





# Bayesian approach integrating correlated foreign information into a multivariate genetic evaluation

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#### Introduction

- √ For some traits (e.g., fatty acids, dry matter intake)
  - Phenotypes are unavailable or difficult to collect internally
- ✓ Most situations
  - Low accuracy of internal evaluations
  - □ Accurate external evaluations for correlated traits (e.g., milk productions evaluations)



#### Introduction

- ✓ Multivariate genetic evaluations
  - □ Correlations among traits
  - □ Prediction of EBV of a trait for which phenotypes are unavailable or difficult to collect internally
  - Improvement of accuracy



#### Aim

- ✓ To develop and test a simultaneous combination of
  - pedigree
  - □ internal phenotypes
  - correlated external information (i.e. EBV and REL)

with a multivariate evaluation using a Bayesian approach



#### **Methods**

✓ Regular mixed model equations

$$\begin{bmatrix} \mathbf{X'} \, \mathbf{R}^{-1} \mathbf{X} & \mathbf{X'} \, \mathbf{R}^{-1} \mathbf{Z} \\ \mathbf{Z'} \, \mathbf{R}^{-1} \mathbf{X} & \mathbf{Z'} \, \mathbf{R}^{-1} \mathbf{Z} + \mathbf{G}^{-1} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}}_{\mathbf{I}} \\ \hat{\mathbf{u}}_{\mathbf{I}} \end{bmatrix} = \begin{bmatrix} \mathbf{X'} \, \mathbf{R}^{-1} \mathbf{y}_{\mathbf{I}} \\ \mathbf{Z'} \, \mathbf{R}^{-1} \mathbf{y}_{\mathbf{I}} \end{bmatrix}$$

- $\mathbf{G}^{-1} = \mathbf{A}^{-1} \otimes \mathbf{G}_0^{-1}$ : inverse of additive genetic (co)variance matrix
- $lue{y}_{I}$ : vector of internal observations
- $\Box$   $\hat{\beta}_{\tau}$ : vector of estimated internal fixed effects
- $\Box$   $\hat{\mathbf{u}}_{\mathbf{T}}$ : vector of internal EBV

# **Assumption**

 $\checkmark$  Prior distribution of  $\mathbf{u}_{\mathbf{I}}$ 

$$\mathbf{p}(\mathbf{u}_{\mathbf{I}}) = MVN(\mathbf{0}, \mathbf{G}) \longrightarrow \mathbf{p}(\mathbf{u}_{\mathbf{I}}|\mathbf{y}_{\mathbf{E}}) = MVN(\hat{\mathbf{u}}_{\mathbf{E}}, \mathbf{D}_{\mathbf{E}})$$

- $\mathbf{y}_{\mathrm{E}}$ : unavailable vector of correlated external phenotypes
- $\square$   $\hat{\mathbf{u}}_{\mathbf{F}}$ : vector of external EBV
- $figspace{1}{2}$   $f D_E$ : prediction error (co)variance matrix of  $\hat{f u}_E$

#### **Methods**

✓ Integration of correlated external information

$$\begin{bmatrix} \mathbf{X'} \, \mathbf{R}^{-1} \mathbf{X} & \mathbf{X'} \, \mathbf{R}^{-1} \mathbf{Z} \\ \mathbf{Z'} \, \mathbf{R}^{-1} \mathbf{X} & \mathbf{Z'} \, \mathbf{R}^{-1} \mathbf{Z} + \mathbf{G}^{-1} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}}_{\mathbf{I}} \\ \hat{\mathbf{u}}_{\mathbf{I}} \end{bmatrix} = \begin{bmatrix} \mathbf{X'} \, \mathbf{R}^{-1} \mathbf{y}_{\mathbf{I}} \\ \mathbf{Z'} \, \mathbf{R}^{-1} \mathbf{y}_{\mathbf{I}} \end{bmatrix}$$



$$\begin{bmatrix} X' R^{-1} X & X' R^{-1} Z \\ Z' R^{-1} X & Z' R^{-1} Z + D_{E}^{-1} \end{bmatrix} \begin{bmatrix} \hat{\beta}_{I} \\ \hat{u}_{I} \end{bmatrix} = \begin{bmatrix} X' R^{-1} y_{I} \\ Z' R^{-1} y_{I} + D_{E}^{-1} \hat{u}_{E} \end{bmatrix}$$

