## Package 'rvalues'

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Type Package

Title R-Values for Ranking in High-Dimensional Settings

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Imports graphics, stats, utils

**Description** A collection of functions for computing ``r-values" from various kinds of user input such as MCMC output or a list of effect size estimates and associated standard errors. Given a large collection of measurement units, the r-value, r, of a particular unit is a reported percentile that may be interpreted as the smallest percentile at which the unit should be placed in the top r-fraction of units.

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batavgs

**Batting Averages Data** 

#### Description

Data set containing number of at-bats and number of hits for Major League baseball players over the 2005 season.

#### Usage

data(batavgs)

#### Format

A data frame with 929 observations on the following 7 variables.

First.Name factor; player's last name

Last.Name factor; player's first name

Pitcher numeric vector; an indicator of whether or not the player is a pitcher

midseasonAB numeric vector; number of at-bats during the first half of the season

midseasonH numeric vector; number of hits during the first half of the season

TotalAB numeric vector; total number of at-bats over the season

TotalH numeric vector; total number of hits over the season

## Details

The 2005 Major League Baseball season was roughly six months starting from the beginning of April and ending at the beginning of October. Data from postseason play is not included. The midseason data were obtained by only considering the first three months of the season.

#### Source

http://projecteuclid.org/DPubS?service=UI&version=1.0&verb=Display&handle=euclid.aoas/1206367815

#### bcwest

#### References

Brown, L. D. (2008), In-Season prediction of batting averages: a field test of empirical Bayes and Bayes Methodologies, *The Annals of Applied Statistics*, **2**, 1, 113–152.

#### Examples

data(batavgs)
head(batavgs)

bcwest

Breast Cancer Gene Expression Data

## Description

Effect size estimates and standard errors obtained from gene expression measurements on 7129 genes across 49 samples.

#### Usage

data(bcwest)

#### Format

A data frame with 7129 observations on the following 2 variables.

estimates a vector of effect size estimates

std.err standard errors associated with effect size estimates

#### Details

A description of the original data may be found in West et al. (2001). For each gene, the effect size estimate was computed from the difference in the mean expression levels of the two groups (i.e., mean(bc-positive group) - mean(bc-negative group)).

#### Source

T. Hothorn, P. Buehlmann, T. Kneib, M. Schmid, and B. Hofner (2013). mboost: Model-Based Boosting, R package version 2.2-3, http://CRAN.R-project.org/package=mboost.

#### References

Mike West, Carrie Blanchette, Holly Dressman, Erich Huang, Seiichi Ishida, Rainer Spang, Harry Zuzan, John A. Olson Jr., Jeffrey R. Marks and Joseph R. Nevins (2001), Predicting the clinical status of human breast cancer by using gene expression profiles, *Proceedings of the National Academy of Sciences*, **98**, 11462-11467.

FDRCurve

#### Description

Estimates the expected proportion of misclassified units when using a given r-value threshold. If plot=TRUE, the curve is plotted before the estimated function is returned.

## Usage

FDRCurve(object, q, threshold = 1, plot = TRUE, xlim, ylim, xlab, ylab, main, ...)

## Arguments

object	An object of class "rvals"
q	A value in between 0 and 1; the desired level of FDR control.
threshold	The r-value threshold.
plot	logical; if TRUE, the estimated FDR curve is plotted.
xlim,ylim	x and y - axis limits for the plot
xlab,ylab	x and y - axis labels
main	the title of the plot
	additional arguments to plot.default

## Details

Consider parameters of interest  $(\theta_1, ..., \theta_n)$  with an effect of size of interest  $\tau$ . That is, a unit is taken to be "null" if  $\theta_i \leq \tau$  and taken to be "non-null" if  $\theta_i > \tau$ .

For r-values  $r_1, ..., r_n$  and a procedure which "rejects" units satisfying  $r_i \leq c$ , the FDR is defined to be

$$FDR(c) = P(\theta_i < \tau, r_i \le c) / P(r_i \le c).$$

FDRCurve estimates FDR(c) for values of c across (0,1) and plots (if plot=TRUE) the resulting curve.

