Package 'midasHLA'

January 28, 2025

Title R package for immunogenomics data handling and association analysis

Version 1.14.0

Description MiDAS is a R package for immunogenetics data transformation and statistical analysis. MiDAS accepts input data in the form of HLA alleles and KIR types, and can transform it into biologically meaningful variables, enabling HLA amino acid fine mapping, analyses of HLA evolutionary divergence, KIR gene presence, as well as validated HLA-KIR interactions. Further, it allows comprehensive statistical association analysis workflows with phenotypes of diverse measurement scales. MiDAS closes a gap between the inference of immunogenetic variation and its efficient utilization to make relevant discoveries related to T cell, Natural Killer cell, and disease biology.

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Encoding UTF-8

LazyData true

Depends R (>= 4.1), MultiAssayExperiment (>= 1.8.3)

Imports assert that (>= 0.2.0), broom (>= 0.5.1), dplyr (>= 0.8.0.1), formattable (>= 0.2.0.1), HardyWeinberg (>= 1.6.3), kableExtra (>= 1.1.0), knitr (>= 1.21), magrittr (>= 1.5), methods, stringi (>= 1.2.4), rlang (>= 0.3.1), S4Vectors (>= 0.20.1), stats, SummarizedExperiment (>= 1.12.0), tibble (>= 2.0.1), utils, qdapTools (>= 1.3.3)

Suggests broom.mixed (>= 0.2.4), cowplot (>= 1.0.0), devtools (>= 2.0.1), ggplot2 (>= 3.1.0), ggpubr (>= 0.2.5), rmarkdown, seqinr (>= 3.4-5), survival (>= 2.43-3), testthat (>= 2.0.1), tidyr (>= 1.1.2)

RoxygenNote 7.1.1

VignetteBuilder knitr

- **Collate** 'asserts.R' 'class.R' 'data.R' 'global.R' 'midasHLA.R' 'parsingFunctions.R' 'stats.R' 'summarise.R' 'transformationFunctions.R' 'utils.R'
- biocViews CellBiology, Genetics, StatisticalMethod

git_url https://git.bioconductor.org/packages/midasHLA

- git_branch RELEASE_3_20
- git_last_commit aa568de

git_last_commit_date 2024-10-29

Contents

Repository Bioconductor 3.20 Date/Publication 2025-01-27 Author Christian Hammer [aut],

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aaVariationToCounts Transform amino acid variation data frame into counts table

Description

aaVariationToCounts convert amino acid variation data frame into counts table.

Usage

aaVariationToCounts(aa_variation)

Arguments

aa_variation Amino acid variation data frame as returned by hlaToAAVariation.

Value

Amino acid counts data frame. First column holds samples ID's, further columns, corresponding to specific amino acid positions, give information on the number of their occurrences in each sample.

adjustPValues Adjust P-values for Multiple Comparisons

Description

Given a set of p-values, returns p-values adjusted using one of several methods.

Usage

```
adjustPValues(p, method, n = length(p))
```

Arguments

р	numeric vector of p-values (possibly with NAs). Any other R object is coerced by as.numeric.
method	correction method. Can be abbreviated.
n	number of comparisons, must be at least length(p); only set this (to non- default) when you know what you are doing! Note that for Bonferroni correction it is possible to specify number lower than length(p).

allele_frequencies

Details

This function modifies stats::p.adjust method such that for Bonferroni correction it is possible to specify n lower than length(p). This feature is useful in cases when knowledge about the biology or redundance of alleles reduces the need for correction.

See p.adjust for more details.

Value

A numeric vector of corrected p-values (of the same length as p, with names copied from p).

allele_frequencies Alleles frequencies scraped from allelefrequencies.net

Description

Accessed on 28.07.20

Usage

allele_frequencies

Format

A data frame with 2096 rows and 3 variables:

var allele number, character

population reference population name, character

frequency allele frequency in reference population, float

Details

A dataset containing allele frequencies across 5697 alleles For details visit the search results page in the allelefrequencies.net database website.

Source

www.allelefrequencies.net

analyzeAssociations Association analysis

Description

analyzeAssociations perform association analysis on a single variable level using a statistical model of choice.

Usage

```
analyzeAssociations(
   object,
   variables,
   placeholder = "term",
   correction = "bonferroni",
   n_correction = NULL,
   exponentiate = FALSE
)
```

Arguments

object	An existing fit from a model function such as lm, glm and many others.
variables	Character vector specifying variables to use in association tests.
placeholder	String specifying term in object's formula which should be substituted with variables during analysis.
correction	String specifying multiple testing correction method. See details for further information.
n_correction	Integer specifying number of comparisons to consider during multiple testing correction calculations. For Bonferroni correction it is possible to specify a number lower than the number of comparisons being made. This is useful in cases when knowledge about the biology or redundance of alleles reduces the need for correction. For other methods it must be at least equal to the number of comparisons being made; only set this (to non-default) when you know what you are doing!
exponentiate	Logical flag indicating whether or not to exponentiate the coefficient estimates. Internally this is passed to tidy. This is typical for logistic and multinomial regressions, but a bad idea if there is no log or logit link. Defaults to FALSE.

Details

correction specifies p-value adjustment method to use, common choice is Benjamini & Hochberg (1995) ("BH"). Internally this is passed to p.adjust.

Value

Tibble containing combined results for all variables. The first column "term" hold the names of variables. Further columns depends on the used model and are determined by associated tidy function. Generally they will include "estimate", "std.error", "statistic", "p.value", "conf.low", "conf.high", "p.adjusted".

Examples

 ${\it analyze Conditional Associations}$

Stepwise conditional association analysis

Description

analyzeConditionalAssociations perform stepwise conditional testing adding the previous topassociated variable as covariate, until there are no more significant variables based on a self-defined threshold.

Usage

```
analyzeConditionalAssociations(
   object,
   variables,
   placeholder = "term",
   correction = "bonferroni",
   n_correction = NULL,
   th,
   th_adj = TRUE,
   keep = FALSE,
   rss_th = 1e-07,
   exponentiate = FALSE
)
```

Arguments

object	An existing fit from a model function such as lm, glm and many others.
variables	Character vector specifying variables to use in association tests.
placeholder	String specifying term to substitute with value from x. Ignored if set to NULL.
correction	String specifying multiple testing correction method. See details for further information.

n_correction	Integer specifying number of comparisons to consider during multiple testing correction calculations. For Bonferroni correction it is possible to specify a number lower than the number of comparisons being made. This is useful in cases when knowledge about the biology or redundance of alleles reduces the need for correction. For other methods it must be at least equal to the number of comparisons being made; only set this (to non-default) when you know what you are doing!
th	Number specifying threshold for a variable to be considered significant.
th_adj	Logical flag indicating if adjusted p-value should be used as threshold criteria, otherwise unadjusted p-value is used.
keep	Logical flag indicating if the output should be a list of results resulting from each selection step. Default is to return only the final result.
rss_th	Number specifying residual sum of squares threshold at which function should stop adding additional variables. As the residual sum of squares approaches 0 the perfect fit is obtained making further attempts at variable selection nonsense. This behavior can be controlled using rss_th.
exponentiate	Logical flag indicating whether or not to exponentiate the coefficient estimates. Internally this is passed to tidy. This is typical for logistic and multinomial regressions, but a bad idea if there is no log or logit link. Defaults to FALSE.

Value

Tibble with stepwise conditional testing results or a list of tibbles, see keep argument. The first column "term" hold the names of variables. Further columns depends on the used model and are determined by associated tidy function. Generally they will include "estimate", "std.error", "statistic", "p.value", "conf.low", "conf.high", "p.adjusted".

Examples

```
# analyzeConditionalAssociations expects model data to be a data.frame
midas_data <- as.data.frame(midas)</pre>
```

applyInheritanceModel Apply inheritance model

Description

Helper function transforming experiment counts to selected inheritance_model.

as.data.frame.MiDAS

Usage

```
applyInheritanceModel(
    experiment,
    inheritance_model = c("dominant", "recessive", "additive", "overdominant")
)
## S3 method for class 'matrix'
applyInheritanceModel(
    experiment,
    inheritance_model = c("dominant", "recessive", "additive", "overdominant")
)
## S3 method for class 'SummarizedExperiment'
applyInheritanceModel(
    experiment,
    inheritance_model = c("dominant", "recessive", "additive", "overdominant")
)
```

Arguments

Details

Under "dominant" model homozygotes and heterozygotes are coded as 1. In "recessive" model homozygotes are coded as 1 and other as 0. In "additive" model homozygotes are coded as 2 and heterozygotes as 1. In "overdominant" homozygotes (both 0 and 2) are coded as 0 and heterozygotes as 1.

Value

experiment converted to specified inheritance model.

as.data.frame.MiDAS Coerce MiDAS to Data Frame

Description

Coerce MiDAS to Data Frame

Usage

```
## S3 method for class 'MiDAS'
as.data.frame(x, ...)
```

Arguments

х	any R object.
	additional arguments to be passed to or from methods.

Value

Data frame representation of MiDAS object. Consecutive columns hold values of variables from MiDAS's experiments and colData. The metadata associated with experiments is not preserved.

backquote Backquote character

Description

backquote places backticks around elements of character vector

Usage

backquote(x)

Arguments

x Character vector.

Details

backquote is useful when using HLA allele numbers in formulas, where '*' and ':' characters have special meanings.

Value

Character vector with its elements backticked.

characterMatches Check if character matches one of possible values

Description

characterMatches checks if all elements of a character vector matches values in choices.

Usage

```
characterMatches(x, choice)
```

Arguments

х	Character vector to test.
choice	Character vector with possible values for x.

Value

Logical indicating if x's elements matches any of the values in choice.

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checkAlleleFormat Check HLA allele format

Description

checkAlleleFormat test if the input character follows HLA nomenclature specifications.

Usage

```
checkAlleleFormat(allele)
```

Arguments

```
allele Character vector with HLA allele numbers.
```

Details

Correct HLA number should consist of HLA gene name followed by "*" and sets of digits separated with ":". Maximum number of sets of digits is 4 which is termed 8-digit resolution. Optionally HLA numbers can be supplemented with additional suffix indicating its expression status. See http://hla.alleles.org/nomenclature/naming.html for more details.