#### Issue

✓ External information only available for external animals

- $\rightarrow$   $\hat{\mathbf{u}}_{\mathrm{E}}$  and  $\mathbf{D}_{\mathrm{E}}^{-1}$ : partially unknown
- → Estimations for internal animals



# Estimation of $\hat{u}_E$

- ✓ Available
  - lue External EBV of external animals (  $\hat{f u}_{E_{
    m F}}$  )
- ✓ Internal animals
  - lue Prediction of external EBV (  $\hat{\mathbf{u}}_{\mathbf{E_{I}}}$  )

$$p(\hat{\mathbf{u}}_{\mathbf{E}_{\mathbf{I}}}|\hat{\mathbf{u}}_{\mathbf{E}_{\mathbf{E}}}) = MVN(\mathbf{G}_{\mathbf{E}_{\mathbf{I}\mathbf{E}}}\mathbf{G}_{\mathbf{E}_{\mathbf{E}\mathbf{E}}}^{-1}\hat{\mathbf{u}}_{\mathbf{E}_{\mathbf{E}}}, (\mathbf{G}^{\mathbf{E}_{\mathbf{I}\mathbf{I}}})^{-1})$$

$$\rightarrow \hat{\mathbf{u}}_{\mathbf{E}} = \begin{bmatrix} \hat{\mathbf{u}}_{\mathbf{E}_{\mathbf{E}}}' & \hat{\mathbf{u}}_{\mathbf{E}_{\mathbf{I}}}' \end{bmatrix}'$$

Correct propagation of external information



# Estimation of $D_E^{-1}$

$$\mathbf{D}_{\mathrm{E}}^{-1} = \mathbf{G}^{-1} + \mathbf{\Lambda}_{\mathrm{E}}$$

$$\begin{aligned} & \boldsymbol{\Lambda}_{\mathbf{E}} = block \ diag \left( \boldsymbol{\Delta}_{\mathbf{j}} \mathbf{R}_{\mathbf{0}}^{-1} \boldsymbol{\Delta}_{\mathbf{j}} \right); \ j = 1, ..., n \ \text{animals} \\ & \text{For external animals} : \boldsymbol{\Delta}_{\mathbf{j}} = diag \left( \sqrt{RE_k} \right); \ k = 1, ..., t \ \text{traits} \\ & \text{For internal animals} : \boldsymbol{\Delta}_{\mathbf{j}} = \mathbf{0} \end{aligned}$$

**Pedigree** 

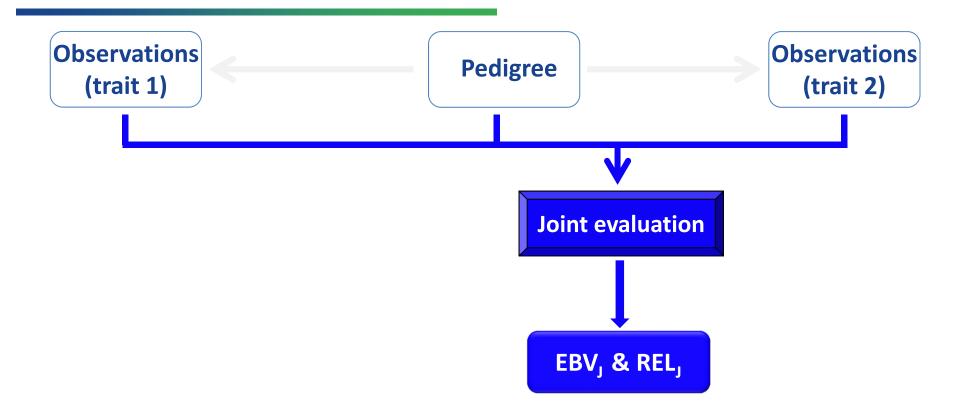
- **▶QMSim** (Sargolzaei and Schenkel, 2009)
  - **▶10** generations
  - **≥2240** animals
  - **➤** Random selection and matings

