

Package ‘bayesCT’

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Type Package

Title Simulation and Analysis of Adaptive Bayesian Clinical Trials

Version 0.99.3

Description

Simulation and analysis of Bayesian adaptive clinical trials for binomial, Gaussian, and time-to-event data types, incorporates historical data and allows early stopping for futility or early success. The package uses novel and efficient Monte Carlo methods for estimating Bayesian posterior probabilities, evaluation of loss to follow up, and imputation of incomplete data. The package has the functionality for dynamically incorporating historical data into the analysis via the power prior or non-informative priors.

LazyLoad yes

License GPL-3

NeedsCompilation no

URL <https://github.com/thevaachandereng/bayesCT/>

BugReports <https://github.com/thevaachandereng/bayesCT/issues/>

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analysis	<i>Analysis wrapper function</i>
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Description

Wrapper function to analyze bayesian trials.

Usage

```
analysis(input, type = "binomial", .data = NULL)
```

Arguments

input	list. Input function for all the analysis.
type	character. Type of analysis to be ran (binomial (default), normal. etc.).
.data	NULL. stores the all the details, please do not fill it in.

Value

a list with results of the analysis of bayesian trial.

prob_of_accepting_alternative scalar. The input parameter of probability of accepting the alternative.

margin scalar. The margin input value of difference between mean estimate of treatment and mean estimate of the control.

alternative character. The input parameter of alternative hypothesis.

N_treatment scalar. The number of patients enrolled in the experimental group for each simulation.

N_control scalar. The number of patients enrolled in the control group for each simulation.

N_enrolled vector. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)

N_complete scalar. The number of patients who completed the trial and had no loss to follow-up.

post_prob_accept_alternative vector. The final probability of accepting the alternative hypothesis after the analysis is done.

est_final scalar. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.

stop_futility scalar. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

stop_expected_success scalar. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

beta_prior	<i>Beta prior for for control and treatment group</i>
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Description

Wrapper function for beta prior $\text{beta}(a_0, b_0)$.

Usage

```
beta_prior(a0 = 1, b0 = 1, .data = NULL)
```

Arguments

<code>a0</code>	numeric. The first shape parameter in the beta distribution ($\text{beta}(a_0, b_0)$).
<code>b0</code>	numeric. The second shape parameter in the beta distribution ($\text{beta}(a_0, b_0)$).
<code>.data</code>	NULL. stores the beta prior rate, please do not fill it in.

Value

a list with vector of beta rate for the beta prior for treatment and control group.

Examples

```
beta_prior(a0 = 1, b0 = 1)
```

binomialBACT	<i>Binomial counts for Bayesian Adaptive Trials</i>
--------------	-----------------------------------------------------

Description

Simulation for binomial counts for Bayesian Adaptive trial with different inputs to control for power, sample size, type 1 error rate, etc.

Usage

```
binomialBACT(
  p_treatment,
  p_control = NULL,
  y0_treatment = NULL,
  N0_treatment = NULL,
  y0_control = NULL,
  N0_control = NULL,
  N_total,
  lambda = 0.3,
  lambda_time = NULL,
```

```

interim_look = NULL,
EndofStudy,
prior = c(1, 1),
block = 2,
rand_ratio = c(1, 1),
prop_loss_to_followup = 0.1,
alternative = "greater",
h0 = 0,
futility_prob = 0.05,
expected_success_prob = 0.9,
prob_ha = 0.95,
N_impute = 10,
number_mcmc = 10000,
discount_function = "identity",
alpha_max = 1,
fix_alpha = FALSE,
weibull_scale = 0.135,
weibull_shape = 3,
method = "fixed"
)

```

Arguments

<code>p_treatment</code>	scalar. Proportion of events under the treatment arm.
<code>p_control</code>	scalar. Proportion of events under the control arm.
<code>y0_treatment</code>	scalar. Number of events for the historical treatment arm.
<code>N0_treatment</code>	scalar. Sample size of the historical treatment arm.
<code>y0_control</code>	scalar. Number of events for the historical control arm.
<code>N0_control</code>	scalar. Sample size of the historical control arm.
<code>N_total</code>	scalar. Total sample size.
<code>lambda</code>	vector. Enrollment rates across simulated enrollment times. See enrollment for more details.
<code>lambda_time</code>	vector. Enrollment time(s) at which the enrollment rates change. Must be same length as <code>lambda</code> . See enrollment for more details.
<code>interim_look</code>	vector. Sample size for each interim look. Note: the maximum sample size should not be included.
<code>EndofStudy</code>	scalar. Length of the study.
<code>prior</code>	vector. Prior values of beta rate, $\text{Beta}(a_0, b_0)$. The default is set to $\text{Beta}(1, 1)$.
<code>block</code>	scalar. Block size for generating the randomization schedule.
<code>rand_ratio</code>	vector. Randomization allocation for the ratio of control to treatment. Integer values mapping the size of the block. See randomization for more details.
<code>prop_loss_to_followup</code>	scalar. Overall proportion of subjects lost to follow-up.
<code>alternative</code>	character. The string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".

<code>h0</code>	scalar. Threshold for comparing two mean values. Default is $h0=0$.
<code>futility_prob</code>	scalar. Probability of stopping early for futility.
<code>expected_success_prob</code>	scalar. Probability of stopping early for success.
<code>prob_ha</code>	scalar. Probability of alternative hypothesis.
<code>N_impute</code>	scalar. Number of imputations for Monte Carlo simulation of missing data.
<code>number_mcmc</code>	scalar. Number of Monte Carlo Markov Chain draws in sampling posterior.
<code>discount_function</code>	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.
<code>alpha_max</code>	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
<code>fix_alpha</code>	logical. Fix alpha at alpha_max? Default value is FALSE.
<code>weibull_scale</code>	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>weibull_shape</code>	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>method</code>	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the <code>bdpsurvival</code> vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.

Value

a list of output for a single trial simulation.

`p_treatment` scalar. The input parameter of proportion of events in the treatment group.

`p_control` scalar. The input parameter of proportion of events in the control group.

