

Package ‘nph’

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Title Planning and Analysing Survival Studies under Non-Proportional Hazards

Version 2.1

Description

Piecewise constant hazard functions are used to flexibly model survival distributions with non-proportional hazards and to simulate data from the specified distributions. A function to calculate weighted log-rank tests for the comparison of two hazard functions is included. Also, a function to calculate a test using the maximum of a set of test statistics from weighted log-rank tests (MaxCombo test) is provided. This test utilizes the asymptotic multivariate normal joint distribution of the separate test statistics. The correlation is estimated from the data. These methods are described in Ristl et al. (2021) [doi:10.1002/pst.2062](https://doi.org/10.1002/pst.2062).

Finally, a function is provided for the estimation and inferential statistics of various parameters that quantify the difference between two survival curves. Eligible parameters are differences in survival probabilities, log survival probabilities, complementary log log (cloglog) transformed survival probabilities, quantiles of the survival functions, log transformed quantiles, restricted mean survival times, as well as an average hazard ratio, the Cox model score statistic (logrank statistic), and the Cox-model hazard ratio. Adjustments for multiple testing and simultaneous confidence intervals are calculated using a multivariate normal approximation to the set of selected parameters.

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logrank.maxtest	<i>Maximum combination (MaxCombo) log-rank test</i>
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Description

Calculates a MaxCombo test for the comparison of two groups based on the maximum of test statistics of a set of weighted log-rank tests

Usage

```
logrank.maxtest(
  time,
  event,
  group,
  alternative = c("two.sided", "less", "greater"),
  rho = c(0, 0, 1),
  gamma = c(0, 1, 0),
  event_time_weights = NULL,
  algorithm = mvtnorm::GenzBretz(maxpts = 50000, abseps = 1e-05, releps = 0)
)
```

Arguments

time	Vector of observed event/censored times
event	logical vector or numeric vector with entries 0 or 1, indicating if an event was observed (TRUE or 1) or the time is censored (FALSE or 0)
group	Vector of group allocations
alternative	Either of "two.sided", "less" or "greater", specifies if two-sided or respective one-sided p-values are calculated. In any case the z test statistic of each included weighted log-rank test is based on the (weighted) sum of expected minus observed events in the group corresponding to the first factor level of group. Hence a small value of the test statistic corresponds to a lower (weighted average) hazard rate in the first group.
rho	Vector of parameter values rho for a set of weighting functions in the rho-gamma family
gamma	Vector of parameter values gamma for a set of weighting functions in the rho-gamma family
event_time_weights	Optional matrix, each column containing a different weighting vector for the event times. These weight vectors need to have one entry per event time (not per event, as multiple events may occur at the same time) and must be sorted by increasing event time.
algorithm	algorithm for the multivariate normal integration to be used in pmvnorm .

Details

To perform a maximum-type combination test, a set of m different weight functions $w_1(t), \dots, w_m(t)$ is specified and the correspondingly weighted logrank statistics z_1, \dots, z_m are calculated. The maximum test statistic is $z_{max} = \max_{i=1, \dots, m} z_i$. If at least one of the selected weight functions results in high power, we may expect a large value of z_{max} . Under the least favorable configuration in H_0 , approximately $(Z_1, \dots, Z_m) \sim N_m(0, \Sigma)$. The p-value of the maximum test, $P_{H_0}(Z_{max} > z_{max}) = 1 - P(Z_1 \leq z_{max}, \dots, Z_m \leq z_{max})$, is calculated based on this multivariate normal approximation via numeric integration. The integration is done using [pmvnorm](#). The default settings in `logrank.maxtest` correspond to greater precision than the original default of [pmvnorm](#). Precision can be set via the argument `algorithm`. Lower precision settings may speed up calculation.

The multivariate normal approach automatically corrects for multiple testing with different weights and does so efficiently since the correlation between the different tests is incorporated in Σ . For actual calculations, Σ is replaced by an estimate.

Value

A list with elements:

`pmult` The two sided p-value for the null hypothesis of equal hazard functions in both groups, based on the multivariate normal approximation for the z-statistics of differently weighted log-rank tests.

`p.Bonf` The two sided p-value for the null hypothesis of equal hazard functions in both groups, based on a Bonferroni multiplicity adjustment for differently weighted log-rank tests.

`tests` Data frame with z-statistics and two-sided unadjusted p-values of the individual weighted log-rank tests

`korr` Estimated correlation matrix for the z-statistics of the differently weighted log-rank tests.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schöler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

Pranab Ghosh, Robin Ristl, Franz König, Martin Posch, Christopher Jennison, Heiko Götte, Armin Schöler, Cyrus Mehta. Robust group sequential designs for trials with survival endpoints and delayed response. *Biometrical Journal*. First published online: 21 December 2021

Tarone RE. On the distribution of the maximum of the logrank statistic and the modified wilcoxon statistic. *Biometrics*. 1981; 37:79-85.

Lee S-H. On the versatility of the combination of the weighted log-rank statistics. *Comput Stat Data Anal*. 2007; 51(12):6557-6564.

Karrison TG et al. Versatile tests for comparing survival curves based on weighted log-rank statistics. *Stata J*. 2016; 16(3):678-690.

