

# Package ‘specmine’

July 10, 2020

**Type** Package

**Title** Metabolomics and Spectral Data Analysis and Mining

**Version** 3.0.1

**Date** 2020-07-10

**Author** Christopher Costa <chrisbcl@hotmail.com> [aut],  
Marcelo Maraschin <mtocsy@gmail.com> [aut],  
Miguel Rocha <mrocha@di.uminho.pt> [aut, cre],  
Sara Cardoso <saracardoso501@gmail.com> [aut],  
Telma Afonso <telma.afonso94@gmail.com> [aut],  
C. Beleites [cph],  
Jie Hao [cph]

**Maintainer** Miguel Rocha <mrocha@di.uminho.pt>

**Depends** R (>= 4.0.0)

**SystemRequirements** Python (>=3.5.2) and the following python module:  
nmrplug.

**Imports** hyperSpec, caret, e1071, ggplot2, impute, ellipse, GGally,  
pcaPP, compare, baseline, MASS, pls, readJDX, speaq, gdata,  
genefilter, RColorBrewer, grDevices, graphics, methods, stats,  
utils, Metrics

**Suggests** ggdendro, reticulate, qdap, scatterplot3d, MAIT, xcms,  
KEGGgraph, KEGGREST, rcytoscapejs, rgl, grid, curl, RCurl

**LazyData** true

**Description** Provides a set of methods for metabolomics data analysis, including data loading in different formats, pre-processing, metabolite identification, univariate and multivariate data analysis, machine learning, feature selection and pathway analysis. Case studies can be found on the website: <<http://bio.di.uminho.pt/metabolomicspackage/index.html>>. This package suggests 'rcytoscapejs', a package not in mainstream repositories. If you need to install it, use: `devtools::install_github('cytoscape/r-cytoscape.js@v0.0.7')`.

**License** GPL (>= 2)

**NeedsCompilation** no

Repository CRAN

Date/Publication 2020-07-10 15:20:03 UTC

## R topics documented:

absorbance_to_transmittance . . . . .	6
aggregate_samples . . . . .	7
aov_all_vars . . . . .	7
apply_by_group . . . . .	8
apply_by_groups . . . . .	9
apply_by_sample . . . . .	9
apply_by_variable . . . . .	10
background_correction . . . . .	11
baseline_correction . . . . .	11
boxplot_variables . . . . .	12
boxplot_vars_factor . . . . .	13
cachexia . . . . .	14
check_dataset . . . . .	14
clustering . . . . .	15
compare_regions_by_sample . . . . .	16
convert_chebi_to_kegg . . . . .	16
convert_from_chemospec . . . . .	17
convert_from_hyperspec . . . . .	18
convert_hmdb_to_kegg . . . . .	18
convert_keggpathway_2_reactiongraph . . . . .	19
convert_multiple_spcmmn_to_kegg . . . . .	19
convert_to_factor . . . . .	20
convert_to_hyperspec . . . . .	20
correlations_dataset . . . . .	21
correlations_test . . . . .	21
correlation_test . . . . .	22
count_missing_values . . . . .	23
count_missing_values_per_sample . . . . .	24
count_missing_values_per_variable . . . . .	24
create_dataset . . . . .	25
create_pathway_with_reactions . . . . .	26
cubic_root_transform . . . . .	27
dataset_from_peaks . . . . .	28
data_correction . . . . .	28
dendrogram_plot . . . . .	29
dendrogram_plot_col . . . . .	30
detect_nmr_peaks_from_dataset . . . . .	30
feature_selection . . . . .	31
filter_feature_selection . . . . .	32
find_equal_samples . . . . .	33
first_derivative . . . . .	33
flat_pattern_filter . . . . .	34

fold_change . . . . .	35
fold_change_var . . . . .	36
get_cpd_names . . . . .	36
get_data . . . . .	37
get_data_as_df . . . . .	38
get_data_value . . . . .	38
get_data_values . . . . .	39
get_files_list_per_assay . . . . .	40
get_metabolights_study . . . . .	40
get_metabolights_study_files_assay . . . . .	41
get_metabolights_study_metadata_assay . . . . .	42
get_metabolights_study_samples_files . . . . .	43
get_MetabolitePath . . . . .	44
get_metabPaths_org . . . . .	44
get_metadata . . . . .	45
get_metadata_value . . . . .	45
get_metadata_var . . . . .	46
get_OrganismsCodes . . . . .	46
get_paths_with_cpds_org . . . . .	47
get_peak_values . . . . .	48
get_samples_names_dx . . . . .	48
get_samples_names_spc . . . . .	49
get_sample_names . . . . .	49
get_type . . . . .	50
get_value_label . . . . .	50
get_x_label . . . . .	51
get_x_values_as_num . . . . .	51
get_x_values_as_text . . . . .	52
group_peaks . . . . .	52
heatmap_correlations . . . . .	53
hierarchical_clustering . . . . .	54
impute_nas_knn . . . . .	54
impute_nas_linapprox . . . . .	55
impute_nas_mean . . . . .	56
impute_nas_median . . . . .	56
impute_nas_value . . . . .	57
indexes_to_xvalue_interval . . . . .	57
is_spectra . . . . .	58
kmeans_clustering . . . . .	58
kmeans_plot . . . . .	59
kmeans_result_df . . . . .	59
kruskalTest_dataset . . . . .	60
ksTest_dataset . . . . .	61
linregression_onevar . . . . .	61
linreg_all_vars . . . . .	62
linreg_coef_table . . . . .	62
linreg_pvalue_table . . . . .	63
linreg_rsquared . . . . .	63

log_transform . . . . .	64
low_level_fusion . . . . .	65
MAIT_identify_metabolites . . . . .	65
mean_centering . . . . .	66
merge_datasets . . . . .	66
merge_data_metadata . . . . .	67
metabolights_studies_list . . . . .	68
metadata_as_variables . . . . .	68
missingvalues_imputation . . . . .	69
msc_correction . . . . .	70
multiClassSummary . . . . .	70
multifactor_aov_all_vars . . . . .	71
multifactor_aov_pvalues_table . . . . .	71
multifactor_aov_varexp_table . . . . .	72
multiplot . . . . .	73
nmr_identification . . . . .	74
normalize . . . . .	76
normalize_samples . . . . .	77
num_samples . . . . .	78
num_x_values . . . . .	78
offset_correction . . . . .	79
pathway_analysis . . . . .	79
pca_analysis_dataset . . . . .	80
pca_biplot . . . . .	81
pca_biplot3D . . . . .	82
pca_importance . . . . .	82
pca_kmeans_plot2D . . . . .	83
pca_kmeans_plot3D . . . . .	84
pca_pairs_kmeans_plot . . . . .	85
pca_pairs_plot . . . . .	85
pca_plot_3d . . . . .	86
pca_robust . . . . .	87
pca_scoresplot2D . . . . .	88
pca_scoresplot3D . . . . .	89
pca_scoresplot3D_rgl . . . . .	89
pca_screplot . . . . .	90
peaks_per_sample . . . . .	91
peaks_per_samples . . . . .	91
plotvar_twofactor . . . . .	92
plot_anova . . . . .	93
plot_fold_change . . . . .	93
plot_kruskaltest . . . . .	94
plot_kstest . . . . .	95
plot_peaks . . . . .	95
plot_regression_coefs_pvalues . . . . .	96
plot_spectra . . . . .	97
plot_spectra_simple . . . . .	97
plot_ttests . . . . .	98

predict_samples . . . . .	99
propolis . . . . .	99
propolisSampleList . . . . .	100
read_Bruker_files . . . . .	101
read_csvs_folder . . . . .	102
read_dataset_csv . . . . .	102
read_dataset_dx . . . . .	103
read_dataset_spc . . . . .	104
read_data_csv . . . . .	105
read_data_dx . . . . .	106
read_data_spc . . . . .	106
read_metadata . . . . .	107
read_ms_spectra . . . . .	107
read_multiple_csvs . . . . .	108
read_spc_nosubhdr . . . . .	109
read_varian_spectra_raw . . . . .	110
recursive_feature_elimination . . . . .	111
remove_data . . . . .	112
remove_data_variables . . . . .	113
remove_metadata_variables . . . . .	114
remove_peaks_interval . . . . .	114
remove_peaks_interval_sample_list . . . . .	115
remove_samples . . . . .	116
remove_samples_by_nas . . . . .	116
remove_samples_by_na_metadata . . . . .	117
remove_variables_by_nas . . . . .	118
remove_x_values_by_interval . . . . .	118
replace_data_value . . . . .	119
replace_metadata_value . . . . .	120
savitzky_golay . . . . .	120
scaling . . . . .	121
scaling_samples . . . . .	122
set_metadata . . . . .	122
set_sample_names . . . . .	123
set_value_label . . . . .	124
set_x_label . . . . .	124
set_x_values . . . . .	125
shift_correction . . . . .	125
smoothing_interpolation . . . . .	126
snv_dataset . . . . .	127
spectra_options . . . . .	127
stats_by_sample . . . . .	128
stats_by_variable . . . . .	129
subset_by_samples_and_xvalues . . . . .	129
subset_metadata . . . . .	130
subset_random_samples . . . . .	131
subset_samples . . . . .	131
subset_samples_by_metadata_values . . . . .	132

subset_x_values . . . . .	133
subset_x_values_by_interval . . . . .	133
summary_var_importance . . . . .	134
sum_dataset . . . . .	135
train_and_predict . . . . .	136
train_classifier . . . . .	137
train_models_performance . . . . .	138
transform_data . . . . .	139
transmittance_to_absorbance . . . . .	140
tTests_dataset . . . . .	140
values_per_peak . . . . .	141
values_per_sample . . . . .	141
variables_as_metadata . . . . .	142
volcano_plot_fc_tt . . . . .	143
xvalue_interval_to_indexes . . . . .	143
x_values_to_indexes . . . . .	144

<b>Index</b>	<b>145</b>
--------------	------------

---

absorbance\_to\_transmittance  
*Convert absorbance to transmittance*

---

## Description

Converts absorbance values to transmittance values.

## Usage

```
absorbance_to_transmittance(dataset)
```

## Arguments

dataset            list representing the dataset from a metabolomics experiment.

## Value

Returns the dataset with the data points converted to transmittance values.

---

aggregate_samples	<i>Aggregate samples</i>
-------------------	--------------------------

---

**Description**

Aggregate samples according to an aggregate function like mean, median, etc.

**Usage**

```
aggregate_samples(dataset, indexes, aggreg.fn = "mean",  
meta.to.remove = c())
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
indexes	index vector with the samples that are going to be aggregated (e.g. c(1,1,2,2), this index vector will aggregate the first two samples and the last two samples).
aggreg.fn	aggregation function (e.g. "mean", "median", etc).
meta.to.remove	metadata's variables to be removed.

**Value**

Returns the dataset with the samples aggregated.

**Examples**

```
## Example of aggregating samples  
propolis_proc = missingvalues_imputation(propolis)  
dataset = aggregate_samples(propolis_proc, as.integer(propolis$metadata$seasons), "mean")
```

---

aov_all_vars	<i>Analysis of variance</i>
--------------	-----------------------------

---

**Description**

Perform analysis of variance of all variables in the dataset.

**Usage**

```
aov_all_vars(dataset, column.class, doTukey = TRUE, write.file = FALSE,  
file.out = "anova-res.csv")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
column.class	string or index indicating what metadata to use.
doTukey	boolean value for do or do not TukeyHSD.
write.file	boolean value indicating if a file with the results is written or not.
file.out	name of the file if write.file is TRUE.

**Value**

Data frame with the results of ANOVA, with p-value, logarithm of p-value, false discovery rate (fdr) and tukey is doTukey is TRUE. The result is ordered by p-value.

**Examples**

```
## Example of ANOVA with TukeyHSD
propolis_proc = missingvalues_imputation(propolis)
propolis_proc = flat_pattern_filter(propolis_proc, "iqr", by.percent = TRUE,
red.value = 75)
result = aov_all_vars(propolis_proc, "seasons", doTukey = FALSE)
```

---

apply_by_group	<i>Apply by group</i>
----------------	-----------------------

---

**Description**

Apply a function to samples from a given metadata's group.

**Usage**

```
apply_by_group(dataset, fn.to.apply, metadata.var, var.value)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
fn.to.apply	function to apply (e.g. mean, max, min).
metadata.var	name of the metadata's variable.
var.value	value of the metadata's variable.

**Value**

Returns a vector with the variables and the value of the applied function.

**Examples**

```
## Example of applying a function to a group
data(cachexia)
apply.group.result = apply_by_group(cachexia, mean, "Muscle.loss",
"control")
```



---

apply_by_groups	<i>Apply by groups</i>
-----------------	------------------------

---

**Description**

Apply a function to samples from a metadata's variable.

**Usage**

```
apply_by_groups(dataset, metadata.var, fn.to.apply = "mean",  
variables = NULL, variable.bounds = NULL)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
metadata.var	name of the metadata's variable.
fn.to.apply	function to apply (e.g. mean, max, min).
variables	allows to define which variables to calculate the stats (if numbers, indexes are assumed).
variable.bounds	allow to define an interval of variables (if numeric).

**Value**

Returns a vector with the variables and the value of the applied function on the metadata's groups.

**Examples**

```
## Example of applying a function to groups  
data(cachexia)  
apply.groups.result = apply_by_groups(cachexia, "Muscle.loss", mean)
```

---

apply_by_sample	<i>Apply function to samples</i>
-----------------	----------------------------------

---

**Description**

Applies a function to the values of each sample

**Usage**

```
apply_by_sample(dataset, fn.to.apply, samples = NULL, ...)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
 fn.to.apply        function to apply (e.g. mean, max, min).  
 samples            if defined restricts the application to a given set of samples.  
 ...                additional parameters to apply function.

**Value**

Returns a vector with the samples and the value of the applied function.

**Examples**

```
## Example of applying a function to variables
data(cachexia)
apply.samples.result = apply_by_sample(cachexia, mean)
```

---

apply\_by\_variable        *Apply function to variables*

---

**Description**

Applies a function to the values of each variable

**Usage**

```
apply_by_variable(dataset, fn.to.apply, variables = NULL,
  variable.bounds = NULL, samples = NULL, ...)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
 fn.to.apply        function to apply (e.g. mean, max, min).  
 variables           allows to define which variables to calculate the stats (if numbers, indexes are assumed).  
 variable.bounds    allow to define an interval of variables (if numeric).  
 samples            if defined restricts the application to a given set of samples.  
 ...                additional parameters to apply function.

**Value**

Returns a vector with the variables and the value of the applied function.

**Examples**

```
## Example of applying a function to variables
data(cachexia)
apply.variables.result = apply_by_variable(cachexia, mean)
```

---

background\_correction *Background correction*

---

**Description**

Perform background correction on the spectra.

**Usage**

```
background_correction(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the dataset with background correction performed on the data.

**Examples**

```
## Example of background correction
library(hyperSpec)
data(flu)
flu.converted = convert_from_hyperspec(flu)
flu.corrected = background_correction(flu.converted)
```

---

baseline\_correction *Baseline correction*

---

**Description**

Performs baseline correction on the dataset.

**Usage**

```
baseline_correction(dataset, method = "modpolyfit", ...)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

method            string representing the baseline correction method. It can be one of these methods:

- "als" Asymmetric Least Squares, baseline correction by 2nd derivative constrained weighted regression

- "fillPeaks" An iterative algorithm using suppression of baseline by means in local windows
- "irls" Iterative Restricted Least Squares, an algorithm with primary smoothing and repeated baseline suppressions and regressions with 2nd derivative constraint
- "lowpass" Low-pass filter, an algorithm for removing baselines based on Fast Fourier Transform filtering
- "medianWindow" an implementation and extension of Mark S. Friedrichs' model-free algorithm
- "modpolyfit" Modified polynomial fitting, an implementation of Chad A. Lieber and Anita Mahadevan-Jansen's algorithm for polynomial fitting
- "peakDetection" A translation from Kevin R. Coombes et al.'s MATLAB code for detecting peaks and removing baselines
- "rfbaseline" Robust Baseline Estimation, Wrapper for Andreas F. Ruckstuhl, Matthew P. Jacobson, Robert W. Field, James A. Dodd's algorithm based on LOWESS and weighted regression
- "rollingBall" Ideas from Rolling Ball algorithm for X-ray spectra by M. A. Kneen and H. J. Annegarn. Variable window width has been left out

... Additional parameters of the baseline correction method.

### Value

Returns the dataset with the data's baseline corrected.

---

boxplot\_variables      *Boxplot of variables*

---

### Description

Boxplot of each variable of the dataset.

### Usage

```
boxplot_variables(dataset, variables = NULL, samples = NULL,
horizontal = TRUE, col = "lightblue", nchar.label = 10,
cex.axis = 0.8, ...)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
variables	vector with the variables names or a NULL value indicating all variables.
samples	vector with the samples names or a NULL value indicating all samples.
horizontal	boolean value indicating if the boxplots should be horizontal.
col	string that represents the color of the bodies of the boxplots.

nchar.label	number of characters to display the variables' names.
cex.axis	numeric value that indicates the amount by which the axis is magnified relative to the default.
...	additional parameters of boxplot function.

### Examples

```
## Example of showing the boxplot of a few variables
data(cachexia)
boxplot_variables(cachexia, variables = c("Creatine", "Serine", "Lactate"))
```

---

boxplot\_vars\_factor     *Boxplot of variables with metadata's variable factors*

---

### Description

Boxplot of variables with metadata's variable factors from the dataset.

### Usage

```
boxplot_vars_factor(dataset, meta.var, variables = NULL,
samples = NULL, horizontal = FALSE, nchar.label = 10, col = NULL,
vec.par = NULL, cex.axis = 0.8, ylabs = NULL, ...)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
meta.var	metadata's variable.
variables	vector with the variables names or a NULL value indicating all variables.
samples	vector with the samples names or a NULL value indicating all samples.
horizontal	boolean value indicating if the boxplots should be horizontal.
nchar.label	number of characters to display the variables' names.
col	string that represents the color of the bodies of the boxplots.
vec.par	vector with the disposition of the boxplots (rows, columns).
cex.axis	numeric value that indicates the amount by which the axis is magnified relative to the default.
ylabs	y-axis labels.
...	additional parameters of boxplot function.

### Examples

```
## Example of showing the boxplot factors of a few variables
data(cachexia)
boxplot_vars_factor(cachexia, "Muscle.loss", variables = c("Creatine", "Serine",
"Lactate"))
```

---

cachexia

*Human Cachexia data*

---

### Description

Cachexia is a complex metabolic syndrome associated with an underlying illness (such as cancer) and characterized by loss of muscle with or without loss of fat mass (Evans et al., 2008). A total of 77 urine samples were collected being 47 of them patients with cachexia, and 30 control patients.

### Usage

```
data(cachexia)
```

### Format

An object of class "list"

### Source

[MetaboAnalyst](#)

### References

Eisner et al. (2010) Learning to predict cancer-associated skeletal muscle wasting from 1h-nmr profiles of urinary metabolites *Metabolomics* 7:25-34

### Examples

```
data(cachexia)
sum_dataset(cachexia)
```

---

check\_dataset

*Check dataset*

---

### Description

Check if the dataset is valid and if not give the proper error message.

### Usage

```
check_dataset(dataset)
```

### Arguments

dataset      list representing the dataset from a metabolomics experiment.

**Value**

Message saying if the dataset is valid or invalid, in the last case also gives the reason.

**Examples**

```
## Example checking a dataset
data(cachexia)
check_dataset(cachexia)
```

---

clustering

*Perform cluster analysis*


---

**Description**

Perform cluster analysis on the dataset.

