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Genomic versus pedigree based inbreeding: 2 sides of the same coin (?)

Nicolò P.P. Macciotta, D. Lourenco, I. Misztal, D. Vicario, L. Degano, C. Dimauro, A. Cesarani

2025 ADSA annual meeting, Louisville()

Inbreeding

- ✓ Probability of two copies of the same random allele to be IBD
- ✓ Correlation between two uniting gametes
- ✓ Main consequence of the strong artificial selection, limited use of influential AI sires
- ✓ Not harmful by itself

✓ **Short term effects:**

- ✓ Inbreeding depression
- ✓ Reduction of fitness
- ✓ Genetic defects

✓ **Long term effects:**

- ✓ Reduction of genetic variability
- ✓ Reduction of genetic response



Impact of GS on inbreeding rate

- ✓ GS expected to decrease inbreeding rate per generation because the best individual animals rather than families with the best genes are selected
- ✓ Increase of the inbreeding rate due to GS (Guinan et al., 2023)
- ✓ More weight on parent average in selecting young animals
- ✓ Shorter generation intervals



Impact of GS on inbreeding rate



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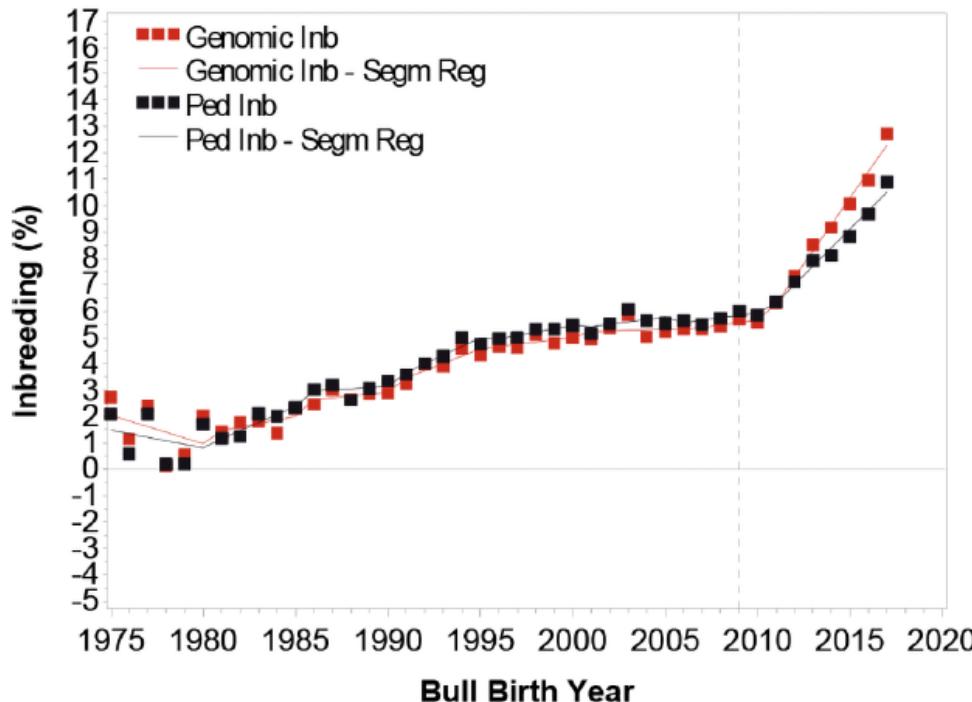
Changes in genetic trends in US dairy cattle since the implementation of genomic selection

F. L. Guinan,^{1*} G. R. Wiggans,² H. D. Norman,² J. W. Dürr,² J. B. Cole,³ C. P. Van Tassell,⁴
I. Misztal,¹ and D. Lourenco¹

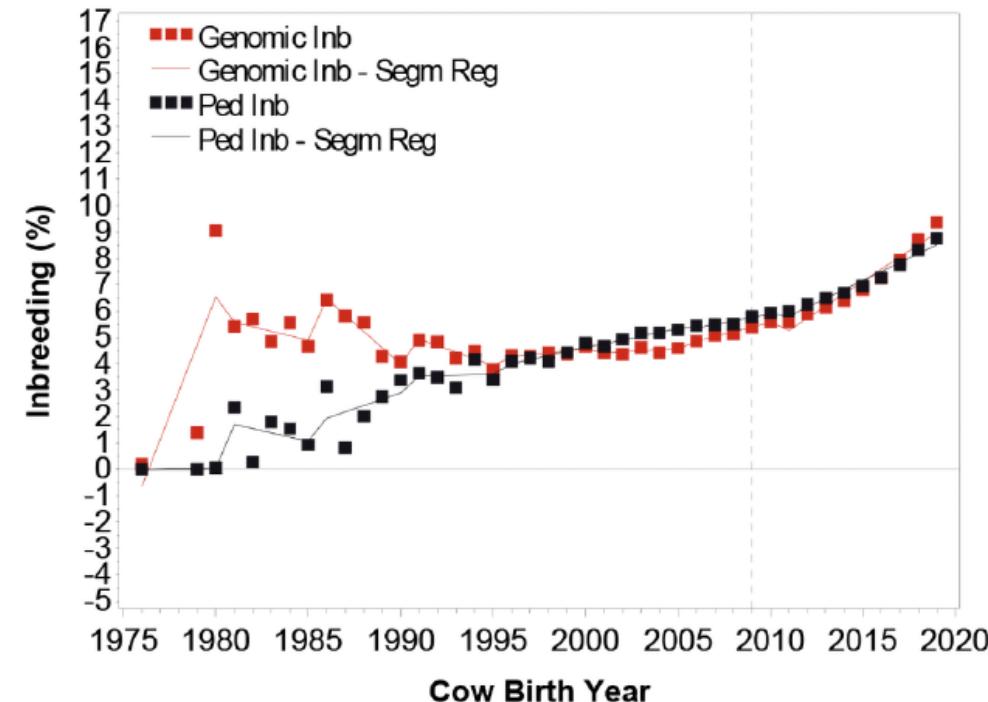
Guinan et al.: CHANGES IN GENETIC TRENDS IN US DAIRY CATTLE

1125

Holstein Bulls



Holstein Cows



Impact of GS on inbreeding measurement

PEDIGREE INBREEDING

- ✓ Traditional
- ✓ Based on pedigree registrations
- ✓ Pedigree depth
- ✓ Consensus on the procedure
- ✓ Expected inbreeding
- ✓ Genome-wide

