Genome Assembly

CHRIS FIELDS

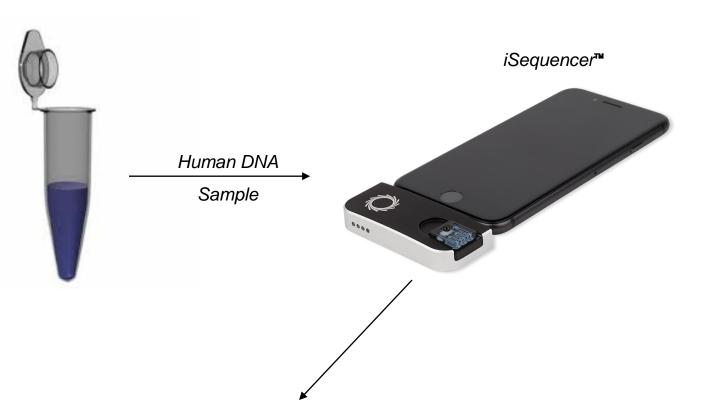
MAYO-ILLINOIS COMPUTATIONAL GENOMICS WORKSHOP JUNE 6, 2022

Overview

- What is a genome assembly?
- Sequencing technologies (2022)
- General steps in a genome assembly
- Planning an assembly project
- Assembly assessment
- Annotation

Ideal World!

I wouldn't need to give this talk!

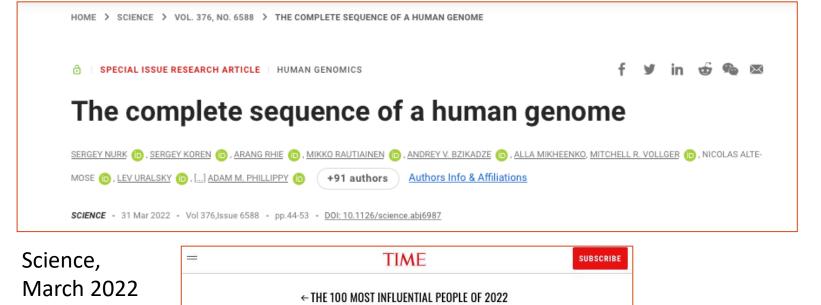


46 complete, haplotyperesolved, chromosome sequences

T. Seemann

Ideal World!

But we may not be too far from this!

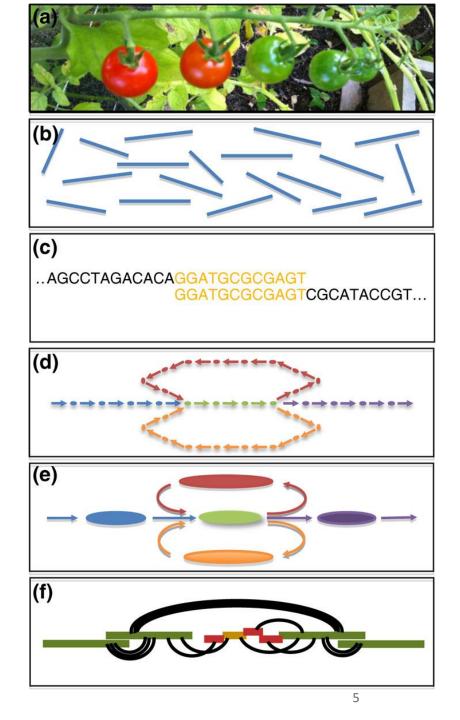


Michael Schatz, Karen Miga, Evan Eichler, and Adam Phillippy

Time, May 2022

- a) DNA is collected from the biological sample, fragmented, and sequenced.
- (b) Output from the sequencer consists of many millions/billions of (possibly short) unordered DNA fragments from random positions in the genome.
- c) Fragments are compared with each other in some way to discover how they overlap.
- d)The overlap relationships are captured in a large assembly graph
- e) The graph is refined to correct errors and simplify
- (f) Finally, additional information such as mates, markers and other longrange information can be used to order and orient the initial assembly (contigs) into large scaffolds

Schatz et al. Genome Biology 2012 13:243



Let's Do a Genome Assembly!

- Sequence a sample, and have the computer do the rest?
- How do you find overlaps between sequences (when you have millions to billions of them)?
 - You compare them all (overlapping pieces)
 - You find shorter perfectly overlapping segments
 - Faster but has a lot of assumptions!!!
- How do you store all this information?
- How long does it take?

Resource needs

- Technology dependent!
- Memory + CPU
- Short reads (billions of reads)
 - Sequencing costs \$\$
 - **Compute costs** \$\$\$\$\$
 - **Results** fragmented, requires significant 'cleanup'

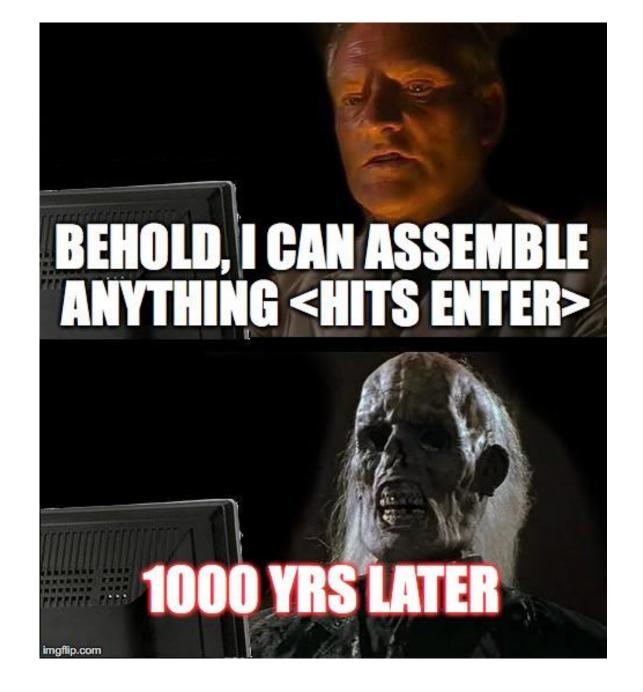
Long error-prone reads

- Sequencing costs \$\$\$\$
- **Compute costs** \$\$\$\$\$
- **Results** better quality, but can't easily phase

• Long accurate reads

- Sequencing costs \$\$\$\$\$
- **Compute costs** \$\$\$
- **Results** (partly) phased diploid assembly***

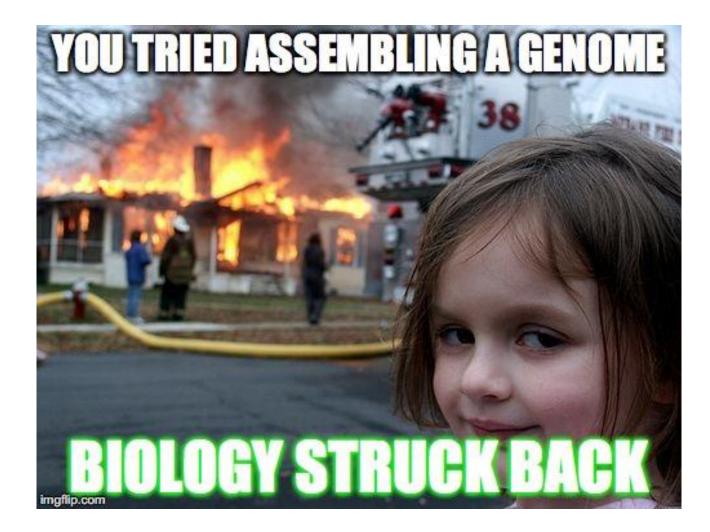
*** - doesn't help much if you have higher ploidy! (though this will likely change)



Results

- You spent your entire grant on getting sequence data and buy a monster multicore high-memory server
- You assemble your genome with your favorite genome assembly tool
- You waited a week to a month and you now have results!
- Wait, why do I have a million scaffolds? And why is my server on fire?!?

Biology



Current Sequencing Technologies

Illumina

Millions to billions of short but highly accurate reads (>99.9%)

Can be paired-end (sequence ends of fragments)

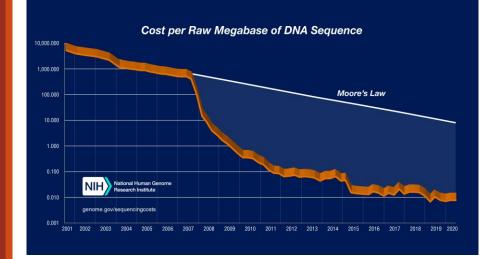
Advantages

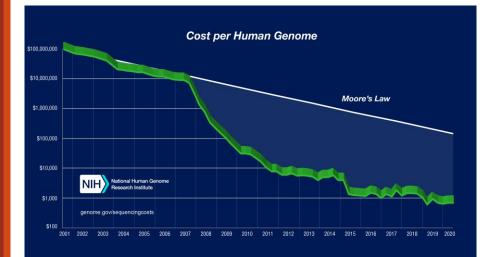
- Highly accurate (~99.9%)
- Relatively even coverage of the genome
- Well-vetted technology
- Most cost-effective, as low as \$10 per **billion** bases
- (Generally) robust to sample issues

Disadvantages

- Requires high depth for many applications (50x + for assembly)
- Sequence length (100-150nt reads) problematic for repeats
- Maximum fragment length (<800bp) is an issue

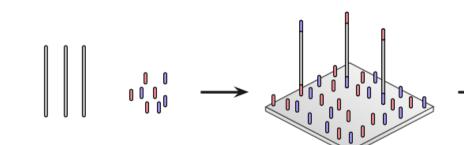


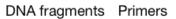




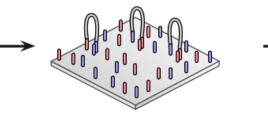
Illumina

https://www.atdbio.com/content/58/N ext-generation-sequencing

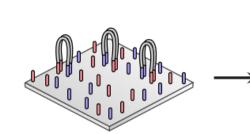


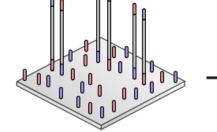


DNA strands are attached to cell surface at one end



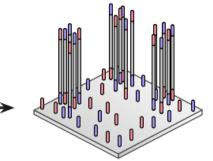
Ends are attached to surface by complimentary primers





