

Exploring pangenome graphs and possible applications

Institute for Medical Biometry and Bioinformatics
21 October 2021

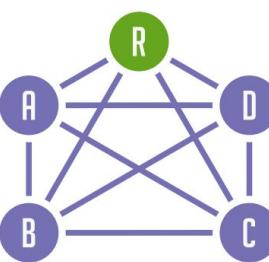
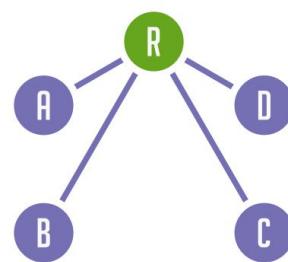
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Jerven Bolleman, Toshiyuki T. Yokoyama, Pjotr Prins,
Sven Nahnsen, Erik Garrison

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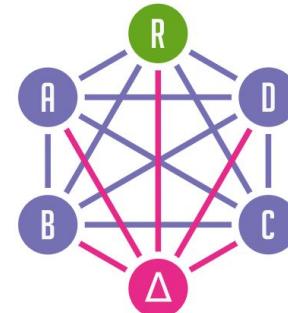
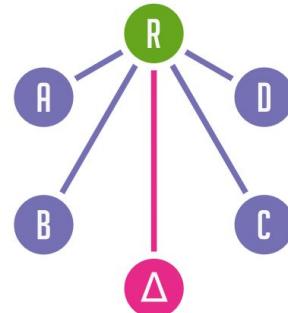


De novo assembly and a pangenomic model

Reference model



Extending the model



Δ: new genome; R: reference genome.

Figure from [Eizenga et al., 2020](#).

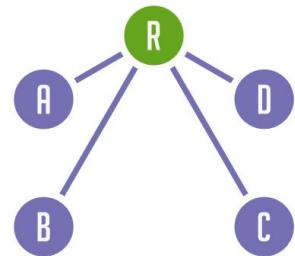
Thanks to advances in sequencing technology, new **telomere-to-telomere** genome assemblies are produced at a high rate.

Pangenomes can **model** the full set of genomic elements in a given species or clade, reducing the **reference-bias**.

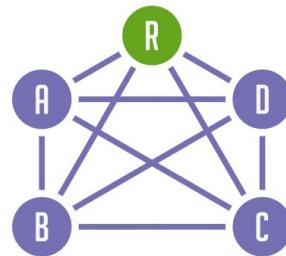
A pangenome encoded as a graph

Reference model

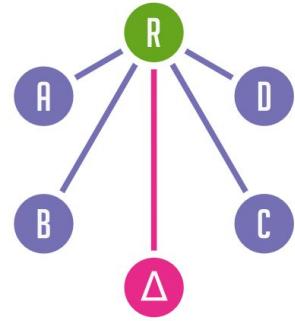
Genomic



Pangenomic

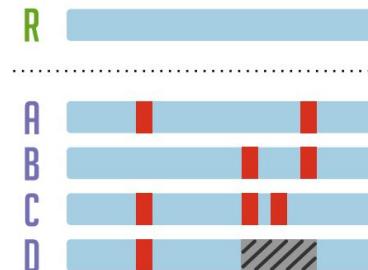


Extending the model



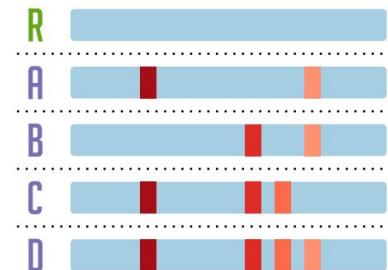
Δ: new genome; R: reference genome.

Figure from [Eizenga et al., 2020](#).



Linear representation
(augmented with variations)

- Reference allele
- Alternative allele
- ▨ Unmapped segment



Graphical (compressed) representation

- Shared segment
- Variant

Figure from [Eizenga et al., 2020](#).

Pangenome graphs - representation

Pangenomes can take many forms, including **graph-based** data structures.

Pangenome graphs compress redundant sequences into a smaller data structure that is still representative of the full set.

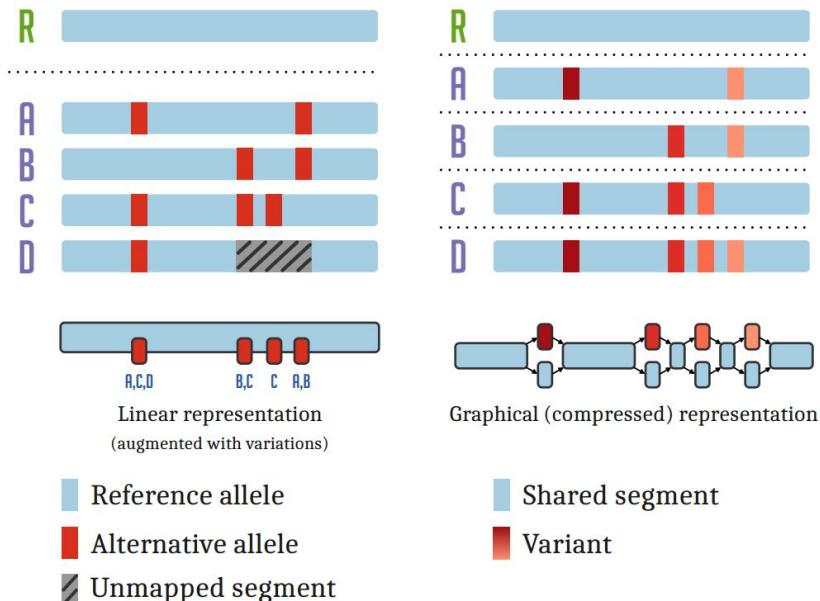
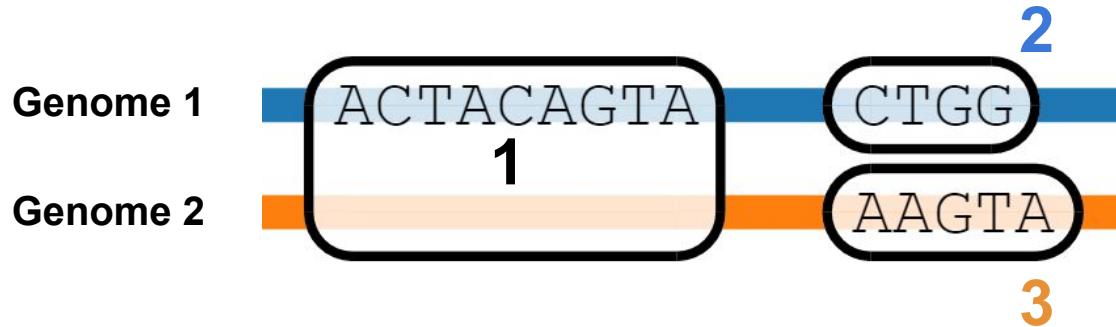


Figure from [Eizenga et al., 2020](#).

Variation graphs

- Genome 1: **ACTACAGTACTGG** Path: 1 2
- Genome 2: **ACTACAGTAAGTA** Path: 1 3

Linear sequences are **paths** through nodes.



Graph topology is not directly shown.

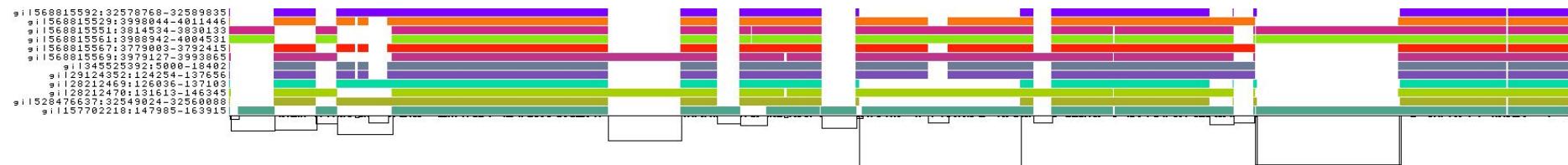
The nodes represent DNA sequences.

Paths can be contigs, haplotypes, reads, or whole chromosomes.

Sketch made using [SequenceTubeMap](#).

1D Graph visualization explained

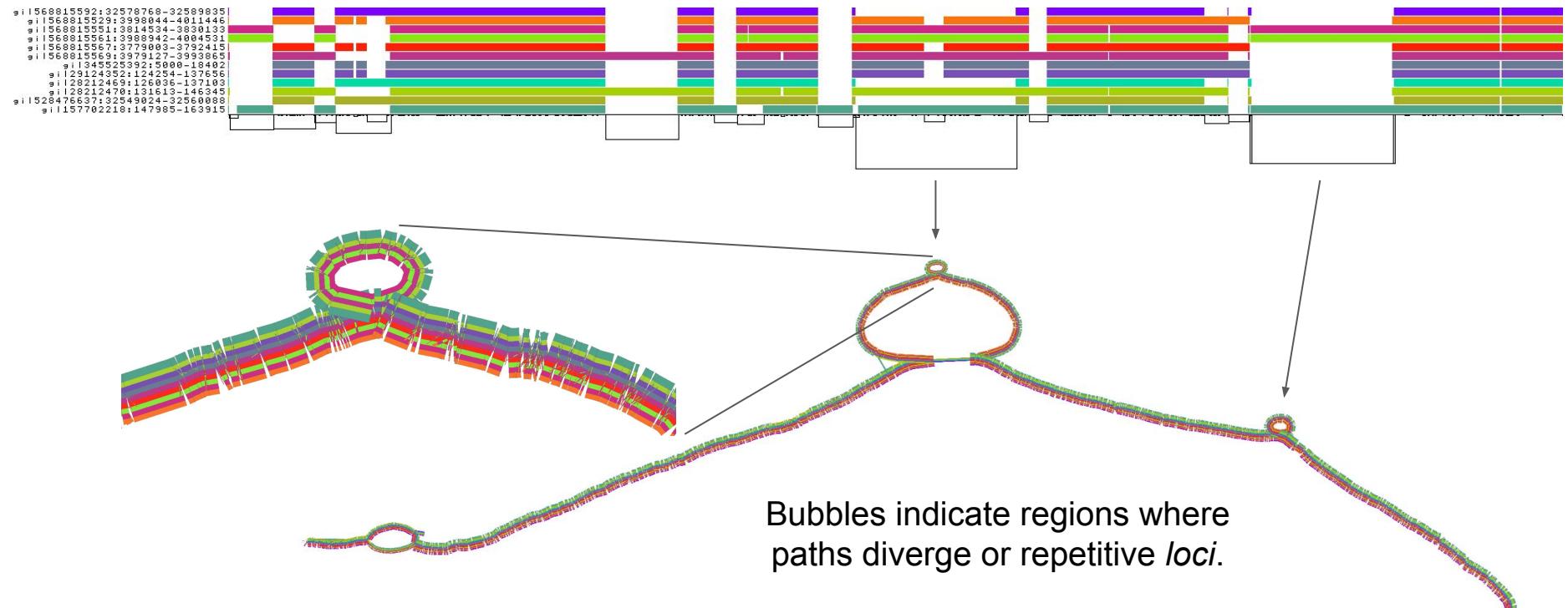
Pangenome graph with 12 ALT sequences of the HLA-DRB1 gene from the GRCh38 reference genome.



- The graph nodes are arranged from left to right, forming the pangenome sequence.
- The colored bars represent the paths versus the pangenome sequences in a binary matrix.
- The path names are placed on the left.
- The black lines under the paths are the links, which represent the graph topology.

2D Graph visualization explained

Pangenome graph with 12 ALT sequences of the HLA-DRB1 gene from the GRCh38 reference genome.



Figures obtained with kind permission from Andrea Guarracino

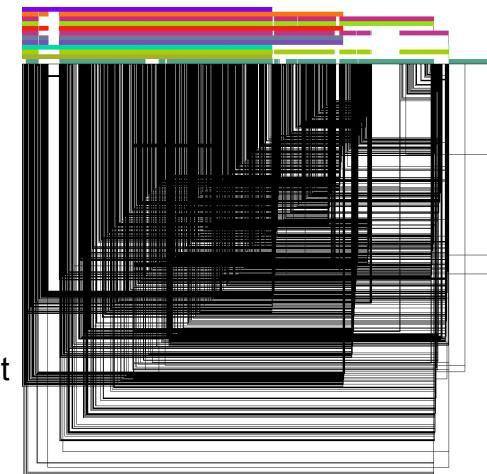
1D Graph Sorting and 2D Layouting by Path-Guided Stochastic Gradient Descent (PG-SGD)

A pangenome graph induced from raw alignments can be very complex and hard to analyse downstream.

Solution Make the graph more linear by reordering of nodes.

- Visualization
- Comparative genomics
- Mapping
- Interpretation

Raw graph built using [seqwish](#).



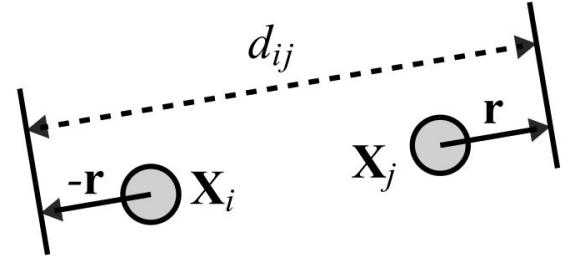
Erik tried more than 10 sorting algorithms, but none of them did the job. The most promising one was [Graph drawing by Stochastic Gradient Descent](#)

1D Graph Sorting by P-SGD - The Algorithm Explained

Objective: Move a single pair of nodes at a time.

Optimizing the disparity between the layout distance of a node pair and the actual nucleotide distance of a path traversing these nodes.

- The first node X_i of a pair is a uniform path step pick from all nodes.

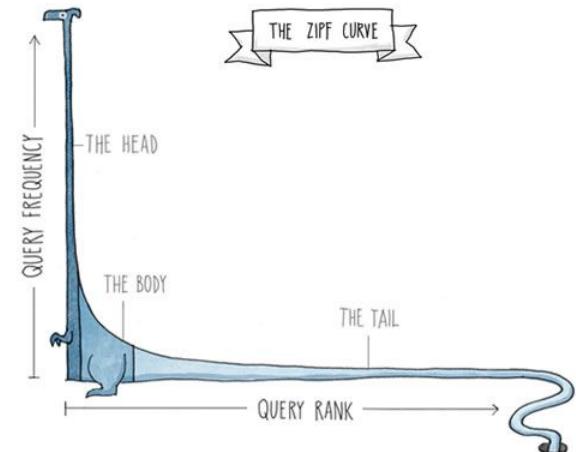
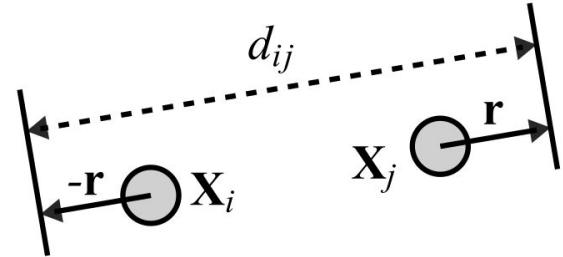


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- The first node X_i of a pair is a uniform path step pick from all nodes.
- The second node X_j of a pair is sampled from the same path following a Zipfian distribution.

