

# Package: iSubGen (via r-universe)

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**Title** Integrative Subtype Generation

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**Description** Multi-data type subtyping, which is data type agnostic and accepts missing data. Subtyping is performed using intermediary assessments created with autoencoders and similarity calculations.

**Depends** R (>= 3.2.3)

**Imports** ConsensusClusterPlus, cluster (>= 1.14.4), keras, tensorflow, philentropy

**Suggests** knitr, rmarkdown

**VignetteBuilder** knitr

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apply.scaling	<i>Apply scaling factors</i>
---------------	------------------------------

---

## Description

Apply scaling factors prior to autoencoder

## Usage

```
apply.scaling(data.matrices, scaling.factors);
```

## Arguments

`data.matrices` list, where each element is a matrix. The list has one matrix for each data type to be scaled

`scaling.factors` list with two elements named: `"center"` and `"scale"`, and each element is a named numerical vector or a list of named numerical vectors. If `scaling.factors$center` or `scaling.factors$scale` are a list then each element needs to correspond to a one of the data matrices. Finally, the named numerical vectors should match the row and rownames from the corresponding data matrix.

## Details

The names for the data matrices and the center and scale lists all must match.

## Value

A list of matrices of the same format as the `data.matrices`

## Author(s)

Natalie Fox

## Examples

```
# Load molecular profiles for three data types and calculate scaling for each
example.molecular.data.dir <- paste0(path.package('iSubGen'),'exdata/');
molecular.data <- list();
scaling.factors <- list();
for(i in c('cna','snv','methy')) {
  # Load molecular profiles from example files saved
  # in the package as <data type>_profiles.txt
  molecular.data[[i]] <- load.molecular.aberration.data(
    paste0(example.molecular.data.dir,i,'_profiles.txt'),
    patients = c(paste0('EP00',1:9), paste0('EP0',10:30))
  );

  scaling.factors[[i]] <- list();

  scaling.factors[[i]]$center <- apply(molecular.data[[i]], 1, mean);
  scaling.factors[[i]]$scale <- apply(molecular.data[[i]], 1, sd);
}

# Example 1: Transform the molecular profiles by the scaling factors
scaled.molecular.data <- apply.scaling(molecular.data, scaling.factors);

# Example 2: Transform one of the data types based on the scaling factors
scaled.molecular.data2 <- apply.scaling(
  molecular.data[[1]],
  scaling.factors[[1]]
);
```

---

calculate.cis.matrix    *Calculate consensus integrative correlation matrix*

---

## Description

Calculate consensus pairwise correlations between patient distances

## Usage

```
calculate.cis.matrix(data.types, data.matrices, dist.metrics,
  correlation.method = "spearman", filter.to.common.patients = FALSE,
  patients.to.return = NULL, patients.for.correlations = NULL,
  patient.proportion = 0.8, feature.proportion = 1, num.iterations = 10,
  print.intermediary.similarity.matrices.to.file = TRUE, print.dir = '.',
  patient.proportion.seeds = seq(1,num.iterations),
  feature.proportion.seeds = seq(1,num.iterations))
```

## Arguments

**data.types**        vector of the IDs for the different data types that are the names of the lists for the data.matrices and dist.metrics

`data.matrices` list of the matrices with features (rows) by patients (columns)

`dist.metrics` list of the distance metrics for comparing patient profiles. ex. euclidean. Options are from `philentropy::distance`

`correlation.method` specifies the type of correlation for similarity comparison. Options are `pearson`, `spearman` or `kendall`.

`filter.to.common.patients` logical, where `TRUE` indicates to filter out patients that don't have all data types

`patients.to.return` vector of patients to calculate CIS for. For example, this is the testing cohort patients when calculating CIS for the testing cohort using the training cohort patients. If `NULL` all patients/columns will be used.

`patients.for.correlations` vector of patients to use to calculate the similarities. For example, this would be the training cohort patients when calculating CIS for the testing cohort. If `NULL` all patients/columns will be used.

`patient.proportion` proportion of `patients.for.correlations` to sample for each iteration (sampled without replacement).

`feature.proportion` proportion of the features to sample for each iteration (sampled without replacement).