HLA alleles with identical sequences across exons encoding the peptide binding domains might be designated with G group allele numbers. Those numbers have additional G or GG suffix. See http://hla.alleles.org/alleles/g_groups.html for more details. They are interpreted as valid HLA alleles designations.

Value

Logical vector specifying if allele elements follows HLA alleles naming conventions.

Examples

```
allele <- c("A*01:01", "A*01:02")
checkAlleleFormat(allele)</pre>
```

checkColDataFormat Assert colData data

Description

checkColDataFormat asserts if the colData data frame has proper format.

Usage

```
checkColDataFormat(data_frame)
```

Arguments

data_frame Data frame containing colData data used to construct MiDAS object.

Value

Logical indicating if data_frame is properly formatted. Otherwise raise an error.

checkHlaCallsFormat Assert hla calls data frame format

Description

checkHlaCallsFormat asserts if hla calls data frame have proper format.

Usage

```
checkHlaCallsFormat(hla_calls)
```

Arguments

hla_calls HLA calls data frame, as returned by readHlaCalls function.

Value

Logical indicating if hla_calls follows hla calls data frame format. Otherwise raise an error.

checkKirCallsFormat Assert KIR counts data frame format

Description

checkKirCallsFormat asserts if KIR counts data frame have proper format.

Usage

```
checkKirCallsFormat(kir_calls)
```

Arguments

kir_calls KIR calls data frame, as returned by readKirCalls function.

Value

Logical indicating if kir_calls follow KIR counts data frame format. Otherwise raise an error.

checkKirGenesFormat Check KIR genes format

Description

checkKirGenesFormat test if the input character follows KIR gene names naming conventions.

Usage

checkKirGenesFormat(genes)

Arguments

genes

Character vector with KIR gene names.

Details

KIR genes: "KIR3DL3", "KIR2DS2", "KIR2DL2", "KIR2DL3", "KIR2DP1", "KIR2DL1", "KIR3DP1", "KIR2DL1", "KIR3DP1", "KIR2DL4", "KIR3DL1", "KIR3DS1", "KIR2DL5", "KIR2DS3", "KIR2DS5", "KIR2DS4", "KIR2DS1", "KIR3DL2".

Value

Logical vector specifying if genes elements follow KIR genes naming conventions.

Examples

checkKirGenesFormat(c("KIR3DL3", "KIR2DS2", "KIR2DL2"))

checkStatisticalModel Assert statistical model

Description

checkStatisticalModel asserts if object is an existing fit from a model functions such as lm, glm and many others. Containing MiDAS object as its data atribute.

Usage

```
checkStatisticalModel(object)
```

Arguments

object An existing fit from a model function such as lm, glm and many others.

Value

Logical indicating if object is an existing fit from a model functions such as lm, glm and many others. Containing MiDAS object as its data attribute. Otherwise raise an error.

colnamesMatches Check column names

Description

colnamesMatches check if data frame's columns are named as specified

Usage

```
colnamesMatches(x, cols)
```

Arguments

х	Data frame to test.
cols	Ordered character vector to test against x's colnames.

Value

Logical indicating if x's colnames equals choice.

```
convertAlleleToVariable
```

Convert allele numbers to additional variables

Description

convertAlleleToVariable converts input HLA allele numbers to additional variables based on the supplied dictionary.

Usage

```
convertAlleleToVariable(allele, dictionary)
```

Arguments

allele	Character vector with HLA allele numbers.
dictionary	Path to file containing HLA allele dictionary or a data frame.

Details

dictionary file should be a tsv format with header and two columns. First column should hold allele numbers, second additional variables (eg. expression level).

Type of the returned vector depends on the type of the additional variable.

Value

Vector containing HLA allele numbers converted to additional variables according to dictionary.

countsToVariables

Examples

```
dictionary <- system.file("extdata", "Match_allele_HLA_supertype.txt", package = "midasHLA")
convertAlleleToVariable(c("A*01:01", "A*02:01"), dictionary = dictionary)</pre>
```

countsToVariables Convert counts table to variables

Description

countsToVariables converts counts table to additional variables.

Usage

```
countsToVariables(counts, dictionary, na.value = NA, nacols.rm = TRUE)
```

Arguments

counts	Data frame with counts, such as returned by hlaCallsToCounts function. First column should contain samples IDs, following columns should contain counts (natural numbers including zero).
dictionary	Path to file containing variables dictionary or data frame. See details for further explanations.
na.value	Vector of length one speciyfing value for variables with no matching entry in dictionary. Default is to use 0.
nacols.rm	Logical indicating if result columns that contain only NA should be removed.

Details

dictionary file should be a tsv format with header and two columns. First column should be named "Name" and hold variable name, second should be named "Expression" and hold expression used to identify variable (eg. "KIR2DL3 & ! KIR2DL2" will match all samples with KIR2DL3 and without KIR2DL2). Optionally a data frame formatted in the same manner can be passed instead.

Dictionaries shipped with the package:

kir_haplotypes KIR genes to KIR haplotypes dictionary.

Value

Data frame with variable number of columns. First column named "ID" corresponds to "ID" column in counts, further columns hold indicators for converted variables. 1 and 0 code presence and absence of a variable respectively.

Examples

```
countsToVariables(MiDAS_tut_KIR, "kir_haplotypes")
```

dfToExperimentMat Helpe

Description

Function deletes 'ID' column from a df, then transpose it and sets the column names to values from deleted 'ID' column.

Usage

dfToExperimentMat(df)

Arguments

df Data frame

Value

Matrix representation of df.

dict_dist_grantham Grantham distance

Description

Integer vector giving Grantham distance values between pairs of amino acid residues.

Usage

dict_dist_grantham

Format

Named integer vector of length 400.

distGrantham

Description

distGrantham calculates normalized Grantham distance between two amino acid sequences. For details on calculations see Grantham R. 1974..

Usage

```
distGrantham(aa1, aa2)
```

Arguments

aa1	Character vector giving amino acid sequence using one letter codings. element must correspond to single amino acid.	Each
aa2	Character vector giving amino acid sequence using one letter codings. element must correspond to single amino acid.	Each

Details

Distance between amino acid sequences is normalized by length of compared sequences. Lengths of aa1 and aa2 must be equal.

Value

Numeric vector of normalized Grantham distance between aa1 and aa2.

experimentMatToDf Helper transform experiment matrix to data frame

Description

Function transpose mat and inserts column names of input mat as a 'ID' column.

Usage

```
experimentMatToDf(mat)
```

Arguments

mat Matrix

Value

Data frame representation of mat.

filterByFrequency Filter MiDAS object by frequency

Description

Filter MiDAS object by frequency

Usage

```
filterByFrequency(
   object,
   experiment,
   lower_frequency_cutoff = NULL,
   upper_frequency_cutoff = NULL,
   carrier_frequency = FALSE
)
```

Arguments

object	MiDAS object.
experiment	String specifying experiment.
lower_frequency	_cutoff
	Number giving lower frequency threshold. Numbers greater than 1 are interpreted as the number of feature occurrences, numbers between 0 and 1 as fractions.
upper_frequency	cutoff
	Number giving upper frequency threshold. Numbers greater than 1 are interpreted as the number of feature occurrences, numbers between 0 and 1 as fractions.
carrier_frequen	су
	Logical flog indicating if coming fragments should be noticed

Logical flag indicating if carrier frequency should be returned.

Value

Filtered MiDAS object.

Examples

filterByOmnibusGroups Filter MiDAS object by omnibus groups

Description

Filter MiDAS object by omnibus groups

Usage

```
filterByOmnibusGroups(object, experiment, groups)
```

Arguments

object	MiDAS object.
experiment	String specifying experiment.
groups	Character vector specifying omnibus groups to select. See getOmnibusGroups for more details.

Value

Filtered MiDAS object.

Examples

filterByVariables Filter MiDAS object by features

Description

Filter MiDAS object by features

Usage

```
filterByVariables(object, experiment, variables)
```

Arguments

object	MiDAS object.
experiment	String specifying experiment.
variables	Character vector specifying features to select.

Value

Filtered MiDAS object.

Examples

filterExperimentByFrequency

Filter experiment by frequency

Description

Helper function for experiments filtering

Usage

```
filterExperimentByFrequency(
  experiment,
  carrier_frequency = FALSE,
  lower_frequency_cutoff = NULL,
  upper_frequency_cutoff = NULL
)
## S3 method for class 'matrix'
filterExperimentByFrequency(
  experiment,
  carrier_frequency = FALSE,
  lower_frequency_cutoff = NULL,
  upper_frequency_cutoff = NULL
)
## S3 method for class 'SummarizedExperiment'
filterExperimentByFrequency(
  experiment,
  carrier_frequency = FALSE,
  lower_frequency_cutoff = NULL,
  upper_frequency_cutoff = NULL
)
```

Arguments

```
experiment Matrix or SummarizedExperiment object.
carrier_frequency
Logical flag indicating if carrier frequency should be returned.
lower_frequency_cutoff
Positive number or NULL. Numbers greater than 1 are interpreted as number of
frequency_cutoff
Positive number or NULL. Numbers greater than 1 are interpreted as number of
feature occurrences, numbers between 0 and 1 as fractions.
```

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Value

Filtered experiment matrix.

filterExperimentByVariables

Filter experiment by variable

Description

Helper function for experiments filtering

Usage

```
filterExperimentByVariables(experiment, variables)
```

```
## S3 method for class 'matrix'
filterExperimentByVariables(experiment, variables)
```

```
## S3 method for class 'SummarizedExperiment'
filterExperimentByVariables(experiment, variables)
```

Arguments

experiment	Matrix or SummarizedExperiment object.
variables	Character vector specifying features to choose.

Value

Filtered experiment object.

filterListByElements Filter list by elements

Description

Filter two level list by its secondary elements and remove empty items

Usage

```
filterListByElements(list, elements)
```

Arguments

list	A list.
elements	Character vector of elements to keep.

