Package 'neatStats'

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Title An R Package for Neat and Painless Statistical Reporting

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Description This package focuses on user-friendly, clear and simple statistics, primarily for publication in psychological science. The main functions are wrappers for other packages, but there are various additions as well. Every relevant step from data aggregation to reportable printed statistics is covered for basic experimental designs.	
<pre>URL https://github.com/gasparl/neatstats</pre>	
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anova_neat

Comparison of Multiple Means: ANOVA

Description

Analysis of variance (ANOVA) F-test results with appropriate Welch's and epsilon corrections where applicable (unless specified otherwise), including partial eta squared effect sizes with confidence intervals (CIs), and inclusion Bayes factor based on matched models (BFs).

Usage

```
anova_neat(data_per_subject, values, between_vars = NULL,
 within_ids = NULL, ci = 0.9, bf_added = TRUE,
  test_title = "--- neat ANOVA ---", welch = TRUE,
 e_correction = NULL)
```

Arguments

data_per_subject

Data frame or name of data frame as string. Should contain all values (measurements/observations) in a single row per each subject.

values

String; column name or names. Multiple column names are also to be given as a single string, separated by commas (e.g., values = 'var1, var2, var3'). (Spaces are ignored.) Each such column should contain a single dependent variable. This means, to test repeated (within-subject) measurements, each specified column should contain one measurement.

between_vars

NULL (default; in case of no between-subject factors) or string; column name or names. Multiple column names are also to be given as a single string, separated by commas (e.g., between_vars = 'grouping1, grouping2'). (Spaces are ignored.) Each such column should contain a single between-subject independent variable (representing between-subject factors).

within_ids

NULL (default), string, or named list. In case of no within-subject factors, leave as NULL. In case of a single within subject factor, a single string may be given to optionally provide custom name for the within-subject factor (note: this is a programming variable name, so it should not contain spaces, etc.); otherwise (if left NULL) this one within-subject factor will always just be named "within_factor". In case of multiple within-subject factors, each factor must be specified as a named list element, each with a vector of strings that distinguish the levels within that factors. The column names given as values should always contain one (and only one) of these strings within each within-subject factor, and thus they will be assigned the appropriate level. For example, values = 'rt_s1_neg,rt_s1_pos,rt_s2_neg,rt_s2_pos' could have within_ids = list(session = c('s1', 's2'), valence = c('pos', 'neg'). (Note: the strings for distinguishing must be unambigous. E.g., for values apple_a and apple_b, do not set levels c('a', 'b'), because 'a' is also found in apple_b. In this case, you could choose levels c('_a', '_b') to make sure the values are correctly distinguished.) See also Examples.

Numeric; confidence level for returned CIs. (Default: .9; Lakens, 2014; Steiger, 2004.)

ci

bf_added Logical. If TRUE (default), inclusion Bayes factor is calculated and displayed.

(Note: with multiple factors and/or larger dataset, the calculation can take con-

siderable time.)

test_title String, "---neat ANOVA ---" by default. Simply displayed in printing preced-

ing the statistics.

welch If TRUE (default), calculates Welch's ANOVA via stats::oneway.test in case

of a single factor (one-way) between-subject design. If FALSE, calculates via ez::ezANOVA in such cases too (i.e., same as in case of every other design).

e_correction NULL (default) or one of the following strings: 'gg', 'hf', or 'none'. If set

to 'gg', Greenhouse-Geisser correction is applied in case of repeated measures (regardless of violation of sphericity). If set to 'hf', Huynh-Feldt correction is applied. If set to 'none', no correction is applied. If NULL, Greenhouse-Geisser correction is applied when Mauchly's sphericity test is significant and the Greenhouse-Geisser epsilon is not larger than .75, while Huynh-Feldt correction is applied when Mauchly's sphericity test is significant and the Greenhouse-

Geisser epsilon is larger than .75 (see Girden, 1992).

Details

The Bayes factor (BF) is always calculated with the default rscaleFixed of 0.5 ("medium") and rscaleRandom of 1 ("nuisance"). BF supporting null hypothesis is denoted as BF01, while that supporting alternative hypothesis is denoted as BF10. When the BF is smaller than 1 (i.e., supports null hypothesis), the reciprocal is calculated (hence, BF10 = BF, but BF01 = 1/BF). When the BF is greater than or equal to 10000, scientific (exponential) form is reported for readability. (The original full BF number is available in the returned named vector as bf.)

