MDstatsDIAMS: Real Data Exploration Using Spectronaut Report

Namgil Lee^{*}, Hojin Yoo[†], Juhyoung Kim[‡], Heejung Yang[§]

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This vignette explores and visualizes real DIA-MS data given in a Spectronaut report. This vignette reproduces figures and tables in Sections S1 and S2 of the Supplementary Material of the paper "A Shrinkage-based Statistical Method for Testing Group Mean Differences in Quantitative Bottom-up Proteomics" written by the authors.

Preparation

1. Set parameters to reduce analysis time

Because the real data size can be too large, users can choose parameters to reduce analysis time in this vignette.

```
# n_protein:
# Size of a subset of proteins randomly selected.
# If -1 or Inf, include all proteins. Default to 100.
n_protein <- 100

# remove_intermediate_reports:
# If TRUE, remove report from environment if not used in further steps.
# Default is TRUE.
remove_intermediate_reports <- TRUE</pre>
```

2. Load packages.

```
library(dplyr)
```

3. Load Report Data

Load a Spectronaut report from a remote repository.

```
df_real <- arrow::read_parquet(
   paste0(
     "/Users/namgil/Documents/Projects/MDstatsDIAMS/data/",</pre>
```

^{*}Kangwon National University, and Bionsight Inc., namgil.lee@kangwon.ac.kr

[†]Bionsight Inc.

[‡]Kangwon National University

[§]Kangwon National University, and Bionsight Inc., heejyang@kangwon.ac.kr

```
"lip_quant_staurosporine_hela_sn_report.parquet"
 )
)
  4. Filter the report and compute log10-transformed precursor quantity.
df_filtered <- df_real %>%
  filter(
    F.ExcludedFromQuantification == 'False'
    & EG.Qvalue < 0.01
    & F.NormalizedPeakArea > 1
  mutate(F.Log10NormalizedPeakArea = log10(F.NormalizedPeakArea)) %>%
  select(
    R.Condition, R.Replicate, PG.ProteinGroups, PG.ProteinNames,
    EG. Modified Sequence, FG. Charge, F. FrgIon, F. FrgLossType, F. Charge,
    EG.Qvalue, F.Log10NormalizedPeakArea
print(dim(df_filtered))
## [1] 13904962
                       11
  5. Set the order of the condition values so that "DMSO" comes before the other conditions.
conditions = unique(df_filtered$R.Condition)
df_filtered <- df_filtered %>%
 mutate(
   R.Condition = factor(
      R.Condition,
      labels = c("DMSO", conditions[-which(conditions == "DMSO")])
    )
  )
print(levels(df_filtered$R.Condition))
## [1] "DMSO" "100pM" "1nM"
                                "10nM" "100nM" "1mM"
                                                          "10mM" "100mM"
  6. Randomly select a subset of the predefined number of proteins to reduce analysis time.
if (
 n_protein < 0
 || is.infinite(n_protein)
  | | n_protein >= n_distinct(df_filtered$PG.ProteinGroups)
) {
 df_subset = df_filtered
} else {
  set.seed(111)
  protein_subset = sample(unique(df_filtered$PG.ProteinGroups), n_protein)
 df_subset = df_filtered %>% filter(PG.ProteinGroups %in% protein_subset)
print(dim(df_subset))
```

[1] 246385 11

7. Remove the large report data from the R environment that will not be used in the next steps.

Summary Statistics

Print summary statistics by condition.

