

# Managing genetic diversity: Strategies for sustainable livestock improvement

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Abstract #: 1115

# Genomic selection: Progress and pitfalls

## Key benefits:

- Genomic selection allows increased accuracy and genetic gain
- Reduced generation interval

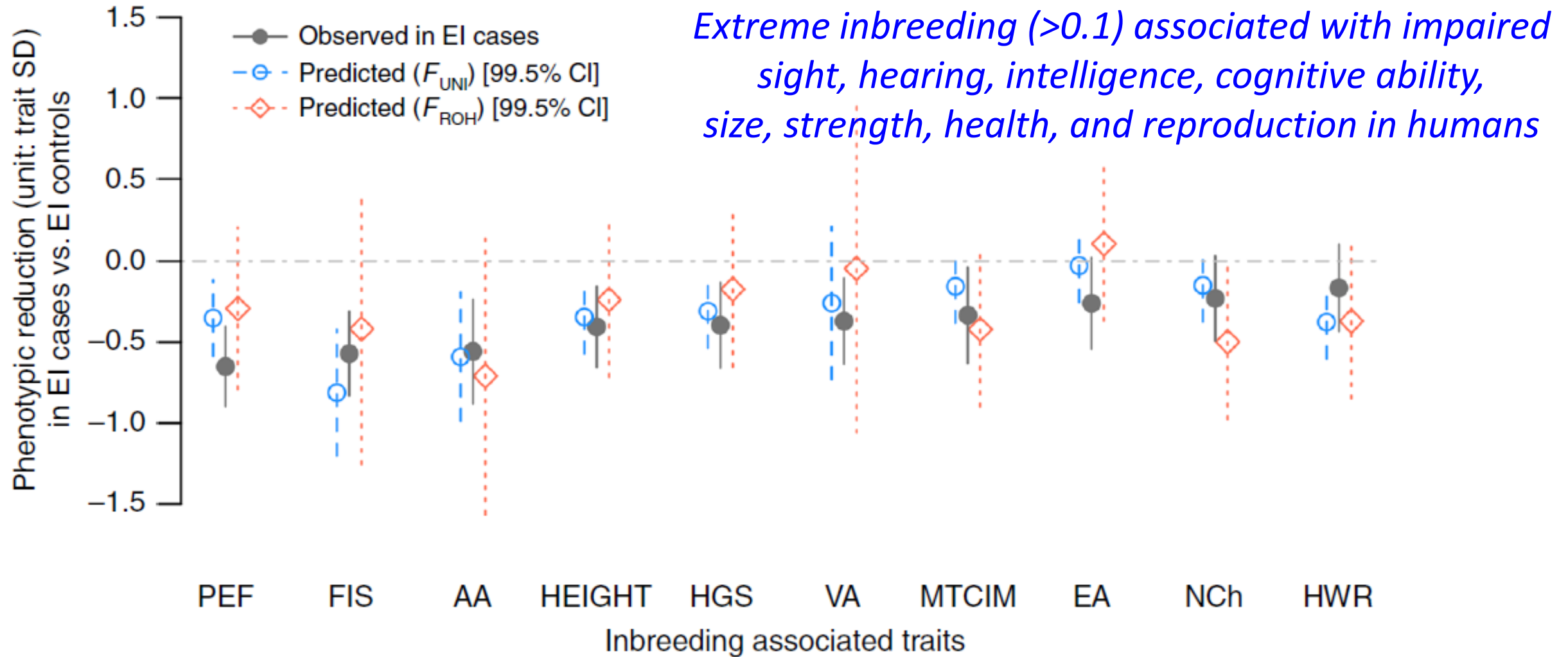


## Key Challenges:

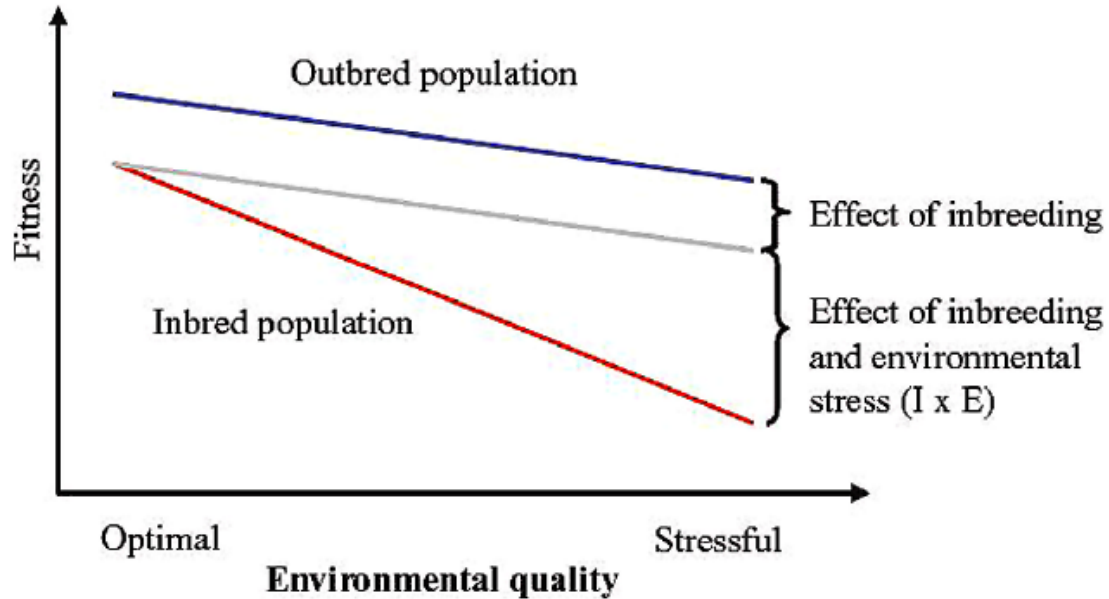
- Accelerated increase in inbreeding
- Increased homozygosity
- Emergence of recessive disorders

Need to balance fast progress / short-term gain with sustainable improvement

# Why genetic diversity matters



# Why genetic diversity matters



**FIGURE 2 | Schematic illustration of fitness effects of inbreeding by environment interactions.** Assuming the effect of inbreeding is independent of the environment, the reduction in fitness as a result of reduced environmental quality will be equal for outbred and inbred populations. The blue and gray lines illustrate fitness of an outbred and an inbred population, respectively, in the absence of inbreeding by environment interactions. Inbreeding depression is, however, often more severe under stressful environmental conditions. Thus, the red line illustrates fitness of an inbred population taking into account the effect of inbreeding by environment interactions (redrawn from Kristensen et al., 2010).

*Short-term:* as populations become smaller and less diverse, relative fitness decreases (i.e., survival and fertility, etc.)

*Long-term consequences:* lower gene diversity means less raw material for adaptations to changing environments, which may affect long-term survival

*Also:* Increased risk of nasty recessives

# Assessment of genetic diversity in (Canadian) dairy cattle

 J. Dairy Sci. 103  
<https://doi.org/10.3168/jds.2019-18013>  
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## Effect of genomic selection on rate of inbreeding and coancestry and effective population size of Holstein and Jersey cattle populations

Bayode O. Makanjuola,<sup>1</sup> Filippo Miglior,<sup>1,2</sup> Emhimad A. Abdalla,<sup>1</sup> Christian Maltecca,<sup>1,3</sup> Flavio S. Schenkel,<sup>1</sup> and Christine F. Baes<sup>1,4\*</sup>

### RESEARCH ARTICLE

### Open Access



## Inbreeding and runs of homozygosity before and after genomic selection in North American Holstein cattle

Mehrnush Forutan<sup>1,2\*</sup>, Saeid Ansari Mahyari<sup>1\*</sup>, Christine Baes<sup>2</sup>, Nina Melzer<sup>3</sup>, Flavio Schramm Schenkel<sup>1,2</sup> and Mehdi Sargolzaei<sup>2,4,5</sup>

### RESEARCH ARTICLE

### Open Access



## Identification of unique ROH regions with unfavorable effects on production and fertility traits in Canadian Holsteins

Bayode O. Makanjuola<sup>1\*</sup>, Christian Maltecca<sup>1,2</sup>, Filippo Miglior<sup>1</sup>, Gabriele Marras<sup>3</sup>, Emhimad A. Abdalla<sup>1</sup>, Flavio S. Schenkel<sup>1</sup> and Christine F. Baes<sup>1,4</sup>



J. Dairy Sci. 103:5302–5313  
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## Symposium review: Exploiting homozygosity in the era of genomics—Selection, inbreeding, and mating programs\*

