

Package ‘spliv’

July 29, 2026

Type Package

Title Patterned Sensitivity Analysis for IV with Fixed Effects

Version 0.1.1

Description Patterned sensitivity analysis for instrumental-variables designs with fixed effects or other residualization steps. The package provides uniform Conley-style sensitivity as a baseline, researcher-specified direct-effect patterns, sensitivity paths and tipping points, and optional confirmatory Beyond Plausibly Exogenous diagnostics.

License GPL-3

Encoding UTF-8

Depends R (>= 4.1)

Imports stats, graphics, fixest

Suggests lfe, testthat (>= 3.0.0), ggplot2, knitr, rmarkdown

Config/testthat/edition 3

VignetteBuilder knitr

URL <https://koren6684.github.io/spliv/>,
<https://github.com/koren6684/spliv>

BugReports <https://github.com/koren6684/spliv/issues>

RoxygenNote 7.3.3

NeedsCompilation no

Author Ore Koren [aut, cre]

Maintainer Ore Koren <okoren@iu.edu>

Repository CRAN

Date/Publication 2026-07-29 16:50:02 UTC

Contents

bpe_design	2
bpe_eval_subset	3
bpe_explore_subsets	4
bpe_find_subset	5
bpe_validate_design	6
demean_fixest	8
demean_lfe	9
embed_prior_into_full_Z	9
estimate_gamma_zero_first_stage	10
iv_inst_names	11
plot_sp_sensitivity	12
spliv	13
spliv_eval_pattern	16
spliv_pattern	17
spliv_sensitivity_path	18
spliv_tipping_point	19
sp_ltz	20
sp_prior_ltz	21
sp_sensitivity_ltz_normal	22
sp_sensitivity_ltz_uniform01_as_normal	23
sp_sensitivity_uci_support	24
sp_uci	26
Index	28

bpe_design	<i>Create a Confirmatory BPE Design Object</i>
------------	--

Description

Creates a researcher-specified design object for confirmatory BPE. The subset must be justified outside the outcome analysis and should represent a theoretically motivated instrument-inactive subset.

Usage

```
bpe_design(
  name,
  subset,
  rationale,
  variables_used = NULL,
  subset_type = NULL,
  pre_specified = TRUE,
  transportability_rationale = NULL,
  notes = NULL
)
```

Arguments

name	Short design name.
subset	Subset specification. Accepts a one-sided formula, a function(data), a logical vector of length nrow(data), or a character string naming a logical column in data.
rationale	Non-empty substantive justification for the subset.
variables_used	Optional character vector describing the design variables used to define the subset. This is required for function-based subsets unless the package can safely infer the variables.
subset_type	Optional short label such as "theory_defined" or "design_based".
pre_specified	Logical indicator for confirmatory use. Confirmatory BPE requires TRUE.
transportability_rationale	Optional description of why the direct effect learned in the subset may be informative for the target sample.
notes	Optional free-form notes.

Value

An object of class "spliv_bpe_design".

Examples

```
design <- bpe_design(
  "Inactive subset", ~ inactive,
  rationale = "The treatment channel is absent in this subset.",
  variables_used = "inactive", pre_specified = TRUE
)
bpe_eval_subset(design, data.frame(inactive = c(TRUE, FALSE)))
```

bpe_eval_subset	<i>Evaluate a Confirmatory BPE Subset</i>
-----------------	---

Description

Evaluates a bpe_design() object on a data frame and returns the resulting logical indicator.

Usage

```
bpe_eval_subset(design, data, max_na_share = 0.05)
```

Arguments

design	A bpe_design() object.
data	Data frame used for evaluation.
max_na_share	Maximum allowable share of NA values in the subset indicator before evaluation fails. The default is 0.05.

Value

A logical vector of length `nrow(data)`.

Examples

```
design <- bpe_design("Inactive", ~ inactive,
  rationale = "The treatment channel is absent.")
bpe_eval_subset(design, data.frame(inactive = c(TRUE, FALSE)))
```

bpe_explore_subsets	<i>Explore Candidate BPE Subsets</i>
---------------------	--------------------------------------

Description

Explores user-supplied candidate subset definitions and reports diagnostics for transparent exploratory work. This function is not valid confirmatory BPE by itself. Confirmatory BPE requires a pre-specified `bpe_design()` object supplied to `spliv()` or `bpe_validate_design()`.