Value

List filtered according to elements argument.

formatResults

Description

formatResults format statistical analysis results table to html or latex format.

Usage

```
formatResults(
  results,
  filter_by = "p.value <= 0.05",
  arrange_by = "p.value",
  select_cols = c("term", "estimate", "std.error", "p.value", "p.adjusted"),
  format = c("html", "latex"),
  header = NULL,
  scroll_box_height = "400px"
)</pre>
```

Arguments

results	Tibble as returned by runMiDAS.
filter_by	Character vector specifying conditional expression used to filter results, this is equivalent to argument passed to filter.
arrange_by	Character vector specifying variable names to use for sorting. Equivalent to argument passed to arrange.
<pre>select_cols</pre>	Character vector specifying variable names that should be included in the output table. Can be also used to rename selected variables, see examples.
format	String "latex" or "html".
header	String specifying header for result table. If NULL no header is added.
scroll_box_height	
	A character string indicating the height of the table.

Value

Character vector of formatted table source code.

Examples

getAAFrequencies

```
format = "html",
header = "HLA allelic associations")
```

End(Not run)

getAAFrequencies Calculate amino acid frequencies

Description

getAAFrequencies calculates amino acid frequencies in amino acid data frame.

Usage

getAAFrequencies(aa_variation)

Arguments

aa_variation Amino acid variation data frame as returned by hlaToAAVariation.

Details

Both gene copies are taken into consideration for frequencies calculation, frequency = n / (2 * j) where n is the number of amino acid occurrences and j is the number of samples in aa_variation.

Value

Data frame with each row holding specific amino acid position, it's count and frequency.

Examples

```
aa_variation <- hlaToAAVariation(MiDAS_tut_HLA)
getAAFrequencies(aa_variation)</pre>
```

getAlleleResolution Infer HLA allele resolution

Description

getAlleleResolution returns the resolution of input HLA allele numbers.

Usage

```
getAlleleResolution(allele)
```

Arguments

allele Character vector with HLA allele numbers.

Details

HLA allele resolution can take the following values: 2, 4, 6, 8. See http://hla.alleles.org/ nomenclature/naming.html for more details.

NA values are accepted and returned as NA.

Value

Integer vector specifying allele resolutions.

Examples

```
allele <- c("A*01:01", "A*01:02")
getAlleleResolution(allele)</pre>
```

Description

List HLA alleles and amino acid residues at a given position.

Usage

```
getAllelesForAA(object, aa_pos)
```

Arguments

object	MiDAS object.
aa_pos	String specifying gene and amino acid position, example "A_9".

Value

Data frame containing HLA alleles, their corresponding amino acid residues and frequencies at requested position.

Examples

```
getAllelesForAA(object = MiDAS_tut_object, aa_pos = "A_9")
```

getExperimentFrequencies

Description

getExperimentFrequencies calculate features frequencies.

Usage

```
getExperimentFrequencies(
  experiment,
  pop_mul = NULL,
  carrier_frequency = FALSE,
  ref = NULL
)
## S3 method for class 'matrix'
getExperimentFrequencies(
  experiment,
  pop_mul = NULL,
  carrier_frequency = FALSE,
  ref = NULL
)
## S3 method for class 'SummarizedExperiment'
getExperimentFrequencies(
  experiment,
  pop_mul = NULL,
  carrier_frequency = FALSE,
  ref = NULL
)
```

Arguments

experiment	Matrix or SummarizedExperiment object.
pop_mul	Number by which number of samples should be multiplied to get the population size.
carrier_frequency	
	Logical flag indicating if carrier frequency should be returned.
ref	Wide format data frame with first column named "var" holding features match- ing experiment and specific populations frequencies in following columns. See getReferenceFrequencies for more details.

Value

Data frame with each row holding specific variable, it's count and frequency.

```
{\tt get} {\tt Experiment} {\tt Population} {\tt Multiplicator}
```

Get experiment's population multiplicator

Description

getExperimentPopulationMultiplicator extracts population multiplicator from experiment's metadata.

Usage

```
getExperimentPopulationMultiplicator(experiment)
```

S3 method for class 'matrix'
getExperimentPopulationMultiplicator(experiment)

S3 method for class 'SummarizedExperiment'
getExperimentPopulationMultiplicator(experiment)

Arguments

experiment Matrix or SummarizedExperiment object.

Value

Experiment's population multiplicator number.

getExperiments Get available experiments in MiDAS object.

Description

Get available experiments in MiDAS object.

Usage

```
getExperiments(object)
```

Arguments

object MiDAS object.

Value

Character vector giving names of experiments in object.

Examples

getExperiments(object = MiDAS_tut_object)

getFrequencies

Description

Calculate features frequencies for a given experiment in MiDAS object.

Usage

```
getFrequencies(
    object,
    experiment,
    carrier_frequency = FALSE,
    compare = FALSE,
    ref_pop = list(hla_alleles = c("USA NMDP African American pop 2", "USA NMDP Chinese",
        "USA NMDP European Caucasian", "USA NMDP Hispanic South or Central American",
        "USA NMDP Japanese", "USA NMDP North American Amerindian",
        "USA NMDP South Asian Indian"), kir_genes = c("USA California African American KIR",
        "USA California Asian American KIR", "USA California Caucasians KIR",
        "USA California Hispanic KIR")),
    ref = list(hla_alleles = allele_frequencies, kir_genes = kir_frequencies)
)
```

Arguments

object	MiDAS object.
experiment	Matrix or SummarizedExperiment object.
carrier_freque	ncy
	Logical flag indicating if carrier frequency should be returned.
compare	Logical flag indicating if hla_calls frequencies should be compared to reference frequencies given in ref.
ref_pop	Named list of character vectors giving names of reference populations in ref to compare with. Optionally vectors can be named, then those names will be used as population names. Each vector should correspond to a specific experiment.
ref	Named list of reference frequencies data frames. Each element should give reference for a specific experiment. See allele_frequencies for an example on how reference frequency data frame should be formatted.

Value

Data frame with features from selected experiment and their corresponding frequencies. Column "term" hold features names, "Counts" hold number of feature occurrences, "Freq" hold feature frequencies. If argument compare is set to TRUE, further columns will hold frequencies in reference populations.

Examples

getFrequencyMask Helper function for filtering frequency data frame

Description

Helper function for filtering frequency data frame

Usage

```
getFrequencyMask(
   df,
   lower_frequency_cutoff = NULL,
   upper_frequency_cutoff = NULL
)
```

Arguments

df

Data frame as returned by getExperimentFrequencies.

lower_frequency_cutoff

Positive number or NULL. Numbers greater than 1 are interpreted as number of feature occurrences, numbers between 0 and 1 as fractions.

```
upper_frequency_cutoff
```

Positive number or NULL. Numbers greater than 1 are interpreted as number of feature occurrences, numbers between 0 and 1 as fractions.

Value

Character vector containing names of variables after filtration.

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getHlaCalls

Description

Get HLA calls from MiDAS object.

Usage

```
getHlaCalls(object)
```

Arguments

object MiDAS object.

Value

HLA calls data frame.

Examples

getHlaCalls(object = MiDAS_tut_object)

getHlaCallsGenes Get HLA calls genes

Description

getHlaCallsGenes get's genes found in HLA calls.

Usage

```
getHlaCallsGenes(hla_calls)
```

Arguments

hla_calls HLA calls data frame, as returned by readHlaCalls function.

Value

Character vector of genes in hla_calls.

getHlaFrequencies Calculate HLA allele frequencies

Description

getHlaFrequencies calculates allele frequencies in HLA calls data frame.

Usage

```
getHlaFrequencies(
  hla_calls,
  carrier_frequency = FALSE,
  compare = FALSE,
  ref_pop = c("USA NMDP African American pop 2", "USA NMDP Chinese",
    "USA NMDP European Caucasian", "USA NMDP Hispanic South or Central American",
    "USA NMDP Japanese", "USA NMDP North American Amerindian",
    "USA NMDP South Asian Indian"),
    ref = allele_frequencies
)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
carrier_frequency	
	Logical flag indicating if carrier frequency should be returned.
compare	Logical flag indicating if hla_calls frequencies should be compared to reference frequencies given in ref.
ref_pop	Character vector giving names of reference populations in ref to compare with. Optionally vector can be named, then those names will be used as population names.
ref	Data frame giving reference allele frequencies. See allele_frequencies for an example.

Details

Both gene copies are taken into consideration for frequencies calculation, frequency = n / (2 * j) where n is the number of allele occurrences and j is the number of samples in hla_calls.

Value

Data frame with each row holding HLA allele, it's count and frequency.

Examples

getHlaFrequencies(MiDAS_tut_HLA)

getHlaKirInteractions Get HLA - KIR interactions

Description

getHlaKirInteractions calculate presence-absence matrix of HLA - KIR interactions.

Usage

```
getHlaKirInteractions(
    hla_calls,
    kir_calls,
    interactions_dict = system.file("extdata", "Match_counts_hla_kir_interactions.txt",
        package = "midasHLA")
)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.	
kir_calls	KIR calls data frame, as returned by readKirCalls function.	
interactions_dict		
	Path to HLA - KIR interactions dictionary.	

Details

hla_calls are first reduced to all possible resolutions and converted to additional variables, such as G groups, using dictionaries shipped with the package.

interactions_dict file should be a tsv format with header and two columns. First column should be named "Name" and hold interactions names, second should be named "Expression" and hold expression used to identify interaction (eg. "C2 & KIR2DL1" will match all samples with C2 and KIR2DL1). The package is shipped with an interactions file based on Pende et al., 2019.

Value

Data frame with variable number of columns. First column named "ID" corresponds to "ID" column in counts, further columns hold indicators for HLA - KIR interactions. 1 and 0 code presence and absence of a variable respectively.

Examples

```
getHlaKirInteractions(
  hla_calls = MiDAS_tut_HLA,
  kir_calls = MiDAS_tut_KIR,
  interactions_dict = system.file(
    "extdata", "Match_counts_hla_kir_interactions.txt",
    package = "midasHLA")
)
```

getKirCalls

Description

Get KIR calls from MiDAS object.

