

HiCzin: Normalizing metagenomic Hi-C data and detecting spurious contacts using zero-inflated negative binomial regression: Supplementary Material

Yuxuan Du¹, Sarah Laperriere², Jed Fuhrman² and Fengzhu Sun ^{*1}

¹Department of Quantitative and Computational Biology, University of Southern California, USA

²Department of Biological Sciences, University of Southern California, USA

Materials

Metagenomic yeast (M-Y) samples are synthetic microbial samples with 16 yeast strains and 13 yeast species. Raw WGS and Hi-C libraries were constructed by Metaphase [1]. WGS dataset contains 85.7 million read pairs (101 bp per read) and Hi-C dataset contains 81 million read pairs (100 bp per read).

Initial processing

We applied a standard cleaning pipeline on both WGS and Hi-C datasets using bbdduk from the BBTools suite(v37.25) [3]. Adaptor sequences were removed by bbdduk with parameter ‘ktrim=r k=23 mink=11 hdist=1 minlen=50 tpe tbo’ and reads were quality-trimmed using bbdduk with parameter ‘trimq=10 qtrim=r ftm=5 minlen=50’. We recommend using softwares(e.g. DeconSeq [4]) to remove noise from human if necessary [5]. Then, the first 10 nucleotides of each read were trimmed by bbdduk with parameter ‘ftl=10’.

Shotgun assembly and Hi-C read alignment

For the shotgun dataset, de novo metagenome assembly was produced by MEGAHIT [6] with parameters ‘-min-contig-len 300 -k-min 21 -k-max 141 -k-step 12 -merge-level 20,0.95’ and contigs shorter than 1 kb were discarded.

For the Hi-C dataset, only paired reads were kept for the downstream analysis. All PCR optical and tile-edge duplicates for Hi-C paired-end reads were removed by ‘clumpify.sh’ from BBTools suite [3] with default parameters. Processed Hi-C paired-end reads were mapped to assembled contigs using BWA-MEM [7] with parameters ‘-5SP’. Then, samtools [8] with parameters ‘view -F 0x904’ were applied on the resulting BAM files to remove unmapped reads (0x4) and supplementary (0x800) and secondary (0x100) alignments. Alignments with low quality (<30 nucleotide match length or mapping score <30) were also filtered out.

By this means, 4,700,202 read pairs were mapped to different contigs for the synthetic M-Y samples.

*Corresponding author: fsun@usc.com

Contact map generation

As contact map reflects the proximity distance within contigs, only pairs of reads aligned on different contigs were kept so as to generate the contact map. Raw contig-contig interactions were aggregated as contacts by counting the number of alignments linking two contigs. Contigs that no Hi-C reads were aligned to were discarded.

Annotating contigs by reference genomes for the M-Y samples

In order to explore experimental biases, the reference genomes of 16 yeast strains were downloaded (Table S1). As analysis was performed at the species level, the genomes of four strains (FY, CEN.PK, RM11-1A and SK1) from the same species (*S. cerevisiae*) were combined into one reference genome. Then, all contigs were aligned to those 13 reference genomes of all known species by BLASTn [9] with parameters: ‘-perc_identity 95 -evalue 1e-30 -word_size 50’. Hence, the true species that assembly contigs came from were determined if there existed any alignment of the contigs to the species’ reference genome; the placement of the alignment was ignored [1].

Evaluation criteria of clustering

Fowlkes-Mallows scores:

The Fowlkes-Mallows score (FM) is defined as the geometric mean of the pairwise precision and recall, i.e.,

$$FM = \sqrt{\frac{TP}{TP+FP} \cdot \frac{TP}{TP+FN}}, \quad (S1)$$

where TP is the number of true positives, FP is the number of false positives, and FN is the number of false negatives.

Adjusted Rand Index:

Define rand index (RI) as a measure of the percentage of correct decisions made by the clustering algorithm, i.e.,

$$RI = \frac{TP+TN}{TP+TN+FP+FN}, \quad (S2)$$

where TP is the number of true positives, TN is the number of true negatives, FP is the number of false positives, and FN is the number of false negatives.

Then, the Adjusted Rand Index (ARI) can be defined as

$$ARI = \frac{RI - E(RI)}{\max(RI) - E(RI)}. \quad (S3)$$

Normalized Mutual Information:

Let U and V denote the class labels and cluster labels. Define the entropy of a label set S as

$$H(S) = - \sum_{i=1}^{|S|} P(i) \log(P(i)), \quad (S4)$$

where $P(i) = |S_i|/N$ is the probability of an object in class S_i .