`num.iterations` number of iterations to take the median from

`print.intermediary.similarity.matrices.to.file` logical, where `TRUE` indicates that created intermediary integrative similarity matrix from each iteration should be printed to file

`print.dir` directory for where to print the intermediary similarity matrices to file

`patient.proportion.seeds` vector of scalars of the length `num.iterations` specifying the seeds used for random sampling for selecting the patient subsets at each iteration

`feature.proportion.seeds` vector of scalars of the length `num.iterations` specifying the seeds used for random sampling for selecting the feature subsets at each iteration

**Value**

CIS matrix where rows are patients and columns are pairs of data types

**Author(s)**

Natalie Fox

**Examples**

```
# Load molecular profiles for three data types from example files saved
# in the package as <data type>_profiles.txt
example.molecular.data.dir <- paste0(path.package('iSubGen'), '/exdata/');
```

```
molecular.data <- list();
for(i in c('cna','snv','methy')) {
  molecular.data[[i]] <- load.molecular.aberration.data(
    paste0(example.molecular.data.dir,i,'_profiles.txt'),
    patients = c(paste0('EP00',1:9), paste0('EP0',10:30))
  );
}

# Example 1: calculate the consensus integrative similarity (CIS) matrix
corr.matrix <- calculate.cis.matrix(
  data.types = names(molecular.data),
  data.matrices = molecular.data,
  dist.metrics = list(
    cna = 'euclidean',
    snv = 'euclidean',
    methy = 'euclidean'
  ),
  print.intermediary.similarity.matrices.to.file = FALSE
);

# Example 2: calculate the CIS matrix for patients EP001 through EP009 in relation
# to patients EP010 through EP030 meaning the profile of EP001 is correlated to
# the profiles of EP010 through EP030 so when assessing new patients, they can be
# compared to the training profiles
corr.matrix2 <- calculate.cis.matrix(
  data.types = names(molecular.data),
  data.matrices = molecular.data,
  dist.metrics = list(
    cna = 'euclidean',
    snv = 'euclidean',
    methy = 'euclidean'
  ),
  patients.to.return = paste0('EP00',1:9),
  patients.for.correlations = paste0('EP0',10:30),
  print.intermediary.similarity.matrices.to.file = FALSE
);

# Example 3: Adjusting the proportion of the features that will be used to correlate
# the patient profiles
corr.matrix3 <- calculate.cis.matrix(
  data.types = names(molecular.data),
  data.matrices = molecular.data,
  dist.metrics = list(
    cna = 'euclidean',
    snv = 'euclidean',
    methy = 'euclidean'
  ),
  patients.to.return = paste0('EP00',1:9),
  patients.for.correlations = paste0('EP0',10:30),
  feature.proportion = 0.6,
  print.intermediary.similarity.matrices.to.file = FALSE
);
```

---

```
calculate.integrative.similarity.matrix
```

*Calculate integrative similarity matrix*

---

### Description

Calculate pairwise correlations between patient distances

### Usage

```
calculate.integrative.similarity.matrix(data.types, data.matrices, dist.metrics,  
correlation.method = "spearman", filter.to.common.patients = FALSE,  
patients.to.return = NULL, patients.for.correlations = NULL)
```

### Arguments

`data.types` vector, where each element is a data type ID matching the names in `data.matrices` and `dist.metrics`

`data.matrices` list, where each element is a matrix with features as rows and patients as columns

`dist.metrics` list, where each element is the distance metric to use for comparing patient profiles. ex. euclidean. Options are from `philterentropy::distance`

`correlation.method` specifies the type of correlation. Options are pearson, spearman or kendall.

`filter.to.common.patients` logical, where TRUE indicates to filter out patients that don't have all data types

`patients.to.return` vector, where each element a patient ID specifying the patients to calculate integrative similarity for. For example, this is the testing cohort patients when calculating integrative similarity for the testing cohort using the training cohort patients. If NULL all patients/columns will be used.

`patients.for.correlations` vector, where each element a patient ID specifying the patients to use to calculate the similarities. For example, this would be the training cohort patients when calculating integrative similarity for the testing cohort. If NULL all patients/columns will be used.

