# Package 'cohetsurr'

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Type Package

Title Assessing Complex Heterogeneity in Surrogacy

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<b>Description</b> Provides functions to assess and test for complex heterogeneity in the strength of a surrogate marker with respect to multiple baseline covariates, using both a parametric model and a semiparametric two-step model. More details are available in: Knowlton, R., Tian, L., & Parast, L. (2025). ``A General Framework to Assess Complex Heterogeneity in the Strength of a Surrogate Marker," Statistics in Medicine, 44(5), e70001 <doi:10.1002 sim.70001="">. Additionally, provides functions to assess complex heterogeneity in the strength of a surrogate marker in observational settings using meta-learners, with options for different base learners. More details will be available in the future in: Knowlton, R., Parast, L. (2025) ``Assessing Surrogate Heterogeneity in Real World Data Using Meta-Learners." A tutorial for this package can be found at <a href="https://www.laylaparast.com/cohetsurr">https://www.laylaparast.com/cohetsurr</a>.</doi:10.1002>
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boot.var

boot.var	Performs bootstrap estimation procedures for the variance of the pro- portion of treatment effect explained, the omnibus test, and identifying a region above a treshold
	a region above a treshold

# Description

Performs bootstrap estimation procedures for the variance of the proportion of treatment effect explained, the omnibus test, and identifying a region above a treshold

# Usage

```
boot.var(data.control, data.treat, W.grid.expand, type, test = FALSE,
data.all = NULL, num.cov = NULL, results.for.test = NULL, threshold = NULL)
```

# **Arguments**

data.control	dataframe containing data from the control group, specifically the outcome Y, the surrogate S, and the baseline covariates W	
data.treat	dataframe containing data from the treamtent group, specifically the outcome Y, the surrogate S, and the baseline covariates W	
W.grid.expand	expanded version of the W grid of baseline covariates, where each row is a specific combination of the covariates for which the estimates should be provided	
type	options are "model", "two step", or "both"; specifies the estimation method that should be used for the proportion of treatment effect explained	
test	TRUE or FALSE, if test for heterogeneity is wanted	
data.all	dataframe containing data from the control and treamtent groups, specifically the outcome Y, the surrogate S, and the baseline covariates W	
num.cov	number of baseline covariates in the matrix W	
results.for.test		
	the grid of estimates for the proportion of treatment effect explained obtained prior to bootstrapping, needed for the omnibus test	
threshold	threshold to flag regions where the estimated proportion of the treatment effect explained is at least that high	

# Value

# A list is returned:

return.grid	grid of variance estimates for the overall treatment effect, the residual treatment effect, and the proportion of treatment effect explained as a function of the baseline covariates, W. If requested by user, includes regions flagged above the threshold.
pval	p-value(s) from the F test and the two step omnibus test for heterogeneity, depending on type argument

complex.heterogeneity Estimates the proportion of treatment effect explained by the surrogate marker as a function of multiple baseline covariates.

# Description

Assesses complex heterogeneity in the utility of a surrogate marker by estimating the proportion of treatment effect explained by the surrogate marker as a function of multiple baseline covariates. Optionally, tests for evidence of heterogeneity overall and flags regions where the proportion of treatment effect explained is above a given threshold.

# Usage

```
complex.heterogeneity(y, s, a, W.mat, type = "model", variance = FALSE,
test = FALSE, W.grid = NULL, grid.size = 4, threshold = NULL)
```

## **Arguments**

у	y, the outcome
S	s, the surrogate marker
a	a, the treatment assignment with 1 indicating the treatment group and 0 indicating the control group
W.mat	matrix of baseline covariate observations, where the first column is $W1$ , second columns is $W2$ , and so on.
type	options are "model", "two step", or "both"; specifies the estimation method that should be used for the proportion of treatment effect explained
variance	TRUE or FALSE, if variance/standard error estimates are wanted
test	TRUE or FALSE, if test for heterogeneity is wanted
W.grid	grid for the baseline covariates W where estimation will be provided
grid.size	number of measures for each baseline covariate to include in the estimation grid, if one is not provided by the user directly
threshold	threshold to flag regions where the estimated proportion of the treatment effect explained is at least that high

#### Value

#### A list is returned:

return.grid grid of estimates for the overall treatment effect, the residual treatment effect, and the proportion of treatment effect explained as a function of the baseline covariates, W. Includes variance estimates and regions flagged above the threshold, if specified by the user.

