Package 'CytoProfile'

March 3, 2025

March 5, 2023
Title Cytokine Profiling Analysis Tool
Version 1.0
Description CytoProfile is a comprehensive tool for cytokine profiling analysis. It supports quality control using biologically meaningful cutoffs on raw cytokine measurements and tests for distributional symmetry to suggest appropriate transformations. The package offers exploratory data analysis with summary statistics, enhanced boxplots, and barplots, along with both univariate and multivariate analysis capabilities for indepth cytokine profiling.
Encoding UTF-8
Roxygen list(markdown = TRUE)
RoxygenNote 7.3.2
Depends R (>= 3.5)
Imports mixOmics, moments, dplyr, tidyr, pROC, plot3D, tidyverse, caret, xgboost, randomForest
Suggests BiocManager, testthat, knitr
NeedsCompilation no
License GPL-2
LazyData true
Contents
cyt.anova cyt.bp cyt.bp2 cyt.dualflashplot cyt.errbp

2 cyt.anova

	cyt.heatn	nap																			6
	cyt.pca.																				7
	cyt.plsda																				8
	cyt.rf																				10
	cyt.skku																				12
	cyt.ttests																				13
	cyt.volc																				14
	cyt.xgb																				15
	cytodata																				17
Index																					18

cyt.anova

ANOVA analysis on all continuous variables within the data.

Description

This function performs an ANOVA for each continuous variable against every categorical predictor in the input data. For each valid predictor (i.e., with more than one level and no more than 10 levels), it conducts Tukey's HSD test and extracts the adjusted p-values for pairwise comparisons.

Usage

```
cyt.anova(x.df)
```

Arguments

x.df

A data frame or matrix containing both categorical and continuous variables. Character columns are converted to factors; all factor columns are used as predictors, while numeric columns are used as continuous outcomes.

Value

A list of adjusted p-values from Tukey's HSD tests for each combination of continuous outcome and categorical predictor. The list elements are named in the format "outcome_predictor".

Examples

```
## Not run:
data("cytodata")
cyt.anova(cytodata[, c(2:4,5:6)])
## End(Not run)
```

cyt.bp 3

cyt.bp	Generating a PDF file to show the measured value by column of the data frame.

Description

This function creates a PDF file containing box plots for the continuous variables in the provided data. If the number of columns in x.df exceeds bin.size, the function splits the plots across multiple pages.

Usage

```
cyt.bp(x.df, Title, bin.size = 25, mfRow = c(1, 1), yLim = NULL)
```

Arguments

x.df	A matrix or data frame containing the raw data to be plotted.
Title	A string representing the name of the PDF file.
bin.size	The maximum number of box plots to display on a single page.
mfRow	A numeric vector of length two specifying the layout (rows and columns) for the plots on each page.
yLim	An optional numeric vector defining the y-axis limits for the plots.

Value

A PDF file containing the box plots of the continuous variables.

cyt.bp2	Boxplot Function Enhanced for Specific Group Comparisons

Description

This function generates a PDF file containing boxplots for each combination of numeric and factor variables in the provided data. It first converts any character columns to factors and checks that the data contains at least one numeric and one factor column. If the scale argument is set to "log2", all numeric columns are log2-transformed. The function then creates boxplots using ggplot2 for each numeric variable grouped by each factor variable.

Usage

```
cyt.bp2(x.df, Title, mfRow = c(1, 1), scale = NULL, yLim = NULL)
```

4 cyt.dualflashplot

Arguments

x.df	A matrix or data frame of raw data.
Title	A string representing the title of the PDF file.
mfRow	A numeric vector of length two specifying the layout (rows and columns) for the plots on each page. Defaults to $c(1,1)$.
scale	Transformation option for continuous variables. Options are NULL (default) and "log2". When set to "log2", numeric columns are transformed using the log2 function.
yLim	An optional numeric vector defining the y-axis limits for the plots.

Value

A PDF file containing the boxplots.

Examples

```
data.df <- cytodata[,-c(1,4)]
cyt.bp2(data.df, Title = "boxplot2.test2.pdf", scale = "log2")</pre>
```

cyt.dualflashplot

Dual flash plot for group comparisons.

