Package 'RUVSeq'

February 21, 2025

Version 1.41.0

Title Remove Unwanted Variation from RNA-Seq Data

Description This package implements the remove unwanted variation (RUV) methods of Risso et al. (2014) for the normalization of RNA-Seq read counts between samples.

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Date 04-15-2014

Imports methods, MASS

Depends Biobase, EDASeq (>= 1.99.1), edgeR

Suggests BiocStyle, knitr, RColorBrewer, zebrafishRNASeq, DESeq2

VignetteBuilder knitr

License Artistic-2.0

LazyLoad yes

biocViews ImmunoOncology, DifferentialExpression, Preprocessing, RNASeq, Software

URL https://github.com/drisso/RUVSeq

BugReports https://github.com/drisso/RUVSeq/issues

RoxygenNote 6.0.1

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

git_branch devel

git_last_commit bb9c230

git_last_commit_date 2024-10-29

Repository Bioconductor 3.21

Date/Publication 2025-02-20

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RUVSeq-package Remove Unwanted Variation from RNA-Seq Data

Description

This package implements the remove unwanted variation (RUV) methods of Risso et al. (2014) for the normalization of RNA-Seq read counts between samples.

Details

| Package: | RUVSeq |
|----------|--------------|
| Type: | Package |
| Version: | 0.99.1 |
| Date: | 2014-04-15 |
| License: | Artistic-2.0 |

The RUVg function implements the RUVg normalization procedure of Risso et al. (2014), by using control genes to remove unwanted variation from the RNA-Seq read counts.

See also RUVr and RUVs for the "residual" and "sample" methods, based, respectively, on residuals (e.g., deviance residuals from a first-pass GLM regression of the unnormalized counts on the covariates of interest) and replicate/negative control samples for which the covariates of interest are constant.

Author(s)

Davide Risso and Sandrine Dudoit

Maintainer: Davide Risso <<risso.davide@gmail.com>>

References

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. Normalization of RNA-seq data using factor analysis of control genes or samples. *Nature Biotechnology*, 2014. (In press).

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. The role of spike-in standards in the normalization of RNA-Seq. In D. Nettleton and S. Datta, editors, *Statistical Analysis of Next Generation Sequence Data*. Springer, 2014. (In press).

makeGroups

See Also

RUVg, RUVr, RUVs

makeGroups

Make a matrix suitable for use with RUVSeq methods such as RUVs.

Description

Each row in the returned matrix corresponds to a set of replicate samples. The number of columns is the size of the largest set of replicates; rows for smaller sets are padded with -1 values.

Usage

makeGroups(xs)

Arguments

xs

A vector indicating membership in a group.

Author(s)

Kamil Slowikowski

See Also

RUVs

Examples

```
makeGroups(c("A", "B", "B", "C", "C", "D", "D", "D", "A"))
```

| residuals.DGEGLM | Deviance and Pearson Residuals for the Negative Binomial Model of |
|------------------|---|
| | edgeR |

Description

This function implements the residuals method for the edgeR function glmFit.

Usage

```
## S3 method for class 'DGEGLM'
residuals(object, type = c("deviance", "pearson"), ...)
```

Arguments

| object | An object of class DGEGLM as created by the glmFit function of edgeR. |
|--------|---|
| type | Compute deviance or Pearson residuals. |
| | Additional arguments to be passed to the generic function. |

Value

A genes-by-samples numeric matrix with the negative binomial residuals for each gene and sample.

Author(s)

Davide Risso

References

McCullagh P, Nelder J (1989). Generalized Linear Models. Chapman and Hall, New York.

Venables, W. N. and Ripley, B. D. (1999). *Modern Applied Statistics with S-PLUS*. Third Edition. Springer.

