vsn

April 20, 2011

---

vsnInput   *Class to contain input data and parameters for vsn functions*

Description

Class to contain input data and parameters for vsn functions

Creating Objects

new("vsnInput")

Slots

x: A numeric matrix with the input data.

reference: An object of vsn, typically this would have been obtained from a previous fit to a set of reference arrays (data).

strata: A factor of length 0 or n. If its length is n, then its levels correspond to different normalization strata (see vsn2).

ordered: Logical scalar; are the rows reordered so that the strata are contiguous.

lts.quantile: Numeric scalar, see vsn2.

subsample: Integer scalar, see vsn2.

verbose: Logical scalar, see vsn2.

calib Character of length 1, see manual page of vsn2.

pstart: A 3D array of size (number of strata) x (number of columns of the data matrix) x 2. It contains the start parameters.

optimpar: List with parameters for the numerical optimiser L-BFGS-B; see the manual page of vsn2.

Methods

[Subset

dim Get dimensions of data matrix.

nrow Get number of rows of data matrix.

ncol Get number of columns of data matrix.

show Print a summary of the object

Author(s)

Wolfgang Huber

See Also

[vsn2](#)

---

vsn *Class to contain result of a vsn fit*

Description

Class to contain result of a vsn fit

Creating Objects

new("vsn") vsn2(x) with x being an [ExpressionSet](#).

Slots

coefficients: A 3D array of size (number of strata) x (number of columns of the data matrix) x 2. It contains the fitted normalization parameters (see vignette).

strata: A factor of length 0 or n. If its length is n, then its levels correspond to different normalization strata (see vignette).

mu: A numeric vector of length n with the fitted parameters $\hat{\mu}_k$, for $k = 1, ..., n$.

sigsq: A numeric scalar, $\hat{\sigma}^2$.

hx: A numeric matrix with 0 or n rows. If the number of rows is n, then hx contains the transformed data matrix.

lbfgsb: An integer scalar containing the return code from the L-BFGS-B optimizer.

hoffset: Numeric scalar, the overall offset c- see manual page of [vsn2](#).

calib: Character of length 1, see manual page of [vsn2](#).

Methods

[Subset

dim Get dimensions of data matrix.

nrow Get number of rows of data matrix.

ncol Get number of columns of data matrix.

show Print a summary of the object

exprs Accessor to slot hx.

coef, coefficients Accessors to slot coefficients.

Author(s)

Wolfgang Huber

See Also

vsn2

Examples

```
data("kidney")
v = vsn2(kidney)
show(v)
dim(v)
v[1:10, ]
```

---

justvsn   *Wrapper functions for vsn*

Description

justvsn is equivalent to calling

```
fit = vsn2(x, ...)
nx = predict(fit, newdata=x, useDataInFit = TRUE)
```

vsnrma is a wrapper around vsn2 and rma.

Usage

```
justvsn(x, ...)
vsnrma(x, ...)
```

Arguments

x   For justvsn, any kind of object for which vsn2 methods exist. For vsnrma,
                an AffyBatch.

...   Further arguments that get passed on to vsn2.

Details

vsnrma does probe-wise background correction and between-array normalization by calling vsn2
on the perfect match (PM) values only. Probeset summaries are calculated with the medianpolish
algorithm of rma.

Value

justvsn returns the vsn-normalised intensities in an object generally of the same class as its first
argument (see the man page of predict for details). It preserves the metadata.

vsnrma returns an ExpressionSet.

Author(s)

Wolfgang Huber

See Also

> vsn2

Examples

```
##-------------------------------------------------
## use "vsn2" to produce a "vsn" object
##-------------------------------------------------
data("kidney")
fit = vsn2(kidney)
nkid = predict(fit, newdata=kidney)

##-------------------------------------------------
## justvsn on ExpressionSet
##-------------------------------------------------
nkid2 = justvsn(kidney)
stopifnot(identical(exprs(nkid), exprs(nkid2)))

##-------------------------------------------------
## justvsn on RGList
##-------------------------------------------------
rg = new("RGList", list(R=exprs(kidney)[,1,drop=FALSE], G=exprs(kidney)[,2,drop=FALSE]))
erge = justvsn(rg)
```

---

kidney	*Intensity data for 1 cDNA slide with two adjacent tissue samples from a nephrectomy (kidney)*

Description

Intensity data for 1 cDNA slide with two adjacent tissue samples from a nephrectomy (kidney)

Usage

```
data(kidney)
```

Format

kidney is an ExpressionSet containing the data from one cDNA chip. The 8704x2 matrix exprs(kidney) contains the spot intensities for the red (635 nm) and green color channels (532 nm) respectively. For each spot, a background estimate from a surrounding region was subtracted.

Details

The chip was produced in 2001 by Holger Sueltmann at the Division of Molecular Genome Analysis at the German Cancer Research Center in Heidelberg.

