Analysis of MS Data in R

Investigating Mass Spectra and Neutral Loss Trigger

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Availability

```
https://github.com/sgibb/talk-odense-20171127/github:@sgibb/twitter:@sebastiangibb
```

Loading Data with MSnbase (Gatto and Lilley 2011)

```
library("MSnbase")

msfile <- file.path(
    "data",
    "neutral-loss",
    "LUM2_01470_KS_L1-5-2_EC17-123-150In2_55oC_NLtrig.mzML.gz"
)

ms <- readMSData(msfile, mode="onDisk")</pre>
```

The MSnExp object

```
ms
# MSn experiment data ("OnDiskMSnExp")
# Object size in memory: 50.52 Mb
# - - - Spectra data - - -
# MS level(s): 1 2 3
# Number of spectra: 108844
# MSn retention times: 0:0 - 120:0 minutes
# - - - Processing information - - -
# Data loaded [Fri Nov 24 14:09:00 2017]
# MSnbase version: 2.5.0
# - - - Meta data - - -
# phenoData
# rowNames:
     LUM2 01470 KS L1-5-2 EC17-123-150In2 55oC NLtrig.mzML.qz
# varLabels: sampleNames
   varMetadata: labelDescription
# Loaded from:
   LUM2_01470_KS_L1-5-2_EC17-123-150In2_55oC_NLtrig.mzML.gz
# protocolData: none
# featureData
  featureNames: F1.S000001 F1.S000002 ... F1.S108844 (108844
     total)
# fvarLabels: fileIdx spIdx ... spectrum (29 total)
# fvarMetadata: labelDescription
# experimentData: use 'experimentData(object)'
```

The MSnExp object 2

```
head(featureNames(ms))
# [1] "F1.S000001" "F1.S000002" "F1.S000003" "F1.S000004" "F1.S000005"
# [6] "F1.S000006"
fData(ms)[1:2, 1:6]
#
            fileIdx spIdx centroided smoothed seqNum acquisitionNum
# F1.S000001
                                 NA
                                         NA
                                                               1
# F1.S000002 1 2
                                 NA
                                         NA
ms[[1]]
# Object of class "Spectrum1"
# Retention time: 0:0
# MSn level: 1
# Total ion count: 32197
# Polarity: 1
```

Plot a Spectrum of an MSnExp object

```
s \leftarrow ms[[95]]
plot(mz(s), intensity(s), type="h")
    1200
intensity(s)
    800
    009
     400
     200
     0
                         500
                                                      1000
                                                                                    1500
                                                       mz(s)
```

Filter by MS level

```
ms1 <- ms[fData(ms)$msLevel == 1]

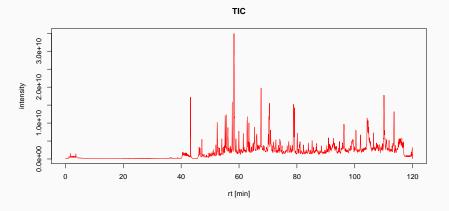
ms1 <- filterMsLevel(ms, 1)
ms2 <- filterMsLevel(ms, 2)
ms3 <- filterMsLevel(ms, 3)</pre>
```

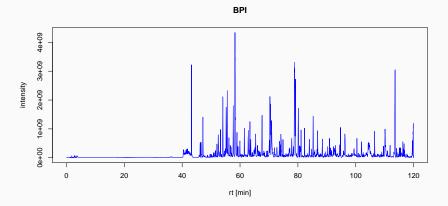
The MSnExp object 3

```
ms1
# MSn experiment data ("OnDiskMSnExp")
# Object size in memory: 1.81 Mb
# - - - Spectra data - - -
# MS level(s): 1
# Number of spectra: 5565
# MSn retention times: 0:0 - 119:58 minutes
# - - - Processing information - - -
# Data loaded [Fri Nov 24 14:09:00 2017]
# Filter: select MS level(s) 1 [Sat Nov 25 17:29:30 2017]
# MSnbase version: 2.5.0
# - - - Meta data - - -
# phenoData
# rowNames:
     LUM2 01470 KS L1-5-2 EC17-123-150In2 55oC NLtrig.mzML.qz
# varLabels: sampleNames
   varMetadata: labelDescription
# Loaded from:
   LUM2_01470_KS_L1-5-2_EC17-123-150In2_55oC_NLtrig.mzML.gz
# protocolData: none
# featureData
# featureNames: F1.S000001 F1.S000002 ... F1.S108817 (5565 total)
# fvarLabels: fileIdx spIdx ... spectrum (29 total)
# fvarMetadata: labelDescription
# experimentData: use 'experimentData(object)'
```

