2.3 Lab: Introduction to R

Code **▼**

Hide

comment

To create a vector of numbers, use the function $\mathbf{c}(\mathbf{0})$ - c for *concatenate*. Any numbers inside the parentheses are joined together.

Hide

[1] 1 3 2 5

We can also save things using = rather than <-

Hide

$$x = c(1,6,2)$$

x

[1] 1 6 2

Hide

$$y = c(1,4,3)$$

[1] 1 4 3

Can check vector length using the length() function. Can add vectors of the same length together.

Hide

length(x)

[1] 3

Hide

length(y)

[1] 3

```
x + y
```

```
[1] 2 10 5
```

Let's see what happens with vectors of different length.

Hide

```
z = c(1,2,3,4)
x + z
```

Warning in x + z: longer object length is not a multiple of shorter object length

```
[1] 2 8 5 5
```

The **Is()** function allows us to look at a list of all the objects, such as data and functions, that we have saved so far. The **rm()** function can be used to delete any that we don't want.

```
ls()
```

```
[1] "A" "Auto" "cylinders" "f" "fa"
[6] "fh" "fh_in" "fh_out" "x" "y"
[11] "z"
```

```
A
Auto
cylinders

f
fa
fh
fh
fh_in
fh_out
x
y
```

```
rm(x,y)
ls()
```

```
[1] "A" "Auto" "cylinders" "f" "fa" "fh" [7] "fh_in" "fh_out" "z"
```

```
A
Auto
cylinders

f
fa
fh
fh
out
z
```

It is also possible to remove all objects at once.

Hide

```
character(0)
```

The ? can be used as a help call on functions.

Hide

```
?matrix
```

rm(list=ls())

ls()

The **matrix()** function takes a number of inputs, but for now we focus on the first three: the data, the number of rows, and the number of columns.

```
x = matrix(data=c(1,2,3,4), nrow=2, ncol=2)
x
```

```
[,1] [,2]
 [1,]
          1
                3
           2
 [2,]
                                                                                                     Hide
 is.matrix(x)
 [1] TRUE
Notice that we can omit the keywords in this call because data, nrow, and ncol are also the first three arguments
to matrix.
                                                                                                     Hide
 y = matrix(c(1,2,3,4),2,2)
 y == x
       [,1] [,2]
 [1,] TRUE TRUE
 [2,] TRUE TRUE
byrow=True can be used to populate the matrix in order of the rows.
                                                                                                     Hide
 matrix(c(1,2,3,4),2,2,byrow=TRUE)
       [,1] [,2]
 [1,]
          1
                2
           3
 [2,]
sqrt() takes the square root. ^ is used for exponents. Notice how vector/matrix operations work very nicely here.
                                                                                                     Hide
 sqrt(x)
            [,1]
                      [,2]
 [1,] 1.000000 1.732051
 [2,] 1.414214 2.000000
                                                                                                     Hide
 x^2
       [,1] [,2]
                9
 [1,]
```

[2,]

16

rnorm() generates a vector of random normal variables, with first argument n the sample size. Each time we call this function, we will (likely) get a different answer. Here we create two correlated sets of numbers, \mathbf{x} and \mathbf{y} , and use the **cor()** function to compute the correlation between them.

```
Hide x = rnorm(50)

y = x + rnorm(50, mean=50, sd=.1)

cor(x,y)

[1] 0.995529
```

By default, **rnorm()** creates standard normal random variables with a mean of 0 and a standard deviation of 1. However, the mean and standard deviation can be altered using the **mean** and **sd** arguments, as illustrated above. Sometimes, we want our code to reported the exact same set of random numbers; we can use the **set.seed()** function to do this. The **set.seed()** function takes an (arbitrary) integer argument.

```
Hide
set.seed(1303)
rnorm(50)
 [1] -1.1439763145 1.3421293656 2.1853904757
                                               0.5363925179 0.0631929665
     0.5022344825 -0.0004167247 0.5658198405 -0.5725226890 -1.1102250073
[11] -0.0486871234 -0.6956562176 0.8289174803
                                               0.2066528551 -0.2356745091
[16] -0.5563104914 -0.3647543571 0.8623550343 -0.6307715354 0.3136021252
[21] -0.9314953177 0.8238676185
                                 0.5233707021 0.7069214120
                                                             0.4202043256
[26] -0.2690521547 -1.5103172999 -0.6902124766 -0.1434719524 -1.0135274099
[31]
     1.5732737361 0.0127465055
                                 0.8726470499
                                               0.4220661905 -0.0188157917
[36]
     2.6157489689 -0.6931401748 -0.2663217810 -0.7206364412
                                                             1.3677342065
[41]
     0.2640073322  0.6321868074 -1.3306509858
                                               0.0268888182
                                                             1.0406363208
[46]
     1.3120237985 -0.0300020767 -0.2500257125
                                               0.0234144857
                                                             1.6598706557
```

We will use **set.seed()** throughout the labs whenever we perform calculations involving random quantities. In general this should allow the user to reproduce the same results.

