# Package 'fishpond'

March 31, 2025

**Title** Fishpond: downstream methods and tools for expression data

Version 2.12.0

Maintainer Michael Love <michaelisaiahlove@gmail.com>

**Description** Fishpond contains methods for differential transcript and gene expression analysis of RNA-seq data using inferential replicates for uncertainty of abundance quantification, as generated by Gibbs sampling or bootstrap sampling. Also the package contains a number of utilities for working with Salmon and Alevin quantification files.

Imports graphics, stats, utils, methods, abind, gtools, qvalue, S4Vectors, IRanges, SummarizedExperiment, GenomicRanges, matrixStats, svMisc, Matrix, SingleCellExperiment, jsonlite

**Suggests** testthat, knitr, rmarkdown, macrophage, tximeta, org.Hs.eg.db, samr, DESeq2, apeglm, tximportData, limma, ensembldb, EnsDb.Hsapiens.v86, GenomicFeatures, AnnotationDbi, pheatmap, Gviz, GenomeInfoDb, data.table

**License** GPL-2 **Encoding** UTF-8

URL https://thelovelab.github.io/fishpond,
 https://thelovelab.com/mikelove/fishpond

BugReports https://support.bioconductor.org/tag/fishpond

biocViews Sequencing, RNASeq, GeneExpression, Transcription, Normalization, Regression, MultipleComparison, BatchEffect, Visualization, DifferentialExpression, DifferentialSplicing, AlternativeSplicing, SingleCell

VignetteBuilder knitr

LazyData true

RoxygenNote 7.2.3

git\_url https://git.bioconductor.org/packages/fishpond

git\_branch RELEASE\_3\_20

git\_last\_commit 10b80e1

git\_last\_commit\_date 2024-10-29

Repository Bioconductor 3.20

2 fishpond-package

# **Contents**

	27
swish	24
splitSwish	
scaleInfReps	22
salmonEC	
plotMASwish	20
plotInfReps	
plotAllelicHeatmap	
plotAllelicGene	15
miniSwish	15
makeTx2Tss	14
makeSimSwishData	
makeInfReps	12
loadFry	10
labelKeep	9
isoformProportions	9
importAllelicCounts	7
getTrace	7
deswish	6
computeInfRV	5
alevinEC	4
addStatsFromCSV	3
fishpond-package	2
	addStatsFromCSV alevinEC. computeInfRV deswish getTrace importAllelicCounts isoformProportions labelKeep loadFry makeInfReps makeSimSwishData makeTx2Tss miniSwish plotAllelicGene plotAllelicHeatmap plotInfReps plotMASwish salmonEC scaleInfReps splitSwish swish swish

fishpond-package

Fishpond: downstream methods and tools for expression data

# Description

This package provides statistical methods and other tools for working with Salmon and Alevin quantification of RNA-seq data. Fishpond contains the Swish non-parametric method for detecting differential transcript expression (DTE). Swish can also be used to detect differential gene expression (DGE), to perform allelic analysis, or to assess changes in isoform proportions.

addStatsFromCSV 3

#### **Details**

The main Swish functions are:

- scaleInfReps scaling transcript or gene expression data
- labelKeep labelling which features have sufficient counts
- swish perform non-parametric differential analysis
- Plots, e.g., plotMASwish, plotInfReps

All software-related questions should be posted to the Bioconductor Support Site:

```
https://support.bioconductor.org
```

The code can be viewed at the GitHub repository, which also lists the contributor code of conduct:

```
https://github.com/mikelove/fishpond
```

#### References

Swish method:

Zhu, A., Srivastava, A., Ibrahim, J.G., Patro, R., Love, M.I. (2019) Nonparametric expression analysis using inferential replicate counts. Nucleic Acids Research. https://doi.org/10.1093/nar/gkz622

Compression, makeInfReps and splitSwish:

Van Buren, S., Sarkar, H., Srivastava, A., Rashid, N.U., Patro, R., Love, M.I. (2020) Compression of quantification uncertainty for scRNA-seq counts. bioRxiv. https://doi.org/10.1101/2020.07.06.189639

addStatsFromCSV

Read statistics and nulls from CSV file

# Description

After running splitSwish and the associated Snakefile, this function can be used to gather and add the results to the original object. See the alevin section of the vignette for an example.

#### Usage

```
addStatsFromCSV(y = NULL, infile, estPi0 = FALSE)
```

#### **Arguments**

У	a SummarizedExperiment (if NULL, function will output a data.frame)
infile	character, path to the summary.csv file
estPi0	logical, see swish

# Value

the SummarizedExperiment with metadata columns added, or if y is NULL, a data.frame of compiled results

4 alevinEC

alevinEC	Construct a sparse matrix of transcript compatibility counts from
	alevin output

#### **Description**

Constructs a UMI count matrix with equivalence class identifiers in the rows and barcode identifiers in the columns. The count matrix is generated from one or multiple 'bfh.txt' files that have been created by running alevin-fry with the –dumpBFH flag. Alevin-fry - https://doi.org/10.1186/s13059-019-1670-y

# Usage

```
alevinEC(
  paths,
  tx2gene,
  multigene = FALSE,
  ignoreTxVersion = FALSE,
  ignoreAfterBar = FALSE,
  quiet = FALSE
)
```

#### **Arguments**

paths 'Charachter' or 'character vector', path specifying the location of the 'bfh.txt'

files generated with alevin-fry.

tx2gene A 'dataframe' linking transcript identifiers to their corresponding gene identi-

fiers. Transcript identifiers must be in a column 'isoform\_id'. Corresponding

gene identifiers must be in a column 'gene\_id'.

multigene 'Logical', should equivalence classes that are compatible with multiple genes be

retained? Default is 'FALSE', removing such ambiguous equivalence classes.

ignoreTxVersion

logical, whether to split the isoform id on the '.' character to remove version information to facilitate matching with the isoform id in 'tx2gene' (default

FALSE).

ignoreAfterBar logical, whether to split the isoform id on the 'l' character to facilitate matching

with the isoform id in 'tx2gene' (default FALSE).

quiet 'Logical', set 'TRUE' to avoid displaying messages.