#### Value

A list with the following two components

fdrcurve	A function which returns the estimated FDR for each r-value threshold.
Rval.cut	The largest r-value cutoff which still gives an estimated FDR less than q.

## Author(s)

Nicholas Henderson and Michael Newton

## fluEnrich

#### See Also

OverlapCurve

#### Examples

```
n <- 500
theta <- rnorm(n)
ses <- sqrt(rgamma(n,shape=1,scale=1))
XX <- theta + ses*rnorm(n)
dd <- cbind(XX,ses)
rvs <- rvalues(dd, family = gaussian)
FDRCurve(rvs, q = .1, threshold = .3, cex.main = 1.5)</pre>
```

fluEnrich

#### Flu Enrichment Data

#### Description

Gene-set enrichment for genes that have been identified as having an effect on influenza-virus replication.

#### Usage

data(fluEnrich)

## Format

A data frame with 5719 observations on the following 3 variables.

nflugenes number of genes both annotated to the given GO term and in the collection of flu genes

setsize number of genes annotated to the given GO term

GO\_terms the GO (gene ontology) term label

#### Details

These data were produced by associating the 984 genes (the collection of flu genes) identified in the Hao et al. (2013) meta-analysis with gene ontology (GO) gene sets (GO terms). In total, 17959 human genes were annotated to at least one GO term and 16572 GO terms were available, though this data set only contains the 5719 terms which annotated between 10 and 1000 human genes.

## References

Hao, L. Q. He, Z. Wang, M. Craven, M. A. Newton, and P. Ahlquist (2013). Limited agreement of independent RNAi screens for virus-required host genes owes more to false-negatives than false-positive factors. *PLoS computational biology*, **9**, 9, e1003235.

HIV Data Set

## Description

These data contain effect size estimates and standard errors obtained from gene expression measurements on 7680 genes across 8 samples.

## Usage

data(hiv)

## Format

A data frame with 7680 observations on the following 2 variables.

estimates a vector of effect size estimates

std.err standard errors associated with effect size estimates

## References

van't Wout, et. al. (2003), Cellular gene expression upon human immuno-deficiency virus type 1 infection of CD4+-T-Cell lines, *Journal ofVirology*, **77**, 1392–1402.

MakeGridGrid Construction

## Description

Computes a grid of points on the interval (0,1). This function is useful for constructing the "alphagrid" used in various r-value computations.

## Usage

```
MakeGrid(nunits, type = "log", ngrid = NULL, lower = 1/nunits, upper = 1 - lower)
```

#### Arguments

nunits	The number of units in the data for which r-values are to be calculated.
type	The type of grid; type can be set to type="uniform", type="log", or type="log.symmetric".
ngrid	a number specifying the number of grid points
lower	the smallest grid point; must be greater than zero
upper	the largest grid point; must be less than one

## hiv

## **MCMCtest**

## Details

If nunits  $\leq 1000$ , the default number of grid points is equal to nunits. When nunits > 1000, the default number of grid points is determined by

 $1000 + 25 * log(nunits - 1000) * (nunits - 1000)^{1/4}$ 

#### Value

A vector of grid points in (0,1).

## Author(s)

Nicholas Henderson and Michael Newton

### See Also

rvalues

## Examples

alpha.grid <- MakeGrid(1000,type="uniform",ngrid=200)</pre>

```
log.grid <- MakeGrid(40,type="log")
log.grid
hist(log.grid)</pre>
```

```
MCMCtest
```

MCMC sample output

## Description

a matrix of test MCMC output

#### Usage

data(MCMCtest)

#### Format

A 2000 x 400 numeric matrix. Data in the ith row should be thought of as a sample from the posterior for the ith case of interest.

## See Also

rvaluesMCMC

mroot

## Description

For a given multi-dimensional function with both a vector of lower bounds and upper bounds, mroot finds a vector such that each component of the function is zero.

#### Usage

```
mroot(f, lower, upper, ..., f.lower = f(lower, ...), f.upper = f(upper, ...),
tol = .Machine$double.eps^0.25, maxiter = 5000)
```

#### Arguments

f	the function for which the root is sought	
lower	a vector of lower end points	
upper	a vector of upper end points	
	additional arguments to be passed to f	
f.lower, f.upper		
	the same as f(lower) and f(upper)	
tol	the convergence tolerance	
maxiter	the maximum number of iterations	

## Details

The function f is from  $\mathbb{R}^n$  to  $\mathbb{R}^n$  with  $f(x_1, \ldots, x_n) = (f_1(x_1), \ldots, f_n(x_n))$ .

