Package 'causalverse'

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```
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Description Causal Verse: An R toolkit expediting causal research & analysis. Streamlines com-
      plex methodologies, empowering users to unveil causal relationships with precision. Your go-
      to for insightful causality exploration..
License GPL-3 | file LICENSE
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Roxygen list(markdown = TRUE)
RoxygenNote 7.2.3
Suggests knitr,
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Imports ggplot2 (>= 3.4.2),
      ggthemes (>= 4.2.4),
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      rio (>= 0.5.29),
      xtable (>= 1.8.4),
      dplyr (>= 1.1.1),
      tidyr (>= 1.3.0),
      scales (>= 1.2.1),
      gridExtra (>= 2.3),
      systemfit (>= 1.1.30),
      Hotelling (>= 1.0.8),
      MatchIt (>= 4.5.4),
      rlang (>= 1.1.1),
      fixest (>= 0.11.1),
      stats (>= 4.2.3),
```

PanelMatch ($\geq 2.0.1$),

2 R topics documented:

```
doParallel (>= 1.0.17),
fastDummies (>= 1.7.3),
magrittr (>= 2.0.3),
did (>= 2.1.2),
synthdid (>= 0.0.9),
plm (>= 2.6.3),
MASS,
foreach (>= 1.5.2)
```

VignetteBuilder knitr

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ama_export_fig	Function to export a figure with custom settings

Description

This function exports a ggplot2 figure to a given path. It exports both an archived version with the current date and a current version without a date. The function supports exporting to PDF and JPG formats.

Usage

```
ama_export_fig(figure, filename, filepath, width = 7, height = 7)
```

Arguments

figure	A ggplot2 object.
filename	A character string specifying the filename without the extension.
filepath	A character string specifying the directory to save the file.
width	The width of the image in inches (default is 7 inches).
height	The height of the image in inches (default is 7 inches).

Examples

```
## Not run:
test_plot <- ggplot(mpg, aes(x=displ, y=hwy)) + geom_point() # Create a ggplot2 plot
filename <- "sample_plot" # Define a filename
filepath <- tempdir() # Define a path using a temporary directory
ama_export_fig(test_plot, filename, filepath) # Call the ama_export_fig function
## End(Not run)</pre>
```

ama_export_tab

Function to export a table with AMA style

Description

This function exports the provided table in both Excel(.xlsx) and LaTeX(.tex) formats. The table is archived with the current date in the filename for the Excel version, while the LaTeX version is saved with just the specified filename.

Usage

```
ama_export_tab(table, filename, filepath, caption = NULL, size = "small")
```

Arguments

table	A data frame or matrix.
filename	A character string specifying the filename without the extension.
filepath	A character string specifying the directory to save the file.
caption	A character string specifying the caption for the table.
size	Latex size including "tiny", or "small"

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Examples

```
## Not run:
data(mtcars) # Load the mtcars dataset
ama_export_tab(mtcars[1:5, 1:5], "sample_table", tempdir(), "Sample Caption for mtcars")
## End(Not run)
```

ama_labs

Custom Label Formatting for ggplot2: American Marketing Association Style

Description

This function provides custom label formatting for ggplot2 based on the guidelines set by the American Marketing Association.

Usage

```
ama_labs(
  title = NULL,
  subtitle = NULL,
  caption = NULL,
  x = NULL,
  y = NULL,
  fill = NULL,
  color = NULL,
  ...
)
```

Arguments

```
title Plot title.

subtitle Plot subtitle.

caption Plot caption.

x X-axis label.

y Y-axis label.

fill Fill legend title.

color Color legend title.

Additional arguments to be passed to ggplot2::labs().
```

Value

Modified labels for a ggplot2 plot.

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Examples

```
## Not run:
library(ggplot2)
ggplot(mtcars, aes(mpg, wt)) + geom_point() +
ama_labs(title = "Sample Plot") +
ama_theme()
## End(Not run)
```

ama_scale_color

Custom Color Scale for ggplot2: American Marketing Association Style

Description

This function provides a custom color scale for ggplot2 plots based on the guidelines set by the American Marketing Association.

Usage

```
ama_scale_color(
  use_color = FALSE,
  palette_name = "OkabeIto",
  grayscale_limits = c(0.2, 0.8)
)
```

Arguments

```
use_color Logical. If TRUE, uses color, otherwise uses grayscale.

palette_name Character. Name of the color palette to use.

grayscale_limits
```

Numeric vector. Limits for the grayscale gradient.

Value

A color scale for a ggplot2 plot.

Examples

```
## Not run:
library(ggplot2)
ggplot(mtcars, aes(mpg, wt, color = gear)) + geom_point(size = 4) + ama_scale_color()
## End(Not run)
```

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ama_scale_fill

Custom Fill Scale for ggplot2: American Marketing Association Style

Description

This function provides a custom fill scale for ggplot2 plots based on the guidelines set by the American Marketing Association.

Usage

```
ama_scale_fill(
  use_color = FALSE,
  palette_name = "OkabeIto",
  grayscale_limits = c(0.2, 0.8)
)
```

Arguments

```
use_color Logical. If TRUE, uses color, otherwise uses grayscale.

palette_name Character. Name of the color palette to use.

grayscale_limits

Numeric vector. Limits for the grayscale gradient.
```

Value

A fill scale for a ggplot2 plot.

Examples

```
## Not run:
library(ggplot2)
ggplot(mtcars, aes(mpg, wt, fill = gear)) +
geom_point(shape = 21, size = 4) +
ama_scale_fill()
## End(Not run)
```

ama_theme

Custom Theme for ggplot2: American Marketing Association Style

Description

This function provides a custom theme for ggplot2 following the guidelines set by the American Marketing Association.

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Usage

```
ama_theme(
  base_size = 16,
  base_family = "sans",
  title_size = ggplot2::rel(1.2),
  axis_title_size = ggplot2::rel(1.2),
  legend_title_size = ggplot2::rel(0.6),
  legend_text_size = ggplot2::rel(0.6),
  axis_text_size = ggplot2::rel(1),
  ...
)
```

Arguments

```
base_size Base font size.

base_family Font family. Use "sans" for Arial and "serif" for Times New Roman.

title_size Title font size as a relative value.

axis_title_size Axis title font size as a relative value.

legend_title_size Legend title font size as a relative value.

legend_text_size Legend text font size as a relative value.

axis_text_size Axis text font size as a relative value.

