Metabolomics data analysis with Bioconductor

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Talk content

- Very short introduction to metabolomics data analysis.
- Focus on pre-processing of LCMS data.
- Focus on the xcms package (new user interface), but other exist too (e.g. yamss).

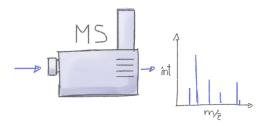
Metabolomics?

- Is the large-scale study of small molecules (metabolites) in a system (cell, tissue or organism).
- Metabolites are intermediates and products of cellular processes (metabolism).
- Simple definition:
 - Genome: what can happen.
 - Transcriptome: what appears to be happening.
 - Proteome: what makes it happen.
 - Metabolome: what actually happened. Influenced by genetic and environmental factors.

How are we measuring that?

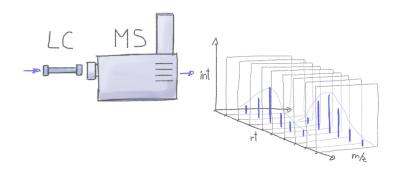
- Nuclear magnetic Resonance (NMR) not covered here.
- Mass spec (MS)-based metabolomics

Mass Spectrometry (MS)



 Problem: unable to distinguish between metabolites with the same mass-to-charge ratio (m/z).

Liquid Chromatography Mass Spectrometry (LCMS)



- Combines physical separation via LC with MS for mass analysis.
- Additional time dimension to separate different ions with same m/z.
- LCMS metabolomics: identify peaks in the m/z rt plane.

LCMS-based metabolomics data pre-processing

- Input: mzML or netCDF files with multiple MS spectra per sample.
- Output: matrix of abundances, rows being features, columns samples.
- feature: ion with a unique mass-to-charge ratio (m/z) and retention time.
- Example: load files from the faahKO data packages, process using xcms.

OnDiskMSnExp: small memory size, loads data on-demand.

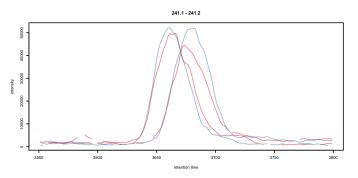
LCMS-based metabolomics data pre-processing

- Chromatographic peak detection.
- Sample alignment.
- Correspondence.

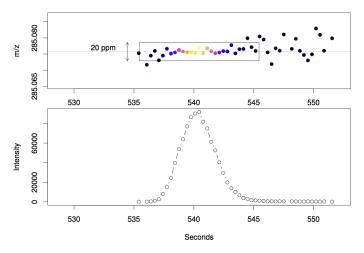
- Goal: Identify chromatographic peaks within slices along mz dimension.
- What type of peaks have to be detected?

```
mzr <- c(241.1, 241.2)
chrs <- extractChromatograms(faahKO, mz = mzr, rt = c(3550, 3800))

cols <- brewer.pal(3, "Set1")[c(1, 1, 2, 2)]
plotChromatogram(chrs, col = paste0(cols, 80))</pre>
```



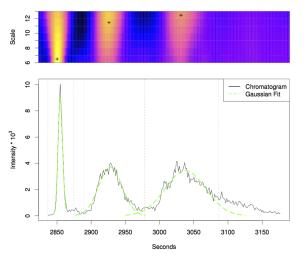
- centWave (Tautenhahn et al. BMC Bioinformatics, 2008):
- Step 1: Detection of regions of interest



mz-rt regions with low mz-variance.



• Step 2: Peak detection using continuous wavelet transform (CWT)



Equivalent to multiple Gaussian fits and choosing the best.



Example: centWave-based peak detection:

```
faahK0 <- findChromPeaks(faahK0, param = CentWaveParam())</pre>
```

