

Package ‘AATtools’

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Type Package

Title Reliability and Scoring Routines for the Approach-Avoidance Task

Version 0.0.2

Description Compute approach bias scores using different scoring algorithms,
compute bootstrapped and exact split-half reliability estimates,
and compute confidence intervals for individual participant scores.

Depends R (\geq 3.6.0)

Imports magrittr, dplyr, doParallel, foreach

License GPL-3

Encoding UTF-8

BugReports <https://github.com/Spiritspeak/AATtools/issues>

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R topics documented:

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aat_alpha

*Compute reliability using Cronbach's Alpha***Description**

aat_alpha computes approach-avoidance scores for each stimulus within each participant. It then computes Cronbach's alpha by treating each stimulus as an item on a questionnaire and measuring how much these scores correlate within each participant.

aat_alpha_jackknife additionally leaves out one stimulus or participant at a time and computes Cronbach's alpha for each such exclusion. This gives a glimpse of which stimuli may have especially confused participants or otherwise induced responses that are unlike those to other stimuli.

Please note that this method does not tolerate missing values - for every movement direction, within every stimulus, within every participant, there must be at least one trial - otherwise either the entire stimulus or participant must be excluded. [q_reliability](#) or [aat_splithalf](#) are recommended instead.

Usage

```
aat_alpha(
  ds,
  subjvar,
  stimvar,
  pullvar,
  rtvar,
  algorithm = c("aat_singlemeandiff", "aat_singlemediandiff"),
  delete.missing = c("subjects", "stimuli", "both", "none")
)

aat_alpha_jackknife(
  ds,
  subjvar,
  stimvar,
  pullvar,
  rtvar,
  algorithm = c("aat_singlemeandiff", "aat_singlemediandiff"),
  delete.missing = c("subjects", "stimuli", "both", "none")
)

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

## S3 method for class 'aat_alpha'
plot(x, ...)

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

## S3 method for class 'aat_alpha_jackknife'
plot(x, exclusions = c("both", "stimulus", "participant"), ...)
```

Arguments

<code>ds</code>	a data.frame
<code>subjvar</code>	character naming the column with participant IDs
<code>stimvar</code>	character naming the column with stimulus IDs
<code>pullvar</code>	character naming the column for movement direction
<code>rtvar</code>	character indicating the column with reaction times
<code>algorithm</code>	character name of the algorithm used to compute per-stimulus approach bias scores
<code>delete.missing</code>	character denoting the deletion strategy when missing values are encountered, which may occur when one participant saw a stimulus which another participant did not see, or when all approach trials for a certain stimulus were excluded for one participant. Defaults to excluding the entire participant's dataset.
<code>x</code>	an <code>aat_alpha</code> or <code>aat_alpha_jackknife</code> object
<code>...</code>	Ignored.
<code>exclusions</code>	Character. Should the function display Cronbach's alpha for individually excluded participants, stimuli, or both?

Value

`aat_alpha` returns an `aat_alpha` S3 object, `aat_alpha_jackknife` returns an `aat_alpha_jackknife` S3 object, both with `print` and `plot` methods.

Author(s)

Sercan Kahveci

References

Cousijn, J., Goudriaan, A. E., & Wiers, R. W. (2011). Reaching out towards cannabis: Approach-bias in heavy cannabis users predicts changes in cannabis use. *Addiction*, 106(9), 1667-1674.

Examples

```
data(erotica)
#We artificially reduce the number of stimuli here because the original
#erotica dataset is not suitable for computing Cronbach's alpha.
erotica$stimulus<- substr(as.character(erotica$stimulus),5,5)

myalpha<-aat_alpha(erotica,"subject","stimulus","is_pull","RT")
print(myalpha)
plot(myalpha)

myalpha2<-aat_alpha_jackknife(erotica,"subject","stimulus","is_pull","RT")
print(myalpha2)
plot(myalpha2)
```

aat_bootstrap

*Compute bootstrapped approach-bias scores***Description**

Compute bootstrapped approach-bias scores with confidence intervals.

Usage

```
aat_bootstrap(
  ds,
  subjvar,
  pullvar,
  targetvar = NULL,
  rtvar,
  iters,
  algorithm = c("aat_doublemeandiff", "aat_doublemediandiff", "aat_dscore",
    "aat_dscore_multiblock", "aat_regression", "aat_standardregression",
    "aat_doublemeanquotient", "aat_doublemedianquotient", "aat_singlemeandiff",
    "aat_singlemediandiff"),
  trialdropfunc = c("prune_nothing", "trial_prune_3SD", "trial_prune_SD_dropcases",
    "trial_recode_SD", "trial_prune_percent_subject", "trial_prune_percent_sample"),
  errortrialfunc = c("prune_nothing", "error_replace_blockmeanplus",
    "error_prune_dropcases"),
  plot = TRUE,
  include.raw = FALSE,
  parallel = TRUE,
  ...
)

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

## S3 method for class 'aat_bootstrap'
plot(x, ...)
```

