

## MStractor Workflow

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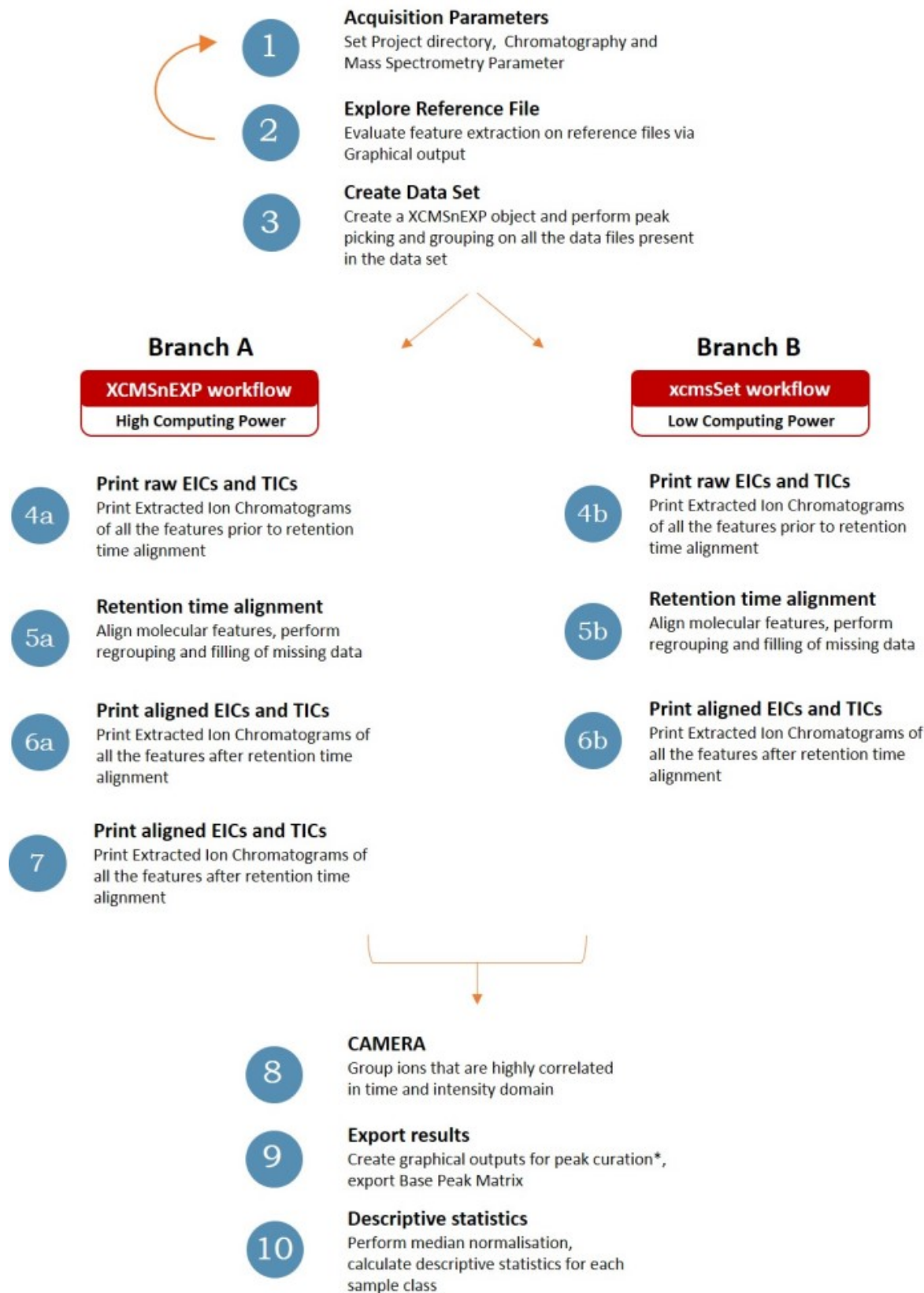
02/07/2020

```
#####
#
#          * *The Australian Wine Research Institute #
#          * * Metabolomics Australia (SA node)      #
#          **                                           #
#          *                                           #
#          *                                           #
#          * /_____| | | | | | | | | | | | | | | | | #
#          /_____| | | | | | | | | | | | | | | | | #
#          / 0.1.0 /_____| | | | | | | | | | | | | #
#          / THE AWRI /_____| | | | | | | | | | | | #
#          ]==/ \ \_____| | | | | | | | | | | | | #
#          \ \_____| | | | | | | | | | | | | | | | #
#          . | | . | . . . . | | * . . . . | . | | | . . : | . . . * . | : : . | | | . | | * : . | | | . | | . | | . #
#          #####
```

