# Package 'scCNAutils'

March 15, 2019

<b>Title</b> Functions to analyze copy number aberrations in single-cell data
Version 0.0.0.9000
<b>Description</b> Functions to analyze copy number aberrations in singlecell data. A bunch of scripts and workflows to read and analyze scRNA-seq data and look at CNA-oriented signal.
<b>Depends</b> R (>= $3.4.4$ )
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LazyData true
Imports Matrix,  dplyr,  magrittr,  tidyr,  rlang,  parallel,  ggplot2,  data.table,  Rtsne,  FNN,  igraph,  RcppHMM,  GenomicRanges,  IRanges,  shiny,  rbokeh,  scales,  ggrepel
Suggests testthat
RoxygenNote 6.1.1
R topics documented:  annotate_cna

2 annotate\_cna

call_cna call_cna_multisamps convert_to_coord define_cycling_cells find_communities gene_info make_metacells merge_samples norm_ge plot_aneuploidy plot_cna plot_communities plot_ccells plot_usne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_tsne run_tsne run_umap smooth_movingw tsne_browser umap_browser umap_browser usinsor zscore	. 4 . <del>6</del>
convert_to_coord  define_cycling_cells find_communities gene_info make_metacells merge_samples norm_ge plot_aneuploidy plot_cna plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
define_cycling_cells find_communities gene_info make_metacells merge_samples norm_ge plot_aneuploidy plot_cna plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	. 7
find_communities gene_info make_metacells merge_samples norm_ge plot_aneuploidy plot_cna plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	. 8
gene_info make_metacells merge_samples norm_ge  plot_aneuploidy plot_cna  plot_communities plot_qc_cells plot_tsne .  plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	. 9
make_metacells merge_samples norm_ge plot_aneuploidy plot_cna plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	. 9
merge_samples norm_ge plot_aneuploidy plot_cna plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser umap_browser winsor	. 10
norm_ge plot_aneuploidy plot_cna plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser umap_browser winsor	
plot_aneuploidy plot_cna  plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	. 12
plot_cna plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
plot_communities plot_qc_cells plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
plot_qc_cells plot_tsne .  plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
plot_tsne plot_umap qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
plot_umap  qc_cells  qc_filter  read_mtx  rebin_cov  rm_cv_outliers  run_louvain  run_pca  run_tsne  run_umap  smooth_movingw  tsne_browser  umap_browser  winsor	
qc_cells qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
qc_filter read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
read_mtx rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
rebin_cov rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
rm_cv_outliers run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
run_louvain run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
run_pca run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
run_tsne run_umap smooth_movingw tsne_browser umap_browser winsor	
run_umap	
smooth_movingw	
tsne_browser	
umap_browser      winsor	
winsor	
zscore	
	. 27
	28

annotate\_cna

Annotate CNAs with gene information

## Description

Add columns with the names of the genes in the region, total and expressed genes only.

## Usage

```
annotate_cna(cna_df, gene_info)
```

auto\_cna\_call 3

#### Arguments

cna\_df a data.frame with CNA calls. E.g. data.frames created by call\_cna\_\* func-

tions.

gene\_info a data.frame with gene information created by the gene\_info function

#### **Details**

The subset of "expressed" genes is made out of genes with non-zero expression in at least 10 expression higher than 0.5.

#### Value

the input data.frame with two new columns with all/expressed genes.

#### Author(s)

Jean Monlong

auto\_cna\_call

Automated pipeline to call CNA

#### **Description**

Automated pipeline to call CNA using metacells.

#### Usage

```
auto_cna_call(ge_df, comm_df, nb_metacells = 10, metacell_size = 3,
  multisamps = TRUE, trans_prob = 0.1, baseline_cells = NULL,
  baseline_communities = NULL, prefix = "scCNAutils_out",
  nb_cores = 1, chrs = c(1:22, "X", "Y"), bin_mean_exp = 3,
  z_wins_th = 3, smooth_wsize = 3)
```

#### **Arguments**

ge\_df normalized gene expression of all cells (e.g. output from norm\_ge.

comm\_df a data.frame with community information, output from find\_communities.

