

# Package ‘psychtoolbox’

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**Title** Tools for psychology research and psychometrics

**Version** 0.0.1

**Description** This package contains functions helping to analyse psychological data.

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**URL** <https://gitlab.com/lukas.novak/psychtoolbox>

**Encoding** UTF-8

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**Imports** coin,  
docextractr,  
effectsize,  
dplyr,  
equaltestMI,  
insight,  
lavaan,  
magrittr,  
rmarkdown,  
rstatix,  
stats,  
tidyr,  
expss,  
stringr,  
tidyselect,  
vctrs,  
purrr,  
reshape2,  
janitor,  
broom,  
car,  
bootnet,  
WRS2,  
nortest,  
rlang

**Suggests** testthat (>= 3.0.0)

**Config/testthat/edition** 3

**Depends** R (>= 2.10)

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clin_sig_chang	<i>Clinically significant change</i>
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Description

This easy function calculates Clinically significant change (clinical cut-off scores) as defined by Jacobson and Truax (1991).

Usage

clin\_sig\_chang(SD\_0, SD\_1, M\_1, M\_0)

Arguments

- |      |   |
|------|---|
| SD_0 | standard deviation of the non-clinical population |
| SD_1 | standard deviation of the clinical population     |
| M_1  | mean of the clinical population                   |
| M_0  | mean of the non-clinical population               |

Format

numeric vector of values

Details

This function computes cut-off score differentiating between the clinical and non-clinical population based on the Jacobson and Truax (1991) formula (p. 13). The mathematical formula can be also found in Biescad & Timulak(2014, p. 150).

Value

numeric vector

Author(s)

Lukas Novak, <lukasjirinovak@gmail.com>

## References

Jacobson, N. S., & Truax, P. (1991). Clinical significance: A statistical approach to defining meaningful change in psychotherapy research. *Journal of Consulting and Clinical Psychology*, 59(1), 12-19, DOI: <https://doi.org/10.1037/0022-006X.59.1.12>

Matus Biescad & Ladislav Timulak (2014). Measuring psychotherapy outcomes in routine practice: Examining Slovak versions of three commonly used outcome instruments, *European Journal of Psychotherapy & Counselling*, 16:2, 140-162, DOI: <https://doi.org/10.1080/13642537.2014.895772>

## See Also

[RCI\(\)](#) function for calculation of the Reliable Change Index

## Examples

```
clin.cut.off=clin_sig_chang(SD_0 = 3.5,
                           SD_1 = 2.1,
                           M_0 = 4.2,
                           M_1 = 12.1)

clin.cut.off
```

---

edge_weight_summary	<i>Summarizing Bootstrapped Network Estimates from 'bootnet'</i>
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---

## Description

Summarizing Bootstrapped Network Estimates from 'bootnet'

## Usage

```
edge_weight_summary(
  bootnet_output,
  include_sample_edge_weight = TRUE,
  include_p_values = "none"
)
```

## Arguments

**bootnet\_output** Output from the 'bootnet' package after bootstrapping network analysis.

**include\_sample\_edge\_weight** Logical, whether to include sample edge weight in the summary table.

**include\_p\_values** String, whether to include p-values as significance stars, add p-value column, or exclude p-values entirely. Options are "stars", "exact", or "none". Default is "none".

## Format

An object of class "tibble".

## Details

This function provides a convenient summary of bootstrapped network estimates from the 'bootnet' package, including edge weights, their confidence intervals, and p-values alternatively. It is tailored to enhance the usability of 'bootnet' output by summarizing critical metrics in a concise format. Importantly, it only summarizes results in edges that were non-zero on sample level. Edges that had zero edge weight in a sample are filtered out.

The empirical p-values are calculated using the following formula:

$$p = \frac{\sum_{i=1}^n \mathbf{1}\{|x_i| \geq |x_0|\}}{n}$$

where  $x_0$  is the original estimate and  $x_i$  are the bootstrap estimates.

## Value

A tibble summarizing edge weights.

## Author(s)

Lukas Novak, <lukasjirinovak@gmail.com>

## References

Epskamp, S., Borsboom, D., & Fried, E. I. (2018). Network analysis: An integrative approach to the structure of psychopathology. *Annual review of clinical psychology*, 14, 91-121.

## Examples

```
# Example testing of function with 'bootnet' output
set.seed(746841)
library(tidyr)
test_dat_1 <- tibble("Group" = rbinom(100, size = 1, prob = .5),
  "y" = ifelse(Group == 0,
    rnorm(100, mean = 50, sd = 10),
    rnorm(100, mean = 40, sd = 12)),
  "z" = rnorm(100, mean = 41, sd = 11),
  "w" = rnorm(100, mean = 40, sd = 10))

estimate <- bootnet::estimateNetwork(test_dat_1, default = "EBICglasso")
boot_results <- bootnet::bootnet(estimate, nBoots = 100, nCores = 1)

summary <- edge_weight_summary(bootnet_output = boot_results)
print(summary)
```

## Description

This function performs linear regression and print results in tibble output. This function aims to provide the results of the regression analysis in the format, which is frequently desired in academic journals.

**Usage**

```
lasy.lin.reg(  
  independent.var,  
  dependent.var,  
  print.cov = FALSE,  
  Z_score_independent = FALSE,  
  check_multicollinearity = TRUE,  
  covariates = NULL,  
  data  
)
```

**Arguments**

independent.var	independent variable/s
dependent.var	dependent variable/s
print.cov	Print effect of covariates, default is FALSE
Z_score_independent	Should independent variables be z-scored? Default is FALSE
check_multicollinearity	Should multicollinearity assumption be checked? Default is TRUE
covariates	covariates to be included in a model
data	data frame or tibble object

**Format**

An object of class "tibble"

**Details**

Currently, this function does not provide model fit indicators such as AIC or BIC.

**Value**

data frame

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

- Welch, B. L. (1947). "The generalization of "Student's" problem when several different population variances are involved". *Biometrika*. 34 (1–2): 28–35.
- Wilcoxon, F., Individual Comparisons by Ranking Methods, *Biometrics Bulletin*, Vol. 1, 1945, pp. 80–83. DOI:10.2307/3001968
- Dunn, O. J. (1961) Multiple comparisons among means. *Journal of the American Statistical Association*. 56, 52–64.
- Games, P. A., Keselman, H. J., & Clinch, J. J. Tests for homogeneity of variance in factorial designs. *Psychological Bulletin*, 86, 978–984

## Examples

```
# data simulation
library(dplyr)
data.PAQ = tibble(.rows = 1000)
data.PAQ <- data.PAQ %>%
  mutate("multiple__exper_1" = rnorm(n = nrow(data.PAQ), mean = 20, sd = 5)) %>%
  mutate("exper_1" = rnorm(n = nrow(data.PAQ), mean = 20, sd = 5)) %>%
  mutate("exper_1" = rnorm(n = nrow(data.PAQ), mean = 20, sd = 5),
         "exper_2" = rnorm(n = nrow(data.PAQ), mean = 20, sd = 5),
         "last_binary_vasdl" = rnorm(n = nrow(data.PAQ), mean = 20, sd = 20),
         "last_binary_val2" = rnorm(n = nrow(data.PAQ), mean = 5, sd = 5),
         "last_binary_val3" = rnorm(n = nrow(data.PAQ), mean = 10, sd = 40),
         "last_binary_val4" = rnorm(n = nrow(data.PAQ), mean = 50, sd = 5),
         "last_binary_val5" = rnorm(n = nrow(data.PAQ), mean = 5, sd = 4),
         "last_binary_val6" = rnorm(n = nrow(data.PAQ), mean = 65, sd = 5))

lin.reg.output <- lasy.lin.reg(independent.var = c("last_binary_vasdl"),
                              covariates = c("last_binary_val6"),
                              dependent.var = c("last_binary_val5", "last_binary_val4"),
                              Z_score_independent = FALSE,
                              check_multicollinearity = TRUE,
                              data = data.PAQ,
                              print.cov = FALSE)

print(lin.reg.output)
```

---

lasy.log.reg

*Lasy logistic regression function*


---

## Description

This function performs logistic regression and print results in tibble output. This function aims to provide the results of the regression analysis in the format, which is frequently desired in academic journals.

## Usage

```
lasy.log.reg(
  independent.var,
  dependent.var,
  print.cov = FALSE,
  covariates = NULL,
  check_multicollinearity = TRUE,
  data
)
```

## Arguments

<code>independent.var</code>	independent variable/s
<code>dependent.var</code>	dependent variable/s
<code>print.cov</code>	Print effect of covariates, default is FALSE

covariates	covariates to be included in a model
check_multicollinearity	Should multicollinearity assumption be checked? Default is TRUE
data	data frame or tibble object

**Format**

An object of class "tibble"

**Details**

Currently, this function does not provide model fit indicators such as AIC or BIC.

