

Package ‘squallms’

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Description squallms is a Bioconductor R package that implements a ``semi-labeled" approach to untargeted mass spectrometry data. It pulls in raw data from mass-spec files to calculate several metrics that are then used to label MS features in bulk as high or low quality. These metrics of peak quality are then passed to a simple logistic model that produces a fully-labeled dataset suitable for downstream analysis.

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URL <https://github.com/wkumler/squallms>

BugReports <https://github.com/wkumler/squallms/issues>

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Contents

squallms-package	2
extractChromMetrics	3
labelFeatsLasso	4
labelFeatsManual	5
logModelFeatProb	7
logModelFeatQuality	8
makeXcmsObjFlat	9
pickyPCA	10
updateXcmsObjFeats	11

Index	13
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squallms-package	<i>squallms: Speedy quality assurance via lasso labeling for LC-MS data</i>
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Description

squallms is a Bioconductor R package that implements a "semi-labeled" approach to untargeted mass spectrometry data. It pulls in raw data from mass-spec files to calculate several metrics that are then used to label MS features in bulk as high or low quality. These metrics of peak quality are then passed to a simple logistic model that produces a fully-labeled dataset suitable for downstream analysis.

Details

squallms is an R package that allows for the identification and removal of low-quality chromatographic features from mass-spectrometry data. This process involves three steps: first, the calculation of peak quality metrics via `extractChromMetrics()`; second, labeling of high and low quality features either one at a time with `labelFeatsManual()` or in bulk with `labelFeatsLasso()`; and finally, probabilistic model construction using `logModelFeatQuality()`.

The package interfaces neatly with XCMS, providing two additional functions that turn an XCMS object into a flat file (`makeXcmsObjFlat()`) and edit the XCMS object's feature list to contain only high quality features (`updateXcmsObjFeats()`). Data can be provided without using XCMS if it is properly formatted (see the help pages).

See the package intro on GitHub at <https://github.com/wkumler/squallms> and explore the vignettes with `vignette("intro_to_squallms", package = "squallms")`

Author(s)

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See Also

Useful links:

- <https://github.com/wkumler/squallms>
- Report bugs at <https://github.com/wkumler/squallms/issues>

extractChromMetrics	<i>Extract metrics of chromatographic peak quality</i>
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Description

This function takes flat-form XC-MS data (i.e. the format produced by 'makeXcmsObjectFlat') and calculates metrics of peak shape and similarity for downstream processing (e.g. by [updateXcmsObjFeats](#)). The core metrics are those described in <https://doi.org/10.1186/s12859-023-05533-4>, corresponding to peak shape (correlation coefficient between the raw data and an idealized bell curve, `beta_cor`) and within-peak signal-to-noise (maximum intensity divided by the standard deviation of the residuals after the best-fitting bell is subtracted from the raw data). Additionally, the function interpolates the raw data to fixed retention time intervals and performs a principal components analysis (PCA) across a file-by-RT matrix which typically extracts good-looking peaks in the first component (PC1). These functions are calculated on the raw MS data as obtained via **RaMS**, so either the filepaths must be included in the `peak_data` object or supplied as `ms1_data`.

Usage

```
extractChromMetrics(  
  peak_data,  
  recalc_betas = FALSE,  
  ms1_data = NULL,  
  verbosity = 0  
)
```

Arguments

<code>peak_data</code>	Flat-form XC-MS data with columns for the bounding box of a chromatographic peak (<code>mzmin</code> , <code>mzmax</code> , <code>rtmin</code> , <code>rtmax</code>) as grouped by a feature ID. Must be provided WITHOUT retention time correction for proper matching to the values in the raw data.
<code>recalc_betas</code>	Scalar boolean controlling whether existing beta values (as calculated in the XCMS object when peakpicked with <code>CentWaveParam(verboseBetaColumns=TRUE)</code>) should be used as-is (the default) or recalculated. See https://github.com/sneumann/xcms/pull/685 for differences in these implementations.
<code>ms1_data</code>	Optional <code>data.table</code> object produced by RaMS containing MS1 data with columns for <code>filename</code> , <code>rt</code> , <code>mz</code> , and <code>int</code> . If not provided, the files are detected from the <code>filepath</code> column in <code>peak_data</code> .
<code>verbosity</code>	Scalar value between zero and two determining how much diagnostic information is produced. 0 should return nothing, 1 should return text-based progress markers, and 2 will return diagnostic plots if available.

