Introduction to R

Session 2 – Data manipulation

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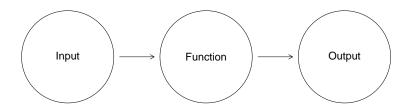
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SCIENCE
DEPARTMENT OF STATISTICS

Functions

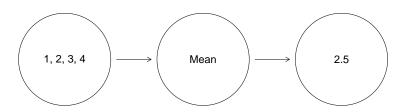
A function is a relationship between a set of inputs (arguments) and a set of outputs. E.g., the function is fed some information on which it operates, the results of which are the output.



This is an essential building block for the R package.

Functions

We have seen many functions, e.g. log, mean, table, with, etc.



Working with functions

- Functions can be user-defined, i.e., you can write your own.
- Output is the last line of the function. You can use return() to specify the output.
- Here is a function calculates the standard error of the mean (SEM).

```
mystder <- function(x) {
    mysd <- sd(x, na.rm = TRUE) # Calc std. deviation
    n <- length(x) # Calc sample size
    mysd/sqrt(n) # Definition of SEM
}
mystder(Patient.df$Height)</pre>
```

```
## [1] 0.02999245
```

 A set of user-defined functions can be bundled together into an R package.

Getting data into R

- Base R includes only functions which read data sets saved in simple file formats, e.g. csv, txt, tab delimited, etc.
- What if your data was saved in another format, e.g. Excel?
- The readxl package for R contains functions that may help, https: //cran.r-project.org/web/packages/readxl/index.html

```
library(readxl)
excel <- read_excel("data.xlsx", sheet = 1)</pre>
```

Getting data into R

- What if your data was saved in another format, e.g. Excel, STATA, SPSS, or SAS spreadsheets?
- The haven package for R contains functions that may help, https: //cran.r-project.org/web/packages/haven/index.html

```
library(haven)
stata <- read_dta("data.dta")
spss <- read_sav("data.sav")
sas <- read_sas("data.sas7bdat")
sasxport <- read_xpt("data.xpt")</pre>
```

However, it is always the easiest and safest to read data into R from a csv file.

Packages

- Currently, the CRAN package repository features 10,098 available packages (4 Jul. 2017). There are about 13,169 CRAN, BioConductor and Github packages in total.
- To install packages from the R GUI, click on Packages → Install Package(s) ... → New Zealand (or whatever region you are located) → Package name
- Or, you can type install.packages(package name), e.g. install.packages("haven").
- After the installation, use library("package name") to load it into R.

Note: Installation is performed only once; however, it must be loaded (i.e. use the command library("package name")) in every R session.

Bioconductor Packages

• To install core packages, type the following in an R command window:

```
source("https://bioconductor.org/biocLite.R")
biocLite()
```

• Install specific packages, e.g., "GenomicFeatures" and "AnnotationDbi", with

```
biocLite(c("GenomicFeatures", "AnnotationDbi"))
```

Note: Installation is performed only once; however, it must be loaded (i.e. use the command library("package name")) in every R session.

which individual does Smoke?

Let's use R's powerful subsetting capabilities to select those cases for which the value of Smoke is 1 (i.e. Yes).

```
index <- which(Patient.df$Smoke == 1)
index</pre>
```

##	[1]	6	7	8	12	15	16	18	
##	[8]	19	20	21	22	23	28	33	
##	[15]	35	38	45	47	48	52	53	
##	[22]	55	56	57	58	64	70	75	
##	[29]	79	80	81	89	90	91	93	
##	[36]	99	108	109	110	116	118	119	
##	[43]	122	123	124	127	136	137	144	
##	[50]	146	148	154	155	162	172	174	
##	[57]	180	182	187	188	189	191	192	
##	[64]	195	198	204	205	210	213	221	
##	[71]	225	229	230	232	233	234	237	
	5								

How many are there?

```
# Use length() to count the number of elements in
# 'index'.
length(index)
```

```
## [1] 4371
```

- Patient.df\$Smoke == 1 gives a vector of True/False for every observations.
- which() gives a vector of the position of TRUE.
- length() tells us how many elements there are in the vector.

How many are there?

Another way is to sum up how many TRUEs in the logical vector Patient.df\$Smoke == 1

```
sum(Patient.df$Smoke == 1)
```

```
## [1] NA
```

How many are there?

Another way is to sum up how many TRUEs in the logical vector Patient.df\$Smoke == 1

```
sum(Patient.df$Smoke == 1, na.rm = TRUE)
```

```
## [1] 4371
```

Who are they?

```
# Use square brackets to extract IDs corresponding
# to the cases numbers contained in 'index'
index <- which(Patient.df$Smoke == 1)</pre>
Patient.df$Patient.ID[index]
        [1]
                19
                                             55
                                                     56
                                                            67
##
                       34
                               44
                                      51
        [8]
                70
                       71
                               72
                                      73
                                             74
##
                                                     90
                                                           110
      [15]
               115
                      120
                             129
                                     139
                                            140
                                                           155
##
                                                   154
##
      [22]
               159
                      161
                             164
                                     165
                                            180
                                                   190
                                                           210
      [29]
##
               217
                      218
                             219
                                     242
                                            249
                                                   251
                                                           255
      [36]
##
               275
                      297
                             298
                                     303
                                            317
                                                   331
                                                           335
      [43]
##
               343
                      344
                             346
                                     353
                                            376
                                                   377
                                                           395
##
      [50]
              400
                      403
                             427
                                     429
                                            445
                                                   470
                                                           474
##
      [57]
              485
                      488
                             503
                                     507
                                            509
                                                   519
                                                           520
##
      [64]
               524
                      528
                             537
                                     539
                                            550
                                                   554
                                                           570
      [71]
##
               578
                      586
                             588
                                     592
                                            593
                                                   595
                                                           604
##
      [72]
               606
                      618
                             620
                                     639
                                            644
                                                   646
                                                           650
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```

Subsetting

First element only
Patient.df\$Height[1]

Square brackets [] are used to extract subsets of data.

```
## [1] 70.4
# All but the first element
Patient.df$Height[-1]
##
       [1] 63.9 61.8 69.8 NA 70.2 62.6 64.4 64.3
##
       [9] 67.6 59.5 71.1 70.2 61.0 68.0 73.4 61.1
##
      [17] 63.8 67.8 61.7 71.1 68.4 65.7 61.4 59.8
      [25] 61.8 60.3 66.1 67.9 64.1 59.1 70.0 62.4
##
##
      [33] 63.7 65.1 64.3 65.4 69.7 67.5 68.0 71.1
##
      [41] 63.6 63.2 64.6 74.8 61.4 64.5 65.2 71.9
      [49] 67.4 66.5 65.7 68.5 60.0 67.2 69.4 65.8
##
```

Subsetting

Square brackets [] are used to extract subsets of data.

```
# Elements 3 through 8
Patient.df$Height[3:8]
```

```
## [1] 61.8 69.8 NA 70.2 62.6 64.4
```

```
# Elements 3 and 8
Patient.df$Height[c(3, 8)]
```

```
## [1] 61.8 64.4
```

Subsetting two-dimensional arrays, such as data frames, requires the use of two indices.

```
# First row or record
Patient.df[1, ]
```

```
## Patient.ID Age Sex Race Weight Height Smoke
## 1 3 21 Male 1 179.5 70.4 NA
```

For data frames, you need to use two indices.

```
# Second column or variable
Patient.df[, 2]
```

```
##
       [1] 21 32 48 35 48 44 42 24 67 56 82 44 50 36
##
      [15] 48 32 66 70 63 37 60 42 58 80 80 23 83 28
##
      [29] 90 86 27 72 34 21 45 84 36 28 69 63 31 25
##
      [43] 41 33 39 55 72 32 27 55 78 65 57 26 31 31
##
      [57] 49 61 65 42 48 80 22 43 72 34 59 43 22 73
      [71] 66 69 79 85 48 32 73 32 28 62 38 56 48 22
##
      [85] 21 66 89 61 55 52 51 66 33 54 24 83 31 42
##
      [99] 24 47 35 21 59 64 41 29 63 33 53 60 76 31
##
##
     [113] 20 77 27 40 30 78 43 24 80 66 48 66 89 42
     [127] 47 46 22 73 46 38 42 81 77 65 24 21 36 73
##
     [141] 81 20 39 38 61 30 76 58 81 67 26 53 51 33
##
     [155] 58 85 35 30 85 72 61 38 29 26 45 34 61 70
##
```

For data frames, you need to use two indices.

```
# Some rows and columns
Patient.df[2:5, 4:7]
```

```
## Race Weight Height Smoke

## 2 1 NA 63.9 NA

## 3 1 149.7 61.8 2

## 4 1 203.5 69.8 NA

## 5 1 155.3 NA 2
```

For data frames, you need to use two indices.

```
# Rows by number, columns by name
Patient.df[1:10, c("Race", "Smoke")]
```

```
##
   Race Smoke
## 1
          1
               NA
               NA
## 2
## 3
               NΑ
## 4
## 5
## 6
## 7
## 8
## 9
               NA
## 10
```

Missing values

-R reserves the object NA (Not Available) for elements of a vector that are missing or unavailable. - Use of is.na() to search for missing values requires that they are recorded as NA. - na will not do because R is case sensitive!

```
sum(is.na(Patient.df$Smoke))
```

[1] 8404

Missing values

The default option of table() ignores NAs when constructing frequency tables. Now that all occurrences of "NA, refused" have been replaced with NA, missing values will no longer be shown in frequency tables constructed using table().

```
table(Patient.df$Smoke)
```

```
## 1 2
## 4371 4255
```

##

Missing values

If you still want to see how many NAs in the frequency tables, you can change the useNA arguement to "always" in table().

```
table(Patient.df$Smoke, useNA = "always")
```

The ifelse() function provides a quick way to convert the Smoke variable in Patient.df from number to words, i.e.

```
ifelse(test, yes, no)
```

- test: a logical test.
- yes, what happens if the test is True.
- no, what happens if the test is False.

```
Patient.df$Smoke.group <-
  with(Patient.df, ifelse(Smoke == 1, "Yes", "No"))
table(Patient.df$Smoke.group)</pre>
```

```
## No Yes
## 4255 4371
```

How about Race? One ifelse() inside another ifelse():

```
##
## African Caucasian Other
## 4860 11612 553
```

head(Patient.df)

```
##
    Patient.ID Age
                      Sex Race Weight Height Smoke
                                179.5
## 1
                21
                     Male
                             1
                                        70.4
                                               NA
                32 Female
                                   NA
                                        63.9
                                               NA
## 2
                48 Female
## 3
                              149.7
                                       61.8
                                               NA
## 4
            10
                35
                     Male
                             1 203.5
                                        69.8
## 5
            11
                48
                     Male
                             1 155.3
                                         NA
                44
                     Male
                             2
                                189.6
                                        70.2
## 6
            19
##
    Smoke.group Race.group
## 1
           <NA>
                 Caucasian
## 2
           <NA> Caucasian
## 3
             No
                 Caucasian
## 4
           <NA> Caucasian
## 5
             No
                 Caucasian
## 6
            Yes
                   African
```

```
str(Patient.df)
```

```
'data.frame': 17030 obs. of 9 variables:
##
    $ Patient.ID : int
                        3 4 9 10 11 19 34 44 45 48 ...
##
    $ Age
                        21 32 48 35 48 44 42 24 67 56 ...
                 : int
                 : chr "Male" "Female" "Female" "Male" ...
##
    $ Sex
##
    $ Race
                 : int 1 1 1 1 1 2 2 1 2 1 ...
##
    $ Weight
                        180 NA 150 204 155 ...
                 : num
##
    $ Height
                        70.4 63.9 61.8 69.8 NA 70.2 62.6 64.4
                 : num
##
    $ Smoke
                        NA NA 2 NA 2 1 1 1 NA 2 ...
                 : int
##
    $ Smoke.group: chr
                        NA NA "No" NA ...
##
    $ Race.group : chr
                        "Caucasian" "Caucasian" "Caucasian" "C
```

Now we have created new variables, Smoke.group and Race.group, the next step is to remove the Smoke and Race variables in the Patient.df.

```
names(Patient.df)
```

```
## [1] "Patient.ID" "Age" "Sex"
## [4] "Race" "Weight" "Height"
## [7] "Smoke" "Smoke.group" "Race.group"
```

Now we have created new variables, Smoke.group and Race.group, the next step is to remove the Smoke and Race variables in the Patient.df.

```
Smoke.index <- which(names(Patient.df) == "Smoke")
Smoke.index
## [1] 7
Race.index <- which(names(Patient.df) == "Race")
Race.index</pre>
```

```
names(Patient.df[, -c(Smoke.index, Race.index)])
```

```
## [1] "Patient.ID" "Age" "Sex"
## [4] "Weight" "Height" "Smoke.group"
```

[1] 4

Data Cleaning with %in% operator

```
names(Patient.df)
## [1] "Patient.ID"
                       "Age"
                                       "Sex"
                       "Weight" "Height"
## [4] "Race"
                       "Smoke.group" "Race.group"
## [7] "Smoke"
names(Patient.df) %in% c("Smoke", "Race")
## [1] FALSE FALSE FALSE TRUE FALSE FALSE
                                                 TRUE.
## [8] FALSE FALSE
names(Patient.df[, !names(Patient.df) %in% c("Smoke",
    "Race")])
## [1] "Patient.ID"
                       "Age"
                                       "Sex"
## [4] "Weight"
                       "Height"
                                       "Smoke.group"
       "Race.group"
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```

Data Cleaning with %in% operator

```
Patient.df <- Patient.df[, !names(Patient.df) %in%
    c("Smoke", "Race")]</pre>
```

Your turn

First to calculate BMI:

$$\mathtt{BMI} = \frac{\mathtt{Weight(kg)}}{\mathtt{Height(m)}^2} = \frac{\mathtt{Weight(pounds)}}{\mathtt{Height(inches)}^2} \times 703$$

Then, create a BMI.group for normal, overweight and obese, where,

- overweight is a BMI greater than or equal to 25; and
- obesity is a BMI greater than or equal to 30.

Your turn

To calculate BMI:

$$\mathtt{BMI} = \frac{\mathtt{Weight(kg)}}{\mathtt{Height(m)}^2} = \frac{\mathtt{Weight(pounds)}}{\mathtt{Height(inches)}^2} \times 703$$

Patient.df\$BMI <- (Patient.df\$Weight/Patient.df\$Height^2)*703
mean(Patient.df\$BMI, na.rm = TRUE)</pre>

[1] 27.03084

Or,

Calculate BMI using Weight and Height
Patient.df\$BMI <- with(Patient.df, (Weight/Height^2)*703)
mean(Patient.df\$BMI, na.rm = TRUE)</pre>

[1] 27.03084

Your turn

```
Patient.df$BMI.group <-
  ifelse(Patient.df$BMI >= 30, "obese",
         ifelse(Patient.df$BMI >= 25, "overweight",
          "normal"))
table(Patient.df$BMI.group)
##
##
       normal
                   obese overweight
##
         6916
                    4185
                                5866
table(Patient.df$BMI.group, useNA = "always")
```

normal

6916

##

##

5866

obese overweight

4185

<NA>

63

```
with(Patient.df, table(Race.group, BMI.group))
```

```
##
             BMI.group
              normal obese overweight
  Race.group
    African
                1858
                      1439
                                1546
##
##
    Caucasian
                4774 2627
                                4168
    Other
                 282
                       119
                                 149
##
```

Produce the last table with Race groups of African and Caucasian, and BMI groups of obese and overweight

```
exclude.rows <- with(Patient.df, which(Race.group ==
    "Other" | BMI.group == "Normal"))
head(Patient.df[-exclude.rows, c("Race.group", "BMI.group")])
```

```
Race.group BMI.group
    Caucasian overweight
## 1
## 2 Caucasian
                     <NA>
## 3 Caucasian overweight
## 4 Caucasian overweight
## 5 Caucasian
                     <NA>
       African overweight
## 6
```

##

```
## BMI.group
## Race.group normal obese overweight
## African 1858 1439 1546
## Caucasian 4774 2627 4168
```

Convert counts to percentages rounded to 1 decimal place.

```
round(prop.table(R.B.tab) * 100, 1)
```

```
## BMI.group
## Race.group normal obese overweight
## African 11.3 8.8 9.4
## Caucasian 29.1 16.0 25.4
```

Easier way

```
R.B.tab1 <- with(Patient.df, table(Race.group, BMI.group))</pre>
dim(R.B.tab1)
## [1] 3 3
row.names(R.B.tab1)
## [1] "African"
                 "Caucasian" "Other"
colnames (R.B.tab1)
## [1] "normal"
                     "obese"
                                   "overweight"
```

Easier way

```
colnames (R.B.tab1)
## [1] "normal"
                     "obese"
                                  "overweight"
which(row.names(R.B.tab1) == "Other")
## [1] 3
which(colnames(R.B.tab1) == "Normal")
## integer(0)
```

Easier way

```
R.B.tab1[-3, -1]

## BMI.group

## Race.group obese overweight

## African 1439 1546

## Caucasian 2627 4168
```

```
## BMI.group
## Race.group obese overweight
## African 14.7 15.8
## Caucasian 26.9 42.6
```

round(prop.table(R.B.tab1[-3, -1]) * 100, 1)

```
# Mean BMI level of females
with(Patient.df, mean(BMI[Sex == "Female"], na.rm = TRUE))
## [1] 27.45053
# Mean BMI level of females and Caucasian
with (Patient.df,
    mean(BMI[Sex == "Male" &
                   Race.group == "Caucasian"],
          na.rm = TRUE))
```

Summary

- Making R functions
- Installing and loading R pacakges
- Subsetting vectors and datasets
- ifelse() function