#### Value

A list with two elements. The first element 'counts' is a sparse count matrix with equivalence class identifiers in the rows and barcode identifiers followed by an underscore and a sample identifier in the columns. The second element 'tx2gene\_matched' allows for linking the equivalence class identifiers to their respective transcripts and genes.

computeInfRV 5

#### **Details**

The resulting count matrix uses equivalence class identifiers as rownames. These can be linked to respective transcripts and genes using the 'tx2gene\_matched' element of the output. Specifically, if the equivalence class identifier reads 1|2|8, then the equivalence class is compatible with the transcripts and their respective genes in rows 1, 2 and 8 of 'tx2gene\_matched'.

#### Author(s)

Jeroen Gilis

 ${\it compute inferential relative variance (InfRV)}$ 

# **Description**

InfRV is a useful quantity for comparing groups of features (transcripts, genes, etc.) by inferential uncertainty. This function provides computation of the mean InfRV over samples, per feature, stored in mcols(y)\$meanInfRV.

#### Usage

```
computeInfRV(y, pc = 5, shift = 0.01, meanVariance, useCounts = FALSE)
```

# Arguments

y a SummarizedExperiment

pc a pseudocount parameter for the denominator

shift a final shift parameter

meanVariance logical, use pre-computed inferential mean and variance assays instead of counts

and computed variance from infReps. If missing, will use pre-computed mean

and variance when present

useCounts logical, whether to use the MLE count matrix for the mean instead of mean of

inferential replicates. this argument is for backwards compatability, as previous

versions used counts. Default is FALSE

### **Details**

InfRV is defined in Zhu et al. (2019) as:  $\max(s^2-\mu,0)/\mu$ , using the inferential sample variance and sample mean. This formulation takes the non-Poisson part of the inferential variance and scales by the mean, which effectively stabilizes inferential uncertainty over mean count. In practice, we also add pc to the denominator and shift to the final quantity, to facilitate visualization.

This function also computes and adds the mean and variance of inferential replicates, which can be useful ahead of plotInfReps. Note that InfRV is not used in the swish statistical method (for generating test statistics, p-values or q-values), it is just for visualization.

# Value

a SummarizedExperiment with meanInfRV in the metadata columns

6 deswish

#### References

Anqi Zhu, Avi Srivastava, Joseph G Ibrahim, Rob Patro, Michael I Love "Nonparametric expression analysis using inferential replicate counts" Nucleic Acids Research (2019). https://doi.org/10.1093/nar/gkz622

deswish

deswish: DESeq2-apeglm With Inferential Samples Helps

#### **Description**

The DESeq2-apeglm With Inferential Samples implementation supposes a hierarchical distribution of log2 fold changes. The final posterior standard deviation is calculated by adding the posterior variance from modeling biological replicates computed by apeg1m, and the observed variance on the posterior mode over inferential replicates. This function requires the DESeq2 and apeglm packages to be installed and will print an error if they are not found.

#### Usage

```
deswish(y, x, coef)
```

#### **Arguments**

y a SummarizedExperiment containing the inferential replicate matrices, as output by tximeta, and then with labelKeep applied. One does not need to run

scaleInfReps as scaling is done internally via DESeq2.

x the design matrix

coef the coefficient to test (see lfcShrink)

# Value

a SummarizedExperiment with metadata columns added: the log2 fold change and posterior SD using inferential replicates, and the original log2 fold change (apeglm) and its posterior SD

#### References

The DESeq and 1fcShrink function in the DESeq2 package:

Zhu, Ibrahim, Love "Heavy-tailed prior distributions for sequence count data: removing the noise and preserving large differences" Bioinformatics (2018).

Love, Huber, Anders "Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2" Genome Biology (2014).

#### **Examples**

```
# a small example... 500 genes, 10 inf reps
y <- makeSimSwishData(m=500, numReps=10)
y <- labelKeep(y)
#y <- deswish(y, ~condition, "condition_2_vs_1")</pre>
```

getTrace 7

getTrace

Obtain a trace of inferential replicates for a sample

#### **Description**

Simple helper function to obtain a trace (e.g. MCMC trace) of the ordered inferential replicates for one samples. Supports either multiple features, idx, or multiple samples, samp\_idx (not both). Returns a tidy data.frame for easy plotting.

#### Usage

```
getTrace(y, idx, samp_idx)
```

#### **Arguments**

```
y a SummarizedExperiment with inferential replicates as assays infRep1 etc.
idx the names or row numbers of the gene or transcript to plot
samp_idx the names or column numbers of the samples to plot
```

#### Value

a data.frame with the counts along the interential replicates, possible with additional columns specifying feature or sample

#### **Examples**

```
y <- makeSimSwishData()
getTrace(y, "gene-1", "s1")</pre>
```

importAllelicCounts

Import allelic counts as a SummarizedExperiment

# **Description**

Read in Salmon quantification of allelic counts from a diploid transcriptome. Assumes that diploid transcripts are marked with the following suffix: an underscore and a consistent symbol for each of the two alleles, e.g. ENST123\_M and ENST123\_P, or ENST123\_alt and ENST123\_ref, etc. importAllelicCounts requires the tximeta package. Further information in Details below.