Usage

```
bpe_explore_subsets(
  data,
  spec,
  rules = NULL,
  seed = 1,
  min_n_S = NULL,
  max_F_S = NULL,
  min_varZ_S = NULL,
  return_all = TRUE
)
```

Arguments

<code>data</code>	Data frame used for estimation.
<code>spec</code>	Either an IV formula ($y \sim X \mid Z$) or a list with at least <code>formula</code> , and optional <code>fe</code> , <code>fe_engine</code> , and <code>z_names</code> .
<code>rules</code>	Candidate subset definitions. Each candidate may be a one-sided formula evaluated in <code>data</code> , a <code>function(data)</code> returning a logical vector, a character string naming a logical column in <code>data</code> , a logical vector of length <code>nrow(data)</code> , or a named list with components <code>name</code> and <code>subset</code> .
<code>seed</code>	Integer seed for deterministic rule evaluation.
<code>min_n_S</code>	Optional exploratory screen for subset size.
<code>max_F_S</code>	Deprecated exploratory screen for the first-stage F-statistic. The first-stage F-statistic is reported as a diagnostic only.
<code>min_varZ_S</code>	Optional exploratory screen for within-subset residualized instrument variance.

`return_all` If TRUE (default), returns all candidates and diagnostics. If FALSE, returns the first candidate in the user-supplied order. No automatic subset search or confirmatory selection is performed.

Details

Warning: this function is exploratory. It should not be used for confirmatory BPE inference. Confirmatory BPE requires a pre-specified `bpe_design()` object with a substantive rationale.

Value

If `return_all = TRUE`, a data frame with one row per rule and a list column subset. If `return_all = FALSE`, a list with the first supplied subset, rule, and diagnostics.

Examples

```
d <- data.frame(y = rnorm(40), x = rnorm(40), z = rnorm(40),
  inactive = rep(c(TRUE, FALSE), each = 20))
suppressWarnings(bpe_explore_subsets(
  d, y ~ x | z, rules = list(inactive = ~ inactive)))
```

bpe_find_subset

Deprecated Exploratory BPE Subset Search

Description

`bpe_find_subset()` is retained for backward compatibility only. Use `bpe_explore_subsets()` for exploratory work and convert any theory-justified subset into a confirmatory `bpe_design()` object before calling `spliv()`.

Usage

```
bpe_find_subset(
  data,
  spec,
  rules = NULL,
  seed = 1,
  min_n_S = NULL,
  max_F_S = NULL,
  min_varZ_S = NULL,
  return_all = TRUE
)
```

Arguments

<code>data</code>	Data frame used for estimation.
<code>spec</code>	Either an IV formula ($y \sim X \mid Z$) or a list with at least <code>formula</code> , and optional <code>fe</code> , <code>fe_engine</code> , and <code>z_names</code> .
<code>rules</code>	Candidate subset definitions. Each candidate may be a one-sided formula evaluated in <code>data</code> , a <code>function(data)</code> returning a logical vector, a character string naming a logical column in <code>data</code> , a logical vector of length <code>nrow(data)</code> , or a named list with components <code>name</code> and <code>subset</code> .
<code>seed</code>	Integer seed for deterministic rule evaluation.
<code>min_n_S</code>	Optional exploratory screen for subset size.
<code>max_F_S</code>	Deprecated exploratory screen for the first-stage F-statistic. The first-stage F-statistic is reported as a diagnostic only.
<code>min_varZ_S</code>	Optional exploratory screen for within-subset residualized instrument variance.
<code>return_all</code>	If TRUE (default), returns all candidates and diagnostics. If FALSE, returns the first candidate in the user-supplied order. No automatic subset search or confirmatory selection is performed.

Value

See `bpe_explore_subsets()`.

Examples

```
d <- data.frame(y = rnorm(40), x = rnorm(40), z = rnorm(40),
  inactive = rep(c(TRUE, FALSE), each = 20))
suppressWarnings(bpe_find_subset(
  d, y ~ x | z, rules = list(inactive = ~ inactive)))
```

<code>bpe_validate_design</code>	<i>Validate a Confirmatory BPE Design</i>
----------------------------------	---

Description

Runs the confirmatory BPE eligibility diagnostics for a pre-specified design without fitting the final SPLIV model.

Usage

```
bpe_validate_design(
  formula,
  data,
  design,
  fe = NULL,
  fe_engine = c("fixest", "lfe"),
  vcov = c("iid", "hc1", "cluster"),
```

```

cluster = NULL,
z_names = NULL,
bpe_min_n_S = 2000,
bpe_min_clusters_S = 30,
bpe_min_varZ_S = 1e-06,
bpe_equiv_margin,
bpe_equiv_level = 0.95,
bpe_transport = c("none", "sampling", "conservative"),
bpe_transport_kappa = 0,
bpe_kappa = 1,
scale_instrument = c("residual_sd", "none")
)