Usage

```
getKirCalls(object)
```

Arguments

object MiDAS object.

Value

KIR calls data frame.

Examples

getKirCalls(object = MiDAS_tut_object)

getKIRFrequencies Calculate KIR genes frequencies

Description

getKIRFrequencies calculates KIR genes frequencies in KIR calls data frame.

Usage

```
getKIRFrequencies(kir_calls)
```

Arguments

kir_calls KIR calls data frame, as returned by readKirCalls function.

Value

Data frame with each row holding KIR gene, it's count and frequency.

Examples

getKIRFrequencies(MiDAS_tut_KIR)

getObjectDetails Get attributes of statistical model object

Description

getObjectDetails extracts some of the statistical model object attributes that are needed for runMiDAS internal calculations.

Usage

```
getObjectDetails(object)
```

Arguments

object An existing fit from a model function such as lm, glm and many others.

Value

List with following elements:

call Object's call

formula_vars Character containing names of variables in object formula

data MiDAS object associated with model

getOmnibusGroups Get omnibus groups from MiDAS object.

Description

Get omnibus groups from MiDAS object.

Usage

```
getOmnibusGroups(object, experiment)
```

Arguments

object	MiDAS object.
experiment	String specifying experiment.

Details

For some experiments features can be naturally divided into groups (here called omnibus groups). For example, in "hla_aa" experiment features can be grouped by amino acid position ("B_46_E", "B_46_A") can be grouped into B_46 group). Such groups can be then used to perform omnibus test, see runMiDAS for more details.

Value

List of omnibus groups for a given experiment.

Examples

getPlaceholder Get placeholder name from MiDAS object.

Description

Get placeholder name from MiDAS object.

Usage

getPlaceholder(object)

Arguments

object MiDAS object.

Value

String giving name of placeholder.

Examples

getPlaceholder(object = MiDAS_tut_object)

getReferenceFrequencies

Helper transforming reference frequencies

Description

Helper transforming reference frequencies

Usage

```
getReferenceFrequencies(ref, pop, carrier_frequency = FALSE)
```

Arguments

ref	Long format data frame with three columns "var", "population", "frequency".
рор	Character giving names of populations to include
carrier_frequency	
	Logical indicating if carrier frequency should be returned instead of frequency.
	Carrier frequency is calculated based on Hardy-Weinberg equilibrium model.

Value

Wide format data frame with population frequencies as columns.

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getVariableAAPos Find variable positions in sequence alignment

Description

getVariableAAPos finds variable amino acid positions in protein sequence alignment.

Usage

```
getVariableAAPos(alignment, varchar = "[A-Z]")
```

Arguments

alignment	Matrix containing amino acid level alignment, as returned by readHlaAlignments,
varchar	Regex matching characters that should be considered when looking for variable
	amino acid positions. See details for further explanations.

Details

The variable amino acid positions in the alignment are those at which different amino acids can be found. As the alignments can also contain indels and unknown characters, the user choice might be to consider those positions as variable or not. This can be achieved by passing appropriate regular expression in varchar. Eg. when varchar = "[A-Z]" occurence of deletion/insertion (".") will not be treated as variability. In order to detect this kind of variability varchar = "[A-Z]" should be used.

Value

Integer vector specifying which alignment columns are variable.

Examples

```
alignment <- readHlaAlignments(gene = "TAP1")
getVariableAAPos(alignment)</pre>
```

hasTidyMethod Check if tidy method for class exist

Description

hasTidyMethod check if there is a tidy method available for a given class.

Usage

```
hasTidyMethod(class)
```

Arguments

class String giving object class.

Value

Logical indicating if there is a tidy method for a given class.

hlaAlignmentGrantham Helper function returning alignment for Grantham distance calculations

Description

Helper function returning alignment for Grantham distance calculations

Usage

hlaAlignmentGrantham(gene, aa_sel = 2:182)

Arguments

gene	Character vector specifying HLA gene.
aa_sel	Numeric vector specifying amino acids that should be extracted.

Value

HLA alignment processed for grantham distance calculation. Processing includes extracting specific amino acids, masking indels, gaps and stop codons.

```
hlaCallsGranthamDistance
```

Calculate Grantham distance between HLA alleles

Description

hlaCallsGranthamDistance calculate Grantham distance between two HLA alleles of a given, using original formula by Grantham R. 1974.

Usage

```
hlaCallsGranthamDistance(
    hla_calls,
    genes = c("A", "B", "C"),
    aa_selection = "binding_groove"
)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
genes	Character vector specifying genes for which allelic distance should be calculated.
aa_selection	String specifying variable region in peptide binding groove which should be con- sidered for Grantham distance calculation. Valid choices includes: "binding_groove", "B_pocket", "F_pocket". See details for more information.

hlaCallsToCounts

Details

Grantham distance is calculated only for class I HLA alleles. First exons forming the variable region in the peptide binding groove are selected. Here we provide option to choose either "binding_groove" - exon 2 and 3 (positions 1-182 in IMGT/HLA alignments, however here we take 2-182 as many 1st positions are missing), "B_pocket" - residues 7, 9, 24, 25, 34, 45, 63, 66, 67, 70, 99 and "F_pocket" - residues 77, 80, 81, 84, 95, 116, 123, 143, 146, 147. Then all the alleles containing gaps, stop codons or indels are discarded. Finally distance is calculated for each pair.

See Robinson J. 2017. for more details on the choice of exons 2 and 3.

Value

Data frame of normalized Grantham distances between pairs of alleles for each specified HLA gene. First column (ID) is the same as in hla_calls, further columns are named as given by genes.

Examples

hlaCallsGranthamDistance(MiDAS_tut_HLA, genes = "A")

hlaCallsToCounts Transform HLA calls to counts table

Description

hlaCallsToCounts converts HLA calls data frame into a counts table.

Usage

hlaCallsToCounts(hla_calls, check_hla_format = TRUE)

Arguments

hla_calls HLA calls data frame, as returned by readHlaCalls function.

check_hla_format

Logical indicating if hla_calls format should be checked. This is useful if one wants to use hlaCallsToCounts with input not adhering to HLA nomenclature standards. See examples.

Value

HLA allele counts data frame. First column holds samples ID's, further columns, corresponding to specific alleles, give information on the number of their occurrences in each sample.

hlaToAAVariation

Description

hlaToAAVariation convert HLA calls data frame to a matrix of variable amino acid positions.

Usage

```
hlaToAAVariation(hla_calls, indels = TRUE, unkchar = FALSE, as_df = TRUE)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
indels	Logical indicating whether indels should be considered when checking variabil- ity.
unkchar	Logical indicating whether unknown characters in the alignment should be con- sidered when checking variability.
as_df	Logical indicating if data frame should be returned. Otherwise a matrix is re- turned.

Details

Variable amino acid positions are found by comparing elements of the alignment column wise. Some of the values in alignment can be treated specially using indels and unkchar arguments. Function processes alignments for all HLA genes found in hla_calls.

Variable amino acid position uses protein alignments from EBI database.

Value

Matrix or data frame containing variable amino acid positions. Rownames corresponds to ID column in hla_calls, and colnames to alignment positions. If no variation is found one column matrix filled with NA's is returned.

Examples

hlaToAAVariation(MiDAS_tut_HLA)

hlaToVariable

Convert HLA calls to variables

Description

hlaToVariable converts HLA calls data frame to additional variables.

hlaToVariable

Usage

```
hlaToVariable(
    hla_calls,
    dictionary,
    reduce = TRUE,
    na.value = 0,
    nacols.rm = TRUE
)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
dictionary	Path to file containing HLA allele dictionary or a data frame.
reduce	Logical indicating if function should try to reduce allele resolution when no matching entry in the dictionary is found. See details.
na.value	Vector of length one speciyfing value for alleles with no matching entry in dictionary. Default is to use 0 .
nacols.rm	Logical indicating if result columns that contain only NA should be removed.

Details

dictionary file should be a tsv format with header and two columns. First column should hold allele numbers and second corresponding additional variables. Optionally a data frame formatted in the same manner can be passed instead.

dictionary can be also used to access dictionaries shipped with the package. They can be referred to by using one of the following strings:

- "allele_HLA_Bw" Translates HLA-B alleles together with A*23, A*24 and A*32 into Bw4 and Bw6 allele groups. In some cases HLA alleles containing Bw4 epitope, on nucleotide level actually carries a premature stop codon. Meaning that although on nucleotide level the allele would encode a Bw4 epitope it's not really there and it is assigned to Bw6 group. However in 4-digit resolution these alleles can not be distinguished from other Bw4 groups. Since alleles with premature stop codons are rare, Bw4 group is assigned.
- "allele_HLA-B_only_Bw" Translates HLA-B alleles (without A*23, A*24 and A*32) into Bw4 and Bw6 allele groups.
- "allele_HLA-C_C1-2" Translates HLA-C alleles into C1 and C2 allele groups.
- "allele_HLA_supertype" Translates HLA-A and HLA-B alleles into supertypes, a classification that group HLA alleles based on peptide binding specificities.
- "allele_HLA_Ggroup" Translates HLA alleles into G groups, which defines amino acid identity only in the exons relevant for peptide binding. Note that alleles DRB1*01:01:01 and DRB1*01:16 match more than one G group, here this ambiguity was removed by deleting matching with DRB5*01:01:01G group.

reduce control if conversion should happen in a greedy way, such that if some HLA number cannot be converted, it's resolution is reduced by 2 and another attempt is taken. This process stops when alleles cannot be further reduced or all have been successfully converted.

Value

Data frame with variable number of columns. First column named "ID" corresponds to "ID" column in hla_calls, further columns holds converted HLA variables.

Examples

hlaToVariable(MiDAS_tut_HLA, dictionary = "allele_HLA_supertype")

HWETest

Test for Hardy Weinberg equilibrium

Description

Test experiment features for Hardy Weinberg equilibrium.

Usage

