

Considerations and Strategies for Imputation Reference Panel Construction

Beef Improvement Federation

Advancements in Genomics and Genetic Prediction

June 2, 2022

Troy Rowan Ph.D.

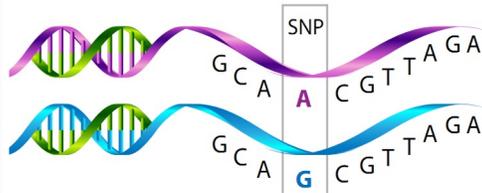
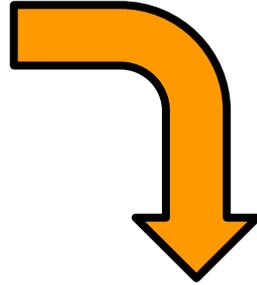
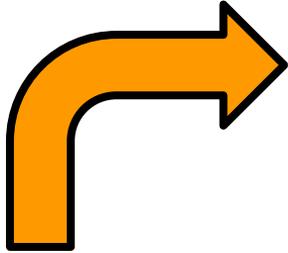
University of Tennessee

A big thank-you



Bob Schnabel

Mizzou



Ranch Tested. Rancher Trusted.
Red Angus
 Red Angus Association of America
 Genomic Data Report

Reg#: _____ Prefix: _____ ID: _____ Birthdate: 3/15/2012 Sex: C
 Name: _____
 Genomic Test Performed: GBL/HD50K Angus/O1 Job Number: _____ Report Date: 12/7/2012

Current (With PF50K Genomic Data)	CED	BW	WW	YW	Mik	TM	ME	HPG	CEM	STAY	MARB	YG	CW	REA	FAT
EPD	-1.8	69	95	28	63	5	11			9	0.31	0.62	23	0.1	0.22
ACC	34	23	26	20		10	10			10	24	24	23	24	35
Top%	26	7	18	2	1	73	32			44	60	99	28	59	99

Previous (Without PF50K Genomic Data)	CED	BW	WW	YW	Mik	TM	ME	HPG	CEM	STAY	MARB	YG	CW	REA	FAT
EPD	0	69	102	23	58	5	11			9	0.33	-0.08	27	-0.08	-0.03
ACC	P+	P+	P+	P		P	10			P	P	P	P	P	P
Top%	65	7	10	11	2	73	32			44	55	35	19	94	4

Where will future improvements to genomics prediction come from?

Journal of
Animal Breeding and Genetics



J. Anim. Breed. Genet. ISSN 0931-2668

EDITORIAL

Is genomic selection now a mature technology?

I. Misztal

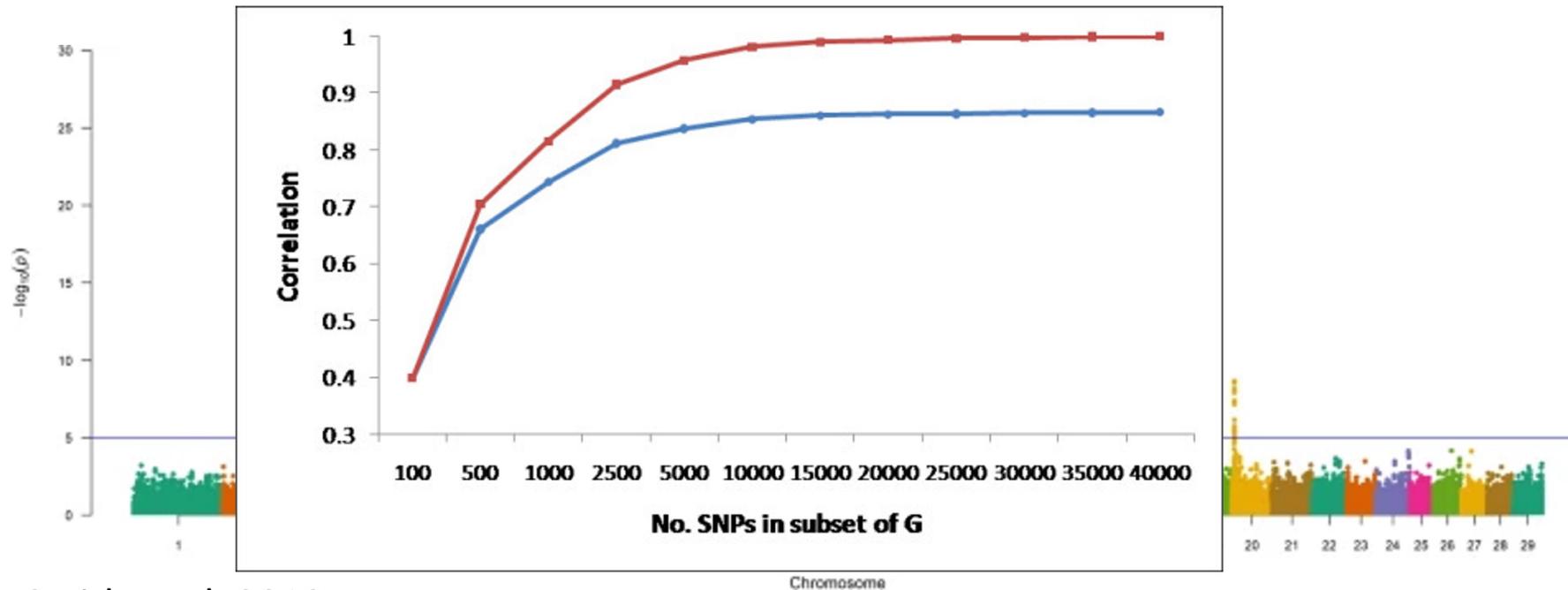
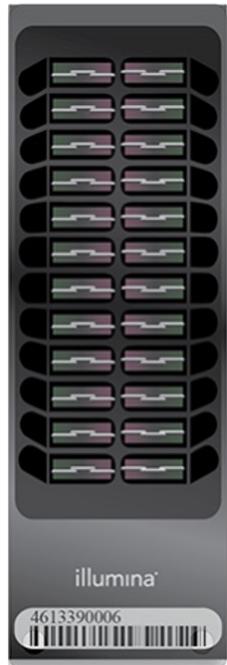
Subject Editor for JABG

University of Georgia

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$$\begin{pmatrix} X'X & X'Z \\ Z'X & Z'Z + G^{-1}\lambda_G \end{pmatrix} \begin{pmatrix} \hat{\beta} \\ \hat{u} \end{pmatrix} = \begin{pmatrix} X'y \\ Z'y \end{pmatrix}$$

