Chromatoplots Manual

StatGraph

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Contents

1 Introduction

Metabolomics experiments often rely on Gas Chromatography - Mass Spectrometry (GC-MS) instruments for measuring the levels of metabolites. We have developed numerical methods and graphical diagnostics for converting raw GC-MS data into a dataset specifying the amount of each metabolite in each sample. In this manual, we present basic workflow and diagnosic windows.

This manual doesn't include the GUI yet, will be added later. And this manual doesn't focus on the algrithm, algrithm will be elucidated in another document later.

Now it's just for developer's communication, will be fullfilled while developing this package.

2 Installation

Please use the latest version of the following package, you can check out "commandr" and "chromatoplots" from their svn, and "xcms", "irange", from Bioconductor¹ dev 2.6 pool.

chromatoplots

svn://had.co.nz/statgraphics/chromatoplots/trunk/chromatoplots

commandr

svn://had.co.nz/statgraphics/commandr/trunk/commandr

xcms

http://www.bioconductor.org/packages/2.6/bioc/html/xcms.html

IRanges

http://www.bioconductor.org/packages/2.6/bioc/html/IRanges.html

Check the system and package dependencies before you install chromatoplots, install xcms first, then commandr, install chromatoplots last.

¹www.bioconductor.org

3 Workflow

3.1 Arrange Your File

First of all, you need to arrange all your raw data based on your experimental design.

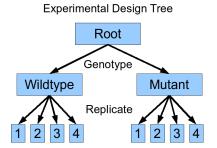


Figure 1: Experimental design

In this study, my experimental design looks like this.

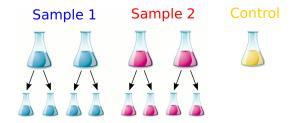


Figure 2: My calibrated data has two treatment, each treatment has four replicates.

So I arranged my file in the following way.

```
./raw/s1:
M0101A.CDF M0101B.CDF M0102A.CDF M0102B.CDF
./raw/s2:
M0201A.CDF M0201B.CDF M0202A.CDF M0202B.CDF
```

3.2 Raw Data Input

Command Line

Specify a CDF data which contains GC-MS processed raw data by using the following command, and generate $profile\ matrix^1$.

```
> file <- "raw/s1/M0101A.CDF"
> raw <- loadSample(file)
> raw_prof <- genProfile(raw) # generate profile matrix</pre>
```

Graphic Diagnose

To call a diagnostic windows for this stage.

> explore(raw_prof)

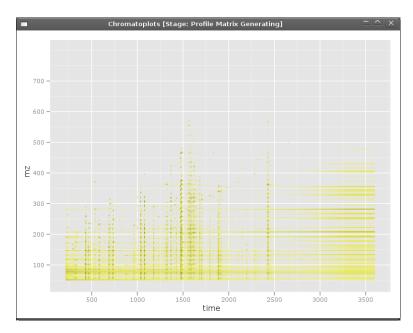


Figure 3: Image of log of the profile matrix. Horizontal axis is time, and vertical axis is m/z. The color scale ranges from yellow to black. The darker the color, the higher the intensity. Gray indicates a time and m/z combination where nothing is detected.

Low Level Graphic Functions

image(object) #profile matrix, slow(more than one minute)

¹Profile matrix has a row for each mass and a column for each scan, in order of time.

3.3 Baseline Subtraction

Command Line

```
> cor_prof <- removeBaseline(raw_prof, "median", scanrad = 100)
#if it works great, then accept correction.
> raw <- cor_prof</pre>
```

Graphic Diagnose

> explore(cor_prof, raw = raw_prof, mz=51)

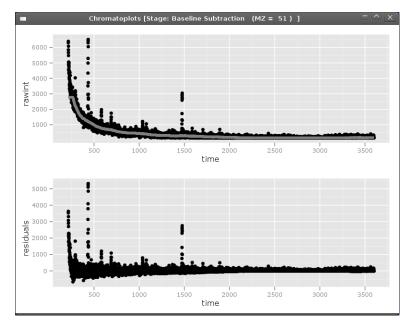


Figure 4: Chromatogram of raw relative intensities from the row of the profile matrix corresponding to 51 m/z. A non-linear baseline is evident.

