

# Applied Time Series

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# Factors

- A factor is a statistical data type used to store categorical variables, i.e. the variables that can have a limited number of values (like gender, for example).
- R works with ordered and unordered factors.
- We create factors in R by using the `factor()` function .

```
student_nationalities <- c("Russian", "Polish", "French",  
                           "American", "Russian",  
                           "Italian", "French")  
nationalities <- factor(student_nationalities)  
nationalities
```

```
## [1] Russian Polish French American Russian Italian  
## Levels: American French Italian Polish Russian
```

# Factors

- To create a vector of ordered factors we must use `ordered` and `levels` options as arguments of the `factor` function.

```
running_speed <- c("Low", "Fast", "Medium", "Fast",  
                  "Low", "Medium", "Fast")  
speed_levels <- factor(running_speed, ordered = TRUE,  
                      levels = c("Low", "Medium", "Fast"))  
speed_levels
```

```
## [1] Low    Fast   Medium Fast   Low    Medium Fast  
## Levels: Low < Medium < Fast
```

```
speed_levels[1] > speed_levels[4]
```

```
## [1] FALSE
```

# Lists

- A list in R is a general form of vector that gathers different objects under one name in an ordered way.
- These objects are called components and they can be matrices, vectors, data frames, even other lists, etc. No interdependence among the list components is necessary.
- We can create a list with the `list()` function.

```
vect1 <- 8:-1  
matr1 <- matrix(1:4, nrow = 2)  
df1 <- people[1:2,]  
new_list <- list(vect1, matr1, df1)  
new_list
```

# Lists: an example

```
## [[1]]
##   [1]  8  7  6  5  4  3  2  1  0 -1
##
## [[2]]
##      [,1] [,2]
## [1,]    1    3
## [2,]    2    4
##
## [[3]]
##      name gender height weight single
## 1  Ivanov   male   175   83.5   TRUE
## 2 Petrova female   164   58.5   TRUE
```

# Lists: components naming

- We can easily name the elements of the list when we create it.

```
new_list <- list(new_vector = vect1, new_matrix = matr1,  
                new_data_frame = df1)
```

- If we want to name the components of the list after the list was created, we can use the `names()` command as we do with vectors.

```
new_list <- list(vect1, matr1, df1)  
names(new_list) <- c("new_vector", "new_matrix",  
                    "new_data_frame")
```

## Lists: components naming(2)

```
new_list
```

```
## $new_vector
```

```
## [1] 8 7 6 5 4 3 2 1 0 -1
```

```
##
```

```
## $new_matrix
```

```
##      [,1] [,2]
```

```
## [1,]    1    3
```

```
## [2,]    2    4
```

```
##
```

```
## $new_data_frame
```

```
##      name gender height weight single
```

```
## 1  Ivanov   male   175   83.5   TRUE
```

```
## 2 Petrova female   164   58.5   TRUE
```

# Selecting a component from a list

- To select a component from a list using an index, we can use double square brackets. All three ways are equivalent:

```
new_list[[1]]
```

```
## [1] 8 7 6 5 4 3 2 1 0 -1
```

```
new_list[["new_vector"]]
```

```
## [1] 8 7 6 5 4 3 2 1 0 -1
```

```
new_list$new_vector
```

```
## [1] 8 7 6 5 4 3 2 1 0 -1
```



# Selecting an element from a list

- To select a particular element from the list we use squared brackets exactly as we do with vectors, matrices and data frames.

```
new_list[[1]][7]
```

```
## [1] 2
```

```
new_list[["new_matrix"]][2,2]
```

```
## [1] 4
```

```
new_list$new_data_frame[2,"name"]
```

```
## [1] Petrova
```

```
## Levels: Ivanov Petrova Vasechkin
```

# Loops

Loops are used for repeated calculations. The R syntax allows different ways to do a loop. Here are two of them.

