



# Application of single-step genomic evaluation for crossbred performance

Tao Xiang Aarhus University,Denmark AgroParisTech, France

Bjarne Nielsen, Guosheng Su, Andres Legarra, Ole F Christensen





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## Background

• **Reference:** *Christensen, O. F et al.* (2014). Genomic evaluation of both purebred and crossbred performances. *Genetics Selection Evolution*, 46(1), 23.

$$\mathbf{y}_{\mathrm{A}} = \mathbf{X}_{\mathrm{A}}\mathbf{\beta}_{\mathrm{A}} + \mathbf{Z}_{\mathrm{A}}\mathbf{a}_{\mathrm{A}} + \mathbf{e}_{\mathrm{A}}$$

$$\mathbf{y}_{\mathrm{B}} = \mathbf{X}_{\mathrm{B}}\boldsymbol{\beta}_{\mathrm{B}} + \mathbf{Z}_{\mathrm{B}}\mathbf{a}_{\mathrm{B}} + \mathbf{e}_{\mathrm{B}}$$

$$\mathbf{y}_{AB} = \mathbf{X}_{AB}\boldsymbol{\beta}_{AB} + \mathbf{c}_{AB} + \mathbf{e}_{AB}$$

$$\mathbf{c}_{AB} = 0.5(\mathbf{Z}_{AB}, \mathbf{A}\mathbf{c}_{A} + \mathbf{Z}_{AB}, \mathbf{B}\mathbf{c}_{B}) + \mathbf{MS}_{AB}$$

$$\operatorname{Var} \begin{bmatrix} \boldsymbol{a}_{A} \\ \boldsymbol{c}_{A} \end{bmatrix} = \begin{bmatrix} \sigma_{a_{A}}^{2} & \sigma_{a_{A},c_{A}} \\ \sigma_{c_{A},a_{A}} & \sigma_{c_{A}}^{2} \end{bmatrix} \otimes \boldsymbol{A}_{A}$$







# Background

$$\mathbf{c}_{AB} = \mathbf{c}_{AB}^{(A)} + \mathbf{c}_{AB}^{(B)}$$
$$\mathbf{y}_{AB} = \mathbf{X}_{AB} \mathbf{\beta}_{AB} + \mathbf{c}_{AB}^{(A)} + \mathbf{c}_{AB}^{(B)} + \mathbf{e}_{AB}$$
$$\operatorname{Var} \begin{bmatrix} \mathbf{c}_{A} \\ \mathbf{c}_{AB}^{(A)} \end{bmatrix} = \mathbf{A}^{(A)} \sigma_{\mathbf{c}_{A}}^{2}$$

$$\operatorname{Var}\begin{bmatrix}\mathbf{a}_{A}\\ *\\ \mathbf{c}_{A}\\ \mathbf{c}_{AB}^{(A)}\end{bmatrix} = \begin{bmatrix}\sigma_{a_{A}}^{2} & \sigma_{a_{A},c_{A}}\\ \sigma_{c_{A},a_{A}} & \sigma_{c_{A}}^{2}\end{bmatrix} \otimes \mathbf{A}^{(A)}$$



How to construct  $\mathbf{A}^{(A)}$  and  $\mathbf{A}^{(B)}$ ?







## Background

Pedigree-based breed A specific

partial relationship matrix

 $\mathbf{A}^{(\mathbf{A})} = \begin{bmatrix} \mathbf{A}_{\mathbf{A}} & \mathbf{A}^{(\mathbf{A})}_{\mathbf{A},\mathbf{AB}} \\ \mathbf{A}^{(\mathbf{A})}_{\mathbf{AB},\mathbf{A}} & \mathbf{A}^{(\mathbf{A})}_{\mathbf{AB},\mathbf{AB}} \end{bmatrix}$ 

Marker-based breed A specific partial relationship matrix

$$G^{(A)} = egin{pmatrix} G^{(A)} & G^{(A)}_{A,A} & G^{(A)}_{A,AB} \ G^{(A)}_{AB,A} & G^{(A)}_{AB,AB} \end{pmatrix}$$