**Usage**

```
clustering(dataset, method = "hc", distance = "euclidean",
type = "samples", num.clusters = 5, clustMethod = "complete")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
method	a string describing the method of clustering. Possible types are: <ul style="list-style-type: none"> <li>• "hc" perform hierarchical clustering.</li> <li>• "kmeans" perform kmeans clustering.</li> </ul>
distance	the distance measure to be used to compute the distances between the rows of a data matrix. Possible types are "euclidean", "manhattan", "pearson" or "spearman". Only for hierarchical clustering.
type	a string indicating if cluster analysis will be performed on samples ("samples") or on variables ("variables").
num.clusters	the number of clusters in k-means cluster analysis.
clustMethod	Cluster method for hierarchical clustering.

**Value**

An object of class kmeans or hclust with the clustering results.

**Examples**

```
## Example of kmeans and hierarchical clustering
data(cachexia)
hc.result = clustering(cachexia, method = "hc",
distance = "euclidean")
kmeans.result = clustering(cachexia, method = "kmeans",
num.clusters = 4)
```

---

compare\_regions\_by\_sample  
*Compare regions by sample*

---

**Description**

Compare two regions of a dataset by samples.

**Usage**

```
compare_regions_by_sample(dataset1, dataset2, fn.to.apply,  
samples = NULL, ...)
```

**Arguments**

dataset1	list representing the dataset from a metabolomics experiment.
dataset2	list representing the dataset from a metabolomics experiment.
fn.to.apply	function to apply (e.g. mean, max, min).
samples	if defined restricts the application to a given set of samples.
...	additional parameters to apply.by.sample function.

**Value**

Returns a data.frame with the results of the function applied to the samples and the ration between the two datasets.

**Examples**

```
## Example of comparing regions by sample  
data(cachexia)  
subset1 = subset_x_values(cachexia, 1:31, by.index = TRUE)  
subset2 = subset_x_values(cachexia, 32:63, by.index = TRUE)  
comp.regions.result = compare_regions_by_sample(subset1, subset2,  
mean)
```

---

convert\_chebi\_to\_kegg *Convert CHEBI codes to KEGG codes.*

---

**Description**

Converts a vector of CHEBI codes into a vector of the corresponding KEGG codes. This is performed by using our internal library used in NMR identification, it will not have all chebi codes.

**Usage**

```
convert_chebi_to_kegg(chebi_codes)
```



**Arguments**

chebi\_codes      Vector with the CHEBI codes (each chebi must be structured like: CHEBI:<number>)

**Value**

Named vector with kegg codes and respective names. Vector names are the compound names and the vector elements the kegg codes.

**Examples**

```
keggs=convert_hmdb_to_kegg(c("CHEBI:15377", "CHEBI:26078", "CHEBI:30168"))
keggs
```

---

convert\_from\_chemospec

*Convert from ChemoSpec*

---

**Description**

Convert the dataset in the ChemoSpec format to a dataset of this package.

**Usage**

```
convert_from_chemospec(csobj, type = "undefined",
description = "")
```

**Arguments**

csobj            ChemoSpec object representing the dataset.  
type            string representing the type of the data.  
description    string representing the description of the dataset.

**Value**

Returns a list representing the dataset converted.

---

convert\_from\_hyperspec  
*Convert from hyperspec*

---

**Description**

Convert the dataset in the hyperspec format to a dataset of this package.

**Usage**

```
convert_from_hyperspec(hsobj, type = "undefined",  
description = "")
```

**Arguments**

hsobj	hyperspec object representing the dataset.
type	string representing the type of the data.
description	string representing the description of the dataset.

**Value**

Returns a list representing the dataset converted.

**Examples**

```
## Example of converting a dataset from hyperspec  
library(hyperSpec)  
data(flu)  
dataset = convert_from_hyperspec(flu, type = "undefined",  
description = "some description")
```

---

convert\_hmdb\_to\_kegg *Convert HMDB codes to KEGG codes.*

---

**Description**

Converts a vector of HMDB codes into a vector of the corresponding KEGG codes. This is performed by using our internal library used in NMR identification, it will not have all chebi codes.

**Usage**

```
convert_hmdb_to_kegg(hmdb_codes)
```

**Arguments**

hmdb_codes	Vector with the HMDB codes (each hmdb code must have 7 digits, e.g., HMDB0000001)
------------	---

**Value**

Named vector with kegg codes and respective names. Vector names are the compound names and the vector elements the kegg codes.

**Examples**

```
keggs=convert_hmdb_to_kegg(c("HMDB0000001", "HMDB0000008", "HMDB0000246"))
keggs
```

---

```
convert_keggpathway_2_reactiongraph
```

*Convert KEGGPathway object to graph object.*

---

**Description**

Converts KEGGPathway object into a graph object.

**Usage**

```
convert_keggpathway_2_reactiongraph(pathObj)
```

**Arguments**

pathObj            KEGGPathway object.

**Value**

Kegg reaction graph.

---

```
convert_multiple_spcnmn_to_kegg
```

*Convert specmine metabolite codes to KEGG codes.*

---

**Description**

Converts a vector of specmine metabolite codes into a vector of the corresponding KEGG codes. This is performed by using our internal library used in NMR identification, it will not have all chebi codes.

**Usage**

```
convert_multiple_spcnmn_to_kegg(spcnmn_codes)
```

**Arguments**

spcnmn\_codes      Vector with the SPCNMN codes (each chebi must be structured like: SPCNMN<number>)

**Value**

Named vector with kegg codes and respective names. Vector names are the compound names and the vector elements the kegg codes.

**Examples**

```
keggs=convert_multiple_spcnmn_to_kegg(c("SPCMNM2111", "SPCMNM2142", "SPCMNM069774"))
keggs
```

---

convert\_to\_factor      *Convert metadata to factor*

---

**Description**

Convert a metadata's variable to factor.

**Usage**

```
convert_to_factor(dataset, metadata.var)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
metadata.var      name of the metadata's variable.

**Value**

Returns the dataset with the metadata's variable converted to factor.

---

convert\_to\_hyperspec    *Convert to hyperspec*

---

**Description**

Convert a dataset to an hyperspec object.

**Usage**

```
convert_to_hyperspec(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns an hyperspec object representing the dataset converted.

---

correlations\_dataset    *Dataset correlations*

---

### Description

Calculate the correlations of all variables or samples in the dataset.

### Usage

```
correlations_dataset(dataset, method = "pearson", by.var = TRUE)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
method	correlation method, it can be "pearson", "kendall" or "spearman".
by.var	if TRUE then the correlations of the variables will be calculated, if not then the correlations of the samples will be calculated.

### Value

Returns the correlation matrix

### Examples

```
## Example of correlations of variables
data(cachexia)
corr.result = correlations_dataset(cachexia,
method = "pearson", by.var = TRUE)
```

---

correlations\_test    *Correlations test*

---

### Description

Performs correlations test to the whole dataset.

### Usage

```
correlations_test(dataset, method = "pearson", by.var = TRUE,
alternative = "two.sided")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
method	correlation method, it can be "pearson", "kendall" or "spearman".
by.var	if TRUE then the correlations of the variables will be calculated, if not then the correlations of the samples will be calculated.
alternative	alternative argument from cor.test of stats package. Can be "two.sided", "less" or "greater".

**Value**

Returns a matrix with the correlation values and the p-values

**Examples**

```
## Example of correlations test of variables
data(cachexia)
corr.result = correlations_test(cachexia,
method = "pearson", by.var = FALSE)
```

---

correlation_test	<i>Correlation test of two variables or samples</i>
------------------	---

---

**Description**

Performs correlations test of two variables or samples from the dataset.

**Usage**

```
correlation_test(dataset, x, y, method = "pearson",
alternative = "two.sided", by.var = TRUE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
x	first variable or sample.
y	second variable or sample.
method	correlation method, it can be "pearson", "kendall" or "spearman".
alternative	alternative argument from cor.test of stats package. Can be "two.sided", "less" or "greater".
by.var	if TRUE then the correlations of the variables will be calculated, if not then the correlations of the samples will be calculated.

**Value**

Returns the correlation result from cor.test function of stats package.

**Examples**

```
## Example of correlations test of variables
data(cachexia)
corr.result = correlation_test(cachexia, "Serine", "Creatine", method = "pearson",
by.var = TRUE)
```

---

count\_missing\_values    *Count missing values*

---

**Description**

Counts the missing values on the dataset.

**Usage**

```
count_missing_values(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the number of missing values on the dataset.

**Examples**

```
## Example of counting the missing values
data(cachexia)
count_missing_values(cachexia)
```

count\_missing\_values\_per\_sample  
*Count missing values per sample*

---

**Description**

Counts the missing values on each sample of the dataset.

**Usage**

```
count_missing_values_per_sample(dataset, remove.zero = TRUE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
remove.zero	boolean value indicating if the results of samples with no missing value are removed.

**Value**

Returns a vector with the number of missing values on each sample.

**Examples**

```
## Example of counting the missing values on each sample  
data(cachexia)  
cachexia$data[10,10] = NA  
count_missing_values_per_sample(cachexia)
```

---

count\_missing\_values\_per\_variable  
*Count missing values per variable*

---

**Description**

Counts the missing values on each variable of the dataset.

**Usage**

```
count_missing_values_per_variable(dataset, remove.zero = TRUE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
remove.zero	boolean value indicating if the results of variables with no missing value are removed.



**Value**

Returns a vector with the number of missing values on each sample.

**Examples**

```
## Example of counting the missing values on each variable
data(cachexia)
cachexia$data[10,10] = NA
count_missing_values_per_variable(cachexia)
```

---

create_dataset	<i>Create dataset</i>
----------------	-----------------------

---

**Description**

Create a dataset from existing objects

**Usage**

```
create_dataset(datamatrix, type = "undefined", metadata = NULL,
description = "", sample.names = NULL, x.axis.values = NULL,
label.x = NULL, label.values = NULL, xSet = NULL)
```

**Arguments**

datamatrix	matrix with numerical data: rows are assumed to be variables and columns assumed to be samples.
type	type of data: string that can be one of the following: <ul style="list-style-type: none"><li>• nmr-spectra</li><li>• nmr-peaks</li><li>• ir-spectra</li><li>• uvv-spectra</li><li>• raman-spectra</li><li>• fluor-spectra</li><li>• ms-spectra</li><li>• lcms-peaks</li><li>• gcms-peaks</li><li>• integrated-data</li><li>• concentrations</li><li>• undefined</li></ul>
metadata	data frame with the dataset's metadata: columns represent each metadata variable and rows represent the value of the metadata for the sample.
description	string with a short description of the dataset.

sample.names	vector with sample names, if NULL then the column names of datamatrix or sequential numbers will be used.
x.axis.values	vector with the x axis values, if NULL then the row names of datamatrix or sequential numbers will be used.
label.x	x axis label.
label.values	values label.
xSet	xcmsSet object from xcms package to store the reading and preprocessing results from MS spectra. Used for metabolite identification purposes.

**Value**

list representing the dataset:

data	matrix with the data
type	type of the data
description	short description of the dataset
metadata	data frame with the metadata variables
labels	list with labels of x axis and values
xSet	xcmsSet object

---

create\_pathway\_with\_reactions

*Creates the pathway, with reactions included in the nodes.*

---

**Description**

Creates a cytoscape pathway, where the reactions between compounds are also included in the nodes.

**Usage**

```
create_pathway_with_reactions(path, path.name, identified_cpds,
                             nodeNames="kegg", nodeTooltip=FALSE,
                             map.zoom=FALSE, map.layout="preset",
                             map.width=NULL, map.height=NULL)
```

**Arguments**

path	KEGGPathway object.
path.name	Name of the pathway.
identified_cpds	Vector of kegg codes to color differently in the map.
nodeNames	How the nodes should be named. If "kegg", nodes are named with kegg codes. If "names", nodes are named with the common names.

nodeTooltip	Boolean value indicating if tooltips of nodes should appear when hovering with the mouse. Does not work for all environments (e.g. can be used in shiny apps).
map.zoom	Boolean value indicating if a zoom widget should appear or not. Does not work for all environments (e.g. can be used in shiny apps).
map.layout	Layout of the map, available values are the ones of cytoscape ("breadthfirst", "preset", "cose", ...)
map.width	width of the map, in percentage (e.g. "80%"). May not work as expected in some environments.
map.height	Height of the map, in px (e.g. "500px"). May not work as expected in some environments.

---

cubic\_root\_transform *Cubic root transformation*

---

## Description

Performs cubic root transformation on the data matrix.

## Usage

```
cubic_root_transform(datamat)
```

## Arguments

datamat          data matrix.

## Value

Returns the data matrix with the cubic root transformation applied.

## Examples

```
## Example of cubic root transformation
data(cachexia)
datamat.cubic = cubic_root_transform(cachexia$data)
```

---

dataset_from_peaks	<i>Dataset from peaks</i>
--------------------	---------------------------

---

**Description**

Converts a peak list to a dataset.

**Usage**

```
dataset_from_peaks(sample.list, metadata = NULL,  
description = "", type = "nmr-peaks")
```

**Arguments**

sample.list	list with the peaks from each sample.
metadata	data frame with the associated metadata.
description	string with the description of the dataset.
type	string that represents the type of the data.

**Value**

Returns the dataset from the peak list.

**Examples**

```
## Example of converting a peak list to a dataset (computationally heavy)  
dataset = dataset_from_peaks(propolisSampleList, metadata = NULL,  
description = "some text", type = "nmr-peaks")
```

---

data_correction	<i>Data correction</i>
-----------------	------------------------

---

**Description**

Perform spectra corrections with 3 different methods.

**Usage**

```
data_correction(dataset, type = "background",  
method = "modpolyfit", ...)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
type	string that represents the type of correction that will be applied to the spectra. The three possible types are: "background", to perform background correction; "offset", to perform offset correction; and "baseline", to perform baseline correction.
method	string parameter of baseline correction indicating the correction method.
...	additional parameters of baseline correction.

**Value**

Returns the dataset with the spectra corrected.

---

dendrogram_plot	<i>Plot dendrogram</i>
-----------------	------------------------

---

**Description**

Plot dendrogram of hierarchical clustering results.

**Usage**

```
dendrogram_plot(dataset, hc.result, column.metadata = 1,
labels = NULL, ...)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
hc.result	object of class hclust with the clustering results.
column.metadata	string or index indicating what metadata to use to name the leafs.
labels	vector with the leaf names (optional).
...	other parameters for plotting.

**Examples**

```
### Example of a dendrogram
hc.result = hierarchical_clustering(cachexia)
dendrogram_plot(cachexia, hc.result)
```

---

dendrogram\_plot\_col *Plot dendrogram*

---

### Description

Plot dendrogram of hierarchical clustering results with different colors

### Usage

```
dendrogram_plot_col(dataset, hc.result, classes.col, colors = NULL, title = "",  
lab.cex = 1, leg.pos = "topright", label_samples=NULL,...)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
hc.result	object of class hclust with the clustering results.
classes.col	string or index indicating what metadata to use to color the leafs.
colors	vector with the corresponding colors of the metadata classes.
title	title of dendrogram.
lab.cex	the magnification to be used for x and y labels relative to the current setting of cex.
leg.pos	position of the legend.
label_samples	string or index indicating what metadata to use to name the leafs. If not provided the name of the leafs will remain the samples names.
...	other parameters for plotting.

### Examples

```
## Example of colored dendrogram  
data(cachexia)  
hc.result = hierarchical_clustering(cachexia)  
dendrogram_plot_col(cachexia, hc.result, "Muscle.loss",  
title = "Example")
```

---

detect\_nmr\_peaks\_from\_dataset

*Detection of the peaks in an NMR spectra dataset.*

---

### Description

This function detects the peaks, that have a minimum intensity of `baseline_tresh`, and performs alignment of those peaks.

**Usage**

```
detect_nmr_peaks_from_dataset(dataset, baseline_tresh=50000,
                              ap.method="own", ap.samp.classes=1, ap.step=0.03)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
baseline_tresh	Minimum intensity value that peaks must have. Peaks with intensity smaller than baseline_tresh will not be considered as detected peaks.
ap.method	Method to used in the alignment of peaks, after they are identified. Can be "own" or "metaboanalyst", which the later is for using the peak alignment used in MetaboAnalyst software.
ap.samp.classes	the metadata's variable to be used in the MetaboAnalyst method.
ap.step	step value for the peak alignment process

**Value**

Returns a dataset with the peaks detected and aligned.

---

feature_selection	<i>Perform feature selection</i>
-------------------	----------------------------------

---

**Description**

Perform feature selection on the dataset.

**Usage**

```
feature_selection(dataset, column.class, method = "rfe",
                  functions, validation = "cv", repeats = 5, number = 10,
                  subsets = 2^(2:4))
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
column.class	string or index indicating what metadata to use.
method	method used for feature selection. Possible values are "rfe" (recursive feature elimination) and "filter" (Selection by filter - sbf) from caret's package.
functions	a list of functions for model fitting, prediction and variable importance/filtering.
validation	the external resampling method: boot, cv, LOOCV or LGOCV (for repeated training/test splits).
repeats	for repeated k-fold cross-validation only: the number of complete sets of folds to compute.
number	either the number of folds or number of resampling iterations.
subsets	a numeric vector of integers corresponding to the number of features that should be retained (rfe only).

**Value**

caret's result from rfe or sbf.

**Examples**

```
## Example of feature selection using rfe and sbf
library(caret)
rfe.result = feature_selection(cachexia, "Muscle.loss",
                              method="rfe", functions = caret::rfFuncs,
                              validation = "cv", number = 3,
                              subsets = 2^(1:6))
sbf.result = feature_selection(cachexia, "Muscle.loss",
                              method="filter", functions = caret::rfSFBF,
                              validation = "cv")
```

---

filter\_feature\_selection

*Perform selection by filter*

---

**Description**

Perform selection by filter using univariate filters, from caret's package.

**Usage**

```
filter_feature_selection(datamat, samples.class,
                        functions = caret::rfSFBF, method = "cv", repeats = 5)
```

**Arguments**

datamat	data matrix from dataset.
samples.class	string or index indicating what metadata to use.
functions	a list of functions for model fitting, prediction and variable filtering.
method	the external resampling method: boot, cv, LOOCV or LGOCV (for repeated training/test splits).
repeats	for repeated k-fold cross-validation only: the number of complete sets of folds to compute.

**Value**

A caret's sbf object with the result of selection by filter.



**Examples**

```
## Example of selection by filter
library(caret)
rfe.result = filter_feature_selection(cachexia$data,
  cachexia$metadata$Muscle.loss, functions = caret::rfSBF,
  method = "cv")
```

---

find\_equal\_samples      *Find equal samples*

---

**Description**

Finds samples that have the same peak values - x and y (equal data frames)

**Usage**

```
find_equal_samples(sample.list)
```

**Arguments**

sample.list      list of data frames with the samples' peaks.

**Value**

Returns a dataframe with two columns indicating which pair of samples are equal.

**Examples**

```
## Example of finding equal samples
data(propolisSampleList)
equal.samples = find_equal_samples(propolisSampleList)
```

---

first\_derivative      *First derivative*

---

**Description**

Calculates the first derivative of the data.

**Usage**

```
first_derivative(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Return the dataset with the first derivative of the data calculated.

---

flat\_pattern\_filter    *Flat pattern filter*

---

**Description**

Performs a flat pattern filter over the dataset.

**Usage**

```
flat_pattern_filter(dataset, filter.function = "iqr",
  by.percent = TRUE, by.threshold = FALSE, red.value = 0)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

filter.function

filter function. It can be:

- **"iqr"** - Interquantile Range
- **"rsd"** - Relative Standard Deviation
- **"sd"** - Standard Deviation
- **"mad"** - Median Absolute Deviation
- **"mean"** - Mean
- **"median"** - Median

by.percent        boolean value, if TRUE the number of variables to filter will be a percentage of the number of variables in the dataset; percentage is given by the "red.value" parameter

by.threshold     boolean value, if TRUE, defines filtering will select variables where values of filtering function are below a given threshold. Threshold is defined by red.value that defines the minimum value of the function needed to keep the variable.

red.value         it can be the percentage or the threshold number. If red.value = "auto", will calculate number of variables to remove automatically

**Value**

Returns the dataset with the data filtered.

## Examples

```
## Example of flat pattern filter
dataset.filtered = flat_pattern_filter(propolis, "iqr", by.percent = TRUE,
  red.value = 20)
```

---

fold_change	<i>Fold change analysis</i>
-------------	-----------------------------

---

## Description

Perform fold change analysis on the dataset.

## Usage

```
fold_change(dataset, metadata.var, ref.value,
  threshold.min.fc = NULL, write.file = FALSE,
  file.out = "fold_change.csv")
```

## Arguments

dataset	list representing the dataset from a metabolomics experiment.
metadata.var	metadata to use to calculate the fold change.
ref.value	class name to indicate the initial value.
threshold.min.fc	minimum threshold of the fold change value.
write.file	boolean value to write or not a file with the results.
file.out	name of the file.

## Value

Table of results with fold change and log<sub>2</sub> of fold change.

## Examples

```
## Example of fold change
data(cachexia)
fold.change.results = fold_change(cachexia, "Muscle.loss",
  "control", write.file = FALSE)
```

---

fold_change_var	<i>Fold change applied on two variables</i>
-----------------	---

---

**Description**

Fold change applied on two variables. Instead of having the difference of the variables on two groups, we have the difference of the groups on two variables.

**Usage**

```
fold_change_var(dataset, metadata.var, variables,  
threshold.min.fc = NULL, write.file = FALSE,  
file.out = "fold_change_reverse.csv")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
metadata.var	metadata to use to calculate the fold change.
variables	vector with two positions containing the name of the variables.
threshold.min.fc	minimum threshold of the fold change value.
write.file	boolean value to write or not a file with the results.
file.out	name of the file.

**Value**

Table of results with fold change and log2 of fold change.

**Examples**

```
## Example of fold change reverse  
data(cachexia)  
fold.change.results = fold_change_var(cachexia, "Muscle.loss",  
c("Creatine", "Serine"))
```

---

get_cpd_names	<i>Get the names of the compounds that correspond to the kegg codes given.</i>
---------------	--

---

**Description**

Gets the common name of the compounds of the kegg codes given.

**Usage**

```
get_cpd_names(kegg_codes)
```

**Arguments**

kegg\_codes      Character vector with kegg codes.

**Value**

Named vector with the names of the compounds. The names of the vector are the compounds' names and the vector elements the kegg codes.

**Examples**

```
get_cpd_names(c("cpd:C00001", "cpd:C00008", "g1:G13099"))
```

---

get\_data

*Get data*

---

**Description**

Get the data matrix from dataset

**Usage**

```
get_data(dataset)
```

**Arguments**

dataset          list representing the dataset from a metabolomics experiment.

**Value**

Returns the data matrix

**Examples**

```
## Example of getting the data matrix  
data(cachexia)  
cachexia.dm = get_data(cachexia)
```

---

get_data_as_df	<i>Get data as data frame</i>
----------------	-------------------------------

---

**Description**

Get the data matrix from the dataset as a data frame.

**Usage**

```
get_data_as_df(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the data matrix from the dataset as a data.frame object.

**Examples**

```
## Example of getting the data matrix as data frame
data(cachexia)
cachexia.dt = get_data_as_df(cachexia)
```

---

get_data_value	<i>Get data value</i>
----------------	-----------------------

---

**Description**

Get a data value given the x-axis labels and the sample

**Usage**

```
get_data_value(dataset, x.axis.val, sample, by.index = FALSE)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
x.axis.val        index or name of the x-axis value.  
sample            index or name of the sample.  
by.index          boolean value indicating if the x-axis value and sample are represented as index or not.

**Value**

Returns a numeric with the data point value.

**Examples**

```
## Example of getting a data value from the dataset
data(cachexia)
data.value = get_data_value(cachexia, "Creatine", "PIF_178",
  by.index = FALSE)
```

---

get_data_values	<i>Get data values</i>
-----------------	------------------------

---

**Description**

Gets the values of all samples in the dataset given a set of x axis names or indexes.

**Usage**

```
get_data_values(dataset, x.axis.val, by.index = FALSE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
x.axis.val	vector with the values of the x axis (could be names or indexes).
by.index	boolean value indicating if the x.axis.val is a vector of indexes or not.

**Value**

Returns a matrix with the values of all samples in the specified x axis.

**Examples**

```
## Example of getting a metadata value
data(cachexia)
data.values = get_data_values(cachexia, c("Creatine", "Serine", "Lactate"),
  by.index = FALSE)
```

```
get_files_list_per_assay
```

*Get list of files per assay for MetaboLights study.*

---

**Description**

Returns a list of the data files in each assay of a MetaboLights study.