`prob_of_accepting_alternative` scalar. The input parameter of probability threshold of accepting the alternative.

margin scalar. The margin input value of difference between mean estimate of treatment and mean estimate of the control.
alternative character. The input parameter of alternative hypothesis.
interim_look vector. The sample size for each interim look.
N_treatment scalar. The number of patients enrolled in the experimental group for each simulation.
N_control scalar. The number of patients enrolled in the control group for each simulation.
N_enrolled vector. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)
N_complete scalar. The number of patients who completed the trial and had no loss to follow-up.
post_prob_accept_alternative vector. The final probability of accepting the alternative hypothesis after the analysis is done.
est_final scalar. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.
stop_futility scalar. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.
stop_expected_success scalar. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.
est_interim scalar. The interim estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.

 binomialdata

Binomial dataset for analyzing adaptive Bayesian trials

Description

A dataset containing the results of 300 patients with binomial outcome, the dataset is filled with loss to follow up.

Usage

```
data(binomialdata)
```

Format

A data frame with 300 rows and 4 variables:

id Patient ID in the trial

treatment treatment assignment for patients, 1 for treatment group 0 for control group

outcome binomial outcome of the trial, 1 for response (success or failure), 0 for no response

complete 1 for complete outcome, 0 for loss to follow up

Examples

```
data(binomialdata)
```

binomial_analysis *Analyzing Bayesian trial for binomial counts*

Description

Function to analyze Bayesian trial for binomial count data which allows early stopping and incorporation of historical data using the discount function approach

Usage

```
binomial_analysis(
  treatment,
  outcome,
  complete = NULL,
  y0_treatment = NULL,
  N0_treatment = NULL,
  y0_control = NULL,
  N0_control = NULL,
  alternative = "greater",
  N_impute = 10,
  h0 = 0,
  number_mcmc = 10000,
  prob_ha = 0.95,
  futility_prob = 0.1,
  expected_success_prob = 0.9,
  prior = c(1, 1),
  discount_function = "identity",
  fix_alpha = FALSE,
  alpha_max = 1,
  weibull_scale = 0.135,
  weibull_shape = 3,
  method = "fixed"
)
```

Arguments

treatment	vector. treatment assignment for patients, 1 for treatment group and 0 for control group
outcome	vector. binomial outcome of the trial, 1 for response (success or failure), 0 for no response.
complete	vector. similar length as treatment and outcome variable, 1 for complete outcome, 0 for loss to follow up. If complete is not provided, the dataset is assumed to be complete.
y0_treatment	scalar. Number of events for the historical treatment arm.
N0_treatment	scalar. Number of observations of the historical treatment group.

<code>y0_control</code>	scalar. Number of events for the historical control arm.
<code>N0_control</code>	scalar. Number of observations of the historical control group.
<code>alternative</code>	character. The string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".
<code>N_impute</code>	scalar. Number of imputations for Monte Carlo simulation of missing data.
<code>h0</code>	scalar. Threshold for comparing two mean values. Default is $h_0=0$.
<code>number_mcmc</code>	scalar. Number of Monte Carlo Markov Chain draws in sampling posterior.
<code>prob_ha</code>	scalar. Probability of alternative hypothesis.
<code>futility_prob</code>	scalar. Probability of stopping early for futility.
<code>expected_success_prob</code>	scalar. Probability of stopping early for success.
<code>prior</code>	vector. Prior values of beta rate, $\text{Beta}(a_0, b_0)$. The default is set to $\text{Beta}(1, 1)$.
<code>discount_function</code>	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.
<code>fix_alpha</code>	logical. Fix alpha at alpha_max? Default value is FALSE.
<code>alpha_max</code>	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
<code>weibull_scale</code>	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>weibull_shape</code>	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>method</code>	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the bdpsurvival vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.

Value

a list of output for the Bayesian trial for binomial count.

prob_of_accepting_alternative scalar. The input parameter of probability of accepting the alternative.

margin scalar. The margin input value of difference between mean estimate of treatment and mean estimate of the control.

alternative character. The input parameter of alternative hypothesis.

N_treatment scalar. The number of patients enrolled in the experimental group for each simulation.

N_control scalar. The number of patients enrolled in the control group for each simulation.

N_enrolled vector. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)

N_complete scalar. The number of patients who completed the trial and had no loss to follow-up.

post_prob_accept_alternative vector. The final probability of accepting the alternative hypothesis after the analysis is done.

est_final scalar. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.

stop_futility scalar. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

stop_expected_success scalar. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

binomial_outcome	<i>Proportion of an event in control and treatment</i>
------------------	--------------------------------------------------------

Description

Wrapper function for proportion of an event in control and treatment group with binomial outcome.

Usage

```
binomial_outcome(p_treatment = NULL, p_control = NULL, .data = NULL)
```

Arguments

p_treatment numeric. The proportion of an event in the treatment group, $0 < p_treatment < 1$.

p_control numeric. The proportion of an event in the control group, $0 < p_control < 1$.

.data NULL. stores the proportion of control and treatment, please do not fill it in.

Value

a list with proportion of control and treatment group.

Examples

```
binomial_outcome(p_control = 0.12, p_treatment = 0.08)
```

data_binomial	<i>Data file for binomial analysis</i>
---------------	----------------------------------------

Description

Wrapper function for data file in binomial analysis.

Usage

```
data_binomial(treatment, outcome, complete, .data = NULL)
```

Arguments

treatment	vector. treatment assignment for patients, 1 for treatment group and 0 for control group
outcome	vector. binomial outcome of the trial, 1 for response (success or failure), 0 for no response.
complete	vector. similar length as treatment and outcome variable, 1 for complete outcome, 0 for loss to follow up. If complete is not provided, the dataset is assumed to be complete.
.data	NULL. stores the binomial data for analysis, please do not fill it in.

Value

a list with treatment, outcome and loss to follow up vector with binomial outcome.

Examples

```
data_binomial(treatment = c(0, 1), outcome = c(1, 1), complete = c(1, 1))
```

data_normal	<i>Data file for normal analysis</i>
-------------	--------------------------------------

Description

Wrapper function for data file in normal analysis.

