

# Package ‘cicero’

December 5, 2024

**Type** Package

**Title** Predict cis-co-accessibility from single-cell chromatin accessibility data

**Version** 1.24.0

**Description** Cicero computes putative cis-regulatory maps from single-cell chromatin accessibility data. It also extends monocle 2 for use in chromatin accessibility data.

**Depends** R (>= 3.5.0), monocle, Gviz (>= 1.22.3)

**License** MIT + file LICENSE

**Encoding** UTF-8

**Imports** assertthat (>= 0.2.0), Biobase (>= 2.37.2), BiocGenerics (>= 0.23.0), data.table (>= 1.10.4), dplyr (>= 0.7.4), FNN (>= 1.1), GenomicRanges (>= 1.30.3), ggplot2 (>= 2.2.1), glasso (>= 1.8), grDevices, igraph (>= 1.1.0), IRanges (>= 2.10.5), Matrix (>= 1.2-12), methods, parallel, plyr (>= 1.8.4), reshape2 (>= 1.4.3), S4Vectors (>= 0.14.7), stats, stringi, stringr (>= 1.2.0), tibble (>= 1.4.2), tidyr, VGAM (>= 1.0-5), utils

**RoxygenNote** 7.2.3

**Suggests** AnnotationDbi (>= 1.38.2), knitr, markdown, rmarkdown, rtracklayer (>= 1.36.6), testthat, vdiff (>= 0.2.3), covr

**VignetteBuilder** knitr

**biocViews** Sequencing, Clustering, CellBasedAssays, ImmunoOncology, GeneRegulation, GeneTarget, Epigenetics, ATACSeq, SingleCell

**LazyData** true

**git\_url** <https://git.bioconductor.org/packages/cicero>

**git\_branch** RELEASE\_3\_20

**git\_last\_commit** d20fd1a

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

**Repository** Bioconductor 3.20

**Date/Publication** 2024-12-05

**Author** Hannah Pliner [aut, cre],  
Cole Trapnell [aut]

**Maintainer** Hannah Pliner <hpliner@uw.edu>

## Contents

cicero-package . . . . .	2
aggregate_by_cell_bin . . . . .	3
aggregate_nearby_peaks . . . . .	3
annotate_cds_by_site . . . . .	4
assemble_connections . . . . .	5
build_gene_activity_matrix . . . . .	6
cell_data . . . . .	7
cicero_data . . . . .	8
compare_connections . . . . .	8
df_for_coords . . . . .	9
estimate_distance_parameter . . . . .	10
find_overlapping_ccans . . . . .	12
find_overlapping_coordinates . . . . .	13
generate_ccans . . . . .	13
generate_cicero_models . . . . .	14
gene_annotation_sample . . . . .	16
human.hg19.genome . . . . .	17
make_atac_cds . . . . .	17
make_cicero_cds . . . . .	18
make_sparse_matrix . . . . .	19
normalize_gene_activities . . . . .	20
plot_accessibility_in_pseudotime . . . . .	21
plot_connections . . . . .	21
ranges_for_coords . . . . .	24
run_cicero . . . . .	25
<b>Index</b>	<b>27</b>

---

cicero-package	<i>cicero</i>
----------------	---------------

---

## Description

Cicero computes putative cis-regulatory maps from single-cell chromatin accessibility data. It also extends monocle 2 for use in chromatin accessibility data.

## Author(s)

**Maintainer:** Hannah Pliner <hpliner@uw.edu>

Authors:

- Cole Trapnell <colettrap@uw.edu>

---

aggregate\_by\_cell\_bin *Aggregate count CDS by groups of cells*

---

### Description

Aggregates a CDS based on an indicator column in the pData table

### Usage

```
aggregate_by_cell_bin(cds, group_col)
```

### Arguments

cds	A CDS object to be aggregated
group_col	The name of the column in the pData table that indicates the cells assignment to its aggregate bin.

### Details

This function takes an input CDS object and collapses cells based on a column in the pData table by summing the values within the cell group.

### Value

A count cds aggregated by group\_col

### Examples

```
data("cicero_data")
#input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
#pData(input_cds)$cell_subtype <- rep(1:10, times=20)
#binned_input_lin <- aggregate_by_cell_bin(input_cds, "cell_subtype")
```

---

aggregate\_nearby\_peaks

*Make an aggregate count cds by collapsing nearby peaks*

---

### Description

Make an aggregate count cds by collapsing nearby peaks

### Usage

```
aggregate_nearby_peaks(cds, distance = 1000)
```

### Arguments

cds	A CellDataSet (CDS) object. For example, output of <a href="#">make_atac_cds</a>
distance	The distance within which peaks should be collapsed

**Details**

This function takes an input CDS object and collapses features within a given distance by summing the values for the collapsed features. Ranges of features are determined by their feature name, so the feature names must be in the form "chr1:1039013-2309023".

**Value**

A CDS object with aggregated peaks.

**Examples**

```
data("cicero_data")
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
agg_cds <- aggregate_nearby_peaks(input_cds, distance = 10000)
```

---

annotate\_cds\_by\_site *Add feature data columns to fData*

---

**Description**

Annotate the sites of your CDS with feature data based on coordinate overlap.