Low Level Graphic Functions

explore_prof_filter(object,protocol,raw,mz,subtract=TRUE)

3.4 Peak Detection

Command Line

```
> peaks <- findPeaks(raw, "gauss") # "gauss" protocol
```

You can read another file and do the previous step by simply running permorn() command.

```
> rep_file <- "raw/s1/M0102A.CDF"
> rep_raw <- perform(pipeline(raw), rep_file)
> rep_peaks <- perform(findPeaksProto(peaks), rep_raw)</pre>
```

Or load all replicates within one treatment at once, the first argument specify the directory contains replicates raw files.

```
> s1_exp <- loadExperiment("raw/s1/", pipeline = pipeline(peaks))
```

Graphic Diagnose

```
> explore(peaks, raw = raw, sample=NA,residuals=TRUE,island=TRUE)
```

"sample" controls which sample you want to show; residuals=TRUE if you want the residual graph, island=TRUE if you want the plot which shows a cutoff line(red) and peaks(blue) they find for a specific MZ value (identified by user).

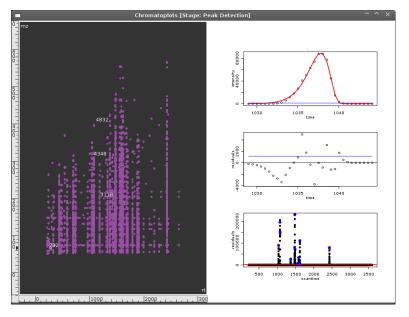


Figure 5: Chromatoplots visualization window for the peak detection stage. The user selects points in the left image to view the baseline-corrected chromatogram at the corresponding m/z. The chromatogram is annotated with the quantile cutoff and the maxima of detected peaks(right bottom). On the right top is a GGobi visu- alization for exploring the peak fits.

Low Level Graphic Functions

plot_peak(object,id, raw, cutoff, sample=NA,residuals = FALSE,island=FALSE)
cplotPeaks(obj,raw,mz) #show island only, ggplot2 function
cplotPeaks2(obj,raw,mz) #show island only, basic R graphic,faster

3.5 Components Detection

Command Line

To detect components, run the following command, the second method is essentially the same with the first one, but you can specify a quantity(TIC) filter and a peaks number filter. The reason here is to reduce components that contains very few peaks or quantity(if you want), these components may be falsely assigned to the wrong group, e.g. those only contains one m/z value maybe identified as one group with each other.

Then you could load all replicates in the whole experimental tree at once.

```
> samples <- c("raw/s1", "raw/s2")
> s1_s2_xset <- perform(pipeline(xset_comps), samples)</pre>
```

Graphic Diagnose

For graphic diagnose, run the following command, you could compare the result with filter.

```
> explore(xset_comps,sample=1,residual=TRUE,island=TRUE)
> explore(xset_comps_filt,sample=1,residual=TRUE,island=TRUE)
```

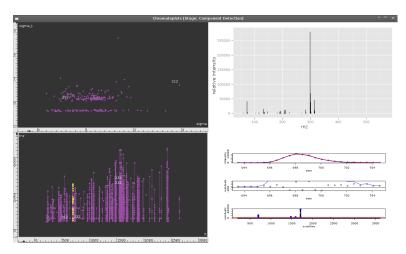


Figure 6: Chromatoplots visualization window for the component detection stage. When the user selects components in the top scatterplot, the mass spectrum for the component is displayed to the right and the corresponding chain of peaks is highlighted in the scatterplot on the bottom left. Selecting a peak in that plot displays the peak fit to the right.

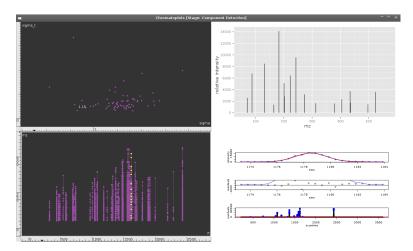


Figure 7: This plot is the same with , but this is after compsfilter, components numbers are reduced a lot.

Low Level Graphic Functions

plot_comp(peaks, mz_range)

3.6 Grouping Components

Command Line

Two arguments used here, default dist.cutoff=0.05, rt_window=-1 which means default setup never use rt windows filter.

```
> xset_groups <- groupComps(s1_s2_xset, "angle")
> xset_groups <- groupComps(s1_s2_xset, "angle",dist.cutoff=0.05,rt_window=1)</pre>
```

Graphic Diagnose

With this interactive graphic, you could diagnose spectra for each components within one group.

> explore(xset_groups)

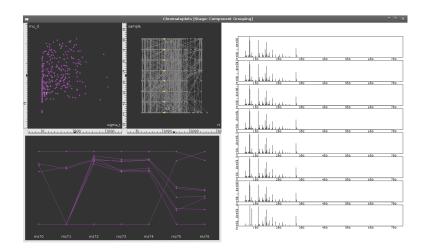


Figure 8: Chromatoplots visualization window for the component grouping stage. The user selects a group in the top scatterplot to highlight the corresponding chain of components in the plot to its right and display the spectra of the components in the parallel coordinates plot at the bottom, and plot spectra for each components inside that group in the right graphic.

