Package 'publipha'

April 4, 2023

Title Bayesian Meta-Analysis with Publications Bias and P-Hacking

Version 0.1.2

Description Tools for Bayesian estimation of meta-analysis models that account
for publications bias or p-hacking. For publication bias, this package
implements a variant of the p-value based selection model of Hedges (1992)
<doi:10.1214/ss/1177011364> with discrete selection probabilities. It also
implements the mixture of truncated normals model for p-hacking described
in Moss and De Bin (2019) <arXiv:1911.12445>.

License GPL-3

Depends methods, R (>= 3.6.0), Rcpp (>= 0.12.19)

Imports rstan (>= 2.18.1), rstantools (>= 1.5.1), loo, truncnorm

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0.3.3.4.0), rstan (>= 2.21.8), StanHeaders (>= 2.21.0-7),
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'data-cuddy2018.R' 'data-dang2018.R' 'data-motyl2017.R'
'densities-helpers.R' 'densities-mpsnorm.R'
'densities-phnorm.R' 'densities-psnorm.R' 'densities-snorm.R'
'ma.R' 'generics.R' 'publipha-package.R' 'stanmodels.R'
'tools.R' 'utility.R'

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R topics documented:

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publipha-package   *The 'publipha' package.*

Description

Meta-analysis that corrects for publication selection bias and p-hacking.

References

Hedges, Larry V. "Modeling publication selection effects in meta-analysis." Statistical Science (1992): 246-255.

Moss, Jonas and De Bin, Riccardo. "Modelling publication bias and p-hacking" (2019) arXiv:1911.12445

Stan Development Team (2018). RStan: the R interface to Stan. R package version 2.18.1. https://mc-stan.org

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dat.anderson2010 *Studies on Effect of Violent Video Games on Negative Outcomes*

Description

Results from 477 studies on the effect of violent video games on negative outcomes.

Usage

```
dat.anderson2010
```

Format

The data frame contains the following columns:

author	character	first author
year	numeric	publication year
outcome	character	one of seven outcomes
best	boolean	if TRUE, the was a best practice study
experimental	boolean	if TRUE, the study was experimental
adult	boolean	if TRUE, the study subjects were adults
country	character	country of study
ni	numeric	sample size
yi	numeric	observed mean difference in outcome (violent vs. non-violent)
vi	numeric	corresponding sampling variance

Source

https://github.com/Joe-Hilgard/Anderson-meta

References

Baskerville, N. B., Liddy, C., & Hogg, W. (2012). Systematic review and meta-analysis of practice facili
Hilgard, J., Engelhardt, C. R., & Rouder, J. N. (2017). Overstated evidence for short-term effects of vio

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dat.baskerville2012 *Studies on Practice Facilitation*

Description

Results from 23 studies on the effect of practice facilitation in a primary care setting.

Usage

```
dat.baskerville2012
```

Format

The tibble contains the following columns:

author	character	first author of study
year	numeric	publication year
design	character	study design (RCT, C-RCT, or CCT)
blinded	boolean	if TRUE, the study was blinded
concealed	boolean	if TRUE, the study was concealed
yi	numeric	observed mean difference in outcome (facilitated vs non-facilitated)
vi	numeric	corresponding sampling variance

Source

Baskerville, N. B., Liddy, C., & Hogg, W. (2012). Systematic review and meta-analysis of practice facilitation within primary care settings. The Annals of Family Medicine, 10(1), 63-74.

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dat.cuddy2018   *Studies on the Effect of Power Posing*

Description

Results from 27 studies related to power posing.

Usage

dat.cuddy2018

Format

The data frame contains the following columns:

author	character	first author
year	numeric	publication year
power	boolean	if TRUE, the outcome was feeling of power
ease	boolean	if TRUE, the outcome was an EASE variable
yi	numeric	standardized mean difference
vi	numeric	corresponding sampling variance

Details

The data points are taken from the p-curve analysis of Cuddy et al. (2018), restricted to 2 cell designs with mean difference as the outcome variable.

Source

https://osf.io/jx3av/

References

Cuddy, A. J., Schultz, S. J., & Fosse, N. E. (2018). P-curving a more comprehensive body of research on po

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dat.dang2018	*Meta-analysis on Ego Depletion*

Description

Results from 150 studies of ego depletion, the claim that self-control is a limited resource which is tapped whenever self-control is exerted.

Usage

```
data(dat.dang2018)
```

Format

The tibble contains the following columns:

author	character	the last name of the first author and the first letter of the last name of the second author;
year	numeric	publication year
in_carter	character	was the study in the meta-analysis of Carter et al. (2015)
study	character	the number given to the study in the original paper (0 = only one study was reported in the original
dv	boolean	the dependent variable
iv	boolean	the independent variable
n1i	numeric	the number of participants in the depletion condition
n2i	numeric	the number of participants in the control condition
yi	numeric	the adjusted standardized mean difference
vi	numeric	the variance

Source

https://link.springer.com/article/10.1007/s00426-017-0862-x#SupplementaryMaterial)

References

Dang, J. (2018). An updated meta-analysis of the ego depletion effect. Psychological Research, 82(4), 645-651.

Carter, E. C., Kofler, L. M., Forster, D. E., & McCullough, M. E. (2015). A series of meta-analytic tests of the depletion effect: self-control does not seem to rely on a limited resource. Journal of Experimental Psychology: General, 144(4), 796.

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dat.motyl2017 *Effect Sizes from 875 Studies in Psychology.*

Description

Effect sizes from 875 studies in psychology. Adopted from Motyl et al. (2017).

Usage

```
data(dat.motyl2017)
```

Format

The tibble contains the following columns:

author	character	first author of study
year	numeric	publication year
study	numeric	the number given to the study in the original paper (0 = only one study was reported in the orig
journal	character	journal where the study was published
concealed	character	design of the study; "Between", "Within", or "Mixed"
experimental	numeric	TRUE for an experimental study
ni	numeric	sample size
yi	numeric	observed mean difference in outcome
vi	numeric	corresponding sampling variance

Source

<https://osf.io/he8mu/>

References

Motyl, M., Demos, A. P., Carsel, T. S., Hanson, B. E., Melton, Z. J., Mueller, A. B., ... & Yantis, C. (2017). The state of social and personality science: Rotten to the core, not so bad, getting better, or getting worse?. Journal of personality and social psychology, 113(1), 34.

ExtractParameters *Extract Parameters from an* mafit *Object*

Description

Extract samples from a model of class [mafit](#) and apply a function fun to them.

Usage

```
extract_theta0(object, fun = mean)

extract_theta(object, fun = mean, i)

extract_tau(object, fun = mean)

extract_eta(object, fun = mean, i)

extract_isq(object, fun = mean)
```

Arguments

object	an object of class `mafit`.
fun	the function to be applied to the fitted parameters.
i	an optional index specifying which parameter to apply `fun` to. Only for `extract_eta` and `extract_theta`.

Details

Support parameters for extraction are: The meta-analytic mean theta0, the individual means theta, the heterogeneity parameter tau, the selection bias parameter eta, and the I squared isq. See Higgins and Thompson (2002) for details about I squared.

All `extract_` functions are wrappers around `rstan::extract`.

Value

The result of `FUN` being applied to all estimated parameters of `object`.

References

Higgins, J. P., & Thompson, S. G. (2002). Quantifying heterogeneity in a meta-analysis. Statistics in medicine, 21(11), 1539-1558.

Examples

```
set.seed(313)
model <- publipha::psma(yi = yi, vi = vi, data = dat.baskerville2012)
extract_theta0(model, mean) # [1] extract_theta0(model, mean)
extract_theta0(model, sd) # [1] 0.1095921
extract_tau(model, mean) # [1] 0.1315312
extract_theta(model, hist, i = 5)
```

---

loo,mafit-method   *Calculate the* loo *for an* ma *object.*

Description

Computes PSIS-LOO CV, approximate leave-one-out cross-validation using Pareto smoothed importance sampling, see [loo](#).

Usage

```
## S4 method for signature 'mafit'
loo(x, ...)
```

Arguments

x   an object of class mafit.

...   passed to [loo](#). Only

Details

... affect the function through two parameters, marginal and lower_bound. When marginalis TRUE, the PSIS-LOO CV is based on the marginal likelihood, i.e. with the dependence on theta integrated out. marginal defaults to TRUE. lower_bound species the lower bound where log-likelihoods are dropped; this is only used in the *p*-hacking model and defaults to -6.

Value

A [loo](#) object.

Examples

```
phma_model <- phma(yi, vi, data = metadat::dat.begg1989)
psma_model <- psma(yi, vi, data = metadat::dat.begg1989)
loo(phma_model)
loo(psma_model)
```

---

ma   *Meta-analysis Correcting for Publication Bias or p-hacking*

Description

Bayesian random effects meta-analysis. Correct for publication bias, correct for p-hacking, or run an ordinary meta-analysis without any correction.

Usage

```
ma(
  yi,
  vi,
  bias = c("publication selection", "p-hacking", "none"),
  data,
  alpha = c(0, 0.025, 0.05, 1),
  prior = NULL,
  tau_prior = c("half-normal", "uniform", "inv_gamma"),
  ...
)

psma(
  yi,
  vi,
  data,
  alpha = c(0, 0.025, 0.05, 1),
  prior = NULL,
  tau_prior = c("half-normal", "uniform", "inv_gamma"),
  ...
)

phma(
  yi,
  vi,
  data,
  alpha = c(0, 0.025, 0.05, 1),
  prior = NULL,
  tau_prior = c("half-normal", "uniform", "inv_gamma"),
  ...
)

cma(
  yi,
  vi,
  data,
  prior = NULL,
  tau_prior = c("half-normal", "uniform", "inv_gamma"),
```

```
  ...
)

allma(
  yi,
  vi,
  data,
  alpha = c(0, 0.025, 0.05, 1),
  prior = NULL,
  tau_prior = c("half-normal", "uniform", "inv_gamma"),
  ...
)
```

Arguments

yi	Numeric vector of length codek with observed effect size estimates.
vi	Numeric vector of length codek with sampling variances.
bias	String; If "publication bias", corrects for publication bias. If "p-hacking", corrects for p-hacking.
data	Optional list or data frame containing yi and vi.
alpha	Numeric vector; Specifies the cutoffs for significance. Should include 0 and 1. Defaults to (0, 0.025, 0.05, 1).
prior	Optional list of prior parameters. See the details.
tau_prior	Which prior to use for tau, the heterogeneity parameter. Defaults to "half-normal"; "uniform" and "inv_gamma are also supported.
...	Passed to rstan::sampling.

Details

ma does a Bayesian meta-analysis with the type of correction used specified by bias. psma is a wrapper for ma with bias = "publication selection", phma is a wrapper with bias = "p-hacking", while cma has bias = "none". The function allma runs all bias options and returns a list.

The bias options are:

1. publication selection: The model of publication bias described in Hedges (1992).

2. p-hacking: The model for *p*-hacking described in Moss & De Bin (2019).

3. none: Classical random effects meta-analysis with no correction for selection bias.

The effect size distribution is normal with mean theta0 and standard deviation tau. The prior for theta0 is normal with parameters theta0_mean (default: 0), theta0_sd (default: 1). eta is the vector of K normalized publication probabilities (publication bias model) or K *p*-hacking probabilities (*p*-hacking model). The prior of eta is Dirchlet with parameter eta0, which defaults to rep(1, K) for the publication bias model and the p-hacking model. eta0 is the prior for the Dirichlet distribution over the non-normalized etas in the publication bias model, and they are forced to be decreasing.

The standard prior for tau is half-normal with parameters tau_mean (default: 0), tau_sd (default: 1). If the uniform prior is used, the parameter are u_min (default: 0), and u_max with a default of 3. The inverse Gamma has parameters shape (default: 1) and scale default: 1.

To change the prior parameters, pass them to prior in a list.

Value

An S4 object of class mafit when ma, psma, phma or cma is run. A list of mafit objects when allma is run.

References

Hedges, Larry V. "Modeling publication selection effects in meta-analysis." Statistical Science (1992): 246-255.

Moss, Jonas and De Bin, Riccardo. "Modelling publication bias and p-hacking" (2019) arXiv:1911.12445

Examples

```
phma_model <- phma(yi, vi, data = metadat::dat.begg1989)

prior <- list(
  eta0 = c(3, 2, 1),
  theta0_mean = 0.5,
  theta0_sd = 10,
  tau_mean = 1,
  tau_sd = 1
)

psma_model <- psma(yi, vi, data = metadat::dat.begg1989, prior = prior)


cma_model <- psma(yi, vi, data = metadat::dat.begg1989, prior = prior)


model <- allma(yi, vi, data = metadat::dat.begg1989, prior = prior)
```

---

mafit-class   *Class* mafit*: Fitted Meta-analysis Model*

Description

Class mafit: Fitted Meta-analysis Model

---

mpsnorm   *Marginal Publication Selection Meta-analysis Model*

Description

Density, distribution, and random variate generation for the marginalized distribution of the publication selection meta-analysis model

Usage

```
dmpsnorm(x, theta0, tau, sigma, alpha = c(0, 0.025, 0.05, 1), eta, log = FALSE)

pmpsnorm(
  q,
  theta0,
  tau,
  sigma,
  alpha = c(0, 0.025, 0.05, 1),
  eta,
  lower.tail = TRUE,
  log.p = FALSE
)

rmpsnorm(n, theta0, tau, sigma, alpha = c(0, 0.025, 0.05, 1), eta)
```

Arguments

x, q	vector of quantiles.
theta0	vector of means.
tau	vector of heterogeneity parameters.
sigma	vector of study standard deviations.
alpha	vector of thresholds for publication bias.
eta	vector of publication probabilities, normalized to sum to 1.
log, log.p	logical; If TRUE, probabilities are given as log(p).
lower.tail	logical; If TRUE (default), the probabilities are $P[X \leq x]$ otherwise, $P[X \geq x]$.
n	number of observations. If length(n) > 1, the length is taken to be the number required.

Details

These functions assume a normal underlying effect size distribution and one-sided selection on the effects. For the fixed effects publication bias model see psnorm.

Value

dmpsnorm gives the density, pmpsnorm gives the distribution function, and rmpsnorm generates random deviates.

References

Hedges, Larry V. "Modeling publication selection effects in meta-analysis." Statistical Science (1992): 246-255.

Moss, Jonas and De Bin, Riccardo. "Modelling publication bias and p-hacking" Forthcoming (2019)

Examples

```
rmpsnorm(100, theta0 = 0, tau = 0.1, sigma = 0.1, eta = c(1, 0.5, 0.1))
```

---

phnorm	*p-hacking Meta-analysis Model*

Description

Density, distribution, and random variate generation for the p-hacking meta- analysis model.

Usage

```
dphnorm(x, theta, sigma, alpha = c(0, 0.025, 0.05, 1), eta, log = FALSE)

rphnorm(n, theta, sigma, alpha = c(0, 0.025, 0.05, 1), eta)

pphnorm(
  q,
  theta,
  sigma,
  alpha = c(0, 0.025, 0.05, 1),
  eta,
  lower.tail = TRUE,
  log.p = FALSE
)
```

Arguments

x, q	vector of quantiles.
theta	vector of means.
sigma	vector of study standard deviations.
alpha	vector of thresholds for p-hacking.
eta	vector of p-hacking probabilities, normalized to sum to 1.

log, log.p	logical; If TRUE, probabilities are given as log(p).
n	number of observations. If length(n) > 1, the length is taken to be the number required.
lower.tail	logical; If TRUE (default), the probabilities are $P[X \leq x]$ otherwise, $P[X \geq x]$.

Details

These functions assume one-sided selection on the effects. alpha contains the selection thresholds and eta the vector of *p*-hacking probabilities. theta is the true effect, while sigma is the true standard deviation before selection.

Value

dphnorm gives the density, pphnorm gives the distribution function, and rphnorm generates random deviates.

References

Moss, Jonas and De Bin, Riccardo. "Modelling publication bias and p-hacking" Forthcoming (2019)

Examples

```
rphnorm(100, theta = 0, sigma = 0.1, eta = c(1, 0.5, 0.1))
```

---

psnorm   *Publication Selection Meta-analysis Model*

Description

Density, distribution, quantile, random variate generation, and expectation calculation for the distribution for the publication selection meta-analysis model

Usage

```
dpsnorm(x, theta, sigma, alpha = c(0, 0.025, 0.05, 1), eta, log = FALSE)

ppsnorm(
  q,
  theta,
  sigma,
  alpha = c(0, 0.025, 0.05, 1),
  eta,
  lower.tail = TRUE,
  log.p = FALSE
)

rpsnorm(n, theta, sigma, alpha = c(0, 0.025, 0.05, 1), eta)
```

Arguments

`x, q`	vector of quantiles.
`theta`	vector of means.
`sigma`	vector of study standard deviations.
`alpha`	vector of thresholds for publication bias.
`eta`	vector of publication probabilities, normalized to sum to 1.
`log, log.p`	logical; If `TRUE`, probabilities are given as `log(p)`.
`lower.tail`	logical; If `TRUE` (default), the probabilities are $P[X \leq x]$ otherwise, $P[X \geq x]$.
`n`	number of observations. If `length(n) > 1`, the length is taken to be the number required.

Details

The effect size distribution for the publication selection model is not normal, but has itself been selected for. These functions assume one-sided selection on the effects. These functions do not assume the existence of an underlying effect size distribution. For these, see `mpsnorm`.

Value

`dpsnorm` gives the density, `ppsnorm` gives the distribution function, and `rpsnorm` generates random deviates.

References

Hedges, Larry V. "Modeling publication selection effects in meta-analysis." Statistical Science (1992): 246-255.

Moss, Jonas and De Bin, Riccardo. "Modelling publication bias and p-hacking" Forthcoming (2019)

Examples

```
rpsnorm(100, theta = 0, sigma = 0.1, eta = c(1, 0.5, 0.1))
```

---

snorm   *Selected Normal Effect Size Distribution*

Description

Density, random variate generation, and expectation calculation for the effect size distribution of the one-sided normal publication bias model.

Usage

```
dsnorm(x, theta0, tau, sigma, alpha = c(0, 0.025, 0.05, 1), eta, log = FALSE)

rsnorm(n, theta0, tau, sigma, alpha = c(0, 0.025, 0.05, 1), eta)

esnorm(theta0, tau, sigma, alpha, eta)
```

Arguments

x	vector of quantiles.
theta0	vector of means.
tau	vector of heterogeneity parameters.
sigma	vector of study standard deviations.
alpha	vector of thresholds for publication bias.
eta	vector of publication probabilities, normalized to sum to 1.
log	logical; If TRUE, probabilities are given as log(p).
n	number of observations. If length(n) > 1, the length is taken to be the number required.

Details

The effect size distribution for the publication selection model is not normal, but has itself been selected for. These functions assume a normal underlying effect size distribution and one-sided selection on the effects.

Value

dsnorm gives the density, psnorm gives the distribution function, and rsnorm generates random deviates.

References

Hedges, Larry V. "Modeling publication selection effects in meta-analysis." Statistical Science (1992): 246-255.

Moss, Jonas and De Bin, Riccardo. "Modelling publication bias and p-hacking" (2019) arXiv:1911.12445

Examples

```
rsnorm(100, theta0 = 0, tau = 0.1, sigma = 0.1, eta = c(1, 0.5, 0.1))
```

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