# Usage

```
importAllelicCounts(
  coldata,
  a1,
  a2,
  format = c("wide", "assays"),
  tx2gene = NULL,
  ...
)
```

8 importAllelicCounts

#### **Arguments**

a data.frame as used in tximeta

the symbol for the effect allele

the symbol for the non-effect allele

format either "wide" or "assays" for whether to combine the allelic counts as columns

(wide) or put the allelic count information in different assay slots (assays). For wide output, the data for the non-effect allele (a2) comes first, then the effect allele (a1), e.g. [a2 | a1]. The ref level of the factor variable se\$allele will be "a2" (so by default comparisons will be: a1 vs a2). For assays output, all of

the original matrices are renamed with a prefix, either a1- or a2-.

tx2gene optional, a data.frame with first column indicating transcripts, second column

indicating genes (or any other transcript grouping). Alternatively, this can be a GRanges object with required columns tx\_id, and group\_id (see makeTx2Tss).

For more information on this argument, see Details.

... any arguments to pass to tximeta

#### **Details**

**Requirements** - There must be exactly two alleles for each transcript, and the --keep-duplicates option should be used in Salmon indexing to avoid removal of transcripts with identical sequence. The output object has half the number of transcripts, with the two alleles either stored in a "wide" object, or as re-named "assays". Note carefully that the symbol provided to a1 is used as the effect allele, and a2 is used as the non-effect allele (see the format argument description and Value description below).

tx2gene - The two columns should include the a1 and a2 suffix for the transcripts and genes/groups, or those will be added internally, if it is detected that the first transcript does not have these suffices. For example if \_alt or \_ref, or \_M or \_P (as indicated by the a1 and a2 arguments) are not present in the table, the table rows will be duplicated with those suffices added on behalf of the user. If tx2gene is not provided, the output object will be transcript-level. Do not attempt to set the tx0ut argument, it will conflict with internal calls to downstream functions. If the a1/a2 suffices are not at the end of the transcript name in the quantification files, e.g. ENST123\_M|<metadata>, then ignoreAfterBar=TRUE can be used to match regardless of the string following | in the quantification files.

skipMeta=TRUE is used, as it is assumed the diploid transcriptome does not match any reference transcript collection. This may change in future iterations of the function, depending on developments in upstream annotations and software.

If tx2gene is a GRanges object, the rowRanges of the output will be the reduced ranges of the grouped input ranges, with tx\_id collapsed into a CharacterList, and TSS positions saved as an IntegerList, if these are not equal among the transcripts of a group. Other metadata columns are not manipulated, just the metadata for the first range is returned.

#### Value

a SummarizedExperiment, with allele counts (and other data) combined into a wide matrix [a2 | a1], or as assays (a1, then a2). The original strings associated with a1 and a2 are stored in the metadata of the object, in the alleles list element. Note the reference level of se\$allele will be "a2", such that comparisons by default will be a1 vs a2 (effect vs non-effect).

isoformProportions 9

#### References

Euphy Wu, Noor P. Singh, Kwangbom Choi, Mohsen Zakeri, Matthew Vincent, Gary A. Churchill, Cheryl L. Ackert-Bicknell, Rob Patro, Michael I. Love. "Detecting isoform-level allelic imbalance accounting for inferential uncertainty" bioRxiv (2022) https://doi.org/10.1101/2022.08.12.503785

isoformProportions

Create isoform proportions from scaled data

#### **Description**

Takes output of scaled (and optionally filtered) counts and returns isoform proportions by dividing out the total scaled count for the gene for each sample. The operation is performed on the counts assay, then creating a new assay called isoProp, and on all of the inferential replicates, turning them from counts into isoform proportions. Any transcripts (rows) from single isoform genes are removed, and the transcripts will be re-ordered by gene ID.

# Usage

```
isoformProportions(y, geneCol = "gene_id", quiet = FALSE)
```

#### **Arguments**

y a SummarizedExperiment

geneCol the name of the gene ID column in the metadata columns for the rows of y

quiet display no messages

#### Value

a SummarizedExperiment, with single-isoform transcripts removed, and transcripts now ordered by gene

labelKeep

Label rows to keep based on minimal count

# **Description**

Adds a column keep to mcols(y) that specifies which rows of the SummarizedExperiment will be included in statistical testing. Rows are not removed, just marked with the logical keep.

# Usage

```
labelKeep(y, minCount = 10, minN = 3, x)
```

10 loadFry

### **Arguments**

y a SummarizedExperiment

minCount the minimum count

minN the minimum sample size at minCount

x the name of the condition variable, will use the smaller of the two groups to set

minN. Similar to edgeR's filterByExpr, as the smaller group grows past 10,

minN grows only by 0.7 increments of sample size

#### Value

a SummarizedExperiment with a new column keep in mcols(y)

# **Examples**

```
y <- makeSimSwishData()
y <- scaleInfReps(y)
y <- labelKeep(y)</pre>
```

loadFry

Load in data from alevin-fry USA mode

# **Description**

Enables easy loading of sparse data matrices provided by alevin-fry USA mode.

#### Usage

```
loadFry(fryDir, outputFormat = "scRNA", nonzero = FALSE, quiet = FALSE)
```

# **Arguments**

fryDir path to the output directory returned by alevin-fry quant command. This di-

rectory should contain a metainfo.json, and an alevin folder which contains

quants\_mat.mtx, quants\_mat\_cols.txt and quants\_mat\_rows.txt

 $output {\tt Format} \qquad {\tt can \, be} \, \textit{either} \, \textbf{be} \, \textbf{a} \, \textbf{list that defines the desired format of the output SingleCellExperiment}$ 

object *or* a string that represents one of the pre-defined output formats, which are "scRNA", "snRNA", "all", "scVelo", "velocity", "U+S+A" and "S+A". See details for the explanations of the pre-defined formats and how to define custom

format.

nonzero whether to filter cells with non-zero expression value across all genes (default

FALSE). If TRUE, this will filter based on all assays. If a string vector of assay names, it will filter based on the matching assays in the vector. If not in USA

mode, it must be TRUE/FALSE/counts.

quiet logical whether to display no messages

# Value

A SingleCellExperiment object that contains one or more assays. Each assay consists of a gene by cell count matrix. The row names are feature names, and the column names are cell barcodes

loadFry 11

#### **Details about loadFry**

This function consumes the result folder returned by running alevin-fry quant in unspliced, spliced, ambiguous (USA) quantification mode, and returns a SingleCellExperiment object that contains a final count for each gene within each cell. In USA mode, alevin-fry quant returns a count matrix contains three types of count for each feature (gene) within each sample (cell or nucleus), which represent the spliced mRNA count of the gene (S), the unspliced mRNA count of the gene (U), and the count of UMIs whose splicing status is ambiguous for the gene (A). For each assay defined by outputFormat, these three counts of a gene within a cell will be summed to get the final count of the gene according to the rule defined in the outputFormat. The returned object will contains the desired assays defined by outputFormat, with rownames as the barcode of samples and colnames as the feature names.