GENOMIC INBREEDING

- ✓ Recent
- ✓ Based on molecular analyses
- ✓ Different metrics and approaches (**G**, ROH)
- ✓ Realized inbreeding
- ✓ Genome-wide, chromosome-wide, and regional
- ✓ Age of inbreeding



US Holstein



https://www.holsteinusa.com/holstein_breed/breedhistory.html

- ✓ 34k genotyped bulls
- ✓ 42,500 SNPs
- ✓ 1 M individuals in **A**

ITA Simmental



<http://dualbreeding.com/it/razze>

- ✓ 19k genotyped cows and bulls
- ✓ 42,100 SNPs

ITA Rendena



<http://dualbreeding.com/it/razze>

- ✓ 2.5k genotyped cows and bulls
- ✓ 120,000 SNPs



Considered metrics

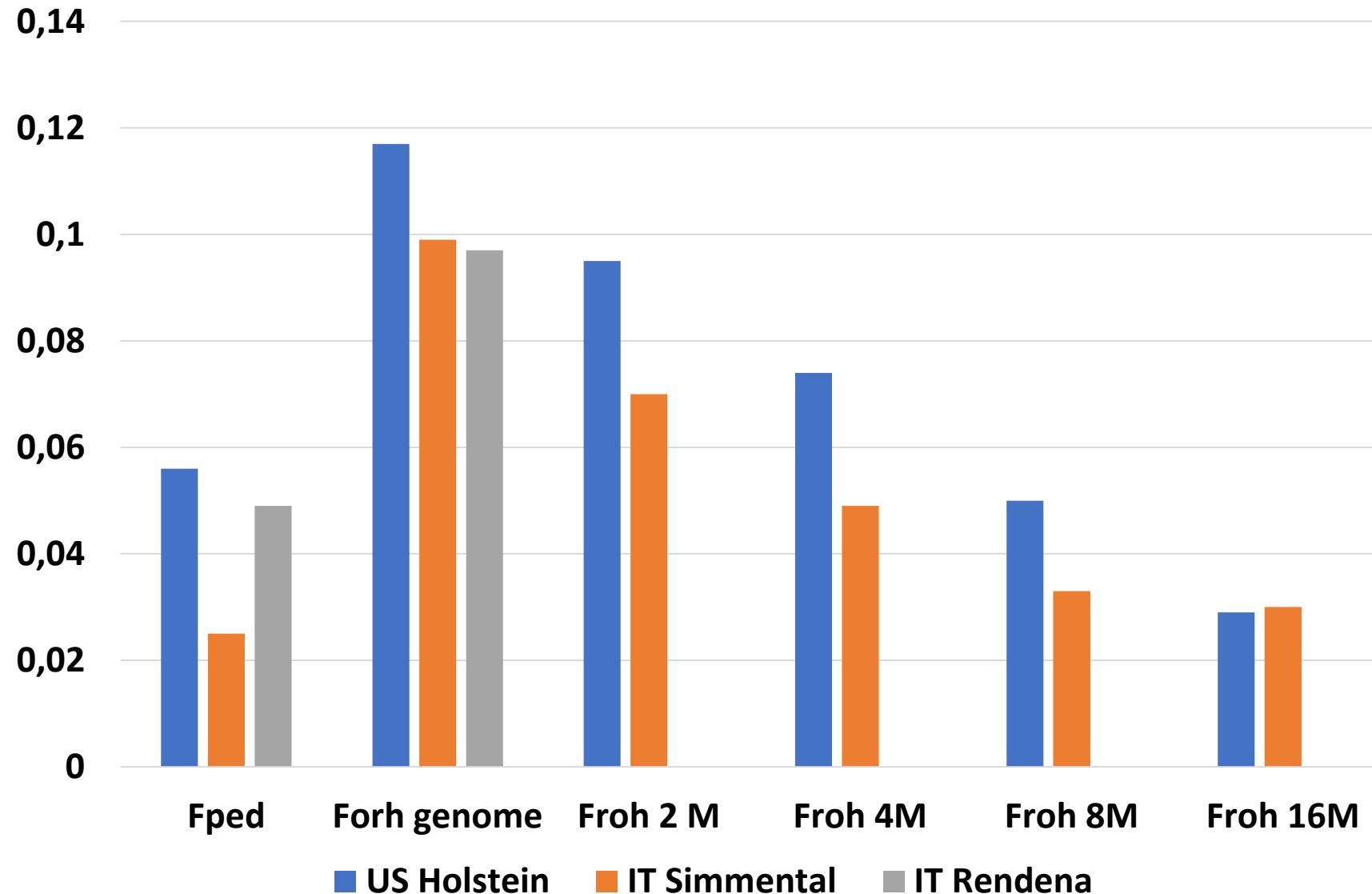
- ✓ Pedigree based inbreeding (F_{PED}), recursive formula (Aguilar and Misztal, 2008)
- ✓ **G** based inbreeding coefficient using observed (F_{GRM}) or fixed at 0.5 ($F_{GRM0.5}$) frequencies (Van Raden 2008)
- ✓ Whole genome ROH based inbreeding coefficient (F_{ROH})
- ✓ Whole genome ROH based inbreeding coefficient calculated with different minimum length classes (2, 4, 8, 16 Mb)

Technical issues in ROH based methods

- ✓ SNP beadchip
- ✓ Panel densities
- ✓ Genotyping errors
- ✓ SNP editing
- ✓ Imputation
- ✓ ROH settings



Pedigree based and genomic inbreeding



Relationship among different inbreeding metrics in Italian Simmental

	F_{PED}	F_{GRM}	$F_{GRM\ 0.5}$	F_{ROH}	$F_{ROH\ 2Mb}$	$F_{ROH\ 4Mb}$	$F_{ROH\ 8Mb}$	$F_{ROH\ 16Mb}$
F_{PED}	1							
F_{GRM}	0.18	1						
$F_{GRM\ 0.5}$	0.44	0.12	1					
F_{ROH}	0.57	0.17	0.85	1				
$F_{ROH\ 2Mb}$	0.58	0.20	0.81	0.98	1			
$F_{ROH\ 4Mb}$	0.59	0.26	0.78	0.95	0.97	1		
$F_{ROH\ 8Mb}$	0.57	0.29	0.73	0.89	0.92	0.96	1	
$F_{ROH\ 16Mb}$	0.49	0.25	0.61	0.77	0.8	0.84	0.89	1



