

Julia Nelson

Ma-331-A

“I pledge my honor that I have  
abided by the Steven’s Honor System.”

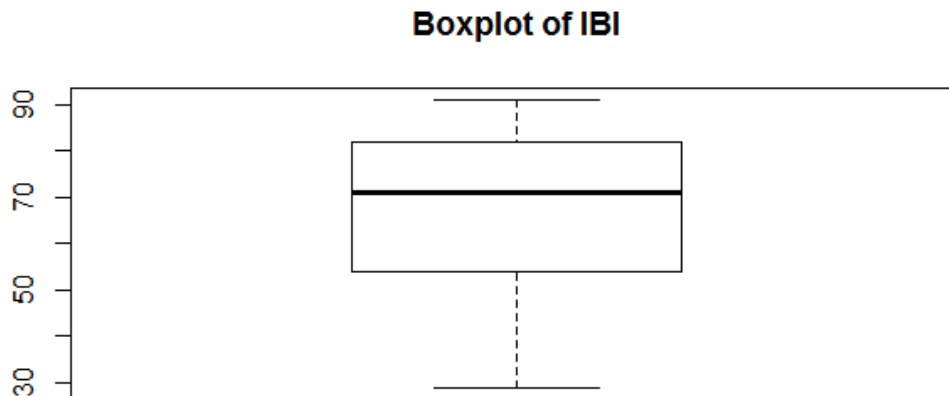
### **Homework 07**

#### Exercise 10.32

- a) In this part of the task, the variable IBI is described using both numeric and graphical tools. After entering the data into R studio and performing some commands on it, the summary below was obtained of the variable IBI.

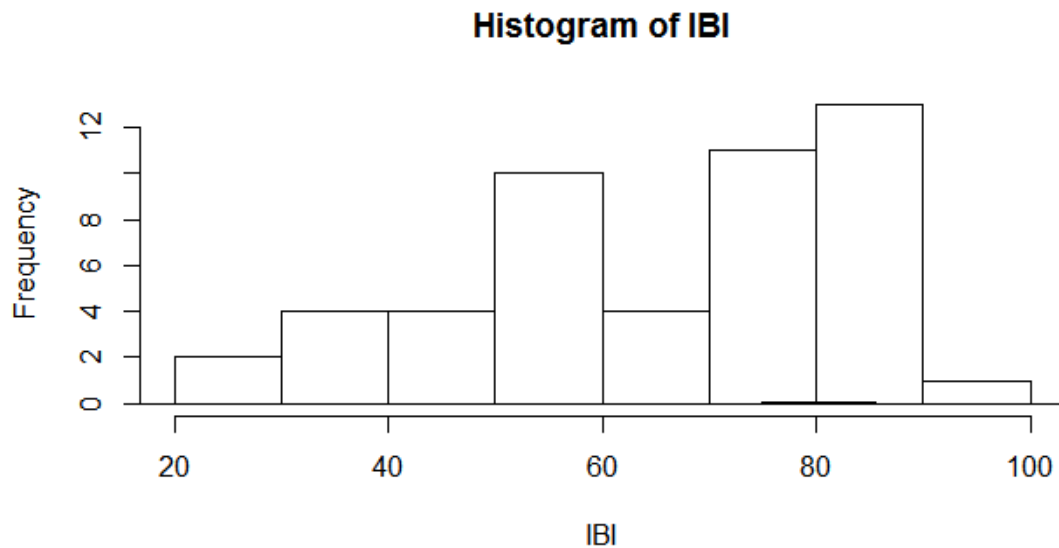
Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
29.00	54.00	71.00	65.53	82.00	91.00

The above output provides the numerical description of the variable IBI. It can be seen that the mean of the variable is 65.53 and its median is 71. Given that the minimum and the maximum value for the variable is 29 and 91 respectively, the data can be presumed to be skewed. In order to be able to visualize the distribution of the data, the boxplot below was generated.



