# R Programming: Worksheet 2

After today you should know:

- the differences between a data frame and a matrix;
- how to access parts of a data frame using subset();
- how to alter the elements of a data frame, including its row names and variable names;
- what a factor is, and how to create one from a character vector;
- how to get boxplots using factors.

## 1. Data Frames

Load the MASS library and take a look at the dataset called survey.

```
> library(MASS)
> head(survey)
```

You can look at the documentation:

```
> ?survey
```

and get a brief summary of each variable:

> summary(survey)

(a) Find the mean pulse rate of the students. What goes wrong here? There are missing values, so

```
> mean(survey$Pulse)
```

leads to NA.

The vector of pulses stored in survey contains some entries which are labelled NA. This is used in R to represent missing data.

(b) Try looking at the documentation for mean() to see how to get around this. Inspection of the documentation shows that you can get around this by using

```
> mean(survey$Pulse, na.rm = TRUE)
## [1] 74.15
```

which ignores missing values for the purposes of calculating the mean.

(c) The ages are recorded as fractions representing a number of months. Change that columns of the data frame so that it contains whole years (the floor() command may be useful).

> survey\$Age = floor(survey\$Age)

Another possibility is:

> survey = within(survey, Age <- floor(Age))</pre>

(d) Find the mean pulse rate for students under 20.

```
> with(survey, mean(Pulse[Age < 20], na.rm = TRUE))
## [1] 75.49</pre>
```

**Subsetting.** Suppose I want to obtain the records of students who are over 190 cm tall. Since data frames allow me to subset just like a matrix, I might just think of typing:

```
> survey[survey$Height > 190, ]
```

What goes wrong here? Every missing height value leads to an NA in the logical vector we're subsetting with. This leads (in each case) to a row of NAs in the data frame. Try instead the following:

> subset(survey, Height > 190)

The subset function ignores missing values, which is usually the behaviour we would prefer. We can also select only some of the fields of the data frame if we prefer:

> subset(survey, Height > 190, select = c("Pulse", "Clap"))

Recall that the & operator does a point-wise logical 'and' comparison.

```
> subset(survey, (Pulse > 70) & (Smoke == "Heavy"))
```

Similarly, | is for 'or', and ! for 'not'.

```
> with(survey, (Pulse > 70) | (Pulse < 45))
> !(survey$Age > 30)
```

(e) Find the mean age of students who write with their right hand. *Either of:* 

```
> mean(subset(survey, W.Hnd == "Right")$Age)
> mean(survey$Age[survey$W.Hnd == "Right"], na.rm = TRUE)
```

In the second case the **na**.**rm**=**TRUE** is necessary, even though no ages are missing.

(f) What proportion of left handers do not clap with their left hand on top?

```
> mean(subset(survey, W.Hnd == "Left", select = Clap) != "Left")
## [1] 0.5
> mean(survey$Clap[survey$W.Hnd == "Left"] != "Left", na.rm = TRUE)
## [1] 0.5
```

(g) Using the plot() command, plot the pulse of the subjects against their age.

```
> plot(survey$Age, survey$Pulse)
```

Try subtracting 10 from the age and taking the logarithm (using the function log()), to obtain a slightly clearer picture.

> with(survey, plot(log(Age - 10), Pulse))

#### 2. More Manipulation

Load the hills dataset from lectures. What happens when you apply the plot() command to this data? It produces pair-wise scatter plots of the three variables.

> plot(hills)

You can access the row names of the hills data with the function rownames(). You can also **change** the rownames this way:

```
> rownames(hills)[1] <- "Redmantle"
> rownames(hills)[1]
```

If at any point you want to revert to the original data frame, type

```
> rm(hills)
```

You can't really edit the original data set in the MASS package, only your own copy. Calling rm() deletes the copy.

(a) The race climbs are measured in feet. Change the data set so that they are measured in metres (there are 2.54 cm in 1 inch, and 12 inches in 1 foot). Use the round() command to set them to the nearest 10 metres.

```
> x = 100/(12 * 2.54) # feet in a meter
> x
## [1] 3.281
> hills$climb = round(hills$climb/x, -1)
```

(b) Obtain a logical vector indicating which races were longer than an hour. How many are this long?

```
> hills$time > 60
##
   [1] FALSE FALSE FALSE FALSE
                                TRUE
                                      TRUE TRUE FALSE FALSE FALSE
                                                                     TRUE
## [12] FALSE TRUE FALSE FALSE
                                TRUE
                                      TRUE
                                            TRUE FALSE FALSE FALSE FALSE
## [23] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
                                                        TRUE FALSE
                                                                     TRUE
## [34] FALSE TRUE
> sum(hills$time > 60)
## [1] 11
```

The which() command is a useful way to turn a logical vector into a numerical one. For example, try

```
> which(hills$dist < 5)
## [1] 1 15 18 19 21 22 24 25 26 34</pre>
```

(c) Use which() to select the first three races under 5 miles.

```
> hills[which(hills$dist < 5)[1:3], ]
## dist climb time
## Greenmantle 2.5 200 16.08
## Eildon Two 4.5 460 26.93
## Knock Hill 3.0 110 78.65</pre>
```

(d) Identify (with an R command, not just by looking) which of the races involves "Meall Ant-Suidhe". I've no idea how to pronounce Meall Ant-Suidhe, so remove it from the data frame (this might not normally be considered good statistical practice).

```
> rm = which(rownames(hills) == "Meall Ant-Suidhe")
> hills = hills[-rm, ]
```

(e) Plot the race times against distance. Can you spot an outlier? (think carefully)

> plot(hills\$dist, hills\$time)

One of the very short races (5 miles) has a record time of over 78 minutes; closer inspection reveals no significant climbs.