#### Details about the output format

The outputFormat argument takes *either* be a list that defines the desired format of the output SingleCellExperiment object *or* a string that represents one of the pre-defined output format.

Currently the pre-defined formats of the output SingleCellExperiment object are:

- "scRNA": This format is recommended for single cell experiments. It returns a counts assay that contains the S+A count of each gene in each cell, and a unspliced assay that contains the U count of each gene in each cell.
- "snRNA", "all" and "U+S+A": These three formats are the same. They return a counts assay that contains the U+S+A count of each gene in each cell without any extra layers. "snRNA" is recommended for single-nucleus RNA-sequencing experiments. "raw" is recommended for mimicing CellRanger 7's behavior, which returns this format for both single-cell and single-nucleus experiments.
- "S+A": This format returns a counts assay that contains the S+A count of each gene in each cell.
- "raw": This format puts the three kinds of counts into three separate assays, which are unspliced, spliced and ambiguous.
- "velocity": This format contains two assays. The spliced assay contains the S+A count of each gene in each cell. The unspliced assay contains the U counts of each gene in each cell.
- "scVelo": This format is for direct entry into velociraptor R package or other scVelo downstream analysis pipeline for velocity analysis in R with Bioconductor. It adds the expected "S"-pliced assay and removes errors for size factors being non-positive.

A custom output format can be defined using a list. Each element in the list defines an assay in the output SingleCellExperiment object. The name of an element in the list will be the name of the corresponding assay in the output object. Each element in the list should be defined as a vector that takes at least one of the three kinds of count, which are U, S and A. See the provided toy example for defining a custom output format.

#### Author(s)

Dongze He, with contributions from Steve Lianoglou, Wes Wilson

# References

alevin-fry publication:

He, D., Zakeri, M., Sarkar, H. et al. "Alevin-fry unlocks rapid, accurate and memory-frugal quantification of single-cell RNA-seq data." Nature Methods 19, 316–322 (2022). https://doi.org/10.1038/s41592-022-01408-3

12 makeInfReps

#### **Examples**

```
# Get path for minimal example avelin-fry output dir
testdat <- fishpond:::readExampleFryData("fry-usa-basic")

# This is exactly how the velocity format defined internally.
custom_velocity_format <- list("spliced"=c("S","A"), "unspliced"=c("U"))

# Load alevin-fry gene quantification in velocity format
sce <- loadFry(fryDir=testdat$parent_dir, outputFormat=custom_velocity_format)
SummarizedExperiment::assayNames(sce)

# Load the same data but use pre-defined, velociraptor R pckage desired format
scvelo_format <- "scVelo"

scev <- loadFry(fryDir=testdat$parent_dir, outputFormat=scvelo_format, nonzero=TRUE)
SummarizedExperiment::assayNames(scev)</pre>
```

makeInfReps

Make pseudo-inferential replicates from mean and variance

#### **Description**

Makes pseudo-inferential replicate counts from mean and variance assays. The simulated counts are drawn from a negative binomial distribution, with mu=mean and size set using a method of moments estimator for dispersion.

# Usage

```
makeInfReps(y, numReps, minDisp = 0.001)
```

#### **Arguments**

y a SummarizedExperiment numReps how many inferential replicates

minDisp the minimal dispersion value, set after method of moments estimation from in-

ferential mean and variance

#### **Details**

Note that these simulated counts only reflect marginal variance (one transcript or gene at a time), and do not capture the covariance of counts across transcripts or genes, unlike imported inferential replicate data. Therefore, makeInfReps should not be used with summarizeToGene to create genelevel inferential replicates if inferential replicates were originally created on the transcript level. Instead, import the original inferential replicates.

# Value

a SummarizedExperiment

makeSimSwishData 13

#### References

Van Buren, S., Sarkar, H., Srivastava, A., Rashid, N.U., Patro, R., Love, M.I. (2020) Compression of quantification uncertainty for scRNA-seq counts. bioRxiv. https://doi.org/10.1101/2020.07.06.189639

#### **Examples**

```
library(SummarizedExperiment)
mean <- matrix(1:4,ncol=2)
variance <- mean
se <- SummarizedExperiment(list(mean=mean, variance=variance))
se <- makeInfReps(se, numReps=50)</pre>
```

makeSimSwishData

Make simulated data for swish for examples/testing

#### **Description**

Makes a small swish dataset for examples and testing. The first six genes have some differential expression evidence in the counts, with varying degree of inferential variance across inferential replicates (1-2: minor, 3-4: some, 5-6: substantial). The 7th and 8th genes have all zeros to demonstrate labelKeep.