From the boxplot above, it is evident that there are no outliers for the variable IBI. It can also be seen clearly that the median value is closer to the upper quartile as compare to the lower quartile. This implies that there are more large values than they are small values for this variable. To get

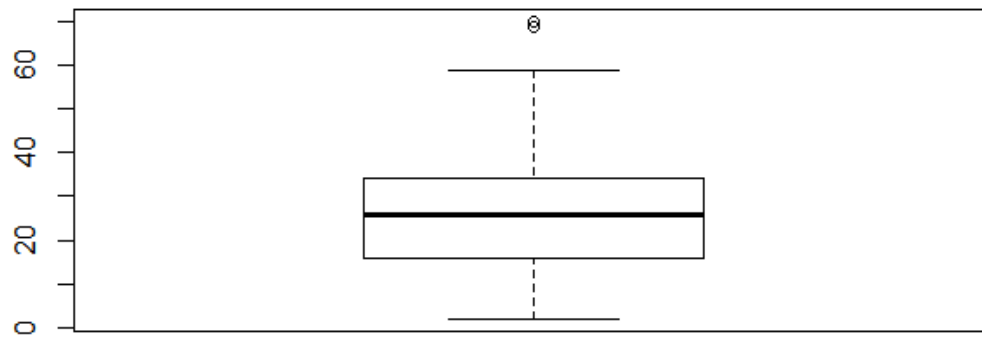
this point even clearer, a histogram of the variable was generated for the data and is as shown below. As seen in the histogram, there are more values to the right hence indicating negative skewness.



Another variable that was of interest to this study was the variable area, which represents the area of watershed in square kilometers for the streams. The numerical description of the data is as shown below. It can be seen that the data is nearly normal since the median and the mean are close to each other, but it indicates that more values are below the mean of the variable. Moreover, a wide gap is noticed between the upper quartile and the maximum value which is an indication of presence of outliers, which can be investigated using a boxplot.

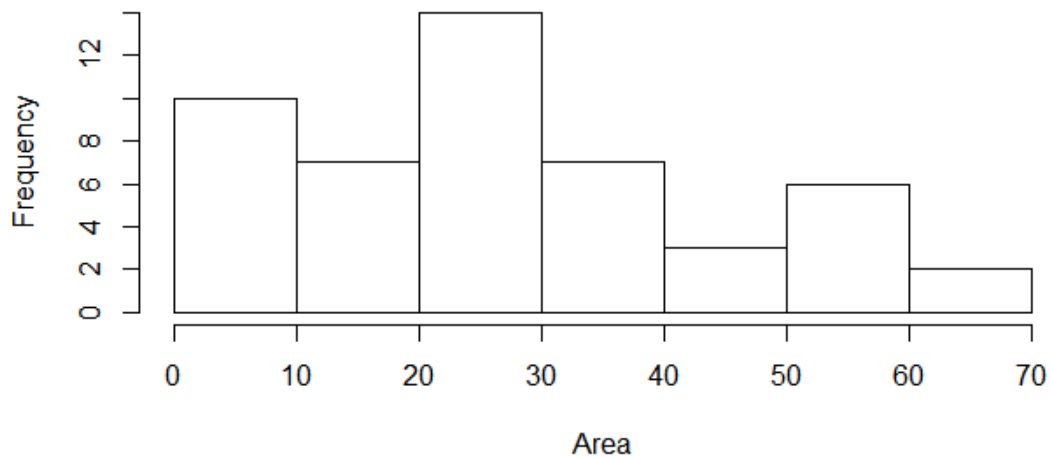
Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
2.00	16.00	26.00	28.29	34.00	70.00

**Boxplot of Area**



The boxplot above shows a nearly normally distributed data that consists of some outliers. Nonetheless, the normality of the data can be investigated using a histogram.

