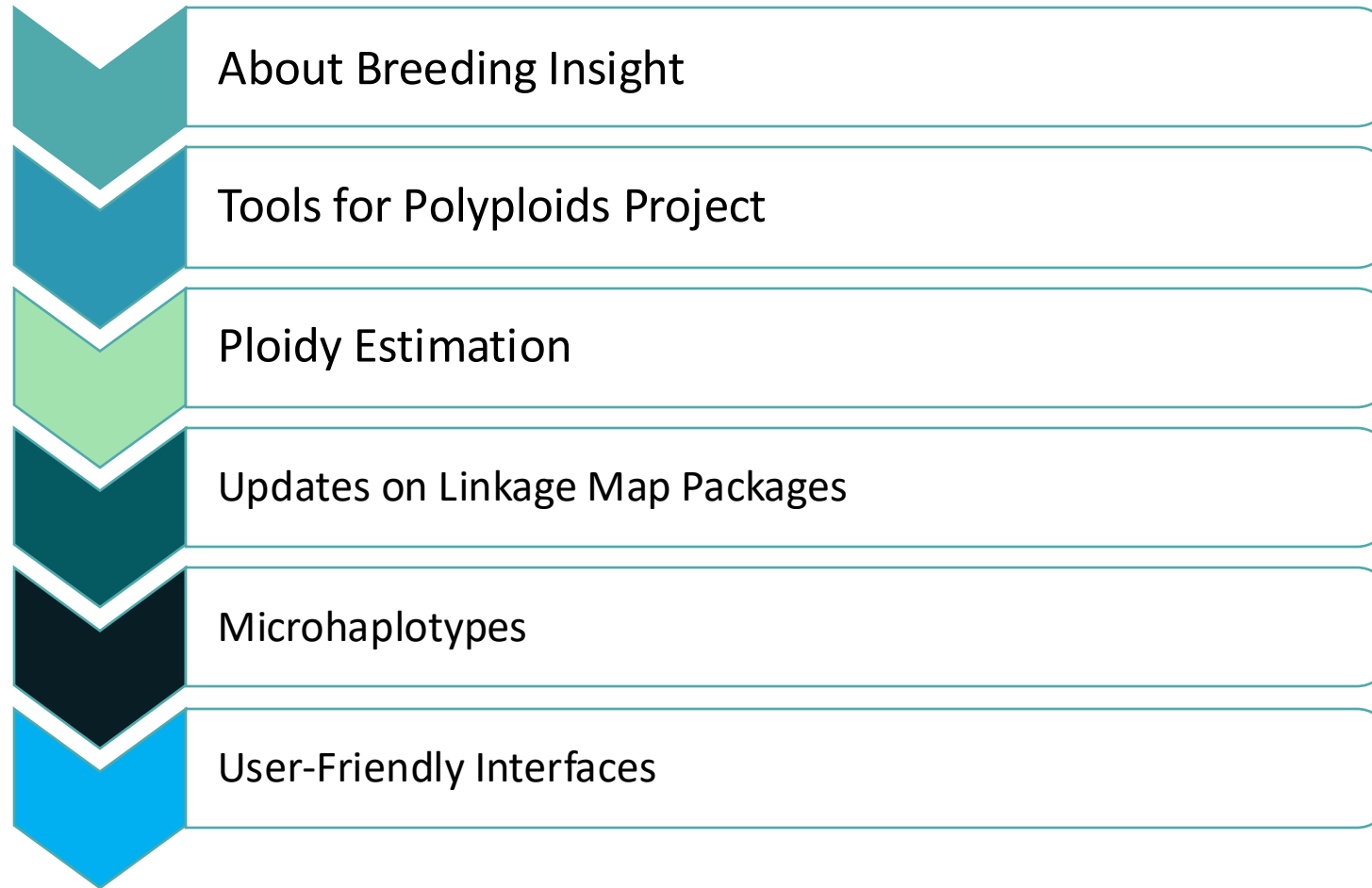


Enhancing Genetic Insights with Polyploid Tools and User-friendly Interfaces

Cris Taniguti
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Genomics Coordinator
Breeding Insight, Cornell University

Outline



Breeding Insight



Mission

Transform breeding
by enabling the implementation
of **genomic insight and selection**
as part of **routine specialty crop and**
livestock breeding across all of
USDA-ARS.

About



Breeding Insight

Who we are:

- We connect and build resources for specialty crop and livestock breeders.
- We are a USDA-ARS-funded initiative located at Cornell University.
- We are a service and technology transfer model providing additional support to USDA specialty breeding programs.
- We are a team made up of scientists and software developers.

About



Breeding Insight

What we are not:

- We are not a private company.
- We are not interested in duplicating established technology, techniques, or patents.
- We are not selling and we are not receiving commission for the technology we provide or recommend.

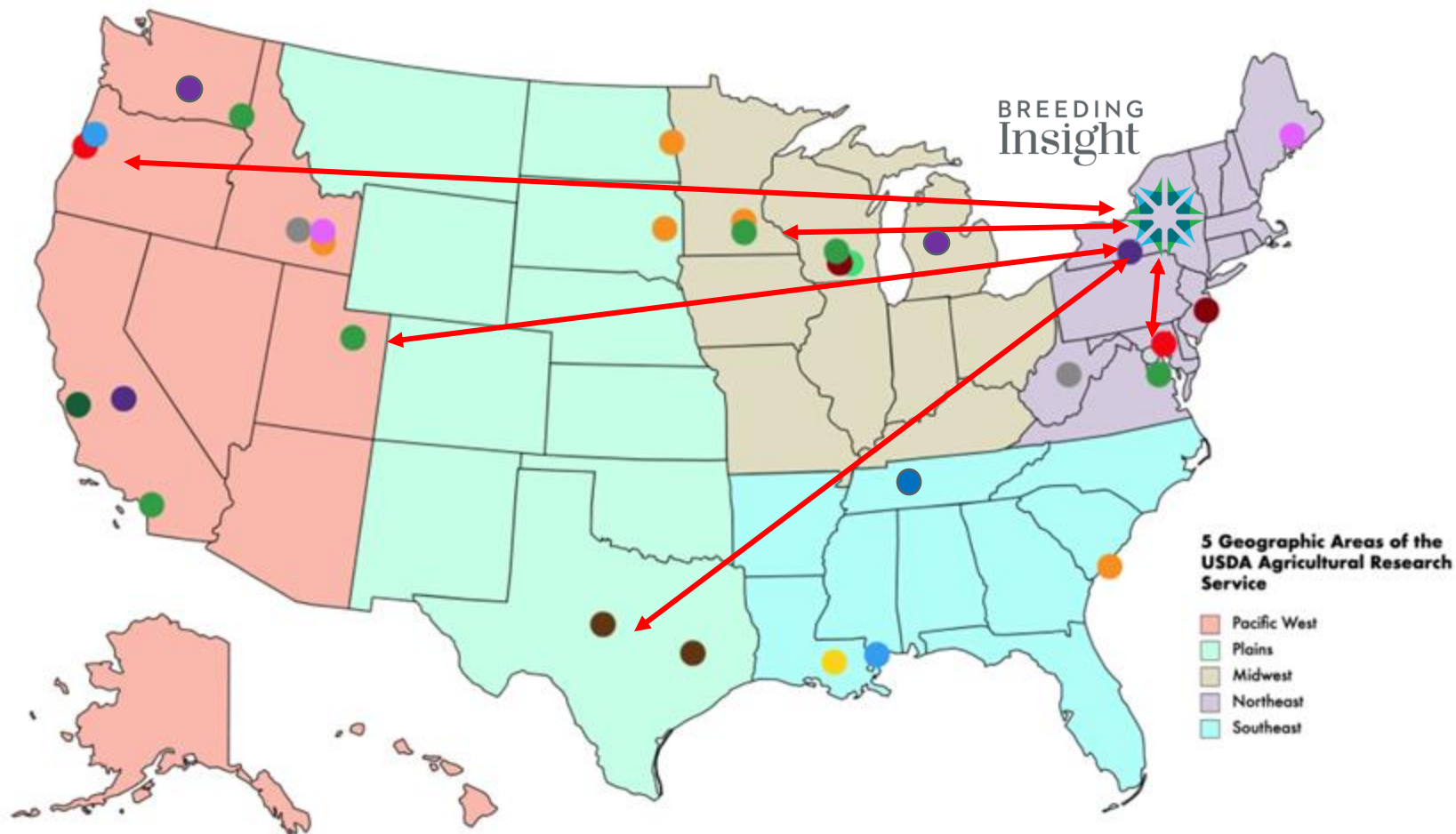
Breeding Insight Tools and Capacity



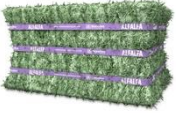
- Phenomics
- Genomics
- Bioinformatics
- Statistics
- Software Developers
- Image Analysis
- Data QC
- Pedigree
- Marker Validation

27 species

59 breeding programs



Exploring Resistance Traits



- Salt resistance
- Drought Stress tolerance
- pH tolerance
- Aphanomyces resistance
- Anthracnose resistance



- Potato Cys Nematore (PCN) resistance



- Nematode resistance
- Fusarium resistance
- Wireworm, Diabrotica, and Systema (WDS) resistance
- Insect resistance



- Mite resistance



- Powdery mildew resistance



- Powdery mildew resistance



- Impatiens necrotic spot virus resistance



- Powdery mildew resistance
- Downy mildew resistance
- Cold resistance



- Shock virus resistance



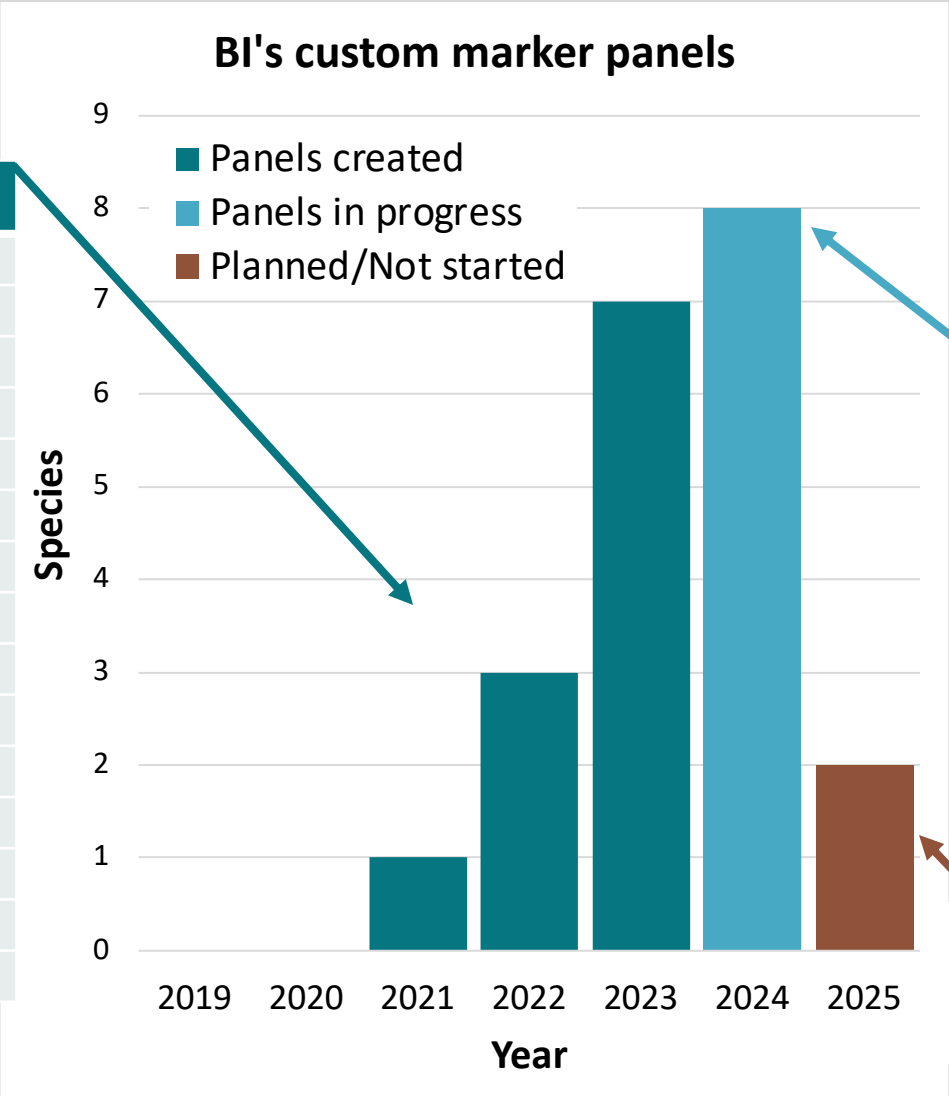
- Resilience to changes in water quality and pathogen load

Genomic Resources

*Purchased by collaborators
 **Designed and purchased by collaborators
 ***Co-designed with collaborators

- ✓ Accessible
- ✓ Reliable
- ✓ Fast
- ✓ Affordable
- ✓ Repeatable

Species	Available Panels
Grape	2K rhAmpSeq**
Alfalfa	3K DArTag
Blueberry	3K DArTag
NA Atlantic Salmon	50K Infinium array*
NA Atlantic Salmon	3K DArTag
Lettuce	3K DArTag
Cucumber	3K DArTag
Pecan	3K DArTag
Cranberry	3K DArTag
Honeybee	81 DArTmp
Sweetpotato weevil	101 DArTmp
Sweetpotato	3K DArTag
Strawberry	3K DArTag**
Strawberry	5K DArTag**
Cranberry v2.0	3K DArTag



Species	In Progress Panels
Grape & Muscadine	3K DArTag***
Hop	3K DArTag
Hemp	3K DArTag
Blackberry	3K DArTag
Red clover	3K DArTag
Lettuce (parentage ver.)	100 KASP
Honeybee	105 AgriSeq
Trout	GTSeq
Sugar beet	5K AgriSeq**

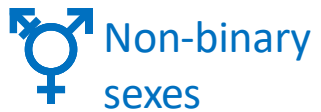
Species	Planned Panels
Raspberry	3K DArTag
Hydrangea	3K DArTag

Genomic Challenges for Specialty Species

Haplodiploid 1x & 2x

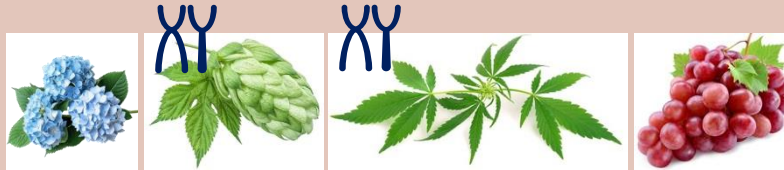


Heteromorphic
sex chromosomes



Allopolyploid
Autopolyploid
Mixed (some level
of preferential
pairing)