References

Huber W, Boer JM, von Heydebreck A, Gunawan B, Vingron M, Fuzesi L, Poustka A, Sueltmann H. Transcription profiling of renal cell carcinoma. Verh Dtsch Ges Pathol. 2002;86:153-64. PMID: 12647365

See Also

vsn

Examples

```
data(kidney)
plot(exprs(kidney), pch=".", log="xy")
abline(a=0,b=1,col="blue")
```

---

lymphoma   *Intensity data for 8 cDNA slides with CLL and DLBL samples from the*
                                  Alizadeh et al. paper in Nature 2000

Description

8 cDNA chips from Alizadeh lymphoma paper

Usage

```
data(lymphoma)
```

Format

lymphoma is an ExpressionSet containing the data from 8 chips from the lymphoma data set
by Alizadeh et al. (see references). Each chip represents two samples: on color channel 1 (CH1,
Cy3, green) the common reference sample, and on color channel 2 (CH2, Cy5, red) the various
disease samples. See pData(lymphoma). The 9216x16 matrix exprs(lymphoma) contains
the background-subtracted spot intensities (CH1I-CH1B and CH2I-CH2B, respectively).

Details

The chip intensity files were downloaded from the Stanford microarray database. Starting from
the link below, this was done by following the links *Published Data -> Alizadeh AA, et al. (2000)*
Nature 403(6769):503-11 -> Data in SMD -> Display Data, and selecting the following 8 slides:

> lc7b019
> lc7b047
> lc7b048
> lc7b056
> lc7b057
> lc7b058
> lc7b069
> lc7b070

Then, the script makedata.R from the scripts subdirectory of this package was run to generate
the R data object.

Source

http://genome-www5.stanford.edu/MicroArray/SMD

References

A. Alizadeh et al., Distinct types of diffuse large B-cell lymphoma identified by gene expression profiling. Nature 403(6769):503-11, Feb 3, 2000.

See Also

vsn

Examples

```
data(lymphoma)
lymphoma
pData(lymphoma)
```

---

meanSdPlot   *Plot row standard deviations versus row means*

Description

Methods for objects of classes matrix, ExpressionSet, vsn and MAList to plot row standard deviations versus row means.

Usage

```
meanSdPlot(x,
           ranks = TRUE,
           xlab  = ifelse(ranks, "rank(mean)", "mean"),
           ylab  = "sd",
           pch   = ".",
           plot  = TRUE,
           ...)
```

Arguments

x	An object of class matrix, ExpressionSet, vsn or MAList.
ranks	Logical, indicating whether the x-axis (means) should be plotted on the original scale (FALSE) or on the rank scale (TRUE). The latter distributes the data more evenly along the x-axis and allows a better visual assessment of the standard deviation as a function of the mean.
xlab	Character, label for the x-axis.
ylab	Character, label for the y-axis.
pch	Plot symbol.
plot	Logical. If TRUE (default), a plot is produced. Calling the function with plot=FALSE can be useful if only its return value is of interest.
...	Further arguments that get passed to plot.default.

Details

Standard deviation and mean are calculated row-wise from the expression matrix (in) x. The scatterplot of these versus each other allows to visually verify whether there is a dependence of the standard deviation (or variance) on the mean. The red dots depict the running median estimator (window-width 10%). If there is no variance-mean dependence, then the line formed by the red dots should be approximately horizontal.

Value

A named list with four components: its elements px and py are the x- and y-coordinates of the individual data points in the plot; its first and second element are the x-coordinates and values of the running median estimator (the red dots in the plot). Depending on the value of plot, the method can also have a side effect, which is to create a plot on the active graphics device.

Author(s)

Wolfgang Huber

See Also

vsn

Examples

```
data(kidney)
log.na = function(x) log(ifelse(x>0, x, NA))

exprs(kidney) = log.na(exprs(kidney))
meanSdPlot(kidney)

## ...try this out with non-logged data, the lymphoma data, your data...
```

---

normalize.AffyBatch.vsn

Wrapper for vsn to be used as a normalization method with expresso

---

Description

Wrapper for vsn2 to be used as a normalization method with the expresso function of the package affy. The expresso function is deprecated, consider using justvsn instead. The normalize.AffyBatch.vsn can still be useful on its own, as it provides some additional control of the normalization process (fitting on subsets, alternate transform parameters).