Arguments

<code>ds</code>	a longformat data.frame
<code>subjvar</code>	Quoted name of the participant identifier column
<code>pullvar</code>	Quoted name of the column indicating pull trials. Pull trials should either be represented by 1, or by the second level of a factor.
<code>targetvar</code>	Name of the column indicating trials featuring the target stimulus. Target stimuli should either be represented by 1, or by the second level of a factor.
<code>rtvar</code>	Name of the reaction time column.
<code>iters</code>	Total number of desired iterations. At least 200 are required to get confidence intervals that make sense.
<code>algorithm</code>	Function (without brackets or quotes) to be used to compute AAT scores. See Algorithms for a list of usable algorithms.

- trialdropfunc** Function (without brackets or quotes) to be used to exclude outlying trials in each half. The way you handle outliers for the reliability computation should mimic the way you do it in your regular analyses. It is recommended to exclude outlying trials when computing AAT scores using the mean double-difference scores and regression scoring approaches, but not when using d-scores or median double-difference scores.
- **prune_nothing** excludes no trials (default)
 - **trial_prune_3SD** excludes trials deviating more than 3SD from the mean per participant.
 - **trial_prune_SD_dropcases** removes trials deviating more than a specific number of standard deviations from the participant's mean, and removes participants with an excessive percentage of outliers. Required arguments:
 - **trialsd** - trials deviating more than **trialsd** standard deviations from the participant's mean are excluded (optional; default is 3)
 - **maxoutliers** - participants with a higher percentage of outliers are removed from the data. (optional; default is .15)
 - **trial_recode_SD** recodes outlying reaction times to the nearest non-outlying value, with outliers defined as reaction times deviating more than a certain number of standard deviations from the participant's mean. Required argument:
 - **trialsd** - trials deviating more than this many standard deviations from the mean are classified as outliers.
 - **trial_prune_percent_subject** and **trial_prune_percent_sample** remove trials below and/or above certain percentiles, on a subject-by-subject basis or sample-wide, respectively. The following arguments are available:
 - **lowerpercent** and **upperpercent** (optional; defaults are .01 and .99).
- errortrialfunc** Function (without brackets or quotes) to apply to an error trial.
- **prune_nothing** removes no errors (default).
 - **error_replace_blockmeanplus** replaces error trial reaction times with the block mean, plus an arbitrary extra quantity. If used, the following additional arguments are required:
 - **blockvar** - Quoted name of the block variable (mandatory)
 - **errorvar** - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
 - **errorbonus** - Amount to add to the reaction time of error trials. Default is 0.6 (recommended by Greenwald, Nosek, & Banaji, 2003)
 - **error_prune_dropcases** removes errors and drops participants if they have more errors than a given percentage. The following arguments are available:
 - **errorvar** - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
 - **maxerrors** - participants with a higher percentage of errors are excluded from the dataset. Default is .15.
- plot** Plot the bias scores and their confidence intervals after computation is complete. This gives a good overview of the data.

<code>include.raw</code>	logical indicating whether raw split-half data should be included in the output object.
<code>parallel</code>	If TRUE (default), will use parallel computing to compute results faster. If a <code>doParallel</code> backend has not been registered beforehand, this function will register a cluster and stop it after finishing, which takes some extra time.
<code>...</code>	Other arguments, to be passed on to the algorithm or outlier rejection functions (see arguments above)
<code>x</code>	An <code>aat.bootstrap</code> object.

Value

A list, containing bootstrapped bias scores, their variance, bootstrapped 95 percent confidence intervals, the number of iterations, and a matrix of bias scores for each iteration.

Author(s)

Sercan Kahveci

Examples

```
# Compute 10 bootstrapped AAT scores.
boot<-aat_bootstrap(ds=erotica[erotica$is_irrelevant==0,], subjvar="subject",
                    pullvar="is_pull", targetvar="is_target",rtvar="RT",
                    iters=10,algorithm="aat_doublemediandiff",
                    trialdropfunc="trial_prune_3SD",
                    plot=FALSE, parallel=FALSE)

plot(boot)
print(boot)
```

aat_compute

Compute simple AAT scores

Description

Compute simple AAT scores, with optional outlier exclusion and error trial recoding.

