

# Introductionary + advanced course

Berry Boessenkool, [berry-b@gmx.de](mailto:berry-b@gmx.de)

Hints and corrections are very welcome!

Download the current slides, source code and datasets at

[github.com/brry/course](https://github.com/brry/course)

These slides are licenced under ,  
so you can use the material freely as long as you cite me.

PDF created on 2018-10-25, 08:01

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

Course plan

# Outline

## R course info

### 1. One-Session-Intro

Intro, objects, vectors

Files, data.frames

Plotting

Packages, regression

### 2. Getting started: Background, GUI and first steps

### 3. Objects

### 4. "Real" data

### 5. Plotting

### 6. Character Operations

### 7. Conditions & loops

### 8. Functions & packages

### 9. Distributions

### 10. Statistics

### 11. Time series handling and analysis

### 12. Spatial data and GIS functionality

### 13. Final tasks

### Course plan

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fan

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: [extremeStat](http://extremeStat),

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: [extremeStat](#), [berryFunctions](#),

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: [extremeStat](#), [berryFunctions](#), [rdwd](#),

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: `extremeStat`, `berryFunctions`, `rdwd`, `OSMscale`

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: `extremeStat`, `berryFunctions`, `rdwd`, `OSMscale`
  - Community: 

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: `extremeStat`, `berryFunctions`, `rdwd`, `OSMscale`
  - Community: 
- These slides were originally based on a one week course held in 2013 for [CaWa](#) with Mathias Seibert (GFZ) in Bishkek, Kyrgyzstan

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: [extremeStat](#), [berryFunctions](#), [rdwd](#), [OSMscale](#)
  - Community: 
- These slides were originally based on a one week course held in 2013 for [CaWa](#) with Mathias Seibert (GFZ) in Bishkek, Kyrgyzstan
- R installation instructions: [github.com/brry/course#install](https://github.com/brry/course#install)

```
print("Hello world!")
```

- Berry Boessenkool → berry-b@gmx.de
- Geoecology @ Potsdam University
- R Fanatic since 2010
  - Teaching: [RclickHandbuch.wordpress.com](http://RclickHandbuch.wordpress.com) (German) + this course
  - Programming: `extremeStat`, `berryFunctions`, `rdwd`, `OSMscale`
  - Community: 
- These slides were originally based on a one week course held in 2013 for [CaWa](#) with Mathias Seibert (GFZ) in Bishkek, Kyrgyzstan
- R installation instructions: [github.com/brry/course#install](https://github.com/brry/course#install)
- If we're proceeding too fast, please interrupt!

# Why programming?

# Why programming?

- Data analysis

# Why programming?

- Data analysis
- Plotting

# Why programming?

- Data analysis
- Plotting        *Excel could do that, but is lacking*

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility
  - complex statistical procedures

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility
  - complex statistical procedures
  - job chance advantage

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility
  - complex statistical procedures
  - job chance advantage

Why R? (and not matlab, python, java, html, C++, fortran)

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility
  - complex statistical procedures
  - job chance advantage

Why R? (and not matlab, python, java, html, C++, fortran)

- Open Source → free!

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility
  - complex statistical procedures
  - job chance advantage

Why R? (and not matlab, python, java, html, C++, fortran)

- Open Source → free!
- Large community → stunning lots of methods!

# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility
  - complex statistical procedures
  - job chance advantage

Why R? (and not matlab, python, java, html, C++, fortran)

- Open Source → free!
- Large community → stunning lots of methods!
- Interpreter language → no compilation, human readable code

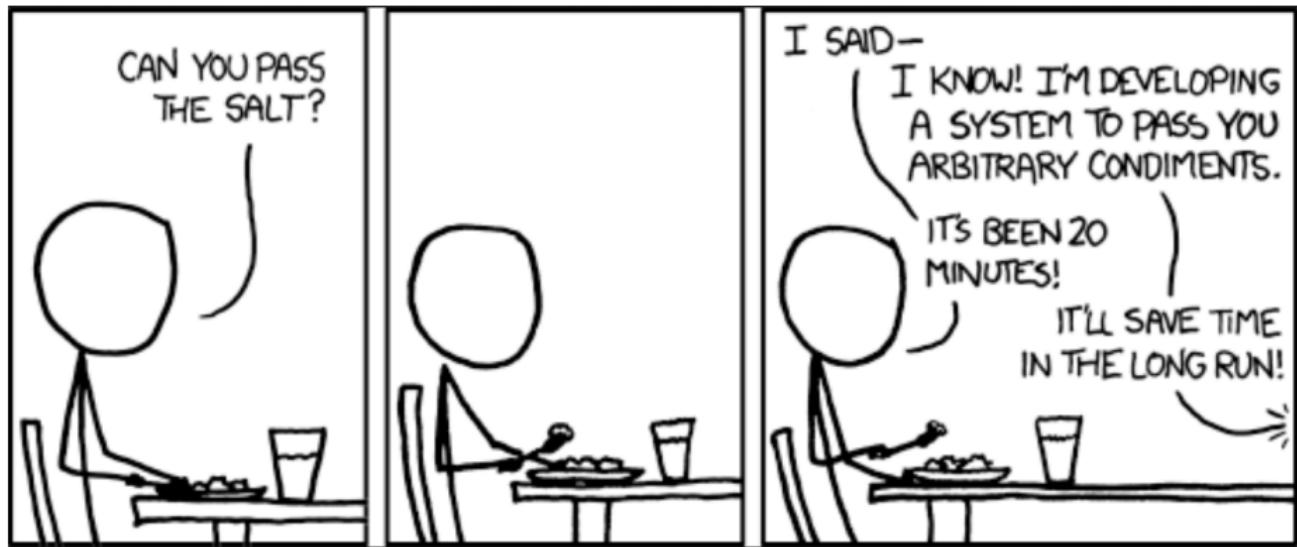
# Why programming?

- Data analysis
- Plotting            *Excel could do that, but is lacking*
- Awesomeness
  - efficiency
  - flexibility
  - automatization
  - transparency, reproducibility
  - complex statistical procedures
  - job chance advantage

## Why R? (and not matlab, python, java, html, C++, fortran)

- Open Source → free!
- Large community → stunning lots of methods!
- Interpreter language → no compilation, human readable code
- The standard for data analysis in many universities and industries

# In the future, programming will save you time!



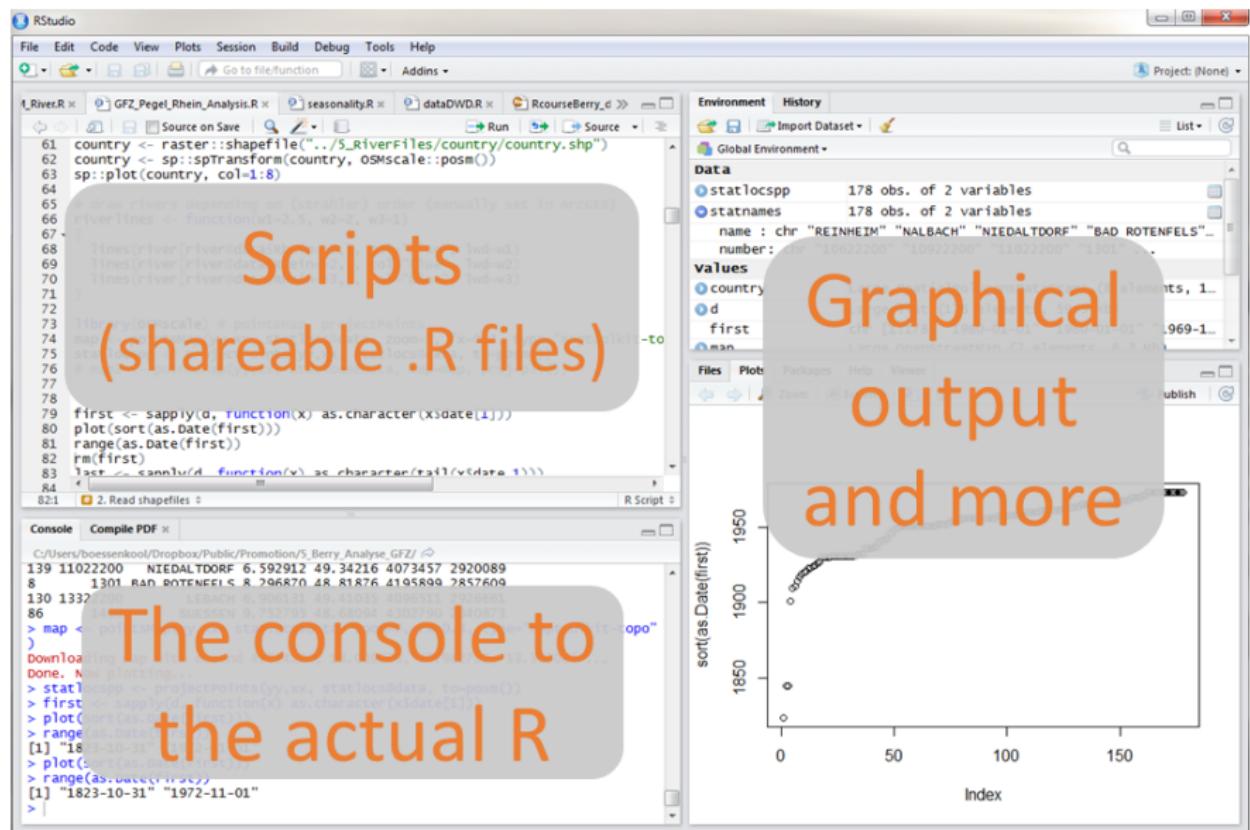
I find that when someone's taking time to do something right in the present, they're a perfectionist with no ability to prioritize, whereas when someone took time to do something right in the past, they're a master artisan of great foresight. ([xkcd.com/974](http://xkcd.com/974))

# Course plan

[Jump to plan Session 1](#)

One slide about Rstudio, and then we'll finally start...

## Integrated Development Environment (IDE): RStudio



# Get started in R

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$  ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$  ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

```
21+21 ; 7*6 ; 0.3/4*sqrt(313600)
```

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$ ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

```
21+21 ; 7*6 ; 0.3/4*sqrt(313600)
```

- objects: assignment with `<-` Rstudio Keyboard shortcut: `ALT + -`  
`nstudents <- 15`  
`nstudents`  
`nstudents > 12`

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$ ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

```
21+21 ; 7*6 ; 0.3/4*sqrt(313600)
```

- objects: assignment with `<-` Rstudio Keyboard shortcut: `ALT + -`  
`nstudents <- 15`  
`nstudents`  
`nstudents > 12`
- What's a good object name?

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$ ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

```
21+21 ; 7*6 ; 0.3/4*sqrt(313600)
```

- objects: assignment with `<-`   Rstudio Keyboard shortcut: `ALT + -`  
`nstudents <- 15`  
`nstudents`  
`nstudents > 12`
- What's a good object name? → short, but explanatory,

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$ ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

```
21+21 ; 7*6 ; 0.3/4*sqrt(313600)
```

- objects: assignment with `<-` Rstudio Keyboard shortcut: `ALT + -`  
`nstudents <- 15`  
`nstudents`  
`nstudents > 12`
- What's a good object name? → short, but explanatory,  
`lowerCamelStandard_or_underscore` are good naming conventions

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$ ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

```
21+21 ; 7*6 ; 0.3/4*sqrt(313600)
```

- objects: assignment with `<-` Rstudio Keyboard shortcut: `ALT + -`  
`nstudents <- 15`  
`nstudents`  
`nstudents > 12`
- What's a good object name? → short, but explanatory,  
`lowerCamelStandard_or_underscore` are good naming conventions
- comments: `# everything after a hashtag is not executed.`

# Get started in R

## Exercise 1: R is an awesome calculator

In the console, calculate  $21+21$ ,  $7*6$  and  $\frac{0,3}{4} * \sqrt{313600}$

*If you don't know how to compute a square root in R, you can google it!*

```
21+21 ; 7*6 ; 0.3/4*sqrt(313600)
```

- objects: assignment with `<-` Rstudio Keyboard shortcut: `ALT + -`  
`nstudents <- 15`  
`nstudents`  
`nstudents > 12`
- What's a good object name? → short, but explanatory,  
`lowerCamelStandard_or_underscore` are good naming conventions
- comments: `# everything after a hashtag is not executed.`
- scripts: Rstudio

## Objects: vectors

- To create a vector with values, use the syntax

```
values <- c(3, -11, 13, 5.94) # concatenate, combine
```

## Objects: vectors

- To create a vector with values, use the syntax

```
values <- c(3, -11, 13, 5.94) # concatenate, combine
```

- To obtain a subset (=indexing), use square brackets: `values[2]`

# Objects: vectors

- To create a vector with values, use the syntax  
`values <- c(3, -11, 13, 5.94) # concatenate, combine`
- To obtain a subset (=indexing), use square brackets: `values[2]`

## Exercise 2: Vector indexing

- ① Create a vector with body sizes of people around you. You can also use the values 1.75, 1.76, 1.83, 1.84, 1.77, 1.76, 1.77, 1.66, 1.86, 1.76. Assign it to an object with a useful name (`YourObject` is not one!).
- ② What does `3:6` create? What does `YourObject[3:6]` do?
- ③ What does `YourObject[-4]` do?
- ④ BONUS (for fast people): Analyze the descriptive statistics:  
`mean(YourObject)`, `median`, `min`, `max`, `range`, `quantile`
- ⑤ BONUS 2: Help your peers - there's no reason to be bored ;-)

# Vector task solutions

```
size <- c(1.75, 1.76, 1.83, 1.84, 1.77, 1.76,
        1.77, 1.66, 1.86, 1.76)
3:6 # A vector with consecutive integers
size[3:6] # Select the corresponding elements of a vector
size[-4] # Select all but the fourth value
```

```
c(mean(size), median(size), min(size), max(size)); range(size); quantile(size)

## [1] 1.776 1.765 1.660 1.860
## [1] 1.66 1.86
##    0%   25%   50%   75%   100%
## 1.660 1.760 1.765 1.815 1.860
```

`mean` and `median` are similar, hinting at a symmetric data distribution  
`range` returns a composite of `minimum` and `maximum`  
`quantile` returns `quantiles` (now that is surprising...)

# Outline

## R course info

### 1. One-Session-Intro

Intro, objects, vectors

Files, `data.frames`

Plotting

Packages, regression

### 2. Getting started: Background, GUI and first steps

### 3. Objects

### 4. "Real" data

### 5. Plotting

### 6. Character Operations

### 7. Conditions & loops

### 8. Functions & packages

### 9. Distributions

### 10. Statistics

### 11. Time series handling and analysis

### 12. Spatial data and GIS functionality

### 13. Final tasks

### Course plan

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

Read the file into R with the command `read.table`, assigning it to an object with a good name.

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

Read the file into R with the command `read.table`, assigning it to an object with a good name.

Use the documentation to find out the correct settings of the arguments:

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

Read the file into R with the command `read.table`, assigning it to an object with a good name.

Use the documentation to find out the correct settings of the arguments:

```
help(read.table), ?read.table, or press F1.
```

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

Read the file into R with the command `read.table`, assigning it to an object with a good name.

Use the documentation to find out the correct settings of the arguments:

```
help(read.table), ?read.table, or press F1.
```

If R tells you "no such file" exists, check the output of `dir()`.

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

Read the file into R with the command `read.table`, assigning it to an object with a good name.

Use the documentation to find out the correct settings of the arguments:

```
help(read.table), ?read.table, or press F1.
```

If R tells you "no such file" exists, check the output of `dir()`.

Check the object with `head(YourObject)`.

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

Read the file into R with the command `read.table`, assigning it to an object with a good name.

Use the documentation to find out the correct settings of the arguments:

```
help(read.table), ?read.table, or press F1.
```

If R tells you "no such file" exists, check the output of `dir()`.

Check the object with `head(YourObject)`.

`str(YourObject)` must yield the column data types: `num`, `num`, `factor`.

# Reading files

## Exercise 3: Reading files

Copy the file `treesize.txt` (*rightclick Raw, save as*) and tell R where to look for it with:

```
setwd("C:/path/to/input") # change back- to forwardslashes
```

Read the file into R with the command `read.table`, assigning it to an object with a good name.

Use the documentation to find out the correct settings of the arguments:

```
help(read.table), ?read.table, or press F1.
```

If R tells you "no such file" exists, check the output of `dir()`.

Check the object with `head(YourObject)`.

`str(YourObject)` must yield the column data types: `num`, `num`, `factor`.

You need to set the argument `header`.

## Solution to exercise 3: Reading files

```
treesize <- read.table(file="treesize.txt", header=TRUE)
```

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:

```
data.frame(count=c(2,6,5), type=c("a","k","k"))
```

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a data.frame

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a data.frame
- If we have the object `df`, we can subset with `df[rows,columns]`

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a data.frame
- If we have the object `df`, we can subset with `df[rows,columns]`
- `df[1,2:4];`

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a data.frame
- If we have the object `df`, we can subset with `df[rows,columns]`
- `df[1,2:4]; df[2, ];`

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a data.frame
- If we have the object `df`, we can subset with `df[rows,columns]`
- `df[1,2:4]; df[2, ]; df[, "name"];`

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a data.frame
- If we have the object `df`, we can subset with `df[rows,columns]`
- `df[1,2:4]; df[2, ]; df[, "name"]; df$name`

## Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type data.frame:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a data.frame
- If we have the object `df`, we can subset with `df[rows,columns]`
- `df[1,2:4]; df[2, ]; df[ , "name"]; df$name`
- Logical values: `vect[c(TRUE,TRUE,FALSE,FALSE,TRUE,FALSE)]`

# Objects: data.frames

- For tables with different data types (numbers, characters, categories, integers), R has the object type `data.frame`:  
`data.frame(count=c(2,6,5), type=c("a","k","k"))`
- `read.table` also returns a `data.frame`
- If we have the object `df`, we can subset with `df[rows,columns]`
- `df[1,2:4]; df[2, ]`; `df[ , "name"]`; `df$name`
- Logical values: `vect[c(TRUE,TRUE,FALSE,FALSE,TRUE,FALSE)]`

## Exercise 4: Data.frame indexing

From the dataset `treesize` from the previous exercise, obtain:

- The first 5 values in column 2
- The maximum "Height" (the maximum of the values in that column)
- For each entry: is the measurement equal to (`==`) A?
- BONUS 1: The height entries for trees older than 23.5 years
- BONUS 2: All rows, excluding rows 3,7,8,9,...,90

## Solution to exercise 4: subsetting data.frames I

```
treesize[1:5, 2]

## [1] 32.7 26.0 27.9 35.4 26.4

max(treesize$height) ; max(treesize[, "age"])

## [1] 40.1
## [1] 24.8

treesize$measurement=="A"

## [1] TRUE TRUE TRUE TRUE FALSE FALSE TRUE TRUE TRUE TRUE TRUE TRUE FALSE
## [13] TRUE TRUE TRUE FALSE TRUE FALSE FALSE FALSE TRUE FALSE TRUE TRUE
## [25] TRUE FALSE TRUE FALSE FALSE FALSE FALSE TRUE TRUE FALSE FALSE FALSE FALSE
## [37] FALSE TRUE FALSE TRUE FALSE FALSE TRUE TRUE TRUE FALSE FALSE FALSE FALSE
## [49] TRUE FALSE TRUE TRUE TRUE TRUE FALSE FALSE TRUE TRUE FALSE FALSE FALSE
## [61] TRUE TRUE FALSE FALSE TRUE FALSE TRUE FALSE TRUE FALSE FALSE FALSE FALSE
## [73] FALSE FALSE FALSE FALSE TRUE FALSE TRUE FALSE FALSE TRUE TRUE TRUE
## [85] FALSE FALSE FALSE FALSE TRUE FALSE TRUE TRUE FALSE TRUE TRUE FALSE
## [97] TRUE FALSE TRUE TRUE
```

## Solution to exercise 4: subsetting data.frames II

```
treesize[ treesize$age >= 23.5 , "height"]  
  
## [1] 32.7 38.4 40.1 34.0 32.3 32.6 34.4 23.2 34.6  
## [10] 35.9
```

```
treesize[ -c(3,7:90) , ]  
  
##      age height measurement  
## 1    24.4    32.7          A  
## 2    13.6    26.0          A  
## 4    21.2    35.4          A  
## 5    19.8    26.4          B  
## 6     7.7     7.2          B  
## 91   23.1    39.0          A  
## 92   24.6    34.6          A  
## 93   7.7     9.2          B  
## 94  10.1    10.3          A  
## 95  23.8    35.9          A  
## 96   7.7     7.4          B  
## 97  17.8    31.6          A  
## ...
```

