Package ‘affyPLM’

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Title Methods for fitting probe-level models
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Suggests affydata, MASS
LinkingTo preprocessCore
Description A package that extends and improves the functionality of the base affy package. Routines that make heavy use of compiled code for speed. Central focus is on implementation of methods for fitting probe-level models and tools using these models. PLM based quality assessment tools.
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R topics documented:

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bg.correct.LESN

Description
This function background corrects PM probe data using LESN - Low End Signal is Noise concepts.

Usage
bg.correct.LESN(object, method=2, baseline=0.25, theta=4)

Arguments
- object: an AffyBatch
- method: an integer code specifying which method to use
- baseline: A baseline value to use
- theta: A parameter used in the background correction process

Details
This method will be more formally documented at a later date.
The basic concept is to consider that the lowest end of intensites is most likely just noise (and should be heavily corrected) and the highest end signals are most likely signal and should have little adjustment. Low end signals are made much smaller while high end signals get less adjustment relative adjustment.

Value
An AffyBatch

Author(s)
Ben Bolstad <bmb@bmbolstad.com>

References

Examples
if (require(affydata)) {
  data(Dilution)
  Dilution.example.bgcorrect <- bg.correct.LESN(Dilution)
}
**Description**

This function converts an AffyBatch into a PLMset by fitting a specified robust linear model to the probe level data.

**Usage**

```r
fitPLM(object, model = PM ~ -1 + probes + samples,
       variable.type = c(default = "factor"),
       constraint.type = c(default = "contr.treatment"),
       subset = NULL,
       background = TRUE, normalize = TRUE, background.method = "RMA.2",
       normalize.method = "quantile", background.param = list(),
       normalize.param = list(), output.param = verify.output.param(),
       model.param = verify.model.param(object, model),
       verbosity.level = 0)
```

**Arguments**

- **object**: an AffyBatch
- **model**: A formula describing the model to fit. This is slightly different from the standard method of specifying formulae in R. Read the description below
- **variable.type**: a way to specify whether variables in the model are factors or standard variables
- **constraint.type**: should factor variables sum to zero or have first variable set to zero (endpoint constraint)
- **subset**: a vector with the names of probesets to be used. If NULL then all probesets are used.
- **normalize**: logical value. If TRUE normalize data using quantile normalization
- **background**: logical value. If TRUE background correct using RMA background correction
- **background.method**: name of background method to use.
- **normalize.method**: name of normalization method to use.
- **background.param**: A list of parameters for background routines
- **normalize.param**: A list of parameters for normalization routines
- **output.param**: A list of parameters controlling optional output from the routine.
- **model.param**: A list of parameters controlling model procedure
- **verbosity.level**: An integer specifying how much to print out. Higher values indicate more verbose. A value of 0 will print nothing
Details

This function fits robust Probe Level linear Models to all the probesets in an \texttt{AffyBatch}. This is carried out on a probeset by probeset basis. The user has quite a lot of control over which model is used and what outputs are stored. For more details please read the vignette.

Value

An \texttt{PLMset}

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

References


See Also

\texttt{expresso}, \texttt{rma}, \texttt{threestep}

Examples

```r
if (require(affydata)) {
  data(Dilution)
  Pset <- fitPLM(Dilution, model=PM ~ -1 + probes + samples)
  se(Pset)[1:5,]

  image(Pset)
  NUSE(Pset)

  # now lets try a wider class of models
  ## Not run: Pset <- fitPLM(Dilution, model=PM ~ -1 + probes + liver,
  ## normalize=FALSE, background=FALSE)
  ## End(Not run)
  ## Not run: coefs(Pset)[1:10,]
  ## Not run: Pset <- fitPLM(Dilution, model=PM ~ -1 + probes + liver +
  ## scanner, normalize=FALSE, background=FALSE)
  ## End(Not run)
  ## Not run: coefs(Pset)[1:10,]

  # try liver as a covariate
  logliver <- log2(c(20,20,10,10))
  ## Not run: Pset <- fitPLM(Dilution, model=PM~-1+probes+logliver+scanner,
  ## normalize=FALSE, background=FALSE, variable.type=c(logliver="covariate"))
  ## End(Not run)
  coefs(Pset)[1:10,]

  # try a different se.type
  ## Not run: Pset <- fitPLM(Dilution, model=PM~-1+probes+scanner,
  ## normalize=FALSE, background=FALSE, model.param=list(se.type=2))
  ## End(Not run)
  se(Pset)[1:10,]
}
```
MAplot

Relative M vs. A plots

Description

Create boxplots of M or M vs A plots. Where M is determined relative to a specified chip or to a pseudo-median reference chip.