A root  $x = (x_1, \ldots, x_n)$  of f satisfies  $f_k(x_k) = 0$  for each component k.

lower =  $(l_1, \ldots, l_n)$  and upper =  $(u_1, \ldots, u_n)$  are both n-dimensional vectors such that, for each  $k, f_k$  changes sign over the interval  $[l_k, u_k]$ .

## Value

a vector giving the estimated root of the function

#### Author(s)

Nicholas Henderson

#### See Also

uniroot

## NBA1314

## Examples

```
ff <- function(x,a) {</pre>
    ans <- qnorm(x) - a
    return(ans)
}
n <- 10000
a <- rnorm(n)
low <- rep(0,n)
up <- rep(1,n)
## Find the roots of ff, first using mroot and
## then by using uniroot inside a loop.
system.time(mr <- mroot(ff, lower = low, upper = up, a = a))</pre>
ur <- rep(0,n)
system.time({
for(i in 1:n) {
   ur[i] <- uniroot(ff, lower = 0, upper = 1, a = a[i])$root</pre>
}
})
```

NBA1314

National Basketball Association, free throw data, 2013-2014 season

## Description

Free throw statistics on 482 active players, 2013-2014 season

## Usage

data(NBA1314)

## Format

A data frame with 482 players (rows) variables including.

RK rank of player by proportion of free throws made

PLAYER name of player

TEAM player's team

GP games played

PPG points per game

FTM0 FTM/GP

FTA0 FTA/GP

FTA free throws attempted

FTM free throws made

FTprop FTA/FTM

#### Details

Data obtained from ESPN.

#### References

See data analyzed in Henderson and Newton, 2015

#### Examples

```
data(NBA1314)
nba.dat <- cbind(NBA1314$FTM, NBA1314$FTA)
rownames(nba.dat) <- NBA1314$PLAYER
```

rvals.nba <- rvalues(nba.dat, family=binomial)</pre>

npmixapply

Apply Functions over estimated unit-specific posterior distributions

## Description

Using a nonparametric estimate of the mixing distribution, computes a posterior quantity of interest for each unit.

#### Usage

npmixapply(object, FUN, ...)

#### Arguments

object	an object of class "npmix"
FUN	a user provided function
	optional arguments to FUN

## Details

object is an object of class "npmix" containing a nonparametric estimate of the mixing distribution F in the following two-level sampling model  $X_i | \theta_i \sim p(x | \theta_i, \eta_i)$  and  $\theta_i \sim F$  for i = 1, ..., n.

Using npmixapply(object, f), then returns the posterior expectation of f:  $E[f(\theta_i)|X_i, \eta_i]$ , for i = 1, ..., n.

## Value

a vector with length equal to n

#### Author(s)

Nicholas Henderson

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## npmle

#### See Also

npmle

#### Examples

```
## Not run:
data(hiv)
npobj <- npmle(hiv, family = gaussian, maxiter = 4)
### Compute unit-specific posterior means
pmean <- npmixapply(npobj, function(x) { x })
### Compute post. prob that \theta_i < .1
pp <- npmixapply(npobj, function(x) { x < .1})
## End(Not run)
```

```
npmle
```

Maximum Likelihood Estimate of a Mixing Distribution.

## Description

Estimates the mixture distribution nonparametrically using an EM algorithm. The estimate is discrete with the results being returned as a vector of support points and a vector of associated mixture probabilities. The available choices for the sampling distribution include: Normal, Poisson, Binomial and t-distributions.