C. Maltecca,<sup>1†</sup> F. Tiezzi,<sup>1</sup> J. B. Cole,<sup>2</sup> and C. Baes<sup>3,4</sup>



J. Dairy Sci. 102:2807–2817  
<https://doi.org/10.3168/jds.2018-15520>  
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## Symposium review: The genomic architecture of inbreeding: How homozygosity affects health and performance\*

Christine F. Baes,<sup>1,2†</sup> Bayode O. Makanjuola,<sup>1</sup> Filippo Miglior,<sup>1,3</sup> Gabriele Marras,<sup>1,4</sup> Jeremy T. Howard,<sup>5,6</sup> Allison Fleming,<sup>1,3</sup> and Christian Maltecca<sup>6</sup>



### RESEARCH ARTICLE

## Effect of recent and ancient inbreeding on production and fertility traits in Canadian Holsteins

Bayode O. Makanjuola<sup>1\*</sup>, Christian Maltecca<sup>1,2</sup>, Filippo Miglior<sup>1</sup>, Flavio S. Schenkel<sup>1</sup> and Christine F. Baes<sup>1,3</sup>



J. Dairy Sci. 108:3045–3067  
<https://doi.org/10.3168/jds.2024-26035>

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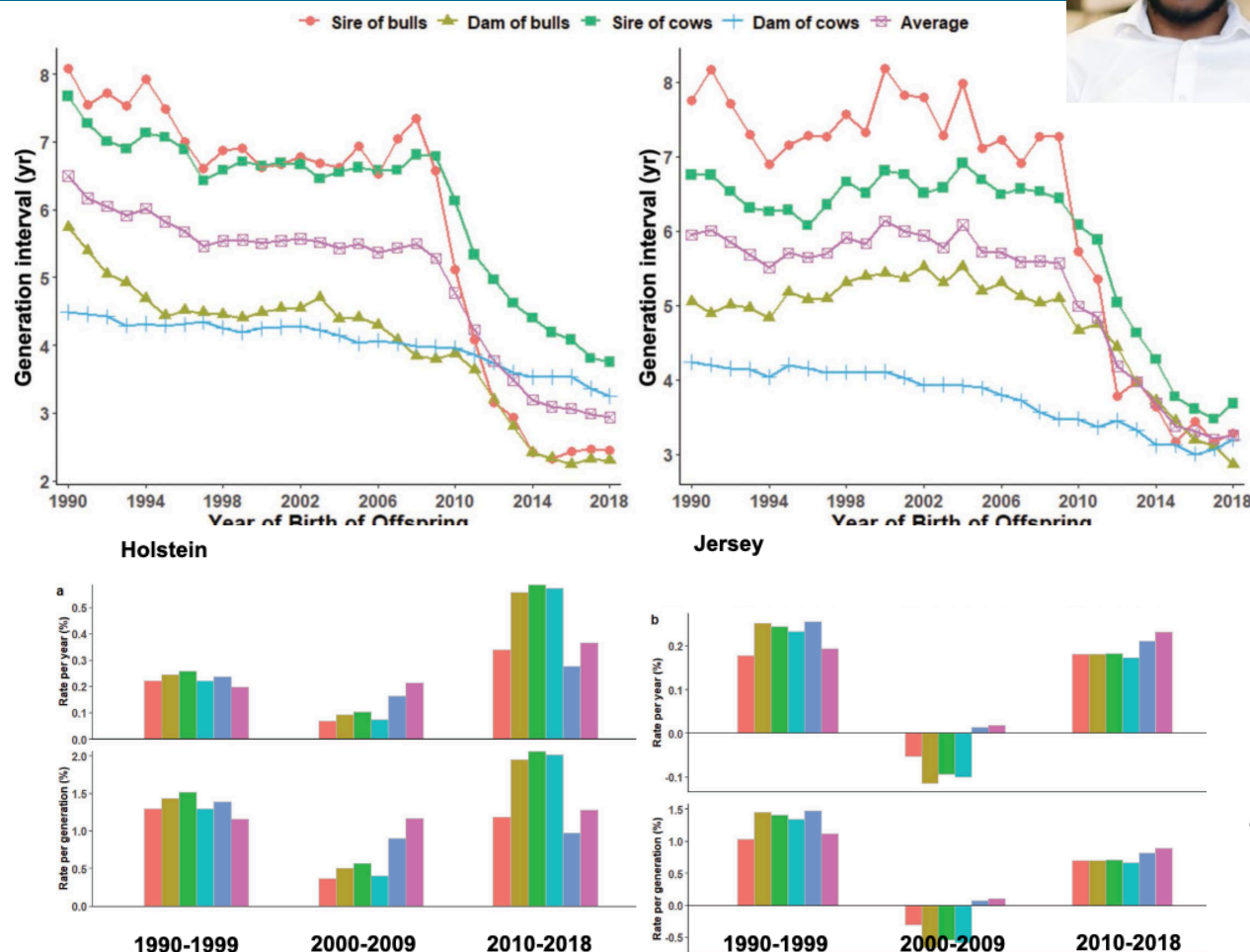
## Invited review: Management of genetic defects in dairy cattle populations

John B. Cole,<sup>1,2,3\*</sup> Christine F. Baes,<sup>4,5</sup> Sophie A. E. Eaglen,<sup>6</sup> Thomas J. Lawlor,<sup>7</sup> Christian Maltecca,<sup>3</sup> M. Sofia Ortega,<sup>8</sup> and Paul M. VanRaden<sup>9</sup>

# Evidence: inbreeding trends



- Estimate the rate of inbreeding and effective population size using both pedigree and genomic information for animals born between 1990 and 2018
- Evaluate the changes in effective population size following genomic selection
- Compare genetic diversity pre and post genomics

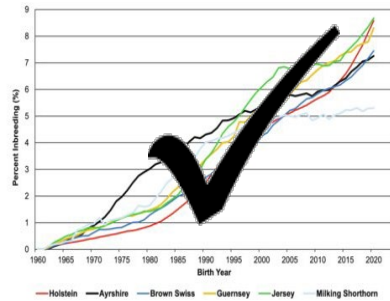


Effective population size	Estimated values
Pedigree ( $N_{ePED}$ )	66
Runs of Homozygosity ( $N_{eROH\_SNP1101}$ )	46



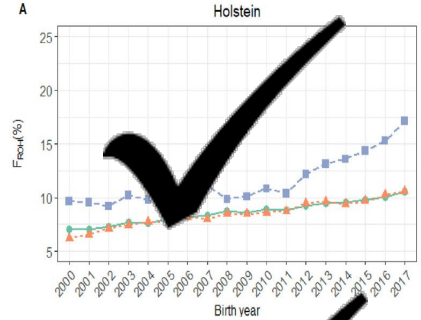
# Rate of Inbreeding is increasing (fact check)

*Is the rate of inbreeding increasing in the US? In all breeds?*



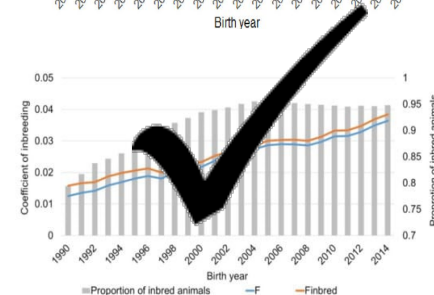
L.F. Brito, N. Bedere, F. Douhard, H.R. Oliveira, M. Arnal, F. Peñagaricano, A.P. Schinckel, C.F. Baes, F. Miglior, (2021) Genetic selection of high-yielding dairy cattle toward sustainable farming systems in a rapidly changing world. *Animal*, (15, 1)

*How about in Australian cows?*



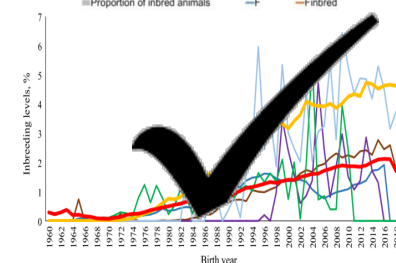
Scott, B. A., M. Haile-Mariam, B. G. Cocks, and J. E. Pryce. 2021. How genomic selection has increased rates of genetic gain and inbreeding in the Australian national herd, genomic information nucleus, and bulls. *Journal of Dairy Science* 104:11832-11849.

