Package 'ADImpute'

November 28, 2024

Type Package

Title Adaptive Dropout Imputer (ADImpute)

Version 1.16.0

Description Single-cell RNA sequencing (scRNA-seq) methods are typically unable to quantify the expression levels of all genes in a cell, creating a need for the computational prediction of missing values ('dropout imputation'). Most existing dropout imputation methods are limited in the sense that they exclusively use the scRNA-seq dataset at hand and do not exploit external gene-gene relationship information. Here we propose two novel methods: a gene regulatory network-based approach using gene-gene relationships learnt from external data and a baseline approach corresponding to a sample-wide average. ADImpute can implement these novel methods and also combine them with existing imputation methods (currently supported: DrImpute, SAVER). ADImpute can learn the best performing method per gene and combine the results from different methods into an ensemble.

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Encoding UTF-8

LazyData true

Depends R (>= 4.0)

Imports checkmate, BiocParallel, data.table, DrImpute, kernlab, MASS, Matrix, methods, rsvd, S4Vectors, SAVER, SingleCellExperiment, stats, SummarizedExperiment, utils

Suggests BiocStyle, knitr, rmarkdown, testthat

biocViews GeneExpression, Network, Preprocessing, Sequencing, SingleCell, Transcriptomics

RoxygenNote 7.1.1

VignetteBuilder knitr

BugReports https://github.com/anacarolinaleote/ADImpute/issues

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

git branch RELEASE 3 20

git_last_commit fc250fa

git_last_commit_date 2024-10-29

Repository Bioconductor 3.20

Date/Publication 2024-11-28

2 Contents

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Contents

Index

ArrangeData
CenterData
CheckArguments_Impute
ChooseMethod
Combine
ComputeMSEGenewise
CreateArgCheck
CreateTrainData
DataCheck_Matrix
DataCheck_Network
DataCheck_SingleCellExperiment
DataCheck_TrLength
demo_data
demo_net
demo_sce
EvaluateMethods
GetDropoutProbabilities
HandleBiologicalZeros
Impute
ImputeBaseline
ImputeDrImpute
ImputeNetParallel
ImputeNetwork
ImputeNPDropouts
ImputePredictiveDropouts
ImputeSAVER 2
MaskData
MaskerPerGene
network.coefficients
NormalizeRPM
NormalizeTPM
PseudoInverseSolution_percell
ReadData
ReturnChoice
ReturnOut
SetBiologicalZeros
SplitData
transcript_length
WriteCSV
WriteTXT

28

ArrangeData 3

Description

ArrangeData finds common genes to the network and provided data and limits both datasets to these

Usage

```
ArrangeData(data, net.coef = NULL)
```

Arguments

data matrix with entries equal to zero to be imputed (genes as rows and samples as

columns)

net.coef matrix; object containing network coefficients

Value

list; data matrix, network coefficients matrix and intercept for genes common between the data matrix and the network

CenterData	Data centering	

Description

CenterData centers expression of each gene at 0

Usage

CenterData(data)

Arguments

data matrix of gene expression to be centered row-wise (genes as rows and samples

as columns)

Value

list; row-wise centers and centered data

4 ChooseMethod

CheckArguments_Impute Argument check to Impute()

Description

CheckArguments_Impute checks whether the arguments passed to Impute are correct.

Usage

```
CheckArguments_Impute(data, method.choice, do, tr.length, labels, cell.clusters, true.zero.thr, drop_thre)
```

Arguments

matrix; raw counts (genes as rows and samples as columns) data method.choice character; best performing method in training data for each gene do character; choice of methods to be used for imputation. Currently supported methods are 'Baseline', 'DrImpute', 'Network', and 'Ensemble'. Defaults to 'Ensemble'. Not case-sensitive. Can include one or more methods. Nonsupported methods will be ignored. tr.length matrix with at least 2 columns: 'hgnc_symbol' and 'transcript_length' labels character; vector specifying the cell type of each column of data cell.clusters integer; number of cell subpopulations true.zero.thr if set to NULL (default), no true zero estimation is performed. Set to numeric value between 0 and 1 for estimation. Value corresponds to the threshold used to determine true zeros: if the probability of dropout is lower than true.zero.thr, the imputed entries are set to zero. drop_thre numeric; between 0 and 1 specifying the threshold to determine dropout values

Value

NULL object

ChooseMethod Method choice per gene	ChooseMethod	Method choice per gene	
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Description

ChooseMethod determines the method for dropout imputation based on performance on each gene in training data

Usage

```
ChooseMethod(real, masked, imputed, write.to.file = TRUE)
```

Combine 5

Arguments

real matrix; original gene expression data, i.e. before masking (genes as rows and

samples as columns)

masked matrix, logical indicating which entries were masked (genes as rows and sam-

ples as columns)

imputed list; list of matrices with imputation results for all considered methods

write.to.file logical; should the output be written to a file?

Details

The imputed values are compared to the real ones for every masked entry in real. The Mean Squared Error is computed for all masked entries per gene and the method with the best performance is chosen for each gene.

Value

character; best performing method in the training set for each gene

See Also

ComputeMSEGenewise

Combine	Combine imputation methods

Description

Combine imputation methods

Usage

```
Combine(data, imputed, method.choice, write = FALSE)
```

Arguments

data matrix with entries equal to zero to be imputed, already normalized (genes as

rows and samples as columns)

imputed list; list of matrices with imputation results for all considered methods method.choice named character; vector with the best performing method per gene write logical; should a file with the imputation results be written?

Details

Combines imputation results from all methods according to training results provided in method. choice

Value

matrix; imputation results combining the best performing method per gene

6 CreateArgCheck

EGenewise Computation of MSE per gene

Description

ComputeMSEGenewise computes the MSE of dropout imputation for a given gene.

Usage

```
ComputeMSEGenewise(real, masked, imputed, baseline)
```

Arguments

numeric; vector of original expression of a given gene (before masking)
masked logical; vector indicating which entries were masked for a given gene

imputed matrix; imputation results for a given imputation method

baseline logical; is this baseline imputation?

Value

MSE of all imputations indicated by masked

CreateArgCheck	Argument check	

Description

CreateArgCheck creates tests for argument correctness.

Usage

```
CreateArgCheck(missing = NULL, match = NULL, acceptable = NULL,
null = NULL)
```

Arguments

missing	named list: logical.	Name corresponds to	variable name.	and corresponding

entry to whether it was missing from the function call.

match named list. Name corresponds to variable name, and corresponding entry to its

value.

acceptable named list. Name corresponds to variable name, and corresponding entry to its

acceptable values.

null named list; logical. Name corresponds to variable name, and corresponding

entry to whether it was NULL in the function call.

Value

argument check object.

Create TrainData 7

CreateTrainData	Preparation of training data for method evaluation	

Description

CreateTrainingData selects a subset of cells to use as training set and sets a portion (mask) of the non-zero entries in each row of the subset to zero

Usage

```
CreateTrainData(data, train.ratio = .7, train.only = TRUE, mask = .1,
write = FALSE)
```

Arguments

data matrix; raw counts (genes as rows and samples as columns)

train.ratio numeric; ratio of the samples to be used for training

train.only logical; if TRUE define only a training dataset, if FALSE writes both training

and validation sets (defaults to TRUE)

numeric; ratio of total non-zero samples to be masked per gene (defaults to .1)

write logical; should the output be written to a file?

Value

list with resulting matrix after subsetting and after masking

DataCheck_Matrix	Data check (matrix)

Description

DataCheck_Matrix tests for potential format and storage issues with matrices. Helper function to ADImpute.

Usage

```
DataCheck_Matrix(data)
```

Arguments

data object to check

Value

data object with needed adjustments

DataCheck_Network

Data check (network)

Description

DataCheck_Network tests for potential format and storage issues with the network coefficient matrix. Helper function to ADImpute.

Usage

DataCheck_Network(network)

Arguments

network

data object containing matrix coefficients

Value

network data object with needed adjustments

DataCheck_SingleCellExperiment

Data check (SingleCellExperiment)

Description

DataCheck_SingleCellExperiment tests for existence of the appropriate assays in sce. Helper function to ADImpute.

Usage

DataCheck_SingleCellExperiment(sce, normalized = TRUE)

Arguments

sce

SingleCellExperiment; data for normalization or imputation

normalized

logical; is the data expected to be normalized?

Value

NULL object.

DataCheck_TrLength 9

DataCheck_TrLength

Data check (transcript length)

Description

DataCheck_TrLength tests for potential format and storage issues with the object encoding transcript length, for e.g. TPM normalization. Helper function to ADImpute.

Usage

```
DataCheck_TrLength(trlength)
```

Arguments

trlength

data object containing transcript length information

Value

transcript length object with needed adjustments

demo_data

Small dataset for example purposes

Description

A small dataset to use on vignettes and examples (50 cells).