Genomic Prediction



Rolf et al. 2012, Smith et al. 2019

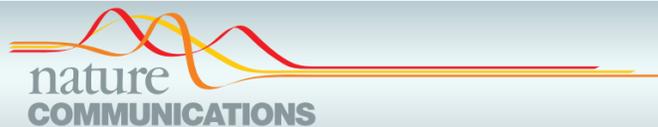
Complex Trait Mapping

Real. Life. Solutions.™

Quantifying the contribution of sequence variants with regulatory and evolutionary significance to 34 bovine complex traits

Ruidong Xiang^{a,b,1}, Irene van den Berg^{a,b}, Iona M. MacLeod^b, Benjamin J. Hayes^{b,c}, Claire P. Prowse-Wilkins^{a,b}, Min Wang^{b,d}, Sunduimijid Bolormaa^b, Zhiqian Liu^b, Simone J. Rochfort^{b,d}, Coralie M. Reich^b, Brett A. Mason^b, Christy J. Vander Jagt^b, Hans D. Daetwyler^{b,d}, Mogens S. Lund^e, Amanda J. Chamberlain^b, and Michael E. Goddard^{a,b}

^aFaculty of Veterinary & Agricultural Science, The University of Melbourne, Parkville, VIC 3052, Australia; ^bAgriculture Victoria, AgriBio, Centre for AgriBiosciences, Bundoora, VIC 3083, Australia; ^cCentre for Animal Science, The University of Queensland, St. Lucia, QLD 4067, Australia; ^dSchool of Applied Systems Biology, La Trobe University, Bundoora, VIC 3083, Australia; and ^eCenter for Quantitative Genetics and Genomics, Department of Molecular Biology and Genetics, Aarhus University, DK-8830 Tjele, Denmark



ARTICLE

<https://doi.org/10.1038/s41467-021-21001-0>

OPEN

Genome-wide fine-mapping identifies pleiotropic and functional variants that predict many traits across global cattle populations

Ruidong Xiang^{1,2✉}, Iona M. MacLeod², Hans D. Daetwyler^{2,3}, Gerben de Jong⁴, Erin O'Connor⁵, Chris Schrooten⁶, Amanda J. Chamberlain² & Michael E. Goddard^{1,2}

Real. Life. Solutions.™



Pausch et al. *Genet Sel Evol* (2017) 49:24
DOI 10.1186/s12711-017-0301-x



RESEARCH ARTICLE

Open Access



Evaluation of the accuracy of imputed sequence variant genotypes and their utility for causal variant detection in cattle

Hubert Pausch^{1*}, Iona M. MacLeod¹, Ruedi Fries², Reiner Emmerling³, Phil J. Bowman^{1,4}, Hans D. Daetwyler^{1,4} and Michael E. Goddard^{1,5}



ORIGINAL RESEARCH
published: 21 January 2021
doi: 10.3389/fgene.2020.603822



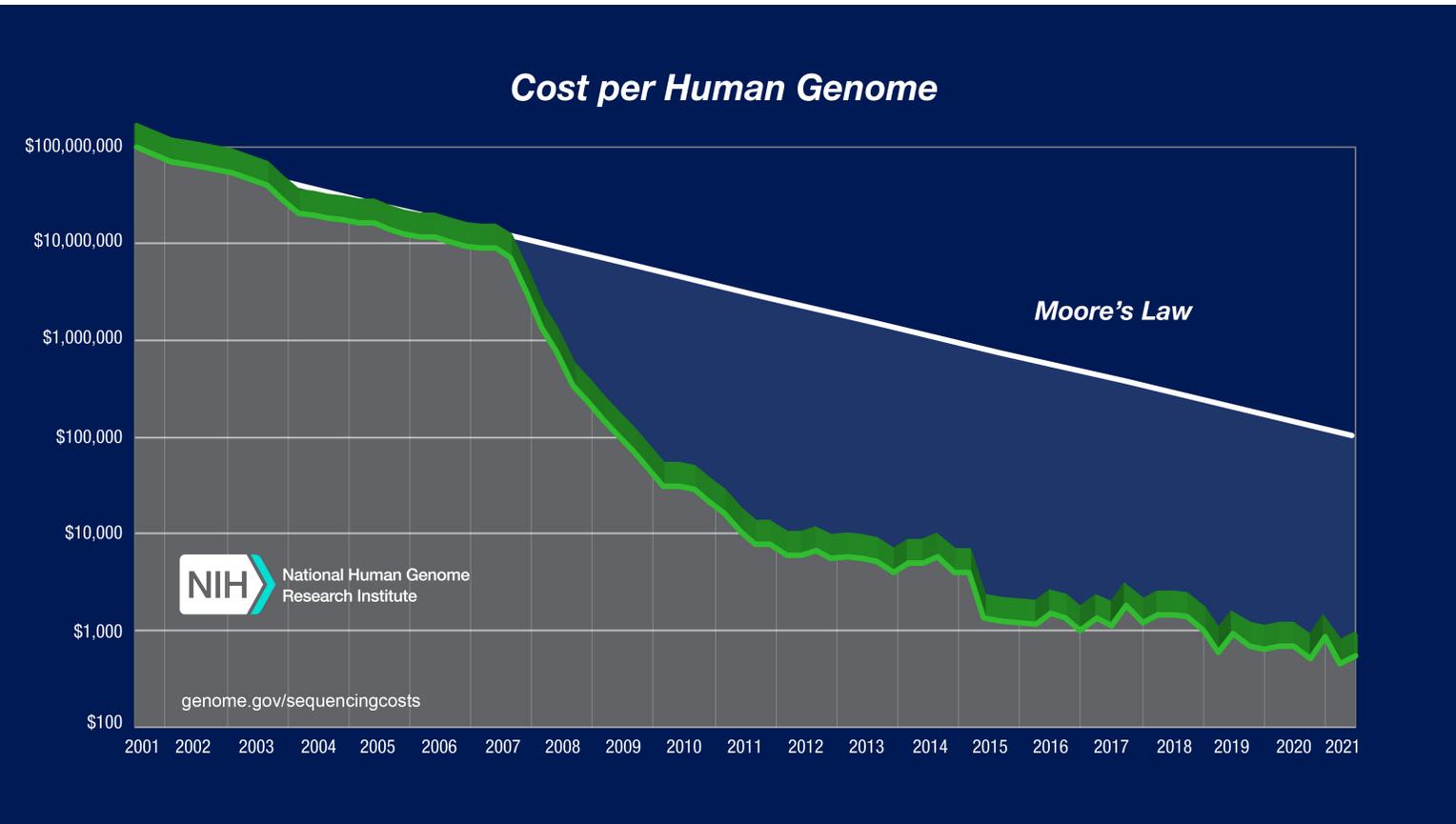
Genomic Prediction Based on SNP Functional Annotation Using Imputed Whole-Genome Sequence Data in Korean Hanwoo Cattle

