Genomic prediction using information from multiple populations

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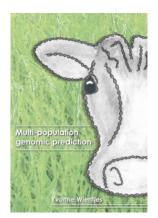




Who am I?

PhD - Wageningen University & Research

- Animal Breeding and Genomics
- `Multi-population genomic prediction'



Postdoc - Wageningen University & Research

- Genomic prediction for crossbred performance
- Long-term effects of Genomic Selection

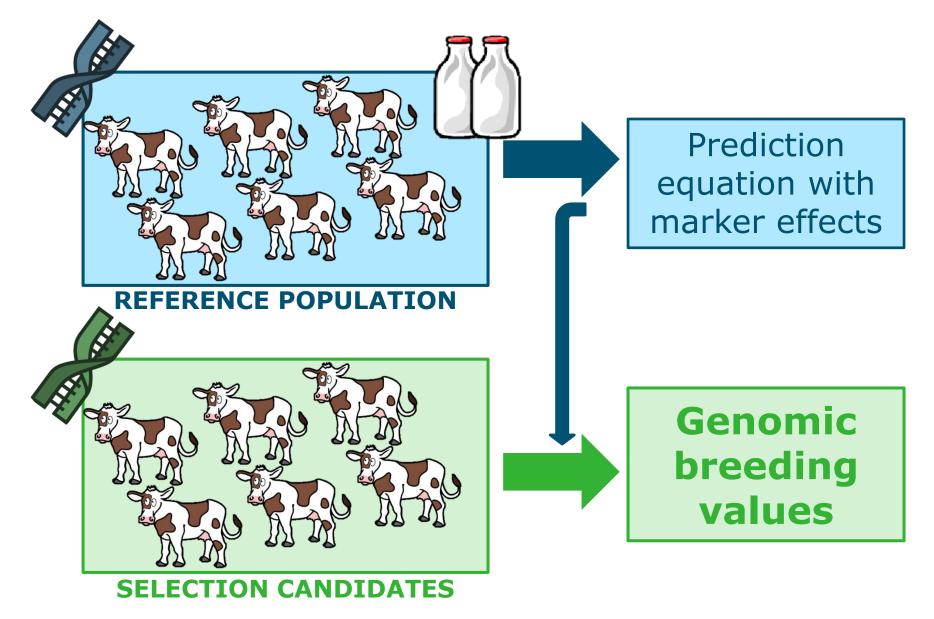




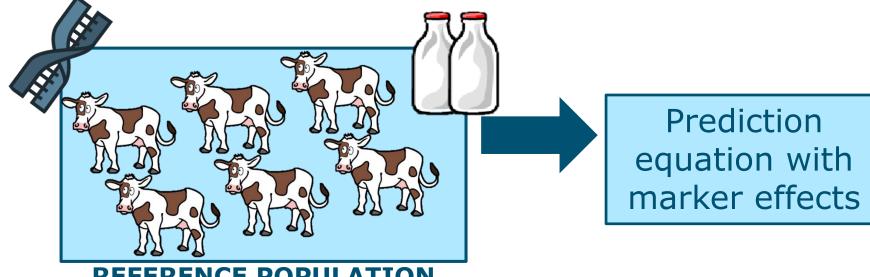




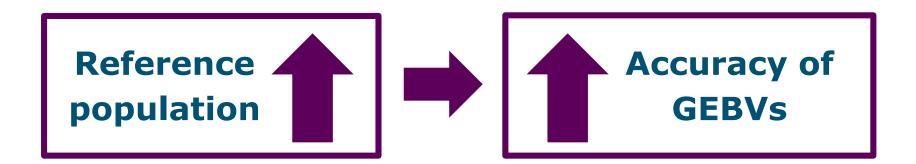
Genomic selection



Reference population



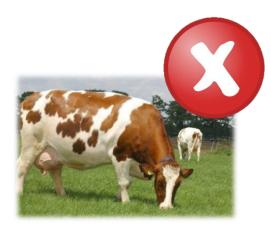
REFERENCE POPULATION





Size of reference population



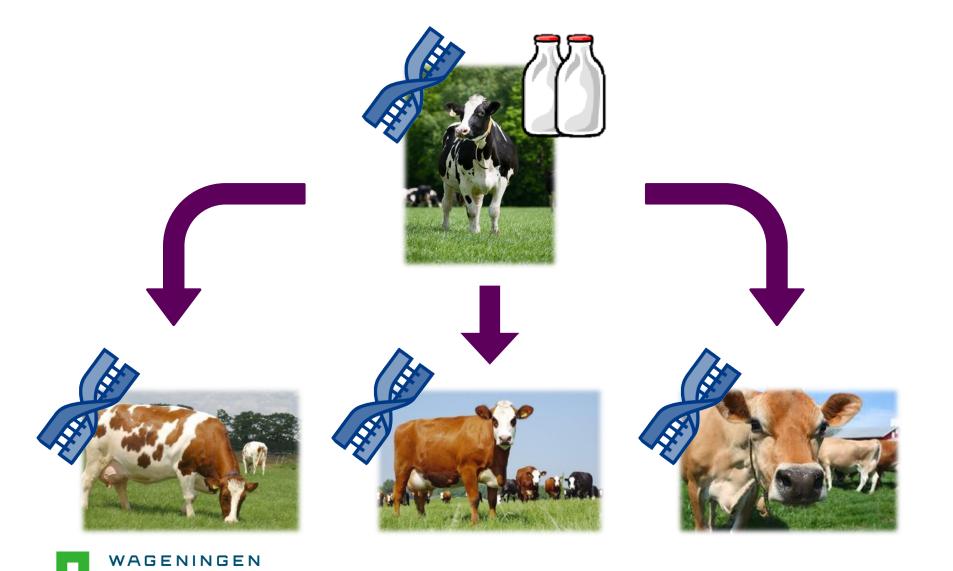








Use information across populations?



RESEARCH

Differences between populations