Description

This function reshapes the input data and computes summary statistics (mean and variance) for each variable grouped by a specified factor column. It then calculates the SSMD (Strictly Standardized Mean Difference) and log2 fold change between two groups (group1 and group2) and categorizes the effect strength as "Strong Effect", "Moderate Effect", or "Weak Effect". A dual flash plot is generated using ggplot2 where the x-axis represents the average log2 fold change and the y-axis represents the SSMD. Additionally, the function prints the computed statistics to the console.

Usage

```
cyt.dualflashplot(
  data,
  group_var,
  group1,
  group2,
  ssmd_thresh = 1,
  log2fc_thresh = 1,
  top_labels = 15
)
```

Arguments

data A data frame containing the input data.

group_var A string specifying the name of the grouping column in the data.

group1 A string representing the name of the first group for comparison.

cyt.errbp 5

group2	A string representing the name of the second group for comparison.
ssmd_thresh	A numeric threshold for the SSMD value used to determine significance. Default is 1.
log2fc_thresh	A numeric threshold for the log2 fold change used to determine significance. Default is 1.
top_labels	An integer specifying the number of top variables (based on absolute SSMD) to label in the plot. Default is 15

Value

A ggplot object representing the dual flash plot for the comparisons between group1 and group2.

cyt.errbp

Error-bar plot for comparison.

Description

This function draws an error-bar plot for comparing groups to a baseline group. It creates a barplot of the central tendency (mean or median) and overlays error bars representing the spread (e.g., standard deviation, MAD, or standard error). Optionally, p-value and effect size labels (based on SSMD) are added, either as symbols or numeric values.

Usage

```
cyt.errbp(
  center.df,
  pLab = TRUE,
  esLab = TRUE,
  classSymbol = TRUE,
  xlab = "",
  ylab = "",
  main = ""
)
```

Arguments

center.df A data frame containing the following columns for each group:

• name: Group names.

• center: Mean or median values.

• spread: Standard deviation, MAD, or standard error.

 $\bullet\,$ p.value: P-value for the comparison.

• effect.size: Effect size based on SSMD.

Note: The first row of center.df must correspond to the baseline group.

pLab Logical. Whether to label the p-values on the plot. Default is TRUE. esLab Logical. Whether to label the effect sizes on the plot. Default is TRUE.

classSymbol Logical. Whether to use symbolic notation for significance and effect size. De-

fault is TRUE.

xlab Character. Label for the x-axis. ylab Character. Label for the y-axis. main Character. Title of the graph.

6 cyt.heatmap

Value

An error-bar plot is produced.

Description

This function creates a heatmap using the numeric columns from the provided data frame. If requested via the scale parameter, the function applies a log2 transformation to the data (with non-positive values replaced by NA). Optionally, if an annotation column is specified and exists in data, the function attempts to generate a color annotation (although the annotation is not passed to heatmap.2 in the current implementation). The heatmap is saved as a file, with the format determined by the file extension in title.

Usage

```
cyt.heatmap(data, scale = NULL, annotation_col_name = NULL, title)
```

Arguments

data	A data frame containing the input data. Only numeric columns will be used to generate the heatmap.					
scale	Character. An optional scaling option. If set to "log2", the numeric data will be log2-transformed (with non-positive values set to NA). Default is NULL.					
annotation_col_name						
	Character. An optional column name from data to be used for generating annotation colors. Default is NULL.					
title	Character. The title of the heatmap and the file name for saving the plot. The file extension (".pdf" or ".png") determines the output format.					

Value

The function does not return a value. It saves the heatmap to a file.

Examples

```
cyt.heatmap(cytodata[,-4], scale = "log2", annotation_col_name = "Group")
```

cyt.pca 7

cyt.pca

Analyze data with Principal Component Analysis (PCA) for cytokines.

Description

This function performs Principal Component Analysis (PCA) on cytokine data and generates several types of plots, including:

- 2D PCA plots using mixOmics's plotIndiv function,
- 3D scatter plots (if style is "3d" or "3D" and comp. num is 3) via the plot3D package,
- Scree plots showing both individual and cumulative explained variance,
- · Loadings plots, and
- Biplots and correlation circle plots.

The function optionally applies a log2 transformation to the numeric data and handles analyses based on either treatment or stimulation groups.

Usage

```
cyt.pca(
  x.df,
  group.col = NULL,
  trt.col = NULL,
  colors = NULL,
  title,
  ellipse = FALSE,
  comp.num = 2,
  scale = NULL,
  pch.values = NULL,
  style = NULL
)
```

Arguments

group.col	Character. The name of the column containing the grouping information. If not specified and trt.col is provided, the treatment column will be used as the grouping variable.
trt.col	Character. The name of the column containing the treatment (or stimulation) information. If not specified and group.col is provided, the grouping column will be used as the treatment variable.
colors	A vector of colors corresponding to the groups. If set to NULL, a palette is generated using rainbow() based on the number of unique groups.
title	A string specifying the file name of the PDF where the PCA plots will be saved.
ellipse	Logical. If TRUE, a 95% confidence ellipse is drawn on the PCA individuals plot. Default is FALSE.
comp.num	Numeric. The number of principal components to compute and display. Default is 2.
scale	Character. If set to "log2", a log2 transformation is applied to the numeric cytokine measurements (excluding the grouping columns). Default is NULL.

8 cyt.plsda

pch.values	A vector of plotting symbols (pch values) to be used in the PCA plots. Default is NULL.
style	Character. If set to "3d" or "3D" and comp.num equals 3, a 3D scatter plot is generated using the plot3D package. Default is NULL.
data.df	A data frame containing cytokine data. It should include at least one column representing grouping information and optionally a second column representing treatment or stimulation.

Value

A PDF file containing the PCA plots is generated and saved.

Examples

```
## Not run:
data <- cytodata[,-c(1,4, 24)]
data.df <- filter(data, Group != "ND" & Treatment != "Unstimulated")
cyt.pca(data.df, title = "Example PCA Analysis.pdf" ,colors = c("black", "red2"),
scale = "log2", comp.num = 3, pch.values = c(16,4), style = "3D",
group.col = "Group", trt.col = "Treatment")
cyt.pca(data.df, title = "Example PCA Analysis 2.pdf" ,colors = c("black", "red2"),
scale = "log2", comp.num = 2, pch.values = c(16,4), group.col = "Group")
## End(Not run)</pre>
cyt.plsda

Analyze data with Sparse Partial Least Squares Discriminant Analysis
```

Analyze data with Sparse Partial Least Squares Discriminant Analysis (sPLS-DA).

Description

This function conducts Sparse Partial Least Squares Discriminant Analysis (sPLS-DA) on the provided data. It uses the specified group.col (and optionally trt.col) to define class labels while assuming the remaining columns contain continuous variables. The function supports a log2 transformation via the scale parameter and generates a series of plots, including classification plots, scree plots, loadings plots, and VIP score plots. Optionally, ROC curves are produced when roc is TRUE. Additionally, cross-validation is supported via LOOCV or Mfold methods. When both group.col and trt.col are provided and differ, the function analyzes each treatment level separately.

Usage

```
cyt.plsda(
  x.df,
  group.col = NULL,
  trt.col = NULL,
  colors = NULL,
  title,
  ellipse = FALSE,
  bg = FALSE,
  conf.mat = FALSE,
  var.num,
```

cyt.plsda 9

```
cv.opt = NULL,
fold.num = 5,
scale = NULL,
comp.num = 2,
pch.values,
style = NULL,
roc = FALSE
)
```

Arguments

x.df	A matrix or data frame containing the variables. Columns not specified by group.col or trt.col are assumed to be continuous variables for analysis.
group.col	A string specifying the column name that contains group information. If trt.col is not provided, it will be used for both grouping and treatment.
trt.col	A string specifying the column name for treatments. Default is NULL.
colors	A vector of colors for the groups or treatments. If NULL, a random palette (using rainbow) is generated based on the number of groups.
title	A string specifying the file name for saving the PDF output.
ellipse	Logical. Whether to draw a 95% confidence ellipse on the figures. Default is FALSE.
bg	Logical. Whether to draw the prediction background in the figures. Default is FALSE.
conf.mat	Logical. Whether to print the confusion matrix for the classifications. Default is FALSE.
var.num	Numeric. The number of variables to be used in the PLS-DA model.
cv.opt	Character. Option for cross-validation method: either "loocv" or "Mfold". Default is NULL.
fold.num	Numeric. The number of folds to use if cv.opt is "Mfold". Default is 5.
scale	Character. Option for data transformation; if set to "log2", a log2 transformation is applied to the continuous variables. Default is NULL.
comp.num	Numeric. The number of components to calculate in the sPLS-DA model. Default is 2.
pch.values	A vector of integers specifying the plotting characters to be used in the plots.
style	Character. If set to "3D" or "3d" and comp. num equals 3, a 3D plot is generated using the plot3D package. Default is NULL.
roc	Logical. Whether to compute and plot the ROC curve for the model. Default is FALSE.

