Computer lab 2

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Instructions

- $\bullet\,$ This lab should be done individually
- ullet It is ok to discuss the problems with each other, but to copy code is **NOT** allowed.
- \bullet The lab should be turned in through ${\bf LISAM}$ as a documented .R file.
- In the .R file two text element (objects) is mandatory:
 - name (your name)
 - liuid (your LiU-ID)
- \bullet The deadline for the lab be found at the course ${\bf webpage}$

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Automatic feedback with markmyassignment

As a complement to get fast feedback on your lab assignments the package markmyssignment has been written. This makes it possible to get automatic feedback on specified assignments, using any computer (altough internet access is required).

To install markmyassignment the package devtools is needed. To install devtools and markmyassignment just run the following code:

```
> install.packages("devtools")
> devtools::install_github("MansMeg/markmyassignment")
```

To get automatic feedback on your assignment you need an assignment path from your teacher. To set the assignment just run the following code

```
> library(markmyassignment)
> set_assignment("[assignment path]")
```

where [assignment path] äis the path given to you by your teacher.

To see which tasks that are included in the lab you can use the function show_tasks() in the following way:

```
> show_tasks()
```

To get automatic feedback you use the function mark_my_assignment(). To get feedback on all assignments just run the function.

```
> mark_my_assignment()
```

Remember that the functions need to be in the global environment in R.

You can also get feedback on specific tasks in the following way:

```
> mark_my_assignment(tasks="foo")
> mark_my_assignment(tasks=c("foo", "bar"))
```

It is also possible to correct an .R file in the following way:

```
> mark_my_assignment(mark_file = "[my search path to file]")
```

where [my search path to file] is the search path to the file to get feedback on.

Note! When an .R-file is checked, the global environment needs to be empty. Use rm(list=ls()) fto clean the global environment.

Chapter 1

Lab assignments

To use markmyassignment run the following code:

```
library(markmyassignment)

Loading required package: methods
Loading required package: yaml
Loading required package: testthat
Loading required package: httr

lab_path <-
"https://raw.githubusercontent.com/MansMeg/AdvRCourse/master/Labs/Tests/lab2.yml"
set_assignment(lab_path)

Assignment set:
Lab2 : Advanced R programming, computer lab 2</pre>
```

1.1 Conditional statements

1.1.1 sheldon_game(player1, player2)

In the series The big bang theory the character Sheldon comes up with a new version the game Rock-paper-scissors, called Rock-paper-scissors-lizard-spock. The detailed rules of the game can be found here.

Create a function you call sheldon_game() that should take two arguments, player1 and player2. Each player should choose one of the choices rock, paper, scissors, lizard or spock. If another choice is made, the function should be stopped (with the stop() or stopifnot() function).

The function should return either "Player 1 wins!", "Player 2 wins!" or "Draw!".

Tip! This can be solved by using the cyclical structure of the game and the modulus operator.

```
> sheldon_game("lizard", "spock")
[1] "Player 1 wins!"
> sheldon_game("rock", "paper")
[1] "Player 2 wins!"
```

1.2 for loops

1.2.1 my_moving_median()

Create a function that will create a rolling median from a given numeric vector. Call the function my_moving_median(). The functions should accept a given numeric vector x, and a numeric scalar n.

The function should check if the arguments is a numeric vector (x) and a numeric scalar (n) and if not, it should stop the function (with the stop() or stopifnot() function). Otherwise the function should return the following vector:

```
y_t = \text{median}(x_t, \dots, x_{t-n})
```

where y_t is the tth elementet. It should also be possible to send argument to the median() function (as na.rm).

```
my_moving_median(x = 1:10, n=2)

[1] 2 3 4 5 6 7 8 9

my_moving_median(x = 5:15, n=4)

[1] 7 8 9 10 11 12 13

my_moving_median(x = c(5,1,2,NA,2,5,6,8,9,9), n=2)

[1] 2 NA NA NA 5 6 8 9

my_moving_median(x = c(5,1,2,NA,2,5,6,8,9,9), n=2, na.rm=TRUE)

[1] 2.0 1.5 2.0 3.5 5.0 6.0 8.0 9.0
```

1.2.2 for_mult_table()

Write a function called for mult_table() with arguments from and to. Both arguments can be assumed to be positive integers. The function should return a multiplication table as a matrix from the argument from to the argument to. The column names and row names should contain the factors specified by the arguments.

The function should stop (using stop()) if the arguments is not numeric scalars. See an example below.

1.2.3 * cor_matrix()

Create a function called <code>cor_matrix()</code>. It should take a numeric <code>data.frame</code> with numerical variables of an arbitrary size as an argument <code>X</code> and calculate the correlation matrix for the given variables. If another object than an data.frame is used as argument, the function should return an error.

In this exercise you are not allowed to use the functions var(), sd() or cor() in your function. See an example below.

```
data(iris)
cor_matrix(iris[,1:4])
         [,1]
                  [,2]
                           [,3]
                                     [,4]
[1,] 1.00000 -0.11757 0.87175
                                0.81794
[2,] -0.11757 1.00000 -0.42844 -0.36613
[3,] 0.87175 -0.42844
                       1.00000 0.96287
[4,] 0.81794 -0.36613 0.96287
data(faithful)
cor_matrix(faithful)
        [,1]
                [,2]
[1,] 1.00000 0.90081
[2,] 0.90081 1.00000
```

1.3 while loops

1.3.1 find_cumsum()

Create a functions that takes two arguments x and $find_sum$. The function should calculate the cumulative sum of the numeric vector x and stop when the cumulative sume is greater than $find_sum$ or if the vector x is traversed and return this value. The function should assert that x is a numeric vector and is a numeric scaler (with stop()) or stopifnot()).