**Usage**

```
get_files_list_per_assay(studyID)
```

**Arguments**

studyID            ID of the metabolights study

**Value**

A list with one or more item. Each item corresponds to an assay of the MetaboLights study. Each item contains a data frame with the names of the samples (column 'Samples') and respective file names (column 'Files').

**References**

MetaboLights database: <https://www.ebi.ac.uk/metabolights/>

**Examples**

```
get_files_list_per_assay('MTBLS346')
```

---

```
get_metabolights_study
```

*Download MetaboLights study files.*

---

**Description**

Download data and metadata files for each assay from the specified MetaboLights database study.

**Usage**

```
get_metabolights_study(studyID, directory)
```



**Arguments**

studyID	ID of the metabolights study to download. For example, 'MTBLS100'
directory	Directory where to download the data.

**Note**

Study's files are stored by assay. Data files from assay 1 of the study will be stored in folder '1'.

Be aware that the study's files may not be structured in the right way to be readily imported with a specmine read function.

Specmine takes into consideration that the names of the data files/folders correspond to the names of the samples. In some studies, data file names do not correspond to the samples' names in the metadata. To overcome this, we create a file called 'samples\_files.csv' matching the sample name to the respective data file/zipped folder.

In some cases, one downloaded zipped data folder may contain more than one sample / replicates, but metabolights information only associates the overall folder as one sample. So manual naming of the folder samples and further changing the metadata file (metadata.csv) may be necessary.

Also, some data formats of some metabolights studies are not yet readable by specmine.

The metadata file(s) are csv file(s) with the metadata information on each sample. There is one metadata file per assay. Metadata file from assay 1 will be named 'metadata1.csv'.

**References**

MetaboLights database: <https://www.ebi.ac.uk/metabolights/>

---

get\_metabolights\_study\_files\_assay

*Download data files from an assay of MetaboLights study*

---

**Description**

Downloads the data files from the assay specified in 'assay' of the MetaboLights study ('studyID')

**Usage**

```
get_metabolights_study_files_assay(studyID, assay, directory)
```

**Arguments**

studyID	ID of the metabolights study to download.
assay	Number of the assay.
directory	Directory where to download the data.

## Details

This function should be used together with `get_metabolights_study_metadata_assay`. See example below.

Be aware that the study's files may not be structured in the right way to be readily imported with a specmine read function.

Specmine takes into consideration that the names of the data files/folders correspond to the names of the samples. In some studies, data file names do not correspond to the samples' names in the metadata. To overcome this, we create a file called 'samples\_files.csv' matching the sample name to the respective data file/zipped folder.

In some cases, one downloaded zipped data folder may contain more than one sample / replicates, but metabolights information only associates the overall folder as one sample. So manual naming of the folder samples and further changing the metadata file (metadata.csv) may be necessary.

Also, some data formats of some metabolights studies are not yet readable by specmine.

## References

MetaboLights database: <https://www.ebi.ac.uk/metabolights/>

## Examples

```
get_metabolights_study_files_assay('MTBLS346', 1, tempdir())
get_metabolights_study_metadata_assay('MTBLS346', 1, tempdir())
```

---

```
get_metabolights_study_metadata_assay
```

*Download metadata file from an assay of MetaboLights study*

---

## Description

Downloads the metadata file from the assay specified in 'assay' of the MetaboLights study ('studyID').

## Usage

```
get_metabolights_study_metadata_assay(studyID, assay, directory)
```

## Arguments

studyID	ID of the metabolights study to download.
assay	Number of the assay.
directory	Directory where to download the data.

### Details

This function should be used together with `get_metabolights_study_files_assay`. See example below.

The metadata file is a csv file with the metadata information on each sample of the study's assay.

### References

MetaboLights database: <https://www.ebi.ac.uk/metabolights/>

### Examples

```
get_metabolights_study_files_assay('MTBLS346', 1, tempdir())
get_metabolights_study_metadata_assay('MTBLS346', 1, tempdir())
```

---

`get_metabolights_study_samples_files`

*Get list of files from an assay of the MetaboLights study and saves it in a csv file.*

---

### Description

Get list of files from an assay of the MetaboLights study and saves it in a csv file.

### Usage

```
get_metabolights_study_samples_files(studyID, assay, directory)
```

### Arguments

<code>studyID</code>	ID of the metabolights study
<code>assay</code>	Number of the assay
<code>directory</code>	Directory path where the file will be saved.

### References

MetaboLights database: <https://www.ebi.ac.uk/metabolights/>

### Examples

```
get_metabolights_study_samples_files('MTBLS346',1, tempdir())
```

---

get_MetabolitePath	Returns an object of KEGGPathway of the pathway specified in pathcode.
--------------------	--

---

**Description**

Returns an object of KEGGPathway of the pathway specified in pathcode (e.g. "hsa00010").

**Usage**

```
get_MetabolitePath(pathcode)
```

**Arguments**

pathcode	Pathway code of the path wanted.
----------	----------------------------------

**Value**

KEGGPathway object.

---

get_metabPaths_org	Get the metabolic pathways present in given organism.
--------------------	---

---

**Description**

Get vector with paths numbers that occur in the given organism, named with the full paths names.

**Usage**

```
get_metabPaths_org(org_code)
```

**Arguments**

org_code	Organism code. The correct code for an organism can be consulted using function <a href="#">get_OrganismsCodes</a> .
----------	--

---

get_metadata	<i>Get metadata</i>
--------------	---------------------

---

**Description**

Get the metadata from the dataset

**Usage**

```
get_metadata(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

returns a data frame with the metadata.

**Examples**

```
## Example of getting the metadata  
data(cachexia)  
cachexia.mt = get_metadata(cachexia)
```

---

get_metadata_value	<i>Get metadata value</i>
--------------------	---------------------------

---

**Description**

Get the metadata value

**Usage**

```
get_metadata_value(dataset, variable, sample)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
variable           index or name of the metadata variable.  
sample             index or name of the sample.

**Value**

Return the corresponding metadata value of the sample.

**Examples**

```
## Example of getting a metadata value
data(cachexia)
metadata.value = get_metadata_value(cachexia, "Muscle.loss", "PIF_178")
```

---

get_metadata_var	<i>Get metadata variable</i>
------------------	------------------------------

---

**Description**

Get the values of a metadata variable from the dataset.

**Usage**

```
get_metadata_var(dataset, var)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
var	index or name of the metadata variable.

**Value**

Returns a vector with the values of the metadata variable.

**Examples**

```
## Example of getting a metadata variable
data(cachexia)
metadata.variable = get_metadata_var(cachexia, "Muscle.loss")
```

---

get_OrganismsCodes	<i>Get all organisms in KEGG.</i>
--------------------	-----------------------------------

---

**Description**

Get code, t number, full name and phylogeny of all organisms in KEGG.

**Usage**

```
get_OrganismsCodes()
```

**Value**

Data frame with t number, organism code, full name and phylogeny for each kegg organism.

## Examples

```
get_OrganismsCodes()
```

---

```
get_paths_with_cpds_org
```

*Get only the paths of the organism that contain one or more of the given compounds.*

---

## Description

Gives only the metabolic paths of the mentioned organism that contain one or more of the given compounds.

## Usage

```
get_paths_with_cpds_org(organism_code, compounds, full.result=TRUE)
```

## Arguments

organism_code	Organism code. The correct code for an organism can be consulted using function <a href="#">get_OrganismsCodes</a> .
compounds	Named vector with kegg codes of compounds and respective names. This vector can be obtained by using the function <a href="#">get_cpd_names</a> or the function <a href="#">convert_hmdb_to_kegg</a> .
full.result	If the full result is to be given. Defaults to TRUE.

## Value

Data frame.

If full result is chosen, the data frame contains information on the pathways of the organism that contains one or more of the given compounds and, for each pathway, the kegg codes (and their names) of the compounds given that are present in that path. The ratio between the number of compounds given compounds present in each pathway and the total number of compounds in each pathway is also given, full result or not.

If full result is not wanted, only the pathways will be given.

## Examples

```
#Get human metabolic paths that have one or more of the three following compounds
keggs=get_cpd_names(c("cpd:C00033", "cpd:C00147", "gl:G13099"))
paths_org_cpds=get_paths_with_cpds_org("hsa", keggs)
paths_org_cpds
```

---

get_peak_values	<i>Get peak values</i>
-----------------	------------------------

---

**Description**

Gets the peak values from a data frame of samples' peaks.

**Usage**

```
get_peak_values(samples.df, peak.val)
```

**Arguments**

samples.df	data frame with the samples' peaks.
peak.val	peak name.

**Value**

Returns a vector with the peak values.

**Examples**

```
## Example of getting the peak values
data(propolis)
peak.values = get_peak_values(propolis$data, 2.11)
```

---

get_samples_names_dx	<i>Get sample's names from DX files</i>
----------------------	---

---

**Description**

Function to get the names of the DX files from a folder.

**Usage**

```
get_samples_names_dx(foldername)
```

**Arguments**

foldername	string with the path of the data folder.
------------	--

**Value**

Returns a vector with the sample's names.



---

get\_samples\_names\_spc *Get sample's names from SPC files*

---

**Description**

Function to get the names of the SPC files from a folder.

**Usage**

```
get_samples_names_spc(foldername)
```

**Arguments**

foldername      string with the path of the data folder.

**Value**

Returns a vector with the sample's names.

---

get\_sample\_names      *Get sample names*

---

**Description**

Get the sample names from the dataset.

**Usage**

```
get_sample_names(dataset)
```

**Arguments**

dataset      list representing the dataset from a metabolomics experiment.

**Value**

Returns a vector with the sample names.

**Examples**

```
## Example of getting the sample names
data(cachexia)
sample.names = get_sample_names(cachexia)
```

---

<code>get_type</code>	<i>Get type of data</i>
-----------------------	-------------------------

---

**Description**

Get the type of the data from the dataset

**Usage**

```
get_type(dataset)
```

**Arguments**

`dataset` list representing the dataset from a metabolomics experiment.

**Value**

Returns a string with the type of the data.

**Examples**

```
## Example of getting the type of the data
data(cachexia)
type = get_type(cachexia)
```

---

<code>get_value_label</code>	<i>Get value label</i>
------------------------------	------------------------

---

**Description**

Get the value label from the dataset

**Usage**

```
get_value_label(dataset)
```

**Arguments**

`dataset` list representing the dataset from a metabolomics experiment.

**Value**

Returns a string with the value label.

**Examples**

```
## Example of getting the value label
data(cassavaPPD)
value.label = get_value_label(propolis)
```

---

get_x_label	<i>Get x-axis label</i>
-------------	-------------------------

---

**Description**

Get the x-axis label from the dataset.

**Usage**

```
get_x_label(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns a string with the x-axis label.

**Examples**

```
## Example of getting the x-axis label
data(cassavaPPD)
x.label = get_x_label(propolis)
```

---

get_x_values_as_num	<i>Get x-axis values as numbers</i>
---------------------	-------------------------------------

---

**Description**

Get the x-axis values from the dataset as numbers.

**Usage**

```
get_x_values_as_num(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns a numeric vector with the x-axis values, if the variable labels are not all numeric then an error message is shown.

**Examples**

```
## Example of getting the x-axis values as numbers
xvalues.numeric = get_x_values_as_num(propolis)
```

---

get\_x\_values\_as\_text    *Get x-axis values as text*

---

**Description**

Get the x-axis values from the dataset as text.

**Usage**

```
get_x_values_as_text(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns a character vector with the x-axis values.

**Examples**

```
## Example of getting the x-axis values as text
xvalues.text = get_x_values_as_text(propolis)
```

---

group\_peaks            *Group peaks*

---

**Description**

Group peaks with peak alignment.

**Usage**

```
group_peaks(sample.list, type, method = "own", metadata = NULL,
samp.classes = 1, description = "", label.x = NULL,
label.values = NULL, step = 0.03)
```

**Arguments**

sample.list        list containing the sample's data.  
type                type of the data.  
method             method of peak alignment. Can be "own" or "metaboanalyst", which the later is for using the peak alignment used in MetaboAnalyst software.  
metadata           data frame containing the metadata.  
samp.classes       the metadata's variable to be used in the MetaboAnalyst method.

description	short description of the data.
label.x	the label for the x values.
label.values	the label for the y values.
step	step value for the peak alignment process.

### Value

Returns a dataset with the peaks of the data aligned.

### Examples

```
## Example of grouping peaks (computationally heavy)
peaks.ds = group_peaks(propolisSampleList, "nmr-peaks", method = "own",
  metadata = NULL, description = "short description",
  label.x = "ppm", label.values = "intensity", step = 0.03)
```

---

heatmap\_correlations *Correlations heatmap*

---

### Description

Plots a heatmap with the correlations.

### Usage

```
heatmap_correlations(correlations, col = NULL, ...)
```

### Arguments

correlations	correlation matrix
col	colors to be used on heatmap.
...	extra parameters to visual purposes.

### Examples

```
## Example of correlations heatmap
data(cachexia)
correlations = correlations_dataset(cachexia)
heatmap_correlations(correlations)
```

---

hierarchical\_clustering

*Perform hierarchical clustering analysis*

---

**Description**

Perform hierarchical clustering analysis on the dataset.

**Usage**

```
hierarchical_clustering(dataset, distance = "euclidean",  
clustMethod = "complete", hc.type = "samples")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
distance	the distance measure to be used to compute the distances between the rows of a data matrix. Possible types are "euclidean", "manhattan", "pearson" or "spearman".
clustMethod	the agglomeration method to be used. Possible values are "ward", "single", "complete", "average", "mcquitty", "median" or "centroid".
hc.type	a string indicating if hierarchical cluster analysis will be performed on samples ("samples") or on variables ("variables")

**Value**

An object of class hclust with the clustering results.

**Examples**

```
## Example of hierarchical clustering  
data(cachexia)  
hc.result = hierarchical_clustering(cachexia,  
distance = "euclidean", clustMethod = "complete",  
hc.type = "samples")
```

---

impute\_nas\_knn

*Impute missing values with KNN*

---

**Description**

Impute missing values with KNN

**Usage**

```
impute_nas_knn(dataset, k = 10, ...)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
k                    the number of nearest neighbors.  
...                 additional values to impute.knn function.

**Value**

Returns the dataset with no missing values.

**Examples**

```
## Example of NA imputation with knn  
data(propolis)  
dataset = impute_nas_knn(propolis, k=10)
```

---

impute\_nas\_linapprox    *Impute missing values with linear approximation*

---

**Description**

Impute missing values with linear approximation.

**Usage**

```
impute_nas_linapprox(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the dataset with no missing values.

**Examples**

```
## Example of NA imputation with linear approximation  
data(propolis)  
dataset = impute_nas_linapprox(propolis)
```

---

impute_nas_mean	<i>Impute missing values with mean</i>
-----------------	--

---

**Description**

Impute missing values with mean

**Usage**

```
impute_nas_mean(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the dataset with no missing values.

**Examples**

```
## Example of NA imputation with mean
data(propolis)
propolis = impute_nas_mean(propolis)
```

---

impute_nas_median	<i>Impute missing values with median</i>
-------------------	--

---

**Description**

Impute missing values with median

**Usage**

```
impute_nas_median(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the dataset with no missing values.

**Examples**

```
## Example of NA imputation with median
data(propolis)
propolis = impute_nas_median(propolis)
```



---

impute_nas_value	<i>Impute missing values with value replacement</i>
------------------	---

---

**Description**

Impute missing values with value replacement.

**Usage**

```
impute_nas_value(dataset, value)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
value	value to replace the missing values.

**Value**

Returns the dataset with no missing values.

**Examples**

```
## Example of NA imputation with value replacing  
data(propolis)  
propolis = impute_nas_value(propolis, 0.0005)
```

---

indexes_to_xvalue_interval	<i>Get the x-values of a vector of indexes</i>
----------------------------	--

---

**Description**

Returns x-values corresponding to a vector of indexes (only to numerical values - spectra)

**Usage**

```
indexes_to_xvalue_interval(dataset, indexes)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
indexes	numeric vector containing the indexes.

**Value**

Returns a numeric vector with the interval of x-values from the indexes vector

**Examples**

```
## Example of getting the interval of x-values from indexes
xvalue.interval = indexes_to_xvalue_interval(propolis, c(10,50))
```

---

is_spectra	<i>Check type of data</i>
------------	---------------------------

---

**Description**

Check if the dataset is from spectral data where x.values are numeric.

**Usage**

```
is_spectra(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns a boolean indicating if the dataset is from spectral data or not.

**Examples**

```
## Example of checking if the dataset is from spectral data
is_spectra(propolis)
```

---

kmeans_clustering	<i>Perform k-means clustering analysis</i>
-------------------	--

---

**Description**

Perform k-means clustering analysis on the dataset.

**Usage**

```
kmeans_clustering(dataset, num.clusters, type = "samples")
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
num.clusters      the number of clusters.  
type                a string indicating if k-means will be performed on samples ("samples") or on variables ("variables")

**Value**

An object of class kmeans with the clustering results.

**Examples**

```
## Example of kmeans clustering
kmeans.result = kmeans_clustering(cachexia,
num.clusters = 4, type = "samples")
```

---

kmeans_plot	<i>Plot kmeans clusters</i>
-------------	-----------------------------

---

**Description**

Plot for each formed cluster, in grey the values of all samples of that cluster and in blue the median of that samples.

**Usage**

```
kmeans_plot(dataset, kmeans.result)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
kmeans.result    object of class kmeans with the clustering results.

**Examples**

```
## Example of kmeans plot - dataset filtered for performance purposes
kmeans.result = kmeans_clustering(cachexia,
num.clusters = 4, type = "samples")
kmeans_plot(cachexia, kmeans.result)
```

---

kmeans_result_df	<i>Show cluster's members</i>
------------------	-------------------------------

---

**Description**

Show for each cluster from kmeans analysis the sample names belonging to them.

**Usage**

```
kmeans_result_df(kmeans.result)
```

**Arguments**

kmeans.result    object of class kmeans with the clustering results.

**Value**

Data frame with the clusters and the samples' names that belong to each one.

**Examples**

```
## Example of showing kmeans cluster's members
kmeans.result = kmeans_clustering(cachexia,
num.clusters = 4, type = "samples")
kmeans_result_df(kmeans.result)
```

---

kruskalTest\_dataset     *Kruskal-Wallis tests on dataset*

---

**Description**

Run Kruskal-Wallis Tests for each row of the data from the dataset.

**Usage**

```
kruskalTest_dataset(dataset, metadata.var, threshold = NULL,
write.file = FALSE, file.out = "kruskal.csv")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
metadata.var	metadata variable to use in the t-tests.
threshold	threshold value of the p-value.
write.file	boolean value to write or not a file with the results.
file.out	name of the file.

**Value**

Table with the results of the Kruskal-Wallis tests, with p-value,  $-\log_{10}(\text{p-value})$  and false discovery rate (fdr).

**Examples**

```
## Example of ks-Tests on dataset
data(cachexia)
kruskaltests.result = kruskalTest_dataset(cachexia, "Muscle.loss",
write.file = FALSE)
```

---

ksTest_dataset	<i>Kolmogorov-Smirnov tests on dataset</i>
----------------	--

---

**Description**

Run Kolmogorov-Smirnov Tests for each row of the data from the dataset.

**Usage**

```
ksTest_dataset(dataset, metadata.var, threshold = NULL,  
write.file = FALSE, file.out = "ks.csv")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
metadata.var	metadata variable to use in the t-tests.
threshold	threshold value of the p-value.
write.file	boolean value to write or not a file with the results.
file.out	name of the file.

**Value**

Table with the results of the Kolmogorov-Smirnov tests, with p-value,  $-\log_{10}(\text{p-value})$  and false discovery rate (fdr).

**Examples**

```
## Example of ks-Tests on dataset  
data(cachexia)  
kstests.result = ksTest_dataset(cachexia, "Muscle.loss",  
write.file = FALSE)
```

---

linregression_onevar	<i>Linear regression on one variable</i>
----------------------	--

---

**Description**

Performs linear regression on one variable of the dataset.

**Usage**

```
linregression_onevar(dataset, x.val, metadata.vars, combination)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
x.val	the x-value to be tested.
metadata.vars	metadata variables to use in linear regression. For example, c('variable1','variable2').
combination	a formula specifying the model. For example, 'variable1+variable2'.

**Value**

Returns a summary of the result from the lm function from stats package.

---

linreg_all_vars	<i>Linear Regression</i>
-----------------	--------------------------

---

**Description**

Performs linear regression analysis over the dataset with the selected metadata's variables.

**Usage**

```
linreg_all_vars(dataset, metadata.vars, combination)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
metadata.vars	metadata variables to use in linear regression. For example, c('variable1','variable2').
combination	a formula specifying the model. For example, 'variable1+variable2'.

**Value**

Returns a list where each element is the linear regression result of a variable on the dataset.

---

linreg_coef_table	<i>Linear regression coefficient table</i>
-------------------	--

---

**Description**

Gets a data.frame with the coefficient values.

**Usage**

```
linreg_coef_table(linreg.results, write.file = FALSE,
file.out = "linreg-coefs.csv")
```

**Arguments**

linreg.results Linear regression results from linreg.all.vars function.  
write.file boolean value to indicate if a file should be written with the results.  
file.out name of the file.

**Value**

Returns a data.frame with the coefficient values.

---

linreg\_pvalue\_table *Linear regression p-values table*

---

**Description**

Gets the p-values table from the linear regression analysis.

**Usage**

```
linreg_pvalue_table(linreg.results, write.file = FALSE,  
file.out = "linreg-pvalues.csv")
```

**Arguments**

linreg.results Linear regression results from linreg.all.vars function.  
write.file boolean value to indicate if a file should be written with the results.  
file.out name of the file.

**Value**

Returns a data.frame with the p-values.

---

linreg\_rsquared *Linear regression r-squared*

---

**Description**

Gets the linear regression r-squared values.

**Usage**

```
linreg_rsquared(linreg.results, write.file = FALSE,  
file.out = "linreg-rsquared.csv")
```

**Arguments**

linreg.results Linear regression results from linreg.all.vars function.  
write.file boolean value to indicate if a file should be written with the results.  
file.out name of the file.

**Value**

Returns a data.frame with the r-squared values.