**Value**

data frame

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

Welch, B. L. (1947). "The generalization of "Student's" problem when several different population variances are involved". *Biometrika*. 34 (1–2): 28–35.

Wilcoxon, F., Individual Comparisons by Ranking Methods, *Biometrics Bulletin*, Vol. 1, 1945, pp. 80–83. DOI:10.2307/3001968

Dunn, O. J. (1961) Multiple comparisons among means. *Journal of the American Statistical Association*. 56, 52–64.

Games, P. A., Keselman, H. J., & Clinch, J. J. Tests for homogeneity of variance in factorial designs. *Psychological Bulletin*, 86, 978–984

**Examples**

```
# data loading
data(paq.validation.study)
regress.output <- lasy.log.reg(independent.var = c("TEQ", "PAQ"),
                             covariates = c("Age"),
                             dependent.var = c("econom_stat_dich",
                                                "family_status",
                                                "edu_dich"),
                             data = paq.validation.study)

print(regress.output)
```

mitab

*Measurement invariance table***Description**

Measurement invariance table

**Usage**

```
mitab(
  group1_nam,
  group2_nam,
  ordered,
  model,
  data,
  std.lv,
  meanstructure,
  group,
  yes_no_results,
  estimator,
  robust = FALSE,
  cfi.difference = TRUE,
  rmsea.difference = TRUE,
  ...
)
```

**Arguments**

group1_nam	name of the first group
group2_nam	name of the second group
ordered	logical, if set to TRUE items will be treated as ordered variables
model	lavaan model to test
data	data frame or tibble
std.lv	logical, if TRUE than standardized loadings are stored in temporal output
meanstructure	logical, if TRUE than model with meanstructure is estimated
group	name of grouping variable
yes_no_results	logical, if TRUE than lasy output indicating difference between models is added, currently working only based on CFI and RMSEA
estimator	name of estimator to be used during fitting procedure
robust	logical, if TRUE, than robust results are printed, working only with estimators providing robust results (e.g. MLR or DWLS)
cfi.difference	logical, if TRUE, delta of the CFI is printed in output
rmsea.difference	logical, if TRUE, delta of RMSEA is printed in output
...	optional arguments for CFA function



**Format**

An object of class "tibble"

**Details**

This function creates table with the key output from measurement invariance testing.

**Value**

data frame

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

Myles Hollander and Douglas A. Wolfe (1973). Nonparametric Statistical Methods. New York: John Wiley & Sons. Pages 27–33 (one-sample), 68–75 (two-sample). Or second edition (1999).

**Examples**

```
# The famous Holzinger and Swineford (1939) example
HS.model <- ' visual  =~ x1 + x2 + x3
textual =~ x4 + x5 + x6
speed   =~ x7 + x8 + x9 '

library(lavaan)
dat <- HolzingerSwineford1939
res.tab.mi <- mitab(
  group1_nam = "Grant-White",
  group2_nam = "Pasteur",
  ordered = FALSE,
  model = HS.model,
  data = dat,
  std.lv = TRUE,
  meanstructure = TRUE,
  group = "school",
  yes_no_results = TRUE,
  estimator = "MLR",
  robust = TRUE,
  cfi.difference = TRUE,
  rmsea.difference = TRUE
)

print(res.tab.mi)
```

mult.g.comp

*Robust multi-group comparison***Description**

This function allows to compare multiple groups in multiple outcome variables with violated parametric assumptions.

**Usage**

```
mult.g.comp(
  df,
  outcome.var,
  groups,
  desc_only = FALSE,
  short_results = TRUE,
  remove_missings = FALSE,
  percent_decimals = 2,
  show_non_significant_results = FALSE,
  diagnostics = FALSE
)
```

**Arguments**

df	data frame or tibble object
outcome.var	continuous variable/s
groups	grouping variable/s
desc_only	print only descriptive statistics, default is FALSE
short_results	prints only significance stars without numerical results, default is TRUE
remove_missings	remove missing values from a table, default is FALSE
percent_decimals	number of decimals used to round percenages, default is 2
show_non_significant_results	if TRUE, p-values from non-significant tests are reported. Default is FALSE.
diagnostics	if TRUE, prints a detailed diagnostic report for each test run. Default is FALSE.

**Format**

An object of class "tibble"

**Details**

This function automatically selects and performs appropriate statistical tests based on the number of groups and the underlying data assumptions. For non-significant results, if `show_non_significant_results = TRUE`, the p-value of the performed test is reported. For multi-group comparisons, this is the Kruskal-Wallis test. For two-group comparisons, this is the p-value from the specific test chosen by the decision tree.