**Usage**

```
annotate_cds_by_site(
  cds,
  feature_data,
  verbose = FALSE,
  maxgap = 0,
  all = FALSE,
  header = FALSE
)
```

**Arguments**

cds	A CDS object.
feature_data	Data frame, or a character path to a file of feature data. If a path, the file should be tab separated. Default assumes no header, if your file has a header, set header = FALSE. For either a data frame or a path, the file should be in bed-like format, with the first 3 columns containing chromosome, start and stop respectively. The remaining columns will be added to the fData table as feature data.
verbose	Logical, should progress messages be printed?
maxgap	The maximum number of base pairs allowed between the peak and the feature for the feature and peak to be considered overlapping. Default = 0 (overlapping). Details in <a href="#">findOverlaps-methods</a> . If maxgap is set to "nearest" then the nearest feature will be assigned regardless of distance.
all	Logical, should all overlapping intervals be reported? If all is FALSE, the largest overlap is reported.
header	Logical, if reading a file, is there a header?

## Details

`annotate_cds_by_site` will add columns to the `fData` table of a CDS object based on the overlap of peaks with features in a data frame or file. An "overlap" column will be added, along with any columns beyond the three required columns in the feature data. The "overlap" column is the number of base pairs overlapping the `fData` site. When `maxgap` is used, the true overlap is still calculated (overlap will be 0 if the two features only overlap because of `maxgap`) NA means that there was no overlapping feature. If a peak overlaps multiple data intervals and all is FALSE, the largest overlapping interval will be chosen (in a tie, the first entry is taken), otherwise all intervals will be chosen and annotations will be collapsed using a comma as a separator.

## Value

A CDS object with updated `fData` table.

## Examples

```
data("cicero_data")
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
feat <- data.frame(chr = c("chr18", "chr18", "chr18", "chr18"),
                  bp1 = c(10000, 10800, 50000, 100000),
                  bp2 = c(10700, 11000, 60000, 110000),
                  type = c("Acetylated", "Methylated", "Acetylated",
                          "Methylated"))
input_cds <- annotate_cds_by_site(input_cds, feat)
```

---

assemble\_connections    *Combine and reconcile cicero models*

---

## Description

Function which takes the output of [generate\\_cicero\\_models](#) and assembles the connections into a data frame with cicero co-accessibility scores.

## Usage

```
assemble_connections(cicero_model_list, silent = FALSE)
```

## Arguments

`cicero_model_list`    A list of cicero output objects, generally, the output of [generate\\_cicero\\_models](#).

`silent`    Logical, should the function run silently?

## Details

This function combines glasso models computed on overlapping windows of the genome. Pairs of sites whose regularized correlation was calculated twice are first checked for qualitative concordance (both zero, positive or negative). If they not concordant, NA is returned. If they are concordant the mean is returned.

**Value**

A data frame of connections with their cicero co-accessibility scores.

**See Also**

[generate\\_cicero\\_models](#)

**Examples**

```
data("cicero_data")
data("human.hg19.genome")
sample_genome <- subset(human.hg19.genome, V1 == "chr18")
sample_genome$V2[1] <- 100000
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds, reduced_coordinates = tsne_coords)
model_output <- generate_cicero_models(cicero_cds,
                                       distance_parameter = 0.3,
                                       genomic_coords = sample_genome)
cicero_cons <- assemble_connections(model_output)
```

---

build\_gene\_activity\_matrix

*Calculate initial Cicero gene activity matrix*

---

**Description**

This function calculates the initial Cicero gene activity matrix. After this function, the activity matrix should be normalized with any comparison matrices using the function [normalize\\_gene\\_activities](#).

**Usage**

```
build_gene_activity_matrix(
  input_cds,
  cicero_cons_info,
  site_weights = NULL,
  dist_thresh = 250000,
  coaccess_cutoff = 0.25
)
```

**Arguments**

`input_cds` Binary sci-ATAC-seq input CDS. The input CDS must have a column in the fData table called "gene" which is the gene name if the site is a promoter, and NA if the site is distal.

`cicero_cons_info` Cicero connections table, generally the output of `run_cicero`. This table is a data frame with three required columns named "Peak1", "Peak2", and "coaccess". Peak1 and Peak2 contain coordinates for the two compared elements, and coaccess contains their Cicero co-accessibility score.

`site_weights` NULL or an individual weight for each site in `input_cds`.

`dist_thresh` The maximum distance in base pairs between pairs of sites to include in the gene activity calculation.

`coaccess_cutoff` The minimum Cicero co-accessibility score that should be considered connected.

**Value**

Unnormalized gene activity matrix.

**Examples**

```
data("cicero_data")
data("human.hg19.genome")
sample_genome <- subset(human.hg19.genome, V1 == "chr18")
sample_genome$V2[1] <- 100000
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- detectGenes(input_cds)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds,
                              reduced_coordinates = tsne_coords)
cons <- run_cicero(cicero_cds, sample_genome, sample_num=2)

data(gene_annotation_sample)
gene_annotation_sub <- gene_annotation_sample[,c(1:3, 8)]
names(gene_annotation_sub)[4] <- "gene"
input_cds <- annotate_cds_by_site(input_cds, gene_annotation_sub)
num_genes <- pData(input_cds)$num_genes_expressed
names(num_genes) <- row.names(pData(input_cds))
unnorm_ga <- build_gene_activity_matrix(input_cds, cons)
```

---

cell\_data

*Metadata for example cells in cicero\_data*

---

**Description**

Metadata information for `cicero_data`

**Usage**

`cell_data`

**Format**

A data frame with 200 rows and 2 variables:

**timepoint** Time at cell collection

**cell** Cell barcode

---

cicero\_data

*Example single-cell chromatin accessibility data*

---

**Description**

A dataset containing a subset of a single-cell ATAC-seq dataset collected on Human Skeletal Muscle Myoblasts. Only includes data from chromosome 18.

