

Statistical Methods for Quantitative MS-based Proteomics: Part I. Preprocessing

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This is part of the online course [Proteomics Data Analysis \(PDA\)](#)

- [Playlist PDA Preprocessing](#)

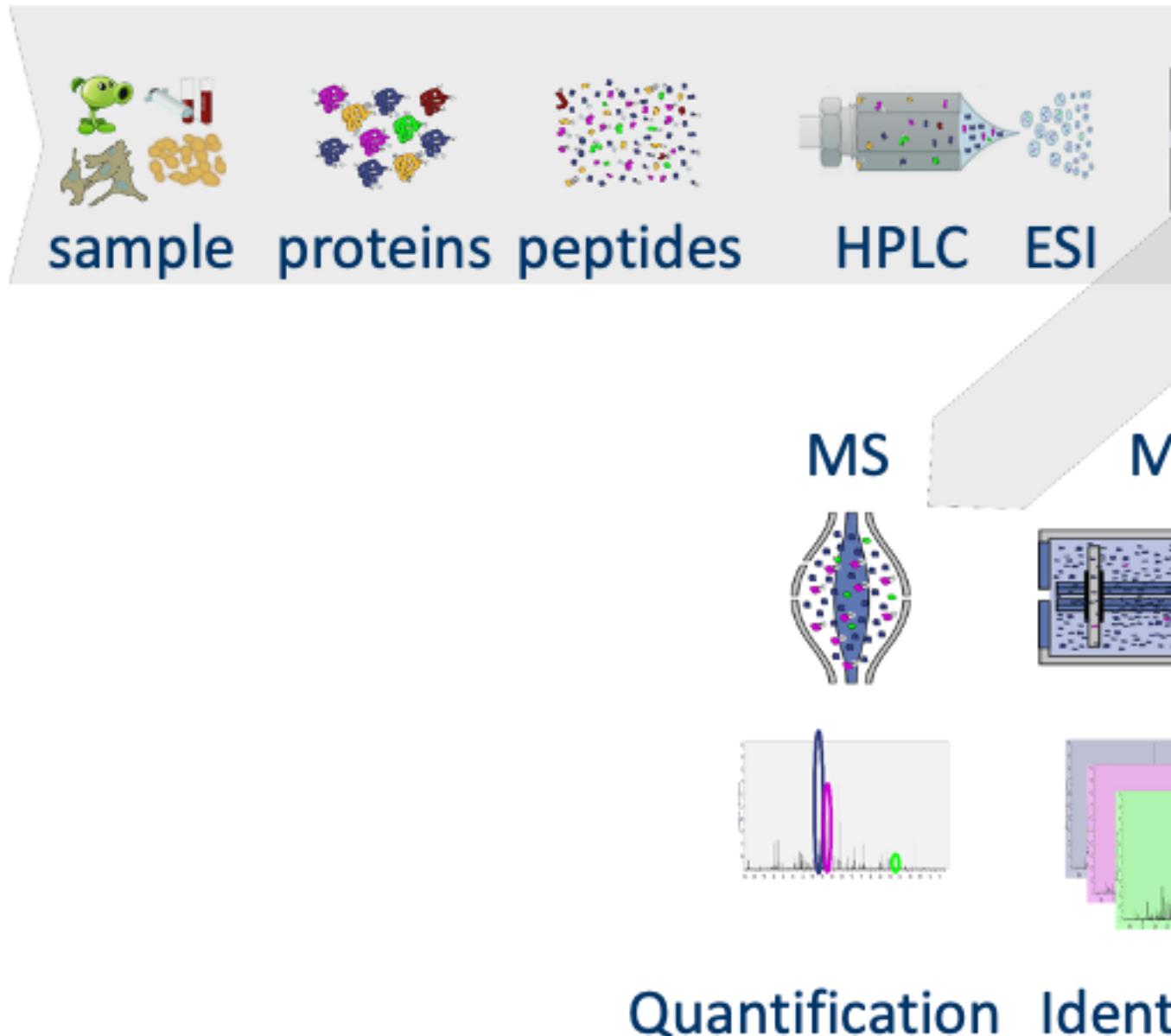
Outline

1. Introduction
2. Preprocessing
 - Log-transformation
 - Filtering
 - Normalization
 - Summarization

Note, that the R-code is included for learners who are aiming to develop R/markdown scripts to automate their quantitative proteomics data analyses. According to the target audience of the course we either work with a graphical user interface (GUI) in a R/shiny App msqrob2gui (e.g. Proteomics Bioinformatics course of the EBI and the Proteomics Data Analysis course at the Gulbenkian institute) or with R/markdowns scripts (e.g. Bioinformatics Summer School at UCLouvain or the Statistical Genomics Course at Ghent University).

1 Intro: Challenges in Label-Free Quantitative Proteomics

1.1 MS-based workflow



- Peptide Characteristics
 - Modifications
 - Ionisation Efficiency: huge variability
 - Identification

- * Misidentification → outliers
- * MS² selection on peptide abundance
- * Context depending missingness
- * Non-random missingness

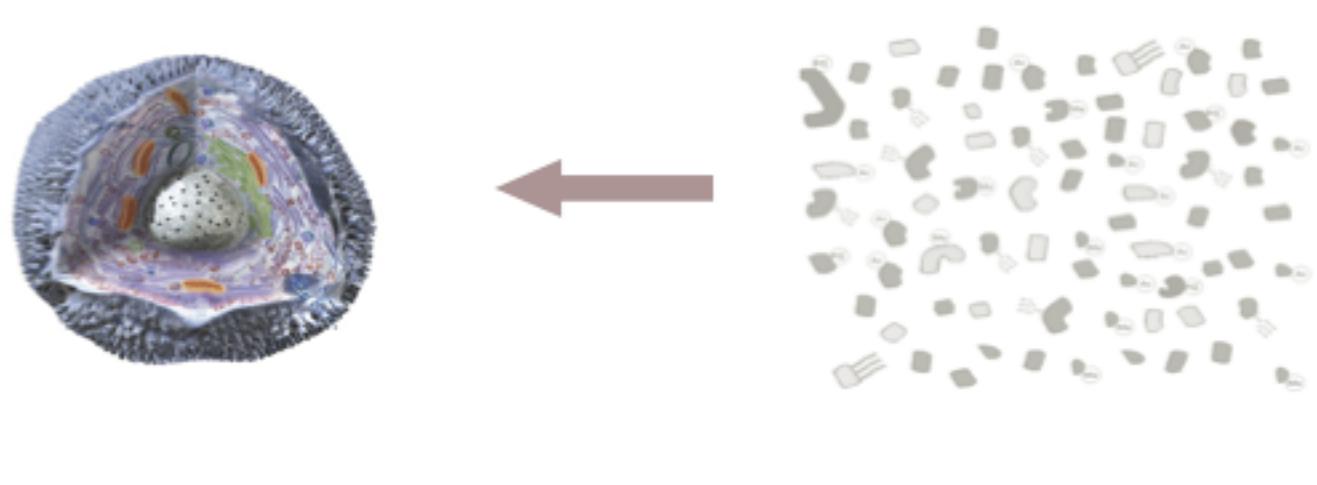
→ Unbalanced peptide identifications across samples and messy data

1.2 Level of quantification

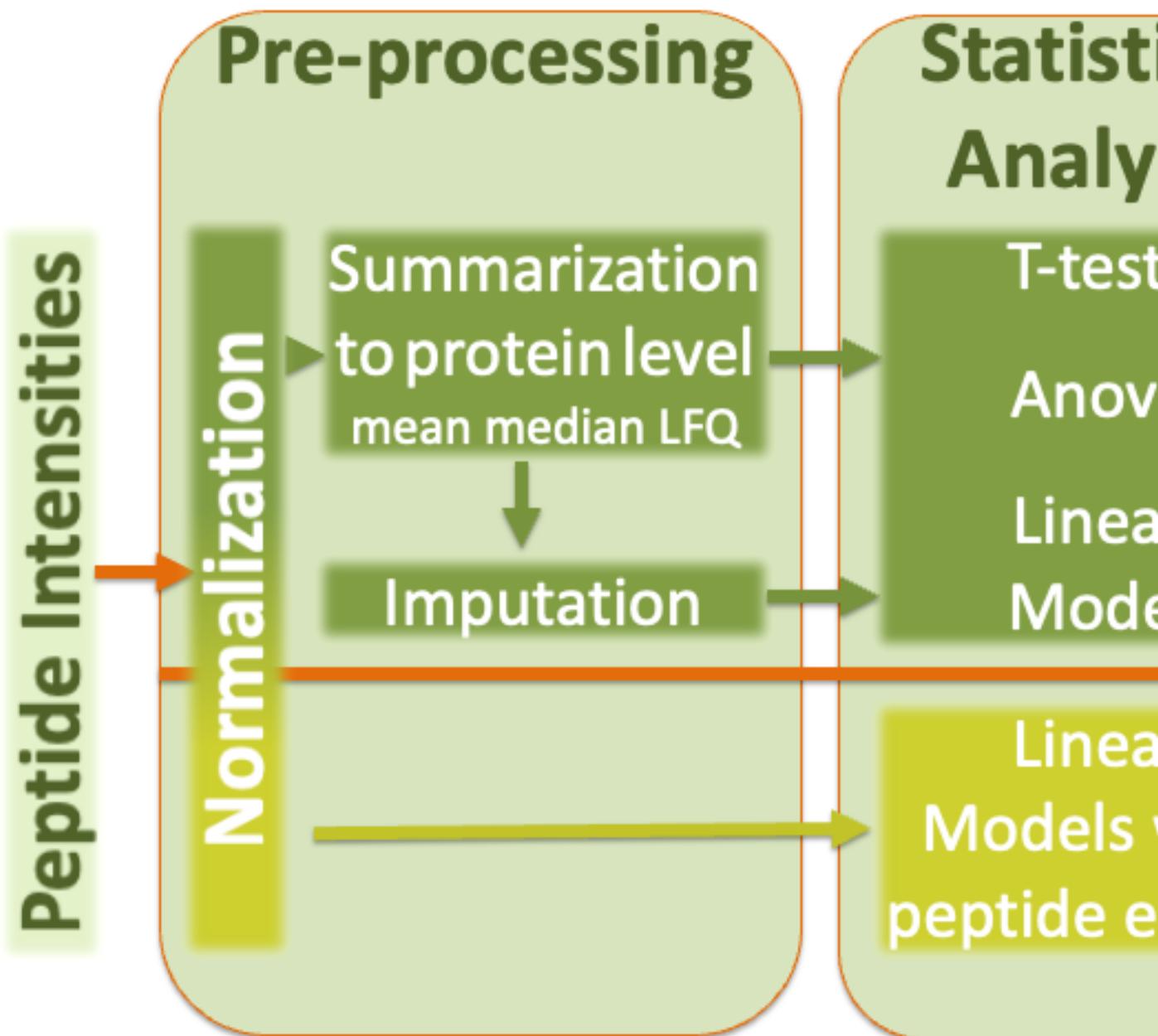
- MS-based proteomics returns peptides: pieces of proteins



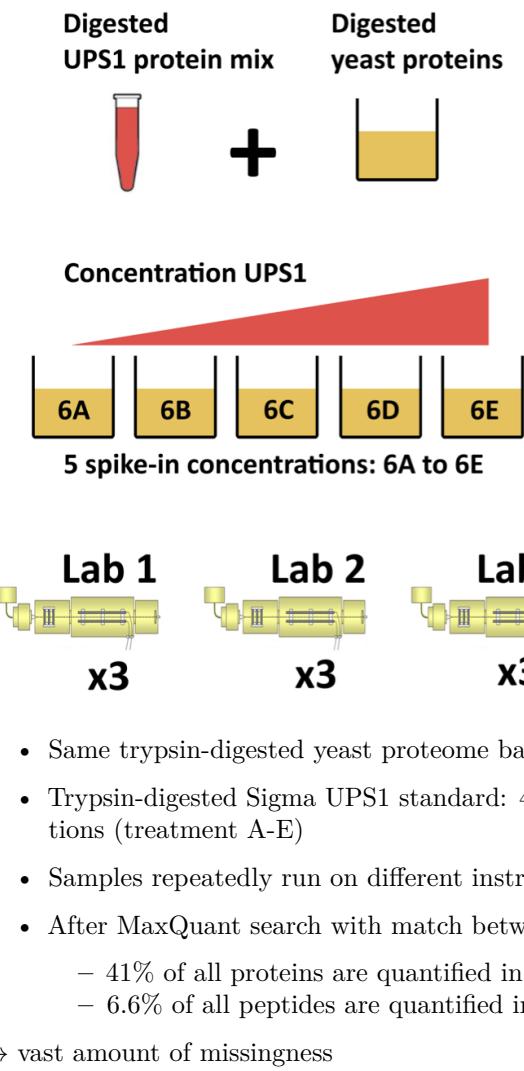
- Quantification commonly required on the protein level



1.3 Label-free Quantitative Proteomics Data Analysis Workflows



1.4 CPTAC Spike-in Study



1.5 Maxquant output

The screenshot shows a file browser window with a toolbar at the top featuring three colored circles (red, yellow, green) and navigation buttons (< >). Below the toolbar are four view mode icons: grid, list, columns, and horizontal list.

