# Package 'proBatch'

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Title Tools for Diagnostics and Corrections of Batch Effects in Proteomics
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Description The proBatch package facilitates batch effects analysis and correction in high-thoughput experiments. It was developed primarily for mass- spectrometry proteomics (DIA/SWATH), but could also be applicable to most omic data with minor adaptations. The package con- tains functions for diagnostics (proteome/genome-wide and feature- level), correction (normalization and batch effects correction) and quality control. Non- linear fitting based approaches were also included to deal with complex, mass spectrometry-specific signal drifts.
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adjust\_batch\_trend adjust batch signal trend with the custom (continuous) fit

#### Description

adjust batch signal trend with the custom (continuous) fit

### Usage

```
adjust_batch_trend(data_matrix, sample_annotation,
    batch_col = "MS_batch", feature_id_col = "peptide_group_label",
    sample_id_col = "FullRunName", measure_col = "Intensity",
    sample_order_col = "order", fit_func = fit_nonlinear,
    abs_threshold = 5, pct_threshold = 0.2, ...)
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data	
<pre>sample_annotat:</pre>	ion	
	data frame with sample ID, technical (e.g. MS batches) and biological (e.g. Diet) covariates	
batch_col	column in sample_annotation that should be used for batch comparison	
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.	
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)	
<pre>measure_col</pre>	if df_long is among the parameters, it is the column with expression/abundance/intensity, otherwise, it is used internally for consistency	
sample_order_col		
	column, determining the order of sample MS run, used as covariate to fit the non-linear fit	
fit_func	function to fit the (non)-linear trend	
abs_threshold	the absolute threshold to filter data for curve fitting	
<pre>pct_threshold</pre>	the percentage threshold to filter data for curve fitting	
	other parameters, usually those of the fit_func	

### Value

list of two items: 1) data\_matrix, adjusted with continious fit; 2) fit\_df, used to examine the fitting curves

### See Also

fit\_nonlinear

### Examples

```
trend_corrected_matrix <- adjust_batch_trend(example_proteome_matrix,
example_sample_annotation, span = 0.7,
abs_threshold = 5, pct_threshold = 0.20)
```

center\_peptide\_batch\_medians

Median centering of the peptides (per batch median)

#### Description

Median centering of the peptides (per batch median)

### Usage

```
center_peptide_batch_medians(df_long, sample_annotation = NULL,
    sample_id_col = "FullRunName", batch_col = "MS_batch",
    feature_id_col = "peptide_group_label", measure_col = "Intensity")
```

#### Arguments

df_long	data frame where each row is a single feature in a single sample. It minimally has a sample_id_col, a feature_id_col and a measure_col, but usually also an m_score (in OpenSWATH output result file)
<pre>sample_annotat:</pre>	ion
	data frame with sample ID, technical (e.g. MS batches) and biological (e.g. Diet) covariates
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
batch_col	column in sample_annotation that should be used for batch comparison
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity, otherwise, it is used internally for consistency

### Value

df\_long-size long format data with batch-effect corrected with per-feature batch median centering in Intensity\_normalized column

#### Examples

```
median_centered_proteome <- center_peptide_batch_medians(
example_proteome, example_sample_annotation)</pre>
```

correct\_batch\_effects Batch correction method allows correction of continuous sigal drift within batch and discrete difference across batches.

### Description

Batch correction method allows correction of continuous sigal drift within batch and discrete difference across batches.

### Usage

```
correct_batch_effects(data_matrix, sample_annotation,
  fitFunc = "loess_regression", discreteFunc = c("MedianCentering",
    "ComBat"), batch_col = "MS_batch",
  feature_id_col = "peptide_group_label",
    sample_id_col = "FullRunName", measure_col = "Intensity",
    sample_order_col = "order", abs_threshold = 5, pct_threshold = 0.2,
    ...)
```

#### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data	
<pre>sample_annotat;</pre>	ion	
	data frame with sample ID, technical (e.g. MS batches) and biological (e.g. Diet) covariates	
fitFunc	function to use for the fit (currently only loess_regression available)	
discreteFunc	function to use for discrete batch correction (MedianCentering or ComBat)	
batch_col	column in sample_annotation that should be used for batch comparison	
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.	
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)	
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity, otherwise, it is used internally for consistency	
sample_order_col		
	column, determining the order of sample MS run, used as covariate to fit the non-linear fit	
abs_threshold	the absolute threshold to filter data for curve fitting	
<pre>pct_threshold</pre>	the percentage threshold to filter data for curve fitting	
	other parameters, usually of normalize_custom_fit, and fit_func	

### Value

data\_matrix-size data matrix with batch-effect corrected by fit and discrete functions

### Examples

```
batch_corrected_matrix <- correct_batch_effects(
example_proteome_matrix, example_sample_annotation,
discreteFunc = 'MedianCentering',
batch_col = 'MS_batch',
span = 0.7,
abs_threshold = 5, pct_threshold = 0.20)</pre>
```

correct\_with\_ComBat Adjusts for discrete batch effects using ComBat

### Description

Standardized input-output ComBat normalization ComBat allows users to adjust for batch effects in datasets where the batch covariate is known, using methodology described in Johnson et al. 2007. It uses either parametric or non-parametric empirical Bayes frameworks for adjusting data for batch effects. Users are returned an expression matrix that has been corrected for batch effects. The input data are assumed to be cleaned and normalized before batch effect removal.

### Usage

```
correct_with_ComBat(data_matrix, sample_annotation,
    sample_id_col = "FullRunName", batch_col = "MS_batch",
    par.prior = TRUE)
```

#### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data
sample_annotation	
	data frame with sample ID, technical (e.g. MS batches) and biological (e.g. Diet) covariates
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
batch_col	column in sample_annotation that should be used for batch comparison
par.prior	whether parametrical or non-parametrical prior should be used

### Value

data\_matrix-size data matrix with batch-effect corrected by ComBat

#### Examples

```
combat_corrected_matrix <- correct_with_ComBat(
example_proteome_matrix, example_sample_annotation)</pre>
```

create\_peptide\_annotation

Prepare peptide annotation from long format data frame Create lightweight peptide annotation data frame for selection of illustrative proteins

### Description

Prepare peptide annotation from long format data frame

Create light-weight peptide annotation data frame for selection of illustrative proteins

### Usage