- If we know the exact number of iterations to be done, we use the `for()` function.

```
for (i in 1:2) {  
  A <- diag(i)  
  print(A)  
}
```

```
##      [,1]  
## [1,]    1  
##      [,1] [,2]  
## [1,]    1    0  
## [2,]    0    1
```

## Loops(2)

- If we want the iterations to continue until a condition is met, we use the `while()` function

```
z <- 0
while (z < 4) {
  ran <- abs(rnorm(1))
  z <- z + ran
  print(c(ran, z))
}
```

```
## [1] 0.5797943 0.5797943
## [1] 0.9099121 1.4897064
## [1] 1.206160 2.695866
## [1] 1.073483 3.769350
## [1] 1.101732 4.871082
```

# Conditions

- For instructing the program to do a set of operations only if a certain condition is satisfied, we make use of the function `if()` (and `else` if necessary). Several conditions can be combined with `&&` (and) and `||` (or).

```
a <- rnorm(1)
a
```

```
## [1] -1.410146
```

```
if (a > 0) {
  print('positive')
} else {
  print('negative')
}
```

```
## [1] "negative"
```

# Functions

- Functions are objects in R that when they are called, execute a certain sequence of operations. The functions are also stored in the workspace. Writing a function is a good way to expand what R can do.
- R allows the user to create his or her own functions in addition to those that already exist in the baseline software.
- A function can be defined with the code of the following form:  
name <- function(arg\_1, arg\_2,...) expression

```
power2 <- function(x) {  
  y <- x^2  
  return(y)  
}  
z <- power2(4)  
z
```

```
## [1] 16
```

# Packages: general information

- R comes with several standard packages but not all of them are automatically attached.
- Besides them, there are many contributed packages, written by people all over the world.
- Packages contain functions, the documentation explaining how to use them, and sample data.
- We can download packages from CRAN and github. Probably, there is a package written by one of the hundreds volunteers that happens to be useful for your task in hand. Install the package and attach it to your current session.

# Packages: installing

- To install a package we go Tools → Install Packages. Alternatively, we can type `install.packages("packagename")` in the command prompt.

```
install.packages("tseries")
```

- Packages are sometimes interdependent. If we install one package intentionally, several others may be installed automatically.

# Packages: attaching

- After downloading a package we have to attach it with the `library()` command (the `require()` function is also possible but not recommended). The name of the package can be written with or without quotes when the library function is called.

```
library(tseries)
```

- To see a list of attached packages we use the `search()` function with no arguments.
- Functions in different packages may happen to have the same names. To avoid confusion we use the following syntax to apply a function from a particular package: `packagename :: functionname`

```
stats :: lag
```



# Datasets

- The package `datasets` contains preloaded data bases. To see a list of them we use the `data()` function with no arguments.
- Other packages can contain datasets as well. To access them we use the same function but with arguments telling R the name of the base and the name of the package.

```
data(NelPlo, package = "tseries")
```

- If a package is attached, the datasets contained in this package become available to the user.
- To see a list of datasets in a package we use the `data` function with an only argument telling R the name of the package.

```
data(package = "tseries")
```

# Editing data

- To edit data in a separate spreadsheet-like window we use the `edit(objectname)` command.

```
people2 <- edit(people)
```

- By clicking the object in the workspace we can just view it but this does not allow to edit it. This is equivalent to the command `View(objectname)`.

```
View(people)
```

# Our first packages

- Install tidyverse package. This is a collection of R packages designed for data analyses. The packages included in the tidyverse set are:
  - \* ``ggplot2`` - to get a nice visualization of data
  - \* ``dplyr`` - to manipulate data quickly and efficiently
  - \* ``readr`` - to read rectangular data (like csv).
  - \* and three other packages
- Each package from this collection may also require other packages which are also installed automatically.
- Do not forget to attach the tidyverse package.

```
library(tidyverse)
```

# Working with time series

The algorithm of working with times series models.

- 1 Importing data that we will do with `readr` or `readxl` packages.
- 2 Data visualisation that we will do with `ggplot2` package (if necessary).
- 3 Data manipulation that we will do with `dplyr` package treating series as data frames.
- 4 Transformation data frames into special time series (ts) format if necessary (some packages work only with ts data).
- 5 Estimation and evaluation of models with special packages developed specially for a certain class of models.

# Entering data from keyboard

- If a dataset is small we just enter it with `c()` or `'data.frame()'` commands and then edit them if necessary. We also can create an empty data frame and then edit it.

```
dataset <- data.frame()  
dataset <- edit(dataset)
```

# Importing data from an external source

- If a dataset is big we can import the data by clicking 'Import Dataset' in the Environment menu and choosing the source. The reading commands are from packages that are attached automatically.
- The same reading commands may be typed by the user as well. Do not forget to attach the respective package first.

```
library(readr) # not necessary if the `tidyverse` package  
                #is already attached.  
data1 <- read_csv("mydata.csv")
```

- There are also importing commands that do not require any packages to be attached (the 'utils' package where they belong to is attached automatically).

```
data2 <- read.csv("mydata.csv", stringsAsFactors = FALSE)
```