# Assembly of large genomes from short WGS reads

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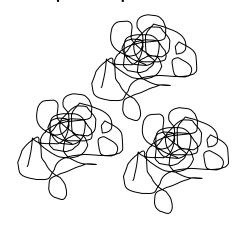




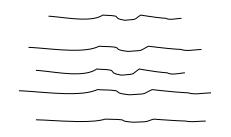


# Whole Genome Shotgun reads

#### Multiple copies of DNA

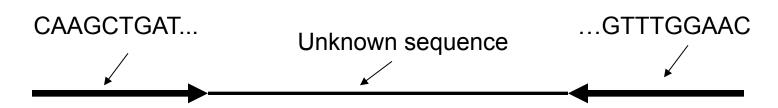


Shred & Size select Fragments of 150 - 200,000 bases



Sequence the ends

Pairs of Reads of 150 – 500 bases each





#### Recently developed assemblers for NGS data

- MSR-CA
  - handles 454, Illumina, and Sanger reads
- Allpaths-LG
- SOAPdenovo
- Velvet
- ABySS
- Contrail
- SGA (also overlap-based)



### Two assembly approaches

- Overlap-Layout-Consensus (OLC)
  - Used by most assemblers for previous generation (Sanger) sequencing
  - Celera Assembler, PCAP, Phusion, Arachne, etc.
- Graph
  - Used by most assemblers for Illumina data
  - SOAPdenovo, Allpaths-LG, Velvet, Abyss, etc
- We use a combined approach that combines the benefits of both OLC and Graph in our MSR-CA assembler



## Assembly approaches: OLC

- OLC: Overlap-Layout-Consensus
  - Compute <u>overlaps</u> of reads

Create <u>layout</u> of *contigs* from overlapping reads

AGTGATTAGATAGTAGA

AGATGATACTAGAGATAGATAGACC

ATAGTAGAGATAGATAGACCACTCATCATAC

Create <u>consensus</u> sequence of contigs

AGTGATTAGATGATAGAGAGATAGACCACTCATCATAC



## 5 billion reads?

# ... that's 12.5 quadrillion overlaps

at 1 million overlaps/second, that would take 400 years





### Two assembly approaches

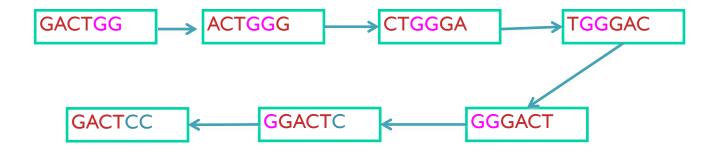
- Overlap-Layout-Consensus (OLC)
  - Used by most assemblers for previous generation (Sanger) sequencing
  - Celera Assembler, PCAP, Phusion, Arachne, etc
- DeBruijn Graph
  - Used by most assemblers for Illumina data
  - SOAPdenovo, Allpaths-LG, Velvet, Abyss, etc
- We use a combined approach that combines the benefits of both OLC and Graph in our MSR-CA assembler



# De Bruijn Graph strategy: find all k-mers, build graph

- Every k-mer becomes a node
- Two nodes are linked with an edge if they share a k-1 mer

#### **GACTGGGACTCC**

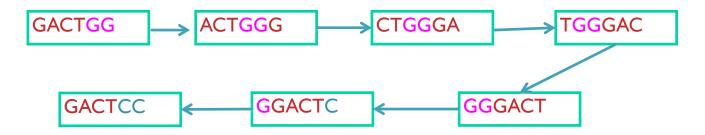




### De Bruijn Graph Strategy

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- An assembly is a path through the graph that visits each edge at least once
- We can only roughly estimate the graph of the genome from reads due to sequencing errors and lack of coverage



### Two assembly approaches

- Overlap-Layout-Consensus (OLC)
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  - Celera Assembler, PCAP, Phusion, Arachne, etc
- Graph
  - Used by most assemblers for Illumina data
  - SOAPdenovo, Allpaths-LG, Velvet, Abyss, etc
- We propose to use a combined approach that combines the benefits of both OLC and Graph in our MSR-CA assembler



# Benefits and drawbacks of OLC and Graph

#### Benefits of OLC

- Can deal with variable length reads and reads from different sequencing platforms
- Overlaps can be long and thus more reliable
- Overlaps do not have to be exact
- Can resolve repeats of up to read size
- Drawbacks of OLC
  - Computationally intensive, number of overlaps grows quickly with the number of reads and coverage

- Benefits of Graph
  - Computationally efficient
- Drawbacks of Graph
  - Errors in the reads create spurious branches in the graph – requires error correction
  - Max. size of k-mer is limited by the shortest read size
  - All overlaps in the graph are exact overlaps of k-1 bases
  - Repeats of longer than kbases cannot be resolved



#### MSR-CA combines benefits of OLC and Graph

#### Benefits of OLC

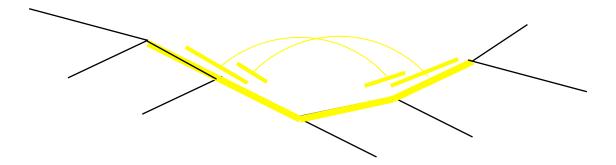
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#### **MSR-CA** strategy

- Error correct Illumina reads
- Create a deBruijn graph: each k-mer is unique in the graph



- Many pairs will extend to the same Super-read
- Replace the pairs by the corresponding Super-reads
- Assemble the Super-reads with an OLC assembler (CABOG)
  using additional long mate pairs for linking the contigs



### MSR-CA design

- Efficient multi-threaded code
- Designed to handle data sets with up to 12B reads
- Development is aimed at WGS assembly of the 24Gb Loblolly Pine genome on a computer with 48 cores and 512Gb of RAM in 1-2 months
- Current version 1.4 and being continually improved





## Results on a pool of 500 Pine fosmids

	Sequence in assembly, bp	N50 contig size, bp	N50 scaffold size	Number of scaffolds>30k b
Allpaths-LG	14,050,574	10,324	26,298	248
SOAPdenovo	13,470,572	1,632	33,389	322
MSR-CA	14,604,209	7,640	22,740	218

Notes: N50 computed from estimated 19.25Mb total sequence, each fosmid was estimated at 38Kb, assemblies used short pairs and jumping library pairs





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