# CeDoSIA SS2020 - Exercise Sheet 1: Introduction to R

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#### 1 Vectors

First, create three named numeric vectors of size 10, 11 and 12 respectively in the following manner:

- One vector with the "colon" approach: from:to
- One vector with the seq() function: seq(from, to)
- And one vector with the seq() function and the by argument: seq(from, to, by)

For easier naming you can use the vector letters or LETTERS which contain the latin alphabet in small and capital, respectively. In order to select specific letters just use e.g. letters[1:4] to get the first four letters. Check their types. What is the outcome? Where do you think the difference comes from?

Then combine all three vectors in a list. Check the attributes of the vectors and the list. What is the difference and why?

**Hint:** If list elements have no names, we can access them with the double brackets and an index, e.g.  $my_list[[1]]$ 

```
# Answer :
# A. Create vectors
vector.1 <- 1:10
names(vector.1) <- letters[vector.1]</pre>
vector.2 \leftarrow seq(1, 11)
names(vector.2) <- letters[vector.2]</pre>
vector.3 < seq(1, 12, by = 1)
names(vector.3) <- letters[vector.3]</pre>
typeof(vector.1)
## [1] "integer"
typeof(vector.2)
## [1] "integer"
typeof(vector.3)
## [1] "double"
# B. Combine in a list
awesome.list <- list(vector.1, vector.2, vector.3)</pre>
attributes(vector.1)
## $names
## [1] "a" "b" "c" "d" "e" "f" "q" "h" "i" "i"
attributes(vector.2)
## $names
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k"
attributes(vector.3)
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" "l"
attributes(awesome.list)
## NULL
attributes(awesome.list[[1]])
```

```
## $names
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j"

## Why is the last vector of type double and not integer?
## By default seq returns integers from:to. But the `by`
## parameter returns always doubles

## myList got no names since we did not assign any compared to our vectors
```

#### 2 Factors

```
f1 <- factor(letters)
levels(f1) <- rev(levels(f1))
f2 <- rev(factor(letters))
f3 <- factor(letters, levels = rev(letters))</pre>
```

The function rev reverses the order of an order-able object. What is the difference between f1, f2 and f3? Why?

```
# Answer :
f1 <- factor(letters)</pre>
levels(f1) <- rev(levels(f1))</pre>
# f1 goes from z - a, but the underlying encoding goes from z = 1 to a = 26
# We create the vector with the letters a to z and the mapped integer
\# structure 1 to 26. THEN we reverse the levels = the mapping. As 1 becomes z
# and a becomes 26 the letters are mapped back to the unchanged integer
# structure and hence reversed.
## [1] zyxwvutsrqponmlkjihgfedcba
## Levels: z y x w v u t s r q p o n m l k j i h g f e d c b a
f2 <- rev(factor(letters))</pre>
# f2 goes from z - a, but the underlying encoding goes from a = 1 to z = 26
# We create the vector with the letters a to z and the mapped integer
# structure 1 to 26. Then we reverse the vector, i.e. the underlying integers,
# hence the vector gets reversed, but not the levels.
f2
## [1] zyxwvutsrqponmlkjihgfedcba
## Levels: a b c d e f g h i j k l m n o p q r s t u v w x y z
f3 <- factor(letters, levels = rev(letters))</pre>
\# f3 goes from a - z, but the underlying encoding goes from z = 1 to a = 26.
# We create the vector with the letters a to z BUT the mapped integer
# structure 26 to 1. Hence the levels but not the vector are reversed.
## [1] abcdefghijklmnopqrstuvwxyz
## Levels: z y x w v u t s r q p o n m l k j i h g f e d c b a
```

```
# Reversing f3 will give f1
rev(f3)
## [1] z y x w v u t s r q p o n m l k j i h g f e d c b a
## Levels: z y x w v u t s r q p o n m l k j i h g f e d c b a
```

# 3 Computation on matrices

Create a 10 by 5 matrix which contains the the numbers from 1 to 50 column-wise. Name the rows as 'row\_n' and columns as 'col\_n'. Compute the mean and sum of each row and column. Add vector seq(60,100,10) as another row to the matrix.

Generate another matrix with the same dimensions, containing random numbers between 1 and 100. Subtract this matrix from the first one.

Plot the covariance matrix of the columns of the resulting matrix with spearman correlation coefficients.

**Hint:** Check out functions paste0(), colMeans(), rowMeans(), colSums(), rowSums(), sample(), cor() and corrplot() (in package 'corrplot')

```
# Answer :
m <- matrix(1:50, ncol = 5)</pre>
rownames(m) <- paste0("row_", 1:nrow(m))</pre>
colnames(m) <- paste0("col_", 1:ncol(m))</pre>
colMeans(m)
## col_1 col_2 col_3 col_4 col_5
## 5.5 15.5 25.5 35.5 45.5
rowMeans(m)
## row_1 row_2 row_3 row_4 row_5 row_6 row_7 row_8 row_9 row_10
      21
           22 23 24
                              25 26
                                           27
                                                     28
                                                           29
colSums(m)
## col_1 col_2 col_3 col_4 col_5
   55 155 255 355 455
rowSums(m)
## row_1 row_2 row_3 row_4 row_5 row_6 row_7 row_8 row_9 row_10
          110
                115 120
                              125
                                    130
                                           135
                                                  140
                                                        145 150
m < - rbind(m, seq(60, 100, 10))
n <- matrix(sample(1:100, length(m), replace=T), ncol = ncol(m))</pre>
M <- cor(m, method = "spearman")</pre>
#corrplot::corrplot(M, method = "number")
```

