# Package 'mobileRNA'

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Type Package

**Title** mobileRNA: Investigate the RNA mobilome & population-scale changes

Version 1.5.1

**Description** Genomic analysis can be utilised to identify differences between RNA populations in two conditions, both in production and abundance. This includes the identification of RNAs produced by multiple genomes within a biological system. For example, RNA produced by pathogens within a host or mobile RNAs in plant graft systems. The mobileRNA package provides methods to pre-process, analyse and visualise the sRNA and mRNA populations based on the premise of mapping reads to all genotypes at the same time.

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**Depends** R (>= 4.3.0)

VignetteBuilder knitr

**Encoding UTF-8** 

LazyData FALSE

**Roxygen** list(markdown = TRUE)

RoxygenNote 7.3.1

Suggests knitr, rmarkdown, BiocStyle

Imports dplyr, tidyr, ggplot2, BiocGenerics, DESeq2, edgeR, ggrepel, grDevices, pheatmap, utils, tidyselect, progress, RColorBrewer, GenomicRanges, rtracklayer, data.table, SimDesign, scales, IRanges, stats, methods, Biostrings, reticulate, S4Vectors, Seqinfo, SummarizedExperiment, rlang, bioseq, grid

**biocViews** Visualization, RNASeq, Sequencing, SmallRNA, GenomeAssembly, Clustering, ExperimentalDesign, QualityControl, WorkflowStep, Alignment, Preprocessing

**BiocType** Software

BugReports https://github.com/KJeynesCupper/mobileRNA/issues

**SystemRequirements** GNU make, ShortStack (>= 4.0), HTSeq, HISAT2, SAMtools, Conda

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mapR	NA mobileRNA pre-processing method for sRNAseq & mRNAseq (align-	

# Description

The mobileRNA workflow includes specific pre-processing guidelines. For sRNAseq, this undertakes alignment with Bowtie and sRNA cluster analysis with ShortStack. For mRNAseq, this undertakes alignment with HISAT2 and HTSeq. All OS software should be installed within a Conda environment.

ment, raw count or cluster analysis)

#### **Usage**

```
mapRNA(
  input = c("mRNA", "sRNA"),
  sampleData = NULL,
  tidy = TRUE,
  input_files_dir,
  output_dir,
  genomefile,
  annotationfile = NULL,
  condaenv,
  threads = 6,
  mmap = "n",
  dicermin = 20,
  dicermax = 24,
  mincov = 0.5,
  pad = 200,
  order = "pos",
  stranded = "no",
  a = 10,
  mode = "union",
  nonunique = "none",
  type = "mRNA",
  idattr = "Name"
)
```

## **Arguments**

input

string; define type of Next-Generation Sequencing data set. "sRNA" for sR-

NAseq data and "mRNA" for mRNAseq data.

sampleData dataframe; stores mRNA sample data where rows represent each sample in the

analysis. Column one stores the sample names, while column two stores the the name(s) of the fastq file(s) for mate 1 (e.g. flyA\_1.fq, flyB\_1.fq) and column three stores the the name(s) of the fastq file(s) for mate 2 (e.g. flyA\_2.fq,flyB\_2.fq). If data is single ended, column three will not hold any values. Only for mRNA

analysis.

tidy logical; removes unnecessary extra output files when set to TRUE.

input\_files\_dir

path; directory containing only the FASTQ samples for analysis. Note that all

samples in this directory will be used by this function.

output\_dir path; directory to store output.

path; path to a FASTA genome reference file. genomefile

path; path to a GFF file. For mRNA analysis only. annotationfile

character; name or directory of the Conda environment to use where OS depencondaenv

dencies are stored.

threads numeric; set the number of threads to use where more threads means a faster

completion time. Default is 6.

character; define how to handle multi-mapped reads. Choose from "u", "f", "r" mmap

> or "n". For core sRNA analysis, use either "u", "f" or "r" options. Where "u" means only uniquely-aligned reads are used as weights for placement of multimapped reads. Where "f" means fractional weighting scheme for placement of

> multi-mapped reads and "r" mean multi-mapped read placement is random. For core mRNA analysis, to include multimapped reads, use any parameter, other than "n". While for mobile sRNA or mRNA, it is important to use "n", to not consider multi-mapped reads, only unique reads as we cannot distinguish which genome the reads mapped to multiple locations in. integer; the minimum size in nucleotides of a valid small RNA. This option

dicermin

sets the bounds to discriminate dicer-derived small RNA loci from other loci.

Default is 20. For sRNA analysis only.

integer; the maximum size in nucleotides of a valid small RNA. This option dicermax

sets the bounds to discriminate dicer-derived small RNA loci from other loci.

Default is 24. For sRNA analysis only.

numeric; minimum alignment depth, in units of reads per million, required to mincov

nucleate a small RNA cluster during de novo cluster search. Must be a number

> 0. Default is 2. For sRNA analysis only.

integer; initial peaks are merged if they are this distance or less from each other. pad

Must  $\geq$  1, default is 75. For sRNA analysis only.

character; either "name" or "pos" to indicate how the input data has been sorted. order

For paired-end data only, this sorts the data either by read name or by alignment

position. Default is "pos", to sort by position. For mRNA analysis only.

stranded whether the data is from a strand-specific assay, either "yes"/"no"/"reverse". De-

fault is "no". For mRNA analysis only.

numeric; skip all reads with alignment quality lower than the given minimum а

value (default: 10). For mRNA analysis only.

mode character; states mode to handle reads overlapping more than one feature. Either

"union", "intersection-strict" and "intersection-nonempty". Default is "union".

For mRNA analysis only.

character; states the mode to handle reads that align to or are assigned to more nonunique

> than one feature in the overlap. Either "none" and "all". Default is "none". For mobile mRNA, ensure the default is utilized to exclude multimapped reads. For

mRNA analysis only.

character; feature type (3rd column in GFF file) to be used, all features of other type

type are ignored. Default is "mRNA". For mRNA analysis only.

idattr character; GFF attribute to be used as feature ID. Several GFF lines with the

> same feature ID will be considered as parts of the same feature. The feature ID is used to identity the counts in the output table. Default is "Name". For mRNA

analysis only.

# **Details**

Please ensure all OS software is installed within a Conda environment. See appendix of vignette for manual pipeline. Alignment statistics are reported for each analysis within log plain text files (log.txt).

In order to align reads, the function will check whether a genome reference index has already been generated and, if not, will generate one. The method varies between sRNA and mRNA analysis depending on the alignment tool. This is generated in the same location as the reference file.

NOTE: This function utilises the reticulate R package to connect to the conda environment. Hence, restart R if you wish to change the employed Conda environment during a session.

**For sRNA analysis** The function invokes a number of OS commands, and is dependent on the installation of ShortStack (>= 4.0) with Conda. Please note that ShortStack is only compatible with Linux and Mac operating systems.

