BIOS 545 Week 3, Lecture 2

Department of Biostatistics and Bioinformatics

Steve Pittard wsp@emory.edu

September 8, 2016

- Merging data frames is possible
- But we don't often encounter two or more similar enough data frames that we can merge them easily.
- Usually we pick and choose columns from a number of data frames to build a new data frame.
- But this isn't easy either since not all observations are from the same population.
- Always work with the minimum amount of data necessary for a given analysis
- There is no reason to create a big data frame if you are only going to be using 20 percent of the attributes/columns.

Merging data frames is possible. You have to select a "key" that is common to both data frames to make this work. Pick a column(s) that links the two data frames

```
df1 \leftarrow data.frame(indiv_id = 1:4, snp1 = c(1,1,0,1), snp2 = c(1,1,0,0))
df2 \leftarrow data.frame(indiv_id = c(1,3,4,6), cov1 = c(1.14,4.50,0.80,1.39),
                                              cov2 = c(74.6, 79.4, 48.2, 68.1))
df1
  indiv_id snp1 snp2
        3
df2
  indiv_id cov1 cov2
        1 1.14 74.6
        3 4.50 79.4
        4 0.80 48.2
        6 1.39 68.1
```

The "indiv_id" column looks like it will do the trick

```
df1
```

```
indiv_id snp1 snp2
        2
        3
              1
df2
  indiv_id cov1 cov2
        1 1.14 74.6
        3 4.50 79.4
        4 0.80 48.2
        6 1.39 68.1
merge(df1, df2, by="indiv_id", all=TRUE)
  indiv_id SNP1 SNP2 cov1 cov2
                   1 1.14 74.6
        1
                       NA
                            NA
        3
                   0 4.50 79.4
```

0 0.80 48.2

NA 1.39 68.1

NA

4

Pay attention to the "all.x" and "all.y" arguments as it helps you specify which records you want to the include

```
df1
  indiv_id snp1 snp2
        3
df2
  indiv_id cov1 cov2
        1 1.14 74.6
        3 4.50 79.4
        4 0.80 48.2
        6 1.39 68.1
merge(df1, df2, by="indiv_id", all.y=T)
```

Note that the merge columns do not have to be named the same thing in each data frame as long as they refer to the same thing

```
names(df2) <- c("id","cov1","cov2")
head(df1,2)
  indiv_id snp1 snp2
head(df2,2)
  id cov1 cov2
1 1 1 1 1 4 74 6
2 3 4.50 79.4
merge(df1,df2,by.x="indiv_id",by.y="id",all=TRUE)
  indiv_id snp1 snp2 cov1 cov2
1
                   1 1.14 74.6
                        NA
                             NA
3
                   0 4.50 79.4
                   0 0.80 48.2
5
             NA
                   NA 1.39 68.1
```

- The split function lets us break up a data frame based on a grouping variable
- Let's say we want to split up mtcars based on the number of cylinders which take on the values 4,6,8
- Use the split command which gives back a list with each element containing a part of the data frame corresponding to each cylinder group
- Without using split you could do:

```
eight.cyl <- mtcars[mtcars$cyl == 8,]
six.cyl <- mtcars[mtcars$cyl == 6, ]
four.cyl <- mtcars[mtcars$cyl == 4, ]</pre>
```

But what if we had 10 categories we wanted to split by ? The **split** function does scale well so use it !

```
hold <- split(mtcars, mtcars$cyl)
str(hold,max.level=1)
List of 3
$ 4: 'data_frame': 11 obs. of 11 variables:
$ 6: 'data.frame': 7 obs. of 11 variables:
$ 8: 'data frame': 14 obs. of 11 variables:
head(hold[[1]].3) # Show the first 3 lines of the first list element
           mpg cyl disp hp drat wt qsec vs am gear carb
Datsun 710 22.8 4 108.0 93 3.85 2.32 18.61 1 1
Merc 240D 24.4 4 146.7 62 3.69 3.19 20.00 1 0
Merc 230 22.8 4 140.8 95 3.92 3.15 22.90 1 0
```