Low Level Graphic Functions

cplotSpec(object,sample=NA,comp=NA,group=NA,...)

3.7 Retention Time Correction

Command Line

This stage based on the grouping result, which means if components are grouped in the wrong way, retention time will not be corrected in the right way, so you should pay attention to the last stage.

> xset_rtcor <- rtcor(xset_groups, "rloess")</pre>

Graphic Diagnose

We offer two ways to view the retention time, one is heatmap, the other one is lines, you can specify x-scale, samples you want to diagnose. (not interactive yet)

> explore(xset_rtcor_filt,raw=xset_groups_filt,geom="heatmap")

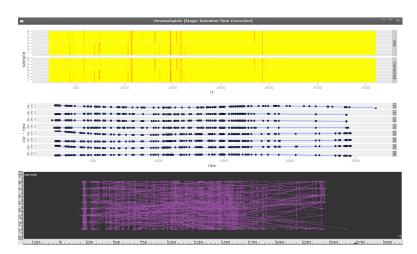


Figure 9: Chromatoplots visualization window for the retention time correction stage. Using "heatmap" method to plot for all the data

> explore(xset_rtcor,raw=xset_groups,sample=1:2,xscale=c(1000,1500))

Low Level Graphic Functions

cplotRT(object,xscale=NA,geom=NA,sample=NA,log=F)#geom=NA or "heatmap"
cplotRtFit(object,raw,xscale=NA,sample=NA) #cor-fit plot

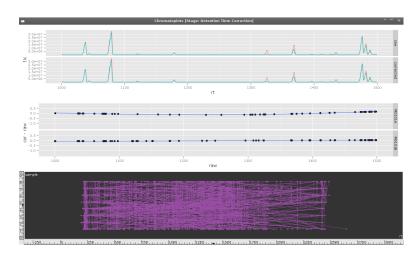


Figure 10: Chromatoplots visualization window for the peak detection stage. Using lines to show small group or scale of data

3.8 Summarization

Command Line

> xset_sum <- summarize(xset_groups,"common")</pre>

Graphic Diagnose

> explore(xset_sum)

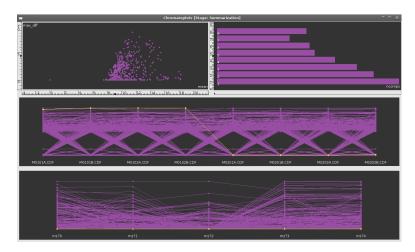


Figure 11: Chromatoplots visualization window for the summarization stage. By selecting a group in the top-left scatterplot, the user can examine the trend of the group over the samples and compare the mass spectra of its components.

3.9 Normalization

Command Line

> xset_norm <- normalize(xset_sum, "scale")</pre>

Graphic Diagnose

No diagnose graphic for this stage yet.

3.10 Metabolite Identification

Command Line

MSP format could be read by AMDIS to identify metabolites based on the NIST library. So far we only offer MSP transform from our final result. Our recommendation is: first read the processed data into explorase, then find the metabolite of interest in your favorite way, pick up the id for those metabolite, then use the following command to finish the transformation. 'id' indicate the metabolite of interest, and 'file' argument specify a file name you want to output to.

```
> dump2msp(xset_norm1_filt,id=c(183,80),file="~/interest.msp")
```

MSP format looks like this, in each parenthesis, left number indicate M/Z ratio, and right number indicate quantity, this list store the information about spectra for specific metabolite.

NAME:ID= 183 COMMENT: RI: CASNO:

RT: 12.31059

SOURCE:

NUM PEAKS: 116 (52 1914) (53 2737) (54 1816) (55 13739) (56 11831) (57 25260) (58 49057) (59 340150) (60 28146) (61 19998) 66 8410) (67 1950) (69 3679) (70 18527) (71 19182) (73 2520939) (74 216013) (75 163938) (76 8531) (77 2867) (82 1302) (83 3869) (84 8377) (85 22143) (86 738410) (87 78753) (88 31147) (89 5262) (95 6570) (97 1239) (98 4172) (99 16189) (100 451355) (101 100048) (102 46618) (103 35702) (104 6098) (105 13001) (106 1567) (109 1304) (112)2416) (113 19520) (114 19720) (115 33595) (116 57323) (117 159767) (118 26505) (119 52013) (120 6254) (121 2898) (123)2658) (128 5611) (129 13716) (130 198813) (131 187114) (132 600902) (134 45430) (136 54134) (133 88626) (135 4035) (142)4978) (143 5232) (144 59164) (145 15154) (147 1876247) (148 288850) (149 144728) (150 14222) (151 3035) (156 1950) (158)100146) (159 28280) (160 37288) (161 12054) (162 3850) (171)3126) (172 80766) (174 8348546) (175 1543325) (176 663500) (177 103438) (178 17975) (179 3491) (187 5887) (188 58814) (189 17957) (190 17487) (191 7612) (192 2102) (202 33007) 5912) (207 (203 6666) (204 41724) (205 11967) (206 1754) (218)7095) (219 1808) (221 1818) (232 4204) (234 1909) (246 120672) (248 1948382) (249 492673) (250 244634) (251 42950) (252 10635) (263 1341) (276 665575) 9950) (262 2754) (264 86643) (279 15637) (280 3916) (291 (277)174492) (278 3736) (292 1191)