Enzymes create double strands

Denaturation forms two separate DNA fragments



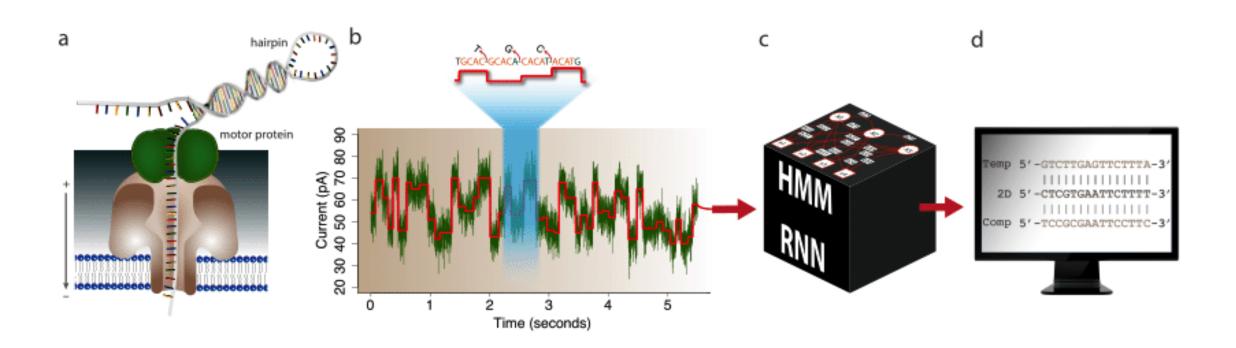
Repetition forms clusters of identical strands

'Long reads'

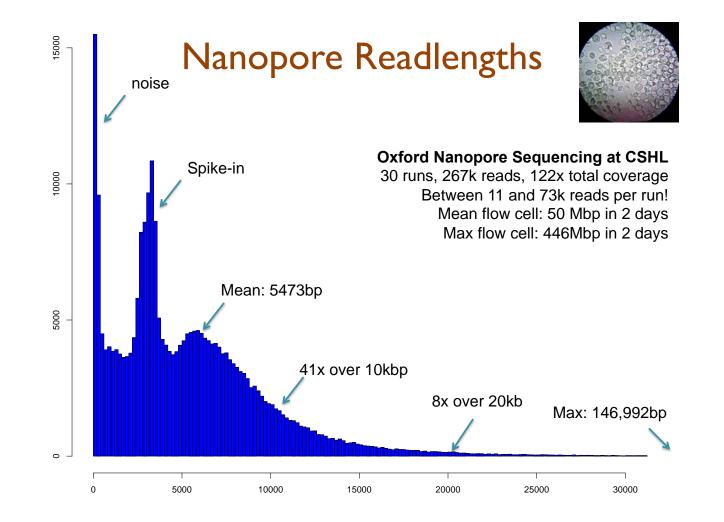
Pacific Biosciences (PacBio)

Oxford Nanopore (ONT)



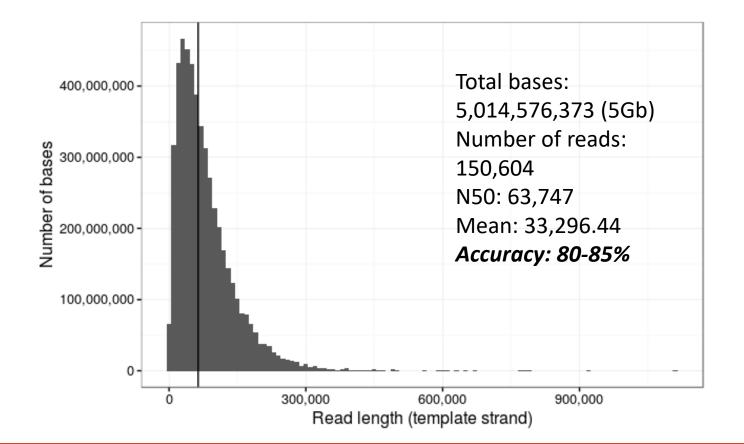


2016-2017



Whale watching: E. coli

2017



N. Loman, ASM Microbe 2017

http://lab.loman.net/2017/03/09/ultrareads-for-nanopore/

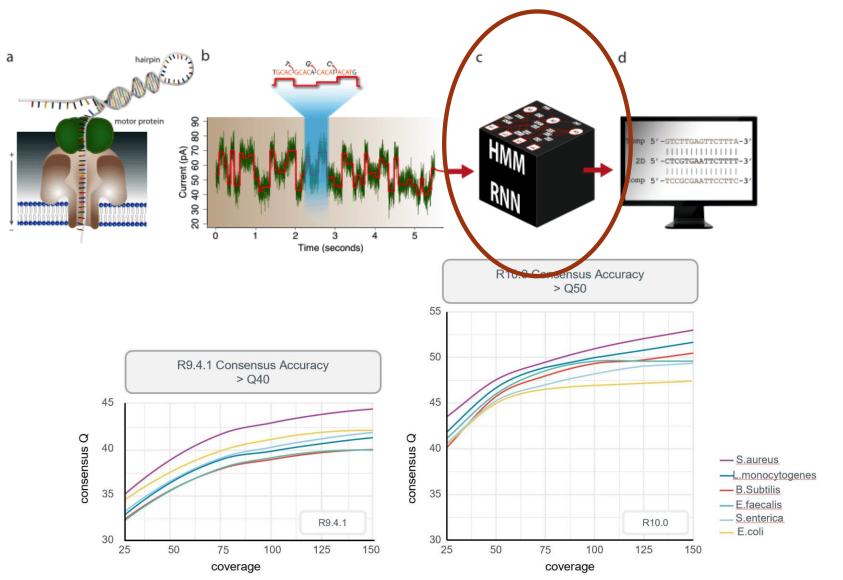
E. coli: genome assembly in 8 reads

1	to 4.641.65	52 (4.6 Mbp)				888,577 to 1,013,761 (125,2 K
	Read	Length	Ref start	Ref end	Time (m)	
	1	876991	4398844	634183	32.48	
	2	696402	470003	1166405	25.79	
	3	799047	1137438	1936485	29.59	miniasm
	4	642071	1759431	2401502	23.78	
	5	826662	2106227	2932889	30.61	N50 4Mb
	6	883962	2699626	3583588	32.73	Time: 1.5s (1 CPU)
	7	825191	3285196	4110387	30.56	
	8	463341	3995967	4459308	17.16	

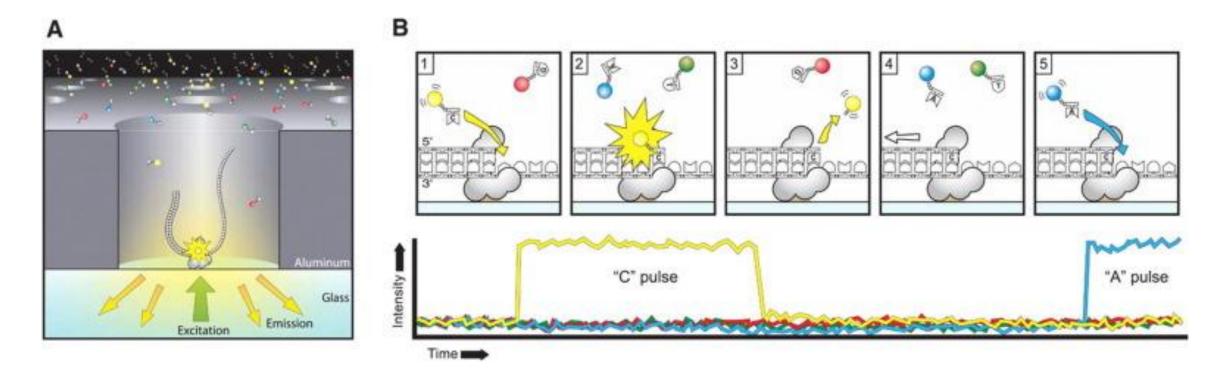
1x coverage!

N. Loman, ASM Microbe 2017 http://lab.loman.net/2017/03/09/ultrareads-for-nanopore/

2022



Note that accuracy is based on comparion to human data!



https://www.sciencedirect.com/science/article/pii/S1672022915001345?via%3Dihub

Pacific Biosciences

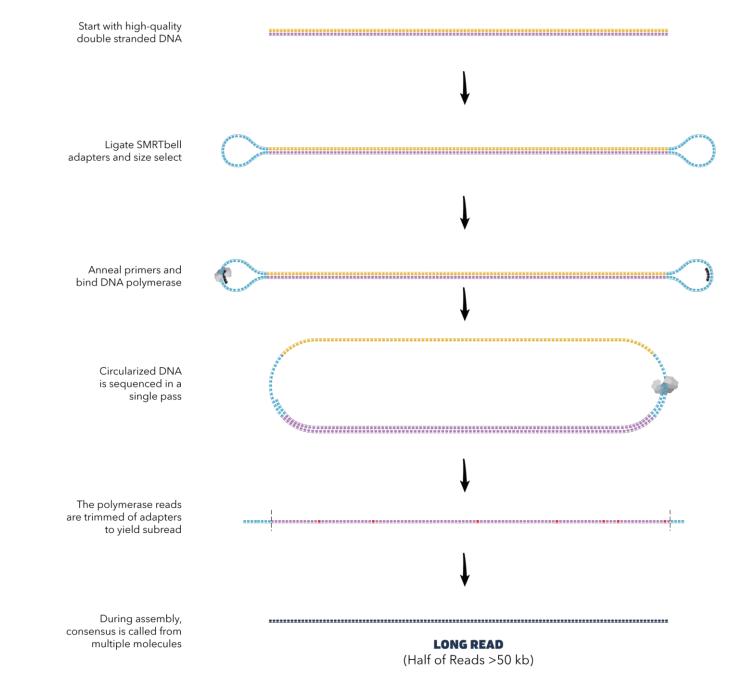
PacBio Continuous Long Read Sequencing (aka PacBio CLR)

Optimized for length

25-50kb long reads

90% accuracy

Yields of ~125Gb+ per SMRT cell
Need ~50-90x coverage
Needs error correction, polishing
1-2 SMRT cells per human sample



PacBio Circular Consensus Sequencing (aka PacBio HiFi)

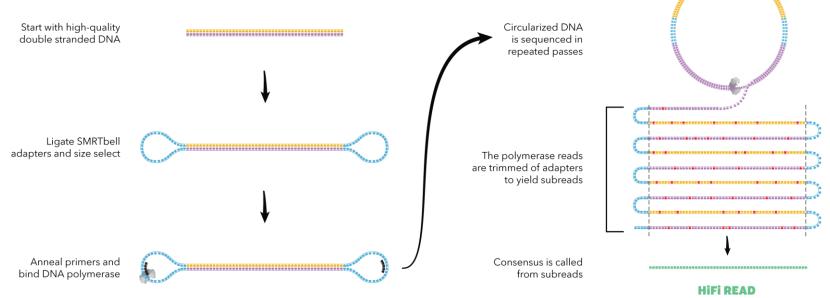
Optimized for accuracy

10-15kb long reads

99% accuracy

Yields of ~25Gb per SMRT cell Need ~25-50x coverage No error correction/polishing required