The **mean()** and **var()** functions can be used to compute the mean and variance of a vector of numbers. Applying **sqrt()** to the output of **var()** will give the standard deviation. Or we can simply use the **sd()** function.

```
Hide

set.seed(3)
y=rnorm(100)
mean(y)

[1] 0.01103557

Hide

var(y)
```

```
[1] 0.7328675

Hide

sqrt(var(y))

[1] 0.8560768

Hide

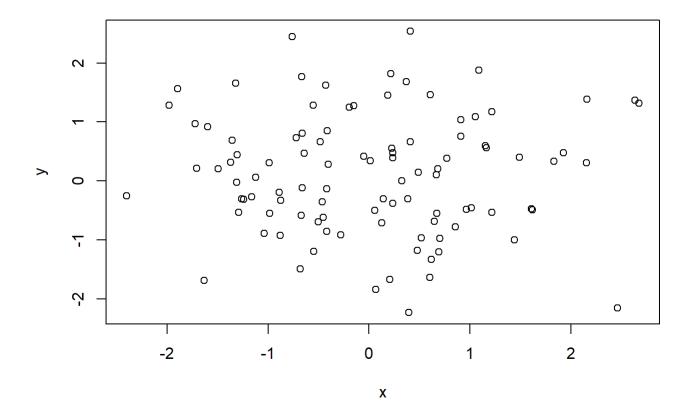
sd(y)

[1] 0.8560768
```

Graphics

The **plot()** function is the primary way to plot data in R. For instance, **plot(x,y)** produces a scatterplot of the numbers in **x** versus the numbers in **y**. There are many additional options that can be passed into the **plot()** function. For example, passing in the argument **xlab** will result in a label on the x-axis. To find out more information about the **plot()** function, type **?plot**.

| X=rnorm(100) y=rnorm(100) plot(x,y)

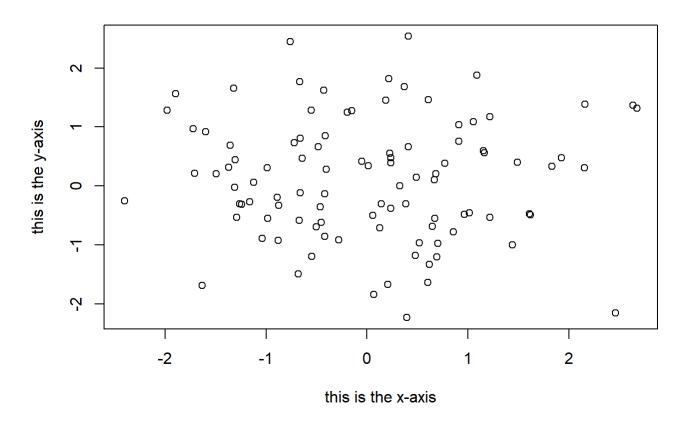


With some keywords...

Hide

 $plot(x,y,xlab='this\ is\ the\ x-axis',\ ylab='this\ is\ the\ y-axis',\ main='Plot\ of\ Y\ vs\ X')$

Plot of Y vs X



We can use the **pdf()** function or **jpeg()** function to save a figure.

Call dev.off() to indicate to R that we are done creating the plot.

```
pdf("Figure.pdf")
plot(x,y,col='green')
dev.off()

png
2
```

seq() can be used to create a sequence of numbers. For instance, **seq(a,b)** makes a vector of integers between **a** and **b**. There are many other options: for instance, **sq(0,1,length=10)** makes a sequence of 10 numbers that are equally spaced between 0 and 1. Typing **3:11** is a shorthand for **seq(3,11)** for integer arguments.

```
Hide seq(1,10)

[1] 1 2 3 4 5 6 7 8 9 10
```

```
1:10
```