Levene's test is returned for between-subject design without Welch's correction, while Mauchly's sphericity test is returned for repeated measures with more than two levels. If Mauchly's test is significant, epsilon correction may be applied (see the e_correction parameter). If Levene's test is significant and the data is unbalanced (unequal group sample sizes), you should either consider the results respectively or choose a different test.

Value

Prints ANOVA statistics (including, for each model, F-test with partial eta squared and its CI, and BF, as specified via the corresponding parameters) in APA style. Furthermore, when assigned, returns a list with up to three elements. First, 'stat_list', a list of named vectors per each effect (main or interaction). Each vector contains the following elements: F (F value), p (p value), petas (partial eta squared), epsilon (epsilon used for correction), and bf (inclusion BF; when bf_added is not FALSE). Second, the ezANOVA object, named ez_anova (calculated even when oneway.test is printed). Third, when bf_added is not FALSE, the anovaBF object, named bf_models; including all models on which the inclusion BFs are based.

Note

All F-tests are calculated via ez::ezANOVA, including Levene's test and Mauchly's sphericity test. (But Welch's ANOVA is calculated in case of one-way between-subject designs via stats::oneway.test, unless the welch parameter is set to FALSE.)

Confidence intervals are calculated, using the F value, via MBESS::conf.limits.ncf, coverting noncentrality parameters to partial eta squared as ncp/(ncp+ df_nom+df_denom+1) (Smithson, 2003).

The inclusion Bayes factor based on matched models is calculated via bayestestR::bayesfactor_inclusion, (with match_models = TRUE, and using an BayesFactor::anovaBF object for models input).

References

Girden, E. (1992). ANOVA: Repeated measures. Newbury Park, CA: Sage.

Kelley, K. (2007). Methods for the behavioral, educational, and social sciences: An R package. Behavior Research Methods, 39(4), 979-984. doi: 10.3758/BF03192993

Lakens, D. (2014). Calculating confidence intervals for Cohen's d and eta-squared using SPSS, R, and Stata [Blog post]. Retrieved from http://daniellakens.blogspot.com/2014/06/calculating-confidence-inhtml

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McDonald, J. H. 2015. Handbook of Biological Statistics (3rd ed.). Sparky House Publishing, Baltimore, Maryland. Retrieved from http://www.biostathandbook.com

Moder, K. (2010). Alternatives to F-test in one way ANOVA in case of heterogeneity of variances (a simulation study). Psychological Test and Assessment Modeling, 52(4), 343-353.

Navarro, D. (2013). Learning Statistics with R: A Tutorial for Psychology Students and Other Beginners (Version 0.6.1). Retrieved from https://learningstatisticswithr.com/

Smithson, M. (2003). Confidence intervals. Thousand Oaks, Calif: Sage Publications.

Steiger, J. H. (2004). Beyond the F test: effect size confidence intervals and tests of close fit in the analysis of variance and contrast analysis. Psychological Methods, 9(2), 164-182. doi: 10.1037/1082989X.9.2.164

