Stitch-seq-tools Documentation

Release 0.2.2

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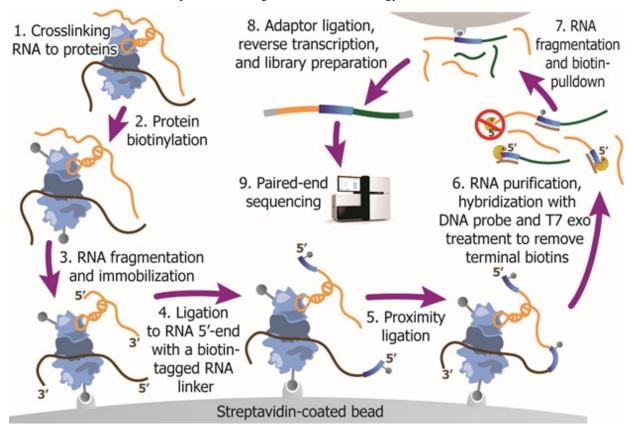
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RNA-STITCH-SEQ-TOOLS 0.2 DOCUMENTATION

1.1 Overview

RNA-Stitch-seq-tools is a set of bioinformatic tools for analysis of a novel DNA sequencing based technology to detect RNA-RNA interactome and RNA-chromatin interactome (RNA-chromatin interactome is coming soon).

Below is a illustration for the experimental design of this new technology



See also:

Offline documentation.

Download a copy of RNA-stitch-seq-tools documentation:

- PDF
- Epub

1.2 Installation

1.2.1 step 1: Install the dependent prerequisites:

- 1. Python libraries [for python 2.x]:
- · Biopython
- · Pysam
- BAM2X
- · Numpy, Scipy
- Parallel python (Only for Select_strongInteraction_pp.py)
- PyCogent (for annotation of RNA types) [see note]
- 2. The Boost.Python C++ library
- 3. Other softwares needed:
- Bowtie (not Bowtie 2)
- · samtools
- NCBI blast+ (use blastn)

Note: the Annotation feature need the development version of PyCogent (install instruction). Since we need the getTranscriptByStableId function which is described here.

1.2.2 Step 2: Download the package

Clone the package from GitHub:

```
git clone http://github.com/yu68/stitch-seq.git
```

1.2.3 Step 3: Add library source to your python path

Add these lines into your ~/.bash_profile or ~/.profile

```
Location="/path/of/RNA-Stitch-seq-tools" # change accordingly export PYTHONPATH="$Location/src:$PYTHONPATH" export PATH="$PATH:$Location/bin" Loc_lib="/path/of/boost_1_xx_0/lib/" # change accordingly export LD_LIBRARY_PATH="$Loc_lib:$LD_LIBRARY_PATH"
```

1.3 Support

For issues related to the use of RNA-Stitch-seq-tools, or if you want to **report a bug or request a feature**, please contact Pengfei Yu <p3yu at ucsd dot edu>

CHAPTER

TWO

ANALYSIS PIPELINE

2.1 Overview

The next generation DNA sequencing based technology utilize RNA proximity ligation to transfrom RNA-RNA interactions into chimeric DNAs. Through sequencing and mapping these chimeric DNAs, it is able to achieve high-throughput mapping of nearly entire interaction networks. RNA linkers were introduced to mark the junction of the ligation and help to split the chimeric RNAs into two interacting RNAs. This bioinformatic pipeline is trying to obtain the strong interactions from raw fastq sequencing data. The major steps are:

- Step 1: Remove PCR duplicates.
- Step 2: Split library based on barcode.txt.
- Step 3: Recover fragments for each library.
- Step 4: Split partners and classify different types of fragments.
- Step 5: Align both parts of "Paired" fragment to the genome.
- Step 6: Determine strong interactions.
- Step 7: Visualization of interactions and coverages.

Other functions:

- 1. Determine the RNA types of different parts within fragments.
- 2. Find linker sequences within the library.

2.2 Pipeline

2.2.1 Step 1: Remove PCR duplicates.

Starting from the raw pair-end sequencing data, PCR duplicates should be removed as the first step if both the 10nt random indexes and the remaining sequences are exactly the same for two pairs. It is achieved by remove_dup_PE.py

```
optional arguments:
   -h, --help show this help message and exit
Library dependency: Bio, itertools
```

The program will generate two fastq/fasta files after removind PCR duplicates and report how many read pairs has been removed. The output are prefixed with 'Rm_dupPE'

Note: One pair is considered as a PCR duplicate only when the sequences of both two ends (including the 10nt random index) are the exactly same as any of other pairs.

