

Genetic Covariances Between Animals

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BLUP Animal Model

- ▶ Traditional prediction of breeding values
 - ▶ i.e. no genomic information available
 - ▶ de-facto standard method
- ▶ Model recap

$$y = X\beta + Za + e$$

where

- y vector of length n of phenotypic information
- β vector of length p of unknown fixed effects
- X $n \times p$ incidence matrix
- a vector of length q of unknown random breeding values
- Z $n \times q$ incidence matrix
- e vector of length n of unknown random residuals

Numerical Example

Calf	Sire	Dam	Herd	WWG
3	1	2	1	4.5
4	1	NA	2	2.9
5	4	3	2	3.9
6	5	2	1	3.5

Animal Model Setup

- ▶ Convert information from data into model
- ▶ Put observations in vector y

$$y = \begin{bmatrix} 4.5 \\ 2.9 \\ 3.9 \\ 3.5 \end{bmatrix}$$

- ▶ Fixed effects such as herd go in vector β and X links observations to components in β

$$X = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}, \beta = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}$$

Breeding Values As Random Effects and Residuals

$$Z = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}, a = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_6 \end{bmatrix}, e = \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \end{bmatrix}$$

Putting Everything Together

$$\begin{bmatrix} 4.5 \\ 2.9 \\ 3.9 \\ 3.5 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_6 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \end{bmatrix}$$

Solution with Mixed Model Equations

- ▶ General form

$$\begin{bmatrix} X^T R^{-1} X & X^T R^{-1} Z \\ Z^T R^{-1} X & Z^T R^{-1} Z + G^{-1} \end{bmatrix} \begin{bmatrix} \hat{\beta} \\ \hat{a} \end{bmatrix} = \begin{bmatrix} X^T R^{-1} y \\ Z^T R^{-1} y \end{bmatrix}$$

- ▶ Assume $R = I * \sigma_e^2$ and we set $G = A * \sigma_a^2$

$$\begin{bmatrix} X^T X & X^T Z \\ Z^T X & Z^T Z + \lambda * A^{-1} \end{bmatrix} \begin{bmatrix} \hat{\beta} \\ \hat{a} \end{bmatrix} = \begin{bmatrix} X^T y \\ Z^T y \end{bmatrix}$$

with $\lambda = \sigma_e^2 / \sigma_a^2$

Components of Mixed Model Equations

$$X^T X = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, X^T Z = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}$$

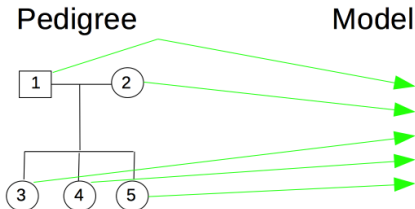
$$Z^T Z = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}, X^T y = \begin{bmatrix} 8 \\ 7 \end{bmatrix}, Z^T y = \begin{bmatrix} 0 \\ 0 \\ 4 \\ 3 \\ 4 \\ 4 \end{bmatrix}$$

Numerator Relationship Matrix

- ▶ Only component missing in mixed model equations: matrix A
- ▶ A is called the **numerator relationship matrix**
- ▶ Properties of A

$$\text{var}(a) = G = A * \sigma_a^2$$

Meaning of A



Variance-Covariance

	a_1	a_2	a_3	a_4	a_5
a_1	$\text{var}(a_1)$	$\text{cov}(a_1, a_2)$	$\text{cov}(a_1, a_3)$	$\text{cov}(a_1, a_4)$	$\text{cov}(a_1, a_5)$
a_2		$\text{var}(a_2)$	$\text{cov}(a_2, a_3)$	$\text{cov}(a_2, a_4)$	$\text{cov}(a_2, a_5)$
a_3			$\text{var}(a_3)$	$\text{cov}(a_3, a_4)$	$\text{cov}(a_3, a_5)$
a_4				$\text{var}(a_4)$	$\text{cov}(a_4, a_5)$
a_5					$\text{var}(a_5)$

Elements of A

- ▶ Diagonal elements $(A)_{ii} = 1 + F_i$ where F_i : inbreeding coefficient of i
- ▶ Offdiagonal elements $(A)_{ij} = cov(a_i, a_j)/\sigma_a^2$
- ▶ Symmetric
- ▶ Definition leads to recursive algorithm for computing components of A