Value

A `data.frame` containing one row for each feature in `peak_data`, with columns containing the median peak shape value (`med_cor`), and the median SNR value (`med_snr`).

Examples

```
library(xcms)
library(MSnbase)
library(dplyr)
mzML_files <- system.file("extdata", package = "RaMS") %>%
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiwrapParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files %>%
  readMSData(msLevel. = 1, mode = "onDisk") %>%
  findChromPeaks(cwp) %>%
  adjustRtime(obp) %>%
  groupChromPeaks(pdp) %>%
  fillChromPeaks(FillChromPeaksParam(ppm = 5))
peak_data <- makeXcmsObjFlat(xcms_filled)
feat_metrics <- extractChromMetrics(peak_data, recalc_betas = TRUE)
```

labelFeatsLasso

Label similar chromatographic features in bulk via interactive selection

Description

This function interpolates multi-file chromatograms to a shared set of retention time points then performs a PCA to place similar chromatograms near each other in a reduced dimensionality space. Features can then be labeled in groups instead of one at a time, massively reducing the burden of creating a high-quality dataset. This implementation relies on R's **shiny** package to provide interactive support in a browser environment and the **plotly** package for selection tools. Classified features are then returned as a simple R object for downstream use.

Usage

```
labelFeatsLasso(
  peak_data,
  ms1_data = NULL,
  rt_window_width = 1,
  ppm_window_width = 5,
  verbosity = 1
)
```

Arguments

peak_data	Flat-form XC-MS data with columns for the bounding box of a chromatographic peak (mzmin, mzmax, rtmin, rtmax) as grouped by a feature ID. Must be provided WITHOUT retention time correction for proper matching to the values in the raw data.
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ms1_data	Optional data.table object produced by RaMS containing MS1 data with columns for filename, rt, mz, and int. If not provided, the files are detected from the filepath column in peak_data.
rt_window_width	The width of the retention time window that should be used for PCA construction, in minutes.
ppm_window_width	The width of the m/z window that should be used for PCA construction, in parts per million.
verbosity	Scalar value between zero and two determining how much diagnostic information is produced. 0 should return nothing, 1 should return text-based progress markers, and 2 will return diagnostic plots if available.

Value

A character vector named with feature IDs containing the classifications of each peak that was viewed during the interactive phase. NA values indicate those features that were not classified.

Examples

```
library(xcms)
library(dplyr)
library(MSNbase)
mzML_files <- system.file("extdata", package = "RaMS") %>%
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiwrapParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files %>%
  readMSData(msLevel. = 1, mode = "onDisk") %>%
  findChromPeaks(cwp) %>%
  adjustRtime(obp) %>%
  groupChromPeaks(pdp) %>%
  fillChromPeaks(FillChromPeaksParam(ppm = 5))
peak_data <- makeXcmsObjFlat(xcms_filled)
if (interactive()) {
  lasso_labels <- labelFeatsLasso(peak_data)
}
```

Description

This function allows the user to view and label individual chromatographic features using the keyboard. Running labelFeatsManual launches a browser which shows a single feature extracted from multiple files. The feature can then be classified by pressing a key (defaults are left=bad,

right=good) on the keyboard which is recorded and a new feature is automatically shown. This implementation relies on R's **shiny** package to provide interactive support in a browser environment and the **keys** package for key binding within a browser. Classified features are then returned as a simple R object for downstream use.

Usage

```
labelFeatsManual(peak_data, ms1_data = NULL, verbosity = 1)
```

Arguments

peak_data	Flat-form XC-MS data with columns for the bounding box of a chromatographic peak (mzmin, mzmax, rtmin, rtmax) as grouped by a feature ID. Must be provided WITHOUT retention time correction for proper matching to the values in the raw data.
ms1_data	Optional data.table object produced by RaMS containing MS1 data with columns for filename, rt, mz, and int. If not provided, the files are detected from the filepath column in peak_data.
verbosity	Scalar value between zero and two determining how much diagnostic information is produced. 0 should return nothing while 1 will report diagnostic messages.

