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add_rows

Broadcasting vector arithmetic

Description

Convenience functions for adding or multiplying each row / column of a matrix by a number.

Usage

```
add_rows(mat, vec)
add_cols(mat, vec)
multiply_rows(mat, vec)
multiply_cols(mat, vec)
```

Arguments

mat Matrix-like object vec Numeric vector

Value

Matrix-like object

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all_matrix_inputs

Get/set inputs to a matrix transform

Description

A matrix object can either be an input (i.e. a file on disk or a raw matrix in memory), or it can represent a delayed operation on one or more matrices. The all_matrix_inputs() getter and setter functions allow accessing the base-level input matrices as a list, and changing them. This is useful if you want to re-locate data on disk without losing your transformed BPCells matrix. (Note: experimental API; potentially subject to revisions).

Usage

```
all_matrix_inputs(x)
all_matrix_inputs(x) <- value</pre>
```

Arguments

x IterableMatrix

value List of IterableMatrix objects

Value

List of IterableMatrix objects. If a matrix m is itself an input object, then all_matrix_inputs(m) will return list(m).

binarize

Convert matrix elements to zeros and ones

Description

Binarize compares the matrix element values to the threshold value and sets the output elements to either zero or one. By default, element values greater than the threshold are set to one; otherwise, set to zero. When strict_inequality is set to FALSE, element values greater than or equal to the threshold are set to one. As an alternative, the <, <=, >, and >= operators are also supported.

```
binarize(mat, threshold = 0, strict_inequality = TRUE)
```

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Arguments

mat IterableMatrix

threshold A numeric value that determines whether the elements of x are set to zero or

one.

strict_inequality

A logical value determining whether the comparison to the threshold is >= (strict_inequality=FALSE) or > (strict_inequality=TRUE).

Value

binarized IterableMatrix object

call_peaks_tile

Call peaks from tiles

Description

Calling peaks from a pre-set list of tiles can be much faster than using dedicated peak-calling soft-ware like macs3. The resulting peaks are less precise in terms of exact coordinates, but should be sufficient for most analyses.

Usage

```
call_peaks_tile(
  fragments,
  chromosome_sizes,
  cell_groups = rep.int("all", length(cellNames(fragments))),
  effective_genome_size = NULL,
  peak_width = 200,
  peak_tiling = 3,
  fdr_cutoff = 0.01,
  merge_peaks = c("all", "group", "none")
)
```

Arguments

fragments IterableFragments object chromosome_sizes

Chromosome start and end coordinates given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

See read_ucsc_chrom_sizes().

cell_groups Grouping vector with one entry per cell in fragments, e.g. cluster IDs

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effective_genome_size

(Optional) effective genome size for poisson background rate estimation. See deeptools for values for common genomes. Defaults to sum of chromosome sizes, which overestimates peak significance

peak_width Width of candidate peaks

peak_tiling Number of candidate peaks overlapping each base of genome. E.g. peak_width

= 300 and peak_tiling = 3 results in candidate peaks of 300bp spaced 100bp

apart

fdr_cutoff Adjusted p-value significance cutoff

merge_peaks How to merge significant peaks with merge_peaks_iterative()

• "all" Merge the full set of peaks

• "group" Merge peaks within each group

• "none" Don't perform any merging

Details

Peak calling steps:

1. Estimate the genome-wide expected insertions per tile based on peak_width, effective_genome_size, and per-group read counts

- 2. Tile the genome with nonoverlapping tiles of size peak_width
- 3. For each tile and group, calculate p_value based on a Poisson model
- 4. Compute adjusted p-values using BH method and using the total number of tiles as the number of hypotheses tested.
- 5. Repeat steps 2-4 peak_tiling times, with evenly spaced offsets
- 6. If merge_peaks is "all" or "group": use merge_peaks_iterative() within each group to keep only the most significant of the overlapping candidate peaks
- 7. If merge_peaks is "all", perform a final round of merge_peaks_iterative(), prioritizing each peak by its within-group significance rank

Value

tibble with peak calls and the following columns:

- chr, start, end: genome coordinates
- group: group ID that this peak was identified in
- p_val, q_val: Poission p-value and BH-corrected p-value
- enrichment: Enrichment of counts in this peak compared to a genome-wide background

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checksum

Calculate the MD5 checksum of an IterableMatrix

Description

Calculate the MD5 checksum of an IterableMatrix and return the checksum in hexidecimal format.

Usage

```
checksum(matrix)
```

Arguments

matrix

IterableMatrix object

Details

checksum() converts the non-zero elements of the sparse input matrix to double precision, concatenates each element value with the element row and column index words, and uses these 16-byte blocks along with the matrix dimensions and row and column names to calculate the checksum. The checksum value depends on the storage order so column- and row-order matrices with the same element values give different checksum values. checksum() uses element and index values in little-endian CPU storage order. It converts to little-endian order on big-endian architecture although this has not been tested.

Value

MD5 checksum string in hexidecimal format.

Examples

```
library(Matrix)
library(BPCells)
m1 <- matrix(seq(1,12), nrow=3)
m2 <- as(m1, 'dgCMatrix')
m3 <- as(m2, 'IterableMatrix')
checksum(m3)</pre>
```

cluster_graph_leiden Cluster an adjacency matrix

Description

Cluster an adjacency matrix

Usage

```
cluster_graph_leiden(snn, resolution = 0.001, seed = 12531, ...)
cluster_graph_louvain(snn, resolution = 1, seed = 12531)
cluster_graph_seurat(snn, resolution = 0.8, ...)
```

Arguments

snn Symmetric adjacency matrix (dgCMatrix) output from e.g. knn_to_snn_graph.

Only the lower triangle is used

resolution Resolution parameter. Higher values result in more clusters

seed Random seed for clustering initialization

... Additional arguments to underlying clustering function

Details

```
cluster_graph_leiden: Leiden graph clustering algorithm igraph::cluster_leiden()
cluster_graph_louvain: Louvain graph clustering algorithm igraph::cluster_louvain()
cluster_graph_seurat: Seurat's clustering algorithm Seurat::FindClusters()
```

Value

Factor vector containing the cluster assignment for each cell.

```
cluster_membership_matrix
```

Convert grouping vector to sparse matrix

Description

Converts a vector of membership IDs into a sparse matrix

```
cluster_membership_matrix(groups, group_order = NULL)
```

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Arguments

groups Vector with one entry per cell, specifying the cell's group

group_order Optional vector listing ordering of groups

Value

cell x group matrix where an entry is 1 when a cell is in a given group

collect_features

Collect features for plotting

Description

Helper function for data on features to plot from a diverse set of data sources.

Usage

```
collect_features(
  source,
  features = NULL,
  gene_mapping = human_gene_mapping,
  n = 1
)
```

Arguments

source Matrix or data frame to pull features from, or a vector of feature values for a

single feature. For a matrix, the features must be rows.

features Character vector of features names to plot if source is not a vector.

gene_mapping An optional vector for gene name matching with match_gene_symbol(). Ig-

nored if source is a data frame.

n Internal-use parameter marking the number of nested calls. This is used for

finding the name of the "source" input variable from the caller's perspective

Details

If source is a data.frame, features will be drawn from the columns. If source is a matrix object (IterableMatrix, dgCMatrix, or matrix), features will be drawn from rows.

Value

Data frame with one column for each feature requested

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convert_matrix_type

Convert the type of a matrix

Description

Convert the type of a matrix

Usage

```
convert_matrix_type(matrix, type = c("uint32_t", "double", "float"))
```

Arguments

matrix IterableMatrix object input

type One of uint32_t (unsigned 32-bit integer), float (32-bit real number), or double

(64-bit real number)

Value

IterableMatrix object

Description

BPCells fragments can be interconverted with GRanges and data.frame R objects. The main conversion method is R's builtin as() function, though the convert_to_fragments() helper is also available. For all R objects except GRanges, BPCells assumes a 0-based, end-exclusive coordinate system. (See genomic-ranges-like reference for details)

```
# Convert from R to BPCells
convert_to_fragments(x, zero_based_coords = !is(x, "GRanges"))
as(x, "IterableFragments")

# Convert from BPCells to R
as.data.frame(bpcells_fragments)
as(bpcells_fragments, "data.frame")
as(bpcells_fragments, "GRanges")
```

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Arguments

Χ

Fragment coordinates given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

- chr, start, end: genomic position
- cell_id: cell barcodes or unique identifiers as string or factor

zero_based_coords

Whether to convert the ranges from a 1-based end-inclusive coordinate system to a 0-based end-exclusive coordinate system. Defaults to true for GRanges and false for other formats (see this archived UCSC blogpost)

Value

convert_to_fragments(): IterableFragments object

discrete_palette

Color palettes

Description

These color palettes are derived from the ArchR color palettes, and provide large sets of distinguishable colors

Usage

```
discrete_palette(name, n = 1)
continuous_palette(name)
```

Arguments

name

Name of the color palette. Valid discrete palettes are: stallion, calm, kelly, bear, ironMan, circus, paired, grove, summerNight, and captain. Valid continuous palettes are bluePurpleDark

n

Minimum number of colors needed

Details

If the requested number of colors is too large, a new palette will be constructed via interpolation from the requested palette

Value

Character vector of hex color codes

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extend_ranges

Extend genome ranges in a strand-aware fashion.

Description

Extend genome ranges in a strand-aware fashion.