See Also

t_neat

```
# assign random data in a data frame for illustration
# (note that the 'subject' is only for illustration; since each row contains the
# data of a single subject, no additional subject id is needed)
dat_1 = data.frame(
   subject = c(1, 2, 3, 4, 5, 6, 7, 8, 9, 10),
   grouping1 = c(1, 1, 1, 1, 2, 2, 2, 2, 2, 2),
   grouping2 = c(1, 2, 1, 2, 2, 1, 1, 1, 2, 1),
   value_1_a = c(36.2, 45.2, 41, 24.6, 30.5, 28.2, 40.9, 45.1, 31, 16.9),
   value_2_a = c(-14.1, 58.5, -25.5, 42.2, -13, 4.4, 55.5, -28.5, 25.6, -37.1),
   value_1_b = c(83, 71, 111, 70, 92, 75, 110, 111, 110, 85),
   value_2b = c(8.024, -14.162, 3.1, -2.1, -1.5, 0.91, 11.53, 18.37, 0.3, -0.59),
   value_1_c = c(27.4, -17.6, -32.7, 0.4, 37.2, 1.7, 18.2, 8.9, 1.9, 0.4),
   value_2c = c(7.7, -0.8, 2.2, 14.1, 22.1, -47.7, -4.8, 8.6, 6.2, 18.2)
head(dat_1) # see what we have
# For example, numbers '1' and '2' in the variable names of the values can
# denote sessions in an experiment, such as '_1' for first session, and '_2 for
\# second session'. The letters '_a', '_b', '_c' could denote three different
\# types of techniques used within each session, to be compared to each other.
# See further below for a more verbose but more meaningful example data.
# get the between-subject effect of 'grouping1'
anova_neat('dat_1', values = 'value_1_a', between_vars = 'grouping1')
# main effects of 'grouping1', 'grouping2', and their interactions
```

```
anova_neat('dat_1', values = 'value_1_a', between_vars = 'grouping1, grouping2')
# repeated measures:
# get the within-subject effect for 'value_1_a' vs. 'value_1_b'
anova_neat('dat_1', values = 'value_1_a, value_1_b')
# same, but give the factor a custom variable name
anova_neat('dat_1', values = 'value_1_a, value_1_b', within_ids = 'a_vs_b')
anova_neat('dat_1', values = 'value_1_a, value_1_b', within_ids = 'letters')
# within-subject effect for 'value_1_a' vs. 'value_1_b' vs. 'value_1_c'
anova_neat('dat_1', values = 'value_1_a, value_1_b, value_1_c')
# within-subject main effect for 'value_1_a' vs. 'value_1_b' vs. 'value_1_c',
# between-subject main effect 'grouping1', and the interaction of these two main
# effects
anova_neat('dat_1', values = 'value_1_a, value_1_b, value_1_c', between_vars = 'grouping1')
# within-subject 'number' main effect for variables with number '1' vs. number
# '2' ('value_1_a' and 'value_1_b' vs. 'value_2_a' and 'value_2_b'), 'letter'
# main effect for variables with final letterr 'a' vs. final letter 'b'
# ('value_1_a' and 'value_2_a' vs. 'value_1_b' and 'value_2_b'), and the
# 'letter' x 'number' interaction
anova_neat('dat_1',
           values = 'value_1_a, value_2_a, value_1_b, value_2_b',
           within_ids = list(
              letters = c('_a', '_b'),
               numbers = c('_1', '_2')
           ))
# same as above, but now including between-subject main effect 'grouping2' and
# its interactions
anova_neat(
    'dat_1',
   values = 'value_1_a, value_2_a, value_1_b, value_2_b',
   within_ids = list(
       letters = c('_a', '_b'),
       numbers = c('_1', '_2')
   ).
   between_vars = 'grouping2'
)
# In real datasets, these could of course be more meaningful. For example, let's
# say participants rated the attractiveness of pictures with low or high levels
# of frightening and low or high levels of disgusting qualities. So there are
# four types of ratings:
# 'low disgusting, low frightening' pictures
# 'low disgusting, high frightening' pictures
# 'high disgusting, low frightening' pictures
# 'high disgusting, high frightening' pictures
# this could be meaningfully assigned e.g. as below
pic_ratings = data.frame(
    subject = c(1, 2, 3, 4, 5, 6, 7, 8, 9, 10),
  rating_fright_low_disgust_low = c(36.2, 45.2, 41, 24.6, 30.5, 28.2, 40.9, 45.1, 31, 16.9),
  rating_fright_high_disgust_low = c(-14.1, 58.5, -25.5, 42.2, -13, 4.4, 55.5, -28.5, 25.6, -37.1),
```

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```
rating_fright_low_disgust_high = c(83, 71, 111, 70, 92, 75, 110, 111, 110, 85),
  rating_fright_high_disgust_high = c(8.024, -14.162, 3.1, -2.1, -1.5, 0.91, 11.53, 18.37, 0.3, -0.59)
head(pic_ratings) # see what we have
# the same logic applies as for the examples above, but now the
# within-subject differences can be more meaningfully specified, e.g.
# 'disgust_low' vs. 'disgust_high' for levels of disgustingness, while
# 'fright_low' vs. 'fright_high' for levels of frighteningness
anova_neat('pic_ratings',
           values = 'rating_fright_low_disgust_low, rating_fright_high_disgust_low,
               rating_fright_low_disgust_high, rating_fright_high_disgust_high',
           within_ids = list(
               disgustingness = c('disgust_low', 'disgust_high'),
               frighteningness = c('fright_low', 'fright_high')
           ))
# the results are the same as for the analogous test for the 'dat_1' data, only
# with different names
# now let's say the ratings were done in two separate groups
pic_ratings = data.frame(
    subject = c(1, 2, 3, 4, 5, 6, 7, 8, 9, 10),
    group_id = c(1, 2, 1, 2, 2, 1, 1, 1, 2, 1),
  rating_fright_low_disgust_low = c(36.2, 45.2, 41, 24.6, 30.5, 28.2, 40.9, 45.1, 31, 16.9),
  rating_fright_high_disgust_low = c(-14.1, 58.5,-25.5, 42.2,-13, 4.4, 55.5,-28.5, 25.6,-37.1),
    rating_fright_low_disgust_high = c(83, 71, 111, 70, 92, 75, 110, 111, 110, 85),
  rating_fright_high_disgust_high = c(8.024, -14.162, 3.1, -2.1, -1.5, 0.91, 11.53, 18.37, 0.3, -0.59)
)
# now test the effect and interactions of 'group_id'
anova_neat(
    'pic_ratings',
    values = 'rating_fright_low_disgust_low, rating_fright_high_disgust_low,
        rating_fright_low_disgust_high, rating_fright_high_disgust_high',
    within_ids = list(
        disgustingness = c('disgust_low', 'disgust_high'),
        frighteningness = c('fright_low', 'fright_high')
    ),
    between_vars = 'group_id'
)
# again, same results as with 'dat_1' (using 'grouping2' as group_id)
```

corr_neat

Correlation Statistics

Description

Pearson correlation results including confidence interval (CI) and correlation Bayes factor (BF).

Usage