2.2.2 Step 2: Split library based on barcode.txt.

After removing PCR duplicates, the libraries from different samples are separated based on 4nt barcodes in the middle of random indexes ("RRRBBBBRRR"; R: random, B: barcode). It is implemented by split_library_pairend.py

```
usage: split_library_pairend.py [-h] [-f | -q] [-v] [-b BARCODE]
                                [-r RANGE [RANGE ...]] [-t] [-m MAX_SCORE]
                                input1 input2
Example: split_library_pairend.py -q Rm_dupPE_example.F1.fastq
        Rm_dupPE_example.R1.fastq -b barcode.txt
positional arguments:
 input1
                        input fastq/fasta file 1 for pairend data (contain
 input2
                        input fastq/fasta file 2 for pairend data
optional arguments:
 -h, --help
                       show this help message and exit
 -f, --fasta
                       add this option for fasta input file
                       add this option for fastq input file
 -q, --fastq
                 show program's version number and exit
 -v, --version
 -b BARCODE, --barcode BARCODE
                       barcode file
 -r RANGE [RANGE ...], --range RANGE [RANGE ...]
                        set range for barcode location within reads, default is
                       full read
 -t, --trim
                       trim sequence of 10nt index
 -m MAX_SCORE, --max_score MAX_SCORE
                        max(mismatch+indel) allowed for barcode match,
                        otherwise move reads into 'unassigned' file
                        default: 2.
Library dependency: Bio
```

Here is a example for barcode.txt

ACCT CCGG

The output of this script are several pairs of fastq/fasta files prefixed with the 4nt barcode sequences, together with another pair of fastq/fasta files prefixed with 'unassigned'.

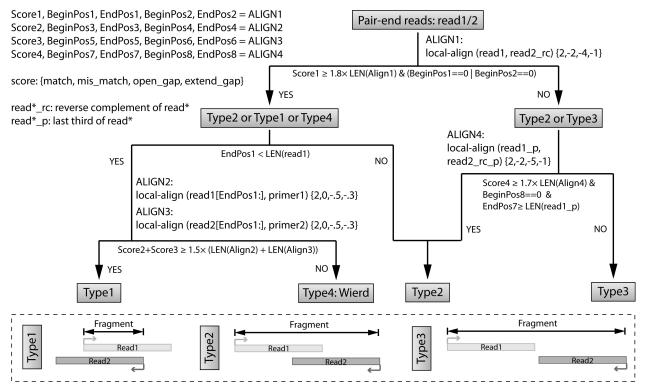
For example, if the input fastq/fasta files are Rm_dupPE_example.F1.fastq and Rm_dupPE_example.R1.fastq, and the barcode file is the same as above, then the output files are:

- ACCT_Rm_dupPE_example.F1.fastq
- ACCT_Rm_dupPE_example.R1.fastq
- CCGG_Rm_dupPE_example.F1.fastq
- CCGG_Rm_dupPE_example.R1.fastq
- GGCG_Rm_dupPE_example.F1.fastq
- GGCG_Rm_dupPE_example.R1.fastq
- unassigned_Rm_dupPE_example.F1.fastq
- unassigned_Rm_dupPE_example.R1.fastq

2.2.3 Step 3: Recover fragments for each library.

After splitting the libraries, the later steps from here (Step 3-7) need to be executed parallelly for each sample.

In this step, we are trying to recover the fragments based on local alignment. The fragments are classifed as several different types as shown in the figure below. The flow chart is also clarified at the top.



We will use a complied program recoverFragment to do that

recoverFragment - recover fragment into 4 different categories from pair-end seq data