Usage

```
aat_compute(
  ds,
  subjvar,
  pullvar,
  targetvar = NULL,
  rtvar,
  algorithm = c("aat_doublemeandiff", "aat_doublemediandiff", "aat_dscore",
    "aat_dscore_multiblock", "aat_regression", "aat_standardregression",
    "aat_doublemeanquotient", "aat_doublemedianquotient", "aat_singlemeandiff",
    "aat_singlemediandiff"),
  trialdropfunc = c("prune_nothing", "trial_prune_3SD", "trial_prune_SD_dropcases",
    "trial_recode_SD", "trial_prune_percent_subject", "trial_prune_percent_sample"),
```

```

    errortrialfunc = c("prune_nothing", "error_replace_blockmeanplus",
      "error_prune_dropcases"),
    ...
  )

```

Arguments

- | | |
|-----------------------|---|
| ds | a long-format data.frame |
| subjvar | column name of subject variable |
| pullvar | column name of pull/push indicator variable, must be numeric or logical (where pull is 1 or TRUE) |
| targetvar | column name of target stimulus indicator, must be numeric or logical (where target is 1 or TRUE) |
| rtvar | column name of reaction time variable |
| algorithm | Function (without brackets or quotes) to be used to compute AAT scores. See Algorithms for a list of usable algorithms. |
| trialdropfunc | <p>Function (without brackets or quotes) to be used to exclude outlying trials in each half. The way you handle outliers for the reliability computation should mimic the way you do it in your regular analyses. It is recommended to exclude outlying trials when computing AAT scores using the mean double-difference scores and regression scoring approaches, but not when using d-scores or median double-difference scores.</p> <ul style="list-style-type: none"> • prune_nothing excludes no trials (default) • trial_prune_3SD excludes trials deviating more than 3SD from the mean per participant. • trial_prune_SD_dropcases removes trials deviating more than a specific number of standard deviations from the participant's mean, and removes participants with an excessive percentage of outliers. Required arguments: <ul style="list-style-type: none"> – trialsd - trials deviating more than trialsd standard deviations from the participant's mean are excluded (optional; default is 3) – maxoutliers - participants with a higher percentage of outliers are removed from the data. (optional; default is .15) • trial_recode_SD recodes outlying reaction times to the nearest non-outlying value, with outliers defined as reaction times deviating more than a certain number of standard deviations from the participant's mean. Required argument: <ul style="list-style-type: none"> – trialsd - trials deviating more than this many standard deviations from the mean are classified as outliers. • trial_prune_percent_subject and trial_prune_percent_sample remove trials below and/or above certain percentiles, on a subject-by-subject basis or sample-wide, respectively. The following arguments are available: <ul style="list-style-type: none"> – lowerpercent and upperpercent (optional; defaults are .01 and .99). |
| errortrialfunc | <p>Function (without brackets or quotes) to apply to an error trial.</p> <ul style="list-style-type: none"> • prune_nothing removes no errors (default). • error_replace_blockmeanplus replaces error trial reaction times with the block mean, plus an arbitrary extra quantity. If used, the following additional arguments are required: |

- **blockvar** - Quoted name of the block variable (mandatory)
- **errorvar** - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
- **errorbonus** - Amount to add to the reaction time of error trials. Default is 0.6 (recommended by Greenwald, Nosek, & Banaji, 2003)
- **error_prune_dropcases** removes errors and drops participants if they have more errors than a given percentage. The following arguments are available:
 - **errorvar** - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
 - **maxerrors** - participants with a higher percentage of errors are excluded from the dataset. Default is .15.
- ... Other arguments, to be passed on to the algorithm or outlier rejection functions (see arguments above)

Examples

```
#Compute the correlation between relevant-feature and irrelevant-feature AAT scores
ds<-erotica[erotica$correct==1,]
relevant <- aat_compute(ds=ds[ds$is_irrelevant==0,],
  pullvar="is_pull",targetvar="is_target",
  rtvar="RT",subjvar="subject",
  trialdropfunc="trial_prune_3SD",
  algorithm="aat_doublemediandiff")

irrelevant <- aat_compute(ds=ds[ds$is_irrelevant==1,],
  pullvar="is_pull",targetvar="is_target",
  rtvar="RT",subjvar="subject",
  trialdropfunc="trial_prune_3SD",
  algorithm="aat_doublemediandiff")

comparison.df <- merge(relevant, irrelevant, by = "subject")
cor(comparison.df$ab.x, comparison.df$ab.y)
# 0.1145726
```

aat_simulate

Simulate AAT datasets and predict parameters

Description

aat_simulate() generates approach-avoidance task datasets.

Usage

```
aat_simulate(
  npps = 40,
  nstims = 32,
  stimreps = 2,
  meanrt = 743,
  meanrt_jitter = 66,
  sdrt = 133,
  sdrt_jitter = 38,
```



```

    pullfx = 25,
    pullfx_jitter = 40,
    stimfx = 10,
    stimfx_jitter = 35,
    biasfx = 35,
    biasfx_jitter = 75
  )

aat_simulate2(
  ...,
  defaults = c("none", "Lender2018_raw", "Lender2018_clean", "Kahveci2021_raw",
    "Kahveci2021_clean"),
  slowols = 0,
  fastols = 0,
  olsd = 3
)

