

## Examination in School of Mathematical Sciences Semester 2, 2020

## 104843 STATS 2107 Statistical Modelling & Inference II

Total Duration: 190 mins

#### NUMBER OF QUESTIONS: 6 TOTAL MARKS: 100

#### Instructions

- Computations must be uploaded to MyUni as high quality images take a photo and email the scan to your computer and upload the file into the MyUni examination (use of CamScanner and Canvas Student on your mobile device is recommended)
- Show all calculations and assumptions
- If you believe that a parameter or an important piece of information has been inadvertently omitted by the examiner, assume a suitable value, clearly stating it, and continue with the solution.
- It is your responsibility to ensure plenty of time for file scanning and uploading at end. You should upload early and then if you have an updated version of your exam with remaining time, upload again. All uploads are retained, your latest entry will be marked.
- You must review your submissions on MyUni before the end of the exam to check that they
  have uploaded, are legible, in the correct order, and correctly oriented. If they are not, then
  you must resubmit.

#### **Materials**

- This is an open-book, open-internet examination. You are permitted to access your notes and any material on MyUni. You are also permitted to access websites like Wolfram Alpha, Symbolab, Desmos, etc
- No discussion or consultation about the exam with any third party is permitted during the exam period.
- You may write your answers on paper and submit a scanned PDF. If you have a tablet and stylus, you may write your answers electronically and submit an exported PDF. No extra time allowance will be given for typesetting answers. If using paper:
  - Make sure you have plenty of blank A4 paper to write your answers on
  - Use a dark pen to write (not a pencil)

#### YOU MAY COMMENCE WRITING ONCE YOU ENTER THE EXAMINATION

1. Suppose  $Y_1, Y_2, \ldots, Y_n$  are independent and identically distributed (i.i.d.) random variables with probability density function (PDF)

$$f_{Y_i}(y) = \frac{1}{\theta^2} y e^{-\frac{y}{\theta}}, \quad \text{for } y > 0,$$

where  $\theta > 0$  is an unknown parameter. Let  $\hat{\theta}_1 = Y_n$  and  $\hat{\theta}_2 = \frac{1}{2}\bar{Y}$  be estimators of  $\theta$ .

(a) Find the bias of  $\hat{\theta}_1$  and  $\hat{\theta}_2$ .

*Hint:* Identify the distribution of  $Y_i$ .

## Solution: [4 Marks]

Observe that  $Y_i \sim \text{Gamma}(2, \theta)$ . Hence  $E(Y_i) = 2\theta$  and  $\text{var}(Y_i) = 2\theta^2$ . [2 marks] It follows that

$$b_{\hat{\theta}_1}(\theta) = E(\hat{\theta}_1) - \theta = E(Y_n) - \theta = 2\theta - \theta = \theta,$$

$$b_{\hat{\theta}_2}(\theta) = E(\hat{\theta}_2) - \theta = E\left(\frac{1}{2}\bar{Y}\right) - \theta = \frac{1}{2}E(\bar{Y}) - \theta = \frac{1}{2}(2\theta) - \theta = \theta - \theta = 0.$$

[2 marks]

(b) Find the mean square error (MSE) of  $\hat{\theta}_1$  and  $\hat{\theta}_2$ .

## Solution: [4 Marks]

$$MSE_{\hat{\theta}_1}(\theta) = \text{var}(\hat{\theta}_1) + b_{\hat{\theta}_1}(\theta)^2 = 2\theta^2 + \theta^2 = 3\theta^2$$

[2 marks]

$$MSE_{\hat{\theta}_2}(\theta) = \operatorname{var}(\hat{\theta}_2) - b_{\hat{\theta}_2}(\theta)^2 = \operatorname{var}\left(\frac{1}{2}\bar{Y}\right) = \frac{1}{4}\left(\frac{2\theta^2}{n}\right) = \frac{\theta^2}{2n}$$

[2 marks]

(c) Which estimator,  $\hat{\theta}_1$  or  $\hat{\theta}_2$ , is preferred by MSE? Justify your answer.

### Solution: [2 Marks]

$$MSE_{\hat{\theta}_2}(\theta) = \left(\frac{1}{2n}\right)\theta^2 < 3\theta^2 = MSE_{\hat{\theta}_1}(\theta).$$

[1 mark]

So  $\hat{\theta}_2$  is preferred as it has a lower MSE.