```

Arguments

formula	IV formula $y \sim X \mid Z$.
data	Data frame.
design	A <code>bpe_design()</code> object.
fe	Optional one-sided formula of fixed effects.
fe_engine	FE demeaning engine, one of "fixest" or "lfe".
vcov	One of "iid", "hc1", or "cluster".
cluster	Cluster ids or one-sided formula when <code>vcov = "cluster"</code> .
z_names	Optional instrument name for BPE. Confirmatory BPE currently supports exactly one instrument.
bpe_min_n_S	Minimum subset size threshold.
bpe_min_clusters_S	Minimum number of clusters required in the subset when clustered covariance is used.
bpe_min_varZ_S	Minimum residualized instrument variance required in the subset.
bpe_equiv_margin	Researcher-specified equivalence margin for the first-stage coefficient. Eligibility is based on the first-stage equivalence interval, not on the first-stage F-statistic.
bpe_equiv_level	Confidence level used for the first-stage equivalence interval.
bpe_transport	One of "none", "sampling", or "conservative". Transportability is an assumption reflected in the reported covariance; it is not established by the subset itself.
bpe_transport_kappa	Non-negative scalar controlling the conservative transport covariance inflation.
bpe_kappa	Positive scalar multiplier applied to the transported BPE covariance before it is embedded into the LTZ prior.
scale_instrument	One of "residual_sd" or "none".

Value

An object of class "spliv_bpe_validation" containing design metadata, subset diagnostics, first-stage equivalence diagnostics, reduced- form direct-effect estimates, and covariance components for confirmatory BPE.

Examples

```
set.seed(2)
d <- data.frame(
  y = rnorm(80), x = rnorm(80), z = rnorm(80),
  inactive = rep(c(TRUE, FALSE), each = 40)
)
design <- bpe_design("Inactive", ~ inactive,
  rationale = "The treatment channel is absent.")
bpe_validate_design(y ~ x | z, d, design,
  bpe_min_n_S = 20, bpe_equiv_margin = 1)
```

demean_fixest

Demean with fixest

Description

Demeans vectors/matrices by high-dimensional fixed effects using `fixest::demean`.

Usage

```
demean_fixest(y, X, Z, W = NULL, fe_fml, data)
```

Arguments

<code>y</code>	Outcome vector.
<code>X</code>	Regressor matrix.
<code>Z</code>	Instrument matrix.
<code>W</code>	Optional controls matrix.
<code>fe_fml</code>	One-sided FE formula.
<code>data</code>	Data frame aligned with the rows of <code>y</code> , <code>X</code> , <code>Z</code> , <code>W</code> .

Value

List with demeaned `y`, `X`, `Z`, `W`, and FE frame.

Examples

```
d <- data.frame(g = factor(rep(1:2, each = 3)))
y <- 1:6; X <- matrix(rnorm(6), 6, 1); Z <- matrix(rnorm(6), 6, 1)
demean_fixest(y, X, Z, fe_fml = ~ g, data = d)
```


demean_lfe

*Demean with lfe***Description**

Demeans vectors/matrices by high-dimensional fixed effects using `lfe::demeanlist`.

Usage

```
demean_lfe(y, X, Z, W = NULL, fe_list)
```

Arguments

<code>y</code>	Outcome vector.
<code>X</code>	Regressor matrix.
<code>Z</code>	Instrument matrix.
<code>W</code>	Optional controls matrix.
<code>fe_list</code>	List/data.frame of FE ids.

Value

List with demeaned `y`, `X`, `Z`, `W`.

Examples

```
d <- data.frame(g = factor(rep(1:2, each = 3)))
y <- 1:6; X <- matrix(rnorm(6), 6, 1); Z <- matrix(rnorm(6), 6, 1)
demean_lfe(y, X, Z, fe_list = d["g"])
```

embed_prior_into_full_Z

*Embed Prior into Full Instrument Space***Description**

Expands a prior estimated on a subset of instruments into the full instrument vector implied by $y \sim X | Z$.

Usage

```
embed_prior_into_full_Z(formula, data, z_names, mu_hat, omega_hat)
```

Arguments

formula	IV formula $y \sim X \mid Z$.
data	Data frame.
z_names	Instrument names with prior moments.
mu_hat	Prior means for z_names.
omega_hat	Prior covariance for z_names.

Value

List with full-length mu, Omega, and instrument names.

Examples

```
d <- data.frame(y = rnorm(30), x = rnorm(30), z1 = rnorm(30), z2 = rnorm(30))
embed_prior_into_full_Z(y ~ x | z1 + z2, d,
  z_names = "z1", mu_hat = 0, omega_hat = matrix(0.1, 1, 1))
```

estimate_gamma_zero_first_stage

Legacy Exploratory BPE Prior Helper

Description

estimate_gamma_zero_first_stage() is retained as a legacy exploratory helper. It is not sufficient for confirmatory BPE inference. Confirmatory BPE requires bpe_design(), bpe_validate_design(), and spliv(method = "bpe", ...).

