.frames, called calibrate and validate. The first corresponds to the calibration sample, the second to the validation sample.

#### Author(s)

Martin Schultze

#### See Also

crossvalidate

#### **Examples**

```
# seeded selection, 25% validation sample
data(fairplayer)
split <- holdout(fairplayer, .75, seed = 55635)
lapply(split, nrow) # check size of samples</pre>
```

kfold 23

kfold	k-Folds Crossvalidation	

#### **Description**

k-Folds crossvalidation for item selection using any approach implemented in STUART.

#### Usage

```
kfold(
  type,
  k = 5,
  max.invariance = "strict",
  seed = NULL,
  seeded.search = TRUE,
   ...,
  remove.details = TRUE
)
```

#### **Arguments**

type A character calling the item-selection procedure. Can be one of "mmas" (see mmas), "gene" (see gene), "bruteforce" (see bruteforce), or randomsamples (see randomsamples). The number of folds. k max.invariance The maximum measurement invariance level which will be tested. Currently there are four options: 'configural', 'weak', 'strong', and 'strict' (the default). All levels below max.invariance are also tested. The random seed. seed seeded.search A logical indicating whether the seed should also be used for the search procedure (the default) or only for the sample splitting. Arguments passed to the item-selection procedure called with type. remove.details A logical indicating whether to remove detailed information such as models and copies of datasets. Reduces output size by approx. 90%. Defaults to TRUE.

#### **Details**

The function splits the provided data into k subsets using holdout and runs the item-selection procedure requested via type on the training datasets. Validation is performed using crossvalidate to check for invariance of the measurement models between the training and validation data up to the invariance level provided via max.invariance. The final item selection is based on the highest value on the objective function in the multiple-group SEM imposing max.invariance between the training and validation data.

24 kfold

#### Value

Returns an object of the class stuartKfold for which specific summary and print methods are available. The results are a list.

call The called function.

subtests A list containing the names of the selected items and their respective subtests.

solution A list of matrices with the choices made in the global-best solution.

final The results of the estimation of the solution leading to best objective value when

cross-validated.

frequencies A list of matrices showing the relative frequencies with which an item was se-

lected across the k-folds.

full A list of the results returned by the k runs of type.

crossvalidations

A list of data.frames showing the fit and model comparisons of all invariance

levels up to max. invariance in each of the k folds.

data A data. frame. The same as the original data. frame provided to data with the

added variable stuartKfold indicating which fold an observation was assigned

to.

#### Author(s)

Martin Schultze

#### See Also

holdout crossvalidate

#### **Examples**

```
# k-Folding for a simple bruteforce selection
data(fairplayer)
fs <- list(ra = names(fairplayer)[53:57])

sel <- kfold('bruteforce', k = 2,
    data = fairplayer, factor.structure = fs,
    capacity = 3, seed = 55635,
    cores = 1)
summary(sel)

### longitudinal example with mmas ----
data(fairplayer)
fs <- list(si1 = names(fairplayer)[83:92],
    si2 = names(fairplayer)[93:102],
    si3 = names(fairplayer)[103:112])

repe <- list(si = c('si1', 'si2', 'si3'))</pre>
```

```
sel_mmas <- kfold('mmas', k = 3,
  data = fairplayer, factor.structure = fs,
  repeated.measures = repe, long.invariance = 'strong',
  capacity = 3, seed = 55635, pbest = .5,
  cores = 1)
summary(sel_mmas)</pre>
```

mmas

Subtest construction using the Max-Min-Ant-System

#### **Description**

Construct subtests from a given pool of items using the classical Max-Min Ant-System (Stützle, 1998). Allows for multiple constructs, occasions, and groups.

### Usage

```
mmas(
  data,
  factor.structure,
  capacity = NULL,
  item.weights = NULL,
  item.invariance = "congeneric",
  repeated.measures = NULL,
  long.invariance = "strict",
 mtmm = NULL,
 mtmm.invariance = "configural",
  grouping = NULL,
  group.invariance = "strict",
  comparisons = NULL,
  auxiliary = NULL,
  use.order = FALSE,
  software = "lavaan",
  cores = NULL,
  objective = NULL,
  ignore.errors = FALSE,
  burnin = 5,
  ants = 16,
  colonies = 256,
  evaporation = 0.95,
  alpha = 1,
  beta = 1,
  pheromones = NULL,
  heuristics = NULL,
  deposit = "ib",
```

```
localization = "nodes",
pbest = 0.005,
tolerance = 0.5,
schedule = "run",
analysis.options = NULL,
suppress.model = FALSE,
seed = NULL,
filename = NULL
```