### Value

matrix where rows are patients and columns are pairs of data types

### Author(s)

Natalie Fox

**Examples**

```

# Load molecular profiles for three data types from example files saved
# in the package as <data type>_profiles.txt
example.molecular.data.dir <- paste0(path.package('iSubGen'), '/exdata/');
molecular.data <- list();
for(i in c('cna', 'snv', 'methy')) {
  molecular.data[[i]] <- load.molecular.aberration.data(
    paste0(example.molecular.data.dir, i, '_profiles.txt'),
    patients = c(paste0('EP00', 1:9), paste0('EP0', 10:30))
  );
}

# Example 1: calculate integrative similarity between pairs of CNA, coding SNVs, methylation data
corr.matrix <- calculate.integrative.similarity.matrix(
  data.types = names(molecular.data),
  data.matrices = molecular.data,
  dist.metrics = list(
    cna = 'euclidean',
    snv = 'euclidean',
    methy = 'euclidean'
  )
);

# Example 2: calculate the integrative similarity for patients EP001 through EP009
# in relation to patients EP010 through EP030 meaning the profile of EP001 is
# correlated to the profiles of EP010 through EP030 so when assessing new patients,
# they can be compared to the training profiles
corr.matrix2 <- calculate.integrative.similarity.matrix(
  data.types = names(molecular.data),
  data.matrices = molecular.data,
  dist.metrics = list(
    cna = 'euclidean',
    snv = 'euclidean',
    methy = 'euclidean'
  ),
  patients.to.return = paste0('EP00', 1:9),
  patients.for.correlations = paste0('EP0', 10:30)
);

# Example 3: Calculate integrative similarity between CNA and methylation data
corr.matrix3 <- calculate.integrative.similarity.matrix(
  data.types=names(molecular.data)[c(1,3)],
  data.matrices=molecular.data[c(1,3)],
  dist.metrics=list(
    cna='euclidean',
    snv='euclidean',
    methy='euclidean'
  )[c(1,3)],
  patients.to.return=paste0('EP00', 1:9),
  patients.for.correlations=paste0('EP0', 10:30)
);

```

---

calculate.scaling      *Calculate scaling factors*

---

### Description

Calculate scaling factors

### Usage

```
calculate.scaling(data.matrices);
```

### Arguments

`data.matrices` list, where each element is a matrix. The list has one matrix for each data type to be scaled

### Details

The names for the data matrices and the center and scale lists all must match.

### Value

a list with two elements named: `"center"` and `"scale"`, and each of these element is a named numerical vector or a list of named numerical vectors. If `scaling.factors$center` or `scaling.factors$scale` are a list then each element will correspond to a one of the data matrices. Finally, the named numerical vectors will match the row and rownames from the data matrices.

### Author(s)

Natalie Fox

### Examples

```
# Load molecular profiles for three data types from example files saved
# in the package as <data type>_profiles.txt
example.molecular.data.dir <- paste0(path.package('iSubGen'), '/exdata/');
molecular.data <- list();
for(i in c('cna', 'snv', 'methy')) {
  molecular.data[[i]] <- load.molecular.aberration.data(
    paste0(example.molecular.data.dir, i, '_profiles.txt'),
    patients = c(paste0('EP00', 1:9), paste0('EP0', 10:30))
  );
}

# Example 1: Calculate scaling factors for all three data types
scaling.factors <- calculate.scaling(molecular.data);

# Example 2: Calculate scaling factors for only the methylation data
scaling.factors2 <- calculate.scaling(molecular.data[['methy']]);
```



---

cluster.patients	<i>Clustering to find patient subtypes</i>
------------------	--

---

**Description**

A wrapper function for using consensus clustering to subtype patients

**Usage**

```
cluster.patients(data.matrix, distance.metric, parent.output.dir,
new.result.dir, subtype.table.file = NULL, max.num.subtypes = 12,
clustering.reps = 1000, proportion.features = 0.8, proportion.patients = 0.8,
verbose = FALSE, consensus.cluster.write.table = TRUE);
```

**Arguments**

data.matrix	matrix with patients as rows and features as columns
distance.metric	distance metric for comparing patient profiles. ex. euclidean
parent.output.dir	directory where the consensus clustering function will create a directory of results
new.result.dir	directory name for consensus clustering results
subtype.table.file	filename for subtype assignment table for different number of clusters
max.num.subtypes	maximum number of clusters to separate patients into
clustering.reps	number of subsamples for consensus clustering function
proportion.features	proportion of features to sample for each clustering iteration
proportion.patients	proportion of patients to sample for each clustering iteration
verbose	logical, where TRUE indicates to print messages to the screen to indicate progress
consensus.cluster.write.table	logical, where TRUE indicates for the ConsensusClusterPlus function to writeTable

**Value**

consensus_cluster_result	consensus clustering function return value
subtype_table	the table written to subtype.table.file