Bryan Irvine M. Lopez^{1†}, Narae An^{1†}, Krishnamoorthy Srikanth¹, Seunghwan Lee², Jae-Don Oh³, Dong-Hyun Shin⁴, Woncheoul Park¹, Han-Ha Chai¹, Jong-Eun Park¹ and Dajeong Lim^{1*}

¹ Division of Animal Genomics and Bioinformatics, National Institute of Animal Science, Rural Development Administration, Wanju, South Korea, ² Department of Animal Science and Biotechnology, Chungnam National University, Daejeon, South Korea, ³ Department of Animal Biotechnology, Chonbuk National University, Jeonju, South Korea, ⁴ Department of Agricultural Convergence Technology, Chonbuk National University, Jeonju, South Korea



Sequencing opportunities



- Declining costs of sequencing
- Evolving tech landscape
- Low-pass
- Imputation

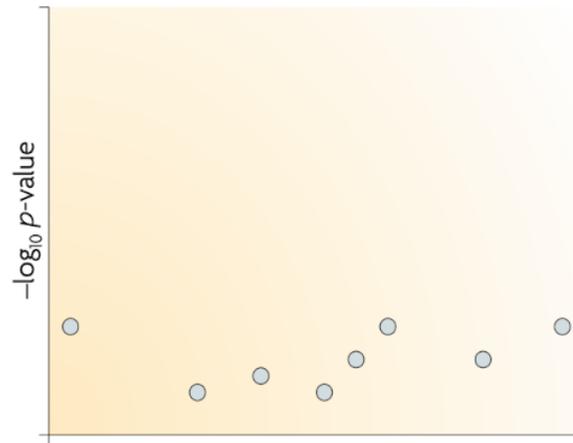
Future utility of sequence information in genetic evaluations

- ***Orders of magnitude*** more information!
- Leveraging functional annotations
- Causal variant discovery & implementation
- Feature selection genomic prediction
- Integration... is another issue

a Genotype data with missing data at untyped SNPs (grey question marks)

1	?	?	?	1	?	1	?	0	2	2	?	?	2	?	0	▲
0	?	?	?	2	?	2	?	0	2	2	?	?	2	?	0	
1	?	?	?	2	?	2	?	0	2	1	?	?	2	?	0	
1	?	?	?	2	?	1	?	1	2	2	?	?	2	?	0	
2	?	?	?	2	?	2	?	1	2	1	?	?	2	?	0	▲
1	?	?	?	1	?	1	?	1	2	2	?	?	2	?	0	
1	?	?	?	2	?	2	?	0	2	1	?	?	2	?	1	
2	?	?	?	1	?	1	?	1	2	1	?	?	2	?	1	
1	?	?	?	0	?	0	?	2	2	2	?	?	2	?	0	▲

