Package ‘idem’

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Title  Inference in Randomized Controlled Trials with Death and Missingness

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Description In randomized studies involving severely ill patients, functional outcomes are often unobserved due to missed clinic visits, premature withdrawal or death. It is well known that if these unobserved functional outcomes are not handled properly, biased treatment comparisons can be produced. In this package, we implement a procedure for comparing treatments that is based on the composite endpoint of both the functional outcome and survival. The procedure was proposed in Wang et al. (2016) <DOI:10.1111/biom.12594>. It considers missing data imputation with a sensitivity analysis strategy to handle the unobserved functional outcomes not due to death.

Depends R (>= 3.4.0), Rcpp (>= 0.12.0), methods

License GPL (>= 3)

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RcppModules stan_fit4idem_mod

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VignetteBuilder knitr

SystemRequirements GNU make
Description

This package contains the functions for drawing inference in randomized clinical trials with death and intermittent missingness.

Notation

Consider a two-arm randomized study. Let $Y_k$ denote outcome measured at time $t_k$ and $Z$ denote a functional endpoint that is a function of $Y$. Let $L$ denote the survival time. Let $X$ denote the baseline covariates and $T$ denote the treatment assignment.
Ranking

If two subjects were both alive at the end of the study, they are ranked based on functional outcome $Z$. If at least one subject was dead at the end of the study, they are ranked based on survival time $L$.

Treatment effect, $\theta$ is defined as the probability that the outcome for a random individual randomized to treatment $T = 0$ is less than the outcome of a random individual randomized to treatment $T = 1$ minus the probability that the outcome for a random individual randomized to treatment $T = 0$ is greater than the outcome of a random individual randomized to treatment $T = 1$.

Missingness

In order to estimate $\theta$ in the presence of missing data, we need to impute $Z$ for subjects alive at the end of the study with $Y_k$ missing for some $k$.

The benchmark assumption we consider for the imputation is the complete case missing value (CCMV) restrictions. We then consider exponential tilting models for introducing sensitivity parameters for evaluating the robustness of the findings with regards to different missing data mechanism assumptions. The models are as follows:

$$f(Y_{\text{mis}}(s)|Y_{\text{obs}}, Y_0, X, T, S = s) \propto \exp(\beta_T Z) f(Y_{\text{mis}}(s)|Y_{\text{obs}}, Y_0, X, T, S = 1)$$

where $S$ denotes the missingness patterns, $S = 1$ denotes the completers and $\beta_T$ denotes the sensitivity parameter for arm $T$.

Graphical user interface (GUI)

This package provides a web-based GUI. See imShiny for details.

References


Example dataset

The Awakening and Breathing Controlled (ABC) trial randomized critically ill patients receiving mechanical ventilation 1:1 within each study site to management with a paired sedation plus ventilator weaning protocol involving daily interruption of sedative through spontaneous awakening trials (SATs) and spontaneous breathing trials (SBTs) or sedation per usual care (UC) and SBTs. The example dataset is from a single site substudy in ABC. The researchers assessed differences in cognitive, psychological and functional outcomes at 3 and 12 months after randomization.
Format

A dataframe with 5 variables:

- **AGE** Age
- **TRT** Treatment assignment. 0: UC + SBT, 1: SAT + SBT
- **SURV** Survival days
- **Y2** Cognitive score at 12 months
- **Y1** Cognitive score at 3 months

References


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**imData**

Create data for IDEM analysis

**Description**

Create a class IDEMDATA object for IDEM analysis

**Usage**

```r
imData(data, trt = NULL, surv = NULL, outcome = NULL, endfml = NULL, y0 = NULL, cov = NULL, duration = 9999, bounds = NULL, trt.label = NULL, unitTime = "days", err.terminate = TRUE, ...)
```

**Arguments**

- **data** Original dataset
- **trt** Variable name for the Control (0) and Intervention (1) treatment assignments in the dataset
- **surv** Variable name for the survival (time to event) variable in the dataset
- **outcome** Chronologically ordered vector of variable names for clinical outcomes in the dataset excluding baseline
- **endfml** R expression indicating the user-specified final outcome of interest. This is the function for Z of one or more of Yk’s
- **y0** Variable name of the baseline clinical outcome
imFitModel

imFitModel

Description

Fit linear imputation models to the observed data from complete survivors for each treatment arm at each time point.