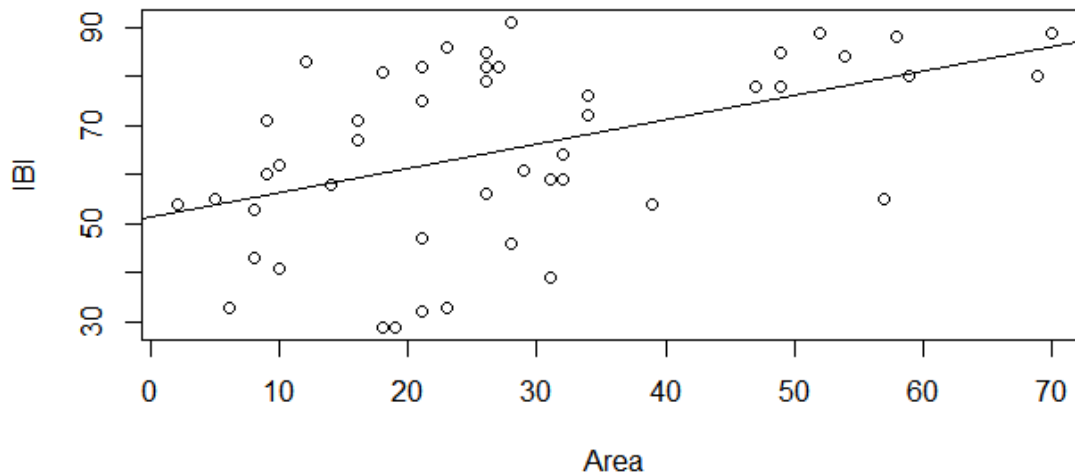
**Histogram of Area**



The above histogram shows that more values are concentrated to the left of the graph, this is an indication of positive skewness in the data.

- b) The relationship between IBI and Area can be investigated using a scatter plot.

**A scatterplot of IBI against Area**



The scatter plot output above shows that the variable IBI and Area have a positive linear relationship. This means that as the value of Area increases, the value of IBI also increases.

- c) The simple linear model for the relationship between IBI and Area can be statistically written as below:

$$IBI = \beta_0 + \beta_1 * Area$$

- d) The null and alternative hypothesis for the model in part (c) can be stated statistically as below

$$H_0: \beta_1 = 0 \text{ vs } H_1: \beta_1 \neq 0$$

- e) Regression results of the model in part (c) above as shown below

```
Call:
lm(formula = IBI ~ Area)

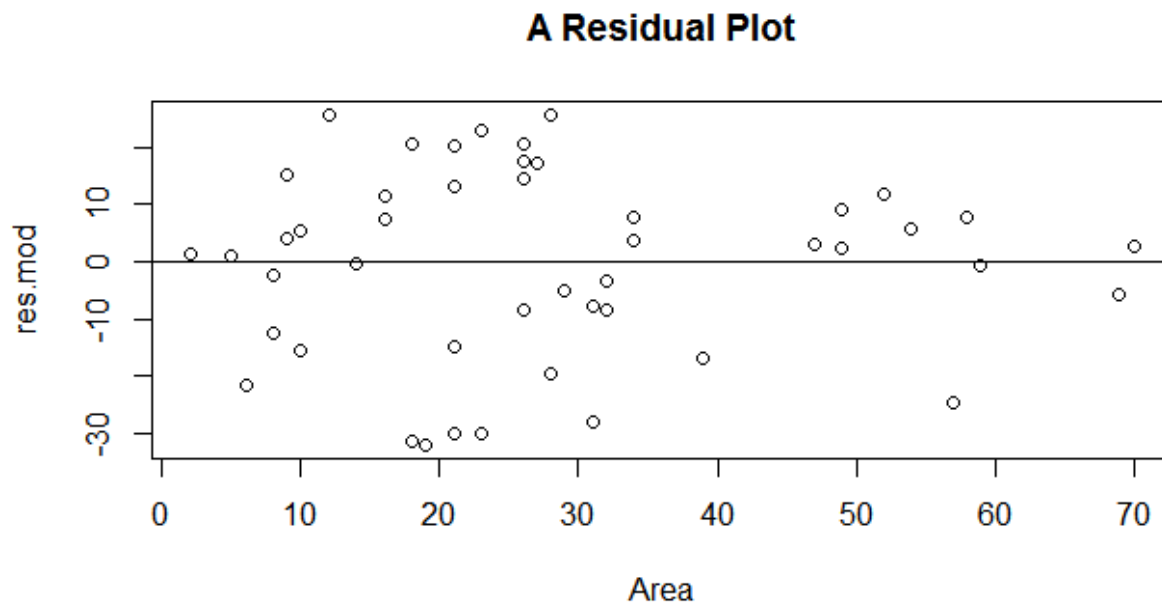
Residuals:
    Min       1Q   Median       3Q      Max
-31.934  -8.399   2.818  11.729  25.611

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  51.5275     4.4074  11.691 1.63e-15 ***
Area         0.4951     0.1324   3.738 0.000502 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 16.25 on 47 degrees of freedom
Multiple R-squared:  0.2292, Adjusted R-squared:  0.2128
F-statistic: 13.97 on 1 and 47 DF, p-value: 0.0005024
```