Favorites

- Google Drive
- iCloud Drive
- Desktop
- Documents
- Dropbox
- AirDrop
- All My Files
- iCloud Drive
- Applications
- Downloads
- Desktop

Name

Name
aifMsms.txt
allPeptides.txt
evidence.txt
libraryMatch.txt
matchedFeatures.txt
modificationSpots.txt
mqpar.xml
ms3Scans.txt
msms.txt
msmsScans.txt
msScans.txt
mzRange.txt
Oxidation (M)S
parameters.txt

2 Import the data in R

2.1 Data infrastructure

Click to see background on data infrastructure used in R to store proteomics data

- We use the `QFeatures` package that provides the infrastructure to
 - store,
 - process,
 - manipulate and
 - analyse quantitative data/features from mass spectrometry experiments.
- It is based on the `SummarizedExperiment` and `MultiAssayExperiment` classes.
- Assays in a `QFeatures` object have a hierarchical relation:
 - proteins are composed of peptides,
 - themselves produced by spectra
 - relations between assays are tracked and recorded throughout data processing

2.2 Import data in R

2.2.1 Load libraries

Click to see code

```
library(tidyverse)
library(limma)
library(QFeatures)
library(msqrrob2)
library(plotly)
library(ggplot2)
library(data.table)
```

2.2.2 Read data

Click to see background and code

1. We use a `peptides.txt` file from MS-data quantified with maxquant that contains MS1 intensities summarized at the peptide level.

```
peptidesTable <- fread("https://raw.githubusercontent.com/statOmics/PDA/data/quantification/fullCptacData")
int64 <- which(sapply(peptidesTable, class) == "integer64")
for (j in int64) peptidesTable[[j]] <- as.numeric(peptidesTable[[j]])
```

2. Maxquant stores the intensity data for the different samples in columns that start with Intensity. We can retrieve the column names with the intensity data with the code below:

```
quantCols <- grep("Intensity ", names(peptidesTable))
```

3. Read the data and store it in `QFeatures` object

```
pe <- readQFeatures(
  assayData = peptidesTable,
  fnames = 1,
  quantCols = quantCols,
  name = "peptideRaw")
```

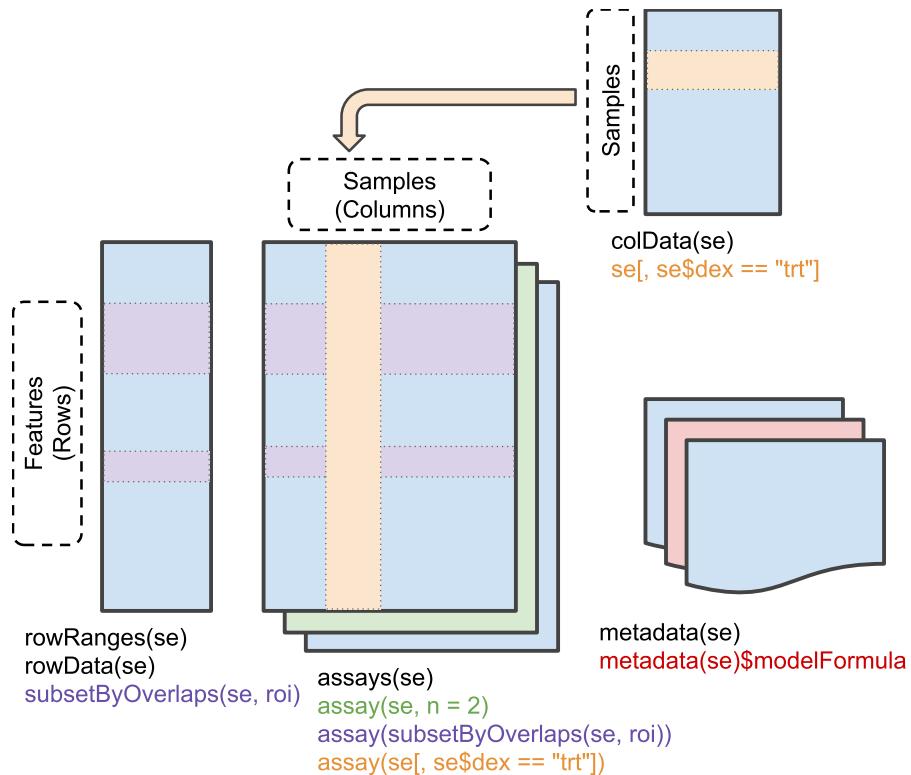


Figure 1: Conceptual representation of a ‘SummarizedExperiment’ object. Assays contain information on the measured omics features (rows) for different samples (columns). The ‘rowData’ contains information on the omics features, the ‘colData’ contains information on the samples, i.e. experimental design etc.

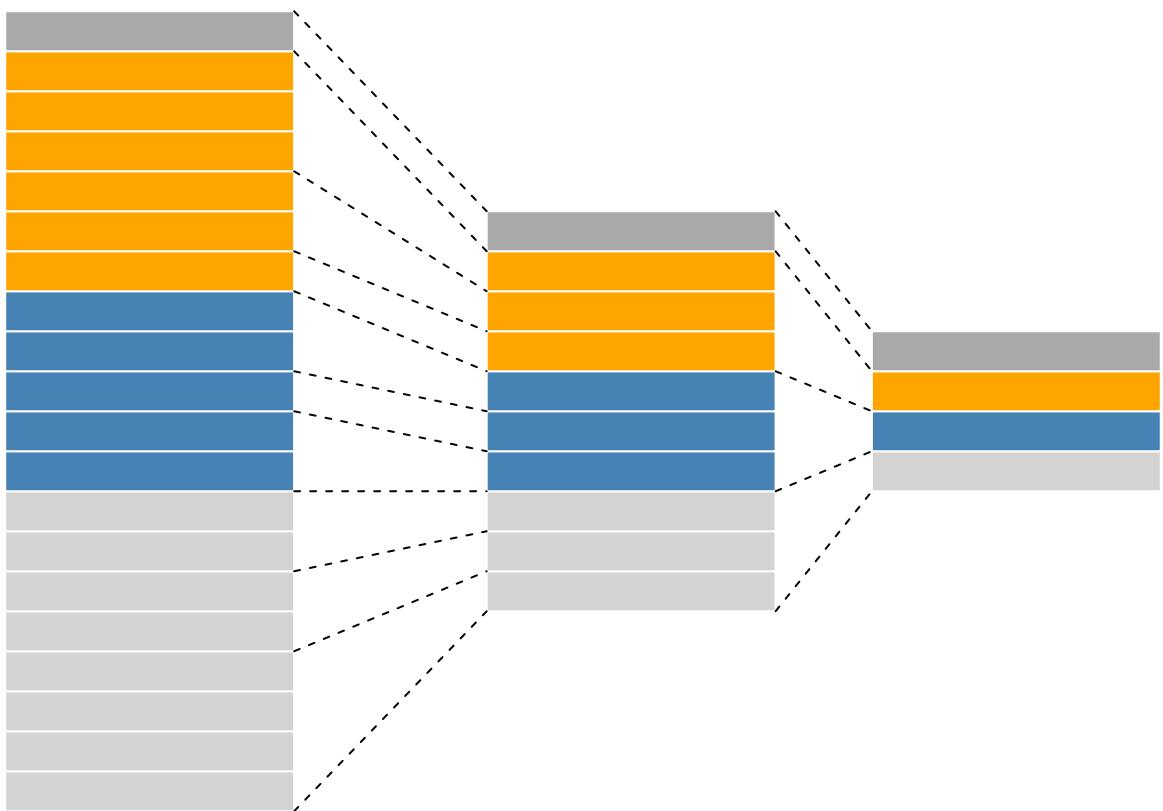


Figure 2: Conceptual representation of a **QFeatures** object and the aggregative relation between different assays.

```

## Checking arguments.

## Loading data as a 'SummarizedExperiment' object.

## Formatting sample annotations (colData).

## Formatting data as a 'QFeatures' object.

## Setting assay rownames.

rm(peptidesTable)
gc()

##           used (Mb) gc trigger (Mb) limit (Mb) max used (Mb)
## Ncells    8013617 428.0    11667507 623.2        NA 11667507 623.2
## Vcells   17741691 135.4    34178546 260.8      16384 34178546 260.8

gc()

##           used (Mb) gc trigger (Mb) limit (Mb) max used (Mb)
## Ncells    8013479 428.0    11667507 623.2        NA 11667507 623.2
## Vcells   17736841 135.4    34178546 260.8      16384 34178546 260.8

```

2.2.3 Explore object

Click to see background and code

- The rowData contains information on the features (peptides) in the assay. E.g. Sequence, protein, ...

```

rowData(pe[["peptideRaw"]])

## DataFrame with 11466 rows and 143 columns
##           Sequence N.term.cleavage.window C.term.cleavage.window
##           <character>          <character>          <character>
## AAAAGAGGAGDSDAVTK AAAAGAGGAG... EHQBHQDEQKAA... DSGDAVTKIG...
## AAAALAGGK         AAAALAGGK    QQQLSKAAKAA... AAALAGGKKS...
## AAAALAGGKK        AAAALAGGKK    QQQLSKAAKAA... AALAGGKKSK...
## AAADALSDLEIK     AAADALSDLE... MPKETPSKAA... ALSDLEIKDS...
## AAADALSDLEIKDSK AAADALSDLE... MPKETPSKAA... DLEIKDSKSN...
## ...
## YYSIYDLGNNAVGLAK YYSIYDLGN... VGDAFLRKYY... NNAVGLAKAI...
## YYTFNGPNYNENETIR YYTFNGPNYN... FKDGFSYPKYY... YNENETIRHI...
## YYTITEVATR        YYTITEVATR    QEWDINERYYY... TITEVATRAK...
## YYTVFDRDNNR       YYTVFDRDNN... LGDVFIGRYY... VFDRDNNRVG...
## YYTVFDRDNNRVGFAEAAR YYTVFDRDNN... LGDVFIGRYY... VGFAEAARL...
##           Amino.acid.before First.amino.acid Second.amino.acid
##           <character>          <character>          <character>
## AAAAGAGGAGDSDAVTK          K            A            A
## AAAALAGGK                 K            A            A
## AAAALAGGKK                K            A            A
## AAADALSDLEIK              K            A            A
## AAADALSDLEIKDSK           K            A            A
## ...
## YYSIYDLGNNAVGLAK          ...          ...          ...
## YYTFNGPNYNENETIR          K            Y            Y
## YYTITEVATR                 R            Y            Y
## YYTVFDRDNNR                  R            Y            Y
## YYTVFDRDNNRVGFAEAAR       R            Y            Y
##           Second.last.amino.acid Last.amino.acid Amino.acid.after