Usage

```
data_normal(treatment, outcome, complete, .data = NULL)
```

Arguments

treatment	vector. treatment assignment for patients, 1 for treatment group and 0 for control group
outcome	vector. normal outcome of the trial.
complete	vector. similar length as treatment and outcome variable, 1 for complete outcome, 0 for loss to follow up. If complete is not provided, the dataset is assumed to be complete.
.data	NULL. stores the normal data for analysis, please do not fill it in.

Value

a list with treatment, outcome and loss to follow up vector with normal outcome.

data_survival	<i>Data file for survival analysis</i>
---------------	----------------------------------------

Description

Wrapper function for data file in survival analysis.

Usage

```
data_survival(time, treatment, event, .data = NULL)
```

Arguments

time	vector. exposure time for the subjects. It must be the same length as the treatment variable.
treatment	vector. treatment assignment for patients, 1 for treatment group and 0 for control group
event	vector. The status indicator, normally 0=alive, 1=dead. Other choices are TRUE/FALSE (TRUE = death) or 1/2 (2=death). For censored data, the status indicator is 0=right censored, 1 = event at time. Although unusual, the event indicator can be omitted, in which case all subjects are assumed to have an event.
.data	NULL. stores the survival data for analysis, please do not fill it in.

Value

a list with time, treatment, and event with time-to-event outcome.

Examples

```
data_survival(time      = c(6.2, 8.2, 8.0, 2.3),
               treatment = c(0, 1, 0, 1),
               event     = c(1, 1, 1, 1))
```

enrollment	<i>Simulating enrollment dates</i>
------------	------------------------------------

Description

This function simulates enrollment dates using either poisson distribution

Usage

```
enrollment(param, N_total, time = NULL)
```

Arguments

param	a vector of lambda in poisson
N_total	a numeric value of total sample size
time	a vector of the length(param) - 1 indicating end of time when a specific lambda is used

Value

a vector of enrollment dates

Examples

```
enrollment(param = c(0.003, 0.7), 100, time = 10)
enrollment(param = c(0.3, 0.5, 0.9, 1.2, 2.1), 200, c(20, 30, 40, 60))
```

enrollment_rate	<i>Enrollment rate wrapper</i>
-----------------	--------------------------------

Description

Wrapper function for enrollment rate.

Usage

```
enrollment_rate(lambda = 0.3, time = NULL, .data = NULL)
```

Arguments

lambda	vector. Vector with different enrollment rate parameters.
time	vector. Vector with different cut-off times for lambda.
.data	NULL. This should not be changed by the user.

Value

a list with enrollment rate information

Examples

```
enrollment_rate(lambda = c(0.3, 1), time = 25)
```

gamma_prior	<i>Gamma prior for for control and treatment group</i>
-------------	--------------------------------------------------------

Description

Wrapper function for gamma prior $\text{Gamma}(a_0, b_0)$.

Usage

```
gamma_prior(a0 = 0.1, b0 = 0.1, .data = NULL)
```

Arguments

`a0` numeric. The shape parameter in the gamma distribution ($\text{beta}(a_0, b_0)$).

`b0` numeric. The scale parameter in the beta distribution ($\text{beta}(a_0, b_0)$).

`.data` NULL. stores the gamma prior rate, please do not fill it in.

Value

a list with vector of gamma rate for the gamma prior for treatment and control group.

Examples

```
gamma_prior(a0 = .1, b0 = .1)
```

historical_binomial	<i>Historical data for binomial distribution</i>
---------------------	--------------------------------------------------

Description

Wrapper function for historical data from binomial outcome.

Usage

```
historical_binomial(
  y0_treatment = NULL,
  N0_treatment = NULL,
  discount_function = "identity",
  y0_control = NULL,
  N0_control = NULL,
  alpha_max = 1,
  fix_alpha = FALSE,
  weibull_scale = 0.135,
  weibull_shape = 3,
  method = "fixed",
  .data = NULL
)
```

Arguments

<code>y0_treatment</code>	scalar. Number of events for the historical treatment arm.
<code>N0_treatment</code>	scalar. Number of observations of the historical treatment group.
<code>discount_function</code>	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.
<code>y0_control</code>	scalar. Number of events for the historical control arm.
<code>N0_control</code>	scalar. Number of observations of the historical control group.
<code>alpha_max</code>	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
<code>fix_alpha</code>	logical. Fix alpha at alpha_max? Default value is FALSE.
<code>weibull_scale</code>	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>weibull_shape</code>	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .

method character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the the `bdpsurvival` vignette `vignette("bdpsurvival-vignette", package="bayesDP")` for more details.

.data NULL. stores the proportion of control and treatment, please do not fill it in.

Value

a list with historical data for control and treatment group with the discount function.

Examples

```
historical_binomial(y0_treatment = 5, N0_treatment = 10, y0_control = 15, N0_control = 23)
historical_binomial(y0_treatment = 5, N0_treatment = 10, y0_control = 15, N0_control = 23,
  discount_function = "weibull", alpha_max = 1, fix_alpha = FALSE,
  weibull_scale = 0.135, weibull_shape = 3)
```

`historical_normal` *Historical data for normal distribution*

Description

Wrapper function for historical data from normal outcome.

Usage

```
historical_normal(
  mu0_treatment = NULL,
  sd0_treatment = NULL,
  N0_treatment = NULL,
  mu0_control = NULL,
  sd0_control = NULL,
  N0_control = NULL,
  discount_function = "identity",
  alpha_max = 1,
  fix_alpha = FALSE,
  weibull_scale = 0.135,
  weibull_shape = 3,
  method = "fixed",
  .data = NULL
)
```


Arguments

<code>mu0_treatment</code>	scalar. Mean of the historical treatment group.
<code>sd0_treatment</code>	scalar. Standard deviation of the historical treatment group.
<code>N0_treatment</code>	scalar. Number of observations of the historical treatment group.
<code>mu0_control</code>	scalar. Mean of the historical control group.
<code>sd0_control</code>	scalar. Standard deviation of the historical control group.
<code>N0_control</code>	scalar. Number of observations of the historical control group.
<code>discount_function</code>	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.
<code>alpha_max</code>	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
<code>fix_alpha</code>	logical. Fix alpha at alpha_max? Default value is FALSE.
<code>weibull_scale</code>	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>weibull_shape</code>	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>method</code>	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the <code>bdpsurvival</code> vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.
<code>.data</code>	NULL. stores the proportion of control and treatment, please do not fill it in.