Usage

```
estimate_gamma_zero_first_stage(
  data,
  y_name,
  z_names,
  controls = character(0),
  subset,
  fe = NULL
)
```

Arguments

data	Data frame.
y_name	Outcome variable name.
z_names	Instrument names whose direct effects are estimated.
controls	Optional character vector of additional controls.

subset	Legacy subset specification. Accepts a <code>bpe_design()</code> object, a one-sided formula, a <code>function(data)</code> , a logical vector, or a character string naming a logical column in data.
fe	Optional one-sided FE formula. If supplied, uses <code>fixest::feols</code> .

Value

List with `mu_hat`, `omega_hat`, and diagnostics.

Examples

```
d <- data.frame(y = rnorm(40), z = rnorm(40), inactive = rep(c(TRUE, FALSE), each = 20))
suppressWarnings(estimate_gamma_zero_first_stage(
  d, y_name = "y", z_names = "z", subset = ~ inactive))
```

iv_inst_names	<i>Instrument Names for Parsed IV Formula</i>
---------------	---

Description

Returns the instrument-matrix column order used internally.

Usage

```
iv_inst_names(formula, data)
```

Arguments

formula	IV formula $y \sim X \mid Z$.
data	Data frame.

Value

Character vector of instrument names.

Examples

```
d <- data.frame(y = 1:4, x = 1:4, z = 4:1)
iv_inst_names(y ~ x | z, d)
```

plot_sp_sensitivity *Plot Patterned Sensitivity Output or a Fitted Object*

Description

Plot Patterned Sensitivity Output or a Fitted Object

Usage

```
plot_sp_sensitivity(
  df_plot,
  ylab = "Effect (beta)",
  main = "Plausibly Exogenous IV Sensitivity",
  ...
)

plot_conley_sensitivity(
  df_plot,
  ylab = "Effect (beta)",
  main = "Plausibly Exogenous IV Sensitivity",
  ...
)
```

Arguments

df_plot	Data frame returned by a sensitivity helper or plausexog_fit object.
ylab	Y-axis label.
main	Plot title.
...	Unused.

Value

Invisibly returns the plotted input.

Examples

```
set.seed(7)
d <- data.frame(y = rnorm(80), x = rnorm(80), z = rnorm(80))
p <- spliv_sensitivity_path(y ~ x | z, d, method = "uci", delta_grid = c(0, 0.1))
plot_sp_sensitivity(p, term = "x")
```

Description

Main estimator for patterned sensitivity analysis of exclusion violations in fixed-effect or residualized IV designs.

Usage

```
plausexog(
  formula,
  data,
  fe = NULL,
  fe_engine = c("fixest", "lfe"),
  vcov = c("iid", "hc1", "cluster"),
  cluster = NULL,
  method = c("ltz", "uci", "bpe"),
  prior = NULL,
  delta = NULL,
  violation_pattern = NULL,
  bpe = FALSE,
  bpe_design = NULL,
  bpe_spec = list(design = NULL, subset = NULL, z_names = NULL),
  bpe_kappa = 1,
  bpe_omega = NULL,
  bpe_min_n_S = 2000,
  bpe_min_clusters_S = 30,
  bpe_max_F_S = NULL,
  bpe_min_varZ_S = 1e-06,
  bpe_equiv_margin = NULL,
  bpe_equiv_level = 0.95,
  bpe_transport = c("sampling", "none", "conservative"),
  bpe_transport_kappa = 0,
  bpe_not_applicable = c("na", "error"),
  scale_instrument = c("residual_sd", "none"),
  grid = list(),
  ...
)

plausexog_iv(...)

spliv(
  formula,
  data,
  fe = NULL,
  fe_engine = c("fixest", "lfe"),
```

```

vcov = c("iid", "hc1", "cluster"),
cluster = NULL,
method = c("ltz", "uci", "bpe"),
prior = NULL,
delta = NULL,
violation_pattern = NULL,
bpe = FALSE,
bpe_design = NULL,
bpe_spec = list(design = NULL, subset = NULL, z_names = NULL),
bpe_kappa = 1,
bpe_omega = NULL,
bpe_min_n_S = 2000,
bpe_min_clusters_S = 30,
bpe_max_F_S = NULL,
bpe_min_varZ_S = 1e-06,
bpe_equiv_margin = NULL,
bpe_equiv_level = 0.95,
bpe_transport = c("sampling", "none", "conservative"),
bpe_transport_kappa = 0,
bpe_not_applicable = c("na", "error"),
scale_instrument = c("residual_sd", "none"),
grid = list(),
...
)