# Outline

## R course info

### 1. One-Session-Intro

Intro, objects, vectors

Files, data.frames

#### Plotting

Packages, regression

### 2. Getting started: Background, GUI and first steps

### 3. Objects

### 4. "Real" data

### 5. Plotting

### 6. Character Operations

### 7. Conditions & loops

### 8. Functions & packages

### 9. Distributions

### 10. Statistics

### 11. Time series handling and analysis

### 12. Spatial data and GIS functionality

### 13. Final tasks

### Course plan

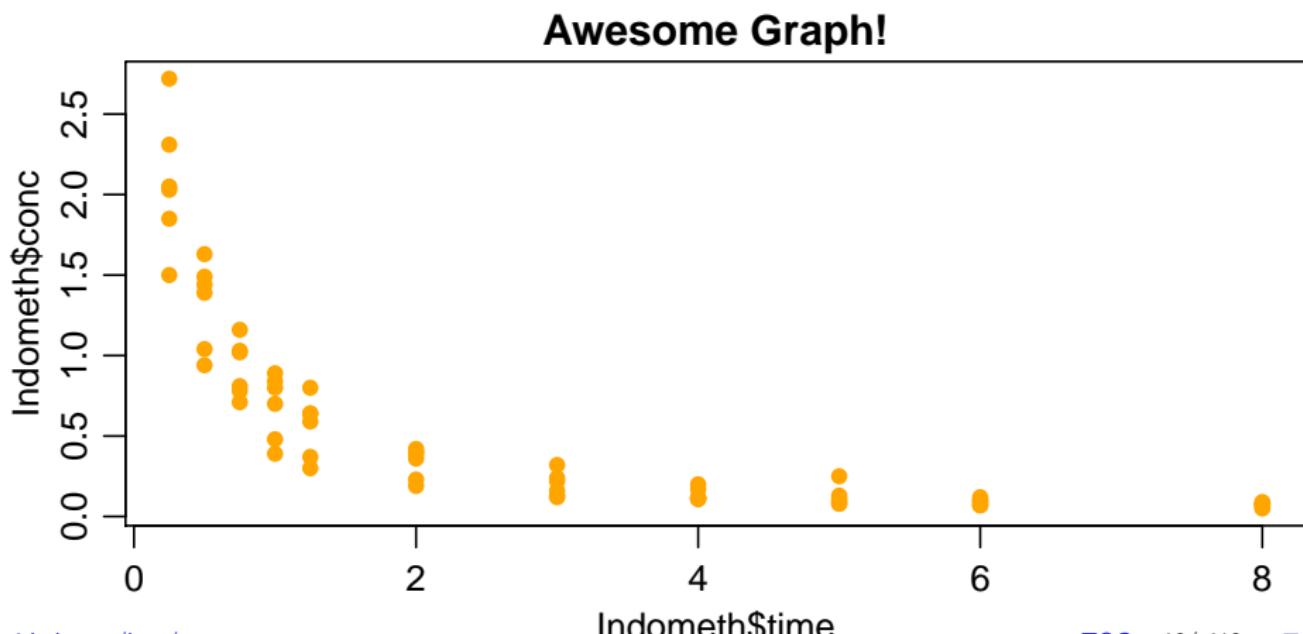
# Plotting I

General code for scatterplots: `plot(x, y, ...)`

# Plotting I

General code for scatterplots: `plot(x, y, ...)`

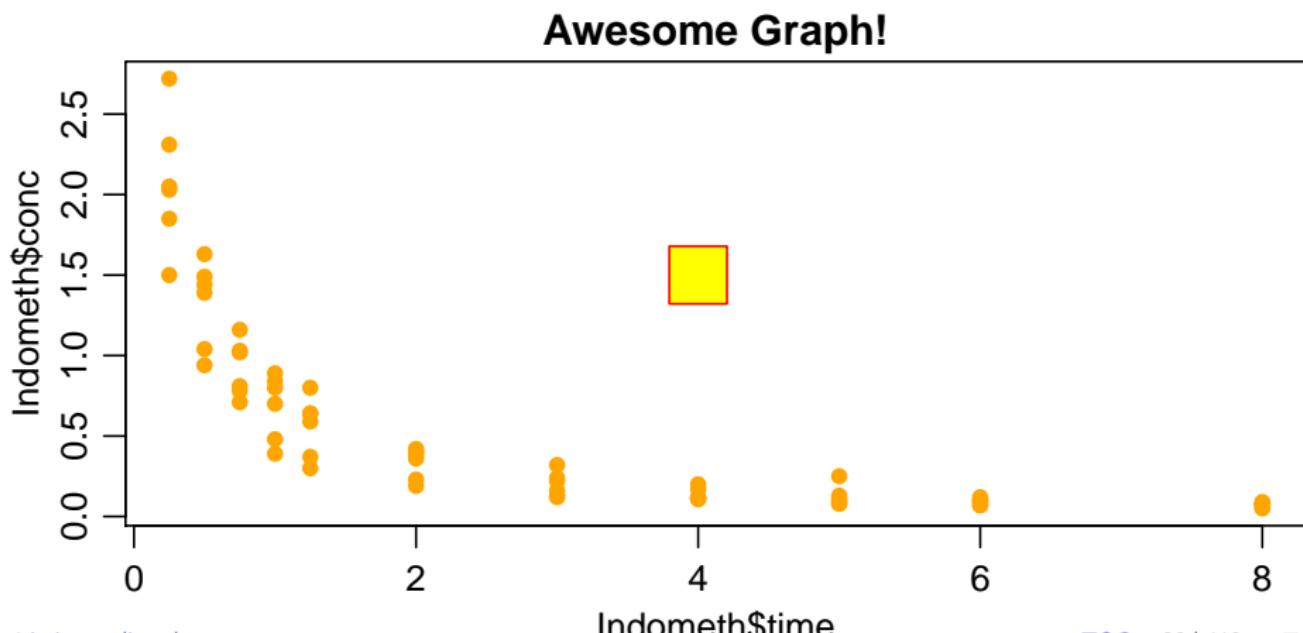
```
plot(x=Indometh$time, y=Indometh$conc,  
      col="orange", pch=16, main="Awesome Graph!")
```



# Plotting I

General code for scatterplots: `plot(x, y, ...)`

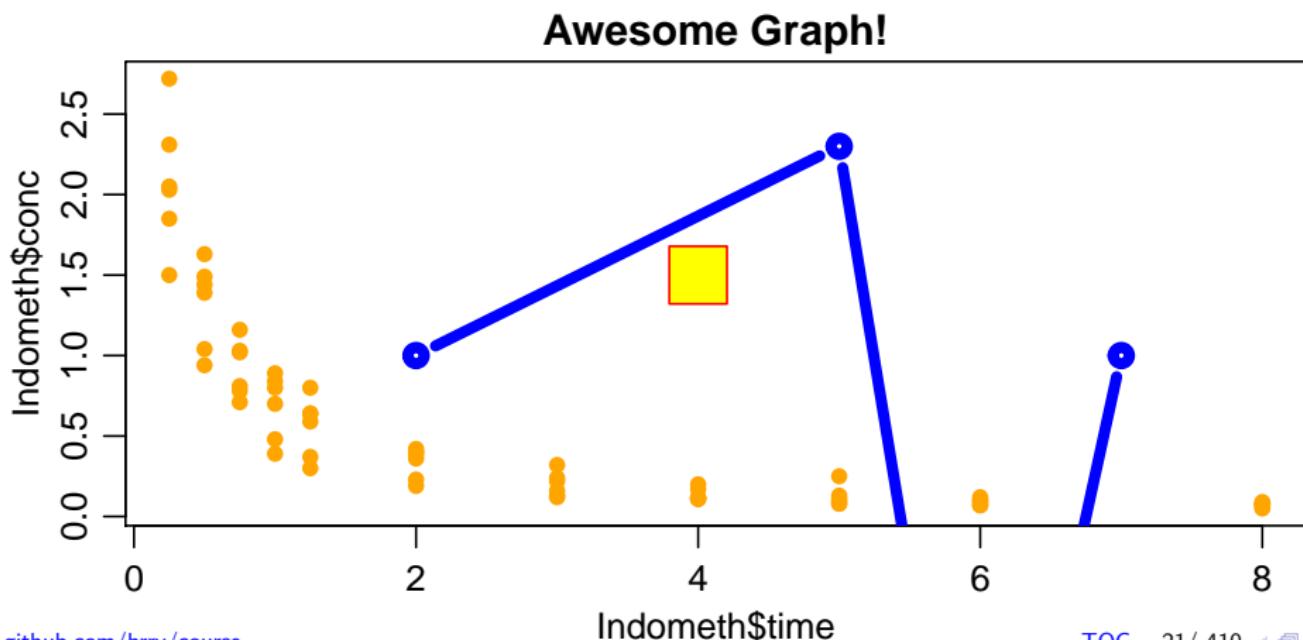
```
points(4, 1.5, pch=22, bg="yellow", cex=4, col="red")  
# PointCharacter, BackGround, Character Expansion
```



# Plotting I

General code for scatterplots: `plot(x, y, ...)`

```
lines(x=c(2,5,6,7), y=c(1,2.3,-3,1),  
      col=4, type="b", lwd=5)
```



## Plotting II: Our treesize dataset

```
treesize <- read.table(file="data/treesize.txt", header=TRUE)
```

## Plotting II: Our treesize dataset

```
treesize <- read.table(file="data/treesize.txt", header=TRUE)
```

```
plot(x=xvalues, y=y_values, xlab="nice axis label",
      main="graph title", las=1)
```

## Plotting II: Our treesize dataset

```
treesize <- read.table(file="data/treesize.txt", header=TRUE)

plot(x=xvalues, y=y_values, xlab="nice axis label",
      main="graph title", las=1)
```

### Exercise 5: Scatterplots, sample distribution graphs

- ① Plot tree height over age.
- ② Add labels to the plot.
- ③ Change the point character (`pch`) and color (`col`).
- ④ BONUS 1: Use a vector for colors, e.g. `treesize$measurement`
- ⑤ BONUS 2: Compare the histogram (`hist`) of the heights with the `boxplot` and `quantile(x, probs=c(0.1, 0.8))`.

## Solution for exercise 5: Scatterplots

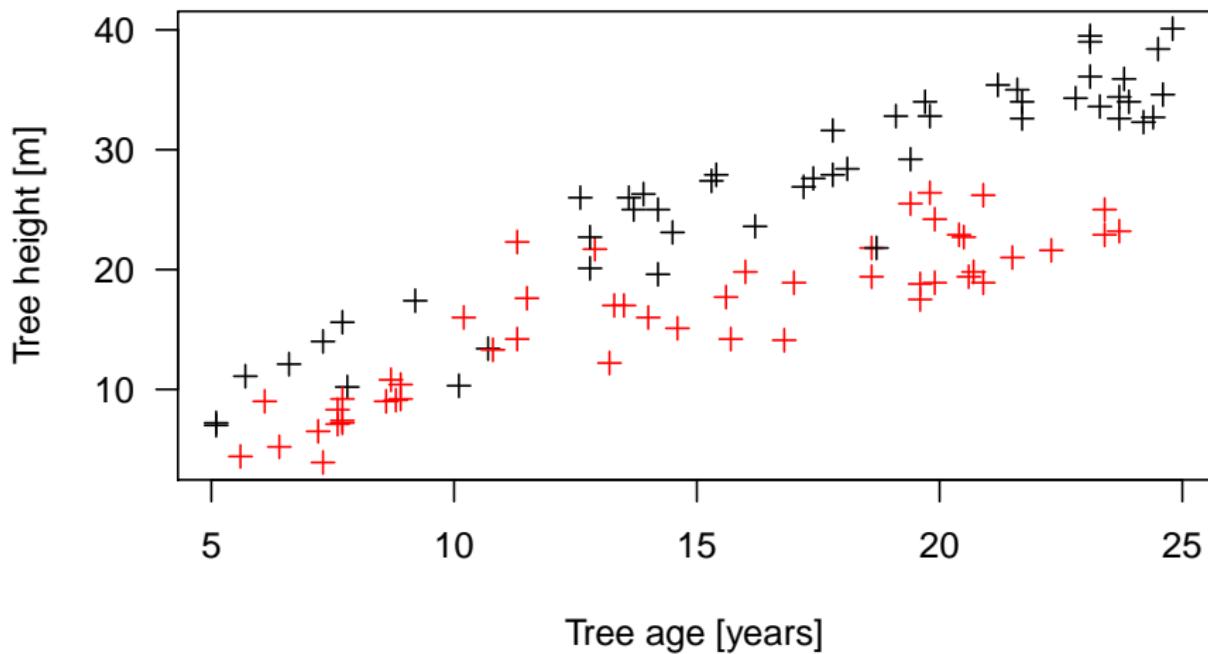
```
treesize <- read.table(file="data/treesize.txt", header=TRUE)

plot(treesize$age, treesize$height)
plot(treesize$age, treesize$height, las=1, ylab="Tree height [m]",
      xlab="Tree age [years]", col=treesize$measurement,
      main="Older trees are larger", pch=3)
quantile(treesize$height, probs=c(0.1, 0.8))

##    10%    80%
##  8.93 32.36
```

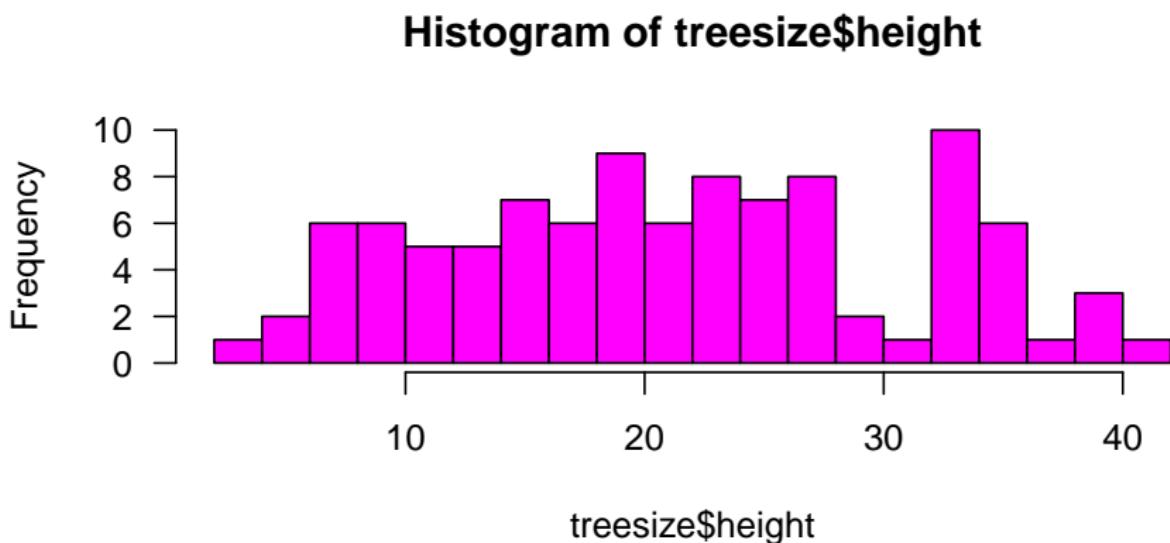
# Solution for exercise 5: Scatterplot

Older trees are larger



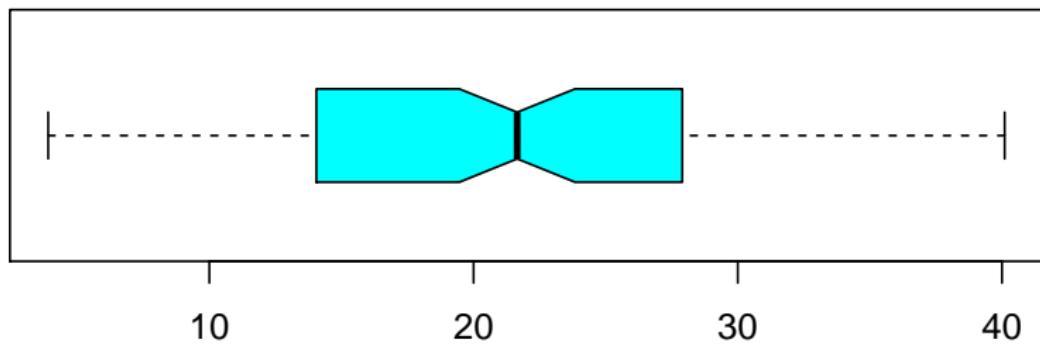
## Solution for exercise 5: Histogram

```
hist(treesize$height, col=6, breaks=20, las=1)
```



## Solution for exercise 5: Boxplot

```
boxplot(treesize$height, col=5, horizontal=TRUE, notch=TRUE)
```



## commonly needed plot arguments

```
plot(x, y, # point coordinates
col="lightblue", # point color
pch=0, # point character (symbol)
xlab="My label [km]", ylab="", # axis labels
main="Graph title", # title
cex=1.8, # character expansion (symbol size)
type="l", # draw lines instead of points
lwd=3, # line width (thickness of lines)
las=1, # label axis style (axis numbers upright)
xaxt="n" # axis type (none to suppress axis)
)
```

# Outline

R course info

## 1. One-Session-Intro

Intro, objects, vectors

Files, data.frames

Plotting

Packages, regression

## 2. Getting started: Background, GUI and first steps

## 3. Objects

## 4. "Real" data

## 5. Plotting

## 6. Character Operations

## 7. Conditions & loops

## 8. Functions & packages

## 9. Distributions

## 10. Statistics

## 11. Time series handling and analysis

## 12. Spatial data and GIS functionality

## 13. Final tasks

Course plan

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network
- Packages for a range of topics: [cran.r-project.org/web/views](http://cran.r-project.org/web/views)

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network
- Packages for a range of topics: [cran.r-project.org/web/views](http://cran.r-project.org/web/views)
- All >10'500 available packages: [cran.r-project.org/web/packages](http://cran.r-project.org/web/packages)

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network
- Packages for a range of topics: [cran.r-project.org/web/views](http://cran.r-project.org/web/views)
- All >10'500 available packages: [cran.r-project.org/web/packages](http://cran.r-project.org/web/packages)
- `install.packages("ggplot2")` to download and install.  
(only needs to be executed once, works on user level, no admin rights required)  
You can do this in Rstudio

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network
- Packages for a range of topics: [cran.r-project.org/web/views](http://cran.r-project.org/web/views)
- All >10'500 available packages: [cran.r-project.org/web/packages](http://cran.r-project.org/web/packages)
- `install.packages("ggplot2")` to download and install.  
(only needs to be executed once, works on user level, no admin rights required)  
You can do this in Rstudio
- `library("ggplot2")` to load it  
(needed in every new R session) Put this in the script for reproducibility

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network
- Packages for a range of topics: [cran.r-project.org/web/views](http://cran.r-project.org/web/views)
- All >10'500 available packages: [cran.r-project.org/web/packages](http://cran.r-project.org/web/packages)
- `install.packages("ggplot2")` to download and install.  
(only needs to be executed once, works on user level, no admin rights required)  
You can do this in Rstudio
- `library("ggplot2")` to load it  
(needed in every new R session) Put this in the script for reproducibility
- Better to use the `package::function` syntax

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network
- Packages for a range of topics: [cran.r-project.org/web/views](http://cran.r-project.org/web/views)
- All >10'500 available packages: [cran.r-project.org/web/packages](http://cran.r-project.org/web/packages)
- `install.packages("ggplot2")` to download and install.  
(only needs to be executed once, works on user level, no admin rights required)  
You can do this in Rstudio
- `library("ggplot2")` to load it  
(needed in every new R session) Put this in the script for reproducibility
- Better to use the `package::function` syntax
- Regularly run `update.packages()` or use the Rstudio button

# R Packages

- Many people write code for specific tasks and publish it on CRAN, the Comprehensive R Archive Network
- Packages for a range of topics: [cran.r-project.org/web/views](http://cran.r-project.org/web/views)
- All >10'500 available packages: [cran.r-project.org/web/packages](http://cran.r-project.org/web/packages)
- `install.packages("ggplot2")` to download and install.  
(only needs to be executed once, works on user level, no admin rights required)  
You can do this in Rstudio
- `library("ggplot2")` to load it  
(needed in every new R session) Put this in the script for reproducibility
- Better to use the `package::function` syntax
- Regularly run `update.packages()` or use the Rstudio button
- Rarely needed: `remove.packages("packagename")`

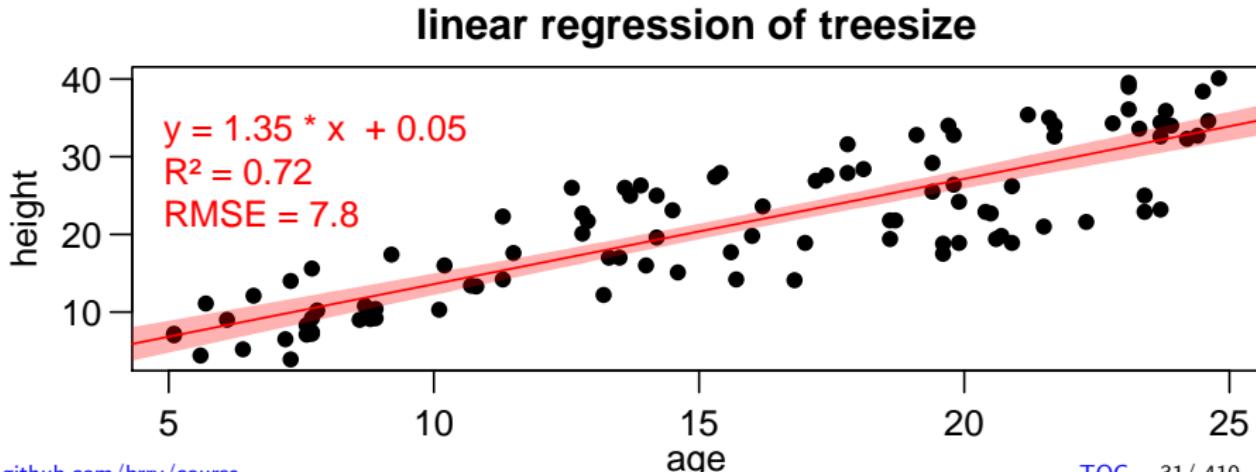
# Packages and linear regression

## Exercise 6: Packages, linear regression

- ① Install and load the package `berryFunctions`
- ② Pass the `treesize` data to `linReg`. Don't use (x,y) coordinates, but a (y~x) formula, read as "y dependent on x". Use the data argument as described in the "Examples" section of the `linReg` manual.
- ③ Describe the resulting graph (height vs age).
- ④ Feed the data into `lm`, assign the output to an object (useful name!).
- ⑤ Briefly explain the `summary` of the linear model.
- ⑥ BONUS1: Look into the source code of `linReg`. You can print the function by calling it without brackets. You can also open it on github with `berryFunctions::funSource`.
- ⑦ BONUS2: What is the backbone for the calculation in `linReg`?
- ⑧ BONUS3: install `rskey` ([via github](#)), set a keyboard shortcut for `funSource` and use it to get the source code of some function.