[10 marks]

- 2. Car crash tests are performed to assess the safety ratings of vehicles. Manufacturers of new cars are to send a few samples to an authorised agent for testing. Let p be the proportion of a certain model of car failing the crash test. Porsche had designed a new model and decided that in order to proceed with production, it needs p to be less than 0.05 for this car model. To test the hypothesis p = 0.05 versus  $H_a: p > 0.05$ , the following procedure was used.
  - i) Two cars of the new model were randomly selected from the production line.
  - ii) If both of them failed the crash test, then the null hypothesis is rejected.
  - iii) If only one of them failed the test, then a third car is randomly selected. If this third car failed the crash test, then the null hypothesis rejected.
  - iv) In all other cases, we do not reject the null hypothesis.
  - (a) Give the definition of Type I and Type II errors. Explain what is meant by committing a Type I error and committing a Type II error in the context of this problem.

### Solution: [6 Marks]

A Type I error is made if the null hypothesis is rejected when the null hypothesis is in fact true. In this case, a Type I error would occur if the actual proportion of this car model that fails the crash test is 0.05 (or less), but the two randomly selected cars failed the crash test or only one of them failed the crash test but the third car also failed the crash test, leading to the rejection of the null hypothesis when it is in fact true.

A Type II error is made if the null hypothesis is not rejected when in fact the null hypothesis is false. In this case, a Type II error would occur if the actual proportion of this car model that fails the crash test is greater than 0.05, but none of the two randomly selected cars failed the crash test or only one of them failed the crash test but the third car did not fail the crash test, leading to the failure to reject the null hypothesis when it is in fact false.

[3 marks]

(b) Find the probability of Type I error.

*Hint:* What is the distribution of the number of cars in the sample failing the crash test?

**Solution:** [5 Marks]

Let Y denote the number of cars in the sample (of 3) that fails the crash test.

Then  $Y \sim B(3, 0.05)$  and the rejection rule of our test is  $RR = \{Y \geq 2\}$ . [2 marks]

$$\alpha = P \text{ (Type I error)}$$

$$= P \text{ (reject } H_0 \mid H_0 \text{ true)}$$

$$= P (Y \ge 2 \mid p = 0.05)$$

$$= P (Y = 2) + P(Y = 3)$$

$$= {3 \choose 2} 0.05^2 0.95^1 + {3 \choose 3} 0.05^3$$

$$= 0.007125 + 0.000125$$

$$= 0.00725$$

[3 marks]

(c) Suppose the true proportion of cars that fails the crash test is  $p^*$ . Find the probability of Type II error.

Solution: [5 Marks]

$$\beta = P \text{ (Type II error)}$$

$$= P \text{ (not reject } H_0 \mid p = p^*)$$

$$= P (Y < 2 \mid p = p^*)$$

$$= P (Y = 0) + P(Y = 1) \text{ where } Y \sim B(3, p^*)$$

$$= {3 \choose 0} (p^*)^0 (1 - p^*)^3 + {3 \choose 1} (p^*)^1 (1 - p^*)^2$$

$$= (1 - p^*)^3 + 3(p^*)(1 - p^*)^2$$

$$= 1 - 3p^* + 2(p^*)^3$$

[5 marks]

(d) Find the power function of this test.

Solution: [2 Marks]

power(p) = 
$$1 - \beta = 1 - (1 - 3p^* + 2(p^*)^3) = 3(p^*)^2 - 2(p^*)^3$$
.

[2 marks]

(e) Show that the value of the power function at p = 0.05 is the same as the value found in part (b).

Solution: [2 Marks]

power
$$(0.05) = 3(0.05)^2 - 2(0.05)^3 = 0.00725$$
.

[2 marks]

[20 marks]

3. Suppose Y is a random variable with PDF

$$f_Y(y) = 2\left(\frac{\theta - y}{\theta^2}\right)$$
 for  $0 < y < \theta$ .

where  $\theta$  is an unknown parameter.

(a) Find the distribution of  $X = \frac{Y}{\theta}$  and explain why it is a pivotal quantity for  $\theta$ . *Hint:* Use the change of variable formula.

## Solution: [5 Marks]

The transformation:  $X = \frac{Y}{\theta}$ 

The inverse:  $Y = \theta X$ 

The Jacobian:  $J = \left| \frac{\partial y}{\partial x} \right| = |\theta|$ 

It follows that

$$f_X(x) = F_Y(x)J = 2\left(\frac{\theta - \theta x}{\theta^2}\right)|\theta| = 2(1 - x)$$
 for  $0 < x < 1$ .

