Practical 2: Review of Linear Model

One-way analysis of variance using the normal linear model.

A cancer researcher obtained measurements of the metabolic rate of glucose utilization for low, intermediate and high grade tumors of a particular type.

Low	54.7	50.5	45.9	59.7	51.5	45.5	47.4	55.6	59.4	56.9	
Intermediate	48.6	55.8	62.6	64.7	54.3	61.3	44.6	59.8	46.0	56.5	42.4
	71.4	62.6	67.9	55.2	47.3	75.9					
High	63.8	65.4	68.1	70.0	63.9	63.6	63.6	59.5	56.3	59.1	60.6

1. Enter these data into a single vector (y):

Combine the individual vectors (low, inter and high) into a single vector:

```
y = c(low, inter, high)
```

2. Create a corresponding vector for grade with values 1, 2 & 3 corresponding to low intermediate and high:

```
grade = c(low*0+1, inter*0+2, high*0+3)
```

Can you explain how the above R-code creates the appropriate coding scheme?

3. Plot metabolic rate by grade. Interpret this plot.

```
plot(grade,y,ylab="Metabolic
Rate",xlab="Grade")
```

MR is higher for High Grade tumors than for both the Low and Intermediate Grades.

MR is more variable for the Intermediate Grade.

4. Prepare a box-plot of metabolic rate by grade. Interpret this plot.

```
boxplot(y~grade)
```

MR is higher for High Grade tumors than both the Low and Intermediate Grades.

MR is more variable for the Intermediate Grade.

MR distribution is symmetric for Low and Intermediate, but skewed for High.

There are no outliers.

5. Use the aov function in R to obtain a one-way ANOVA of the metabolic rate with grade being the factor. Interpret the summary output.

```
mr.aov = aov (y ~ factor (grade))
```

```
summary(mr.aov)
```

There is a statistically significant difference between Grade (p = 0.0105).

6. What is the statistical model for this ANOVA? What are the assumptions that must be satisfied/checked?

$$y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \epsilon_i \qquad \epsilon_i \sim NID(0, \sigma^2)$$
where
$$x_1 = 1 \text{ if Grade} = 2 & 0 \text{ otherwise}$$

$$x_2 = 1 \text{ if Grade} = 3 & 0 \text{ otherwise}$$

$$i = 1 \text{ to } 38$$

Errors are Normally distributed (with mean 0). Errors are Independent. Errors have constant variance.

7. Plot the model residuals by grade. Interpret this plot.

```
plot(grade, mr.aov$residuals)
abline(h = 0)
```

The residuals do not have constant variance. There is greater variability in the intermediate group.

8. Obtain a normal probability plot and interpret.

```
qqnorm(mr.aov$residuals)
qqline(resid(mr.aov))
```

Normal Probability Plot: Good adherence to the reference line.

Minor departures.

Conclusion: Normal distribution.

9. Obtain a histogram for the residuals and interpret.

```
hist(resid(mr.aov))
```

Histogram: Almost perfectly symmetric. Close to Normal/bell-shape.

Conclusion: Normal distribution.

10. We will now repeat the ANOVA using an explicit linear model and the lm function in R. Create variables x2 and x3 which are indicators of Intermediate and High grade cases.

```
x2 = ifelse(grade==2, 1, 0)
x3 = ifelse(grade==3, 1, 0)
```

Combine the y, grade, x2 and x3 vectors into a matrix:

```
cbind(y, grade, x2, x3)
```

What type of variables are x2 and x3?

11. Fit the linear model $Y_i = \beta_0 + \beta_1 X_{i2} + \beta_2 X_{i3} + e_i$, $e_i \sim NID(0, \sigma^2)$:

```
mr.lm = lm(y \sim x2 + x3)
summary(mr.lm)
```

12. What do the parameters β_0 , β_1 and β_2 represent?

```
\beta_0: mean MR for Low Grade (code 1).

\beta_1: the difference in mean MR between Intermediate and Low.

\beta_2: the difference in mean MR between High and Low.
```

13. What inferences can be drawn about β_1 and β_2 and hence the 3 grades of tumor?

```
Accept H_0: \beta_1 = 0
Intermediate Grade is not significantly different to Low Grade (p = 0.11530).
```

Reject H₀: $\beta_2 = 0$: High Grade is significantly higher, by 10.372, than Low Grade (p = 0.00282).

14. Fit the linear model $Y_i = \beta_0 + e_i$, $e_i \sim NID(0, \sigma^2)$. What does this model imply?

```
mr2.lm <- lm(y ~ 1)
summary(mr2.lm)</pre>
```

The model implies that there is no difference between Grades – a common mean.

The Null model.