TIC

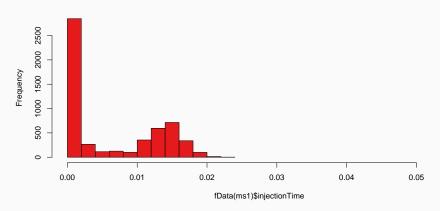
```
plot(rtime(ms1) / 60, tic(ms1),
    type="l", col="red",
    main="TIC", xlab="rt [min]", ylab="intensity")
```





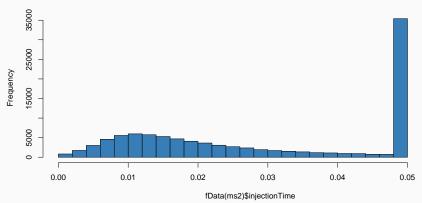
IIT MS1

```
hist(fData(ms1)$injectionTime,
    col=col["MS1"],
    main="Ion Injection Time MS1", xlim=c(0, 0.055))
```



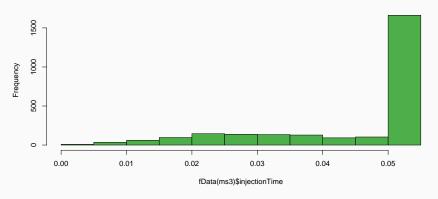
IIT MS2

```
hist(fData(ms2)$injectionTime,
    col=col["MS2"],
    main="Ion Injection Time MS2", xlim=c(0, 0.055))
```



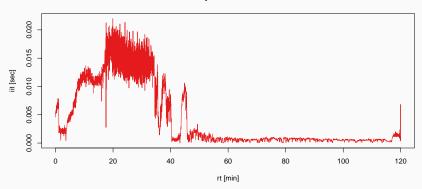
IIT MS3

```
hist(fData(ms3)$injectionTime,
    col=col["MS3"],
    main="Ion Injection Time MS3", xlim=c(0, 0.055))
```



IIT vs RT MS1

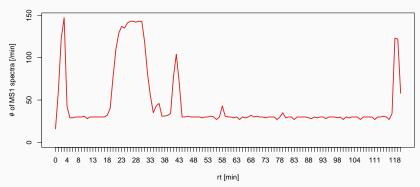
```
plot(rtime(ms1) / 60, fData(ms1)$injectionTime,
     type="l", col=col["MS1"],
    main="Ion Injection Time MS1",
    xlab="rt [min]", ylab="iit [sec]")
```



MS1 per minute

```
plot(table(round(rtime(ms1) / 60)),
    type="l", col=col["MS1"], main="# MS1/min",
    xlab="rt [min]", ylab="# of MS1 spectra [/min]")
```

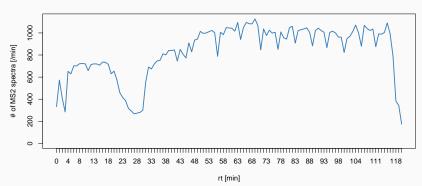




MS2 per minute

```
plot(table(round(rtime(ms2) / 60)),
    type="1", col=col["MS2"], main="# MS2/min",
    xlab="rt [min]", ylab="# of MS2 spectra [/min]")
```

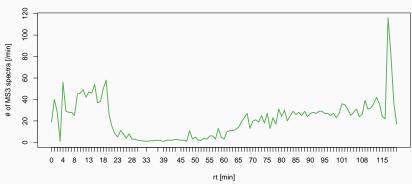
MS2/min



MS3 per minute

```
plot(table(round(rtime(ms3) / 60)),
    type="1", col=col["MS3"], main="# MS3/min",
    xlab="rt [min]", ylab="# of MS3 spectra [/min]")
```

