

R—Tutorial

Halil Bisgin, PhD



Installation

- Rstudio—Nice IDE
 - -<u>https://www.rstudio.com/products/rstudio/download2/</u>
- R
 - -https://cran.r-project.org/src/base/R-3/



Nuts & Bolts

- Entering Input
 - -x < -1
 - -print(x)
 - -msg <- "hello"
 - -print(msg)
- Evaluation
 - -x < -5
 - -x



R Objects

- Data Types
 - • character
 - • numeric (real numbers)
 - • integer
 - • complex
 - ● logical (True/False)
- The most basic type of R object is a vector.
- Empty vectors can be created with the vector() function
- You can always check the data type with: class() function.



Attributes

- R objects can have attributes, which are like metadata for the object:
 - -names, dimnames
 - -dimensions (e.g. matrices, arrays)
 - -class (e.g. integer, numeric)
 - -length
 - -other user-defined attributes/metadata



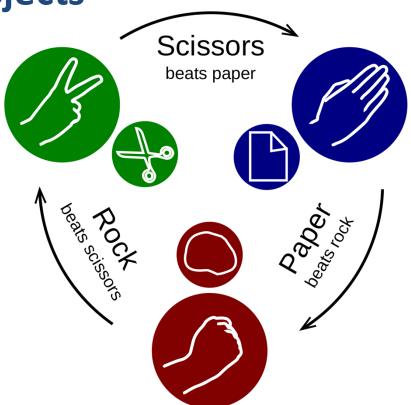
Creating Vectors

- > x <- c(0.5, 0.6) ## numeric
- > x <- c(TRUE, FALSE) ## logical
- > x <- c(T, F) ## logical
- > x <- c("a", "b", "c") ## character
- > x <- 9:29 ## integer
- > x <- c(1+0i, 2+4i) ## complex



Mixing Objects

- > y <- c(1.7, "a") ## character
- > y <- c(TRUE, 2) ## numeric
- > y <- c("a", TRUE) ## character





Explicit Coercion

```
> x < -0:6
```

$$>$$
 class(x)

> as.numeric(x)

[1] 0 1 2 3 4 5 6

> as.logical(x)

[1] FALSE TRUE TRUE TRUE TRUE TRUE TRUE

> as.character(x)

[1] "0" "1" "2" "3" "4" "5" "6"



Matrices

```
> m <- matrix(nrow = 2, ncol = 3)
> m
[,1] [,2] [,3]
[1,] NA NA NA
[2,] NA NA NA
> dim(m)
[1] 2 3
> attributes(m)
$dim
[1] 2 3
```



Matrices



Matrices

• Matrices can be created by column-binding or row-binding with the cbind() and rbind() functions.

```
> x <- 1:3
> y <- 10:12
> cbind(x, y)
x y
[1,] 1 10
[2,] 2 11
[3,] 3 12

> rbind(x, y)
[,1] [,2] [,3]
x 1 2 3
y 10 11 12
```



Lists

• Lists are a special type of vector that can contain elements of different classes.

```
x < -list(1, "a", TRUE, 1 + 4i)
```

 We can also create an empty list of a pre-specified length with the vector() function

```
x <- vector("list", length = 5)</pre>
```



Factors

- Factors are used to represent categorical data and can be unordered or ordered.
- One can think of a factor as an integer vector where each integer has a label.
- Factors are important in statistical modeling and are treated specially by modelling functions like lm().
- Often factors will be automatically created for you when you read a dataset in using a function like read.table().

```
x <- factor(c("yes", "yes", "no", "yes", "no"))
table(x)
unclass(x)</pre>
```



Missing Values

- Missing values are denoted by NA or NaN for q undefined mathematical operations.
 - is.na() is used to test objects if they are NA
 - is.nan() is used to test for NaN
 - NA values have a class also, so there are integer NA, character NA, etc.
 - A NaN value is also NA but the converse is not true

```
>x <- c(1, 2, NA, 10, 3)
> is.na(x)
[1] FALSE FALSE TRUE FALSE FALSE
> is.nan(x)
[1] FALSE FALSE FALSE FALSE FALSE
```



Data Frames

- Data frames are used to store tabular data in R.
- Data frames are represented as a special type of list where every element of the list has to have the same length
- Unlike matrices, data frames can store different classes of objects in each column
- Data frames have a special attribute called row.names
- Data frames can be converted to a matrix by calling data.matrix().(as.matrix() also works)



Data Frames

- > var1 <- 1:5
- > var2 <- (1:5)/10
 - ➤var3 <- c("R", "and", "Data Mining", "Examples", "Case
 Studies")</pre>
 - >df1 <- data.frame(col1=var1, col2=var2, col3=var3)</pre>
 - > names(df1) <- c("VarInt", "VarReal", "VarChar")</pre>



Data Frame

```
> m <- matrix(1:4, nrow = 2, ncol = 2)
> dimnames(m) <- list(c("a", "b"), c("c", "d"))
> m
c d
a 1 3
b 2 4
```

- > colnames(m) <- c("h", "f")
- > rownames(m) <- c("x", "z")



FOR LOOPS

- Not much different.
- Need to indicate the index and its range, though

```
for(j in 3:13) {
  fib[j] <- fib[j-1] + fib[j-2]
}
fib</pre>
```



Other flow control

• R has a while function as well as a for function. This allows iterations of flexible length.

```
while(expression)
{
    code block
}
```

- should be read as
 - while the expression is true, iteratively execute the code block and reevaluate the expression based on the result.