**Usage**

cicero\_data

**Format**

A data frame with 35137 rows and 3 variables:

**Peak** Peak information

**Cell** Cell ID

**Count** Reads per cell per peak

---

compare\_connections

*Compare Cicero connections to other datasets*

---

**Description**

Compare two sets of connections and return a vector of logicals for whether connections in one are present in the other.

**Usage**

```
compare_connections(conns1, conns2, maxgap = 0)
```

**Arguments**

**conns1** A data frame of Cicero connections, like those output from `assemble_connections`. The first two columns must be the coordinates of peaks that are connected.

**conns2** A data frame of connections to be searched for overlap. The first two columns must be coordinates of genomic sites that are connected.

**maxgap** The number of base pairs between peaks allowed to be called overlapping. See [findOverlaps-methods](#) in the IRanges package for further description.



**Value**

A vector of logicals of whether the Cicero pair is present in the alternate dataset.

**Examples**

```
## Not run:  
cons$in_dataset <- compare_connections(conns, alt_data)  
  
## End(Not run)
```

---

df_for_coords	<i>Construct a data frame of coordinate info from coordinate strings</i>
---------------	--

---

**Description**

Construct a data frame of coordinate info from coordinate strings

**Usage**

```
df_for_coords(coord_strings)
```

**Arguments**

`coord_strings` A list of coordinate strings (each like "chr1:500000-1000000")

**Details**

Coordinate strings consist of three pieces of information: chromosome, start, and stop. These pieces of information can be separated by the characters ":", "\_", or "-". Commas will be removed, not used as separators (ex: "chr18:8,575,097-8,839,855" is ok).

**Value**

data.frame with three columns, chromosome, starting base pair and ending base pair

**Examples**

```
df_for_coords(c("chr1:2,039-30,239", "chrX:28884:101293"))
```

---

```
estimate_distance_parameter
```

*Calculate distance penalty parameter*

---

## Description

Function to calculate distance penalty parameter (`distance_parameter`) for random genomic windows. Used to choose `distance_parameter` to pass to [generate\\_cicero\\_models](#).

## Usage

```
estimate_distance_parameter(
  cds,
  window = 5e+05,
  maxit = 100,
  s = 0.75,
  sample_num = 100,
  distance_constraint = 250000,
  distance_parameter_convergence = 1e-22,
  max_elements = 200,
  genomic_coords = cicero::human.hg19.genome,
  max_sample_windows = 500
)
```

## Arguments

<code>cds</code>	A cicero CDS object generated using <a href="#">make_cicero_cds</a> .
<code>window</code>	Size of the genomic window to query, in base pairs.
<code>maxit</code>	Maximum number of iterations for <code>distance_parameter</code> estimation.
<code>s</code>	Power law value. See details for more information.
<code>sample_num</code>	Number of random windows to calculate <code>distance_parameter</code> for.
<code>distance_constraint</code>	Maximum distance of expected connections. Must be smaller than <code>window</code> .
<code>distance_parameter_convergence</code>	Convergence step size for <code>distance_parameter</code> calculation.
<code>max_elements</code>	Maximum number of elements per window allowed. Prevents very large models from slowing performance.
<code>genomic_coords</code>	Either a data frame or a path (character) to a file with chromosome lengths. The file should have two columns, the first is the chromosome name (ex. "chr1") and the second is the chromosome length in base pairs. See <code>data(human.hg19.genome)</code> for an example. If a file, should be tab-separated and without header.
<code>max_sample_windows</code>	Maximum number of random windows to screen to find <code>sample_num</code> windows for distance calculation. Default 500.

## Details

The purpose of this function is to calculate the distance scaling parameter used to adjust the distance-based penalty function used in Cicero's model calculation. The scaling parameter, in combination with the power law value  $s$  determines the distance-based penalty.

This function chooses random windows of the genome and calculates a distance\_parameter. The function returns a vector of values calculated on these random windows. We recommend using the mean value of this vector moving forward with Cicero analysis.

The function works by finding the minimum distance scaling parameter such that no more than 5 distance\_constraint have non-zero entries after graphical lasso regularization and such that fewer than 80 nonzero.

If the chosen random window has fewer than 2 or greater than max\_elements sites, the window is skipped. In addition, the random window will be skipped if there are insufficient long-range comparisons (see below) to be made. The max\_elements parameter exist to prevent very dense windows from slowing the calculation. If you expect that your data may regularly have this many sites in a window, you will need to raise this parameter.

Calculating the distance\_parameter in a sample window requires peaks in that window that are at a distance greater than the distance\_constraint parameter. If there are not enough examples at high distance have been found, the function will return the warning "Warning: could not calculate sample\_num distance\_parameters - see documentation details". When looking for sample\_num example windows, the function will search max\_sample\_windows windows. By default this is set at 500, which should be well beyond the 100 windows that need to be found. However, in very sparse datasets, increasing max\_sample\_windows may help avoid the above warning. Increasing max\_sample\_windows may slow performance in sparse datasets. If you are still not able to get enough example windows, even with a large max\_sample\_windows parameter, this may mean your window parameter needs to be larger or your distance\_constraint parameter needs to be smaller. A less likely possibility is that your max\_elements parameter needs to be larger. This would occur if your data is particularly dense.

