# Compatibility of pedigree-based and marker-based relationships for single-step genomic prediction

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EAAP 2012, Bratislava

### Single-step genetic evaluation

 Combines phenotypes, genomic and pedigree information using a combined relationship matrix (Misztal, Legarra, Aguilar + coworkers, Christensen and Lund).

 $\hfill\square$  Inverse of this matrix

$$H^{-1} = \begin{bmatrix} G^{-1} - A_{11}^{-1} & 0 \\ 0 & 0 \end{bmatrix} + A^{-1}$$

 $\square$   $A_{11}$  and G need to be "compatible".

 Aim here: Provide explanation; Show a possible way to handle it.

#### The idea behind single-step methods

 $\Box$  Two types of data: phenotypes y and markers m ( -1, 0, 1)

 $\Box$  Some animals genotyped  $(m^{obs})$  but others are not  $(m^{miss})$ .

 $\hfill\square$  This is a "missing data problem" !

 $\Box$  A model is specified for the "full data":  $f(y, m^{obs}, m^{miss})$ 

□ Marginalisation:

$$f(y, m^{obs}) = \int f(y, m^{obs}, m^{miss}) dm^{miss}$$

 $\Box$   $f(y, m^{obs})$  should be used for inference.

#### The model behind single-step methods

 $\Box$  Phenotypes conditional on all markers m:

 $y = \mu + a + e$ 

where  $a \sim N(0, \sigma_a^2 G(m))$  with

$$G(m) = \sum_{j} (m_j - (2\rho_j - 1))(m_j - (2\rho_j - 1))^T / s$$

and  $\rho_j$ 's are allele frequencies.

 $\Box$  Markers:  $m_j \sim N((2\rho_j - 1)1, v_j A)$ 

#### The model behind single-step methods

By marginalisation (integrating  $m^{miss}$ )

 $\Box$  Phenotypes conditional on observed markers  $m^{obs}$ :

$$y = \mu + a + e$$

where  $\operatorname{Var}(a) = \sigma_a^2 H$  with

$$H^{-1} = \begin{bmatrix} G(m^{obs})^{-1} - A_{11}^{-1} & 0\\ 0 & 0 \end{bmatrix} + A^{-1}$$

and

$$G(m^{obs}) = \sum_{j} (m_j^{obs} - (2\rho_j - 1))(m_j^{obs} - (2\rho_j - 1))^T / s$$

 $\Box$  Observed markers:  $m_j^{obs} \sim N((2\rho_j - 1)1, v_j A_{11})$ 

## Compatibility issue

- $\Box \ G(m^{obs})$  and  $A_{11}$  need to be "compatible"
- $\Box$  Allele frequencies  $\rho_j$  and scaling  $s = \sum_j v_j$  used to make compatible.
- □ log-Likelihood for parameter estimation:

$$\ell_{y,m^{obs}}(\sigma_{a}^{2},\sigma_{e}^{2},\rho,s) = \ell_{y|m^{obs}}(\sigma_{a}^{2},\sigma_{e}^{2},\rho,s) + \ell_{m^{obs}}(\rho,v)$$

- □ Allele frequencies  $\rho$  enter into both terms !, but maximising  $\ell_{y,m^{obs}}$  numerically is not feasible computationally.
- $\Box$  Various adjustments of  $G(m^{obs})$  used in practice.

#### Alternative approach: adjusting A instead

 $\Box$  Phenotypes conditional on observed markers  $m^{obs}$ :

$$y = \mu + a + e$$

where  $\operatorname{Var}(a) = \sigma_a^2 \tilde{H}$  with

$$\tilde{H}^{-1} = \begin{bmatrix} \tilde{G} - (\tilde{A}_{11}(\alpha))^{-1} & 0 \\ 0 & 0 \end{bmatrix} + (\tilde{A}(\alpha))^{-1}$$

with

$$\tilde{G} = \sum_{j} (m_{j}^{obs}) (m_{j}^{obs})^{T} / \tilde{s}$$

 $\Box$  Observed markers:  $m_j^{obs} \sim N(0, (\tilde{s}/p)\tilde{A}_{11}(\alpha))$ 

## **Relationship matrix** $\tilde{A}(\alpha)$

- □ Founders in the pedigree are related (coefficient= $\alpha$ ) and inbreed (coefficient= $\alpha/2$ ).
- $\Box$   $\tilde{A}(\alpha)$  is defined recursively in the usual way.

 $\Box$  Inverse:  $\left[ (\tilde{A}_{\alpha}) \right]$ 

$$\tilde{A}^{-1} = (T^{-1})^{\mathrm{T}} \begin{bmatrix} (\tilde{A}_0)^{-1} & 0\\ 0 & \tilde{D}^{-1} \end{bmatrix} T^{-1}.$$

 $\Box$  Colleau algorithm for computing  $A_{11}(\alpha)$  also exist.

 $\hfill\square$  Fast computing procedure still exists !

## Compatibility issue when adjusting $\boldsymbol{A}$

 $\Box~\tilde{G}=\sum_{j}(m_{j}^{obs})(m_{j}^{obs})^{T}/\tilde{s}$  and  $\tilde{A}_{11}(\alpha)$  need to be "compatible"

- $\Box$  Two parameters,  $\alpha$  and scaling parameter  $\tilde{s}$  used to make compatible.
- □ log-Likelihood for parameter estimation:

$$\ell_{y,m^{obs}}(\sigma_a^2,\sigma_e^2,\alpha,\tilde{s}) = \ell_{y|m^{obs}}(\sigma_a^2,\sigma_e^2,\alpha,\tilde{s}) + \ell_{m^{obs}}(\alpha,\tilde{s})$$

□ Parameters  $\alpha$  and  $\tilde{s}$  enter into both terms, and maximising  $\ell_{y,m^{obs}}$  numerically is computationally feasible.

## Conclusion

 $\Box$  Compatibility of G and A: The meaning is that certain parameters should be fitted to data (in theory both phenotypes and observed markers).

 $\Box$  An approach where A contains parameter(s) provides an interesting alternative.