Why is this useful? Well we might want to focus in on only the cars occupying a certain cylinder group while ignoring the rest. So if we wanted only the 8 cylinder cars:

```
hold <- split(mtcars, mtcars$cyl)
sapply(hold,nrow)
4 6 8
11 7 14
eight.cyl <- hold$`8`
# -OR-
eight.cyl <- hold[[3]]</pre>
```

Because what we get back is a list we can use lapply to look at the first few records of each element which is a data frame

```
hold <- split(mtcars, mtcars$cyl)

str(hold,max.level=1)

List of 3

$ 4:'data.frame': 11 obs. of 11 variables:
$ 6:'data.frame': 7 obs. of 11 variables:
$ 8:'data.frame': 14 obs. of 11 variables:
```

```
lapply(hold,head,3)
$`4`
          mpg cyl disp hp drat wt qsec vs am gear carb
Datsum 710 22.8 4 108.0 93 3.85 2.32 18.61 1 1
Merc 240D 24.4 4 146.7 62 3.69 3.19 20.00 1 0
Merc 230 22.8 4 140.8 95 3.92 3.15 22.90 1 0
$`6`
              mpg cyl disp hp drat wt qsec vs am gear carb
Mazda RX4
             21.0
                    6 160 110 3.90 2.620 16.46
Mazda RX4 Wag 21.0 6 160 110 3.90 2.875 17.02 0 1
Hornet 4 Drive 21.4
                      258 110 3.08 3.215 19.44 1
$`8`
                 mpg cyl disp hp drat wt qsec vs am gear carb
Hornet Sportabout 18.7
                      8 360.0 175 3.15 3.44 17.02 0
                14.3 8 360.0 245 3.21 3.57 15.84 0 0
                                                             4
Duster 360
```

Merc 450SE

16.4 8 275.8 180 3.07 4.07 17.40 0

3

We could write our own summary function. While it is an advanced idea at this point, it is good for you too see this kind of approach as it is common in R. This example gives the mean MPG for each cylinder group:

```
# We create our own function and apply it to each element of "hold"
```

```
hold <- split(mtcars,mtcars$cyl)
sapply(hold, function(x) mean(x$mpg))</pre>
```

```
4 6 8
26.66364 19.74286 15.10000
```

Data Frames - order/sort

Let's take a look at what the order command does. It returns the record/row numbers of the data frame from lowest MPG to highest. So record #15 must be the lowest MPG automobile in the set. And record #20 must have the highest MPG

```
order(mtcars$mpg)
[1] 15 16 24 7 17 31 14 23 22 29 12 13 11 6 5 10 25 30
[19] 1 2 4 32 21 3 9 8 27 26 19 28 18 20

mtcars[15,]

mpg cyl disp hp drat wt qsec vs am gear carb
Cadillac Fleetwood 10.4 8 472 205 2.93 5.25 17.98 0 0 3 4

mtcars[20,]

mpg cyl disp hp drat wt qsec vs am gear carb
Toyota Corolla 33.9 4 71.1 65 4.22 1.835 19.9 1 1 4 1
```

Data Frames - order/sort

Ordering and sorting data frames is an important technique

```
# sort by mpg (ascending)
newdata <- mtcars[order(mtcars$mpg),]</pre>
```

```
        mpg cyl
        disp hp drat
        wt qsec vs am gear carb

        Cadillac Fleetwood
        10.4
        8 472.0
        205 2.93 5.250 17.98 0 0 0 3 4

        Lincoln Continental
        10.4
        8 460.0 215 3.00 5.424 17.82 0 0 0 3 4

        Camaro Z28
        13.3
        8 350.0 245 3.73 3.840 15.41 0 0 3 3 4

        Duster 360
        14.3
        8 360.0 245 3.21 3.570 15.84 0 0 3 3 4
```

newdata <- mtcars[rev(order(mtcars\$mpg)),]</pre>

	mpg	cyl	disp	hp	drat	wt	qsec	٧s	am	gear	carb
Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2

