

# Advanced object types

Theoretical and Empirical Research Methodology, Exercise 3 - Part 2

**Prof. Dr. Claudius Gräbner-Radkowitzsch**

**Europa-University Flensburg, Department of Pluralist Economics**

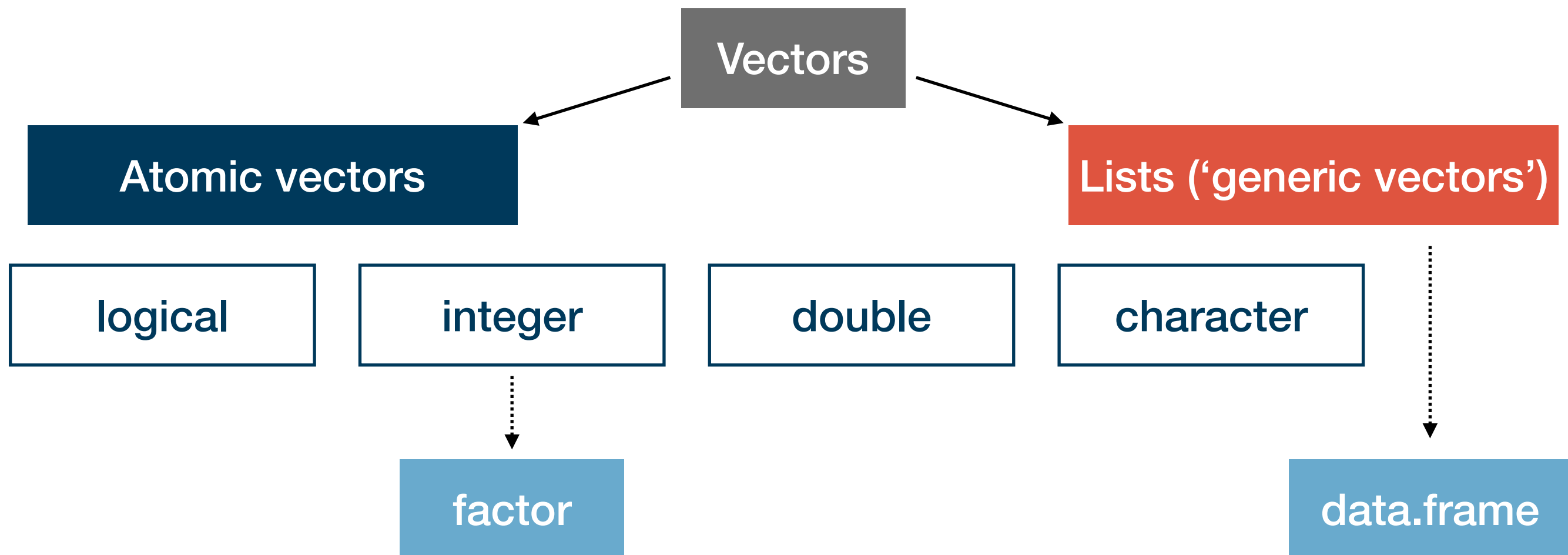
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# Goals for today

- I. Learn about three advanced object types: `factor`, `data.frame`, and `tibble`
- II. Understand and their relation to the basic types discussed previously

# Advanced object types in R

# Summary and outlook



# On more advanced object types

- While there are many object types in R, understanding the basics is key
  - These are by far the most common ones
  - All other object types are somehow ‘built upon’ the basic types by adding **attributes**
- Among the special types, two stand out in their prevalence:

## Categorical data: factor

- Can also take a pre-specified number of values: levels
- Classical example: Male, Female, Diverse
- Created using the function `factor()`

## Data frames: `data.frame` & `tibble`

- A kind of ‘table’ in which different variables are stored as vectors
  - A table-like form of `data.frame`
  - Tibbles as a new version of `data.frame` that “do less and do it better”
  - Created using `data.frame()` or `tibble::tibble()`
- |   | gender | height |
|---|--------|--------|
| 1 | male   | 189    |
| 2 | male   | 175    |
| 3 | male   | 180    |
| 4 | female | 166    |
| 5 | female | 150    |

- Others that we will not cover here are, e.g., matrices, durations, or dates

# Digression: some remarks on attributes

- To turn our basic object types into something more fancy we can give them **attributes**, one of which is called **class**
  - This changes their behaviour when functions are applied to them
  - Technically, adding a class attribute changes the class but not the type:

```
ff <- factor(c("F", "M", "M"), levels = c("F", "M", "D"))  
typeof(ff)  
class(ff)
```
- The class **factor** is an integer with two attributes:

```
attributes(ff)
```
- Not too important for us right now, but good to keep in mind!

# Factors

- Factors are used to represent ordinal or categorical data
  - Elements of factors can take one out of several pre-specified values: levels
  - Factors are integers with the attributes `levels` and `class`
- We create factors using the function `factor()`, which takes a vector and an optional argument `levels`:

```
f_1 <- factor(c(rep("F", 4), rep("D", 5), rep("M", 3)),  
              levels = c("D", "F", "M"))
```

## Your turn

- What happens if we do not specify `levels` explicitly?
- What happens if the vector contains elements not pre-specified as levels?

# Factors

- Usually levels are not ordered, but for ordinal data you can use the argument `ordered`:

```
f_2 <- factor(c("high", "high", "low"),  
              levels = c("low", "mid", "high"),  
              ordered=TRUE)
```
- There are some useful factor-specific functions such as `table()`.

## What does the function `table()` do?

Try it on the factors you defined so far!

How can you make sure the frequency of elements that do not show up in the vector is displayed as zero?