Diploids 2x



Polyploids

4x



6x



8x



~10x



≤12x



Auto



Auto



Allo



Allo



Auto



Auto

Auto



Auto/Allo/Mix

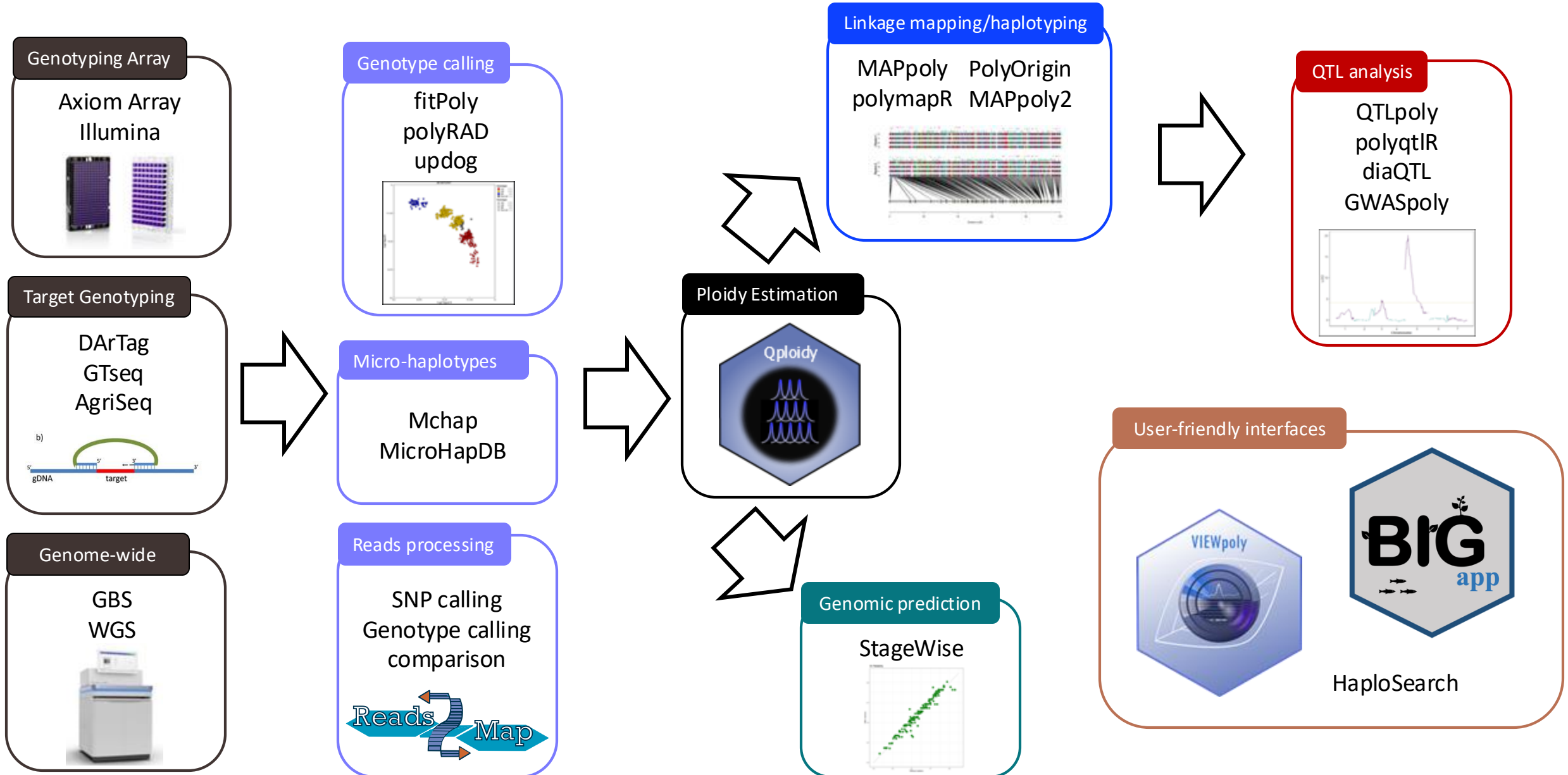


Tools for Polyploids Project

- ▶ Computational Tools Development and Improvement
- ▶ Collaboration between developers and breeders
- ▶ Website: www.polyploids.org
- ▶ Workshops 2021 – 2024 (records available)



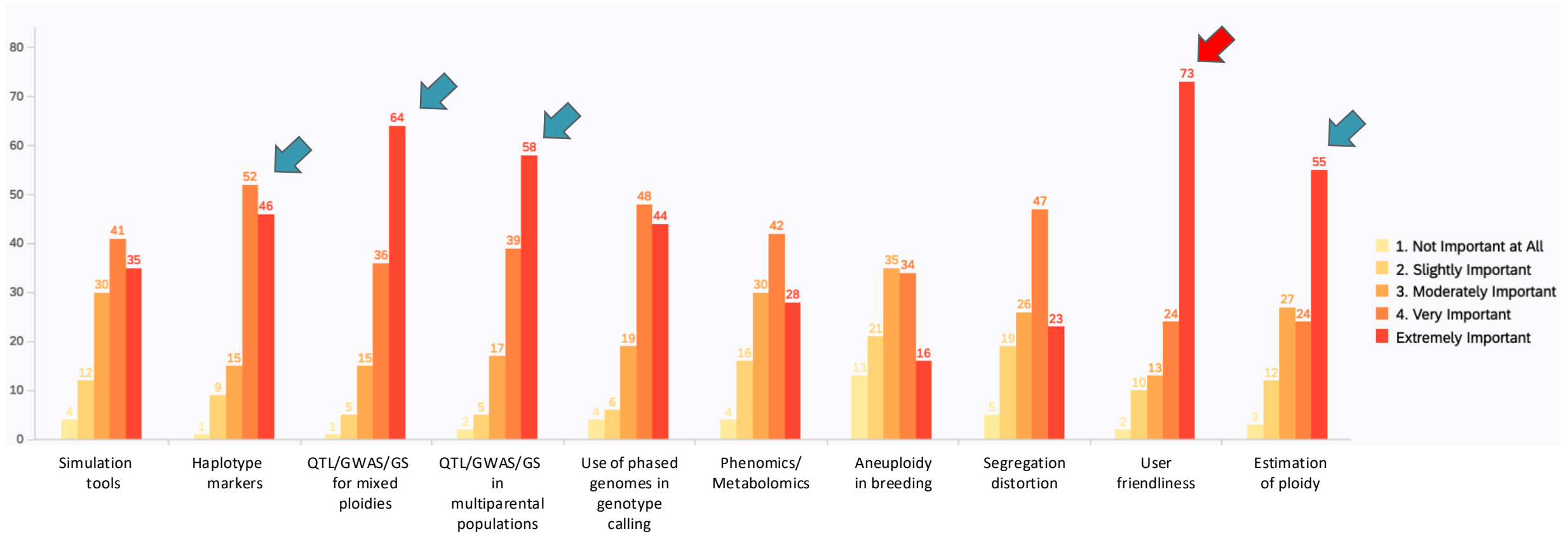
Genetic Analysis Tools for Polyploid



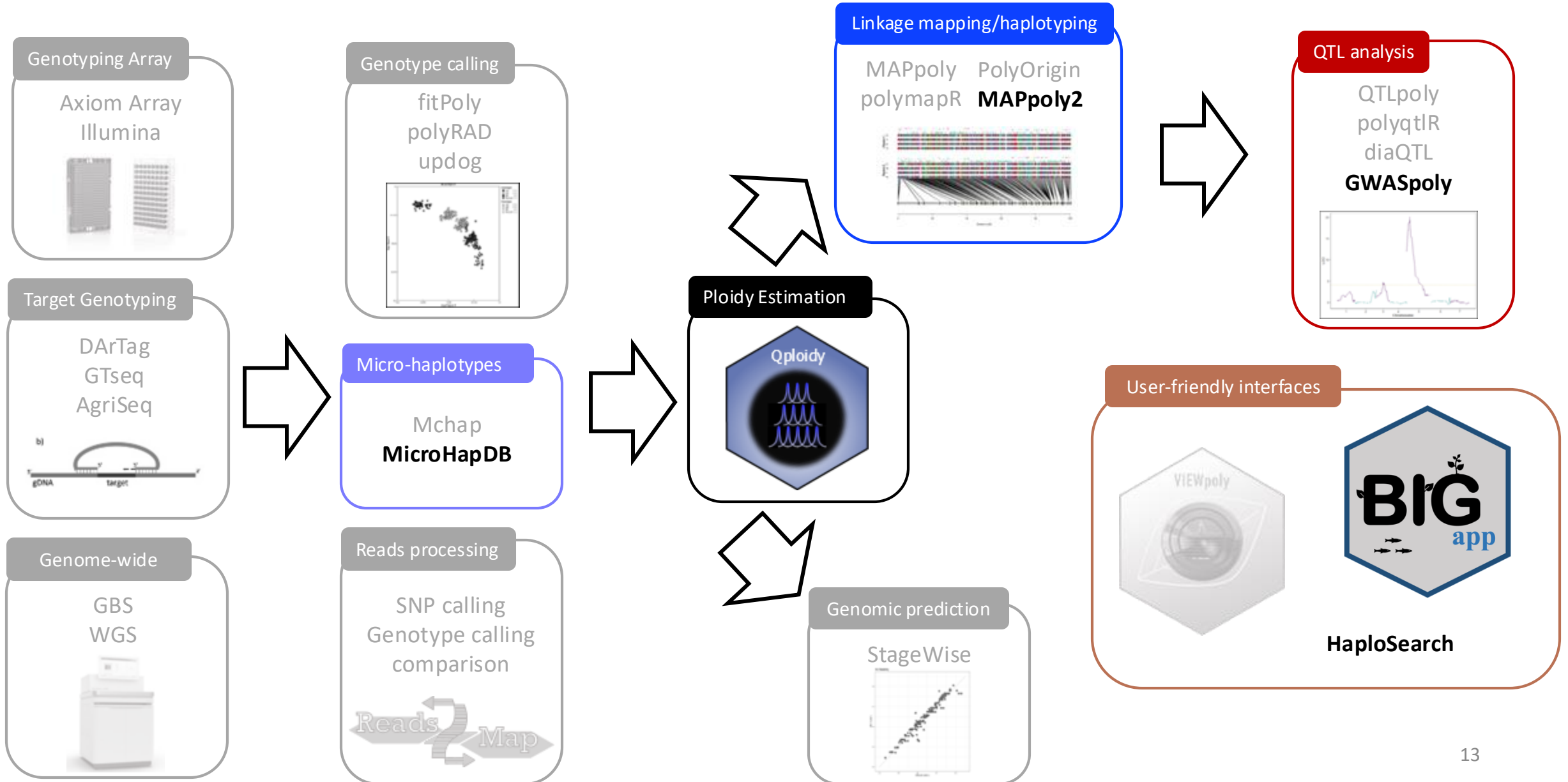
Tools for Polyploids Project Survey



David Byrne
Texas A&M University



Genetic Analysis Tools for Polyploid



Ploidy Estimation

- Requisite for all downstream analysis
- Texas A&M postdoc
 - Roses
 - Axiom Array data
 - 89% agreement
 - 0.83 Cohen's Kappa coefficient
 - MS submitted!
 - Package and tutorial already available!



[Cristianetaniguti/Qploidy](https://github.com/Cristianetaniguti/Qploidy)



TEXAS A&M
UNIVERSITY



Oscar Riera-Lizarazu



Tessa Hocchaus



Jeekin Lau

		Based on past reports		
Based on Qploidy	Ploidy level	2x	3x	4x
	2x	78	1	0
	3x	4	27	0
	4x	2	15	90
	5x	0	0	1

Input File Formats

Sequencing

VCF
(w/allele depth)

Array

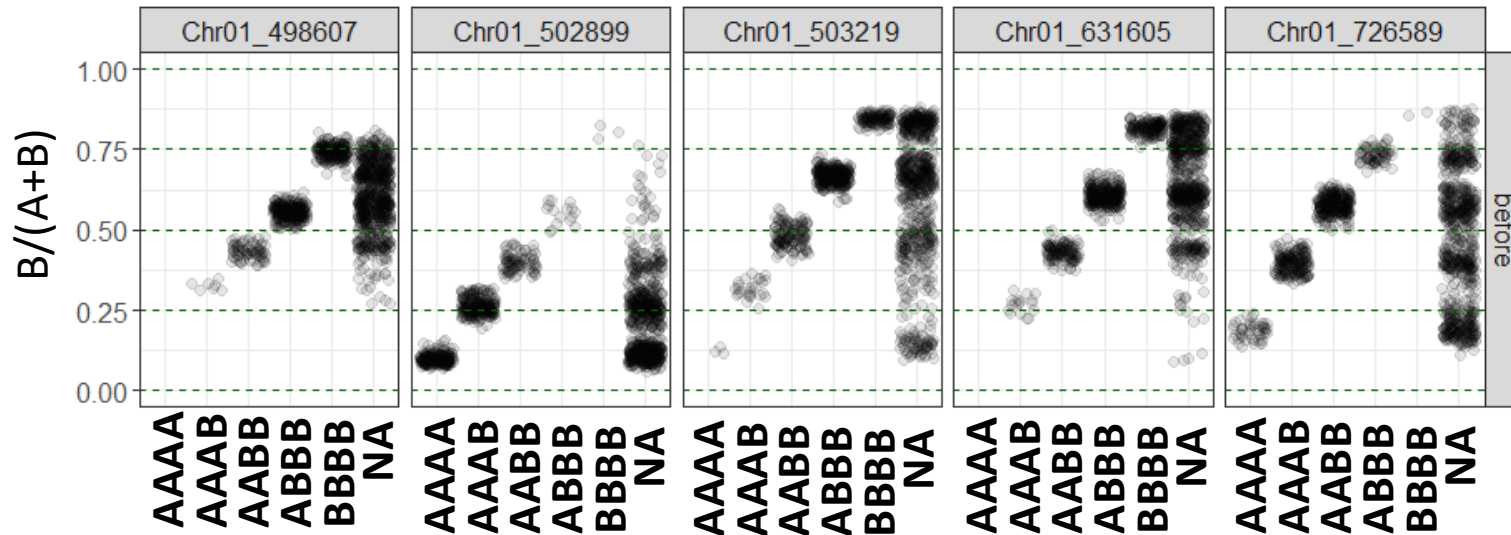
Illumina
Axiom

Qploidy Method

- B allele frequency = $B/(A+B)$

1 marker: Total depth/intensity = 100; ideal ratios:

AAAA	AAAB	AABB	ABBB	BBBB
A = 100	A = 75	A = 50	A = 25	A = 0
B = 0	B = 25	B = 50	B = 75	B = 100
Ratio = 0	Ratio = 0.25	Ratio = 0.5	Ratio = 0.75	Ratio = 1