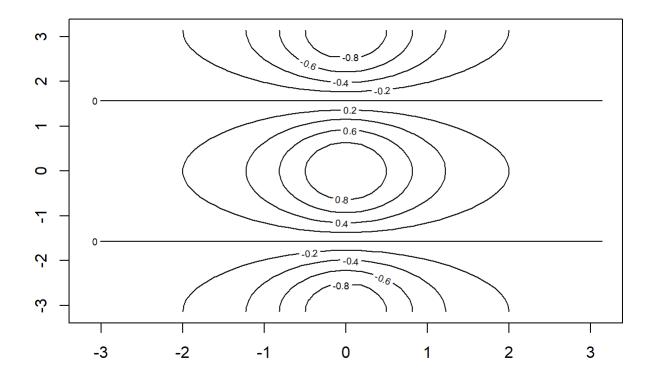
```
[1] 1 2 3 4 5 6 7 8 9 10
```

```
seq(-pi,pi,length=50)
```

```
[1] -3.14159265 -3.01336438 -2.88513611 -2.75690784 -2.62867957
[6] -2.50045130 -2.37222302 -2.24399475 -2.11576648 -1.98753821
[11] -1.85930994 -1.73108167 -1.60285339 -1.47462512 -1.34639685
[16] -1.21816858 -1.08994031 -0.96171204 -0.83348377 -0.70525549
[21] -0.57702722 -0.44879895 -0.32057068 -0.19234241 -0.06411414
[26]
     0.06411414 0.19234241 0.32057068
                                         0.44879895 0.57702722
[31]
     0.70525549  0.83348377  0.96171204
                                         1.08994031
                                                    1.21816858
[36]
     1.34639685
                 1.47462512 1.60285339
                                         1.73108167
                                                    1.85930994
[41]
     1.98753821
                2.11576648 2.24399475
                                         2.37222302
                                                    2.50045130
[46]
     2.62867957
                 2.75690784
                             2.88513611
                                         3.01336438
                                                    3.14159265
```

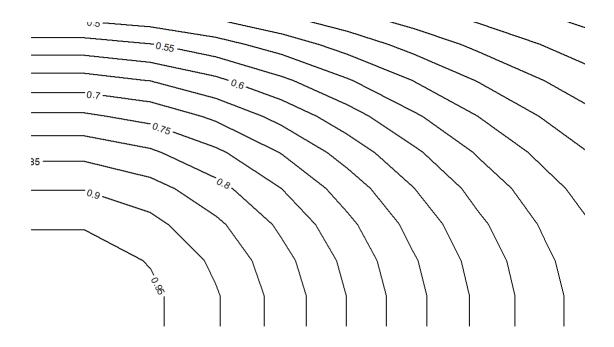
countour() produces a contour plot in order to represent three-dimensional data; it is like a topographical map. It takes three arguments: 1. A vector of the \mathbf{x} values (the first dimension), 2. A vector of the \mathbf{y} values (the second dimension), and 3. A matrix whose elements correspond to the \mathbf{z} values (the third dimension) for each pair of (\mathbf{x}, \mathbf{y}) coordinates.

```
x=y=seq(-pi,pi,length=50)
f=outer(x,y,function(x,y)cos(y)/(1+x^2))
contour(x,y,f)
```



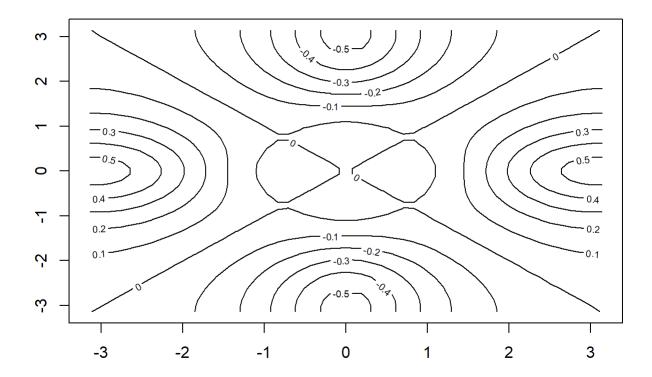
As with the **plot()** function, there are many other inputs that can be used to fine-tune the output of the **contour()** function.

```
plot.new()
contour(x,y,f,nlevels=45,add=T)
```