 $\label{lem:nb_metacells} \mbox{ the number of metacells per comunity.}$ 

metacell\_size the number of cells in a metacell.

multisamps use the multi-sample version of the HMM segmentation? Default is TRUE. See

details

trans\_prob the transition probability for the HMM.

baseline\_cells cells to use as baseline.

baseline\_communities

communities to use as baseline. Used if baseline.cells is NULL.

4 auto\_cna\_signal

prefix the prefix to use for the files created by this function (e.g. graphs).

nb\_cores the number of processors to use.

chrs the chromosome names to keep. NULL to include all the chromosomes.

bin\_mean\_exp the desired minimum mean expression in the bin.

z\_wins\_th the threshold to winsorize Z-score. Default is 3

smooth\_wsize the window size for smoothing. Default is 3.

#### **Details**

Once the metacells are created there are two ways to call CNA. First, if multisamps=FALSE, to call CNA on each metacell and merge the result per community, keeping the information about how many metacell support the CNA. Second, if multisamps=TRUE (default), to run the HMM on all the metacells for a community. The multi-sample approach should be more robust.

The transition probability (trans\_prob) is going to affect the HMM segmentation. Smaller values will create longer segments. One approach, often advocated by HMM aficionados, is to try different values and use the ones that gives the best results, for example based on the QC graphs (TODO). Another approach is to use a loose transition probability and then filter short segments ('length' column or 'pass.filter' column).

#### Value

a data.frame with CNAs

## Author(s)

Jean Monlong

auto\_cna\_signal

Automated pipeline to compute CNA signal from scRNA expression

## Description

Goes from reading raw gene counts to CNA-level signal, tSNE and community detection.

#### Usage

```
auto_cna_signal(data, genes_coord, prefix = "scCNAutils_out",
   nb_cores = 1, pause_after_qc = FALSE, use_cache = TRUE,
   sample_names = NULL, info_df = NULL, max_mito_prop = 0.2,
   min_total_exp = 0, cells_sel = NULL, chrs = c(1:22, "X", "Y"),
   cell_cycle = NULL, bin_mean_exp = 3, rm_cv_quant = NULL,
   z_wins_th = 3, smooth_wsize = 3, cc_sd_th = 3, nb_pcs = 10,
   comm_k = 100, viz = c("tsne", "umap", "both"))
```

auto\_cna\_signal 5

#### **Arguments**

data a data.frame with gene expression or the path to the folder with the 'matrix.mtx',

'genes.tsv' and 'barcodes.tsv' files. A list if multiple samples.

genes\_coord either a file name or a data.frame with coordinates and gene names.

prefix the prefix to use for the files created by this function (e.g. graphs).

nb\_cores the number of processors to use.

pause\_after\_qc pause after the QC to pick custom QC thresholds.

use\_cache should intermediate files used and avoid redoing steps?

sample\_names the names of each sample. If NULL, tries to use data's names.

info\_df a data.frame with information about cells.

max\_mito\_prop the maximum proportion of mitochondrial RNA.

min\_total\_exp the minimum total cell expression

cells\_sel consider only these cells. Other cells filtered no matter what.

chrs the chromosome names to keep. NULL to include all the chromosomes.

cell\_cycle if non-null, either a file or data.frame to compute cell cycle scores. See details.

bin\_mean\_exp the desired minimum mean expression in the bin.

rm\_cv\_quant the quantile threshold to remove CV outlier. Default NULL (i.e. not used).

z\_wins\_th the threshold to winsorize Z-score. Default is 3

smooth\_wsize the window size for smoothing. Default is 3.

cc\_sd\_th the number of SD used for the thresholds when defining cycling cells.

nb\_pcs the number of PCs used in the community detection or tSNE.

comm\_k the number of nearest neighbor for the KNN graph. Default 100.

viz which method to use for visualization ('tsne', 'umap' or 'both'). Default is

'tsne'.