Data Frames - order/sort

Ordering and sorting data frames is an important technique

```
newdata <- mtcars[order(-mtcars$mpg),]</pre>
```

head(newdata)

```
mpg cyl disp hp drat wt qsec vs am gear carb
Toyota Corolla 33.9 4 71.1 65 4.22 1.835 19.90 1 1 4 1
Fiat 128 32.4 4 78.7 66 4.08 2.200 19.47 1 1 4 1
Honda Civic 30.4 4 75.7 52 4.93 1.615 18.52 1 1 4 2
Lotus Europa 30.4 4 95.1 113 3.77 1.513 16.90 1 1 5 2
```

Data Frames - sample

The **sample()** function is quite useful when you want to take, well, a sample of your data. You can sample with or without replacement. The basic function works as follows:

```
# Create a vector of numbers from 1 to 20
my_vec <- 1:20
sample(my_vec,10,replace=TRUE) # Repetition is possible
 [1] 3 20 16 14 16 10 18 7 7 6
sample(my_vec, 10, replace=TRUE) # Different results each time
 [1] 5 1 2 2 19 8 20 11 3 19
sample(my_vec, 10, replace=FALSE) # Don't replace sampled numbers
 [1] 2 8 9 6 17 18 3 5 14 15
sample(1:20, 10, replace=FALSE) # Short cut
 [1] 13 6 4 14 3 19 16 17 20 12
```

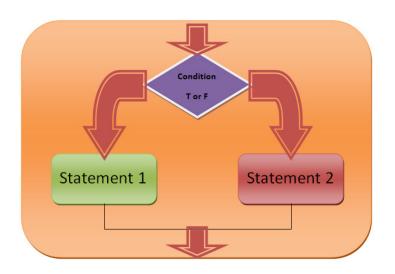
Data Frames - sample

But how do you sample from a data frame ? We want a random sample of 10 records from mtcars

```
my_records <- sample(1:nrow(mtcars), 10, replace = FALSE)
my_records
[1] 21 6 9 30 29 28 3 11 12 1
sample_of_ten <- mtcars[my_records,]</pre>
```

```
mpg cyl
                      disp hp drat wt qsec vs am gear carb
             21.5
                            97 3.70 2.465 20.01
Toyota Corona
                                                       3
Valiant.
             18.1
                    6 225.0 105 2.76 3.460 20.22 1
          22.8
                    4 140.8 95 3.92 3.150 22.90 1
Merc 230
Ferrari Dino 19.7
                    6 145.0 175 3.62 2.770 15.50 0 1
Ford Pantera L 15.8
                    8 351.0 264 4.22 3.170 14.50 0 1
Lotus Europa
             30.4
                       95.1 113 3.77 1.513 16.90 1
Datsun 710
             22.8
                    4 108.0 93 3.85 2.320 18.61 1
Merc 280C
          17.8
                    6 167.6 123 3.92 3.440 18.90 1
                                                   0
Merc 450SE
             16.4
                    8 275.8 180 3.07 4.070 17.40
                                                   0
                                                       3
                                                            3
Mazda RX4
             21.0
                    6 160.0 110 3.90 2.620 16.46
```

Programming Structures



Programming Structures

Goals for this Session:

- Understand the for-loop structure and how to use it
- "Walk" through a vector while accumulating a sum, computing a product, or some other operation.
- "Walk" though a matrix by row, (or column), while accumulating a sum, computing a product or some other arithmetic operation.
- "Walk" through a data frame by row to compute something. Also process the results of a previous "split" operation.
- Understand the if statement and how to branch based on the value of a vector or matrix element.
- Also use the if statement in conjunction with the for loop to do some processing.