pval p-value(s) from the F test and the two step omnibus test for heterogeneity, de-

pending on type argument

# Author(s)

Rebecca Knowlton

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#### References

Knowlton, R., Tian, L., & Parast, L. (2025). A General Framework to Assess Complex Heterogeneity in the Strength of a Surrogate Marker. Statistics in Medicine, 44(5), e70001

## **Examples**

exampledata

Example data

## **Description**

Example data

## Usage

```
data("exampledata")
```

#### **Format**

A list with 7 elements representing 1000 observations from a treatment group and 1000 observations from a control group, and a grid of baseline covariate values at which to calculate estimates:

- y the outcome
- s the surrogate marker
- a the treatment assignment, where 1 indicates treatment and 0 indicates control
- w1 the first baseline covariate of interest
- w2 the second baseline covariate of interest
- w1.grid the grid of first baseline covariate values to provide estimates for
- w2.grid the grid of second baseline covariate values to provide estimates for

## **Examples**

```
data(exampledata)
names(exampledata)
```

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obs.boot.var Calculate bootstrapped variance estimates in an observational setting.
---

## **Description**

Calculates bootstrapped variance estimates of delta, delta.s, and R.s, and optionally calculates p-values for for identifying individual for whom the surrogate is strong.

#### Usage

obs.boot.var(df.train, df.test, type, numeric\_predictors, categorical\_predictors, threshold, use.actual.control.S, gam.smoothers, tree.tuners)

#### **Arguments**

df.train A dataframe containing training data.
df.test A dataframe containing testing data.

type Options are "linear", "gam", "trees", or "all"; type of base learners to use.

numeric\_predictors

The column names in the dataframes that represent numeric baseline covariates.

categorical\_predictors

The column names in the dataframes that represent categorical baseline covari-

ates.

threshold An optional threshold to test individuals for the null hypothesis that PTE >

threshold.

use.actual.control.S

TRUE or FALSE, if user prefers to use the actual observed values for the surro-

gate in the control group instead of predicting values from the base learners.

 ${\tt gam.\,smoothers} \quad A \ list \ of \ smoothing \ parameters \ to \ use \ for \ GAM \ base \ learners, \ so \ they \ are \ not$ 

retuned with bootstrapping iterations ("m1sp", "m0sp", "m1ssp", "m0ssp", "s0")

tree.tuners A list of tuning parameters to use for tree base learners, so they are not retuned

with bootstrapping iterations ("m1sp", "m0sp", "m1ssp", "m0ssp", "s0")

# Value

A dataframe is returned, which is the df.test argument with new columns appended for the estimated variances of delta, delta.s, and R.s, as well as p-values if a threshold is provided.

obs.estimate.PTE	Estimate the proportion of the treatment effect explained in an observational setting.

#### **Description**

Fits base learners using the specified type of model on the training data, and uses those models to calculate delta, delta.s, and R.s on the testing set.

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#### Usage

obs.estimate.PTE(df.train, df.test, type, numeric\_predictors, categorical\_predictors, use.actual.control.S, gam.smoothers, tree.tuners, want.smooth, want.tune)

#### **Arguments**

df.train A dataframe containing training data.
df.test A dataframe containing testing data.

type Options are "linear", "gam", "trees", or "all"; type of base learners to use.

numeric\_predictors

The column names in the dataframes that represent numeric baseline covariates.

 ${\tt categorical\_predictors}$ 

The column names in the dataframes that represent categorical baseline covari-

ates.

use.actual.control.S

TRUE or FALSE, if user prefers to use the actual observed values for the surrogate in the control group instead of predicting values from the base learners.

gam. smoothers A list of smoothing parameters to use for GAM base learners, so they are not

retuned with bootstrapping iterations ("m1sp", "m0sp", "m1ssp", "m0ssp", "s0")

tree.tuners A list of tuning parameters to use for tree base learners, so they are not retuned

with bootstrapping iterations ("m1sp", "m0sp", "m1ssp", "m0ssp", "s0")

want. smooth TRUE or FALSE, if smoothing parameters for GAM should be saved

want.tune TRUE or FALSE, if tuning parameters for trees should be saved

#### Value

# A list is returned:

df.test argument with new columns appended for the estimates of delta, delta.s,

and R.s

smooth\_params A list of smoothing parameters used for GAM base learners ("m1sp", "m0sp",

"m1ssp", "m0ssp", "s0")

tuner\_params A list of tuning parameters used for tree base learners ("m1sp", "m0sp", "m1ssp",

"m0ssp", "s0")

obs.het.surr Estimate the proportion of the treatment effect explained by the surro-

gate marker as a function of multiple baseline covariates in an obser-

vational setting

## Description

Assesses surrogate heterogeneity in real world data by estimating the proportion of the treatment effect explained as a function of baseline covariates. Optionally tests individuals for strong surrogacy based on a threshold.