Usage

imFitModel(im.data)
**Arguments**

- `im.data`: A class IDEMDATA object generated by `imData`

**Value**

A class IDEMFIT list of modeling fitting results with the following items:

- `im.data`: Original class IDEMDATA object
- `rst.mdl`: A list of modeling fitting results for each model with
  - `lm`: results from function `lm`
  - `formula`: model formula
  - `coef`: model coefficients
  - `res`: residuals
  - `h`: bandwidth of residuals for kernel density estimation

**See Also**

- `imData`, `idem-package`

**Examples**

```r
im.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=NULL, endfml="Y2",
trt.label = c("UC+SBT", "SAT+SBT"),
cov=c("AGE"), duration=365, bounds=c(0,100));
im.fit <- imFitModel(im.abc);
```

---

**Description**

Conduct imputation under benchmark assumptions or for sensitivity analysis for a given set of subjects using the model fitting results

**Usage**

```r
imImpAll(fit.rst, data.all = NULL, deltas = 0, normal = TRUE,
n.imp = 5, endponly = TRUE, update.progress = NULL,
imputeNone = FALSE, ..., seed = NULL)
```
Arguments

fit.rst  A class IDEMFIT results generated by \texttt{imFitModel}.
data.all  A dataframe containing subjects with missing data. The default value is NULL, in which case the function will impute missing data for subjects in the original dataset in the class IDEMFIT object \texttt{fit.rst}
deltas  Vector of imputation sensitivity parameters
normal  Logical variable indicating whether normality assumption should be made for the residuals
n.imp  Number of complete datasets required
deltas  Vector of imputation sensitivity parameters

Value

If \texttt{imputeNone} is TRUE, return a dataset with the original data for the subset of subjects who died at the end of the study or had no missing outcomes.
Otherwise, return a class IDEMIMP list with components

\textbf{lst.var} List of parameters

\textbf{complete} A dataset with the original data for the subset of subjects who died at the end of the study or had no missing outcomes and the \texttt{n.imp} imputed missing outcomes for subjects who need missing value imputation.

\textbf{n.imp} Number of imputed complete datasets

\textbf{deltas} Imputation sensitivity parameters

\textbf{org.data} Original dataset

\textbf{normal} Normal assumption for the imputation

\textbf{stan.par} STAN options

Examples

```r
## Not run:
rst.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"), 
y0=NULL, endfml="Y2", 
trt.label = c("UC+SBT", "SAT+SBT"),
```
imImpSingle

```r
cov=c("AGE"), duration=365, bounds=c(0,100));
rst.fit <- imFitModel(rst.abc);
rst.imp <- imImpAll(rst.fit, deltas=c(-0.25,0,0.25),
                   normal=TRUE, chains = 2, iter = 2000, warmup = 1000);
## End(Not run)
```

---

**imImpSingle**  
Impute missing data for MCMC convergence checking

**Description**

Call STAN model to impute missing data for an individual subject under benchmark assumption for MCMC convergence checking

**Usage**

```r
imImpSingle(dsub, fit.rst, normal = TRUE, chains = 4, iter = 5000,
            warmup = 1000, control = list(adapt_delta = 0.95), ..., 
            seed = NULL)
```

**Arguments**

- **dsub** original individual subject data
- **fit.rst** A class IDEMFIT results generated by `imFitModel`
- **normal** Logical variable indicating whether normality assumption should be made for the residuals
- **chains** STAN parameter. Number of Markov chains
- **iter** STAN parameter. Number of iterations
- **warmup** STAN parameter. Number of burnin.
- **control** STAN parameter. See `rstan::stan` for details.
- **...** other options to call STAN sampling such as `thin`, `algorithm`. See `rstan::sampling` for details.
- **seed** Random seed

**Value**

NULL if there is no missing data in `dsub`
Otherwise, return a class IDEMSINGLE object that contains a list with components

- **dsub** original data of the subject
- **rst.stan** A stan.fit class result returned from `rstan::sampling`
- **complete** A dataframe with complete data for the selected subject
Examples

```r
im.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1", "Y2"),
y0=NULL, endfml="Y2",
trt.label = c("UC+SBT", "SAT+SBT"),
cov=c("AGE"), duration=365, bounds=c(0,100));
im.fit <- imFitModel(im.abc);
im.imp <- imImpSingle(abc[1,], im.fit, chains = 4, iter = 200, warmup = 100);
```

---

**imInfer**

*Treatment effect estimation and hypothesis testing*

**Description**

Estimate treatment effect size. Estimate variation and conduct hypothesis testing by bootstrap analysis.

