

Package ‘cytominer’

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Title Methods for Image-Based Cell Profiling

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Description Typical morphological profiling datasets have millions of cells and hundreds of features per cell. When working with this data, you must clean the data, normalize the features to make them comparable across experiments, transform the features, select features based on their quality, and aggregate the single-cell data, if needed. 'cytominer' makes these steps fast and easy. Methods used in practice in the field are discussed in Caicedo (2017) <[doi:10.1038/nmeth.4397](https://doi.org/10.1038/nmeth.4397)>. An overview of the field is presented in Caicedo (2016) <[doi:10.1016/j.copbio.2016.04.003](https://doi.org/10.1016/j.copbio.2016.04.003)>.

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aggregate	<i>Aggregate data based on given grouping.</i>
-----------	--

Description

aggregate aggregates data based on the specified aggregation method.

Usage

```
aggregate(
  population,
  variables,
  strata,
  operation = "mean",
  univariate = TRUE,
  ...
)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.
strata	character vector specifying grouping variables for aggregation.
operation	optional character string specifying method for aggregation, e.g. "mean", "median", "mean+sd". A sequence can comprise only of univariate functions.
univariate	boolean specifying whether the aggregation function is univariate or multivariate.
...	optional arguments passed to aggregation operation

Value

aggregated data of the same class as population.

Examples

```
population <- tibble::tibble(
  Metadata_group = c(
    "control", "control", "control", "control",
    "experiment", "experiment", "experiment",
    "experiment"
  ),
  Metadata_batch = c("a", "a", "b", "b", "a", "a", "b", "b"),
  AreaShape_Area = c(10, 12, 15, 16, 8, 8, 7, 7)
)
variables <- c("AreaShape_Area")
strata <- c("Metadata_group", "Metadata_batch")
aggregate(population, variables, strata, operation = "mean")
```

correlation_threshold *Remove redundant variables.*

Description

correlation_threshold returns list of variables such that no two variables have a correlation greater than a specified threshold.

Usage

```
correlation_threshold(variables, sample, cutoff = 0.9, method = "pearson")
```

Arguments

variables	character vector specifying observation variables.
sample	tbl containing sample used to estimate parameters.
cutoff	threshold between [0,1] that defines the minimum correlation of a selected feature.
method	optional character string specifying method for calculating correlation. This must be one of the strings "pearson" (default), "kendall", "spearman".

Details

correlation_threshold is a wrapper for `caret::findCorrelation`.

Value

character vector specifying observation variables to be excluded.

Examples

```
suppressMessages(suppressWarnings(library(magrittr)))
sample <- tibble::tibble(
  x = rnorm(30),
  y = rnorm(30) / 1000
)

sample %<>% dplyr::mutate(z = x + rnorm(30) / 10)
variables <- c("x", "y", "z")

head(sample)
cor(sample)

# `x` and `z` are highly correlated; one of them will be removed
correlation_threshold(variables, sample)
```

count_na_rows

Count the number of NAs per variable.

Description

count_na_rows counts the number of NAs per variable.

Usage

```
count_na_rows(population, variables)
```

Arguments

population tbl with grouping (metadata) and observation variables.
 variables character vector specifying observation variables.

Value

data frame with frequency of NAs per variable.

Examples

```
population <- tibble::tibble(
  Metadata_group = c(
    "control", "control", "control", "control",
    "experiment", "experiment", "experiment", "experiment"
  ),
  Metadata_batch = c("a", "a", "b", "b", "a", "a", "b", "b"),
  AreaShape_Area = c(10, 12, 15, 16, 8, 8, 7, 7),
  AreaShape_length = c(2, 3, NA, NA, 4, 5, 1, 5)
)
variables <- c("AreaShape_Area", "AreaShape_length")
count_na_rows(population, variables)
```

 covariance

Compute covariance matrix and vectorize.

Description

covariance computes the covariance matrix and vectorize it.

Usage

```
covariance(population, variables)
```

Arguments

population tbl with grouping (metadata) and observation variables.
 variables character vector specifying observation variables.

Value

data frame of 1 row comprising vectorized covariance matrix.

Examples

```
population <- tibble::tibble(  
  x = rnorm(30),  
  y = rnorm(30),  
  z = rnorm(30)  
)  
  
variables <- c("x", "y")  
  
covariance(population, variables)
```

drop_na_columns	<i>Remove variables with NA values.</i>
-----------------	---

Description

drop_na_columns returns list of variables which have greater than a specified threshold number of NAs.

Usage

```
drop_na_columns(population, variables, cutoff = 0.05)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.
cutoff	threshold between [0,1]. Variables with an NA frequency > cutoff are returned.

Value

character vector specifying observation variables to be excluded.