[3 marks]

Since X is a function of  $\theta$  and the sample Y,  $\theta$  is the only unknown parameter, and the distribution of X does not depend on  $\theta$ , then X is a pivotal quantity of  $\theta$ . [2 marks]

(b) Find the 90% upper confidence interval for  $\theta$  based on X, where the upper limit U is such that  $P(\theta \le U) = 0.90$ .

# Solution: [4 Marks]

$$0.90 = P(\theta \le U)$$

$$= P\left(\frac{1}{\theta} \ge \frac{1}{U}\right)$$

$$= P\left(\frac{Y}{\theta} \ge \frac{Y}{U}\right)$$

$$= P\left(X \ge \frac{Y}{U}\right)$$

$$= \int_{\frac{y}{U}}^{1} 2(1-x)dx$$

$$= \left[2x - x^{2}\right]_{\frac{y}{U}}^{1}$$

$$= \left[2 - 1 - 2\left(\frac{y}{U}\right) + \left(\frac{y}{U}\right)^{2}\right]$$

$$= 1 - \frac{1}{U}(2y - y^{2})$$

$$0.10 = \frac{1}{U}(2y - y^{2})$$

$$U = 10y(2 - y)$$

Please turn over for page 6

```
Hence, the 90% upper confidence interval for \theta is (0, 10y(2-y)). [1 mark]

(c) Give the 90% upper confidence interval for \theta when y = 0.6.

Solution: [1 Marks]

Observe that U = 10(0.6)(2 - 0.6) = 8.4.

Hence CI = (0, 8.4). [1 mark]
```

4. An analysis was performed to investigate whether ice-cream consumption is related to the price, the temperature of the day, and the income of the consumer. A multiple linear regression model was fitted to the data using R and the following output was obtained. Note that some of the entries in the R output are missing and you will be asked to calculate these in the question.

```
##
## Call:
## lm(formula = Consumption ~ Price + Temperature + Income, data = ic2)
##
## Residuals:
##
        Min
                    10
                          Median
                                        3Q
                                                 Max
## -0.115062 -0.025694 0.003299
                                  0.022995
                                            0.128025
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.3462222 0.1166343
                                       2.968 0.00341 **
## Price
                ???????? 0.3533218
                                     -4.840 2.83e-06 ***
## Temperature 0.0033775 0.0002239
                                      15.083 < 2e-16 ***
                                       ????? 2.43e-10 ***
## Income
                0.0037759 0.0005619
## ---
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
##
## Residual standard error: 0.04151 on 176 degrees of freedom
## Multiple R-squared: 0.652, Adjusted R-squared: 0.646
## F-statistic: 109.9 on 3 and 176 DF, p-value: < 2.2e-16
(a) What was the size of the sample?
   Solution: [1 Marks]
```

From the R output, we have n-p=176 and p=4, so n=176+4=180. [1 mark]

(b) What is the estimate of the coefficient of Price,  $\beta_1$ ?

```
Solution: [2 Marks]
```

```
\hat{\beta}_1 = 0.3533218 \times -4.840 = -1.710078.
```

(c) What is the value of the t-statistic for  $\hat{\beta}_3$ ?

# Solution: [2 Marks]

$$t_{\hat{\beta}_3} = \frac{0.0037759}{0.0005619} = 6.719879.$$

(d) Perform a statistical test to determine if there is a significant (linear) relationship between Consumption and the three predictors (Price, Temperature, and Income) at the 5% significance level. State the null and alternative hypotheses, the value of the test statistic, the P-value, and your conclusion.

### Solution: [6 Marks]

 $H_0: \beta_1 = \beta_2 = \beta_3 = 0$  vs  $H_a:$  at least one of  $\beta_1, \beta_2$ , and  $\beta_3$  is not zero. [1 mark] where  $\beta_1, \beta_2$ , and  $\beta_3$  is the coefficient associated with the variable Price, Temperature, and Income, respectively. [1 mark]

The value of the test statistic is 109.9.

[1 mark]

The P-value is less than  $2.2 \times 10^{-16}$ .

[1 mark]

Hence we have sufficient evidence to reject the null hypothesis at the 5% significance level and hence conclude that there is a significance linear relationship between Consumption and the three predictors (Price, Temperature, and Income). [2 marks]

(e) Calculate the 95% confidence interval for  $\beta_2$ , the coefficient for Temperature, taking into account the other predictor variables. You may use R to calculate the appropriate critical value.

