Package ‘flowStats’

January 31, 2017

Type Package
Title Statistical methods for the analysis of flow cytometry data
Version 3.32.0
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Description Methods and functionality to analyse flow data that is beyond the
basic infrastructure provided by the flowCore package.
Depends R (>= 2.10), flowCore, fda (>= 2.2.6), cluster, flowWorkspace,
ncdfFlow(>= 2.19.5)
Suggests xtable
Imports BiocGenerics, MASS, flowViz, flowCore, fda (>= 2.2.6),
Biobase, methods, grDevices, graphics, stats, utils,
KernSmooth, lattice,ks
Enhances RBGL,ncdfFlow,graph
License Artistic-2.0
Lazyload yes
biocViews FlowCytometry, CellBasedAssays
NeedsCompilation no

R topics documented:

<table>
<thead>
<tr>
<th>Topic</th>
<th>Pages</th>
</tr>
</thead>
<tbody>
<tr>
<td>flowStats-package</td>
<td>2</td>
</tr>
<tr>
<td>autoGate</td>
<td>2</td>
</tr>
<tr>
<td>BackGating</td>
<td>3</td>
</tr>
<tr>
<td>binByRef</td>
<td>4</td>
</tr>
<tr>
<td>calcPBChiSquare</td>
<td>4</td>
</tr>
<tr>
<td>calcPearsonChi</td>
<td>5</td>
</tr>
<tr>
<td>curv1Filter-class</td>
<td>6</td>
</tr>
<tr>
<td>curv2Filter-class</td>
<td>8</td>
</tr>
<tr>
<td>curvPeaks</td>
<td>10</td>
</tr>
<tr>
<td>densityId</td>
<td>11</td>
</tr>
<tr>
<td>gaussNorm</td>
<td>13</td>
</tr>
<tr>
<td>gpaSet</td>
<td>15</td>
</tr>
<tr>
<td>idFeaturesByBackgating</td>
<td>17</td>
</tr>
</tbody>
</table>
flowStats-package  Statistical methods for flow cytometry data analysis

Description
Functions, methods and classes implementing algorithms for statistical analysis of flow cytometry data. This involves mostly data normalization and automated gating.

Details

Package: flowStats
Type: Package
Version: 1.0
License: Artistic-2.0
Lazyload: yes

Author(s)

Florian Hahne
Maintainer: Florian Hahne <fahne@fhcrc.org>

autoGate  Automated gating of single populations in 2D

Description
This function tries to fit a single norm2Filter based on a rough preselection of the data. This function is considered internal. Please use the API provided by lymphGate.

Usage

autoGate(x, ..., scale = 2.5)
**Arguments**

- **x** (An object of class `flowSet`)
- **...** (Named arguments or a list of the ranges used for the initial rough preselection. This gets passed on to `rectangleGate`, see its documentation for details.)
- **scale** (The scale parameter that gets passed on to `norm2Filter`.)

**Details**

The `flowSet` is first filtered using a `rectangleGate` and the `norm2Filter` is subsequently fitted to the remaining subset.

**Value**

A list with items:

- **x** (The filtered `flowSet`.)
- **n2gate** (The `norm2Filter` object.)
- **n2gateResults** (The `filterResult` after applying the `norm2Filter` on the `flowSet`.)

**Author(s)**

Florian Hahne

**See Also**

`lymphGate`, `norm2Filter`

**Examples**

```r
data(GvHD)
flowStats::autoGate(GvHD[10:15], "FSC-H"=c(100,500), "SSC-H"=c(0, 400))
```

---

**BackGating**

**Sample backgating results**

**Description**

A data frame containing the sub-populations of ITN dataset corresponding to the high-density areas on "FSC" and "SSC" channels. This dataset is yielded by backGating on channel CD3, CD8, and CD4 of the ITN sample data.

**Usage**

```r
data(BackGating)
```

**Source**

Results from executing the following code:

```r
data(ITN)
flowStats:::backGating(ITN, xy=c("FSC", "SSC"), channels=c("CD3", "CD8", "CD4"))
```
Bin a test data set using bins previously created by probability binning a control dataset

Description

The bins generated by probability binning a control data set can be applied to a test data set to perform statistical comparisons by methods such as the Chi-squared test or the probability binning statistic.

Usage

\[
\text{binByRef}(\text{binRes}, \text{data})
\]

Arguments

- \text{binRes}: The result generated by calling the \text{probBin} function on a control dataset.
- \text{data}: An object of class \text{flowFrame}

Value

An environment containing the matrices for each bin of the test data set

Author(s)

Nishant Gopalakrishnan

See Also

\text{plotBins}, \text{proBin}

Examples

```r
data(GvHD)
resCtrl<-\text{proBin}(GvHD[1],200)
resSample<-\text{binByRef}(resCtrl,GvHD[2])
\text{ls}(\text{resSample})
```

Probability binning metric for comparing the probability binned datasets

Description

This function calculates the Probability binning metric proposed by Baggerly et al. The function utilizes the data binned using the \text{proBin} and \text{binByRef} functions.

Usage

\[
\text{calcPBChiSquare}(\text{ctrlRes}, \text{sampRes}, \text{ctrlCount}, \text{sampCount})
\]
calcPearsonChi

Arguments

ctrlRes The result generated by calling the probBin function on a control dataset.
sampRes The result generated by calling the byByRef function on a test sample dataset
ctrlCount The number of events in the control sample
sampCount The number of events in the test sample being compared

Value
A list containing the statistic, p.value, observed, expected counts and the residuals

Author(s)
Nishant Gopalakrishnan

See Also
probBin, calcPBChiSquare

Examples

```r
data(GvHD)
# flow frame 1 is treated as control dataset and used to generate bins
resCtrl<-probBin(GvHD[[1]][,c("FSC-H","SSC-H","Time")],200)
plotBins(resCtrl,GvHD[[1]],channels=c("FSC-H","SSC-H","Time"),title="Binned control data")
# Same bins are applied to flowFrame 16
resSample<-binByRef(resCtrl,GvHD[[16]][,c("FSC-H","SSC-H","Time")])
ctrlCount<-nrow(GvHD[[1]])
sampCount<-nrow(GvHD[[16]])
stat<-calcPBChiSquare(resCtrl,resSample,ctrlCount,sampCount)
```

Description
This function calculates the Pearson's chi-squared statistic for comparing data binned using the probBin and binByRef functions. Internally, the function utilizes the chisq.test function.

Usage

calcPearsonChi(ctrlRes,sampRes)

Arguments

ctrlRes The result generated by calling the probBin function on a control dataset.
sampRes The result generated by calling the byByRef function on a sample dataset

Value
A list containing the statistic, p.value, observed, expected counts and the residuals
**Examples**

```r
data(GvHD)
# flow frame 1 is treated as control dataset and used to generate bins
resCtrl<-proBin(GvHD[[1]][,c("FSC-H","SSC-H","Time")],200)
plotBins(resCtrl,GvHD[[1]],channels=c("FSC-H","SSC-H","Time"),title="Binned control data")
# Same bins are applied to flowFrame 16
resSample<-binByRef(resCtrl,GvHD[[16]][,c("FSC-H","SSC-H","Time")])
stat<-calcPearsonChi(resCtrl,resSample)
```

---

**Description**

Class and constructor for data-driven filter objects that selects high-density regions in one dimension.

**Usage**

```r
curv1Filter(x, bwFac=1.2, gridsize=rep(401, 2), filterId="defaultCurv1Filter")
```

**Arguments**

- **x**
  Character giving the name of the measurement parameter on which the filter is supposed to work on. This can also be a list containing a single character scalar for programmatic access.

- **filterId**
  An optional parameter that sets the filterId slot of this filter. The object can later be identified by this name.

- **bwFac, gridsize**
  Numerics of length 1 and 2, respectively, used to set the bwFac and gridsize slots of the object.