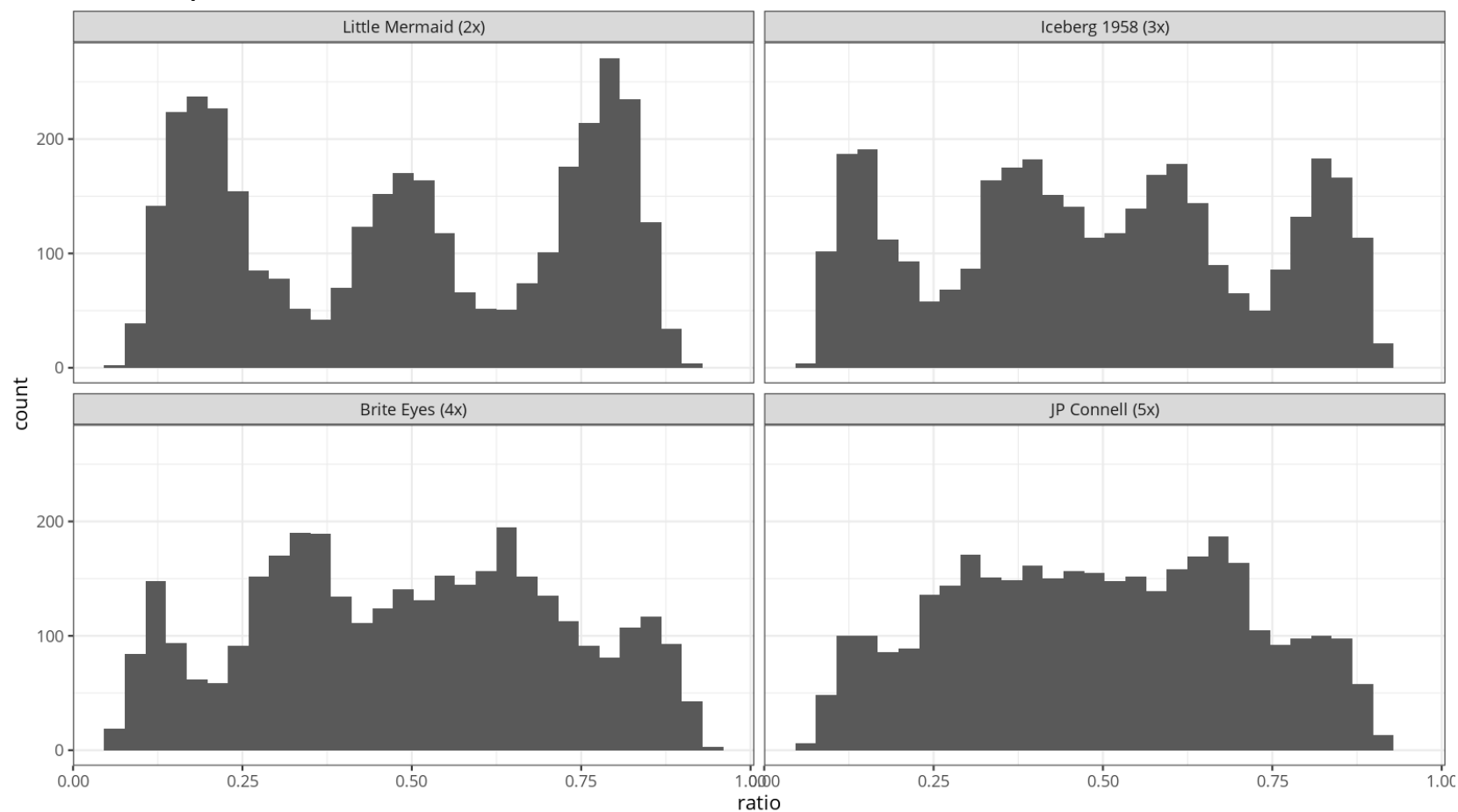


Reality for Arrays and Target Sequencing:

- High variation in marker-to-marker ratio
- Low variation in sample-to-sample ratio

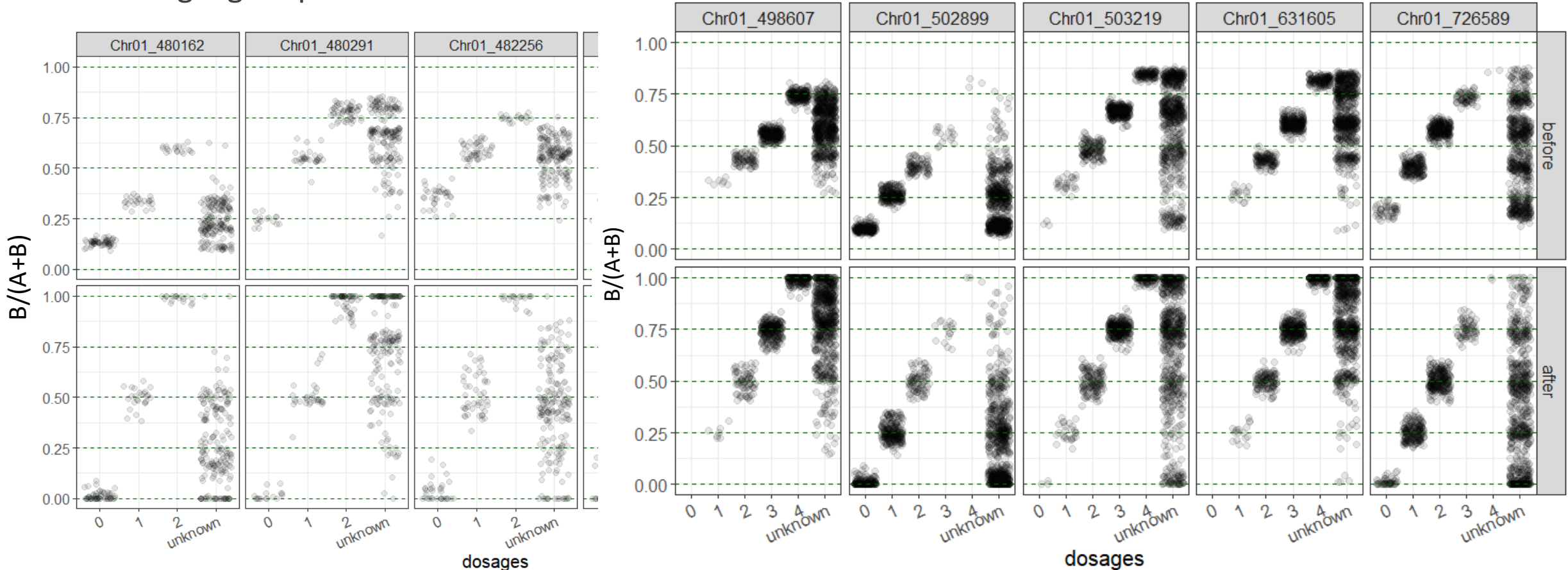
Sample-level ploidy determination

4 samples and all markers in chromosome



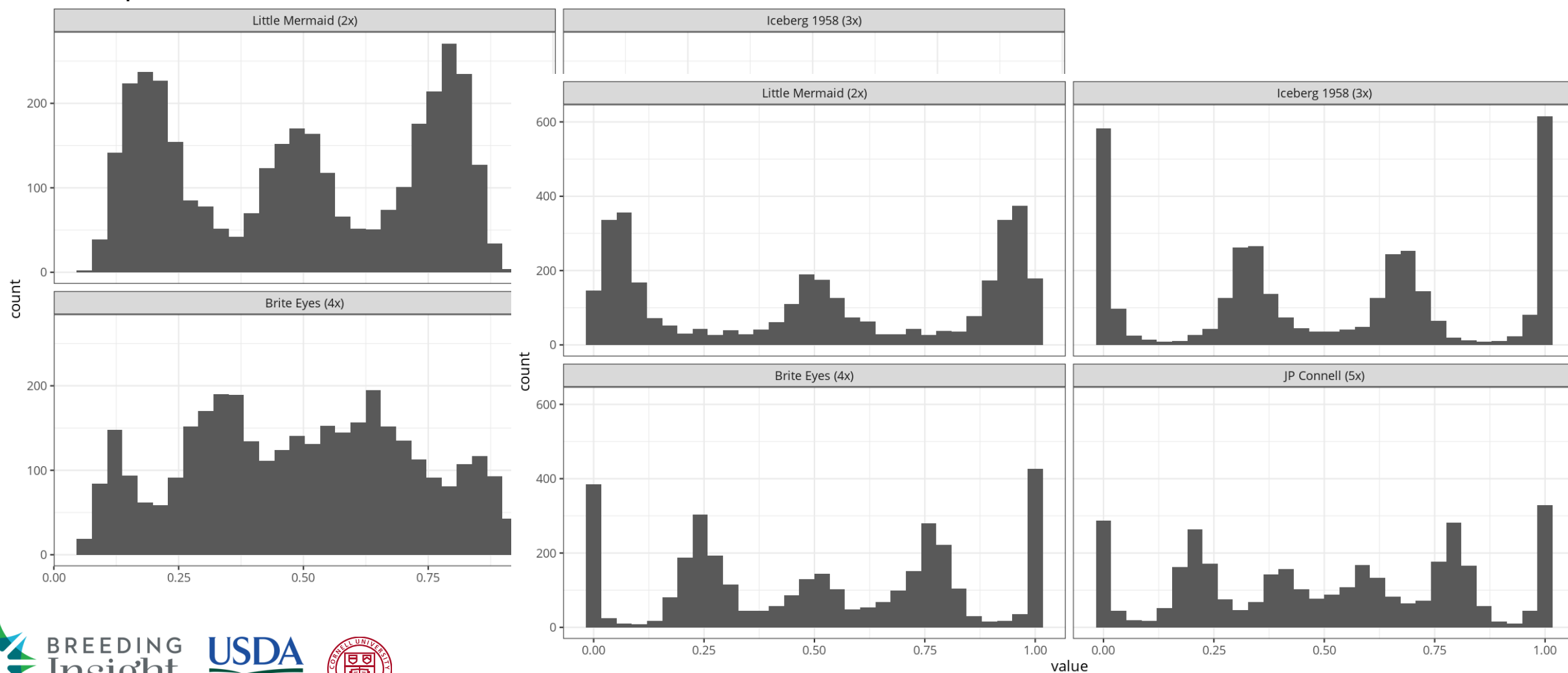
Human Genetics

- CNV studies in humans: PennCNV, QuantiSNP, Birdsuite, ...
- They always have diploid samples as reference
- Qploidy is the expansion of the PennCNV method (Wang et al., 2007)
- Using higher ploidies as reference



Sample-level ploidy determination

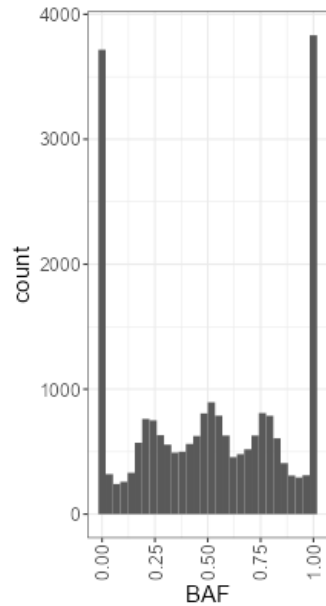
4 samples and all markers



The improvement – chromosomal-level estimations

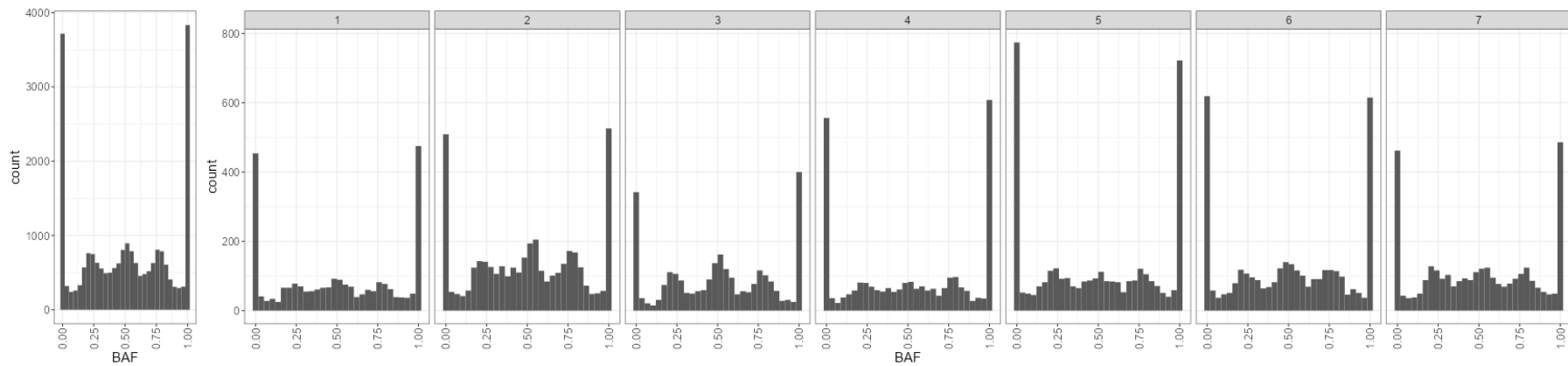
90_MBxBE (4x)