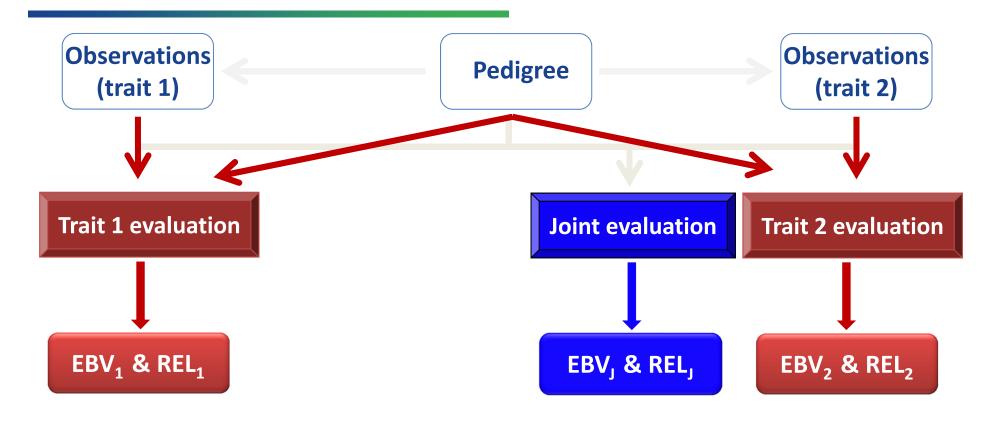


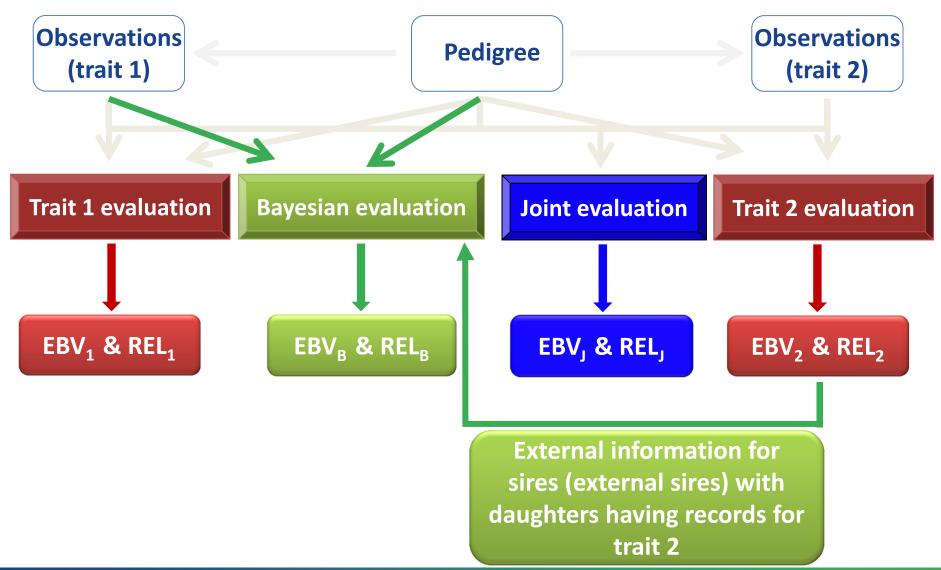
Observations (trait 1) Pedigree Observations (trait 2)