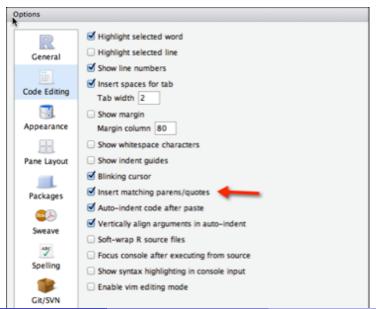
Programming Structures

Some things to keep in mind for getting better with programming structures

- Put the statements in the Editor window of RStudio and perfect them there. You can highlight sections of code and hit the "run" button
- You will most definitely make mistakes when writing loops. It is guaranteed. Better to get familiar with the most common mistakes ahead of time
- Work through the labs!
- The next assignment will assume facility with these structures

```
Run 🕪 Dource 🕶 🗐
      # Compute the grades
   3 score = c(74,68,98,90,100,67,59)
   4 for (ii in 1:length(score)) {
      if (score[ii] == 100) {
       arade = "A+"
       } else if (score[ii] >= 90) {
           grade = "A"
       } else if (score[ii] >= 80) {
  10
            grade = "B"
  11  } else if (score[ii] >= 70) {
  12
           grade = "C"
  13 -
        } else if (score[ii] >= 60) {
            arade = "D"
  14
  15
  16 -
       else {
          grade = "F"
  17
  18
  19
        print(grade)
  20
  21
 11:32
      (Top Level) $
                                                          R Script $
```

Go to Preferences -> Code Editing to turn on "insert matching parens/braces"



September 8, 2016

This is a looping construct that let's you do some things for a specific number of times.

"name" is some index variable that takes on values returned by "expr_1", which is almost always some type of sequence. It could represent the length of a vector or a row of a matrix.

```
for (name in expr_1) {
    expr_2
}

for (ii in 1:3) {
   print(ii)
}

[1] 1
[1] 2
[1] 3
```

Better to generalize this - use the **length()** function so the loop will work with a vector of any size

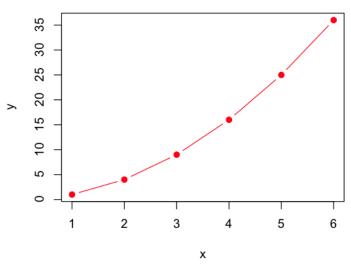
```
x <- rnorm(3)
for (ii in 1:length(x)) {
 print(ii)
}
[1] 1
[1] 2
Γ1 3
# Here we access the actual values of x
x <- rnorm(3)
for (ii in 1:length(x)) {
   print(x[ii])
[1] -0.6257547
   0.39524
[1] -1.955594
```

Consider the example wherein we have a \times values that we want to provide as input into some function that will generate y values.

```
y <- vector()  # A blank vector
x <- 1:6
for (ii in 1:length(x)) {
   y[ii] <- x[ii]^2
}

x
[1] 1 2 3 4 5 6
y
[1] 1 4 9 16 25 36
plot(x,y,main="Super Cool Data Plot",type="b",pch=19,col="red")</pre>
```

Super Cool Data Plot



Here we use a for-loop to add up the elements in a vector and find the average. There are already functions in R to do this but knowing how to do it yourself is important

```
x < -rnorm(1000, 20, 4) + 1,000 random elements from a N(20, 4)
mysum <- 0
for (ii in 1:length(x)) {
  mysum <- mysum + x[ii]</pre>
avg <- mysum / length(x)</pre>
cat("The average of this vector is: ",avg, "\n")
[1] "The average of this vector is: 20.1320691898645"
We could clean up the output a little bit
cat("The average of this vector is:",round(avg,2),"\n")
[1] "The average of this vector is: 20.13"
```