## 0.1 Introduction

**MStractor** combines the functions for molecular feature extraction present in XCMS and CAMERA with user friendly dedicated GUIs for LC and MS parameter input, graphical QC outputs and descriptive statistics calculations.

A schematic representation of the workflow is displayed in **Figure 1**, reported below. The workflow consists of 10 steps with the possibility of running the workflow using either the most recent xcms functions specific for *XCMSnExp* object (available from xcms version 1.51.5) (**branch a**) or the set of functions for *xcmsSet* objects (**branch b**). The 2 branches perform the same data processing steps and produce similar outputs, however, it is recommended to use most recent xcms functions. Branch a is for a high computing power environment, whereas branch b is suitable for low computing power environments. Specifically, the authors observed that step **6a** tends to be quite time-consuming in particular when dealing with complex data set. Therefore, in a case where computing power is limited and the data set is large, the authors suggest using branch b of the workflow. It is, however, recommended to follow branch a and use the most updated xcms functions. Step 8 is common to both branches, where the data set is processed using CAMERA. Since the latest release available for CAMERA does not support XCMSnEXP objects, the data set object is required to be converted to the xcmsSet format prior to this step.



\*Manual Curation Needed

**Figure 1 :** Schematic representation of MStractor Workflow

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Each step of the framework displayed in Figure 1 involves using 1 or more functions. MStractor functions can be classified into 3 different groups:

**1.MStractor specific functions:** developed by the authors to provide the user with a friendly and seamless experience. The functions belonging to this class mainly involve the use of GUIs allowing parameter input and storage to be used later in the workflow. This group also include functions to normalise data and calculate descriptive statistics.

**2.Wrapper functions:** includes wrapper functions of xcms and CAMERA. The function backbone is the xcms (or CAMERA) function and additional code is used to automate routine operations, reduce parameter input errors and produce graphical outputs within the selected working directory.

**3.xcms and CAMERA functions:** these functions are basic CAMERA and xcms functions that are used along the workflow. However, the user is not required to update the function arguments because all the necessary parameters are input in the previous steps of the workflow

## 0.2 *Data Preparation and File Naming*

Preparing the data and creating the correct working directory and folder structure (outside the R environment) is required prior to executing the workflow. The raw data files need to be converted to an appropriate format which can be easily achieved using various tools freely available, such as Proteowizard (ProteWizard, <http://proteowizard.sourceforge.net/>) and Massmatrix (MassMatrix, <http://www.massmatrix.net/>). MStractor supported file formats include mzDATA, mzML and mzXML. After conversion, files should be grouped in folders according to their class of belonging. Every class-folder needs to be stored within the directory ‘MSfiles’. The script will use the ‘MSfiles’ subfolder downstream to automatically determine the sample classes required for processing. If the folder ‘MSfiles’ is not present, the user will encounter errors. For each class-folder a minimum of 2 replicates is required. An example dataset (see Figure 2) is provided with the package.

```
path to project/MSfiles
./Treatment1
./Treatment1_R1.mzXML
./Treatment1_R2.mzXML
./Treatment1_R3.mzXML
./Mix
./Mix_R1.mzXML
./Mix_R2.mzXML
./Mix_R3.mzXML
```

**Figure 2 :** Project subfolder structure

File naming is a key aspect for correct visualisation of graphical output downstream. A specific function of the workflow assigns different colours and symbols to the different sample classes. In this respect, the filename has to correspond to the folder name followed by an underscore and a number indicating the biological replicate, as displayed in **Figure 2**.