### Solution: [4 Marks]

The critical value is  $t_{176,0.025} = 1.973534$  using the R command qt(0.975,176).

[1 mark]

It follows that the required confidence interval is given by

CI = 
$$\hat{\beta}_2 \pm t_{n-p,\alpha/2} SE(\hat{\beta}_2)$$
  
= 0.0033775 \pm 1.973534 \times 0.0002239  
\pi (0.002935626, 0.003819374).

[3 marks]

[15 marks]

- 5. A study was conducted to investigate the growth characteristics of blue mussel in New South Wales. Samples were taken from two distinct locations. Their age (in weeks) and weight (in grams) were recorded. An analysis was performed using R. The commands and outputs are given in Appendix A.
  - (a) Consider the scatterplot of weight against age given in Figure 1. Describe the relationship.

#### Solution: [4 Marks]

There is strong, positive, linear relationship between weight and age. [3 marks] The lines do not look parallel, with Location 2 having a larger slope and also a higher intercept. [1 mark]

(b) Four observations from the data are:

Weight	Age	Location	
0.44	3	1	
0.50	3	1	
1.76	3	2	
2.38	4	2	

Write down the design matrix for fitting a separate regression model of Weight on Age and Location, for these these four observations only.

Solution: [2 Marks]

$$\boldsymbol{X} = \begin{bmatrix} 1 & 3 & 0 & 0 \\ 1 & 3 & 0 & 0 \\ 1 & 3 & 1 & 3 \\ 1 & 4 & 1 & 4 \end{bmatrix}$$

[2 marks]

(c) Consider the separate regression model. Write down the two lines of best fit for the relationship between weight and age: one for Location 1 and one for Location 2.

Solution: [3 Marks]

For Location 1, we have

Weight = 
$$-0.62660 + 0.37219 \times Age$$

[1 mark]

For Location 2, we have

$$\texttt{Weight} = (-0.62660 + 0.22773) + (0.37219 + 0.29714) \times \texttt{Age} = -0.39887 + 0.66933 \times \texttt{Age} = -0.0007 + 0.0007 +$$

[2 marks]

(d) Based on the separate regression model, estimate the weight of an 8 weeks old mussel at Location 2.

Solution: [2 Marks]

$$\texttt{Weight} = -0.39887 + 0.66933 \times 8 = 4.95577g$$

[2 marks]

(e) Test for a statistically significant interaction term in the separate regression model at the 5% significance level. Remember to include the null and alternative hypotheses, the value of the test statistic, the P-value, and your conclusion.

## Solution: [4 Marks]

 $H_0: \beta_3 = 0 \text{ vs } H_1: \beta_3 \neq 0,$ 

[1 mark]

where  $\beta_3$  is the coefficient associated with the interaction term.

The value of the test statistic is 12.592.

[1 mark]

The P-value is less than  $2 \times 10^{-16}$ .

[1 mark]

We have sufficient evidence to reject the null hypothesis at the 5% significance level and hence conclude that there is evidence that the interaction term is necessary for the model. [1 mark]

(f) Using the Bayesian Information Criterion which model fits the data best? Justify your answer.

## Solution: [1 Marks]

The best model appears to be the separate regression model as this has the smallest BIC with a value of 533.9220.

(g) Assess the assumptions of the linear model used in the parallel model. The plots given in Figure 2 may be used where appropriate.

## Solution: [4 Marks]

**Linearity**: Residual versus fitted (top left) shows very reasonable random scatter so reasonable. There are possible outliers (numbers 96, 142, 176). [1 mark]

**Homoscedascity**: Standardised residual versus fitted (bottom left) shows somewhat equal spread as we move from left to right so reasonable. [1 mark]

**Normality**: Residual QQ-plot (top right) is roughly linear so reasonable except for the points at the far left and far right. [1 mark]

**Independence**: The weight of one mussel should not affect the weight of another mussel so this is reasonable. [1 mark]

[20 marks]

6. Suppose  $Y_1, Y_2, \ldots, Y_n$  are i.i.d. random variables with PDF

$$f_{Y_i}(y;\theta) = \frac{2y}{\theta}e^{-\frac{1}{\theta}y^2}, \quad y > 0.$$

where  $\theta > 0$  is an unknown parameter. The mean and variance of  $Y_i$  is given by

$$E[Y_i] = \sqrt{\frac{\pi\theta}{4}}$$
 and  $var(Y_i) = \frac{\theta(4-\pi)}{4}$ ,

respectively, for  $i = 1, 2, \dots, n$ .