Usage

```
normalize.AffyBatch.vsn(
    abatch,
    reference,
    strata = NULL,
    subsample = if (nrow(exprs(abatch))>30000L) 30000L else 0L,
    subset,
    log2scale = TRUE,
```

```
log2asymp=FALSE,
...)
```

Arguments

abatch   An object of type [AffyBatch](#).

reference   Optional, a 'vsn' object from a previous fit. If this argument is specified, the
              data in 'x' are normalized "towards" an existing set of reference arrays whose
              parameters are stored in the object 'reference'. If this argument is not specified,
              then the data in 'x' are normalized "among themselves". See [vsn2](#) for details.

strata   The 'strata' functionality is not supported, the parameter is ignored.

subsample   Is passed on to [vsn2](#).

subset   This allows the specification of a subset of expression measurements to be used
              for the vsn fit. The transformation with the parameters of this fit is then, how-
              ever, applied to the whole dataset. This is useful for excluding expression mea-
              surements that are known to be differentially expressed or control probes that
              may not match the vsn model, thus avoiding that they influence the normaliza-
              tion process. This operates at the level of probesets, not probes. Both 'subset'
              and 'subsample' can be used together.

log2scale   If TRUE, this will perform a global affine transform on the data to put them
              on a similar scale as the original non-transformed data. Many users prefer this.
              Fold-change estimates are not affected by this transform. In some situations,
              however, it may be helpful to turn this off, e.g., when comparing independently
              normalized subsets of the data.

log2asymp   If TRUE, this will perform a global affine transform on the data to make the gen-
              eralized log (asinh) transform be asymptotically identical to a log base 2 trans-
              form. Some people find this helpful. Only **one** of 'log2scale' or 'log2asymp'
              can be set to TRUE. Fold-change estimates are not affected by this transform.

...   Further parameters for [vsn2](#).

Details

Please refer to the *Details* and *References* sections of the man page for [vsn2](#) for more details about
this method.

Important note: after calling [vsn2](#), the function normalize.AffyBatch.vsn **exponenti-
ates** the data (base 2). This is done in order to make the behavior of this function similar to the
other normalization methods in affy. That packages uses the convention of taking the logarithm to
base in subsequent analysis steps (e.g. in [medpolish](#)).

Value

An object of class [AffyBatch](#). The vsn object returned, which can be used as reference for
subsequent fits, is provided by description(abatch)@preprocessing$vsnReference.

Author(s)

D. P. Kreil <http://bioinf.boku.ac.at/>, Wolfgang Huber

See Also

[vsn2](#)

Examples

```
## Please see vignette.
```

---

sagmbSimulateData *Simulate data and assess vsn's parameter estimation*

Description

Functions to validate and assess the performance of vsn through simulation of data.

Usage

```
sagmbSimulateData(n=8064, d=2, de=0, up=0.5, nrstrata=1,  miss=0, log2scale=FALS
sagmbAssess(h1, sim)
```

Arguments

n	Numeric. Number of probes (rows).
d	Numeric. Number of arrays (columns).
de	Numeric. Fraction of differentially expressed genes.
up	Numeric. Fraction of up-regulated genes among the differentially expressed genes.
nrstrata	Numeric. Number of probe strata.
miss	Numeric. Fraction of data points that is randomly sampled and set to NA.
log2scale	Logical. If TRUE, glog on base 2 is used, if FALSE, (the default), then base e.
h1	Matrix. Calibrated and transformed data, according, e.g., to vsn
sim	List. The output of a previous call to sagmbSimulateData, see Value

Details

Please see the vignette.

Value

For sagmbSimulateData, a list with four components: hy, an n x d matrix with the true (=simulated) calibrated, transformed data; y, an n x d matrix with the simulated uncalibrated raw data - this is intended to be fed into vsn; is.de, a logical vector of length n, specifying which probes are simulated to be differentially expressed. strata, a factor of length n.

For sagmbSimulateData, a number: the root mean squared difference between true and estimated transformed data.

Author(s)

Wolfgang Huber

References

Wolfgang Huber, Anja von Heydebreck, Holger Sueltmann, Annemarie Poustka, and Martin Vingron (2003) "Parameter estimation for the calibration and variance stabilization of microarray data", Statistical Applications in Genetics and Molecular Biology: Vol. 2: No. 1, Article 3. http://www.bepress.com/sagmb/vol2/iss1/art3

See Also

[vsn](vsn)

Examples

```
sim <- sagmbSimulateData(nrstrata=4)
ny  <- vsn(sim$y, strata=sim$strata)
res <- sagmbAssess(exprs(ny), sim)
res
```

---

scalingFactorTransformation

The transformation that is applied to the scaling parameter of the vsn model

---

Description

The transformation that is applied to the scaling parameter of the vsn model

Usage

```
scalingFactorTransformation(b)
```

Arguments

b   Real vector.

Value

A real vector of same length as b, with transformation `f` applied (see vignette *Likelihood Calculations for vsn*).

