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yacrd and fpa: upstream tools for long-read genome assembly

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Abstract

Motivation: Genome assembly is increasingly performed on long, uncorrected reads. Assembly quality may be degraded due to unfiltered chimeric reads; also, the storage of all read overlaps can take up to terabytes of disk space. **Results:** We introduce two tools, <code>yacrd</code> and <code>fpa</code>, to respectively perform chimera removal/read scrubbing, and filter out spurious overlaps. We show that <code>yacrd</code> results in higher-quality assemblies and is two orders of magnitude faster than the best available alternative.

Availability: https://github.com/natir/yacrd and https://github.com/natir/fpa

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Supplementary information: Supplementary data are available online.

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1 INTRODUCTION

Third-generation DNA sequencing (PacBio, Oxford Nanopore) is increasingly becoming a go-to technology for the construction of reference genomes (*de novo* assembly). New bioinformatics methods for this type of data are rapidly emerging.

Some long-read assemblers perform error-correction on reads prior to assembly. Correction helps reduce the high error rate of third-generation reads and make assembly tractable, but is also a time and memory-consuming step. Recent assemblers (e.g. Li (2016); Ruan and Li (2019) among others) have found ways to directly assemble raw uncorrected reads. Here we will therefore focus only on **correction-free assembly**. In this setting, assembly quality may become affected by e.g. chimeric reads and highly-erroneous regions (Myers, 2015), as we will see next.

The DASCRUBBER program (Myers, 2017) introduced the concept of read "scrubbing", which consists of quickly removing problematic regions in reads without attempting to otherwise correct bases. The idea is that scrubbing reads is a more lightweight operation than correction, and is therefore suitable for high-performance and correction-free genome assemblers.

DASCRUBBER performs all-against-all mapping of reads and constructs a pileup for each read. Mapping quality is then analyzed to determinate putatively high error rate regions, which are replaced by equivalent and higher-quality regions from other reads in the pileup. MiniScrub (LaPierre et al., 2018) is another scrubbing tool that uses a modified version of Minimap2 (Li, 2017) to record positions of the anchors used in overlap detection. For each read, MiniScrub converts anchors positions to an image. A convolutional neural network then detects and removes of low quality read regions.

Another problem that is even more upstream of read scrubbing is the computation of overlaps between reads. The storage of overlaps is disk-intensive and to the best of our knowledge, there has never been an attempt at optimizing its potentially high disk space.

In this paper we present two tools that together optimize the early steps of long-read assemblers. One is yacrd (for Yet Another Chimeric Read Detector) for fast and effective scrubbing of reads, and the other is fpa (for Filter Pairwise Alignment) which filters overlaps found between reads.

2 MATERIALS & METHODS

Similarly to DASCRUBBER and MiniScrub, **yacrd** is based on the assumption that low quality regions in reads are not well-supported by other reads. To detect such regions yacrd performs all-against-all read mapping using Minimap2 and then computes the base coverage of each read. Contrarily to DASCRUBBER and MiniScrub, yacrd only uses approximate positional mapping information given by Minimap2, which avoids the time-expensive alignment step. This comes at the expense of not having base-level alignments, but this will turn out to be sufficient for performing scrubbing. Reads are split at any location where coverage drops below a certain threshold (set to 4 by default), and the low-coverage region is removed entirely. A read is completely discarded if less than 40% of its length is below the coverage threshold. yacrd time complexity is linear in the number of overlaps.

yacrd performance is directly linked to the overlapper performance. We tuned Minimap2 parameters (especially the maximal distance between two minimizers, -g parameter) to find similar regions between reads and not to create bridges over low quality regions (see Supplementary Section 3). yacrd takes reads and their overlaps as input, and produces scrubbed reads, as well as a report.

fpa operates between the overlapper and the assembler. It filters out overlaps based on a highly customizable set of parameters, such