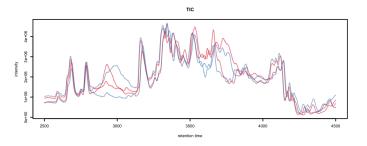
 Result: XCMSnExp, container for LC/GC-MS results, extends OnDiskMSnExp.

```
head(chromPeaks(faahK0))
```

```
mz mzmin mzmax
                   rt
                         rtmin
                                  rtmax
                                              into
                                                         intb maxo
[1.7] 425.9 425.9 425.9 2520.158 2510.768 2527.982
                                                 9999.769
                                                             9984 120
                                                                       741
[2,] 464.3 464.3 464.3 2518.593 2504.508 2532.677 32103.270
                                                            32082.926
                                                                      1993
[3,] 499.1 499.1 499.1 2524.852 2520.158 2527.982 4979.194 4904.709
                                                                      883
[4,] 572.7 572.7 572.7 2524.852 2520.158 2527.982 2727.446 2721.187
                                                                      559
[5,] 579.8 579.8 579.8 2524.852 2520.158 2527.982 2450.477 2444.218
                                                                       468
[6,] 453.2 453.2 453.2 2506.073 2501.378 2527.982 1007408.973 1007380.804 38152
sn sample is_filled
Г1. Т
      740
Γ2.7 1992
[3,] 13
[4,] 558 1
[5,]
     467
[6,] 38151
```

LCMS pre-processing: Alignment

- Goal: Adjust retention time differences/shifts between samples.
- Total ion chromatogram (TIC) representing the sum of intensities across a spectrum.



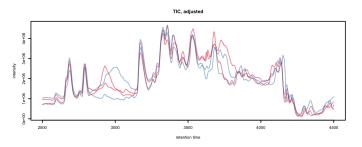
- Overview of algorithms: (Smith et al. Brief Bioinformatics 2013).
- xcms: peak groups (Smith et. al Anal Chem 2006), obiwarp (Prince et al. Anal Chem, 2006),

LCMS pre-processing: Alignment

• Example: use obiwarp to align samples.

```
faahKO <- adjustRtime(faahKO, param = ObiwarpParam())</pre>
```

• TIC after adjustment:



- Assumptions:
 - Samples relatively similar (either similar chromatograms or a set of common metabolites present in all).
 - Analyte elution order same in all samples.

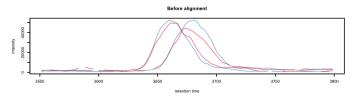


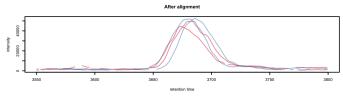
LCMS pre-processing: Alignment

• Example: effect of alignment on example peak.

```
chrs_adj <- extractChromatograms(faahKO, mz = mzr, rt = c(3550, 3800))

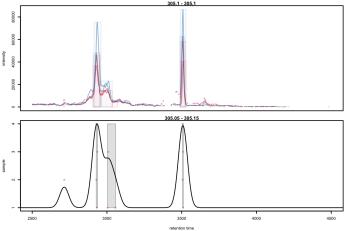
par(mfrow = c(2, 1))
plotChromatogram(chrs, col = paste0(cols, 80), main = "Before alignment")
plotChromatogram(chrs_adj, col = paste0(cols, 80), main = "After alignment")</pre>
```





LCMS pre-processing: Correspondence

- Goal: Group detected chromatographic peaks across samples.
- Peaks that are close in rt (and m/z) are grouped to a feature.
- xcms: peak density method:



LCMS pre-processing: Correspondence

Example: peak grouping.

```
faahKO <- groupChromPeaks(faahKO, param = PeakDensityParam())
```

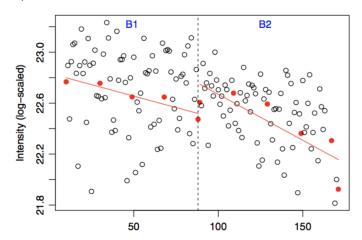
• featureValues: extract values for each feature from each sample.

```
## Access feature intensities
head(featureValues(faahKO, value = "into"))
      ko15.CDF ko16.CDF wt15.CDF wt16.CDF
FT0001 6029 945
                     NA 4586 527
FT0002 1144.015
                     NA 1018.815
FT0003
            NA 774 576 1275 475
                                      NA
FT0004
            NA
                     NA 1284 728
                                 1220 7
FT0005 2759.095 3872.963
                                       NA
FT0006 7682 585 3806 080
                             NA
                                      NA
```

• Fill-in values for missing peaks: fillChromPeaks.

What next? Data normalization

- Adjust within and between batch differences.
- MetNorm RUV for metabolomics (Livera et al. Anal Chem 2015).
- Injection order dependent signal drift (Wehrens et al. *Metabolomics* 2016).



What next? Identification

- Annotate features to metabolites.
- Each metabolite can be represented by multiple features (ion adducts, isotopes).
- Starting point: CAMERA package.
- On-line spectra databases (e.g. MassBank).

Finally...

• Hands on in the afternoon workshop.

thank you for your attention!