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#### Usage

```
obs.het.surr(df.train, df.test, type, var.want = FALSE, threshold = NULL,
  use.actual.control.S = FALSE)
```

## **Arguments**

df.train	A dataframe containing training data. Must have columns $G$ (treatment assignment), $S$ (surrogate marker), and $Y$ (primary outcome), in addition to the baseline covariates of interest.
df.test	A dataframe containing testing data. Must contain the same baseline covariate columns as the training data.
type	Options are "linear", "gam", "trees", or "all"; type of base learners to use.
var.want	TRUE or FALSE, if variance estimates are wanted.
threshold	An optional threshold to test individuals for the null hypothesis that PTE > threshold; must have var.want = TRUE to return p-values.
use.actual.control.S	

TRUE or FALSE, if user prefers to use the actual observed values for the surrogate in the control group instead of predicting values from the base learners.

#### Value

A dataframe is returned, which is the df.test argument with new columns appended for the estimates and corresponding variances of delta, delta.s, and R.s. If a threshold is specified, returns a p-value for the null hypothesis that PTE > threshold.

#### Author(s)

Rebecca Knowlton

## References

Knowlton, R. and Parast, L. (2025) "Assessing Surrogate Heterogeneity in Real World Data Using Meta-Learners." Under Review.

#### **Examples**

```
data(obs_exampledata_train)
data(obs_exampledata_test)
obs.het.surr(df.train = obs_exampledata_train, df.test = obs_exampledata_test,
type = "linear", var.want = FALSE)
```

## **Description**

Example testing data for observational setting

#### Usage

```
data("obs_exampledata_test")
```

#### **Format**

A data frame with 200 observations on the following 9 variables.

X1 a numeric baseline covariate of interest

X2 a numeric baseline covariate of interest

X3 a numeric baseline covariate of interest

X4 a numeric baseline covariate of interest

X5 a numeric baseline covariate of interest

X6 a numeric baseline covariate of interest

G the treatment assignment, where 1 indicates treated and 0 indicates control

S the surrogate marker

Y the primary outcome

#### **Examples**

```
data(obs_exampledata_test)
names(obs_exampledata_test)
```

## Description

Example training data for observational setting

## Usage

```
data("obs_exampledata_train")
```

#### **Format**

A data frame with 1800 observations on the following 9 variables.

X1 a numeric baseline covariate of interest

X2 a numeric baseline covariate of interest

X3 a numeric baseline covariate of interest

X4 a numeric baseline covariate of interest

X5 a numeric baseline covariate of interest

X6 a numeric baseline covariate of interest

G the treatment assignment, where 1 indicates treated and 0 indicates control

S the surrogate marker

Y the primary outcome

#### **Examples**

```
data(obs_exampledata_train)
names(obs_exampledata_train)
```

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parametric.est	Estimates the proportion of treatment effect explained as a function of multiple baseline covariates, W, using a parametric model

# Description

Estimates the proportion of treatment effect explained as a function of multiple baseline covariates, W, using a parametric model

## Usage

```
parametric.est(data.control, data.treat, W.grid.expand)
```

## **Arguments**

data.control	dataframe containing data from the control group, specifically the outcome Y, the surrogate S, and the baseline covariates W
data.treat	dataframe containing data from the treamtent group, specifically the outcome Y, the surrogate S, and the baseline covariates W
W.grid.expand	expanded version of the W grid of baseline covariates, where each row is a specific combination of the covariates for which the estimates should be provided

## Value

grid of estimates of the proportion of treatment effect explained, the overall treatment effect, and the residual treatment effect for the given baseline covariate combinations

two.step.est	Estimates the proportion of treatment effect explained as a function of multiple baseline covariates, W, using a two step, semiparametric model

# Description

Estimates the proportion of treatment effect explained as a function of multiple baseline covariates, W, using a two step, semiparametric model

## Usage

```
two.step.est(data.control, data.treat, W.grid.expand.function)
```

# Arguments

data.control	dataframe containing data from the control group, specifically the outcome $Y$ , the surrogate $S$ , and the baseline covariates $W$	
data.treat	dataframe containing data from the treamtent group, specifically the outcome Y, the surrogate S, and the baseline covariates W	
W.grid.expand.function		
	expanded version of the W grid of baseline covariates, where each row is a specific combination of the covariates for which the estimates should be provided	

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# Value

grid of estimates of the proportion of treatment effect explained, the overall treatment effect, and the residual treatment effect for the given baseline covariate combinations

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