Given a vector find the smallest value without using the "min" function:

```
set.seed(188)
x \leftarrow rnorm(1000) + 1,000 random elements from a N(20,4)
# Set the min to the first element of x. Unless we are very lucky then
# this will change as we walk through the vector
mymin \leftarrow x[1]
for (ii in 1:length(x)) {
  if (x[ii] < mymin) {</pre>
     mymin <- x[ii]</pre>
mymin
[1] -3.422185
min(mymin)
                  # The internal R function matches what we got
[1] -3.422185
```

We can loop through data frames also. Let's see if we can compute the mean of the MPG for all cars. Note that we use the **nrow** function to get the number of rows to loop over

```
mpgsum <- 0
for (ii in 1:nrow(mtcars)) {
   mpgsum <- mpgsum + mtcars[ii,"mpg"]
}

mpgmean <- mpgsum/nrow(mtcars)  # Divide the sum by the # of records

cat("Mean MPG for all cars is:",mpgmean,"\n")
Mean MPG for all cars is: 20.09062

mean(mtcars$mpg)
[1] 20.09062</pre>
```

Remember the split command ? We can work with the output of that also. Relative to mtcars we let's split up the data frame by cylinder number, which is (4,6, or 8)

```
mysplits <- split(mtcars, mtcars$cyl)
str(mysplits, max.level=1)
List of 3
$ 4:'data.frame': 11 obs. of 11 variables:
$ 6:'data.frame': 7 obs. of 11 variables:
$ 8:'data.frame': 14 obs. of 11 variables:</pre>
```

We get back a list that contains 3 elements each of which has a data frame corresponding to the number of cylinders. We could summarize each of these data frame elements using a for loop

```
mysplits
$`4`
              mpg cyl disp hp drat wt gsec vs am gear carb
Merc 240D
             24.4
                    4 146.7 62 3.69 3.190 20.00 1
Merc 230
             22.8
                    4 140.8 95 3.92 3.150 22.90
Toyota Corona 21.5 4 120.1 97 3.70 2.465 20.01
. .
. .
$`6`
               mpg cyl disp hp drat wt qsec vs am gear carb
Hornet 4 Drive 21.4
                     6 258.0 110 3.08 3.215 19.44
Valiant
              18.1 6 225.0 105 2.76 3.460 20.22
Merc 280
              19.2 6 167.6 123 3.92 3.440 18.30
Merc 280C
             17.8 6 167.6 123 3.92 3.440 18.90
. .
$`8`
                    mpg cyl disp hp drat wt qsec vs am gear carb
Hornet Sportabout
                   18.7
                          8 360.0 175 3.15 3.440 17.02
Duster 360
                   14.3
                          8 360.0 245 3.21 3.570 15.84
                                                                    4
Merc 450SE
                   16.4
                          8 275.8 180 3.07 4.070 17.40
Merc 450SL
                          8 275.8 180 3.07 3.730 17.60
                                                                    3
                   17.3
```

```
mysplit <- split(mtcars,mtcars$cyl)</pre>
for (ii in 1:length(mysplit)) {
   print(nrow(mysplit[[ii]]))
}
Γ1 11
[1] 7
[1] 14
# This is equivalent to
sapply(mysplit, nrow)
   6 8
11 7 14
```

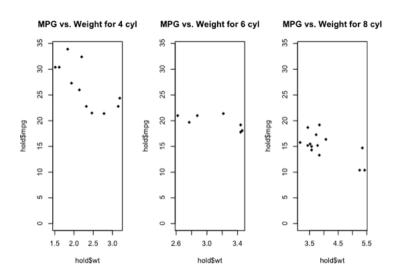
```
mysplit <- split(mtcars,mtcars$cyl)</pre>
for (ii in 1:length(mysplit)) {
   splitname <- names(mysplit[ii])</pre>
   cat("mean for", splitname, "cylinders is", mean(mysplit[[ii]] $mpg), "\n")
}
mean for 4 cylinders is 26.66364
mean for 6 cylinders is 19.74286
mean for 8 cylinders is 15.1
# This is basically equivalent to
sapply(mysplit, function(x) mean(x$mpg))
26.66364 19.74286 15.10000
```

