Package 'ltmle'

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Type Package

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Description Targeted Maximum Likelihood Estimation (TMLE) of
treatment/censoring specific mean outcome or marginal structural model for
point-treatment and longitudinal data.

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'BinaryToCensoring.R' 'DeterministicFunctions.R' 'zzz.R'

URL https://github.com/joshuaschwab/ltmle

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ltmle-package | *Targeted Maximum Likelihood Estimation for Longitudinal Data*

Description

Targeted Maximum Likelihood Estimation (TMLE) of treatment/censoring specific mean outcome or marginal structural model for point-treatment and longitudinal data. Also provides Inverse Probability of Treatment/Censoring Weighted estimate (IPTW) and maximum likelihood based G-computation estimate (G-comp). Can be used to calculate additive treatment effect, risk ratio, and odds ratio.

Author(s)

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References

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See Also

ltmle

Examples

```
## For examples see examples(ltmle)
```

BinaryToCensoring   *BinaryToCensoring*

Description

Helper function for creating censoring columns as factors.

Usage

```
BinaryToCensoring(is.censored, is.uncensored)
```

Arguments

is.censored   binary vector: 0=uncensored, 1=censored

is.uncensored   binary vector: 0=censored, 1=uncensored

Details

Exactly one of is.censored and is.uncensored must be specified as a *named* argument. All elements of the input vector must be 0, 1, or NA

Value

an object of class "factor" with levels "censored" and "uncensored"

Author(s)

Joshua Schwab <jschwab77@berkeley.edu>

See Also

factor

Examples

```
  BinaryToCensoring(is.censored=c(0, 1, 1, 0, NA))
  BinaryToCensoring(is.uncensored=c(1, 0, 0, 1, NA))   #the same

 ## Not run:
 BinaryToCensoring(c(0, 1))   #error because the input must be named

 ## End(Not run)
```

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deterministic.g.function_template

Deterministic g/Q functions - examples and templates

Description

Template for the `deterministic.g.function` argument to [ltmle](#) or [ltmleMSM](#).

Usage

```
deterministic.g.function_template(data, current.node, nodes)

deterministic.Q.function_template(data, current.node, nodes,
  called.from.estimate.g)
```

Arguments

data	the 'data' data.frame passed to `ltmle` or `ltmleMSM`
current.node	the column index of data corresponding to the A or C node (for g) or L or Y node (for Q)
nodes	list of column indicies, components:

- A Anodes (treatment)
- C Cnodes (censoring)
- L Lnodes (time-varying covariates)
- Y Ynodes (events)
- AC Anodes and Cnodes combined and sorted
- LY Lnodes and Ynodes combined, sorted, "blocks" removed - see [ltmle](#)

called.from.estimate.g

TRUE or FALSE - your function will be called with `called.from.estimate.g=TRUE` during estimation of g and `called.from.estimate.g=FALSE` during estimation of Q.

Details

`MaintainTreatment` and `MaintainControl` are two commonly used `deterministic.g.function`s.

The intended use of the templates is for the user to copy and paste the function arguments and body and then fill in the required sections. They will not run as-is. Note that there are no comments in the functions as saved. Versions with comments may be found in Examples section below.

MaintainTreatment and MaintainControl may be passed as-is for the `deterministic.g.function` argument to [ltmle](#) or [ltmleMSM](#)

Note that censoring nodes in `data` may be passed as binaries but they are converted to the preferred format of factors with levels "censored" and "uncensored" before deterministic functions are called. Also note that nodes may be passed to ltmle as either the names of nodes or numerical column indicies, but they are all converted to numerical indicies before deterministic functions are called. If the `survivalFunction` argument to ltmle or ltmleMSM is TRUE, the package automatically assumes that once Y jumps to 1, all future Y nodes stay 1 and treatment does not change. It is not necessary to specify this in deterministic functions.