# Solution for exercise 6: Linear regression

```
library("berryFunctions")
linReg(height~age, data=treesize)
linReg
browseURL("https://github.com/brry/berryFunctions")
# go to R/linReg.R - working horse: lm
linear_model <- lm(height~age, data=treesize)
summary(linear_model)
blog.yhathq.com/posts/r-lm-summary.html
stats.stackexchange.com/questions/5135/interpretation-of-rs-lm-output
```



# Time to put all of this into practice

## Exercise 7: Reading files, subsetting, comparison

- ① Start a new session of R. Read in the tree size data.
- ② Check that everything is read correctly with `str`.
- ③ `which` rows of the `data.frame` have measurement equal to (`==`) B?
- ④ Plot the linear regression for measurement method B, then add the points and regression line for A in a different color and shape.
- ⑤ BONUS 1: use `boxplot` with a formula (`height~measurement`) for an automatic boxplot comparison with median difference notches.
- ⑥ BONUS 2: Visually compare the effect of Girth and Height on Volume in the dataset `trees`. Here's one idea (of many possible): Plot with two panels below each other (`par(mfrow=c(2,1))`), with the linear regression plots (`berryFunctions::linReg`) for each variable. Make it look nice with the graphical parameters `mar` and `mgp`, as well as by specifying some of the `linReg` arguments.

## Solution for exercise 7: Reading files, subsetting

```
setwd("D:/my/path")
treesize <- read.table(file="data/treesize.txt", header=TRUE)
str(treesize)

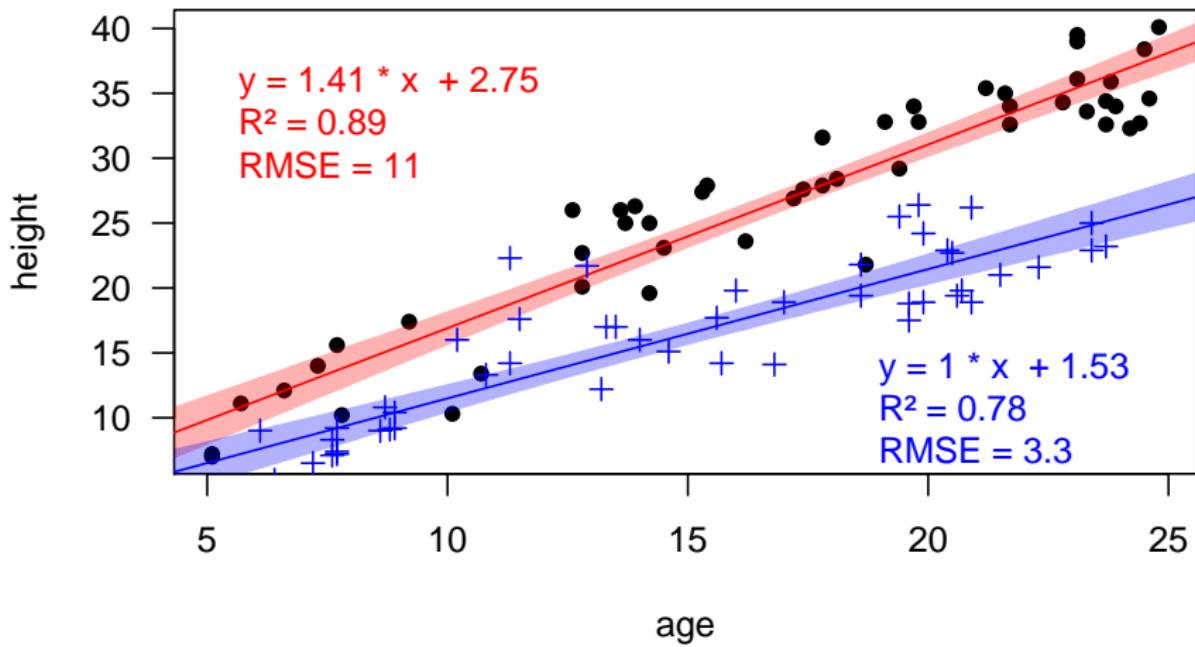
which(treesize$measurement=="B")

## [1] 5 6 12 16 18 19 20 22 26 28 29 30 31 34 35
## [16] 36 37 39 41 42 46 47 48 50 55 56 59 60 63 64
## [31] 66 68 70 71 72 73 74 75 76 78 80 81 85 86 87
## [46] 88 90 93 96 98

treeA <- treesize[treesize$measurement=="A", ]
treeB <- treesize[treesize$measurement=="B", ]
library("berryFunctions")
linReg(height~age, data=treeA, pos1="topleft")
linReg(height~age, data=treeB, add=T, col=4,
       pos1="bottomright", inset=0.05)
points(height~age, data=treeB, pch=3, col=4)
```

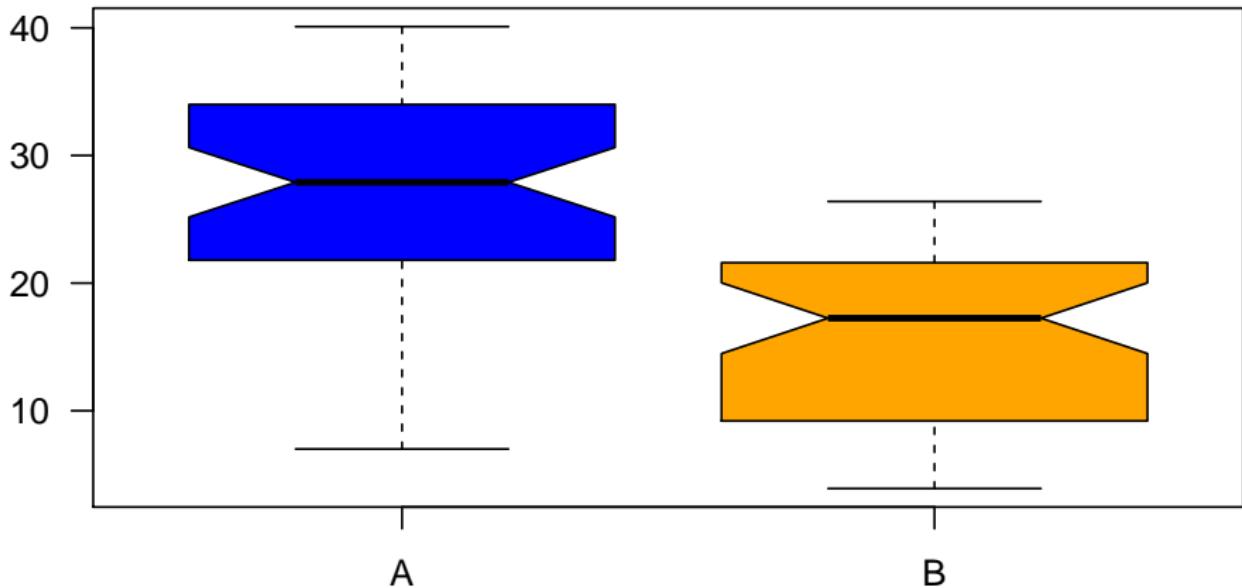
# Solution for exercise 7: Plotting, linear Regression

## linear regression of treeA



# Solution for exercise 7.BONUS 1

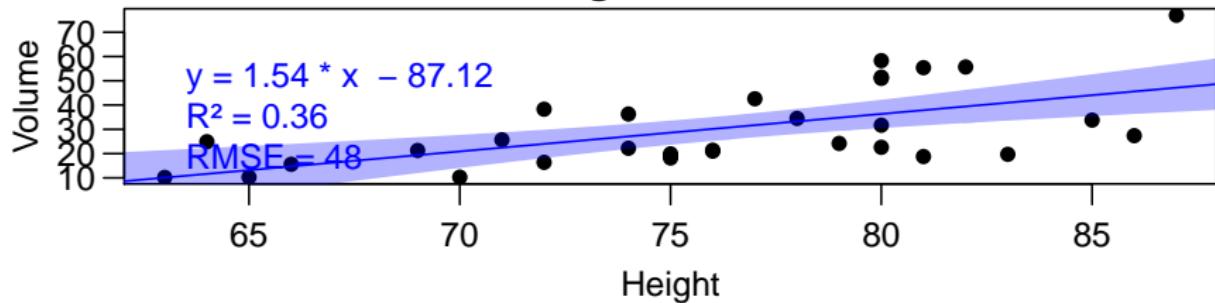
```
boxplot(height~measurement, data=treesize,  
        col=c("blue","orange"), notch=TRUE)
```



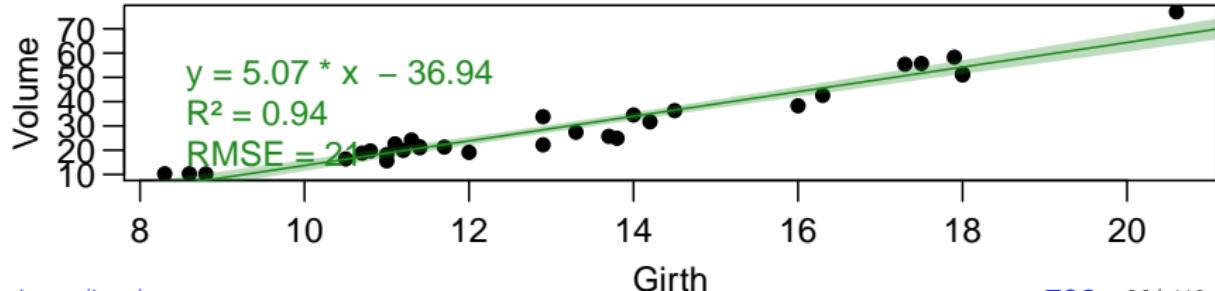
# Solution for exercise 7.BONUS 2

```
par(mfrow=c(2,1), mar=c(3,3,1.5,0.5), mgp=c(1.9,0.7,0), las=1)
linReg(Volume~Height, data=trees, col="blue", pos1="topleft")
linReg(Volume~Girth, data=trees, col="forestgreen", pos1="topleft")
```

linear regression of trees



linear regression of trees



# Outline

R course info

1. One-Session-Intro

2. Getting started: Background, GUI and first steps

    Background information

        Graphical user interfaces

        Getting help

        R syntax

        Source code of R commands

3. Objects

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

# R - What and Why?

# R - What and Why?

- R is a standard environment for statistical computing
- and an object-oriented programming language.
- It is developed and maintained by the R development core team
- and has thousands of add-on packages written by the community.
- There are mailing lists where users (& developers) help users
- and blogs to communicate analysis and coding to the public.
- R allows complex data analysis of big data sets with current methods
- and the creation of advanced professional (even interactive) graphics.
- Coding and programming forces you to keep track of your work
- and allows you to efficiently customize and automatize your analysis.
- R is very versatile: you can write your own functions and packages,
- implement new methods and make them available to everyone.
- Being capable of programming helps finding a job
- and the best part: **R is open source and free!**

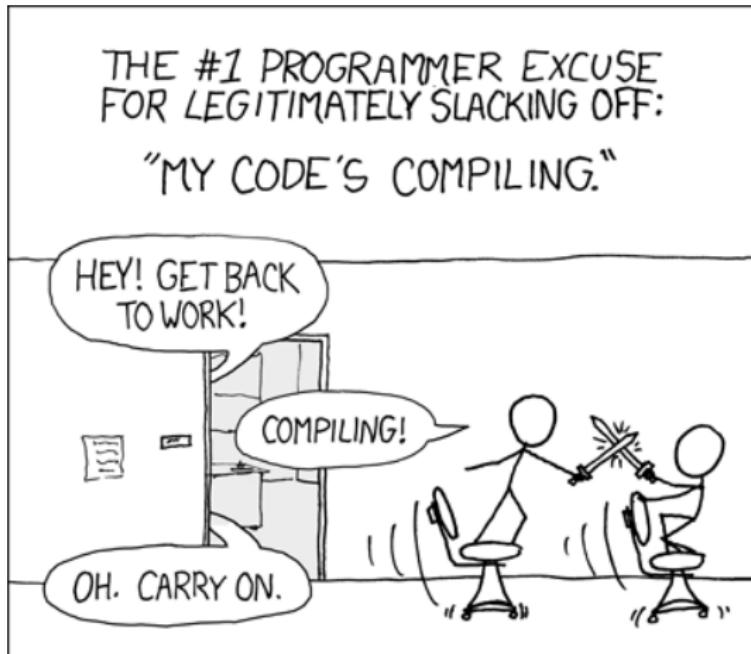
[fantasyfootballanalytics.net/2014/01/why-r-is-better-than-excel.html](http://fantasyfootballanalytics.net/2014/01/why-r-is-better-than-excel.html)

[thegrantlab.org/bio3d/tutorials/79-static-content/85-why-use-r](http://thegrantlab.org/bio3d/tutorials/79-static-content/85-why-use-r)

[github.com/bryf/course](https://github.com/bryf/course)

# interpreter language - no compilation

# interpreter language - no compilation



[xkcd.com/303](http://xkcd.com/303)

# Where to get R

Instructions online: [github.com/brry/course#install](https://github.com/brry/course#install)

- Editor alternative to Rstudio: [Tinn-R](#)
- Only for Windows OS, but flexible window arrangement
- Needs some initial tweaking, see [Berry's TinnR Page](#)
- Since Rstudio has become so incredibly awesome, you should use it!

# Outline

R course info

1. One-Session-Intro

2. Getting started: Background, GUI and first steps

    Background information

    Graphical user interfaces

    Getting help

    R syntax

    Source code of R commands

3. Objects

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

# RStudio

**SCRIPTS**

**RUN CODE**

**CODE EXECUTION**

**OBJECTS IN WORKSPACE**

**DOCUMENTATION**

**PLOTS**

The screenshot shows the RStudio interface with several red boxes highlighting specific areas:

- A red box highlights the "Run" button in the top toolbar.
- A red box highlights the "Help" menu in the top toolbar.
- A red box highlights the "Index" axis of a scatter plot in the bottom right corner.

The workspace contains the following objects:

- country**: A raster object with 178 observations and 1 variable.
- statlocsp**: A data frame with 178 observations and 1 variable.
- statnames**: A character vector with 178 elements.
- values**: A data frame with 178 observations and 1 variable.
- country**: A data frame with 178 observations and 1 variable.
- d**: A data frame with 178 observations and 1 variable.
- first**: A character vector with 178 elements.
- man**: A data frame with 178 observations and 1 variable.

The console output shows the execution of R code to read shapefiles and project points onto a map, followed by a scatter plot of date versus index.

```
1 River.R x GFZ_Pegel_Rhein_Analysis.R x seasonality.R x dataDWDR x RsourceBerry d >>
File Edit Code View Plots Session Build Debug Tools Help
Source On Save Run Source
61 country <- raster::shapefile("../5_RiverFiles/country", wld=TRUE)
62 country <- sp::spTransform(country, osmscale())
63 sp::plot(country, col=1:8)
64
65 # draw rivers depending on (strahler
66 riverlines <- function(wl=2.5, w2=2, w3=1)
67 {
68   lines(river[river$data$Rhein==1], col="blue", lwd=w1)
69   lines(river[river$data$Rhein==2], col="blue", lwd=w2)
70   lines(river[river$data$Rhein==3], col="blue", lwd=w3)
71 }
72
73 library(osmscale) # pointsMap, projectPoints
74 map <- pointsMap(yy,xx, statlocs@data, zoom=7, fx=0.3, type="maptoolkit-to-
75 statlocsp <- projectPoints(yy,xx, statlocs@data, to=psom())
76 # map2 <- pointsMap(yy,xx, statlocs@data, map=map, proj=pl1())
77
78
79 first <- supply(d, function(x) as.character(x$date[1]))
80 plot(sort(as.Date(first)))
81 range(as.Date(first))
82 rm(first)
83 last <- supply(d, function(v) as.character(v$date[1]))
84
85
86 2. Read shapefiles :
87
88
89
90
91
92
93
94
95
96
97
98
99
100
101
102
103
104
105
106
107
108
109
110
111
112
113
114
115
116
117
118
119
120
121
122
123
124
125
126
127
128
129
130
131
132
133
134
135
136
137
138
139 11022200 NIEDALTENDORF 6.592912 49.34216 4073457 2920089
8 1301 BAD ROTENFELS 8.296870 48.81876 4195899 2857609
130 13322200 LEBACH 6.906131 49.41035 4096511 2926661
86 1409 SUESSEN 9.752795 48.68094 4302790 2840873
> map <- pointsMap(yy,xx, statlocs@data, zoom=7, fx=0.3, type="maptoolkit-topo"
)
Downloading map with extend 4.594204, 13.009235, 45.982782, 53.363084 ...
Done. Now plotting...
> statlocsp <- projectPoints(yy,xx, statlocs@data, to=psom())
> first <- supply(d, function(x) as.character(x$date[1]))
> plot(sort(as.Date(first)))
> range(as.Date(first))
> last <- supply(d, function(v) as.character(v$date[1]))
> [1] "1823-10-31" "1972-11-01"
> plot(sort(as.Date(first)))
> range(as.Date(first))
> [1] "1823-10-31" "1972-11-01"
>
```

# RStudio configuration

keyboard shortcuts (ALT+SHIFT+K)

# RStudio configuration

keyboard shortcuts (ALT+SHIFT+K)

Recommended settings for reproducible research under

Tools - Global Options - General

**OFF:** Restore .Rdata into workspace at startup

Save workspace to .RData on exit: **NEVER**

# RStudio configuration

keyboard shortcuts (ALT+SHIFT+K)

Recommended settings for reproducible research under

Tools - Global Options - General

**OFF:** Restore .Rdata into workspace at startup

Save workspace to .RData on exit: **NEVER**

*Instead use `save(object, file="object.Rdata")` after long computations. You can load them later with `load("object.Rdata")`.*

# RStudio configuration

## keyboard shortcuts (ALT+SHIFT+K)

Recommended settings for reproducible research under

**Tools - Global Options - General**

**OFF:** Restore .Rdata into workspace at startup

Save workspace to .RData on exit: **NEVER**

*Instead use `save(object, file="object.Rdata")` after long computations. You can load them later with `load("object.Rdata")`.*