```
HWETest(
   object,
   experiment = c("hla_alleles", "hla_aa", "hla_g_groups", "hla_supertypes",
        "hla_NK_ligands"),
   HWE_group = NULL,
   HWE_cutoff = NULL,
   as.MiDAS = FALSE
)
```

Arguments

object	MiDAS object.
experiment	String specifying experiment to test. Valid values includes "hla_alleles", "hla_aa", "hla_g_groups", "hla_supertypes", "hla_NK_ligands".
HWE_group	Expression defining samples grouping to test for Hardy Weinberg equilibrium. By default samples are not grouped.
HWE_cutoff	Number specifying p-value threshold. When HWE_group is specified both groups are thresholded.
as.MiDAS	Logical flag indicating if MiDAS object should be returned.

Details

Setting as . MiDAS to TRUE will filter MiDAS object based on p-value cut-off given by HWE_cutoff.

Value

Data frame with Hardy Weinberg Equilibrium test results or a filtered MiDAS object.

Examples

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isCharacterOrNULL

```
# get HWE in groups defined by disease status
# grouping by `disease == 1` will divide samples into two groups:
# `disease == 1` and `not disease == 1`
HWETest(midas, experiment = "hla_alleles", HWE_group = disease == 1)
# filter MiDAS object by HWE test p-value
HWETest(midas, experiment = "hla_alleles", HWE_cutoff = 0.05, as.MiDAS = TRUE)
```

isCharacterOrNULL Check if object is character vector or NULL

Description

isCharacterOrNULL checks if the object is a character vector or NULL.

Usage

isCharacterOrNULL(x)

Arguments

x object to test.

Value

Logical indicating if object is character vector or NULL

isClass

Check if object is of class x

Description

isClassOrNULL checks if object is an instance of a specified class or is null.

Usage

```
isClass(x, class)
```

Arguments

х	object to test.
class	String specifying class to test.

Value

Logical indicating if x is an instance of class.

isClassOrNULL

Description

isClassOrNULL checks if object is an instance of a specified class or is null.

Usage

```
isClassOrNULL(x, class)
```

Arguments

х	object to test.
class	String specifying class to test.

Value

Logical indicating if x is an instance of class.

isCountOrNULL Check if object is count or NULL

Description

isCountOrNULL check if object is a count (a single positive integer) or NULL.

Usage

```
isCountOrNULL(x)
```

Arguments

x object to test.

Value

Logical indicating if object is count or NULL

isCountsOrZeros Check if vector contains only counts or zeros

Description

isCountsOrZeros checks if vector contains only positive integers or zeros.

Usage

```
isCountsOrZeros(x, na.rm = TRUE)
```

Arguments

Х	Numeric vector or object that can be unlist to numeric vector.
na.rm	Logical indicating if NA values should be accepted.

Value

Logical indicating if provided vector contains only positive integers or zeros.

```
isExperimentCountsOrZeros
```

Check if frequencies can be calculated for an experiment

Description

isExperimentCountsOrZeros checks if experiment contains only positive integers or zeros.

Usage

```
isExperimentCountsOrZeros(x, na.rm = TRUE)
```

Arguments

х	Matrix or SummarizedExperiment object.
na.rm	Logical indicating if NA values should be accepted.

Value

Logical indicating if x contains only positive integers or zeros.

isExperimentInheritanceModelApplicable

Check if experiment is inheritance model applicable

Description

isExperimentInheritanceModelApplicable check experiment's metadata for presence of "inheritance_model_app flag, indicating if inheritance model can be applied.

Usage

isExperimentInheritanceModelApplicable(experiment)

S3 method for class 'matrix'
isExperimentInheritanceModelApplicable(experiment)

S3 method for class 'SummarizedExperiment'
isExperimentInheritanceModelApplicable(experiment)

Arguments

experiment Matrix or SummarizedExperiment object.

Value

Logical flag.

isFlagOrNULL Check if object is flag or NULL

Description

isFlagOrNULL checks if object is flag (a length one logical vector) or NULL.

Usage

isFlagOrNULL(x)

Arguments

x object to test.

Value

Logical indicating if object is flag or NULL

isNumberOrNULL

Description

isNumberOrNULL checks if object is number (a length one numeric vector) or NULL.

Usage

isNumberOrNULL(x)

Arguments

x object to test.

Value

Logical indicating if object is number or NULL

isStringOrNULL Check if object is string or NULL

Description

isStringOrNULL checks if object is string (a length one character vector) or NULL.

Usage

isStringOrNULL(x)

Arguments

x object to test.

Value

Logical indicating if object is string or NULL

isTRUEorFALSE

Description

isTRUEorFALSE check if object is a flag (a length one logical vector) except NA.

Usage

isTRUEorFALSE(x)

Arguments

х

object to test.

Value

Logical indicating if object is TRUE or FALSE flag

Description

iterativeLRT performs likelihood ratio test in an iterative manner over groups of variables given in omnibus_groups.

Usage

iterativeLRT(object, placeholder, omnibus_groups)

Arguments

object	An existing fit from a model function such as lm, glm and many others.
placeholder	String specifying term to substitute with value from x. Ignored if set to NULL.
omnibus_groups	List of character vectors giving sets of variables for which omnibus test should be applied.

Value

Data frame containing summarised likelihood ratio test results.

iterativeModel

Description

Information about variable statistic from each model is extracted using tidy function.

Usage

```
iterativeModel(object, placeholder, variables, exponentiate = FALSE)
```

Arguments

object	An existing fit from a model function such as lm, glm and many others.
placeholder	String specifying term to substitute with value from x. Ignored if set to NULL.
variables	Character vector specifying variables to use in association tests.
exponentiate	Logical flag indicating whether or not to exponentiate the coefficient estimates. Internally this is passed to tidy. This is typical for logistic and multinomial regressions, but a bad idea if there is no log or logit link. Defaults to FALSE.

Value

Tibble containing per variable summarised model statistics. The exact output format is model dependent and controlled by model's dedicated tidy function.

kableResults

Create association analysis results table in HTML or LaTeX

Description

kableResults convert results table (runMiDAS output) to HTML or LaTeX format.

Usage

```
kableResults(
  results,
  colnames = NULL,
  header = "MiDAS analysis results",
  pvalue_cutoff = NULL,
  format = getOption("knitr.table.format"),
  scroll_box_height = "400px"
)
```

Arguments

results	Tibble as returned by runMiDAS.
colnames	Character vector of form c("new_name" = "old_name"), used to rename results colnames.
header	String specifying results table header.
<pre>pvalue_cutoff</pre>	Number specifying p-value cutoff for results to be included in output. If NULL no filtering is done.
format	String "latex" or "html".
scroll_box_height	
	A character string indicating the height of the table.

Value

Association analysis results table in HTML or LaTeX.

Examples

kir_frequencies KIR genes frequencies scraped from allelefrequencies.net

Description

Accessed on 28.08.20

Usage

kir_frequencies

Format

A data frame with 3744 rows and 3 variables:

var allele number, character

population reference population name, character

frequency KIR genes carrier frequency in reference population, float

Details

A dataset containing KIR genes frequencies across 16 genes. For details visit the search results page in the allelefrequencies.net database website.

Source

www.allelefrequencies.net

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lapply_tryCatch *lapply with tryCatch routine*

Description

Used to run function iteratively over list, while using tryCatch to catch warnings and errors to finally present a summary of issues rather than error on each and every one. Used in iterativeLRT and iterativeModel.

Usage

lapply_tryCatch(X, FUN, err_res, ...)

Arguments

Х	a vector (atomic or list) or an expression object. Other objects (including classed objects) will be coerced by base::as.list.
FUN	the function to be applied to each element of X: see 'Details'. In the case of functions like +, %*%, the function name must be backquoted or quoted.
err_res	Function creating a result that should be output in case of error.
	optional arguments to FUN.

Value

List of elements as returned by FUN.

listMiDASDictionaries List HLA alleles dictionaries

Description

listMiDASDictionaries lists dictionaries shipped with the MiDAS package. See hlaToVariable for more details on dictionaries.

Usage

```
listMiDASDictionaries(pattern = "allele", file.names = FALSE)
```

Arguments

pattern	String used to match dictionary names, it can be a regular expression. By default all names are matched.
file.names	Logical value. If FALSE, only the names of dictionaries are returned. If TRUE their paths are returned.

Value

Character vector giving names of available HLA alleles dictionaries.

LRTest

Description

LRTest carry out an asymptotic likelihood ratio test for two models.

Usage

LRTest(mod0, mod1)

Arguments

mod0	An existing fit from a model function such as lm, glm and many others.
mod1	Object of the same class as mod0 with extra terms included.

Details

mod0 have to be a reduced version of mod1. See examples.

Value

Data frame with the results of likelihood ratio test of the supplied models.

Column term holds new variables appearing in mod1, df difference in degrees of freedom between models, logLik difference in log likelihoods, statistic Chisq statistic and p.value corresponding p-value.

MiDAS-class MiDAS class

Description

The MiDAS class is a MultiAssayExperiment object containing data and metadata required for MiDAS analysis.

Valid MiDAS object must have unique features names across all experiments and colData. It's metadata list needs to have a placeholder element, which is a string specifying name of column in colData used when defining statistical model for downstream analyses (see runMiDAS for more details). Optionally the object's metadata can also store 'hla_calls' and 'kir_calls' data frames (see prepareMiDAS for more details).

Usage

```
## S4 method for signature 'MiDAS'
getExperiments(object)
## S4 method for signature 'MiDAS'
getHlaCalls(object)
## S4 method for signature 'MiDAS'
```