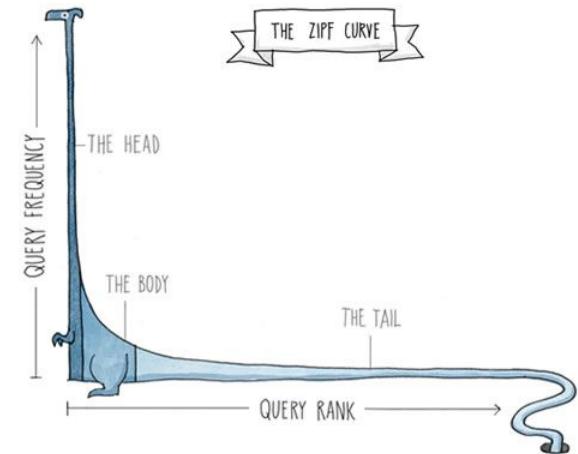
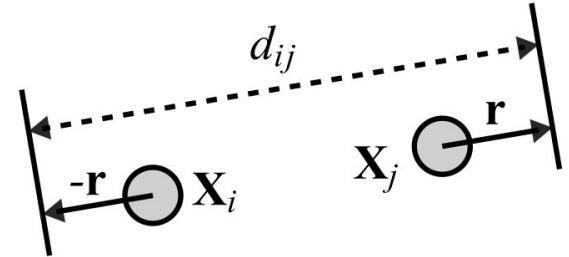


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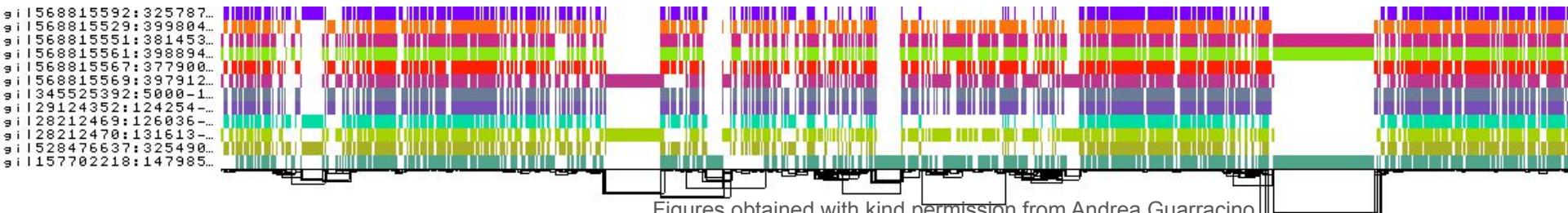
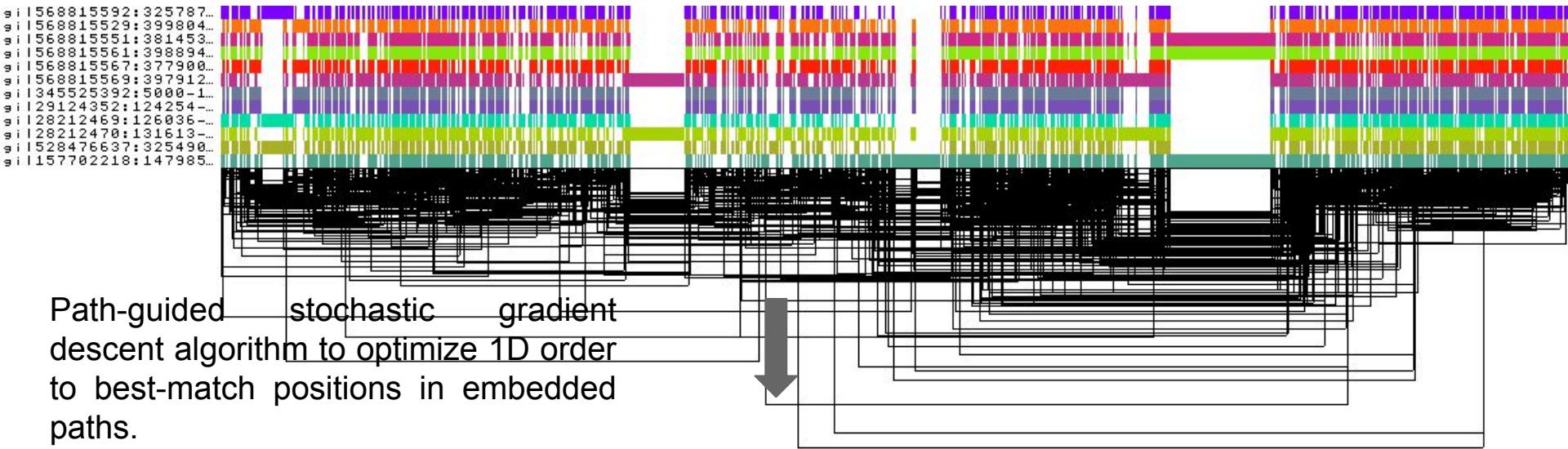
Optimizing the disparity between the layout distance of a node pair and the actual nucleotide distance of a path traversing these nodes.

- The first node X_i of a pair is a uniform path step pick from all nodes.
- The second node X_j of a pair is sampled from the same path following a Zipfian distribution.
- The path nucleotide distance of the nodes in the pair guides the actual layout distance d_{ij} update of these nodes. The magnitude r of the update depends on the current learning rate of the SGD.



1D Graph sorting by PG-SGD - Hogwild!

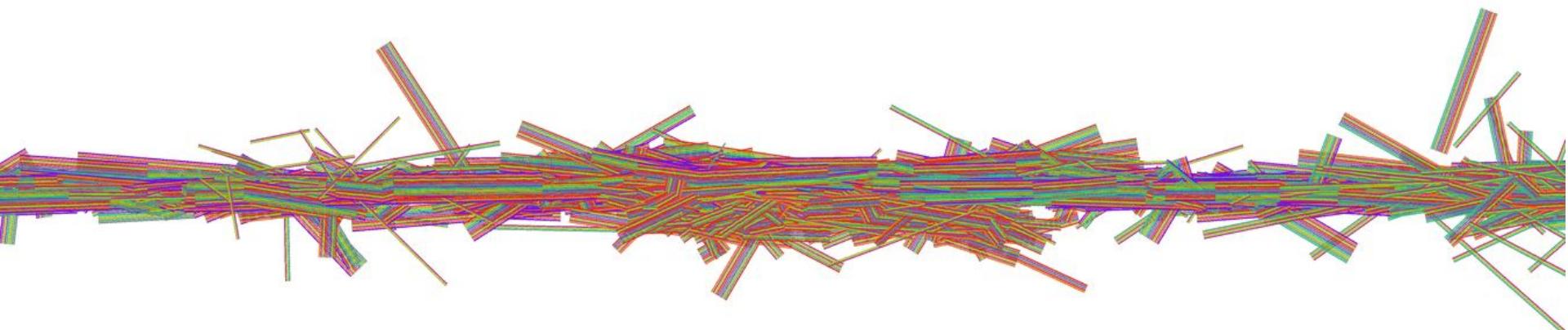
Pangenome graph with 12 ALT sequences of the HLA-DRB1 gene from the GRCh38 reference genome.



Figures obtained with kind permission from Andrea Guarracino

Bonus: 2D Graph layout by PG-SGD - Also Hogwild!

Pangenome graph with 12 ALT sequences of the HLA-DRB1 gene from the GRCh38 reference genome.



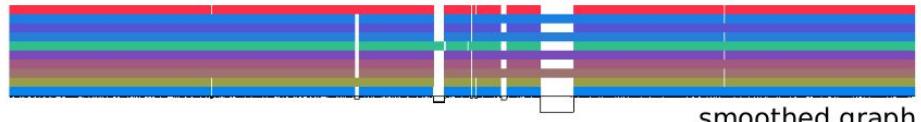
Path-guided stochastic gradient descent algorithm to optimize 2D layout. Path-labeled rendering with odgi draw.

The layout can be plugged into [gfaestus](#) for interactive visualization.

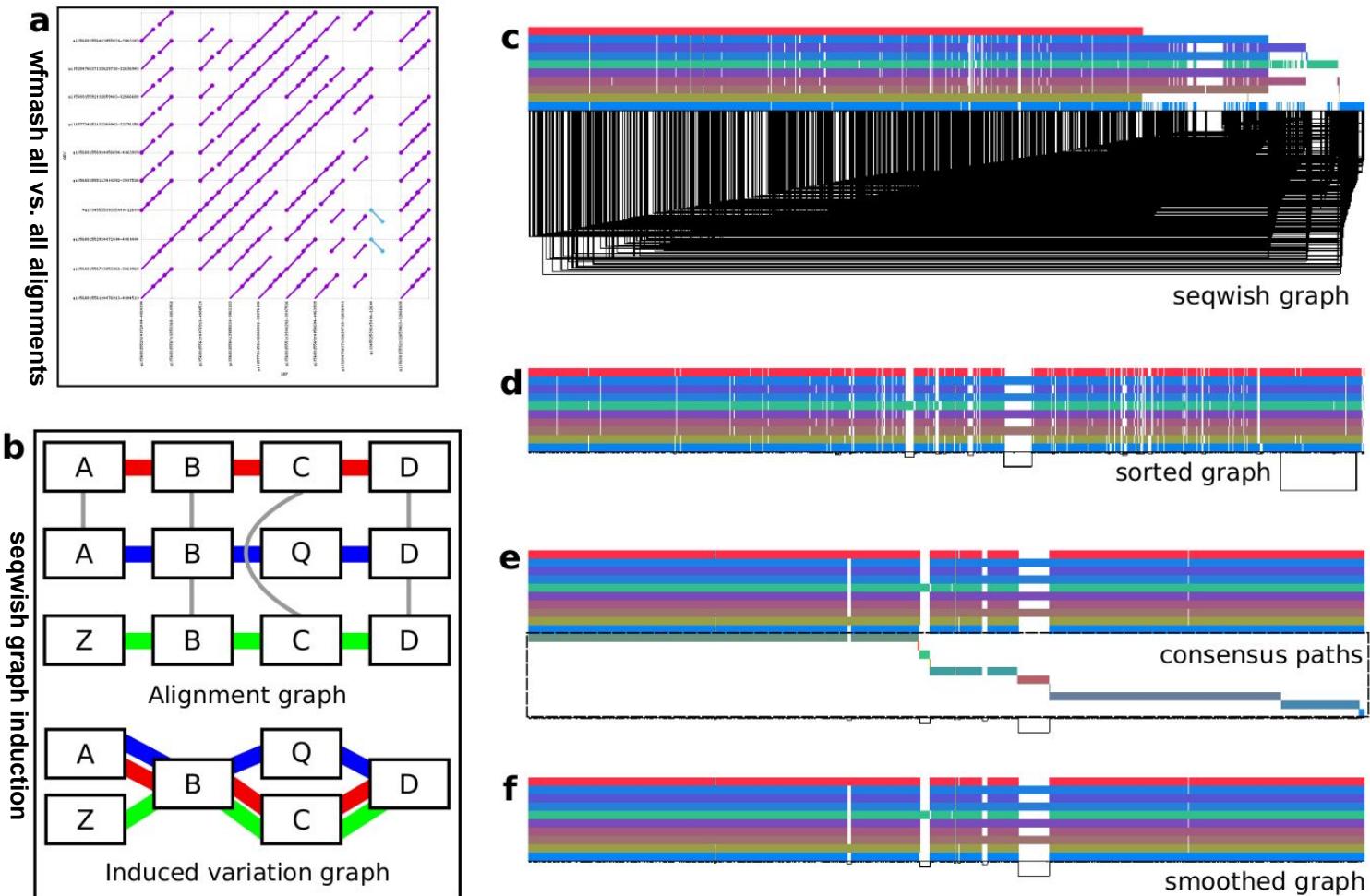
1D PG-SGD implementation is **the key step** in pangenome graph simplification pipeline [smoothxq](#)

- [smoothxq](#) runs Partial Order Alignment ([POA](#)) for each block of paths that are collinear within a [seqwish](#) induced variation graph.
- A prerequisite is that the graph nodes are sorted according to their occurrence in the graph's embedded paths
- Our 1D path-guided SGD algorithm is designed to provide this kind of sort.

PanGenome Graph Builder ([PGGB](#))



PanGenome Graph Builder (PGGB)



Figures obtained with kind permission from Erik Garrison

[nf-core/pangenome](#)

```
[heumos@savasana pangenome]$ pggb -i data/HLA/DRB1-3123.fa.gz -N -w 50000 -s 500  
0 -I 0 -p 80 -n 10 -k 8 -t 16 -v -L -o out
```

```
[heumos@savasana pangenome]$ nextflow run main.nf -profile docker -with-docker heumos/pangenome:latest --input ~/git/pggb/data/HLA/DRB1-3123.fa.gz --alignment_no_splits --smoothxg_max_block_weight 50000 --alignment_segment_length 5000 --smoothxg_block_id_min 0 --alignment_map_pct_id 80 --alignment_n_secondary 10 --seqwish_min_match_length 8 --do_viz --outdir potato_out/ --wfmask --max-cpus 16 --max-memory 24G --file_name_prefix "pggb" -c savasana.config
```

Parameters

- Input/output options
- Alignment options
- Seqwish options
- Smoothxg options**
 - smoothxg_max_block_weight
 - smoothxg_max_path_jump
 - smoothxg_max_edge_jump
 - smoothxg_max_poa_length
 - smoothxg_consensus_spec
 - smoothxg_block_id_min**
 - smoothxg_ratio_contain
 - smoothxg_poa_params
- Visualization options

command

```
nextflow run nf-core/pangenome -r
```

clones in last 11 months

stars

last release

open issues

collaborators

watchers

last updated

pull requests



```
[heumos@savasana pangenome]$ ls potato_out/  
alignment    multiqc_report.html  odgi_stats  pipeline_info  smoothxg  
multiqc_data  odgi_build       odgi_viz     seqwish
```



nf-core/pangenome - MultiQC



A modular tool to aggregate results from bioinformatics analyses across many samples into a single report.

Report generated on 2021-10-18, 11:56 based on data in: /home/heumos/pggb/DRB1

ODGI

ODGI is an optimized dynamic graph/genome implementation, for efficient analysis and manipulation of pangenome graphs structured in the variation graph model.