Additional theme elements to be passed to ggplot2::theme().
```

Value

A ggplot2 theme.

Examples

```
## Not run:
library(ggplot2)
# Using Arial font
ggplot(mtcars, aes(mpg, wt)) + geom_point() + ama_theme()
# Using Times New Roman font
ggplot(mtcars, aes(mpg, wt)) + geom_point() + ama_theme(base_family = "serif")
## End(Not run)
```

balance_assessment

Assess balance between treated and control groups

Description

This function performs a balance assessment between treated and control groups using Seemingly Unrelated Regression (SUR) and Hotelling's T-squared test.

Usage

```
balance_assessment(data, treatment_col, ...)
```

Arguments

data A dataframe containing the data to be assessed.

treatment_col The name of the column that contains the treatment indicator (0 for control, 1 for treated).

... Names of the dependent variables.

Value

A list with two elements: 'SUR' (results of the SUR) and 'Hotelling' (results of the Hotelling's T-squared test).

Examples

```
## Not run:
set.seed(123)
data = mtcars %>%
    dplyr::select(mpg, cyl, disp, hp, wt) %>%
    dplyr::rowwise() %>%
    dplyr::mutate(treatment = sample(c(0,1), 1, replace = TRUE)) %>%
    dplyr::ungroup()

results <- balance_assessment(data, "treatment", "mpg", "cyl")
print(results$SUR)
print(results$Hotelling)
## End(Not run)</pre>
```

balance_scatter_custom

Custom function to visualize the balance between treatment and control groups

Description

Custom function to visualize the balance between treatment and control groups

Usage

```
balance_scatter_custom(
  matched_set_list,
  set.names = NULL,
  show.legend = TRUE,
  legend.title = "Type",
  legend.position = "right",
  xlim = c(0, 0.8),
  ylim = c(0, 0.8),
  main = "Standardized Mean Difference of Covariates",
```

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```
pchs = NULL,
dot.size = NULL,
covariates,
data,
  x.axis.label = "Before Refinement",
  y.axis.label = "After Refinement",
  theme_use = causalverse::ama_theme(),
  ...
)
```

Arguments

matched_set_list

List of matched sets

set.names Vector of names for matched sets. Defaults to NULL.

show. legend Boolean to determine if legend should be shown. Defaults to TRUE.

legend.title Legend title. Defaults to "Type".

legend.position

Position of legend. Defaults to "right".

xlim Vector defining x-axis limits. Defaults to c(0, 0.8). ylim Vector defining y-axis limits. Defaults to c(0, 0.8).

main Main title for the plot. Defaults to "Standardized Mean Difference of Covari-

ates".

pchs Plot characters. Defaults to NULL.

dot.size Size of dots in the scatter plot. Defaults to NULL.

covariates Covariates for calculating balance.

data Dataset for balance calculation.

x.axis.label x-axis label. Defaults to "Before Refinement".y.axis.label y-axis label. Defaults to "After Refinement".

theme_use Custom theme that follows ggplots2. Defaults to causalverse::ama_theme().

.. Additional arguments passed to the labs() function

Value

ggplot object

Examples

```
## Not run:
library(PanelMatch)
# Maha 4-year lag, up to 5 matches
PM.results.maha.4lag.5m <- PanelMatch::PanelMatch(
    lag = 4,
    time.id = "year",
    unit.id = "wbcode2",
    treatment = "dem",
    refinement.method = "mahalanobis",
    data = PanelMatch::dem,
    match.missing = TRUE,
    covs.formula = ~ I(lag(tradewb, 1:4)) + I(lag(y, 1:4)),</pre>
```

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```
size.match = 5,
   qoi = "att",
   outcome.var = "y",
   lead = 0:4,
   forbid.treatment.reversal = FALSE,
   use.diagonal.variance.matrix = TRUE
)
# Maha 4-year lag, up to 10 matches
PM.results.maha.4lag.10m <- PanelMatch::PanelMatch(
   lag = 4,
   time.id = "year",
   unit.id = "wbcode2",
   treatment = "dem",
   refinement.method = "mahalanobis",
   data = PanelMatch::dem,
   match.missing = TRUE,
   covs.formula = \sim I(lag(tradewb, 1:4)) + I(lag(y, 1:4)),
   size.match = 10,
   qoi = "att",
   outcome.var = "y",
   lead = 0:4,
   forbid.treatment.reversal = FALSE,
   use.diagonal.variance.matrix = TRUE
# Using the function
balance_scatter_custom(
   matched_set_list = list(PM.results.maha.4lag.5m$att, PM.results.maha.4lag.10m$att),
   set.names = c("Maha 4 Lag 5 Matches", "Maha 4 Lag 10 Matches"),
   data = dem,
   covariates = c("y", "tradewb")
)
## End(Not run)
```

get_balanced_panel

Extract a Balanced Panel

Description

This function extracts a balanced panel from the data for a specific adoption cohort. It drops units with missing observations at any point within its entire specified window (leads + adoption time, adoption time + lags). Units treated in the same period as the adoption cohort are marked as "treated," and units with their first treatment after the leads time of the specified adoption cohort are marked as "control."

Usage

```
get_balanced_panel(
  data = data,
  adoption_cohort,
  lags,
  leads,
```

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```
time_var,
unit_id_var,
treated_period_var,
filter_units = TRUE
)
```

Arguments

data The dataset to be used.

adoption_cohort

Numeric, the specific adoption cohort.

lags Numeric, the number of lags.

leads Numeric, the number of leads.

time_var String, the name of the time variable.
unit_id_var String, the name of the unit ID variable.

treated_period_var

String, the name of the treated period variable.

filter_units Logical, whether to filter only units with data on all time periods within the

specified time window. Defaults to TRUE.