```

Arguments

npps	Number of participants
nstim	Number of stimuli
stimreps	Number of repetitions of each stimulus within each group (i.e. within approach target, avoid target, approach control, avoid control)
meanrt	Mean sample reaction time
meanrt_jitter	Extent by which participants' mean RTs deviate from mean sample RT.
sdrt	Standard deviation of samplewide RTs, ignoring effects of movement, stimulus, and approach bias. In essence, this represents the amount of pure noise present in the data.
sdrt_jitter	Extent by which standard deviations of individual participants' RTs are larger or smaller than the samplewide SD.
pullfx	size of the effect of approach-versus-avoidance, in milliseconds
pullfx_jitter	Individual variation in the effect of approach-versus-avoidance
stimfx	size of the effect of stimulus category, in milliseconds
stimfx_jitter	Individual variation in the effect of stimulus category
biasfx	Size of the approach bias effect, in milliseconds
biasfx_jitter	Individual variation in the approach bias effect
...	Any parameters of <code>aat_simulate</code> provided here will override the defaults from the <code>defaults</code> parameter.
defaults	Which set of default values should be used?
slowols	Number of slow outliers to insert per participant
fastols	Number of fast outliers to insert per participant
olsd	Number of standard deviations by which (slow) outliers deviate

Details

Defaults of `aat_simulate()` are based on Kahveci, Van Alebeek, Berking, & Blechert (2021).

"Lender2018" parameters are taken from the relevant-feature AAT of Lender, Meule, Rinck, Brockmeyer, & Blechert (2018). "Kahveci2021" parameters are taken from Kahveci, Van Alebeek, Berking, & Blechert (in review).

Lender, A., Meule, A., Rinck, M., Brockmeyer, T., & Blechert, J. (2018). Measurement of food-related approach-avoidance biases: Larger biases when food stimuli are task relevant. *Appetite*, 125, 42-47.

Kahveci, S., Van Alebeek, H., Berking, M., & Blechert, J. (in review). Touchscreen based assessment of food approach biases: investigation of reliability and stimulus-specific effects.

Value

`aat_simulate()` returns a `data.frame` with the following columns: `subj` (participant ID), `stim` (stimulus number), `rep` (stimulus repetition number), `is_pull` (0 = avoid, 1 = approach), `is_target` (0 = control stimulus, 1 = target stimulus), `meanrt` (participant's mean RT), `sdrt` (participant's residual standard deviation), `pullfx` (participant approach-avoidance effect size in ms), `stimfx` (participant stimulus category effect size in ms), `biasfx` (participant approach bias effect size in ms), and `rt` (trial reaction time). Additionally, the `data.frame` has the attribute `population_reliability` which represents the expected reliability of the data given the provided parameters.

Examples

```
ts<- aat_simulate(pullfx = 50, stimfx = 10, biasfx = 100)
mod<-lm(rt~is_pull*is_target,data=ts)
coef(mod) #these should be somewhat close to the provided coefficients
print(attr(ts,"population_reliability"))
print(q_reliability(ts,"subj",rt~is_pull*is_target,"is_pull:is_target"))
#these two should be not too far apart,
# and should converge when the process is repeated a bunch

# Here's how to derive the parameters used in this function from a real dataset
## Not run:
mod<-lmer(decisiontime ~ is_pull * is_food + (is_pull * is_food | subjectid),data=dsa)
fixef(mod) # from here, all the fx and mean RTs are derived
ranef(mod)$subjectid %>% apply(2,sd) #from here, all the fx jitters are derived
dsa %>% group_by(subjectid) %>% summarise(sd=sd(resid)) %>%
summarise(m=mean(sd),s=sd(sd)) # from here, sdrt_jitter is derived

## End(Not run)
hist(aat_simulate2(defaults="Lender2018_raw",slowols=10,fastols=10)$rt)
```

`aat_splithalf`

Compute the bootstrapped split-half reliability for approach-avoidance task data

Description

Compute bootstrapped split-half reliability for approach-avoidance task data.

Usage

```

aat_splithalf(
  ds,
  subjvar,
  pullvar,
  targetvar = NULL,
  rtvar,
  iters,
  algorithm = c("aat_doublemeandiff", "aat_doublemediandiff", "aat_dscore",
    "aat_dscore_multiblock", "aat_regression", "aat_standardregression",
    "aat_doublemedianquotient", "aat_doublemeanquotient", "aat_singlemeandiff",
    "aat_singlemediandiff"),
  trialdropfunc = c("prune_nothing", "trial_prune_3SD", "trial_prune_SD_dropcases",
    "trial_recode_SD", "trial_prune_percent_subject", "trial_prune_percent_sample"),
  errortrialfunc = c("prune_nothing", "error_replace_blockmeanplus",
    "error_prune_dropcases"),
  casedropfunc = c("prune_nothing", "case_prune_3SD"),
  plot = TRUE,
  include.raw = FALSE,
  parallel = TRUE,
  ...
)

## S3 method for class 'aat_splithalf'
print(x, coef = c("Raju", "FlanaganRulon", "SpearmanBrown"), ...)

## S3 method for class 'aat_splithalf'
plot(x, type = c("median", "minimum", "maximum", "random"), ...)