```
create_peptide_annotation(df_long,
  feature_id_col = "peptide_group_label",
  annotation_col = c("ProteinName"))
```

### Arguments

df_long	data frame where each row is a single feature in a single sample. It minimally has a sample_id_col, a feature_id_col and a measure_col, but usually also an m_score (in OpenSWATH output result file)
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
annotation_col	one or more columns contatining protein ID

#### Value

data frame containing petpide annotations

#### See Also

plot\_peptides\_of\_one\_protein, plot\_protein\_corrplot

### Examples

```
generated_peptide_annotation <- create_peptide_annotation(
example_proteome, feature_id_col = "peptide_group_label",
annotation_col = c("ProteinName" ))</pre>
```

dates\_to\_posix

#### Description

convert date/time column of sample\_annotation to POSIX format required to keep number-like behaviour

#### Usage

```
dates_to_posix(sample_annotation, time_column = c("RunDate", "RunTime"),
    new_time_column = "DateTime", dateTimeFormat = c("%b_%d",
    "%H:%M:%S"))
```

#### Arguments

sample\_annotation

data matrix with:

- 1. sample\_id\_col (this can be repeated as row names)
- 2. biological covariates
- 3. technical covariates (batches etc)
- time\_column name of the column(s) where run date & time are specified. These will be used
  to determine the run order
  new\_time\_column
  name of the new column to which date&time will be converted to
- dateTimeFormat POSIX format of the date and time. See as.POSIXct from base R for details

#### Value

sample annotation file with a new column new\_time\_column with POSIX-formatted date

#### Examples

```
date_to_posix <- dates_to_posix(example_sample_annotation,
time_column = c('RunDate','RunTime'),
new_time_column = 'DateTime',
dateTimeFormat = c("%b_%d", "%H:%M:%S"))
```

date\_to\_sample\_order Convert date/time to POSIXct and rank samples by it

### Description

Converts date/time columns fo sample\_annotation to POSIXct format and calculates sample run rank in order column

#### Usage

```
date_to_sample_order(sample_annotation, time_column = c("RunDate",
    "RunTime"), new_time_column = "DateTime",
    dateTimeFormat = c("%b_%d", "%H:%M:%S"),
    new_order_col = "order", instrument_col = "instrument")
```

### Arguments

sample\_annotation

data matrix with:

- 1. sample\_id\_col (this can be repeated as row names)
- 2. biological covariates
- 3. technical covariates (batches etc)

time_column	name of the column(s) where run date & time are specified. These will be used	
	to determine the run order	
new_time_column		
	name of the new column to which date&time will be converted to	
dateTimeFormat	POSIX format of the date and time. See as.POSIXct from base R for details	
new_order_col	name of column with generated the order of sample run based on time columns	
instrument_col	column, denoting different instrument used for measurements	

#### Value

sample annotation file with a new column new\_time\_column with POSIX-formatted date & new\_order\_col
used in some diagnostic plots (e.g. plot\_iRT, plot\_sample\_mean)

### Examples

```
sample_annotation_wOrder <- date_to_sample_order(
example_sample_annotation,
time_column = c('RunDate','RunTime'),
new_time_column = 'new_DateTime',
dateTimeFormat = c("%b_%d", "%H:%M:%S"),
new_order_col = 'new_order',
instrument_col = NULL)</pre>
```

example\_peptide\_annotation Peptide annotation data

### Description

This is data from Aging study annotated with gene names

### Usage

example\_peptide\_annotation

#### Format

A data frame with 535 rows and 10 variables:

peptide\_group\_label peptide group label ID, identical to peptide\_group\_label in example\_proteome
Gene HUGO gene ID

ProteinName protein group name as specified in example\_proteome

example\_proteome Example protein data in long format

#### Description

This is data from Aging study with all iRT, spike-in peptides, few random peptides and QTL proteins for biological signal improvement demonstration

#### Usage

example\_proteome

#### Format

A data frame with 124655 rows and 5 variables:

- **peptide\_group\_label** peptide ID, which is regular feature level. This column is mostly used as feature\_id\_col
- Intensity peptide group intensity in given sample. Used in function as measure\_col
- **ProteinName** Protein group ID, specified as N/UniProtID1|UniProtID2|..., where N is number of protein peptide group maps to. If 1/UniProtID, then this is proteotypic peptide
- Gene conventional gene name of corresponding ProteinName
- FullRunName name of the file, in most functions used for sample\_id\_col ...

#### Source

PRIDE ID will be added in future

example\_proteome\_matrix

Example protein data in matrix

#### Description

This is measurement data from Aging study with columns representing samples and rows representing peptides

### Usage

```
example_proteome_matrix
```

#### Format

A matrix with 534 rows and 233 columns:

#### Source

PRIDE ID will be added in future

example\_sample\_annotation

Sample annotation data version 1

### Description

This is data from BXD aging study with mock instruments to show how instrument-specific functionality works

### Usage

example\_sample\_annotation

### Format

A data frame with 233 rows and 11 variables:

FullRunName name of the file, in most functions used for sample\_id\_col

MS\_batch mass-spectrometry batch: 7-level factor of manually annotated batches

EarTag mouse ID, i.e. ID of the biological object

Strain mouse strain ID - biological covariate #1

**Diet** diet - either HFD = 'High Fat Diet' or CD = 'Chow Diet'. Mix stands for mixture of several samples

Sex mice sex - 3-level biological covariate. Possible values - "

- RunDate mass-spectrometry running date. In combination with RunTime used for running order determination
- RunTime mass-spectrometry running time. In combination with RunDate used for running order determination

DateTime numeric date and time generated by date\_to\_sample\_order

order order of samples generated by sorting DateTime in date\_to\_sample\_order

digestion\_batch peptide digestion batch: 5-level factor of manually annotated batches ...

log\_transform

### Description

Log transformation of the data, ensuring that the row and column names are retained

### Usage

