

# Multitrait across country genomic evaluations for EuroGenomics countries

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Hanni Kärkkäinen<sup>1</sup>, Vincent Ducrocq<sup>2</sup>, Sören Borchersen<sup>3</sup>,  
Gert Aamand<sup>4</sup>, Reinhard Reents<sup>5</sup>, Esa Mäntysaari<sup>1</sup>

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<sup>1</sup>Natural Resources Institute Finland,

<sup>2</sup>French National Institute for Agricultural Research (INRA), France

<sup>3</sup>EuroGenomics Cooperation, Denmark

<sup>4</sup>Nordic Cattle Genetic Evaluation (NAV), Denmark

<sup>5</sup>Vereinigte Informationssysteme Tierhaltung w.V. (vit), Germany

- EuroGenomics<sup>1</sup> multitrait across country evaluation -project<sup>2</sup> started on May 2018
- Contrary to the Melbourne project, EuroGenomics countries share bull genotypes  
⇒ Possible to build a true multitrait across country SNP BLUP evaluation using pseudo phenotypes from all countries directly

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<sup>1</sup>Germany (DEU), Nordic countries Denmark, Finland and Sweden (DFS), France (FRA), The Netherlands (NLD), Spain (ESP) and Poland (POL). Order and abbreviations from Interbull practice.

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# Shared EuroGenomics data

## Genotypes

- 46,342 SNP genotypes for total of  $\sim 35,000$  bulls with a record
- Imputed genotypes received from NAV ( $\rightarrow$  common set of markers)

## Phenotypes

- Protein yield, somatic cell score and female fertility
- Around 11,000 – 3,700 records within countries
- EBVs the countries send to Interbull evaluation + EDC  $\rightarrow$  DRP
- Heritability estimates from countries

## Pedigree

Genetic correlation estimates from Interbull

## First phase

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- Our first goal was to demonstrate and validate the performance of EuroGenomics SNP MACE
- We have accomplished that based on the shared bull genotypes, and shown that it is feasible and benefits the participants

# SNP MACE Model

- Basic SNP MACE model  $\mathbf{y} = \boldsymbol{\mu} + \mathbf{Z}\mathbf{g} + \mathbf{e}$

$$\Leftrightarrow \begin{bmatrix} \mathbf{y}_1 \\ \vdots \\ \mathbf{y}_c \end{bmatrix} = \begin{bmatrix} \mu_1 \mathbf{1}^{n_1} \\ \vdots \\ \mu_c \mathbf{1}^{n_c} \end{bmatrix} + \begin{bmatrix} \mathbf{Z}_1 \mathbf{g}_1 \\ \vdots \\ \mathbf{Z}_c \mathbf{g}_c \end{bmatrix} + \begin{bmatrix} \mathbf{e}_1 \\ \vdots \\ \mathbf{e}_c \end{bmatrix} \quad (1)$$

- $\mathbf{y}_i \in \mathbb{R}^{n_i}$  is the pseudo phenotype (deregressed national breeding value, later DYD) for country  $i \in [1, \dots, c]$  with  $n_i$  observations
- $\mu_i$  the general mean for country  $i$
- $\mathbf{1}$  vector of  $n_i$  ones
- $\mathbf{Z}_i \in \mathbb{R}^{n_i \times m}$  design matrix for genotypes ( $m$  is the number of markers, all countries have the same set of markers with same 0,1,2 coding)

- $\mathbf{g}_i \in \mathbb{R}^m$  estimated SNP effects for country  $i$
- $\mathbf{e}_i \in \mathbb{R}^{n_i}$  residual effects for country  $i$  individuals
- $\text{Var}(\mathbf{e}_i) = \sigma_{e_i}^2 \text{diag}(1/\text{EDC}_{ik}) = \mathbf{R}_i \forall i$ , for animals  $k \in [1, \dots, n_i]$
- $\text{Cov}(\mathbf{e}_i, \mathbf{e}_{i^+}) = 0 \forall i \neq i^+$
- $\text{Var}(\mathbf{g}_i) = \mathbf{I}^m \sigma_{s_i}^2 \theta_i$ , where  $\theta_i = 1 / \sum_{j=1}^m 2p_{ij}(1 - p_{ij})$   
with  $p_{ij}$  = allele frequency of locus  $j$  in country  $i$ ,  $\sigma_{s_i}^2$  = sire variance of country  $i$  and  $\mathbf{I}^m \in \mathbb{R}^{m \times m}$  identity matrix
- $\text{Cov}(\mathbf{g}_i, \mathbf{g}_{i^+}) = \mathbf{I}^m \sigma_{ii^+} \sqrt{\theta_i \theta_{i^+}}$ , where  $\sigma_{ii^+} = \rho_{ii^+} \times \sigma_{s_i} \sigma_{s_{i^+}}$ , with  $\rho_{ii^+}$  = genetic correlation between countries  $i$  and  $i^+$

