HW6Q2

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Load the Lawstat Library to Access the Brown-Forscythe-Levene Test

```
library(lawstat)
```

Read in the Data

```
Advertising = read.csv("Advertising.csv")
attach(Advertising)
```

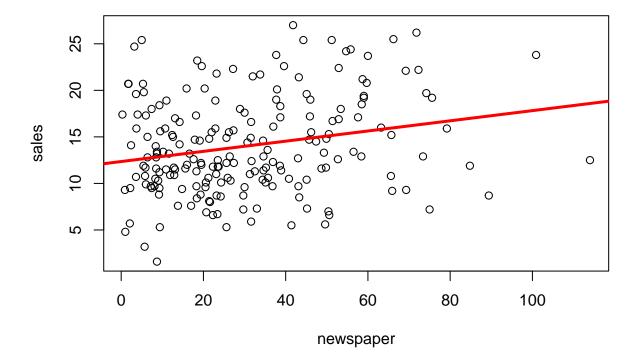
Question 2: Building an SLR Model Sales ~ Newspaper (Initial Results)

Construct a Simple Linear Model with Predictor: Newspaper and Response: Sales

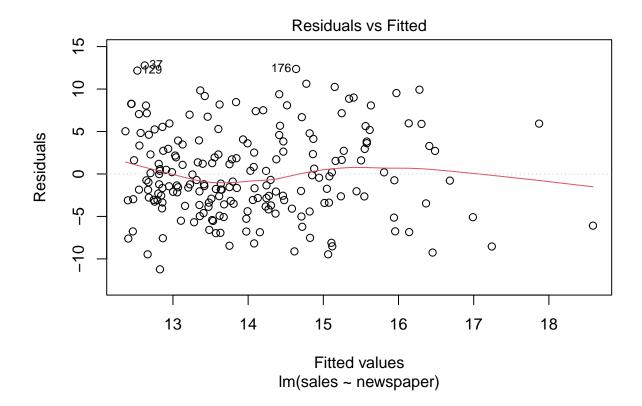
```
lm.fit.news = lm(sales~newspaper, data = Advertising)
```

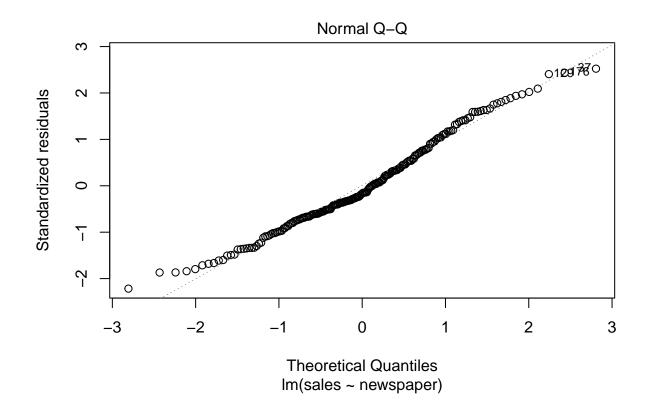
```
##
## Call:
## lm(formula = sales ~ newspaper, data = Advertising)
##
## Residuals:
##
       \mathtt{Min}
                  1Q
                       Median
                                    3Q
                                            Max
## -11.2272 -3.3873 -0.8392
                                3.5059
                                       12.7751
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
                           0.62142
                                     19.88
                                           < 2e-16 ***
## (Intercept) 12.35141
## newspaper
                0.05469
                           0.01658
                                      3.30 0.00115 **
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 5.092 on 198 degrees of freedom
## Multiple R-squared: 0.05212,
                                    Adjusted R-squared: 0.04733
## F-statistic: 10.89 on 1 and 198 DF, p-value: 0.001148
```

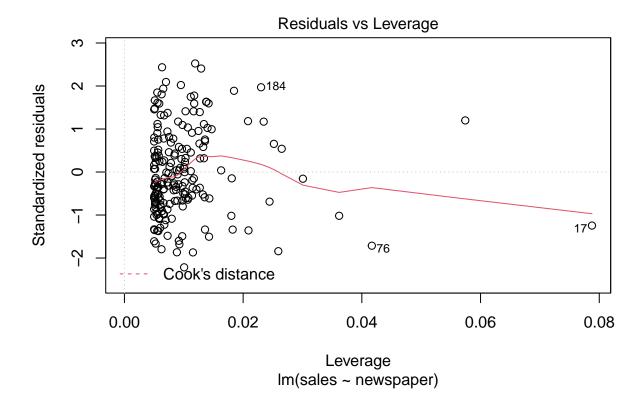
Construct a Scatter Plot with the Calculated Linear Model "Eye-Test" for Abnormalities (Non-linearity, etc.)