```

```

## <character> <character> <character>
## AAAAGAGGAGDSDGDAVTK T K I
## AAAALAGGK G K K
## AAAALAGGKK K K S
## AAADALSLEIK I K D
## AAADALSLEIKDSK S K S
## ... ...
## YYSIYDLGNNAVGLAK A K A
## YYTFNGPYNENETIR I R H
## YYTITEVATR T R A
## YYTVFDRDNNR N R V
## YYTVFDRDNNRVGFAEAAR A R L
## A.Count R.Count N.Count D.Count C.Count Q.Count
## <integer> <integer> <integer> <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 7 0 0 2 0 0
## AAAALAGGK 5 0 0 0 0 0
## AAAALAGGKK 5 0 0 0 0 0
## AAADALSLEIK 4 0 0 2 0 0
## AAADALSLEIKDSK 4 0 0 3 0 0
## ... ...
## YYSIYDLGNNAVGLAK 2 0 2 1 0 0
## YYTFNGPYNENETIR 0 1 4 0 0 0
## YYTITEVATR 1 1 0 0 0 0
## YYTVFDRDNNR 0 2 2 2 0 0
## YYTVFDRDNNRVGFAEAAR 3 3 2 2 0 0
## E.Count G.Count H.Count I.Count L.Count K.Count
## <integer> <integer> <integer> <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 0 5 0 0 0 1
## AAAALAGGK 0 2 0 0 1 1
## AAAALAGGKK 0 2 0 0 1 2
## AAADALSLEIK 1 0 0 1 2 1
## AAADALSLEIKDSK 1 0 0 1 2 2
## ... ...
## YYSIYDLGNNAVGLAK 0 2 0 1 2 1
## YYTFNGPYNENETIR 2 1 0 1 0 0
## YYTITEVATR 1 0 0 1 0 0
## YYTVFDRDNNR 0 0 0 0 0 0
## YYTVFDRDNNRVGFAEAAR 1 1 0 0 0 0
## M.Count F.Count P.Count S.Count T.Count W.Count
## <integer> <integer> <integer> <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 0 0 0 1 1 0
## AAAALAGGK 0 0 0 0 0 0
## AAAALAGGKK 0 0 0 0 0 0
## AAADALSLEIK 0 0 0 1 0 0
## AAADALSLEIKDSK 0 0 0 2 0 0
## ... ...
## YYSIYDLGNNAVGLAK 0 0 0 1 0 0
## YYTFNGPYNENETIR 0 1 1 0 2 0
## YYTITEVATR 0 0 0 0 3 0
## YYTVFDRDNNR 0 1 0 0 1 0
## YYTVFDRDNNRVGFAEAAR 0 2 0 0 1 0
## Y.Count V.Count U.Count Length Missed.cleavages
## <integer> <integer> <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 0 1 0 18 0

```

```

## AAAALAGGK          0      0      0      9      0
## AAAALAGGKK         0      0      0     10      1
## AAADALSDLEIK       0      0      0     12      0
## AAADALSDLEIKDSK    0      0      0     15      1
## ...
## YYSIYDLGNNAVGLAK   3      1      0     16      0
## YYTFNGPYNENETIR    3      0      0     16      0
## YYTITEVATR          2      1      0     10      0
## YYTVFDRDNNR         2      1      0     11      1
## YYTVFDRDNNRVGFAEAAR 2      2      0     19      2
##                         Mass      Proteins Leading.razor.protein
##                         <numeric> <character>      <character>
## AAAAGAGGAGDSDAVTK  1445.675 sp|P38915|...      sp|P38915|...
## AAAALAGGK           728.418 sp|Q3E792|...      sp|Q3E792|...
## AAAALAGGKK          856.513 sp|Q3E792|...      sp|Q3E792|...
## AAADALSDLEIK        1215.635 sp|P09938|...      sp|P09938|...
## AAADALSDLEIKDSK    1545.789 sp|P09938|...      sp|P09938|...
## ...
## YYSIYDLGNNAVGLAK   1759.88  sp|P07267|...      sp|P07267|...
## YYTFNGPYNENETIR    1993.88  sp|Q00955|...      sp|Q00955|...
## YYTITEVATR          1215.61  sp|P38891|...      sp|P38891|...
## YYTVFDRDNNR         1461.66  P07339ups|...      P07339ups|...
## YYTVFDRDNNRVGFAEAAR 2263.08  P07339ups|...      P07339ups|...
## Start.position End.position Unique..Groups.
##                         <integer> <integer> <character>
## AAAAGAGGAGDSDAVTK      97      114      yes
## AAAALAGGK                13      21      yes
## AAAALAGGKK                13      22      yes
## AAADALSDLEIK              9      20      yes
## AAADALSDLEIKDSK            9      23      yes
## ...
## YYSIYDLGNNAVGLAK        388      403      yes
## YYTFNGPYNENETIR         1275      1290      yes
## YYTITEVATR                311      320      yes
## YYTVFDRDNNR                 225      235      yes
## YYTVFDRDNNRVGFAEAAR      225      243      yes
## Unique..Proteins. Charges      PEP      Score
##                         <character> <character> <numeric> <numeric>
## AAAAGAGGAGDSDAVTK      yes      2 1.1843e-05  82.942
## AAAALAGGK                 no      2 7.4562e-06  134.810
## AAAALAGGKK                 no      2 3.3094e-09  143.730
## AAADALSDLEIK                yes      2 9.1593e-23  182.230
## AAADALSDLEIKDSK               yes      3 1.5319e-04  73.927
## ...
## YYSIYDLGNNAVGLAK        yes      2 7.7415e-37  174.240
## YYTFNGPYNENETIR         yes      2 4.2208e-21  147.750
## YYTITEVATR                  yes      2 1.3566e-04  109.160
## YYTVFDRDNNR                  yes      2 6.1425e-04  110.930
## YYTVFDRDNNRVGFAEAAR      yes      3 8.9859e-04  59.728
## Identification.type.6A_1 Identification.type.6A_2
##                         <character>      <character>
## AAAAGAGGAGDSDAVTK      By matchin...      By MS/MS
## AAAALAGGK                 By matchin...      By matchin...
## AAAALAGGKK                 By matchin...      By matchin...

```

## AAADALSDLEIK	By MS/MS	By MS/MS
## AAADALSDLEIKDSK	By matchin...	By matchin...
##
## YYSIYDLGNNAVGLAK	By matchin...	By matchin...
## YYTFNGPYNENETIR	By matchin...	By matchin...
## YYTITEVATR	By MS/MS	By matchin...
## YYTVFDRDNNR	By matchin...	By matchin...
## YYTVFDRDNNRVGFAEAAR	By matchin...	By matchin...
## Identification.type.6A_3	Identification.type.6A_4	
	<character>	<character>
## AAAAGAGGAGDSDAVTK	By matchin...	By MS/MS
## AAAALAGGK	By matchin...	By MS/MS
## AAAALAGGKK	By matchin...	By MS/MS
## AAADALSDLEIK	By matchin...	By MS/MS
## AAADALSDLEIKDSK	By matchin...	By MS/MS
##
## YYSIYDLGNNAVGLAK	By matchin...	By MS/MS
## YYTFNGPYNENETIR	By matchin...	By MS/MS
## YYTITEVATR	By matchin...	By matchin...
## YYTVFDRDNNR	By matchin...	By matchin...
## YYTVFDRDNNRVGFAEAAR	By matchin...	By matchin...
## Identification.type.6A_5	Identification.type.6A_6	
	<character>	<character>
## AAAAGAGGAGDSDAVTK	By matchin...	By matchin...
## AAAALAGGK	By matchin...	By matchin...
## AAAALAGGKK	By matchin...	By matchin...
## AAADALSDLEIK	By MS/MS	By MS/MS
## AAADALSDLEIKDSK	By MS/MS	By MS/MS
##
## YYSIYDLGNNAVGLAK	By MS/MS	By MS/MS
## YYTFNGPYNENETIR	By MS/MS	By MS/MS
## YYTITEVATR	By matchin...	By matchin...
## YYTVFDRDNNR	By matchin...	By matchin...
## YYTVFDRDNNRVGFAEAAR	By matchin...	By matchin...
## Identification.type.6A_7	Identification.type.6A_8	
	<character>	<character>
## AAAAGAGGAGDSDAVTK	By MS/MS	By MS/MS
## AAAALAGGK	By MS/MS	By MS/MS
## AAAALAGGKK	By MS/MS	By MS/MS
## AAADALSDLEIK	By MS/MS	By matchin...
## AAADALSDLEIKDSK	By MS/MS	By MS/MS
##
## YYSIYDLGNNAVGLAK	By matchin...	By matchin...
## YYTFNGPYNENETIR	By matchin...	By matchin...
## YYTITEVATR	By MS/MS	By matchin...
## YYTVFDRDNNR	By matchin...	By matchin...
## YYTVFDRDNNRVGFAEAAR	By matchin...	By matchin...
## Identification.type.6A_9	Identification.type.6B_1	
	<character>	<character>
## AAAAGAGGAGDSDAVTK	By MS/MS	By matchin...
## AAAALAGGK	By MS/MS	By MS/MS
## AAAALAGGKK	By MS/MS	By matchin...
## AAADALSDLEIK	By MS/MS	By MS/MS
## AAADALSDLEIKDSK	By MS/MS	By matchin...

```

## ...
## YYSIYDLGNNAVGLAK      ...      ...
## YYTFNGPYNENETIR       By matchin... By matchin...
## YYTITEVATR            By matchin... By matchin...
## YYTVFDRDNNR           By matchin... By MS/MS
## YYTVFDRDNNRVGFAEAAR  By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR  By matchin... By matchin...
## Identification.type.6B_2 Identification.type.6B_3
## <character>          <character>
## AAAAGAGGAGDSDGDAVTK  By matchin... By matchin...
## AAAALAGGK              By matchin... By matchin...
## AAAALAGGKK             By MS/MS    By MS/MS
## AAADALSLEIK            By MS/MS    By matchin...
## AAADALSDLEIKDSK        By matchin... By matchin...
## ...
## YYSIYDLGNNAVGLAK      ...      ...
## YYTFNGPYNENETIR       By matchin... By matchin...
## YYTITEVATR            By matchin... By matchin...
## YYTVFDRDNNR           By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR  By matchin... By matchin...
## Identification.type.6B_4 Identification.type.6B_5
## <character>          <character>
## AAAAGAGGAGDSDGDAVTK  By matchin... By matchin...
## AAAALAGGK              By matchin... By matchin...
## AAAALAGGKK             By matchin... By matchin...
## AAADALSLEIK            By MS/MS    By MS/MS
## AAADALSDLEIKDSK        By MS/MS    By MS/MS
## ...
## YYSIYDLGNNAVGLAK      ...      ...
## YYTFNGPYNENETIR       By MS/MS    By MS/MS
## YYTITEVATR            By MS/MS    By MS/MS
## YYTVFDRDNNR           By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR  By matchin... By matchin...
## Identification.type.6B_6 Identification.type.6B_7
## <character>          <character>
## AAAAGAGGAGDSDGDAVTK  By matchin... By matchin...
## AAAALAGGK              By matchin... By MS/MS
## AAAALAGGKK             By matchin... By MS/MS
## AAADALSLEIK            By MS/MS    By MS/MS
## AAADALSDLEIKDSK        By MS/MS    By MS/MS
## ...
## YYSIYDLGNNAVGLAK      ...      ...
## YYTFNGPYNENETIR       By matchin... By matchin...
## YYTITEVATR            By matchin... By matchin...
## YYTVFDRDNNR           By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR  By matchin... By matchin...
## Identification.type.6B_8 Identification.type.6B_9
## <character>          <character>
## AAAAGAGGAGDSDGDAVTK  By MS/MS    By MS/MS
## AAAALAGGK              By MS/MS    By MS/MS
## AAAALAGGKK             By MS/MS    By MS/MS
## AAADALSLEIK            By matchin... By matchin...
## AAADALSDLEIKDSK        By MS/MS    By MS/MS
## ...
## YYSIYDLGNNAVGLAK      ...      ...