 Linkage disequilibrium between markers and causal variants

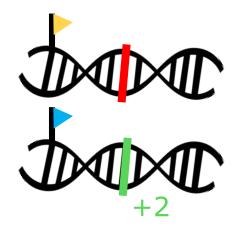


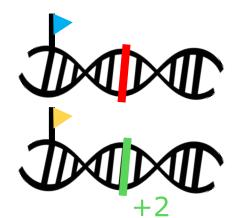
Differences between populations in LD

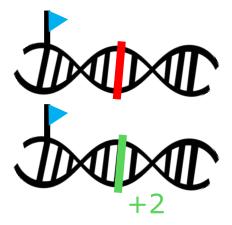














Differences between populations

 Linkage disequilibrium between markers and causal variants

- Allele frequencies of causal variants
- Effects of causal variants
 - Environment different
 - Non-additive effects
 - → Genetic correlation between populations



Differences between populations

 Linkage disequilibrium between markers and causal variants

- Allele frequencies of causal variants
- Effects of causal variants
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Close family relationships are absent

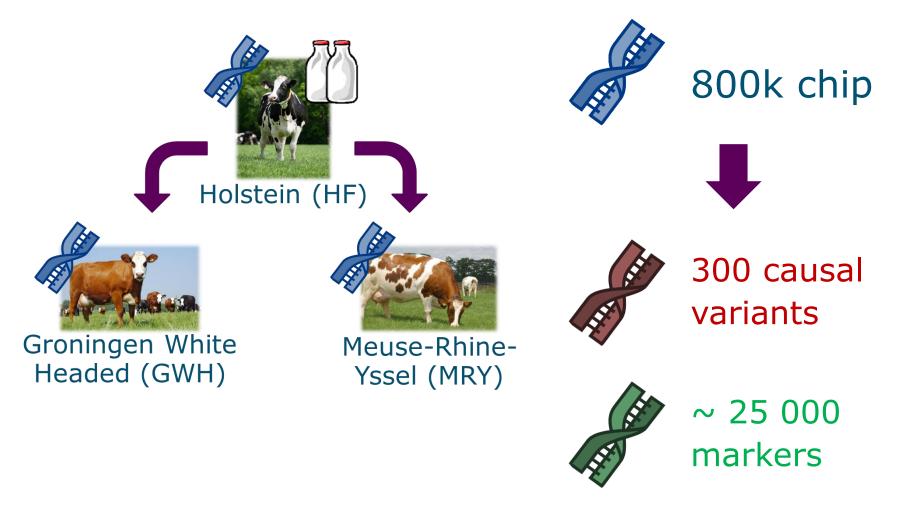




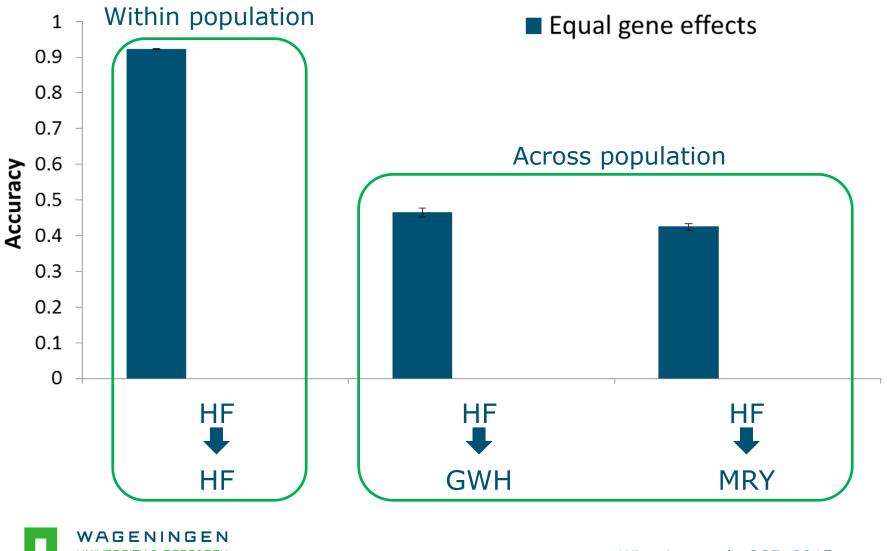
105 Groningen White Headed (GWH)

