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Animal Production

A SNP MACE model for international genomic evaluation: - technical challenges and possible solutions

Zengting Liu and Mike E. Goddard
vit, IT solutions for animal production, Germany
University of Melbourne, Melbourne 3083, VIC, Australia

Overview

Current status of (inter)national conventional and genomic evaluations

- A SNP MACE model for international evaluation
- Solving algorithms for the SNP MACE model
- Approximation of prediction error (co)variances
- Further development and extension
 - Different SNP sets across countries
 - Countries use MACE info as phenotypes



International evaluation based on SNP effects

- Concept of a SNP based international evaluation (Goddard, 2011)
- A SNP-Focus Model replacing the Animal Model (Schaeffer, 2014)
- Interbull estimation of SNP effects (Goddard, 2016)
- A SNP MACE model proposed (Goddard, 2017)
 - Interbull Technical Workshop, Slovenia
- Interbull project on the SNP MACE model



A SNP MACE model

- A SNP genomic model for Multiple Across Country Evaluation

$$\mathbf{g}_i^N \implies \mathbf{g}_i \quad [1]$$

for country i ($i = 1, \dots, c$)

- A SNP BLUP model for national genomic evaluation

$$\mathbf{y}_i = \mu_i \mathbf{1} + \mathbf{Z}_i \mathbf{g}_i^N + \mathbf{e}_i \quad [2]$$

where \mathbf{y}_i is phenotype after absorbing all other effects (including the residual polygenic effect)

$$\text{var}(\mathbf{e}_i) = \mathbf{R}_i^{-1} = \text{diag}\{n_{ik} \sigma_{e_i}^{-2}\}$$

$$\text{var}(\mathbf{g}_i) = \mathbf{B}_i \sigma_i^2 \quad \text{with DGV variance } \sigma_i^2$$

$$\mathbf{B}_i = \frac{1}{\sum_j 2p_{ij}(1-p_{ij})} \mathbf{I} = \theta_i \mathbf{I}$$



SNP genetic (co)variances between countries (I)

- A (co)variance matrix for countries ($i = 1, \dots, c$) **for a single SNP marker**

$$\mathbf{G}_{cou} = \begin{bmatrix} \sigma_1^2 \theta_1 & r_{12} \sigma_1 \sigma_2 \sqrt{\theta_1 \theta_2} & \cdots & r_{1c} \sigma_1 \sigma_c \sqrt{\theta_1 \theta_c} \\ & \sigma_2^2 \theta_2 & \cdots & r_{2c} \sigma_2 \sigma_c \sqrt{\theta_2 \theta_c} \\ & & \ddots & \vdots \\ & & & \sigma_c^2 \theta_c \end{bmatrix} = \begin{bmatrix} g_{11} & g_{12} & \cdots & g_{1c} \\ & g_{22} & \cdots & g_{2c} \\ & & \ddots & \vdots \\ \text{symm.} & & & g_{cc} \end{bmatrix}$$

- Its inverse $\mathbf{G}_{cou}^{-1} = \begin{bmatrix} g^{11} & g^{12} & \cdots & g^{1c} \\ & g^{22} & \cdots & g^{2c} \\ & & \ddots & \vdots \\ \text{symm.} & & & g^{cc} \end{bmatrix}$



SNP genetic (co)variances between countries (II)

- Genetic (co)variance matrix for **ALL** SNP effects (**ordered by countries**)

$$\text{var} \begin{bmatrix} \mathbf{g}_1 \\ \mathbf{g}_2 \\ \vdots \\ \mathbf{g}_c \end{bmatrix} = \mathbf{G} = \mathbf{G}_{\text{cou}} \otimes \mathbf{I} = \begin{bmatrix} g_{11}\mathbf{I} & g_{12}\mathbf{I} & \cdots & g_{1c}\mathbf{I} \\ & g_{22}\mathbf{I} & \cdots & g_{2c}\mathbf{I} \\ & & \ddots & \vdots \\ \text{symm.} & & & g_{cc}\mathbf{I} \end{bmatrix}$$

- Its inverse $\mathbf{G}^{-1} = \mathbf{G}_{\text{cou}}^{-1} \otimes \mathbf{I} = \begin{bmatrix} g^{11}\mathbf{I} & g^{12}\mathbf{I} & \cdots & g^{1c}\mathbf{I} \\ & g^{22}\mathbf{I} & \cdots & g^{2c}\mathbf{I} \\ & & \ddots & \vdots \\ \text{symm.} & & & g^{cc}\mathbf{I} \end{bmatrix}$