The results above were obtained after running the model in c) in R. According to the results, the model is significant at 0.05 level of significance;  $F(1,47) = 13.97$ ,  $p=0.000$ . Moreover, the results show that the predictor variable Area is a significant predictor of IBI at 0.05 level of significance;  $p=0.000$ . Therefore, IBI can be predicted using the Area of watershed in square kilometers for the streams. When the Area of the watershed increases by one square kilometer, the IBI increases by 0.495. Moreover, the model has a relatively weak coefficient of determination of 0.2292.

f) The plot of residuals against area is as shown below



The distribution of points in the residual plot above is generally random and distributed all over. This indicates the presence of unbiasedness, homoscedasticity and linearity.

- g) It can be presumed that the data is normally distributed bearing in mind that the data points are randomly and evenly distributed on both sides of the 0 axis.
- h) The Gauss Markov assumptions of a linear regression model are reasonable assumptions that have to be checked when performing a linear regression analysis. For instance, one of the assumptions is that the independent variable should be linearly related to the dependent variable. This is an important assumption since the model is a linear model and the two

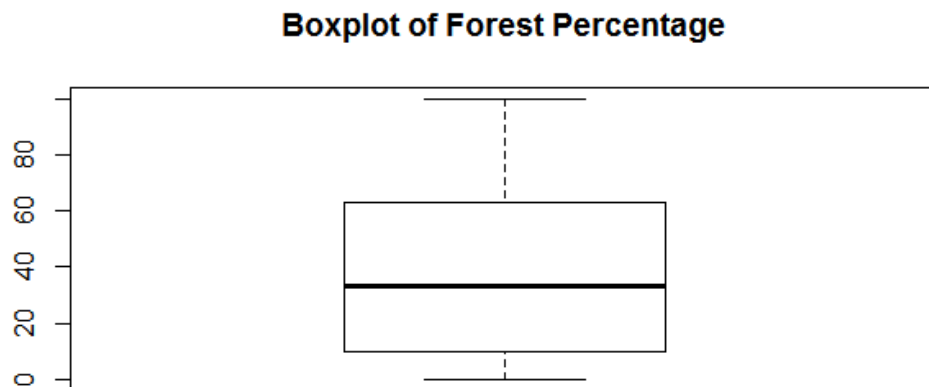
variables should have a linear correlation to bring sense to the model. Another assumption is the normality of the data. Linear regression works better with a data that is normally distributed. A linear regression is not useful for data that is not normal and thus attaining this assumption is crucial before using the data to run a regression analysis.