Value

A data frame with the balanced panel.

Examples

lee_bounds

Summarize Lee Bounds for Always-Takers

Description

Computes and summarizes the Lee bounds on the average direct effect for always-takers (ATs) for whom there is a direct effect of treatment (D) on the outcome (Y). This function utilizes compute_bounds_ats to calculate initial bounds and applies bootstrapping to estimate the standard deviation of these estimates, providing a summary in a data frame format.

med_ind

Usage

```
lee_bounds(
   df,
   d,
   m,
   y,
   cluster = NULL,
   c_at_ratio = NULL,
   units = "",
   numdraws = 1000
)
```

Arguments

df	A data frame containing the data.
d	Name of the treatment variable in df.
m	Name of the mediator variable in df.
у	Name of the outcome variable in df.
cluster	(Optional) The name of the cluster variable for clustered bootstrapping.
c_at_ratio	(Optional) Specifies the ratio of $EY(1,1) \mid C/EY(1,1) \mid AT$. If this is specified, the direct effect for ATs is point-identified.
units	A string denoting the units of the outcome variable (for labeling purposes).
numdraws	The number of bootstrap draws for estimating the standard deviation.

Value

A data frame summarizing the computed bounds with terms, estimates, and standard errors.

Examples

```
## Not run:
data(example_data)
summarized_bounds <- lee_bounds(df = example_data, d = "treatment", m = "mediator", y = "outcome")
## End(Not run)</pre>
```

med_ind Estimate Mediation Indirect Effects

Description

med_ind estimates the indirect effects of an independent variable on a dependent variable through a mediator using Monte Carlo simulations (Selig & Preacher, 2008). It calculates the distribution of the product of path coefficients (a*b) and provides confidence intervals for the indirect effect, along with a ggplot histogram for visualization.

med_ind

Usage

```
med_ind(
    a,
    b,
    var_a,
    var_b,
    cov_ab,
    ci = 95,
    iterations = 20000,
    seed = 1,
    theme = causalverse::ama_theme()
)
```

Arguments

a	The regression coefficient for the effect of the independent (causal) variable on the mediator.
b	The regression coefficient for the effect of the mediator on the dependent (outcome) variable.
var_a	The variance of the coefficient a.
var_b	The variance of the coefficient b.
cov_ab	The covariance between coefficients a and b.
ci	The confidence interval width for the indirect effect (default is 95 for a 95% CI).
iterations	The number of iterations for the Monte Carlo simulation (default is 20000).
seed	The seed for random number generation to ensure reproducibility (default is 1).
theme	Custom theme that follows ggplots2 (default is AMA style)

Value

A list containing the lower quantile, upper quantile, raw simulation data, and histogram plot of the indirect effects.

References

Selig, J. P., & Preacher, K. J. (2008, June). Monte Carlo method for assessing mediation: An interactive tool for creating confidence intervals for indirect effects Computer software. Available from http://quantpsy.org/.

Examples

```
## Not run:
result <- med_ind(a = 0.5, b = 0.7, var_a = 0.04, var_b = 0.05, cov_ab = 0.01)
result$lower_quantile
result$upper_quantile
result$plot
## End(Not run)</pre>
```

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nice_tab

Nice Tabulation Function

Description

Create a custom function that takes a data frame and a number of decimal places as input, rounds all numeric columns in the data frame to the specified number of decimal places, and returns the modified data frame.

Usage

```
nice_tab(data, digit_decimal = 2)
```

Arguments

data A data frame.

digit_decimal A number of decimal places.

Value

A data frame with all numeric columns rounded to the specified number of decimal places.

panel_estimate

Panel Estimate Function

Description

This function computes estimates and standard errors for panel data using selected estimators. It allows the user to select specific estimators and set parameters for Monte Carlo replications and seed.

Usage

```
panel_estimate(
   setup,
   selected_estimators = setdiff(names(panel_estimators), "mc"),
   mc_replications = 200,
   seed = 1
)
```

Arguments

setup A list containing matrices Y, N0, and T0 for panel data analysis.

selected_estimators

A character vector specifying which estimators to use. For example, c("synthdid", "did", "sc", "difp", "mc", "sc_ridge", "difp_ridge") or names(panel_estimators). Defaults to all available estimators except 'mc'.

mc_replications

The number of Monte Carlo replications for computing standard errors. Applicable if the 'mc' estimator is used. Defaults to 200.

seed An integer value to set the random seed for reproducibility. Defaults to 1.

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Value

A list where each element corresponds to an estimator and contains its estimate and standard error.

Examples

```
## Not run:
data('california_prop99')
setup = panel.matrices(california_prop99)
results_all = panel_estimate(setup)
results_selected = panel_estimate(setup, selected_estimators = c("did", "sc"))
summary(results_selected$did$estimate)
## End(Not run)
```

Description

This function generates coefplots or iplots based on fixest outputs, allowing the user to visualize interaction coefficients with ease.

Usage

```
plot_coef_par_trends(
   data,
   dependent_vars,
   time_var,
   unit_treatment_status,
   unit_id_var,
   plot_type = "coefplot",
   combined_plot = TRUE,
   legend_position = "bottomleft",
   legend_title = "Legend Title",
   legend_args = list(),
   plot_args = list()
```

Arguments

data Data frame containing the data to be used in the model.

dependent_vars Named list of dependent variables to model and their respective labels.

time_var Name of the time variable in the data.

unit_treatment_status

Name of the treatment status variable.

unit_id_var Name of the unit identification variable.

plot_type Type of plot to generate. Either "coefplot" or "iplot".

combined_plot Logical indicating whether to combine plots for all dependent variables.

```
legend_position
Position of the legend on the plot.

legend_title Title for the legend.

legend_args List of additional arguments to customize the legend.

plot_args List of additional arguments to customize the plot.
```

Value

A plot visualizing interaction coefficients.