~2-3 SMRT cells per human sample



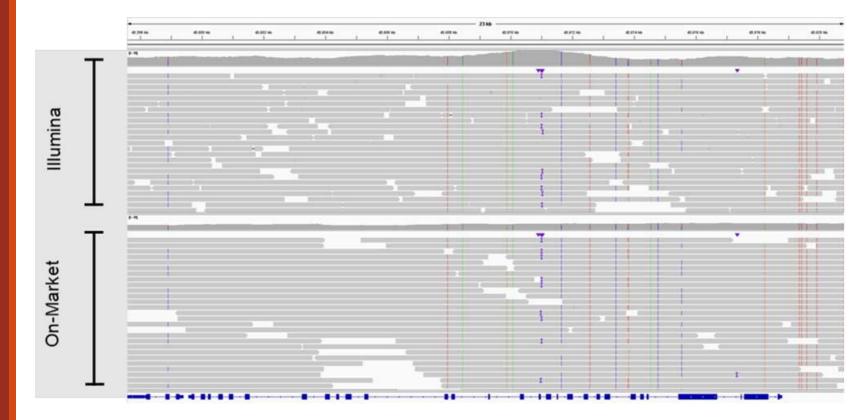
Illumina Infinity

Announced early 2022: "... we are developing a novel, high performance long read assay, code named 'Infinity' that will accelerate access to the remaining ~5% of genic regions that are challenging to map"

- Contiguous reads up to 10 kb
- ~10x the throughput compared to traditional long read technologies
- 90% less DNA input compared to current Long Read methods
- Fully automatable workflow

Early access 2H of 2022

Very little known about this one so far



Illumina Infinity announcement

'Long Reads'

Advantages

- Reads can be very long (1kb 100kb)
- Relatively even coverage of the genome
- (PacBio HiFi) Highly accurate (99%)
- (Oxford) real-time sequencing
- (Oxford) portable

Disadvantages

- Expensive compared to Illumina short reads
- Need very high quality, high MW DNA samples
- Least expensive options are *error-prone*
- Depending on technology, can have systematic errors (homopolymer issues), but getting better

Basic Steps for Genome Assembly

Steps

1. Basic DNA sequence cleanup and evaluation (pre-assembly)

2. Contig building

3. Scaffolding

4. Post-assembly processing and analyses

Basic cleanup and evaluation

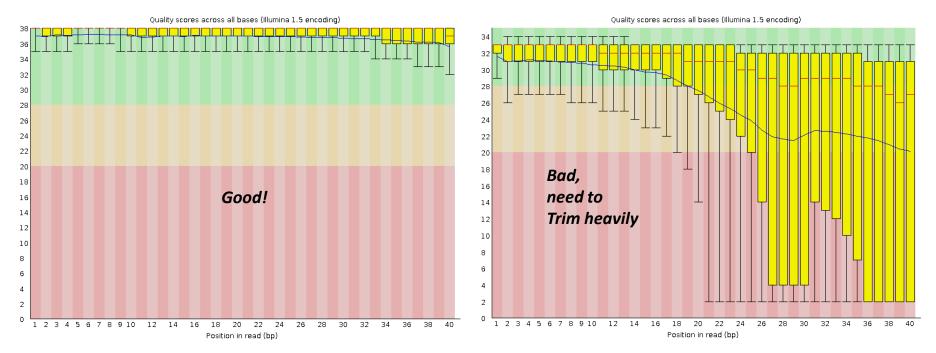
• Is the DNA sequence high quality?

• Does it need to be trimmed?

Evaluate libraries for read 'coverage'

Any additional sequence preparation steps

DNA Quality (FASTQC)



Illumina Data

https://www.bioinformatics.babraham.ac.uk/projects/fastqc/

Adapters

Overrepresented sequences

Sequence	Count	Percentage	Possible Source
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGATCT	8122	8.122	Illumina Paired End PCR Primer 2 (100% over 40bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGATCGGAAG	5086	5.086	Illumina Paired End PCR Primer 2 (97% over 36bp)
AATGATACGGCGACCACCGAGATCTACACTCTTTCCCTAC	1085	1.085	Illumina Single End PCR Primer 1 (100% over 40bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGGAAG	508	0.508	Illumina Paired End PCR Primer 2 (97% over 36bp)
AATTATACGGCGACCACCGAGATCTACACTCTTTCCCTAC	242	0.242	Illumina Single End PCR Primer 1 (97% over 40bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAAGATCGGAA	235	0.2350000000000000	Illumina Paired End Adapter 2 (96% over 31bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCGAGATCGGAAGA	228	0.227999999999999998	Illumina Paired End Adapter 2 (96% over 28bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGGACG	205	0.2050000000000002	Illumina Paired End PCR Primer 2 (97% over 36bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGGATCGGAA	183	0.183	Illumina Paired End Adapter 2 (100% over 32bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGGTCGGAAG	183	0.183	Illumina Paired End Adapter 2 (100% over 32bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGAACT	164	0.164	Illumina Paired End PCR Primer 2 (97% over 40bp)
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGGTCT	129	0.129	Illumina Paired End PCR Primer 2 (97% over 40bp)
AATTATACTTCTACCACCTATATCTACACTCTTTCCCTAC	123	0.123	No Hit
GATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGGACT	122	0.122	Illumina Paired End PCR Primer 2 (97% over 36bp)
CGGTTCAGCAGGAATGCCGAGATCGGAAGAGCGGTTCAGC	113	0.1129999999999999999	Illumina Paired End PCR Primer 2 (96% over 25bp)

https://www.bioinformatics.babraham.ac.uk/projects/fastqc/

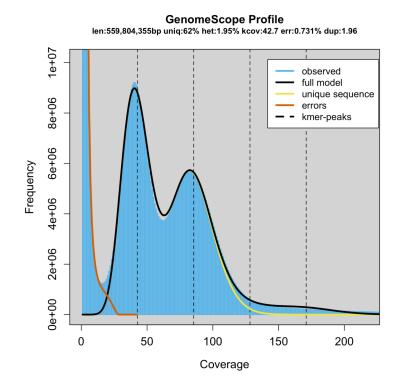
Other pre-assembly steps

Depending on the assembler and technology you use, you may want to:

Join paired-end reads

Assess reads for contaminants

• Error correction of reads (e.g. fix sequencing errors)



Starting the assembly

Contig building

Greedy assembly

Seed and extend

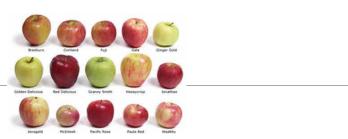
Overlap graph

de Bruijn graphs

String graphs

..etc etc

... all essentially doing similar things, but taking different 'shortcuts' based on needs

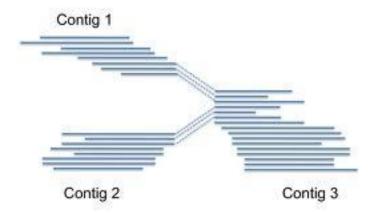


Contigs

Contiguous, unambiguous stretches of assembled DNA sequence

Contigs ends correspond to

- Real ends (for linear DNA molecules)
- Dead ends (missing sequence)
- Decision points (forks in the road)



Assembly recipe



Find all overlaps between reads

- hmm, sounds like a lot of work...

Build a graph

a picture of read connections

Simplify the graph

- sequencing errors will mess it up a lot

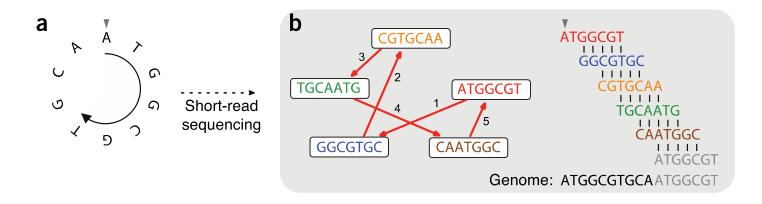
- Traverse the graph
 - trace a sensible path to produce a consensus

Graph

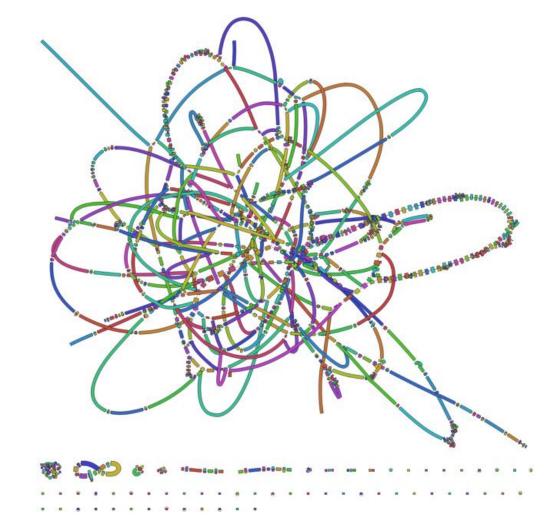
Review: A structure where objects are related to one another somehow

Nodes/Vertices = objects (sequence)

Edges = relationship (overlap)

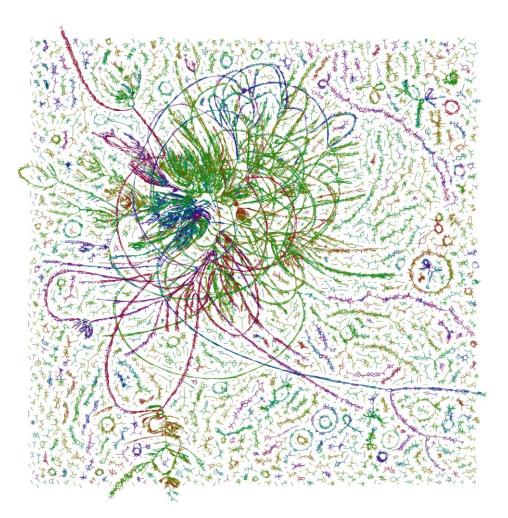


Compeau et al, Nature Biotech, 29(11), 2011; https://en.wikipedia.org/wiki/Graph_(discrete_mathematics)



https://github.com/rrwick/Bandage/wiki/Effect-of-kmer-size

Simple?