Value

a list with historical data for control and treatment group with the discount function.

Examples

```
historical_normal(mu0_treatment = 15, sd0_treatment = 2, N0_treatment = 10,
                 mu0_control = 17, sd0_control = 3, N0_control = 20)
```

historical_survival *Historical data for survival analysis*

Description

Wrapper function for historical data from time-to-event outcome.

Usage

```
historical_survival(
  time = NULL,
  treatment = NULL,
  event = NULL,
  discount_function = "identity",
  alpha_max = 1,
  fix_alpha = FALSE,
  weibull_scale = 0.135,
  weibull_shape = 3,
  method = "fixed",
  .data = NULL
)
```

Arguments

<code>time</code>	vector. exposure time for the subjects. It must be the same length as the treatment variable.
<code>treatment</code>	vector. treatment assignment for patients, 1 for treatment group and 0 for control group
<code>event</code>	vector. The status indicator, normally 0=alive, 1=dead. Other choices are TRUE/FALSE (TRUE = death) or 1/2 (2=death). For censored data, the status indicator is 0=right censored, 1 = event at time. Although unusual, the event indicator can be omitted, in which case all subjects are assumed to have an event.
<code>discount_function</code>	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.

alpha_max	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
fix_alpha	logical. Fix alpha at alpha_max? Default value is FALSE.
weibull_scale	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when discount_function = "identity".
weibull_shape	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when discount_function = "identity".
method	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the the <code>bdpsurvival</code> vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.
.data	NULL. stores the historical time, treatment and event , please do not fill it in.

Value

a list with historical data for time-to-event outcome with the discount function.

Examples

```
historical_survival(time      = rexp(10, 0.01),
                   treatment = rep(10, 1),
                   event     = rep(10, 1))
```

hypothesis

Hypothesis wrapper

Description

Wrapper function for the hypothesis in the trial.

Usage

```
hypothesis(
  delta = 0,
  futility_prob = 0.05,
  prob_accept_ha = 0.95,
  expected_success_prob = 0.9,
  alternative = "greater",
  .data = NULL
)
```

Arguments

`delta` numeric. Threshold set for margin in null hypothesis. The default is set to 0.

`futility_prob` numeric. Probability of futility. The default is 0.05.

`prob_accept_ha` numeric. Posterior probability of accepting alternative hypothesis. The default is 0.95.

`expected_success_prob` numeric. Probability of expected success.

`alternative` character. The string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".

`.data` NULL. This should not be changed by the user.

Value

a list with information of hypothesis testing (threshold, futility probability, probability of accepting the alternative hypothesis, and probability of expected success).

Examples

```
hypothesis(delta = 0, futility_prob = 0.05, prob_accept_ha = 0.95,
           expected_success_prob = 0.90, alternative = "greater")
hypothesis(delta = 0.2, futility_prob = 0.1, prob_accept_ha = 0.975,
           expected_success_prob = 0.80, alternative = "less")
```

impute

Imputation wrapper

Description

Wrapper function for `no_of_impute`.

Usage

```
impute(no_of_impute = 10000, number_mcmc = 10000, .data = NULL)
```

Arguments

no_of_impute integer. Number of Monte Carlo imputation for missing data.
 number_mcmc scalar. Number of Monte Carlo Markov Chain draws from posterior distribution.
 .data NULL. This should not be changed by the user.

Value

a list with number of imputation

Examples

```
impute(no_of_impute = 100, number_mcmc = 1000)
```

 normalBACT

Normal distribution for Bayesian Adaptive Trials

Description

Simulation of normally distributed data for Bayesian adaptive trials with various inputs to control for power, sample size, type I error rate, etc.

Usage

```
normalBACT(
  mu_treatment,
  sd_treatment,
  mu_control = NULL,
  sd_control = NULL,
  mu0_treatment = NULL,
  sd0_treatment = NULL,
  N0_treatment = NULL,
  mu0_control = NULL,
  sd0_control = NULL,
  N0_control = NULL,
  N_total,
  lambda = 0.3,
  lambda_time = NULL,
  interim_look = NULL,
  EndofStudy,
  block = 2,
  rand_ratio = c(1, 1),
  discount_function = "identity",
  alternative = "greater",
  prop_loss_to_followup = 0.15,
  h0 = 0,
  futility_prob = 0.05,
```

```

    expected_success_prob = 0.9,
    prob_ha = 0.95,
    N_impute = 10,
    number_mcmc = 10000,
    alpha_max = 1,
    fix_alpha = FALSE,
    weibull_scale = 0.135,
    weibull_shape = 3,
    method = "fixed"
)

```

Arguments

<code>mu_treatment</code>	scalar. Mean outcome in the treatment arm.
<code>sd_treatment</code>	scalar. Standard deviation of outcome in the treatment
<code>mu_control</code>	scalar. Mean outcome in the control arm.
<code>sd_control</code>	scalar. Standard deviation of outcome in the control arm. arm.
<code>mu0_treatment</code>	scalar. Mean of the historical treatment group.
<code>sd0_treatment</code>	scalar. Standard deviation of the historical treatment group.
<code>N0_treatment</code>	scalar. Number of observations of the historical treatment group.
<code>mu0_control</code>	scalar. Mean of the historical control group.
<code>sd0_control</code>	scalar. Standard deviation of the historical control group.
<code>N0_control</code>	scalar. Number of observations of the historical control group.
<code>N_total</code>	scalar. Total sample size.
<code>lambda</code>	vector. Enrollment rates across simulated enrollment times. See enrollment for more details.
<code>lambda_time</code>	vector. Enrollment time(s) at which the enrollment rates change. Must be same length as <code>lambda</code> . See enrollment for more details.
<code>interim_look</code>	vector. Sample size for each interim look. Note: the maximum sample size should not be included.
<code>EndofStudy</code>	scalar. Length of the study.
<code>block</code>	scalar. Block size for generating the randomization schedule.
<code>rand_ratio</code>	vector. Randomization allocation for the ratio of control to treatment. Integer values mapping the size of the block. See randomization for more details.
<code>discount_function</code>	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is <code>"identity"</code> . See bdpnormal for more details.
<code>alternative</code>	character. The string specifying the alternative hypothesis, must be one of <code>"greater"</code> (default), <code>"less"</code> or <code>"two.sided"</code> .

