## Livestock Breeding and Genomics - Exercise 11

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## **Problem 1 Variance Components Estimation**

The simplest form of variance components estimation is based on the residuals of a fitted linear model and is shown in the summary results of the R-function lm(). Let us assume that we are given the dataset in the table shown below to which we fit a simple sire model.

Table 1: Example Dataset for Variance Components EstimationBased on Residuals Using a Sire Model

Animal	Sire	WWG
4	2	4.5
5	1	2.9
6	1	3.9
7	2	3.5
8	1	5.0

The sire model is simplified to have a common mean  $\mu$ . For a moment we are setting the sire effects to be fixed effects. This leads to the following model with  $var(e) = I * \sigma_e^2$ 

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

Using the above shown dataset we can use the R-function lm() to fit this simple linear model. Because, we want to have the sires as fixed effects, we have to convert them into factors before calling lm().

```
tbl_data_sol12p02$Sire <- as.factor(tbl_data_sol12p02$Sire)
lm_data_sol12p02 <- lm( WWG ~ 1 + Sire, data = tbl_data_sol12p02 )
summary(lm_data_sol12p02)</pre>
```

```
##
## Call:
## lm(formula = WWG ~ 1 + Sire, data = tbl_data_sol12p02)
##
## Residuals:
##
          1
                   2
                            3
                                      4
                                               5
##
   0.50000 -1.03333 -0.03333 -0.50000 1.06667
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 3.93333
                                      7.172 0.00558 **
                           0.54840
## Sire2
                0.06667
                           0.86709
                                      0.077 0.94356
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

##
## Residual standard error: 0.9499 on 3 degrees of freedom
## Multiple R-squared: 0.001967, Adjusted R-squared: -0.3307
## F-statistic: 0.005911 on 1 and 3 DF, p-value: 0.9436

From the output of summary() we are given the residual standard error to be 0.9499. This residual standard error is an estimate of  $\sigma_e$ . The question is where does it come from. The least-squares procedure does not yield this estimate for  $\sigma_e$ . The answer is that this estimate comes from the residuals r of the model. For our model the vector r of residuals is defined as

$$r = y - X\hat{\mu} - Z_s\hat{s}$$

where  $\hat{\mu}$  and  $\hat{s}$  can be taken from the ouput of the summary() function. They correspond to

$$\widehat{\mu} = 3.933333$$

$$\widehat{s} = \begin{bmatrix} 0 \\ 0.06666666666666666679 \end{bmatrix}$$

The estimate  $\widehat{\sigma_e^2}$  for  $\sigma_e^2$  is obtained by

$$\widehat{\sigma_e^2} = \frac{1}{n-p} \sum_{i=1}^n r_i^2$$

where n is the total number of observations and p is the number of parameters that are estimated by lm which is 2 for our sire model. The term n - p is also called **degrees of freedom** (df). What is given as residual standard error by the output of summary() is the square root of  $\widehat{\sigma_e^2}$ .

## Your Task

Verify for the above given dataset and the proposed sire model the residual standard error given by summary() by using the computation based on the residuals shown above.