#### Usage

```
npmle(data, family = gaussian, maxiter = 500, tol = 1e-4,
      smooth = TRUE, bass = 0, nmix = NULL)
```

## Arguments

data	A data frame or a matrix with the number of rows equal to the number of sam- pling units. The first column should contain the main estimates, and the second column should contain the nuisance terms.
family	family determining the sampling distribution (see <i>family</i> )
maxiter	the maximum number of EM iterations
tol	the convergence tolerance
smooth	logical; whether or not to smooth the estimated cdf
bass	controls the smoothness level; only relevant if smooth=TRUE. Values of up to 10 indicate increasing smoothness.
nmix	optional; the number of mixture components

#### Details

Assuming the following two-level sampling model  $X_i | \theta_i \sim p(x | \theta_i, \eta_i)$  and  $\theta_i \sim F$  for i = 1, ..., n. The function npmle seeks to find an estimate of the mixing distribution F which maximizes the marginal log-likelihood

$$l(F) = \sum_{i} \int p(X_i|\theta, \eta_i) dF(\theta).$$

The distribution function maximizing l(F) is known to be discrete; and thus, the estimated mixture distribution is returned as a set of support points and associated mixture probabilities.

## Value

An object of class npmix which is a list containing at least the following components

support	a vector of estimated support points
mix.prop	a vector of estimated mixture proportions
Fhat	a function; obtained through interpolation of the estimated discrete cdf
fhat	a function; estimate of the mixture density
loglik	value of the log-likelihood at each iteration
convergence	0 indicates convergence; 1 indicates that convergence was not achieved
numiter	the number of EM iterations required

#### Author(s)

Nicholas Henderson and Michael Newton

### References

Laird, N.M. (1978), Nonparametric maximum likelihood estimation of a mixing distribution, *Journal of the American Statistical Association*, **73**, 805–811.

Lindsay, B.G. (1983), The geometry of mixture likelihoods: a general theory. *The Annals of Statistics*, **11**, 86–94

#### See Also

npmixapply

## Examples

```
## Not run:
data(hiv)
npobj <- npmle(hiv, family = tdist(df=6), maxiter = 25)
### Generate Binomial data with Beta mixing distribution
n <- 3000
theta <- rbeta(n, shape1 = 2, shape2 = 10)
ntrials <- rpois(n, lambda = 10)</pre>
```

## **OverlapCurve**

```
x <- rbinom(n, size = ntrials, prob = theta)
### Estimate mixing distribution
dd <- cbind(x,ntrials)
npest <- npmle(dd, family = binomial, maxiter = 25)
### compare with true mixture cdf
tt <- seq(1e-4,1 - 1e-4, by = .001)
plot(npest, lwd = 2)
lines(tt, pbeta(tt, shape1 = 2, shape2 = 10), lwd = 2, lty = 2)
## End(Not run)</pre>
```

OverlapCurve Overlap Curve

#### Description

Estimates the expected proportion of units in the top fraction and those deemed to be in the top fraction by the r-value procedure. If plot=TRUE, the curve is plotted before the estimated function is returned.

#### Usage

```
OverlapCurve(object, plot = TRUE, xlim, ylim, xlab, ylab, main, ...)
```

#### Arguments

object	An object of class "rvals"
plot	logical. If TRUE, the estimated overlap curve is plotted.
xlim, ylim	x and y - axis limits for the plot
xlab,ylab	x and y - axis labels
main	the title of the plot
	additional arguments to plot.default

#### Details

For parameters of interest  $\theta_1, ..., \theta_n$  and corresponding r-values  $r_1, ..., r_n$ , the overlap at a particular value of  $\alpha$  is defined to be

$$pverlap(\alpha) = P(\theta_i \ge \theta_\alpha, r_i \le \alpha),$$

where the threshold  $\theta_{\alpha}$  is the upper- $\alpha$ th quantile of the distribution of the  $\theta_i$  (i.e.,  $P(\theta_i \ge \theta_{\alpha}) = \alpha$ ). OverlapCurve estimates this overlap for values of alpha across (0,1) and plots (if plot=TRUE) the resulting curve.

#### Value

A function returning estimated overlap values.