#### **Arguments**

data

A data frame containing all relevant data.

factor.structure

A list linking factors to items. The names of the list elements correspond to the factor names. Each list element must contain a character-vector of item names that are indicators of this factor.

capacity

A list containing the number of items per subtest. This must be in the same order as the factor.structure provided. If a single number, it is applied to all subtests. If NULL all items are evenly distributed among the subtests.

item.weights

A placeholder. Currently all weights are assumed to be one.

item.invariance

A character vector of length 1 or the same length as factor.structure containing the desired invariance levels between items pertaining to the same subtest. Currently there are five options: 'congeneric', 'ess.equivalent', 'ess.parallel', 'equivalent', and 'parallel', the first being the default.

repeated.measures

A list linking factors that are repeated measures of each other. Repeated factors must be in one element of the list - other sets of factors in other elements of the list. When this is NULL (the default) a cross-sectional model is estimated.

long.invariance

A character vector of length 1 or the same length as repeated.measures containing the longitudinal invariance level of repeated items. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When repeated.measures=NULL this argument is ignored.

mt.mm

A list linking factors that are measurements of the same construct with different methods. Measurements of the same construct must be in one element of the list - other sets of methods in other elements of the list. When this is NULL (the default) a single method model is estimated.

mtmm.invariance

A character vector of length 1 or the same length as mtmm containing the invariance level of MTMM items. Currently there are five options: 'none', 'configural', 'weak', 'strong', and 'strict'. Defaults to 'configural'. With 'none' differing items are allowed for different methods. When mtmm=NULL this argument is ignored.

grouping

The name of the grouping variable. The grouping variable must be part of data provided and must be a numeric variable.

group.invariance			
	A single value describing the assumed invariance of items across groups. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When grouping=NULL this argument is ignored.		
comparisons	A character vector containing any combination of 'item', 'long', 'mtmm', and 'group' indicating which invariance should be assessed via model comparisons. The order of the vector dictates the sequence in which model comparisons are performed. Defaults to NULL meaning that no model comparisons are performed.		
auxiliary	The names of auxiliary variables in data. These can be used in additional modeling steps that may be provided in analysis.options\$model.		
use.order	A logical indicating whether or not to take the selection order of the items into account. Defaults to FALSE.		
software	The name of the estimation software. Can currently be 'lavaan' (the default) or 'Mplus'. Each option requires the software to be installed.		
cores	The number of cores to be used in parallel processing. If NULL (the default) the result of detectCores will be used. On Unix-y machines parallel processing is implemented via mclapply, on Windows machines it is realized via parLapply.		
objective	A function that converts the results of model estimation into a pheromone. See 'details' for details.		
ignore.errors	A logical indicating whether or not to ignore estimation problems (such as non positive-definite latent covariance matrices). Defaults to FALSE.		
burnin	Number of colonies for which to use fixed objective function before switching to empirical objective. Ignored if objective is not of class stuartEmpiricalObjetive. Defaults to 5.		
ants	The number of ants per colony to be estimated. Can either be a single value or an array with two columns for parameter scheduling. See 'details'.		
colonies	The maximum number of colonies estimated since finding the latest global-best solution before aborting the process. Can either be a single value or an array with two columns for parameter scheduling. See 'details'.		
evaporation	The evaporation coefficient. Can either be a single value or an array with two columns for parameter scheduling. See 'details'.		
alpha	The nonlinearity coefficient of the pheromone-trail's contribution to determining selection probabilities. Defaults to 1 (linear). Can either be a single value or an array with two columns for parameter scheduling. See 'details'.		
beta	The nonlinearity coefficient of the heuristics' contribution to determining selection probabilities. Defaults to 1 (linear). Can either be a single value or an array with two columns for parameter scheduling. See 'details'.		
pheromones	A list of pheromones as created by mmas. This can be used to continue previous runs of this function.		
heuristics	An object of the class stuartHeuristic as provided by heuristics which contains heuristic information to be used in determining selection probabilities. If NULL (the default) selection probabilities are determined solely by the		

pheromones.

