Package ‘oligo’

March 28, 2024

Version 1.66.0

Title Preprocessing tools for oligonucleotide arrays

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Depends R (>= 3.2.0), BiocGenerics (>= 0.13.11), oligoClasses (>= 1.29.6), Biobase (>= 2.27.3), Biostrings (>= 2.35.12)

Imports affyio (>= 1.35.0), affxparser (>= 1.39.4), DBI (>= 0.3.1), ff, graphics, methods, preprocessCore (>= 1.29.0), RSQLite (>= 1.0.0), splines, stats, stats4, utils, zlibbioc

Enhances doMC, doMPI

LinkingTo preprocessCore

Suggests BSgenome.Hsapiens.UCSC.hg18, hapmap100kxba, pd.hg.u95av2, pd.mapping50k.xba240, pd.huex.1.0.st.v2, pd.hu18.60mer.expr, pd.hugene.1.0.st.v1, maqcExpression4plex, genefilter, limma, RColorBrewer, oligoData, BiocStyle, knitr, RUnit, biomaRt, AnnotationDbi, ACME, RCurl

VignetteBuilder knitr

Description A package to analyze oligonucleotide arrays (expression/SNP/tiling/exon) at probe-level. It currently supports Affymetrix (CEL files) and NimbleGen arrays (XYS files).

License LGPL (>= 2)

LazyLoad Yes

biocViews Microarray, OneChannel, TwoChannel, Preprocessing, SNP,
DifferentialExpression, ExonArray, GeneExpression, DataImport

git_url https://git.bioconductor.org/packages/oligo

git_branch RELEASE_3_18

git_last_commit e6e7283

git_last_commit_date 2023-10-24

Repository Bioconductor 3.18

Date/Publication 2024-03-27

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The oligo package: a tool for low-level analysis of oligonucleotide arrays

Description

The oligo package provides tools to preprocess different oligonucleotide arrays types: expression, tiling, SNP and exon chips. The supported manufacturers are Affymetrix and NimbleGen.

It offers support to large datasets (when the bigmemory is loaded) and can execute preprocessing tasks in parallel (if, in addition to bigmemory, the snow package is also loaded).

Details

The package will read the raw intensity files (CEL for Affymetrix; XYS for NimbleGen) and allow the user to perform analyses starting at the feature-level.

Reading in the intensity files require the existence of data packages that contain the chip specific information (X/Y coordinates; feature types; sequence). These data packages packages are built using the pdInfoBuilder package.

For Affymetrix SNP arrays, users are asked to download the already built annotation packages from BioConductor. This is because these packages contain metadata that are not automatically created. The following annotation packages are available:

50K Xba - pd.mapping50kxba.240 50K Hind - pd.mapping50khind.240 250K Sty - pd.mapping250k.sty 250K Nsp - pd.mapping250k.nsp GenomeWideSnp 5 (SNP 5.0) - pd.genomewidesnp.5 GenomeWideSnp 6 (SNP 6.0) - pd.genomewidesnp.6

For users interested in genotype calls for SNP 5.0 and 6.0 arrays, we strongly recommend the use use the crlmm package, which implements a more efficient version of CRLMM.

Author(s)

Benilton Carvalho - <carvalho@bclab.org>
References


### basecontent

**Sequence Base Contents**

<table>
<thead>
<tr>
<th>description</th>
<th>Function to compute the amounts of each nucleotide in a sequence.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Usage</strong></td>
<td>basecontent(seq)</td>
</tr>
<tr>
<td><strong>Arguments</strong></td>
<td>seq character vector of length n containing a valid sequence (A/T/C/G)</td>
</tr>
<tr>
<td><strong>Value</strong></td>
<td>matrix with n rows and 4 columns with the counts for each base.</td>
</tr>
<tr>
<td><strong>Examples</strong></td>
<td>sequences &lt;- c(&quot;ATATATCCCG&quot;, &quot;TTTCCGAGC&quot;) basecontent(sequences)</td>
</tr>
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</table>

### basicPLM

**Simplified interface to PLM.**

<table>
<thead>
<tr>
<th>description</th>
<th>Simplified interface to PLM.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Description</strong></td>
<td>Simplified interface to PLM.</td>
</tr>
<tr>
<td><strong>Usage</strong></td>
<td>basicPLM(pmMat, pnVec, normalize = TRUE, background = TRUE, transfo = log2, method = c('plm', 'plmr', 'plmrr', 'plmrc'), verbose = TRUE)</td>
</tr>
</tbody>
</table>
Arguments

- **pmMat**: Matrix of intensities to be processed.
- **pnVec**: Probeset names
- **normalize**: Logical flag: normalize?
- **background**: Logical flag: background adjustment?
- **transfo**: function: function to be used for data transformation prior to summarization.
- **method**: Name of the method to be used for normalization. 'plm' is the usual PLM model; 'plmr' is the (row and column) robust version of PLM; 'plmrr' is the row-robust version of PLM; 'plmrc' is the column-robust version of PLM.
- **verbose**: Logical flag: verbose.

Value

A list with the following components:

- **Estimates**: A (length(pnVec) x ncol(pmMat)) matrix with probeset summaries.
- **StdErrors**: A (length(pnVec) x ncol(pmMat)) matrix with standard errors of `Estimates`.
- **Residuals**: A (nrow(pmMat) x ncol(pmMat)) matrix of residuals.

Note

Currently, only RMA-bg-correction and quantile normalization are allowed.

Author(s)

Benilton Carvalho

See Also

`rcModelPLM`, `rcModelPLMr`, `rcModelPLMrr`, `rcModelPLMrc`, `basicRMA`

Examples

```r
set.seed(1)
pms <- 2^matrix(rnorm(1000), nc=20)
colnames(pms) <- paste("sample", 1:20, sep="")
pns <- rep(letters[1:10], each=5)
res <- basicPLM(pms, pns, TRUE, TRUE)
res[['Estimates']][1:4, 1:3]
res[['StdErrors']][1:4, 1:3]
res[['Residuals']][1:20, 1:3]
```
basicRMA

Simplified interface to RMA.