Usage

```
extend_ranges(
  ranges,
  upstream = 0,
  downstream = 0,
  metadata_cols = c("strand"),
  chromosome_sizes = NULL,
  zero_based_coords = !is(ranges, "GRanges")
)
```

Arguments

ranges

Genomic regions given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

upstream

Number of bases to extend each range upstream (negative to shrink width)

downstream

Number of bases to extend each range downstream (negative to shrink width)

metadata_cols

Optional list of metadata columns to require & extract

chromosome_sizes

(optional) Size of chromosomes as a genomic-ranges object

zero_based_coords

If true, coordinates start and 0 and the end coordinate is not included in the range. If false, coordinates start at 1 and the end coordinate is included in the range

Details

Note that ranges will be blocked from extending past the beginning of the chromosome (base 0), and if chromosome_sizes is given then they will also be blocked from extending past the end of the chromosome

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footprint

Get footprints around a set of genomic coordinates

Description

Get footprints around a set of genomic coordinates

Usage

```
footprint(
  fragments,
  ranges,
  zero_based_coords = !is(ranges, "GRanges"),
  cell_groups = rlang::rep_along(cellNames(fragments), "all"),
  cell_weights = rlang::rep_along(cell_groups, 1),
  flank = 125L,
  normalization_width = flank%/%10L
)
```

Arguments

fragments

IterableFragments object

ranges

Footprint centers given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

- chr, start, end: genomic position
- strand: +/- or TRUE/FALSE for positive or negative strand

"+" strand ranges will footprint around the start coordinate, and "-" strand ranges around the end coordinate.

zero_based_coords

If true, coordinates start and 0 and the end coordinate is not included in the range. If false, coordinates start at 1 and the end coordinate is included in the range

cell_groups Character or factor assigning a group to each cell, in order of cellNames(fragments)

cell_weights Numeric vector assigning weight factors (e.g. inverse of total reads) to each cell,

in order of cellNames(fragments)

flank Number of flanking basepairs to include on either side of the motif

normalization_width

Number of basepairs at the upstream + downstream extremes to use for calculating enrichment

Value

tibble::tibble() with columns group, position, and count, enrichment

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fragments_identical

Check if two fragments objects are identical

Description

Check if two fragments objects are identical

Usage

```
fragments_identical(fragments1, fragments2)
```

Arguments

fragments1 First IterableFragments to compare fragments2 Second IterableFragments to compare

Value

boolean for whether the fragments objects are identical

gene_region

Find gene region

Description

Conveniently look up the region of a gene by gene symbol. The value returned by this function can be used as the region argument for trackplot functions such as trackplot_coverage() or trackplot_gene()

Usage

```
gene_region(
  genes,
  gene_symbol,
  extend_bp = c(10000, 10000),
  gene_mapping = human_gene_mapping
)
```

Arguments

genes

Transcipt features given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

- chr, start, end: genomic position
- $\bullet\,$ strand: +/- or TRUE/FALSE for positive or negative strand
- gene_name: Symbol or gene ID

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gene_symbol Name of gene symbol or ID

extend_bp Bases to extend region upstream and downstream of gene. If length 1, exten-

sion is symmetric. If length 2, provide upstream extension then downstream

extension as positive distances.

gene_mapping Named vector where names are gene symbols or IDs and values are canonical

gene symbols

Value

List of chr, start, end positions for use with trackplot functions.

```
gene_score_tiles_archr
```

Calculate gene-tile distances for ArchR gene activities

Description

ArchR-style gene activity scores are based on a weighted sum of each tile according to the signed distance from the tile to a gene body. This function calculates the signed distances according to ArchR's default parameters.

Usage

```
gene_score_tiles_archr(
  genes,
  chromosome_sizes = NULL,
  tile_width = 500,
  addArchRBug = FALSE
)
```

Arguments

genes Gene coordinates given as GRanges, data.frame, or list. See help("genomic-ranges-like")

for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

• strand: +/- or TRUE/FALSE for positive or negative strand

chromosome_sizes

(optional) Size of chromosomes as a genomic-ranges object

tile_width Size of tiles to consider

addArchRBug Replicate ArchR bug in handling nested genes

Details

ArchR's tile distance algorithm works as follows

- 1. Genes are extended 5kb upstream
- 2. Genes are linked to any tiles 1kb-100kb upstream + downstream, but tiles beyond a neighboring gene are not considered

Value

Tibble with one range per tile, with additional metadata columns gene_idx (row index of the gene this tile corresponds to) and distance.

Distance is a signed distance calculated such that if the tile has a smaller start coordinate than the gene and the gene is on the + strand, distance will be negative. The distance of adjacent but non-overlapping regions is 1bp, counting up from there.

```
{\it gene\_score\_weights\_archr} \\ {\it Calculate~Gene Activity Scores}
```

Description

Gene activity scores can be calculated as a distance-weighted sum of per-tile accessibility. The tile weights for each gene can be represented as a sparse matrix of dimension genes x tiles. If we multiply this weight matrix by a corresponding tile matrix (tiles x cells), then we can get a gene activity score matrix of genes x cells. gene_score_weights_archr() calculates the weight matrix (best if you have a pre-computed tile matrix), while gene_score_archr() provides a easy-to-use wrapper.

```
gene_score_weights_archr(
 genes,
  chromosome_sizes,
 blacklist = NULL,
  tile_width = 500,
  gene_name_column = "gene_id",
  addArchRBug = FALSE
)
gene_score_archr(
  fragments,
  genes,
  chromosome_sizes,
 blacklist = NULL,
  tile_width = 500,
  gene_name_column = "gene_id",
  addArchRBug = FALSE,
  tile_max_count = 4,
  scale_factor = 10000,
  tile_matrix_path = tempfile(pattern = "gene_score_tile_mat")
)
```

Arguments

genes

Gene coordinates given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

- chr, start, end: genomic position
- strand: +/- or TRUE/FALSE for positive or negative strand

chromosome_sizes

Chromosome start and end coordinates given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

See read_ucsc_chrom_sizes().

blacklist

Regions to exclude from calculations, given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

tile_width Size of tiles to consider

gene_name_column

If not NULL, a column name of genes to use as row names

addArchRBug Replicate ArchR bug in handling nested genes

fragments Input fragments object

tile_max_count Maximum value in the tile counts matrix. If not null, tile counts higher than

this will be clipped to tile_max_count. Equivalent to ceiling argument of

ArchR::addGeneScoreMatrix()

scale_factor If not null, counts for each cell will be scaled to sum to scale_factor. Equiv-

alent to scaleTo argument of ArchR::addGeneScoreMatrix()

tile_matrix_path

Path of a directory where the intermediate tile matrix will be saved

Details

gene_score_weights_archr:

Given a set of tile coordinates and distances returned by gene_score_tiles_archr(), calculate a weight matrix of dimensions genes x tiles. This matrix can be multiplied with a tile matrix to obtain ArchR-compatible gene activity scores.

Value

gene_score_weights_archr

Weight matrix of dimension genes x tiles

gene_score_archr

Gene score matrix of dimension genes x cells.

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genomic-ranges-like Genomic range formats

Description

BPCells accepts a flexible set of genomic ranges-like objects as input, either GRanges, data.frame, lists, or character vectors. These objects must specify chromosome, start, and end coordinates along with optional metadata about each range. With the exception of GenomicRanges::GRanges objects, BPCells assumes all objects use a zero-based, end-exclusive coordinate system (see below for details).

Valid Range-like objects:

BPCells can interpret the following types as ranges:

- list(), data.frame(), with columns:
 - chr: Character or factor of chromosome names
 - start: Start coordinates (0-based)
 - end: End coordinates (exclusive)
 - (optional) strand: "+"/"-" or TRUE/FALSE for pos/neg strand
 - (optional) Additional metadata as named list entries or data. frame columns
- GenomicRanges::GRanges
 - start(x) is interpreted as a 1-based start coordinate
 - end(x) is interpreted as an inclusive end coordinate
 - strand(x): "*" entries are interpeted as postive strand
 - (optional) mcols(x) holds additional metadata
- character
 - Given in format "chr1:1000-2000" or "chr1:1,000-2,000"
 - Uses 0-based, end-exclusive coordinate system
 - Cannot be used for ranges where additional metadata is required

Range coordinate systems:

There are two main conventions for the coordinate systems:

One-based, end-inclusive ranges

- The first base of a chromosome is numbered 1
- The last base in a range is equal to the end coordinate
- e.g. 1-5 describes the first 5 bases of the chromosome
- · Used in formats such as SAM, GTF
- In BPCells, used when reading or writing GenomicRanges::GRanges objects

Zero-based, end-exclusive ranges

- The first base of a chromosome is numbered 0
- The last base in a range is one less than the end coordinate
- e.g. 0-5 describes the first 5 bases of the chromosome
- Used in formats such as BAM, BED
- In BPCells, used for all other range objects

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human_gene_mapping

Gene Symbol Mapping data

Description

Mapping of the canonical gene symbols corresponding to each unambiguous alias, previous symbol, ensembl ID, or entrez ID.

Usage

```
human_gene_mapping
mouse_gene_mapping
```

Format

human_gene_mapping

A named character vector. Names are aliases or IDs and values are the corresponding canonical gene symbol

mouse_gene_mapping

A named character vector. Names are aliases or IDs and values are the corresponding canonical gene symbol

Details

See the source code in data-raw/human_gene_mapping.R and data-raw/mouse_gene_mapping.R for exactly how these mappings were made.

Source

human_gene_mapping

```
http://ftp.ebi.ac.uk/pub/databases/genenames/hgnc/tsv/non_alt_loci_set.txt
```

mouse_gene_mapping

http://www.informatics.jax.org/downloads/reports/MGI_EntrezGene.rpthttp://www.informatics.jax.org/downloads/reports/MRK_ENSEMBL.rpt

```
import_matrix_market Import MatrixMarket files
```

Description

Read a sparse matrix from a MatrixMarket file. This is a text-based format used by 10x, Parse, and others to store sparse matrices. Format details on the NIST website.

Usage

```
import_matrix_market(
 mtx_path,
 outdir = tempfile("matrix_market"),
  row_names = NULL,
  col_names = NULL,
  row_major = FALSE,
  tmpdir = tempdir(),
 load_bytes = 4194304L,
  sort_bytes = 1073741824L
)
import_matrix_market_10x(
 mtx_dir,
 outdir = tempfile("matrix_market"),
  feature_type = NULL,
  row_major = FALSE,
  tmpdir = tempdir(),
 load_bytes = 4194304L,
  sort_bytes = 1073741824L
)
```

Arguments

mtx_path

	e
outdir	Directory to store the output
row_names	Character vector of row names
col_names	Character vector of col names
row_major	If true, store the matrix in row-major orientation
tmpdir	Temporary directory to use for intermediate storage
load_bytes	The minimum contiguous load size during the merge sort passes
sort_bytes	The amount of memory to allocate for re-sorting chunks of entries
mtx_dir	Directory holding matrix.mtx.gz, barcodes.tsv.gz, and features.tsv.gz
feature_type	String or vector of feature types to include. (cellranger 3.0 and newer)

Path of mtx or mtx.gz file

Details

Import MatrixMarket mtx files to the BPCells format. This implementation ensures fixed memory usage even for very large inputs by doing on-disk sorts. It will be much slower than hdf5 inputs, so only use MatrixMarket format when absolutely necessary.

As a rough speed estimate, importing the 17GB Parse 1M PBMC DGE_1M_PBMC.mtx file takes about 4 minutes and 1.3GB of RAM, producing a compressed output matrix of 1.5GB. mtx.gz files will be slower to import due to gzip decompression.

When importing from 10x mtx files, the row and column names can be read automatically using the import_matrix_market_10x() convenience function.

Value

MatrixDir object with the imported matrix

IterableFragments-methods

IterableFragments methods

Description

Methods for IterableFragments objects

Usage

```
## $4 method for signature 'IterableFragments'
show(object)

cellNames(x)

cellNames(x, ...) <- value

chrNames(x)

chrNames(x, ...) <- value</pre>
```

Arguments

```
object IterableFragments object

x an IterableFragments object

value Character vector of new names
```

Details

- cellNames<- It is only possible to replace names, not add new names.
- chrNames<- It is only possible to replace names, not add new names.

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Value

- cellNames() Character vector of cell names, or NULL if none are known
- chrNames(): Character vector of chromosome names, or NULL if none are known

Functions

- show(IterableFragments): Print IterableFragments
- cellNames(): Get cell names
- cellNames(x, ...) <- value: Set cell names
- chrNames(): Set chromosome names
- chrNames(x, ...) <- value: Set chromosome names

IterableMatrix-methods

IterableMatrix methods

Description

Generic methods and built-in functions for IterableMatrix objects

```
matrix_type(x)

storage_order(x)

## S4 method for signature 'IterableMatrix'
show(object)

## S4 method for signature 'IterableMatrix'
t(x)

## S4 method for signature 'IterableMatrix,matrix'
x %*% y

## S4 method for signature 'IterableMatrix'
rowSums(x)

## S4 method for signature 'IterableMatrix'
colSums(x)

## S4 method for signature 'IterableMatrix'
rowMeans(x)

## S4 method for signature 'IterableMatrix'
rowMeans(x)
```

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```
colMeans(x)
## S4 method for signature 'IterableMatrix'
log1p(x)
log1p_slow(x)
## S4 method for signature 'IterableMatrix'
expm1(x)
expm1_slow(x)
## S4 method for signature 'IterableMatrix,numeric'
e1 ^ e2
pow_slow(x, exponent)
## S4 method for signature 'numeric, IterableMatrix'
e1 < e2
## S4 method for signature 'IterableMatrix,numeric'
## S4 method for signature 'numeric, IterableMatrix'
## S4 method for signature 'IterableMatrix, numeric'
e1 >= e2
## S4 method for signature 'IterableMatrix'
round(x, digits = 0)
## S4 method for signature 'IterableMatrix, numeric'
e1 * e2
## S4 method for signature 'IterableMatrix,numeric'
## S4 method for signature 'IterableMatrix, numeric'
e1 / e2
## S4 method for signature 'IterableMatrix,numeric'
e1 - e2
```

Arguments

x IterableMatrix objectobject IterableMatrix object

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y matrix

Value

- t() Transposed object
- x %*% y: dense matrix result
- rowSums(): vector of row sums
- colSums(): vector of col sums
- rowMeans(): vector of row means
- colMeans(): vector of col means

Functions

- matrix_type(): Get the matrix data type (mat_uint32_t, mat_float, or mat_double for now)
- storage_order(): Get the matrix storage order ("row" or "col")
- show(IterableMatrix): Display an IterableMatrix
- t(IterableMatrix): Transpose an IterableMatrix
- x %*% y: Multiply by a dense matrix
- rowSums(IterableMatrix): Calculate rowSums
- colSums(IterableMatrix): Calculate colSums
- rowMeans(IterableMatrix): Calculate rowMeans
- colMeans(IterableMatrix): Calculate colMeans
- log1p(IterableMatrix): Calculate log(x + 1)
- $log1p_slow()$: Calculate log(x + 1) (non-SIMD version)
- expm1(IterableMatrix): Calculate exp(x) 1
- expm1_slow(): Calculate exp(x) 1 (non-SIMD version)
- e1^e2: Calculate x^y (elementwise)
- pow_slow(): Calculate x^y (elementwise, non-SIMD version)
- e1 < e2: Binarize matrix according to numeric < matrix comparison
- e1 > e2: Binarize matrix according to matrix > numeric comparison
- e1 <= e2: Binarize matrix according to numeric <= matrix comparison
- e1 >= e2: Binarize matrix according to matrix >= numeric comparison
 round(IterableMatrix): round to nearest integer (digits must be 0)
- e1 * e2: Multiply by a constant, or multiply rows by a vector length nrow(mat)
- e1 + e2: Add a constant, or row-wise addition with a vector length nrow(mat)
- e1 / e2: Divide by a constant, or divide rows by a vector length nrow(mat)
- e1 e2: Subtract a constant, or row-wise subtraction with a vector length nrow(mat)

knn_hnsw 25

knn_hnsw

Get a knn matrix from reduced dimensions

Description

Search for approximate nearest neighbors between cells in the reduced dimensions (e.g. PCA), and return the k nearest neighbors (knn) for each cell. Optionally, we can find neighbors between two separate sets of cells by utilizing both data and query.

Usage

```
knn_hnsw(
 data,
  query = NULL,
  k = 10,
 metric = c("euclidean", "cosine"),
  verbose = TRUE,
  threads = 1,
  ef = 100
)
knn_annoy(
  data,
  query = data,
 k = 10,
 metric = c("euclidean", "cosine", "manhattan", "hamming"),
 n_{trees} = 50,
  search_k = -1
)
```

Arguments

data	cell x dims matrix for reference dataset
query	cell x dims matrix for query dataset (optional)
k	number of neighbors to calculate
metric	distance metric to use
verbose	whether to print progress information during search
threads	Number of threads to use. Note that result is non-deterministic if threads > 1
ef	ef parameter for RccppHNSW::hnsw_search. Increase for slower search but improved accuracy $% \left(\frac{1}{2}\right) =\frac{1}{2}\left(\frac{1}{2}\right) =\frac{1}{2$
n_trees	Number of trees during index build time. More trees gives higher accuracy
search_k	Number of nodes to inspect during the query, or -1 for default value. Higher number gives higher accuracy

26 knn_to_graph

Details

knn_hnsw: Use RcppHNSW as knn engineknn_annoy: Use RcppAnnoy as knn engine

Value

List of 2 matrices – idx for cell x K neighbor indices, dist for cell x K neighbor distances. If no query is given, nearest neighbors are found mapping the data matrix to itself, prohibiting self-neighbors

knn_to_graph

K Nearest Neighbor (KNN) Graph

Description

Convert a KNN object (e.g. returned by knn_hnsw() or knn_annoy()) into a graph. The graph is represented as a sparse adjacency matrix.

Usage

```
knn_to_graph(knn, use_weights = FALSE, self_loops = TRUE)
knn_to_snn_graph(
  knn,
  min_val = 1/15,
  self_loops = FALSE,
  return_type = c("matrix", "list")
)
knn_to_geodesic_graph(knn, return_type = c("matrix", "list"), threads = 0L)
```

Arguments

knn	List of 2 matrices – idx for cell x K neighbor indices, dist for cell x K neighbor distances
use_weights	boolean for whether to replace all distance weights with 1
self_loops	Whether to allow self-loops in the output graph
min_val	minimum jaccard index between neighbors. Values below this will round to 0
return_type	Whether to return a sparse adjacency matrix or an edge list
threads	Number of threads to use during calculations

marker_features 27

Details

knn_to_graph Create a knn graph

knn_to_snn_graph Convert a knn object into a shared nearest neighbors adjacency matrix. This follows the algorithm that Seurat uses to compute SNN graphs

knn_to_geodesic_graph Convert a knn object into an undirected weighted graph, using the same geodesic distance estimation method as the UMAP package. This matches the output of umap._umap.fuzzy_simplicial_se from the umap-learn python package, used by default in scanpy.pp.neighbors. Because this only re-weights and symmetrizes the KNN graph, it will usually use less memory and return a sparser graph than knn_to_snn_graph which computes 2nd-order neighbors. Note: when cells don't have themselves listed as the nearest neighbor, results may differ slightly from umap._umap.fuzzy_simplicial_set, which assumes self is always successfully found in the approximate nearest neighbor search.

Value

knn_to_graph Sparse matrix (dgCMatrix) where mat[i,j] = distance from cell i to cell j, or 0 if cell j is not in the K nearest neighbors of i

knn_to_snn_graph

- return_type == "matrix": Sparse matrix (dgCMatrix) where mat[i,j] = jaccard index of the overlap in nearest neigbors between cell i and cell j, or 0 if the jaccard index is < min_val. Only the lower triangle is filled in, which is compatible with the BPCells clustering methods
- return_type == "list": List of 3 equal-length vectors i, j, and weight, along with an integer dim. These correspond to the rows, cols, and values of non-zero entries in the lower triangle adjacency matrix. dim is the total number of vertices (cells) in the graph

knn_to_geodesic_graph

- return_type == "matrix": Sparse matrix (dgCMatrix) where mat[i,j] = normalized similarity between cell i and cell j. Only the lower triangle is filled in, which is compatible with the BPCells clustering methods
- return_type == "list": List of 3 equal-length vectors i, j, and weight, along with an integer dim. These correspond to the rows, cols, and values of non-zero entries in the lower triangle adjacency matrix. dim is the total number of vertices (cells) in the graph

marker_features

Test for marker features

Description

Given a features x cells matrix, perform one-vs-all differential tests to find markers.

```
marker_features(mat, groups, method = "wilcoxon")
```

28 match_gene_symbol

Arguments

mat	IterableMatrix object of dimensions features x cells
groups	Character/factor vector of cell groups/clusters. Length #cells
method	Test method to use. Current options are:

• wilcoxon: Wilconxon rank-sum test a.k.a Mann-Whitney U test

Details

Tips for using the values from this function:

- Use dplyr::mutate() to add columns for e.g. adjusted p-value and log fold change.
- Use dplyr::filter() to get only differential genes above some given threshold
- To get adjusted p-values, use R p.adjust(), recommended method is "BH"
- To get log2 fold change: if your input matrix was already log-transformed, calculate (foreground_mean background_mean)/log(2). If your input matrix was not log-transformed, calculate log2(forground_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean/background_mean

Value

tibble with the following columns:

- foreground: Group ID used for the foreground
- background: Group ID used for the background (or NA if comparing to rest of cells)
- feature: ID of the feature
- p_val_raw: Unadjusted p-value for differential test
- foreground_mean: Average value in the foreground group
- background_mean: Average value in the background group

match_gene_symbol Gene symbol matching

Description

Correct alias gene symbols, Ensembl IDs, and Entrez IDs to canonical gene symbols. This is useful for matching gene names between different datasets which might not always use the same gene naming conventions.

```
match_gene_symbol(query, subject, gene_mapping = human_gene_mapping)
canonical_gene_symbol(query, gene_mapping = human_gene_mapping)
```

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Arguments

query Character vector of gene symbols or IDs subject Vector of gene symbols or IDs to index into

gene_mapping Named vector where names are gene symbols or IDs and values are canonical

gene symbols

Value

match_gene_symbol

Integer vector of indices v such that subject[v] corresponds to the gene symbols in query

canonical_gene_symbol

Character vector of canonical gene symbols for each symbol in query

matrix_R_conversion

Convert between BPCells matrix and R objects.

Description

BPCells matrices can be interconverted with Matrix package dgCMatrix sparse matrices, as well as base R dense matrices (though this may result in high memory usage for large matrices)

Usage

```
# Convert to R from BPCells
as(bpcells_mat, "dgCMatrix") # Sparse matrix conversion
as.matrix(bpcells_mat) # Dense matrix conversion
# Convert to BPCells from R
as(dgc_mat, "IterableMatrix")
```

matrix_stats

Calculate matrix stats

Description

Calculate matrix stats

```
matrix_stats(
  matrix,
  row_stats = c("none", "nonzero", "mean", "variance"),
  col_stats = c("none", "nonzero", "mean", "variance"),
  threads = 0L
)
```

30 merge_cells

Arguments

matrix Input matrix object

row_stats Which row statistics to compute col_stats Which col statistics to compute

threads Number of threads to use during execution

Details

The statistics will be calculated in a single pass over the matrix, so this method is desirable to use for efficiency purposes compared to the more standard rowMeans or colMeans if multiple statistics are needed. The stats are ordered by complexity: nonzero, mean, then variance. All less complex stats are calculated in the process of calculating a more complicated stat. So to calculate mean and variance simultaneously, just ask for variance, which will compute mean and nonzero counts as a side-effect

Value

List of row_stats: matrix of n_stats x n_rows, col_stats: matrix of n_stats x n_cols

Description

Peak and tile matrix calculations can be sped up by reducing the number of cells. For cases where the outputs are going to be added together afterwards, this can provide a performance improvement

Usage

```
merge_cells(fragments, cell_groups)
```

Arguments

fragments Input fragments object

cell_groups Character or factor vector providing a group for each cell. Ordering is the same

as cellNames(fragments)

merge_peaks_iterative

Description

Merge peaks according to ArchR's iterative merging algorithm. More details on the ArchR website

Usage

```
merge_peaks_iterative(peaks)
```

Arguments

peaks

Peaks given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

Must be ordered by priority and have columns chr, start, end.

Details

Properties of merged peaks:

- No peaks in the merged set overlap
- Peaks are prioritized according to their order in the original input
- The output peaks are a subset of the input peaks, with no peak boundaries changed

Value

tibble::tibble() with a nonoverlapping subset of the rows in peaks. All metadata columns are preserved

min_scalar

Elementwise minimum

Description

min_scalar: Take minumum with a global constant

min_by_row: Take the minimum with a per-row constantmin_by_col: Take the minimum with a per-col constant

32 normalize_ranges

Usage

```
min_scalar(mat, val)
min_by_row(mat, vals)
min_by_col(mat, vals)
```

Arguments

mat IterableMatrix

val Single positive numeric value

Details

Take the minimum value of a matrix with a per-row, per-col, or global constant. This constant must be >0 to preserve sparsity of the matrix. This has the effect of capping the maximum value in the matrix.

Value

IterableMatrix

normalize_ranges

Normalize an object representing genomic ranges

Description

Normalize an object representing genomic ranges

Usage

```
normalize_ranges(
  ranges,
  metadata_cols = character(0),
  zero_based_coords = !is(ranges, "GRanges"),
  n = 1
)
```

Arguments

ranges

Genomic regions given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

metadata_cols Optional list of metadata columns to require & extract zero_based_coords

If true, coordinates start and 0 and the end coordinate is not included in the range. If false, coordinates start at 1 and the end coordinate is included in the range

nucleosome_counts 33

Value

data frame with zero-based coordinates, and elements chr (factor), start (int), and end (int). If ranges does not have chr level information, chr levels are the sorted unique values of chr.

If strand is in metadata_cols, then the output strand element will be TRUE for positive strand, and FALSE for negative strand. (Converted from a character vector of "+"/"-" if necessary)

nucleosome_counts

Count fragments by nucleosomal size

Description

Count fragments by nucleosomal size

Usage

```
nucleosome_counts(fragments, nucleosome_width = 147)
```

Arguments

```
fragments Fragments object nucleosome_width
```

Integer cutoff to use as nucleosome width

Details

Shorter than nucleosome_width is subNucleosomal, nucleosome_width to 2*nucleosome_width-1 is monoNucleosomal, and anything longer is multiNucleosomal. The sum of all fragments is given as nFrags

Value

List with names subNucleosomal, monoNucleosomal, multiNucleosomal, and nFrags, containing the count vectors of fragments in each class per cell.

open_fragments_10x

Read/write a 10x fragments file

Description

10x fragment files come in a bed-like format, with columns chr, start, end, cell_id, and pcr_duplicates. Unlike a standard bed format, the format from cellranger has an *inclusive* end-coordinate, meaning the end coordinate itself is what should be counted as the tagmentation site, rather than offset by 1.

Usage

```
open_fragments_10x(path, comment = "#", end_inclusive = TRUE)
write_fragments_10x(
  fragments,
  path,
  end_inclusive = TRUE,
  append_5th_column = FALSE
)
```

Arguments

path File path (e.g. fragments.tsv or fragments.tsv.gz)

comment Skip lines at beginning of file which start with comment string

end_inclusive Whether the end coordinate of the bed is inclusive – i.e. there was an insertion

at the end coordinate rather than the base before the end coordinate. This is the

10x default, though it's not quite standard for the bed file format.

fragments Input fragments object

append_5th_column

Whether to include 5th column of all 0 for compatibility with 10x fragment file

outputs (defaults to 4 columns chr,start,end,cell)

Details

open_fragments_10x

No disk operations will take place until the fragments are used in a function

write_fragments_10x

Fragments will be written to disk immediately, then returned in a readable object.

Value

10x fragments file object

open_matrix_10x_hdf5 Read/write a 10x feature matrix

Description

Read/write a 10x feature matrix

Usage

```
open_matrix_10x_hdf5(path, feature_type = NULL, buffer_size = 16384L)
write_matrix_10x_hdf5(
    mat,
    path,
    barcodes = colnames(mat),
    feature_ids = rownames(mat),
    feature_names = rownames(mat),
    feature_types = "Gene Expression",
    feature_metadata = list(),
    buffer_size = 16384L,
    chunk_size = 1024L,
    gzip_level = 0L,
    type = c("uint32_t", "double", "float", "auto")
)
```

Arguments

path Path to the hdf5 file on disk

feature_type Optional selection of feature types to include in output matrix. For multiome

data, the options are "Gene Expression" and "Peaks". This option is only com-

patible with files from cellranger 3.0 and newer.

buffer_size For performance tuning only. The number of items to be buffered in memory

before calling writes to disk.

mat IterableMatrix

barcodes Vector of names for the cells
feature_ids Vector of IDs for the features
feature_names Vector of names for the features
feature_types String or vector of feature types

feature_metadata

Named list of additional metadata vectors to store for each feature

chunk_size For performance tuning only. The chunk size used for the HDF5 array storage.

gzip_level Gzip compression level. Default is 0 (no compression)

type Data type of the output matrix. Default is uint32_t to match a matrix of 10x

UMI counts. Non-integer data types include float and double. If auto, will

use the data type of mat.

Details

The 10x format makes use of gzip compression for the matrix data, which can slow down read performance. Consider writing into another format if the read performance is important to you.

Input matrices must be in column-major storage order, and if the rownames and colnames are not set, names must be provided for the relevant metadata parameters. Some of the metadata parameters are not read by default in BPCells, but it is possible to export them for use with other tools.

Value

BPCells matrix object

Description

Read or write a sparse matrix from an anndata hdf5 file. These functions will automatically transpose matrices when converting to/from the AnnData format. This is because the AnnData convention stores cells as rows, whereas the R convention stores cells as columns. If this behavior is undesired, call t() manually on the matrix inputs and outputs of these functions.

Usage

```
open_matrix_anndata_hdf5(path, group = "X", buffer_size = 16384L)
write_matrix_anndata_hdf5(
  mat,
  path,
  group = "X",
  buffer_size = 16384L,
  chunk_size = 1024L,
  gzip_level = 0L
)
```

Arguments

path	Path to the hdf5 file on disk
group	The group within the hdf5 file to write the data to. If writing to an existing hdf5 file this group must not already be in use
buffer_size	For performance tuning only. The number of items to be buffered in memory before calling writes to disk.
chunk_size	For performance tuning only. The chunk size used for the HDF5 array storage.
gzip_level	Gzip compression level. Default is 0 (no compression)

Value

AnnDataMatrixH5 object, with cells as the columns.

order_ranges 37

order_ranges	Get end-sorted ordering for genome ranges

Description

Use this function to order regions prior to calling peak_matrix() or tile_matrix().

Usage

```
order_ranges(ranges, chr_levels, sort_by_end = TRUE)
```

Arguments

ranges
Genomic regions given as GRanges, data.frame, or list. See help("genomic-ranges-like")
for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

chr_levels
Ordering of chromosome names

sort_by_end
If TRUE (defualt), sort by (chr, end, start). Else sort by (chr, start, end)

Value

Numeric vector analogous to the order function. Provides an index selection that will reorder the input ranges to be sorted by chr, end, start

peak_matrix Calculate ranges x cells overlap matrix	peak_matrix	Calculate ranges x cells overlap matrix	
-----------------------------------------------------	-------------	-----------------------------------------	--

Description

Calculate ranges x cells overlap matrix

Usage

```
peak_matrix(
  fragments,
  ranges,
  mode = c("insertions", "fragments", "overlaps"),
  zero_based_coords = !is(ranges, "GRanges"),
  explicit_peak_names = TRUE
)
```

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Arguments

fragments Input fragments object. Must have cell names and chromosome names defined

ranges Peaks/ranges to overlap, given as GRanges, data.frame, or list. See help("genomic-ranges-like")

for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

mode Mode for counting peak overlaps. (See "value" section for more details)

zero_based_coords

Whether to convert the ranges from a 1-based end-inclusive coordinate system to a 0-based end-exclusive coordinate system. Defaults to true for GRanges and false for other formats (see this archived UCSC blogpost)

explicit_peak_names

Boolean for whether to add rownames to the output matrix in format e.g chr1:500-1000, where start and end coords are given in a 0-based coordinate system. Note that either way, peak names will be written when the matrix is saved.

Value

Iterable matrix object with dimension ranges x cells. When saved, the column names of the output matrix will be in the format chr1:500-1000, where start and end coords are given in a 0-based coordinate system.

mode options

- "insertions": Start and end coordinates are separately overlapped with each peak
- "fragments": Like "insertions", but each fragment can contribute at most 1 count to each peak, even if both the start and end coordinates overlap
- "overlaps": Like "fragments", but an overlap is also counted if the fragment fully spans the peak even if neither the start or end falls within the peak

Note

When calculating the matrix directly from a fragments tsv, it's necessary to first call select_chromosomes() in order to provide the ordering of chromosomes to expect while reading the tsv.

plot_dot

Description

Plot feature levels per group or cluster as a grid of dots. Dots are colored by z-score normalized average expression, and sized by percent non-zero.

plot_embedding 39

Usage

```
plot_dot(
    source,
    features,
    groups,
    group_order = NULL,
    gene_mapping = human_gene_mapping,
    colors = c("lightgrey", "#4682B4"),
    return_data = FALSE,
    apply_styling = TRUE
)
```

Arguments

source Feature x cell matrix or data.frame with features. For best results, features

should be sparse and log-normalized (e.g. run log1p() so zero raw counts map

to zero)

features Character vector of features to plot

groups Vector with one entry per cell, specifying the cell's group

group_order Optional vector listing ordering of groups

gene_mapping An optional vector for gene name matching with match_gene_symbol().

colors Color scale for plot

return_data If true, return data from just before plotting rather than a plot.

plot_embedding

Plot UMAP or embeddings

Description

Plot one or more features by coloring cells in a UMAP plot.

Usage

```
plot_embedding(
  source,
  embedding,
  features = NULL,
  quantile_range = c(0.01, 0.99),
  randomize_order = TRUE,
  smooth = NULL,
  smooth_rounds = 3,
  gene_mapping = human_gene_mapping,
  size = NULL,
```

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```
rasterize = FALSE,
raster_pixels = 512,
legend_continuous = c("auto", "quantile", "value"),
labels_quantile_range = TRUE,
colors_continuous = c("lightgrey", "#4682B4"),
legend_discrete = TRUE,
labels_discrete = TRUE,
colors_discrete = discrete_palette("stallion"),
return_data = FALSE,
return_plot_list = FALSE,
apply_styling = TRUE
)
```

Arguments

source Matrix, or data frame to pull features from, or a vector of feature values for a

single feature. For a matrix, the features must be rows.

embedding A matrix of dimensions cells x 2 with embedding coordinates features Character vector of features to plot if source is not a vector.

quantile_range (optional) Length 2 vector giving the quantiles to clip the minimum and maxi-

mum color scale values, as fractions between 0 and 1. NULL or NA values to

skip clipping

randomize_order

If TRUE, shuffle cells to prevent overplotting biases. Can pass an integer instead

to specify a random seed to use.

smooth (optional) Sparse matrix of dimensions cells x cells with cell-cell distance weights

for smoothing.

smooth_rounds Number of multiplication rounds to apply when smoothing.

gene_mapping An optional vector for gene name matching with match_gene_symbol(). Ig-

nored if source is a data frame.

size Point size for plotting

rasterize Whether to rasterize the point drawing to speed up display in graphics programs.

raster_pixels Number of pixels to use when rasterizing. Can provide one number for square

dimensions, or two numbers for width x height.

legend_continuous

Whether to label continuous features by quantile or value. "auto" labels by quantile only when all features are continuous and quantile_range is not NULL. Quantile labeling adds text annotation listing the range of displayed values.

labels_quantile_range

Whether to add a text label with the value range of each feature when the legend is set to quantile

colors_continuous

Vector of colors to use for continuous color palette

legend_discrete

Whether to show the legend for discrete (categorical) features.

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```
labels_discrete
```

Whether to add text labels at the center of each group for discrete (categorical) features.

colors_discrete

Vector of colors to use for discrete (categorical) features.

return_data If true, return data from just before plotting rather than a plot. return_plot_list

If TRUE, return multiple plots as a list, rather than a single plot combined using

patchwork::wrap_plots()

apply_styling If false, return a plot without pretty styling applied

Details

Smoothing:

Smoothing is performed as follows: first, the smoothing matrix is normalized so the sum of incoming weights to every cell is 1. Then, the raw data values are repeatedly multiplied by the smoothing matrix and re-scaled so the average value stays the same.

Value

By default, returns a ggplot2 object with all the requested features plotted in a grid. If return_data or return_plot_list is called, the return value will match that argument.

```
plot_fragment_length Fragment size distribution
```

Description

Plot the distribution of fragment lengths, with length in basepairs on the x-axis, and proportion of fragments on the y-axis. Typical plots will show 10-basepair periodicity, as well as humps spaced at multiples of a nucleosome width (about 150bp).

Usage

```
plot_fragment_length(
   fragments,
   max_length = 500,
   return_data = FALSE,
   apply_styling = TRUE
)
```

Arguments

fragments Fragments object

max_length Maximum length to show on the plot

return_data If true, return data from just before plotting rather than a plot.

apply_styling If false, return a plot without pretty styling applied

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Value

Numeric vector where index i contans the number of length-i fragments

```
plot_read_count_knee Knee plot of single cell read counts
```

Description

Plots read count rank vs. number of reads on a log-log scale.

Usage

```
plot_read_count_knee(
  read_counts,
  cutoff = NULL,
  return_data = FALSE,
  apply_styling = TRUE
)
```

Arguments

read_counts Vector of read counts per cell

cutoff (optional) Read cutoff to mark on the plot

return_data If true, return data from just before plotting rather than a plot.

Details

Performs logarithmic downsampling to reduce the number of points plotted

Value

```
ggplot2 plot object
```

plot_tf_footprint 43

Description

Plot the footprinting around TF motif sites

Usage

```
plot_tf_footprint(
   fragments,
   motif_positions,
   cell_groups = rlang::rep_along(cellNames(fragments), "all"),
   flank = 250L,
   smooth = 0L,
   zero_based_coords = !is(genes, "GRanges"),
   colors = discrete_palette("stallion"),
   return_data = FALSE,
   apply_styling = TRUE
)
```

Arguments

fragments IterableFragments object motif_positions Coordinate ranges for motifs (must include strand) and have constant width Character or factor assigning a group to each cell, in order of cellNames(fragments) cell_groups flank Number of flanking basepairs to include on either side of the motif smooth (optional) Sparse matrix of dimensions cells x cells with cell-cell distance weights for smoothing. zero_based_coords If true, coordinates start and 0 and the end coordinate is not included in the range. If false, coordinates start at 1 and the end coordinate is included in the range return_data If true, return data from just before plotting rather than a plot. If false, return a plot without pretty styling applied apply_styling

See Also

```
footprint(), plot_tss_profile()
```

44 plot_tss_profile

Description

Plot the enrichment of insertions relative to transcription start sites (TSS). Typically, this plot shows strong enrichment of insertions near a TSS, and a small bump downstream around 220bp downstream of the TSS for the +1 nucleosome.

Usage

```
plot_tss_profile(
   fragments,
   genes,
   cell_groups = rlang::rep_along(cellNames(fragments), "all"),
   flank = 2000L,
   smooth = 0L,
   zero_based_coords = !is(genes, "GRanges"),
   colors = discrete_palette("stallion"),
   return_data = FALSE,
   apply_styling = TRUE
)
```

Arguments

IterableFragments object fragments Coordinate ranges for genes (must include strand) genes cell_groups Character or factor assigning a group to each cell, in order of cellNames(fragments) Number of flanking basepairs to include on either side of the motif flank Number of bases to smooth over (rolling average) smooth zero_based_coords If true, coordinates start and 0 and the end coordinate is not included in the range. If false, coordinates start at 1 and the end coordinate is included in the range If true, return data from just before plotting rather than a plot. return_data If false, return a plot without pretty styling applied apply_styling

See Also

```
footprint(), plot_tf_footprint()
```

plot_tss_scatter 45

plot_tss_scatter

TSS Enrichment vs. Fragment Counts plot

Description

Density scatter plot with log10(fragment_count) on the x-axis and TSS enrichment on the y-axis. This plot is most useful to select which cell barcodes in an experiment correspond to high-quality cells

Usage

```
plot_tss_scatter(
   atac_qc,
   min_frags = NULL,
   min_tss = NULL,
   bins = 100,
   apply_styling = TRUE
)
```

Arguments

atac_qc Tibble as returned by qc_scATAC(). Must have columns nFrags and TSSEnrichment

min_fragsMinimum fragment count cutoffmin_tssMinimum TSS Enrichment cutoffbinsNumber of bins for density calculation

prefix_cell_names

Add sample prefix to cell names

Description

Rename cells by adding a prefix to the names. Most commonly this will be a sample name. All cells will recieve the exact text of prefix added to the beginning, so any separator characters like "_" must be included in the given prefix. Use this prior to merging fragments from different experiments with c() in order to help prevent cell name clashes.

Usage

```
prefix_cell_names(fragments, prefix)
```

Arguments

fragments Input fragments object.

prefix String to add as the prefix

46 qc_scATAC

Value

Fragments object with prefixed names

qc_scATAC

Calculate ArchR-compatible per-cell QC statistics

Description

Calculate ArchR-compatible per-cell QC statistics

Usage

```
qc_scATAC(fragments, genes, blacklist)
```

Arguments

fragments IterableFragments object

genes Gene coordinates given as GRanges, data.frame, or list. See help("genomic-ranges-like")

for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

blacklist Blacklisted regions given as GRanges, data.frame, or list. See help("genomic-ranges-like")

for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

Details

This implementation mimics ArchR's default parameters. For uses requiring more flexibility to tweak default parameters, the best option is to re-implement this function with required changes. Output columns of data.frame:

- cellName: cell name for each cell
- nFrags: number of fragments per cell
- subNucleosomal, monoNucleosomal, multiNucleosomal: number of fragments of size 1-146bp, 147-254bp, and 255bp + respectively. equivalent to ArchR's nMonoFrags, nDiFrags, nMultiFrags respectively
- TSSEnrichment: AvgInsertInTSS / max(AvgInsertFlankingTSS, 0.1), where AvgInsertInTSS is ReadsInTSS / 101 (window size), and AvgInsertFlankingTSS is ReadsFlankingTSS / (100*2) (window size). The max(0.1) ensures that very low-read cells do not get assigned spuriously high TSSEnrichment.
- ReadsInPromoter: Number of reads from 2000bp upstream of TSS to 101bp downstream of TSS
- ReadsInBlacklist: Number of reads in the provided blacklist region
- ReadsInTSS: Number of reads overlapping the 101bp centered around each TSS
- ReadsFlankingTSS: Number of reads overlapping 1901-2000bp +/- each TSS

Differences from ArchR: Note that ArchR by default uses a different set of annotations to derive TSS sites and promoter sites. This function uses just one annotation for gene start+end sites, so must be called twice to exactly re-calculate the ArchR QC stats.

ArchR's PromoterRatio and BlacklistRatio are not included in the output, as they can be easily calculated from ReadsInPromoter / nFrags and ReadsInBlacklist / nFrags. Similarly, ArchR's NucleosomeRatio can be calculated as (monoNucleosomal + multiNucleosomal) / subNucleosomal.

Value

data.frame with QC data

range_distance_to_nearest

Find signed distance to nearest genomic ranges

Description

Given a set of genomic ranges, find the distance to the nearest neighbors both upstream and downstream.

Usage

```
range_distance_to_nearest(
  ranges,
  addArchRBug = FALSE,
  zero_based_coords = !is(ranges, "GRanges")
)
```

Arguments

ranges

Genomic regions given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

- chr, start, end: genomic position
- strand: +/- or TRUE/FALSE for positive or negative strand

addArchRBug

boolean to reproduce ArchR's bug that incorrectly handles nested genes

zero_based_coords

If true, coordinates start and 0 and the end coordinate is not included in the range. If false, coordinates start at 1 and the end coordinate is included in the range

Value

A 2-column data frame with columns upstream and downstream, containing the distances to the nearest neighbor in the respective directions. For ranges on + or * strand, distance is calculated as:

```
• upstream = max(start(range) - end(upstreamNeighbor), 0)
```

48 read_bed

• downstream = max(start(downstreamNeighbor) - end(range), 0)

For ranges on - strand, the definition of upstream and downstream is flipped. Note that this definition of distance is one off from GenomicRanges::distance(), as ranges that neighbor but don't overlap are given a distance of 1 rather than 0.

read_bed

Read a bed file into a data frame

Description

Bed files can contain peak or blacklist annotations. These utilities help read thos annotations

Usage

```
read_bed(
  path,
  additional_columns = character(0),
  backup_url = NULL,
  timeout = 300
)

read_encode_blacklist(
  dir,
  genome = c("hg38", "mm10", "hg19", "dm6", "dm3", "ce11", "ce10"),
  timeout = 300
)
```

Arguments

path Path to file (or desired save location if backup_url is used) additional_columns

Names for additional columns in the bed file

backup_url If path does not exist, provides a URL to download the gtf from timeout Maximum time in seconds to wait for download from backup_url

dir Output directory to cache the downloaded gtf file

genome genome name

Details

read_bed

Read a bed file from disk or a url.

read_encode_blacklist

Downloads the Boyle Lab blacklist, as described in https://doi.org/10.1038/s41598-019-45839-z

read_gtf 49

Value

Data frame with coordinates using the 0-based convention.

See Also

```
read_gtf(), read_gencode_genes()
```

read_gtf

Read GTF gene annotations

Description

Read gene annotations from gtf format into a data frame. The source can be a URL, a gtf file on disk, or a gencode release version.

Usage

```
read_gtf(
 path,
  attributes = c("gene_id"),
  tags = character(0),
  features = c("gene"),
  keep_attribute_column = FALSE,
 backup_url = NULL,
  timeout = 300
)
read_gencode_genes(
  dir,
  release = "latest",
 annotation_set = c("basic", "comprehensive"),
  gene_type = "lncRNA|protein_coding|IG_.*_gene|TR_.*_gene",
  attributes = c("gene_id", "gene_type", "gene_name"),
  tags = character(0),
  features = c("gene"),
  timeout = 300
)
read_gencode_transcripts(
  dir,
  release = "latest",
  transcript_choice = c("MANE_Select", "Ensembl_Canonical", "all"),
  annotation_set = c("basic", "comprehensive"),
  gene_type = "lncRNA|protein_coding|IG_.*_gene|TR_.*_gene",
 attributes = c("gene_id", "gene_type", "gene_name", "transcript_id"),
features = c("transcript", "exon"),
  timeout = 300
)
```

50 read_gtf

Arguments

path Path to file (or desired save location if backup_url is used)
attributes Vector of GTF attribute names to parse out as columns
tags Vector of tags to parse out as boolean presence/absence

features List of features types to keep from the GTF (e.g. gene, transcript, exon, intron)

keep_attribute_column

Boolean for whether to preserve the raw attribute text column

backup_url If path does not exist, provides a URL to download the gtf from timeout Maximum time in seconds to wait for download from backup_url

dir Output directory to cache the downloaded gtf file

release release version (prefix with M for mouse versions). For most recent version, use

"latest" or "latest_mouse"

annotation_set Either "basic" or "comprehensive" annotation sets (see details section).

gene_type Regular expression with which gene types to keep. Defaults to protein_coding,

lncRNA, and IG/TR genes

transcript_choice

Method for selecting representative transcripts. Choices are:

- MANE_Select: human-only, most conservative
- Ensembl_Canonical: human+mouse, superset of MANE_Select for human
- all: Preserve all transcript models (not recommended for plotting)

Details

read_gtf

Read gtf from a file or URL

read_gencode_genes

Read gene annotations directly from GENCODE. The file name will vary depending on the release and annotation set requested, but will be of the format gencode.v42.annotation.gtf.gz. GENCODE currently recommends the basic set: https://www.gencodegenes.org/human/. In release 42, both the comprehensive and basic sets had identical gene-level annotations, but the comprehensive set had additional transcript variants annotated.

read_gencode_transcripts

Read transcript models from GENCODE, for use with trackplot_gene()

Value

Data frame with coordinates using the 0-based convention. Columns are:

- chr
- source
- feature
- start

- end
- score
- strand
- frame
- attributes (optional; named according to listed attributes)
- tags (named according to listed tags)

See Also

```
read_bed(), read_encode_blacklist()
```

```
read_ucsc_chrom_sizes Read UCSC chromosome sizes
```

Description

Read chromosome sizes from UCSC and return as a tibble with one row per chromosome. The underlying data is pulled from here: https://hgdownload.soe.ucsc.edu/downloads.html

Usage

```
read_ucsc_chrom_sizes(
   dir,
   genome = c("hg38", "mm39", "mm10", "mm9", "hg19"),
   keep_chromosomes = "chr[0-9]+|chrX|chrY",
   timeout = 300
)
```

rotate_x_labels

Rotate ggplot x axis labels

Description

Rotate ggplot x axis labels

Usage

