GENOME ASSEMBLY: WHERE DO I Start

Marcela Uliano-Silva, PhD



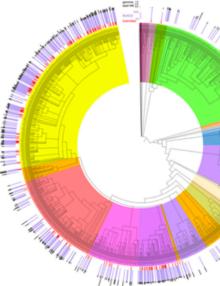


- Senior Bioinformatician Wellcome Sanger Institute Darwin Tree of Life Project. Tree of Life Assembly Team (ToLA)
- Churchill College Postdoctoral By-Fellow, University of Cambridge
- ► Horizon2020 Marie Curie PostDoc Fellow (2017-2019), IZW, BenGenDiv, Germany
- ► PhD in Biophysics (2017) IBCCF UFRJ, Brazil
- MSc in Biophysics (2013) IBCCF UFRJ, Brazil
- BSc in Biology (2010) UFSC, Brazil
- TED Fellow













Darwin TREE

Tree of Life: Major Projects

Collaborating widely to deliver across diversity











Darwin Tree of Life Project \star

70,000 species from Britain and Ireland [Phase 1: 2,000 species]

Aquatic Symbiosis Genomics

1,000 species (500 symbiotic systems) from marine and freshwater

★ Vertebrate Genomes Project

Realising VGP Phase 1 (ordinal - 260 species) and Phase 2 (family) goals

European Reference Genome Atlas

Sequencing the genomes of all species in 0 the European continent - Pilot 25 species

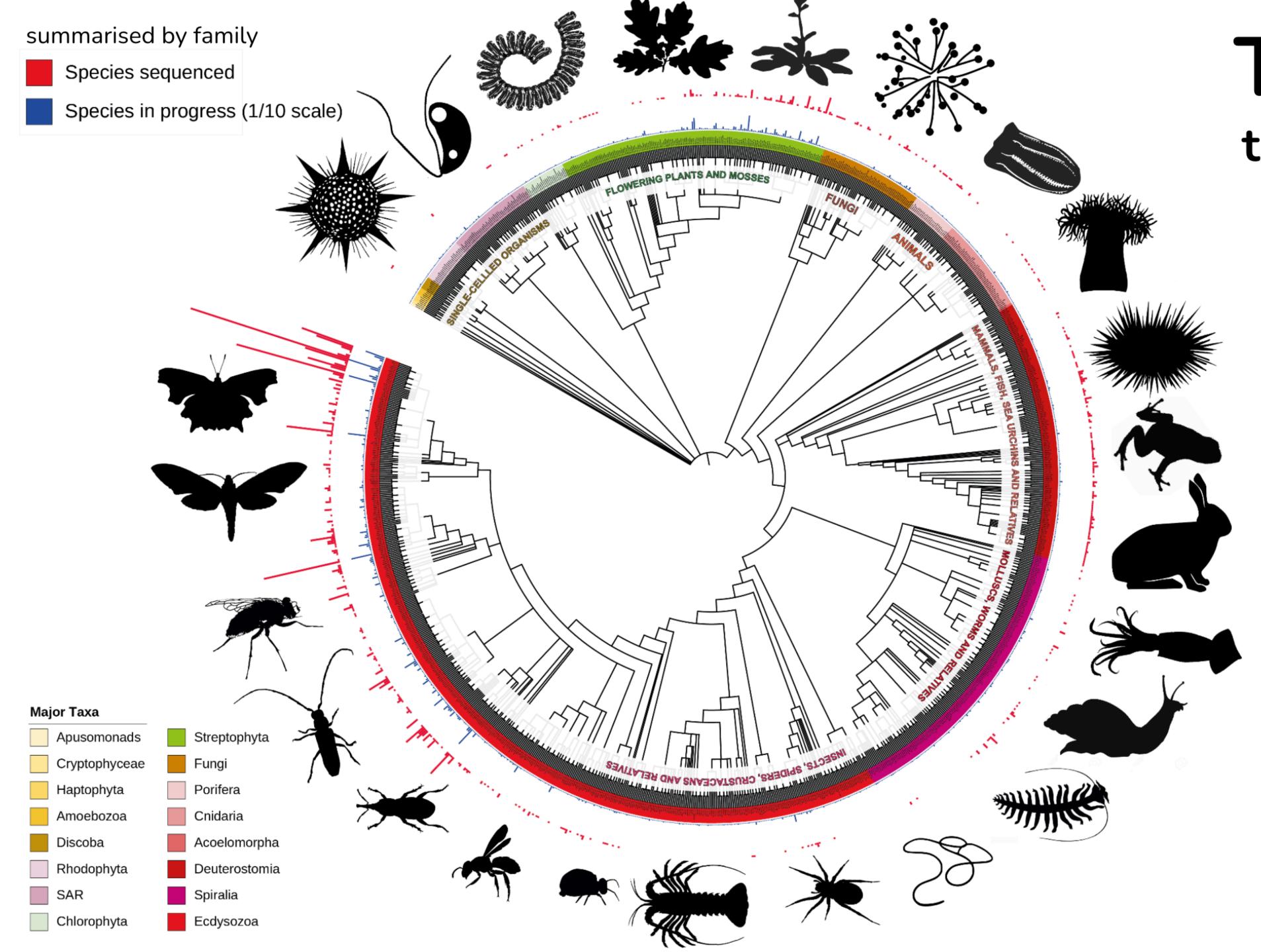
Earth BioGenome Project

Working to deliver Phase 1 (family) goals, and to "sequence all life for the future of life"

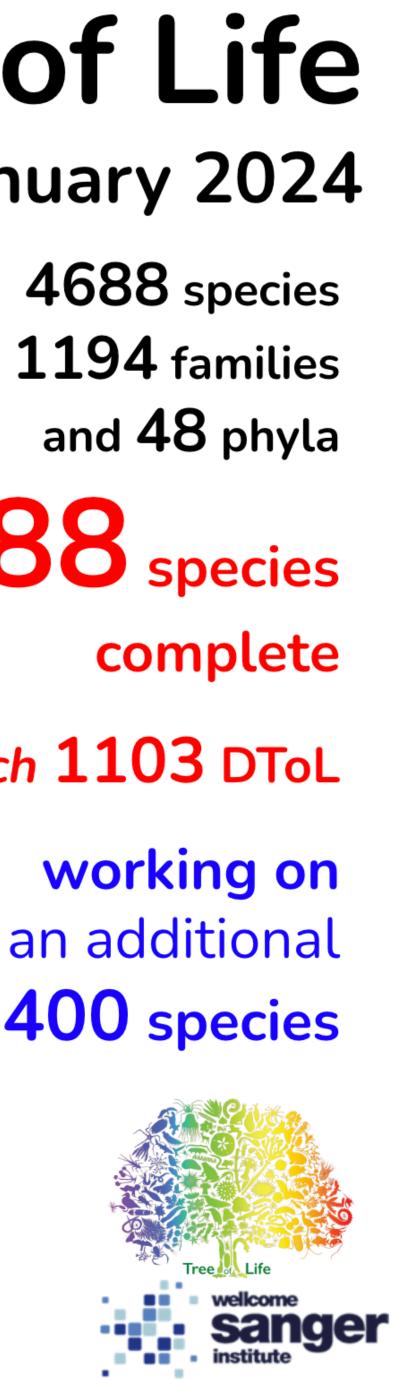






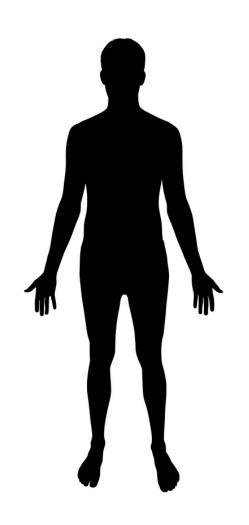


Tree of Life to 01 January 2024 from **1194** families **1288** species of which **1103** DToL 3400 species



Genome assembly: what is my goal?

• Understand variation in populations (disease-related SNPs etc...)



Genome re-sequencing Assembly by mapping to a reference • Study the molecular profile of a species never before sequenced (evolutionary studies etc..)



De novo assembly

ASSEMBLY GRAPHS

Overlap Layout Consensus

"Both representations share the idea that a **genome assembly corresponds to** <u>a path in the graph</u>: for this reason, the step following the construction of such a graph is the extraction of relevant paths. <u>Under ideal conditions, such</u> as the absence of errors and repeats, we can reconstruct only one relevant path in such graph (that is, there is only one possible assembly). " (Rizzi et al, 2019)

De Brujin

ASSEMBLY GRAPHS

Overlap Layout Consensus

- Reads of Length L and overlap cutoff
- LongReads

GTAGTATAGTCAGTATCA Sequence reads GTAGTA TAGTAT AGTATA GTATAG TATAGT ATAGTC TAGTCA AGTCAG GTCAGT TCAGTA CAGTAT AGTATC GTATCA Consensus overlap assembly GTAGTA TAGTAT AGTATA GTATAG TATAGT ATAGTC TAGTCA AGTCAG GTCAGT

TCAGTA

GTAGTATAGTCAGTATCA

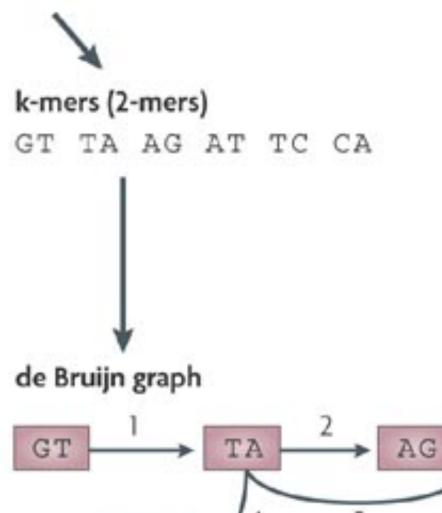
CAGTAT

AGTATC

GTATCA

"Both representations share the idea that a **genome assembly corresponds to** <u>a path in the graph</u>: for this reason, the step following the construction of such a graph is the extraction of relevant paths. <u>Under ideal conditions, such</u> as the absence of errors and repeats, we can reconstruct only one relevant path in such graph (that is, there is only one possible assembly). " (Rizzi et al, 2019)

Original sequence

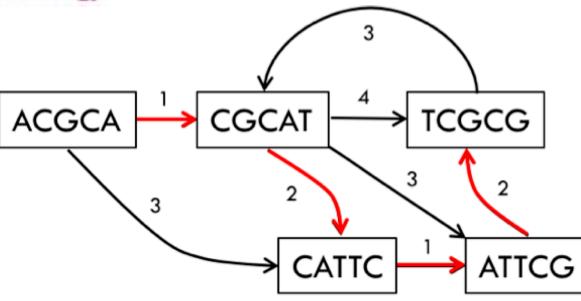


AT

De Brujin

- Kmers _
- Short reads

Nature Reviews | Microbiology



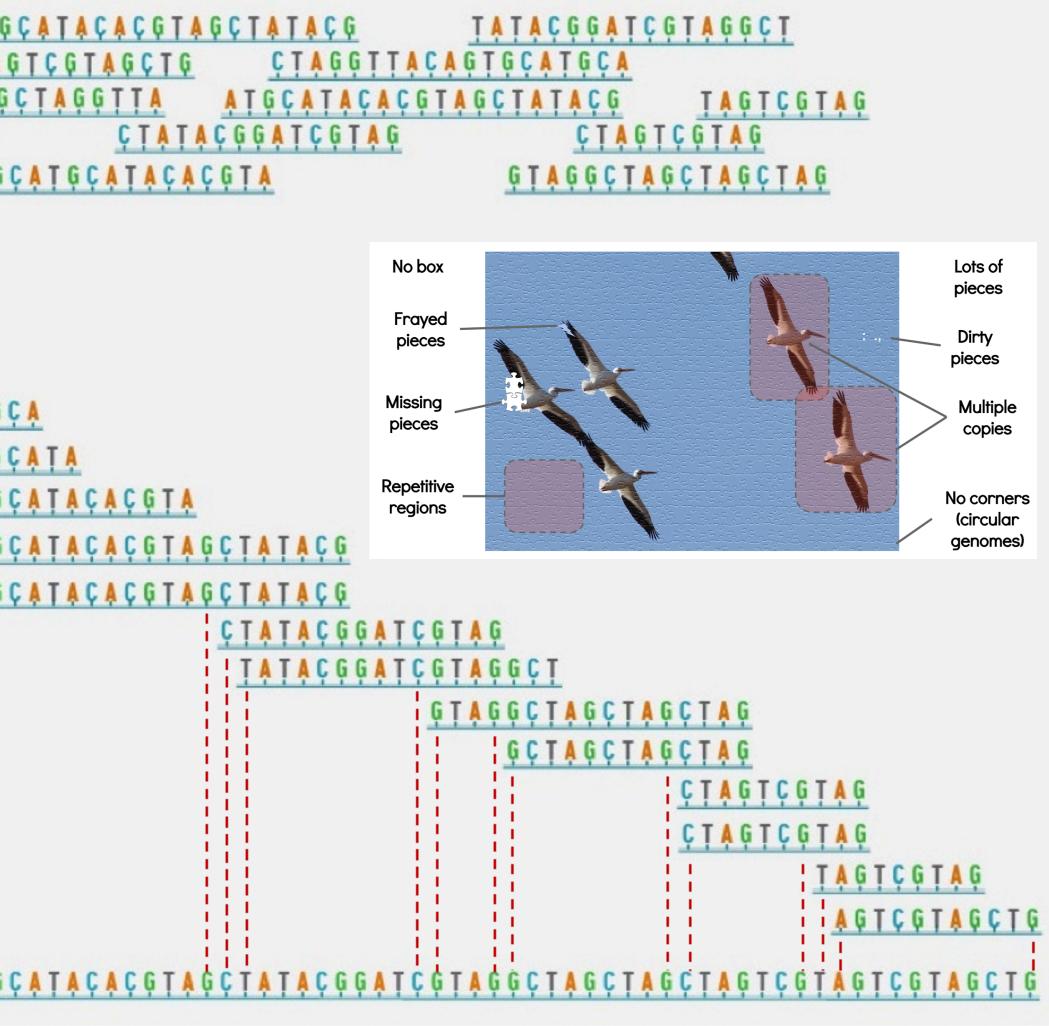


I WANT TO TALK TO YOU ABOUT Long Read Sequencing

The Naïve Genome Assembly Approach

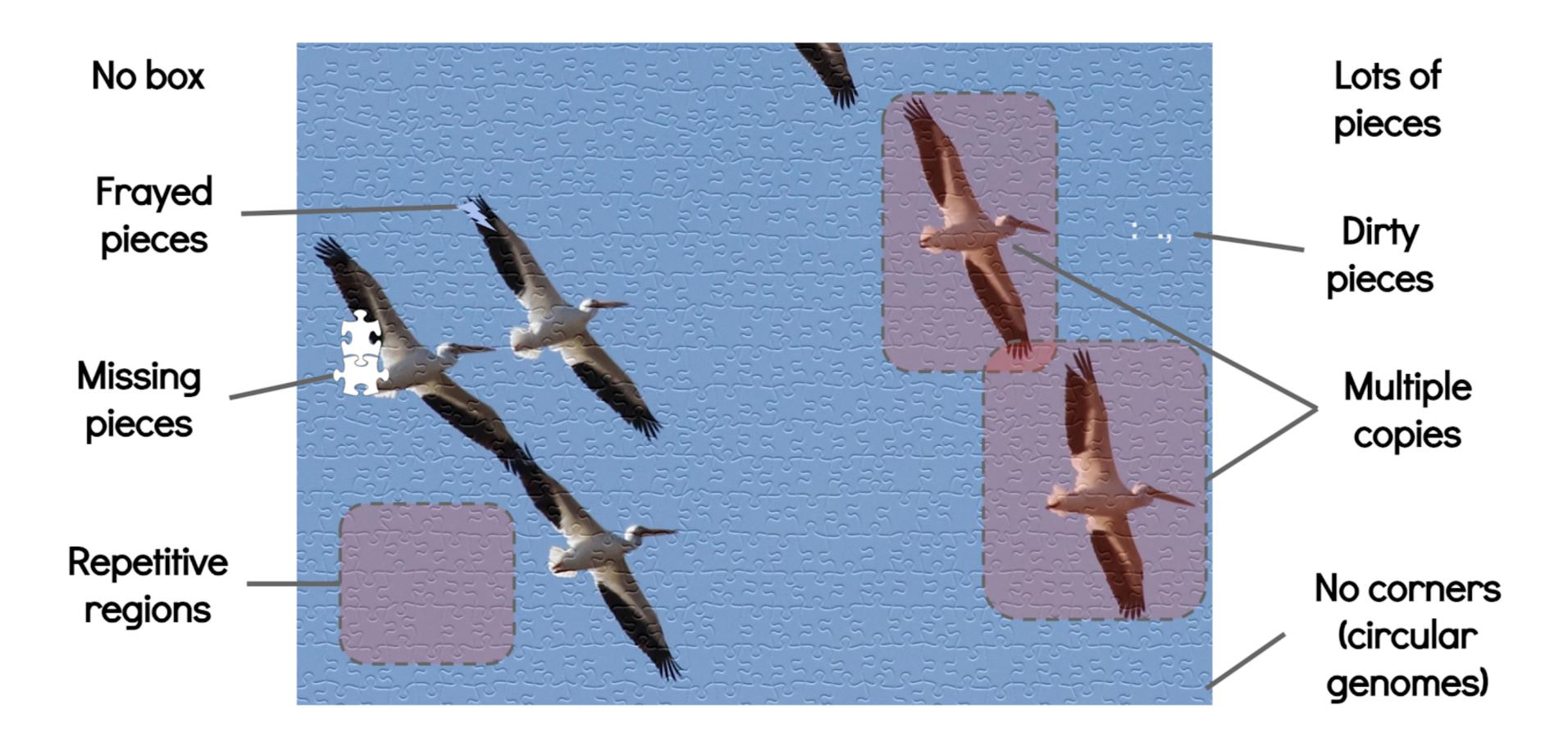
DNA sequence reads	AÇÇAĞĞÇTAĞĞTTA ATĞÇ GÇTAĞÇTAĞÇTAĞ AĞT ÇTAĞŢÇĞTAĞ GÇTAÇÇAĞĞÇ GĞTTAÇAĞŢĞÇATĞÇATĂ GÇT GÇTAÇÇAĞĞÇTAĞĞŢTA GÇT
	<u>GÇTAÇÇAGGÇTAGGTTA</u> <u>GÇTAÇÇAGGÇTAGGTTA</u> <u>Aççaggçtaggtta</u> <u>CTAGGTTAÇAGTGÇATGÇA</u>
Assembly of DNA sequence reads	ĢĢŢŢĂÇĂĢŢĢÇĂŢĢÇĂ ĢÇĂŢĢÇĂ ĂŢĢÇĂ
	<u>Atgç</u>
Assembled DNA sequence	ĢÇTĂÇÇĂĢĢÇTĂĢĢTTĂÇĂĢTĢÇĂŢĢÇI







What makes a jigsaw puzzle hard?



 What helps? Larger pieces (read leng reads). fewer repeats and copies...

• What helps? Larger pieces (read length); fewer dirty or frayed pieces (errors in

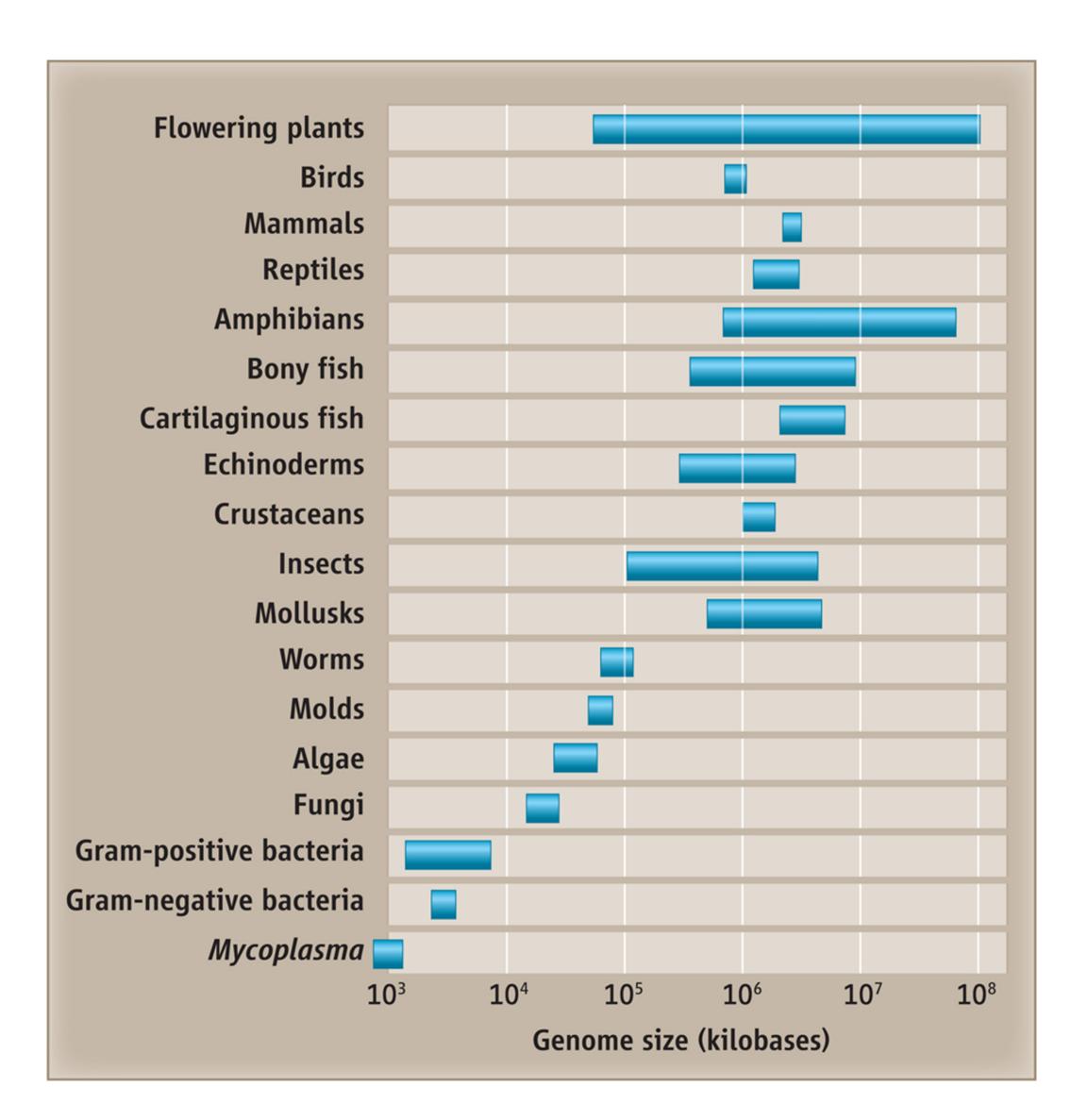
WHAT IS THE PROBLEM WITH SHORT READS?