Description

Simple interface to RMA.

Usage

basicRMA(pmMat, pnVec, normalize = TRUE, background = TRUE, bgversion = 2, destructive = FALSE, verbose = TRUE, ...)

Arguments

pmMat Matrix of intensities to be processed.

pnVec Probeset names.

normalize Logical flag: normalize?

background Logical flag: background adjustment?

bgversion Version of background correction.

destructive Logical flag: use destructive methods?

verbose Logical flag: verbose.

... Not currently used.

Value

Matrix.

Examples

set.seed(1)
pms <- 2^matrix(rnorm(1000), nc=20)
colnames(pms) <- paste("sample", 1:20, sep="")
pns <- rep(letters[1:10], each=5)
res <- basicRMA(pms, pns, TRUE, TRUE)
res[, 1:3]
Description

Boxplot for observed (log-)intensities in a FeatureSet-like object (ExpressionFeatureSet, ExonFeatureSet, SnpFeatureSet, TilingFeatureSet) and ExpressionSet.

Usage

## S4 method for signature 'FeatureSet'
boxplot(x, which=c("pm", "mm", "bg", "both", "all"), transfo=log2, nsample=10000, target = "mps1", ...)

## S4 method for signature 'ExpressionSet'
boxplot(x, which, transfo=identity, nsample=10000, ...)

Arguments

x
  a FeatureSet-like object or ExpressionSet object.
which
  character defining what probe types are to be used in the plot.
transfo
  a function to transform the data before plotting. See 'Details'.
nsample
  number of units to sample and build the plot.
...
  arguments to be passed to the default boxplot method.

Details

The 'transfo' argument will set the transformation to be used. For raw data, 'transfo=log2' is a common practice. For summarized data (which are often in log2-scale), no transformation is needed (therefore 'transfo=identity').

Note

The boxplot methods for FeatureSet and Expression use a sample (via sample) of the probes/probesets to produce the plot. Therefore, the user interested in reproducibility is advised to use set.seed.

See Also

hist, image, sample, set.seed
chromosome

Accessor for chromosome information

Description

Returns chromosome information.

Usage

pmChr(object)

Arguments

object TilingFeatureSet or SnpCallSet object

Details

chromosome() returns the chromosomal information for all probes and pmChr() subsets the output to the PM probes only (if a TilingFeatureSet object).

Value

Vector with chromosome information.

crlmm

Genotype Calls

Description

Performs genotype calls via CRLMM (Corrected Robust Linear Model with Maximum-likelihood based distances).

Usage

```r
  crlmm(filenames, outdir, batch_size=40000, balance=1.5,
         minLLRforCalls=c(5, 1, 5), recalibrate=TRUE,
         verbose=TRUE, pkgname, reference=TRUE)
  justCRLMM(filenames, batch_size = 40000, minLLRforCalls = c(5, 1, 5),
            recalibrate = TRUE, balance = 1.5, phenoData = NULL, verbose = TRUE,
            pkgname = NULL, tmpdir=tempdir())
```
**Arguments**

- **filenames** character vector with the filenames.
- **outdir** directory where the output (and some tmp files) files will be saved.
- **batch_size** integer defining how many SNPs should be processed at a time.
- **recalibrate** Logical - should recalibration be performed?
- **balance** Control parameter to balance homozygotes and heterozygotes calls.
- **minLLRforCalls** Minimum thresholds for genotype calls.
- **verbose** Logical.
- **phenoData** phenoData object or NULL
- **pkgname** alt. pdInfo package to be used
- **reference** logical, defaulting to TRUE ...
- **tmpdir** Directory where temporary files are going to be stored at.

**Value**

SnpCallSetPlus object.

---

**darkColors**  
Create set of colors, interpolating through a set of preferred colors.

**Description**

Create set of colors, interpolating through a set of preferred colors.

**Usage**

darkColors(n)  
seqColors(n)  
seqColors2(n)  
divColors(n)

**Arguments**

- **n** integer determining number of colors to be generated

**Details**

darkColors is based on the Dark2 palette in RColorBrewer, therefore useful to describe qualitative features of the data.

seqColors is based on Blues and generates a gradient of blues, therefore useful to describe quantitative features of the data. seqColors2 behaves similarly, but it is based on OrRd (white-orange-red).

divColors is based on the RdBu pallete in RColorBrewer, therefore useful to describe quantitative features ranging on two extremes.
Examples

```r
x <- 1:10
y <- 1:10
cols1 <- darkColors(10)
cols2 <- seqColors(10)
cols3 <- divColors(10)
cols4 <- seqColors2(10)
plot(x, y, col=cols1, xlim=c(1, 13), pch=19, cex=3)
points(x+1, y, col=cols2, pch=19, cex=3)
points(x+2, y, col=cols3, pch=19, cex=3)
points(x+3, y, col=cols4, pch=19, cex=3)
abline(0, 1, lty=2)
abline(-1, 1, lty=2)
abline(-2, 1, lty=2)
abline(-3, 1, lty=2)
```

fitProbeLevelModel  

**Tool to fit Probe Level Models.**

Description

Fits robust Probe Level linear Models to all the (meta)probesets in an `FeatureSet`. This is carried out on a (meta)probeset by (meta)probeset basis.

Usage

```r
fitProbeLevelModel(object, background=TRUE, normalize=TRUE, target="core", method="plm", verbose=TRUE, S4=TRUE, ...)
```

Arguments

- `object` `FeatureSet` object.
- `background` Do background correction?
- `normalize` Do normalization?
- `target` character vector describing the summarization target. Valid values are: 'probeset', 'core' (Gene/Exon), 'full' (Exon), 'extended' (Exon).
- `method` summarization method to be used.
- `verbose` verbosity flag.
- `S4` return final value as an S4 object (`oligoPLM`) if TRUE. If FALSE, final value is returned as a list.
- `...` subset to be passed down to `getProbeInfo` for subsetting. See `subset` for details.

Value

`fitProbeLevelModel` returns an `oligoPLM` object, if S4=TRUE; otherwise, it will return a list.
getAffinitySplineCoefficients

Note

This is the initial port of fitPLM to oligo. Some features found on the original work by Ben Bolstad (in the affyPLM package) may not be yet available. If you found one of this missing characteristics, please contact Benilton Carvalho.