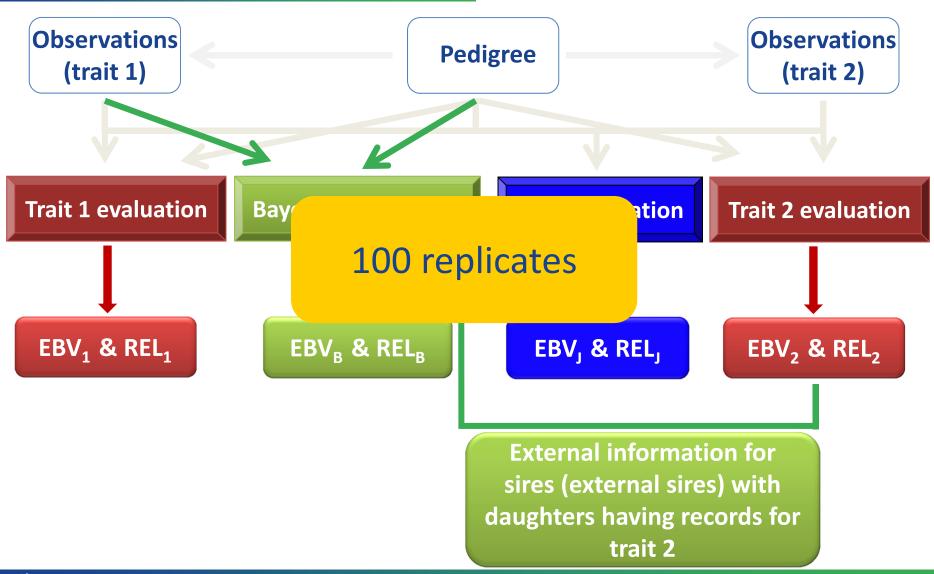
- **≻**Observations
  - **≥2** traits
  - **≻**Only for females
  - > Fixed effect: herd effect (randomly attributed)
  - **▶** Residual correlation: 0.00
  - **→** Genetic correlations: 0.10, 0.25, 0.50, 0.75 and 0.90

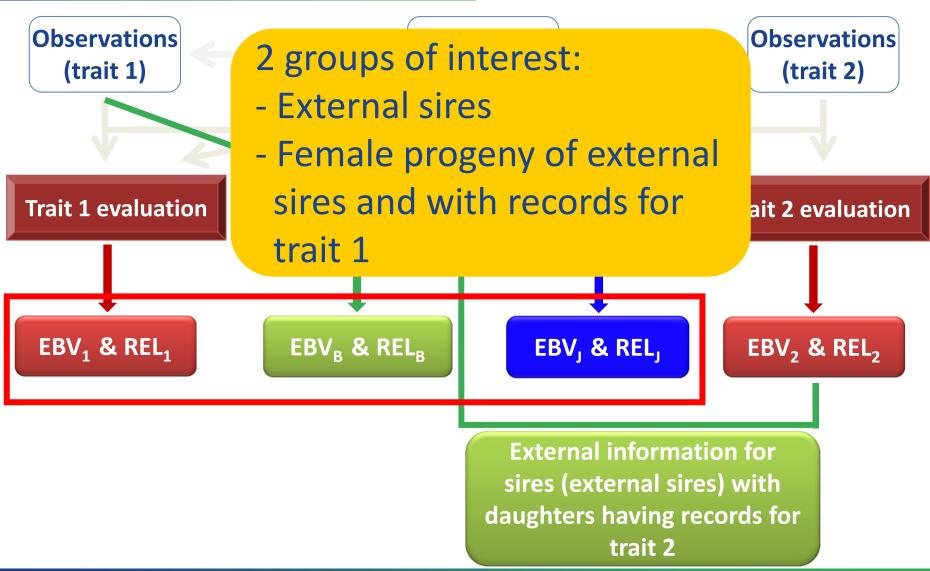
- **≻**Trait of interest **≻**3 herds
- $\geq$  2 herds  $\qquad \qquad \geq h^2 = 35\%$
- $h^2 = 10 \%$