More at [github.com/brry/course#settings](https://github.com/brry/course#settings)

# Tinn R

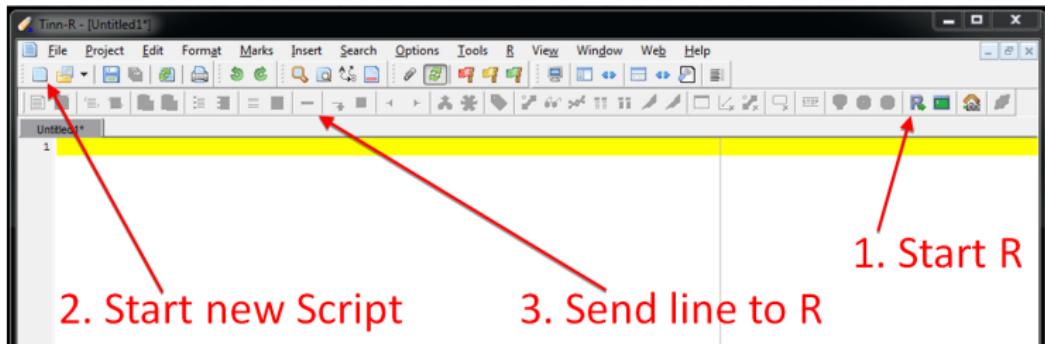


Figure: screenshot Tinn-R

```
# enable sending of blocks up to the next blank line:  
.trPaths <- paste(Sys.getenv('LOCALAPPDATA'),  
'\\Temp\\Tinn-R', c('', 'search.txt', 'objects.txt',  
'file.r', 'selection.r', 'block.r','lines.r'),sep='\\')  
# write this in C:/Program Files/R/R-3.2.2/etc/Rprofile.site
```

# Outline

R course info

1. One-Session-Intro

**2. Getting started: Background, GUI and first steps**

Background information

Graphical user interfaces

**Getting help**

R syntax

Source code of R commands

3. Objects

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

# Getting help: Inside R

## R-Help

```
help("apply") # get help for the function
?apply # The same. saves typing effort.
# Press F1 while the cursor is at a command.

apropos("apply")
help.search("apply")
??apply

help.start() # html help Manuals, Material, search engine
```

# Getting help: Outside R

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#),  
RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#),  
[Tanbakuchi](#), [cuni](#) or [Joergensen](#)

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#),  
RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#),  
[Tanbakuchi](#), [cuni](#) or [Joergensen](#)
- [rseek.org](#) for R related internet search

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#),  
RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#),  
[Tanbakuchi](#), [cuni](#) or [Joergensen](#)
- [rseek.org](#) for R related internet search
- [StackOverflow](#) for programming questions <- **main resource**

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#),  
RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#),  
[Tanbakuchi](#), [cuni](#) or [Joergensen](#)
- [rseek.org](#) for R related internet search
- [StackOverflow](#) for programming questions <- **main resource**
- [CrossValidated](#) for statistical questions

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#),  
RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#),  
[Tanbakuchi](#), [cuni](#) or [Joergensen](#)
- [rseek.org](#) for R related internet search
- [StackOverflow](#) for programming questions <- **main resource**
- [CrossValidated](#) for statistical questions
- Mailing lists: ask other experienced users for help

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#),  
RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#),  
[Tanbakuchi](#), [cuni](#) or [Joergensen](#)
- [rseek.org](#) for R related internet search
- [StackOverflow](#) for programming questions <- **main resource**
- [CrossValidated](#) for statistical questions
- Mailing lists: ask other experienced users for help
  - [R-help mailing list](#)

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#),  
RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#),  
[Tanbakuchi](#), [cuni](#) or [Joergensen](#)
- [rseek.org](#) for R related internet search
- [StackOverflow](#) for programming questions <- **main resource**
- [CrossValidated](#) for statistical questions
- Mailing lists: ask other experienced users for help
  - [R-help mailing list](#)
  - [Nabble](#): Forum-like view of the r-help mailing list

# Getting help: Outside R

- [R-Manuals](#) for introduction to the language
- R Wiki: [rwiki.sciviews.org](http://rwiki.sciviews.org)
- Ref Cards: [base](#) and [advanced](#) cheatsheets from [Rstudio](#), RefCards originally by Tom Short from [Stein](#) (recommended), [Baggot](#), [Tanbakuchi](#), [cuni](#) or [Joergensen](#)
- [rseek.org](#) for R related internet search
- [StackOverflow](#) for programming questions <- **main resource**
- [CrossValidated](#) for statistical questions
- Mailing lists: ask other experienced users for help
  - [R-help mailing list](#)
  - [Nabble](#): Forum-like view of the r-help mailing list
  - ONLY post on mailing lists when all other options are exploited and you are sure there's no answer out there. FIRST read the [posting guide](#).

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)

# Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)
- M. Crawley (2007): The R-book

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)
- M. Crawley (2007): The R-book
- H. Wickham (2014) [Advanced R](#)

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)
- M. Crawley (2007): The R-book
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)
- M. Crawley (2007): The R-book
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)
- domain specific: [Chapman and Hall R Series](#)

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)
- M. Crawley (2007): The R-book
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)
- domain specific: [Chapman and Hall R Series](#)
- Many more listed at [github.com/RomanTsegelskyi/rbooks](https://github.com/RomanTsegelskyi/rbooks)

## Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)
- M. Crawley (2007): The R-book
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)
- domain specific: [Chapman and Hall R Series](#)
- Many more listed at [github.com/RomanTsegelskyi/rbooks](https://github.com/RomanTsegelskyi/rbooks)
- Review list at [ecotope.org](http://ecotope.org), through the wayback machine

# Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): R in a Nutshell
- U. Ligges (2008): Programmieren mit R (German)
- M. Crawley (2007): The R-book
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)
- domain specific: [Chapman and Hall R Series](#)
- Many more listed at [github.com/RomanTsegelskyi/rbooks](https://github.com/RomanTsegelskyi/rbooks)
- Review list at [ecotope.org](http://ecotope.org), through the wayback machine

Tutorials:

# Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): [R in a Nutshell](#)
- U. Ligges (2008): [Programmieren mit R \(German\)](#)
- M. Crawley (2007): [The R-book](#)
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)
- domain specific: [Chapman and Hall R Series](#)
- Many more listed at [github.com/RomanTsegelskyi/rbooks](#)
- Review list at [ecotope.org](#), through the wayback machine

## Tutorials:

- Microsoft + Datacamp (best course I know of, with login, but free)

# Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): [R in a Nutshell](#)
- U. Ligges (2008): [Programmieren mit R \(German\)](#)
- M. Crawley (2007): [The R-book](#)
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)
- domain specific: [Chapman and Hall R Series](#)
- Many more listed at [github.com/RomanTsegelskyi/rbooks](#)
- Review list at [ecotope.org](#), through the wayback machine

## Tutorials:

- Microsoft + Datacamp (best course I know of, with login, but free)
- [tryR.codeschool.com](#) (interactive)

# Good References, books

- Grolemund & Wickham (2017) [R for Data Science](#)
- J. Adler (2010): [R in a Nutshell](#)
- U. Ligges (2008): [Programmieren mit R \(German\)](#)
- M. Crawley (2007): [The R-book](#)
- H. Wickham (2014) [Advanced R](#)
- H. Wickham (2015) [R Packages](#)
- domain specific: [Chapman and Hall R Series](#)
- Many more listed at [github.com/RomanTsegelskyi/rbooks](#)
- Review list at [ecotope.org](#), through the wayback machine

## Tutorials:

- Microsoft + Datacamp (best course I know of, with login, but free)
- [tryR.codeschool.com](#) (interactive)
- STAT 545

# Outline

R course info

1. One-Session-Intro

2. Getting started: Background, GUI and first steps

- Background information

- Graphical user interfaces

- Getting help

**R syntax**

- Source code of R commands

3. Objects

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

## get started, comments

Please start R via the editor you like.

Comments start with a hash key #

```
17-42 # This is a comment, thus not executed
```

```
## [1] -25
```

# Please calculate the following tasks

## Exercise 8: Basic operators

- ①  $4 * 9$
- ②  $2 + 4,8$  (International decimal sign: ".")
- ③  $3^2$
- ④  $\frac{\pi}{2}$
- ⑤ sinus of  $15^\circ$ . Must be 0.26. Remember:  $\text{radians} = \text{degrees} * \frac{\pi}{180^\circ}$
- ⑥  $\sqrt{81}$  (square root)
- ⑦  $| -12 |$  (absolute value)
- ⑧  $\log(100)$  ( $\rightarrow$  should return 2, as  $10^2 = 100$  )
- ⑨  $5!$  (factorial)
- ⑩  $e^3$  (exponential function)
- ⑪ Bonus: How is  $3.91 * 10^{-3}$  written in scientific notation?

## Solution for exercise 8.1-8: Basic operators

```
4*9 # this is our first calculation with a comment
2 + 4.8
3^2
pi/2
sin(15*pi/180)
?sqrt # get the offline help about the function sqrt
sqrt(81) # square root
abs(-12) # absolute value
log(100)
help(log) # natural logarithm
log10(100) # log base 10
log(100, base=10) # another way
```

## Solution for exercise 8.9-11: Basic operators

```
factorial(5)
e^3 # does not work
??exponential
exp(3) # exponential function
exp(1) # the value of e
3.91 * 10^-3
3.91e-3 # no spaces possible in scientific notation
help.start() # offline search, manuals, etc
```

# Outline

R course info

1. One-Session-Intro

2. Getting started: Background, GUI and first steps

Background information

Graphical user interfaces

Getting help

R syntax

Source code of R commands

3. Objects

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

Source code: [github.com/wch/r-source/src/library/base](https://github.com/wch/r-source/src/library/base) (or stats, utils, ...)

CRAN packages source code: [github.com/cran](https://github.com/cran)

HTML Documentation: [www.rdocumentation.org](http://www.rdocumentation.org)

[excellent post](#)

`ncol - .Primitive("dim") - NROW head.matrix`

<https://github.com/wch/r-source/blob/trunk/src/library/utils/R/head.R>

<https://github.com/wch/r-source/blob/trunk/src/library/base/R/matrix.R>

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps

## 3. Objects

Assignments

Vectors, indexing

Data.frames

Matrices

Lists, lapply/sapply functions

Classes, methods, functions

## 4. "Real" data

## 5. Plotting

## 6. Character Operations

## 7. Conditions & loops

## 8. Functions & packages

## 9. Distributions

## 10. Statistics

## 11. Time series handling and analysis

## 12. Spatial data and GIS functionality

## 13. Final tasks

Course plan

# Assignment: Create an object with data in the workspace I

```
a <- 15.4
```

# Assignment: Create an object with data in the workspace I

```
a <- 15.4
```

```
a
```

```
## [1] 15.4
```

# Assignment: Create an object with data in the workspace I

```
a <- 15.4
```

```
a
```

```
## [1] 15.4
```

*a > 2 # is a bigger than 2? --> logical value (boolean)*

```
## [1] TRUE
```

# Assignment: Create an object with data in the workspace I

```
a <- 15.4
```

```
a
```

```
## [1] 15.4
```

*a > 2 # is a bigger than 2? --> logical value (boolean)*

```
## [1] TRUE
```

*A # not an existing object (R is case sensitive)*

```
## Error in eval(expr, envir, enclos): object 'A' not found
```

## Assignment: Create an object with data in the workspace II

Don't assign to existing objects or functions like `mean`, `c`, `T`, `data`, `sin`; Later on, you will hardly be able to distinguish what object (original or custom) you are referring to.

## Assignment: Create an object with data in the workspace II

Don't assign to existing objects or functions like `mean`, `c`, `T`, `data`, `sin`; Later on, you will hardly be able to distinguish what object (original or custom) you are referring to.

As R is Case Sensitive, `Data` would be OK, but it's not useful. Use object names that are short, but still meaningful.

## Assignment: Create an object with data in the workspace II

Don't assign to existing objects or functions like `mean`, `c`, `T`, `data`, `sin`; Later on, you will hardly be able to distinguish what object (original or custom) you are referring to.

As R is Case Sensitive, `Data` would be OK, but it's not useful. Use object names that are short, but still meaningful.

One convention is lowerCamel:

`dailyRainBerlin` is short and readable through capital letters.

## Assignment: Create an object with data in the workspace II

Don't assign to existing objects or functions like `mean`, `c`, `T`, `data`, `sin`; Later on, you will hardly be able to distinguish what object (original or custom) you are referring to.

As R is Case Sensitive, `Data` would be OK, but it's not useful. Use object names that are short, but still meaningful.

One convention is lowerCamel:

`dailyRainBerlin` is short and readable through capital letters.

```
# Rstudio: press 'ALT' + '-' for " <- "
```

## Assignment: Create an object with data in the workspace II

Don't assign to existing objects or functions like `mean`, `c`, `T`, `data`, `sin`; Later on, you will hardly be able to distinguish what object (original or custom) you are referring to.

As R is Case Sensitive, `Data` would be OK, but it's not useful. Use object names that are short, but still meaningful.

One convention is lowerCamel:

`dailyRainBerlin` is short and readable through capital letters.

```
# Rstudio: press 'ALT' + '-' for " <- "
```

Don't use `=` for assignments. Most R users connotate "`=`" with function arguments and prefer "`<-`" for objects ([style guide](#)). Also, "`=`" is only allowed on the top level:

## Assignment: Create an object with data in the workspace II

Don't assign to existing objects or functions like `mean`, `c`, `T`, `data`, `sin`; Later on, you will hardly be able to distinguish what object (original or custom) you are referring to.

As R is Case Sensitive, `Data` would be OK, but it's not useful. Use object names that are short, but still meaningful.

One convention is lowerCamel:

`dailyRainBerlin` is short and readable through capital letters.

```
# Rstudio: press 'ALT' + '-' for " <- "
```

Don't use `=` for assignments. Most R users connotate "`=`" with function arguments and prefer "`<-`" for objects ([style guide](#)). Also, "`=`" is only allowed on the top level:

`median(x <- 1:10)` also creates the object in `globalenv()`, while `median(x=1:10)` does not. This is used in `NCOL`, for example.

## Assignment: Create an object with data in the workspace II

Don't assign to existing objects or functions like `mean`, `c`, `T`, `data`, `sin`; Later on, you will hardly be able to distinguish what object (original or custom) you are referring to.

As R is Case Sensitive, `Data` would be OK, but it's not useful. Use object names that are short, but still meaningful.

One convention is lowerCamel:

`dailyRainBerlin` is short and readable through capital letters.

```
# Rstudio: press 'ALT' + '-' for " <- "
```

Don't use `=` for assignments. Most R users connotate "`=`" with function arguments and prefer "`<-`" for objects ([style guide](#)). Also, "`=`" is only allowed on the top level:

`median(x <- 1:10)` also creates the object in `globalenv()`, while  
`median(x=1:10)` does not. This is used in `NCOL`, for example.

<http://blog.revolutionanalytics.com/2008/12/use-equals-or-arrow-for-assignment.html>

<https://csgillespie.wordpress.com/2010/11/16/assignment-operators-in-r-vs/>

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps

## 3. Objects

Assignments

Vectors, indexing

Data.frames

Matrices

Lists, lapply/sapply functions

Classes, methods, functions

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

# Vectors I

In R, Vectors are not geometric constructs, but an ordered set of values. Create a vector with `c` (stands for Combine or Concatenate), separating each entry with a comma:

# Vectors I

In R, Vectors are not geometric constructs, but an ordered set of values. Create a vector with `c` (stands for Combine or Concatenate), separating each entry with a comma:

```
b <- c(3, 7, -2.5, 11, 3.8, 9)
```

# Vectors I

In R, Vectors are not geometric constructs, but an ordered set of values. Create a vector with `c` (stands for Combine or Concatenate), separating each entry with a comma:

```
b <- c(3, 7, -2.5, 11, 3.8, 9)
```

spaces make code easier to read:

```
b<-c(3,7,-2.5,11,3.8,9)  
b <- c(3, 7, -2.5, 11, 3.8, 9)
```

# Vectors I

In R, Vectors are not geometric constructs, but an ordered set of values. Create a vector with `c` (stands for Combine or Concatenate), separating each entry with a comma:

```
b <- c(3, 7, -2.5, 11, 3.8, 9)
```

spaces make code easier to read:

```
b<-c(3,7,-2.5,11,3.8,9)  
b <- c(3, 7, -2.5, 11, 3.8, 9)
```

```
print(b) # or just write b  
## [1] 3.0 7.0 -2.5 11.0 3.8 9.0
```

## Vectors II: Other ways to generate vectors

## Vectors II: Other ways to generate vectors

```
1:5 # integer numbers from : to  
## [1] 1 2 3 4 5
```

## Vectors II: Other ways to generate vectors

```
1:5 # integer numbers from : to
```

```
## [1] 1 2 3 4 5
```

```
rep(1:4, times=3) # repeat entries several times
```

```
## [1] 1 2 3 4 1 2 3 4 1 2 3 4
```

## Vectors II: Other ways to generate vectors

```
1:5 # integer numbers from : to
```

```
## [1] 1 2 3 4 5
```

```
rep(1:4, times=3) # repeat entries several times
```

```
## [1] 1 2 3 4 1 2 3 4 1 2 3 4
```

```
v <- rep(1:3, each=3, times=2)
```

```
v
```

```
## [1] 1 1 1 2 2 2 3 3 3 1 1 1 2 2 2 3 3 3
```

## Vectors III: Yet other ways to generate vectors

## Vectors III: Yet other ways to generate vectors

```
seq(from=5, to=-1, by=-0.5) # Sequence  
# by must be negative for descending sequences  
  
## [1] 5.0 4.5 4.0 3.5 3.0 2.5 2.0 1.5 1.0  
## [10] 0.5 0.0 -0.5 -1.0
```

## Vectors III: Yet other ways to generate vectors

```
seq(from=5, to=-1, by=-0.5) # Sequence  
# by must be negative for descending sequences  
  
## [1] 5.0 4.5 4.0 3.5 3.0 2.5 2.0 1.5 1.0  
## [10] 0.5 0.0 -0.5 -1.0
```

```
seq(1.32, 6.1, length.out=15) # 15 elements  
  
## [1] 1.320000 1.661429 2.002857 2.344286 2.685714  
## [6] 3.027143 3.368571 3.710000 4.051429 4.392857  
## [11] 4.734286 5.075714 5.417143 5.758571 6.100000
```

## Vectors III: Yet other ways to generate vectors

```
seq(from=5, to=-1, by=-0.5) # Sequence  
# by must be negative for descending sequences  
  
## [1] 5.0 4.5 4.0 3.5 3.0 2.5 2.0 1.5 1.0  
## [10] 0.5 0.0 -0.5 -1.0
```

```
seq(1.32, 6.1, length.out=15) # 15 elements  
  
## [1] 1.320000 1.661429 2.002857 2.344286 2.685714  
## [6] 3.027143 3.368571 3.710000 4.051429 4.392857  
## [11] 4.734286 5.075714 5.417143 5.758571 6.100000
```

```
seq(1.32, 6.1, len=15) # abbreviate argument names  
?seq # read more about it...
```

## Indexing on vectors (subsets) I: square brackets

```
b # remember b?  
## [1] 3.0 7.0 -2.5 11.0 3.8 9.0  
  
b[1] # first element  
## [1] 3
```

## Indexing on vectors (subsets) I: square brackets

```
b # remember b?  
## [1] 3.0 7.0 -2.5 11.0 3.8 9.0  
  
b[1] # first element  
## [1] 3
```

```
b[2:4] # several elements  
## [1] 7.0 -2.5 11.0  
  
b[ c(2,5,1,6,1) ] # flexible  
## [1] 7.0 3.8 3.0 9.0 3.0
```

## Indexing on vectors (subsets) II

```
b
```

```
## [1] 3.0 7.0 -2.5 11.0 3.8 9.0
```

```
b[-2] # all but the second element
```

```
## [1] 3.0 -2.5 11.0 3.8 9.0
```

## Indexing on vectors (subsets) II

```
b
```

```
## [1] 3.0 7.0 -2.5 11.0 3.8 9.0
```

```
b[-2] # all but the second element
```

```
## [1] 3.0 -2.5 11.0 3.8 9.0
```

```
a <- seq(from=1, to=100, by=0.1)
```

```
head(a) # only first 6 elements (or rows)
```

```
## [1] 1.0 1.1 1.2 1.3 1.4 1.5
```

## Indexing on vectors (subsets) III

```
head(a)

## [1] 1.0 1.1 1.2 1.3 1.4 1.5

a[2] <- 87 # manipulate single elements of object
head(a)

## [1] 1.0 87.0 1.2 1.3 1.4 1.5
```

```
ls() # list objects in workspace

## [1] "a" "b" "v"

rm(b) # remove an object
```

# Data types

R can handle data of a variety of types: Numbers like `42.6`, characters like `"R rocks"`, complex numbers like `8+4i`, categories with `as.factor(c("blue", "green"))`, logical or boolean values `c(TRUE, FALSE)` and a few others. They should usually not be mixed. For example, you can't perform algebraic computation on characters:

# Data types

R can handle data of a variety of types: Numbers like `42.6`, characters like `"R rocks"`, complex numbers like `8+4i`, categories with `as.factor(c("blue", "green"))`, logical or boolean values `c(TRUE, FALSE)` and a few others. They should usually not be mixed. For example, you can't perform algebraic computation on characters:

```
(p9 <- "1.3")      # show assignment result
2*p9                # error, as p9 is a character
2*as.numeric(p9)    # make it a number first
p9                  # p9 itself has not changed
is.numeric(p9)      # FALSE
mode(p9)            # character
```

## Integer vs decimal numbers: arithmetic precision

```
6 == 3 + 3           # TRUE
```

## Integer vs decimal numbers: arithmetic precision

```
6 == 3 + 3      # TRUE
```

```
0.5 - 0.2 == 0.3      # TRUE
```

## Integer vs decimal numbers: arithmetic precision

```
6 == 3 + 3      # TRUE
```

```
0.5 - 0.2 == 0.3      # TRUE
```

```
0.4 - 0.1 == 0.3      # FALSE
```

## Integer vs decimal numbers: arithmetic precision

```
6 == 3 + 3           # TRUE
```

```
0.5 - 0.2 == 0.3   # TRUE
```

```
0.4 - 0.1 == 0.3   # FALSE
```

```
print(0.4-0.1 , digits=22)
```

```
## [1] 0.30000000000000004
```

## Integer vs decimal numbers: arithmetic precision

```
6 == 3 + 3           # TRUE
```

```
0.5 - 0.2 == 0.3   # TRUE
```

```
0.4 - 0.1 == 0.3   # FALSE
```

```
print(0.4-0.1, digits=22)
```

```
## [1] 0.3000000000000004
```

```
round(0.4-0.1, digits=6) == round(0.3, 6)
```

```
## [1] TRUE
```

## Integer vs decimal numbers: arithmetic precision

```
6 == 3 + 3           # TRUE
```

```
0.5 - 0.2 == 0.3    # TRUE
```

```
0.4 - 0.1 == 0.3    # FALSE
```

```
print(0.4-0.1, digits=22)
```

```
## [1] 0.3000000000000004
```

```
round(0.4-0.1, digits=6) == round(0.3, 6)
```

```
## [1] TRUE
```

Test for equality only with integers / with rounding (see also `signif`), or use `all.equal`.