Relationship among different inbreeding metrics in US Holstein

	F_{PED}	F_{ROH}	$F_{ROH\ 2Mb}$	$F_{ROH\ 4Mb}$	$F_{ROH\ 8Mb}$	$F_{ROH\ 16Mb}$
F_{PED}	1					
F_{ROH}	0.61	1				
$F_{ROH\ 2Mb}$	0.60	0.99	1			
$F_{ROH\ 4Mb}$	0.56	0.95	0.97	1		
$F_{ROH\ 8Mb}$	0.51	0.87	0.90	0.95	1	
$F_{ROH\ 16Mb}$	0.53	0.74	0.76	0.81	0.87	1

Impact of inbreeding on production, fertility, and health traits in German Holstein dairy cattle utilizing various inbreeding estimators

Julius Mugambe,^{1,*} Rana H. Ahmed,¹ Georg Thaller,¹ and Christin Schmidtmann^{1,2} 

Institute of Animal Breeding and Husbandry, Christian-Albrechts-University Kiel, 24098 Kiel, Germany

Table 3. Correlations between inbreeding coefficients from the different estimators

Item	F _{vr1}	F _{vr0.5}	F _{hat1}	F _{hat2}	F _{hat3}	F _{roh}	F _{roh >1}	F _{roh >2}	F _{roh >4}	F _{roh >8}	F _{roh >16}
F _{ped}	0.05	0.51	0.24	0.28	0.44	0.41	0.40	0.41	0.41	0.39	0.39
F _{vr1}		0.46	0.83	-0.13	0.85	0.32	0.32	0.32	0.33	0.36	0.34
F _{vr0.5}			-0.01	0.76	0.83	0.94	0.94	0.94	0.93	0.88	0.75
F _{hat1}				-0.63	0.51	-0.15	-0.15	-0.15	-0.13	-0.07	-0.02
F _{hat2}					0.35	0.82	0.82	0.82	0.79	0.72	0.58
F _{hat3}						0.72	0.72	0.72	0.72	0.71	0.62
F _{roh}							1.00	1.00	0.98	0.92	0.78
F _{roh(<2)}								1.00	0.98	0.92	0.78
F _{roh(2–4)}									0.98	0.92	0.78
F _{roh(4–8)}										0.94	0.79
F _{roh(8–16)}											0.85



| Relationships between different F metrics in cattle

- ✓ Medium - large correlations (>0.5) between Pedigree-based and ROH based F.
- ✓ F based on Grm with 0.5 allelic frequencies showed largest correlations with other metrics
- ✓ F based on Grm with actual allelic frequencies less related to other metrics
- ✓ Similar pattern in selected, dual purpose and local breeds



Exploring inbreeding metrics covariance structure



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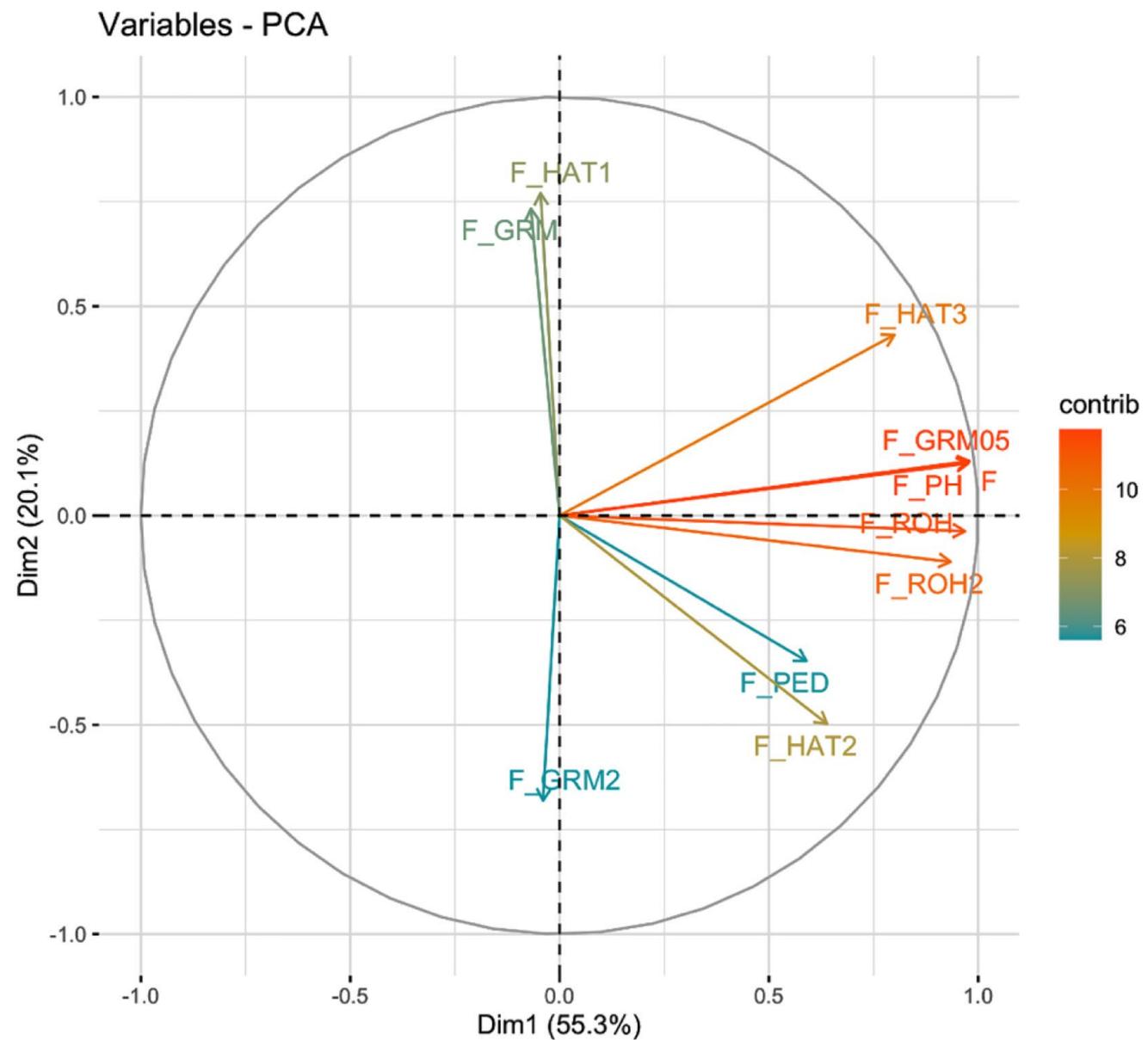
Genomic inbreeding coefficients using imputed genotypes: Assessing different estimators in Holstein-Friesian dairy cows

