Package 'BRACE'

July 16, 2022

mcmc_theta . prop_best_arm RuleI RuleII RuleIII sim_osfd_data	MCMC function to generate posterior of \$\text{heta}\$	
mcmc_theta . prop_best_arm RuleI RuleII RuleIII sim_osfd_data success_prob		
mcmc_theta . prop_best_arm RuleI RuleII RuleIII sim_osfd_data		
mcmc_theta . prop_best_arm RuleI RuleII RuleIII		
mcmc_theta . prop_best_arm RuleI RuleII		
mcmc_theta . prop_best_arm RuleI		
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	runanarm	
ity, e.g. the endpoint of organ support-free days (OSFD) for sepsis patients admitted to the inte sive care units (ICU). The package proposes a Bayesian mixture model to model the composite endpoint of OSFD and implements three response adaptive randomization (RAR) adaptation rules for patient allocation based on the estimated posterior probability from the Bayesian model at the interim analysis. License GPL (>= 2)		
Description Some clinic	cal trials have a composite endpoint of both mortality and morbid-	
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Endpoint (BRACE Version 1.0	e Adaptive randomization design for Composite ().	

Run MCMC and generate posterior samples of the estimand (mean transformed OSFD \$\theta\$)

prop_best_arm

Usage

```
mcmc_theta(d, tau, B = 2000, burnin = 500, thin = 10)
```

Arguments

d Data input of transformed OSFD tau Data input of death indicator

B Number of MCMC iterations burnin Number of burn-in iterations

thin A value thin = k would result in keeping every kth value and discarding all other

values of the MCMC chain.

Value

A vector of \$\theta\$ values of all kept MCMC samples.

Examples

```
## Not run:
theta_ctrl <- mcmc_theta(data_ctrl$d,data_ctrl$tau)
## End(Not run)</pre>
```

prop_best_arm

Calculate the proportion of patients assigned to the best arm

Description

Calculate the proportion of patients assigned to the best arm in response-adaptive randomization (RAR) design.

Usage

```
prop_best_arm(data_all, data_best_arm)
```

Arguments

```
data_all A list of dataframes from all arms data_best_arm A dataframe from the best arm
```

Value

The proportion of patients assigned to the best arm, ranging from 0 to 1.

Examples

```
## Not run:
    prop_best_arm(data_all,data_best_arm)
## End(Not run)
```

RuleI 3

Description

Calculate the number of patients allocated to each arm at the interim stage following BRACE RAR Rule I.

Usage

```
RuleI(nj, n0, pvec)
```

Arguments

nj Total sample size at jth interim stage
 n0 Sample size for the control arm
 pvec Vector of allocation probability for all treatment arms

Value

A vector of number of patients allocated to each arm (for 4 arms scenario only).

Examples

```
nj <- 200
ctrl_prop <- 1/4
pvec <- c(0.8,0.1,0.1)
n_assign <- RuleI(nj=nj,n0=nj*ctrl_prop,pvec=pvec)</pre>
```

RuleII

BRACE RAR Rule II

Description

Calculate the number of patients allocated to each arm at the interim stage following BRACE RAR Rule II.

Usage

```
RuleII(nj, pvec)
```

Arguments

nj Total sample size at jth interim stage

pvec Vector of allocation probability for all treatment arms

Value

A vector of number of patients allocated to each arm.

4 sim_osfd_data

Examples

```
nj <- 200
pvec <- c(0.8,0.1,0.1)
n_assign <- RuleII(nj=nj,pvec=pvec)</pre>
```

RuleIII

BRACE RAR Rule III

Description

Calculate the number of patients allocated to each arm at the interim stage following BRACE RAR Rule III.

Usage

```
RuleIII(nj, pvec)
```

Arguments

nj Total sample size at jth interim stage

pvec Vector of allocation probability for all treatment arms

Value

A vector of number of patients allocated to each arm.

Examples

```
nj <- 200
pvec <- c(0.8,0.1,0.1)
n_assign <- RuleIII(nj=nj,pvec=pvec)</pre>
```

sim_osfd_data

Simulate organ support-free days (OSFD) data

Description

Simulate organ support-free days (OSFD) data with given sample size and parameter values.

Usage

```
sim_osfd_data(n, mu, sigma2, omega, lambda)
```

Arguments

n	Sample size

mu Mean of transformed OSFD among non-censored patients sigma2 Variance of transformed OSFD among non-censored patients

omega Censoring prportion of patients

lambda Mortality rate

success_prob 5

Value

Simulated data consisting of OSFD on original scale (y), death indicator (\$\tau\$) and transformed OSFD (d).

Examples

```
n0 <- 50
mean_osfd <- 20
mu_ctrl_true <- -log(30-mean_osfd)
sigma2_true <- 0.8^2
omega_true <- 0.3
lambda_ctrl_true <- 0.2
data_ctrl <- sim_osfd_data(n=n0, mu=mu_ctrl_true, sigma2=sigma2_true, omega=omega_true, lambda=lambda_ctrl_true)</pre>
```

success_prob

Success probability of the trial

Description

Calculate the success probability of the trial, defined as the probability of difference in estimand (mean transformed OSFD \$\theta\$) between the best treatment arm and control arm passing a clinically meaningful difference.

Usage

```
success_prob(theta_best, theta_ctrl, clin_diff)
```

Arguments

theta_best The posterior samples of \$\theta\$ in the best treatment arm
theta_ctrl The posterior samples of \$\theta\$ in the control arm
clin_diff Clinically meaningful difference

Value

Success probaility of the trial, ranging from 0 to 1.

Examples

```
## Not run:
    clin_diff <- 0.0647669
    success_prob(theta_best,theta_ctrl,clin_diff)
## End(Not run)</pre>
```

Index

```
*Topic \textasciitildekwd1
prop_best_arm, 2
*Topic \textasciitildekwd2
prop_best_arm, 2

mcmc_theta, 1
prop_best_arm, 2

RuleI, 3
RuleII, 3
RuleIII, 4

sim_osfd_data, 4
success_prob, 5
```