##	#	A tibble: 8	x 4		
##		${\tt R.Condition}$	num_protein	num_precursor	num_fragment_ion
##		<fct></fct>	<int></int>	<int></int>	<int></int>
##	1	DMSO	99	2871	9064
##	2	100pM	100	2841	8972
##	3	1nM	99	2822	8907
##	4	10nM	98	2818	8861
##	5	100nM	99	2850	8967
##	6	1mM	99	2822	8895
##	7	10mM	98	2836	8951
##	8	100mM	99	2854	8912
##	#	A tibble: 8	x 3		
## ##				ean_quantity p	recursor_sd_quantity
				ean_quantity production of the contract of the	recursor_sd_quantity <dbl></dbl>
## ##		R.Condition			
## ## ##	1	R.Condition <fct></fct>		<dbl></dbl>	<dbl></dbl>
## ## ## ##	1 2	R.Condition <fct> DMSO</fct>		<db1> 4.43</db1>	<dbl> 0.707</dbl>
## ## ## ##	1 2 3	R.Condition <fct> DMSO 100pM</fct>		<dbl> 4.43 4.44</dbl>	<dbl> 0.707 0.712</dbl>
## ## ## ## ##	1 2 3 4	R.Condition <fct> DMSO 100pM 1nM</fct>		<dbl> 4.43 4.44 4.45</dbl>	<dbl> 0.707 0.712 0.710</dbl>
## ## ## ## ##	1 2 3 4 5	R.Condition <fct> DMSO 100pM 1nM 10nM</fct>		<db1> 4.43 4.44 4.45 4.45</db1>	<dbl> 0.707 0.712 0.710 0.721</dbl>
## ## ## ## ## ##	1 2 3 4 5 6	R.Condition <fct> DMSO 100pM 1nM 10nM 100nM</fct>		<db1> <db1> 4.43 4.44 4.45 4.45 4.44</db1></db1>	<dbl> 0.707 0.712 0.710 0.721 0.709</dbl>
## ## ## ## ## ##	1 2 3 4 5 6 7	R.Condition <fct> DMSO 100pM 1nM 10nM 100nM 1mM</fct>		<db1> 4.43 4.44 4.45 4.45 4.44 4.44</db1>	<dbl> 0.707 0.712 0.710 0.721 0.709 0.713</dbl>

Precursor Quantity Distribution

1. Precursor quantity distribution at fixed condition

Distribution of precursor quantity: all precursors measured in four replicates under fixed condition.

Empirical Density Pearson r = 0.995Normal Density 9 500 Sample Quantiles 2 Frequency 4 300 က α 100 0 0 2 4 6 8 10 -3 -1 1 2 3 Peptide Quantity, log10 Theoretical Quantiles

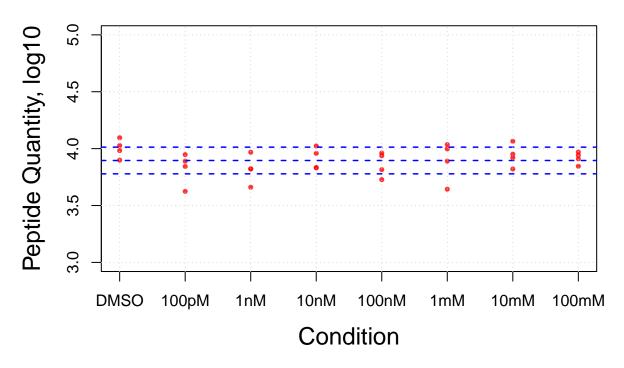
Condition: DMSO

2. An Example of Dose-Response Plot for A Selected Precursor

Condition: DMSO

Distribution of precursor quantity: a selected precursor across all conditions.

SGK1_HUMAN:_HLLEGLLQK_.2

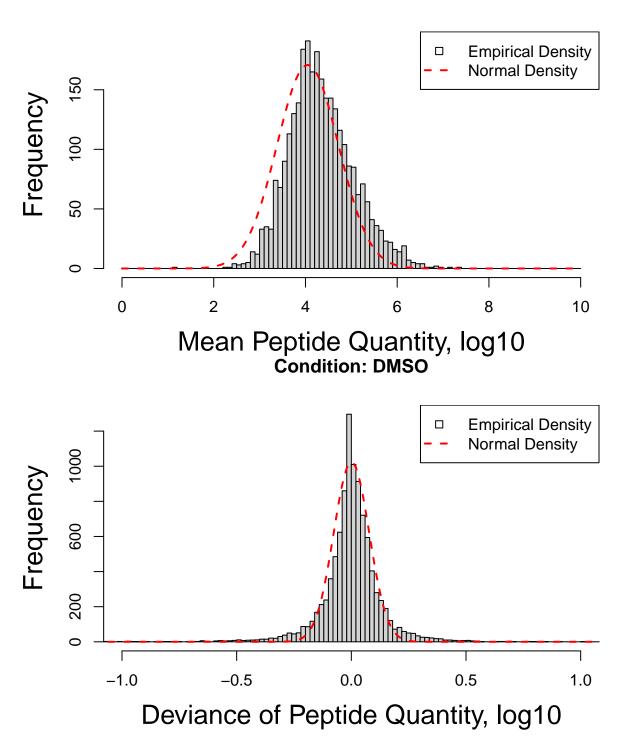


Sampling Distribution for Hierarchical Model

Distribution of precursor mean and deviance:

- precursor mean: all precursors averaged over replicates under fixed condition.
- precursor deviance: all precursors subtracted by the mean.