#### Exercise 10.33

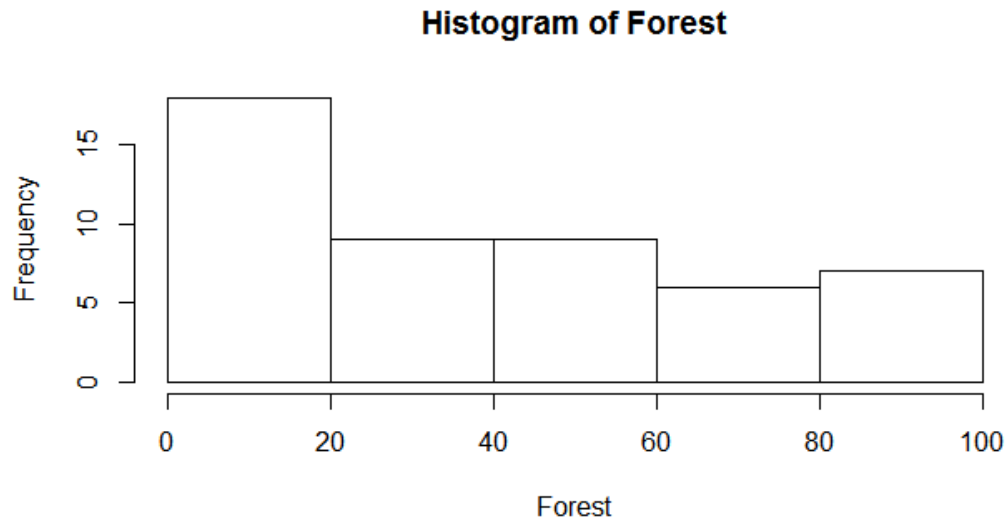
- a) Since the summary of IBI was already obtained in the previous part exercise, the summary of the forest percentage will be done in this section.

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.00	10.00	33.00	39.39	63.00	100.00

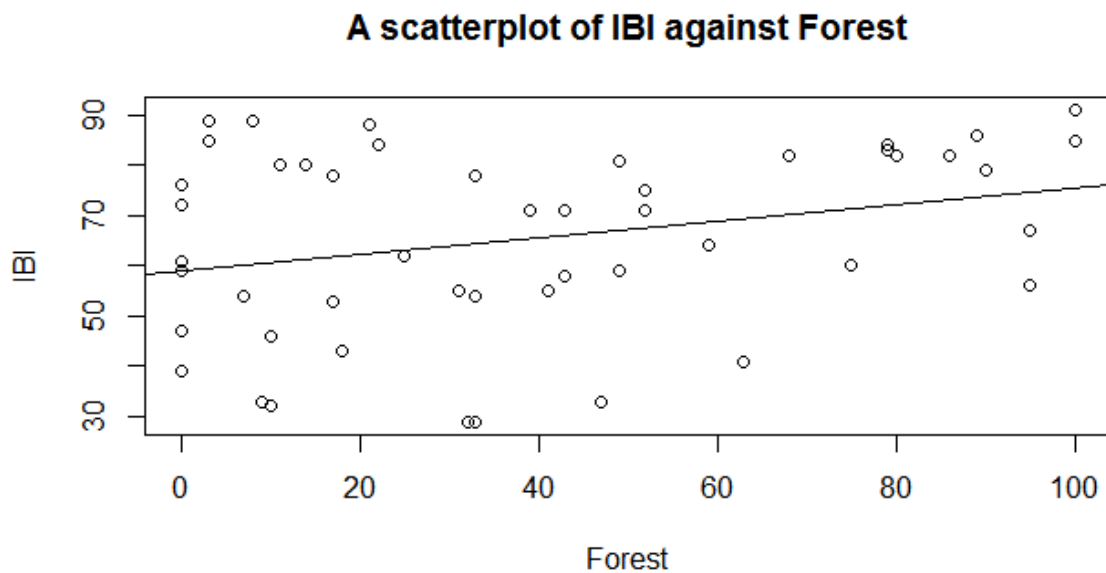
The summary above shows that the median and mean of the variable is 33 and 39.39 respectively. This shows that more data values are concentrated to the lower side of the mean. To put thing into more perspective, a boxplot was generated.