```

Arguments

<code>ds</code>	a longformat data.frame
<code>subjvar</code>	Quoted name of the participant identifier column
<code>pullvar</code>	Quoted name of the column indicating pull trials. Pull trials should either be represented by 1, or by the second level of a factor.
<code>targetvar</code>	Name of the column indicating trials featuring the target stimulus. Target stimuli should either be represented by 1, or by the second level of a factor.
<code>rtvar</code>	Name of the reaction time column.
<code>iters</code>	Total number of desired iterations. At least 200 are recommended for reasonable confidence intervals; If you want to see plots of your data, 1 iteration is enough.
<code>algorithm</code>	Function (without brackets or quotes) to be used to compute AAT scores. See Algorithms for a list of usable algorithms.
<code>trialdropfunc</code>	Function (without brackets or quotes) to be used to exclude outlying trials in each half. The way you handle outliers for the reliability computation should mimic the way you do it in your regular analyses. It is recommended to exclude outlying trials when computing AAT scores using the mean double-difference scores and regression scoring approaches, but not when using d-scores or median double-difference scores. <ul style="list-style-type: none"> • <code>prune_nothing</code> excludes no trials (default)

- `trial_prune_3SD` excludes trials deviating more than 3SD from the mean per participant.
- `trial_prune_SD_dropcases` removes trials deviating more than a specific number of standard deviations from the participant's mean, and removes participants with an excessive percentage of outliers. Required arguments:
 - `trialsd` - trials deviating more than `trialsd` standard deviations from the participant's mean are excluded (optional; default is 3)
 - `maxoutliers` - participants with a higher percentage of outliers are removed from the data. (optional; default is .15)
- `trial_recode_SD` recodes outlying reaction times to the nearest non-outlying value, with outliers defined as reaction times deviating more than a certain number of standard deviations from the participant's mean. Required argument:
 - `trialsd` - trials deviating more than this many standard deviations from the mean are classified as outliers.
- `trial_prune_percent_subject` and `trial_prune_percent_sample` remove trials below and/or above certain percentiles, on a subject-by-subject basis or sample-wide, respectively. The following arguments are available:
 - `lowerpercent` and `upperpercent` (optional; defaults are .01 and .99).

`errortrialfunc` Function (without brackets or quotes) to apply to an error trial.

- `prune_nothing` removes no errors (default).
- `error_replace_blockmeanplus` replaces error trial reaction times with the block mean, plus an arbitrary extra quantity. If used, the following additional arguments are required:
 - `blockvar` - Quoted name of the block variable (mandatory)
 - `errorvar` - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
 - `errorbonus` - Amount to add to the reaction time of error trials. Default is 0.6 (recommended by Greenwald, Nosek, & Banaji, 2003)
- `error_prune_dropcases` removes errors and drops participants if they have more errors than a given percentage. The following arguments are available:
 - `errorvar` - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
 - `maxerrors` - participants with a higher percentage of errors are excluded from the dataset. Default is .15.

`casedropfunc` Function (without brackets or quotes) to be used to exclude outlying participant scores in each half. The way you handle outliers here should mimic the way you do it in your regular analyses.

- `prune_nothing` excludes no participants (default)
- `case_prune_3SD` excludes participants deviating more than 3SD from the sample mean.

`plot` Create a scatterplot of the AAT scores computed from each half of the data from the last iteration. This is highly recommended, as it helps to identify outliers that can inflate or diminish the reliability.

<code>include.raw</code>	logical indicating whether raw split-half data should be included in the output object.
<code>parallel</code>	If TRUE (default), will use parallel computing to compute results faster. If a <code>doParallel</code> backend has not been registered beforehand, this function will register a cluster and stop it after finishing, which takes some extra time.
<code>...</code>	Other arguments, to be passed on to the algorithm or outlier rejection functions (see arguments above)
<code>x</code>	an <code>aat_splithalf</code> object
<code>coef</code>	Optional character argument, indicating which reliability coefficient should be printed. Defaults to Raju's beta.
<code>type</code>	Character argument indicating which iteration should be chosen. Must be an abbreviation of "median" (default), "minimum", "maximum", or "random".

Details

The calculated split-half coefficients are described in Warrens (2016).

Value

A list, containing the mean bootstrapped split-half reliability, bootstrapped 95 a list of data.frames used over each iteration, and a vector containing the split-half reliability of each iteration.

Author(s)

Sercan Kahveci

References

Warrens, M. J. (2016). A comparison of reliability coefficients for psychometric tests that consist of two parts. *Advances in Data Analysis and Classification*, 10(1), 71-84.

