

INTRODUCTION TO STATISTICAL LEARNING

QUESTION-1:

For this question, use of simple linear regression on the 'Auto' data set.

(a) Use the `lm()` function to perform a simple linear regression with **mpg** as the response and **horsepower** as the predictor. Use the `summary()` function to print the results to include in your submission. Comment on the output.

- i. Is there a relationship between the predictor and the response?
- ii. How strong is the relationship between the predictor and the response?
- iii. Is the relationship between the predictor and the response positive or negative?
- iv. What is the predicted **mpg** associated with a **horsepower** of 98? What are the associated 95% confidence and prediction intervals?

(b) Plot the response and the predictor. Use the `abline()` function to display the least squares regression line. Include this graph.

Ans:

Loading the data

```
R Console

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

> ISLAs1<-read.csv("C:\\Users\\sudheesha\\Desktop\\ISL 1\\Auto-rev.csv")
> ISLAs1
  mpg cylinders displacement horsepower weight acceleration year origin
1  18.0         8      307.0        130   3504         12.0    70      1
2  15.0         8      350.0        165  3693         11.5    70      1
3  18.0         8      318.0        150  3436         11.0    70      1
4  16.0         8      304.0        150  3433         12.0    70      1
5  17.0         8      302.0        140  3449         10.5    70      1
6  15.0         8      429.0        198  4341         10.0    70      1
7  14.0         8      454.0        220  4354          9.0    70      1
8  14.0         8      440.0        215  4312          8.5    70      1
9  14.0         8      455.0        225  4425         10.0    70      1
10 15.0         8      390.0        190  3850          8.5    70      1
11 15.0         8      383.0        170  3563         10.0    70      1
12 14.0         8      340.0        160  3609          8.0    70      1
13 15.0         8      400.0        150  3761          9.5    70      1
14 14.0         8      455.0        225  3086         10.0    70      1
15 24.0         4      113.0         95  2372         15.0    70      3
16 22.0         6      198.0         95  2833         15.5    70      1
17 18.0         6      199.0         97  2774         15.5    70      1
18 21.0         6      200.0         85  2587         16.0    70      1
19 27.0         4       97.0         88  2130         14.5    70      3
20 26.0         4       97.0         46  1835         20.5    70      2
21 25.0         4      110.0         87  2672         17.5    70      2
22 24.0         4      107.0         90  2430         14.5    70      2
23 25.0         4      104.0         95  2375         17.5    70      2
24 26.0         4      121.0        113  2234         12.5    70      2
25 21.0         6      199.0         90  2648         15.0    70      1
26 10.0         8      360.0        215  4615         14.0    70      1
```

- a. Using `lm` function, we performed a linear regression in which `mpg` is the response and `horsepower` is the predictor. The data obtained from this is stored in `lmdata` variable and the output can be obtained by executing the following statement `summary(lmdata)`;

```
> lm.fit<-lm(mpg~horsepower,data=ISLA1)
> summary(lm.fit)

Call:
lm(formula = mpg ~ horsepower, data = ISLA1)

Residuals:
    Min       1Q   Median       3Q      Max
-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  <2e-16 ***
---
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
```

- i. Yes, It's clearly evident that there is relationship between `mpg` and `horsepower`. We can say that by the p-value ($<2.2e-16$) corresponding to the F-statistic value.
- ii. From the above data, It is clear that $R\text{-square} = 0.6059$. That means almost 60.59% variation in `mpg` can be explained by using `horsepower`.
- iii. You can see that, the co-efficient of `horsepower` is -0.157845 from the screenshot attached above, which shows that the relationship between predictor and `horsepower` is negative.
- iv. We can have a glance at the responses for the question, in the screenshot below.

```
> predict(lm.fit,data.frame(horsepower=c(98)),interval="prediction")
      fit      lwr      upr
1 24.46708 14.8094 34.12476
> predict(lm.fit,data.frame(horsepower=c(98)),interval="confidence")
      fit      lwr      upr
1 24.46708 23.97308 24.96108
```

- b. Below screenshot is for the response and the predictor using the `abline()` function to display the least squares regression line.

Firstly, we have attached the data [`attach(assignment1)`] and then plotted the graph for `horsepower` vs `mpg` [`plot(horsepower~mpg)`]

Introduction to Statistical Learning – Assignment 1