**Details**

Areas of high local density in one dimensions are identified by detecting significant curvature regions. See Duong, T. and Cowling, A. and Koch, I. and Wand, M.P., Computational Statistics and Data Analysis 52/9, 2008 for details. The constructor curv1Filter is a convenience function for object instantiation. Evaluating a curv1Filter results in potentially multiple sub-populations, an hence in an object of class multipleFilterResult. Accordingly, curv1Filters can be used to split flow cytometry data sets.
**curv1Filter-class**

**Value**

Returns a curv1Filter object for use in filtering flowFrames or other flow cytometry objects.

**Extends**

Class "parameterFilter", directly.
Class "concreteFilter", by class parameterFilter, distance 2.
Class "filter", by class parameterFilter, distance 3.

**Slots**

- **bwFac**: Object of class "numeric". The bandwidth factor used for smoothing of the density estimate.
- **gridsize**: Object of class "numeric". The size of the bins used for density estimation.
- **parameters**: Object of class "character", describing the parameter used to filter the flowFrame.
- **filterId**: Object of class "character", referencing the filter.

**Objects from the Class**

Objects can be created by calls of the form `new("curvFilter", ...)` or using the constructor `curv1Filter`. Using the constructor is the recommended way of object instantiation:

**Methods**

- `%in%` signature(x = "flowFrame", table = "curv1Filter"): The workhorse used to evaluate the filter on data. This is usually not called directly by the user, but internally by calls to the filter methods.
- `show` signature(object = "curv1Filter"): Print information about the filter.

**Note**

See the documentation in the flowViz package for plotting of curv1Filters.

**Author(s)**

Florian Hahne

**See Also**

curv2Filter, flowFrame, flowSet, filter for evaluation of curv1Filters and split for splitting of flow cytometry data sets based on that.

**Examples**

```r
library(flowStats)
## Loading example data
dat <- read.FCS(system.file("extdata","0877408774.B08", package="flowCore"))

## Create directly. Most likely from a command line
curv1Filter("FSC-H", filterId="myCurv1Filter", bwFac=2)
```
To facilitate programmatic construction we also have the following:

c1f <- curv1Filter(filterId="myCurv1Filter", x=list("FSC-H"), bwFac=2)

Filtering using curv1Filter:
fres <- filter(dat, c1f)
fres
summary(fres)
names(fres)

The result of curv1 filtering are multiple sub-populations:

split(dat, fres)

We can limit the splitting to one or several sub-populations:

split(dat, fres, population="rest")
split(dat, fres, population=list(keep=c("peak 2", "peak 3")))

curv2Filter-class Class “curv2Filter”

Description

Class and constructor for data-driven filter objects that selects high-density regions in two dimensions.

Usage

curv2Filter(x, y, filterId="defaultCurv2Filter", bwFac=1.2,
gridsize=rep(151, 2))

Arguments

x, y Characters giving the names of the measurement parameter on which the filter is supposed to work on. y can be missing in which case x is expected to be a character vector of length 2 or a list of characters.

filterId An optional parameter that sets the filterId slot of this filter. The object can later be identified by this name.

bwFac, gridsize Numerics of length 1 and 2, respectively, used to set the bwFac and gridsize slots of the object.

Details

Areas of high local density in two dimensions are identified by detecting significant curvature regions. See Duong, T. and Cowling, A. and Koch, I. and Wand, M.P., Computational Statistics and Data Analysis 52/9, 2008 for details. The constructor curv2Filter is a convenience function for object instantiation. Evaluating a curv2Filter results in potentially multiple sub-populations, hence in an object of class multipleFilterResult. Accordingly, curv2Filters can be used to split flow cytometry data sets.
Value

Returns a `curv2Filter` object for use in filtering `flowFrames` or other flow cytometry objects.

Extends

Class "parameterFilter", directly.
Class "concreteFilter", by class `parameterFilter`, distance 2.
Class "filter", by class `parameterFilter`, distance 3.

Slots

`bwFac`: Object of class "numeric". The bandwidth factor used for smoothing of the density estimate.
`gridsize`: Object of class "numeric". The size of the bins used for density estimation.
`parameters`: Object of class "character", describing the parameters used to filter the `flowFrame`.
`filterId`: Object of class "character", referencing the filter.

Objects from the Class

Objects can be created by calls of the form `new("curv2Filter", ...)` or using the constructor `curv2Filter`. The constructor is the recommended way of object instantiation.

Methods

`%in%` signature(`x = "flowFrame", table = "curv2Filter")`: The workhorse used to evaluate the filter on data. This is usually not called directly by the user, but internally by calls to the `filter` methods.
`show` signature(`object = "curv2Filter"`): Print information about the filter.

Note

See the documentation in the `flowViz` package for plotting of `curv2Filters`.

Author(s)

Florian Hahne

See Also

`curv1Filter`, `flowFrame`, `flowSet`, `filter` for evaluation of `curv2Filters` and `split` for splitting of flow cytometry data sets based on that.

Examples

```r
## Loading example data
dat <- read.FCS(system.file("extdata","0877408774.B08", package="flowCore"))

## Create directly. Most likely from a command line
curv2Filter("FSC-H", "SSC-H", filterId="myCurv2Filter")

## To facilitate programmatic construction we also have the following
```
c2f <- curv2Filter(filterId="myCurv2Filter", x=list("FSC-H", "SSC-H"), bwFac=2)
c2f <- curv2Filter(filterId="myCurv2Filter", x=c("FSC-H", "SSC-H"), bwFac=2)

## Filtering using curv2Filter
fres <- filter(dat, c2f)
fres
summary(fres)
names(fres)

## The result of curv2 filtering are multiple sub-populations
## and we can split our data set accordingly
split(dat, fres)

## We can limit the splitting to one or several sub-populations
split(dat, fres, population="rest")
split(dat, fres, population=list(keep=c("area 2", "area 3")))

curv2Filter("FSC-H", "SSC-H", filterId="test filter")

---

### curvPeaks

**Parse curv1Filter output**

**Description**

Parse the output of `curv1Filter` and find modes and midpoints of the high-density regions. This function is considered to be internal.

**Usage**

```r
curvPeaks(x, dat, borderQuant = 0.01, n = 201, from, to, densities=NULL)
```

**Arguments**

- `x` A `multipleFilterResult` produced by a `curv1Filter` operation.
- `dat` The corresponding `flowFrame`.
- `borderQuant` A numeric in $[0, 1]$ giving the extreme quantiles for which high-density regions are ignored.
- `n, from, to` Arguments are passed on to `density`.
- `densities` The optional y values of the density estimate computed for the respective data.

**Value**

A list with items

- `peaks` x and y locations of the modes of the regions in the density estimates.
- `regions` the left and right margins of the regions.
density1d

midpoints  the mean of regions.
regPoints  x and y locations of the outline of the significant density regions.
densFuns an approximation function of the density estimate

Author(s)
Florian Hahne

See Also
landmarkMatrix

Examples

data(GvHD)
tmp <- filter(GvHD[[10]], curv1Filter("FSC-H"))
res <- flowStats:::curvPeaks(tmp, exprs(GvHD[[10]])[, "FSC-H"])

density1d  Find most likely separation between positive and negative populations in 1D

Description

The function tries to find a reasonable split point between the two hypothetical cell populations "positive" and "negative". This function is considered internal, please use the API provided by rangeGate.

Usage
density1d(x, stain, alpha = "min", sd = 2, plot = FALSE, borderQuant = 0.1, absolute = TRUE, inBetween = FALSE, refLine=NULL, rare=FALSE, bwFac=1.2, sig=NULL, peakNr=NULL, ...)