#### Author(s)

Nicholas Henderson and Michael Newton

#### References

Henderson, N.C. and Newton, M.A. (2016). *Making the cut: improved ranking and selection for large-scale inference*. J. Royal Statist. Soc. B., 78(4), 781-804. doi: 10.1111/rssb.12131 https://arxiv.org/abs/1312.5776

## Examples

```
n <- 500
theta <- rnorm(n)
ses <- sqrt(rgamma(n,shape=1,scale=1))
XX <- theta + ses*rnorm(n)
dd <- cbind(XX,ses)
rvs <- rvalues(dd, family = gaussian)
OverlapCurve(rvs, cex.main = 1.5)</pre>
```

PostPercentile Posterior expected percentiles

#### Description

Computes posterior expected percentiles for both parametric and nonparametric models.

#### Usage

```
PostPercentile(object)
```

#### Arguments

object An object of class "rvals"

### Details

With parameters of interest  $\theta_1, ..., \theta_n$  the rank of the ith parameter (when we set the ranking so that the largest  $\theta_i$  gets rank 1) is defined as  $rank(\theta_i) = sum_j(\theta_j \ge \theta_i)$  and the associated percentile is  $perc(\theta_i) = rank(\theta_i)/(n+1)$ . The posterior expected percentile for the ith unit (see e.g., Lin et. al. (2006)) is simply the expected value of  $perc(\theta_i)$  given the data.

The function PostPercentile computes an asymptotic version of the posterior expected percentile, which is defined as

$$P(\theta_i \le \theta | data),$$

where  $\theta$  has the same distribution as  $\theta_i$  and is independent of both  $\theta_i$  and the data. See Henderson and Newton (2014) for additional details.

#### PostSummaries

## Value

A vector of estimated posterior expected percentiles.

## Author(s)

Nicholas Henderson and Michael Newton

#### References

Henderson, N.C. and Newton, M.A. (2016). *Making the cut: improved ranking and selection for large-scale inference*. J. Royal Statist. Soc. B., 78(4), 781-804. doi: 10.1111/rssb.12131 https://arxiv.org/abs/1312.5776

Lin, R., Louis, T.A., Paddock, S.M., and Ridgeway, G. (2006). Loss function based ranking in two-stage, hierarchical models. *Bayesian Analysis*, **1**, 915–946.

## See Also

#### rvalues

#### Examples

```
n <- 3000
theta <- rnorm(n, sd = 3)
ses <- sqrt(rgamma(n, shape = 1, scale = 1))
XX <- theta + ses*rnorm(n)
dd <- cbind(XX,ses)
rv <- rvalues(dd, family = gaussian)
perc <- PostPercentile(rv)
plot(rv$rvalues, perc)
```

PostSummaries *R-values from posterior summary quantities* 

#### Description

Computes r-values assuming that, for each parameter of interest, the user supplies a value for the posterior mean and the posterior standard deviation. The assumption here is that the posterior distributions are Normal.

#### Usage

```
PostSummaries(post.means, post.sds, hypers = NULL, qtheta = NULL, alpha.grid = NULL,
    ngrid = NULL, smooth = 0)
```

## Arguments

post.means	a vector of posterior means
post.sds	a vector of posterior standard deviations
hypers	a list with two elements: mean and sd. These represent the parameters in the (Normal) prior which was used to generate the posterior means and sds. If hypers is not supplied then one must supply the quantile function qtheta.
qtheta	a function which returns the quantiles (for upper tail probs.) of theta. If this is not supplied, the hyperparameter must be supplied.
alpha.grid	grid of values in $(0,1)$ ; used for the discrete approximation approach for computing r-values.
ngrid	number of grid points for alpha.grid; only relevant when alpha.grid = NULL
smooth	either smooth="none" or smooth takes a value between 0 and 10; this determines the level of smoothing applied to the estimate of $\lambda(\alpha)$ ; if smooth is given a number, the number is used as the bass argument in supsmu.

## Value

An object of class "rvals"

## Author(s)

Nicholas Henderson and Michael Newton

## See Also

rvalues

## Examples

```
n <- 500
theta <- rnorm(n)
sig_sq <- rgamma(n,shape=1,scale=1)
X <- theta + sqrt(sig_sq)*rnorm(n)</pre>
```

```
pm <- X/(sig_sq + 1)
psd <- sqrt(sig_sq/(sig_sq + 1))</pre>
```

```
rvs <- PostSummaries(post.means=pm,post.sds=psd,hypers=list(mean=0,sd=1))
hist(rvs$rvalues)</pre>
```

rvalueBoot

## Description

Estimates a new prior for each bootstrap replications ... (need to add)

## Usage

rvalueBoot(object, statistic = median, R, type = "nonparametric")

## Arguments

object	An object of class "rvals"
statistic	The statistic used to summarize the bootstrap replicates.
R	Number of bootstrap replicates
type	Either type="nonparametric" or type="parametric"; the nonparametric type corresponds to the usual bootstrap where units are sampled with replacement.

