SCR week 4: exercises

Exercises part 1

1.1 Write your own Songtext (Part 1)

In this exercise we will program a function to produce a file with the lyrics of the '99 bottles song' in three different versions; based on lapply(), sapply() and (perhaps?) replicate().

a.

Take a look at the following code:

```
in_stock <- 99
paragraph <- paste0(
  in_stock, " bottles of beer on the wall, ",
  in_stock, " bottles of beer. \n",
  "Take one down, pass it around, ",
  in_stock - 1," bottles of beer on the wall...\n\n"
)
cat(paragraph)</pre>
```

```
## 99 bottles of beer on the wall, 99 bottles of beer.
## Take one down, pass it around, 98 bottles of beer on the wall...
```

class(bttls10to2_sapply) # = atomic (one mode) in a vector structure.

Use lapply(), sapply(), to produce the paragraphs for B = 10 to B = 2 bottles of beer on the wall. What is the class of the output of sapply() what is the class of the output of lapply()?

```
CreateParagraphsBottlesBeer <- function(bttls) {
  paragraph <- paste0(
    bttls, " bottles of beer on the wall, ",
    bttls, " bottles of beer. \n",
    "Take one down, pass it around, ",
    bttls - 1, " bottles of beer on the wall...\n\n"
)
  # cat(paragraph)
  return(paragraph)
  return(paragraph)
}
bttls10to2_lapply <- lapply(10:3, CreateParagraphsBottlesBeer)
class(bttls10to2_lapply)

## [1] "list"

bttls10to2_sapply <- sapply(10:3, CreateParagraphsBottlesBeer)</pre>
```

```
## [1] "character"
```

b.

Since replicate() is a wrapper around the sapply() (and thus the lapply()) function, would it be "natural" the use the replicate() function to write the paragraphs of the N Bottles of Beers song?

Answer:

Nope, as is stated in the helpfile: "replicate() is for repeated evaluation of an expression". For each paragraph the expression changes, i.e. the number of bottles changes.

c.

Time the functions lapply() and sapply() for B = 10 to B = 2 "bottles on the wall" with (only) the use of system.time(). Can you say which of the two functions is slower?

Answer:

```
system.time(
  bttls10to2_lapply <- lapply(10:3, CreateParagraphsBottlesBeer)
)
##
      user
           system elapsed
##
         0
                  0
system.time(
  bttls10to2_sapply <- sapply(10:3, CreateParagraphsBottlesBeer)</pre>
##
      user system elapsed
##
             0.000
                      0.001
     0.000
```

Nope they seem equally fast.

d.

Use the replicate() function to measure the running time when the expressions of the previous subquestions where each B = 1000 times repeated. Do you know have a more definite answer on which of the two functions is slower?

```
B <- 1e3
system.time(
  replicate(B, bttls10to2_lapply <- lapply(10:3, CreateParagraphsBottlesBeer))
)

### user system elapsed
## 0.043 0.005 0.047

system.time(
  replicate(B, bttls10to2_sapply <- sapply(10:3, CreateParagraphsBottlesBeer))
)</pre>
```

```
## user system elapsed
## 0.050 0.003 0.053
```

Jep, sapply() is slower, which makes sense since it is a wrapper around lapply()

e.

In this case using simplify2array or unlist would not make any difference for the output of lapply() in this exercise. Could you tell which of the two functions is faster?

Answer:

```
system.time(replicate(B, unlist(bttls10to2_lapply)))
##
      user system elapsed
##
     0.002
            0.000
                     0.003
system.time(replicate(B, simplify2array(bttls10to2_lapply)))
##
      user
            system elapsed
##
     0.008
             0.000
                     0.008
Here, unlist() is faster.
```

f.

Write a function that can put the luries of the

Write a function that can put the lyrics of the song into a .txt file and can store it to any destination on your computer (given that the destination exists) with use of the cat() function. The input parameters (arguments) of your functions are: B, the number of bottles your start with; and the address for your "to be created" lyrics file. The function should return: invisible(NULL).

Use (if you like) the following variables / objects inside your function:

```
header_song <- pasteO("\n", N = 99, " Bottles of Beers \n\n")
end_song <- c("2 bottles of beer on the wall, 2 bottles of beer\n",
   "Take one down, pass it round 1 bottles of beer\n\n",
   " 1 bottle of beer on the wall, 1 bottle of beer\n",
   "No more bottles of beer\n"
)</pre>
```

Last, apply your created function to write your lyrics to the folder O_data in your working directory.

```
SingTheBottleSong <- function(
  bottle_nr,
  address
) {
  each_bottle_nr <- bottle_nr:3</pre>
```

```
header_song <- paste0("\n", bottle_nr, " Bottles of Beers \n\n")
  CreateParagraphsBottlesBeer <- function(bttls) {</pre>
    paragraph <- paste0(</pre>
      bttls, " bottles of beer on the wall, ",
      bttls, " bottles of beer. n,
      "Take one down, pass it around, ",
      bttls - 1, " bottles of beer on the wall...\n\n"
    )
    # cat(paragraph)
    return(paragraph)
  firstpart_song <- lapply(each_bottle_nr, CreateParagraphsBottlesBeer)</pre>
  firstpart_song <- unlist(firstpart_song)</pre>
  end_song <- c(
    "2 bottles of beer on the wall, 2 bottles of beer.n",
    "Take one down, pass it round 1 bottle of beer.\n\n",
    "1 bottle of beer on the wall, 1 bottle of beer.\n",
    "No more bottles of beer.\n"
  )
  lyrics <- c(</pre>
    header_song,
    firstpart_song,
    end_song
  cat(lyrics, sep = "", file = address)
 return(invisible(NULL))
SingTheBottleSong(
 bottle_nr = 5,
  address = "0_data/my_bottlesong_lyrics.txt"
)
```