What are eukaryotic genomes made of?

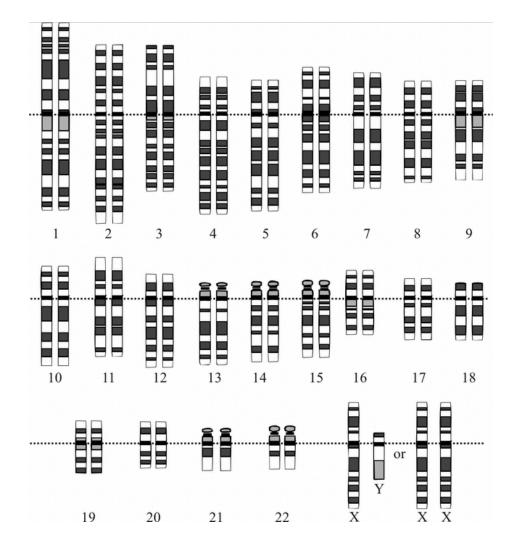
Repeats:

- Centromeres (Tandem arrays of repeated sequence studded with transposable elements (plants, humans)
- Telomeres (tandem arrays of simple repeats)
- Mobile elements
- Segmental duplications
- rRNAs

But why? (Fedoroff, 2012)

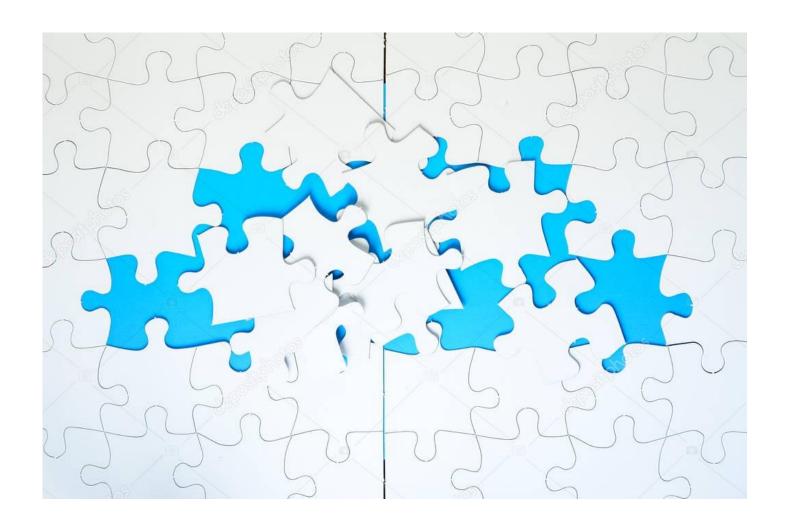


WHEN WE ASSEMBLE A GENOME . . .



What we would like to have

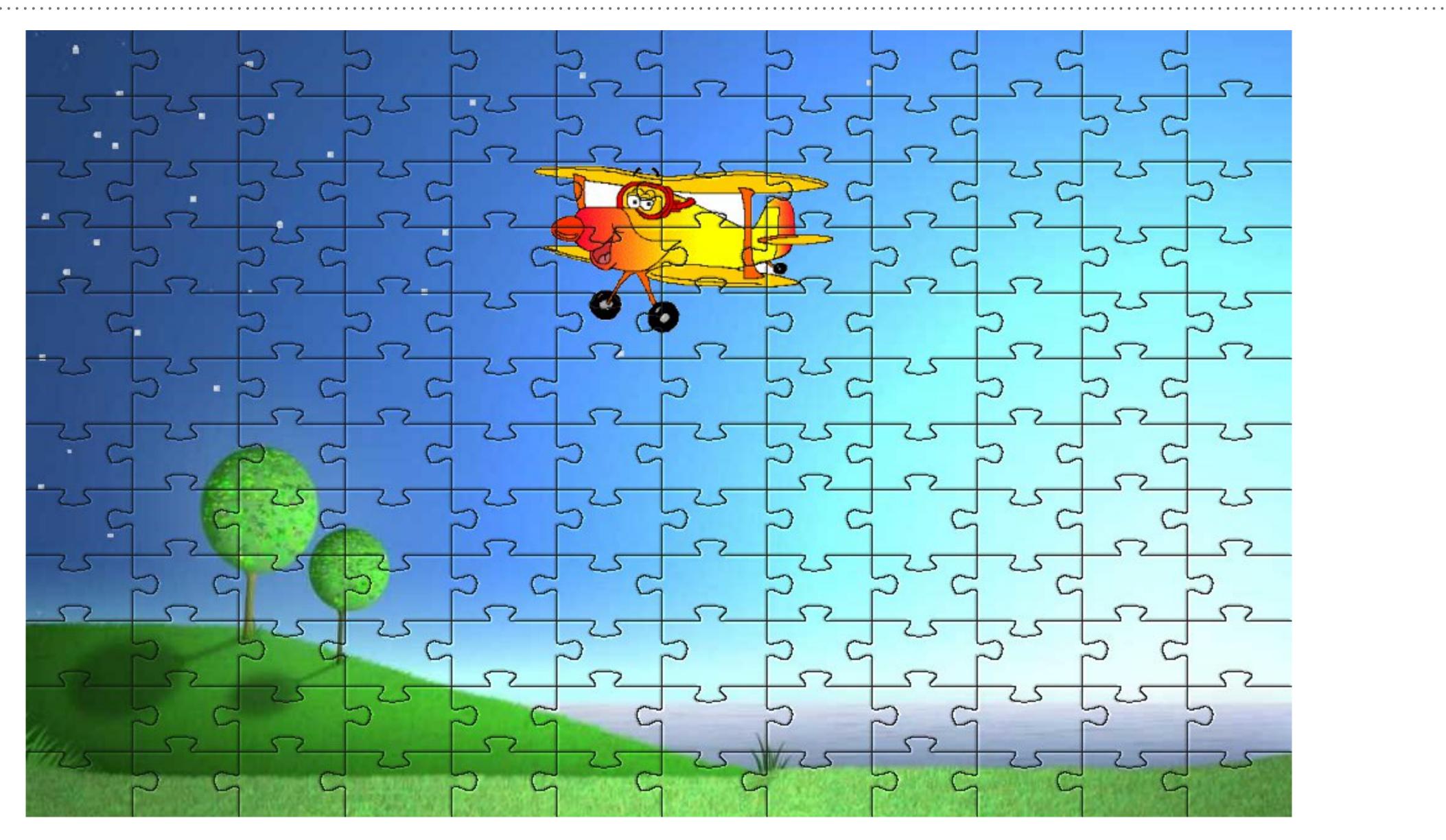
- One DNA sequence for each chromosome



What we really have - Contigs, scaffolds, gaps, N50s



I'M A EUKARYOTIC GENOME – THE BLUE AND GREEN ARE MY REPEATS

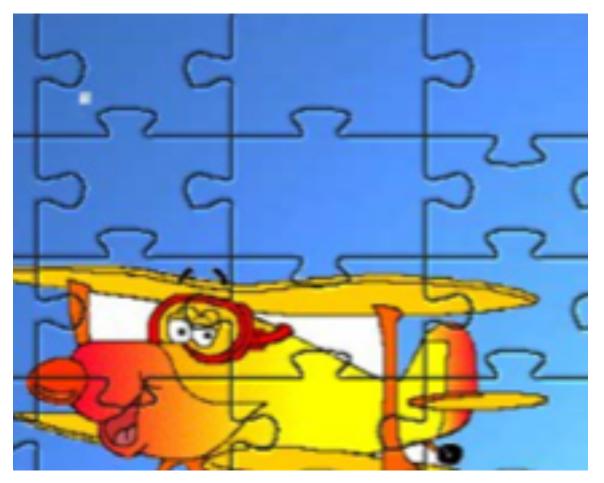


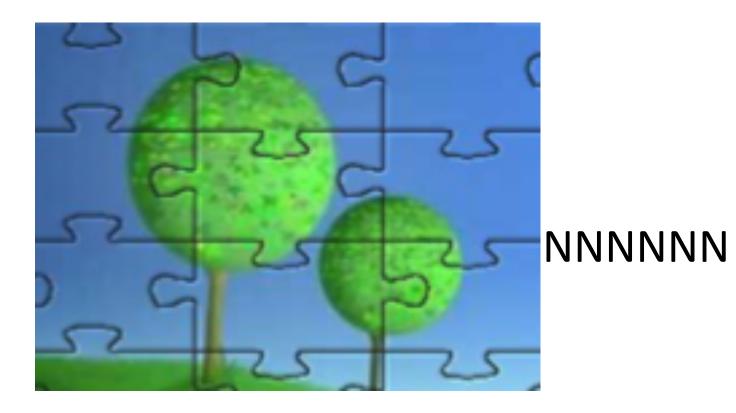
THIS IS A SHORT-READS GENOME ASSEMBLY OF ME



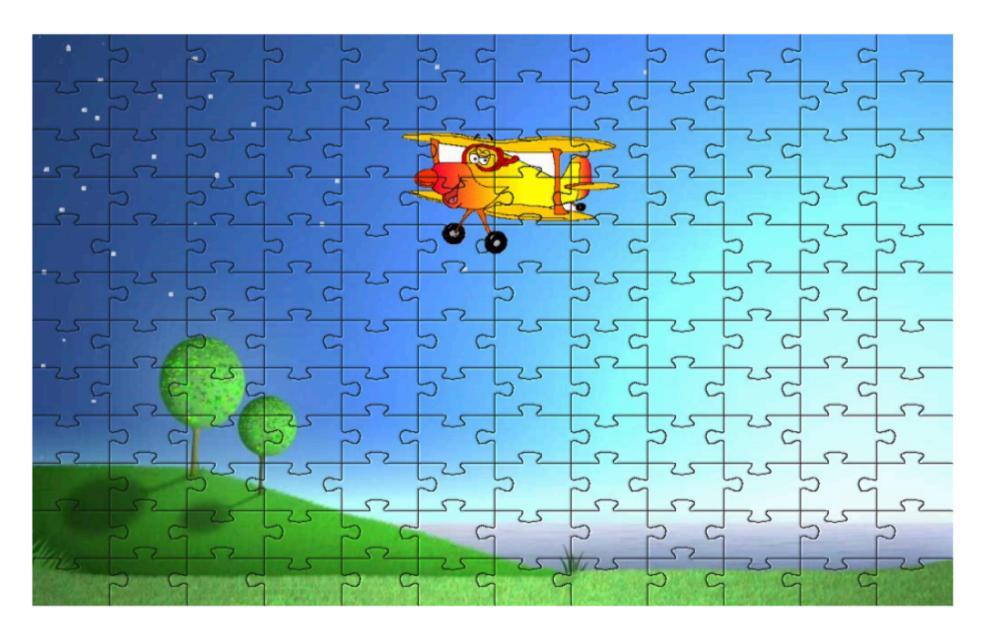
. . . .

NNNNN



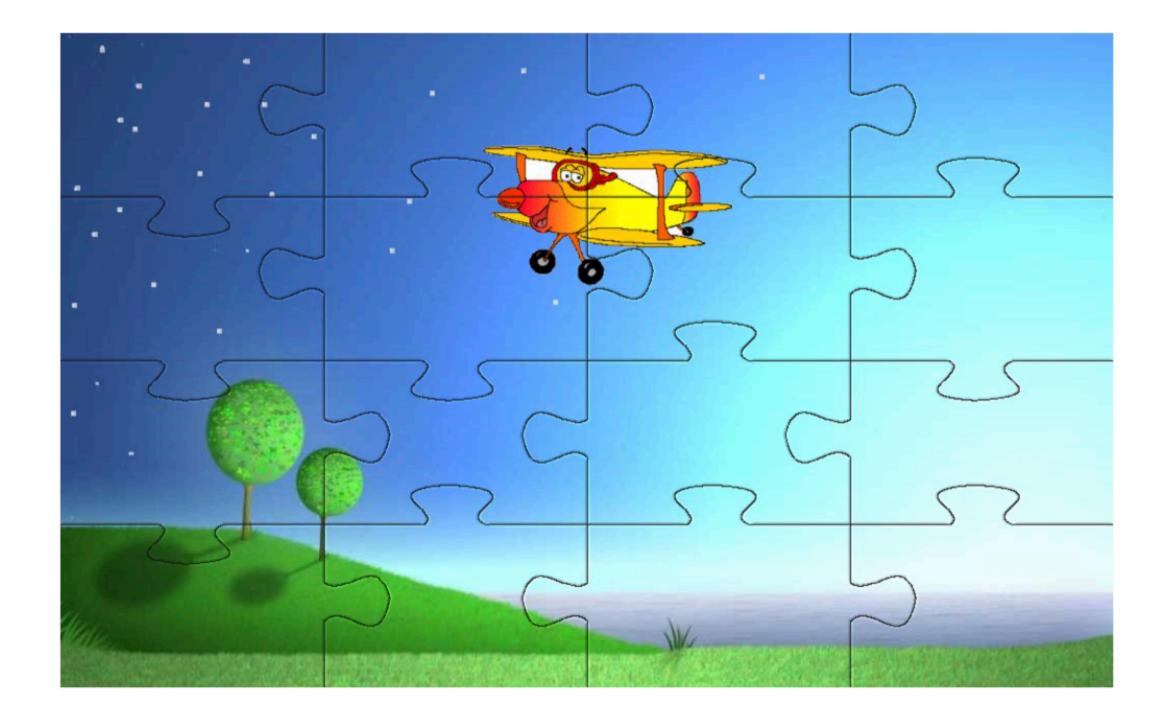






Assembling with long reads

Assembling with short reads



YOUR GENOME ASSEMBLY **PROJECT STARTS IN THE LAB**

High Molecular Weight DNA extraction is key

No one-size-fits-all protocol!

FEATURES



protocols.io

OCT 02, 2023

Sanger Tree of Life Wet Laboratory Protocol Collection

DOI

PLANS

BLOG

dx.doi.org/10.17504/protocols.io.8epv5xxy6g1b/v1 Amy Denton¹, Halyna Yatsenko¹, Jessie Jay¹, kh¹, Caroline Howard¹

¹Tree of Life, Wellcome Sanger Institute, Hinxton, Cambridgeshire, **CB10 1SA**



CASE STUDY

Tree of Life at the Wellcome Sanger Institute

Tree of Life Genome Note Editor





Scan me!







OXFORD

Benchmarking ultra-high molecular weight DNA preservation methods for long-read and long-range sequencing

Hollis A. Dahn^{1,†}, Jacquelyn Mountcastle^{2,†}, Jennifer Balacco², Sylke Winkler³, Iliana Bista^{4,5}, Anthony D. Schmitt⁶, Olga Vinnere Pettersson¹⁰⁷, Giulio Formenti¹⁰², Karen Oliver⁴, Michelle Smith¹⁰⁴, Wenhua Tan¹⁰³, Anne Kraus³, Stephen Mac⁶, Lisa M. Komoroske [©]⁸, Tanya Lama [©]⁸, Andrew J. Crawford [©]⁹, Robert W. Murphy [©]¹, Samara Brown [©]², Alan F. Scott [©]¹⁰, Phillip A. Morin^[1], Erich D. Jarvis^{[2,12} and Olivier Fedrigo^[2,*]

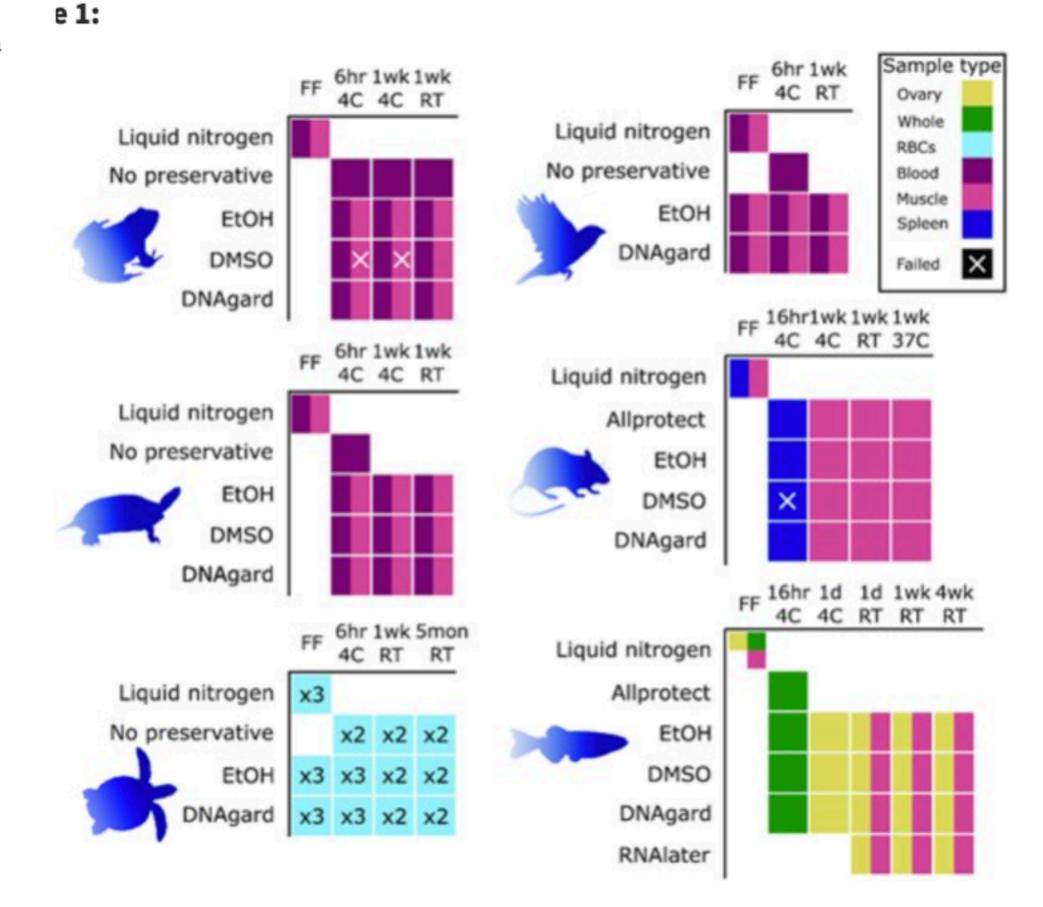
No one-sizefits-all protocol!

slack Channel: all.things.up.to.assembly

GigaScience, 2022, **11**, 1–13 DOI: 10.1093/gigascience/giac068 RESEARCH



PROJECT OF THE G10K CON SORTIUM



I EXTRACTED HMW DNA: WHAT DO I DO NOW?