```

Arguments

formula	IV formula $y \sim X \mid Z$.
data	Data frame.
fe	Optional one-sided formula of fixed effects.
fe_engine	FE demeaning engine, one of "fixest" or "lfe".
vcov	One of "iid", "hc1", or "cluster".
cluster	Cluster ids or one-sided formula, required when <code>vcov = "cluster"</code> .
method	One of "ltz", "uci", or "bpe".
prior	Optional prior list with μ and Ω (or ω) for LTZ. When <code>violation_pattern</code> is supplied, patterned LTZ currently requires a scalar prior over the pattern coefficient.
delta	Optional non-negative scalar sensitivity magnitude. With <code>method = "uci"</code> , <code>delta</code> implies theta bounds $[-\text{delta}, +\text{delta}]$ unless explicit bounds are supplied in <code>grid</code> . With <code>method = "ltz"</code> and no explicit prior, <code>delta</code> induces a zero-mean normal LTZ prior.
violation_pattern	Optional <code>spliv_pattern()</code> object describing how the direct effect of the instrument may vary across observations. If omitted, LTZ/UCI retain the package's backward-compatible uniform direct-effect behavior. This argument is currently supported for LTZ and UCI, but not for confirmatory BPE.

bpe	Logical; when TRUE, learn prior moments from a confirmatory bpe_design() and run LTZ.
bpe_design	Optional bpe_design() object for confirmatory BPE.
bpe_spec	Optional list. Prefer bpe_spec = list(design = my_design). For backward compatibility, bpe_spec\$subset may also be supplied, but it must represent an explicit researcher-supplied subset and should be paired with a non-empty rationale, explicit pre_specified = TRUE, and any relevant metadata such as variables_used or subset_type. Supply exactly one confirmatory subset source: bpe_design, bpe_spec\$design, or bpe_spec\$subset. Exploratory subset_rule inputs are not accepted for confirmatory estimation.
bpe_kappa	Positive scalar multiplier applied to the confirmatory BPE covariance after transport adjustment.
bpe_omega	Deprecated. Confirmatory BPE now uses the full reduced-form covariance from the subset together with bpe_transport.
bpe_min_n_S	Minimum subset size required for BPE eligibility. Default 2000.
bpe_min_clusters_S	Minimum number of clusters required in subset S when vcov = "cluster". Default 30.
bpe_max_F_S	Deprecated. The first-stage F-statistic is reported for diagnostics only and no longer determines confirmatory BPE eligibility.
bpe_min_varZ_S	Minimum residualized instrument variance required in subset S. Default 1e-6.
bpe_equiv_margin	Researcher-specified first-stage equivalence margin. Confirmatory BPE currently supports exactly one instrument.
bpe_equiv_level	Confidence level for the first-stage equivalence check.
bpe_transport	One of "none", "sampling", or "conservative".
bpe_transport_kappa	Non-negative scalar controlling the conservative transport covariance inflation.
bpe_not_applicable	Behavior when subset diagnostics fail. One of "na" (default) to return NA estimates, or "error" to stop.
scale_instrument	One of "residual_sd" (default) or "none".
grid	List controlling UCI bounds or other tuning parameters. For backward-compatible scalar UCI, if grid\$delta is supplied and grid\$gmin/grid\$gmax are omitted, the package interprets delta as a direct-effect bound of [-delta, +delta] under the chosen scale_instrument. When violation_pattern is supplied, grid\$delta instead refers to theta bounds over the pattern-scaled direct effect.
...	Reserved.

Details

spliv implements patterned sensitivity analysis for exclusion violations in IV designs with fixed effects or other residualization steps. Researchers can supply a theoretically motivated spliv_pattern()

object to specify where direct effects of an instrument are expected to be larger or smaller, and the package then scales LTZ/UCI sensitivity along that pattern.

The package does not estimate an unrestricted direct-effect field. Instead, researchers supply a structured pattern and ask whether conclusions survive direct effects scaled along that pattern.

In applied work, users should usually vary delta over a range with `spliv_sensitivity_path()` rather than report one arbitrary sensitivity value.

Patterned sensitivity currently supports one endogenous treatment, one excluded instrument, and one researcher-specified pattern at a time.

Confirmatory BPE is not a subgroup-search procedure. The researcher must supply a pre-specified `bpe_design()` object, the package validates that subset, and BPE proceeds only if the confirmatory eligibility checks pass.

The first-stage F-statistic is still reported for diagnostics, but confirmatory BPE eligibility is determined by the pre-specification checks, subset size, cluster count, residualized instrument variation, and a first-stage equivalence interval.