What about looping over each split and pulling out only those cars with an manual transmission ? (am ==1)

```
data(mtcars)
mysplit <- split(mtcars,mtcars$cyl)</pre>
mylist <- list() # Setup a blank list to contain the subset results
for (ii in 1:length(mysplit)) {
  mylist[[ii]] <- subset(mysplit[[ii]], am == 1)</pre>
}
mylist
# Equivalent to:
lapply(mysplit, subset, am == 1)
```

Let's say we want to plot MPG vs. Weight for each cylinder group. Check it out:

```
mysplits <- split(mtcars, mtcars$cyl)</pre>
par(mfrow=c(1,3)) # This relates to plotting
for (ii in 1:length(mysplits)) {
  hold <- mysplits[[ii]]
  plot(hold$wt, hold$mpg, pch = 18, main=paste("MPG vs. Weight for",
       names(mysplits[ii]), "cyl",sep=" "),ylim=c(0,34))
}
NOTE:
names(mysplits[1])
[1] "4"
names(mysplits[2])
[1] "6"
names(mysplits[3])
[1] "8"
```



Programming Structures - Using for with matrices

The for loop structure generalizes to matrices. set.seed(123) mymat <- matrix(round(rnorm(6),2).3.2)</pre> for (ii in 1:nrow(mymat)) { cat("The sum of row", ii, "is", sum(mymat[ii,]), "\n") } The sum of row 1 is -0.49The sum of row 2 is -0.1

The sum of row 3 is 3.28

This is an easy structure. It tests a logical expression, which results in either or a TRUE or FALSE condition, and, based on that, executes a specific block of code

```
if (logical_expression) {
    do something
    ...
}
if (logical_expression) {
    do something
    ..
} else {
    do something else
    ...
}
```

```
Here is a basic example
(x < -3)
[1] 3
if (is.numeric(x)) {
   print("x is a number")
[1] "x is a number"
if (x != 3) {
     print("x is not equal to 3")
} else {
    print("guess what ? x is in fact equal to 3")
}
[1] "guess what ? x is in fact equal to 3"
```

Here is a more involved if statement that tests for several conditions. It uses the "else" keyword in addition to "if". Note that an "if" statement does not require an "else" statement but an "else" statement requires a "parent" if statement.

```
if (some.num < 3) {  # A more involved if statement
    print("Less than 3")
} else if (some.num > 3) {
    print("Greater than 3")
} else {
    print("Must be equal to 3")
}
[1] "Must be equal to 3"
```

if/else statements show up a lot in functions. Checking for valid arguments is a common practice.

```
x < -4
v <- 5
if (!is.numeric(x) | !is.numeric(y)) {
   stop("I need numeric values to do this")
} else {
   if (x == y) {
       print("Equal")
   } else {
      print("Not equal")
[1] "Not equal"
```

- R supports a command called ifelse that is desgined to work specifically on vectors.
- It works well for very large vectors. The format is **ifelse(test,yes,no)** where "test" is a logical expression to be evaluated.
- If it is TRUE then the action specified in the "yes" position will be executed. If the evaluated expression is FALSE then the action specified in the "no" position is executed.

```
some.data = rnorm(10000,0,2)
colors = ifelse(some.data < 0,"RED","GREEN")
plot(some.data,col=colors)

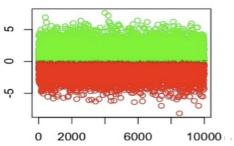
# This would be the same as:

for (ii in 1:length(some.data)) {
   if (some.data[ii] < 0) {
      colors[ii] = "RED"
   } else {
      colors[ii] = "GREEN"</pre>
```

```
some.data <- rnorm(10000,0,2)
colors <- ifelse(some.data < 0,"RED","GREEN")
plot(some.data,col=colors)

# This would be the same as:

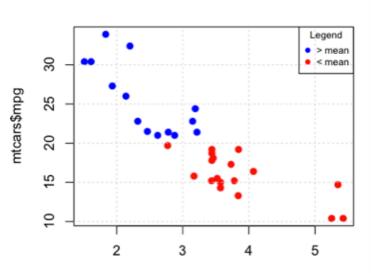
for (ii in 1:length(some.data)) {
    if (some.data[ii] < 0) {
        colors[ii] <- "RED"
    } else {
        colors[ii] <- "GREEN"
    }
}</pre>
```



