# MH3510 Assignment 1

# Question

The following data indicate the relationship between the amount of  $\beta$ -erythroidine in an aqueous solution and the colorimeter reading of the turbidity:

Concentration	Colorimeter Reading
40	69
50	175
60	272
70	335
80	490
90	415
40	72
60	265
80	492
50	180

- (a) Fit a simple regression to the data.
- (b) Comments on the model adequacy using  $\mathcal F$  test and  $\mathcal R^2$  statistics.

## Solution

#### **Data Preparation**

We saved the data in data/data.csv. Then we named the columns as X and y.

```
## 2 50 175
## 3 60 272
## 4 70 335
## 5 80 490
## 6 90 415
```

### Simple Linear Regression

We first fit the simple linear regression model using the  ${\tt lm}$  function.

```
model <- lm(y ~ X, data = data)
summary(model)</pre>
```

```
##
## Call:
## lm(formula = y ~ X, data = data)
##
## Residuals:
                                    3Q
##
       Min
                  1Q
                       Median
                                            Max
  -100.312 -15.080
                        3.203
                                10.880
                                         61.978
##
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
  (Intercept) -252.2971
                            58.7508
                                    -4.294 0.00264 **
                            0.9153
                                     9.318 1.43e-05 ***
## X
                  8.5290
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 48.09 on 8 degrees of freedom
## Multiple R-squared: 0.9156, Adjusted R-squared: 0.9051
## F-statistic: 86.83 on 1 and 8 DF, p-value: 1.434e-05
```

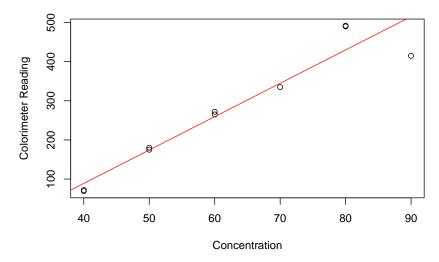
We can see that the estimated regression line is

$$\hat{y} = -252.2971 + 8.5290X$$

with  $\mathcal{F} \approx 86.829$ ,  $p \approx 1.434 \times 10^{-5}$  and  $\mathcal{R}^2 \approx 0.9156$ .

We can also plot the data and the regression line.

#### Simple Linear Regression



We can also calculate the regression coefficients manually.

We can calculate  $\overline{X}$ ,  $\overline{y}$  first.

```
X_bar <- mean(data$X)
y_bar <- mean(data$y)
cat(paste("X_bar =", X_bar, "\ny_bar =", y_bar))</pre>
```

## X\_bar = 62 ## y\_bar = 276.5

Then we calculate  $S_{xx}$ ,  $S_{xy}$  and  $S_{yy}$  by:

$$S_{xx} = \sum_{i=1}^{n} (x_i - \overline{x})^2$$

$$S_{xy} = \sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})$$

$$S_{yy} = \sum_{i=1}^{n} (y_i - \overline{y})^2$$

```
S_xx <- sum((data$X - X_bar)^2)
S_xy <- sum((data$X - X_bar) * (data$y - y_bar))
S_yy <- sum((data$y - y_bar)^2)
cat(paste("S_xx =", S_xx, "\nS_xy =", S_xy, "\nS_yy =", S_yy))</pre>
```

## S\_xx = 2760 ## S\_xy = 23540 ## S\_yy = 219270.5

Then we can calculate the  $\hat{\beta}_1$  and  $\hat{\beta}_0$  by:

$$\begin{cases} \hat{\beta}_1 = \frac{S_{xy}}{S_{xx}} \\ \hat{\beta}_0 = \overline{y} - \hat{\beta}_1 \overline{x} \end{cases}$$

```
beta_1 <- S_xy / S_xx
beta_0 <- y_bar - beta_1 * X_bar
cat(paste("beta_1 =", beta_1, "\nbeta_0 =", beta_0))</pre>
```

## beta\_1 = 8.52898550724638 ## beta\_0 = -252.297101449275

So we can write

$$\hat{y} = -252.2971 + 8.5290X$$

Which is the same as the result from 1m function.

#### **ANOVA** Table

It is easy to see the  $\mathcal{F}$  statistic and  $\mathcal{R}^2$  statistic from the summary function. We can also calculate them manually.

We can calculate SSR and SSE by:

$$SSR = \sum_{i=1}^{n} (\hat{y}_i - \overline{y})^2$$

$$SSE = \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$

```
y_hat <- beta_0 + beta_1 * data$X
SSR <- sum((y_hat - y_bar)^2)
SSE <- sum((data$y - y_hat)^2)
cat(paste("SSR =", SSR, "\nSSE =", SSE))</pre>
```

## SSR = 200772.31884058 ## SSE = 18498.1811594203

 $\operatorname{And}$ 

$$MS_{Reg} = SSR$$

$$s^2 = \frac{1}{n-2}SSE$$

```
MS_Reg <- SSR
s2 <- SSE / 8
cat(paste("MS_Reg =", MS_Reg, "\ns^2 =", s2))</pre>
```

## MS\_Reg = 200772.31884058 ## s^2 = 2312.27264492754

After that, we can obtain the  $\mathcal{F}$  statistic by

$$\mathcal{F} = \frac{\mathrm{MS}_{\mathrm{Reg}}}{s^2}$$

```
F <- MS_Reg / s2
cat(paste("F =", F))
```

## F = 86.8289988557434

So we can calculate the *p*-value for  $\mathcal{H}_0$  is  $\beta_1 = 0$  is

$$p = \mathbb{P}(X, y \mid \beta_1 = 0) = \int_{\mathcal{F}}^{+\infty} F_{1,8}(f) df \approx 1.434 \times 10^{-5}$$

```
p <- 1 - pf(86.829, 1, 8)
cat(paste("p =", p))</pre>
```

## p = 1.43437158113269e-05

The p value is quite small, so we can reject the null hypothesis and say that there is a significant relationship between X (concentration) and y (colorimeter reading).

For  $\mathbb{R}^2$  statistic, we can calculate it by

$$\mathcal{R}^2 = \frac{S_{xy}^2}{S_{xx}S_{yy}}$$

```
R2 <- S_xy^2 / (S_xx * S_yy)
cat(paste("R2 =", R2))
```

## R2 = 0.915637620384775

So the  $\mathbb{R}^2$  statistic is about 0.9156. Which means that there is a strong linear relationship between X and y.

#### Conclusion

For question (a), we have fitted a simple linear regression model to the data by:

$$\hat{y} = -252.2971 + 8.5290X$$

For question (b), we have tested the model adequacy using  $\mathcal{F}$  test and  $\mathcal{R}^2$  statistics. The  $\mathcal{F}$  statistic is 86.829 and the p value is  $1.434 \times 10^{-5}$ , which means that there is a significant relationship between X and y. The  $\mathcal{R}^2$  statistic is 0.9156, which means that there is a strong linear relationship between X and y.