The parameter  $s$  is a constant that captures the power-law distribution of contact frequencies between different locations in the genome as a function of their linear distance. For a complete discussion of the various polymer models of DNA packed into the nucleus and of justifiable values for  $s$ , we refer readers to (Dekker et al., 2013) for a discussion of justifiable values for  $s$ . We use a value of 0.75 by default in Cicero, which corresponds to the "tension globule" polymer model of DNA (Sanborn et al., 2015). This parameter must be the same as the  $s$  parameter for generate\_cicero\_models.

Further details are available in the publication that accompanies this package. Run citation("cicero") for publication details.

## Value

A list of results of length sample\_num. List members are numeric distance\_parameter values.

## References

- Dekker, J., Marti-Renom, M.A., and Mirny, L.A. (2013). Exploring the three-dimensional organization of genomes: interpreting chromatin interaction data. Nat. Rev. Genet. 14, 390–403.
- Sanborn, A.L., Rao, S.S.P., Huang, S.-C., Durand, N.C., Huntley, M.H., Jewett, A.I., Bochkov, I.D., Chinnappan, D., Cutkosky, A., Li, J., et al. (2015). Chromatin extrusion explains key features of loop and domain formation in wild-type and engineered genomes. Proc. Natl. Acad. Sci. U. S. A. 112, E6456–E6465.

**See Also**

[generate\\_cicero\\_models](#)

**Examples**

```
data("cicero_data")
data("human.hg19.genome")
sample_genome <- subset(human.hg19.genome, V1 == "chr18")
sample_genome$V2[1] <- 100000
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds, reduced_coordinates = tsne_coords)
distance_parameters <- estimate_distance_parameter(cicero_cds,
                                                    sample_num=5,
                                                    genomic_coords = sample_genome)
```

---

find\_overlapping\_ccans

*Find CCANs that overlap each other in genomic coordinates*

---

**Description**

Find CCANs that overlap each other in genomic coordinates

**Usage**

```
find_overlapping_ccans(ccan_assignments, min_overlap = 1)
```

**Arguments**

ccan\_assignments

A data frame where the first column is the peak and the second is the CCAN assignment. For example, output of `generate_ccans`.

min\_overlap

The minimum base pair overlap to count as overlapping.

**Value**

A data frame with two columns, CCAN1 and CCAN2. CCANs in this list are overlapping. The data frame is reciprocal (if CCAN 2 overlaps CCAN 1, there will be two rows, 1,2 and 2,1).

**Examples**

```
ccan_df <- data.frame(peak = c("chr18_1408345_1408845", "chr18_1779830_1780330",
                              "chr18_1929095_1929595", "chr18_1954501_1954727",
                              "chr18_2049865_2050884", "chr18_2083726_2084102",
                              "chr18_2087935_2088622", "chr18_2104705_2105551",
                              "chr18_2108641_2108907"),
                     CCAN = c(1,2,2,2,3,3,3,3,2))
```

```
olap_ccans <- find_overlapping_ccans(ccan_df)
```

---

```
find_overlapping_coordinates
```

*Find peaks that overlap a specific genomic location*

---

### Description

Find peaks that overlap a specific genomic location

### Usage

```
find_overlapping_coordinates(coord_list, coord, maxgap = 0)
```

### Arguments

coord_list	A list of coordinates to be searched for overlap in the form chr_100_2000.
coord	The coordinates that you want to find in the form chr1_100_2000.
maxgap	The maximum distance in base pairs between coord and the coord_list that should count as overlapping. Default is 0.

### Value

A character vector of the peaks that overlap coord.

### Examples

```
test_coords <- c("chr18_10025_10225", "chr18_10603_11103",
                "chr18_11604_13986",
                "chr18_157883_158536", "chr18_217477_218555",
                "chr18_245734_246234")
find_overlapping_coordinates(test_coords, "chr18:10,100-1246234")
```

---

```
generate_ccans
```

*Generate cis-co-accessibility networks (CCANs)*

---

### Description

Post process cicero co-accessibility scores to extract modules of sites that are co-accessible.

### Usage

```
generate_ccans(
  connections_df,
  coaccess_cutoff_override = NULL,
  tolerance_digits = 2
)
```

**Arguments**

`connections_df` Data frame of connections with columns: Peak1, Peak2, coaccess. Generally, the output of `run_cicero` or `assemble_connections`

`coaccess_cutoff_override`  
Numeric, co-accessibility score threshold to impose. Overrides automatic calculation.

`tolerance_digits`  
The number of digits to calculate cutoff to. Default is 2 (0.01 tolerance)

**Details**

CCANs are calculated by first specifying a minimum co-accessibility score and then using the Louvain community detection algorithm on the subgraph induced by excluding edges below this score. For this function, either the user can specify the minimum co-accessibility using `coaccess_cutoff_override`, or the cutoff can be calculated automatically by optimizing for CCAN number. The cutoff calculation can be slow, so users may wish to use the `coaccess_cutoff_override` after initially calculating the cutoff to speed future runs.

**Value**

Data frame with two columns - Peak and CCAN. CCAN column indicates CCAN assignment. Peaks not included in a CCAN are not returned.