1.2 Wine Equality: <*>apply() family and aggregate

If you haven't done so, make sure you have finished the swirl() module: 11: vapply and tapply.

a.

Read the file winequality-red.ssv into a data.frame object and write code that checks the type (or mode) of each column in the data.frame.

```
wine_quality <- read.csv("0_data/winequality-red.ssv", sep = ";", header = TRUE )
apply(wine_quality, 2, typeof)

## fixed.acidity volatile.acidity citric.acid
## "double" "double" "double"</pre>
```

```
##
         residual.sugar
                                     chlorides free.sulfur.dioxide
##
                "double"
                                      "double"
                                                            "double"
## total.sulfur.dioxide
                                      density
               "double"
                                      "double"
                                                            "double"
##
##
              sulphates
                                       alcohol
                                                             quality
##
               "double"
                                      "double"
                                                            "double"
```

b.

Categorize the columns density and pH into two new factors columns with each two categories: higher than the median, and lower or equal to the median.

Answer:

```
wine_quality$dens_fctr <- factor(
   wine_quality$density > median(wine_quality$density),
   labels = c("low", "high")
)
wine_quality$pH_fctr <- factor(
   wine_quality$pH > median(wine_quality$pH),
   labels = c("low", "high")
)
```

c.

Use tapply() to create a table for the means of fixed.acidity in a cross-table for "low and high density" with "low and high pH".

Answer:

d.

Use aggregate() to find the means for fixed.acidity of all combinations of "low and high density" with "low and high pH".

```
aggregate(
    x = wine_quality$fixed.acidity,
    by = list(
        Density = wine_quality$dens_fctr,
        pH = wine_quality$pH_fctr
```

```
),
  FUN = mean
##
     Density
               рΗ
         low low 8.245079
## 1
## 2
        high low 9.953452
## 3
         low high 6.943648
## 4
        high high 7.858131
e.
Repeat the sub-exercises \mathbf{c} and \mathbf{d}, but use an anonymous function that outputs the mean as well as the
standard deviation. Does the output of both tapply() and aggregate() makes sense?
Answer:
aggregate(
  x = wine_quality$fixed.acidity,
  by = list(
    Density = wine_quality$dens_fctr,
    pH = wine_quality$pH_fctr
  ),
  FUN = function(x) {
         m <- mean(x, na.rm = TRUE)</pre>
         s <- sd(x, na.rm = TRUE)
         return(c(mean = m, sd = s))
       }
)
##
     Density
              рΗ
                      x.mean
         low low 8.2450794 1.0290368
## 1
## 2
        high low 9.9534517 1.7092179
## 3
        low high 6.9436475 0.8825252
## 4
        high high 7.8581315 1.0434042
```

```
## pH
## Density low high
## low Numeric,2 Numeric,2
## high Numeric,2 Numeric,2
```

The output of aggregate seems to be more user-friendly. The tapply() output is a matrix with elements of mode list. Each element of the matrix is a list of 1 element / component that contains the two numeric values: the mean and the standard deviation.

Btw, also notice the difference in speed of the code:

user system elapsed

0.198

0.004

##

0.193

```
system.time(replicate(1e3, {
  obj_agg <- aggregate(</pre>
    x = wine_quality$fixed.acidity,
    by = list(
      Density = wine_quality$dens_fctr,
      pH = wine_quality$pH_fctr
    ),
    FUN = function(x) {
         m <- mean(x, na.rm = TRUE)
         s <- sd(x, na.rm = TRUE)
         return(c(mean = m, sd = s))
       }
  )
}))
##
      user system elapsed
##
            0.042
     1.313
                     1.358
system.time(replicate(1e3, {
  obj_tap <- tapply(X = wine_quality$fixed.acidity,</pre>
     INDEX = list(
       Density = wine_quality$dens_fctr,
       pH = wine_quality$pH_fctr
     ),
     FUN = function(x) {
        m <- mean(x, na.rm = TRUE)</pre>
        s <- sd(x, na.rm = TRUE)
        return(c(mean = m, sd = s))
     }
  )
}))
##
```

Exercises: part 2

2.1 From <*>pply towards for

In the Week 2 exercises you created a list object:

```
set.seed(20181005)
my_list <- replicate(10, list(rnorm(5)))</pre>
```

This object replicated the list(rnorm(5)) command 10 times, and put the results into a list (of length 10). On each of these ten entries we calculated the mean of the 5 random normal draws, this was done by the code:

```
my_list_means <- lapply(my_list, mean)</pre>
```

 \mathbf{a}

Create a similar my_list object, but instead of using the replicate function, use a for loop. Do not forget to first create a list of length 10, using for example:

```
my_list <- vector(length = 10, mode = "list")</pre>
```

Due to the fact that you're sampling from a normal distribution, the values in your list will be different from the list of the Week 2 exercises.