Arguments

x A flowSet or flowFrame.
stain A character scalar giving the flow parameter for which to compute the separation.
alpha A tuning parameter that controls the location of the split point between the two populations. This has to be a numeric in the range [0,1], where values closer to 0 will shift the split point closer to the negative population and values closer to 1 will shift towards the positive population. Additionally, the value of alpha can be "min", in which case the split point will be selected as the area of lowest local density between the two populations.
sd For the case where there is only a single population, the algorithm falls back to estimating the mode of this population and a robust measure of the variance of it distribution. The sd tuning parameter controls how far away from the mode the split point is set.
Create a plot of the results of the computation.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>borderQuant</td>
<td>Usually the instrument is set up in a way that the positive population is somewhere on the high end of the measurement range and the negative population is on the low end. This parameter allows to disregard populations with mean values in the extreme quantiles of the data range. It’s value should be in the range $[0, 1]$.</td>
</tr>
<tr>
<td>absolute</td>
<td>Logical controlling whether to classify a population (positive or negative) relative to the theoretical measurement range of the instrument or the actual range of the data. This can be set to TRUE if the alignment of the measurement range is not optimal and the bulk of the data is on one end of the theoretical range.</td>
</tr>
<tr>
<td>inBetween</td>
<td>Force the algorithm to put the separator in between two peaks. If there are more than two peaks, this argument is ignored.</td>
</tr>
<tr>
<td>reflexLine</td>
<td>Either NULL or a numeric of length 1. If NULL, this parameter is ignored. When it is set to a numeric, the minor sub-population (if any) below this reference line will be ignored while determining the separator between positive and negative.</td>
</tr>
<tr>
<td>rare</td>
<td>Either TRUE or FALSE, assumes that there is one major peak, and that the rare positive population is to the right of it. Uses a robust estimate of mean and variance to gate the positive cells.</td>
</tr>
<tr>
<td>bwFac</td>
<td>The bandwidth for smoothing the density estimate. User-tunable</td>
</tr>
<tr>
<td>sig</td>
<td>a value of c(NULL,&quot;L&quot;,&quot;R&quot;), when sig is not NULL, use the half (left or right) of signal to estimate the std and mean</td>
</tr>
<tr>
<td>peakNr</td>
<td>when peakNr is not NULL, drop the less significant peaks by their heights</td>
</tr>
</tbody>
</table>

**Details**

The algorithm first tries to identify high density regions in the data. If the input is a `flowSet`, density regions will be computed on the collapsed data, hence it should have been normalized before (see `warpSet` for one possible normalization technique). The high density regions are then classified as positive and negative populations, based on their mean value in the theoretical (or absolute if argument `absolute=TRUE`) measurement range. In case there are only two high-density regions the lower one is usually classified as the negative populations, however the heuristics in the algorithm will force the classification towards a positive population if the mean value is already very high. The `absolute` and `borderQuant` arguments can be used to control this behaviour. The split point between populations will be drawn at the value of minimum local density between the two populations, or, if the `alpha` argument is used, somewhere between the two populations where the value of alpha forces the point to be closer to the negative ($\theta \sim 0.5$) or closer to the positive population ($\theta \sim 0.5$).

If there is only a single high-density region, the algorithm will fall back to estimating the mode of the distribution (`hubers`) and a robust measure of it’s variance and, in combination with the `sd` argument, set the split point somewhere in the right or left tail, depending on the classification of the region.

For more than two populations, the algorithm will still classify each population into positive and negative and compute the split point between those clusteres, similar to the two population case.

**Value**

A numeric indicating the split point between positive and negative populations.
Author(s)
Florian Hahne

See Also
warpSet, rangeGate

Examples

data(GvHD)
dat <- GvHD[pData(GvHD)$Patient==10]
dat <- transform(dat, "FL4-H"=asinh('FL4-H'), "FL3-H"=asinh('FL3-H'))
d <- flowStats:::density1d(dat, "FL4-H", plot=TRUE)
if(require(flowViz))
densityplot(~"FL4-H", dat, reline=d)

## tweaking the location
flowStats:::density1d(dat, "FL4-H", plot=TRUE, alpha=0.8)

## only a single population
flowStats:::density1d(dat, "FL3-H", plot=TRUE)
flowStats:::density1d(dat, "FL3-H", plot=TRUE, sd=2)

---

**gaussNorm**  
*Per-channel normalization based on landmark registration*

**Description**

This function normalizes a set of flow cytometry data samples by identifying and aligning the high density regions (landmarks or peaks) for each channel. The data of each channel is shifted in such a way that the identified high density regions are moved to fixed locations called base landmarks.

**Usage**

gaussNorm (flowset, channel.names, max.lms=2, base.lms=NULL, peak.density.thr=0.05, peak.distance.thr=0.05, debug=FALSE, fname='')

**Arguments**

- **flowset**: A `flowSet`
- **channel.names**: A character vector of flow parameters in `flowset` to be normalized.
- **max.lms**: A numeric vector of the maximum number of base landmarks to be used for normalizing each channel. If it has only one value that will be used as the maximum number of base landmarks for all the channels.
- **base.lms**: A list of vector for each channel that contains the base landmarks for normalizing that channel. If not specified the base landmarks are computed from the set of extracted landmarks.
peak.density.thr
The peaks with density value less than "peak.density.thr times maximum peak
density" are discarded.

peak.distance.thr
The sequences of peaks that are located closer than "peak.distance.thr times
range of data" are identified. Then for each sequence only one peak (the one
with the highest intensity value) is used as a landmark. In other words no two
landmarks are located closer than "peak.distance.thr times range of data" to each
other.

default Logical. Forces the function to draw before and after normalization plots for
each sample. The plot of the i-th sample is stored in paste(fname, i) file.

fname The pre- and post-normalization plots of the i-th sample is stored in paste(fname, i)
file if debug is set to TRUE. If default value is used the plots are drawn on sepa-
rate X11 windows for each sample. In this case, the function waits for a user
input to draw the plots for the next sample.

Details
Normalization is archived in three phases: (i) identifying high-density regions (landmarks) for each
flowFrame in the flowSet for a single channel; (ii) computing the best matching between the
landmarks and a set of fixed reference landmarks for each channel called base landmarks; (iii)
manipulating the data of each channel in such a way that each landmark is moved to its matching
base landmark. Please note that this normalization is on a channel-by-channel basis. Multiple
channels are normalized in a loop.

Value
A list with items flowset: normalized flowSet. confidence: a confidence measure of the nor-
malization procedure.

Author(s)
Alireza Hadj Khodabakhshi

Examples

```
data(ITN)
dat <- transform(ITN, "CD4"=asinh(CD4), "CD3"=asinh(CD3), "CD8"=asinh(CD8))
lg <- lymphGate(dat, channels=c("CD3", "SSC"),
preselection="CD4", scale=1.5)
dat <- Subset(dat, lg$n2gate)
datr <- gaussNorm(dat, "CD8")$flowset
if(require(flowViz)){
d1 <- densityplot(~CD8, dat, main="original", filter=curv1Filter("CD8"))
d2 <- densityplot(~CD8, datr, main="normalized", filter=curv1Filter("CD8"))
plot(d1, split=c(1,1,2,1))
plot(d2, split=c(2,1,2,1), newpage=FALSE)
}
```
**gpaSet**

Multi-dimensional normalization of flow cytometry data

### Description

This function performs a multi-dimensional normalization of flow cytometry data (flowSets) using a generalized Procrustes analysis (GPA) method.

### Usage

```r
gpaSet(x, params, register="backgating", bgChannels=NULL, 
bg=NULL, rotation.only=TRUE, 
downweight.missingFeatures=FALSE, thres.sigma=2.5, 
show.workflow=FALSE, 
ask=names(dev.cur())!="pdf")
```

### Arguments

- `x` A flowSet.
- `params` A character vector of length 2 describing the channels of interest.
- `register` A character indicating the method to be used for identifying features. Available method only includes "backgating" at the point.
- `bgChannels` A character vector indicating the channels used for backgating. If NULL, backGating will find the appropriate backgating channels.
- `bg` A data frame as the returning value of the backGating function. If not NULL, gpaSet will skip the backGating process and use the given data frame to extract potential features.
- `rotation.only` Logical for coarsing a reflection matrix to a rotation matrix.
- `downweight.missingFeatures` Logical. If TRUE, the missing features, labeled as bogus features, are down-weighted to zero. See details.
- `thres.sigma` A numerical value indicating the threshold of where to cut the tree, e.g., as resulting from diana, into several clusters. It is default to 2.5 sigma of the distribution of the heights of the cluster points.
- `show.workflow` Logical. If TRUE, the workflow of gpaSet will be displayed.
- `ask` Logical. If TRUE, the display operates in interactive mode.