b Testing association at typed SNPs may not lead to a clear signal



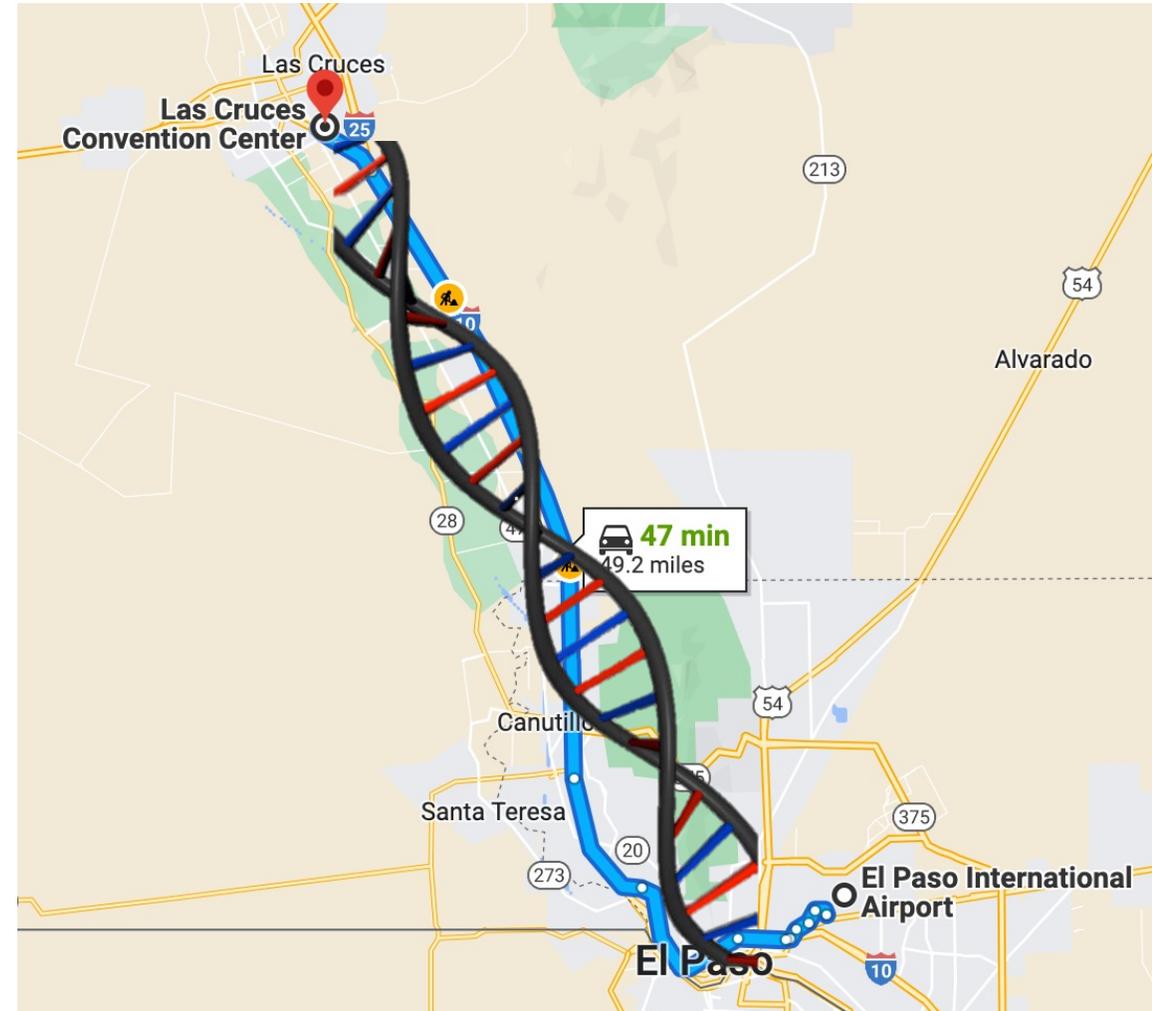
Imputation visualized

49.2 miles

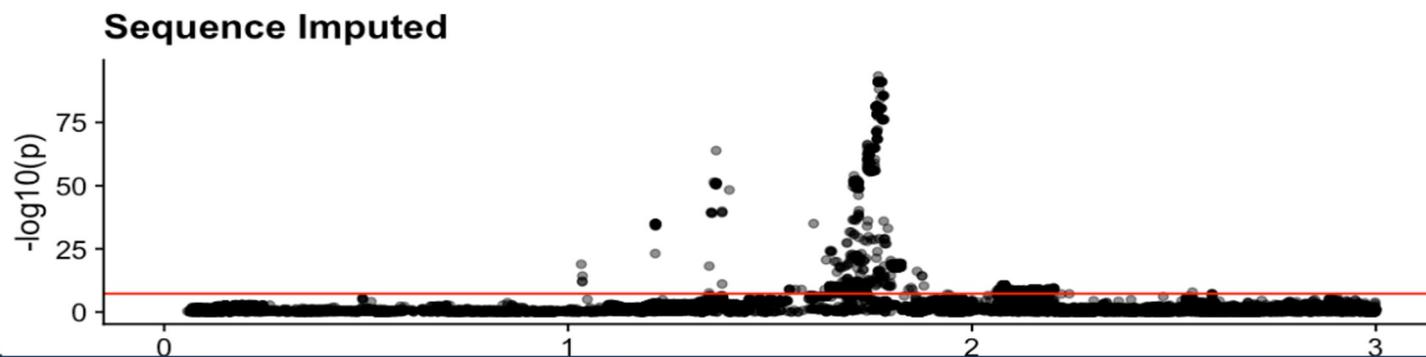
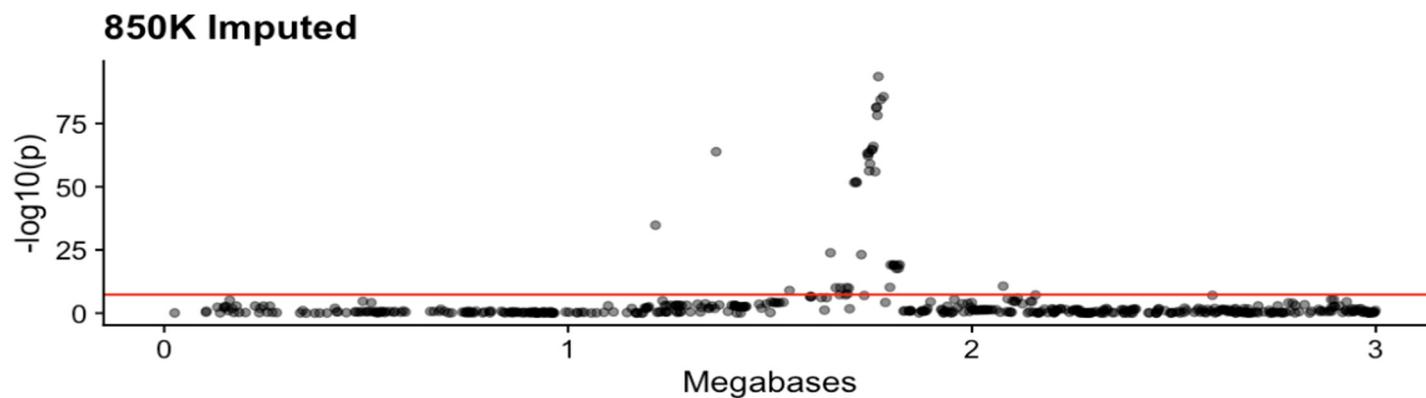
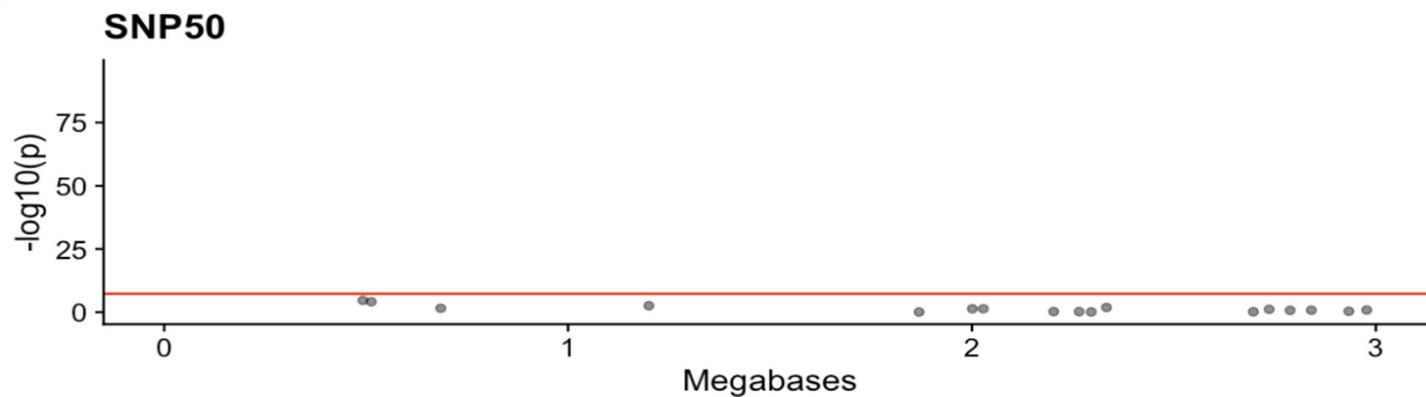
47 minutes

Unimputed chip genotypes:
Look for 1-second every ~5 minutes!

Sequence-level genotypes:
A 1-second look every ~5 seconds!