*What about German cows?*



Wirth A, Duda J, Distl O. Genetic Diversity and the Impact of the Breed Proportions of US Brown Swiss in German Brown Cattle. *Animals*. 2021; 11(1):152

*But what about red cows?*



Nyman, S., Johansson, A.M., Palucci, V. et al. Inbreeding and pedigree analysis of the European red dairy cattle. *Genet Sel Evol* 54, 70 (2022). <https://doi.org/10.1186/s12711-022-00761-3>

(...etc)

Rates of inbreeding are increasing faster in the genomic era

Implementation of genomic selection decreasing effective population size

Not sure what long-term implications are

# Evidence: inbreeding trends



- Estimate the effect of inbreeding on production and fertility traits
- Assess the effect of recent and ancient inbreeding on production and fertility traits

**Table 4** Estimates of inbreeding depression for all significant traits, expressed as the difference (Diff) in predicted phenotype between lowly inbred (5% percentile) and highly inbred (95% percentile) from the mean for  $F_{PED}$ ,  $F_{ROH}$  and  $F_{GRM}$

Traits <sup>a</sup>	$F_{PED}^b$			$F_{ROH}^b$		
	Low	High	Diff	Low	High	Diff
MY (kg)	9031.89	8887.20	144.69	9152.62	8809.77	342.85
FY (kg)	357.35	354.09	3.26	359.29	354.25	5.04
PY (kg)	294.74	292.17	2.57	296.79	289.71	7.08
AFS (day)	450.51	456.91	-6.40	451.04	457.75	-6.71
NS <sup>c</sup>	37.03	39.48	-2.45	35.34	40.21	-4.87
NRR <sup>d</sup>	70.27	69.55	0.72	70.61	68.05	2.56
FSTC (day)	19.47	22.30	-2.83	18.22	22.34	-4.12
NS <sup>c</sup> (%)	52.43	55.99	-3.56	53.67	54.71	-1.04
NRR <sup>d</sup> (%)	54.50	52.58	1.92	54.60	53.46	1.14

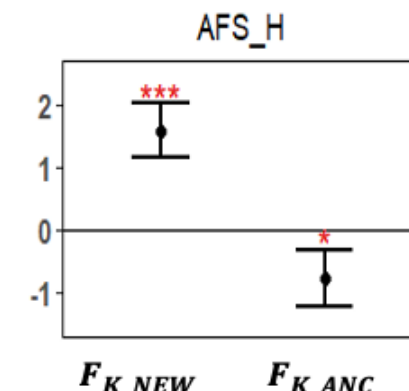
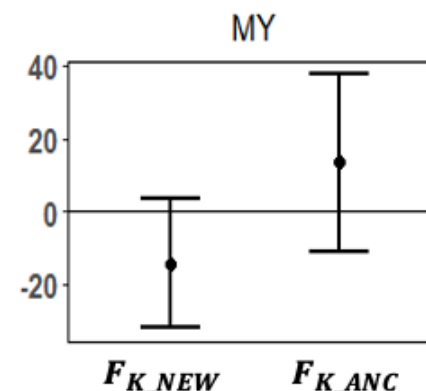
<sup>a</sup>MY milk yield, FY fat yield, PY protein yield, AFS\_H age at first service for heifers, NS\_H number of service for heifers, NRR\_H 56-day non-return rate for heifers, FSTC\_H first service to conception for heifers, CTFSC\_C conception to first service for cows, NS\_C number of service for cows, NRR\_C 56-day non-return rate for cows, FSTC\_C first service to conception for cows

<sup>b</sup> $F_{PED}$  = pedigree inbreeding;  $F_{ROH}$  = segment-based genomic inbreeding;  $F_{GRM}$  = marker-by-marker-based genomic inbreeding

<sup>c</sup>Incidence of more than one service after first

<sup>d</sup>Incidence of no subsequent service between 15 and 56 days following the first service

Effect of 1% increase in new and ancient inbreeding





# Estimated inbreeding depression (fact check)

*Were there detrimental effects of inbreeding in American Holsteins in the 90's?*

*How about in French  
cows?*

## What about Jerseys?

*Surely not in Dutch cows, though, right?*

(etc...)

Variables	Rajasthan herds						Goats herds					
	MS	SD	SE	MS	SD	SE	MS	SD	SE	MS	SD	SE
Female goats, N	15			15			15			15		
BSOVF field visits, #	-0.53	0.13	0.03	-0.44	0.17	0.03	0.23	0.15	0.03	2.50	0.51	0.10
BSOVF field visits, %	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 1	-1.14	0.13	0.03	0.19	0.30	0.03	0.11	0.09	0.03	0.00	0.10	0.01
Number 2	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 3	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 4	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 5	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 6	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 7	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 8	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 9	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 10	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 11	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 12	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 13	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 14	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 15	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 16	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 17	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 18	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 19	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
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Number 34	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
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Number 40	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 41	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 42	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
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Number 49	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 50	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 51	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 52	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
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Number 64	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
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Number 68	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
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Number 71	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
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Number 79	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 80	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 81	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 82	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 83	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 84	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 85	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 86	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00	0.00	0.02	0.00
Number 87	0.00	0.02	0.00	0.00</								

Trait	Herds		
	Holstein	Jersey	Normande
Milk (kg)	-0.001 (0.001)	-0.001 (0.001)	-32 (0.04)
Fat (kg)	-1.7 (0.4)	0 (0.4)	-1.4 (0.5)
Protein (kg)	-1.2 (0.3)	0.1 (0.3)	-1.2 (0.5)
Fat content (%)	-0.0003 (0.008)	-0.0003 (0.001)	-0.001 (0.02)
Protein content (%)	-0.0001 (0.006)	-0.0001 (0.001)	-0.001 (0.02)
SCS	0.006 (0.02)	0.006 (0.02)	0.001 (0.3)
CR (%)	-0.27 (0.1)	-0.44 (0.9)	-0.28 (0.8)
DPS (d)	0.12 (0.01)	0.12 (0.01)	0.12 (0.01)

Covariate	Milk			Fat			Protein		
	Estimated regression (tension $\text{kg F}_{0.05}$ )	$P$ -value	$R^2$	Estimated regression (tension $\text{kg F}_{0.05}$ )	$P$ -value	$R^2$	Estimated regression (tension $\text{kg F}_{0.05}$ )	$P$ -value	$R^2$
Equal slope (Model 1)	$F_{0.05}$	15.79 $\pm$ 4.4	<0.01	0.54 $\pm$ 0.1	<0.01	0.54	0.54 $\pm$ 0.1	<0.01	0.54
	$F_{0.05}^{\text{Milk}}$	-2.54 $\pm$ 0.3	<0.01	-0.09 $\pm$ 0.01	<0.01	0.90	-0.09 $\pm$ 0.01	<0.01	0.90
	$F_{0.05}^{\text{Fat}}$	0.06 $\pm$ 0.01	<0.01	0.00 $\pm$ 0.01	<0.01	0.92	0.00 $\pm$ 0.01	<0.01	0.92
Heterogeneous slopes (Model 2)	$F_{0.05}$	17.20 $\pm$ 2.3	<0.01	0.73 $\pm$ 0.3	0.87	0.53	0.83 $\pm$ 0.3	<0.01	0.83
	$F_{0.05}^{\text{Milk}}$	-0.07 $\pm$ 0.01	<0.01	-0.08 $\pm$ 0.01	0.88	0.27	-0.08 $\pm$ 0.01	<0.01	0.73
	$F_{0.05}^{\text{Fat}}$	0.07 $\pm$ 0.02	<0.01	1.69 $\pm$ 0.4	<0.01	1.36	1.36 $\pm$ 0.4	<0.01	1.36
	$F_{0.05}^{\text{Protein}}$	-0.04 $\pm$ 0.01	<0.01	1.74 $\pm$ 0.3	<0.01	1.28	1.28 $\pm$ 0.3	<0.01	1.28
	$F_{0.05}^{\text{Milk}} \times F_{0.05}^{\text{Fat}}$	-0.04 $\pm$ 0.01	<0.01	-0.14 $\pm$ 0.02	0.51	0.33	-0.14 $\pm$ 0.02	<0.01	0.42
	$F_{0.05}^{\text{Fat}} \times F_{0.05}^{\text{Protein}}$	0.01 $\pm$ 0.01	<0.01	-0.09 $\pm$ 0.02	<0.01	0.07	-0.07 $\pm$ 0.02	<0.01	0.07
	$F_{0.05}^{\text{Milk}} \times F_{0.05}^{\text{Protein}}$	0.00 $\pm$ 0.01	<0.01	0.00 $\pm$ 0.01	<0.01	0.00	0.00 $\pm$ 0.01	<0.01	0.00