---

log_transform	<i>Logarithmic transformation.</i>
---------------	------------------------------------

---

**Description**

Performs logarithmic transformation on the data matrix.

**Usage**

```
log_transform(datamat)
```

**Arguments**

datamat data matrix.

**Value**

Returns the data matrix with the logarithmic transformation applied.

**Examples**

```
## Example of logarithmic transformation  
propolis_proc = missingvalues_imputation(propolis)  
datamat.log = log_transform(propolis_proc$data)
```



---

low_level_fusion	<i>Low level fusion</i>
------------------	-------------------------

---

**Description**

Low level fusion method for integrate different datasets (only samples with the same name on all datasets will be merged)

**Usage**

```
low_level_fusion(datasets)
```

**Arguments**

datasets	list containing the datasets to be fused (each dataset is a list from the pre defined format for the dataset).
----------	--

**Value**

Return a single dataset with all the datasets merged.

---

MAIT_identify_metabolites	<i>MAIT metabolite identification</i>
---------------------------	---------------------------------------

---

**Description**

Performs metabolite identification using MAIT.

**Usage**

```
MAIT_identify_metabolites(dataset, metadata.variable,  
xSet = NULL, data.folder = NULL, features = NULL,  
mass.tolerance = 0.5)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
metadata.variable	metadata's variable.
xSet	xcmsSet object that can be passed. Stored in dataset\$xSet.
data.folder	string indicating the data folder.
features	features that can be used to help to identify the metabolites.
mass.tolerance	mass tolerance.

**Details**

After running the MAIT\_identify\_metabolites function, the results table can be accessed by:

```
mait.metab.table = mait.metabolites@FeatureInfo@metaboliteTable
```

where 'mait.metabolites' is the result obtained from running MAIT\_identify\_metabolites.

**Value**

Returns an object resulted from identifyMetabolites function from MAIT package.

**References**

<http://www.bioconductor.org/packages/release/bioc/html/MAIT.html>

---

mean_centering	<i>Mean centering</i>
----------------	-----------------------

---

**Description**

Performs mean centering on the dataset.

**Usage**

```
mean_centering(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the dataset with mean centering applied.

---

merge_datasets	<i>Merge two datasets</i>
----------------	---------------------------

---

**Description**

Merges two datasets with the same variables and metadata's variables.

**Usage**

```
merge_datasets(dataset1, dataset2)
```

**Arguments**

- dataset1      list representing the first dataset from a metabolomics experiment.  
dataset2      list representing the second dataset from a metabolomics experiment.

**Value**

Returns one dataset with the data from the two datasets merged.

---

merge\_data\_metadata      *Merge data and metadata*

---

**Description**

Merges the data and metadata from the dataset into a single data.frame.

**Usage**

```
merge_data_metadata(dataset, samples = NULL,  
metadata.vars = NULL, x.values = NULL, by.index = FALSE)
```

**Arguments**

- dataset      list representing the dataset from a metabolomics experiment.  
samples      vector with indexes or names of the samples to select  
metadata.vars      metadata's variables.  
x.values      vector with the desired x-values to get from the dataset.  
by.index      if TRUE, the values of the x.values argument are indexes.

**Value**

Returns a data.frame with the data and metadata from the dataset merged.

**Examples**

```
## Example of merging data and metadata  
data(cachexia)  
dt.merged = merge_data_metadata(cachexia)
```

---

```
metabolights_studies_list
```

*List the study IDs available in the MetaboLights database.*

---

**Description**

Gives the IDs of the studies available in the MetaboLights database.

**Usage**

```
metabolights_studies_list()
```

**Value**

Vector with the different study IDs available in the MetaboLights database.

**References**

MetaboLights database: <https://www.ebi.ac.uk/metabolights/>

**Examples**

```
## metabolights_studies_list()
```

---

```
metadata_as_variables Metadata as variables
```

---

**Description**

Use one or more metadata variables as variables.

**Usage**

```
metadata_as_variables(dataset, metadata.vars, by.index = FALSE)
```

**Arguments**

<code>dataset</code>	list representing the dataset from a metabolomics experiment.
<code>metadata.vars</code>	name or index of the metadata variables that are going to be used as data variables.
<code>by.index</code>	boolean value indicating if the metadata variables are indexes or names

**Value**

Returns the dataset with the metadata variables removed from the metadata and added on the data matrix.

## Examples

```
## Example of using a metadata variable as data variable
data(propolis)
propolis = metadata_as_variables(propolis, "seasons", by.index = FALSE)
```

---

```
missingvalues_imputation
      Missing values imputation
```

---

## Description

Treats the missing values of a dataset according to a specific method.

## Usage

```
missingvalues_imputation(dataset, method = "value",
value = 5e-04, k = 5)
```

## Arguments

dataset	list representing the dataset from a metabolomics experiment.
method	imputation method. It can be: <ul style="list-style-type: none"><li>• <b>"value"</b> - replaces the missing values with a specific value</li><li>• <b>"mean"</b> - replaces the missing values with the mean of the variables' values</li><li>• <b>"median"</b> - replaces the missing values with the median of the variables' values</li><li>• <b>"knn"</b> - replaces the missing values with k nearest neighbor averaging</li><li>• <b>"linapprox"</b> - replaces the missing values with linear approximation</li></ul>
value	the value to replace the missing values if the method is "value".
k	the number of neighbors if the method is "knn".

## Value

Returns the dataset with no missing values.

## Examples

```
## Example of impute missing values
data(propolis)
dataset = missingvalues_imputation(propolis, method = "value",
value = 0.0005)
```

msc\_correction      *Multiplicative scatter correction*

---

**Description**

Perform multiplicative scatter correction on the spectra.

**Usage**

```
msc_correction(dataset)
```

**Arguments**

dataset      list representing the dataset from a metabolomics experiment.

**Value**

Return the dataset with the multiplicative scatter correction employed on the data.

---

multiClassSummary      *Multi Class Summary*

---

**Description**

Summary function for caret to compute AUC.

**Usage**

```
multiClassSummary(data, lev = NULL, model = NULL)
```

**Arguments**

data      data parameter.  
lev      lev parameter.  
model      model parameter.

**References**

[www.r-bloggers.com/error-metrics-for-multi-class-problems-in-r-beyond-accuracy-and-kappa/](http://www.r-bloggers.com/error-metrics-for-multi-class-problems-in-r-beyond-accuracy-and-kappa/)

---

`multifactor_aov_all_vars`*Multifactor ANOVA*

---

**Description**

Perform multi-factor ANOVA on all variables with the selected metadata variables.

**Usage**

```
multifactor_aov_all_vars(dataset, metadata.vars, combination)
```

**Arguments**

`dataset` list representing the dataset from a metabolomics experiment.  
`metadata.vars` metadata variables to use in ANOVA.  
`combination` a formula specifying the model.

**Value**

List where each element is the multifactor anova result of a variable on the dataset.

**Examples**

```
## Example of multifactor ANOVA on all variables
data(propolis)
propolis = missingvalues_imputation(propolis, "value", value = 0.00005)
m.aov.results = multifactor_aov_all_vars(propolis,
c("seasons","agroregions"), "seasons*agroregions")
```

---

`multifactor_aov_pvalues_table`*Multifactor ANOVA p-values table*

---

**Description**

Gets the p-values table from the multifactor ANOVA results.

**Usage**

```
multifactor_aov_pvalues_table(multifactor.aov.results,
write.file = FALSE, file.out = "multi-anova-pvalues.csv")
```





## Examples

```
## Example of multifactor ANOVA variability explained table
data(propolis)
propolis = missingvalues_imputation(propolis, "value", value = 0.00005)
m.aov.results = multifactor_aov_all_vars(propolis,
c("seasons", "agroregions"), "seasons*agroregions")
m.aov.varepx = multifactor_aov_varexp_table(m.aov.results)
```

---

multiplot

*Multiplot*

---

## Description

Multiplot from ggplot2 - function taken from (see references).

## Usage

```
multiplot(plots, plotlist = NULL, file, cols = 1, layout = NULL)
```

## Arguments

plots	list with the plots to display.
plotlist	plot list.
file	file.
cols	number of columns.
layout	layout of the plot.

## References

[http://www.cookbook-r.com/Graphs/Multiple\\_graphs\\_on\\_one\\_page\\_](http://www.cookbook-r.com/Graphs/Multiple_graphs_on_one_page_)

## Examples

```
## Example of multiplot
data(cachexia)
pca.result = pca_analysis_dataset(cachexia)
plot1 = pca_scoresplot2D(cachexia, pca.result, "Muscle.loss",
ellipses = TRUE)
plot2 = pca_scoresplot2D(cachexia, pca.result, "Muscle.loss",
ellipses = FALSE, labels = TRUE)
plts = list(plot1, plot2)
multiplot(plts, cols = 2)
```

---

nmr\_identification      *NMR metabolite identification*

---

### Description

This function performs metabolite identification on a dataset of nmr peaks.

### Usage

```
nmr_identification(dataset, ppm.tol, frequency_scores, solvent_scores, organism_scores,
                  method='Match_uniq', per.sample=FALSE, tresh_zero=0, alpha=10e-4)
```

### Arguments

dataset	List representing the dataset from an nmr peaks metabolomics experiment.
ppm.tol	ppm tolerance when matching reference peaks to the dataset peaks.
per.sample	Logical indicating whether identification should be done for each sample (for each sample, only the peaks present in that sample will be used in the identification) or for all samples together (all peaks will be used together in the identification).
tresh_zero	Intensity value above which peaks are considered present. Only necessary if per.sample=TRUE
method	Identification method to use. There are three: 'Match_uniq' (default), 'Hyper' or 'Hyper_uniq'. For more information, see details below.
alpha	A metabolite with a corrected p.value above alpha will not be considered identified. Only necessary if method is either 'Hyper' or 'Hyper_uniq'.
frequency_scores	Each frequency in the library should be given a score from 0 to 1, according to the frequency in which the samples were acquired. A score of 1 should be given to the frequency under which the samples were acquired. Argument should be in the form of a list, like: list('400'=0, '500'=1, '600'=0, '700'=0). For more information, see details below.
solvent_scores	Each solvent in the library should be given a score from 0 to 1, according to the solvent with which the samples were acquired. A score of 1 should be given to the solvent with which the samples were acquired. Argument should be in the form of a list, like: list(CD3OD=0, D2O=1, Water=1, CDC13=0, 'Acetone-d6'=0, 'Acetone'=0, "DMSO-d6"=0, "100%_DMSO"=0, "5%_DMSO"=0, C=0, C6D6=0, CD3CN=0, C2D2C14=0, CD2C12=0, CDC3OD=0, Ethanol=0). For more information, see details below.
organism_scores	This gives the opportunity to score each reference according to the presence or not of the respective metabolite in the organism(s) or group(s) of organisms under study, so that metabolites not present in an organism/group of interest will have a lower score and thus be less likely to be identified. The presence of a metabolite in an organism/group is evaluated using the information

in the KEGG database. Thus, all organism/group codes should be present in KEGG. Argument should be in the form of a list, like: list('hsa'=1, 'other'=0, 'not\_in\_kegg'=0). For more information, see details below.

## Details

There are three methods implemented to perform metabolite identification. The default one is Matched Ratio with uniqueness score ('Match\_uniq'):

- Hypergeometric Test: it calculates the probability of a group of  $k$  peaks matching to a certain reference spectrum not being caused by chance. A  $p$ -value over  $\alpha$  denotes that the metabolite corresponding to the reference spectrum in question is not present in the sample. After all reference metabolites are matched to the samples, the  $p$ -values are adjusted for multiple testing using the False Discovery Rate (FDR) method. For those with  $p$ .value under  $\alpha$ , the score is transformed into a scale of 0 to 1, by applying the following:  $1-(p.value/\alpha)$ .

- Hypergeometric Test with Uniqueness score ('Hyper\_uniq'): the final score of a reference spectrum is obtained by calculating the average between the hypergeometric test score and the uniqueness score, if the hypergeometric test score is not null. The uniqueness score of a reference is the average of the uniqueness rate of all peaks in that reference. The uniqueness rate of a peak is calculated by dividing 1 by the number of reference spectra that peak is in, based on the reference library used.

- Matched ratio scores with uniqueness score ('Match\_uniq'): the final score of a reference spectrum is obtained by calculating the average between the matched ratio score and the uniqueness score, if the matched ratio score is not null. The matched ratio score gives the ratio of peaks from a reference that matched the sample. It is calculated by dividing the number of different peaks matched between the reference and the sample by the total number of different peaks in such reference.

After scoring a match between a reference and a sample, the reference is further scored regarding the conditions under which it was acquired. To do so, each frequency and solvent represented in the library must be given a score by the user. The user also has the possibility to score each reference according to the presence or not of the respective metabolite in the organism(s) or group(s) of organisms under study, so that metabolites not present in an organism/group of interest will have a lower score and thus be less likely to be identified. The presence of a metabolite in an organism/group is evaluated using the information in the KEGG database.

## Value

If `per.sample=FALSE`, it gives a list with two items: `results_table` and `more_results`. If `per.sample=TRUE`, it gives a list with as many items as the number of samples. Each sample contains a list with the two items `results_table` and `more_results`.

`results_table` is a data.frame with the information on the metabolites matched. Each row corresponds to a spectrum from the library that matched the dataset:

**SPCMNM** ID of the metabolite in our library.

**Name** Name of the metabolite.

**SPCMNS** ID of the respective spectrum in our library.

**Final\_Score** Final score.

**match\_score** Matching score.

**hypergeometric\_score** Hypergeometric score, if method='Hyper' or 'Hyper\_uniq'.

**ratio** matched ratio score, if method='Match\_uniq'

**uniqueness\_score** uniqueness\_score, if method = 'Hyper\_uniq' or 'Match\_uniq'

**score\_frequency** Score given to the frequency of that spectrum.

**score\_solvents** Score given to the solvent of that spectrum.

**score\_organisms** The organism score, according to the metabolite's presence in one of the organisms/groups given by the user in organism\_scores argument.

**n.peaks.matched** Number of peaks from the metabolite's spectrum that matched the sample.

**detailed\_results\_id** ID to access the more detailed results in the item more\_results.

more\_results is a list whose items are identified by an ID that is specified in the detailed\_results\_ID column of the results\_table. Each item is a list with the following information:

**matched\_peaks\_ref** Vector with the peaks from the reference spectrum that matched the sample. Reference peak in the ith position matched the sample peak in the ith position of the vector matched\_peaks\_samp.

**matched\_peaks\_samp** Vector with the peaks from the sample that matched the reference spectrum. Sample peak in the ith position matched the reference peak in the ith position of the vector matched\_peaks\_ref.

**reference\_peaks** Vector with all the peaks in the reference spectrum.

## Examples

```
propolis_mv=missingvalues_imputation(propolis)
freq_scores = list('400'=0, '500'=0, '600'=1, '700'=0)
solv_scores = list(CD30D=0, D20=.8, Water=.8, CDC13=0, 'Acetone-d6'=0, 'Acetone'=0, "DMSO-d6"=0,
                  '100%_DMSO'=0,
                  '5%_DMSO'=0, C=0, C6D6=0, CD3CN=0, C2D2C14=0, CD2C12=0,
                  CDC30D=1, Ethanol=0)
org_scores = list('Eudicots'=1, 'Monocots'=1, 'ame'=.9, 'other'=0, 'not_in_kegg'=0)
id_res=nmr_identification(propolis_mv, ppm.tol=0.03, freq_scores, solv_scores, org_scores)
```

---

normalize

*Normalize data*

---

## Description

Normalize the data from the dataset with a specific method.

## Usage

```
normalize(dataset, method, ref = NULL, constant = 1000)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
method	string specifying the normalization method. The possible values are: <ul style="list-style-type: none"><li>• <b>"sum"</b> normalization by sum.</li><li>• <b>"median"</b> normalization by median.</li><li>• <b>"ref.sample"</b> normalization by reference sample.</li><li>• <b>"ref.feature"</b> normalization by reference feature.</li></ul>
ref	the reference if method is "ref.sample" or "ref.feature".
constant	the constant value if method is "sum".

**Value**

Returns the dataset with the data normalized.

---

normalize_samples	<i>Normalize samples</i>
-------------------	--------------------------

---

**Description**

Normalize the data from a datamatrix with a specific method.

**Usage**

```
normalize_samples(datamat, method, ref = NULL, constant = 1000)
```

**Arguments**

datamat	data matrix.
method	string specifying the normalization method. The possible values are: <ul style="list-style-type: none"><li>• <b>"sum"</b> normalization by sum.</li><li>• <b>"median"</b> normalization by median.</li><li>• <b>"ref.sample"</b> normalization by reference sample.</li><li>• <b>"ref.feature"</b> normalization by reference feature.</li></ul>
ref	the reference if method is "ref.sample" or "ref.feature".
constant	the constant value if method is "sum".

**Value**

Returns the data matrix normalized.

---

num_samples	<i>Get number of samples</i>
-------------	------------------------------

---

**Description**

Get the number of samples from a dataset.

**Usage**

```
num_samples(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns an integer with the number of samples in the dataset.

**Examples**

```
## Example of getting the number of samples  
data(cachexia)  
number.of.samples = num_samples(cachexia)
```

---

num_x_values	<i>Get number of x values</i>
--------------	-------------------------------

---

**Description**

Get the number of x-axis values.

**Usage**

```
num_x_values(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns an integer with the number of x-axis values.

**Examples**

```
## Example of getting the number of x-axis values  
number.x.values = num_x_values(propolis)
```

---

offset_correction	<i>Offset correction</i>
-------------------	--------------------------

---

**Description**

Perform offset correction on the data.

**Usage**

```
offset_correction(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the dataset

---

pathway_analysis	<i>Creates the metabolic pathway wanted. If any of the given compounds is present in the pathway, it is coloured differently.</i>
------------------	---

---

**Description**

The pathway created contains the compounds, reactions and other paths that it connects to as nodes.

The compounds given in compounds are colored in blue, while the rest of the compounds are colored in grey.

The other paths that it may connect to are colored in orange.

Reversible reactions are colored in green and the irreversible ones in red.

**Usage**

```
pathway_analysis(compounds, pathway,  
                 nodeName="kegg", nodeTooltip=FALSE,  
                 map.zoom=FALSE, map.layout="preset", map.width=NULL, map.height=NULL)
```

**Arguments**

compounds	Vector of compounds of interest, in kegg codes.
pathway	KEGG code (e.g., "hsa00010") of the path wanted.
nodeNames	How the nodes should be named. If "kegg", nodes are named with kegg codes. If "names", nodes are named with the common names.
nodeTooltip	If a tooltip should appear when hovering a node.
map.zoom	If the map should have the zoom in and out option.
map.layout	Layout of the map, available values are the ones of cytoscape ("breadthfirst", "preset", "cose", ...)
map.width	Width of the map, in percentage
map.height	Width of the map, in px (e.g. "500px")

**Value**

Shows the pathway created.

**Examples**

```

cpds=c("HMDB0000042", "HMDB0000050", "HMDB0000452", "HMDB0001851", "HMDB0001964", "HMDB0000055",
"HMDB0000094", "HMDB0000092", "HMDB0000174", "HMDB0000639", "HMDB0000500", "HMDB0002092",
"HMDB0000176", "HMDB0000744", "HMDB0000223", "HMDB0000239", "HMDB0001545", "HMDB0001431",
"HMDB0001257", "HMDB0000254", "HMDB0000254", "HMDB0000258", "HMDB0002085", "HMDB0000975",
"HMDB0000300")
compounds=convert_hmdb_to_kegg(cpds)
##You can also convert CHEBI and SPCNMN (specmine code for metabolites) codes to KEGG
pathway_analysis(compounds, "mus00750", nodeNames="names", map.layout="preset")

```

---

pca\_analysis\_dataset *PCA analysis (classical)*

---

**Description**

Performs a classical PCA analysis over the dataset.

**Usage**

```

pca_analysis_dataset(dataset, scale = TRUE, center = TRUE,
write.file = FALSE, file.out = "pca", ...)

```



**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
scale	boolean value indicating if the variables are going to be scaled or not.
center	boolean value indicating if the variables are going to be centered or not.
write.file	boolean value that indicates if the results from PCA analysis are going to be written on a file.
file.out	name of the file that will store the results.
...	additional parameters to ggplot function.

**Value**

object of class 'prcomp' with the results from the PCA analysis.

**Examples**

```
## Example of performing a classical PCA analysis
data(cachexia)
pca.results = pca_analysis_dataset(cachexia)
```

---

pca\_biplot

*PCA biplot*

---

**Description**

Shows a PCA biplot.

**Usage**

```
pca_biplot(dataset, pca.result, cex = 0.8, legend.cex = 0.8,
x.colors = 1, inset = c(0, 0), legend.place = "topright", ...)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
cex	cex value.
legend.cex	cex value of the legend.
x.colors	colors of a metadata's variable.
inset	inset parameter of legend function.
legend.place	legend place.
...	additional parameters passed to biplot function.

**Examples**

```
## Example of a PCA biplot
data(cachexia)
pca.result = pca_analysis_dataset(cachexia)
pca_biplot(cachexia, pca.result, cex = 0.8)
```

---

pca\_biplot3D                    *3D PCA biplot (interactive)*

---

**Description**

Shows a interactive 3D PCA biplot.

**Usage**

```
pca_biplot3D(dataset, pca.result, column.class = NULL,
pcas = c(1, 2, 3))
```

**Arguments**

dataset                    list representing the dataset from a metabolomics experiment.  
pca.result                prcomp object with the PCA results.  
column.class              metadata's variable.  
pcas                        the three principal components.

**Examples**

```
### Example of a 3D PCA biplot
pca.result = pca_analysis_dataset(cachexia)
pca_biplot3D(cachexia, pca.result, "Muscle.loss", pcas = c(1,2,3))
```

---

pca\_importance                *PCA importance*

---

**Description**

Gets the importance from the PCs.

**Usage**

```
pca_importance(pca.res, pcs = 1:length(pca.res$sdev), sd = TRUE,
prop = TRUE, cumul = TRUE, min.cum = NULL)
```

**Arguments**

pca.res	prcomp object with the PCA results.
pcs	vector with the PCs to get.
sd	boolean value indicating if standard deviation will be returned or not.
prop	boolean value that indicates if the proportion of variance is returned or not.
cumul	boolean value that indicates if the cumulative variance is returned or not.
min.cum	allows to define minimum cumulative % of variance

**Value**

Returns the information about the importance of the PCs.

**Examples**

```
## Example of performing a classical PCA analysis
data(cachexia)
pca.result = pca_analysis_dataset(cachexia)
pca_importance(pca.result, pcs = 1:5)
```

---

pca\_kmeans\_plot2D      *2D PCA k-means plot*

---

**Description**

Groups the points with the clusters given by k-means in a 2D PCA scores plot.

**Usage**

```
pca_kmeans_plot2D(dataset, pca.result, num.clusters = 3,
pcas = c(1, 2), kmeans.result = NULL, labels = FALSE, bw=FALSE,
ellipses = FALSE, leg.pos = "right", xlim = NULL, ylim = NULL)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
num.clusters	number of clusters of k-means.
pcas	vector with the principal components to be plotted.
kmeans.result	result from k-means. If null k-means is performed in the function.
labels	boolean value indicating if the samples' labels will be shown.
ellipses	boolean value that indicates if an ellipse will be drawn on each group of the metadata's variable. Ellipses will not be drawn if bw=TRUE.
bw	if TRUE, it will be displayed a black and white plot. It defaults to FALSE.
leg.pos	legend position.
xlim	vector with two positions with the x-axis limits.
ylim	vector with two positions with the y-axis limits.