**Two group comparison:**

A decision tree based on normality (Shapiro-Wilk for  $N \leq 5000$ , Anderson-Darling for  $N > 5000$ ) and homogeneity of variances (Fligner-Killeen) is used. Depending on the assumptions, a Student's t-test, Welch's t-test, Wilcoxon test, or Yuen's test on trimmed means is performed.

**Three and more groups comparison:**

For comparisons involving three or more groups, a Kruskal-Wallis test is first performed. If significant, post-hoc tests (Dunn's test for equal variances or Games-Howell for unequal variances) are conducted.

**Value**

data frame

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

Welch, B. L. (1947). "The generalization of "Student's" problem when several different population variances are involved". *Biometrika*. 34 (1–2): 28–35.

Wilcoxon, F., Individual Comparisons by Ranking Methods, *Biometrics Bulletin*, Vol. 1, 1945, pp. 80–83. DOI:10.2307/3001968

Dunn, O. J. (1961) Multiple comparisons among means. *Journal of the American Statistical Association*. 56, 52–64.

Games, P. A., Keselman, H. J., & Clinch, J. J. Tests for homogeneity of variance in factorial designs. *Psychological Bulletin*, 86, 978–984

Cohen, J. (1988). *Statistical power analysis for the behavioral sciences* (2nd ed.). Hillsdale, NJ: Erlbaum.

Lakens, D. (2013). Calculating and reporting effect sizes to facilitate cumulative science: a practical primer for t-tests and ANOVAs. *Frontiers in psychology*, 4, 863.

Wilcox, R. R., & Tian, T. S. (2011). Measuring effect size: A robust heteroscedastic approach for two or more groups. *Journal of Applied Statistics*, 38(7), 1359-1368.

**Examples**

```
# data loading
tab.1=mult.g.comp(df = paq.validation.study, outcome.var = c("PAQ", "G_DIF", "G_DDF", "G_EOT"),
groups = c("economical_status",
           "Gender",
           "education",
           "family_status"))

# printing the output
print(tab.1)
```

---

`node_predictability_summary`*Summarizing Node Predictability*

---

**Description**

Summarizing Node Predictability

**Usage**

```
node_predictability_summary(predict_function_output)
```

**Arguments**

`predict_function_output`

output from a predictability analysis function.

**Format**

An object of class "tibble".

**Details**

This function provides a summary of node predictability metrics. It is designed to extract and format predictability information from the output of a predictability analysis function.

**Value**

A tibble summarizing node predictability.

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

Epskamp, S., Borsboom, D., & Fried, E. I. (2018). Network analysis: An integrative approach to the structure of psychopathology. *Annual review of clinical psychology*, 14, 91-121.

**Examples**

```
# Example testing of function with 'bootnet' and predictability output
set.seed(746841)
library(tidyr)
test_dat_1 = tibble("Group" = rbinom(1:100, size = 0:1, prob = .5),
  "y" = ifelse(Group == 0,
    rnorm(n = 1:100, mean = 50, sd = 10),
    rnorm(n = 1:100, mean = 40, sd = 12)),
  "z" = rnorm(n = 1:100, mean = 41, sd = 11),
  "w" = rnorm(n = 1:100, mean = 40, sd = 10))

estimate_dat_1 <- bootnet::estimateNetwork(test_dat_1, default = "mgm")
predict_output_dat_1 <- predict(estimate_dat_1$results, estimate_dat_1$data)
```

```
predictability_summary <- node_predictability_summary(predict_output_dat_1)
print(predictability_summary)
```

---

paq.validation.study    *paq.validation.study*

---

## Description

This dataset contains data which were used for validation of the Perth Alexithymia Questionnaire in the Czech Republic.