SYNOPSIS

DESCRIPTION

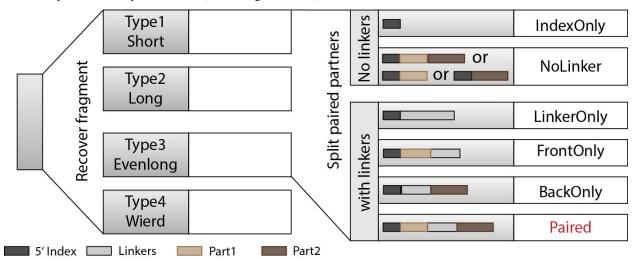
-h, --help

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```
Displays this help message.
    --version
          Display version information
    -I, --inputs STR
          input of forward and reverse fastq file, path of two files separated by SPACE
    -p, --primer STR
          fasta file contianing two primer sequences
    -v, --verbose
          print alignment information for each alignment
EXAMPLES
    recoverFragment -I read_1.fastq read_2.fastq -p primer.fasta
          store fragment using fasta/fastg into 4 output files
          'short_*', 'long_*','evenlong_*','wierd_*'
VERSION
    recoverFragment version: 0.1
    Last update August 2013
```

2.2.4 Step 4: Split partners and classify different types of fragments.

When we recovered the fragments, the next we are goting to do is to find parts that are separated by the linkers, and from here, we will be able to classify the fragments into different types: "IndexOnly", "NoLinker", "LinkerOnly", "BackOnly", "FrontOnly", "Paired". (see the figure below).



This will be done by split_partner.py

```
type3_2
                        read_2 for evenlong (type3) fastq file
optional arguments:
 -h, --help
                        show this help message and exit
  -e EVALUE, --evalue EVALUE
                        cutoff evalues, only choose alignment with evalue less
                        than this cutoffs (default: 1e-5).
 --linker_db LINKER_DB
                        BLAST database of linker sequences
  --blast_path BLAST_PATH
                        path for the local blast program
 -o OUTPUT, --output OUTPUT
                        output file containing sequences of two sepatated
                        part.s
 -t TRIM, --trim TRIM trim off the first this number of nt as index,
                        default:10
 -b BATCH, --batch BATCH
                        batch this number of fragments for BLAST at a time.
                        default: 100000
 -1 LENGTH, --length LENGTH
                        shortest length to be considered for each part of the
                        pair, default: 15
Library dependency: Bio, itertools
```

The linker fasta file contain sequences of all linkers

```
>L1
CTAGTAGCCCATGCAATGCGAGGA
>L2
AGGAGCGTAACGTACCCGATGATC
```

The output fasta files will be the input file name with different prefix ("NoLinker", "LinkerOnly", "BackOnly", "FrontOnly", "Paired") for different types. The other output file specified by $-\circ$ contains information of aligned linker sequences for each Type1/2 fragment.

For example, if the commend is

```
split_partner.py fragment_ACCT.fasta evenlong_ACCTRm_dupPE_stitch_seq_1.fastq
    evenlong_ACCTRm_dupPE_stitch_seq_2.fastq
    -o fragment_ACCT_detail.txt --linker_db linker.fa
```

Then, the output files will be:

- backOnly_fragment_ACCT.fasta
- NoLinker_fragment_ACCT.fasta
- frontOnly_fragment_ACCT.fasta
- Paired1_fragment_ACCT.fasta
- Paired2_fragment_ACCT.fasta
- fragment ACCT detail.txt

The format of the last output file fragment_ACCT_detail.txt will be "Name | linker_num | linker_loc | Type | linker_order". Here are two examples:

```
HWI-ST1001:238:HONYEADXX:1:1101:10221:1918 L1:2;L2:1 19,41;42,67;68,97 None L2;L1;L1 HWI-ST1001:238:HONYEADXX:1:1101:4620:2609 L1:2 28,46;47,79 Paired L1;L1
```

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In the **first** fragment, there are three regions can be aligned to linkers, 2 for L1 and 1 for L2, the order is L2, L1, L1. And they are aligned in region [19,41], [42,67], [68,97] of the fragment. "None" means this fragment is either 'LinkerOnly' or 'IndexOnly' (in this case it is 'LinkerOnly'). This fragment won't be written to any of the output fasta files.

In the **second** fragment, two regions can be aligned to linkers, and they are both aligned to L1. The two regions are in [28,46], [47,79] of the fragment. the fragment is "Paired" because on both two sides flanking the linker aligned regions, the length is larger than 15nt. The left part will be writen in Paired1_fragment_ACCT.fasta and the right part in Paired2 fragment ACCT.fasta

2.2.5 Step 5: Align both parts of "Paired" fragment to the genome.

In this step, we will use the Paired1* and Paired2* fasta files output from the previous step. The sequences of part1 and part2 are aligned to the mouse genome mm9 with Bowtie and the pairs with both part1 and part2 mappable are selected as output. We also annotate the RNA types of each part in this step. All of these are implemented using script Stitch-seq_Aligner.py.