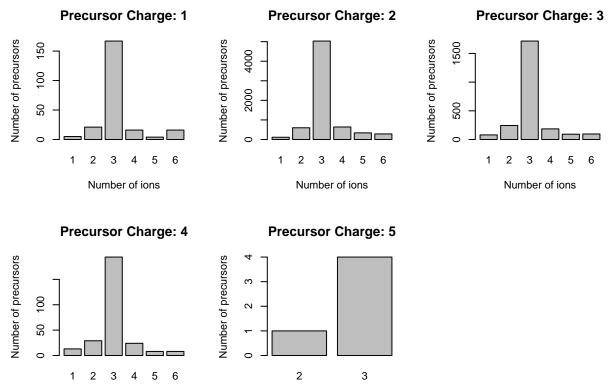
Condition: DMSO



Distribution of Ionization Efficiency

1. Distribution of the number of fragment ions of each precursor

• The number of fragment ions of each precursor is computed. The precursors are separated by their charges to inspect distribution specifically.



2. Calculate ionization efficiency distribution

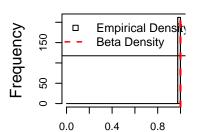
Number of ions

- A fragment ion quantity is calculated by its normalized peak area.
- The ionization efficiency of a fragment ion is the proportion of the fragment ion quantity out of the sum of the fragment ion quantities in its corresponding precursor.

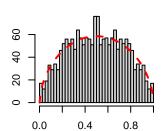
Number of ions

• For analysis, the precursors are separated by their numbers of fragment ions to inspect distribution specifically.

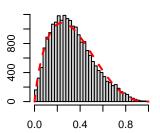
Precursors num ions: 1



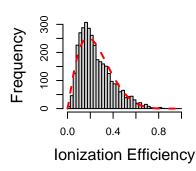
Precursors num ions: 2



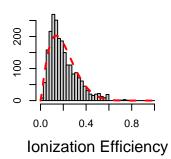
Precursors num ions: 3



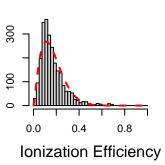
Precursors num ions: 4



Precursors num ions: 5

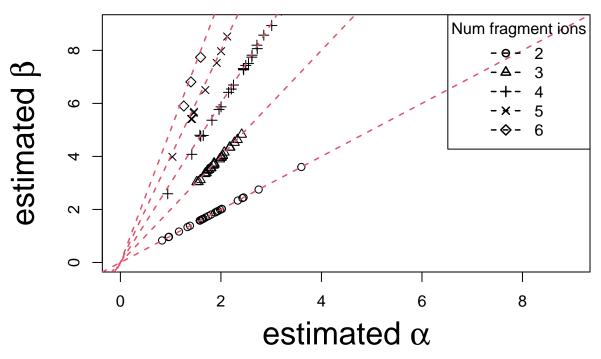


Precursors num ions: 6



- 3. Fit Beta distribution to an ionization efficiency distribution
- A beta distribution has two shape parameters, alpha and beta. The mean of a Beta distribution is equal to alpha / (alpha + beta).
- In this analysis, 30 proteins are randomly selected among the proteins having at least 30 peptides. For each protein, alpha and beta are estimated by ionization efficiency distribution of its fragment ions.

Estimated Beta shape parameters for ionization efficiency



- In the figure, each dot is a protein, and there are 30 points on each straight line.
- The dashed lines are y = x, y = 2x, ..., y = 5x.
- The figure shows that the estimated (alpha, beta) values agree with the dashed lines.
- 4. Correlation of Ionization Efficiency Between Conditions
- Investigate correlation between fragment ion quantities of two conditions with the same precursors, fragment ions, and replicate