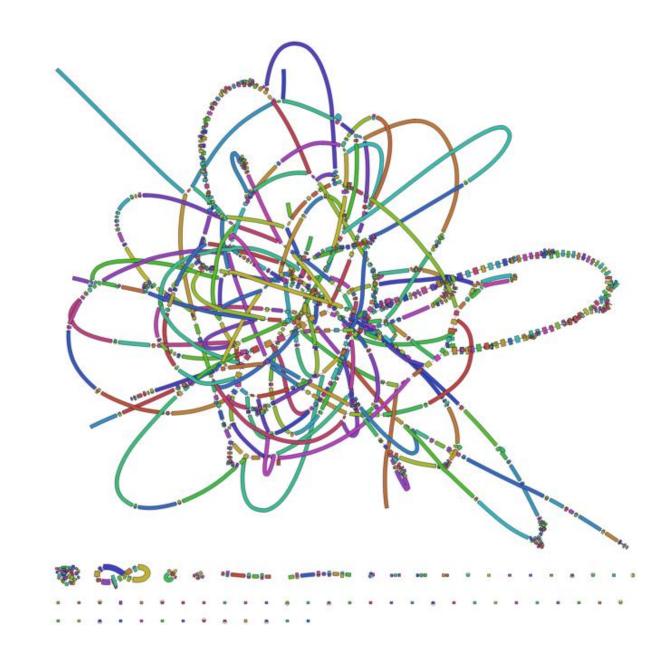
http://armbrustlab.ocean.washington.edu/seastar



In essence...

For each unconnected graph:

- **Find a path** which visits each node once
 - This is referred to as a Hamiltonian path/cycle
- Form consensus sequences from paths
 - use all the overlap alignments
 - each of these collapsed paths is a *contig*

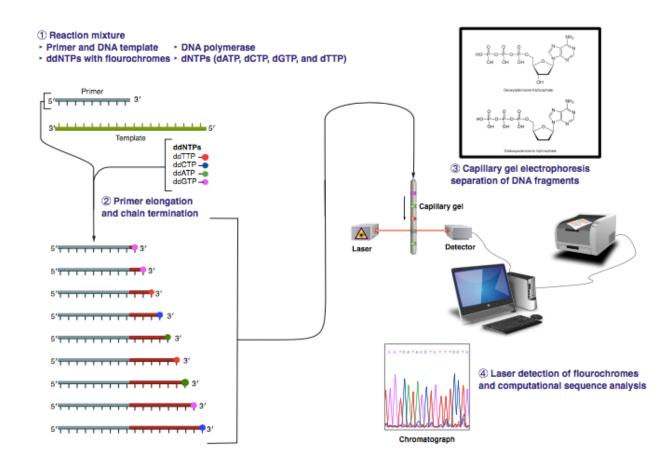


Overlap Layout Consensus Assembly

Used for longer read data

Sanger

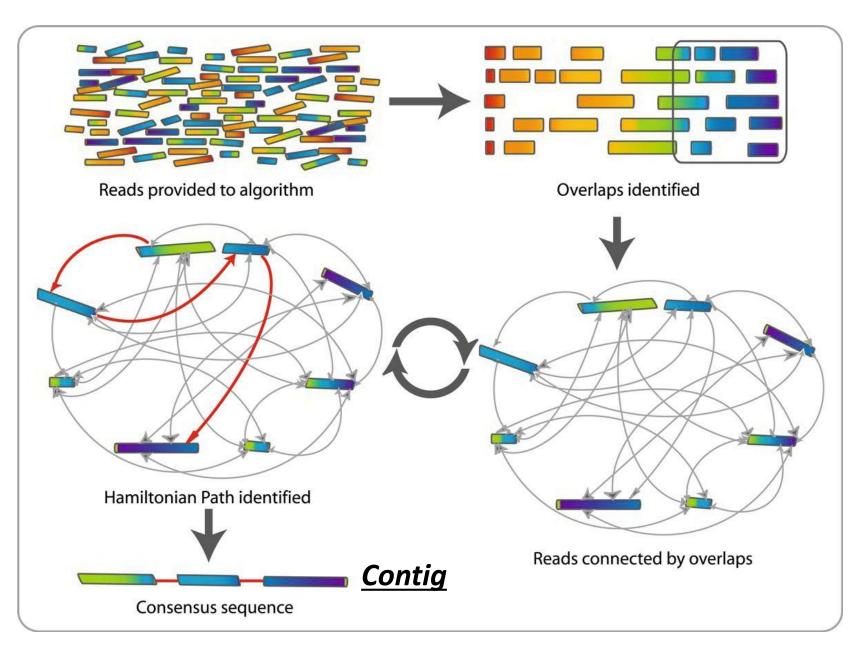
Newer variants for PacBio and Oxford Nanopore



By Estevezj - Own work, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=23264166 For each unconnected graph, at least one per replicon in original sample

Find a path which visits each node once

Form consensus Sequences from paths



OLC assembly steps

Calculate *overlays*

 Can use BLAST-like methods, but finding common strings (k-mers) more efficient

Assemble *layout* graph, try to simplify graph and remove nodes (reads) – find Hamiltonian path

Generate *consensus* from the alignments between reads (overlays)

Some OLC-based assemblers

Canu – is a fork of the Celera Assembler designed for high-noise single-molecule sequencing (PacBio, Oxford Nanopore)

HiCanu – PacBio HiFi assembler

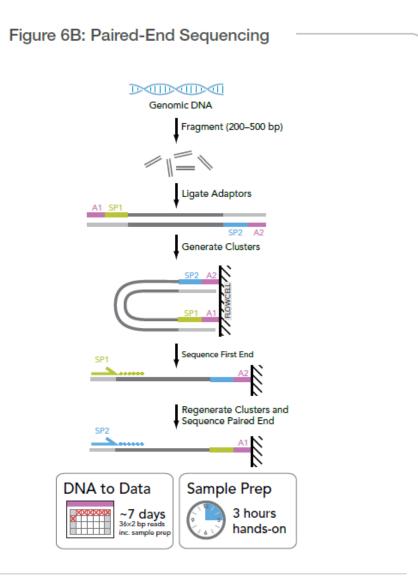
Newbler, a.k.a. GS de novo Assembler - designed for 454 sequences, but works with Sanger reads

Hifiasm – a hybrid *diploid* assembler

De Bruijn graph assemblers

Developed to deal with highthroughput highly accurate short-read data

Uses shotgun data (generally paired-end fragments of 300-500nt)



Adapters containing attachment sequences (A1 & A2) and sequencing primer sites (SP1 & SP2) are ligated onto DNA fragments (e.g., genomic DNA). The resulting library of single molecules is attached to a flow cell. Each end of every template is read sequentially.

Shredded Book Reconstruction

Dickens accidentally shreds the first printing of <u>A Tale of Two Cities</u>
 – Text printed on 5 long spools

It was the vas the bestinges intervas was was worstor softimes, it was the aggo of wisis domiti was he age of it do is softimes, ...

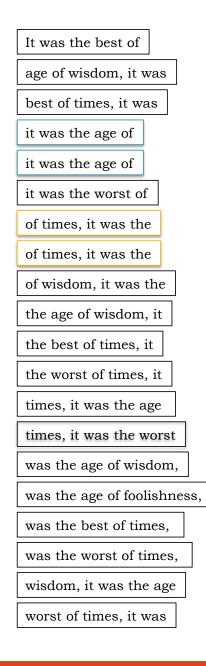
It was the vassthe bestimiesnits was the worst of times, it was the the agenois disd, out with the vase the fage is infostishness,

It was they state besimesinter as wascheoustrof times, it was the age of wisdom, it it as as they age of of is bolishness, ...

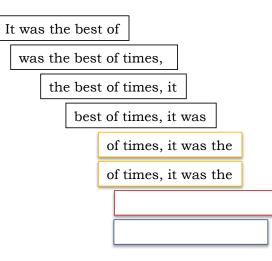
It was it the sorted of times, was abeliever strof times, it was the age of vivided onit, it availed age of sorted on the second strong age for the strong age of th

It | walt the shire of since s

- How can he reconstruct the text?
 - 5 copies x 138, 656 words / 5 words per fragment = 138k fragments
 - The short fragments from every copy are mixed together
 - Some fragments are identical



Greedy Reconstruction



The repeated sequence make the correct reconstruction ambiguous

• It was the best of times, it was the [worst/age]

Model the assembly problem as a graph problem

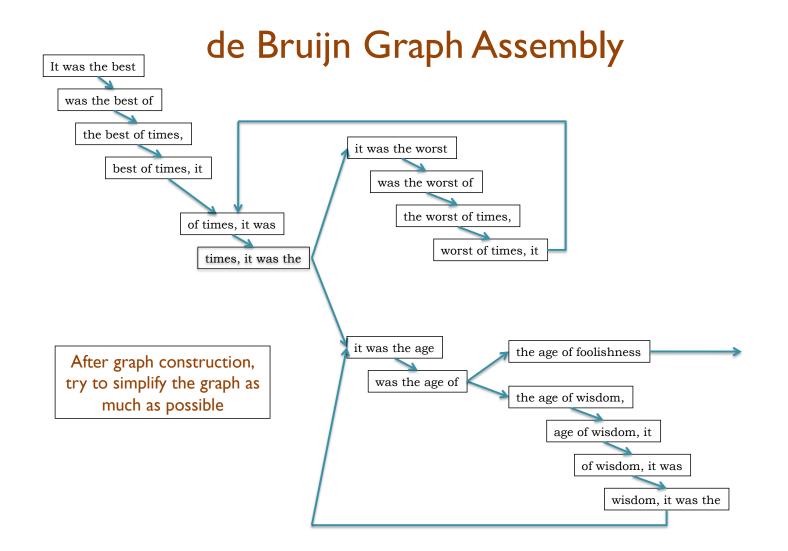
de Bruijn Graph Construction

- $D_k = (V, E)$
 - V = All length-k subfragments (k < l)
 - E = Directed edges between consecutive subfragments
 - Nodes overlap by k-1 words

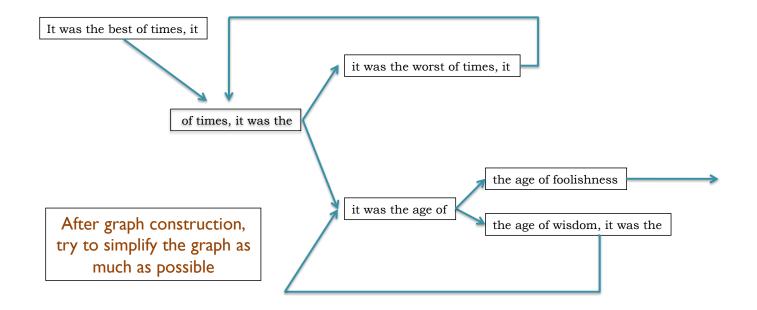


- Locally constructed graph reveals the global sequence structure
 - Overlaps between sequences implicitly computed

de Bruijn, 1946 Idury and Waterman, 1995 Pevzner, Tang, Waterman, 2001

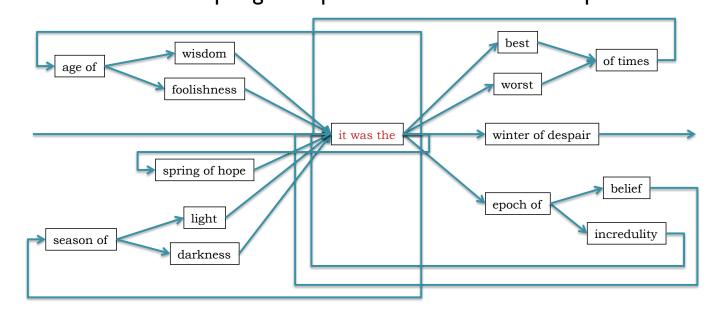


de Bruijn Graph Assembly



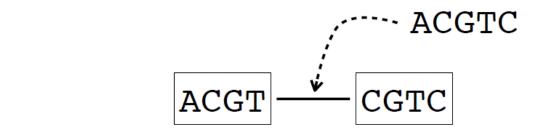
The full tale

... it was the best of times it was the worst of times ...
... it was the age of wisdom it was the age of foolishness ...
... it was the epoch of belief it was the epoch of incredulity ...
... it was the season of light it was the season of darkness ...
... it was the spring of hope it was the winder of despair ...



