

# BACS HW11

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## Question 1

a. Let's dig into what regression is doing to compute model fit

```
pts <- data.frame(  
  x = c(  
    -4.704724, 3.966620, 3.448928, 11.861426, 11.473157,  
    20.662193, 18.462001, 27.909883, 25.709691, 35.416420,  
    32.957382, 41.887572, 41.369880, 49.652955, 7.331619  
  ),  
  y = c(  
    4.682789, -1.593332, 14.096971, 7.472177, 22.465133,  
    16.537685, 31.879314, 25.951867, 41.293496, 29.089927,  
    45.826250, 39.550129, 50.010331, 48.615637, 5.728810  
  )  
)
```

i. Plot Scenario 2, storing the returned points

```
regr <- lm(y ~ x, data = pts)  
summary(regr)
```

ii. Run a linear model of x and y points to confirm the R2 value reported by the simulation

```
##  
## Call:  
## lm(formula = y ~ x, data = pts)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -10.299  -6.970  -2.898   6.492  12.215   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept)   4.9885     3.6651   1.361   0.197
```

```
## x          0.9370      0.1362    6.878 1.12e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.298 on 13 degrees of freedom
## Multiple R-squared:  0.7844, Adjusted R-squared:  0.7678
## F-statistic: 47.3 on 1 and 13 DF,  p-value: 1.123e-05
```

### iii. Add line segments to the plot

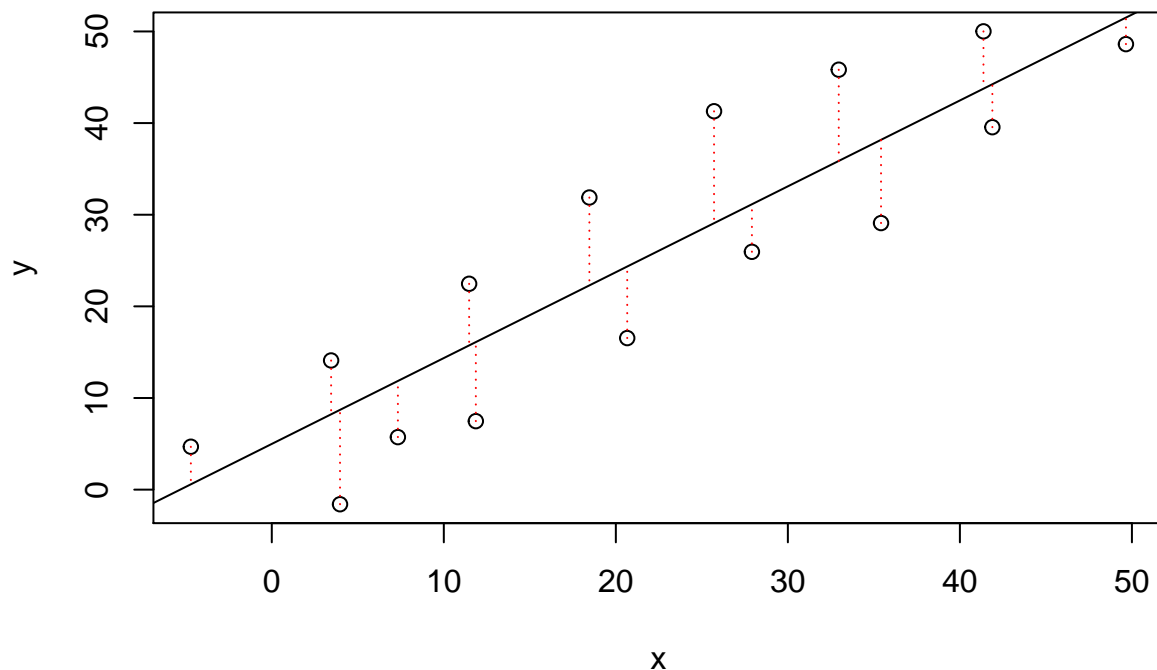
1. Get values of  $\hat{y}$  (estimated values)

```
y_hat <- regr$fitted.values
y_hat
```

```
##          1          2          3          4          5          6          7
## 0.5801637 8.7052424 8.2201632 16.1027022 15.7388929 24.3490507 22.2874633
##          8          9         10         11         12         13         14
## 31.1401607 29.0785733 38.1738112 35.8696843 44.2373025 43.7522233 51.5134926
##          15
## 11.8582578
```

2. Add segments

```
plot(pts)
abline(lm(pts$y ~ pts$x))
segments(pts$x, pts$y, pts$x, y_hat, col="red", lty="dotted")
```



```
sse <- sum((fitted(regr) - mean(pts$y))^2)
ssr <- sum((fitted(regr) - pts$y)^2)
sst <- sse + ssr
r2 <- 1 - (ssr / sst)
```

iv. Use only `pts$x`, `pts$y`, `y_hat` and `mean(pts$y)` to compute SSE, SSR and SST, and verify  $R^2$

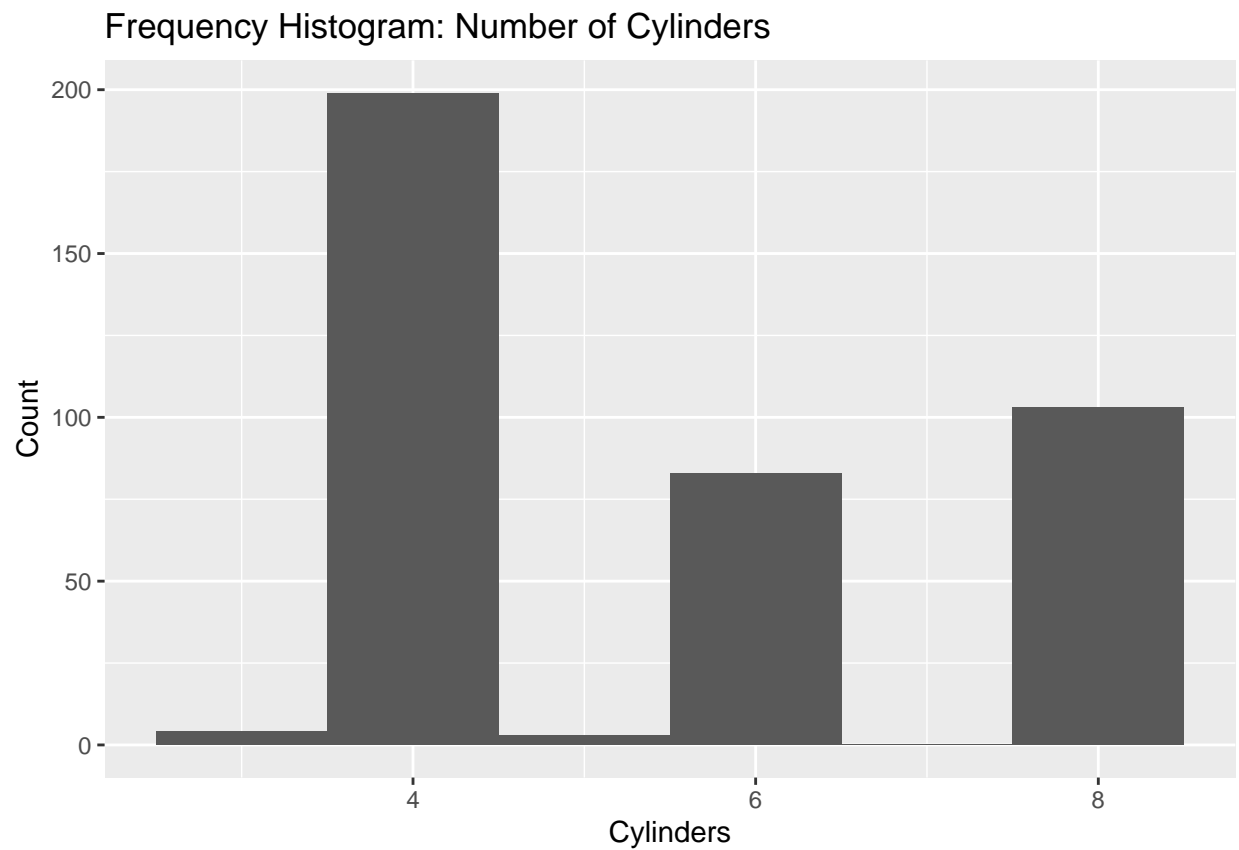
**b. Comparing scenarios 1 and 2, which do we expect to have a stronger  $R^2$ ?** For the first scenario,  $R^2$  will be very close to +1 since most of the data points are sitting at or close to the increasing regression line. However, the second scenario's  $R^2$  value won't be as high as the first scenario's, but it still will be near 1.

In this case, the first scenario will have a stronger  $R^2$ .

**c. Comparing scenarios 3 and 4, which do we expect to have a stronger  $R^2$ ?** In the third scenario, the  $R^2$  will be close to  $-1$  since most of the data points are sitting on or close to the decreasing regression line. However, the fourth scenario's  $R^2$  value won't be as high as the third scenario's, but it still will be near  $-1$ .

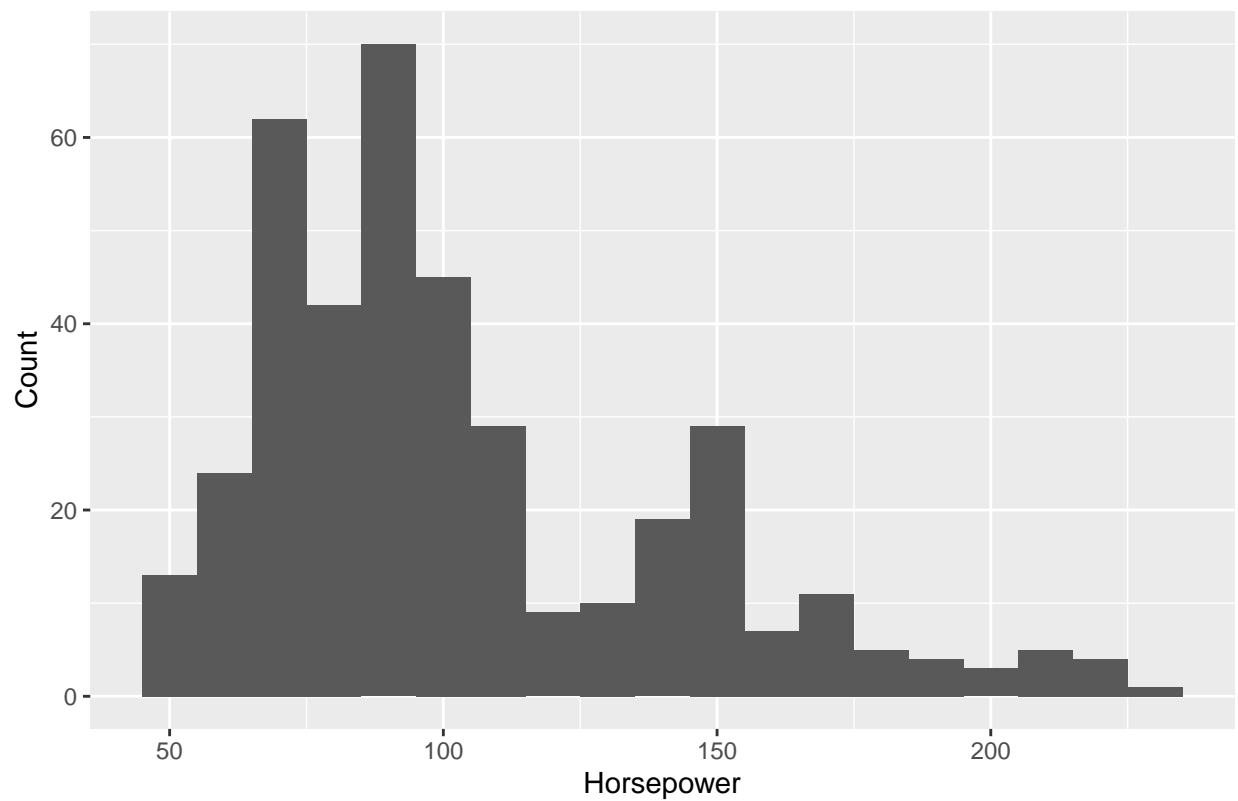
In this case, the third scenario will have a stronger  $R^2$ , but in a decreasing manner.

```
qplot(auto$cylinders, xlab = 'Cylinders', ylab = 'Count',  
      main='Frequency Histogram: Number of Cylinders', binwidth = 1)
```



```
qplot(auto$horsepower, xlab = 'Horsepower', ylab = 'Count', binwidth = 10,  
      main='Frequency Histogram: Horsepower')
```

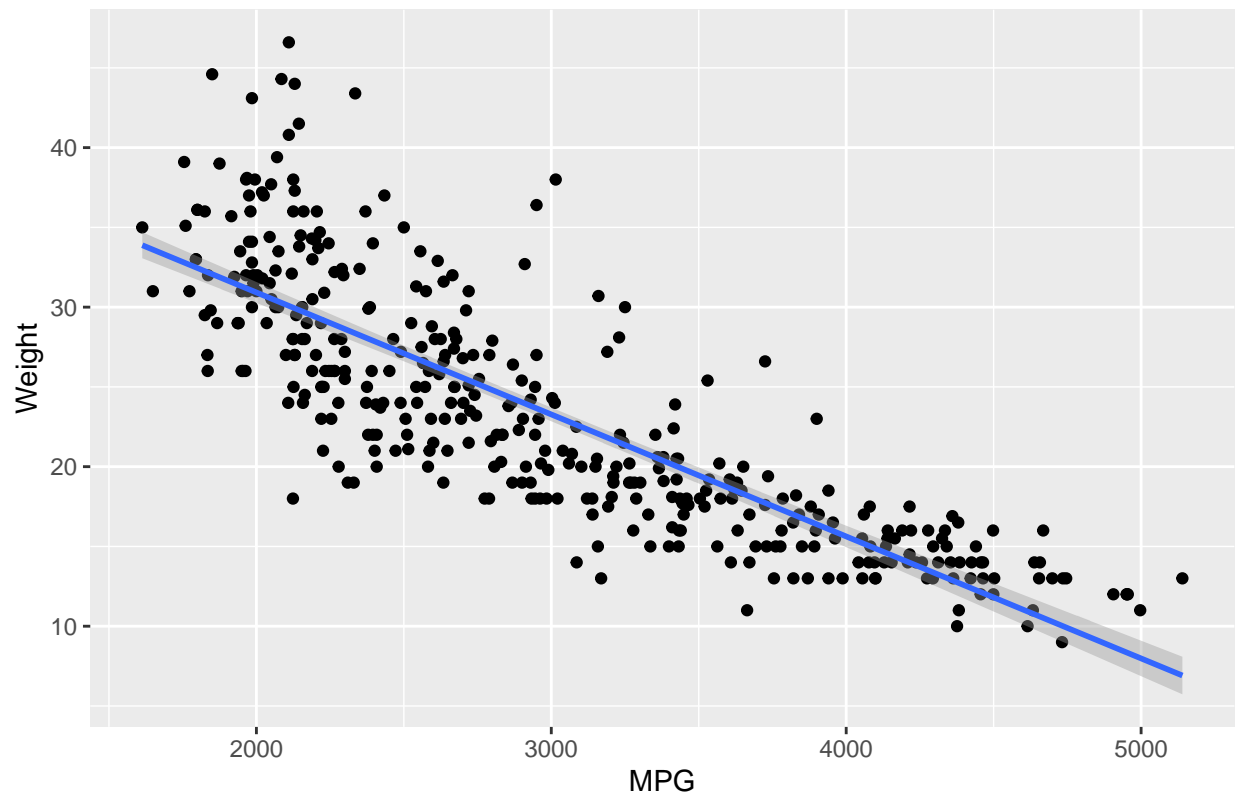
Frequency Histogram: Horsepower



```
ggplot(data = auto, aes(x = weight, y = mpg)) +  
  geom_point() +  
  geom_smooth(method = lm) +  
  xlab('MPG') +  
  ylab('Weight') +  
  ggtitle('MPG vs. Weight: Entire Sample')
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

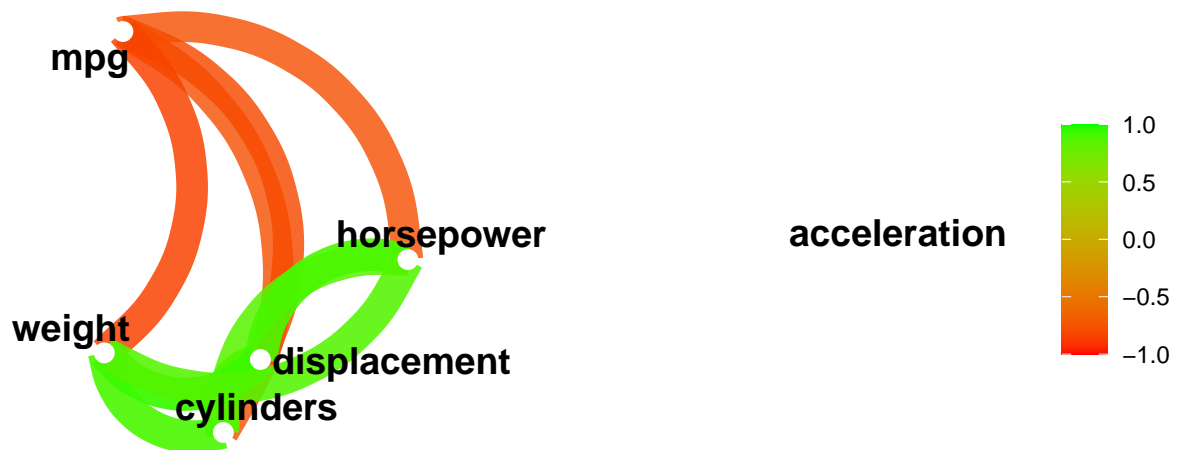
MPG vs. Weight: Entire Sample



```
library(corr)
variables <-
  auto[,
    c("mpg",
      "cylinders",
      "displacement",
      "horsepower",
      "weight",
      "acceleration")
  ]
variables %>%
  correlate() %>%
  network_plot(min_cor = 0.7, colors = c("red", "green"), legend = TRUE)
```

ii. Report a correlation table of all variables, rounding to two decimal places

```
##
## Correlation method: 'pearson'
## Missing treated using: 'pairwise.complete.obs'
```



```
library("PerformanceAnalytics")
```

```
## Loading required package: xts
```

```
## Loading required package: zoo
```

```
##
```

```
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##   as.Date, as.Date.numeric
```

```
##
```

```
## Attaching package: 'xts'
```

```
## The following objects are masked from 'package:dplyr':
```

```
##
```

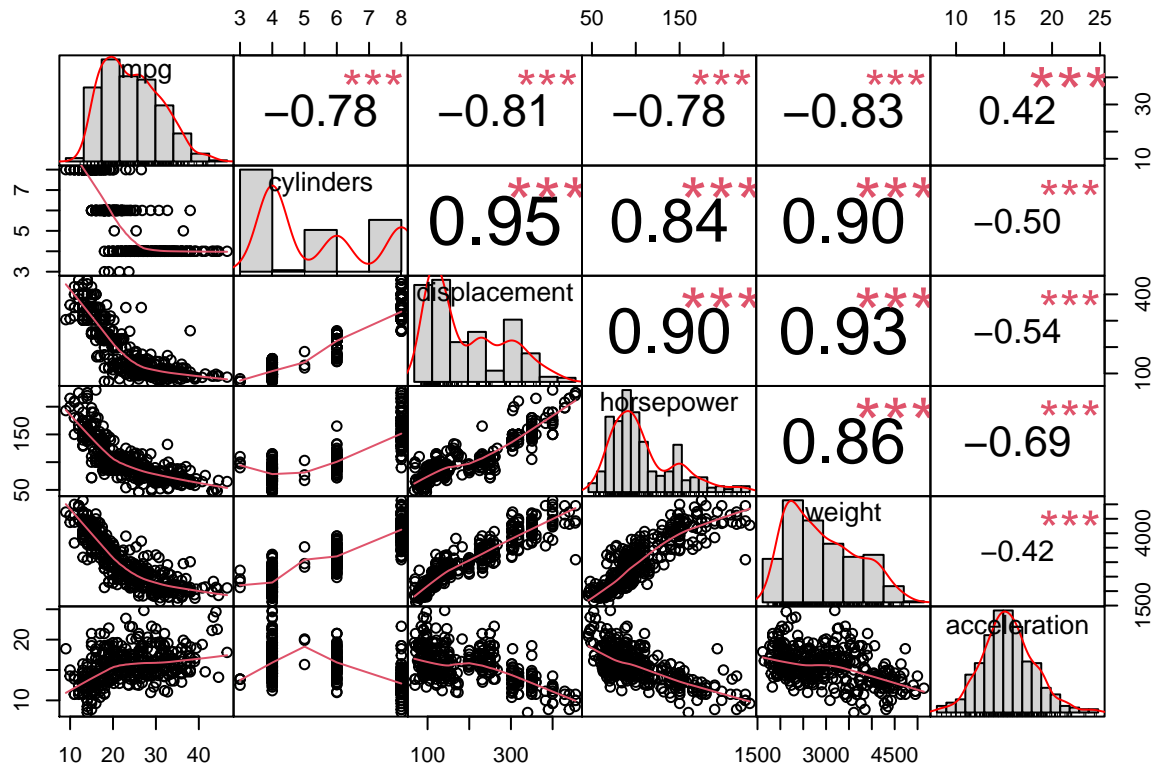
```
##   first, last
```

```
##
```

```
## Attaching package: 'PerformanceAnalytics'
```

```
## The following object is masked from 'package:graphics':
##
##     legend
```

```
chart.Correlation(variables, histogram=TRUE, pch=19)
```



## Reference:

R SQUARED: SST, SSE AND SSR The Correlation Coefficient (r)