## 0.3 *Installation*

To install the package from GitHub, make sure the package ‘remotes’ is installed and run the following,

```
library(remotes)

Sys.setenv(R_REMOTES_NO_ERRORS_FROM_WARNINGS="true")

remotes::install_github("MetabolomicsSA/MStractor")

# Alternatively, Download the tar.gz package from
# https://github.com/MetabolomicsSA/MStractor/releases
# and run the following

library(remotes)

Sys.setenv(R_REMOTES_NO_ERRORS_FROM_WARNINGS="true")

setRepositories(ind=1:2)

remotes::install_local("C:/pathtoPackage/MStractor_0.1.0.tar.gz",
  dependencies=NA)
```

After successful installation, load the MStractor library.

```
library(MStractor)
```

## 0.4 1. *Acquisition Parameters*

The first step of the workflow consists of running 6 functions that allow the input of chromatographic, mass spectrometry and peak picking parameters as well as loading the data and automatically defining the visualization settings for graphical QC outputs.

The default input values provided in the package functions are based on the settings of the acquisition instrument used by the authors and, therefore, suitable for processing the example data set available within the package. These values should be optimised for the user's LCMS system configuration, since they can dramatically influence the data processing outcome.

### 0.4.1 1.1 *Define Working Directory and Reference Files*

The function 'Project()' doesn't require arguments. Its execution opens a GUI (Figure 3) allowing the selection of the working directory and 2 reference files chosen from one of the folders within the MSfiles directory such as 2 pooled biological QC replicates (labeled as 'Mix' in the example data set). The function also creates a QC folder, that stores the graphical outputs to evaluate and QC input parameters at various steps of the workflow.

```
Project()
#If using the example dataset provided within the package set the
#working directory and click 'cancel' on the following 2 R-prompts asking
#to define reference files
```

use the following lines of code to load the example dataset:

```
path<-system.file("extdata",package = "MStractor")
files <- dir(path, pattern = ".mzXML", full.names = TRUE)
```



**Figure 3: GUI for project folder selection**

### 0.4.2 1.2 *Input of Chromatographic and Mass Spectrometry Parameters*

ChromParam() allows the user to input the chromatographic parameters related to the analytical data set to be processed and that will be stored and used in the later stages of the workflow. The values to be entered are the retention time range of the data set (rt start and rt end), the maximum retention time drift observed (rtDelta) and the minimum and maximum fullwidth at half maximum (FWHM min and max).

MassSpecParam() is mainly used to input mass spectrometry related parameters. The criteria to be entered are the acquisition mode (negative as default); the mass range to be considered (mz start and mz end); the number of expected charges (default value set at 3); the maximum number of chromatographic peaks (EICsMax) expected for a single EIC, the sensitivity and the file type (default set to 'mzXML') .

Values are entered using GUIs as displayed in Figure 4.

Both the functions return the input parameters.

For more details about the parameters consult the xcms manual: <https://www.bioconductor.org/packages/release/bioc/manuals/xcms/man/xcms.pdf> )

#leave default values if using the example dataset included in the package

ChromParam()

```
## [[1]]
## [1] 1
##
## [[2]]
## [1] "max"
##
## [[3]]
## [1] 10
##
## [[4]]
## [1] 90
##
## [[5]]
## [1] 32
```

MassSpecParam()

```
## [[1]]
## [1] "negative"
##
## [[2]]
## [1] 100
##
## [[3]]
## [1] 1650
##
## [[4]]
## [1] 0.01
##
## [[5]]
## [1] 3
##
## [[6]]
## [1] 30
##
## [[7]]
## [1] 0.7
##
## [[8]]
## [1] ".mzXML"
```