De Bruijn graphs - concept

- de Bruijn graph
 - k-dimensional graph over four symbols {A, C, G, T}
 - vertex: k-mer -- a string of k nucleotides
 - edge: (k+1)-mer



k=4

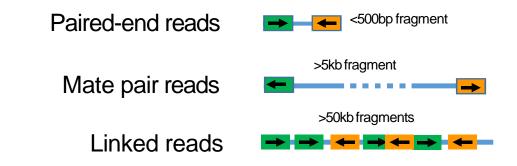
M. Schatz, Feb 2015 Course, JHU

Scaffolding

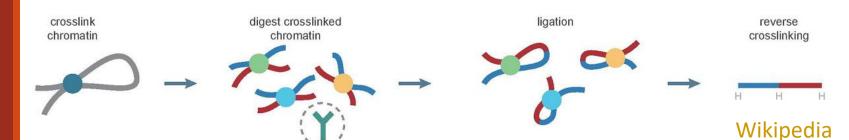
- Now, you have a huge pile of contigs but you want to make them larger. How?
- Add context!
- Link together contigs using *other* genomic information
 - Infer contigs position on the genome relative to one another

Linking Contigs via DNA Seq

Illumina sequencing



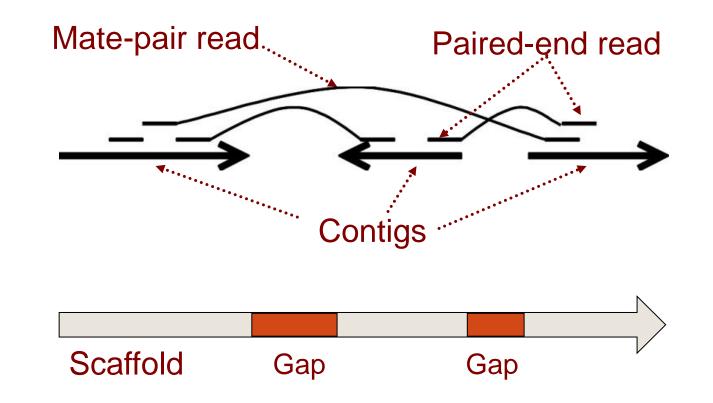
HiC (Chromosome Conformation Capture)



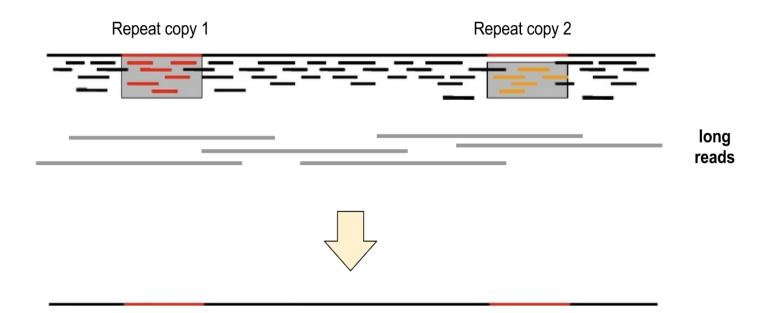
PacBio/ONT long-reads

10-100 kb+

Contigs to scaffolds



Long reads





HiC

Chromosome Conformation Technology

https://dovetailgenomics.com/omni-c/

<u>Wikipedia</u>



Omni-C[™] libraries start with endogenous chromatin.

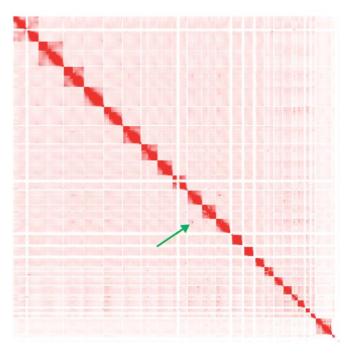
Crosslinking (red lines) the chromatin creates a stabilized nucleosome (blue circles) scaffold.

Non-specific endonuclease digests the cross-linked chromatin.

Proximity ligation with a biotin (green dots) tagged bridge between DNA ends (black lines) creates chimeric molecules (ex. 1 and 2).

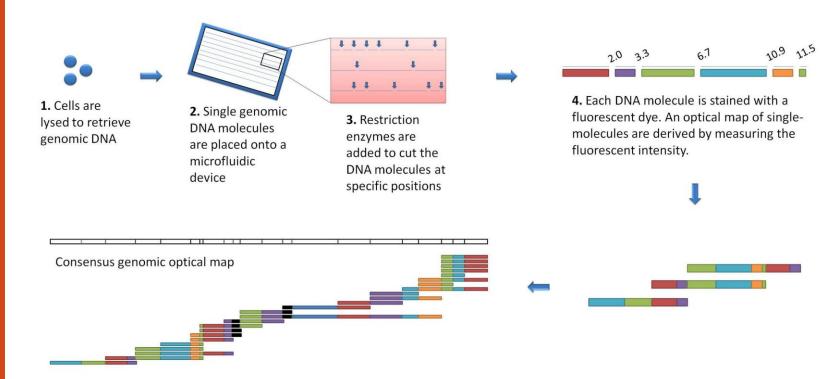
The crosslinks are reversed.

DNA is purified and enriched for ligation-containting chimeric molecules. Libraries sequenced as pair-end short reads.



Optical Mapping

Using high resolution single-molecule restriction mapping combined with fluorescent dyes and fluorescence microscopy to produce a genomic map



5. Overlapping of the multiple single-molecule maps gives us the consensus genomic optical map

Starting a new assembly project

Planning a genome sequencing project?

BUDGET!!!

- Technological costs
- Computational costs
- Person costs (time)!

Biology!

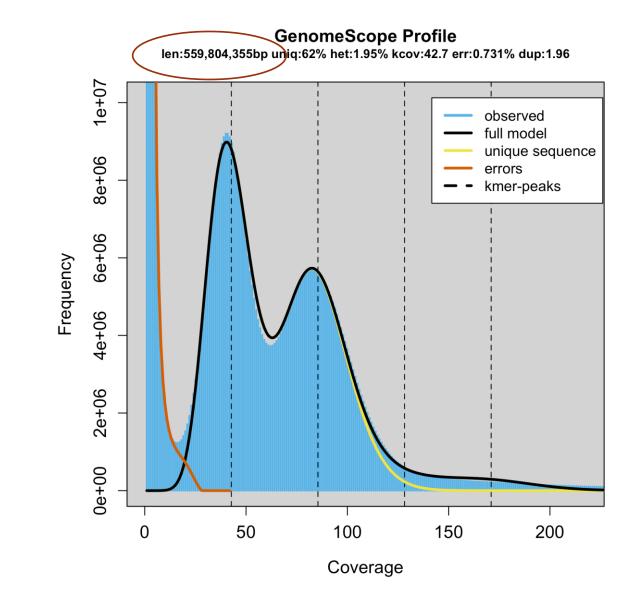
- **Size**: how large and/or complex is my genome?
- **Ploidy:** number of sets of chromosomes of the genome?
- Multinucleated: can cells have more than one nucleus?
- **Repetitive:** How much of the genome is repetitive? Repeat size distribution?
- Heterozygosity: Is my genome highly heterozygous? Inbred (homozygous)?
- **Public data:** Is a good quality genome of a related species available?

How large is my genome?

The size and complexity of the genome can be estimated from the ploidy of the organism and the DNA content per cell

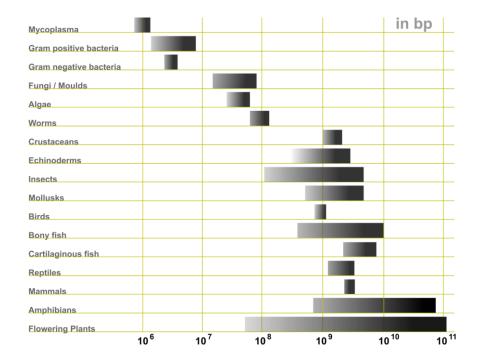
This will affect:

- How many reads will be required to attain sufficient coverage (typically 10x to 100x, depending on read length)
- What sequencing technology to use (short vs. long reads)
- What computational resources will be needed (generally amount of memory needed and length of time resources will be used)



Oyster: http://qb.cshl.edu/genomescope/genomescope2.0/

Genome size/complexity



By Abizar at English Wikipedia, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=19537795

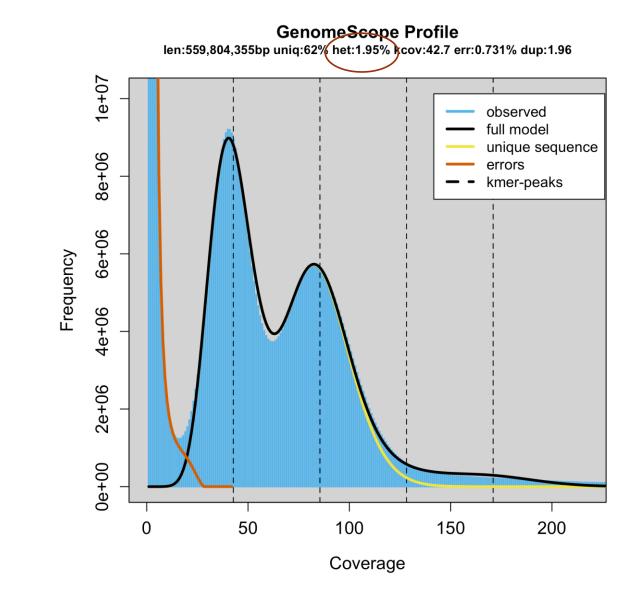
Heterozygosity

Heterozygous – Locus-specific; diploid organism has two different alleles at the same locus.