**Usage**

```r
imInfer(imp.rst, n.boot = 0, n.cores = 1, update.progress = NULL,
effect.quantiles = 0.5, quant.ci = c(0.025, 0.975), ..., seed = NULL)
```

**Arguments**

- `imp.rst` A class IDEMIMP object returned from `imImpAll`
- `n.boot` Number of bootstrap samples
- `n.cores` Number of cores for parallel computation. Fixed at 1 for Windows.
- `update.progress` Parameter reserved for run idem in GUI mode
- `effect.quantiles` Composite quantiles of interest for measuring treatment effect
- `quant.ci` Quantiles for extracting bootstrap confidence intervals
- `...` Extra options for ranking subjects using the composite endpoint that include
  - `cut.z`: Clinically meaningful difference in the functional outcome
  - `cut.surv`: Clinically meaningful difference in survival time
- `seed` Random seed

**Details**

If `n.boot`=0, bootstrap analysis will not be conducted. Instead, only the treatment effect size will be estimated using the imputed data.
Value

A class IDEMTEST list containing

**lst.var**  List of specification parameters

**deltas**  Vector of sensitivity parameters

**theta**  A data frame with columns

- Delta0: Sensitivity parameter for control arm
- Delta1: Sensitivity parameter for intervention arm
- Theta: Estimated $\theta$
- SD: Standard deviation (when n.boot >0)
- PValue: p-value (when n.boot >0)

**effect.quantiles**  A data frame with columns

- Delta: Sensitivity parameter
- TRT: Treatment arm
- Q: Quantiles of the composite endpoint to be estimated
- QuantY: Estimated quantiles if the quantiles correspond to functional outcome (when n.boot >0)
- QuantSurv: Estimated quantiles if the quantiles correspond to survival days (when n.boot >0)
- Q: Bootstrap quantiles for the QuantY (when n.boot >0)
- QSurv: Bootstrap quantiles for the QuantSurv (when n.boot >0)

**bootstrap**  A list with length n.boot. The $i$th item is the class IDEMEST list corresponding to the $i$th bootstrap sample

Examples

```r
## Not run:
rst.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=NULL, endfml="Y2",
trt.label = c("UC+SBT", "SAT+SBT"),
cov=c("AGE"), duration=365, bounds=c(0,100));
rst.fit <- imFitModel(rst.abc);
rst.imp <- imImpAll(rst.fit, deltas=c(-0.25,0,0.25),
                   normal=TRUE, chains = 2, iter = 2000, warmup = 1000);
rst.est <- imInfer(rst.imp, n.boot = 0, effect.quantiles = c(0.25,0.5,0.75));
rst.test <- imInfer(rst.imp, n.boot = 100, effect.quantiles = c(0.25,0.5,0.75));
## End(Not run)
```
**imShiny**

**Run Web-Based idem application**

**Description**

Call Shiny to run idem as a web-based application.

**Usage**

`imShiny()`

**Details**

A web browser will be brought up for users to access the GUI of idem.

**Examples**

```r
## Not run:
run.idem()
## End(Not run)
```

---

**plot.IDEMDATA**

**Plot of IDEMDATA object**

**Description**

Generate different types of plots for class IDEMDATA objects.

**Usage**

```r
## S3 method for class 'IDEMDATA'
plot(x, opt = c("survivor", "missing", "KM"),
     cols = c("black", "blue"), fname = NULL, ...)
```

**Arguments**

- `x` A class IDEMDATA object generated by `imData`
- `opt` Types of the plot
  - `survivor`: Spaghetti plot for subjects alive at the end of the study
  - `missing`: Plot the missing patterns of the observed data
  - `KM`: Plot Kaplan-Meier survival curves
- `cols` Curve colors of the treatment and control arm for survival plot or colors of the observed and missing data for missingness plot.
description: Plot method of the class IDEMFIT to generate model fitting diagnosis plots

usage: ## S3 method for class 'IDEMFIT'
plot(x, trt = NULL, mfrow = NULL, ...)

arguments: x A class IDEMFIT object generated by imFitModel
trt Treatment arm selected for the diagnostic plots. If NULL, all treatment arms are included
mfrow Plot option
...