```
corr_neat(var1, var2, ci = 0.95, bf_added = TRUE, direction = NULL,
  round_r = 3, for_table = FALSE)
```

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Arguments

var1	Numeric vector; numbers of the first variable.
var2	Numeric vector; numbers of the second variable.
ci	Numeric; confidence level for the returned CI, as implemented in cor.test.
bf_added	Logical. If TRUE (default), Bayes factor is calculated and displayed.
direction	NULL or string; optionally specifies one-sided test: either "negative" (negative correlation expected) or "positive" (positive correlation expected). (Short forms also work, e.g. "p", "pos", "neg", etc.) If NULL (default), the test is two-sided.
round_r	Number to round to the correlation and its CI.
for_table	Logical. If TRUE, omits the confidence level display from the printed text.

Details

The Bayes factor (BF) is always calculated with the default r-scale of \emptyset . 707. BF supporting null hypothesis is denoted as BF01, while that supporting alternative hypothesis is denoted as BF10. When the BF is smaller than 1 (i.e., supports null hypothesis), the reciprocal is calculated (hence, BF10 = BF, but BF01 = 1/BF). When the BF is greater than or equal to 10000, scientific (exponential) form is reported for readability. (The original full BF number is available in the returned named vector as bf.)

Value

Prints correlation statistics (including CI and BF) in APA style. Furthermore, when assigned, returns a named vector with the following elements: r (Pearson correlation), p (p value), bf (Bayes factor).

Note

```
The correlation and CI is calculated via stats::cor.test.

The Bayes factor is calculated via BayesFactor::correlationBF.
```

See Also

t_neat

```
# assign two variables
v1 = c(11, 15, 19, 43, 53, -4, 34, 8, 33, -1, 54)
v2 = c(4, -2, 23, 13, 32, 16, 3, 29, 37, -4, 65)

corr_neat(v1, v2) # prints statistics

# one-sided, and omitting the "95% CI" part
corr_neat(v1, v2, direction = 'pos', for_table = TRUE)

# print statistics and assign main results
results = corr_neat(v1, v2, direction = 'pos')

results['p'] # get precise p value
```

8 mon_conv

dems_neat

Demographics

Description

Prints participant count, age mean and SD, and gender ratio, from given dataset.

Usage

```
dems_neat(data_per_subject, group_by = NULL, percent = TRUE,
  round_perc = 0)
```

Arguments

data_per_subject

Data frame from which demographics are to be calculated. Must contain columns named precisely as "age" and as "gender". The age column must contain numbers, while gender column must contain 1 (= male) and 2 (= female) only (either

as numbers or as strings).

group_by A vector of factors by which the statistics are grouped, typically a column from

the data frame provided as data_per_subject.

percent Logical. If TRUE (default), gender ratio is presented as percent of males. If

FALSE, presented as count of males.

round_perc Number to round to, when using percents.

Examples

mon_conv

Monitor Screen Unit Conversion

Description

Given a specific monitor object, converts specified screen units to other specified units. The possible units to convert from and to: "cm" (centimeters), "pix" (pixels), or "deg" (degrees of visual angle).

Usage