See Also

[logrank.test](#)

Examples

```
A <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
```

```

B <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.1, 0.6, 0.1), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.04, 0.04), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
dat <- sample_fun(A, B, r0 = 0.5, eventEnd = 30,
  lambdaRecr = 0.5, lambdaCens = 0.25 / 365,
  maxRecrCalendarTime = 2 * 365,
  maxCalendar = 4 * 365)
logrank.maxtest(dat$y, dat$event, dat$group)

```

logrank.test	<i>Weighted log-rank test</i>
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Description

Calculates a weighted log-rank test for the comparison of two groups.

Usage

```

logrank.test(
  time,
  event,
  group,
  alternative = c("two.sided", "less", "greater"),
  rho = 0,
  gamma = 0,
  event_time_weights = NULL
)

```

Arguments

time	Vector of observed event/censored times
event	logical vector or numeric vector with entries 0 or 1, indicating if an event was observed (TRUE or 1) or the time is censored (FALSE or 0)
group	Vector of group allocations
alternative	Either of "two.sided", "less" or "greater", specifies if two-sided or respective one-sided p-values are calculated. In any case the z test statistic of each included weighted log-rank test is based on the (weighted) sum of expected minus observed events in the group corresponding to the first factor level of group. Hence a small value of the test statistic corresponds to a lower (weighted average) hazard rate in the first group.
rho	Parameter to calculate weights in the rho-gamma family
gamma	Parameter to calculate weights in the rho-gamma family

event_time_weights

Optional vector of user defined weights. This weight vector needs to have one entry per event time (not per event, as multiple events may occur at the same time) and must be sorted by increasing event time.

Details

For a given sample, let \mathcal{D} be the set of unique event times. For a time-point $t \in \mathcal{D}$, let $n_{t,ctr}$ and $n_{t,trt}$ be the number of patients at risk in the control and treatment group and let $d_{t,ctr}$ and $d_{t,trt}$ be the respective number of events. The expected number of events in the control group is calculated under the least favorable configuration in H_0 , $\lambda_{ctr}(t) = \lambda_{trt}(t)$, as $e_{t,ctr} = (d_{t,ctr} + d_{t,trt}) \frac{n_{t0}}{n_{t0} + n_{t1}}$. The conditional variance of $d_{t,ctr}$ is calculated from a hypergeometric distribution as $var(d_{t,ctr}) = \frac{n_{t0}n_{t1}(d_{t0} + d_{t1})(n_{t0} + n_{t1} - d_{t0} - d_{t1})}{(n_{t0} + n_{t1})^2(n_{t0} + n_{t1} - 1)}$. Further define a weighting function $w(t)$. The weighted logrank test statistic for a comparison of two groups is

$$z = \sum_{t \in \mathcal{D}} w(t)(d_{t,ctr} - e_{t,ctr}) / \sqrt{\sum_{t \in \mathcal{D}} w(t)^2 var(d_{t,ctr})}$$

Under the the least favorable configuration in H_0 , the test statistic is asymptotically standard normally distributed and large values of z are in favor of the alternative.

The function consider particular weights in the Fleming-Harrington $\rho - \gamma$ family $w(t) = \hat{S}(t-)^{\rho}(1 - \hat{S}(t-))^{\gamma}$. Here, $\hat{S}(t) = \prod_{s \in \mathcal{D}: s \leq t} 1 - \frac{d_{t,ctr} + d_{t,trt}}{n_{t,ctr} + n_{t,trt}}$ is the pooled sample Kaplan-Meier estimator. (Note: Prior to package version 2.1, $S(t)$ was used in the definition of $\rho - \gamma$ weights, this was changed to $S(t-)$ with version 2.1.) Weights $\rho = 0, \gamma = 0$ correspond to the standard logrank test with constant weights $w(t) = 1$. Choosing $\rho = 0, \gamma = 1$ puts more weight on late events, $\rho = 1, \gamma = 0$ puts more weight on early events and $\rho = 1, \gamma = 1$ puts most weight on events at intermediate time points.

Value

A list with elements:

- D** A data frame event numbers, numbers at risk and expected number of events for each event time
- test** A data frame containing the z and chi-squared statistic for the one-sided and two-sided test, respectively, of the null hypothesis of equal hazard functions in both groups and the p-value for the one-sided test.
- var** The estimated variance of the sum of expected minus observed events in the first group.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

References

- Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schöler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.
- Thomas R Fleming and David P Harrington. *Counting processes and survival analysis*. John Wiley & Sons, 2011

See Also[logrank.maxtest](#)**Examples**

```
A <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
B <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.1, 0.6, 0.1), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.04, 0.04), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
dat <- sample_fun(A, B, r0 = 0.5, eventEnd = 30,
  lambdaRecr = 0.5, lambdaCens = 0.25 / 365,
  maxRecrCalendarTime = 2 * 365,
  maxCalendar = 4 * 365)
logrank.test(dat$y, dat$event, dat$group)
```

m2r

Transform median time into rate

Description

This helper function calculates the hazard rate per day of an exponential distribution from the median given in months.

Usage

```
m2r(x)
```

Arguments

x The median time in months to be transformed into rate

nphparams

*Simultaneous Inference For Parameters Quantifying Differences Between Two Survival Functions***Description**

Hypothesis tests with parametric multiple testing adjustment and simultaneous confidence intervals for a set of parameters, which quantify differences between two survival functions. Eligible parameters are differences in survival probabilities, log survival probabilities, complementary log log (cloglog) transformed survival probabilities, quantiles of the survival functions, log transformed quantiles, restricted mean survival times, as well as an average hazard ratio, the Cox model score statistic (logrank statistic), and the Cox-model hazard ratio.