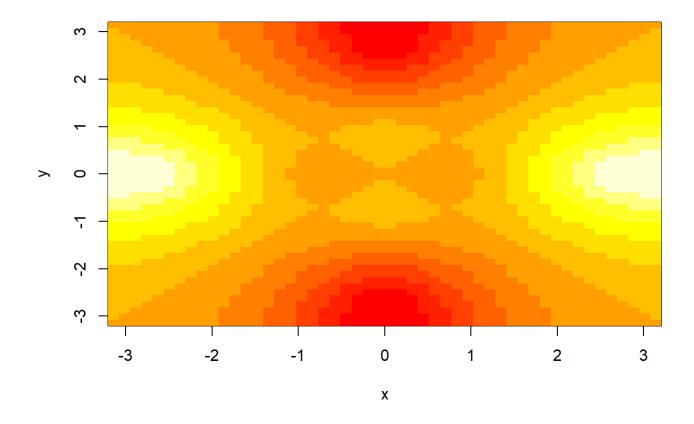
Something, something countour...

```
fa=(f-t(f))/2
contour(x,y,fa,nlevels=15)
```

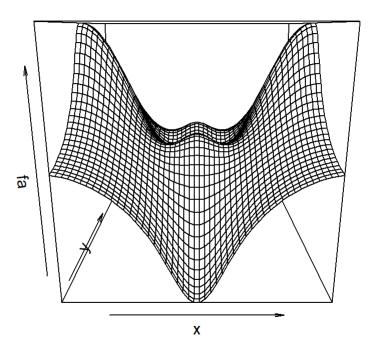


image() works the same way as contour(), except that it produces a color-coded plot whose colors depend on the z value. This is known as a heatmap, and is sometimes used to plot temperature in weather forecasts.
 Alternatively, persp() can be used to produce a three-dimensional plot. The arguments theta and phi control the angles at which the plot is viewed.

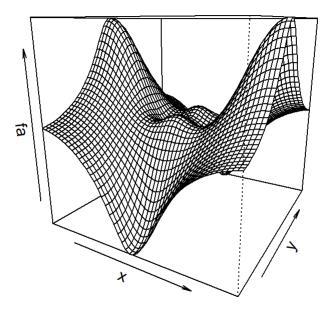
```
plot.new()
image(x,y,fa)
```



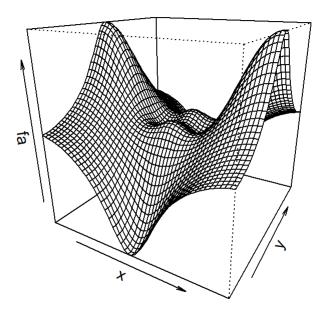
persp(x,y,fa)



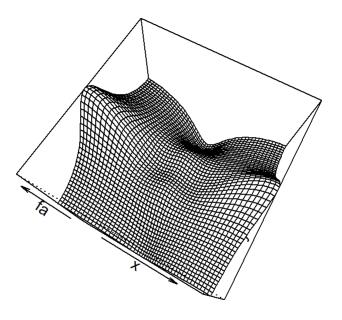
persp(x,y,fa,theta=30)



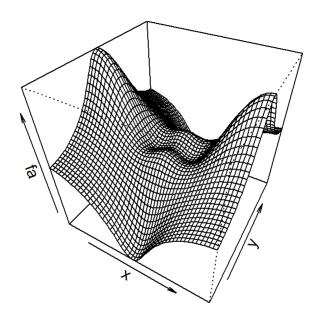
persp(x,y,fa,theta=30,phi=20)



persp(x,y,fa,theta=30,phi=70)



persp(x,y,fa,theta=30,phi=40)



Indexing Data

We often wish to examine part of a set of data. Suppose that our data is stored in the matrix A.

```
A = matrix(1:16,4,4)
A
```

```
[,1] [,2] [,3] [,4]
[1,]
                       13
        2
[2,]
             6
                 10
                       14
        3
             7
[3,]
                 11
                       15
[4,]
                 12
                       16
```

Then, by typing **A[2,3]**, we can select the element corresponding of the second row and the third column. The first number after the open-bracket symbol [always refers to the rows, and the second number always refers to the column.

Notice that indexing is one-based.

Hide

A[2,3]

```
[1] 10
```

We can also select multiple rows and columns at a time, by providing vecotrs as the indices.

Hide

```
A[c(1,3), c(2,4)]
```

```
[,1] [,2]
[1,] 5 13
[2,] 7 15
```

Hide

A[1:3,2:4]

```
[,1] [,2] [,3]
[1,] 5 9 13
[2,] 6 10 14
[3,] 7 11 15
```

Hide

A[1:2,]

```
[,1] [,2] [,3] [,4]
[1,] 1 5 9 13
[2,] 2 6 10 14
```

Hide

A[,1:2]

```
[,1] [,2]
[1,] 1 5
[2,] 2 6
[3,] 3 7
[4,] 4 8
```

The use of a negative sign - in the index tells R to keep all rows or columns except those indicated in the index.