See Also

[q_reliability](#)

Examples

```
split <- aat_splithalf(ds=erotica[erotica$is_irrelevant==0,],
  subjvar="subject",pullvar="is_pull",targetvar="is_target",
  rtvar="RT",iters=10,trialdropfunc="trial_prune_3SD",
  casedropfunc="case_prune_3SD",algorithm="aat_dscore",
  plot=FALSE,parallel=FALSE)

print(split)
#Mean reliability: 0.521959
#Spearman-Brown-corrected r: 0.6859041
#95%CI: [0.4167018, 0.6172474]

plot(split)

#Regression Splithalf
```

```

aat_splithalf(ds=erotica[erotica$is_irrelevant==0,],
  subjvar="subject", pullvar="is_pull", targetvar="is_target",
  rtvar="RT", iters=10, trialdropfunc="trial_prune_3SD",
  casedropfunc="case_prune_3SD", algorithm="aat_regression",
  formula = RT ~ is_pull * is_target, aatterm = "is_pull:is_target",
  plot=FALSE, parallel=FALSE)
#Mean reliability: 0.5313939
#Spearman-Brown-corrected r: 0.6940003
#95%CI: [0.2687186, 0.6749176]

```

Algorithms

AAT score computation algorithms

Description

- `aat_doublemeandiff` computes a mean-based double-difference score:
 $(\text{mean}(\text{push_target}) - \text{mean}(\text{pull_target})) - (\text{mean}(\text{push_control}) - \text{mean}(\text{pull_control}))$
- `aat_doublemediandiff` computes a median-based double-difference score:
 $(\text{median}(\text{push_target}) - \text{median}(\text{pull_target})) - (\text{median}(\text{push_control}) - \text{median}(\text{pull_control}))$
- `aat_dscore` computes D-scores for a 2-block design (see Greenwald, Nosek, and Banaji, 2003):
 $((\text{mean}(\text{push_target}) - \text{mean}(\text{pull_target})) - (\text{mean}(\text{push_control}) - \text{mean}(\text{pull_control}))) / \text{sd}(\text{participant_reaction_times})$
- `aat_dscore_multiblock` computes D-scores for pairs of sequential blocks and averages the resulting score (see Greenwald, Nosek, and Banaji, 2003). Requires extra `blockvar` argument, indicating the name of the block variable.
- `aat_regression` and `aat_standardregression` fit regression models to participants' reaction times and extract a term that serves as AAT score. `aat_regression` extracts the raw coefficient, equivalent to a mean difference score. `aat_standardregression` extracts the t-score of the coefficient, standardized on the basis of the variability of the participant's reaction times. These algorithms can be used to regress nuisance variables out of the data before computing AAT scores. When using these functions, additional arguments must be provided:
 - `formula` - a formula to fit to the data
 - `aatterm` - the term within the formula that indicates the approach bias; this is usually the interaction of the pull and target terms.
- `aat_doublemeanquotient` and `aat_doublemedianquotient` compute a log-transformed ratio of approach to avoidance for both stimulus categories and subtract these ratios:
 $\log(\text{mean}(\text{pull_target}) / \text{mean}(\text{push_target})) - \log(\text{mean}(\text{pull_control}) / \text{mean}(\text{push_control}))$
- `aat_singlemeandiff` and `aat_singlemediandiff` subtract the mean or median approach reaction time from the mean or median avoidance reaction time. These algorithms are only sensible if the supplied data contain a single stimulus category.

Usage

```

aat_doublemeandiff(ds, subjvar, pullvar, targetvar, rtvar, ...)

aat_doublemediandiff(ds, subjvar, pullvar, targetvar, rtvar, ...)

aat_dscores(ds, subjvar, pullvar, targetvar, rtvar, ...)

aat_dscores_multiblock(ds, subjvar, pullvar, targetvar, rtvar, blockvar, ...)

aat_regression(ds, subjvar, formula, aatterm, ...)

aat_standardregression(ds, subjvar, formula, aatterm, ...)

aat_doublemedianquotient(ds, subjvar, pullvar, targetvar, rtvar, ...)

aat_doublemeanquotient(ds, subjvar, pullvar, targetvar, rtvar, ...)

aat_singlemeandiff(ds, subjvar, pullvar, rtvar, ...)

aat_singlemediandiff(ds, subjvar, pullvar, rtvar,...)

```

Arguments

<code>ds</code>	A long-format data.frame
<code>subjvar</code>	Column name of the participant identifier variable
<code>pullvar</code>	Column name of the movement variable (0: avoid; 1: approach)
<code>targetvar</code>	Column name of the stimulus category variable (0: control stimulus; 1: target stimulus)
<code>rtvar</code>	Column name of the reaction time variable
<code>...</code>	Other arguments passed on by functions (ignored)
<code>blockvar</code>	name of the variable indicating block number
<code>formula</code>	A regression formula to fit to the data to compute an AAT score
<code>aatterm</code>	A character naming the formula term representing the approach bias. Usually this is the interaction of the movement-direction and stimulus-category terms.