(a) Write down the likelihood function.

Solution: [2 Marks]

$$L(\theta; \boldsymbol{y}) = \prod_{i=1}^{n} f_{Y_i}(y_i; \theta) = \left(\frac{2}{\theta}\right)^n \left(\prod_{i=1}^{n} y_i\right) e^{-\frac{1}{\theta} \sum_{i=1}^{n} y_i^2}$$

(b) Write down the log-likelihood function.

Solution: [2 Marks]

$$\ell(\theta; \boldsymbol{y}) = \log L(\theta; \boldsymbol{y}) = n \log \left(\frac{2}{\theta}\right) + \sum_{i=1}^{n} \log(y_i) - \frac{1}{\theta} \sum_{i=1}^{n} y_i^2$$

(c) Write down the score function.

Solution: [2 Marks]

$$S(\theta; \boldsymbol{y}) = \frac{\partial \ell(\theta; \boldsymbol{y})}{\partial \theta} = -\frac{n}{\theta} + \frac{1}{\theta^2} \sum_{i=1}^{n} y_i^2$$

(d) Find the maximum likelihood estimate  $\hat{\theta}$  of  $\theta$ .

Solution: [3 Marks]

Solve for  $\theta$  in  $0 = S(\theta; \boldsymbol{y})$ :

$$0 = -\frac{n}{\theta} + \frac{1}{\theta^2} \sum_{i=1}^{n} y_i^2$$
$$\frac{n}{\theta} = \frac{1}{\theta^2} \sum_{i=1}^{n} y_i^2$$

$$\hat{\theta} = \frac{1}{n} \sum_{i=1}^{n} y_i^2.$$

(e) Find the Fisher information. You may assume that the usual regularity conditions hold.

Solution: [3 Marks]

Observe that 
$$E[Y_i^2] = var(Y_i) + E[Y_i]^2 = \frac{\theta(4-\pi)}{4} - \frac{\pi\theta}{4} = \theta$$
.

$$I(\theta) = E\left[-\frac{\partial^2 \ell}{\partial \theta^2}\right]$$

$$= E\left[-\frac{n}{\theta^2} + \frac{2}{\theta^3} \sum_{i=1}^n Y_i^2\right]$$

$$= -\frac{n}{\theta^2} + \frac{2}{\theta^3} \sum_{i=1}^n E\left[Y_i^2\right]$$

$$= -\frac{n}{\theta^2} + \frac{2}{\theta^3}(n\theta)$$

$$= \frac{n}{\theta^2}$$

(f) State the asymptotic distribution of  $\hat{\theta}$  and hence give a  $100(1-\alpha)\%$  confidence interval for  $\theta$ .

## Solution: [5 Marks]

By the properties of MLE,  $\hat{\theta} \sim N\left(\theta, \frac{\theta^2}{n}\right)$  asymptotically. [3 marks] It follows that the  $100(1-\alpha)\%$  confidence interval for  $\theta$  is

$$CI = \hat{\theta} \pm z_{\frac{\alpha}{2}} \sqrt{I(\hat{\theta})^{-1}} = \hat{\theta} \pm z_{\frac{\alpha}{2}} \left(\frac{\hat{\theta}}{\sqrt{n}}\right).$$

[2 marks]

- (g) Suppose we want to test  $H_0: \theta = \theta_0$  versus  $H_a: \theta \neq \theta_0$ .
  - (i) State the score test statistic and its asymptotic distribution.
  - (ii) State the test statistic and the critical region of the likelihood ratio test.

#### Solution: [6 Marks]

#### The score test:

The score test statistic is

$$U = \frac{S(\theta; \boldsymbol{y})}{\sqrt{I(\theta_0)}} = \frac{-\frac{n}{\theta_0} + \frac{1}{\theta_0^2} \sum_{i=1}^n y_i^2}{\sqrt{\frac{n}{\theta_0^2}}} = -\sqrt{\frac{n}{\theta_0}} + \frac{1}{\theta_0 \sqrt{n}} \sum_{i=1}^n y_i^2.$$

[2 marks]

Asymptotically,  $U \sim N(0, 1)$ .