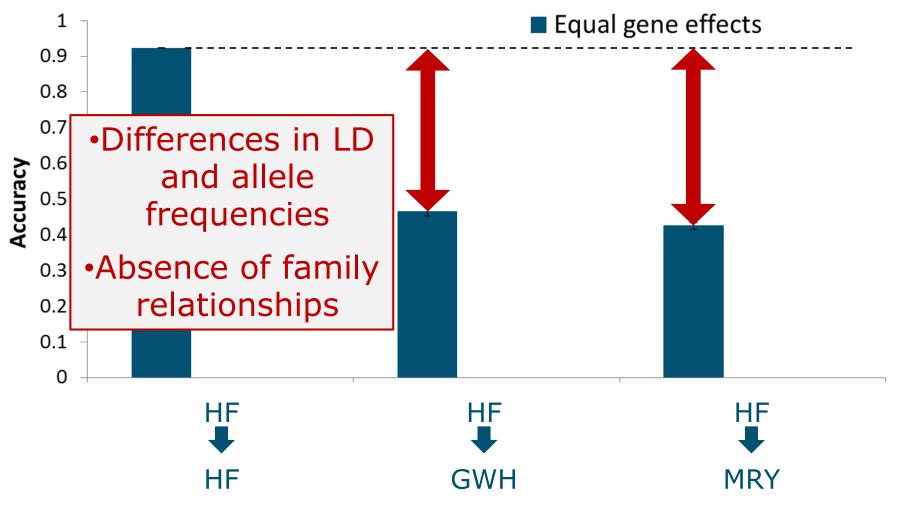
147 Meuse-Rhine-Yssel (MRY)











WAGENINGEN UNIVERSITY & RESEARCH

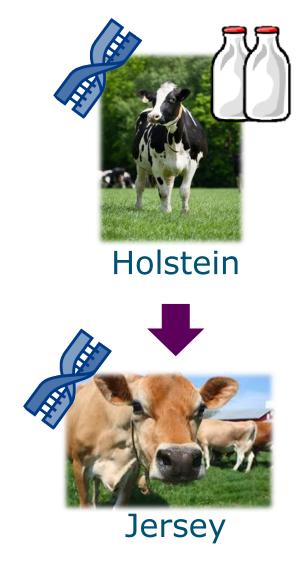


Accuracy of across-population genomic prediction is much lower than within-population genomic prediction

 Differences in effects of causal variants reduce accuracy



Effect of properties causal variants





Causal variants

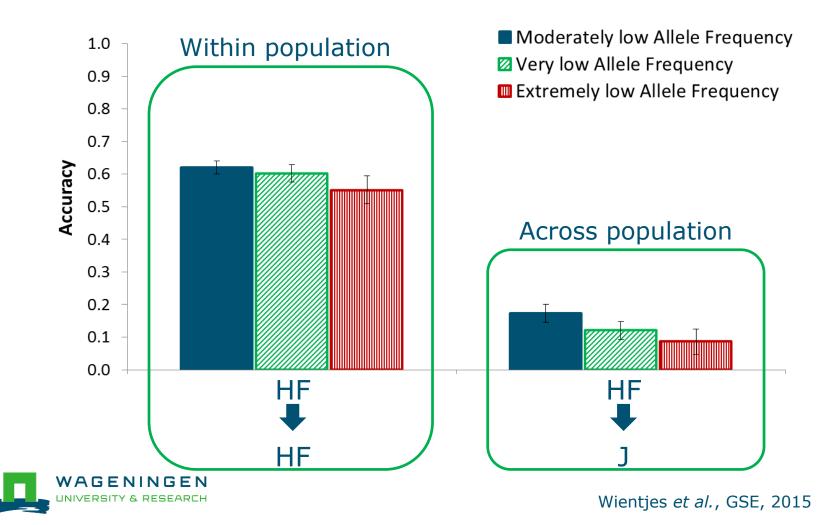
Moderately low allele frequency (~1 in 8)

Very low allele frequency (~1 in 13)

Extremely low allele frequency (~1 in 63)

Accuracies of predicting breeding values

Gene effects randomly sampled



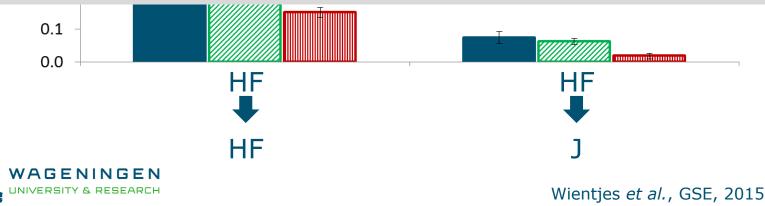
Accuracies of predicting breeding values

Larger effects for genes with lower frequency





Accuracy of genomic prediction depends on genetic architecture of trait



Accuracy of across-population genomic prediction is low...