```
log_transform(data_matrix, log_base = 2)
```

### Arguments

data_matrix	raw data matrix (features in rows and samples in columns)
log_base	base of the logarithm for transformation

#### Value

data\_matrix-size matrix, with columns log2 transformed

#### Examples

log\_transformed\_matrix <- log\_transform(example\_proteome\_matrix)</pre>

long\_to\_matrix Long to wide conversion

### Description

Convert from a long data frame representation to a wide matrix representation

### Usage

```
long_to_matrix(df_long, feature_id_col = "peptide_group_label",
    measure_col = "Intensity", sample_id_col = "FullRunName")
```

#### Arguments

df_long	data frame where each row is a single feature in a single sample. It minimally has a sample_id_col, a feature_id_col and a measure_col, but usually also an m_score (in OpenSWATH output result file)
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
<pre>measure_col</pre>	if df_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)

matrix\_to\_long

### Value

data\_matrix (proBatch) like matrix (features in rows, samples in columns)

### See Also

Other matrix manipulation functions: matrix\_to\_long

### Examples

```
proteome_matrix <- long_to_matrix(example_proteome)</pre>
```

matrix\_to\_long Wide to long conversion

### Description

Convert from wide matrix to a long data frame representation

### Usage

```
matrix_to_long(data_matrix, sample_annotation = NULL,
    feature_id_col = "peptide_group_label", measure_col = "Intensity",
    sample_id_col = "FullRunName", step = NULL)
```

#### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data
sample_annotati	on
	data matrix with:
	1. sample_id_col (this can be repeated as row names)
	2. biological covariates
	3. technical covariates (batches etc)
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
step	normalization step (e.g. Raw or Quantile_normalized or qNorm_ComBat). Use- ful if consecutive steps are compared in plots. Note that in plots these are usu- ally ordered alphabetically, so it's worth naming with numbers, e.g. 1_raw, 2_quantile

### Value

df\_long (proBatch) like data frame

#### See Also

Other matrix manipulation functions: long\_to\_matrix

### Examples

```
proteome_long <- matrix_to_long(example_proteome_matrix,
example_sample_annotation)
```

normalize

Data normalization methods

### Description

Data normalization methods

#### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
measure_col	if 'df_long' is among the parameters, it is the column with expression/abundance/intensity, otherwise, it is used internally for consistency

normalize_data	Normalization brings the samples to the same scale

### Description

Normalization brings the samples to the same scale

### Usage

```
normalize_data(data_matrix, normalizeFunc = c("quantile",
    "medianCentering"), log_base = NULL)
```

### Arguments

data_matrix	raw data matrix (features in rows and samples in columns)
normalizeFunc	global batch normalization method ('quantile' or 'MedianCentering')
log_base	whether to log transform data matrix before normalization ('NULL', '2' or '10')

### Value

data\_matrix-size matrix, with columns normalized

### Examples

```
quantile_normalized_matrix <- normalize_data(example_proteome_matrix,
normalizeFunc = "quantile", log_base = 2)
```

normalize\_sample\_medians

Normalization by centering sample medians to global median of the data

### Description

Normalization by centering sample medians to global median of the data

### Usage

```
normalize_sample_medians(df_long, sample_id_col = "FullRunName",
    measure_col = "Intensity")
```

### Arguments

df_long	log transformed long format data matrix (see 'df_long')
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
measure_col	if 'df_long' is among the parameters, it is the column with expression/abundance/intensity, otherwise, it is used internally for consistency

### Value

'df\_long'-size matrix, with intensity scaled to global median

### Examples

median\_normalized\_matrix <- normalize\_sample\_medians(example\_proteome)</pre>

plot\_heatmap

Plot the heatmap of samples

### Description

Plot the heatmap of samples

### Usage

```
plot_heatmap(data_matrix, sample_annotation = NULL,
  sample_id_col = "FullRunName", sample_annotation_col = NULL,
  sample_annotation_row = NULL, fill_the_missing = TRUE,
  cluster_rows = TRUE, cluster_cols = FALSE,
  annotation_color_list = NA,
  heatmap_color = colorRampPalette(rev(RColorBrewer::brewer.pal(n = 7,
  name = "RdYlBu")))(100), color_for_missing = "black", filename = NA,
  plot_title = NA, ...)
```

#### Arguments

data\_matrix features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. in most function, it is assumed that this is the log transformed version of the original data

sample\_annotation

### data matrix with

- 1. sample\_id\_col (this can be repeated as row names)
- 2. biological and
- 3. technical covariates (batches etc)

; each column of sample annotation will get it's own row. If cluster_cols =
T this will indicate, whether sample proximity is driven by one of biolical or
technical factors

- sample\_id\_col name of the column in sample\_annotation file, where the filenames (colnames of the data matrix are found)
- sample\_annotation\_col

biological or technical factors to be annotated in heatmap columns

sample\_annotation\_row

```
biological or technical factors to be annotated in heatmap rows
```

### fill\_the\_missing

boolean value determining if missing values should be substituted with -1 (and colored with black)

cluster\_rows boolean value determining if rows should be clustered

```
cluster_cols boolean value determining if columns should be clustered
```

annotation\_color\_list

```
list specifying colors for columns (samples). Best created by sample_annotation_to_colors
```

heatmap\_color vector of colors used in heatmap (typicall a gradient)

```
color_for_missing
```

```
special color to make missing values. Usually black or white, depending on heatmap_color
```

- filename filepath where to save the image
- plot\_title Title of the plot (usually, processing step + representation level (fragments, transitions, proteins))
- ... other parameters of link[pheatmap]{pheatmap}

### Value

object returned by link[pheatmap]{pheatmap}

#### plot\_hierarchical\_clustering

#### See Also

sample\_annotation\_to\_colors, pheatmap

#### Examples

```
color_scheme <- sample_annotation_to_colors (example_sample_annotation,
factor_columns = c('MS_batch','EarTag', "Strain",
"Diet", "digestion_batch", "Sex"),
not_factor_columns = 'DateTime',
numeric_columns = c('order'))
heatmap_plot <- plot_heatmap(example_proteome_matrix,
example_sample_annotation,
sample_annotation_col = c("MS_batch", "digestion_batch", "Diet"),
cluster_cols = TRUE,
annotation_color_list = color_scheme$list_of_colors,
show_rownames = FALSE, show_colnames = FALSE)
```

