Applied Statistical Methods – Solution 2

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Problem 1: Multiple Linear Regression Model

During the lecture the regression model was explained using the dataset on body weight and breast circumference. To improve the regression model for body weight, two additional conformation traits BCS and HEI are also considered. The new dataset is given in Table 1.

Table 1: Dataset for Multiple Linear Regression of Body Weight on Breast Circumference and two Conformation Traits for ten Animals

Animal	Breast Circumference	Body Weight	BCS	HEI
1	176	471	5.0	161
2	177	463	4.2	121
3	178	481	4.9	157
4	179	470	3.0	165
5	179	496	6.8	136
6	180	491	4.9	123
7	181	518	4.4	163
8	182	511	4.4	149
9	183	510	3.5	143
10	184	541	4.7	130

The same dataset is also available from the website at https://charlotte-ngs.github.io/gelasmss2021/data/asm_w03_ex02_bw_mult_reg.csv.

Your Task

- Setup the linear regression model with an intercept for the data given in Table 1
- Compute the solution for the unknown parameter b
- Verify the result with the output from the function lm() in R

Solution

The linear regression model is given by the following equation

$$y = X * b + \epsilon$$

where y is a vector of body weights, X is a matrix with two columns. The first column of X is all ones and the second column contains the breast circumference values, b is the vector with the intercept and the

unknown regression coefficient and ϵ is the vector of unknown random residuals. The least squares estimate \hat{b} can be computed as

$$\hat{b} = (X^T X)^{-1} X^T y$$

The matrix X and the vector y are extracted from the dataframe and have the following form

$$X = \begin{bmatrix} 1 & 176 & 5 & 161 \\ 1 & 177 & 4.2 & 121 \\ 1 & 178 & 4.9 & 157 \\ 1 & 179 & 3 & 165 \\ 1 & 179 & 6.8 & 136 \\ 1 & 180 & 4.9 & 123 \\ 1 & 181 & 4.4 & 163 \\ 1 & 182 & 4.4 & 149 \\ 1 & 183 & 3.5 & 143 \\ 1 & 184 & 4.7 & 130 \end{bmatrix}, \quad y = \begin{bmatrix} 471 \\ 463 \\ 481 \\ 470 \\ 496 \\ 491 \\ 518 \\ 511 \\ 510 \\ 541 \end{bmatrix}$$

The result for \hat{b} is then

```
xtx <- crossprod(mat_x)</pre>
n_hat_b <- solve(xtx,crossprod(mat_x, vec_y))</pre>
n_hat_b
##
                   [,1]
## [1,] -1313.0788097
## [2,]
             9.6492685
## [3,]
             8.6331873
             0.2267639
## [4,]
We can verify this result using the lm() function of R
lm_bwbc <- lm(`Body Weight` ~ `Breast Circumference` + BCS + HEI, data = tbl_reg)</pre>
summary(lm_bwbc)
##
```

```
## Call:
## lm(formula = `Body Weight` ~ `Breast Circumference` + BCS + HEI,
##
       data = tbl_reg)
##
## Residuals:
              1Q Median
                            3Q
## -7.686 -5.001 -2.190 5.715
                                9.613
##
## Coefficients:
##
                            Estimate Std. Error t value Pr(>|t|)
                                                 -6.273 0.000763 ***
## (Intercept)
                          -1313.0788
                                        209.3310
## `Breast Circumference`
                              9.6493
                                          1.0958
                                                   8.805 0.000119 ***
## BCS
                              8.6332
                                          2.8939
                                                   2.983 0.024533 *
```

Problem 2: Prediction

Given the measurement of the trait Breast Circumference for two additional animals. The measurements are shown in the following table

Table 2: Breast Circumference Measurements For Two Animals Used To Predict Body Weight

Animal	Breast Circumference
Animal 11	181.2
Calf 12	99.5

We want to use the results of the simple linear regression of body weight (BW) on breast circumference (BC), as shown in Problem 2 of Exercise 1 to computed the predicted values for Body Weight for the two animals. The observed value for Breast Circumference of "Calf 12" is outside of the range of the values used in Problem 1. Predicting values of response variables based on predictors that are outside of the range of values used for the parameter estimation is called **extrapolation**. Based on the result of the predicted value of the trait Body Weight for "Calf 12" what can be said about the process of extrapolation?

Your Tasks

- Compute the predicted value of Body Weight for "Animal 11"
- Compute the predicted value of Body Weight for "Calf 12"
- Make a statement about the validity of the extrapolated value of Body Weight for "Calf 12"

Solution

The equation to predict Body Weight from Breast Circumference is based on the regression equation that was derived in Problem 1.

$$\widehat{y_k} = \widehat{b_0} + \widehat{b_1} * x_k$$

where $\widehat{b_0}$ and $\widehat{b_1}$ are the estimates of the intercept and the regression coefficient from Problem 1. The variable x_k is the Breast Circumference for the newly measured animal k.

Assume that the results from Problem 1 are stored in a variable n_hat_b, and that the measured values for Breast Circumference are stored in a dataframe with the name tbl_new_ani in a column called 'Breast Circumference. The value $\hat{y_k}$ for the two newly measured animals "Animal 11" and "Calf 12" can be computed as

```
vec_x_k <- tbl_new_ani$`Breast Circumference`
vec_y_k <- n_hat_b[1] + n_hat_b[2] * vec_x_k; vec_y_k</pre>
```

[1] 435.3686 -352.9766

Collecting these results in a table leads to

Table 3: Prediction Results

Animal	Breast Circumference	Predicted Body Weight
Animal 11	181.2	435
Calf 12	99.5	-353

The result of the predicted Body Weight for "Animal 11" corresponds to 435 which is a plausible result. The predicted Body Weight for "Calf 12" is -353 which does not make any sense. As a consequence, we can say that extrapolation of response values based on predictors that are so far away from the range of predictors used to estimate the regression equation is not allowed.