Diploids as reference

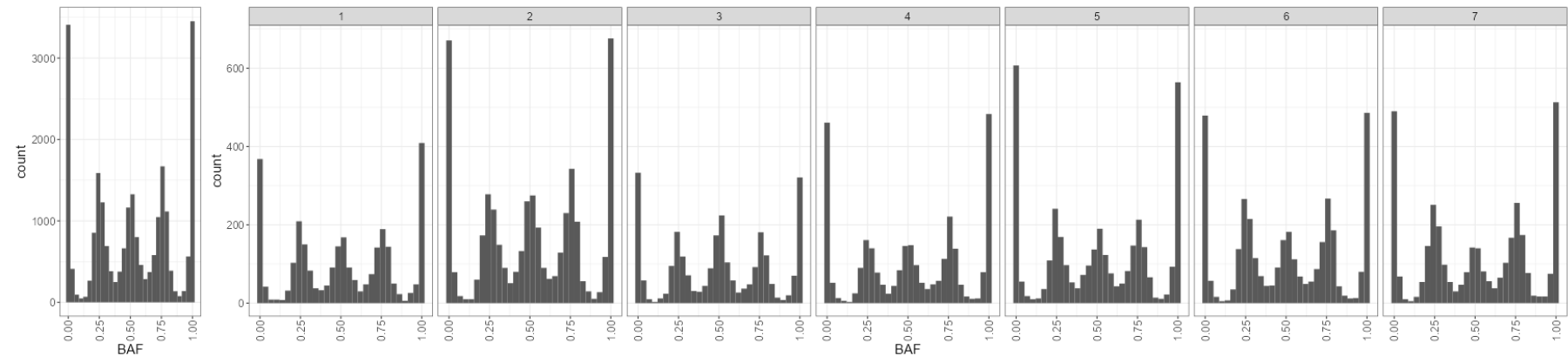


The improvement – chromosomal-level estimations

Diploids as reference



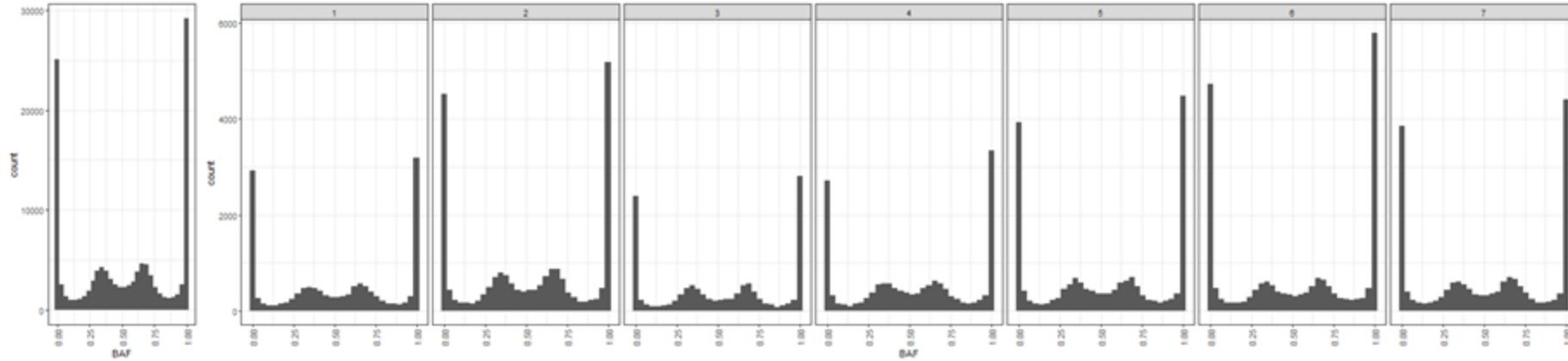
Tetraploids as reference



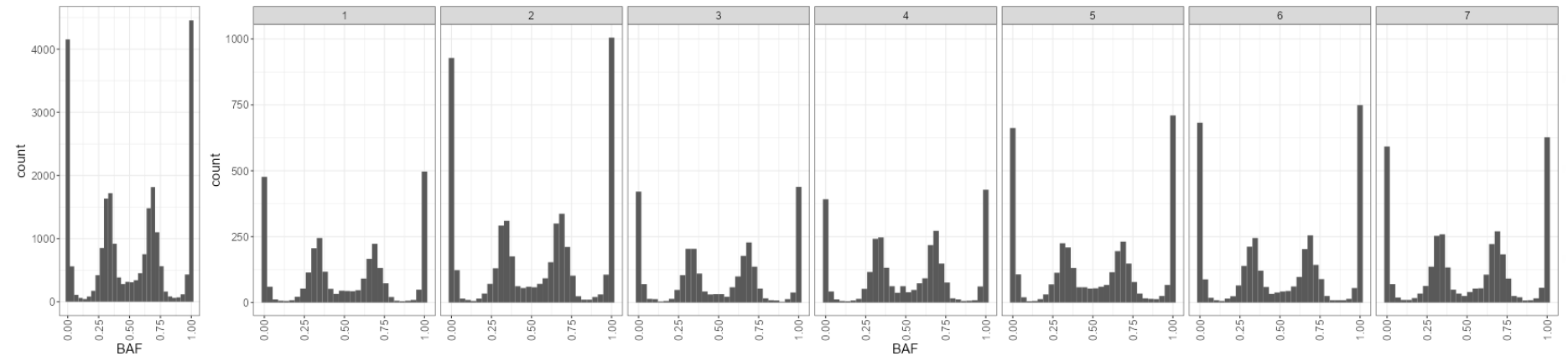
The improvement – chromosomal-level estimations

A triploid sample using diploids as reference

Sally Holmes (3x)

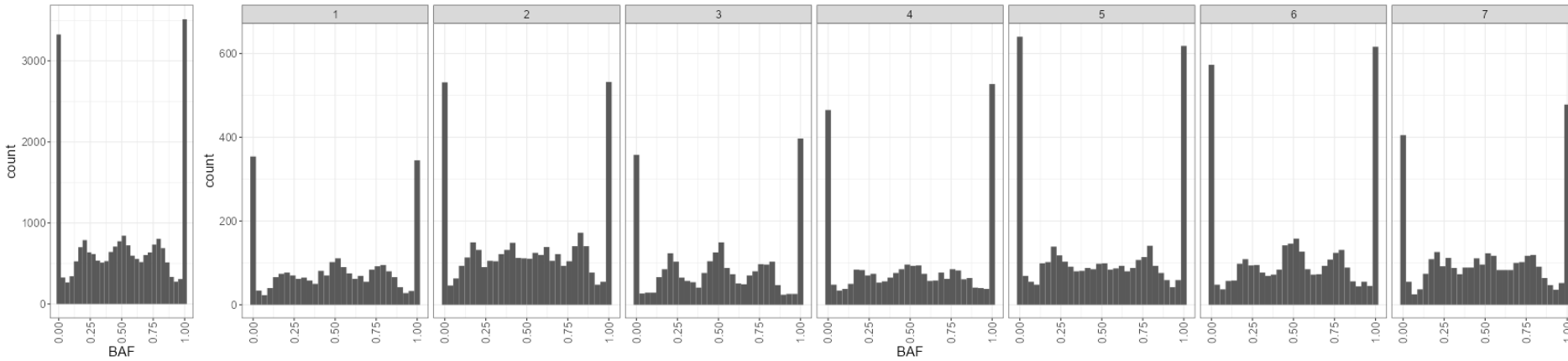


Tetraploids as reference



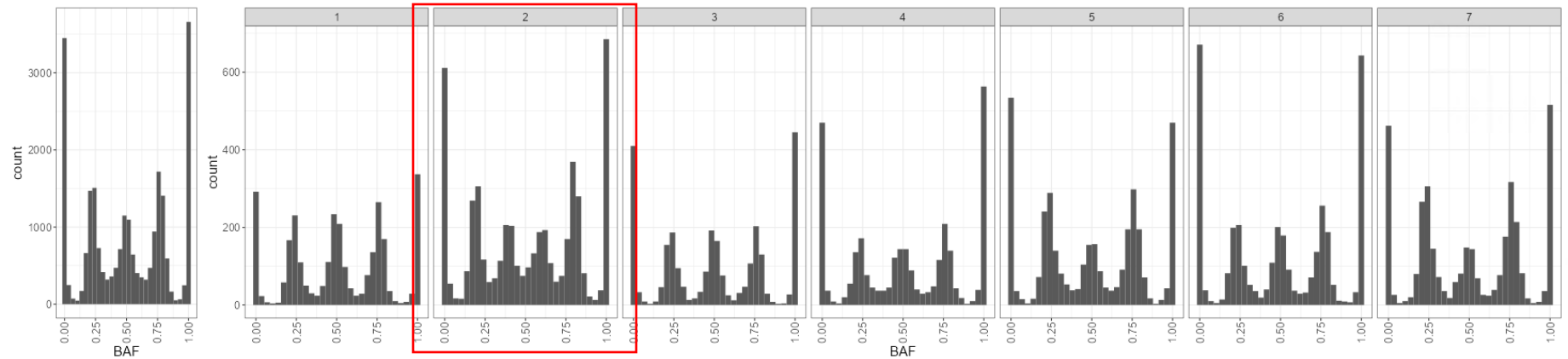
The improvement – chromosomal-level estimations

Diploids as reference

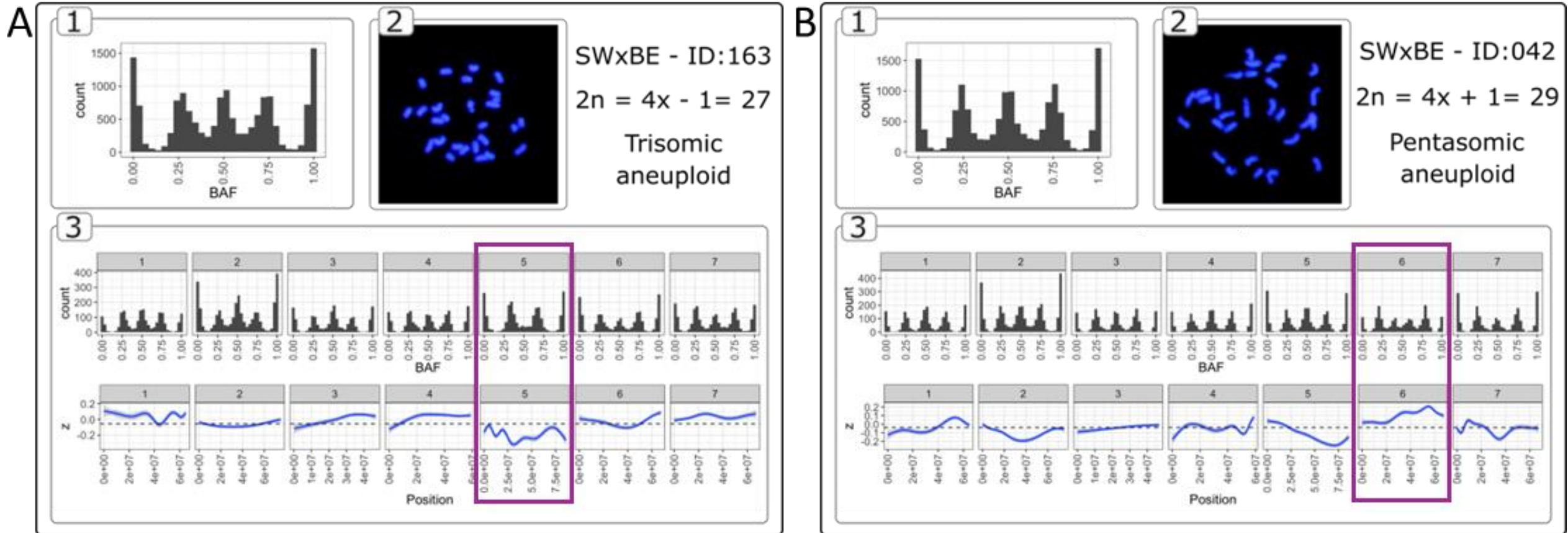


16401_N025 (4x + 1)

Tetraploids as reference

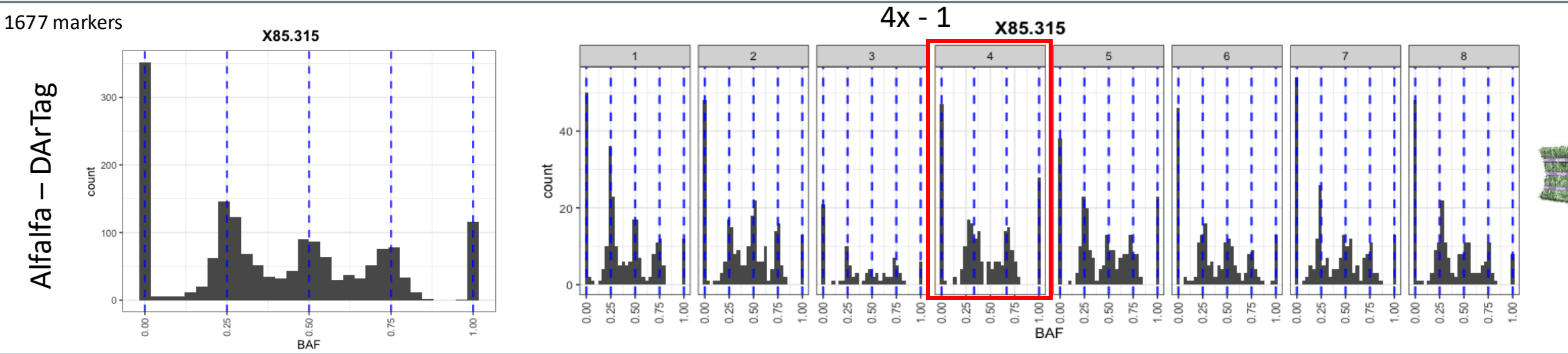
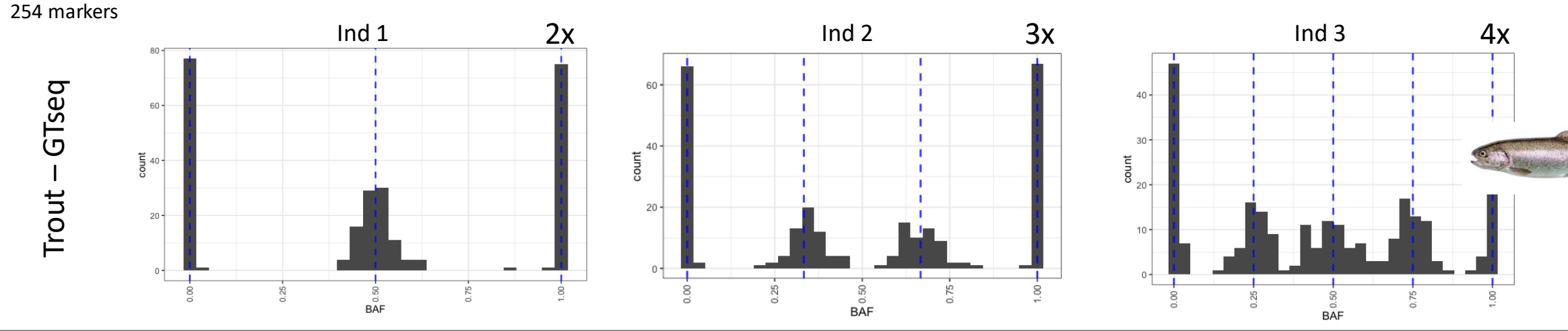


Z Score of the Total Allele Intensity/Depth



Other Species and Technologies

--- Expected peak position



When to use

When does Qploidy work?