Author(s)

This is a simplified port from Ben Bolstad’s work implemented in the affyPLM package. Problems with the implementation in oligo should be reported to Benilton Carvalho.

References


See Also

rma, summarizationMethods, subset

Examples

```r
if (require(oligoData)){
  data(nimbleExpressionFS)
  fit <- fitProbeLevelModel(nimbleExpressionFS)
  image(fit)
  NUSE(fit)
  RLE(fit)
}
```

getAffinitySplineCoefficients

*Estimate affinity coefficients.*

Description

Estimate affinity coefficients using sequence information and splines.

Usage

getAffinitySplineCoefficients(intensities, sequences)

Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>intensities</td>
<td>Intensity matrix</td>
</tr>
<tr>
<td>sequences</td>
<td>Probe sequences</td>
</tr>
</tbody>
</table>
Value
Matrix with estimated coefficients.

See Also
getBaseProfile

getBaseProfile
Compute and plot nucleotide profile.

Description
Computes and, optionally, lots nucleotide profile, describing the sequence effect on intensities.

Usage
getBaseProfile(coefs, probeLength = 25, plot = FALSE, ...)

Arguments
coefs     affinity spline coefficients.
probeLength length of probes
plot      logical. Plots profile?
...       arguments to be passed to matplot.

Value
Invisibly returns a matrix with estimated effects.

getContainer
Get container information for NimbleGen Tiling Arrays.

Description
Get container information for NimbleGen Tiling Arrays. This is useful for better identification of control probes.

Usage
getContainer(object, probeType)

Arguments
object     A TilingFeatureSet or TilingFeatureSet object.
probeType  String describing which probes to query ('pm', 'bg')
getCrlmmSummaries

Value

'character' vector with container information.

Description

This will read the summaries written to disk and return them to the user as a SnpCallSetPlus or SnpCnvCallSetPlus object.

Usage

getcrlmmSummaries(tmpdir)

Arguments

tmpdir directory where CRLMM saved the results to.

Value

If the data were from SNP 5.0 or 6.0 arrays, the function will return a SnpCnvCallSetPlus object. It will return a SnpCallSetPlus object, otherwise.

getNetAffx

NetAffx Biological Annotations

Description

Gets NetAffx Biological Annotations saved in the annotation package (Exon and Gene ST Affymetrix arrays).

Usage

getcrlmmSummaries(object, type = "probeset")

Arguments

object 'ExpressionSet' object (eg., result of rma())
type Either 'probeset' or 'transcript', depending on what type of summaries were obtained.
**getNgsColorsInfo**

**Details**

This retrieves NetAffx annotation saved in the (pd) annotation package - annotation(object). It is only available for Exon ST and Gene ST arrays.

The 'type' argument should match the summarization target used to generate 'object'. The 'rma' method allows for two targets: 'probeset' (target='probeset') and 'transcript' (target='core', target='full', target='extended').

**Value**

'AnnotatedDataFrame' that can be used as featureData(object)

**Author(s)**

Benilton Carvalho

---

**getNgsColorsInfo**

*Helper function to extract color information for filenames on NimbleGen arrays.*

**Description**

This function will (try to) extract the color information for NimbleGen arrays. This is useful when using read.xysfiles2 to parse XYS files for Tiling applications.

**Usage**

getNgsColorsInfo(path = ".", pattern1 = "_532", pattern2 = "_635", ...)

**Arguments**

- **path**: path where to look for files
- **pattern1**: pattern to match files supposed to go to the first channel
- **pattern2**: pattern to match files supposed to go to the second channel
- ... extra arguments for list.xysfiles

**Details**

Many NimbleGen samples are identified following the pattern sampleID_532.XYS / sampleID_635.XYS. The function suggests sample names if all the filenames follow the standard above.

**Value**

A data.frame with, at least, two columns: 'channel1' and 'channel2'. A third column, 'sampleNames', is returned if the filenames follow the sampleID_532.XYS / sampleID_635.XYS standard.

**Author(s)**

Benilton Carvalho <bcarvalh@jhsph.edu>
getPlatformDesign  

Retrieve Platform Design object

**Description**
Retrieve platform design object.

**Usage**
getPlatformDesign(object)
getPD(object)

**Arguments**

object  
FeatureSet object

**Details**
Retrieve platform design object.

**Value**
platformDesign or PDInfo object.

getProbeInfo  

Probe information selector.

**Description**
A tool to simplify the selection of probe information, so user does not need to use the SQL approaches.

**Usage**
getProbeInfo(object, field, probeType = "pm", target = "core", sortBy = c("fid", "man_fsetid", "none"),

**Arguments**

object  
FeatureSet object.

field  
character string with names of field(s) of interest to be obtained from database.

probeType  
character string: 'pm' or 'mm'

target  
Used only for Exon or Gene ST arrays: 'core', 'full', 'extended', 'probeset'.

sortBy  
Field to be used for sorting.

...  
Arguments to be passed to subset
### Description

Accessors for physical array coordinates.

### Usage

```r
gGetX(object, type)
gGetY(object, type)
```

### Arguments

- **object** FeatureSet object
- **type** `character` defining the type of the probes to be queried. Valid options are `pm`, `mm`, `bg`

### Value

A vector with the requested coordinates.
Examples

```r
## Not run:
x <- read.celfiles(list.celfiles())
theXpm <- getX(x, "pm")
theYpm <- getY(x, "pm")

## End(Not run)
```

Description

Plot the density estimates for each sample

Usage

```r
## S4 method for signature 'FeatureSet'
hist(x, transfo=log2, which=c("pm", "mm", "bg", "both", "all"),
    nsample=10000, target = "mps1", ...)

## S4 method for signature 'ExpressionSet'
hist(x, transfo=identity, nsample=10000, ...)
```

Arguments

- `x` FeatureSet or ExpressionSet object
- `transfo` a function to transform the data before plotting. See 'Details'.
- `nsample` number of units to sample and build the plot.
- `which` set of probes to be plotted ("pm", "mm", "bg", "both", "all").
- `...` arguments to be passed to `matplot`

Details

The `transfo` argument will set the transformation to be used. For raw data, `transfo=log2` is a common practice. For summarized data (which are often in log2-scale), no transformation is needed (therefore `transfo=identity`).