[1 mark]

#### The likelihood ratio test:

The likelihood ratio test statistic is

$$G^{2} = -2 \left[ \ell(\theta_{0}; \boldsymbol{y}) - \ell(\hat{\theta}; \boldsymbol{y}) \right]$$

$$= -2 \left[ n \log \left( \frac{2}{\theta_{0}} \right) + \sum_{i=1}^{n} \log(y_{i}) - \frac{1}{\theta_{0}} \sum_{i=1}^{n} y_{i}^{2} - n \log \left( \frac{2}{\hat{\theta}} \right) - \sum_{i=1}^{n} \log(y_{i}) + \frac{1}{\hat{\theta}} \sum_{i=1}^{n} y_{i}^{2} \right]$$

$$= -2 \left[ n \log \left( \frac{\hat{\theta}}{\theta_{0}} \right) - \left( \frac{1}{\theta_{0}} - \frac{1}{\hat{\theta}} \right) \sum_{i=1}^{n} y_{i}^{2} \right]$$

[2 marks]

We reject  $H_0$  if  $G^2 \ge \chi^2_{1,\alpha}$ .

[1 mark]

(h) An alternative form of the distribution is

$$f_{Y_i}(y; \sigma^2) = \frac{y}{\sigma^2} e^{-\frac{y^2}{2\sigma^2}}, \quad y > 0.$$

Give an expression for the maximum likelihood estimate of  $\sigma^2$ .

Solution: [2 Marks]

Observe that  $\sigma^2 = \frac{\theta}{2}$ .

By the invariance property of MLE, we have

$$\hat{\sigma}^2 = \frac{\hat{\theta}}{2} = \frac{1}{2n} \sum_{i=1}^n y_i^2.$$

[25 marks]

[Exam total: 100]

#### Appendix A

Load the data

```
library(tidyverse)
mussel <- read.csv("mussel.csv", header=T)
mussel$Location <- factor(mussel$Location)</pre>
```

#### Visualise data

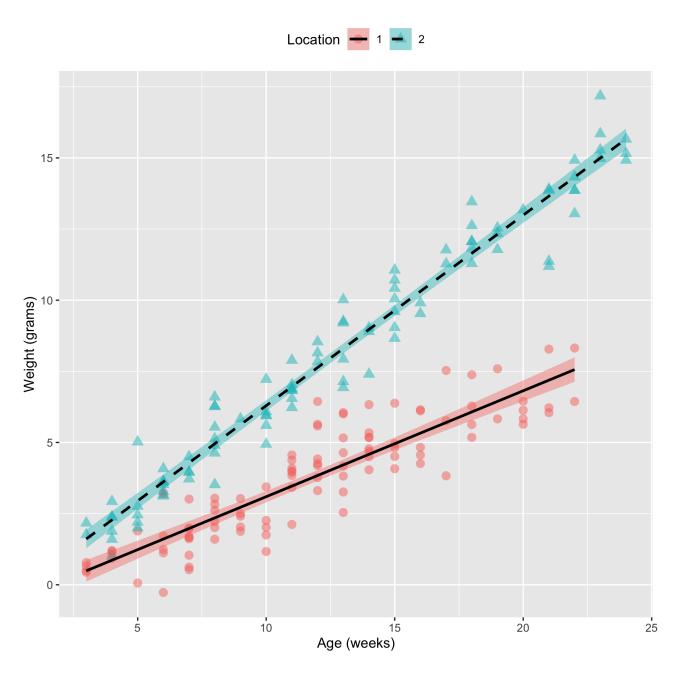


Figure 1: Scatterplot of Weight against Age for the mussel dataset. Colour and shape of points indicates Location.