```
usage: Stitch-seq_Aligner.py [-h] [-s samtool_path] [-a ANNOTATION]
                                 [-A DB_DETAIL]
                                miRNA_reads mRNA_reads bowtie_path miRNA_ref
                                mRNA_ref
Align miRNA-mRNA pairs for Stitch-seq. print the alignable miRNA-mRNA pairs
with coordinates
positional arguments:
 part1_reads paired part1 fasta file
part2_reads paired part2 fasta file
bowtie_path path for the bowtie program
part1_ref reference genomic seq for part1
  part2_ref
                         reference genomic seq for part2
optional arguments:
  -h, --help
                         show this help message and exit
  -s samtool_path, --samtool_path samtool_path
                          path for the samtool program
  -a ANNOTATION, --annotation ANNOTATION
                           If specified, include the RNA type annotation for each
                           aligned pair, need to give bed annotation RNA file
  -A DB_DETAIL, --annotationGenebed DB_DETAIL
                           annotation bed12 file for lincRNA and mRNA with intron
                           and exon
Library dependency: Bio, pysam, itertools
```

An annotation file for different types of RNAs in mm9 genome (bed format, 'all_RNAs-rRNA_repeat.txt.gz') was included in Data folder. The annotation bed12 file for lincRNA and mRNA ('Ensembl_mm9.genebed.gz') was also included in Data folder. One can use the option -a ../Data/all_RNAs-rRNA_repeat.txt.gz -A ../Data/Ensembl_mm9.genebed.gz for annotation.

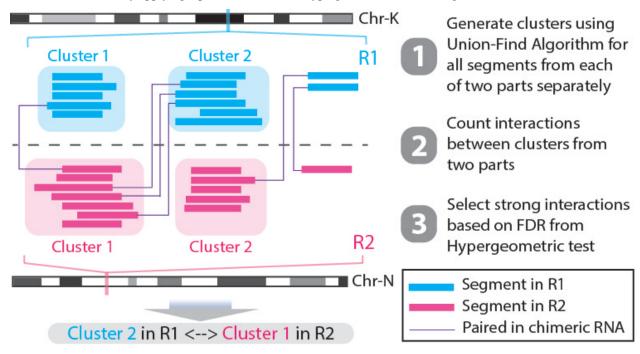
Here is a example:

The format for the output file ACCT_fragment_paired_align.txt will be:

Column ¹	Description
1	chromosome name of part1
2,3	start/end position of part1
4	strand information of part1
5	sequence of part1
6	RNA type for part1
7	RNA name for part1
8	RNA subtype ² for part1
9	name of the pair

2.2.6 Step 6: Determine strong interactions.

In this step, we will generate clusters with high coverage separately for all part1 (R1) an part2 (R2) segments. Then based on the pairing information, we count the interactions between clusters from part1 and part2. The strong interactions can be selected by applying a p-value cutoff from hypergeometric test. (See figure below)



We will use the script Select_strongInteraction_pp.py, parallel computing are implemented for clustering parallelly on different chromosomes:

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¹column 10-17 are the same as column 1-8 except they are for part2 instead of part1.

²subtype can be intron/exon/utr5/utr3 for lincRNA and mRNA (protein-coding), '.' for others

```
minimum number of segments allowed in each cluster,
default:5

-m MIN_INTERACTION, --min_interaction MIN_INTERACTION
minimum number of interactions to support a strong
interaction, default:3

-p P_VALUE, --p_value P_VALUE
the p-value based on hypergeometric distribution to
call strong interactions, default: 0.05

-o OUTPUT, --output OUTPUT
specify output file
-P PARALLEL, --parallel PARALLEL
number of workers for parallel computing, default: 5
-F, --FDR Compute FDR if specified

need Scipy for hypergeometric distribution
```

The input of the script is the output of Step 5 (ACCT_fragment_paired_align.txt in the example). "annotated_bed" class is utilized in this script.

Here is a example:

```
Select_strongInteraction.py -i ACCT_fragment_paired_align.txt -o ACCT_interaction_clusters.txt
```

The column description for output file ACCT_interaction_clusters.txt is:

Column	Description	
1	chromosome name of cluster in part1	
2,3	start/end position of cluster in part1	
4	RNA type for cluster in part1	
5	RNA name for cluster in part1	
6	RNA subtype for cluster in part1	
7	# of counts for cluster in part1	
8-14	Same as 1-7, but for cluster in part2	
15	# of interactions between these two clusters	
16	log(p-value) of the hypergeometric testing	

2.2.7 Step 7: Visualization of interactions and coverages.

There are two ways of visulization provided (LOCAL and GLOBAL):

- Visualization of local interactions.
- Visualization of global interactome.

2.3 Other functions

2.3.1 Determine the RNA types of different parts within fragments.

2.3.2 Find linker sequences within the library.

2.3.3 Find intersections between two different interaction sets.

The script tool 'intersectInteraction.py' could be used to identify overlap of interactions between two interaction set from independent experiments (two replicates or treatment v.s. control)