The mutual information (MI) between U and V is calculated by:

$$\text{MI}(U, V) = \sum_{i=1}^{|U|} \sum_{j=1}^{|V|} P(i, j) \log\left(\frac{P(i, j)}{P(i) + P'(j)}\right), \quad (\text{S5})$$

where $P(i, j) = |U_i \cap V_j|/N$, $P(i) = |U_i|/N$, and $P'(j) = |V_j|/N$.

Then, the Normalized Mutual Information (NMI) is defined as

$$\text{NMI}(U, V) = \frac{2 \times \text{MI}(U, V)}{H(U) + H(V)}. \quad (\text{S6})$$

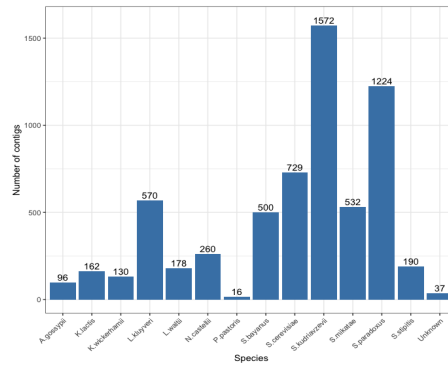


Figure S1: Number of assembled contigs from 13 species.

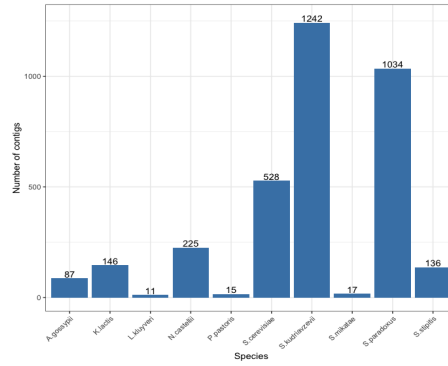


Figure S2: Number of contigs labeled by TAXAssign.

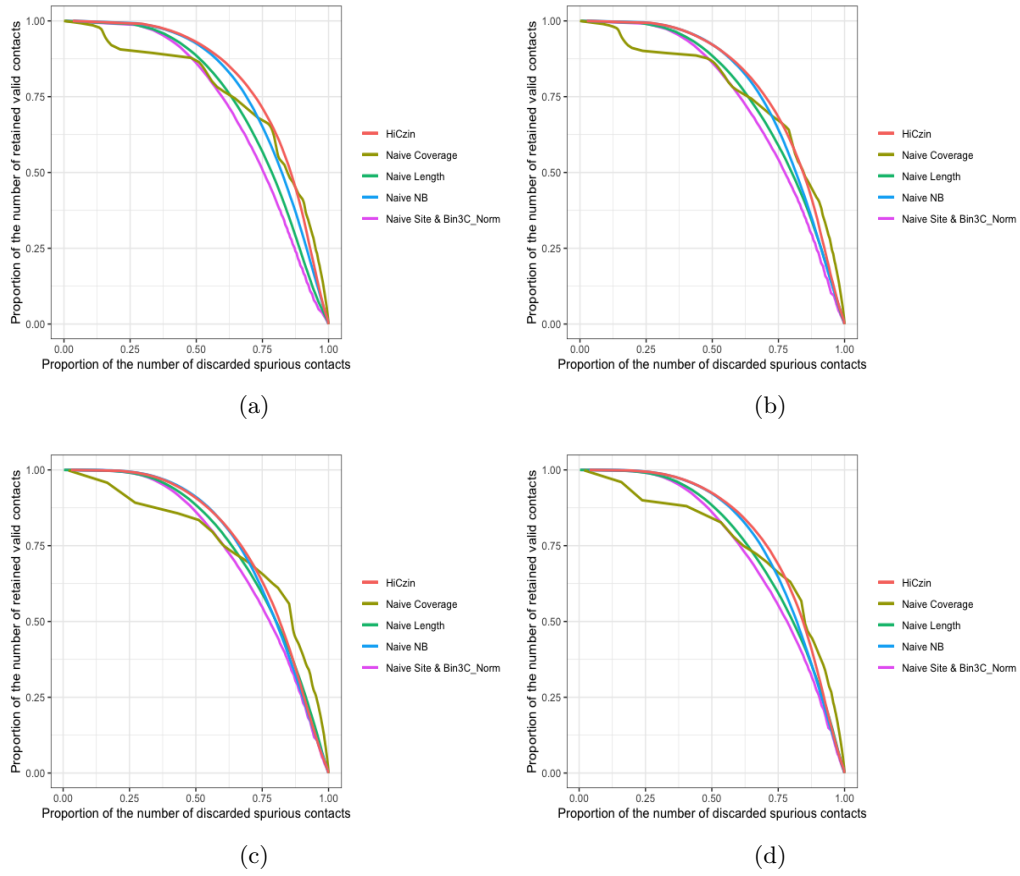


Figure S3: (a) DR curve using 6 species and 2156 contigs; (b) DR curve using 5 species and 1122 contigs; (c) DR curve using 4 species and 594 contigs; (d) DR curve using 3 species and 369 contigs.