Value

A deterministic.g.function should return a list with components:

`is.deterministic`
vector of logicals, length=nrow(data)

`prob1`   the probability that data[is.deterministic, current.node] == 1, vector of length 1 or length(which(is.deterministic))

A deterministic.Q.function should return a list with components:

`is.deterministic`
vector of logicals, length=nrow(data)

`Q.value`   the iterated expectation of the final Y, vector of length 1 or length(which(is.deterministic))

NOTE: The `Q.value` component is not used or required when `called.from.estimate.g` is TRUE

Functions

- `deterministic.Q.function_template`: Template for the `deterministic.Q.function` argument to [ltmle](#) or [ltmleMSM](#).

Author(s)

Joshua Schwab <jschwab77@berkeley.edu>

See Also

[ltmle](#), [ltmleMSM](#)

Examples

```
# Show template for a deterministic.g.function (comments will not be
# shown, see below for comments)
deterministic.g.function_template

# Show template for a deterministic.Q.function (comments will not be
# shown, see below for comments)
deterministic.Q.function_template

# Use MaintainTreatment
set.seed(1)
rexpit <- function(x) rbinom(n = length(x), size = 1, prob = plogis(x))
n <- 100
W <- rnorm(n)
A1 <- rexpit(W)
A2 <- as.numeric(rexpit(W) | A1)  #treatment at time 1 implies treatment at time 2
Y <- rexpit(W + A1 + A2 + rnorm(n))
data <- data.frame(W, A1, A2, Y)

result <- ltmle(data, Anodes = c("A1", "A2"), Ynodes = "Y", abar = c(1, 1),
    deterministic.g.function = MaintainTreatment)

# deterministic.g.function_template with comments:

deterministic.g.function_template <- function(data, current.node, nodes) {
    # data: the 'data' data.frame passed to ltmle/ltmleMSM current.node: the
    # column index of data corresponding to the A or C node (see
    # is.deterministic below) nodes: list of column indicies, components: A,
    # C, L, Y, AC (Anodes and Cnodes combined and sorted), LY (Lnodes and
    # Ynodes combined, sorted, 'blocks' removed - see ?ltmle) Note that nodes
    # may be passed to ltmle as either the names of nodes or numerical column
    # indicies, but they are all converted to numerical indicies before
    # deterministic.g.function is called

    # deterministic.g.function will be called at all Anodes and Cnodes
    # return(NULL) is equivalent to return(list(is.deterministic=rep(FALSE,
    # nrow(data)), prob1=numeric(0)))

    # define is.deterministic here: vector of logicals, length=nrow(data)
    # define prob1 here: the probability that data[is.deterministic,
    # current.node] == 1, vector of length 1 or
    # length(which(is.deterministic))
    is.deterministic <- stop("replace me!")
    prob1 <- stop("replace me!")
    return(list(is.deterministic = is.deterministic, prob1 = prob1))
}

# deterministic.Q.function_template with comments:

deterministic.Q.function_template <- function(data, current.node, nodes,
    called.from.estimate.g) {
```

```
    # data: the 'data' data.frame passed to ltmle/ltmleMSM current.node: the
    # column index of data corresponding to the A or C node (see
    # is.deterministic below) nodes: list of column indicies, components: A,
    # C, L, Y, AC (Anodes and Cnodes combined and sorted), LY (Lnodes and
    # Ynodes combined, sorted, 'blocks' removed - see ?ltmle)
    # called.from.estimate.g: TRUE or FALSE - your function will be called
    # with called.from.estimate.g=TRUE during estimation of g and
    # called.from.estimate.g=FALSE during estimation of Q. During estimation
    # of g, only the is.deterministic element of the return list will be
    # used.  Note that nodes may be passed to ltmle as either the names of
    # nodes or numerical column indicies, but they are all converted to
    # numerical indicies before deterministic.Q.function is called

    # It is not necessary to specify that deterministic Y events (Y==1)
    # indicate a deterministic Q value of 1; this is automatic
    # if the survivalFunction input to ltmle/ltmleMSM is TRUE.
    # deterministic.Q.function will be called at all Lnodes and Ynodes (after
    # removing 'blocks') and Anodes and Cnodes (see called.from.estimate.g
    # above) return(NULL) is equivalent to
    # return(list(is.deterministic=rep(FALSE, nrow(data)),
    # Q.value=numeric(0)))

    # define is.deterministic here: vector of logicals, length=nrow(data)
    # define Q.value here: the iterated expectation of the final Y, vector of
    # length 1 or length(which(is.deterministic))
    is.deterministic <- stop("replace me!")
    Q.value <- stop("replace me!")
    return(list(is.deterministic = is.deterministic, Q.value = Q.value))
}
```