Value

A character vector named with feature IDs containing the classifications of each peak that was viewed during the interactive phase. NA values indicate those features that were not classified.

Examples

```
library(xcms)
library(dplyr)
library(MSNbase)
mzML_files <- system.file("extdata", package = "RaMS") %>%
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiwrapParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files %>%
  readMSData(msLevel. = 1, mode = "onDisk") %>%
  findChromPeaks(cwp) %>%
  adjustRtime(obp) %>%
  groupChromPeaks(pdp) %>%
  fillChromPeaks(FillChromPeaksParam(ppm = 5))
peak_data <- makeXcmsObjFlat(xcms_filled)
if (interactive()) {
  manual_labels <- labelFeatsManual(peak_data)
}
```

logModelFeatProb

*Model feature quality using a logistic regression***Description**

After metrics have been extracted (typically with ‘extractChromMetrics’) and labeling has occurred (typically with ‘labelFeatsManual’ or ‘labelFeatsLasso’), a model can be built to robustly classify the peaks that were not labeled based on the metrics. The default regression equation fits the med_cor and med_snr columns in feature_metrics to the feature_labels, but additional metrics can be passed as well (but be careful of overfitting!).

Usage

```
logModelFeatProb(
  feature_metrics,
  feature_labels,
  log_formula = feat_class ~ med_cor + med_snr,
  verbosity = 2
)
```

Arguments

feature_metrics	A data.frame with columns used to construct the feature quality model.
feature_labels	A character vector named with feature IDs and entries corresponding to the peak quality (either "Good", "Bad", or NA).
log_formula	The formula to use when predicting feature quality from the feature metrics. This formula is passed to ‘glm’ as-is, so make sure that the predictive features exist in the feature_metrics data.frame.
verbosity	Scalar value between zero and two determining how much diagnostic information is produced. 0 should return nothing, 1 should return text-based progress markers, and 2 will return diagnostic plots if available.

Value

A numeric vector of probabilities returned by the logistic model named by feature ID

Examples

```
library(xcms)
library(MSnbase)
mzML_files <- system.file("extdata", package = "RaMS") |>
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiwrapParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files |>
  readMSData(msLevel. = 1, mode = "onDisk") |>
```

```

    findChromPeaks(cwp) |>
    adjustRtime(obp) |>
    groupChromPeaks(pdp) |>
    fillChromPeaks(FillChromPeaksParam(ppm = 5))
peak_data <- makeXcmsObjFlat(xcms_filled)
feat_metrics <- extractChromMetrics(peak_data, verbosity = 0)

# Load demo labels previously assigned using the lasso method
lasso_classes <- readRDS(system.file("extdata", "intro_lasso_labels.rds", package = "squallms"))
feat_probs <- logModelFeatProb(feat_metrics, lasso_classes)

```

logModelFeatQuality *Turn 0-1 likelihood values into categorical (good/bad) classifications*

Description

This function wraps ‘logModelFeatProb’ for modeling of feature quality to estimate the quality of every feature in the dataset and classify them as good/bad based on whether they exceed the provided ‘likelihood_threshold’. Lower likelihood thresholds (0.01, 0.1) will produce more false positives (noise peaks included when they shouldn’t be). Higher likelihood thresholds (0.9, 0.99) will produce more false negatives (good peaks removed when they shouldn’t be).

Usage

```

logModelFeatQuality(
  feature_metrics,
  feature_labels,
  log_formula = feat_class ~ med_cor + med_snr,
  likelihood_threshold = 0.5,
  verbosity = 2
)

```

Arguments

feature_metrics	A data.frame with columns used to construct the feature quality model.
feature_labels	A character vector named with feature IDs and entries corresponding to the peak quality (either "Good", "Bad", or NA).
log_formula	The formula to use when predicting feature quality from the feature metrics. This formula is passed to ‘glm’ as-is, so make sure that the predictive features exist in the feature_metrics data.frame.
likelihood_threshold	A scalar numeric above which features will be kept if their predicted probability exceeds.
verbosity	Scalar value between zero and two determining how much diagnostic information is produced. 0 should return nothing, 1 should return text-based progress markers, and 2 will return diagnostic plots if available.

