

# Multiple Traits

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## So Far ...

- ▶ Prediction of Breeding Values for **one trait**

→ **univariate** analyses

- ▶ In Livestock Breeding, populations are improved with respect to several traits

→ **multi-trait** or multiple trait

- ▶ Different selection strategies and different approaches of how data is analysed are possible

# Multiple Trait Selection

In principle, there are three possible strategies

- ▶ Selection index theory provides a tool for optimal integration of different sources of information (3)
- ▶ But still other strategies are applied
  - ▶ Tandem selection (1)
  - ▶ Selection based on independent thresholds (2)

# Tandem Selection

- ▶ Improve one trait at the time until they all reach a certain threshold
- ▶ Problem: For traits which are not improved
  - ▶ only correlated selection responses
  - ▶ can be negative  $\implies$  **while improving one trait, other traits would get worse**
- ▶ Populations with long generation intervals, response per year is very small

**Example for problematic situation: trait1: milk yield and trait2: fertility in cattle**

**1. increasing milk yield would decrease fertility**

**2. improve fertility would lower milk yield**

$\implies$  **infinite loop without significant progress**

# Independent Selection Thresholds

## Traits to improve

- mastitis resistance (above 95%)
- fertility (NRR above 80%)
- milk yield (above +400kg PBV)

- ▶ Applied before selection index
- ▶ Define selection thresholds in each of the traits
- ▶ Select animals as parents which are above thresholds for all traits

# Example

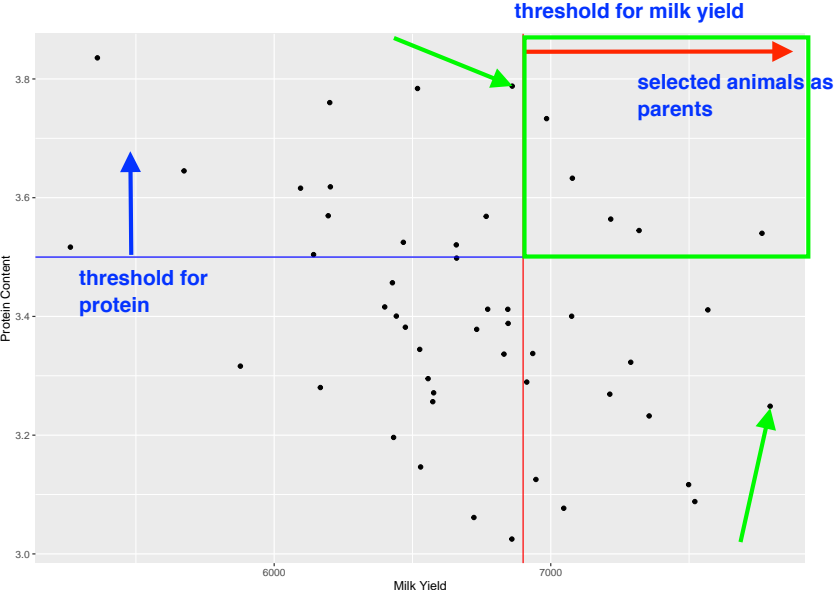


Figure 1: Milk Yield and Protein Content For Dairy Cows

# Pros and Cons

- ▶ Selection response in all traits
- ▶ Thresholds often set to only positive predicted breeding values in all traits

→ exclusion of very many animals and reduction in genetic variability

- ▶ Genetic relationships between traits ignored

→ genetic gain will not be as expected

3. Differences in the economic relevance ignored.

→ threshold in all traits above positive predicted breeding values emphasizes traits with high heritability

# Aggregate Genotype (Gesamtzuchtwert)

preferred selection strategy when populations are to be improved with respect to several traits at the same time.