```

```

## YYTFNGPYNENETIR By matchin... By matchin...
## YYTITEVATR By MS/MS By matchin...
## YYTVFDRDNNR By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin... By matchin...
## Identification.type.6C_1 Identification.type.6C_2
## <character> <character>
## AAAAGAGGAGDSDAVTK By matchin... By matchin...
## AAAALAGGK By matchin... By MS/MS
## AAAALAGGKK By matchin... By MS/MS
## AAADALSLEIK By MS/MS By matchin...
## AAADALSDEIKDSK By matchin... By matchin...
## ...
## YYSIYDLGNNAVGLAK By matchin... ...
## YYTFNGPYNENETIR By matchin... By matchin...
## YYTITEVATR By matchin... By matchin...
## YYTVFDRDNNR By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin... By matchin...
## Identification.type.6C_3 Identification.type.6C_4
## <character> <character>
## AAAAGAGGAGDSDAVTK By matchin... By matchin...
## AAAALAGGK By matchin... By MS/MS
## AAAALAGGKK By matchin... By MS/MS
## AAADALSLEIK By MS/MS By MS/MS
## AAADALSDEIKDSK By matchin... By MS/MS
## ...
## YYSIYDLGNNAVGLAK By matchin... ...
## YYTFNGPYNENETIR By matchin... By MS/MS
## YYTITEVATR By MS/MS By matchin...
## YYTVFDRDNNR By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin... By matchin...
## Identification.type.6C_5 Identification.type.6C_6
## <character> <character>
## AAAAGAGGAGDSDAVTK By MS/MS By matchin...
## AAAALAGGK By matchin... By matchin...
## AAAALAGGKK By matchin... By matchin...
## AAADALSLEIK By MS/MS By MS/MS
## AAADALSDEIKDSK By MS/MS By MS/MS
## ...
## YYSIYDLGNNAVGLAK By MS/MS ...
## YYTFNGPYNENETIR By matchin... By matchin...
## YYTITEVATR By matchin... By matchin...
## YYTVFDRDNNR By matchin... By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin... By matchin...
## Identification.type.6C_7 Identification.type.6C_8
## <character> <character>
## AAAAGAGGAGDSDAVTK By MS/MS By matchin...
## AAAALAGGK By MS/MS By MS/MS
## AAAALAGGKK By MS/MS By MS/MS
## AAADALSLEIK By matchin... By MS/MS
## AAADALSDEIKDSK By MS/MS By MS/MS
## ...
## YYSIYDLGNNAVGLAK By matchin... ...
## YYTFNGPYNENETIR By matchin... By matchin...
## YYTITEVATR By matchin... By MS/MS

```

```

## YYTVFDRDNNR          By matchin...          By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin...          By matchin...
##                               Identification.type.6C_9 Identification.type.6D_1
##                               <character>           <character>
## AAAAGAGGAGDSDGDAVTK By matchin...          By matchin...
## AAAALAGGK             By MS/MS            By matchin...
## AAAALAGGKK            By MS/MS            By matchin...
## AAADALSLEIK           By MS/MS            By MS/MS
## AAADALSDLEIKDSK      By MS/MS            By MS/MS
## ...
## YYSIYDLGNNAVGLAK     By matchin...
## YYTFNGPYNENETIR      By matchin...
## YYTITEVATR            By MS/MS            By matchin...
## YYTVFDRDNNR           By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin...
##                               Identification.type.6D_2 Identification.type.6D_3
##                               <character>           <character>
## AAAAGAGGAGDSDGDAVTK By matchin...          By matchin...
## AAAALAGGK              By matchin...          By matchin...
## AAAALAGGKK             By matchin...          By matchin...
## AAADALSLEIK           By matchin...          By matchin...
## AAADALSDLEIKDSK       By MS/MS            By matchin...
## ...
## YYSIYDLGNNAVGLAK     By matchin...
## YYTFNGPYNENETIR      By matchin...
## YYTITEVATR             By MS/MS            By MS/MS
## YYTVFDRDNNR           By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin...
##                               Identification.type.6D_4 Identification.type.6D_5
##                               <character>           <character>
## AAAAGAGGAGDSDGDAVTK By matchin...          By matchin...
## AAAALAGGK              By matchin...          By matchin...
## AAAALAGGKK             By MS/MS            By matchin...
## AAADALSLEIK           By MS/MS            By MS/MS
## AAADALSDLEIKDSK       By MS/MS            By MS/MS
## ...
## YYSIYDLGNNAVGLAK     By MS/MS            By MS/MS
## YYTFNGPYNENETIR      By MS/MS            By MS/MS
## YYTITEVATR            By matchin...
## YYTVFDRDNNR           By matchin...
## YYTVFDRDNNRVGFAEAAR By matchin...
##                               Identification.type.6D_6 Identification.type.6D_7
##                               <character>           <character>
## AAAAGAGGAGDSDGDAVTK By MS/MS            By matchin...
## AAAALAGGK              By matchin...          By MS/MS
## AAAALAGGKK             By matchin...          By MS/MS
## AAADALSLEIK           By MS/MS            By matchin...
## AAADALSDLEIKDSK       By matchin...          By MS/MS
## ...
## YYSIYDLGNNAVGLAK     By MS/MS            By matchin...
## YYTFNGPYNENETIR      By MS/MS            By matchin...
## YYTITEVATR            By matchin...
## YYTVFDRDNNR           By matchin...          By MS/MS
## YYTVFDRDNNRVGFAEAAR By matchin...

```

```

## Identification.type.6D_8 Identification.type.6D_9
## <character> <character>
## AAAAGAGGAGDSGDAVTK By matchin... By matchin...
## AAAALAGGK By MS/MS By MS/MS
## AAAALAGGKK By MS/MS By MS/MS
## AAADALSLEIK By MS/MS By MS/MS
## AAADALSLEIKDSK By MS/MS By MS/MS
## ...
## YYSIYDLGNNAVGLAK By matchin... ...
## YYTFNGPNYNENETIR By matchin... By matchin...
## YYTITEVATR By MS/MS By matchin...
## YYTVFDRDNNR By MS/MS By matchin...
## YYTVFDRDNNRVGFEEAAR By matchin... By matchin...
## Identification.type.6E_1 Identification.type.6E_2
## <character> <character>
## AAAAGAGGAGDSGDAVTK By matchin... By matchin...
## AAAALAGGK By matchin... By matchin...
## AAAALAGGKK By matchin... By matchin...
## AAADALSLEIK By MS/MS By MS/MS
## AAADALSLEIKDSK By MS/MS By MS/MS
## ...
## YYSIYDLGNNAVGLAK By matchin... ...
## YYTFNGPNYNENETIR By matchin... By matchin...
## YYTITEVATR By matchin... By matchin...
## YYTVFDRDNNR By matchin... By matchin...
## YYTVFDRDNNRVGFEEAAR By matchin... By matchin...
## Identification.type.6E_3 Identification.type.6E_4
## <character> <character>
## AAAAGAGGAGDSGDAVTK By matchin... By matchin...
## AAAALAGGK By matchin... By MS/MS
## AAAALAGGKK By matchin... By matchin...
## AAADALSLEIK By matchin... By MS/MS
## AAADALSLEIKDSK By MS/MS By matchin...
## ...
## YYSIYDLGNNAVGLAK By matchin... ...
## YYTFNGPNYNENETIR By matchin... By MS/MS
## YYTITEVATR By matchin... By matchin...
## YYTVFDRDNNR By matchin... By MS/MS
## YYTVFDRDNNRVGFEEAAR By matchin... By matchin...
## Identification.type.6E_5 Identification.type.6E_6
## <character> <character>
## AAAAGAGGAGDSGDAVTK By matchin... By matchin...
## AAAALAGGK By matchin... By matchin...
## AAAALAGGKK By matchin... By matchin...
## AAADALSLEIK By MS/MS By matchin...
## AAADALSLEIKDSK By MS/MS By MS/MS
## ...
## YYSIYDLGNNAVGLAK By MS/MS ...
## YYTFNGPNYNENETIR By MS/MS By MS/MS
## YYTITEVATR By matchin... By matchin...
## YYTVFDRDNNR By MS/MS By matchin...
## YYTVFDRDNNRVGFEEAAR By matchin... By MS/MS
## Identification.type.6E_7 Identification.type.6E_8
## <character> <character>

```

```

## AAAAGAGGAGDSDGDAVTK By matchin... By matchin...
## AAAALAGGK By MS/MS By MS/MS
## AAAALAGGKK By MS/MS By MS/MS
## AAADALSDLEIK By MS/MS By MS/MS
## AAADALSDLEIKDSK By matchin... By MS/MS
## ... ...
## YYSIYDLGNNAVGLAK By matchin... By matchin...
## YYTFNGPYNENETIR By matchin... By matchin...
## YYTITEVATR By matchin... By matchin...
## YYTVFDRDNNR By MS/MS By MS/MS
## YYTVFDRDNNRVGFEEAAR By matchin... By matchin...
## Identification.type.6E_9 Experiment.6A_1 Experiment.6A_2
## <character> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK By matchin... NA 1
## AAAALAGGK By MS/MS NA 1
## AAAALAGGKK By MS/MS NA 1
## AAADALSDLEIK By MS/MS 1 1
## AAADALSDLEIKDSK By MS/MS 1 1
## ... ...
## YYSIYDLGNNAVGLAK By matchin... NA NA
## YYTFNGPYNENETIR By MS/MS NA NA
## YYTITEVATR By matchin... 1 NA
## YYTVFDRDNNR By MS/MS NA NA
## YYTVFDRDNNRVGFEEAAR By matchin... NA NA
## Experiment.6A_3 Experiment.6A_4 Experiment.6A_5
## <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK NA 1 1
## AAAALAGGK 2 1 1
## AAAALAGGKK NA 1 NA
## AAADALSDLEIK 1 1 1
## AAADALSDLEIKDSK NA 1 1
## ... ...
## YYSIYDLGNNAVGLAK NA 1 1
## YYTFNGPYNENETIR NA 1 1
## YYTITEVATR 1 NA NA
## YYTVFDRDNNR NA NA NA
## YYTVFDRDNNRVGFEEAAR NA NA NA
## Experiment.6A_6 Experiment.6A_7 Experiment.6A_8
## <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 1 1 1
## AAAALAGGK 1 2 1
## AAAALAGGKK 1 1 1
## AAADALSDLEIK 1 1 1
## AAADALSDLEIKDSK 1 1 1
## ... ...
## YYSIYDLGNNAVGLAK 1 NA NA
## YYTFNGPYNENETIR 1 1 NA
## YYTITEVATR 1 1 NA
## YYTVFDRDNNR NA NA NA
## YYTVFDRDNNRVGFEEAAR NA NA NA
## Experiment.6A_9 Experiment.6B_1 Experiment.6B_2
## <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 1 NA NA
## AAAALAGGK 1 1 1

