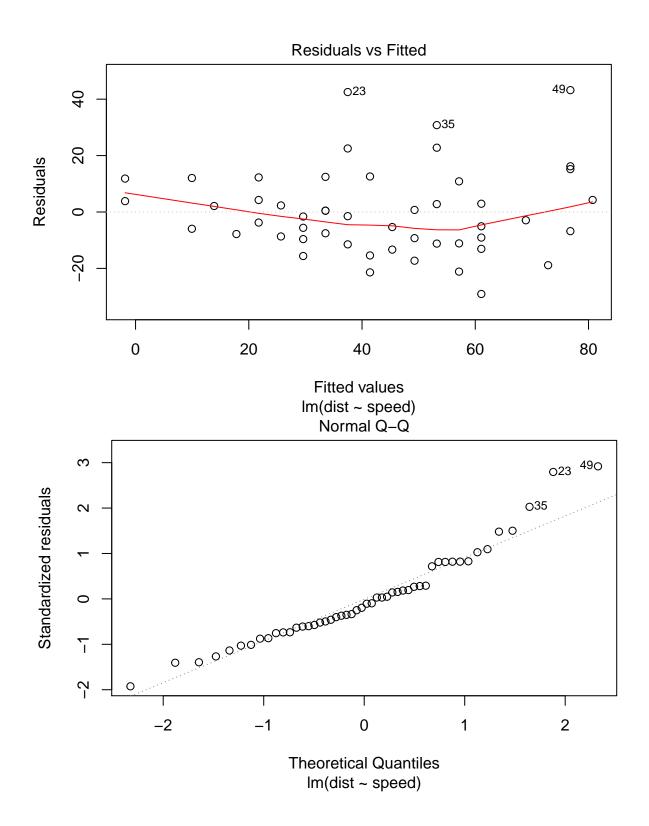
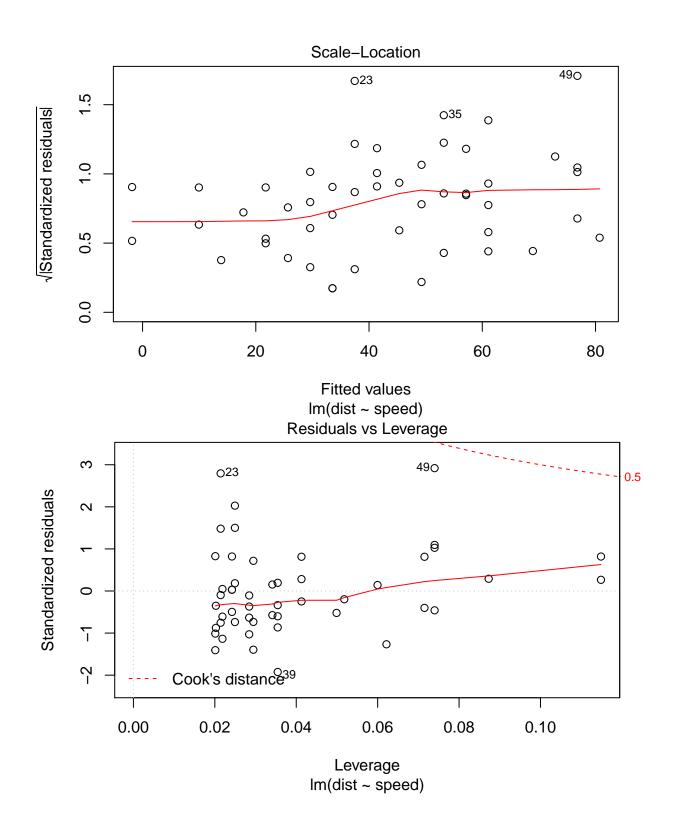
Stat 230: Linear Models Homework 4 Professor Ding Lev Golod

QUESTION 3, PART (A)

We see a strong cone shape in the residual plot, and a weaker cone in the plot of square roots of studentized residuals. This is evidence of heteroskedasticity.

```
lm_model <- lm(dist ~ speed, data = cars)</pre>
# ## Residuals vs Fitted Values
\# plt1a_resid <- ggplot(data.frame(x=lm_model$fitted.values, y=lm_model$residuals),
                        aes(x,y)) +
   geom_point() +
#
   xlab('Fitted Values') +
   ylab('Residuals') +
   ggtitle('Question 3 (A) : Residual vs. Fitted')
# plt1a_resid
# ## Square Root of Studentized Residuals vs Fitted Values
# plt1a_stdresid <- ggplot(data.frame(x=lm_model$fitted.values,</pre>
                                      y=abs(rstandard(lm model))^0.5),
#
                        aes(x,y)) +
#
   qeom_point() +
  xlab('Fitted Values') +
   ylab('Square Root of Studentized Residuals') +
# ggtitle('Question 3 (A) : Square Root of Studentized Residuals vs. Fitted')
# plt1a_stdresid
plot(lm_model)
```





QUESTION 3, PART (B)

The p-value > 0.05; we fail to reject the null hypothesis of homoskedasticity at the level of alpha = 0.05. However, based on the graph in Part (A) I think that there is probably some heteroskedasticity after all.

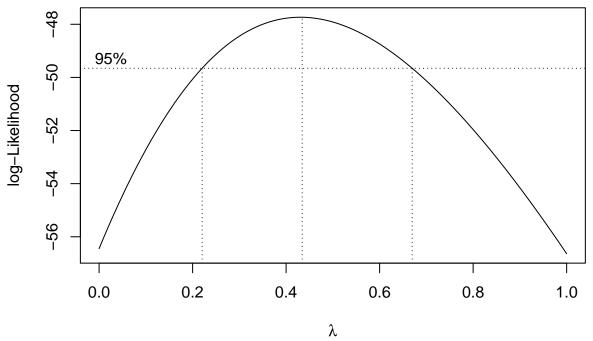
bptest(lm_model)

```
##
## studentized Breusch-Pagan test
##
## data: lm_model
## BP = 3.21, df = 1, p-value = 0.073
```

Question 3, Part (C)

Looking at the Residual and Square-Root of Studentized Residuals plots, we see that there is now less of a fan shape. This means less heteroskedasticity.

boxcox(lm_model,lambda=seq(0,1,by=.05))



```
1 <- 0.433
y_transform <- (cars$dist^1 - 1)/1
newdat <- data.frame(y = y_transform, x = cars$speed)
lm2<- lm(y ~ x, data = newdat)
plot(lm2)</pre>
```

