Package ‘specr’

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example_data  Example data set

Description
This simulated data set can be used to explore the major function of `specr`.

Usage
data(example_data)

Format
A tibble

Examples
data(example_data)
head(example_data)

icc_specs  Compute intraclass correlation coefficient

Description
This function extracts intraclass correlation coefficients (ICC) from a multilevel model. It can be used to decompose the variance in the outcome variable of a specification curve analysis (e.g., the regression coefficients). This approach summarises the relative importance of analytical choices by estimating the share of variance in the outcome (e.g., the regression coefficient) that different analytical choices or combinations therefor account for. To use this approach, one needs to estimate a multilevel model that includes all analytical choices as grouping variables (see examples).

Usage
icc_specs(model, percent = TRUE)
plot_choices

Arguments

model  a multilevel (i.e., mixed effects) model that captures the variances of the specification curve.

percent a logical value indicating whether the ICC should also be printed as percentage. Defaults to TRUE.

Value

a tibble including the grouping variable, the random effect variances, the raw intraclass correlation coefficient (ICC), and the ICC in percent.

References


See Also

plot_variance() to plot the variance decomposition.

Examples

# Step 1: Run spec curve analysis
results <- run_specs(df = example_data,
            y = c("y1", "y2"),
            x = c("x1", "x2"),
            model = c("lm"))

# Step 2: Estimate a multilevel model without predictors
model <- lme4::lmer(estimate ~ 1 + (1|x) + (1|y), data = results)

# Step 3: Estimate intra-class correlation
icc_specs(model)

Description

This function plots how analytical choices affect the obtained results (i.e., the rank within the curve). Significant results are highlighted (negative = red, positive = blue, grey = nonsignificant). This function creates the lower panel in plot_specs().
Usage

plot_choices(
  df,
  choices = c("x", "y", "model", "controls", "subsets"),
  desc = FALSE,
  null = 0
)

Arguments

df a data frame resulting from run_specs().
choices a vector specifying which analytical choices should be plotted. By default, all choices are plotted.
desc logical value indicating whether the curve should be arranged in a descending order. Defaults to FALSE.
null Indicate what value represents the 'null' hypothesis (Defaults to zero).

Value
a ggplot object.

Examples

# Run specification curve analysis
results <- run_specs(df = example_data,
  y = c("y1", "y2"),
  x = c("x1", "x2"),
  model = c("lm"),
  controls = c("c1", "c2"),
  subsets = list(group1 = unique(example_data$group1),
                 group2 = unique(example_data$group2)))

# Plot simple table of choices
plot_choices(results)

# Plot only specific choices
plot_choices(results,
  choices = c("x", "y", "controls"))

---

plot_curve Plot ranked specification curve

Description

This function plots the a ranked specification curve. Confidence intervals can be included. Significant results are highlighted (negative = red, positive = blue, grey = nonsignificant). This function creates the upper panel in plot_specs().
Usage

plot_curve(
  df,
  desc = FALSE,
  ci = TRUE,
  ribbon = FALSE,
  legend = FALSE,
  null = 0
)

Arguments

df  a data frame resulting from run_specs().
desc logical value indicating whether the curve should the arranged in a descending order. Defaults to FALSE.
ci   logical value indicating whether confidence intervals should be plotted.
ribbon logical value indicating whether a ribbon instead should be plotted.
legend logical value indicating whether the legend should be plotted. Defaults to FALSE.
null  Indicate what value represents the null hypothesis (Defaults to zero)

Value

a ggplot object.

Examples

# load additional library
library(ggplot2) # for further customization of the plots

# Run specification curve analysis
results <- run_specs(df = example_data,
  y = c("y1", "y2"),
  x = c("x1", "x2"),
  model = c("lm"),
  controls = c("c1", "c2"),
  subsets = list(group1 = unique(example_data$group1),
                 group2 = unique(example_data$group2)))

# Plot simple specification curve
plot_curve(results)

# Ribbon instead of CIs and customize further
plot.curve(results, ci = FALSE, ribbon = TRUE) +
  geom_hline(yintercept = 0) +
  geom_hline(yintercept = median(results$estimate),
             linetype = "dashed") +
  theme_linedraw()
plot_decisiontree  

Plot decision tree

Description

This function plots a simple decision tree that is meant to help understanding how few analytical choices may result in a large number of specifications. It is somewhat useless if the final number of specifications is very high.

Usage

plot_decisiontree(df, label = FALSE, legend = FALSE)

Arguments

df  
data frame resulting from run_specs().

label  
Logical. Should labels be included? Defaults to FALSE. Produces only a reasonable plot if the number of specifications is low.

legend  
Logical. Should specific decisions be identifiable. Defaults to FALSE.

Value

a ggplot object.

Examples

results <- run_specs(df = example_data,  
y = c("y1", "y2"),  
x = c("x1", "x2"),  
model = c("lm"),  
controls = c("c1", "c2"))

# Basic, non-labelled decisions tree  
plot_decisiontree(results)

# Labelled decisions tree  
plot_decisiontree(results, label = TRUE)

# Add legend  
plot_decisiontree(results, label = TRUE, legend = TRUE)
plot_samplesizes

Plot sample sizes

Description

This function plots a histogram of sample sizes per specification. It can be added to the overall
specification curve plot (see vignettes).