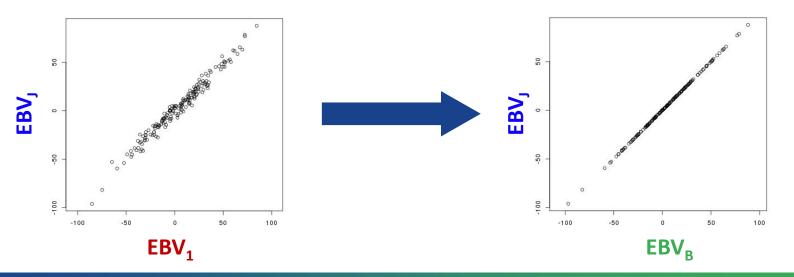




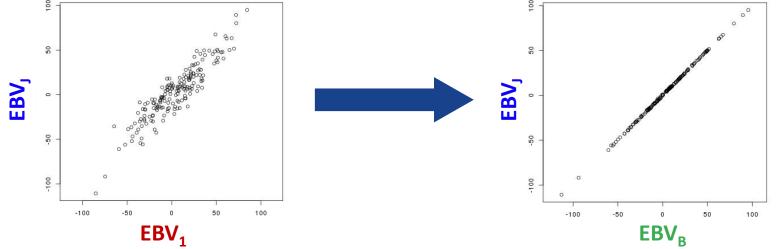




Fuelvetions	Genetic correlations					
Evaluations	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.987 (0.004)					
Bayesian	>0.999 (0.000)					

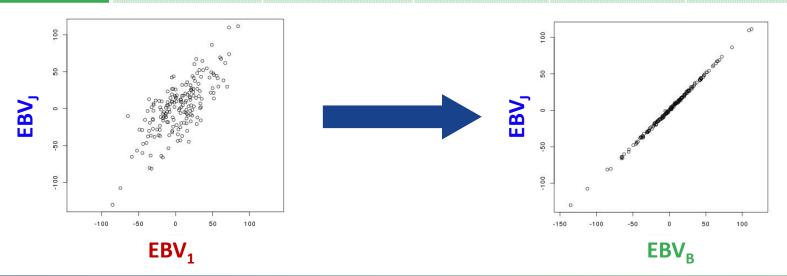


Evaluations	Genetic correlations					
	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.987 (0.004)	0.927 (0.020)				
Bayesian	>0.999 (0.000)	>0.999 (0.000)				
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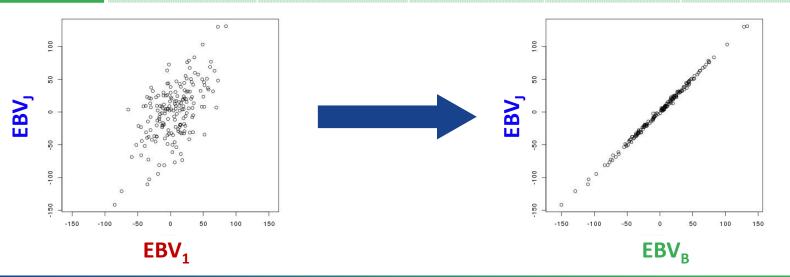


Evaluations	Genetic correlations					
	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.987 (0.004)	0.927 (0.020)	0.777 (0.053)			
Bayesian	>0.999 (0.000)	>0.999 (0.000)	0.999 (0.000)			



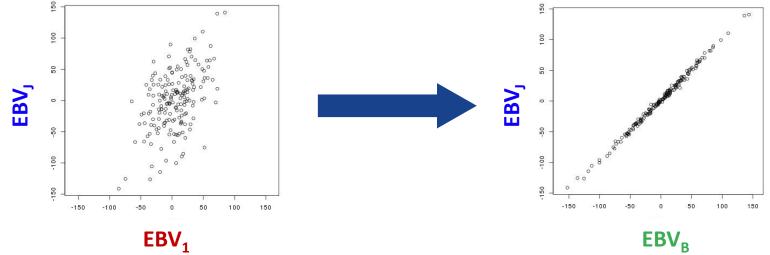


Evaluations	Genetic correlations					
	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.987 (0.004)	0.927 (0.020)	0.777 (0.053)	0.634 (0.079)		
Bayesian	>0.999 (0.000)	>0.999 (0.000)	0.999 (0.000)	0.999 (0.000)		