```
plot_hierarchical_clustering
```

cluster the data matrix to visually inspect which confounder dominates

#### Description

cluster the data matrix to visually inspect which confounder dominates

#### Usage

```
plot_hierarchical_clustering(data_matrix, color_df,
    distance = "euclidean", agglomeration = "complete",
    label_samples = TRUE, label_font = 0.2, plot_title = NULL, ...)
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. in most function, it is assumed that this is the log transformed version of the original data
color_df	data frame of colors, as created by sample_annotation_to_colors
distance	distance metric used for clustering
agglomeration	agglomeration methods as used by hclust
label_samples	if TRUE sample IDs (column names of data_matrix) will be printed
label_font	size of the font. Is active if label_samples is TRUE, ignored otherwise
plot_title	Title of the plot (usually, processing step + representation level (fragments, tran- sitions, proteins))
	other parameters of plotDendroAndColors from WGCNA package

### Value

No return

#### See Also

```
hclust, sample_annotation_to_colors, plotDendroAndColors
```

#### Examples

```
color_scheme <- sample_annotation_to_colors (example_sample_annotation,
factor_columns = c('MS_batch','EarTag', "Strain", "Diet", "digestion_batch", "Sex"),
not_factor_columns = 'DateTime',
numeric_columns = c('order'))
```

color\_annotation <- color\_scheme\$color\_df</pre>

```
hiarchical_clustering_plot <- plot_hierarchical_clustering(
example_proteome_matrix, color_annotation,
distance = "euclidean", agglomeration = 'complete',
label_samples = FALSE)</pre>
```

plot\_iRT

Plot iRT measurements

#### Description

Creates a iRT facetted ggplot2 plot of the value in measure\_col vs order\_col using plot\_single\_feature. Additionally, the resulting plot can also be facetted by batch.

#### Usage

```
plot_iRT(df_long, sample_annotation, peptide_annotation = NULL,
    protein_col = "ProteinName", order_col = "order",
    irt_pattern = "iRT", sample_id_col = "FullRunName",
    batch_col = "MS_batch", measure_col = "Intensity",
    feature_id_col = "peptide_group_label", color_by_batch = FALSE,
    color_scheme = "brewer", facet_by_batch = FALSE,
    color_by_col = NULL, color_by_value = NULL,
    plot_title = "iRT peptide profile", ...)
```

#### Arguments

df\_long data frame where each row is a single feature in a single sample. It minimally has a sample\_id\_col, a feature\_id\_col and a measure\_col, but usually also an m\_score (in OpenSWATH output result file)

#### sample\_annotation

data matrix with:

- 1. sample\_id\_col (this can be repeated as row names)
- 2. biological covariates
- 3. technical covariates (batches etc)

peptide\_annotation

```
long format data with peptide ID and their corresponding protein annotations
```

protein\_col column where protein names are specified

plot\_PCA

order_col	column in sample_annotation that determines sample order. It is used for certain diagnostics and normalisations.
irt_pattern	substring used to identify irts proteins in the column 'ProteinName'
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
batch_col	column in sample_annotation that should be used for batch comparison
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
color_by_batch	(logical) whether to color points by batch
color_scheme	color scheme for ggplot representation
facet_by_batch	(logical) whether to plot each batch in its own facet
color_by_col	column to color by certain value denoted by color_by_value
color_by_value	value in color_by_col to color
plot_title	the string indicating the source of the peptides
	additional arguments to plot_single_feature function

### Value

ggplot2 type plot of measure\_col vs order\_col, faceted by irt\_pattern containing proteins and (optionally) by batch\_col

### See Also

Other feature-level diagnostic functions: plot\_peptides\_of\_one\_protein, plot\_single\_feature, plot\_spike\_in, plot\_with\_fitting\_curve

#### Examples

```
irt_plot <- plot_iRT(example_proteome,
example_sample_annotation,
protein_col = 'Gene', irt_pattern = "BOVINE_A1ag")
```

plot\_PCA plot PCA plot

### Description

plot PCA plot

### Usage

```
plot_PCA(data_matrix, sample_annotation,
    feature_id_col = "peptide_group_label", color_by = "MS_batch",
    PC_to_plot = c(1, 2), fill_the_missing = 0,
    colors_for_factor = NULL, theme = "classic", plot_title = NULL)
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. in most function, it is assumed that this is the log transformed version of the original data	
sample_annotati	ion	
	data matrix with 1) sample_id_col (this can be repeated as row names) 2) bio- logical and 3) technical covariates (batches etc)	
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.	
color_by	column name (as in sample_annotation) to color by	
PC_to_plot	principal component numbers for x and y axis	
fill_the_missing		
	boolean value determining if missing values should be substituted with -1 (and colored with black)	
colors_for_factor		
	named vector of colors for the color_by variable	
theme	ggplot theme, by default classic. Can be easily overriden (see examples)	
plot_title	Title of the plot (usually, processing step + representation level (fragments, tran- sitions, proteins))	

#### Value

ggplot scatterplot colored by factor levels of column specified in factor\_to\_color

#### See Also

autoplot.pca\_common, ggplot

### Examples

```
pca_plot <- plot_PCA(example_proteome_matrix, example_sample_annotation,
color_by = 'MS_batch', plot_title = "PCA colored by MS batch")
```

### Description

Creates a spike-in facetted ggplot2 plot of the value in measure\_col vs order\_col using plot\_single\_feature. Additionally, the resulting plot can also be facetted by batch.

### Usage