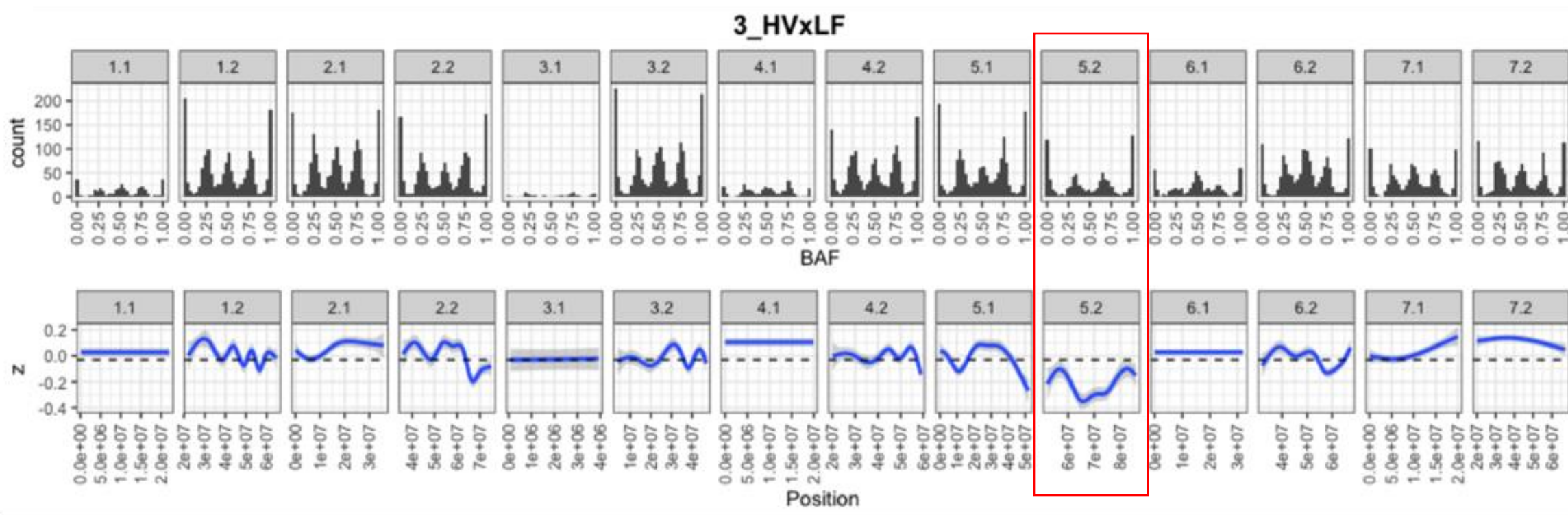
- Marker data derived from **Array** or **Target sequencing** platforms
- All DNA samples prepared following **same library preparation**
- **Known ploidy** of at least 60 samples or **known most common ploidy**
- Heterozygous
- Sample collected **represents whole individual**

When Qploidy does NOT work?

- RADseq or GBS libraries
- Combined datasets from genotyping in **different batches**
- **Do not have a subset of samples with known ploidy** or **lack a predominant ploidy**
- Samples consist of **inbred lines**
- Sample is possible from **chimeric or mosaic** tissue

What is next for Qploidy?

- Integrate Z score with BAF (HMM, deep learning?)
- Locus-level resolution



Linkage Mapping

- Outcrossing populations
- Autopolyploids
- Interconnected full-sib populations
- Combination of ploidies
- Speed optimization
- Documentation
- Prepared to be implemented in an interface



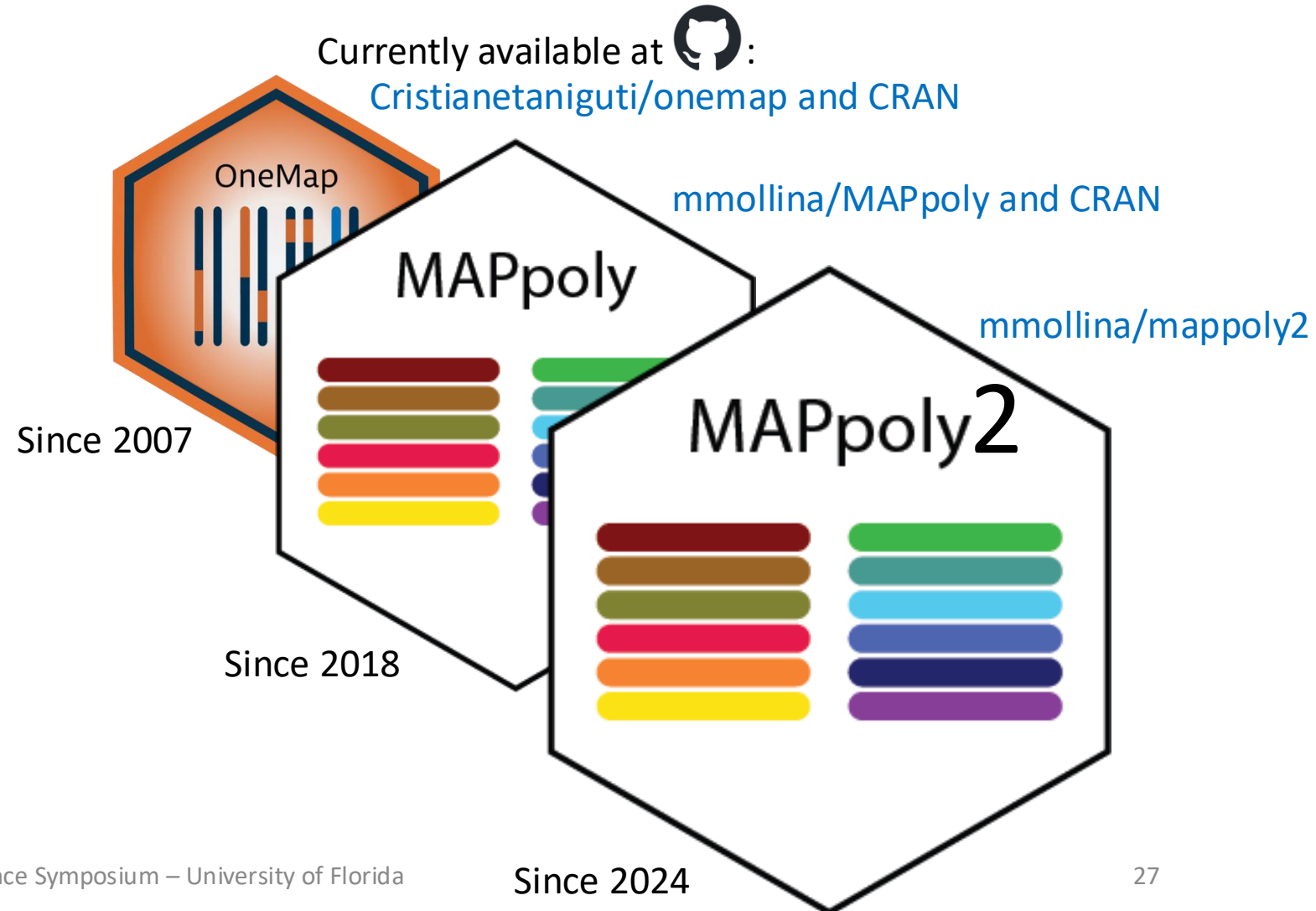
Augusto Garcia
University of São Paulo



Marcelo Mollinari
NCState University



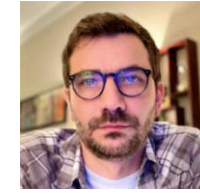
Gabriel Gesteira
NCState University



Linkage Mapping



Augusto Garcia
University of São Paulo

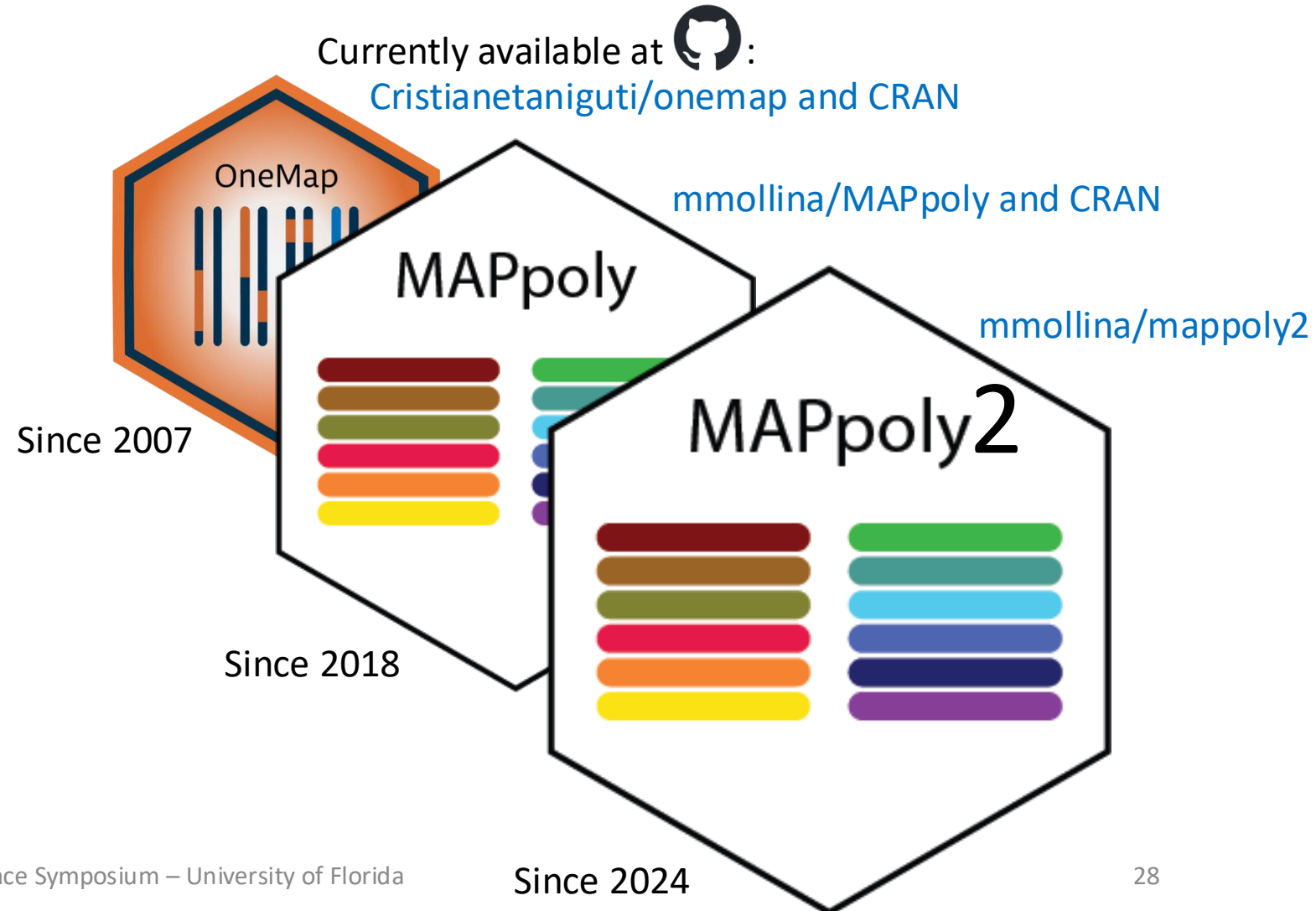


Marcelo Mollinari
NCState University



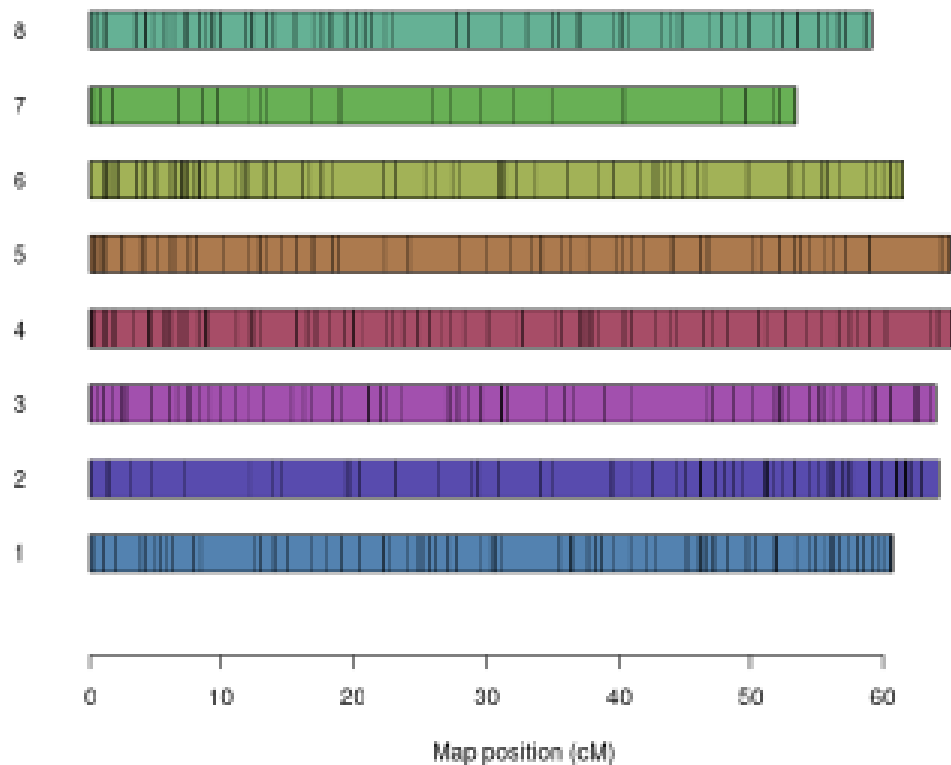
Gabriel Gesteira
NCState University

- OneMap: no more new features since 3.2.0
 - Still the best option for:
 - Recombinant Inbred Lines
 - Dominant and multiallelic markers
- OneMap and MAPpoly are stable
- Development efforts are now in MAPpoly2



Linkage Mapping - Alfalfa

Alfalfa genetic map



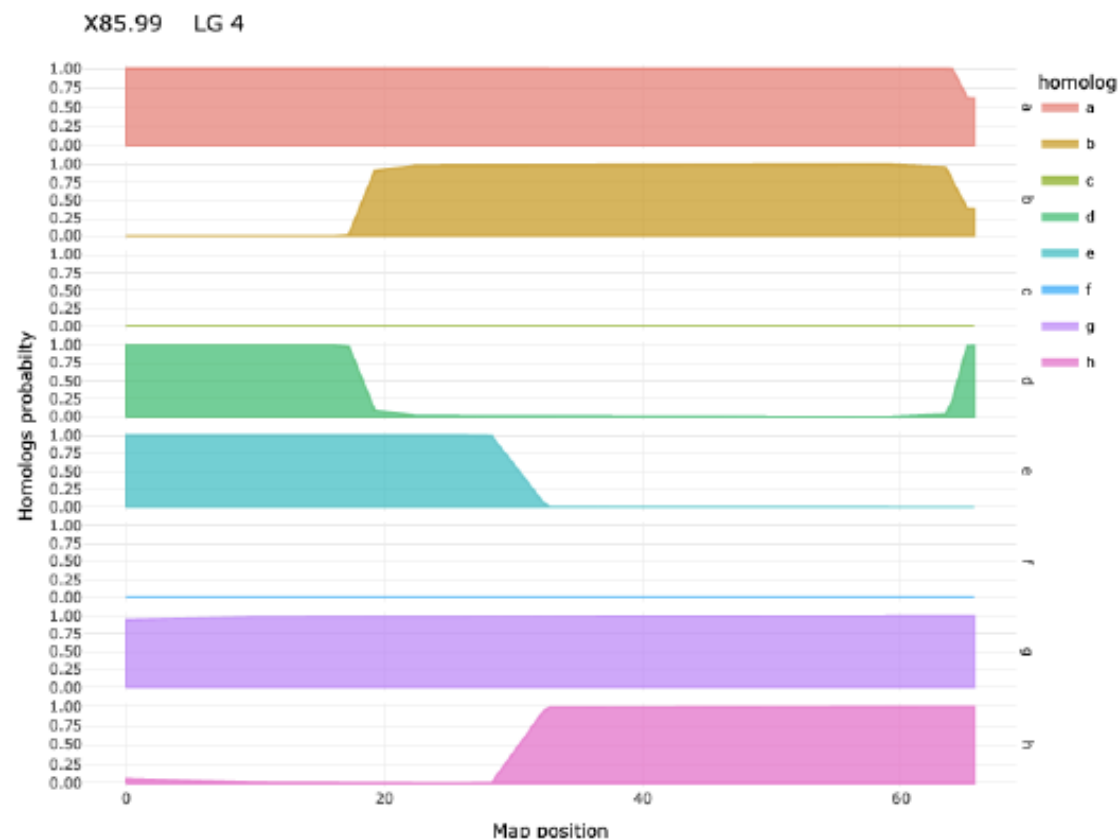
Summary

LG	Map length (cM)	Markers/cM	Simplex	Double-simplex	Multiplex	Total	Max gap
1	60.69	2.39	37	27	81	145	4.53
2	64.12	1.89	39	24	58	121	5.01
3	63.89	1.77	35	23	55	113	5.81
4	65.66	2.15	51	22	68	141	3.23
5	65.14	1.58	40	17	46	103	5.32
6	61.58	2.24	47	32	59	138	3.86
7	53.39	0.75	15	7	18	40	7.49
8	59.18	2.23	46	19	67	132	4.86
Total	493.65	1.88	310	171	452	933	5.01