prop_loss_to_followup	scalar. Overall proportion of subjects lost to follow-up.
h0	scalar. Threshold for comparing two mean values. Default is $h_0=0$.
futility_prob	scalar. Probability of stopping early for futility.
expected_success_prob	scalar. Probability of stopping early for success.
prob_ha	scalar. Probability of alternative hypothesis.
N_impute	scalar. Number of imputations for Monte Carlo simulation of missing data.
number_mcmc	scalar. Number of Monte Carlo Markov Chain draws in sampling posterior.
alpha_max	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
fix_alpha	logical. Fix alpha at alpha_max? Default value is FALSE.
weibull_scale	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when discount_function = "identity".
weibull_shape	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when discount_function = "identity".
method	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the the <code>bdpsurvival</code> vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.

Value

a list of output for a single trial simulation.

`mu_treatment` scalar. The input parameter of mean value of the outcome in the treatment group.

`p_control` scalar. The input parameter of mean value of the outcome in the control group.

`sd_treatment` scalar. The input parameter of standard deviation of the outcome in the control group.

`sd_control` scalar. The input parameter of standard deviation of the outcome in the control group.

`prob_of_accepting_alternative` scalar. The input parameter of probability threshold of accepting the alternative.

`margin` scalar. The margin input value of difference between mean estimate of treatment and mean estimate of the control.

alternative character. The input parameter of alternative hypothesis.
interim_look vector. The sample size for each interim look.
N_treatment scalar. The number of patients enrolled in the experimental group for each simulation.
N_control scalar. The number of patients enrolled in the control group for each simulation.
N_enrolled vector. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)
N_complete scalar. The number of patients who completed the trial and had no loss to follow-up.
post_prob_accept_alternative vector. The final probability of accepting the alternative hypothesis after the analysis is done.
est_final scalar. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.
stop_futility scalar. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.
stop_expected_success scalar. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.
est_interim scalar. The interim estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.

 normaldata

Gaussian dataset for analyzing adaptive Bayesian trials

Description

A dataset containing the results of 300 patients with continuous (normal) outcome, the dataset is filled with loss to follow up.

Usage

```
data(normaldata)
```

Format

A data frame with 300 rows and 4 variables:

id Patient ID in the trial

treatment treatment assignment for patients, 1 for treatment group 0 for control group

outcome continuous outcome of the trial (gaussian distribution)

complete 1 for complete outcome, 0 for loss to follow up

Examples

```
data(normaldata)
```

normal_analysis	<i>Analyzing Bayesian trial for normal mean data</i>
-----------------	------------------------------------------------------

Description

Function to analyze Bayesian trial for normal mean data which allows early stopping and incorporation of historical data using the discount function approach

Usage

```
normal_analysis(  
  treatment,  
  outcome,  
  complete = NULL,  
  mu0_treatment = NULL,  
  sd0_treatment = NULL,  
  N0_treatment = NULL,  
  mu0_control = NULL,  
  sd0_control = NULL,  
  N0_control = NULL,  
  alternative = "greater",  
  N_impute = 100,  
  h0 = 0,  
  number_mcmc = 10000,  
  prob_ha = 0.95,  
  futility_prob = 0.1,  
  expected_success_prob = 0.9,  
  discount_function = "identity",  
  fix_alpha = FALSE,  
  alpha_max = 1,  
  weibull_scale = 0.135,  
  weibull_shape = 3,  
  method = "fixed"  
)
```

Arguments

treatment	vector. treatment assignment for patients, 1 for treatment group and 0 for control group
outcome	vector. normal outcome of the trial.
complete	vector. similar length as treatment and outcome variable, 1 for complete outcome, 0 for loss to follow up. If complete is not provided, the dataset is assumed to be complete.
mu0_treatment	scalar. Mean of the historical treatment group.
sd0_treatment	scalar. Standard deviation of the historical treatment group.

<code>N0_treatment</code>	scalar. Number of observations of the historical treatment group.
<code>mu0_control</code>	scalar. Mean of the historical control group.
<code>sd0_control</code>	scalar. Standard deviation of the historical control group.
<code>N0_control</code>	scalar. Number of observations of the historical control group.
<code>alternative</code>	character. The string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".
<code>N_impute</code>	scalar. Number of imputations for Monte Carlo simulation of missing data.
<code>h0</code>	scalar. Threshold for comparing two mean values. Default is $h0=0$.
<code>number_mcmc</code>	scalar. Number of Monte Carlo Markov Chain draws in sampling posterior.
<code>prob_ha</code>	scalar. Probability of alternative hypothesis.
<code>futility_prob</code>	scalar. Probability of stopping early for futility.
<code>expected_success_prob</code>	scalar. Probability of stopping early for success.
<code>discount_function</code>	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.
<code>fix_alpha</code>	logical. Fix alpha at <code>alpha_max</code> ? Default value is FALSE.
<code>alpha_max</code>	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
<code>weibull_scale</code>	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>weibull_shape</code>	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>method</code>	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the bdpsurvival vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.

Value

a list of output for the analysis of Bayesian trial for normal mean.

prob_of_accepting_alternative scalar. The input parameter of probability of accepting the alternative.

margin scalar. The margin input value of difference between mean estimate of treatment and mean estimate of the control.

alternative character. The input parameter of alternative hypothesis.

N_treatment scalar. The number of patients enrolled in the experimental group for each simulation.

N_control scalar. The number of patients enrolled in the control group for each simulation.