## Examples

```
## Example of a 2D PCA k-means plot
pca.result = pca_analysis_dataset(cachexia)
pca_kmeans_plot2D(cachexia, pca.result, num.clusters = 3, pcas = c(1,2))
```

---

pca\_kmeans\_plot3D      *3D PCA k-means plot (interactive)*

---

## Description

Groups the points with the clusters given by k-means in a interactive 3D PCA scores plot.

## Usage

```
pca_kmeans_plot3D(dataset, pca.result, num.clusters = 3,
pcas = c(1, 2, 3), kmeans.result = NULL, labels = FALSE,
size = 1, ...)
```

## Arguments

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
num.clusters	number of clusters of k-means.
pcas	vector with the principal components to be plotted.
kmeans.result	result from k-means. If null k-means is performed in the function.
labels	boolean value indicating if the samples' labels will be shown.
size	parameter of plot3d from rgl package.
...	additional parameters of plot3d function from rgl package.

## Examples

```
### Example of a 3D PCA k-means plot
pca.result = pca_analysis_dataset(cachexia)
pca_kmeans_plot3D(cachexia, pca.result, num.clusters = 3,
pcas = c(1,2,3))
```

---

pca\_pairs\_kmeans\_plot *PCA k-means pairs plot*

---

### Description

Groups the points with the clusters from k-means in a PCA pairs plot.

### Usage

```
pca_pairs_kmeans_plot(dataset, pca.result, num.clusters = 3,  
kmeans.result = NULL, pcas = c(1, 2, 3, 4, 5))
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
num.clusters	number of clusters of k-means.
kmeans.result	result from k-means. If null k-means is performed in the function.
pcas	vector with the principal components to be plotted.

### Examples

```
## Example of a PCA k-means pairs plot  
pca.result = pca_analysis_dataset(cachexia)  
kmeans.res = kmeans_clustering(cachexia, 3)  
pca_pairs_kmeans_plot(cachexia, pca.result, num.clusters = 3,  
kmeans.result = kmeans.res, pcas = c(1,2,3,4,5))
```

---

pca\_pairs\_plot *PCA pairs plot*

---

### Description

Shows a PCA pairs plot.

### Usage

```
pca_pairs_plot(dataset, pca.result, column.class = NULL,  
pcas = c(1, 2, 3, 4, 5), ...)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
column.class	metadata's variable.
pcas	the principal components to be shown.
...	additional parameters to ggpairs function from GGally package.

**Examples**

```
## Example of a PCA pairs plot
data(cachexia)
pca.result = pca_analysis_dataset(cachexia)
pca_pairs_plot(cachexia, pca.result, "Muscle.loss", pcas = c(1,2,3))
```

---

pca_plot_3d	<i>3D pca plot</i>
-------------	--------------------

---

**Description**

3D plot from 3 components

**Usage**

```
pca_plot_3d(dataset, model, var.class, pcas = 1:3, colors = NULL,
legend.place = "topright", ...)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
model	model with pca scores (pls model).
var.class	metadata column class.
pcas	the components to be plotted.
colors	colors of the groups.
legend.place	legend place.
...	additional parameters to legend function.

**Examples**

```
### Example of a 3d pca plot
data(cachexia)
train.result = train_models_performance(cachexia, "pls",
"Muscle.loss", "cv")
pca_plot_3d(cachexia, train.result$final.models$pls, "Muscle.loss")
```

---

pca_robust	<i>PCA analysis (robust)</i>
------------	------------------------------

---

## Description

Performs a robust PCA analysis.

## Usage

```
pca_robust(dataset, center = "median", scale = "mad", k = 10,  
write.file = FALSE, file.out = "robpca", ...)
```

## Arguments

dataset	list representing the dataset from a metabolomics experiment.
center	indicates how the data is to be centered. Can be a function or a vector with the center values of each column.
scale	indicates how the data is to be rescaled. Can be a function or a vector with the scale value of each column.
k	the desired number of components to compute
write.file	boolean value that indicates if the results from PCA analysis are going to be written on a file.
file.out	name of the file that will store the results.
...	additional parameters pass to or from other functions.

## Value

Returns an object of class 'princomp' with the PCA results.

## Examples

```
## Example of performing a robust PCA analysis  
data(cachexia)  
pca.results = pca_robust(cachexia, center = "mean", scale = "mad",  
k = 10)
```

---

pca\_scoresplot2D      *2D PCA scores plot*

---

### Description

Shows a 2D PCA scores plot of two principal componets.

### Usage

```
pca_scoresplot2D(dataset, pca.result, column.class,  
pcas = c(1, 2), labels = FALSE, ellipses = FALSE, bw=FALSE,  
palette = 2, leg.pos = "right", xlim = NULL, ylim = NULL)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
column.class	metadata's variable.
pcas	vector of two elements with the PCs that will be plotted.
labels	boolean value indicating if the sample's labels will be displayed.
ellipses	boolean value that indicates if an ellipse will be drawn on each group of the metadata's variable. Ellipses will not be drawn if bw=TRUE.
bw	if TRUE, it will be displayed a black and white plot. It defaults to FALSE.
palette	parameter of scale_colour_brewer from ggplot2.
leg.pos	position of the legend.
xlim	vector with two numeric values indicating the limits of the x axis.
ylim	vector with two numeric values indicating the limits of the y axis.

### Examples

```
## Example of a 2D PCA scores plot  
data(cachexia)  
pca.result = pca_analysis_dataset(cachexia)  
pca_scoresplot2D(cachexia, pca.result, "Muscle.loss", pcas = c(1,2),  
                  ellipses = TRUE)
```



---

pca\_scoresplot3D      *3D PCA scores plot*

---

### Description

Shows a 3D PCA scores plot of three principal componets.

### Usage

```
pca_scoresplot3D(dataset, pca.result, column.class,  
pcas = c(1, 2, 3))
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
column.class	metadata's variable.
pcas	vector with the principal components to be plotted.

### Examples

```
### Example of a 3D PCA scores plot  
data(cachexia)  
pca.result = pca_analysis_dataset(cachexia)  
pca_scoresplot3D(cachexia, pca.result, "Muscle.loss", pcas = c(1,2,3))
```

---

pca\_scoresplot3D\_rgl      *3D PCA scores plot (interactive)*

---

### Description

Shows a interactive 3D PCA scores plot of three principal components.

### Usage

```
pca_scoresplot3D_rgl(dataset, pca.result, column.class,  
pcas = c(1, 2, 3), size = 1, labels = FALSE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
pca.result	prcomp object with the PCA results.
column.class	metadata's variable.
pcas	vector with the principal components to be plotted.
size	parameter of plot3d from rgl package.
labels	boolean value indicating if the samples' labels will be shown.

**Examples**

```
### Example of a 3D PCA scores plot
pca.result = pca_analysis_dataset(cachexia)
pca_scoresplot3D_rgl(cachexia, pca.result, "Muscle.loss",
                    pcas = c(1,2,3), labels = TRUE)
```

---

pca\_screepplot

*PCA scree plot*


---

**Description**

PCA scree plot with the proportion and cumulative variance of the PCs.

**Usage**

```
pca_screepplot(pca.result, num.pcs = NULL, cex.legend = 0.8,
               leg.pos = "right", lab.text = c("individual percent",
               "cumulative percent"), fill.col = c("blue", "red"),
               ylab = "Percentage", xlab = "Principal components", ...)
```

**Arguments**

pca.result	prcomp object with the PCA results.
num.pcs	number of principal components.
cex.legend	cex value of legend.
leg.pos	legend position.
lab.text	legend's labels.
fill.col	color of the legend's boxes.
ylab	y-axis label.
xlab	x-axis label
...	additional parameters to matplot.

**Examples**

```
## Example of a scree plot
data(cachexia)
pca.result = pca_analysis_dataset(cachexia)
pca_screepplot(pca.result)
```

---

peaks_per_sample	<i>Peaks per sample</i>
------------------	-------------------------

---

**Description**

Counts number of peaks in a sample (given its index).

**Usage**

```
peaks_per_sample(sample.list, sample.index)
```

**Arguments**

sample.list     list of data frames with the samples' peaks.  
sample.index    sample index.

**Value**

Return a integer value with the number of peaks in the sample.

**Examples**

```
## Example of counting the peaks in a sample
data(propolisSampleList)
num.peaks.sample = peaks_per_sample(propolisSampleList, 4)
```

---

peaks_per_samples	<i>Peaks per samples</i>
-------------------	--------------------------

---

**Description**

Calculates the number of peaks on each sample.

**Usage**

```
peaks_per_samples(sample.list)
```

**Arguments**

sample.list     list of data frames with the samples' peaks.

**Value**

Returns a numeric vector with the number of peaks on each sample.

**Examples**

```
## Example of counting the peaks in each sample
data(propolisSampleList)
num.peaks.samples = peaks_per_samples(propolisSampleList)
```

---

plotvar\_twofactor      *Plot variable distribution on two factors*

---

**Description**

Plot variable distribution on two factors from the dataset.

**Usage**

```
plotvar_twofactor(dataset, variable, meta.var1, meta.var2,
  colour = "darkblue", title = "", xlabel = NULL, ylabel = NULL)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
variable	variable's name.
meta.var1	first metadata's variable.
meta.var2	second metadata's variable.
colour	colours of the bodies of the boxplots.
title	title of the plot.
xlabel	x-axis label.
ylabel	y-axis label.

**Value**

Returns the plot object from ggplot function.

**Examples**

```
## Example of plotting a variable's distribution with 2 factors
propolis_proc = missingvalues_imputation(propolis)
plotvar_twofactor(propolis_proc, "0.46", "seasons", "agroregions")
```

---

plot_anova	<i>Plot ANOVA results</i>
------------	---------------------------

---

### Description

Function for plotting the results from ANOVA.

### Usage

```
plot_anova(dataset, anova.results, anova.threshold = 0.01,  
reverse.x = FALSE)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
anova.results	ANOVA results.
anova.threshold	ANOVA threshold for the p-value.
reverse.x	boolean value to indicate if the x-axis is plotted in reverse.

### Examples

```
## Example of plotting the ANOVA results - first filter the  
## dataset to reduce computation time  
propolis_proc = missingvalues_imputation(propolis)  
propolis_proc = flat_pattern_filter(propolis_proc, "iqr", by.percent = TRUE,  
red.value = 75)  
anova.results = aov_all_vars(propolis_proc, "seasons", doTukey = FALSE)  
plot_anova(propolis_proc, anova.results)
```

---

plot_fold_change	<i>Plot fold change results</i>
------------------	---------------------------------

---

### Description

Function for plotting the results from fold change.

### Usage

```
plot_fold_change(dataset, fc.results, fc.threshold, plot.log = TRUE,  
var = FALSE, xlab = "")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
fc.results	fold change results.
fc.threshold	fold change threshold for the p-value.
plot.log	boolean value to determine if the fold change values are transformed logarithmically or not.
var	boolean value, if TRUE it uses the xlab argument to represent the xlabel of the plot.
xlab	string with the x axis description.

**Examples**

```
## Example of plotting the fold change results
data(cachexia)
fc.results = fold_change(cachexia, "Muscle.loss",
"control")
plot_fold_change(cachexia, fc.results, 2)
```

---

plot_kruskaltest	<i>Plot Kruskal-Wallis tests results</i>
------------------	--

---

**Description**

Function for plotting the results from Kruskal-Wallis tests.

**Usage**

```
plot_kruskaltest(dataset, kr.results, kr.threshold = 0.01)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
kr.results	Kruskal-Wallis tests results.
kr.threshold	Kruskal-Wallis test threshold for the p-value.

**Examples**

```
## Example of plotting the Kolmogorov-Smirnov tests results
data(cachexia)
kr.results = kruskalTest_dataset(cachexia, "Muscle.loss",
write.file = FALSE)
plot_kruskaltest(cachexia, kr.results, 0.05)
```

---

plot_kstest	<i>Plot Kolmogorov-Smirnov tests results</i>
-------------	--

---

### Description

Function for plotting the results from Kolmogorov-Smirnov tests.

### Usage

```
plot_kstest(dataset, ks.results, ks.threshold = 0.01)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
ks.results	Kolmogorov-Smirnov tests results.
ks.threshold	Kolmogorov-Smirnov test threshold for the p-value.

### Examples

```
## Example of plotting the Kolmogorov-Smirnov tests results
data(cachexia)
ks.results = ksTest_dataset(cachexia, "Muscle.loss",
write.file = FALSE)
plot_kstest(cachexia, ks.results, 0.05)
```

---

plot_peaks	<i>Plot the peaks of a MS or NMR dataset.</i>
------------	---

---

### Description

Function returns a plot where each point represents the intensity of a peak in a sample. Peaks are coloured according to a metadata class.

### Usage

```
plot_peaks(dataset, column.class, samples = NULL, variable.bounds = NULL,
xlab = NULL, ylab = NULL, legend.place = "topright", cex = 0.8,
reverse.x = FALSE, p.size=0.5, ...)
```

**Arguments**

<code>dataset</code>	list representing the dataset from a metabolomics experiment.
<code>column.class</code>	string indicating the metadata's variable.
<code>samples</code>	vector with samples' names, if NULL all the samples will be considered.
<code>variable.bounds</code>	numeric vector with two elements indicating the interval of x-values to plot.
<code>xlab</code>	x-axis label.
<code>ylab</code>	y-axis label.
<code>legend.place</code>	string indicating the place that the legend's box will be placed.
<code>cex</code>	numeric value that indicates the amount by which the legend is magnified relative to the default.
<code>reverse.x</code>	boolean value indicating if the x-axis will be shown reversed or not.
<code>p.size</code>	numeric value indicating the amount by which the plot points are magnified relative to the default.
<code>...</code>	additional parameters to <code>matplot</code> .

**Examples**

```
data(propolis)
plot_peaks(propolis, "seasons", variable.bounds = c(0,3), samples=c("XX_au", "XX_sm", "XX_wi"))
```

---

```
plot_regression_coefs_pvalues
Plot regression coefficient and p-values
```

---

**Description**

Plots the linear regression coefficient and the p-values.

**Usage**

```
plot_regression_coefs_pvalues(linreg.results, bar.col = NULL,
coef.size = 5, ...)
```

**Arguments**

<code>linreg.results</code>	linear regression results.
<code>bar.col</code>	color of the bars.
<code>coef.size</code>	coefficient font size.
<code>...</code>	additional parameters to <code>geom_text</code> and <code>geom_bar</code> from <code>ggplot</code> .



---

plot_spectra	<i>Plot spectra</i>
--------------	---------------------

---

**Description**

Plot spectra from dataset.

**Usage**

```
plot_spectra(dataset, column.class, func = NULL, samples = NULL,
  variable.bounds = NULL, xlab = NULL, ylab = NULL, lty = 1,
  legend.place = "topright", cex = 0.8, reverse.x = FALSE, ...)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
column.class	string indicating the metadata's variable.
func	function to compute the summary statistics to apply to the data.
samples	vector with samples' names, if NULL all the samples will be considered.
variable.bounds	numeric vector with two elements indicating the interval of x-values to plot.
xlab	x-axis label.
ylab	y-axis label.
lty	parameter of matplot.
legend.place	string indicating the place that the legend's box will be placed.
cex	numeric value that indicates the amount by which the legend is magnified relative to the default.
reverse.x	boolean value indicating if the x-axis will be shown reversed or not.
...	additional parameters to matplot.

---

plot_spectra_simple	<i>Plot spectra (simple)</i>
---------------------	------------------------------

---

**Description**

Plot spectra from dataset (simple version).

**Usage**

```
plot_spectra_simple(dataset, samples = NULL,
  variable.bounds = NULL, xlab = NULL, ylab = NULL,
  lty = 1, lwd = 1, col = 1, reverse.x = FALSE, ...)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
samples	vector with samples' names, if NULL all the samples will be considered.
variable.bounds	numeric vector with two elements indicating the interval of x-values to plot.
xlab	x-axis label.
ylab	y-axis label
lty	parameter of matplot.
lwd	parameter of matplot.
col	parameter of matplot.
reverse.x	boolean value indicating if the x-axis will be shown reversed or not.
...	additional parameters to pass to matplot.

---

plot\_ttests

*Plot t-tests results*

---

**Description**

Function for plotting the results from t-tests.

**Usage**

```
plot_ttests(dataset, tt.results, tt.threshold = 0.01)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
tt.results	t-tests results.
tt.threshold	t-test threshold for the p-value.

**Examples**

```
## Example of plotting the t-tests results
data(cachexia)
ttests.results = tTests_dataset(cachexia, "Muscle.loss")
plot_ttests(cachexia, ttests.results, 0.05)
```

---

predict_samples	<i>Predict samples</i>
-----------------	------------------------

---

**Description**

Predict new samples.

**Usage**

```
predict_samples(train.result, new.samples)
```

**Arguments**

```
train.result    result from training a classifier.
new.samples     dataframe with new samples.
```

**Value**

Returns a data frame with the samples and the predicted class.

**Examples**

```
## Example of predicting samples
training.result = train_models_performance(cachexia, "pls",
  "Muscle.loss", "cv")
result = predict_samples(training.result$final.models$pls, cachexia$data)
```

---

propolis	<i>Brazilian Propolis from different Harvest Seasons and different Agroecological Regions (dataset)</i>
----------	---

---

**Description**

Propolis or bee glue is a sticky dark-colored substance produced from the collected buds or exudates of plants (resin) by bees (*Apis mellifera* L.). The resin is masticated, salivary enzymes are added, and the partially digested material is mixed with beeswax and used in the hive to seal the walls, strengthen the borders of combs, and embalm dead invaders (Wollenweber et al., 1990). The propolis samples are from NMR data and were collected in the autumn (AU), winter (WI), spring (SP), and summer (SM) of 2010 from *Apis mellifera* hives located in southern Brazil (Santa Catarina State). A total of 59 samples were collected, and the distribution of samples by seasons being: SM - 16 samples, AU and SP - 15 samples, WI - 13 samples. Also, three agroecological regions were defined for the different apiaries, and one distributed as follows: Highlands - 12 samples, Plain - 11 samples, Plateau - 36 samples.

**Usage**

```
data(propolis)
```

**Format**

An object of class "list"

**References**

E. Wollenweber, B. M. Hausen, and W. Greenaway. Phenolic constituents and sensitizing properties of propolis, poplar balsam and balsam of peru. *Bulletin de Groupe Polyphenol*, 15:112-120, 1990. M. Maraschin, A. Somensi-Zeggio, S. K. Oliveira, S. Kuhnen, M. M. Tomazzoli, A. C. M. Zeri, R. Carreira, and M. Rocha. A machine learning and chemometrics assisted interpretation of spectroscopic data - a nmr-based metabolomics platform for the assessment of brazilian propolis. 2012

**Examples**

```
data(propolis)
sum_dataset(propolis)
```

---

propolisSampleList	<i>Brazilian Propolis from different Harvest Seasons and different Agroecological Regions (sample list)</i>
--------------------	---

---

**Description**

Propolis or bee glue is a sticky dark-colored substance produced from the collected buds or exudates of plants (resin) by bees (*Apis mellifera* L.). The resin is masticated, salivary enzymes are added, and the partially digested material is mixed with beeswax and used in the hive to seal the walls, strengthen the borders of combs, and embalm dead invaders (Wollenweber et al., 1990). The propolis samples are from NMR data and were collected in the autumn (AU), winter (WI), spring (SP), and summer (SM) of 2010 from *Apis mellifera* hives located in southern Brazil (Santa Catarina State). A total of 59 samples were collected, and the distribution of samples by seasons being: SM - 16 samples, AU and SP - 15 samples, WI - 13 samples. Also, three agroecological regions were defined for the different apiaries, and one distributed as follows: Highlands - 12 samples, Plain - 11 samples, Plateau - 36 samples.

**Usage**

```
data(propolisSampleList)
```

**Format**

An object of class "list"

## References

E. Wollenweber, B. M. Hausen, and W. Greenaway. Phenolic constituents and sensitizing properties of propolis, poplar balsam and balsam of peru. *Bulletin de Groupe Polyphenol*, 15:112-120, 1990. M. Maraschin, A. Somensi-Zeggio, S. K. Oliveira, S. Kuhnén, M. M. Tomazzoli, A. C. M. Zeri, R. Carreira, and M. Rocha. A machine learning and chemometrics assisted interpretation of spectroscopic data - a nmr-based metabolomics platform for the assessment of brazilian propolis. 2012

## Examples

```
data(propolisSampleList)
propolisSampleList[[1]]
```

---

read_Bruker_files	<i>Read Bruker processed spectra.</i>
-------------------	---------------------------------------

---

## Description

This functions read a directory containing directories where each one corresponds to a Bruker spectrum directory.

A CSV file with the names of the samples and to which Bruker spectrum directory (directory name) they correspond to should be given, unless directories' names correspond to the samples names.