Author(s)

Wolfgang Huber

Examples

```
b  = seq(-3, 2, length=20)
fb = scalingFactorTransformation(b)
if(interactive())
  plot(b, fb, type="b", pch=16)
```

---

vsn2   *Fit the vsn model*

Description

`vsn2` fits the vsn model to the data in `x` and returns a `vsn` object with the fit parameters and the transformed data matrix. The data are, typically, feature intensity readings from a microarray, but this function may also be useful for other kinds of intensity data that obey an additive-multiplicative error model. To obtain an object of the same class as `x`, containing the normalised data and the same metdata as `x`, use

```
fit = vsn2(x, ...)
nx = predict(fit, newdata=x)
```

or the wrapper `justvsn`. Please see the vignette *Introduction to vsn*.

Usage

```
vsnMatrix(x,
          reference,
          strata,
          lts.quantile = 0.9,
          subsample    = 0L,
          verbose      = interactive(),
          returnData   = TRUE,
          calib        = "affine",
          pstart,
          minDataPointsPerStratum = 42L,
          optimpar     = list(),
          defaultpar   = list(factr=5e7, pgtol=2e-4, maxit=60000L,
                              trace=0L, cvg.niter=7L, cvg.eps=0))

## S4 method for signature 'ExpressionSet':
vsn2(x, reference, strata, ...)

## S4 method for signature 'AffyBatch':
vsn2(x, reference, strata, subsample, ...)

## S4 method for signature 'NChannelSet':
vsn2(x, reference, strata, backgroundsubtract=FALSE,
       foreground=c("R","G"), background=c("Rb", "Gb"), ...)

## S4 method for signature 'RGList':
vsn2(x, reference, strata, ...)
```

Arguments

x   An object containing the data to which the model is fitted.

reference   Optional, a [vsn](vsn) object from a previous fit. If this argument is specified, the
                     data in x are normalized "towards" an existing set of reference arrays whose pa-
                     rameters are stored in the object reference. If this argument is not specified,
                     then the data in x are normalized "among themselves". See Details for a more
                     precise explanation.

strata   Optional, a factor or integer whose length is nrow(x). It can be used
                     for stratified normalization (i.e. separate offsets a and factors b for each level of
                     strata). If missing, all rows of x are assumed to come from one stratum. If
                     strata is an integer, its values must cover the range $1, \ldots, n$, where n is the
                     number of strata.

lts.quantile   Numeric of length 1. The quantile that is used for the resistant least trimmed
                     sum of squares regression. Allowed values are between 0.5 and 1. A value of 1
                     corresponds to ordinary least sum of squares regression.

subsample   Integer of length 1. If its value is greater than 0, the model parameters are es-
                     timated from a subsample of the data of size subsample only, yet the fitted
                     transformation is then applied to all data. For large datasets, this can substan-
                     tially reduce the CPU time and memory consumption at a negligible loss of
                     precision. Note that the AffyBatch method of vsn2 sets a value of 30000
                     for this parameter if it is missing from the function call - which is different from
                     the behaviour of the other methods.

backgroundsubtract
                     Logical of length 1: should local background estimates be subtracted before
                     fitting vsn?

foreground, background
                     Aligned character vectors of the same length, naming the channels of x that
                     should be used as foreground and background values.

verbose   Logical. If TRUE, some messages are printed.

returnData   Logical. If TRUE, the transformed data are returned in a slot of the resulting
                     [vsn](vsn) object. Setting this option to FALSE allows saving memory if the data are
                     not needed.

calib   Character of length 1. Allowed values are affine and none. The default,
                     affine, corresponds to the behaviour in package versions <= 3.9, and to what
                     is described in references [1] and [2]. The option none is an experimental new
                     feature, in which no affine calibration is performed and only two global variance
                     stabilisation transformation parameters a and b are fitted. This functionality
                     might be useful in conjunction with other calibration methods, such as quantile
                     normalisation - see the vignette *Introduction to vsn*.

pstart   Optional, a three-dimensional numeric array that specifies start values for the
                     iterative parameter estimation algorithm. If not specified, the function tries
                     to guess useful start values. The first dimension corresponds to the levels of
                     strata, the second dimension to the columns of x and the third dimension
                     must be 2, corresponding to offsets and factors.

minDataPointsPerStratum
                     The minimum number of data points per stratum. Normally there is no need for
                     the user to change this; refer to the vignette for further documentation.

optimpar   Optional, a list with parameters for the likelihood optimisation algorithm. De-
                     fault parameters are taken from defaultpar. See details.

defaultpar   The default parameters for the likelihood optimisation algorithm. Values in
                     optimpar take precedence over those in defaultpar. The purpose of this

argument is to expose the default values in this manual page - it is not intended to be changed, please use `optimpar` for that.

`...` Arguments that get passed on to `vsnMatrix`.

Value

An object of class vsn.