Note

The hist methods for FeatureSet and Expression use a sample (via sample) of the probes/probesets to produce the plot (unless nsample > nrow(x)). Therefore, the user interested in reproducibility is advised to use `set.seed`. 

Display a pseudo-image of a microarray chip

**Description**

Produces a pseudo-image (graphics::image) for each sample.

**Usage**

```r
## S4 method for signature 'FeatureSet'
image(x, which, transfo=log2, ...)

## S4 method for signature 'PLMset'
image(x, which=0,
    type=c("weights", "resids", "pos.resids", "neg.resids", "sign.resids"),
    use.log=TRUE, add.legend=FALSE, standardize=FALSE,
    col=FALSE, main, ...)
```

**Arguments**

- `x` FeatureSet object
- `which` integer indices of samples to be plotted (optional).
- `transfo` function to be applied to the data prior to plotting.
- `type` Type of statistics to be used.
- `use.log` Use log.
- `add.legend` Add legend.
- `standardize` Standardize residuals.
- `col` Colors to be used.
- `main` Main title.
- `...` parameters to be passed to `image`

**Examples**

```r
if(require(oligoData) & require(pd.hg18.60mer.expr)){
  data(nimbleExpressionFS)
  par(mfrow=c(1, 2))
  image(nimbleExpressionFS, which=4)
  ## fit <- fitPLM(nimbleExpressionFS)
  ## image(fit, which=4)
  plot(1) ## while fixing fitPLM TODO
}
```
justSNPRMA

Summarization of SNP data

Description

This function implements the SNPRMA method for summarization of SNP data. It works directly with the CEL files, saving memory.

Usage

```r
justSNPRMA(filenames, verbose = TRUE, phenoData = NULL, normalizeToHapmap = TRUE)
```

Arguments

- `filenames`: character vector with the filenames.
- `verbose`: logical flag for verbosity.
- `phenoData`: a phenoData object or NULL.
- `normalizeToHapmap`: Normalize to Hapmap? Should always be TRUE, but it’s kept here for future use.

Value

SnpQSet or a SnpCnvQSet, depending on the array type.

Examples

```r
## snprmaResults <- justSNPRMA(list.celfiles())
```

list.xysfiles

List XYS files

Description

Lists the XYS files.

Usage

```
list.xysfiles(...)  
```

Arguments

- `...`: parameters to be passed to `list.files`
Details

The functions interface `list.files` and the user is asked to check that function for further details.

Value

Character vector with the filenames.

See Also

`list.files`

Examples

`list.xysfiles()`

MAplot

MA plots

Description

Create MA plots using a reference array (if one channel) or using channel2 as reference (if two channel).

Usage

`MAplot(object, ...)`

```r
## S4 method for signature 'FeatureSet'
MAplot(object, what=pm, transfo=log2, groups,
       refSamples, which, pch=".", summaryFun=rowMedians,
       plotFun=smoothScatter, main="vs pseudo-median reference chip",
       pairs=FALSE, ...)

## S4 method for signature 'TilingFeatureSet'
MAplot(object, what=pm, transfo=log2, groups,
       refSamples, which, pch=".", summaryFun=rowMedians,
       plotFun=smoothScatter, main="vs pseudo-median reference chip",
       pairs=FALSE, ...)

## S4 method for signature 'PLMset'
MAplot(object, what=coefs, transfo=identity, groups,
       refSamples, which, pch=".", summaryFun=rowMedians,
       plotFun=smoothScatter, main="vs pseudo-median reference chip",
       pairs=FALSE, ...)

## S4 method for signature 'matrix'
MAplot(object, what=identity, transfo=identity,
```
MAplot

```r
MAplot(groups, refSamples, which, pch=".", summaryFun=rowMedians,
plotFun=smoothScatter, main="vs pseudo-median reference chip",
pairs=FALSE, ...)
```

## S4 method for signature 'ExpressionSet'

```r
MAplot(object, what=exprs, transfo=identity,
groups, refSamples, which, pch=".", summaryFun=rowMedians,
plotFun=smoothScatter, main="vs pseudo-median reference chip",
pairs=FALSE, ...)
```

### Arguments

- **object**: FeatureSet, PLMset or ExpressionSet object.
- **what**: function to be applied on object that will extract the statistics of interest, from which log-ratios and average log-intensities will be computed.
- **transfo**: function to transform the data prior to plotting.
- **groups**: factor describing groups of samples that will be combined prior to plotting. If missing, MvA plots are done per sample.
- **refSamples**: integers (indexing samples) to define which subjects will be used to compute the reference set. If missing, a pseudo-reference chip is estimated using `summaryFun`.
- **which**: integer (indexing samples) describing which samples are to be plotted.
- **pch**: same as `pch` in `plot` summaryFun
- **summaryFun**: function that operates on a matrix and returns a vector that will be used to summarize data belonging to the same group (or reference) on the computation of grouped-stats.
- **plotFun**: function to be used for plotting. Usually `smoothScatter`, `plot` or `points`.
- **main**: string to be used in title.
- **pairs**: logical flag to determine if a matrix of MvA plots is to be generated
- **...**: Other arguments to be passed downstream, like `plot` arguments.

### Details

MAplot will take the following extra arguments:

1. **subset**: indices of elements to be plotted to reduce impact of plotting 100’s thousands points (if `pairs=FALSE` only);
2. **span**: see `loess`;
3. **family.loess**: see `loess`;
4. **addLoess**: logical flag (default `TRUE`) to add a loess estimate;
5. **parParams**: list of params to be passed to `par()` (if `pairs=TRUE` only);

### Value

Plot
Author(s)
Benilton Carvalho - based on Ben Bolstad’s original MAplot function.