```

```

## AAAALAGGKK 1 NA 1
## AAADALSDLEIK 1 1 NA 1
## AAADALSDLEIKDSK 1 NA NA 1
## ... ...
## YYSIYDLGNNAVGLAK NA NA NA NA
## YYTFNGPNYNENETIR 1 NA NA NA
## YYTITEVATR NA 1 1 1
## YYTVFDRDNNR NA NA NA NA
## YYTVFDRDNNRVGFAEAAR NA NA NA NA
## Experiment.6B_3 Experiment.6B_4 Experiment.6B_5
## <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK NA NA 1
## AAAALAGGK 1 2 1
## AAAALAGGKK 1 1 NA
## AAADALSDLEIK 1 1 1
## AAADALSDLEIKDSK NA 1 1
## ... ...
## YYSIYDLGNNAVGLAK NA 1 1 1
## YYTFNGPNYNENETIR NA 1 1 1
## YYTITEVATR 1 1 1 1
## YYTVFDRDNNR NA NA NA NA
## YYTVFDRDNNRVGFAEAAR NA NA NA NA
## Experiment.6B_6 Experiment.6B_7 Experiment.6B_8
## <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 1 NA 1
## AAAALAGGK NA 2 1
## AAAALAGGKK NA 1 1
## AAADALSDLEIK 1 1 1
## AAADALSDLEIKDSK 1 1 1
## ... ...
## YYSIYDLGNNAVGLAK 1 NA NA NA
## YYTFNGPNYNENETIR 1 1 NA NA
## YYTITEVATR 1 NA 1 1
## YYTVFDRDNNR NA NA NA NA
## YYTVFDRDNNRVGFAEAAR NA NA NA NA
## Experiment.6B_9 Experiment.6C_1 Experiment.6C_2
## <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK 1 NA NA
## AAAALAGGK 2 NA 1
## AAAALAGGKK 1 NA 1
## AAADALSDLEIK 1 1 1
## AAADALSDLEIKDSK 1 1 1
## ... ...
## YYSIYDLGNNAVGLAK NA NA NA NA
## YYTFNGPNYNENETIR NA NA NA NA
## YYTITEVATR NA 1 1 1
## YYTVFDRDNNR NA NA NA NA
## YYTVFDRDNNRVGFAEAAR NA NA NA NA
## Experiment.6C_3 Experiment.6C_4 Experiment.6C_5
## <integer> <integer> <integer>
## AAAAGAGGAGDSDGDAVTK NA 1 1
## AAAALAGGK 2 2 NA
## AAAALAGGKK NA 1 NA
## AAADALSDLEIK 1 1 1

```

## AAADALSDLEIKDSK	1	1	1
##
## YYSIYDLGNNAVGLAK	NA	1	1
## YYTFNGPYNENETIR	NA	1	1
## YYTITEVATR	1	1	NA
## YYTVFDRDNNR	NA	NA	NA
## YYTVFDRDNNRVGFAEAAR	NA	NA	NA
	Experiment.6C_6	Experiment.6C_7	Experiment.6C_8
	<integer>	<integer>	<integer>
## AAAAGAGGAGDSDAVTK	1	1	1
## AAAALAGGK	NA	2	1
## AAAALAGGKK	NA	1	1
## AAADALSDLEIK	1	1	1
## AAADALSDLEIKDSK	1	1	1
##
## YYSIYDLGNNAVGLAK	1	NA	NA
## YYTFNGPYNENETIR	1	1	1
## YYTITEVATR	1	NA	1
## YYTVFDRDNNR	1	NA	1
## YYTVFDRDNNRVGFAEAAR	NA	NA	NA
	Experiment.6C_9	Experiment.6D_1	Experiment.6D_2
	<integer>	<integer>	<integer>
## AAAAGAGGAGDSDAVTK	1	NA	NA
## AAAALAGGK	1	NA	1
## AAAALAGGKK	1	NA	NA
## AAADALSDLEIK	1	1	1
## AAADALSDLEIKDSK	1	1	1
##
## YYSIYDLGNNAVGLAK	NA	NA	NA
## YYTFNGPYNENETIR	1	NA	NA
## YYTITEVATR	1	NA	1
## YYTVFDRDNNR	NA	NA	NA
## YYTVFDRDNNRVGFAEAAR	NA	NA	NA
	Experiment.6D_3	Experiment.6D_4	Experiment.6D_5
	<integer>	<integer>	<integer>
## AAAAGAGGAGDSDAVTK	NA	1	1
## AAAALAGGK	1	1	1
## AAAALAGGKK	NA	1	NA
## AAADALSDLEIK	1	1	1
## AAADALSDLEIKDSK	1	1	1
##
## YYSIYDLGNNAVGLAK	NA	1	1
## YYTFNGPYNENETIR	NA	1	1
## YYTITEVATR	1	1	1
## YYTVFDRDNNR	NA	1	1
## YYTVFDRDNNRVGFAEAAR	NA	1	NA
	Experiment.6D_6	Experiment.6D_7	Experiment.6D_8
	<integer>	<integer>	<integer>
## AAAAGAGGAGDSDAVTK	1	1	NA
## AAAALAGGK	NA	2	1
## AAAALAGGKK	NA	1	1
## AAADALSDLEIK	1	1	1
## AAADALSDLEIKDSK	1	1	1
##

```

## YYSIYDLGNNAVGLAK           1           1           NA
## YYTFNGPYNENETIR            1           1           1
## YYTITEVATR                  1           NA          1
## YYTVFDRDNNR                 1           1           1
## YYTVFDRDNNRVGFAEAAR        NA          NA          NA
##                               Experiment.6D_9 Experiment.6E_1 Experiment.6E_2
##                               <integer>    <integer>    <integer>
## AAAAGAGGAGDSDGDAVTK        NA          NA          1
## AAAALAGGK                   2           NA          1
## AAAALAGGKK                  1           NA          NA
## AAADALSDLEIK                1           1           1
## AAADALSDLEIKDSK             1           1           1
## ...
## YYSIYDLGNNAVGLAK           NA          NA          NA
## YYTFNGPYNENETIR            1           NA          NA
## YYTITEVATR                  NA          NA          1
## YYTVFDRDNNR                 1           1           NA
## YYTVFDRDNNRVGFAEAAR        NA          NA          NA
##                               Experiment.6E_3 Experiment.6E_4 Experiment.6E_5
##                               <integer>    <integer>    <integer>
## AAAAGAGGAGDSDGDAVTK        NA          NA          1
## AAAALAGGK                   2           2           1
## AAAALAGGKK                  NA          1           NA
## AAADALSDLEIK                1           1           1
## AAADALSDLEIKDSK             1           1           1
## ...
## YYSIYDLGNNAVGLAK           1           1           1
## YYTFNGPYNENETIR            NA          1           1
## YYTITEVATR                  1           1           1
## YYTVFDRDNNR                 1           1           1
## YYTVFDRDNNRVGFAEAAR        NA          1           1
##                               Experiment.6E_6 Experiment.6E_7 Experiment.6E_8
##                               <integer>    <integer>    <integer>
## AAAAGAGGAGDSDGDAVTK        1           NA          NA
## AAAALAGGK                   NA          2           2
## AAAALAGGKK                  NA          1           1
## AAADALSDLEIK                1           1           1
## AAADALSDLEIKDSK             1           NA          1
## ...
## YYSIYDLGNNAVGLAK           1           NA          NA
## YYTFNGPYNENETIR            1           1           1
## YYTITEVATR                  NA          NA          NA
## YYTVFDRDNNR                 1           1           1
## YYTVFDRDNNRVGFAEAAR        1           1           1
##                               Experiment.6E_9 Intensity     Reverse Potential.contaminant
##                               <integer> <numeric> <character>      <character>
## AAAAGAGGAGDSDGDAVTK        NA          1190800
## AAAALAGGK                   1          280990000
## AAAALAGGKK                  1          33360000
## AAADALSDLEIK                1          54622000
## AAADALSDLEIKDSK             1          18910000
## ...
## YYSIYDLGNNAVGLAK           NA          2145900
## YYTFNGPYNENETIR            1          5608800

```

```

## YYTITEVATR NA 13034000
## YYTVFDRDNNR 1 8702500
## YYTVFDRDNNRVGFAEAAR 1 2391100
## id Protein.group.IDs Mod..peptide.IDs Evidence.IDs
## <integer> <character> <character> <character>
## AAAAGAGGAGDSDAVTK 0 859 0 0;1;2;3;4;...
## AAAALAGGK 1 230 1 24;25;26;2...
## AAAALAGGKK 2 230 2 74;75;76;7...
## AAADALSDLEIK 3 229 3 99;100;101...
## AAADALSDLEIKDSK 4 229 4 144;145;14...
## ...
## YYSIYDLGNNAVGLAK 11461 196 12240 331367;331...
## YYTFNGPNYNENETIR 11462 1254 12241 331384;331...
## YYTITEVATR 11463 854 12242 331411;331...
## YYTVFDRDNNR 11464 34 12243 331439;331...
## YYTVFDRDNNRVGFAEAAR 11465 34 12244 331455;331...
## MS.MS.IDs Best.MS.MS Oxidation..M..site.IDs MS.MS.Count
## <character> <integer> <character> <integer>
## AAAAGAGGAGDSDAVTK 0;1;2;3;4;... 0 10
## AAAALAGGK 10;11;12;1... 21 18
## AAAALAGGKK 30;31;32;3... 31 21
## AAADALSDLEIK 51;52;53;5... 72 29
## AAADALSDLEIKDSK 85;86;87;8... 94 32
## ...
## YYSIYDLGNNAVGLAK 169138;169... 169147 ...
## YYTFNGPNYNENETIR 169151;169... 169159 ...
## YYTITEVATR 169165;169... 169173 ...
## YYTVFDRDNNR 169177;169... 169180 7
## YYTVFDRDNNRVGFAEAAR 169184 169184 1

```

- The colData contains information on the samples

```
colData(pe)
```

```
## DataFrame with 45 rows and 0 columns
```

- No information is stored yet on the design.

```
pe |> colnames()
```

```
## CharacterList of length 1
## [["peptideRaw"]] Intensity 6A_1 Intensity 6A_2 ... Intensity 6E_9
```

- Note, that the sample names include the spike-in condition.
- They also end on a number.
 - 1-3 is from lab 1,
 - 4-6 from lab 2 and
 - 7-9 from lab 3.

- We update the colData with information on the design

```
colData(pe)$lab <- rep(
  rep(
    paste0("lab", 1:3),
    each=3), 5) |>
  as.factor()
```

```

colData(pe)$condition <- pe[["peptideRaw"]] |>
  colnames() |>
  substr(12,12) |>
  as.factor()

colData(pe)$spikeConcentration <- rep(
  c(A = 0.25, B = 0.74, C = 2.22, D = 6.67, E = 20),
  each = 9)

```

- We explore the colData again

```

colData(pe)

## DataFrame with 45 rows and 3 columns
##           lab condition spikeConcentration
##       <factor> <factor>     <numeric>
## Intensity 6A_1   lab1      A         0.25
## Intensity 6A_2   lab1      A         0.25
## Intensity 6A_3   lab1      A         0.25
## Intensity 6A_4   lab2      A         0.25
## Intensity 6A_5   lab2      A         0.25
## ...          ...
## Intensity 6E_5   lab2      E         20
## Intensity 6E_6   lab2      E         20
## Intensity 6E_7   lab3      E         20
## Intensity 6E_8   lab3      E         20
## Intensity 6E_9   lab3      E         20

```

3 Preprocessing

3.1 Log-transformation

3.1.1 Explore the data with plots

Peptide AALEELVK from spiked-in UPS protein P12081. We only show data from lab1.