```
getKirCalls(object)
## S4 method for signature 'MiDAS'
getPlaceholder(object)
## S4 method for signature 'MiDAS'
getOmnibusGroups(object, experiment)
## S4 method for signature 'MiDAS'
getFrequencies(
  object,
  experiment,
  carrier_frequency = FALSE,
  compare = FALSE,
 ref_pop = list(hla_alleles = c("USA NMDP African American pop 2", "USA NMDP Chinese",
   "USA NMDP European Caucasian", "USA NMDP Hispanic South or Central American",
    "USA NMDP Japanese", "USA NMDP North American Amerindian",
  "USA NMDP South Asian Indian"), kir_genes = c("USA California African American KIR",
    "USA California Asian American KIR", "USA California Caucasians KIR",
    "USA California Hispanic KIR")),
  ref = list(hla_alleles = allele_frequencies, kir_genes = kir_frequencies)
)
## S4 method for signature 'MiDAS'
filterByFrequency(
  object,
  experiment,
  lower_frequency_cutoff = NULL,
  upper_frequency_cutoff = NULL,
  carrier_frequency = FALSE
)
## S4 method for signature 'MiDAS'
filterByOmnibusGroups(object, experiment, groups)
## S4 method for signature 'MiDAS'
filterByVariables(object, experiment, variables)
## S4 method for signature 'MiDAS'
getAllelesForAA(object, aa_pos)
```

Arguments

object	MiDAS object.
experiment	String specifying experiment.
carrier_freque	ncy
	Logical flag indicating if carrier frequency should be returned.
compare	Logical flag indicating if hla_calls frequencies should be compared to reference frequencies given in ref.
ref_pop	Named list of character vectors giving names of reference populations in ref to compare with. Optionally vectors can be named, then those names will be used as population names. Each vector should correspond to a specific experiment.

ref	Named list of reference frequencies data frames. Each element should give reference for a specific experiment. See allele_frequencies for an example on how reference frequency data frame should be formatted.
lower_frequency	y_cutoff
	Number giving lower frequency threshold. Numbers greater than 1 are interpreted as the number of feature occurrences, numbers between 0 and 1 as fractions.
upper_frequency_cutoff	
	Number giving upper frequency threshold. Numbers greater than 1 are interpreted as the number of feature occurrences, numbers between 0 and 1 as fractions.
groups	Character vector specifying omnibus groups to select. See getOmnibusGroups for more details.
variables	Character vector specifying features to select.
aa_pos	String specifying gene and amino acid position, example "A_9".

Value

Instance of class MiDAS

midasToWide

Transform MiDAS to wide format data.frame

Description

Transform MiDAS to wide format data.frame

Usage

midasToWide(object, experiment)

Arguments

object	Object of class MiDAS
experiment	Character specifying experiments to include

Value

Data frame representation of MiDAS object. Consecutive columns holds values of variables from MiDAS's experiments and colData. The metadata associated with experiments is not preserved.

MiDAS_tut_HLA

Description

Example HLA calls data used in MiDAS tutorial

Usage

MiDAS_tut_HLA

Format

Data frame with 1000 rows and 19 columns. First column holds samples ID's, following columns holds HLA alleles calls for different genes.

- **ID** Character sample ID
- A_1 Character
- A_2 Character
- **B_1** Character
- **B_2** Character
- C_1 Character
- C_2 Character
- DPA1_1 Character
- DPA1_2 Character
- DPB1_1 Character
- DPB1_2 Character
- DQA1_1 Character
- DQA1_2 Character
- DQB1_1 Character
- DQB1_2 Character
- DRA_1 Character
- DRA_2 Character
- DRB1_1 Character
- DRB1_2 Character

MiDAS_tut_KIR

Description

Example KIRR presence/absence data used in MiDAS tutorial

Usage

MiDAS_tut_KIR

Format

Data frame with 1000 rows and 17 columns. First column holds samples ID's, following columns holds presence/absence indicators for different KIR genes.

- ID Character sample IDKIR3DL3 Integer
- KIR2DS2 Integer
- KIR2DL2 Integer
- KIR2DL3 Integer
- KIR2DP1 Integer
- KIR2DL1 Integer
- KIR3DP1 Integer
- KIR2DL4 Integer
- KIR3DL1 Integer
- KIR3DS1 Integer
- KIR2DL5 Integer
- KIR2DS3 Integer
- KIR2DS5 Integer
- KIR2DS4 Integer
- KIR2DS1 Integer
- KIR3DL2 Integer

MiDAS_tut_object MiDAS tutorial MiDAS object

Description

Example MiDAS object created with data used in MiDAS tutorial: MiDAS_tut_HLA, MiDAS_tut_KIR, MiDAS_tut_pheno. Used in code examlpes and unit tests.

Usage

MiDAS_tut_object

Format

MiDAS object with following experiments defined:

hla_alleles SummarizedExperiment with 447 rows and 1000 columns
hla_aa SummarizedExperiment with 1223 rows and 1000 columns
hla_g_groups SummarizedExperiment with 46 rows and 1000 columns
hla_supertypes SummarizedExperiment with 12 rows and 1000 columns
hla_NK_ligands SummarizedExperiment with 5 rows and 1000 columns
kir_genes SummarizedExperiment with 16 rows and 1000 columns
kir_haplotypes SummarizedExperiment with 6 rows and 1000 columns
hla_kir_interactions SummarizedExperiment with 29 rows and 1000 columns
hla_divergence matrix with 4 rows and 1000 columns
hla_het SummarizedExperiment with 9 rows and 1000 columns

MiDAS_tut_pheno MiDAS tutorial phenotype data

Description

Example phenotype data used in MiDAS tutorial

Usage

MiDAS_tut_pheno

Format

Data frame with 1000 rows and 4 columns.

ID Character sample ID disease Integer lab_value Numeric outcome Integer objectHasPlaceholder Check if placeholder is present in object formula

Description

isTRUEorFALSE check if object is a flag (a length one logical vector) except NA.

Usage

```
objectHasPlaceholder(object, placeholder)
```

Arguments

object	statistical model to test.
placeholder	string specifying name of placeholder.

Value

Logical indicating if placeholder is present in object formula.

omnibusTest	Omnibus test	

Description

OmnibusTest calculates overall p-value for linear combination of variables using likelihood ratio test.

Usage

```
omnibusTest(
   object,
   omnibus_groups,
   placeholder = "term",
   correction = "bonferroni",
   n_correction = NULL
)
```

Arguments

object	An existing fit from a model function such as lm, glm and many others.
	List of character vectors giving sets of variables for which omnibus test should be applied.
placeholder	String specifying term in object's formula which should be substituted with variables during analysis.
correction	String specifying multiple testing correction method. See details for further information.

n_correction Integer specifying number of comparisons to consider during multiple testing correction calculations. For Bonferroni correction it is possible to specify a number lower than the number of comparisons being made. This is useful in cases when knowledge about the biology or redundance of alleles reduces the need for correction. For other methods it must be at least equal to the number of comparisons being made; only set this (to non-default) when you know what you are doing!

Details

Likelihood ratio test is conducted by comparing a model given in an object with an extended model, that is created by including the effect of variables given in variables as their linear combination.

Value

Data frame with columns:

- "group" Omnibus group name
- "term" Elements of omnibus group added to base model
- "df" Difference in degrees of freedom between base and extended model
- · "logLik" Difference in log likelihoods between base and extended model
- "statistic" Chisq statistic
- "p.value" P-value
- "p.adjusted" Adjusted p-value

Examples

prepareMiDAS

Construct a MiDAS object

Description

prepareMiDAS transform HLA alleles calls and KIR calls according to selected experiments creating a MiDAS object.

Usage

```
prepareMiDAS(
  hla_calls = NULL,
  kir_calls = NULL,
  colData,
  experiment = c("hla_alleles", "hla_aa", "hla_g_groups", "hla_supertypes",
    "hla_NK_ligands", "kir_genes", "kir_haplotypes", "hla_kir_interactions",
"hla_divergence", "hla_het", "hla_custom", "kir_custom"),
  placeholder = "term",
  lower_frequency_cutoff = NULL,
  upper_frequency_cutoff = NULL,
  indels = TRUE,
  unkchar = FALSE,
  hla_divergence_aa_selection = "binding_groove",
  hla_het_resolution = 8,
  hla_dictionary = NULL,
  kir_dictionary = NULL
)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.	
kir_calls	KIR calls data frame, as returned by readKirCalls function.	
colData	Data frame holding additional variables like phenotypic observations or covari- ates. It have to contain 'ID' column holding samples identifiers corresponding to identifiers in hla_calls and kir_calls. Importantly rows of hla_calls and kir_calls without corresponding phenotype are discarded.	
experiment	Character vector indicating analysis type for which data should be prepared. Valid choices are "hla_alleles", "hla_aa", "hla_g_groups", "hla_supertypes", "hla_NK_ligands", "kir_genes", "hla_kir_interactions", "hla_divergence", "hla_het". See details for further explanations.	
placeholder	String giving name for dummy variable inserted to colData. This variable can be than used to define base statistical model used by runMiDAS.	
lower_frequency	y_cutoff	
	Number giving lower frequency threshold. Numbers greater than 1 are inter- preted as the number of feature occurrences, numbers between 0 and 1 as frac- tions.	
upper_frequency		
	Number giving upper frequency threshold. Numbers greater than 1 are inter- preted as the number of feature occurrences, numbers between 0 and 1 as frac- tions.	
indels	Logical indicating whether indels should be considered when checking amino acid variability in 'hla_aa' experiment.	
unkchar	Logical indicating whether unknown characters in the alignment should be con- sidered when checking amino acid variability in 'hla_aa' experiment.	
hla_divergence_aa_selection		
	String specifying variable region in peptide binding groove which should be con- sidered for Grantham distance calculation. Valid choices includes: "binding_groove", "B_pocket", "F_pocket". See details for more information.	

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hla_het_resolution		
	Number specifying HLA alleles resolution used to calculate heterogeneity in "hla_het" experiment.	
hla_dictionary	Data frame giving HLA allele dictionary used in 'hla_custom' experiment. See hlaToVariable for more details.	
kir_dictionary	Data frame giving KIR genes dictionary used in 'kir_custom' experiment. See countsToVariables for more details.	

Details

experiment specifies analysis types for which hla_calls and kir_call should be prepared.