Usage

demo_data

Format

matrix; a subset of the Grun pancreas dataset, obtained with the scRNAseq R package, to use in the vignette and examples.

References

Grun D et al. (2016). De novo prediction of stem cell identity using single-cell transcriptome data. Cell Stem Cell 19(2), 266-277.

10 demo_sce

demo_net

Small regulatory network for example purposes

Description

Subset of the Gene Regulatory Network used by ADImpute's Network imputation method.

Usage

 $demo_net$

Format

matrix; subset of the Gene Regulatory Network installed along with ADImpute.

demo_sce

Small dataset for example purposes

Description

A small dataset to use on vignettes and examples (50 cells).

Usage

demo_sce

Format

SingleCellExperiment; a subset of the Grun pancreas dataset, obtained with the $scRNAseq\ R$ package, to use in the vignette and examples.

References

Grun D et al. (2016). De novo prediction of stem cell identity using single-cell transcriptome data. Cell Stem Cell 19(2), 266-277.

EvaluateMethods 11

EvaluateMethods	Imputation method evaluation on training set	

Description

EvaluateMethods returns the best-performing imputation method for each gene in the dataset

Usage

```
EvaluateMethods(data, sce = NULL, do = c('Baseline', 'DrImpute',
'Network'), write = FALSE, train.ratio = .7, train.only = TRUE,
mask.ratio = .1, outdir = getwd(), scale = 1, pseudo.count = 1,
labels = NULL, cell.clusters = 2, drop_thre = NULL, type = 'count',
cores = BiocParallel::bpworkers(BPPARAM),
BPPARAM = BiocParallel::SnowParam(type = "SOCK"),
net.coef = ADImpute::network.coefficients, net.implementation = 'iteration',
tr.length = ADImpute::transcript_length, bulk = NULL, ...)
```

Arguments

data	matrix; normalized counts, not logged (genes as rows and samples as columns)
sce	SingleCellExperiment; normalized counts and associated metadata.
do	character; choice of methods to be used for imputation. Currently supported methods are 'Baseline', 'DrImpute' and 'Network'. Not case-sensitive. Can include one or more methods. Non- supported methods will be ignored.
write	logical; write intermediary and imputed objects to files?
train.ratio	numeric; ratio of samples to be used for training
train.only	logical; if TRUE define only a training dataset, if FALSE writes and returns both training and validation sets (defaults to TRUE)
mask.ratio	numeric; ratio of samples to be masked per gene
outdir	character; path to directory where output files are written. Defaults to working directory
scale	integer; scaling factor to divide all expression levels by (defaults to 1)
pseudo.count	integer; pseudo-count to be added to expression levels to avoid log(0) (defaults to 1)
labels	character; vector specifying the cell type of each column of data
cell.clusters	integer; number of cell subpopulations
drop_thre	numeric; between 0 and 1 specifying the threshold to determine dropout values
type	A character specifying the type of values in the expression matrix. Can be 'count' or 'TPM'
cores	integer; number of cores used for paralell computation
BPPARAM	parallel back-end to be used during parallel computation. See BiocParallelParam-class.
net.coef	matrix; network coefficients. Please provide if you don't want to use ADImpute's network model. Must contain one first column 'O' acconting for the intercept of the model and otherwise be an adjacency matrix with hgnc_symbols in rows and columns. Doesn't have to be squared. See ADImpute::demo_net

for a small example.

net.implementation

character; either 'iteration', for an iterative solution, or 'pseudoiny', to use Moore-Penrose pseudo-inversion as a solution. 'pseudoiny' is not advised for big data

big data.

tr.length matrix with at least 2 columns: 'hgnc_symbol' and 'transcript_length' bulk vector of reference bulk RNA-seq, if available (average across samples)

... additional parameters to pass to network-based imputation

Details

For each gene, a fraction (mask.ratio) of the quantified expression values are set to zero and imputed according to 3 different methods: scImpute, baseline (average gene expression across all cells) or a network-based method. The imputation error is computed for each of the values in the original dataset that was set to 0, for each method. The method resulting in a lowest imputation error for each gene is chosen.

