
NucleoMiner2 Documentation

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README / DOCUMENTATION FOR *NUCLEOMINER2*

NucleoMiner2 offers Python API and R package allowing to perform quantitative analysis of epigenetic marks on individual nucleosomes. It was developed to detect natural Single-Nucleosome Epi-Polymorphisms (SNEP) from MNase-seq and ChIP-seq data.

1.1 License

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- Gael YVERT

This software is a computer program which purpose is to perform quantitative analysis of epigenetic marks at single nucleosome resolution.

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1.2 Installation Instructions

1.2.1 Links

NucleoMiner2 home page and documentation are available here:

- <https://forge.cbp.ens-lyon.fr/redmine/projects/nucleominer>

The Yvert lab web page is accessible here:

- <http://www.ens-lyon.fr/LBMC/gisv/>

1.2.2 Installation

The first installation step is to retrieve the source code of *NucleoMiner2*. You can do this by typing the following command in a terminal.

```
git clone http://forge.cbp.ens-lyon.fr/git/nucleominer
```

Prerequisites

To work properly, *NucleoMiner2* needs that the following free software are installed and made available on your system:

- Bowtie2 <http://bowtie-bio.sourceforge.net/bowtie2>
- SAMtools <http://samtools.sourceforge.net>
- bedtools <http://code.google.com/p/bedtools/>
- TemplateFilter <http://compbio.cs.huji.ac.il/NucPosition/TemplateFiltering>

It also requires the following R packages to be installed on your system:

- fork
- rjson
- seqinr
- plotrix
- DESeq

These packages can be installed by typing the following command in an R console:

```
install.packages(c("fork", "rjson", "seqinr", "plotrix"))
source("http://bioconductor.org/biocLite.R")
biocLite("DESeq")
```

Finally, by typing the git command above, you downloaded specific R packages provided with *NucleoMiner2* that you now need to install:

- cachecache <https://forge.cbp.ens-lyon.fr/redmine/projects/cachecache>
- bot <https://forge.cbp.ens-lyon.fr/redmine/projects/bot>
- nucleominer <https://forge.cbp.ens-lyon.fr/redmine/projects/nucleominer>

To do so, type the following command in your terminal:

```
cd nucleominer
R CMD INSTALL doc/Chuffart_NM2_workdir/deps/bot_0.14.tar.gz\
  doc/Chuffart_NM2_workdir/deps/cachecache_0.1.tar.gz\
  build/nucleominer_2.3.46.tar.gz
```


TUTORIAL

This tutorial describes steps allowing to perform quantitative analysis of epigenetic marks on individual nucleosomes. We assume that files are organised according to a given hierarchy and that all command lines are launched from the project's root directory.

This tutorial is divided into two main parts. The first part covers the python script *wf.py* that aligns and converts short sequence reads. The second part covers the R scripts that extracts nucleosome-level information (nucleosome position and indicators) from the dataset.

2.1 Experimental Dataset, Working Directory and Configuration File

2.1.1 Working Directory Organisation

After having installed NucleoMiner2 environment (Previous section), go to the root working directory of the tutorial by typing the following command in a terminal:

```
cd doc/Chuffart_NM2_workdir/
```

2.1.2 Retrieving Experimental Dataset

The MNase-seq and MN-ChIP-seq raw data are available at ArrayExpress (<http://www.ebi.ac.uk/arrayexpress/>) under accession number E-MTAB-2671.

\$\$\$ TODO explain how organise Experimental Dataset into the *data* directory of the working directory.

In this tutorial, we want to compare nucleosomes of 2 yeast strains: BY and RM. For each strain Mnase-Seq was performed as well as ChIP-Seq using an antibody recognizing the H3K14ac epigenetic mark. Illumina sequencing was done in single-read of 50 bp long.

The dataset is composed of 55 files organised as follows:

- 3 replicates for BY MNase Seq
 - sample 1 (5 fastq.gz files)
 - sample 2 (5 fastq.gz files)
 - sample 3 (4 fastq.gz files)
- 3 replicates for RM MNase Seq
 - sample 4 (4 fastq.gz files)
 - sample 5 (4 fastq.gz files)

- sample 6 (5 fastq.gz files)
- 3 replicates for BY ChIP Seq H3K14ac
 - sample 36 (5 fastq.gz files)
 - sample 37 (5 fastq.gz files)
 - sample 53 (9 fastq.gz files)
- 2 replicates for RM ChIP Seq H3K14ac
 - sample 38 (5 fastq.gz files)
 - sample 39 (4 fastq.gz files)

2.1.3 Python and R Common Configuration File

First, we need to define useful configuration variables that will be passed to python and R scripts. These variables are contained in file *configurator.py*. The execution of this python script dumps variables into the *nucleominer_config.json* file that will then be used by both R and python scripts.

The initialization of this variables is done in the *configurator.py* file. If you need to adapt variable values (path, default parameters...) you need to edit this file. Then, go to the root directory of your project and run the following command to dump the configuration file:

```
python src/current/configurator.py
```

2.2 Preprocessing Illumina Fastq Reads for Each Sample

Once variables and design have been specified, the script *wf.py* will automatically run all the analysis. You don't need to do anything. To run the full analysis, run the following command:

```
python src/current/wf.py
```

The details of the steps performed by this script are explained below. This preprocessing consists of 4 steps embedded in the *wf.py* script. They are described bellow. As a preamble, this script computes *samples*, *samples_mnase* and *strains* that will be used along the 4 steps.

```
wf.samples = []
```

List of samples where a sample is identified by an id (key: *id*) and a strain name (key *strain*).

```
wf.samples_mnase = []
```

List of Mnase samples.

```
wf.strains = []
```

List of reference strains.

2.2.1 Creating Bowtie Index from each Reference Genome

For each strain, the script *wf.py* then creates bowtie index. Bowtie index of a strain is a tree view of the genome of this strain. It will be used by bowtie to align reads. The part of the script performing this is the following:

```
for strain in strains:
    per_strain_stats[strain] = create_bowtie_index(strain,
        config["FASTA_REFERENCE_GENOME_FILES"][strain], config["INDEX_DIR"],
        config["BOWTIE_BUILD_BIN"])
```

As an indication, the following table summarizes the file sizes and process durations that we experienced when running this step on a Linux server***.

strain	fasta genome file size	bowtie index file size	process duration
BY	12 Mo	25 Mo	11 s.
RM	12 Mo	24 Mo	9 s.

2.2.2 Aligning Reads to Reference Genome

Next, the *wf.py* script launches bowtie to align reads to the reference genome. It produces a *.sam* file that is converted into a *.bed* file. Binaries for *bowtie*, *samtools* and *bedtools* are wrapped using python *subprocess* class. This step is performed by the following part of the script:

```
for sample in samples:
    per_sample_align_stats["sample_%s" % sample["id"]] = align_reads(sample,
        config["ALIGN_DIR"], config["LOG_DIR"], config["INDEX_DIR"],
        config["ILLUMINA_OUTPUTFILE_PREFIX"], config["BOWTIE2_BIN"],
        config["SAMTOOLS_BIN"], config["BEDTOOLS_BIN"])
```

2.2.3 Convert Aligned Reads into TemplateFilter Format

TemplateFilter uses particular input formats for reads, so it is necessary to convert the *.bed* files. TemplateFilter expect reads in the following format: *chr, coord, strand* and *#read* where:

- *chr* is the number of the chromosome;
- *coord* is the coordinate of the reads;
- *strand* is *F* for forward and *R* for reverse;
- *#reads* the number of reads covering this position.

