# Package 'eisaR'

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**Title** Exon-Intron Split Analysis (EISA) in R

**Version** 1.19.0

**Description** Exon-intron split analysis (EISA) uses ordinary RNA-seq data to measure changes in mature RNA and pre-mRNA reads across different experimental conditions to quantify transcriptional and post-transcriptional regulation of gene expression. For details see Gaidatzis et al., Nat Biotechnol 2015. doi: 10.1038/nbt.3269. eisaR implements the major steps of EISA in R.

**Depends** R (>= 4.1)

License GPL-3

**biocViews** Transcription, GeneExpression, GeneRegulation, FunctionalGenomics, Transcriptomics, Regression, RNASeq

**Encoding UTF-8** 

**Roxygen** list(markdown = TRUE)

RoxygenNote 7.3.1

Suggests knitr, rmarkdown, testthat, BiocStyle, QuasR, Rbowtie, Rhisat2, Biostrings, BSgenome, BSgenome.Hsapiens.UCSC.hg38, ensembldb, AnnotationDbi, GenomicFeatures, txdbmaker, rtracklayer

VignetteBuilder knitr

**Imports** graphics, stats, GenomicRanges, S4Vectors, IRanges, limma, edgeR (>= 4.0), methods, SummarizedExperiment, BiocGenerics, utils

URL https://github.com/fmicompbio/eisaR

BugReports https://github.com/fmicompbio/eisaR/issues

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2 eisaR-package

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# **Contents**

| eisal | eisaR-package           |   |  |  |  | eisaR: Exon-Intron Split Analysis (EISA) in R |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |    |  |
|-------|-------------------------|---|--|--|--|---|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|----|--|
| Index |                         |   |  |  |  |   |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | 13 |  |
|       | runEISA .               |   |  |  |  |   |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |    |  |
|       | plotEISA                |   |  |  |  |   |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |    |  |
|       | getRegions<br>getTx2Gen |   |  |  |  |   |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |    |  |
|       | getFeaturel             | _ |  |  |  |   |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |    |  |
|       | eisaR-pack<br>exportToG | _ |  |  |  |   |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |    |  |

### **Description**

Exon-intron split analysis (EISA) uses ordinary RNA-seq data to measure changes in mature RNA and pre-mRNA reads across different experimental conditions to quantify transcriptional and post-transcriptional regulation of gene expression. For details see Gaidatzis et al., Nat Biotechnol 2015. doi: 10.1038/nbt.3269. eisaR implements the major steps of EISA in R.

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# See Also

Useful links:

- https://github.com/fmicompbio/eisaR
- Report bugs at https://github.com/fmicompbio/eisaR/issues

exportToGtf 3

exportToGtf

Export GRangesList to GTF

## **Description**

Export the features in a GRangesList generated by getFeatureRanges to a GTF file. The function will represent each row of each of the entries as an "exon", each individual entry as a "transcript", and aggregate all features belonging to the same gene as a "gene" entry in the GTF file.

## Usage

```
exportToGtf(grl, filepath)
```

## **Arguments**

grl GRangesList object, typically generated by getFeatureRanges filepath Path to output GTF file

#### Value

Does not return anything, generates a GTF file

#### Author(s)

Charlotte Soneson

```
## Get feature ranges
grl <- getFeatureRanges(
   gtf = system.file("extdata/small_example.gtf", package = "eisaR"),
   featureType = c("spliced", "intron"),
   intronType = "separate",
   flankLength = 5L,
   joinOverlappingIntrons = FALSE,
   verbose = TRUE
)

## Export GTF
exportToGtf(grl = grl, filepath = file.path(tempdir(), "exported.gtf"))</pre>
```

4 getFeatureRanges

getFeatureRanges

Generate a GRangesList object with feature ranges

#### **Description**

Generate a GRangesList object with genomic ranges for (any combination of) spliced transcripts, unspliced transcripts and introns.

## Usage

```
getFeatureRanges(
  gtf,
  featureType = c("spliced", "intron"),
  intronType = "separate",
  flankLength = 90L,
  joinOverlappingIntrons = FALSE,
  collapseIntronsByGene = FALSE,
  keepIntronInFeature = FALSE,
  verbose = TRUE
)
```

#### **Arguments**

Path to gtf file. gtf

Character vector indicating the type(s) of features to extract, any subset of c("spliced", featureType

"intron", "unspliced").

intronType Character vector indicating how to define the introns (only used if "intron" is

part of featureType). Has to be either "separate" (introns are defined for each transcript separately) or "collapse" (transcripts of the same gene are first col-

lapsed before introns are defined as any non-exonic part of the gene locus).