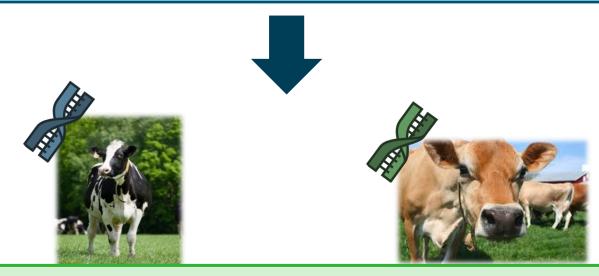
, but what if we combine populations in one reference population?



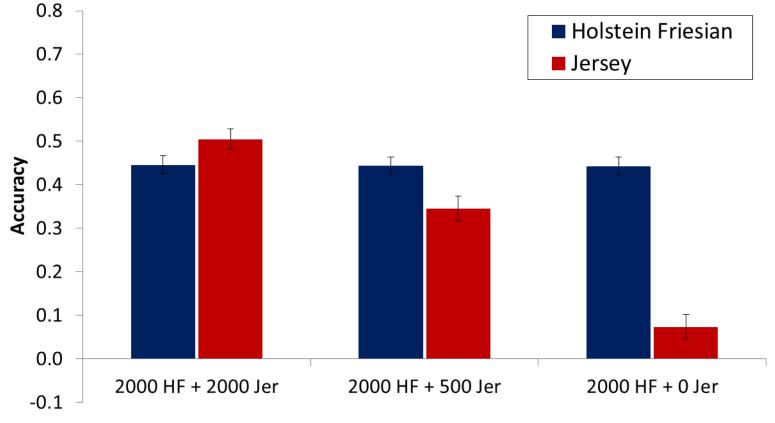




Prediction equation with marker effects

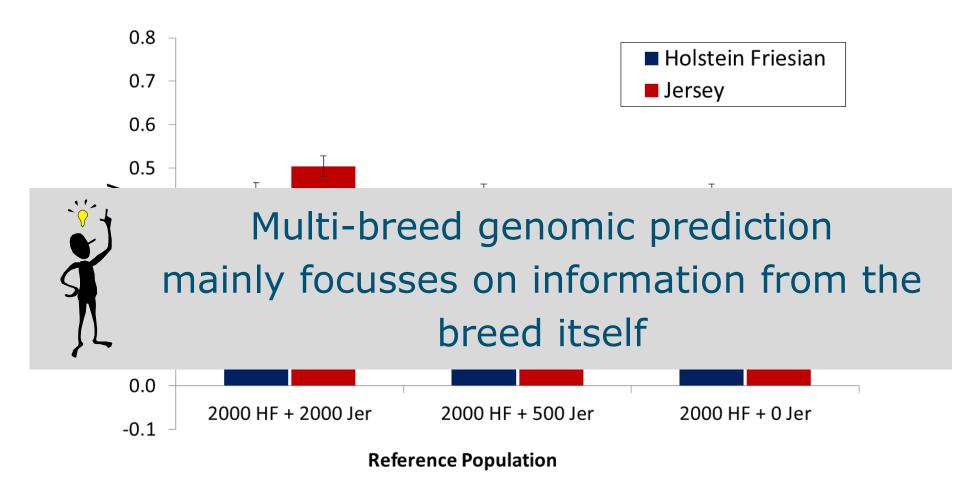


Genomic breeding values



Reference Population





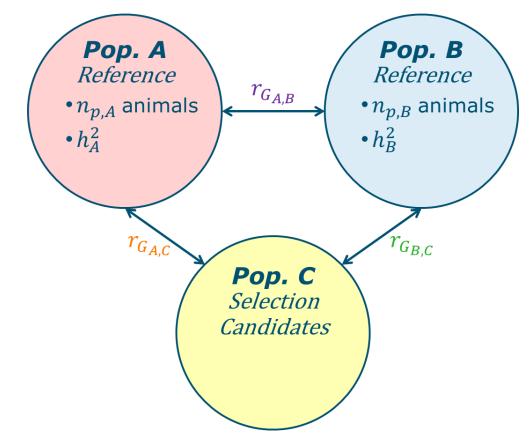


Can we predict accuracy?

