

Package ‘TTE’

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Description Design and analysis tools for target trial emulation using longitudinal observational data. Functions are provided for checking person-period data, expanding longitudinal data into sequentially nested trials, estimating inverse probability weights for intention-to-treat and per-protocol analyses, and assessing weight distributions and covariate balance. Additional functions fit weighted pooled discrete-time outcome models, obtain standardized risks and treatment contrasts, and estimate weighted Kaplan-Meier and Aalen-Johansen curves. Two worked examples based on fully synthetic data illustrate an active-comparator new-user study comparing sodium-glucose cotransporter 2 inhibitors with dipeptidyl peptidase-4 inhibitors and an analysis of sequentially nested trials comparing angiotensin receptor blocker and calcium channel blocker strategies.

URL <https://github.com/nomahi/TTE/>

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TTE-package	<i>Design and Analysis Tools for Target Trial Emulation</i>
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Description

Design and analysis tools for target trial emulation using longitudinal observational data.

Details

The main workflow comprises the following steps:

- check person-period data with `check_tte()`;
- expand longitudinal data into sequentially nested trials with `seqdesign_tte()`;
- estimate and combine inverse probability weights with `est_wt()` and `combine_wt()`;
- assess covariate balance and diagnose weight distributions with `balance_wt()` and `diagnose_wt()`;
- fit weighted pooled discrete-time outcome models with `discsurvreg()`;
- obtain standardized risks and treatment contrasts with `std_tte()`; and
- estimate and plot weighted Kaplan-Meier and Aalen-Johansen curves with `curve_tte()`.

The package includes two pairs of fully synthetic datasets:

- SGLT2 and SGLT2_baseline; and
- ARB and ARB_baseline.

No records from the motivating studies are included.

Author(s)

Hisashi Noma

References

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- Noma, H., Kurita, N., Fukuma, S., Fujisawa, T., Oda, F., Maeda, M., and Fukuda, H. (2026). Heart failure and renal outcomes with angiotensin receptor blockers compared with calcium channel blockers in patients with chronic kidney disease: a target trial emulation. *Heart*. doi:10.1136/heartjnl2026328193.

Examples

```
library(TTE)
data(SGLT2)

fit <- discsurvreg(
  Y_death ~ A + splines::ns(time, df = 3) + trial_period,
  data = SGLT2,
  id = id,
  weights = w_itt,
  var_method = "standard"
)
fit
```

 ARB

Synthetic ARB versus CCB Sequential-Trial Example

Description

Fully synthetic baseline and person-month data representing sequentially nested new-user trials comparing angiotensin receptor blocker (ARB) and calcium channel blocker (CCB) strategies among people with chronic kidney disease. Heart failure hospitalization is the primary endpoint, and death is a competing event.

Usage

```
data(ARB_baseline)
data(ARB)
```

Format

ARB_baseline is a data frame with 900 person-trial entries and the following variables:

id Original individual identifier.

trial Identifier for the sequential trial in which the individual entered.

A Binary treatment-strategy indicator: 0 for CCB and 1 for ARB.

treatment Treatment-strategy label: "CCB" or "ARB".

age Age at trial baseline, in years.

female Indicator for female sex: 0 for no and 1 for yes.

bmi Body mass index at trial baseline, in kg/m².

sbp Systolic blood pressure at trial baseline, in mm Hg.

dbp Diastolic blood pressure at trial baseline, in mm Hg.

egfr Estimated glomerular filtration rate at trial baseline, in mL/min/1.73 m².

proteinuria Indicator for proteinuria at trial baseline: 0 for no and 1 for yes.

diabetes Indicator for diabetes at trial baseline: 0 for no and 1 for yes.

prior_heart_failure Indicator for a history of heart failure at trial baseline: 0 for no and 1 for yes.

prior_stroke Indicator for a history of stroke at trial baseline: 0 for no and 1 for yes.

trial_period Calendar-period category for the trial baseline.

ps_oracle Probability of ARB initiation generated by the simulation treatment model.

w_iptw Stabilized baseline inverse probability of treatment weight.

ARB is a person-month data frame containing the variables in ARB_baseline, except ps_oracle, together with the following variables:

time Zero-based follow-up-month index; time = 0 denotes the first month after trial baseline.

Y_hf Indicator of heart failure hospitalization during the interval: 0 for no and 1 for yes.