(f) In fact there is a race whose time was recorded as one hour larger than it ought to have been. Correct this.

```
> which(hills$dist < 4 & hills$time > 50)
```

## [1] 18

> hills[18, ]

## dist climb time
## Knock Hill 3 110 78.65
> hills[18, 3] = hills[18, 3] - 60

You can turn the data frame into a matrix like this:

> hillsMat = as.matrix(hills)

- (g) Print the two objects hills and hillsMat by typing their names. Can you see any difference? Superficially they look the same. print.default() tells a different story.
- (h) Use is() to see that they are different. What happens if you use plot() on the matrix? Compare this behaviour with that of the command pairs(). plot() when applied to a numeric matrix just uses the first two columns. pairs() does the right thing regardless.
- (i) What happens if you try to turn the survey data into a matrix? Everything in a matrix has to be of the same type, so it makes everything into a string (character vector).

#### 3. Factors

(Extension of Exercise 3.5 from Lectures)

Take a look at the birthwt data from the MASS package.

- (a) How is race stored in these data? Is this sensible? It's stored as a numeric vector, which isn't sensible since these data are categorical.
- (b) Turn this into a factor with level names as indicated in the documentation. One way is

```
> birthwt$race <- factor(birthwt$race, labels = c("white", "black", "other"))</pre>
```

but you might also turn it into a character vector first:

> birthwt\$race <- as.factor(c("white", "black", "other")[birthwt\$race])</pre>

(c) Use the table() command to count the number of babies of each race.

```
> table(birthwt$race)
##
## white black other
## 96 26 67
```

(d) Try the following command:

> tab = with(birthwt, table(smoke, low))

What do the results in tab suggest? Comparing

```
> tab[1, 2]/sum(tab[1, ])
## [1] 0.2522
> tab[2, 2]/sum(tab[2, ])
## [1] 0.4054
```

it seems that a much higher proportion of babies whose mother smoked during pregnancy were of low birthweight (although the documentation doesn't specify how the data were collected, so we have to be careful what we conclude from this.)

### 4. Reading In Data

I have emailed you a dataset called **sprays.dat**. The data represent insect counts in agricultural experiments treated with different insecticides. Save the file onto your local drive.

(a) Read the data into R using the command

```
> dat = read.table("sprays.dat", header = TRUE)
```

Check that the first row is as you expect.

```
> head(dat)
##
     count spray
## 1
         10
                 Α
## 2
         7
                 А
## 3
         20
                 А
## 4
         14
                 А
## 5
         14
                 А
## 6
         12
                 А
```

What happens if you omit the header=TRUE part? The first row ends up being the variable names.

- (b) Look at the two variables in this new data frame; what types do they have?difference between them do you notice? The variable spray has automatically become a factor, but count is a numeric vector. Is their format appropriate? Yes, the factor is sensible for categorical data.
- (c) Find the mean number of insects for each different experimental unit.

```
> mean(dat$count[dat$spray == "A"])
## [1] 14.5
> mean(dat$count[dat$spray == "B"])
## [1] 15.33
> # ETC
```

(d) Look at the documentation for the command tapply(). Now repeat the previous question with a single command.

(e) Try using the command quantile() on the counts. Use tapply() to find the upper and lower quartiles of the data broken down by spray type.

```
> quantile(dat$count)
                       75%
                           100%
##
      0%
           25%
                 50%
         3.00 7.00 14.25 26.00
##
    0.00
> tapply(dat$count, dat$spray, quantile)
##
   $A
##
      0%
           25%
                 50%
                       75%
                           100%
##
   7.00 11.50 14.00 17.75 23.00
##
## $B
##
    0% 25% 50% 75% 100%
   7.0 12.5 16.5 17.5 21.0
##
##
## $C
         25%
             50%
                  75% 100%
##
    0%
##
    0.0 1.0
             1.5 3.0 7.0
##
##
   $D
##
                       75% 100%
      0%
           25%
                 50%
##
    2.00
         3.75 5.00
                      5.00 12.00
##
## $E
        25% 50%
                  75% 100%
##
     0%
## 1.00 2.75 3.00 5.00 6.00
##
## $F
     0% 25% 50% 75% 100%
##
##
  9.0 12.5 15.0 22.5 26.0
```

(f) \* Can you write a command which gives you just a vector of the six upper quartiles? [Hint: how can you pass the argument probs=0.75 to quantile() when you are using tapply()?]

(g) Use the plot() command to produce separate boxplots for each level (as we did in class for the height data).

> plot(dat\$spray, dat\$count)