

# Package ‘calm’

October 2, 2022

**Type** Package

**Title** Covariate Assisted Large-scale Multiple testing

**Version** 1.10.0

**Description** Statistical methods for multiple testing with covariate information. Traditional multiple testing methods only consider a list of test statistics, such as p-values. Our methods incorporate the auxiliary information, such as the lengths of gene coding regions or the minor allele frequencies of SNPs, to improve power.

**License** GPL (>=2)

**Encoding** UTF-8

**LazyData** false

**Imports** mgcv, stats, graphics

**Suggests** knitr, rmarkdown

**VignetteBuilder** knitr

**biocViews** Bayesian, DifferentialExpression, GeneExpression, Regression, Microarray, Sequencing, RNASeq, MultipleComparison, Genetics, ImmunoOncology, Metabolomics, Proteomics, Transcriptomics

**RoxygenNote** 6.1.1

**BugReports** <https://github.com/k22liang/calm/issues>

**git\_url** <https://git.bioconductor.org/packages/calm>

**git\_branch** RELEASE\_3\_15

**git\_last\_commit** a76133d

**git\_last\_commit\_date** 2022-04-26

**Date/Publication** 2022-10-02

**Author** Kun Liang [aut, cre]

**Maintainer** Kun Liang <kun.liang@uwaterloo.ca>

R topics documented:

|                          |          |
|--------------------------|----------|
| calm . . . . .           | 2        |
| CLfdr . . . . .          | 3        |
| EstFDR . . . . .         | 5        |
| EstNullProp_RB . . . . . | 5        |
| pso . . . . .            | 6        |
| <b>Index</b>             | <b>7</b> |

---

|      |                                                        |
|------|--------------------------------------------------------|
| calm | <i>Covariate Assisted Large-scale Multiple testing</i> |
|------|--------------------------------------------------------|

---

Description

Statistical methods for multiple testing with covariate information.

Details

|           |            |
|-----------|------------|
| Package:  | calm       |
| Type:     | Package    |
| Version:  | 0.9.0      |
| Date:     | 2019-06-22 |
| License:  | GPL (>= 2) |
| LazyLoad: | yes        |

Author(s)

Kun Liang <kun.liang@uwaterloo.ca>  
Maintainer: Kun Liang <kun.liang@uwaterloo.ca>

References

Liang, K (2019) *Empirical Bayes analysis of RNA sequencing experiments with auxiliary information.*

See Also

[CLfdr](#)

---

|       |                                      |
|-------|--------------------------------------|
| CLfdr | <i>Conditional local FDR (CLfdr)</i> |
|-------|--------------------------------------|

---

## Description

CLfdr returns the local false discovery rate (FDR) conditional on auxiliary covariate information

## Usage

```
CLfdr(x, y, pval = NULL, pi0.method = "RB", bw.init = NULL,
      bw = NULL, reltol = 1e-04, n.subsample = NULL, check.gam = FALSE,
      k.gam = NULL, info = TRUE)
```

## Arguments

|                          |                                                                                                                                                                              |
|--------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| <code>x</code>           | covariates, could be a vector of length $m$ or a matrix with $m$ rows.                                                                                                       |
| <code>y</code>           | a vector of $z$ -values of length $m$ .                                                                                                                                      |
| <code>pval</code>        | a vector of $p$ -values of length $m$ . The $p$ -values are only used to computed the overall true null proportion when <code>pi0.method="RB"</code> .                       |
| <code>pi0.method</code>  | method to estimate the overall true null proportion ( $\pi_0$ ). "RB" for the right-boundary procedure (Liang and Nettleton, 2012, JRSSB) or "JC" (Jin and Cai, 2007, JASA). |
| <code>bw.init</code>     | initial values for bandwidth, optional. If not specified, normal-reference rule will be used.                                                                                |
| <code>bw</code>          | bandwidth values.                                                                                                                                                            |
| <code>reltol</code>      | relative tolerance in optim function.                                                                                                                                        |
| <code>n.subsample</code> | size of the subsample when esitimating bandwidth.                                                                                                                            |
| <code>check.gam</code>   | indicator to perform gam.check function on the nonparametric fit.                                                                                                            |
| <code>k.gam</code>       | tuning parameter for mgcv::gam.                                                                                                                                              |
| <code>info</code>        | indicator to print out fitting information.                                                                                                                                  |

## Details

In many multiple testing applications, the auxiliary information is widely available and can be useful. Such information can be summary statistics from a similar experiment or disease, the lengths of gene coding regions, and minor allele frequencies of SNPs.

`y` is a vector of  $m$   $z$ -values, one of each hypothesis under test. The  $z$ -values follow  $N(0,1)$  if their corresponding null hypotheses are true. Other types of test statistics, such as  $t$ -statistics and  $p$ -values can be transformed to  $z$ -values. In practice, if the distribution of  $z$ -values is far from  $N(0,1)$ , recentering and rescaling of the  $z$ -values may be necessary.

