

# Package ‘EBCHS’

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**Type** Package

**Title** An Empirical Bayes Method for Chi-Squared Data

**Version** 0.1.0

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**Description** We provide the main R functions to compute the posterior interval for the noncentrality parameter of the chi-squared distribution. The skewness estimate of the posterior distribution is also available to improve the coverage rate of posterior intervals. Details can be found in Du and Hu (2020) <[doi:10.1080/01621459.2020.1777137](https://doi.org/10.1080/01621459.2020.1777137)>.

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**Encoding** UTF-8

**URL** <https://github.com/dulilun/EBCHS>

**RoxygenNote** 7.1.1

**Imports** stats, pracma, splines, fda

**Suggests** testthat

**NeedsCompilation** no

**Repository** CRAN

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## Contents

density_LS . . . . .	2
density_PLS . . . . .	2
EB_CS . . . . .	3

## Index

5

**density\_LS***log-density derivatives–parametric approach***Description**

Assuming the log density of the chi-squared statistics admits a parametric form, this function estimates up to the fourth order log density derivatives.

**Usage**

```
density_LS(x)
```

**Arguments**

<code>x</code>	a sequence of chi-squared test statistics
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**Value**

a list: the first-to-fourth log density derivatives

**Examples**

```
p = 1000
k = 7
# the prior distribution for lambda
alpha = 2
beta = 10
# lambda
lambda = rep(0, p)
pi_0 = 0.8
p_0 = floor(p*pi_0)
p_1 = p-p_0
lambda[(p_0+1):p] = stats::rgamma(p_1, shape = alpha, rate=1/beta)
# Generate a Poisson RV
J = sapply(1:p, function(x){rpois(1, lambda[x]/2)})
X = sapply(1:p, function(x){rchisq(1, k+2*J[x])})
out = density_LS(X)
```

**density\_PL***Penalized least-squares method***Description**

The semiparametric model is employed to estimate the log density derivatives of the chi-squared statistics.

**Usage**

```
density_PLS(x, qq)
```

**Arguments**

x	a sequence of chi-squared test statistics
qq	the quantiles used for splines

**Value**

a list: the first and second density derivatives

**Examples**

```
p = 1000
k = 7
# the prior distribution for lambda
alpha = 2
beta = 10
# lambda
lambda = rep(0, p)
pi_0 = 0.5
p_0 = floor(p*pi_0)
p_1 = p-p_0
lambda[(p_0+1):p] = stats::rgamma(p_1, shape = alpha, rate=1/beta)
# Generate a Poisson RV
J = sapply(1:p, function(x){rpois(1, lambda[x]/2)})
X = sapply(1:p, function(x){rchisq(1, k+2*J[x])})
qq = c(0.2, 0.4, 0.6, 0.8)
out = density_PLS(X, qq)
```

EB\_CS

*Main function used in the paper (Du and Hu, 2020)*

**Description**

Give a sequence of chi-squared statistic values, the function computes the posterior mean, variance, and skewness of the noncentrality parameter given the data.

**Usage**

```
EB_CS(
  x,
  df,
  qq = c(0.2, 0.4, 0.6, 0.8),
  method = c("LS", "PLS", "g_model"),
  mixture = FALSE
)
```

## Arguments

x	a sequence of chi-squared test statistics
df	the degrees of freedom
qq	the quantiles used in spline basis
method	LS: parametric least-squares; PLS: penalized least-squares; g-model: g-modeling
mixture	default is FALSE: there is no point mass at zero.

## Value

a list: posterior mean, variance, and skewness estimates

## References

Du and Hu (2020), *An Empirical Bayes Method for Chi-Squared Data*, *Journal of American Statistical Association*, forthcoming.

## Examples

```
p = 1000
k = 7
# the prior distribution for lambda
alpha = 2
beta = 10
# lambda
lambda = rep(0, p)
pi_0 = 0.8
p_0 = floor(p*pi_0)
p_1 = p-p_0
lambda[(p_0+1):p] = rgamma(p_1, shape = alpha, rate=1/beta)
# Generate a Poisson RV
J = sapply(1:p, function(x){rpois(1, lambda[x]/2)})
X = sapply(1:p, function(x){rchisq(1, k+2*J[x])})
qq_set = seq(0.01, 0.99, 0.01)
out = EB_CS(X, k, qq=qq_set, method='LS', mixture = TRUE)
E = out$E_lambda
V = out$V_lambda
S = out$S_lambda
```

# Index

[density\\_LS](#), [2](#)

[density\\_PLS](#), [2](#)

[EB\\_CS](#), [3](#)