Each entry is *tab*-separated.

WARNING for reverse strands, bowtie returns the position of the first nucleotide on the left hand side, whereas TemplateFilter expects the first one on the right hand side. This is taken into account in NucleoMiner2 by adding the read length (in our case 50) to the reverse reads coordinates.

This step is performed by the following part of the *wf.py* script:

```
for sample in samples:
    per_sample_convert_stats["sample_%s" % sample["id"]] = split_fr_4_TF(sample,
        config["ALIGN_DIR"], config["FASTA_INDEXES"], config["AREA_BLACK_LIST"],
        config["READ_LENGTH"], config["MAPQ_THRES"])
```

The following table summarizes the number of reads, the involved file sizes and process durations that we experienced when running the two last steps. In our case, alignment process were multithreaded over 3 cores.

id	Illumina reads	aligned and filtered reads	ratio	.bed file size	TF input file size	process duration
1	16436138	10199695	62,06%	1064 Mo	60 Mo	383 s.
2	16911132	12512727	73,99%	1298 Mo	64 Mo	437 s.
3	15946902	12340426	77,38%	1280 Mo	65 Mo	423 s.
4	13765584	10381903	75,42%	931 Mo	59 Mo	352 s.
5	15168268	11502855	75,83%	1031 Mo	64 Mo	386 s.
6	18850820	14024905	74,40%	1254 Mo	69 Mo	482 s.
36	17715118	14092985	79,55%	1404 Mo	68 Mo	483 s.
37	17288466	7402082	42,82%	741 Mo	48 Mo	339 s.
38	16116394	13178457	81,77%	1101 Mo	63 Mo	420 s.
39	14241106	10537228	73,99%	880 Mo	57 Mo	348 s.
53	40876476	33780065	82,64%	3316 Mo	103 Mo	1165 s.

2.2.4 Run TemplateFilter on Mnase Samples

Finally, for each sample we perform TemplateFilter analysis.

WARNING TemplateFilter returns a list of nucleosomes. Each nucleosome is defined by its center and its width. An odd width leads us to consider non- integer lower and upper bound.

WARNING TemplateFilter was not designed to handle replicates. So we recommend to keep a maximum of nucleosomes and filter the aberrant ones afterwards using the benefits of having replicates. To do this, we set a low correlation threshold parameter (0.5) and a particularly high value of overlap (300%).

This step is performed by the following part of the *wf.py* script:

```
for sample in samples_mnase:
    per_mnase_sample_stats["sample_%s" % sample["id"]] = template_filter(sample,
        config["ALIGN_DIR"], config["LOG_DIR"], config["TF_BIN"],
        config["TF_TEMPLATES_FILE"], config["TF_CORR"], config["TF_MINW"],
        config["TF_MAXW"], config["TF_OL"])
```

id	strain	found nucs	nuc file size	process duration
1	BY	96214	68 Mo	1022 s.
2	BY	91694	65 Mo	1038 s.
3	BY	91205	65 Mo	1036 s.
4	RM	88076	62 Mo	984 s.
5	RM	90141	64 Mo	967 s.
6	RM	87517	62 Mo	980 s.

2.3 Inferring Nucleosome Position and Extracting Read Counts

The second part of the tutorial uses R (<http://http://www.r-project.org>). NucleoMiner2 contains a set of R scripts that will be sourced in R from a console launched at the root of your project. These scripts are:

- headers.R
- extract_maps.R
- translate_common_wp.R
- split_samples.R
- count_reads.R
- get_size_factors

- `launch_deseq.R`

2.3.1 The Script headers.R

The script `headers.R` is included in all other R scripts. It is in charge of:

- launching libraries used in the scripts
- launching configuration (design, strain, marker...)
- computing and caching Common Uninterrupted Regions (CURs). Caching means storing the information in the computer's memory.

Note that you can customize the function “translate”. This function allows you to use the alignments between genomes when performing various tasks.

- You may want to analyze data of a single strain (e.g. treatment/control, or only few mutations). In this case, the genome is identical across all samples and you do not need to define particular CURs (CURs are chromosomes). Simply use the default translate function which is neutral.
- If you are analyzing data from two or more strains (as NucleoMiner2 was designed for), then you need to translate coordinates of one genome into the coordinates of another one. You must do this by aligning the two genomes, which will produce a `.c2c` file (see Appendice “Generate `.c2c` Files”). then use it to produce the list of regions and customise “translate”.

In our tutorial, we are in the second case and to perform all these steps run the following command line in your R console:

```
source("src/current/headers.R")
```

2.3.2 The Script `extract_maps.R`

This script is in charge of extracting Maps for well-positioned and sensitive nucleosomes. First of all, this script computes intra and inter-strain matches of nucleosome maps for each CUR. This step can be executed in parallel on many cores using the BoT library. Next, it collects results and produces maps of well-positioned nucleosomes, sensitive nucleosomes and Unaligned Nucleosomal Regions .

The map of well-positioned nucleosomes for BY is collected in the result directory and is called *BY_wp.tab*. It is composed of following columns:

- `chr`, the number of the chromosome
- `lower_bound`, the lower bound of the nucleosome
- `upper_bound`, the upper bound of the nucleosome
- `cur_index`, index of the CUR
- `index_nuc`, the index of the nucleosome in the CUR
- `wp`, 1 if it is a well positioned nucleosome, 0 otherwise
- `nb_reads`, the number of reads that support this nucleosome
- `nb_nucs`, the number of TemplateFilter nucleosome across replicates (= the number of replicates in which it is a well-positioned nucleosome)
- `llr_1`, for a well-positioned nucleosome, it is the LLR1 (log-likelihood ratio) between the first and the second TemplateFilter nucleosome on the chain.

- `llr_2`, for a well-positioned nucleosome, it is the LLR1 between the second and the third TemplateFilter nucleosome on the chain.
- `wp_llr`, for a well-positioned nucleosome, it is the LLR2 that compares consistency of the positioning over all TemplateFilter nucleosomes.
- `wp_pval`, for a well-positioned nucleosome, it is the p-value chi square test obtained from LLR2 ($1-pchisq(2.LLR2, df=4)$)
- `dyad_shift`, for a well-positioned nucleosome, it is the shift between the two extreme TemplateFilter nucleosome dyad positions.