# SNP MACE Model – (Co)variance matrices

$$\text{Var} \begin{bmatrix} \mathbf{g}_1 \\ \vdots \\ \mathbf{g}_c \end{bmatrix} = \begin{bmatrix} \mathbf{I}^m \sigma_{s_1}^2 \theta_1 & \dots & \mathbf{I}^m \sigma_{1c} \sqrt{\theta_1 \theta_c} \\ & \ddots & \vdots \\ \text{symm.} & & \mathbf{I}^m \sigma_{s_c}^2 \theta_c \end{bmatrix} = \mathbf{D} \quad (2)$$

$\in \mathbb{R}^{(c \times m) \times (c \times m)}$  and it's inverse

$$\mathbf{D}^{-1} = \begin{bmatrix} \mathbf{D}^{11} & \dots & \mathbf{D}^{1c} \\ & \ddots & \vdots \\ \text{symm.} & & \mathbf{D}^{cc} \end{bmatrix} \quad (3)$$

$\in \mathbb{R}^{(c \times m) \times (c \times m)}$ .

$$\text{Var} \begin{bmatrix} \mathbf{e}_1 \\ \vdots \\ \mathbf{e}_c \end{bmatrix} = \begin{bmatrix} \mathbf{R}_1 & \dots & \mathbf{0} \\ & \ddots & \vdots \\ \text{symm.} & & \mathbf{R}_c \end{bmatrix} = \mathbf{R} \quad (4)$$

$\in \mathbb{R}^{n \times n}$ , where  $n = \sum_{i=1}^c n_i$  and

$$\mathbf{R}_i = \sigma_{e_i}^2 \text{diag}(1/\text{EDC}_{ik}).$$

# SNP MACE Model – Mixed Model Equations

$$\begin{bmatrix} \ddots & & & & & & & \\ & \ddots & & & & & & \\ & & \begin{bmatrix} \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{1} & \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{Z}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{1} & \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{Z}_i + \mathbf{D}^{ii} \end{bmatrix} & \cdots & & & & \\ & & & \ddots & & & & \\ & & & & \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{D}^{i+} \end{bmatrix} & & & \\ & & & & & \ddots & & \\ & & & & & & \begin{bmatrix} \mathbf{1}'\mathbf{R}_{i+}^{-1}\mathbf{1} & \mathbf{1}'\mathbf{R}_{i+}^{-1}\mathbf{Z}_{i+} \\ \mathbf{Z}_{i+}'\mathbf{R}_{i+}^{-1}\mathbf{1} & \mathbf{Z}_{i+}'\mathbf{R}_{i+}^{-1}\mathbf{Z}_{i+} + \mathbf{D}^{i+i+} \end{bmatrix} & \cdots \\ & & & & & & & \ddots \end{bmatrix} \times \begin{bmatrix} \vdots \\ \hat{\boldsymbol{\mu}}_i \\ \hat{\mathbf{g}}_i \\ \vdots \\ \hat{\boldsymbol{\mu}}_{i+} \\ \hat{\mathbf{g}}_{i+} \\ \vdots \end{bmatrix} = \begin{bmatrix} \vdots \\ \begin{bmatrix} \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{y}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{y}_i \end{bmatrix} \\ \vdots \\ \begin{bmatrix} \mathbf{1}'\mathbf{R}_{i+}^{-1}\mathbf{y}_{i+} \\ \mathbf{Z}_{i+}'\mathbf{R}_{i+}^{-1}\mathbf{y}_{i+} \end{bmatrix} \\ \vdots \end{bmatrix} \quad (5)$$