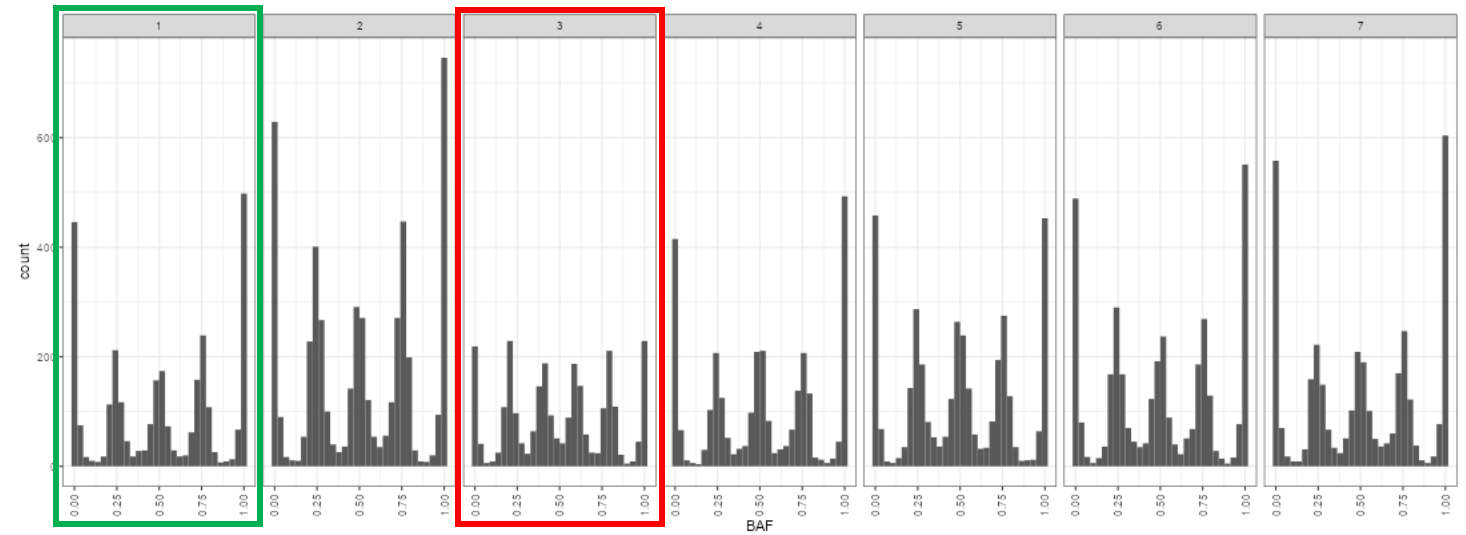
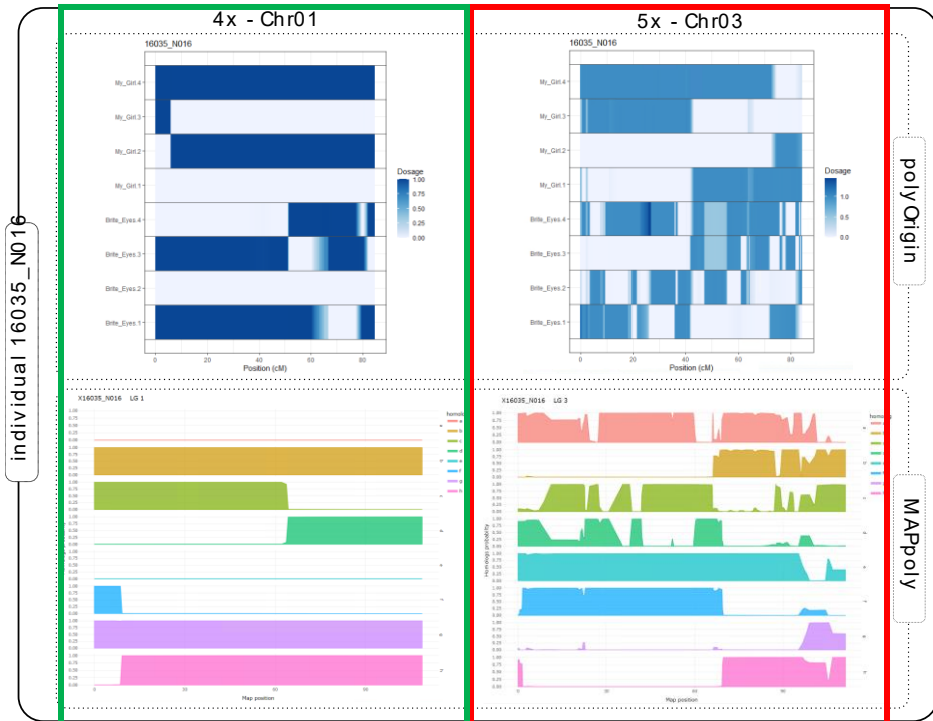
Based on bi-parental F1 population of parents susceptible (I195) and resistant (J432) to *Aphanomyces euteiches*

Haplotype Reconstruction

Example of haplotype reconstruction in individual 85_99, LG4



Aneuploids and Linkage Maps

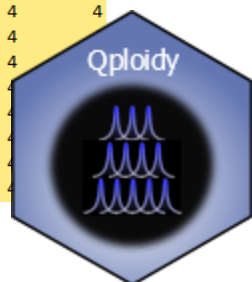


Impact on Linkage Maps

filter_aneuploid(
mappoly.obj,
Qploidy.out.csv)



	A	B	C	D	E	F	G	H
1	samples	Chr1	Chr2	Chr3	Chr4	Chr5	Chr6	Chr7
2	MBxGV_1	3	4	4	4	4	4	4
3	MBxGV_28	4	4	3	4	4	4	4
4	MBxGV_76	4	4	3	4	5	4	4
5	MBxGV_16	4	4	4	3	4	4	4
6	MBxGV_20	4	4	4	5	4	4	4
7	MBxGV_6	4	4	4	5	4	4	4
8	MBxGV_103	4	5	4	4	4	4	4
9	MBxGV_44	4	5	4	4	4	4	4
10	MBxGV_52	4	5	4	4	4	4	4
11	MBxGV_10	4	4	4	4	4	4	4
12	MBxGV_10.1	4	4	4	4	4	4	4
13	MBxGV_100	4	4	4	4	4	4	4
14	MBxGV_101	4	4	4	4	4	4	4
15	MBxGV_102	4	4	4	4	4	4	4
16	MBxGV_105	4	4	4	4	4	4	4



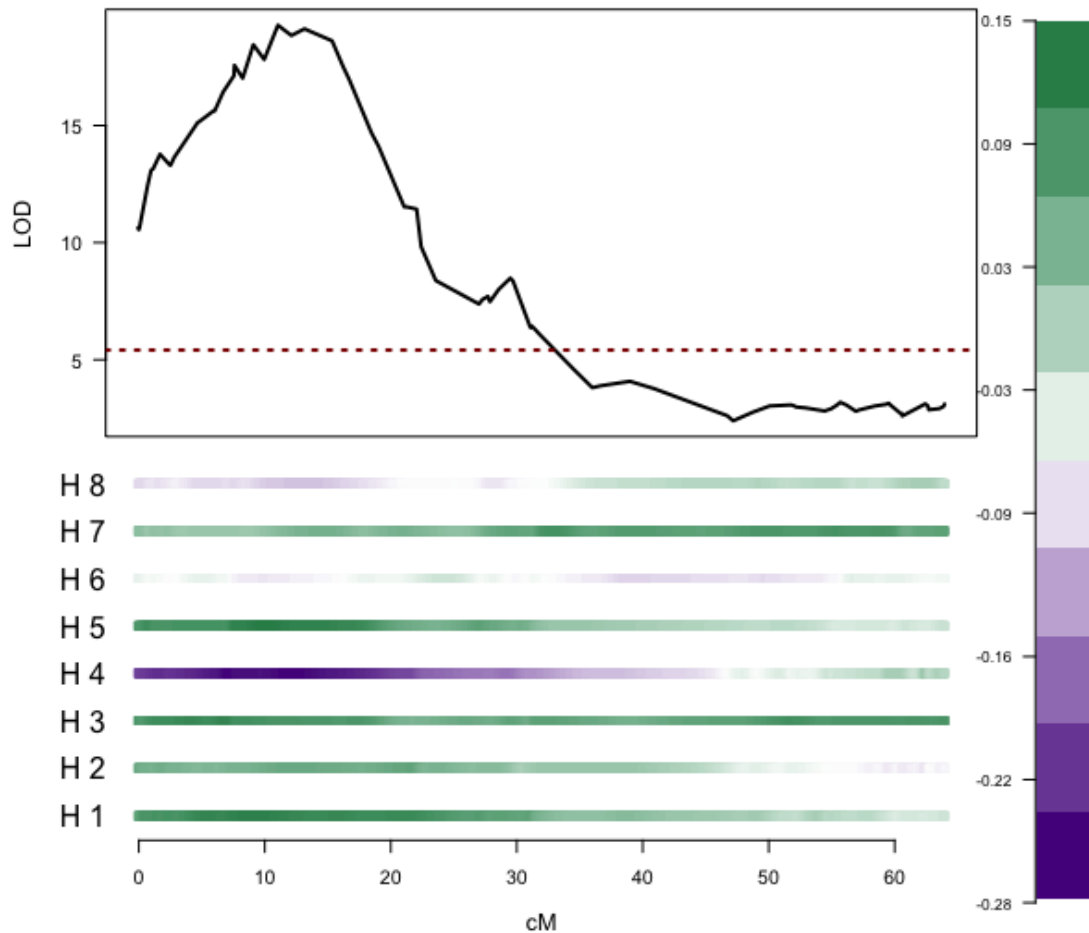
Population ID	# of samples	Female parent	Male parent	Aneuploid individuals	Aneuploid chromosomes
MBxBE	95	Morden Blush	Brite Eyes	18 (18.94 %)	2.3 %
HVxLF	91	High Voltage	Lemon Fizz	13 (14.28 %)	1.9 %
MBxRR	94	Morden Blush	Ramblin Red	14 (14.89 %)	2.0 %
MBxGV	94	Morden Blush	George Vancouver	9 (9.57 %)	1.3 %
BExMG	121	Brite Eyes	My Girl	20 (16.53 %)	2.6 %
SWxBE	208	Stormy Weather	Brite Eyes	51 (24.52 %)	3.3 %
BExMG	66	My Girl	Brite Eyes	7 (10.6 %)	1.5 %



BREEDING
Insight



Alfalfa - Aphanomyces Resistance QTL Analysis



LG	Position (cM)	LOD	PVE (BLUE*)	PVE (SMA**)
3	11.04	19.28	38.61	16.13

PVE - Percentage of variation explained

*BLUE: Best linear unbiased estimates

** Single marker analysis

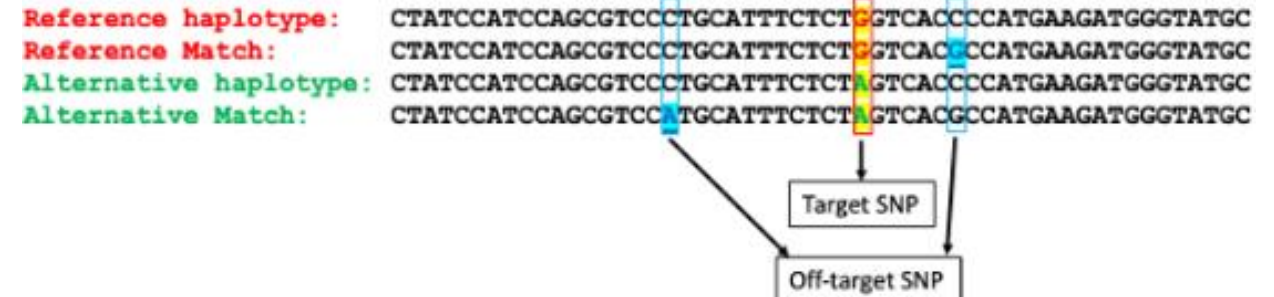
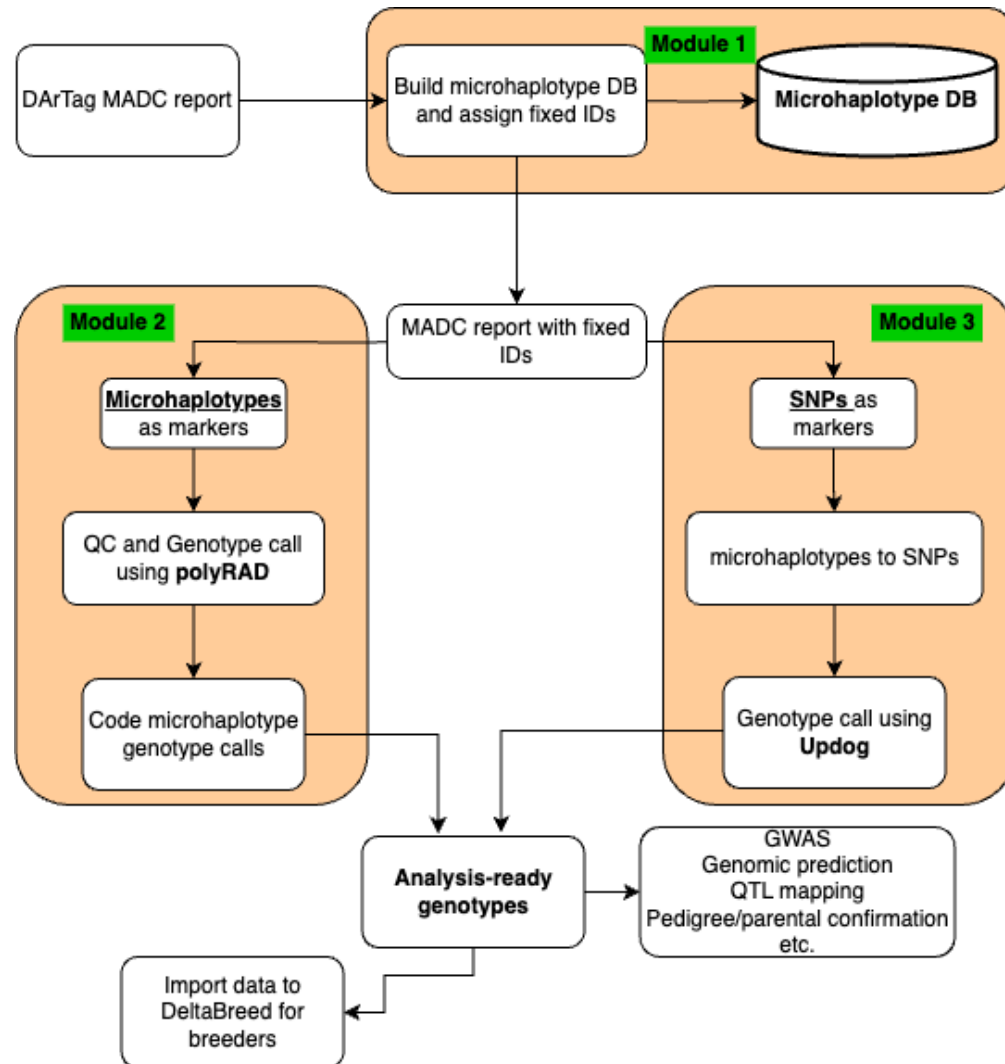
- The alleles originating from homologues 1, 2, 3 of parent 1 and homologues 5 and 7 of parent 2 contributes positively to the trait.
- The most appropriate model is oooQ x oQQQ.