Important when designing breeding programs

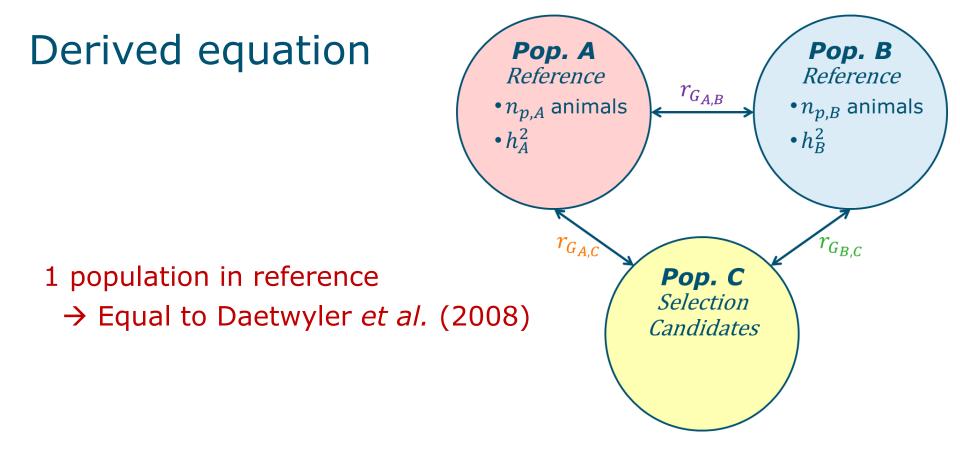
• Who to genotype?

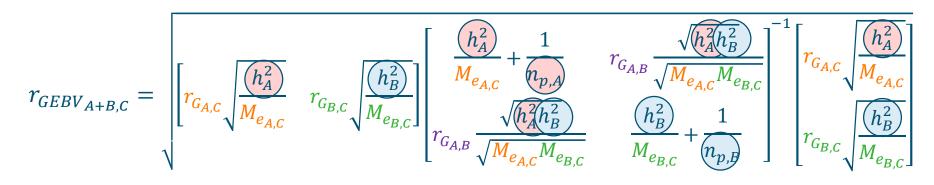
Derived equation



$$r_{GEBV_{A+B,C}} = \sqrt{\begin{bmatrix} r_{G_{A,C}} \sqrt{\frac{h_A^2}{M_{e_{A,C}}}} & r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}} \begin{bmatrix} \frac{h_A^2}{M_{e_{A,C}}} + \frac{1}{n_{p,A}} & r_{G_{A,B}} \frac{\sqrt{h_A^2 h_B^2}}{\sqrt{M_{e_{A,C}} M_{e_{B,C}}}} \end{bmatrix}^{-1} \begin{bmatrix} r_{G_{A,C}} \sqrt{\frac{h_A^2}{M_{e_{A,C}}}} \\ r_{G_{A,B}} \frac{\sqrt{h_A^2 h_B^2}}{\sqrt{M_{e_{A,C}} M_{e_{B,C}}}} & \frac{h_B^2}{M_{e_{B,C}}} + \frac{1}{n_{p,B}} \end{bmatrix}^{-1} \begin{bmatrix} r_{G_{A,C}} \sqrt{\frac{h_A^2}{M_{e_{A,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \\ r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{B,C}}}} \end{bmatrix}^{-1} \end{bmatrix} = \frac{1}{n_{B,C}} \begin{bmatrix} r_{G_{B,C}} \sqrt{\frac{h_B^2}{M_{e_{$$

Wientjes et al., Genetics, 2016

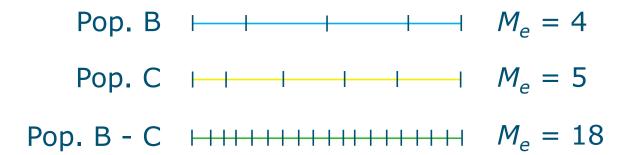




Wientjes et al., Genetics, 2016

M_e across populations

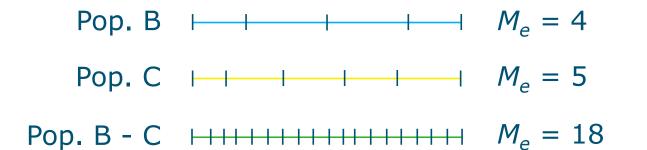
'Effective number of estimated effects'



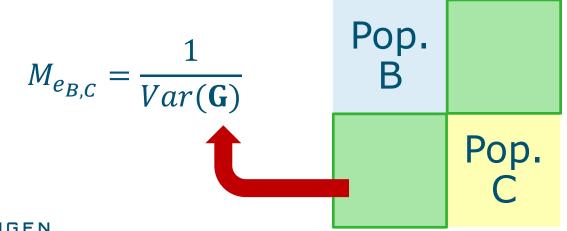


M_e across populations

'Effective number of estimated effects'