Answer:

```
for (i in 1:10) {
  my_list[[i]] <- rnorm(5)
}</pre>
```

b

We can time the code from lecture 2 as follows:

```
system.time(replicate(1e4, {
  set.seed(20181005)
  my_list <- replicate(10, list(rnorm(5)))
}))</pre>
```

```
## user system elapsed
## 0.500 0.023 0.524
```

Which is of the two code chunks is faster? the code with the for loop or the code from the exercise of lecture 2?

```
system.time(replicate(1e4, {
  my_list <- vector(length = 10, mode = "list")
  for (i in 1:length(my_list)) {
    my_list[[i]] <- rnorm(5)
  }
}))</pre>
```

```
## user system elapsed
## 0.224 0.029 0.254
```

The for loop is faster.

 \mathbf{c}

We assign the mean of each entry in my_list towards the object my_means as follows:

```
my_means1 <- lapply(my_list, mean)</pre>
```

Create a for loop with which we can create the same my_means2 object. Can you conclude which of the two ways of coding is faster, is it the code with the lapply function, or your code with the for loop?

Answer:

```
my_means2 <- vector(mode = "list", length = length(my_list))
for (i in 1:length(my_list)) {
   my_means2[[i]] <- mean(my_list[[i]])
}
all.equal(my_means2, my_means1)</pre>
```

```
## [1] TRUE
```

The computing time needed with the for loop:

```
system.time(replicate(1e4, {
  my_means2 <- vector(mode = "list", length = length(my_list))
  for (i in 1:length(my_list)) {
     my_means2[[i]] <- mean(my_list[[i]])
  }
}))</pre>
```

```
## user system elapsed
## 0.251 0.001 0.252
```

The computing time needed with the lapply function:

```
system.time(replicate(1e4, {
  my_means1 <- lapply(my_list, mean)
}))</pre>
```

```
## user system elapsed
## 0.283 0.002 0.285
```

The difference is very hard to detect. The code with lapply() function seems to be slightly faster...

\mathbf{d}

##

0.818

Quicker code would be to combine to calculate the mean directly on the five samples generated from the standard normal distribution.

The replicate() and lapply() together would give us

0.833

```
system.time(replicate(1e4, {
   set.seed(20181005)
   my_list <- replicate(10, list(rnorm(5)))
   my_means <- lapply(my_list, mean)
}))

## user system elapsed</pre>
```

whereas using lapply() only, gives us:

0.012

```
system.time(replicate(1e4, {
  set.seed(20181005)
  my_means <- lapply(1:10, function(b) {
     mean(rnorm(5))
  })
})</pre>
```

```
## user system elapsed
## 0.647 0.020 0.671
```

Can you rewrite the second R code chunk into a for loop? That is, system.time() a replication of the for loop for 1e3 times.

Answer:

```
system.time(replicate(1e3, {
    set.seed(20181005)
    my_means <- vector(length = 10, mode = "list")
    for (i in 1:10) {
        my_means[[i]] <- mean(rnorm(5))
    }
}))</pre>
```

```
## user system elapsed
## 0.051 0.000 0.051
```

It is interestingly to see that the for loop seems to the "fastest" here.

When using an anonymous function for the FUN argument in lapply(), then, the more complex this function becomes, we see that a "correctly applied" explicit loop (such as the for loop) has a speed gain over lapply()

2.3 An improvement of the 99 Bottles of Beer song?

Step-by step we are going to improve the SingTheBottleSong() function in this exercise (see Exercises part 1).

Remember, the main function inside SingTheBottleSong() was:

```
CreateParagraphsBottlesBeer <- function(bttls) {
   paragraph <- pasteO(
     bttls, " bottles of beer on the wall, ",
     bttls, " bottles of beer. \n",
     "Take one down, pass it around, ",
     bttls - 1, " bottles of beer on the wall...\n\n"
   )
   # cat(paragraph)
   return(paragraph)
}</pre>
```

for the complete lyrics, take a look at:

http://99-bottles-of-beer.net/lyrics.html

a.

The first improvement of the function CreateParagraphsBottlesBeer() is to make sure that bttls can be coerced to an integer using:

```
bttls <- suppressWarnings(as.integer(bttls))</pre>
```

Take a look at the help file of the functions stop() and supressWarnings().

If bttls is not coerced to an integer, it probably gives NA. If this is the case, then use stop() to exit the function by leaving an appropriate message. Hint: Look at the help-file or internet for examples on how to use stop() in a function.

Answer:

Add the following two lines in the beginning of the body of the CreateParagraphsBottlesBeer() function:

```
bttls <- suppressWarnings(as.integer(bttls))
if(is.na(bttls)) stop("bttls should be able to be coerced to integer")</pre>
```

b.

Note that CreateParagraphsBottlesBeer() does not make sense for bttls = 2 or bttls = 1. Use an if(cond) cons1.expr else if cons2.expr else alt.expr

expression to create the three needed different paragraphs inside the updated function.