Note on overall scale and location of the glog transformation

The data are returned on a $glog$ scale to base 2. More precisely, the transformed data are subject to the transformation $glog_2(f(b)*x+a)+c$, where the function $glog_2(u) = log_2(u + \sqrt{u*u+1}) = asinh(u)/\log(2)$ is called the generalised logarithm, the offset a and the scaling parameter b are the fitted model parameters (see references), and $f(x) = \exp(x)$ is a parameter transformation that allows ensuring positivity of the factor in front of x while using an unconstrained optimisation over b [4]. The overall offset c is computed from the b's such that for large x the transformation approximately corresponds to the log_2 function. This is done separately for each stratum, but with the same value across arrays. More precisely, if the element `b[s,i]` of the array b is the scaling parameter for the `s`-th stratum and the `i`-th array, then `c[s]` is computed as `log2(2*f(mean(b[,i])))`. The offset c is inconsequential for all differential expression calculations, but many users like to see the data in a range that they are familiar with.

Specific behaviour of the different methods

vsn2 methods exist for ExpressionSet, NChannelSet, AffyBatch (from the `affy` package), RGList (from the `limma` package), `matrix` and `numeric`. If x is an NChannelSet, then vsn2 is applied to the matrix that is obtained by horizontally concatenating the color channels. Optionally, available background estimates can be subtracted before. If x is an RGList, it is converted into an NChannelSet using a copy of Martin Morgan's code for RGList to NChannelSet coercion, then the NChannelSet method is called.

Standalone versus reference normalisation

If the `reference` argument is *not* specified, then the model parameters μ_k and σ are fit from the data in x. This is the mode of operation described in [1] and that was the only option in versions 1.X of this package. If `reference` is specified, the model parameters μ_k and σ are taken from it. This allows for 'incremental' normalization [4].

Convergence of the iterative likelihood optimisation

`L-BFGS-B` uses three termination criteria:

1. `(f_k - f_{k+1}) / max(|f_k|, |f_{k+1}|, 1) <= factr * epsmch` where `epsmch` is the machine precision.
2. `|gradient| < pgtol`
3. `iterations > maxit`

These are set by the elements `factr`, `pgtol` and `maxit` of `optimpar`. The remaining elements are

`trace` An integer between 0 and 6, indicating the verbosity level of `L-BFGS-B`, higher values create more output.

`cvg.niter` The number of iterations to be used in the least trimmed sum of squares regression.

`cvg.eps` Numeric. A convergence threshold for the least trimmed sum of squares regression.

Author(s)

Wolfgang Huber

References

[1] Variance stabilization applied to microarray data calibration and to the quantification of differ-ential expression, Wolfgang Huber, Anja von Heydebreck, Holger Sueltmann, Annemarie Poustka, Martin Vingron; Bioinformatics (2002) 18 Suppl.1 S96-S104.

[2] Parameter estimation for the calibration and variance stabilization of microarray data, Wolfgang Huber, Anja von Heydebreck, Holger Sueltmann, Annemarie Poustka, and Martin Vingron; Statisti-cal Applications in Genetics and Molecular Biology (2003) Vol. 2 No. 1, Article 3. http://www.bepress.com/sagmb/vol2/i

[3] L-BFGS-B: Fortran Subroutines for Large-Scale Bound Constrained Optimization, C. Zhu, R.H. Byrd, P. Lu and J. Nocedal, Technical Report, Northwestern University (1996).

[4] Package vignette: Likelihood Calculations for vsn

See Also

justvsn, predict

Examples

```
data("kidney")

fit = vsn2(kidney)                  ## fit
nkid = predict(fit, newdata=kidney)  ## apply fit

plot(exprs(nkid), pch=".")
abline(a=0, b=1, col="red")
```

---

vsn2trsf   *Apply the vsn transformation to data*

Description

Apply the vsn transformation to data.

Usage

```
## S4 method for signature 'vsn':
predict(object, newdata, strata=object@strata, log2scale=TRUE, useDataInFit=FALS
```

Arguments

object	An object of class vsn that contains transformation parameters and strata infor-mation, typically this is the result of a previous call to vsn2.
newdata	Object of class ExpressionSet, NChannelSet, AffyBatch (from the affy package), RGList (from the limma package), matrix or numeric, with the data to which the fit is to be applied to.
strata	Optional, a factor or integer that aligns with the rows of newdata; see the strata argument of vsn2.

log2scale   If TRUE, the data are returned on the glog scale to base 2, and an overall offset
             c is added (see *Value* section of the vsn2 manual page). If FALSE, the data are
             returned on the glog scale to base e, and no offset is added.

useDataInFit If TRUE, then no transformation is attempted and the data stored in object
             is transferred appropriately into resulting object, which otherwise preserves the
             class and metadata of newdata. This option exists to increase performance in
             constructs like

```
fit = vsn2(x, ...)
nx = predict(fit, newdata=x)
```

             and is used, for example, in the justvsn function.

Value

An object typically of the same class as newdata. There are two exceptions: if newdata is an
RGList, the return value is an NChannelSet, and if newdata is numeric, the return value is a
matrix with 1 column.