The pipeline undertakes de novo detection of sRNA-producing loci and alignment, where the output of each are stored in their respective folders in the users desired location. The de novo detection of sRNA-producing loci analyses each sample to identify de novo sRNA-producing loci (ie. sRNA clusters), and joins these results into a single file called "locifile.txt". The alignment step aligns and clusters each sample to the genome reference along with the file containing the de novo sRNA clusters. The final reports are imported into R using RNAimport().

#### For mRNA analysis

The function invokes a number of OS commands, and is dependent on the installation of HISAT2,HTSeq and SAMtools with Conda. The pipeline can undertake single- or pair-end analysis, and to do so requires a data frame stating the sample information where each row represents a sample. The reads are mapped using HISAT and then the raw counts are estimated by HTSeq. The output alignment file (BAM) and raw counts file for each sample are stored within the samples own folder within the desired directory.

#### Value

- \*\* For sRNA analysis\*\* The OS commands generate output into the users desired location, generating two folders:
  - 1\_de\_novo\_detection: Stores output from the detection of de novo sRNA-producing loci
  - 2 sRNA results: Stores results

The first folder stores the alignment (BAM) and the de novo sRNA-producing loci for each sample (.txt) within the samples respective folder. The analyses joins the de novo sRNA clusters across the experimental design which is stored in "locifile.txt". The second folder stores the final clustering results for each sample, and as before the results of each sample are stored within it's respective folder. These results (.txt) are imported into R for downstream analysis by utilizing the RNAimport() function.

The function generates a number of extra files for each sample and are not required for the down-stream analysis. These are generated by ShortStack, see documentation for more information (https://github.com/MikeAxtell/ShortStack). As default these files are deleted. This is determined by the tidy argument.

\*\* For mRNA analysis\*\* For mRNA analysis, generate a new folder which stores the results in the users desired output location, known as "1\_mRNA\_preprocessing". Within this folder, there will contain one folder per sample storing it's sorted alignment file (BAM) and raw counts file ("Result.txt"). Note, that the function excludes multi-mapped mRNAs.

#### References

ShortStack https://github.com/MikeAxtell/ShortStack, HISAT2 https://anaconda.org/bioconda/hisat2, HTSeq https://htseq.readthedocs.io/en/master/install.html, SAM-tools https://anaconda.org/bioconda/samtools

# **Examples**

```
## Not run:
## EXAMPLE 1 - sRNAseq
samples <- file.path(system.file("extdata/sRNAseq",package="mobileRNA"))</pre>
```

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```
GenomeRef <- system.file("extdata","reduced_chr12_Eggplant.fa.gz", package="mobileRNA")</pre>
output_location <- tempdir()</pre>
mapRNA(input = "sRNA",
input_files_dir = samples,
output_dir = output_location,
genomefile = GenomeRef,
condaenv = "ShortStack4",
mmap = "n")
## EXAMPLE 2 - mRNAseq
# create sample data including name, and file mates:
sampleData <- data.frame(sample = c("selfgraft_1", "selfgraft_2",</pre>
                                     "heterograft_1", "heterograft_2"),
                    mate1 = c("selfgraft_mRNAdemo_1.fq.gz", "selfgraft_mRNAdemo_2.fq.gz",
                         "heterograft_mRNAdemo_1.fq.gz", "heterograft_mRNAdemo_2.fq.gz"))
# location of samples:
samples <- system.file("extdata/mRNAseq", package="mobileRNA")</pre>
# location to store output
output_location <- tempdir()</pre>
# run alignment
mapRNA(input = "mRNA",
      input_files_dir = samples,
      output_dir = output_location,
      genomefile = output_assembly_file,
      annotationfile = output_annotation_file,
      sampleData = sampleData,
      condaenv = "/Users/user-name/miniconda3")
## End(Not run)
```

mobileRNA

mobileRNA: Explore RNA mobilome & population-scale changes

# Description

Uses small RNA or messenger RNA sequencing data in two conditions and identifies changes in the RNA population. mobileRNA was primarily designed for the identification of a putative RNA mobilome in a chimeric system. For example, in plant graft systems. As input, mobileRNA takes sRNAseq or mRNAseq fastq files. Output consists of a data frame with putative differences between two conditions along with a number of plots.

#### Details

The most important functions in the **mobileRNA** are:

```
RNAmergeGenomes Merge two genome assembly files (FASTA).

RNAmergeAnnotations Merge two genome annotation files (GFF).

mapRNA Pre-processing of sRNAseq and mRNAseq (alignment, raw count, cluster analysis).
```

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RNAimport Reads the pre-processing report files into a dataframe for all conditions

RNAdicercall Calculates the consensus sRNA dicercall class.

RNAsubset Subsets the data set based on the sRNA class.

RNAdifferentialAnalysis Undertakes differential analysis with either the edgeR or DESeq2 method.

RNAmobile Identify putative RNA molecules produced by the non-tissue sample genome

RNApopulation Identify gained/lost RNA populations between treatment and control conditions.

RNAsummary Summarise the differential abundance of RNAs.

RNAreorder Reorder the data frame for differential analysis, ensuring control verse treatment comparison.

RNAsequences Extract RNA sequence from sRNA clusters.

RNAattributes Overlap the genomic features related to the sRNA clusters.

RNAdistribution Plot the distribution of sRNA classes based on nucleotide length.

plotHeatmap Heatmap of log-transformed normalization data.

plotSampleDistance Plots a sample distance heatmap for quality control.

plotSamplePCA Plots a PCA plot, customize ratio, colours and shapes.

RNAfeatures Summarise the distribution of sRNA clusters across genomic features.

RNAdf2se Convert a mobileRNA dataframe to a SummarizedExperiment object.

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#### See Also

See vignette("mobileRNA", package = "mobieRNA") for an overview of the package.

 $mRNA\_data$ 

mRNA\_data: simulated messenger RNA data for biological replicates

#### **Description**

Simulated mRNAseq dataset

#### Usage

data(mRNA\_data)

## Details

Simulates data is taken from eggplant and tomato mRNAseq samples and created to simulate to movement of mRNA molecules from an Tomato rootstock to an Eggplant Scion. Two Eggplant replicates were spiked with the same 150 tomato mRNA clusters, and named "heterograft\_" 1 to 2. The analysis compares these heterografts to two Eggplant self-grafts which are the original unspiked Eggplant replicates, called "selfgraft\_" 1 to 2.

This data was imported and organised by the RNAimport() function.

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#### Value

Dataframe in global environment

#### **Examples**

```
data("mRNA_data")
```

plotHeatmap

Heatmap of log-transformed normalization data

#### **Description**

Undertakes normalisation of RPM/FPKM using a pseudocount and transforms the data using log-scale, to enable visualization of the differences and patterns in expression across samples using a heatmap.

## Usage