- Our recipe working across the Tree of Life:
- Chromosome level genomes
- ► 25x Pachio HiFi
- ► 100x Hi-C (Arima/Qiagen)

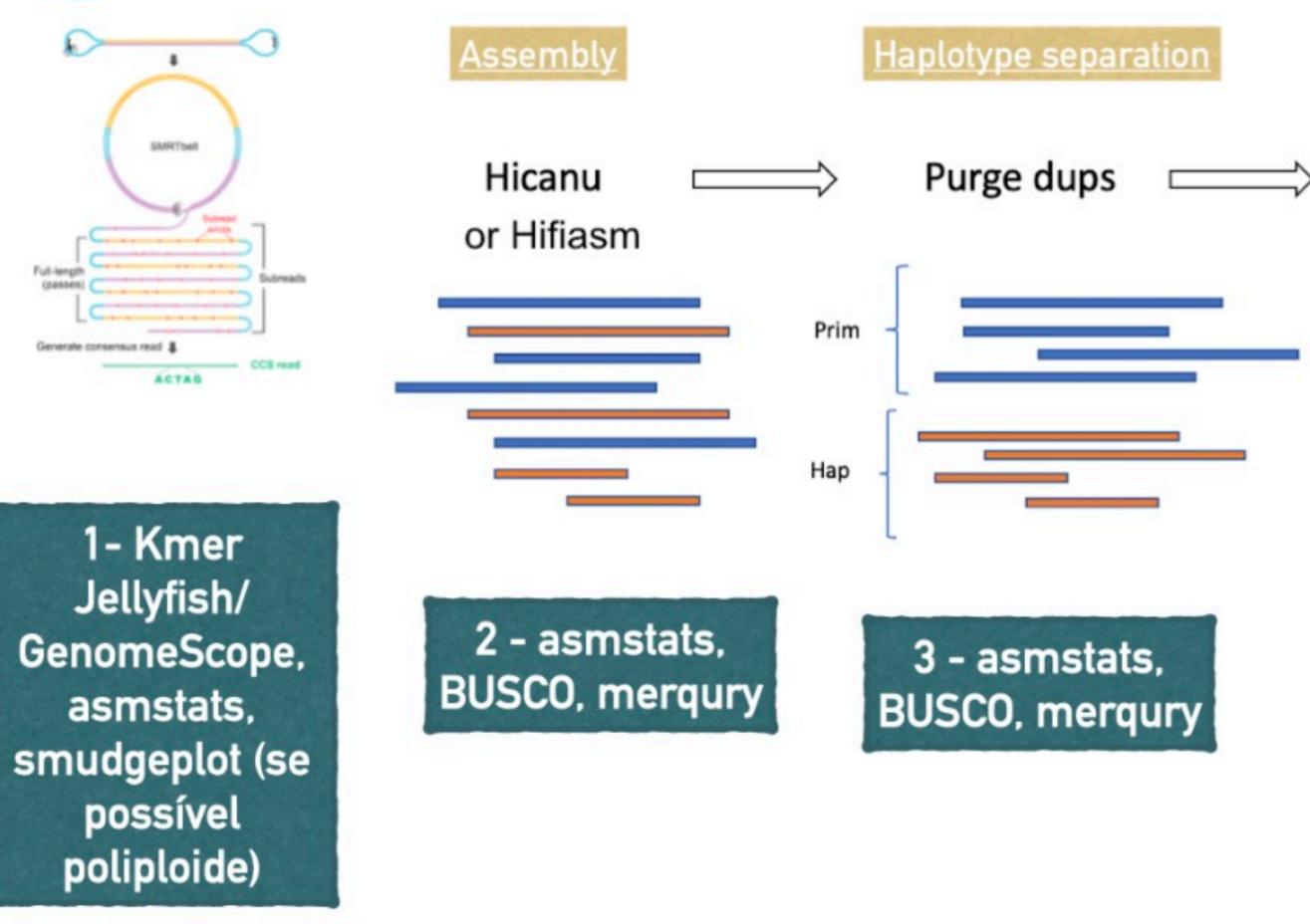
- T2T (Telomere to Telomere) genomes
- ➤ The above plus 25x ONT Ultra Long (>100Kb reads)





DToL Current Pipeline

Sequencing technologies: PacBio HiFi + HiC (Arima or Qiagen) ٠





For mitochondria genome assembly



Yahsscaffolding (Arima or Qiagen HiC)



Curated assembly

4 - asmstats, BUSCO, mergury, HiC heatmap

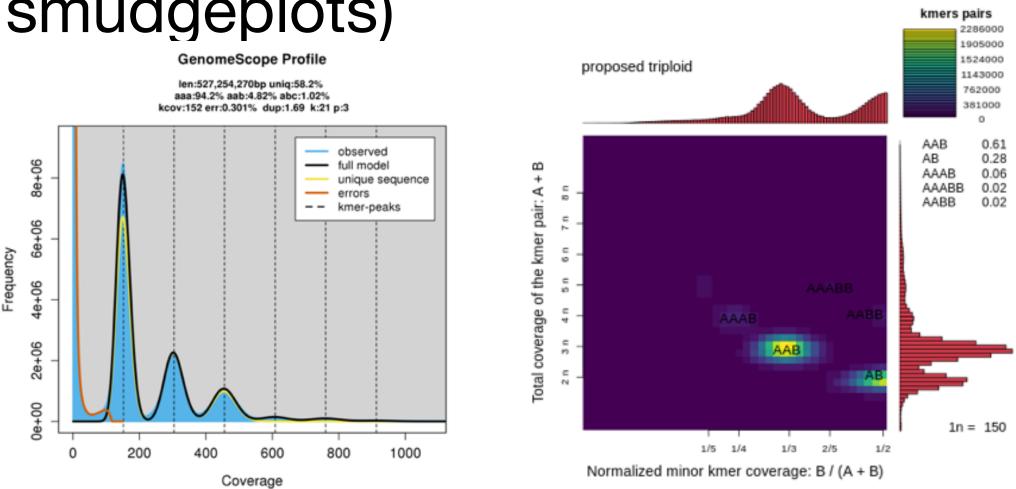
5 - asmstats, BUSCO, mergury, HiC heatmap



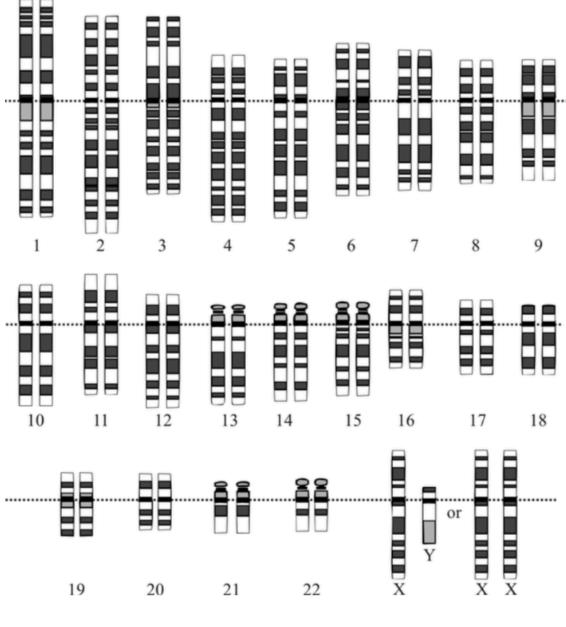


Key considerations to start your genome assembly project

- Genome size (flow cytometry, Kmer analysis, GoaT) https://goat.genomehubs.org/
- Heterozygosity (kmer analyses: jellyfish, genomescope)
- Repetitive content (kmer analyses: jellyfish, genomescope)
- Ploidy (kmer analyses: smudgeplots)







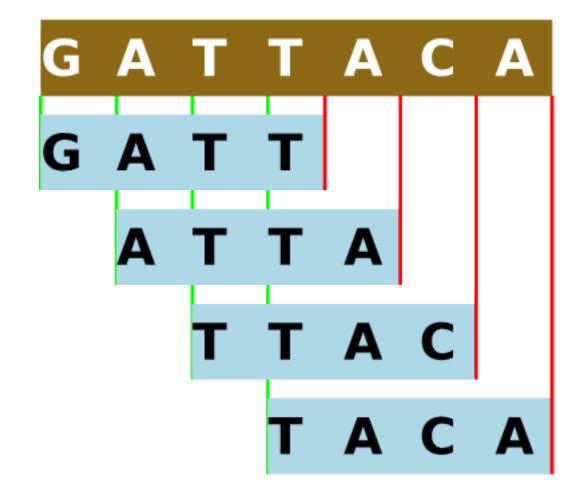
I HAVE HIGH-QUALITY DATA (ILLUMINA, PACBIO HIFI, DUPLEX NANOPORE)

I WILL do a kmer analysis first thing





KMER ANALYSIS



WHAT ARE K-MERS ?

► In biology, k-mers are unique subsequences of a sequence of length k

So, by way of example, the sequence ATCGATCAC contains the following 3-mers (k-mer of size 3):

Sequence: ATCGATCAC 3-mer #0: ATC 3-mer #1: TCG 3-mer #2: CGA 3-mer #3: GAT 3-mer #4: ATC 3-mer #5: TCA 3-mer #6: CAC

APPLICATIONS OF K-MER ANALYSIS

- Genome assembly: K-mers used to construct De Brujin graphs
- Detect bacterial contamination on eukaryotic genome assembly (CG content) discrepancies)
- Correcting NSG data
- Detect horizontal gene transfers
- Identification of CpG Islands
- Estimation of genome size and heterozygosity
- Genome assembly k-mer completeness

WHY ARE K-MERS SO POPULAR?

"Decomposing a sequence into its *k-mers* for analysis allows this set of fixed-size chunks to be analysed rather than the sequence, and this can be more efficient." (Bernardo Cavijo)

https://bioinfologics.github.io/post/2018/09/17/k-mer-counting-part-i-introduction/

.

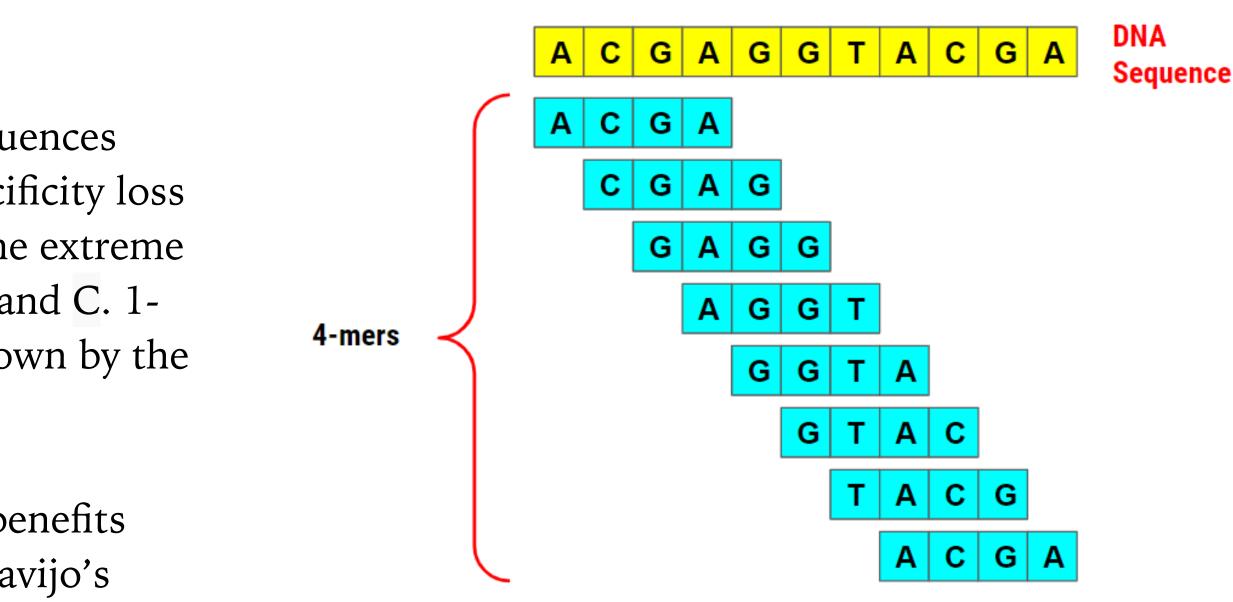
Choosing k: specificity vs. Sensitivity

- Using a *k* that is too small will result in many unrelated sequences being composed of the same *k*-mers, in a textbook case of specificity loss because there being very few possible *k*-mers of that size. In the extreme of the small k, k=1 only distinguishes two canonical *k*-mers: A and C. 1mer analysis is incredibly popular in biology, but it is best known by the name of GC content analysis.

- Using extremely large *k* values would sacrifice many of the benefits and sensitivity of *k-mer* analyses in the first place. (Bernado Cavijo's post)

Why do we chose k=31 so often?

One reason is: it is specific enough that a large number of them are unique both in mammalian-sized genomes and in bacterial genome databases.



ce

COUNT AND HISTO

Counting *k-mers* in a (small) genome

We will start with an easy example first: the phi-X174 genome has 5386 bp and is a simple non-repetitive genome.

We can use kat hist to count 27-mers on the genome and check how many times each 27-mer appears (we start with k = 27 because KAT uses that as default):

```
$ kat hist -o phiX.hist phiX.fasta
```

Checking the phiX.hist histogram (A.K.A. kmer spectrum) file, every 27-mer in the genome appears only once. After the header lines starting with #, every line has a copy number (A.K.A. frequency) and a number of k-mers.

```
# Title:27-mer spectra for: phiX.fasta
# XLabel:27-mer frequency
# YLabel:# distinct 27-mers
# Kmer value:27
# Input 1:../genomes/phiX.fasta
###
1 5360
2 0
3 0
4 0
...
```

Bernardo Cavijo's post

COUNT AND HISTO

\$ kat hist -o phiX_9mer.hist -m 9 phiX.fasta

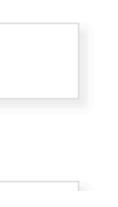
Then the phiX_9mer.hist file looks like this:

Title:9-mer spectra for: phiX.fasta # XLabel:9-mer frequency \$ kat hist -o phiX_8mer.hist -m 8 phiX.fasta # YLabel:# distinct 9-mers # Kmer value:9 Now the histogram file looks like this: # Input 1:phiX.fasta ### # Title:8-mer spectra for: phiX.fasta 1 4972 # XLabel:8-mer frequency 2 189 # YLabel:# distinct 8-mers 38 # Kmer value:8 41 # Input 1:phiX.fasta 50 ### 60 1 4159 70 2 491 80 3 67 90 48 . . . 51 60 70

80

90

Here, only 4159 8-mers are unique, out of 4726 distinct 8-mers, that are present in the genome's 5377 total 8-mers.

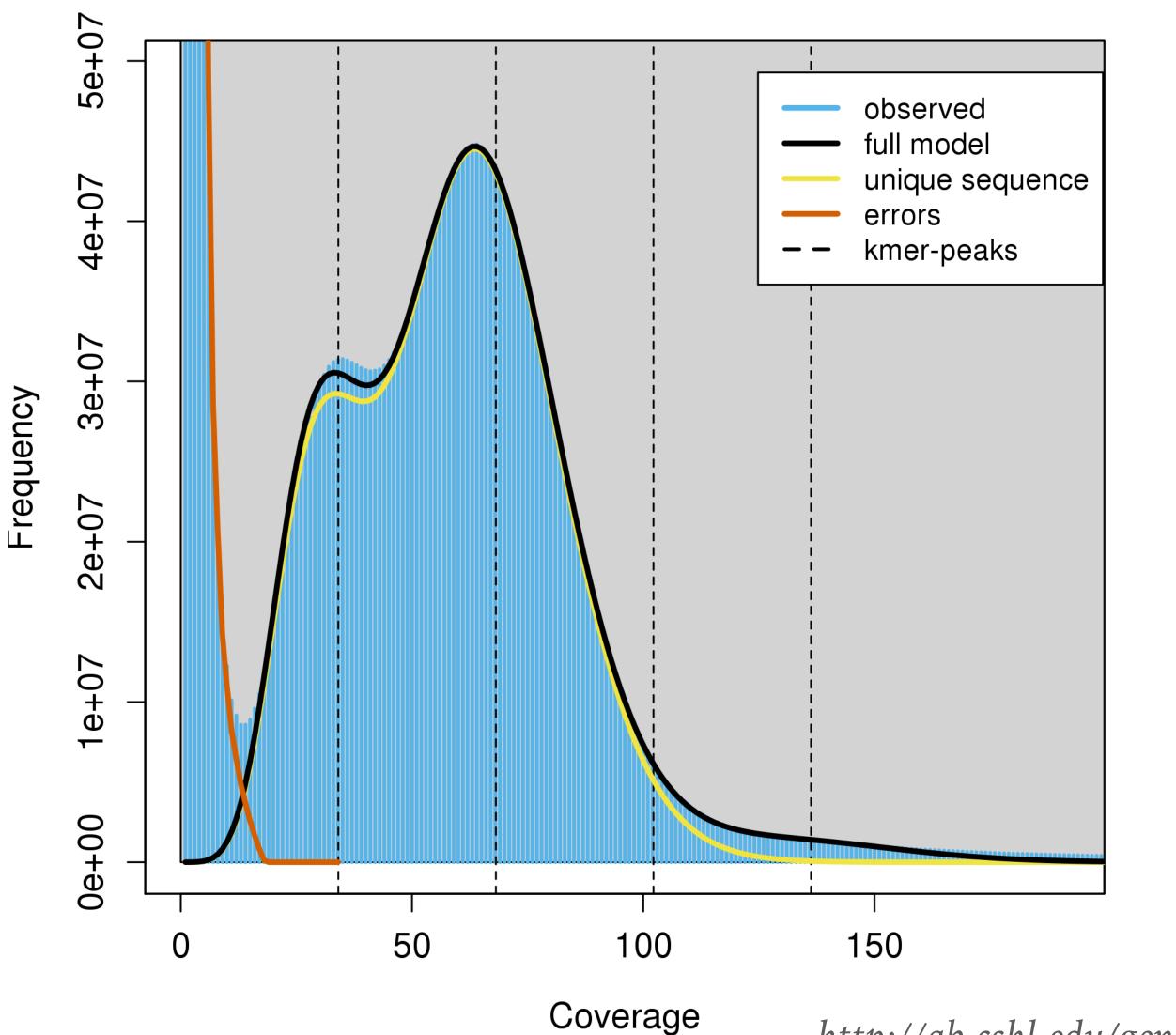


Bernardo Cavijo's post

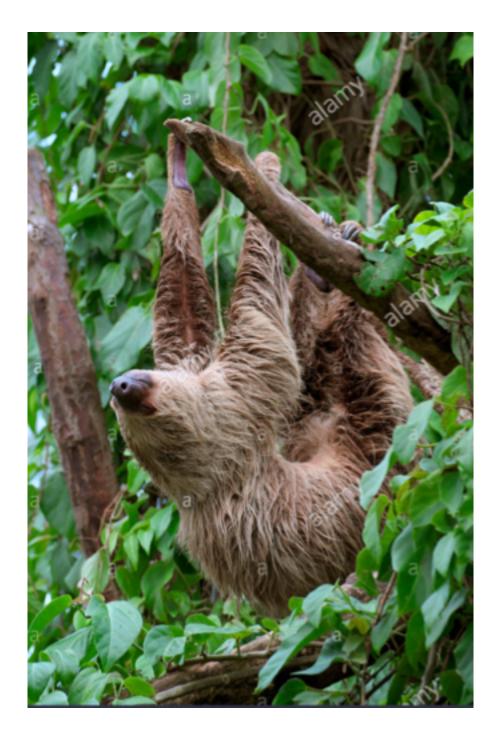
A TYPICAL KMER PLOT FOR A DIPLOID SPECIES

GenomeScope Profile

len:3,249,909,355bp uniq:64.2% het:0.947% kcov:34.1 err:0.385% dup:2.79% k:21



Choloepus didactylus (VGP)



http://qb.cshl.edu/genomescope/analysis.php?code=bVuZNlhwn2tVCHhRN71I

A TYPICAL KMER PLOT FOR A DIPLOID SPECIES WITH <u>HIGH HETEROZYGOSITY</u>

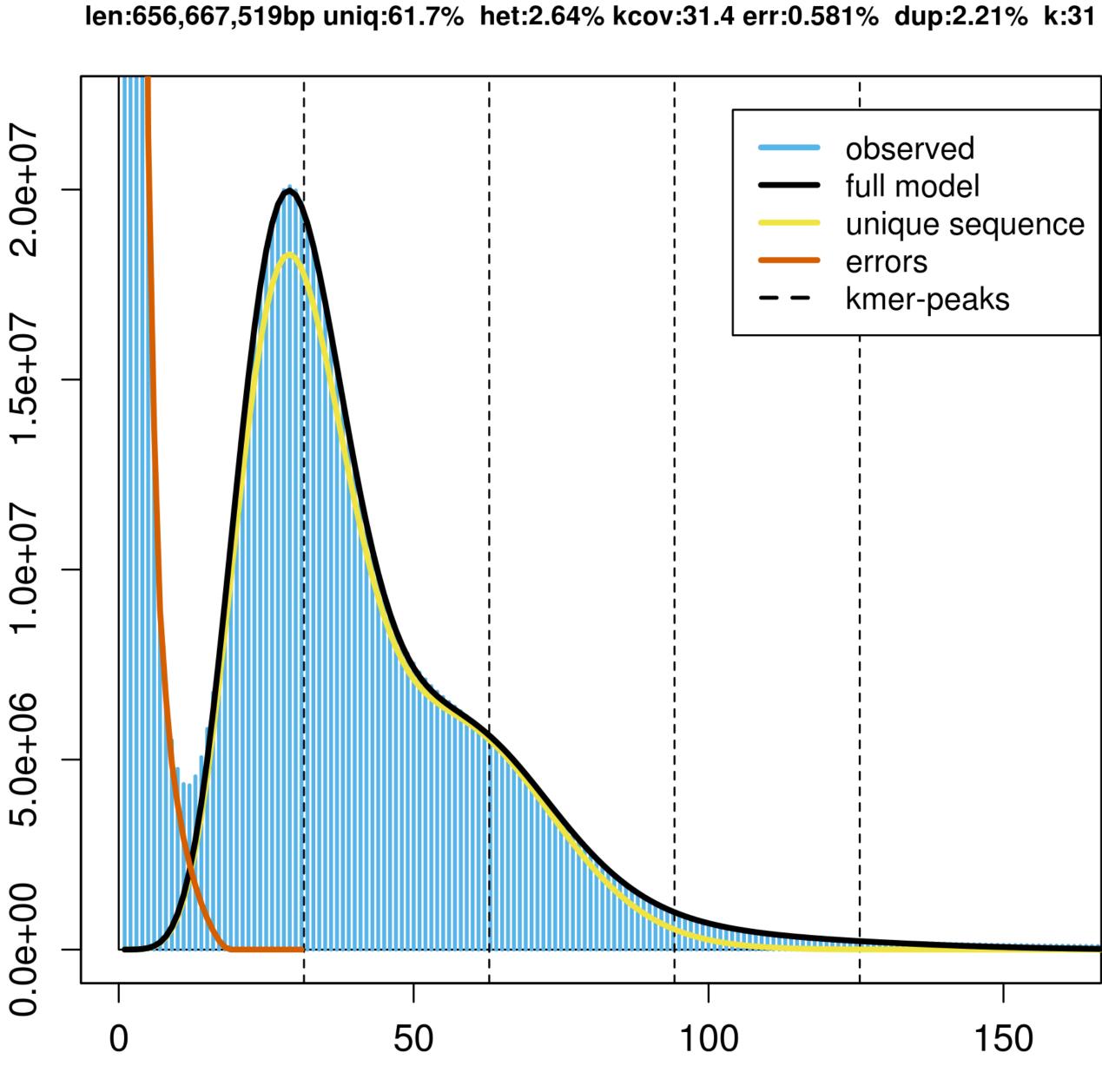
Blastobasis lacticolella (DToL)