Value

A character vector of feature quality assessments returned by the logistic model and named by feature ID

Examples

```
library(xcms)
library(dplyr)
library(MSnbase)
mzML_files <- system.file("extdata", package = "RaMS") %>%
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiWarpParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files %>%
  readMSData(msLevel = 1, mode = "onDisk") %>%
  findChromPeaks(cwp) %>%
  adjustRtime(obp) %>%
  groupChromPeaks(pdp) %>%
  fillChromPeaks(FillChromPeaksParam(ppm = 5))
peak_data <- makeXcmsObjFlat(xcms_filled)
feat_metrics <- extractChromMetrics(peak_data, verbosity = 0)

# Load demo labels previously assigned using the lasso method
lasso_classes <- readRDS(system.file("extdata", "intro_lasso_labels.rds", package = "squallms"))
feat_classes <- logModelFeatQuality(feat_metrics, lasso_classes)
```

makeXcmsObjFlat	<i>Make an XCMS object flat</i>
-----------------	---------------------------------

Description

XCMSnExp objects are complicated S4 objects that make it difficult to access information about the features and their associated peaks. This function turns the output into a "flat" file format (a data.frame) so it can interact with tidyverse functions more easily.

Usage

```
makeXcmsObjFlat(xcms_obj, revert_rts = TRUE, verbosity = 0)
```

Arguments

xcms_obj	A XCMSnExp object produced by the typical XCMS workflow which should include retention time correction and peak correspondence and filling
revert_rts	Scalar boolean controlling whether the adjusted retention times found in the XCMS object are propagated or returned as-is
verbosity	Scalar numeric. Will report XCMS info if greater than zero or run silently if not (the default).

Value

A data.frame with columns for feature information (from featureDefinitions: feature, feat_mzmed, feat_rtmed, feat_npeaks, and peakidx) and peak information (from chromPeaks: mz, mzmin, mzmax, rt, rtmin, rtmax, into, intb, maxo, sn, sample) as well as the full path to the associated file and the file name alone (filepath and filename).

Examples

```
library(xcms)
library(dplyr)
library(MSnbase)
mzML_files <- system.file("extdata", package = "RaMS") %>%
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiwrapParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files %>%
  readMSData(msLevel = 1, mode = "onDisk") %>%
  findChromPeaks(cwp) %>%
  adjustRtime(obp) %>%
  groupChromPeaks(pdp) %>%
  fillChromPeaks(FillChromPeaksParam(ppm = 5))
makeXcmsObjFlat(xcms_filled)
```

pickyPCA

Perform a PCA on multi-file chromatographic data

Description

Internal function, mostly.

Usage

```
pickyPCA(
  peak_data,
  ms1_data,
  rt_window_width = NULL,
  ppm_window_width = NULL,
  verbosity = 1
)
```

Arguments

peak_data	Flat-form XC-MS data with columns for the bounding box of a chromatographic peak (mzmin, mzmax, rtmin, rtmax) as grouped by a feature ID. Must be provided WITHOUT retention time correction for proper matching to the values in the raw data.
ms1_data	Optional data.table object produced by RaMS containing MS1 data with columns for filename, rt, mz, and int. If not provided, the files are detected from the filepath column in peak_data.
rt_window_width	The width of the retention time window that should be used for PCA construction, in minutes.

ppm_window_width	The width of the m/z window that should be used for PCA construction, in parts per million.
verbosity	Scalar value between zero and two determining how much diagnostic information is produced. 0 should return nothing, 1 should return text-based progress markers, and 2 will return diagnostic plots if available.

Value

A list with two named components, `interp_df` and `pcamat`. `interp_df` is the MS1 data associated with each peak interpolated to a retention time spacing shared across the feature. `pcamat` is the result of cleaning this object up, pivoting it wider, and converting it to a matrix so that a PCA can be performed.