Y_death Indicator of death during the interval: 0 for no and 1 for yes.

event_code Competing-event code: 0 for no event, 1 for heart failure hospitalization, and 2 for death.

stay_ltfu Interval-specific indicator of remaining uncensored with respect to loss to follow-up: 0 for no and 1 for yes.

stay_adherent Interval-specific indicator of adherence to the baseline treatment strategy: 0 for no and 1 for yes.

pp_at_risk Indicator that the person-trial remains in the per-protocol risk set: 0 for no and 1 for yes.

w_itt Combined analysis weight for the intention-to-treat analysis.

w_pp Combined analysis weight for the per-protocol analysis.

In ARB, sbp, dbp, and egfr are updated over follow-up; the other clinical covariates are carried forward from trial baseline.

Details

The datasets were independently simulated and include no participant records from the motivating study. They are intended solely for education, software testing, and reproducible examples; numerical results obtained from them have no clinical interpretation.

References

Noma, H., Kurita, N., Fukuma, S., Fujisawa, T., Oda, F., Maeda, M., and Fukuda, H. (2026). Heart failure and renal outcomes with angiotensin receptor blockers compared with calcium channel blockers in patients with chronic kidney disease: a target trial emulation. *Heart*. doi:10.1136/heartjnl2026328193.

Examples

```
data(ARB_baseline)
data(ARB)
head(ARB_baseline)
table(ARB$event_code)
```

balance_wt	<i>Assess Covariate Balance Before and After Weighting</i>
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Description

Computes unweighted and weighted means, standard deviations, and standardized mean differences using externally supplied analysis weights.

Usage

```
balance_wt(formula, data, weights = NULL, absolute = TRUE)
```

Arguments

- formula A formula with a binary treatment variable on the left-hand side and baseline covariates on the right-hand side.
- data A data frame containing the variables in formula.
- weights Optional numeric analysis weights, the name of a weight column in data, or an object returned by est_wt() or combine_wt().
- absolute Logical; whether to report standardized mean differences in absolute value.

Value

An object of class balance_wt containing unweighted and weighted covariate summaries and standardized mean differences. A summary() method provides a formatted balance table.

See Also

```
est_wt, diagnose_wt
```

Examples

```
data(SGLT2_baseline)
wt <- est_wt(
  A ~ age + female + bmi + hba1c + egfr + prior_heart_failure +
    prior_stroke + recent_hospitalization + trial_period,
  data = SGLT2_baseline, type = "treatment"
)
bal <- balance_wt(
  A ~ age + female + bmi + hba1c + egfr + prior_heart_failure +
    prior_stroke + recent_hospitalization + trial_period,
  data = SGLT2_baseline, weights = wt
)
summary(bal)
```

boot_tte	<i>Individual-Cluster Bootstrap</i>
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Description

Resamples original individuals while retaining all person-period rows and repeated trial entries belonging to each sampled individual.

Usage

```
boot_tte(data, id, statistic, R = 500L,
  conf_level = 0.95, seed = NULL, ...)
```

Arguments

data	Analysis data containing all rows required by statistic.
id	Original individual identifier used to define bootstrap clusters.
statistic	A function whose first argument is a bootstrap data frame and whose return value is a named numeric vector.
R	Number of bootstrap replicates.
conf_level	Confidence level for percentile intervals.
seed	Optional random-number seed.
...	Additional arguments passed to statistic.

Details

Sampling is performed at the level of the original individual. Consequently, all follow-up rows and all sequential-trial entries associated with a sampled individual are retained together.

Value

An object of class `boot_tte` containing the estimate from the original data, bootstrap replicates, and percentile confidence intervals.

Examples

```

data(SGLT2)
small <- subset(SGLT2, id <= 80 & time < 24)
stat_fun <- function(d) {
  fit <- discsurvreg(
    Y_death ~ A + time, data = d,
    id = id, weights = w_itt,
    var_method = "standard"
  )
  c(log_hr = unname(coef(fit)["A"]))
}
b <- boot_tte(small, id = id, statistic = stat_fun, R = 10, seed = 1)
b

```

check_tte

*Check a Target-Trial Person-Period Dataset***Description**

Detects duplicate intervals, time-ordering problems, rows occurring after an event, within-trial treatment changes, and invalid analysis weights.

Usage

```

check_tte(data, id, time, trial = NULL, event = NULL,
  treatment = NULL, weights = NULL, expected_start = 0,
  allow_gaps = FALSE)

```

Arguments

data	A person-period data frame.
id	Original individual identifier.
time	Integer-valued time index since trial baseline.
trial	Optional trial identifier.
event	Optional binary event indicator or competing-event code.
treatment	Optional baseline treatment variable.
weights	Optional numeric analysis weights or the name of a weight column in data.
expected_start	Expected first value of the time index within each person-trial.
allow_gaps	Logical; whether gaps in integer time indices are allowed.