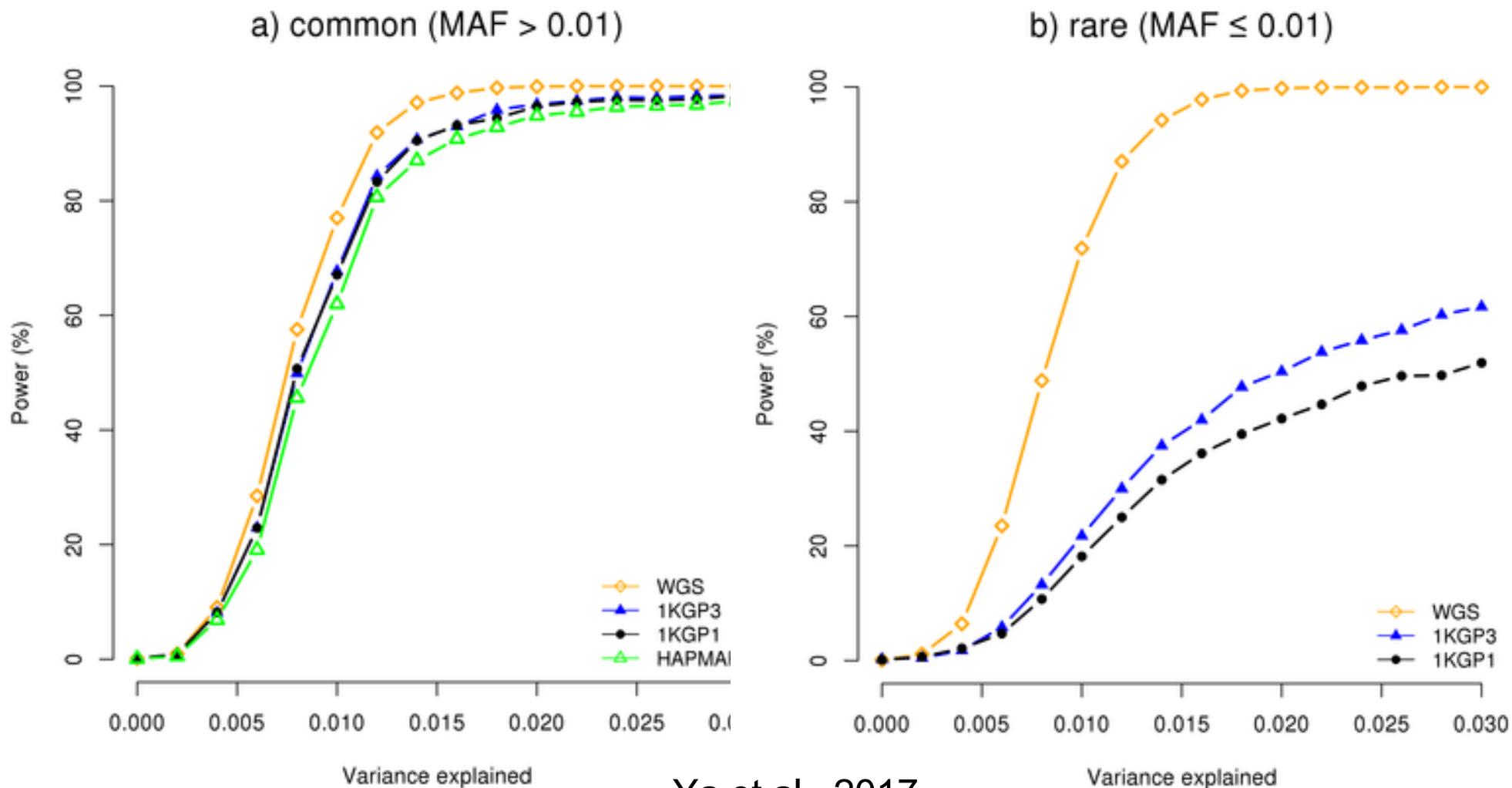


The Power of Imputation



Rowan et al. 2021

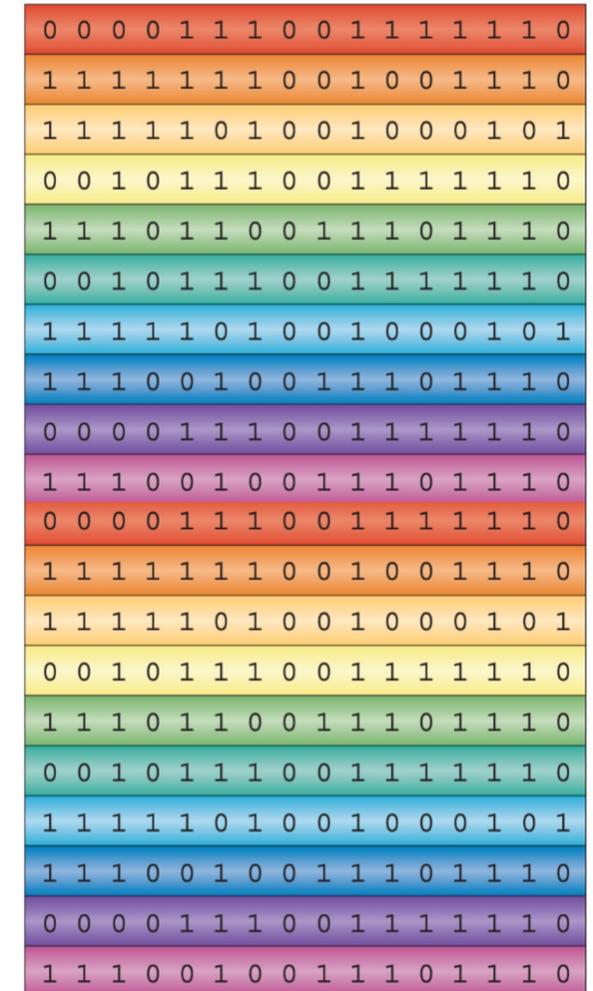
If our goal is to identify causal mutations, Imputation quality matters!



Ya et al., 2017

What does accurate imputation need?

- A large set of accurate reference haplotypes
 - Higher density chip or sequence
 - More is always better
 - Crap in crap out
- Good reference genome
- Recombination map