Arguments

... Additional parameters for the routine
A A vector to plot along the horizontal axis
M A vector to plot along vertical axis
subset A set of indices to use when drawing the loess curve
show.statistics If true some summary statistics of the M values are drawn
span span to be used for loess fit.
family.loess "gaussian" or "symmetric" as in loess.
cex Size of text when writing summary statistics on plot

See Also

mva.pairs

normalize.ExpressionSet

Normalization applied to ExpressionSets

Description

Allows the user to apply normalization routines to ExpressionSets.

Usage

normalize.ExpressionSet.quantiles(eset, transfn=c("none","log","antilog"))
normalize.ExpressionSet.loess(eset, transfn=c("none","log","antilog"),...)
normalize.ExpressionSet.contrasts(eset, span = 2/3,
choose.subset=TRUE, subset.size=5000, verbose=TRUE, family=\"symmetric\",
transfn=c("none","log","antilog"))
normalize.ExpressionSet.qspline(eset, transfn=c("none","log","antilog"),...)
normalize.ExpressionSet.invariantset(eset,prd.td=c(0.003, 0.007),
verbose=FALSE, transfn=c("none","log","antilog"),
baseline.type=c("mean","median","pseudo-mean","pseudo-median"))
normalize.ExpressionSet.scaling(eset, trim=0.02, baseline=-1,
transfn=c("none","log","antilog"))
normalize.ExpressionSet

Arguments

eset  An ExpressionSet parameter to be passed to the function loess.
span  use a subset of values to establish the normalization relationship
choose.subset number to use for subset
subset.size verbosity flag
verbose parameter to be passed to the function loess.
family greedy parameter (details in the bibliographic reference)
prd.td How much to trim from the top and bottom before computing the mean when using the scaling normalization
trim baseline Index of array to use as baseline, negative values (-1,-2,-3,-4) control different baseline selection methods
baseline transfn Transform the ExpressionSet before normalizing. Useful when dealing with expression values that are log-scale
baseline.type A method of selecting the baseline array
... Additional parameters that may be passed to the normalization routine

Details

This function carries out normalization of expression values. In general you should either normalize at the probe level or at the expression value level, not both.

Typing normalize.ExpressionSet.methods should give you a list of methods that you may use. note that you can also use the normalize function on ExpressionSets. Use method to select the normalization method.

Value

A normalized ExpressionSet.

Author(s)

Ben Bolstad, <bmb@bmbolstad.com>

References


Examples

if (require(affydata)) {
  data(Dilution)
  eset <- rma(Dilution, normalize=FALSE, background=FALSE)
  normalize(eset)
}
normalize.quantiles.probeset

Quantile Normalization applied to probesets

Description

Using a normalization based upon quantiles, this function normalizes a matrix of probe level intensities.

Usage

normalize.AffyBatch.quantiles.probeset(abatch, type=c("separate", "pmonly", "mmonly", "together"), use.median=FALSE, use.log=TRUE)

Arguments

abatch An AffyBatch
type how should MM and PM values be handled
use.median use median rather than mean
use.log take logarithms, then normalize

Details

This function applies the quantile method in a probeset specific manner.

In particular a probeset summary is normalized using the quantile method and then the probes adjusted accordingly.

Value

A normalized AffyBatch.

Author(s)

Ben Bolstad, <bmb@bmbolstad.com>

References


See Also

normalize.quantiles
normalize.scaling  Scaling normalization

Description

Allows the user to apply scaling normalization.

Usage

normalize.scaling(X, trim=0.02, baseline=-1, log.scalefactors=FALSE)
normalize.AffyBatch.scaling(abatch,
   type=c("together","pmonly","mmonly","separate"),
   trim=0.02, baseline=-1, log.scalefactors=FALSE)

Arguments

X A matrix. The columns of which are to be normalized.
abatch An AffyBatch
type A parameter controlling how normalization is applied to the Affybatch.
trim How much to trim from the top and bottom before computing the mean when using the scaling normalization.
baseline Index of array to use as baseline, negative values (-1,-2,-3,-4) control different baseline selection methods.
log.scalefactors Compute the scale factors based on log2 transformed data.