Assumption: known alleles origin

#### • $G^{(A)}$ is adjusted, compatible with $A^{(A)}$

Inverse scale of combined breed A partial relationship  

$$\left(H^{(A)}\right)^{-1} = \begin{pmatrix} \left(G^{(A)}_{\omega}\right)^{-1} - \left(A^{(A)}_{11}\right)^{-1} & 0 \\ 0 & 0 \end{pmatrix} + \left(A^{(A)}\right)^{-1} \end{pmatrix}$$







#### Phenotypic data

 Trait: TNB at the first parity of sows (h<sup>2</sup> is around 0.1)
 Records: 293,339 Landrace (LL); 180,112 Yorkshire(YY); 10,974 crossbreds (LY)
 Purebreds farrowed 2003~2013; Crossbreds farrowed 2010-2013

#### Pedigree

Traced back to year 1994 by DMU Trace (Madsen, 2010)
 332,929 LL & 210,554 YY







#### • Models:

Purebreds:  $y = hys + month + hybrid + b_1^*age + b_2^*age^2 + a + sb + e$ 

a ~ N(0, $\mathbf{H}_{(L)}\sigma_{a_L}^2$ ) or a ~ N(0, $\mathbf{H}_{(Y)}\sigma_{a_Y}^2$ ), depending on LL and YY; sb ~ N(0, $\mathbf{I}\sigma_{sb}^2$ ); e ~ N(0, $\mathbf{I}\sigma_{e}^2$ )

Crossbreds: y= hys + month + b1\*age +b2\*age<sup>2</sup> +  $\underline{c^{(L)}}$  +  $\underline{c^{(Y)}}$  +  $\underline{e}$ 

 $c^{(L)} \sim N(0, \mathbf{H}_{(L)}\sigma_{c_L}^2), c^{(Y)} \sim N(0, \mathbf{H}_{(Y)}\sigma_{c_Y}^2), e \sim N(0, \mathbf{I}\sigma_e^2)$ 







#### Genotypes

>41,009 SNPs (60K) genotyped in 7,723 LL & 7,785 YY

7,940 SNPs (8K) genotyped in 5,203 LY
Imputation 41,009 genotyped crossbred SNPs

#### Crossbred alleles tracing

Two phased imputed crossbred alleles segmented to segments
 LL and YY segmented to small segments same length, position
 Crossbred phased segment should match to purebred segments
 Proportions of matched segment counted in LL and YY separately

Known which phased LY allele originated from LL or YY







#### Scenarios (SC):

Scenarios	Genotypes	Phenotypes	Targets
Nogen_SC	NO, pedigree only	All phenotypic records	<ul> <li>VC estimation</li> <li>Model reliabilites for crossbred performance in purebred boars</li> </ul>
Genpure_SC	7723LL, 7785YY		
Genall_SC	7723LL, 7785YY, 5203LY		

## • $r^2 = \sum_{i=1}^{n} (1 - SEP_i^2 / \sigma_c^2) / n$







## **Results & Discussion**

#### Genetic correlation between purebred and crossbred performance

	Nogen_SC	Genpure_SC	Genall_SC
Landrace	0,700 (0,119)	0,736 (0,114)	0,785 (0,090)
Yorkshire	0,570 (0,129)	0,591 (0,124)	0,676 (0,100)

#### **Confirm the existence of a moderate correlation.**

**Crossbred genomic information reduce standard errors.** 







## **Results & Discussion**

# • Model reliabilities for purebred boars for cp.



 R<sup>2</sup> generally low, increases with genomic info included
 Non-genotyped animals benefit from genomic info on genotyped animals, especially crossbred info







#### Conclusions

- New single-step model is applicable for genetic evaluation for crossbred performance
- Model with marker information perform better than pedigree based model
- Crossbred genomic information is as essential as purebred genotypes
- Confirm moderate, positive genetic correlation between purebred and crossbred performance in TNB







# Thank you for your attention!