Further, a histogram was generated for the same and it provided the evidence that more values were to the left side of the mean. Beside showing that the data is skewed to the left, the histogram also showed that the data is not normally distributed.



- b) The relationship between IBI and Forest percentage can be investigated using a scatter plot.



The above scatter plot indicates that IBI and Forest have a positive linear relationship, but it can also be seen that not so many data points are close to the linear trend-line. This is an indication of a weak relationship between the two variables.

- c) The simple linear model for the relationship between IBI and Area can be statistically written as below:

$$IBI = \beta_0 + \beta_1 * Forest$$

- d) The null and alternative hypothesis for the model in part (c) can be stated statistically as below

$$H_0: \beta_1 = 0 \text{ vs } H_1: \beta_1 \neq 0$$

- e) Regression results of the model in part (c) above as shown below

```
Call:
lm(formula = IBI ~ Forest)

Residuals:
    Min       1Q   Median       3Q      Max
-35.469 -10.798   3.374  13.013  29.515

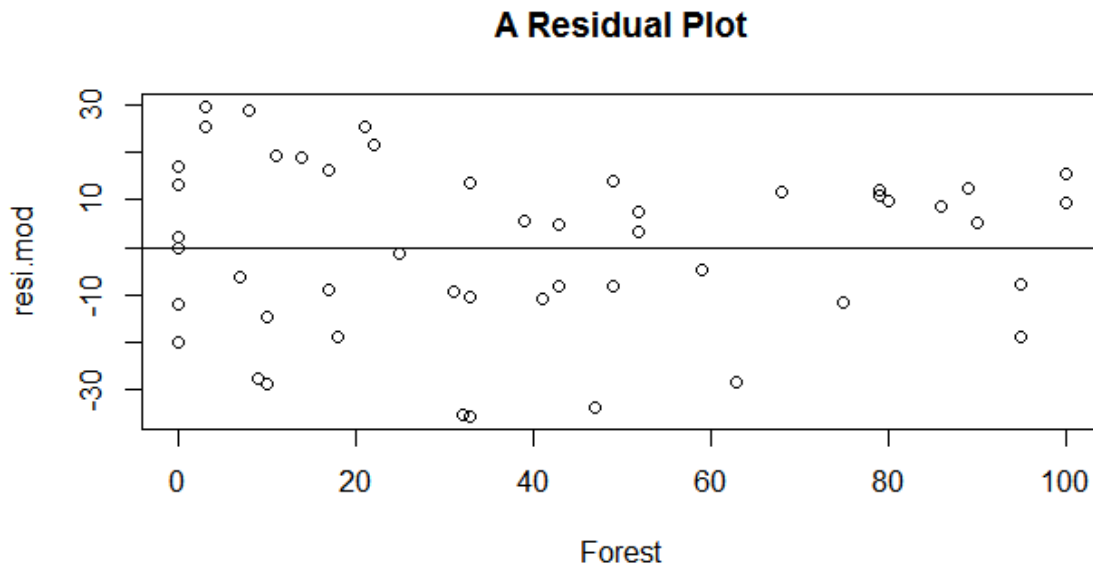
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  58.98657    4.02087   14.670  <2e-16 ***
Forest         0.16614    0.07936    2.094   0.0417 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 17.71 on 47 degrees of freedom
Multiple R-squared:  0.08531, Adjusted R-squared:  0.06585
F-statistic: 4.383 on 1 and 47 DF, p-value: 0.04171
```

From the summary of the model above, it can be seen that the model is significant at 0.05 level of significance and hence the null hypothesis is rejected;  $F(1, 47) = 4.383$ ,  $p=0.0417$ . This implies that the percentage of watershed area that was forest for each stream is a significant predictor of IBI. Also, the model has a very weak goodness of fit coefficient of 0.08531.