deposit Which deposit rule to use. Can be either 'ib' (the default) for an iteration-best

deposit rule, or 'gb' for a global-best deposit rule.

localization Which localization to use when depositing pheromones. Can be either 'nodes'

(the default) for depositing pheromones on selected nodes or 'arcs' for deposit-

ing on selection arcs.

pbest The desired overall probability of constructing the global-best solution when the

algorithm convergels. Can either be a single value or an array with two columns

for parameter scheduling. See 'details'.

tolerance The tolerance of imprecision when comparing the pheromones to the upper and

lower limits. Can either be a single value or an array with two columns for

parameter scheduling. See 'details'.

schedule The counter which the scheduling of parameters pertains to. Can be either 'run'

(the default), for a continuous schedule, 'colony', for a schedule that is restarted every time a new global best is found, or 'mixed' for a schedule that restarts its

current phase every time a new global best is found. See 'details'.

analysis.options

A list additional arguments to be passed to the estimation software. The names of list elements must correspond to the arguments changed in the respective estimation software. E.g. analysis.options\$model can contain additional

modeling commands - such as regressions on auxiliary variables.

suppress.model A logical indicating whether to suppress the default model generation. If TRUE

a model must be provided in analysis.options\$model.

seed A random seed for the generation of random samples. See Random for more

details.

filename The stem of the filenames used to save inputs, outputs, and data files when

software='Mplus'. This may include the file path. When NULL (the default) files will be saved to the temporary directory, which is deleted when the R ses-

sion is ended.

#### **Details**

The pheromone function provided via objective is used to assess the quality of the solutions. These functions can contain any combination of the fit indices provided by the estimation software. When using Mplus these fit indices are 'rmsea', 'srmr', 'cfi', 'tli', 'chisq' (with 'df' and 'pvalue'), 'aic', 'bic', and 'abic'. With lavaan any fit index provided by inspect can be used. Additionally 'crel' provides an aggregate of composite reliabilites, 'rel' provides a vector or a list of reliability coefficients for the latent variables, 'con' provides an aggregate consistency estimate for MTMM analyses, and 'lvcor' provides a list of the latent variable correlation matrices. For more detailed objective functions 'lambda', 'theta', 'psi', 'alpha', and 'nu' provide the model-implied matrices. Per default a pheromone function using 'crel', 'rmsea', and 'srmr' is used. Please be aware that the objective must be a function with the required fit indices as (correctly named) arguments.

Using model comparisons via the comparisons argument compares the target model to a model with one less degree of assumed invariance (e.g. if your target model contains strong invariance, the comparison model contain weak invariance). Adding comparisons will change the preset for the objective function to include model differences. With comparisons, a custom objective function (the recommended approach) can also include all model fit indices with a preceding delta. to

indicate the difference in this index between the two models. If more than one type of comparison is used, the argument of the objective function should end in the type of comparison requested (e.g. delta.cfi.group to use the difference in CFI between the model comparison of invariance across groups).

The scheduling of parameters is possible for the arguments ants, colonies, evaporation, pbest, alpha, beta, tolerance, and deposit. For all of these parameter scheduling is done when an array with two columns is provided. The first column of the array contains the timer, i.e. when to switch between parameter settings, the second column contains the values. The argument schedule can be used to select an absolute schedule (schedule='run'), a relative schedule which resets completely after a new global best is found (schedule='colony'), or a mixed version which resets the current phase of the schedule after a new global best is found (schedule='mixed'). When providing a parameter schedule for iterations 0, 3, and 10 using 'run' will result in a change after the third and the tenth iteration - irrespective of whether global best solutions were found. In contrast, using 'colony' will result in the first setting being used again once a new global best is found. This setting will then be used until iteration 3 (if no new best solution is found) before a switch occurs. If a new global best is found the setting will begin the sequence from the beginning. Using 'mixed' will result in the first setting being used until three consecutive iterations cannot produce a new global best. After this the second setting is used. If a new global best is found, the second setting is kept, but for the purpose of the schedule it is now iteration 3 again, meaning that the third setting will be used later than in a 'run' schedule.

#### Value

Returns an object of the class stuartOutput for which specific summary and plot methods are available. The results are a list.

call The called function.

software The software used to fit the CFA models.

parameters A list of the ACO parameters used.

analysis.options

A list of the additional arguments passed to the estimation software.

timer An object of the class proc\_time which contains the time used for the analysis.

log A data. frame containing the optimization history.

log\_mat A list of matrices (e.g. lvcor) relevant to the estimation history, if any.