- Inter-SNP genetic correlations: within or between countries are all 0
- Intra-SNP genetic correlations between countries to be estimated
 - Set to country correlations as in current MACF



Solving the mixed model equations

- MME of the SNP MACE model have special structures:
 - Data contribution by country, zero residual covariances (off-diagonals)
 - SNP genetic contribution: only diagonal and sub-diagonals $\neq 0$
 - Block-diagonal matrix in the SNP-major order
- Identical processes for every country or every SNP \rightarrow parallel computing
- PCG algorithm using multiple cores

$$\mathbf{C} \mathbf{v} = \{ \mathbf{Z}'_i \mathbf{R}_i^{-1} \mathbf{Z}_i \} \mathbf{v} + \{ \mathbf{G}_{cou}^{-1} \} \mathbf{v}$$

for country i

parallelised by countries

for every SNP marker j

parallelised by SNP markers

- Conditioner may be the inverted diagonal block for country i

$$\mathbf{M}_i = (\mathbf{Z}'_i \mathbf{R}_i^{-1} \mathbf{Z}_i + \mathbf{G}^{ii})^{-1}$$

the matrix \mathbf{M}_i is also used in reliability calculation.



Calculation of prediction error (co)variances

- Countries need to calculate reliabilities of DGV (sum of all SNP effects)
- Not only reliabilities of MACE SNP effect estimates but also (co)reliabilities between the SNP effect estimates
 - the whole PEC block of 50k x 50k

- Absorbing all the other countries into own SNP effects

$$\begin{aligned}
 \mathbf{C}_i &= (\mathbf{Z}'_i \mathbf{R}_i^{-1} \mathbf{Z}_i + \mathbf{G}^{ii}) - \sum_{j \neq i} \mathbf{G}^{ij} (\mathbf{Z}'_j \mathbf{R}_j^{-1} \mathbf{Z}_j + \mathbf{G}^{jj})^{-1} \mathbf{G}^{ji} \\
 &= (\mathbf{Z}'_i \mathbf{R}_i^{-1} \mathbf{Z}_i + g^{ii} \mathbf{I}) - \sum_{j \neq i} g^{ij} \mathbf{I} (\mathbf{Z}'_j \mathbf{R}_j^{-1} \mathbf{Z}_j + \mathbf{G}^{jj})^{-1} g^{ji} \mathbf{I} \\
 &= (\mathbf{Z}'_i \mathbf{R}_i^{-1} \mathbf{Z}_i + g^{ii} \mathbf{I}) - (g^{ij})^2 \sum_{j \neq i} \mathbf{M}_j
 \end{aligned}$$

- Invert the own block matrix \mathbf{C}_i^{-1}
- Provide the PEC matrix to countries \mathbf{C}_i^{-1}



Three methods for handling different sets of SNP markers

- Method 1: conversion of country SNP effects to a common set
- Method 2: conversion of SNP effects for GBLUP models
- Method 3: direct modelling heterogeneous sets of SNP markers



Method 1: Conversion of country SNP effects to a common set of SNP markers

- SNP effects of national set of SNP markers for i -th country:
- SNP effects of a common set of SNP markers:
- Define DGV of all reference animals with own set of SNP markers:

$$\mathbf{g}_c^N$$

$$\mathbf{g}_c^N$$

$$\mathbf{u}_i = \mathbf{Z}_i \mathbf{g}_i^N$$

- A SNP BLUP model is fitted to model the DGV of reference animals:

$$\mathbf{u}_i = \mathbf{Z}_i^c \mathbf{g}_c^N + \xi$$

$$(\mathbf{Z}_i^{c'} \mathbf{R}_i^{-1} \mathbf{Z}_i^c + \sigma_i^{-2} \mathbf{B}_c^{-1}) \mathbf{g}_c^N = \mathbf{Z}_i^{c'} \mathbf{R}_i^{-1} \mathbf{u}_i$$

$$\mathbf{g}_c^N = (\mathbf{Z}_i^{c'} \mathbf{R}_i^{-1} \mathbf{Z}_i^c + \sigma_i^{-2} \mathbf{B}_c^{-1})^{-1} \mathbf{Z}_i^{c'} \mathbf{R}_i^{-1} (\mathbf{Z}_i \mathbf{g}_i^N)$$