Value

- if sce is provided: returns a SingleCellExperiment with the best performing method per gene stored as row-features. Access via SingleCellExperiment::int_elementMetadata(sce)\$ADImpute\$methods.
- if sce is not provided: returns a character with the best performing method in the training set for each gene

See Also

ImputeBaseline, ImputeDrImpute, ImputeNetwork

Examples

```
# Normalize demo data
norm_data <- NormalizeRPM(ADImpute::demo_data)
method_choice <- EvaluateMethods(norm_data, do = c('Baseline','DrImpute'),
cores = 2)</pre>
```

GetDropoutProbabilities

Get dropout probabilities

Description

GetDropoutProbabilities computes dropout probabilities (probability of being a dropout that should be imputed rather than a true biological zero) using an adaptation of scImpute's approach

Usage

```
GetDropoutProbabilities(data, thre, cell.clusters, labels = NULL,
type = 'count', cores, BPPARAM, genelen = ADImpute::transcript_length)
```

Arguments

data matrix; original data before imputation

thre numeric; probability threshold to classify entries as biological zeros

cell.clusters integer; number of cell subpopulations

labels character; vector specifying the cell type of each column of data

type A character specifying the type of values in the expression matrix. Can be

'count' or 'TPM'

cores integer; number of cores used for paralell computation

BPPARAM parallel back-end to be used during parallel computation. See BiocParallelParam-class.

genelen matrix with at least 2 columns: 'hgnc_symbol' and 'transcript_length'

Details

This function follows scImpute's model to distinguish between true biological zeros and dropouts, and is based on adapted code from the scImpute R package.

Value

matrix with same dimensions as data containing the dropout probabilities for the corresponding entries

HandleBiologicalZeros Get dropout probabilities

Description

GetDropoutProbabilities computes dropout probabilities (probability of being a dropout that should be imputed rather than a true biological zero) using an adaptation of scImpute's approach

Usage

```
HandleBiologicalZeros(data, imputed, thre = 0.5, cell.clusters,
labels = NULL, type = 'count', cores = BiocParallel::bpworkers(BPPARAM),
BPPARAM = BiocParallel::SnowParam(type = "SOCK"),
genelen = ADImpute::transcript_length, prob.mat = NULL)
```

Arguments

data matrix; original data before imputation imputed list; imputation results for considered methods

thre numeric; between 0 and 1 specifying the threshold to determine dropout values

cell.clusters integer; number of cell subpopulations

labels character; vector specifying the cell type of each column of data

type A character specifying the type of values in the expression matrix. Can be

'count' or 'TPM'

cores integer; number of cores used for paralell computation

BPPARAM parallel back-end to be used during parallel computation. See BiocParallelParam-class.

genelen matrix with at least 2 columns: 'hgnc_symbol' and 'transcript_length'

prob.mat matrix with same dimensions as data containing the dropout probabilities for

the corresponding entries

14 Impute

Details

This function follows scImpute's model to distinguish between true biological zeros and dropouts, and is based on adapted code from the scImpute R package.

Value

list with 2 components: zerofiltered, a list equivalent to imputed but with entries of imputed likely biological zeros set back to zero, and dropoutprobabilities matrix with same dimensions as data containing the dropout probabilities for the corresponding entries

Impute

Dropout imputation using different methods

Description

Impute performs dropout imputation on normalized data, based on the choice of imputation methods.

Usage

```
Impute(data, sce = NULL, do = 'Ensemble', write = FALSE,
outdir = getwd(), method.choice = NULL, scale = 1, pseudo.count = 1,
labels = NULL, cell.clusters = 2, drop_thre = NULL, type = 'count',
tr.length = ADImpute::transcript_length,
cores = BiocParallel::bpworkers(BPPARAM),
BPPARAM = BiocParallel::SnowParam(type = "SOCK"),
net.coef = ADImpute::network.coefficients, net.implementation = 'iteration',
bulk = NULL, true.zero.thr = NULL, prob.mat = NULL, ...)
```

Arguments

data	matrix; raw counts (genes as rows and samples as columns)
sce	SingleCellExperiment; normalized counts and associated metadata.
do	character; choice of methods to be used for imputation. Currently supported methods are 'Baseline', 'DrImpute', 'Network', and 'Ensemble'. Defaults to 'Ensemble'. Not case-sensitive. Can include one or more methods. Nonsupported methods will be ignored.
write	logical; write intermediary and imputed objects to files?
outdir	character; path to directory where output files are written. Defaults to working directory
method.choice	character; best performing method in training data for each gene
scale	integer; scaling factor to divide all expression levels by (defaults to 1)
pseudo.count	integer; pseudo-count to be added to expression levels to avoid $\log(0)$ (defaults to 1)
labels	character; vector specifying the cell type of each column of data
cell.clusters	integer; number of cell subpopulations
drop_thre	numeric; between 0 and 1 specifying the threshold to determine dropout values