A list of matrices with the choices made in the global-best solution.

pheromones A list of matrices with the pheromones of each choice.

subtests A list containing the names of the selected items and their respective subtests.

final The results of the estimation of the global-best solution.

### Author(s)

Martin Schultze

#### References

Stützle, T. (1998). Local search algorithms for combinatorial problems: Analysis, improvements, and new applications. Unpublished doctoral dissertation. Darmstadt: Fachbereich Informatik, Universität Darmstadt.

30 objective matrices

#### See Also

bruteforce, gene, randomsamples, heuristics

#### **Examples**

```
# MMAS in a simple situation
# requires lavaan
# number of cores set to 1 in all examples
data(fairplayer)
fs <- list(si = names(fairplayer)[83:92])</pre>
# minimal example
sel <- mmas(fairplayer, fs, 4,
  colonies = 0, ants = 10, # minimal runtime, remove for application
  seed = 55635, cores = 1)
summary(sel)
# longitudinal example
data(fairplayer)
fs <- list(si1 = names(fairplayer)[83:92],</pre>
  si2 = names(fairplayer)[93:102],
  si3 = names(fairplayer)[103:112])
repe <- list(si = c('si1', 'si2', 'si3'))
# change evaporation rate after 10 and 20 colonies
sel <- mmas(fairplayer, fs, 4,</pre>
  repeated.measures = repe, long.invariance = 'strong',
  evaporation = cbind(c(0, 10, 20), c(.95, .8, .5)),
  seed = 55635, cores = 1)
```

objectivematrices

Generate matrix-components for objective functions.

#### **Description**

Generate objects of the correct structure for use in custom objective functions.

### Usage

```
objectivematrices(
  data,
  factor.structure,
  capacity = NULL,
  matrices = c("lvcor"),
  n.random = 0,
```

objectivematrices 31

```
item.invariance = "congeneric",
  repeated.measures = NULL,
  long.invariance = "strict",
 mtmm = NULL,
 mtmm.invariance = "configural",
  grouping = NULL,
  group.invariance = "strict",
  comparisons = NULL,
  auxiliary = NULL,
  use.order = FALSE,
  software = "lavaan",
  cores = NULL,
  objective = NULL,
  ignore.errors = FALSE,
  analysis.options = NULL,
  suppress.model = FALSE,
)
```

#### **Arguments**

data

A data frame containing all relevant data.

factor.structure

A list linking factors to items. The names of the list elements correspond to the factor names. Each list element must contain a character-vector of item names that are indicators of this factor.

capacity

A list containing the number of items per subtest. This must be in the same order as the factor.structure provided. If a single number, it is applied to all subtests. If NULL all items are evenly distributed among the subtests.

matrices

Which matrix to extract. Can be one of 'lvcor' (the default) for latent correlations, 'lambda', 'theta', 'psi', or 'alpha' for the model-implied matrices.

n.random

The number of random draws to base values on. If 0 (the default) values are in the matrices are set to 0 and can be overwritten later. If any value larger than 0, the mean from n.random random solutions is used.

item.invariance

A character vector of length 1 or the same length as factor.structure containing the desired invariance levels between items pertaining to the same subtest. Currently there are five options: 'congeneric', 'ess.equivalent', 'ess.parallel', 'equivalent', and 'parallel', the first being the default.

repeated.measures

A list linking factors that are repeated measures of each other. Repeated factors must be in one element of the list - other sets of factors in other elements of the list. When this is NULL (the default) a cross-sectional model is estimated.

long.invariance

A character vector of length 1 or the same length as repeated.measures containing the longitudinal invariance level of repeated items. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When repeated.measures=NULL this argument is ignored.