**Author(s)**

Natalie Fox

**Examples**

```
## Not run:

# For this example instead of clustering CIS and IRF matrices,
# create a data matrix to see how the function works without
# running through the whole iSubGen process.
# This example is created with to have 4 distinct clusters
set.seed(5);
ex.matrix <- matrix(
  c(
    sample(c(0,1), 30, replace = TRUE), rep(1,75), rep(0,25),
    sample(c(0,1), 30, replace = TRUE), rep(1,75), rep(0,25),
    sample(c(0,1), 30, replace = TRUE), rep(1,75), rep(0,25),
    sample(c(0,1), 30, replace = TRUE), rep(1,100),
    sample(c(0,1), 30, replace = TRUE), rep(1,100),
    sample(c(0,1), 30, replace = TRUE), rep(1,100),
    sample(c(0,1), 30, replace = TRUE), rep(0,100),
    sample(c(0,1), 30, replace = TRUE), rep(0,100),
    sample(c(0,1), 30, replace = TRUE), rep(0,100),
    sample(c(0,1), 30, replace = TRUE), rep(0,75), rep(1,25),
    sample(c(0,1), 30, replace = TRUE), rep(0,75), rep(1,25),
    sample(c(0,1), 30, replace = TRUE), rep(0,75), rep(1,25)
  ),
  nrow=130);
rownames(ex.matrix) <- paste0('gene',1:130);
colnames(ex.matrix) <- paste0('patient',LETTERS[1:12]);

# Use Consensus clustering to subtype the patient profiles
subtyping.results <- cluster.patients(
  data.matrix = ex.matrix,
  distance.metric = 'euclidean',
  parent.output.dir = './',
  new.result.dir = 'example_subtyping',
  max.num.subtypes = 6,
  clustering.reps = 50,
  consensus.cluster.write.table = FALSE
);

## End(Not run)
```

---

combine.integrative.features

*Combine iSubGen integrative features*

---

**Description**

Combine a independent reduced features matrix (ex. from autoencoders) and pairwise integrative similarity matrices into one integrative feature matrix.

**Usage**

```
combine.integrative.features(irf.matrix, cis.matrix,
  irf.rescale.recenter = NA, cis.rescale.recenter = NA,
  irf.rescale.denominator = NA, cis.rescale.denominator = NA,
  irf.weights = rep(1, ncol(irf.matrix)),
  cis.weights = rep(1, ncol(cis.matrix)))
```

**Arguments**

<code>irf.matrix</code>	matrix of independent reduced features with patients as rows and features as columns
<code>cis.matrix</code>	matrix of consensus integrative similarity or integrative similarity features with patients as rows and features as columns
<code>irf.rescale.recenter</code>	either NA, "mean", a single number or a vector of numbers of length equal to the number of columns of irf
<code>cis.rescale.recenter</code>	either NA, "mean", a single number or a vector of numbers of length equal to the number of columns of cis
<code>irf.rescale.denominator</code>	either NA, "sd", a single number or a vector of numbers of length equal to the number of columns of irf
<code>cis.rescale.denominator</code>	either NA, "sd", a single number or a vector of numbers of length equal to the number of columns of cis
<code>irf.weights</code>	single number or vector of numbers of length equal to the number of columns of irf
<code>cis.weights</code>	single number or vector of numbers of length equal to the number of columns of cis

**Details**

The recenter values determine the how column centering is performed. If NA, no recentering is done. If the values equal "mean", then the mean of each column will be used. Otherwise, the numeric values specified will be used. The denominator values determine how column scaling is performed. If NA, no recentering is done. If the denominator values equal "sd", then the standard deviation of each column will be used. Otherwise, the numeric values specified will be used. The values used are returned by the function along with the compressed feature matrix to be recorded for reproducibility purposes.

**Value**

<code>integrative.feature.matrix</code>	a matrix of compressed features with patients as rows and features as columns
<code>irf.rescale.recenter</code>	a numeric vector with length equal to the number of columns of irf
<code>cis.rescale.recenter</code>	a numeric vector with length equal to the number of columns of cis

`irf.rescale.denominator`  
 a numeric vector with length equal to the number of columns of `irf`

`cis.rescale.denominator`  
 a numeric vector with length equal to the number of columns of `cis`

`irf.weights` a numeric vector with length equal to the number of columns of `irf`

`cis.weights` a numeric vector with length equal to the number of columns of `cis`

