

Package ‘ebci’

May 27, 2020

Title Robust empirical Bayes confidence intervals

Version 0.0.0.9002

Description Computes empirical Bayes confidence estimators and confidence intervals (EBCIs) in a normal means model. The EBCIs are robust in the sense that they achieve correct coverage regardless of the distribution of the means. If the means are treated as fixed, the EBCIs have an average coverage guarantee.

Depends R (>= 3.6.0)

License MIT + file LICENSE

Encoding UTF-8

LazyData true

Suggests spelling,
testthat (>= 2.1.0),
lpSolve,
parallel,
knitr,
rmarkdown

Language en-US

URL <https://github.com/kolesarm/ebci>

BugReports <https://github.com/kolesarm/ebci/issues>

Roxygen list(markdown = TRUE)

RoxygenNote 7.1.0

VignetteBuilder knitr

R topics documented:

ct	2
cva	3
cva_tbl	4
cz	4
ebci	5

w_eb	7
w_opt	8

Index	10
--------------	-----------

ct	<i>Neighborhood effects data from Chetty and Hendren (2018)</i>
----	---

Description

This dataset contains a subset of the publicly available data from Chetty and Hendren (2018). It contains raw estimates and standard errors of neighborhood effects at the county level

Usage

ct

Format

A data frame with 741 rows corresponding to counties and 12 columns corresponding to the variables:

ct County ID

cz ID of commuting zone to which the county belongs.

ctname County name

state 2-digit state code

pop Population of the county according to the year 2000 Census

pop_cz Population of the CZ to which the county belongs according to the year 2000 Census

theta25 Fixed-effect estimate of the causal effect of living in the county for one year on children's percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for children growing up with parents at the 25th percentile of national income distribution

theta75 Fixed-effect estimate of the causal effect of living in the county for one year on children's percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for children growing up with parents at the 75th percentile of national income distribution

se25 Standard error of theta25

se75 Standard error of theta75

stayer25 Average percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for stayers (children who grew up in the county and did not move) with parents at the 25th percentile of national income distribution.

stayer75 Average percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for stayers (children who grew up in the county and did not move) with parents at the 75th percentile of national income distribution.

Source

https://opportunityinsights.org/data/?paper_id=599

References

Chetty, R., & Hendren, N. (2018). *The Impacts of Neighborhoods on Intergenerational Mobility II: County-Level Estimates*. *The Quarterly Journal of Economics*, 133(3), 1163–1228. <https://doi.org/10.1093/qje/qjy006>

cva	<i>Compute average coverage critical value under moment constraints.</i>
-----	--

Description

Computes the critical value $cva_{\alpha}(m_2, \kappa)$ from Armstrong, Kolesár, and Plagborg-Møller (2020).

Usage

```
cva(m2, kappa = Inf, alpha = 0.05, check = TRUE)
```

Arguments

m2	Bound on second moment of the normalized bias, m_2
kappa	Bound on the kurtosis of the normalized bias, κ
alpha	Determines confidence level, $1 - \alpha$.
check	If TRUE, verify accuracy of the solution by checking that the implied least favorable distribution satisfies the m2 and kappa constraints and yields the same non-coverage rate. If this fails (perhaps due to numerical accuracy issues), solve a finite-grid approximation (by discretizing the support of the normalized bias) to the primal linear programming problem, and check that it agrees with the dual solution.

Value

Returns a list with 4 components:

cv Critical value for constructing two-sided confidence intervals.

alpha The argument alpha.

x Support points for the least favorable distribution for the squared normalized bias, b^2 .

p Probabilities associated with the support points.

References

Armstrong, Timothy B., Kolesár, Michal, and Plagborg-Møller, Mikkel (2020): *Robust Empirical Bayes Confidence Intervals*, <https://arxiv.org/abs/2004.03448>

Examples

```
## Critical value without imposing a constraint on kurtosis
cva(1, kappa=Inf)
## With a constraint
cva(1, kappa=3)
```

cva_tbl	<i>Table of pre-computed critical values</i>
---------	--

Description

Table of pre-computed critical values

Usage

```
cva_tbl
```

Format

A data frame with 4 variables:

m2 Bound on the second moment of the normalized bias

kappa Bound on the kurtosis of the normalized bias

cv Critical value

alpha Determines confidence level, $1 - \alpha$.

