Practical 2: Basic Quantitative Genetics

Time schedule of practical session 2:

11:15	Question to lectures, multiple-choice question and follow up on previous multiple choice questions
11:30	Today's exercise and assignment to groups
12:00	15 minutes break
12:30	Go through excercises using final word
12:50	Repeat multiple-choice questions
13:00	End of practical session 2

Introduction:

In this practical we use R for explorative data analyses of two quantitative traits, body weight and blood glucose levels, observed in the F2 mouse population. We will be characterizing and investigating the potential effects of a single marker locus. This includes computation of allele and genotype frequencies, evaluating different genetic models, and estimation of the breeding values and genetic variances for the single marker locus.

Furthermore you may also want to explore these shinyapps that could help understand some of the basic concepts of quantitative genetics:

https://neyhartj.shinyapps.io/qgshiny/

https://shiny.cnsgenomics.com/Falconer2/

Let's continue explore our mouse data

The mouse data set can be loaded using the following command:

mouse <- readRDS(url("https://github.com/psoerensen/bgcourse/raw/main/data/mouseqtl.rds"))</pre>

Question 1: How many observations and which variables do we have in the data set? To get a fast overview of the data set you are working with you can use the str function:

Answer:

Question 2: How many observations do the two marker variables have in each genotype class? Use the table function to explore the two marker variables:

Answer:

Question 2: What are the genotype and allele frequencies for M227? Include the allele and genotype frequencies for M227 in the following table:

Variable	M227
f_{AA}	
f_{AB}	
f_{BB}	
f_A	
f_B	

Question 3: Does the marker variable M227 potentially influence body weight and glucose? Use the boxplot function to visualize the potential effect of the marker variable M227 on the two traits:

Answer:

To best answer these question we can fit a linear model that also include the effect of the marker variable in addition to sex and reps. This can be done using the lm function:

```
fit <- lm(BW~ sex + reps + M227, data=mouse)
```

To test the effect of the variables in the model use the anova function on the fit object from the lm function:

```
anova(fit)
```

```
## Analysis of Variance Table
##
## Response: BW
##
              Df Sum Sq Mean Sq F value
## sex
               1 20542.9 20542.9 1203.352 < 2.2e-16 ***
               2 2195.9 1097.9
                                   64.315 < 2.2e-16 ***
## reps
## M227
               2 1660.3
                           830.1
                                   48.627 < 2.2e-16 ***
## Residuals 1168 19939.3
                            17.1
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

Question 4: Based on the linear model results do marker variable M227 influence body weight?

Answer:

The additive effect is modeled by a variable, add, with levels that is coded as -1, 0, and 1 (corresponding to -a, 0, a) for the genotypes AA, AB, and BB. The following lines of R code create a the add variable, fit the linear model and test the effects:

```
alleles \leftarrow c(-1,0,1)
names(alleles) <- c("AA", "AB", "BB")</pre>
mouse$add <- alleles[mouse$M227]</pre>
fit <- lm(BW~ sex + reps + add, data=mouse)
summary(fit)
##
## Call:
## lm(formula = BW ~ sex + reps + add, data = mouse)
## Residuals:
##
        Min
                  1Q
                       Median
                                     30
                                             Max
## -15.9743 -2.6780 -0.0483
                                 2.5625
                                        19.7455
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 34.3598
                            0.2396 143.399
                                              <2e-16 ***
                 8.4133
## sexMale
                            0.2415 34.838
                                              <2e-16 ***
## reps2
                -0.3787
                            0.2852
                                    -1.328
                                               0.184
## reps3
                 2.8966
                            0.3043
                                      9.518
                                              <2e-16 ***
## add
                 1.7381
                            0.1790
                                      9.713
                                              <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.135 on 1169 degrees of freedom
     (3 observations deleted due to missingness)
## Multiple R-squared: 0.5492, Adjusted R-squared: 0.5477
## F-statistic: 356.1 on 4 and 1169 DF, p-value: < 2.2e-16
```

The summary(fit) command produced

- parameter estimates (or Coefficients) $\widehat{\mu}$ and $\widehat{\beta}$,
- their standard errors (SE) (estimates for square root of the sampling variance of the parameter estimates),
- t-statistic (estimate/SE) and
- P-value under the null hypothesis that the parameter is 0 and errors are uncorrelated and have distribution $N(0, \sigma^2)$.

Under the assumptions of linear model, sampling distribution of t-statistic is t-distribution and hence q% confidence intervals are determined as $\widehat{\beta} \pm a \times SE$, where a is the q/2% quantile of t-distribution with n-2 degrees of freedom. To get a confidence interval use the confint function:

```
confint(fit,parm="add")
## 2.5 % 97.5 %
## add 1.387014 2.089222
```

The regression coefficient for the variable add is 1.74. The coefficient corresponds to the allele substitution effect (α) . Previously we have estimated allele and genotype frequencies for M227. The following table summarizes all genotypic values, all breeding values and the dominance deviations.

Genotyp	Genotypic value	Breeding Value	Dominance Deviation
A_iA_j	GV_{ij}	BV_{ij}	D_{ij}
A_1A_1	a	$2q\alpha$	$-2q^2d$
A_1A_2	d	$(q-p)\alpha$	2pqd
A_2A_2	-a	$-2p\alpha$	$-2p^2d$

Question 5: What are the breeding values for body weight based on the M227 locus?

Answer:

Now we want to compute the genetic variance associated with marker M227. The formula below shows that genetic variance for a single locus model σ_G^2 consists of two components. The first component σ_A^2 is called the **genetic additive variance** and the second component σ_D^2 is termed **dominance variance**. Here σ_A^2 corresponds to the variance of the breeding values. The variance of breeding values is also called the additive genetic variance, because as we have already seen the breeding values are additive in the number of favorable alleles. In populations where there is no additive genetic variance, individuals all have the same breeding value. Therefore, they will produce offspring with the same expected advantage (zero), and selection cannot generate any improvement over generations. Because σ_D^2 corresponds to the variance of the dominance deviation effects it is called dominance variance.

$$\sigma_G^2 = 2pq\alpha^2 + (2pqd)^2$$
$$= \sigma_A^2 + \sigma_D^2$$

Question 6: What is the additive genetic variance associated with M227 for body weight?

Answer:

Question 7: Should you have considered other factors in the linear model specified above?

Answer:

Now we will fit the full genetic model to locus M227 including both additive and dominance effects. The additive effect is modeled as previously shown by a variable add that is coded as -1, 0, and 1 (corresponding to -a, 0, a) for the genotypes AA, AB, and BB. The dominance effect is modeled by a variable dom that is coded as 0, 1, and 0 (corresponding to 0,d,0) for the genotypes AA, AB, and BB. The corresponding R code is shown below:

```
alleles \leftarrow c(-1,0,1)
names(alleles) <- c("AA", "AB", "BB")</pre>
mouse$add <- alleles[mouse$M227]</pre>
mouse$dom <- as.numeric(mouse$add==1)</pre>
fit <- lm(BW~sex + reps + add+dom, data=mouse)</pre>
summary(fit)
##
## Call:
## lm(formula = BW ~ sex + reps + add + dom, data = mouse)
##
## Residuals:
##
        Min
                  1Q
                      Median
                                     3Q
                                             Max
## -16.1773 -2.7642 -0.0437
                                 2.5549
                                         20.1121
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 34.5549
                             0.2666 129.635 < 2e-16 ***
## sexMale
                 8.4130
                             0.2413 34.863 < 2e-16 ***
## reps2
                -0.3706
                             0.2850 -1.300
                                              0.1937
## reps3
                 2.9062
                             0.3041
                                      9.555 < 2e-16 ***
                             0.2580
                                      7.937 4.82e-15 ***
## add
                 2.0479
## dom
                -0.8811
                             0.5290 - 1.665
                                              0.0961 .
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 4.132 on 1168 degrees of freedom
     (3 observations deleted due to missingness)
## Multiple R-squared: 0.5503, Adjusted R-squared: 0.5484
## F-statistic: 285.8 on 5 and 1168 DF, p-value: < 2.2e-16
confint(fit,parm="add")
          2.5 %
##
                  97.5 %
## add 1.541656 2.554078
confint(fit,parm="dom")
##
           2.5 %
                    97.5 %
## dom -1.919038 0.1569128
```

The results from the linear model analysis suggest that only the additive genetic effect, add, is significantly different from 0. However in the following exercise we will be using the both the additive effect (add) and dominance effect (dom) estimated for locus M227, and the frequency of the positive allele (p) to explore the effect of changes in allele frequency.

Use the following shinyapp, https://shiny.cnsgenomics.com/Falconer2/, to understand the relationship between allelic substitution effect (α) and additive gene action (a), dominance gene action (d), and allele frequency (p).

Question 8: Use the estimated gene actions (Question 7) and the estimated allele frequency (Question 2) to obtain the predicted allelic substitution effect? Use rounded values if necessary.
Answer:
Question 9: Does the value of α match the estimate of the (marginal) additive effect from Question 5?
Answer:
Question 10: How does α depend on a larger dominance gene action d (e.g., maximum value, 10)?
Answer:
Question 11: How does α depend on a different allele frequency p (e.g., 0.95)?
Answer:
Question 12: Under that new value of p , how does α depend on d (e.g., from the initial value of d to the maximum value)?
Answer: