Package 'BRGenomics'

December 29, 2019

```
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2 binNdimensions

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PausingI		vSign		• •																								1	(
_	. 11	, ~-5-	ıal .																									1	4
	naices																											1	4
Stranded	Covera	ge .																										. 1.	4
ort-func	tions																											. 1	
ceGRang	gesBPr	es .																										1	8
geGRan	igesDat	a																										. 1	(
aSubsan	nple .																											2	(
aSubsan	npleMa	ıtrix																										. 2	4
Oseq .																												2	4
Oseq_pa	ired .																											. 2	4
sampleC	Range	s																										2	
setRegic	nsByS	ignal																										2	
Chromo	somes																											2	•
dm6_cl	nr4																											. 2	
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Description

This function takes in data along 1 or more dimensions, and for each dimension the data is divided into evenly-sized bins from the minimum value to the maximum value, and bin numbers are returned. For instance, if each index of the input data were a gene, the input dimensions would be various quantitative measures of that gene, e.g. expression level, number of exons, length, etc. If plotted in cartesian coordinates, each gene would be a single datapoint, and each measurement would be a separate dimension. The bin numbers for each datapoint in each dimension are returned in a dataframe, with a column for each dimension and a row for each index.

Usage

```
binNdimensions(..., nbins = 10)
```

genebodies 3

Arguments

... A single dataframe, or any number of lists or vectors containing different mea-

surements across the same datapoints. If a dataframe is given, columns should correspond to measurements (dimensions). If lists or vectors are given, they must all have the same lengths. Other input classes will be coerced into a single

dataframe.

nbins Either a number giving the number of bins to use for all dimensions (default =

10), or a vector containing the number of bins to use for each dimension of input

data given.

Value

A dataframe containing indices in 1:nbins for each datapoint in each dimension.

Author(s)

Mike DeBerardine

4 genebodies

Description

This function returns ranges that are defined relative to the strand-specific start and end sites of regions of interest (usually genes). Unlike GenomicRanges::promoters, distances can be upstream or downstream based on the sign, and both the start and end of the returned regions can be defined in terms of either the start of end site of the input ranges. For example, genebodies(txs,-50,150, fix.end = "start") is equivalent to promoters(txs,50,150). The default arguments return ranges that begin 300 bases downstream of the original start positions, and end 300 bases upstream of the original end positions.

Usage

```
genebodies(
  genelist,
  start = 300,
  end = -300,
  fix.start = "start",
  fix.end = "end",
  min.window = 0
)
```

Arguments

genelist	A GRanges object containing genes of interest.
start	Depending on fix.start, the distance from either the strand-specific start or end site to begin the returned ranges. If positive, the returned range will begin downstream of the reference position; negative numbers are used to return sites upstream of the reference. Set start = 0 to return the reference position.
end	Identical to the start argument, but defines the strand-specific end position of returned ranges. end must be downstream of start.
fix.start	The reference point to use for defining the strand-specific start positions of returned ranges, either "start" or "end".
fix.end	The reference point to use for defining the strand-specific end positions of returned ranges, either "start" or "end". Cannot be set to "start" if fix.start = "end".
min.window	When fix.start = "start" and fix.end = "end", min.window defines the minimum size (width) of a returned range. However, when fix.end = fix.start, all returned ranges have the same width, and min.window simply size-filters the input ranges.

Value

A GRanges object that may be shorter than genelist due to loss of short ranges.

Author(s)

Mike DeBerardine

See Also

```
intra-range-methods
```

getCountsByPositions 5

Examples

getCountsByPositions Get signal counts at each position within regions of interest

Description

Generate a matrix containing a row for each region of interest, and columns for each position (each base if binsize = 1) within each region.

Usage

```
getCountsByPositions(
  dataset.gr,
  regions.gr,
  binsize = 1,
  FUN = sum,
  simplify.multi.widths = c("list", "pad 0", "pad NA"),
  field = "score",
  ncores = detectCores()
)
```

Arguments

dataset.gr A GRanges object in which signal is contained in metadata (typically in the "score" field).

regions.gr A GRanges object containing all the regions of interest.

6 getCountsByPositions

binsize Size of bins (in bp) to use for counting within each range of regions.gr. Note

that counts will *not* be length-normalized.

FUN If binsize > 1, the function used to aggregate the signal within each bin. By

default, the signal is summed, but any function operating on a numeric vector

can be used.

simplify.multi.widths

A string indicating the output format if the ranges in regions.gr have variable

widths. Default = "list". See details below.

field The metadata field of dataset.gr to be counted. If length(field) > 1, the

output is a list whose elements contain the output for generated each field.

ncores Multiple cores can only be used if length(field) > 1.

Details

If the widths of all ranges in regions.gr are equal, a matrix is returned containing a row for each range in regions.gr, and a column for each bin. For input regions.gr with varying widths, setting simplify.multi.widths = "list" will output a list of variable-length vectors, with each vector corresponding to an input region. If simplify.multi.widths = "pad 0" or "pad NA", the output is a matrix containing a row for each range in regions.gr, and a column for each position in each range. The number of columns is determined by the largest range in regions.gr, and columns corresponding to positions outside of each range are either set to 0 or NA, depending on the argument.

Author(s)

Mike DeBerardine

See Also

getCountsByRegions

getCountsByRegions 7

getCountsByRegions

Get signal counts in regions of interest

Description

Returns a vector the same length as regions.gr containing signal found in each range.

Usage

```
getCountsByRegions(
  dataset.gr,
  regions.gr,
  field = "score",
  ncores = detectCores()
)
```

Arguments

dataset.gr A GRanges object in which signal is contained in metadata (typically in the

"score" field).

regions.gr A GRanges object containing all the regions of interest.

field The metadata field of dataset.gr to be counted. If length(field) > 1, a

dataframe is returned containing the counts for each region in each field.

ncores Multiple cores can only be used if length(field) > 1.

Author(s)

Mike DeBerardine

See Also

```
getCountsByPositions
```

8 getDESeqDataSet

Examples

```
data("PROseq") # load included PROseq data
data("txs_dm6_chr4") # load included transcripts

counts <- getCountsByRegions(PROseq, txs_dm6_chr4)
length(txs_dm6_chr4)
length(counts)
head(counts)

# Assign as metadata to the transcript GRanges
txs_dm6_chr4$PROseq <- counts
txs_dm6_chr4[1:6]</pre>
```

getDESeqDataSet

Get DESeqDataSet objects for downstream analysis

Description

This is a convenience function for generating DESeqDataSet objects, but this function also adds support for counting reads across non-contiguous regions.

Usage

```
getDESeqDataSet(
  dataset.list,
  regions.gr,
  sample_names = names(dataset.list),
  gene_names = NULL,
  sizeFactors = NULL,
  field = "score",
  ncores = detectCores(),
  quiet = FALSE
)
```

Arguments

dataset.list A list of GRanges datasets that can be individually passed to getCountsByRegions.

regions.gr A GRanges object containing regions of interest.

sample_names Names for each dataset in dataset.list are required, and by default the names

of the list elements are used. The names must each contain the string "_rep#", where "#" is a single character (usually a number) indicating the replicate. Sam-

ple names across different replicates must be otherwise identical.

gene_names An optional character vector giving gene names, or any other identifier over

which reads should be counted. Gene names are required if counting is to be performed over non-contiguous ranges, i.e. if any genes have multiple ranges. If supplied, gene names are added to the resulting DESeqDataSet object.

sizeFactors DESeq2 sizeFactors can be optionally applied in to the DESeqDataSet object

in this function, or they can be applied later on, either by the user or in a call to getDESeqResults. Applying the sizeFactors later is useful if multiple sets of factors will be explored, although sizeFactors can be overwritten at any time.

getDESeqDataSet 9

field	Argument passed to	getCountsByRegions.
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ncores Number of cores to use for read counting across all samples. Default is the total

number of cores available.

quiet If TRUE, all output messages from call to DESeqDataSet will be suppressed.

Value

A DESeqData object in which rowData are given as rowRanges, which are equivalent to regions.gr, unless there are non-contiguous gene regions (see note below). Samples (as seen in colData) are factored so that samples are grouped by replicate and condition, i.e. all non-replicate samples are treated as distinct, and the DESeq2 design = ~condition.

Use of non-contiguous gene regions

In DESeq2, genes must be defined by single, contiguous chromosomal locations. This function allows individual genes to be encompassed by multiple distinct ranges in regions.gr. To use non-contiguous gene regions, provide gene_names in which some names are duplicated. For each unique gene in gene_names, this function will generate counts across all ranges for that gene, but be aware that it will only keep the largest range for each gene in the resulting DESeqDataSet object's rowRanges.

A note on DESeq2 sizeFactors

DESeq2 sizeFactors are sample-specific normalization factors that are applied by division, i.e. $counts_{norm,i} = counts_i/sizeFactor_i$. This is in contrast to normalization factors as defined in this package (and commonly elsewhere), which are applied by multiplication. Also note that DESeq2's "normalizationFactors" are not sample specific, but rather gene specific factors used to correct for ascertainment bias across different genes (e.g. as might be relevant for GSEA or Go analysis).

Author(s)

Mike DeBerardine

See Also

```
DESeq2::DESeqDataSet, getDESeqResults
```

```
suppressPackageStartupMessages(require(DESeq2))
data("PROseq") # import included PROseq data
data("txs_dm6_chr4") # import included transcripts

# divide PROseq data into 6 toy datasets
ps_a_rep1 <- PROseq[seq(1, length(PROseq), 6)]
ps_b_rep1 <- PROseq[seq(2, length(PROseq), 6)]
ps_c_rep1 <- PROseq[seq(3, length(PROseq), 6)]

ps_a_rep2 <- PROseq[seq(4, length(PROseq), 6)]
ps_b_rep2 <- PROseq[seq(5, length(PROseq), 6)]
ps_c_rep2 <- PROseq[seq(6, length(PROseq), 6)]
ps_list <- list(A_rep1 = ps_a_rep1,</pre>
```

10 getDESeqResults

getDESeqResults

Get DESeq2 results using reduced dispersion matrices

Description

This function calls DESeq2::DESeq and DESeq2::results on a pre-existing DESeqDataSet object and returns a DESeqResults table for one or more pairwise comparisons. However, unlike a standard call to DESeq2::results using the contrast argument, this function subsets the dataset so that DESeq2 only estimates dispersion for the samples being compared, and not for all samples present.

Usage

```
getDESeqResults(
   dds,
   contrast.numer,
   contrast.denom,
   comparisons.list = NULL,
   sizeFactors = NULL,
   alpha = 0.1,
   args.DESeq = NULL,
   args.results = NULL,
   ncores = detectCores(),
   quiet = FALSE
)
```

Arguments

dds A DESeqDataSet object, produced using either getDESeqDataSet from this

package or DESeqDataSet from DESeq2. If dds was not created using getDESeqDataSet,

dds must be made with design = \sim condition such that a unique condition

level exists for each sample/treatment condition.

contrast.numer A string naming the condition to use as the numerator in the DESeq2 compar-

ison, typically the perturbative condition.

contrast.denom A string naming the condition to use as the denominator in the DESeq2 com-

parison, typically the control condition.

getDESeqResults 11

comparisons.list

As an optional alternative to supplying a single contrast. numer and contrast. denom,

users can supply a list of character vectors containing numerator-denominator

pairs, e.g. list(c("B", "A"), c("C", "A"), c("C", "B")).

sizeFactors A vector containing DESeq2 sizeFactors to apply to each sample. Each sam-

ple's readcounts are divided by its respective DESeq2 sizeFactor. A warning will be generated if the DESeqDataSet already contains sizeFactors, and the

previous sizeFactors will be over-written.

alpha The significance threshold passed to DESeqResults. This won't affect the out-

put results, but is used as a performance optimization by DESeq2.

args.DESeq Additional arguments passed to DESeq, given as a list of argument-value pairs,

e.g. list(test = "LRT", fitType = "local"). All arguments given here will be passed to DESeq except for object and parallel. If no arguments are given,

all defaults will be used.

args.results Additional arguments passed to DESeq2::results, given as a list of argument-

value pairs, e.g. list(altHypothesis = "greater",lfcThreshold = 1.5). All arguments given here will be passed to results except for object, contrast, alpha, and parallel. If no arguments are given, all defaults will be used.

ncores The number of cores to use for parallel processing. Multicore processing is only

used if more than one comparison is being made (i.e. argument comparisons.list is used), and the number of cores utilized will not be greater than the number of

comparisons being performed.

quiet If TRUE, all output messages from calls to DESeq and results will be sup-

pressed, although passing option quiet in args.DESeq will supersede this op-

tion for the call to DESeq.

Value

For a single comparison, the output is the DESeqResults result table. If a comparisons.list is used to make multiple comparisons, the output is a named list of DESeqResults objects, with elements named following the pattern "X_vs_Y", where X is the name of the numerator condition, and Y is the name of the denominator condition.

Author(s)

Mike DeBerardine

See Also

```
getDESeqDataSet, DESeq2::results
```

```
#-----#
# getDESeqDataSet
#-------#
suppressPackageStartupMessages(require(DESeq2))
data("PROseq") # import included PROseq data
data("txs_dm6_chr4") # import included transcripts

# divide PROseq data into 6 toy datasets
ps_a_rep1 <- PROseq[seq(1, length(PROseq), 6)]
ps_b_rep1 <- PROseq[seq(2, length(PROseq), 6)]</pre>
```

```
ps_c_rep1 <- PROseq[seq(3, length(PROseq), 6)]</pre>
ps_a_rep2 <- PROseq[seq(4, length(PROseq), 6)]</pre>
ps_b_rep2 <- PROseq[seq(5, length(PROseq), 6)]</pre>
ps_c_rep2 <- PROseq[seq(6, length(PROseq), 6)]</pre>
ps_list <- list(A_rep1 = ps_a_rep1,</pre>
               A_rep2 = ps_a_rep2,
               B_rep1 = ps_b_rep1,
               B_rep2 = ps_b_rep2,
               C_{rep1} = ps_c_{rep1}
               C_{rep2} = ps_c_{rep2}
# make flawed dataset (ranges in txs_dm6_chr4 not disjoint)
    this means there is double-counting
# also using discontinuous gene regions, as gene_ids are repeated
dds <- getDESeqDataSet(ps_list,</pre>
                      txs_dm6_chr4,
                      gene_names = txs_dm6_chr4$gene_id,
                      ncores = 2)
dds
#-----#
# getDESeqResults
#-----#
res <- getDESeqResults(dds, "B", "A")</pre>
res
reslist <- getDESeqResults(dds,</pre>
                          comparisons.list = list(c("B", "A"), c("C", "A")),
                          ncores = 1)
names(reslist)
reslist[[1]]
```

getMaxPositionsBySignal

Find sites with max signal in regions of interest

Description

For each signal-containing region of interest, find the single site with the most signal. Sites can be found at base-pair resolution, or defined for larger bins.

Usage

```
getMaxPositionsBySignal(
  regions.gr,
  dataset.gr,
  binsize = 1,
  bin.centers = FALSE,
```

```
field = "score",
  keep.score = FALSE
)
```

Arguments

regions.gr A GRanges object containing regions of interest.

dataset.gr A GRanges object in which signal is contained in metadata (typically in the

"score" field).

binsize The size of bin in which to calculate signal scores.

bin.centers Logical indicating if the centers of bins are returned, as opposed to the entire

bin. If TRUE,

field The metadata field of dataset.gr to be counted.

keep. score Logical indicating if the signal value at the max site should be reported. If set to

TRUE, the values are kept as a new metadata column in regions.gr.

Value

Output is a GRanges object with regions.gr metadata, but each range only contains the site within each regions.gr range that had the most signal. If binsize > 1, the entire bin is returned, unless bin.centers = TRUE, in which case a single-base site is returned. The site is set to the center of the bin, and if the binsize is even, the site is rounded to be closer to the beginning of the range.

If keep.score = TRUE, the output will also contain metadata for the signal at the max site. The output is *not* necessarily same length as regions.gr, as regions without signal are not returned. If *no regions* have signal (e.g. as could happen if running this function on a single region), the function will return an empty GRanges object with intact metadata columns.

Author(s)

Mike DeBerardine

See Also

```
getCountsByPositions
```

14 getPausingIndices

```
#-----#
# max sites in 5 bp bins
#-----#
getMaxPositionsBySignal(pr[1:3], PROseq, binsize = 5, keep.score = TRUE)
```

getPausingIndices

Calculate pausing indices from user-supplied promoters & genebodies

Description

Pausing index (PI) is calculated for each gene (within matched promoters.gr and genebodies.gr) as promoter-proximal (or pause region) signal counts divided by genebody signal counts. If length.normalize = TRUE (recommended), the signal counts within each range in promoters.gr and genebodies.gr are divided by their respective range widths (region lengths) before pausing indices are calculated.

Usage

```
getPausingIndices(
  dataset.gr,
  promoters.gr,
  genebodies.gr,
  field = "score",
  length.normalize = TRUE,
  remove.empty = FALSE,
  ncores = detectCores()
)
```

Arguments

dataset.gr A GRanges object in which signal is contained in metadata (typically in the

"score" field).

promoters.gr A GRanges object containing promoter-proximal regions of interest.

genebodies.gr A GRanges object containing genebody regions of interest.

field The metadata field of dataset.gr to be counted. If length(field) > 1, a

dataframe is returned containing the pausing indices for each region in each

field.

length.normalize

A logical indicating if signal counts within regions of interest should be length normalized. The default is TRUE, which is recommended, especially if input

regions don't all have the same width.

remove.empty A logical indicating if genes without any signal in promoters.gr should be

removed. No genes are filtered by default.

ncores Multiple cores can only be used if length(field) > 1.

Value

A vector of length given by the length of the genelist (or possibly shorter if remove.empty = TRUE). If length(field) > 1, a dataframe is returned, containing a column for each field.

getStrandedCoverage 15

Author(s)

Mike DeBerardine

See Also

getCountsByRegions

```
data("PROseg") # load included PROseg data
data("txs_dm6_chr4") # load included transcripts
#-----#
# Get promoter-proximal and genebody regions
# genebodies from +300 to 300 bp before the poly-A site
gb <- genebodies(txs_dm6_chr4, 300, -300, min.window = 400)</pre>
# get the transcripts that are large enough (>1kb in size)
txs <- subset(txs_dm6_chr4, tx_name %in% gb$tx_name)</pre>
# for the same transcripts, promoter-proximal region from 0 to +100
pr <- promoters(txs, 0, 100)</pre>
#-----#
# Calculate pausing indices
#-----#
pidx <- getPausingIndices(PROseq, pr, gb)</pre>
length(txs_dm6_chr4)
length(pidx)
head(pidx)
#-----#
# Without length normalization
head( getPausingIndices(PROseq, pr, gb, length.normalize = FALSE) )
# Removing empty means the values no longer match the genelist
#----#
pidx_signal <- getPausingIndices(PROseq, pr, gb, remove.empty = TRUE)</pre>
length(pidx_signal)
```

16 getStrandedCoverage

Description

Computes strand-specific coverage signal, and returns a GRanges object with signal in the "score" metadata column. Function also works for non-strand-specific data. Note that output is not automatically converted into a "basepair-resolution" GRanges object.

Usage

```
getStrandedCoverage(dataset.gr, field = "score")
```

Arguments

dataset.gr A GRanges object either containing ranges for each read, or one in which read-

counts for individual ranges are contained in metadata (typically in the "score"

field).

field The name of the metadata field that contains readcounts. If no metadata field

contains readcounts, and each range represents a single read, set to NULL.

Author(s)

Mike DeBerardine

See Also

makeGRangesBPres

```
#-----#
# Using included full-read data
#-----#
data("PROseq_paired")
PROseq_paired[1:6]
getStrandedCoverage(PROseq_paired)[1:6]
#-----#
# Re-creating score for included single-base data
data("PROseq")
PR0seq[1:6]
# undo coverage for the first 100 positions
ps <- PROseq[1:100]</pre>
ps_reads <- rep(ps, times = ps$score)</pre>
mcols(ps_reads) <- NULL</pre>
ps_reads[1:6]
# re-create coverage
getStrandedCoverage(ps_reads, field = NULL)[1:6]
```

import-functions 17

import-functions

Import basepair-resolution files

Description

Import basepair-resolution files

Usage

```
import_bigWig(
 plus_file,
 minus_file,
  genome = NULL,
 keep.X = TRUE,
 keep.Y = TRUE,
 keep.M = FALSE,
 keep.nonstandard = FALSE
)
import_bedGraph(
 plus_file,
 minus_file,
 genome = NULL,
  keep.X = TRUE,
 keep.Y = TRUE,
 keep.M = FALSE,
 keep.nonstandard = FALSE
)
```

Arguments

plus_file, minus_file

Paths for strand-specific input files.

genome

Optional string for UCSC reference genome, e.g. "hg38". If given, non-standard chromosomes are trimmed, and options for sex and mitochondrial chromosomes are applied.

keep.X, keep.Y, keep.M, keep.nonstandard

Logicals indicating which non-autosomes should be kept. By default, sex chromosomes are kept, but mitochondrial and non-standard chromosomes are removed.

Details

Imports a GRanges object containing base-pair resolution data, with the score metadata column indicating the number of reads represented by each range.

import_bedGraph is useful for when both 5'- and 3'-end information is to be maintained for each sequenced molecule. It effectively imports the entire read.

For import_bigWig, all ranges are of width = 1.

18 makeGRangesBPres

Author(s)

Mike DeBerardine

See Also

```
tidyChromosomes, rtracklayer::import
```

Examples

```
# get local address for included bigWig files
p.bw <- system.file("extdata", "PROseq_dm6_chr4_plus.bw", package = "BRGenomics")
m.bw <- system.file("extdata", "PROseq_dm6_chr4_minus.bw", package = "BRGenomics")
# import bigWigs
PROseq <- import_bigWig(p.bw, m.bw, genome = "dm6")</pre>
```

makeGRangesBPres

Make base-pair resolution GRanges object

Description

Splits up all ranges in gr to be each 1 basepair wide. All information is preserved, including all metadata. To wit, length(output.gr) = sum(width(dataset.gr)).

Usage

