# Package 'PathwaySplice'

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Type Package

Title An R Package for Unbiased Splicing Pathway Analysis

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Author Aimin Yan, Xi Chen, Lily Wang

Maintainer Aimin Yan <aimin.at.work@gmail.com>

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Suggests testthat, knitr, rmarkdown

VignetteBuilder knitr

biocViews AlternativeSplicing, DifferentialSplicing,

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# **Description**

Pathway analysis of alternative splicing would be biased without accounting for the different number of exons associated with each gene, because genes with higher number of exons are more likely to be included in the 'significant' gene list in alternative splicing.

PathwaySplice is an R package that:

- (1) performs pathway analysis that explicitly adjusts for the number of exons associated with each gene
- (2) visualizes selection bias due to different number of exons for each gene
- (3) formally tests for presence of bias using logistic regression
- (4) supports gene sets based on the Gene Ontology terms, as well as more broadly defined gene sets (e.g. MSigDB) or user defined gene sets
- (5) identifies the significant genes driving pathway significance
- (6) organizes significant pathways with an enrichment map, where pathways with large number of overlapping genes are grouped together in a network graph

License LGPL(>=2) LazyData TRUE RoxygenNote 6.0.1 NeedsCompilation no 2 compareResults

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compareResults

compareResults

# **Description**

This function helps with visualizing the effects of bias adjustment in pathway analysis, by comparing the distributions of bias factors (e.g. number of exon bins) in genes associated with the most significant gene sets, before and after adjusting for bias factors in splicing pathway analysis.

# Usage

```
compareResults(n.go, adjusted, unadjusted, gene.based.table,
  output.dir = tempdir(), type.boxplot = c("All", "Only3"))
```

# **Arguments**

n.go	Distributions of bias factor in genes associated with the most significant n.go gene sets will be compared
adjusted	An object returned by runPathwaySplice, should correspond to gene set anlaysis results adjusting for biases in splicing analysis
unadjusted	An object returned by runPathwaySplice, should correspond to gene set analysis results NOT adjusting for biases
gene.based.tabl	le
	An object returned by makeGeneTable, should correspond to a table with one p-value for each gene
output.dir	Directory for output files
type.boxplot	Options are 'All' and 'Only3', corresponding to drawing 5 boxplots or 3 boxplots.
	5 boxplots: all genesets, sig.adjusted (sig gene sets in adjusted analysis), sig.unadjusted (sig gene sets in unadjusted analysis), sig.adjusted.only (sig gene sets in adjusted analysis only), sig.unadjusted.only (sig gene sets in unadjusted analysis only)
	3 boxplots: all genesets, adjusted.sig, unadjusted.sig

#### Value

The output include 3 files in output.dir: (1) a venn diagram comparing significant gene sets before and after adjusting for bias factors (2) a .csv file with gene set names belonging to different sections of the venn diagram (3) a box plot showing the distributions of number of features within all genes in significant gene sets, with and without adjusting for bias factors

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#### **Examples**

```
dir.name <- system.file('extdata', package='PathwaySplice')</pre>
hallmark.pathway.file <- file.path(dir.name, 'h.all.v6.0.symbols.gmt.txt')
hallmark <- gmtGene2Cat(hallmark.pathway.file,genomeID='hg19')</pre>
gene.based.table <- makeGeneTable(featureBasedData)</pre>
res.adj <- runPathwaySplice(gene.based.table,genome='hg19',</pre>
                          id='ensGene',gene2cat=hallmark,
                          go.size.limit = c(5, 200),
                          method='Wallenius', output.file=tempfile())
res.unadj <- runPathwaySplice(gene.based.table,genome='hg19',</pre>
                          id='ensGene',gene2cat=hallmark,go.size.limit = c(5, 200),
                          method='Hypergeometric',output.file=tempfile())
compareResults(20, res.adj, res.unadj, gene.based.table, type.boxplot='Only3')
## Not run:
# illustrate specification of output directory
compare Results (20, res. adj, res. unadj, gene. based. table, type. boxplot='Only3', output. dir='C:/Temp') \\
output.file.dir <- '~/OutputTestPathwaySplice'</pre>
compareResults(20,res.adj, res.unadj,gene.based.table,output.file.dir,type.boxplot='Only3')
## End(Not run)
```

enrichmentMap

enrichmentMap

#### **Description**

This function draws an enrichment map based on the overlap of gene sets as measured by the Jaccard Coefficient(JC)

# Usage

```
enrichmentMap(pathway.res, n = 50, fixed = TRUE, node.label.font = 1,
 similarity.threshold, scaling.factor = 1, output.file.dir = tempdir(),
 label.node.by.index = FALSE, add.numSIGInCat = FALSE, ...)
```