- Data was split into learning and validation sets by bulls' birth date
  - The youngest 10% from each country → validation set
- Under SNP MACE the animal solutions (DGV) were computed as  $\hat{a}_{ik} = \mathbf{z}_{ik}\hat{\mathbf{g}}_i$  for animal  $k$  in country  $i$
- The bias  $b_1$  was tested with a weighted linear regression of  $\text{DRP}_v$  on predicted  $\text{DGV}_v$ , using  $\text{EDC}_v$  as weights
- Validation reliability was defined as  $R_v^2 = (\text{cor}(\text{DRP}_v, \text{DGV}_v))^2 / R_{\text{DRP}_v}^2$ ,
  - Records with  $R_{\text{DRP}_v}^2 \geq 0.5$  were used in validation  
(except for Poland fertility trait  $R_{\text{DRP}_v}^2 \geq 0.3$ , due to limited no. of records)

Two reference methods:

1. Country-wise single trait model
  - Compares to situation where country uses only their own geno- and phenotypes
2. Current EuroGenomics practice *i.e.* using MACE proofs for all exchange bulls
  1. Run MACE BLUP → solutions for all animals
  2. Estimate reliabilities / EDC for all records
  3. → Deregressed proofs for all animals
  4. → National DGV:s by single trait GBLUP

### Residual polygenic component

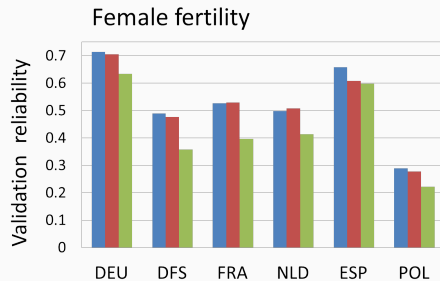
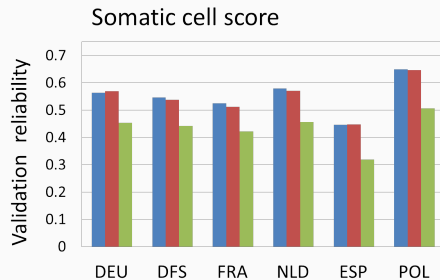
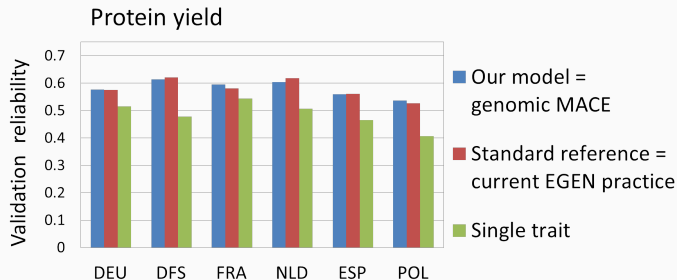
- Benefits all models (genomic MACE and both reference methods)
- We tested 10, 20 and 30% of polygenic effect  $\Rightarrow$  On average 20% best choice
- Genomic MACE  $R_V^2$  rises on average 6%, also bias diminishes noticeably

### Estimation of genetic correlations

- Estimated with MTG2 program (Lee & *al.*)
- Done with the "official Interbull style"  
= variance ratio kept constant, genetic covariances and sire variances estimated
- Not much different values than Interbull estimates  $\iff$  Not much different reliabilities

# Results

Validation reliability  $R_v^2$  of DGV predicted by the genomic MACE, current EuroGenomics MACE and a single trait model, all models with 20% polygenic effect and Interbull variance components.