Wakely's dowd



frequency

ilBlaLact1 GenomeScope Profile



k-mer coverage

A TYPICAL KMER PLOT FOR A DIPLOID SPECIES WITH LOW Heterozygosity

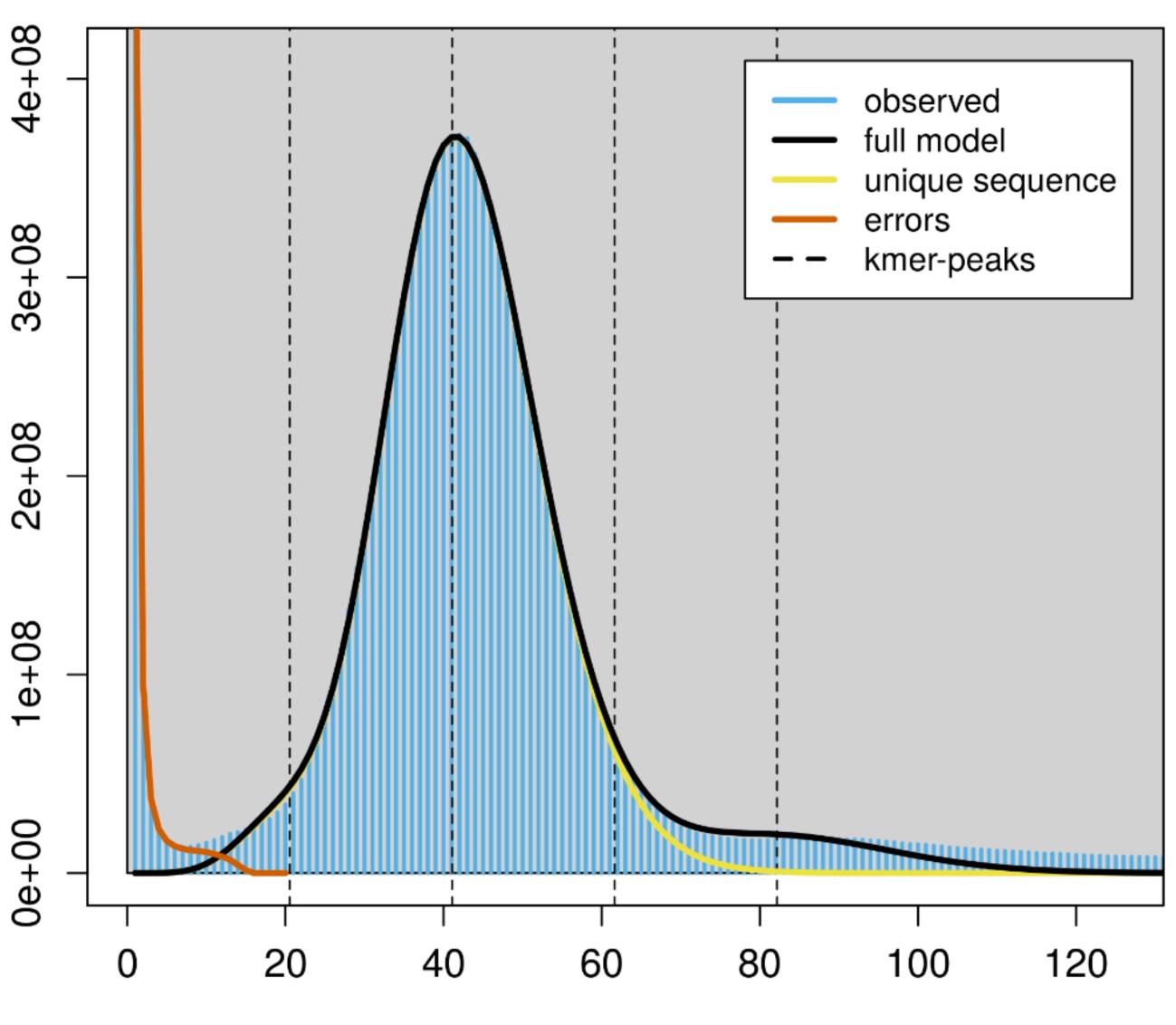
Urtica urens



Coverage*Frequency

GenomeScope Profile

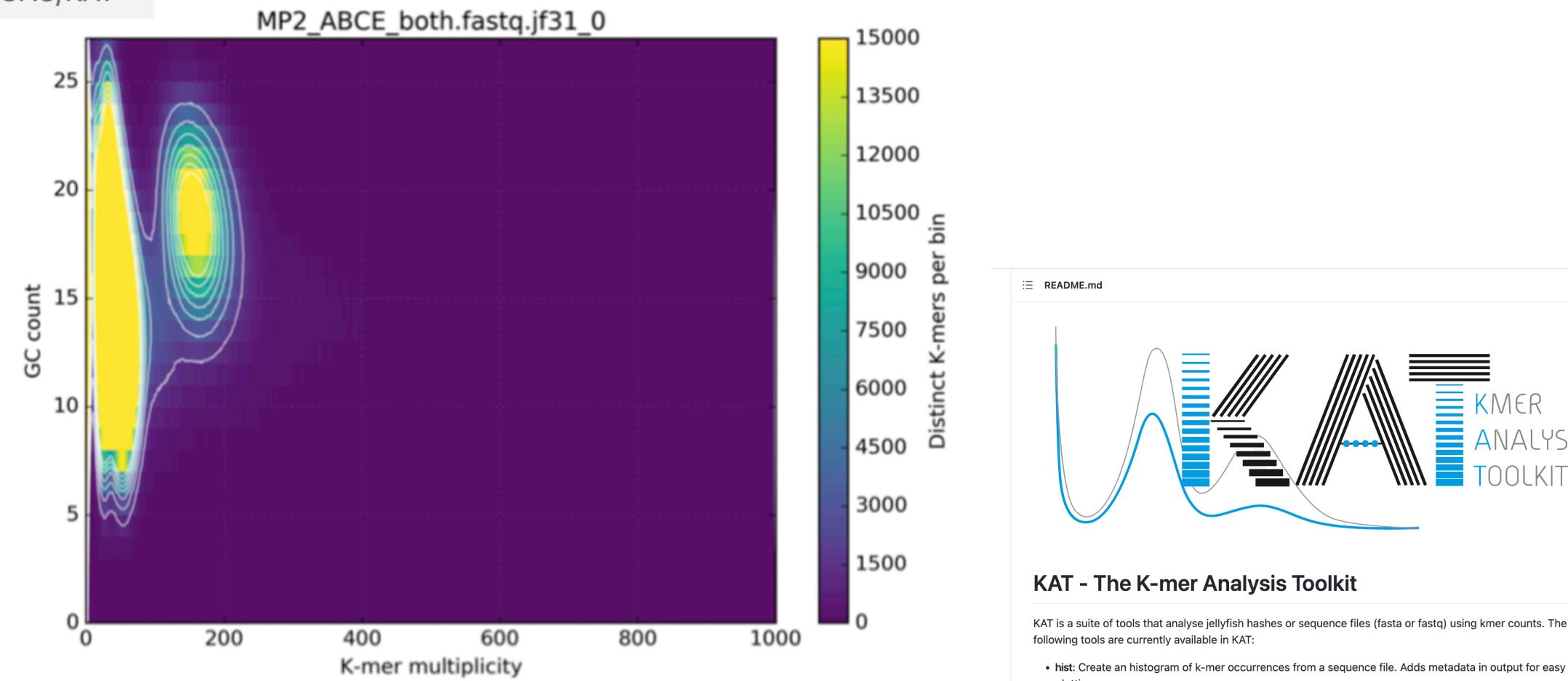
len:438,762,965bp uniq:51.8% aa:99.8% ab:0.183% kcov:20.5 err:0.135% dup:1.2 k:31 p:2



Coverage

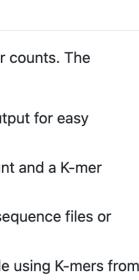
SPOTTING BACTERIAL CONTAMINATION: KMER AND ITS GC CONTENT

github.com/TGAC/KAT



You can use KAT to plot this!

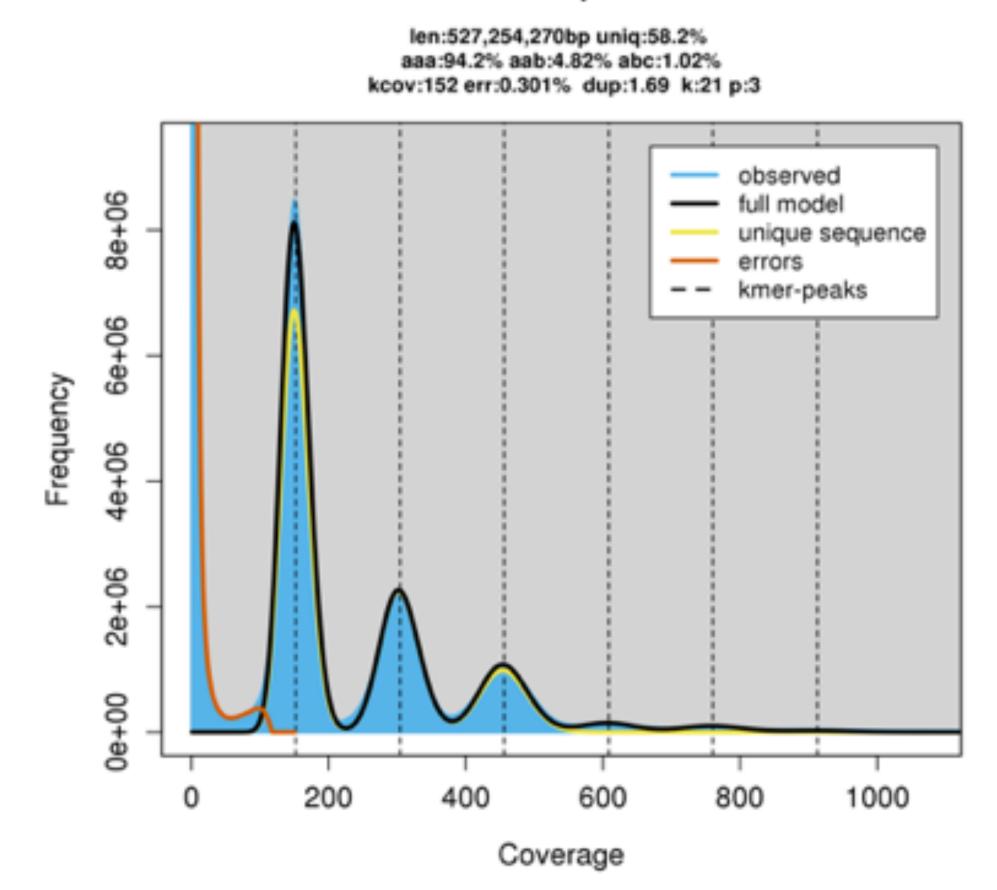
- hist: Create an histogram of k-mer occurrences from a sequence file. Adds metadata in output for easy plotting.
- gcp: K-mer GC Processor. Creates a matrix of the number of K-mers found given a GC count and a K-mer count.
- comp: K-mer comparison tool. Creates a matrix of shared K-mers between two (or three) sequence files or hashes.
- sect: SEquence Coverage estimator Tool. Estimates the coverage of each sequence in a file using K-mers from another sequence file.



Tubastraea tagusensis

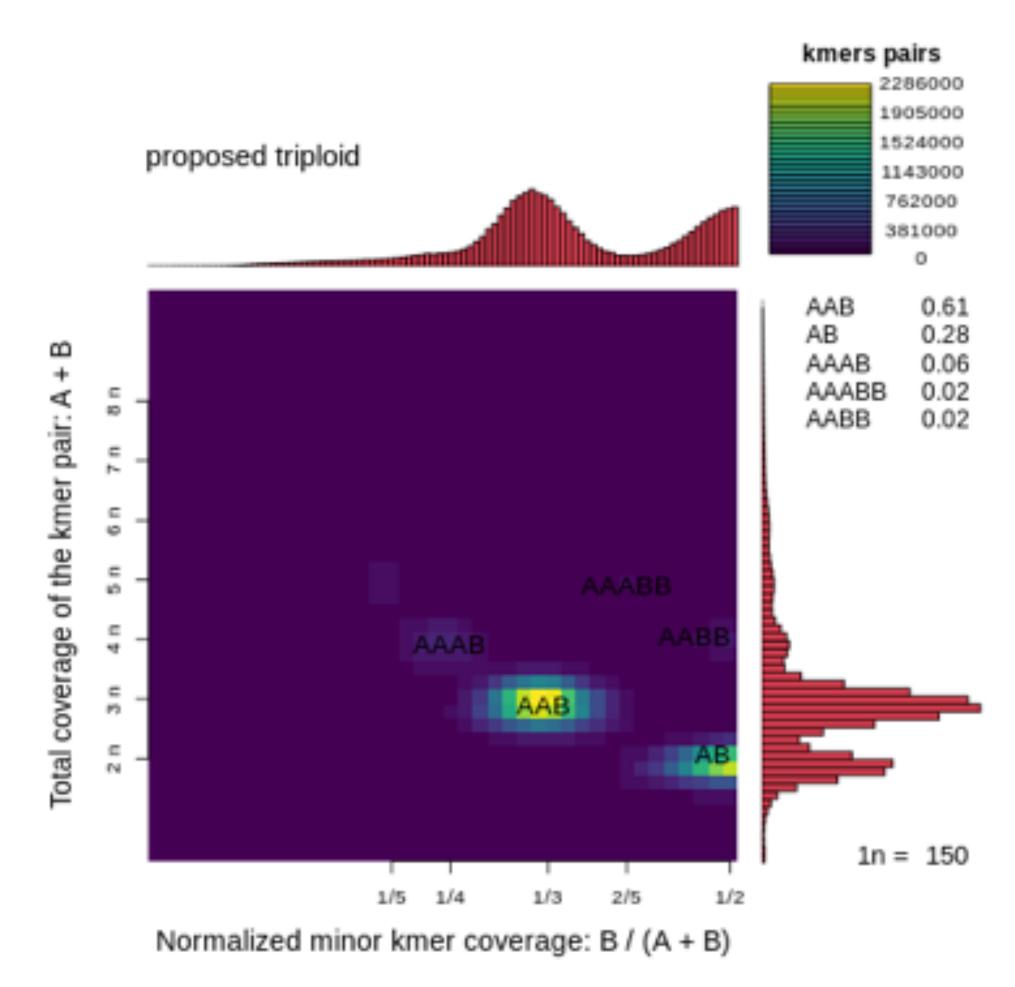
KMER PROFILE FOR A TRIPLOID SPECIES

GenomeScope Profile

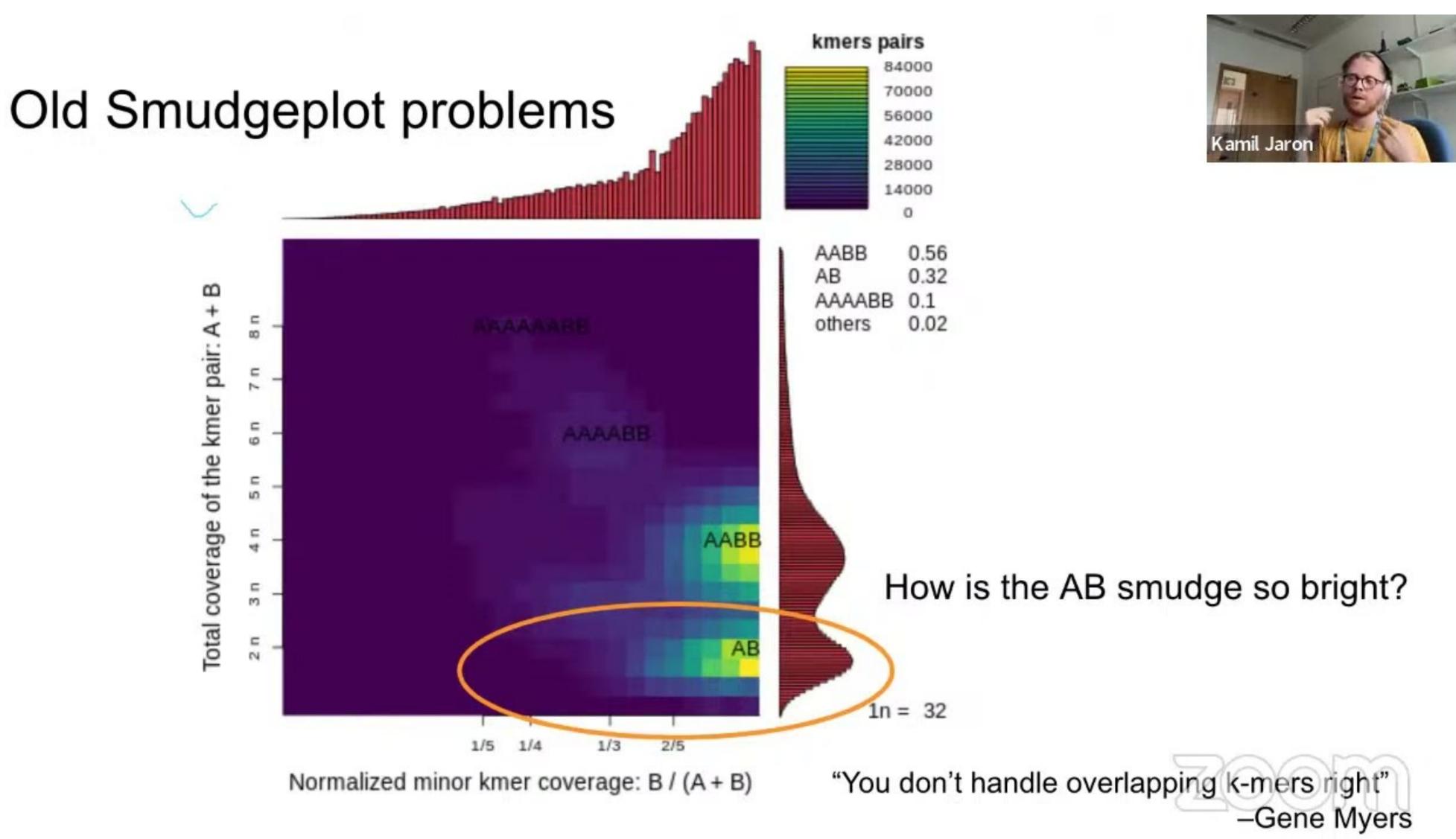








MORE ON SMUDGEPLOTS



github.com/thegenemyers/MERQURY.FK

:= README.md

MerquryFK & KatFK: Fast & Simple

Authors: Gene Myers & Arang Rhie First: Feb 24, 2021 Current: Aug 11, 2021

- Introduction
 - HAPmaker
 - CNplot
 - ASMplot
 - HAPplot
 - MerguryFK
 - KatComp
 - KatGC
 - PloidyPlot

Introduction

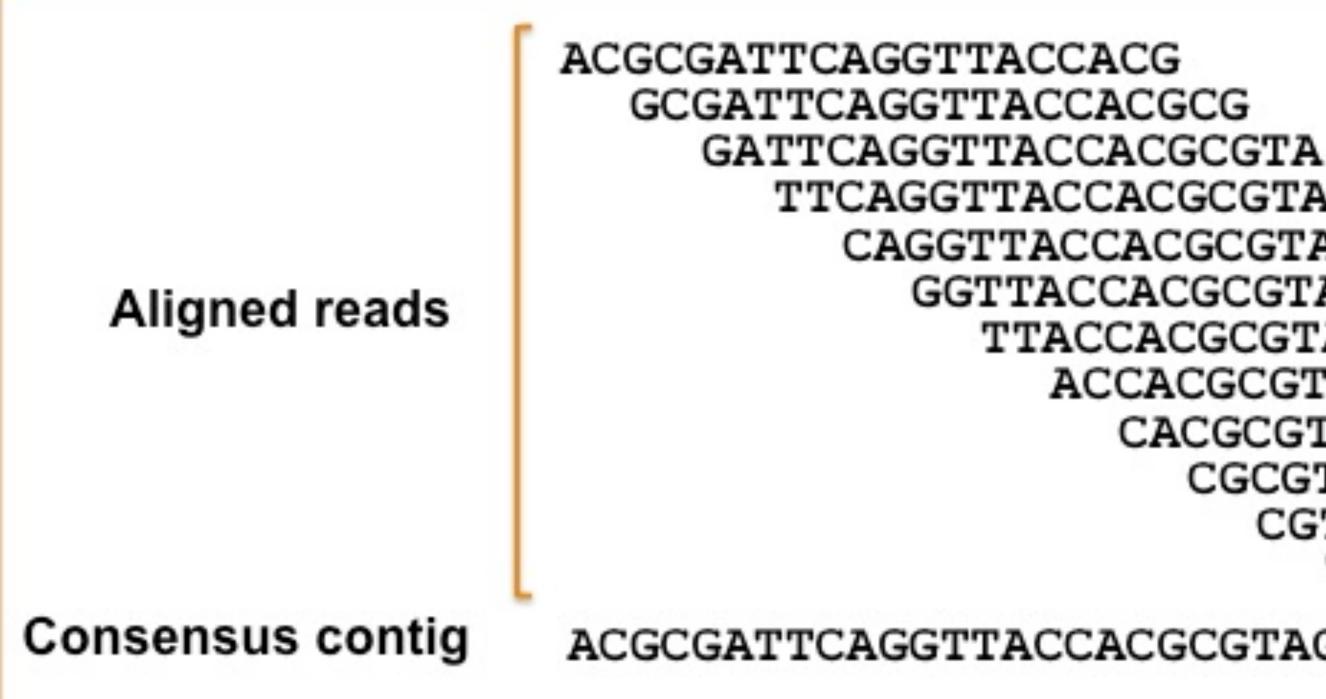
The original Mergury is a collection of R, Java, and shell scripts for producing k-mer analysis plots of genomic sequence data and assemblies with meryl as its core k-mer counter infra-structure. MerguryFK replaces meryl with the FastK k-mer counter suite to considerably speed up analyses. Moreover, all the R, Java, and shell scripts have been refactored into a typical collection of UNIX command line tools that the user will hopefully experience as easier to comprehend and invoke. In addition, we have realized some analyses, KatComp and KatGC, that one finds only in the somewhat similar KAT k-mer suite developed at the Earlham Institute. Lastly, we include in this collection, **PloidyPlot** which is an improved version of the ploidy plotting tool SmudgePlot.