Usage

plot_samplesizes(df, desc = FALSE)

Arguments

df a data frame resulting from run_specs().
desc logical value indicating whether the curve should be arranged in a descending
order. Defaults to FALSE.

Value

a ggplot object.

Examples

# load additional library
library(ggplot2) # for further customization of the plots

# run specification curve analysis
results <- run_specs(df = example_data,
y = c("y1", "y2"),
x = c("x1", "x2"),
model = c("lm"),
controls = c("c1", "c2"),
subsets = list(group1 = unique(example_data$group1),
               group2 = unique(example_data$group2)))

# plot ranked bar chart of sample sizes
plot_samplesizes(results)

# customize
plot_samplesizes(results) +
  geom_hline(yintercept = median(results$obs),
              color = "darkgrey",
              linetype = "dashed") +
  theme_linedraw()
Description

This function plots an entire visualization of the specification curve analysis. The function uses the entire tibble that is produced by run_specs() to create a standard visualization of the specification curve analysis. Alternatively, one can also pass two separately created ggplot objects to the function. In this case, it simply combines them using cowplot::plot_grid. Significant results are highlighted (negative = red, positive = blue, grey = nonsignificant).

Usage

plot_specs(
  df = NULL,
  plot_a = NULL,
  plot_b = NULL,
  choices = c("x", "y", "model", "controls", "subsets"),
  labels = c("A", "B"),
  rel_heights = c(2, 3),
  desc = FALSE,
  null = 0,
  ci = TRUE,
  ribbon = FALSE,
  sample_perc = 1,
  ...
)

Arguments

df a data frame resulting from run_specs().
plot_a a ggplot object resulting from plot_curve() (or plot_choices() respectively).
plot_b a ggplot object resulting from plot_choices() (or plot_curve() respectively).
choices a vector specifying which analytical choices should be plotted. By default, all choices are plotted.
lables labels for the two parts of the plot
rel_heights vector indicating the relative heights of the plot.
desc logical value indicating whether the curve should the arranged in a descending order. Defaults to FALSE.
null Indicate what value represents the 'null' hypothesis (defaults to zero).
ci logical value indicating whether confidence intervals should be plotted.
ribbon logical value indicating whether a ribbon instead should be plotted.
sample_perc
d numeric value denoting what percentage of the specifications should be plotted. Needs to be strictly greater than 0 and smaller than 1. Defaults to 1 (= all specifications). Drawing a sample from all specifications usually makes only sense of the number of specifications is very large and one wants to simplify the visualization.

... additional arguments that can be passed to plot_grid().

Value

a ggplot object.

See Also

- plot_curve() to plot only the specification curve.
- plot_choices() to plot only the choices panel.
- plot_samplesizes() to plot a histogram of sample sizes per specification.

Examples

# load additional library
library(ggplot2) # for further customization of the plots

# run spec analysis
results <- run_specs(example_data,
y = c("y1", "y2"),
x = c("x1", "x2"),
model = "lm",
controls = c("c1", "c2"),
subset = list(group1 = unique(example_data$group1)))

# plot results directly
plot_specs(results)

# Customize each part and then combine
p1 <- plot_curve(results) +
  geom_hline(yintercept = 0, linetype = "dashed", color = "grey") +
  ylim(-3, 12) +
  labs(x = "", y = "regression coefficient")

p2 <- plot_choices(results) +
  labs(x = "specifications (ranked)")

plot_specs(plot_a = p1,  # arguments must be called directly!
            plot_b = p2,
            rel_height = c(2, 2))
plot_summary

Create box plots for given analytical choices

Description
This function provides a convenient way to visually investigate the effect of individual choices on the estimate of interest. It produces box-and-whisker plot(s) for each provided analytical choice.

Usage
plot_summary(df, choices = c("x", "y", "model", "controls", "subsets"))

Arguments
df
   a data frame resulting from run_specs().
choices
   a vector specifying which analytical choices should be plotted. By default, all choices are plotted.

Value
   a ggplot object.

See Also
   summarise_specs() to investigate the affect of analytical choices in more detail.

Examples
   # run spec analysis
   results <- run_specs(example_data,
                        y = c("y1", "y2"),
                        x = c("x1", "x2"),
                        model = "lm",
                        controls = c("c1", "c2"),
                        subset = list(group1 = unique(example_data$group1)))

   # plot boxplot comparing specific choices
   plot_summary(results, choices = c("subsets", "controls", "y"))
Description

This function creates a simple barplot that visually displays how much variance in the outcome (e.g., the regression coefficient) different analytical choices or combinations thereof account for. To use this approach, one needs to estimate a multilevel model that includes all analytical choices as grouping variables (see examples and vignettes). This function uses `icc_specs()` to compute the intraclass correlation coefficients (ICCs), which provides the data basis for the plot (see examples).