N_enrolled vector. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)

N_complete scalar. The number of patients who completed the trial and had no loss to follow-up.

post_prob_accept_alternative vector. The final probability of accepting the alternative hypothesis after the analysis is done.

est_final scalar. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.

stop_futility scalar. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

stop_expected_success scalar. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

normal_outcome

Parameters for treatment and control in normal case

Description

Wrapper function for mean and standard deviation with normal outcome.

Usage

```
normal_outcome(
  mu_control = NULL,
  sd_control = NULL,
  mu_treatment = NULL,
  sd_treatment = NULL,
  data = NULL
)
```

Arguments

mu_control	numeric. The mean for the control group.
sd_control	numeric. The standard deviation for the control group.
mu_treatment	numeric. The mean for the treatment group.
sd_treatment	numeric. The standard deviation for the treatment group.
data	NULL. stores the proportion of control and treatment, please do not fill it in.

Value

a list with proportion of control and treatment group.

Examples

```
normal_outcome(mu_control = 12, mu_treatment = 8, sd_treatment = 2.2, sd_control = 1.6)
```

pw_exp_impute	<i>Imputes time-to-event outcomes.</i>
---------------	----------------------------------------

Description

Imputation of time-to-event outcomes using the piecewise constant hazard exponential function.

Usage

```
pw_exp_impute(time, hazard, maxtime = NULL, cutpoint = NULL)
```

Arguments

time	vector. The observed time for patient that have had no event or passed maxtime.
hazard	vector. The constant hazard rates for exponential failures.
maxtime	scalar. maximum time before end of study.
cutpoint	vector. The change-point vector indicating time when the hazard rates change.

Value

a dataset with simulated follow-up time (time) and respective event indicator (1 = event, 0 = censoring)

Examples

```
pw_exp_impute(time = c(120), c(0.005, 0.001), 110, 40)
pw_exp_impute(time = c(10, 20, 30), c(0.005, 0.01, 0.02), 100, c(40, 80))
pw_exp_impute(time = c(40, 30), c(0.005, 0.01), 120, c(50))
```

pw_exp_sim *Simulates time-to-event outcomes.*

Description

Simulation of time-to-event outcomes using the piecewise constant hazard exponential function.

Usage

```
pw_exp_sim(hazard, n, maxtime = NULL, cutpoint = NULL)
```

Arguments

hazard	vector. The constant hazard rates for exponential failures.
n	scalar. The number of outcomes for simulation.
maxtime	scalar. maximum time before end of study.
cutpoint	vector. The change-point vector indicating time when the hazard rates change.

Value

a dataset with simulated follow-up time (time) and respective event indicator (1 = event, 0 = censoring)

Examples

```
pw_exp_sim(c(0.02, 0.01, 0.005), 100, 100, c(10, 20))
pw_exp_sim(0.015, 100, 100)
```

randomization *Randomization allocation*

Description

Implements a randomization allocation for control and treatment arms with different randomization ratios and block sizes.

Usage

```
randomization(N_total, block = 2, allocation = c(1, 1))
```

Arguments

`N_total` an integer value of total sample size for randomization allocation.
`block` a vector value of the block size for randomization. Note that it needs to be a multiple of the sum of allocation.
`allocation` a numeric vector of the randomization allocation in the order `c(control, treatment)`.

Value

the randomization allocation with 0, 1 for control and treatment

Examples

```
# Implementing treatment allocation for control to treatment with 1:1.5 randomization ratio
randomization(N_total = 100, block = 5, allocation = c(2, 3))

# Treatment allocation with 2:1 for control to treatment
randomization(N_total = 70, block = 9, allocation = c(2, 1))

# Treatment allocation for control to treatment with 1:2 for control to treatment with
# multiple block sizes c(3, 9, 6)
randomization(N_total = 100, block = c(3, 9, 6), allocation = c(1, 2))

# For complete randomization set the N_total to block size
randomization(N_total = 100, block = 100, allocation = c(1, 1))
```

randomize

Randomization scheme wrapper

Description

Wrapper function for the randomization scheme in the trial.

Usage

```
randomize(block_size = 2, randomization_ratio = c(1, 1), .data = NULL)
```

Arguments

`block_size` integer. Block size for the complete randomization in a block.
`randomization_ratio` vector. The randomization allocation for control to treatment.
`.data` NULL. This should not be changed by the user.

Value

a list with randomization details (block size and ratio).

Examples

```
randomize(block_size = 100, randomization_ratio = c(2, 3))
randomize(block_size = 10, randomization_ratio = c(1, 4))
```

simulate	<i>Simulation wrapper for binomial and normal.</i>
----------	----------------------------------------------------

Description

Wrapper function for complete binomial and normal function to compute power and type 1 error.

Usage

```
simulate(input, no_of_sim = 10000, .data = NULL)
```

Arguments

input	list. Input function for all inputs in binomial and normal .
no_of_sim	numeric. Number of simulations to run
.data	NULL. stores the proportion of control and treatment, please do not fill it in.

Value

a list with results of the simulation (power and type I error) and the input.

input A list of input values used in the trial simulation.

power_data_frame. A data frame with the interim look and power at each look.

type1_error scalar. The type 1 error or the number of times the trial rejects the null when the parameters are simulated under the null hypothesis.

est_final vector. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group for all the simulation.

post_prob_accept_alternative vector. The final probability of accepting the alternative for the simulations.

N_enrolled vector. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)

stop_futility vector. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

stop_expected_success vector. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

study_details	<i>Details of the clinical study</i>
---------------	--------------------------------------

Description

Wrapper function for details of the clinical trial simulation.

Usage

```
study_details(
  total_sample_size,
  study_period,
  interim_look = NULL,
  prop_loss_to_followup = 0.1,
  .data = NULL
)
```

Arguments

total_sample_size	integer. The number of sample size needed.
study_period	integer. The length of the study.
interim_look	vector. Vector with interim looks.
prop_loss_to_followup	integer. The proportion of loss to follow up.
.data	NULL. This should not be changed by the user.