The sensitive map for BY is collected in the result directory and is called *BY_fuzzy.tab*. It is composed of following columns:

- `chr`, the number of the chromosome
- `lower_bound`, the lower bound of the nucleosome
- `upper_bound`, the upper bound of the nucleosome
- `cur_index`, index of the CUR

The map of common well-positioned nucleosomes aligned between the BY and RM strains is collected in the result directory and is called *BY_RM_common_wp.tab*. It is composed of following columns:

- `cur_index`, the index of the CUR
- `index_nuc_BY`, the index of the BY nucleosome in the CUR
- `index_nuc_RM`, the index of the RM nucleosome in the CUR
- `llr_score`, , the LLR3 score that estimates conservation between the positions in BY and RM
- `common_wp_pval`, the p-value chi square test obtained from LLR3 ($1-pchisq(2.LLR3, df=2)$)
- `diff`, the dyads shift between the positions in the two strains (in bp)

The common UNR map for BY and RM strains is collected in the result directory and is called *BY_RM_common_unr.tab*. It is composed of the following columns:

- `cur_index`, the index of the CUR
- `index_nuc_BY`, the index of the BY nucleosome in the CUR
- `index_nuc_RM`, the index of the RM nucleosome in the CUR

To execute this script, run the following command in your R console:

```
source("src/current/extract_maps.R")
```

2.3.3 The Script `translate_common_wp.R`

This script is used to translate common well-positioned nucleosome positions from a strain to another strain and stores it into a table.

For example, the file *results/2014-04/RM_wp_tr_2_BY.tab* contains RM well-positioned nucleosomes translated into the BY genome coordinates. It is composed of following columns:

- `strain_ref`, the reference genome (in which positioned are defined)
- `begin`, the translated lower bound of the nucleosome
- `end`, the translated upper bound of the nucleosome
- `chr`, the number of chromosomes for the reference genome (in which positioned are defined)

- `length`, the length of the nucleosome (could be negative)
- `cur_index`, the index of the CUR
- `index_nuc`, the index of the nucleosome in the CUR

To execute this script, run the following command in your R console:

```
source("src/current/translate_common_wp.R")
```

2.3.4 The Script `split_samples.R`

To optimize memory space usage, we split and compress TemplateFilter input files according to their corresponding chromosome. for example, *sample_1_TF.tab* will be split into :

- *sample_1_chr_1_splited_sample.tab.gz*
- *sample_1_chr_2_splited_sample.tab.gz*
- ...
- *sample_1_chr_17_splited_sample.tab.gz*

To execute this script, run the following command in your R console:

```
source("src/current/split_samples.R")
```

2.3.5 The Script `count_reads.R`

To associate a number of observations (read) to each nucleosome we run the script *count_reads.R*. It produces the files *BY_RM_H3K14ac_wp_and_nbreads.tab*, *BY_RM_H3K14ac_unr_and_nbreads.tab*, *BY_RM_Mnase_Seq_wp_and_nbreads.tab* and *BY_RM_Mnase_Seq_unr_and_nbreads.tab* for H3K14ac common well-positioned nucleosomes, H3K14ac UNRs, Mnase common well-positioned nucleosomes and Mnase UNRs respectively.

For example, the file *BY_RM_H3K14ac_unr_and_nbreads.tab* contains counted reads for well-positioned nucleosomes with the experimental condition ChIP H3K14ac. It is composed of the following columns:

- `chr_BY`, the number of the chromosome for BY
- `lower_bound_BY`, the lower bound of the nucleosome for BY
- `upper_bound_BY`, the upper bound of the nucleosome for BY
- `index_nuc_BY`, the index of the BY nucleosome in the CUR for BY
- `chr_RM`, the number of the chromosome for RM
- `lower_bound_RM`, the lower bound of the nucleosome for RM
- `upper_bound_RM`, the upper bound of the nucleosome for RM
- `index_nuc_RM`, the index of the RM nucleosome in the CUR for RM
- `cur_index`, index of the CUR
- `BY_H3K14ac_36`, the number of reads for the current nucleosome for the sample 36
- `BY_H3K14ac_37`, #reads for sample 37
- `BY_H3K14ac_53`, #reads for sample 53
- `RM_H3K14ac_38`, #reads for sample 38

- RM_H3K14ac_39, #reads for sample 39

To execute this script, run the following command in your R console:

```
source("src/current/count_reads.R")
```

2.3.6 The Script `get_size_factors.R`

This script uses the DESeq function *estimateSizeFactors* to compute the size factor of each sample. It corresponds to normalisation of read counts from sample to sample, as determined by DESeq. When a sample has n reads for a nucleosome or a UNR, the normalised count is n/f where f is the factor contained in this file. The script dumps computed size factors into the file *size_factors.tab*. This file has the form:

sample_id	wp	unr	wpunr
1	0.87396	0.88097	0.87584
2	1.07890	1.07440	1.07760
3	1.06400	1.05890	1.06250
4	0.85782	0.87948	0.86305
5	0.97577	0.96590	0.97307
6	1.19630	1.18120	1.19190
36	0.93318	0.92762	0.93166
37	0.48315	0.48453	0.48350
38	1.11240	1.11210	1.11230
39	0.89897	0.89917	0.89903
53	2.22650	2.22700	2.22660

sample_id are given in file *samples.csv*

If you don't know which column to use for normalization, we recommend using *wpunr*.

Here are the details of the factors produced:

- *unr*: factor computed from data of UNR regions. These regions are defined for every pairs of aligned genomes (e.g. BY_RM)
- *wp*: same, but for well-positioned nucleosomes.
- *wpunr*: both types of regions.

To execute this script, run the following command in your R console:

```
source("src/current/get_size_factors.R")
```

2.3.7 The Script `launch_deseq.R`

Finally, the script *launch_deseq.R* perform statistical analysis on each nucleosome using *DESeq*. It produces files:

- results/current/BY_RM_H3K14ac_wp_snep.tab
- results/current/BY_RM_H3K14ac_unr_snep.tab
- results/current/BY_RM_H3K14ac_wpunr_snep.tab
- results/current/BY_RM_H3K14ac_wp_mnase.tab
- results/current/BY_RM_H3K14ac_unr_mnase.tab
- results/current/BY_RM_H3K14ac_wpunr_mnase.tab

These files are organised with the following columns (see file *BY_RM_H3K14ac_wp_snep.tab* for an example):

- `chr_BY`, the number of the chromosome for BY
- `lower_bound_BY`, the lower bound of the nucleosome for BY
- `upper_bound_BY`, the upper bound of the nucleosome for BY
- `index_nuc_BY`, the index of the BY nucleosome in the CUR for BY
- `chr_RM`, the number of the chromosome for RM
- `lower_bound_RM`, the lower bound of the nucleosome for RM
- `upper_bound_RM`, the upper bound of the nucleosome for RM
- `index_nuc_RM`, the index of the RM nucleosome in the CUR for RM
- `cur_index`, index of the CUR
- `form`
- `BY_Mnase_Seq_1`, the number of reads for the current nucleosome for the sample 1

Next columns concern indicators for each sample:

- `BY_Mnase_Seq_2`, #reads for sample 2
- `BY_Mnase_Seq_3`, #reads for sample 3
- `RM_Mnase_Seq_4`, #reads for sample 4
- `RM_Mnase_Seq_5`, #reads for sample 5
- `RM_Mnase_Seq_6`, #reads for sample 6
- `BY_H3K14ac_36`, #reads for sample 36
- `BY_H3K14ac_37`, #reads for sample 37
- `BY_H3K14ac_53`, #reads for sample 53
- `RM_H3K14ac_38`, #reads for sample 38
- `RM_H3K14ac_39`, #reads for sample 39

The 5 last columns concern DESeq analysis:

- `manip[a_manip] strain[a_strain] manip[a_strain]:strain[a_strain]`, the `manip` (marker) effect, the `strain` effect and the `sneep` effect. These are the coefficients of the fitted generalized linear model.
- `pvalsGLM`, the pvalue resulting from the comparison of the GLM model considering the interaction term *marker:strain* to the GLM model that does not consider it. This is the statistical significance of the interaction term and therefore the statistical significance of the SNEP.
- `sneep_index`, a boolean set to TRUE if the `pvalueGLM` value is under the threshold computed with FDR function with a rate set to 0.0001.

To execute this script, run the following command in your R console:

```
source("src/current/launch_deseq.R")
```

2.4 Results: Number of SNEPs

Here are the number of computed SNEPs for each forms.

form	strains	#nucs	H3K14ac
wp	BY-RM	30464	3549
unr	BY-RM	9497	1559
wpunr	BY-RM	39961	5240

2.5 APPENDICE: Generate .c2c Files

The .c2c files is a simple table that describes how two genome sequences are aligned. This file can be generated by using scripts that were developed in NucleoMiner 1.0 (Nagarajan et al. PLoS Genetics 2010) and which we provide in this release of NucleoMiner2.

To use these scripts on your UNIX/LINUX computer you need first to install MUMmer which is designed to rapidly align entire genomes, whether in complete or draft form.

2.5.1 Installing MUMmer

Get the last version of MUMmer archive on your computer (MUMmer3.23.tar.gz is provided in the directory deps of your working directory). Copy it in a dedicated directory. Install it locally into the src folder of you working directory by typing (working directory):

```
tar -xvzf MUMmer3.23.tar.gz
```

```
cd src
tar xfvz ../deps/MUMmer3.23.tar.gz
cd MUMmer3.23
make check
make install
```

2.5.2 Installing NucleoMiner 1.0 scripts

Get the nucleominer-1.0.tar.gz archive on your computer (this archive is provided in the directory deps of your working directory). Install it locally into the src folder of you working directory by typing (working directory):

```
cd src
tar xfvz ../deps/nucleominer-1.0.tar.gz
cd ..
```

This creates a directory that contains NucleoMiner 1.0 scripts (src/nucleominer-1.0/scripts).

2.5.3 Generate .c2c Files

To generate .c2c files you need to type the following command in a terminal:

```
export PATH=$PATH:src/MUMmer3.23:src/nucleominer-1.0/scripts
export PERL5LIB=$PERL5LIB:src/nucleominer-1.0/scripts/
NMgxcomp data/saccharomyces_cerevisiae_BY_S288c_chromosomes.fasta \
  data/saccharomyces_cerevisiae_rm11-1a_1_supercontigs.fasta \
  data/byxrm 2>NMgxcomp.log
```

After execution, the directory *data* will hold the .c2c files.

REFERENCES

3.1 Python Reference

`configurator.CSV_SAMPLE_FILE = None`
Path to csv file that contains sample information.

`configurator.BOWTIE_BUILD_BIN = None`
Path for bowtie2 build bin.

`configurator.BOWTIE2_BIN = None`
Path for bowtie2 bin.

`configurator.SAMTOOLS_BIN = None`
Path for samtools bin.

`configurator.BEDTOOLS_BIN = None`
Path for bedtools bin.

`configurator.TF_BIN = None`
Path for TemplateFilter bin.

`configurator.TF_TEMPLATES_FILE = None`
Path for TemplateFilter templates file.

`configurator.ILLUMINA_OUTPUTFILE_PREFIX = None`
Prefix for Illumina fastq output files.

`configurator.INDEX_DIR = None`
Path for index dir.

`configurator.ALIGN_DIR = None`
Path for align dir.

`configurator.LOG_DIR = None`
Path for log dir

`configurator.CACHE_DIR = None`
Path for cache dir.

`configurator.RESULTS_DIR = None`
Path for results dir

`configurator.FASTA_REFERENCE_GENOME_FILES = None`
Dictionary where each fasta reference genomes is indexed by reference strain that it corresponds.

`configurator.AREA_BLACK_LIST = None`
Dictionary where keys are strain and values are black listed of genome region.

`configurator.FASTA_INDEXES = None`

Dictionary of strain that indexes dictionaries where keys are chromosome reference from Fastq file and value are its correspondance for Templatefilter.

`configurator.C2C_FILES = None`

Dictionary where each strain combination indexes genome alignment.

`configurator.READ_LENGTH = None`

Length of Illumina reads.

`configurator.MAPQ_THRES = None`

Alignment quality threshold.

`configurator.TF_CORR = None`

TemplateFilter Template correlation threshold.

`configurator.TF_MINW = None`

TemplateFilter minimum width of a nucleosome.

`configurator.TF_MAXW = None`

TemplateFilter maximum width of a nucleosome.

`configurator.TF_OL = None`

TemplateFilter maximum allowed overlap for two nucleosomes.

`wf.json_conf_file = 'src/current/nucleominer_config.json'`

Path to the json configuration file.

`wf.samples = []`

List of samples where a sample is identified by an id (key: *id*) and a strain name (key *strain*).

`wf.samples_mnase = []`

List of Mnase samples.

`wf.strains = []`

List of reference strains.

`libcoverage.create_bowtie_index(strain, strain_fasta_ref, index_dir, bowtie_build_bin)`

Creates bowtie index for a strain *strain*.

Parameters

- **strain** – the strain reference.
- **strain_fasta_ref** – fasta reference genome.
- **index_dir** – directories where to put bowtie index.
- **bowtie_build_bin** – bowtie2 build binary.

`libcoverage.align_reads(sample, align_dir, log_dir, index_dir, illumina_outputfile_prefix, bowtie2_bin, samtools_bin, bedtools_bin)`

Aligns reads to reference genomes. It produces .sam files, that are converted to .bam, that are then converted to .bed.

Parameters

- **sample** – a dict that describe a sample.
- **align_dir** – directory where aligned reads will be stored.
- **log_dir** – directory where logs will be stored.
- **illumina_outputfile_prefix** – prefix of Illumina sequencer fastq.gz output files.
- **bowtie2_bin** – bowtie2 binary.

- **samtools_bin** – samtools binary.
- **bedtools_bin** – bedtools binary.
- **index_dir** – bowtie index directory.

`libcoverage.split_fr_4_TF(sample, align_dir, fasta_indexes, area_black_list, read_length, mapq_thres)`

Create TemplateFilter input files from bed files. This function appends in two times. First, it collects reads from bed files and feeds a datastructure

Parameters

- **sample** – a dict that describe a sample.
- **align_dir** – directory where aligned reads will be stored.
- **fasta_index** – the chr reference from the illumina output file.
- **area_black_list** – the description of genome that will be omit.
- **read_length** – Length of Illumina reads.
- **mapq_thres** – mapping quality criterion threshold, see MAPQ in BED/BAM file format.

`libcoverage.template_filter(sample, align_dir, log_dir, tf_bin, tf_templates_file, corr, minw, maxw, ol)`

Run TemplateFilter on a specific sample. It produces .tab file.

Parameters

- **sample** – a dict that describe a sample.
- **align_dir** – directory where aligned reads will be stored.
- **log_dir** – directory where logs will be stored.
- **tf_bin** – path to the TemplateFilter binary.
- **tf_templates_file** – path to the TemplateFilter templates file.
- **corr** – correlation threshold transmits to TemplateFilter.
- **minw** – minimum width of a nuc, transmits to TemplateFilter.
- **maxw** – maximum width of a nuc, transmits to TemplateFilter.
- **ol** – maximum overlaps for 2 nuc, transmits to TemplateFilter.

3.2 R Reference

3.2.1 Arabic to Roman pair list.

Description

Utility to convert Arabic numbers to Roman numbers

Usage

```
ARAB2ROM()
```

Author(s)

Florent Chuffart

R: False Discovery Rate

3.2.2 False Discovery Rate

Description

From a vector x of independent p-values, extract the cutoff corresponding to the specified FDR. See Benjamini & Hochberg 1995 paper

Usage

```
FDR(x, FDR)
```

Arguments

x

A vector x of independent p-values.

FDR

The specified FDR.

Value

Return the the corresponding cutoff.

Author(s)

Gael Yvert, Florent Chuffart

Examples

```
print("example")
```

R: Roman to Arabic pair list.

3.2.3 Roman to Arabic pair list.

Description

Utility to convert Roman numbers into Arabic numbers

Usage

```
ROM2ARAB ()
```

Author(s)

Florent Chuffart

R: Aggregate replicated sample's nucleosomes.

3.2.4 Aggregate replicated sample's nucleosomes.

Description

This function aggregates nucleosomes from replicated samples. It uses TemplateFilter output of each sample as replicate. Each sample owns a set of nucleosomes computed using TemplateFilter and ordered by the position of their center (dyad). A chain of nucleosomes is built across all replicates. Adjacent nucleosomes of the chain are compared two by two. Comparison is based on a log likelihood ratio (LLR). Depending on the LLR value nucleosomes are merged (low LLR) or separated (high LLR). Finally the function returns a list of clusters and all computed llr_scores. Each cluster owns an attribute wp for "well positioned". This attribute is set to TRUE if the cluster is composed of exactly one nucleosome of each sample.

Usage

```
aggregate_intra_strain_nucs(samples, llr_thres = 20, coord_max = 2e+07)
```

Arguments

`samples`

A list of samples. Each sample is a list like *sample = list(id=..., marker=..., strain=..., roi=..., inputs=..., outputs=...)* with *roi = list(name=..., begin=..., end=..., chr=..., genome=...)*.

`llr_thres`

Log likelihood ratio threshold to decide between merging and separating

`coord_max`

A too big value to be a coord for a nucleosome lower bound.

Value

Returns a list of clusterized nucleosomes, and all computed llr scores.

Author(s)

Florent Chuffart

Examples

```
# Dealing with a region of interest
roi = list(name="example", begin=1000, end=1300, chr="1", genome=rep("A",301))
samples = list()
for (i in 1:3) {
  # Create TF output
  tf_nuc = list("chr"=paste("chr", roi$chr, sep=""), "center"=(roi$end + roi$begin)/2, "width"= 150)
  outputs = dfadd(NULL,tf_nuc)
  outputs = filter_tf_outputs(outputs, roi$chr, roi$begin, roi$end)
  # Generate corresponding reads
  nb_reads = round(runif(1,170,230))
  reads = round(rnorm(nb_reads, tf_nuc$center,20))
  u_reads = sort(unique(reads))
  strands = sample(c(rep("R",ceiling(length(u_reads)/2)),rep("F",floor(length(u_reads)/2))))
  counts = apply(t(u_reads), 2, function(r) { sum(reads == r)})
  shifts = apply(t(strands), 2, function(s) { if (s == "F") return(-tf_nuc$width/2) else return(tf_nuc$width/2)})
  u_reads = u_reads + shifts
  inputs = data.frame(list("V1" = rep(roi$chr, length(u_reads)),
                           "V2" = u_reads,
                           "V3" = strands,
                           "V4" = counts), stringsAsFactors=FALSE)
  samples[[length(samples) + 1]] = list(id=1, marker="Mnase_Seq", strain="strain_ex", total_reads = length(u_reads))
}
print(aggregate_intra_strain_nucs(samples))
```

R: Aligns nucleosomes between 2 strains.

3.2.5 Aligns nucleosomes between 2 strains.

Description

This function aligns nucleosomes between two strains for a given genome region.

Usage

```
align_inter_strain_nucs(replicates, wp_nucs_strain_ref1 = NULL,
  wp_nucs_strain_ref2 = NULL, corr_thres = 0.5, llr_thres = 100,
  config = NULL, ...)
```

Arguments

`replicates`

Set of replicates, ideally 3 per strain.

`wp_nucs_strain_ref1`

List of aggregates nucleosome for strain 1. If it's NULL this list will be computed.

`wp_nucs_strain_ref2`

List of aggregates nucleosome for strain 2. If it's NULL this list will be computed.

`corr_thres`

Correlation threshold.

llr_thres

Log likelihood ratio threshold to decide between merging and separating

config

GLOBAL config variable

...

A list of parameters that will be passed to *aggregate_intra_strain_nucs* if needed.

Value

Returns a list of clusterized nucleosomes, and all computed llr scores.