There are some general conventions for our tools programmed for your convenience. First, suffix extensions need not be given for arguments of a known type. For example, if an argument is a fasta or fastq with root name "foo" without extensions, then our commands will look for foo.fasta, foo.fa, foo.fastq, and foo.fq if you specify foo as the argument. Second, option arguments (those that begin with a '-') can be in any order and in any position relative to the non-optional primary arguments (which must be given in the order specified). We find this pretty convenient when for example you have typed out an entire CNplot command (2. below) but forgot that you wanted .pdf's. All you do is append -pdf to what you've already typed and then hit return. So for example, CNplot -w4 -h3 Assembly -ls Reads -pdf is acceptable input.

For the tools that take a FastK k-mer table as an input, we use the syntax <name>[.ktab], to describe it on the command line indicating that the .ktab extension is optional as per the convention above. Regardless of whether the extension is given, it is expected that the associated histogram file <name>.hist is also present (this file is always produced by a run of FastK that produces a k-mer table). Also note carefully that these tables must be produced with the option -t or -t1 set so that all k-mers that occur 1 or more times in the underlying data set are in the table.

KNOWING THE CHALLENGE, YOU GO AND BUILD CONTIGS WITH ASSEMBLERS

CONTIG

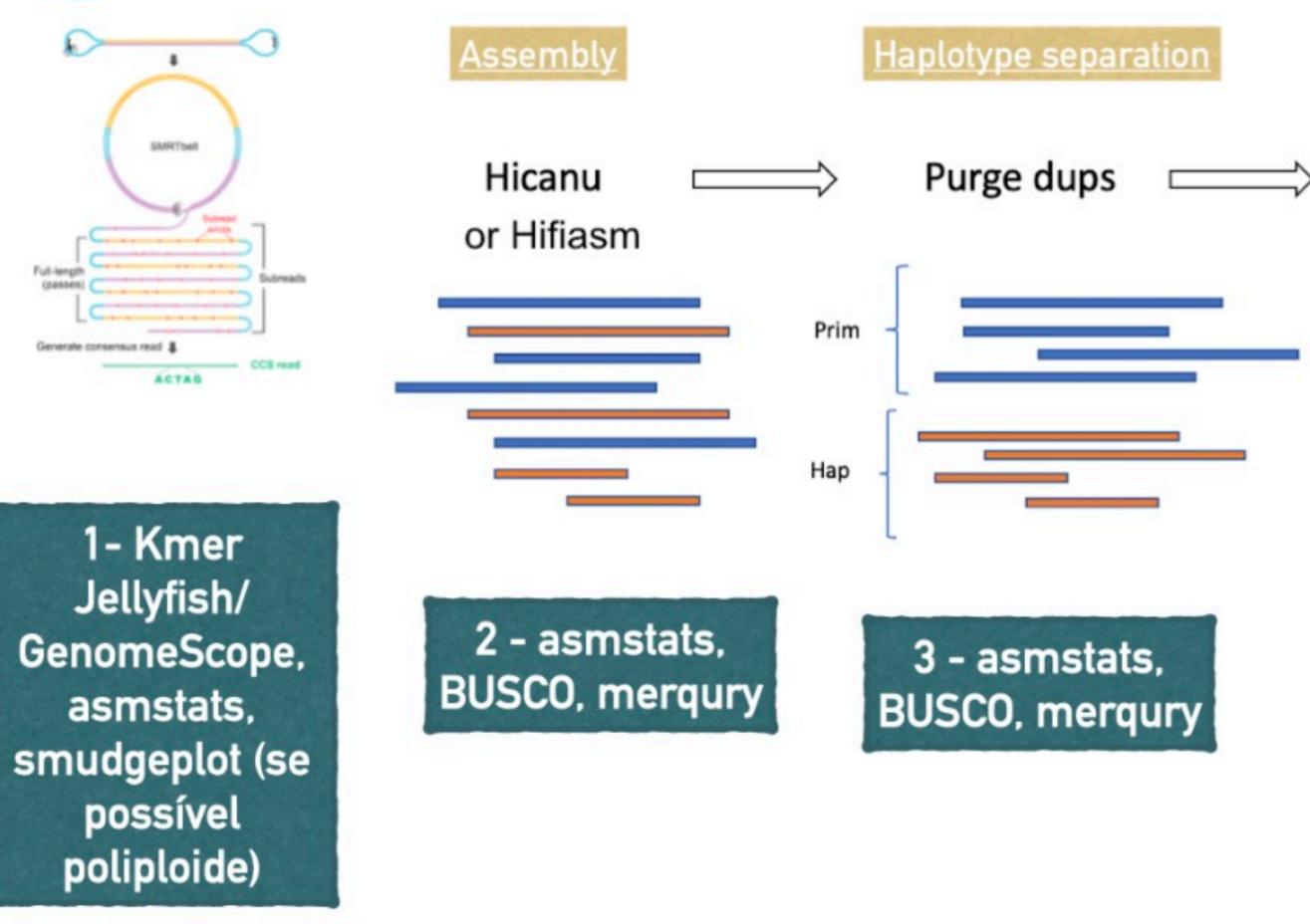


TTCAGGTTACCACGCGTAGC CAGGTTACCACGCGTAGCGC GGTTACCACGCGTAGCGCAT TTACCACGCGTAGCGCATTA ACCACGCGTAGCGCATTACA CACGCGTAGCGCATTACACA CGCGTAGCGCATTACACAGA CGTAGCGCATTACACAGATT TAGCGCATTACACAGATTAG

ACGCGATTCAGGTTACCACGCGTAGCGCATTACACAGATTAG

DToL Current Pipeline

Sequencing technologies: PacBio HiFi + HiC (Arima or Qiagen) ٠





For mitochondria genome assembly



Yahsscaffolding (Arima or Qiagen HiC)



Curated assembly

4 - asmstats, BUSCO, mergury, HiC heatmap

5 - asmstats, BUSCO, mergury, HiC heatmap

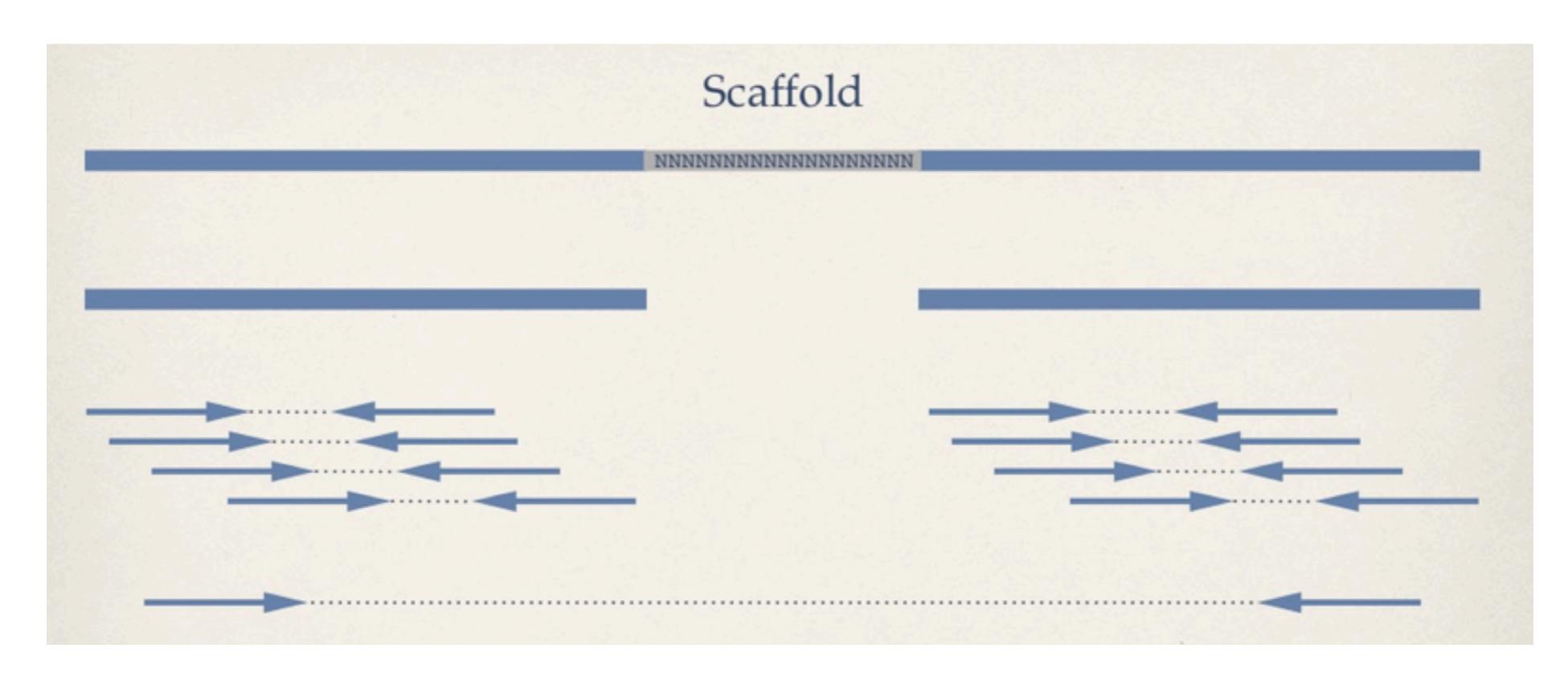




DOES MY ASSEMBLED SIZE CORRESPONDS WITH MY ESTIMATED GENOME SIZE?

Genomics is a game of going back and forth

Scaffolding methods



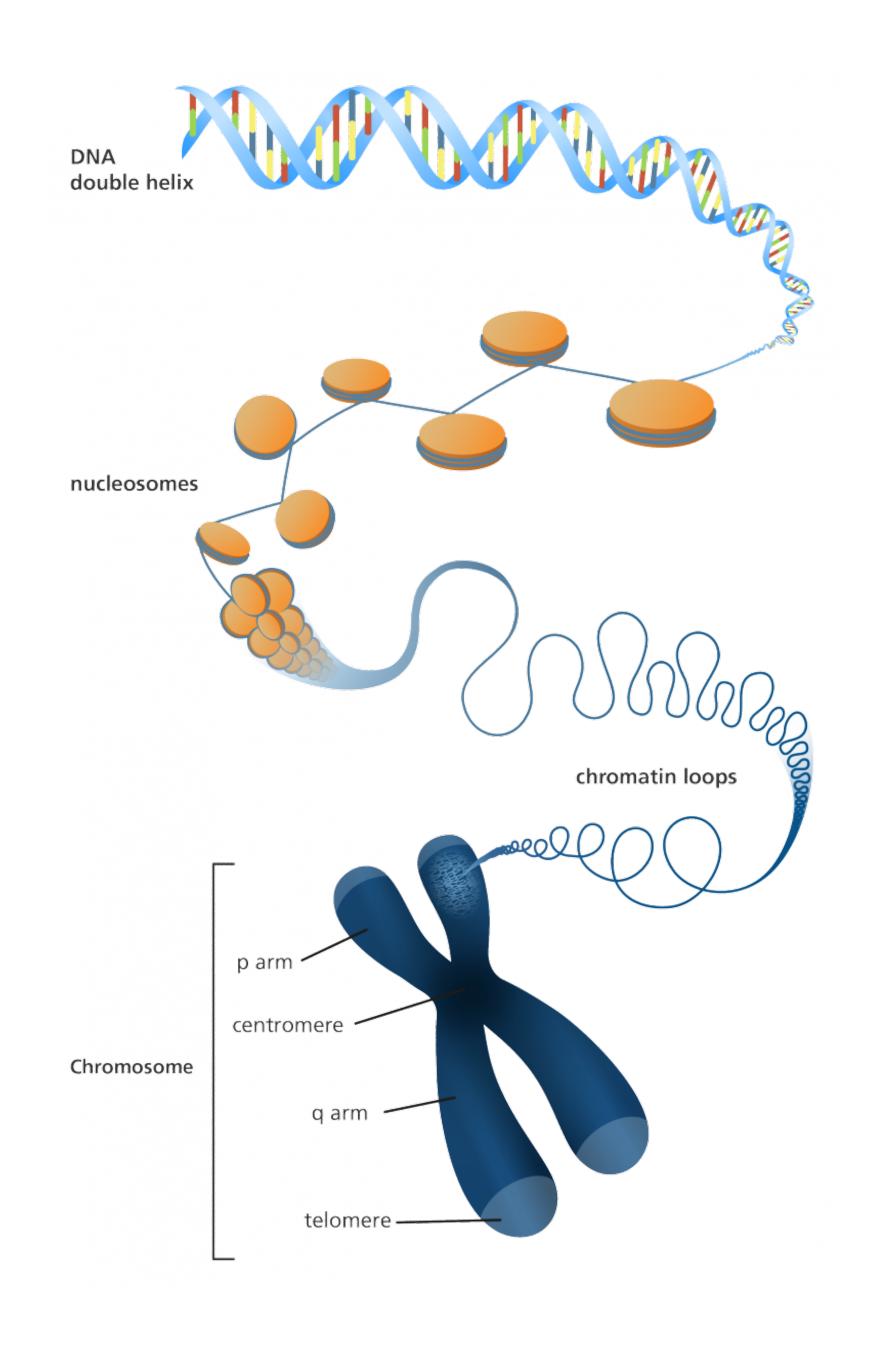
Scaffold: joining and orienting contigs Scaffolding methods: mate-pairs (blerg), optical maps (bionano), <u>Hi-C</u>, Nanopore UltraLong reads



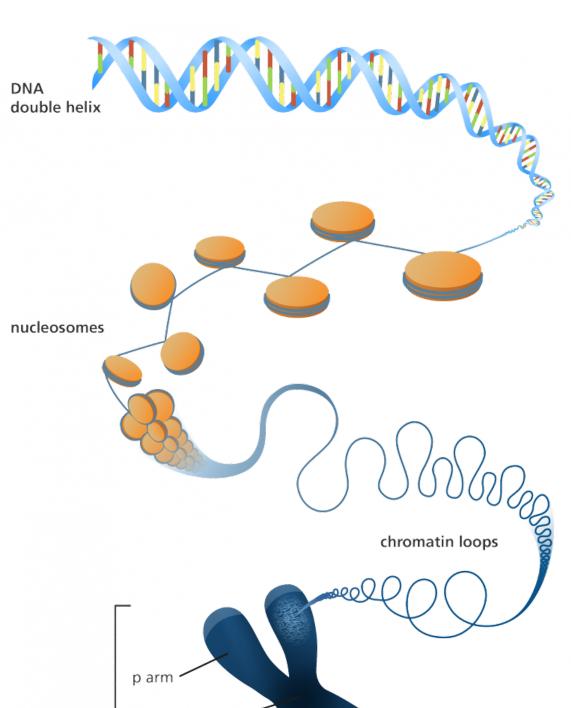
HOW DO I BUILD UP SCAFFOLDS AND CHROMOSOMES?

Hi-C and Ultralong Nanopore

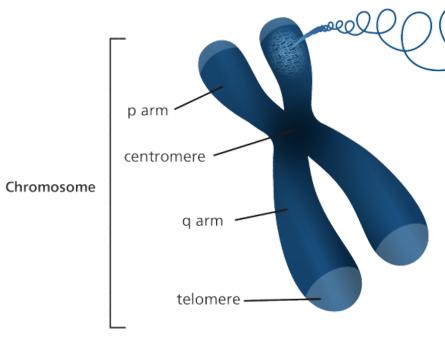
The human genome consists of over 3 billion nucleotides and is contained within 23 pairs of chromosomes. If the chromosomes were aligned end to end and the DNA stretched, the genome would measure roughly 2 meters long. Yet the genome functions within a sphere smaller than a tenth of the thickness of a human hair (10 micron). ... the genome does not exist as a simple one-dimensional polymer; instead the genome folds into a complex compact threedimensional structure. (Lajoie et al 2015)



- The organisation of the chromatin in the nucleus is extremely relevant to biological function at the gene level as well as the global nuclear level.
- The study of the packaging and organisation of chromatin in the nucleus will shed light on:
- the spatial aspects of gene regulation
- chromosome morphogenesis
- genome stability
- genome transmission
- biophysics of chromatin
- pathologies related to genome instability or nuclear morphology



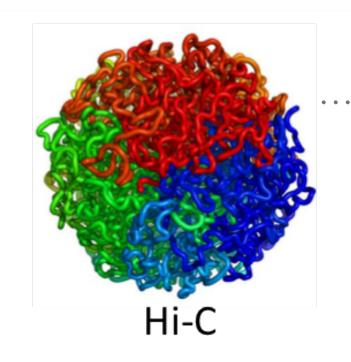




Published in final edited form as: Science. 2009 October 9; 326(5950): 289-293. doi:10.1126/science.1181369.

Comprehensive mapping of long range interactions reveals folding principles of the human genome

Erez Lieberman-Aiden^{1,2,3,4,*}, Nynke L. van Berkum^{5,*}, Louise Williams¹, Maxim Imakaev², Tobias Ragoczy^{6,7}, Agnes Telling^{6,7}, Ido Amit¹, Bryan R. Lajoie⁵, Peter J. Sabo⁸, Michael O. Dorschner⁸, Richard Sandstrom⁸, Bradley Bernstein^{1,9}, M. A. Bender¹⁰, Mark Groudine^{6,7}, Andreas Gnirke¹, John Stamatoyannopoulos⁸, Leonid A. Mirny^{2,11}, Eric S. Lander^{1,12,13,†}, and Job Dekker^{5,†} ¹ Broad Institute of Harvard and MIT, Cambridge, Massachusetts 02139, USA.



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⁵ Program in Gene Function and Expression and Department of Biochemistry and Molecular Pharmacology, University of Massachusetts Medical School, Worcester, Massachusetts 01605, USA.

⁶ Fred Hutchinson Cancer Research Center, Seattle, Washington 98109, USA.

⁷ Department of Radiation Oncology, University of Washington School of Medicine, University of Washington, Seattle, Washington 98195, USA. ⁸ Department of Genome Sciences, University of Washington, Seattle, Washington 98195, USA

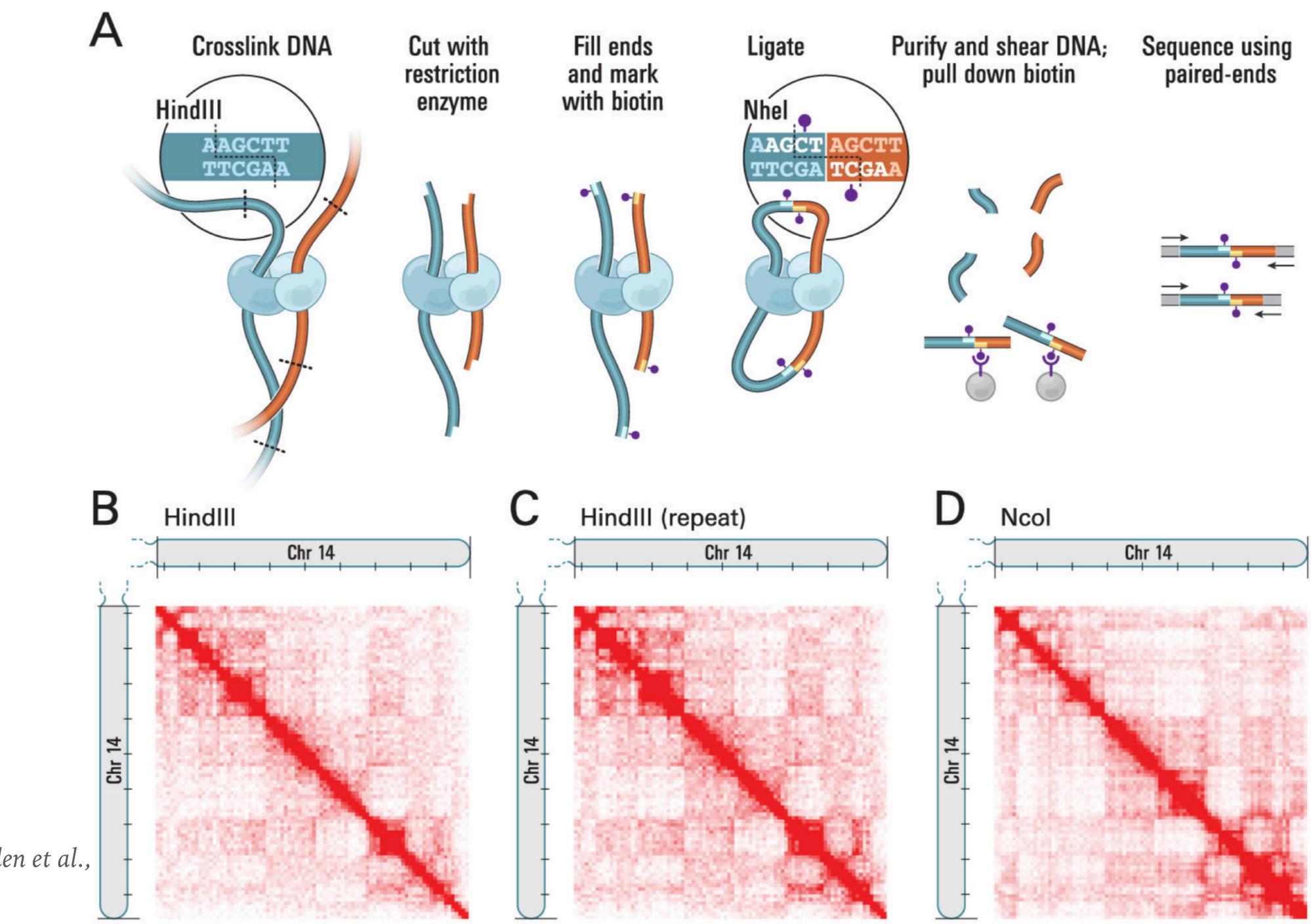
⁹ Department of Pathology, Harvard Medical School, Boston, Massachusetts 02115, USA.