Trait	$F_{ED}$ Estimate	SE	$F_{ROH}$ Estimate	SE	$F_{GRM}$ Estimate	SE
MY	-37.95**	3.66	-36.35**	3.66	-48.87**	2.81
FFY	-1.54**	0.14	-1.34**	0.09	-1.60**	0.11
PI	-1.27**	0.11	-1.20**	0.09	-1.55**	0.09
CY	0.46*	0.23	0.43*	0.15	0.62**	0.10
KCF	0.11	0.19	0.06	0.06	0.09	0.05
RL	0.13	0.19	0.19	0.12	0.42**	0.10
CR	-0.31*	0.12	0.02	0.08	-0.39**	0.13
SC5150	0.52**	0.44	0.56**	0.28	0.44	0.34
SC5400	0.63**	0.43	0.59**	0.28	1.15**	0.43

MY, 305-day milk yield (kg); FFY, first lactation peak yield (kg); PI, peak yield (kg); CY, calving interval (days); KCF, interval calving to first insemination (days); RL, interval from first insemination to calving (days); CR, calving interval (days); SC5150, somatic cell score (days  $\times 10^3$ ); SC5400, somatic cell score (days  $\times 10^3$ )  $\times 100$  (log<sub>10</sub> of cells/ml); F<sub>ED</sub>, pedigree inbreeding based on all genes compared with allele frequencies of 0.5; F<sub>ROH</sub>, genomic inbreeding based on all regions of homozygosity; F<sub>GRM</sub>, inbreeding based on genomic relationship matrix. Significance for non-matry is indicated by asterisks: \*P < 0.05, \*\*P < 0.01.

Smith, L. A., B. G. Cassell, and R. E. Pearson. 1998. The effects of inbreeding on the lifetime performance of dairy cattle. *Journal of Dairy Science* 81:2729-2737

Dezetter, C., H. Leclerc, S. Mattalia, A. Barbat, D. Boichard, and V. Ducrocq. 2015. Inbreeding and crossbreeding parameters for production and fertility traits in Holstein, Montbéliarde, and Normande cows. *Journal of Dairy Science* 98:4904-4913.

Gulisija, D., D. Gianola, K. A. Weigel, and M. A. Toro. 2006. *Between-founder heterogeneity in inbreeding depression for production in Jersey cows. Livestock Science 104:244-253.*

Doekes, H. P., R. F. Veerkamp, P. Bijma, G. de Jong, S. J. Hiemstra, and J. J. Windig. Inbreeding depression due to recent and ancient inbreeding in Dutch Holstein-Friesian dairy cattle. *Genetics Selection Evolution* (2019) 51:54.

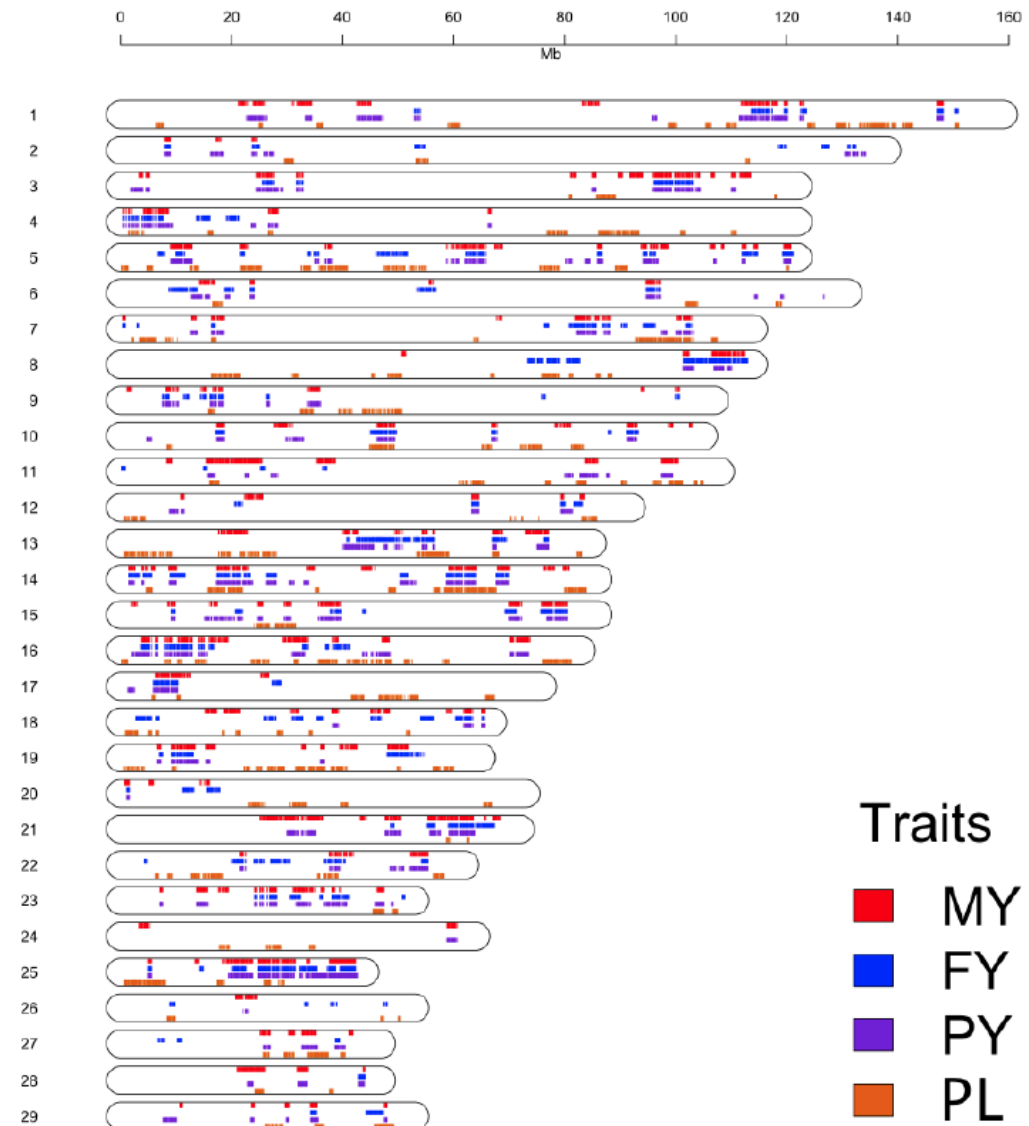
## Negative effect of inbreeding observed on many traits

Recent inbreeding had more detrimental effect compared to ancient inbreeding

Little evidence of purging  
due to selection  
(Gulisija, Crow and Weigel 2006  
is an exception... others?)

# ROH used to find undesirable (not lethal) regions

- Some mutations in the population are **sub-lethal**: they harm performance but don't kill animals
- They are more difficult to find than lethal haplotypes
- May appear in ROH



# Evidence: inbreeding trends



- Identify specific genetic regions where homozygosity has negative effects on production and fertility traits
- Identify unique ROH with unfavorable effects across multiple traits

Trait	BTA	ROH effects	$-\log_{10}$ (P value)
MY (kg)	8	-545.69	6.64
FY (kg)	14	-23.59	6.65
PY (kg)	8	-16.34	6.25
NS	6	0.23	5.40
NRR	1	-0.10	3.41
FSTC (day)	6	7.80	4.84