- f) The residual plot is as shown below





This residual plot shows a slightly abnormal plot with more values to the left side of the graph compared to the right.

- g) The residuals do not appear to be approximately normal because data points are not evenly distributed across the plot.

#### Exercise 10.34

The two models have turned out to be all significant at 0.05 level of significance. Even though the two models can predict IBI, it is evident from the results that one model is better than the other. When looking at the obtained p-values of the two model the first model of Area has a smaller p-value of 0.000, while the second model has a p-value of 0.0417. This indicates that the first model is more significant compared to the second one. The best measure of a powerful model is, however, not the p-value but the coefficient of determination. The model of Area has a coefficient of determination of 0.2292 which is larger than the second model which has a coefficient of 0.08531. Since the first model has a larger coefficient of determination, it has a better predictive power and hence better as compared to the second one.

### Exercise 10.35

For this exercise, I change case 11 and 30 of the IBI data and run a regression on R and obtained the output below

Call:

```
lm(formula = ibi ~ forest)
```

Residuals:

Min	1Q	Median	3Q	Max
-69.274	-11.182	1.045	17.055	30.485

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	58.18233	5.05088	11.519	2.75e-15 ***
forest	0.11092	0.09968	1.113	0.272

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 22.24 on 47 degrees of freedom

Multiple R-squared: 0.02566, Adjusted R-squared: 0.004934

F-statistic: 1.238 on 1 and 47 DF, p-value: 0.2715

The output above shows that the model has not become totally insignificant;  $p=0.2715$ . This shows that outliers can affect the significance of a model in a negative way. Having outliers affects the distribution of the data and affect its linearity. Presence of outliers makes data to be less linear and thus unable to produce a suitable linear regression model.

### Exercise 10.36

a) Confidence interval at 95% level of confidence

	fit	lwr	upr
1	61.92376	56.86528	66.98224
2	68.35952	63.44646	73.27257
3	54.49788	46.94304	62.05271
4	74.79528	67.96292	81.62764
5	56.47811	49.72863	63.22760
6	75.78540	68.55498	83.01581
7	62.91387	58.03500	67.79275
8	67.36940	62.59452	72.14428
9	57.46823	51.09270	63.84376
10	59.44846	53.74454	65.15238
11	65.88423	61.20914	70.55931
12	75.78540	68.55498	83.01581
13	65.38917	60.71734	70.06100
14	55.48800	48.34436	62.63164
15	79.74586	70.78212	88.70961

```

16 55.98306 49.03881 62.92730
17 66.87434 62.14749 71.60120
18 56.47811 49.72863 63.22760
19 61.92376 56.86528 66.98224
20 64.39905 59.68831 69.10979
21 66.87434 62.14749 71.60120
22 77.27057 69.41307 85.12807
23 61.92376 56.86528 66.98224
24 55.48800 48.34436 62.63164
25 60.43858 55.02285 65.85432
26 54.00282 46.23684 61.76880
27 60.43858 55.02285 65.85432
28 64.39905 59.68831 69.10979
29 64.89411 60.21036 69.57786
30 64.39905 59.68831 69.10979
31 67.36940 62.59452 72.14428
32 52.51765 44.09939 60.93590
33 80.73598 71.31335 90.15861
34 80.24092 71.04873 89.43311
35 60.93364 55.64772 66.21956
36 58.45835 52.43281 64.48388
37 59.44846 53.74454 65.15238
38 55.98306 49.03881 62.92730
39 62.91387 58.03500 67.79275
40 65.38917 60.71734 70.06100
41 68.35952 63.44646 73.27257
42 86.18162 74.12586 98.23739
43 85.68657 73.87597 97.49716
44 78.26069 69.96866 86.55272
45 70.83481 65.36041 76.30921
46 55.98306 49.03881 62.92730
47 61.92376 56.86528 66.98224
48 78.26069 69.96866 86.55272
49 64.39905 59.68831 69.10979