## Usage

```
paq.validation.study
```

## Format

A data frame with 848 rows and 53 variables:

```
P_DIF double COLUMN_DESCRIPTION
P_DDF double COLUMN_DESCRIPTION
P_DAF double COLUMN_DESCRIPTION
N_DIF double COLUMN_DESCRIPTION
N_DDF double COLUMN_DESCRIPTION
N_DAF double COLUMN_DESCRIPTION
G_EOT double COLUMN_DESCRIPTION
G_DIF double COLUMN_DESCRIPTION
G_DDF double COLUMN_DESCRIPTION
G_DAF double COLUMN_DESCRIPTION
family_status integer COLUMN_DESCRIPTION
Gender integer COLUMN_DESCRIPTION
TEQ_1 double COLUMN_DESCRIPTION
TEQ_3 double COLUMN_DESCRIPTION
TEQ_5 double COLUMN_DESCRIPTION
TEQ_16 double COLUMN_DESCRIPTION
TEQ_CON_2 double COLUMN_DESCRIPTION
TEQ_CON_4 double COLUMN_DESCRIPTION
TEQ_CON_14 double COLUMN_DESCRIPTION
TEQ double COLUMN_DESCRIPTION
Age double COLUMN_DESCRIPTION
age_group double COLUMN_DESCRIPTION
age_quality double COLUMN_DESCRIPTION
ethnicity integer COLUMN_DESCRIPTION
```

```

education integer COLUMN_DESCRIPTION
economical_status integer COLUMN_DESCRIPTION
PAQ_1 double COLUMN_DESCRIPTION
PAQ_2 double COLUMN_DESCRIPTION
PAQ_3 double COLUMN_DESCRIPTION
PAQ_4 double COLUMN_DESCRIPTION
PAQ_5 double COLUMN_DESCRIPTION
PAQ_6 double COLUMN_DESCRIPTION
PAQ_7 double COLUMN_DESCRIPTION
PAQ_8 double COLUMN_DESCRIPTION
PAQ_9 double COLUMN_DESCRIPTION
PAQ_10 double COLUMN_DESCRIPTION
PAQ_11 double COLUMN_DESCRIPTION
PAQ_12 double COLUMN_DESCRIPTION
PAQ_13 double COLUMN_DESCRIPTION
PAQ_14 double COLUMN_DESCRIPTION
PAQ_15 double COLUMN_DESCRIPTION
PAQ_16 double COLUMN_DESCRIPTION
PAQ_17 double COLUMN_DESCRIPTION
PAQ_18 double COLUMN_DESCRIPTION
PAQ_19 double COLUMN_DESCRIPTION
PAQ_20 double COLUMN_DESCRIPTION
PAQ_21 double COLUMN_DESCRIPTION
PAQ_22 double COLUMN_DESCRIPTION
PAQ_23 double COLUMN_DESCRIPTION
PAQ_24 double COLUMN_DESCRIPTION
PAQ double COLUMN_DESCRIPTION
edu_dich integer COLUMN_DESCRIPTION
econom_stat_dich integer COLUMN_DESCRIPTION

```

### **dichotomization of variables**

```

paq.validation.study <- paq.validation.study %>% dplyr::mutate(edu_dich = as.factor(ifelse( education == "University master or higher", "University", "lower_edu" )), econom_stat_dich = as.factor(ifelse( economical_status == "Student", "Student", "non_student" )))

```

---

RCI	<i>Reliable Change Index (RCI)</i>
-----	------------------------------------

---

**Description**

This function calculates Reliable Change Index (RCI) as modified by Wiger and Solberg (2001, p.148).

**Usage**

```
RCI(SD_0, test.ret.rel)
```

**Arguments**

SD_0	standard deviation of the non-clinical population
test.ret.rel	test-retest reliability of the instrument

**Format**

numeric vector of values

**Details**

This function computes value corresponding to "the minimum amount of change that could not be attributed to the error of measurement" (Biescad & Timulak, 2014, p. 150). If score change from before to post treatment is lower than value resulting from this function, then change in client score can not be attributed to the effectiveness of the therapy but rather other factors such as a measurement error (Biescad & Timulak, 2014). This function is a result of modification of the original Jacobson and Truax (1991) formula by Wiger and Solberg (2001, p.148).

**Value**

numeric vector

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

- Jacobson, N. S., & Truax, P. (1991). Clinical significance: A statistical approach to defining meaningful change in psychotherapy research. *Journal of Consulting and Clinical Psychology*, 59(1), 12-19, DOI: <https://doi.org/10.1037/0022-006X.59.1.12>
- Matus Biescad & Ladislav Timulak (2014). Measuring psychotherapy outcomes in routine practice: Examining Slovak versions of three commonly used outcome instruments, *European Journal of Psychotherapy & Counselling*, 16:2, 140-162, DOI: <https://doi.org/10.1080/13642537.2014.895772>
- Wiger, D. E., & Solberg, K. B. (2001). *Tracking Mental Health Outcomes: A Therapist's Guide to Measuring Client Progress, Analyzing Data, and Improving Your Practice* (1., Vol. 2001). Wiley.