#### Value

a data.frame with QC, community and tSNE for each cell.

#### Author(s)

6 call\_cna

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Merge consecutive genes into expressed bins

## **Description**

Merge consecutive genes into expressed bins

## Usage

```
bin_genes(ge_df, mean_exp = 3, nb_cores = 1)
```

## **Arguments**

ge\_df the input gene expression with coordinate columns (chr, start, end) and then one

column per cell.

mean\_exp the desired minimum mean expression in the bin.

nb\_cores the number of processors to use.

## Value

a data.frame with bin expression.

## Author(s)

Jean Monlong

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cal	1	cna

Call CNA

## **Description**

Calls CNA using a HMM approach.

#### Usage

```
call_cna(z_df, trans_prob = 1e-04, nb_cores = 1, mc_info = NULL)
```

#### **Arguments**

z\_df the Z-scores, from zscore.

trans\_prob the transition probability for the HMM.

nb\_cores the number of processor to use.

mc\_info the information about the metacells, if relevant. Default is NULL.

call\_cna\_multisamps 7

## Value

a data.frame with the CNA calls.

## Author(s)

Jean Monlong

call\_cna\_multisamps Call CNA

## Description

Calls CNA using a HMM approach considering multiple samples at the same time.

## Usage

```
call_cna_multisamps(z_df, mc_info, trans_prob = 1e-04, nb_cores = 1)
```

## Arguments

z\_df the Z-scores, from zscore.

mc\_info the information about the metacells, if relevant. Default is NULL.

trans\_prob the transition probability for the HMM.

nb\_cores the number of processor to use.

## Value

a data.frame with the CNA calls.

#### Author(s)

8 convert\_to\_coord

convert_to_coord	Convert gene symbols to coordinates	

## **Description**

Convert the 'symbol' column (gene names) into three columns with gene coordinates 'chr', 'start' and 'end'.

## Usage

```
convert_to_coord(ge_df, genes_coord, chrs = c(1:22, "X", "Y"),
    rm_dup = TRUE)
```

## **Arguments**

ge\_df the data.frame with gene expression and one column 'symbol' with gene names.

genes\_coord either a file name or a data.frame with coordinates and gene names.

chrs the chromosome names to keep. NULL to include all the chromosomes.

rm\_dup remove duplicated coordinates? Default is TRUE.

#### **Details**

If *genes\_coord* is a filename, the file is expected to be a tab-delimited file with four columns: 'chr', 'start', 'end', 'symbol'. The order of the columns is not important.

The gene names in column 'symbol' should match the gene names in the input  $ge\_df$ .

## Value

a data.frame with columns 'chr', 'start', 'end' columns with genes coordinates (and still one column per barcode).

## Author(s)

define\_cycling\_cells 9

## **Description**

Using cell cycle scores, identify cells that are cycling.

## Usage

```
define_cycling_cells(qc_df, sd_th = 3)
```

#### **Arguments**

qc\_df the output data.frame from qc\_cells (ran with a non-null *cell\_cycle* parameter) sd th the number of SD used for the thresholds.

#### Value

a list with

cells.noc a vector with the names of non-cycling cells

graphs a list of ggplot2 graphs

## Author(s)

Jean Monlong

find\_communities Community detection

#### **Description**

Build a KNN graph and run Louvain algorithm for community detection.

#### Usage

```
find_communities(pca_o, nb_pcs = 10, k = 100, gamma = 1, nreps = 1,
   nb_cores = 1)
```

## Arguments

pca_o	the output of run_pca
-------	-----------------------

nb\_pcs the number of PCs to use. Default 10.

k the number of nearest neighbor for the KNN graph. Default 100.

gamma a vector of gamma. Default 1.

nreps the number of repetition for each gamma, Default 1.

nb\_cores the number of processors to use. Default is 1.