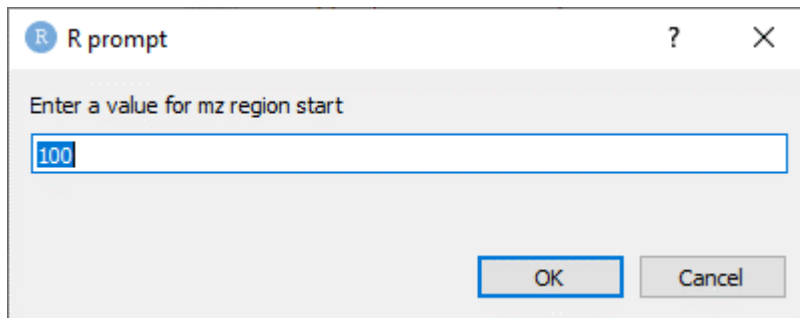


Figure 4: Example of GUI input parameters

#### 0.4.3 1.3 Load the Dataset

The function returns a list of the loaded files. It creates a timestamp tracking the amount of time to complete the data processing and creates the object 'ClassType' which associates every data file with the class of belonging.

```
#Skip this step if using the example dataset provided within the package
LoadData()
```

#### 0.4.4 1.4 Define Class Identifiers

The argument of the function is the object 'ClassType'. The function defines and returns specific symbols and colours for each sample class, which are then used to produce graphical outputs.

```
## don't run if using the the provided example dataset
```

```

ClassType<-c('Mix','Treatment1')

DefineClassAttributes(ClassType)

## [[1]]
## [1] "Mix"          "Mix"          "Mix"          "Treatment1" "Treatment1"
## [6] "Treatment1"
##
## [[2]]
## [1] "#FF0000" "#FF0000" "#FF0000" "#00FFFF" "#00FFFF" "#00FFFF"
##
## [[3]]
## [1] 1 1 1 2 2 2

```

#### 0.4.5 1.5 *Define Peak Picking Parameters*

The function uses GUIs to define xcms peak picking parameters using the centwave algorithm. These include: minimum and maximum peakwidth, minimum and maximum m/z error (in ppm), values for the integration threshold (set at 2000) and signal to noise threshold. The pick picking can be set to gaussian and the integration method can be 1 or 2, as for the xcms documentation. The default values for minimum and maximum m/z error and signal to noise (S/N) threshold are set to 'none' and correspond, respectively, to 3.03 and 50 ppm and to 1000 for S/N value. Detailed information about the mentioned parameters can be found in the xcms documentation. <https://www.bioconductor.org/packages/release/bioc/manuals/xcms/man/xcms.pdf>

```

#leave default values if using the package dataset
PeakPickingParam()

```

```

## [[1]]
## [1] 10
##
## [[2]]
## [1] 20
##
## [[3]]
## [1] 3.030303
##
## [[4]]
## [1] 50
##
## [[5]]
## [1] 26.51515
##
## [[6]]
## [1] 0.002
##
## [[7]]
## [1] FALSE
##
## [[8]]
## [1] 1

```

## 0.5 2. *Parameter Evaluation*

This is a QC step that allows testing the suitability of the input parameters on the reference files selected in the first step.

First, a GUI is used to define the settings of xcms functions `findChromPeaks` and `groupChromPeaks` which perform peak picking and grouping of molecular features. Secondly, data files are read using the xcms function `readMSData` and, lastly, peak picking and grouping is carried out. The output of the function is the XCMSnEXP object 'x\_refs'.

```
#not to be executed if using the example dataset
exploreRefs()
```

Note: In case the example dataset is used, the following needs to be run.

```
dir.create("./QC")
SampleGroup<-c("Mix", "Mix")
symbol<-c(1,1)
ClassCol<-c("#FF0000FF", "#FF0000FF")
path<-system.file("extdata",package = "MStractor")
files <- dir(path, pattern = ".mzXML", full.names = TRUE)
test<-files[1:2]
pd <- data.frame(sample_name = sub(basename(test), pattern = filetype,
  replacement = "", fixed = TRUE),sample_group = SampleGroup,
  stringsAsFactors = FALSE)
ref_data <- readMSData(test, pdata = new("NAnnotatedDataFrame",
  pd), mode = "onDisk")
expRefData(ref_data)
```