## First phase conclusions

After the first phase we have learned that

- Fitting genomic MACE with individual animal genotypes is feasible, and countries gain from cooperation
- The genomic MACE produces on average slightly higher validation reliability and is slightly less biased (higher  $b_1$ ) than the current EuroGenomics MACE
- Under all of the tested models the equivalent GBLUP has better convergence properties than the SNP BLUP
- Residual polygenic component seems useful
- Genetic correlations estimated by Interbull can be utilized in genomic MACE

## Second phase

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## Including cow information

- EuroGenomics countries want to include cow reference information *without* sharing the cow genotypes
- We are developing a method to use all the information, including cows
  - requires only SNP-solutions (computed with full national reference population) and PEVs of the shared bulls
- Procedure includes
  1. PEV  $\xrightarrow{\text{iterative approximation}}$  EDC for bulls
  2. SNP-solutions & EDC & genotypes  $\rightarrow$  RHS
  3. RHS & EDC & genotypes  $\rightarrow$  Multitrait SNP-effects
- We call this "SNP information approximation approach"

# SNP MACE Model – Mixed Model Equations

$$\begin{bmatrix} \ddots & & & & \\ & \vdots & & & \\ & \begin{bmatrix} \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{1} & \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{Z}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{1} & \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{Z}_i + \mathbf{D}^{ii} \end{bmatrix} & \cdots & \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{D}^{ii+} \end{bmatrix} & \cdots \\ & & \ddots & \vdots & \\ & & & \begin{bmatrix} \mathbf{1}'\mathbf{R}_{i+}^{-1}\mathbf{1} & \mathbf{1}'\mathbf{R}_{i+}^{-1}\mathbf{Z}_{i+} \\ \mathbf{Z}_{i+}'\mathbf{R}_{i+}^{-1}\mathbf{1} & \mathbf{Z}_{i+}'\mathbf{R}_{i+}^{-1}\mathbf{Z}_{i+} + \mathbf{D}^{i+i+} \end{bmatrix} & \cdots \\ & & & & \ddots \end{bmatrix} \times \begin{bmatrix} \vdots \\ \hat{\boldsymbol{\mu}}_i \\ \hat{\mathbf{g}}_i \\ \vdots \\ \hat{\boldsymbol{\mu}}_{i+} \\ \hat{\mathbf{g}}_{i+} \\ \vdots \end{bmatrix} = \begin{bmatrix} \vdots \\ \begin{bmatrix} \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{y}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{y}_i \end{bmatrix} \\ \vdots \\ \begin{bmatrix} \mathbf{1}'\mathbf{R}_{i+}^{-1}\mathbf{y}_{i+} \\ \mathbf{Z}_{i+}'\mathbf{R}_{i+}^{-1}\mathbf{y}_{i+} \end{bmatrix} \\ \vdots \end{bmatrix}$$



## Estimation of country wise EDC — Background

Since EuroGenomics countries share the bull genotypes, we can construct the left hand side matrices

$$\begin{bmatrix} \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{1} & \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{Z}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{1} & \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{Z}_i + \mathbf{D}_i^{-1} \end{bmatrix}$$

— if we know the  $\mathbf{R}_i^{-1}$

- Matrix  $\mathbf{R}_i^{-1}$  holds weights for each bull
- Until now we have used  $\mathbf{R}_i^{-1} = \text{diag}\{EDC_{ij}\sigma_{e_i}^{-2}\}$ ,  
with  $EDC_{ij}$  being the number of daughters of bull  $j$  in country  $i$
- On the other hand, the prediction error variance of bull  $j$  from country  $i$

$$PEV_{ij} \simeq [(\mathbf{R}_i^{-1} + \mathbf{G}_i^{-1}\sigma_{s_i}^{-2})^{-1}]_{j,j}, \text{ where } \mathbf{G}_i = \sigma_{s_i}^{-2} \times \mathbf{Z}_i\mathbf{D}_i\mathbf{Z}_i'$$

- Equivalently the same can be attained using SNP model

$$PEV_{ij} \simeq \mathbf{z}_j[(\mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{Z}_i + \mathbf{D}_i^{-1})^{-1}]\mathbf{z}_j'$$

## Estimation of country wise EDC – II

- However, we are restricted to genotypes and EDCs that countries exchange  
⇒ can't compute  $(\mathbf{Z}'_i \mathbf{R}_i^{-1} \mathbf{Z}_i + \mathbf{D}_i^{-1})^{-1}$  the countries use
- But, if we would get for each exchanged bull

$$PEV_{ij} = \mathbf{z}_j [(\mathbf{Z}'_i \mathbf{R}_i^{-1} \mathbf{Z}_i + \mathbf{D}_i^{-1})^{-1} \mathbf{z}'_j]$$

from the countries

- We could equate it to

$$PEV_{ij} = [(\mathfrak{R}_i^{-1} + \mathbf{G}_i^{-1} \sigma_{s_i}^{-2})^{-1}]_{j,j}$$

where  $\mathfrak{R}_i^{-1}$  would consists of weights of the (exchange) bulls that would lead into the same PEV that the country has computed with all animals (including) females in national genomic evaluation.