See Also
plot, smoothScatter

Examples
if(require(oligoData) & require(pd.hg18.60mer.expr)){
  data(nimbleExpressionFS)
  nimbleExpressionFS
  groups <- factor(rep(c('brain', 'UnivRef'), each=3))
  data.frame(sampleNames(nimbleExpressionFS), groups)
  MAplot(nimbleExpressionFS, pairs=TRUE, ylim=c(-.5, .5), groups=groups)
}

---

Accessors and replacement methods for the intensity/PM/MM/BG matrices.

Description
Accessors and replacement methods for the PM/MM/BG matrices.

Usage
intensity(object)
mm(object, subset = NULL, target='core')
pm(object, subset = NULL, target='core')
bg(object, subset = NULL)
mm(object, subset = NULL, target='core')<-value
pm(object, subset = NULL, target='core')<-value
bg(object)<-value

Arguments
object FeatureSet object.
subset Not implemented yet.
value matrix object.
target One of 'probeset', 'core', 'full', 'extended'. This is ignored if the array design is something other than Gene ST or Exon ST.
mmindex

Accessors for PM, MM or background probes indices.

Description

Extracts the indexes for PM, MM or background probes.

Usage

mmindex(object, ...)

pmindex(object, ...)

bgindex(object, ...)

Arguments

object FeatureSet or DBPDInfo object

... Extra arguments, not yet implemented

Details

The indices are ordered by 'fid', i.e. they follow the order that the probes appear in the CEL/XYS files.

Value

A vector of integers representing the rows of the intensity matrix that correspond to PM, MM or background probes.
Examples

```r
## How pm() works
## Not run:
x <- read.celfiles(list.celfiles())
pms0 <- pm(x)
pmi <- pmindex(x)
pms1 <- exprs(x)[pmi,]
identical(pms0, pms1)

## End(Not run)
```

**mmSequence**

### Probe Sequences

Accessor to the (PM/MM/background) probe sequences.

**Usage**

```r
mmSequence(object)
```

**Arguments**

- `object` FeatureSet, AffySNPPDInfo or DBPInfo object
- `...` additional arguments

**Value**

A DNAStringSet containing the PM/MM/background probe sequence associated to the array.

**oligo-defunct**

### Defunct Functions in Package 'oligo'

**Description**

The functions or variables listed here are no longer part of 'oligo'

**Usage**

```r
fitPLM(...)
coefs(...)
resids(...)
```
Arguments

Arguments.

Details

fitPLM was replaced by fitProbeLevelModel, allowing faster execution and providing more specific models. fitPLM was based in the code written by Ben Bolstad in the affyPLM package. However, all the model-fitting functions are now in the package preprocessCore, on which fitProbeLevelModel depends.

coops and residS, like fitPLM, were inherited from the affyPLM package. They were replaced respectively by coef and residuals, because this is how these statistics are called everywhere else in R.

Description

A class to represent Probe Level Models.

Objects from the Class

Objects can be created by calls of the form fitProbeLevelModel(FeatureSetObject), where FeatureSetObject is an object obtained through read.celfiles or read.xysfiles, representing intensities observed for different probes (which are grouped in probesets or meta-probesets) across distinct samples.

Slots

- chip.coefs: "matrix" with chip/sample effects - probeset-level
- description: "MIAME" compliant description information.
- phenoData: "AnnotatedDataFrame" with phenotypic data.
- protocolData: "AnnotatedDataFrame" with protocol data.
- probe.coefs: "numeric" vector with probe effects
- weights: "matrix" with weights - probe-level
- residuals: "matrix" with residuals - probe-level
- se.chip.coefs: "matrix" with standard errors for chip/sample coefficients
- se.probe.coefs: "numeric" vector with standard errors for probe effects
- residualSE: scale - residual standard error
- geometry: array geometry used for plots
- method: "character" string describing method used for PLM
- manufacturer: "character" string with manufacturer name
annotation: "character" string with the name of the annotation package
narrays: "integer" describing the number of arrays
nprobes: "integer" describing the number of probes before summarization
nprobesets: "integer" describing the number of probesets after summarization

Methods

annotation signature(object = "oligoPLM"): accessor/replacement method to annotation slot
boxplot signature(x = "oligoPLM"): boxplot method
coeff signature(object = "oligoPLM"): accessor/replacement method to coeff slot
coeffs.probe signature(object = "oligoPLM"): accessor/replacement method to coefs.probe slot
geometry signature(object = "oligoPLM"): accessor/replacement method to geometry slot
image signature(x = "oligoPLM"): image method
manufacturer signature(object = "oligoPLM"): accessor/replacement method to manufacturer slot
method signature(object = "oligoPLM"): accessor/replacement method to method slot
ncol signature(x = "oligoPLM"): accessor/replacement method to ncol slot
nprobes signature(object = "oligoPLM"): accessor/replacement method to nprobes slot
nprobesets signature(object = "oligoPLM"): accessor/replacement method to nprobesets slot
residuals signature(object = "oligoPLM"): accessor/replacement method to residuals slot
residualSE signature(object = "oligoPLM"): accessor/replacement method to residualSE slot
se signature(object = "oligoPLM"): accessor/replacement method to se slot
se.probe signature(object = "oligoPLM"): accessor/replacement method to se.probe slot
show signature(object = "oligoPLM"): show method
weights signature(object = "oligoPLM"): accessor/replacement method to weights slot
NUSE signature(x = "oligoPLM"): Boxplot of Normalized Unscaled Standard Errors (NUSE) or NUSE values.
RLE signature(x = "oligoPLM"): Relative Log Expression boxplot or values.
opset2eset signature(x = "oligoPLM"): Convert to ExpressionSet.

Author(s)

This is a port from Ben Bolstad’s work implemented in the affyPLM package. Problems with the implementation in oligo should be reported to the package’s maintainer.

References


See Also

rma, summarize
Examples

```r
## TODO: review code and fix broken
## Not run:
if (require(oligoData)){
  data(nimbleExpressionFS)
  fit <- fitProbeLevelModel(nimbleExpressionFS)
  image(fit)
  NUSE(fit)
  RLE(fit)
}
## End(Not run)
```

---

### paCalls

**Methods for P/A Calls**

#### Description

Methods for Present/Absent Calls are meant to provide means of assessing whether or not each of the (PM) intensities are compatible with observations generated by background probes.