```
plotHeatmap(
   data,
   value = "RPM",
   pseudocount = 1e-06,
   colours = (grDevices::colorRampPalette(RColorBrewer::brewer.pal(9, "GnBu")))(100),
   cluster = TRUE,
   scale = "none",
   clustering_method = "complete",
   row.names = FALSE,
   border.color = NA,
   column.angle = 45,
   title = NA
)
```

## **Arguments**

data data.frame; originally generated by RNAimport()
value character; state the values to plot, either FPKM or RPM.

pseudocount numeric; pseudo count, default is 1e-6

colours character; colors. Default is grDevices::colorRampPalette(RColorBrewer::brewer.pal(9,

"GnBu"))(100)

cluster logical; include hierarchical clustering when default cluster= TRUE

scale character; indicating whether the values should be centered & scaled in either

the row direction or the column direction, or none. Respective options are "row",

"column" & "none". Default is scale="none".

clustering\_method

character; clustering method used. Accepts the same values as helust. Default

clustering\_method= "complete"

row.names logical; indicated whether to include rownames from column 1 of data frame.

Default row.names=FALSE

border.color character; the border colour. Default is no border (NA).

column angle  $\alpha$  numeric; angle of the column labels, choose from 0, 45, 90, 270 or 315.

title character; states plot title, placed at the top center

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#### **Details**

Undertakes FPKM/RPM normalisation using a pseudocount and then transforms the normalised-RPM data using log-scale.

This function expects to receive a data frame containing FPKM/RPM data. This function employs the use of a pseudo count during normalisation as the function is expected to be used when identifying mobile sRNAs in a chimeric system. In such system, it is expected that control replicates will contain zero values for the candidate mobile sRNA clusters.

#### Value

Produces a list objects storing the heatmap plot and the data.

#### **Examples**

```
data("sRNA_data")

# vector of control names
controls <- c("selfgraft_1", "selfgraft_2" , "selfgraft_3")

# Locate potentially mobile sRNA clusters associated to tomato, no
# statistical analysis
sRNA_data_mobile <- RNAmobile(input = "sRNA", data = sRNA_data,
controls = controls, genome.ID = "B_", task = "keep", statistical = FALSE)

# plot heatmap of potential mobile sRNAs
p1 <- plotHeatmap(sRNA_data_mobile)</pre>
```

plotRNAfeatures

Plots the distribution of genomic features in the genome and those that overlap with sRNA clusters

## **Description**

Using the RNAfeatures() function, this function plots the percentage distribution of genomic features in the provided genome annotation and percentage the distibution of genomic features which overlap sRNA clusters. This is illustrated as a stacked bar plot.

#### Usage

```
plotRNAfeatures(
  data,
  annotation,
  repeats = NULL,
  promoterRegions = 1000,
  repeat.type = NULL,
  brewerPalette = "Spectral",
  x.axis.text = c("Genome", "Dataset"),
  legend.position = "bottom"
)
```

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#### **Arguments**

data data.frame; generated by RNAimport()

annotation path; URL or connection to a GFFFile object. A genome reference annotation

file (.gff/.gff1/.gff2/.gff3). Can be in compressed format (gzip).

repeats path; URL or connection to a GFFFile object. A genome reference annota-

tion file, which only contains information on repeat sequences in the genome (.gff/.gff1/.gff2/.gff3). By default, this is not required, however if there is a specific repeats annotation file for the genome it is suggested to supply it. Can be

in compressed format (gzip).

promoterRegions

numeric; defines the upstream promoter region of genes. Default is 1000, which

refers to promoters set at 1Kb upstream of genes

repeat.type character; features type in annotation file to represent repeats or transposable

elements when repeats not supplied. Default is c("transposable\_element", transposable\_ele

which represent the transposable element features in the TAIR10 genome anno-

tation.

brewerPalette vector; colour scales from ColorBrewer to support the ggplot2::scale\_fill\_brewer

function. Default is "Spectral".

x.axis.text vector; labs to represent the genome and the dataset bars in the plot. Default is

x.axis.text=c("Genome", "Dataset").

legend.position

character; position of legend. Either "none", "left", "right", "bottom", "top",

"inside".

## **Details**

RNAfeatures calculates the number or percentage of sRNA clusters which overlap with genomic features based on their genomic coordinates.

# Value

Returns a table containing the number or percentage of overlaps in the supplied sRNA data set with specific regions in the genome annotation such as genes, repeats, introns, exons.

## See Also

RNAmergeAnnotations() to merge 2 GFF files into 1.

## **Examples**

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|--|

# Description

Draws a simple hierarchical clustered heatmap to observe sample distance.

# Usage

```
plotSampleDistance(
  data,
  colours = (grDevices::colorRampPalette(rev(RColorBrewer::brewer.pal(9, "GnBu"))))(255),
  vst = FALSE,
  cellheight = 40,
  cellwidth = 40
)
```

# Arguments

data	data.frame; originally generated by RNAimport()
colours	character; the colour palette. Default is $grDevices::colorRampPalette(RColorBrewer::brewer."GnBu"))(100)$
vst	logical; to undertake variance stabilizing transformation. By default, the function uses a regularized log transformation on the data set, however, this will not suit all experimental designs.
cellheight	numeric; individual cell height in points. Default is 40.

# **Details**

cellwidth

In special conditions, regularized log transformation will not suit the experimental design. For example, an experimental design without replicates. In this instance, it is preferable to change the default setting and switch to a variance stabilizing transformation method (vst=TRUE).

numeric; individual cell width in points. Default is 40.

# Value

A blue/green scale heatmap illustrating the sample distance.

## **Examples**

```
data("sRNA_data")
p1 <- plotSampleDistance(sRNA_data)</pre>
```

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# Description

Draws a principal component analysis (PCA) plot of PC1 and PC2. The function undertakes rlog transformation of the data in an unbiased manner (blind=TRUE).

# Usage

```
plotSamplePCA(
  data,
  group,
  vst = FALSE,
  labels = TRUE,
  boxed = TRUE,
  legend.title = "Conditions",
  size.ratio = 2,
  colours = NULL,
  point.shape = TRUE,
  ggplot.theme = NULL,
  label.box.padding = 1,
  title = "PCA plot",
  legend.position = "top",
  legend.direction = "horizontal"
)
```

# Arguments

٤	Suments	
	data	data.frame; originally generated by RNAimport()
	group	character; contains experimental conditions for each replicate.
		<i>IMPORTANT</i> : Ensure this is in the same order as the replicates are found in the data frame supplied to the data argument (from left to right).
	vst	logical; variance stabilizing transformation. By default, the function uses a regularized log transformation on the data set, however, this will not suit all experimental designs.
	labels	logical; include sample name labels on PCA. Default labels=TRUE
	boxed	logical; add a box around each sample name label. Default boxed=TRUE
	legend.title	character; title for legend key. Default legend.title = "Conditions"
	size.ratio	numeric; set plot ratio, broadens axis dimensions by ratio. Default size.ratio=2, double the plot dimension.
	colours	character; vector of HEX colour codes. Must match the number of conditions.
	point.shape	logical; set whether the point shapes should be different for each condition.
	ggplot.theme	character; state the ggplot2 theme (without () brackets). For example, ggplot.theme=ggplot2::the
label.box.padding		
		numeric; Amount of padding around bounding box, as a unit or number. Defaults to 1.
	title	character; title for plot.

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#### **Details**

This function uses the DESeq2 package to organise and plot the data. It organises the data into a DESeqDataSet which undergoes log-transformation where the results are used to undertake the PCA analysis. The results are plotted against the principal components 1 and 2.

In special conditions, regularized log transformation will not suit the experimental design. For example, an experimental design without replicates. In this instance, it is preferable to change the default setting and switch to a variance stabilizing transformation method (`vst=TRUE`).

#### Value

A PCA plot to show sample distance.

## **Examples**