Heterozygosity is a metric used to denote the probability an individual will be heterozygous at a given allele.

Higher heterozygosity == more diverse == harder to assemble

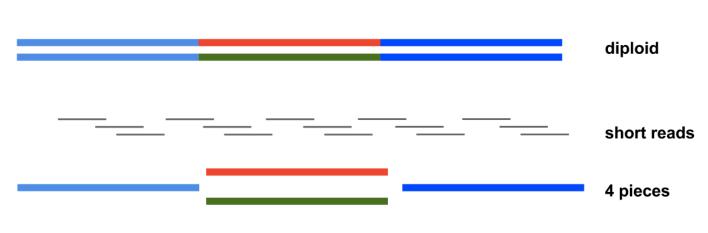
Unfortunately, assemblies are represented (for now) as haploid. So this is a major problem!



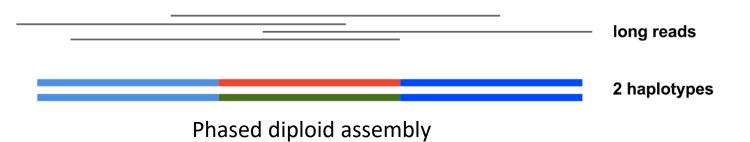
Oyster: http://qb.cshl.edu/genomescope/genomescope2.0/

Heterozygosity

Short reads - initial assembly has mix of homozygous and heterozygous regions



Unphased haploid assembly Haplotypes are separate contigs (haplotigs)

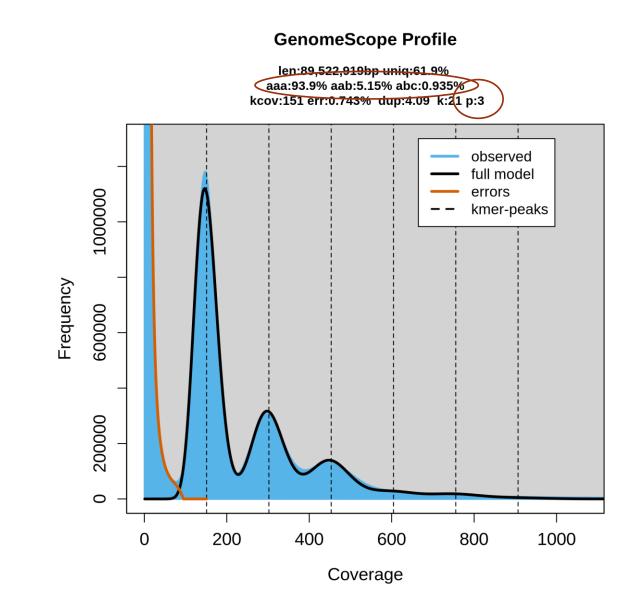




Ploidy

Number of sets of chromosomes in a cell (N)

- Bacteria 1N
- Vertebrates 2N (human, mouse, rat)
- Amphibians 2N to 12N
- Plants 2N to ??? (wheat is 6N)



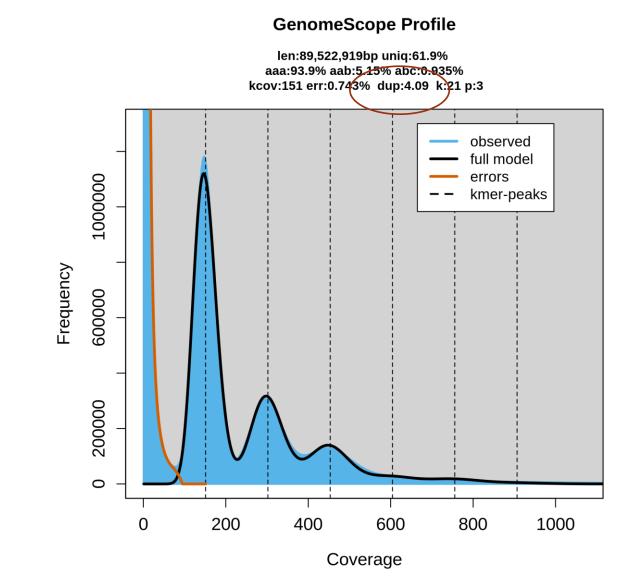
Root knot nematode: http://qb.cshl.edu/genomescope/genomescope2.0/

Repetitive sequences

Most common source of assembly errors

If sequencing technology produces reads > repeat size, impact is much smaller

Most common solution: generate reads or mate pairs with spacing > largest known repeat



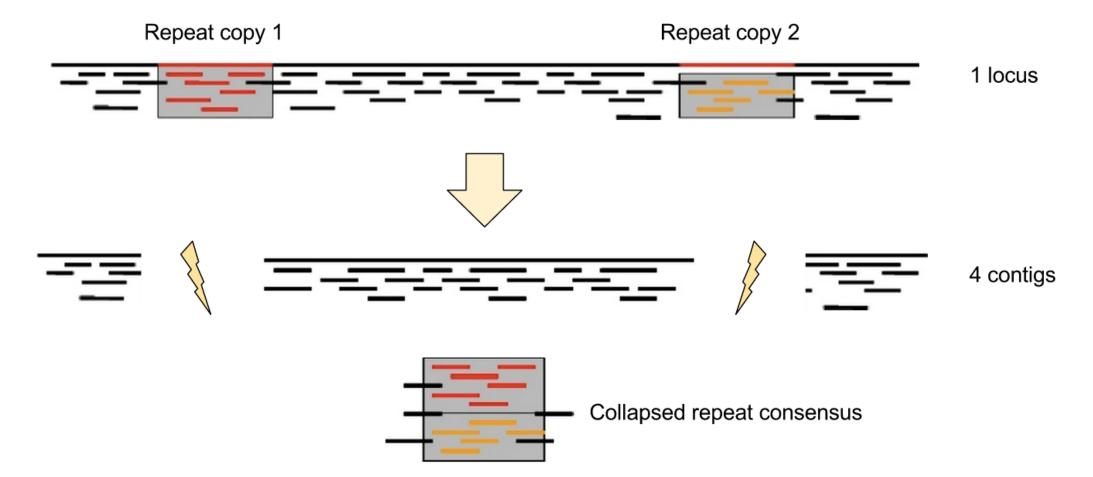
Root knot nematode: http://qb.cshl.edu/genomescope/genomescope2.0/

What is a repeat?

A segment of DNA which occurs more than once in the genome sequence



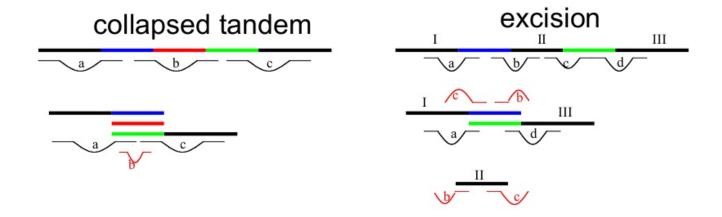
- Very common
 - Transposons (self replicating genes)
 - Satellites (repetitive adjacent patterns)
 - Gene duplications (paralogs)