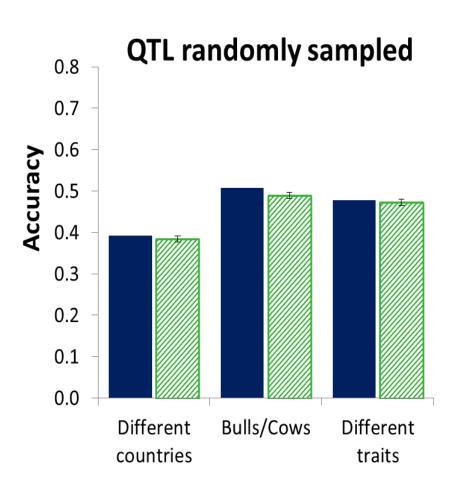






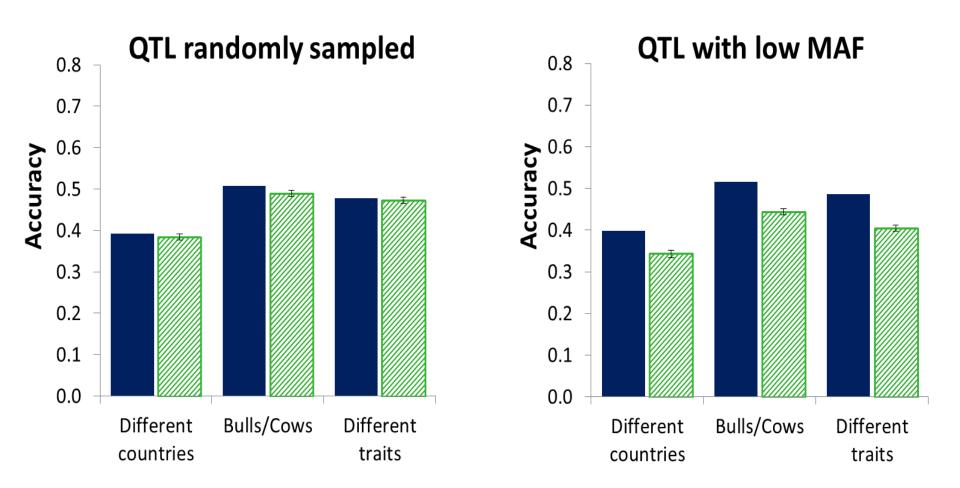


Results of validation



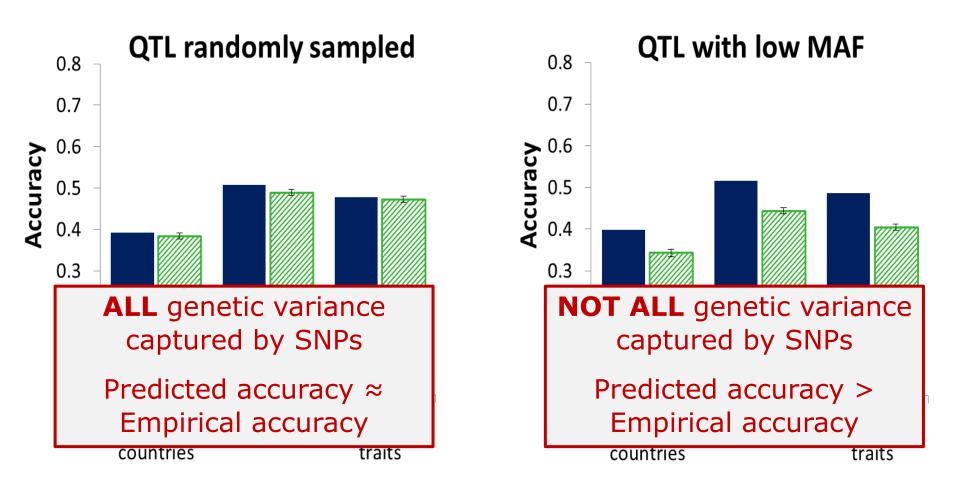
Predicted accuracy Empirical accuracy

Results of validation



Predicted accuracy Empirical accuracy

Results of validation



Predicted accuracy Empirical accuracy

Combining populations

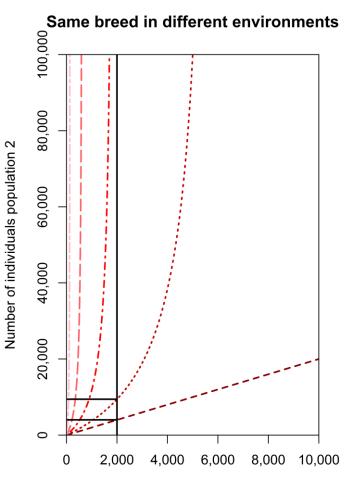
Populations of:

- Different countries
- Different breeds
- Bulls/Cows
- Measured for different traits
- Different generations





Combining populations?