Value

An object of class `check_tte` containing the detected data-quality problems and the rows or person-trials affected by each problem.

See Also[seqdesign_tte](#)**Examples**

```
data(ARB)
chk <- check_tte(
  ARB,
  id = id, trial = trial, time = time,
  event = event_code, treatment = A, weights = w_itt
)
chk
```

combine_wt

*Combine Inverse Probability Weight Components***Description**

Multiplies treatment, loss-to-follow-up, and adherence weight components and optionally truncates and normalizes the resulting analysis weights.

Usage

```
combine_wt(..., truncate = c(0.01, 0.99),
  normalize = c("none", "mean1", "sum_n"))
```

Arguments

...	Numeric vectors or objects returned by <code>est_wt()</code> or <code>combine_wt()</code> . All components must have equal length.
truncate	Lower and upper quantiles used to truncate the product. The default truncates at the 1st and 99th percentiles.
normalize	Normalization applied after truncation: "none" for no normalization, "mean1" for a mean of 1, or "sum_n" for a sum equal to the number of observations.

Value

An object of class `combine_wt` containing the combined weights and metadata describing truncation and normalization. Use `weights()` to extract the numeric weight vector.

See Also[est_wt](#), [diagnose_wt](#)

Examples

```
data(ARB)
w <- combine_wt(
  ARB$w_iptw,
  ARB$w_itt / ARB$w_iptw,
  normalize = "mean1"
)
summary(weights(w))
```

curve_tte

Estimate Weighted Kaplan-Meier or Aalen-Johansen Curves

Description

Estimates weighted survival or cumulative-incidence curves from person-period risk sets. Curves are represented as right-continuous step functions.

Usage

```
curve_tte(data, time, event, treatment, weights = NULL,
  type = c("km", "aj"), cause = 1, id = NULL, trial = NULL,
  labels = NULL, bootstrap = 0L, conf_level = 0.95,
  seed = NULL)
```

Arguments

data	Person-period data with one at-risk row per interval.
time	Follow-up interval index.
event	Binary event indicator or competing-event code.
treatment	Treatment variable defining the curves to be compared.
weights	Optional numeric analysis weights, the name of a weight column in data, or an object returned by <code>est_wt()</code> or <code>combine_wt()</code> .
type	Curve type: "km" for Kaplan-Meier survival or "aj" for Aalen-Johansen cumulative incidence.
cause	Event code of interest for Aalen-Johansen estimation.
id	Original individual identifier, required when <code>bootstrap > 0</code> .
trial	Optional trial identifier used together with <code>id</code> to define person-trial records.
labels	Optional display labels for the treatment groups.
bootstrap	Number of individual-cluster bootstrap replicates.
conf_level	Confidence level for percentile intervals.
seed	Optional random-number seed.

Details

The time variable indexes follow-up intervals, whereas summary and plotting times are expressed as elapsed follow-up time. For example, interval indices 0, ..., 59 represent the 60 intervals ending at elapsed times 1, ..., 60. Thus, `summary(x, time = 60)` reports the estimate after 60 follow-up intervals.

Value

An object of class `curve_tte` containing treatment-specific curve estimates and, when requested, bootstrap confidence intervals. A `summary()` method returns estimates at specified elapsed follow-up times.

See Also

[std_tte](#), [boot_tte](#)

Examples

```
data(SGLT2)
km <- curve_tte(
  SGLT2, time = time, event = Y_death,
  treatment = treatment, weights = w_itt,
  type = "km", id = id, trial = trial
)
summary(km, time = 60)

data(ARB)
aj <- curve_tte(
  ARB, time = time, event = event_code,
  treatment = treatment, weights = w_itt,
  type = "aj", cause = 1, id = id, trial = trial
)
summary(aj, time = 60)
```

diagnose_wt

Diagnose Inverse Probability Weights

Description

Summarizes weight distributions and effective sample sizes. When analysis data are supplied, treatment-group and risk-set diagnostics are also produced.

Usage

```
diagnose_wt(weights, data = NULL, treatment = NULL, time = NULL,
  id = NULL, trial = NULL, extreme = c(0.01, 0.99))
```

Arguments

<code>weights</code>	Numeric analysis weights, the name of a weight column in data, or an object returned by <code>est_wt()</code> or <code>combine_wt()</code> .
<code>data</code>	Optional data frame containing variables used for stratified or risk-set diagnostics.
<code>treatment</code>	Optional treatment variable.
<code>time</code>	Optional follow-up interval index.
<code>id</code>	Optional original individual identifier.
<code>trial</code>	Optional trial identifier.
<code>extreme</code>	Lower and upper quantiles used to flag extreme weights.