# **Arguments**

Pathway analysis results, an object returned by runPathwaySplice pathway.res The top n most significant gene sets are shown on enrichment map fixed If set to FALSE, will invoke tkplot (an interactive graphing facility in R) that allows one to draw an interactive enrichment map. Users can then manually

adjust the layout of the enrichment map. Note: on OS X system, users need to have XQuartz installed to run this function. tcltk R package is also required, but in most distributions of R tcltk is already included

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```
ront size of node label
similarity.threshold
Gene sets with Jaccard Coefficient > similarity.threshold will be connected on the enrichment map
scaling.factor Scaling factor that users can use to adjust the edge thickness of the network output.file.dir
Output files directory, see Details section below.
label.node.by.index
Options for labeling nodes on network.
FALSE indicates to label nodes by gene set names
TRUE indicates to label nodes by the index of gene sets
.... Additional parameter
```

#### **Details**

In the enrichment map,

- the *node colors* are controlled by gene set p-values, where smaller p-values correspond to dark red color.
- node sizes are controlled by the number of significant genes in gene set.
- thickness of the edges correspond to Jaccard similarity coefficient between two gene sets.
- the numbers after ':' indicates the nubmer of significant genes in the gene set.

The Jaccard similarity coefficient ranges from 0 to 1. JC=0 indicates there are no overlapping genes between two gene sets, JC=1 indicates two gene sets are identical.

The output directory will include the following files:

(1) a network file (in GML format) that can be used as an input for Cytoscape software (2) when label.node.by.index=TRUE, also a gene set information file that includes full names of the gene sets and the gene set indices shown on the network.

#### Value

A list with edge and node information used to plot enrichment map

#### Author(s)

Aimin created this function based on enrichMap function in G Yu's DOSE R package

#### **Examples**

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featureBasedData

featureBasedData

# **Description**

This dataset includes analysis results of RNA-seq data in Dolatshad et al. (2015), which compared transcriptome of CD34+ cells from myelodysplastic syndrome (MDS) patients with SF3B1 mutations vs. healthy controls using RNA sequencing. The JunctionSeq package was used to assess differential usage of counting bins, which are non-overlapping segments of the exons or splicing junctions (see Fig 1 in Anders et al. (2012)). Because of the size limit, only counting bins associated with a subset of genes were included here for demonstration.

# Usage

```
data(featureBasedData)
```

#### **Format**

A data frame with variables for gene identifier (geneID), gene feature identifier (countbinID), and p-value for gene feature (pvalue). Here we used "gene feature" and "counting bin" interchangeably

#### References

H Dolatshad, A Pellagatti, M Fernandez-Mercado1, B H Yip, L Malcovati, M Attwood, B Przychodzen N Sahgal, A A Kanapin, H Lockstone, L Scifo, P Vandenberghe, E Papaemmanuil, C W J Smith, P J Campbell, S Ogawa1, J P Maciejewski, M Cazzola, K I Savage1 and J Boultwood1 (2015) Disruption of SF3B1 results in deregulated expression and splicing of key genes and pathways in myelodysplastic syndrome hematopoietic stem and progenitor cells. Leukemia (2015) 29, 1092-1103

Anders S, Reyes A, Huber W (2012) Dececting differential usage of exons from RNA-seq data. Genome Research 22(10): 2008-2017

gmtGene2Cat

gmtGene2Cat

gmtGene2Cat

#### **Description**

Obtains all pathways associated with a set of genes

# Usage

```
gmtGene2Cat(pathway.file, gene.anno.file = NULL, genomeID = c("mm10",
   "hg19", "hg38"))
```

#### **Arguments**

pathway.file Input file for the gene sets in GMT format, must be in *gene symbols*gene.anno.file Gene annotation file that facilitate gene id conversions when gene ids in RNASeq data and pathway.file differ. If not specified, gmtGene2cat relies on gene
annotations provided by R package AnnotationHub.

genomeID Genome to be used. Options are 'mm10','hg19' or 'hg38'.

#### **Details**

This function reads a gene set file in GMT format, and returns a list with its name being a gene id, and each element of the list being the pathways associated with the gene. When gene ids in RNA-Seq data differ from those in pathway database, gene.anno.file facilitate gene id conversions. Users can prepare this file based on the format of the example gene annotation file at this link.

#### Value

A list where each entry is named by a gene and contains a vector of all the pathways associated with the gene

# **Examples**

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lrTestBias lrTestBias

#### **Description**

This function tests presence of selection bias using logistic regression, and produces a boxplot that compares distributions of bias factors (e.g. number of exons) for significant genes and non-significant genes.

# Usage