Author(s)

Florent Chuffart

Examples

```
# Define new translate_cur function...
translate_cur = function(roi, strain2, big_cur=NULL, config=NULL) {
  return(roi)
}
# Binding it by uncomment following lines.
unlockBinding("translate_cur", as.environment("package:nucleominer"))
unlockBinding("translate_cur", getNamespace("nucleominer"))
assign("translate_cur", translate_cur, "package:nucleominer")
assign("translate_cur", translate_cur, getNamespace("nucleominer"))
lockBinding("translate_cur", getNamespace("nucleominer"))
lockBinding("translate_cur", as.environment("package:nucleominer"))

# Dealing with a region of interest
roi =list(name="example", begin=1000, end=1300, chr="1", genome=rep("A",301), strain_ref1 = "STRAIN1", strain_ref2 = "STRAIN2")
roi2 = translate_cur(roi, roi$strain_ref1)
replicates = list()
for (j in 1:2) {
  samples = list()
  for (i in 1:3) {
    # Create TF output
    tf_nuc = list("chr"=paste("chr", roi$chr, sep=""), "center"=(roi$end + roi$begin)/2, "width"=roi$end - roi$begin)
    outputs = dfadd(NULL,tf_nuc)
    outputs = filter_tf_outputs(outputs, roi$chr, roi$begin, roi$end)
    # Generate corresponding reads
    nb_reads = round(runif(1,170,230))
    reads = round(rnorm(nb_reads, tf_nuc$center,20))
    u_reads = sort(unique(reads))
    strands = sample(c(rep("R",ceiling(length(u_reads)/2)),rep("F",floor(length(u_reads)/2))))
    counts = apply(t(u_reads), 2, function(r) { sum(reads == r)})
    shifts = apply(t(strands), 2, function(s) { if (s == "F") return(-tf_nuc$width/2) else return(0)})
    u_reads = u_reads + shifts
    inputs = data.frame(list("V1" = rep(roi$chr, length(u_reads)),
                           "V2" = u_reads,
                           "V3" = strands,
                           "V4" = counts), stringsAsFactors=FALSE)
```

```
        samples[[length(samples) + 1]] = list(id=1, marker="Mnase_Seq", strain=paste("strain_ex", j, sep=""))
    }
    replicates[[length(replicates) + 1]] = samples
}
print(aligned_inter_strain_nucs(replicates))
```

R: Launch DESeq methods.

3.2.6 Launch DESeq methods.

Description

This function is based on DESeq example. It normalizes data, fit data to GLM model with and without interaction term and compares the two models.

Usage

```
analyse_design(snep_design, reads)
```

Arguments

snep_design

The design to consider.

reads

The data to consider.

Author(s)

Florent Chuffart

R: Stage replicates data

3.2.7 Stage replicates data

Description

This function loads in memory the data corresponding to the given experiments.

Usage

```
build_replicates(expe, roi, only_fetch = FALSE, get_genome = FALSE,
  all_samples, config = NULL)
```

Arguments

`expe`

a list of vectors corresponding to replicates.

`roi`

the region that we are interested in.

`only_fetch`

filter or not inputs.

`get_genome`

Load or not corresponding genome.

`all_samples`

Global list of samples.

`config`

GLOBAL config variable.

Author(s)

Florent Chuffart

Examples

```
# library(rjson)
# library(nucleominer)
#
# # Read config file
# json_conf_file = "nucleominer_config.json"
# config = fromJSON(paste(readLines(json_conf_file), collapse=""))
# # Read sample file
# all_samples = get_content(config$CSV_SAMPLE_FILE, "cvs", sep=";", head=TRUE, stringsAsFactors=FALSE)
# # here are the sample ids in a list
# expes = list(c(1))
# # here is the region that we want to see the coverage
# cur = list(chr="8", begin=472000, end=474000, strain_ref="BY")
# # it displays the coverage
# replicates = build_replicates(expes, cur, all_samples=all_samples, config=config)
# out = watch_samples(replicates, config$READ_LENGTH,
#   plot_coverage = TRUE,
#   plot_squared_reads = FALSE,
#   plot_ref_genome = FALSE,
#   plot_arrow_raw_reads = FALSE,
#   plot_arrow_nuc_reads = FALSE,
#   plot_gaussian_reads = FALSE,
#   plot_gaussian_unified_reads = FALSE,
#   plot_ellipse_nucs = FALSE,
#   plot_wp_nucs = FALSE,
#   plot_wp_nuc_model = FALSE,
#   plot_common_nucs = FALSE,
#   height = 50)
```

R: Extract a sub part of the corresponding c2c file

3.2.8 Extract a sub part of the corresponding c2c file

Description

This fonction allows to access to a specific part of the c2c file.

Usage

```
c2c_extraction(strain1, strain2, chr = NULL, lower_bound = NULL,
               upper_bound = NULL, config = NULL)
```

Arguments

strain1

the key strain

strain2

the target strain

chr

if defined, the c2c will be filtered according to the chromosome value

lower_bound

if defined, the c2c will be filtered for part of the genome upper than lower_bound

upper_bound

if defined, the c2c will be filtered for part of the genome lower than upper_bound

config

GLOBAL config variable

Author(s)

Florent Chuffart

R: reformat an “apply manipulated” list of regions

3.2.9 reformat an “apply manipulated” list of regions

Description

Utils to reformat an “apply manipulated” list of regions

Usage

```
collapse_regions(regions)
```

Arguments

regions	
---------	--

Author(s)

Florent Chuffart

R: Compute Common Uninterrupted Regions (CUR)

3.2.10 Compute Common Uninterrupted Regions (CUR)

Description

CURs are regions that can be aligned between the genomes

Usage

```
compute_inter_all_strain_curs(diff_allowed = 30, min_cur_width = 4000,
  config = NULL)
```

Arguments

diff_allowed

the maximum indel width allowed in a CUR

min_cur_width

The minimum width of a CUR

config

GLOBAL config variable

Author(s)

Florent Chuffart

R: Crop bound of regions according to region of interest bound

3.2.11 Crop bound of regions according to region of interest bound

Description

The function is no more necessary since we remove “big_cur” bug in translate_cur function.

Usage

```
crop_fuzzy(tmp_fuzzy_nucs, roi, strain, config = NULL)
```

Arguments

`tmp_fuzzy_nucs`

the regions to be cropped.

`roi`

The region of interest.

`strain`

The strain to consider.

`config`

GLOBAL config variable

Author(s)

Florent Chuffart

R: Adding list to a dataframe.

3.2.12 Adding list to a dataframe.

Description

Add a list *l* to a dataframe *df*. Create it if *df* is *NULL*. Return the dataframe *df*.

Usage

```
dfadd(df, l)
```

Arguments

`df`

A dataframe

`l`

A list

Value

Return the dataframe *df*.

Author(s)

Florent Chuffart

Examples

```
## Here dataframe is NULL
print(df)
df = NULL

# Initialize df
df = dfadd(df, list(key1 = "value1", key2 = "value2"))
print(df)

# Adding elements to df
df = dfadd(df, list(key1 = "value1'", key2 = "value2'"))
print(df)
```

R: Prefetch data

3.2.13 Prefetch data

Description

Fetch and filter inputs and outputs per region of interest. Organize it per replicates.

Usage

```
fetch_mnase_replicates(strain, roi, all_samples, config = NULL,
    only_fetch = FALSE, get_genome = FALSE, get_ouputs = TRUE)
```

Arguments

strain

The strain we want mnase replicatesList of replicates. Each replicates is a vector of sample ids.

roi

Region of interest.

all_samples

Global list of samples.

config

GLOBAL config variable

only_fetch

If TRUE, only fetch and not filtering. It is used tio load sample files into memory before forking.

get_genome

If TRUE, load corresponding genome sequence.

get_ouputs

If TRUE, get also ouput corresponding TF output files.