MS3/min



NL trigger (Jiang et al. 2017)

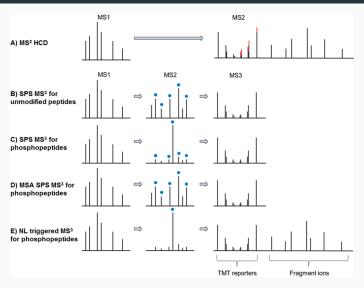


Figure: (Fig. 1 in Jiang et al. 2017)

NL trigger/Thermo Orbitrap Lumos



Figure

NL trigger

Own NL trigger

MS2: mz of top 5 peaks - precursor mz \approx 80/97 [Da]

Thermo NL trigger

 $MS2 \leftrightarrow MS3$

Verification

Comparison of MS2 scan numbers

NL trigger/How to find our own trigger? 1

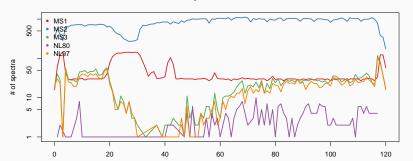
```
findTop5mz <- function(x) {</pre>
   mz(x)[order(intensity(x), decreasing=TRUE)[1:5]]
top5mz <- spectrapply(ms2, findTop5mz)</pre>
top5mz <- do.call(rbind, top5mz)</pre>
top5mz[1:2, ]
#
                [,1] [,2] [,3] [,4] [,5]
# F1.S000003 507.9660 619.1350 427.9571 699.134 601.090
# F1.S000004 676.1671 427.9181 756.1360 508.006 409.902
deltamz <- precursorMz(ms2) - top5mz
deltamz[1:2,]
#
                 [,1] [,2] [,3] [,4] [,5]
# F1.S000003 55.58902 -55.57996 135.5980 -135.57898 -37.53497
# F1.S000004 -84.10120 164.14780 -164.0702 84.05984 182.16382
```

NL trigger/How to find our own trigger? 2

```
nl <- c(phosphate=80, phosphoNL=97.9763)</pre>
tolerance <- 0.5 # [Da]
# How many phosphates?
n180 <- as.logical(rowSums(
    abs(deltamz - nl["phosphate"]) <= tolerance)</pre>
sum(n180)
# [1] 276
# How many phosphoNL?
n197 <- as.logical(rowSums(
    abs(deltamz - nl["phosphoNL"]) <= tolerance)</pre>
sum(n197)
# [1] 2124
```

NL trigger/Spectra + NL vs RT log

spectra vs rtime



NL trigger/Find device trigger; the tricky part 1

	scanld	pcCid
F1.S000003	3	563.5551@cid35.00
F1.S000004	4	592.0658@cid35.00

	scanld	pcCid
F1.S000040	40	933.4402@cid35.00
F1.S000054	54	850.0895@cid35.00

NL trigger/Find device trigger; the tricky part 2

pcCid	scanId.ms3	scanld.ms2	delta
933.4402@cid35.00	40	38	2
850.0895@cid35.00	54	52	2
644.5992@cid35.00	67	11506	-11439
644.5992@cid35.00	67	11081	-11014
644.5992@cid35.00	67	4763	-4696
644.5992@cid35.00	67	4364	-4297

NL trigger/Find device trigger; the tricky part 3

```
nldf <- nldf[nldf$delta > 0,]
nldf <- nldf[order(nldf$scanId.ms3, nldf$delta),]
nldf <- nldf[!duplicated(nldf$scanId.ms3),]</pre>
```

	pcCid	scanId.ms3	scanId.ms2	delta
1	933.4402@cid35.00	40	38	2
2	850.0895@cid35.00	54	52	2
10	644.5992@cid35.00	67	65	2
12	699.0558@cid35.00	82	80	2
15	636.8463@cid35.00	83	81	2
16	757.3283@cid35.00	108	106	2