`x` contains auxiliary covariate information. For a single covariate, `x` should be a vector of length  $m$ . For multiple covariates, `x` should be a matrix with  $m$  rows. The covariates can be either continuous or ordered.

`pi0.method` specifies the method used to estimate the overall true null proportion. If the  $z$ -values are generated from the normal means model, the "JC" method from Jin and Cai (2007) JASA can be a good candidate. Otherwise, the right-boundary procedure ("RB", Liang and Nettleton, 2012, JRSSB) is used.

`bw` are bandwidth values for estimating local alternative density. Suppose there are  $p$  covariates, then `bw` should be a vector of  $p+1$  positive numerical values. By default, these bandwidth values are chosen by cross-validation to minimize a certain error measure. However, finding the optimal bandwidth values by cross-validation can be computationally intensive, especially when  $p$  is not small. If good estimates of bandwidth values are available, for example, from the analysis of a similar dataset, the bandwidth values can be specified explicitly to save time.

`reltol` specifies the relative convergence tolerance when choosing the bandwidth values (`bw`). It will be passed on to `stats::optim()`. For most analyses, the default value of  $1e-4$  provides reasonably good results. A smaller value such as  $1e-5$  or  $1e-6$  could be used for further improvement at the cost of more computation time.

### Value

|                      |                                                                                                                                                                                                                                                                |
|----------------------|----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| <code>fdr</code>     | a vector of local FDR estimates. <code>fdr[i]</code> is the posterior probability of the $i$ th null hypothesis is true given all the data. <code>1-fdr[i]</code> is the posterior probability of being a signal (the corresponding null hypothesis is false). |
| <code>FDR</code>     | a vector of FDR values (q-values), which can be used to control FDR at a certain level by thresholding the FDR values.                                                                                                                                         |
| <code>pi0</code>     | a vector of true null probability estimates. This contains the prior probabilities of being null.                                                                                                                                                              |
| <code>bw</code>      | a vector of bandwidths for conditional alternative density estimation                                                                                                                                                                                          |
| <code>fit.gam</code> | an object of <code>mgcv::gam</code>                                                                                                                                                                                                                            |

### Author(s)

Kun Liang, <kun.liang@uwaterloo.ca>

### References

Liang (2019), Empirical Bayes analysis of RNA sequencing experiments with auxiliary information, to appear in Annals of Applied Statistics

### Examples

```
data(pso)
ind.nm <- is.na(pso$tval_mic)
x <- pso$len_gene[ind.nm]
# normalize covariate
x <- rank(x)/length(x)
y <- pso$zval[ind.nm]
# assign names to the z-values helps to give names to the output variables
names(y) <- row.names(pso)[ind.nm]

fit.nm <- CLfdr(x=x, y=y)
fit.nm$fdr[1:5]
```

---

EstFDR*FDR estimation*

---

**Description**

False discovery rate (FDR) estimation from local FDR

**Usage**

```
EstFDR(fdr)
```

**Arguments**

fdr                      vector of local FDR

**Value**

the estimate of the FDR

**Examples**

```
lfdr <- c(runif(900), rbeta(100, 1, 10))
FDR <- EstFDR(lfdr)
sum(FDR < 0.05)
```

---

EstNullProp\_RB*Right-boundary procedure*

---

**Description**

True null proportion ( $\pi_0$ ) estimator of Liang and Nettleton (2012), JRSSB

**Usage**

```
EstNullProp_RB(pval, lambda.vec = 0.05 * seq_len(19))
```

**Arguments**

pval                      vector of p-values  
lambda.vec                vector of lambda candidates (excluding 0 and 1)

**Value**

the estimate of the overall true null proportion

**Examples**

```
pval <- c(runif(900), rbeta(100, 1, 10))
EstNullProp_RB(pval)
```

---

|     |                                  |
|-----|----------------------------------|
| pso | <i>Psoriasis RNA-seq dataset</i> |
|-----|----------------------------------|

---

**Description**

A dataset containing the test statistics to analyze an RNA-seq study of psoriasis.

**Usage**

pso

**Format**

A dataset with the following vectors:

**zval** 16490 z-values of genes with matching microarray data

**len\_gene** 16490 gene coding region length for zval

**tval\_mic** 16490 matching microarray t-statistics

**Source**

Liang (2019), Empirical Bayes analysis of RNA sequencing experiments with auxiliary information, to appear in *Annals of Applied Statistics*;

**Examples**

```
data(pso)
dim(pso)
# total number of genes without matching microarray data
sum(is.na(pso$tval_mic))
```

# Index

- \* **datasets**

- pso, [6](#)

- \* **package**

- calm, [2](#)

calm, [2](#)

CLfdr, [2](#), [3](#)

EstFDR, [5](#)

EstNullProp\_RB, [5](#)

pso, [6](#)

stats::optim(), [4](#)