Examples

head(res)

```
library(edgeR)
library(zebrafishRNASeq)
data(zfGenes)
## run on a subset genes for time reasons
## (real analyses should be performed on all genes)
genes <- rownames(zfGenes)[grep("^ENS", rownames(zfGenes))]</pre>
spikes <- rownames(zfGenes)[grep("^ERCC", rownames(zfGenes))]</pre>
set.seed(123)
idx <- c(sample(genes, 1000), spikes)</pre>
seq <- newSeqExpressionSet(as.matrix(zfGenes[idx,]))</pre>
x <- as.factor(rep(c("Ctl", "Trt"), each=3))</pre>
design <- model.matrix(~x)</pre>
y <- DGEList(counts=counts(seq), group=x)</pre>
y <- calcNormFactors(y, method="upperquartile")</pre>
y <- estimateGLMCommonDisp(y, design)</pre>
y <- estimateGLMTagwiseDisp(y, design)</pre>
fit <- glmFit(y, design)</pre>
res <- residuals(fit, type="deviance")</pre>
```

RUVg-methods

Description

This function implements the RUVg method of Risso et al. (2014).

Usage

RUVg(x, cIdx, k, drop=0, center=TRUE, round=TRUE, epsilon=1, tolerance=1e-8, isLog=FALSE)

Arguments

| x | Either a genes-by-samples numeric matrix or a SeqExpressionSet object con- taining the read counts. |
|-----------|---|
| cIdx | A character, logical, or numeric vector indicating the subset of genes to be used as negative controls in the estimation of the factors of unwanted variation. |
| k | The number of factors of unwanted variation to be estimated from the data. |
| drop | The number of singular values to drop in the estimation of the factors of un- wanted variation. This number is usually zero, but might be set to one if the first singular value captures the effect of interest. It must be less than k. |
| center | If TRUE, the counts are centered, for each gene, to have mean zero across samples. This is important to ensure that the first singular value does not capture the average gene expression. |
| round | If TRUE, the normalized measures are rounded to form pseudo-counts. |
| epsilon | A small constant (usually no larger than one) to be added to the counts prior to the log transformation to avoid problems with $log(0)$. |
| tolerance | Tolerance in the selection of the number of positive singular values, i.e., a sin- gular value must be larger than tolerance to be considered positive. |
| isLog | Set to TRUE if the input matrix is already log-transformed. |

Details

The RUVg procedure performs factor analysis of the read counts based on a suitably-chosen subset of negative control genes known a priori not be differentially expressed (DE) between the samples under consideration.

Several types of controls can be used, including housekeeping genes, spike-in sequences (e.g., ERCC), or "in-silico" empirical controls (e.g., least significantly DE genes based on a DE analysis performed prior to RUV normalization).

Note that one can relax the negative control gene assumption by requiring instead the identification of a set of positive or negative controls, with a priori known expression fold-changes between samples. RUVg can then simply be applied to control-centered log counts, as detailed in the vignette.

Methods

signature(x = "matrix", cIdx = "ANY", k = "numeric") It returns a list with

- A samples-by-factors matrix with the estimated factors of unwanted variation (W).
- The genes-by-samples matrix of normalized expression measures (possibly rounded) obtained by removing the factors of unwanted variation from the original read counts (normalizedCounts).

signature(x = "SeqExpressionSet", cIdx = "character", k="numeric") It returns a SeqExpressionSet with

- The normalized counts in the normalizedCounts slot.
- The estimated factors of unwanted variation as additional columns of the phenoData slot.

Author(s)

Davide Risso

References

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. Normalization of RNA-seq data using factor analysis of control genes or samples. *Nature Biotechnology*, 2014. (In press).

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. The role of spike-in standards in the normalization of RNA-Seq. In D. Nettleton and S. Datta, editors, *Statistical Analysis of Next Generation Sequence Data*. Springer, 2014. (In press).

See Also

RUVr, RUVs.