Impute 15

type A character specifying the type of values in the expression matrix. Can be

'count' or 'TPM'

tr.length matrix with at least 2 columns: 'hgnc_symbol' and 'transcript_length'

cores integer; number of cores used for paralell computation

BPPARAM parallel back-end to be used during parallel computation. See BiocParallelParam-class.

net.coef matrix; network coefficients. Please provide if you don't want to use ADIm-

pute's network model. Must contain one first column 'O' acconting for the intercept of the model and otherwise be an adjacency matrix with hgnc_symbols in rows and columns. Doesn't have to be squared. See ADImpute::demo_net

for a small example.

net.implementation

character; either 'iteration', for an iterative solution, or 'pseudoinv', to use Moore-Penrose pseudo-inversion as a solution. 'pseudoinv' is not advised for

big data.

bulk vector of reference bulk RNA-seq, if available (average across samples)

true.zero.thr if set to NULL (default), no true zero estimation is performed. Set to numeric

value between 0 and 1 for estimation. Value corresponds to the threshold used to determine true zeros: if the probability of dropout is lower than true.zero.thr,

the imputed entries are set to zero.

prob.mat matrix of the same size as data, filled with the dropout probabilities for each

gene in each cell

. . . additional parameters to pass to network-based imputation

Details

Values that are 0 in data are imputed according to the best-performing methods indicated in method.choice. Currently supported methods are:

- Baseline: imputation with average expression across all cells in the dataset. See ImputeBaseline.
- Previously published approaches: DrImpute and SAVER.
- Network: leverages information from a gene regulatory network to predicted expression of genes that are not quantified based on quantified interacting genes, in the same cell. See ImputeNetwork.
- Ensemble: is based on results on a training subset of the data at hand, indicating which method best predicts the expression of each gene. These results are supplied via method.choice. Applies the imputation results of the best performing method to the zero entries of each gene.

If 'Ensemble' is included in do, method. choice has to be provided (use output from EvaluateMethods()). Impute can create a directory imputation containing the imputation results of all methods in do. If true.zero.thr is set, dropout probabilities are computed using scImpute's framework. Expression values with dropout probabilities below true.zero.thr will be set back to 0 if imputed, as they likely correspond to true biological zeros (genes not expressed in cell) rather than technical dropouts (genes expressed but not captured). If see is set, imputed values by the different methods are added as new assays to see. Each assay corresponds to one imputation method. If true.zero.thr is set, only the values after filtering for biological zeros will be added. This is different from the output if see is not set, where the original values before filtering and the dropout probability matrix are returned.

16 ImputeBaseline

Value

• if sce is not set: returns a list of imputation results (normalized, log-transformed) for all selected methods in do. If true.zero.thr is defined, returns a list of 3 elements: 1) a list, imputations, containing the direct imputation results from each method; 2) a list, zerofiltered, containing the results of imputation in imputations after setting biological zeros back to zero; 3) a matrix, dropoutprobabilities, containing the dropout probability matrix used to set biological zeros.

• if sce is set: returns a SingleCellExperiment with new assays, each corresponding to one of the imputation methods applied. If true.zero.thr is defined, the assays will contain the results after imputation and setting biological zeros back to zero.

See Also

EvaluateMethods, ImputeBaseline, ImputeDrImpute, ImputeNetwork, ImputeSAVER

Examples

```
# Normalize demo data
norm_data <- NormalizeRPM(demo_data)
# Impute with particular method(s)
imputed_data <- Impute(do = 'Network', data = norm_data[,1:10],
net.coef = ADImpute::demo_net)
imputed_data <- Impute(do = 'Network', data = norm_data[,1:10],
net.implementation = 'pseudoinv', net.coef = ADImpute::demo_net)</pre>
```

 ${\tt ImputeBaseline}$

Impute using average expression across all cells

Description

 $Impute Baseline\ imputes\ dropouts\ using\ gene\ averages\ across\ cells.\ Zero\ values\ are\ excluded\ from\ the\ mean\ computation.$

Usage

```
ImputeBaseline(data, write = FALSE, ...)
```

Arguments

data	matrix with entries equal to zero to be imputed, normalized and log2-transformed
	(genes as rows and samples as columns)
write	logical; should a file with the imputation results be written?
	additional arguments to saveRDS

Value

matrix; imputation results considering the average expression values of genes

ImputeDrImpute 17

Description

ImputeDrImpute uses the DrImpute package for dropout imputation

Usage

```
ImputeDrImpute(data, write = FALSE)
```

Arguments

data matrix with entries equal to zero to be imputed, normalized and log2-transformed

(genes as rows and samples as columns)

write logical; should a file with the imputation results be written?