```

>

After running the analysis in R the output above was produced indicating that the confidence interval of the IBI for an area of 40km squared is (65.39, 70.06). This implies that there is a 95% confidence that the true population mean response lies between 65.39 and 70.06.

b) Prediction interval at 95% level of confidence

	fit	lwr	upr
1	61.92376	28.83634	95.01118
2	68.35952	35.29402	101.42502
3	54.49788	20.93801	88.05775
4	74.79528	41.39063	108.19992
5	56.47811	23.09032	89.86591
6	75.78540	42.29707	109.27372
7	62.91387	29.85344	95.97431
8	67.36940	34.32415	100.41465
9	57.46823	24.15402	90.78244
10	59.44846	26.25624	92.64069
11	65.88423	32.85325	98.91520
12	75.78540	42.29707	109.27372
13	65.38917	32.35865	98.41968
14	55.48800	22.01830	88.95769

15	79.74586	45.84103	113.65070
16	55.98306	22.55535	89.41076
17	66.87434	33.83600	99.91269
18	56.47811	23.09032	89.86591
19	61.92376	28.83634	95.01118
20	64.39905	31.36301	97.43509
21	66.87434	33.83600	99.91269
22	77.27057	43.64128	110.89987
23	61.92376	28.83634	95.01118
24	55.48800	22.01830	88.95769
25	60.43858	27.29466	93.58250
26	54.00282	20.39479	87.61085
27	60.43858	27.29466	93.58250
28	64.39905	31.36301	97.43509
29	64.89411	31.86190	97.92631
30	64.39905	31.36301	97.43509
31	67.36940	34.32415	100.41465
32	52.51765	18.75293	86.28236
33	80.73598	46.70695	114.76501
34	80.24092	46.27498	114.20687
35	60.93364	27.81069	94.05659
36	58.45835	25.20934	91.70735
37	59.44846	26.25624	92.64069
38	55.98306	22.55535	89.41076
39	62.91387	29.85344	95.97431
40	65.38917	32.35865	98.41968
41	68.35952	35.29402	101.42502
42	86.18162	51.33151	121.03174
43	85.68657	50.92050	120.45263
44	78.26069	44.52722	111.99416
45	70.83481	37.68126	103.98837
46	55.98306	22.55535	89.41076
47	61.92376	28.83634	95.01118
48	78.26069	44.52722	111.99416
49	64.39905	31.36301	97.43509

After running the analysis in R, it emerged that the IBI when area is 40km squared has a prediction interval of (65.39, 98.42) at 95% level of confidence. This indicates that a predicted future value of IBI when area is 40km squared will lie within this interval with 95% confidence level.

- c) After running the analysis in R the output above was produced indicating that the confidence interval of the IBI for an area of 40km squared is (65.39, 70.06). This implies that there is a 95% confidence that the true population mean response of Ozark highland streams lies between 65.39 and 70.06. Moreover, it emerged that the IBI when area is 40km squared has a prediction interval of (65.39, 98.42) at 95% level of confidence. This indicates that a predicted future value of IBI when area is 40km squared will lie within this interval with 95% confidence level.

- d) These results cannot be applied to other streams in Arkansas because the sample was not large enough to be used to make generalization and the sampling was not done randomly across all rivers in Arkansas.

#### Exercise 10.37

The estimates of the index of biotic integrity can be computed for 10km squared area and 63% forest as shown below

Area 10km squared

$$IBI = 51.53 + 0.5 * Area$$

$$IBI = 51.53 + 0.5 * 10 = 56.53$$

Forest 63%

$$IBI = 58.99 + 0.17 * forest$$

$$IBI = 58.99 + 0.17 * 63 = 69.7$$

The two estimates differ due to the rounding off issue. The percentages were rounded off and hence end up giving larger values as compared to the actual measurement.