**Examples**

```
## Not run:
data("cicero_data")
set.seed(18)
data("human.hg19.genome")
sample_genome <- subset(human.hg19.genome, V1 == "chr18")
sample_genome$V2[1] <- 100000
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds, reduced_coordinates = tsne_coords)
cicero_cons <- run_cicero(cicero_cds, sample_genome, sample_num = 2)
ccan_assigns <- generate_ccans(cicero_cons)

## End(Not run)
```

---

generate\_cicero\_models

*Generate cicero models*

---

**Description**

Function to generate graphical lasso models on all sites in a CDS object within overlapping genomic windows.

**Usage**

```
generate_cicero_models(
  cds,
  distance_parameter,
  s = 0.75,
  window = 5e+05,
  max_elements = 200,
  genomic_coords = cicero::human.hg19.genome
)
```

**Arguments**

<code>cds</code>	A cicero CDS object generated using <a href="#">make_cicero_cds</a> .
<code>distance_parameter</code>	Distance based penalty parameter value. Generally, the mean of the calculated <code>distance_parameter</code> values from <a href="#">estimate_distance_parameter</a> .
<code>s</code>	Power law value. See details.
<code>window</code>	Size of the genomic window to query, in base pairs.
<code>max_elements</code>	Maximum number of elements per window allowed. Prevents very large models from slowing performance.
<code>genomic_coords</code>	Either a data frame or a path (character) to a file with chromosome lengths. The file should have two columns, the first is the chromosome name (ex. "chr1") and the second is the chromosome length in base pairs. See <code>data(human.hg19.genome)</code> for an example. If a file, should be tab-separated and without header.

**Details**

The purpose of this function is to compute the raw covariances between each pair of sites within overlapping windows of the genome. Within each window, the function then estimates a regularized correlation matrix using the graphical LASSO (Friedman et al., 2008), penalizing pairs of distant sites more than proximal sites. The scaling parameter, `distance_parameter`, in combination with the power law value `s` determines the distance-based penalty.

The parameter `s` is a constant that captures the power-law distribution of contact frequencies between different locations in the genome as a function of their linear distance. For a complete discussion of the various polymer models of DNA packed into the nucleus and of justifiable values for `s`, we refer readers to (Dekker et al., 2013) for a discussion of justifiable values for `s`. We use a value of 0.75 by default in Cicero, which corresponds to the “tension globule” polymer model of DNA (Sanborn et al., 2015). This parameter must be the same as the `s` parameter for [estimate\\_distance\\_parameter](#).

Further details are available in the publication that accompanies this package. Run `citation("cicero")` for publication details.

**Value**

A list of results for each window. Either a `glasso` object, or a character description of why the window was skipped. This list can be directly input into [assemble\\_connections](#) to create a reconciled list of cicero co-accessibility scores.

## References

- Dekker, J., Marti-Renom, M.A., and Mirny, L.A. (2013). Exploring the three-dimensional organization of genomes: interpreting chromatin interaction data. *Nat. Rev. Genet.* 14, 390–403.
- Friedman, J., Hastie, T., and Tibshirani, R. (2008). Sparse inverse covariance estimation with the graphical lasso. *Biostatistics* 9, 432–441.
- Sanborn, A.L., Rao, S.S.P., Huang, S.-C., Durand, N.C., Huntley, M.H., Jewett, A.I., Bochkov, I.D., Chinnappan, D., Cutkosky, A., Li, J., et al. (2015). Chromatin extrusion explains key features of loop and domain formation in wild-type and engineered genomes. *Proc. Natl. Acad. Sci. U. S. A.* 112, E6456–E6465.

## See Also

[estimate\\_distance\\_parameter](#)

## Examples

```
data("cicero_data")
data("human.hg19.genome")
sample_genome <- subset(human.hg19.genome, V1 == "chr18")
sample_genome$V2[1] <- 100000
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds, reduced_coordinates = tsne_coords)
model_output <- generate_cicero_models(cicero_cds,
                                       distance_parameter = 0.3,
                                       genomic_coords = sample_genome)
```

---

gene\_annotation\_sample

*Example gene annotation information*

---

## Description

Gencode gene annotation data from chromosome 18 of the human genome (hg19).

## Usage

```
gene_annotation_sample
```

## Format

A data frame with 15129 rows and 8 variables:

**chromosome** Chromosome

**start** Exon starting base



**end** Exon ending base  
**strand** Exon mapping direction  
**feature** Feature type  
**gene** Gene ID  
**transcript** Transcript ID  
**symbol** Gene symbol

---

human.hg19.genome      *Chromosome lengths from human genome hg19*

---

### Description

A list of the chromosomes in hg19 and their lengths in base pairs.

### Usage

```
human.hg19.genome
```

### Format

A data frame with 93 rows and 2 variables:

**V1** Chromosome  
**V2** Chromosome length, base pairs

---

make\_atac\_cds      *Make ATAC CDS object*

---

### Description

This function takes as input a data frame or a path to a file in a sparse matrix format and returns a properly formatted CellDataSet (CDS) object.

### Usage

```
make_atac_cds(input, binarize = FALSE)
```

### Arguments

**input**            Either a data frame or a path to input data. If a file, it should be a tab-delimited text file with three columns and no header. For either a file or a data frame, the first column is the peak coordinates in the form "chr10\_100013372\_100013596", the second column is the cell name, and the third column is an integer that represents the number of reads from that cell overlapping that peak. Zero values do not need to be included (sparse matrix format).

**binarize**        Logical. Should the count matrix be converted to binary?

**Value**

A CDS object containing your ATAC data in proper format.