### Details

Normalization is achieved by first identifying features for each flowFrame in the flowSet for designated channels using backgating, subsequently labeling features, and finally aligning the features to a reference feature in the sense of minimizing the Frobenius norm of

\[ ||sFQ - \bar{F}||, \]

where \( s \) is a scalar, \( Q \) a rotational matrix, \( F \) the matrix of features, and \( \bar{F} \) the reference feature. Both \( s \) and \( Q \) are solved by using singular value decomposition (SVD).

Note that if feature \( F_{ij} \) is missing, it is given a bogus value as \( \bar{F}_{ij} \).
If `downweight.missingFeatures` is TRUE, the cost function becomes

$$||sW_0FQ - W_0\bar{F}||,$$

where the weighting function $W_0$ is zero if the corresponding feature is bogus.

**Value**

The normalized `flowSet` with "GPA" attribute.

**Author(s)**

C. J. Wong <cwon2@fhcrc.org>

**References**

in progress

**Examples**

```r
## Example 1: calling up gpaSet directly
data(ITN)
data(BackGating)

tl <- transformList(colnames(ITN)[3:7], asinh, transformationId="asinh")
dat <- transform(ITN, tl)

xy = c("FSC", "SSC")
bgChannels = c("CD8", "CD4", "CD3")

## bg <- flowStats:::backGating(dat, xy=xy, channels=bgChannels)
## using pre-generated backgating results: BackGating
s <- gpaSet(dat, params=xy, bgChannels=bgChannels, bg=BackGating)

if(require(flowViz)) {
  d1 <- densityplot(~., s, channels=c("FSC", "SSC"),
                   layout=c(2,1), main="After GPA using bg")
  d2 <- xyplot(FSC ~ SSC, as(s, "flowFrame"),
               channels=c("FSC", "SSC"), main="All flowFrames")
  plot(d1)
  plot(d2)
}

## view "GPA" attribute
attr(s, "GPA")
```

```r
## Not run:
## Example 2: using work flow and normalization objects
data(ITN)
ITN <- ITN[1:8, ]
wf <- workFlow(ITN)

if(require(flowViz)) {
  tl <- transformList(colnames(ITN)[3:7], asinh, transformationId="asinh")
  add(wf, tl)

  x <- Data(wf[['asinh']])
  norm <- normalize(normFun=function(x, parameters, ...)
                     gpaSet(x, parameters, ...),
                  parameters = c("FSC", "SSC"),
```
idFeaturesByBackgating

(Internal use only) Identify features of flow cytometry data using backgating

Description

Identify and labeling significant features using divisive clustering method such as diana.

Usage

idFeaturesByBackgating(bg, nDim, thres.sigma=2.5, lambda=0.1,
reference.method="median",
plot.workflow=FALSE, ask=names(dev.cur())!="pdf")

Arguments

bg A data frame containing subpopulations on channels of interests. Must be a returning result from flowStats:::backGating
nDim An integer indicating the length of channels of interest.
thres.sigma An numerical value indicating the threshold at which to cut tree, e.g., as resulting from 'diana', into several clusters.
lambda A numerical value indicating the percentage of the potential features that is used as a threshold for deciding outlier clusters. The default value is 0.1.
reference.method A character vector indicating the method for computing the reference features. If median, the reference feature is defined by the median of each cluster of features. Valid methods include median and mean only.
plot.workflow Logical. If TRUE, display the workflow of feature identification.
ask Logical. If TRUE, the display operates in interactive mode.
iProcrustes

Details

Using the resulting data frame from backGating as potential features, the algorithm follows four major steps: (i) centering the potential features, which yields the returning value TransMatrix, (ii) using diana to compute a clustering of the potential features, (iii) cutting the tree into several clusters, and (iv) accessing outliers and rendering the final registered features with labels.

In step three, the threshold for cutting the tree is computed by

\[ sd \times \text{thres}.\sigma, \]

where \( sd \) is the standard deviation of the distribution of the height between entities computed by diana.

A cluster is determined as an outlier if the number of its members is less than the median of the numbers of all clusters’ members times \( \lambda \).

Value

register  
A list containing registered features for each sample.

Author(s)

Chao-Jen Wong

See Also

diana, BackGating, gpaSet.

Examples

## Not run:
data(ITN)
wfl <- workFlow(ITN)
tl <- transformList(colnames(ITN)[3:7], asinh, transformationId="asinh")
dat <- transformList(ITN, tl)
bg <- backGating(dat, xy=c("FSC", "SSC"), channels="CD3")
## End(Not run)
data(BackGating)
results <- flowStats::idFeaturesByBackgating(bg=BackGating, nDim=2,
plot.workflow=TRUE, ask=TRUE)

iProcrustes

Procrustes analysis. Using singular value decomposition (SVD) to determine a linear transformation to align the points in X to the points in a reference matrix Y.

Description

Based on generalized Procrustes analysis, this function determines a linear transformation (rotation/reflection and scaling) of the points in matrix x to align them to their reference points in matrix xbar. The alignment is carried out by minimizing the distance between the points in x and xbar.
**Usage**

```r
iProcrustes(x, xbar, rotation.only=TRUE, scalling=TRUE, translate=FALSE)
```

**Arguments**

- `x`: A numerical matrix to be align to points in `xbar`, the second argument. The columns represents the coordinates of the points. The matrices `x` and `xbar` must have the same dimensions.

- `xbar`: A numerical, reference matrix to which points in matrix `x` are to be aligned.

- `rotation.only`: Logical. When `rotation.only` is `TRUE`, it allows the function to lose reflection component of the linear transformation. Although it might not give the best-fitting alignment, when dealing with flow cytometry data alignment, a non-reflection transformation is preferred. When `rotation.only` is `FALSE`, it allows the function to retain the reflection component.

- `scalling`: Logical. When `scalling` is `FALSE`, it allows the function to calculate the linear transformation without a scalling factor. That is, the returning scalling factor is set to 1.

- `translate`: Logical. Set `translate` to `FALSE` when the points in matrices `x` and `xbar` are already centralized prior to applying this function. When `translate` is `TRUE`, it allows the function to translate the centroid the points in matrix `x` to that of points in `xbar`.

**Details**

Suppose the points in matrix $X$ and $\bar{X}$ are centralized (meaning their centroids are at the origin). The linear transformation of $X$ for aligning $X$ to its reference matrix $\bar{X}$, i.e., $\min ||sXQ - \bar{X}||_F$, is given by:

$$Q = VU^T,$$

and

$$s = \text{trace}(\bar{X}^T X Q)/\text{trace}(X^T X),$$

where $V$ and $U$ are the singular value vectors of $\bar{X}^T X$ (that is, $\bar{X}^T X = U\Sigma V^T$), and $s$ is the scalling factor.

**Value**

A list of the linear transformation with items

- `Q`: An orthogonal, rotation/reflection matrix.

- `scal`: A scalling factor

- `T` (optional): A translation vector used to shift the centroid of the points in matrix `x` to the origin. Returned when `translate` is `TRUE`.

- `T.xbar` (optional): Centered `xbar` (that is, the centroid of the points in `xbar` is translated to the origin). Returned when `translate` is `TRUE`.