We can use ifelse when we want to turn some continuous quantity within a data frame into a factor that we can then use to group by

```
mtcars$rating <- ifelse(mtcars$mpg >= mean(mtcars$mpg), "blue", "red")
head(mtcars)
                 mpg cyl disp hp drat wt qsec vs am gear carb rating
Mazda RX4
                21.0
                      6 160 110 3.90 2.620 16.46 0
                                                                blue
Mazda RX4 Wag
                21.0
                      6 160 110 3.90 2.875 17.02 0 1
                                                               blue
Datsun 710
            22.8
                      4 108 93 3.85 2.320 18.61 1 1
                                                               blue
Hornet 4 Drive
                21.4
                      6 258 110 3.08 3.215 19.44 1 0
                                                            1 blue
Hornet Sportabout 18.7
                      8 360 175 3.15 3.440 17.02 0 0
                                                            2 red
Valiant
                18.1
                      6 225 105 2.76 3.460 20.22 1 0
                                                            1 red
plot(mtcars$mpg~mtcars$wt,col=mtcars$rating,pch=19, main="MPG vs wt")
grid()
legend("topright", c("> mean", "< mean"), pch=19,</pre>
        col=c("blue", "red"), title="Legend", cex=0.7)
```





We can use ifelse when we want to turn some continuous quantity within a data frame into a factor that we can then use to group by

```
mtcars$rating <- ifelse(mtcars$mpg >= mean(mtcars$mpg), "blue", "red")
head(mtcars)
                 mpg cyl disp hp drat wt qsec vs am gear carb rating
Mazda RX4
                21.0
                      6 160 110 3.90 2.620 16.46 0
                                                                blue
Mazda RX4 Wag
                21.0
                      6 160 110 3.90 2.875 17.02 0 1
                                                                blue
Datsun 710
             22.8
                      4 108 93 3.85 2.320 18.61 1 1
                                                               blue
Hornet 4 Drive
                21.4
                      6 258 110 3.08 3.215 19.44 1 0
                                                            1 blue
Hornet Sportabout 18.7
                      8 360 175 3.15 3.440 17.02 0 0
                                                            2 red
Valiant
                18.1
                      6 225 105 2.76 3.460 20.22 1 0
                                                            1 red
plot(mtcars$mpg~mtcars$wt,col=mtcars$rating,pch=19, main="MPG vs wt")
grid()
legend("topright", c("> mean", "< mean"), pch=19,</pre>
        col=c("blue", "red"), title="Legend", cex=0.7)
```

```
score <- c(74,68,98,90,100,67,59) # Exam scores to be graded
for (ii in 1:length(score)) {
  if (score[ii] == 100) {
      grade <- "A+"
  } else if (score[ii] >= 90) {
      grade <- "A"
  } else if (score[ii] >= 80) {
      grade <- "B"
  } else if (score[ii] >= 70) {
      grade <- "C"
  } else if (score[ii] >= 60) {
      grade <- "D"
  else {
    grade <- "F"
  print(grade)
    "C"
[1]
    "D"
[1]
    " A "
    " A "
[1]
    "A+"
Γ17
    "D"
[1]
    "F"
```

```
set.seed(123)
x <- round(runif(9,1,20))
                                # Are the elements in x odd or even
[1] 6 16 9 18 19 2 11 18 11
for (ii in 1:length(x)) {
    if (x[ii] \%\% 2 == 0) {
        print(TRUE)
    else {
        print(FALSE)
   TRUE
[1] TRUE
[1] FALSE
[1] TRUE
[1] FALSE
[1] TRUE
[1] FALSE
[1] TRUE
[1] FALSE
```

