

Livestock Breeding and Genomics - Solution 12

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2023-12-01

Problem 1: Accuracy of Predicted Breeding Values

Use the dataset given below to predict breeding values for the response variable `Weight` using a BLUP animal model. The model contains `Herd` as fixed effect and `BreastCircumference` as regression covariate. Compute reliabilities ($B\%$) for all predicted breeding values.

Hints

- The phenotypic variance σ_p^2 can be computed from the variance of the weight values given in the dataset.
- Heritability (h^2) is assumed to be 0.25 for the trait `Weight`.

Data

The dataset is available from

https://charlotte-ngs.github.io/lbgfs2023/data/beef_data_blup.csv

Solution

- Read the data

```
tbl_beef <- readr::read_delim(s_beef_data_path, delim = ",")

## Rows: 15 Columns: 6
## -- Column specification -----
## Delimiter: ","
## dbf (6): Animal, Sire, Dam, BreastCircumference, Herd, Weight
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

- Specify the linear mixed effects model

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

with vectors y : observations, β : fixed regression coefficient of **BreastCircumference** and fixed herd-effects, u : random breeding values, e : random residuals. Matrix X links covariates and fixed effects to observations and matrix Z related breeding values to observations.

Expected values of the random components are

$$E \begin{bmatrix} y \\ u \\ e \end{bmatrix} = \begin{bmatrix} X\beta \\ 0 \\ 0 \end{bmatrix}$$

Variance-Covariance matrices of the random components are

$$var \begin{bmatrix} y \\ u \\ e \end{bmatrix} = \begin{bmatrix} V & ZG & R \\ GZ^T & G & 0 \\ R & 0 & R \end{bmatrix}$$

with $R = I * \sigma_e^2$, $G = A\sigma_u^2$ and $V = ZGZ^T + R$. Matrix A is the numerator relationship matrix based on the pedigree and σ_e^2 and σ_u^2 are given variance components.

- Setup the mixed model equations (MME)

Mixed model equations for BLUE of β and BLUP of u are given by

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

Information from data are inserted to the MME. Start with matrix X

```
matX <- model.matrix(lm(Weight ~ 0 + BreastCircumference + as.factor(Herd), data = tbl_beef))
attr(matX, "assign") <- NULL
attr(matX, "contrasts") <- NULL
dimnames(matX) <- NULL
```

For matrix Z , we have to complete the pedigree

```
vec_fnd_sire <- setdiff(tbl_beef$Sire, tbl_beef$Animal)
vec_fnd_sire <- vec_fnd_sire[!is.na(vec_fnd_sire)]
vec_fnd_dam <- setdiff(tbl_beef$Dam, tbl_beef$Animal)
vec_fnd_dam <- vec_fnd_dam[!is.na(vec_fnd_dam)]
vec_fnd <- c(vec_fnd_sire, vec_fnd_dam)
n_nr_fnd <- length(vec_fnd)
n_nr_obs <- nrow(tbl_beef)
matZ <- cbind(matrix(rep(0, n_nr_fnd*n_nr_obs), nrow = n_nr_obs),
               diag(nrow = n_nr_obs))
```

The inverse numerator relationship A^{-1} is computed based on the pedigree

```
n_nr_anl <- n_nr_fnd + n_nr_obs
ped <- pedigreeemm::pedigree(sire = c(rep(NA, n_nr_fnd), tbl_beef$Sire),
                             dam = c(rep(NA, n_nr_fnd), tbl_beef$Dam),
                             label = c(1:n_nr_anl))
matA_inv <- pedigreeemm::getAInv(ped = ped)
```

```
## 'as(<dtTMatrix>, "dtCMatrix")' is deprecated.
## Use 'as(., "CsparseMatrix")' instead.
## See help("Deprecated") and help("Matrix-deprecated").
```

The coefficient matrix and the right-hand-sides of the MME are

```
lambda <- (1-n_h2_weight) / n_h2_weight
mat_coef <- rbind(cbind(crossprod(matX), crossprod(matX, matZ)),
                  cbind(crossprod(matZ, matX), crossprod(matZ) + lambda * matA_inv))
vec_y <- tbl_beef$Weight
mat_rhs <- rbind(crossprod(matX, vec_y),
                 crossprod(matZ, vec_y))
mat_sol <- solve(mat_coef, mat_rhs)
mat_sol
```

```
##           [,1]
## [1,]  8.732302e+00
## [2,] -1.068146e+03
## [3,] -1.083529e+03
## [4,] -1.101438e+03
## [5,]  1.313321e+00
## [6,] -2.114677e+00
## [7,]  1.524946e-01
## [8,]  2.757089e-01
## [9,] -6.930179e-01
## [10,] -1.594902e-01
## [11,]  2.757169e-01
## [12,]  9.499436e-01
## [13,]  8.854023e-01
## [14,]  6.245173e-01
## [15,]  8.252461e-01
## [16,]  1.478646e+00
## [17,]  5.187407e-01
## [18,]  4.999237e-01
## [19,] -1.838960e+00
## [20,] -3.993777e-02
## [21,] -1.157909e+00
## [22,] -2.820549e+00
## [23,] -1.219582e+00
## [24,]  1.775756e+00
## [25,]  1.325188e+00
## [26,]  1.405062e+00
## [27,]  5.855898e-01
```

- Get estimates of fixed effects and predictions of breeding values from solutions of MME

Estimates for the fixed effects are:

```
n_nr_sol <- nrow(mat_sol)
n_nr_fix <- n_nr_sol - n_nr_ani
mat_sol[1:n_nr_fix,]
```

```
## [1]      8.732302 -1068.145601 -1083.529349 -1101.438311
```

Predicted breeding values are:

```
mat_sol[(n_nr_fix + 1):n_nr_sol,]
```

```
## [1] 1.31332085 -2.11467663 0.15249457 0.27570891 -0.69301795 -0.15949022
## [7] 0.27571692 0.94994357 0.88540227 0.62451727 0.82524613 1.47864635
## [13] 0.51874075 0.49992374 -1.83895972 -0.03993777 -1.15790882 -2.82054922
## [19] -1.21958208 1.77575637 1.32518772 1.40506154 0.58558977
```

- Compute accuracies from inverse of coefficient matrix of MME