Answer:

replace the paragraph <- pasteO(___) inside the body of the function with the following lines of code:

```
if(bttls >= 3) {
  paragraph <- paste0(</pre>
    bttls, " bottles of beer on the wall, ",
    bttls, " bottles of beer. \n",
    "Take one down, pass it around, ",
    bttls - 1, " bottles of beer on the wall...\n\"
} else if (bttls==2) {
 paragraph <- paste0(</pre>
    bttls, " bottles of beer on the wall, ",
    bttls, " bottles of beer. \n",
    "Take one down, pass it around, ",
    bttls - 1, " bottle of beer on the wall...\n\n"
} else (bttls == 1) {
 paragraph <- paste0(</pre>
    bttls, " bottle of beer on the wall, ",
    bttls, " bottle of beer. n,
    "Take one down, pass it around, ",
    "no more bottles of beer on the wall.\n\"
```

c.

Also, use stop() with an appropriate message to exit the function when bttls has a value < 1.

Answer:

Adding

if(bttls < 1) stop("bttls should be >= 1")

inside the function gives us the following updated function:

```
CreateParagraphsBottlesBeer <- function(bttls) {</pre>
  bttls <- suppressWarnings(as.integer(bttls))</pre>
  if (is.na(bttls)) stop("bttls should be able to be coerced to integer")
  if (bttls < 1) stop("bttls should be >= 1")
  if (bttls >= 3) {
    paragraph <- paste0(</pre>
      bttls, " bottles of beer on the wall, ",
      bttls, " bottles of beer. \n",
      "Take one down, pass it around, ",
      bttls - 1, " bottles of beer on the wall...\n\n"
  } else if (bttls == 2) {
    paragraph <- paste0(</pre>
      bttls, " bottles of beer on the wall, ",
      bttls, " bottles of beer. \n",
      "Take one down, pass it around, ",
      bttls - 1, " bottle of beer on the wall...\n\n"
```

```
} else {
   paragraph <- pasteO(
     bttls, " bottle of beer on the wall, ",
     bttls, " bottle of beer. \n",
     "Take one down, pass it around, ",
     "no more bottles of beer on the wall.\n\n"
   )
}
return(paragraph)
}</pre>
```

The function of the full song now becomes:

```
SingTheBottleSong <- function(
  bottle_nr,
  address
) {
  each_bottle_nr <- bottle_nr:1 # notice the change here!!
  header_song <- pasteO("\n", bottle_nr, " Bottles of Beers \n\n")
  body_song <- unlist(lapply(each_bottle_nr, CreateParagraphsBottlesBeer))
  lyrics <- c(
    header_song,
    body_song
)
  cat(lyrics, sep = "", file = address)
  return(invisible(NULL))
}
SingTheBottleSong(
  bottle_nr = 5,
  address = "O_data/my_bottlesong_lyrics.txt"
)</pre>
```

2.4 Functions and if else statements | About your SCR grade

a.

Look at the help-file of the function round, can you spot in the help-file where it is written that a number of 8.5 would be rounded to an 8, instead of a 9? What does it say?

Answer

See the text "Note that for rounding off a 5, the IEC 60559 standard is expected to be used, go to the even digit."

b.

Complete the function MakeRound on the underscores ___, such that it rounds the number 8.5 to a 9.

```
MakeRound <- function(x) {
   tmp <- trunc(x)
   if(___) {
      out <- ceiling(x)
   } else {
      out <- floor(x)
   }
   return(out)
}</pre>
MakeRound(8.5)
```

Answer:

```
MakeRound <- function(x) {
   tmp <- trunc(x)
   if (x - tmp >= 0.5) {
      out <- ceiling(x)
   } else {
      out <- floor(x)
   }
   return(out)
}</pre>
MakeRound(8.5)
```

[1] 9

c.

Final grades for the courses in the Statistical Science programme are rounded to whole and half numbers except the grade 5.5 (e.g. .., 4.5, 5, 6, 6.5, 7, ..). Modify the MakeRound function such that it rounds towards whole and half numbers. Show that your modified function rounds 8.25 towards a 8.5 and not towards round (8.25*2)/2. The 5.49, however, should round towards 5, and a 5.5 towards 6.

You may want to use the following pogramming structure inside the body of your function:

```
if(x >= 5 && x < 5.5) {
    ---
} else if(x >= 5.5 & x < 5.75) {
    ---
} else if(___) {
    ---
} else {
    ---
}</pre>
```

```
MakeRound <- function(x) {
  tmp <- trunc(2*x)
  if (x >= 5 && x < 5.5) {
    out <- 5
  } else if (x >= 5.5 & x < 5.75) {</pre>
```

```
out <- 6
} else if (2*x - tmp >= 0.5) {
   out <- ceiling(2*x)/2
} else {
   out <- floor(2*x)/2
}
   return(out)
}
MakeRound(5.49)

## [1] 5

MakeRound(5.50)

## [1] 6</pre>
MakeRound(8.25)
```

[1] 8.5

d.