---

ltmle   *Longitudinal Targeted Maximum Likelihood Estimation*

Description

ltmle is Targeted Maximum Likelihood Estimation (TMLE) of treatment/censoring specific mean outcome for point-treatment and longitudinal data. ltmleMSM adds Marginal Structural Models. Both always provide Inverse Probability of Treatment/Censoring Weighted estimate (IPTW) as well. Maximum likelihood based G-computation estimate (G-comp) can be obtained instead of TMLE. ltmle can be used to calculate additive treatment effect, risk ratio, and odds ratio.

Usage

```
ltmle(data, Anodes, Cnodes = NULL, Lnodes = NULL, Ynodes,
  survivalOutcome = NULL, Qform = NULL, gform = NULL, abar,
  rule = NULL, gbounds = c(0.01, 1), Yrange = NULL,
  deterministic.g.function = NULL, stratify = FALSE,
  SL.library = "glm", SL.cvControl = list(), estimate.time = TRUE,
```

```
    gcomp = FALSE, iptw.only = FALSE, deterministic.Q.function = NULL,
    variance.method = "tmle", observation.weights = NULL, id = NULL)

ltmleMSM(data, Anodes, Cnodes = NULL, Lnodes = NULL, Ynodes,
    survivalOutcome = NULL, Qform = NULL, gform = NULL,
    gbounds = c(0.01, 1), Yrange = NULL,
    deterministic.g.function = NULL, SL.library = "glm",
    SL.cvControl = list(), regimes, working.msm, summary.measures,
    final.Ynodes = NULL, stratify = FALSE, msm.weights = "empirical",
    estimate.time = TRUE, gcomp = FALSE, iptw.only = FALSE,
    deterministic.Q.function = NULL, variance.method = "tmle",
    observation.weights = NULL, id = NULL)
```

Arguments

data	data frame following the time-ordering of the nodes. See 'Details'.
Anodes	column names or indicies in data of treatment nodes
Cnodes	column names or indicies in data of censoring nodes
Lnodes	column names or indicies in data of time-dependent covariate nodes
Ynodes	column names or indicies in data of outcome nodes
survivalOutcome	If TRUE, then Y nodes are indicators of an event, and if Y at some time point is 1, then all following should be 1. Required to be TRUE or FALSE if outcomes are binary and there are multiple Ynodes.
Qform	character vector of regression formulas for Q. See 'Details'.
gform	character vector of regression formulas for g or a matrix/array of prob(A=1). See 'Details'.
abar	binary vector (numAnodes x 1) or matrix (n x numAnodes) of counterfactual treatment or a list of length 2. See 'Details'.
rule	a function to be applied to each row (a named vector) of data that returns a numeric vector of length numAnodes. See 'Details'.
gbounds	lower and upper bounds on estimated cumulative probabilities for g-factors. Vector of length 2, order unimportant.
Yrange	NULL or a numerical vector where the min and max of Yrange specify the range of all Y nodes. See 'Details'.
deterministic.g.function	optional information on A and C nodes that are given deterministically. See 'Details'. Default NULL indicates no deterministic links.
stratify	if TRUE stratify on following abar when estimating Q and g. If FALSE, pool over abar.
SL.library	optional character vector of libraries to pass to [SuperLearner](). NULL indicates [glm]() should be called instead of [SuperLearner](). 'default' indicates a standard set of libraries. May be separately specified for Q and g. See 'Details'.
SL.cvControl	optional list to be passed as cvControl to [SuperLearner]()