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		H. sapiens chr1 (ONT ultra-long R9.4)			C. elegans (Pacbio P6-C4)		
		raw	dascrubber	yacrd	raw	dascrubber	yacrd
Reads	# reads	1,075,867	819,798	1,044,848	740,776	660,766	751,750
	Relative # bases	1.00	0.71	0.80	1.00	0.84	0.84
	N50	10,568	9,858	9,520	16,572	15,667	15,845
	# chimera	25,888	6 %	20 %	71,704	13 %	21 %
	Time		3 days 2 hours	27 mins		1 day 20 hours	33 mins
Miniasm	# contigs	184	184	394	226	131	154
	NGA50	96,225	410,37	453,748	432,112	544,677	440,776
	Asm/Ref size	81 %	78 %	81 %	113 %	108 %	110 %
	# misassemblies	1,745	209	432	1,396	754	1,015
Wtdbg2	# contigs	810	496	485	139	100	122
	NGA50	1,513,450	545,902	1,482,513	565,278	578,041	593,039
	Asm/Ref size	87 %	80 %	84 %	106 %	104 %	106 %
	# misassembly	1,316	177	582	614	485	577

Table 1. Performance of yacrd compared to DASCRUBBER on an ONT and a PacBio dataset. Relative #bases indicates the proportion of raw read bases kept after scrubbing. # chimera indicates the number of chimeric reads detected in the dataset using Minimap2 (see Supplementary Section 4) and the proportion of remaining chimeric reads after scrubbing. NGA50 is the N50 of aligned contigs, and # misassemblies are the number of misassemblies, both metrics were computed by QUAST (Gurevich et al., 2013). Asm/Ref size indicates the relative length of the assembly divided the reference length.

as overlap length, length of reads names, etc. fpa can remove self-overlaps, end-to-end overlaps, containment overlaps, internal matches (when e.g. two reads share a repetitive region) as defined in (Li, 2016). fpa supports the PAF or BLASR m4 formats as inputs and outputs, with optional compression. fpa can also rename reads, generate an index of overlaps and output an overlap graph in GFA format

yacrd and fpa are evaluated on several datasets (details provided in Supplementary Section 1), and here we highlight their performance on two of them: *H. sapiens* chromosome 1 Oxford Nanopore (ONT) ultra-long reads, and *C. elegans* PacBio reads. All tools were run on a single cluster node with recommended parameters (see Supplementary Section 2). Scrubbed reads were then assembled using both Miniasm and Wtdbg2 with recommended parameters for each sequencing technology.

3 RESULT & DISCUSSION

Table 1 compares the results of **yacrd** and DASCRUBBER. We also evaluated MiniScrub (see Supplementary Section 2 and 5), but its memory usage exceeded 256 GB on the two datasets of Table 1.

The main feature of yacrd is that its total execution time, which is essentially that of Minimap2, is two orders of magnitude faster than DASCRUBBER. We next evaluate whether running yacrd results in higher-quality reads and assemblies. yacrd removes 20-27% of the bases in raw reads, comparably to DASCRUBBER. Both scrubbers significantly reduce chimeras: only 6-13% of those in raw reads remain with DASCRUBBER and 18-20% with yacrd. The impact of removing chimeras is directly seen on assembly metrics: both scrubbers produce significantly less misassemblies with Miniasm and Wtdbg2 than with direct assembly of raw reads. Both yacrd and DASCRUBBER resulted in increased contiguity (NGA50) with Miniasm, and equivalent (or significantly degraded for DASCRUBBER) contiguity with Wtdbg2, and comparable assembly lengths.

On ONT reads, DASCRUBBER reduces the number of misassemblies by a factor of 2-3 more than yacrd. However,

given that all assemblies in Table 1 completed in less than an hour and DASCRUBBER took 3 days, running this tool on larger datasets would become a significant performance bottleneck. In Supplementary Section 3 we examine the behavior of yacrd across its parameter space. We observe that different parameters worked best for different datasets, one of which is actually a parameter for Minimap2.

fpa reduced the size of reads overlap file (PAF file produced by Minimap2) by 40-79% on the evaluated datasets, without any significant effect on quality assembly. As a consequence this reduces the memory usage of Miniasm by 13-67%. Other performance metrics are presented in Supplementary Table 5.

Finally, we examine the effect of combining both yacrd and fpa. We propose a pipeline based on Miniasm (Supplementary Section 7) and show that it results in improved assembly contiguity, comparable assembly size, less mismatches and indels, less misassemblies, at the cost of a reasonable increase in running time (around 2x).

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