Source

Computed using the cva function in this package

cZ	<i>Neighborhood effects data from Chetty and Hendren (2018)</i>
----	---

Description

This dataset contains a subset of the publicly available data from Chetty and Hendren (2018). It contains raw estimates and standard errors of neighborhood effects at the commuting zone level

Usage

```
cZ
```

Format

A data frame with 741 rows corresponding to commuting zones (CZ) and 10 columns corresponding to the variables:

cz Commuting zone ID

czname Name of CZ

state 2-digit state code

pop Population according to the year 2000 Census

theta25 Fixed-effect estimate of the causal effect of living in the CZ for one year on children's percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for children growing up with parents at the 25th percentile of national income distribution

theta75 Fixed-effect estimate of the causal effect of living in the CZ for one year on children's percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for children growing up with parents at the 75th percentile of national income distribution

se25 Standard error of theta25

se75 Standard error of theta75

stayer25 Average percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for stayers (children who grew up in the CZ and did not move) with parents at the 25th percentile of national income distribution.

stayer75 Average percentile rank in the national distribution of household earnings at age 26 relative to others in the same birth cohort for stayers (children who grew up in the CZ and did not move) with parents at the 75th percentile of national income distribution.

Source

https://opportunityinsights.org/data/?paper_id=599

References

Chetty, R., & Hendren, N. (2018). *The Impacts of Neighborhoods on Intergenerational Mobility II: County-Level Estimates*. *The Quarterly Journal of Economics*, 133(3), 1163–1228. <https://doi.org/10.1093/qje/qjy006>

ebci

Compute empirical Bayes confidence intervals by shrinking toward regression

Description

Computes empirical Bayes estimators based on shrinking towards a regression, and associated robust empirical Bayes confidence intervals (EBCIs), as well as length-optimal robust EBCIs.

Usage

```
ebci(
  formula,
  data,
  se,
  weights = NULL,
  alpha = 0.1,
  kappa = NULL,
  wopt = FALSE,
  fs_correction = "PMT",
  tstat = FALSE,
  cores = max(parallel::detectCores() - 1L, 1L)
)
```

Arguments

formula	object of class "formula" (or one that can be coerced to that class) of the form $Y \sim \text{predictors}$, where Y is a preliminary unbiased estimator, and predictors are predictors X that guide the direction of shrinkage. For shrinking toward the grand mean, use $Y \sim 1$, and for shrinking toward \emptyset use $Y \sim \emptyset$
data	optional data frame, list or environment (or object coercible by <code>as.data.frame</code> to a data frame) containing the preliminary estimator Y and the predictors. If not found in data, these variables are taken from <code>environment(formula)</code> , typically the environment from which the function is called.
se	Standard errors σ associated with the preliminary estimates Y
weights	An optional vector of weights to be used in the fitting process in computing δ , μ_2 and κ . Should be <code>NULL</code> or a numeric vector.
alpha	Determines confidence level, $1 - \alpha$.
kappa	If non- <code>NULL</code> , use pre-specified value for the kurtosis κ of $\theta - X'\delta$ (such as <code>Inf</code>), instead of computing it.
wopt	If <code>TRUE</code> , also compute length-optimal robust EBCIs. These are robust EBCIs centered at length-optimal shrinkage factors <code>w_opt</code> .
fs_correction	Finite-sample correction method used to compute μ_2 and κ . These corrections ensure that we do not shrink the preliminary estimates Y all the way to zero. If "PMT", use posterior mean truncation, if "FPLIB" use limited information Bayesian approach with a flat prior, and if "none", truncate the estimates at \emptyset for μ_2 and 1 for κ .
tstat	If <code>TRUE</code> , shrink the t-statistics Y/se rather than the preliminary estimates Y .
cores	Number of cores to use when computing length-optimal robust EBCIs. By default, the computation of the length-optimal shrinkage factors <code>w_opt</code> is parallelized if there are more than 30 observations to speed up the calculations.

Value

Returns a list with the following components:

`sqrt_mu2` Square root of the estimated second moment of $\theta - X'\delta$, $\sqrt{\mu_2}$. Vector of length 2, the first element corresponds to the estimate after the finite-sample correction as specified by `fs_correction`, the second element is the uncorrected estimate.

`kappa` Estimated kurtosis κ of $\theta - X'\delta$. Vector of length 2 with the same structure as `sqrt_mu2`.

`delta` Estimated regression coefficients δ

`df` Data frame with components described below.