```
plot_peptides_of_one_protein(protein_name, protein_col = "ProteinName",
    df_long, sample_annotation, peptide_annotation = NULL,
    order_col = "order", sample_id_col = "FullRunName",
    batch_col = "MS_batch", measure_col = "Intensity",
    feature_id_col = "peptide_group_label", color_by_batch = FALSE,
    color_scheme = "brewer", facet_by_batch = FALSE,
    color_by_col = NULL, color_by_value = NULL,
    plot_title = sprintf("Peptides of %s protein", protein_name), ...)
```

### Arguments

protein_name	name of the protein as defined in ProteinName
protein_col	column where protein names are specified
df_long	data frame where each row is a single feature in a single sample. It minimally has a sample_id_col, a feature_id_col and a measure_col, but usually also an m_score (in OpenSWATH output result file)
sample_annotati	on
	data matrix with:
	1. sample_id_col (this can be repeated as row names)
	2. biological covariates
	3. technical covariates (batches etc)
<pre>peptide_annotat</pre>	
	long format data with peptide ID and their corresponding protein annotations
order_col	column in sample_annotation that determines sample order. It is used for certain diagnostics and normalisations.
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
batch_col	column in sample_annotation that should be used for batch comparison
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
color_by_batch	(logical) whether to color points by batch
color_scheme	color scheme for ggplot representation
facet_by_batch	(logical) whether to plot each batch in its own facet
color_by_col	column to color by certain value denoted by color_by_value
color_by_value	value in color_by_col to color
plot_title	the string indicating the source of the peptides
	additional arguments to plot_single_feature function

### Value

ggplot2 type plot of measure\_col vs order\_col, faceted by spike\_ins containing proteins and (optionally) by batch\_col

### See Also

Other feature-level diagnostic functions: plot\_iRT, plot\_single\_feature, plot\_spike\_in, plot\_with\_fitting\_cum

### Examples

```
peptides_of_one_protein_plot <- plot_peptides_of_one_protein (
protein_name = "Haao",
protein_col = "Gene", df_long = example_proteome,
example_sample_annotation,
order_col = 'order', sample_id_col = 'FullRunName',
batch_col = 'MS_batch')</pre>
```

```
plot_peptide_corr_distribution
```

*Plot distribution of peptide correlations within one protein and between proteins* 

#### Description

Plot distribution of peptide correlations within one protein and between proteins

### Usage

```
plot_peptide_corr_distribution(data_matrix, peptide_annotation,
    protein_col = "ProteinName", feature_id_col = "peptide_group_label",
    plot_title = "Distribution of peptide correlation",
    theme = "classic")
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data
peptide_annotat	tion
	long format data with peptide ID and their corresponding protein annotations
protein_col	the column name in peptide_annotation with protein names
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
plot_title	Title of the plot, usually processing step
theme	ggplot theme, by default classic. Can be easily overriden
	parameters for the ggplot visualisation

### Value

ggplot type object with violin plot for each plot\_param

plot\_protein\_corrplot

### Examples

```
peptide_corr_distribution <- plot_peptide_corr_distribution(
example_proteome_matrix,
example_peptide_annotation, protein_col = 'Gene')</pre>
```

plot\_protein\_corrplot Peptide correlation matrix (heatmap)

### Description

Plots correlation plot of peptides from a single protein

### Usage

```
plot_protein_corrplot(data_matrix, protein_name, peptide_annotation,
    protein_col = "ProteinName", feature_id_col = "peptide_group_label",
    flavor = c("pheatmap", "corrplot"), filename = NULL, width = NA,
    height = NA, unit = c("cm", "in", "mm"),
    plot_title = sprintf("Peptide correlation matrix of %s protein",
    protein_name), ...)
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data
protein_name	the name of the protein
<pre>peptide_annotat</pre>	ion
	df with peptides and their corresponding proteins
protein_col	the column name in peptide_annotation with protein names
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
flavor	either corrplot from 'corrplot' package or heatmap, as in 'pheatmap'
filename	path where the results are saved. If null the object is returned to the active window; otherwise, the object is save into the file. Currently only pdf and png format is supported
width	option determining the output image width
height	option determining the output image width
unit	units: 'cm', 'in' or 'mm'
plot_title	The title of the plot
	parameters for the corrplot visualisation

### Value

corrplot or pheatmap object depending on flavor

### Examples

plot\_PVCA

### Plot variance distribution by variable

#### Description

Plot variance distribution by variable

### Usage

```
plot_PVCA(data_matrix, sample_annotation, sample_id_col = "FullRunName",
    feature_id_col = "peptide_group_label",
    technical_covariates = c("MS_batch", "instrument"),
    biological_covariates = c("cell_line", "drug_dose"),
    fill_the_missing = 0, threshold_pca = 0.6, threshold_var = 0.01,
    colors_for_bars = NULL, theme = "classic", plot_title = NULL)
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. in most function, it is assumed that this is the log transformed version of the original data	
sample_annotati		
	data matrix with 1) sample_id_col (this can be repeated as row names) 2) bio- logical and 3) technical covariates (batches etc)	
sample_id_col	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)	
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.	
technical_covariates		
	vector sample_annotation column names that are technical covariates	
biological_covariates		
	vector sample_annotation column names, that are biologically meaningful co-variates	
fill_the_missin	g	
	numeric value that the missing values are substituted with	
threshold_pca	the percentile value of the minimum amount of the variabilities that the selected principal components need to explain	
threshold_var	the percentile value of weight each of the covariates needs to explain (the rest will be lumped together)	
colors_for_bars		
	four-item color vector, specifying colors for the following categories: c('residual', 'biological', 'biol:techn', 'technical')	

#### plot\_sample\_corr\_distribution

theme	ggplot theme, by default classic. Can be easily overriden (see examples)
plot_title	Title of the plot (usually, processing step + representation level (fragments, tran- sitions, proteins))

### Value

list of two items: plot =gg, df = pvca\_res

#### See Also

sample\_annotation\_to\_colors, ggplot

### Examples

```
matrix <- example_proteome_matrix[1:50, ]
pvca_plot <- plot_PVCA(matrix, example_sample_annotation,
technical_covariates = c('MS_batch', 'digestion_batch'),
biological_covariates = c("Diet", "Sex", "Strain"))</pre>
```

plot\_sample\_corr\_distribution

Create violin plot of correlation distribution

### Description

Useful to visualize within batch vs within replicate vs non-related sample correlation

### Usage

```
plot_sample_corr_distribution(data_matrix, sample_annotation,
    repeated_samples = NULL, sample_id_col = "FullRunName",
    batch_col = "MS_batch", biospecimen_id_col = "EarTag",
    plot_title = "Correlation distribution",
    plot_param = "batch_replicate")
```

