

## Problem 1

Solution:

- a. Quantitative: mpg" "cylinders" "displacement" "horsepower" "weight"  
"acceleration" "year"

Qualitative: "origin" "name"

- b. Range of each quantitative Variables

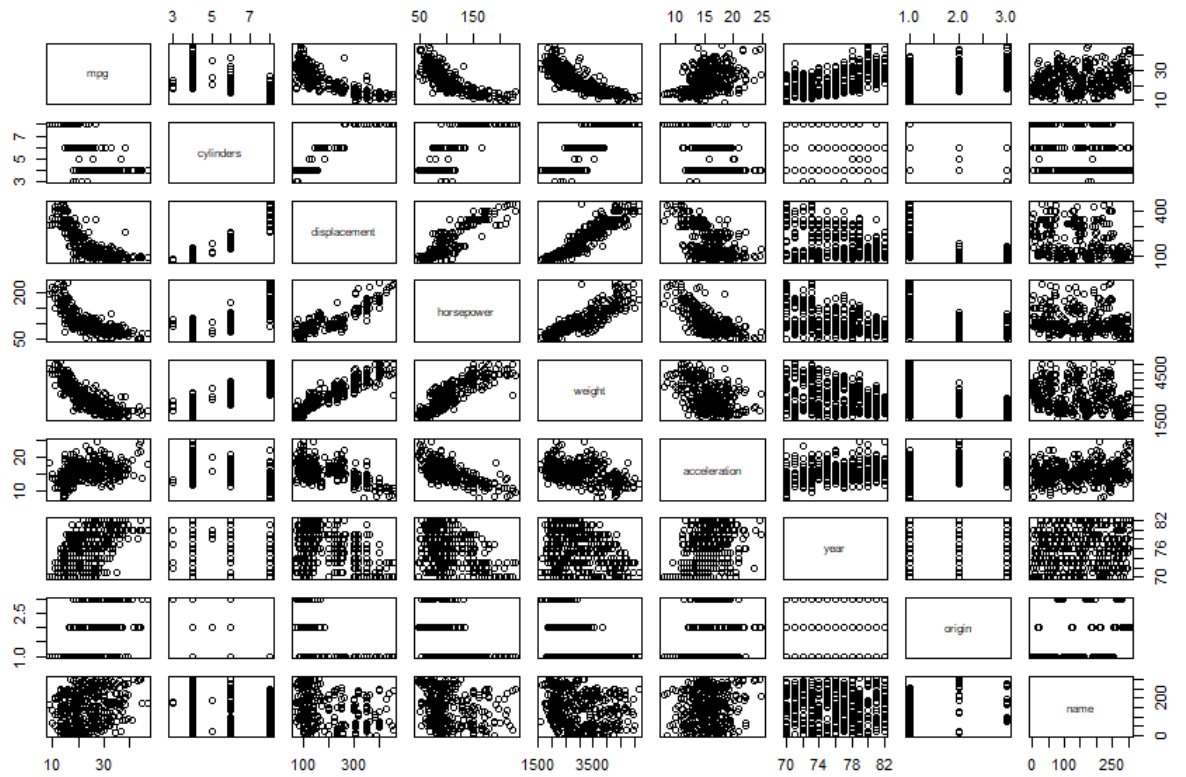
mpg1	mpg2	cylinders1	cylinders2	displacement1		
9.0	46.6	3.0	8.0	68.0		
displacement2	horsepower1	horsepower2	weight1	weight2		
455.0	46.0	230.0	1613.0	5140.0		
acceleration1	acceleration2	year1	year2			
8.0	24.8	70.0	82.0			