`df` has the following components:

`w_eb` EB shrinkage factors, $\mu_2/(\mu_2 + \sigma_i^2)$

`w_opt` Optimal shrinkage factors `w_opt`

`ncov_pa` Maximal non-coverage of parametric EBCIs

`len_eb` Half-length of robust EBCIs based on EB shrinkage, so that the intervals take the form `cbind(th_eb-len_eb, th_eb+len_eb)`

`len_op` Half-length of robust EBCIs based on length-optimal shrinkage, so that the intervals take the form `cbind(th_op-len_op, th_op+len_op)`

`len_pa` Half-length of parametric EBCIs, which take the form `cbind(th_eb-len_pa, th_eb+len_a)`

`len_us` Half-length of unshrunk CIs, which take the form `cbind(th_us-len_us, th_us+len_us)`

`th_us` Unshrunk estimate Y

`th_eb` EB estimate.

`th_eb` Estimate based on length-optimal shrinkage.

`se` Standard error σ , as supplied by the argument `se`.

References

Armstrong, Timothy B., Kolesár, Michal, and Plagborg-Møller, Mikkel (2020): *Robust Empirical Bayes Confidence Intervals*, <https://arxiv.org/abs/2004.03448>

Examples

```
ebci(theta25 ~ stayer25, cz, se25, 1/se25^2, tstat=FALSE, cores=1)
```

w_eb

Empirical Bayes estimator and confidence intervals

Description

Compute empirical Bayes shrinkage factor w_{eb} and the normalized half-length of of the associated robust Empirical Bayes confidence interval (EBCI).

Usage

```
w_eb(S, kappa = Inf, alpha = 0.05)
```

Arguments

S	Signal-to-noise ratio μ_2/σ^2 , where μ_2 is the second moment of $\theta - X'\delta$, and σ^2 is the variance of the preliminary estimator.
kappa	Kurtosis of $\theta - X'\delta$.
alpha	Determines confidence level, $1 - \alpha$.

Value

Returns a list with 3 components:

w Empirical Bayes shrinkage factor w_{eb}

length Normalized half-length of the corresponding confidence interval, so that the interval obtains by taking the estimator based on shrinkage given by w, and adding and subtracting length times the standard error σ of the preliminary estimator.

m2 Second moment of the normalized bias, $1/m2$.

References

Armstrong, Timothy B., Kolesár, Michal, and Plagborg-Møller, Mikkel (2020): Robust Empirical Bayes Confidence Intervals, <https://arxiv.org/abs/2004.03448>

Examples

```
w_opt(1, 3)
## No constraint on kurtosis yields doesn't affect shrinkage, but yields
## larger half-length
w_opt(1, Inf)
```

w_opt

Optimal shrinkage for Empirical Bayes confidence intervals

Description

Compute linear shrinkage factor w_{opt} to minimize the length of the resulting Empirical Bayes confidence interval (EBCI).

Usage

```
w_opt(S, kappa, alpha = 0.05, cv_tbl = NULL)
```


Arguments

S	Signal-to-noise ratio μ_2/σ^2 , where μ_2 is the second moment of $\theta - X'\delta$, and σ^2 is the variance of the preliminary estimator.
kappa	Kurtosis of $\theta - X'\delta$.
alpha	Determines confidence level, $1 - \alpha$.
cv_tbl	Optionally, supply a data frame of critical values. <code>cva(m2,kappa,alpha)</code> , for different values of m2, such as a subset of the data frame <code>cva_tbl</code> , that matches the supplied values of alpha and kappa. The data frame needs to contain two variables, m2, corresponding to the value of second moment of the normalized bias, and cv, with the corresponding value of <code>cva(m2,kappa,alpha)</code> . If non NULL, for the purposes of optimizing the shrinkage factor, compute the critical value cva by interpolating between the critical values in this data frame, instead of computing them from scratch. This can speed up the calculations.

Value

Returns a list with 3 components:

w	Optimal shrinkage factor w_{opt}
length	Normalized half-length of the corresponding confidence interval, so that the interval obtains by taking the estimator based on shrinkage given by w, and adding and subtracting length times the standard error σ of the preliminary estimator.
m2	The second moment of the normalized bias, $(1/w - 1)^2 S$

References

Armstrong, Timothy B., Kolesár, Michal, and Plagborg-Møller, Mikkel (2020): *Robust Empirical Bayes Confidence Intervals*, <https://arxiv.org/abs/2004.03448>

Examples

```
w_opt(1, 3)
## Use precomputed critical value table
w_opt(1, 3, cv_tbl=cva_tbl[cva_tbl$kappa==3 & cva_tbl$alpha==0.05, ])
```

Index

*Topic **datasets**

ct, [2](#)

cva_tbl, [4](#)

cz, [4](#)

ct, [2](#)

cva, [3](#)

cva_tbl, [4](#), [9](#)

cz, [4](#)

ebci, [5](#)

w_eb, [7](#)

w_opt, [6](#), [7](#), [8](#)