Usage

```
nphparams(
  time,
  event,
  group,
  data = parent.frame(),
  param_type,
  param_par = NA,
  param_alternative = NA,
  lvl = 0.95,
  closed_test = FALSE,
  alternative_test = "two.sided",
  alternative_CI = "two.sided",
  haz_method = "local",
  rhs = 0,
  perturb = FALSE,
  Kpert = 500
)
```

Arguments

time	vector of observed event/censored times.
event	Vector with entries 0 or 1 (or FALSE/TRUE) indicating if an event was observed (1) or the time is censored (0).
group	group indicator, must be a vector with entries 0 or 1 indicating the allocation of a subject to one of two groups. Group 0 is regarded as reference group when calculating parameters.
data	an optional data frame containing the time, event and group data.
param_type	character vector defining the set of parameters that should be analysed. Possible entries are "S", "logS", "cloglogS", "Q", "logQ", "RMST", "avgHR", "score" and "HR", representing differences in survival probabilities, log survival probabilities, complementary log log (cloglog) transformed survival probabilities,

quantiles of the survival functions, log transformed quantiles, restricted mean survival times, as well as an average hazard ratio, the Cox model score statistic (logrank statistic), and the Cox-model hazard ratio.

param_par	numeric vector which contains the time points at which the requested parameters are evaluated (e.g. x-year survival or RMST after x-years), or, in case of analysing quantiles, the according probability. May be NA for parameter types "RMST", "avgHR", "score" or "HR". In this case, the minimum of the largest event times of the two groups is used. Also, times greater than this minimum are replaced by this minimum for "RMST", "avgHR", "score" or "HR".
param_alternative	optional character vector with entries "less" or "greater", defining the alternative for each parameter. Only required if one-sided tests or one-sided confidence intervals are requested. Note that group 0 is regarded as reference group when calculating parameters and therefore whether "greater" or "less" corresponds to a benefit may depend on the type of parameter. In general, to show larger survival in group 1 compared to group 0, alternatives would be "greater" for parameter types "S", "logS", "Q", "logQ" and "RMST" and would be "less" for parameters types "cloglogS", "avgHR", "HR", and "score". (The score test is defined here such that alternative "less" corresponds to smaller hazard (and better survival) in group 1 compared to group 0.)
lvl	Confidence level. Applies to, both, unadjusted and multiplicity adjusted (simultaneous) confidence intervals.
closed_test	logical indicating whether p-values should be adjusted using a closed testing procedure. Default is FALSE, and in this case p-values will be adjusted by a single step procedure. With k hypotheses this involves the computation of 2^k tests, which may require considerable computation time.
alternative_test	character with possible values "two.sided" (default) or "one.sided". Specifies whether hypothesis tests should be two-sided or one-sided. In the #' latter case, param_alternative must be defined.
alternative_CI	character with possible values "two.sided" (default) or "one.sided". Specifies whether confidence intervals should be two-sided or one-sided. In the latter case, param_alternative must be defined.
haz_method	character with possible values "local" or "muhaz". Specifies whether local hazard should be calculated under a local constant hazard assumption (default) #' or using the function <code>muhaz</code> from the <code>muhaz</code> package. Only relevant when median or log(median) survival times are analysed.
rhs	right-hand side vector of null hypotheses. Refers to log-scaled difference for ratios. Default is to consider for all null hypothesis a difference of 0.
perturb	logical, indicating whether the perturbation based estimate should be used instead of the asymptotic estimate to calculate the covariance matrix. Defaults to FALSE.
Kpert	The number of perturbation samples to be used with the perturbation approach for covariance estimation.

Value

A list of class `nphparams` with elements:

`est` Estimated differences (at log-scale in case of ratios).

`V` Estimated covariance matrix of differences.

`tab` A data frame with analysis results. Contains the parameter type (`Parameter`) and settings (`Time_or_which_quantile`), the estimated difference (`Estimate`), its standard error (`SE`), unadjusted confidence interval lower and upper bounds (`lwr_unadjusted`, `upr_unadjusted`), unadjusted p-values (`p_unadj`), multiplicity adjusted confidence interval lower and upper bounds (`lwr_adjusted`, `upr_adjusted`), single-step multiplicity adjusted p-values (`p_adj`), closed-test adjusted p-values, if requested (`p_adjusted_closed_test`) and for comparison Bonferroni-Holm adjusted p-values (`p_Holm`).

`param` The used parameter settings. If `param_par` was `NA` for "HR", "avgHR" or "RMST", it is replaced by `minmaxt` here.

`paramin` The parameter settings as provided to the function. The only difference to `param` is in `param_par`, as `NA` is not replaced here.

`dat0` A data frame with information on all observed events in group 0. Contains time (`t`), number of events (`ev`), Nelson-Aalen estimate (`NAsurv`) and Kaplan-Meier estimate (`KMSurv`) of survival, and the number at risk (`atrisk`).

`dat1` A data frame with information on all observed events in group 1. Contains time (`t`), number of events (`ev`), Nelson-Aalen estimate (`NAsurv`) and Kaplan-Meier estimate (`KMSurv`) of survival, and the number at risk (`atrisk`).

`minmaxt` Minimum of the largest event times of the two groups.

`est0` Estimated parameter values in group 0.

`est1` Estimated parameter values in group 1.

`V0` Estimated covariance matrix of parameter estimates in group 0.

`V1` Estimated covariance matrix of parameter estimates in group 1.