10 gene\_info

#### Value

a list with:

comm a data.frame with two columns: 'cell' and 'community'.

comm. all a matrix with communities for each gamma

gamma the list of input gamma corresponding to each comm.all column.

best.gamma the gamma resulting on the highest ARI mean ari.df data.frame with ARI stats for each gamma

## Author(s)

Jean Monlong

gene\_info Gene information

### **Description**

Computes information at the gene level that can be used to annotate the results later (e.g. CNA calls).

#### Usage

```
gene_info(ge_df, genes_coord)
```

## **Arguments**

ge\_df a data.frame or path to a file with gene expression information across cells. E.g.

after norm.ge.

genes\_coord either a file name or a data.frame with coordinates and gene names.

## **Details**

If the input data.frame has a symbol column, 'genes\_coord' is used to add the coordinates, later used to overlap with region/CNA calls. If the input data.frame has coordinates, 'genes\_coord' is used to retrieve the gene name. In both cases a row in the input data.frame must be a gene, not a bin

It's better to run this function after normalization (norm\_ge but before binning (bin\_genes).

## Value

a data.frame with summary stats for each gene.

#### Author(s)

make\_metacells 11

_metacells

## Description

Randomly select cells in each community and merge them to create metacells with higher resolution.

## Usage

```
make_metacells(ge_df, comm_df, nb_metacells = 10, metacell_size = 3,
  baseline_cells = NULL, nb_cores = 1)
```

## Arguments

ge_df	normalized gene expression of all cells (e.g. output from norm_ge.
comm_df	a data.frame with community information, output from $\label{find_communities} \textbf{find_communities}.$
nb_metacells	the number of metacells per comunity.
metacell_size	the number of cells in a metacell.
baseline_cells	the cells to use for baseline communities.
nb_cores	the number of processor to use.

## Value

a list with

ge a data.frame with coordinates and gene expression for each metacell.

info information about which metacell correspond to which community.

mc\_cells information about which cells were used for each metacell.

#### Author(s)

norm\_ge

merge_	samp.	Les

Merge expression of multiple samples

## **Description**

The expression of multiple samples are merged. New cell names are produced as SAMPLE\_CELL.

## Usage

```
merge_samples(ge_list, sample_names = NULL)
```

## Arguments

```
ge_list a list of ge_df (e.g. read from read_mtx).
```

sample\_names the names of each sample. If NULL, tries to use ge\_list's names.

#### Value

a list with

ge the merged gene expression data.frame

info a data.frame with new and original cell names, and corresponding sample name

#### Author(s)

Jean Monlong

norm_	ge
-------	----

Normalize gene expression

## **Description**

The expression of each cell is normalized to account for depth differences.

## Usage

```
norm_ge(ge_df, method = c("tmm", "total"), nb_cores = 1)
```

#### **Arguments**

ge\_df the input gene expression method the normalization method

nb\_cores the number of processors to use.

plot\_aneuploidy 13

## Value

a data.frame with the normalized expression.

#### Author(s)

Jean Monlong

plot\_aneuploidy

Aneuploidy graph

#### **Description**

Graphs showing the median expression in each chromosome for each community.

## Usage

```
plot_aneuploidy(ge_df, comm_df = NULL, baseline_cells = NULL,
  baseline_communities = NULL, max_cells = 100, chrs_order = c(1:22,
  "X", "Y"))
```

## Arguments

ge\_df a data.frame with gene expression (better if binned and normalized).

comm\_df a data.frame with community information for each cell.

baseline\_cells cells to use as baseline.

baseline\_communities

the communities to use as baseline.

max\_cells the maximum number of cells to consider in the boxplot of each community.

Default: 100.

chrs\_order order of the chromosomes in the graph.

## Value

a list of ggplot2 object, one for each chromosome.

#### Author(s)

plot\_communities

plot\_cna

Heatmap of CNA

## **Description**

Heatmap of CNA

## Usage

```
plot_cna(cna, chrs_order = c(1:22, "X", "Y"))
```

## **Arguments**

cna CNAs from call\_cna.

chrs\_order order of the chromosomes in the graph.