Integer scalar indicating the length of the flanking sequence added to each side flankLength

of each extracted intron (only used if "intron" is included in featureType).

joinOverlappingIntrons

Logical scalar indicating whether two introns that overlap (after adding the

flanking sequences) should be joined into one feature.

collapseIntronsByGene

Logical scalar indicating whether to collapse overlapping intron ranges by gene after extraction.

keepIntronInFeature

Logical scalar indicating whether introns (after adding the flank length) should be restricted to stay within the boundaries of the feature they were generated

from. For backwards compatibility, the default is FALSE.

verbose Logical scalar, whether to print out progress messages. getRegionsFromTxDb 5

## Value

Returns a GRangesList object where each element represents one extracted feature. The metadata of this object contains two data. frames mapping corresponding identifiers between the different feature types, as well as a list of all features for each type.

# Author(s)

Charlotte Soneson

```
## Get feature ranges
grl <- getFeatureRanges(</pre>
  gtf = system.file("extdata/small_example.gtf", package = "eisaR"),
  featureType = c("spliced", "intron"),
  intronType = "separate",
  flankLength = 5L,
  joinOverlappingIntrons = FALSE,
  collapseIntronsByGene = FALSE,
  verbose = TRUE
)
## GRangesList
grl
## Corresponding transcript/gene IDs
S4Vectors::metadata(grl)$corrtx
S4Vectors::metadata(grl)$corrgene
## List of features of different types
{\tt S4Vectors::metadata(grl)\$featurelist}
## Get feature sequences
if (requireNamespace("BSgenome", quietly = TRUE) &&
    requireNamespace("GenomicFeatures", quietly = TRUE)) {
  library(BSgenome)
  genome <- Biostrings::readDNAStringSet(</pre>
    system.file("extdata/small_example_genome.fa", package = "eisaR"))
  seqs <- GenomicFeatures::extractTranscriptSeqs(x = genome,</pre>
                                                   transcripts = grl)
  seas
}
```

#### **Description**

From a transcript database package (TxDb), extract exonic and gene body ranges for use with EISA. These regions can be used to quantify RNA-seq alignments in exons and gene bodies, respectively. Intronic counts can then be obtained from the difference between gene bodies and exonic region counts

## Usage

```
getRegionsFromTxDb(txdb, exonExt = 10L, strandedData = TRUE)
```

#### **Arguments**

txdb a TxDb or an EnsDb object with the transcript annotations.

exonExt numeric (default = 10L). Exonic ranges will be extended on either side by this

many nucleotides, in order to avoid "bleed-over" of exonic alignments into ad-

jacent intronic regions.

strandedData logical(1). If TRUE, the RNA-seq data is assumed to be strand-specific, and

therefore only overlapping genes that are on the same strand will be filtered out. If FALSE, also genes overlapping on opposite strands will be filtered out.

#### **Details**

The exonic regions are generated as follows:

- 1. extract exons by gene from the txdb
- 2. extend each exon by exonExt
- 3. combine overlapping exons within each gene
- 4. create gene body ranges from the most extreme exonic coordinates
- 5. filter out genes that have only a single exon (no intron), have exons on more than a single chromosome or on both strands, or that overlap other genes

#### Value

a list with elements "exons" and "genebodies", containing named GenomicRanges objects with ranges for exons and gene bodies, respectively.

#### Author(s)

Michael Stadler

#### See Also

 $\mathsf{TxDb}$  for details on  $\mathsf{TxDb}$  objects and the  $\mathsf{txdbmaker}$  package for how to create them, e.g. from .gtf files.

getTx2Gene 7

#### **Examples**

```
if (requireNamespace("AnnotationDbi", quietly = TRUE)) {
   txdb <- AnnotationDbi::loadDb(system.file("extdata", "hg19sub.sqlite", package = "eisaR"))
   regL <- getRegionsFromTxDb(txdb)
   lengths(regL)
}</pre>
```

getTx2Gene

Generate a transcript-to-gene mapping from a GRangesList

#### **Description**

Generate a data. frame mapping transcript IDs to gene IDs, based on a GRangesList object generated by getFeatureRanges.

# Usage

```
getTx2Gene(grl, filepath = NULL)
```

## **Arguments**

grl GRangesList object, typically generated by getFeatureRanges

filepath Either NULL or the path to a file where the transcript-to-gene mapping will be

written

#### Value

Invisibly returns a data. frame with the transcript-to-gene mapping.