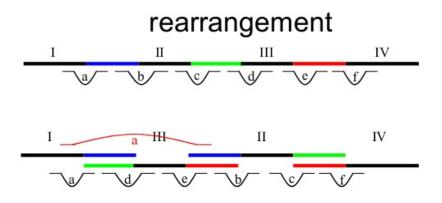


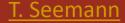
Assembling repeats

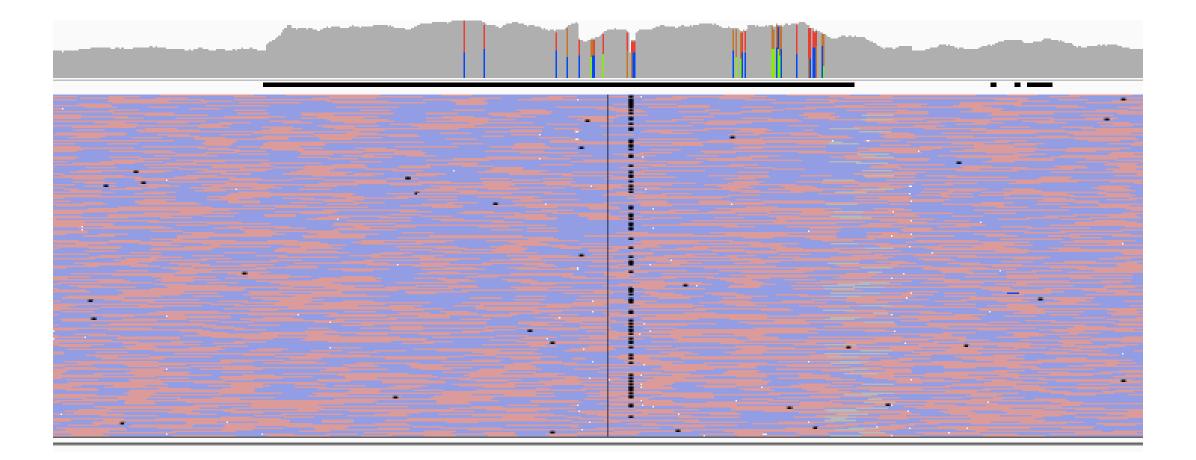
T. Seemann

Repeat mis-assembly

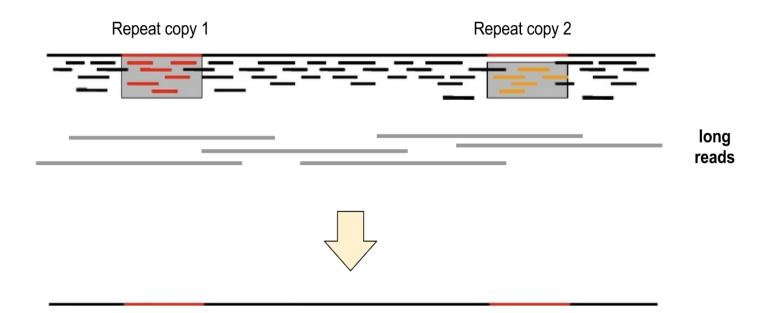




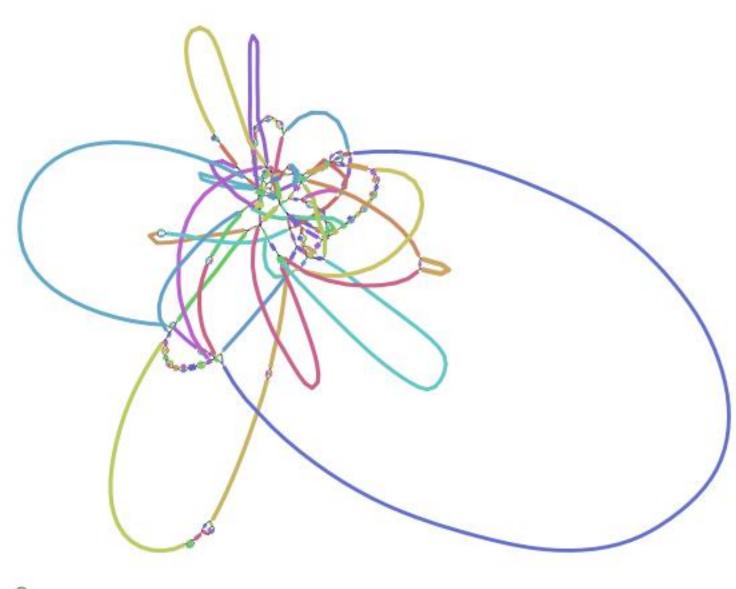


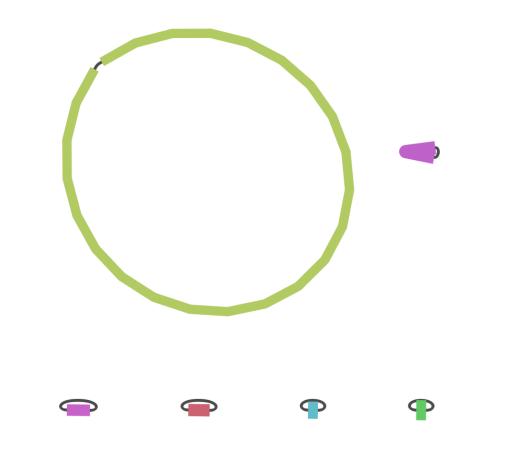


Long reads









$\sum_{i=1}^{n}$
$\bigvee \bigvee \bigvee$
<u> </u>

Genome(s) from related species

Preferably of good quality, with large reliable scaffolds

Help verifying the completeness of the assembly

Can themselves be improved in some cases

Help guiding the assembly of the target species

- But to be used with caution can cause errors when genome architecture is different!
- Large-scale genomic rearrangement in particular is a problem

Typical sequencing strategies

Small genomes (bacteria, fungal)

- If you can can get HMW DNA!
 - PacBio HiFi
 - **Oxford Nanopore** sequences at 40-50x coverage, 'polish' with hybrid correction (using Illumina data) and assembly using Unicycler, Canu, Flye
 - This may be changing with newer flow cells (R10.4.1 + 'kit14', as of May 2022)
- 2 x 300bp overlapping paired-end reads from Illumina MiSeq works okay but will get fragments

Larger genomes

- If you can afford it and can get HMW DNA
 - PacBio HiFi
 - HiC for scaffolding

T2T strategy

- Human assemblies
- HMW DNA preps
- 50x PacBio HiFi reads or higher
- 15-30x Oxford ultralong reads (>100kb)
- This is also in flux!
- \$\$\$\$\$\$\$\$



Science, _ March 2022



Assembly strategies and algorithms

For long reads (>500 nt), Overlap/Layout/Consensus (OLC) algorithms work best.

- Examples: hifiasm (PacBio HiFi only), Canu, Redbean, Flye, Shasta
- Hifiasm is generally recommended for PacBio HiFi data

For short reads, De Bruijn graph-based assemblers are most widely used

• Examples: MEGAHIT, SPAdes

Key points:

- There is no simple solution, best to try different assemblers and strategies
- Use simple metrics to gauge quality of assembly
- The field is rapidly evolving, like the sequencing technology

NEXT YEAR THIS PRESENTATION WILL CHANGE AGAIN!

Assessing your assembly

How good is my assembly?

How much total sequence is in the assembly relative to estimated genome size?

How many pieces, and what is their size distribution?

Are the contigs assembled correctly?

Are the scaffolds connected in the right order / orientation?

How were the repeats handled?

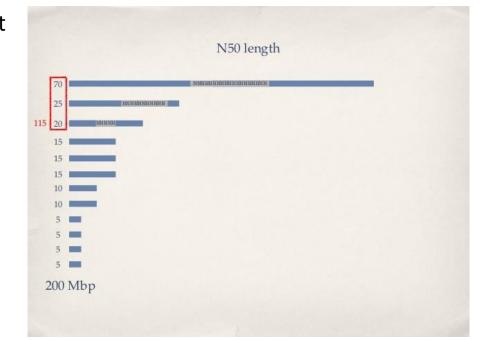
Are all the genes I expected in the assembly?

N50: the most common measure of assembly quality

N50 = length of the shortest contig in a set making up 50% of the total assembly length (*Larger is better*)

NG50 = length of the shortest contig in a set making up 50% of the estimated genome size

NG50 is generally better



What, Me Worry?

N50 concerns

Optimizing for N50

- Encourages mis-assemblies!
- Encourages 'gaming' the stats

An aggressive assembler may over-join:

- 1,1,3,5,8,12,20 (previous)
- 1,1,3,5,**20**,20 (now)
- 1+1+3+5+20+20 = 50 (unchanged)

N50 is the "halfway sum" (still 25)

1+1+3+5+20= 30 (≥ 25) so N50 is 20 (was 12)

You can also filter contigs below a certain (arbitrary) size, which lowers overall assembly size (and increases N50)



Comparative analysis

Compare against

- A close reference genome
- Results from another assembler
- Self-comparison
- Versions of the same assembly

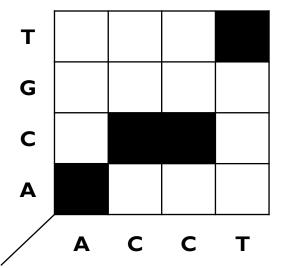
Whole genome alignment

- MUMmer
- Lastz

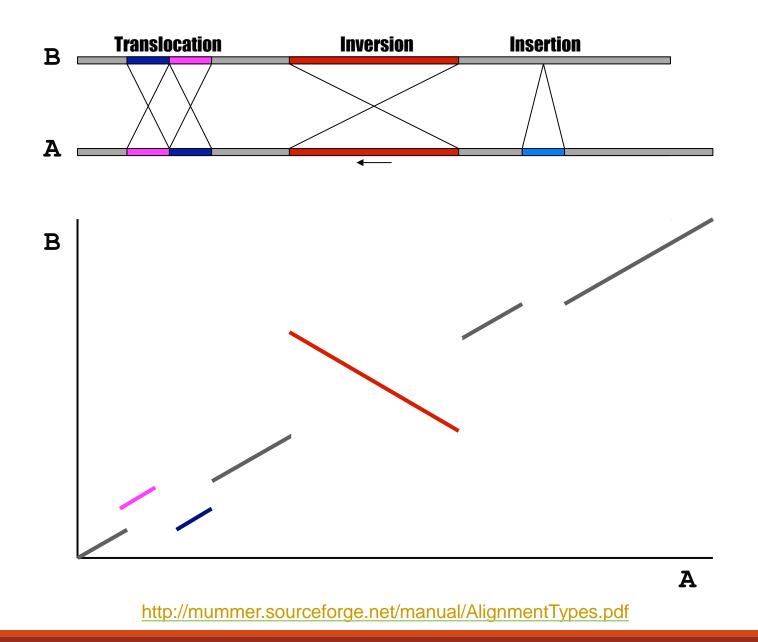
Generates an alignment and a *dot plot*

Dot Plot

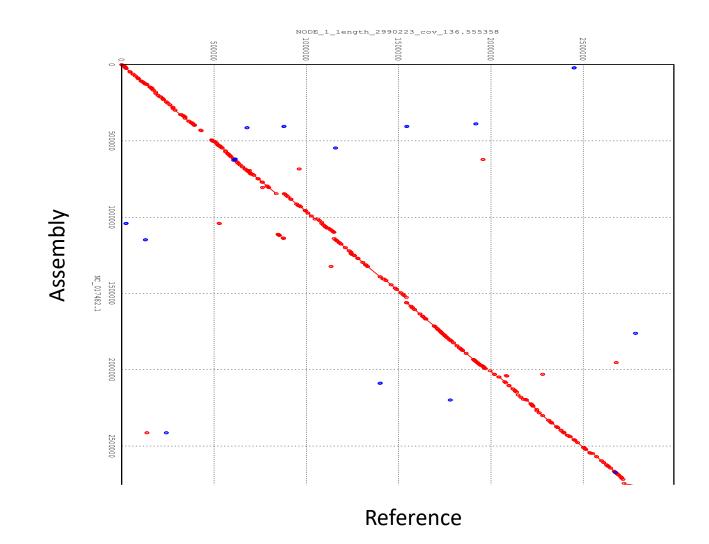
- How can we visualize *whole* genome alignments?
- With an alignment dot plot
 - $-N \times M$ matrix
 - Let *i* = position in genome A
 - Let *j* = position in genome *B*
 - Fill cell (i,j) if A_i shows similarity to B_j