**See Also**

`clin_sig_chang()` function for calculation of the clinical cut-off scores

**Examples**

```
re.ch.in = RCI(SD_0 = 4.87, test.ret.rel = 0.66)
re.ch.in
```

---

small\_boot\_table

*Summarizing Bootstrapped Network Estimates from 'bootnet'*


---

**Description**

Summarizing Bootstrapped Network Estimates from 'bootnet'

**Usage**

```
small_boot_table(
  bootnet_output,
  predict_function_output = NULL,
  include_sample_edge_weight = TRUE,
  include_p_values = "none"
)
```

**Arguments**

**bootnet\_output** output from the 'bootnet' package after bootstrapping network analysis.

**predict\_function\_output** optional, output from a predictability analysis function.

**include\_sample\_edge\_weight** Logical, whether to include sample edge weight in the summary table.

**include\_p\_values** String, whether to include p-values as significance stars, add p-value column, or exclude p-values entirely. Options are "stars", "exact", or "none". Default is "none".

**Format**

An object of class "tibble".

**Details**

This function provides a convenient summary of bootstrapped network estimates from the 'bootnet' package, including edge weights, their confidence intervals, and optionally p-values and node predictability metrics if provided. It is tailored to enhance the usability of 'bootnet' output by summarizing critical metrics in a concise format. Importantly, it only summarizes results in edges that were non-zero on sample level. Edges that had zero edge weight in a sample are filtered out.

The empirical p-values are calculated using the following formula:

$$p = \frac{\sum_{i=1}^n \mathbf{1}\{|x_i| \geq |x_0|\}}{n}$$

where  $x_0$  is the original estimate and  $x_i$  are the bootstrap estimates.



**Value**

A tibble summarizing edge weights and, optionally, node predictability.

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

Epskamp, S., Borsboom, D., & Fried, E. I. (2018). Network analysis: An integrative approach to the structure of psychopathology. *Annual review of clinical psychology*, 14, 91-121.

**Examples**

```
# Load required package

# Simulate data using base R
set.seed(654899)
Sigma <- matrix(0.5, nrow = 5, ncol = 5) + diag(0.5, 5)
chol_decomp <- chol(Sigma)
z <- matrix(rnorm(100 * 5), nrow = 100, ncol = 5)
sim_data <- z %%% chol_decomp
sim_data_df <- as.data.frame(sim_data)

# Estimate a network using 'EBICglasso'
estimate <- bootnet::estimateNetwork(sim_data_df, default = "EBICglasso")

# Perform bootstrapping on the estimated network
boot_results <- bootnet::bootnet(estimate, nBoots = 100, nCores = 1)

# Summarize the bootstrapped network
summary <- small_boot_table(bootnet_output = boot_results)
print(summary)
```

---

two.g.comp

*Automatic two-groups comparison*


---

**Description**

Automatic two-groups comparison

**Usage**

```
two.g.comp(df, y, group.var)
```

**Arguments**

df	data frame or tibble with one socio-demographic variable and one continuous variable
y	continuous variable
group.var	binary grouping variable

**Format**

An object of class "tibble"

**Details**

This function computes either Wilcox test or t-test depending on whether homogeneity of variances assumption is met or not.

**Value**

data frame

**Author(s)**

Lukas Novak, <lukasjirinovak@gmail.com>

**References**

Myles Hollander and Douglas A. Wolfe (1973). Nonparametric Statistical Methods. New York: John Wiley & Sons. Pages 27–33 (one-sample), 68–75 (two-sample). Or second edition (1999).

**Examples**

```
Sex = rbinom(1:100, size = 0:1, prob = .5)
test.dat = data.frame(cbind(Sex), Empathy = ifelse(Sex == 0,
                                                    rnorm(n = 1:100, mean = 50, sd = 10),
                                                    rnorm(n = 1:100, mean = 10, sd = 25)))

# running the function
two.g.comp.out.EC = two.g.comp(df = test.dat, y = "Empathy", group.var = "Sex")
# printing the output
print(two.g.comp.out.EC)
```

---

word2pdf

*word to pdf*

---

**Description**

Conversion of word document to pdf using either R Markdown package or Libre office. The latter represents higher quality approach - in general.

**Usage**

```
word2pdf(imp_file, out_file)
```

**Arguments**

imp_file	name of the word document to convert - without docx suffix
out_file	name of output pdf file without - without pdf suffix

**Format**

An object of class "pdf"

**Details**

this function is currently running only on windows

**Value**

pdf file

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**Examples**

```
# example from word do pdf
#word2pdf(imp_file = "example.docx",out_file = "example1.pdf")
```

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