```
rotate_x_labels(degrees = 45)
```

Arguments

degrees

Number of degrees to rotate by

52 sctransform_pearson

Description

Calculate pearson residuals of a negative binomial sctransform model. Normalized values are calculated as (X - mu) / sqrt(mu + mu^2/theta). mu is calculated as cell_read_counts * gene_beta.

Usage

```
sctransform_pearson(
  mat,
  gene_theta,
  gene_beta,
  cell_read_counts,
  min_var = -Inf,
  clip_range = c(-10, 10),
  columns_are_cells = TRUE,
  slow = FALSE
)
```

Arguments

Details

slow

The parameterization used is somewhat simplified compared to the original SCTransform paper, in particular it uses a linear-scale rather than log-scale to represent the cell_read_counts and gene_beta variables. It also does not support the addition of arbitrary cell metadata (e.g. batch) to add to the negative binomial regression.

If TRUE, use a 10x slower but more precise implementation (default FALSE)

Value

IterableMatrix

select_cells 53

select_cells

Subset, translate, or reorder cell IDs

Description

Subset, translate, or reorder cell IDs

Usage

```
select_cells(fragments, cell_selection)
```

Arguments

fragments Input fragments object

 ${\tt cell_selection} \ \ List \ of \ chromosme \ IDs \ (numeric), \ or \ names \ (character). \ \ The \ output \ cell \ ID \ n$

will be taken from the input cell with ID/name cell_selection[n].

select_chromosomes

Subset, translate, or reorder chromosome IDs

Description

Subset, translate, or reorder chromosome IDs

Usage

```
select_chromosomes(fragments, chromosome_selection)
```

Arguments

fragments Input fragments object

chromosome_selection

List of chromosme IDs (numeric), or names (character). The output chromosome ID n will be taken from the input fragments chromosome with ID/name chromosome_selection[n].

54 set_trackplot_label

select_regions

Subset fragments by genomic region

Description

Fragments can be subset based on overlapping (or not overlapping) a set of regions

Usage

```
select_regions(
  fragments,
  ranges,
  invert_selection = FALSE,
  zero_based_coords = !is(ranges, "GRanges")
```

Arguments

fragments

Input fragments object.

ranges

Peaks/ranges to overlap, given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

invert_selection

If TRUE, select fragments *not* overlapping selected regions instead of only fragments overlapping the selected regions.

zero_based_coords

Whether to convert the ranges from a 1-based end-inclusive coordinate system to a 0-based end-exclusive coordinate system. Defaults to true for GRanges and false for other formats (see this archived UCSC blogpost)

Value

Fragments object filtered according to the selected regions

Description

Adjust labels and heights on trackplots. Labels are set as facet labels in ggplot2, and heights are additional properties read by trackplot_combine() to determine relative height of input plots.

shift_fragments 55

Usage

```
set_trackplot_label(plot, labels)
set_trackplot_height(plot, height)
get_trackplot_height(plot)
```

Arguments

plot ggplot object

labels character vector of labels – must match existing number of facets in plot

height New height. If numeric, adjusts relative height. If ggplot2::unit or grid::unit

sets absolute height in specified units. "null" units are interpreted as relative

height.

Value

set_trackplot_label: ggplot object with adjusted facet labels
set_trackplot_height: ggplot object with adjusted trackplot height
get_trackplot_height: ggplot2::unit object with height setting

shift_fragments

Shift start or end coordinates

Description

Shifts start or end of fragments by a fixed amount, which can be useful to correct the Tn5 offset.

Usage

```
shift_fragments(fragments, shift_start = 0L, shift_end = 0L)
```

Arguments

fragments Input fragments object

shift_start How many basepairs to shift the start coords shift_end How many basepairs to shift the end coords

Details

The correct Tn5 offset is +/- 4bp since the Tn5 cut sites on opposite strands are offset by 9bp. However, +4/-5 bp is often applied to bed-format files, since the end coordinate in bed files is 1 past the last basepair of the sequenced DNA fragment. This results in a bed-like format except with inclusive end coordinates.

56 svds

Value

Shifted fragments object

subset_lengths

Subset fragments by length

Description

Subset fragments by length

Usage

```
subset_lengths(fragments, min_len = 0L, max_len = NA_integer_)
```

Arguments

fragments Input fragments object

min_len Minimum bases in fragment (inclusive)
max_len Maximum bases in fragment (inclusive)

Details

Fragment length is calculated as end-start

Value

Fragments object

svds

Calculate svds

Description

Use the C++ Spectra solver (same as RSpectra package), in order to compute the largest k values and corresponding singular vectors. Empirically, memory usage is much lower than using irlba::irlba(), likely due to avoiding R garbage creation while solving due to the pure-C++ solver. This documentation is a slightly-edited version of the RSpectra::svds() documentation.

Usage

```
svds(A, k, nu = k, nv = k, opts = list(), threads=0L, ...)
```

svds 57

Arguments

A The matrix whose truncated SVD is to be computed.

k Number of singular values requested.

nu Number of right singular vectors to be computed. This must be between 0 and

'k'. (Must be equal to 'k' for BPCells IterableMatrix)

opts Control parameters related to computing algorithm. See *Details* below threads Control threads to use calculating mat-vec producs (BPCells specific)

Details

When RSpectra is installed, this function will just add a method to RSpectra::svds() for the IterableMatrix class.

The opts argument is a list that can supply any of the following parameters:

ncv Number of Lanzcos basis vectors to use. More vectors will result in faster convergence, but with greater memory use. ncv must be satisfy $k < ncv \le p$ where p = min(m, n). Default is min(p, max(2*k+1, 20)).

tol Precision parameter. Default is 1e-10.

maxitr Maximum number of iterations. Default is 1000.

center Either a logical value (TRUE/FALSE), or a numeric vector of length n. If a vector c is supplied, then SVD is computed on the matrix A-1c', in an implicit way without actually forming this matrix. center = TRUE has the same effect as center = colMeans(A). Default is FALSE. Ignored in BPCells

scale Either a logical value (TRUE/FALSE), or a numeric vector of length n. If a vector s is supplied, then SVD is computed on the matrix (A-1c')S, where c is the centering vector and S=diag(1/s). If scale = TRUE, then the vector s is computed as the column norm of A-1c'. Default is FALSE. Ignored in BPCells

Value

A list with the following components:

d A vector of the computed singular values.

u An m by nu matrix whose columns contain the left singular vectors. If nu == 0,

NULL will be returned.

An n by nv matrix whose columns contain the right singular vectors. If nv == 0,

NULL will be returned.

nconv Number of converged singular values.

niter Number of iterations used.

nops Number of matrix-vector multiplications used.

References

Qiu Y, Mei J (2022). RSpectra: Solvers for Large-Scale Eigenvalue and SVD Problems. R package version 0.16-1, https://CRAN.R-project.org/package=RSpectra.

58 tile_matrix

tile_matrix

Calculate ranges x cells tile overlap matrix

Description

Calculate ranges x cells tile overlap matrix

Usage

```
tile_matrix(
  fragments,
  ranges,
  zero_based_coords = !is(ranges, "GRanges"),
  explicit_tile_names = FALSE
)
```

Arguments

fragments

Input fragments object

ranges

Tiled regions given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

- chr, start, end: genomic position
- tile_width: Size of each tile in this region in basepairs

Must be non-overlapping and sorted by (chr, start), with chromosomes ordered according to the chromosome names of fragments

zero_based_coords

Whether to convert the ranges from a 1-based end-inclusive coordinate system to a 0-based end-exclusive coordinate system. Defaults to true for GRanges and false for other formats (see this archived UCSC blogpost)

explicit_tile_names

Boolean for whether to add rownames to the output matrix in format e.g chr1:500-1000, where start and end coords are given in a 0-based coordinate system. For whole-genome Tile matrices the names will take ~5 seconds to generate and take up 400MB of memory. Note that either way, tile names will be written when the matrix is saved.

Value

Iterable matrix object with dimension ranges x cells. When saved, the column names will be in the format chr1:500-1000, where start and end coords are given in a 0-based coordinate system.

Note

When calculating the matrix directly from a fragments tsv, it's necessary to first call select_chromosomes() in order to provide the ordering of chromosomes to expect while reading the tsv.

trackplot_combine 59

trackplot_combine

Combine track plots

Description

Combines multiple track plots of the same region into a single grid. Uses the patchwork package to perform the alignment.

Usage

```
trackplot_combine(
  tracks,
  side_plot = NULL,
  title = NULL,
  side_plot_width = 0.3
)
```

Arguments

tracks List of tracks in order from top to bottom, generally ggplots as output from the

other trackplot_*() functions.

side_plot Optional plot to align to the right (e.g. RNA expression per cluster). Will be

aligned to a trackplot_coverage() output if present, or else the first generic ggplot in the alignment. Should be in horizontal orientation and in the same

cluster ordering as the coverage plots.

title Text for overarching title of the plot

side_plot_width

Fraction of width that should be used for the side plot relative to the main track

area

Value

A plot object with aligned genome plots. Each aligned row has the text label, y-axis, and plot body. The relative height of each row is given by heights. A shared title and x-axis are put at the top.

See Also

```
trackplot_coverage(), trackplot_gene(), trackplot_loop(), trackplot_scalebar()
```

60 trackplot_coverage

trackplot_coverage

Pseudobulk coverage trackplot

Description

Plot a pseudobulk genome track, showing the number of fragment insertions across a region for each cell type or group.

Usage