Value

A PDF file containing the classification figures, component figures with Variable of Importance in Projection (VIP) scores, and classifications based on VIP scores greater than 1. ROC curves and confusion matrices are also produced if requested.

10 cyt.rf

Examples

cyt.rf

Run Random Forest Classification on Cytokine Data

Description

This function trains and evaluates a Random Forest classification model on cytokine data. It includes feature importance visualization, cross-validation for feature selection, and performance metrics like accuracy, sensitivity, and specificity. Optionally, for binary classification, the function also plots the ROC curve and calculates the AUC.

Usage

```
cyt.rf(
  data,
  group_col,
  ntree = 500,
  mtry = 5,
  train_fraction = 0.7,
  plot_roc = FALSE,
  k_folds = 5,
  step = 0.5,
  run_rfcv = TRUE
)
```

Arguments

data	A data frame containing the cytokine data, with one column as the grouping variable and the rest as numerical features.
group_col	A string representing the name of the column with the grouping variable (i.e., the target variable for classification).
ntree	An integer specifying the number of trees to grow in the forest (default is 500).

cyt.rf 11

An integer specifying the number of variables randomly selected at each split (default is 5).

train_fraction A numeric value between 0 and 1 representing the proportion of data to use for training (default is 0.7).

plot_roc A logical value indicating whether to plot the ROC curve and calculate the AUC for binary classification (default is FALSE).

k_folds An integer specifying the number of folds for cross-validation (default is 5).

step A numeric value specifying the fraction of variables to remove at each step during cross-validation for feature selection (default is 0.5).

run_rfcv A logical value indicating whether to run Random Forest cross-validation for

feature selection (default is TRUE).

Details

The function fits a Random Forest model to the provided data, splitting it into training and test sets. It calculates key performance metrics such as accuracy, sensitivity, and specificity for both the training and test sets. For binary classification tasks, it can also plot the ROC curve and calculate the AUC. If run_rfcv is set to TRUE, Random Forest cross-validation is performed to identify the optimal number of features for classification.

Value

A list containing:

model The trained Random Forest model.

confusion_matrix

The confusion matrix of the test set predictions.

importance_plot

A ggplot object showing the variable importance plot based on Mean Decrease

Gini.

rfcv_result Results from Random Forest cross-validation for feature selection (if run_rfcv

is TRUE).

importance_data

A data frame containing the variable importance based on the Gini index.

Examples

```
# Example usage:
results <- cyt.rf(data = cytodata, group_col = "Group", ntree = 500, plot_roc = TRUE)
print(results$confusion_matrix)
print(results$importance_plot)</pre>
```

12 cyt.skku

cyt.skku

Distribution of the Data Set Shown by Skewness and Kurtosis

Description

This function computes summary statistics—including sample size, mean, standard error, skewness, and kurtosis—for each numeric measurement column in a data set. If grouping columns are provided via group.cols, the function computes the metrics separately for each group defined by the combination of these columns (using the first element as the treatment variable and the second as the grouping variable, or the same column for both if only one is given). When no grouping columns are provided, the entire data set is treated as a single group ("Overall"). A log2 transformation (using a cutoff equal to one-tenth of the smallest positive value in the data) is applied to generate alternative metrics. Histograms showing the distribution of skewness and kurtosis for both raw and log2-transformed data are then generated and saved to a PDF if a file name is provided.