```
mon_conv(mon_obj, value, from, to)
```

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Arguments

mon_obj Monitor object, as assigned with mon_neat.

value Number; value (magnitude) of the given unit to convert from.

from String; unit ("cm", "pix", or "deg") to convert from.

to String; unit ("cm", "pix", or "deg") to convert to.

Value

Number (magnitude) in the given output (to) unit.

See Also

```
mon_neat, mon_params
```

Examples

```
# assign monitor with 50 cm distance, screen width 52 cm and 1920 pixels
my_mon = mon_neat(distance = 50, mon_width_cm = 52, mon_width_pixel = 1920)
# convert 30.4 pixels to degrees of visual angle, for the specified monitor
mon_conv(my_mon, 30.4, 'pix', 'deg') # returns 0.9434492 (degrees)
# convert 0.94 degrees of visual angle to pixels
mon_conv(my_mon, 0.94, 'deg', 'pix') # returns 30.28885 (pixels)
# convert 10 degrees of visual angle to cm
mon_conv(my_mon, 10, from = 'deg', to = 'cm')
# convert 8.748866 cm to pixels
mon_conv(my_mon, 8.748866, from = 'cm', to = 'pix')
```

mon_neat

Monitor Object

Description

Assigns a monitor object, storing distance and width parameters.

Usage

```
mon_neat(distance, mon_width_cm, mon_width_pixel)
```

Arguments

```
distance Viewing distance in cm (from eyes to screen).

mon_width_cm Monitor screen width in cm.

mon_width_pixel Monitor screen width in pixels.
```

Value

A monitor object with the specified parameters, to be used in the mon_conv function.

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See Also

```
mon_conv, mon_params
```

Examples

```
\# assign monitor with 57 cm viewing distance, screen width 52 cm and 1920 pixels my_mon = mon_neat(distance = 57, mon_width_cm = 52, mon_width_pixel = 1920)
```

mon_params

Monitor parameters

Description

Prints the parameters of a given monitor object.

Usage

```
mon_params(mon_obj)
```

Arguments

mon_obj

Monitor object, as assigned with mon_neat.

Value

Returns nothing, just prints all parameters.

See Also

```
mon_neat, mon_conv
```

```
# assign monitor with 57 cm distance, and screen width 52 cm and 1920 pixels
my_mon = mon_neat(distance = 57, mon_width_cm = 52, mon_width_pixel = 1920)
# print the above given parameters
mon_params( my_mon )
```

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m_neat Neat Descriptives

Description

Returns means (or medians) and SDs per group for given variable. Primarily for use in the table_neat function.

Usage

```
m_neat(values, round_to = 0, new_name = NULL, group_by = NULL,
    medians = FALSE)
```

Arguments

values	A vector of numbers from which the statistics are to be calculated.
round_to	Number of digits after the decimal point to round to.
new_name	NULL (default) or string. String specifies new name for the variable to be used as column title. If NULL, the input variable name is used.
group_by	A vector of factors by which the statistics are grouped.
medians	Logical. If TRUE, medians are calculated, otherwise (default: FALSE) means.

Value

A data frame with the statistics per group. Furthermore, prints statistics per group, unless used within the table_neat function, in which case nothing is printed.

See Also

table_neat to create full tables using multiple variables

```
data("mtcars") # load base R example dataset

res1 = m_neat(mtcars$wt ) # overall means and SDs for wt (Weight)

res2 = m_neat(mtcars$wt, 2, new_name = 'weight') # rounded to 2 and renamed

res3 = m_neat(mtcars$wt, 2, group_by = mtcars$cyl) # grouped by cyl (Number of cylinders)

res4 = m_neat(mtcars$wt, 2, group_by = mtcars$cyl, medians = TRUE) # medians
```

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|--|

Description

Unconditional exact test results for the comparison of two independent proportions, including confidence interval (CI) for the proportion difference, and corresponding independent multinomial contingency table Bayes factor (BF). Cohen's h and its CI are also calculated.

Usage