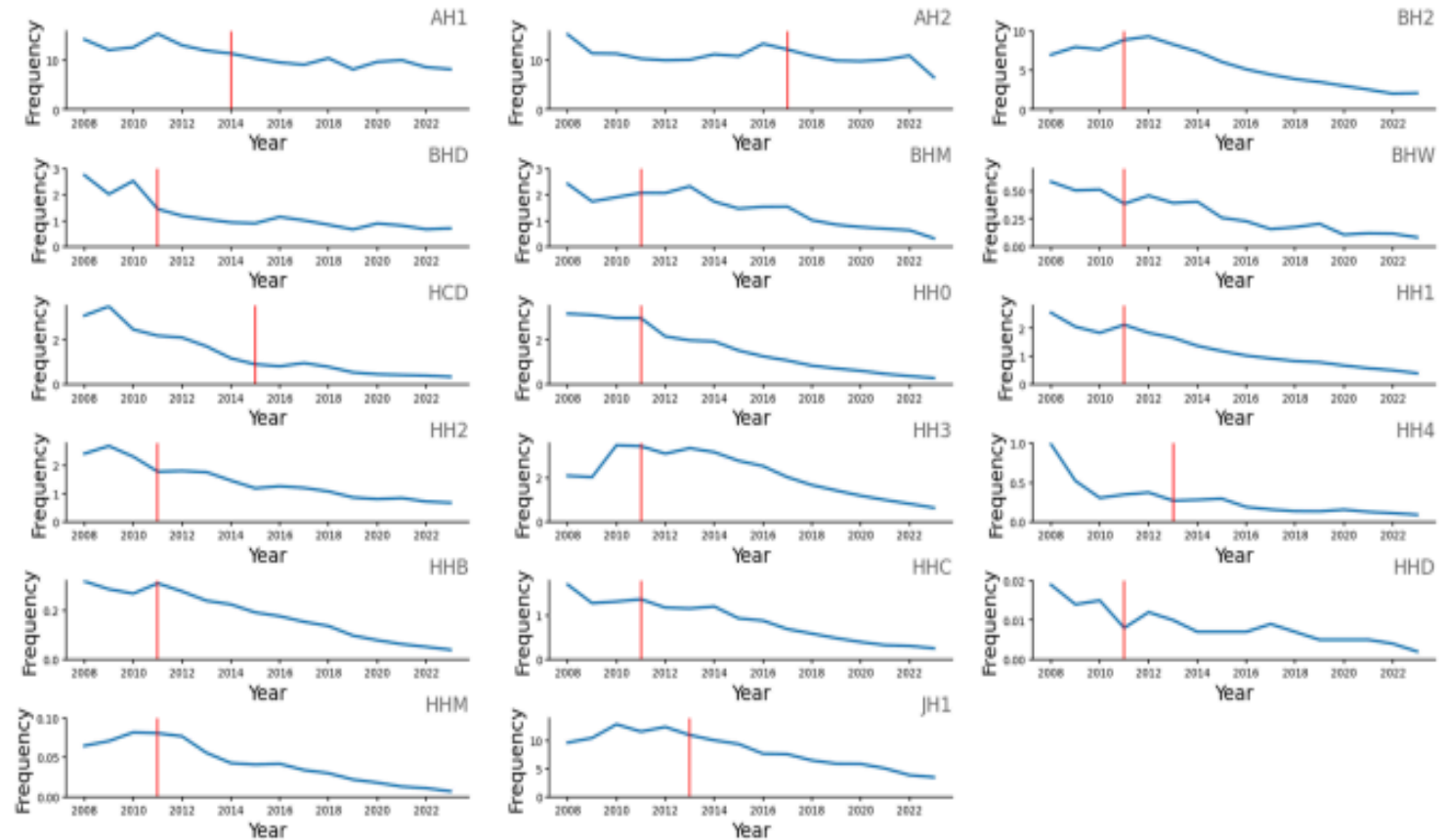
Unique ROH regions with unfavorable effects within and across traits identified

Candidate genes with negative association detected

Measures to manage the frequency of detrimental regions need to be considered

# Haplotype frequencies by year (2008–2023)

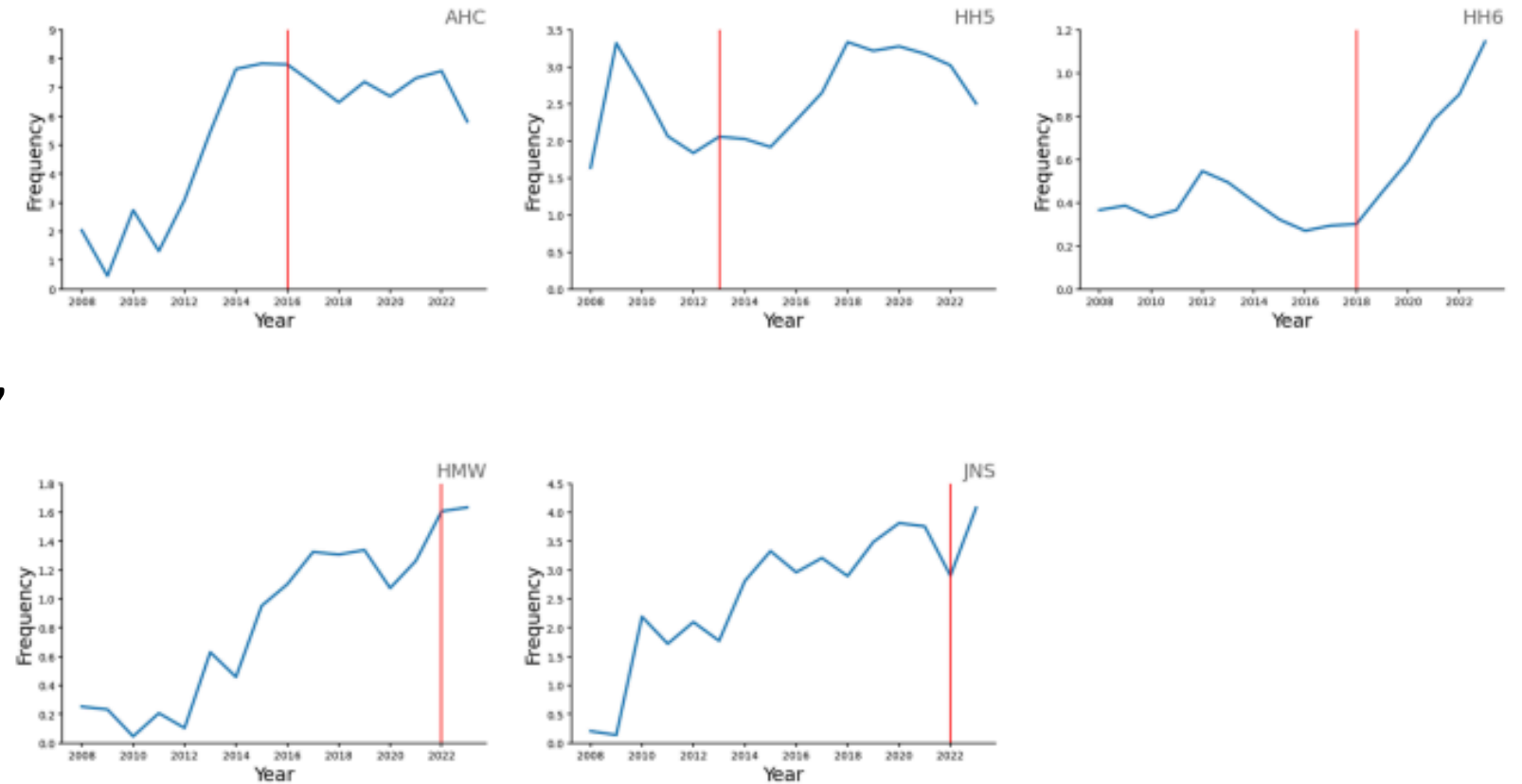
- Genomics facilitates identification of recessive defects
- Impacts: pregnancy loss, malformations, early mortality, etc.
- Examples: Cholesterol Deficiency, Brachyspina, etc.



**Figure 2.** Haplotype frequencies by year (2008–2023) for recessive genetic defects in US Ayrshire, Brown Swiss, Holstein, and Jersey cattle, tracked by the Council on Dairy Cattle Breeding, that are under control (Al-Khudhair et al., 2024a). Vertical red lines indicate the year each haplotype was first published. AH1 = Ayrshire Haplotype 1 (*PIRM/UBE3B*), AH2 = Ayrshire Haplotype 2 (*RPAP2*), BH2 = Brown Swiss Haplotype 2 (*TUBD1*), BHD = spinal dysmyelination (SDM; *SPAST*), BHM = spinal muscular atrophy (SMA; *KDSR/FVT1*), BHW = Weaver (*PNPLA8*), HCD = cholesterol deficiency (*APOB*), HH0 = Holstein Haplotype 0 (brachyspina/*FANCI*), HH1 = Holstein Haplotype 1 (*APAF1*), HH2 = Holstein Haplotype 2 (*IFT80*), HH3 = Holstein Haplotype 3 (*SMC2*), HH4 = Holstein Haplotype 4 (*GART*), HHB = bovine leukocyte adhesion deficiency (*ITGB2*), HHC = complex vertebral malformation (*SLC35A3*), HHD = deficiency of uridine monophosphate synthase (*UMPS*), HHM = mulefoot (syndactyly; *LRP4*), and JH1 = Jersey haplotype 1 (*CWC15*). The year the haplotype test was introduced is not the time of initial discovery of a genetic defect; some defects were discovered decades before haplotype tests became available.

# Haplotype frequencies by year (2008–2023)

- Higher homozygosity unmasks recessive defects
- Impacts: pregnancy loss, malformations, early mortality, etc.
- Examples: Cholesterol Deficiency, Brachyspina, etc.