Examples

library(zebrafishRNASeq)
data(zfGenes)

```
## run on a subset of genes for time reasons
## (real analyses should be performed on all genes)
genes <- rownames(zfGenes)[grep("^ENS", rownames(zfGenes))]
spikes <- rownames(zfGenes)[grep("^ERCC", rownames(zfGenes))]
set.seed(123)
idx <- c(sample(genes, 1000), spikes)
seq <- newSeqExpressionSet(as.matrix(zfGenes[idx,]))</pre>
```

```
# RUVg normalization
seqRUVg <- RUVg(seq, spikes, k=1)</pre>
```

pData(seqRUVg)
head(normCounts(seqRUVg))

```
plotRLE(seq, outline=FALSE, ylim=c(-3, 3))
plotRLE(seqRUVg, outline=FALSE, ylim=c(-3, 3))
```

```
barplot(as.matrix(pData(seqRUVg)), beside=TRUE)
```

RUVr-methods

Description

This function implements the RUVr method of Risso et al. (2014).

Usage

RUVr(x, cIdx, k, residuals, center=TRUE, round=TRUE, epsilon=1, tolerance=1e-8, isLog=FALSE)

Arguments

| x | Either a genes-by-samples numeric matrix or a SeqExpressionSet object con- taining the read counts. |
|-----------|--|
| cIdx | A character, logical, or numeric vector indicating the subset of genes to be used as negative controls in the estimation of the factors of unwanted variation. |
| k | The number of factors of unwanted variation to be estimated from the data. |
| residuals | A genes-by-samples matrix of residuals obtained from a first-pass regression of the counts on the covariates of interest, usually the negative binomial deviance residuals obtained from edgeR with the residuals method. |
| center | If TRUE, the residuals are centered, for each gene, to have mean zero across samples. |
| round | If TRUE, the normalized measures are rounded to form pseudo-counts. |
| epsilon | A small constant (usually no larger than one) to be added to the counts prior to the log transformation to avoid problems with $log(0)$. |
| tolerance | Tolerance in the selection of the number of positive singular values, i.e., a sin- gular value must be larger than tolerance to be considered positive. |
| isLog | Set to TRUE if the input matrix is already log-transformed. |

Details

The RUVr procedure performs factor analysis on residuals, such as deviance residuals from a firstpass GLM regression of the counts on the covariates of interest using **edgeR**. The counts may be either unnormalized or normalized with a method such as upper-quartile (UQ) normalization.

Methods

signature(x = "matrix", cIdx = "ANY", k = "numeric", residuals = "matrix") It returns a list
 with

- A samples-by-factors matrix with the estimated factors of unwanted variation (W).
- The genes-by-samples matrix of normalized expression measures (possibly rounded) obtained by removing the factors of unwanted variation from the original read counts (normalizedCounts).

```
signature(x = "SeqExpressionSet", cIdx = "character", k="numeric", residuals = "matrix")
It returns a SeqExpressionSet with
```

- The normalized counts in the normalizedCounts slot.
- The estimated factors of unwanted variation as additional columns of the phenoData slot.

Author(s)

Davide Risso

References

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. Normalization of RNA-seq data using factor analysis of control genes or samples. *Nature Biotechnology*, 2014. (In press).

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. The role of spike-in standards in the normalization of RNA-Seq. In D. Nettleton and S. Datta, editors, *Statistical Analysis of Next Generation Sequence Data*. Springer, 2014. (In press).

See Also

RUVg, RUVs, residuals.

Examples

```
library(edgeR)
library(zebrafishRNASeq)
data(zfGenes)
## run on a subset of genes for time reasons
## (real analyses should be performed on all genes)
genes <- rownames(zfGenes)[grep("^ENS", rownames(zfGenes))]</pre>
spikes <- rownames(zfGenes)[grep("^ERCC", rownames(zfGenes))]</pre>
set.seed(123)
idx <- c(sample(genes, 1000), spikes)</pre>
seq <- newSeqExpressionSet(as.matrix(zfGenes[idx,]))</pre>
# Residuals from negative binomial GLM regression of UQ-normalized
# counts on covariates of interest, with edgeR
x <- as.factor(rep(c("Ctl", "Trt"), each=3))</pre>
design <- model.matrix(~x)</pre>
y <- DGEList(counts=counts(seq), group=x)</pre>
y <- calcNormFactors(y, method="upperquartile")</pre>
y <- estimateGLMCommonDisp(y, design)</pre>
y <- estimateGLMTagwiseDisp(y, design)</pre>
fit <- glmFit(y, design)</pre>
res <- residuals(fit, type="deviance")</pre>
# RUVr normalization (after UQ)
seqUQ <- betweenLaneNormalization(seq, which="upper")</pre>
controls <- rownames(seq)</pre>
seqRUVr <- RUVr(seqUQ, controls, k=1, res)</pre>
```

