

Package ‘affyPLM’

October 8, 2015

Version 1.44.0

Title Methods for fitting probe-level models

Author Ben Bolstad <bmb@bmbolstad.com>

Maintainer Ben Bolstad <bmb@bmbolstad.com>

Depends R (>= 2.6.0), BiocGenerics (>= 0.3.2), affy (>= 1.11.0),
Biobase (>= 2.17.8), gcrma, stats, preprocessCore (>= 1.5.1)

Imports BiocGenerics, zlibbioc, graphics, grDevices, methods

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.

License GPL (>= 2)

URL <http://bmbolstad.com>

biocViews Microarray, OneChannel, Preprocessing, QualityControl

LazyLoad yes

NeedsCompilation yes

R topics documented:

bg.correct.LESN	2
fitPLM	3
MAplot	5
normalize.ExpressionSet	6
normalize.quantiles.probeset	7
normalize.scaling	8
PLMset-class	9
PLMset2exprSet	11
preprocess	12

pseudo.coloring	14
ReadRMAExpress	15
rmaPLM	15
threestep	17
threestepPLM	19

Index	21
--------------	-----------

bg.correct.LESN	<i>LESN - Low End Signal is Noise Background corrections</i>
-----------------	--

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 intensities 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

Bolstad, BM (2004) *Low Level Analysis of High-density Oligonucleotide Array Data: Background, Normalization and Summarization*. PhD Dissertation. University of California, Berkeley.

Examples

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

fitPLM

*Fit a Probe Level Model to Affymetrix Genechip Data.***Description**

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

Usage

```
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 ver-
 bose. A value of 0 will print nothing

Details

This function fits robust Probe Level linear Models to all the probesets in an [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 [PLMset](#)

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

References

Bolstad, BM (2004) *Low Level Analysis of High-density Oligonucleotide Array Data: Background, Normalization and Summarization*. PhD Dissertation. University of California, Berkeley.

See Also

[expresso](#), [rma](#), [threestep](#)

Examples

```
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)
  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,m odel.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"))
```

Arguments

eset	An ExpressionSet
span	parameter to be passed to the function loess .
choose.subset	use a subset of values to establish the normalization relationship
subset.size	number to use for subset
verbose	verbosity flag
family	parameter to be passed to the function loess .
prd.td	cutoff parameter (details in the bibliographic reference)
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
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

Bolstad, BM (2004) *Low Level Analysis of High-density Oligonucleotide Array Data: Background, Normalization and Summarization*. PhD Dissertation. University of California, Berkeley.

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.
```

Arguments

abatch	An <code>AffyBatch</code>
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

Bolstad, B (2001) *Probe Level Quantile Normalization of High Density Oligonucleotide Array Data*. Unpublished manuscript <http://oz.berkeley.edu/~bolstad/stuff/qnorm.pdf>

Bolstad, B. M., Irizarry R. A., Astrand, M, and Speed, T. P. (2003) *A Comparison of Normalization Methods for High Density Oligonucleotide Array Data Based on Bias and Variance*. *Bioinformatics* 19(2), pp 185-193. <http://www.stat.berkeley.edu/~bolstad/normalize/normalize.html>

See Also

[normalize.quantiles](#)

normalize.scaling	<i>Scaling normalization</i>
-------------------	------------------------------

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", "ponly", "monly", "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.

`se.probe.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.

`se.chip.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.

se.const.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.
- RLEI** 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

Bolstad, BM (2004) *Low Level Analysis of High-density Oligonucleotide Array Data: Background, Normalization and Summarization*. PhD Dissertation. University of California, Berkeley.

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)
}
```

preprocess

Background correct and Normalize

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 verbose. 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, Normalization and Summarization*. PhD Dissertation. University of California, Berkeley.

See Also

[expresso](#), [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=FA
```

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 pacakge.

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

See Also

[AffyBatch](#), [read.affybatch](#)

ReadRMAExpress	<i>Read RMAExpress computed expression values</i>
----------------	---

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	<i>Fit a RMA to Affymetrix Genechip Data as a PLMset</i>
--------	--

Description

This function converts an [AffyBatch](#) into an [PLMset](#) by fitting a multichip model. In particular we concentrate on the RMA model.

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

Bolstad, BM (2004) *Low Level Analysis of High-density Oligonucleotide Array Data: Background, Normalization and Summarization*. PhD Dissertation. University of California,

Irizarry RA, Bolstad BM, Collin F, Cope LM, Hobbs B and Speed TP (2003) *Summaries of Affymetrix GeneChip probe level data* Nucleic Acids Research 31(4):e15

Bolstad, BM, Irizarry RA, Astrand, M, and Speed, TP (2003) *A Comparison of Normalization Methods for High Density Oligonucleotide Array Data Based on Bias and Variance*. Bioinformatics 19(2):185-193

See Also

[expresso](#), [rma](#), [threestep](#), [fitPLM](#), [threestepPLM](#)

Examples

```

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

```

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.

Value

An [ExpressionSet](#)

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

References

Bolstad, BM (2004) *Low Level Analysis of High-density Oligonucleotide Array Data: Background, Normalization and Summarization*. PhD Dissertation. University of California, Berkeley.

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](#)

Author(s)

Ben Bolstad <bmb@bmbolstad.com>

References

Bolstad, BM (2004) *Low Level Analysis of High-density Oligonucleotide Array Data: Background, Normalization and Summarization*. PhD Dissertation. University of California, Berkeley.