Conditional execution in R

- The syntax of the if function in R is
- if(expression) { code block}
- which is executed as "if the expression is TRUE, execute the code block, otherwise skip it".
- This is handy when the code block can only be executed when some condition holds.
- There is also an if else combination.
- if(expression) { code block}
- else {other code block}



Functions

- R has a rich collection of functions for performing calculations necessary for doing statistics.
- Much of learning R comes down to acquiring a large enough vocabulary of functions to solve your problems.
- The actual structure of the programs may be quite simple, but you still need to learn the names of the functions that do the required work.
- However, even for everyday users need to write their own functions.



Functions

- Functions are written in a library file with a .R extension and the files are loaded into the R Markdown document doing an analysis with the source command.
- Putting the code into its own function saves retyping the commands and ensures reproducibility by forcing exactly the same code to be run when it is used.



Function definition

```
myfun <- function (a, b, c)
{
  do something ...
  do another thing ...
  output #this line will be without <- or =
}</pre>
```



Function example

```
mymean <-function (x,y)
{
  output = (x+y)/2
  output
}</pre>
```



Default arguments

```
mymean <-function (x=0,y=0)
{
  output = (x+y)/2
  output
}</pre>
```



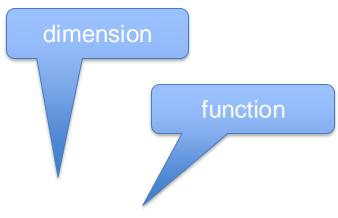
Using built-in functions for rows/columns

```
set.seed(123)
```

v < - sample(x = -10:10, size = 12)

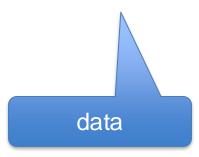
M1 <- matrix(v, nrow=4)

M1



apply

—Example: sum_stats <- apply(expr_mat, 1, summary)</pre>





- Read data from and write data to
 - -R native formats (incl. Rdata and RDS)
 - -CSV files
 - -EXCEL files
 - -ODBC databases
 - -SAS databases



```
a <- 1:10
save(a, file = "./data/dumData.Rdata")
rm(a)
a
## Error in eval(expr, envir, enclos): object 'a' not found
load("./data/dumData.Rdata")
a
    [1] 1 2 3 4 5 6 7 8 9 10
```



```
library(xlsx)
xlsx.file <- "./data/dummmyData.xlsx"</pre>
write.xlsx(df2, xlsx.file, sheetName = "sheet1", row.names = F)
df3 <- read.xlsx(xlsx.file, sheetName = "sheet1")
df3
    VarInt VarReal VarChar
##
## 1
             0.1
                          R.
## 2 2 0.2
                        and
## 4 4 0.4
                    Examples
## 5
        5 0.5 Case Studies
```



```
read.delim(file, header = TRUE, sep = "\t")
read.csv(file, header = TRUE, sep = ",")
read.table(file, header = FALSE, sep = "")
```

- The read.table() function is one of the most commonly used functions for reading data
- file: filename in quotation marks (may need to include path)
- header: if your file has headers (column names)
- sep: separator; each function has its default separator
- There are further options in these functions if you need
- Built in datasets: mtcars, iris



- -file, the name of a file, or a connection
- -header, logical indicating if the file has a header line
- -sep, a string indicating how the columns are separated
- -colClasses, a character vector indicating the class of each column in the dataset
- -nrows, the number of rows in the dataset. By default read.table() reads an entire file.
- Built in datasets: mtcars, iris
- write.table



Working Directory

- > getwd() # get current working directory
 - You can select a different working directory with the function setwd(), and thus avoid entering the full path of the data files.
- setwd("<new path>") # set working directory
 - -Note that the forward slash should be used as the path separator even on Windows platform.
 - > setwd("C:/MyDoc")



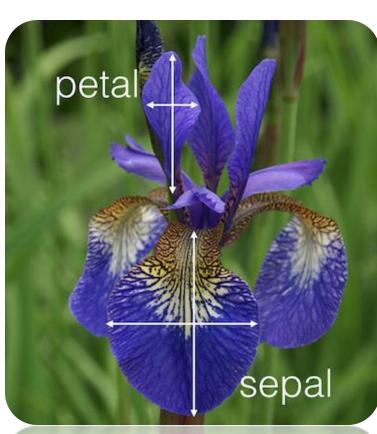
Install and import a library

- install.packages("ggplot2")
- library(ggplot2)



Exploring Data

- dim(iris)
- str(iris)
- names(iris)
- attributes(iris)
- head(iris, 3)
- tail(iris, 3)
- iris[1:10, "Sepal.Length"]
- summary(iris)
- library(Hmisc)– describe(iris[, c(1, 5)])
- range(iris\$Sepal.Length)
- quantile(iris\$Sepal.Length, c(0.1, 0.3, 0.65))



sepa