- ▶ Define the set of important traits for which population should be improved
- ▶ Determine economic values  $w$  for these traits
- ▶ Aggregate genotype  $H$  follows as

Selection of animals as parents is best done when animals can be ranked according to a certain criterion. In our case this criterion is a single number. In univariate case with only one trait, this criterion is the predicted breeding value of each animal in the single trait.

$$H = w^T u$$

$u$  is the vector of true breeding values



## Selection Index

True Aggregate Genotype  $H$  cannot be observed, because the true breeding values  $u$ , cannot be observed. But we can predict  $H$  based on the PBV. This prediction of  $H$  is done using a method that is called selection index theory and the predicted  $H$  is called index  $I$

- ▶ Use index  $I$  to estimate  $H$  where  $I$  is a linear combination of information sources

$$I = b^T \hat{u}$$

- ▶ Index weights  $b$  are determined using selection index theory as

$$b = P^{-1} G w$$

- ▶ Information sources are predicted breeding values
- ▶ If traits in  $u$  and  $\hat{u}$  are the same and  $\hat{u}$  were estimated using BLUP, then  $b = w$

# Implementations

- ▶ First possible implementation
  - ▶ Do univariate predictions of breeding values using BLUP animal model
  - ▶ Combine  $\hat{u}$  with appropriate  $b$ -values
- ▶ Improvement
  - ▶ get  $\hat{u}$  from multivariate analysis

**second possibility of predicting breeding values with a multi-trait BLUP approach is preferred to the first one.**

# Multivariate Analysis

Here 'Multivariate' means: prediction of breeding values for more than one trait in the same analysis.  
Example: two traits  $y_1$  and  $y_2$  and they can be something like birth-weight and weaning weight

- ▶ Given two traits with univariate models

$$y_1 = X_1\beta_1 + Z_1u_1 + e_1$$

$$y_2 = X_2\beta_2 + Z_2u_2 + e_2$$

- ▶ Combine both univariate models by stacking one on top of the other, resulting in

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} Z_1 & 0 \\ 0 & Z_2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \end{bmatrix}$$

$y$  =  $X$   $\beta$

## Multivariate Model

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} Z_1 & 0 \\ 0 & Z_2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \end{bmatrix}$$

can be written as

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

$$\text{with } y = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}, \beta = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}, u = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}, e = \begin{bmatrix} e_1 \\ e_2 \end{bmatrix}$$

$$X = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix}, Z = \begin{bmatrix} Z_1 & 0 \\ 0 & Z_2 \end{bmatrix}$$

# Multivariate Variance-Covariance Matrices

Variance-Covariance Matrix between traits 1 and 2

genetic additive variance of trait 1

$$G_0 = \begin{bmatrix} \sigma_{g_1}^2 & \sigma_{g_1, g_2} \\ \sigma_{g_1, g_2} & \sigma_{g_2}^2 \end{bmatrix} = \begin{bmatrix} g_{11} & g_{12} \\ g_{21} & g_{22} \end{bmatrix}$$

genetic covariance between traits 1 and 2

genetic additive variance of trait 2

$$\text{var}(u) = \text{var} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} g_{11}A & g_{12}A \\ g_{21}A & g_{22}A \end{bmatrix} = G_0 \otimes A = G$$

When just looking at trait 2, this was  $A \cdot \sigma_u^2$  with  $A$  being the numerator relationship matrix

$$R = \text{var}(e) = \text{var} \begin{bmatrix} e_1 \\ e_2 \end{bmatrix} = \begin{bmatrix} r_{11}I_n & r_{12}I_n \\ r_{21}I_n & r_{22}I_n \end{bmatrix} = R_0 \otimes I_n$$

# Solutions

► Mixed Model Equations

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

**In a multivariate analysis, the simplified version of the mixed model equations are no longer usable.**

# Advantages

of the approach shown:

- \* prediction of breeding value using a multi-trait BLUP model
- \* combining multi-trait predicted breeding values into a selection index  $I$  to estimate the aggregate genotype.

- ▶ some traits have lower heritability than others
- ▶ environmental correlations exist between traits measured on the same animal
- ▶ some traits are available only a subset of all animals
- ▶ some traits were used for a first round of selection
- ▶ accuracies are higher in multivariate analyses

## Disadvantages

- \* more complexity
- \* more parameters to estimate from data (correlations  $G_0$  and  $R_0$ )
- \* more computing resources