**Figure 3.** Haplotype frequencies by year (2008–2023) for recessive genetic defects in US Ayrshire, Holstein, and Jersey cattle, tracked by the Council on Dairy Cattle Breeding, which have unfavorable trends (Al-Khudhair et al., 2024a). Vertical red lines indicate the year the haplotype test associated with each defect was first published. AHC = Ayrshire Haplotype C (*CHRNA1*), HH5 = Holstein Haplotype 5 (*TFB1M*), HH6 = Holstein Haplotype 6 (*SDE2*), HMW = early-onset muscle weakness (*CACNA1S*), and JNS = Jersey neuropathy with splayed forelimbs (*UCHL1*). The year the haplotype test was introduced is not the time of initial discovery of a genetic defect; some defects were discovered decades before haplotype tests became available.

"We don't have a way to quickly identify if a bull has a bad mutation we don't know about"

"Are our top 10 and 100 lists doing us any favours?"

"My cows milk a lot already – do I really have to push them more?"

"The client is responsible for mating decisions, we just help them choose bulls"

"We can always crossbreed, but we haven't reached that level of pain yet"

"I sit down with my rep and he gives me a list to choose from"

"How do producers make decisions?"

"I don't want malformed calves"

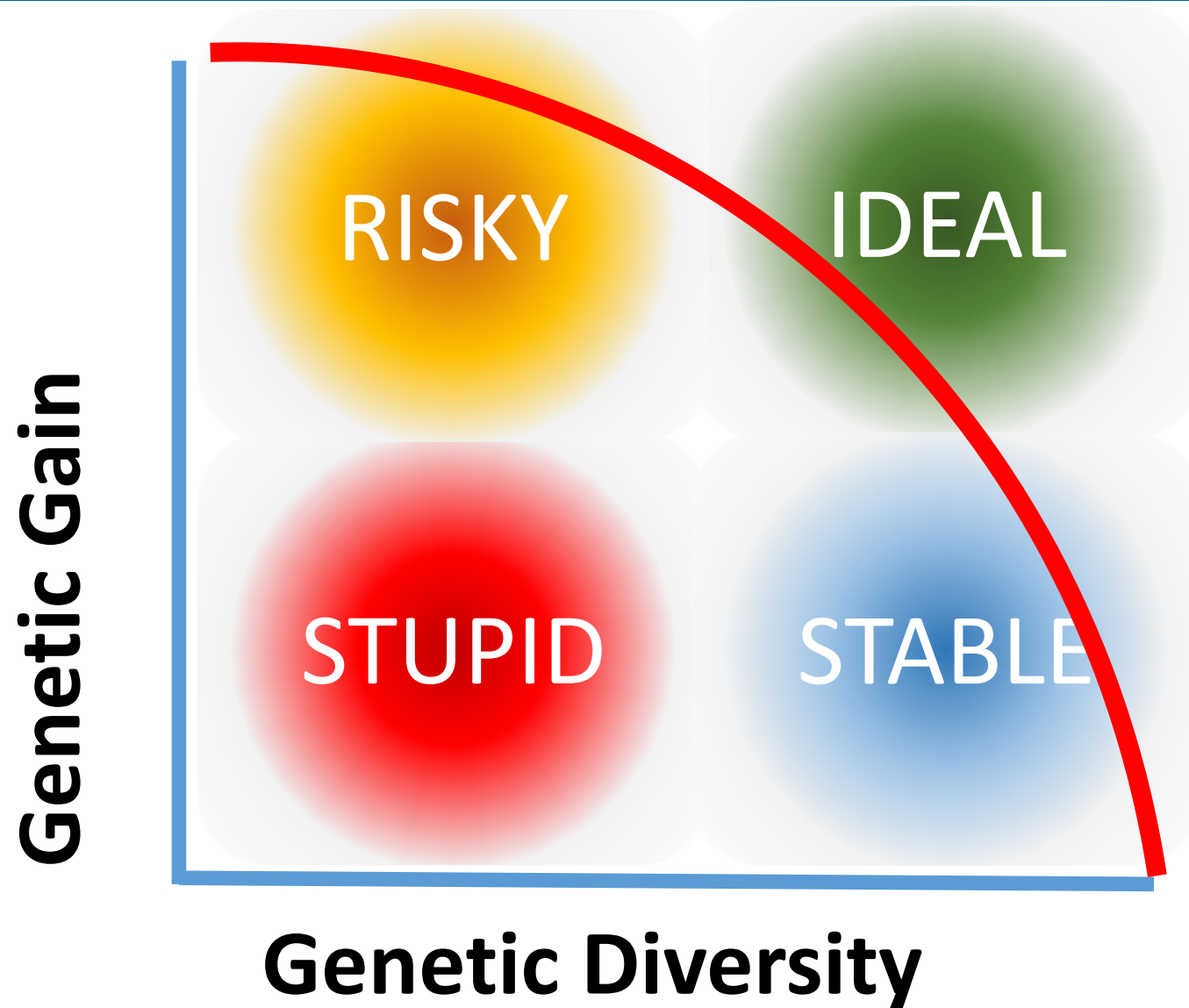
"How do we create demand for outcross bulls?"

"All bulls are good these days, I just need healthy cows"



**What should we do now?**

# The breeders dilemma



# Tools / strategies currently available



software for preventing inbreeding in livestock



All



Images



News



Videos



Shopping



More

Tools

About 127,000 results (0.47 seconds)

## Reality:

- Optimal Contribution Selection rarely used in practice
- Top lists dominate conversations / decisions / semen tanks
- "outcrosses" interesting, but generally unpopular
- Cross breeding, etc. mainly seen as "Experimenting"

## Project

# Detecting detrimental genetics in the Canadian dairy herd

*The objective of this project is to develop a national strategy to rapidly identify, understand, and manage detrimental genetic mutations in the Canadian dairy herd. This research aims at improving detection and management of detrimental genetics in Canadian dairy cattle for greater efficiency and profitability.*

## Research team

**Christine Baes, University of Guelph**

Flavio Schenkel, University of Guelph; Filippo Miglior, Lactanet

## Funders

**Agriculture and  
Agri-Food Canada**

**DairyGen partners  
(DFC, Semex  
Alliance, Holstein  
Canada and  
Lactanet)**

**Budget: \$899,990**



Shelby Duggan  
(Undergrad)



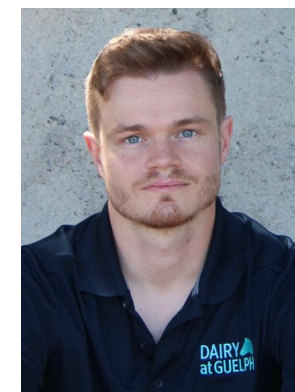
Gabriella Condello  
(PhD Student)



Bayode Makanjuola  
(Research Associate)



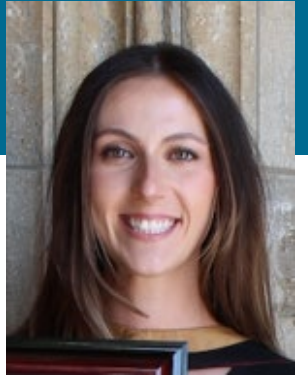
Ricarda Jahnel  
(Project Manager)



Colin Lynch

(Lactanet, Brian Wickam Young  
Person Exchange Program))

# Survey producers to understand needs



Survey to ensure the developed reporting system meets the needs of the Dairy Industry:

1. Understand current limitations affecting reporting genetic conditions
2. Experience with suspected genetic conditions and reporting
3. Outlook on genetic testing for health traits

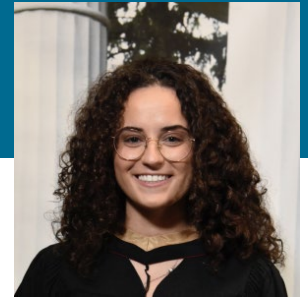
**Poster 2350: Developing a Rapid Monitoring System to Identify Genetic Conditions in Dairy Cattle**