NAME:ID= 80

COMMENT:

RI: CASNO:

RT: 17.92231

SOURCE:

NUM PEAKS: 30

(68	14601)(82	1207) (84	3577)(86	90239) (98	5911)
(100	55826) (112	6396) (114	15901)(126	2180) (140	15216)
(142	43639) (144	24571) (154	1505) (156	21899) (158	14960)
(170	3668) (171	1918) (172	57686) (174	1541131) (175	285225)
(176	121220) (186	4869) (200	3673) (214	13901) (246	79569)
(247	17734) (248	6937) (288	5245) (302	1693) (304	591876)

4 Other Graphic Function

Keep updating ...

cplotTIC (object,xintercept=NA,xscale=NA,color=NA) # plot TIC

5 Workflow Fast Reference

```
> file <- "raw/s1/M0101A.CDF"</pre>
> raw <- loadSample(file) # load raw data
> raw_prof <- genProfile(raw) # generate profile matrix</pre>
> explore(raw_prof)
                               # view profile matrix, slow
> cor_prof <- removeBaseline(raw_prof, "median", scanrad = 100)</pre>
> raw <- cor_prof #if it works great, then accept correction.
> explore(cor_prof, raw = raw_prof, mz=51)
> peaks <- findPeaks(raw, "gauss") # "gauss" protocol
> rep_file <- "raw/s1/M0102A.CDF"</pre>
> rep_raw <- perform(pipeline(raw), rep_file)</pre>
> rep_peaks <- perform(findPeaksProto(peaks), rep_raw)</pre>
> s1_exp <- loadExperiment("raw/s1/", pipeline = pipeline(peaks))</pre>
> explore(peaks, raw = raw, sample=NA,residuals=TRUE,island=TRUE)
> xset_comps <- findComps(s1_exp, "sigma")</pre>
# you can use filter here
> xset_comps_filt <- findComps(s1_exp,"sigma_filt",</pre>
                  tic.cutoff=0.69,npeaks.cutoff=0.7)
> samples <- c("raw/s1", "raw/s2")</pre>
> s1_s2_xset <- perform(pipeline(xset_comps), samples)</pre>
> explore(xset_comps,sample=1,residual=TRUE,island=TRUE)
> xset_groups <- groupComps(s1_s2_xset, "angle")</pre>
# or use dist.cutoff and rt_window filter at the same time.
> xset_groups <- groupComps(s1_s2_xset, "angle",dist.cutoff=0.05,rt_window=1)
> explore(xset_groups)
> xset_rtcor <- rtcor(xset_groups, "rloess")</pre>
> explore(xset_rtcor_filt,raw=xset_groups_filt,geom="heatmap")
# or use other ways to view this stage
> explore(xset_rtcor,raw=xset_groups,sample=1:2,xscale=c(1000,1500))
> xset_sum <- summarize(xset_groups,"common") #summarization</pre>
> explore(xset_sum)
> xset_norm <- normalize(xset_sum, "scale") #normalization
> library(explorase)
# use explorase to pick up interest id.
> explorase(as(xset_norm, "ExpressionSet"))
# transfrom this id to MSP format
> dump2msp(xset_norm1_filt,id=c(183,80),file="~/interest.msp")
# Identify your metabolite
```

6 Update List

$6.1 \quad 2009-11-1$

• Metabolite identification: dump2msp() transform final data to MSP format

- Remove legend for profile matrix.
- Using stack_plots to show profile matrix.
- Baseline subtraction: add arguments mz to explore() and related low level function, could be used for interactive from raw profile matrix later.
- Find Peaks: Dignostic windows fixed, add interactive plot for mz, and find a way to embed three figures in one box.
- Find comps: remove 'prof' and 'raw' argument, compute them from the object based on protocol@alpha, add some other argument.
- ProcessProtoRaw,desn't work so far, so use sample argument to retrieve the file path from phenoData slot.
- Retention time add some diagnostic graphic
- Grouping components: Add interactive spectra diagnostic window
- remove undifined protocol from namespace, so the package can be complied
- Add retetion time to feature data