Sequence analysis

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Basic4Cseq: an R/Bioconductor package for analyzing 4C-seq data

Carolin Walter^{1,*}, Daniel Schuetzmann², Frank Rosenbauer² and Martin Dugas¹

¹Institute of Medical Informatics, University of Münster, 48149 Münster, Germany, and ²Institute of Molecular Tumorbiology, University of Münster, 48149 Münster, Germany

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ABSTRACT

Summary: Basic4Cseq is an R/Bioconductor package for basic filtering, analysis and subsequent near-cis visualization of 4C-seq data. The package processes aligned 4C-seq raw data stored in binary alignment/map (BAM) format and maps the short reads to a corresponding virtual fragment library. Functions are included to create virtual fragment libraries providing chromosome position and further information on 4C-seq fragments (length and uniqueness of the fragment ends, and blindness of a fragment) for any BSGenome package. An optional filter is included for BAM files to remove invalid 4C-seq reads, and further filter functions are offered for 4C-seq fragments. Additionally, basic quality controls based on the read distribution are included. Fragment data in the vicinity of the experiment's viewpoint are visualized as coverage plot based on a running median approach and a multi-scale contact profile. Wig files or csv files of the fragment data can be exported for further analyses and visualizations of interactions with other programs.

Availability and implementation: Basic4Cseq is implemented in R and available at http://www.bioconductor.org/. A vignette with detailed descriptions of the functions is included in the package.

Contact: Carolin.Walter@uni-muenster.de

Supplementary information: Supplementary data are available at *Bioinformatics* online.

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1 INTRODUCTION

Circular chromosome conformation capture combined with high-throughput sequencing (4C-seq) is a method that allows the identification of chromosomal interactions between one potential interaction partner, called viewpoint, and virtually any other part of the genome (Gheldof *et al.*, 2012). Two rounds of digestion and ligation with two distinct restriction enzymes are performed to create a 4C-seq library. Unlike regular next-generation sequencing data, valid 4C-seq reads can only originate from precisely defined points in the genome. Any 4C-seq read will map to the end of a so-called 4C-seq fragment, i.e. a genomic region flanked by two primary restriction sites. When a secondary restriction site is not present, a fragment is called blind; a fragment end is defined as the region between a primary restriction site and the nearest secondary restriction site (van de Werken *et al.*, 2012a). Blind fragments are expected to yield

less reads than non-blind fragments; the length of a fragment end in general can cause variations in the read count. Any repetitive fragment end is to be interpreted with caution (van de Werken et al., 2012a). Consequently, these properties have to be considered during the analysis of 4C-seq data to prevent biases or misinterpretation. van de Werken et al.'s 4Cseqpipe is a pipeline to analyze 4C-seq data, which respects these issues (van de Werken et al., 2012a). However, it is not embedded into the R/Bioconductor environment, has little transparency regarding internal data structures, limited output options and less flexibility for smaller interactions with a length of up to 10 kb. The package r3Cseq (Thongjuea et al., 2013) works on regular binary alignment/map (BAM) files and can detect and visualize interactions, but does not differentiate between fragment types, which is a source of bias. The recently published fourSig method (Williams et al., 2014) includes fragment filtering options and detects different types of interactions, but does not provide more complex visualization routines for the data profile in the viewpoint region. In contrast, Basic4Cseq offers routines for the analysis of 4C-seq fragment data that are easy to customize. Provided functionality includes fragment filtering, quality control, explorative near-cis visualization and data export/import functions.

2 AVAILABLE FUNCTIONALITY

2.1 Preprocessing and virtual fragment library creation

Basic4Cseq expects 4C-seq data stored in BAM format. SAMtools (Li et al., 2009) and BEDtools (Quinlan and Hall, 2010) can be used to remove reads on very short or blind fragments directly from the BAM file. Basic4Cseq uses BSgenome packages to split a given genome at primary restriction sites. The resulting fragments are scanned for the presence of a secondary restriction site, and the ends of each fragment are checked for uniqueness. Fragment data are stored as a virtual fragment library file, which can be applied for 4C-seq experiments with the same underlying genome, restriction enzyme combination, and read length.