```
data("sRNA_data")
groups <- c("Heterograft", "Heterograft", "Selfgraft", "Selfgraft", "Selfgraft")

p <- plotSamplePCA(data = sRNA_data, group = groups)
plot(p)</pre>
```

RNAattributes

Overlap the genomic features related to the sRNA clusters

## **Description**

Overlap the genomic features related to the sRNA clusters

# Usage

```
RNAattributes(
  data,
  annotation,
  match = c("within", "genes"),
  bufferRegion = 1000
)
```

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#### **Arguments**

data data.frame; originally generated by RNAimport() or containing chr, start and

end columns.

annotation path; URL, connection or GFFFile object. A genome reference annotation file

(.gff/.gff1/.gff2/.gff3).

match character; must be either "within" or "genes". Where "within" will return matches

where the clusters can be found within any annotation, while "genes" will return

matches where the clusters can be found within only genes.

bufferRegion numeric; a buffer region in base-pairs to extend the start and end coordinates

upstream and downstream respectively.

#### **Details**

Based on genomic coordinates, assign sRNA clusters with matching annotation information. This function can be used to find the genomic features from which the sRNA clusters originate from. This includes genes or repetitive regions. An additional buffer region at the start/end of the gene is added to improve hits, and align with the assumptions about promoter regions.

It is important that any alteration which were made to the genome reference (FASTA) used for aligment/clustering, such as alterations to the chromosome name, must be carried forth to the genome annotation file. See RNAmergeGenomes() and RNAmergeAnnotations() for more information.

#### Value

Appends the attribute columns from the GFF file to the supplied data based on overlapping genomic regions.

## **Examples**

RNAdf2se

Convert a mobileRNA dataframe to a SummarizedExperiment object

#### **Description**

Convert any mobileRNA output dataframe into a SummarizedExperiment object.

#### Usage

```
RNAdf2se(input = c("sRNA", "mRNA"), data)
```

# **Arguments**

input character; must be either "sRNA" or "mRNA" data data.frame produced by the **mobileRNA** package.

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#### **Details**

The function relies on the naming structure of columns created by functions in the **mobileRNA** package. It is able to extract the sample names based on these additions, and organise the data appropriately.

#### Value

A SummarizedExperiment object containing information from working data frame.

#### #'For sRNAseq data

- The rownames contain the locus name and the cluster name.
- The assays represent the additional information including DicerCall, Count, RPM, Major-RNA.
- The rowData includes the Cluster ID, the DicerCounts & the DicerConsensus.
- The colnames represents the sample replicate names.

## For mRNAseq data

- The rownames contain the gene names.
- The assays represent the additional information including Count & FPKM.
- The rowData includes the gene & the SampleCounts.
- The colnames represents the sample replicate names.

## **Examples**

```
# load data.frame
data("sRNA_data")
se <- RNAdf2se(input = "sRNA", data = sRNA_data)</pre>
```

RNAdicercall

Define the consensus dicercall for each sRNA cluster

# Description

The sRNA dicercall represents the length in nucleotides of the most abundant sRNA sequence within a cluster. The function calculates the consensus dicercall classification.

## Usage

```
RNAdicercall(
  data,
  conditions = NULL,
  ties.method = NULL,
  tidy = FALSE,
  chimeric = FALSE,
  controls = NULL,
  genome.ID = NULL
)
```

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## **Arguments**

data data.frame; originally generated by RNAimport().

conditions character; vector containing sample replicate names. When supplied, the data

from the named replicates will be the only ones used to calculate the dicercall consensus for each sRNA cluster. Each string should represent a sample name

present in the dataframe supplied to the data argument.

ties.method character; string specifying how ties are handled, choose either "exclude" or

"random". When using random, if there is a tie one of the classes will be chosen at random. While, when using exclude if there is a tie the class is set to undefined, however, if there is a tie between a undefined and a known class, the known class takes president (eg 3x24-nt and 3xN-nt, then it will be classed as

24nt). Default setting ties.method="exclude".

tidy logical; tidy-up data by removing sRNA clusters with an unknown or unclassi-

fied result. Default setting tidy=FALSE, removes excess background noise.

chimeric logical; state whether the system is chimeric and contains multiple genomes/genotypes.

controls character; vector of control condition sample names.

genome. ID character; chromosome identifier of the genome representing either the origin of

mobile molecules or the other genome in the chimeric system.

#### **Details**

For each sample, the alignment/clustering step predicted the sRNA dicercall for each cluster. This value is stored in the columns starting with "DicerCall\_". This value represents the length of nucleotides of the most abundant sRNA within the cluster. For some clusters, there is no particular sRNA which is more abundant than another, hence, it is stated as "NA" or "N", which is referred to as unclassified. The RNAdicercall() function calculate the consensus dicercall for each sRNA cluster based on the values across replicates. There are several parameters which will alter the output, including the handling of ties and the method to draw the consensus from.

When ties.method = "random", as per default, ties are broken at random. In this case, the determination of a tie assumes that the entries are probabilities: there is a relative tolerance of 1e-5, relative to the largest (in magnitude, omitting infinity) entry in the row.

When ties.method = "exclude", ties between sRNA classification are ruled as unclassified ("N"). However, when there is a tie between the choice of a class or unclassified result the exclude option will always select the class choice over the unclassified result.

If users are working with a chimeric system, utilise the chimeric=TRUE parameter and state genome. ID and controls parameter variables. This will remove any potential mapping errors which could introduce false interpretation.

To remove excess data noise, tidy=TRUE can be used to removed unclassified ("N") sRNA clusters, resulting in a reduced data set size.

#### Value

The original input data with two additional columns appended known as DicerCounts and DicerConsensus. The DicerCounts column stores the number of replicates that contributed to defining the consensus dicer-derived sRNA class. Note that when utilising the exclude ties methods, the DicerCounts will be represented as 0 when a tie is identified. While, the DicerConsensus stores the consensus dicercall.

#### **Examples**

RNAdifferentialAnalysis

Differential Analysis using DESeq2 or edgeR

## **Description**

RNAdifferentialAnalysis function computes the differential analysis with DESeq2 or edgeR of sRNA or mRNA data produced by the mobileRNA package pipeline.

# Usage