¹⁰ Department of Pediatrics, University of Washington, Seattle, Washington 98195, USA.

¹¹ Department of Physics, MIT, Cambridge, Massachusetts 02139, USA.

¹² Department of Biology, MIT, Cambridge, Massachusetts 02139, USA.

¹³ Department of Systems Biology, Harvard Medical School, Boston MA 02115.



Lieberman-Aiden et al., 2010

I-C

Intrachromosomal contact probability is on average much higher than interchromosomal. Interaction probability rapidly decays with increasing genomic distance.

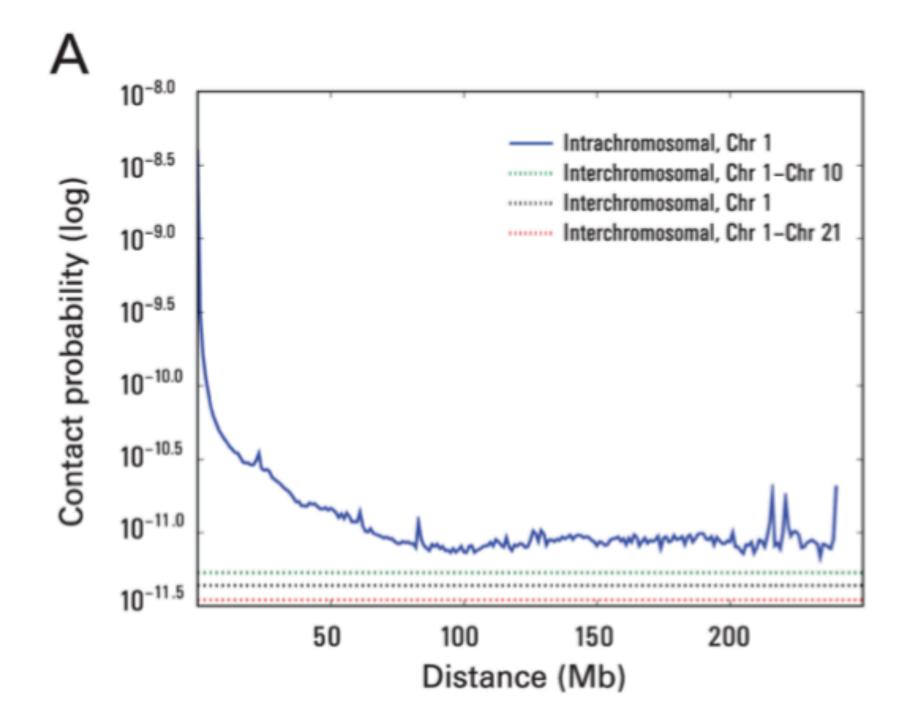
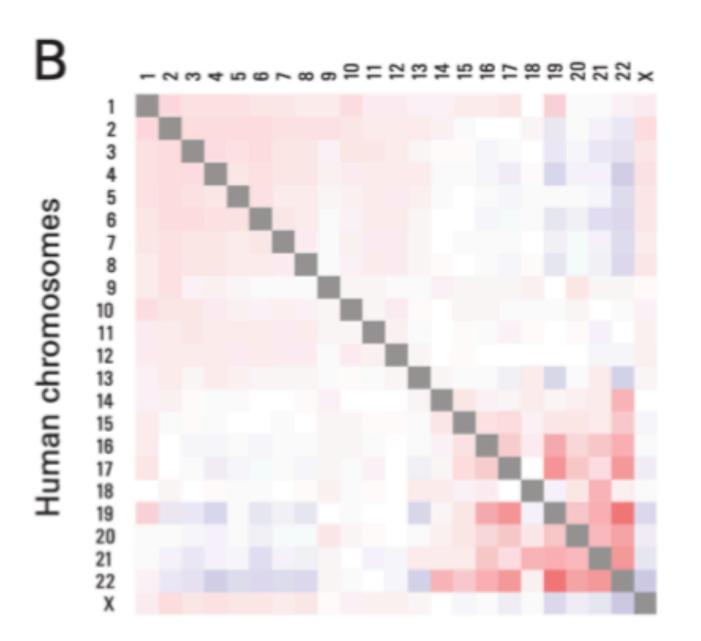
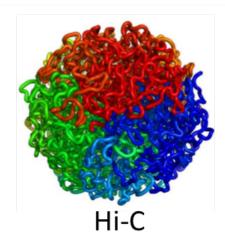


Fig. 2.

The presence and organization of chromosome territories. (A) Probability of contact decreases as a function of genomic distance on chromosome 1, eventually reaching a plateau at ~90M (blue). The level of interchromosomal contact (black dashes) differs for different pairs of chromosomes; loci on chromosome 1 are most likely to interact with loci on chromosome 10 (green dashes) and least likely to interact with loci on chromosome 21 (red dashes). Interchromosomal interactions are depleted relative to intrachromosomal interactions. (B) Observed/expected number of interchromosomal contacts between all pairs of chromosomes. Red indicates enrichment, and blue indicates depletion (up to twofold). Small, gene-rich chromosomes tend to interact more with one another.



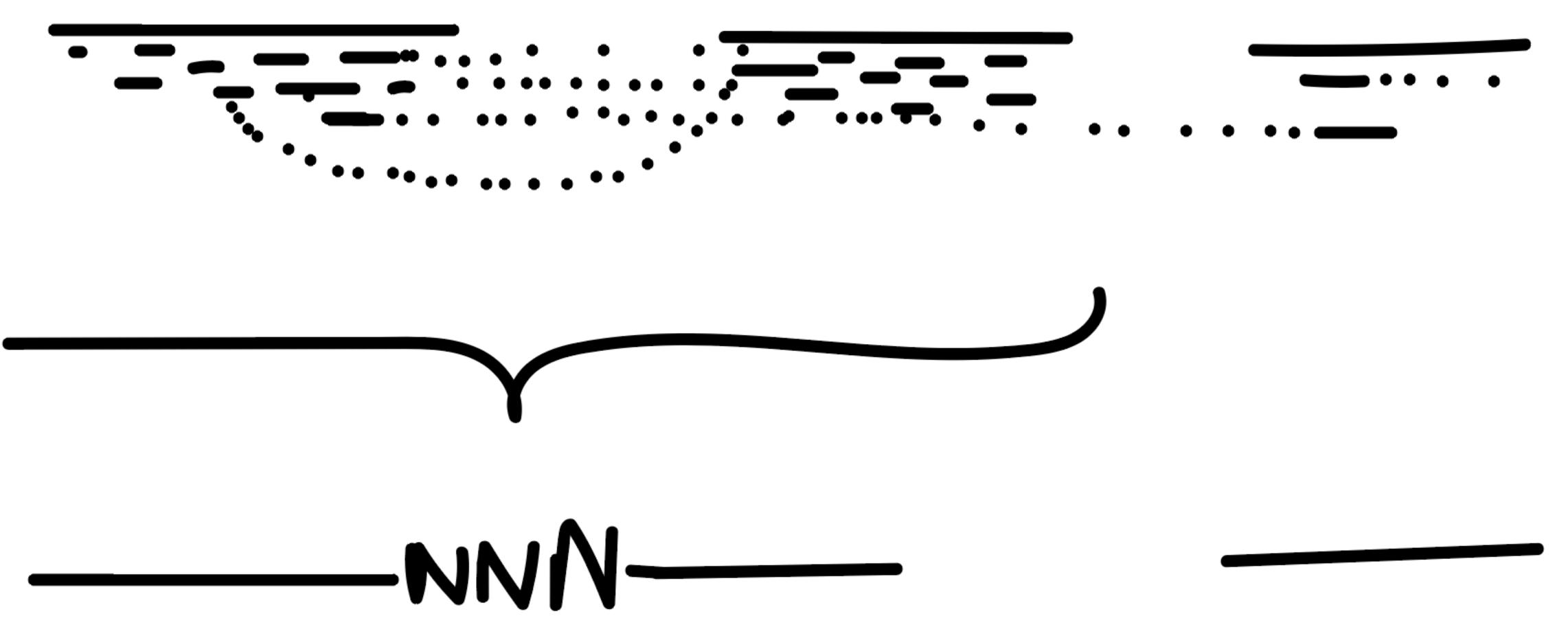


Human chromosomes

HOW TO DO HI-C SEQUENCING

- You have a protocol for Hi-C extraction
- This is sequenced as short Illumina reads
- -5SP)
- Ran YaHS and/or Salsa for scaffolding
- Build and look at Hi-C HeatMaps

> You map the Hi-C data to your built contigs (Arima Mapping pipeline or BWA men



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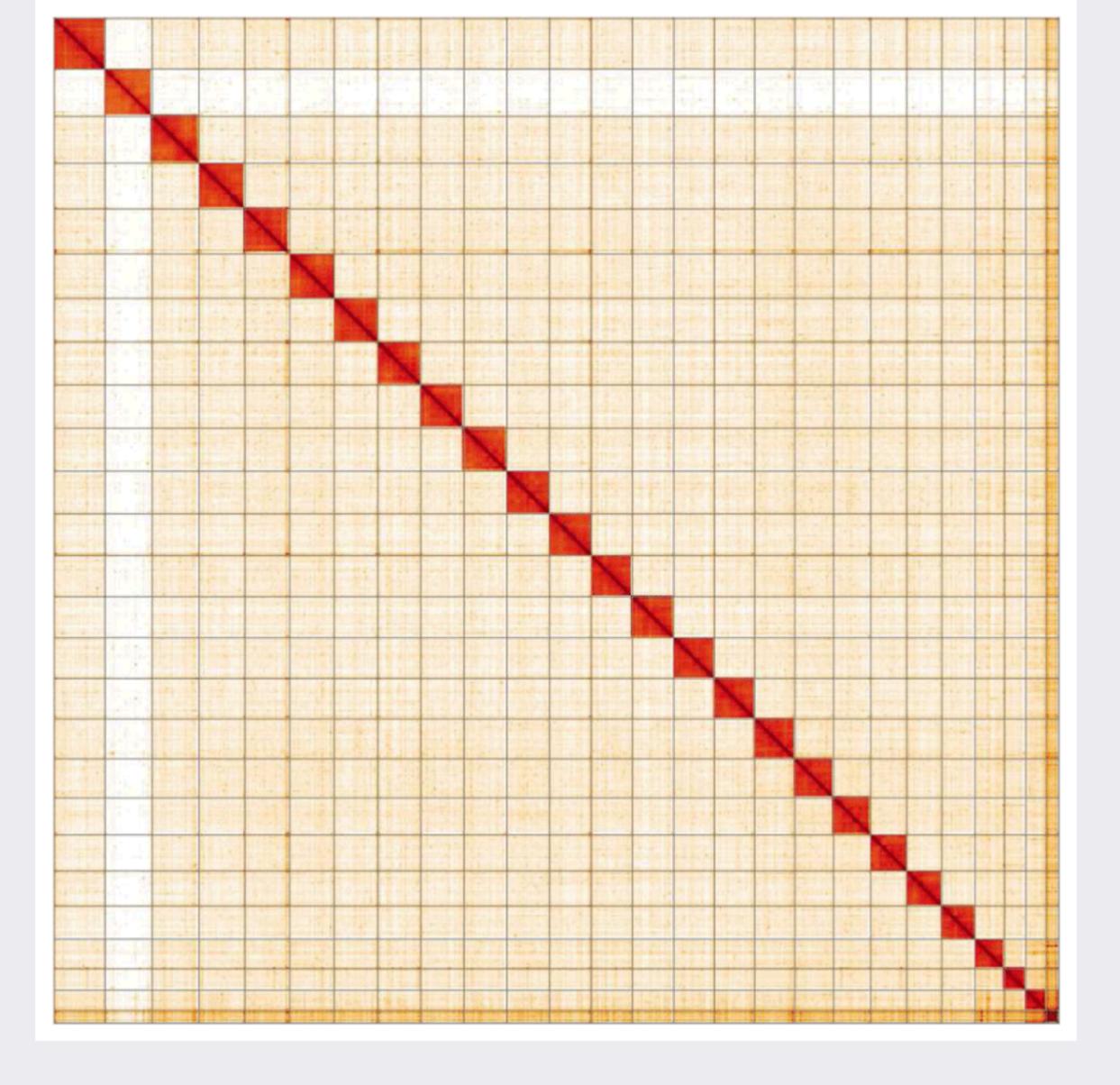
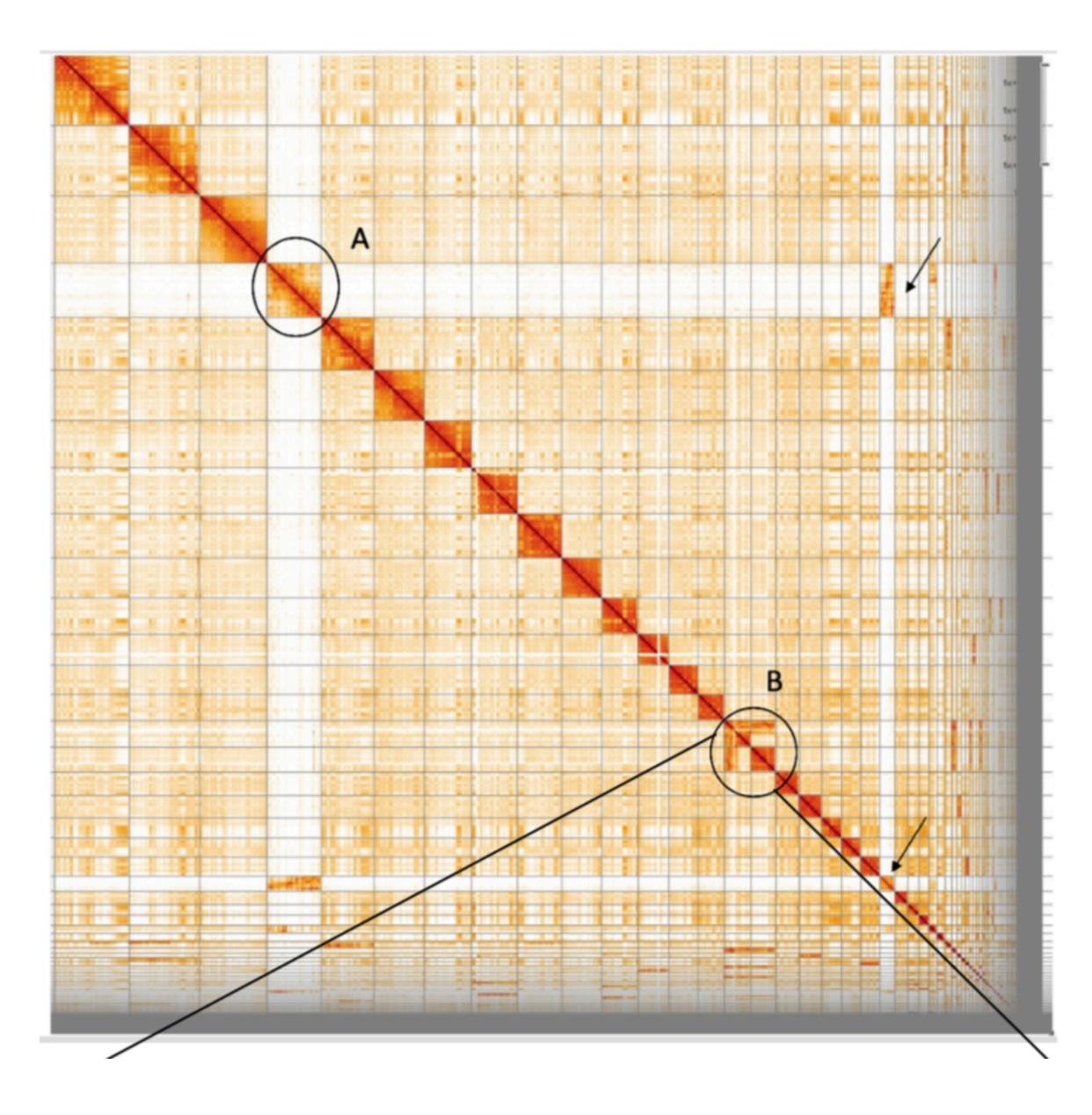


Figure 5. Genome assembly of *Pieris rapae*, ilPieRapa1.1: Hi-C contact map.

Hi-C contact map of the ilPieRapa1.1 assembly, visualised in HiGlass. Chromosomes are given in size order from left to right and top to bottom.

YOU DO MORE THAN SCAFFOLDING WITH HI-C: YOU SEE BIOLOGY



Choloepus didactylus VGP Non-curated output 3.2 Gb, 281 scaffolds, N50 = 161 Mb

SHARED SEQUENCES: CENTROMERES

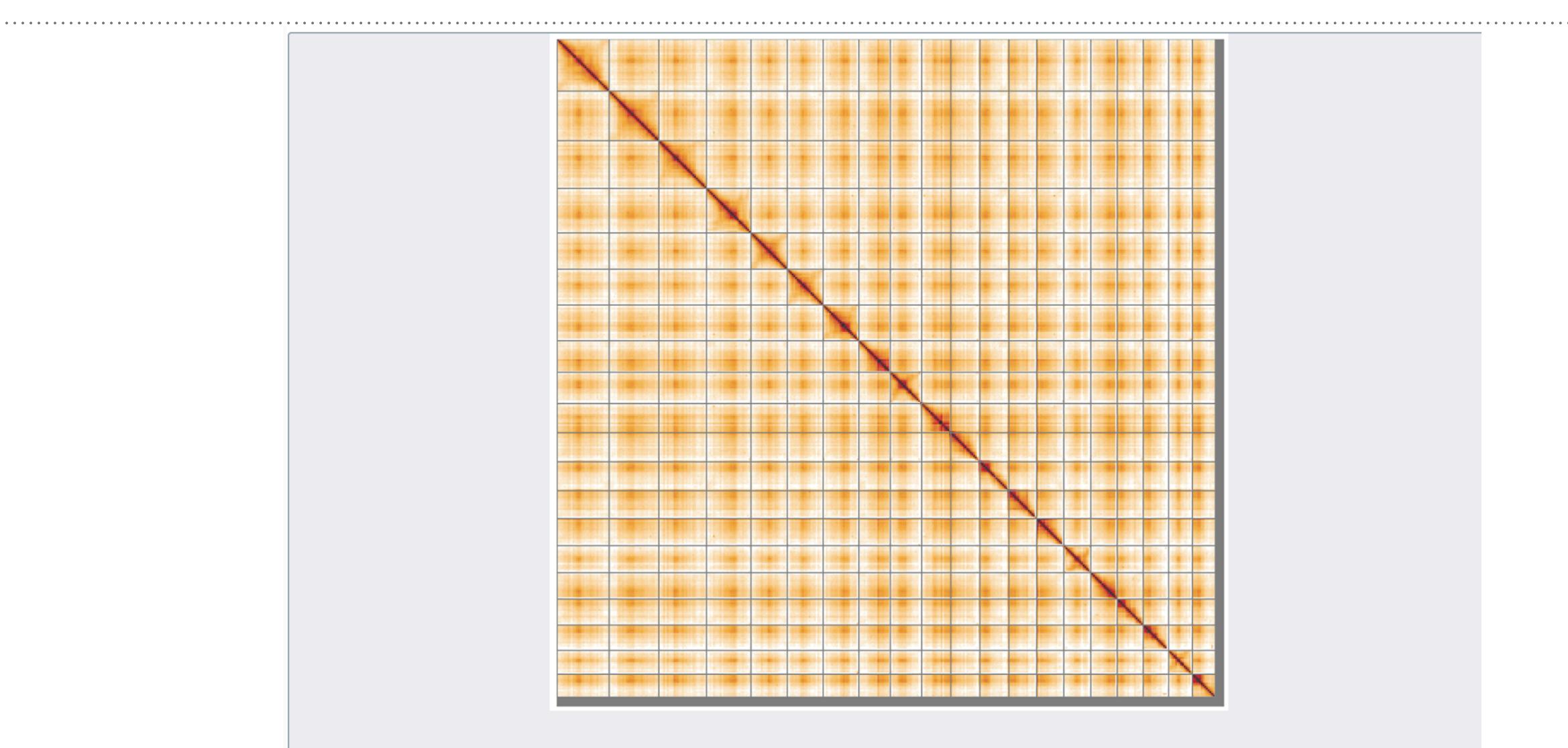


Figure 5. Genome assembly of *llex aquifolium*, drlleAqui2.1: Hi-C contact map of the drlleAqui2.1 assembly, visualised using HiGlass.