- Additional data needed for the conversion

$$\mathbf{Z}_i' \mathbf{R}_i^{-1} \mathbf{Z}_i \quad \mathbf{Z}_i^{c'} \mathbf{R}_i^{-1} \mathbf{Z}_i^c$$

in addition to

$$\mathbf{Z}_i' \mathbf{R}_i^{-1} \mathbf{Z}_i$$

- Back conversion of MACE SNP effect estimates to the own SNP set



Method 2: Conversion of country SNP effects for GBLUP models

- Country uses a GBLUP model with its own SNP set
- Assumption: equal GEBV for reference animals with both SNP sets
- For all reference animals: GEBV vector \mathbf{u}_i^*
- Genomic relationship matrix for all reference animals is **invertible**:
- Estimate SNP effects of the common set

$$\mathbf{G}_{rel}^{-1}$$

$$\mathbf{g}_i^c = (1 - k)\mathbf{B}_c \mathbf{Z}_i^c{}' \mathbf{G}_{rel}^{-1} \mathbf{u}_i^*$$

- Equal genomic relationship matrices

$$(1 - k)\mathbf{Z}_i^c{}' \mathbf{B}_c \mathbf{Z}_i^c + k\mathbf{A}_i = (1 - k)\mathbf{Z}_i{}' \mathbf{B}_i \mathbf{Z}_i + k\mathbf{A}_i$$



Summary

- The SNP MACE model is an efficient tool for utilizing phenotype info of foreign reference animals
 - Particularly useful for new traits with large-scale genotyped cows
- No requirement for direct access to original national genotype and phenotype data
 - Keep the current infra-structure of national evaluation systems
- Parallel computing for efficiently solving the SNP MACE equations
 - No more pedigree relationship matrix, difficult to be parallelized
- Direct modelling different sets of SNP reduces the need for conversion to a common set of SNP markers
- A gain in accuracy of prediction is expected, especially for novel traits



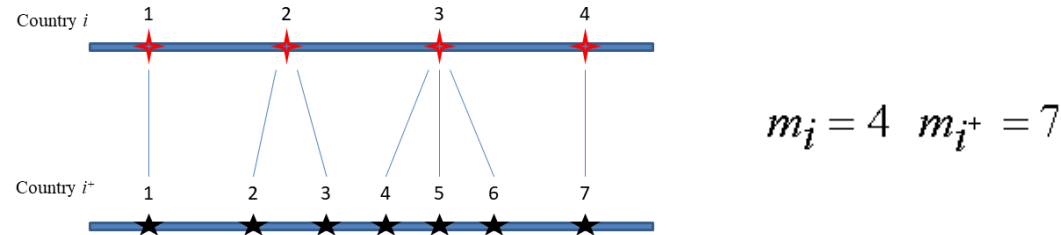


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Method 3: Direct modelling of SNP effects of different marker sets

- Cross reference two SNP sets for a country pair (same allele coding)



- SNP effect covariance matrix

$$\sigma_{i,i^+} \mathbf{B}_{i,i^+} = \sigma_{i,i^+} \sqrt{\theta_i \theta_{i^+}} \mathbf{E}_{i,i^+}$$

$$\mathbf{E}_{i,i^+} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

- Same modelling of **SNP/DGV variances** as in national genomic evaluations
- Correct **covariance of DGV** for any pair of countries
- But **covariances of SNP** are only correct for countries with fewer markers



Considering the SNP array differences

- Modelling SNP effect covariances between countries
 - Country A with more SNPs with unequal SNP covariance: less optimal
 - Country B with fewer SNPs with equal SNP covariance: exact modelling
 - But **DGV covariance** is correct for both countries
- For every one of all country pairs:
 - Set up a SNP cross-reference table
 - Determine the SNP covariances for each country pair
 - In case of a change in SNP arrays in one country, re-do the SNP cross-referencing with all the other countries
- Advantages of the procedure Method 3:
 - Countries do not have to be forced to use the same SNP arrays
 - SNP effects conversion to the common SNP set is not needed
- Disadvantages of Method 3:
 - More work of the SNP MACE, particularly when countries change their SNP sets
 - Approximate inverse of **G** matrix