Christos Dadousis,^{1*} Michela Ablondi,¹ Claudio Cipolat-Gotet,¹ Jan-Thijs van Kaam,² Maurizio Marusi,² Martino Cassandro,² Alberto Sabbioni,² and Andrea Summer¹

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PCA of different measures of inbreeding in Italian Simmental

Metrics	Prin1	Prin2	Prin3	Prin4	Prin5	Prin6	Prin7	Prin8
F _{PED}	0.29	0.05	-0.83	0.46	0.12	0.02	-0.01	-0.01
F _{GRM0.5}	0.31	-0.33	0.47	0.57	0.48	-0.13	0.06	0.04
F _{GRM}	-0.12	0.92	0.23	0.30	0.06	-0.01	0.01	0.01
F _{ROH}	0.41	-0.04	0.16	0.12	-0.38	0.44	-0.46	-0.48
F _{ROH 2MB}	0.42	0.01	0.09	0.01	-0.39	0.21	0.12	0.78
F _{ROH 4MB}	0.42	0.05	0.03	-0.12	-0.26	-0.28	0.71	-0.39
F _{ROH 8MB}	0.41	0.10	-0.02	-0.29	0.04	-0.69	-0.51	0.07
F _{ROH 16MB}	0.37	0.14	-0.04	-0.51	0.62	0.44	0.08	-0.01
Eigenvalue (%)	69	13	8	6	3	0.008	0.001	0.001

PCA of different measures of inbreeding in US Holstein

Metrics	Prin1	Prin2	Prin3	Prin4	Prin5	Prin6
F _{PED}	0.30	0.92	0.26	-0.03	-0.01	0.01
F _{ROH}	0.43	-0.01	-0.41	0.36	0.46	0.56
F _{ROH 2MB}	0.44	-0.03	-0.36	0.23	0.06	-0.79
F _{ROH 4MB}	0.44	-0.09	-0.22	-0.13	-0.82	0.25
F _{ROH 8MB}	0.43	-0.20	0.13	-0.80	0.35	-0.08
F _{ROH 16MB}	0.39	-0.33	0.76	0.41	-0.02	0.01
Eigenvalue (%)	82	11	5	1	0.7	0.3

Factor pattern of different measures of inbreeding in Italian Simmental

Metrics	Factor1	Factor2	Factor3	Factor4	Factor5
F _{PED}	0.34	0.92	0.11	-0.01	0.14
F _{GRM0.5}	0.43	0.11	0.88	-0.3	0.12
F _{GRM}	0.01	-0.01	-0.08	0.99	0.02
F _{ROH}	0.86	0.24	0.40	-0.01	0.13
F _{ROH 2MB}	0.89	0.25	0.31	-0.01	0.19
F _{ROH 4MB}	0.87	0.27	0.24	0.01	0.29
F _{ROH 8MB}	0.80	0.26	0.18	0.02	0.47
F _{ROH 16MB}	0.62	0.20	0.15	0.04	0.73
Explained variance (%)	58	12	11	10	9

Factor pattern of different measures of inbreeding in US Holstein

Metrics	Factor1	Factor2	Factor3	Factor4	Factor5
F _{PED}	0.28	0.16	0.94	0.04	0.01
F _{ROH}	0.88	0.35	0.31	0.01	-0.07
F _{ROH 2MB}	0.87	0.38	0.30	0.04	-0.01
F _{ROH 4MB}	0.83	0.45	0.28	0.12	0.14
F _{ROH 8MB}	0.69	0.59	0.23	0.34	0.01
F _{ROH 16MB}	0.43	0.88	0.18	0.02	0.01
Explained variance (%)	49	27	21	2	1

PCA results

- ✓ A first PC is equally related to all inbreeding metrics
- ✓ Subsequent PC related to Fped, FGRM, and to ROH based inbreeding
- ✓ ROH based metrics are partitioned by PCA according to minimum length (inbreeding age?)

MFA results

- ✓ Factors represent the kind of inbreeding metrics
- ✓ Longer ROH factor (inbreeding age?)