Examples

```
## Not run:
library(fixest)
data("base_did")
# Sample call to the function:
plot_coef_par_trends(
  data = base_did,
  dependent_vars = c(y = "Outcome 1", x1 = "Outcome 2"),
  time_var = "period",
  unit_treatment_status = "treat",
  unit_id_var = "id",
  plot_type = "coefplot",
  combined_plot = TRUE,
  plot_args = list(main = "Interaction coefficients Plot"),
  legend_title = "Metrics",
  legend_position = "bottomleft"
plot_coef_par_trends(
  data = base_did,
  dependent_vars = c(y = "Outcome 1", x1 = "Outcome 2"),
  time_var = "period",
  unit_treatment_status = "treat",
  unit_id_var = "id",
 plot_type = "coefplot",
  combined_plot = FALSE
## End(Not run)
```

plot_covariate_balance_pretrend

Plot Covariate Balance Over Pre-Treatment Period

Description

This function visualizes the covariate balance over the pre-treatment period. It's particularly designed for outputs from methods like PanelMatch.

Usage

```
plot_covariate_balance_pretrend(
   balance_data,
   y_limits = c(-1, 1),
   theme_use = causalverse::ama_theme(),
   xlab = "Time to Treatment",
   ylab = "Balance (in SD unit)",
   main_title = "Covariate Balance Over Pre-Treatment Period",
   legend_title = "Covariate",
   show_legend = TRUE,
   ...
)
```

Arguments

balance_data	A matrix containing the covariate balance data over the pre-treatment period.
y_limits	A numeric vector of length 2 defining the y-axis limits.
theme_use	A ggplot2 theme. By default, it uses causalverse::ama_theme().
xlab	A string indicating the label for the x-axis.
ylab	A string indicating the label for the y-axis.
main_title	A string for the main title of the plot.
legend_title	A string for the legend title.
show_legend	A logical; if TRUE, the legend is displayed, otherwise, it's hidden.
	Additional arguments passed to the ggplot labs.

Value

A ggplot2 object.

Examples

```
## Not run:
   balance_data_sample <- matrix(rnorm(20), nrow = 5)
   plot_covariate_balance_pretrend(balance_data_sample)
## End(Not run)</pre>
```

```
plot_density_by_treatment
```

Plot Density by Treatment

Description

This function creates a list of ggplot density plots for specified variables by treatment groups.

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Usage

```
plot_density_by_treatment(
  data,
  var_map,
  treatment_var,
  show_legend = TRUE,
  theme_use = ggplot2::theme_minimal(),
  ...
)
```

Arguments

A data frame containing the variables to plot and a treatment variable.

A named list mapping the column names in the data to display names for plot-

tin

treatment_var A named vector where the name is the treatment column in the data and the

value is the legend title.

show_legend A logical value indicating whether to show the legend. Defaults to TRUE.

theme_use ggplot2 theme. Defaults to ggplot2::theme_minimal().
... Additional arguments to be passed to geom_density.

Value

A list of ggplot objects for each variable in var_map.

Examples

plot_PanelEstimate

Plot Estimated Effects of Treatment Over Time

Description

This function takes an object (result of PanelEstimate or similar) and plots its estimates over time.

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Usage

```
plot_PanelEstimate(
   pe.object,
   ylab = "Estimated Effect of Treatment",
   xlab = "Time Since Treatment",
   main = "Estimated Effects of Treatment Over Time",
   ylim = NULL,
   theme_use = causalverse::ama_theme(),
   ...
)
```

Arguments

pe.object	The object with the estimation results.
ylab	The y-axis label.
xlab	The x-axis label.
main	The main title for the plot.
ylim	The limits for the y-axis.
theme_use	The theme to use for the plot. Defaults to causalverse::ama_theme().
	Additional parameters to pass to labs() function.

Value

A ggplot object with the desired plot.

Examples

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Description

Plots parallel trends for given metrics.

Usage

```
plot_par_trends(
   data,
   metrics_and_names,
   treatment_status_var,
   time_var,
   conf_level = 0.95,
   non_negative = FALSE,
   display_CI = TRUE,
   output_format = "plot",
   smoothing_method = NULL,
   title_prefix = "Parallel Trends for",
   theme_use = causalverse::ama_theme()
)
```

Arguments

data A data frame containing the data to plot. metrics_and_names A named list of metrics to plot. treatment_status_var The variable indicating treatment status. time_var The variable indicating time. conf_level Confidence level for confidence intervals (default is 0.95). non_negative Logical; if TRUE, sets negative lower confidence bounds to 0. display_CI Logical; if TRUE, displays confidence intervals. output_format Format of the output; "plot" returns a list of ggplots, "data.frame" returns a data frame. smoothing_method Method to use for smoothing; NULL means no smoothing. title_prefix A character string specifying the prefix for the plot title (default is "Parallel Trends for").

Custom theme that follows ggplots2

Value

theme_use

A list of ggplot objects or a data frame.

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Examples

```
## Not run:
library(tidyverse)
data <- expand.grid(entity = 1:100, time = 1:10) %>%
  dplyr::arrange(entity, time) %>%
  dplyr::mutate(
    treatment = ifelse(entity <= 50, "Treated", "Control"),</pre>
    outcome1 = 0.5 * time + rnorm(n(), 0, 2) + ifelse(treatment == "Treated", 0, 0),
    outcome2 = 3 + 0.3 * time + rnorm(n(), 0, 1) + ifelse(treatment == "Treated", 0, 2)
results <- plot_par_trends(</pre>
  data = data,
  metrics_and_names = list(outcome1 = "Outcome 1", outcome2 = "Outcome 2"),
  treatment_status_var = "treatment",
  time_var = list(time = "Time"),
  smoothing_method = "loess"
library(gridExtra)
gridExtra::grid.arrange(grobs = results, ncol = 1)
## End(Not run)
```

plot_rd_aa_share

Plot RD Always-assigned Share

Description

This function creates a plot for the share of always-assigned units in a Regression Discontinuity (RD) design, either Sharp RD (SRD) or Fuzzy RD (FRD). It provides options to include various confidence intervals and reference lines.