Value

An object of class `diagnose_wt` containing distributional summaries, effective sample sizes, and any requested treatment-group or risk-set diagnostics.

See Also

[est_wt](#), [combine_wt](#), [balance_wt](#)

Examples

```
data(ARB)
diag <- diagnose_wt(
  w_itt, data = ARB,
  treatment = A, time = time, id = id, trial = trial
)
diag
```

discsurvreg

Fit a Weighted Pooled Discrete-Time Outcome Model

Description

Fits a weighted pooled generalized linear model and computes cluster-robust inference with clustering at the level of the original individual.

Usage

```
discsurvreg(formula, data, id, weights = NULL,
  family = stats::quasibinomial(link = "cloglog"),
  var_method = c("standard", "MBN"), eform = TRUE,
  conf_level = 0.95)
```

Arguments

<code>formula</code>	Model formula for the interval-specific outcome probability.
<code>data</code>	Person-period data.
<code>id</code>	Original individual identifier used to define variance-estimation clusters.
<code>weights</code>	Optional numeric analysis weights, the name of a weight column in data, or an object returned by <code>est_wt()</code> or <code>combine_wt()</code> .
<code>family</code>	A generalized linear model family. The default is a quasibinomial model with complementary log-log link.
<code>var_method</code>	Cluster-robust variance estimator: "standard" for the ordinary sandwich estimator or "MBN" for the Morel-Bokossa- Neerchal finite-sample correction.
<code>eform</code>	Logical; whether the default printed coefficient table reports exponentiated estimates.
<code>conf_level</code>	Confidence level used in printed and interval estimates.

Value

An object of class `discsurvreg` containing the fitted `glm` object, the cluster-robust covariance matrix, model coefficients, and analysis settings. Methods for `coef()`, `vcov()`, and `confint()` are available.

References

Morel, J. G., Bokossa, M. C., and Neerchal, N. K. (2003). Small sample correction for the variance of GEE estimators. *Biometrical Journal* **45**, 395–409. doi:[10.1002/bimj.200390021](https://doi.org/10.1002/bimj.200390021).

See Also

[std_tte](#)

Examples

```
data(SGLT2)
fit <- discsurvreg(
  Y_death ~ A + splines::ns(time, df = 3) + trial_period,
  data = SGLT2,
  id = id, weights = w_itt,
  family = stats::quasibinomial(link = "cloglog"),
  var_method = "standard"
)
fit
confint(fit, parm = "A", eform = TRUE)
```

est_wt

*Estimate Inverse Probability Weights***Description**

Estimates baseline treatment weights or cumulative censoring and adherence weights while retaining intermediate predicted probabilities and interval-specific weight factors.

Usage

```
est_wt(formula, data, numerator = NULL,
       type = c("treatment", "censoring", "adherence"),
       id = NULL, trial = NULL, time = NULL, cumulative = NULL,
       lag = NULL, stabilize = TRUE, truncate = c(0.01, 0.99),
       eps = 1e-06)
```

Arguments

formula	Denominator probability model.
data	A data frame containing the model variables.
numerator	Optional numerator probability model used to construct stabilized weights.
type	Weight type: "treatment", "censoring", or "adherence".
id	Original individual identifier, required for cumulative weights.
trial	Optional trial identifier used together with id to define person-trial records.
time	Follow-up interval index used to order rows before cumulative multiplication.
cumulative	Logical; whether interval-specific factors are multiplied within person-trial. When NULL, the setting is determined by type.
lag	Number of intervals by which cumulative weights are lagged before being assigned to outcome risk sets.
stabilize	Logical; whether numerator probabilities are used to construct stabilized weights.
truncate	Lower and upper quantiles used to truncate the final weights. The default truncates at the 1st and 99th percentiles.
eps	Lower probability bound used to avoid division by zero; predicted probabilities are bounded away from 0 and 1 by this amount.

Details

For type = "treatment", the function estimates a baseline inverse probability of treatment weight. For type = "censoring" or type = "adherence", interval-specific factors may be multiplied within person-trial to obtain cumulative weights. Weight truncation is an analysis choice; the default applies truncation at the 1st and 99th percentiles.