#### Usage

```r
paCalls(object, method, ..., verbose=TRUE)
## S4 method for signature 'ExonFeatureSet'
paCalls(object, method, verbose = TRUE)
## S4 method for signature 'GeneFeatureSet'
paCalls(object, method, verbose = TRUE)
## S4 method for signature 'ExpressionFeatureSet'
paCalls(object, method, ..., verbose = TRUE)
```

#### Arguments

- `object`: Exon/Gene/Expression-FeatureSet object.
- `method`: String defining what method to use. See 'Details'.
- `...`: Additional arguments passed to MAS5. See 'Details'
- `verbose`: Logical flag for verbosity.

#### Details

For Whole Transcript arrays (Exon/Gene) the valid options for method are 'DABG' (p-values for each probe) and 'PSDABG' (p-values for each probeset). For Expression arrays, the only option currently available for method is 'MAS5'.

**ABOUT MAS5 CALLS:**

The additional arguments that can be passed to MAS5 are:

1. `alpha1`: a significance threshold in (0, alpha2);
2. alpha2: a significance threshold in (alpha1, 0.5);
3. tau: a small positive constant;
4. ignore.saturated: if TRUE, do the saturation correction described in the paper, with a saturation level of 46000;

This function performs the hypothesis test:
H0: median(Ri) = tau, corresponding to absence of transcript
H1: median(Ri) > tau, corresponding to presence of transcript

where Ri = (PMi - MMi) / (PMi + MMi) for each i a probe-pair in the probe-set represented by data.

The p-value that is returned estimates the usual quantity:
Pr(observing a more "present looking" probe-set than data | data is absent)

So that small p-values imply presence while large ones imply absence of transcript. The detection call is computed by thresholding the p-value as in:
call "P" if p-value < alpha1 call "M" if alpha1 <= p-value < alpha2 call "A" if alpha2 <= p-value

Value

A matrix (of dimension dim(PM) if method="DABG" or "MAS5"; of dimension length(unique(probeNames(object))) x ncol(object) if method="PSDABG") with p-values for P/A Calls.

Author(s)

Benilton Carvalho

References


Examples

```r
## Not run:
if (require(oligoData) & require(pd.huex.1.0.st.v2)){
  data(affyExonFS)
  ## Get only 2 samples for example
dabgP = paCalls(affyExonFS[, 1:2])
dabgPS = paCalls(affyExonFS[, 1:2], "PSDABG")
}
plotM-methods

Methods for Log-Ratio plotting

Description

The plotM methods are meant to plot log-ratios for different classes of data.

Methods

object = "SnpQSet", i = "character"  Plot log-ratio for SNP data for sample i.
object = "SnpQSet", i = "integer"    Plot log-ratio for SNP data for sample i.
object = "SnpQSet", i = "numeric"   Plot log-ratio for SNP data for sample i.
object = "TilingQSet", i = "missing" Plot log-ratio for Tiling data for sample i.

pmAllele

Access the allele information for PM probes.

Description

Accessor to the allelic information for PM probes.

Usage

pmAllele(object)

Arguments

object  SnpFeatureSet or PDInfo object.
pmFragmentLength

*Access the fragment length for PM probes.*

**Description**

Accessor to the fragment length for PM probes.

**Usage**

```r
pmFragmentLength(object, enzyme, type=c('snp', 'cn'))
```

**Arguments**

- `object`: PDInfo or SnpFeatureSet object.
- `enzyme`: Enzyme to be used for query. If missing, all enzymes are used.
- `type`: Type of probes to be used: 'snp' for SNP probes; 'cn' for Copy Number probes.

**Value**

A list of length equal to the number of enzymes used for digestion. Each element of the list is a data.frame containing:

- `row`: the row used to link to the PM matrix;
- `length`: expected fragment length.

**Note**

There is not a 1:1 relationship between probes and expected fragment length. For one enzyme, a given probe may be associated to multiple fragment lengths. Therefore, the number of rows in the data.frame may not match the number of PM probes and the `row` column should be used to match the fragment length with the PM matrix.

pmPosition

*Accessor to position information*

**Description**

`pmPosition` will return the genomic position for the (PM) probes.

**Usage**

```r
pmPosition(object)
pmOffset(object)
```
**pmStrand**

**Arguments**

- `object` AffySNPPDInfo, TilingFeatureSet or SnpCallSet object

**Details**

- `pmPosition` will return genomic position for PM probes on a tiling array.
- `pmOffset` will return the offset information for PM probes on SNP arrays.

**Description**

Returns the strand information for PM probes (0 - sense / 1 - antisense).

**Usage**

- `pmStrand(object)`

**Arguments**

- `object` AffySNPPDInfo or TilingFeatureSet object

**probeNames**

**Accessor to feature names**

**Description**

Accessors to featureset names.

**Usage**

- `probeNames(object, subset = NULL, ...)`
- `probesetNames(object, ...)`

**Arguments**

- `object` FeatureSet or DBPDInfo
- `subset` not implemented yet.
- `...` Arguments (like 'target') passed to downstream methods.

**Value**

`probeNames` returns a string with the probeset names for *each probe* on the array. `probesetNames`, on the other hand, returns the *unique probeset names*.
**read.celfiles**

*Parser to CEL files*

---

**Description**

Reads CEL files.

**Usage**

```r
read.celfiles(..., filenames, pkgname, phenoData, featureData, 
experimentData, protocolData, notes, verbose=TRUE, sampleNames, 
rm.mask=FALSE, rm.outliers=FALSE, rm.extra=FALSE, checkType=TRUE)
```