Details

These function carries out scaling normalization of expression values.

Value

A normalized ExpressionSet.

Author(s)

Ben Bolstad, <bmb@bmbolstad.com>

Examples

if (require(affydata)) {
   data(Dilution)
   normalize.AffyBatch.scaling(Dilution)
}
PLMset-class

Class PLMset

Description

This is a class representation for Probe level Linear Models fitted to Affymetrix GeneChip probe level data.

Objects from the Class

Objects can be created using the function `fitPLM`

Slots

- `probe.coefs`: Object of class "matrix". Contains model coefficients related to probe effects.
- `seprobe.coefs`: Object of class "matrix". Contains standard error estimates for the probe coefficients.
- `chip.coefs`: Object of class "matrix". Contains model coefficients related to chip (or chip level) effects for each fit.
- `sechip.coefs`: Object of class "matrix". Contains standard error estimates for the chip coefficients.
- `const.coefs`: Object of class "matrix". Contains model coefficients related to intercept effects for each fit.
- `secconst.coefs`: Object of class "matrix". Contains standard error estimates for the intercept estimates
- `model.description`: Object of class "character". This string describes the probe level model fitted.
- `weights`: List of objects of class "matrix". Contains probe weights for each fit. The matrix has columns for chips and rows are probes.
- `phenoData`: Object of class "phenoData". This is an instance of class `phenoData` containing the patient (or case) level data. The columns of the pData slot of this entity represent variables and the rows represent patients or cases.
- `annotation`: A character string identifying the annotation that may be used for the ExpressionSet instance.
- `experimentData`: Object of class "MIAME". For compatibility with previous version of this class description can also be a "character". The class `characterOrMIAME` has been defined just for this.
- `cdfName`: A character string giving the name of the cdfFile.
- `nrow`: Object of class "numeric". Number of rows in chip.
- `ncol`: Object of class "numeric". Number of cols in chip.
- `narrays`: Object of class "numeric". Number of arrays used in model fit.
- `normVec`: Object of class "matrix". For storing normalization vector(s). Not currently used
- `varcov`: Object of class "list". A list of variance/covariance matrices.
- `residualSE`: Object of class "matrix". Contains residual standard error and df.
- `residuals`: List of objects of class "matrix". Contains residuals from model fit (if stored).
- `model.call`: Object of class "call"
Methods

weights<- signature(object = "PLMset"): replaces the weights.
weights signature(object = "PLMset"): extracts the model fit weights.
coefs<- signature(object = "PLMset"): replaces the chip coefs.
coefs signature(object = "PLMset"): extracts the chip coefs.
se signature(object = "PLMset"): extracts the standard error estimates of the chip coefs.
se<- signature(object = "PLMset"): replaces the standard error estimates of the chip coefs.
coefs.probe signature(object = "PLMset"): extracts the probe coefs.
se.probe signature(object = "PLMset"): extracts the standard error estimates of the probe coefs.
coefs.const signature(object = "PLMset"): extracts the intercept coefs.
se.const signature(object = "PLMset"): extracts the standard error estimates of the intercept coefs.
getCdfInfo signature(object = "PLMset"): retrieve the environment that defines the location of probes by probe set.
image signature(x = "PLMset"): creates an image of the robust linear model fit weights for each sample.
indexProbes signature(object = "PLMset", which = "character"): returns a list with locations of the probes in each probe set. The list names defines the probe set names. which can be "pm", "mm", or "both". If "both" then perfect match locations are given followed by mismatch locations.
Mbox signature(object = "PLMset"): gives a boxplot of M's for each chip. The M's are computed relative to a "median" chip.
normvec signature(x = "PLMset"): will return the normalization vector (if it has been stored).
residSE signature(x = "PLMset"): will return the residual SE (if it has been stored).
boxplot signature(x = "PLMset"): Boxplot of Normalized Unscaled Standard Errors (NUSE).
NUSE signature(x = "PLMset"): Boxplot of Normalized Unscaled Standard Errors (NUSE) or NUSE values.
RLE signature(x = "PLMset"): Relative Log Expression boxplot or values.

Note

This class is better described in the vignette.

Author(s)

B. M. Bolstad <bmb@bmbolstad.com>

References

PLMset2exprSet

Convert a PLMset to an ExpressionSet

Description

This function converts a PLMset to an ExpressionSet. This is often useful since many Bioconductor functions operate on ExpressionSet objects.