BPE reduced-form covariance is propagated as a full covariance matrix and can optionally be inflated via `bpe_transport`.

Value

Object of class `plausexog_fit`.

Deprecated Wrappers

`plausexog()` and `plausexog_iv()` are deprecated compatibility aliases for `spliv()`.

Examples

```
set.seed(1)
d <- data.frame(y = rnorm(60), x = rnorm(60), z = rnorm(60), w = rnorm(60))
fit <- spliv(y ~ x + w | z + w, d, method = "uci", delta = 0.1)
fit$estimates
```

<code>spliv_eval_pattern</code>	<i>Evaluate a Direct-Effect Pattern</i>
---------------------------------	---

Description

Evaluates a `spliv_pattern()` object on a data frame and returns a numeric vector after optional centering and normalization.

Usage

```
spliv_eval_pattern(pattern, data)
```


Arguments

pattern	A spliv_pattern() object.
data	Data frame used for evaluation.

Value

Numeric vector of length nrow(data).

Examples

```
d <- data.frame(exposure = seq(-1, 1, length.out = 5))
p <- spliv_pattern("Exposure", ~ exposure,
  rationale = "Illustration of a monotone exposure pattern.")
spliv_eval_pattern(p, d)
```

spliv_pattern	<i>Create a Patterned Exclusion-Violation Object</i>
---------------	--

Description

Creates a researcher-specified direct-effect pattern for patterned sensitivity analysis. The pattern determines where direct effects of the instrument are expected to be larger or smaller in the estimation sample.

Usage

```
spliv_pattern(
  name,
  pattern,
  rationale,
  variables_used = NULL,
  pattern_type = NULL,
  normalize = c("max_abs", "sd", "none"),
  center = FALSE,
  pre_specified = TRUE,
  notes = NULL
)
```

Arguments

name	Short pattern name.
pattern	Pattern specification. Accepts a one-sided formula, a function(data), a numeric vector of length nrow(data), a logical vector of length nrow(data), or a character string naming a numeric or logical column in data.
rationale	Substantive justification for the proposed pattern.
variables_used	Optional character vector describing the variables used to define the pattern.

pattern_type	Optional short label such as "theory_defined" or "design_based".
normalize	One of "max_abs" (default), "sd", or "none".
center	Logical; if TRUE, subtracts the estimation-sample mean before normalization.
pre_specified	Logical indicator for whether the pattern was chosen before examining outcome results.
notes	Optional free-form notes.

Value

An object of class "spliv_pattern".

Examples

```
p <- spliv_pattern(
  name = "Exposure", pattern = ~ exposure,
  rationale = "The alternative channel follows exposure.",
  variables_used = "exposure", pattern_type = "theory_defined"
)
p$name
```

spliv_sensitivity_path

Sensitivity Path over Delta Grid

Description

Runs `spliv()` repeatedly over a user-supplied `delta_grid` and returns a tidy sensitivity-path object. This is the recommended workflow for patterned or uniform sensitivity analysis: users should usually report how intervals change over a range of `delta` values rather than selecting one arbitrary sensitivity level.

Usage

```
spliv_sensitivity_path(
  formula,
  data,
  method = c("uci", "ltz"),
  delta_grid = seq(0, 0.2, by = 0.01),
  violation_pattern = NULL,
  stop_on_error = TRUE,
  ...
)
```

Arguments

formula	IV formula $y \sim X \mid Z$.
data	Data frame.
method	One of "uci" or "ltz". <code>spliv_sensitivity_path()</code> intentionally excludes confirmatory BPE.
delta_grid	Non-negative sensitivity grid. For UCI, each value d implies theta bounds $[-d, +d]$.
violation_pattern	Optional <code>spliv_pattern()</code> object. If omitted, the path uses the package's backward-compatible uniform direct-effect pattern.
stop_on_error	Logical; if TRUE (default), stop on the first failed fit. If FALSE, record NA rows and store the error message in the returned error column.
...	Additional named arguments passed through to <code>spliv()</code> , such as <code>fe</code> , <code>vcov</code> , <code>cluster</code> , <code>scale_instrument</code> , or <code>grid = list(level = 0.95)</code> .

Value

A data frame with class `c("spliv_sensitivity_path", "data.frame")`. The returned object includes path columns such as `delta`, `method`, `estimate`, `conf_low`, `conf_high`, `contains_zero`, and pattern metadata, plus attributes containing the original call, the supplied grid, and a tipping-point summary.