**Examples**

```
data("cicero_data")
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
```

---

make_cicero_cds	<i>Create cicero input CDS</i>
-----------------	--------------------------------

---

**Description**

Function to generate an aggregated input CDS for cicero. `run_cicero` takes as input an aggregated cicero CDS object. This function will generate the CDS given an input CDS (perhaps generated by `make_atac_cds`) and a value for `k`, which is the number of cells to be aggregated per bin. The default value for `k` is 50.

**Usage**

```
make_cicero_cds(
  cds,
  reduced_coordinates,
  k = 50,
  summary_stats = NULL,
  size_factor_normalize = TRUE,
  silent = FALSE,
  return_agg_info = FALSE
)
```

**Arguments**

<code>cds</code>	Input CDS object.
<code>reduced_coordinates</code>	A data frame with columns representing the coordinates of each cell in reduced dimension space (generally 2-3 dimensions). <code>row.names(reduced_coordinates)</code> should match the cell names in the CDS object. If dimension reduction was done using <code>monocle</code> , tSNE coordinates can be accessed by <code>t(reducedDimA(cds))</code> , and <code>DDRTree</code> coordinates can be accessed by <code>t(reducedDimS(cds))</code> .
<code>k</code>	Number of cells to aggregate per bin.
<code>summary_stats</code>	Which numeric <code>pData(cds)</code> columns you would like summarized (mean) by bin in the resulting CDS object.
<code>size_factor_normalize</code>	Logical, should accessibility values be normalized by size factor?
<code>silent</code>	Logical, should warning and info messages be printed?
<code>return_agg_info</code>	Logical, should a list of the assignments of cells to aggregated bins be output? When <code>TRUE</code> , this function returns a list of two items, first, the aggregated CDS object and second, a data.frame with the binning information.

**Details**

Aggregation of similar cells is done using a k-nearest-neighbors graph and a randomized "bagging" procedure. Details are available in the publication that accompanies this package. Run `citation("cicero")` for publication details. KNN is calculated using [knn.index](#)

**Value**

Aggregated CDS object. If `return_agg_info` is TRUE, a list of the aggregated CDS object and a data.frame of aggregation info.

**Examples**

```
## Not run:
data("cicero_data")

input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds, reduced_coordinates = tsne_coords)

## End(Not run)
```

---

make\_sparse\_matrix      *Make a symmetric square sparse matrix from data frame*

---

**Description**

Convert a data frame into a square sparse matrix (all versus all)

**Usage**

```
make_sparse_matrix(data, i.name = "Peak1", j.name = "Peak2", x.name = "value")
```

**Arguments**

<code>data</code>	data frame
<code>i.name</code>	name of i column
<code>j.name</code>	name of j column
<code>x.name</code>	name of value column

**Value**

sparse matrix

---

```
normalize_gene_activities
```

*Normalize gene activities*

---

### Description

Normalize the output of `build_gene_activity_matrix`. Input is either one or multiple gene activity matrices. Any gene activities to be compared amongst each other should be normalized together.

### Usage

```
normalize_gene_activities(activity_matrices, cell_num_genes)
```

### Arguments

`activity_matrices`

A gene activity matrix, output from `build_gene_activity_matrix`, or a list of gene activity matrices to be normalized together.

`cell_num_genes`

A named vector of the total number of accessible sites per cell. Names should correspond to the cell names in the activity matrices. These values can be found in the "num\_genes\_expressed" column of the pData table of the CDS used to calculate the gene activity matrix.

### Value

Normalized activity matrix or matrices.

### Examples

```
data("cicero_data")
data("human.hg19.genome")
sample_genome <- subset(human.hg19.genome, V1 == "chr18")
sample_genome$V2[1] <- 100000
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- detectGenes(input_cds)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds,
                             reduced_coordinates = tsne_coords)
cons <- run_cicero(cicero_cds, sample_genome, sample_num=2)

data(gene_annotation_sample)
gene_annotation_sub <- gene_annotation_sample[,c(1:3, 8)]
names(gene_annotation_sub)[4] <- "gene"
input_cds <- annotate_cds_by_site(input_cds, gene_annotation_sub)
num_genes <- pData(input_cds)$num_genes_expressed
names(num_genes) <- row.names(pData(input_cds))
unnorm_ga <- build_gene_activity_matrix(input_cds, cons)
cicero_gene_activities <- normalize_gene_activities(unnorm_ga, num_genes)
```

---

plot\_accessibility\_in\_pseudotime  
*Plot accessibility by pseudotime*

---

### Description

Make a barplot of chromatin accessibility across pseudotime

### Usage

```
plot_accessibility_in_pseudotime(cds_subset, breaks = 10)
```

### Arguments

cds_subset	Subset of the CDS object you want to plot. The CDS must have a column in the pData table called "Pseudotime".
breaks	Number of breaks along pseudotime. Controls the coarseness of the plot.

### Details

This function plots each site in the CDS subset by cell pseudotime as a barplot. Cells are divided into bins by pseudotime (number determined by breaks) and the percent of cells in each bin that are accessible is represented by bar height. In addition, the black line represents the pseudotime-dependent average accessibility from a smoothed binomial regression.

