

# Coding Hours

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Introduction to R  
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# What is R?

- Programming language & free software environment
- Statistical computing & graphics
- First appeared: August 1993
- Written in C, Fortran and R itself

# What can R be used for?

- Data manipulation
- Data visualization
- Analyzing data
  - Advanced statistical methods
  - Machine learning
- And much more!

# Why you should learn R for data science?

- R is easy to learn
- Very powerful
- Popular in academia
- Excellent package support
- Open source

# RStudio

- Download RStudio or use RStudio Cloud
- Available on all computers at the EUR
- Preferably work on a Windows computer

The screenshot displays the RStudio environment with the following components:

- Source Editor:** Contains an R script with the following code:
 

```
1
2 rm(list = ls())
3 N <- 1000
4 u <- rnorm(N)
5 x1 <- -2 + rnorm(N)
6 x2 <- 1 + x1 + rnorm(N)
7 y <- 1 + x1 + x2 + u
8 r1 <- ln(y ~ x1 + x2)
9
10 |
```
- Console:** Shows the execution of the script with the following output:
 

```
10:1 (Top Level) >
Tapez <Entrée> pour voir le graphique suivant :
Tapez <Entrée> pour voir le graphique suivant :
Tapez <Entrée> pour voir le graphique suivant :
>
> ?ln
> rm(list = ls())
> N <- 1000
> u <- rnorm(N)
> x1 <- -2 + rnorm(N)
> x2 <- 1 + x1 + rnorm(N)
> y <- 1 + x1 + x2 + u
> r1 <- ln(y ~ x1 + x2)
>
```
- Workspace:** Lists the objects created in the environment:
 

Object	Class
N	1000
r1	lm[12]
u	numeric[1000]
x1	numeric[1000]
x2	numeric[1000]
y	numeric[1000]
- Help Pane:** Displays the documentation for the `lm` function, titled "Fitting Linear Models".
 

**Description**

`lm` is used to fit linear models. It can be used to carry out regression, single stratum analysis of variance and analysis of covariance (although `ancov` may provide a more convenient interface for these).

**Usage**

```
lm(formula, data, subset, weights,
    method = "qr", model = TRUE, x =
    singular.ok = TRUE, contrasts =
```

# Learning R with DataCamp

- DataCamp is a website that offers courses for learning R, Python and SQL
- Complete the Introduction to R tutorial



## Continuing the Journey with r-statistics

- **r-statistics** is an educational resource for those seeking knowledge related to machine learning and statistical computing in R.
- Read the **R Tutorial** on r-statistics.
- Continue with the articles about **Statistical Tests** and **Linear Regressions** in R

# A New Way of Testing

1. Import the MASS package
2. Find the maximum price of all cars in the Cars93 data in the MASS package.
3. Check whether the mean of the horsepower of the Cars93 data significantly different from 140.
4. Do the same as for the two questions above for the data on the cars with airbags.
5. Do the same as for the last three questions but now for the cars with more than 4 cylinders that cost less than 20.

# Your First Linear Regression in R

- Consider the *Boston* data from the MASS package.
- Scatter plot the following columns:
  - crim
  - zn
  - indus
  - age
- Regress crim on zn, indus and age.
- Add a dummy variable for dis larger than 5 in the regression above.
- Plot the results of your regression together with crim in one figure.

## Reinvent OLS in R (1/3)

1. Write a *Backsubstitution*(**A**, **b**) that takes a  $m \times n$  matrix **A** of full rank and a  $m \times 1$  vector **b** and returns the unique  $n \times 1$  vector **x** which solves  $\mathbf{Ax} = \mathbf{b}$  using backsubstitution. As a hint, the way to solve it is

$$x_n = b_n/a_{nn}, \quad x_i = \left( b_i - \sum_{j>i} a_{ij}x_j \right) / a_{ii}, \quad i \in \{1, 2, \dots, n-1\}.$$

It might be easier to first define

$$s = \sum_{j>i} a_{ij}x_j,$$

and for all  $x_n, x_{n-1}, \dots, x_1$  use the same formula for solving.