```
usage: intersectInteraction.py [-h] -a FILEA -b FILEB [-s START] [-n NBASE]
                               [-o OUTPUT] [-c]
find intersections (overlaps) between two interaction sets
optional arguments:
 -h, --help
                        show this help message and exit
 -a FILEA, --filea FILEA
                        file for interaction set a
 -b FILEB, --fileb FILEB
                        file for interaction set b
 -s START, --start START
                        start column number of the second part in each
                        interaction (0-based), default:7
 -n NBASE, --nbase NBASE
                        number of overlapped nucleotides for each part of
                      interactions to call intersections, default: 1
 -o OUTPUT, --output OUTPUT
                        specify output file
                        Use a set of different 'nbase' to call overlaps and
 -c, --compare
                        find the best one. if nbase=-200, then choose from
                        [0, -10, -20, \ldots, -200]
require numpy and matplotlib if set '-c'
```

if "-c" option is set, then the program will also output a plot to show different numbers of intersections (overlaps) given a sequence of different "-n" parameters.

2.3. Other functions

VISUALIZATION OF LOCAL RNA-RNA INTERACTIONS

3.1 Prerequirement

This program require python modules: xplib, matplotlib, numpy, bx-python

3.2 Run the program to generate visualization

The script "Plot_interaction.py" will be used for this purpose,

```
usage: Plot_interaction.py [-h] [-n N] [-s START [START ...]] [-d DISTANCE]
                           [-g GENEBED] [-w PHYLOP_WIG] [-p PAIR_DIST] [-S]
                           [-o OUTPUT]
                           interaction linkedPair
plot linked pairs around a given interaction. information of linked pairs are
stored in file '*_fragment_paired_align.txt'
positional arguments:
                        Interaction file from output of
 interaction
                        'Select_strongInteraction_pp.py'
 linkedPair
                        file for information of linked pairs, which is output
                        of 'Stitch-seq_Aligner.py'
optional arguments:
 -h, --help
                        show this help message and exit
 -n N
                        Choose region to plot, it can be a number (around n-th
                        interaction in the interaction file) or one/two
                        regions with format 'chr:start-end', default=1
 -s START [START ...], --start START [START ...]
                        start column number of the second region in
                        interaction file and linkedPair file, default=(7,8)
 -d DISTANCE, --distance DISTANCE
                        the plus-minus distance (unit: kbp) flanking the
                        interaction regions to be plotted, default=10
 -g GENEBED, --genebed GENEBED
                        the genebed file from Ensembl, default:
                        ../Data/Ensembl_mm9.genebed
 -w PHYLOP_WIG, --phyloP_wig PHYLOP_WIG
                        the bigWig file for phyloP scores, defualt:
                        mouse.phyloP30way.bw
 -p PAIR_DIST, --pair_dist PAIR_DIST
                        two interacted parts within this distance are
```

```
considered as self-ligated and they are marked or eliminated (see option -s for slim mode), default:
200bp
-S, --Slim set slim mode to eliminate self ligated interactions
-o OUTPUT, --output OUTPUT output plot file, can be format of emf, eps, pdf, png, ps, raw, rgba, svg, svgz
```

Note: linkedPair file is the output *_fragment_paired_align.txt from *Step5:Stitch-seq_Aligner.py* of the pipeline; Interaction txt file is the output of *Step6:Select_strongInteraction_pp.py*.

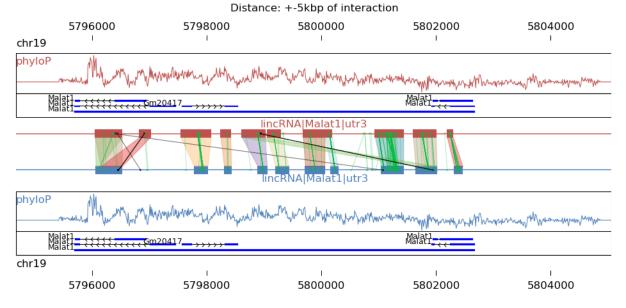
3.3 Example of result graph

Example code:

```
python Plot_interaction.py
    ACCT_interaction_clusters_rmrRNA.txt \
    ACCT_fragment_paired_align_rmRNA_sort.txt.gz \
    -n 2412 \
    -d 5 \
    -o local_interaction.pdf
```

Result figure:

chr19:5799714-5800053 <-> chr19:5799677-5800057



Explanation:

CHAPTER

FOUR

VISUALIZATION OF GLOBAL RNA-RNA INTERACTOME

4.1 Prerequirement

This program is powered by RCircos.

Required R packages (our program will check for the presence of these packages and install/load them automatically if not present):

• argparse, RCircos, biovizBase, rtracklayer

The program also require a python script "bam2tab.py" (already in /bin/ folder) to call coverage from BAM2X

4.2 Run the program to generate visualization

We will use the script "Plot_Circos.R" for this purpose.

```
usage: Plot_Circos.R [-h] [-g GENOME] [-b BIN] [-o OUTPUT]
                  interaction part1 part2
positional arguments:
 interaction
                       the interaction file, [required]
 part1
                       aligned BAM file for part1, [required]
 part2
                       aligned BAM file for part2, [required]
optional arguments:
 -h, --help
                        show this help message and exit
 -g GENOME, --genome GENOME
                        genome information, choice: mm9/mm10/hg19 et.al.,
                        [default: mm9]
 -b BIN, --bin BIN
                       window size for the bins for coverage calling, [default: 100000.0]
 -o OUTPUT, --output OUTPUT
                        output pdf file name, [default: Interactome_view.pdf]
```

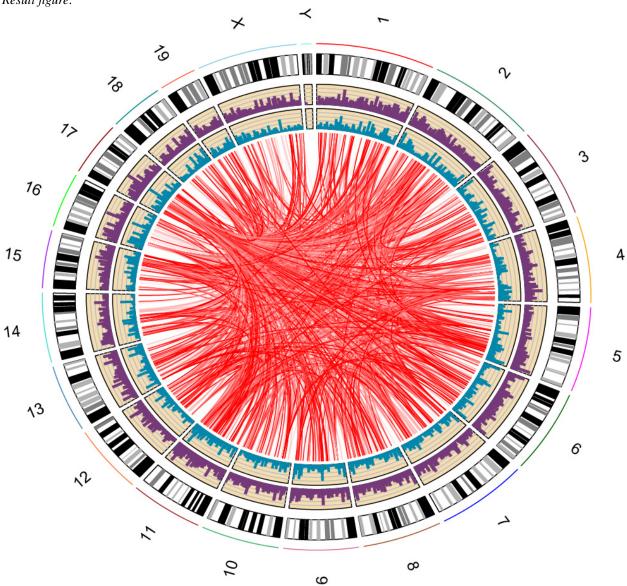
Note: part1, part2 BAM files are the ones generated from *Step5:Stitch-seq_Aligner.py* of the pipeline; Interaction txt file is the output of *Step6:Select_strongInteraction_pp.py*.

4.3 Example of result graph

Example code:

Rscript Plot_Circos.R GGCG_interaction_clusters.txt
 sort_Paired1_fragment_GGCG.bam sort_Paired2_fragment_GGCG.bam
 -b 100000 -o Interactome_GGCG.pdf

Result figure:



Explanation:

CHAPTER

FIVE

PYTHON APIS CREATED FOR THIS PROJECT

5.1 Annotation module

For the purpose of annotating RNA types for genomic regions.

```
Annotation.overlap(bed1, bed2)
```

This function compares overlap of two Bed object from same chromosome

Parameters

- bed1 A Bed object from xplib.Annotation.Bed (BAM2X)
- bed2 A Bed object from xplib.Annotation.Bed (BAM2X)

Returns boolean - True or False

Example:

```
>>> from xplib.Annotation import Bed
>>> from Annotation import overlap
>>> bed1=Bed(["chr1",10000,12000])
>>> bed2=Bed(["chr1",9000,13000])
>>> print overlap(bed1,bed2)
True
```

Annotation. Subtype (bed1, genebed)

This function determines intron or exon or utr from a BED12 file.

Parameters

- **bed1** A Bed object defined by xplib.Annotation.Bed (BAM2X)
- genebed A Bed12 object representing a transcript defined by xplib Annotaton.Bed with information of exon/intron/utr from an BED12 file

Returns str – RNA subtype. "intron"/"exon"/"utr3"/"utr5"/"."

Example:

```
>>> from xplib.Annotation import Bed
>>> from xplib import DBI
>>> from Annotation import Subtype
>>> bed1=Bed(["chr13",40975747,40975770])
>>> a=DBI.init("../../Data/Ensembl_mm9.genebed.gz","bed")
>>> genebed=a.query(bed1).next()
>>> print Subtype(bed1,genebed)
"intron"
```

Annotation.annotation(bed, ref_allRNA, ref_detail, ref_repeat)

This function is based on overlap() and Subtype() functions to annotate RNA type/name/subtype for any genomic region.

Parameters

- **bed** A Bed object defined by xplib.Annotation.Bed (in BAM2X).
- ref_allRNA the DBI.init object (from BAM2X) for bed6 file of all kinds of RNA
- ref_detail the DBI.init object for bed12 file of lincRNA and mRNA with intron, exon, UTR
- **genome** the Genome object from PyCogent.

Returns list of str – [type,name,subtype]

Example:

```
>>> from xplib.Annotation import Bed
>>> from xplib import DBI
>>> from Annotation import annotation
>>> bed=Bed(["chr13",40975747,40975770])
>>> ref_allRNA=DBI.init("../../Data/all_RNAs-rRNA_repeat.txt.gz","bed")
>>> ref_detail=DBI.init("../../Data/Ensembl_mm9.genebed.gz","bed")
>>> ref_repeat=DBI.init("../../Data/mouse.repeat.txt.gz","bed")
>>> print annotation(bed,ref_allRNA,ref_detail,ref_repeat)
["protein_coding","gcnt2","intron"]
```

5.2 "annotated bed" data class

```
class data structure.annotated bed(x=None, **kwargs)
```

To store, compare, cluster for the genomic regions with RNA annotation information. Utilized in the program Select_stronginteraction_pp.py

Cluster(c)

Store cluster information of self object

Parameters c – cluster index

Example:

```
>>> a=annotated_bed(chr="chr13", start=40975747, end=40975770)
>>> a.Cluster(3)
>>> print a.cluster
3
```

Note: a.cluster will be the count information when a become a cluster object in *Select_stronginteraction_pp.py*

Update (S, E)

Update the upper and lower bound of the cluster after adding segments using Union-Find.

Parameters

- \mathbf{S} start loc of the newly added genomic segment
- E end loc of the newly added genomic segment

Example:

```
>>> a=annotated_bed(chr="chr13", start=40975747, end=40975770)
    >>> a.Update(40975700,40975800)
    >>> print a.start, a.end
    40975700 40975800
___init__ (x=None, **kwargs)
    Initiation example:
    >>> str="chr13 40975747
                                      40975770
                                                       +
                                                                ATTAAG...TGA
                                                                                  protein_coding gcnt
    >>> a=annotated_bed(str)
    >>> a=annotated_bed(chr="chr13",start=40975747,end=40975770,strand='+',type="protein_coding"
 _lt___(other)
    Compare two objects self and other when they are not overlapped
        Parameters other - another annotated_bed object
        Returns boolean – "None" if overlapped.
    Example:
    >>> a=annotated_bed(chr="chr13", start=40975747, end=40975770)
    >>> b=annotated_bed(chr="chr13", start=10003212, end=10005400)
    >>> print a>b
    False
__str__()
    Use print function to output the cluster information (chr, start, end, type, name, subtype, cluster)
    Example:
    >>> str="chr13 40975747
                                      40975770
                                                        ATTAAG...TGA
                                                                       protein_coding gcnt2
    >>> a=annotated_bed(str)
    >>> a.Cluster(3)
    >>> a.Update(40975700,40975800)
    >>> print a
    "chr13 40975700
                                              protein_coding gcnt2 intron 3"
                              40975800
overlap (other)
    Find overlap between regions
        Parameters other – another annotated_bed object
        Returns boolean
```

Note: RNA-Stitch-seq-tools benifits a lot from BAM2X, a convenient python interface for most common NGS datatypes. Try BAM2X now!

CHAPTER

SIX

UPDATES

New features in 0.2.2:

- "Plot_interaction.py" function to plot local RNA-RNA interactions.
- "intersectInteraction.py" function to call overlap between two independent interaction sets.

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