Note that the return values of this function do not include the transformed matrix `scal * x * Q` or `scal * (x - IT) * Q`, where $T$ is the translation vector and $I$ is an $n - by - 1$ vector with elements 1.
Author(s)
C. J. Wong <cwon2@fhcrc.org>

See Also

gpaSet

Examples

## Example 1

```r
x <- matrix(runif(20), nrow=10, ncol=2)+ 1.4
s <- matrix(c(cos(60), -sin(60), sin(60), cos(60)),
    nrow=2, ncol=2, byrow=TRUE)
xbar <- 2.2 *(x %*% s) - 0.1

lt <- iProcrustes(x, xbar, translate=TRUE) ## return linear transformation
lt

## showing result
I <- matrix(1, nrow=nrow(x), ncol=1)
tx <- x - I %*% lt$T
## get the transformed matrix xnew
xnew <- lt$scal * (tx %*% lt$Q)
if (require(lattice)) {
    xyplot(V1 ~ V2,
        do.call(make.groups, lapply(list(x=x, xbar=xbar, T.xbar=lt$T.xbar,
            xnew=xnew),as.data.frame)),
        group=which, aspect=c(0.7), pch=c(1,3,2,4), col.symbol="black",
        main="Align the points in x to xbar",
        key=list(points=list(pch=c(1,3,2,4), col="black"), space="right",
            text=list(c("x", "xbar", "T.xbar", "xnew"))))
}
```

## Example 2. centralized x and xbar prior to using iProcrustes

```r
x <- matrix(runif(10), nrow=5, ncol=2)
s <- matrix(c(cos(60), -sin(60), sin(60), cos(60)),
    nrow=2, ncol=2, byrow=TRUE)
xbar <- 1.2 *(x %*% s) - 2
I <- matrix(1, nrow=nrow(x), ncol=1)
x <- x-(I %*% colMeans(x)) ## shift the centroid of points in x to the origin
xbar <- xbar - (I %*% colMeans(xbar)) ## shift centroid to the origin
lt <- iProcrustes(x, xbar, translate=FALSE) ## return linear transformation
## only return the rotation/reflection matrix and scalling factor
lt

xnew=lt$scal *(x %*% lt$Q) ## transformed matrix aligned to centralized xbar
if (require(lattice)) {
    xyplot(V1 ~ V2,
        do.call(make.groups, lapply(list(x=x, xbar=xbar,
            xnew=xnew), as.data.frame)),
        group=which, auto.key=list(space="right"))
}
```
Sample flow cytometry data

Description

A flowSet containing data from 15 patients.

Usage

data(ITN)

Format

A flowSet containing 15 flowFrames. There are 3 patient groups with 5 samples each.

Source

Immune Tolerance Network

landmarkMatrix

Compute and cluster high density regions in 1D

Description

This functions first identifies high-density regions for each flowFrame in a flowSet and subsequently tries to cluster these regions, yielding the landmarks matrix that needs to be supplied to landmarkreg. The function is considered to be internal.

Usage

landmarkMatrix(data, fres, parm, border=0.05, peakNr=NULL, densities = NULL, n = 201, indices=FALSE)

Arguments

data A flowSet.
fres A list of filterResultList objects generated by a filtering operation using a curv1Filter. Each list item represents the results for one of the flow parameters in parm.
parm Character scalar of flow parameter to compute landmarks for.
border A numeric in [0,1]. Ignore all high-density regions with mean values in the extreme percentiles of the data range.
peakNr Force a fixed number of peaks.
densities An optional matrix of y values of the density estimates for the flowSet. If this is not present, density estimates will be calculated by the function.
n Number of bins used for the density estimation.
indices Return matrix of population indices instead of landmark locations. These indices can be used to point into the populations identified by the curv1Filter.
Details

In order to normalize the data using the `landmarkreg` function in the `fda`, a set of landmarks has to be computed for each `flowFrame` in a `flowSet`. The number of landmarks has to be the same for each frame. This function identifies high-density regions in each frame, computes a simple clustering and returns a matrix of landmark locations. Missing landmarks of individual frames are substituted by the mean landmark location of the respective cluster.

Value

A matrix of landmark locations. Columns are landmarks and rows are `flowFrames`.

Author(s)

Florian Hahne

See Also

`landmarkreg`, `warpSet`

Examples

data(GvHD)
tmp <- list("FSC-H"=filter(GvHD[1:3], curv1Filter("FSC-H")))
res <- flowStats:::landmarkMatrix(GvHD[1:3], tmp, "FSC-H")

lymphFilter-class

Automated gating of elliptical cell populations in 2D.

Description

Cell populations of roughly elliptical shape in two-dimensional projections are of huge interest in many flow cytometry applications. This function identifies a single such population, potentially from a mixture of multiple populations.

Usage

`lymphGate(x, channels, preselection=NULL, scale=2.5, bwFac=1.3, filterId="defaultLymphGate", evaluate=TRUE, plot=FALSE, ...)`

Arguments

- `x`: An object of class `flowSet`.
- `channels`: A character vector of length 2 of valid flow parameters in `x`.
- `preselection`: Either `NULL`, in which case this boils down to fitting a regular `norm2Filter`, a character scalar giving one of the flow parameters in `x`, or a named list of numerics specifying the initial rough preselection. The latter gets passed on to `rectangleGate`, see its documentation for details.
- `scale`: The `scaleFactor` parameter that gets passed on to `norm2Filter`.
- `bwFac`: The bandwidth factor that gets passed on to `curv1Filter`. 
**lymphFilter-class**

- **filterId**: A character used as filterId.
- **evaluate**: A logical indicating whether the filter should be resolved (computation of the `filterResult` and the subset).
- **plot**: Logical. Produce plots of filter results

**Details**

This algorithm does not apply real mixture modelling, however it is able to identify a single elliptical cell population from a mixture of multiple such populations. The idea is to first define a rough rectangular preselection and, in a second step, fit a bivariate normal distribution to this subset only. Depending on the value of `preselection`, the initial rough selection is either

**NULL**: No preselection at all

**character scalar**: Preselection based on cells that are positive for a single marker only. This allows for back-gating, for instances by selecting CD4+ T-cells and using this information to back-gate lymphocytes in FSC and SSC. Positive cells are identified using a `curv1Filter`.

**a named list of numerics**: Preselection by a rectangular gate. The items of the list have to be numerics of length one giving the gate boundaries in the respective dimensions.

**Value**

A list with items

- **x**: The filtered `flowSet`.
- **n2gate**: The `norm2Filter` object.
- **n2gateResults**: The `filterResult` after applying the `norm2Filter` on the `flowSet`.

for the `lymphGate` function. Note that `x` and `n2gateResults` are NULL when `eval=False`.

**Extends**

- Class `parameterFilter`, directly.
- Class `concreteFilter`, by class "parameterFilter", distance 2.
- Class `filter`, by class "parameterFilter", distance 3.

**Slots**

See Arguments section for details.

- **preselection**: Object of class `character`, the name of the flow parameter used for preselection.
- **rectDef**: Object of class `list`, the initial rectangular selection.
- **scale**: Object of class `numeric`.
- **bwFac**: Object of class `numeric`.
- **parameters**: Object of class `parameters`, the flow parameters to operate on.
- **filterId**: Object of class "character", the filter identifier.

**Objects from the Class**

Objects can be created by calls of the form `new("lymphFilter", parameters, ...)` or using the constructor `lymphFilter`. The constructor is the recommended way of object instantiation.
normalize-methods

Author(s)

Florian Hahne

See Also

norm2Filter, curv1Filter

Examples

data(GvHD)
dat <- GvHD[pData(GvHD)$Patient==10]
dat <- transform(dat, "FL4-H"=asinh(Var("FL4-H")))
lg <- lymphGate(dat, channels=c("FSC-H", "SSC-H"), preselection="FL4-H", scale=1.5)

if(require(flowViz))
  xyplot('SSC-H~FSC-H', dat, filter=lg$n2gate)

normalize-methods

normalize a GatingSet imported with flowWorkspace, using sequential normalization on the manual gates in the GatingHierarchy.