## Reinvent OLS in R (2/3)

1. Generate 20 observations from  $y = \mathbf{X}\beta + \sigma\varepsilon$  with  $\beta = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ ,  $\mathbf{X} = [1 \ u_1 \ u_2]$ ,  
 $u_i \sim \mathcal{N}(0, 1)$  for  $i \in \{1, 2\}$ ,  $\varepsilon \in \mathcal{N}(0, 1)$ ,  $\sigma = 0.25$ .

2. Estimate the OLS estimator  $\hat{\beta}$  using
  - 2.1 using direct matrix multiplication;
  - 2.2 using your elimination + backsubstitution, noting that

$$\mathbf{b} \equiv \mathbf{X}'\mathbf{y} = \mathbf{X}'\mathbf{X}\hat{\beta} \equiv \mathbf{A}\mathbf{x};$$

- 2.3 using a prepackaged function.

Write three functions *EstimateMM()*, *EstimateEB()*, *EstimatePF()*, which implement the three options, and check that the results indeed are the same.

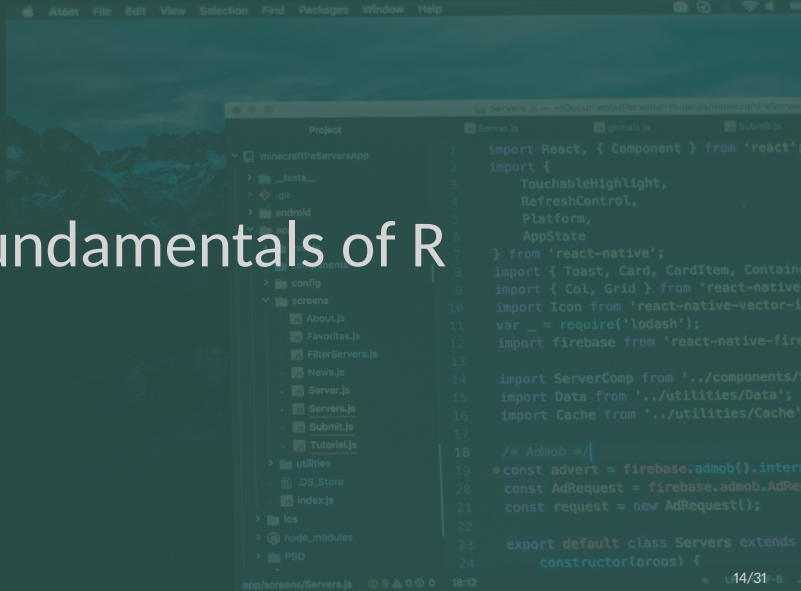
## Reinvent OLS in R (3/3)

3. Eventually, we might also be interested in

$$\begin{aligned} \mathbf{e} &= \mathbf{y} - \mathbf{X}\hat{\beta}, & \hat{\sigma}^2 &= \frac{1}{n-k} \mathbf{e}'\mathbf{e} = \frac{1}{n-k} \sum_{i=1}^n e_i^2, \\ \hat{\Sigma} &= \hat{\sigma}^2 (\mathbf{X}'\mathbf{X})^{-1}, & s(\hat{\beta}) &= \text{diag}(\hat{\Sigma})^{1/2}, \end{aligned}$$

with  $n$  and  $k$  the size of  $\mathbf{y}$  and  $\beta$ , respectively. Also the  $t$ -statistics,  $t = \hat{\beta}_i / s(\hat{\beta}_i)$ , could be of interest. Build a version of your program which also computes  $s(\hat{\beta}_i)$  and the  $t$ -value, and outputs this together with  $\hat{\beta}$ .

# Fundamentals of R



# Creating variables

```
a <- 10 # assign 10 to 'a'  
a = 10  # same as above  
10 -> a # assign 10 to 'a'  
10 = a  # Wrong!
```



# Classes or Data types

- Variable types:
  - **character** - String
  - **integer** - Integers
  - **numeric** - Integers + Fractions
  - **logical** - Boolean
- Data types:
  - **vector** - A collection of elements of the same class
  - **matrix** - Simple matrix
  - **data.frame** - Complex table
  - **list** - A collection of elements of different classes

# Classes or Data types

```
a <- 10  
class(a)    # numeric  
a <- as.character(a)  
print(a)    # ???  
class(a)    # character
```