#### Value

a ggplot2 graph

## Author(s)

Jean Monlong

plot\_communities

Community graphs

## Description

Graphs about the communities found by find\_communities. For example the size of the communities or the distribution of QC metrics in each community.

#### Usage

```
plot_communities(comm_df, qc_df = NULL, info_df = NULL)
```

#### **Arguments**

comm df	the output	data frama	from find	communities
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qc\_df a data.frame with QC metrics (output from qc\_cells). Default is NULL (i.e.

not used)

info\_df a data.frame with sample merge info (output from merge\_samples). Default is

NULL (i.e. not used)

plot\_qc\_cells 15

#### **Details**

If the QC data.frame is provided, the distribution of QC metrics is shown to investigate if some communities are batch effects.

If multiple samples were merged (merge\_samples), the proportion of cells from each sample of origin can be shown if the info\_df data.frame is provided.

If qc\_df and/or info\_df are null but their columns present in comm\_df, their corresponding graphs will be generated. Hence a merged version of comm\_df, qc\_df and info\_df works (e.g. output of auto\_cna\_signal.

#### Value

```
a list of ggplot2 graphs.
```

#### Author(s)

Jean Monlong

## **Examples**

```
## Not run:
ggp.l = plot_communities(comm_df, qc_df)

## Print first graph
ggpl.l[[1]]

## Customize ggplot
ggpl.l[[1]] + ggtitle('First graph about communities')

## End(Not run)
```

plot\_qc\_cells

QC graphs

## **Description**

```
QC graphs
```

## Usage

```
plot_qc_cells(qc_df, info_df = NULL)
```

## Arguments

qc\_df the output data.frame from qc\_cells

info\_df a data.frame with sample merge info (output from merge\_samples). Default is

NULL (i.e. not used)

plot\_tsne

#### Value

```
a list of ggplots
```

#### Author(s)

Jean Monlong

## **Examples**

```
## Not run:
ggp.l = plot_qc_cells(qc_df)

## Print first graph
ggpl.l[[1]]

## Customize ggplot
ggpl.l[[1]] + ggtitle('First QC graph')

## End(Not run)
```

plot\_tsne

tSNE graphs

## **Description**

tSNE graphs colored according to QC metrics or sample labels.

## Usage

```
plot_tsne(tsne_df, qc_df = NULL, comm_df = NULL, info_df = NULL)
```

#### **Arguments**

tsne_df	the output data.frame from run_tsne (columns: cell, tsne1, tsne2)
qc_df	a data.frame with QC metrics (output from $qc\_cells$ ). Default is NULL (i.e. not used)
comm_df	a data.frame with communities (output from find_communities). Default is NULL (i.e. not used)
info_df	a data.frame with sample merge info (output from merge_samples).

#### **Details**

If the QC data.frame is provided, the distribution of QC metrics is shown to investigate if some communities are batch effects.

If multiple samples were merged (merge\_samples), the points can be colored by sample of origin by providing the info\_df data.frame.

If any qc\_df/comm\_df/info\_df are null but their columns present in tsne\_df, their corresponding graphs will be generated. Hence a merged version of tsne\_df, comm\_df, qc\_df and info\_df works (e.g. output of auto\_cna\_signal.

plot\_umap 17

#### Value

```
a list of ggplot objects
```

#### Author(s)

Jean Monlong

## **Examples**

```
## Not run:
ggp.l = plot_tsne(tsne_df, qc_df, comm_df)

## Print first graph
ggpl.l[[1]]

## Customize ggplot
ggpl.l[[1]] + ggtitle('First tSNE graph')

## End(Not run)
```

plot\_umap

UMAP graphs

## **Description**

UMAP graphs colored according to QC metrics or sample labels.

