

BLUP

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Recap: Linear Mixed Effects Model

$$y = X \cdot b + Z \cdot u + e$$

where

- y vector of phenotypic observations of all animals
- b vector of unknown fixed effects and unknown regression coefficients
- u vector of random breeding values of all animals
- e vector of random environmental effects
- X design matrix with regression covariates or indicator values for fixed effects
- Z design matrix relating breeding values to observations

Solution - BLUP

- ▶ Solution to correction problem in selection index: BLUP
- ▶ Estimates environmental effects at the same time as breeding values are predicted
- ▶ Linear mixed effects model
- ▶ Meaning of BLUP
 - ▶ **B** stands for **best** → correlation between true (u) and its prediction (\hat{u}) is maximal or the prediction error variance ($\text{var}(u - \hat{u})$) is minimal.
 - ▶ **L** stands for **linear** → predicted breeding values are linear functions of the observations (y)
 - ▶ **U** stands for **unbiased** → expected values of the predicted breeding values are equal to the true breeding values
 - ▶ **P** stands for **prediction**

Example Dataset I

Animal	Sire	Dam	Herd	Weaning Weight	Breast Circumference
14	1	5	1	2.44	1.48
15	1	5	1	2.41	1.47
16	1	6	2	2.51	1.50
17	1	6	2	2.55	1.47
20	2	8	1	2.34	1.46
21	2	8	1	1.99	1.38

Example Dataset II

Animal	Sire	Dam	Herd	Weaning Weight
12	1	4	1	2.61
13	1	4	1	2.31
14	1	5	1	2.44
15	1	5	1	2.41
16	1	6	2	2.51
17	1	6	2	2.55
18	1	7	2	2.14
19	1	7	2	2.61
20	2	8	1	2.34
21	2	8	1	1.99
22	2	9	1	3.10
23	2	9	1	2.81
24	2	10	2	2.14
25	2	10	2	2.41
26	3	11	2	2.54
27	3	11	2	3.16

Linear Models

- ▶ Simple linear model

$$y_{ij} = \mu + \textit{herd}_j + e_{ij}$$

- ▶ Result: Estimate of effect of herd j
- ▶ Try with given dataset

Linear Mixed Effects Model

- ▶ What about breeding value u_i for animal i ?
 - ▶ Problem: breeding values have a variance σ_u^2 and co-variances
 - ▶ Cannot be specified in simple linear model

→ **Linear Mixed Effects Model (LME)**

$$y_{ijk} = \mu + \beta_j + u_i + e_{ijk}$$

Matrix-Vector Notation

- ▶ LME for all animals of a population

→ use matrix-vector notation

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

where

- y vector of length n of all observations
- β vector of length p of all fixed effects
- X $n \times p$ design matrix linking the fixed effects to the observations
- u vector of length n_u of random effects
- Z $n \times n_u$ design matrix linking random effect to the observations
- e vector of length n of random residual effects.

Expected Values and Variances

- ▶ Expected values

$$E(u) = 0 \text{ and } E(e) = 0 \rightarrow E(y) = X\beta$$

- ▶ Variances

$$\text{var}(u) = U \text{ and } \text{var}(e) = R$$

with $\text{cov}(u, e^T) = 0$,

$$\text{var}(y) = Z * \text{var}(u) * Z^T + \text{var}(e) = ZUZ^T + R = V$$

Estimates of unknown Parameters

Use properties of BLUP to design predictor \hat{u}

- ▶ Linear: predictor \hat{u} is a linear function of data

$$\hat{u} = M \cdot (y - k)$$

- ▶ Unbiased: $E(\hat{u}) = E(u)$. Above we have specified $E(u) = 0$

$$k = E(y)$$

- ▶ Best: Prediction error variance $\text{var}(u - \hat{u})$ is minimal

$$M = UZ^T V^{-1}$$

Result

- ▶ Summary of above results, using $E(y) = X\beta$

$$\hat{u} = UZ^T V^{-1}(y - X\beta)$$

- ▶ For u and y following a Gaussian normal distribution

$$\hat{u} = E(u|y)$$

- ▶ β unknown, replace it by general least squares estimate

$$\hat{\beta} = (X^T V^{-1} X)^{-1} X^T V^{-1} y$$

Mixed Model Equations

- ▶ Problem: V^{-1}
- ▶ Same solutions obtained with following set of equations

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

Sire Model

- ▶ Breeding value of sire as random effect:

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

- * Expected values

$$E(s) = 0 \text{ and } E(e) = 0 \rightarrow E(y) = X\beta$$

- ▶ Variances

$$\text{var}(s) = U_s \text{ and } \text{var}(e) = R$$

- ▶ Unrelated sires: $\text{var}(s) = U_s = I\sigma_s^2$

Example

$$\begin{bmatrix} 2.61 \\ 2.31 \\ 2.44 \\ 2.41 \\ 2.51 \\ 2.55 \\ 2.14 \\ 2.61 \\ 2.34 \\ 1.99 \\ 3.1 \\ 2.81 \\ 2.14 \\ 2.41 \\ 2.54 \\ 3.16 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ 0 & 1 \\ 0 & 1 \\ 0 & 1 \\ 0 & 1 \\ 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ 0 & 1 \\ 0 & 1 \\ 0 & 1 \\ 0 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ s_3 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \\ e_6 \\ e_7 \\ e_8 \\ e_9 \\ e_{10} \\ e_{11} \\ e_{12} \\ e_{13} \\ e_{14} \\ e_{15} \\ e_{16} \end{bmatrix}$$

Animal Model

- ▶ Breeding value for all animals as random effects

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

- ▶ Expected values

$$E(u) = 0 \text{ and } E(e) = 0 \rightarrow E(y) = X\beta$$

- ▶ Variances

$$\text{var}(u) = U = A * \sigma_u^2 \text{ and } \text{var}(e) = R$$

- ▶ Matrix A : numerator relationship matrix and σ_u^2 the genetic additive variance