Value

a list with sample size, length of the study, interim looks and proportion loss to follow up

Examples

```
study_details(total_sample_size = 300, study_period = 50, interim_look = c(210, 240, 270))
```

survivalBACT	<i>Time-to-event outcome for Bayesian Adaptive Trials</i>
--------------	-----------------------------------------------------------

Description

Simulation for time-to-event outcome for Bayesian Adaptive trial with different inputs to control for power, sample size, type 1 error rate, etc.

Usage

```

survivalBACT(
  hazard_treatment,
  cutpoint = NULL,
  hazard_control = NULL,
  N_total,
  breaks = NULL,
  time0 = NULL,
  treatment0 = NULL,
  event0 = NULL,
  lambda = 0.3,
  lambda_time = NULL,
  interim_look = NULL,
  EndofStudy,
  prior = c(0.1, 0.1),
  block = 2,
  rand_ratio = c(1, 1),
  prop_loss_to_followup = 0.1,
  alternative = "greater",
  h0 = 0,
  futility_prob = 0.05,
  expected_success_prob = 0.9,
  prob_ha = 0.95,
  N_impute = 10,
  number_mcmc = 10000,
  discount_function = "identity",
  alpha_max = 1,
  fix_alpha = FALSE,
  weibull_scale = 0.135,
  weibull_shape = 3,
  method = "fixed"
)

```

Arguments

hazard_treatment	vector. Constant hazard rates under the treatment arm.
cutpoint	vector. The change-point vector indicating time when the hazard rates change.
hazard_control	vector. Constant hazard rates under the control arm.
N_total	scalar. Total sample size.
breaks	vector. Breaks (interval starts) used to compose the breaks of the piecewise exponential model. Do not include zero. Default breaks are the quantiles of the input times.
time0	vector. Historical exposure time for the subjects. It must be the same length as the treatment variable.
treatment0	vector. the historical treatment assignment for patients, 1 for treatment group and 0 for control group.

event0	vector. Historical status indicator, normally 0=alive, 1=dead. Other choices are TRUE/FALSE (TRUE = death) or 1/2 (2=death). For censored data, the status indicator is 0=right censored, 1 = event at time. Although unusual, the event indicator can be omitted, in which case all subjects are assumed to have an event.
lambda	vector. Enrollment rates across simulated enrollment times. See enrollment for more details.
lambda_time	vector. Enrollment time(s) at which the enrollment rates change. Must be same length as lambda. See enrollment for more details.
interim_look	vector. Sample size for each interim look. Note: the maximum sample size should not be included.
EndofStudy	scalar. Length of the study.
prior	vector. Prior values of the gamma rate, Gamma(a0, b0). The default is set to Gamma(.1, .1).
block	scalar. Block size for generating the randomization schedule.
rand_ratio	vector. Randomization allocation for the ratio of control to treatment. Integer values mapping the size of the block. See randomization for more details.
prop_loss_to_followup	scalar. Overall proportion of subjects lost to follow-up.
alternative	character. The string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".
h0	scalar. Threshold for comparing two mean values. Default is $h_0=0$.
futility_prob	scalar. Probability of stopping early for futility.
expected_success_prob	scalar. Probability of stopping early for success.
prob_ha	scalar. Probability of alternative hypothesis.
N_impute	scalar. Number of imputations for Monte Carlo simulation of missing data.
number_mcmc	scalar. Number of Monte Carlo Markov Chain draws in sampling posterior.
discount_function	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.
alpha_max	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
fix_alpha	logical. Fix alpha at alpha_max? Default value is FALSE.

<code>weibull_scale</code>	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>weibull_shape</code>	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>method</code>	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the <code>bdpsurvival</code> vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.

Value

a list of output for a single trial simulation.

`lambda_treatment` vector. The input parameter of constant hazard rates in the treatment group.

`cutpoint_treatment` vector. The change-point vector when the constant hazard rate(s) changes for the treatment group.

`lambda_control` vector. The input parameter of constant hazard rates in the control group.

`cutpoint_control` vector. The change-point vector when the constant hazard rate(s) changes for the control group.

`prob_of_accepting_alternative` scalar. The input parameter of probability threshold of accepting the alternative.

`margin` scalar. The margin input value of difference between mean estimate of treatment and mean estimate of the control.

`alternative` character. The input parameter of alternative hypothesis.

`interim_look` vector. The sample size for each interim look.

`N_treatment` scalar. The number of patients enrolled in the treatment group for each simulation.

`event_treatment` scalar. The number of events in the treatment group for each simulation.

`N_control` scalar. The number of patients enrolled in the control group for each simulation.

`event_control` scalar. The number of events in the control group for each simulation.

`N_enrolled` vector. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)

`N_complete` scalar. The number of patients who completed the trial and had no loss to follow-up.

`post_prob_accept_alternative` vector. The final probability of accepting the alternative hypothesis after the analysis is done.

`est_final` scalar. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.

`stop_futility` scalar. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

`stop_expected_success` scalar. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

`est_interim` scalar. The interim estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.

survivaldata

Time-to-event dataset for analyzing adaptive Bayesian trials

Description

A dataset containing the results of 100 patients with time-to-event outcome, the dataset is filled with treatment assignment and status (0 = censored, 1 = not censored).

Usage

```
data(survivaldata)
```

Format

A data frame with 100 rows and 4 variables:

id Patient ID in the trial

treatment treatment assignment for patients, 1 for treatment group 0 for control group

time the follow up time for patients

event The status indicator, normally 0=alive, 1=dead or 0 = no event, 1 = event occurred

Examples

```
data(survivaldata)
```

survival_analysis *Analyzing Bayesian trial for time-to-event data*

Description

Function to analyze Bayesian trial for time-to-event data which allows early stopping and incorporation of historical data using the discount function approach

Usage

```
survival_analysis(  
  time,  
  treatment,  
  event = NULL,  
  time0 = NULL,  
  treatment0 = NULL,  
  event0 = NULL,  
  surv_time = NULL,  
  h0 = 0,  
  breaks = NULL,  
  alternative = "greater",  
  N_impute = 10,  
  number_mcmc = 10000,  
  prob_ha = 0.95,  
  futility_prob = 0.1,  
  expected_success_prob = 0.9,  
  prior = c(0.1, 0.1),  
  discount_function = "identity",  
  fix_alpha = FALSE,  
  alpha_max = 1,  
  weibull_scale = 0.135,  
  weibull_shape = 3,  
  method = "fixed"  
)
```