## General remark on using factors in practice

In my experience, it's usually better to store categorical data as character, and only transform them to factors if necessary



# Data frames

- Data frames are special lists of vectors where the length of each vector is equal!  
→ Most list operations also work for data.frames

- We usually represent data frames as tables:

	gender	height	Names of the vectors
1	male	189	
2	male	175	
3	male	180	
4	female	166	
5	female	150	

**vector 1** &  
**vector 2**

- To create a data frame from scratch use `data.frame()`:  

```
df_1 <- data.frame(  
  "gender" = c(rep("male", 3), rep("female", 2)),  
  "height" = c(189, 175, 180, 166, 150)  
)
```
- To create a data frame from a list use `as.data.frame()`
- If you read in data into R, it almost always starts off as a data.frame
- How to transform them is the main subject of the sessions on **data wrangling**

# Data frames and tibbles

- A modern version of the `data.frame` is the **tibble** (from the package *tibble*)
  - We will mostly use **tibbles** in this course, but make sure you familiarise yourself with the differences to the `data.frame`, which continues to be widespread (see the tutorial reading)

```
df_1 <- data.frame(  
  "gender" = c(rep("male", 3), rep("female", 2)),  
  "height" = c(189, 175, 180, 166, 150)  
)
```

- To transform a `data.frame` (or a `list`) into a **tibble**, use `tibble::as_tibble()`:

```
tb_1 <- tibble::as_tibble(df_1)
```

- To extract single columns use the `[` or `[[` operators
  - What's the difference between the two?
  - How do you think you can test for the type of a column vector?

# Data frames and tibbles

- To get a quick overview about the content, use `dplyr::glimpse()` or `head()`
- A complete overview can be obtained via `View()`
- Data frames are among the most widely used data types
  - There different approaches of how to handle and transform them, each associated with an R **dialect**
  - We mainly rely on the **tidyverse** dialect, which is the easiest to learn and comprehend → built upon **tibbles**
  - Alternatives are the **base** (classical) and **data.table** (fastest) dialect, which mainly use **data.frames** and **data.tables**
- This is useful to keep in mind when searching help in the internet

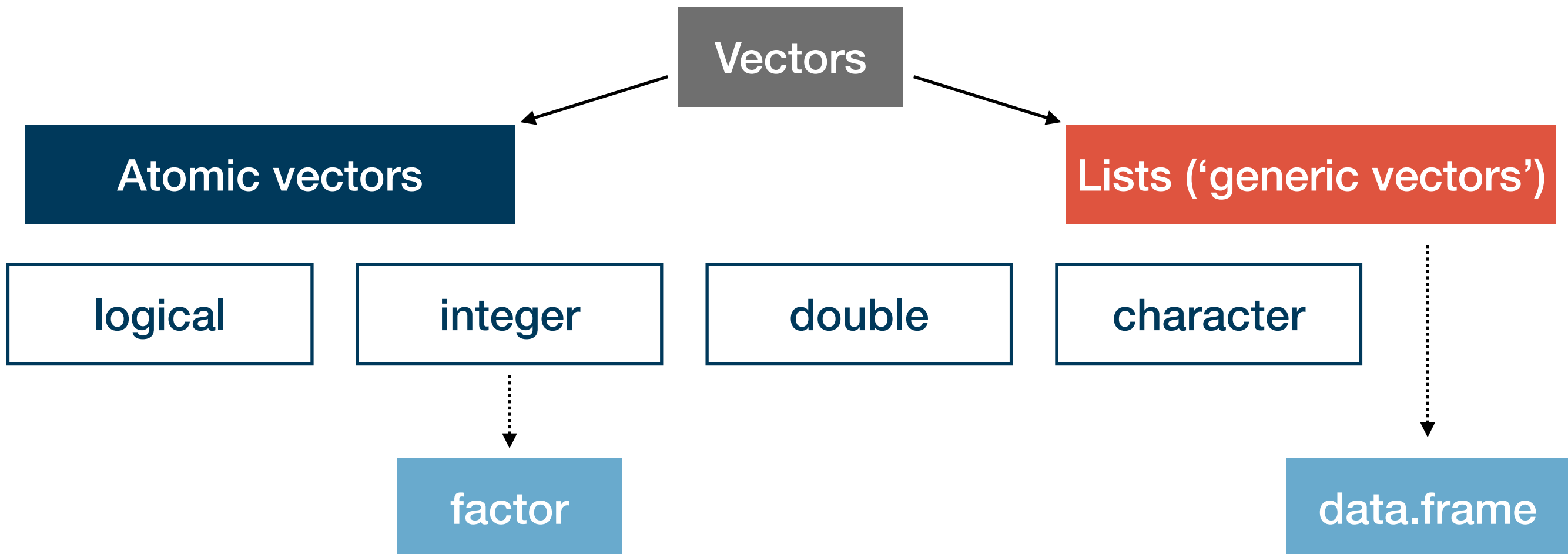
# Final exercises

- Create a factor with the levels "still", "medium", "sparkling", and arbitrary instances of the three levels
- Get the relative frequencies for "medium" of this factor
- Create a data frame with two columns, one called "nb" containing the numbers 1 to 5 as double, the other called "char" containing the numbers 6 to 10 as character
- Transform this data frame into a **tibble**!
- Extract the second column of this **tibble** such that you have a vector



# Summary and outlook

- This was the last session on the fundamentals of R
- We learned about the most important object types in R
- Functions do different things when applied to different objects → understanding object types is absolutely fundamental



# Summary and outlook

- Next session will be dedicated to recap and practicing
- I will explain unclear concepts or answer open questions → use the Moodle forum
- We will do some exercises together in class

## Tasks until next time:

1. Fill in the **quick feedback survey** on Moodle
2. Read the **tutorials** posted on the course page
3. Do the **exercises** provided on the course page and **discuss problems** and difficulties via the Moodle forum