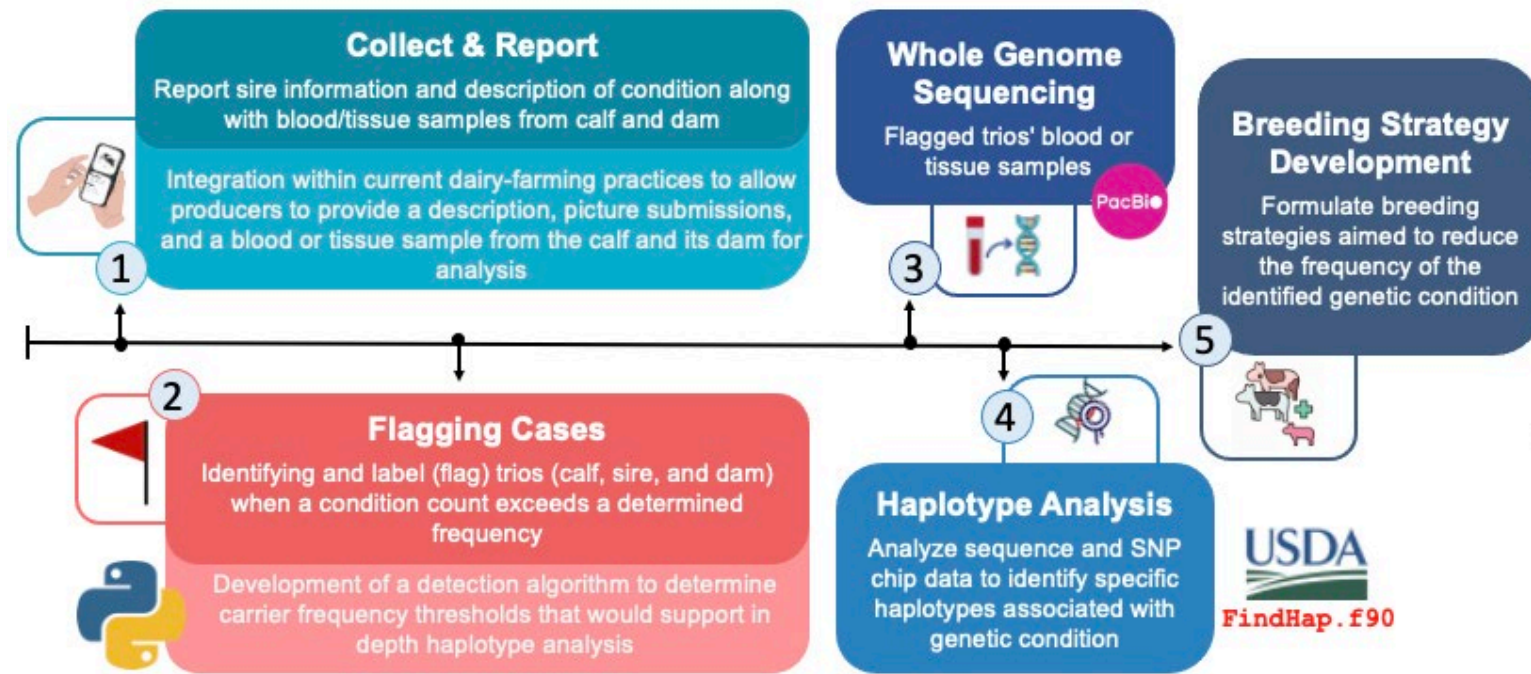
Survey approved by University of Guelph Research Ethics Board, Approval #64



# Developing a monitoring system



- Industry portal: report photo, description, tissue sample
- Not all congenital defects are genetic in nature: Auto-flag cases beyond threshold → WGS + haplotype mapping
- Breeding recommendations fed back to AI companies & producers



**Poster 2350: Developing a Rapid Monitoring System to Identify Genetic Conditions in Dairy Cattle**

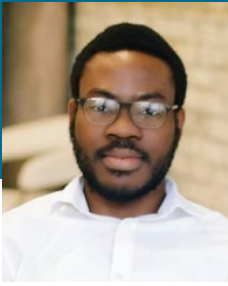


# Measuring relatedness

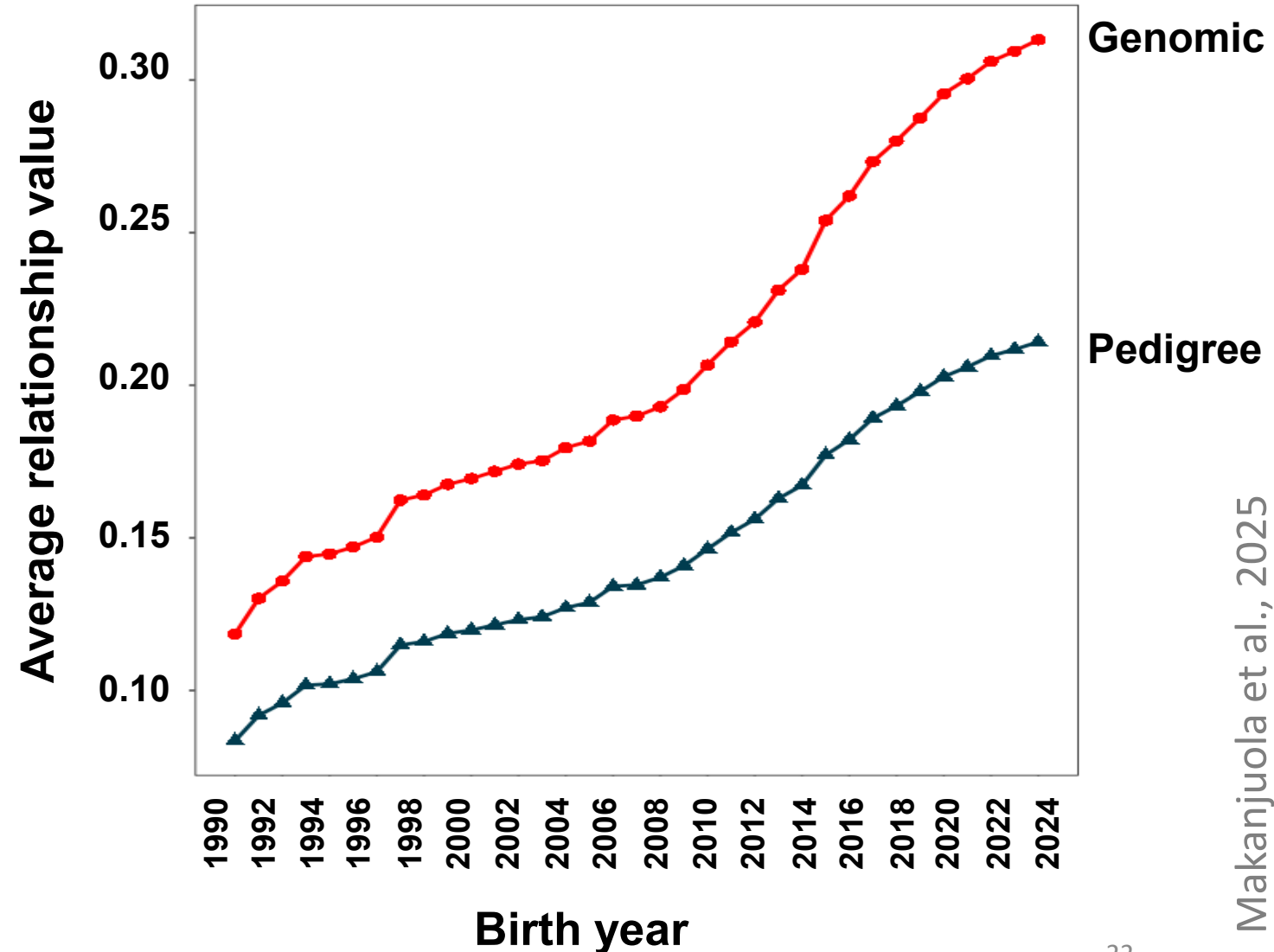


- **Relationship value (R-Value)** represents the percentage of DNA a sire shares with active females within a given population (*Van Doormaal et al., 2003*)
- Pedigree relationship (R) vs. Genomic relationship (GR)
- Expected Future Inbreeding (EFI/GFI) =  $\frac{1}{2} \times \text{R-Value}$
- Genomic R-Value  $\approx 1.5 \times \text{R-Value}$  in modern populations
- Diversity KPI for AI companies?
- Can calculate R-Values for a bull with:
  - A population of females
  - All females in a herd

# Measuring relatedness



- Average R-Value increasing over time
  - ~50 % since 2000
- Increase mirrors the observed increase in inbreeding
- Genomic R-Values  $\approx 1.5 \times R$  in modern populations