```
makeGRangesBPres(dataset.gr)
```

Arguments

```
dataset.gr A disjoint GRanges object
```

Details

Note that this function doesn't perform any transformation on the metadata in the input; for any ranges of width > 1, the metadata is simply copied to the daughters of that range (whose widths are all equal to 1).

This function is intended to work on datasets at single-base resolution. Data of this type is often formatted as a bigWig file, and any data imported from a bigWig file by rtracklayer is suitable for processing. bigWig files will typically use run-length compression on the data signal (the 'score' column), such that when imported by rtracklayer, adjacent bases sharing the same signal will combined into a single range. The base-pair resolution GRanges objects produced by this function remove this compression, resulting in each index (each range) of the GRanges object addressing a single genomic position.

To properly use base-pair resolution information, the user should be selecting a single-base from each read, which can be accomplished using GenomicRanges::resize(). Then, single-base coverage can be calculated using getStrandedCoverage.

Author(s)

Mike DeBerardine

mergeGRangesData 19

See Also

```
getStrandedCoverage, GenomicRanges::resize()
```

Examples

```
data("PROseq") # load included PROseq data
range(width(PROseq))

# simulate the format of a bigWig file, using arbitrary scores
bw <- reduce(PROseq)
score(bw) <- score(PROseq)[seq_along(bw)]
range(width(bw))
length(bw)

gr <- makeGRangesBPres(bw)
range(width(gr))
length(gr)</pre>
```

mergeGRangesData

Merge base-pair resolution GRanges objects

Description

Merges 2 or more GRanges objects. For each object, the range widths must all be 1, and the score metadata column contains coverage information at each site. This function returns a single GRange object containing all sites of the input objects, and the sum of all scores at all sites.

Usage

```
mergeGRangesData(..., field = "score", ncores = detectCores())
```

Arguments

... Any number of GRanges objects in which signal (e.g. readcounts) are contained

within metadata.

field One or more metadata fields to be combined, typically the "score" field. Fields

typically contain coverage information.

ncores More than one core can be used to coerce non-single-width GRanges objects

using makeGRangesBPres.

Author(s)

Mike DeBerardine

See Also

makeGRangesBPres

20 metaSubsample

Examples

metaSubsample

Iterative Subsampling for Metaplotting

Description

This function performs bootstrap subsampling of mean readcounts at different positions within regions of interest. Mean signal counts can be estimated at base-pair resolution, or smoothed over larger bins.

Usage

```
metaSubsample(
  dataset.gr,
  regions.gr,
  binsize = 1,
  first.output.xval = 1,
  sample.name = deparse(substitute(dataset.gr)),
  n.iter = 1000,
  prop.sample = 0.1,
  lower = 0.125,
  upper = 0.875,
  NF = 1,
  field = "score",
  remove.empty = FALSE,
  ncores = 1
)
```

Arguments

dataset.gr A GRanges object in which signal is contained in metadata (typically in the "score" field).

metaSubsample 21

regions.gr A GRanges object containing intervals over which to metaplot. All ranges must

have the same width.

binsize The size of bin (number of columns, e.g. basepairs) to use for metaplotting.

Especially important for metaplots over large/sparse regions.

first.output.xval

The relative start position of the first bin, e.g. if regions.gr begins at 50 bases upstream of the TSS, set first.output.xval = -50. This number only affects

the x-values that are returned, which are provided as a convenience.

sample.name Defaults to the name of dataset.gr. This is included in the output as a conve-

nience for row-binding outputs from different samples.

n.iter Number of random subsampling iterations to perform. Default is 1000.

prop.sample The proportion of the ranges in regions.gr (e.g. the proportion of genes) to

subsample in each iteration. The default is 0.1 (10 percent).

lower The lower quantile of subsampled signal means to return. The default is 0.125

(12.5th percentile).

upper The upper quantile of subsampled signal means to return. The default is 0.875

(85.5th percentile).

NF Optional normalization factor by which to multiply the counts.

field The metadata field of dataset.gr to be counted.

remove.empty A logical indicating whether regions without signal should be removed from the

analysis.

ncores Number of cores to use for parallel computation. No parallel processing is used

by default, as there's no performance benefit for typical usage with short com-

putation times.

Value

Dataframe containing x-values, means, lower quantiles, upper quantiles, and the sample name (as a convenience for row-binding multiple of these dataframes).

Author(s)

Mike DeBerardine

See Also

metaSubsampleMatrix, getCountsByPositions

metaSubsampleMatrix Iterative Subs

Iterative Subsampling for Metaplotting (On Count Matrices)

Description

In the most general sense, this function performs iterations of randomly subsampling rows of a matrix, and returns a summary of mean values calculated for each column. The typical application is for generating metaplots, with the typical input being a matrix in which each row is a gene or other region of interest, each column is a position within that gene (either a specific basepair or a bin), and element values are signal (e.g. read counts) within those positions.

Usage