#### Author(s)

Charlotte Soneson

```
## Get feature ranges
grl <- getFeatureRanges(
   gtf = system.file("extdata/small_example.gtf", package = "eisaR"),
   featureType = c("spliced", "intron"),
   intronType = "separate",
   flankLength = 5L,
   joinOverlappingIntrons = FALSE,
   verbose = TRUE
)

## Get transcript-to-gene mapping
t2g <- getTx2Gene(grl = grl)
t2g</pre>
```

8 plotEISA

plotEISA

Visualize the results from an exon-intron split analysis.

#### **Description**

plotEISA takes the return value from runEISA and generates a scatterplot of intronic versus exonic

#### Usage

```
plotEISA(
 Х,
  contrast = c("ExIn", "none"),
 minLfc = NULL,
 maxFDR = 0.05,
 genecolors = c("#E41A1C", "#497AB3", "#222222"),
)
```

### **Arguments**

list with EISA results, typically the return value from runEISA contrast one of "ExIn" or "none". If "ExIn" (the default), genes that significantly differ between exonic and intronic changes are highlighted. "none" turns off gene

highlighting.

minLfc NULL or numeric(1) with the minimal absolute log2 fold change to color a

gene. If NULL (the default), no fold changes are not used to select genes for

highlighting.

maxFDR numeric(1) with maximal false discovery rate for gene highlighting.

genecolors Vector of length three specifying the colors to use for genes that are significantly

up, down or unchanged.

further arguments past to plot(). Parameters that will be set automatically . . .

unless given in the arguments are:

pch : plot symbol (default: ".")

cex : plot symbol expansion factor (default: 2)

col : plot symbol color (default: according to contrast and genecolors)

xlab/ylab : axis labels

#### Value

NULL (invisibly)

# Author(s)

Michael Stadler

#### **Examples**

```
# see the help for runEISA() for a full example
```

runEISA

Run Exon-Intron Split Analysis.

# Description

Starting from count tables with exonic and intronic counts for two conditions, perform all the steps in EISA (normalize, identify quantifyable genes, calculate contrasts and their significance).

#### Usage

```
runEISA(
  cntEx,
  cntIn,
  cond,
  method = NULL,
  modelSamples = TRUE,
  geneSelection = c("filterByExpr", "none", "Gaidatzis2015"),
  statFramework = c("QLF", "LRT"),
  legacyQLF = FALSE,
  effects = c("predFC", "Gaidatzis2015"),
  pscnt = 2,
  sizeFactor = c("exon", "intron", "individual"),
  recalcNormFactAfterFilt = TRUE,
  recalcLibSizeAfterFilt = FALSE,
  ...
)
```

#### **Arguments**

| cntEx  | Gene by sample matrix with exonic counts, OR a SummarizedExperiment with two assays named exon and intron, containing exonic and intronic counts, respectively. If cntEx is a SummarizedExperiment, cntIn will be disregarded.   |
|--------|--|
| cntIn  | Gene by sample matrix with intronic counts. Must have the same structure as cntEx (same number and order of rows and columns) if cntEx is a matrix. Will be disregarded if cntEx is a SummarizedExperiment.  |
| cond   | numeric, character or factor with two levels that groups the samples (columns of cntEx and cntIn) into two conditions. The contrast will be defined as secondLevel - firstLevel.   |
| method | One of NULL (the default) or "Gaidatzis2015". If "Gaidatzis2015", gene filtering, statistical analysis and calculation of contrasts is performed as described in Gaidatzis et al. 2015, and the statistical analysis is based on glmFit and glmLRT. This is done by setting the arguments modelSamples, geneSelection, |

effects, pscnt, statFramework, sizeFactor, recalcNormFactAfterFilt and recalcLibSizeAfterFilt to appropriate values (see details), overriding the defaults or any value passed to these arguments. If NULL, the default values of the arguments will be used instead (recommended).

modelSamples

Whether to include a sample identifier in the design matrix of the statistical model. If TRUE, potential sample effects that affect both exonic and intronic counts of that sample will be taken into account, which could result in higher sensitivity (default: TRUE).

geneSelection

Controls how to select quantifyable genes. One of the following:

"filterByExpr": (default) First, counts are normalized using calcNormFactors, treating intronic and exonic counts as individual samples. Then, filterByExpr is used with default parameters to select quantifyable genes.

"none": This will use all the genes provided in the count tables, assuming that an appropriate selection of quantifyable genes has already been done.