# Usage

```
makeSimSwishData(
  m = 1000,
  n = 10,
  numReps = 20,
  null = FALSE,
  meanVariance = FALSE,
  allelic = FALSE,
  diffAI = FALSE,
  dynamicAI = FALSE
)
```

#### **Arguments**

m number of genesn number of samples

numReps how many inferential replicates to generate null logical, whether to make an all null dataset

meanVariance logical, whether to output only mean and variance of inferential replicates

allelic logical, whether to make an allelic sim dataset

diffAI logical, whether to make a differential allelic sim dataset dynamicAI logical, whether to make a dynamic allelic sim dataset

14 makeTx2Tss

#### Value

a SummarizedExperiment

# **Examples**

```
library(SummarizedExperiment)
y <- makeSimSwishData()
assayNames(y)</pre>
```

makeTx2Tss

Make a GRanges linking transcripts to TSS within gene

# **Description**

This helper function takes either a TxDb/EnsDb or GRanges object as input and outputs a GRanges object where transcripts are aggregated to the gene + TSS (transcription start site). For nearby TSS that should be grouped together, see maxgap.

# Usage

```
makeTx2Tss(x, maxgap = 0)
```

# Arguments

x either TxDb/EnsDb or GRanges object. The GRanges object should have meta-

data columns tx\_id and gene\_id

maxgap integer, number of basepairs to use determining whether to combine nearby TSS

#### Value

GRanges with columns tx\_id, tss, and group\_id

# **Examples**

```
## Not run:
library(EnsDb.Hsapiens.v86)
edb <- EnsDb.Hsapiens.v86
t2t <- makeTx2Tss(edb)
## End(Not run)</pre>
```

miniSwish 15

miniSwish

Helper function for distributing Swish on a subset of data

# **Description**

This function is called by the Snakefile that is generated by splitSwish. See alevin example in the vignette. As such, it doesn't need to be run by users in an interactive R session.

#### Usage

```
miniSwish(
   infile,
   outfile,
   numReps = 20,
   lengthCorrect = FALSE,
   overwrite = FALSE,
   ...
)
```

# **Arguments**

infile path to an RDS file of a SummarizedExperiment

outfile a CSV file to write out

numReps how many inferential replicates to generate

lengthCorrect logical, see scaleInfReps, and Swish vignette. As this function is primarily for

alevin, the default is FALSE

overwrite logical, whether outfile should overwrite an existing file

... arguments passed to swish

#### Details

Note that the default for length correction is FALSE, as opposed to the default in scaleInfReps which is TRUE. The default for numReps here is 20.

# Value

nothing, files are written out

plotAllelicGene

Plot allelic counts in a gene context using Gviz

#### **Description**

Plot allelic data (allelic proportions, isoform propostions) in a gene context leveraging the Gviz package. See the allelic vignette for example usage. TPM and count filters are used by default to clean up the plot of features with minimal signal; note that the isoform proportion displayed at the bottom of the plot is among the features that pass the filtering steps. If the function is not responding, it is likely due to issues connecting to UCSC servers to draw the ideogram, in this case set ideogram=FALSE.

16 plotAllelicGene

#### Usage

```
plotAllelicGene(
  у,
  gene,
  db,
  region = NULL,
  symbol = NULL,
  genome = NULL,
  tpmFilter = 1,
  isoPropFilter = 0.05,
  countFilter = 10,
  pc = 1,
  transcriptAnnotation = "symbol",
  labels = list(a2 = "a2", a1 = "a1"),
  qvalue = TRUE,
  log2FC = TRUE,
  ideogram = FALSE,
  cov = NULL,
  covFacetIsoform = FALSE,
  allelicCol = c("dodgerblue", "goldenrod1"),
  isoformCol = "firebrick",
  statCol = "black",
  gridCol = "grey80",
  baselineCol = "black",
  titleCol = "black",
  titleAxisCol = "black",
  titleBgCol = "white",
  geneBorderCol = "darkblue",
  geneFillCol = "darkblue",
  genomeAxisCol = "black",
  innerFontCol = "black",
)
```

# Arguments

У	a SummarizedExperiment (see swish)
gene	the name of the gene of interest, requires a column gene_id in the metadata columns of the rowRanges of y
db	either a TxDb or EnsDb object to use for the gene model
region	GRanges, the region to be displayed in the Gviz plot. if not specified, will be set according to the gene plus 20 of the total gene extent on either side
symbol	alternative to gene, to specify the gene of interest according to a column symbol in the metadata columns of the rowRanges of y
genome	UCSC genome code (e.g. "hg38", if not specified it will use the GenomeInfoDb::genome() of the rowRanges of y
tpmFilter	minimum TPM value (mean over samples) to keep a feature
isoPropFilter	minimum percent of isoform proportion to keep a feature
countFilter	minimum count value (mean over samples) to keep a feature
рс	pseudocount to avoid dividing by zero in allelic proportion calculation

plotAllelicGene 17

transcriptAnnotation

argument passed to Gviz::GeneRegionTrack("symbol", "gene", "transcript",

etc.)

labels list, labels for a2 (non-effect) and a1 (effect) alleles

logical, whether to inclue qualue track qvalue log2FC logical, whether to include log2FC track

logical, whether to include ideogram track ideogram

character specifying a factor or integer variable to use to facet the allelic proporcov

tion plots, should be a column in colData(y)

covFacetIsoform

logical, if cov is provided, should it also be used to facet the isoform proportion

track, in addition to the allelic proportion track

allelicCol the colors of the points and lines for allelic proportion isoformCol the colors of the points and lines for isoform proportion

statCol the color of the lollipops for q-value and log2FC

gridCol the color of the grid in the data tracks

baselineCol the color of the horizontal baseline for q-value and lo2gFC

titleCol font color of the side titles (track labels) titleAxisCol axis color of the side titles (track labels)

titleBgCol background color of the side titles (track labels) geneBorderCol the color of the borders and font in gene region track

geneFillCol the color of the fill in the gene region track

line color of the genome axis track genomeAxisCol

innerFontCol font color of genome axis track, ideogram, and allelic proportion legend

additional arguments passed to Gviz::plotTracks()

#### Value

nothing, a plot is displayed

#### References

The methods for allelic expression analysis are described in:

Euphy Wu, Noor P. Singh, Kwangbom Choi, Mohsen Zakeri, Matthew Vincent, Gary A. Churchill, Cheryl L. Ackert-Bicknell, Rob Patro, Michael I. Love. "Detecting isoform-level allelic imbalance accounting for inferential uncertainty" bioRxiv (2022) https://doi.org/10.1101/2022.08.12. 503785

This function makes use of the Gviz package that is described in:

Hahne, F., Ivanek, R. (2016). Visualizing Genomic Data Using Gviz and Bioconductor. In: Mathé, E., Davis, S. (eds) Statistical Genomics. Methods in Molecular Biology, vol 1418. Humana Press, New York, NY. https://doi.org/10.1007/978-1-4939-3578-9\_16

18 plotAllelicHeatmap

plotAllelicHeatmap Pl

Plot allelic ratio heatmap

#### **Description**

Plot allelic ratio heatmap over features and samples using the pheatmap package. The a1/(a2 + a1) ratio is displayed.

# Usage

```
plotAllelicHeatmap(
   y,
   idx,
   breaks = NULL,
   cluster_cols = FALSE,
   main = "Allelic ratio",
   stripAfterChar = "-",
   ...
)
```

#### **Arguments**

```
y a SummarizedExperiment (see swish)

idx a numeric or logical vector of which features to plot

breaks breaks passed along to pheatmap

cluster_cols logical, passed to pheatmap

main title of the plot

stripAfterChar for the column names, if specified will strip allelic identifiers after this character, default is hyphen. set to NULL to avoid this action
```

#### Value

nothing, a plot is displayed

# References

The methods for allelic expression analysis are described in:

Euphy Wu, Noor P. Singh, Kwangbom Choi, Mohsen Zakeri, Matthew Vincent, Gary A. Churchill, Cheryl L. Ackert-Bicknell, Rob Patro, Michael I. Love. "Detecting isoform-level allelic imbalance accounting for inferential uncertainty" bioRxiv (2022) https://doi.org/10.1101/2022.08.12.503785

This function makes use of the pheatmap package:

Kolde, Raivo. "Pheatmap: pretty heatmaps." R package version 1.2 (2012): 726.

other arguments passed to pheatmap

plotInfReps 19

plotInfReps Plot inferential replicates for a gene or transcript	plotInfReps	Plot inferential replicates for a gene or transcript	
--	-------------	--	--

# Description

For datasets with inferential replicates, boxplots are drawn for the two groups and potentially grouped by covariates. For datasets with only mean and variance, points and intervals (95 approximation) are drawn. Additionally, for numeric x values, points and intervals will be drawn and computeInfRV should be run first in order to add the mean and variance statistics.

# Usage

```
plotInfReps(
  у,
  idx,
  Х,
  cov = NULL,
 colsDrk = c("dodgerblue", "goldenrod4", "royalblue4", "red3", "purple4", "darkgreen"),
  colsLgt = c("lightblue1", "goldenrod1", "royalblue1", "salmon1", "orchid1",
    "limegreen"),
  xaxis,
  xlab,
  ylim,
  main,
  mainCol,
  legend = FALSE,
  legendPos = "topleft",
  legendTitle = FALSE,
  legendCex = 1,
  useMean = TRUE,
  q = qnorm(0.975),
  applySF = FALSE,
  reorder,
  thin,
  shiftX
)
```

### **Arguments**

У	a SummarizedExperiment (see swish)
idx	the name or row number of the gene or transcript
х	the name of the condition variable for splitting and coloring the samples or cells. Also can be a numeric, e.g. pseudotime, in which case, cov can be used to designate groups for coloring
cov	the name of the covariate for adjustment
colsDrk	dark colors for the lines of the boxes
colsLgt	light colors for the inside of the boxes
xaxis	logical, whether to label the sample numbers. default is TRUE if there are less than 30 samples

20 plotMASwish

xlab the x-axis label ylim y limits main title

mainCol name of metadata column to use for title (instead of rowname)

legend logical, show simple legend (default FALSE)
legendPos character, position of the legend (default "topleft")

legendTitle logical, whether to add the name of the grouping variable as a title on the legend

(default FALSE)

legendCex numeric, size of the legend (default 1)

useMean logical, when inferential replicates are not present or when x is continuous,

whether to use the mean assay or the counts assay for plotting

numeric, the quantile to use when plotting the intervals when inferential repli-

cates are not present or when x is continuous. Default is qnorm(.975) ~= 1.96

corresponding to 95 intervals

applySF logical, when inferential replicates are not present, should y\$sizeFactor be

divided out from the mean and interval plots (default FALSE)

reorder logical, should points within a group defined by condition and covariate be re-

ordered by their count value (default is FALSE, except for alevin data)

thin integer, should the mean and interval lines be drawn thin (the default switches

from 0 [not thin] to 1 [thinner] at n=150 cells, and from 1 [thinner] to 2 [thinnest]

at n=400 cells)

shiftX when x is continuous and cov is provided, the amount to shift the values on the

x-axis to improve visibility of the point and line ranges (will be subtracted from

the first level of cov and added to the second level of cov)

#### Value

nothing, a plot is displayed

#### **Examples**

```
y <- makeSimSwishData()
plotInfReps(y, 3, "condition")

y <- makeSimSwishData(n=40)
y$batch <- factor(rep(c(1,2,3,1,2,3),c(5,10,5,5,10,5)))
plotInfReps(y, 3, "condition", "batch")</pre>
```

plotMASwish

MA plot - log fold change over average counts

#### **Description**

MA plot - log fold change over average counts

#### Usage

```
plotMASwish(y, alpha = 0.05, sigcolor = "blue", ...)
```

salmonEC 21

### **Arguments**

```
y a SummarizedExperiment (see swish)
alpha the FDR threshold for coloring points
sigcolor the color for the significant points
... passed to plot
```