Value

An object of class `est_wt` containing the estimated weight vector, predicted numerator and denominator probabilities, interval-specific factors, and settings used in weight construction. The numeric weights can be extracted with `weights()`, and a `summary()` method provides diagnostics.

See Also

[combine_wt](#), [diagnose_wt](#), [balance_wt](#)

Examples

```
data(SGLT2_baseline)
wt <- est_wt(
  A ~ age + female + bmi + hba1c + egfr +
    prior_heart_failure + prior_stroke +
    recent_hospitalization + trial_period,
  data = SGLT2_baseline,
  type = "treatment"
)
summary(wt)

data(ARB)
wc <- est_wt(
  stay_ltfu ~ A + time + I(time^2) + age + female + egfr +
    prior_heart_failure,
  numerator = stay_ltfu ~ A + time + I(time^2),
  data = ARB,
  type = "censoring", id = id, trial = trial, time = time,
  cumulative = TRUE, lag = 1
)
summary(wc)
```

seqdesign_tte

Expand Longitudinal Data into Sequentially Nested Trials

Description

Creates a trial entry at each eligible treatment decision time and expands each entry into follow-up person-period records.

Usage

```
seqdesign_tte(data, id, calendar_time, eligible, treatment,
  outcome, censor = NULL, max_follow = Inf, induction = 0L,
  keep = names(data))
```

Arguments

<code>data</code>	Longitudinal source data with one row per individual and calendar interval.
<code>id</code>	Original individual identifier.
<code>calendar_time</code>	Ordered calendar-time variable.
<code>eligible</code>	Logical eligibility indicator or expression evaluated at each potential trial baseline.
<code>treatment</code>	Treatment strategy observed at the eligible trial baseline.
<code>outcome</code>	Binary outcome indicator.
<code>censor</code>	Optional censoring indicator.
<code>max_follow</code>	Maximum number of retained follow-up intervals, including the interval indexed by time 0.
<code>induction</code>	Number of initial follow-up intervals excluded from outcome ascertainment.
<code>keep</code>	Names of source variables to retain in the expanded data.

Details

An individual may enter more than one sequential trial when eligible at multiple treatment decision times. Follow-up for each trial entry is obtained from subsequent rows of the original longitudinal data and ends according to the supplied outcome, censoring, and maximum-follow-up definitions.

Value

A person-period data frame with added `trial`, `time`, `A`, and `Y` variables, together with the retained source variables and indicators used to define follow-up.

See Also

[check_tte](#)

Examples

```
data(SGLT2)
source_data <- subset(SGLT2, id <= 8)
source_data$eligible <- source_data$time == 0
expanded <- seqdesign_tte(
  source_data, id = id, calendar_time = time,
  eligible = eligible, treatment = A, outcome = Y_death,
  max_follow = 24, induction = 1
)
head(expanded)
```

SGLT2

*Synthetic SGLT2i versus DPP-4i Target-Trial Example***Description**

Fully synthetic baseline and person-month data representing an active-comparator new-user target trial emulation comparing sodium-glucose cotransporter 2 inhibitors (SGLT2i) with dipeptidyl peptidase-4 inhibitors (DPP-4i) among older adults with type 2 diabetes. The primary endpoint is all-cause death after a one-month induction period.

Usage

```
data(SGLT2_baseline)
data(SGLT2)
```

Format

SGLT2_baseline is a data frame with 700 trial entries and the following variables:

`id` Original individual identifier.

`trial` Trial identifier.

`A` Binary treatment-strategy indicator: 0 for DPP-4i and 1 for SGLT2i.

`treatment` Treatment-strategy label: "DPP-4i" or "SGLT2i".

`age` Age at trial baseline, in years.

`female` Indicator for female sex: 0 for no and 1 for yes.

`bmi` Body mass index at trial baseline, in kg/m².

`hba1c` Glycated hemoglobin at trial baseline, in percent.

`egfr` Estimated glomerular filtration rate at trial baseline, in mL/min/1.73 m².

`proteinuria` Indicator for proteinuria at trial baseline: 0 for no and 1 for yes.

`prior_heart_failure` Indicator for a history of heart failure at trial baseline: 0 for no and 1 for yes.

`prior_stroke` Indicator for a history of stroke at trial baseline: 0 for no and 1 for yes.

`recent_hospitalization` Indicator for a recent hospitalization at trial baseline: 0 for no and 1 for yes.

`trial_period` Calendar-period category for the trial baseline.

`ps_oracle` Probability of SGLT2i initiation generated by the simulation treatment model.