estimate.time	if TRUE, run an initial estimate using only 50 observations and use this to print a very rough estimate of the total time to completion. No action if there are fewer than 50 observations.
gcomp	if TRUE, run the maximum likelihood based G-computation estimate *instead* of TMLE
iptw.only	by default (iptw.only = FALSE), both TMLE and IPTW are run in `ltmle` and `ltmleMSM`. If `iptw.only = TRUE`, only IPTW is run, which is faster.
deterministic.Q.function	
	optional information on Q given deterministically. See 'Details'. Default NULL indicates no deterministic links.
variance.method	
	Method for estimating variance of TMLE. One of "ic", "tmle", "iptw". If "tmle", compute both the robust variance estimate using TMLE and the influence curve based variance estimate (use the larger of the two). If "iptw", compute both the robust variance estimate using IPTW and the influence curve based variance estimate (use the larger of the two). If "ic", only compute the influence curve based variance estimate. "ic" is fastest, but may be substantially anti-conservative if there are positivity violations or rare outcomes. "tmle" is slowest but most robust if there are positivity violations or rare outcomes. "iptw" is a compromise between speed and robustness. variance.method="tmle" or "iptw" are not yet available with non-binary outcomes, gcomp=TRUE, stratify=TRUE, or deterministic.Q.function.
observation.weights	
	observation (sampling) weights. Vector of length n. If NULL, assumed to be all 1.
id	Household or subject identifiers. Vector of length n or NULL. Integer, factor, or character recommended, but any type that can be coerced to factor will work. NULL means all distinct ids.
regimes	binary array: n x numAnodes x numRegimes of counterfactual treatment or a list of 'rule' functions
working.msm	character formula for the working marginal structural model
summary.measures	
	array: num.regimes x num.summary.measures x num.final.Ynodes - measures summarizing the regimes that will be used on the right hand side of `working.msm` (baseline covariates may also be used in the right hand side of `working.msm` and do not need to be included in `summary.measures`)
final.Ynodes	vector subset of Ynodes - used in MSM to pool over a set of outcome nodes
msm.weights	projection weights for the working MSM. If "empirical", weight by empirical proportions of rows matching each regime for each final.Ynode, with duplicate regimes given zero weight. If NULL, no weights. Or an array of user-supplied weights with dimensions c(n, num.regimes, num.final.Ynodes) or c(num.regimes, num.final.Ynodes).

Details

The estimates returned by `ltmle` are of a treatment specific mean, $E[Y_{\bar{a}}]$, the mean of the final treatment node, where all treatment nodes, A, are set to $\bar{a}$ (abar) and all censoring nodes C are

set to 1 (uncensored). The estimates returned by `ltmleMSM` are similar but are the parameters in a working marginal structural model.

`data` should be a data frame where the order of the columns corresponds to the time-ordering of the model.

- in censoring columns (Cnodes): factor with two levels: "censored" and "uncensored". The helper function `BinaryToCensoring` can be used to create these factors.
- in treatment columns (Anodes): 1 = treated, 0 = untreated (must be binary)
- in event columns (Ynodes): If `survivalOutcome` is TRUE, then Y nodes are treated as indicators of a one-time event. See details for `survivalOutocme`. If `survivalOutcome` is FALSE, Y nodes are treated as binary if all values are 0 or 1, and are treated as continuous otherwise. If Y nodes are continuous, they may be automatically scaled. See details for `Yrange`.
- time-dependent covariate columns (Lnodes): can be any numeric data
- Data in `Cnodes`, `Anodes`, `Lnodes` and `Ynodes` are not used after (to the right of) censoring (or an event when `survivalOutcome==TRUE`) and may be coded as NA or any other value.
- Columns in `data` that are before (to the left of) the first of `Cnodes` or `Anodes` are treated as baseline variables, even if they are specified as `Lnodes`.
- After the first of `Cnodes`, `Anodes`, `Ynodes`, or `Lnodes`, every column must be in one of `Cnodes`, `Anodes`, `Ynodes`, or `Lnodes`.