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Evaluations	Genetic correlations					
Evaluations	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.987 (0.004)	0.927 (0.020)	0.777 (0.053)	0.634 (0.079)	0.563 (0.091)	
Bayesian	>0.999 (0.000)	>0.999 (0.000)	0.999 (0.000)	0.999 (0.000)	0.998 (0.000)	
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✓ Average rank correlations of  $EBV_J$  with  $EBV_1$  or  $EBV_B$  for external sires (N = 181.0 ± 1.1)

Evaluations	Genetic correlations					
	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.987 (0.004)	0.927 (0.020)	0.777 (0.053)	0.634 (0.079)	0.563 (0.091)	
Bayesian	>0.999 (0.000)	>0.999 (0.000)	0.999 (0.000)	0.999 (0.000)	0.998 (0.000)	

→ Rankings of Bayesian evaluations similar to rankings of joint evaluations



#### ✓ Average REL

Fuelvetions	Genetic correlations					
Evaluations	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.10 (0.00)	0.10 (0.00)	0.10 (0.00)	0.10 (0.00)	0.10 (0.00)	
Bayesian	0.10 (0.00)	0.12 (0.00)	0.17 (0.00)	0.26 (0.00)	0.34 (0.00)	
Joint	0.10 (0.00)	0.12 (0.00)	0.17 (0.00)	0.26 (0.00)	0.33 (0.00)	

- → Retrieving almost all correlated information
- → Still some double counting



# Results: female progeny

✓ Average rank correlations of  $EBV_J$  with  $EBV_1$  or  $EBV_B$  for female progeny (N = 241.2 ± 47.1) of external sires

Fuelvetions	Genetic correlations					
Evaluations	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.992 (0.002)	0.954 (0.009)	0.844 (0.029)	0.721 (0.048)	0.652 (0.057)	
Bayesian	0.997 (0.001)	0.983 (0.003)	0.946 (0.010)	0.910 (0.017)	0.892 (0.021)	

→ Rankings of Bayesian evaluations more similar to rankings of joint evaluations



# Results: female progeny

#### ✓ Average REL

Evaluations	Genetic correlations					
Evaluations	0.10	0.25	0.50	0.75	0.90	
Trait 1	0.14 (0.00)	0.14 (0.00)	0.14 (0.00)	0.14 (0.00)	0.14 (0.00)	
Bayesian	0.14 (0.00)	0.14 (0.00)	0.15 (0.00)	0.18 (0.00)	0.20 (0.00)	
Joint	0.14 (0.00)	0.14 (0.00)	0.17 (0.00)	0.21 (0.00)	0.25 (0.00)	

→ Propagation of sires' external information to progeny



# **Perspectives**

- ✓ Combination of information for traits having different
  - variance components heritabilities
    - Milk yields in different countries
  - units of measurement
    - Milk yields expressed in kg or lb



# Perspectives

- ✓ Combination of information for traits having different
  - variance components heritabilities
    - Milk yields in different countries
  - units of measurement
    - Milk yields expressed in kg or lb
  - models
    - □ Random regressions test-day or lactation models



### **Perspectives**

- ✓ Combination of information for traits having different
  - variance components heritabilities
    - Milk yields in different countries
  - units of measurement
    - Milk yields expressed in kg or lb
  - models
    - □ Random regressions test-day or lactation models
  - genotype by environment interactions



#### **Conclusions**

- ✓ Good integration of correlated external information
  - even with low genetic correlations
- ✓ Rankings of the Bayesian evaluations more similar to rankings of the joint evaluations
  - for animals with external information
  - for their progeny
- ✓ Numerous possible applications



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