Value

A data.frame containing participant number and computed AAT score.

erotica	<i>AAT examining approach bias for erotic stimuli</i>
---------	---

Description

AAT

Usage

```
erotica
```

Format

An object of class "data.frame"

Source

[osf.io repository](#)

References

Kahveci, S., Van Bockstaele, B.D., & Wiers, R.W. (in preparation). Pulling for Pleasure? Erotic Approach-Bias Associated With Porn Use, Not Problems. DOI:10.17605/OSF.IO/6H2RJ

Preprocessing

Pre-processing rules

Description

These are pre-processing rules that can be used in [aat_splithalf](#), [aat_bootstrap](#), and [aat_compute](#).

- The following rules are to be used for the `trialdropfunc` argument. The way you handle outliers for the reliability computation and bootstrapping more broadly should mimic the way you do it in your regular analyses. It is recommended to exclude outlying trials when computing AAT scores using the mean double-difference scores and regression scoring approaches, but not when using d-scores or median double-difference scores.
 - `prune_nothing` excludes no trials (default)
 - `trial_prune_3SD` excludes trials deviating more than 3SD from the mean per participant.
 - `trial_prune_SD_dropcases` removes trials deviating more than a specific number of standard deviations from the participant's mean, and removes participants with an excessive percentage of outliers. Required arguments:
 - * `trialsd` - trials deviating more than `trialsd` standard deviations from the participant's mean are excluded (optional; default is 3)
 - * `maxoutliers` - participants with a higher percentage of outliers are removed from the data. (optional; default is .15)
 - `trial_recode_SD` recodes outlying reaction times to the nearest non-outlying value, with outliers defined as reaction times deviating more than a certain number of standard deviations from the participant's mean. Required argument:
 - * `trialsd` - trials deviating more than this many standard deviations from the mean are classified as outliers.
 - `trial_prune_percent_subject` and `trial_prune_percent_sample` remove trials below and/or above certain percentiles, on a subject-by-subject basis or sample-wide, respectively. The following arguments are available:
 - * `lowerpercent` and `upperpercent` (optional; defaults are .01 and .99).
- The following pre-processing rules are to be used for the `errortrialfunc` argument. They determine what is to be done with errors - remove or recode?
 - `prune_nothing` removes no errors (default).

- `error_replace_blockmeanplus` replaces error trial reaction times with the block mean, plus an arbitrary extra quantity. If used, the following additional arguments are required:
 - * `blockvar` - Quoted name of the block variable (mandatory)
 - * `errorvar` - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
 - * `errorbonus` - Amount to add to the reaction time of error trials. Default is 0.6 (recommended by Greenwald, Nosek, & Banaji, 2003)
- `error_prune_dropcases` removes errors and drops participants if they have more errors than a given percentage. The following arguments are available:
 - * `errorvar` - Quoted name of the error variable, where errors are 1 or TRUE and correct trials are 0 or FALSE (mandatory)
 - * `maxerrors` - participants with a higher percentage of errors are excluded from the dataset. Default is .15.
- These are pre-processing rules to be used for the `casedropfunc` argument. The way you handle outliers here should mimic the way you do it in your regular analyses.
 - `prune_nothing` excludes no participants (default)
 - `case_prune_3SD` excludes participants deviating more than 3SD from the sample mean.

Usage

```
prune_nothing(ds, ...)

trial_prune_percent_subject(
  ds,
  subjvar,
  rtvar,
  lowerpercent = 0.01,
  upperpercent = 0.99,
  ...
)

trial_prune_percent_sample(
  ds,
  rtvar,
  lowerpercent = 0.01,
  upperpercent = 0.99,
  ...
)

trial_prune_3SD(ds, subjvar, rtvar, ...)

trial_prune_SD_dropcases(
  ds,
  subjvar,
  rtvar,
  trialsd = 3,
  maxoutliers = 0.15,
  ...
)
```

```

trial_recode_SD(ds, subjvar, rtvar, trialsd = 3, ...)

case_prune_3SD(ds, ...)

error_replace_blockmeanplus(
  ds,
  subjvar,
  rtvar,
  blockvar,
  errorvar,
  errorbonus,
  ...
)

error_prune_dropcases(ds, subjvar, errorvar, maxerrors = 0.15, ...)

```

Arguments

<code>ds</code>	A data.frame.
<code>...</code>	Other arguments (ignored).
<code>subjvar</code>	The name of the subject variable.
<code>rtvar</code>	The name of the reaction time variable.
<code>lowerpercent, upperpercent</code>	for <code>trial_prune_percent_subject</code> and <code>trial_prune_percent_sample</code> , the lower and upper proportions beyond which trials are considered outliers and removed (defaults to .01 and .99).
<code>trialsd</code>	The amount of deviation from the participant mean (in SD) after which a trial is considered an outlier and excluded (defaults to 3).
<code>maxoutliers</code>	for <code>trial_prune_SD_dropcases</code> , the maximum percentage of outliers, after which a participant is excluded from the data.
<code>blockvar</code>	The name of the block variable.
<code>errorvar</code>	The name of the error variable.
<code>errorbonus</code>	for <code>error_replace_blockmeanplus</code> , the amount of seconds to add to the block mean and use as a replacement for error trial reaction times (default is 0.6).
<code>maxerrors</code>	for <code>error_prune_dropcases</code> , the maximum percentage of errors, after which a participant is excluded from the data.