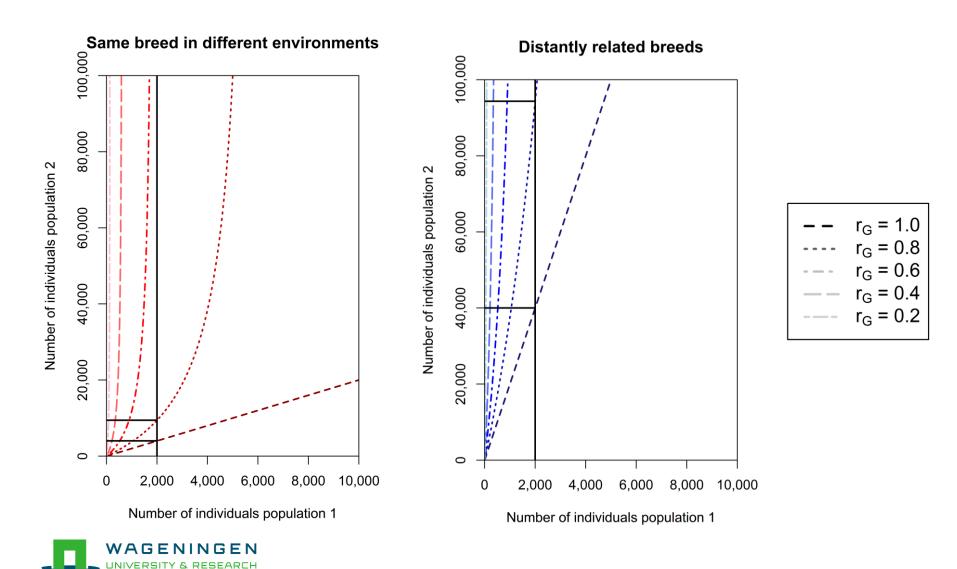


Number of individuals population 1



 $r_{G} = 1.0$ $r_{G} = 0.8$ $r_{G} = 0.6$ $r_{G} = 0.4$
 r _G = 0.4 r _G = 0.2

Combining populations?



Combining populations?

Beneficial, when:

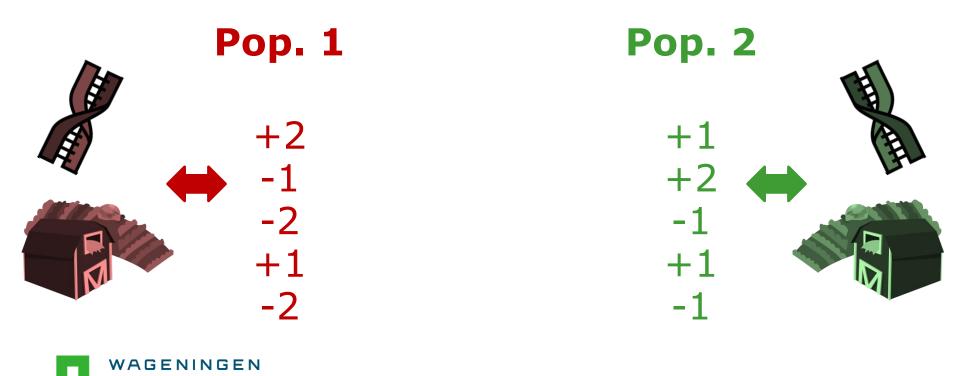
- Closely related populations
- Population itself is small
- A large number of individuals is added

Depends on genetic correlation between populations!!



Genetic correlation between populations (r_q)

Correlation between breeding values of two individuals with the same genotype in the two populations'



Estimation of rg

Multi-trait model Each population different trait REML

Requires relationships between populations



Relationships between populations

Theoretically defined relationships:

$$\mathbf{G}_{New} = \begin{bmatrix} \frac{\mathbf{Z}_{1}\mathbf{Z}_{1}'}{\sum 2p_{1i}(1-p_{1i})} & \frac{\mathbf{Z}_{1}\mathbf{Z}_{2}'}{\sqrt{\sum 2p_{1i}(1-p_{1i})}\sqrt{\sum 2p_{2i}(1-p_{2i})}} \\ \frac{\mathbf{Z}_{2}\mathbf{Z}_{1}'}{\sqrt{\sum 2p_{1i}(1-p_{1i})}\sqrt{\sum 2p_{2i}(1-p_{2i})}} & \frac{\mathbf{Z}_{2}\mathbf{Z}_{2}'}{\sum 2p_{2i}(1-p_{2i})} \end{bmatrix}$$