**Author(s)**

Natalie Fox

**Examples**

```
# Create matrices for combining
irf.matrix <- matrix(runif(25*4), ncol = 4);
rownames(irf.matrix) <- c(paste0('EP00',1:9), paste0('EP0',10:25));
cis.matrix <- matrix(runif(25*6), ncol=6);
rownames(cis.matrix) <- c(paste0('EP00',1:9), paste0('EP0',10:25));

# Example 1: Join the matrices without any weighting adjustments
isubgen.feature.matrix <- combine.integrative.features(
  irf.matrix,
  cis.matrix
)$integrative.feature.matrix;

# Example 2: Combine matrices after scaling each column by subtracting the mean
# and dividing by the standard deviation of the column
isubgen.feature.matrix.rescaled.result <- combine.integrative.features(
  irf.matrix,
  cis.matrix,
  irf.rescale.recenter = 'mean',
  cis.rescale.recenter = 'mean',
  irf.rescale.denominator = 'sd',
  cis.rescale.denominator = 'sd'
);
isubgen.feature.matrix.2 <- isubgen.feature.matrix.rescaled.result$integrative.feature.matrix;

# Example 3: Combine matrices
isubgen.feature.matrix.reweighted.result <- combine.integrative.features(
  irf.matrix,
  cis.matrix,
  irf.weights = 1/4,
  cis.weights = 1/6
);
isubgen.feature.matrix.3 <- isubgen.feature.matrix.reweighted.result$integrative.feature.matrix;
```

---

create.autoencoder      *Create an autoencoder for dimensionality reduction*

---

### Description

Create an autoencoder for dimensionality reduction using keras and tensorflow packages

### Usage

```
create.autoencoder(data.type, data.matrix, encoder.layers.node.nums = c(15,2),
  autoencoder.activation = 'tanh', optimization.loss.function = 'mean_squared_error',
  model.file.output.dir = '.')
```

### Arguments

`data.type`            data type ID. The ID will be used for naming the output file

`data.matrix`        matrix with data features as rows and patients as columns

`encoder.layers.node.nums`  
                      vector with the number of nodes for each layer when the reducing the feature dimensions within the autoencoder. The autoencoder will be made symmetrically so the number of nodes in each layer will be used in reverse, not repeating the last layer to re encode the features in the autoencoder

`autoencoder.activation`  
                      activation function to use in the autoencoder

`optimization.loss.function`  
                      loss function used for optimization while fitting the autoencoder

`model.file.output.dir`  
                      file location for the autoencoder file

### Value

`autoencoder`        the autoencoder created by the keras package

`autoencoder.file`  
                      the hdf5 file that the model was saved in and can be loaded from

### Author(s)

Natalie Fox

### Examples

```
## Not run:

example.molecular.data.dir <- paste0(path.package('iSubGen'), '/exdata/');

ae.result <- create.autoencoder(
  data.type = 'cna',
```

```
data.matrix = load.molecular.aberration.data(  
  paste0(example.molecular.data.dir, 'cna_profiles.txt'),  
  patients = c(paste0('EP00',1:9), paste0('EP0',10:30))  
),  
encoder.layers.node.nums = c(15,5,2)  
);  
  
## End(Not run)
```

---

```
create.autoencoder.irf.matrix
```

*Create matrix of independent reduced features*

---

## Description

Create matrix of independent reduced features using autoencoders

## Usage

```
create.autoencoder.irf.matrix(data.types, data.matrices,  
autoencoders, filter.to.common.patients = FALSE,  
patients.to.return = NULL)
```

## Arguments

<code>data.types</code>	vector, where each element is a data type ID matching the names in <code>data.matrices</code> and <code>dist.metrics</code>
<code>data.matrices</code>	list, where each element is a matrix with features as rows and patients as columns
<code>autoencoders</code>	list, where each element is an autoencoder corresponding to each data type. Can be either a keras autoencoder object or the file where the autoencoder was saved.
<code>filter.to.common.patients</code>	logical, where TRUE indicates to filter out patients that don't have all data types.
<code>patients.to.return</code>	vector of patients to return correlations for. If NULL all patients/columns will be used.