see also: imFitModel
Examples

```r
im.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"), 
y0=NULL, endmf="Y2", 
trt.label = c("UC+SBT", "SAT+SBT"), 
cov=c("AGE"), duration=365, bounds=c(0,100));
im.fit <- imFitModel(im.abc);
plot(im.fit, mfrow=c(2,4));
```

Description

Generate different types of plots for class IDEMIMP objects generated by `imImpAll`

Usage

```r
## S3 method for class 'IDEMIMP'
plot(x, opt = c("imputed", "composite"), 
     fname = NULL, ...)
```

Arguments

- `x` A class IDEMIMP object returned from `imImpAll`
- `opt` Types of the plot
  - `imputed`: Plot density of imputed values and the density of the observed outcomes
  - `composite`: Generate cumulative plot of the composite survival and functional outcome
- `fname` File name of the result pdf file. If `fname` is null, result pdf file will not be generated
- `...` Options for generating the plots.
  - `type = imputed`
    - `deltas`: Imputation sensitivity parameter for which to generate the results
    - `endp`: If TRUE, plot the densities of the imputed functional outcomes. Otherwise, plot the densities of the imputed outcomes
    - `adjdensity` estimation option
    - `colsplot` option for colors
    - `ltysplot` options for line types
    - `xlimplot` options
    - `ylimplot` options
    - `mfrowplot` options
type = composite

• at.surv: Sets the range of the survival times to plot in the cumulative distribution function. By default the range is the range of survival values up to the duration of the study
• at.z: Sets the range of the functional outcome to plot in the cumulative distribution function. By default this is the range of the functional outcomes plus the buffer amount to improve visibility in the transition from survival to functional outcome
• p.death: Proportion of the plot width devoted to Survival. By default the cumulative distribution will devote horizontal space to the survival portion that is proportional to the number of subjects who die prior to duration
• buffer: Small horizontal gap used to better visually distinguish the transition from survival to functional outcome
• delta: Imputation sensitivity parameter for which to generate the results
• seg.lab: Labels for the two components of the composite outcome
• main: plot options

See Also

imImpAll

Examples

```r
## Not run:
im.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=NULL, endfml="Y2",
    trt.label = c("UC+SBT", "SAT+SBT"),
    cov=c("AGE"), duration=365, bounds=c(0,100));
rst.fit <- imFitModel(im.abc);
rst.imp <- imImpAll(rst.fit, deltas=c(-0.25,0.0,0.25),
    normal=TRUE, chains = 2, iter = 2000, warmup = 1000);
plot(rst.imp, opt = "imputed"),
plot(rst.imp, opt = "composite")
## End(Not run)
```

plot.IDEMINFER  
*Plot hypothesis testing results*

Description

Generate contour plot of p-values or treatment effect theta for sensitivity analysis results

Usage

```r
## S3 method for class 'IDEMINFER'
plot(x, con.v = 0.05, nlevels = 30,
    opt = c("pvalue", "effect"), ...)
```
Arguments

x  A class IDEMINFER list generated by \texttt{imInfer}

\texttt{conv}  Levels of contour plot

\texttt{nlevels}  Levels of color scale

\texttt{opt}  contour plots of pvalue or effect

\ldots  Options for \texttt{filled.contour}

Details

The plot will only be generated when bootstrap analysis has been conducted.

Examples

\begin{verbatim}
## Not run:
rst.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=NULL, endfml="Y2",
trt.label = c("UC\+SBT", "SAT\+SBT"),
cov=c("AGE"), duration=365, bounds=c(0,100));
rst.fit <- imFitModel(rst.abc);
rst.imp <- imImpAll(rst.fit, deltas=c(-0.25,0,0.25),
                     normal=TRUE, chains = 2, iter = 2000, warmup = 1000);
rst.est <- imInfer(rst.imp, n.boot = 100);
plot(rst.est);
## End(Not run)
\end{verbatim}

\section*{Description}

Plot method of the class IDEMSINGLE to generate traceplot of the imputed missing outcomes

Usage

\begin{verbatim}
## S3 method for class 'IDEMSINGLE'
plot(x, ...)
\end{verbatim}

Arguments

\begin{itemize}
  \item \texttt{x}  A class IDEMSINGLE object returned from \texttt{imImpSingle}
  \item \ldots  Additional arguments
\end{itemize}

See Also

\texttt{imImpSingle}
Examples

```r
im.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=NULL, endfml="Y2",
trt.label = c("UC+SBT", "SAT+SBT"),
cov=c("AGE"), duration=365, bounds=c(0,100));
im.fit <- imFitModel(im.abc);
im.imp.single <- imImpSingle(abc[1,], im.fit,
    chains = 4, iter = 200, warmup = 100);
plot(im.imp.single);
```