It is not allowed to use a for loop in this assignment.

```
find_cumsum(x=1:100, find_sum=500)
[1] 528
find_cumsum(x=1:10, find_sum=1000)
[1] 55
```

1.3.2 while_mult_table()

Create a functions that calulates a given mutiplication table similar to the exercise 1.2.2 but now you shoul use a (nested) while loop instead of a for-loop. The result should be identical in all other respect.

```
while_mult_table(from = 3, to = 5)
   3 4 5
3 9 12 15
4 12 16 20
5 15 20 25
while_mult_table(from = 7, to = 12)
    7 8
          9
             10 11 12
  49 56
          63
             70
                  77
                      84
  56 64
         72
             80
                  88
                     96
  63 72
         81
             90
                 99 108
         90 100 110 120
10 70 80
11 77 88 99 110 121 132
12 84 96 108 120 132 144
```

1.3.3 * trial_division_factorization()

Write a function that can factorize any integer number x into its prime components. The function should return a vector of all the prime factors as a numeric vector. The idea is to simply test all integers (or prime numbers) from 2 to \sqrt{n} . For more detailed information on how to factorize primes using the trial division algorithm can be found here.

```
trial_division_factorization(x = 2^3 * 13 * 17 * 31)
[1] 2 2 2 13 17 31
trial_division_factorization(x = 47 * 91 * 97)
[1] 7 13 47 97
```

1.4 repeat and loop controls

1.4.1 repeat_find_cumsum()

Implement a function similar to find_cumsum() in 1.3.1 but using repeat{} instead of a while loop.

```
repeat_find_cumsum(x=1:100, find_sum=500)
[1] 528
repeat_find_cumsum(x=1:10, find_sum=1000)
[1] 55
```

1.4.2 repeat_my_moving_median()

Implement a function similar to my_moving_average() in 1.3.1 but using repeat{} instead of a while loop.

```
repeat_my_moving_median(x = 1:10, n=2)
[1] 2 3 4 5 6 7 8 9
repeat_my_moving_median(x = 5:15, n=4)
[1] 7 8 9 10 11 12 13
repeat_my_moving_median(x = c(5,1,2,NA,2,5,6,8,9,9), n=2)
[1] 2 NA NA NA 5 6 8 9
```

1.5 Environment

1.5.1 in_environment()

Create a function called <code>in_environment()</code> that return the contents of an environment as a text vector. The function shoulde take the argument <code>env</code> that takes an environment name as a text element.

```
env <- search()[length(search())]
env
[1] "package:base"</pre>
```

1.5.2 * where()

This function is similar to the where() function described in Advanced R. But unlike the implementation in the book you are not allowed to use parent.env() to traverse the search path. The function should take a name and return the name of the environment where the object is found by following the search path.

If the function is not found, the function should return the text element "[fun] not found!". Assert that fun is a character vector of length 1.

```
where(fun = "sd")
[1] "package:stats"
where(fun = "read.table")
[1] "package:utils"
where(fun = "non_existant_function")
[1] "non_existant_function not found!"
```

1.6 Functionals

1.6.1 cov()

This exercise is taken from the Advanced R book. Use lapply() and an anonymous function to find the coefficient of variation (the standard deviation divided by the mean) for all columns in a given data.frame. The function should return a vector of the coefficients of variation.

Assert that X is a data.frame.

1.7 Clojures

1.7.1 moment()

This exercise has been taken from the Advanced R book. Create a function that creates a function for the ith central moment for a given variable. Definitions of the central moments can be found **here**.

```
m1 <- moment(i=1)
m2 <- moment(i=2)
m1(1:100)</pre>
```

```
[1] 0
m2(1:100)
[1] 833.25
```

1.7.2 * mcmc_counter_factory()

In markov chain monte carlo methods we often want to use counters for the sampling algorithms. Create a counter factory that creats a clojure with a counter. The factory should take two arguments, thin and burnin. Each counter should contain three objects:

- iteration (numeric) scalar
- store_sample (boolean) scalar
- samples (numeric) scalar

Assume that the counter is called in each iteration so in each call of the counter (without argument) should bump up the iterations object.

The store_sample boolean indicates whether the current iterations should be stored. Iterations after the number of burnin iterations should be stored every thin iteration. So if we run 20 iterations with burnin = 10 and thin = 3 we would store iteration $\{13, 16, 19\}$. The samples scalar should keep track of the number of stored iterations (three as the example). In each iteration the values should be returned as a list.

Assert that burnin is a nonnegative value and that thin is a positive value. Below is an example.

```
mcmccnt <- mcmc_counter_factory(burnin = 3, thin = 2)</pre>
mcmccnt()
[[1]]
[1] 1
[[2]]
[1] FALSE
[[3]]
[1] 0
mcmccnt()
[[1]]
[1] 2
[[2]]
[1] FALSE
[[3]]
[1] 0
mcmccnt()
[[1]]
[1] 3
[[2]]
[1] FALSE
[[3]]
[1] 0
```

```
mcmccnt()
[[1]]
[1] 4
[[2]]
[1] FALSE
[[3]]
[1] 0
mcmccnt()
[[1]]
[1] 5
[[2]]
[1] TRUE
[[3]]
[1] 1
mcmccnt()
[[1]]
[1] 6
[[2]]
[1] FALSE
[[3]]
[1] 1
```

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