```
metaSubsampleMatrix(
   counts.mat,
   binsize = 1,
   first.output.xval = 1,
   sample.name = deparse(substitute(counts.mat)),
   n.iter = 1000,
   prop.sample = 0.1,
   lower = 0.125,
   upper = 0.875,
   NF = 1,
   ncores = 1
)
```

Arguments

counts.mat A matrix of signal counts in which rows are regions of interest and columns are sites/bins in each region.

binsize The size of bin (number of columns, e.g. basepairs) to use for metaplotting.

Especially important for metaplots over large/sparse regions.

first.output.xval

The relative start position of the first bin, e.g. if regions.gr begins at 50 bases upstream of the TSS, set first.output.xval = -50. This number only affects the x-values that are returned, which are provided as a convenience.

sample.name Defaults to the name of dataset.gr.

n.iter Number of random subsampling iterations to perform. Default is 1000.

prop. sample The proportion of rows to subsample in each iteration. The default is 0.1.

lower The lower quantile of subsampled signal means to return. The default is 0.125

(12.5th percentile).

upper The upper quantile of subsampled signal means to return. The default is 0.875

(85.5th percentile).

NF Optional normalization factor by which to multiply the counts.

ncores Number of cores to use for parallel computation. As of writing, parallel pro-

cessing doesn't show any benefit for short computation times (e.g. <1 minute

for our typical experience on a laptop).

Value

Dataframe containing x-values, means, lower quantiles, upper quantiles, and the sample name (as a convenience for row-binding multiple of these dataframes).

Author(s)

Mike DeBerardine

See Also

```
metaSubsample, getCountsByPositions
```

PROseq_paired

PR0seq

PRO-seq data from Drosophila S2 cells

Description

PRO-seq data of Drosophila S2 cells, chromosome 4.

Usage

PR0seq

Format

A disjoint GRanges object with 47533 ranges with 1 metadata column:

score coverage of PRO-seq read 3'-ends ...

Details

Hojoong Kwak, Nicholas J. Fuda, Leighton J. Core, John T. Lis (2013). Precise Maps of RNA Polymerase Reveal How Promoters Direct Initiation and Pausing. Science, 339(6122), 950–953. https://doi.org/10.1126/science.1229386

Source

GEO Accession GSM1032758, run SRR611828.

PROseq_paired

Paired PRO-seq data from Drosophila S2 cells

Description

PRO-seq data of Drosophila S2 cells, chromosome 4. Entire mapped reads kept.

Usage

PROseq_paired

Format

A GRanges object with 52464 ranges with 1 metadata column:

score number of reads sharing the same mapped 5' and 3' ends ...

Details

Hojoong Kwak, Nicholas J. Fuda, Leighton J. Core, John T. Lis (2013). Precise Maps of RNA Polymerase Reveal How Promoters Direct Initiation and Pausing. Science, 339(6122), 950–953. https://doi.org/10.1126/science.1229386

Source

GEO Accession GSM1032758, run SRR611828.

subsampleGRanges 25

subsampleGRanges	Randomly subsample reads from GRanges dataset
------------------	---

Description

Random subsampling is not performed on ranges, but on reads. Readcounts should be given as a metadata field (usually "score"), and should normally be integers. If normalized readcounts are given, an attempt will be made to infer the normalization factor based on the least-common-multiple of the signal found in the specified field. This function can also subsample ranges directly if field = NULL, but the sample function can be used in this scenario.

Usage

```
subsampleGRanges(dataset.gr, n = NULL, prop = NULL, field = "score")
```

Arguments

dataset.gr A GRanges object in which signal (e.g. readcounts) are contained within meta-

data.

n Number of reads to subsample. Either n or prop can be given.

prop Proportion of total signal to subsample.

field The metadata field of dataset.gr that contains readcounts for reach position.

If each range represents a single read, set field = NULL

Author(s)

Mike DeBerardine

Examples

```
data("PROseq") # load included PROseq data
length(PROseq)
sum(score(PROseq))
# sample 10% of the reads
ps_sample <- subsampleGRanges(PROseq, prop = 0.1)
length(ps_sample)
sum(score(ps_sample)) # 1/10th the score is sampled</pre>
```

subsetRegionsBySignal Subset regions of interest by quantiles of overlapping signal

Description

A convenience function to subset regions of interest by the amount of signal they contain, according to their quantile (i.e. their signal ranks).

Usage

