

# Variance Components Estimation

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# Why

- ▶ Predictions of breeding values using BLUP requires variance components  $\sigma_u^2$  or  $\sigma_s^2$  and  $\sigma_e^2$
- ▶ So far we have assumed that they are known
- ▶ In reality: must be estimated from data

# Sire Model

- ▶ Start with a simple sire model

$$y = X\beta + Z_s s + e$$

with  $\text{var}(e) = R$ ,  $\text{var}(s) = A_s \sigma_s^2$  and  $\text{var}(y) = Z_s A_s Z_s^T \sigma_s^2 + R$

- ▶  $A_s$ : numerator relationship for sires
- ▶  $\sigma_s^2$  corresponds to  $0.25 * \sigma_u^2$
- ▶  $R = I * \sigma_e^2$

→ estimate  $\sigma_s^2$  and  $\sigma_e^2$  from data

# Analysis of Variance (ANOVA)

Source	Degrees of Freedom ( $df$ )	Sums of Squares ( $SSQ$ )
Overall ( $\mu$ )	$Rank(X) = 1$	$y^T X (X^T X)^{-1} X^T y = F$
Sires ( $s$ )	$Rank(Z_s) - Rank(X) = q - 1$	$y^T Z_s (Z_s^T Z_s)^{-1} Z_s^T y - y^T X (X^T X)^{-1} X^T y = S$
Residual ( $e$ )	$n - Rank(Z_s) = n - q$	$y^T y - y^T Z_s (Z_s^T Z_s)^{-1} Z_s^T y = R$
Total	$n$	$y^T y$

## Sums of Squares

$$F = y^T X(X^T X)^{-1} X^T y = \frac{1}{n} \left[ \sum_{i=1}^n y_i \right]^2$$

$$S = y^T Z_s(Z_s^T Z_s)^{-1} Z_s^T y - y^T X(X^T X)^{-1} X^T y = \sum_{i=1}^q \frac{1}{n_i} \left[ \sum_{j=1}^{n_i} y_{ij} \right]^2 - F$$

$$R = y^T y - y^T Z_s(Z_s^T Z_s)^{-1} Z_s^T y = \sum_{i=1}^n y_i^2 - S - F$$

# Estimates

- ▶  $\beta$  and  $s$  fixed
- ▶ Estimates of  $\sigma_e^2$  and  $\sigma_s^2$  are based on observed sums of squares  $S$  and  $R$
- ▶ Set their expected values equal to the observed sums of squares

$$E(R) = (n - q)\sigma_e^2$$

$$E(S) = (q - 1)\sigma_e^2 + \text{tr}(Z_s M Z_s)\sigma_s^2$$

where  $M = I - X(X^T X)^{-1}X^T$  and  $q$  is the number of sires.

$$\rightarrow \widehat{\sigma_e^2} = \frac{R}{n-q} \text{ and } \widehat{\sigma_s^2} = \frac{S - (q-1)\widehat{\sigma_e^2}}{\text{tr}(Z_s M Z_s)}$$

# Numerical Example

Table 1: Small Example Dataset for Variance Components Estimation Using a Sire Model

Animal	Sire	WWG
4	2	2.9
5	1	4.0
6	3	3.5
7	2	3.5

► Model

$$y_{ij} = \mu + s_j + e_i$$

## Design Matrices

$$X = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, Z_s = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$



# ANOVA

An analysis of variance can be constructed as

Source	Degrees of Freedom ( $df$ )	Sums of Squares ( $SSQ$ )
Overall ( $\mu$ )	$Rank(X) = 1$	$F = 48.3025$
Sires ( $s$ )	$Rank(Z_s) - Rank(X) = q - 1$	$S = 0.4275$
Residual ( $e$ )	$n - Rank(Z_s) = n - q$	$R = 0.18$

## Estimates

$$M = \begin{bmatrix} 0.75 & -0.25 & -0.25 & -0.25 \\ -0.25 & 0.75 & -0.25 & -0.25 \\ -0.25 & -0.25 & 0.75 & -0.25 \\ -0.25 & -0.25 & -0.25 & 0.75 \end{bmatrix}$$

$$Z_s^T M Z = \begin{bmatrix} 0.75 & -0.50 & -0.25 \\ -0.50 & 1.00 & -0.50 \\ -0.25 & -0.50 & 0.75 \end{bmatrix}$$

## Results

$$\hat{\sigma}_e^2 = R = 0.18$$

$$\hat{\sigma}_s^2 = \frac{S - (q - 1)\hat{\sigma}_e^2}{\text{tr}(Z_s^T M Z_s)} = \frac{0.4275 - 2 * 0.18}{2.5} = 0.027$$