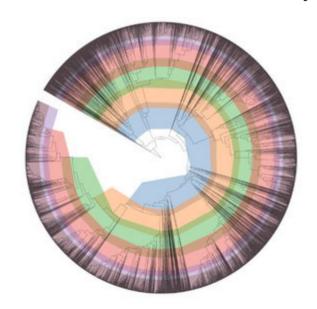
### **ENVIRONMENTAL METAGENOMICS**

Physalia course, online, 11-15 November 2024

Metagenome de-novo assembly and quality control

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NB: original course material courtesy: Dr. Antti Karkman, University of Helsinki

Dr. Igor Pessi, Finnish Environment Institute (SYKE)

## Typical analysis methods used in metagenomics

### 1) Alignment:







### 2) Classification:

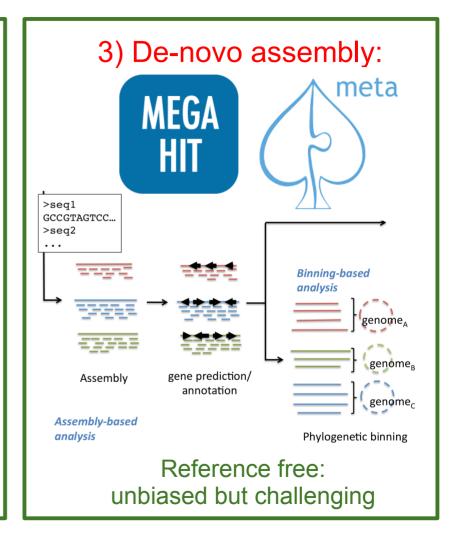


Centrifuge

MetaPhlan

Clark

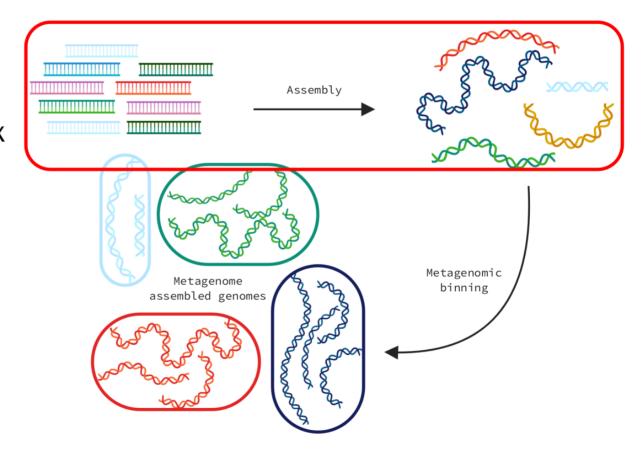
Reference based: assume similarity to reference



## De novo assembly

Assemble short nucleotide sequences into longer sequences by finding their overlap / consensus without reference genome

- No reference available
- Uneven and complex communities





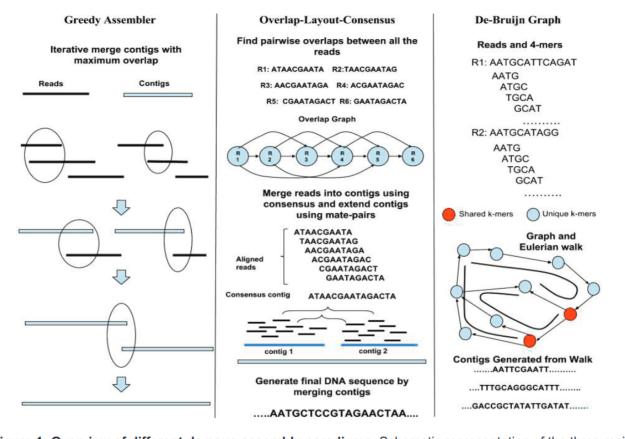
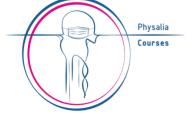


Figure 1: Overview of different de novo assembly paradigms. Schematic representation of the three main paradigms for genome assembly – Greedy, Overlap-Layout-Consensus, and de Bruijn. In Greedy assembler, reads with maximum overlaps are iteratively merged into contigs. In Overlap-Layout-Consensus approach, a graph is constructed by finding overlaps between all pairs of reads. This graph is further simplified and contigs are constructed by finding branch-less paths in the graph, and taking the consensus sequence of the overlapping reads implied by the corresponding paths. Contigs are further organized and extended using mate pair information. In de Bruijn graph assemblers, reads are chopped into short overlapping segments (k-mers) which are organized in a de Bruijn graph structure based on their co-occurrence across reads. The graph is simplified to remove artifacts due to sequencing errors, and branch-less paths are reported as contigs.