Author(s)

Wolfgang Huber

Examples

```
data("kidney")

## nb: for random subsampling, the 'subsample' argument of vsn
##   provides an easier way to do this
fit = vsn2(kidney[sample(nrow(kidney), 500), ])
tn = predict(fit, newdata=exprs(kidney))
```

---

vsnh   *A function that transforms a matrix of microarray intensities.*

Description

A function that transforms a matrix of microarray intensities. This function works in conjunction
with vsn. vsn and vsnh have been superseded by vsn2 and the predict method for vsn
objects. The functions vsn and vsnh remain in the package for backward compatibility, but for
new projects, please use vsn2 and predict.

Usage

```
vsnh(y, p, strata)
```

Arguments

y   A numeric matrix containing intensity values from an array experiment. It may contain NA values.

p   An array with the transformation parameters. If `strata` is specified, it must be a 3d array, dim(p)[1] must be greater than or equal to the maximum of `strata`, dim(p)[2] must be ncol(y), and dim(p)[3] must be 2. If `strata` is missing, then the first dimension may be omitted. NA values are not allowed. See Details.

strata   Integer vector of length nrow(y). See `vsn` for details.

Details

The transformation is:

vsnh(y, p, s)[k, i] = asinh(p[s[k], i, 1] + p[s[k], i, 2] * y[k, i]) - log(2*p[s[1], 1, 2])

where k=1:nrow(y) counts over the probes, i=1:ncol(y) counts over the samples, p[s[k], i, 1] is the calibration offset for stratum s[k] in sample i, p[s[k], i, 2] is the calibration factor for stratum s[k] in sample i, and s[k] is the stratum of the the k-th probe.

The constant offset - log(2*p[s[1], 1, 2]) is there to make sure that for large y, vsnh(y) for the first stratum on the first chip is approximately the same as log(y). This has no effect on the generalized log-ratios (glog-ratios), which are differences between transformed intensities, but some users are more comfortable with the absolute values that are obtained this way, since they are more comparable to the log scale.

Value

A numeric matrix of the same size as y, with the transformed data.

Author(s)

Wolfgang Huber

References

Variance stabilization applied to microarray data calibration and to the quantification of differential expression, Wolfgang Huber, Anja von Heydebreck, Holger Sueltmann, Annemarie Poustka, Martin Vingron; Bioinformatics (2002) 18 Suppl.1 S96-S104.

Parameter estimation for the calibration and variance stabilization of microarray data, Wolfgang Huber, Anja von Heydebreck, Holger Sueltmann, Annemarie Poustka, and Martin Vingron; Statistical Applications in Genetics and Molecular Biology (2003) Vol. 2 No. 1, Article 3. http://www.bepress.com/sagmb/vol2/iss1

See Also

`vsn`

Examples

```
data(kidney)
y      = exprs(kidney)
p      = array(c(-0.2, -0.1, 0.1, 0.2, 0.0026, 0.0028, 0.0030, 0.0032), dim=c(2,2,2))
strata = sample(1:2, nrow(y), replace=TRUE)
res1   = vsnh(exprs(kidney), p, strata)

res2   = asinh(p[strata,,1] + p[strata,,2] * y) - log(2*p[strata,1,2])
```

```
stopifnot(max(abs(res1 - res2)) < 1e-10)
```

---

logLik-methods *Calculate the log likelihood and its gradient for the vsn model*

Description

logLik calculates the log likelihood and its gradient for the vsn model. plotVsnLogLik makes a false color plot for a 2D section of the likelihood landscape.

Usage

```
## S4 method for signature 'vsnInput':
logLik(object, p, mu = numeric(0), sigsq=as.numeric(NA), calib="affine")

plotVsnLogLik(object,
              p,
              whichp = 1:2,
              expand = 1,
              ngrid = 31L,
              fun = logLik,
              main = "log likelihood",
              ...)
```

Arguments

object	A vsnInput object.
p	For plotVsnLogLik, a vector or a 3D array with the point in parameter space around which to plot the likelihood. For logLik, a matrix whose columns are the set of parameters at which the likelihoods are to be evaluated.
mu	Numeric vector of length 0 or nrow(object). If the length is 0, there is no reference and sigsq must be NA (the default value). See vsn2.
sigsq	Numeric scalar.
calib	as in vsn2.
whichp	Numeric vector of length 2, with the indices of those two parameters in p along which the section is to be taken.
expand	Numeric vector of length 1 or 2 with expansion factors for the plot range. The range is auto-calculated using a heuristic, but manual adjustment can be useful; see example.
ngrid	Integer scalar, the grid size.
fun	Function to use for log-likelihood calculation. This parameter is exposed only for testing purposes.
main	This parameter is passed on levelplot.
...	Arguments that get passed on to fun, use this for mu, sigsq, calib.

Details

logLik is an R interface to the likelihood computations in vsn (which are done in C).