q_reliability

Compute psychological experiment reliability

Description

This function can be used to compute an exact reliability score for a psychological task whose results involve a difference score. The resulting intraclass correlation coefficient is equivalent to the average all possible split-half reliability scores. It ranges from -1 to 1, with -1 implying that all variance in the data is explained by within-subjects variability, 1 implying that all variance is explained by between-subjects variability, and 0 implying that within-subjects and between-subjects variability contribute equally to the total variance in the sample.

Usage

```

q_reliability(ds, subjvar, formula, aatterm = NA)

q_reliability2(ds, subjvar, splitvars, rtvar, na.rm = F)

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

## S3 method for class 'qreliability'
plot(x, ...)

```

Arguments

<code>ds</code>	a long-format data.frame
<code>subjvar</code>	name of the subject variable
<code>formula</code>	a formula predicting the participant's reaction time using trial-level variables such as movement direction and stimulus category
<code>aatterm</code>	a string denoting the term in the formula that contains the participant's approach bias
<code>splitvars</code>	Vector of column names over which to split the data to compute difference scores. This can be used to compute the reliability of single, double, or even triple difference scores.
<code>rtvar</code>	Column name of the variable containing reaction times
<code>na.rm</code>	If true, remove rows with missing values from the data
<code>x</code>	a qreliability object
<code>...</code>	Other arguments passed to the generic <code>print</code> and <code>plot</code> functions.

Value

a qreliability object, containing the reliability coefficient, and a data.frame with participants' bias scores and score variance.

Please note that the valence of the bias scores may or may not correspond with approach and avoidance. If you plan to use these scores in your analyses, always verify that they are in the right direction by correlating them with independently calculated bias scores, for example using `aat_compute()`.

Author(s)

Sercan Kahveci

Examples

```

# Double-difference score reliability
q_reliability(ds=erotica,subjvar="subject",
              formula= RT ~ is_pull * is_target, aatterm = "is_pull:is_target")

# Single-difference reliability for target stimuli
q_reliability(ds=erotica[erotica$is_target ==1,],subjvar="subject",
              formula= RT ~ is_pull, aatterm = "is_pull")

# Reliability of the mean reaction time of approaching target stimuli (no difference score)

```

```
q_reliability(ds=erotica[erotica$is_target ==1 & erotica$is_pull ==1,],subjvar="subject",
             formula= RT ~ 1, aatterm = "1")

q_reliability2(ds=erotica,subjvar="subject",
              splitvars=c("is_pull", "is_target"),rtvar="RT")
```

reliability-coefficients

Correct a correlation coefficient for being based on only a subset of the data

Description

Correct a correlation coefficient for being based on only a subset of the data

Usage

```
SpearmanBrown(
  corr,
  ntests = 2,
  fix.negative = c("nullify", "bilateral", "none")
)

FlanaganRulon(
  x1,
  x2,
  x = NULL,
  fix.negative = c("nullify", "bilateral", "none")
)

RajuCoefficient(x1, x2, prop, fix.negative = c("nullify", "bilateral", "none"))
```

Arguments

<code>corr</code>	To-be-corrected correlation coefficient
<code>ntests</code>	An integer indicating how many times larger the full test is, for which the corrected correlation coefficient is being computed. When <code>ntests=2</code> , the formula will compute what the correlation coefficient would be if the test were twice as long.
<code>fix.negative</code>	Determines how to deal with a negative value. "nullify" sets it to zero, "bilateral" applies the correction as if it were a positive number, and then sets it to negative. "none" gives the raw value. It should be noted that negative values are not supposed to occur, and there is no commonly accepted way to deal with them when they do occur.
<code>x1</code>	scores from half 1
<code>x2</code>	scores from half 2
<code>x</code>	scores from the full sample (more accurate if provided)
<code>prop</code>	Proportion of the first half to the complete sample

Value

Spearman-Brown-corrected correlation coefficient.

Functions

- **SpearmanBrown**: Perform a Spearman-Brown correction on the provided correlation score.
- **FlanaganRulon**: Compute the true reliability using the Flanagan-Rulon formula, which takes into account unequal variances between split halves
- **RajuCoefficient**: Compute split-half reliability using the Raju formula, which takes into account unequal split-halves and variances.

Examples

```
SpearmanBrown(.5)
FlanaganRulon(a<-rnorm(50),rnorm(50)+a*.5,fix.negative="bilateral")
a<-rnorm(50)
b<-rnorm(50)+a*.5
RajuCoefficient(a,b,prop=.4,fix.negative="bilateral")
```

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