- 'hla_alleles' hla_calls are transformed to counts matrix describing number of allele occurrences for each sample. This experiment is used to test associations on HLA alleles level.
- 'hla_aa' hla_calls are transformed to a matrix of variable amino acid positions. See hlaToAAVariation for more details. This experiment is used to test associations on amino acid level.
- "hla_g_groups" hla_calls are translated into HLA G groups and transformed to matrix describing number of G group occurrences for each sample. See hlaToVariable for more details. This experiment is used to test associations on HLA G groups level.
- "hla_supertypes" hla_calls are translated into HLA supertypes and transformed to matrix describing number of G group occurrences for each sample. See hlaToVariable for more details. This experiment is used to test associations on HLA supertypes level.
- "hla_NK_ligands" hla_calls are translated into NK ligands, which includes HLA Bw4/Bw6 and HLA C1/C2 groups and transformed to matrix describing number of their occurrences for each sample. See hlaToVariable for more details. This experiment is used to test associations on HLA NK ligands level.
- "kir_genes" kir_calls are transformed to counts matrix describing number of KIR gene occurrences for each sample. This experiment is used to test associations on KIR genes level.
- "hla_kir_interactions" hla_calls and kir_calls are translated to HLA KIR interactions as defined in Pende et al., 2019.. See getHlaKirInteractions for more details. This experiment is used to test associations on HLA KIR interactions level.
- "hla_divergence" Grantham distance for class I HLA alleles is calculated based on hla_calls using original formula by Grantham R. 1974.. See hlaCallsGranthamDistance for more details. This experiment is used to test associations on HLA divergence level measured by Grantham distance.
- "hla_het" hla_calls are transformed to heterozygosity status, where 1 designates a heterozygote and 0 homozygote. Heterozygosity status is calculated only for classical HLA genes (A, B, C, DQA1, DQB1, DRA, DRB1, DPA1, DPB1). This experiment is used to test associations on HLA divergence level measured by heterozygosity.

Value

Object of class MiDAS

Examples

prepareMiDAS_hla_aa Prepare MiDAS data on HLA amino acid level

Description

Prepare MiDAS data on HLA amino acid level

Usage

```
prepareMiDAS_hla_aa(hla_calls, indels = TRUE, unkchar = FALSE, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
indels	Logical indicating whether indels should be considered when checking variabil- ity.
unkchar	Logical indicating whether unknown characters in the alignment should be con- sidered when checking variability.
	Not used

Value

SummarizedExperiment

prepareMiDAS_hla_alleles

Prepare MiDAS data on HLA allele level

Description

Prepare MiDAS data on HLA allele level

Usage

```
prepareMiDAS_hla_alleles(hla_calls, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
	Not used

Value

prepareMiDAS_hla_custom

Prepare MiDAS data on custom HLA level

Description

Prepare MiDAS data on custom HLA level

Usage

```
prepareMiDAS_hla_custom(hla_calls, hla_dictionary, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
hla_dictionary	Data frame giving HLA allele dictionary. See hlaToVariable for more details.
	Not used

Value

Matrix

prepareMiDAS_hla_divergence

Prepare MiDAS data on HLA divergence level

Description

Prepare MiDAS data on HLA divergence level

Usage

```
prepareMiDAS_hla_divergence(
    hla_calls,
    hla_divergence_aa_selection = "binding_groove",
    ...
)
```

Arguments

```
      hla_calls
      HLA calls data frame, as returned by readHlaCalls function.

      hla_divergence_aa_selection
      String specifying variable region in peptide binding groove which should be considered for Grantham distance calculation. Valid choices includes: "binding_groove", "B_pocket", "F_pocket". See details for more information.

      ...
      Not used
```

Value

```
prepareMiDAS_hla_g_groups
```

Prepare MiDAS data on HLA allele's G groups level

Description

Prepare MiDAS data on HLA allele's G groups level

Usage

```
prepareMiDAS_hla_g_groups(hla_calls, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
	Not used

Value

Matrix

prepareMiDAS_hla_het Prepare MiDAS data on HLA heterozygosity level

Description

Prepare MiDAS data on HLA heterozygosity level

Usage

```
prepareMiDAS_hla_het(hla_calls, hla_het_resolution = 8, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.	
hla_het_resolution		
	Number specifying HLA alleles resolution used to calculate heterogeneity.	
	Not used	

Value

prepareMiDAS_hla_kir_interactions

Prepare MiDAS data on HLA - KIR interactions level

Description

Prepare MiDAS data on HLA - KIR interactions level

Usage

```
prepareMiDAS_hla_kir_interactions(hla_calls, kir_calls, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
kir_calls	KIR calls data frame, as returned by readKirCalls function.
	Not used

Value

Matrix

prepareMiDAS_hla_NK_ligands Prepare MiDAS data on HLA allele's groups level

Description

Prepare MiDAS data on HLA allele's groups level

Usage

```
prepareMiDAS_hla_NK_ligands(hla_calls, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
	Not used

Value

```
prepareMiDAS_hla_supertypes
```

Prepare MiDAS data on HLA allele's supertypes level

Description

Prepare MiDAS data on HLA allele's supertypes level

Usage

```
prepareMiDAS_hla_supertypes(hla_calls, ...)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
	Not used

Value

Matrix

```
prepareMiDAS_kir_custom
```

Prepare MiDAS data on custom KIR level

Description

Prepare MiDAS data on custom KIR level

Usage

```
prepareMiDAS_kir_custom(kir_calls, kir_dictionary, ...)
```

Arguments

kir_calls	KIR calls data frame, as returned by readKirCalls function.
kir_dictionary	Data frame giving KIR genes dictionary. See countsToVariables for more details.
	Not used

Value

prepareMiDAS_kir_genes

```
Prepare MiDAS data on KIR genes level
```

Description

Prepare MiDAS data on KIR genes level

Usage

```
prepareMiDAS_kir_genes(kir_calls, ...)
```

Arguments

kir_calls	KIR calls data frame, as returned by readKirCalls function.
	Not used

Value

Matrix

```
prepareMiDAS_kir_haplotypes

Prepare MiDAS data on KIR haplotypes level
```

Description

Prepare MiDAS data on KIR haplotypes level

Usage

```
prepareMiDAS_kir_haplotypes(kir_calls, ...)
```

Arguments

kir_calls	KIR calls data frame, as returned by readKirCalls function.
	Not used

Value

readHlaAlignments Read HLA allele alignments

Description

readHlaAlignments read HLA allele alignments from file.

Usage

```
readHlaAlignments(file, gene = NULL, trim = FALSE, unkchar = "")
```

Arguments

file	Path to input file.
gene	Character vector of length one specifying the name of a gene for which alignment is required. See details for further explanations.
trim	Logical indicating if alignment should be trimmed to start codon of the mature protein.
unkchar	Character to be used to represent positions with unknown sequence.

Details

HLA allele alignment file should follow EBI database format, for details see ftp://ftp.ebi.ac. uk/pub/databases/ipd/imgt/hla/alignments/README.md.

All protein alignment files from the EBI database are shipped with the package. They can be easily accessed using gene parameter. If gene is set to NULL, file parameter is used instead and alignment is read from the provided file. In EBI database alignments for DRB1, DRB3, DRB4 and DRB5 genes are provided as a single file, here they are separated.

Additionally, for the alleles without sequence defined in the original alignment files we have infered thier sequence based on known higher resolution alleles.

Value

Matrix containing HLA allele alignments.

Rownames correspond to allele numbers and columns to positions in the alignment. Sequences following the termination codon are marked as empty character (""). Unknown sequences are marked with a character of choice, by default "". Stop codons are represented by a hash (X). Insertion and deletions are marked with period (.).

Examples

```
hla_alignments <- readHlaAlignments(gene = "A")</pre>
```

readHlaCalls

Description

readHlaCalls read HLA allele calls from file

Usage

```
readHlaCalls(file, resolution = 4, na.strings = c("Not typed", "-", "NA"))
```

Arguments

file	Path to input file.
resolution	Number specifying desired resolution.
na.strings	a character vector of strings which are to be interpreted as NA values. Blank fields are also considered to be missing values in logical, integer, numeric and complex fields. Note that the test happens <i>after</i> white space is stripped from the input, so na.strings values may need their own white space stripped in advance.

Details

Input file has to be a tsv formatted table with a header. First column should contain sample IDs, further columns hold HLA allele numbers. See system.file("extdata", "MiDAS_tut_HLA.txt", package = "midasHLA") file for an example.

resolution parameter can be used to reduce HLA allele numbers. If reduction is not needed resolution can be set to 8. resolution parameter can take the following values: 2, 4, 6, 8. For more details about HLA allele numbers resolution see http://hla.alleles.org/nomenclature/ naming.html.

Value

HLA calls data frame. First column hold sample IDs, further columns hold HLA allele numbers.

Examples

```
file <- system.file("extdata", "MiDAS_tut_HLA.txt", package = "midasHLA")
hla_calls <- readHlaCalls(file)</pre>
```

readKirCalls

Description

readKirCalls read KIR calls from file.

Usage

```
readKirCalls(file, na.strings = c("", "NA", "uninterpretable"))
```

Arguments

file	Path to input file.
na.strings	a character vector of strings which are to be interpreted as NA values. Blank fields are also considered to be missing values in logical, integer, numeric and complex fields. Note that the test happens <i>after</i> white space is stripped from the input, so na.strings values may need their own white space stripped in advance.

Details

Input file has to be a tsv formatted table. First column should be named "ID" and contain samples IDs, further columns should hold KIR genes presence/absence indicators. See system.file("extdata", "MiDAS_tut_KIR", package = "midasHLA") for an example.

Value

Data frame containing KIR gene's counts. First column hold samples IDs, further columns hold KIR genes presence / absence indicators.

Examples

```
file <- system.file("extdata", "MiDAS_tut_KIR.txt", package = "midasHLA")
readKirCalls(file)</pre>
```

reduceAlleleResolution

Reduce HLA alleles

Description

reduceAlleleResolution reduce HLA allele numbers resolution.

Usage

```
reduceAlleleResolution(allele, resolution = 4)
```

reduceHlaCalls

Arguments

allele	Character vector with HLA allele numbers.
resolution	Number specifying desired resolution.

Details

In cases when allele number contain additional suffix their resolution can not be unambiguously reduced. These cases are returned unchanged. Function behaves in the same manner if resolution is higher than resolution of input HLA allele numbers.

NA values are accepted and returned as NA.