Value

For `logLik`, a numeric matrix of size `nrow(p)+1` by `ncol(p)`. Its columns correspond to the columns of p. Its first row are the likelihood values, its rows `2...nrow(p)+1` contain the gradients. If `mu` and `sigsq` are specified, the ordinary negative log likelihood is calculated using these parameters as given. If they are not specified, the profile negative log likelihood is calculated.

For `plotVsnLogLik`, a dataframe with the 2D grid coordinates and log likelihood values.

Author(s)

Wolfgang Huber

See Also

[vsn2](#)

Examples

```
data("kidney")

v = new("vsnInput", x=exprs(kidney),
  pstart=array(as.numeric(NA), dim=c(1, ncol(kidney), 2)))

fit = vsn2(kidney)
print(coef(fit))

p = sapply(seq(-1, 1, length=31), function(f) coef(fit)+c(0,0,f,0))

ll = logLik(v, p)

plot(p[3, ], ll[1, ], type="l", xlab=expression(b[1]), ylab=expression(-log(L)))
abline(v=coef(fit)[3], col="red")

plotVsnLogLik(v, coef(fit), whichp=c(1,3), expand=0.2)
```

`vsn-package`   *vsn*

Description

vsn

Details

The main function of the package is [vsn2](#). Interesting for its applications are also `predict` and the wrapper function [justvsn](#).

[vsn2](#) can be applied to objects of class [ExpressionSet](#), [NChannelSet](#), [AffyBatch](#) (from the `affy` package) and [RGList](#) (from the `limma` package), `matrix` and `vector`. It returns an object of class [vsn](#), which contains the results of fitting the `vsn` model to the data.

The most common use case is that you will want to construct a new data object with the vsn-normalized data whose class is the same as that of the input data and which preserves the metadata. This can be achieved by

```
        fit = vsn2(x, ...)
        nx = predict(fit, newdata=x)
```

To simplify this, there exists also a simple wrapper justvsn.

Author(s)

Wolfgang Huber

---

vsnPlotPar	*Plot trajectories of calibration and transformation parameters for a vsn fit*

Description

Plot trajectories of calibration and transformation parameters for a vsn fit

Usage

```
vsnPlotPar(x, what, xlab="iter", ylab=what, ...)
```

Arguments

x	An object of class ExpressionSet-class which has been created by the function vsn.
what	Character, should either be "factors" or "offsets".
xlab	Character, label for the x-axis.
ylab	Character, label for the y-axis.
...	Further arguments that get passed to plot.default.

Details

The plot that is created by this function may help in assessing whether the parameter estimation in vsn was sufficiently converged.

Value

The function is called for its side effect, creating a plot on the active graphics device.

Author(s)

Wolfgang Huber

See Also

vsn

Examples

```
   ## see example for vsn
```

| vsn.old | *Variance stabilization and calibration for microarray data.* |

Description

Robust estimation of variance-stabilizing and calibrating transformations for microarray data. This function has been superseded by vsn2. The function vsn remains in the package for backward compatibility, but for new projects, please use vsn2.

Usage

```
vsn(intensities,
    lts.quantile = 0.5,
    verbose      = interactive(),
    niter        = 10,
    cvg.check    = NULL,
    describe.preprocessing = TRUE,
    subsample,
    pstart,
    strata)
```

Arguments

intensities   An object that contains intensity values from a microarray experiment. The intensities are assumed to be the raw scanner data, summarized over the spots by an image analysis program, and possibly "background subtracted". The intensities must not be logarithmically or otherwise transformed, and not thresholded or "floored". NAs are not accepted. See details.

lts.quantile   Numeric. The quantile that is used for the resistant least trimmed sum of squares regression. Allowed values are between 0.5 and 1. A value of 1 corresponds to ordinary least sum of squares regression.

verbose   Logical. If TRUE, some messages are printed.

niter   Integer. The number of iterations to be used in the least trimmed sum of squares regression.

cvg.check   List. If non-NULL, this allows finer control of the iterative least trimmed sum of squares regression. See details.

pstart   Array. If not missing, user can specify start values for the iterative parameter estimation algorithm. See vsnh for details.

describe.preprocessing
                Logical. If TRUE, calibration and transformation parameters, plus some other information are stored in the preprocessing slot of the returned object. See details.

subsample   Integer. If specified, the model parameters are estimated from a subsample of the data only, the transformation is then applied to all data. This can be useful for performance reasons.

strata   Integer vector. Its length must be the same as nrow(intensities). This parameter allows for the calibration and error model parameters to be stratified within each array, e.g to take into account probe sequence properties, print-tip or plate

effects. If `strata` is not specified, one pair of parameters is fitted for every sample (i.e. for every column of `intensities`). If `strata` is specified, a pair of parameters is fitted for every stratum within every sample. The strata are coded for by the different integer values. The integer vector `strata` can be obtained from a factor `fac` through `as.integer(fac)`, from a character vector `str` through `as.integer(factor(fac))`.