Click to see code to make plot

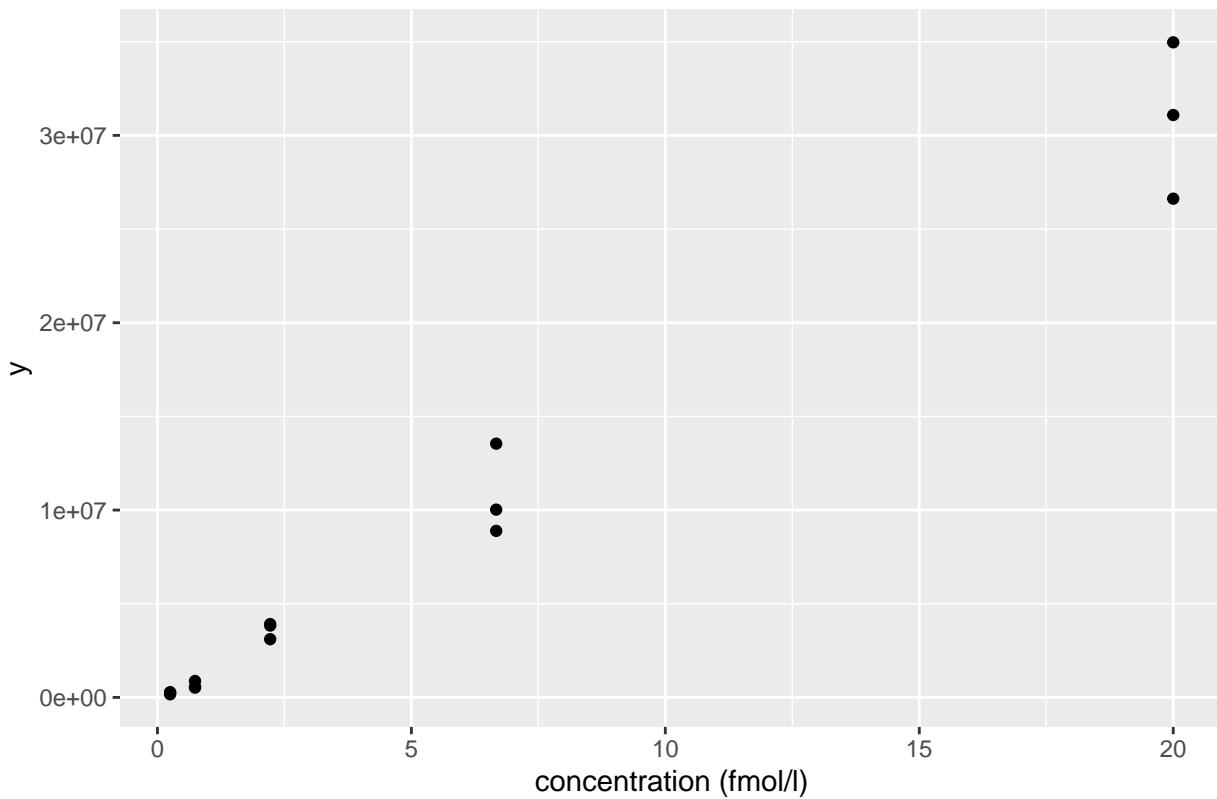
```

subset <- pe[["AALEELVK", colData(pe)$lab=="lab1"]
plotWhyLog <- data.frame(concentration = colData(subset)$spikeConcentration,
                           y = assay(subset[["peptideRaw"]]) |> c()
                           ) |>
  ggplot(aes(concentration, y)) +
  geom_point() +
  xlab("concentration (fmol/l)") +
  ggtitle("peptide AALEELVK in lab1")

```

```
plotWhyLog
```

peptide AALEELVK in lab1



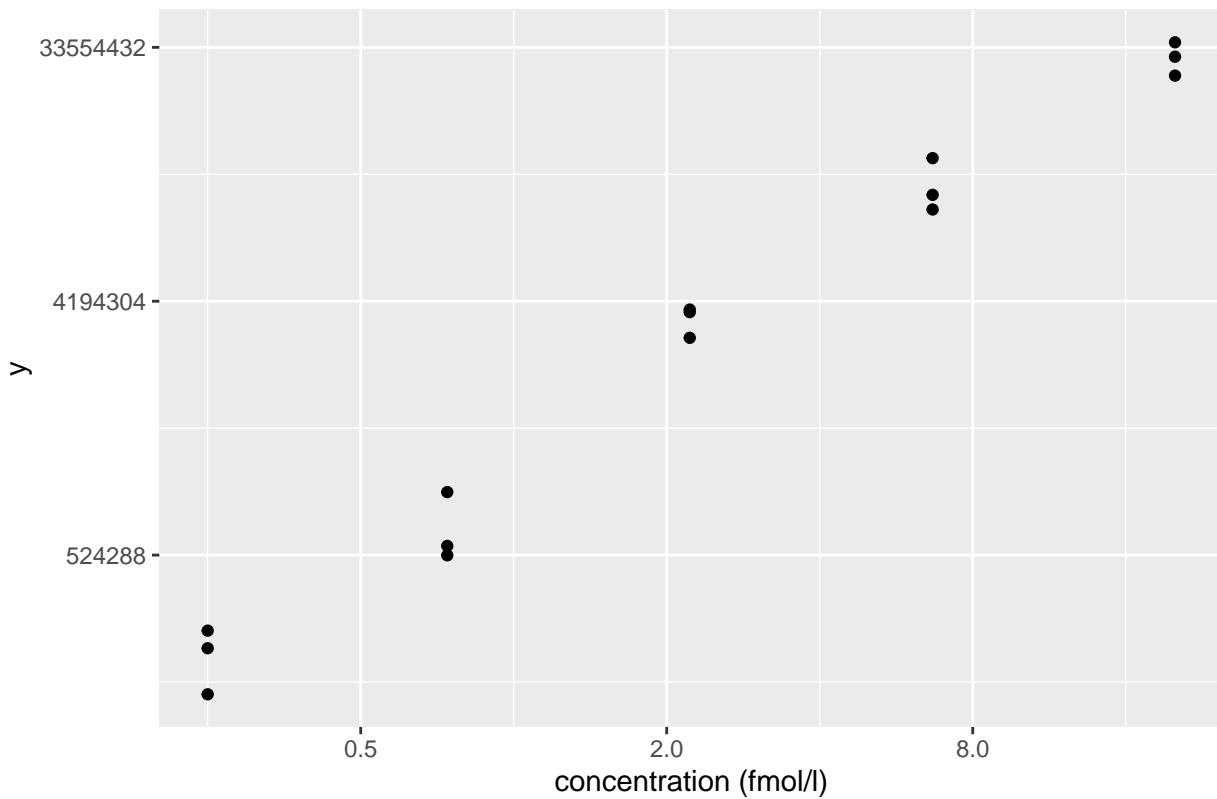
- Variance increases with the mean → Multiplicative error structure

Click to see code to make plot

```
plotLog <- data.frame(concentration = colData(subset)$spikeConcentration,
                      y = assay(subset[["peptideRaw"]]) |> c()
                     ) |>
  ggplot(aes(concentration, y)) +
  geom_point() +
  scale_x_continuous(trans='log2') +
  scale_y_continuous(trans='log2') +
  xlab("concentration (fmol/l)") +
  ggtitle("peptide AALEELVK in lab1 with axes on log scale")
```

```
plotLog
```

peptide AALEELVK in lab1 with axes on log scale



- Data seems to be homoscedastic on log-scale → log transformation of the intensity data
- In quantitative proteomics analysis on \log_2

→ Differences on a \log_2 scale: \log_2 fold changes

$$\log_2 B - \log_2 A = \log_2 \frac{B}{A} = \log FC_B - A$$

$$\begin{aligned}\log_2 FC = 1 &\rightarrow FC = 2^1 = 2 \\ \log_2 FC = 2 &\rightarrow FC = 2^2 = 4\end{aligned}$$

3.1.2 log-transformation of the data

Click to see code to log-transfrom the data

- Peptides with zero intensities are missing peptides and should be represent with a NA value rather than 0.

```
pe <- zeroIsNA(pe, "peptideRaw") # convert 0 to NA
```

- Logtransform data with base 2

```
pe <- logTransform(pe, base = 2, i = "peptideRaw", name = "peptideLog")
```

3.2 Filtering

- Reverse sequences
- Only identified by modification site (only modified peptides detected)

- Razor peptides: non-unique peptides assigned to the protein group with the most other peptides
- Contaminants
- Peptides few identifications
- Proteins that are only identified with one or a few peptides
- FDR of identification
- ...

Filtering does not induce bias if the criterion is independent from the downstream data analysis!

Click to see code to filter the data

1. Remove peptides that map to multiple proteins

We remove PSMs that could not be mapped to a protein or that map to multiple proteins (the protein identifier contains multiple identifiers separated by a ;).

```
pe <- filterFeatures(
  pe, ~ Proteins != "" & ## Remove failed protein inference
  !grepl(";", Proteins)) ## Remove protein groups

## 'Proteins' found in 2 out of 2 assay(s).
```

2. Remove reverse sequences (decoys) and contaminants

We now remove the contaminants, peptides that map to decoy sequences, and proteins which were only identified by peptides with modifications.

```
pe <- filterFeatures(pe, ~Reverse != "+")

## 'Reverse' found in 2 out of 2 assay(s).

pe <- filterFeatures(pe, ~ Potential.contaminant != "+")

## 'Potential.contaminant' found in 2 out of 2 assay(s).
```

3. Drop peptides that were only identified in one sample

We keep peptides that were observed at least three times. We tolerate the following proportion of NAs: pNA = (n-3)/n.

```
n0bs <- 3
n <- ncol(pe[["peptideLog"]])
pNA <- (n-n0bs)/n
pe <- filterNA(pe, pNA = pNA, i = "peptideLog")
nrow(pe[["peptideLog"]])

## [1] 10091
```

We keep 10091 peptides upon filtering.

3.3 Normalization

Click to see code to make plot

```
densityConditionD <- pe[["peptideLog"]][, colData(pe)$condition == "D"] |>
  assay() |>
  as.data.frame() |>
  gather(sample, intensity) |>
  mutate(lab = colData(pe)[sample, "lab"]) |>
  ggplot(aes(x=intensity, group=sample, color=lab)) +
```

```

geom_density() +
ggtitle("condition D")

densityLab2 <- pe[["peptideLog"]][, colData(pe)$lab=="lab2"] |>
assay() |>
as.data.frame() |>
gather(sample, intensity) |>
mutate(condition = colData(pe)[sample, "condition"]) |>
ggplot(aes(x=intensity, group=sample, color=condition)) + 
geom_density() +
ggtitle("lab2")

```

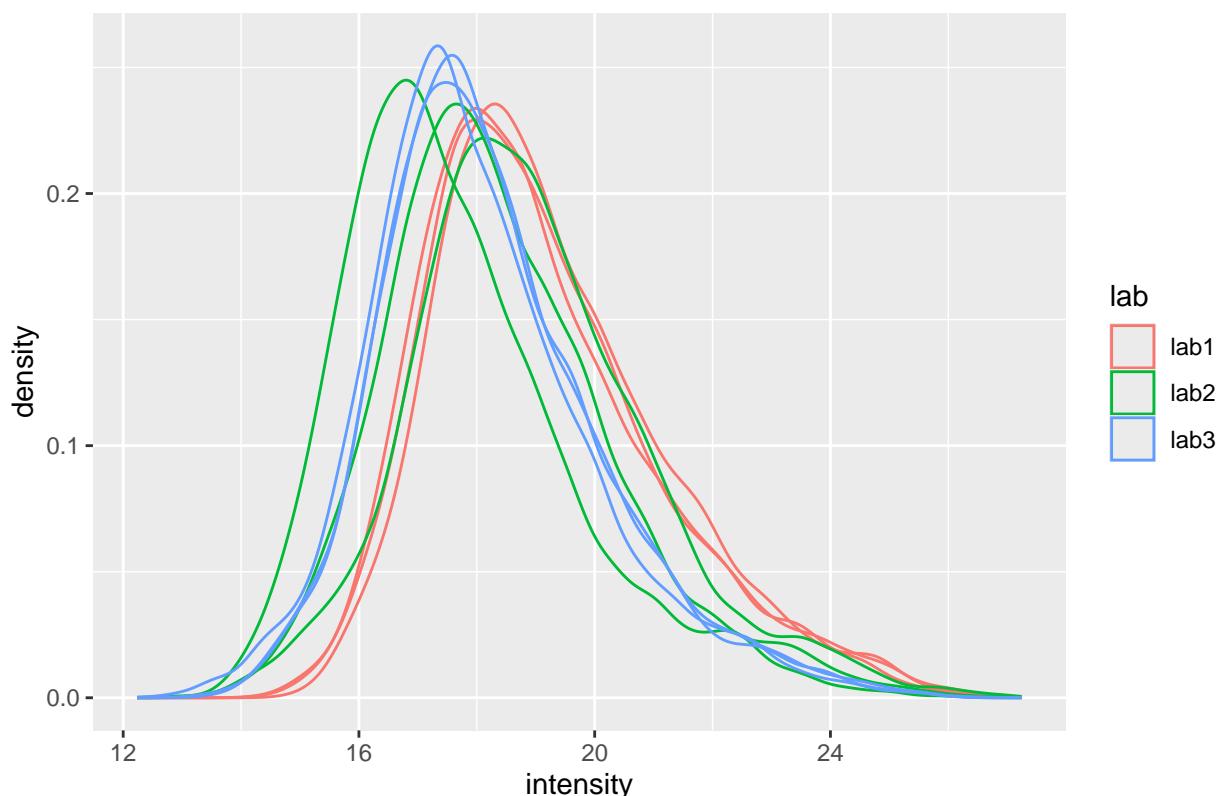
densityConditionD

```

## Warning: Removed 37590 rows containing non-finite outside the scale range
## (`stat_density()`).