```
RNAdifferentialAnalysis(
  data,
  group,
  method = c("edgeR", "DESeq2"),
  dispersionValue = NULL
)
```

## **Arguments**

data data.frame; originally generated by RNAimport().

group character; the condition of each sample in the experimental design, formatted in

the order of samples shown in the data object from left to right.

method character; method to undertaken differential analysis, choose from methods of

either DESeq2::DESeq or edgeR::edgeR. Must be stated as either "DESeq2" or

"edgeR" in the function.

dispersionValue

numeric; value to represent the dispersion value for the edgeR::edgeR method. Recommended for analysis in experiments without biological replicates.

## **Details**

The user has the flexibility to choose the method that best suits their data. In this function, the DESeq2 method, calculates the differentials based on the normalized count data based on the size factors. Whereas the edgeR method, calculates normalization factors estimates dispersion, and returns the common dispersion. After normalization, the mean expression levels across samples are calculated, and differential expression analysis is performed using the exact test within groups, and the adjusted p-values are calculated using the Benjamini-Hochberg method. Note that this function

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is only capable of handling one replicate per condition with the edgeR method. This requires setting a suitable dispersion value. The dispersion value is other wise known as the common Biological squared coefficient of variation. See the User's Guide for the edgeR package for more details, edgeR::edgeR.

#### Value

Undertakes differential analysis, based on a specified method, and appends the results to the supplied data frame. This includes:

- Count mean
- · Log fold change
- p-value
- · Adjusted p-value
- · Comparison order

#### **Examples**

RNAdistribution

Plot the distribution of sRNA classes based on nucleotide length

#### **Description**

RNAdistribution plots the distribution of dicer-derived sRNA classes across samples or the sRNA consensus determined by the RNAdicercall() function. This can be displayed as a line or bar plot.

# Usage

```
RNAdistribution(
  data,
  samples = NULL,
  style,
  data.type = "samples",
```

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```
facet = TRUE,
facet.arrange = 3,
colour = "#0868AC",
outline = "black",
wrap.scales = "fixed",
overlap = TRUE,
relative = FALSE
)
```

## **Arguments**

data	data.frame; generated originally by RNAimport()
samples	character; states a subset of samples to plot. This argument is based on the sample names within the data.
style	character; style="line" or style="bar". Instructs how to plot the data. Where style="line" plots a line graph and style="bar" plots a bar graph.
data.type	character; either plotting "samples" or the "consensus" stored in the 'DicerConsensus' column.
facet	logical; forms a matrix of panels defined by row and column faceting variables. It plots the results for each sample as a bar chart and contains it within a single plot. The number of rows in the facet can be changed using the argument facet.arrange. Default facet = TRUE, plots each sample separately when facet = FALSE.
facet.arrange	numeric; value supplied to define the number of columns to include in the facet. This argument is piped into the ncol argument in ggplot2::facet_wrap() to define the number of columns. By default, this is set to 3.
colour	character; fill colour, Default is "#0868AC".
outline	character; states the outline colour for a box plot. Default is "black".
wrap.scales	character; scales be fixed ("fixed", the default), free ("free"), or free in one dimension ("free_x", "free_y")
overlap	logical; generates a single line graph, containing all sample replicate information. Default overlap=TRUE.
relative	logical; used in conjunction with data.type="consensus". Instructs plotting of the relative frequency of all sRNA classes across the data defined by the consensus dicer-derived sRNA column (see RNAdicercall() function for more information).

#### **Details**

The function can be used to plot a variety of different comparisons and plots. It can be used to plot the distribution of sRNA classes within each sample replicate, which can be represented as a bar chart style="bar" or a line graph style="line". These plots can be represented individually or in a single facet plot when facet="TRUE".

Additionally, there is the option to plot the distribution of sRNA classes within individual samples or to plot the distribution of the consensus dicer-derived sRNA classes determined by the RNAdicercall() function and stored in the column DicerConsensus when data.type="consensus". When plotting samples individually, there is the option to overlap the results onto a single line graph when overlap=TRUE. This is not an option for bar plots.

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#### Value

The function returns a list containing the results: a data frame and the plot(s). To access an element, simply use the "\$" symbol, and the elements "data" and "plot" will appear.

#### **Examples**

RNAfeatures

Summarise the distribution of sRNA clusters across genomic features

# Description

Calculates the number of genomic features within the supplied annotations and calculates the number of sRNA clusters which overlap with these genomic features. Features include promoter regions, repeat regions, exons, introns, and untranslated regions. This can be summarised as the absolute or relative values.

#### Usage

```
RNAfeatures(
  data,
  annotation,
  repeats = NULL,
  promoterRegions = 1000,
  percentage = TRUE,
  repeat.type = NULL
)
```

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#### **Arguments**

data data.frame; generated by RNAimport()

annotation path; URL or connection to a GFFFile object. A genome reference annotation

file (.gff/.gff1/.gff2/.gff3). Can be in compressed format (gzip).

repeats path; URL or connection to a GFFFile object. A genome reference annota-

tion file, which only contains information on repeat sequences in the genome (.gff/.gff1/.gff2/.gff3). By default, this is not required, however if there is a specific repeats annotation file for the genome it is suggested to supply it. Can be

in compressed format (gzip).

promoterRegions

numeric; defines the upstream promoter region of genes. Default is 1000, which

refers to promoters set at 1Kb upstream of genes

percentage logical; define whether to return the results as a percentage of the total or re-

turned as a count value representing the number of sRNA clusters that overlap

with a given genomic feature. Default is TRUE.

repeat.type character; features type in annotation file to represent repeats or transposable

elements when repeats not supplied. Default is c("transposable\_element", transposable\_ele

which represent the transposable element features in the TAIR10 genome anno-

tation.

#### **Details**

RNAfeatures calculates the number or percentage of sRNA clusters which overlap with genomic features based on their genomic coordinates.

#### Value

Returns a table containing the number or percentage of overlaps in the supplied sRNA data set with specific regions in the genome annotation such as genes, repeats, introns, exons.

# See Also

RNAmergeAnnotations() to merge 2 GFF files into 1.

## **Examples**

RNAimport

Import and organise sRNAseq & mRNAseq data sets

## **Description**

Load and organise either sRNAseq or mRNAseq pre-processing results into a single dataframe containing all experimental replicates specified where rows represent either a sRNA cluster (ie. sRNA producing-locus) or gene, respectively. Based on using the mobileRNA pre-processing method (See mapRNA()).

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#### **Usage**

```
RNAimport(
  input = c("sRNA", "mRNA"),
  directory,
  samples,
  analysisType = "mobile",
  annotation,
  idattr = "Name",
  FPKM = FALSE,
  featuretype = "mRNA"
)
```

#### **Arguments**

input string; define type of dataset. "sRNA" for sRNAseq data and "mRNA" for mR-

NAseq data.

directory path; directory containing of sample folders generated by ShortStack

samples character; vector naming samples correlating to outputted folders within the

directory path.

analysisType character; either "core" or "mobile" to represent the sRNA analysis workflow.