#### Value

nothing, a plot is displayed

# **Examples**

```
y <- makeSimSwishData()
y <- scaleInfReps(y)
y <- labelKeep(y)
y <- swish(y, x="condition")
plotMASwish(y)</pre>
```

salmonEC

Construct a sparse matrix of transcript compatibility counts from salmon output

# **Description**

Constructs a count matrix with equivalence class identifiers in the rows. The count matrix is generated from one or multiple 'eq\_classes.txt' files that have been created by running salmon with the -dumpEq flag. Salmon - https://doi.org/10.1038/nmeth.4197

# Usage

```
salmonEC(
  paths,
  tx2gene,
  multigene = FALSE,
  ignoreTxVersion = FALSE,
  ignoreAfterBar = FALSE,
  quiet = FALSE
)
```

# **Arguments**

paths 'Character' or 'character vector', path specifying the location of the 'eq\_classes.txt'

files generated with salmon.

tx2gene A 'dataframe' linking transcript identifiers to their corresponding gene identi-

fiers. Transcript identifiers must be in a column 'isoform\_id'. Corresponding

gene identifiers must be in a column 'gene\_id'.

multigene 'Logical', should equivalence classes that are compatible with multiple genes be

retained? Default is 'FALSE', removing such ambiguous equivalence classes.

22 scaleInfReps

ignoreTxVersion

logical, whether to split the isoform id on the '.' character to remove version information to facilitate matching with the isoform id in 'tx2gene' (default FALSE).

ignoreAfterBar logical, whether to split the isoform id on the 'l' character to facilitate matching with the isoform id in 'tx2gene' (default FALSE).

quiet

'Logical', set 'TRUE' to avoid displaying messages.

#### Value

A list with two elements. The first element 'counts' is a sparse count matrix with equivalence class identifiers in the rows. If multiple paths are specified, the columns are in the same order as the paths. The second element 'tx2gene\_matched' allows for linking those identifiers to their respective transcripts and genes.

#### **Details**

The resulting count matrix uses equivalence class identifiers as rownames. These can be linked to respective transcripts and genes using the 'tx2gene\_matched' element of the output. Specifically, if the equivalence class identifier reads 1|2|8, then the equivalence class is compatible with the transcripts and their respective genes in rows 1, 2 and 8 of 'tx2gene\_matched'.

# Author(s)

Jeroen Gilis

scaleInfReps

Scale inferential replicate counts

# **Description**

A helper function to scale the inferential replicates to the mean sequencing depth. The scaling takes into account a robust estimator of size factor (median ratio method is used). First, counts are corrected per row using the effective lengths (for gene counts, the average transcript lengths), then scaled per column to the geometric mean sequence depth, and finally are adjusted per-column up or down by the median ratio size factor to minimize systematic differences across samples.

#### Usage

```
scaleInfReps(
  lengthCorrect = TRUE,
 meanDepth = NULL,
 sfFun = NULL,
 minCount = 10,
 minN = 3,
  saveMeanScaled = FALSE,
  quiet = FALSE
```

splitSwish 23

#### **Arguments**

y a SummarizedExperiment with: infReps a list of inferential replicate count

matrices, counts the estimated counts matrix, and length the effective lengths

matrix

lengthCorrect whether to use effective length correction (default is TRUE)

meanDepth (optional) user can specify a different mean sequencing depth. By default the

geometric mean sequencing depth is computed

sfFun (optional) size factors function. An alternative to the median ratio can be pro-

vided here to adjust the scaledTPM so as to remove remaining library size dif-

ferences. Alternatively, one can provide a numeric vector of size factors

minCount for internal filtering, the minimum count

minN for internal filtering, the minimum sample size at minCount

saveMeanScaled store the mean of scaled inferential replicates as an assay 'meanScaled'

quiet display no messages

#### Value

a SummarizedExperiment with the inferential replicates as scaledTPM with library size already corrected (no need for further normalization). A column log10mean is also added which is the log10 of the mean of scaled counts across all samples and all inferential replicates.

#### **Examples**

```
y <- makeSimSwishData()
y <- scaleInfReps(y)</pre>
```

splitSwish

Function for splitting SummarizedExperiment into separate RDS files

#### **Description**

The splitSwish function splits up the y object along genes and writes a Snakefile that can be used with Snakemake to distribute running swish across genes. This workflow is primarily designed for large single cell datasets, and so the default is to not perform length correction within the distributed jobs. See the alevin section of the vignette for an example. See the Snakemake documention for details on how to run and customize a Snakefile: https://snakemake.readthedocs.io

# Usage

```
splitSwish(y, nsplits, prefix = "swish", snakefile = NULL, overwrite = FALSE)
```

#### Arguments

y a SummarizedExperiment

nsplits integer, how many pieces to break y into

prefix character, the path of the RDS files to write out, e.g. prefix="/path/to/swish"

will generate swish.rds files at this path

24 swish

snakefile character, the path of a Snakemake file, e.g. Snakefile, that should be written out. If NULL, then no Snakefile is written out

overwrite logical, whether the snakefile and RDS files (swish1.rds, ...) should overwrite existing files

#### Value

nothing, files are written out

#### References

Compression and splitting across jobs:

Van Buren, S., Sarkar, H., Srivastava, A., Rashid, N.U., Patro, R., Love, M.I. (2020) Compression of quantification uncertainty for scRNA-seq counts. bioRxiv. https://doi.org/10.1101/2020.07.06.189639

Snakemake:

Koster, J., Rahmann, S. (2012) Snakemake - a scalable bioinformatics workflow engine. Bioinformatics. https://doi.org/10.1093/bioinformatics/bts480

swish

Swish method: differential expression accounting for inferential uncertainty

# Description

The Swish method, or "SAMseq With Inferential Samples Helps". Performs non-parametric inference on rows of y for various experimental designs. See References for details.