```

condition D

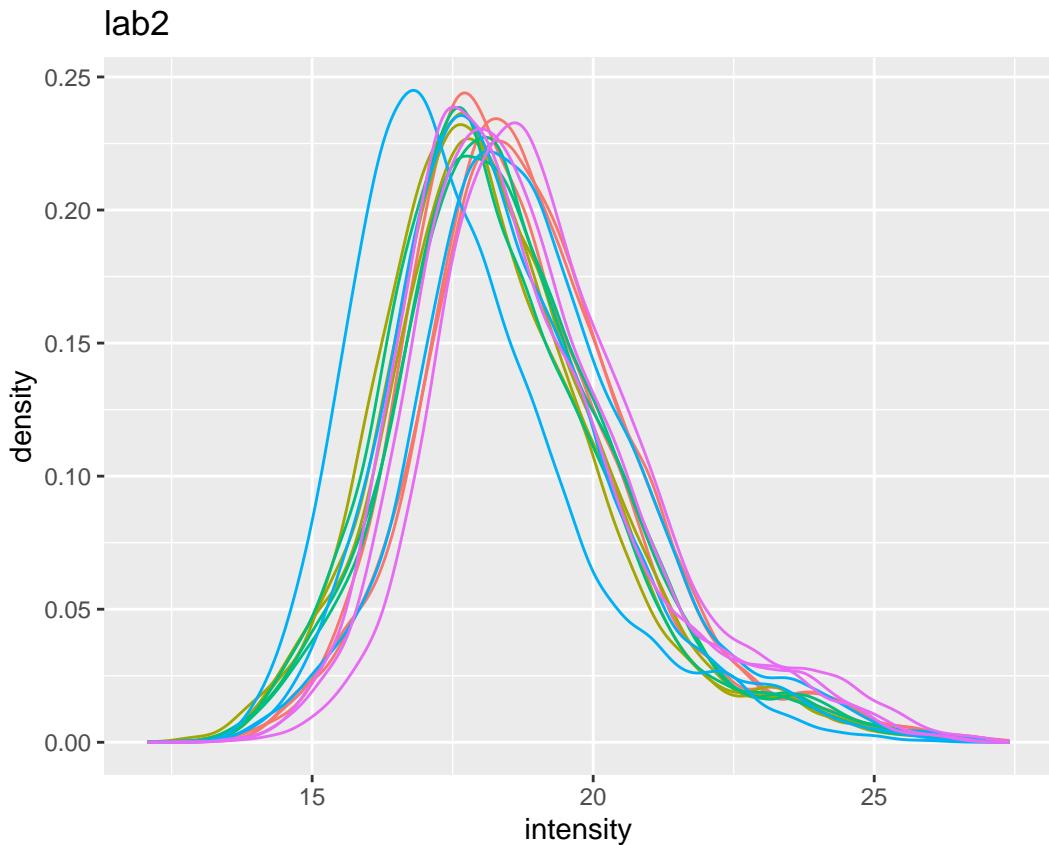


densityLab2

```

## Warning: Removed 42228 rows containing non-finite outside the scale range
## (`stat_density()`).

```



- Even in very clean synthetic dataset (same background, only 48 UPS proteins can be different) the marginal peptide intensity distribution across samples can be quite distinct

- Considerable effects between and within labs for replicate samples
- Considerable effects between samples with different spike-in concentration

→ Normalization is needed

3.3.1 Mean or median?

- Miller and Fishkin (1997) reported that over a period of 30 years males would like to have on average 64.3 partners and females 2.8.
- Miller and Fishkin (1997) reported that the median number of partners someone would like to have over a period of 30 years males is 1 for both males and females.

Mean is very sensitive to outliers!

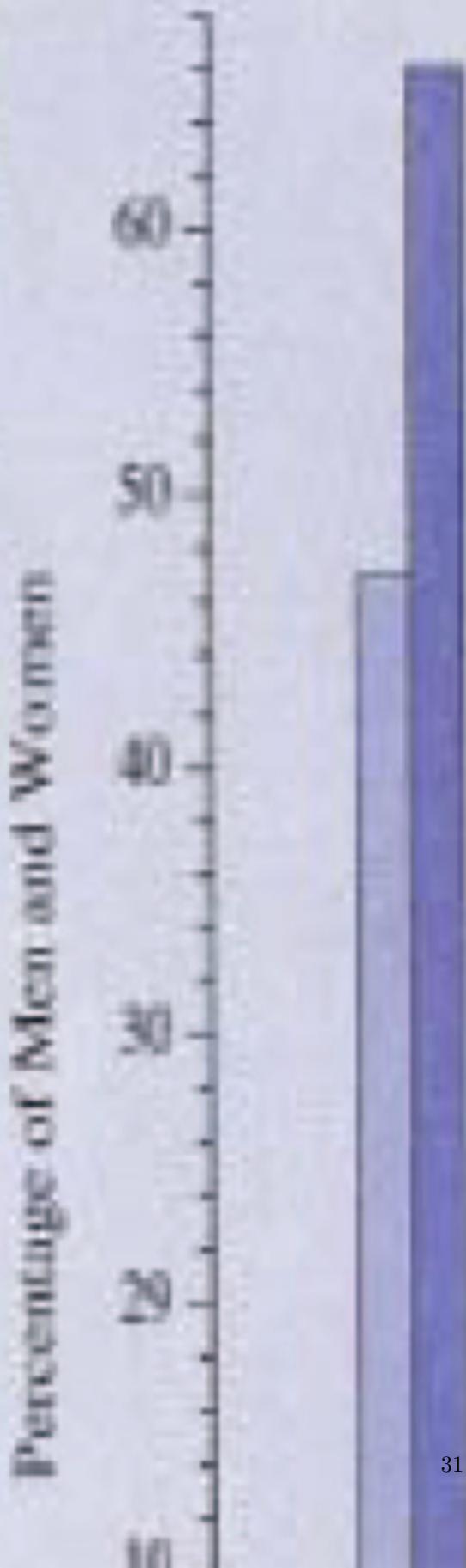


FIGURE 2-8 Distribution of the number of children ever born by age 30 for men and women over 30 years. Note: To the right of the bars, the tail of these distributions is apparent. It is apparent that the tail is longer for women than for men here. [From Miller & Simpson, 1990, p. 11. Copyright © 1990 by Wadsworth Publishing Company, Inc. Reprinted by permission of the author.]

3.3.2 Normalization of the data by median centering

$$y_{ip}^{\text{norm}} = y_{ip} - \hat{\mu}_i$$

with $\hat{\mu}_i$ the median intensity over all observed peptides in sample i .

Click to see R-code to normalize the data

```
pe <- normalize(pe,
                 i = "peptideLog",
                 name = "peptideNorm",
                 method = "center.median")
```

3.3.3 Plots of normalized data

Click to see code to make plot

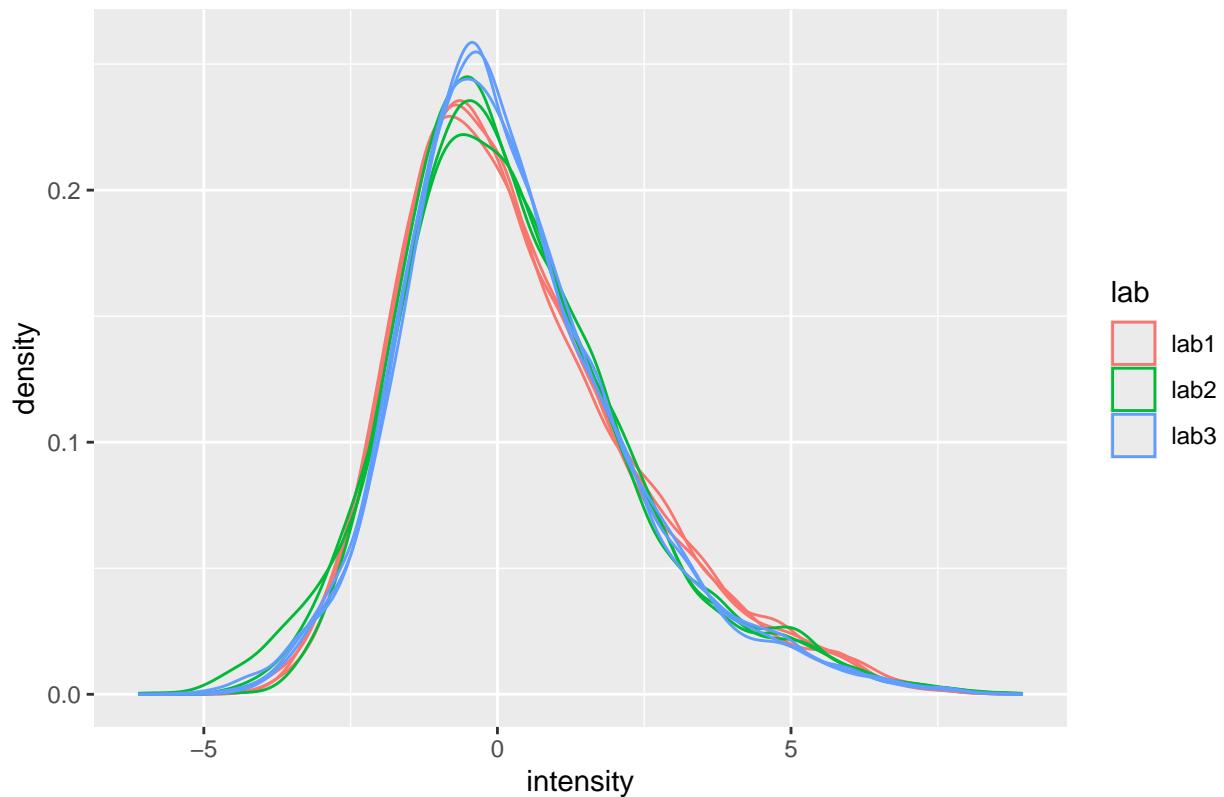
```
densityConditionDNorm <- pe[["peptideNorm"]][, colData(pe)$condition=="D"] |>
  assay() |>
  as.data.frame() |>
  gather(sample, intensity) |>
  mutate(lab = colData(pe)[sample, "lab"]) |>
  ggplot(aes(x=intensity, group=sample, color=lab)) +
  geom_density() +
  ggtitle("condition D")

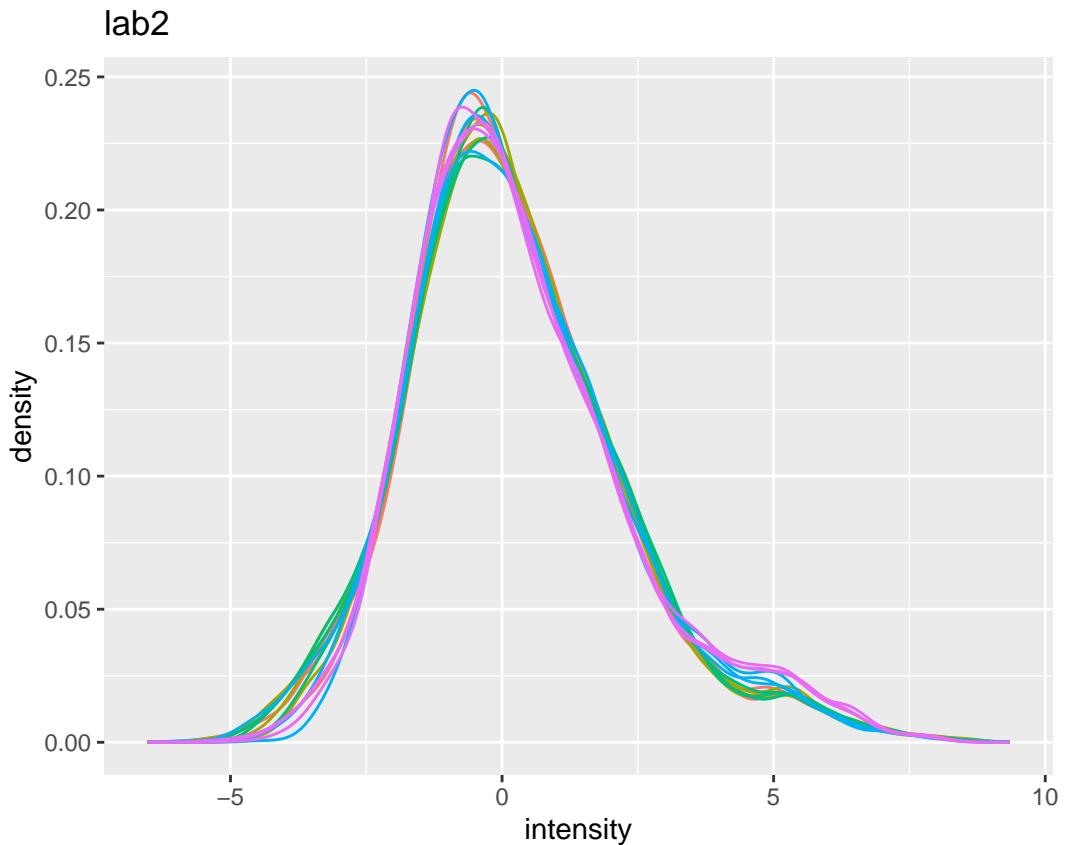
densityLab2Norm <- pe[["peptideNorm"]][, colData(pe)$lab=="lab2"] |>
  assay() |>
  as.data.frame() |>
  gather(sample, intensity) |>
  mutate(condition = colData(pe)[sample, "condition"]) |>
  ggplot(aes(x=intensity, group=sample, color=condition)) +
  geom_density() +
  ggtitle("lab2")

densityConditionDNorm

## Warning: Removed 37590 rows containing non-finite outside the scale range
## (`stat_density()`).
```