Marchini et al. 2010

Coverage considerations

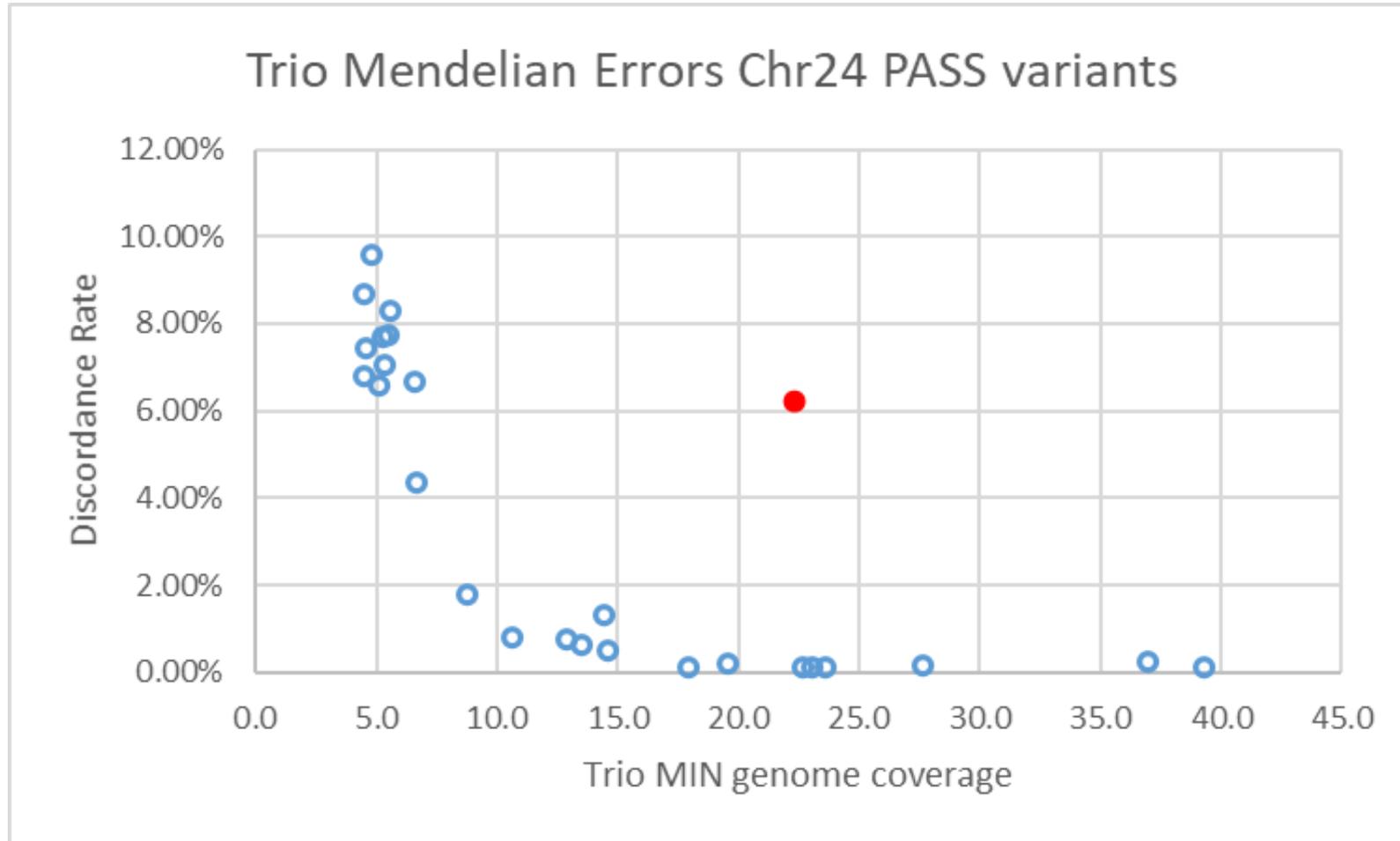


Figure courtesy of Bob Schnabel (Mizzou)

Evaluating Imputation Accuracy

- Metrics
 - Concordance: There's a SNP here. Did I impute the right genotype?
 - Deceiving at low minor allele frequencies
 - Correlation
 - Imputation quality statistic
- Truth set validation vs. internal software values

How to build a reference panel?



Breed Specific

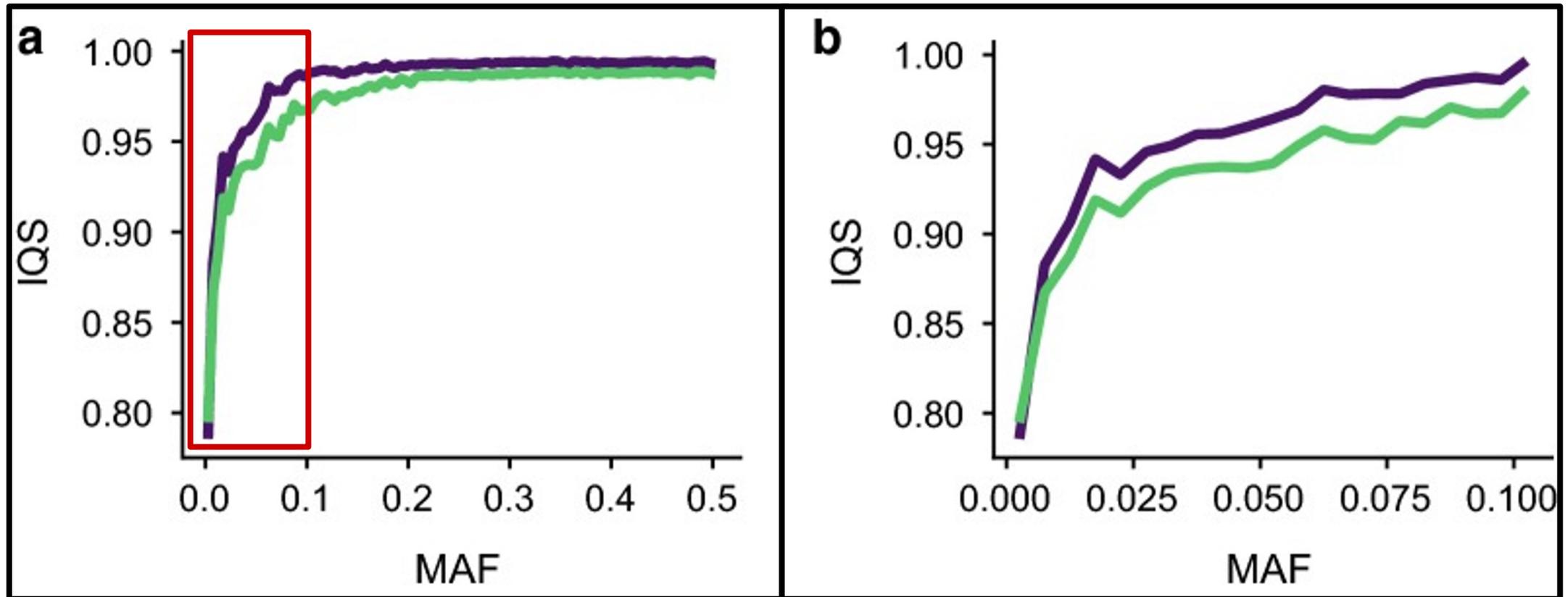


Multi-breed

1) It's really hard to impute a haplotype that you don't observe!