## Usage

```
plot_umap(umap_df, qc_df = NULL, comm_df = NULL, info_df = NULL)
```

#### **Arguments**

umap_df	the output data.frame from run_umap (columns: cell, umap1, umap2)
qc_df	a data.frame with QC metrics (output from qc_cells). Default is NULL (i.e. not used)
comm_df	a data.frame with communities (output from find_communities). Default is NULL (i.e. not used)
info_df	a data.frame with sample merge info (output from merge_samples).

## **Details**

If the QC data.frame is provided, the distribution of QC metrics is shown to investigate if some communities are batch effects.

If multiple samples were merged (merge\_samples), the points can be colored by sample of origin by providing the info\_df data.frame.

If any qc\_df/comm\_df/info\_df are null but their columns present in umap\_df, their corresponding graphs will be generated. Hence a merged version of umap\_df, comm\_df, qc\_df and info\_df works (e.g. output of auto\_cna\_signal.

18 qc\_cells

#### Value

a list of ggplot objects

#### Author(s)

Jean Monlong

#### **Examples**

```
## Not run:
ggp.l = plot_umap(umap_df, qc_df, comm_df)
## Print first graph
ggpl.l[[1]]
## Customize ggplot
ggpl.l[[1]] + ggtitle('First umap graph')
## End(Not run)
```

qc\_cells

Compute quality control metrics for each cell

## Description

From raw gene expression, a few QC metrics are computed.

#### Usage

```
qc_cells(ge_df, cell_cycle = NULL)
```

#### **Arguments**

ge\_df the input gene expression with a 'symbol' column and then one column per cell. cell\_cycle if non-null, either a file or data.frame to compute cell cycle scores. See details.

## **Details**

If cell\_cycle is provided it should be a data.frame (or a tsv file) with two columns: 'symbol' with gene names, and 'phase' with the cell cycle phase (e.g. either 'G1.S' or 'G2.M').

## Value

a data.frame with qc metrics per cell.

#### Author(s)

qc\_filter 19

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Filter cells based on QC results

## **Description**

Filter cells based on QC results

#### Usage

```
qc_filter(ge_df, qc_df, max_mito_prop = 0.2, min_total_exp = 0,
    cells_sel = NULL)
```

## **Arguments**

ge\_df the input gene expression with a 'symbol' column and then one column per cell.

qc\_df the output data.frame from qc\_cells

max\_mito\_prop the maximum proportion of mitochondrial RNA.

min\_total\_exp the minimum total cell expression

cells\_sel consider only these cells. Other cells filtered no matter what.

#### Value

ge\_df with only the cells that passed the filters

#### Author(s)

Jean Monlong

read\_mtx

Read a trio of genes, barcodes and mtx files.

## **Description**

Read a trio of genes, barcodes and mtx files.

## Usage

```
read_mtx(mtx_file = "matrix.mtx", genes_file = "genes.tsv",
  barcodes_file = "barcodes.tsv", path = ".", rm_dup = TRUE,
  genes_col = 2)
```

20 rebin\_cov

#### **Arguments**

mtx\_file the path to the mtx file genes\_file the path to the genes file. barcodes\_file the path to the barcodes file

path the path to the folder containing the files

rm\_dup remove duplicated gene names? Default is TRUE. genes\_col the column to use in genes\_file. Default is 2.

#### Value

a data.frame with a 'symbol' column with gene names and one column per barcode.

#### Author(s)

Jean Monlong

rebin\_cov

Re-bin coverage data

## **Description**

The new bins are overlapped with the regions in cov\_df to compute a weight. The coverage in the new bin is the weighted sum of the coverage in overlapping regions of cov\_df.

#### Usage

```
rebin_cov(cov_df, bins)
```

## **Arguments**

cov\_df a data.frame with coverage information.

bins a data.frame or GRanges object with the new bins.

## **Details**

Coordinates in data.frame are expected to be defined by columns names 'chr', 'start' and 'end'.