#### Arguments

features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data		
ion		
data matrix with 1) sample_id_col (this can be repeated as row names) 2) bio- logical and 3) technical covariates (batches etc)		
repeated_samples		
if NULL, only repeated sample correlation is plotted		
name of the column in sample_annotation file, where the filenames (colnames of the data matrix) are found		
column in sample_annotation that should be used for batch comparison		

biospecimen_id_col		
	column in sample_annotation that captures the biological sample, that (possibly) was profiled several times as technical replicates. Tip: if such ID is absent, but can be defined from several columns, create new biospecimen_id column	
plot_title	Title of the plot (usually, processing step + representation level (fragments, tran- sitions, proteins))	
plot_param	columns, defined in correlation_df, which is output of get_sample_corr_distrib, specifically, $\#$	
	1. replicate	
	2. batch_the_same	
	3. batch_replicate	
	4. batches	
	;	

### Value

ggplot type object with violin plot for each plot\_param

### See Also

get\_sample\_corr\_distrib, ggplot

### Examples

```
sample_corr_distribution_plot <- plot_sample_corr_distribution(
example_proteome_matrix,
example_sample_annotation, batch_col = 'MS_batch',
biospecimen_id_col = "EarTag",
plot_param = 'batch_replicate')</pre>
```

plot\_sample\_corr\_heatmap

Sample correlation matrix (heatmap)

### Description

Plot correlation of selected samples

### Usage

```
plot_sample_corr_heatmap(data_matrix, samples_to_plot = NULL,
  flavor = c("pheatmap", "corrplot"), filename = NULL, width = NA,
  height = NA, unit = c("cm", "in", "mm"),
  plot_title = sprintf("Correlation matrix of sample %s",
  samples_to_plot), ...)
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data
samples_to_plot	:
	string vector of samples in data_matrix to be used in the plot
flavor	either corrplot from 'corrplot' package or heatmap, as in 'pheatmap'
filename	path where the results are saved. If null the object is returned to the active window; otherwise, the object is save into the file. Currently only pdf and png format is supported
width	option determining the output image width
height	option determining the output image width
unit	units: 'cm', 'in' or 'mm'
plot_title	Title of the plot (usually, processing step + representation level (fragments, tran- sitions, proteins))
	parameters for the corrplot.mixed or pheatmap visualisation, for details see examples and help to corresponding functions

### Value

corrplot or pheatmap object depending on flavor

### See Also

pheatmap, corrplot.mixed

### Examples

```
specified_samples = example_sample_annotation$FullRunName[
which(example_sample_annotation$order %in% 110:115)]
```

```
sample_corr_heatmap <- plot_sample_corr_heatmap(example_proteome_matrix,
samples_to_plot = specified_samples,
flavor = 'pheatmap',
cluster_rows= FALSE, cluster_cols=FALSE,
annotation_names_col = TRUE, annotation_legend = FALSE,
show_colnames = FALSE)
```

plot\_sample\_mean\_or\_boxplot

*Plot per-sample mean or boxplot (showing median and quantiles) vs order (if the real running order available)* 

### Description

Plot per-sample mean or boxplot (showing median and quantiles) vs order (if the real running order available)

#### Usage

```
plot_sample_mean(data_matrix, sample_annotation = NULL,
    sample_id_col = "FullRunName", order_col = "order",
    batch_col = "MS_batch", facet_col = NULL, color_by_batch = FALSE,
    color_scheme = "brewer", theme = "classic", plot_title = NULL,
    order_per_facet = FALSE, vline_color = "grey", ylimits = NULL)
plot_boxplot(df_long, sample_annotation = NULL,
```

```
sample_id_col = "FullRunName", measure_col = "Intensity",
order_col = "order", batch_col = "MS_batch", facet_col = NULL,
color_by_batch = TRUE, color_scheme = "brewer", theme = "classic",
plot_title = NULL, order_per_facet = FALSE)
```

### Arguments

data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. in most function,			
<pre>sample_annotat:</pre>	ion			
	data matrix with 1) sample_id_col (this can be repeated as row names) 2) bio- logical and 3) technical covariates (batches etc)			
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)			
order_col	column where running order is specified.			
batch_col	column in sample_annotation that should be used for batch comparison			
facet_col	recommended if more than one batch covariate is present. Faceting is most suited to examine instruments separately			
color_by_batch	should the each batch be represented with its own color?			
color_scheme	named vector, names corresponding to unique batch values as specified in sample_annotation			
theme	ggplot theme, by default classic. Can be easily overriden (see examples)			
plot_title	Title of the plot (usually, processing step + representation level (fragments, tran- sitions, proteins))			
order_per_face	order_per_facet			
	if order is defined ignoring facets (usually instrument), re-define order per-batch			
vline_color	color of vertical lines, typically denoting different MS batches in ordered runs; should be NULL for experiments without intrinsic order			
ylimits	range of y-axis to plot feature-level trends			
df_long	data frame where each row is a single feature in a single sample, thus it has minimally, sample_id_col, feature_id_col and measure_col, but usually also m_score (in OpenSWATH output result file)			
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity, otherwise, it is used internally for consistency			

### Details

functions for quick visual assessment of trends associated, overall or specific covariate-associated (see batch\_col and facet\_col)

### Value

ggplot2 class object. Thus, all aesthetics can be overriden

#### plot\_single\_feature

#### See Also

ggplot

### Examples

```
mean_plot <- plot_sample_mean(example_proteome_matrix, example_sample_annotation,
order_col = 'order', batch_col = "MS_batch")
```

```
boxplot <- plot_boxplot(example_proteome, example_sample_annotation,
batch_col = "MS_batch")
```

plot\_single\_feature Plot peptide measurements

#### Description

Creates a peptide facetted ggplot2 plot of the value in measure\_col vs order\_col. Additionally, the resulting plot can also be facetted by batch.

### Usage