The `refTic` function returns a plot of the overlaid TICs for the reference files

```
refTic(x_refs)

## null device
##           1
```

**Figure 5: Non-aligned overlaid TICs**

The `get100` function randomly picks 100 features detected across the retention time domain and plots them into a matrix. Using this output, it is easy to check whether all the features are correctly integrated.

```
get100(x_refs)
```

**Figure 6: EICs of 100 features randomly picked across the retention time**

## 0.6 3. *Create XCMSnExp Dataset*

An XCMSnEXP dataset containing all the raw files to be processed is automatically created based on the folder structure defined in the data preparation step. The function performs the same steps described for `exploreRefs()`, with the only difference that the processing is applied to the whole dataset.

It is important that the input parameters used in this step match the ones defined for the reference files. After performing peak picking on each datafile, peaks are matched across all samples and grouped using the xcms functions `findChromPeaks` and `groupChromPeaks`. The results are stored in the XCMSnEXP object 'xdata'



```
peakPickGroup() # don't run if using the example dataset
```

In case the example dataset is used, the following needs to be run

```
ClassType<-c('Mix', 'Treatment1')
SampleGroup<-c("Mix", "Mix", "Mix",
               "Treatment1", "Treatment1", "Treatment1")
symbol<-c(1, 1, 1, 2, 2, 2)
ClassCol<- c("#FF0000FF", "#FF0000FF", "#FF0000FF",
             "#00FFFFFF", "#00FFFFFF", "#00FFFFFF")
path<-system.file("extdata", package = "MStractor")
files <- dir(path, pattern = ".mzXML", full.names = TRUE)
pd <- data.frame(sample_name = sub(basename(files), pattern = filetype,
    replacement = "", fixed = TRUE), sample_group = SampleGroup,
    stringsAsFactors = FALSE)
raw_data <- readMSData(files =files, pdata = new("NAnnotatedDataFrame",
    pd), mode = "onDisk")
ppgExData(raw_data)
```

## 0.7 Workflow Branching

The workflow forks at step 4 in branch a and branch b, as displayed in **Figure 1**. From this point on, the user can choose to either proceed with the XCMSnEXP object or perform a conversion into a xcmsSet object. Workflow branch A requires higher computing power than branch b. **Branches a** and **b** should not be run together.

## 0.8 Branch A

### 0.8.1 4a Print raw EICs and TICs

Step 4 produces graphical outputs that can be used to visualise macroscopic differences among chromatographic profiles. For both branches plots of overlaid TIC (**Figure7**) chromatograms and for individual EICs are stored in the QC directory.

```
OverlaidTICs(xdata, 'raw') #raw indicates non-retention time aligned signals
printEICs(xdata, 'raw')
```

**Figure 7: Overlaid TICs**

### 0.8.2 5a Retention time alignment, grouping and peak filling

Retention time correction parameters are entered via GUI. The retention time alignment method currently supported is 'loess' (see xcms manual), while Obiwrap will be possibly implemented in the future MStractor releases. A plot of the retention time deviation is also generated and stored in the QC folder. After correction, features are regrouped and the filling of missing signals is performed.

```
RTalign(xdata, 'loess')

xdata <- groupChromPeaks(xdata, param = PeakDensityParam(sampleGroups =
    xdata$sample_group, minFraction = 0.3, bw = 20))

xfilled <- fillChromPeaks(xdata, param =(FillChromPeaksParam(ppm = 50,
    expandMz = 0.5)))
```

### 0.8.3 6a *Print aligned EICs and TICs*

Step 6 replicates the series of functions already described for step 4a, with the only difference that retention time aligned TICs and EICs are generated

```
OverlaidTICs(xdata, 'aligned')
printEICs(xfilled, 'filled')
# 'aligned' and 'filled' indicate retention time aligned signals
```

### 0.8.4 7a *Create Summary Datamatrix*

Step 7 is specific for **branch a** and collates feature information generated by the `xcms featureDefinition()` function with their corresponding response (either the integrated peak area 'into' or maximum intensity 'maxo'). The arguments of the function are the filled XCMSnEXP object and the type of instrumental response desired (either 'into' or 'maxo'). After running step seven, the workflow merges again and the user has to proceed with step 8.

```
CreateDM(xfilled, 'maxo')
```