## Usage

```
read_Bruker_files(bruker_directory, metadata_file=NULL,
                 m.header_col=TRUE, m.header_row=TRUE, m.sep=",",
                 samples.names=NULL, zipped=TRUE,
                 description="", label.x="ppm", label.values="intensity")
```

## Arguments

bruker_directory	Path of the directory with all the directories of the Bruker spectra.
metadata_file	Path of the metadata file.
m.header_col	Boolean value indicating if the metadata CSV file contains a header column with the name of the metadata variables.
m.header_row	Boolean value indicating if the metadata CSV file contains a header row with the name of the samples.
m.sep	The separator character of the metadata file.
samples.names	CSV file where the first column represents the samples names and in the second column the names of the spectra directories to which they correspond. If NULL, it will be considered that the directories names are the samples names (it has to be the same names that appear in the metadata file).
zipped	Boolean value indicating if the spectra directories are zipped or not.

description	A short text describing the dataset.
label.x	The label for the x values.
label.values	The label for the y values.

**Value**

Returns a list representing a dataset for specmine.

---

read_csvs_folder	<i>Read CSVs from folder</i>
------------------	------------------------------

---

**Description**

Reads multiple CSV files in a given folder.

**Usage**

```
read_csvs_folder(foldername, ...)
```

**Arguments**

foldername	string with the name of the folder.
...	additional parameters to read.csv function.

**Value**

Returns a list of data frames.

---

read_dataset_csv	<i>Read dataset from CSV</i>
------------------	------------------------------

---

**Description**

Reads the data from a CSV file and creates the dataset.

**Usage**

```
read_dataset_csv(filename.data, filename.meta = NULL,
  type = "undefined", description = "", label.x = NULL,
  label.values = NULL, sample.names = NULL, format = "row",
  header.col = TRUE, header.row = TRUE, sep = ",",
  header.col.meta = TRUE, header.row.meta = TRUE, sep.meta = ",")
```

**Arguments**

filename.data	name of the data file.
filename.meta	name of the metadata file.
type	type of the data.
description	a short text describing the dataset.
label.x	the label for the x values.
label.values	the label for the y values.
sample.names	the names of the samples.
format	format which the data are in the CSV file. It can be "row" if the samples are in the rows or "col" if the samples are in the columns.
header.col	boolean value indicating if the CSV contains a header column with the names of the samples or variables.
header.row	boolean value indicating if the CSV contains a header row with the names of the samples or variables.
sep	the separator character.
header.col.meta	boolean value indicating if the metadata CSV file contains a header column with the name of the metadata variables.
header.row.meta	boolean value indicating if the metadata CSV file contains a header row with the name of the samples.
sep.meta	the separator character of the metadata file.

**Value**

Returns the dataset from the CSV file.

---

read_dataset_dx	<i>Read dataset from (J)DX files</i>
-----------------	--------------------------------------

---

**Description**

Reads the data from the (J)DX files and creates the dataset.

**Usage**

```
read_dataset_dx(folder.data, filename.meta = NULL,
type = "undefined", description = "", label.x = NULL,
label.values = NULL, header.col.meta = TRUE,
header.row.meta = TRUE, sep.meta = ",")
```

**Arguments**

folder.data	string containing the path of the data folder.
filename.meta	name of the metadata file.
type	type of the data.
description	a short text describing the dataset.
label.x	the label for the x values.
label.values	the label for the y values.
header.col.meta	boolean value indicating if the metadata CSV file contains a header column with the name of the metadata variables.
header.row.meta	boolean value indicating if the metadata CSV file contains a header row with the name of the samples.
sep.meta	the separator character of the metadata file.

**Value**

Returns the dataset from the (J)DX files.

---

read_dataset_spc	<i>Read dataset from SPC files</i>
------------------	------------------------------------

---

**Description**

Reads the data from the SPC files and creates the dataset.

**Usage**

```
read_dataset_spc(folder.data, filename.meta = NULL,
  type = "undefined", description = "", nosubhdr = FALSE,
  label.x = NULL, label.values = NULL, header.col.meta = TRUE,
  header.row.meta = TRUE, sep.meta = ",")
```

**Arguments**

folder.data	string containing the path of the data folder.
filename.meta	name of the metadata file.
type	type of the data.
description	a short text describing the dataset.
nosubhdr	if TRUE, it won't read the subheader.
label.x	the label for the x values.
label.values	the label for the y values.



header.col.meta	boolean value indicating if the metadata CSV file contains a header column with the name of the metadata variables.
header.row.meta	boolean value indicating if the metadata CSV file contains a header row with the name of the samples.
sep.meta	the separator character of the metadata file.

**Value**

Returns the dataset from the SPC files.

---

read_data_csv	<i>Read CSV data</i>
---------------	----------------------

---

**Description**

Reads the data from the CSV file.

**Usage**

```
read_data_csv(filename, format = "row", header.col = TRUE,
header.row = TRUE, sep = ",")
```

**Arguments**

filename	name of the file with the data.
format	format which the data are in the CSV file. It can be "row" if the samples are in the rows or "col" if the samples are in the columns.
header.col	boolean value indicating if the CSV contains a header column with the names of the samples or variables.
header.row	boolean value indicating if the CSV contains a header row with the names of the samples or variables.
sep	the separator character.

**Value**

Returns a numeric matrix with the data.

---

read_data_dx	<i>Read data from (J)DX files</i>
--------------	-----------------------------------

---

**Description**

Reads the data from the (J)DX files.

**Usage**

```
read_data_dx(foldername, debug = 0)
```

**Arguments**

foldername	string with the path of the data folder.
debug	debug option for readJDX's readJDX function.

**Value**

Returns a list with the samples and the respective data points.

---

read_data_spc	<i>Read data from SPC files</i>
---------------	---------------------------------

---

**Description**

Reads the data from the SPC files.

**Usage**

```
read_data_spc(foldername, nosubhdr = FALSE)
```

**Arguments**

foldername	string with the path of the data folder.
nosubhdr	if TRUE, it won't read the subheader.

**Value**

Returns a list with the samples and the respective data points.

---

read_metadata	<i>Read metadata</i>
---------------	----------------------

---

**Description**

Read the metadata from a file.

**Usage**

```
read_metadata(filename, header.col = TRUE, header.row = TRUE,  
sep = ",")
```

**Arguments**

filename	name of the file with the metadata.
header.col	boolean value indicating if the metadata CSV file contains a header column with the name of the metadata variables.
header.row	boolean value indicating if the metadata CSV file contains a header row with the name of the samples.
sep	the separator character.

**Value**

Returns a data frame with the metadata.

---

read_ms_spectra	<i>Read MS spectra</i>
-----------------	------------------------

---

**Description**

Read the data from the MS files and creates the dataset.

**Usage**

```
read_ms_spectra(folder.name, type = "undefined",  
filename.meta = NULL, description = "", prof.method = "bin",  
fwhm = 30, bw = 30, intvalue = "into", header.col.meta = TRUE,  
header.row.meta = TRUE, sep.meta = ",")
```

**Arguments**

folder.name	string containing the path of the data folder.
type	type of the data.
filename.meta	name of the metadata file.
description	a short text describing the dataset.
prof.method	profmethod parameter from xcmsSet function from xcms package.
fwhm	fwhm parameter from xcmsSet function from xcms package. A commonly used value is 30 (seconds) for LC-MS and 4 (seconds) for GC-MS spectra.
bw	bw parameter from group function from xcms package.
intvalue	value parameter from groupval function from xcms package. It can be: <ul style="list-style-type: none"> <li>• <b>"into"</b> - integrated area of original (raw) peak</li> <li>• <b>"intf"</b> - integrated area of filtered peak.</li> <li>• <b>"maxo"</b> - maximum intensity of original (raw) peak.</li> <li>• <b>"maxf"</b> - maximum intensity of filtered peak.</li> </ul>
header.col.meta	boolean value indicating if the metadata CSV file contains a header column with the name of the metadata variables.
header.row.meta	boolean value indicating if the metadata CSV file contains a header row with the name of the samples.
sep.meta	the separator character of the metadata file.

**Value**

Returns a dataset from the MS files.

---

read\_multiple\_csvs      *Read multiple CSVs*

---

**Description**

Reads multiple CSVs, each one with a sample.

**Usage**

```
read_multiple_csvs(filenamees, ext = ".csv", ...)
```

**Arguments**

filenamees	list of file names of the files to read.
ext	extension name.
...	additional parameters to read.csv function.

**Value**

returns a list of dataframes.

---

read_spc_nosubhdr	<i>Import for Thermo Galactic's spc file format These functions allow to import .spc files.</i>
-------------------	---

---

## Description

Import for Thermo Galactic's spc file format These functions allow to import .spc files.

## Usage

```
read_spc_nosubhdr(filename, keys.hdr2data = c("fexper", "fres", "fsource"),
  keys.hdr2log = c("fdate", "fpeakpt"), keys.log2data = FALSE,
  keys.log2log = TRUE, log.txt = TRUE, log.bin = FALSE,
  log.disk = FALSE, hdr = list(), no.object = FALSE)
```

## Arguments

filename	The complete file name of the .spc file.
keys.hdr2data, keys.hdr2log, keys.log2data, keys.log2log	character vectors with the names of parameters in the .spc file's log block (log2xxx) or header (hdr2xxx) that should go into the extra data (yyy2data) or into the long.description field of the returned hyperSpec object's log (yyy2log). All header fields specified in the .spc file format specification (see below) are imported and can be referred to by their de-capitalized names.
log.txt	Should the text part of the .spc file's log block be read?
log.bin, log.disk	Should the normal and on-disk binary parts of the .spc file's log block be read? If so, they will be put as raw vectors into the hyperSpec object's log.
hdr	A list with fileheader fields that overwrite the settings of actual file's header. Use with care, and look into the source code for detailed insight on the elements of this list.
no.object	If TRUE, a list with wavelengths, spectra, labels, log and data are returned instead of a hyperSpec object. This parameter will likely be subject to change in future - use with care.

## Value

If the file contains multiple spectra with individual wavelength axes, read.spc returns a list of hyperSpec objects. Otherwise the result is a hyperSpec object.

read.spc.KaiserMap returns a hyperSpec object with data columns x, y, and z containing the stage position as recorded in the .spc files' log.

**Note**

Only a restricted set of test files was available for development. Particularly, the w-planes feature could not be tested.

If you have .spc files that cannot be read with these function, don't hesitate to contact the package maintainer with your code patch or asking advice.

**Author(s)**

C. Beleites

**See Also**

[textio](#)

---

read\_varian\_spectra\_raw

*Function that reads raw spectra (intensity over time spectra) from the varian format and processes them to ppm spectra.*

---

**Description**

This function read raw spectra (i.e. intensity over time spectra) from the varian format and processess them to intensity over ppm spectra. For this function to work, in each spectrum directory should be present a fid and procpa files.

Python 3 with modules nmrglue must be installed.

**Usage**

```
read_varian_spectra_raw(varian_spectra_directory,  
metadata_file=NULL, m.header_col=TRUE, m.header_row=TRUE, m.sep=",",  
samples.names=NULL, zero_filling=TRUE, apodization=TRUE, zipped=TRUE,  
description="", label.x="ppm", label.values="intensity")
```

**Arguments**

varian_spectra_directory	Path of the directory with all the directories of the varian spectra.
metadata_file	Path of the metadata file.
m.header_col	Boolean value indicating if the metadata CSV file contains a header column with the name of the metadata variables.
m.header_row	Boolean value indicating if the metadata CSV file contains a header row with the name of the samples.
m.sep	The separator character of the metadata file.

<code>samples.names</code>	CSV file where the first column represents the samples names and in the second column the names of the spectra directories to which they correspond. If NULL, it will be considered that the directories names are the samples names (it has to be the same names that appear in the metadata file).
<code>zero_filling</code>	boolean value indicating whether zero-filling should be performed or not when processing the fid spectra. Defaults to TRUE.
<code>apodization</code>	boolean value indicating whether apodization should be performed or not when processing the fid spectra. Defaults to TRUE.
<code>zipped</code>	Boolean value indicating if the spectra directories are zipped or not.
<code>description</code>	A short text describing the dataset.
<code>label.x</code>	The label for the x values.
<code>label.values</code>	The label for the y values.

**Value**

Returns a list representing a dataset for specmine.

**Warning**

You must not call this function unless you have Python ( $\geq 3.5.2$ ) installed in your machine and the module `nmgglue`.

---

```
recursive_feature_elimination
```

*Perform recursive feature elimination*

---

**Description**

Perform recursive feature elimination on the dataset using `caret`'s package.

**Usage**

```
recursive_feature_elimination(datamat, samples.class,
  functions = caret::rfFuncs, method = "cv", repeats = 5,
  number = 10, subsets = 2^(2:4))
```

**Arguments**

<code>datamat</code>	data matrix from dataset.
<code>samples.class</code>	string or index indicating what metadata to use.
<code>functions</code>	a list of functions for model fitting, prediction and variable importance.
<code>method</code>	the external resampling method: <code>boot</code> , <code>cv</code> , <code>LOOCV</code> or <code>LGOCV</code> (for repeated training/test splits).
<code>repeats</code>	for repeated k-fold cross-validation only: the number of complete sets of folds to compute.

number either the number of folds or number of resampling iterations.  
subsets a numeric vector of integers corresponding to the number of features that should be retained.

### Value

A caret's rfe object with the result of recursive feature selection.

### Examples

```
## Example of recursive feature elimination
data(cachexia)
library(caret)
rfe.result = recursive_feature_elimination(cachexia$data,
  cachexia$metadata$Muscle.loss, functions = caret::rfFuncs,
  method = "cv", number = 3, subsets = 2^(1:6))
```

---

remove\_data

*Remove data*

---

### Description

Remove data from the dataset.

### Usage

```
remove_data(dataset, data.to.remove, type = "sample",
  by.index = FALSE, rebuild.factors = TRUE)
```

### Arguments

dataset list representing the dataset from a metabolomics experiment.  
data.to.remove vector with the elements' names to remove  
type type of the element to remove. It can be:

- **"sample"** to remove samples.
- **"data"** to remove variables.
- **"metadata"** to remove metadata's variables.

by.index if TRUE, the values of the data.to.remove argument are indexes in case the type is "data".  
rebuild.factors if TRUE, rebuilds the factors from metadata.

### Value

Returns the dataset with the specified data removed.



**Examples**

```
## Example of removing data
data(cachexia)
dataset = remove_data(cachexia, c("Creatine","Serine"), type = "data",
  by.index = FALSE)
```

---

remove\_data\_variables *Remove data variables*

---

**Description**

Remove data variables from the dataset.

**Usage**

```
remove_data_variables(dataset, variables.to.remove,
  by.index = FALSE)
```

**Arguments**

`dataset` list representing the dataset from a metabolomics experiment.

`variables.to.remove` vector with the variables' names to remove.

`by.index` if TRUE, the values of the `variables.to.remove` argument are indexes.

**Value**

Returns the dataset with the specified data variables removed.

**Examples**

```
## Example of removing data variables
data(cachexia)
dataset = remove_data_variables(cachexia, c("Creatine","Serine"),
  by.index = FALSE)
```

---

remove\_metadata\_variables  
*Remove metadata's variables*

---

**Description**

Remove metadata's variables from the dataset

**Usage**

```
remove_metadata_variables(dataset, variables.to.remove)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
variables.to.remove    vector with the metadata's variables to remove.

**Value**

Returns the dataset with the selected metadata's variables removed.

**Examples**

```
## Example of removing metadata's variables  
dataset = remove_metadata_variables(propolis, c("seasons"))
```

---

remove\_peaks\_interval *Remove interval of peaks*

---

**Description**

Removes peaks from a given interval.

**Usage**

```
remove_peaks_interval(sample.df, peak.val.min, peak.val.max)
```

**Arguments**

sample.df            data frame with the samples' peaks.  
peak.val.min        minimum peak value.  
peak.val.max        maximum peak value.

**Value**

Returns a data frame with the filtered samples' peaks.

**Examples**

```
## Example of removing a interval of peaks
data(propolisSampleList)
samples.df = remove_peaks_interval(propolisSampleList[[1]], 2, 8.05)
```

---

remove\_peaks\_interval\_sample\_list

*Remove interval of peaks (sample list)*

---

**Description**

Removes peaks on a sample list given a peak interval.

**Usage**

```
remove_peaks_interval_sample_list(sample.list, peak.val.min,
peak.val.max)
```

**Arguments**

sample.list	list of data frames with the samples' peaks.
peak.val.min	minimum peak value.
peak.val.max	maximum peak value.

**Value**

Returns the sample list with the filtered peaks.

**Examples**

```
## Example of removing a interval of peaks in a sample list
data(propolisSampleList)
samples.list = remove_peaks_interval_sample_list(propolisSampleList, 2, 8.05)
```

---

remove_samples	<i>Remove samples</i>
----------------	-----------------------

---

**Description**

Remove samples from the dataset.

**Usage**

```
remove_samples(dataset, samples.to.remove, rebuild.factors = TRUE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
samples.to.remove	vector with the sample's names to remove.
rebuild.factors	if TRUE, rebuilds the factors from metadata.

**Value**

Returns the dataset with the specified samples removed.

**Examples**

```
## Example of removing samples
data(cachexia)
cachexia = remove_samples(cachexia, c("PIF_178", "PIF_090"))
```

---

remove_samples_by_nas	<i>Remove samples by NAs</i>
-----------------------	------------------------------

---

**Description**

Remove samples from the dataset by the number of NAs

**Usage**

```
remove_samples_by_nas(dataset, max.nas = 0, by.percent = FALSE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
max.nas	number of NAs (or percentage) to which samples with more NAs will be removed.
by.percent	if TRUE the max.nas argument is a percentage.

**Value**

Returns the dataset with the samples with more NAs than the limit removed.

**Examples**

```
## Example of removing samples by NAs
data(propolis)
propolis = remove_samples_by_nas(propolis, 40, by.percent = TRUE)
```

---

remove\_samples\_by\_na\_metadata

*Remove samples by NA on metadata*

---

**Description**

Remove samples from the dataset with the metadata's variable value with NAs.

**Usage**

```
remove_samples_by_na_metadata(dataset, metadata.var)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
metadata.var      metadata's variable.

**Value**

Returns the dataset with the samples with NA on metadata's variable removed.

**Examples**

```
## Example of removing samples that have NAs on metadata's variable
data(cachexia)
cachexia$metadata$Muscle.loss[10] = NA
cachexia = remove_samples_by_na_metadata(cachexia, "Muscle.loss")
```

---

remove\_variables\_by\_nas

*Remove variables by NAs*

---

**Description**

Remove variables from the dataset by the number of NAs

**Usage**

```
remove_variables_by_nas(dataset, max.nas = 0, by.percent = FALSE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
max.nas	number of NAs (or percentage) to which samples with more NAs will be removed.
by.percent	if TRUE the max.nas argument is a percentage.

**Value**

Returns the dataset with the variables with more NAs than the limit removed.

**Examples**

```
## Example of removing variables by NAs
data(propolis)
propolis = remove_variables_by_nas(propolis, 40, by.percent = TRUE)
```

---

remove\_x\_values\_by\_interval

*Remove x-values by interval*

---

**Description**

Remove an interval of x-values from the dataset.

**Usage**

```
remove_x_values_by_interval(dataset, min.value, max.value)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
min.value	minimum value of the interval.
max.value	maximum value of the interval.

**Value**

Returns the dataset with the specified interval of x-values removed.

---

replace\_data\_value      *Replace data value*

---

**Description**

Replace a data value for a new value on the dataset.

**Usage**

```
replace_data_value(dataset, x.axis.val, sample, new.value,  
by.index = FALSE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
x.axis.val	variable index or name.
sample	sample name.
new.value	new value to replace the old value.
by.index	boolean value to indicate if the x.axis.val is an index or not.

**Value**

Returns the dataset with the data value replaced.

**Examples**

```
## Example of replacing a data value from the dataset  
data(cachexia)  
dataset = replace_data_value(cachexia, "Creatine", "PIF_178", 10.3,  
by.index = FALSE)
```

---

`replace_metadata_value`*Replace metadata's value*

---

**Description**

Replace a metadata's variable value of a sample.

**Usage**

```
replace_metadata_value(dataset, variable, sample, new.value)
```

**Arguments**

<code>dataset</code>	list representing the dataset from a metabolomics experiment.
<code>variable</code>	metadata's variable.
<code>sample</code>	name of the sample.
<code>new.value</code>	new value of the metadata's variable.

**Value**

Returns the dataset with the metadata updated.

**Examples**

```
## Example of replacing metadata's variable value of a sample
data(cachexia)
dataset = replace_metadata_value(cachexia, "Muscle.loss", "PIF_178",
  "control")
```

---

`savitzky_golay`*Savitzky-golay transformation*

---

**Description**

Smoothing and derivative of the data using Savitzky-Golay.

**Usage**

```
savitzky_golay(dataset, p.order, window, deriv = 0)
```



**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
p.order	integer value representing the polynomial order.
window	odd number indicating the window size.
deriv	integer value indicating the differentiation order.

**Value**

Returns the dataset with the spectra smoothed using Savitzky-Golay.

---

scaling	<i>Scale dataset</i>
---------	----------------------

---

**Description**

Performs scaling according to a method.

**Usage**

```
scaling(dataset, method = "auto")
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
method	string specifying the scaling method. The possible values are: <ul style="list-style-type: none"><li>• <b>"auto"</b> auto scaling.</li><li>• <b>"range"</b> range scaling.</li><li>• <b>"pareto"</b> pareto scaling.</li><li>• <b>"tointerval"</b> scaling to an interval.</li></ul>

**Value**

Returns the dataset scaled.

---

scaling_samples	<i>Scale data matrix</i>
-----------------	--------------------------

---

**Description**

Performs scaling according to a method.

**Usage**

```
scaling_samples(datamat, method = "auto")
```

**Arguments**

datamat	data matrix.
method	string specifying the scaling method. The possible values are: <ul style="list-style-type: none"><li>• <b>"auto"</b> auto scaling.</li><li>• <b>"range"</b> range scaling.</li><li>• <b>"pareto"</b> pareto scaling.</li><li>• <b>"tointerval"</b> scaling to an interval.</li></ul>

**Value**

Returns the data matrix scaled.