If `survivalOutcome` is TRUE, all Y values are indicators of an event (e.g. death) at or before the current time, where 1 = event and 0 = no event. The events in Ynodes must be of the form where once Y jumps to 1, Y remains 1 at subsequent nodes.

For continuous outcomes, (`survivalOutcome==FALSE` and some Y nodes are not 0 or 1,) Y values are truncated at the minimum and maximum of `Yrange` if specified, and then transformed and scaled to be in [0,1]. That is, transformed to `(Y-min(Yrange))/(max(Yrange)-min(Yrange))`. If `Yrange` is NULL, it is set to the range of all Y nodes. In that case, Y nodes are only scaled if any values fall outside of [0,1]. For intervention specific means (`ltmle`), parameter estimates are transformed back based `Yrange`.

`Qform` should be NULL, in which case all parent nodes of each L and Y node will be used as regressors, or a named character vector that can be coerced to class "[formula](#)". The length of `Qform` must be equal to `length(Lnodes) + length(Ynodes)**` and the names and order of the formulas must be the same as the names and order of the L and Y nodes in `data`. The left hand side of each formula should be "Q.kplus1". If `SL.library` is NULL, `glm` will be called using the elements of `Qform`. If `SL.library` is specified, [SuperLearner](#) will be called after a design matrix is created using `Qform`.

** If there is a "block" of L and Y nodes not separated by A or C nodes, only one regression is required at the first L/Y node in a block. You can pass regression formulas for the other L/Y nodes, but they will be ignored (with a message). See example 5.

`gform` should be NULL, in which case all parent nodes of each L and Y node will be used as regressors, or a character vector that can be coerced to class "[formula](#)", or a matrix/array of Prob(A=1). If `gform` is a character vector, the length of `gform` must be equal to `length(Anodes) + length(Cnodes)` and the order of the formulas must be the same as the order the A and C nodes appear in `data`. The left hand side of each formula should be the name of the Anode or Cnode. If `SL.library` is NULL, `glm` will be called using the elements of `gform`. If `SL.library` is specified, [SuperLearner](#) will be called after a design matrix is created using `gform`.

In ltmle, gform can also be a n x numACnodes matrix where entry (i, j) is the probability that the ith observation of the jth A/C node is 1 (if an Anode) or uncensored (if a Cnode), conditional on following abar up to that node. In ltmleMSM, gform can similarly be a n x numACnodes x numRegimes array, where entry (i, j, k) is the probability that the ith observation of the jth A/C node is 1 (if an Anode) or uncensored (if a Cnode), conditional on following regime k up to that node. If gform is a matrix/array, deterministic.g.function will not be used and should be NULL.

abar specifies the counterfactual values of the Anodes, using the order they appear in data and should have the same length (if abar is a vector) or number of columns (if abar is a matrix) as Anodes.

rule can be used to specify a dynamic treatment rule. rule is a function applied to each row of data which returns a numeric vector of the same length as Anodes.

abar and rule cannot both be specified. If one of them if a list of length 2, additive treatment effect, risk ratio, and odds ratio can be computed using summary.ltmleEffectMeasures.

regimes can be a binary array: n x numAnodes x numRegimes of counterfactual treatment or a list of 'rule' functions as described above for the rule argument for the ltmle function

deterministic.g.function can be a function used to specify model knowledge about value of Anodes and/or Cnodes that are set deterministically. For example, it may be the case that once a patient starts treatment, they always stay on treatment. For details on the form of the function and examples, see deterministic.g.function_template

deterministic.Q.function can be a function used to specify model knowledge about the final event state. For example, it may be the case that a patient can complete the study at some intermediate time point, in which case the probability of death is 0 (assuming they have not died already). For details on the form of the function and examples, see deterministic.Q.function_template