32 objectivematrices

mtmm

A list linking factors that are measurements of the same construct with different methods. Measurements of the same construct must be in one element of the list - other sets of methods in other elements of the list. When this is NULL (the default) a single method model is estimated.

mtmm.invariance

A character vector of length 1 or the same length as mtmm containing the invariance level of MTMM items. Currently there are five options: 'none', 'configural', 'weak', 'strong', and 'strict'. Defaults to 'configural'. With 'none' differing items are allowed for different methods. When mtmm=NULL this argument is ignored.

grouping

The name of the grouping variable. The grouping variable must be part of data provided and must be a numeric variable.

group.invariance

A single value describing the assumed invariance of items across groups. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When grouping=NULL this argument is ignored.

comparisons

A character vector containing any combination of 'item', 'long', 'mtmm', and 'group' indicating which invariance should be assessed via model comparisons. The order of the vector dictates the sequence in which model comparisons are performed. Defaults to NULL meaning that no model comparisons are performed.

auxiliary

The names of auxiliary variables in data. These can be used in additional modeling steps that may be provided in analysis.options\$model.

use.order

A logical indicating whether or not to take the selection order of the items into account. Defaults to FALSE.

software

The name of the estimation software. Can currently be 'lavaan' (the default) or 'Mplus'. Each option requires the software to be installed.

cores

The number of cores to be used in parallel processing. If NULL (the default) the result of detectCores will be used. On Unix-y machines parallel processing is implemented via mclapply, on Windows machines it is realized via parLapply.

objective

A function that converts the results of model estimation into a pheromone. See 'details' for... details.

ignore.errors

A logical indicating whether or not to ignore estimation problems (such as non positive-definite latent covariance matrices). Defaults to FALSE.

analysis.options

A list additional arguments to be passed to the estimation software. The names of list elements must correspond to the arguments changed in the respective estimation software. E.g. analysis.options\$model can contain additional modeling commands - such as regressions on auxiliary variables.

suppress.model

A logical indicating whether to suppress the default model generation. If TRUE a model must be provided in analysis.options\$model.

... Additional arguments passed either to randomsamples or to lavaan.

#### Value

Returns an object of class stuartFixedObjective

#### Author(s)

Martin Schultze

#### See Also

```
empirical objective, extract objective, objective matrices
```

### **Examples**

```
# Extract latent correlation matric
# requires lavaan
# number of cores set to 1 in all examples
data(sups)
fs <- list(pro = names(sups)[2:13],
  fee = names(sups)[14:20])

mats <- objectivematrices(sups, fs, 3,
    cores = 1)
mats

mats$lvcor$use[,] <- FALSE
mats$lvcor$use[2, 1] <- TRUE</pre>
mats$lvcor$use
```

randomsamples

Generating random samples of Subtests

#### **Description**

Construct a defined number of random subtests from a given pool of items.

### Usage

```
randomsamples(
  data,
  factor.structure,
  capacity = NULL,
  item.invariance = "congeneric",
  repeated.measures = NULL,
  long.invariance = "strict",
  mtmm = NULL,
  mtmm.invariance = "configural",
  grouping = NULL,
  group.invariance = "strict",
  comparisons = NULL,
```

```
auxiliary = NULL,
  use.order = FALSE,
  software = "lavaan",
  cores = NULL,
  objective = NULL,
  ignore.errors = FALSE,
  analysis.options = NULL,
  suppress.model = FALSE,
  seed = NULL,
  request.override = 10000,
  filename = NULL,
 n = 1000,
 percentile = 100
)
```

#### **Arguments**

data

A data.frame containing all relevant data.

factor.structure

A list linking factors to items. The names of the list elements correspond to the factor names. Each list element must contain a character-vector of item names that are indicators of this factor.

capacity

A list containing the number of items per subtest. This must be in the same order as the factor. structure provided. If a single number, it is applied to all subtests. If NULL all items are evenly distributed among the subtests.

item.invariance

A character vector of length 1 or the same length as factor. structure containing the desired invariance levels between items pertaining to the same subtest. Currently there are five options: 'congeneric', 'ess.equivalent', 'ess.parallel', 'equivalent', and 'parallel', the first being the default.

repeated.measures

A list linking factors that are repeated measures of each other. Repeated factors must be in one element of the list - other sets of factors in other elements of the list. When this is NULL (the default) a cross-sectional model is estimated.

long.invariance

A character vector of length 1 or the same length as repeated.measures containing the longitudinal invariance level of repeated items. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When repeated measures=NULL this argument is ignored.

mtmm

A list linking factors that are measurements of the same construct with different methods. Measurements of the same construct must be in one element of the list - other sets of methods in other elements of the list. When this is NULL (the default) a single method model is estimated.