Use your last MakeRound function to create a modified version of the lecture's CalculateGrade function such that it rounds the final grade towards a whole or half number. Here's the

```
CalculateGrade <- function(A, E1, E2) {
  A_comp <- A * (1 / 3)
  E_comp <- mean(c(max(E1, E2), E2)) * (2 / 3)
  grade <- A_comp + E_comp
  return(grade = grade)
}</pre>
```

```
CalculateGrade <- function(A, E1, E2) {
    A_comp <- A * (1 / 3)
    E_comp <- mean(c(max(E1, E2), E2)) * (2 / 3)

MakeRound <- function(x) {
    tmp <- trunc(2*x)
    if (x >= 5 && x < 5.5) {
      out <- 5
    } else if (x >= 5.5 & x < 5.75) {
      out <- 6
    } else if (2*x - tmp >= 0.5) {
      out <- ceiling(2*x)/2
    } else {
      out <- floor(2*x)/2
    }
    return(out)</pre>
```

```
grade <- MakeRound(x = A_comp + E_comp)
return(grade = grade)
}</pre>
```

2.5 SingTheBottleSong2() based on for

 \mathbf{a}

Rewrite SingTheBottleSong() into the function SingTheBottleSong2() that uses the for loop instead of the implicit lapply().

Answer:

```
SingTheBottleSong2 <- function(</pre>
  bottle_nr,
  address
) {
  header_song <- paste0("\n", bottle_nr, " Bottles of Beers \n\n")
  body_song <- vector(length(bottle_nr), mode = "list")</pre>
  each_bottle_nr <- bottle_nr:1</pre>
  for (bttls in each_bottle_nr) {
    body_song[[bottle_nr - bttls + 1]] <- CreateParagraphsBottlesBeer(bttls)</pre>
  body_song <- unlist(body_song)</pre>
  lyrics <- c(
    header_song,
    body_song
  cat(lyrics, sep = "", file = address)
  return(invisible(NULL))
}
```

b

Use system.time() to measure the time of both functions. Can you choose a number s.t. a distinction in time evaluation become visible?

Note that for the model answers we could not easily find such a distinction. The timing is quite instable. Around 1e5 bottles of beer the lapply() based function seems to be faster.

```
lyrics_addr <- "0_data/my_bottlesong_lyrics.txt"
if (file.exists(lyrics_addr)) file.remove(lyrics_addr)</pre>
```

```
## [1] TRUE
```

```
system.time(
  SingTheBottleSong(
    bottle_nr = 1e5,
    address = lyrics_addr
  )
)
##
      user system elapsed
     1.413
            0.303
##
                    1.745
file.remove(lyrics_addr)
## [1] TRUE
system.time(
  SingTheBottleSong2(
    bottle_nr = 1e5,
    address = lyrics_addr
  )
)
##
           system elapsed
      user
##
     1.479
             0.299
                     1.807
file.remove(lyrics_addr)
```

[1] TRUE

Now, it seems that the using the implicit loop with lapply and unlist()

When using a pre-definied function as FUN argument in lapply(), most of the times you have a speed gain over for or any other types of explicit loops in R.

2.6 Bubble sort

Bubble sort (<- click on Bubble sort) is a sorting algorithm that works by stepping through a vector to be sorted, comparing each pair of adjacent elements and swapping them if they are in the wrong order. The walk through the vector is repeated until no swaps are needed. The algrithm sorts a vector in ascending order.

Write a function BubbleSort() that implements the Bubble sort algorithm, and show that it can sort a vector (e.g. sample(50)) from its smallest value to its highest value.

Tip: it is common to see the following ERROR when swapping values, at least once:

```
x[2] <- x[3]
x[3] <- x[2]
```

Check for yourself why the above swap does not work.

Here's a function template to help you get started!

```
BubbleSort <- function(x, na.rm = TRUE) {</pre>
    # Description of function..
    # Args:
    # arg1: Description.
    # arg2: Description + default
    # Returns:
    # x sorted from smallest to highest values
    # perfrom some checks here...
    n <- length(x)</pre>
    repeat {
        swaps <- 0
        for (i in 1:(n-1)) {
            . . . . . .
        }
        if (!swaps)
            break
    }
    return(x)
}
```

```
BubbleSort <- function(x, na.rm = TRUE) {</pre>
    # Description of function..
    # Args:
    # arg1: Description.
    # arg2: Description + default
    # Returns:
    # x sorted from smallest to highest values
    # perfrom some checks here...
    n <- length(x)
    repeat {
        swaps <- 0
        for (i in 1:(n - 1)) {
          ###
          # solution part:
          if (x[i] > x[i + 1]) {
            x[c(i, i + 1)] \leftarrow x[c(i + 1, i)]
            swaps <- T
          }
```

```
}
    if (!swaps)
        break
}
    return(x)
}
BubbleSort(sample(50))
```

```
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 ## [24] 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 ## [47] 47 48 49 50
```

