

Assembly of long, error-prone reads using repeat graphs

Mikhail Kolmogorov, Jeffrey Yuan, Yu Lin, and Pavel A. Pevzner July 2, 2021

Presented by: Johannes Hausmann, Luis Kress

 \bullet Assembly: reconstruct target sequence from the reads

- Assembly: reconstruct target sequence from the reads
- Different assemblers, different graph structures

- Assembly: reconstruct target sequence from the reads
- Different assemblers, different graph structures
- ullet Repeats o assembly fragmentation



Figure 1: Tangled assembly graph of *E.coli*[1]

- Assembly: reconstruct target sequence from the reads
- Different assemblers, different graph structures
- ullet Repeats o assembly fragmentation



Figure 1: Tangled assembly graph of *E.coli*[1]

 \bullet Small differences between repeat copies \to hard to resolve with error-prone reads

• Most assemblers spend much time on correct contig assembly

- Most assemblers spend much time on correct contig assembly
- Flye uses a different approach [1]:

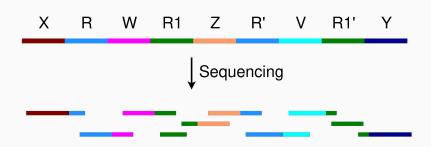
- Most assemblers spend much time on correct contig assembly
- Flye uses a different approach [1]:
 - we don't care about the repeats

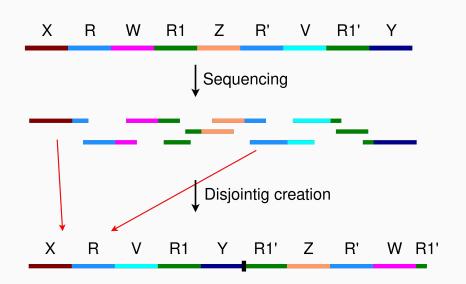


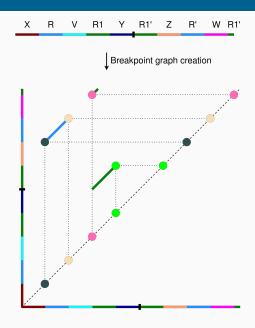
- Most assemblers spend much time on correct contig assembly
- Flye uses a different approach [1]:
 - we don't care about the repeats (at least at the initial stage)

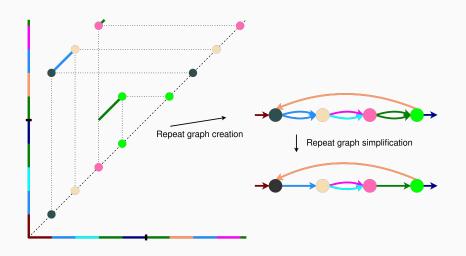


ullet Generate paths from overlapping reads without checking for correct repeat resolution o Disjointigs









Results

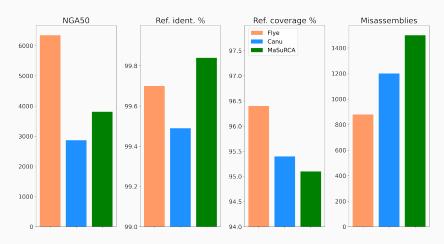


Figure 2: Results for HUMAN testset

References i

- M. Kolmogorov, J. Yuan, Y. Lin, and P. A. Pevzner. **Assembly of long, error-prone reads using repeat graphs.** *Nature Biotechnology*, 37(5):540–546, May 2019.
- Y. Lin, S. Nurk, and P. A. Pevzner.

 What is the difference between the breakpoint graph and the de Bruijn graph?

 BMC genomics, 15 Suppl 6:S6, 2014.
- P. A. Pevzner, P. A. Pevzner, H. Tang, and G. Tesler.

 De novo repeat classification and fragment assembly.

 Genome Research, 14(9):1786–1796, Sept. 2004.

References ii



B. J. Walker, T. Abeel, T. Shea, M. Priest, A. Abouelliel, S. Sakthikumar, C. A. Cuomo, Q. Zeng, J. Wortman, S. K. Young, and A. M. Earl.

Pilon: an integrated tool for comprehensive microbial variant detection and genome assembly improvement. PloS One, 9(11):e112963, 2014.

Git (presentation and poster)





Assembly of long, error-prone reads using repeat graphs

Mikhail Kolmogorov, Jeffrey Yuan, Yu Lin, and Pavel A. Pevzner July 2, 2021

Presented by: Johannes Hausmann, Luis Kress

Appendix

Dot plot creation

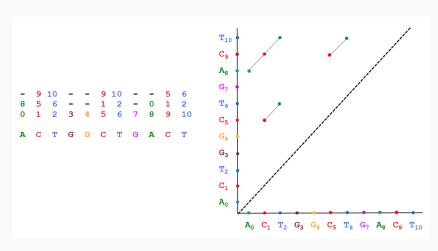


Figure 3: Dot plot creation

Repeat graphs

- generalization of de bruijn graphs
- structure
- creation
 - from disjointigs = random walk of reads on the repeat graph
 - means the repeat graph hasn't to be known

Difference repeat graph de Bruijn graph

- A-Bruijn graph (alignments) generalizes the de Bruijn graph
- "We thus argue that the time has come to explain that the breakpoint graphs and the de Bruijn graphs are two identical data structures (if one ignores a cosmetic difference between them) as they both represent specific instances of a general notion of the A-Bruijn graph introduced in [2]. The A-Bruijn graphs are based on representing genomes as sets of labeled paths and further gluing identically labeled edges (breakpoint graphs) or vertices (de Bruijn graphs) in the resulting paths." [3]
- de Bruijn graphs need correct bases
- otherwise tangled graph

Segmental duplications

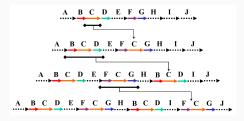


Figure 4: Segmental Duplications

- Segmental duplications are duplicated blocks of genomic DNA typically ranging in size from 1-200 kb (IHGSC 2001)
- They often contain sequence features such as high-copy repeats and gene sequences with intron-exon structure.

Repeat resolution

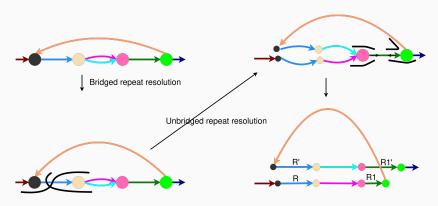


Figure 5: Repeat resolution of the given example

Repeat resolution

"Pilon is a software tool which can be used to automatically improve draft assemblies or to find variation among strains, including large event detection. Pilon requires as input a FASTA file of the genome along with one or more BAM files of reads aligned to the input FASTA file. Pilon uses read alignment analysis to identify inconsistencies between the input genome and the evidence in the reads. It then attempts to make improvements to the input genome, including" [4]:

- Single base differences
- Small indels
- Larger indel or block substitution events
- Gap filling
- Identification of local misassemblies, including optional opening of new gaps

Contigity improvement



Figure 6: Contigity improvements

 \bullet colors are contigs \to change in color means fragmentation