Hide

A[-c(1,3),]

```
[,1] [,2] [,3] [,4]
[1,] 2 6 10 14
[2,] 4 8 12 16
```

```
A[-c(1,3), -c(1,3,4)]
```

```
[1] 6 8
```

The dim() function outputs the number of rows followed by the number of columns of a given matrix.

Hide

```
dim(A)
```

[1] 4 4

Loading Data

read.table() is used to import a data set into R. write.table() is used to export data.

fix() can be used to view data in a spreadsheet-like window. However, the indow must be closed before further R commands can be entered.

Hide

```
\label{thm:local_condition} Auto=read.table('D:\GoogleDrive\Introduction to Statistical Learning with Applications in R\data-sets\Auto.data') fix(Auto)
```

This data set has not been loaded correctly, because R has assumed that the variable names are part of the data and so has included them in the first row. The data set also includes a number of missing ovservations, indicated by a question mark? Missing values are a common occurrence in real data sets. Using the option **header=T** (or **header=TRUE**) in the **read.table()** function tells R that the first line of the file contains the variable names, and using the options **na.strings** tells R that any time it sees a particular character or set of characters (such as a question mark), it should be treated as a missing element of the data matrix.

Hide

```
fh='D:\\GoogleDrive\\Introduction to Statistical Learning with Applications in R\\data-s
ets\\Auto.data'
Auto=read.table(fh, header=T, na.strings='?')
fix(Auto)
```

Excel is a common-format data storage program. An easy way to load such data into R is to svae it as a csv (comma separated value) file and then use the **read.csv()** function to load it in.

```
fh='D:\\GoogleDrive\\Introduction to Statistical Learning with Applications in R\\data-s
ets\\Auto.csv'
Auto=read.csv(fh, header=T, na.strings="?")
dim(Auto)
```

```
[1] 397 9
```

We can use the **na.omit()** function to remove all rows with missing data.