Description

Generate a plot of survivor only and survivor average causal effect values

Usage

```r
## S3 method for class 'summary.IDEMINFER'
plot(x, opt = c("pvalue", "effect"),
    by.sace = TRUE, delta0 = 0, delta1 = 0, sace.delta = NULL, ...)
```

Arguments

- `x`: A class `summary.IDEMINFER` object generated by `summary` of `IDEMINFER`
- `opt`: contour plots of `pvalue` or `effect`
- `by.sace`: Logical value. If `TRUE`, create a contour plot for given SACE sensitivity parameter. Otherwise, create a plot for treatment effect for given imputation sensitivity parameters
- `delta0`: Selected treatment arm 0 sensitivity parameters
- `delta1`: Selected treatment arm 1 sensitivity parameters
- `sace.delta`: Single SACE sensitivity parameter
- `...`: Options for `plot`

Details

The plot function will only generate the contour plot of p-values or treatment effects on functional outcomes for survivors only analyses.

For SACE analysis, the plot function generates contour plot of line plot based on the value of `by.sace`.
Examples

```r
## Not run:
rst.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=0, endfm1="Y2",
trt.label = c("UC+SBR", "SAT+SBR"),
cov=c("AGE"), duration=365, bounds=c(0,100));

rst.fit <- imFitModel(rst.abc);

rst.imp <- imImpAll(rst.fit, deltas=c(-0.25,0,0.25),
normal=TRUE, chains = 2, iter = 2000, warmup = 1000);

rst.infer <- imInfer(rst.imp, n.boot = 100, effect.quantiles = c(0.25,0.5,0.75));

rst.survivors <- summary(rst.infer, opt="survivor");
plot(rst.survivors);
## End(Not run)
```

---

### print.IDEMDATA

**Print IDEMDATA object**

#### Description

Print the specification details of class IDEMDATA objects generated by `imData`

#### Usage

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

#### Arguments

- `x` A class IDEMDATA object generated by `imData`
- `...` Additional arguments

#### See Also

`imData`

### print.IDEMERROR

**Print error messages**

#### Description

Print error messages in the parameter specifications generated by `imData`

#### Usage

```r
## S3 method for class 'IDEMERROR'
print(x, html = FALSE, ...)
```
Arguments

- **x**: A class IDEMERROR object returned by `imData` when there are misspecifications
- **html**: Logical indicator for the format of the error messages. When TRUE, the error messages are formatted in HTML format
- **...**: Additional arguments

See Also

- `imData`

Examples

```r
## Not run:
rst.data <- imData(abc, trt="TRT", outcome=c("Y1","Y2"), y0=NULL,
                    endfml="Y3", bounds=c(10,20), duration=365,
                    err.terminate=FALSE);
print(rst.data);
## End(Not run)
```

---

**print.IDEMFIT**  
*Print model fitting results*

**Description**

Print method of the class IDEMFIT generated by `imFitModel`

**Usage**

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

**Arguments**

- **x**: A class IDEMFIT object generated by `imFitModel`
- **...**: Additional arguments

**Details**

Print the results from `lm` for all the models

**See Also**

- `imFitModel`
### print.IDEMIMP

**Print imputation results**

**Description**

Print method for class IDEMIMP objects generated by `imImpAll`

**Usage**

```r
## S3 method for class 'IDEMIMP'
print(x, ...)  # A class IDEMIMP object returned from imImpAll

...  # Extra arguments
```

**See Also**

`imImpAll`

### print.IDEINFER

**Print inference results**

**Description**

Print method of class IDEMINFER for treatment effect estimation and hypothesis testing results

**Usage**

```r
## S3 method for class 'IDEMINFER'
print(x, delta0 = NULL, delta1 = NULL, ...)  # A class IDEMINFER list generated by imInfer
```

**Arguments**

- `x`: A class IDEMINFER list generated by `imInfer`
- `delta0`: Selected treatment arm 0 sensitivity parameters
- `delta1`: Selected treatment arm 1 sensitivity parameters
- `...`: Extra arguments
Examples

```r
## Not run:
rst.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=NULL, endfm="Y2",
trt.label = c("UC+SBT", "SAT+SBT"),
cov=c("AGE"), duration=365, bounds=c(0,100));
rst.fit <- imFitModel(rst.abc);
rst.imp <- imImpAll(rst.fit, deltas=c(-0.25,0,0.25),
                   normal=TRUE, chains = 2, iter = 2000, warmup = 1000);
rst.test <- imInfer(rst.imp, n.boot = 100);
print(rst.test, delta0 = 0, delta1 = 0.15)
## End(Not run)
```

Description

Print method for class IDEMSINGLE objects generated by `imImpSingle`.