## Integer vs decimal numbers: arithmetic precision

```
6 == 3 + 3           # TRUE
```

```
0.5 - 0.2 == 0.3    # TRUE
```

```
0.4 - 0.1 == 0.3    # FALSE
```

```
print(0.4-0.1, digits=22)
```

```
## [1] 0.3000000000000004
```

```
round(0.4-0.1, digits=6) == round(0.3, 6)
```

```
## [1] TRUE
```

Test for equality only with integers / with rounding (see also `signif`), or use `all.equal`. More pitfalls like this in the [R inferno](#).

## Exercise 9: Vectors and statistics

- ① Repeat your favorite number 6 times using `rep`.
- ② With `seq`uence, generate one vector from -1 to 4 in steps of 0.5 and one with a length of 21 elements. Then add a number to the vector.
- ③ Assign a vector with the body heights of a few people around you to an object with a useful name.
- ④ Print only the 4th and 2nd element (one single vector). BONUS: All but the 2nd to 4th elements.
- ⑤ `sort` the numbers decreasingly.
- ⑥ Calculate the average size (`mean`), the variance (`var`) and the standard deviation (`sd`) of this dataset, as well as the `min`, `max`, `median` and 80% `quantile` (20% of the entries are larger than this value).
- ⑦ BONUS 1: Explain what `length` does.
- ⑧ BONUS 2: Generate the numbers from 1 to 10 in three different ways.
- ⑨ BONUS 3: Explain and relate to each other: `mean`, `var`, `sd`, `median`, `quantile`.

## Solution for exercise 9.1-2: Vectors

```
# repeat and sequence:  
rep(5, times=6)  
seq(from=-1, to=4, by=0.5)  
vec2 <- seq(-1, 4, length.out=21)  
seq(-1, 4, len=21)
```

You can leave argument names away, if you specify their content in the right order.  
If it is unambiguous, you can use partial argument names.

```
# add a number to a vector:  
c(vec2, 8) # does not change vec2  
vec3 <- c(vec2, 8) # creates a new object  
vec2 <- c(vec2, 8) # changes the object vec2  
# Another way:  
vec2 <- seq(-1, 4, length.out=21)  
vec2[21] <- 5 # changing one element  
vec2  
vec2[22] # returns NA - there is no value here  
vec2[26] <- 8  
vec2 # is now changed
```

## Solution for exercise 9.3-6: Vectors

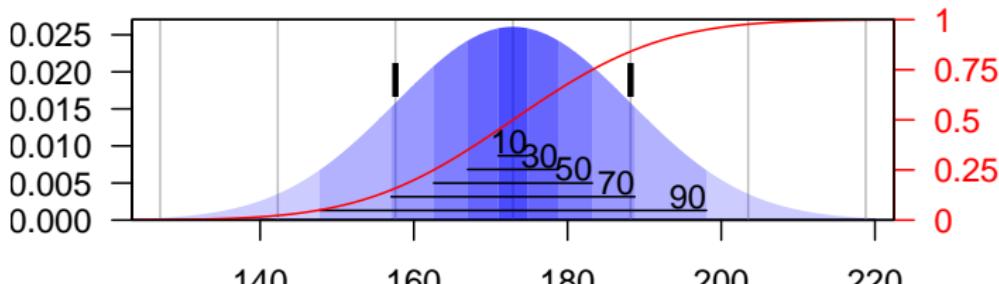
```
# vector with body heights:  
heights <- c(192, 171, 178, 175, 185, 145, 164) # cm  
# print selected elements:  
heights[ c(4,2) ]  
# sort:  
sort(heights, decreasing=T)  
# statistics:  
mean(heights)  
var(heights) # cm^2  
sd(heights) # cm  
min(heights)  
max(heights)  
median(heights)  
quantile(heights, probs=0.80)
```

# Solution for exercise 9.BONUS: Vectors

```
length(heights) # number of elements in vector  
# Numbers from 1 to 10:  
1:10 ; seq(1,10) ; c(1,2,3,4,5,6,7,8,9,10)  
# mean vs median:  
mean(replace(heights, 1, 292)) # subject to outliers  
median(replace(heights, 1, 292)) # indifferent to extrema
```

70% of values in a normal distribution lie between  $\text{mean}-\text{sd}$  and  $\text{mean}+\text{sd}$ :

**Normal density with  
mean = 170 and sd = 15**



# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps

## 3. Objects

Assignments

Vectors, indexing

### **Data.frames**

Matrices

Lists, lapply/sapply functions

Classes, methods, functions

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

## `data.frames`

`data.frames` are the most common form of tables in R.

## `data.frames`

`data.frames` are the most common form of tables in R. Each column can have a different data type (numeric, character, factor, etc), but within a column, it stays the same.

## `data.frames`

`data.frames` are the most common form of tables in R. Each column can have a different data type (numeric, character, factor, etc), but within a column, it stays the same.

Columns as well as rows can have names (`colnames`, `rownames`). If these start with a number, an "X" will be prefixed.

## `data.frames`

`data.frames` are the most common form of tables in R. Each column can have a different data type (numeric, character, factor, etc), but within a column, it stays the same.

Columns as well as rows can have names (`colnames`, `rownames`). If these start with a number, an "X" will be prefixed. This may happen when reading data, which is usually stored as a `data.frame` in R.

## `data.frames`

`data.frames` are the most common form of tables in R. Each column can have a different data type (numeric, character, factor, etc), but within a column, it stays the same.

Columns as well as rows can have names (`colnames`, `rownames`). If these start with a number, an "X" will be prefixed. This may happen when reading data, which is usually stored as a `data.frame` in R.

```
?data.frame
```

# Creating data.frames

```
df <- data.frame(col_a=11:14, ColName2=letters[1:4],  
                  LogicalCol=(1:4)>2)  
df  
  
##   col_a ColName2 LogicalCol  
## 1     11      a     FALSE  
## 2     12      b     FALSE  
## 3     13      c      TRUE  
## 4     14      d      TRUE  
  
str(df)  
  
## 'data.frame': 4 obs. of  3 variables:  
##   $ col_a    : int  11 12 13 14  
##   $ ColName2 : Factor w/ 4 levels "a","b","c","d": 1 2 3 4  
##   $ LogicalCol: logi  FALSE FALSE TRUE TRUE
```

## Subsetting data.frames by index numbers

```
nrow(df)  
## [1] 4  
  
ncol(df) # see also dim(df)  
## [1] 3
```

## Subsetting data.frames by index numbers

```
nrow(df)  
## [1] 4  
  
ncol(df) # see also dim(df)  
## [1] 3
```

```
df[ 3 , 1] # 3rd row, 1st column  
## [1] 13  
  
df[ , 2] # all rows, 2nd column (as vector)  
## [1] a b c d  
## Levels: a b c d
```

## Subsetting data.frames by names

```
df[ "LogicalCol" ] # no comma -> data.frame  
  
## LogicalCol  
## 1 FALSE  
## 2 FALSE  
## 3 TRUE  
## 4 TRUE
```

Bad practice, unclear whether rows or columns are indexed!

```
df[ , "LogicalCol", drop=FALSE ] # better
```

## Subsetting data.frames by names

```
df[ "LogicalCol" ] # no comma -> data.frame  
  
## LogicalCol  
## 1 FALSE  
## 2 FALSE  
## 3 TRUE  
## 4 TRUE
```

Bad practice, unclear whether rows or columns are indexed!

```
df[ , "LogicalCol", drop=FALSE ] # better
```

```
df$col_a  
  
## [1] 11 12 13 14
```

## Subsetting data.frames by names

```
df[ "LogicalCol" ] # no comma -> data.frame  
  
## LogicalCol  
## 1 FALSE  
## 2 FALSE  
## 3 TRUE  
## 4 TRUE
```

Bad practice, unclear whether rows or columns are indexed!

```
df[ , "LogicalCol", drop=FALSE ] # better
```

```
df$col_a  
  
## [1] 11 12 13 14
```

```
df[2, ] # full row  
  
## col_a ColName2 LogicalCol  
## 2 12 b FALSE
```

# Some hints about data.frames

- If `nrow(df)` returns `NULL`, `df` might in fact be a vector.

## Some hints about data.frames

- If `nrow(df)` returns `NULL`, `df` might in fact be a vector.
- `NROW(df)` will show `length(df)` if `df` is a vector.

# Some hints about data.frames

- If `nrow(df)` returns `NULL`, `df` might in fact be a vector.
- `NROW(df)` will show `length(df)` if `df` is a vector.
- Also, remember you can check the object with `class` and `mode`, as well as `is.vector`.

## Some hints about data.frames

- If `nrow(df)` returns `NULL`, `df` might in fact be a vector.
- `NROW(df)` will show `length(df)` if `df` is a vector.
- Also, remember you can check the object with `class` and `mode`, as well as `is.vector`.
- `colnames(df)` will show the column names. (see also `rownames`)

## Exercise 10: data.frames

Working with the dataset `state.x77`

- ① Change the object type to a `data.frame`
- ② What are the 3 ways to select the population column? Which one doesn't work for Life expectancy? Why not?
- ③ How can you subset a single column, without the result being simplified to a vector? (you want to keep it as a `data.frame`, but with one single column only. Hint: `?"[ ]"`)
- ④ How could the `$` subsetting method be dangerous when misspelling columns?
- ⑤ By which two ways can you select the Income and Illiteracy columns?
- ⑥ How can you select all the states (=rows) with an income above 5000?
- ⑦ BONUS: Why does `state.x77$Income` not work on the original dataset? (use `rm(state.x77$Income)` to remove your changed object from the workspace, if you didn't assign it to a different name.)

## Solution to exercise 10: subsetting data.frames

```
state <- as.data.frame(state.x77)
state[,1] # not safe, columns could be in different position
state[ , "Population"] # safe and readable
state$Population # quicker to type, has autocomplete
# $ doesn't work (at least not without quotation marks)
# with a space, it's an invalid column name
state[ , "Population", drop=FALSE]
state$Iilleeteeraacyy # does not give an error!!!
state[ , c("Income","Illiteracy")]
state[ , 2:3]
state[state$Income>=5000, ]
# Because it is a matrix, not a data.frame
```

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps

## 3. Objects

Assignments

Vectors, indexing

Data.frames

## Matrices

Lists, lapply/sapply functions

Classes, methods, functions

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

# Creating matrices

```
matrix(data=1:6 , nrow=2, ncol=3)

##      [,1] [,2] [,3]
## [1,]    1    3    5
## [2,]    2    4    6
```

# Creating matrices

```
matrix(data=1:6 , nrow=2, ncol=3)

##      [,1] [,2] [,3]
## [1,]    1    3    5
## [2,]    2    4    6
```

```
m <- matrix(1:6 , nrow=2, ncol=3, byrow=T)
print(m)
```

```
##      [,1] [,2] [,3]
## [1,]    1    2    3
## [2,]    4    5    6
```

## Manipulating matrices, row and column names

```
colnames(m) <- c("A", "B", "C")
m[1,1] <- c(989)
print(m)
```

```
##          A B C
## [1,]  989 2 3
## [2,]    4 5 6
```

# Manipulating matrices, row and column names

```
colnames(m) <- c("A", "B", "C")
m[1,1] <- c(989)
print(m)
```

```
##          A B C
## [1,] 989 2 3
## [2,]    4 5 6
```

```
rownames(m) <- c("row1", "row2")
```

```
m
```

```
##          A B C
## row1 989 2 3
## row2    4 5 6
```

# Matrices only have one single data type

```
m[1,1] <- "a"  
m  
  
##      A     B     C  
## row1 "a"  "2"  "3"  
## row2 "4"  "5"  "6"
```

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps

## 3. Objects

Assignments

Vectors, indexing

Data.frames

Matrices

### Lists, lapply/sapply functions

Classes, methods, functions

4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

Course plan

# Lists

Lists store every R data type (vectors, data.frames, ...) in one variable

```
list("df"=df[c(1:2), ], matrix=m, character="TEXT")  
  
## $df  
##   col_a ColName2 LogicalCol  
## 1     11         a     FALSE  
## 2     12         b     FALSE  
##  
## $matrix  
##       A     B     C  
## row1 "a"  "2"  "3"  
## row2 "4"  "5"  "6"  
##  
## $character  
## [1] "TEXT"
```

# List basics I

```
# create a list: an object with several 'objects' inside it
L1 <- list(first="result of function 'list'",  
           second=345:287, third=rnorm(10),  
           fourth=data.frame(a=6:3, b=rexp(4)),  
           fifth=LETTERS, sixth=letters,  
           seventh=sample(300), eighth="goodbye")
```

# List basics I

```
# create a list: an object with several 'objects' inside it
L1 <- list(first="result of function 'list'",  
           second=345:287, third=rnorm(10),  
           fourth=data.frame(a=6:3, b=rexp(4)),  
           fifth=LETTERS, sixth=letters,  
           seventh=sample(300), eighth="goodbye")
```

```
# print the whole thing:
```

```
L1
```

# List basics I

```
# create a list: an object with several 'objects' inside it
L1 <- list(first="result of function 'list'",  
           second=345:287, third=rnorm(10),  
           fourth=data.frame(a=6:3, b=rexp(4)),  
           fifth=LETTERS, sixth=letters,  
           seventh=sample(300), eighth="goodbye")
```

```
# print the whole thing:
```

```
L1
```

```
# print only the first 4 elements of the list
head(L1,4)
tail(L1) # or the last n elements (by default 6)
```

## List basics II

```
# Structure of the list
str(L1, max.level=1, vec.len=3)

## List of 8
## $ first  : chr "result of function 'list'"
## $ second : int [1:59] 345 344 343 342 341 340 339 338 ...
## $ third  : num [1:10] 1.083 0.953 0.269 1.076 ...
## $ fourth : 'data.frame': 4 obs. of  2 variables:
##   $ fifth  : chr [1:26] "A" "B" "C" ...
##   $ sixth  : chr [1:26] "a" "b" "c" ...
##   $ seventh: int [1:300] 19 216 192 48 161 46 73 117 ...
##   $ eighth : chr "goodbye"
```

## List basics II

```
# str down to each level of the list
str(L1, vec.len=3) # see the df for the difference

## List of 8
## $ first  : chr "result of function 'list'"
## $ second : int [1:59] 345 344 343 342 341 340 339 338 ...
## $ third  : num [1:10] 1.083 0.953 0.269 1.076 ...
## $ fourth : 'data.frame': 4 obs. of  2 variables:
##   ..$ a: int [1:4] 6 5 4 3
##   ..$ b: num [1:4] 0.275 0.441 1.693 0.96
## $ fifth  : chr [1:26] "A" "B" "C" ...
## $ sixth  : chr [1:26] "a" "b" "c" ...
## $ seventh: int [1:300] 19 216 192 48 161 46 73 117 ...
## $ eighth : chr "goodbye"
```

# List indexing I

```
# Assessing list elements (indexing) with double square brackets:  
L1[[3]]  
  
## [1] 1.08270524 0.95297954 0.26906649  
## [4] 1.07597085 -0.86401751 0.06248087  
## [7] 0.13752332 0.01085314 0.75895050  
## [10] -1.11800498
```

# List indexing I

*# Assessing list elements (indexing) with double square brackets:*

```
L1[[3]]
```

```
## [1] 1.08270524 0.95297954 0.26906649  
## [4] 1.07597085 -0.86401751 0.06248087  
## [7] 0.13752332 0.01085314 0.75895050  
## [10] -1.11800498
```

*# What's different if you use only single brackets?*

```
L1[3]
```

```
## $third  
## [1] 1.08270524 0.95297954 0.26906649  
## [4] 1.07597085 -0.86401751 0.06248087  
## [7] 0.13752332 0.01085314 0.75895050  
## [10] -1.11800498
```

*# HINT: look at the class of the results*

## List indexing II

```
# Assessing elements by name
L1$eight # NULL if not existent, so watch out with spelling!

## NULL

L1$eighth

## [1] "goodbye"
```

## List indexing II

```
# Assessing elements by name  
L1$eight # NULL if not existent, so watch out with spelling!  
  
## NULL  
  
L1$eighth  
  
## [1] "goodbye"
```

```
names(L1) # as with names(VECTOR) or rownames(DATFRAME)  
  
## [1] "first"    "second"   "third"    "fourth"  
## [5] "fifth"    "sixth"    "seventh"  "eighth"  
  
names(L1)[2] <- "second elem."
```

## List indexing III

```
# Several elements of the list (the result is still a list):  
L1[4:5]
```

```
## $fourth  
##   a           b  
## 1 6 0.2754568  
## 2 5 0.4412032  
## 3 4 1.6930600  
## 4 3 0.9602851  
##  
## $fifth  
## [1] "A"  "B"  "C"  "D"  "E"  "F"  "G"  "H"  "I"  "J"  "K"  
## [12] "L"  "M"  "N"  "O"  "P"  "Q"  "R"  "S"  "T"  "U"  "V"  
## [23] "W"  "X"  "Y"  "Z"
```

## List indexing IV

```
# changing elements:  
L1[c(3,6)] <- c(77777777, 3333333)  
head(L1)
```

# L-apply

```
typeof(L1[[1]])
```

```
## [1] "character"
```

## L-apply

```
typeof(L1[[1]])
```

```
## [1] "character"
```

*# Apply a function to each element of the list:*

```
lapply(X=L1, FUN=typeof)
```

# L-apply

```
typeof(L1[[1]])
```

```
## [1] "character"
```

*# Apply a function to each element of the list:*

```
lapply(X=L1, FUN=typeof)
```

```
## $first                  ## $fifth
## [1] "character"        ## [1] "character"
## 
## $second elem.~          ## $sixth
## [1] "integer"            ## [1] "character"
## 
## $third                  ## $seventh
## [1] "double"              ## [1] "integer"
## 
## $fourth                 ## $eigth
## [1] "list"                ## [1] "character"
```

# S-apply |

```
# simplify that output (if possible):
sapply(X=L1, FUN=typeof)

##      first second elem.      third      fourth
## "character"    "integer"    "double"    "list"
##      fifth       sixth      seventh      eighth
## "character"  "character"  "integer"  "character"
```

# S-apply |

```
# simplify that output (if possible):
sapply(X=L1, FUN=typeof)

##      first second elem.      third      fourth
## "character"    "integer"    "double"    "list"
##      fifth       sixth      seventh      eighth
## "character"  "character"  "integer"  "character"
```

```
# simplifying not possible if FUN(L1[[1]]) and FUN(L1[[2]])
# return different results:
sapply(X=L1, FUN=head)
```

## S-apply II

```
# List with vectors:  
L2 <- list(3:7, 8:4, 5:9)  
L2  
  
## [[1]]  
## [1] 3 4 5 6 7  
##  
## [[2]]  
## [1] 8 7 6 5 4  
##  
## [[3]]  
## [1] 5 6 7 8 9
```

## S-apply II

```
# List with vectors:  
L2 <- list(3:7, 8:4, 5:9)  
  
L2  
  
## [[1]]  
## [1] 3 4 5 6 7  
##  
## [[2]]  
## [1] 8 7 6 5 4  
##  
## [[3]]  
## [1] 5 6 7 8 9
```

```
sapply(L2, I)
```

```
##      [,1] [,2] [,3]  
## [1,]    3    8    5  
## [2,]    4    7    6  
## [3,]    5    6    7  
## [4,]    6    5    8  
## [5,]    7    4    9
```

I basically returns the input as is.  
This is useful for tapply results!

## S-apply III

```
# That doesn't work for vectors of different lengths:  
L3 <- list(AB=c(6,9,2,6), BC=1:8, CD=c(-3,2) )  
sapply(L3, I)  
  
## $AB  
## [1] 6 9 2 6  
##  
## $BC  
## [1] 1 2 3 4 5 6 7 8  
##  
## $CD  
## [1] -3 2
```

## S-apply IV

```
# the solution: berryFunctions::l2df
# list to data.frame conversion
if(!require(berryFunctions)) install.packages("berryFunctions")
library(berryFunctions)

l2df(L3)

##      V1 V2 V3 V4 V5 V6 V7 V8
## AB    6  9  2  6 NA NA NA NA
## BC    1  2  3  4  5  6  7  8
## CD   -3  2 NA NA NA NA NA NA

# vectors are padded with NAs, names are kept
```

# S-apply V

L3

```
## $AB  
## [1] 6 9 2 6  
##  
## $BC  
## [1] 1 2 3 4 5 6 7 8  
##  
## $CD  
## [1] -3 2
```

*# nested subsetting: get the third element of each vector:*

```
sapply(L3, "[", 3)
```

```
## AB BC CD  
## 2 3 NA
```

## apply functions to vectors & further arguments

```
sample(x=1:100, size=5)
```

```
## [1] 49 39 35 98 74
```

## apply functions to vectors & further arguments

```
sample(x=1:100, size=5)
```

```
## [1] 49 39 35 98 74
```

```
lapply(c(6,2,13), sample, x=1:100)
```

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

```
## [1] 10 35 87 80 88 39
```

```
##
```

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

```
## [1] 76 98
```

```
##
```

```
## [[3]]
```

```
## [1] 16 80 73 9 90 8 64 74 99 45 53 71 77
```

# Practice matrices and lists

## Exercise 11: matrix, list

- ① Create a matrix with some numbers you like, with a useful amount of rows and columns.
- ② Create a list (with meaningful element names!) that contains the previous matrix, your name, the current `Sys.time()`, the number of this exercise, and the haircolors of at least two of your neighbours.
- ③ BONUS 1: Using `sapply`, get the `length` for each element. What is surprising about the result?
- ④ BONUS 2: Make the haircolors element a `factor`. Inspect the result with `str`.

## matrix + list: solutions

```
mat <- matrix(c(42,007,123,7777,-17,88), ncol=2)
mylist <- list(matrix=mat, creator="Berry",
               created=Sys.time(), exerciseNumber=11,
               haircolor=c("brunette", "red", "blonde", "blonde"))
sapply(mylist, length) # matrix length is 6 (=ncol*nrow)
mylist$haircolor <- as.factor(mylist$haircolor)
str(mylist)
```

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps

## 3. Objects

Assignments

Vectors, indexing

Data.frames

Matrices

Lists, lapply/sapply functions

**Classes, methods, functions**

4. "Real" data

5. Plotting

6. Character Operations

7. Conditions & loops

8. Functions & packages

9. Distributions

10. Statistics

11. Time series handling and analysis

12. Spatial data and GIS functionality

13. Final tasks

Course plan

# Determine the class of an object

Everything is an object that belongs to an existing class!!

```
class(v)  
## [1] "integer"  
  
class(df)  
## [1] "data.frame"  
  
class(m)  
## [1] "matrix"
```

# Usage of class in functions

```
x <- 10
class(x) <- c("a", "b")
# error:
if( class(x) == "b" ) message("class of x contains b")

## Warning in if (class(x) == "b") message("class of x
contains b"): the condition has length > 1 and only the
first element will be used

# use this instead
if( inherits(x, "b") ) message("good coding!")

## good coding!
```

# Examining objects

Every object class has attributes and methods:

```
str(v)
```

```
##  int [1:18] 1 1 1 2 2 2 3 3 3 1 ...
```

```
str(df)
```

```
## 'data.frame': 4 obs. of 3 variables:
```

```
## $ col_a      : int 11 12 13 14
```

```
## $ ColName2   : Factor w/ 4 levels "a","b","c","d": 1 2 3 4
```

```
## $ LogicalCol: logi FALSE FALSE TRUE TRUE
```

```
str(m)
```

```
##  chr [1:2, 1:3] "a" "4" "2" "5" "3" ...
```

```
## - attr(*, "dimnames")=List of 2
```

```
##   ..$ : chr [1:2] "row1" "row2"
```

```
##   ..$ : chr [1:3] "A" "B" "C"
```

# Overview of some useful functions

```
length(v)  
## [1] 18  
  
names(df)  
## [1] "col_a"      "ColName2"    "LogicalCol"  
  
dim(m)  
## [1] 2 3  
  
which(v > 4)  
## integer(0)
```

Please note ...

Parentheses and square brackets of functions and objects

Please note ...

## Parentheses and square brackets of functions and objects

objects are variables and contain your data. The indexing (accessing of items) is done by brackets [ ]

# Please note ...

## Parentheses and square brackets of functions and objects

**objects** are variables and contain your data. The indexing (accessing of items) is done by brackets [ ]

**functions** are methods (built-in or self-made) and they perform an analysis or start a process. When functions are called, you have to use parentheses ( ), in which the arguments are put.

## Please note ...

### Parentheses and square brackets of functions and objects

objects are variables and contain your data. The indexing (accessing of items) is done by brackets [ ]

functions are methods (built-in or self-made) and they perform an analysis or start a process. When functions are called, you have to use parentheses ( ), in which the arguments are put.

With `newtable <- edit(olddata)`, you can change cells in a table by hand. **As this is not reproducible, this is bad practice!**. `fix` is even worse, as it changes the object itself.

With `View`, you can look at a table in an external viewer. This is useful for large tables, as the console is not cluttered with data.

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

[adv-r.had.co.nz/Data-structures](http://adv-r.had.co.nz/Data-structures.html).

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

[adv-r.had.co.nz/Data-structures.html](http://adv-r.had.co.nz/Data-structures.html). `as.character(3.14)` converts a data type;

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

[adv-r.had.co.nz/Data-structures](http://adv-r.had.co.nz/Data-structures.html). `as.character(3.14)` converts a data type; `is.integer(4:6)` checks.

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

[adv-r.had.co.nz/Data-structures](http://adv-r.had.co.nz/Data-structures.html). `as.character(3.14)` converts a data type; `is.integer(4:6)` checks. `str` shows an abbreviation of `class`.

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

[adv-r.had.co.nz/Data-structures](http://adv-r.had.co.nz/Data-structures.html). `as.character(3.14)` converts a data type; `is.integer(4:6)` checks. `str` shows an abbreviation of `class`. `mode` (for users) is like `typeof` (R internal), but combines integer and double to numeric (& closure, special and builtin to function).

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

[adv-r.had.co.nz/Data-structures](http://adv-r.had.co.nz/Data-structures.html). `as.character(3.14)` converts a data type; `is.integer(4:6)` checks. `str` shows an abbreviation of `class`. `mode` (for users) is like `typeof` (R internal), but combines integer and double to numeric (& closure, special and builtin to function). Other rare typeofs: raw, environment, promise, ...

# Overview: data types

In order of coercion (if mixed, TRUE is converted to 1, 3.14 to "3.14" etc)

Description	example	typeof	class
empty set	NULL	NULL	NULL
not available	NA	logical	logical
logical	c(T, F, FALSE, TRUE)	logical	logical
category	factor("left")	integer	<b>factor</b>
integer number	4:6	integer	integer
decimal	8.7	double	<b>numeric</b>
complex	5+3i	complex	complex
character string	"homer rocks"	character	character
time	Sys.time()	double	<b>POSIXct</b>
date	as.Date("2017-05-02")	double	<b>Date</b>
function	ncol	closure	<b>function</b>

[adv-r.had.co.nz/Data-structures](http://adv-r.had.co.nz/Data-structures.html). `as.character(3.14)` converts a data type; `is.integer(4:6)` checks. `str` shows an abbreviation of `class`. `mode` (for users) is like `typeof` (R internal), but combines integer and double to numeric (& closure, special and builtin to function). Other rare typeofs: raw, environment, promise, ... When mixing date/time with others, the order of appearance determines the output class.

# Overview: Object types

Object	example	typeof	class
vector	<i>see data types</i>	...	...
matrix	<code>matrix(9:15, ncol=2)</code>	...	matrix
array	<code>array(letters[1:24], dim=c(2,6,4))</code>	...	array
data.frame	<code>data.frame(C1=4:5, C2=c("a","b"))</code>	list	data.frame
list	<code>list(el1=7:15, el2="big")</code>	list	list
function	<code>function(x) 12+0.5*x</code>	closure	function
...	<code>lm(b ~ a)</code>	list	lm

# Overview: Object types

Object	example	typeof	class
vector	<i>see data types</i>	...	...
matrix	<code>matrix(9:15, ncol=2)</code>	...	matrix
array	<code>array(letters[1:24], dim=c(2,6,4))</code>	...	array
data.frame	<code>data.frame(C1=4:5, C2=c("a","b"))</code>	list	data.frame
list	<code>list(el1=7:15, el2="big")</code>	list	list
function	<code>function(x) 12+0.5*x</code>	closure	function
...	<code>lm(b ~ a)</code>	list	lm

A `matrix` consists of only one data type. If you accidentally change one element to a character, all are converted and calculations are not possible any more (See coercion order in previous slide).

# Overview: Object types

Object	example	typeof	class
vector	see <i>data types</i>	...	...
matrix	<code>matrix(9:15, ncol=2)</code>	...	matrix
array	<code>array(letters[1:24], dim=c(2,6,4))</code>	...	array
data.frame	<code>data.frame(C1=4:5, C2=c("a","b"))</code>	list	data.frame
list	<code>list(el1=7:15, el2="big")</code>	list	list
function	<code>function(x) 12+0.5*x</code>	closure	function
...	<code>lm(b ~ a)</code>	list	lm

A `matrix` consists of only one data type. If you accidentally change one element to a character, all are converted and calculations are not possible any more (See coercion order in previous slide).

`data.frames` can have multiple data types, but a column in itself also has only one type.

## Overview: Object types

Object	example	typeof	class
vector	see <i>data types</i>	...	...
matrix	<code>matrix(9:15, ncol=2)</code>	...	matrix
array	<code>array(letters[1:24], dim=c(2,6,4))</code>	...	array
data.frame	<code>data.frame(C1=4:5, C2=c("a","b"))</code>	list	data.frame
list	<code>list(el1=7:15, el2="big")</code>	list	list
function	<code>function(x) 12+0.5*x</code>	closure	function
...	<code>lm(b ~ a)</code>	list	lm

A **matrix** consists of only one data type. If you accidentally change one element to a character, all are converted and calculations are not possible any more (See coercion order in previous slide).

**data.frames** can have multiple data types, but a column in itself also has only one type.

**lists** can combine anything, even other lists.

# Overview: Object types

Object	example	typeof	class
vector	see <i>data types</i>	...	...
matrix	<code>matrix(9:15, ncol=2)</code>	...	matrix
array	<code>array(letters[1:24], dim=c(2,6,4))</code>	...	array
data.frame	<code>data.frame(C1=4:5, C2=c("a","b"))</code>	list	data.frame
list	<code>list(el1=7:15, el2="big")</code>	list	list
function	<code>function(x) 12+0.5*x</code>	closure	function
...	<code>lm(b ~ a)</code>	list	lm

A `matrix` consists of only one data type. If you accidentally change one element to a character, all are converted and calculations are not possible any more (See coercion order in previous slide).

`data.frames` can have multiple data types, but a column in itself also has only one type.

`lists` can combine anything, even other lists.

`is.atomic(object)` returns TRUE (vector, matrix, array) or FALSE

# Overview: Object types

Object	example	typeof	class
vector	see <i>data types</i>	...	...
matrix	<code>matrix(9:15, ncol=2)</code>	...	matrix
array	<code>array(letters[1:24], dim=c(2,6,4))</code>	...	array
data.frame	<code>data.frame(C1=4:5, C2=c("a","b"))</code>	list	data.frame
list	<code>list(el1=7:15, el2="big")</code>	list	list
function	<code>function(x) 12+0.5*x</code>	closure	function
...	<code>lm(b ~ a)</code>	list	lm

A `matrix` consists of only one data type. If you accidentally change one element to a character, all are converted and calculations are not possible any more (See coercion order in previous slide).

`data.frames` can have multiple data types, but a column in itself also has only one type.

`lists` can combine anything, even other lists.

`is.atomic(Object)` returns TRUE (vector, matrix, array) or FALSE

`as.matrix(Object)` converts the class of an object by force.

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
    - Reading and writing data
      - Factors, tapply function
      - Merging data.frames
      - Management of NAs, apply function
      - Reproducible coding
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Data I/O functions (Input / Output)

R **reads and writes** data formats like ASCII files, GEOTIFF, GRIB, NETCDF, HDF ...

# Data I/O functions (Input / Output)

R **reads and writes** data formats like ASCII files, GEOTIFF, GRIB, NETCDF, HDF ...

```
read.csv("relative/path/to/input/file.csv")
# Combined with a variable assignment:
mydata <- read.table("C:/path/to/input/file.txt")
write.table(x=mynewdata, file="output.txt")
```

Windows users: change all \ to \\ or /

# Data I/O functions (Input / Output)

R **reads and writes** data formats like ASCII files, GEOTIFF, GRIB, NETCDF, HDF ...

```
read.csv("relative/path/to/input/file.csv")
# Combined with a variable assignment:
mydata <- read.table("C:/path/to/input/file.txt")
write.table(x=mynewdata, file="output.txt")
```

Windows users: change all \ to \\ or /

Relative file names (minimal changing of working directory):

```
setwd("C:/Users/berry/Desktop") # set working directory
dir() # show files in current folder
mydata <- read.table("filename.txt", header=TRUE, sep="\t")
# use a better object name!
```

# Folder management

```
setwd(..) # sets working directory path up one level
getwd() # shows the current wd path
dir(recursive=TRUE) # also list items in subfolders
dir("../other_subfolder") # list files in different folder
file.create() # some file operations
file.rename()
file.remove() # see also unlink()
file.copy()
dir.create() # some folder (directory) operations
file.exists()
dir.exists()
```

## Some of the arguments for `read.table`

`header = TRUE`

read first line as column names

## Some of the arguments for `read.table`

`header = TRUE`

read first line as column names

`dec = ","`

comma as decimal mark

## Some of the arguments for `read.table`

`header = TRUE`

read first line as column names

`dec = ","`

comma as decimal mark

`sep = "_"`

underscore as column separator ("\t" for tabstop)

## Some of the arguments for `read.table`

`header = TRUE`

read first line as column names

`dec = ","`

comma as decimal mark

`sep = "_"`

underscore as column separator ("\t" for tabstop)

`fill = T`

fill incomplete rows with NAs at the end

## Some of the arguments for `read.table`

`header = TRUE`

read first line as column names

`dec = ","`

comma as decimal mark

`sep = "_"`

underscore as column separator ("\t" for tabstop)

`fill = T`

fill incomplete rows with NAs at the end

`skip = 12`

ignore the first 12 lines (eg with meta data)

## Some of the arguments for `read.table`

`header = TRUE`

read first line as column names

`dec = ","`

comma as decimal mark

`sep = "_"`

underscore as column separator ("\t" for tabstop)

`fill = T`

fill incomplete rows with NAs at the end

`skip = 12`

ignore the first 12 lines (eg with meta data)

`comment.char = "%"`

omit (rest of) lines that start with % (like R's #)

## Some of the arguments for `read.table`

<code>header = TRUE</code>	read first line as column names
<code>dec = ","</code>	comma as decimal mark
<code>sep = "_"</code>	underscore as column separator ("\t" for tabstop)
<code>fill = T</code>	fill incomplete rows with NAs at the end
<code>skip = 12</code>	ignore the first 12 lines (eg with meta data)
<code>comment.char = "%"</code>	omit (rest of) lines that start with % (like R's #)
<code>na.strings = c(-999, "NN")</code>	identify NA entries (missing values)

## Some of the arguments for `read.table`

`header = TRUE`

`dec = ","`

`sep = "_"`

`fill = T`

`skip = 12`

`comment.char = "%"`

`na.strings = c(-999, "NN")`

`stringsAsFactors=FALSE`

read first line as column names

comma as decimal mark

underscore as column separator ("\t" for tabstop)

fill incomplete rows with NAs at the end

ignore the first 12 lines (eg with meta data)

omit (rest of) lines that start with % (like R's #)

identify NA entries (missing values)

do not convert characters to factors

## Some of the arguments for `read.table`

`header = TRUE`

`dec = ","`

`sep = "_"`

`fill = T`

`skip = 12`

`comment.char = "%"`

`na.strings = c(-999, "NN")`

`stringsAsFactors=FALSE`

`as.is=TRUE`

read first line as column names

comma as decimal mark

underscore as column separator ("\t" for tabstop)

fill incomplete rows with NAs at the end

ignore the first 12 lines (eg with meta data)

omit (rest of) lines that start with % (like R's #)

identify NA entries (missing values)

do not convert characters to factors

the same, but less typing, and potentially columnwise

# Some of the arguments for `read.table`

<code>header = TRUE</code>	read first line as column names
<code>dec = ","</code>	comma as decimal mark
<code>sep = "_"</code>	underscore as column separator ("\t" for tabstop)
<code>fill = T</code>	fill incomplete rows with NAs at the end
<code>skip = 12</code>	ignore the first 12 lines (eg with meta data)
<code>comment.char = "%"</code>	omit (rest of) lines that start with % (like R's #)
<code>na.strings = c(-999, "NN")</code>	identify NA entries (missing values)
<code>stringsAsFactors=FALSE</code>	do not convert characters to factors
<code>as.is=TRUE</code>	the same, but less typing, and potentially columnwise

Alternatives to `read.table`:

# Some of the arguments for `read.table`

<code>header = TRUE</code>	read first line as column names
<code>dec = ","</code>	comma as decimal mark
<code>sep = "_"</code>	underscore as column separator ("\t" for tabstop)
<code>fill = T</code>	fill incomplete rows with NAs at the end
<code>skip = 12</code>	ignore the first 12 lines (eg with meta data)
<code>comment.char = "%"</code>	omit (rest of) lines that start with % (like R's #)
<code>na.strings = c(-999, "NN")</code>	identify NA entries (missing values)
<code>stringsAsFactors=FALSE</code>	do not convert characters to factors
<code>as.is=TRUE</code>	the same, but less typing, and potentially columnwise

Alternatives to `read.table`:

`scan` At the core of `read.table` - for complicated things

# Some of the arguments for `read.table`

<code>header = TRUE</code>	read first line as column names
<code>dec = ","</code>	comma as decimal mark
<code>sep = "_"</code>	underscore as column separator ("\t" for tabstop)
<code>fill = T</code>	fill incomplete rows with NAs at the end
<code>skip = 12</code>	ignore the first 12 lines (eg with meta data)
<code>comment.char = "%"</code>	omit (rest of) lines that start with % (like R's #)
<code>na.strings = c(-999, "NN")</code>	identify NA entries (missing values)
<code>stringsAsFactors=FALSE</code>	do not convert characters to factors
<code>as.is=TRUE</code>	the same, but less typing, and potentially columnwise

Alternatives to `read.table`:

`scan` At the core of `read.table` - for complicated things

`read.csv` comma separated values (different defaults than `read.table`)

# Some of the arguments for `read.table`

<code>header = TRUE</code>	read first line as column names
<code>dec = ","</code>	comma as decimal mark
<code>sep = "_"</code>	underscore as column separator ("\t" for tabstop)
<code>fill = T</code>	fill incomplete rows with NAs at the end
<code>skip = 12</code>	ignore the first 12 lines (eg with meta data)
<code>comment.char = "%"</code>	omit (rest of) lines that start with % (like R's #)
<code>na.strings = c(-999, "NN")</code>	identify NA entries (missing values)
<code>stringsAsFactors=FALSE</code>	do not convert characters to factors
<code>as.is=TRUE</code>	the same, but less typing, and potentially columnwise

Alternatives to `read.table`:

`scan` At the core of `read.table` - for complicated things

`read.csv` comma separated values (different defaults than `read.table`)

`read.fwf` fixed width formatted data

# Some of the arguments for `read.table`

<code>header = TRUE</code>	read first line as column names
<code>dec = ","</code>	comma as decimal mark
<code>sep = "_"</code>	underscore as column separator ("\t" for tabstop)
<code>fill = T</code>	fill incomplete rows with NAs at the end
<code>skip = 12</code>	ignore the first 12 lines (eg with meta data)
<code>comment.char = "%"</code>	omit (rest of) lines that start with % (like R's #)
<code>na.strings = c(-999, "NN")</code>	identify NA entries (missing values)
<code>stringsAsFactors=FALSE</code>	do not convert characters to factors
<code>as.is=TRUE</code>	the same, but less typing, and potentially columnwise

Alternatives to `read.table`:

`scan` At the core of `read.table` - for complicated things

`read.csv` comma separated values (different defaults than `read.table`)

`read.fwf` fixed width formatted data

`read_excel` Excel files (install package, see [github.com/hadley/readxl](https://github.com/hadley/readxl))

# Practice reading tables and subsetting data.frames

## Exercise 12: data.frames - working with tables

- ① Read the file `animals.txt` (*rightclick Raw, save as*) with the correct settings for each needed argument and assign it to an object.
- ② Check whether everything is read correctly with `str`.
- ③ Print only the first column (BONUS: in 3 different ways of subsetting).
- ④ `which` element of this vector is equal to `(==)` "Sparrow"?
- ⑤ Can the elephant fly? Find out with a logical query.
- ⑥ Find out how many legs the mammals have on average. Interpret the result.
- ⑦ BONUS 1: Use `tapply` to find the average cuteness score for each animal class.
- ⑧ BONUS 2: Read the file "`days.txt`" in a way that days, months, and years each have their own column. What other function besides `read.table` might be useful for that?

## Solution for exercise 12.1-2: data.frames

```
# set working directory to path with files:  
setwd("C:/Dokumente und Einstellungen/hydro/Desktop")  
# which files are there:  
dir()  
# step by step, specify the arguments of read.table...  
animals <- read.table("data/animals.txt", sep="\t",  
                      header=TRUE, dec=",")  
animals  
str(animals) # ... until all numbers are num.
```

## Solution for exercise 12.3-5: data.frames

```
# first column, and which is Sparrow:  
animals$name  
animals[, "Name"] # more typing  
animals[, 1] # hard to interpret later on  
animals$name == "Sparrow"  
which( animals$name == "Sparrow" )  
  
# Can the elephant fly? Automatic query:  
animals[4,3]  
animals[ animals$name == "Elephant" , 3 ]  
animals[ animals$name == "Elephant" , "can.fly" ]  
animals[ animals$name == "Sparrow" , "can.fly" ]
```

## Solution for exercise 12.6-7: data.frames

```
# How many legs do the mammals have on average?  
animals[ animals$Class == "Mammal" , "n.legs" ]  
mean( animals[ animals$Class == "Mammal" , "n.legs" ] )  
# How cute are they, on average?  
mean( animals[ animals$Class == "Mammal" , "cuteness" ] )  
  
# How cute is each category of animals?  
tapply(X=animals$cuteness, INDEX=animals$Class, FUN=mean)  
# separates cuteness into groups, categorized by Class,  
# then calculates mean for each group
```

## Solution for exercise 12.8: data.frames

```
# Read "days.txt" with columns for days, months and years
dir()
days <- read.table("days.txt", sep=".")
head(days)
str(days)

# custom names of columns
colnames(days) <- c("day", "month", "year")
head(days)

# other way:
read.fwf("data/days.txt", widths=c(2,1,2,1,4))

# usage example: return months for e.g. vegetation index:
days$month
```

## Frequent mistakes with `read.table`

Error in `scan(file, what, nmax, sep, dec, quote, skip, nlines, na.strings, : line _ did not have _ elements`

usually `header`, `sep` or `fill` is given incorrectly.

## Frequent mistakes with `read.table`

<p>Error in <code>scan(file, what, nmax, sep, dec, quote, skip, nlines, na.strings, : line _ did not have _ elements</code></p>	usually <code>header</code> , <code>sep</code> or <code>fill</code> is given incorrectly.
<p>Warning message: <code>In read.table("_.txt", _) : incomplete final line found by readTableHeader on 'C:\_.txt'</code></p>	Open the file and at the end, add a line break (with ENTER) = line feed = newline = carriage return.

## Frequent mistakes with `read.table`

<p>Error in <code>scan(file, what, nmax, sep, dec, quote, skip, nlines, na.strings, : line _ did not have _ elements</code></p>	usually <code>header</code> , <code>sep</code> or <code>fill</code> is given incorrectly.
<p>Warning message: <code>In read.table("_.txt", _) : incomplete final line found by readTableHeader on 'C:\_.txt'</code></p>	Open the file and at the end, add a line break (with ENTER) = line feed = newline = carriage return.
<p><code>str(DataframeIreadIn)</code> yields "factor" in some numeric columns</p>	Check data file! <code>dec</code> may need to be corrected

## Frequent mistakes with `read.table`

Error in <code>scan(file, what, nmax, sep, dec, quote, skip, nlines, na.strings, : line _ did not have _ elements</code>	usually <code>header</code> , <code>sep</code> or <code>fill</code> is given incorrectly.
Warning message: <code>In read.table("_.txt", _) :</code> incomplete final line found by <code>readTableHeader</code> on 'C:\_.txt'	Open the file and at the end, add a line break (with ENTER) = line feed = newline = carriage return.
<code>str(DataframeIreadIn)</code> yields "factor" in some numeric columns	Check data file! <code>dec</code> may need to be corrected

```
as.numeric( as.character(df$col) )
as.numeric( gsub(pattern=",", replacement=". ", x=df$col) )
# If you try to read data in the format "3,590.18", run
df$col <- as.numeric(gsub(", ", "", df$col))
```

## Frequent mistakes with `read.table`

Error in <code>scan(file, what, nmax, sep, dec, quote, skip, nlines, na.strings, : line _ did not have _ elements</code>	usually <code>header</code> , <code>sep</code> or <code>fill</code> is given incorrectly.
Warning message: <code>In read.table("_.txt", ...)</code> : incomplete final line found by <code>readTableHeader</code> on 'C:\_.txt'	Open the file and at the end, add a line break (with ENTER) = line feed = newline = carriage return.
<code>str(DataframeIreadIn)</code> yields "factor" in some numeric columns	Check data file! <code>dec</code> may need to be corrected

```
as.numeric( as.character(df$col) )
as.numeric( gsub(pattern=",", replacement=". ", x=df$col) )
# If you try to read data in the format "3,590.18", run
df$col <- as.numeric(gsub(", ", "", df$col))
```

Note: **errors** stop function execution, **warnings** can influence function performance and should only be ignored if you know what they really mean.

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
    - Reading and writing data
    - Factors, tapply function
    - Merging data.frames
    - Management of NAs, apply function
    - Reproducible coding
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Factors = categorical variables

```
iris$Species[63] # how factors are printed  
  
## [1] versicolor  
## Levels: setosa versicolor virginica  
  
class(iris$Species)  
  
## [1] "factor"  
  
levels(iris$Species)  
  
## [1] "setosa"      "versicolor"   "virginica"  
  
# R saves factors as numbers internally  
as.numeric(iris$Species)[c(1,63,105)]  
  
## [1] 1 2 3
```

factors, explained with [animals.txt](#): the danger

## factors, explained with `animals.txt`: the danger

```
ac <- read.table(file="data/animals.txt", sep="\t", header=T)$cuteness
ac # R calls print.factor , see methods(print)

## [1] 1,3 6,5 9,8 6,3 5,8 4,5
## Levels: 1,3 4,5 5,8 6,3 6,5 9,8
```

## factors, explained with `animals.txt`: the danger

```
ac <- read.table(file="data/animals.txt", sep="\t", header=T)$cuteness
ac # R calls print.factor , see methods(print)

## [1] 1,3 6,5 9,8 6,3 5,8 4,5
## Levels: 1,3 4,5 5,8 6,3 6,5 9,8

class(ac); typeof(ac)

## [1] "factor"
## [1] "integer"
```

## factors, explained with `animals.txt`: the danger

```
ac <- read.table(file="data/animals.txt", sep="\t", header=T)$cuteness
ac # R calls print.factor , see methods(print)

## [1] 1,3 6,5 9,8 6,3 5,8 4,5
## Levels: 1,3 4,5 5,8 6,3 6,5 9,8

class(ac); typeof(ac)

## [1] "factor"
## [1] "integer"

as.numeric(ac) # don't ever do this!!

## [1] 1 5 6 4 3 2
```

## factors, explained with `animals.txt`: the danger

```
ac <- read.table(file="data/animals.txt", sep="\t", header=T)$cuteness
ac # R calls print.factor , see methods(print)

## [1] 1,3 6,5 9,8 6,3 5,8 4,5
## Levels: 1,3 4,5 5,8 6,3 6,5 9,8

class(ac); typeof(ac)

## [1] "factor"
## [1] "integer"

as.numeric(ac) # don't ever do this!

## [1] 1 5 6 4 3 2

char <- as.character(ac)
as.numeric(sub(",",".",char))

## [1] 1.3 6.5 9.8 6.3 5.8 4.5
```

factors, explained with [animals.txt](#): the solution

## factors, explained with `animals.txt`: the solution

```
df <- read.table(file="data/animals.txt", sep="\t", header=T, as.is=TRUE)
str(df)

## 'data.frame': 6 obs. of 6 variables:
## $ Name      : chr "Frog" "Merl" "Doe" "Elephant" ...
## $ Class     : chr "Amphibian" "Bird" "Mammal" "Mammal" ...
## $ can.fly   : logi FALSE TRUE FALSE FALSE TRUE FALSE
## $ color     : chr "green" "black" "brown" "gray" ...
## $ n.legs    : int 4 2 4 4 4 2
## $ cuteness: chr "1,3" "6,5" "9,8" "6,3" ...
```

Now Name and cuteness are character strings instead of factors.

`as.is=TRUE` is a special case of `stringsAsFactors=FALSE`: with the first you can specify the latter per column (I use it because I can type it more quickly).

factors are very useful to apply a function per group

factors are very useful to apply a function per group

```
str(PlantGrowth)

## 'data.frame': 30 obs. of  2 variables:
## $ weight: num  4.17 5.58 5.18 6.11 4.5 4.61 5.17 4.53 5.33 5.14 ...
## $ group : Factor w/ 3 levels "ctrl","trt1",...: 1 1 1 1 1 1 1 1 1 1 ...
```

factors are very useful to apply a function per group

```
str(PlantGrowth)

## 'data.frame': 30 obs. of  2 variables:
## $ weight: num  4.17 5.58 5.18 6.11 4.5 4.61 5.17 4.53 5.33 5.14 ...
## $ group : Factor w/ 3 levels "ctrl","trt1",...: 1 1 1 1 1 1 1 1 1 1 ...
```

```
tapply(X=PlantGrowth$weight, INDEX=PlantGrowth$group, FUN=mean)

##   ctrl  trt1  trt2
## 5.032 4.661 5.526
```

factors are very useful to apply a function per group

```
str(PlantGrowth)

## 'data.frame': 30 obs. of  2 variables:
## $ weight: num  4.17 5.58 5.18 6.11 4.5 4.61 5.17 4.53 5.33 5.14 ...
## $ group : Factor w/ 3 levels "ctrl","trt1",...: 1 1 1 1 1 1 1 1 1 1 ...
```

```
tapply(X=PlantGrowth$weight, INDEX=PlantGrowth$group, FUN=mean)

##   ctrl   trt1   trt2
## 5.032 4.661 5.526
```

This part of Hadley's book is a good intro to factors.

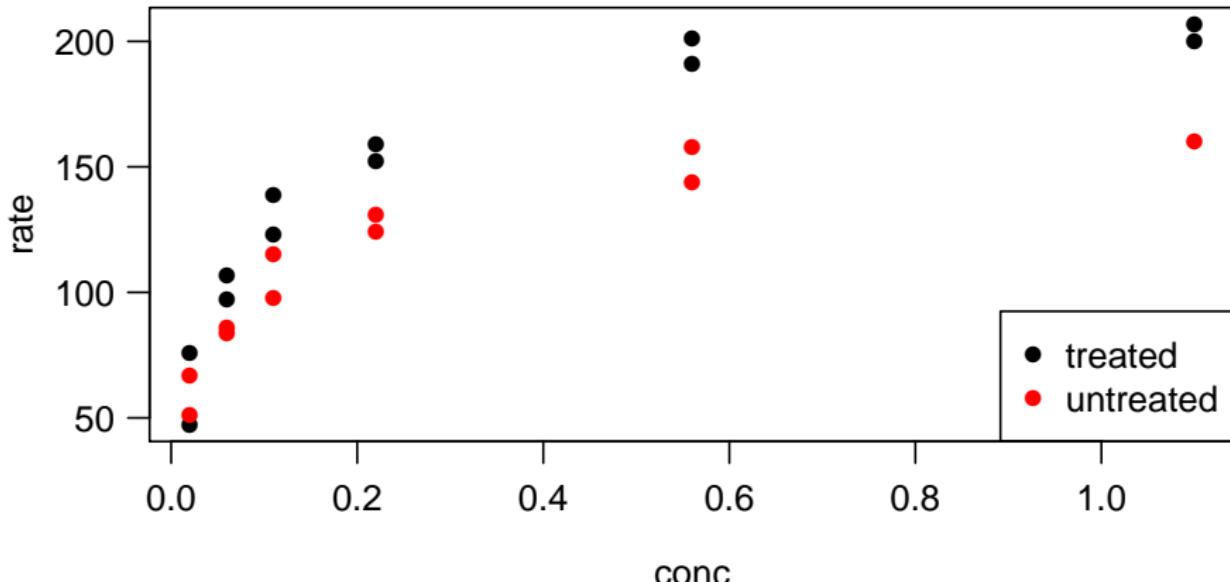
factors are also useful for plotting specifics

col can be a factor vector, integers correspond to `palette()`

factors are also useful for plotting specifics

col can be a factor vector, integers correspond to palette()

```
plot(rate~conc, data=Puromycin, col=state, pch=16, las=1)
legend("bottomright", levels(Puromycin$state), col=1:2, pch=16)
```



# practice working with categories

## Exercise 13: Factors, tapply

- ① Plot the Puromycin dataset, but use the 3rd and 4th color of the default `palette()`
- ② BONUS 1: Use custom colors.  
Hint: create a vector with the colors, then index from it.
- ③ BONUS 2: In the `iris` dataset, add a row for the *Iris sintenisii*.  
Hint: add a level to the factor first.
- ④ Using `tapply`, find out what you should feed chickens according to the measurements in `chickwts`.

## Solution for exercise 13: factors, tapply

```
plot(rate~conc, data=Puromycin, col=as.integer(state)+2, pch=16)

mycol <- c("orange", "mediumpurple2")
plot(rate~conc, data=Puromycin, col=mycol[state], pch=16)

levels(iris$Species) <- c(levels(iris$Species), "sintensii" )
iris[151, ] <- c(6,3,5,2,"sintensii")
```

## Solution for exercise 13: factors, tapply

```
plot(rate~conc, data=Puromycin, col=as.integer(state)+2, pch=16)

mycol <- c("orange", "mediumpurple2")
plot(rate~conc, data=Puromycin, col=mycol[state], pch=16)

levels(iris$Species) <- c(levels(iris$Species), "sintensii" )
iris[151, ] <- c(6,3,5,2,"sintensii")
```

```
sort(tapply(chickwts$weight, chickwts$feed, median))
```

```
## horsebean    linseed    soybean   meatmeal sunflower    casein
##      151.5     221.0     248.0     263.0     328.0     342.0
```

```
sort(tapply(chickwts$weight, chickwts$feed, min))
```

```
## horsebean    linseed   meatmeal    soybean    casein sunflower
##      108        141       153        158        216      226
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. **"Real" data**
    - Reading and writing data
    - Factors, tapply function
    - Merging data.frames**
    - Management of NAs, apply function
    - Reproducible coding
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Merge (join) tables

If you haven't already, now is a good time to download (and unzip!) all datasets.

## Exercise 14: Merge

- ① Read the files `mergeApples1.txt` (*rightclick Raw, save as*) and `mergeApples2.txt` into separate data.frames.
- ② Use `rbind` to combine both tables line-wise and assign the output to e.g. `apples`.
- ③ Read `mergeAge.txt` into e.g. `age`.
- ④ Describe what `merge(age, apples, by.x="Name", by.y="Person")` does.
- ⑤ What happens if you set `all` or `all.x = TRUE`?
- ⑥ BONUS 1: What is the difference between `cbind` and `rbind`?
- ⑦ BONUS 2: Why does `cbind` not work with `age` and `apples`?

## Solution for exercise 14: Merge

```
ma1 <- read.table("data/mergeApples1.txt", header=T)
ma2 <- read.table("data/mergeApples2.txt", header=T)
apples <- rbind(ma1, ma2)
apples
age <- read.table("data/mergeAge.txt", header=T)

merge(age, apples, by.x="Name", by.y="Person")
# combine the information from both datasets
# for names that occur in both tables

merge(age, apples, by.x="Name", by.y="Person", all=T)
# Keep all entries, pad with NAs

# rbind: combine rows below each other
# cbind: bind columns beside each other
# dimensions and names must be correct
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  - 4. "Real" data**
    - Reading and writing data
    - Factors, tapply function
    - Merging data.frames
    - Management of NAs, apply function**
    - Reproducible coding
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# NA-values (NA=not available)

Creating a table with **NA** values

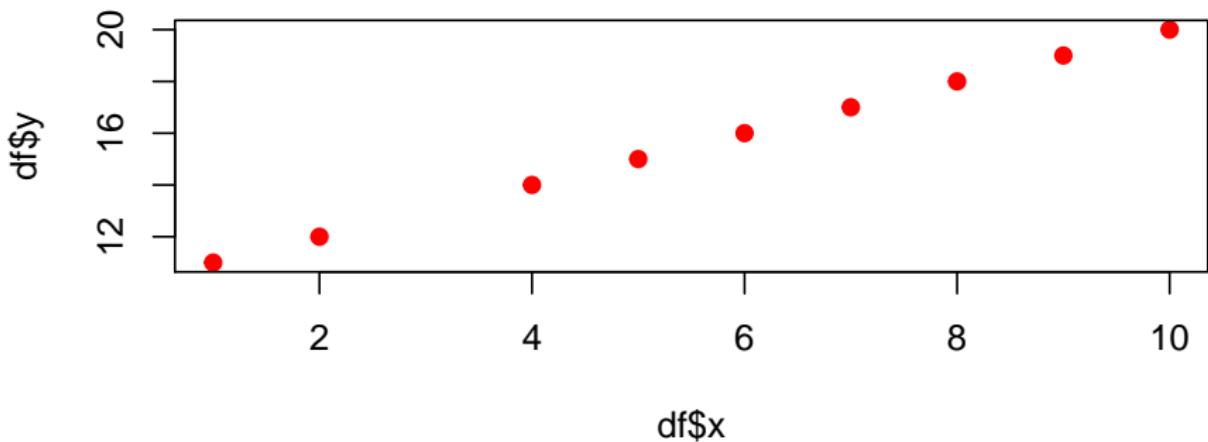
```
# create some test data. "c" around ":" is not necessary
df <- data.frame(x= c(1:10), y=11:20 )
df[3,2] <- NA
head(df)

##   x   y
## 1 1 11
## 2 2 12
## 3 3 NA
## 4 4 14
## 5 5 15
## 6 6 16
```