Description

The method will step through the gating hierarchy in a breadth first search manner and normalize each dimension and gate not explicitly excluded in skipdims, or skipgates. The normalization approach is based on warpSet, but uses sequential normalization to alternately normalize then perform gating of the cell populations. This often helps with feature registration of populations lower in the gating hierarchy. FSC and SSC, as well as time are generally excluded by default. The rule of thumb, is to only normalize a channel in a gate if it is absolutely warranted.

Usage

normalize(data,x,...)

Arguments

data  The GatingSet to be normalized.
x     missing. Not used in here.
...   Arguments passed to downstream functions.
target: The target sample to normalize the other samples in the gating set to. A character vector. Must match a sample name in x, otherwise NULL will use the mean (average) of the peaks identified in all samples
populations: A character vector of population names that are to be normalized.
dims: A character vector of parameter names to be normalized.
chunksize: For a memory-efficient implementation of normalization, set the chunksize (an integer), which will perform normalization on chunks of the data of size chunksize.
nPeaks: A list of integer or an integer vector that specifies the expected number of peaks for each sample. Can be omitted to keep all peaks.

bwFac: The bandwidth for density estimation, a numeric. Affects the sensitivity for smoothing and detecting distinct peaks.

Details
This function implements sequential normalization using a GatingSet and a set of manual gates. For each gate in the gating hierarchy, the algorithm checks if the gate should be normalized, and which dimensions in the gate should be normalized. If normalization is warranted, this is performed prior to gating. After gating, the counts for the gate in the GatingSet are updated, and the next gate is processed. This is useful in the application of template gates to data that has staining variability in one or more channels.

Value
Returns a GatingSet of normalized data.

Author(s)
Greg Finak <gfinak@fhcrc.org>

See Also
See also GatingSet-class, GatingHierarchy-class, ncdfflowSet

Examples
## Not run:
#gs is a GatingSet
gs_norm <- normalize(gs,
  target = "M+T panel_903997-25.fcs"
  , populations = "cd27gate"
  , dims = "<Violet A 610/20-A>"
  , minCountThreshold = 100
  , nPeaks = list("cd27gate" = 2)
  , chunksize = 10
  , bwFac = 2
)
#show the population statistics for before and after normalization
getPopStats(gs)
getPopStats(gs_norm)

#plot the gate to see the effects of normalization
grid.arrange(
  plotGate(gs, "cd27gate", type = "densityplot", stack = T)
  , plotGate(gs_norm, "cd27gate", type = "densityplot", stack = T)
)
## End(Not run)
Description

Create QA plots for a flow cytometry normalization process.

Usage

```r
normQA(data, morph = c("^fsc", "^ssc"), channels, odat = NULL, ask = names(dev.cur()) != "pdf", grouping = NULL, tag.outliers = FALSE, peaksOnly = TRUE)
```

Arguments

data: a normalized `flowSet`.
morph: A character vector of channel names to use for the backgating into the morphological channels.
channels: The channels for which to create plots. Defaults to all normalized channels.
mdat: The original data set, always needed if there are no warping functions available.
ask: Ask before creating a new plot.
grouping: A grouping variable in `data`'s `phenoData` slot.
tag.outliers: Logical. Add sample name to outliers in the plots.
peaksOnly: Logical. Only use data when a peak was detected in a particular sample. If set to FALSE, an average peak location is estimated.

Details

This function assumes that the necessary information has been added as attributes to `data` during the normalization procedure. Depending on the available information, a set of QA plots is generated. Available plots are:

- Amount of peak adjustment
- Warping functions
- Landmark classification confidence
- Backgating of peak events in morphological channels

Value

This function is called for its side effect of generating plots.

Author(s)

Florian Hahne
**plotBins**

Plots the probability bins overlaid with flowFrame data

**Description**

This function is useful in visualizing the differences between the binned control and sample datasets. The bins generated from the control dataset are overlaid with the sample dataset. An optional argument residuals can be used to shade each bin based on a calculated statistical measure of difference between the number of events in each bin.

**Usage**

```r
gplotBms(binRes, data, channels, title, residuals, shadeFactor)
```

**Arguments**

- **binRes**: The result generated by calling the `probBin` function on a control dataset.
- **data**: An object of class `flowFrame` `sample(dataset)`
- **channels**: The flow parameters to be plotted. In cases where more than two parameters are binned from the control set, the `plotBins` function plots the projections of the hyperplanes in 2 dimensions.
- **title**: Optional title for the plot generated
- **residuals**: A vector of length equal to the number of bins generated that can be used to shade each bin. The residuals from the `calcPearsonChi` function or the `calcPBChiSquare` function can be used to highlight the bins that are different between control and sample datasets.
- **shadeFactor**: Optional argument between 0 and 1 that controls the intensity of the shading of bins.

**Author(s)**

Nishant Gopalakrishnan

**See Also**

- `probBin`, `calcPearsonChi`, `calcPBChiSquare`

**Examples**

```r
data(GvHD)
# flow frame 1 is treated as control dataset and used to generate bins
resCtrl<-probBin(GvHD[[1]],200,channels=c("FSC-H","SSC-H"))
plotBms(resCtrl,GvHD[[1]],channels=c("FSC-H","SSC-H"),title="Binned control data")
# Same bins are applied to flowFrame 16
resSample<-binByRef(resCtrl,GvHD[[16]])
stat<-calcPearsonChi(resCtrl,resSample)
dev.new()
plotBms(resCtrl,data=GvHD[[16]],channels=c("FSC-H","SSC-H","Time"),title="Comparision 1 & 16",
        residuals=stat$residuals[2,],shadeFactor=0.7)
```
**Description**

This function divides the flowframe events into bins such that each bin contains the same number of events. The number of events falling into each bin can then be compared across the control and test samples using statistical methods such as the Chi-squared test.

**Usage**

```r
proBin(m, minEvents=500, channels=NULL)
```

**Arguments**

- `m`: An object of class `flowFrame`
- `minEvents`: The minimum number of events in each bin. (i.e. the termination criterion for the probability binning algorithm)
- `channels`: A character vector for the fluorescence channels on which probability binning is to be performed. Defaults is `NULL`, in which case, all fluorescence channels are used for probability binning. (Time information, if provided in the flowFrame is discarded)

**Details**

The `flowSet` is first filtered using a `rectangleGate` and the `norm2Filter` is subsequently fitted to the remaining subset.

**Value**

A list with items:

- `table`: A data.frame that stores information regarding each node of the tree generated during each stage of the probability binning algorithm. Each row in the table represents a node, the first row representing the original `flowFrame` matrix. The `dataIndx` column provides indexes for retrieving the matrices during each stage of the binning process from the environment `data`. The parent field indicates the row number in the table that holds the parent information for the corresponding node. The left and right columns indicates the row numbers in the table that stores information regarding the children of that particular node. The leaf nodes that hold the binned data can be identified by the nodes with the left of right values of zero (i.e. no children nodes) The visited column is used internally by the algorithm to check if a particular node has been visited during the computation process.

- `data`: An environment in which the matrices generated during each stage of the probability binning process is stored. The matrices stored at the leaf nodes represent the binned events obtained after the stop criterion of `minEvents` has been achieved. These can be identified by the corresponding `dataIndx` fields provided by the rows in the table with the left or right column values of zero.
### quadrantGate

**limits**  
A list containing the boundaries of each hyperplane generated during probability binning.

**splitPars**  
A `data.frame` containing two columns `splitCol` - indicates the column number of the `flowFrame`, the split was performed.  
`splitMed` - The median value which was used as the threshold for splitting the `flowFrame`  
The `splitCol` and `splitMed` parameters are utilized by the `plotBins` and `shadeBins` functions in visualizing the differences between control and test sample cases.

**Author(s)**  
Nishant Gopalakrishnan

**See Also**  
`plotBins`, `binByRef`

**Examples**

```r
data(GvHD)
res<-proBin(GvHD[[1]],200,channels=c("FSC-H","SSC-H","FL1-H","FL4-H"))
```

---

### quadrantGate  
Automated quad gating

**Description**

This function tries to find the most likely separation of two-dimensional flow cytometry in four quadrants.