"Gaidatzis2015": First, intronic and exonic counts are linearly scaled to the mean library size (estimated as the sum of all intronic or exonic counts, respectively). Then, quantifyable genes are selected as the genes with counts x that fulfill log2(x + 8) > 5 in both exons and introns.

statFramework

Selects the framework within edgeR that is used for the statistical analysis. One of:

"QLF": (default) Quasi-likelihood F-test using glmQLFit and glmQLFTest. This framework is highly recommended as it gives stricter error rate control by accounting for the uncertainty in dispersion estimation.

"LRT": Likelihood ratio test using glmFit and glmLRT.

legacyQLF

Whether to use the 'legacy' version of glmQLFit. See glmQLFit for more details. If FALSE, the new method introduced in edgeR 4.0.0 is used.

effects

How the effects (contrasts or log2 fold-changes) are calculated. One of:

"predFC": (default) Fold-changes are calculated using the fitted model with predFC with prior.count = pscnt. Please note that if a sample factor is included in the model (modelSamples=TRUE), effects cannot be obtained from that model. In that case, effects are obtained from a simpler model without sample effects.

"Gaidatzis2015": Fold-changes are calculated using the formula log2((x + pscnt)/(y + pscnt)). If pscnt is not set to 8, runEISA will warn that this deviates from the method used in Gaidatzis et al., 2015.

pscnt

numeric(1) with pseudocount to add to read counts (default: 2). For method = "Gaidatzis2015", it is set to 8. It is added to scaled read counts used in geneSelection = "Gaidatzis2015" and effects = "Gaidatzis2015", or else used in cpm(..., prior.count = pscnt) and predFC(..., prior.count = pscnt).

sizeFactor

How the size factors are calculated in the analysis. If 'exon' (default), the exonderived size factors are used also for the columns corresponding to intronic counts. If 'intron', the intron-derived size factors are used also for the columns corresponding to exonic counts. If 'individual', column-wise size factors are calculated.

recalcNormFactAfterFilt

Logical, indicating whether normalization factors should be recalculated after filtering out lowly expressed genes.

recalcLibSizeAfterFilt

Logical, indicating whether library sizes should be recalculated after filtering out lowly expressed genes.

... additional arguments passed to the DGEList constructor, such as lib.size or genes.

#### **Details**

Setting method = "Gaidatzis2015" has precedence over other argument values and corresponds to setting: modelSamples = FALSE, geneSelection = "Gaidatzis2015", statFramework = "LRT", effects = "Gaidatzis2015", pscnt = 8, sizeFactor = "individual", recalcNormFactAfterFilt = TRUE, recalcLibSizeAfterFilt = FALSE.

#### Value

a list with elements

fracIn fraction intronic counts in each sample

contrastName contrast name

**contrasts** contrast matrix for quantifyable genes, with average log2 fold-changes in exons (Dex), in introns (Din), and average difference between log2 fold-changes in exons and introns (Dex.Din)

**DGEList** DGEList object used in model fitting

**tab.ExIn** statistical results for differential changes between exonic and intronic contrast, an indication for post-transcriptional regulation.

**contr.ExIn** contrast vector used for testing the difference between exonic and intronic contrast (results in tab.ExIn)

**designMatrix** design matrix used for testing the difference between exonic and intronic contrast (results in tab.ExIn)

params a list with parameter values used to run EISA

#### Author(s)

Michael Stadler

#### References

Analysis of intronic and exonic reads in RNA-seq data characterizes transcriptional and post-transcriptional regulation. Dimos Gaidatzis, Lukas Burger, Maria Florescu and Michael B. Stadler Nature Biotechnology, 2015 Jul;33(7):722-9. doi: 10.1038/nbt.3269.

#### See Also

DGEList for DGEList object construction, calcNormFactors for normalization, filterByExpr for gene selection, glmFit and glmQLFit for statistical analysis.

# **Index**

```
* internal
    eisaR-package, 2
calcNormFactors, 10, 11
DGEList, 11
eisaR(eisaR-package), 2
\verb|eisaR-package|, 2
exportToGtf, 3
filterByExpr, 10, 11
getFeatureRanges, 4
{\tt getRegionsFromTxDb}, {\tt 5}
getTx2Gene, 7
glmFit, 9–11
glmLRT, 9, 10
glmQLFit, 10, 11
glmQLFTest, 10
plotEISA, 8
predFC, 10
runEISA, 8, 9
TxDb, 6
```