Package: BCHM (via r-universe)

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Title Clinical Trial Calculation Based on BCHM Design

Version 1.00

Author Nan Chen and J. Jack Lee

Description Users can estimate the treatment effect for multiple subgroups basket trials based on the Bayesian Cluster Hierarchical Model (BCHM). In this model, a Bayesian non-parametric method is applied to dynamically calculate the number of clusters by conducting the multiple cluster classification based on subgroup outcomes. Hierarchical model is used to compute the posterior probability of treatment effect with the borrowing strength determined by the Bayesian non-parametric clustering and the similarities between subgroups. To use this package, 'JAGS' software and 'rjags' package are required, and users need to pre-install them.

Depends R (>= 3.5), rjags

License LGPL-2

Maintainer J. Jack Lee <jjlee@mdanderson.org>

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Suggests testthat

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Description

The BCHM computation is based on the Bayesian Cluster Hierarchical Model (BCHM) to perform the non-parametric Bayesian clustering and posterior probability calculation with the Bayesian hierarchical model for binary response data in several subgroups. Due to the potential heterogeneity among subgroups, the exchangeability assumption across all subgroups may not hold. A Bayesian non-parametric method is applied to calculate the number of clusters by conducting the multiple cluster classification based on subgroup outcomes. Due to the MCMC sampling, the clustering result is dynamic. A cluster matrix (Similarity Matrix) is constructed to depict the strength of association between any two subgroups to be classified into the same cluster. The Bayesian hierarchical model is used to compute the posterior probability of treatment effect with the borrowing strength determined by the similarity matrix values calculated from the Bayesian non-parametric clustering.

Usage

```
BCHM(
  nDat,
  xDat.
 mu = 0.2,
  sigma02 = 10,
  sigmaD2 = 0.001,
  alpha = 1e-60,
  d0 = 0.05
  alpha1 = 50,
  beta1 = 10,
  tau2 = 0.1,
  phi1 = 0.1,
  deltaT = 0.05,
  thetaT = 0.6.
  burnIn = 10000,
 MCIter = 20000.
 MCNum = 20000,
  seed = 1000
)
```

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Arguments

nDat Number of patients in each subgroup.

xDat Number of responses in each subgroup.

mu Hyperprior mean for the cluster.

sigma02 Hyperprior variance for the cluster.
sigmaD2 Variance of subgroup response rate.

alpha Alpha value of the Dirichlet Process determining number of clusters.

d0 Minimum value for the similarity matrix.

alpha1 Prior for borrowing strength gamma(alpha1, beta1) in the hierarchical model.

beta1 Prior for borrowing strength gamma(alpha1, beta1) in the hierarchical model.

tau2 Hyperprior precision parameter of subgroup means in the hierarchical model

phi1 The response rate of the standard treatment.

deltaT The expected improvement in the response rate over the standard treatment.

thetaT Threshold value for the inference claiming efficacy.

burnIn Number of burn_in in MCMC.

MCIter Number of MCMC iterations.

MCNum Number of MCMC iterations in the hierarchical model.

seed Random seed.

Value

The return is a list including three elements: Samples, SMatrix, and Result.

The return list element Samples is the sampled posterior response rates of all subgroups.

The return list element SMatrix is the similarity matrix of all subgroups.

The return list element Result is the calculation results of all subgroups. It has seven columns: the number of responses of each subgroups, the number of patients in each subgroups, the observe response rates of each subgroups, the cluster index of each subgroups, the posterior mean response rates of each subgroups, the probability of Pr(P>Phi1+deltaT) of each subgroups, and the final decision (0: Not rejected the null, 1 Rejected the null). Note: Because a MCMC computation is applied in the clustering calculation, there are many possible clustering configurations. The cluster index in column 4 is the most possible clustering configuration.

See Also

BCHMplot_cluster Plot the clustering results of subgroups.

BCHMplot_post_value Plot the posterior response of subgroups.

BCHMplot_post_dist Plot the posterior distributions of subgroups.

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Examples

```
nDat = c(25, 25, 25, 25) \# total number of patients
xDat = c(2, 3, 8, 6) # number of responses
alpha <- 1e-20
d0 <- 0.0
alpha1 = 50
beta1 = 10
tau2 <- 0.1
phi1 <- 0.1
deltaT <- 0.2
thetaT <- 0.60
res <- BCHM(nDat = nDat,
            xDat = xDat,
            alpha = alpha,
            d0 = d0,
            alpha1 = alpha1,
            beta1 = beta1,
            tau2 = tau2,
            phi1 = phi1,
            deltaT = deltaT,
            thetaT = thetaT,
            burnIn = 100,
            MCIter = 200,
            MCNum = 1000,
            seed = 1000
)
print(res$SMatrix)
print(res$Result)
col <- res$Result[,4]</pre>
BCHMplot_cluster(res, col, pch=16)
BCHMplot_post_value(res, col, HPD = 0.8)
BCHMplot_post_dist(res, col, lty=1:length(nDat), lwd =3, xlim=c(0, 0.8))
```

BCHMplot_cluster

Plot the clustering results of subgroups.