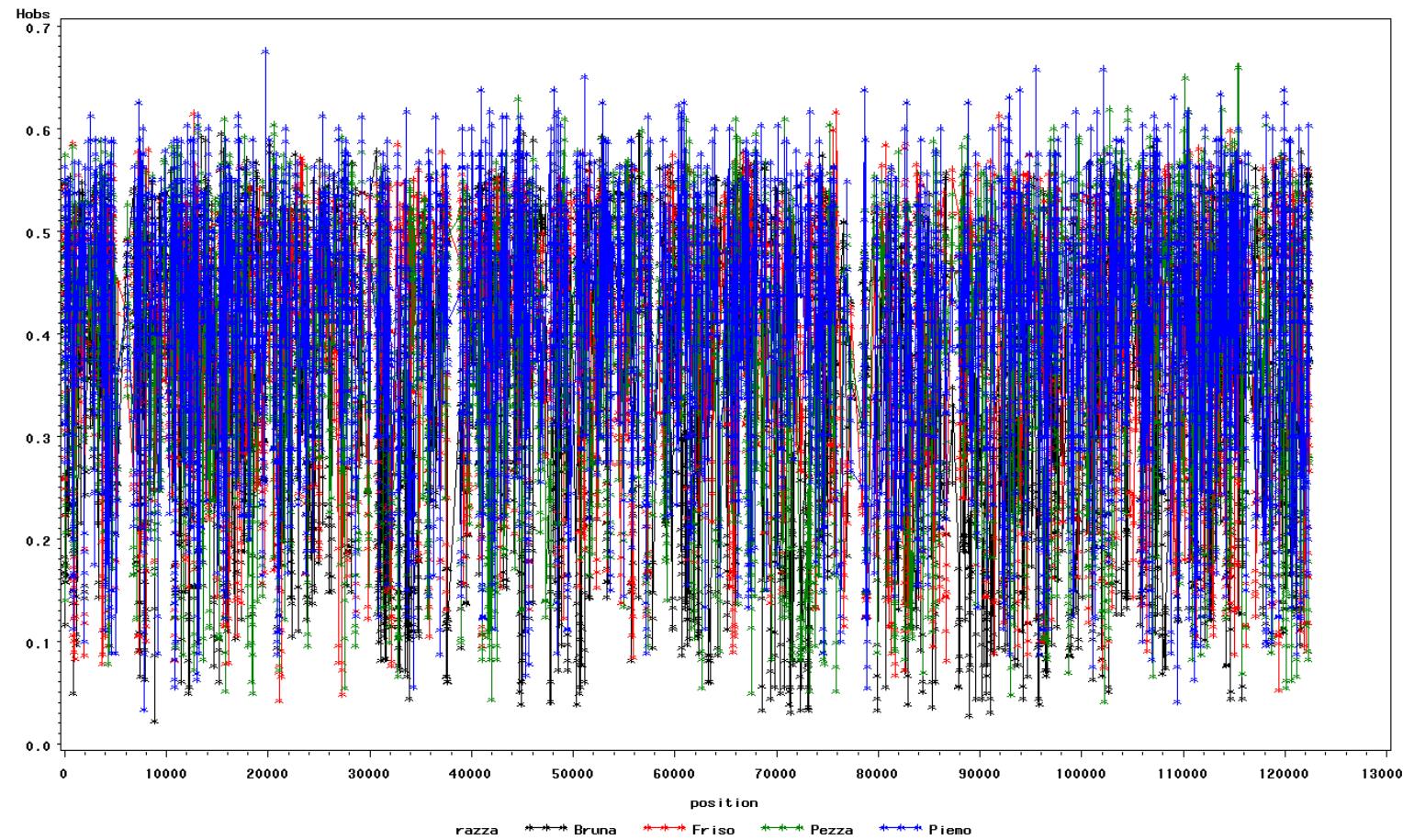
Regional inbreeding

- ✓ Molecular based inbreeding allows for estimation of chromosomal inbreeding
- ✓ the Inbreeding patterns vary between different regions of the genome
- ✓ Possibility of distinguish between favourable and unfavourable haplotypes
- ✓ Age of inbreeding may help in studying inbreeding depression (> for recent inbreeding)



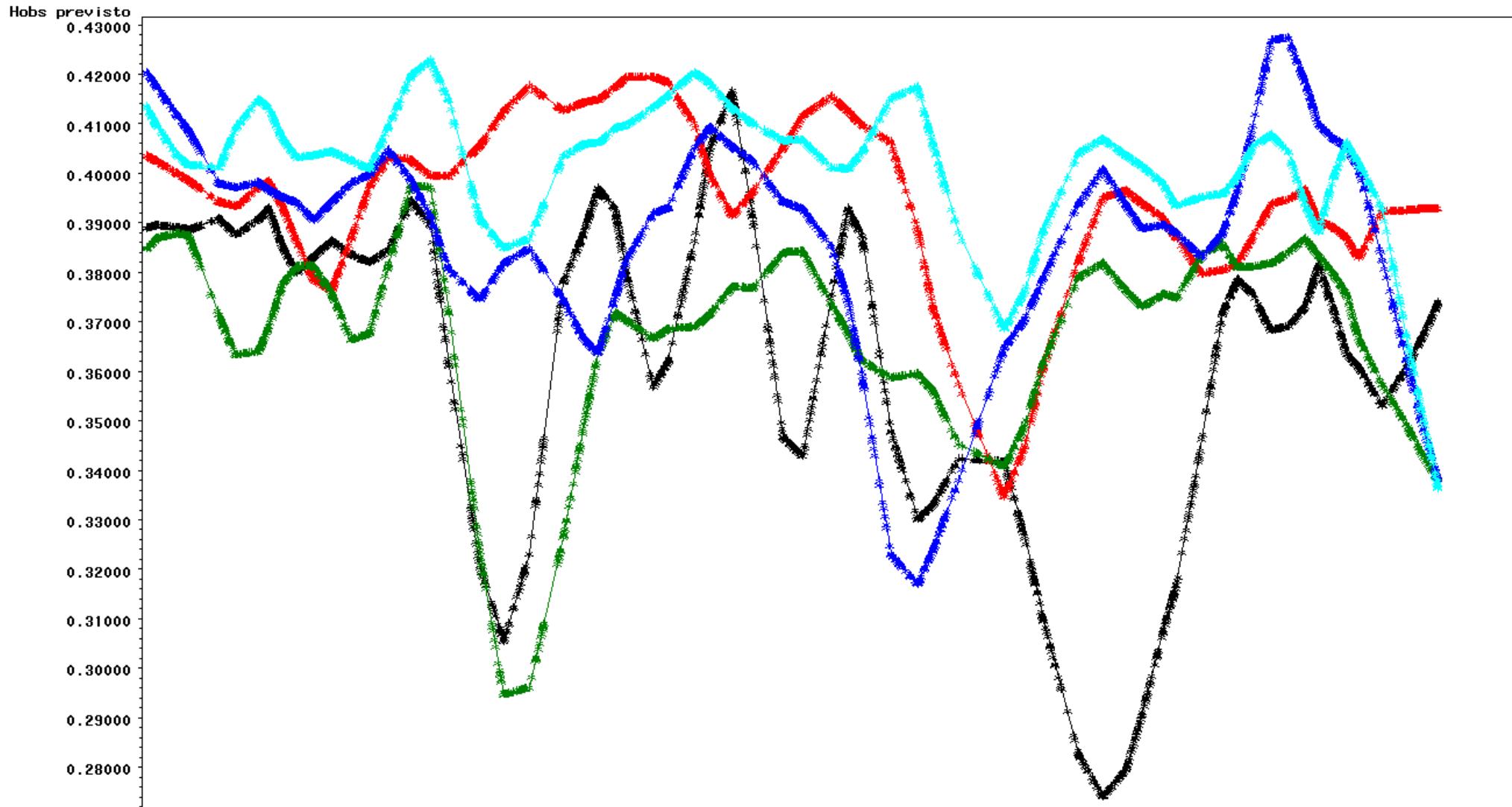
Accumulation (decrease) of homozygosity (heterozygosity)