#### Value

a data.frame with new bins

#### Author(s)

rm\_cv\_outliers 21

rm_cv_outliers	Remove outliers based on the coefficient of variation
----------------	---

## Description

Compute the coefficient of variation for each gene/bin and remove the ones with the highest values, either based on a quantile or SD-based threshold. The genes/bins that satisfy both quantile and SD-based thresholds are removed.

## Usage

```
rm_cv_outliers(ge_df, ol_quant_th = 0.99, ol_sd_th = 5)
```

## Arguments

ge\_df the expression data.frame.

ol\_quant\_th the quantile threshold. Default is 0.99 (removes the top 1% with highest values).

ol\_sd\_th the SD-based treshold. Default is 5.

#### Value

a subset of the ge\_df data.frame

## Author(s)

Jean Monlong

run_louvain Python wrapper to run Louvain	
---	--

## Description

Louvain on an igraph object.

#### Usage

```
run_louvain(graph, gamma = 1, nreps = 1, nb_cores = 1)
```

## **Arguments**

graph a igraph object

gamma a vector of gamma. Default 1.

nreps the number of repetition for each gamma, Default 1.

nb\_cores the number of processors to use. Default is 1.

zu\_pca

#### **Details**

This functions depends on Python and louvain being installed. Make sure igraph, louvain and numpy are installed. For example with something like: 'pip install python-igraph louvain numpy'.

#### Value

a list with

comm a data.frame with the community for each gamma

gamma the input gammas corresponding to the columns of comm

#### Author(s)

Jean Monlong

|--|

#### **Description**

PCA analysis, eventually using a subset of core cells for the PC construction.

#### Usage

```
run_pca(z_df, core_cells = NULL, out_pcs = 100)
```

#### **Arguments**

z\_df a data.frame with z-scores for each cell

core\_cells if non-NULL, a vector with the names of the cells to use as core cells. See

details. Default is NULL.

out\_pcs the number of top PCs to report. Default is 100.

#### **Details**

Cells in core\_cells are used to build the principal components to which all cells are then projected to. Usually used to reduce the effect of cell cycle in the PCA, by using only cells that don't cycle (see qc\_cells) as *core\_cells*.

The graph (*sdev.graph*) shows the standard deviation for the top 50 PCs. To show more/less PCs, add xlim(1,N) to the *sdev.graph*. See examples.

#### Value

a list with

x the PC matrix

sdev the standard deviations of the PCs

sdev.graph a ggplot graph of the sdev

run\_tsne 23

#### Author(s)

Jean Monlong

## **Examples**

```
## Not run:
pca.o = run_pca(z)
## Zoom in to the top 20 PCs
pca.o$sdev.graph + xlim(1,20)
## End(Not run)
```

run\_tsne

Run tSNE

## Description

tSNE from PCA results.

## Usage

```
run_tsne(pca_o, nb_pcs = 10, nb_it = 1000, tsne_init = NULL)
```

## Arguments

pca\_o the output of run\_pca
nb\_pcs the number of PCs to use. Default 10.

nb\_it the number of iterations. Default 1000.

tsne\_init previous tSNE results to use as starting point. Not used is NULL (default).

#### Value

```
a data.frame with columns: cell, tsne1, tsne2
```

#### Author(s)

24 smooth\_movingw

run_umap	Run UMAF

## **Description**

UMAP on the PCA results.

## Usage

```
run_umap(pca_o, nb_pcs = 10, nb_neighbors = 5)
```

## **Arguments**

pca\_o the output of run\_pca

nb\_pcs the number of PCs to use. Default 10.
nb\_neighbors the number of neighbors. Default 5.

#### **Details**

This functions depends on Python and UMAP being installed. Make sure umap-mearn, sklearn, numpy, scipy and pandas are installed. For example with something like: 'pip install sklearn numpy scipy pandas umap-learn'.

## Value

```
a data.frame with columns: cell, umap1, umap2
```

#### Author(s)

Jean Monlong

|--|

## **Description**

The expression/score of a gene/bin is replaced by a summary of bins around. For example the median across 3 bins.