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

[1] 392 9
```

We can use names() to check the variable names of the data.

names(Auto)

[1] "mpg" "cylinders" "displacement" "horsepower"
[5] "weight" "acceleration" "year" "origin"
[9] "name"

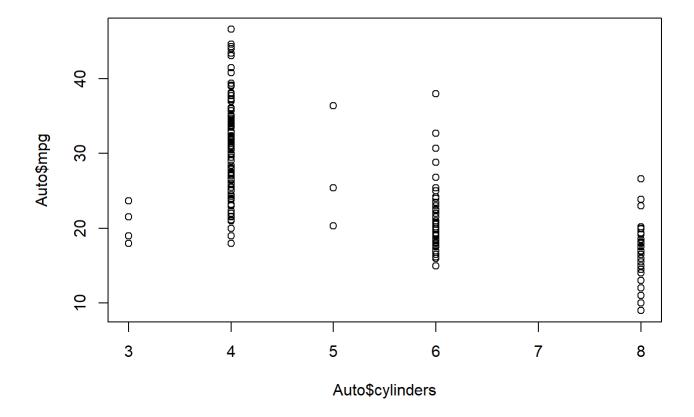
mpg
cylinders
displacement
horsepower
weight
acceleration
year
origin
name

Additional Graphical and Numerical Summaries

To refer to a variable in a data set, we must type the data set and the variable name joined with a \$ symbol. Alternatively, we can use the **attach()** function in order to tell R to make the variables in this data frame available by name.

Hide

plot(Auto\$cylinders, Auto\$mpg)



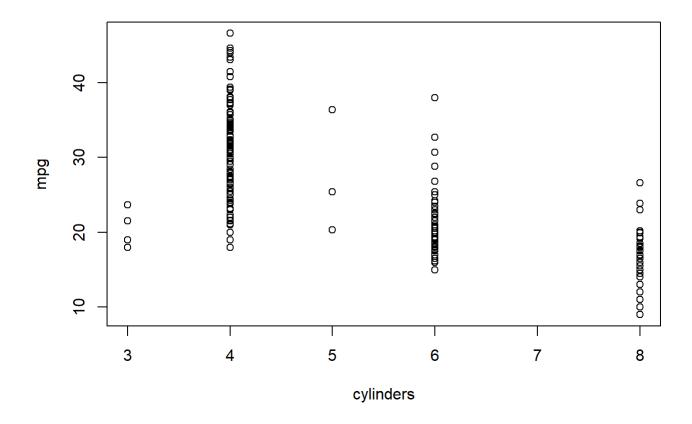
attach(Auto)

The following objects are masked from Auto (pos = 4):

acceleration, cylinders, displacement, horsepower, mpg, name, origin, weight, year

Hide

plot(cylinders, mpg)

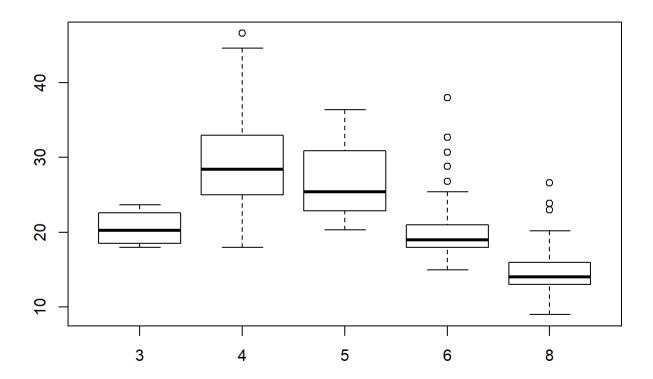


cylinders is stored as anumeric vector, so R has treated it as quantitative. However, since there are only a small number of possible values for **cylinders**, it may be better to treat it qualitatively. The **as.factor()** function converts quantitative variables into qualitative ones.

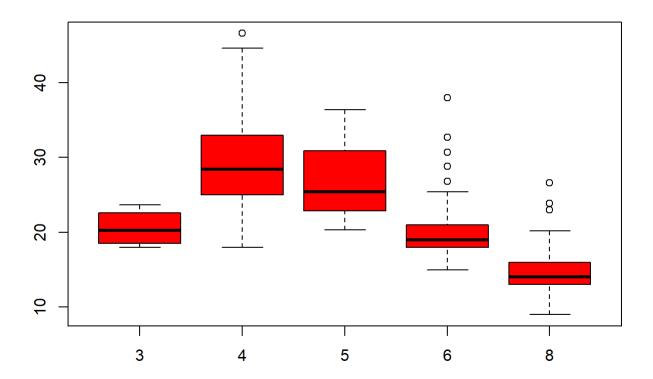
```
cylinders=as.factor(cylinders)
```

If the variable plotted on the *x*-axis is categorical, then *boxplots* will automatically be produced by the **plot()** function. As usual, a number of options can be specified in order to customize the plots.

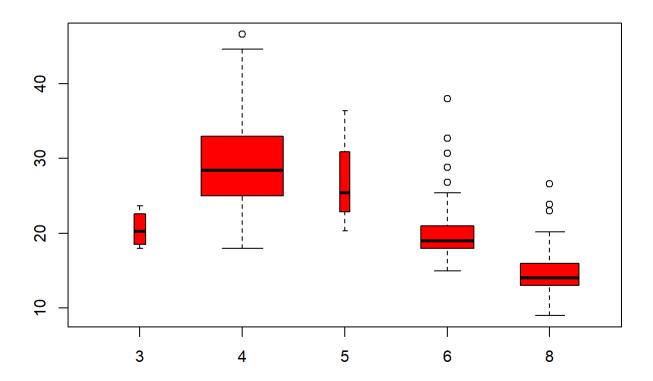
```
plot(cylinders, mpg)
```



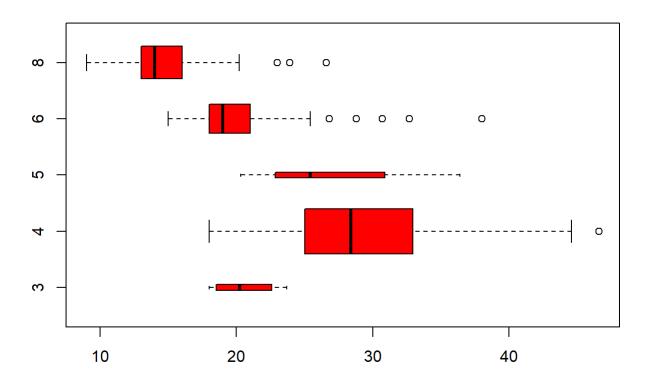
plot(cylinders, mpg, col='red')



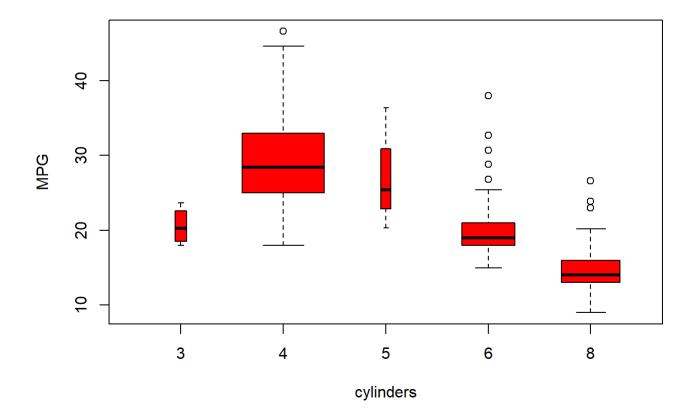
plot(cylinders, mpg, col='red', varwidth=T)