RUVs-methods

```
pData(seqRUVr)
head(normCounts(seqRUVr))
```

| RUVs-methods | Remove Unwanted Variation Using Replicate/Negative Control Sam- |
|--------------|---|
| | ples |

Description

This function implements the RUVs method of Risso et al. (2014).

Usage

RUVs(x, cIdx, k, scIdx, round=TRUE, epsilon=1, tolerance=1e-8, isLog=FALSE)

Arguments

| x | Either a genes-by-samples numeric matrix or a SeqExpressionSet object con- taining the read counts. |
|-----------|---|
| cIdx | A character, logical, or numeric vector indicating the subset of genes to be used as negative controls in the estimation of the factors of unwanted variation. |
| k | The number of factors of unwanted variation to be estimated from the data. |
| scIdx | A numeric matrix specifying the replicate samples for which to compute the count differences used to estimate the factors of unwanted variation (see details). |
| round | If TRUE, the normalized measures are rounded to form pseudo-counts. |
| epsilon | A small constant (usually no larger than one) to be added to the counts prior to the log transformation to avoid problems with $log(0)$. |
| tolerance | Tolerance in the selection of the number of positive singular values, i.e., a sin- gular value must be larger than tolerance to be considered positive. |
| isLog | Set to TRUE if the input matrix is already log-transformed. |

Details

The RUVs procedure performs factor analysis on a matrix of count differences for replicate/negative control samples, for which the biological covariates of interest are constant.

Each row of scIdx should correspond to a set of replicate samples. The number of columns is the size of the largest set of replicates; rows for smaller sets are padded with -1 values.

For example, if the sets of replicate samples are (1,11,21),(2,3),(4,5),(6,7,8), then scIdx should be

1 11 21 2 3 -1

- 45-1
- 678

Methods

signature(x = "matrix", cIdx = "ANY", k = "numeric", scIdx = "matrix") It returns a list with

- A samples-by-factors matrix with the estimated factors of unwanted variation (W).
- The genes-by-samples matrix of normalized expression measures (possibly rounded) obtained by removing the factors of unwanted variation from the original read counts (normalizedCounts).

signature(x = "SeqExpressionSet", cIdx = "character", k="numeric", scIdx = "matrix")
It returns a SeqExpressionSet with

- The normalized counts in the normalizedCounts slot.
- The estimated factors of unwanted variation as additional columns of the phenoData slot.

Author(s)

Davide Risso (building on a previous version by Laurent Jacob).

References

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. Normalization of RNA-seq data using factor analysis of control genes or samples. *Nature Biotechnology*, 2014. (In press).

D. Risso, J. Ngai, T. P. Speed, and S. Dudoit. The role of spike-in standards in the normalization of RNA-Seq. In D. Nettleton and S. Datta, editors, *Statistical Analysis of Next Generation Sequence Data*. Springer, 2014. (In press).

See Also

RUVg, RUVr.

Examples

```
library(zebrafishRNASeq)
data(zfGenes)
```

```
## run on a subset of genesfor time reasons
## (real analyses should be performed on all genes)
genes <- rownames(zfGenes)[grep("^ENS", rownames(zfGenes))]
spikes <- rownames(zfGenes)[grep("^ERCC", rownames(zfGenes))]
set.seed(123)
idx <- c(sample(genes, 1000), spikes)
seq <- newSeqExpressionSet(as.matrix(zfGenes[idx,]))</pre>
```

```
# RUVs normalization
controls <- rownames(seq)
differences <- matrix(data=c(1:3, 4:6), byrow=TRUE, nrow=2)
seqRUVs <- RUVs(seq, controls, k=1, differences)</pre>
```

pData(seqRUVs)
head(normCounts(seqRUVs))

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