#### Fit models

```
## Identical regression model
identical.model <- lm(Weight ~ Age, data = mussel)</pre>
summary(identical.model)
##
## Call:
## lm(formula = Weight ~ Age, data = mussel)
## Residuals:
      Min
              1Q Median
                            30
                                  Max
## -5.0800 -1.7364 0.1562 1.8067 5.0823
##
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## Age
            0.57775
                      0.02914 19.830 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.269 on 193 degrees of freedom
## Multiple R-squared: 0.6708, Adjusted R-squared: 0.6691
## F-statistic: 393.2 on 1 and 193 DF, p-value: < 2.2e-16
## Parallel regression model ----
parallel.model <- lm(Weight ~ Age + Location, data = mussel)</pre>
summary(parallel.model)
##
## Call:
## lm(formula = Weight ~ Age + Location, data = mussel)
##
## Residuals:
      Min
              1Q Median
                           3Q
                                  Max
## -2.9364 -0.8978 0.0312 0.8541 3.4195
##
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
##
0.54959
                       0.01562 35.19 <2e-16 ***
## Age
## Location2 3.83452 0.17426 22.00 <2e-16 ***
## ---
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.212 on 192 degrees of freedom
## Multiple R-squared: 0.9065, Adjusted R-squared: 0.9055
## F-statistic:
                 931 on 2 and 192 DF, p-value: < 2.2e-16
## Separate regression model ----
separate.model <- lm(Weight ~ Age * Location, data = mussel)</pre>
summary(separate.model)
##
## Call:
## lm(formula = Weight ~ Age * Location, data = mussel)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                  3Q
                                          Max
## -2.46706 -0.51737 -0.00505 0.52470 2.60030
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -0.62660 0.23265 -2.693 0.0077 **
                0.37219
                          0.01823 20.413 <2e-16 ***
## Age
## Location2 0.22773 0.31421 0.725 0.4695
## Age:Location2 0.29714 0.02360 12.592 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8984 on 191 degrees of freedom
## Multiple R-squared: 0.9489, Adjusted R-squared: 0.9481
## F-statistic: 1183 on 3 and 191 DF, p-value: < 2.2e-16
Model Selection
BIC(identical.model, parallel.model, separate.model)
##
                  df
                          BIC
## identical.model 3 886.7381
## parallel.model 4 646.5005
```

## separate.model 5 533.9220

Assumption checking

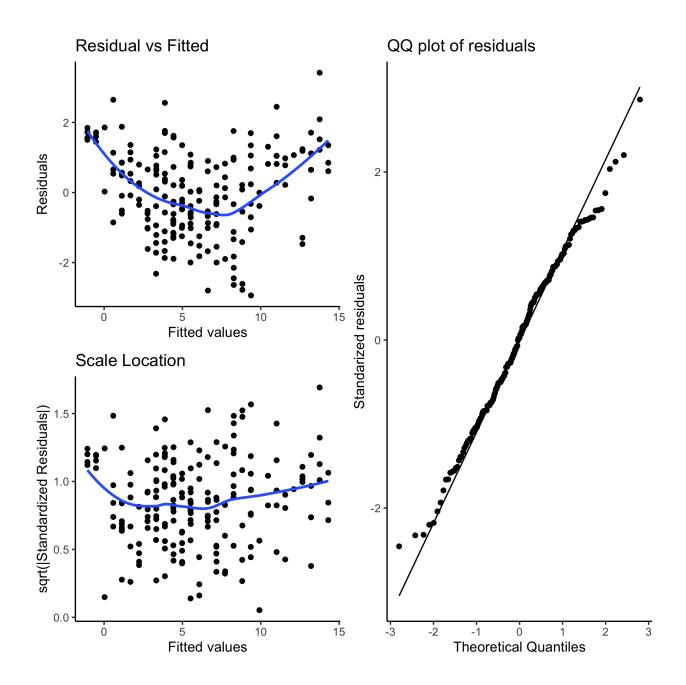


Figure 2: Assumption plots of the separate model for the mussel dataset.

# Appendix B

Distribution	Probability mass function / probability density function	Expectation	Variance
Binomial	$p(x) = \binom{n}{x} p^x (1-p)^{n-x}$ for $x = 0, 1, 2, \dots, n$	np	np(1-p)
Geometric	$p(x) = p(1-p)^{x-1}$ for $x = 1, 2,$	$\frac{1}{p}$	$\frac{1-p}{p^2}$
Poisson	$p(x) = rac{e^{-\lambda}\lambda^x}{x!}$ for $x = 0, 1, 2, \dots$	$\lambda$	$\lambda$
Uniform	$f(x) = \frac{1}{b-a} \text{ for } a < x < b$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$
Exponential	$f(x) = \lambda e^{-\lambda x} \text{ for } x > 0$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$
Gamma	$f(x) = \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha - 1} e^{-\lambda x}$ for $x > 0$	$\frac{\alpha}{\lambda}$	$\frac{\alpha}{\lambda^2}$
Normal	$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}}e^{-(1/2\sigma^2)(x-\mu)^2}$ for $-\infty < x < \infty$	$\mu$	$\sigma^2$
Beta	$f(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$ for $0 < x < 1$	$\frac{\alpha}{\alpha + \beta}$	$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$