Limited to Mapping Populations Designs

Development of Microhaplotype Database



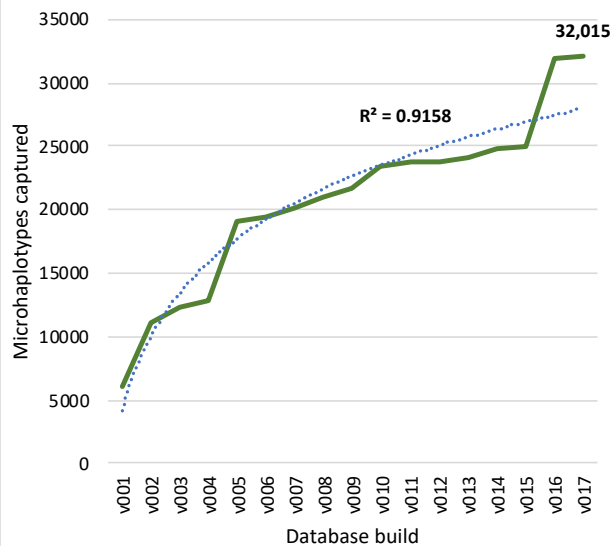
DONGYAN ZHAO
Genomics Lead



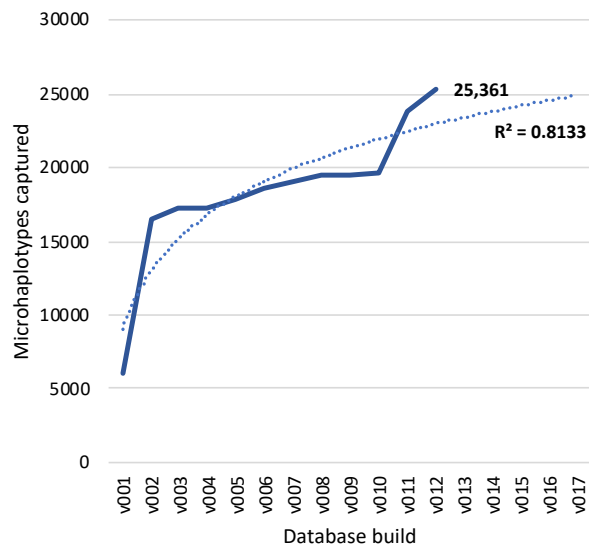
- Establish best workflow for filtering and processing genotype data
- Create file formats that work with other applications for downstream analysis

Microhaplotype Database allele capture curves

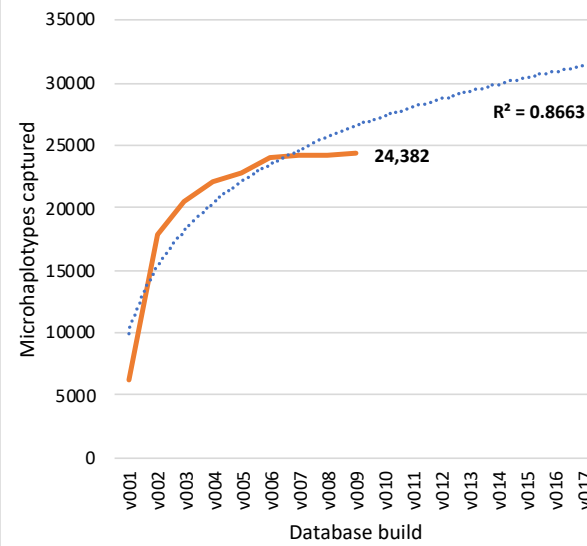
Alfalfa (14,194 samples)



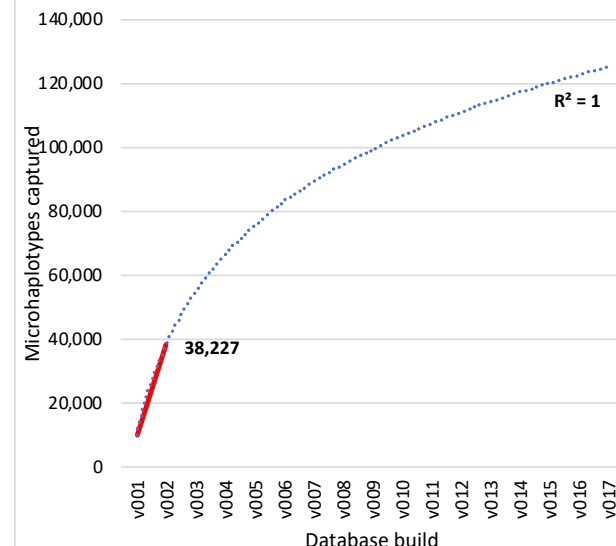
Blueberry (8742 samples)



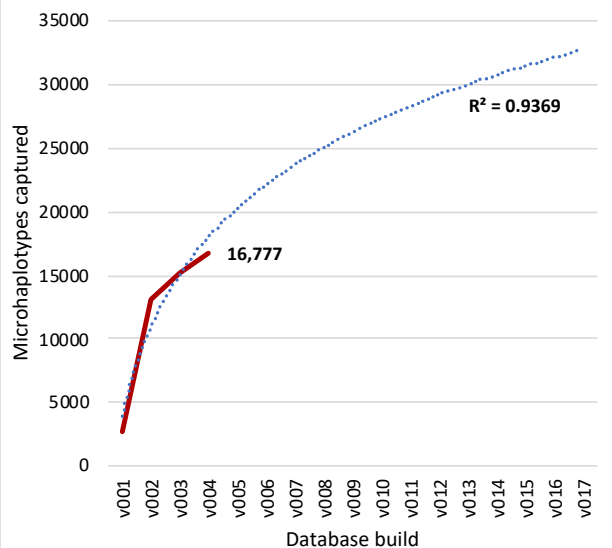
Sweetpotato (6392 samples)



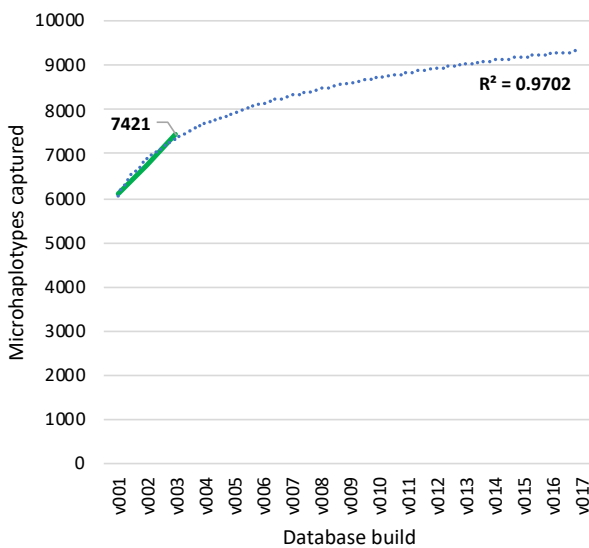
Strawberry (1880 samples)



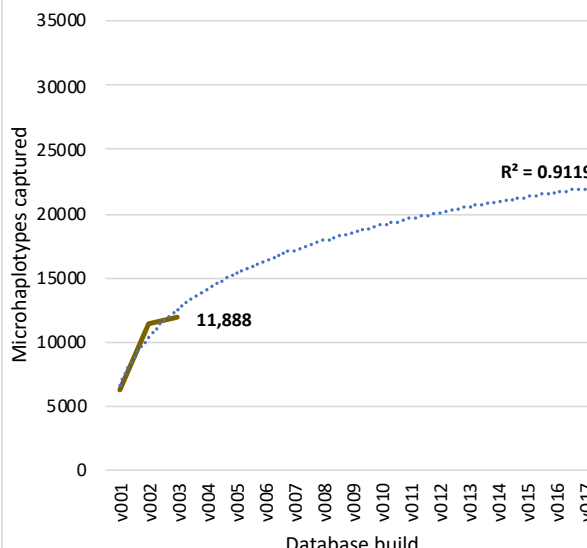
Cranberry (1880 samples)



Cucumber (1880 samples)

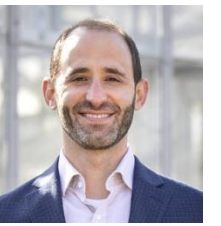


Pecan (940 samples)

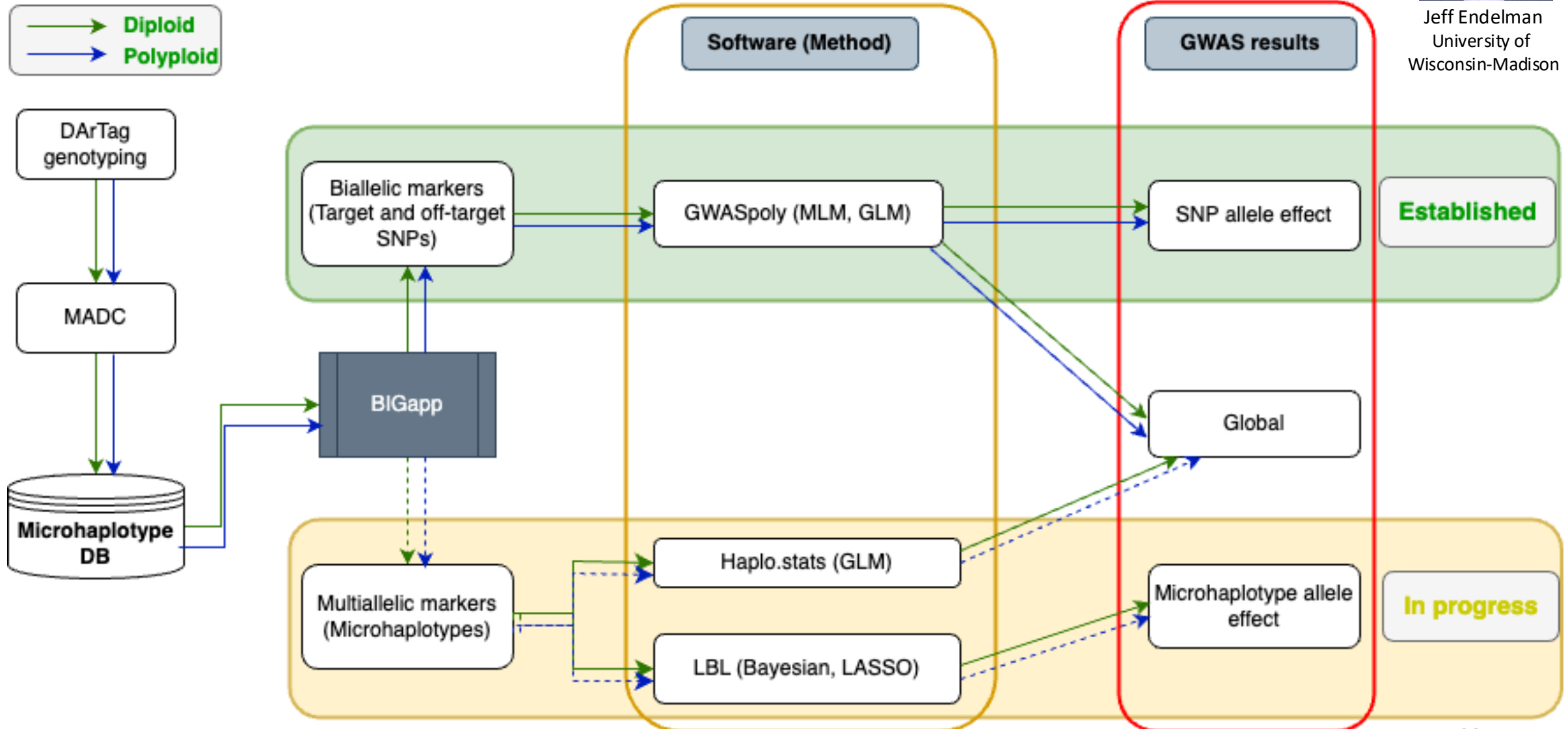


- 156,076 alleles across 7 species
- Allele capture curves are expected to follow a diminishing returns pattern. The slope is very steep in early builds but begins to asymptote in later builds as fewer new alleles are discovered.
- Databases are considered mature when they asymptote.