Examples

```
population <- tibble::tibble(  
  Metadata_group = c(  
    "control", "control", "control", "control",  
    "experiment", "experiment", "experiment", "experiment"  
  ),  
  Metadata_batch = c("a", "a", "b", "b", "a", "a", "b", "b"),  
  AreaShape_Area = c(10, 12, 15, 16, 8, 8, 7, 7),  
  AreaShape_Length = c(2, 3, NA, NA, 4, 5, 1, 5)  
)  
variables <- c("AreaShape_Area", "AreaShape_Length")  
drop_na_columns(population, variables)
```

drop_na_rows	<i>Drop rows that are NA in all specified variables.</i>
--------------	--

Description

drop_na_rows drops rows that are NA in all specified variables.

Usage

```
drop_na_rows(population, variables)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.

Value

population without rows that have NA in all specified variables.

Examples

```
population <- tibble::tibble(  
  Metadata_group = c(  
    "control", "control", "control", "control",  
    "experiment", "experiment", "experiment", "experiment"  
  ),  
  Metadata_batch = c("a", "a", "b", "b", "a", "a", "b", "b"),  
  AreaShape_Area = c(10, 12, NA, 16, 8, 8, 7, 7),  
  AreaShape_Length = c(2, 3, NA, NA, 4, 5, 1, 5)  
)  
variables <- c("AreaShape_Area", "AreaShape_Length")  
drop_na_rows(population, variables)
```

extract_subpopulations	<i>Extract subpopulations.</i>
------------------------	--------------------------------

Description

extract_subpopulations identifies clusters in the reference and population sets and reports the frequency of points in each cluster for the two sets.

Usage

```
extract_subpopulations(population, reference, variables, k)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
reference	tbl with grouping (metadata) and observation variables. Columns of population and reference should be identical.
variables	character vector specifying observation variables.
k	scalar specifying number of clusters.

Value

list containing clusters centers (subpop_centers), two normalized histograms specifying frequency of each clusters in population and reference (subpop_profiles), and cluster prediction and distance to the predicted cluster for all input data (population_clusters and reference_clusters).

Examples

```
data <- tibble::tibble(
  Metadata_group = c(
    "control", "control", "control", "control",
    "experiment", "experiment", "experiment", "experiment"
  ),
  AreaShape_Area = c(10, 12, NA, 16, 8, 8, 7, 7),
  AreaShape_Length = c(2, 3, NA, NA, 4, 5, 1, 5)
)
variables <- c("AreaShape_Area", "AreaShape_Length")
population <- dplyr::filter(data, Metadata_group == "experiment")
reference <- dplyr::filter(data, Metadata_group == "control")
extract_subpopulations(
  population = population,
  reference = reference,
  variables = variables,
  k = 3
)
```

generalized_log

Generalized log transform data.

Description

generalized_log transforms specified observation variables using $x = \log((x + \sqrt{x^2 + \text{offset}^2})/2)$.

Usage

```
generalized_log(population, variables, offset = 1)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.
offset	optional offset parameter for the transformation.

Value

transformed data of the same class as population.

Examples

```
population <- tibble::tibble(
  Metadata_Well = c("A01", "A02", "B01", "B02"),
  Intensity_DNA = c(8, 20, 12, 32)
)
variables <- c("Intensity_DNA")
generalized_log(population, variables)
```

generate_component_matrix

A sparse matrix for sparse random projection.

Description

generate_component_matrix generates the sparse random component matrix for performing sparse random projection. If density is the density of the sparse matrix and n_components is the size of the projected space, the elements of the random matrix are drawn from

Usage

```
generate_component_matrix(n_features, n_components, density)
```

Arguments

n_features	the dimensionality of the original space.
n_components	the dimensionality of the projected space.
density	the density of the sparse random matrix.

Details

$-\sqrt{1 / (\text{density} * \text{n_components})}$ with probability density / 2
 0 with probability $1 - \text{density}$
 $\sqrt{1 / (\text{density} * \text{n_components})}$ with probability density / 2

Value

A sparse random matrix of size (n_features, n_components).

Examples

```
generate_component_matrix(500, 100, 0.3)
```

normalize	<i>Normalize observation variables.</i>
-----------	---

Description

normalize normalizes observation variables based on the specified normalization method.

Usage

```
normalize(
  population,
  variables,
  strata,
  sample,
  operation = "standardize",
  ...
)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.
strata	character vector specifying grouping variables for grouping prior to normalization.
sample	tbl containing sample that is used by normalization methods to estimate parameters. sample has same structure as population. Typically, sample corresponds to controls in the experiment.
operation	optional character string specifying method for normalization. This must be one of the strings "standardize" (default), "robustize".
...	arguments passed to normalization operation

Value

normalized data of the same class as population.