Usage

PLMset2exprSet(pset)
pset2eset(pset)

Arguments

pset The PLMset to convert to ExpressionSet.

Details

These functions convert PLMset objects to ExpressionSet objects. This is often useful since many Bioconductor functions operate on ExpressionSet objects. Note that the function pset2eset is a wrapper for PLMset2exprSet.

Value

returns a ExpressionSet

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

See Also

ExpressionSet

Examples

if (require(affydata)) {
  data(Dilution)
  Pset <- fitPLM(Dilution)
  eset <- pset2eset(Pset)
}
Description
This function pre-processes an AffyBatch.

Usage
preprocess(object, subset=NULL, normalize=TRUE, background=TRUE, 
background.method="RMA.2", normalize.method="quantile", 
background.param=list(), normalize.param=list(), 
verbosity.level=0)

Arguments
object      an AffyBatch
subset     a vector with the names of probesets to be used. If NULL then all probesets are 
           used.
normalize   logical value. If TRUE normalize data using quantile normalization
background  logical value. If TRUE background correct using RMA background correction
background.method
             name of background method to use.
normalize.method
             name of normalization method to use.
background.param
             list of parameters for background correction methods
normalize.param
             list of parameters for normalization methods
verbosity.level
             An integer specifying how much to print out. Higher values indicate more ver-
            bose. A value of 0 will print nothing

Details
This function carries out background correction and normalization pre-processing steps. It does not 
summarize to produce gene expression measures. All the same pre-processing methods supplied by 
threestep are supported by this function.

Value
An AffyBatch

Author(s)
Ben Bolstad <bmb@bmbolstad.com>

References
Bolstad, BM (2004) Low Level Analysis of High-density Oligonucleotide Array Data: Background, 
pseudo.coloring

See Also
epresso, rma

Examples

if (require(affydata)) {
  data(Dilution)

  # should be equivalent to the bg and norm of rma()
  abatch.preprocessed <- preprocess(Dilution)
}

pseudo.coloring  Coloring pseudo chip images

Description

These are routines used for coloring pseudo chip images.

Usage

  pseudoPalette(low = "white", high = c("green", "red"), mid = NULL, k = 50)
  pseudoColorBar(x, horizontal = TRUE, col = heat.colors(50), scale = 1:length(x), k = 11, log.ticks = FALSE, ...)

Arguments

  low  color at low end of scale
  high color at high end of scale
  mid  color at exact middle of scale
  k    number of colors to have
  x    A data series
  horizontal  If TRUE then color bar is to be draw horizontally
  col    colors for color bar
  scale  tickmarks for x if x is not numeric
  log.ticks  use a log type transformation to assign the colors
  ...    additional parameters to plotting routine

Details

Adapted from similar tools in maPlots package.

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

See Also

  AffyBatch, read.affybatch
ReadRMAExpress

Description
Read RMAExpress computed binary output files into a matrix or ExpressionSet

Usage
ReadRMAExpress(filename, return.value=c("ExpressionSet","matrix"))

Arguments
filename
The name of the file containing RMAExpress output to be read in
return.value
should a matrix or an ExpressionSet be returned

Value
returns an ExpressionSet

Author(s)
Ben Bolstad <bmb@bmbolstad.com>

References
http://rmaexpress.bmbolstad.com

rmaPLM

Description
Fit a RMA to Affymetrix Genechip Data as a PLMset

Usage
rmaPLM(object, subset=NULL, normalize=TRUE, background=TRUE,
background.method="RMA.2", normalize.method="quantile",
background.param=list(), normalize.param=list(), output.param=list(),
model.param=list(), verbosity.level=0)
Arguments

object an AffyBatch
subset a vector with the names of probesets to be used. If NULL then all probesets are used.
normalize logical value. If TRUE normalize data using quantile normalization
background logical value. If TRUE background correct using RMA background correction
background.method name of background method to use.
normalize.method name of normalization method to use.
background.param A list of parameters for background routines
normalize.param A list of parameters for normalization routines
output.param A list of parameters controlling optional output from the routine.
model.param A list of parameters controlling model procedure
verbosity.level An integer specifying how much to print out. Higher values indicate more verbose. A value of 0 will print nothing

Details

This function fits the RMA as a Probe Level Linear models to all the probesets in an AffyBatch.