Relationships between populations

Theoretically defined relationships:

$$\mathbf{G}_{-}\text{New} = \begin{bmatrix} \mathbf{Z}_{1}\mathbf{Z}_{1}' \\ \hline \Sigma 2p_{1i}(1-p_{1i}) \\ \mathbf{Z}_{2}\mathbf{Z}_{1}' \\ \hline \sqrt{\sum 2p_{1i}(1-p_{1i})} \sqrt{\sum 2p_{2i}(1-p_{2i})} \end{bmatrix}$$

$$\frac{\mathbf{Z}_{1}\mathbf{Z}_{2}'}{\sqrt{\sum 2p_{1i}(1-p_{1i})}\sqrt{\sum 2p_{2i}(1-p_{2i})}} \frac{\mathbf{Z}_{2}\mathbf{Z}_{2}'}{\sum 2p_{2i}(1-p_{2i})}$$

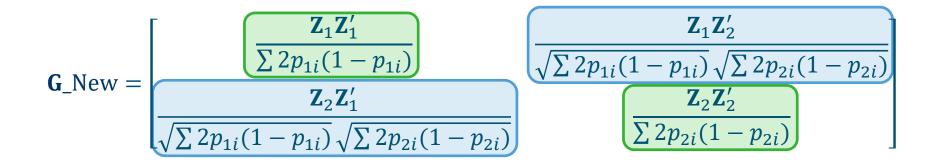
VanRaden – method 1



Wientjes et al., Genetics, 2017

Relationships between populations

Theoretically defined relationships:

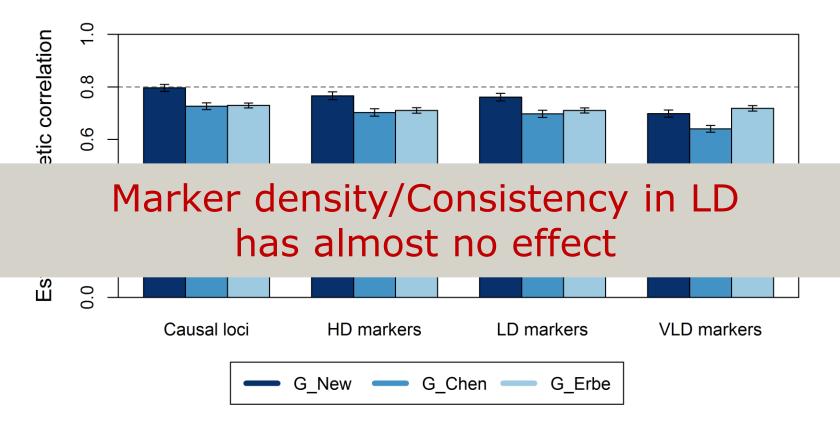


G_Chen: Scaling factor across populations: $\sum 2\sqrt{p_{1i}(1-p_{1i})p_{2i}(1-p_{2i})}$ **G_Erbe:** Base when populations separated



Estimated genetic correlation

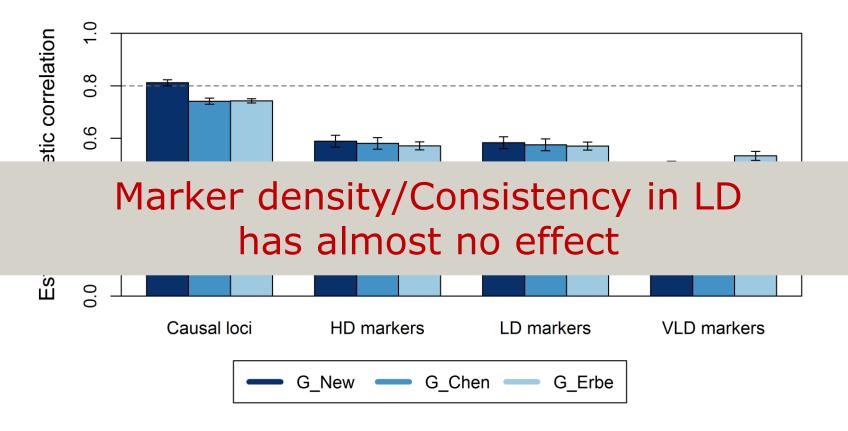
Allele frequency differences between populations of causal loci <u>REPRESENTED</u> by markers





Estimated genetic correlation

Allele frequency differences between populations of causal loci <u>NOT REPRESENTED</u> by markers





Conclusion

Combining populations beneficial when:

- Closely related populations
- Population itself is small
- Many individuals are added

Genetic correlation between populations:

- Important parameter
- Unbiasedly estimated with G_New
 - markers represent properties causal variants





Acknowledgements



Mario Calus Roel Veerkamp Piter Bijma Henk Bovenhuis Pascal Duenk Jeremie Vandenplas



Ben Hayes Mike Goddard



Chris Schrooten

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