**Usage**

```r
quadrantGate(x, stains, alpha=c("min", "min"), sd=c(2, 2), plot=FALSE,
filterId="defaultQuadGate", refLine.1=NULL, refLine.2=NULL
,rare=c(FALSE,FALSE)
,sig=c(NULL,NULL)
,...)
```

**Arguments**

- **x**  
  A `flowSet` or `flowFrame`.

- **stains**  
  A character vector of length two giving the two flow parameters for which the quad gate is to be computed.

- **alpha, sd**  
  Tuning factors to control the computation of the gate boundaries. See `rangeGate` for details.

- **plot**  
  Logical. Produce plots of intermediate results.

- **filterId**  
  Character, the name assigned to the resulting filter.
refLine.1 Either NULL or a numeric of length 1. If NULL, this parameter is ignored. When it is set to a numeric, the minor sub-population (if any) below this reference line in the first stain channel will be ignored while determining the separator between positive and negative.

refLine.2 Either NULL or a numeric of length 1. If NULL, this parameter is ignored. When it is set to a numeric, the minor sub-population (if any) below this reference line in the second stain channel will be ignored while determining the separator between positive and negative.

rare logical flags for two channels. Refer to density1d for more details.

sig parameters for two channels. Refer to density1d for more details.

... Additional arguments

Details

The most likely separation between positive and negative stains for two-dimensional data is computed based on density estimates. Essentially, the gate parameters are first fitted separately for the two parameters and later combined. See the documentation for rangeGate for details. There is a certain amount of heuristics involved in this process. The algorithm can be slightly tweaked using the alpha and sd arguments. Their values will be recycled for the two dimensions unless explicitly given as vectors of length 2.

Value

An object of class quadGate.

Author(s)

Florian Hahne

See Also

quadGate, rangeGate

Examples

```r
## Not run:
data(GvHD)
dat <- GvHD[pData(GvHD)$Patient==10]
dat <- transform(dat, "FL4-H"=asinh("FL4-H"), "FL2-H"=asinh("FL2-H"))
qg <- quadrantGate(dat, c("FL2-H", "FL4-H"))
qg
if(require(flowViz))
xyplot("FL2-H"~"FL4-H", dat, filter=qg)
qg <- quadrantGate(dat, c("FL2-H", "FL4-H"), alpha=c(0.1, 0.9), plot=TRUE)
qg
split(dat, qg)
## End(Not run)
```
rangeGate

Find most likely separation between positive and negative populations in 1D

Description

The function tries to find a reasonable split point between the two hypothetical cell populations "positive" and "negative".

Usage

rangeGate(x, stain, alpha="min", sd=2, plot=FALSE, borderQuant=0.1, absolute=TRUE, filterId="defaultRectangleGate", positive=TRUE, reflLine=NULL, simple = FALSE,...)

rangeFilter(stain, alpha="min", sd=2, borderQuant=0.1, filterId="defaultRangeFilter")

Arguments

x A flowSet or flowFrame.
stain A character scalar giving the flow parameter for which to compute the separation.
alpha A tuning parameter that controls the location of the split point between the two populations. This has to be a numeric in the range [0,1], where values closer to 0 will shift the split point closer to the negative population and values closer to 1 will shift towards the positive population. Additionally, the value of alpha can be "min", in which case the split point will be selected as the area of lowest local density between the two populations.
sd For the case where there is only a single population, the algorithm falls back to estimating the mode of this population and a robust measure of the variance of it distribution. The sd tuning parameter controls how far away from the mode the split point is set.
plot Create a plot of the results of the computation.
borderQuant Usually the instrument is set up in a way that the positive population is somewhere on the high end of the measurement range and the negative population is on the low end. This parameter allows to disregard populations with mean values in the extreme quantiles of the data range. It’s value should be in the range [0,1].
absolute Logical controlling whether to classify a population (positive or negative) relative to the theoretical measurement range of the instrument or the actual range of the data. This can be set to TRUE if the alignment of the measurement range is not optimal and the bulk of the data is on one end of the theoretical range.
filterId Character, the name assigned to the resulting filter.
positive Create a range gate that includes the positive (TRUE) or the negative (FALSE) population.
reflLine Either NULL or a numeric of length 1. If NULL, this parameter is ignored. When it is set to a numeric, the minor sub-population (if any) below this reference line will be ignored while determining the separator between positive and negative.
rangeGate

simple logical scalar indicating whether to use a simple peak finding version of density1d algorithm.

Details

The algorithm first tries to identify high density regions in the data. If the input is a flowSet, density regions will be computed on the collapsed data, hence it should have been normalized before (see warpSet for one possible normalization technique). The high density regions are then classified as positive and negative populations, based on their mean value in the theoretical (or absolute if argument absolute=TRUE) measurement range. In case there are only two high-density regions the lower one is usually classified as the negative populations, however the heuristics in the algorithm will force the classification towards a positive population if the mean value is already very high. The absolute and borderQuant arguments can be used to control this behaviour. The split point between populations will be drawn at the value of minimum local density between the two populations, or, if the alpha argument is used, somewhere between the two populations where the value of alpha forces the point to be closer to the negative (0 - 0.5) or closer to the positive population (0.5 - 1).

If there is only a single high-density region, the algorithm will fall back to estimating the mode of the distribution (hubers) and a robust measure of its variance and, in combination with the sd argument, set the split point somewhere in the right or left tail, depending on the classification of the region.

For more than two populations, the algorithm will still classify each population into positive and negative and compute the split point between those clusters, similar to the two population case.

The rangeFilter class and constructor provide the means to treat rangeGate as regular flowCore filters.

Value

A range gate, more explicitly an object of class rectangleGate.

Methods

\%in\% signature(x = "flowFrame", table = "rangeFilter"): the work horse for doing the actual filtering. Internally, this simply calls the rangeGate function.

Author(s)

Florian Hahne, Kyongryun Lee

See Also

warpSet, rangeGate, rectangleGate

Examples

data(GvHD)
dat <- GvHD[pData(GvHD)$Patient==10]
dat <- transform(dat, "FL4-H"=asinh(FL4-H), "FL3-H"=asinh(FL3-H))
rg <- rangeGate(dat, "FL4-H", plot=TRUE)
rg
split(dat, rg)
## Test rangeGate when setting refLine=0; it does not do anything since there is no sub-population below zero.

```r
rangeGate(dat, "FL4-H", plot=FALSE, refLine=0)
```

```r
rf <- rangeFilter("FL4-H")
filter(dat, rf)
```

---

### singletGate

Creates a singlet polygon gate using the prediction bands from a robust linear model.

#### Description

The function constructs a singlet polygonGate using the prediction bands from a robust linear model trained on the area and height of one flow parameter (usually forward scatter).

#### Usage

```r
singletGate(x, area = "FSC-A", height = "FSC-H", sidescatter = NULL, prediction_level = 0.99, subsample_pct = NULL, wider_gate = FALSE, filterId = "singlet", maxit = 5, ...)
```

#### Arguments

- `x`: a `flowFrame` object
- `area`: character giving the channel name that records the signal intensity as peak area
- `height`: character giving the channel name that records the signal intensity as peak height
- `sidescatter`: character giving an optional channel name for the sidescatter signal. By default, ignored.
- `prediction_level`: a numeric value between 0 and 1 specifying the level to use for the prediction bands
- `subsample_pct`: a numeric value between 0 and 1 indicating the percentage of observations that should be randomly selected from `x` to construct the gate. By default, no subsampling is performed.
- `wider_gate`: logical value. If TRUE, the prediction bands used to construct the singlet gate use the robust fitted weights, which increase prediction uncertainty, especially for large FSC-A. This leads to wider gates, which are sometimes desired.
- `filterId`: the name for the filter that is returned
- `maxit`: the limit on the number of IWLS iterations passed to `rlm`
- `...`: additional arguments passed to `rlm`
Details

We construct a singlet gate by applying a robust linear model with rlm. By default, we model the forward-scatter height (FSC-H) as a function of forward-scatter area (FSC-A). If sidescatter is given, forward-scatter height is as a function of area + sidescatter + sidescatter / area.