Residual covariances to be modelled as the procedure for genetic covariances



Introduction

- Interbull MACE / GMACE evaluation for bulls / genomic bulls
 - Based on national conventional / genomic evaluation

- National genomic evaluation uses MACE EBV of foreign reference bulls
 - Significant increase in accuracy of genomic prediction
 - Fear of domination of foreign reference bulls on own SNP effects
 - Negative impact of genomic pre-selection on conventional EBV of bulls
 - Single-step national evaluation beneficial

- LD info of foreign reference cows NOT used in own SNP effect estimation
 - More countries add cows into national reference population
 - No MACE for cows, exchanging genotype of millions of cows infeasible

- Novel traits have relatively small national reference population
 - MACE bull evaluation perhaps not ready yet
 - Expected to have the largest gain in accuracy of prediction



An example with 2 countries and 3 SNP markers (1)

- Data contribution: least squares parts

$$\begin{bmatrix}
 \varphi_{\mu_1} & \varphi_{\mu_1 s_1} & \varphi_{\mu_1 s_2} & \varphi_{\mu_1 s_3} & 0 & 0 & 0 & 0 \\
 \varphi_{\mu_1 s_1} & \varphi_{1 s_{11}} & \varphi_{1 s_{12}} & \varphi_{1 s_{13}} & 0 & & & \\
 \varphi_{\mu_1 s_2} & \varphi_{1 s_{21}} & \varphi_{1 s_{22}} & \varphi_{1 s_{23}} & 0 & & & \\
 \varphi_{\mu_1 s_3} & \varphi_{1 s_{31}} & \varphi_{1 s_{32}} & \varphi_{1 s_{33}} & 0 & & & \\
 0 & 0 & 0 & 0 & \varphi_{\mu_2} & \varphi_{\mu_2 s_1} & \varphi_{\mu_2 s_2} & \varphi_{\mu_2 s_3} \\
 0 & & & & \varphi_{\mu_2 s_1} & \varphi_{2 s_{11}} & \varphi_{2 s_{12}} & \varphi_{2 s_{13}} \\
 0 & & & & \varphi_{\mu_2 s_2} & \varphi_{2 s_{21}} & \varphi_{2 s_{22}} & \varphi_{2 s_{23}} \\
 0 & & & & \varphi_{\mu_2 s_3} & \varphi_{2 s_{31}} & \varphi_{2 s_{32}} & \varphi_{2 s_{33}}
 \end{bmatrix}
 \begin{bmatrix}
 \hat{\mu}_1 \\
 \hat{g}_{1s_1} \\
 \hat{g}_{1s_2} \\
 \hat{g}_{1s_3} \\
 \hat{\mu}_2 \\
 \hat{g}_{2s_1} \\
 \hat{g}_{2s_2} \\
 \hat{g}_{2s_3}
 \end{bmatrix}
 =
 \begin{bmatrix}
 \Delta_{\mu_1} \\
 \Delta_{1s_1} \\
 \Delta_{1s_2} \\
 \Delta_{1s_3} \\
 \Delta_{\mu_2} \\
 \Delta_{2s_1} \\
 \Delta_{2s_2} \\
 \Delta_{2s_3}
 \end{bmatrix}$$

- Only the products of matrices or vectors are available, not the matrices or vectors themselves

$$\mathbf{Z}_i' \mathbf{R}_i^{-1} \mathbf{Z}_i \rightarrow \{\varphi_{1 s_{11}}\}$$

$$\mathbf{Z}_i' \mathbf{R}_i^{-1} \mathbf{y}_i \rightarrow \{\Delta_{2 s_3}\}$$



An example with 2 countries and 3 SNP markers (2)