2.7 Coding a for loop into while or repeat

Take again (a quick) look at the following for loop example:

```
grades <- trunc(runif(100, 10, 100))/10 # uniform grades
n <- length(grades)
grd_rnd <- numeric(n)
grd_tbl <- numeric(10)
names(grd_tbl) <- 1:10

for(i in 1:n) {
   grade <- grades[i]
   tmp <- trunc(grade)
   if(grade - tmp >= 0.5) {
     grd_rnd[i] <- ceiling(grade)
   } else {
     grd_rnd[i] <- floor(grade)
   }
   grd_tbl[grd_rnd[i]] <- grd_tbl[grd_rnd[i]] + 1
}
grd_tbl</pre>
```

Recode this example into a while or repeat loop.

```
grades <- trunc(runif(100, 10, 100))/10 # uniform grades
n <- length(grades)
grd_rnd <- numeric(n)
grd_tbl <- numeric(10)
names(grd_tbl) <- 1:10
i <- 1
while (i <= n) {
    grade <- grades[i]
    tmp <- trunc(grade)
    if (grade - tmp >= 0.5) {
        grd_rnd[i] <- ceiling(grade)
    } else {
        grd_rnd[i] <- floor(grade)</pre>
```

```
}
grd_tbl[grd_rnd[i]] <- grd_tbl[grd_rnd[i]] + 1
# print(i)
i <- i + 1
}</pre>
```

Selfstudy Exercises

3.1 Vectorization: All roads lead to Rome... some are more efficient.

a.

Rewrite the following code into code which uses a [*]pply() function, and into a vectorized solution.

```
a <- numeric(5)
for (i in 1:5) a[i] <- i + 3
rm(i)
a</pre>
## [1] 4 5 6 7 8
```

Answer:

```
a <- sapply(1:5, function(i) i ) + 3
a</pre>
```

```
## [1] 4 5 6 7 8
```

```
a <- 1:5 + 3
a
```

```
## [1] 4 5 6 7 8
```

b.

The following code creates 2 vectors of length 100, with i.i.d. random values from a standard normal distribution (x) and from a normal distribution with mean = 2, and sd = 3.

inefficient code:

```
x <- y <- NULL
for (i in 1:100) {
    x[i] <- rnorm(1)
    y[i] <- rnorm(1, 2, 3)
}</pre>
```

Although you probably feel that it is not a wise decision to use a loop for this coding problem, try to recode this for (i in 1:100) {} without a growing object, And delete i at the end with rm() as in the lecture notes.

```
x <- y <- numeric(100)
for (i in 1:100) {
    x[i] <- rnorm(1)
    y[i] <- rnorm(1, 2, 3)
}
rm(i)</pre>
```

c.

Rewrite the code of exercise b twice by using one of the [*]apply functions (which can be tricky!) AND using a vectorized solution.

Answer using sapply:

```
x <- sapply(1:100, function(elem) rnorm(1))
y <- sapply(1:100, function(elem) rnorm(1))*3 + 2</pre>
```

Answer using a vectorized solution:

```
x <- rnorm(100)
y <- rnorm(100)*3 + 2
```

 \mathbf{d}

Take a look at the help file of the vectorized pmin function. We can use it to find the parallel minima of the elements of the objects x and y. See the following code:

```
head(pmin(x,y))
```

```
## [1] 0.03586826 0.05496134 -1.28159602 -0.18110238 -1.14483591 -1.45179128
```

Now try to rewrite this code example into your most cumbersome code possible by using the for loop and if() else {} control statements.

Answer (that will lower your grade if you would use it at your exam or assignments):

```
pminoutresult <- NULL # ugly long name
for (i in 1:100) {
   tmpx <- x[i] # unnecessary storage of object
   tmpy <- y[i] # unnecessary storage of object
   if ( tmpx < tmpy) {
      pminoutresult[i] <- tmpx # unnecessary assigning
   } else {
      pminoutresult[i] <- tmpy # unnecessary assigning
   }
   rm(list = c("tmpx", "tmpy")) # nice, but would not have been needed
}
rm(i) # not necessary if nowhere else used
all(pminoutresult == pmin(x,y))</pre>
```

```
## [1] TRUE
```

A better example using the for loop that would not have lost you points:

```
pmout <- numeric(100)
for (i in 1:100) {
  pmout[i] <- if (x[i] < y[i]) {
    x[i]</pre>
```

```
} else {
    y[i]
}

rm(i) # not necessary if nowhere else used
all(pmout == pmin(x,y))
```

[1] TRUE

Without the if-else statement:

```
pmout <- numeric(100)
for (i in 1:length(pmout)) {
   pmout[i] <- c(x[i], y[i])[(x[i] > y[i]) + 1]
}
rm(i) # not necessary if nowhere else used
all(pmout == pmin(x,y))
```

[1] TRUE

e.

We want to create the matrix with ij'th element equal to 3 times the ith element of 'x' minus 2 times the square of the jth element of 'y'

 $in {\it efficient \ code:}$

```
system.time(replicate(1e2, {
    rslt1e <- matrix(0, 100, 100)
    for (i in 1:100) {
        for (j in 1:100) {
            rslt1e[i, j] <- 3 * x[i] - 2 * y[j]^2
        }
    }
}</pre>
```

```
## user system elapsed
## 0.206 0.002 0.209
```

Carefully study what happens in the above code. Could you rewrite it using one (or even two) sapply implicit loop(s)? If yes which of the pieces of code is faster? Use the function system.time (and possibly replicate) to compare the codes on speed (check the help-file and internet, for example: http://stats.stackexchange.com/questions/3235/timing-functions-in-r).