eterozigosi osservata BTA6

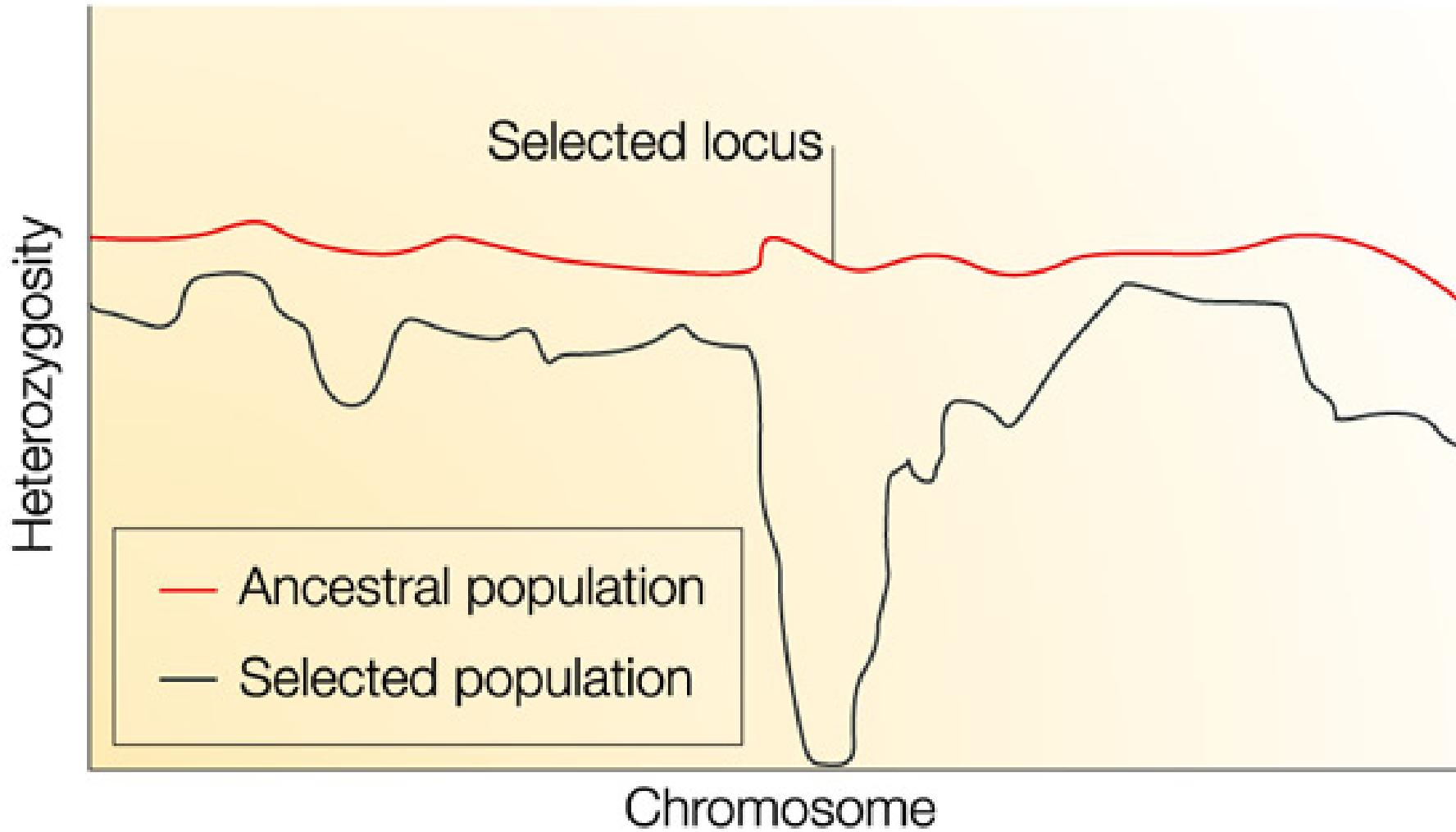


Accumulation (decrease) of homozygosity (heterozygosity)