mtmm.invariance

A character vector of length 1 or the same length as mtmm containing the invariance level of MTMM items. Currently there are five options: 'none', 'configural', 'weak', 'strong', and 'strict'. Defaults to 'configural'. With 'none' differing

items are allowed for different methods. When mtmm=NULL this argument is ignored.

grouping

The name of the grouping variable. The grouping variable must be part of data provided and must be a numeric variable.

group.invariance

A single value describing the assumed invariance of items across groups. Currently there are four options: 'configural', 'weak', 'strong', and 'strict'. Defaults to 'strict'. When grouping=NULL this argument is ignored.

comparisons

A character vector containing any combination of 'item', 'long', 'mtmm', and 'group' indicating which invariance should be assessed via model comparisons. The order of the vector dictates the sequence in which model comparisons are performed. Defaults to NULL meaning that no model comparisons are performed.

auxiliary

The names of auxiliary variables in data. These can be used in additional modeling steps that may be provided in analysis.options\$model.

use.order

A logical indicating whether or not to take the selection order of the items into account. Defaults to FALSE.

software

The name of the estimation software. Can currently be 'lavaan' (the default), 'Mplus', or 'Mplus Demo'. Each option requires the software to be installed.

cores

The number of cores to be used in parallel processing. If NULL (the default) the result of detectCores will be used. On Unix-y machines parallel processing is implemented via mclapply, on Windows machines it is realized via parLapply.

objective

A function that converts the results of model estimation into a pheromone. See mmas for details.

ignore.errors

A logical indicating whether or not to ignore estimation problems (such as non positive-definite latent covariance matrices). Defaults to FALSE.

analysis.options

A list additional arguments to be passed to the estimation software. The names of list elements must correspond to the arguments changed in the respective estimation software. E.g. analysis.options\$model can contain additional modeling commands - such as regressions on auxiliary variables.

suppress.model

A logical indicating whether to suppress the default model generation. If TRUE a model must be provided in analysis.options\$model.

seed

A random seed for the generation of random samples. See Random for more details.

request.override

The maximum number of combinations for which the estimation is performed immediately, without an additional override request.

filename

The stem of the filenames used to save inputs, outputs, and data files when software='Mplus'. This may include the file path. When NULL (the default) files will be saved to the temporary directory, which is deleted when the R session is ended.

n

The number of random samples to be drawn.

percentile

The percentile of the final solution reported among the viable solutions. Defaults to 100 (the best solution found).

#### **Details**

The pheromone function provided via objective is used to assess the quality of the solutions. These functions can contain any combination of the fit indices provided by the estimation software. When using Mplus these fit indices are 'rmsea', 'srmr', 'cfi', 'tli', 'chisq' (with 'df' and 'pvalue'), 'aic', 'bic', and 'abic'. With lavaan any fit index provided by inspect can be used. Additionally 'crel' provides an aggregate of composite reliabilites, 'rel' provides a vector or a list of reliability coefficients for the latent variables, 'con' provides an aggregate consistency estimate for MTMM analyses, and 'lvcor' provides a list of the latent variable correlation matrices. For more detailed objective functions 'lambda', 'theta', 'psi', 'alpha', and 'nu' provide the model-implied matrices. Per default a pheromone function using 'crel', 'rmsea', and 'srmr' is used. Please be aware that the objective must be a function with the required fit indices as (correctly named) arguments.

Using model comparisons via the comparisons argument compares the target model to a model with one less degree of assumed invariance (e.g. if your target model contains strong invariance, the comparison model contain weak invariance). Adding comparisons will change the preset for the objective function to include model differences. With comparisons, a custom objective function (the recommended approach) can also include all model fit indices with a preceding delta. to indicate the difference in this index between the two models. If more than one type of comparison is used, the argument of the objective function should end in the type of comparison requested (e.g. delta.cfi.group to use the difference in CFI between the model comparison of invariance across groups).

#### Value

Returns an object of the class stuartOutput for which specific summary and plot methods are available. The results are a list.

call The called function.

software used to fit the CFA models.

parameters A list of the parameters used.

analysis.options

A list of the additional arguments passed to the estimation software.

timer An object of the class proc\_time which contains the time used for the analysis.

log A data. frame containing the estimation history.

log\_mat A list of matrices (e.g. lvcor) relevant to the estimation history, if any.

solution NULL pheromones NULL

subtests A list containing the names of the selected items and their respective subtests.

final The results of the estimation of the global-best solution.