Answer using two sapply loops:

```
system.time(replicate(1e2, {
   rslt1e <- sapply(y, function(y_elem) {
      sapply(x, function(x_elem) {
        3 * x_elem - 2 * y_elem^2
      })
   })
})</pre>
```

```
## user system elapsed
## 0.928 0.010 0.945
```

Answer using one sapply loop:

```
system.time(replicate(1e3, {
    rslt1e <- sapply(y, function(y_elem) {
        3*x - 2*y_elem 2 sapply + vectorization calculation
})
})
## user system elapsed
## 0.416 0.109 0.527</pre>
```

f.

Check the help file of the function outer(). Run the examples and get comfortable with the functions example(outer). If the examples do not make you feel comfortable on what the function does, explore the world wide web for more examples.

When comfortable with outer, rewrite the code problem 1e. into a vectorized solution by using the function outer

Answer:

```
rslt1e <- outer(x, y, function(x_elem, y_elem) {
   3*x_elem - 2*y_elem ^ 2
})</pre>
```

 $\mathbf{g}.$

Rewrite the following code into a vectorized solution that uses only the set.seed(42) and the function sample.

```
set.seed(42)
a1 <- replicate(12, sample(c(0,1), 1))</pre>
```

Answer:

```
set.seed(42)
a2 <- sample(c(0,1), 12, replace = TRUE)
all.equal(a1,a2)</pre>
```

```
## [1] TRUE
```

h.

Check the helpfile of ifelse(). Could you rewrite the lines that start from obtained_ects in the following code chunk into only one line of code using the ifelse() function?

Code to be rewritten:

```
set.seed(20181005)
ideal_grades <- rnorm(42, mean = 7, sd = 1.5)
obtained_ects <- character(length(ideal_grades))
for (student in 1:length(ideal_grades)) {
   if (ideal_grades[student] > 5.5) {
     obtained_ects[student] <- "yes"
   } else {
     obtained_ects[student] <- "no"
   }
}</pre>
```

Answer:

```
set.seed(20181005)
ideal_grades <- rnorm(42, 7, sd = 1.5)
obtained_ects <- ifelse(ideal_grades > 5.5, "yes", "no")
```

i.

Take a look at the source of ifelse() function. Although it is vectorized, it still uses an if statements, i.e.

```
if (any(test[ok]))
    ans[test & ok] <- rep(yes, length.out = length(ans))[test & ok]</pre>
```

Instead of using the ifelse() function, or if statments, only [brackets and filtering to create obtained_ects. Given that you have ideal_grades already in your workspace, you should be able to do so in two lines.

Answer:

```
obtained_ects <- rep(FALSE, length(ideal_grades))
obtained_ects[ideal_grades > 5.5] <- TRUE</pre>
```

j.

Could yo create a function that can replace SingTheBottlesSong() and SingTheBottlesSong2(), and does not use any implicit or explicit loops.

Hint: Check the live coding and the paste function

Answer:

The following lines of code should bring you towards an answers:

```
bttls <- suppressWarnings(as.integer(bttls))
if(is.na(bttls)) stop("bttls should be able to be coerced to integer")
if(bttls < 1) stop("bttls should be >= 1")

if(bttls > 2) {
   paragraphs_plural <- paste0(
   bttls:3, " bottles of beer on the wall, ",
   bttls:3, " bottles of beer. \n",</pre>
```

```
"Take one down, pass it around, ",
    (bttls - 1):2, " bottles of beer on the wall...\n\n"
} else {
 paragraphs_plural <- NULL</pre>
if(bttls > 1) {
 paragraph_2bttls <- paste0(</pre>
    2, " bottles of beer on the wall, ",
    2, " bottles of beer. \n",
    "Take one down, pass it around, ",
    1, " bottle of beer on the wall...\n\"
 )
}
paragraph_last <- paste0(</pre>
 1, " bottle of beer on the wall, ",
  1, " bottle of beer. \n",
  "Take one down, pass it around, ",
  "no more bottles of beer on the wall...\n\"
body_song <- c(paragraphs_plural, paragraph_2bttls, paragraph_last)</pre>
```

3.3 A Real (Non-Statistical) Programming Assignment

In this exercise we will make a data.frame containing functions from the package stats

 \mathbf{a}

Make a character vector called functs that contains all the objects from the package stats using the code ls(envir = as.environment("package:stats")). Take a careful look at what each individual part does (run each part separately). Try to think/guess why this works.

Answer:

```
functs <- ls(envir = as.environment("package:stats"))</pre>
```

 \mathbf{b}

Loop over the vector functs and use the function get to check for each of the objects from package stats if it is of the class function or not. Save only the function objects.

Answer:

```
which 不需要
```

 \mathbf{c}

We are now going to create a look-up table. We are going to make a data.frame that one can use to see which functions use a particular argument, and which arguments a particular function uses. We are going

to create an entry for *each* function, and denote with TRUE or FALSE whether an argument is used in that particular function or not.