Value

matrix; imputation results from DrImpute

See Also

DrImpute

Description

ImputeNetParallel implements network-based imputation in parallel

Usage

```
ImputeNetParallel(drop.mat, arranged, cores =
BiocParallel::bpworkers(BPPARAM), type = 'iteration', max.iter = 50,
BPPARAM = BiocParallel::SnowParam(type = "SOCK"))
#'
```

Arguments

drop.mat	matrix, logical; dropout entries in the data matrix (genes as rows and samples as columns)
arranged	list; output of ArrangeData
cores	integer; number of cores used for paralell computation
type	character; either 'iteration', for an iterative solution, or 'pseudoinv', to use Moore-Penrose pseudo-inversion as a solution.
max.iter	numeric; maximum number of iterations for network imputation. Set to -1 to remove limit (not recommended)
BPPARAM	parallel back-end to be used during parallel computation. See BiocParallelParam-class.

18 ImputeNetwork

Value

matrix; imputation results incorporating network information

ImputeNetwork Network-based imputation

Description

Network-based imputation

Usage

```
ImputeNetwork(data, net.coef = NULL,
cores = BiocParallel::bpworkers(BPPARAM),
BPPARAM = BiocParallel::SnowParam(type = "SOCK"),
type = 'iteration', write = FALSE, ...)
```

Arguments

data	matrix with entries equal to zero to be imputed, normalized and log2-transformed (genes as rows and samples as columns)
net.coef	matrix; network coefficients.
cores	integer; number of cores to use
BPPARAM	$parallel\ back-end\ to\ be\ used\ during\ parallel\ computation.\ See\ {\tt BiocParallelParam-class}.$
type	character; either 'iteration', for an iterative solution, or 'pseudoinv', to use Moore-Penrose pseudo-inversion as a solution.
write	logical; should a file with the imputation results be written?
	additional arguments to ImputeNetParallel

Details

Imputes dropouts using a gene regulatory network trained on external data, as provided in net.coef. Dropout expression values are estimated from the expression of their predictor genes and the network coefficients.

Value

matrix; imputation results incorporating network information

See Also

ImputeNetParallel

ImputeNPDropouts 19

ImputeNPDropouts	Helper function to PseudoInverseSolution percell

Description

ImputeNPDropouts computes the non-dropout- dependent solution of network imputation for each cell

Usage

```
ImputeNPDropouts(net, expr)
```

Arguments

net matrix, logical; network coefficients for all dropout (to be imputed) genes that

are predictive of the expression of other dropout genes

expr numeric; vector of gene expression for all genes in the cell at hand

Value

vector; imputation results for the non-dropout-dependent genes

ImputePredictiveDropouts

Helper function to PseudoInverseSolution_percell

Description

ImputePredictiveDropouts applies Moore-Penrose pseudo-inversion to compute the dropout-dependent solution of network imputation for each cell

Usage

```
ImputePredictiveDropouts(net, thr = 0.01, expr)
```

Arguments

net matrix, logical; network coefficients for all dropout (to be imputed) genes that

are predictive of the expression of other dropout genes

thr numeric; tolerance threshold to detect zero singular values

expr numeric; vector of gene expression for all genes in the cell at hand

Value

vector; imputation results for the dropout-dependent genes

20 MaskData

ImputeSAVER <i>l</i>	<i>Jse SAVER</i>
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Description

ImputeSAVER uses the SAVER package for dropout imputation

Usage

```
ImputeSAVER(data, cores, try.mean = FALSE, write = FALSE)
```

Arguments

data matrix with entries equal to zero to be imputed, normalized (genes as rows and

samples as columns)

cores integer; number of cores to use

try.mean logical; whether to additionally use mean gene expression as prediction

write logical; should a file with the imputation results be written?