- Data contribution: least squares parts + SNP genetic parts

$$\begin{bmatrix}
 \varphi_{\mu_1} & \varphi_{\mu_1 s_1} & \varphi_{\mu_1 s_2} & \varphi_{\mu_1 s_3} & 0 & 0 & 0 & 0 \\
 \varphi_{\mu_1 s_1} & \varphi_{1 s_{11}} & \varphi_{1 s_{12}} & \varphi_{1 s_{13}} & 0 & 0 & 0 & 0 \\
 \varphi_{\mu_1 s_2} & \varphi_{1 s_{21}} & \varphi_{1 s_{22}} & \varphi_{1 s_{23}} & 0 & 0 & 0 & 0 \\
 \varphi_{\mu_1 s_3} & \varphi_{1 s_{31}} & \varphi_{1 s_{32}} & \varphi_{1 s_{33}} & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & \varphi_{\mu_2} & \varphi_{\mu_2 s_1} & \varphi_{\mu_2 s_2} & \varphi_{\mu_2 s_3} \\
 0 & & & & \varphi_{\mu_2 s_1} & \varphi_{2 s_{11}} & \varphi_{2 s_{12}} & \varphi_{2 s_{13}} \\
 0 & & & & \varphi_{\mu_2 s_2} & \varphi_{2 s_{21}} & \varphi_{2 s_{22}} & \varphi_{2 s_{23}} \\
 0 & & & & \varphi_{\mu_2 s_3} & \varphi_{2 s_{31}} & \varphi_{2 s_{32}} & \varphi_{2 s_{33}}
 \end{bmatrix}
 \begin{bmatrix}
 \hat{\mu}_1 \\
 \hat{g}_{1s_1} \\
 \hat{g}_{1s_2} \\
 \hat{g}_{1s_3} \\
 \hat{\mu}_2 \\
 \hat{g}_{2s_1} \\
 \hat{g}_{2s_2} \\
 \hat{g}_{2s_3}
 \end{bmatrix}
 =
 \begin{bmatrix}
 \Delta_{\mu_1} \\
 \Delta_{1s_1} \\
 \Delta_{1s_2} \\
 \Delta_{1s_3} \\
 \Delta_{\mu_2} \\
 \Delta_{2s_1} \\
 \Delta_{2s_2} \\
 \Delta_{2s_3}
 \end{bmatrix}$$

The matrix is partitioned into blocks. The top-left block is a 4x4 matrix of SNP effects for country 1. The top-right block is a 4x4 zero matrix. The bottom-left block is a 4x4 zero matrix. The bottom-right block is an 8x8 matrix of SNP effects for country 2. The diagonal blocks are augmented with identity matrices: $g^{11}I$ for the top-left 4x4 block and $g^{22}I$ for the bottom-right 8x8 block.

- Order: SNP markers within country (in country-major order)

An example with 2 countries and 3 SNP markers (3)



$$\begin{array}{c}
 \left[\begin{array}{cc|cc|c}
 \varphi_{\mu_1} & & \varphi_{\mu_1 s_1} & & \varphi_{\mu_1 s_2} & & \varphi_{\mu_1 s_3} & & \\
 & \varphi_{\mu_2} & & \varphi_{\mu_2 s_1} & & \varphi_{\mu_2 s_2} & & \varphi_{\mu_2 s_3} & \\
 \hline
 \varphi_{\mu_1 s_1} & & \varphi_{1 s_{11}} + \mathbf{G}_{cou}^{-1} & & \varphi_{1 s_{12}} & & \varphi_{1 s_{13}} & & \\
 & \varphi_{\mu_2 s_1} & & \varphi_{2 s_{11}} & & \varphi_{2 s_{12}} & & \varphi_{2 s_{13}} & \\
 \hline
 \varphi_{\mu_1 s_2} & & \varphi_{1 s_{21}} & & \varphi_{1 s_{22}} + \mathbf{G}_{cou}^{-1} & & \varphi_{1 s_{23}} & & \\
 & \varphi_{\mu_2 s_2} & & \varphi_{2 s_{21}} & & \varphi_{2 s_{22}} & & \varphi_{2 s_{23}} & \\
 \hline
 \varphi_{\mu_1 s_3} & & \varphi_{1 s_{31}} & & \varphi_{1 s_{32}} & & \varphi_{1 s_{33}} + \mathbf{G}_{cou}^{-1} & & \\
 & \varphi_{\mu_2 s_3} & & \varphi_{2 s_{31}} & & \varphi_{2 s_{32}} & & \varphi_{2 s_{33}} & \\
 \hline
 \end{array} \right.
 \end{array}$$

- Order: countries within SNP marker (in SNP-major order)

$$\mathbf{G}_{cou}^{-1} = \begin{bmatrix} \sigma_1^2 \theta_1 & \sigma_{12} \sqrt{\theta_1 \theta_2} \\ \sigma_{12} \sqrt{\theta_1 \theta_2} & \sigma_2^2 \theta_2 \end{bmatrix}^{-1}$$



Why not fitting the residual polygenic effect?