---

set_metadata	<i>Set new metadata</i>
--------------	-------------------------

---

**Description**

Updates the dataset's metadata with a new one.

**Usage**

```
set_metadata(dataset, new.metadata)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
new.metadata	matrix or dataframe with the new metadata.

**Value**

Returns the dataset with the updated metadata.

**Examples**

```
## Example of setting a new metadata to the dataset
data(cachexia)
new.metadata = c(rep("meta1", 39), rep("meta2", 38))
new.metadata = data.frame(var_meta = new.metadata)
rownames(new.metadata) = get_sample_names(cachexia)
cachexia = set_metadata(cachexia, new.metadata)
```

---

set_sample_names	<i>Set samples names</i>
------------------	--------------------------

---

**Description**

Set new samples names to the dataset.

**Usage**

```
set_sample_names(dataset, new.sample.names)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
new.sample.names	vector with the new samples names.

**Value**

Returns the dataset with the updated samples names.

**Examples**

```
## Example of setting a new value label to the dataset
data(cachexia)
new.samples.names = as.character(1:77)
cachexia = set_sample_names(cachexia, new.samples.names)
```

---

set_value_label	<i>Set value label</i>
-----------------	------------------------

---

**Description**

Set a new value label for the dataset.

**Usage**

```
set_value_label(dataset, new.val.label)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
new.val.label    string with the new value label.

**Value**

Returns the dataset with the updated value label.

**Examples**

```
## Example of setting a new value label to the dataset  
data(cachexia)  
cachexia = set_value_label(cachexia, "new value label")
```

---

set_x_label	<i>Set x-label</i>
-------------	--------------------

---

**Description**

Set a new x-label to the dataset.

**Usage**

```
set_x_label(dataset, new.x.label)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
new.x.label        string with the x-label.

**Value**

Returns the dataset with the updated x-label.

**Examples**

```
## Example of setting a new x-label to the dataset
data(cachexia)
cachexia = set_x_label(cachexia, "new x-label")
```

---

set_x_values	<i>Set new x-values</i>
--------------	-------------------------

---

**Description**

Set new x-values to the dataset

**Usage**

```
set_x_values(dataset, new.x.values, new.x.label = NULL)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
new.x.values	vector with the new x-values.
new.x.label	string with the new x-label (can be NULL).

**Value**

Returns the dataset with the updated x-values.

**Examples**

```
## Example of setting new x-values to the dataset
data(cachexia)
new.xvalues = 1:63
cachexia = set_x_values(cachexia, new.xvalues, new.x.label = NULL)
```

---

shift_correction	<i>Shift correction</i>
------------------	-------------------------

---

**Description**

Shifts the spectra according to a specific method.

**Usage**

```
shift_correction(dataset, method = "constant", shift.val = 0,
interp.function = "linear", ref.limits = NULL)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
method	string that indicates the shifting method. It can be: <ul style="list-style-type: none"> <li>• "constant" uses a constant shift that is added to the x-values</li> <li>• "interpolation" uses interpolation according to "interp.function"</li> </ul>
shift.val	value of the shift (for constant and interpolation methods); can be a single value for all spectra, can be the string "auto", the shifts are automatically determined or a vector with the size of the number of samples with the shifts for each spectra.
interp.function	string that represents the interpolation function, can be "linear" or "spline".
ref.limits	vector with 2 elements that represents the reference limits to calculate the shifts.

**Value**

Returns the dataset with the spectras shifted.

---

smoothing\_interpolation  
*Smoothing interpolation*

---

**Description**

Performs smoothing interpolation according to a specific method.

**Usage**

```
smoothing_interpolation(dataset, method = "bin",
  reducing.factor = 2, x.axis = NULL, p.order = 3,
  window = 11, deriv = 0)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
method	string specifying the smoothing method. The three possible methods are: "bin", "loess" and "savitzky.golay".
reducing.factor	numeric value indicating the reducing factor for bin method.
x.axis	numeric vector representing the new x-axis for loess method.
p.order	numeric value representing the polynomial order for savitzky-golay method.
window	numeric value indicating the size of the window for savitzky-golay method. (must be an odd number)
deriv	numeric value indicating the differentiation order for savitzky-golay method.

**Value**

Returns the dataset with the spectra smoothed.

---

snv_dataset	<i>Standard Normal Variate</i>
-------------	--------------------------------

---

**Description**

Performs Standard Normal Variate on the dataset.

**Usage**

```
snv_dataset(dataset)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.

**Value**

Returns the dataset with the data normalized by SNV.

---

spectra_options	<i>Information on the library of NMR reference spectra in our package.</i>
-----------------	--

---

**Description**

This dataset provides all the information on the library of NMR spectra used as references in NMR metabolite identification.

**Usage**

```
data("spectra_options")
```

**Format**

A data frame with 1816 observations on the following 9 variables. Each observation corresponds to a spectrum in our library.

SPCMNS a character vector with the spectra IDs.

SPCMNM a character vector with the metabolite IDs of the corresponding spectra.

FREQUENCY a character vector with the frequencies under which the spectra were obtained.

NUCLEUS a character vector mentioned the nucleus examined. All observations are '1H'.

PH a character vector with the pH of the samples from which the spectra were obtained. May contain missing values.

TEMPERATURE a character vector with the temperature under which the spectra were obtained. May contain missing values.

SOLVENT a character vector with the solvent of the samples from which the spectra were obtained.

ORIGINAL\_DATABASE\_ID whenever available, a character vector with the ID of the corresponding spectra from the database it was originally acquired from.

DATABASE a character vector specifying from which database the spectra were taken from.

## References

The spectra were taken from the following databases: HMDB (<https://hmdb.ca>), BMRB (<http://www.bmrw.wisc.edu>) and SDBS (<https://sdfs.db.aist.go.jp>). Some spectra were internally acquired and are mentioned as OUR in the DATABASE variable.

## Examples

```
data(spectra_options)
```

---

stats_by_sample	<i>Statistics of samples</i>
-----------------	------------------------------

---

## Description

Get a summary of statistics of the samples.

## Usage

```
stats_by_sample(dataset, samples = NULL)
```

## Arguments

dataset	list representing the dataset from a metabolomics experiment.
samples	if defined restricts the application to a given set of samples.

## Value

Returns a vector with the a summary of statistics of the samples.

## Examples

```
## Example of getting stats of samples  
data(cachexia)  
samples.stats.result = stats_by_sample(cachexia)
```



---

stats_by_variable	<i>Statistics of variables</i>
-------------------	--------------------------------

---

**Description**

Get a summary of statistics of the variables.

**Usage**

```
stats_by_variable(dataset, variables = NULL,  
variable.bounds = NULL)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
variables	allows to define which variables to calculate the stats (if numbers, indexes are assumed).
variable.bounds	allow to define an interval of variables (if numeric).

**Value**

Returns a vector with the a summary of statistics of the variables.

**Examples**

```
## Example of getting stats of variables  
data(cachexia)  
variable.stats.result = stats_by_variable(cachexia)
```

---

subset_by_samples_and_xvalues	<i>Subset by samples and x-values</i>
-------------------------------	---------------------------------------

---

**Description**

Gets a subset of specific samples and x-values.

**Usage**

```
subset_by_samples_and_xvalues(dataset, samples, variables = NULL,  
by.index = FALSE, variable.bounds = NULL, rebuild.factors = TRUE)
```

**Arguments**

**dataset** list representing the dataset from a metabolomics experiment.  
**samples** vector with indexes or names of the samples to select  
**variables** vector with the desired variables to get from the dataset.  
**by.index** if TRUE, the values of the variables argument are indexes.  
**variable.bounds** variable bounds used if by.index is FALSE and variables are NULL.  
**rebuild.factors** if TRUE the metadata factors are rebuilt.

**Value**

Returns the dataset with the selected samples and x-values.

**Examples**

```
## Example of subsetting samples and x-values
data(cachexia)
subset = subset_by_samples_and_xvalues(cachexia, c("PIF_178", "NETL_022_V1"),
  variables = c("Creatine", "Serine"))
```

---

subset_metadata	<i>Subset metadata</i>
-----------------	------------------------

---

**Description**

Subsets the metadata according to the specified metadata's variables.

**Usage**

```
subset_metadata(dataset, variables)
```

**Arguments**

**dataset** list representing the dataset from a metabolomics experiment.  
**variables** metadata's variables.

**Value**

Returns the dataset with the metadata subsetted.

**Examples**

```
## Example of subsetting samples
data(propolis)
subset = subset_metadata(propolis, c("seasons"))
```

---

subset\_random\_samples *Subset random samples*

---

**Description**

Gets a subset of random samples from the dataset.

**Usage**

```
subset_random_samples(dataset, nsamples)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
nsamples           integer representing the number of samples that we want to get.

**Value**

Returns the dataset with a number of random samples.

**Examples**

```
## Example of subsetting random samples  
data(cachexia)  
subset = subset_random_samples(cachexia, 15)
```

---

subset\_samples            *Subset samples*

---

**Description**

Gets a subset of specific samples from the dataset.

**Usage**

```
subset_samples(dataset, samples, rebuild.factors = TRUE)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
samples            vector with indexes or names of the samples to select  
rebuild.factors    if TRUE the metadata factors are rebuilt.

**Value**

Returns the dataset with the selected set of samples.

**Examples**

```
## Example of subsetting samples
data(cachexia)
subset = subset_samples(cachexia, c("PIF_178", "PIF_132"))
```

---

subset\_samples\_by\_metadata\_values

*Subset samples by metadata values*

---

**Description**

Gets a subset of specific samples according to metadata's values from the dataset.

**Usage**

```
subset_samples_by_metadata_values(dataset, metadata.varname,
values)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
metadata.varname	name of the metadata's variable.
values	values of the metadata's variable.

**Value**

Returns the dataset with the set of samples according to the metadata's values.

**Examples**

```
## Example of subsetting samples by metadata's values
subset_samples_by_metadata_values(propolis, "seasons",
c("sm", "au"))
```

---

subset_x_values	<i>Subset x-values</i>
-----------------	------------------------

---

**Description**

Gets a subset of specific x-values from the dataset.

**Usage**

```
subset_x_values(dataset, variables, by.index = FALSE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
variables	vector with the desired variables to get from the dataset.
by.index	if TRUE, the values of the variables argument are indexes.

**Value**

Returns the dataset with the selected set of x-values.

**Examples**

```
## Example of subsetting x-values  
data(cachexia)  
subset = subset_x_values(cachexia, c(1,2,10,20), by.index = TRUE)
```

---

subset_x_values_by_interval	<i>Subset x-values by interval</i>
-----------------------------	------------------------------------

---

**Description**

Gets a subset of a specific interval of x-values.

**Usage**

```
subset_x_values_by_interval(dataset, min.value, max.value)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
min.value	the minimum value of the interval.
max.value	the maximum value of the interval.

**Value**

Returns the dataset with the selected interval of x-values.

**Examples**

```
## Example of subsetting x-values by an interval
subset = subset_x_values_by_interval(propolis, 1, 3)
```

---

summary\_var\_importance

*Summary of variables importance*

---

**Description**

Summary of variables importance of the models

**Usage**

```
summary_var_importance(performances, number.rows)
```

**Arguments**

performances    the result from training the models.  
number.rows    number of variables to display.

**Value**

Returns list with the variables importance of each model.

**Examples**

```
## Example of getting a summary of variables importance
data(cachexia)
training.result = train_models_performance(cachexia, "pls",
  "Muscle.loss", "cv")
result = summary_var_importance(training.result, 10)
```

---

sum_dataset	<i>Dataset summary</i>
-------------	------------------------

---

**Description**

Returns a summary with its main features.

**Usage**

```
sum_dataset(dataset, stats = TRUE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
stats	if TRUE prints some global statistics of the data values.

**Value**

Returns the summary of the dataset containing:

- Description
- Type of data
- Number of samples
- Number of datapoints
- Number of metadata variables if metadata not null
- Labels of x axis values and data points if labels not null

If the parameter 'stats' is TRUE then returns also:

- Number of missing values in data
- Mean of data values
- Median of data values
- Standard deviation
- Range of values
- Quantiles

**Examples**

---

train_and_predict	<i>Train and predict</i>
-------------------	--------------------------

---

### Description

Train a model and predict new unlabeled samples with that model.

### Usage

```
train_and_predict(dataset, new.samples, column.class, model,
  validation, num.folds = 10, num.repeats = 10, tunelength = 10,
  tunegrid = NULL, metric = NULL, summary.function =
  defaultSummary)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
new.samples	dataframe with new samples to predict the class label.
column.class	metadata column class.
model	model to be used in training.
validation	validation method.
num.folds	number of folds in cross validation.
num.repeats	number of repeats.
tunelength	number of levels for each tuning parameters.
tunegrid	dataframe with possible tuning values.
metric	metric used to evaluate the model's performance. Can be "Accuracy" or "ROC".
summary.function	summary function. For "ROC" the multiClassSummary function must be used.

### Value

Returns a list with the training result and the predictions result.

### Examples

```
## Example of training and predicting
data(cachexia)
result = train_and_predict(cachexia, new.samples = cachexia$data,
  "Muscle.loss", "pls", "cv")
```



---

train_classifier	<i>Train classifier</i>
------------------	-------------------------

---

## Description

Train a specific classifier.

## Usage

```
train_classifier(dataset, column.class, model, validation,  
num.folds = 10, num.repeats = 10, tunelength = 10,  
tunegrid = NULL, metric = NULL,  
summary.function = defaultSummary, class.in.metadata = TRUE)
```

## Arguments

dataset	list representing the dataset from a metabolomics experiment.
column.class	metadata column class.
model	model to be used in training.
validation	validation method.
num.folds	number of folds in cross validation.
num.repeats	number of repeats.
tunelength	number of levels for each tuning parameters.
tunegrid	dataframe with possible tuning values.
metric	metric used to evaluate the model's performance. Can be "Accuracy" or "ROC".
summary.function	summary function. For "ROC" the multiClassSummary function must be used.
class.in.metadata	boolean value to indicate if the class is in metadata.

## Value

Returns the train result object from caret.

## Examples

```
## Example of training a classifier  
data(cachexia)  
result = train_classifier(cachexia, "Muscle.loss", "pls", "cv")
```

---

 train\_models\_performance

*Train models*


---

### Description

Train various models.

### Usage

```
train_models_performance(dataset, models, column.class,
  validation, num.folds = 10, num.repeats = 10, tunelength = 10,
  tunegrid = NULL, metric = NULL, summary.function = "default",
  class.in.metadata = TRUE, compute.varimp = TRUE)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
models	models to be used in training.
column.class	metadata column class.
validation	validation method.
num.folds	number of folds in cross validation.
num.repeats	number of repeats.
tunelength	number of levels for each tuning parameters.
tunegrid	dataframe with possible tuning values.
metric	metric used to evaluate the model's performance. Can be "Accuracy" or "ROC".
summary.function	summary function. For "ROC" the multiClassSummary function must be used.
class.in.metadata	boolean value to indicate if the class is in metadata.
compute.varimp	boolean value to indicate if the var importance is calculated.

### Value

Returns a list with the results from training

performance	The results from the best tunes of the models
vips	The variable importance from the models
full.results	The full results from the tuning parameters of each model
best.tunes	The best tune of each model
confusion.matrices	The confusion matrices of the models (only in classification)
final.models	The final models

## Examples

```
## Example of training models
data(cachexia)
result = train_models_performance(cachexia, "pls",
  "Muscle.loss", "cv")
```

---

transform_data	<i>Transform data</i>
----------------	-----------------------

---

## Description

Performs data transformation according to a method.

## Usage

```
transform_data(dataset, method = "log")
```

## Arguments

dataset	list representing the dataset from a metabolomics experiment.
method	string specifying the transformation method. The possible values are: <ul style="list-style-type: none"><li>• <b>"log"</b> logarithmic transformation.</li><li>• <b>"cubicroot"</b> cubic root transformation.</li></ul>

## Value

Returns the dataset with the data transformation applied.

## Examples

```
## Example of logarithmic transformation
data(cachexia)
dataset.log = transform_data(cachexia, "log")
```

---

transmittance\_to\_absorbance  
*Convert transmittance to absorbance*

---

**Description**

Converts transmittance values to absorbance values.

**Usage**

```
transmittance_to_absorbance(dataset, percent = TRUE)
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
percent            boolean value to indicate if the absorbance values are in a percentage or not.

**Value**

Returns the dataset with the data points converted to absorbance values.

---

tTests\_dataset            *t-Tests on dataset*

---

**Description**

Run t-Tests for each row of the data from the dataset.

**Usage**

```
tTests_dataset(dataset, metadata.var, threshold = NULL,  
write.file = FALSE, file.out = "ttests.csv")
```

**Arguments**

dataset            list representing the dataset from a metabolomics experiment.  
metadata.var        metadata variable to use in the t-tests.  
threshold          threshold value of the p-value.  
write.file         boolean value to write or not a file with the results.  
file.out            name of the file.

**Value**

Table with the results of the t-tests, with p-value,  $-\log_{10}(\text{p-value})$  and false discovery rate (fdr).

**Examples**

```
## Example of t-Tests on dataset
data(cachexia)
ttests.result = tTests_dataset(cachexia, "Muscle.loss",
write.file = FALSE)
```

---

values_per_peak	<i>Values per peak</i>
-----------------	------------------------

---

**Description**

Gets the number of values on each peak.

**Usage**

```
values_per_peak(samples.df)
```

**Arguments**

samples.df      data frame with the samples' peaks.

**Value**

Returns a vector with the number of values for each peak.

**Examples**

```
## Example of getting the number of values for each peak
data(propolis)
num.peaks = values_per_peak(propolis$data)
```

---

values_per_sample	<i>Values per peak</i>
-------------------	------------------------

---

**Description**

Gets the number of values on each sample.

**Usage**

```
values_per_sample(samples.df)
```

**Arguments**

samples.df      data frame with the samples' peaks.

**Value**

Returns a vector with the number of values for each sample.

**Examples**

```
## Example of getting the number of values for each sample
data(propolis)
num.values.samples = values_per_sample(propolis$data)
```

---

variables\_as\_metadata *Variables as metadata*

---

**Description**

Use one or more data variables as metadata variables.

**Usage**

```
variables_as_metadata(dataset, variables, by.index = FALSE)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
variables	name or index of the variables that are going to be used as metadata variables.
by.index	boolean value indicating if the variables are indexes or names

**Value**

Returns the dataset with the variables removed from the data and added on the metadata.

**Examples**

```
## Example of using a variable as metadata variable
data(cachexia)
dataset = variables_as_metadata(cachexia, "Creatine", by.index = FALSE)
```

---

volcano\_plot\_fc\_tt      *Volcano plot*

---

### Description

Volcano plot to intersect the results from t-tests and fold change.

### Usage

```
volcano_plot_fc_tt(dataset, fc.results, tt.results,  
fc.threshold = 2, tt.threshold = 0.01)
```

### Arguments

dataset	list representing the dataset from a metabolomics experiment.
fc.results	fold change results.
tt.results	t-tests results.
fc.threshold	fold change threshold value.
tt.threshold	t-test p-value threshold.

### Value

Returns the name of the samples which intersects both fold change and t-tests results above the given thresholds.

### Examples

```
## Example of a volcano plot  
data(cachexia)  
foldchange.results = fold_change(cachexia, "Muscle.loss", "control")  
ttests.results = tTests_dataset(cachexia, "Muscle.loss")  
volcano_plot_fc_tt(cachexia, foldchange.results, ttests.results,  
fc.threshold = 2, tt.threshold = 0.01)
```

---

xvalue\_interval\_to\_indexes

*Get indexes of an interval of x-values*

---

### Description

Returns indexes corresponding to an interval of x-values (only to numerical values - spectra)

### Usage

```
xvalue_interval_to_indexes(dataset, min.value, max.value)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
min.value	minimum x-value of the interval.
max.value	maximum x-value of the interval.

**Value**

Returns a numeric vector with the indexes of the x-values interval

**Examples**

```
## Example of getting the indexes of an interval of x-values
data(propolis)
indexes.interval = xvalue_interval_to_indexes(propolis, 2.0, 5.5)
```

---

x\_values\_to\_indexes    *Get x-values indexes*

---

**Description**

Returns the indexes corresponding to a vector of x-values (only to numerical values - spectra)

**Usage**

```
x_values_to_indexes(dataset, x.values)
```

**Arguments**

dataset	list representing the dataset from a metabolomics experiment.
x.values	vector of x-values.

**Value**

Returns a numeric vector with the indexes of the x-values.