Usage

```
plot_rd_aa_share(
   data,
   rd_type = "SRD",
   x_label = "Share of Always-assigned Units",
   y_label = "ATE",
   plot_title = "",
   theme_use = causalverse::ama_theme(),
   tau = TRUE,
   tau_CI = FALSE,
   bounds_CI = TRUE,
   ref_line = 0,
   ...
)
```

Arguments

data

The output object from the rdbounds function.

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rd_type	The type of RD design, either "SRD" for Sharp RD or "FRD" for Fuzzy RD. Default is "SRD".
x_label	The label for the x-axis. Default is "Share of Always-assigned Units".
y_label	The label for the y-axis. Default is "ATE".
plot_title	The title of the plot. Default is an empty string.
theme_use	A ggplot 2 theme function to apply to the plot. Default is causalverse: : ama_theme().
tau	Logical, whether to include a vertical line at the estimated treatment effect. Default is TRUE.
tau_CI	Logical, whether to include confidence intervals for the treatment effect estimate. Default is FALSE.
bounds_CI	Logical, whether to include confidence intervals for the manipulation bounds. Default is TRUE.
ref_line	The y-intercept for a reference line. Default is 0.
	Additional arguments passed to labs in ggplot2.

Value

A ggplot object.

Examples

```
## Not run:
set.seed(1)
data <- rdbounds::rdbounds_sampledata(10000, covs = FALSE)</pre>
rdbounds_est_tau <- rdbounds::rdbounds(</pre>
   y = data$y,
    x = data$x,
    treatment = data$treatment,
    c = 0,
    discrete_x = FALSE,
    discrete_y = FALSE,
    bwsx = c(.2, .5),
    bwy = 1,
    kernel = "epanechnikov",
    orders = 1,
    evaluation_ys = seq(from = 0, to = 15, by = 1),
    refinement_A = TRUE,
    refinement_B = TRUE,
    right_effects = TRUE,
    potential_taus = c(.025, .05, .1, .2),
    yextremes = c(0, 15),
    num\_bootstraps = 5
)
plot_rd_aa_share(rdbounds_est_tau)
## End(Not run)
```

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plot_treat_time

Plot number of treated units over time or return a dataframe.

Description

Plot number of treated units over time or return a dataframe.

Usage

```
plot_treat_time(
  data,
  time_var,
  unit_treat,
  outlier_method = "iqr",
  show_legend = FALSE,
  theme_use = causalverse::ama_theme(),
  legend_title = "Point Type",
legend_labels = c("Regular", "Outlier"),
  regular_size = 3,
  outlier_size = 5,
  regular_color = "black",
  outlier_color = "red",
  regular_shape = 16,
  outlier_shape = 17,
  title = "Random Time Assignment",
  xlab = "Time",
  ylab = "Number of Treated Units",
  output = "plot",
)
```

Arguments

data	Dataframe containing data.
time_var	Time variable for aggregating the number of treated units.
unit_treat	Variable indicating if the unit was treated in a specific time period.
outlier_method	Method for outlier detection ("iqr" or "z-score").
show_legend	Logical indicating whether to show legend.
theme_use	ggplot2 theme to use.
legend_title	Title for legend.
legend_labels	Labels for regular and outlier points.
regular_size	Size of regular points.
outlier_size	Size of outlier points.
regular_color	Color of regular points.
outlier_color	Color of outlier points.
regular_shape	Shape of regular points.
outlier_shape	Shape of outlier points.

```
title Plot title.

xlab X-axis label.

ylab Y-axis label.

output Type of output ("plot" or "dataframe").

... Additional arguments to pass to ggplot2::labs.
```

Value

ggplot2 object or dataframe.

Examples

```
# Example usage:
## Not run:
data <- data.frame(time = c(1,1,2,2,3,3), treat = c(0,1,1,1,0,0))
plot_treat_time(data, time_var = time, unit_treat = treat)
plot_treat_time(data, time_var = time, unit_treat = treat, output = "dataframe")
## End(Not run)</pre>
```

```
plot_trends_across_group
```

Custom Faceted Line Plot with Optional Standard Error

Description

This function generates a faceted line plot for a given dataset, allowing the user to specify the x-axis, y-axis, grouping variable, and facet variable. Additionally, users can include standard errors and customize labels.

Usage

```
plot_trends_across_group(
   data,
   x_var,
   y_var,
   grouping_var,
   facet_var,
   se = NULL,
   include_legend = TRUE,
   title = "Dependent Variable across Years by Group and Industry",
   x_label = "Year",
   y_label = "Dependent Variable",
   theme = causalverse::ama_theme(),
   ...
)
```

Arguments

data A data frame containing the data to be plotted.

x_var A character string specifying the x-axis variable.

y_var A character string specifying the y-axis variable.

grouping_var A character string specifying the grouping variable.

facet_var A character string specifying the facet variable.

se A character string specifying the standard error variable, or NULL (default) if

not provided.

include_legend Logical. If TRUE, includes the legend, otherwise it does not.

title Character string specifying the main plot title.

x_label Character string specifying the x-axis label.

y_label Character string specifying the y-axis label.

theme A ggplot2 theme. Defaults to ama_theme.

... Additional arguments passed to labs.

Value

A ggplot object.

Examples

```
## Not run:
# Create a small sample dataset
sample_data <- data.frame(</pre>
  year = rep(2001:2005, each = 2),
  dependent_variable = rnorm(10, mean = 50, sd = 10),
  group = rep(c("treated", "control"), times = 5),
  industry = rep(c("Tech", "Healthcare"), each = 5)
)
# Use the function
plot_trends_across_group(data = sample_data,
                        x_{var} = "year",
                         y_var = "dependent_variable",
                         grouping_var = "group",
                         facet_var = "industry",
                         title = "Sample Title")
## End(Not run)
```

process_panel_estimate

Process Panel Estimate

Description

This function processes the output from panel_estimate() for panel estimates and returns a formatted data frame. It takes a list of results, each corresponding to a different method, and combines them into a single data frame. The data frame includes the method name, estimate, and standard error for each method.