This example mimics the bracket notation

```
set.seed(123)
x \leftarrow round(runif(10,1,20))
[1] 7 13 16 8 13 19 9 11 13 11
logvec <- vector()</pre>
                                # Setup an empty vector
for (ii in 1:length(x)) {
    if (x[ii] \% 2 == 0) {
        logvec[ii] <- TRUE</pre>
    }
    else {
        logvec[ii] <- FALSE</pre>
logvec
     TRUE
                                      TRUE FALSE
           TRUE FALSE
                        TRUE FALSE
                                                   TRUE FALSE
                                                                 TRUE
x[logvec]
[1] 6 16 18 2 18 10
```

One can easily "break" out of a for loop based on some condition. Normally you should clean your data before processing but perhaps not

Let's say that you are processing elements of a vector and if you encounter a value of NA then you want to stop the for loop

```
my.vec <- c(1,2,3,NA,5,6,7,8,9,10)

for (ii in 1:length(my.vec)) {
    if (is.na(my.vec[ii])) {
        break
    }
    cat("element is ",ii,"\n")
}

element is 1
element is 2
element is 3</pre>
```

Here we want to "catch" the the missing value and then "skip over it". To do this we would use the "next" statement.

```
my.vec \leftarrow c(1,2,3,NA,5,6,7,8,9,10)
for (ii in 1:length(my.vec)) {
    if (is.na(my.vec[ii])) {
       next
    }
    cat("element is ",ii,"\n")
}
element is
            10
```

Here is an example that will be useful when processing things like genetic sequences. Let's say we have a string of text we wish to "encode" by changing all vowels to something else. This would not be a tough code to break but let's see what is involved. In our code we:

```
We'll change a to s, e to t, i to u, o to v, u to w
```

So a string like:

```
sequence <- "Hello my name is Ed. Happy to meet you" would come out like:
```

"Htllv my nsmt us td. Hsppy tv mttt yvw"

```
sequence <- "Hello my name is Ed. Happy to meet you"
seq <- unlist(strsplit(sequence,""))</pre>
[1] "H" "e" "l" "l" "o" " " "m" "v" " " "n" "a" "m" "e" " "
    "i" "s" " "E" "d" "." " "H" "a" "p" "p" "v" " "t"
    "o" " "m" "e" "e" "t" " "v" "o" "u"
sequence <- "Hello my name is Ed. Happy to meet you"
seq < unlist(strsplit(sequence,""))</pre>
for (ii in 1:length(seq)) {
 # Write code to inspect each element of seq to determine if it is
 # a candidate for changing.
"Htllv my nsmt us Ed. Hsppy tv mttt yvw"
```

The **while** loop is similar to the for loop. We have some expression that must be evaluated until some condition is met. This is useful when we are writing code that must converge on something.

```
sum <- 0
n < -1000
i <- 1
while (i \le n) \{
    sim < - sim + i
    i < -i + 1
SIIM
Γ17 500500
    The following is equivalent
sum <- 0
n <- 1000
for (i in 1:n) {
   sum <- sum + 1
}
sum
```

Taking the square root of a number and then taking the square root of that result and so will eventually converge to 1. We can use a while loop to do this

```
num <- 13
sqrtval <- sqrt(num)</pre>
# Loop until the sqrt value becomes equal to 1
while (sqrtval != 1) {
  sgrtval <- sgrt(sgrtval)</pre>
# sprintf allows us to format a variable according to a pattern
# See http://www.cookbook-r.com/Strings/Creating_strings_from_variables/
 print(sprintf("%2.12f",sqrtval))
}
```