Description

plot the observed response rate versus subgroup ID with clusters coded by the color of dots.

Usage

```
BCHMplot_cluster(
  res,
  col = res$Result[, 4],
  pch = 16,
```

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```
xlim = c(0, dim(res$Result)[1] + 2),
ylim = c(0, 1),
cex = 2,
...
)
```

Arguments

res	BCHM calculation results.
col	Color vector
pch	pch vector
xlim	X-axis range
ylim	Y-axis range
cex	size of points
	other options

Value

None

See Also

```
BCHM Perform the analysis based on the BCHM design.

BCHMplot_post_value Plot the posterior response of subgroups.

BCHMplot_post_dist Plot the posterior distributions of subgroups.
```

Examples

```
nDat = c(25, 25, 25, 25) # total number of patients
xDat = c(2, 3, 8, 6) # number of responses
alpha <- 1e-20
d0 <- 0.0
alpha1 = 50
beta1 = 10
tau2 <- 0.1
phi1 <- 0.1
deltaT <- 0.2
thetaT <- 0.60
res <- BCHM(nDat = nDat,
            xDat = xDat,
            alpha = alpha,
            d0 = d0,
            alpha1 = alpha1,
            beta1 = beta1,
            tau2 = tau2,
            phi1 = phi1,
            deltaT = deltaT,
            thetaT = thetaT,
```

BCHMplot_post_dist

BCHMplot_post_dist

Plot the posterior distributions of subgroups.

Description

plot the posterior distribution by subgroup ID

Usage

```
BCHMplot_post_dist(
    res,
    col = res$Result[, 4],
    lty = 1:dim(res$Result)[1],
    lwd = 2,
    xlim = c(0, 1),
    ylim = c(0, 20),
    ...
)
```

Arguments

res	BCHM calculation results.
col	Color vector
lty	line types
lwd	line width
xlim	X-axis range
ylim	Y-axis range
	other options

Value

None

See Also

```
BCHM Perform the analysis based on the BCHM design.

BCHMplot_cluster Plot the clustering results of subgroups.

BCHMplot_post_value Plot the posterior response of subgroups.
```

Examples

```
nDat = c(25, 25, 25, 25) # total number of patients
xDat = c(2, 3, 8, 6) # number of responses
alpha <- 1e-20
d0 <- 0.0
alpha1 = 50
beta1 = 10
tau2 <- 0.1
phi1 <- 0.1
deltaT <- 0.2
thetaT <- 0.60
res <- BCHM(nDat = nDat,
            xDat = xDat,
            alpha = alpha,
            d0 = d0,
            alpha1 = alpha1,
            beta1 = beta1,
            tau2 = tau2,
            phi1 = phi1,
            deltaT = deltaT,
            thetaT = thetaT,
            burnIn = 100,
            MCIter = 200,
            MCNum = 1000,
            seed = 1000
print(res$SMatrix)
print(res$Result)
col <- res$Result[,4]</pre>
BCHMplot_post_dist(res, col, lty=1:length(nDat), lwd =3, xlim=c(0, 0.8))
```

Description

plot the posterior response rate with its highest probability density (HPD) interval by subgroup ID

Usage

```
BCHMplot_post_value(
    res,
    col = res$Result[, 4],
    pch = c(19, 4),
    cex = 2,
    HPD = 0.95,
    xlim = c(0, dim(res$Result)[1] + 2),
    ylim = c(0, 1),
    ObsMean = FALSE,
    ...
)
```

Arguments

res	BCHM calculation results.
col	Color vector
pch	pch vector pch[1] Posterior mean pch[2] Observed mean
cex	size of points
HPD	Highest Posterior Density level for drawing (NA: No HPD drawing)
xlim	X-axis range
ylim	Y-axis range
ObsMean	Draw the observed mean of subgroups if this parameter is TRUE
	other options

Value

None

See Also

```
BCHM Perform the analysis based on the BCHM design.

BCHMplot_cluster Plot the clustering results of subgroups.

BCHMplot_post_dist Plot the posterior distributions of subgroups.
```

Examples

```
nDat = c(25, 25, 25, 25) # total number of patients xDat = c(2, 3, 8, 6) # number of responses alpha <- 1e-20 d0 <- 0.0 alpha1 = 50 beta1 = 10 tau2 <- 0.1 phi1 <- 0.1 deltaT <- 0.2
```

```
thetaT <- 0.60
res <- BCHM(nDat = nDat,
            xDat = xDat,
            alpha = alpha,
            d0 = d0,
            alpha1 = alpha1,
            beta1 = beta1,
            tau2 = tau2,
            phi1 = phi1,
            deltaT = deltaT,
            thetaT = thetaT,
            burnIn = 100,
            MCIter = 200,
            MCNum = 1000,
            seed = 1000
)
print(res$SMatrix)
print(res$Result)
col <- res$Result[,4]</pre>
BCHMplot_post_value(res, col, HPD = 0.8)
```

Index

```
BCHM, 2, 5, 7, 8 
BCHMplot_cluster, 3, 4, 7, 8 
BCHMplot_post_dist, 3, 5, 6, 8 
BCHMplot_post_value, 3, 5, 7, 7
```