# Usage

```
swish(
 у,
 х,
 cov = NULL,
 pair = NULL,
  interaction = FALSE,
 cor = c("none", "spearman", "pearson"),
 nperms = 100,
 estPi0 = FALSE,
 qvaluePkg = "qvalue",
 pc = 5,
 nRandomPairs = 30,
  fast = NULL,
 returnNulls = FALSE,
  quiet = FALSE
)
```

swish 25

#### **Arguments**

Х

y a SummarizedExperiment containing the inferential replicate matrices of medianratio-scaled TPM as assays 'infRep1', 'infRep2', etc.

the name of the condition variable. A factor with two levels for a two group analysis (possible to adjust for covariate or matched samples, see next two arguments). The log fold change is computed as non-reference level over reference

level (see vignette: 'Note on factor levels')

cov the name of the covariate for adjustment. If provided a stratified Wilcoxon in

performed. Cannot be used with pair, unless using either interaction or cor

pair the name of the pair variable, which should be the number of the pair. Can be an

integer or factor. If specified, a signed rank test is used to build the statistic by default. **Note:** For simple paired designs, see use of fast=1 for a much faster implementation of paired testing using a one-sample z-score test statistic. All samples across x must be pairs if this argument is specified. Cannot be used

with cov, unless using either interaction or cor

interaction logical, whether to perform a test of an interaction between x and cov. Can use

pair or not. See Details.

cor character, whether to compute correlation of x with the log counts, and signi-

fance testing on the correlation as a test statistic. Either "spearman" or "pearson" correlations can be computed. For Spearman the correlation is computed over ranks of x and ranks of inferential replicates. For Pearson, the correlation is computed for x and log2 of the inferential replicates plus pc. Default is "none", e.g. two-group comparison using the rank sum test or other alternatives listed above. Additionally, correlation can be computed between a continuous variable

cov and log fold changes across x matched by pair

nperms the number of permutations. if set above the possible number of permutations,

the function will print a message that the value is set to the maximum number

of permutations possible

estPi0 logical, whether to estimate pi0

qvaluePkg character, which package to use for q-value estimation, samr or qvalue

pc pseudocount for finite estimation of log2FC, not used in calculation of test statis-

tics, locfdr or qvalue

nRandomPairs the number of random pseudo-pairs (only used with interaction=TRUE and

un-matched samples) to use to calculate the test statistic

fast an integer (0 or 1), toggles different methods based on speed, currently only

relevant for simple paired analysis. For simple paired design, fast=1 triggers the use of a one-sample z-score instead of a signed rank statistic. The one-sample z-score is much faster (can be >10x faster), by avoiding the expensive re-computation of ranks during permutations. fast=1 is not relevant for

interaction or cor type designs

returnNulls logical, only return the stat vector, the log2FC vector, and the nulls matrix

(default FALSE)

quiet display no messages

#### **Details**

**interaction:** The interaction tests are different than the other tests produced by swish, in that they focus on a difference in the log2 fold change across levels of x when comparing the two levels in

26 swish

cov. If pair is specified, this will perform a Wilcoxon rank sum test on the two groups of matched sample LFCs. If pair is not included, multiple random pairs of samples within the two groups are chosen, and again a Wilcoxon rank sum test compared the LFCs across groups.

#### Value

a SummarizedExperiment with metadata columns added: the statistic (either a centered Wilcoxon Mann-Whitney or a signed rank statistic, aggregated over inferential replicates), a log2 fold change (the median over inferential replicates, and averaged over pairs or groups (if groups, weighted by sample size), the local FDR and q-value, as estimated by the samr package.

#### References

The citation for swish method is:

Anqi Zhu, Avi Srivastava, Joseph G Ibrahim, Rob Patro, Michael I Love "Nonparametric expression analysis using inferential replicate counts" Nucleic Acids Research (2019). https://doi.org/10.1093/nar/gkz622

The swish method builds upon the SAMseq method, and extends it by incorporating inferential uncertainty, as well as providing methods for additional experimental designs (see vignette).

For reference, the publication describing the SAMseq method is:

Jun Li and Robert Tibshirani "Finding consistent patterns: A nonparametric approach for identifying differential expression in RNA-Seq data" Stat Methods Med Res (2013). https://doi.org/10.1177/0962280211428386

#### **Examples**

```
library(SummarizedExperiment)
set.seed(1)
y <- makeSimSwishData()</pre>
y <- scaleInfReps(y)</pre>
y <- labelKeep(y)</pre>
y <- swish(y, x="condition")</pre>
# histogram of the swish statistics
hist(mcols(y)$stat, breaks=40, col="grey")
cols = rep(c("blue","purple","red"),each=2)
for (i in 1:6) {
  arrows(mcols(y)$stat[i], 20,
         mcols(y)$stat[i], 10,
         col=cols[i], length=.1, lwd=2)
}
# plot inferential replicates
plotInfReps(y, 1, "condition")
plotInfReps(y, 3, "condition")
plotInfReps(y, 5, "condition")
```

# **Index**

```
* package
    fishpond-package, 2
addStatsFromCSV, 3
alevinEC, 4
computeInfRV, 5, 19
deswish, 6
fishpond-package, 2
getTrace, 7
{\tt importAllelicCounts}, {\tt 7}
\verb|isoformProportions|, 9
labelKeep, 3, 9
loadFry, 10
makeInfReps, 12
makeSimSwishData, 13
makeTx2Tss, 14
miniSwish, 15
plotAllelicGene, 15
\verb|plotAllelicHeatmap|, 18|
plotInfReps, 3, 5, 19
plotMASwish, 3, 20
salmonEC, 21
scaleInfReps, 3, 15, 22
splitSwish, 3, 15, 23
swish, 3, 15, 24
```