Examples

```
suppressMessages(suppressWarnings(library(magrittr)))
population <- tibble::tibble(
  Metadata_group = c(
    "control", "control", "control", "control",
    "experiment", "experiment", "experiment", "experiment"
  ),
  Metadata_batch = c("a", "a", "b", "b", "a", "a", "b", "b"),
  AreaShape_Area = c(10, 12, 15, 16, 8, 8, 7, 7)
)
variables <- c("AreaShape_Area")
```

```
strata <- c("Metadata_batch")
sample <- population %>% dplyr::filter(Metadata_group == "control")
cytominer::normalize(population, variables, strata, sample, operation = "standardize")
```

replicate_correlation *Measure replicate correlation of variables.*

Description

‘replicate_correlation’ measures replicate correlation of variables.

Usage

```
replicate_correlation(
  sample,
  variables,
  strata,
  replicates,
  replicate_by = NULL,
  split_by = NULL,
  cores = NULL
)
```

Arguments

sample	tbl containing sample used to estimate parameters.
variables	character vector specifying observation variables.
strata	character vector specifying grouping variables for grouping prior to normalization.
replicates	number of replicates.
replicate_by	optional character string specifying column containing the replicate id.
split_by	optional character string specifying column by which to split the sample into batches; replicate correlations will be calculate per batch.
cores	optional integer specifying number of CPU cores used for parallel computing using doParallel.

Value

data frame of variable quality measurements

Examples

```

set.seed(123)
x1 <- rnorm(10)
x2 <- x1 + rnorm(10) / 100
y1 <- rnorm(10)
y2 <- y1 + rnorm(10) / 10
z1 <- rnorm(10)
z2 <- z1 + rnorm(10) / 1

batch <- rep(rep(1:2, each = 5), 2)

treatment <- rep(1:10, 2)

replicate_id <- rep(1:2, each = 10)

sample <-
  tibble::tibble(
    x = c(x1, x2), y = c(y1, y2), z = c(z1, z2),
    Metadata_treatment = treatment,
    Metadata_replicate_id = replicate_id,
    Metadata_batch = batch
  )

head(sample)

# `replicate_correlation` returns the median, min, and max
# replicate correlation (across batches) per variable
replicate_correlation(
  sample = sample,
  variables = c("x", "y", "z"),
  strata = c("Metadata_treatment"),
  replicates = 2,
  split_by = "Metadata_batch",
  replicate_by = "Metadata_replicate_id",
  cores = 1
)

```

sparse_random_projection

Reduce the dimensionality of a population using sparse random projection.

Description

sparse_random_projection reduces the dimensionality of a population by projecting the original data with a sparse random matrix. Generally more efficient and faster to compute than a Gaussian random projection matrix, while providing similar embedding quality.

Usage

```
sparse_random_projection(population, variables, n_components)
```

Arguments

population tbl with grouping (metadata) and observation variables.
 variables character vector specifying observation variables.
 n_components size of the projected feature space.

Value

Dimensionality reduced population.

Examples

```
population <- tibble::tibble(
  Metadata_Well = c("A01", "A02", "B01", "B02"),
  AreaShape_Area_DNA = c(10, 12, 7, 7),
  AreaShape_Length_DNA = c(2, 3, 1, 5),
  Intensity_DNA = c(8, 20, 12, 32),
  Texture_DNA = c(5, 2, 43, 13)
)
variables <- c("AreaShape_Area_DNA", "AreaShape_Length_DNA", "Intensity_DNA", "Texture_DNA")
sparse_random_projection(population, variables, 2)
```

 svd_entropy

Feature importance based on data entropy.

Description

svd_entropy measures the contribution of each feature in decreasing the data entropy.

Usage

```
svd_entropy(variables, sample, cores = NULL)
```

Arguments

variables character vector specifying observation variables.
 sample tbl containing sample used to estimate parameters.
 cores optional integer specifying number of CPU cores used for parallel computing using doParallel.

Value

data frame specifying the contribution of each feature in decreasing the data entropy. Higher values indicate more information.

Examples

```
sample <- tibble::tibble(
  AreaShape_MinorAxisLength = c(10, 12, 15, 16, 8, 8, 7, 7, 13, 18),
  AreaShape_MajorAxisLength = c(35, 18, 22, 16, 9, 20, 11, 15, 18, 42),
  AreaShape_Area = c(245, 151, 231, 179, 50, 112, 53, 73, 164, 529)
)
variables <- c("AreaShape_MinorAxisLength", "AreaShape_MajorAxisLength", "AreaShape_Area")
svd_entropy(variables, sample, cores = 1)
```

transform	<i>Transform observation variables.</i>
-----------	---

Description

transform transforms observation variables based on the specified transformation method.