```
plot_single_feature(pep_name, df_long, sample_annotation,
    order_col = "order", sample_id_col = "FullRunName",
    batch_col = "MS_batch", measure_col = "Intensity",
    feature_id_col = "peptide_group_label", geom = c("point", "line"),
    color_by_batch = FALSE, color_scheme = "brewer",
    facet_by_batch = FALSE, color_by_col = NULL, color_by_value = NULL,
    plot_title = NULL, vline_color = "red", theme = "classic")
```

### Arguments

pep_name	name of the peptide for diagnostic profiling
df_long	data frame where each row is a single feature in a single sample. It minimally has a sample_id_col, a feature_id_col and a measure_col, but usually also an m_score (in OpenSWATH output result file)
<pre>sample_annotat</pre>	ion
	data matrix with:
	<ol> <li>sample_id_col (this can be repeated as row names)</li> <li>biological covariates</li> </ol>
	3. technical covariates (batches etc)
order_col	column in sample_annotation that determines sample order. It is used for certain diagnostics and normalisations.
sample_id_col	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
batch_col	column in sample_annotation that should be used for batch comparison
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency

feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
geom	whether to show the feature as points and/or connect by lines
color_by_batch	(logical) whether to color points by batch
color_scheme	color scheme for ggplot representation
<pre>facet_by_batch</pre>	(logical) whether to plot each batch in its own facet
color_by_col	column to color by certain value denoted by color_by_value
color_by_value	value in color_by_col to color
plot_title	the string indicating the source of the peptides
vline_color	color of vertical lines, typically denoting different MS batches in ordered runs; should be NULL for experiments without intrinsic order
theme	plot theme (default is 'classical'; other options not implemented)

### Value

ggplot2 type plot of measure\_col vs order\_col, faceted by pep\_name and (optionally) by batch\_col

### See Also

Other feature-level diagnostic functions: plot\_iRT, plot\_peptides\_of\_one\_protein, plot\_spike\_in, plot\_with\_fitting\_curve

#### Examples

single\_feature\_plot <- plot\_single\_feature(
pep\_name = "46213\_NVGVSFYADKPEVTQEQK\_2",
df\_long = example\_proteome, example\_sample\_annotation,
color\_by\_col = NULL)</pre>

plot\_spike\_in Plot spike-in measurements

### Description

Creates a spike-in facetted ggplot2 plot of the value in measure\_col vs order\_col using plot\_single\_feature. Additionally, the resulting plot can also be facetted by batch.

### Usage

```
plot_spike_in(df_long, sample_annotation, peptide_annotation = NULL,
    protein_col = "ProteinName", order_col = "order",
    spike_ins = "BOVIN", sample_id_col = "FullRunName",
    batch_col = "MS_batch", measure_col = "Intensity",
    feature_id_col = "peptide_group_label", color_by_batch = FALSE,
    color_scheme = "brewer", facet_by_batch = FALSE,
    color_by_col = NULL, color_by_value = NULL,
    plot_title = "Spike-in BOVINE protein peptides", ...)
```

#### Arguments

df\_long data frame where each row is a single feature in a single sample. It minimally has a sample\_id\_col, a feature\_id\_col and a measure\_col, but usually also an m\_score (in OpenSWATH output result file)

#### sample\_annotation

data matrix with:

- 1. sample\_id\_col (this can be repeated as row names)
- 2. biological covariates
- 3. technical covariates (batches etc)

i	ne	nt	tί	de	annotation
		ν	ιт	uc_	

long format data with peptide ID and their corresponding protein annotations

- protein\_col column where protein names are specified
- order\_col column in sample\_annotation that determines sample order. It is used for certain diagnostics and normalisations.
- spike\_ins substring used to identify spike-in proteins in the column 'ProteinName'
- sample\_id\_col name of the column in sample\_annotation file, where the filenames (colnames of the data matrix are found)
- batch\_col column in sample\_annotation that should be used for batch comparison
- measure\_col if df\_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency
- feature\_id\_col name of the column with feature/gene/peptide/protein ID used in the long format representation df\_long. In the wide formatted representation data\_matrix this corresponds to the row names.
- color\_by\_batch (logical) whether to color points by batch
- color\_scheme color scheme for ggplot representation
- facet\_by\_batch (logical) whether to plot each batch in its own facet
- color\_by\_col column to color by certain value denoted by color\_by\_value
- color\_by\_value value in color\_by\_col to color
- plot\_title the string indicating the source of the peptides

... additional arguments to plot\_single\_feature function

### Value

ggplot2 type plot of measure\_col vs order\_col, faceted by spike\_ins containing proteins and (optionally) by batch\_col

#### See Also

Other feature-level diagnostic functions: plot\_iRT, plot\_peptides\_of\_one\_protein, plot\_single\_feature, plot\_with\_fitting\_curve

### Examples

```
spike_in_plot <- plot_spike_in(example_proteome, example_sample_annotation,
protein_col = 'Gene', spike_ins = "BOVINE_A1ag",
plot_title = "Spike-in BOVINE protein peptides")
```

#### plot\_with\_fitting\_curve

Plot peptide measurements across multi-step analysis

#### Description

Plot Intensity of a few representative peptides for each step of the analysis including the fitting curve

#### Usage