SL.library may be a character vector of libraries (or 'glm' or 'default'), in which case these libraries are used to estimate both Q and g OR a list with two components, Q and g, where each is a character vector of libraries (or 'glm' or 'default'). 'glm' indicates glm should be called instead of SuperLearner If SL.library is the string 'default', SL.library is set to list("SL.glm", "SL.stepAIC", "SL.bayesglm", c("SL.glm","screen.corP"), c("SL.step", "screen.corP"), c("SL.step.forward","screen.corP"), c("SL.stepAIC", "screen.corP"), c("SL.step.interaction","screen.corP"), c("SL.bayesglm", "screen.corP"). Note that the default set of libraries consists of main terms models. It may be advisable to include squared terms, interaction terms, etc in gform and Qform or include libraries that consider non-linear terms.

If attr(SL.library, "return.fit") == TRUE, then fit$g and fit$Q will return full SuperLearner or speedglm objects. If not, only a summary matrix will be returned to save memory.

The print method for ltmle objects only prints the tmle estimates.

Value

ltmle returns an object of class "ltmle" (unless abar or rule is a list, in which case it returns an object of class ltmleSummaryMeasures, which has the same components as ltmleMSM.) The function summary (i.e. summary.ltmle) can be used to obtain or print a summary of the results. An object of class "ltmle" is a list containing the following components:

estimates   a named vector of length 4 with elements, each an estimate of $E[Y_{bara}]$:

- tmle - Targeted Maximum Likelihood Estimate [NULL if gcomp is TRUE]
- iptw - Inverse Probability of Treatment/Censoring Weighted estimate

- gcomp - maximum likelihood based G-computation estimate [NULL if gcomp is FALSE]

IC   a list with the following components of Influence Curve values

- `tmle` - vector of influence curve values for Targeted Maximum Likelihood Estimate [NULL if gcomp is TRUE]

- `iptw` - vector of influence curve values for Inverse Probability of Treatment/Censoring Weighted estimate

- `gcomp` - vector of influence curve values for Targeted Maximum Likelihood Estimate without updating [NULL if gcomp is FALSE]

cum.g   cumulative g, after bounding: for ltmle, n x numACnodes, for ltmleMSM, n x numACnodes x num.regimes

cum.g.unbounded

                    cumulative g, before bounding: for ltmle, n x numACnodes, for ltmleMSM, n x numACnodes x num.regimes

cum.g.used   binary - TRUE if an entry of cum.g was used in the updating step (note: even if cum.g.used is FALSE, a small value of cum.g.unbounded may still indicate a positivity problem): for ltmle, n x numACnodes, for ltmleMSM, n x numACnodes x num.regimes

call   the matched call

gcomp   the gcomp input

formulas   a `list` with elements `Qform` and `gform`

fit   a list with the following components

- `g` - list of length numACnodes - glm or SuperLearner (see Details) return objects from fitting g regressions

- `Q` - list of length numLYnodes - glm or SuperLearner (see Details) return objects from fitting Q regressions

- `Qstar` - list of length numLYnodes - glm (or numerical optimization if glm fails to solve the score equation) return objects from updating the Q fit

ltmleMSM returns an object of class "ltmleMSM" The function summary (i.e. summary.ltmleMSM) can be used to obtain or print a summary of the results. An object of class "ltmleMSM" is a list containing the following components:

beta   parameter estimates for working.msm using TMLE (GCOMP if gcomp input is TRUE)

beta.iptw   parameter estimates for working.msm using IPTW

IC   matrix, n x numBetas - influence curve values for TMLE (without updating if gcomp input is TRUE)

IC.iptw   matrix, n x numBetas - influence curve values for IPTW

msm   object of class glm - the result of fitting the working.msm

cum.g   array, n x numACnodes x numRegimes - cumulative g, after bounding

cum.g.unbounded

        array, n x numACnodes x numRegimes - cumulative g, before bounding

call   the matched call

gcomp   the gcomp input

formulas   a `list` with elements `Qform` and `gform`

fit   a list with the following components

- g - list of length numRegimes of list of length numACnodes - glm or SuperLearner (see Details) return objects from fitting g regressions