## Estimation of $\mathfrak{R}_i^{-1}$

- Values of  $\mathfrak{R}_i^{-1}$  can be estimated iteratively
- We have used a Newton method, where the new EDC are estimated as:

$$edc_{k+1} = edc_k - \mathbf{C}_k^{-1}(PEV_k - PEV), \text{ where}$$

- $edc_k$  and  $edc_{k+1}$  are the current and the subsequent EDC estimates, respectively,
- $PEV_k$  is the PEV computed as  $\text{diag}(LHS^{-1})$  using the current ( $k$ th) estimate of EDC,
- $PEV$  consists of the PEVs the country has computed with the full national reference population and
- $\mathbf{C}$  is the value of the partial derivative of  $(PEV_k - PEV)$  with respect to  $edc$  at point  $edc_k$ , that can be simplified into  $\mathbf{C} = LHS^{-1} \circ LHS^{-1}$ ,

corresponding to the general description of the Newton method

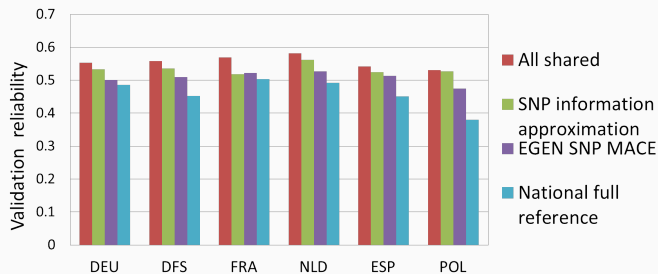
$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$$

## Pilot testing the model with current EuroGenomics data

1. Divide data for every country:
  - Validation (youngest 10%) & "full national reference population"
  - "Full reference"  $\xrightarrow{\text{randomly 1:1}}$  "Shared bulls" + "Cows"
2. SNP-solutions and PEVs by country using the "full national reference"  
 $\rightarrow$  "shared SNP-solutions" & "PEVs of the shared bulls"
3. Country  $i$  PEVs  $\xrightarrow{\text{Newton iteration}}$  Country  $i$  EDCs
4. Country  $i$  SNP-solutions  $\rightarrow$  Country  $i$  RHS
  - using "shared" genotypes (= half of the animals)
  - estimated  $EDC_i^{-1}$  as weights
5. All country wise RHS  $\rightarrow$  multitrait SNP BLUP  $\rightarrow$  SNP-solutions  $\rightarrow$  DGV
6. Validate by using the youngest 10%

# Pilot for protein yield

Validation reliability  $R_v^2$  of DGV predicted by multitrait SNP BLUP with full reference population, SNP information approximation, MT SNP BLUP with shared reference and national full reference SNP BLUP.



## MULTITRAIT ACROSS COUNTRY

- **All shared:** countries would share everything, including cow genotypes
- **SNP information approximation:** countries share bull genotypes (as currently) + SNP-solutions & PEVs of shared bulls
- **EGEN SNP MACE:** countries share bull genotypes — phase I model

## SINGLE TRAIT

- **National full reference:** countries use all national information, but do not share anything

## EDC estimation

- The Newton iteration is computationally feasible
  - Requires 2 inverses of matrix size *number\_of\_animals* / iteration round
- In the pilot study the iteration was run until convergence

## National SNP-solutions → RHS → MT SNP BLUP

- Is implemented for testing purposes as part of our MiX99 suite
- Works quite nicely
  - Converges and behaves similar to "normal" multitrait SNP BLUP runs
- Could be even more advantageous for low heritability traits
- **Pilot study!** ⇒ With actual cow data behaves probably differently



**Thank You !**