Following this, the `xsetConvert` function (already described in step 4) reverts the XCMSnEXP object into the `xcmsSet` one. This is necessary for using CAMERA in the subsequent steps.

```
xsetConvert(xfilled)
```

## Note: you might want to set/adjust the 'sampclass' of the returned `xcmsSet` object before proceeding with

```
sampnames(xset)<-spn
```

## 0.9 *Branch B*

The user needs to revert the object to a `xcmsSet` using the following code. This is required to execute function listed in the remaining sections of this document.

```
xsetConvert(xdata)
sampnames(xset)<-spn
```

### 0.9.1 4b *Print raw EICs and TICs*

```
getTICs(xcmsSet= xset, pngName= "./QC/TICs_raw.png", rt= "raw")
#raw indicates non-retention time aligned signals
printEICsXset(xset, 'raw')
```

### 0.9.2 5b *Retention time alignment, grouping and peak filling*

Retention time correction parameters are entered via GUI. The retention time alignment method currently supported is 'loess' (see `xcms` manual), while Obiwarped will be implemented in the future releases. A plot of the retention time deviation is also generated and stored in the QC folder. After correction, features are regrouped and the filling of missing signals is performed.

```
RTalign_xset(xset, 'loess')
xsAlign <- group(xsAlign, method= "nearest", mzVsRTbalance= 10, mzCheck=
  mzErrAbs, rtCheck= rtDelta, kNN=10)
xsFilled <- fillPeaks(xsAlign, method="chrom", expand.mz=0.5)
```

### 0.9.3 6b *Print aligned EICs and TICs*

Step 6 replicates the series of functions already described for stage 4, with the only difference that retention time aligned TICs and EICs are generated

```
getTICs(xcmsSet= xsAlign, pngName= "./QC/TICs_Aligned.png", rt= "corrected")
printEICsXset(xsFilled,'corrected')
# 'corrected' indicates retention time aligned signals
```

## 0.10 Workflow Merging

### 0.11 8. *Spectra Reconstruction*

From step 8, the workflow is the same for branches **a** and **b**. Spectra reconstruction is carried out using the function `autoCAMERA()` that includes the 3 CAMERA functions `groupFWHM()`, `findIsotopes()` and `groupCorr()`. A GUI allows the input of the function parameters. Consult CAMERA manual for more information: <https://www.bioconductor.org/packages/release/bioc/manuals/CAMERA/man/CAMERA.pdf>. Using `autoCamera`, the peak table embedded within the `xcmsSet` object is extracted and features arranged into pseudospectra groups according to their retention times. Then, based on the maximum expected charge test, isotopic patterns are located and the features belonging to coeluting compounds are resolved using a correlation matrix.

```
xset

## An "xcmsSet" object with 6 samples
##
## Time range: 11.7-8505 seconds (0.2-141.8 minutes)
## Mass range: 96.9573-1565.3684 m/z
## Peaks: 15284 (about 2547 per sample)
## Peak Groups: 2242
## Sample classes: Mix, Treatment1
##
## Feature detection:
##   o Peak picking performed on MS1.
##   o Scan range limited to 1 - 4400
## Profile settings: method = bin
##                   step = 0.1
##
## Memory usage: 6.81 MB

autoCamera(xset)

## Start grouping after retention time.
## Created 649 pseudospectra.
## Generating peak matrix!
## Run isotope peak annotation
## % finished: 10 20 30 40 50 60 70 80 90 100
## Found isotopes: 310
## Start grouping after correlation.
## Generating EIC's ..
## Warning: Found NA peaks in selected sample.
##
```

```

## Calculating peak correlations in 649 Groups...
## % finished: 10 20 30 40 50 60 70 80 90 100
##
## Calculating peak correlations across samples.
## % finished: 10 20 30 40 50 60 70 80 90 100
##
## Calculating isotope assignments in 649 Groups...
## % finished: 10 20 30 40 50 60 70 80 90 100
## Calculating graph cross linking in 649 Groups...
## % finished: 10 20 30 40 50 60 70 80 90 100
## New number of ps-groups: 1054
## xsAnnotate has now 1054 groups, instead of 649

```

The function generates the object `PksAn`, a datamatrix containing the results of the spectra reconstruction.