See Also

[expresso](#), [rma](#), [threestep](#), [rmaPLM](#), [fitPLM](#)

Examples

```
if (require(affydata)) {  
  data(Dilution)  
  
  # should be equivalent to rma()  
  ## Not run: eset <- threestepPLM(Dilution)  
}
```

Index

- *Topic **classes**
 - PLMset-class, 9
- *Topic **hplot**
 - MAplot, 5
- *Topic **manip**
 - bg.correct.LESN, 2
 - fitPLM, 3
 - normalize.ExpressionSet, 6
 - normalize.quantiles.probeset, 7
 - normalize.scaling, 8
 - PLMset2exprSet, 11
 - preprocess, 12
 - pseudo.coloring, 14
 - ReadRMAExpress, 15
 - rmaPLM, 15
 - threestep, 17
 - threestepPLM, 19
- AffyBatch, 2–4, 7, 8, 12–17, 19
- annotation, PLMset-method (PLMset-class), 9
- bg.correct.LESN, 2
- boxplot, PLMset-method (PLMset-class), 9
- cdfName, PLMset-method (PLMset-class), 9
- coefs (PLMset-class), 9
- coefs, PLMset-method (PLMset-class), 9
- coefs.const (PLMset-class), 9
- coefs.const, PLMset-method (PLMset-class), 9
- coefs.probe (PLMset-class), 9
- coefs.probe, PLMset-method (PLMset-class), 9
- coefs<- (PLMset-class), 9
- coefs<-, PLMset-method (PLMset-class), 9
- description, PLMset-method (PLMset-class), 9
- ExpressionSet, 6, 7, 9, 11, 12, 15, 17, 18
- expresso, 4, 13, 16, 18, 20
- fitPLM, 3, 9, 16, 20
- getCdfInfo, PLMset-method (PLMset-class), 9
- image, PLMset-method (PLMset-class), 9
- indexProbes, PLMset, character-method (PLMset-class), 9
- indexProbesProcessed (PLMset-class), 9
- indexProbesProcessed, PLMset-method (PLMset-class), 9
- loess, 5, 6
- MAplot, 5
- MAplot, PLMset-method (MAplot), 5
- matrix, 15
- Mbox (PLMset-class), 9
- Mbox, PLMset-method (PLMset-class), 9
- model.description (PLMset-class), 9
- model.description, PLMset-method (PLMset-class), 9
- mva.pairs, 5
- normalize.AffyBatch.quantiles.probeset (normalize.quantiles.probeset), 7
- normalize.AffyBatch.scaling (normalize.scaling), 8
- normalize.ExpressionSet, 6
- normalize.quantiles, 8
- normalize.quantiles.probeset, 7
- normalize.scaling, 8
- normvec (PLMset-class), 9
- normvec, PLMset-method (PLMset-class), 9
- NUSE (PLMset-class), 9
- nuse (PLMset-class), 9
- NUSE, PLMset-method (PLMset-class), 9
- nuse, PLMset-method (PLMset-class), 9

- pData, PLMset-method (PLMset-class), 9
- pData<-, PLMset, data.frame-method (PLMset-class), 9
- phenoData, PLMset-method (PLMset-class), 9
- phenoData<-, PLMset, AnnotatedDataFrame-method (PLMset-class), 9
- PLMset, 3, 4, 11, 15, 16, 19, 20
- PLMset (PLMset-class), 9
- PLMset-class, 9
- PLMset2exprSet, 11
- preprocess, 12
- pset2eset (PLMset2exprSet), 11
- pseudo.coloring, 14
- pseudoColorBar (pseudo.coloring), 14
- pseudoPalette (pseudo.coloring), 14

- quantile, 8

- read.affybatch, 14
- ReadRMAExpress, 15
- resid (PLMset-class), 9
- resid, PLMset-method (PLMset-class), 9
- resid<- (PLMset-class), 9
- resid<-, PLMset-method (PLMset-class), 9
- residSE (PLMset-class), 9
- residSE, PLMset-method (PLMset-class), 9
- residuals (PLMset-class), 9
- residuals, PLMset-method (PLMset-class), 9
- residuals<- (PLMset-class), 9
- residuals<-, PLMset-method (PLMset-class), 9
- RLE (PLMset-class), 9
- RLE, PLMset-method (PLMset-class), 9
- rma, 4, 13, 16, 18, 20
- rmaPLM, 15, 20

- sampleNames, PLMset-method (PLMset-class), 9
- sampleNames<- (PLMset-class), 9
- sampleNames<-, PLMset, character-method (PLMset-class), 9
- se (PLMset-class), 9
- se, PLMset-method (PLMset-class), 9
- se.const (PLMset-class), 9
- se.const, PLMset-method (PLMset-class), 9
- se.probe (PLMset-class), 9
- se.probe, PLMset-method (PLMset-class), 9
- se<- (PLMset-class), 9
- se<-, PLMset-method (PLMset-class), 9
- show, PLMset-method (PLMset-class), 9
- summary, PLMset-method (PLMset-class), 9
- threestep, 4, 13, 16, 17, 19, 20
- threestepPLM, 16, 19
- varcov (PLMset-class), 9
- varcov, PLMset-method (PLMset-class), 9

- weights (PLMset-class), 9
- weights, PLMset-method (PLMset-class), 9
- weights<- (PLMset-class), 9
- weights<-, PLMset-method (PLMset-class), 9