Value

matrix; imputation results from SAVER

See Also

saver

MaskData	Masking of entries for performance evaluation

Description

MaskData sets a portion (mask) of the non-zero entries of each row of data to zero

Usage

```
MaskData(data, write.to.file = FALSE, mask = .1)
```

Arguments

data matrix; raw counts (genes as rows and samples as columns)

write.to.file logical; should the output be written to a file?

mask numeric; ratio of total non-zero samples to be masked per gene (defaults to .1)

Details

Sets a portion (mask) of the non-zero entries of each row of data to zero. Result is written to filename.

MaskerPerGene 21

Value

matrix containing masked raw counts (genes as rows and samples as columns)

MaskerPerGene

Helper mask function

Description

Helper mask function, per feature.

Usage

MaskerPerGene(x, rowmask)

Arguments

x logical; data to mask

rowmask numeric; number of samples to be masked per gene

Value

logical containing positions to mask

network.coefficients

Transcriptome wide gene regulatory network

Description

Gene Regulatory Network used by ADImpute's Network imputation method. First column, 0, corresponds to the intercept of a gene--specific prediction model. The remaining rows and columns correspond to the adjacency matrix of the inferred network, where rows are target genes and columns are predictors. Genes are identified by their hgnc_symbol.

Usage

network.coefficients

Format

dgCMatrix

22 NormalizeTPM

|--|

Description

NormalizeRPM performs RPM normalization, with possibility to log the result

Usage

```
NormalizeRPM(data, sce = NULL, log = FALSE, scale = 1,
pseudo.count = 1)
```

Arguments

data matrix; raw data (genes as rows and samples as columns)

sce SingleCellExperiment; raw data

log logical; log RPMs?

scale integer; scale factor to divide RPMs by

pseudo.count numeric; if log = TRUE, value to add to RPMs in order to avoid taking log(0)

Value

matrix; library size normalized data

Examples

```
demo <- NormalizeRPM(ADImpute::demo_data)</pre>
```

NormalizeTPM	TPM normalization

Description

NormalizeTPM performs TPM normalization, with possibility to log the result

Usage

```
NormalizeTPM(data, sce = NULL, tr_length = NULL, log = FALSE, scale = 1, pseudo.count = 1)
```

Arguments

data matrix; raw data (genes as rows and samples as columns)

sce SingleCellExperiment; raw data

tr_length data.frame with at least 2 columns: 'hgnc_symbol' and 'transcript_length'

log logical; log TPMs?

scale integer; scale factor to divide TPMs by

pseudo.count numeric; if log = T, value to add to TPMs in order to avoid taking log(0)

Details

Gene length is estimated as the median of the lengths of all transcripts for each gene, as obtained from biomaRt. Genes for which length information cannot be found in biomaRt are dropped.

Value

matrix; normalized data (for transcript length and library size)

Examples

```
demo <- NormalizeTPM(ADImpute::demo_data)</pre>
```

PseudoInverseSolution_percell

Network-based parallel imputation - Moore-Penrose pseudoinversion

Description

PseudoInverseSolution_percell applies Moore-Penrose pseudo-inversion to compute the solution of network imputation for each cell

Usage

```
PseudoInverseSolution_percell(expr, net, drop_ind, thr = 0.01)
```

Arguments

expr numeric; expression vector for cell at hand

net matrix; network coefficients

drop_ind logical; dropout entries in the cell at hand

thr numeric; tolerance threshold to detect zero singular values

Value

matrix; imputation results incorporating network information

24 ReturnChoice

ReadData

Data read

Description

ReadData reads data from raw input file (.txt or .csv)

Usage

```
ReadData(path, ...)
```

Arguments

path character; path to input file

... additional arguments to data.table::fread()

Value

matrix; raw counts (genes as rows and samples as columns)

ReturnChoice

Wrapper for return of EvaluateMethods()

Description

ReturnChoice Adjusts the output of EvaluateMethods to a character vector or a SingleCellExperiment object. Helper function to ADImpute.

Usage

```
ReturnChoice(sce, choice)
```

Arguments

sce SingleCellExperiment; a SingleCellExperiment object if available; NULL oth-

erwise

choice character; best performing method in the training set for each gene

Value

- if sce is provided: returns a SingleCellExperiment with the best performing method per gene stored as row-features. Access via SingleCellExperiment::int_elementMetadata(sce)\$ADImpute\$methods.
- if sce is not provided: returns a character with the best performing method in the training set for each gene

ReturnOut 25

Description

ReturnOut Adjusts the output of Impute to a list of matrices or a SingleCellExperiment object. Helper function to ADImpute.