```
lrTestBias(genewise.table, boxplot.width = 0.1)
```

# **Arguments**

```
genewise.table A dataframe with genewise p-value for each gene, returned from makeGeneTable() boxplot.width width of boxplot
```

#### **Details**

To determine presentce of selection bias, we fit the following logistic regression model:

```
Pr(a gene is significant) ~ number of features within the gene
```

Here features refer to exon bins or splicing junction bins, depending on how genewise pvalues were obtained in the genewise.table

# Value

Nothing to be returned

#### **Examples**

```
gene.based.table <- makeGeneTable(featureBasedData)
lrTestBias(gene.based.table)</pre>
```

makeGeneTable

makeGeneTable

#### **Description**

This function obtains genewise p-values, by representing each gene with the smallest p-value among its features, and then determines genes status as significant or not.

# Usage

```
makeGeneTable(feature.table, sig.threshold = 0.05, stat = "pvalue")
```

# Arguments

feature.table An featureBasedData object.
sig.threshold Significance threshold used to determine whether the gene is significant or not stat The statistic used to select significant genes. Options are 'pvalue' or 'fdr'

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#### Value

Returns a genewised table with several variables (columns)

geneID Gene identifiers in ensembl gene IDs

geneWisePvalue each gene is represented by the smallest p-value among its features

sig.gene a gene is significant (1) or not (0)

mostSigDeFeature

the most significant gene feature

numFeature number of gene features within the gene

#### **Examples**

```
data(featureBasedData)
gene.based.table <- makeGeneTable(featureBasedData)</pre>
```

runPathwaySplice

runPathwaySplice

#### **Description**

This function identifies pathways that are enriched with signficant genes, while accounting for different number of gene features (e.g. exons) associated with each gene

#### Usage

```
runPathwaySplice(genewise.table, genome, id, gene2cat = NULL,
  test.cats = c("GO:CC", "GO:BP", "GO:MF"), go.size.limit = c(10, 200),
  method = "Wallenius", repcnt = 2000, use.genes.without.cat = FALSE,
  binsize = "auto", output.file = tempfile())
```

#### **Arguments**

genewise.table data frame returned from function makeGeneTable genome Genome to be used, options are 'hg19' or 'mm10' id GeneID, options are 'entrezgene' or 'ensembl\_gene\_id'

gene2cat Get sets to be tested, these are defined by users, can be obtained from gmtGene2Cat

function

test.cats Default gene ontology gene sets to be tested if gene2cat is not defined

go.size.limit Size limit of the gene sets to be tested

method the method used to calculate pathway enrichment p value. Options are 'Walle-

nius', 'Sampling', and 'Hypergeometric'

report Number of random samples to be calculated when 'Sampling' is used, this ar-

gument ignored unless method='Sampling'

 ${\tt use.genes.without.cat}$ 

Whether genes not mapped to any gene\_set tested are included in analysis. Default is set to FALSE, where genes not mapped to any tested categories are ig-

nored in analysis.

binsize The number of genes in each gene bin in the bias plot

output.file File name for the analysis result in .csv format.

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#### **Details**

This function implements the methodology described in Young et al. (2011) to adjust for different number of gene features (column numFeature in gene.based.table). For example, gene features can be non-overlapping exon counting bins associated with each gene (Fig 1 in Anders et al. 2012). In the bias plot, the genes are grouped by numFeature in genewise.table into gene bins, the proportions of signficant genes are then plotted against the gene bins.

#### Value

runPathwaySplice returns a tibble with the following information:

gene\_set Name of the gene set. Note in this document we used the terms gene\_set, cate-

gory, and pathway interchangeably

over\_represented\_pvalue

P-vaue for the associated gene\_set being over-represented among significant

genes

under\_represented\_pvalue

P-vaue for the associated gene\_set being under-represented among significant

genes

numSIGInCat The number of significant genes in the gene\_set

numInCat The total number of genes in the gene\_set

description Description of the gene gene\_set

ontology The domain of the gene ontology terms if GO categories were tested. Go cate-

gories can be classified into three domains: cellular component, biological pro-

cess, molecular function.

SIGgene\_ensembl

Ensembl gene ID of significant genes in the gene\_set

SIGgene\_symbol Gene symbols of signficant genes in the gene\_set

Ave\_value\_all\_gene

The average value for numFeature for all the genes in the gene\_set, note that

numFeature is the bias factor adjusted by PathwaySplice

These information are also saved in the file output.file

# References

Young MD, Wakefield MJ, Smyth GK, Oshlack A (2011) Gene ontology analysis for RNA-seq: accounting for selection bias. Genome Biology 11:R14

Anders S, Reyes A, Huber W (2012) Dececting differential usage of exons from RNA-seq data. Genome Research 22(10): 2008-2017

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