Author(s)

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See Also

[print.nphparams](#), [plot.nphparams](#)

Examples

```
data(pembro)
set1<-nphparams(time=time, event=event, group=group,data=pembro,
param_type=c("score","S"),
param_par=c(3.5,2),
param_alternative=c("less","greater"),
closed_test=TRUE,alternative_test="one.sided")
print(set1)
plot(set1,trt_name="Pembrolizumab",ctr_name="Cetuximab")
```

```

set2<-nphparams(time=time, event=event, group=group, data=pembro,
param_type=c("S","S","S","Q","RMST"),
param_par=c(0.5,1,2,0.5,3.5))
print(set2)
plot(set2,showlines=TRUE,show_rmst_diff=TRUE)

#Create a summary table for set2, showing parameter estimates for each group and the
#estimated differences between groups. Also show unadjusted and multiplicity adjusted
#confidence intervals using the multivariate normal method and, for comparison,
#Bonferroni adjusted confidence intervals:

set2Bonf<-nphparams(time=time, event=event, group=group, data=pembro,
param_type=c("S","S","S","Q","RMST"),
param_par=c(0.5,1,2,0.5,3.5),
lv1=1-0.05/5)
KI_paste<-function(x,r) {
x<-round(x,r)
paste("[",x[,1],", ", " ,x[,2],"]",sep="")
}
r<-3
tab<-data.frame(
Parameter=paste(set2$tab[,1],set2$tab[,2]),
Pembrolizumab=round(set2$est1,r),
Cetuximab=round(set2$est0,r),
Difference=round(set2$tab$Estimate,r),
CI_undadj=KI_paste(set2$tab[,5:6],r),
CI_adj=KI_paste(set2$tab[,8:9],r),
CI_Bonf=KI_paste(set2Bonf$tab[,c(5:6)],r))
tab

```

nph_gui

Launch a GUI (shiny app) for the nph package

Description

Launch a GUI (shiny app) for the nph package

Usage

```
nph_gui()
```

Details

The packages shinycssloaders, formatR and styler are required for correct display of the GUI. The package rmarkdown with access to pandoc is required to save reports.

pchaz

*Calculate survival for piecewise constant hazard***Description**

Calculates hazard, cumulative hazard, survival and distribution function based on hazards that are constant over pre-specified time-intervals.

Usage

```
pchaz(Tint, lambda)
```

Arguments

Tint	vector of length $k + 1$, for the boundaries of k time intervals (presumably in days) with piecewise constant hazard. The boundaries should be increasing and the first one should be 0, the last one should be larger than the assumed trial duration.
lambda	vector of length k with the piecewise constant hazards for the intervals specified via Tint.

Details

Given k time intervals $[t_{j-1}, t_j), j = 1, \dots, k$ with $0 = t_0 < t_1 < \dots < t_k$, the function assume constant hazards λ_j at each interval. The resulting hazard function is $\lambda(t) = \sum_{j=1}^k \lambda_j 1_{t \in [t_{j-1}, t_j)}$, the cumulative hazard function is $\Lambda(t) = \int_0^t \lambda(s) ds = \sum_{j=1}^k ((t_j - t_{j-1}) \lambda_j 1_{t > t_j} + (t - t_{j-1}) \lambda_j 1_{t \in [t_{j-1}, t_j)})$ and the survival function $S(t) = e^{-\Lambda(t)}$. The output includes the functions values calculated for all integer time points between 0 and the maximum of Tint. Additionally, a list with functions is also given to calculate the values at any arbitrary point t .

Value

A list with class `mipch` containing the following components:

`haz` Values of the hazard function over discrete times `t`.

`cumhaz` Values of the cumulative hazard function over discrete times `t`.

`S` Values of the survival function over discrete times `t`.

`F` Values of the distribution function over discrete times `t`.

`t` Time points for which the values of the different functions are calculated.

`Tint` Input vector of boundaries of time intervals.

`lambda` Input vector of piecewise constant hazards.

`funcs` A list with functions to calculate the hazard, cumulative hazard, survival, pdf and cdf over arbitrary continuous times.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>, Nicolas Ballarini

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schöler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[subpop_pchaz](#), [pop_pchaz](#), [plot.mixpch](#)

Examples

```
pchaz(Tint = c(0, 40, 100), lambda=c(.02, .05))
```

pembro

Reconstructed Data Set Based On Survival Curves In Burtess et al. 2019

Description

The data set was approximately reconstructed from the survival curves shown in Figure 2D of Burtess et al. 2019 (see references). It contains survival times and event #' indicator under treatment with pembrolizumab (group 1) versus cetuximab with chemotherapy (group 0).

Usage

```
data(pembro)
```

Format

A data frame.

References

Burtess, B., Harrington, K. J., Greil, R., Soulières, D., Tahara, M., de Castro Jr, G., ... & Abel, E. (2019). Pembrolizumab alone or with chemotherapy versus cetuximab with chemotherapy for recurrent or metastatic squamous cell carcinoma of the head and neck (KEYNOTE-048): a randomised, open-label, phase 3 study. *The Lancet*, 394(10212), 1915-1928.

Examples

```
data(pembro)
head(pembro)
```

Description

Plots survival and other functions stored in mixpch objects versus time.