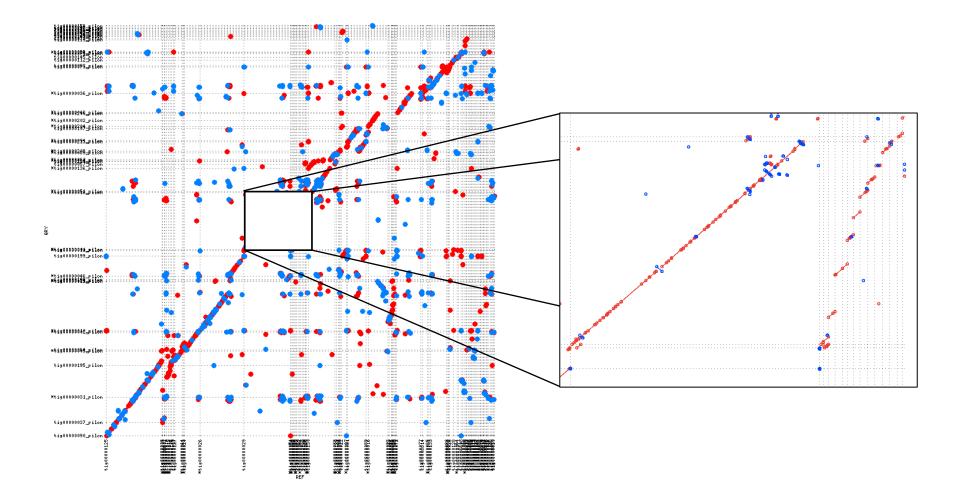


 A perfect alignment between A and B would completely fill the positive diagonal

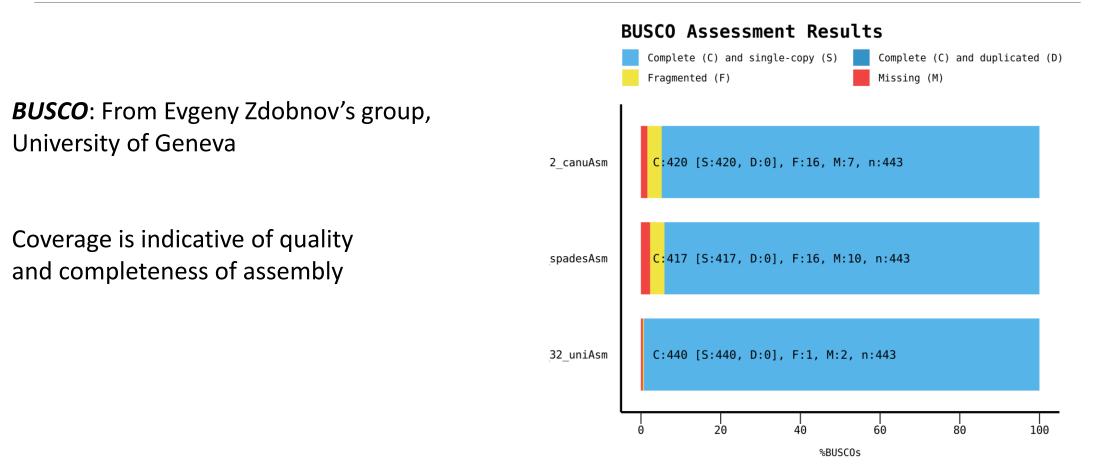


From M. Schatz and A. Phillipy : Alignment and Assembly Lecture





BUSCO: conserved gene sets



QUAST

QUality ASsessment Tool

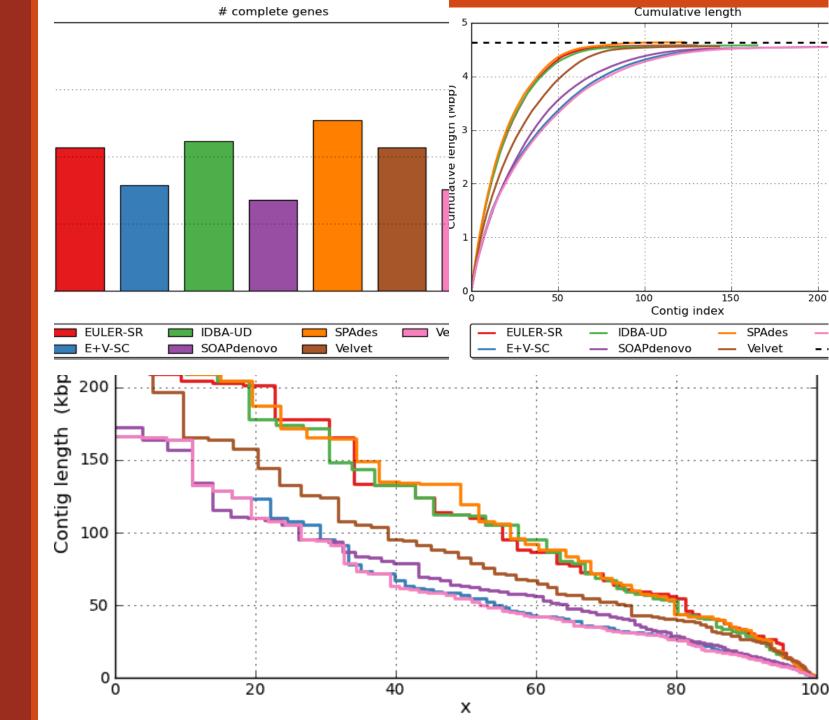
- Small (bacterial, fungal) and large (eukaryotic) genomes
- Metagenomes
- Icarus for contig alignment visualization

Can compare multiple assemblies against one another

Compare against a known (or close) reference

Optional: Predict genes or include annotations (checks for odd issues like frameshifts)

Generates a summary HTML report



Even the best genomes are not perfect

nature	
Explore content 🗸 Journal information 🖌 Publish with us 🖌 Subscribe	
nature > news > article	
NEWS 04 June 2021	
A complete human genome	
sequence is close: how scientists	
filled in the gaps	
Researchers added 200 million DNA base pairs and 115 protein-coding genes – but th	ey've

yet to entirely sequence the Y chromosome.

Genome graphs

Genome graphs

With the release of the latest human genome reference, there is more pressure to represent more data with a genome.

Current representations are mainly **haploid** (one copy)

Newer representations are *genome graphs*, where variant information is retained (e.g. heterozygosity)

Tools are still catching up, but many new assemblers (e.g. hifiasm) generate a *diploid* assembly now

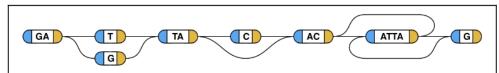
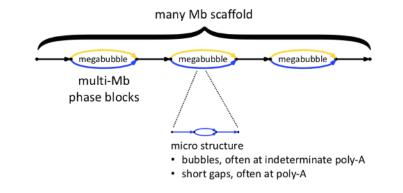


Figure 1: Example sequence graphs. Each node holds a string of bases. An edge can connect, at each of its ends, to a base on either the left (5', blue) or the right (3', yellow) side of the node. When reading through a thread to form a DNA sequence, a valid walk must leave each node via the opposite side from that through which it was entered; a node's sequence is read as reverse-complemented if the node is entered on the 3' side. One thread that this graph spells out (reading from the left side of the leftmost sequence to the right side of the rightmost sequence, along the nodes drawn in the middle) is the sequence "GATTACACATTAG". Straying from this path, there are three variants available: a substitution of "G" for "T", a deletion of a "C", and an inversion of "ATTA". If all of these detours are taken, the sequence produced is "GAGTAACTAATG". All 8 possible threads from the leading G to the trailing G are allowed.



Novak et al, bioRxiv: https://doi.org/10.1101/101378

10x Genomics

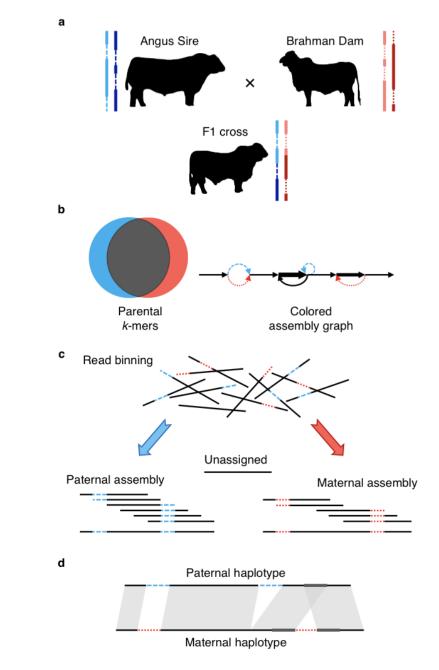
Genome graphs

One interesting application of graphs: *trio assembly*

Short- and long-read sequencing

Advantage:

Better assembly Phased variants Structural variants!



Koren et al, Nature Biotech. 2018, Oct 22

Genome Annotation

Methods for genome annotation

Ab initio

- i.e. based on sequence alone
- INFERNAL/rFAM (RNA genes), miRBase (miRNAs), RepeatMasker (repeat families), many gene prediction algorithms (e.g. AUGUSTUS, Glimmer, GeneMark, ...)

Evidence-based

- Transcriptome data for the target organism (the more the better)
- Proteins of interest
- Align trx/protein sequences to assembly, generate gene models

Combined approaches

Most common

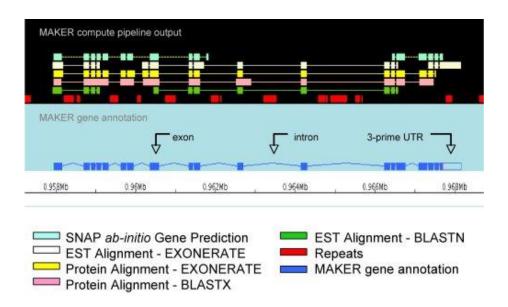
General steps for biological annotation

- 1. Predict gene models using *ab initio*-based tools
 - May require considerable tuning and a bootstrapping step
- 2. Using closely-related protein/transcripts, BLAST against assembly to find locations
- **3**. Find potential splice junctions of BLAST hits
- 4. Combine all evidence and make consensus gene (and possibly transcript/isoform) predictions, with annotation metrics for confidence of matches
 - Can include UTR regions if RNA-Seq is included
- 5. Assess completeness of annotation (run BUSCO, but on proteins/transcripts)
- 6. Run InterProScan of predicted proteins against databases of protein domains (Pfam, Prosite, HAMAP, PANTHER, ...)

MAKER, integration framework for genome annotation

MAKER runs many software tools on the assembled genome and collates the outputs

See http://gmod.org/wiki/MAKER



Example Pipelines

Bacterial

- Prokka (bacterial/archaeal/viral)
- NCBI Prokaryotic Genome Annotation Pipeline (PGAP)
- Joint Genome Institute IMG/ER (Integrated Microbial Genome Expert Review)
 - Online only

Eukaryotic

- NCBI RefSeq pipeline
 - Have to submit to NCBI (and make public) to use
 - Requires RNA-Seq
- MAKER
- Braker2

Acknowledgements

Materials from this slide deck include figures and slides from many publications, Web pages and presentations by:

- Carver Biotechnology Center (HPCBio, DNA Sequencing Core)
- M. Schatz, A. Phillipy, T. Seemann, S. Salzberg, K. Bradnam, D. Zerbino, M. Pop, G. Sutton, Nick Loman, Carson Holt, Ryan Wick.
- I highly recommend Ben Langmead's teaching materials; he has a ton fabulous (and much more in-depth) notes on his lab page: <u>http://www.langmead-</u> <u>lab.org/teaching-materials/</u>
- Thank you!