Question 8

a) The full
summary is as
follows:

I)
Yes, there is
a relationship
between the
response and
the predictor,
as we have 2
non-zero
coefficients

```
Call:
lm(formula = mpg ~ horsepower, data = Auto)
Residuals:
    Min
                  Median
              10
                               30
-13.5710 -3.2592 -0.3435
                           2.7630 16.9240
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 39.935861 0.717499 55.66 <2e-16 ***
horsepower -0.157845 0.006446 -24.49
Signif. codes:
0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 4.906 on 390 degrees of freedom
Multiple R-squared: 0.6059, Adjusted R-squared: 0.6049
F-statistic: 599.7 on 1 and 390 DF, p-value: < 2.2e-16
```

(B0 = 39.935 and B1 = -.0158), each of which are significant at a level of p-value = essentially 0 (as shown by the coefficients section of the table)

- ii) The relationship is fairly strong between the response and the predictor, as we have an R^2 value of .6059. This is not super strong, but is strong enough to suggest a relationship, as around 60% of the variation in mpg can be accounted for by the change in horsepower
- iii) The relationship is a negative relationship between mpg and horsepower, since the B1 coefficient is negative. As horsepower goes up, mpg goes down.

Other requests)

I extracted the coefficients as such:

```
coeffs = summary(model)$coefficients
B0 = coeffs["(Intercept)", "Estimate"]
B1 = coeffs["horsepower", "Estimate"]
```

RSE: 4.906

> print(sigma(model)) [1] 4.905757

95% confidence intervals:

RSE for B0: .717 RSE for B1: .006 Coefficients:

Estimate Std. Error (Intercept) 39.935861 0.717499 horsepower -0.157845 0.006446

B0: [38.5, 41.37] [1] "B0 Confidence interval: ---" [1] "[B0 - 2 * RSEB0, B0 + 2 * RSEB0]" [1] 38.50086 [1] 41.37086

B1: [-.17, -.145]

[1] "B1 Confidence interval: ---"

[1] "[B1 - 2 * RSEB1, B1 + 2 * RSEB1]"

[1] -0.1707357

[1] -0.1449537