## Details

When type="nonparametric", the prior is re-estimated (using the resampled data) in each bootstrap replication, and r-values are re-computed with respect to this new model. When type="parametric",

#### Value

A list with the following two components

rval.repmat	A matrix where each column corresponds to a separate bootstrap replication.
rval.boot	A vector of r-values obtained by applying the statistic to each row of rval.repmat

## Author(s)

Nicholas Henderson and Michael Newton

#### References

Henderson, N.C. and Newton, M.A. (2016). Making the cut: improved ranking and selection for large-scale inference. J. Royal Statist. Soc. B., 78(4), 781-804. doi: 10.1111/rssb.12131 https: //arxiv.org/abs/1312.5776

## See Also

rvalues

rvalues

#### Examples

```
## Not run:
n <- 3000
theta <- rnorm(n, sd = 3)
ses <- sqrt(rgamma(n, shape = 10, rate = 1))
XX <- theta + ses*rnorm(n)
dd <- cbind(XX,ses)
rv <- rvalues(dd, family = gaussian, prior = "conjugate")
rvb <- rvalueBoot(rv, R = 10)
summary(rvb$rval.repmat[512,])
## End(Not run)
```

rvalues

R-values

#### Description

Given data on a collection of units, this function computes r-values which are percentiles constructed to maximize the agreement between the reported percentiles and the percentiles of the effect of interest. Additional details about r-values are provided below and can also be found in the listed references.

### Usage

#### Arguments

data	A data frame or a matrix with the number of rows equal to the number of sam- pling units. The first column should contain the main estimates, and the second column should contain the nuisance terms.
family	An argument which determines the sampling distribution; this could be either family = gaussian, family = tdist, family = binomial, family = poisson
hypers	values of the hyperparameters; only meaningful when the conjugate prior is used; if set to "estimate", the hyperparameters are found through maximum like- lihood; if not set to "estimate" the user should supply a vector of length two.
prior	the form of the prior; either prior="conjugate" or prior="nonparametric".
alpha.grid	a numeric vector of points in $(0,1)$ ; this grid is used in the discrete approximation of r-values
ngrid	number of grid points for alpha.grid; only relevant when alpha.grid=NULL

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#### rvalues

smooth	either smooth="none" or smooth takes a value between 0 and 10; this determines the level of smoothing applied to the estimate of $\lambda(\alpha)$ (see below for the definition of $\lambda(\alpha)$ ); if smooth is given a number, the number is used as the bass argument in supsmu.
control	a list of control parameters for estimation of the prior; only used when the prior is nonparametric

#### Details

The r-value computation assumes the following two-level sampling model  $X_i|\theta_i \sim p(x|\theta_i, \eta_i)$  and  $\theta_i \sim F$ , for i = 1, ..., n, with parameters of interest  $\theta_i$ , effect size estimates  $X_i$ , and nuisance terms  $\eta_i$ . The form of  $p(x|\theta_i, \eta_i)$  is determined by the family argument. When family = gaussian, it is assumed that  $X_i|\theta_i, \eta_i \sim N(\theta_i, \eta_i^2)$ . When family = binomial, the  $(X_i, \eta_i)$  represent the number of successes and number of trials respectively, and it is assumed that  $X_i|\theta_i, \eta_i \sim \text{Binomial}(\theta_i, \eta_i)$ . When family = poisson, the  $X_i$  should be counts, and it is assumed that  $X_i|\theta_i, \eta_i \sim \text{Poisson}(\theta_i * \eta_i)$ .