Extract from the functions, whose names you've collected in functs, the arguments. Create an empty data.frame called stats.fncts and add an entry for each function in the stats package: the first column will contain the name, the following columns (as many columns as there are (unique) arguments in the stats package) will contain a TRUE or FALSE depending on whether the argument is used by the function or not.

Example result (using only two functions):

```
data.frame(
  fnames = c("anova", "aov"),
  contrasts = c(F, T),
  data = c(F, T),
  formula = c(F, T),
  object = c(T, F),
  projections = c(F, T),
  qr = c(F, T)
)
```

```
## fnames contrasts data formula object projections qr
## 1 anova FALSE FALSE FALSE TRUE FALSE FALSE
## 2 aov TRUE TRUE TRUE FALSE TRUE TRUE
```

Hint: as always, divide the task up in several smaller pieces. What do we need to do this task? We need all the functions from the stats package, we already have those in functs! We need to extract from each function its arguments (go to the slides to see how to do this!). We don't want a look-up table with duplicate entries, so we need a way to get all the unique(!) entries!

Answer:

```
# First get all the formals of all the functions, then use unique to get only the unique ones, then sor
stat.args <- sort(unique(names(unlist(lapply(functs, formals)))))

#First create the entire sized data.frame, to avoid growing an object:
stat.mat <- matrix(F, ncol = length(stat.args), nrow = length(functs))
#using a for loop check for each function, which arguments in the entire list of arguments, match argum
for (i in 1:length(functs)) {
    stat.mat[i, ] <- stat.args %in% names(formals(functs[i]))
}

#collect the function names and the other information into a single data.frame
stat.fncts <- data.frame(
    fnames = functs,
    stat.mat
)

#give nice names
colnames(stat.fncts) <- c("fnames", stat.args)
#str(stat.fncts)</pre>
```

 \mathbf{d}

There are functions that have the exact same arguments. How big is the largest group of functions that share the exact same arguments?

Hint: work out this exercise first on the first 10 rows of the datafram stat.fncts to see how to make the code for the big data set. Use a (statistical) programmer's strategy...

Answer:

[1] 41

[1] 41

```
grp_max <- stat.fncts[indx, ]
grp_max$fnames</pre>
```

```
[1] aggregate
                       ansari.test
                                     ar.burg
                                                    ar.yw
                                                                  as.hclust
##
    [6] as.stepfun
                                     bartlett.test biplot
                                                                  cor.test
                      as.ts
## [11] cycle
                       deltat
                                     density
                                                    diffinv
                                                                  end
## [16] fligner.test
                      formula
                                     frequency
                                                    ftable
                                                                  getCall
## [21] kernapply
                      kruskal.test
                                                    loadings
                                                                  model.tables
                                     lag
  [26] monthplot
                      mood.test
                                     naprint
                                                                  prcomp
                                                    ppr
## [31] princomp
                                                    screeplot
                       quantile
                                     reorder
                                                                  start
## [36] t.test
                       terms
                                     time
                                                    var.test
                                                                  wilcox.test
## [41] window
## 437 Levels: acf acf2AR add.scope add1 addmargins ... xtabs
```

3.4 Gradient Descent

There are various methods of estimating the parameters of a statistical model e.g. least squares or maximum likelihood estimation. In maximum likelihood estimation a maximum of a real function (the likelihood function) needs to be found by means of optimization. Various optimization methods exist for finding the minimum or maximum of a real function.

Gradient descent is an iterative algorithm that can be used to find a local minimum of a function. If a function f(x) is defined and differentiable in a point x_0 , then f(x) decreases fastest if one goes from x_0 in

the direction of the negative gradient of f at x_0 , denote - $f'(x_0)$. It follows that, if $x_1 = x_0 - \alpha f'(x_0)$ for small enough α , then $f(x_0) \ge f(x_1)$. A minimum of f(x) can be found by starting with x_0 and considering the sequence x_0, x_1, x_2, \ldots such that

$$x_{n+1} = x_n - \alpha f(x_n).$$

Write a function that implements the gradient descent algorithm for the polynomial

$$f(x) = x^2 - 3x + 2$$

using repeat and $\alpha = 0.01$. Use an auxiliary function for evaluating the first derivative f'(x). The while loop stops when the absolute difference between x_{n+1} and x_n is smaller than 1/1e4. Demonstrate the algorithm using the start values $x_0 = -2$ and $x_0 = 3$. Report the number of iterations and the x-value for which f(x) is minimal.

```
dvtv <- function(x){</pre>
  return(2*x - 3)
gradD <- function(x.init, alpha = 0.01, conv = 1/1e4){</pre>
  # x0 = , alpha = 0.01, conv = 1/1e4
  x <- x.init; it <- 0
  repeat {
    x <- x.init - alpha*dvtv(x.init)</pre>
    it <- it + 1
    \#cat("cost: ", x^2 - 3*x + 2, " it:", it, "\n")
    if ( abs(x.init - x) < conv) break</pre>
    x.init \leftarrow x
  list(x = x, it = it)
gradD(x.init = +3)
## $x
## [1] 1.504834
##
## $it
## [1] 284
gradD(x.init = -2)
## $x
## [1] 1.495172
##
## $it
## [1] 326
```