### Value

ggplot object

### Examples

```
## Not run:
plot_accessibility_in_pseudotime(input_cds_lin[c("chr18_38156577_38158261",
                                                "chr18_48373358_48374180",
                                                "chr18_60457956_60459080")])

## End(Not run)
```

---

plot\_connections      *Plot connections*

---

### Description

Plotting function for Cicero connections. Uses [plotTracks](#) as its basis

**Usage**

```

plot_connections(
  connection_df,
  chr,
  minbp,
  maxbp,
  coaccess_cutoff = 0,
  peak_color = "#B4656F",
  connection_color = "#7F7CAF",
  connection_color_legend = TRUE,
  alpha_by_coaccess = FALSE,
  connection_width = 2,
  connection_ymax = NULL,
  gene_model = NULL,
  gene_model_color = "#81D2C7",
  gene_model_shape = c("smallArrow", "box"),
  comparison_track = NULL,
  comparison_coaccess_cutoff = 0,
  comparison_peak_color = "#B4656F",
  comparison_connection_color = "#7F7CAF",
  comparison_connection_color_legend = TRUE,
  comparison_connection_width = 2,
  comparison_ymax = NULL,
  collapseTranscripts = FALSE,
  include_axis_track = TRUE,
  return_as_list = FALSE,
  viewpoint = NULL,
  comparison_viewpoint = TRUE,
  viewpoint_color = "#F0544F",
  viewpoint_fill = "#EFD8D7",
  viewpoint_alpha = 0.5
)

```

**Arguments**

connection_df	Data frame of connections, which must include the columns 'Peak1', 'Peak2', and 'coaccess'. Generally, the output of run_cicero or assemble_connections.
chr	The chromosome of the region you would like to plot in the form 'chr10'.
minbp	The base pair coordinate of the start of the region to be plotted.
maxbp	The base pair coordinate of the end of the region to be plotted.
coaccess_cutoff	The minimum cicero co-accessibility score you would like to be plotted. Default is 0.
peak_color	Color for peak annotations - a single color, the name of a column containing color values that correspond to Peak1, or the name of column containing a character or factor to base peak colors on.
connection_color	Color for connection lines. A single color, the name of a column containing color values, or the name of a column containing a character or factor to base connection colors on.

connection_color_legend	Logical, should connection color legend be shown?
alpha_by_coaccess	Logical, should the transparency of connection lines be scaled based on co-accessibility score?
connection_width	Width of connection lines.
connection_ymax	Connection y-axis height. If NULL, chosen automatically.
gene_model	Either NULL or a data.frame. The data.frame should be in a form compatible with the Gviz function <a href="#">GeneRegionTrack-class</a> (cannot have NA as column names).
gene_model_color	Color for gene annotations.
gene_model_shape	Shape for gene models, passed to <a href="#">GeneRegionTrack-class</a> . Options described at <a href="#">GeneRegionTrack-class</a> .
comparison_track	Either NULL or a data frame. If a data frame, a second track of connections will be plotted based on this data. This data frame has the same requirements as connection_df (Peak1, Peak2 and coaccess columns).
comparison_coaccess_cutoff	The minimum cicero co-accessibility score you would like to be plotted for the comparison dataset. Default = 0.
comparison_peak_color	Color for comparison peak annotations - a single color, the name of a column containing color values that correspond to Peak1, or the name of a column containing a character or factor to base peak colors on.
comparison_connection_color	Color for comparison connection lines. A single color, the name of a column containing color values, or the name of a column containing a character or factor to base connection colors on.
comparison_connection_color_legend	Logical, should comparison connection color legend be shown?
comparison_connection_width	Width of comparison connection lines.
comparison_ymax	Connection y-axis height for comparison track. If NULL, chosen automatically.
collapseTranscripts	Logical or character scalar. Can be one in gene, longest, shortest or meta. Variable is passed to the <a href="#">GeneRegionTrack-class</a> function of Gviz. Determines whether and how to collapse related transcripts. See Gviz documentation for details.
include_axis_track	Logical, should a genomic axis be plotted?
return_as_list	Logical, if TRUE, the function will not plot, but will return the plot components as a list. Allows user to add/customize Gviz components and plot them separately using <a href="#">plotTracks</a> .

viewpoint NULL or Coordinates in form "chr1\_10000\_10020". Use viewpoint if you would like to plot cicero connections "4C-seq style". Only connections originating in the viewpoint will be shown. Ideal for comparisons with 4C-seq data. If comparison\_viewpoint is TRUE, any comparison track will be subsetted as well.

comparison\_viewpoint Logical, should viewpoint apply to comparison track as well?

viewpoint\_color Color for the highlight border.

viewpoint\_fill Color for the highlight fill.

viewpoint\_alpha Alpha value for the highlight fill.

### Value

A gene region plot, or list of components if return\_as\_list is TRUE.

### Examples

```
cicero_cons <- data.frame(
  Peak1 = c("chr18_10034652_10034983", "chr18_10034652_10034983",
            "chr18_10034652_10034983", "chr18_10034652_10034983",
            "chr18_10087586_10087901", "chr18_10120685_10127115",
            "chr18_10097718_10097934", "chr18_10087586_10087901",
            "chr18_10154818_10155215", "chr18_10238762_10238983",
            "chr18_10198959_10199183", "chr18_10250985_10251585"),
  Peak2 = c("chr18_10097718_10097934", "chr18_10087586_10087901",
            "chr18_10154818_10155215", "chr18_10238762_10238983",
            "chr18_10198959_10199183", "chr18_10250985_10251585",
            "chr18_10034652_10034983", "chr18_10034652_10034983",
            "chr18_10034652_10034983", "chr18_10034652_10034983",
            "chr18_10087586_10087901", "chr18_10120685_10127115"),
  coaccess = c(0.0051121787, 0.0016698617, 0.0006570246,
              0.0013466927, 0.0737935011, 0.3264019452,
              0.0051121787, 0.0016698617, 0.0006570246,
              0.0013466927, 0.0737935011, 0.3264019452))
plot_connections(cicero_cons, chr = "chr18",
  minbp = 10034652,
  maxbp = 10251585,
  peak_color = "purple")
```