Usage

```
transform(population, variables, operation = "generalized_log", ...)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.
operation	optional character string specifying method for transform. This must be one of the strings "generalized_log" (default), "whiten".
...	arguments passed to transformation operation.

Value

transformed data of the same class as population.

Examples

```
population <- tibble::tibble(
  Metadata_Well = c("A01", "A02", "B01", "B02"),
  Intensity_DNA = c(8, 20, 12, 32)
)
variables <- c("Intensity_DNA")
transform(population, variables, operation = "generalized_log")
```

variable_importance *Measure variable importance.*

Description

variable_importance measures importance of variables based on specified methods.

Usage

```
variable_importance(  
  sample,  
  variables,  
  operation = "replicate_correlation",  
  ...  
)
```

Arguments

sample	tbl containing sample used to estimate parameters.
variables	character vector specifying observation variables.
operation	optional character string specifying method for computing variable importance. Currently, only "replicate_correlation" (default) is implemented.
...	arguments passed to variable importance operation.

Value

data frame containing variable importance measures.

Examples

```
set.seed(123)  
x1 <- rnorm(10)  
x2 <- x1 + rnorm(10) / 100  
y1 <- rnorm(10)  
y2 <- y1 + rnorm(10) / 10  
z1 <- rnorm(10)  
z2 <- z1 + rnorm(10) / 1  
  
batch <- rep(rep(1:2, each = 5), 2)  
  
treatment <- rep(1:10, 2)  
  
replicate_id <- rep(1:2, each = 10)  
  
sample <-  
  tibble::tibble(  
    x = c(x1, x2), y = c(y1, y2), z = c(z1, z2),  
    Metadata_treatment = treatment,
```

```

    Metadata_replicate_id = replicate_id,
    Metadata_batch = batch
  )

head(sample)

# `replicate_correlation` returns the median, min, and max
# replicate correlation (across batches) per variable
variable_importance(
  sample = sample,
  variables = c("x", "y", "z"),
  operation = "replicate_correlation",
  strata = c("Metadata_treatment"),
  replicates = 2,
  split_by = "Metadata_batch",
  replicate_by = "Metadata_replicate_id",
  cores = 1
)

```

variable_select	<i>Select observation variables.</i>
-----------------	--------------------------------------

Description

variable_select selects observation variables based on the specified variable selection method.

Usage

```

variable_select(
  population,
  variables,
  sample = NULL,
  operation = "variance_threshold",
  ...
)

```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.
sample	tbl containing sample that is used by some variable selection methods. sample has same structure as population.
operation	optional character string specifying method for variable selection. This must be one of the strings "variance_threshold", "correlation_threshold", "drop_na_columns".
...	arguments passed to selection operation.

Value

variable-selected data of the same class as population.

Examples

```

# In this example, we use `correlation_threshold` as the operation for
# variable selection.

suppressMessages(suppressWarnings(library(magrittr)))
population <- tibble::tibble(
  x = rnorm(100),
  y = rnorm(100) / 1000
)

population %<>% dplyr::mutate(z = x + rnorm(100) / 10)

sample <- population %>% dplyr::slice(1:30)

variables <- c("x", "y", "z")

operation <- "correlation_threshold"

cor(sample)

# `x` and `z` are highly correlated; one of them will be removed

head(population)

futile.logger::flog.threshold(futile.logger::ERROR)

variable_select(population, variables, sample, operation) %>% head()

```

variance_threshold *Remove variables with near-zero variance.*

Description

variance_threshold returns list of variables that have near-zero variance.

Usage

```
variance_threshold(variables, sample)
```

Arguments

variables character vector specifying observation variables.
sample tbl containing sample used to estimate parameters.

Details

variance_threshold is a reimplementaion of caret::nearZeroVar, using the default values for freqCut and uniqueCut.

Value

character vector specifying observation variables to be excluded.

Examples

```
sample <- tibble::tibble(
  AreaShape_Area = c(10, 12, 15, 16, 8, 8, 7, 7, 13, 18),
  AreaShape_Euler = c(0, 0, 0, 0, 0, 0, 0, 0, 0, 0)
)
variables <- c("AreaShape_Area", "AreaShape_Euler")
variance_threshold(variables, sample)
```

whiten

Whiten data.

Description

whiten transforms specified observation variables by estimating a whitening transformation on a sample and applying it to the population.

Usage

```
whiten(population, variables, sample, regularization_param = 1)
```

Arguments

population	tbl with grouping (metadata) and observation variables.
variables	character vector specifying observation variables.
sample	tbl containing sample that is used by the method to estimate whitening parameters. sample has same structure as population. Typically, sample corresponds to controls in the experiment.
regularization_param	optional parameter used in whitening to offset eigenvalues to avoid division by zero.

Value

transformed data of the same class as population.

Examples

```
population <- tibble::tibble(
  Metadata_Well = c("A01", "A02", "B01", "B02"),
  Intensity_DNA = c(8, 20, 12, 32),
  Texture_DNA = c(5, 2, 43, 13)
)
variables <- c("Intensity_DNA", "Texture_DNA")
whiten(population, variables, population, 0.01)
```

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