

# Chapter 2 Problem 9

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The exercise involves the Auto data set studied in the lab. Make sure that the missing values have been removed from the data.

```
#Remove missing values from data
library(ISLR)
data=Auto
auto=na.omit(Auto)
```

a. Which of the predictors are quantitative, and which are qualitative?

All predictors except name and origin are quantitative. It's clear as to why that is from the summary of the Auto data.

```
summary(auto)
```

##	mpg	cylinders	displacement	horsepower
##	Min. : 9.00	Min. : 3.000	Min. : 68.0	Min. : 46.0
##	1st Qu.: 17.00	1st Qu.: 4.000	1st Qu.: 105.0	1st Qu.: 75.0
##	Median : 22.75	Median : 4.000	Median : 151.0	Median : 93.5
##	Mean : 23.45	Mean : 5.472	Mean : 194.4	Mean : 104.5
##	3rd Qu.: 29.00	3rd Qu.: 8.000	3rd Qu.: 275.8	3rd Qu.: 126.0
##	Max. : 46.60	Max. : 8.000	Max. : 455.0	Max. : 230.0
##				
##	weight	acceleration	year	origin
##	Min. : 1613	Min. : 8.00	Min. : 70.00	Min. : 1.000
##	1st Qu.: 2225	1st Qu.: 13.78	1st Qu.: 73.00	1st Qu.: 1.000
##	Median : 2804	Median : 15.50	Median : 76.00	Median : 1.000
##	Mean : 2978	Mean : 15.54	Mean : 75.98	Mean : 1.577
##	3rd Qu.: 3615	3rd Qu.: 17.02	3rd Qu.: 79.00	3rd Qu.: 2.000
##	Max. : 5140	Max. : 24.80	Max. : 82.00	Max. : 3.000
##				
##		name		
##	amc matador	: 5		
##	ford pinto	: 5		
##	toyota corolla	: 5		
##	amc gremlin	: 4		
##	amc hornet	: 4		
##	chevrolet chevette	: 4		
##	(Other)	: 365		

b. What is the range of each quantitative predictor?

```
sapply(auto[,1:7], range) #range
```

##	mpg	cylinders	displacement	horsepower	weight	acceleration	year
## [1,]	9.0	3	68	46	1613	8.0	70
## [2,]	46.6	8	455	230	5140	24.8	82

c. What is the mean and standard deviation of each quantitative predictor?

```
sapply(auto[,1:7], mean) #means
```

```
##      mpg      cylinders displacement  horsepower      weight
## 23.445918    5.471939    194.411990    104.469388 2977.584184
## acceleration      year
## 15.541327    75.979592
```

```
sapply(auto[,1:7], sd) #standard deviations
```

```
##      mpg      cylinders displacement  horsepower      weight
##  7.805007    1.705783    104.644004    38.491160   849.402560
## acceleration      year
##  2.758864    3.683737
```

d. Now, remove the 10th through 85th observations. What is the range, mean, and standard deviation of each predictor in the subset of the data that remains?

```
#Delete 10th to 85th observations
```

```
auto2=auto[-(10:85),]
```

```
sapply(auto2[,1:7], range) #range
```

```
##      mpg cylinders displacement horsepower weight acceleration year
## [1,] 11.0         3          68         46   1649          8.5   70
## [2,] 46.6         8         455        230   4997         24.8   82
```

```
sapply(auto2[,1:7], mean) #means
```

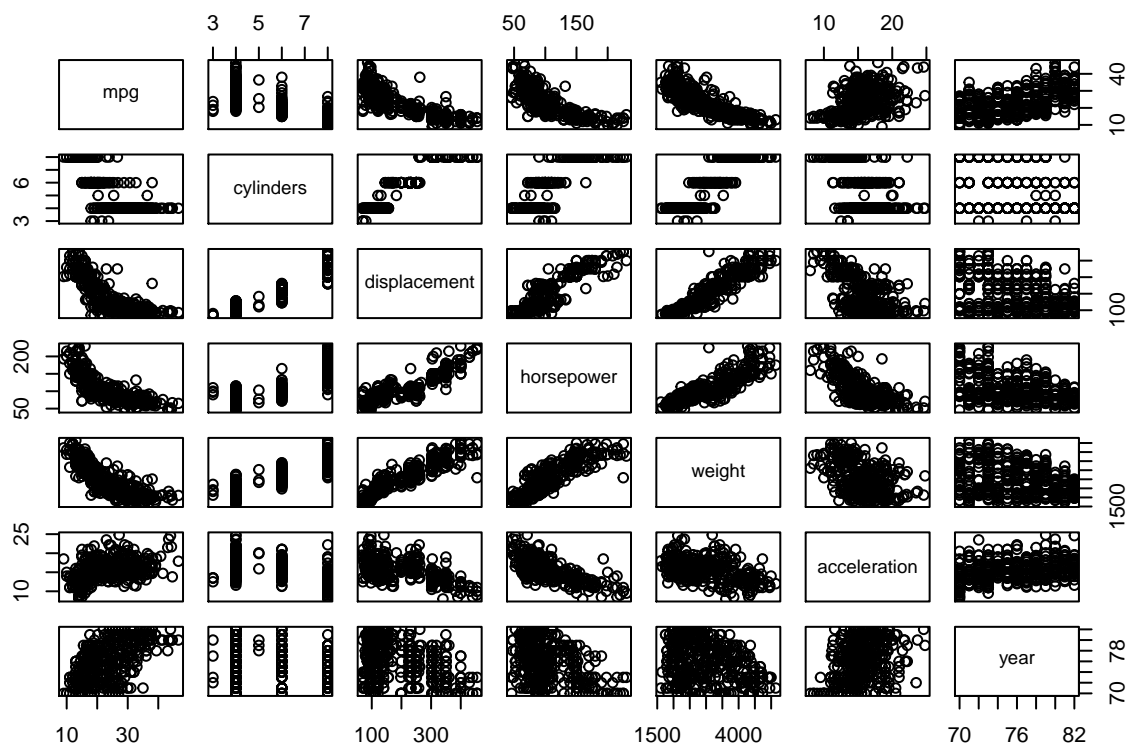
```
##      mpg      cylinders displacement  horsepower      weight
## 24.404430    5.373418    187.240506    100.721519 2935.971519
## acceleration      year
## 15.726899    77.145570
```

```
sapply(auto2[,1:7], sd) #standard deviations
```

```
##      mpg      cylinders displacement  horsepower      weight
##  7.867283    1.654179    99.678367    35.708853   811.300208
## acceleration      year
##  2.693721    3.106217
```

e. Using the full data set, investigate predictors graphically using scatterplots or other tools of your choice. Create some plots highlighting the relationships among the predictors. Comment on your findings.

```
pairs(auto[,1:7])
```



Positive correlations: mpg with years

Negative correlations: mpg with displacement, horsepower, weight

f. Suppose that we wish to predict the gas mileage based on other variables. Do your plots suggest that any of the other variables might be useful in predicting mpg? Justify your answer.

Yes, we were able to see relationships between mpg and other predictors (see above).