Where the "core" sRNA analysis imports all reads (unique & multi-map), while "mobile" sRNA analysis imports only uniquely aligned read counts. Only for

sRNA data, default is "mobile".

annotation path; directory to genome annotation (GFF) file used for pre-processing. Only

for mRNA data.

idattr character; GFF attribute to be used as feature ID containing mRNA names. Sev-

eral GFF lines with the same feature ID will be considered as parts of the same feature. The feature ID is used to identity the counts in the output table. Default

is "Name". Only for mRNA data.

FPKM logical; calculate the FPKM for each sample. Default is FALSE.

 $\mbox{ feature type } \mbox{ character; type of feature. Default is "mRNA", only for mRNA data.}$ 

#### **Details**

The RNAimport() function requires the user to supply a directory path and a character vector. The path must be to the pre-processing output.

Following the mobileRNA method, for sRNA analysis, the path will be to the 2\_alignment\_results folder. While for mRNA analysis, the path will be to the 2\_raw\_counts folder. Both folders are generated by the mapRNA() function. The vector should contain strings that represent and mirror the names of the sample replicate folders in the above directory.

Together this information allows the function to extract the information stored in "Result.txt" files of each sample.

#### Value

**For sRNAseq:** A dataframe where rows represent sRNA clusters and columns represent replicate information extracted from the ShortStack output. Replicate information includes Dicercall, Counts, and MajorRNA sequence. Each replicate information is distinguishable as the replicate name is joined as a suffix to each column name. For example, for a sample called "Sample1", the columns will include DicerCall\_Sample1, Count\_Sample1, MajorRNA\_Sample1 and RPM\_Sample1.

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The breakdown of each column:

• Locus : sRNA cluster locus

• chr : Chromosome

• start : start coordinate of cluster

• end : end coordinate of cluster

• Cluster: name of cluster

- DicerCall\_: the size in nucleotides of most abundant sRNA in the cluster
- Count\_: number of uniquely aligned sRNA-seq reads that overlap the locus
- MajorRNA\_: RNA sequence of the most abundant sRNA in the cluster
- RPM\_: reads per million
- FPKM\_: Fragments Per Kilobase of transcript per Million mapped reads (only if option activated)

**For mRNAseq:** A dataframe where rows represent genes and columns represent replicate information extracted from HTseq result. Replicate information includes Counts and FPKM. For example, for a sample called "Sample1", the columns will include Count\_Sample1, and FPKM\_Sample1.

The breakdown of each column:

• mRNA: Name of mRNA

· Locus: Genomic loci of mRNA

· chr : Chromosome

• start : start coordinate

• end : end coordinate

• width: width in nucleotides of regions

- Count\_: number of uniquely aligned mRNA-seq reads that overlap the locus
- FPKM\_: Fragments Per Kilobase of transcript per Million mapped reads

#### References

ShortStack https://github.com/MikeAxtell/ShortStack, HISAT2 https://anaconda.org/bioconda/hisat2, HTSeq https://htseq.readthedocs.io/en/master/install.html, SAM-tools https://anaconda.org/bioconda/samtools

## **Examples**

RNAmergeAnnotations

Merge two genome annotation files (GFF Format)

#### **Description**

Merges two genomes annotation files (GFF) into one single GFF format file saved to the desired directory. This function adds a unique prefix to the chromosome names in each genome annotation to ensure each is distinguishable within the merged file.

## Usage

```
RNAmergeAnnotations(
  annotationA,
  annotationB,
  output_file,
  AnnoA.ID = "A",
  AnnoB.ID = "B",
  format = "gff3"
)
```

# Arguments

```
annotationA path; path to a genome annotation assembly file in GFF format.

annotationB path; path to a genome annotation assembly file in GFF format.

output_file path; a character string or a base::connections() open for writing. Including file output name, and must have a GFF file extension.

AnnoA.ID character; string to represent prefix added to existing chromosome names in annotationA. Default set as "A".

AnnoB.ID character; string to represent prefix added to existing chromosome names in annotationB. Default set as "B".

format format of GFF output, either "gff", "gff1", "gff2", "gff3." Default is "gff3".
```

#### **Details**

As default, the function removes periods and adds a prefix to the existing chromosome names. The prefix is separated from the original chromosome name by an underscore. For example, based on the default settings, it will add the prefix " $A_{-}$ " to the chromosome names in annotationA, for instance,  $A_{-}$ 0,  $A_{-}$ 1,  $A_{-}$ 2 etc.

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The merged genome is saved to the specified output directory, and requires the user to set the name with a GFF format.

**IMPORTANT:** The genome reference and annotation of a species must have chromosomes with matching names. It is critical that if you used the RNAmergeGenomes() function to create a merged reference genome, that you treat the input annotations in the same way.

#### Value

A GFF format file containing the annotations of two genomes distinguishable by the appended prefixes.

# Examples

RNAmergeGenomes

Merge two genome reference assemblies (FASTA format)

#### **Description**

Merges two reference genomes (FASTA) into one single reference with modified chromosome names.

Typically, use genomeA as the origin tissue genome assembly, and genomeB as the genome from which mobile RNAs are produced by.

#### Usage

```
RNAmergeGenomes(
  genomeA,
  genomeB,
  output_file,
  GenomeA.ID = "A",
  GenomeB.ID = "B",
  compress.output = FALSE
)
```

#### **Arguments**

```
genomeA path; path to a genome reference assembly file in FASTA format.

genomeB path; path to a genome reference assembly file in FASTA format.

output_file path; a character string or a base::connections() open for writing. Including file output name, and file extension of .fa or .fasta.
```

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GenomeA.ID	character; string to represent prefix added to existing chromosome names in
	genomeA. Default set as "A"
GenomeB.ID	character; string to represent prefix added to existing chromosome names in genomeB. Default set as "B".
compress.output	
	logical; state whether the output file should be in a compressed format (gzip)

#### Details

The function merges two FASTA files, however, when merging genomic files it is critical that the two genomes are distinguishable by the chromosome names. As a default setting, the function extracts the chromosome names for the given FASTA files and alters adds a unique prefix while retaining the identifying number. Plus, removes any periods.

As default, the function will rename the chromosome names in genome\_A to "A" and separates the prefix and the existing chromosome names with an underscore ("\_"). For example, A\_0, A\_1, A\_2 etc.

Please note that the underscore is added automatically, hence, when setting a custom prefix just includes character values.

**IMPORTANT:** The genome reference and annotation of the same species/accession/variety must have chromosomes with matching names. It is critical that if you use the RNAmergeAnnotations() function to create a merged genome annotation, that you treat the input references in the same way.

#### Value

Returns a single FASTA format file containing both genome assemblies with edited chromosome names (prefixes, and removal of periods) to the given directory.