**Examples**

```
## Example of getting the indexes of the x-values
data(propolis)
indexes = x_values_to_indexes(propolis, c(1.3, 3.51, 6.03))
```



# Index

- \* **IO**
  - read\_spc\_nosubhdr, 109
- \* **Kolmogorov-Smirnov**
  - ksTest\_dataset, 61
  - plot\_kstest, 95
- \* **Kruskal-Wallis**
  - kruskalTest\_dataset, 60
  - plot\_kruskaltest, 94
- \* **MAIT**
  - MAIT\_identify\_metabolites, 65
- \* **~boxplot**
  - boxplot\_vars\_factor, 13
- \* **~factor**
  - boxplot\_vars\_factor, 13
- \* **absorbance**
  - absorbance\_to\_transmittance, 6
  - transmittance\_to\_absorbance, 140
- \* **aggregation**
  - aggregate\_samples, 7
- \* **alignment**
  - group\_peaks, 52
- \* **anova**
  - aov\_all\_vars, 7
  - multifactor\_aov\_all\_vars, 71
  - multifactor\_aov\_pvalues\_table, 71
  - multifactor\_aov\_varexp\_table, 72
  - plot\_anova, 93
- \* **apply**
  - apply\_by\_group, 8
  - apply\_by\_groups, 9
  - apply\_by\_sample, 9
  - apply\_by\_variable, 10
- \* **background**
  - background\_correction, 11
- \* **baseline**
  - baseline\_correction, 11
- \* **bin**
  - smoothing\_interpolation, 126
- \* **biplot**
  - pca\_biplot, 81
  - pca\_biplot3D, 82
- \* **boxplot**
  - boxplot\_variables, 12
- \* **caret**
  - multiClassSummary, 70
- \* **centering**
  - mean\_centering, 66
- \* **chemospec**
  - convert\_from\_chemospec, 17
- \* **classifier**
  - train\_classifier, 137
- \* **clustering**
  - clustering, 15
  - dendrogram\_plot, 29
  - dendrogram\_plot\_col, 30
  - hierarchical\_clustering, 54
  - kmeans\_clustering, 58
  - kmeans\_plot, 59
  - kmeans\_result\_df, 59
- \* **coefficient**
  - linreg\_coef\_table, 62
- \* **correction**
  - background\_correction, 11
  - baseline\_correction, 11
  - data\_correction, 28
  - offset\_correction, 79
  - shift\_correction, 125
- \* **correlations**
  - heatmap\_correlations, 53
- \* **correlation**
  - correlation\_test, 22
  - correlations\_dataset, 21
  - correlations\_test, 21
- \* **csv**
  - read\_csvs\_folder, 102
  - read\_dataset\_csv, 102
  - read\_metadata, 107
  - read\_multiple\_csvs, 108

- \* **cubicroot**
  - cubic\_root\_transform, 27
  - transform\_data, 139
- \* **dataframe**
  - get\_data\_as\_df, 38
- \* **datasets**
  - cachexia, 14
  - propolis, 99
  - propolisSampleList, 100
  - spectra\_options, 127
- \* **dataset**
  - check\_dataset, 14
  - convert\_from\_chemospec, 17
  - convert\_from\_hyperspec, 18
  - convert\_to\_hyperspec, 20
  - create\_dataset, 25
  - data\_correction, 28
  - dataset\_from\_peaks, 28
  - first\_derivative, 33
  - get\_data, 37
  - get\_data\_value, 38
  - get\_metadata, 45
  - get\_sample\_names, 49
  - get\_type, 50
  - get\_value\_label, 50
  - get\_x\_values\_as\_num, 51
  - get\_x\_values\_as\_text, 52
  - is\_spectra, 58
  - low\_level\_fusion, 65
  - merge\_datasets, 66
  - msc\_correction, 70
  - num\_samples, 78
  - num\_x\_values, 78
  - read\_data\_spc, 106
  - read\_dataset\_csv, 102
  - read\_dataset\_dx, 103
  - read\_dataset\_spc, 104
  - replace\_data\_value, 119
  - replace\_metadata\_value, 120
  - scaling, 121
  - set\_metadata, 122
  - set\_sample\_names, 123
  - set\_value\_label, 124
  - set\_x\_label, 124
  - set\_x\_values, 125
  - sum\_dataset, 135
- \* **data**
  - get\_data\_values, 39
  - merge\_data\_metadata, 67
  - normalize, 76
  - read\_data\_csv, 105
  - read\_data\_dx, 106
  - remove\_data, 112
  - scaling\_samples, 122
- \* **dendrogram**
  - dendrogram\_plot, 29
  - dendrogram\_plot\_col, 30
- \* **derivative**
  - first\_derivative, 33
- \* **dx**
  - get\_samples\_names\_dx, 48
- \* **equal**
  - find\_equal\_samples, 33
- \* **factor**
  - convert\_to\_factor, 20
  - plotvar\_twofactor, 92
- \* **featureselection**
  - feature\_selection, 31
- \* **file**
  - read\_spc\_nosubhdr, 109
- \* **filters**
  - feature\_selection, 31
  - filter\_feature\_selection, 32
- \* **filter**
  - flat\_pattern\_filter, 34
- \* **flatpattern**
  - flat\_pattern\_filter, 34
- \* **foldchange**
  - fold\_change, 35
  - fold\_change\_var, 36
  - plot\_fold\_change, 93
- \* **fusion**
  - low\_level\_fusion, 65
- \* **ggplot**
  - multiplot, 73
- \* **groups**
  - apply\_by\_groups, 9
- \* **group**
  - apply\_by\_group, 8
- \* **hclust**
  - clustering, 15
  - dendrogram\_plot, 29
  - dendrogram\_plot\_col, 30
  - hierarchical\_clustering, 54
- \* **heatmap**
  - heatmap\_correlations, 53

- \* **hyperspec**
  - convert\_from\_hyperspec, 18
  - convert\_to\_hyperspec, 20
- \* **importance**
  - pca\_importance, 82
- \* **indexes**
  - indexes\_to\_xvalue\_interval, 57
  - x\_values\_to\_indexes, 144
  - xvalue\_interval\_to\_indexes, 143
- \* **interval**
  - subset\_x\_values\_by\_interval, 133
  - xvalue\_interval\_to\_indexes, 143
- \* **idx**
  - read\_data\_dx, 106
  - read\_dataset\_dx, 103
- \* **kendall**
  - correlations\_dataset, 21
- \* **kmeans**
  - clustering, 15
  - kmeans\_clustering, 58
  - kmeans\_plot, 59
  - kmeans\_result\_df, 59
  - pca\_kmeans\_plot2D, 83
  - pca\_kmeans\_plot3D, 84
  - pca\_pairs\_kmeans\_plot, 85
- \* **label**
  - get\_value\_label, 50
  - get\_x\_label, 51
  - set\_value\_label, 124
  - set\_x\_label, 124
- \* **linear**
  - linreg\_all\_vars, 62
  - linregression\_onevar, 61
- \* **loess**
  - smoothing\_interpolation, 126
- \* **log**
  - log\_transform, 64
  - transform\_data, 139
- \* **mass**
  - read\_ms\_spectra, 107
- \* **matrix**
  - get\_data, 37
  - get\_data\_as\_df, 38
  - read\_data\_csv, 105
- \* **mean**
  - mean\_centering, 66
- \* **merge**
  - merge\_datasets, 66
- \* **metabolite**
  - MAIT\_identify\_metabolites, 65
- \* **metadata**
  - convert\_to\_factor, 20
  - get\_metadata, 45
  - get\_metadata\_value, 45
  - get\_metadata\_var, 46
  - merge\_data\_metadata, 67
  - metadata\_as\_variables, 68
  - read\_metadata, 107
  - remove\_metadata\_variables, 114
  - remove\_samples\_by\_na\_metadata, 117
  - replace\_metadata\_value, 120
  - set\_metadata, 122
  - subset\_metadata, 130
  - subset\_samples\_by\_metadata\_values, 132
  - variables\_as\_metadata, 142
- \* **missingvalue**
  - remove\_samples\_by\_nas, 116
  - remove\_variables\_by\_nas, 118
- \* **missing**
  - count\_missing\_values, 23
  - count\_missing\_values\_per\_sample, 24
  - count\_missing\_values\_per\_variable, 24
  - impute\_nas\_knn, 54
  - impute\_nas\_linapprox, 55
  - impute\_nas\_mean, 56
  - impute\_nas\_median, 56
  - impute\_nas\_value, 57
  - missingvalues\_imputation, 69
- \* **model**
  - train\_models\_performance, 138
- \* **msc**
  - msc\_correction, 70
- \* **multifactor**
  - multifactor\_aov\_all\_vars, 71
  - multifactor\_aov\_pvalues\_table, 71
- \* **multiplot**
  - multiplot, 73
- \* **names**
  - get\_samples\_names\_spc, 49
- \* **normalization**
  - normalize, 76
  - normalize\_samples, 77
  - snv\_dataset, 127

- \* **offset**
  - offset\_correction, 79
- \* **pairs**
  - pca\_pairs\_kmeans\_plot, 85
  - pca\_pairs\_plot, 85
- \* **pca**
  - pca\_analysis\_dataset, 80
  - pca\_biplot, 81
  - pca\_biplot3D, 82
  - pca\_importance, 82
  - pca\_kmeans\_plot2D, 83
  - pca\_kmeans\_plot3D, 84
  - pca\_pairs\_plot, 85
  - pca\_plot\_3d, 86
  - pca\_robust, 87
  - pca\_scoresplot2D, 88
  - pca\_scoresplot3D, 89
  - pca\_scoresplot3D\_rgl, 89
  - pca\_screepplot, 90
- \* **peaklist**
  - dataset\_from\_peaks, 28
- \* **peak**
  - get\_peak\_values, 48
  - group\_peaks, 52
  - peaks\_per\_sample, 91
  - peaks\_per\_samples, 91
  - remove\_peaks\_interval, 114
  - remove\_peaks\_interval\_sample\_list, 115
  - values\_per\_peak, 141
- \* **pearson**
  - correlations\_dataset, 21
- \* **plotting**
  - kmeans\_plot, 59
- \* **plot**
  - plot\_anova, 93
  - plot\_fold\_change, 93
  - plot\_kruskaltest, 94
  - plot\_kstest, 95
  - plot\_regression\_coefs\_pvalues, 96
  - plot\_spectra, 97
  - plot\_spectra\_simple, 97
  - plot\_ttests, 98
  - plotvar\_twofactor, 92
  - volcano\_plot\_fc\_tt, 143
- \* **pls**
  - pca\_plot\_3d, 86
- \* **predict**
  - predict\_samples, 99
  - train\_and\_predict, 136
- \* **pvalue**
  - linreg\_pvalue\_table, 63
- \* **random**
  - subset\_random\_samples, 131
- \* **read**
  - read\_csvs\_folder, 102
  - read\_multiple\_csvs, 108
- \* **region**
  - compare\_regions\_by\_sample, 16
- \* **regression**
  - linreg\_all\_vars, 62
  - linreg\_coef\_table, 62
  - linreg\_pvalue\_table, 63
  - linreg\_rsquared, 63
  - linregression\_onevar, 61
  - plot\_regression\_coefs\_pvalues, 96
- \* **remove**
  - remove\_data, 112
  - remove\_data\_variables, 113
  - remove\_metadata\_variables, 114
  - remove\_samples, 116
  - remove\_samples\_by\_na\_metadata, 117
  - remove\_samples\_by\_nas, 116
  - remove\_variables\_by\_nas, 118
  - remove\_x\_values\_by\_interval, 118
- \* **reverse**
  - fold\_change\_var, 36
- \* **rfe**
  - recursive\_feature\_elimination, 111
- \* **robust**
  - pca\_robust, 87
- \* **rsquared**
  - linreg\_rsquared, 63
- \* **samples**
  - num\_samples, 78
  - predict\_samples, 99
- \* **sample**
  - aggregate\_samples, 7
  - apply\_by\_sample, 9
  - compare\_regions\_by\_sample, 16
  - find\_equal\_samples, 33
  - get\_peak\_values, 48
  - get\_sample\_names, 49
  - get\_samples\_names\_dx, 48
  - normalize\_samples, 77
  - peaks\_per\_sample, 91

- peaks\_per\_samples, 91
- remove\_peaks\_interval, 114
- remove\_peaks\_interval\_sample\_list, 115
- remove\_samples, 116
- set\_sample\_names, 123
- stats\_by\_sample, 128
- subset\_by\_samples\_and\_xvalues, 129
- subset\_samples, 131
- values\_per\_peak, 141
- values\_per\_sample, 141
- \* **savitzky-golay**
  - savitzky\_golay, 120
  - smoothing\_interpolation, 126
- \* **sbf**
  - filter\_feature\_selection, 32
- \* **scaling**
  - scaling, 121
  - scaling\_samples, 122
- \* **scoresplot**
  - pca\_scoresplot2D, 88
  - pca\_scoresplot3D, 89
  - pca\_scoresplot3D\_rgl, 89
- \* **screepplot**
  - pca\_screepplot, 90
- \* **shift**
  - shift\_correction, 125
- \* **smoothing**
  - savitzky\_golay, 120
  - smoothing\_interpolation, 126
- \* **snv**
  - snv\_dataset, 127
- \* **spc**
  - get\_samples\_names\_spc, 49
  - read\_data\_spc, 106
  - read\_dataset\_spc, 104
- \* **spearman**
  - correlations\_dataset, 21
- \* **spectral**
  - is\_spectra, 58
- \* **spectra**
  - plot\_spectra, 97
  - plot\_spectra\_simple, 97
- \* **spectrometry**
  - read\_ms\_spectra, 107
- \* **statistic**
  - stats\_by\_sample, 128
  - stats\_by\_variable, 129
- \* **subset**
  - subset\_by\_samples\_and\_xvalues, 129
  - subset\_metadata, 130
  - subset\_random\_samples, 131
  - subset\_samples, 131
  - subset\_samples\_by\_metadata\_values, 132
  - subset\_x\_values, 133
  - subset\_x\_values\_by\_interval, 133
- \* **summaryfunction**
  - multiClassSummary, 70
- \* **summary**
  - sum\_dataset, 135
  - summary\_var\_importance, 134
- \* **test**
  - correlation\_test, 22
  - correlations\_test, 21
- \* **train**
  - train\_and\_predict, 136
  - train\_classifier, 137
  - train\_models\_performance, 138
- \* **transformation**
  - cubic\_root\_transform, 27
  - log\_transform, 64
- \* **transmittance**
  - absorbance\_to\_transmittance, 6
  - transmittance\_to\_absorbance, 140
- \* **tttest**
  - plot\_tttests, 98
  - tTests\_dataset, 140
- \* **tukey**
  - aov\_all\_vars, 7
- \* **type**
  - get\_type, 50
- \* **unsupervised**
  - pca\_analysis\_dataset, 80
- \* **values**
  - count\_missing\_values, 23
  - count\_missing\_values\_per\_sample, 24
  - count\_missing\_values\_per\_variable, 24
  - get\_data\_values, 39
  - impute\_nas\_knn, 54
  - impute\_nas\_linapprox, 55
  - impute\_nas\_mean, 56
  - impute\_nas\_median, 56
  - impute\_nas\_value, 57

- missingvalues\_imputation, 69
- values\_per\_sample, 141
- \* **value**
  - get\_data\_value, 38
  - get\_metadata\_value, 45
  - replace\_data\_value, 119
- \* **varexp**
  - multifactor\_aov\_varexp\_table, 72
- \* **variables**
  - boxplot\_variables, 12
- \* **variable**
  - apply\_by\_variable, 10
  - get\_metadata\_var, 46
  - metadata\_as\_variables, 68
  - remove\_data\_variables, 113
  - stats\_by\_variable, 129
  - variables\_as\_metadata, 142
- \* **vip**
  - summary\_var\_importance, 134
- \* **volcano**
  - volcano\_plot\_fc\_tt, 143
- \* **wrappers**
  - feature\_selection, 31
  - recursive\_feature\_elimination, 111
- \* **xaxis**
  - get\_x\_label, 51
  - get\_x\_values\_as\_num, 51
  - get\_x\_values\_as\_text, 52
  - num\_x\_values, 78
- \* **xvalues**
  - indexes\_to\_xvalue\_interval, 57
  - set\_x\_values, 125
  - subset\_x\_values, 133
  - x\_values\_to\_indexes, 144
- \* **xvalue**
  - remove\_x\_values\_by\_interval, 118
  - subset\_by\_samples\_and\_xvalues, 129
- absorbance\_to\_transmittance, 6
- aggregate\_samples, 7
- aov\_all\_vars, 7
- apply\_by\_group, 8
- apply\_by\_groups, 9
- apply\_by\_sample, 9
- apply\_by\_variable, 10
- background\_correction, 11
- baseline\_correction, 11
- boxplot\_variables, 12
- boxplot\_vars\_factor, 13
- cachexia, 14
- check\_dataset, 14
- clustering, 15
- compare\_regions\_by\_sample, 16
- convert\_chebi\_to\_kegg, 16
- convert\_from\_chemospec, 17
- convert\_from\_hyperspec, 18
- convert\_hmdb\_to\_kegg, 18, 47
- convert\_keggpathway\_2\_reactiongraph, 19
- convert\_multiple\_spcnm\_to\_kegg, 19
- convert\_to\_factor, 20
- convert\_to\_hyperspec, 20
- correlation\_test, 22
- correlations\_dataset, 21
- correlations\_test, 21
- count\_missing\_values, 23
- count\_missing\_values\_per\_sample, 24
- count\_missing\_values\_per\_variable, 24
- create\_dataset, 25
- create\_pathway\_with\_reactions, 26
- cubic\_root\_transform, 27
- data\_correction, 28
- dataset\_from\_peaks, 28
- dendrogram\_plot, 29
- dendrogram\_plot\_col, 30
- detect\_nmr\_peaks\_from\_dataset, 30
- feature\_selection, 31
- filter\_feature\_selection, 32
- find\_equal\_samples, 33
- first\_derivative, 33
- flat\_pattern\_filter, 34
- fold\_change, 35
- fold\_change\_var, 36
- get\_cpd\_names, 36, 47
- get\_data, 37
- get\_data\_as\_df, 38
- get\_data\_value, 38
- get\_data\_values, 39
- get\_files\_list\_per\_assay, 40
- get\_metabolights\_study, 40
- get\_metabolights\_study\_files\_assay, 41, 43
- get\_metabolights\_study\_metadata\_assay, 42, 42

- get\_metabolights\_study\_samples\_files, 43
- get\_MetabolitePath, 44
- get\_metabPaths\_org, 44
- get\_metadata, 45
- get\_metadata\_value, 45
- get\_metadata\_var, 46
- get\_OrganismsCodes, 44, 46, 47
- get\_paths\_with\_cpds\_org, 47
- get\_peak\_values, 48
- get\_sample\_names, 49
- get\_samples\_names\_dx, 48
- get\_samples\_names\_spc, 49
- get\_type, 50
- get\_value\_label, 50
- get\_x\_label, 51
- get\_x\_values\_as\_num, 51
- get\_x\_values\_as\_text, 52
- group\_peaks, 52
  
- heatmap\_correlations, 53
- hierarchical\_clustering, 54
  
- impute\_nas\_knn, 54
- impute\_nas\_linapprox, 55
- impute\_nas\_mean, 56
- impute\_nas\_median, 56
- impute\_nas\_value, 57
- indexes\_to\_xvalue\_interval, 57
- is\_spectra, 58
  
- kmeans\_clustering, 58
- kmeans\_plot, 59
- kmeans\_result\_df, 59
- kruskalTest\_dataset, 60
- ksTest\_dataset, 61
  
- linreg\_all\_vars, 62
- linreg\_coef\_table, 62
- linreg\_pvalue\_table, 63
- linreg\_rsquared, 63
- linregression\_onevar, 61
- log\_transform, 64
- low\_level\_fusion, 65
  
- MAIT\_identify\_metabolites, 65
- mean\_centering, 66
- merge\_data\_metadata, 67
- merge\_datasets, 66
  
- metabolights\_studies\_list, 68
- metadata\_as\_variables, 68
- missingvalues\_imputation, 69
- msc\_correction, 70
- multiClassSummary, 70
- multifactor\_aov\_all\_vars, 71
- multifactor\_aov\_pvalues\_table, 71
- multifactor\_aov\_varexp\_table, 72
- multiplot, 73
  
- nmr\_identification, 74
- normalize, 76
- normalize\_samples, 77
- num\_samples, 78
- num\_x\_values, 78
  
- offset\_correction, 79
  
- pathway\_analysis, 79
- pca\_analysis\_dataset, 80
- pca\_biplot, 81
- pca\_biplot3D, 82
- pca\_importance, 82
- pca\_kmeans\_plot2D, 83
- pca\_kmeans\_plot3D, 84
- pca\_pairs\_kmeans\_plot, 85
- pca\_pairs\_plot, 85
- pca\_plot\_3d, 86
- pca\_robust, 87
- pca\_scoresplot2D, 88
- pca\_scoresplot3D, 89
- pca\_scoresplot3D\_rgl, 89
- pca\_screplot, 90
- peaks\_per\_sample, 91
- peaks\_per\_samples, 91
- plot\_anova, 93
- plot\_fold\_change, 93
- plot\_kruskaltest, 94
- plot\_kstest, 95
- plot\_peaks, 95
- plot\_regression\_coefs\_pvalues, 96
- plot\_spectra, 97
- plot\_spectra\_simple, 97
- plot\_ttests, 98
- plotvar\_twofactor, 92
- predict\_samples, 99
- propolis, 99
- propolisSampleList, 100
  
- read\_Bruker\_files, 101

read\_csvs\_folder, 102  
read\_data\_csv, 105  
read\_data\_dx, 106  
read\_data\_spc, 106  
read\_dataset\_csv, 102  
read\_dataset\_dx, 103  
read\_dataset\_spc, 104  
read\_metadata, 107  
read\_ms\_spectra, 107  
read\_multiple\_csvs, 108  
read\_spc\_nosubhdr, 109  
read\_varian\_spectra\_raw, 110  
recursive\_feature\_elimination, 111  
remove\_data, 112  
remove\_data\_variables, 113  
remove\_metadata\_variables, 114  
remove\_peaks\_interval, 114  
remove\_peaks\_interval\_sample\_list, 115  
remove\_samples, 116  
remove\_samples\_by\_na\_metadata, 117  
remove\_samples\_by\_nas, 116  
remove\_variables\_by\_nas, 118  
remove\_x\_values\_by\_interval, 118  
replace\_data\_value, 119  
replace\_metadata\_value, 120

savitzky\_golay, 120  
scaling, 121  
scaling\_samples, 122  
set\_metadata, 122  
set\_sample\_names, 123  
set\_value\_label, 124  
set\_x\_label, 124  
set\_x\_values, 125  
shift\_correction, 125  
smoothing\_interpolation, 126  
snv\_dataset, 127  
spectra\_options, 127  
stats\_by\_sample, 128  
stats\_by\_variable, 129  
subset\_by\_samples\_and\_xvalues, 129  
subset\_metadata, 130  
subset\_random\_samples, 131  
subset\_samples, 131  
subset\_samples\_by\_metadata\_values, 132  
subset\_x\_values, 133  
subset\_x\_values\_by\_interval, 133  
sum\_dataset, 135  
summary\_var\_importance, 134

textio, 110  
train\_and\_predict, 136  
train\_classifier, 137  
train\_models\_performance, 138  
transform\_data, 139  
transmittance\_to\_absorbance, 140  
tTests\_dataset, 140

values\_per\_peak, 141  
values\_per\_sample, 141  
variables\_as\_metadata, 142  
volcano\_plot\_fc\_tt, 143

x\_values\_to\_indexes, 144  
xvalue\_interval\_to\_indexes, 143