---

ranges\_for\_coords      *Construct GRanges objects from coordinate strings*

---

### Description

Construct GRanges objects from coordinate strings

### Usage

```
ranges_for_coords(coord_strings, meta_data_df = NULL, with_names = FALSE)
```



**Arguments**

- coord\_strings A list of coordinate strings (in the form "chr1:500000-1000000")
- meta\_data\_df A data frame with any meta data columns you want included with the ranges. Must be in the same order as coord\_strings.
- with\_names logical - should meta data include coordinate string (field coord\_string)?

**Details**

Coordinate strings consist of three pieces of information: chromosome, start, and stop. These pieces of information can be separated by the characters ":", "\_", or "-". Commas will be removed, not used as separators (ex: "chr18:8,575,097-8,839,855" is ok).

**Value**

GRanges object of the input strings

**See Also**

[GRanges-class](#)

**Examples**

```
ran1 <- ranges_for_coords("chr1:2039-30239", with_names = TRUE)
ran2 <- ranges_for_coords(c("chr1:2049-203902", "chrX:489249-1389389"),
  meta_data_df = data.frame(dat = c("1", "X")))
ran3 <- ranges_for_coords(c("chr1:2049-203902", "chrX:489249-1389389"),
  with_names = TRUE,
  meta_data_df = data.frame(dat = c("1", "X"),
    stringsAsFactors = FALSE))
```

---

run\_cicero

*Run Cicero*

---

**Description**

A wrapper function that runs the primary functions of the Cicero pipeline with default parameters. Runs [estimate\\_distance\\_parameter](#), [generate\\_cicero\\_models](#) and [assemble\\_connections](#). See the manual pages of these functions for details about their function and parameter options. Defaults in this function are designed for mammalian data, those with non-mammalian data should read about parameters in the above functions.

**Usage**

```
run_cicero(
  cds,
  genomic_coords,
  window = 5e+05,
  silent = FALSE,
  sample_num = 100
)
```

## Arguments

<code>cds</code>	Cicero CDS object, created using <code>make_cicero_cds</code>
<code>genomic_coords</code>	Either a data frame or a path (character) to a file with chromosome lengths. The file should have two columns, the first is the chromosome name (ex. "chr1") and the second is the chromosome length in base pairs. See <code>data(human.hg19.genome)</code> for an example. If a file, should be tab-separated and without header.
<code>window</code>	Size of the genomic window to query, in base pairs.
<code>silent</code>	Whether to print progress messages
<code>sample_num</code>	How many sample genomic windows to use to generate <code>distance_parameter</code> parameter. Default: 100.

## Value

A table of co-accessibility scores

## Examples

```
data("cicero_data")
data("human.hg19.genome")
sample_genome <- subset(human.hg19.genome, V1 == "chr18")
sample_genome$V2[1] <- 100000
input_cds <- make_atac_cds(cicero_data, binarize = TRUE)
input_cds <- reduceDimension(input_cds, max_components = 2, num_dim=6,
                             reduction_method = 'tSNE',
                             norm_method = "none")
tsne_coords <- t(reducedDimA(input_cds))
row.names(tsne_coords) <- row.names(pData(input_cds))
cicero_cds <- make_cicero_cds(input_cds, reduced_coordinates = tsne_coords)
cons <- run_cicero(cicero_cds, sample_genome, sample_num = 2)
```

# Index

## \* datasets

- cell\_data, [7](#)
- cicero\_data, [8](#)
- gene\_annotation\_sample, [16](#)
- human.hg19.genome, [17](#)

- aggregate\_by\_cell\_bin, [3](#)
- aggregate\_nearby\_peaks, [3](#)
- annotate\_cds\_by\_site, [4](#)
- assemble\_connections, [5](#), [14](#), [15](#), [25](#)

- build\_gene\_activity\_matrix, [6](#), [20](#)

- cell\_data, [7](#)
- cicero (cicero-package), [2](#)
- cicero-package, [2](#)
- cicero\_data, [8](#)
- compare\_connections, [8](#)

- df\_for\_coords, [9](#)

- estimate\_distance\_parameter, [10](#), [15](#), [16](#),  
[25](#)

- find\_overlapping\_ccans, [12](#)
- find\_overlapping\_coordinates, [13](#)

- gene\_annotation\_sample, [16](#)
- generate\_ccans, [13](#)
- generate\_cicero\_models, [5](#), [6](#), [10](#), [12](#), [14](#), [25](#)

- human.hg19.genome, [17](#)

- knn.index, [19](#)

- make\_atac\_cds, [3](#), [17](#)
- make\_cicero\_cds, [10](#), [15](#), [18](#), [26](#)
- make\_sparse\_matrix, [19](#)

- normalize\_gene\_activities, [6](#), [20](#)

- plot\_accessibility\_in\_pseudotime, [21](#)
- plot\_connections, [21](#)
- plotTracks, [21](#), [23](#)

- ranges\_for\_coords, [24](#)
- run\_cicero, [7](#), [14](#), [25](#)