```
plot_with_fitting_curve(pep_name, df_long, sample_annotation, fit_df,
  fit_value_var = "fit", order_col = "order",
  sample_id_col = "FullRunName", batch_col = "MS_batch",
  measure_col = "Intensity", feature_id_col = "peptide_group_label",
  geom = c("point", "line"), color_by_batch = FALSE,
  color_scheme = "brewer", facet_by_batch = FALSE,
  plot_title = sprintf("Fitting curve of %s peptide", pep_name),
  color_by_col = NULL, color_by_value = NULL, theme = "classic",
  vline_color = "grey", ...)
```

### Arguments

pep_name	name of the peptide for diagnostic profiling
df_long	data frame where each row is a single feature in a single sample. It minimally has a sample_id_col, a feature_id_col and a measure_col, but usually also an m_score (in OpenSWATH output result file)
sample_annotat:	ion
	data matrix with:
	1. sample_id_col (this can be repeated as row names)
	2. biological covariates
	3. technical covariates (batches etc)
fit_df	data frame typically output generated from nonlinear curve fitting by normalize_custom_fit
fit_value_var	column denoting intensity values, typically fitted to curve
order_col	column in sample_annotation that determines sample order. It is used for certain diagnostics and normalisations.
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
batch_col	column in sample_annotation that should be used for batch comparison
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
geom	for the intensity measure_col profile
color_by_batch	(logical) whether to color points by batch
color_scheme	color scheme for ggplot representation

#### proBatch

facet_by_batch	(logical) whether to plot each batch in its own facet
<pre>plot_title</pre>	the string indicating the source of the peptides
color_by_col	column to color by certain value denoted by color_by_value
color_by_value	value in color_by_col to color
theme	plot theme (default is 'classical'; other options not implemented)
vline_color	color of vertical lines, typically denoting different MS batches in ordered runs; should be NULL for experiments without intrinsic order
	additional arguments to plot_single_feature function

### Value

ggplot-class plot with minimally two facets (before and after non-linear fit) with measure\_col (Intensity) vs order\_col (injection order) for selected peptides (specified in pep\_name)

### See Also

Other feature-level diagnostic functions: plot\_iRT, plot\_peptides\_of\_one\_protein, plot\_single\_feature, plot\_spike\_in

#### Examples

```
loess_fit_70 <- adjust_batch_trend(example_proteome_matrix,
example_sample_annotation, span = 0.7)
```

```
fitting_curve_plot <- plot_with_fitting_curve(
pep_name = "10231_QDVDVWLWQQEGSSK_2",
df_long = example_proteome, example_sample_annotation,
fit_df = loess_fit_70$fit_df, plot_title = "Curve fitting with 70% span")</pre>
```

proBatch

proBatch: A package for diagnostics and correction of batch effects, primarily in proteomics

### Description

The proBatch package contains functions for analyzing and correcting batch effects and other unwanted technical variation from high-thoughput experiments. Although the package has primarily been developed for mass spectrometry proteomics (DIA/SWATH), it should also be applicable to most omic data with minor adaptations. It addresses the following needs:

- prepare the data for analysis
- Visualize batch effects in sample-wide and feature-level;
- Normalize and correct for batch effects.

### Arguments

df_long	data frame where each row is a single feature in a single sample. It minimally has a sample_id_col, a feature_id_col and a measure_col, but usually also an m_score (in OpenSWATH output result file)
data_matrix	features (in rows) vs samples (in columns) matrix, with feature IDs in rownames and file/sample names as colnames. Usually the log transformed version of the original data
<pre>sample_annotati</pre>	ion
	data matrix with:
	<ol> <li>sample_id_col (this can be repeated as row names)</li> <li>biological covariates</li> <li>table is the particular table at the set of the set of</li></ol>
	3. technical covariates (batches etc)
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)
batch_col	column in sample_annotation that should be used for batch comparison
order_col	column in sample_annotation that determines sample order. It is used for certain diagnostics and normalisations.
measure_col	if df_long is among the parameters, it is the column with expression/abundance/intensity; otherwise, it is used internally for consistency
feature_id_col	name of the column with feature/gene/peptide/protein ID used in the long format representation df_long. In the wide formatted representation data_matrix this corresponds to the row names.
plot_title	Title of the plot (usually, processing step + representation level (fragments, tran- sitions, proteins))
theme	ggplot theme, by default classic. Can be easily overriden

### Details

To learn more about proBatch, start with the vignettes: browseVignettes(package = "proBatch")

### Section

Common arguments to the functions.

quantile_normalize	Quantile normalization of the data, ensuring that the row and column
	names are retained

### Description

Quantile normalization of the data, ensuring that the row and column names are retained

### Usage

```
quantile_normalize(data_matrix)
```

### Arguments

data\_matrix log transformed data matrix (features in rows and samples in columns)

sample\_annotation\_to\_colors

### Value

 ${\tt data\_matrix-size\ matrix,\ with\ columns\ quantile-normalized}$ 

#### Examples

quantile\_normalized\_matrix <- quantile\_normalize(example\_proteome\_matrix)</pre>

sample\_annotation\_to\_colors

Generate colors for sample annotation

### Description

Convert the sample annotation data frame to list of colors the list is named as columns included to use in potting functions

### Usage

```
sample_annotation_to_colors(sample_annotation,
    columns_for_plotting = NULL, sample_id_col = "FullRunName",
    factor_columns = c("MS_batch", "EarTag", "Strain", "Diet", "Sex"),
    not_factor_columns = "DateTime", numeric_columns = "order",
    rare_categories_to_other = TRUE, numeric_palette_type = "brewer",
    granularity = 10)
```

### Arguments

sample_annotation		
	data matrix with:	
	1. sample_id_col (this can be repeated as row names)	
	2. biological covariates	
	3. technical covariates (batches etc)	
columns_for_plo	otting	
	only consider these columns from sample_annotation	
<pre>sample_id_col</pre>	name of the column in sample_annotation file, where the filenames (colnames of the data matrix are found)	
factor_columns	columns of sample_annotation to be treated as factors. Note that factor and character columns are treated as factors by default.	
not_factor_colu	Jmns	
	don't treat these columns as factors. This can be used to override the default behaviour of considering factors and character columns as factors.	
numeric_columns	5	
	columns of sample_annotation to be treated as continuous numeric values.	
rare_categories_to_other		
	if True rare categories will be merged as 'other'	
numeric_palette_type		
	palette to be used for numeric values coloring	
granularity	number of colors to map to the number vector (equally spaced between mini- mum and maximum)	

### Value

list of colors

### Examples

```
color_scheme <- sample_annotation_to_colors (example_sample_annotation,
factor_columns = c('MS_batch','EarTag', "Strain",
"Diet", "digestion_batch", "Sex"),
not_factor_columns = 'DateTime',
numeric_columns = c('order'))
```

sample\_color\_scheme Sample color annotation

### Description

This is an color scheme generated from example sample annotation

### Usage

sample\_color\_scheme

### Format

A list of 3 components: list\_of\_colors, color\_df and sample\_annotation

- color\_df a data frame with 233 samples and 11 variables describing a color for each component
- **sample\_anotation** a data frame containing 233 samples and 11 variables annotating samples to facilitate conversion to a color scheme

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