```
trackplot_coverage(
  fragments,
  region,
  groups,
  cell_read_counts,
  group_order = NULL,
  bins = 500,
  clip_quantile = 0.999,
  colors = discrete_palette("stallion"),
  legend_label = "group",
  zero_based_coords = !is(region, "GRanges"),
  return_data = FALSE
)
```

Arguments

fragments Fragments object

region Region to plot, e.g. output from gene_region(). String of format "chr1:100-

200", or list/data.frame/GRanges of length 1 specifying chr, start, end. See

help("genomic-ranges-like") for details

groups Vector with one entry per cell, specifying the cell's group

cell_read_counts

Numeric vector of read counts for each cell (used for normalization)

group_order Optional vector listing ordering of groups bins Number of bins to plot across the region

clip_quantile (optional) Quantile of values for clipping y-axis limits. Default of 0.999 will

crop out just the most extreme outliers across the region. NULL to disable

clipping

colors Character vector of color values (optionally named by group)

legend_label Custom label to put on the legend

zero_based_coords

Whether to convert the ranges from a 1-based end-inclusive coordinate system to a 0-based end-exclusive coordinate system. Defaults to true for GRanges and

false for other formats (see this archived UCSC blogpost)

return_data If true, return data from just before plotting rather than a plot.
scale_bar Whether to include a scale bar in the top track (TRUE or FALSE)

trackplot_gene 61

Value

Returns a combined plot of pseudobulk genome tracks. For compatability with draw_trackplot_grid(), the extra attribute \$patches\$labels will be added to specify the labels for each track. If return_data or return_plot_list is TRUE, the return value will be modified accordingly.

See Also

trackplot_combine(), trackplot_gene(), trackplot_loop(), trackplot_scalebar()

trackplot_gene

Plot transcript models

Description

Plot transcript models

Usage

```
trackplot_gene(
  transcripts,
  region,
  exon_size = 2.5,
  gene_size = 0.5,
  label_size = 11 * 0.8/ggplot2::.pt,
  track_label = "Genes",
  return_data = FALSE
)
```

Arguments

transcripts

Transcipt features given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

- chr, start, end: genomic position
- strand: +/- or TRUE/FALSE for positive or negative strand
- feature: Only entries marked as "transcript" or "exon" will be considered
- gene_name: Symbol or gene ID to display
- transcript_id: Transcritp identifier to link transcripts and exons

Usually given as the output from read_gencode_transcripts()

region

Region to plot, e.g. output from gene_region(). String of format "chr1:100-200", or list/data.frame/GRanges of length 1 specifying chr, start, end. See help("genomic-ranges-like") for details

exon_size size for exon lines in units of mm

label_size size for transcript labels in units of mm

62 trackplot_loop

labels Character vector with labels for each item in transcripts. NA for items that

should not be labeled

transcript_size

size for transcript lines in units of mm

Value

Plot of gene locations

See Also

```
trackplot_combine(), trackplot_coverage(), trackplot_loop(), trackplot_scalebar()
```

trackplot_loop

Plot loops

Description

Plot loops

Usage

```
trackplot_loop(
  loops,
  region,
  color_by = NULL,
  colors = c("#bfd3e6", "#8c96c6", "#88419d", "#4d004b"),
  allow_truncated = TRUE,
  curvature = 0.75,
  track_label = "Links",
  return_data = FALSE
)
```

Arguments

loops Genomic regions given as GRanges, data.frame, or list. See help("genomic-ranges-like") for details on format and coordinate systems. Required attributes:

• chr, start, end: genomic position

region Region to plot, e.g. output from gene_region(). String of format "chr1:100-

200", or list/data.frame/GRanges of length 1 specifying chr, start, end. See

help("genomic-ranges-like") for details

color_by Name of a metadata column in loops to use for coloring, or a numeric vector

with same length as loops

colors Vector of hex color codes to use for the color gradient

trackplot_scalebar 63

allow_truncated

If FALSE, remove any loops that are not fully contained within region

curvature Curvature value between 0 and 1. 1 is a 180-degree arc, and 0 is flat lines.

Value

Plot of loops connecting genomic coordinates

See Also

```
trackplot_combine(), trackplot_coverage(), trackplot_gene(), trackplot_scalebar()
```

trackplot_scalebar

Plot scale bar

Description

Plots a human-readable scale bar and coordinates of the region being plotted

Usage

```
trackplot_scalebar(region, font_pt = 11)
```

Arguments

region Region to plot, e.g. output from gene_region(). String of format "chr1:100-

200", or list/data.frame/GRanges of length 1 specifying chr, start, end. See

help("genomic-ranges-like") for details

font_pt Font size for scale bar labels in units of pt.

Value

Plot with coordinates and scalebar for plotted genomic region

See Also

```
trackplot_combine(), trackplot_coverage(), trackplot_gene(), trackplot_loop()
```

transpose_storage_order

Transpose the storage order for a matrix

Description

Transpose the storage order for a matrix

Usage

```
transpose_storage_order(
  matrix,
  outdir = tempfile("transpose"),
  tmpdir = tempdir(),
  load_bytes = 4194304L,
  sort_bytes = 1073741824L
)
```

Arguments

matrix	Input matrix
outdir	Directory to store the output
tmpdir	Temporary directory to use for intermediate storage
load_bytes	The minimum contiguous load size during the merge sort passes
sort_bytes	The amount of memory to allocate for re-sorting chunks of entries

Details

This re-sorts the entries of a matrix to change the storage order from row-major to col-major. For large matrices, this can be slow – around 2 minutes to transpose a 500k cell RNA-seq matrix The default load_bytes (4MiB) and sort_bytes (1GiB) parameters allow ~85GB of data to be sorted with two passes through the data, and ~7.3TB of data to be sorted in three passes through the data.

Value

MatrixDir object with a copy of the input matrix, but the storage order flipped

```
write_fragments_memory
```

Read/write BPCells fragment objects

Description

BPCells fragments can be read/written in compressed (bitpacked) or uncompressed form in a variety of storage locations: in memory (as an R object), in an hdf5 file, or in a directory on disk (containing binary files).

Usage

```
write_fragments_memory(fragments, compress = TRUE)
write_fragments_dir(
  fragments,
  dir,
  compress = TRUE,
 buffer_size = 1024L,
  overwrite = FALSE
)
open_fragments_dir(dir, buffer_size = 1024L)
write_fragments_hdf5(
  fragments,
  path,
  group = "fragments",
  compress = TRUE,
 buffer_size = 8192L,
  chunk\_size = 1024L,
 overwrite = FALSE,
  gzip_level = 0L
)
open_fragments_hdf5(path, group = "fragments", buffer_size = 16384L)
```

Arguments

fragments	Input fragments object
compress	Whether or not to compress the data. With compression, storage size is be about half the size of a gzip-compressed 10x fragments file.
dir	Directory to read/write the data from
buffer_size	For performance tuning only. The number of items to be bufferred in memory before calling writes to disk.

overwrite If TRUE, write to a temp dir then overwrite existing data. Alternatively, pass a

temp path as a string to customize the temp dir location.

path Path to the hdf5 file on disk

group The group within the hdf5 file to write the data to. If writing to an existing hdf5

file this group must not already be in use

chunk_size For performance tuning only. The chunk size used for the HDF5 array storage.

gzip_level Gzip compression level. Default is 0 (no compression). This is recommended

when both compression and compatibility with outside programs is required. Otherwise, using compress=TRUE is recommended as it is >10x faster with

often similar compression levels.

Details

Saving in a directory on disk is a good default for local analysis, as it provides the best I/O performance and lowest memory usage. The HDF5 format allows saving within existing hdf5 files to group data together, and the in memory format provides the fastest performance in the event memory usage is unimportant.

Value

Fragment object

write_matrix_memory

Read/write sparse matrices

Description

BPCells matrices are stored in sparse format, meaning only the non-zero entries are stored. Matrices can store integer counts data or decimal numbers (float or double). See details for more information.

Usage

```
write_matrix_memory(mat, compress = TRUE)
write_matrix_dir(
   mat,
   dir,
   compress = TRUE,
   buffer_size = 8192L,
   overwrite = FALSE
)
open_matrix_dir(dir, buffer_size = 8192L)
write_matrix_hdf5(
   mat,
```

write_matrix_memory 67

```
path,
  group,
  compress = TRUE,
  buffer_size = 8192L,
  chunk_size = 1024L,
  overwrite = FALSE,
  gzip_level = 0L
)

open_matrix_hdf5(path, group, buffer_size = 16384L)
```

Arguments

compress Whether or not to compress the data.

dir Directory to save the data into

buffer_size For performance tuning only. The number of items to be buffered in memory

before calling writes to disk.

overwrite If TRUE, write to a temp dir then overwrite existing data. Alternatively, pass a

temp path as a string to customize the temp dir location.

path Path to the hdf5 file on disk

group The group within the hdf5 file to write the data to. If writing to an existing hdf5

file this group must not already be in use

chunk_size For performance tuning only. The chunk size used for the HDF5 array storage.

gzip_level Gzip compression level. Default is 0 (no compression). This is recommended

when both compression and compatibility with outside programs is required. Otherwise, using compress=TRUE is recommended as it is >10x faster with

often similar compression levels.

matrix Input matrix, either IterableMatrix or dgCMatrix

Details

Storage locations:

Matrices can be stored in a directory on disk, in memory, or in an HDF5 file. Saving in a directory on disk is a good default for local analysis, as it provides the best I/O performance and lowest memory usage. The HDF5 format allows saving within existing hdf5 files to group data together, and the in memory format provides the fastest performance in the event memory usage is unimportant.

Bitpacking Compression:

For typical RNA counts matrices holding integer counts, this bitpacking compression will result in 6-8x less space than an R dgCMatrix, and 4-6x smaller than a scipy csc_matrix. The compression will be more effective when the count values in the matrix are small, and when the rows of the matrix are sorted by rowMeans. In tests on RNA-seq data optimal ordering could save up to 40% of storage space. On non-integer data only the row indices are compressed, not the values themselves so space savings will be smaller.

For non-integer data matrices, bitpacking compression is much less effective, as it can only be applied to the indexes of each entry but not the values. There will still be some space savings, but far less than for counts matrices.

Value

BPCells matrix object

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