Because rlm relies on iteratively reweighted least squares (IRLS), the runtime to construct a singlet gate is dependent in part on the number of observations in x. To improve the runtime, we provide an option to subsample randomly a subset of x. A percentage of observations to subsample can be given in subsample_pct. By default, no subsampling is applied.

Value

polygonGate.

Author(s)

Greg Finak, John Ramey

See Also

rangeGate, polygonGate

Examples

## Not run:
# fr is a flowFrame
sg <- singletGate(fr, area = "FSC-A", height = "FSC-H")
sg
# plot the gate
xyplot(FSC-H ~ FSC-A, fr, filter = sg)
## End(Not run)

warpSet

Normalization based on landmark registration

Description

This function will perform a normalization of flow cytometry data based on warping functions computed on high-density region landmarks for individual flow channels.

Usage

warpSet(x, stains, grouping = NULL, monwrd = TRUE, subsample=NULL, peakNr=NULL, clipRange=0.01, nbreaks=11, fres, bwFac=2, warpFuns=FALSE, target=NULL, ...)

warpSet
Arguments

x A flowSet.
stains A character vector of flow parameters in x to be normalized.
grouping A character indicating one of the phenotypic variables in the phenoData slot of x used as a grouping factor. The within-group and between-group variance is computed and a warning is issued in case the latter is bigger than the former, indicating the likely removal of signal by the normalization procedure.
monwrd Logical. Compute strictly monotone warping functions. This gets directly passed on to landmarkreg.
subsample Numeric. Reduce the number of events in each flowSet by sub sampling for all density estimation steps and the calculation of the warping functions. This can increase computation time for large data sets, however it might reduce the accuracy of the density estimates. To be used with care.
peakNr Numeric scalar. Force a fixed number of peaks to use for the normalization.
clipRange Only use peaks within a clipped data range. Essentially, the number indicates the percent of clipping on both sides of the data range, e.g. \( \min(x) - 0.01 \times \text{diff(range}(x)) \).
nbreaks The number of spline sections used to approximate the data. Higher values produce more accurate results, however this comes with the cost of increased computing times. For most data, the default setting is good enough.
fres A named list of filterResultList objects. This can be used to speed up the process since the curv1filter step can take quite some time.
bwFac Numeric of length 1 used to set the bandwidth factor by curv1filter for smoothing of the density estimate.
warpFuns Logical indicating whether to return the normalized flowSet or a list of warping functions.
target Character vector specifying the target sample to which other samples in the flowSet should be normalized. If NULL, then the mean of the peaks is used.
...

Details

Normalization is achieved by first identifying high-density regions (landmarks) for each flowFrame in the flowSet for a single channel and subsequently by computing warping functions for each flowFrame that best align these landmarks. This is based on the algorithm implemented in the landmarkreg function in the fda package. An intermediate step classifies the high-density regions, see landmarkMatrix for details.

Please note that this normalization is on a channel-by-channel basis. Multiple channels are normalized in a loop.

Value

The normalized flowSet if warpFuns is FALSE, otherwise a list of warping functions. Additional information is attached as the warping attribute to the flowSet in form of a list.

Note

We currently use a patched fda version.
Author(s)
Florian Hahne

References

See Also
curv1Filter, landmarkMatrix

Examples

data(ITN)
dat <- transform(ITN, "CD4"=asinh(CD4), "CD3"=asinh(CD3), "CD8"=asinh(CD8))
lg <- lymphGate(dat, channels=c("CD3", "SSC"), preselection="CD4", scale=1.5)
dat <- Subset(dat, lg$n2gate)
datr <- warpSet(dat, "CD8", grouping="GroupID", monwrd=TRUE)
if(require(flowViz)){
d1 <- densityplot(~CD8, dat, main="original", filter=curv1Filter("CD8"))
d2 <- densityplot(~CD8, datr, main="normalized", filter=curv1Filter("CD8"))
plot(d1, split=c(1,1,2,1))
plot(d2, split=c(2,1,2,1), newpage=FALSE)
}
Index

*Topic classes
  curv1Filter-class, 6
curv2Filter-class, 8
lymphFilter-class, 22

*Topic datasets
  BackGating, 3
  ITN, 21

*Topic methods
  curv1Filter-class, 6
curv2Filter-class, 8
normalize-methods, 24

*Topic misc
  autoGate, 2
  binByRef, 4
calcPBChiSquare, 4
calcPearsonChi, 5
idFeaturesByBackgating, 17
lymphFilter-class, 22
plotBins, 27
proBin, 28

*Topic package
  flowStats-package, 2
%in%,flowFrame,curv1Filter-method (curv1Filter-class), 6
%in%,flowFrame,curv2Filter-method (curv2Filter-class), 8
%in%,flowFrame,lymphFilter-method (lymphFilter-class), 22
%in%,flowFrame,rangeFilter-method (rangeGate), 31

autoGate, 2
BackGating, 3, 18
binByRef, 4, 29
calcPBChiSquare, 4, 5, 6, 27
calcPearsonChi, 5, 27
concreteFilter, 7, 9, 23
curv1Filter, 9, 10, 21, 22, 24, 35, 36
curv1Filter (curv1Filter-class), 6
curv1Filter-class, 6
curv2Filter, 7, 9
curv2Filter (curv2Filter-class), 8
curv2Filter-class, 8
density, 10
densityId, 11, 30
diana, 17, 18
fda, 22, 35
filter, 6–9, 23
filterResult, 3, 23
filterResultList, 21
flowFrame, 4, 7, 9–11, 14, 15, 21, 27–29, 31,
  33, 35
flowFrames, 21
flowSet, 3, 7, 9, 11, 13, 15, 21, 22, 26, 29, 31,
  35
flowStats (flowStats-package), 2
flowStats-package, 2
flowViz, 7, 9
gaussNorm, 13
gpaSet, 15, 18, 20

hubers, 12, 32
idFeatures (idFeaturesByBackgating), 17
idFeaturesByBackgating, 17
iProcrustes, 18
ITN, 21
landmarkMatrix, 11, 21, 35, 36
landmarkreg, 21, 22, 35
lymphFilter (lymphFilter-class), 22
lymphFilter-class, 22
lymphGate, 2, 3
lymphGate (lymphFilter-class), 22

multipleFilterResult, 6, 8, 10
ncdfFlowSet, 25
norm2Filter, 3, 22, 24
normalize (normalize-methods), 24
normalize, GatingSet, missing-method
  (normalize-methods), 24

37
normalize,GatingSetInternal,missing-method
(normalize-methods), 24
normalize-methods, 24
normQA, 26

oneDGate (rangeGate), 31

parameterFilter, 7, 9, 23
plotBins, 4, 27, 29
polygonGate, 34
proBin, 4–6, 27, 28

quadGate, 30
quadrantGate, 29

rangeFilter (rangeGate), 31
rangeFilter-class (rangeGate), 31
rangeGate, 11, 13, 29, 30, 31, 32, 34
rectangleGate, 3, 22, 32
rlm, 33, 34

show,curv1Filter-method
(curv1Filter-class), 6
show,curv2Filter-method
(curv2Filter-class), 8
singletGate, 33
split, 7, 9
summarizeFilter,multipleFilterResult,curv1Filter-method
(curv1Filter-class), 6
summarizeFilter,multipleFilterResult,curv2Filter-method
(curv2Filter-class), 8
warpSet, 12, 13, 22, 32, 34
warpSetGS (warpSet), 34
warpSetNCDF (warpSet), 34
warpSetNCDFLowMem (warpSet), 34