```r
read.celfiles2(channel1, channel2, pkgname, phenoData, featureData, 
experimentData, protocolData, notes, verbose=TRUE, sampleNames, 
rm.mask=FALSE, rm.outliers=FALSE, rm.extra=FALSE, checkType=TRUE)
```

**Arguments**

- `...`: names of files to be read.
- `filenames`: a character vector with the CEL filenames.
- `channel1`: a character vector with the CEL filenames for the first ‘channel’ on a Tiling application.
- `channel2`: a character vector with the CEL filenames for the second ‘channel’ on a Tiling application.
- `pkgname`: alternative data package to be loaded.
- `phenoData`: phenoData.
- `featureData`: featureData.
- `experimentData`: experimentData.
- `protocolData`: protocolData.
- `notes`: notes.
- `verbose`: logical. Print progress messages ?
- `sampleNames`: character vector with sample names (usually better descriptors than the filenames).
- `rm.mask`: logical. Read masked?
- `rm.outliers`: logical. Remove outliers?
- `rm.extra`: logical. Remove extra?
- `checkType`: logical. Check type of each file? This can be time consuming.
Details

When using 'affyio' to read in CEL files, the user can read compressed CEL files (CEL.gz). Additionally, 'affyio' is much faster than 'affxparser'.

The function guesses which annotation package to use from the header of the CEL file. The user can also provide the name of the annotation package to be used (via the pkgname argument). If the annotation package cannot be loaded, the function returns an error. If the annotation package is not available from BioConductor, one can use the pdInfoBuilder package to build one.

Value

ExpressionFeatureSet
  if Expression arrays
ExonFeatureSet  if Exon arrays
SnpFeatureSet   if SNP arrays
TilingFeatureSet if Tiling arrays

See Also

list.celfiles, read.xysfiles

Examples

if(require(pd.mapping50k.xba240) & require(hapmap100kxba)){
  celPath <- system.file("celFiles", package="hapmap100kxba")
  celFiles <- list.celfiles(celPath, full.name=TRUE)
  affySnpFeatureSet <- read.celfiles(celFiles)
}

--

read.xysfiles                  Parser to XYS files
--

Description

NimbleGen provides XYS files which are read by this function.

Usage

read.xysfiles(..., filenames, pkgname, phenoData, featureData,
  experimentData, protocolData, notes, verbose=TRUE, sampleNames,
  checkType=TRUE)

read.xysfiles2(channel1, channel2, pkgname, phenoData, featureData,
  experimentData, protocolData, notes, verbose=TRUE, sampleNames,
  checkType=TRUE)
Arguments

- filenames: character vector with filenames.
- channel1: a character vector with the XYS filenames for the first 'channel' on a Tiling application.
- channel2: a character vector with the XYS filenames for the second 'channel' on a Tiling application.
- pkgname: character vector with alternative PD Info package name.
- phenoData: phenoData.
- featureData: featureData.
- experimentData: experimentData.
- protocolData: protocolData.
- notes: notes.
- verbose: verbose.
- sampleNames: character vector with sample names (usually better descriptors than the filenames).
- checkType: logical. Check type of each file? This can be time consuming.

Details

The function will read the XYS files provided by NimbleGen Systems and return an object of class FeatureSet.

The function guesses which annotation package to use from the header of the XYS file. The user can also provide the name of the annotation package to be used (via the pkgname argument). If the annotation package cannot be loaded, the function returns an error. If the annotation package is not available from BioConductor, one can use the pdInfoBuilder package to build one.

Value

- ExpressionFeatureSet
- TilingFeatureSet

See Also

list.xysfiles, read.celfiles

Examples

```r
if (require(maqcExpression4plex) & require(pd.hg18.60mer.expr)){
  xysPath <- system.file("extdata", package="maqcExpression4plex")
  xysFiles <- list.xysfiles(xysPath, full.name=TRUE)
  ngsExpressionFeatureSet <- read.xysfiles(xysFiles)
}
```
readSummaries  

Read summaries generated by crlmm

Description

This function reads the different summaries generated by crlmm.

Usage

readSummaries(type, tmpdir)

Arguments

type: type of summary of character class:
  - 'alleleA'
  - 'alleleB'
  - 'alleleA-sense'
  - 'alleleA-antisense'
  - 'alleleB-sense'
  - 'alleleB-antisense'
  - 'calls'
  - 'llr'
  - 'conf'

tmpdir: directory containing the output saved by crlmm

Details

On the 50K and 250K arrays, given a SNP, there are probes on both strands (sense and antisense). For this reason, the options 'alleleA-sense', 'alleleA-antisense', 'alleleB-sense' and 'alleleB-antisense' should be used **only** with such arrays (XBA, HIND, NSP or STY).

On the SNP 5.0 and SNP 6.0 platforms, this distinction does not exist in terms of algorithm (note that the actual strand could be queried from the annotation package). For these arrays, options 'alleleA', 'alleleB' are the ones to be used.

The options calls, llr and conf will return, respectively, the CRLMM calls, log-likelihood ratios (for development purpose **only**) and CRLMM confidence calls matrices.

Value

Matrix with values of summaries.

rma-methods  

RMA - Robust Multichip Average algorithm

Description

Robust Multichip Average preprocessing methodology. This strategy allows background subtraction, quantile normalization and summarization (via median-polish).
### Usage

#### S4 method for signature 'ExonFeatureSet'
```r
rma(object, background=TRUE, normalize=TRUE, subset=NULL, target="core")
```

#### S4 method for signature 'HTAFeatureSet'
```r
rma(object, background=TRUE, normalize=TRUE, subset=NULL)
```

#### S4 method for signature 'ExpressionFeatureSet'
```r
rma(object, background=TRUE, normalize=TRUE, subset=NULL)
```

#### S4 method for signature 'GeneFeatureSet'
```r
rma(object, background=TRUE, normalize=TRUE, subset=NULL, target="core")
```

#### S4 method for signature 'SnpCnvFeatureSet'
```r
rma(object, background=TRUE, normalize=TRUE, subset=NULL)
```

### Arguments

- **object**: Exon/HTA/Expression/Gene/SnpCnv-FeatureSet object.
- **background**: Logical - perform RMA background correction?
- **normalize**: Logical - perform quantile normalization?
- **subset**: To be implemented.
- **target**: Level of summarization (only for Exon/Gene arrays)

### Methods

- **signature(object = "ExonFeatureSet")**: When applied to an ExonFeatureSet object, rma can produce summaries at different levels: probeset (as defined in the PGF), core genes (as defined in the core.mps file), full genes (as defined in the full.mps file) or extended genes (as defined in the extended.mps file). To determine the level for summarization, use the target argument.

- **signature(object = "ExpressionFeatureSet")**: When used on an ExpressionFeatureSet object, rma produces summaries at the probeset level (as defined in the CDF or NDF files, depending on the manufacturer).

- **signature(object = "GeneFeatureSet")**: When applied to a GeneFeatureSet object, rma can produce summaries at different levels: probeset (as defined in the PGF) and 'core genes' (as defined in the core.mps file). To determine the level for summarization, use the target argument.

- **signature(object = "HTAFeatureSet")**: When applied to a HTAFeatureSet object, rma can produce summaries at different levels: probeset (as defined in the PGF) and 'core genes' (as defined in the core.mps file). To determine the level for summarization, use the target argument.

- **signature(object = "SnpCnvFeatureSet")**: If used on a SnpCnvFeatureSet object (ie., SNP 5.0 or SNP 6.0 arrays), rma will produce summaries for the CNV probes. Note that this is an experimental feature for internal (and quick) assessment of CNV probes. We recommend the use of the 'crlmm' package, which contains a Copy Number tool specifically designed for these data.
runDate

References


See Also

snprma

Examples