eterozigosi osservata BTA6

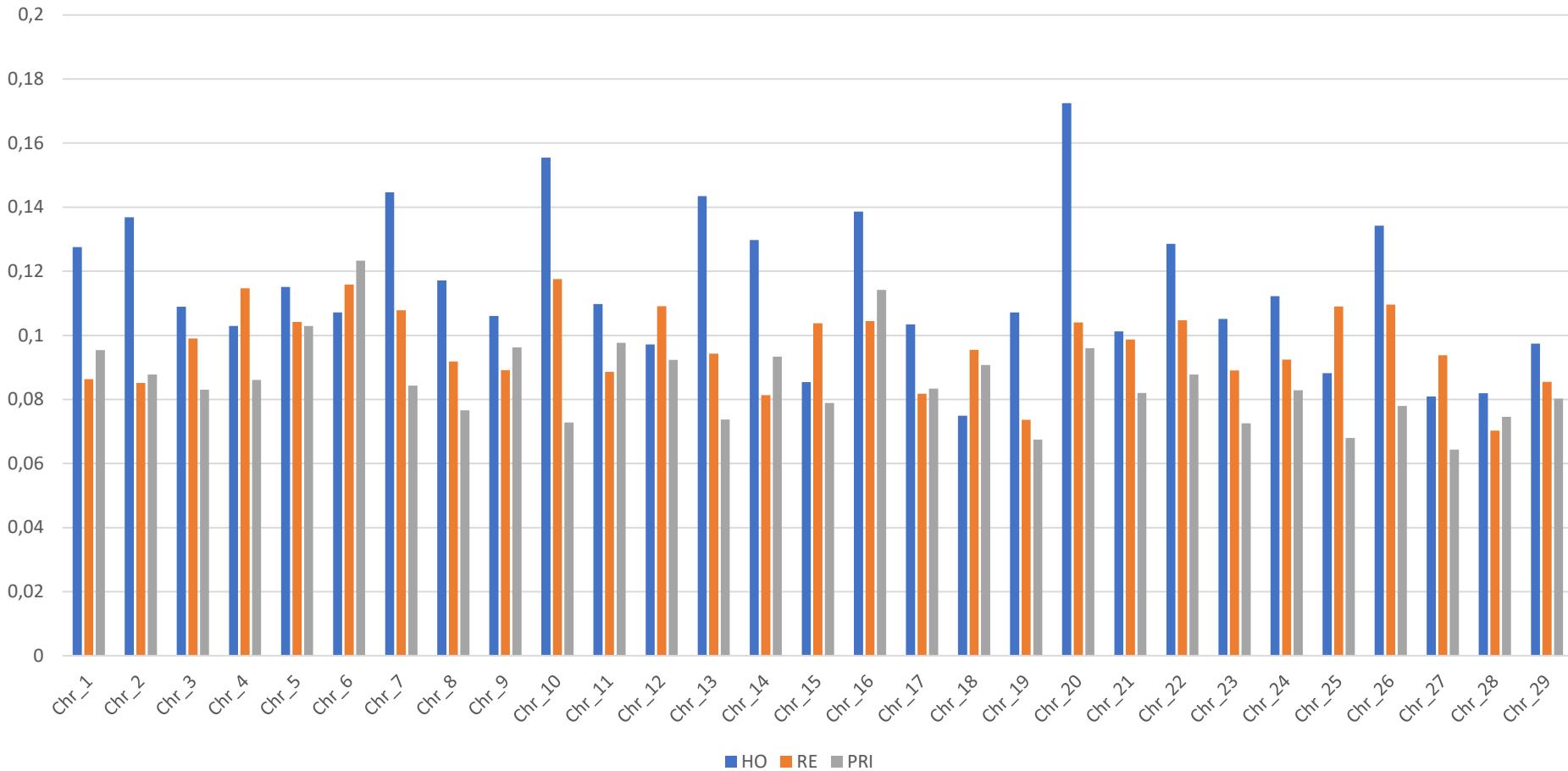


Anderson and Georges, 2004

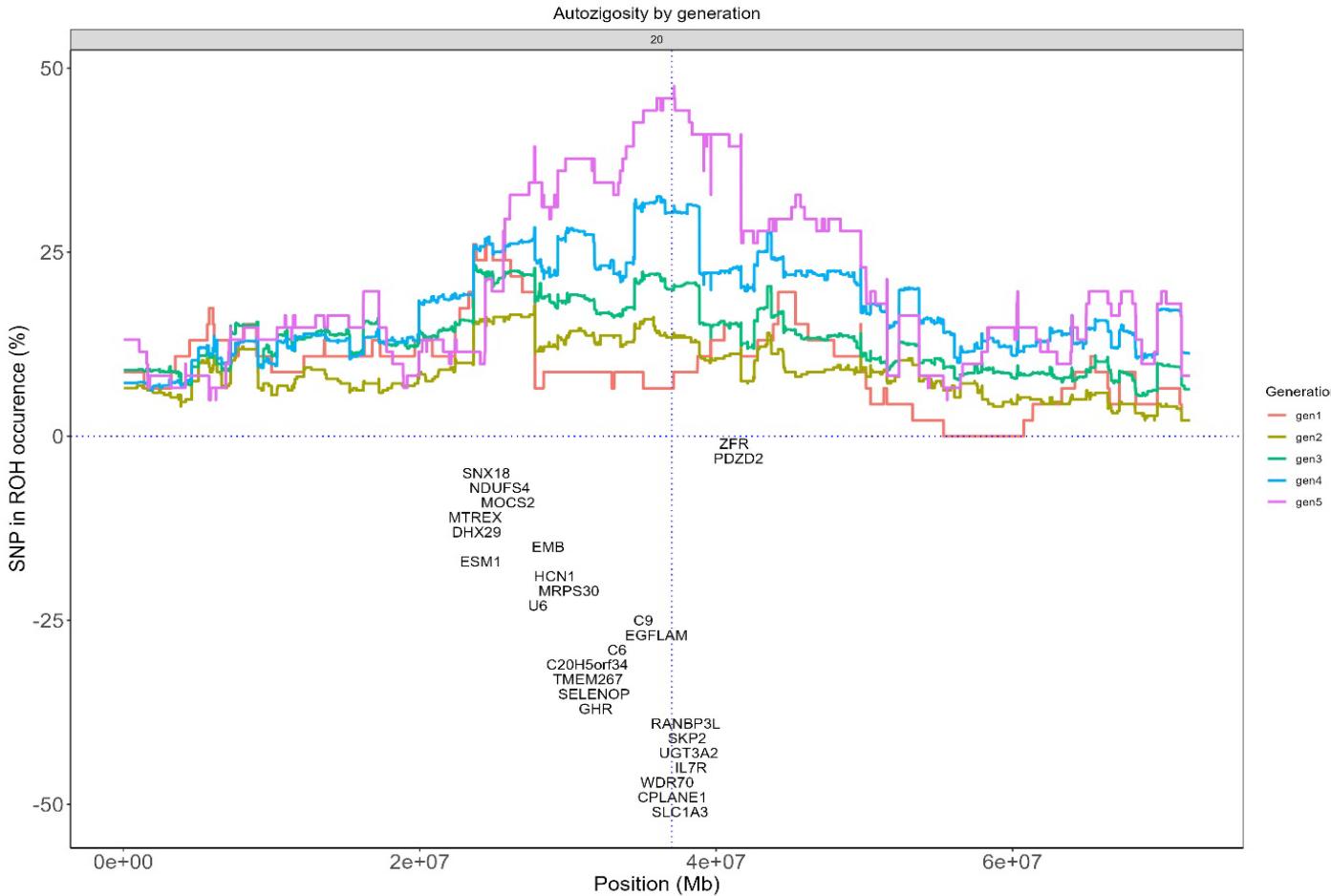


$$F_{ROH} = \frac{L_{ROH}}{L_{AUT}}$$

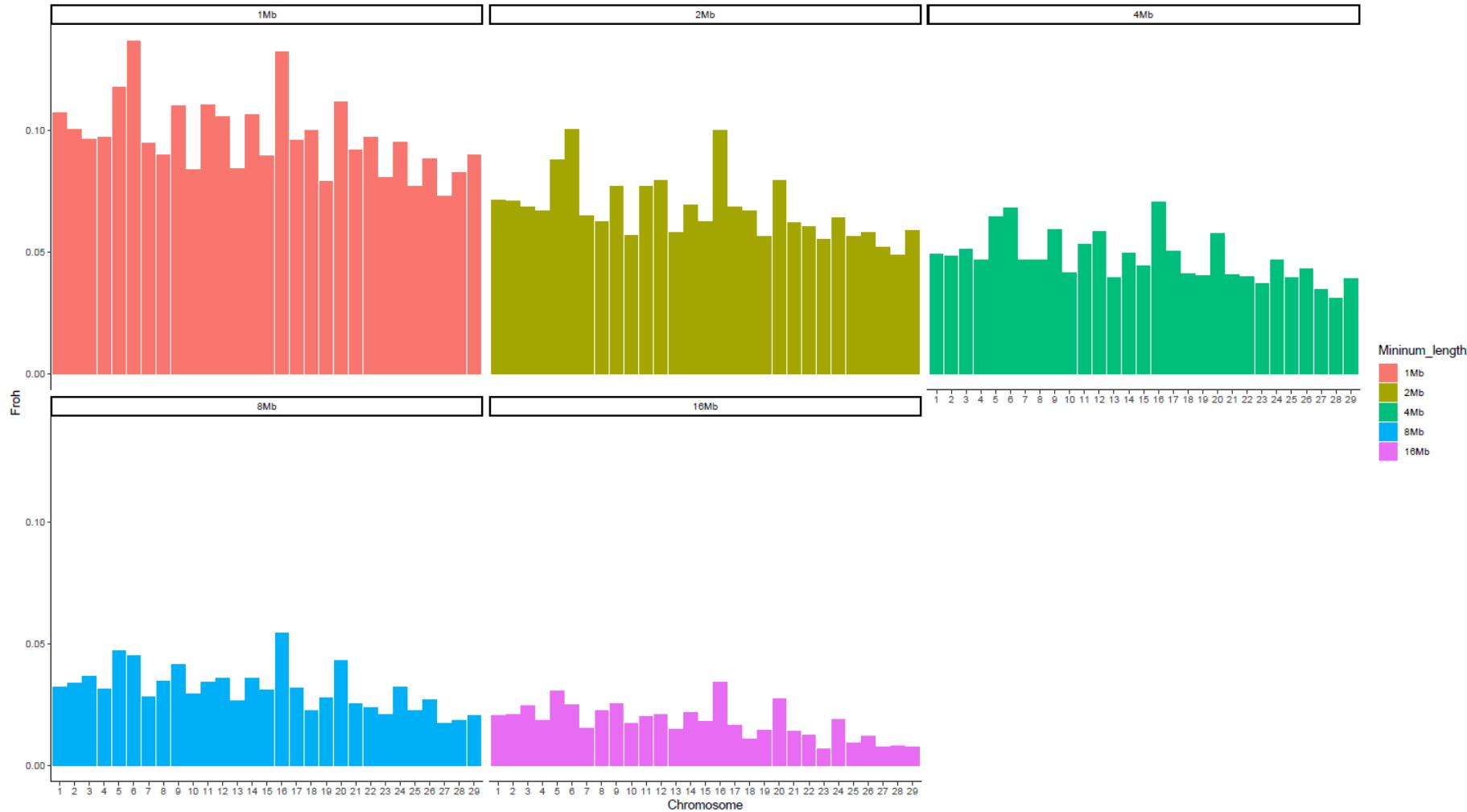
Genome distribution of FROH genome



Accumulation of genomic inbreeding across generations



Chromosomal F_{ROH} with different minimum lengths



Chromosomal inbreeding

- ✓ Different distribution across chromosome
- ✓ F_{ROH} based does not change according to minimum length
- ✓ Relationships with selection
- ✓ Accumulation across generations



Age of inbreeding and inbreeding depression

- ✓ Longer ROH (> 4Mb) contribute more to inbreeding depression to shorter ROH (Makanjuola et al., 2020; Scott et al.)
- ✓ More recent inbreeding contributes more than remote inbreeding to inbreeding depression
- ✓ Purging effect of selection may have eliminated deleterious variants

Inbreeding depression in Italian Simmental

	305 d MY	P value
Fped	-16.92	n.s.
Fgrm	-7.17	<0.05
Fgrm0.5	-9.20	n.s.
Froh1	-18.07	<0.05
Froh2	-19.36	<0.05
Froh4	-21.46	<0.05
Froh8	-22.60	<0.05
Froh16	-30.92	<0.05

DOES the G matrix contain other informations?



RESEARCH ARTICLE

Identification of key contributors in complex population structures

Markus Neuditschko^{1,2*}, Herman W. Raadsma², Mehar S. Khatkar², Elisabeth Jonas^{2,3}, Eike J. Steinig⁴, Christine Flury⁵, Heidi Signer-Hasler⁵, Mirjam Frischknecht^{1,6}, Ruedi von Niederhäusern¹, Tosso Leeb⁶, Stefan Rieder¹

- ✓ Selection of reference individuals for genotype imputation
- ✓ Identification of key ancestors through pedigree or genomic relationship information to maximize genetic diversity

DOES the G matrix contain other informations?