Multiallelic Genotypes for GWAS



Jeff Endelman
University of
Wisconsin-Madison



HaploSearch: user interface to search for microhaplotypes



TYLER SLONECKI

Biosoftware Coordinator



MENG LIN

Bioinformatics
Coordinator

Welcome ▾

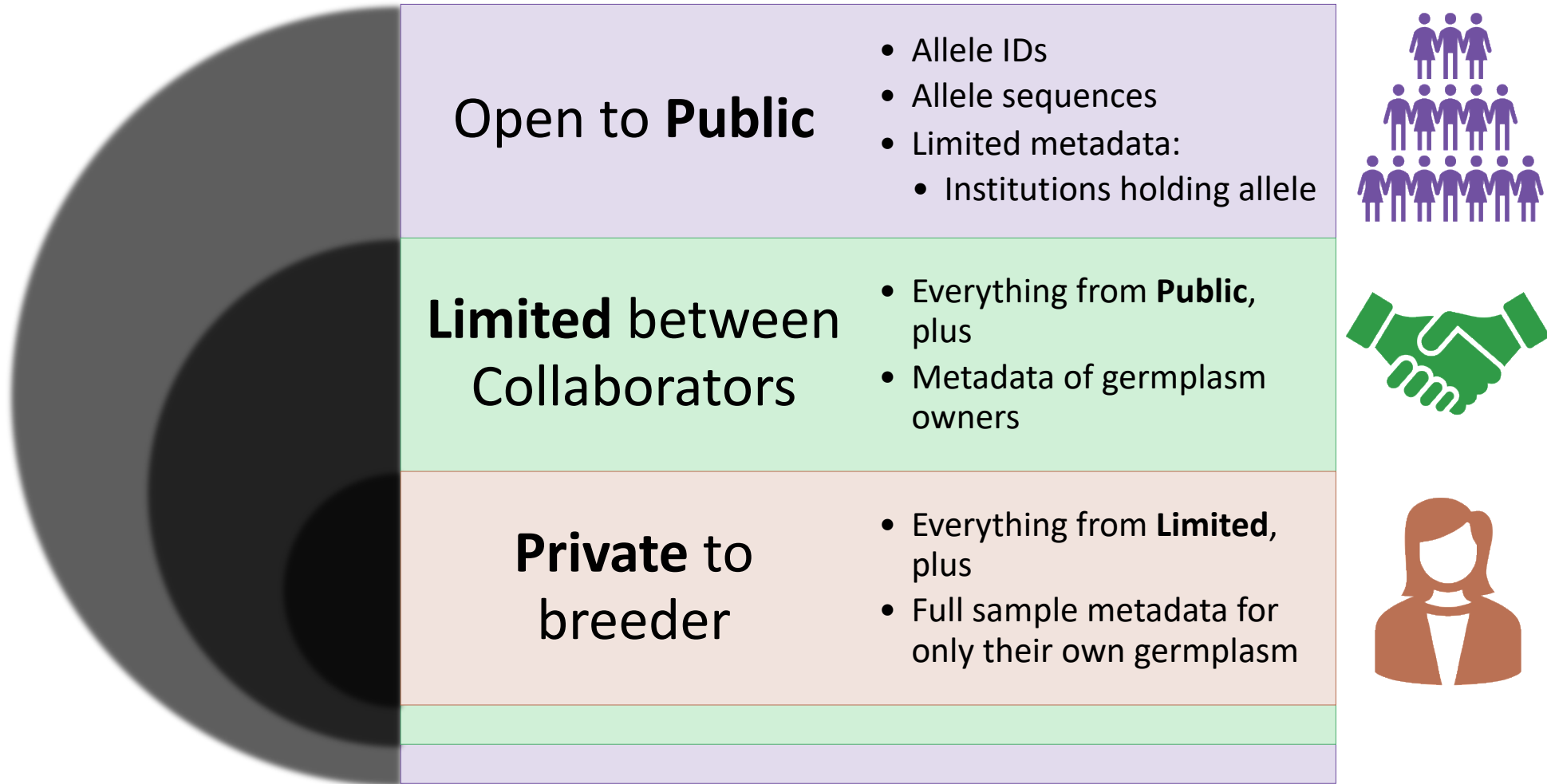
Home
Database Report
Admin
Query

Query Unique Microhaplotypes

Species: Alfalfa ▾
Keyword Search
→ View Details

	Allele ID	Info	Associated Trait	Allele Sequence
<input type="checkbox"/>	<input type="text" value="Search Allele ID"/> ▾	<input type="text" value="Search Info"/> ▾	<input type="text" value="Search Associated T"/> ▾	<input type="text" value="Search Allele Sequence"/> ▾
<input type="checkbox"/>	chr1.1_000194324 Ref_0001			CGAAATAATAACCCAAGTCTGCCAGTTTATGTTAAACTTTTCTTACATGGTACAAGTTCGGTGACAACCTAACAAAGTAA
<input type="checkbox"/>	chr1.1_000194324 Alt_0002			CGAAATAATAACCCAAGTCTGCCAGTTTATGTTAAACTTTTCTTACAAGGTACAAGTTCGGTGACAACCTAACAAAGTAA
<input type="checkbox"/>	chr1.1_000194324 AltMatch_0001			CGAAATAATAACCCAAGTCTGCCAGTTTATGTTAAACTTTTCTTACAAGGTACAAGTTCGGTGACAAC
<input checked="" type="checkbox"/>	chr1.1_000309952 Ref_0001			ATCTCGACCACTTATAAACACATATTCATATCATATGTA
<input type="checkbox"/>	chr1.1_000309952 Alt_0002		drought resistance	ATCTTGACCACTTATAAACACATATTCATATCATATGTA
<input type="checkbox"/>	chr1.1_000309952 RefMatch_0001	likely paralogous		ATCTCGACCACTTATAAACACATATTC
<input type="checkbox"/>	chr1.1_000309952 AltMatch_0001			ATCTTGACCACTTATAAACACATATTC
<input checked="" type="checkbox"/>	chr1.1_000452961 Ref_0001			TTCTCTTTTACCCGACCGGCTCCCTGCAGACCAGAAAGCCCAA
<input type="checkbox"/>	chr1.1_000452961 Alt_0002			TTACGAGATCGCGAAGTTCGTTCTTTCTTTATCTTTCTCTTTTACCCGACCGGCTCCCTGCAGACCAGAAAGCCCAA
<input type="checkbox"/>	chr1.1_000532584 Ref_0001		anthracnose race1 resistance	CAACGGAACATATAAGATATCCACTTCTCTTGGAGCTTGATGATACTTATAATGTTGGGGATTGTGTTTGCAGGATT
<input type="checkbox"/>	chr1.1_000532584 Alt_0002			CAACGGAACATATAAGATATCCACTTCTCTTGGAGCTTGATGATACTTATAATGTTGGGGATTGTGTTTGCAGGATT
<input type="checkbox"/>	chr1.1_000532584 RefMatch_0001	likely paralogous		CAACGGAACATATAAGATATCCACTTCTCTTGGAGCTTGATAACTTATAATGTTGGGGATTGTGTT
<input type="checkbox"/>	chr1.1_000532584 RefMatch_0002			CAACGGAACATATAAGATATCCACTTCTCTTGGGGCTTGATAACTTATAATGTTGGGGATTGTGTT

Data sharing model



Introducing BIGapp

BIGapp is a web-based application to make BI genomics and bioinformatics analyses

BIGapp's Key Design Principles:

1. Accessibility:

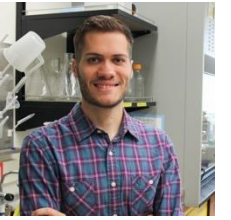
- Intuitive interface; no coding experience needed.

2. Integrated Help:

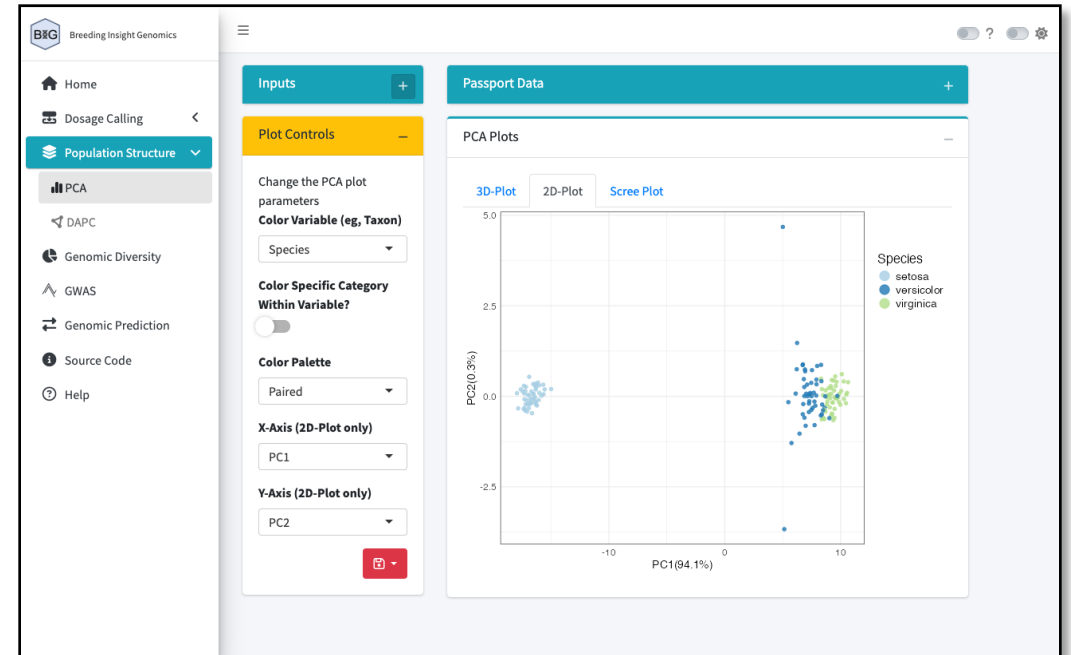
- Tutorials and help pages for guidance and result interpretation.

3. Broad Support:

- Designed for all species and ploidy levels.



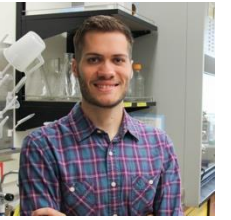
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<https://github.com/Breeding-Insight/BIGapp>

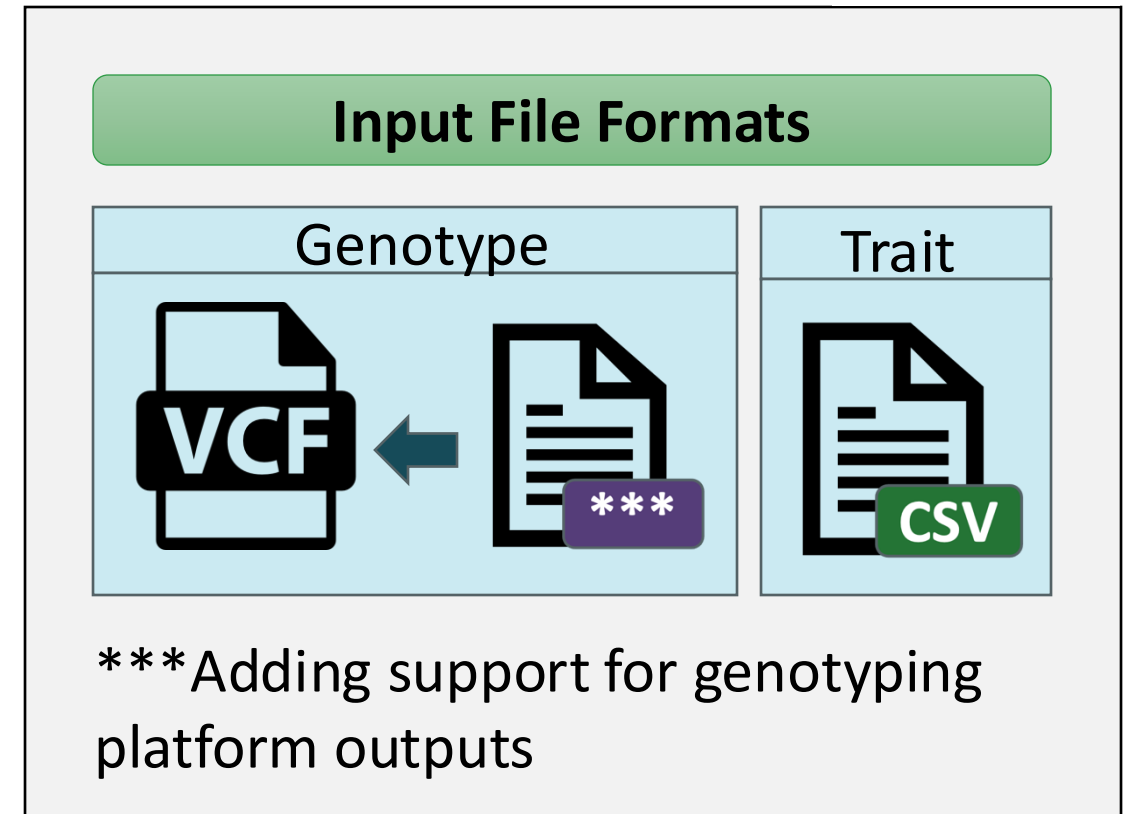
<https://github.com/Breeding-Insight/BIGr>

BIGapp v1.0 Features

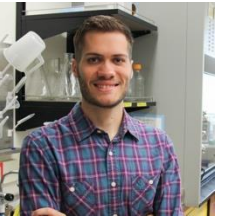


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- Genotype processing
 - Dosage call from read counts
 - SNP filtering
 - Sample filtering
- Summary metrics
 - SNP minor allele frequency
 - Sample observed heterozygosity
 - Population dosage ratios
- Population Structure
 - PCA
 - DAPC
- Genome-wide association studies (GWAS)
- Genomic Selection (GS)



Availability and Future Plans



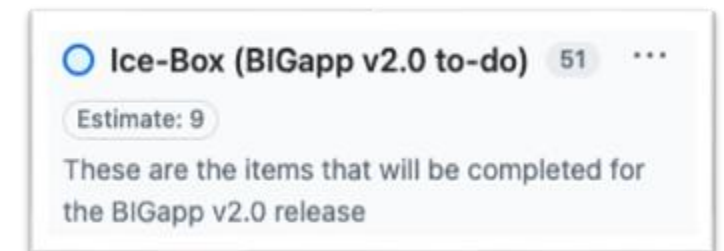
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How to get BIGapp?

- Visit <https://github.com/Breeding-Insight> and install it like an R package.

Future Plans:

1. Integrate ploidy estimation support (Qploidy)
2. Integrate linkage mapping support (MAPpoly2).
3. Integrate support for GWAS analysis with microhaplotypes



Highlights

- New Tools available:
 - Qploidy
 - MAPpoly2
 - BIGapp
- Keep an eye out for updates and releases:
 - Qploidy
 - MAPpoly2
 - BIGapp
 - MicroHaplotypeDB
 - HaploSearch
 - GWASpoly
- Shiny Power: Implementing Genomic Tools in user-friendly interfaces

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GitHub: <https://github.com/Breeding-Insight>

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Funding and Support

U.S. Department of Agriculture – Agriculture Research Service
through Cornell University