Details

Overview: The function calibrates for sample-to-sample variations through shifting and scaling, and transforms the intensities to a scale where the variance is approximately independent of the mean intensity. The variance stabilizing transformation is equivalent to the natural logarithm in the high-intensity range, and to a linear transformation in the low-intensity range. In an intermediate range, the *arsinh* function interpolates smoothly between the two. For details on the transformation, please see the help page for `vsnh`. The parameters are estimated through a robust variant of maximum likelihood. This assumes that for the majority of genes the expression levels are not much different across the samples, i.e., that only a minority of genes (less than a fraction `1-lts.quantile`) is differentially expressed.

Even if most genes on an array are differentially expressed, it may still be possible to use the estimator: if a set of non-differentially expressed genes is known, e.g. because they are external controls or reliable 'house-keeping genes', the transformation parameters can be fitted with `vsn` from the data of these genes, then the transformation can be applied to all data with `vsnh`.

Format: The format of the matrix of intensities is as follows: for the **two-color printed array technology**, each row corresponds to one spot, and the columns to the different arrays and wavelengths (usually red and green, but could be any number). For example, if there are 10 arrays, the matrix would have 20 columns, columns 1...10 containing the green intensities, and 11...20 the red ones. In fact, the ordering of the columns does not matter to `vsn`, but it is your responsibility to keep track of it for subsequent analyses. For **one-color arrays**, each row corresponds to a probe, and each column to an array.

Performance: This function is slow. That is due to the nested iteration loops of the numerical optimization of the likelihood function and the heuristic that identifies the non-outlying data points in the least trimmed squares regression. For large arrays with many tens of thousands of probes, you may want to consider random subsetting: that is, only use a subset of the e.g. 10-20,000 rows of the data matrix `intensities` to fit the parameters, then apply the transformation to all the data, using `vsnh`. An example for this can be seen in the function `normalize.AffyBatch.vsn`, whose code you can inspect by typing `normalize.AffyBatch.vsn` on the R command line.

Iteration control: By default, if `cvg.check` is `NULL`, the function will run the fixed number `niter` of iterations in the least trimmed sum of squares regression. More fine-grained control can be obtained by passing a list with elements `eps` and `n`. If the maximum change between transformed data values is smaller than `eps` for `n` subsequent iterations, then the iteration terminates.

Estimated transformation parameters: If `describe.preprocessing` is `TRUE`, the transformation parameters are returned in the `preprocessing` slot of the `experimentData` slot of the resulting `ExpressionSet` object, in the form of a `list` with three elements

- `vsnParams`: the parameter array (see `vsnh` for details)

- `vsnParamsIter`: an array with dimensions `c(dim(vsnParams, niter))` that contains the parameter trajectory during the iterative fit process (see also `vsnPlotPar`).

- `vsnTrimSelection`: a logical vector that for each row of the intensities matrix reports whether it was below (TRUE) or above (FALSE) the trimming threshold.

If `intensities` has class `ExpressionSet`, and its `experimentData` slot has class `MIAME`, then this list is appended to any existing entries in the `preprocessing` slot. Otherwise, the `experimentData` object and its `preprocessing` slot are created.

Value

An object of class `ExpressionSet`. Differences between the columns of the transformed intensities are "generalized log-ratios", which are shrinkage estimators of the natural logarithm of the fold change. For the transformation parameters, please see the Details.

Author(s)

Wolfgang Huber

References

Variance stabilization applied to microarray data calibration and to the quantification of differential expression, Wolfgang Huber, Anja von Heydebreck, Holger Sueltmann, Annemarie Poustka, Martin Vingron; Bioinformatics (2002) 18 Suppl.1 S96-S104.

Parameter estimation for the calibration and variance stabilization of microarray data, Wolfgang Huber, Anja von Heydebreck, Holger Sueltmann, Annemarie Poustka, and Martin Vingron; Statistical Applications in Genetics and Molecular Biology (2003) Vol. 2 No. 1, Article 3. http://www.bepress.com/sagmb/vol2/iss1

See Also

`vsnh`, `vsnPlotPar`, `ExpressionSet-class`, `MIAME-class`, `normalize.AffyBatch.vsn`

Examples

```
data(kidney)
log.na = function(x) log(ifelse(x>0, x, NA))

plot(log.na(exprs(kidney)), pch=".", main="log-log")

vsnkid = vsn(kidney)   ## transform and calibrate
plot(exprs(vsnkid), pch=".", main="h-h")
meanSdPlot(vsnkid)

## this should always hold true
params = preproc(description(vsnkid))$vsnParams
stopifnot(all(vsnh(exprs(kidney), params) == exprs(vsnkid)))
```

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