Usage

```
## S3 method for class 'mixpch'
plot(
  x,
  fun = c("S", "F", "haz", "cumhaz"),
  add = FALSE,
  ylab = fun,
  xlab = "Time",
  ...
)
```

Arguments

x	an object of class mixpch.
fun	character string in c("S", "F", "haz", "cumhaz") indicating which function to plot. Select "S" for the survival function, "F" for the distribution function, "haz" for the hazard function or "cumhaz" for the cumulative hazard function.
add	logical, indicates if the drawing should be added to an existing plot.
ylab	label of the y-axis
xlab	label of the x-axis
...	further arguments passed to the plotting functions

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schüler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[pchaz](#), [subpop_pchaz](#), [pop_pchaz](#)

Examples

```
A <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
plot(A)
plot(A, "haz", add = TRUE)
```

plot.nphparams

Plot nphparams Objects

Description

Plots the estimated survival distributions, shows numbers at risk and indicates the requested parameters for quantifying differences between the survival curves.

Usage

```
## S3 method for class 'nphparams'
plot(
  x,
  xlim = NULL,
  ylim = c(0, 1),
  trt_name = "Treatment",
  ctr_name = "Control",
  xlab = "Time",
  ylab = "Survival",
  main = "",
  col_ctr = 1,
  col_trt = 2,
  atrisktimes = 0:3,
  bold = 2,
  showlines = FALSE,
  show_rmst_diff = FALSE,
  ...
)
```

Arguments

x	an object of class nphparams.
xlim	limits of the x-axis, must be a numeric vector of length two
ylim	limits of the y-axis, must be a numeric vector of length two
trt_name	character, an optional name for group 1 to be shown with the number at risk table in the plot. Default is "Treatment".

ctr_name	character, an optional name for group 0 to be shown with the number at risk table in the plot. Default is "Control".
xlab	character, an optional label for the x-axis. Default is "Time".
ylab	character, an optional label for the x-axis. Default is "Survival".
main	character, an optional title of the plot. Default is "", showing no title.
col_ctr	the color of the survival curve estimate of group 0. Default is 1 (black).
col_trt	the color of the survival curve estimate of group 1. Default is 2 (red).
atrisktimes	numeric vector of time-points for which the number at risk is displayed.
bold	numeric, passed to linewidth and font settings. Default is 2, resulting in lines of width 2 and boldfont. Use 1 for line-width 1 and standard font.
showlines	logical, indicating whether the time-points or the quantile-probabilites defined for the requested parametes should be shown in terms of vertical or horizontal lines. Default is FALSE.
show_rmst_diff	logical, indicating whether the estiamted difference in restricted mean survival times should by visualized by a gray background area.
...	further arguments, not used

Details

When setting `show_lines`, line type 2 (dashed) is used for survival probabilities and quantiles, line type 3 (dotted) is used for the score test, average hazard ratio and Cox model hazard ratio and line type 5 (long dashed) is used for restricted mean survival time.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

See Also

[nphparams](#), [plot.nphparams](#)

Examples

```
data(pembro)
set1<-nphparams(time=time, event=event, group=group,data=pembro,
param_type=c("score","S"),
param_par=c(3.5,2),
param_alternative=c("less","greater"),
closed_test=TRUE,alternative_test="one.sided")
print(set1)
plot(set1,trt_name="Pembrolizumab",ctr_name="Cetuximab")

set2<-nphparams(time=time, event=event, group=group, data=pembro,
param_type=c("S","S","S","Q","RMST"),
param_par=c(0.5,1,2,0.5,3.5))
print(set2)
plot(set2,showlines=TRUE,show_rmst_diff=TRUE)
```

plot_diagram	<i>Draw a state space figure</i>
--------------	----------------------------------

Description

A figure that shows the states and the possible transitions between them.

Usage

```
plot_diagram(  
  A,  
  B,  
  A_subgr_labels = "",  
  B_subgr_labels = "",  
  which = c("Both", "Experimental", "Control"),  
  treatment_labels = c("Experimental", "Control"),  
  colors = "default",  
  show.rate = TRUE  
)
```

Arguments

A	An object of class <code>mixpch</code> , resembling the survival function in treatment group 0
B	An object of class <code>mixpch</code> , resembling the survival function in treatment group 1
A_subgr_labels	A character vector with the same length as <code>A\$p</code> . It indicates names for the subgroups in A
B_subgr_labels	A character vector with the same length as <code>B\$p</code> . It indicates names for the subgroups in B
which	Which treatment arm should be shown? One of "Both", "Experimental", "Control".
treatment_labels	A character vector of length 2 indicating the treatment labels.
colors	Either a vector of length two with colors for A and B, or "default".
show.rate	A logical indicating whether the rate should be shown in the diagram

plot_shhr	<i>Plot of survival, hazard and hazard ratio of two groups as a function of time</i>
-----------	--

Description

A convenience function that uses the generic plot function in the `nph` package to plot the three functions in a layout of 3 columns and 1 row.