Usage

```
ReturnOut(result, sce)
```

Arguments

result list; imputation result

sce SingleCellExperiment; a SingleCellExperiment object if available; NULL oth-

erwise

Value

imputation results. A SingleCellExperiment if !is.null(sce), or a list with imputed results in matrix format otherwise.

SetBiologicalZeros Set biological zeros

Description

SetBiologicalZeros sets some of the entries back to zero after dropout imputation, as they likely correspond to true biological zeros (genes not expressed in given cell)

Usage

SetBiologicalZeros(imputation, drop_probs, thre = .2, was_zero)

Arguments

imputation matrix; imputed values

drop_probs matrix; dropout probabilities for each entry in imputation. 0 means certain

biological zero, while 1 means certain dropout to be imputed

thre numeric; probability threshold to classify entries as biological zeros

was_zero matrix; logical matrix: was the corresponding entry of imputation originally a

zero?

Details

Entries which were originally zero and have dropout probability below thre are considered biological zeros and, if they were imputed, are set back to 0.

26 transcript_length

Value

matrix containing likely biological zeros set back to 0.

SplitData

Selection of samples for training

Description

SplitData selects a portion (ratio) of samples (columns in data) to be used as training set

Usage

```
SplitData(data, ratio = .7, write.to.file = FALSE, train.only = TRUE)
```

Arguments

data matrix; raw counts (genes as rows and samples as columns)
ratio numeric; ratio of the samples to be used for training

write.to.file logical; should the output be written to a file?

train.only logical; if TRUE define only a training dataset, if FALSE writes both training

and validation sets (defaults to TRUE)

Details

Selects a portion (ratio) of samples (columns in data) to be used as training set and writes to file 'training_raw.txt'.

Value

matrix containing raw counts (genes as rows and samples as columns)

transcript_length

Table for transcript length calculations

Description

A data.frame to be used for transcript length computations. May be necessary upon TPM normalization, or as input to scImpute. All data was retrieved from biomaRt.

Usage

transcript_length

Format

A data.frame with 2 columns:

hgnc_symbol Gene symbol identifier
transcript length Length of transcript

WriteCSV 27

WriteCSV

Write csv file

Description

WriteCSV writes data to a comma-delimited output file

Usage

```
WriteCSV(object, file)
```

Arguments

object R object to write

file character; path to output file

Value

Returns NULL

Examples

```
file <- tempfile()
WriteCSV(iris, file = file)</pre>
```

WriteTXT

Write txt file

Description

WriteTXT writes data to a tab-delimited output file

Usage

```
WriteTXT(object, file)
```

Arguments

object R object to write

file character; path to output file

Value

Returns NULL

Examples

```
file <- tempfile()
WriteTXT(iris, file = file)</pre>
```

Index

* datasets	NormalizeRPM, 22
demo_data,9	NormalizeTPM, 22
demo_net, 10	
demo_sce, 10	PseudoInverseSolution_percell, 23
network.coefficients, 21	
transcript_length, 26	ReadData, 24
, – ,	ReturnChoice, 24
ArrangeData, 3, 17	ReturnOut, 25
CenterData, 3	saver, 20
CheckArguments_Impute, 4	SetBiologicalZeros, 25
ChooseMethod, 4	SplitData, 26
Combine, 5	
ComputeMSEGenewise, 5, 6	transcript_length, 26
CreateArgCheck, 6	Weite COV 27
CreateTrainData, 7	WriteCSV, 27
	WriteTXT, 27
DataCheck_Matrix, 7	
DataCheck_Network, 8	
DataCheck_SingleCellExperiment, 8	
DataCheck_TrLength, 9	
demo_data, 9	
demo_net, 10	
demo_sce, 10	
DrImpute, 17	
or impute, 17	
EvaluateMethods, 11, 16	
GetDropoutProbabilities, 12	
HandleBiologicalZeros, 13	
Impute, 14	
ImputeBaseline, <i>12</i> , <i>15</i> , <i>16</i> , 16	
ImputeDrImpute, 12, 16, 17	
ImputeNetParallel, 17, 18	
ImputeNetwork, 12, 15, 16, 18	
ImputeNPDropouts, 19	
ImputePredictiveDropouts, 19	
ImputeSAVER, 16, 20	
MaskData, 20	
MaskerPerGene, 21	
,	
network.coefficients, 21	