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YaHS: yet another Hi-C scaffolding tool

Chenxi Zhou^{1,2}, Shane A. McCarthy^{1,2}, and Richard Durbin^{1,2,*}

¹ Department of Genetics, University of Cambridge, Downing Street, Cambridge, CB2 3EH, UK ² Wellcome Sanger Institute, Wellcome Genome Campus, Hinxton, Cambridge, CB10 1SA, UK

* Correspondence: rd109@cam.ac.uk

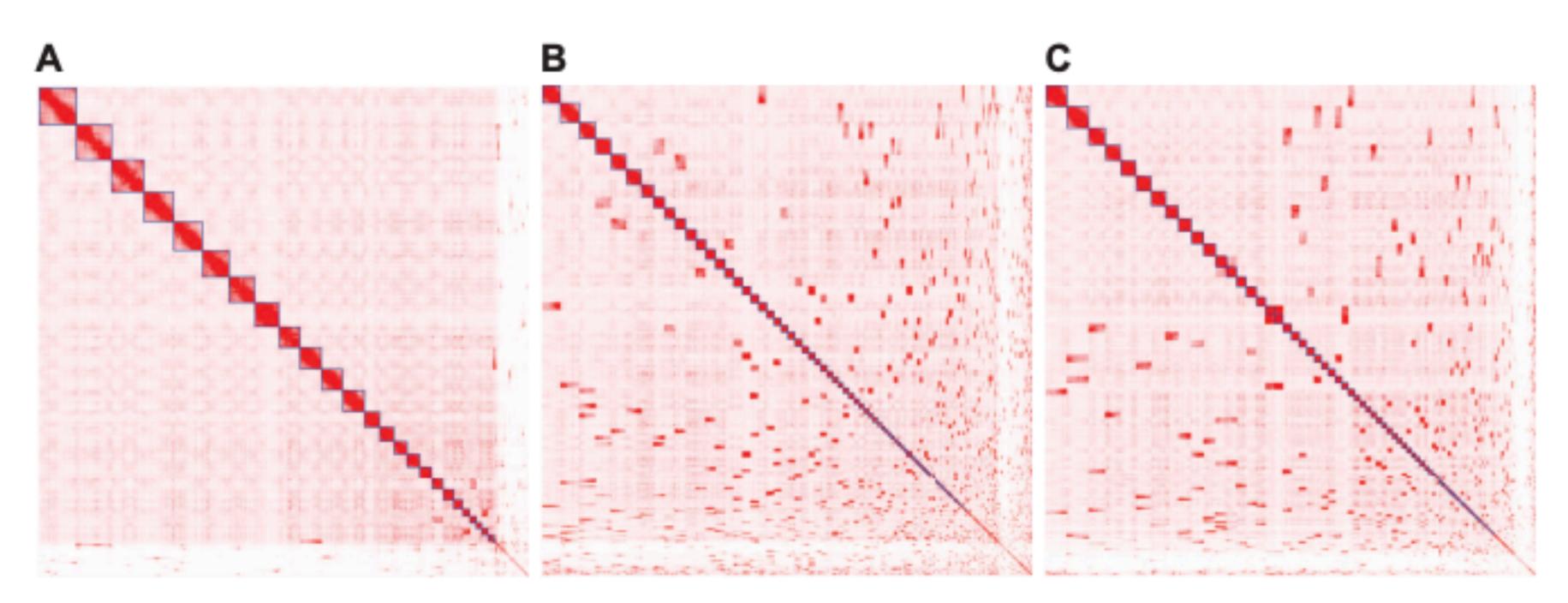
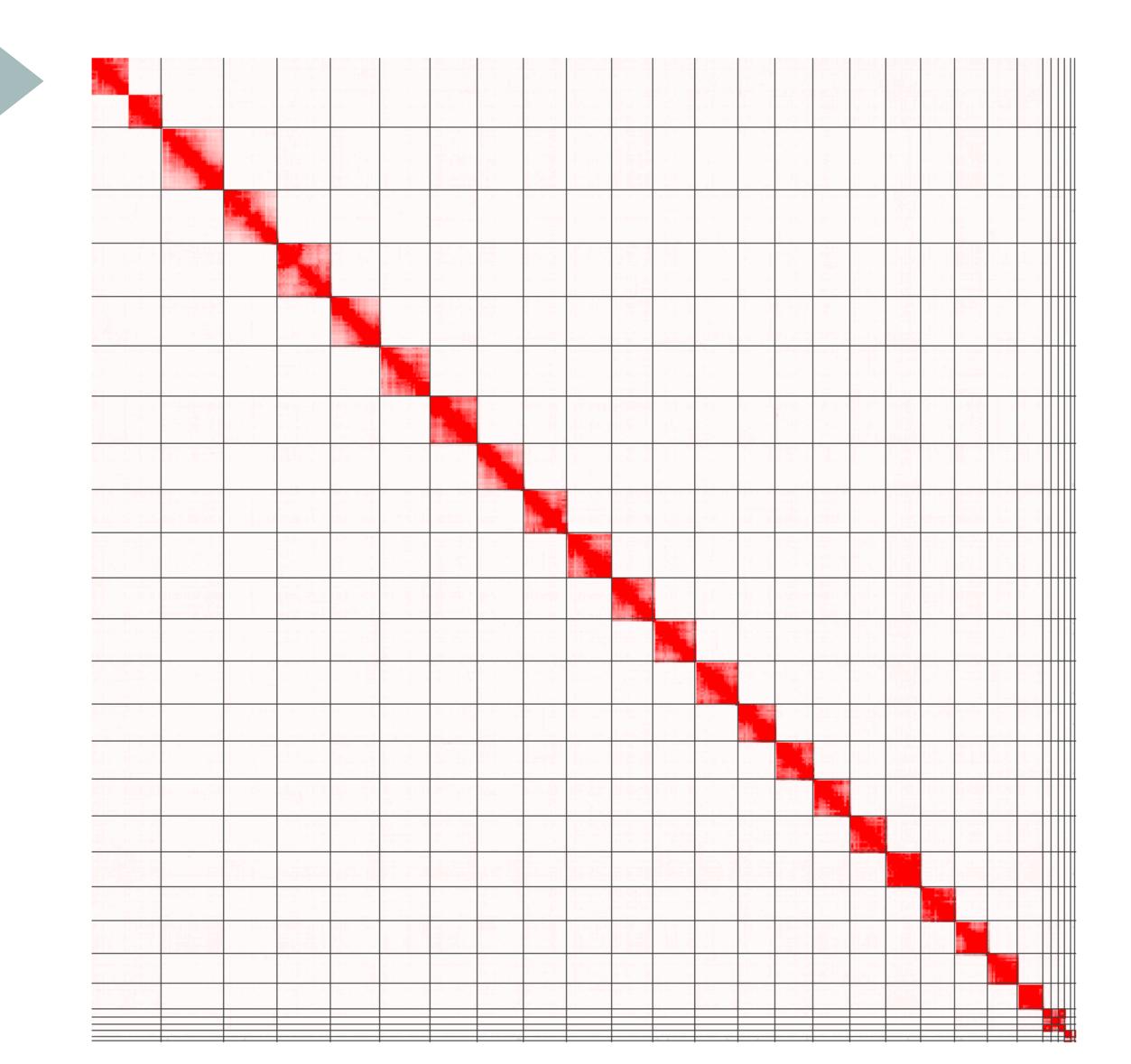


Figure 1. Hi-C contact maps of genome assemblies constructed with YaHS (A), SALSA2 (B) and pin_hic (C) for the simulated T2T data without contig errors. The blocks highlighted with blue squares in diagonal line are scaffolds. The contact maps were plotted with Juicebox (Durand et al., 2016).

HI-C: DETECTING MISASSEMBLES

Look at me!!!!





Lycaena phlaeas - ilLycPhla1

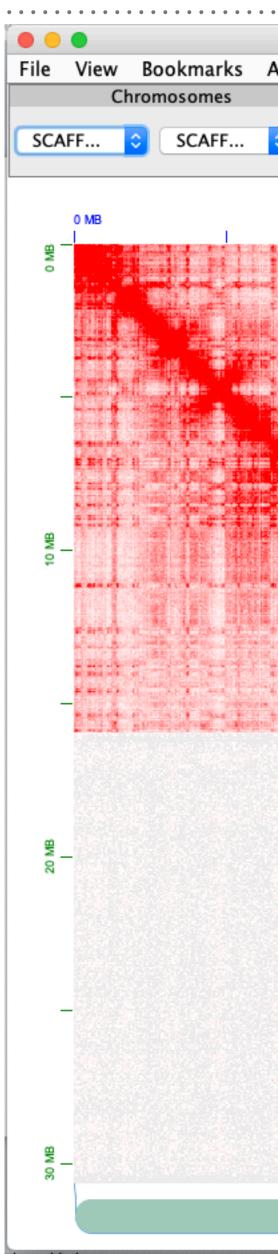


HI-C: DETECTING MISASSEMBLES



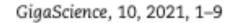
Lycaena phlaeas - ilLycPhla1





[Juicebox 1.9.8] Hi-C Map <8				
ssembly Dev Show	Normalization	Resolutio		
		Kesolutio		
Observed 📀	None	2.5 MB 500 KB 100 KB		
10 MB	20 MB	30 MB		





doi: 10.1093/gigascience/giaa153 Review

REVIEW

Significantly improving the quality of genome assemblies through curation

Kerstin Howe[®], William Chow[®], Joanna Collins[®], Sarah Pelan[®], Damon-Lee Pointon[®], Ying Sims[®], James Torrance[®], Alan Tracey[®] and Jonathan Wood[®]

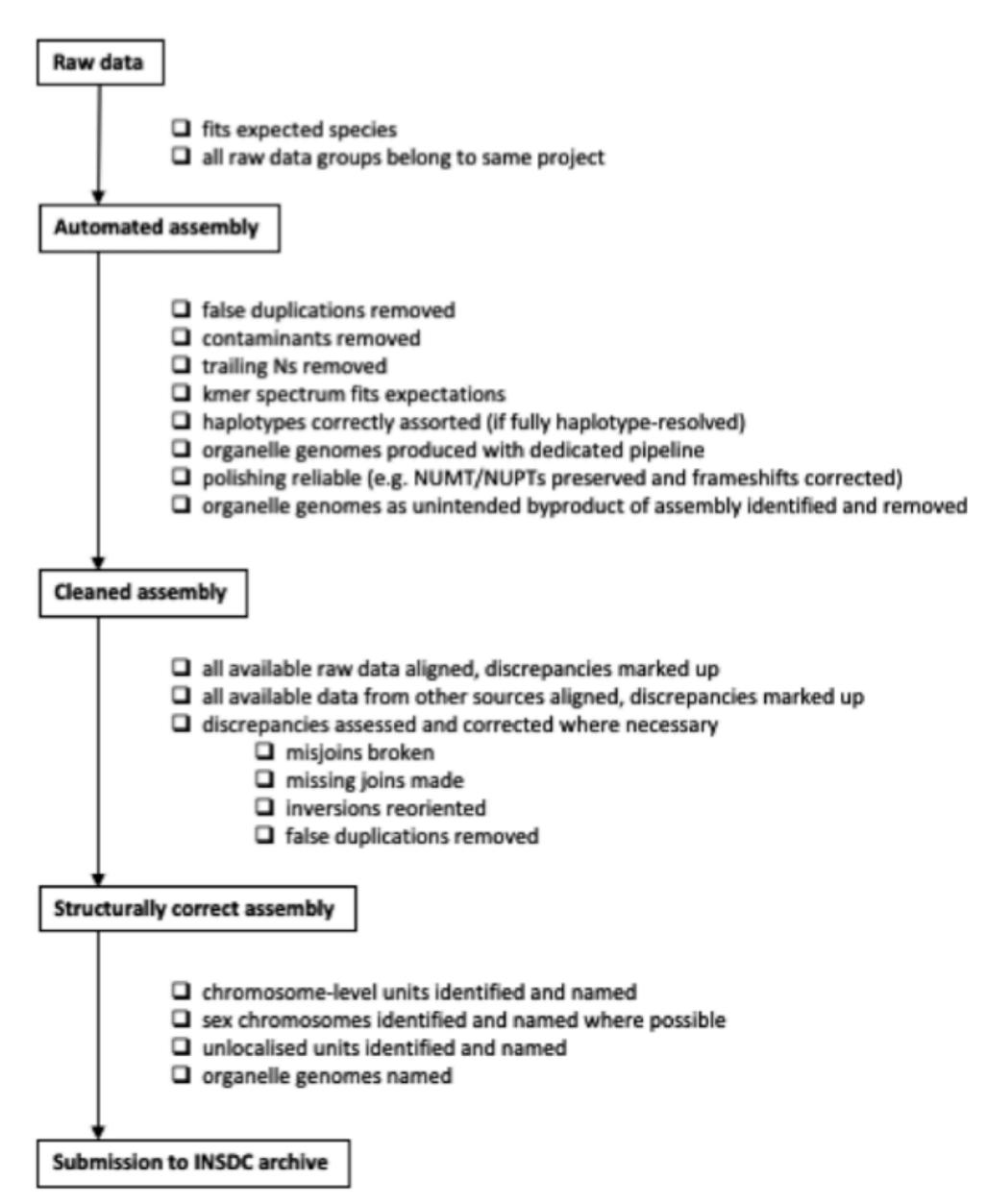
Tree of Life, Wellcome Sanger Institute, Cambridge CB10 1SA, UK

*Correspondence address. Kerstin Howe, Tree of Life, Wellcome Sanger Institute, Cambridge CB10 1SA, UK. E-mail: kerstin@sanger.ac.uk ⁽³⁾ http://orcid.org/0000-0003-2237-513X

Abstract

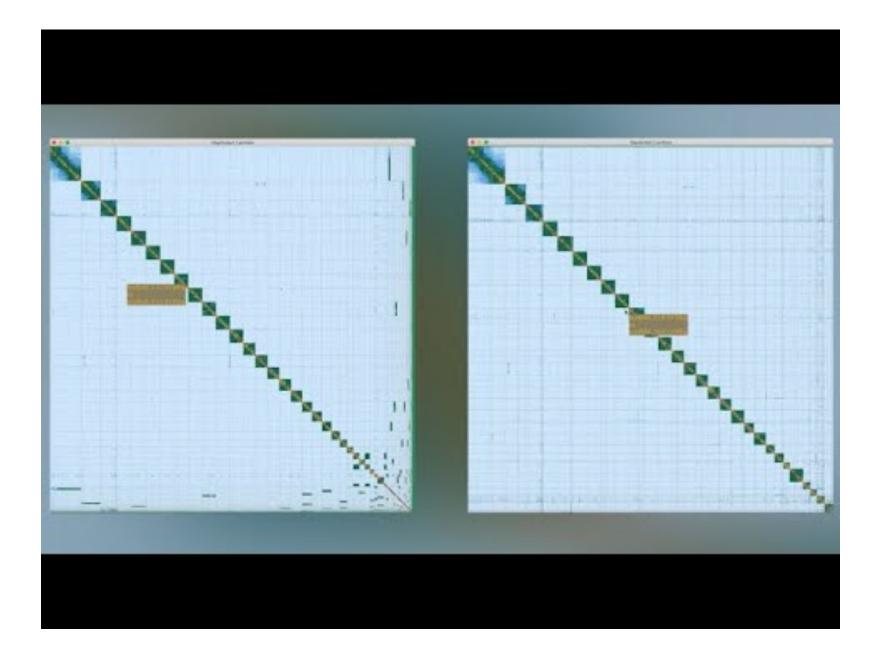
Genome sequence assemblies provide the basis for our understanding of biology. Generating error-free assemblies is therefore the ultimate, but sadly still unachieved goal of a multitude of research projects. Despite the ever-advancing improvements in data generation, assembly algorithms and pipelines, no automated approach has so far reliably generated near error-free genome assemblies for eukaryotes. Whilst working towards improved datasets and fully automated pipelines, assembly evaluation and curation is actively used to bridge this shortcoming and significantly reduce the number of assembly errors. In addition to this increase in product value, the insights gained from assembly curation are fed back into the automated assembly strategy and contribute to notable improvements in genome assembly quality. We describe our tried and tested approach for assembly curation using gEVAL, the genome evaluation browser. We outline the procedures applied to genome curation using gEVAL and also our recommendations for assembly curation in a gEVAL-independent context to facilitate the uptake of genome curation in the wider community.

Keywords: genome; assembly; curation; gEVAL



LOTS OF MATERIALS FROM THE CURATION TEAM

https://gitlab.com/wtsi-grit/rapid-curation/-/blob/main/Interpreting_HiC_Maps_guide.pdf

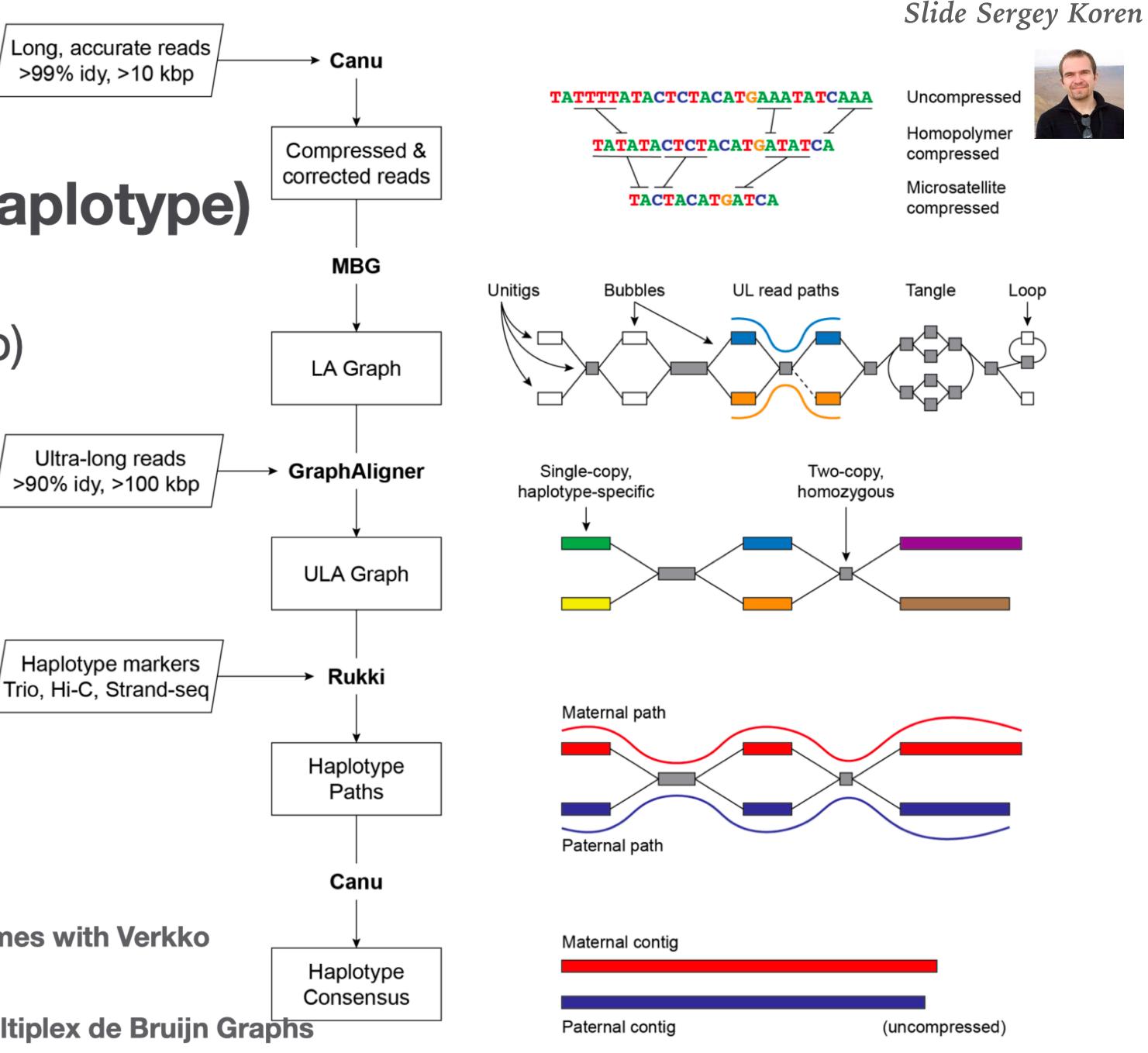


https://bga23.org/manual-curation/

https://docs.google.com/presentation/d/1g9Ubjjjpl4Vxvw-HSOlodJUDJBaicBbU/edit#slide=id.p13

THE ONLY TYPE OF HYBRID ASSEMBLY YOU SHOULD USE IS VERKKO





Sequencing recipe (per haplotype)

- 25 PacBio HiFi (20 kb)
- 25x ONT ultra-long (>100 kb)
- 30x Illumina Trio or Hi-C

Telomere-to-telomere assembly of diploid chromosomes with Verkko Rautiainen, et al. Nat Biotech (2023)



LJA: Assembling Long and Accurate Reads Using Multiplex de Bruijn Graphs Bankevich, et al. Nat Biotech (2021)



Slide Sergey Koren



Software Open Access Published: 18 July 2023

MitoHiFi: a python pipeline for mitochondrial genome assembly from PacBio high fidelity reads

<u>Marcela Uliano-Silva</u> ⊡, <u>João Gabriel R. N. Ferreira</u>, <u>Ksenia Krasheninnikova</u>, <u>Darwin Tree of Life</u> <u>Consortium</u>, <u>Giulio Formenti</u>, <u>Linelle Abueg</u>, <u>James Torrance</u>, <u>Eugene W. Myers</u>, <u>Richard Durbin</u>, <u>Mark</u> <u>Blaxter</u> & <u>Shane A. McCarthy</u>

BMC Bioinformatics 24, Article number: 288 (2023) Cite this article

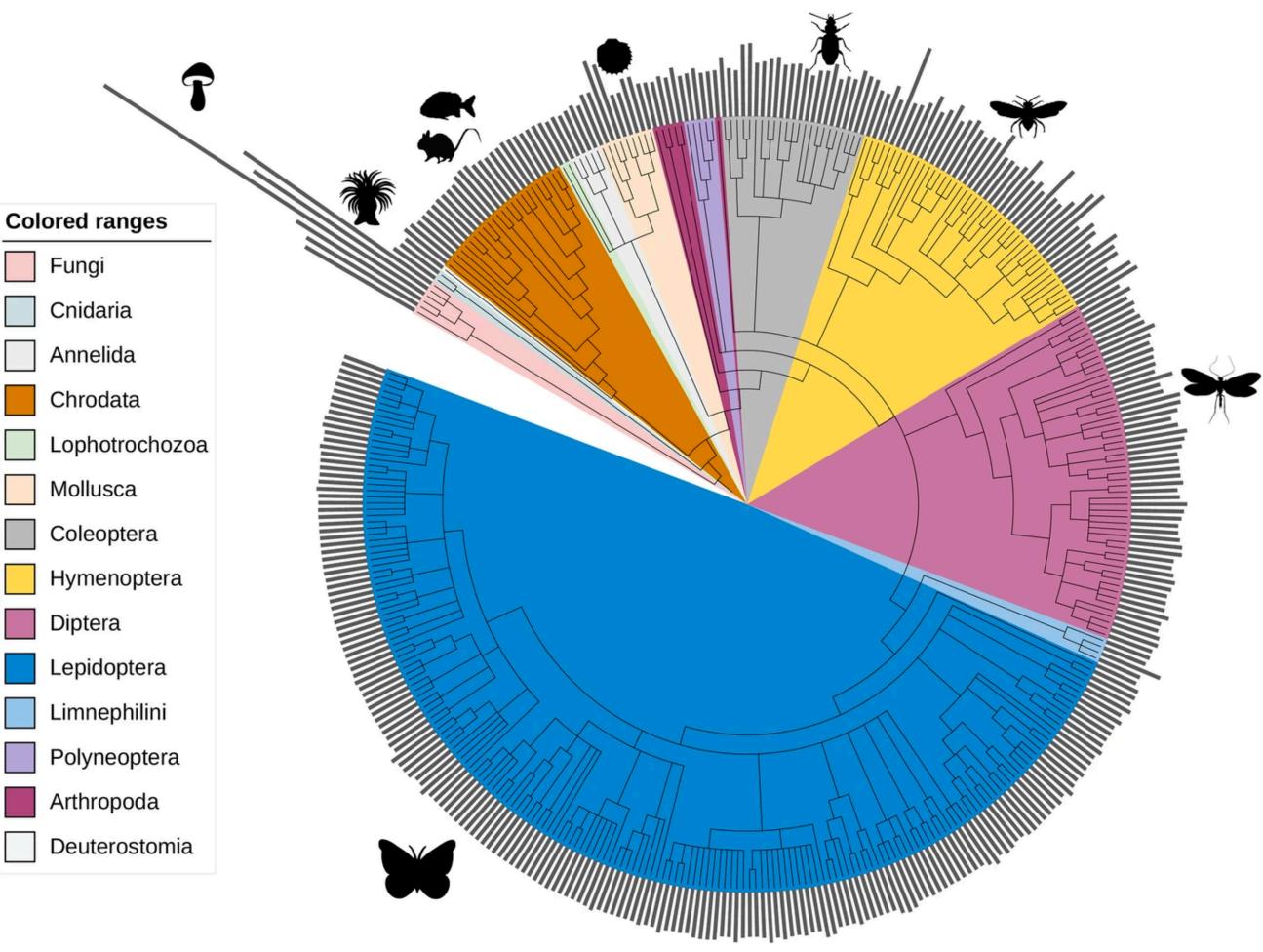
1576 Accesses | **19** Citations | **41** Altmetric | <u>Metrics</u>

Use the docker container!!!