plot(cylinders, mpg, col='red', varwidth=T, horizontal=T)



plot(cylinders, mpg, col='red', varwidth=T, xlab='cylinders', ylab='MPG')

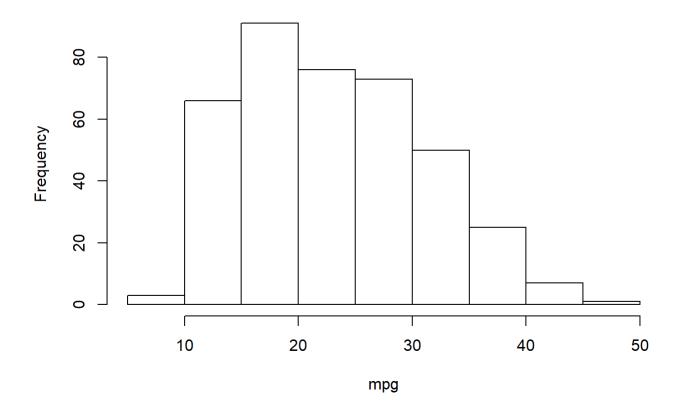


The hist() function can be used to plot a histogram. Note that col=2 has the same effect as col="red".

Hide

hist(mpg)

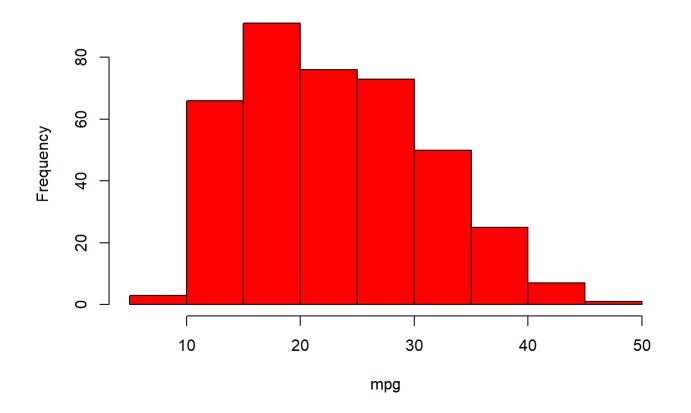
Histogram of mpg



Hide

hist(mpg, col=2)

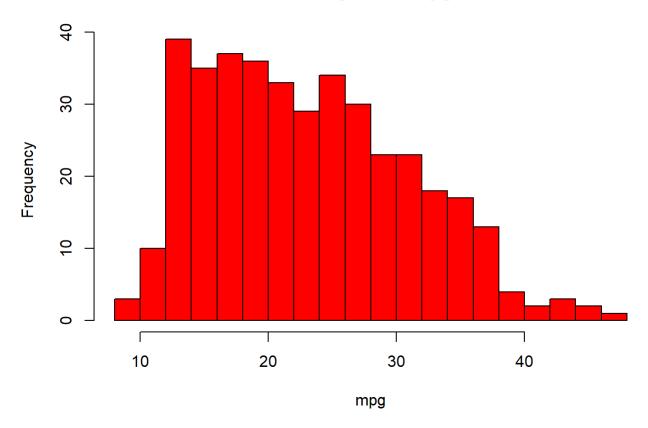
Histogram of mpg



Hide

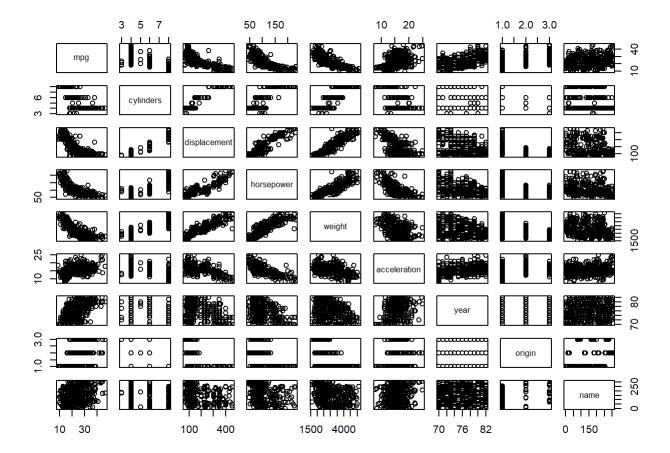
hist(mpg, col=2, breaks=15)

Histogram of mpg

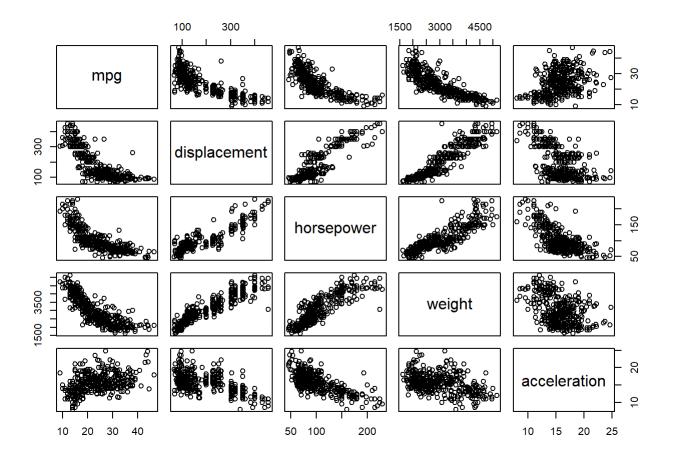


The **pairs()** function creates a *scatterplot matrix* i.e. a scatterplot for every pair of variables for any given data set. We can also produce scatterplots for just a subset of the variables.

pairs(Auto)



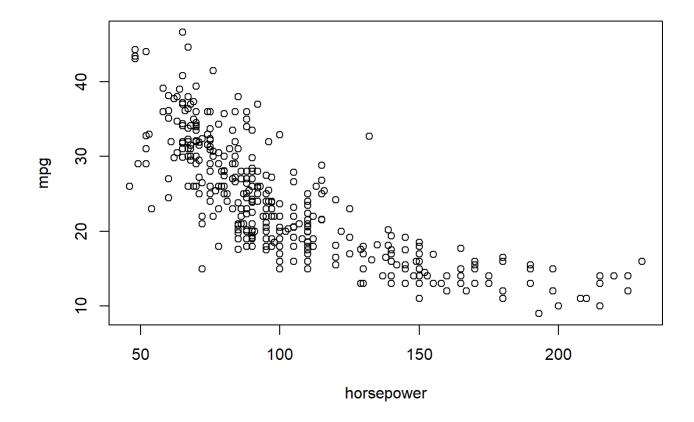
pairs(~ mpg + displacement + horsepower + weight + acceleration, Auto)



In conjuction with the **plot()** function, **identify()** provides a useful interactive method for identifying the value for a particular variable for points on a plot. We pass in three arguments to **identify()**: the *x*-axis variable, the*y*-axis variable, and the variable whose values we would like to see printed for each point. Then clicking on a given point in the plot will cause R to print the value of the variable of interset. Right-clicking on the plot will exist the **identify()** function. The numbers printed under the **indetify()** function correspond to the rows for the selected points.

Hide

plot(horsepower, mpg)
identify(horsepower, mpg, name)



integer(0)

The summary() function produces a numerical summary of each variable in a particular data set.

Hide

summary(Auto)

```
cylinders
                                  displacement
                                                    horsepower
     mpg
Min. : 9.00
                                        : 68.0
                                                         : 46.0
                Min.
                        :3.000
                                 Min.
                                                  Min.
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.:275.8
                3rd Qu.:8.000
                                                  3rd Qu.:126.0
Max.
       :46.60
                Max.
                        :8.000
                                 Max.
                                        :455.0
                                                  Max.
                                                         :230.0
    weight
                acceleration
                                                     origin
                                     year
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
       :2978
                       :15.54
                                       :75.98
                                                 Mean :1.577
Mean
               Mean
                                Mean
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
                      5
ford pinto
toyota corolla
                      5
                      4
amc gremlin
amc hornet
                      4
                     4
chevrolet chevette:
(Other)
                   :365
```

For qualitative variables such as **name**, R will list the number of observations that fall in each category. We can also produce a summary of just a single variable.

```
Hide

summary(mpg)

Min. 1st Qu. Median Mean 3rd Qu. Max.
9.00 17.00 22.75 23.45 29.00 46.60
```