## Value

matrix where rows are patients and columns are pairs of data types

## Author(s)

Natalie Fox

**Examples**

```
## Not run:

# Load three data types and create an autoencoder for each
example.molecular.data.dir <- paste0(path.package('iSubGen'),'exdata/');
molecular.data <- list();
ae.result <- list();
for(i in c('cna','snv','methy')) {
  molecular.data[[i]] <- load.molecular.aberration.data(
    paste0(example.molecular.data.dir,i,'_profiles.txt'),
    patients = c(paste0('EP00',1:9), paste0('EP0',10:30))
  );
  ae.result[[i]] <- create.autoencoder(
    data.type = i,
    data.matrix = molecular.data[[i]],
    encoder.layers.node.numms = c(10,2)
  )$autoencoder;
}

# Create a matrix of the bottleneck layers
irf.matrix <- create.autoencoder.irf.matrix(
  data.types = names(molecular.data),
  data.matrices = molecular.data,
  autoencoders = ae.result
);

## End(Not run)
```

---

```
load.molecular.aberration.data
      Load molecular aberration data
```

---

**Description**

Load the molecular aberration profiles/feature annotation

**Usage**

```
load.molecular.aberration.data(file, patients = NULL, annotation.fields = NULL);
```

**Arguments**

file	file name of the matrix containing molecular and annotation data. If it does not contain an <code>_absolute_</code> path, the file name is <code>_relative_</code> to the current working directory, <code>'getwd()'</code> as in <code>read.table</code> .
patients	vector of patients IDs. Must match colnames from aberration file
annotation.fields	vector referencing the column names for the feature annotation columns

## Details

The `annotation.fields` argument will look for any colnames which contain the values specified in `annotation.fields` and then the column will be renamed to the value that matched from `annotation.fields`.

## Value

If the `patients` argument is specified then the patient molecular aberration profiles are returned. If the `annotation.fields` argument is specified then the feature annotation is returned. If both are specified then the two matrices are returned in a list. If neither is specified then the entire matrix with the mix of patients and annotation is returned.

## Author(s)

Natalie Fox

## Examples

```
example.aberration.data <- paste0(
  path.package('iSubGen'),
  '/exdata/cna_profiles.txt'
);

# Load the CNA profiles for patients EP001 through EP030
cna.profiles <- load.molecular.aberration.data(
  example.aberration.data,
  patients = c(paste0('EP00',1:9), paste0('EP0',10:30))
);

# Load feature annotation for the CNA data
cna.annotation <- load.molecular.aberration.data(
  example.aberration.data,
  annotation.fields = c('gene','start','end')
);
```

---

`read.scaling.factors` *Read scaling factors from file*

---

## Description

Read scaling factors from file

## Usage

```
read.scaling.factors(scaling.factor.files.dir,data.types);
```



**Arguments**

scaling.factor.files.dir  
the directory where the files were saved

data.types a vector of the data types with saved scaling factors

**Details**

One scale and one center file is saved per data type

**Value**

a list with a key \"center\" list and a key \"scale\" list. The center and scale list keys match the data.matrices list keys

**Author(s)**

Natalie Fox

**Examples**

```
# Get the path for the scaling provided in this R package
example.molecular.data.dir <- paste0(path.package('iSubGen'), '/exdata/');

# Example #1: reading scaling factors for a single data type
scaling.factors <- read.scaling.factors(example.molecular.data.dir, 'cna');

# Example #2: reading scaling factors for multiple data types
scaling.factors <- read.scaling.factors(example.molecular.data.dir, c('cna', 'snv', 'methy'));
```

---

write.scaling.factors *Write scaling factors to file*

---

**Description**

Write scaling factors to file

**Usage**

```
write.scaling.factors(scaling.factors, scaling.factor.files.dir=NULL)
```

**Arguments**

scaling.factors  
list with the scaling factors created by calculate.scaling

scaling.factor.files.dir  
directory to output scaling factor files

**Details**

Creates two files for each data type key. One file for the recentering values and one file for the rescaling values. Files have the names `<data type>_gene_recenter.txt` or `<data type>_gene_rescale.txt`

**Value**

No return value, called for side effects

**Author(s)**

Natalie Fox

**Examples**

```
## Not run:

# load the aberration profiles for three data types
example.molecular.data.dir <- paste0(path.package('iSubGen'), '/exdata/');
molecular.data <- list();
for(i in c('cna', 'snv', 'methy')) {
  molecular.data[[i]] <- load.molecular.aberration.data(
    paste0(example.molecular.data.dir, i, '_profiles.txt'),
    patients = c(paste0('EP00', 1:9), paste0('EP0', 10:30))
  );
}

# calculate scaling factors for all three data types
scaling.factors <- calculate.scaling(molecular.data);

# save the scaling factors to file
write.scaling.factors(scaling.factors);

## End(Not run)
```

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