- Q - list of length numLYnodes - glm or SuperLearner (see Details) return objects from fitting Q regressions

- Qstar - list of length numLYnodes - glm (or numerical optimization if glm fails to solve the score equation) return objects from updating the Q fit

Functions

- `ltmleMSM`: Longitudinal Targeted Maximum Likelihood Estimation for a Marginal Structural Model

Author(s)

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See Also

[summary.ltmle](#), [summary.ltmleMSM](#), [SuperLearner](#), [deterministic.g.function_template](#), [deterministic.Q.functi](#)

Examples

```
# See vignette for more examples.

rexpit <- function(x) rbinom(n=length(x), size=1, prob=plogis(x))

# Single time point Example
n <- 1000
W <- rnorm(n)
A <- rexpit(-1 + 2 * W)
Y <- rexpit(W + A)
data <- data.frame(W, A, Y)

result1 <- ltmle(data, Anodes="A", Ynodes="Y", abar=1)
summary(result1)
summary(result1, estimator="iptw")
# MSM Example
# Given data over 3 time points where A switches to 1 once and then stays 1. We want to know
# how death varies as a function of gender, time and an indicator of whether a patient's
# intended regime was to switch before time.
# Note that working.msm includes time and switch.time, which are columns of
```

```
# summary.measures; working.msm also includes male, which is ok because it is a baseline
# covariate (it comes before any A/C/L/Y nodes).
data(sampleDataForLtmleMSM)
Anodes <- grep("^A", names(sampleDataForLtmleMSM$data))
Lnodes <- c("CD4_1", "CD4_2")
Ynodes <- grep("^Y", names(sampleDataForLtmleMSM$data))
msm.weights <- matrix(1:12, nrow=4, ncol=3) #just an example (can also use a 200x3x4 array),
                                   #or NULL (for no weights), or "empirical" (the default)

result2 <- ltmleMSM(sampleDataForLtmleMSM$data, Anodes=Anodes, Lnodes=Lnodes, Ynodes=Ynodes,
                    survivalOutcome=TRUE,
                    regimes=sampleDataForLtmleMSM$regimes,
             summary.measures=sampleDataForLtmleMSM$summary.measures, final.Ynodes=Ynodes,
                    working.msm="Y ~ male + time + I(pmax(time - switch.time, 0))",
                    msm.weights=msm.weights, estimate.time=FALSE)
print(summary(result2))
```

---

sampleDataForLtmleMSM   *Sample data, regimes, and summary measures*

Description

Sample data for use with ltmleMSM. Data: n=1000: male age CD4_1 A1 Y1 CD4_2 A2 Y2 CD4_3 A3 Y3 A1..A3 are treatment nodes, Y1..Y3 are death, CD4_1..CD4_3 are time varying covariates. We are interested in static regimes where a patient switches at some time. In summary.measures, switch.time is first time where At is 1 (4 if never switch), time is the horizon.

Format

List with three components: data, regimes, summary.measures

Details

regimes: 200 x 3 x 4 [n x numACnodes x numRegimes] summary.measures: 4 x 2 x 3 [numRegimes x numSummaryMeasures x numFinalYnodes]

Source

simulated data

Examples

```
data(sampleDataForLtmleMSM)
```

---

summary.ltmle	*Get standard error, p-value, and confidence interval for one ltmle object Summarizing results from Longitudinal Targeted Maximum Likelihood Estimation (ltmle)*

Description

These functions are methods for class ltmle or summary.ltmle objects.