2) A breed label doesn't guarantee share (or not shared) haplotypes

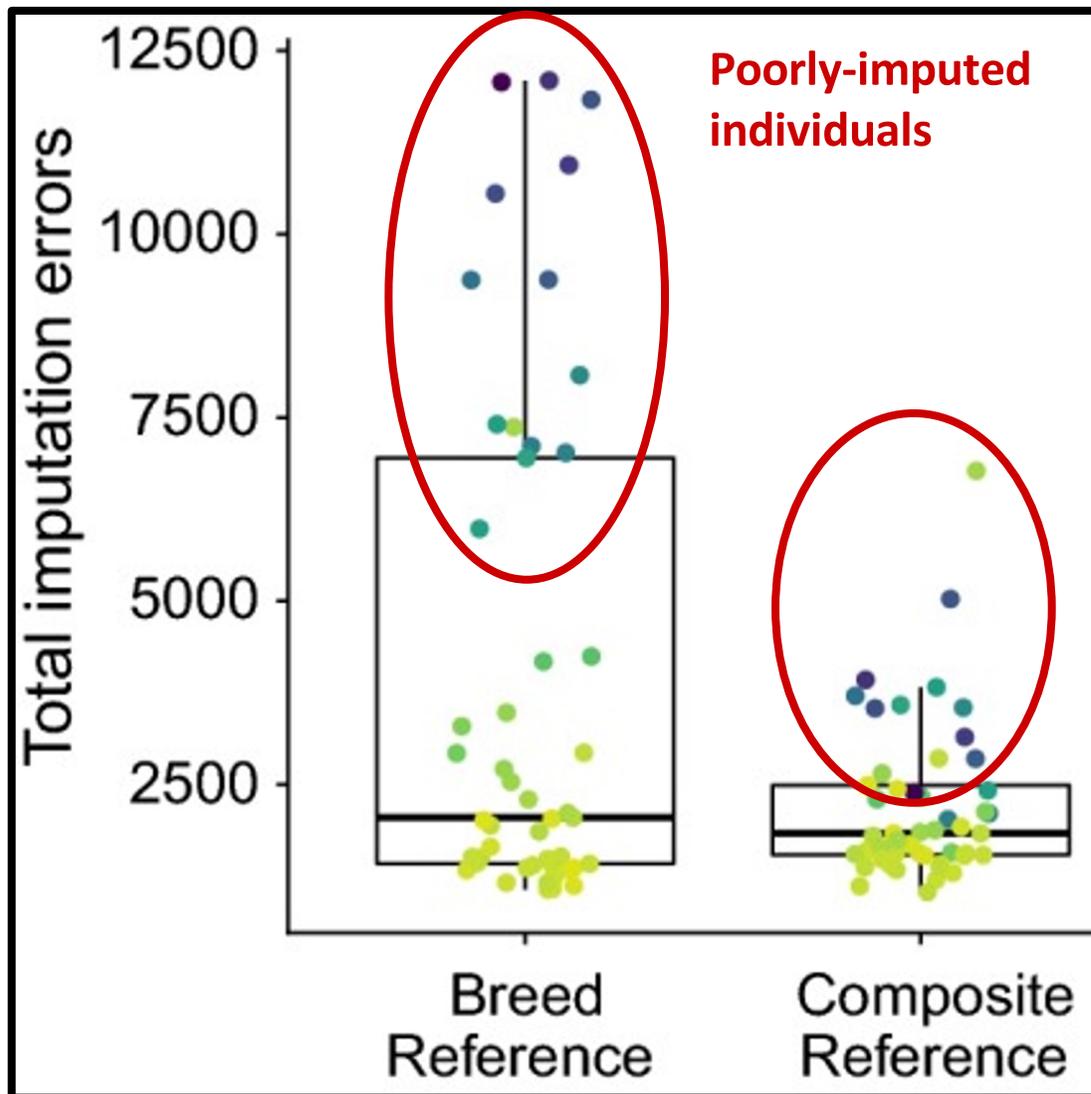
Diverse samples need a diverse imputation reference panel