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Usage

```
process_panel_estimate(results_selected)
```

Arguments

```
results_selected
```

A list of results from panel_estimate(). Each element in the list should be an object containing the results for a particular estimation method. Each object must have an estimate and a std.error attribute.

Value

A data frame with columns Method, Estimate, and SE, representing the method name, the estimate value, and the standard error, respectively. The data frame is formatted using causalverse::nice_tab().

Examples

```
## Not run:
library(synthdid)
setup = synthdid::panel.matrices(synthdid::california_prop99)
results_selected = panel_estimate(setup, selected_estimators = c("did", "sc"))
results_table = process_panel_estimate(results_selected)
print(results_table)
## End(Not run)
```

stack_data

Stacked Data for Staggered DiD Analysis

Description

stack_data processes datasets used in staggered Difference-in-Differences (DiD) designs. Staggered DiD designs arise when different units (e.g., firms, regions, countries) get treated at different time periods. This function creates cohorts based on the provided treatment period variable and stacks them together to create a comprehensive longitudinal format suitable for staggered DiD analyses.

Usage

```
stack_data(
   treated_period_var,
   time_var,
   pre_window,
   post_window,
   data,
   control_type = c("both", "never-treated", "not-yet-treated")
)
```

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Arguments

treated_period_var

A character string indicating the column name of the treatment period variable.

time_var A character string indicating the column name for time.

pre_window An integer indicating the number of periods before the treatment to consider

(i.e., leads).

post_window An integer indicating the number of periods after the treatment to consider (i.e.,

lags).

data A data frame containing the dataset to be processed.

control_type A character string indicating which control type to use. One of "both", "never-

treated", or "not-yet-treated".

Details

The function emphasizes the importance of having a control group, which should be represented by the value 10000 in the treated_period_var column of the provided dataset. The output data will be augmented with relative period dummy variables for ease of subsequent analysis.

Value

A data frame with the stacked data, augmented with relative period dummy variables, suitable for staggered DiD analysis.

Examples

```
## Not run:
    library(did)
    library(tidyverse)
    library(fixest)
    data(base_stagg)
    stacked_data <- stack_data("year_treated", "year", 3, 3, base_stagg, control_type = "both")
    feols_result <- feols(as.formula(paste0(
        "y ~ ",
        paste(paste0("`rel_period_", c(-3:-2, 0:3), "`"), collapse = " + "),
        " | id ^ df + year ^ df"
    )), data = stacked_data)
    print(feols_result)
## End(Not run)</pre>
```

 $synthdid_est$

Synthetic DID Estimation Using synthdid Package

Description

This function estimates synthetic difference-in-differences using the synthdid package. It offers a choice among synthdid_estimate, did_estimate, and sc_estimate methods for estimation, defaulting to synthdid_estimate. It calculates treatment effects (TEs) for each period instead of a single TE for all treated periods.

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Usage

```
synthdid_est(
  data,
  adoption_cohort,
  subgroup = NULL,
  lags,
  leads,
  time_var,
  unit_id_var,
  treated_period_var,
  treat_stat_var,
  outcome_var,
  seed = 1,
  method = "synthdid"
)
```

Arguments

data Data frame to analyze.

adoption_cohort

Cohort in data to use as treated.

subgroup (Optional) List of IDs to use as treated subgroup.

lags Number of lags to use pre-treatment.

leads Number of post-treatment periods (0 for only the treatment period).

time_var Name of the calendar time column.

unit_id_var Name of the unit ID column.

treated_period_var

Name of the treatment time period column.

treat_stat_var Name of the treatment indicator column.

outcome_var Name of the outcome variable column.

seed A numeric value for setting the random seed (only for placebo SE). Default is 1.

method The estimation method to be used. Methods include:

- 'did': Difference-in-Differences.
- 'sc': Synthetic Control Method.
- 'sc_ridge': Synthetic Control Method with Ridge Penalty. It adds a ridge regularization to the synthetic control method when estimating the synthetic control weights.
- 'difp': De-meaned Synthetic Control Method, as proposed by Doudchenko and Imbens (2016) and Ferman and Pinto (2021).
- 'difp_ridge': De-meaned Synthetic Control with Ridge Penalty. It adds a ridge regularizationd when estimating the synthetic control weights.
- 'synthdid': Synthetic Difference-in-Differences, a method developed by Arkhangelsky et al. (2021) Defaults to 'synthdid'.

Value

A list containing the estimated treatment effects, standard errors, observed and predicted outcomes, synthetic control lambda weights, and counts of treated and control units.

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References

Ferman, B., & Pinto, C. (2021). Synthetic controls with imperfect pretreatment fit. Quantitative Economics, 12(4), 1197-1221.

Doudchenko, Nikolay, and Guido W. Imbens. 2016. "Balancing, Regression, Difference-in-Differences and Synthetic Control Methods: A Synthesis." NBER Working Paper 22791.

Arkhangelsky, D., Athey, S., Hirshberg, D. A., Imbens, G. W., & Wager, S. (2021). Synthetic difference-in-differences. American Economic Review, 111(12), 4088-4118.

Examples

```
## Not run:
  library(tidyverse)
  library(causalverse)
  library(synthdid)
  data <- get_balanced_panel(</pre>
    data = fixest::base_stagg,
    adoption_cohort = 5,
    lags = 2,
    leads = 3,
    time_var = "year";
    unit_id_var = "id",
    treated_period_var = "year_treated"
  ) |>
    dplyr::mutate(treatvar = if_else(time_to_treatment >= 0, 1, 0)) |>
    dplyr::mutate(treatvar = as.integer(if_else(year_treated > (5 + 2), 0, treatvar)))
  synthdid_est(
    data,
    adoption_cohort = 5,
    lags = 2,
    leads = 3,
    time_var = "year"
    unit_id_var = "id",
    treated_period_var = "year_treated",
    treat_stat_var = "treatvar",
    outcome_var = "y"
  )
## End(Not run)
```

synthdid_est_ate

Estimate the SynthDiD ATEs and Standard Errors

Description

This function uses an adapted SynthDiD method (Arkhangelsky et al., 2021) to estimate the average treatment effect for staggered adoption scenarios. It combines cohort-level ATT estimates, similar to the approach in Ben-Michael et al. (2022), for synthetic controls with staggered adoption. The function is designed to handle various cohorts, lags, leads, placebo tests, and pooled analyses.