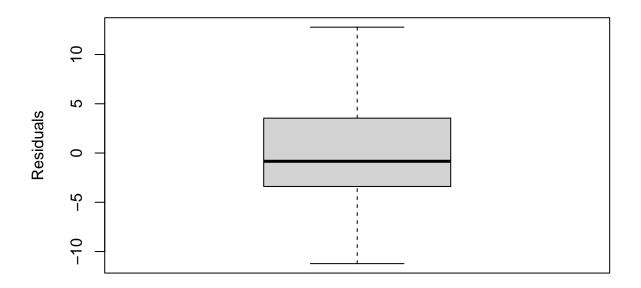
Construct Plots to Check Diagnostics



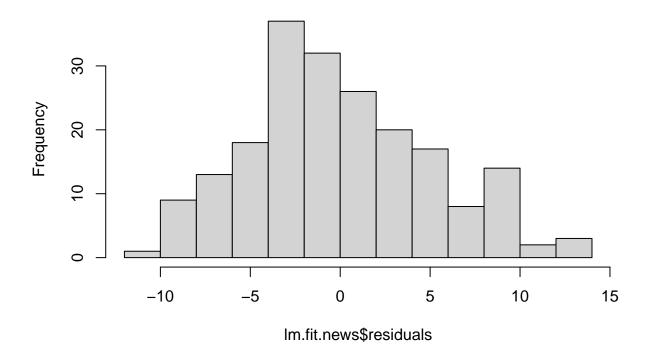




Check Distribution of Residuals for Obvious Deviations



Histogram of Im.fit.news\$residuals



Run Shapiro-Wilk and Brown-Forscythe-Levene Tests

shapiro.test(lm.fit.news\$residuals)

```
##
## Shapiro-Wilk normality test
##
## data: lm.fit.news$residuals
## W = 0.98197, p-value = 0.0114
levene.test(lm.fit.news$residuals, Advertising$Group, location = c("median"))
```

```
##
## Modified robust Brown-Forsythe Levene-type test based on the absolute
## deviations from the median
##
## data: lm.fit.news$residuals
## Test Statistic = 2.7715, p-value = 0.09754
```

Conclusion From the Shapiro-Wilks Test, we conclude that the data is not from a normal distribution. However, from the Levene Test, we conclude that the data is homoscedastic at the .05 significance level, although not very convincingly. Therefore, we must transform the data in hopes of obtaining approximately normal and more convincing homoscedastic data to draw statistically significant conclusions from this data.

Data Transformation Test Results

${\bf Response} \sim {\bf Predictor}$	Shapiro-Wilks P-Value	Levene P-Value
Sales ~ Newspaper	.0114	.09754
$sqrt(Sales) \sim Newspaper$.308	.3703
$log10(Sales) \sim Newspaper$	5.41e-06	.9783
$(1/Sales) \sim Newspaper$	2.2e-16	.2057
$log10(Sales) \sim sqrt(Newspaper)$	8.661 e-06	.9341
$\operatorname{sqrt}(\operatorname{Sales}) \sim \operatorname{sqrt}(\operatorname{Newspaper})$.2702	.4016
$(1/Sales) \sim sqrt(Newspaper)$	2.2e-16	.185
$log10(Sales) \sim log10(Newspaper)$	8.478e-06	.9743
$sqrt(Sales) \sim log10(Newspaper)$.1401	.3659
$(1/Sales) \sim log10(Newspaper)$	2.2e-16	.1831

Conclusion From transforming our data, we obtain more convincing homoscedastic data in all cases, some of which also yielding approximate normally distributed data. The three transformations that show this best are 1. $\operatorname{sqrt}(\operatorname{Sales}) \sim \operatorname{sqrt}(\operatorname{Newspaper})$, 2. $\operatorname{sqrt}(\operatorname{Sales}) \sim \operatorname{Newspaper}$, and 3. $\operatorname{sqrt}(\operatorname{Sales}) \sim \operatorname{log10}(\operatorname{Newspaper})$. Ultimately we will use the $\operatorname{sqrt}(\operatorname{Sales}) \sim \operatorname{sqrt}(\operatorname{Newspaper})$ transformation to analyze our data because it yielded the greatest p-value for the Levene-Test of these three transformations

Analyzing the $sqrt(Sales) \sim sqrt(Newspaper)$ Model

```
summary(lm.fit.news.6)
```

```
##
## Call:
  lm(formula = sqrtSales ~ sqrtNews, data = Advertising)
##
## Residuals:
##
       Min
                  1Q
                       Median
                                    3Q
                                            Max
  -2.25846 -0.43085 -0.06373 0.51934
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                3.31633
                           0.13396
                                    24.756
                                             <2e-16 ***
                                             0.0042 **
## sqrtNews
                0.07019
                           0.02424
                                     2.896
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.697 on 198 degrees of freedom
## Multiple R-squared: 0.04065,
                                    Adjusted R-squared:
                                                         0.0358
## F-statistic: 8.389 on 1 and 198 DF, p-value: 0.0042
```

Interpretation of the Model From the Summary of our Simple Linear Model, we obtain a regression equation of Y = 3.31633 + .07019X, with both coefficients being statistically significant at the .05 significance level. This means that there is an association between the level of Newspaper Advertising and Number of Unit Sales. Therefore the company should increase its Newspaper Advertising budget to indirectly increase Sales. In fact, for every increment of 1 in level of sqrt(Newspaper), (in thousands of dollars), we expect that the response, sqrt(Sales), (in thousands of dollars), increases by .07019. However, the interpretation of our intercept does not make sense. This is because at sqrt(Newspaper) = 0, it would be expected that sqrt(Sales) = 0 by the fact that it would be impossible to have sales with no produced product.