Usage

```
plot_shhr(A, B, main = "", xmax = NULL, ymax_haz = NULL, ymax_hr = NULL)
```

Arguments

A	An object of class <code>mixpch</code> , resembling the survival function in treatment group 0
B	An object of class <code>mixpch</code> , resembling the survival function in treatment group 1
main	An overall title for the plot
xmax	A maximum value for the x-axis. The plot is drawn using <code>xlim = c(0, xmax)</code>
ymax_haz	A maximum value for the y-axis for the hazards plot. The plot is drawn using <code>ylim = c(0, ymax_haz)</code>
ymax_hr	A maximum value for the y-axis for the hazards ratio plot. The plot is drawn using <code>ylim = c(0, ymax_hr)</code>

plot_subgroups	<i>Draw a population composition plot</i>
----------------	---

Description

A figure that shows the composition of the population under study though time

Usage

```
plot_subgroups(
  A,
  B,
  colors = "default",
  max_time = max(A$Tint),
  position = c("stack", "fill"),
  title = ""
)
```

Arguments

A	An object of class <code>mixpch</code> , resembling the survival function in treatment group 0
B	An object of class <code>mixpch</code> , resembling the survival function in treatment group 1
colors	Either a vector of length four with colors for A and B and subgroup 1 and 2, or "default".
max_time	the maximum value for the x-axis.
position	Either "stack" or "fill". By default (stack), the total population decreases through time. If position="fill", the size of the population is rescaled to show conditional percentages.
title	The text for the title.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>, Nicolas Ballarini

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schüler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[pop_pchaz](#)

Examples

```
A <- pop_pchaz(Tint = c(0, 90, 365),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
B <- pop_pchaz(Tint = c(0, 90, 365),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.1, 0.6, 0.1), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.04, 0.04), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
plot_subgroups(A, B, title = "position='stack'")
plot_subgroups(A, B, position='fill', title = "position='fill'")
```

pop_pchaz	<i>Calculate survival for piecewise constant hazards with change after random time and mixture of subpopulations</i>
-----------	--

Description

Calculates hazard, cumulative hazard, survival and distribution function based on hazards that are constant over pre-specified time-intervals

Usage

```
pop_pchaz(
  Tint,
  lambdaMat1,
  lambdaMat2,
  lambdaProgMat,
  p,
  timezero = FALSE,
  int_control = list(rel.tol = .Machine$double.eps^0.4, abs.tol = 1e-09),
  discrete_approximation = FALSE
)
```

Arguments

Tint	vector of length $k + 1$, for the boundaries of k time intervals (presumably in days) with piecewise constant hazard. The boundaries should be increasing and the first one should be 0, the last one should be larger than the assumed trial duration.
lambdaMat1	matrix of dimension m -by- k , each row contains the vector of piecewise constant hazards for one subpopulation before the changing event happens, for the intervals specified via Tint.
lambdaMat2	matrix of dimension m -by- k , each row contains the vector piecewise constant hazards for one subpopulation after the changing event has happened, for the intervals specified via Tint.
lambdaProgMat	matrix of dimension m -by- k , each row contains the vector of piecewise constant hazards for one subpopulation for the changing event, for the intervals specified via Tint.
p	vector of length m for relative sizes (proportions) of the subpopulations. They should sum up to 1.
timezero	logical, indicating whether after the changing event the timecount, governing which interval in Tint and which according value in lambda2 is used, should restart at zero. This argument is either of length 1 (applying the same to all subgroups) or the same length as the number of subgroups.
int_control	A list with additional parameters to be passed to the integrate function.

discrete_approximation

if TRUE, the function uses an approximation based on discretizing the time, instead of integrating. This speeds up the calculations

Details

Given m subgroups with relative sizes p_1, \dots, p_m and subgroup-specific survival functions $S_l(t)$, the marginal survival function is the mixture $S(t) = \sum_{l=1}^m p_l S_l(t)$. Note that the respective hazard function is not a linear combination of the subgroup-specific hazard functions. It may be calculated by the general relation $\lambda(t) = -\frac{dS(t)}{dt} \frac{1}{S(t)}$. In each subgroup, the hazard is modelled as a piecewise constant hazard, with the possibility to also model disease progression. Therefore, each row of the hazard rates is used in [subpop_pchaz](#). See [pchaz](#) and [subpop_pchaz](#) for more details. The output includes the function values calculated for all integer time points between 0 and the maximum of Tint.

Note: this function may be very slow in cases where many time points need to be calculated. If this happens, use `discrete_approximation = TRUE`.

Value

A list with class `mixpch` containing the following components:

`haz` Values of the hazard function.

`cumhaz` Values of the cumulative hazard function.

`S` Values of the survival function.

`F` Values of the distribution function.

`t` Time points for which the values of the different functions are calculated.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>, Nicolas Ballarini

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schöler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[pchaz](#), [subpop_pchaz](#), [plot.mixpch](#)

Examples

```
pop_pchaz(Tint = c(0, 40, 100),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2),
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2),
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2),
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
```

print.nphparams *Print nphparams Objects*

Description

Prints the results table of an nphparams object.

Usage

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

Arguments

x an object of class nphparams.
... further arguments, not used.

Details

Estimates corresponding to differences at a log scale are transformed by taking `exp()`, and are labelled as ratios. I.e. differences in log survival probabilities, differences in log quantiles, cloglog survival differences (equivalent to the log cumulative hazard ratio), log average hazard ratios or Cox model log hazard ratios are transformed to survival probability ratios, quantile ratios, cumulative hazard ratios, average hazard ratios or Cox model hazard ratios, respectively. In the output, the standard error at the backtransformed scale is calculated by the delta-method. Confidence interval bounds are calculated at the log-scale, though, and then transformed by taking `exp()`.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

See Also

[nphparams](#), [plot.nphparams](#)

Examples

```
data(pembro)  
set1<-nphparams(time=time, event=event, group=group,data=pembro,  
param_type=c("score","S"),  
param_par=c(3.5,2),  
param_alternative=c("less","greater"),  
closed_test=TRUE,alternative_test="one.sided")  
print(set1)  
plot(set1,trt_name="Pembrolizumab",ctr_name="Cetuximab")  
  
set2<-nphparams(time=time, event=event, group=group, data=pembro,  
param_type=c("S","S","S","Q","RMST"),
```

```
param_par=c(0.5,1,2,0.5,3.5))
print(set2)
plot(set2,showlines=TRUE,show_rmst_diff=TRUE)
```

rSurv_conditional_fun *Draw conditional random survival times from mixpch object.*

Description

Draws independent random survival times from mixpch objects conditional on observed time.