Ghurye et al., Yale Journal of Biology and Medicine, 2016

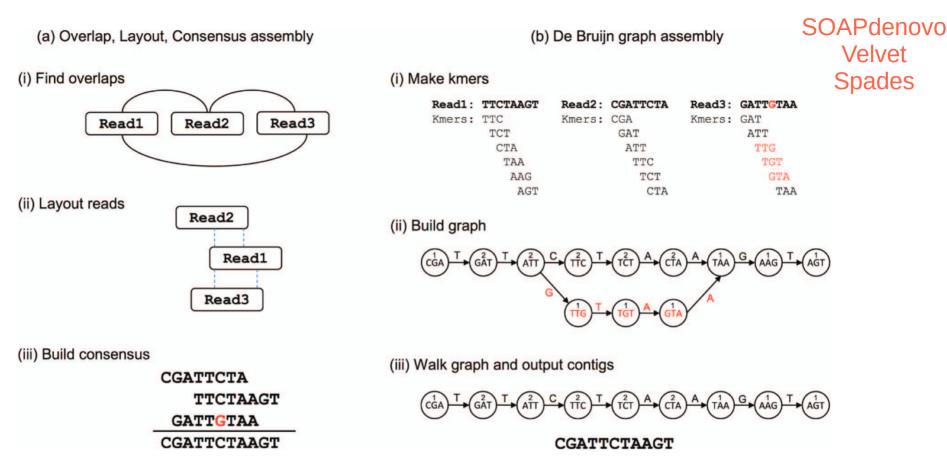


Figure 1. Two different approaches to genome assembly: (a) in Overlap, Layout, Consensus assembly, (i) overlaps are found between reads and an overlap graph constructed (edges indicate overlapping reads). (ii) Reads are laid out into contigs based on the overlaps (dashed lines indicate overlapping portions). (iii) The most likely sequence is chosen to construct consensus sequence. (b) In dBg assembly, (i) reads are decomposed into kmers by sliding a window of size k across the reads. (ii) The kmers become vertices in the dBg, with edges connecting overlapping kmers. Polymorphisms (red) form branches in the graph. A count is kept of how many times a kmer is seen, shown here as numbers above kmers. (iii) Contigs are built by walking the graph from edge nodes. A variety of heuristics handle branches in the graphs—for example, low coverage paths, as shown here, may be ignored.

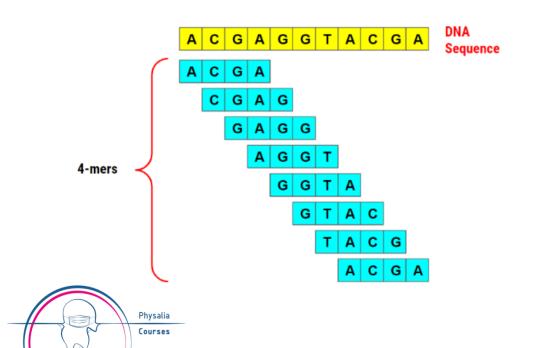
#### Popular de Bruijn graph de-novo metagenomic assemblers for short Illumina reads



We are going to use Megahit in the exercises

# De novo assembly using MEGAHIT

**MEGAHIT**: de Bruijn-graph assembler using a distribution of different k-mer lengths inferred from the length of the sequencing data



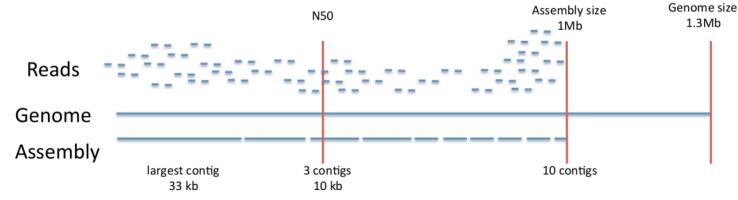
reasons for using MEGAHIT:

- low-memory footprint
- has little issues with the presence of ancient DNA damage
- works with single-end data

BUT: lower assembly quality than other assemblers for modern sequencing data (see CAMI II challenge; DOI: 10.1038/s41592-022-01431-4)

# **Assembly metrics**

- assembly size
- number of contigs, largest contig
- N50

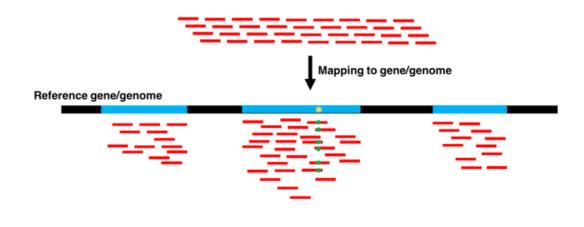




# Alignment against the contigs

Many of the following steps require the alignment of the short-read data against the de novo assembled contigs, e.g.

- correction of the contig sequences
- binning of the contigs into MAGs (coverage along the contigs)
- quantification of the presence of ancient DNA damage





## Taxonomic classification - on contig level

The likely taxonomic origin of contigs can be determined by aligning them against a

reference database.

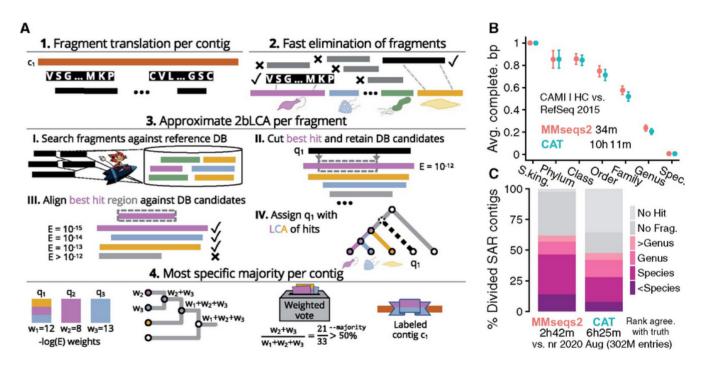
#### available aligners:

- BLAST/DIAMOND
- Kraken2
- Centrifuge
- MMSeqs2

#### available databases:

- NCBI NT/RefSeq
- GTDB





Mirdita et al. (Bioinformatics, 2021; doi: bioinformatics/btab184) Fig. 1