condition D





- Upon normalization the marginal distributions of the peptide intensities across samples are much more comparable
- We still see deviations
- This can be due to technical variability
- In micro-array literature, quantile normalisation is used to force the median and all other quantiles to be equal across samples
- In proteomics quantile normalisation often introduces artifacts due to a difference in missing peptides across samples
- More advanced methods should be developed for normalizing proteomics data
- If there are differences in the width of the marginal distributions of the data across samples. They can also be standardized by using a robust estimator for location and scale, i.e.

$$y_{ip}^{\text{norm}} = \frac{y_{ip} - \mu_i}{s_i}$$

3.4 Summarization

- We illustrate summarization issues using a subset of the cptac study (Lab 2, condition A and E) for a spiked protein (UPS P12081).

Click to see code to make plot

```
summaryPlot <- pe[["peptideNorm"]][
  rowData(pe[["peptideNorm"]])$Proteins == "P12081ups|SYHC_HUMAN_UPS",
  colData(pe)$lab=="lab2"&colData(pe)$condition %in% c("A","E")] |>
  assay() |>
  as.data.frame() |>
```

```

rownames_to_column(var = "peptide") |>
gather(sample, intensity, -peptide) |>
mutate(condition = colData(pe)[sample, "condition"]) |>
ggplot(aes(x = peptide, y = intensity, color = sample, group = sample, label = condition), show.legend = FALSE) +
geom_line(show.legend = FALSE) +
geom_text(show.legend = FALSE) +
theme_minimal() +
theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust = 1)) +
xlab("Peptide") +
ylab("Intensity (log2)")

```

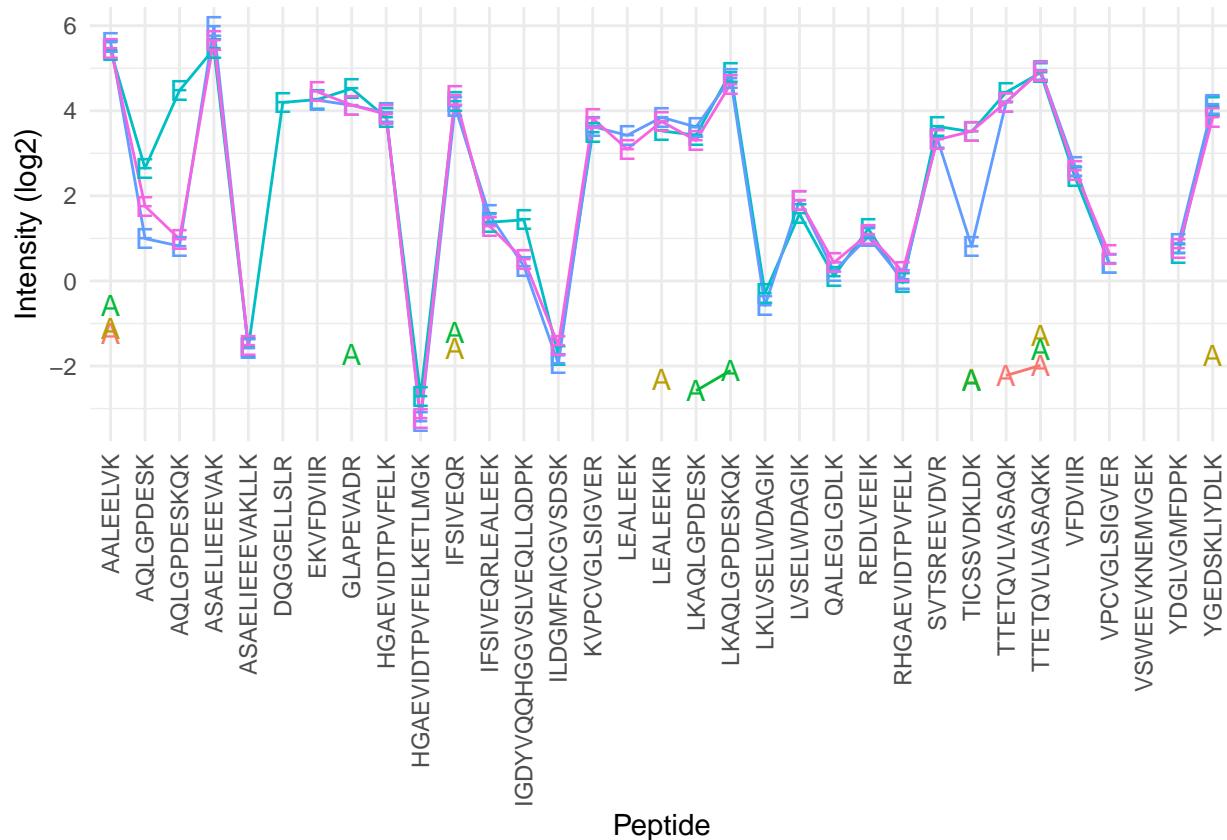
summaryPlot

```

## Warning: Removed 10 rows containing missing values or values outside the scale range
## (`geom_line()`).

## Warning: Removed 90 rows containing missing values or values outside the scale range
## (`geom_text()`).

```



We observe:

- intensities from multiple peptides for each protein in a sample
- Strong peptide effect -Unbalanced peptide identification
- Pseudo-replication: peptide intensities from a particular protein in the same sample are correlated, i.e. they more alike than peptide intensities from a particular protein between samples.

→ Summarize all peptide intensities from the same protein in a sample into a single protein expression value

Commonly used methods are

- Mean summarization

$$y_{ip} = \beta_i^{\text{samp}} + \epsilon_{ip}$$

- Median summarization

- Maxquant's maxLFQ summarization (in protein groups file)

- Model based summarization:

$$y_{ip} = \beta_i^{\text{samp}} + \beta_p^{\text{pep}} + \epsilon_{ip}$$

Click to see R-code to normalize the data

We use the standard summarization in `aggregateFeatures`, which is robust model based summarization.

```
pe <- aggregateFeatures(pe,
  i = "peptideNorm",
  fcol = "Proteins",
  na.rm = TRUE,
  name = "protein")
```

```
## Your quantitative and row data contain missing values. Please read the
## relevant section(s) in the aggregateFeatures manual page regarding the
## effects of missing values on data aggregation.
```

```
## Aggregated: 1/1
```

Other summarization methods can be implemented by using the `fun` argument in the `aggregateFeatures` function.

- `fun = MsCoreUtils::medianPolish()` to fits an additive model (two way decomposition) using Tukey's median polish_ procedure using `stats::medpolish()`
- `fun = MsCoreUtils::robustSummary()` to calculate a robust aggregation using `MASS::rlm()` (default)
- `fun = base::colMeans()` to use the mean of each column
- `fun = matrixStats::colMedians()` to use the median of each column
- `fun = base::colSums()` to use the sum of each column

3.5 Filtering at protein level

We want to have at least 4 observed proteins so that most proteins have at least 2 observations in each group. So we tolerate a proportion of $(n-4)/n$ NAs.

```
nObs <- 4
n <- ncol(pe[["protein"]])
pNA <- (n-nObs)/n
pe <- filterNA(pe, pNA = pNA, i = "protein")
```

4 Exercise

1. We will evaluate different summarization methods (Maxquant maxLFQ, median and robust model based) in the tutorial session before discussing on their advantages/disadvantages.
2. Can you anticipate on potential problems related to the summarization?

5 Software & code

- Our R/Bioconductor package [msqrob2](#) can be used in R markdown scripts or with GUI/shinyApps [QFeaturesGUI](#) and [msqrob2gui](#).
- The GUIs are intended as a introduction to the key concepts of proteomics data analysis for users who have no experience in R.
- However, learning how to code data analyses in R markdown scripts is key for open en reproducible science and for reporting your proteomics data analyses and interpretation in a reproducible way.
- More information on our tools can be found in our papers (L. J. Goeminne, Gevaert, and Clement 2016), (L. J. E. Goeminne et al. 2020), (Sticker et al. 2020) and (Vandenbulcke and Clement 2025). Please refer to our work when using our tools.

References

- Goeminne, L. J. E., A. Sticker, L. Martens, K. Gevaert, and L. Clement. 2020. “MSqRob Takes the Missing Hurdle: Uniting Intensity- and Count-Based Proteomics.” *Anal Chem* 92 (9): 6278–87.
- Goeminne, L. J., K. Gevaert, and L. Clement. 2016. “Peptide-level Robust Ridge Regression Improves Estimation, Sensitivity, and Specificity in Data-dependent Quantitative Label-free Shotgun Proteomics.” *Mol Cell Proteomics* 15 (2): 657–68.
- Sticker, A., L. Goeminne, L. Martens, and L. Clement. 2020. “Robust Summarization and Inference in Proteome-wide Label-free Quantification.” *Mol Cell Proteomics* 19 (7): 1209–19.
- Vandenbulcke, C., S. Vanderaa, and L. Clement. 2025. “msqrob2TMT: Robust Linear Mixed Models for Inferring Differential Signals in Tandem Mass Tag-Based Proteomics.” *Molecular & Cellular Proteomics* 24 (3): e10101–1. <https://doi.org/10.1016/j.mcpro.2025.00101-X>.