Detailed ODGI stats table.

Sample Name	Length	Nodes	Edges	Paths	Components	A	C	T	G	N	% GC
cons@10000_y_0_1000000	13 180	1	0	1	1	3 944	3 073	3 539	2 624	0	43.2%
cons@1000_y_0_1000000	13 180	1	0	1	1	3 944	3 073	3 539	2 624	0	43.2%
cons@100_y_0_1000000	14 125	14	18	4	1	3 944	3 073	3 540	2 624	944	40.3%
cons@10_y_0_1000000	14 260	44	60	10	1	3 980	3 115	3 566	2 655	944	40.5%
seqwish	37 415	1 192	1 596	12	1	11 031	8 124	9 769	7 547	944	41.9%
smooth	22 517	4 860	6 649	13	1	6 422	4 932	5 659	4 560	944	42.2%

nf-core/pangenome - Future scaling up



wfmash: Split all versus all run into one versus all wfmash runs so each becomes one execution job.

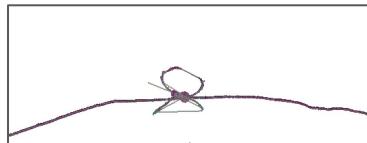
seqwish: Generate seqwish runs by steering the number of bp to use for transitive closure batch and pack each batch into one execution job.

smoothxg: Each abPOA alignment becomes one execution job.

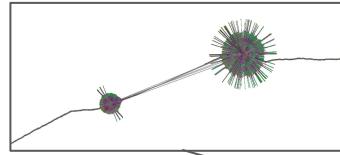
Improvement: Pangenome building of one chromosome can scale across a whole cluster and not just one node.

A PGGB human chr8 pangenome graph

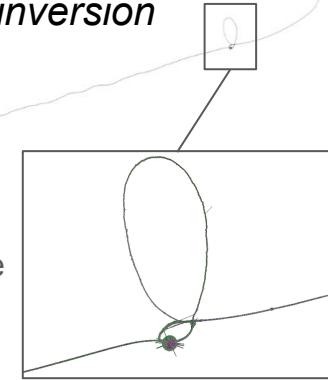
neocentromere



centromere

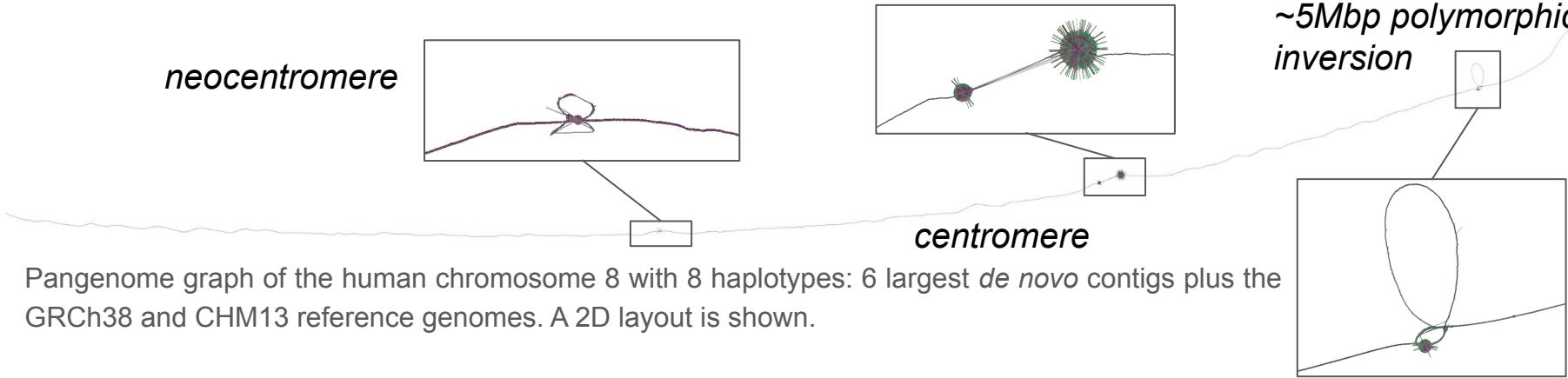


beta-defensin gene cluster and flanking ~5Mbp polymorphic inversion



Pangenome graph of the human chromosome 8 with 8 haplotypes: 6 largest *de novo* contigs plus the GRCh38 and CHM13 reference genomes. A 2D layout is shown.

A PGGB human chr8 pangenome graph



PGGB and nf-core/pangenome:

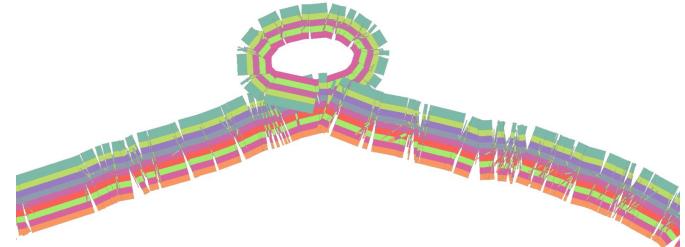
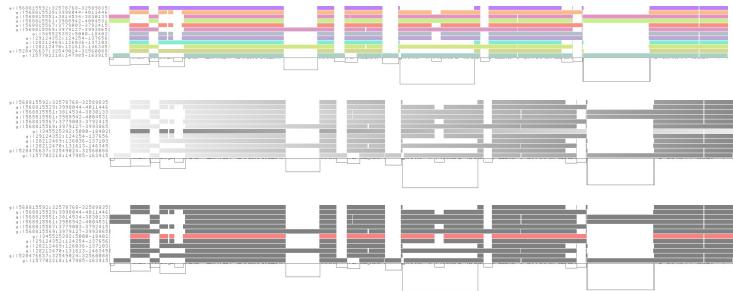
- Building a pangenome graph from all versus all alignments.
- No reference bias!
- All versus all relationship in one, compressed data structure.
- We can reconstruct the centromeres and **ultra deep complex regions!**

Our solution: a new suite of tools for pangenome graphs

To overcome these problems, we have developed an **Optimized Dynamic Genome/Graph Implementation** ([ODGI](#)), a new suite of tools to work with pangenome graphs structured in the variation graph model.

- ODGI supports GFA version 1 ([GFAv1](#))
- The majority of ODGI's tools are index-free
- Path manipulation in parallel

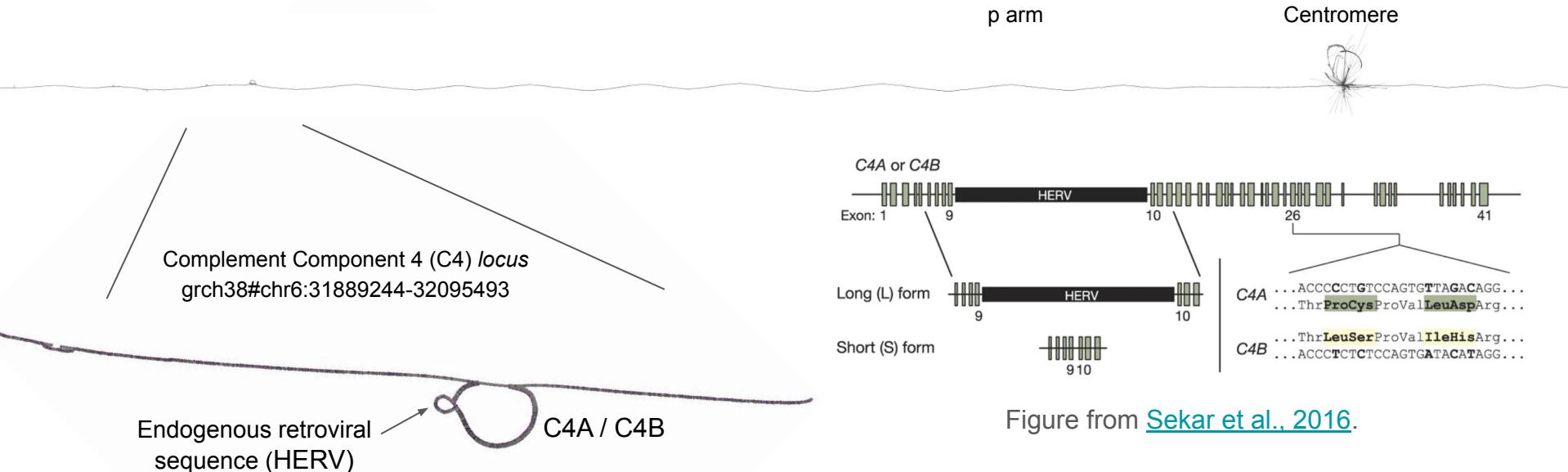
ODGI offers more than [30 tools](#) for graph interrogation, manipulation, and visualization.



Dissecting pangenome graphs - odgi extract

Downstream analyses may require focusing on specific *loci* in the pangenome.

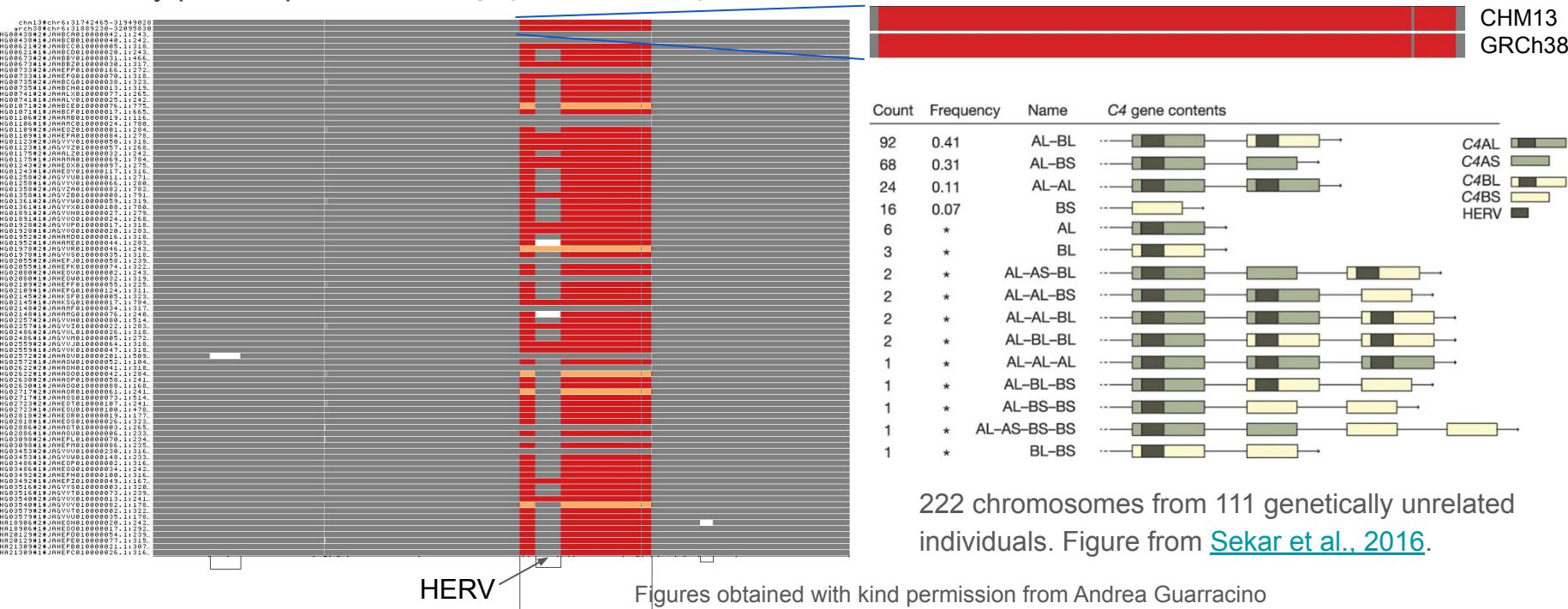
Pangenome graph of the human chromosome 6 with 90 haplotypes (44 diploid *de novo* assemblies plus the GRCh38 and CHM13 reference genomes). A portion of the 2D layout is shown.



Dissecting pangenome graphs - odgi extract

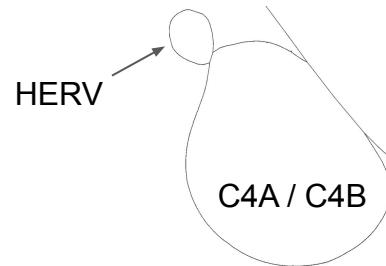
Pangenome graph of the C4 locus with 90 haplotypes (44 diploid *de novo* assemblies plus the GRCh38 and CHM13 reference genomes).

Colored by path depth (white = 0x, grey ~ 1x, red ~ 2x, yellow ~ 3x)



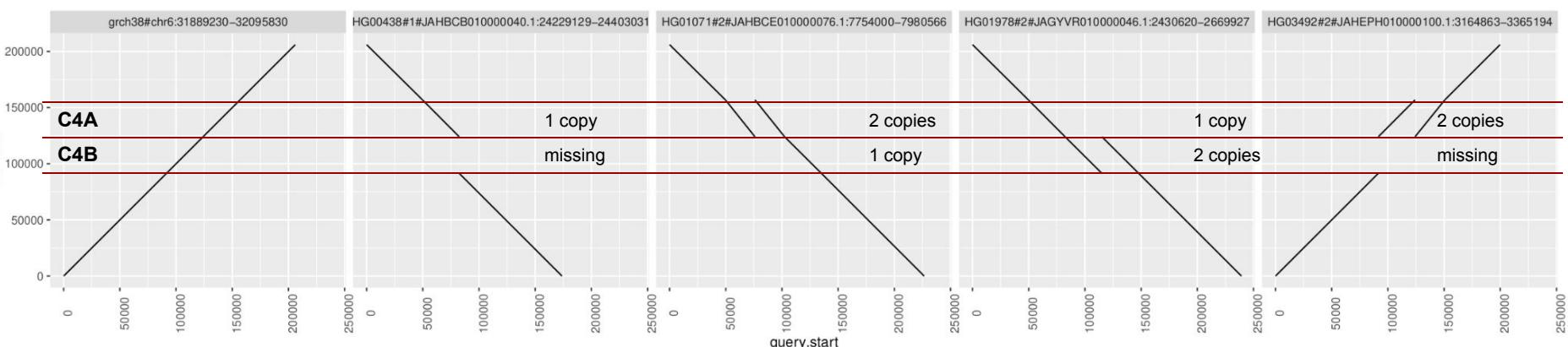
Untangling pangenome graphs - odgi untangle

Repetitive sequences produce collapsed repeats in the pangenome graphs.



Paired BED output

query.name	query.start	query.end	ref.name	ref.start	ref.end	score	inv	self.cov	nth.best
HG03492#2#JAHEPH010000100.1:3164863-3365194	0	91469	chml3#chr6:31742465-31949028	0	91441	0.998121	+	1	1
HG03492#2#JAHEPH010000100.1:3164863-3365194	91469	124145	chml3#chr6:31742465-31949028	124001	156853	0.99149	+	1.80362	1
HG03492#2#JAHEPH010000100.1:3164863-3365194	124145	150633	chml3#chr6:31742465-31949028	124001	156853	0.802825	+	1.99185	1
HG03492#2#JAHEPH010000100.1:3164863-3365194	150633	199828	chml3#chr6:31742465-31949028	156853	206060	0.997848	+	1.00026	1



Haplotypes representing the most frequent configurations found at the C4 locus in the [HPRC dataset](#).

Figures obtained with kind permission from Andrea Guaraccino

Annotating pangenome graphs - odgi position

Input: Chr8 human **consensus graph** originating from 44 *de novo* assemblies from the HPRC. Both references CHM13 and GRCh38 are fully preserved in the graph.

Tool: odgi position - Annotation lift over from an annotated reference path in the graph to the nodes in the graph via e.g. **BED**

Output: **TSV** file with the gene annotation per node for e.g. visualization.

Chr8 consensus graph - cytobands annotation

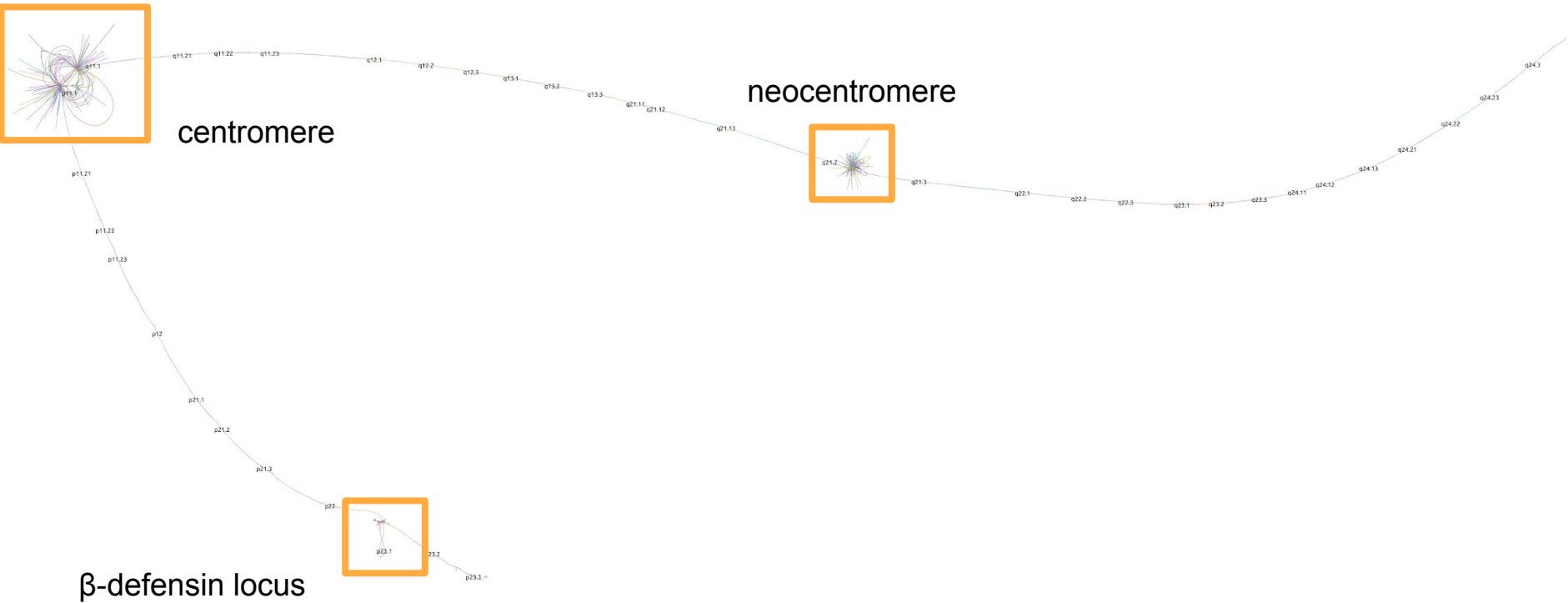
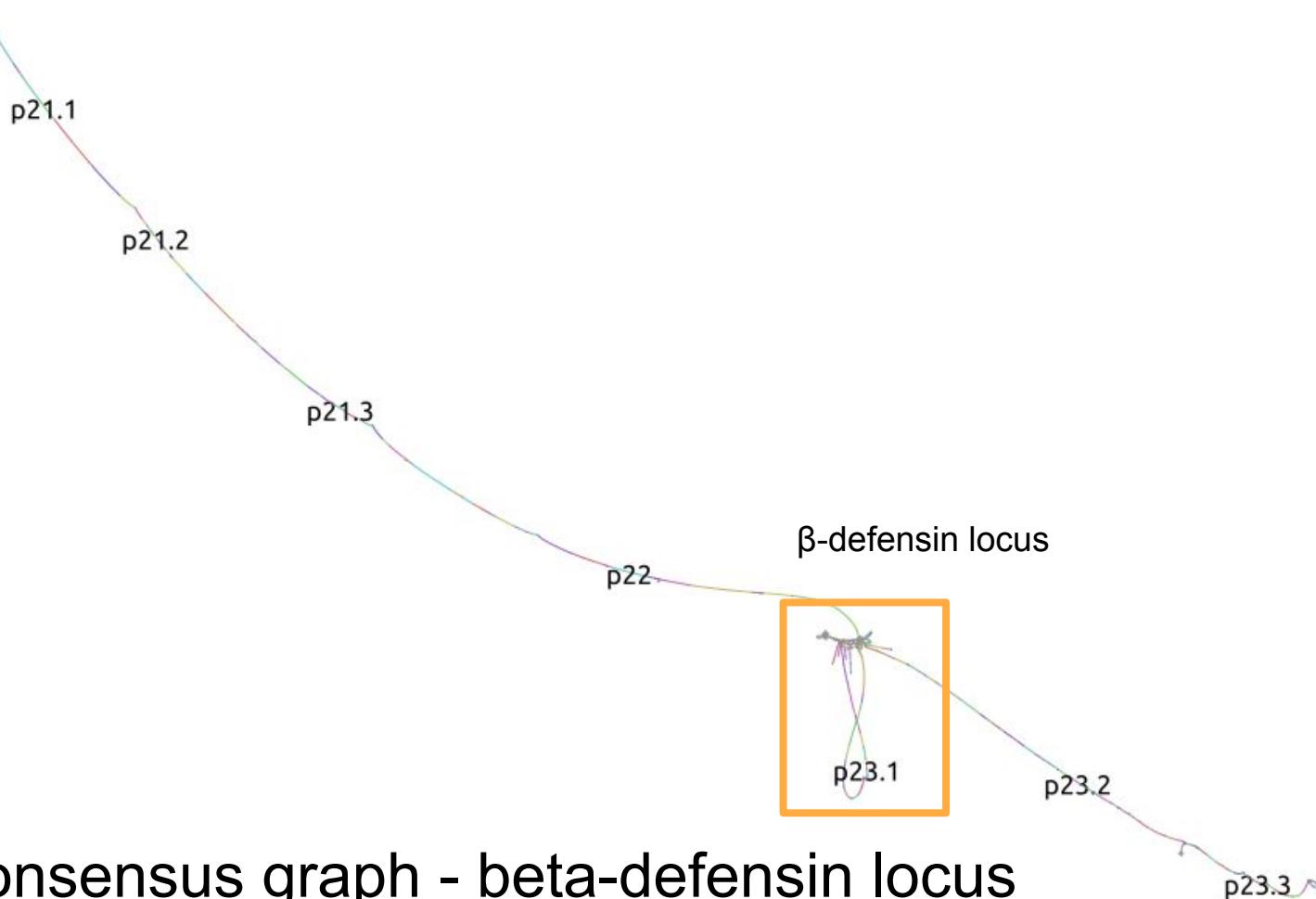


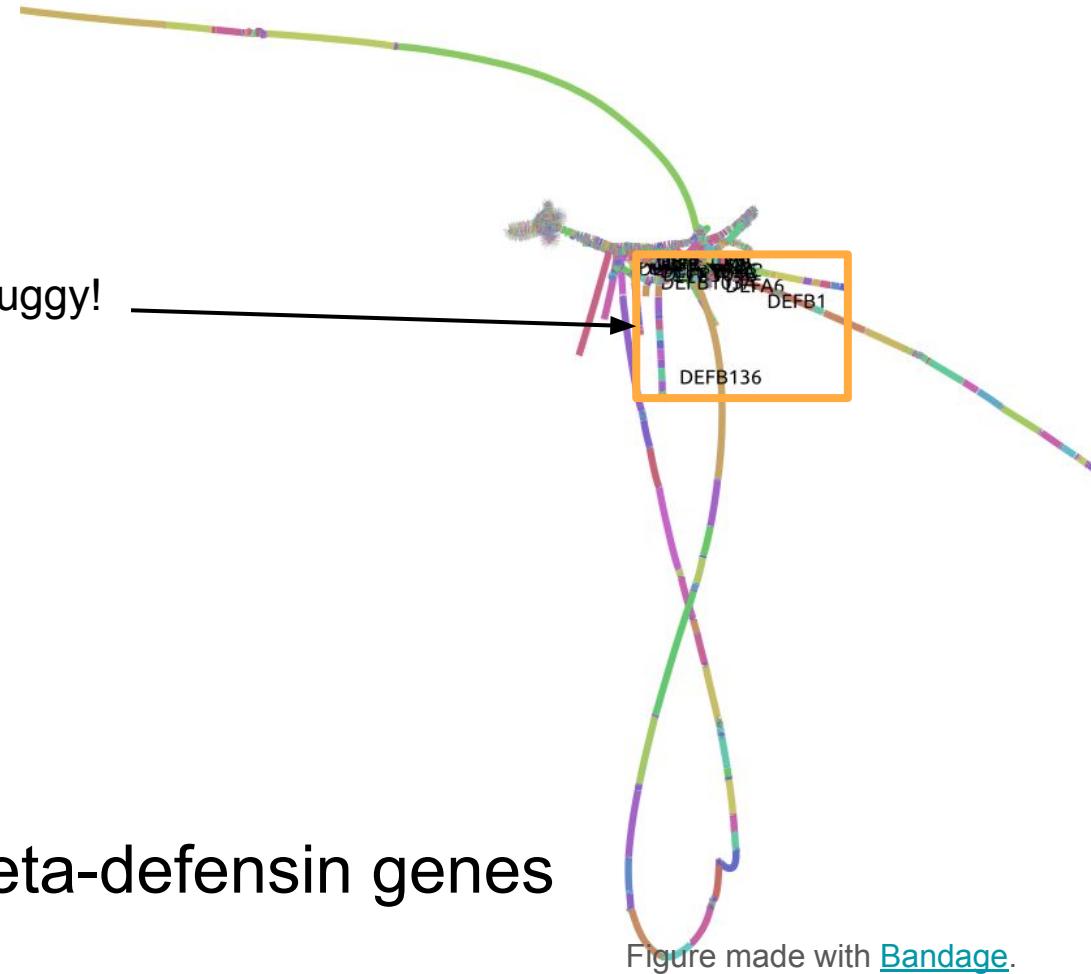
Figure made with [Bandage](#).



Chr8 consensus graph - beta-defensin locus

Figure made with [Bandage](#).

This graph must be buggy!



Identifying assembly breakpoints relative to the references

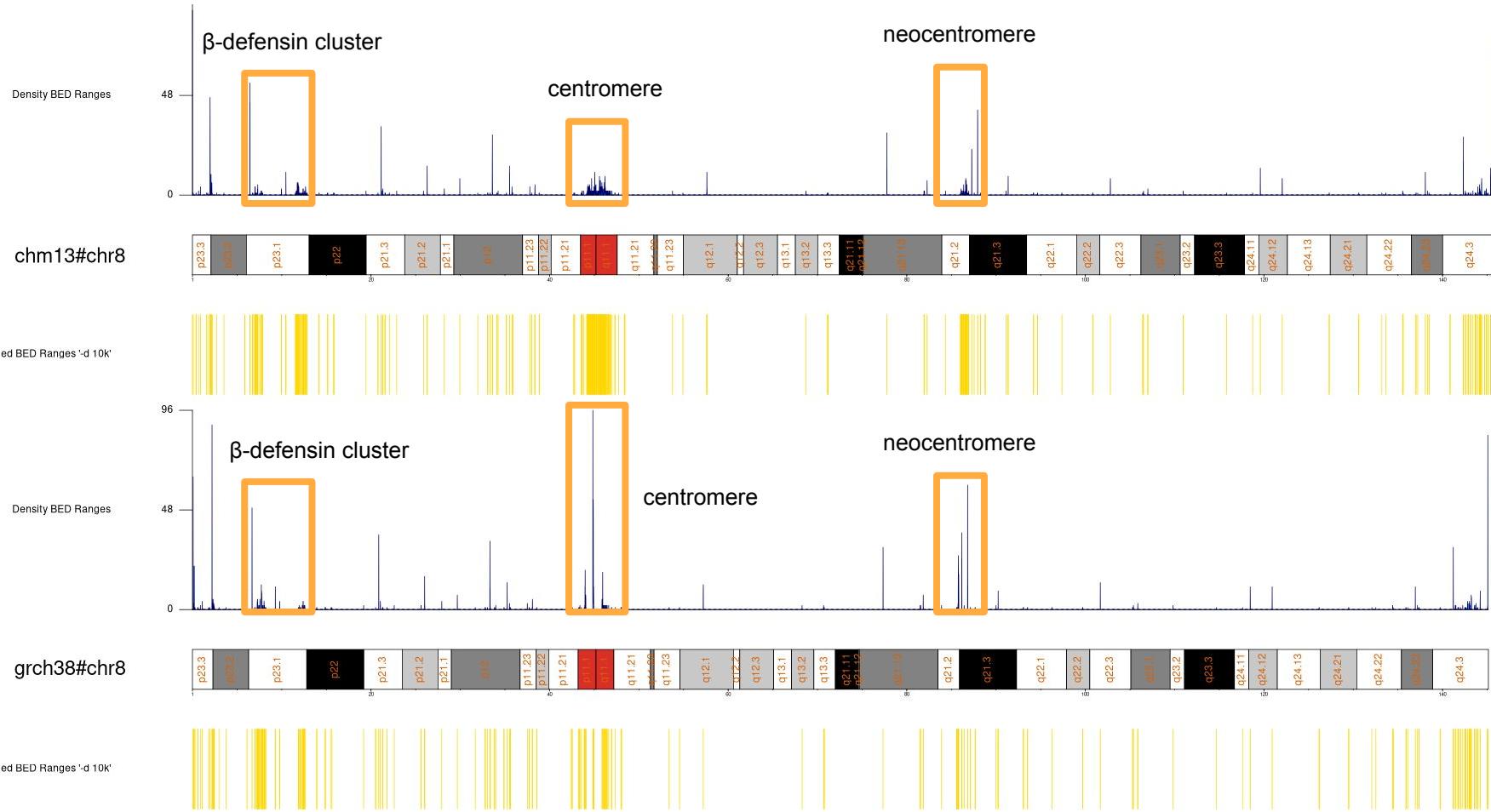
Where do our contigs' ends match the references? - Detecting regions that are difficult to assemble.

Input: Chr8 human pangenome graph made with contigs from 44 *de novo* assemblies from the HPRC adding CHM13 and GRCh38 - 90 haplotypes.

Tool: odgi tips - Walking from the ends of a contig until a reference node is found. For each contig range (e.g. a tip) we look at each possible reference window and find the most-similar one.

Output: BED file with the best reference hit and position for each contigs' ends.

Assembly Breakpoint Ranges of the Contigs in chr8 in the HPRC PGGB RC1 Graph relative to CHM13 and GRCh38



Discussion



[ODGI repository](#)



[ODGI documentation](#)

ODGI: State-of-the-art tool box to transform, analyse, simplify, validate, and visualize pangenome graphs at large scale.

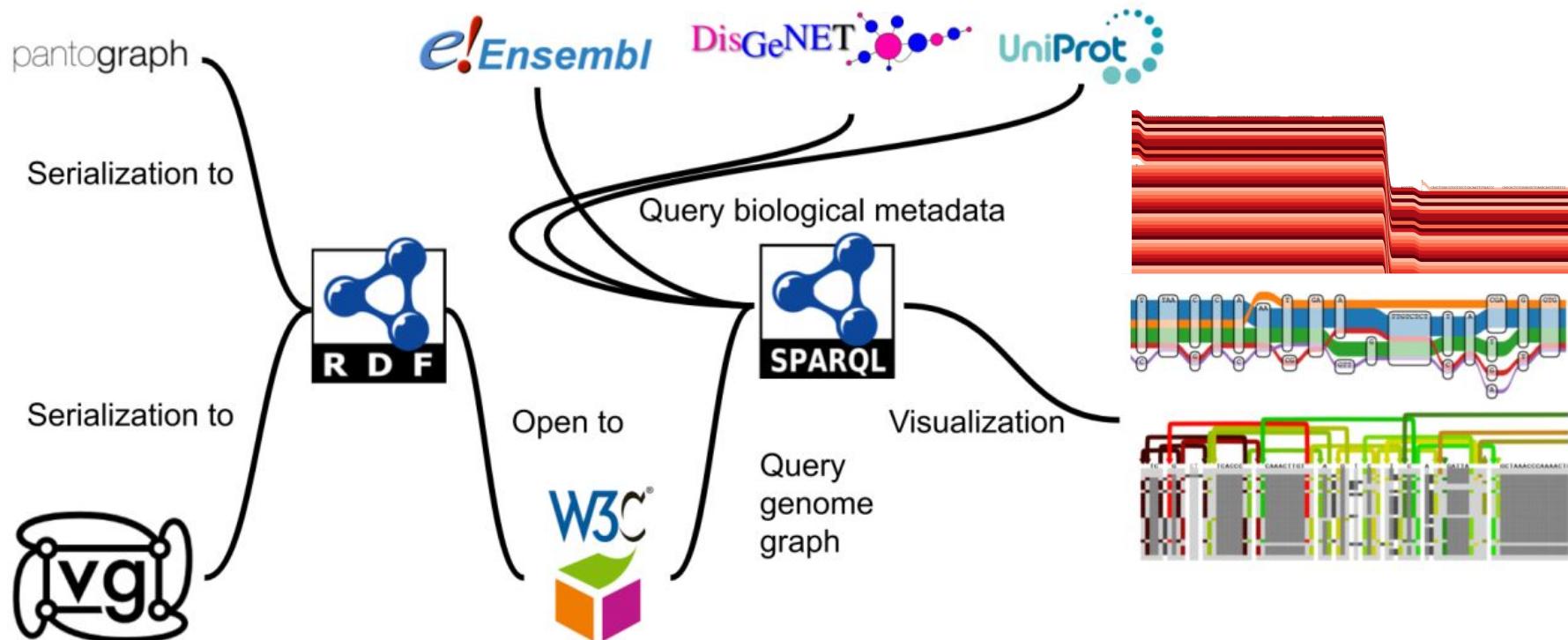
Bridge between linear reference genome analysis and pangenome graphs:
Subgraph extraction, lifting over annotations, linearizing nested graph structures

Discover the underlying biology of pangenome graphs:
Detect complex regions, identify assembly breakpoints

The tools already are the backbone of pipelines such as the Pangenome Graph Builder ([PGGB](#)) or [nf-core/pangenome](#).

Future work: RNA and protein support, expand metadata capabilities

The overview of Pangenome ontologies



<http://biohackathon.org/resource/vg>

Resource Description Framework (RDF)

- **IRI**
Synonym: URI is often an URL
- **Literal**
“Strings”, 1, 1.001
- **Bnode**
Placeholder identifiers: Something exists, but you don't know its identity
- A **Triple** is a statement
Subject → Predicate → Object

Resource Description Framework (RDF)

- A **Triple** is a statement
Subject → Predicate → Object

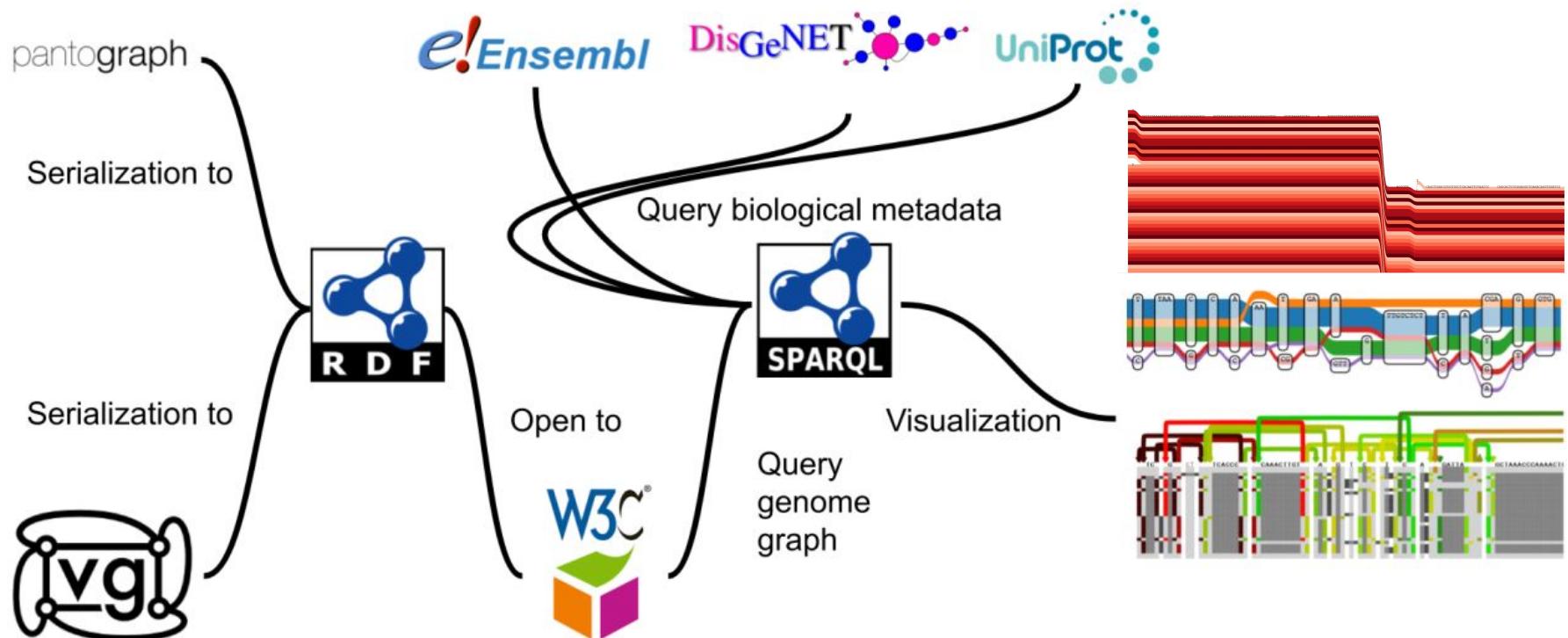
I → love → my cat

<https://uni-tuebingen.de/forschung/forschungsinfrastruktur/zentrum-fuer-quantitative-biologie-qbic/team/simon-heumos/> ↗

<https://www.dictionary.com/browse/love> ↗

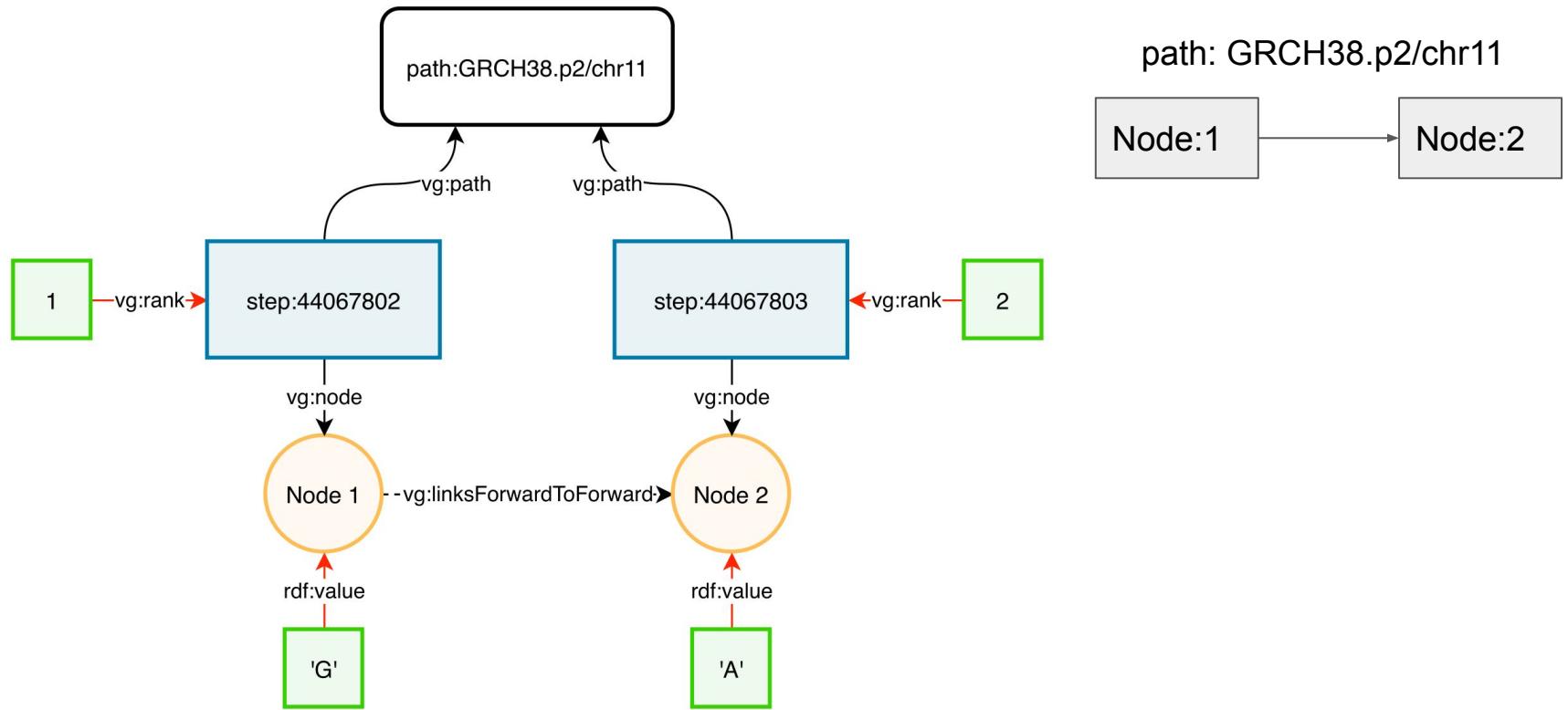


The overview of Pangenome ontologies

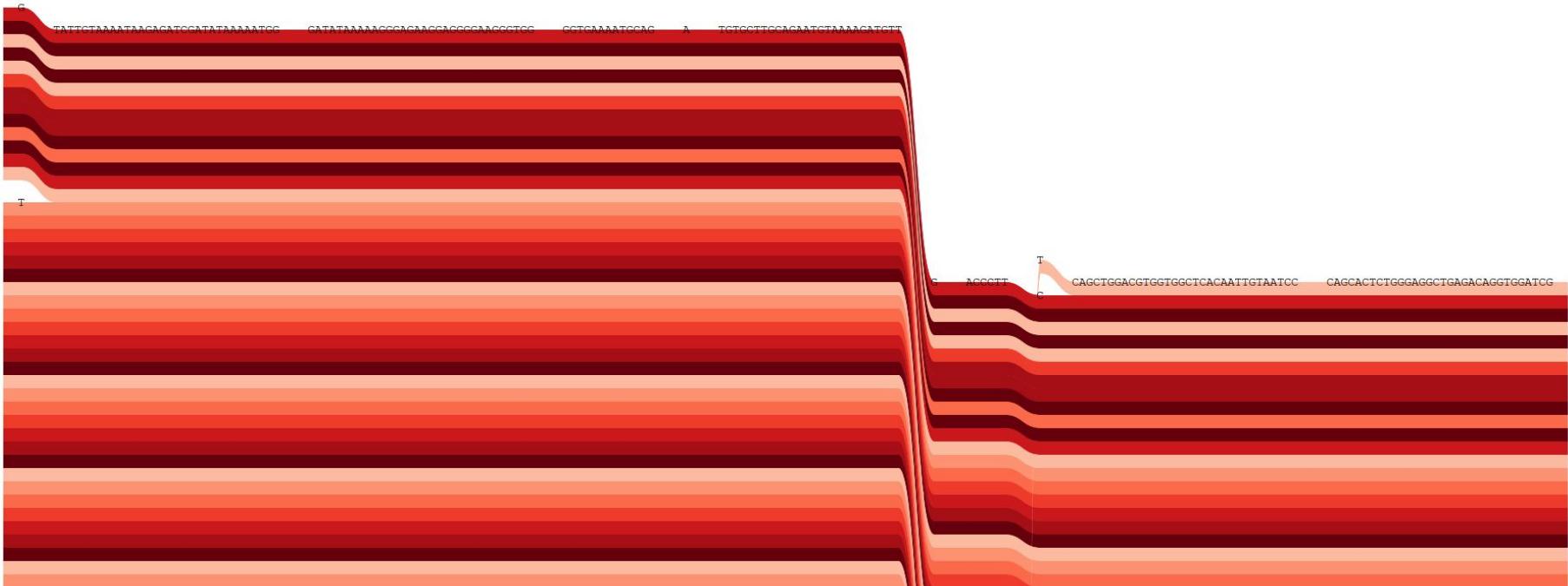


<http://biohackathon.org/resource/vg>

Variation graph ontology

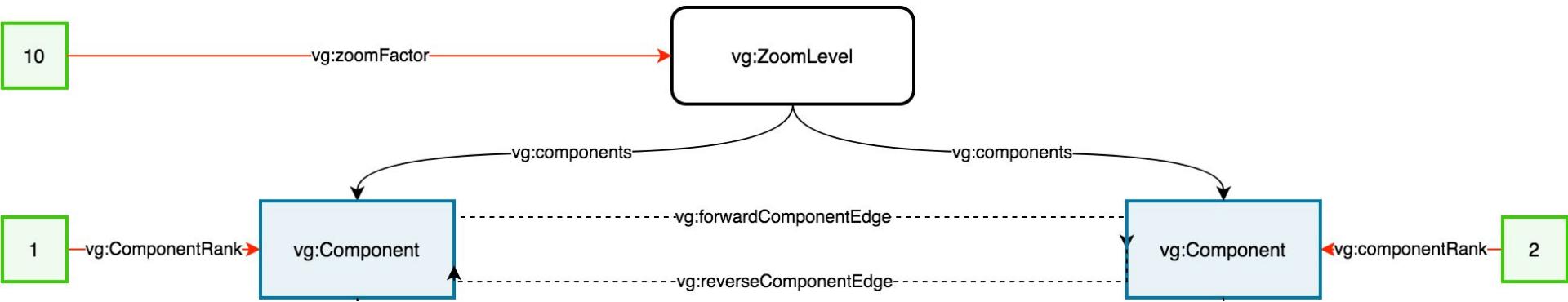


SPARQL Querying a Semantic Genome Graph Japan BH 2019 - Example Visualization¹

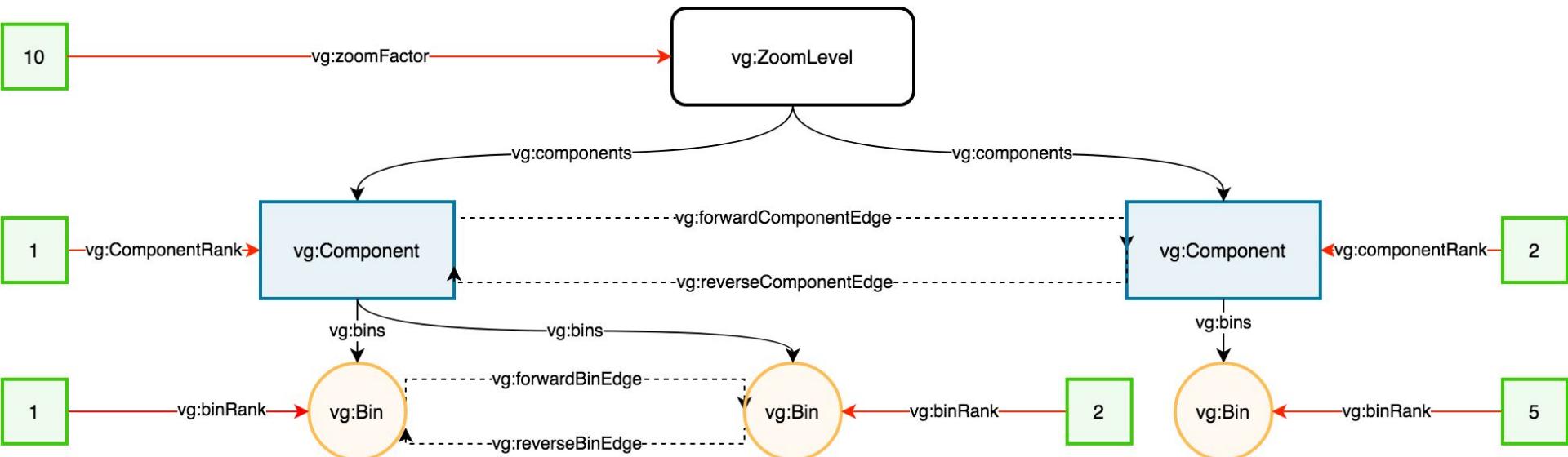


¹<https://twitter.com/simonheumos/status/1169884828860239874>

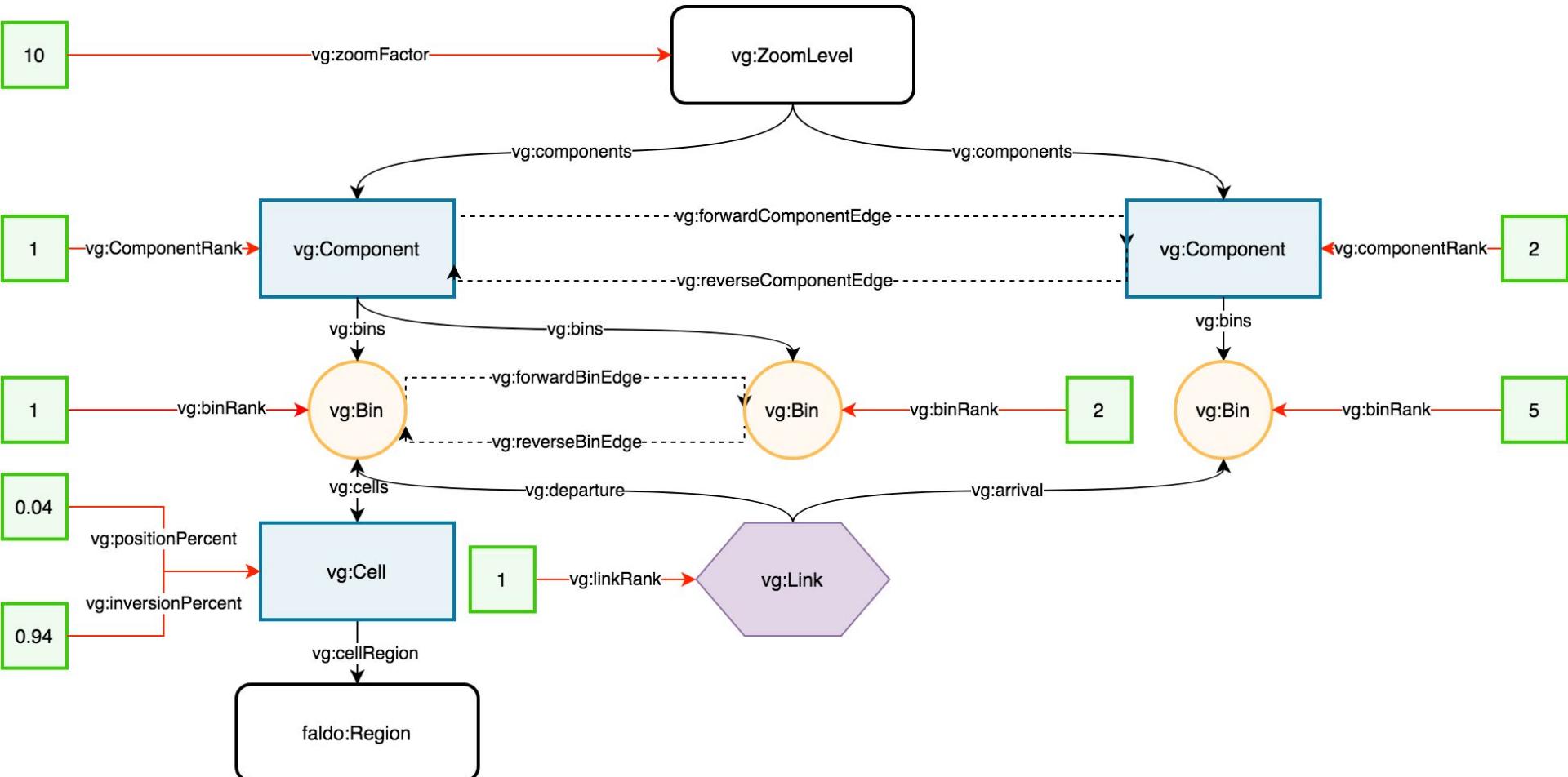
Each ZoomLevel has multiple Components



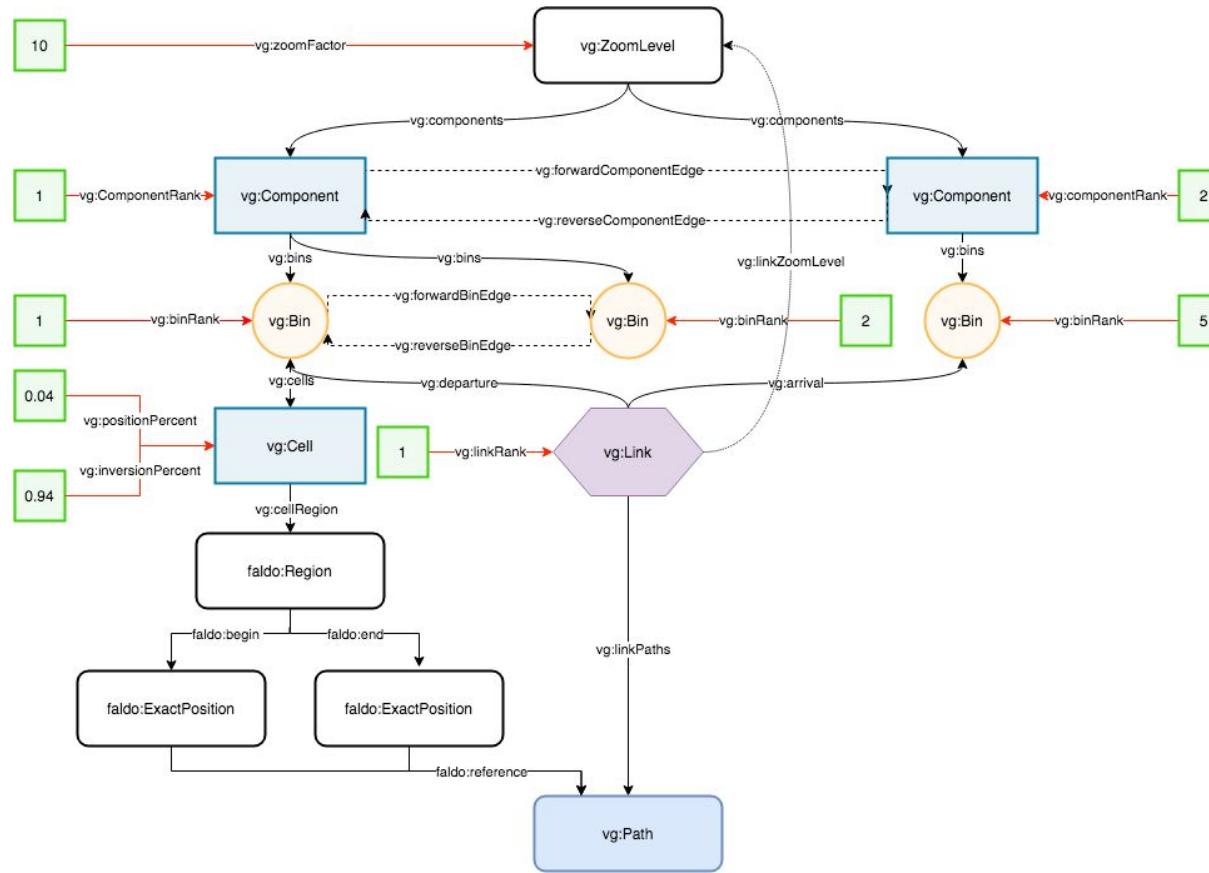
Each Component has multiple Bins

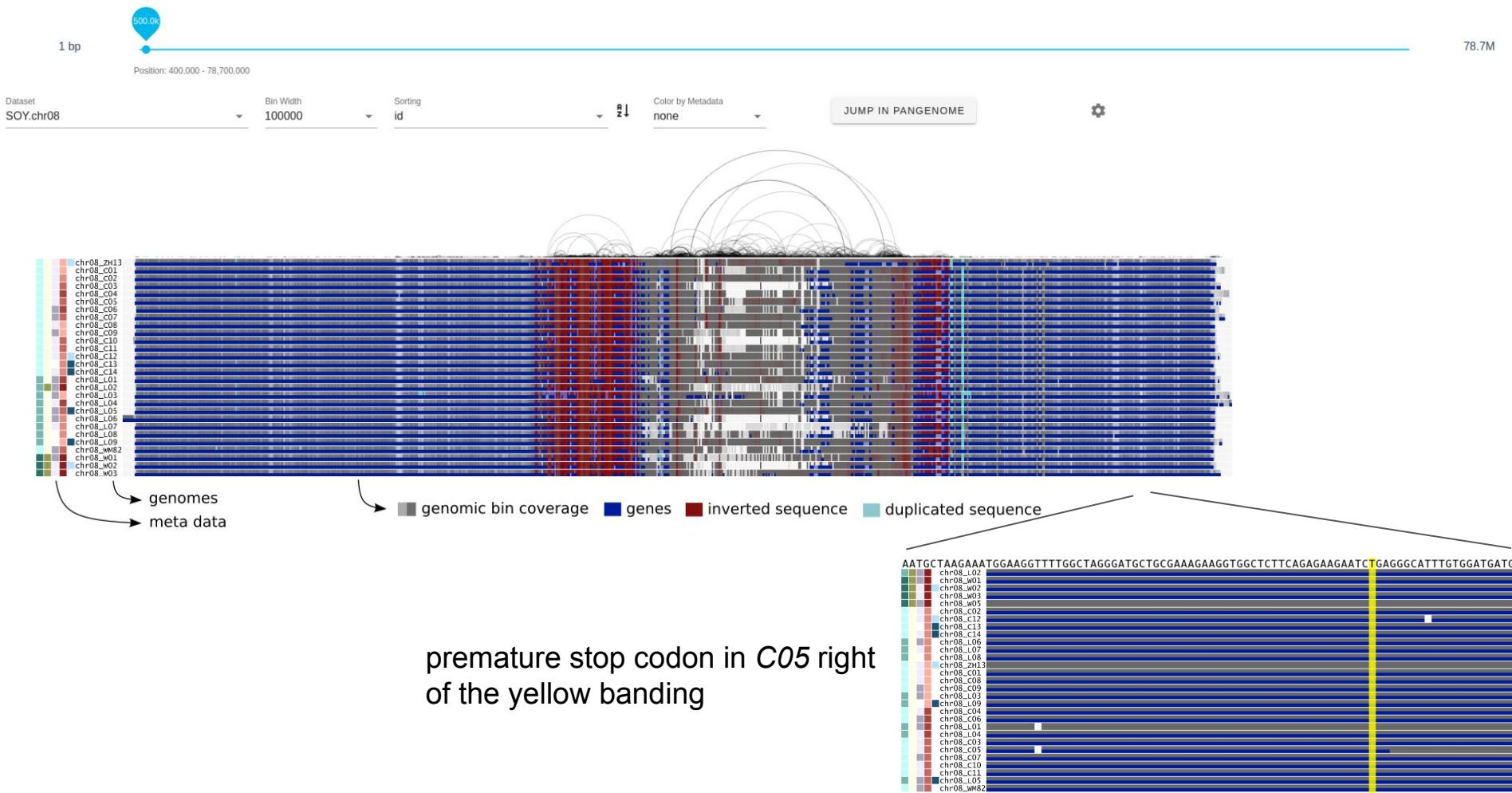


Each Bin is connected by Link, and has Cells

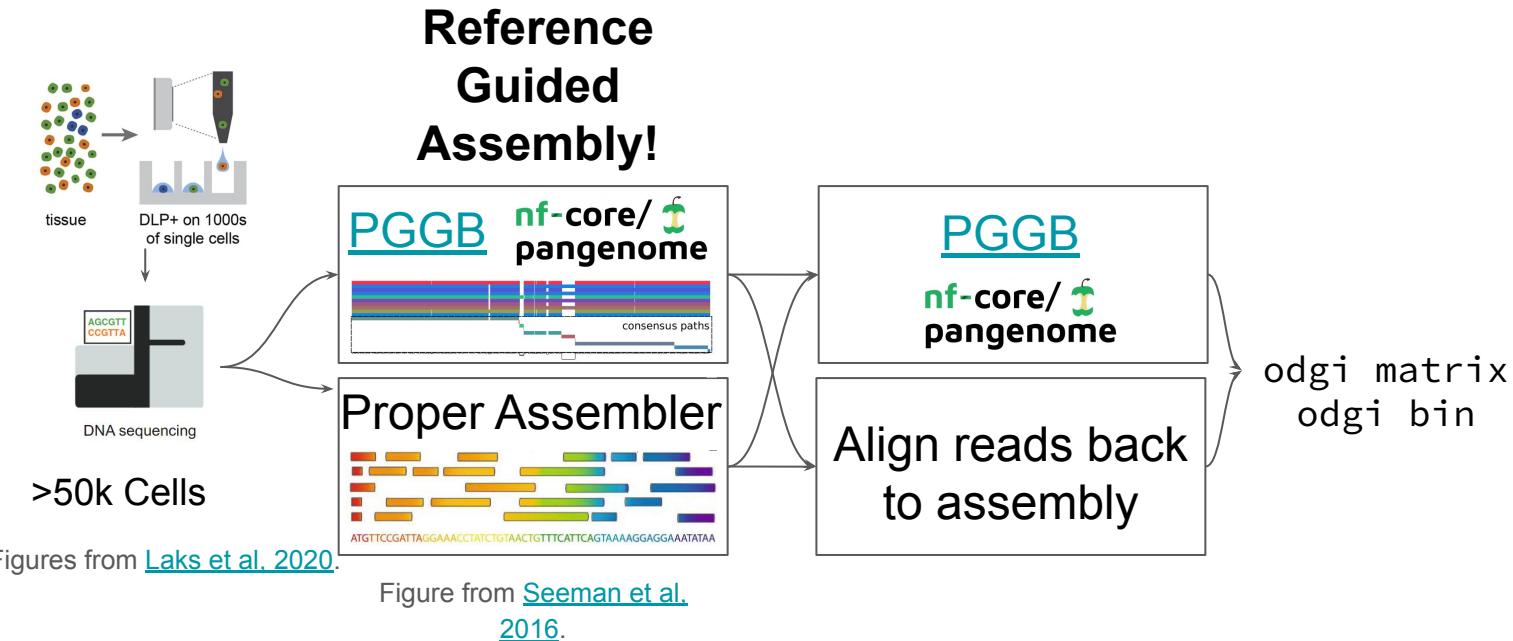


vg ontology is referenced via Link and Cell





Brain exercise: Building pangenome graphs from cancer scWGS data



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BONUS SLIDES

- Simon: [Detecting complex regions with ODGI depth](#)
- Christian Kubica: [Evaluation of short read mappers to a pangenome graph with *Arabidopsis thaliana*](#)
- Andrea: [Chromosomal communities](#)
- Flavia Villani: [Chr 19 mouse pangenome graph](#)
- Jerven: [Zero Extra Costs SPARQLable Pangenome SpOdgi OR SapFhir](#)
- Jörg: [Pantograph Demo](#)
- Christian Fischer: [gfaestus Demo](#)
- Simon: [Which contigs travel T2T and are centromere assemblers?](#)

Which contigs travel T2T and are centromere assemblers?

Reminder: [odgi tips](#) - Where do our contigs' ends match the reference?

T2T classes:

- **centromere**: One of the contigs' end must match chm13 in the p arm, the other contig's end must match chm13 in the q arm. Both matches are outside of the centromere as given by the chm13 centromere cytobands.
- **centromere_plus_X**: We artificially extend the size of the centromere by **X** nucleotides in both the p and the q arm. The higher **X**, the more confident we can be that we identified full centromere-assembling contigs!
- **telomere_plus_X**: We define a nucleotide length **X** which we travel from both chromosome arms. Contigs' ends matching chm13 within these ranges are full T2T contigs!

Detecting complex regions

Human chromosomes have large regions of highly identical repeats:

- Clusters centromeres
- Regions of segmental duplication
- In the acrocentric short arms of chromosomes.

[Logsdon et al., Nature 2021](#): Chr8 carries a modestly sized centromere of approximately 1.5–2.2 Mb, in which AT-rich, 171-base-pair (bp) α -satellite repeats are organized into a well-defined higher-order repeat (HOR) array.

ODGI offers tools to detect and explore such regions.

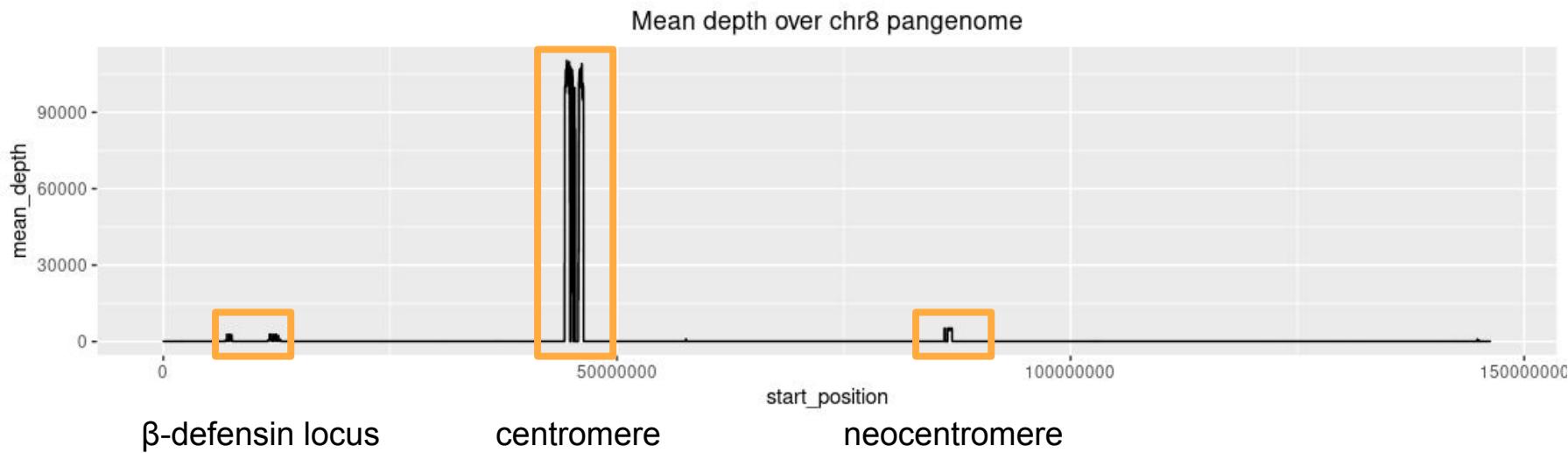
Detecting complex regions - odgi depth

Input: Chr8 human pangenome graph made with 44 *de novo* assemblies from the HPRC adding CHM13 and GRCh38 - 90 haplotypes.

Tool: odgi depth - calculate node depth: For each node, we record the number of times a node is crossed by all the paths present in the graph.

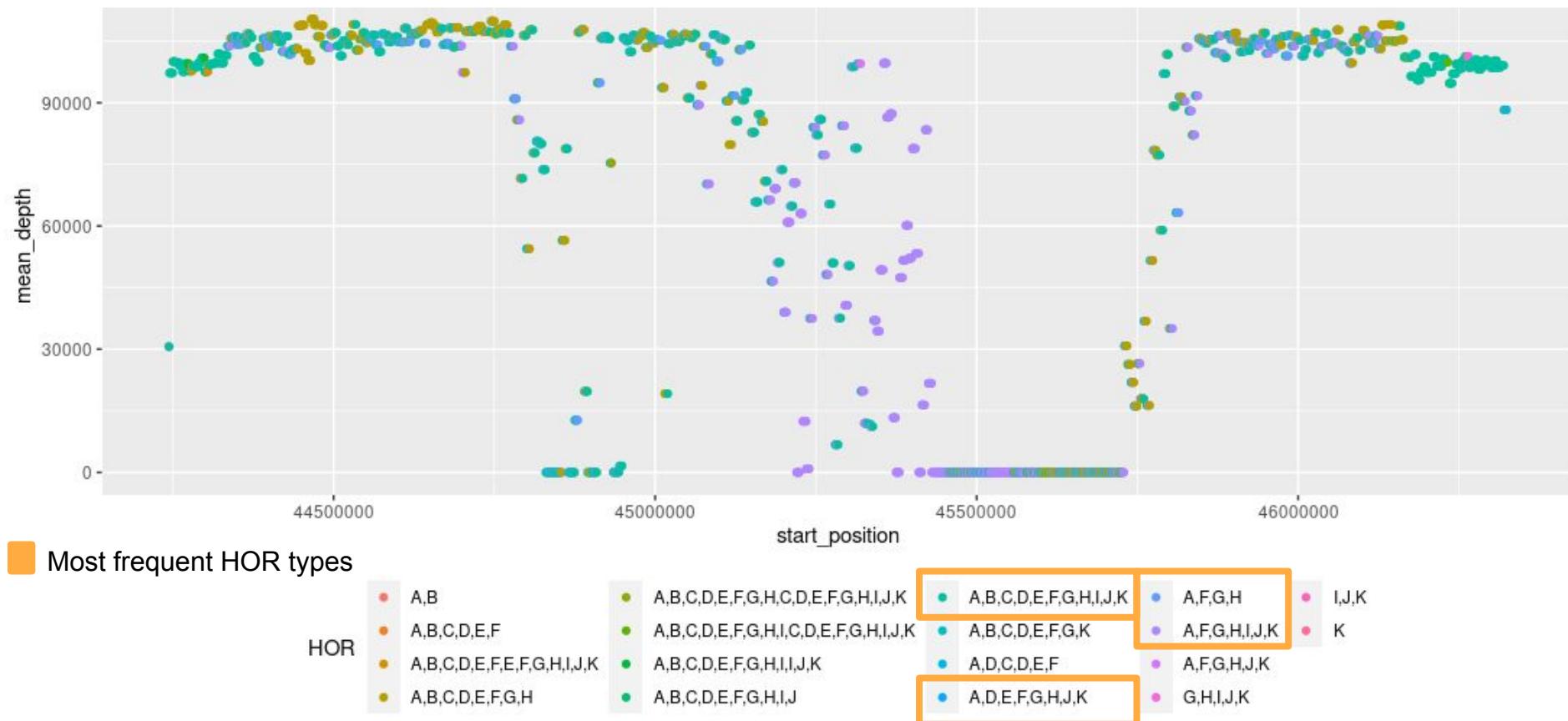
Output: BED file with the mean node depth distribution across windows of the pangenomic CHM13 positions.

Detecting complex regions - Mean depth distribution



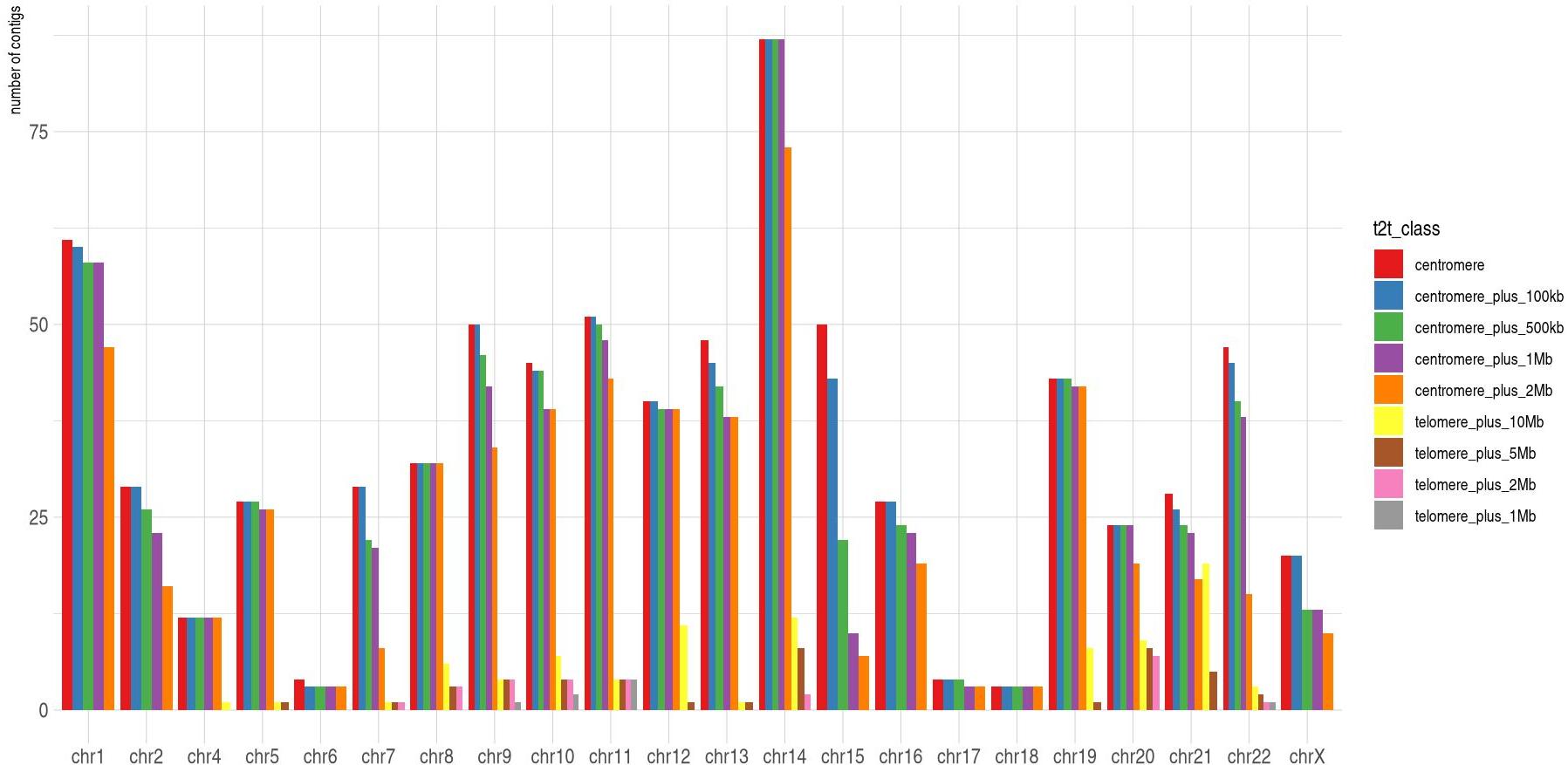
Chromosome 8 does not only possess a centromeric region, but a complex beta-defensin gene locus and a VNTR that can function as a neocentromere.

Mean depth of the centromeric HOR array in Chr8



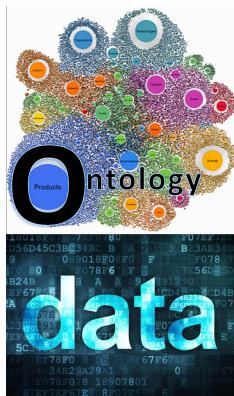
Every letter indicates an alpha-satellite monomer in the HOR: For example A,B,C,D,E,F,G,H,I,J,K would indicate an HOR with 11 alpha-satellite monomers. The mean depth drop falls into the hypomethylated and CENP-A-enriched regions, that have the highest consistent entropy in the entire array. This agrees with the [Loqsdon et al., Nature 2021](#) publication.

Number of telomere-to-telomere and centromer-crossing contigs by chromosome and T2T class

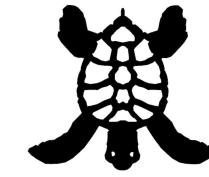


Zero Extra Costs SPARQLable Pangenome SpOdgi

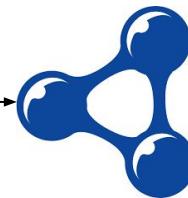
vg Ontology



SpOdgi



Triples



SPARQL Endpoint

Variation Graph

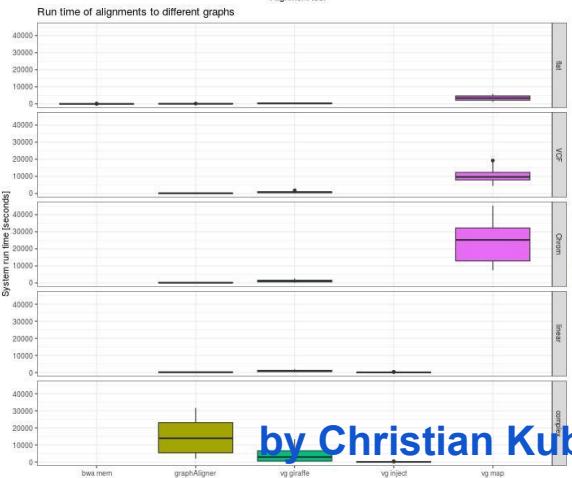
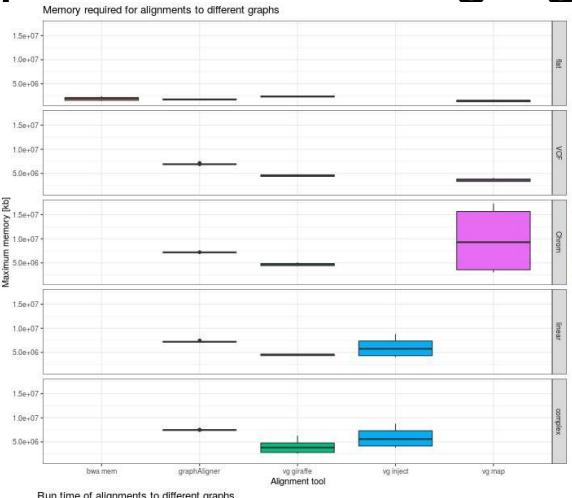
by Jerven Bolleman

Short read alignments to complex graphs are challenging

Alignments of 12 *A.thaliana* short read sets to graphs of different complexity

- using a set of 12 short read sets (6x N90 100 bp; 6x N90 250 bp)
- Using 5 different alignment tools
 - [bwa mem](#) (as control on the flat sequence)
 - [graphaligner](#)
 - [vg map](#)
 - [vg giraffe](#)
 - [bwa mem](#) + [vg inject](#) (using all genomes concatenated)

Flat graph	VCF graph	Chrom pggb	Linear pggb	Complex pggb
Flat reference genome imported without variation	Variation called from the complex graph used to build a vg graph	Independent chromosome graphs concatenated into one graph.	Graph build from all chromosomes and contigs. Enforcing linearity over compression.	Graph build from all chromosomes and contigs. Trying to compress repeated sequences.
Graph size: 119 mb	Graph size:: 220 mb	Graph size: 197 mb	Graph size:: 195 mb	Graph size: 168 mb
Nodes: 3,829,320 Edges: 3,829,313	Nodes: 5,601,134 Edges: 8,307,979	Nodes: 6,247,306 Edges: 8,491,459	Nodes: 6,282,045 Edges: 8,540,670	Nodes: 6,694,806 Edges: 9,147,110

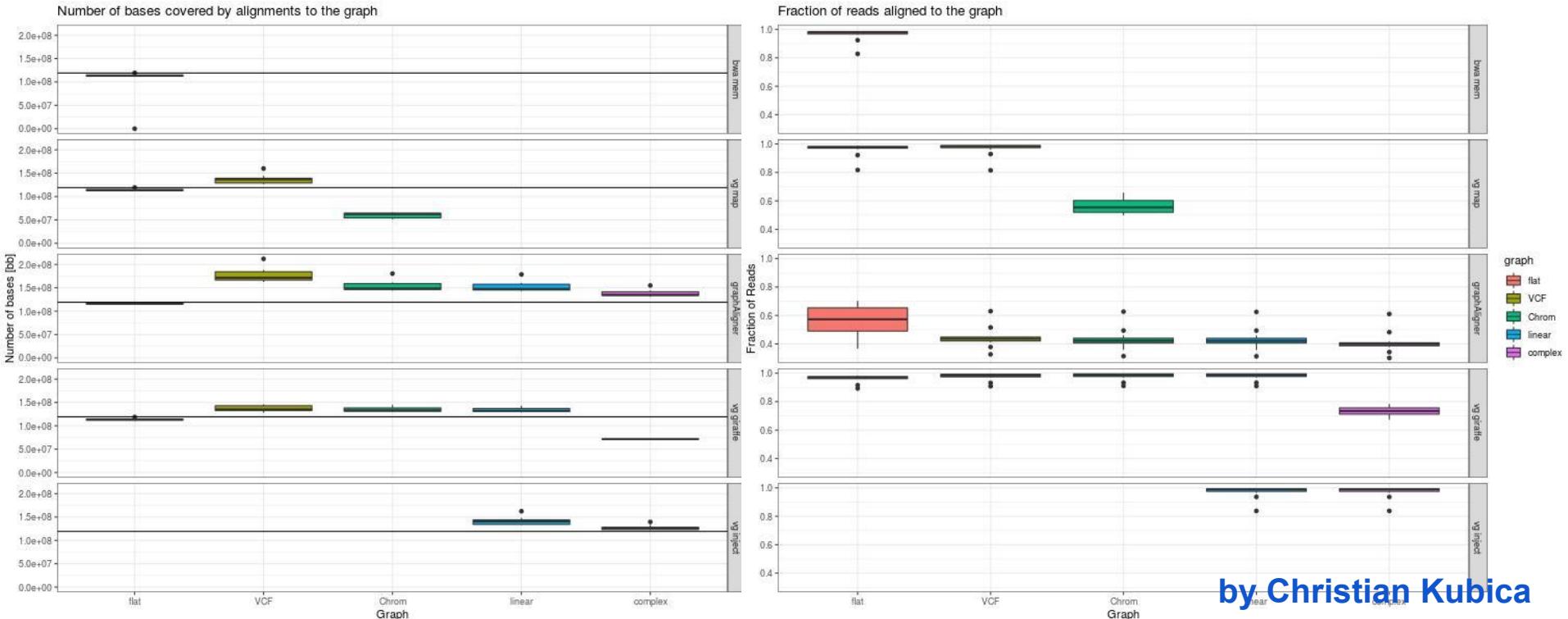


by Christian Kubica

Short read alignments to complex graphs are challenging

vg mappers are not made for WGA derived graphs and lose accuracy. *graphaligner* struggles with short reads.

Current workaround: mapping to flat references and injecting into the graph. Only known variation can be covered



Chromosome 19 mouse pangenome

BXD Family

Ashbrook, D. G., et al. Cell Systems (2021)

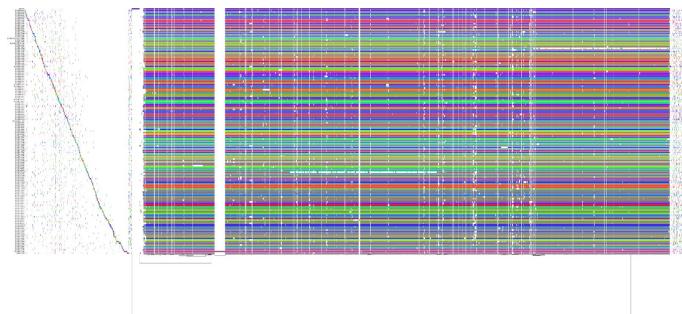
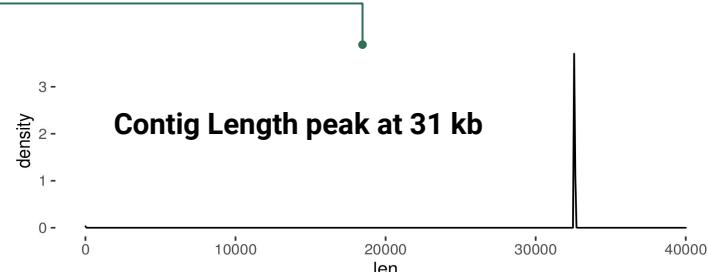
148 BXD strains assembled
(10x technology)

Pangenome of Chr19

148 BXD strains assembled

Nucleotides from pangenome		
A	87,307,913	14,49 %
C	63,277,613	10,49 %
G	63,151,974	10,47 %
N	302,016,734	50,10 %
T	87,048,427	14,45 %
tot	602,802,661	

Pangenome made of
5.5M nodes and 8.6M edges
(total length 264,380,676 bp
with 82,695 paths)

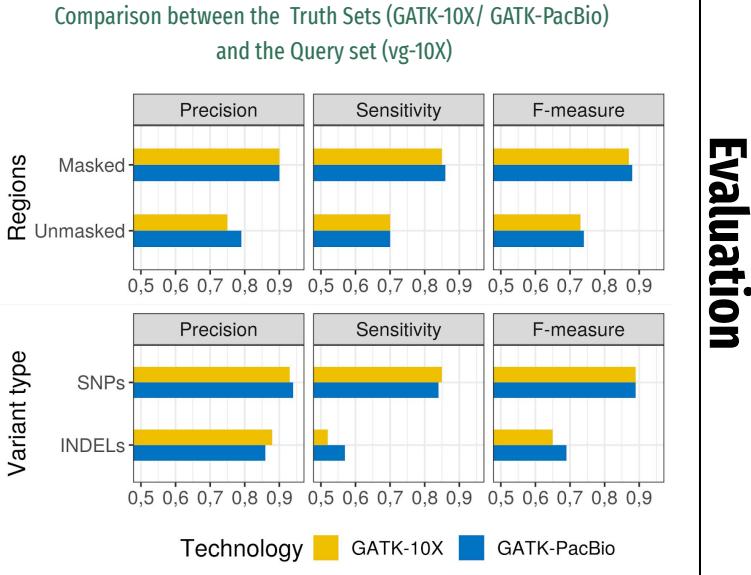


by Flavia Villani

Chromosome 19 mouse pangenome

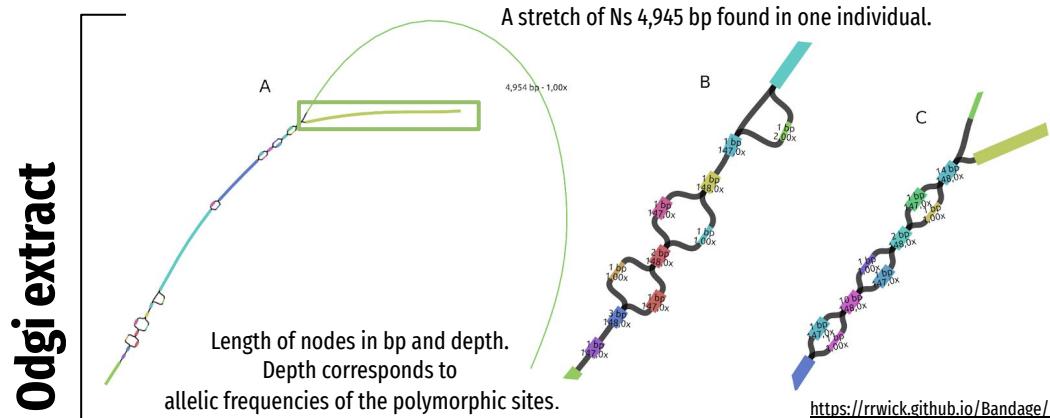
A

Variant Calling of chr19 (DBA/2J sample)



B

Odgi extract of Gene 2700046G09Rik



<https://sines.eimb.ru/>

by Flavia Villani

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