## 0.12 9. *Export Matrix*

Step 9 provides a series of 4 functions that generate the final results of the workflow. The second argument of the function “`collectBP_EICs`” is either “filled” for workflow branch a or “corrected” for workflow branch b

By using `FilterDM()` pseudospectra containing less than a fixed number of ions (2 by default) are removed from the data matrix. Then, only the most intense ion for each pseudospectrum is retained (base peak response). This function was implemented to obtain a simplified data matrix avoiding redundant information and allowing for quick and efficient downstream data processing.

The output is the `BasePks` object, a matrix summarizing the results of data filtering.

`CollectBP_EICs()` prints the extracted ion chromatograms related to the base peak matrix in a ‘png’ format. The function also returns the time required for processing the dataset. The png files are duplicated into 2 directories, named ‘`EICs_BasePeaks`’ and ‘`EICs_BasePeaks_Curated`’. The former is used as data back-up, while the latter allows the user to perform a final curation on the EICs, by removing those that do not contain useful information. This represents the only step of the entire workflow where a manual input of the user is required. An example is displayed in **Figure 8**.

`BasePks_Curated()` generates an updated data matrix (`BasePksCur`) with the discarded features removed. The final matrix is saved in a ‘.tsv’ file called ‘`Pks,BPs_Curated`’ which is saved within the working directory.

Lastly, by using `MedianNormalize()` the curated matrix is normalized on the median value. This is done to minimize possible minimal run-to-run variation in the instrument performances. The output is saved in the .tsv file ‘`Normalized Matrix`’.

```
FilterDM(PksAn, xset)
```

```

#the second argument of the function below is either 'filled' (branch a)
#or 'corrected' (branch b)
CollectBP_EICs(BasePks,'filled')

```

```
## Time difference of 56.29606 mins
```

```

#after this step manual curation is necessary
BasePks_Curated(BasePks)
MedianNormalize(BasePksCur, xset)

```

**Figure 8: Curation example**

### 0.13 10. *Descriptive Statistics*

An additional step of the workflow allows calculating descriptive statistics (average value, standard deviation and % CV) on the analytical replicates of each sample class. The output is stored in separate .tsv file for each class.

```
StatsByClass(ClassType, xset)
```

