

# Package ‘tbd’

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**Title** Estimation of causal effects with outcomes truncated by death

**Version** 0.0.0.9000

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**Description** Estimation of the survivor average causal effect under outcomes truncated by death, which requires the existence of a substitution variable. It can be applied to both experimental and observational data.

**Depends** R (>= 3.3.2), numDeriv, stats

**License** GPL-3

**Encoding** UTF-8

**URL** <https://github.com/KillingVectorField/causal-inference-truncated-by-death>

**LazyData** true

**RoxygenNote** 5.0.1

**NeedsCompilation** no

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boot.ci	<i>Estimate the confidence interval of SACE using bootstrap.</i>
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## Description

Give quantiles of bootstrap samples SACE.

**Usage**

```
boot.ci(object, nboot = 1000, seed = 100:(100 + nboot - 1), alpha = 0.05,
        max.step = 1000, singular.ok = FALSE, print.progress = TRUE)
```

**Arguments**

object	an object of class sace.
nboot	a positive integer. The number of bootstrap samples desired.
seed	an integer vector with length nboot. Seed to generate samples.
alpha	confidence level.
max.step	see documentation of <a href="#">sace</a> .
singular.ok	see documentation of <a href="#">sace</a> .
print.progress	logical. Need progress be printed?

**Value**

a list with 4 elements:

nskip	number of failures during bootstrap.
sace.boot.record	a vector with length nboot-skip. SACE estimates of all bootstrap samples.
boot.sd	scaler. Standard deviation of SACE estimates of all bootstrap samples.
ci	a vector with length 2. Estimated confidence interval.

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print.sace	<i>Print results of sace</i>
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**Description**

print.sace prints estimation of the SACE (survivor average causal effect).

**Usage**

```
## S3 method for class 'sace'
print(x, ...)
```

**Arguments**

x	an object of class sace.
...	additional arguments.

**Value**

the input object is returned silently.

sace

*Estimation of causal effects with outcomes truncated by death***Description**

sace estimates survivor average causal effects (SACE) with outcomes truncated by death.

**Usage**

```
sace(Z, S, Y, X, A, subset, optim.method = "BFGS", max.step = 1000,
     singular.ok = TRUE, need.variance = TRUE, hessian = TRUE)
```

**Arguments**

Z	a logical vector. Exposure indicator. Conventionally, 1 means treatment and 0 means control. Must not have missing values.
S	a logical vector. Survival indicator. 1 means survival and 0 means death. Must not have missing values.
Y	a numeric vector. (Univariate) outcomes. May have NA where $S = 0$ (since $Y$ is not well-defined where $S = 0$ ).
X	an optional numeric matrix or vector. Baseline covariates.
A	an optional numeric matrix or vector. Substitution variable(s) which satisfies the assumptions of "exclusion restriction" and "substitution relevance". See references. If $A == \text{NULL}$ , then the naive method, namely OLS, will be used.
subset	an optional vector specifying a subset of observations to be used.
optim.method	The method to be used for maximum likelihood optimization. See <a href="#">optim</a> .
max.step	integer. Maximum iterating steps of maximum likelihood optimization.
singular.ok	logical. Refers to the OLS estimation of the coefficients $\alpha_1$ and $\alpha_2$ using <a href="#">lm</a> . If FALSE (default), a singular fit raises an error.
need.variance	logical. Is variance of parameters and estimators needed? See details.
hessian	logical. If TRUE, the hessian returned by <a href="#">optim</a> will be used to compute the information matrix. If FALSE, the matrix will be calculated by an explicit formula.

**Details**

This function `sace`, gives estimation of average causal effects (ACE) with outcomes truncated by death. The identification of SACE relies on the existence of a substitution variable and requires the assumptions of monotonicity, ignorability, exclusion restriction, and relevance. While the naive estimates given by the coefficient of  $Z$  from  $\text{lm}(Y \sim Z + X + A, \text{subset} = S == 1)$  are restricted among survivors and therefore may be subject to selection bias, this method gives consistent estimates of the SACE (survivor average causal effect), defined as the average causal effect among the subgroup consisting of subjects who would survive under either exposure, i.e. among the always-survivor group ( $G = LL$ ). See references for details of the assumptions and the model parameterizations.

Parameters  $\beta$  and  $\gamma$  are estimated by MLE, using [optim](#).

If `need.variance == TRUE`, the asymptotic variance estimators of both parameters and estimators will be given. This requires the **numDeriv** package.