Contents lists available at ScienceDirect

Animal
The international journal of animal biosciences



Strategies for choosing core animals in the algorithm for proven and young and their impact on the accuracy of single-step genomic predictions in cattle



A. Cesarani ^{a,b,*}, M. Bermann ^b, C. Dimauro ^a, L. Degano ^c, D. Vicario ^c, D. Lourenco ^b, N.P.P. Macciotta ^a

$$G_{\text{CONTR}} = \sum_{i=1}^{i=k} \left(\frac{a_{ij}}{\sqrt{\lambda_i}} \right)^2$$

Best accuracies when animals with lower Gcontribution were included in the core.

Table 4

Correlations between the genomic estimated breeding values assessed using or not the algorithm for proven and young animals in the simulated and cattle datasets.

Scenario	Core definition	Simulated			Simmental
		Trait 1	Trait 2	Trait 3	Milkability
V90	Bottom	0.77	0.81	0.78	0.81
	Top	0.52	0.55	0.56	0.71
	Random	0.69	0.70	0.72	0.76
V95	Bottom	0.86	0.88	0.83	0.86
	Top	0.63	0.64	0.64	0.81
	Random	0.75	0.78	0.81	0.86
V98	Bottom	0.93	0.94	0.91	0.91
	Top	0.77	0.78	0.72	0.90
	Random	0.88	0.89	0.88	0.92
V99	Bottom	0.95	0.96	0.95	0.95
	Top	0.83	0.84	0.80	0.95
	Random	0.93	0.94	0.93	0.95

Abbreviations: Bottom = animals with lowest values of contribution to the genomic relationship matrix; Top = animals with highest values of contribution to the genomic relationship matrix; Random = animals randomly selected.



| Are genomic and pedigree inbreeding two sides of the same coin?

- ✓ Maybe
- ✓ Moderate large relationship between F_{PED} and F_{ROH}
- ✓ Genomic inbreeding provides several useful insights for managing inbreeding and diversity
- ✓ Evaluation of other indicators (e.g. Gcontr)
- ✓ Combination of the two approaches

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