TODO here we give such warning when alleles have G or GG suffix (see http://hla.alleles.org/alleles/g_groups.html) "Reducing G groups alleles, major allele gene name will be used." I dond't really remember why we are doing this xd These allele numbers are processed as normal alleles (without suffix). Let me know if this warning is relevant or we could go without it. If we want to leave it lets also add text in documentation.

Value

Character vector containing reduced HLA allele numbers.

Examples

```
reduceAlleleResolution(c("A*01", "A*01:24", "C*05:24:55:54"), 2)
```

reduceHlaCalls Reduce HLA calls resolution

Description

reduceHlaCalls reduces HLA calls data frame to specified resolution.

Usage

```
reduceHlaCalls(hla_calls, resolution = 4)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
resolution	Number specifying desired resolution.

Details

Alleles with resolution greater than resolution or optional suffixes are returned unchanged.

Value

HLA calls data frame reduced to specified resolution.

Examples

```
reduceHlaCalls(MiDAS_tut_HLA, resolution = 2)
```

runMiDAS

Description

runMiDAS perform association analysis on MiDAS data using statistical model of choice. Function is intended for use with prepareMiDAS. See examples section.

Usage

```
runMiDAS(
 object,
 experiment,
 inheritance_model = NULL,
 conditional = FALSE,
 omnibus = FALSE,
 omnibus_groups_filter = NULL,
 lower_frequency_cutoff = NULL,
 upper_frequency_cutoff = NULL,
 correction = "bonferroni",
 n_correction = NULL,
 exponentiate = FALSE,
  th = 0.05,
  th_adj = TRUE,
 keep = FALSE,
 rss_th = 1e-07
)
```

Arguments

object	An existing fit from a model function such as lm, glm and many others.
experiment	String indicating the experiment associated with object's MiDAS data to use. Valid values includes: "hla_alleles", "hla_aa", "hla_g_groups", "hla_supertypes", "hla_NK_ligands", "kir_genes", "kir_haplotypes", "hla_kir_interactions", "hla_divergence", "hla_het", "hla_custom", "kir_custom". See prepareMiDAS for more information.
inheritance_mod	lel
	String specifying inheritance model to use. Available choices are "dominant", "recessive", "additive".
conditional	Logical flag indicating if conditional analysis should be performed.
omnibus omnibus_groups_	Logical flag indicating if omnibus test should be used. filter
	Character vector specifying omnibus groups to use.
lower_frequency	_cutoff
	Number giving lower frequency threshold. Numbers greater than 1 are inter- preted as the number of feature occurrences, numbers between 0 and 1 as frac- tions.
upper_frequency	_cutoff
	Number giving upper frequency threshold. Numbers greater than 1 are interpreted as the number of feature occurrences, numbers between 0 and 1 as fractions.

correction	String specifying multiple testing correction method. See details for further information.
n_correction	Integer specifying number of comparisons to consider during multiple testing correction calculations. For Bonferroni correction it is possible to specify a number lower than the number of comparisons being made. This is useful in cases when knowledge about the biology or redundance of alleles reduces the need for correction. For other methods it must be at least equal to the number of comparisons being made; only set this (to non-default) when you know what you are doing!
exponentiate	Logical flag indicating whether or not to exponentiate the coefficient estimates. Internally this is passed to tidy. This is typical for logistic and multinomial regressions, but a bad idea if there is no log or logit link. Defaults to FALSE.
th	Number specifying threshold for a variable to be considered significant.
th_adj	Logical flag indicating if adjusted p-value should be used as threshold criteria, otherwise unadjusted p-value is used.
keep	Logical flag indicating if the output should be a list of results resulting from each selection step. Default is to return only the final result.
rss_th	Number specifying residual sum of squares threshold at which function should stop adding additional variables. As the residual sum of squares approaches 0 the perfect fit is obtained making further attempts at variable selection nonsense. This behavior can be controlled using rss_th.

Details

By default statistical analysis is performed iteratively on each variable in selected experiment. This is done by substituting placeholder in the object's formula with each variable in the experiment.

Setting conditional argument to TRUE will cause the statistical analysis to be performed in a stepwise conditional testing manner, adding the previous top-associated variable as a covariate to object's formula. The analysis stops when there is no more significant variables, based on self-defined threshold (th argument). Either adjusted or unadjusted p-values can be used as the selection criteria, which is controlled using th_adj argument.

Setting omnibus argument to TRUE will cause the statistical analysis to be performed iteratively on groups of variables (like residues at particular amino acid position) using likelihood ratio test.

Argument inheritance_model specifies the inheritance model that should be applyed to experiment's data. Following choices are available:

- "dominant" carrier status is sufficient for expression of the phenotype (non-carrier: 0, heterozygous & homozygous carrier: 1).
- "recessive" two copies are required for expression of the phenotype (non-carrier & heterozygous carrier: 0, homozygous carrier: 1).
- "additive" allele dosage matters, homozygous carriers show stronger phenotype expression or higher risk than heterozygous carriers (non-carrier = 0, heterozygous carrier = 1, homozygous carrier = 2).
- "overdominant" heterozygous carriers are at higher risk compared to non-carriers or homozygous carriers (non-carrier & homozygous carrier = 0, heterozygous carrier = 1).

correction specifies p-value adjustment method to use, common choice is Benjamini & Hochberg (1995) ("BH"). Internally this is passed to p.adjust.

Analysis results, depending on the parameters:

- conditional=FALSE, omnibus=FALSE Tibble with first column "term" holding names of tested variables (eg. alleles). Further columns depends on the used model and are determined by associated tidy function. Generally they will include "estimate", "std.error", "statistic", "p.value", "conf.low", "conf.high", "p.adjusted".
- conditional=TRUE, omnibus=FALSE Tibble or a list of tibbles, see keep argument. The first column "term" hold names of tested variables. Further columns depends on the used model and are determined by associated tidy function. Generally they will include "estimate", "std.error", "statistic", "p.value", "conf.low", "conf.high", "p.adjusted".
- conditional=FALSE, omnibus=TRUE Tibble with first column holding names of tested omnibus
 groups (eg. amino acid positions) and second names of variables in the group (eg. residues).
 Further columns are: "df" giving difference in degrees of freedom between base and extended
 model, "statistic" giving Chisq statistic, "p.value" and "p.adjusted".
- conditional=TRUE, omnibus=TRUE Tibble or a list of tibbles, see keep argument. The first column hold names of tested omnibus groups (eg. amino acid positions), second column hold names of variables in the group (eg. residues). Further columns are: "df" giving difference in degrees of freedom between base and extended model, "statistic" giving Chisq statistic, "p.value" and "p.adjusted".

Examples

```
# create MiDAS object
midas <- prepareMiDAS(hla_calls = MiDAS_tut_HLA,</pre>
                      colData = MiDAS_tut_pheno,
                      experiment = c("hla_alleles", "hla_aa")
)
# construct statistical model
object <- lm(disease ~ term, data = midas)</pre>
# run analysis
runMiDAS(object, experiment = "hla_alleles", inheritance_model = "dominant")
# omnibus test
# omnibus_groups_filter argument can be used to restrict omnibus test only
# to selected variables groups, here we restrict the analysis to HLA-A
# positions 29 and 43.
runMiDAS(
  object,
  experiment = "hla_aa",
  inheritance_model = "dominant",
  omnibus = TRUE,
  omnibus_groups_filter = c("A_29", "A_43")
)
```

runMiDASGetVarsFreq Get variables frequencies from MiDAS

stringMatches

Description

Helper getting variables frequencies from MiDAS object. Additionally for binary test covariate frequencies per phenotype are added. Used in scope of runMiDAS.

Usage

```
runMiDASGetVarsFreq(midas, experiment, test_covar)
```

Arguments

midas	MiDAS object.
experiment	String specifying experiment from midas.
test_covar	String giving name of test covariate.

Value

Data frame with variable number of columns. First column, "term" holds experiment's variables, further columns hold number of variable occurrence and their frequencies.

stringMatches	Check if string matches one of possible values
---------------	--

Description

stringMatches checks if string is equal to one of the choices.

Usage

```
stringMatches(x, choice)
```

Arguments

х	string to test.
choice	Character vector with possible values for x.

Value

Logical indicating if x matches one of the strings in choice.

summariseAAPosition Summarize amino acid position

Description

List HLA alleles and amino acid residues at a given position.

Usage

```
summariseAAPosition(hla_calls, aa_pos, aln = NULL, na.rm = FALSE)
```

Arguments

hla_calls	HLA calls data frame, as returned by readHlaCalls function.
aa_pos	String specifying gene and amino acid position, example "A_9".
aln	Matrix containing amino acid sequence alignments as returned by readHlaAlignments function. By default function will use alignment files shipped with the package.
na.rm	Logical flag indicating if NA values should be considered for frequency calcula- tions.

Value

Data frame containing HLA alleles, their corresponding amino acid residues and frequencies at requested position.

Examples

summariseAAPosition(MiDAS_tut_HLA, "A_9")

updateModel

Extend and Re-fit a Model Call

Description

updateModel adds new variables to model and re-fit it.

Usage

```
updateModel(object, x, placeholder = NULL, backquote = TRUE, collapse = " + ")
```

Arguments

object	An existing fit from a model function such as lm, glm and many others.
х	Character vector specifying variables to be added to model.
placeholder	String specifying term to substitute with value from x. Ignored if set to NULL.
backquote	Logical indicating if added variables should be quoted. Elements of this vector are recycled over x.
collapse	String specifying how variables should be combined. Defaults to " + " ie. linear combination.

Value

Updated fitted object.

validateFrequencyCutoffs

Validate frequency cutoffs

Description

validateFrequencyCutoffs checks if lower_frequency_cutoff and upper_frequency_cutoff are valid.

Usage

```
validateFrequencyCutoffs(lower_frequency_cutoff, upper_frequency_cutoff)
```

Arguments

lower_frequency_cutoff Number upper_frequency_cutoff Number

Details

lower_frequency_cutoff and upper_frequency_cutoff should be a positive numbers, giving either frequency or counts. lower_frequency_cutoff has to be lower than upper_frequency_cutoff.

Value

Logical indicating if lower_frequency_cutoff and upper_frequency_cutoff are valid.

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