```
subsetRegionsBySignal(
  regions.gr,
  dataset.gr,
  quantiles = c(0.5, 1),
  field = "score",
  order.by.rank = FALSE,
  density = FALSE
)
```

Arguments

regions.gr A GRanges object containing regions of interest.

dataset.gr A GRanges object in which signal is contained in metadata (typically in the

"score" field).

quantiles A value pair giving the lower quantile and upper quantile of regions to keep.

Regions with signal quantiles below than the lower quantile are removed, while regions with signal quantiles above the upper quantile are removed. Quantiles must be in range (0,1). An empty GRanges object is returned if lower

quantile = 1 or upper quantile = 0.

field The metadata field of dataset.gr to be counted.

order.by.rank If TRUE, the output regions are sorted based on the amount of signal contained

(in decreasing order). If FALSE (the default), genes are sorted by their positions.

density A logical indicating whether signal counts should be normalized to the width of

ranges in regions.gr. By default, the function only considers the total signal

in each range.

Details

Typical uses may include removing the 5 signal (lower_quantile = 0.05) and the 5 (upper_quantile = 0.95), or returning the middle 50 signal (lower_quantile = 0.25, upper_quantile = 0.75). If lower_quantile = 0 and upper_quantile = 1, all regions are returned, but the returned regions will be sorted by position, or by score if order.by.rank = TRUE.

Value

A GRanges object of length length(regions.gr) * (upper_quantile -lower_quantile).

Author(s)

Mike DeBerardine

See Also

```
getCountsByRegions
```

```
data("PROseq") # load included PROseq data
data("txs_dm6_chr4") # load included transcripts
txs_dm6_chr4
```

tidyChromosomes 27

tidyChromosomes

Remove odd chromosomes from GRanges objects

Description

This convenience function removes non-standard, mitochondrial, and/or sex chromosomes from any GRanges object. For the chromosomes being removed, any ranges found on those chromosomes are removed, and the chromosomes are also removed from seqinfo. Standard chromosomes are defined using the standardChromosomes function from the GenomeInfoDb package.

Usage

```
tidyChromosomes(
  gr,
  keep.X = TRUE,
  keep.Y = TRUE,
  keep.M = FALSE,
  keep.nonstandard = FALSE
)
```

Arguments

gr Any GRanges object, however the object should have a standard genome set, e.g. genome(gr) <-"hg38"

keep.X, keep.Y, keep.M, keep.nonstandard

Logicals indicating which non-autosomes should be kept. By default, sex chromosomes are kept, but mitochondrial and non-standard chromosomes are removed.

Author(s)

Mike DeBerardine

28 txs_dm6_chr4

See Also

GenomeInfoDb::standardChromosomes

Examples

```
data("PROseq") # load included PROseq data
# (only data on chr4, so nothing actually changes)
PROseq_tidy <- tidyChromosomes(PROseq)</pre>
```

txs_dm6_chr4

Ensembl transcripts for Drosophila melanogaster, dm6, chromosome 4.

Description

Transcripts obtained from annotation package TxDb.Dmelanogaster.UCSC.dm6.ensGene, which was in turn made by the Bioconductor Core Team from UCSC resources on 2019-04-25. Metadata columns were obtained from "TXNAME" and "GENEID" columns. Data exported from the TxDb package using GenomicFeatures version 1.35.11 on 2019-12-19.

Usage

```
txs_dm6_chr4
```

Format

A GRanges object with 339 ranges and 2 metadata columns:

tx_name Flybase unique identifiers for transcriptsgene_id FLybase unique identifiers for the associated genes

Source

TxDb.Dmelanogaster.UCSC.dm6.ensGene version 3.4.6

Index

```
* datasets
    PROseq, 24
    PROseq_paired, 24
    txs_dm6_chr4, 28
 GenomeInfoDb::standardChromosomes, 28
binNdimensions, 2
DESeq, 11
DESeq2::DESeq, 10
DESeq2::DESeqDataSet, 9
DESeq2::results, 10, 11
DESeqDataSet, 9, 10
genebodies, 3
GenomicRanges::promoters, 4
GenomicRanges::resize(), 18, 19
getCountsByPositions, 5, 7, 13, 21, 23
getCountsByRegions, 6, 7, 8, 15, 26
getDESeqDataSet, 8, 10, 11
getDESeqResults, 9, 10
getMaxPositionsBySignal, 12
getPausingIndices, 14
getStrandedCoverage, 15, 18, 19
import-functions, 17
import_bedGraph(import-functions), 17
import_bigWig (import-functions), 17
makeGRangesBPres, 16, 18, 19
mergeGRangesData, 19
metaSubsample, 20, 23
metaSubsampleMatrix, 21, 22
PROseq, 24
PROseq_paired, 24
rtracklayer::import, 18
standardChromosomes, 27
subsampleGRanges, 25
subsetRegionsBySignal, 25
tidyChromosomes, 18, 27
txs_dm6_chr4, 28
```