- Needs to know the identifications of reference animals
- Needs to directly access genotypes and phenotypes of reference animals
 - Keep the infra-structure of current national evaluation systems
- In future, millions of cows will be added to ref. pop. worldwide
 - Exchange of genotypes of millions of reference cows may be infeasible
 - Estimating RPG of the millions of cows for all countries is challenging



A SNP MACE model: SNP effect covariances

Country i ; SNP marker k ; reference animal j $k = 1, \dots, n$

- Countries may have different sets of SNP markers

$$\text{var} \begin{bmatrix} \mathbf{g}_1 \\ \mathbf{g}_2 \\ \vdots \\ \mathbf{g}_c \end{bmatrix} = \begin{bmatrix} \sigma_1^2 \mathbf{B}_1 & \sigma_{12} \mathbf{B}_{12} & \cdots & \sigma_{1c} \mathbf{B}_{1c} \\ & \sigma_2^2 \mathbf{B}_2 & \cdots & \sigma_{2c} \mathbf{B}_{2c} \\ & & \ddots & \vdots \\ & & & \sigma_c^2 \mathbf{B}_c \end{bmatrix} = \mathbf{G}$$

symm.

DGV variance of country i ,

DGV covariance between countries i and i^+ .

$$\mathbf{B}_{i,i^+} = \frac{1}{\sqrt{\sum_j 2p_{ij}(1-p_{ij})} \sqrt{\sum_j 2p_{i^+j}(1-p_{i^+j})}} \mathbf{I} = \sqrt{\theta\theta} \mathbf{I}$$

for same SNP set

\mathbf{B}_{i,i^+} is $\sqrt{\theta\theta} \mathbf{I}$ squared matrix for two different SNP sets



National data for the SNP MACE evaluation: replacing the deregression step of national bull EBV in MACE

- $$\begin{bmatrix} \mathbf{1}'\mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{Z}_i\mathbf{1} & \mathbf{1}'\mathbf{R}_i^{-1}\mathbf{Z}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{1} & \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{Z}_i + \sigma_i^{-2}\mathbf{B}_i^{-1} \end{bmatrix} \begin{bmatrix} \hat{\mu}_i \\ \hat{\mathbf{g}}_i^N \end{bmatrix} = \begin{bmatrix} \mathbf{1}'\mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{y}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{y}_i \end{bmatrix}$$

- Re-written as:

$$\left(\Psi_i + \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \sigma_i^{-2}\mathbf{B}_i^{-1} \end{bmatrix} \right) \begin{bmatrix} \hat{\mu}_i \\ \hat{\mathbf{g}}_i^N \end{bmatrix} = \Delta_i$$

- Least-square part of the LHS of MME:

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

- Right-hand-side of the MME:

$$\Delta_i = \begin{bmatrix} \mathbf{1}'\mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{y}_i \\ \mathbf{Z}_i'\mathbf{R}_i^{-1}\mathbf{y}_i \end{bmatrix}$$



Method 2: Conversion of country SNP effects using the genomic relationship matrix: **DGV**

- Country uses a SNP BLUP model with its own SNP set
- Assumption: equal DGV for reference animals with both SNP sets
- DGV genomic relationship matrix for all reference animals is **invertible**:

$$\mathbf{G}_{rel_c}^{-1} = (\mathbf{Z}_i^c \mathbf{B}_c \mathbf{Z}_i^c)^{-1} = (\mathbf{Z}_i \mathbf{B}_i \mathbf{Z}_i)^{-1} = \mathbf{G}_{rel_i}^{-1}$$

- SNP marker cross-referencing (7 own SNPs, 4 common SNPs)

$$\mathbf{B}_{c,i} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

- SNP effects converted to the common set:

$$\mathbf{g}_i^c = \mathbf{B}_{c,i} \mathbf{Z}_i \mathbf{G}_{rel_i}^{-1} (\mathbf{Z}_i \mathbf{g}_i)$$

- Back conversion of MACE SNP effects to own SNP set:

$$\mathbf{g}_i = \mathbf{B}_{c,i} \mathbf{Z}_c \mathbf{G}_{rel_c}^{-1} (\mathbf{Z}_c \mathbf{g}_i^c)$$