#### 4 Data frame manipulation

Create a 3 by 4 matrix that contains the numbers 1 to 12 and then convert it into a data frame. Assign zero to the elements at row 2 which are greater than 4. Set the rownames to "row1", "row2", "row3" and column names to "col1", "col2", "col3" and "col4". Assign 0 to all elements in columns "col3" and "col4". Add a new column named "Letters" with values c(``A'', ``B'', ``C''). Inspect the structure of the data frame.

```
#Answer :
x <- matrix(1:12,3,4)
x <- as.data.frame(x)
x
## V1 V2 V3 V4
## 1  1  4  7 10
## 2  2  5  8 11
## 3  3  6  9 12
x[2, x[2,]>4] <- 0
rownames(x) <- paste0("row",1:3)
colnames(x) <- paste0("col",1:4)
x[,paste0("col",3:4)] <- 0
x$Letters <- c("A","B","C")</pre>
```

## 5 Data frame operations

Compute the number of women who survived the Titanic. Start by loading the data into a data frame using the following command:

```
tab <- read.csv("../../extdata/titanic.csv")
## Error in file(file, "rt"): cannot open the connection

head(tab)
## Error in head(tab): object 'tab' not found
# number of females survived the Titanic
nrow(tab[tab$sex == "female" & tab$survived == 1,])
## Error in nrow(tab[tab$sex == "female" & tab$survived == 1, ]): object 'tab' not found
# total number of passangers
nrow(tab)
## Error in nrow(tab): object 'tab' not found</pre>
```

### 6 Looping and writing your own functions

Write a function named "generateDataFrameSummary" which takes a data frame as input and outputs the medians of the rows and columns (NA values are discarded), and number of NA values in each row and column as a list.

```
#Answer :
generateDataFrameSummary <- function(inputDF){</pre>
 return(list( rowMedians = apply(inputDF, 1, median, na.rm=T),
           colMedians = apply(inputDF, 2, median, na.rm=T),
           rowNANum = apply(inputDF, 1, function(x){ sum(is.na(x)) }),
           colNANum = apply(inputDF, 2, function(x){ sum(is.na(x)) })))
}
# test the function
df <- as.data.frame(matrix(rnorm(1000), nrow=50))</pre>
df[sample(c(T, F), nrow(df), replace=TRUE),
 sample(c(T, F), ncol(df), replace=TRUE)] <- NA</pre>
head(df)
         V1
##
                 V2
                                      V5
                                               V6
                          V3
                                  V4
## 1 -0.1247252 -0.03369697 -1.63846635 1.1846881 0.4077047 -1.566437 0.2125627
## 2 NA NA -0.23444217 NA
                                      NA NA -1.5233418
## 3
        NA
                 NA 0.09825175
                                  NA
                                           NA
                                                  NA 0.2689837
                 NA 0.79159661
        NA
                                  NA
                                           NA
                                                  NA -0.7469679
## 5 -0.3251337 -1.86513161 -0.82402457 -0.5296698 -1.8333851 -1.387791 0.3189131
        NA NA -0.34129167 NA
                                          NA NA -1.3462898
##
         V8
                 V9 V10
                                  V11
                                          V12
                                                  V13
## 1 1.24074196 -0.4362921 0.45200986 0.40081830 -0.1770069 0.02173649
## 2 1.46491260 NA 1.01327830 -0.39540103 -0.7386880
## 3 -0.05691632
                 NA -0.09312204 -0.79214624 0.1943725
## 4 0.18395047
                 NA 0.91309656 0.30116232 1.5539355
## 5 -0.41124726 -0.5265563 0.33640203 -0.77532851 -0.7823704 0.78016726
## 6 1.12799626 NA -0.06148710 0.06715794 0.6334482 NA
    V14
               V15 V16 V17 V18
## 1 -0.4892880 0.8505296 0.06721411 0.91042305 -0.54559218 1.120220 1.4093308
## 2 -0.3530381 -1.9535444 0.73151091 -0.08363822
                                      NA
                                                 NA -1.7891092
## 3 0.9920929 -0.7307767 0.66080007 -0.93735494
                                           NA
                                                  NA -0.9820880
## 4 -1.4683217 -1.4955113 -1.67329594 1.07616010
                                          NA
                                                  NA 0.1954109
## 5 1.3293946 2.3617666 -0.29409099 -0.66651539 -0.03642664 0.617565 -0.8845018
## 6 0.7857377 -1.3475107 0.45137769 1.52312828 NA NA 1.4605889
generateDataFrameSummary(df)
## $rowMedians
## [1] 0.139888400 -0.353038054 -0.056916323 0.195410866 -0.468901757
## [6] 0.451377692 -0.255290391 0.073613852 -0.059536615 -0.342390893
## [26] 0.032161261 -0.163713527 -0.007021512 0.522737486 -0.289883752
## [41] -0.033901956 -0.546574424 -0.262689241 0.213831060 -0.209326882
## $colMedians
```

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```
## V1 V2 V3 V4 V5 V6
## 0.04109641 0.02235865 -0.22179269 -0.34394111 -0.78903514 -0.62003420
  V7 V8 V9 V10 V11 V12
## -0.09087695 -0.02330022 -0.27192588 -0.00485598 -0.09377590 0.45040919
## V13 V14 V15 V16 V17
\#\# 0.03812900 -0.15934218 0.08750901 0.01058144 0.06757430 0.34593335
  V19 V20
## 0.11126602 -0.23716149
##
## $rowNANum
## [39] 0 9 9 9 0 9 9 0 9 0 0 9
##
## $colNANum
## V1 V2 V3 V4 V5 V6 V7 V8 V9 V10 V11 V12 V13 V14 V15 V16 V17 V18 V19 V20
```