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Usage

```
synthdid_est_ate(
  data,
  adoption_cohorts,
  lags,
  leads,
  time_var,
  unit_id_var,
  treated_period_var,
  treat_stat_var,
  outcome_var,
  placebo = F,
 pooled = F,
  subgroup = NULL,
  conf_level = 0.95,
  seed = 1.
 method = "synthdid"
)
```

Arguments

data A data frame in long format to be analyzed.

adoption_cohorts

Vector of cohorts to use for adoption times.

lags Integer, number of lags of adoption time to analyze.

leads Integer, number of leads of adoption time to analyze.

time_var String, column name of time variables.

unit_id_var String, ID column of units.

treated_period_var

String, column with adoption time of each unit.

treat_stat_var String, column name indicating treatment status.

outcome_var String, column of outcome to analyze.

placebo Logical, whether to run placebo analysis.

pooled Logical, whether to run pooled analysis of all treated units.

subgroup Vector, IDs for subgroup analysis.

conf_level Numeric, confidence level for the interval estimation (Default: 95%).

seed A numeric value for setting the random seed (for placebo SE and placebo anal-

ysis). Default is 1.

method The estimation method to be used. Methods include:

- 'did': Difference-in-Differences.
- 'sc': Synthetic Control Method.
- 'sc_ridge': Synthetic Control Method with Ridge Penalty. It adds a ridge regularization to the synthetic control method when estimating the synthetic control weights.
- 'difp': De-meaned Synthetic Control Method, as proposed by Doudchenko and Imbens (2016) and Ferman and Pinto (2021).
- 'difp_ridge': De-meaned Synthetic Control with Ridge Penalty. It adds a ridge regularizationd when estimating the synthetic control weights.

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• 'synthdid': Synthetic Difference-in-Differences, a method developed by Arkhangelsky et al. (2021) Defaults to 'synthdid'.

Value

A list containing the following elements:

- time: Vector of time periods used in estimation from -lags to leads (relative to the adoption period)
- TE_mean: Vector of ATT in each time period
- SE mean: Vector of Standard error of ATT each time period
- TE_mean_lower: Vector of Lower C.I. for ATT per period
- TE_mean_upper: Vector of Upper C.I. for ATT per period
- TE_mean_w, SE_mean_w, TE_mean_w_lower, TE_mean_w_upper: Weighted versions of the above metrics by the number of treated units in each time period
- Ntr: Number of treated units
- Nco: Number of control units
- TE: Treatment effect for each cohort in each time period
- SE: Standard error of TE of each cohort in each time period
- y_obs: Observed outcomes of treated units
- y_pred: Predicted outcomes of treated units
- col names: Column names for TE and SE matrices (times and ATTs)

References

Arkhangelsky, D., Athey, S., Hirshberg, D. A., Imbens, G. W., & Wager, S. (2021). Synthetic difference-in-differences. American Economic Review, 111(12), 4088-4118. American Economic Association 2014 Broadway, Suite 305, Nashville, TN 37203.

Ben-Michael, E., Feller, A., & Rothstein, J. (2022). Synthetic controls with staggered adoption. Journal of the Royal Statistical Society Series B: Statistical Methodology, 84(2), 351-381. Oxford University Press.

Ferman, B., & Pinto, C. (2021). Synthetic controls with imperfect pretreatment fit. Quantitative Economics, 12(4), 1197-1221.

Doudchenko, Nikolay, and Guido W. Imbens. 2016. "Balancing, Regression, Difference-in-Differences and Synthetic Control Methods: A Synthesis." NBER Working Paper 22791.

Arkhangelsky, D., Athey, S., Hirshberg, D. A., Imbens, G. W., & Wager, S. (2021). Synthetic difference-in-differences. American Economic Review, 111(12), 4088-4118.

Examples

```
## Not run:
    library(tidyverse)
    data <- fixest::base_stagg |>
        mutate(treatvar = if_else(time_to_treatment >= 0, 1, 0)) |>
        mutate(treatvar = as.integer(if_else(year_treated > (5 + 2), 0, treatvar)))

synthdid_est_ate(
    data = data,
    adoption_cohorts = 5:7,
```

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```
lags = 2,
leads = 2,
time_var = "year",
unit_id_var = "id",
treated_period_var = "year_treated",
treat_stat_var = "treatvar",
pooled = F,
outcome_var = "y"
)
## End(Not run)
```

synthdid_est_per

Estimate Treatment Effects for Each Period

Description

Given the output from the synthdid::synthdid_estimate method, this function computes the treatment effects (TEs) for each post-treatment period, along with the cumulative average treatment effect (ATE). It also provides observed and predicted outcomes for treated units, synthetic control weights, and counts of treated and control units.

Usage

```
synthdid_est_per(Y, N0, T0, weights)
```

Arguments

Y Data matrix with units as rows and time periods as columns.

N0 Number of control units.

T0 Number of pre-treatment periods.

weights Output from synthdid, containing lambda and omega weights.

Value

A list containing:

- est: TEs for each post-treatment period and cumulative ATEs.
- y_obs: Observed outcomes for treated units.
- y_pred: Predicted outcomes for treated units.
- lambda.synth: Synthetic control lambda weights.
- Ntr: Number of treated units.
- Nco: Number of control units.

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Examples