NL trigger/Find common trigger/differences 1

Set Size

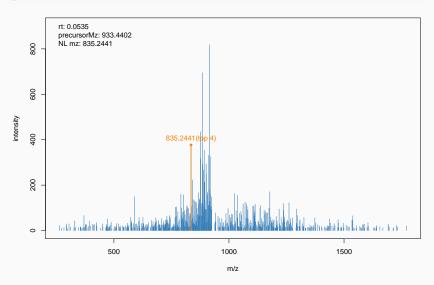
```
ms2idNL <- list(own=scanIndex(ms2)[nl80 | nl97],
                  thermo=nldf$scanId.ms2)
library("UpSetR")
upset(fromList(ms2idNL), order.by="freq")
                  2500
                                 2289
                  2000
                  1000
                  500
   2000
```

NL trigger/Find common trigger/differences 2

```
common <- intersect(ms2idNL$own, ms2idNL$thermo)
uniqueOwn <- setdiff(ms2idNL$own, ms2idNL$thermo)
uniqueThermo <- setdiff(ms2idNL$thermo, ms2idNL$own)</pre>
```

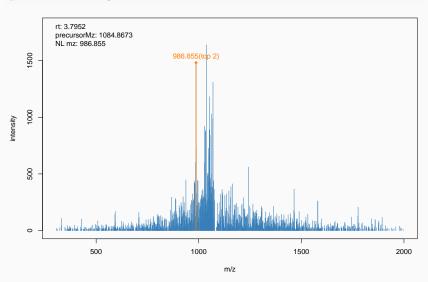
NL trigger/Common trigger

plotMs2(ms[[common[1]]])



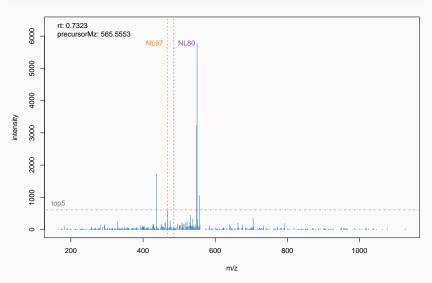
NL trigger/Should have been triggered

plotMs2(ms[[unique0wn[1]]])



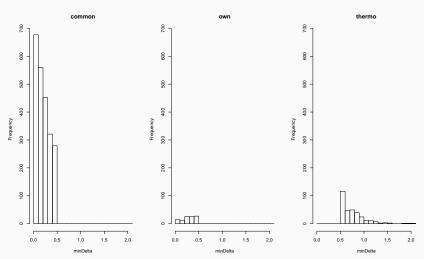
NL trigger/Was triggered but shouldn't

plotMs2(ms[[uniqueThermo[1]]])



NL trigger/Trigger tolerance

```
1 <- list(common=common, own=uniqueOwn, thermo=uniqueThermo)
i <- lapply(1, match, scanIndex(ms2))
for (j in seq(along=1)) {
    minDelta <- apply(deltamz[i[[j]], ], 1, function(r)min(abs(r - rep(nl, each=length(r)))))
    hist(minDelta, breaks=seq(0, 2.1, by=0.1), ylim=c(0, 675), main=names(1)[j])
}</pre>
```



Thanks

- MSnbase: @lgatto, @jotsetung
- mzR: @sneumann, @lgatto, @jotsetung

Availability

```
https://github.com/sgibb/talk-odense-20171127/github:@sgibb/twitter:@sebastiangibb
```

References

Gatto, Laurent, and Kathryn S. Lilley. 2011. "MSnbase-an R/Bioconductor Package for Isobaric Tagged Mass Spectrometry Data Visualization, Processing and Quantitation." *Bioinformatics* 28 (2). Oxford University Press (OUP): 288–89. doi:10.1093/bioinformatics/btr645.

Jiang, Xiaoyue, Ryan Bomgarden, Joseph Brown, Devin L. Drew, Aaron M. Robitaille, Rosa Viner, and Andreas R. Huhmer. 2017. "Sensitive and Accurate Quantitation of Phosphopeptides Using TMT Isobaric Labeling Technique." *Journal of Proteome Research* 16 (11). American Chemical Society (ACS): 4244–52. doi:10.1021/acs.jproteome.7b00610.