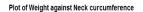
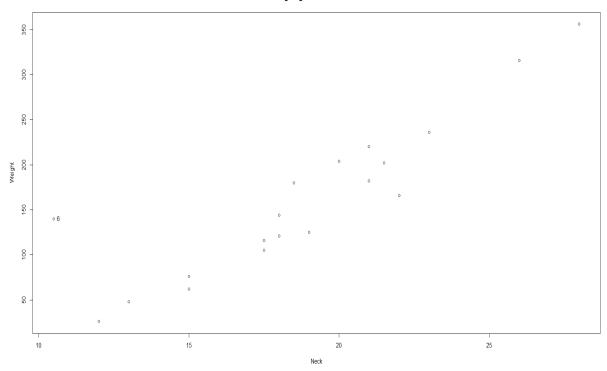
## Stat 6021: Homework Set 10

- 1. In this question, you will revisit the swiss data set that you worked on in Homeworks 6 and 7. The data set contains information regarding a standardized fertility measure and socio-economic indictors for each of the 47 French-speaking provinces of Switzerland around the year 1888. In Homework 5, you found that the model with just three predictors: *Education*, *Catholic*, and *Infant Mortality* was preferred to a model with all the predictors. Fit the model with the three predictors, and answer the following questions.
  - (a) Are there any observations that are outlying? Be sure to show your work and explain how you arrived at your answer.
  - (b) Are there any observations that have high leverage? Be sure to show your work and explain how you arrived at your answer.
  - (c) Are there any influential observations based on DFFITs and Cook's Distance?
  - (d) Briefly describe the difference in what DFFITS and Cook's distance are measuring.

2. (No R Required) Data from n=19 bears of varying ages are used to develop an equation for estimating Weight from Neck circumference. From a visual inspection of the scatterplot, it appears observation 6 may be an outlier.





The output below comes from fitting the linear regression model on the data.

## ##with all 19 bears Coefficients:

Residual standard error: 40.13 on 17 degrees of freedom Multiple R-squared: 0.793, Adjusted R-squared: 0.7809 F-statistic: 65.14 on 1 and 17 DF, p-value: 3.235e-07

The output below comes from fitting the linear regression model on the data, with the outlier removed.

```
##with outlier removed, so 18 bears
Coefficients:
```

Residual standard error: 22.6 on 16 degrees of freedom

Multiple R-squared: 0.938, Adjusted R-squared: 0.9342

F-statistic: 242.2 on 1 and 16 DF, p-value: 4.394e-11

The output below displays the values of the predictor and response for the 6th observation.

> data[6,]
 Neck Weight

6 10.5 140

Some additional information from R, regarding ordinary residuals,  $e_i$ , and leverages,  $h_{ii}$  shown below, from the full data.

## > result\$residuals ##residuals

7	6	5	4	3	2	1
-32.803200	120.829070	-2.276933	23.828133	22.880666	-48.066801	-25.276933
14	13	12	11	10	9	8
34.143331	40.248397	-15.119334	-21.803200	25.249333	-38.224400	-18.592131
		19	18	17	16	15
		-13.539598	-19.434532	4.985732	-33.434532	-3.593068

## > tmp\$hat ##leverages

5 3 7 1 6 0.05422642 0.08132161 0.06633278 0.05682064 0.05422642 0.23960510 0.05700079 9 12 13 10 11  $0.17788427 \ 0.05278518 \ 0.05282121 \ 0.05700079 \ 0.06633278 \ 0.28626504 \ 0.19604381$ 15 16 17 18 19 0.07314261 0.09141025 0.10178713 0.09141025 0.14358291

- (a) Calculate the externally studentized residual,  $t_i$ , for observation 6. Will this be considered outlying in the response?
- (b) What is the leverage for observation 6? Based on the criterion that leverages greater than  $\frac{2p}{n}$  are considered outlying in the predictor(s), is this observation high leverage?
- (c) Calculate the DFFITS for observation 6. Briefly describe the role of leverages in DFFITS.
- (d) Calculate Cook's distance for observation 6.

3. (No R Required) Cook's distance has the equivalent formulae

$$D_{i} = \frac{\left(\hat{\boldsymbol{\beta}} - \hat{\boldsymbol{\beta}}_{(i)}\right)'(\boldsymbol{X}'\boldsymbol{X})\left(\hat{\boldsymbol{\beta}} - \hat{\boldsymbol{\beta}}_{(i)}\right)}{p \text{MSres}}$$
(1)

$$= \frac{r_i^2}{p} \frac{h_{ii}}{1 - h_{ii}}.$$
 (2)

where  $r_i$  denotes studentized residuals. Show that (1) and (2) are equivalent. You may use the following without proof:

$$\hat{\boldsymbol{\beta}} - \hat{\boldsymbol{\beta}}_{(i)} = (1 - h_{ii})^{-1} (\boldsymbol{X}' \boldsymbol{X})^{-1} \boldsymbol{X}_i \boldsymbol{e}_i.$$
 (3)

4. This question is optional (No R Required) Recall that leverage,  $h_{ii}$ , is defined by

$$h_{ii} = \boldsymbol{X_i}' \left( \boldsymbol{X'X} \right)^{-1} \boldsymbol{X_i},$$

where

$$m{X} = \left[ egin{array}{cccc} 1 & x_{1,1} & \cdots & x_{1,k} \\ 1 & x_{2,1} & \cdots & x_{2,k} \\ dots & & & & \\ 1 & x_{n,1} & \cdots & x_{n,k} \end{array} 
ight]$$

and

$$m{X_i} = \left[ egin{array}{c} 1 \\ x_{i,1} \\ dots \\ x_{i,k} \end{array} 
ight].$$

X is called the design matrix, containing the values of the k predictors for all observations, with 1s appended in the first column for the intercept, and  $X_i$  is a column vector that contains the values of the k predictors for observation i, with 1 appended in the first entry. It turns out that the sum of the leverages is

$$\sum_{i=1}^{n} h_{ii} = p,\tag{4}$$

where p denotes the number of parameters in the regression model.

(a) Show that (4) is true in simple linear regression, i.e. when p=2. In case you forgot, the inverse of a  $2\times 2$  matrix is

$$\left[\begin{array}{cc} a & b \\ c & d \end{array}\right]^{-1} = \frac{1}{ad-bc} \left[\begin{array}{cc} d & -b \\ -c & a \end{array}\right].$$

Hint: In your intermediate step, you will need to show that

$$(\boldsymbol{X'X})^{-1} = \begin{bmatrix} \frac{1}{n} + \frac{\bar{x}^2}{\sum_{\substack{i=1\\ -\bar{x}}}^n (x_i - \bar{x})^2} & \frac{-\bar{x}}{\sum_{\substack{i=1\\ i=1}}^n (x_i - \bar{x})^2} \\ \frac{\bar{x}^n}{\sum_{\substack{i=1\\ i=1}}^n (x_i - \bar{x})^2} & \frac{1}{\sum_{\substack{i=1\\ i=1}}^n (x_i - \bar{x})^2} \end{bmatrix}$$

(b) Show that

$$\sum_{i=1}^{n} \sigma^2 \{\hat{y}_i\} = p\sigma^2. \tag{5}$$

In other words, that the sum of the variances of  $\hat{y}_i$  is  $p\sigma^2$ . Hint: how are  $\hat{y}_i$  and  $y_i$  related? You may also use (4).