```
props_neat(case1, case2, n1, n2, greater = NULL, ci = NULL,
    bf_added = TRUE, h_added = FALSE, for_table = FALSE)
```

Arguments

case1	Number of 'cases' (as opposed to 'controls'; e.g. positive outcomes vs. negative outcomes) in 'group 1'.
case2	Number of 'cases' in 'group 2'.
n1	Number; sample size of 'group 1'.
n2	Number; sample size of 'group 2'.
greater	NULL or string (or number); optionally specifies one-sided exact test: either "1" (case1/n1 proportion expected to be greater than case2/n2 proportion) or "2" (case2/n2 proportion expected to be greater than case1/n1 proportion). If NULL (default), the test is two-sided.
ci	Numeric; confidence level for the returned CIs (proportion difference and Cohen's h).
bf_added	Logical. If TRUE (default), Bayes factor is calculated and displayed. (Always two-sided.)
h_added	Logical. If TRUE, Cohen's h and its CI are calculated and displayed. (FALSE by default.)
for_table	Logical. If TRUE, omits the confidence level display from the printed text.

Details

The Bayes factor (BF) is always calculated with the default r-scale of 0.707. BF supporting null hypothesis is denoted as BF01, while that supporting alternative hypothesis is denoted as BF10. When the BF is smaller than 1 (i.e., supports null hypothesis), the reciprocal is calculated (hence, BF10 = BF, but BF01 = 1/BF). When the BF is greater than or equal to 10000, scientific (exponential) form is reported for readability. (The original full BF number is available in the returned named vector as bf.)

Value

Prints exact test statistics (including proportion difference with CI, and BF) in APA style. Furthermore, when assigned, returns a named vector with the following elements: z (Z), p (p value), prop_diff (raw proportion difference), h (Cohen's h), bf (Bayes factor).

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Note

Barnard's unconditional exact test is calculated via Exact::exact.test ("z-pooled").

The CIs for the proportion difference is calculated based on the p value, as described by Altman and Bland (2011).

The Bayes factor is calculated via BayesFactor::contingencyTableBF, with sampleType = "indepMulti", as appropriate when both sample sizes (n1 and n2) are known in advance (as it normally happens). (For details, see contingencyTableBF, or e.g. 'Chapter 17 Bayesian statistics' in Navarro, 2019.)

References

Altman, D. G., & Bland, J. M. (2011). How to obtain the confidence interval from a P value. Bmj, 343(d2090). doi: 10.1136/bmj.d2090

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Navarro, D. (2019). Learning statistics with R. https://learningstatisticswithr.com/

Suissa, S., & Shuster, J. J. (1985). Exact unconditional sample sizes for the 2 times 2 binomial trial. Journal of the Royal Statistical Society: Series A (General), 148(4), 317-327. doi: 10.2307/2981892

Examples

```
props_neat(
    case1 = 35,
    case2 = 48,
    n1 = 80,
    n2 = 77,
    h_added = TRUE
)

props_neat(
    case1 = 35,
    case2 = 48,
    n1 = 80,
    n2 = 77,
    greater = "2"
)
```

ro

Neat rounding

Description

Rounds a given number to given number of digits after the decimal point, returning it as string, with trailing zeros when applicable.

Usage

```
ro(num, round_to = 2)
```

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Arguments

num Number to be rounded.

round_to Number of fractional digits (i.e., digits after the decimal point), to round to.

Value

Number as string: num rounded to round_to digits, with trailing zeros when applicable.

Examples

```
ro( 1.2345 ) # returns "1.23"
ro( 0.12345, 1 ) # returns "0.1"
ro( 12.3, 4 ) # returns "12.3000"
```

roc_neat

Difference of Two Areas Under the Curves

Description

Comparison of two areas under the receiver operating characteristic curves (AUCs).

Usage

```
roc_neat(roc1, roc2, pair = FALSE, greater = NULL)
```

Arguments

roc1 Receiver operating characteristic (ROC) object.
roc2 Receiver operating characteristic (ROC) object.

pair Logical. If TRUE, the test is conducted for paired samples. Otherwise (default)

for independent samples.

greater NULL or string (or number); optionally specifies one-sided test: either "1" (roc1

AUC expected to be greater than roc2 AUC) or "2" (roc2 AUC expected to be

greater than roc2 AUC). If NULL (default), the test is two-sided.

Value

Prints DeLong's test results for the comparison of the two given AUCs in APA style. Furthermore, when assigned, returns a named vector with the following two elements: stat (D value), p (p value).

Note

The test statistics are calculated via pROC::roc.test as DeLong's test (for both paired and unpaired). The roc_neat function merely prints it in APA style.

The ROC object may be calculated via t_neat, or directly with pROC::roc.

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References

DeLong, E. R., DeLong, D. M., & Clarke-Pearson, D. L. (1988). Comparing the areas under two or more correlated receiver operating characteristic curves: a nonparametric approach. Biometrics, 44(3), 837-845. doi: 10.2307/2531595

Robin, X., Turck, N., Hainard, A., Tiberti, N., Lisacek, F., Sanchez, J. C., & Muller, M. (2011). pROC: an open-source package for R and S+ to analyze and compare ROC curves. BMC bioinformatics, 12(1), 77. doi: 10.1186/147121051277

See Also

t_neat

Examples

```
# calculate first AUC (from v1 and v2)
v1 = c(191, 115, 129, 43, 523,-4, 34, 28, 33,-1, 54)
v2 = c(4,-2, 23, 13, 32, 16, 3, 29, 37,-4, 65)
results1 = t_neat(v1, v2, auc_added = TRUE)

# calculate second AUC (from v3 and v4)
v3 = c(14.1, 58.5, 25.5, 42.2, 13, 4.4, 55.5, 28.5, 25.6, 37.1)
v4 = c(36.2, 45.2, 41, 24.6, 30.5, 28.2, 40.9, 45.1, 31, 16.9)
results2 = t_neat(v3, v4, auc_added = TRUE)

# one-sided comparison of the two AUCs
roc_neat(results1$roc_obj, results2$roc_obj, greater = "1")
```

 $script_path$

Script Path

Description

Gives, in RStudio, the path to the script file in which it is executed.

Usage

```
script_path(subdir = "")
```

Arguments

subdir

String, optional. Subdirectory relative to the script's path.

Value

Script file's path as string. If subdir is given, it is appended to the original path.

```
setwd( script_path() ) # sets working directory to the script's path, e.g. "C:/script_folder/"
script_path('my_subdir/misc/') # returns "C:/script_folder/my_subdir/misc/"
```

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|--|

Description

Creates a neat means (or medians) and standard deviations table, using m_neat() function(s) as arguments.

Usage

```
table_neat(values_list, group_by = NULL, group_per = "rows",
  to_clipboard = FALSE, medians = FALSE)
```

Arguments

values_list	Values returned (as data frames) from the m_neat function: variables from which the statistics for the table are to be calculated. The group_by and medians parameters are ignored when they are given in the table_neat function; see Details.
group_by	A vector of factors by which the statistics are grouped. (Overwrites group_by in m_neat; see Details.)
group_per	String, "rows" or "columns". If set to "columns" (or just "c" or "col", etc.), each column contains statistics for one group. Otherwise (default), each row contains statistics for one group.
to_clipboard	Logical. If TRUE, the table is copied to the clipboard (default: FALSE).
medians	Logical. If TRUE, medians are calculated, otherwise (default: FALSE) means. (Overwrites medians in m_neat; see Details.)

Details

The values, round_to, and new_name arguments given in the m_neat function are always applied. However, the group_by or medians given in the m_neat function are only applied when no arguments are given in the table_neat function for the identical parameters (group_by or medians). If either parameter is given in the table_neat function, all separately given respective argument(s) in the m_neat function(s) are ignored.

Value

Returns a data frame with means or medians and SDs per variable and per group.

See Also

```
m_neat for more related details
```

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t_neat

Difference of Two Means and Area Under the Curve

Description

Welch's t-test results including Cohen's d with confidence interval (CI), Bayes factor (BF), and area under the receiver operating characteristic curve (AUC).

Usage