Value

An PLMset

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

References


See Also

expresso, rma, threestep, fitPLM, threestepPLM
threestep

Examples

```r
if (require(affydata)) {
  # A larger example testing weight image function
data(Dilution)
  ## Not run: Pset <- rmaPLM(Dilution, output.param=list(weights=TRUE))
  ## Not run: image(Pset)
}
```

---

threestep  
Three Step expression measures

Description

This function converts an AffyBatch into an ExpressionSet using a three step expression measure.

Usage

```r
threestep(object, subset=NULL, normalize=TRUE, background=TRUE, 
  background.method="RMA.2", normalize.method="quantile", 
  summary.method="median.polish", background.param=list(), 
  normalize.param=list(), summary.param=list(), verbosity.level=0)
```

Arguments

- **object**: an AffyBatch.
- **subset**: a vector with the names of probesets to be used. If NULL, then all probesets are used.
- **normalize**: logical value. If TRUE normalize data using quantile normalization.
- **background**: logical value. If TRUE background correct using RMA background correction.
- **background.method**: name of background method to use.
- **normalize.method**: name of normalization method to use.
- **summary.method**: name of summary method to use.
- **background.param**: list of parameters for background correction methods.
- **normalize.param**: list of parameters for normalization methods.
- **summary.param**: list of parameters for summary methods.
- **verbosity.level**: An integer specifying how much to print out. Higher values indicate more verbose. A value of 0 will print nothing.

Details

This function computes the expression measure using threestep methods. Greater details can be found in a vignette.
threestep

Value

An ExpressionSet

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

References


See Also

expresso, rma

Examples

if (require(affydata)) {
  data(Dilution)

  # should be equivalent to rma()
  eset <- threestep(Dilution)

  # Using Tukey Biweight summarization
  eset <- threestep(Dilution, summary.method="tukey.biweight")

  # Using Average Log2 summarization
  eset <- threestep(Dilution, summary.method="average.log")

  # Using IdealMismatch background and Tukey Biweight and no normalization.
  eset <- threestep(Dilution, normalize=FALSE, background.method="IdealMM", summary.method="tukey.biweight")

  # Using average.log summarization and no background or normalization.
  eset <- threestep(Dilution, background=FALSE, normalize=FALSE, background.method="IdealMM", summary.method="tukey.biweight")

  # Use threestep methodology with the rlm model fit
  eset <- threestep(Dilution, summary.method="rlm")

  # Use threestep methodology with the log of the average
  eset <- threestep(Dilution, summary.method="log.average")

  # Use threestep methodology with log 2nd largest method
  eset <- threestep(Dilution, summary.method="log.2nd.largest")

  eset <- threestep(Dilution, background.method="LESN2")
}
threestepPLM

Three Step expression measures returned as a PLMset

**Description**

This function converts an `AffyBatch` into an `PLMset` using a three step expression measure.

**Usage**

threestepPLM(object, subset=NULL, normalize=TRUE, background=TRUE, background.method="RMA.2", normalize.method="quantile", summary.method="median.polish", background.param = list(), normalize.param=list(), output.param=list(), model.param=list(), verbosity.level=0)

**Arguments**

- **object**: an `AffyBatch`
- **subset**: a vector with the names of probesets to be used. If NULL then all probesets are used.
- **normalize**: logical value. If TRUE normalize data using quantile normalization
- **background**: logical value. If TRUE background correct using RMA background correction
- **background.method**: name of background method to use.
- **normalize.method**: name of normalization method to use.
- **summary.method**: name of summary method to use.
- **background.param**: list of parameters for background correction methods
- **normalize.param**: list of parameters for normalization methods
- **output.param**: list of parameters for output methods
- **model.param**: list of parameters for model methods
- **verbosity.level**: An integer specifying how much to print out. Higher values indicate more verbose. A value of 0 will print nothing

**Details**

This function computes the expression measure using threestep methods. It returns a `PLMset`. The most important difference is that the PLMset allows you to access the residuals which the `threestep` function does not do.

**Value**

An `PLMset`
threestepPLM

Author(s)
Ben Bolstad <bmb@bmbolstad.com>

References

See Also
expresso, rma, threestep, rmaPLM, fitPLM

Examples
if (require(affydata)) {
  data(Dilution)

  # should be equivalent to rma()
  ## Not run: eset <- threestepPLM(Dilution)
}
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