Usage

```r
plot_variance(model)
```

Arguments

- `model` a multilevel model that captures the variances of the specification curve (based on the data frame resulting from `run_specs`).

Value

- a `ggplot` object.

See Also

- `icc_specs()` to produce a tibble that details the variance decomposition.

Examples

```r
# Step 1: Run spec curve analysis
results <- run_specs(df = example_data,
                     y = c("y1", "y2"),
                     x = c("x1", "x2"),
                     model = c("lm"))

# Step 2: Estimate multilevel model
library(lme4, quietly = TRUE)
model <- lmer(estimate ~ 1 + (1|x) + (1|y), data = results)

# Step 3: Plot model
plot_variance(model)
```
run_specs

Estimate all specifications

Description

This is the central function of the package. It runs the specification curve analysis. It takes the data frame and vectors for analytical choices related to the dependent variable, the independent variable, the type of models that should be estimated, the set of covariates that should be included (none, each individually, and all together), as well as a named list of potential subsets. The function returns a tidy tibble which includes relevant model parameters for each specification. The function tidy is used to extract relevant model parameters. Exactly what tidy considers to be a model component varies across models but is usually self-evident.

Usage

run_specs(
  df,
  x,
  y,
  model = "lm",
  controls = NULL,
  subsets = NULL,
  conf.level = 0.95,
  keep.results = FALSE
)

Arguments

df  a data frame that includes all relevant variables
x   a vector denoting independent variables
y   a vector denoting the dependent variables
model a vector denoting the model(s) that should be estimated.
controls a vector denoting which control variables should be included. Defaults to NULL.
subsets a named list that includes potential subsets that should be evaluated (see examples). Defaults to NULL.
conf.level the confidence level to use for the confidence interval. Must be strictly greater than 0 and less than 1. Defaults to .95, which corresponds to a 95 percent confidence interval.
keep.results a logical value indicating whether the complete model object should be kept. Defaults to FALSE.

Value

a tibble that includes all specifications and a tidy summary of model components.
References


See Also

plot_specs() to visualize the results of the specification curve analysis.

Examples

```r
# run specification curve analysis
results <- run_specs(df = example_data,
                      y = c("y1", "y2"),
                      x = c("x1", "x2"),
                      model = c("lm"),
                      controls = c("c1", "c2"),
                      subsets = list(group1 = unique(example_data$group1),
                                      group2 = unique(example_data$group2)))

# Check results frame
results
```

setup_specs

Set up specifications

Description

This function creates a tibble that includes all possible specifications based the dependent and independent variables, model types, and control variables that are specified. This function simply produces a tibble of all combinations. It can be used to check the specified analytical choices. This function is called within run_specs(), which estimates all specified models based on the data that are provided.

Usage

```r
setup_specs(x, y, model, controls = NULL)
```

Arguments

- **x**: a vector denoting independent variables
- **y**: a vector denoting the dependent variables
- **model**: a vector denoting the model(s) that should be estimated.
- **controls**: a vector of the control variables that should be included. Defaults to NULL.
Value

a tibble that includes all possible specifications based on combinations of the analytical choices.

See Also

run_specs() to run the specification curve analysis.

Examples

setup_specs(y = c("y1"),
            x = c("x1", "x2"),
            model = c("lm"),
            controls = c("c1", "c2"))

summarise_specs

summarise_specs

Summarise specifications

Description

This function allows to inspect results of the specification curves by returning a comparatively simple summary of the results. This summary can be produced for various specific analytical choices and customized summary functions.

Usage

summarise_specs(
  df,
  ...,
  var = .data$estimate,
  stats = list(median = median, mad = mad, min = min, max = max, q25 = function(x) quantile(x, prob = 0.25), q75 = function(x) quantile(x, prob = 0.75))
)

Arguments

df
  a data frame resulting from run_specs().

...   one or more grouping variables (e.g., subsets, controls,...) that denote the available analytical choices.

var
  which variable should be evaluated? Defaults to estimate (the effect sizes computed by run_specs()).

stats
  named vector or named list of summary functions (individually defined summary functions can included). If it is not named, placeholders (e.g., "fn1") will be used as column names.
summarise_specs

Value
	a tibble.

See Also

plot_summary() to visually investigate the affect of analytical choices.

Examples

# Run specification curve analysis
results <- run_specs(df = example_data,
                     y = c("y1", "y2"),
                     x = c("x1", "x2"),
                     model = c("lm"),
                     controls = c("c1", "c2"),
                     subsets = list(group1 = unique(example_data$group1),
                                     group2 = unique(example_data$group2)))

# overall summary
summarise_specs(results)

# Summary of specific analytical choices
summarise_specs(results, # data frame
                 x, y) # analytical choices

# Summary of other parameters across several analytical choices
summarise_specs(results,
                 subsets, controls,
                 var = p.value,
                 stats = list(median = median,
                              min = min,
                              max = max))

# Unnamed vector instead of named list passed to 'stats'
summarise_specs(results,
                 controls,
                 stats = c(mean, median))
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