**Value**

a list with following elements:

CALL	function call.
data	data used (within the specified subset).
optim.method	method used for optimization.
need.variance	is variance of parameters and estimators needed?
n	sample size.
mu_0_LL	average potential outcomes among control group, $E[Y(0) G = LL]$ .
mu_1_LL	average potential outcomes among treatment group, $E[Y(1) G = LL]$ .
sace	survivor average causal effect, equals mu_1_LL-mu_0_LL.
beta	$PrS(1) = 1 X, A = \text{expit}(\beta_0 + X'\beta_1 + A\beta_2)$ , estimated by MLE.
gamma	$PrS(0) = 1 X, A/PrS(1) = 1 X, A = \text{expit}(\gamma_0 + X'\gamma_1 + A\gamma_2)$ , estimated by MLE.
beta_gamma.convergence	indicator of convergence of MLE optimization of beta and gamma. 0 means convergence. See <a href="#">optim</a> .
alpha_1	$E[Y(0) Z = 0, G = LL, X, A] = \alpha_{10} + X'\alpha_{11} + A\alpha_{12}$ , coefficients of $\text{lm}(Y \sim 1 + X + A, \text{subset} = Z == 0)$ .
alpha_2	$E[Y(1) Z = 1, G = LL, X, A] = \alpha_{20} + X'\alpha_{21} + A\alpha_{22}$ , coefficients of $\text{lm}(Y \sim 1 + X + W.\text{expit}, \text{subset} = (Z == 1 \& S == 1))$ .

The following items will be given only if need.variance == TRUE:

beta.var	estimated asymptotic covariance matrix of beta.
gamma.var	estimated asymptotic covariance matrix of gamma.
relevance.Pvalue	P value of the asymptotic chi-squared test on the relevance assumption for the substitution variable. A large P value suggests that the relevance assumption may not hold, namely, the substitution variable(s) may have little impact on the latent survival type.
alpha_1.var	estimated asymptotic covariance matrix of alpha_1.
alpha_2.var	estimated asymptotic covariance matrix of alpha_2.
mu_0_LL.var	estimated asymptotic variance of mu_0_LL.
mu_1_LL.var	estimated asymptotic variance of mu_1_LL.
sace.var	estimated asymptotic variance of the SACE.

**Note**

The length of vectors Z, Y, S, as well as the row number of matrix X and A must equal the sample size n.

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## References

Linbo Wang, Xiao-Hua Zhou, Thomas S. Richardson; Identification and estimation of causal effects with outcomes truncated by death, *Biometrika*, Volume 104, Issue 3, 1 September 2017, Pages 597-612, <https://doi.org/10.1093/biomet/asx034>

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simulated_data	<i>Simulated data with known SACE</i>
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## Description

This simulated dataset is to illustrate how to use [sace](#) to estimate the SACE, and compare it with other naive methods. In this simulated data, by design, there is confounding between Z and Y caused by X, and confounding between S and Y caused by X.

## Format

A data frame with 5000 observations and 7 variables. Z, A, Y, S are 1-dimensional, and X is 3-dimensional. The variables are as follows:

**Z** Binary treatment

**X.X1** A factor covariate with 2 levels (1 and -1)

**X.V2** A continuous covariate

**X.V3** A continuous covariate

**A** The substitution variable which is continuous

**Y** The continuous outcome. NA where  $S = 0$

**S** The survival indicator. 1 means survival and 0 means death.

## Source

The dataset is generated by the simulation design of Wang *et al.* 2017 with  $\delta_1 = 1$  and  $\delta_0 = 1$ , which allows confounding between Z and Y caused by X, and confounding between S and Y caused by X.

## References

Linbo Wang, Xiao-Hua Zhou, Thomas S. Richardson; Identification and estimation of causal effects with outcomes truncated by death, *Biometrika*, Volume 104, Issue 3, 1 September 2017, Pages 597-612, <https://doi.org/10.1093/biomet/asx034>

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summary.sace	<i>Summarize results of sace</i>
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**Description**

summary.sace summary estimation of the SACE (survivor average causal effect) and all other model parameters.

**Usage**

```
## S3 method for class 'sace'  
summary(object, ...)
```

**Arguments**

object	an object of class sace.
...	additional arguments.

**Value**

the input object is returned silently.

**Note**

If need.variance is TRUE, sace must have been called with need.variance == TRUE, so that the information needed was recorded.

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