```
t_neat(var1, var2, pair = FALSE, greater = NULL, ci = NULL,
    bf_added = TRUE, auc_added = FALSE, r_added = TRUE,
    for_table = FALSE, test_title = "Descriptives:", round_descr = 2,
    round_auc = 3, auc_greater = "1")
```

Arguments

var1	Numeric vector; numbers of the first variable.
var2	Numeric vector; numbers of the second variable.
pair	Logical. If TRUE, all tests (t, BF, AUC) are conducted for paired samples. If FALSE (default) for independent samples.
greater	NULL or string (or number); optionally specifies one-sided tests (t and BF): either "1" (var1 mean expected to be greater than var2 mean) or "2" (var2 mean expected to be greater than var1 mean). If NULL (default), the test is two-sided.
ci	Numeric; confidence level for returned CIs for Cohen's d and AUC.
bf_added	Logical. If TRUE (default), Bayes factor is calculated and displayed.
auc_added	Logical. If TRUE, AUC is calculated and displayed. (FALSE by default.)
r_added	Logical. If TRUE (default), Pearson correlation is calculated and displayed in case of paired comparison.
for_table	Logical. If TRUE, omits the confidence level display from the printed text.

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test_title String, "Descriptives:" by default. Simply displayed in printing preceding the

descriptive statistics. (Useful e.g. to distinguish several different comparisons

inside a function or a for loop.)

round_descr Number to round to the descriptive statistics (means and SDs).

round_auc Number to round to the AUC and its CI.

auc_greater String (or number); specifies which variable is expected to have greater values

for 'cases' as opposed to 'controls': "1" (default; var1 expected to be greater for 'cases' than var2 mean) or "2" (var2 expected to be greater for 'cases' than

var1). Not to be confused with one-sided tests; see Details.

Details

The Bayes factor (BF) is always calculated with the default r-scale of \emptyset . 707. BF supporting null hypothesis is denoted as BF01, while that supporting alternative hypothesis is denoted as BF10. When the BF is smaller than 1 (i.e., supports null hypothesis), the reciprocal is calculated (hence, BF10 = BF, but BF01 = 1/BF). When the BF is greater than or equal to 10000, scientific (exponential) form is reported for readability. (The original full BF number is available in the returned named vector as bf.)

The original pROC::auc function, by default, always returns an AUC greater than (or equal to) .5, assuming that the prediction based on values in the expected direction work correctly at least at chance level. This however may be confusing. Consider an example where we measure the heights of persons in a specific small sample and expect that greater height predicts masculine gender. The results are, say, 169, 175, 167, 164 (cm) for one gender, and 176, 182, 179, 165 for the other. If the expectation is correct (the second, greater values are for males), the AUC is .812. However, if in this particular population females are actually taller than males, the AUC is in fact .188. To keep things clear, the t_neat function always makes an assumption about which variable is expected to be greater for correct classification ("1" by default; i.e., var1; to be specified as auc_greater = "2" for var2 to be expected as greater). For this example, if the first (smaller) variables are given as var1 for females, and second (larger), variables are given as var2 for males, we have to specify auc_greater = "2" to indicate the expectation of larger values for males. (Or, easier, just add the expected larger values as var1.)

Value

Prints t-test statistics (including Cohen'd with CI, BF, and AUC, as specified via the corresponding parameters) in APA style. Furthermore, when assigned, returns a list, that contains a named vector 'stats' with the following elements: t (t value), p (p value), d (Cohen's d), bf (Bayes factor), auc (AUC), accuracy (accuracy using the most optimal classification threshold). The latter two is NULL when auc_added is FALSE. When auc_added is TRUE, there are also two additional elements of the list. One is 'roc_obj', which is a roc object, to be used e.g. with the roc_neat function. The other is 'best_thresholds', which contains the best threshold value(s) for classification, along with corresponding specificity and sensitivity.

Note

The Welch's t-test is calculated via stats::t.test.

Cohen's d and its confidence interval are calculated, using the t value, via MBESS::ci.smd for independent samples (as standardized mean difference) and via MBESS::ci.sm for paired samples (as standardized mean).

The Bayes factor is calculated via BayesFactor::ttestBF.

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The correlation and its CI are calculated via stats::cor.test, and is always two-sided, always with 95 percent CI. For more, use corr_neat.

The AUC and its CI are calculated via pROC::auc, and the accuracy at most optimal threshold via pROC::coords (x = "best"); both using the object pROC::roc.

References

Delacre, M., Lakens, D., & Leys, C. (2017). Why psychologists should by default use Welch's t-test instead of Student's t-test. International Review of Social Psychology, 30(1). doi: 10.5334/irsp.82

Kelley, K. (2007). Methods for the behavioral, educational, and social sciences: An R package. Behavior Research Methods, 39(4), 979-984. doi: 10.3758/BF03192993

Robin, X., Turck, N., Hainard, A., Tiberti, N., Lisacek, F., Sanchez, J. C., & Muller, M. (2011). pROC: an open-source package for R and S+ to analyze and compare ROC curves. BMC bioinformatics, 12(1), 77. doi: 10.1186/147121051277

See Also

```
corr_neat, roc_neat
```

```
# assign two variables (numeric vectors)
v1 = c(191, 115, 129, 43, 523,-4, 34, 28, 33,-1, 54)
v2 = c(4,-2, 23, 13, 32, 16, 3, 29, 37,-4, 65)

t_neat(v1, v2) # prints results as independent samples
t_neat(v1, v2, pair = TRUE) # as paired samples (r added by default)
t_neat(v1, v2, pair = TRUE, greater = "1") # one-sided
t_neat(v1, v2, pair = TRUE, auc_added = TRUE) # AUC included

# print results and assign returned list
results = t_neat(v1, v2, pair = TRUE)

results$stats['bf'] # get precise BF value
```

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