Usage

```
cyt.skku(
  data.df,
  group.cols = NULL,
  Title = NULL,
  printResRaw = FALSE,
  printResLog = FALSE)
```

Arguments

group.cols	A character vector specifying the names of the grouping columns. When provided, the first element is treated as the treatment variable and the second as the group variable. If not provided, the entire data set is treated as one group.
Title	A character string specifying the file name for the PDF file in which the histograms will be saved. If omitted, the plots are generated on the current graphics device.
printResRaw	Logical. If TRUE, the function returns and prints the computed summary statistics for the raw data. Default is FALSE.
printResLog	Logical. If TRUE, the function returns and prints the computed summary statistics for the log2-transformed data. Default is FALSE.
x.df	A matrix or data frame containing the raw data. If group.cols is specified, the columns with names in group.cols are treated as grouping variables and all other columns are assumed to be numeric measurement variables.

Details

A cutoff is computed as one-tenth of the minimum positive value among all numeric measurement columns to avoid taking logarithms of zero. When grouping columns are provided, the function loops over unique treatment levels (using the first element of group.cols as treatment and the second as group, if available) and computes the metrics for each measurement column within each subgroup. Without grouping columns, the entire data set is analyzed as one overall group.

cyt.ttests 13

Value

The function generates histograms of skewness and kurtosis for both raw and log2-transformed data. Additionally, if either printResRaw and/or printResLog is TRUE, the function returns the corresponding summary statistics as a data frame or a list of data frames.

Examples

```
## Not run:
# Example with grouping columns (e.g., "Group" and "Treatment")
data(cytodata)
cyt.skku(cytodata[,-c(1,3,4)], Title = "Skew_and_Kurtosis.pdf", group.cols = c("Group"))
# Example without grouping columns (analyzes the entire data set)
cyt.skku(cytodata[,-c(1,4)], Title = "Skew_and_Kurtosis_Overall.pdf")
## End(Not run)
```

cyt.ttests

Two Sample T-test Comparisons

Description

This function performs pairwise comparisons between two groups for each combination of a categorical predictor (with exactly two levels) and a continuous outcome variable. It first converts any character variables in x.df to factors and applies a log2 transformation to the continuous variables if specified. Depending on the value of scale, the function conducts either a two-sample t-test or a Mann-Whitney U test and prints the resulting p-values. An error is thrown if a categorical variable does not have exactly two levels.

Usage

```
cyt.ttests(x.df, scale = NULL)
```

Arguments

x.df A matrix or data frame containing continuous variables and categorical vari-

ables.

scale A character value specifying a transformation for continuous variables. Options

are NULL (default) and "log2". When scale = "log2", a log2 transformation is applied and a two-sample t-test is used; when scale is NULL, a Mann-Whitney

U test is performed.

Value

A list of p-values from the statistical tests for each combination of continuous outcome and categorical predictor is returned.

14 cyt.volc

Examples

```
## Not run:
data.df <- cytodata[,-c(1,4)]
data.df <- filter(data.df, Group != "ND", Treatment != "Unstimulated")
cyt.ttests(data.df, scale = "log2")
cyt.ttests(data.df)
## End(Not run)</pre>
```

cyt.volc

Volcano Plot

Description

This function subsets the numeric columns from the input data and compares them based on a selected grouping column. It computes the fold changes (as the ratio of means) and associated p-values (using two-sample t-tests) for each numeric variable between two groups. The results are log2-transformed (for fold change) and -log10-transformed (for p-values) to generate a volcano plot.

Usage

```
cyt.volc(
  data,
  group_col,
  cond1 = NULL,
  cond2 = NULL,
  fold_change_thresh = 2,
  p_value_thresh = 0.05,
  top_labels = 10
)
```

Arguments

group_col A character string specifying the column name used for comparisons (e.g., group,

treatment, or stimulation).

cond1 A character string specifying the name of the first condition for comparison.

Default is NULL.

cond2 A character string specifying the name of the second condition for comparison.

Default is NULL.

fold_change_thresh

A numeric threshold for the fold change. Default is 2.

p_value_thresh A numeric threshold for the p-value. Default is 0.05.

top_labels An integer specifying the number of top variables to label on the plot. Default

is 10.

x.df A matrix or data frame containing the data to be analyzed.

cyt.xgb

Value

A list of volcano plots (as ggplot objects) for each pairwise comparison. Additionally, the function prints the data frame used for plotting (excluding the significance column) from the final comparison.

Note

If cond1 and cond2 are not provided, the function automatically generates all possible pairwise combinations of groups from the specified group_col for comparisons.