- c. Mean and Standard deviation

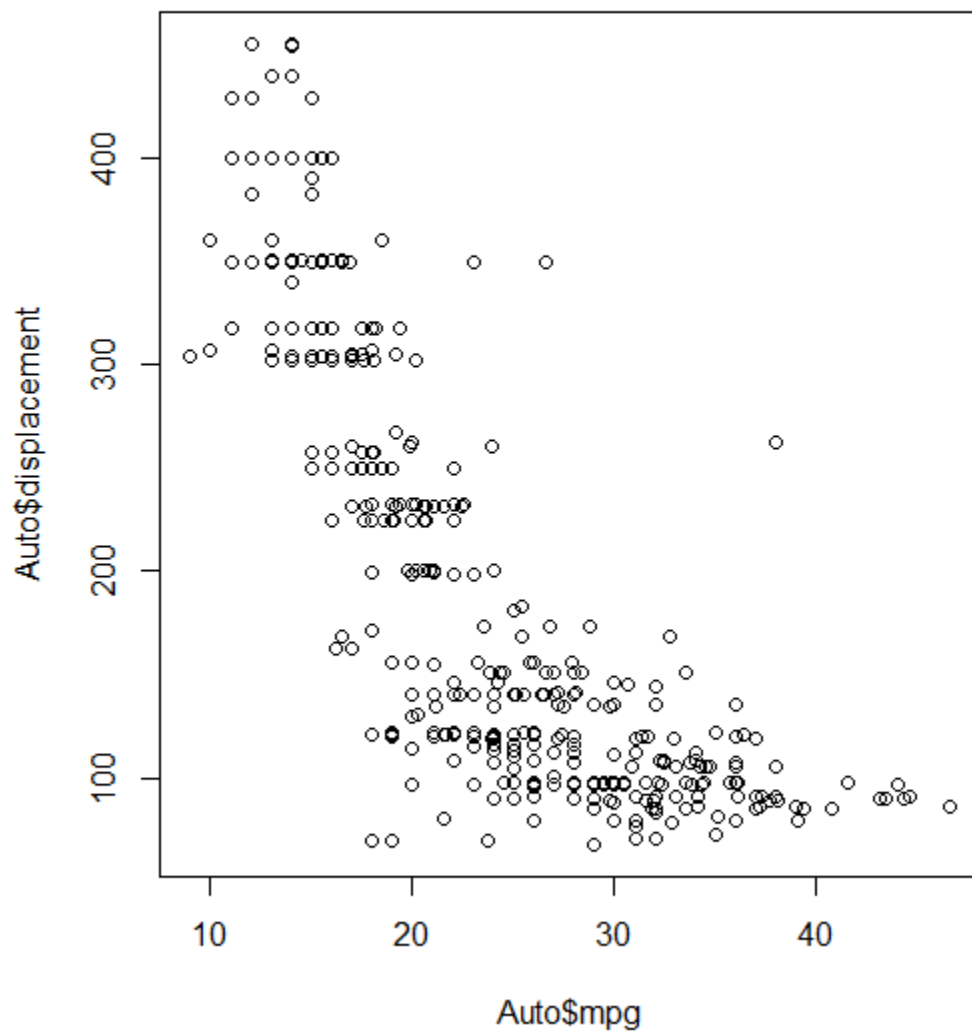
Variables	Mean	Standard Deviation
mpg	23.445918	7.805007
cylinders	5.471939	1.705783
displacement	194.411990	104.644004
horsepower	104.469388	38.491160
weight	2977.584184	849.402560
acceleration	15.541327	2.758864
year	75.979592	3.683737

- d.
- |      |       |           |              |            |         |              |       |
|------|-------|-----------|--------------|------------|---------|--------------|-------|
|      | mpg   | cylinders | displacement | horsepower | weight  | acceleration | year  |
| min  | 11.00 | 3.00      | 68.00        | 46.00      | 1649.00 | 8.50         | 70.00 |
| max  | 46.60 | 8.00      | 455.00       | 230.00     | 4997.00 | 24.80        | 82.00 |
| mean | 24.40 | 5.37      | 187.24       | 100.72     | 2935.97 | 15.73        | 77.15 |
| sd   | 7.87  | 1.65      | 99.68        | 35.71      | 811.30  | 2.69         | 3.11  |

e.

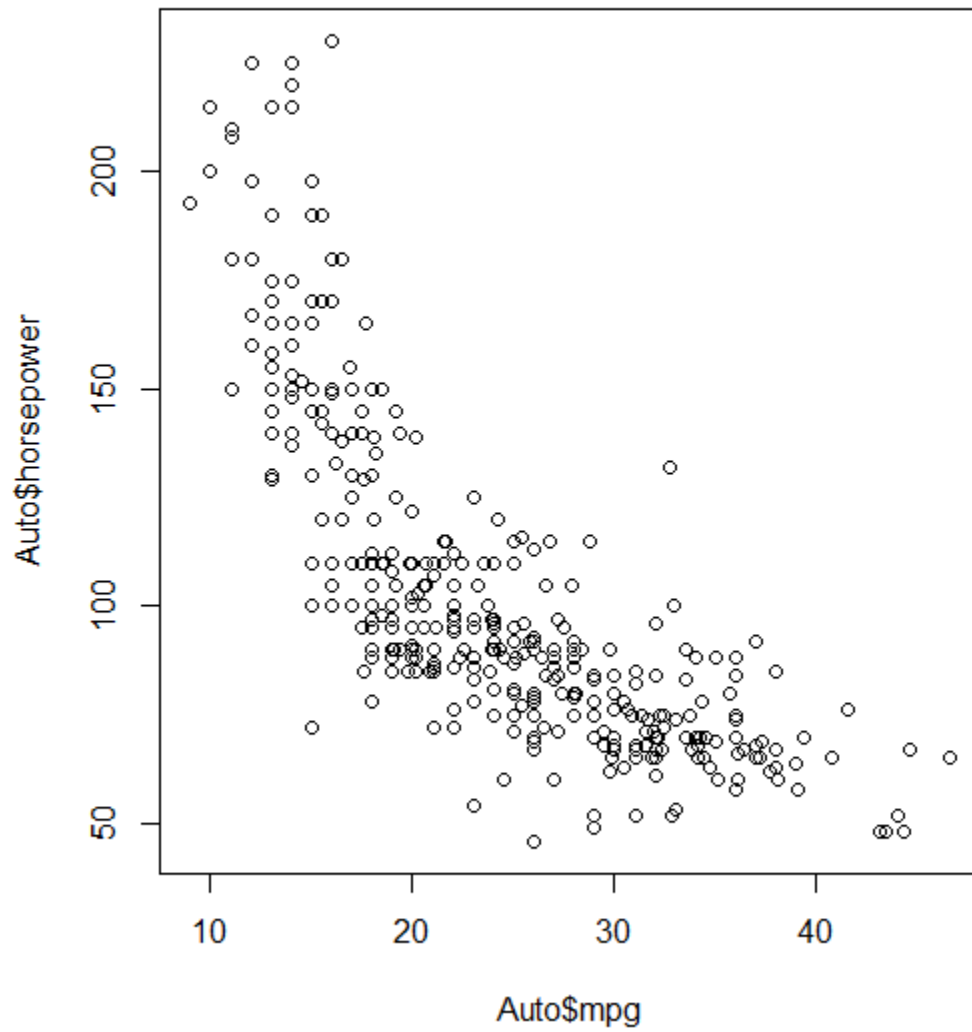


Some additional plots  
Mpg vs displacement



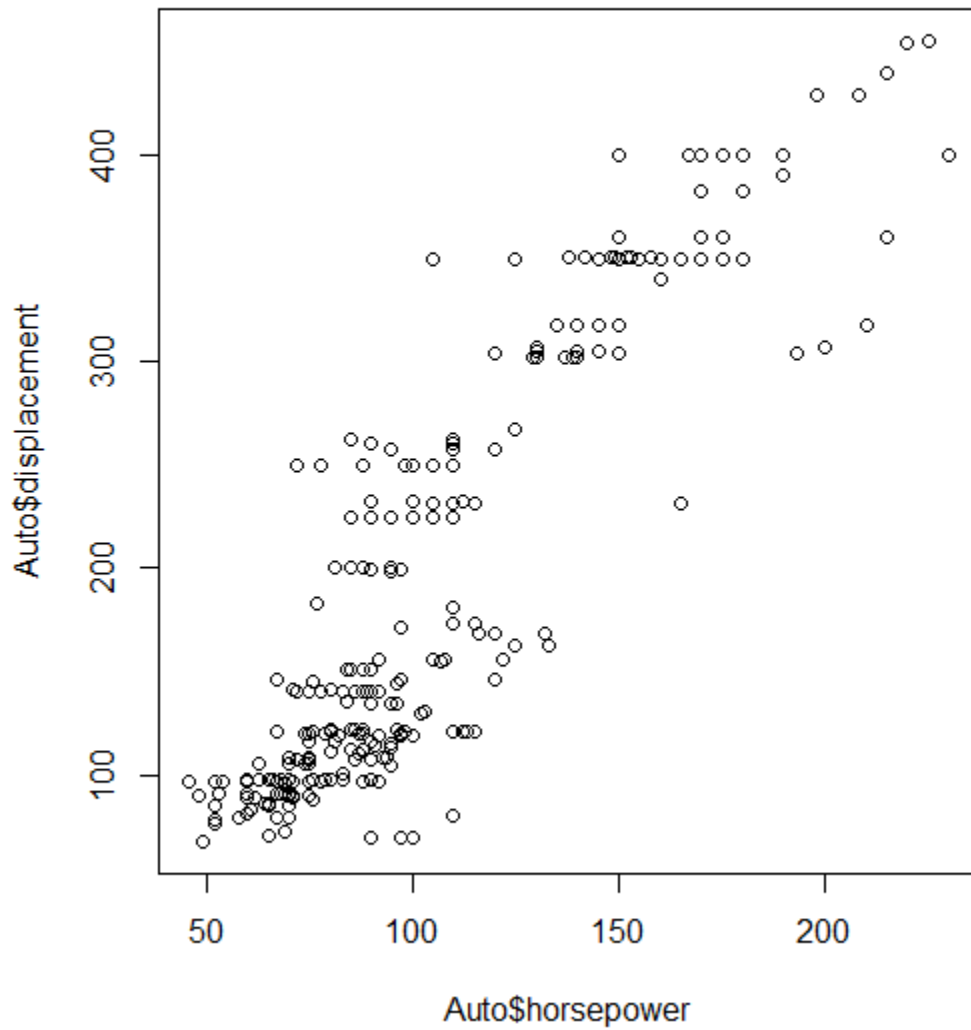
Showing decreasing trend which means negatively correlated.

Mpg vs horsepower



Showing decreasing trend which means negatively correlated.

Horsepower vs displacement

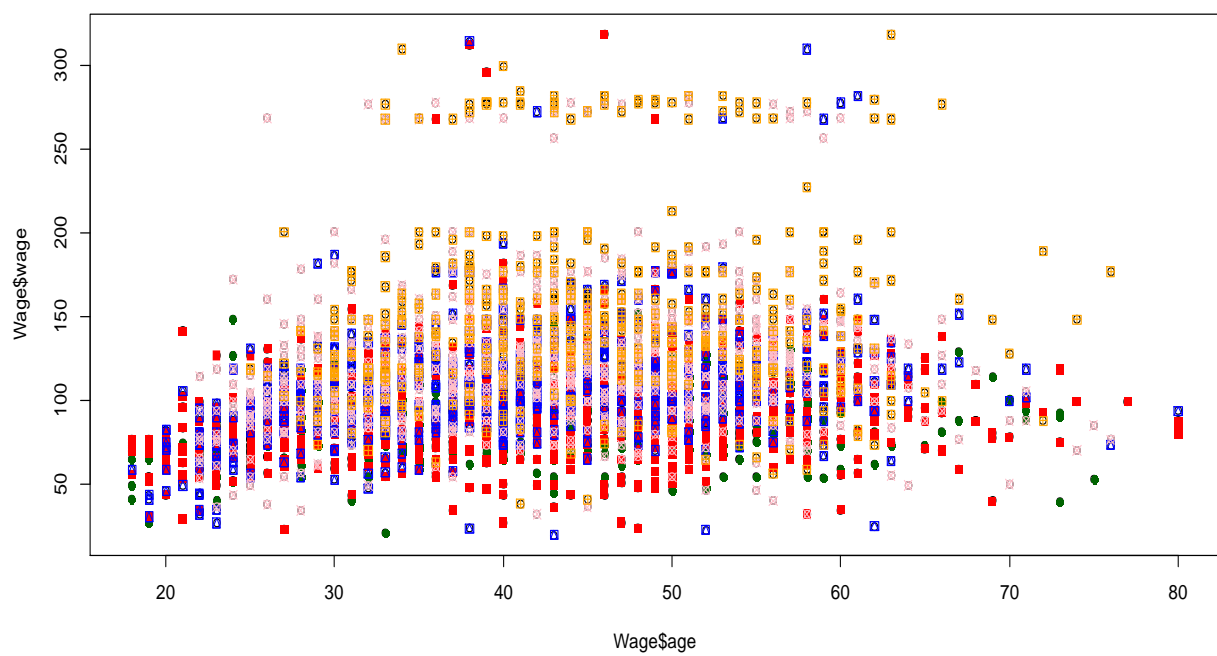


From this plot we can see a strong linear increasing trend in between these two predictors. Positively correlated.

- f. Yes, year, acceleration, and origin would be good predictors of mpg.

## Problem 2

Solution: Scatter plot



Problem 3

Solution:

**1. < HS Grad 2. HS Grad 3. Some College 4. College Grad 5. Advanced Degree**

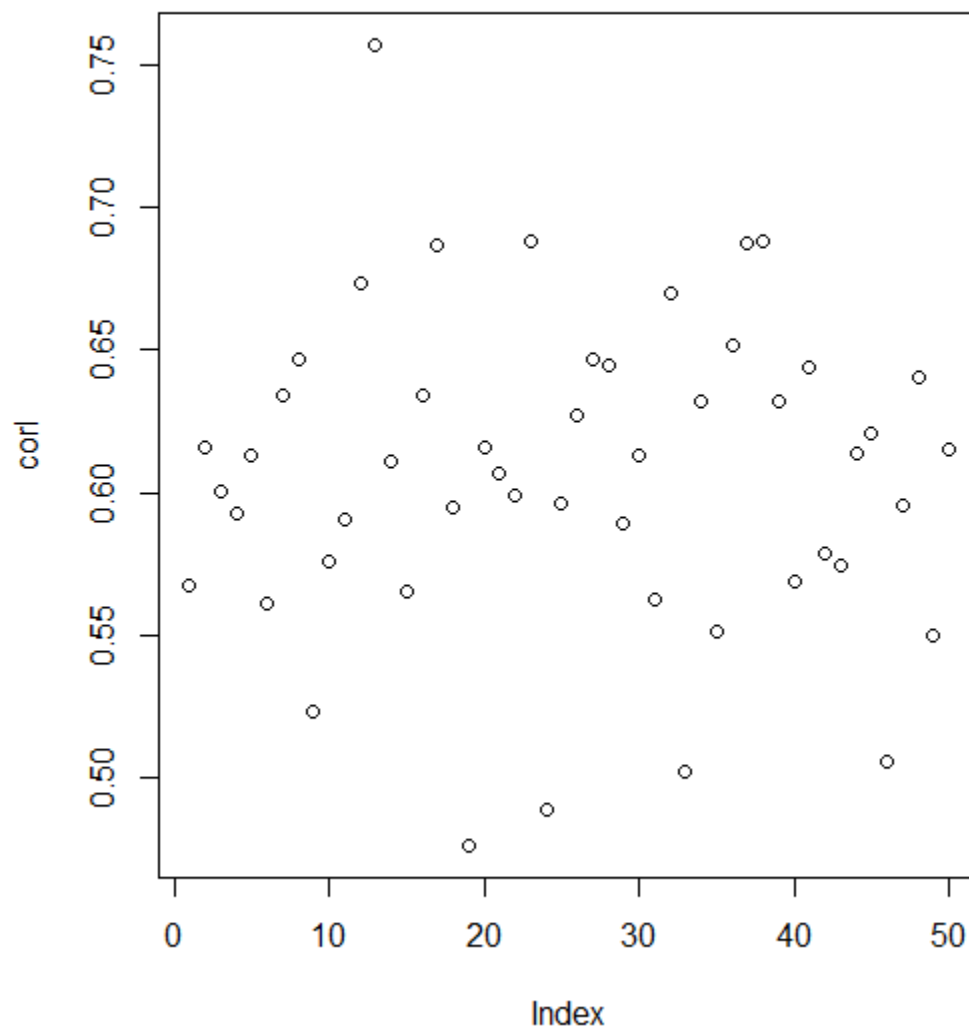
<b>1. Industrial</b>	<b>190</b>	<b>636</b>	<b>342</b>	<b>274</b>	<b>102</b>
<b>2. Information</b>	<b>78</b>	<b>335</b>	<b>308</b>	<b>411</b>	<b>324</b>

#### Problem 4

Solution:

Correlation: 0.594998

After repeating 50 times



Theoretical value should be approximately 0.60.



R-Code

#####Problem 1#####

```
Auto=read.csv ("Auto.csv")
```

```
fix(Auto)
```

```
Auto=read.csv ("Auto.csv", header =T,na.strings ="?")
```

```
fix(Auto)
```

```
dim(Auto)
```

```
Auto=na.omit(Auto)
```

```
dim(Auto)
```

```
names(Auto)
```

```
head(Auto)
```

```
attach(Auto)
```

```
sapply(Auto, class)
```

```
summary(Auto)
```

```
Auto$origin <- factor(Auto$origin, levels=1:3, labels=c("U.S.", "Europe", "Japan"))
```

```
sapply(Auto, class)
```

```
quantitative <- sapply(Auto, is.numeric)
```

```
quantitative
```

```
x=c("mpg"=range(mpg),"cylinders"=range(cylinders),"displacement"=range(displacement)
```

```
,"horsepower"=range(horsepower)
```

```
,"weight"=range(weight)
```

```
,"acceleration"=range(acceleration)
```

```
,"year"=range(Auto[,7]))
```

```
summary(Auto)
```

```
mpgms=c(mean=mean(mpg), standarddeviation=sd(mpg))
```

```
cylindersms=c(mean=mean(cylinders), standarddeviation=sd(cylinders))
```

```
displacementms=c(mean=mean(displacement), standarddeviation=sd(displacement))
```

```
horsepowerms=c(mean=mean(horsepower), standarddeviation=sd(horsepower))
```

```
weightms= c(mean=mean(weight), standarddeviation=sd(weight))
```

```
accelerationms=c(mean=mean(acceleration), standarddeviation=sd(acceleration))
```

```
yearms= c(mean=mean(Auto[,7]), standarddeviation=sd(Auto[,7]))
```

```
MeanandStad=c(mpgms,cylindersms,displacementms,horsepowerms,weightms,accelerationms,yearms)
```

```
MeanandStad
```

```
newmsd <- sapply(Auto[-10:-85, quantitative], function(x) round(c(range(x), mean(x), sd(x)), 2))
```

```
rownames(newmsd) <- c("min", "max", "mean", "sd")
```

```
newmsd
```

```
pairs(Auto)
```

```
plot(Auto$mpg ,Auto$displacement)
```

```
plot(Auto$mpg ,Auto$horsepower)
```

```
plot(Auto$horsepower ,Auto$displacement)
```

#####Problem 2#####

```
require(ISLR)
```

```
data(Wage)
```

```
head(Wage)
```

```
plot(Wage$age, Wage$wage)
```

```
levels(Wage$education)
```

```
gp1 = (Wage$education == "1. < HS Grad")
```

```
points(Wage[gp1,2], Wage[gp1,11], pch = 16, col="darkgreen")
```

```
gp2 = (Wage$education == "2. HS Grad")
```

```
points(Wage[gp2,2], Wage[gp2,11], pch = 15, col="red")
```

```
gp3 = (Wage$education == "3. Some College")
```

```
points(Wage[gp3,2], Wage[gp3,11], pch = 14, col="blue")
```

```
gp4 = (Wage$education == "4. College Grad" )
```

```
points(Wage[gp4,2], Wage[gp4,11], pch = 13, col="pink")
```

```
gp5 = (Wage$education == "5. Advanced Degree")
```

```
points(Wage[gp5,2], Wage[gp5,11], pch = 12, col="orange")
```

#####Problem 3#####

```
mytable=table(Wage$jobclass, Wage$education)
```

```
fable(mytable)
```

#####Problem 4#####

```
X=rnorm(100)
```

```
Y=rnorm(100)
```

```
var1= 2*X+Y
```

```
var2= 2*X-Y
```

```
corr=cor(var1,var2)
```

```
corr
```

#####2nd Part###

```
nrep=50
```

```
for(i in 1:nrep){
```

```
  X=rnorm(100)
```

```
  Y=rnorm(100)
```

```
  var1= 2*X+Y
```

```
  var2= 2*X-Y
```

```
  corl[i]=cor(var1,var2)
```

```
}
```

```
plot(corl)
```