#### Author(s)

Martin Schultze

#### See Also

bruteforce, mmas, gene

sia 37

### **Examples**

```
# Random samples in a simple situation
# requires lavaan
# number of cores set to 1 in all examples
data(fairplayer)
fs <- list(si = names(fairplayer)[83:92])
# 10 random solutions, report median solution
sel <- randomsamples(fairplayer, fs, 4,
    n = 10, percentile = 50,
    seed = 55635, cores = 1)
summary(sel)</pre>
```

sia

Data from a German Meaning of Work Scale.

#### Description

Self-reports from a scale construction study for a German Meaning of Work Scale (Feser et al., 2019) with three facets: Self-realization (17 items), belonging (9 items), justification (8 Items). The data additionally include assessments on the Work and Meaning Inventory (Steger et al., 2012), alienation from work (Fischer and Kohr, 2014).

### Usage

sia

#### **Format**

A data frame with 257 observations on 62 variables.

#### **Details**

- lfdn. Participant ID.
- self1 self17. 17 items of the self-realization facet.
- belong1 belong9. 9 items of the belonging facet.
- just1 just8. 8 items of the justification facet.
- wami1 wami10. 10 items of the Work and Meaning Inventory.
- alien1 alien10. 10 item of the alienation from work sacle.
- age. Age in years.
- sex. Gender with 1 = female, 2 = other, 3 = male.
- work. Type of employment 1 = employed, 2 = self-employed, 3 = temp-work, 4 = civil servant.

38 sups

- wokrhours. Weekly work hours.
- tenure1. Years at the current place of employment.
- tenure2. Years of experience in current job.
- kldb2010. German Classification of Occupations (first code number).

#### Source

Feser, M., Lorenz, T., & Mainz, E. (2019). Meaning of work: A culture based approach towards the construction of a German questionnaire. Poster presented at the 19th Congress of The European Association for Work & Organizational Psychology. Turin, Italy.

Fischer, A., & Kohr, H. (2014). Entfremdung von der Arbeit. Zusammentstellung Sozialwissenschaftlicher Items Und Skalen, ZIS. https://doi.org/https://doi.org/10.6102/zis8

Steger, M. F., Dik, B. J., & Duffy, R. D. (2012). Measuring Meaningful Work. Journal of Career Assessment, 20(3), 322–337. https://doi.org/10.1177/1069072711436160

sups

Data from a scale for Supervisor Support

#### **Description**

A scale for supervisor support with 19 items. The scale consists of two subscales: carreer promotion (items 1 through 12) and feedback and goal setting (items 13 through 19).

### **Format**

A data frame with 411 observations on 20 variables. The first variable indicates the person ID, the following 19 all stem from the scale for Supervisor Support

#### Source

Janssen, A.B., Schultze, M., & Grötsch, A. (2015). Following the ants: Development of short scales for proactive personality and supervisor support by Ant Colony Optimization. European Journal of Psychological Assessment.

## **Index**

```
* ACO subtests
                                                     parLapply, 5, 15, 27, 32, 35
    bruteforce, 4
                                                     Random, 17, 22, 28, 35
    combinations, 7
                                                     randomsamples, 2, 7, 19, 23, 30, 32, 33
    crossvalidate, 9
    gene, 13
                                                     sia, 2, 37
    heuristics, 20
                                                     stuart (stuart-package), 2
    holdout, 22
                                                     stuart-package, 2
    mmas, 25
                                                     sups, 2, 38
    randomsamples, 33
* datasets
    fairplayer, 12
    sia, 37
    sups, 38
as.stuartFixedObjective, 3
bruteforce, 2, 4, 8, 10, 19, 23, 30, 36
combinations, 2, 7, 7
crossvalidate, 2, 9, 22-24
detectCores, 5, 15, 27, 32, 35
empiricalobjective, 2, 3, 10, 13, 33
extractobjective, 11, 13, 33
fairplayer, 2, 12
fixedobjective, 2, 3, 11, 12
gene, 2, 7, 8, 13, 23, 30, 36
heuristics, 2, 20, 27, 30
holdout, 2, 10, 22, 23, 24
inspect, 6, 17, 28, 36
kfold, 2, 23
mclapply, 5, 15, 27, 32, 35
mmas, 2, 5, 7, 8, 10, 19-21, 23, 25, 27, 35, 36
objectivematrices, 2, 11, 13, 30, 33
```