The distribution of the effect sizes F may be a parametric distribution that is conjugate to the corresponding family argument, or it may be estimated nonparametrically. When it is desired that F be parametric (i.e., prior = "conjugate"), the default is to estimate the hyperparameters (i.e., hypers = "estimate"), but these may be supplied by the user as a vector of length two. To estimate F nonparametrically, one should use prior = "nonparametric" (see npmle for further details about nonparametric estimation of F).

The *r*-value,  $r_i$ , assigned to the ith case of interest is determined by  $r_i = \inf[0 < \alpha < 1 : V_{\alpha}(X_i, \eta_i) \ge \lambda(\alpha)]$  where  $V_{\alpha}(X_i, \eta_i) = P(\theta_i \ge \theta_{\alpha} | X_i, \eta_i)$  is the posterior probability that  $\theta_i$  exceeds the threshold  $\theta_{\alpha}$ , and  $\lambda(\alpha)$  is the upper- $\alpha$ th quantile associated with the marginal distribution of  $V_{\alpha}(X_i, \eta_i)$  (i.e.,  $P(V_{\alpha}(X_i, \eta_i) \ge \lambda(\alpha)) = \alpha$ ). Similarly, the threshold  $\theta_{\alpha}$  is the upper- $\alpha$ th quantile of F (i.e.,  $P(\theta_i \ge \theta_{\alpha}) = \alpha$ ).

#### Value

An object of class "rvals" which is a list containing at least the following components:

main	a data frame containing the r-values, the r-value rankings along with the rankings from several other common procedures
aux	a list containing other extraneous information
rvalues	a vector of r-values

#### Author(s)

Nicholas C. Henderson and Michael A. Newton

#### References

Henderson, N.C. and Newton, M.A. (2016). *Making the cut: improved ranking and selection for large-scale inference*. J. Royal Statist. Soc. B., 78(4), 781-804. doi: 10.1111/rssb.12131 https://arxiv.org/abs/1312.5776

## See Also

rvaluesMCMC, PostSummaries, Valpha

## Examples

```
## Not run:
### Binomial example with Beta prior:
data(fluEnrich)
flu.rvals <- rvalues(fluEnrich, family = binomial)
hist(flu.rvals$rvalues)
### look at the r-values for indices 10 and 2484
fig_indices <- c(10,2484)
fluEnrich[fig_indices,]
flu.rvals$rvalues[fig_indices]
### Gaussian sampling distribution with nonparametric prior
### Use a maximum of 5 iterations for the nonparam. estimate
data(hiv)
hiv.rvals <- rvalues(hiv, prior = "nonparametric")</pre>
```

## End(Not run)

rvaluesMCMC *R-values from MCMC output.* 

#### Description

Returns r-values from an array of MCMC output.

## Usage

```
rvaluesMCMC(output, qtheta, alpha.grid = NULL, ngrid = NULL, smooth = "none")
```

## Arguments

output	a matrix contatining mcmc ouput. The ith row should represent a sample from the posterior of the ith parameter of interest.
qtheta	either a function which returns the quantiles (for upper tail probs.) of theta or a vector of theta-quantiles.
alpha.grid	grid of values in (0,1); used for the discrete approximation approach for com- puting r-values.
ngrid	number of grid points for alpha.grid; only relevant when alpha.grid = NULL
smooth	either smooth="none" or smooth takes a value between 0 and 10; this determines the level of smoothing applied to the estimate of $\lambda(\alpha)$ ; if smooth is given a number, the number is used as the bass argument in supsmu.

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### *rvaluesMCMC*

## Value

An object of class "rvals" which is a list containing at least the following components:

main	a data frame containing the r-values, the r-value rankings along with the rankings from several other common procedures
aux	a list containing other extraneous information
rvalues	a vector of r-values

## Author(s)

Nicholas Henderson and Michael Newton

## References

Henderson, N.C. and Newton, M.A. (2016). *Making the cut: improved ranking and selection for large-scale inference*. J. Royal Statist. Soc. B., 78(4), 781-804. doi: 10.1111/rssb.12131 https://arxiv.org/abs/1312.5776