# What is a R package?

- Packages are collections of
  - R functions
  - R data
  - compiled code
- Can be created by anyone
- Academics often create their own packages

# Adding packages

- Using the RStudio GUI
- From the R console:

```
install.packages("car") # install 'car' package  
library(car)           # initialize the 'car' package  
require(car)           # another way to initialize
```

# Getting help

- Build-in support
- Google
- Source code

```
help(merge) # get help page for 'merge'
?merge      # lookup 'merge' from installed pkgs
??merge     # vague search
example(merge) # show code examples
help
```

# Importing and exporting data

- RStudio GUI
- Using the R console:

```
library(readxl) # initialize the required library
mydata <- read_excel("~/Downloads/mydata.xlsx") # import an Excel file
library(xlsx) # initialize the required library
write.xlsx(mydata, "~/Downloads/mydata.xlsx") # export a variable as an
Excel file
```

# How to create a vector?

- Combine function `c()`
- Only data of one type

```
vec1 <- c(1, 2, 3) # numeric vector  
vec2 <- c("a", "b", "c") # character vector  
vec3 <- c(1, "a") # error!
```

# How to reference elements of a vector?

- Elements of a vector can be accessed with its index
- First element has index 1
- Last element has index *length(vectorName)*

```
length(vec1)    # 4
print(vec1)     # 1 2 3
print(vec2[3])  # "c"
print(vec2[2:3]) # "b" "c"
```



# How to manipulate vectors

- Subsetting

```
logic1 <- vec1 < 15 # create a logical vector, TRUE if value < 15
vec1[logic1] # elements in TRUE positions will be included in subset
vec1[1:2] # returns elements in 1 & 2 positions.
vec1[c(1,3)] # returns elements in 1 & 3 positions
vec1[-1] # returns all elements except in position 1.
```

- Sorting

```
sort(vec1) # ascending sort
sort(vec1, decreasing = TRUE) # Descending sort
```

# How to create a data frame?

- Data frames are used for storing data tables

```
myDf1 <- data.frame(vec1, vec2) # make data frame with 2 columns
myDf2 <- data.frame(vec1, vec3, vec4)
myDf3 <- data.frame(vec1, vec2, vec3)
```

- Working with build-in datasets

```
library(datasets) # initialize
library(help=datasets) # display the datasets
```

# Basic operations for a data frame

```
class(airquality) # get class
sapply(airquality, class) # get class of all columns
str(airquality) # structure
summary(airquality) # summary of airquality
head(airquality) # view the first 6 obs
fix(airquality) # view spreadsheet like grid
rownames(airquality) # row names
colnames(airquality) # columns names
nrow(airquality) # number of rows
ncol(airquality) # number of columns
```

## Basic operations for a data frame

```
cbind(myDf1, myDf2) # columns append DFs with same no. rows
rbind(myDf1, myDf1) # row append DFs with same no. columns
myDf1$vec1 # vec1 column
myDf1[, 1] # df[row.num, col.num]
myDf1[, c(1,2)] # columns 1 and 3
myDf1[c(1:5), c(2)] # first 5 rows in column 2
```

## If-Else condition

```
if(checkConditionIfTrue) {  
    ....statements..  
    ....statements..  
} else {    # place the 'else' in same line as '}'  
    ....statements..  
    ....statements..  
}
```

# Loops in R

- Standard for-loop:

```
for(counterVar in c(1:n)){  
  .... statements..  
}
```

- The apply family

- **apply()**: Apply FUN through a data frame or matrix by rows or columns.

```
myData <- matrix(seq(1,16), 4, 4) # make a matrix  
apply(myData, 1, FUN=min) # apply 'min' by rows  
# => [1] 1 2 3 4
```

```
apply(myData, 2, FUN=min) # apply 'min' by columns  
# => [1] 4 8 12 16
```

# Functions in R

- Reusable block of code

```
myfunction <- function(arg1, arg2, ... ){  
  .... statements  
  return(object)  
}
```

# Useful Resources

- [DataCamp R](#)
- [Learn R with Swirl](#)
- [R-statistics Tutorial](#)
- [Statmethods Tutorial](#)
- [Linear Regression](#)



## Survey



<https://nl.surveymonkey.com/r/CSWEUR2019>

*Erasmus*