- Multi-breed reference
- Within-breed reference

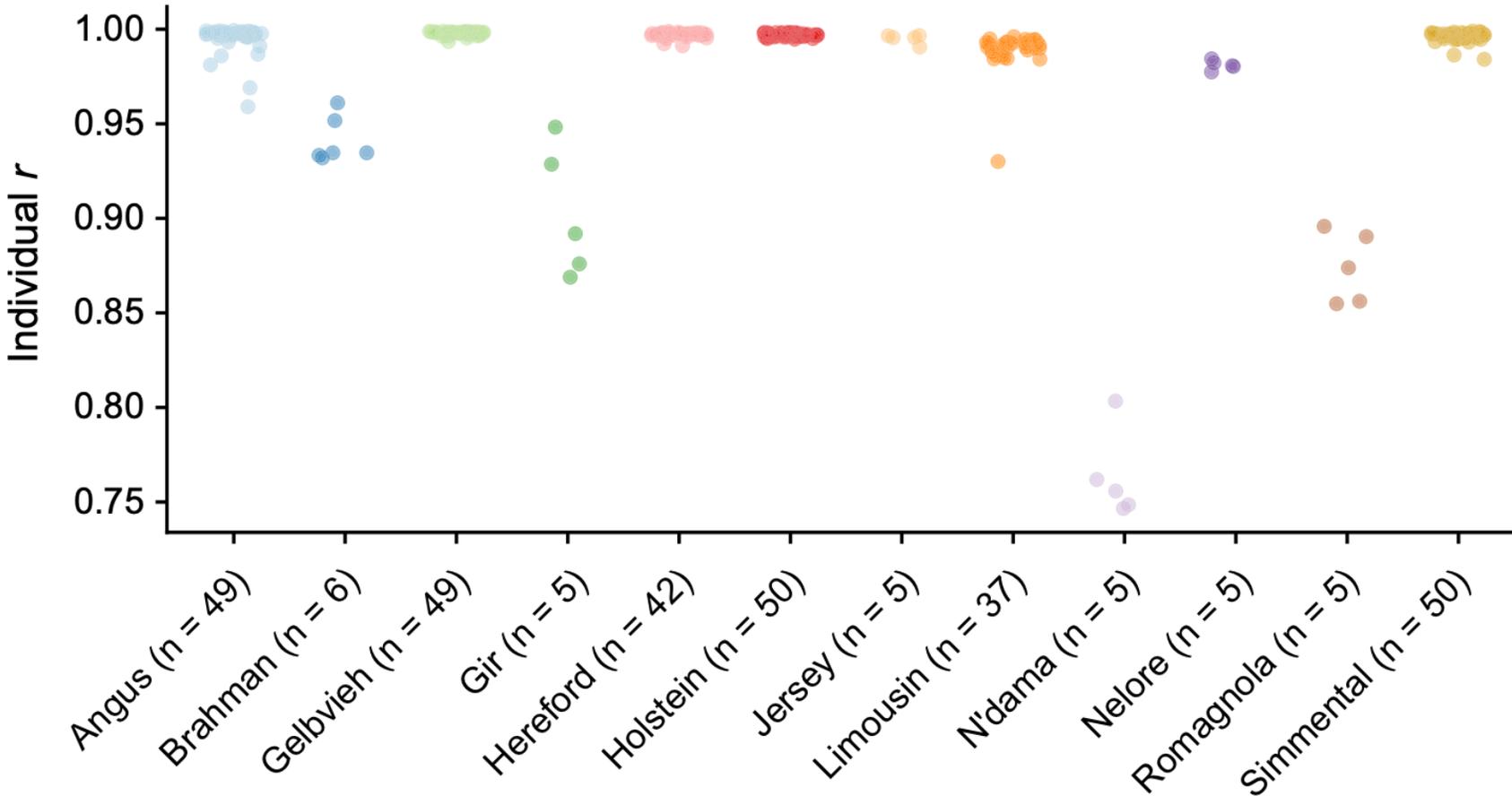
Rowan et al. 2019

Imputation is just pattern matching!



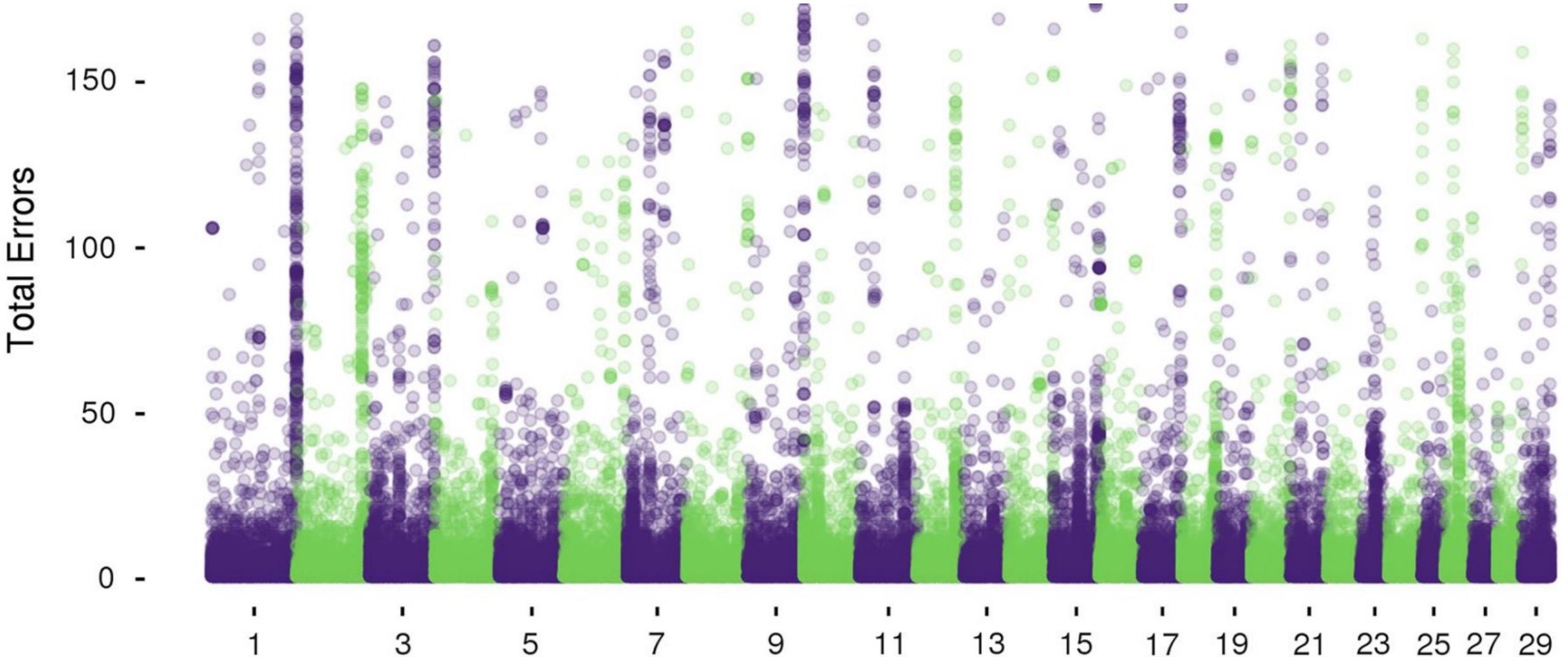
n =50 Gelbvieh

Representation matters!



Breed	Mean	Min	Max
Gelbvieh	0.998	0.994	0.999
Hereford	0.997	0.991	0.999
Holstein	0.997	0.995	0.998
Simmental	0.996	0.984	0.999
Angus	0.995	0.959	0.999
Jersey	0.995	0.991	0.997
Limousin	0.989	0.930	0.996
Nelore	0.981	0.977	0.984
Brahman	0.941	0.932	0.961
Gir	0.903	0.869	0.948
Romagnola	0.874	0.855	0.896
N'Dama	0.763	0.747	0.803

Room for improvement



Rowan et al. 2019

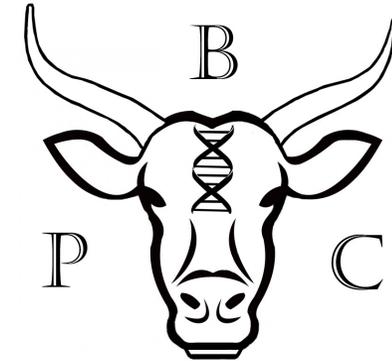
Choosing animals for a reference

- Highest progeny number?
 - I'd argue we should sequence these regardless
- Sire/son sequencing = wasted resources
- Let genotype data drive sequencing



Future opportunities

- Pangenome-aware imputation
 - Imputing structural variation
 - Further breed diversity
- Software developments
 - Genotype callers
 - Imputation algorithms
- Rare alleles



Troy's recommendations to take or leave

- Sequence more animals!
 - Be intentional about representing novel haplotypes
 - And make the data public!
- Minimum of 10X coverage
- Don't assume: Evaluate imputation accuracy
- Get creative with the extra ~30 million SNPs

Takeaways

Future improvements to genetic evaluations will rely on sequence-level genotype information

Whether low-pass or using chips, imputation to sequence level can add value to genomic resources



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Imputation is pattern matching. Relies on high-quality reference.

Continually improving references is essential. Minimum of 10X, choose animals with a plan