Table S1: M-Y species list in the sample.

Genus	Species	Strain in sample	Reference strain
Saccharomyces	cerevisiae	FY4H	FY
Saccharomyces	cerevisiae	CEN.PK	CEN.PK
Saccharomyces	cerevisiae	RM11-1A	RM11-1A
Saccharomyces	cerevisiae	SK1	SK1
Saccharomyces	paradoxus	YDG613	
Saccharomyces	mikatae	FM356	IFO 1815
Saccharomyces	kudriavzevii	FM527	IFO 1802
Saccharomyces	bayanus var. uvarum	YZB5-113	CBS 7001
Naumovozyma	castellii	4310	NRRL Y-12630
Lachancea	waltii	Kwaltii ura3	NRRL Y-8285
Lachancea	kluyveri	FM628	CBS 3082
Kluyveromyces	lactis	MW98-8C	NRRL Y-1140
Kluyveromyces	wickerhamii	Y-8286	UCD 54-210
Ashbya	gossypii	WT	ATCC 10895
Scheffersomyces	stipitis	Y-11545	CBS 6054
Pichia	pastoris	JC308	GS115

Table S2: Number of remaining species, number of remaining contigs and proportion of assigned contigs in each step.

# of remaining species	# of remaining contigs	Proportion of assigned contigs
6	2156	34%
5	1122	18%
4	594	10%
3	369	6%

Note: # of remaining species and remaining contigs mean the number of remaining species and remaining contigs.

Table S3: Pearson correlation coefficients(absolute value) between normalized valid contacts and the product of each of the three factors of explicit biases, and AUDRC for different proportions of labeled contigs.

Proportion of labeled contigs	site	length	coverage	AUDRC
34%	0.004	0.006	0.059	0.801
18%	0.003	0.005	0.099	0.793
10%	0.011	0.010	0.065	0.771
6%	0.014	0.014	0.037	0.794

Table S4: Pearson correlation coefficients(absolute value) between normalized valid contacts and the product of each of the three factors of explicit biases, and AUDRC for different generalized HiCzin models.

	site	length	coverage	AUDRC
HiCzin_LC	0.006	0.002	0.097	0.812
HiCzin_GC	0.008	0.003	0.131	0.816
Unlabeled HiCzin	0.114	0.105	0.079	

References

- [1] Joshua N Burton, Ivan Liachko, Maitreya J Dunham, and Jay Shendure. Species-level deconvolution of metagenome assemblies with hi-c-based contact probability maps. *G3: Genes, Genomes, Genetics*, 4(7):1339–1346, 2014.
- [2] Maximilian O Press, Andrew H Wiser, Zev N Kronenberg, Kyle W Langford, Migun Shakya, Chien-Chi Lo, Kathryn A Mueller, Shawn T Sullivan, Patrick SG Chain, and Ivan Liachko. Hi-c deconvolution of a human gut microbiome yields high-quality draft genomes and reveals plasmid-genome interactions. *bioRxiv*, page 198713, 2017.
- [3] Brian Bushnell. Bbmap: a fast, accurate, splice-aware aligner. Technical report, Lawrence Berkeley National Lab.(LBNL), Berkeley, CA (United States), 2014.
- [4] Robert Schmieder and Robert Edwards. Fast identification and removal of sequence contamination from genomic and metagenomic datasets. *PloS one*, 6(3):e17288, 2011.
- [5] Eitan Yaffe and David A Relman. Tracking microbial evolution in the human gut using hi-c reveals extensive horizontal gene transfer, persistence and adaptation. *Nature Microbiology*, 5(2):343–353, 2020.
- [6] Dinghua Li, Chi-Man Liu, Ruibang Luo, Kunihiko Sadakane, and Tak-Wah Lam. Megahit: an ultra-fast single-node solution for large and complex metagenomics assembly via succinct de bruijn graph. *Bioinformatics*, 31(10):1674–1676, 2015.
- [7] Heng Li. Aligning sequence reads, clone sequences and assembly contigs with bwa-mem. *arXiv preprint arXiv:1303.3997*, 2013.
- [8] Heng Li, Bob Handsaker, Alec Wysoker, Tim Fennell, Jue Ruan, Nils Homer, Gabor Marth, Goncalo Abecasis, and Richard Durbin. The sequence alignment/map format and samtools. *Bioinformatics*, 25(16):2078–2079, 2009.
- [9] Stephen F Altschul, Warren Gish, Webb Miller, Eugene W Myers, and David J Lipman. Basic local alignment search tool. *Journal of molecular biology*, 215(3):403–410, 1990.