## Lab 13

# David Wiley / Duy Truong March 27, 2019

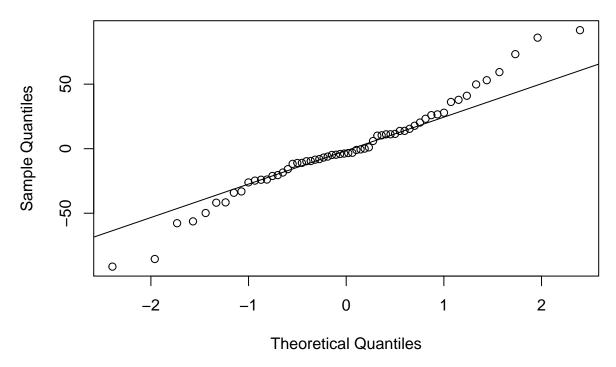
#### Reading in Data

```
# Reading in data
dat = read.csv("/home/david/Documents/2019 Spring/Applied Regression/Labs_HW/Data_Sets/Appendices/data-
fit = lm(MORT -., dat)
summary(fit)
##
## Call:
## lm(formula = MORT ~ ., data = dat)
##
## Residuals:
             1Q Median
##
     Min
                           3Q
                                 Max
## -91.38 -18.97 -3.56 16.00 91.83
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 995.63646   91.64099   10.865   3.35e-15 ***
## PRECIP
                1.40734
                          0.68914
                                   2.042 0.046032 *
## EDUC
              -14.80139
                           7.02747 -2.106 0.039849 *
## NONWHITE
                3.19909
                           0.62231
                                    5.141 3.89e-06 ***
               -0.10797
                           0.13502 -0.800 0.427426
## NOX
## S02
                0.35518
                           0.09096
                                   3.905 0.000264 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 37.09 on 54 degrees of freedom
## Multiple R-squared: 0.6746, Adjusted R-squared: 0.6444
## F-statistic: 22.39 on 5 and 54 DF, p-value: 4.407e-12
```

#### Plotting Normality of Residuals

```
qqnorm(fit$residuals)
qqline(fit$residuals)
```

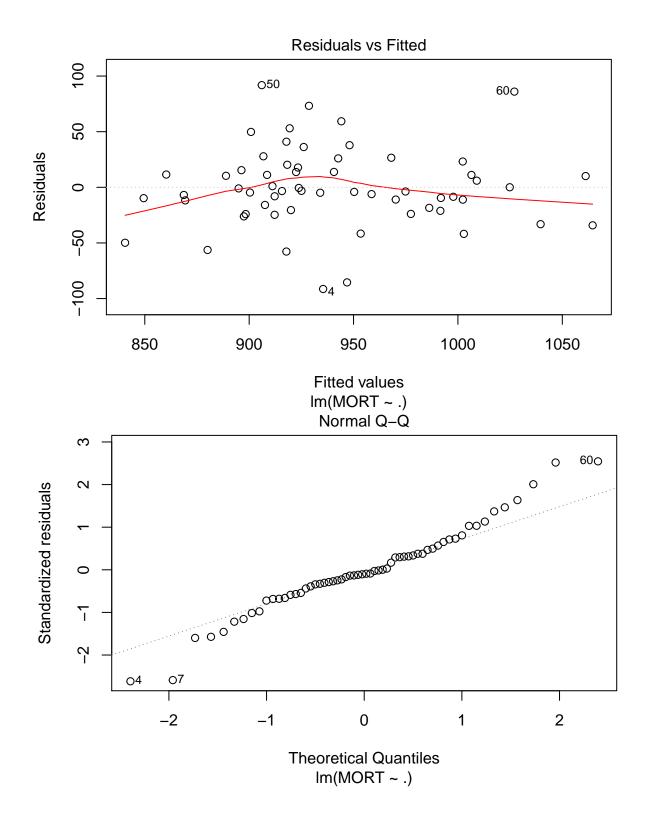
## Normal Q-Q Plot

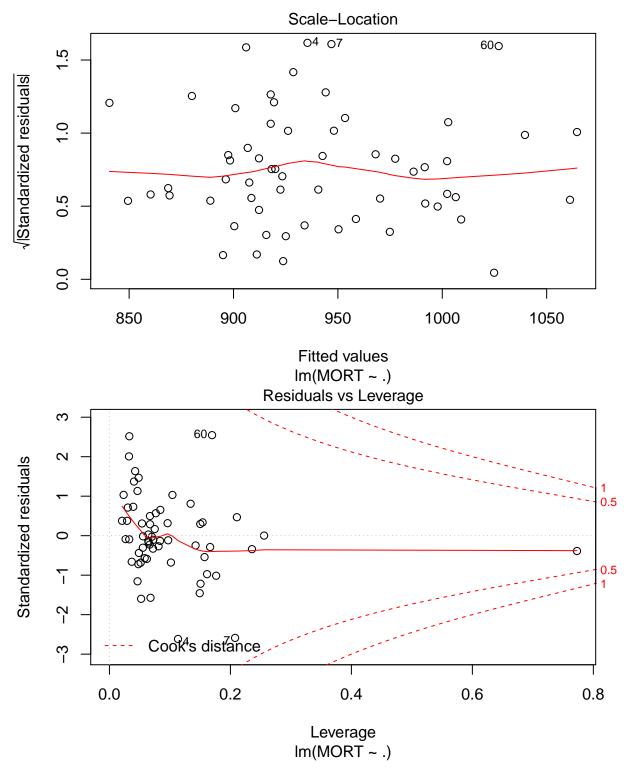


Looking at the plot, we can see there is some light-tailed distribution leading us to believe there may not be normality in the data.

### Plotting Residuals vs Fitted

plot(fit)





From observing that graph of Res vs Fit, we see there is a double bow happening. There means there is non-linearity in the data. This could be corrected by applying an appropriate transformation to the regressor or the response variable or use a method of weight least squares.