Using theoretical and realized accuracies to estimate changes in heritabilities and genetic correlations

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# Changes under genomic selection

- Under selection:
  - Heritabilities decline
  - Antagonistic correlations intensify
- Accelerated changes under genomic selection
  - heritability halved for growth, negative correlations almost doubled (Hidalgo et al, 2019)
- With wrong parameters
  - Selection index inaccurate
  - Genetic gain not as expected
  - Possible deterioration of fitness traits
- How to estimate parameter change for last generation/year with commercial data sets?

### How to estimate parameter change by generation?

- If REML/Gibss sampling
  - Base population parameters
  - Impossible computations with too many genotypes
  - Biases with too few generations (Cesarani et al., 2019)



Data sampling in Hidalgo et al.(2019)

# **Requirements for new estimation**

- All data including genomic
- Parameters by last generation/ year
- Reasonable computing cost



Last generation parameters

### Realized and theoretical accuracies

Realized accuracy 
$$acc = corr(y - Xb, \hat{u})/h$$

Legarra et al. (2008)

y-Xb - adjusted phenotype  $\hat{u}$  - breeding value obtained without that phenotype  $h^2$  - heritability

Theoretical accuracy 
$$acc = \sqrt{\frac{Nh^2}{Nh^2 + M_e}}$$

Daetwyler et al. (2008)

N- number of genotyped animals with phenotypes  $M_{\rm e}-$  number of independent chromosome segments

Me ≈ 5k (chickens, pigs), 10k (beef), 15k (Holsteins) Pocrnic et al. (2017)

### **Derivations of realized accuracies**

$$acc = corr(u, \hat{u}) = \frac{cov(u, \hat{u})}{\sqrt{\sigma_u^2 \sigma_{\hat{u}}^2}} \qquad y = Xb + u + e, \qquad y - Xb = u + e$$

$$corr(y - Xb, \hat{u}) = \frac{cov(u + e, \hat{u})}{\sqrt{\sigma_p^2 \sigma_{\hat{u}}^2}} = \frac{cov(u, \hat{u})}{\sqrt{\sigma_p^2 \sigma_{\hat{u}}^2}} = \frac{cov(u, \hat{u})}{\sqrt{\sigma_p^2 \sigma_{\hat{u}}^2}} = \frac{cov(u, \hat{u})}{\sqrt{\frac{\sigma_u^2}{\sigma_u^2} \sigma_p^2 \sigma_{\hat{u}}^2}} = \frac{cov(u, \hat{u})}{\sqrt{\frac{\sigma_u^2}{\sigma_u^2} \sigma_p^2 \sigma_{\hat{u}}^2}}$$

$$=\frac{cov(u,\hat{u})}{\sqrt{\frac{1}{h^2}}\sigma_u^2\sigma_{\hat{u}}^2}} = acc h$$

# Pig data set

150k records on growth (h<sup>2</sup>=0.21) 25k records on fitness (h<sup>2</sup>=0.05)

53k genotyped animals





Hollifield et al., 2021

JSK genotypeu animais

$$\frac{\text{growth}}{53k * 0.21} = 0.83$$

$$\frac{\frac{\text{fitness}}{25k + 0.05}}{25k + 0.05 + 5k} = 0.44$$

**Theoretical accuracy** 

0.82

0.41

# **Broiler chicken data set**

820k phenotypes for growth h<sup>2</sup>=0.3 150k genotyped

Theoretical accuracy  $\sqrt{\frac{1}{150}}$ Realized accuracy

$$\int \frac{150k * 0.3}{150k * 0.3 + 5k} = 0.94$$
0.58

Both accuracy same 0.89 if  $h^2 = 0.13$ 

Company was using 0.14! (Breen, 2022)





Hidalgo et al., 2021

### Formulas for estimating heritability

$$\widehat{h^2}: \sqrt{\frac{Nh^2}{Nh^2 + M_e}} = corr(y - Xb, \widehat{u})/h$$

$$\widehat{h^2} = \frac{c^2 + \sqrt{c^4 + 4c^2M_e/N}}{2}, c = corr(y - Xb, \widehat{u})$$

$$SE(\widehat{h^2}) \approx \frac{1}{\sqrt{N_{val}}} \left[ c + \frac{2c^2 + \frac{4M_e}{N}}{\sqrt{c^2 + \frac{4M_e}{N}}} \right] \widehat{h^2} \approx \frac{3c}{\sqrt{N_{val}}}$$

N - # animals in reference  $N_{val}$  – number of animals in validation

# Heritability for milk in Holsteins



#### J. Dairy Sci. 104:5843–5853 https://doi.org/10.3168/jds.2020-19789

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### Genomic predictions for yield traits in US Holsteins with unknown parent groups

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# animals with phenotypes and genotypes	580k
# animals with validation	381k
Assumed M <sub>e</sub>	15k
Initial h <sup>2</sup>	0.35
Predictivity	0.55
Calculated h <sup>2</sup>	0.33

### Predictivity with large populations

$$acc = corr(y - Xb, \hat{u})/h < 1$$

$$h^2 > [corr(y - Xb, \hat{u})]^2$$

If 
$$corr(y - Xb, \widehat{u}) = 0.55$$
 then  $h^2 > 0.30$ 

### How to estimate genetic correlations?

Predictivity for trait i

$$corr(y_i - Xb_i, \widehat{u}_i) = acc_i h_i$$

What is predictivity from trait i to trait j?

$$corr(y_i - Xb_i, \widehat{u_j}) = ?$$

....

....

$$corr(y_{i} - Xb_{i}, \widehat{u_{j}}) = acc_{j} \ corr_{ij} \ h_{i}$$
$$corr_{ij} = \frac{corr(y_{i} - Xb_{i}, \widehat{u_{j}})}{h_{i} \ acc_{j}} \qquad SD(corr_{ij}) \approx \frac{1}{h_{i} \ acc_{j} \sqrt{N_{val}}}$$

# Test by simulation

	Trait 1		Trait 2
Validation population	10k		10k
Heritability	0.3		0.2
Correlation		-0.5	
Accuracy	0.8		0.3
Estimate			

corr <sub>12</sub>	-0.50±0.06
corr <sub>21</sub>	-0.50±0.03

# Procedure

- Select reference population, number of genotyped N > 10,000
- Select validation population, number of genotyped  $N_{val} > 5000$
- Calculate predictivities within and across traits
- Recalculate heritabilities
- Recalculate accuracies

$$acc = corr(y - Xb, \hat{u})/h$$

 $\widehat{h^2} = \frac{c^2 + \sqrt{c^4 + 4c^2 M_e/N}}{2}$ 

• Calculate genetic correlations

$$corr_{ij} = \frac{corr(y_i - Xb_i, \widehat{u_j})}{h_i \ acc_j} \pm \frac{1}{h_i \ acc_j \sqrt{N}}$$

Select correlation with smaller SE

# Conclusions

- Estimate of heritability from theoretical and realized accuracies
- Estimate of genetic correlations by predictivity across traits
- Any data size and data slice
- Low standard errors with large reference and validation populations



# **UGA AB&G team**