Usage

```
rSurv_conditional_fun(x, y)
```

Arguments

x	An object of class mixpch
y	A vector of observed right censored times

Details

Note that the mixpch object stores the survival function up to some time T. For random times equal or larger than T, the survival function is zero.

Value

A vector of random survival times, conditional on the observed censored times.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schöler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[rSurv_fun](#), [sample_fun](#), [sample_conditional_fun](#)

Examples

```
A <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
rSurv_conditional_fun(x = A, y = c(10,15,9,2,1))
```

rSurv_fun

Draw random survival times from mixpch object.

Description

Draws independent random survival times from mixpch objects.

Usage

```
rSurv_fun(n, x)
```

Arguments

n	Number of random draws
x	An object of class mixpch

Details

The mixpch object stores the survival function up to some time T. For random times equal or larger T, the value T is returned.

Value

A vector of random survival times.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schüler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[rSurv_conditional_fun](#), [sample_fun](#), [sample_conditional_fun](#)

Examples

```
A <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
rSurv_fun(n = 10, x = A)
```

sample_conditional_fun

Draw conditional survival times based on study settings

Description

Simulates data for a randomized controlled survival study conditional on observed interim data.

Usage

```
sample_conditional_fun(
  dat,
  A,
  B,
  r0 = 0.5,
  eventEnd,
  lambdaRecr,
  lambdaCens,
  maxRecrCalendarTime,
  maxCalendar
)
```

Arguments

dat	A data frame with the same structure and column names as the output of <code>sample_fun</code> , containing the data to condition on
A	An object of class <code>mixpch</code> , resembling the survival function in treatment group 0
B	An object of class <code>mixpch</code> , resembling the survival function in treatment group 1
r0	Allocation ratio to group 1 (must be a number between 0 and 1)
eventEnd	Number of events, after which the study stops
lambdaRecr	Rate per day for recruiting patients, assuming recruiting follows a Poisson process
lambdaCens	Rate per day for random censoring, assuming censoring times are exponential

maxRecrCalendarTime Maximal duration of recruitment in days
 maxCalendar Maximal total study duration in days, after which the study stops

Details

For simulating the data, patients are allocated randomly to either group (unrestricted randomization).

Value

A data frame with each line representing data for one patient and the following columns:

group Treatment group
 inclusion Start of observation in terms of calendar time
 y Observed survival/censored time
 yCalendar End of observation in terms of calendar time.
 event logical, TRUE indicates the observation ended with an event, FALSE corresponds to censored times
 adminCens logical, True indicates that the observation is subject to administrative censoring, i.e. the subject was observed until the end of the study without an event.
 cumEvents Cumulative number of events over calendar time of end of observation

The data frame is ordered by yCalendar

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schüler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[rSurv_fun](#), [rSurv_conditional_fun](#), [sample_fun](#)

Examples

```
A <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
B <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
```

```

lambdaMat2 = matrix(c(0.5, 0.1, 0.6, 0.1), 2, 2) / 365,
lambdaProg = matrix(c(0.5, 0.5, 0.04, 0.04), 2, 2) / 365,
p = c(0.8, 0.2),
timezero = FALSE, discrete_approximation = TRUE)
datinterim <- sample_fun(A, B, r0 = 0.5, eventEnd = 30, lambdaRecr = 1,
  lambdaCens = 0.25 / 365,
  maxRecrCalendarTime = 3 * 365,
  maxCalendar = 4 * 365)
datcond <- sample_conditional_fun(datinterim, A, B, r0 = 0.5, eventEnd = 60,
  lambdaRecr = 1, lambdaCens = 0.25 / 365, maxRecrCalendarTime = 3 * 365,
  maxCalendar = 4 * 365)

```

sample_fun

Draw survival times based on study settings

Description

Simulates data for a randomized controlled survival study.

Usage

```

sample_fun(
  A,
  B,
  r0 = 0.5,
  eventEnd,
  lambdaRecr,
  lambdaCens,
  maxRecrCalendarTime,
  maxCalendar
)

```

Arguments

A	An object of class <code>mixpch</code> , resembling the survival function in treatment group 0
B	An object of class <code>mixpch</code> , resembling the survival function in treatment group 1
r0	Allocation ratio to group 0 (must be a number between 0 and 1)
eventEnd	Number of events, after which the study stops
lambdaRecr	Rate per day for recruiting patients, assuming recruiting follows a Poisson process
lambdaCens	Rate per day for random censoring, assuming censoring times are exponential
maxRecrCalendarTime	Maximal duration of recruitment in days
maxCalendar	Maximal total study duration in days, after which the study stops

Details

For simulating the data, patients are allocated randomly to either group (unrestricted randomization).