Usage

```
## S3 method for class 'ltmle'
summary(object, estimator = ifelse(object$gcomp, "gcomp",
  "tmle"), ...)

## S3 method for class 'ltmleEffectMeasures'
summary(object,
  estimator = ifelse(object$gcomp, "gcomp", "tmle"), ...)

## S3 method for class 'ltmleMSM'
summary(object, estimator = ifelse(object$gcomp,
  "gcomp", "tmle"), ...)

## S3 method for class 'summary.ltmleMSM'
print(x, digits = max(3, getOption("digits") -
  3), signif.stars = getOption("show.signif.stars"), ...)

## S3 method for class 'summary.ltmle'
print(x, ...)

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

## S3 method for class 'summary.ltmleEffectMeasures'
print(x, ...)

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

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

Arguments

object an object of class "ltmle" or "ltmleMSM" or "ltmleEffectMeasures", usually a result of a call to [ltmle] or [ltmleMSM].

estimator	character; one of "tmle", "iptw", "gcomp". The estimator for which to get effect measures. "tmle" is valid iff the original ltmle/ltmleMSM call used gcomp=FALSE. "gcomp" is valid iff the original ltmle/ltmleMSM call used gcomp=TRUE
...	further arguments passed to or from other methods.
x	an object of class "summary.ltmle" or "summary.ltmleMSM" or "ltmleEffectMeasures", usually a result of a call to summary.ltmle or summary.ltmleMSM.
digits	the number of significant digits to use when printing.
signif.stars	logical. If TRUE, significance stars are printed for each coefficient.

Details

summary.ltmle returns the parameter value of the estimator, the estimated variance, a 95 percent confidence interval, and a p-value.

summary.ltmleEffectMeasures returns the additive treatment effect for each of the two objects in the abar list passed to ltmle. Relative risk, and odds ratio are also returned, along with the variance, confidence interval, and p-value for each.

summary.ltmleMSM returns a matrix of MSM parameter estimates.

Value

summary.ltmle returns an object of class "summary.ltmle", a list with components

treatment	a list with components summarizing the estimate of object
	• estimate - the parameter estimate of $E[Y_d]$
	• std.dev - estimated standard deviation of parameter
	• p.value - two-sided p-value
	• CI - vector of length 2 with 95 percent confidence interval
call	the matched call to ltmle for object
estimator	the estimator input argument
variance.estimate.ratio	
	ratio of the TMLE based variance estimate to the influence curve based variance estimate

summary.ltmleEffectMeasures returns an object of class "summary.ltmleEffectMeasures", a list with same components as summary.ltmle above, but also includes:

effect.measures

a list with components, each with the same components as treatment in summary.ltmle above

- treatment - corresponds to the first in the list abar (or rule) passed to ltmle
- control - corresponds to the second in the list abar (or rule) passed to ltmle
- ATE - average treatment effect
- RR - relative risk
- OR - odds ratio

summary.ltmleMSM returns an object of class "summary.ltmleMSM", a matrix with rows for each MSM parameter and columns for the point estimate, standard error, 2.5percent confidence interval, 97.5percent confidence interval, and p-value.

See Also

ltmle, summary

Examples

```
rexpit <- function(x) rbinom(n = length(x), size = 1, prob = plogis(x))

# Compare the expected outcomes under two counterfactual plans: Treatment plan:
# set A1 to 1 if W > 0, set A2 to 1 if W > 1.5, always set A3 to 1 Control plan:
# always set A1, A2, and A3 to 0
W <- rnorm(1000)
A1 <- rexpit(W)
A2 <- rexpit(W + 2 * A1)
A3 <- rexpit(2 * A1 - A2)
Y <- rexpit(W - A1 + 0.5 * A2 + 2 * A3)
data <- data.frame(W, A1, A2, A3, Y)
treatment <- cbind(W > 0, W > 1.5, 1)
control <- matrix(0, nrow = 1000, ncol = 3)
result <- ltmle(data, Anodes = c("A1", "A2", "A3"), Ynodes = "Y", abar = list(treatment,
    control))
print(summary(result))

## For examples of summary.ltmle and summary.ltmleMSM, see example(ltmle)
```

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