# **Examples**

```
fasta_1 <- system.file("extdata","reduced_chr12_Eggplant.fa.gz",
package="mobileRNA")

fasta_2 <-system.file("extdata","reduced_chr2_Tomato.fa.gz",
package="mobileRNA")

output_file <- file.path(tempfile("merged_annotation", fileext = ".fa"))

merged_ref <- RNAmergeGenomes(genomeA = fasta_1,
genomeB = fasta_2,
output_file = output_file)</pre>
```

 ${\tt RNAmobile}$ 

Identify putative RNA molecules produced by the non-tissue sample genome

# **Description**

A function to identify the putative sRNA or mRNA molecules produced by the non-tissue sample genome. Includes putative RNA mobilome or RNAs not expected to be found within the tissue of origin.

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#### Usage

```
RNAmobile(
  input = c("sRNA", "mRNA"),
  data,
  controls,
  genome.ID,
  task = NULL,
  statistical = FALSE,
  alpha = 0.1,
  threshold = NULL
)
```

## **Arguments**

input character; must be either "sRNA" or "mRNA" to represent the type of data.

data.frame; generated through the **mobileRNA** method.

controls character vector; containing names of control samples.

genome . ID character; string or chromosome identifier related to the chromosomes in a given

genome. A distinguishing feature of the genome of interest or non-interest in the

chromosome name (chr column).

task character; string to set the method to keep or remove the chromosomes contain-

ing the identifying string. To keep the chromosomes with the ID, set task=keep.

To remove, set task="remove". As default, task is set to keep.

statistical If TRUE, will undertake statistical filtering based on the a p-value or adjusted

p-value threshold stated by alpha. Default set at FALSE. Requires presence of columns containing statistical data. In order to filter by the adjusted p-value, a column named padjusted must be present. See RNAdifferentialAnalysis()

to calculate statistical values.

alpha numeric; adjusted p-value cutoff as the target FDR for independent filtering.

Default is 0.1. Only mobile molecules with adjusted p-values equal or lower

than specified are returned.

threshold numeric; set a threshold level. For sRNA analysis, this represents filtering by

the minimum number of replicates that defined the consensus dicercall which is stored in the DicerCounts column. While, for mRNA analysis this represents the number of replicates which contained reads for the mRNA molecule which

is stored in the SampleCounts column.

## **Details**

The function identifies candidate sRNAs or mRNAs produced by a specific genome/genotype. It does so by either keeping or removing those mapped to a given genome. To do so, it requires a common pre-fix across chromosomes of the given genome. SeeRNAmergeGenomes() for more information. In addition, it removes RNAs which were likely to be falsely mapped. These are those which were mapped to the non-tissue genotype in the control samples.

**For sRNAseq:** A greater confidence in the sRNA candidates can be achieved by setting a threshold that considers the number of replicates which contributed to defining the consensus dicercall (ie. consensus sRNA classification). This parameter filters based on the DicerCounts column introduced by the RNAdicercall() function.

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**For mRNAseq:** A greater confidence in the mRNA candidates can be achieved by setting a threshold that considers the number of replicates which contained reads for the mRNA molecule. This parameter filters based on the SampleCounts column introduced by the RNAimport() function.

**Statistical Analysis** The function also allows for filtering using statistical inference generated from the differential analysis of the total data set using the function RNAdifferentialAnalysis(). When statistical=TRUE, the feature is enabled and selects molecules that meet the adjusted p-value cutoff defined by alpha.

#### Value

A data frame containing candidate mobile sRNAs or mRNAs, which could have been further filtered based on statistical significance and the ability to by-pass the thresholds which determine the number of replicates that defined the consensus dicercall (sRNA) or contributed to reads counts (mRNA).

## **Examples**

```
data("sRNA_data")

# vector of control names
controls <- c("selfgraft_1", "selfgraft_2" , "selfgraft_3")

# Locate potentially mobile sRNA clusters associated to tomato, no
# statistical analysis
mobile_df1 <- RNAmobile(input = "sRNA", data = sRNA_data,
controls = controls, genome.ID = "B_", task = "keep", statistical = FALSE)</pre>
```

RNApopulation

Identify gained/lost RNA populations between treatment and control conditions

## **Description**

Identify unique sRNA or mRNA populations within a set of samples, typically within the same condition, compared to the other samples in the analysis. This can be known as lost or gained populations due to a treatment.

#### Usage

```
RNApopulation(
  data,
  conditions,
  statistical = FALSE,
  alpha = 0.05,
  chimeric = FALSE,
  controls = NULL,
  genome.ID = NULL,
  dual = TRUE
)
```

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#### **Arguments**

data	data.frame; generated by RNAimport()
conditions	character; containing names of samples within conditions to locate unique sRNA clusters.
statistical	logical; If TRUE, will undertake statistical filtering based on the adjusted p-value cutoff stated by alpha. Default setting statistical=FALSE. Requires presence of columns containing statistical data. To filter by the adjusted p-value, a column named padjusted must be present. See RNAdifferentialAnalysis() to calculate statistical values.
alpha	numeric; user-defined numeric value to represent the adjusted p-value threshold to define statistical significance. Defaults setting alpha=0.05. Returns sRNA clusters or mRNA with adjusted p-values equal or lower than the threshold value.
chimeric	logical; state whether system is chimeric containing multiple genomes/genotypes.
controls	character; vector of control condition sample names.
genome.ID	character; chromosome identifier of the genome representing either the origin of mobile molecules or the other genome in the chimeric system.
dual	logical; works in corporation when chimeric=TRUE and removes sRNA clusters mapped to the genome of mobile molecules.

#### **Details**

The function selects RNA which are unique to a given condition and absent in the samples within the other condition(s). For instance, a treatment might encourage the production of unique sRNAs which are not produced in the control samples. The function can also select the unique populations which show statistical significance, based on a adjusted p-value cutoff. This must have been calculated previously, see RNAdifferentialAnalysis() function.

If users are working with a chimeric system, utilise the chimeric=TRUE parameter and state genome. ID and controls parameter variables. This will remove any potential mapping errors which could lead to false interpretation.

#### Value

A subset of the supplied data and prints summary metric of results including:

- the total number of sRNA clusters or mRNA in the data set
- the number & percentage of unique sRNA clusters or mRNA to your condition
- the samples in the condition

#### **Examples**

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RNAreorder	Reorder the data frame for differential analysis, ensuring control verse treatment comparison
	T

## **Description**

Re-organise the working data frame, placing control samples before treatment samples. This ensures differential analysis comparison between controls and treatment are in the correct arrangement.

#### Usage

```
RNAreorder(data, controls)
```

## **Arguments**

data data.frame; generated by RNAimport()

controls character; vector of control condition sample names.

#### Value

A re-ordered/re-organised working data frame with control samples after the 5 cluster information columns, and treatment sample columns after the control sample columns.