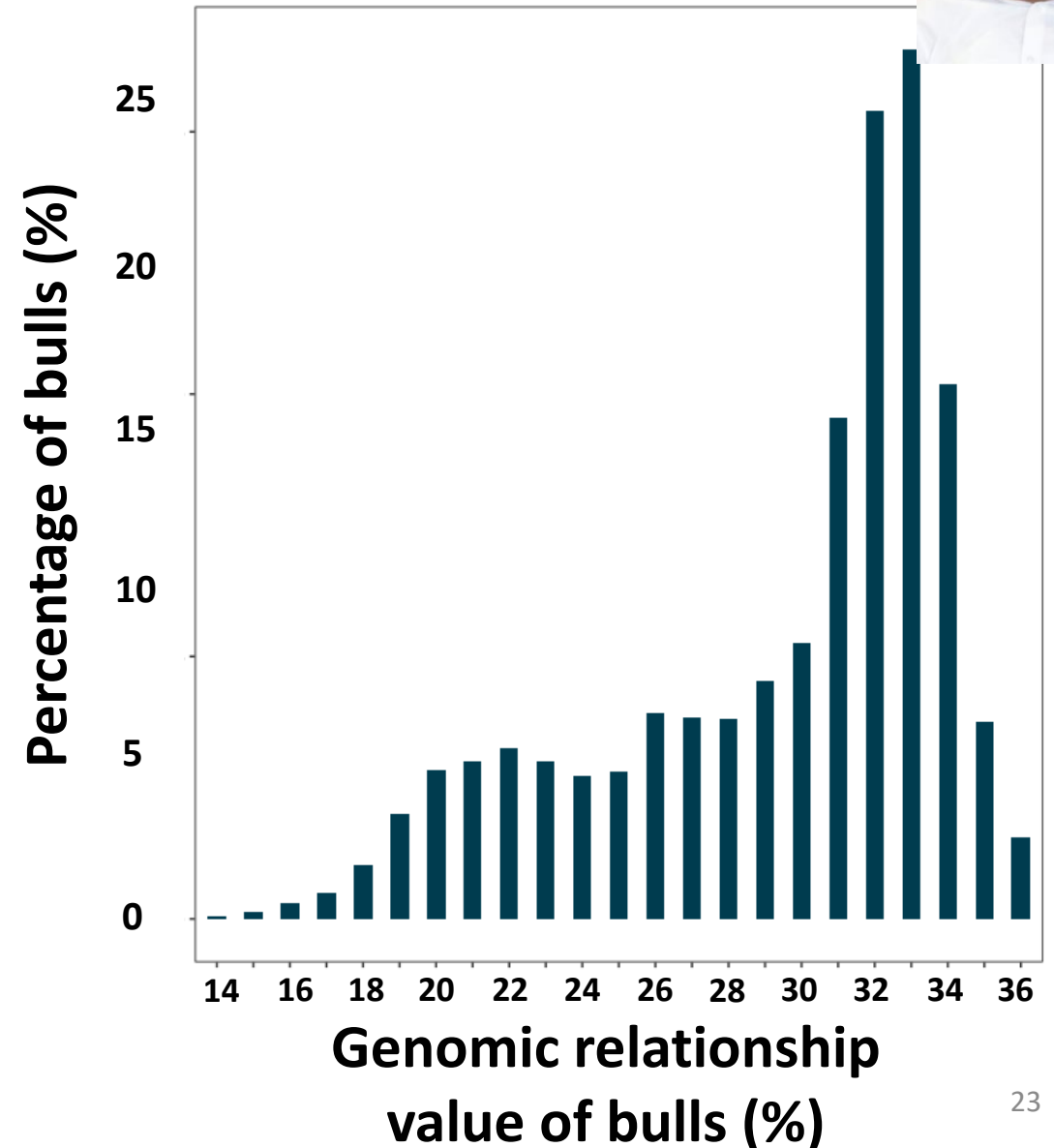


# A herd-level tool to guide sire selection?



- R-values: 10% to 25%
- Genomic R-values: 14% to 36%
- For top 100 LPI bulls:
  - R-values: 19% to 25%
  - Genomic R-values: 28% to 36%
- Relationship dashboards to guide sire selection
- Encourage AI companies to publish diversity KPIs

**Highlights opportunity to select less related bulls**



# Other approaches / safeguards

## Genetic conservation toolbox

- Gene banks: semen, embryos, somatic cells
- Strategic crossbreeding / introgression for rare alleles
- Monitor Effective Population Size ( $N_e$ ) to guide action

## Future-proofing with new technology?

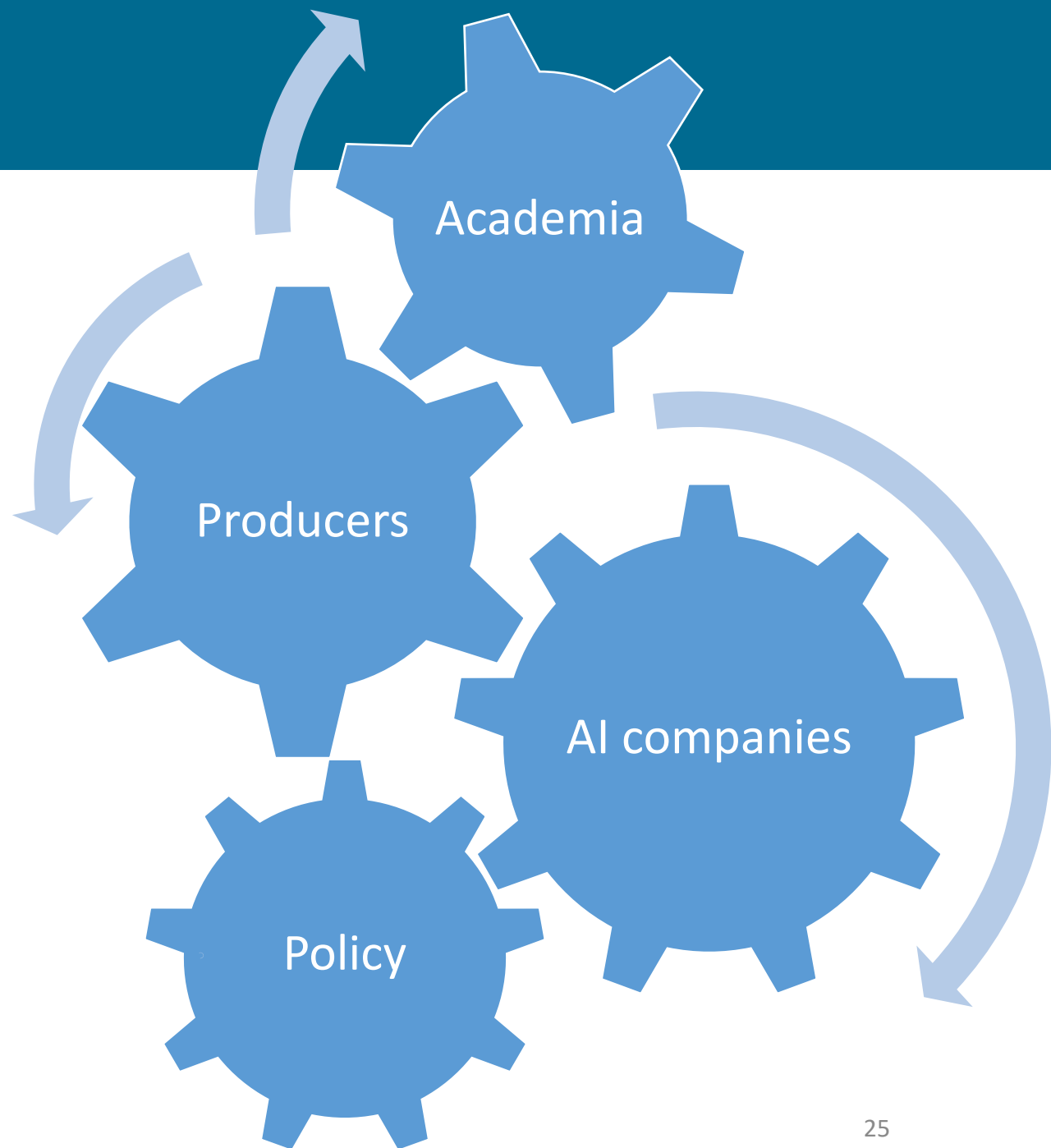
- CRISPR knock-in of favourable alleles w/out added Genomic R
- Targeted recombination to de-intensify ROH blocks
- Regulatory & societal considerations ahead

# Roadmap for stakeholders

**Understand and acknowledge** that this is **our collective responsibility**,

-but there are already great tools we can use to do better!

- Academia: develop technology, metrics & algorithms
- AI companies: publish diversity KPIs, diversify sire teams
- Producers & Associations: adopt mate-selection tools, report defects
- Policy: incentive programs for conservation & transparency (Gene banks: semen, embryos, somatic cells)



# Key messages

- Genetic diversity is our insurance for climate & market change
- Need to balance fast progress / short-term gain with sustainable improvement
- Genomics offers a lot of opportunities to manage genetic conditions
- Collaboratively Monitor • Manage • Innovate



# Acknowledgements

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- Lactanet
- Semex
- Holstein Canada

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- Dairy Cluster 4
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Agriculture and  
Agri-Food Canada