`w_iptw` Stabilized baseline inverse probability of treatment weight.

SGLT2 is a person-month data frame containing the variables in SGLT2_baseline, except `ps_oracle`, together with the following variables:

`time` Zero-based follow-up-month index; `time = 0` denotes the first month after trial baseline.

`Y_death` Indicator of death during the interval: 0 for no and 1 for yes.

event_code Event code: 0 for no event and 1 for death.

stay_ltfu Interval-specific indicator of remaining uncensored with respect to loss to follow-up: 0 for no and 1 for yes.

stay_adherent Interval-specific indicator of adherence to the baseline treatment strategy: 0 for no and 1 for yes.

pp_at_risk Indicator that the trial entry remains in the per-protocol risk set: 0 for no and 1 for yes.

w_itt Combined analysis weight for the intention-to-treat analysis.

w_pp Combined analysis weight for the per-protocol analysis.

The baseline covariates are repeated across follow-up months in SGLT2.

Details

The datasets were independently simulated and include no participant records from the motivating study. They are intended solely for education, software testing, and reproducible examples; numerical results obtained from them have no clinical interpretation.

References

Noma, H., Goto, A., Sugimoto, T., Sunada, H., Oda, F., Maeda, M., and Fukuda, H. (2026). Real-world effectiveness of SGLT2 inhibitors in adults aged 75 years or older: a target trial emulation. *Age and Ageing* **55**, afag246.

Examples

```
data(SGLT2_baseline)
data(SGLT2)
head(SGLT2_baseline)
head(SGLT2)
```

std_tte	<i>Standardize a Discrete-Time Model to Marginal Risks</i>
---------	--

Description

Creates counterfactual copies of a target population, predicts interval-specific event probabilities, and obtains standardized survival, risk, cumulative incidence, risk differences, risk ratios, and time-specific numbers needed to treat or harm.

Usage

```
std_tte(fit, data, treatment, time, times,
        values = NULL, labels = NULL, competing_fit = NULL,
        target_weights = NULL)
```

Arguments

<code>fit</code>	A <code>discsurvreg</code> object for the event of interest.
<code>data</code>	Target population over which predictions are standardized.
<code>treatment</code>	Treatment variable used in the fitted model.
<code>time</code>	Follow-up interval index used in the fitted model.
<code>times</code>	Ordered interval-index values over which predictions are computed.
<code>values</code>	Treatment values defining the counterfactual strategies.
<code>labels</code>	Optional display labels for the treatment strategies.
<code>competing_fit</code>	Optional <code>discsurvreg</code> object for a competing event.
<code>target_weights</code>	Optional nonnegative weights used to average predictions over the target population.

Details

The values supplied in `times` index follow-up intervals, whereas summary horizons are expressed as elapsed follow-up time. For example, `times = 0:59` represents the 60 intervals ending at elapsed times 1 through 60, so `summary(x, horizon = 60)` reports standardized estimates after 60 follow-up intervals.

When `competing_fit` is omitted, risks are obtained from the fitted event model. When a competing-event model is supplied, the function combines the event-specific probabilities to obtain standardized cumulative incidence.

Value

An object of class `std_tte` containing strategy-specific standardized curves and treatment contrasts. A `summary()` method returns estimates at a specified elapsed follow-up horizon.

See Also

[discsurvreg](#), [curve_tte](#)

Examples

```
data(SGLT2)
data(SGLT2_baseline)
fit <- discsurvreg(
  Y_death ~ A + splines::ns(time, df = 3) + trial_period,
  data = SGLT2, id = id, weights = w_itt,
  var_method = "standard"
)
std <- std_tte(
  fit, data = SGLT2_baseline,
  treatment = A, time = time, times = 0:59,
  labels = c("DPP-4i", "SGLT2i")
)
summary(std, horizon = 60)
```

```
# Competing-risk example
data(ARB)
data(ARB_baseline)
fit_hf <- discsurvreg(
  Y_hf ~ A + splines::ns(time, df = 3) + trial_period,
  data = ARB, id = id, weights = w_itt,
  var_method = "standard"
)
fit_death <- discsurvreg(
  Y_death ~ A + splines::ns(time, df = 3) + trial_period,
  data = ARB, id = id, weights = w_itt,
  var_method = "standard"
)
std_hf <- std_tte(
  fit_hf, data = ARB_baseline,
  treatment = A, time = time, times = 0:59,
  competing_fit = fit_death, labels = c("CCB", "ARB")
)
summary(std_hf, horizon = 60)
```

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