Pleeease....

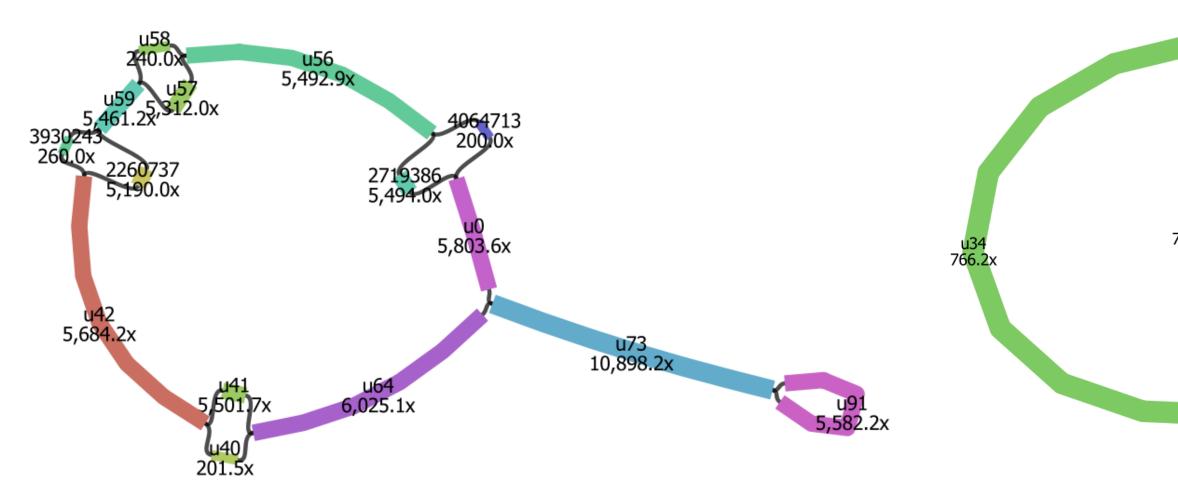




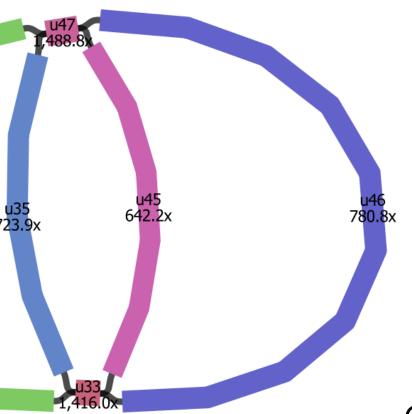


PLANTS CHLOROPLASTS AND MITOGENOMES

Oatk: an organelle assembly toolkit



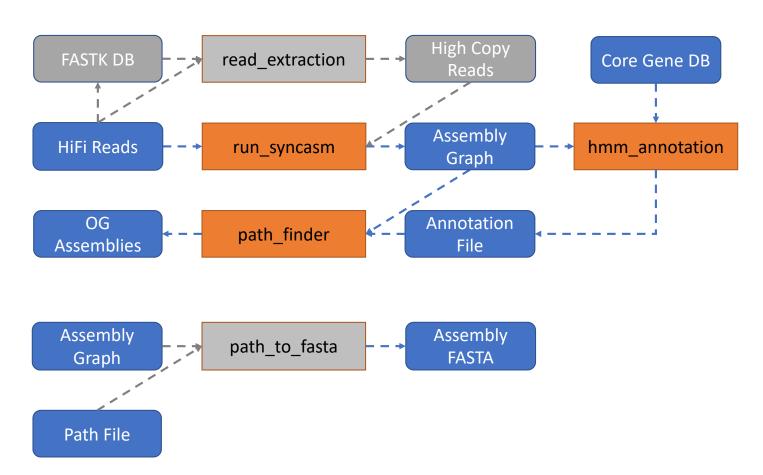
- Classification of graph components
 - Mitochondria _
 - Plastid _
 - others
- For each organelle graph component, ulletlist all possible paths and do selection
- For circular chloroplast assemblies, do • rotation with reference to the conserved gene order





Chenxi Zhou

Oatk workflow







Mutual aid: How you can join, and how you can benefit.

• We would like to sequence your species of interest to support your future plans in genomic Tell us through the form at <u>tinyurl.com/dtol-suggest</u>

https://darwintreeoflife.org



understanding of biodiversity - be it population genetics, conservation, ecosystems, evolution, ...

To conclude

- There is no one-size-fits all protocol for DNA extraction
- Long reads will be ideal for chromosome level genome assembly
- You MUST investigate the genome size, heterozygosity and repeat content at the beginning of your genome assembly project
- Hi-C is useful not only for scaffolding, but for genome curation as well
- All Tree of Life is data Open for the use of all!



https://wellcomeopenresearch.org/gateways/treeoflife

Wellcome Open Research / Gateways



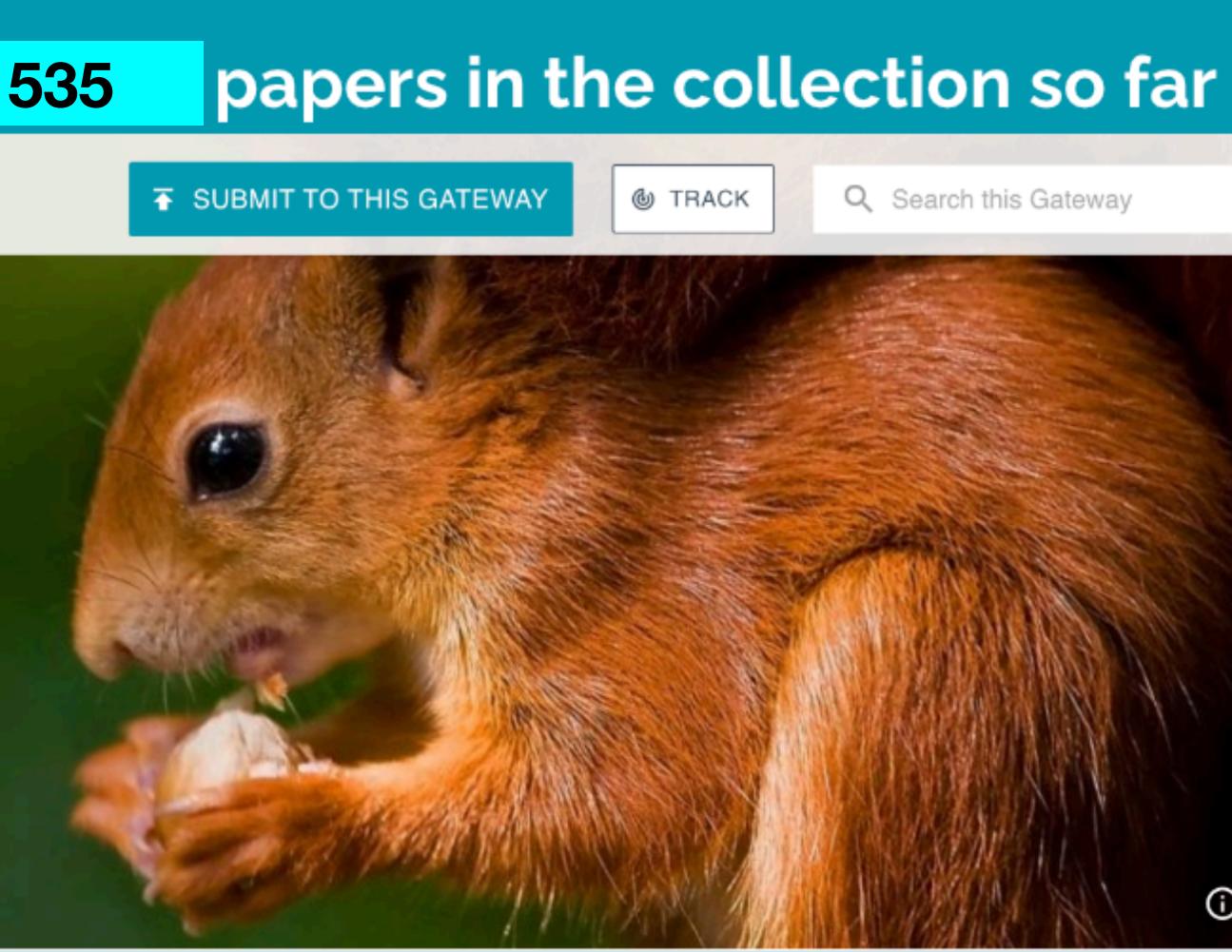
The Tree of Life Programme

This gateway collates genome sequences released by the Wellcome Sanger Institute as part of the Darwin Tree of Life project (sequencing the genomes of all known species of animals, plants, fungi and protists in Britain and Ireland) and other initiatives.



Read more in the blog →

https://wellcomeopenresearch.org/treeoflife



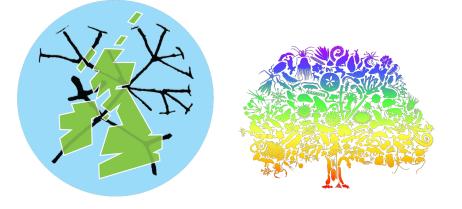


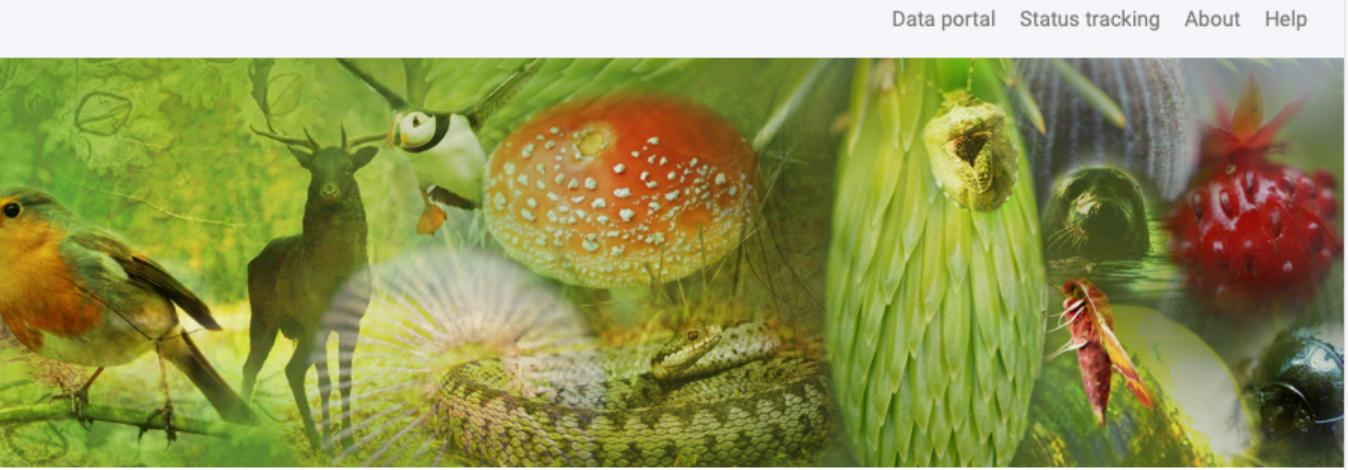


portal.darwintreeoflife.org

🍘 🎬 Darwin Tree of Life Data Portal

Sequencing all 60,000 eukaryotic species of Britain and Ireland







THINGS I HAVEN'T MENTIONED THAT YOU MUST LEARN ABOUT

Research Open access Published: 27 September 2022

Widespread false gene gains caused by duplication errors in genome assemblies

Byung June Ko, Chul Lee, Juwan Kim, Arang Rhie, Dong Ahn Yoo, Kerstin Howe, Jonathan Wood, Seoae Cho, Samara Brown, Giulio Formenti, Erich D. Jarvis ⊠ & Heebal Kim ⊠

Genome Biology 23, Article number: 205 (2022) Cite this article

4164 Accesses 8 Citations 14 Altmetric Metrics

"Whole genome alignments revealed that 4 to 16% of the sequences are falsely duplicated in the previous assemblies, impacting hundreds to thousands of genes. These lead to overestimated gene family expansions.

The main source of the false duplications is heterotype duplications, where the haplotype sequences were relatively more divergent than other parts of the genome leading the assembly algorithms to classify them as separate genes or genomic regions." Kim et al, 2022

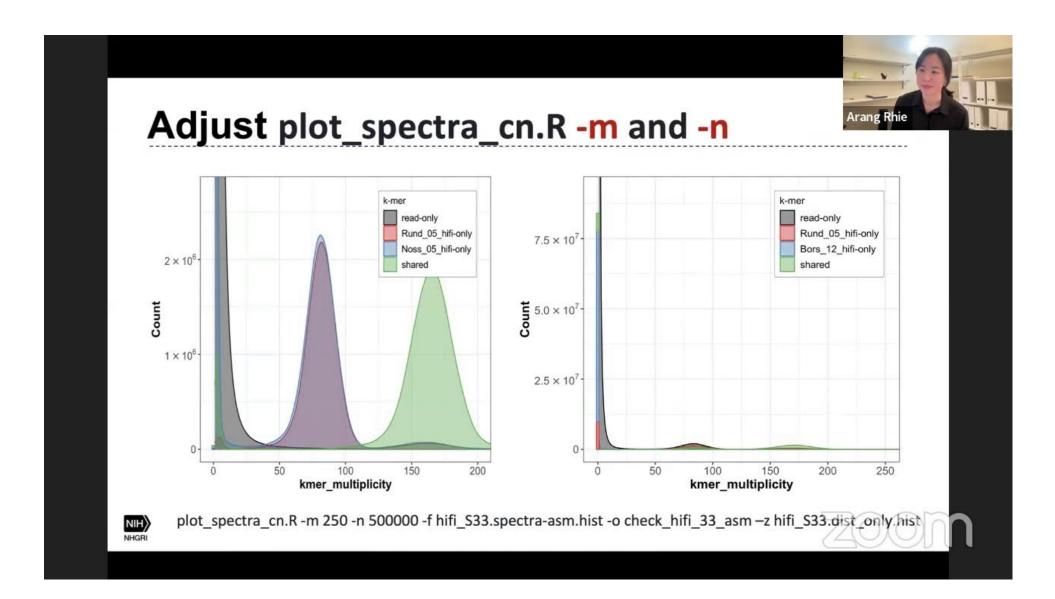
HOW TO IDENTIFY RETAINED HAPLOTIGS? PURGING AND MERQURY!!!!!

Bioinformatics, 36(9), 2020, 2896–2898 doi: 10.1093/bioinformatics/btaa025 Advance Access Publication Date: 23 January 2020 Applications Note

Genome analysis Identifying and removing haplotypic duplication in primary genome assemblies

Dengfeng Guan^{1,2}, Shane A. McCarthy (**b**², Jonathan Wood³, Kerstin Howe (**b**³, Yadong Wang^{1,*} and Richard Durbin ()^{2,3,*}

¹Department of Computer Science and Technology, Center for Bioinformatics, Harbin Institute of Technology, Harbin 150001, China, ²Department of Genetics, University of Cambridge, Cambridge CB2 3EH, UK and ³Wellcome Sanger Institute, Wellcome Genome Campus, Cambridge CB10 1SA, UK







Rhie et al. Genome Biology (2020) 21:245 https://doi.org/10.1186/s13059-020-02134-9

Genome Biology

METHOD

Mergury: reference-free quality, completeness, and phasing assessment for genome assemblies



Open Access

Arang Rhie[®], Brian P. Walenz, Sergey Koren and Adam M. Phillippy

* Correspondence: arang.rhie@nih.

Genome Informatics Section, Computational and Statistical Genomics Branch, National Human Genome Research Institute, National Institutes of Health, Bethesda, MD, USA

Abstract

Recent long-read assemblies often exceed the quality and completeness of available reference genomes, making validation challenging. Here we present Mergury, a novel tool for reference-free assembly evaluation based on efficient k-mer set operations. By comparing k-mers in a de novo assembly to those found in unassembled high-accuracy reads, Mergury estimates base-level accuracy and completeness. For trios, Mergury can also evaluate haplotype-specific accuracy, completeness, phase block continuity, and switch errors. Multiple visualizations, such as k-mer spectrum plots, can be generated for evaluation. We demonstrate on both human and plant genomes that Merqury is a fast and robust method for assembly validation.

Keywords: Genome assembly, Assembly validation, Benchmarking, K-mers, Haplotype phasing, Trio binning



Pesquisar





Biodiversity Genomics Academy 2023

Biodiversity Genomics Academy

26 vídeos 711 visualizações Última atualização em 1 d...



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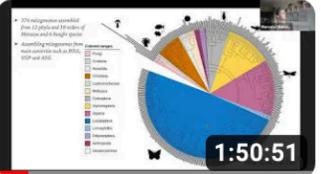






Fri 8 Sep, 11:00 - The Treeval pipeline: Generating evidence for manual curation

Biodiversity Genomics Academy • 273 visualizações • Transmitido há 4 meses



1.	BlobDir viewed in Viewer	Viewer
2	blobtools creates a BlobDir from	bioticols
	intermediate files	Naces Res
3.	pipeline runs intermediate steps to create intermediate files and final BlobDir	1:12:00

But, other species may be present

1:07:49





Fri 8 Sep, 13:00 - Assembling Mitogenomes from PacBio HiFi reads using MitoHiFi

Biodiversity Genomics Academy • 257 visualizações • Transmitido há 4 meses

Mon 11 Sep, 08:00 - Visualising genome assembly cobionts by running BlobToolKit locally

Biodiversity Genomics Academy • 150 visualizações • Transmitido há 4 meses

Mon 11 Sep, 17:00 - Checking for cobionts in public genomes using BlobToolKit

Biodiversity Genomics Academy • 90 visualizações • Transmitido há 4 meses

Tue 12 Sep, 14:00 - Understanding k-mers and ploidy using Smudgeplot

Biodiversity Genomics Academy • 274 visualizações • Transmitido há 4 meses

Wed 13 Sep, 09:00 - Starting a comparative genome study from CNGBdb

Biodiversity Genomics Academy • 86 visualizações • Transmitido há 4 meses

2

1

3

5

6



Obrigada! Thank you!

Bill Baker Ester Gaya Paul Kersey Ilia Leitch Greg Palmer



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Ying Sims

Scientific Operations

Carol Smee Catherine McCarthy Elizabeth Cook Emma Betteridge Iraad Bronner Michelle Smith Mike Quail Naomi Park Alex Dove Barbora Pardubska Carlos Jimenez Verdejo Craig Corton Emily Gallagher Emma Taluy Esther Mellado Harriet Johnson Hermione Blomfield-Smith Irene Fabiola James Uphill John Tushabe Karen Oliver Michelle Smith Robin Moll Tracey Chillongworth

Collaborators

Jonas Korlach et al. Pacific Biosciences Dan Turner et al. Oxford Nanopore

Team301

Chris Laumer Claudia Weber Emmelein Vancaester Erna King Lewis Stevens Max Brown Pablo Gonzalez **Rich Challis**



Obrigada! Thank you!



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Scientific Operations

Team301





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Probability of incorrect base call	Base call accura
1 in 1000	99.9%
1 in 10,000	99.99%
1 in 100,000	99.999%
	base call 1 in 1000 1 in 10,000