```r
if (require(maqcExpression4plex) & require(pd.hg18.60mer.expr)){
  xysPath <- system.file("extdata", package="maqcExpression4plex")
  xysFiles <- list.xysfiles(xysPath, full.name=TRUE)
  ngsExpressionFeatureSet <- read.xysfiles(xysFiles)
  summarized <- rma(ngsExpressionFeatureSet)
  show(summarized)
}
```

---

runDate   
Date of scan

Description

Retrieves date information in CEL/XYS files.

Usage

```r
runDate(object)
```

Arguments

- **object** : 'FeatureSet' object.
sequenceDesignMatrix  

*Create design matrix for sequences*

**Description**

Creating design matrix for sequences.

**Usage**

```r
sequenceDesignMatrix(seqs)
```

**Arguments**

- `seqs` character vector of 25-mers.

**Details**

This assumes all sequences are 25bp long.

The design matrix is often used when the objective is to adjust intensities by sequence.

**Value**

Matrix with length(seqs) rows and 75 columns.

**Examples**

```r
genSequence <- function(x)
    paste(sample(c("A", "T", "C", "G"), 25, rep=TRUE), collapse="", sep="")
seqs <- sapply(1:10, genSequence)
X <- sequenceDesignMatrix(seqs)
Y <- rnorm(10, mean=12, sd=2)
Ydemean <- Y-mean(Y)
X[1:10, 1:3]
fit <- lm(Ydemean~X)
coef(fit)
```

---

**snprma**  

*Preprocessing SNP Arrays*

**Description**

This function preprocess SNP arrays.

**Usage**

```r
snprma(object, verbose = TRUE, normalizeToHapmap = TRUE)
```
summarize

Arguments

- object: SnpFeatureSet object
- verbose: Verbosity flag, logical
- normalizeToHapmap: internal

Value

A SnpQSet object.

Description

These are tools to preprocess microarray data. They include background correction, normalization and summarization methods.

Usage

backgroundCorrectionMethods()
normalizationMethods()
summarizationMethods()
backgroundCorrect(object, method=backgroundCorrectionMethods(), copy=TRUE, extra, subset=NULL, target='core', verbose=TRUE, ...)
summarize(object, probes=rownames(object), method="medianpolish", verbose=TRUE, ...)
## S4 method for signature 'FeatureSet'
summarize(object, method=normalizationMethods(), copy=TRUE, subset=NULL, target='core', verbose=TRUE, ...)
## S4 method for signature 'matrix'
summarize(object, method=normalizationMethods(), copy=TRUE, verbose=TRUE, ...)
## S4 method for signature 'ff_matrix'
summarize(object, method=normalizationMethods(), copy=TRUE, verbose=TRUE, ...)
normalizeToTarget(object, targetDist, method="quantile", copy=TRUE, verbose=TRUE)

Arguments

- object: Object containing probe intensities to be preprocessed.
- method: String determining which method to use at that preprocessing step.
- targetDist: Vector with the target distribution
- probes: Character vector that identifies the name of the probes represented by the rows of object.
- copy: Logical flag determining if data must be copied before processing (TRUE), or if data can be overwritten (FALSE).
- subset: Not yet implemented.
- target: One of the following values: 'core', 'full', 'extended', 'probeset'. Used only with Gene ST and Exon ST designs.
extra Extra arguments to be passed to other methods.
verbose Logical flag for verbosity.
... Arguments to be passed to methods.

Details
Number of rows of object must match the length of probes.

Value
backgroundCorrectionMethods and normalizationMethods will return a character vector with the methods implemented currently.
backgroundCorrect, normalize and normalizeToTarget will return a matrix with same dimensions as the input matrix. If they are applied to a FeatureSet object, the PM matrix will be used as input.
The summarize method will return a matrix with length(unique(probes)) rows and ncol(object) columns.

Examples
ns <- 100
nps <- 1000
np <- 10
intensities <- matrix(rnorm(ns*nps*np, 8000, 400), nc=ns)
ids <- rep(as.character(1:nps), each=np)
bgCorrected <- backgroundCorrect(intensities)
normalized <- normalize(bgCorrected)
summarizationMethods()
expression <- summarize(normalized, probes=ids)
intensities[1:20, 1:3]
expression[1:20, 1:3]
target <- rnorm(np*nps)
normalizedToTarget <- normalizeToTarget(intensities, target)

if (require(oligoData) & require(pd.hg18.60mer.expr)){
  ## Example of normalization with real data
data(nimbleExpressionFS)
boxplot(nimbleExpressionFS, main='Original')
for (mtd in normalizationMethods()){  
  message('Normalizing with ', mtd)
  res <- normalize(nimbleExpressionFS, method=mtd, verbose=FALSE)
  boxplot(res, main=mtd)
}  
}


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