2.2 Filtering of 4C-seq data

Owing to sequencing errors or SNPs, the alignment of raw 4C-seq data needs to allow a certain number of mismatches. However, mismatches in the first restriction enzyme sequence can lead to mapped reads outside the predefined 4C-seq fragment ends. Overlaps of these reads with valid fragment ends may

^{*}To whom correspondence should be addressed.

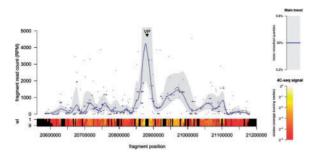


Fig. 1. Coverage plot based on a running median approach of RPM-normalized fetal liver sample data around the experiment's viewpoint (annotated as 'VP') at the Myb promoter, with an additional multiscale contact intensity profile. Fragment end-based resolution increases the visibility of small interactions. The corresponding raw data were taken from Stadhouders *et al.* (2012)

distort the true 4C-seq signal. Basic4Cseq can optionally discard any read with mismatches in the restriction enzyme sequence, if this sequence is present. Additional filtering options are available for the virtual fragment library. Very short or long fragments as well as blind fragments can be discarded; all repetitive fragment ends are removed per default to prevent ambiguous data.

2.3 Visualization

After preprocessing, the aligned short reads are mapped to the virtual fragment library. Basic4Cseq analyzes per default data from non-blind, unique fragment ends to prevent bias. It provides a visualization of the viewpoint region similar to the pipeline from van de Werken et al. (2012a): Fragment-based data are reads per million (RPM)-normalized and smoothed to counter the effects of single, overrepresented fragment ends. Smoothing options include a running median and a running mean approach; quantiles are further smoothed and interpolated with R's loess function. Fragments adjacent to the viewpoint of the 4C-seq experiment are removed to prevent bias through overrepresented sequences caused by self-ligation. Basic4Cseq visualizes both raw fragment data and running medians, plus quantile data in a single plot. Additionally, multi-scale contact profiles of the chosen fragment data are created for running median or running mean windows with varying window sizes. Read counts per fragment are normalized to [0, 1], and the resulting intensity values are expressed on a log 2 scale to allow for better visibility of distant interactions. Basic4Cseq can annotate custom regions of interest for easier interpretation and comparison of experiments. Import and near-cis visualization of fragment-based data (e.g. fourSig fragment data, which are easily convertible) is possible as well. Figure 1 shows a near-cis interaction profile with the typical high coverage at the experiment's viewpoint. Further peaks suggest high contact intensities between the underlying genomic regions and the viewpoint. Comparisons between filter options and alternative tools are added as supplement.

For regions located either more remote from the viewpoint or on other chromosomes, it is advisable to use a statistical enrichment approach for the analysis of the 4C-seq fragment data (Splinter *et al.*, 2012). Basic4Cseq can export filtered 4C-seq data as csv or wig files for use with other programs, e.g. Splinter *et al.*'s domainogram and spider-plot functions (Splinter *et al.*, 2012). Additionally, a visualization routine based on RCircos (Zhang *et al.*, 2013) is included for imported trans interaction data intervals.

2.4 Read distribution and quality control

The distribution of reads throughout 4C-seq fragment ends can provide information on the quality of the experiment data (van de Werken *et al.*, 2012b). Basic4Cseq provides the following quality control statistics: number of total reads, cis/overall ratio of reads and the percentage of covered fragment ends within a certain distance around the experiment's viewpoint. Reference values for high-quality experiments are >1 million reads total, a cis/overall ratio of >40% and a large fraction of covered fragment ends in the viewpoint's vicinity (van de Werken *et al.*, 2012b).

3 CONCLUSION

Basic4Cseq enables users to analyze 4C-seq experimental data in the R/Bioconductor environment. Quality control statistics can be calculated, and virtual fragment libraries with relevant fragment data generated. Functions for near-cis visualizations are included; both coverage profiles and heatmap-like multiscale contact intensity visualizations allow the exploration of contact profiles around the viewpoint. The package allows users to import fragment data for visualization, and to export filtered 4C-seq reads as csv or wig files for visualization or further analysis of significant interactions.

Conflicts of Interest: none declared.

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