Arguments

time	vector. exposure time for the subjects. It must be the same length as the treatment variable.
treatment	vector. treatment assignment for patients, 1 for treatment group and 0 for control group
event	vector. The status indicator, normally 0=alive, 1=dead. Other choices are TRUE/FALSE (TRUE = death) or 1/2 (2=death). For censored data, the status indicator is 0=right censored, 1 = event at time. Although unusual, the event indicator can be omitted, in which case all subjects are assumed to have an event.

time0	vector. Historical exposure time for the subjects. It must be the same length as the treatment variable.
treatment0	vector. the historical treatment assignment for patients, 1 for treatment group and 0 for control group.
event0	vector. Historical status indicator, normally 0=alive, 1=dead. Other choices are TRUE/FALSE (TRUE = death) or 1/2 (2=death). For censored data, the status indicator is 0=right censored, 1 = event at time. Although unusual, the event indicator can be omitted, in which case all subjects are assumed to have an event.
surv_time	scalar. scalar. Survival time of interest for computing the probability of survival for a single arm (OPC) trial. Default is overall, i.e., current+historical, median survival time.
h0	scalar. Threshold for comparing two mean values. Default is $h_0=0$.
breaks	vector. Breaks (interval starts) used to compose the breaks of the piecewise exponential model. Do not include zero. Default breaks are the quantiles of the input times.
alternative	character. The string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".
N_impute	scalar. Number of imputations for Monte Carlo simulation of missing data.
number_mcmc	scalar. Number of Monte Carlo Markov Chain draws in sampling posterior.
prob_ha	scalar. Probability of alternative hypothesis.
futility_prob	scalar. Probability of stopping early for futility.
expected_success_prob	scalar. Probability of stopping early for success.
prior	vector. Prior values of the gamma rate, $\text{Gamma}(a_0, b_0)$. The default is set to $\text{Gamma}(.1, .1)$.
discount_function	character. If incorporating historical data, specify the discount function. Currently supports the Weibull function (<code>discount_function="weibull"</code>), the scaled-Weibull function (<code>discount_function="scaledweibull"</code>), and the identity function (<code>discount_function="identity"</code>). The scaled-Weibull discount function scales the output of the Weibull CDF to have a max value of 1. The identity discount function uses the posterior probability directly as the discount weight. Default value is "identity". See bdpnormal for more details.
fix_alpha	logical. Fix alpha at alpha_max? Default value is FALSE.
alpha_max	scalar. Maximum weight the discount function can apply. Default is 1. For a two-arm trial, users may specify a vector of two values where the first value is used to weight the historical treatment group and the second value is used to weight the historical control group.
weibull_scale	scalar. Scale parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 0.135. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .

<code>weibull_shape</code>	scalar. Shape parameter of the Weibull discount function used to compute alpha, the weight parameter of the historical data. Default value is 3. For a two-arm trial, users may specify a vector of two values where the first value is used to estimate the weight of the historical treatment group and the second value is used to estimate the weight of the historical control group. Not used when <code>discount_function = "identity"</code> .
<code>method</code>	character. Analysis method with respect to estimation of the weight parameter alpha. Default method "mc" estimates alpha for each Monte Carlo iteration. Alternate value "fixed" estimates alpha once and holds it fixed throughout the analysis. See the <code>bdpsurvival</code> vignette <code>vignette("bdpsurvival-vignette", package="bayesDP")</code> for more details.

Value

	a list of output for the Bayesian trial for time-to-event.
<code>prob_of_accepting_alternative</code>	scalar. The input parameter of probability of accepting the alternative.
<code>margin</code>	scalar. The margin input value of difference between mean estimate of treatment and mean estimate of the control.
<code>alternative</code>	character. The input parameter of alternative hypothesis.
<code>alpha_max</code>	scalar. The <code>alpha_max</code> input.
<code>N_treatment</code>	scalar. The number of patients enrolled in the experimental group for each simulation.
<code>event_treatment</code>	scalar. The number of events in the experimental group for each simulation.
<code>N_control</code>	scalar. The number of patients enrolled in the control group for each simulation.
<code>event_control</code>	scalar. The number of events in the control group for each simulation.
<code>N_enrolled</code>	scalar. The number of patients enrolled in the trial (sum of control and experimental group for each simulation.)
<code>N_complete</code>	scalar. The number of patients whose time passes the <code>surv_time</code> .
<code>alpha_discount</code>	vector. The alpha discount function used for control and treatment.
<code>post_prob_accept_alternative</code>	vector. The final probability of accepting the alternative hypothesis after the analysis is done.
<code>est_final</code>	scalar. The final estimate of the difference in posterior estimate of treatment and posterior estimate of the control group.
<code>stop_futility</code>	scalar. Did the trial stop for futility during imputation of patient who had loss to follow up? 1 for yes and 0 for no.
<code>stop_expected_success</code>	scalar. Did the trial stop for early success during imputation of patient who had loss to follow up? 1 for yes and 0 for no.

survival_outcome	<i>Piecewise constant hazard rates and the cutpoint for control and treatment group</i>
------------------	-----------------------------------------------------------------------------------------

Description

Wrapper function for the piecewise constant hazard rates and the cutpoint for control and treatment group.

Usage

```
survival_outcome(  
  hazard_treatment = NULL,  
  cutpoint = NULL,  
  hazard_control = NULL,  
  .data = NULL  
)
```

Arguments

`hazard_treatment` vector. Constant hazard rates under the treatment arm.

`cutpoint` vector. The change-point vector indicating time when the hazard rates change.

`hazard_control` vector. Constant hazard rates under the control arm.

`.data` NULL. stores the hazard rates and cutpoint, please do not fill it in.

Value

a list with hazard rates and cutpoint for control and treatment group.

Examples

```
survival_outcome(hazard_treatment = 0.06,  
                 hazard_control   = 0.08,  
                 cutpoint         = NULL )
```


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