### 0.14 *Session information*

```
sessionInfo()
```

```
## R version 4.0.2 (2020-06-22)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 18363)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=English_Australia.1252 LC_CTYPE=English_Australia.1252
## [3] LC_MONETARY=English_Australia.1252 LC_NUMERIC=C
## [5] LC_TIME=English_Australia.1252
##
## attached base packages:
## [1] stats4      parallel  stats      graphics  grDevices  utils      datasets
## [8] methods     base
##
## other attached packages:
## [1] MStractor_0.1.0      xcms_3.11.3          MSnbase_2.15.2
## [4] ProtGenerics_1.21.0 S4Vectors_0.27.12    mzR_2.23.0
## [7] Rcpp_1.0.4.6         BiocParallel_1.23.0  Biobase_2.49.0
## [10] BiocGenerics_0.35.4
##
## loaded via a namespace (and not attached):
## [1] colorspace_1.4-1      ellipsis_0.3.1
## [3] htmlTable_2.0.0       XVector_0.29.2
## [5] GenomicRanges_1.41.5  base64enc_0.1-3
## [7] rmutil_1.1.5          clue_0.3-57
## [9] rstudioapi_0.11       affyio_1.59.0
## [11] statip_0.2.3          codetools_0.2-16
## [13] splines_4.0.2         ncdf4_1.17
## [15] doParallel_1.0.15     impute_1.63.0
## [17] robustbase_0.93-6     knitr_1.29
## [19] Formula_1.2-3         cluster_2.1.0
## [21] vsn_3.57.0            png_0.1-7
## [23] stabledist_0.7-1      graph_1.67.1
## [25] BiocManager_1.30.10   compiler_4.0.2
## [27] backports_1.1.8       Matrix_1.2-18
## [29] limma_3.45.7          acepack_1.4.1
## [31] htmltools_0.5.0       tools_4.0.2
## [33] igraph_1.2.5          modeest_2.4.0
## [35] gtable_0.3.0          glue_1.4.1
## [37] GenomeInfoDbData_1.2.3 affy_1.67.0
```

```

## [39] RANN_2.6.1                dplyr_1.0.0
## [41] MALDIquant_1.19.3         vctrs_0.3.1
## [43] multtest_2.45.0           preprocessCore_1.51.0
## [45] iterators_1.0.12          timeDate_3043.102
## [47] xfun_0.15                 stringr_1.4.0
## [49] spatial_7.3-12            lifecycle_0.2.0
## [51] gtools_3.8.2              XML_3.99-0.3
## [53] DEoptimR_1.0-8            zlibbioc_1.35.0
## [55] MASS_7.3-51.6             scales_1.1.1
## [57] timeSeries_3062.100       pcaMethods_1.81.0
## [59] SummarizedExperiment_1.19.5 RBGL_1.65.0
## [61] MassSpecWavelet_1.55.0    RColorBrewer_1.1-2
## [63] yaml_2.2.1                gridExtra_2.3
## [65] ggplot2_3.3.2             fBasics_3042.89.1
## [67] rpart_4.1-15              latticeExtra_0.6-29
## [69] stringi_1.4.6             foreach_1.5.0
## [71] checkmate_2.0.0           stable_1.1.4
## [73] svGUI_1.0.0               GenomeInfoDb_1.25.2
## [75] svDialogs_1.0.0           rlang_0.4.6
## [77] pkgconfig_2.0.3           matrixStats_0.56.0
## [79] bitops_1.0-6              mzID_1.27.0
## [81] evaluate_0.14             fda_5.1.4
## [83] lattice_0.20-41           purrr_0.3.4
## [85] htmlwidgets_1.5.1         tidyselect_1.1.0
## [87] plyr_1.8.6                magrittr_1.5
## [89] R6_2.4.1                  IRanges_2.23.10
## [91] snow_0.4-3                generics_0.0.2
## [93] Hmisc_4.4-0               DelayedArray_0.15.5
## [95] sm_2.2-5.6                pillar_1.4.4
## [97] foreign_0.8-80            berryFunctions_1.19.1
## [99] nnet_7.3-14               survival_3.2-3
## [101] abind_1.4-5               RCurl_1.98-1.2
## [103] tibble_3.0.1              CAMERA_1.45.1
## [105] crayon_1.3.4              rmarkdown_2.3
## [107] jpeg_0.1-8.1              grid_4.0.2
## [109] data.table_1.12.8         digest_0.6.25
## [111] tidyr_1.1.0               munsell_0.5.0
## [113] tcltk_4.0.2

```

## 0.15 *References*

- 1)Smith, C.A. and Want, E.J. and O'Maille, G. and Abagyan,R. and Siuzdak, G.: XCMS:Processing mass spectrometry data for metabolite profiling using nonlinear peak alignment, matching and identification, Analytical Chemistry, 78:779-787 (2006)
- 2)Ralf Tautenhahn, Christoph Boettcher, Steffen Neumann: Highly sensitive feature detection for high resolution LC/MS BMC Bioinformatics, 9:504 (2008)
- 3)H. Paul Benton, Elizabeth J. Want and Timothy M. D. Ebbels Correction of mass calibration gaps in liquid chromatography-mass spectrometry metabolomics data Bioinformatics, 26:2488 (2010)
- 4)Kuhl, C., Tautenhahn, R., Boettcher, C., Larson, T. R. and Neumann, S. CAMERA: an integrated strategy for compound spectra extraction and annotation of liquid chromatography/mass spectrometry data sets. Analytical Chemistry, 84:283-289 (2012)