## See Also

rvalues, PostSummaries

#### Examples

data(MCMCtest)

```
### For the MCMC output in MCMC_test, the prior assumed for the effect sizes of
### interest was a mixture of two t-distributions. The function qthetaTMix
### computes the quantiles for this prior.
```

```
qthetaTMix <- function(p) {</pre>
    ### function to compute quantiles (for upper tail probabilities) for a
    ### mixture of two t-distributions
    mu <- c(.35,-.12)
    sig <- c(.2,.08)
    mix.prop <- c(.25,.75)
    ff <- function(x,pp) {</pre>
        prob_less <- 0</pre>
        for(k in 1:2) {
         prob_less <- prob_less + pt((x - mu[k])/sig[k],df=4,lower.tail=FALSE)*mix.prop[k]</pre>
        }
        return(prob_less - pp)
    }
    nn <- length(p)</pre>
    ans <- numeric(nn)</pre>
    for(i in 1:nn) {
        ans[i] <- uniroot(ff,interval=c(-5,5),tol=1e-6,pp=p[i])$root</pre>
    }
    return(ans)
```

rvs <- rvaluesMCMC(MCMCtest, qtheta = qthetaTMix)</pre>

tdist

}

#### t-distribution family object

## Description

A t-distribution family object which allows one to specify a t-density for the sampling distribution. Modeled after family objects often used in the glm function.

## Usage

tdist(df)

## Arguments

df

vector containing the degrees of freedom

## Value

An object of class "newfam", which is a list containing the following components

family	The family name
df	The degrees of freedom

## Author(s)

Nicholas Henderson and Michael Newton

## See Also

family, glm, npmle

#### Examples

a <- tdist(df=5)

TopList

#### Description

Returns a list of the top units ranked according to "r-value" or another specified statistic.

## Usage

```
TopList(object, topnum = 10, sorted.by = c("RValue","PostMean","MLE","PVal"))
```

## Arguments

object	An object of class "rvals"
topnum	The length of the top list.
sorted.by	The statistic by which to sort; this could be sorted.by = "RValue", sorted.by = "PostMean", sorted.by = "MLE", or sorted.by = "PVal"

## Value

a data frame with topnum rows and columns containing the r-value, mle, posterior mean, and p-value rankings.

## Author(s)

Nicholas Henderson and Michael Newton

## See Also

rvalues

## Examples

```
n <- 500
theta <- rnorm(n)
ses <- sqrt(rgamma(n,shape=1,scale=1))
XX <- theta + ses*rnorm(n)
dd <- cbind(XX,ses)
rvs <- rvalues(dd, family = gaussian)
TopList(rvs, topnum = 12)
TopList(rvs, topnum = 15, sorted.by = "MLE")</pre>
```

Valpha

### Description

Computes r-values directly from a "Valpha" matrix V where each column of Valpha contains posterior tail probabilities relative to a threshold indexed by alpha.

## Usage

```
Valpha(V, alpha.grid, smooth = "none")
```

## Arguments

V	a numeric vector with $(i,j)$ entry: $V[i,j] = P(\text{theta}_i \ge \text{theta}_i] \text{data})$
alpha.grid	grid of values in (0,1); used for the discrete approximation approach for computing r-values.
smooth	either smooth="none" or smooth takes a value between 0 and 10; this determines the level of smoothing applied to the estimate of $\lambda(\alpha)$ ; if smooth is given a number, the number is used as the bass argument in supsmu.

## Value

A list with the following components

rvalues	a vector of computed r-values
Vmarginals	The estimated V-marginals along the alpha grid points
Vmarginals.smooth	
	a function obtained through interpolation and smoothing (if desired) the Vmarginals;
	i.e., an estimate of $\lambda(\alpha)$ (see rvalues)

## Author(s)

Nicholas Henderson and Michael Newton

## References

Henderson, N.C. and Newton, M.A. (2016). *Making the cut: improved ranking and selection for large-scale inference*. J. Royal Statist. Soc. B., 78(4), 781-804. doi: 10.1111/rssb.12131 https://arxiv.org/abs/1312.5776

## See Also

rvalues rvaluesMCMC

Valpha

## Examples

```
## Not run:
data(fluEnrich)
rvobj <- rvalues(fluEnrich, family = binomial)
Vrvals <- Valpha(rvobj$aux$V, rvobj$aux$alpha.grid)</pre>
```

## End(Not run)

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