Examples

```
## Not run:
    cyt.volc(cytodata, group_col = "Group")
    cyt.volc(cytodata, group_col = "Group", fold_change_thresh = 2, top_labels = 15)
## End(Not run)
```

cyt.xgb

Run XGBoost Classification on Cytokine Data

Description

This function trains and evaluates an XGBoost classification model on cytokine data. It allows for hyperparameter tuning, cross-validation, and visualizes feature importance.

Usage

```
cyt.xgb(
 data,
  group_col,
  train_fraction = 0.7,
 nrounds = 500,
 max_depth = 6,
 eta = 0.1,
 nfold = 5,
  cv = FALSE,
 objective = "multi:softprob",
 early_stopping_rounds = NULL,
 eval_metric = "mlogloss",
 gamma = 0,
 colsample_bytree = 1,
  subsample = 1,
 min_child_weight = 1,
  top_n_features = 10,
  verbose = 1,
  plot_roc = FALSE
```

16 cyt.xgb

Arguments

data A data frame containing the cytokine data, with one column as the grouping

variable and the rest as numerical features.

group_col A string representing the name of the column with the grouping variable (i.e.,

the target variable for classification).

train_fraction A numeric value between 0 and 1 representing the proportion of data to use for

training (default is 0.7).

nrounds An integer specifying the number of boosting rounds (default is 500).

max_depth An integer specifying the maximum depth of the trees (default is 6).

eta A numeric value representing the learning rate (default is 0.1).

nfold An integer specifying the number of folds for cross-validation (default is 5).

cv A logical value indicating whether to perform cross-validation (default is FALSE).

objective A string specifying the XGBoost objective function (default is "multi:softprob"

for multi-class classification).

early_stopping_rounds

An integer specifying the number of rounds with no improvement to stop train-

ing early (default is NULL).

eval_metric A string specifying the evaluation metric (default is "mlogloss").

gamma A numeric value for the minimum loss reduction required to make a further

partition (default is 0).

colsample_bytree

A numeric value specifying the subsample ratio of columns when constructing

each tree (default is 1).

subsample A numeric value specifying the subsample ratio of the training instances (default

is 1).

min_child_weight

A numeric value specifying the minimum sum of instance weight needed in a

child (default is 1).

top_n_features An integer specifying the number of top features to display in the importance

plot (default is 10).

verbose An integer specifying the verbosity of the training process (default is 1).

plot_roc A logical value indicating whether to plot the ROC curve and calculate the AUC

for binary classification (default is FALSE).

Details

The function allows for training an XGBoost model on cytokine data, splitting the data into training and test sets. If cross-validation is enabled (cv = TRUE), it performs k-fold cross-validation and prints the best iteration based on the evaluation metric. The function also visualizes the top N important features using xgb.ggplot.importance().

Value

A list containing:

model The trained XGBoost model.

confusion_matrix

The confusion matrix of the test set predictions.

cytodata 17

importance The feature importance matrix for the top features.

class_mapping A named vector showing the mapping from class labels to numeric values used

for training.

cv_results Cross-validation results, if cross-validation was performed (otherwise NULL).

plot A ggplot object showing the feature importance plot.

Examples

```
# Example usage:
results <- cyt.xgb(data = cytodata, group_col = "Group", train_fraction = 0.8, cv = TRUE)
print(results$confusion_matrix)
print(results$plot)</pre>
```

cytodata

Cytokine Profiling Data

Description

Contains observed values of cytokines and their respective treatment and groups.

Usage

cytodata

Format

cytodata:

A data frame with 297 rows and 29 columns:

Group Group assigned to the subjects.

Stimulation Stimulation recieved by subjects.

Treatment Treatment recieved by subjects.

Source

Example data put together for cytokine profiling.

Examples

data(cytodata)

Index

```
*\ datasets
     cytodata, 17
cyt.anova, 2
cyt.bp, 3
cyt.bp2, 3
\verb|cyt.dualflashplot|, 4|\\
cyt.errbp, 5
\operatorname{cyt.heatmap}, 6
cyt.pca, 7
cyt.plsda,8
\texttt{cyt.rf}, \textcolor{red}{10}
cyt.skku, 12
cyt.ttests, 13
\verb"cyt.volc", \frac{14}{}
cyt.xgb, 15
cytodata, 17
```