Author(s)

Florent Chuffart

R: Filter TemplateFilter inputs

3.2.14 Filter TemplateFilter inputs

Description

This function filters TemplateFilter inputs according genome area observed properties. It takes into account reads that are at the frontier of this area and the strand of these reads.

Usage

```
filter_tf_inputs(inputs, chr, x_min, x_max, nuc_width = 160,
  only_f = FALSE, only_r = FALSE, filter_for_coverage = FALSE)
```

Arguments

inputs

TF inputs to be filtered.

chr

Chromosome observed, here chr is an integer.

x_min

Coordinate of the first bp observed.

x_max

Coordinate of the last bp observed.

nuc_width

Nucleosome width.

only_f

Filter only F reads.

only_r

Filter only R reads.

filter_for_coverage

Does it filter for plot coverage?

Value

Returns filtred inputs.

Author(s)

Florent Chuffart

R: Filter TemplateFilter outputs

3.2.15 Filter TemplateFilter outputs**Description**

This function filters TemplateFilter outputs according, not only genome area observed properties, but also correlation and overlapping threshold.

Usage

```
filter_tf_outputs(tf_outputs, chr, x_min, x_max, nuc_width = 160,  
                  ol_bp = 59, corr_thres = 0.5)
```

Arguments

tf_outputs

TemplateFilter outputs.

chr

Chromosome observed, here chr is an integer.

x_min

Coordinate of the first bp observed.

x_max

Coordinate of the last bp observed.

nuc_width

Nucleosome width.

ol_bp

Overlap Threshold.

corr_thres

Correlation threshold.

Value

Returns filtered TemplateFilter Outputs

Author(s)

Florent Chuffart

R: to flat aggregate_intra_strain_nucs function output

3.2.16 to flat aggregate_intra_strain_nucs function output

Description

This function builds a dataframe of all clusters obtain from aggregate_intra_strain_nucs function.

Usage

```
flat_aggregated_intra_strain_nucs(partial_strain_maps, cur_index,  
    nb_tracks = 3)
```

Arguments

partial_strain_maps

the output of aggregate_intra_strain_nucs function

cur_index

the index of the roi involved

nb_tracks

the number of replicates

Value

Returns a dataframe of all clusters obtain from aggregate_intra_strain_nucs function.

Author(s)

Florent Chuffart

R: flat reads

3.2.17 flat reads

Description

Extract reads coordinates from TemplateFilter input sequence

Usage

```
flat_reads(reads, nuc_width)
```

Arguments

reads

TemplateFilter input reads

nuc_width

Width used to shift F and R reads.

Value

Returns a list of F reads, R reads and joint/shifted F and R reads.

Author(s)

Florent Chuffart

R: Retrieve Reads

3.2.18 Retrieve Reads

Description

Retrieve reads for a given marker, combi, form.

Usage

```
get_all_reads(marker, combi, form = "wp", config = NULL)
```

Arguments

marker

The marker to considere.

combi

The starin combination to considere.

form

The nuc form to considere.

config

GLOBAL config variable

Author(s)

Florent Chuffart

R: get comp strand

3.2.19 get comp strand

Description

Compute the complementatry strand.

Usage

```
get_comp_strand(strand)
```

Arguments

strand

The original strand.

Value

Returns the complementatry strand.

Author(s)

Florent Chuffart

R: Build the design for DESeq

3.2.20 Build the design for DESeq

Description

This function build the design according sample properties.

Usage

```
get_design(marker, combi, all_samples)
```

Arguments

marker

The marker to considere.

combi

The starin combination to considere.

all_samples

Global list of samples.

Author(s)

Florent Chuffart

R: Compute the fuzzy list for a given strain.

3.2.21 Compute the fuzzy list for a given strain.**Description**

This function grabs the nucleosomes detected by `template_filter` that have been rejected by `aggregate_intra_strain_nucs` as well positions.

Usage

```
get_intra_strain_fuzzy(wp_map, roi, strain, config = NULL)
```

Arguments

`wp_map`

Well positionned nucleosomes map.

`roi`

The region of interest.

`strain`

The strain we want to extract the fuzzy map.

`config`

GLOBAL config variable.

Author(s)

Florent Chuffart

R: Compute the list of SNEPs for a given set of marker, strain...

3.2.22 Compute the list of SNEPs for a given set of marker, strain combination and nuc form.**Description**

This function uses

Usage

```
get_sneps(marker, combi, form, all_samples, FDR = 1e-04, config = NULL)
```

Arguments

marker

The marker involved.

combi

The strain combination involved.

form

the nuc form involved.

all_samples

Global list of samples.

FDR

config

GLOBAL config variable

Author(s)

Florent Chuffart

Examples

```
marker = "H3K4me1"
combi = c("BY", "YJM")
form = "wpunr" # "wp" | "unr" | "wpunr"
# foo = get_sneps(marker, combi, form)
# foo = get_sneps("H4K12ac", c("BY", "RM"), "wp")
```

R: Compute the unaligned nucleosomal regions (UNRs).

3.2.23 Compute the unaligned nucleosomal regions (UNRs).

Description

This function aggregate non common wp nucs for each strain and subtract common wp nucs. It does not take care about the size of the resulting UNR. It will be take into account in the count read part of the pipeline.

Usage

```
get_unrs(combi, roi, cur_index, wp_maps, fuzzy_maps, common_nuc_results,
         config = NULL)
```

Arguments

`combi`

The strain combination to consider.

`roi`

The region of interest.

`cur_index`

The region of interest index.

`wp_maps`

Well positionned nucleosomes maps.

`fuzzy_maps`

Fuzzy nucleosomes maps.

`common_nuc_results`

Common wp nuc maps

`config`

GLOBAL config variable

Author(s)

Florent Chuffart

R: Returns the intersection of 2 list on regions.

3.2.24 Returns the intersection of 2 list on regions.

Description

This function...

Usage

```
intersect_region(region1, region2)
```

Arguments

`region1`

Original regions.

`region2`

Regions to intersect.

Author(s)

Florent Chuffart

R: Likelihood ratio

3.2.25 Likelihood ratio**Description**

Compute the log likelihood ratio of two or more set of value.

Usage

```
llr_score_nvecs(xs)
```

Arguments

`xs`

list of vectors.

Value

Returns the log likelihood ratio.

Author(s)

Florent Chuffart

Examples

```
# LLR score for 2 set of values
mean1=5; sd1=2; card2 = 250
mean2=6; sd2=3; card1 = 200
x1 = rnorm(card1, mean1, sd1)
x2 = rnorm(card2, mean2, sd2)
min = floor(min(c(x1,x2)))
max = ceiling(max(c(x1,x2)))
hist(c(x1,x2), xlim=c(min, max), breaks=min:max)
lines(min:max, dnorm(min:max, mean1, sd1) * card1, col=2)
lines(min:max, dnorm(min:max, mean2, sd2) * card2, col=3)
lines(min:max, dnorm(min:max, mean(c(x1,x2)), sd(c(x1,x2))) * card2, col=4)
llr_score_nvecs(list(x1,x2))
```

R: nm

3.2.26 nm

Description

It provides a set of useful functions allowing to perform quantitative analysis of nucleosomal epigenome.