#### **Examples**

```
# load data
data("sRNA_data")

controlReps <- c("selfgraft_1", "selfgraft_2", "selfgraft_3")

reorder_df <- RNAreorder(data = sRNA_data, controls = controlReps)</pre>
```

**RNAsequences** 

Extract RNA sequence from sRNA clusters

# Description

RNAs equences extrapolates the RNA sequence for sRNA clusters through two different methods utilising the RNA sequence of the most abundant transcript identified within each replicate. This can either be determined by extracting the consensus sequence across replicates or by comparing the sequences across replicates and selecting the most abundant. In this second method ties between sequences can be seen, hence, the user must decide whether a sequence is then chosen at random from the most abundant or will exclude any sequence determination.

The function also calculates the RNA & DNA complementary sequences, as well as stating the length/width of the sequence.

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#### Usage

```
RNAsequences(
  data,
  original = FALSE,
  method = c("consensus", "set"),
  match.threshold = 1,
  duplicates = "random",
  tidy = FALSE
)
```

## **Arguments**

data data.frame; generated by RNAimport()

original logical; output results to working date as additional columns (original=TRUE),

or output as new data frame (original=FALSE). by default, FALSE

method character; string to define method. Either "consensus" or "set". The "consen-

sus" method identifies the consensus sequences across replicated based on the bioseq package method. Whereas the "set" is based on the fixed sequences

calculated for each replicate and whether they are exact matches or not.

match.threshold

numeric; the minimum number of replicates required to share the sRNA se-

quence to count as a match. Default is 1. Only applicable to the "set" method.

duplicates character; string to define how to deal with a tie, "random" as default. Options

include "random" and "exclude". Only applicable to the "set" method.

tidy logical; tidy-up data set by removing sRNA clusters with a unknown or un-

classified consensus sRNA sequence result. By default, tidy=FALSE, while tidy=TRUE removes sRNA clusters with an undetermined consensus RNA se-

quence.

## **Details**

The set method checks whether each sample in the data set shares the same major sRNA sequence for a given sRNA cluster. If at least two replicates share the same sRNA sequence, the sequence is pulled and the complementary DNA and RNA sequences are calculated. Using the match.threshold parameter, we can alter the minimum number of replicates required to share the RNA sequence to count as a match. For example, if set as match.threshold=3, at least 3 replicates must contain the same sequence. As a general rule, if only one replicate has determined a sRNA sequence it is noted that there is no match, but the sequence is pulled and the complementary sequences calculated.

The match column can either return "Yes", "No" or "Duplicate". If a match between replicates is found, "Yes" is supplied, if not, "No". While if there is a tie between sequences "Duplicate" is supplied. For examples, if an equal number of replicates have sequence "x" and sequence "y".

In the situation where duplicates are identified, as default, at random a consensus sRNA sequence is selected. This parameter can be changed to "exclude", and under this parameter no consensus sequence is pulled.

Whereas with the consensus method, the consensus sequence is pulled from all replicates.

#### Value

The results can be added as additional columns to the working data frame supplied to the function or stored as a new data frame containing only the results from the function. The results includes:

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- Match: whether the RNA sequence is consistent across replicates
- Sequence: character; sequence of the most abundant sRNA within a cluster
- Complementary\_RNA: character; complementary RNA nucleotide sequence
- Complementary\_DNA: character; complementary DNA nucleotide sequence
- Width: numeric; length of nucleotide sequence

#### **Examples**

```
data("sRNA_data")

# vector of control names
controls <- c("selfgraft_1", "selfgraft_2" , "selfgraft_3")

# Locate potentially mobile sRNA clusters associated to tomato, no
# statistical analysis
sRNA_data_mobile <- RNAmobile(input = "sRNA", data = sRNA_data,
controls = controls, genome.ID = "B_", task = "keep", statistical = FALSE)

mobile_sequences <- RNAsequences(sRNA_data_mobile, method = "consensus")</pre>
```

RNAsubset

Subset sRNA data based on dicercall size

## **Description**

Subset the existing dataframe to contain only the desired sRNA class(s) based on the consensus dicercall determination and statistical significance.

# Usage

```
RNAsubset(data, class, statistical = FALSE, alpha = 0.05)
```

#### **Arguments**

data	data.frame; generated by RNAimport(), containing the additional information generated by RNAdicercall() which defines the consensus sRNA class for each sRNA dicer-derived clusters
class	numeric; sRNA dicercall class(es) to select. Based on size in nucleotides.
statistical	logical; filter and select statistically significant sRNA. If statistical=TRUE, sRNA clusters are selected based on p-adjusted alpha threshold.
alpha	numeric; user-defined numeric value to represent the adjusted p-value threshold to define statistical significance. Defaults setting alpha=0.05.

# Details

See RNAdicercall() for information on defining the consensus sRNA dicercall class for each cluster. The function allows the choice to filtered the data by statistical significance based on differential expression analysis, see RNAdifferentialAnalysis(). Set statistical=TRUE to filtered by statistical significance (p-adjusted).

It is important to consider what point in your analysis the data is subset as it will drastically reduce your sample size, altering the output of statistical analyse.

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#### Value

A data frame containing only sRNA clusters defined with a specific sRNA dicer-derived consensus size.

## **Examples**

RNAsummary

Summarise differential analysis results

# Description

Print a summary of the statistical analysis of sRNA clusters (sRNAseq) or mRNAs (mRNAseq) from a mobileRNA analysis

# Usage

```
RNAsummary(
  data,
  alpha = 0.1,
  chimeric = FALSE,
  controls = NULL,
  genome.ID = NULL
)
```

## **Arguments**

data	data.frame; generated by RNAdifferentialAnalysis()
alpha	numeric; user-defined numeric value to represent the adjusted p-value threshold to define statistical significance. Defaults setting alpha=0.1.
chimeric	logical; state whether system is chimeric: contains multiple genomes/genotypes.
controls	character; vector of control condition sample names.
genome.ID	character; chromosome identifier of the genome representing either the origin of mobile molecules or the other genome in the chimeric system.

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#### **Details**

To look only at the differential abundance from RNAs in the mobilome, use the chimeric=TRUE parameter and supply the chromosome identifier of the genome from which mobile molecules originate from to the genome. ID parameter & the control condition samples names to the controls parameter.

#### Value

Prints a summary of the RNAs which align with the adjusted p-value cutoff and states the number which has a positive and negative log-fold change. Where a positive log-fold change represents an increase in abundance and a negative value represents a decrease in abundance between the conditions.

#### **Examples**

sRNA\_data

sRNA\_data: simulated data for biological replicates

## **Description**

Simulated sRNAseq dataset

# Usage

```
data(sRNA_data)
```

#### **Details**

Simulates data is taken from eggplant and tomato sRNAseq samples and created to simulate to movement of sRNA molecules from an Tomato rootstock to an Eggplant Scion. Three Eggplant replicates were spiked with the same 150 tomato sRNA clusters, and named "heterograft\_" 1 to 3. The analysis compares these heterografts to three Eggplant self-grafts which are the original un-spiked Eggplant replicates, called "selfgraft\_" 1 to 3.

This data was imported and organised by the RNAimport() function.

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# Value

Dataframe in global environment

# Examples

data("sRNA\_data")

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