Examples

```
set.seed(5)
d <- data.frame(y = rnorm(80), x = rnorm(80), z = rnorm(80))
p <- spliv_sensitivity_path(y ~ x | z, d, method = "uci",
  delta_grid = c(0, 0.1), vcov = "hcl")
head(p)
```

<code>spliv_tipping_point</code>	<i>Extract Tipping-Point Delta from a Sensitivity Path</i>
----------------------------------	--

Description

Returns the tipping-point attribute stored on a `spliv_sensitivity_path()` object. For each term, the tipping point is the smallest `delta` on the supplied grid whose interval includes zero.

Usage

```
spliv_tipping_point(x)
```

Arguments

<code>x</code>	A <code>spliv_sensitivity_path</code> object.
----------------	---

Value

Named numeric vector of tipping-point values.

Examples

```
set.seed(6)
d <- data.frame(y = rnorm(80), x = rnorm(80), z = rnorm(80))
p <- spliv_sensitivity_path(y ~ x | z, d, method = "uci", delta_grid = c(0, 0.1))
spliv_tipping_point(p)
```

sp_ltz

Local-to-Zero Inference for Plausibly Exogenous IV

Description

Formula-level wrapper over the matrix core LTZ implementation.

Usage

```
sp_ltz(
  formula,
  data,
  omega,
  mu,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL
)

conley_ltz(
  formula,
  data,
  omega,
  mu,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL
)
```

Arguments

formula	IV formula $y \sim X \mid Z$.
data	Data frame.
omega	Prior covariance matrix over instrument direct effects.
mu	Prior mean vector over instrument direct effects.
level	Confidence level.
vcov	One of "iid", "hc0", "hc1", "cluster".
cluster	Cluster ids when vcov = "cluster".

Value

Data frame with adjusted estimates and confidence intervals.

Examples

```
set.seed(3)
d <- data.frame(y = rnorm(80), x = rnorm(80), z = rnorm(80))
prior <- conley_prior_ltz(y ~ x | z, d, inst_vary = "z", sd = 0.1)
sp_ltz(y ~ x | z, d, prior$omega, prior$mu)
```

sp_prior_ltz

Build LTZ Prior Matrices for Chosen Instruments

Description

Build LTZ Prior Matrices for Chosen Instruments

Usage

```
sp_prior_ltz(
  formula,
  data = NULL,
  inst_vary,
  mean = 0,
  sd = 1,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL
)

conley_prior_ltz(
  formula,
  data = NULL,
  inst_vary,
  mean = 0,
  sd = 1,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL
)
```

Arguments

formula	Formula or <code>plausexog_fit</code> .
data	Data frame when formula is a formula.
inst_vary	Instrument(s) with non-degenerate prior.
mean	Prior mean(s).
sd	Prior sd(s).
vcov	Vcov type.
cluster	Optional cluster ids.

Value

List with mu, omega, and instrument names.

Examples

```
set.seed(9)
d <- data.frame(y = rnorm(60), x = rnorm(60), z = rnorm(60))
sp_prior_ltz(y ~ x | z, d, inst_vary = "z", sd = 0.1)
```

sp_sensitivity_ltz_normal

LTZ Sensitivity over Delta Grid

Description

LTZ Sensitivity over Delta Grid

Usage

```
sp_sensitivity_ltz_normal(
  formula,
  data = NULL,
  term,
  inst_vary,
  delta_grid,
  mean_fun = function(delta) 0,
  sd_fun = function(delta) delta,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL,
  scale_instrument = c("residual_sd", "none")
)

conley_sensitivity_ltz_normal(
  formula,
  data = NULL,
  term,
  inst_vary,
  delta_grid,
  mean_fun = function(delta) 0,
  sd_fun = function(delta) delta,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL,
  scale_instrument = c("residual_sd", "none")
)
```

Arguments

formula	Formula or <code>plausexog_fit</code> .
data	Data frame when formula is a formula.
term	Coefficient name to track.
inst_vary	Instrument(s) with plausible violation.
delta_grid	Delta grid.
mean_fun	Function mapping delta to prior mean.
sd_fun	Function mapping delta to prior sd.
level	Confidence level.
vcov	Vcov type.
cluster	Optional cluster ids.
scale_instrument	One of "residual_sd" (default) or "none".

Value

Data frame with sensitivity path.

sp_sensitivity_ltz_uniform01_as_normal

LTZ Sensitivity with Normal Approximation to $U(0, \delta)$

Description

LTZ Sensitivity with Normal Approximation to $U(0, \delta)$

Usage

```
sp_sensitivity_ltz_uniform01_as_normal(
  formula,
  data = NULL,
  term,
  inst_vary,
  delta_grid,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL,
  scale_instrument = c("residual_sd", "none")
)

conley_sensitivity_ltz_uniform01_as_normal(
  formula,
  data = NULL,
```

```

    term,
    inst_vary,
    delta_grid,
    level = 0.95,
    vcov = c("hc1", "hc0", "iid", "cluster"),
    cluster = NULL,
    scale_instrument = c("residual_sd", "none")
  )

```

Arguments

formula	Formula or <code>plausexog_fit</code> .
data	Data frame when formula is a formula.
term	Coefficient name to track.
inst_vary	Instrument(s) with plausible violation.
delta_grid	Delta grid.
level	Confidence level.
vcov	Vcov type.
cluster	Optional cluster ids.
scale_instrument	One of "residual_sd" (default) or "none".