Details

Package:	nucleominer
Maintainer:	Florent Chuffart <florent.chuffart@ens-lyon.fr>
Author:	Florent Chuffart
Version:	2.3.52
License:	CeCILL
Title:	nm
Depends:	seqinr, plotrix, DESeq, cachecache, dplyr

Author(s)

Florent Chuffart

R: Plot the distribution of reads.

3.2.27 Plot the distribution of reads.

Description

This function use the DESeq normalization feature to compare qualitatively the distribution.

Usage

```
plot_dist_samples(strain, marker, res, all_samples, NEWPLOT = TRUE)
```

Arguments

strain

The strain to considere.

marker

The marker to considere.

res

Data

all_samples

Global list of samples.

NEWPLOT

If FALSE the curve will be add to the current plot.

Author(s)

Florent Chuffart

R: sign from strand

3.2.28 sign from strand

Description

Get the sign of strand

Usage

```
sign_from_strand(strands)
```

Arguments

strands	
---------	--

Value

If strand in forward then returns 1 else returns -1

Author(s)

Florent Chuffart

R: Substract to a list of regions an other list of regions that...

3.2.29 Substract to a list of regions an other list of regions that intersect it.

Description

This fuction embed a recursive part. It occurs when a substracted region split an original region on two.

Usage

```
substract_region(region1, region2)
```

Arguments

region1

Original regions.

region2

Regions to substract.

Author(s)

Florent Chuffart

R: Switch a pairlist

3.2.30 Switch a pairlist**Description**

Take a pairlist key:value and return the switched pairlist value:key.

Usage

```
switch_pairlist(l)
```

Arguments

`l`

The pairlist to switch.

Value

The switched pairlist.

Author(s)

Florent Chuffart

Examples

```
l = list(key1 = "value1", key2 = "value2")
print(switch_pairlist(l))
```

R: Translate coords of a genome region.

3.2.31 Translate coords of a genome region.**Description**

This function is used in the examples, usually you have to define your own translation function and overwrite this one using *unlockBinding* features. Please, refer to the example.

Usage

```
translate_cur(roi, strain2, config = NULL, big_cur = NULL)
```

Arguments

roi

Original genome region of interest.

strain2

The strain in wich you want the genome region of interest.

config

GLOBAL config variable

big_cur

A largest region than roi use to filter c2c if it is needed.

Author(s)

Florent Chuffart

Examples

```
# Define new translate_cur function...
translate_cur = function(roi, strain2, config) {
  strain1 = roi$strain_ref
  if (strain1 == strain2) {
    return(roi)
  } else {
    stop("Here is my new translate_cur function...")
  }
}

# Binding it by uncomment follwing lines.
# unlockBinding("translate_cur", as.environment("package:nm"))
# unlockBinding("translate_cur", getNamespace("nm"))
# assign("translate_cur", translate_cur, "package:nm")
# assign("translate_cur", translate_cur, getNamespace("nm"))
# lockBinding("translate_cur", getNamespace("nm"))
# lockBinding("translate_cur", as.environment("package:nm"))
```

R: Translate a list of regions from a strain ref to another.

3.2.32 Translate a list of regions from a strain ref to another.

Description

This function is an elaborated call to `translate_cur`.

Usage

```
translate_regions(regions, combi, cur_index, config = NULL, roi)
```

Arguments

`regions`

Regions to be translated.

`combi`

Combination of strains.

`cur_index`

The region of interest index.

`config`

GLOBAL config variable

`roi`

The region of interest.

Author(s)

Florent Chuffart

R: Aggregate regions that intersect themselves.

3.2.33 Aggregate regions that intersect themselves.

Description

This function is based on sort of lower bounds to detect regions that intersect. We compare lower bound and upper bound of the porevious item. This function embed a while loop and break break regions list become stable.

Usage

```
union_regions(regions)
```

Arguments

`regions`

The Regions to be aggregated

Author(s)

Florent Chuffart

R: Watching analysis of samples

3.2.34 Watching analysis of samples

Description

This function allows to view analysis for a particular region of the genome.

Usage

```
watch_samples(replicates, read_length, plot_ref_genome = TRUE,
  plot_arrow_raw_reads = TRUE, plot_arrow_nuc_reads = TRUE,
  plot_squared_reads = TRUE, plot_coverage = FALSE, plot_gaussian_reads = TRUE,
  plot_gaussian_unified_reads = TRUE, plot_ellipse_nucs = TRUE,
  change_col = TRUE, plot_wp_nucs = TRUE, plot_fuzzy_nucs = FALSE,
  plot_wp_nuc_model = TRUE, plot_common_nucs = FALSE, plot_common_unrs = FALSE,
  plot_wp_nucs_4_nonmnase = FALSE, plot_chain = FALSE, plot_sample_id = FALSE,
  aggregated_intra_strain_nucs = NULL, aligned_inter_strain_nucs = NULL,
  height = 10, main = NULL, xlab = NULL, ylab = "#reads (per million reads)",
  config = NULL)
```

Arguments

`replicates`

replicates under the form...

`read_length`

length of the reads

`plot_ref_genome`

Plot (or not) reference genome.

`plot_arrow_raw_reads`

Plot (or not) arrows for raw reads.

`plot_arrow_nuc_reads`

Plot (or not) arrows for reads associated to a nucleosome.

`plot_squared_reads`

Plot (or not) reads in the square fashion.

`plot_coverage`

Plot (or not) reads in the coverage fashion. fashion.

`plot_gaussian_reads`

Plot (or not) gaussian model of a F and R reads.

`plot_gaussian_unified_reads`

Plot (or not) gaussian model of a nuc.

`plot_ellipse_nucs`

Plot (or not) ellipse for a nuc.

`change_col`

Change the color of each nucleosome.

`plot_wp_nucs`

Plot (or not) cluster of nucs

`plot_fuzzy_nucs`

Plot (or not) cluster of fuzzy

`plot_wp_nuc_model`

Plot (or not) gaussian model for a cluster of nucs

`plot_common_nucs`

Plot (or not) aligned reads.

`plot_common_unrs`

Plot (or not) unaligned nucleosomal regions (UNRs).

`plot_wp_nucs_4_nonmnase`

Plot (or not) clusters for non inputs samples.

`plot_chain`

Plot (or not) clusterised nuceosomes between mnase samples.

`plot_sample_id`

Plot (or not) the sample id for each sample.

`aggregated_intra_strain_nucs`

list of aggregated intra strain nucs. If NULL, it will be computed.

`aligned_inter_strain_nucs`

list of aligned inter strain nucs. If NULL, it will be computed.

`height`

Number of reads in per million read for each sample, graphical parametre for the y axis.

`main`

main title of the produced plot

`xlab`

xlab of the produced plot

`ylab`

ylab of the produced plot

`config`

GLOBAL config variable

Author(s)

Florent Chuffart

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