Examples

```
library(xcms)
library(dplyr)
library(MSnbase)
mzML_files <- system.file("extdata", package = "RaMS") %>%
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiWarpParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files %>%
  readMSData(msLevel. = 1, mode = "onDisk") %>%
  findChromPeaks(cwp) %>%
  adjustRtime(obp) %>%
  groupChromPeaks(pdp) %>%
  fillChromPeaks(FillChromPeaksParam(ppm = 5))
peak_data <- makeXcmsObjFlat(xcms_filled)
msdata <- RaMS::grabMSdata(unique(peak_data$filepath), grab_what = "MS1", verbosity = 0)
pixel_pca <- pickyPCA(peak_data, msdata$MS1)
```

updateXcmsObjFeats	<i>Update features in an XCMS object</i>
--------------------	--

Description

After metrics have been extracted (typically with `extractChromMetrics`) and labeling has occurred (typically with `labelFeatsManual` or `labelFeatsLasso`), low quality features can be removed from the XCMS object to improve downstream processing. This function wraps `logModelFeatQuality` and automatically edits the features within the provided XCMS object.

Usage

```
updateXcmsObjFeats(
  xcms_obj,
  feature_metrics,
```

```

    feature_labels,
    log_formula = feat_class ~ med_cor + med_snr,
    likelihood_threshold = 0.5,
    verbosity = 2
  )

```

Arguments

<code>xcms_obj</code>	The XCMS object from which features were initially extracted, usually with ‘extractChromMetrics’
<code>feature_metrics</code>	A data.frame with columns used to construct the feature quality model.
<code>feature_labels</code>	A character vector named with feature IDs and entries corresponding to the peak quality (either "Good", "Bad", or NA).
<code>log_formula</code>	The formula to use when predicting feature quality from the feature metrics. This formula is passed to ‘glm’ as-is, so make sure that the predictive features exist in the feature_metrics data.frame.
<code>likelihood_threshold</code>	A scalar numeric above which features will be kept if their predicted probability exceeds.
<code>verbosity</code>	Scalar value between zero and two determining how much diagnostic information is produced. 0 should return nothing, 1 should return text-based progress markers, and 2 will return diagnostic plots if available.

Value

The initial xcms_obj but only containing features that exceed the provided quality threshold (as established in likelihood space)

Examples

```

library(xcms)
library(dplyr)
library(MSnbase)
mzML_files <- system.file("extdata", package = "RaMS") %>%
  list.files(full.names = TRUE, pattern = "[A-F].mzML")
register(BPPARAM = SerialParam())
cwp <- CentWaveParam(snthresh = 0, extendLengthMSW = TRUE, integrate = 2)
obp <- ObiwrapParam(binSize = 0.1, response = 1, distFun = "cor_opt")
pdp <- PeakDensityParam(
  sampleGroups = 1:3, bw = 12, minFraction = 0,
  binSize = 0.001, minSamples = 0
)
xcms_filled <- mzML_files %>%
  readMSData(msLevel. = 1, mode = "onDisk") %>%
  findChromPeaks(cwp) %>%
  adjustRtime(obp) %>%
  groupChromPeaks(pdp) %>%
  fillChromPeaks(FillChromPeaksParam(ppm = 5))
peak_data <- makeXcmsObjFlat(xcms_filled)
feat_metrics <- extractChromMetrics(peak_data, verbosity = 0)
lasso_classes <- readRDS(system.file("extdata", "intro_lasso_labels.rds", package = "squallms"))
xcms_filled <- updateXcmsObjFeats(xcms_filled, feat_metrics, lasso_classes)

```

Index

* **internal**

squallms-package, [2](#)

extractChromMetrics, [3](#)

extractChromMetrics(), [2](#)

labelFeatsLasso, [4](#)

labelFeatsLasso(), [2](#)

labelFeatsManual, [5](#)

labelFeatsManual(), [2](#)

logModelFeatProb, [7](#)

logModelFeatQuality, [8](#)

logModelFeatQuality(), [2](#)

makeXcmsObjFlat, [9](#)

makeXcmsObjFlat(), [2](#)

pickyPCA, [10](#)

squallms (squallms-package), [2](#)

squallms-package, [2](#)

updateXcmsObjFeats, [3](#), [11](#)

updateXcmsObjFeats(), [2](#)