Value

A data frame with each line representing data for one patient and the following columns:

group Treatment group

inclusion Start of observation in terms of calendar time

y Observed survival/censored time

yCalendar End of observation in terms of calendar time.

event logical, TRUE indicates the observation ended with an event, FALSE corresponds to censored times

adminCens logical, True indicates that the observation is subject to administrative censoring, i.e. the subject was observed until the end of the study without an event.

cumEvents Cumulative number of events over calendar time of end of observation

The data frame is ordered by yCalendar

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schüler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[rSurv_fun](#), [rSurv_conditional_fun](#), [sample_conditional_fun](#)

Examples

```
A <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.2, 0.6, 0.2), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.4, 0.4), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
B <- pop_pchaz(Tint = c(0, 90, 1500),
  lambdaMat1 = matrix(c(0.2, 0.1, 0.4, 0.1), 2, 2) / 365,
  lambdaMat2 = matrix(c(0.5, 0.1, 0.6, 0.1), 2, 2) / 365,
  lambdaProg = matrix(c(0.5, 0.5, 0.04, 0.04), 2, 2) / 365,
  p = c(0.8, 0.2),
  timezero = FALSE, discrete_approximation = TRUE)
plot(A)
```

```
plot(B, add = TRUE)
dat <- sample_fun(A, B, r0 = 0.5, eventEnd = 30, lambdaRecr = 0.5,
  lambdaCens = 0.25 / 365, maxRecrCalendarTime = 2 * 365,
  maxCalendar = 4 * 365)
```

subpop_pchaz	<i>Calculate survival for piecewise constant hazards with change after random time</i>
--------------	--

Description

Calculates hazard, cumulative hazard, survival and distribution function based on hazards that are constant over pre-specified time-intervals

Usage

```
subpop_pchaz(
  Tint,
  lambda1,
  lambda2,
  lambdaProg,
  timezero = FALSE,
  int_control = list(rel.tol = .Machine$double.eps^0.4, abs.tol = 1e-09),
  discrete_approximation = FALSE
)
```

Arguments

Tint	vector of length $k + 1$, for the boundaries of k time intervals (presumably in days) with piecewise constant hazard. The boundaries should be increasing and the first one should be 0, the last one should be larger than the assumed trial duration.
lambda1	vector of length k for piecewise constant hazards before the changing event happens, for the intervals specified via T.
lambda2	vector of length k for piecewise constant hazards after the changing event has happened, for the intervals specified via T.
lambdaProg	vector of length k for piecewise constant hazards for the changing event, for the intervals specified via T.
timezero	logical, indicating whether after the changeing event the timecount, governing which interval in Tint and which according value in lambda2 is used, should restart at zero.
int_control	A list with the rel.tol and abs.tol paramaters to be passed to the integrate function.
discrete_approximation	if TRUE, the function uses an approximation based on discretizing the time, instead of integrating. This speeds up the calculations

Details

We assume that the time to disease progression T_{PD} is governed by a separate process with hazard function $\eta(t)$, which does not depend on the hazard function for death $\lambda(t)$. $\eta(t)$, too, may be modelled as piecewise constant or, for simplicity, as constant over time. We define $\lambda_{prePD}(t)$ and $\lambda_{postPD}(t)$ as the hazard functions for death before and after disease progression. Conditional on $T_{PD} = s$, the hazard function for death is $\lambda(t|T_{PD} = s) = \lambda_{prePD}(t)I_{t \leq s} + \lambda_{postPD}(t)I_{t > s}$. The conditional survival function is $S(t|T_{PD} = s) = \exp(-\int_0^t \lambda(t|T_{PD} = s)ds)$. The unconditional survival function results from integration over all possible progression times as $S(t) = \int_0^t S(t|T_{PD} = s)dP(T_{PD} = s)$. The output includes the function values calculated for all integer time points between 0 and the maximum of Tint. Additionally, a list with functions is also given to calculate the values at any arbitrary point t .

Value

A list with class `mixpch` containing the following components:

`haz` Values of the hazard function.

`cumhaz` Values of the cumulative hazard function.

`S` Values of the survival function.

`F` Values of the distribution function.

`t` Time points for which the values of the different functions are calculated.

`Tint` Input vector of boundaries of time intervals.

`lambda1` Input vector of piecewise constant hazards before the changing event happen.

`lambda2` Input vector of piecewise constant hazards after the changing event happen.

`lambdaProg` Input vector of piecewise constant hazards for the changing event.

`funs` A list with functions to calculate the hazard, cumulative hazard, survival, and cdf over arbitrary continuous times.

Author(s)

Robin Ristl, <robin.ristl@meduniwien.ac.at>, Nicolas Ballarini

References

Robin Ristl, Nicolas Ballarini, Heiko Götte, Armin Schöler, Martin Posch, Franz König. Delayed treatment effects, treatment switching and heterogeneous patient populations: How to design and analyze RCTs in oncology. *Pharmaceutical statistics*. 2021; 20(1):129-145.

See Also

[pchaz](#), [pop_pchaz](#), [plot.mixpch](#)

Examples

```
subpop_pchaz(Tint = c(0, 40, 100), lambda1 = c(0.2, 0.4), lambda2 = c(0.1, 0.01),
lambdaProg = c(0.5, 0.4), timezero = FALSE, discrete_approximation = TRUE)
subpop_pchaz(Tint = c(0, 40, 100), lambda1 = c(0.2, 0.4), lambda2 = c(0.1, 0.01),
lambdaProg = c(0.5, 0.4), timezero = TRUE, discrete_approximation = TRUE)
```

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