## Usage

```
smooth_movingw(df, wsize = 3, nb_cores = 1, FUN = stats::median)
```

tsne\_browser 25

#### **Arguments**

df the input data frame with coordinate columns (chr, start, end) and then one col-

umn per cell

wsize the window size. Default is 3. nb\_cores the number of processors to use.

FUN the function to apply to each window. Default is median.

#### Value

a data.frame with smoothed signal.

#### Author(s)

Jean Monlong

tsne\_browser Shiny application to visualize tSNE results

#### **Description**

Interactive application to visualize the tSNE results: zoom, hover information, different colors.

#### Usage

```
tsne_browser(cells_df, nb_points = 5000, plot_dim = 800)
```

## Arguments

cells\_df the data.frame with tSNE and other information for each cell

nb\_points the default number of points to show. See details.

plot\_dim the dimension of the plot in pixels.

#### **Details**

Drawing thousands of points in a web-browser can be demanding. To reduce the number of points (cells) to draw, close-by cells are merged into bigger points. The merging is done separately for different samples/communities to be able to color them if necessary. The user can decide how many points to draw with the 'nb\_points' parameter or directly within the application. In practice, increase the number of points until the app gets too slow.

#### Value

opens a Shiny app in a web-browser.

#### Author(s)

26 winsor

umap_browser	Shiny application to visualize UMAP results	

## Description

Interactive application to visualize the UMAP results: zoom, hover information, different colors.

#### Usage

```
umap_browser(cells_df, nb_points = 5000, plot_dim = 800)
```

### **Arguments**

cells\_df the data.frame with UMAP and other information for each cell

nb\_points the default number of points to show. See details.

plot\_dim the dimension of the plot in pixels.

#### **Details**

Drawing thousands of points in a web-browser can be demanding. To reduce the number of points (cells) to draw, close-by cells are merged into bigger points. The merging is done separately for different samples/communities to be able to color them if necessary. The user can decide how many points to draw with the 'nb\_points' parameter or directly within the application. In practice, increase the number of points until the app gets too slow.

#### Value

opens a Shiny app in a web-browser.

#### Author(s)

Jean Monlong

winsor	Winsorize

## Description

Convenience function to winsorize a vector.

## Usage

```
winsor(x, u = NULL, l = NULL, uq = NULL)
```

zscore 27

## **Arguments**

X	input vector
u	upper limit
1	lower limit

uq the quantile for the upper limit. Used is u is NULL.

## Value

winsorized vector

## Author(s)

Jean Monlong

## Description

Transform gene expression into a scaled score, either using all cells or a subset of cells as baseline.

## Usage

```
zscore(ge_df, wins_th = 3, method = c("z", "norm"), normals = NULL)
```

## Arguments

ge_at	the input expression data.frame
wins_th	the threshold to winsorize Z-score. Default is 3
method	the normalization method. Either 'z' or norm'.

normals the cells to use as normals. If NULL (default) all cells are used as normals

## Value

a data.frame with Z-scores.

## Author(s)

## **Index**

```
umap_browser, 26
annotate_cna, 2
auto_cna_call, 3
                                                 winsor, 26
auto_cna_signal, 4, 15-17
                                                 zscore, 6, 7, 27
bin_genes, 6
call_cna, 6, 14
call_cna_multisamps, 7
convert_to_coord, 8
define_cycling_cells, 9
find_communities, 3, 9, 11, 14, 16, 17
gene_info, 10
make_metacells, 11
merge_samples, 12, 14–17
norm_ge, 3, 11, 12
plot_aneuploidy, 13
plot_cna, 14
plot_communities, 14
plot_qc_cells, 15
plot_tsne, 16
plot_umap, 17
qc_cells, 9, 14-17, 18, 22
qc_filter, 19
read_mtx, 12, 19
rebin_cov, 20
rm_cv_outliers, 21
run_louvain, 21
run_pca, 9, 22, 23, 24
run_tsne, 16, 23
run_umap, 17, 24
smooth_movingw, 24
tsne_browser, 25
```