Usage

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

Arguments

- `x` A class IDEMSINGLE object returned from `imImpSingle`
- `...` Additional arguments

See Also

- `imImpSingle`

Description

Print survivors only or SACE analysis results
Usage

```r
# S3 method for class 'summary.IDEMINFER'
print(x, delta0 = NULL, delta1 = NULL, ...)
```

Arguments

- `x`: A class IDEMINFER list generated by `imInfer`
- `delta0`: Selected treatment arm 0 sensitivity parameters
- `delta1`: Selected treatment arm 1 sensitivity parameters
- `...`: Extra arguments

Description

Summarize the missing data information for class `IDEMDATA` objects generated by `imData`.

Usage

```r
# S3 method for class 'IDEMDATA'
summary(object, opt = c("misstable", "missid", "trt"),
         endponly = FALSE, ...)
```

Arguments

- `object`: A class IDEMDATA object returned by `imData`
- `opt`: Types of the summarization
  - `misstable`: Summarize the frequencies of each missing pattern
  - `missid`: Summarize the indices of subjects that need imputation, i.e. survivors with functional endpoint missing
  - `trt`: Treatment arms
- `endponly`: Logical variable that indicates whether clinical outcomes not used in calculating the functional outcome are considered as missing and should be imputed. The default is `FALSE`, indicating that all missing clinical outcomes will be imputed sequentially
- `...`: Extra arguments

Value

A dataframe with frequencies of each missing pattern when `opt` is `misstable`.
A vector of indices of subjects that need imputation when `opt` is `missid`.
A vector of treatment arms in the data when `opt` is `trt`.

See Also

- `imData`
Description

Summarize survivors only or Survivor Averaged Causal Effect (SACE) based on the imputation and bootstrap analysis

Usage

```r
## S3 method for class 'IDEMINFER'
summary(object, opt = c("survivor", "SACE"),
         sace.deltas = NULL, ...)
```

Arguments

- `object`: A class IDEMINFER list generated by `imInfer`
- `opt`: Types of the summary
  - `survivor`: Survivors only analysis
  - `SACE`: Survivor Averaged Causal Effect
- `sace.deltas`: Vector of sensitivity parameters for SACE estimation. If `NULL`, the values will be generated based on the standard deviations of the estimated differences in the functional outcomes between the treatment and control arms
- `...`: Optional arguments for summary

Details

For SACE, the default sensitivity parameters will be determined by the standard deviation of the treatment effect size on the functional outcomes.

Value

A class `summary.IDEMINFER` list containing

- `deltas`: imputation sensitivity parameters
- `n.boot`: number of bootstrap samples in bootstrap analysis
- `sace.deltas`: SACE sensitivity parameters when `opt = SACE`
- `rst`: A data frame with columns
  - `Delta0`: Imputation sensitivity parameter for control arm,
  - `Delta1`: Imputation sensitivity parameter for intervention arm
  - `SACE_Delta`: SACE sensitivity parameter when `opt = SACE`
  - `Effect`: SACE estimate
  - `LB`: Lower bound of the 95
  - `UB`: Upper bound of the 95
  - `PValue`: p-value when `n.boot > 0` in the IDEMINFER object
References


Examples

```r
## Not run:
rst.abc <- imData(abc, trt="TRT", surv="SURV", outcome=c("Y1","Y2"),
y0=NULL, end.fml="Y2",
trt.label = c("UC+SBT", "SAT+SBT"),
cov=c("AGE"), duration=365, bounds=c(0,100));
rst.fit <- imFitModel(rst.abc);
rst.imp <- imImpAll(rst.fit, deltas=c(-0.25,0,0.25),
normal=TRUE, chains = 2, iter = 2000, warmup = 1000);
rst.infer <- imInfer(rst.imp, n.boot = 100, effect.quantiles = c(0.25,0.5,0.75));
rst.sace <- summary(rst.infer, opt = "SACE")
## End(Not run)
```
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