```
## Not run:
library(tidyverse)
library(synthdid)
library(fixest)
setup <- base_did |>
  mutate(
   id = as.factor(id),
   period = as.integer(period),
   y = as.double(y),
   post = as.integer(post)
  ) |>
  # Correct treatment
  dplyr::mutate(treatment = as.integer(if_else(treat == 0, 0, post))) |>
 synthdid::panel.matrices(unit = "id", time = "period", outcome = "y", treatment = "treatment")
sdid <- synthdid::synthdid_estimate(setup$Y, setup$N0, setup$T0)</pre>
synthdid_est_per(setup$Y, setup$N0, setup$T0, weights = attr(sdid, 'weights'))
## End(Not run)
```

synthdid_plot_ate

Create ATE Plot Using ggplot2

Description

This function creates a ggplot for visualizing Average Treatment Effect (ATE) from a given estimation object.

Usage

```
synthdid_plot_ate(
  est,
  show_CI = TRUE,
  title = "",
  xlab = "Relative Time Period",
  ylab = "ATE",
  y_intercept = 0,
  theme = causalverse::ama_theme(),
  fill_color = "lightgrey"
)
```

Arguments

```
est Estimation object from synthdid_est_ate.

show_CI Logical; if TRUE, shows confidence intervals on the plot.

title String; title of the plot.

xlab String; label for the x-axis.

ylab String; label for the y-axis.

y_intercept Numeric; value at which a horizontal line is drawn.
```

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```
theme ggplot theme; default is set to causalverse::ama_theme().
fill_color String; color used for the confidence interval shading.
```

Value

A ggplot object representing the ATE plot.

Examples

```
## Not run:
  # Load required libraries
  library(ggplot2)
  library(tidyverse)
  library(causalverse)
  library(tidyverse)
  data <- fixest::base_stagg |>
   dplyr::mutate(treatvar = if_else(time_to_treatment >= 0, 1, 0)) |>
   dplyr::mutate(treatvar = as.integer(if\_else(year\_treated > (5 + 2), \ 0, \ treatvar)))
   synthdid_est_ate(
      data = data,
      adoption_cohorts = 5:7,
      lags = 2,
      leads = 2,
      time_var = "year"
      unit_id_var = "id",
      treated_period_var = "year_treated",
      treat_stat_var = "treatvar",
      pooled = FALSE,
      placebo = FALSE,
      outcome_var = "y"
   )
  # Generate the plot
  synthdid_plot_ate(est, show_CI = TRUE, title = "Sample ATE Plot")
## End(Not run)
```

Description

Computes the standard error of estimates using the jackknife method. It is specifically tailored for use with synthetic difference-in-differences estimates from the synthdid package. This function supports both the usual jackknife estimate of variance and the fixed-weights jackknife estimate as described by Arkhangelsky et al.

Usage

```
synthdid_se_jacknife(estimate, weights = attr(estimate, "weights"), seed = 1)
```

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Arguments

estimate A synthdid estimate object.

weights Optional; custom weights for the fixed-weights jackknife. If NULL, the usual

jackknife estimate is calculated.

seed A numeric value for setting the random seed (only for placebo SE). Default is 1.

Value

Returns the standard error of the provided estimate.

References

Arkhangelsky, D., Athey, S., Hirshberg, D. A., Imbens, G. W., & Wager, S. (2021). Synthetic difference-in-differences. American Economic Review, 111(12), 4088-4118.

Examples

```
## Not run:
setup <- get_balanced_panel(</pre>
  data = fixest::base_stagg,
  adoption\_cohort = 5,
  lags = 2,
  leads = 3.
  time_var = "year",
  unit_id_var = "id",
  treated_period_var = "year_treated"
  dplyr::mutate(treatvar = if_else(time_to_treatment >= 0, 1, 0)) |>
  dplyr::mutate(treatvar = as.integer(if_else(year_treated > (5 + 2), 0, treatvar))) |>
  synthdid::panel.matrices(
    unit = "id",
    time = "year"
    outcome = "y",
    treatment = "treatvar"
estimate <- synthdid::synthdid_estimate(setup$Y, setup$N0, setup$T0)</pre>
se_results <- synthdid_se_jacknife(estimate, seed = 123)</pre>
## End(Not run)
```

synthdid_se_placebo

Calculate Placebo Standard Errors for Synthetic DID

Description

Computes placebo standard errors for synthetic difference-in-differences (DID) estimates. This function is based on the methodology described in Arkhangelsky et al. (2021). It is particularly useful when there is only one treated unit and performs a bootstrap procedure to estimate the standard errors.

Usage

```
synthdid_se_placebo(estimate, replications = 10000, seed = 1)
```

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Arguments

estimate An estimate object obtained from synthetic DID estimation.

replications The number of bootstrap replications to perform. Defaults to 500.

seed A numeric value for setting the random seed. Default is 1.

Value

A vector of standard errors corresponding to the input estimates.

References

Arkhangelsky, D., Athey, S., Hirshberg, D. A., Imbens, G. W., & Wager, S. (2021). Synthetic Difference-in-Differences. American Economic Review, 111(12), 4088-4118. American Economic Association 2014 Broadway, Suite 305, Nashville, TN 37203.

Examples

```
## Not run:
setup <- get_balanced_panel(</pre>
  data = fixest::base_stagg,
  adoption_cohort = 5,
  lags = 2,
  leads = 3,
  time_var = "year"
  unit_id_var = "id"
  treated_period_var = "year_treated"
) |>
  # get treatment status
  dplyr::mutate(treatvar = if_else(time_to_treatment >= 0, 1, 0)) |>
  \mbox{\#} correct those control units to have treatment status to be 0
  dplyr::mutate(treatvar = as.integer(if_else(year_treated > (5 + 2), 0, treatvar))) |>
  synthdid::panel.matrices(
    unit = "id",
    time = "year",
    outcome = "y",
    treatment = "treatvar"
  )
estimate <- synthdid::synthdid_estimate(setup$Y, setup$N0, setup$T0)</pre>
se_results <- synthdid_se_placebo(estimate, replications = 1000)</pre>
## End(Not run)
```

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