Value

Data frame with sensitivity path.

sp_sensitivity_uci_support

UCI Sensitivity over Delta Grid

Description

UCI Sensitivity over Delta Grid

Usage

```

sp_sensitivity_uci_support(
  formula,
  data = NULL,
  term,
  inst_vary,
  delta_grid,
  gmin_fun = function(delta) -delta,
  gmax_fun = function(delta) delta,
  grid = 41,

```



```

    level = 0.95,
    vcov = c("hc1", "hc0", "iid", "cluster"),
    cluster = NULL,
    scale_instrument = c("residual_sd", "none")
)

conley_sensitivity_uci_support(
  formula,
  data = NULL,
  term,
  inst_vary,
  delta_grid,
  gmin_fun = function(delta) -delta,
  gmax_fun = function(delta) delta,
  grid = 41,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL,
  scale_instrument = c("residual_sd", "none")
)

```

Arguments

formula	Formula or <code>plausexog_fit</code> .
data	Data frame when formula is a formula.
term	Coefficient name to track.
inst_vary	Instrument(s) whose gamma bounds vary.
delta_grid	Delta grid.
gmin_fun	Function mapping delta to lower bound.
gmax_fun	Function mapping delta to upper bound.
grid	Points per instrument bound.
level	Confidence level.
vcov	Vcov type.
cluster	Optional cluster ids.
scale_instrument	One of "residual_sd" (default) or "none".

Value

Data frame with sensitivity path.

sp_uci

*Union of Confidence Intervals for Plausibly Exogenous IV***Description**

Formula-level wrapper over the matrix core UCI implementation.

Usage

```
sp_uci(
  formula,
  data,
  inst,
  gmin,
  gmax,
  grid = 21,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL
)
```

```
conley_uci(
  formula,
  data,
  inst,
  gmin,
  gmax,
  grid = 21,
  level = 0.95,
  vcov = c("hc1", "hc0", "iid", "cluster"),
  cluster = NULL
)
```

Arguments

formula	IV formula $y \sim X Z$.
data	Data frame.
inst	Instrument names to vary.
gmin	Lower bound(s) on gamma.
gmax	Upper bound(s) on gamma.
grid	Grid size per varied instrument.
level	Confidence level.
vcov	One of "iid", "hc0", "hc1", "cluster".
cluster	Cluster ids when vcov = "cluster".

Value

Data frame with union confidence intervals.

Examples

```
set.seed(4)
d <- data.frame(y = rnorm(80), x = rnorm(80), z = rnorm(80))
sp_uci(y ~ x | z, d, inst = "z", gmin = -0.1, gmax = 0.1, grid = 5)
```

Index

bpe_design, [2](#)
bpe_eval_subset, [3](#)
bpe_explore_subsets, [4](#)
bpe_find_subset, [5](#)
bpe_validate_design, [6](#)

conley_ltz (sp_ltz), [20](#)
conley_prior_ltz (sp_prior_ltz), [21](#)
conley_sensitivity_ltz_normal
 (sp_sensitivity_ltz_normal), [22](#)
conley_sensitivity_ltz_uniform01_as_normal
 (sp_sensitivity_ltz_uniform01_as_normal),
 [23](#)
conley_sensitivity_uci_support
 (sp_sensitivity_uci_support),
 [24](#)
conley_uci (sp_uci), [26](#)

demean_fixest, [8](#)
demean_lfe, [9](#)

embed_prior_into_full_Z, [9](#)
estimate_gamma_zero_first_stage, [10](#)

iv_inst_names, [11](#)

plausexog (spliv), [13](#)
plausexog_iv (spliv), [13](#)
plot_conley_sensitivity
 (plot_sp_sensitivity), [12](#)
plot_sp_sensitivity, [12](#)

sp_ltz, [20](#)
sp_prior_ltz, [21](#)
sp_sensitivity_ltz_normal, [22](#)
sp_sensitivity_ltz_uniform01_as_normal,
 [23](#)
sp_sensitivity_uci_support, [24](#)
sp_uci, [26](#)
spliv, [13](#)
spliv_eval_pattern, [16](#)
spliv_pattern, [17](#)
spliv_sensitivity_path, [18](#)
spliv_tipping_point, [19](#)