# NA-values

## Handling of **NA** values

```
plot(df$x, df$y, pch=19, col="red")
```



# NA-values

## Checking for **NA** values

```
is.na(df)

##           x     y
## [1,] FALSE FALSE
## [2,] FALSE FALSE
## [3,] FALSE  TRUE
## [4,] FALSE FALSE
## [5,] FALSE FALSE
## [6,] FALSE FALSE
## [7,] FALSE FALSE
## [8,] FALSE FALSE
## [9,] FALSE FALSE
## [10,] FALSE FALSE
```

# NA-values

## Checking for **NA** values

```
is.na(df)
```

```
##           x     y
## [1,] FALSE FALSE
## [2,] FALSE FALSE
## [3,] FALSE  TRUE
## [4,] FALSE FALSE
## [5,] FALSE FALSE
## [6,] FALSE FALSE
## [7,] FALSE FALSE
## [8,] FALSE FALSE
## [9,] FALSE FALSE
## [10,] FALSE FALSE
```

```
na.omit(df)
```

```
##           x     y
## 1        1    11
## 2        2    12
## 4        4    14
## 5        5    15
## 6        6    16
## 7        7    17
## 8        8    18
## 9        9    19
## 10      10   20
```

# NA-values

```
##      x  y
## 1    1 11
## 2    2 12
## 3    3 NA
## 4    4 14
## 5    5 15
## 6    6 16
## 7    7 17
## 8    8 18
## 9    9 19
## 10   10 20
```

# NA-values

```
##      x  y
## 1    1 11
## 2    2 12
## 3    3 NA
## 4    4 14
## 5    5 15
## 6    6 16
## 7    7 17
## 8    8 18
## 9    9 19
## 10   10 20
```

```
mean(df$x) ; mean(df$y)
```

```
## [1] 5.5
## [1] NA
```

# NA-values

```
##      x  y
## 1    1 11
## 2    2 12
## 3    3 NA
## 4    4 14
## 5    5 15
## 6    6 16
## 7    7 17
## 8    8 18
## 9    9 19
## 10   10 20
```

```
mean(df$x) ; mean(df$y)
```

```
## [1] 5.5
## [1] NA
```

```
mean(df$y, na.rm=TRUE) # dangerous with sums
```

```
## [1] 15.77778
```

apply: execute a function on each row (col) of a matrix

apply: execute a function on each row (col) of a matrix

```
# Compute the column- and row sums of a data.frame (df)
apply(X=df, MARGIN=2, FUN=sum) # column wise

##   x   y
## 55 NA
```

apply: execute a function on each row (col) of a matrix

```
# Compute the column- and row sums of a data.frame (df)
apply(X=df, MARGIN=2, FUN=sum) # column wise
```

```
##   x   y
## 55 NA
```

```
apply(df, 1, sum) # line by line
```

```
## [1] 12 14 NA 18 20 22 24 26 28 30
```

## apply: execute a function on each row (col) of a matrix

```
# Compute the column- and row sums of a data.frame (df)
apply(X=df, MARGIN=2, FUN=sum) # column wise
```

```
##   x   y
## 55 NA
```

```
apply(df, 1, sum) # line by line
```

```
## [1] 12 14 NA 18 20 22 24 26 28 30
```

```
apply(df, 2, sum, na.rm=TRUE) # with further arguments
```

```
##   x   y
## 55 142
```

apply: execute a function on each row (col) of a matrix

```
# Compute the column- and row sums of a data.frame (df)
apply(X=df, MARGIN=2, FUN=sum) # column wise
```

```
##   x   y
## 55 NA
```

```
apply(df, 1, sum) # line by line
```

```
## [1] 12 14 NA 18 20 22 24 26 28 30
```

```
apply(df, 2, sum, na.rm=TRUE) # with further arguments
```

```
##   x   y
## 55 142
```

## Exercise 15: apply

Compute the median of each column in the dataset `iris`.

Hint: leave out the non-numeric column.

## apply practice solution

```
apply(iris[ , -5], MARGIN=2, median)  
## Sepal.Length  Sepal.Width Petal.Length  Petal.Width  
##           5.80        3.00        4.35        1.30
```

## apply practice solution

```
apply(iris[ , -5], MARGIN=2, median)
```

```
## Sepal.Length  Sepal.Width Petal.Length  Petal.Width  
##           5.80          3.00          4.35          1.30
```

```
apply(iris[ , -5], MARGIN=2, mean)
```

```
## Sepal.Length  Sepal.Width Petal.Length  Petal.Width  
##      5.843333  3.057333  3.758000  1.199333
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  - 4. "Real" data**
    - Reading and writing data
    - Factors, tapply function
    - Merging data.frames
    - Management of NAs, apply function
  - Reproducible coding**
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Working with Scripts

For any real data analysis, you should work with scripts that save your code.

# Working with Scripts

For any real data analysis, you should work with scripts that save your code.  
For that reason, you don't have to save your workspace when you close R.

# Working with Scripts

For any real data analysis, you should work with scripts that save your code. For that reason, you don't have to save your workspace when you close R. If you save it, you don't know where it is stored by default. In the next R session, all objects will be available again, which is convenient, but dangerous, as it reduces reproducibility.

# Working with Scripts

For any real data analysis, you should work with scripts that save your code. For that reason, you don't have to save your workspace when you close R. If you save it, you don't know where it is stored by default. In the next R session, all objects will be available again, which is convenient, but dangerous, as it reduces reproducibility. Scripts can easily be exchanged with research partners, clients or employers. They may also develop into packages...

# Working with Scripts

For any real data analysis, you should work with scripts that save your code. For that reason, you don't have to save your workspace when you close R. If you save it, you don't know where it is stored by default.

In the next R session, all objects will be available again, which is convenient, but dangerous, as it reduces reproducibility.

Scripts can easily be exchanged with research partners, clients or employers. They may also develop into packages...

Before sending, close R, open a clean (!) session, and check whether everything is running.

# Working with Scripts

For any real data analysis, you should work with scripts that save your code.  
For that reason, you don't have to save your workspace when you close R.  
If you save it, you don't know where it is stored by default.

In the next R session, all objects will be available again, which is  
convenient, but dangerous, as it reduces reproducibility.

Scripts can easily be exchanged with research partners, clients or employers.  
They may also develop into packages...

Before sending, close R, open a clean (!) session, and check whether  
everything is running.

Scripts should have lots of comments!

# Working with Scripts

For any real data analysis, you should work with scripts that save your code.  
For that reason, you don't have to save your workspace when you close R.  
If you save it, you don't know where it is stored by default.

In the next R session, all objects will be available again, which is  
convenient, but dangerous, as it reduces reproducibility.

Scripts can easily be exchanged with research partners, clients or employers.  
They may also develop into packages...

Before sending, close R, open a clean (!) session, and check whether  
everything is running.

Scripts should have lots of comments! # seriously.

# Creating documents, Organizing your files

Rstudio - file - new -

- R HTML (no  $\text{\LaTeX}$  required on system)
- R markdown (easy to learn, creates .pdf or .docx)
- R sweave (very powerful through  $\text{\LaTeX}$ . This is what I use for these slides.)

# Creating documents, Organizing your files

Rstudio - file - new -

- R HTML (no  $\text{\LaTeX}$  required on system)
- R markdown (easy to learn, creates .pdf or .docx)
- R sweave (very powerful through  $\text{\LaTeX}$ . This is what I use for these slides.)

I use knitr, as it is more advanced than the default sweave. For that, set:

Rstudio - tools - global options - Sweave - weave using: knitr

Save as .Rnw file. See the knitr demo files: [yihui.name/knitr/demo](http://yihui.name/knitr/demo)

# Creating documents, Organizing your files

Rstudio - file - new -

- R HTML (no  $\text{\LaTeX}$  required on system)
- R markdown (easy to learn, creates .pdf or .docx)
- R sweave (very powerful through  $\text{\LaTeX}$ . This is what I use for these slides.)

I use knitr, as it is more advanced than the default sweave. For that, set:

Rstudio - tools - global options - Sweave - weave using: knitr

Save as .Rnw file. See the knitr demo files: [yihui.name/knitr/demo](http://yihui.name/knitr/demo)

There are tips online on how to organize data, scripts, plots etc.

[stackoverflow: How to organize large R programs](#)

[stackexchange: How to efficiently manage a statistical analysis project?](#)

Here are some tips by Hadley: <http://r-pkgs.had.co.nz/style.html>

# Real Datasets online

- NOAA (USA) weather data: [www1.ncdc.noaa.gov/pub/data/normals](http://www1.ncdc.noaa.gov/pub/data/normals)
- DWD Climate Data Center: [CDC/observations\\_germany/climate](https://CDC/observations_germany/climate)

```
library(rdwd) # github.com/brry/rdwd
clim <- dataDWD(selectDWD(id=02080, res="daily", var="kl",
                           per="hist"), dir="DWDdata")
```

- [data.fivethirtyeight.com](http://data.fivethirtyeight.com)
- [data.unicef.org](http://data.unicef.org) (potentially add .xls etc as file extension)
- [github.com/okulbilisim/awesome-datascience#data-sets](https://github.com/okulbilisim/awesome-datascience#data-sets)
- [github.com/caesar0301/awesome-public-datasets](https://github.com/caesar0301/awesome-public-datasets)
- [ropensci.org/packages/index.html](https://ropensci.org/packages/index.html)
- [dataviz.tools/category/data-sources/](https://dataviz.tools/category/data-sources/)
- StorytellingWithData dataviz challenge 2018

## Exercise 16: Real Data

Download, read and (visually) analyze a dataset that interests you.

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
    - Syntax, adjustments, labeling
    - Lines
    - Limits, adding to graphics
    - saving plots, multiple figures
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Plot I: the dataset

```
head(beaver2, 4)

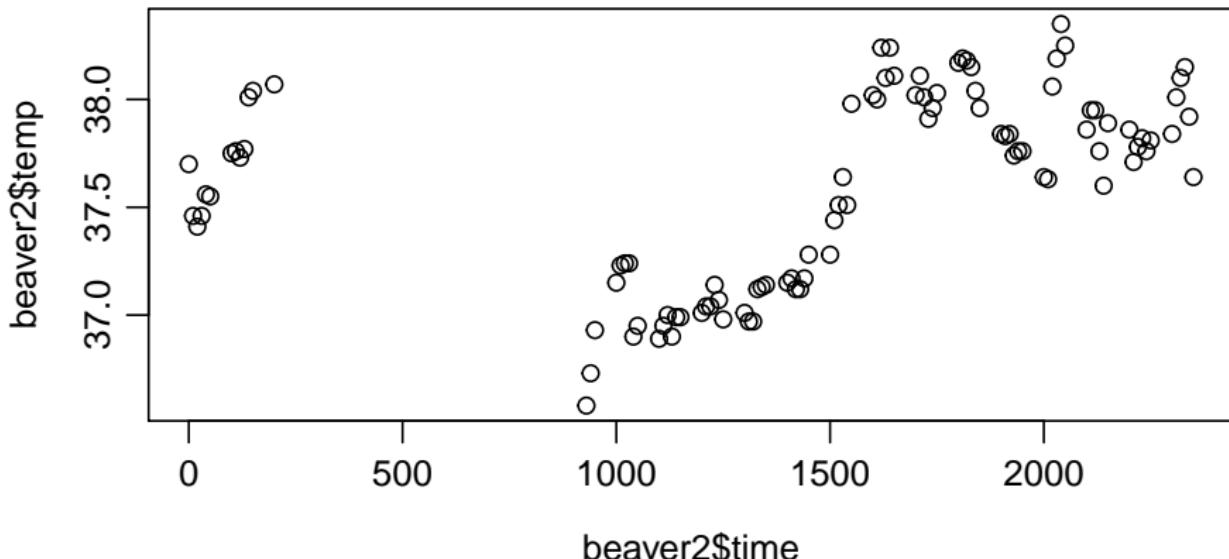
##   day time  temp activ
## 1 307  930 36.58     0
## 2 307  940 36.73     0
## 3 307  950 36.93     0
## 4 307 1000 37.15     0

str(beaver2)

## 'data.frame': 100 obs. of  4 variables:
## $ day  : num  307 307 307 307 307 307 307 307 307 ...
## $ time : num  930 940 950 1000 1010 1020 1030 1040 1050 ...
## $ temp : num  36.6 36.7 36.9 37.1 37.2 ...
## $ activ: num  0 0 0 0 0 0 0 0 0 ...
```

## Plot II

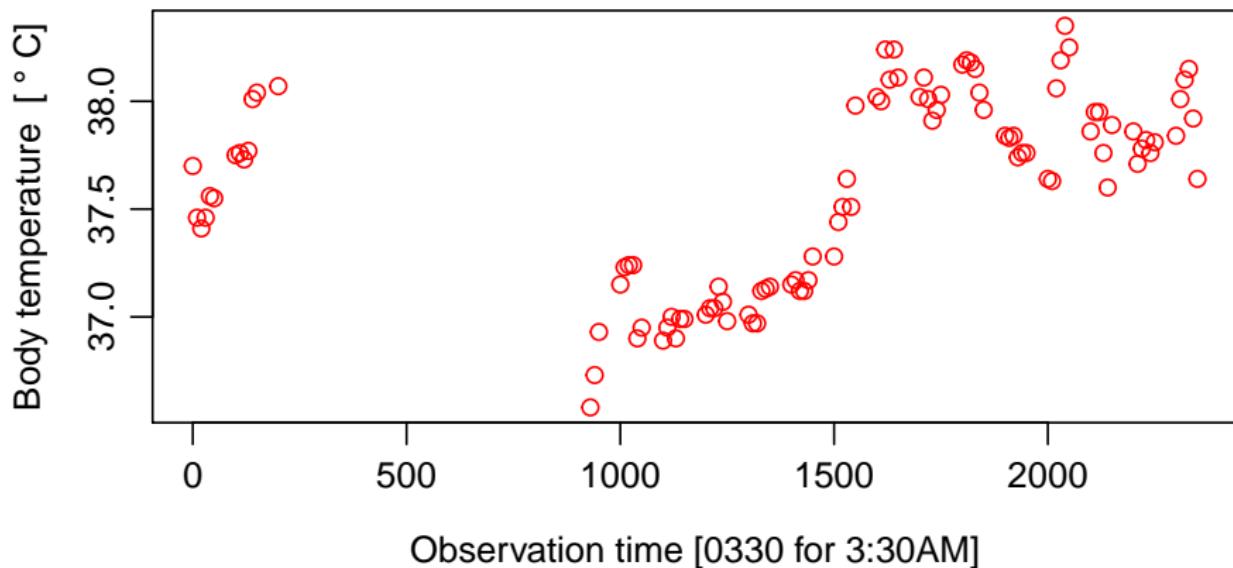
```
# The basic syntax is plot(x, y, adjustments)
plot(beaver2$time, beaver2$temp)
#
```



# Plot III

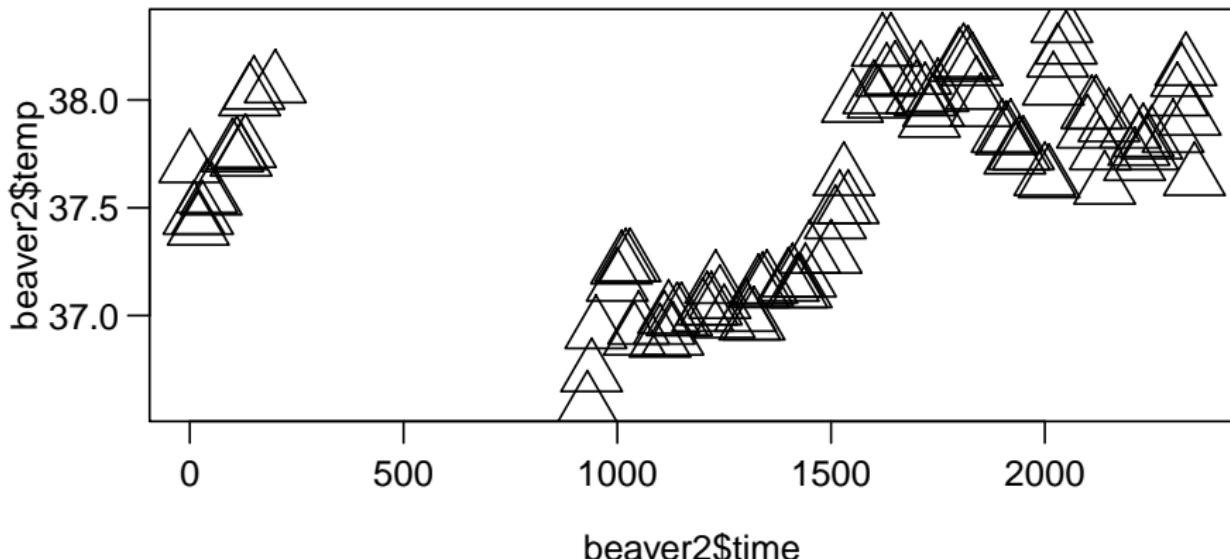
```
plot(beaver2$time, beaver2$temp, ylab="Body temperature [ \u00b0C ]",  
      main="Beavers get warm late in the evening",  
      xlab="Observation time [0330 for 3:30AM]", col="red")
```

**Beavers get warm late in the evening**



## Plot IV

```
plot(beaver2$time, beaver2$temp, pch=2, cex=2.8, las=1)
# pch: PointCharacter, cex: CharacterExpansion
# las: LabelAxisStyle (numbers upright)
```



# Overview of Point Characters (more in bF Anhang)

**plot ( x, y, pch = \_ )**

0	1	2	3	4	5	6	
□	○	△	+	×	◇	▽	
7	8	9	10	11	12	13	14
⊗	*	◇	⊕	⊗	田	⊗	□
15	16	17	18	19	20		
■	●	▲	◆	●	●		
21	22	23	24	25	21:25 fill color with bg		
●	■	◆	▲	▽			

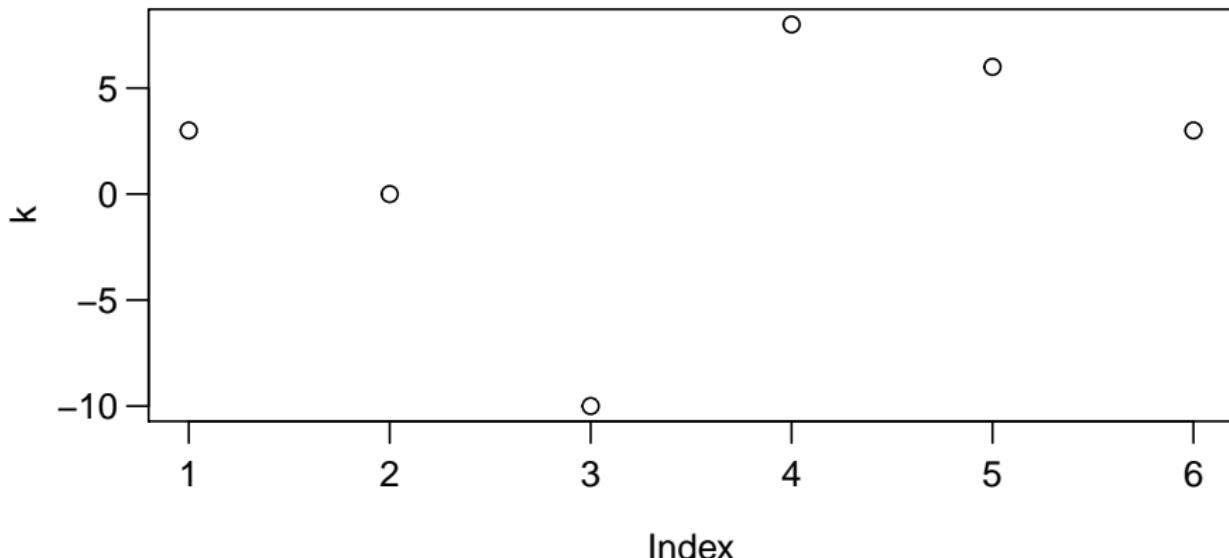
# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
    - Syntax, adjustments, labeling
    - Lines
      - Limits, adding to graphics
      - saving plots, multiple figures
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

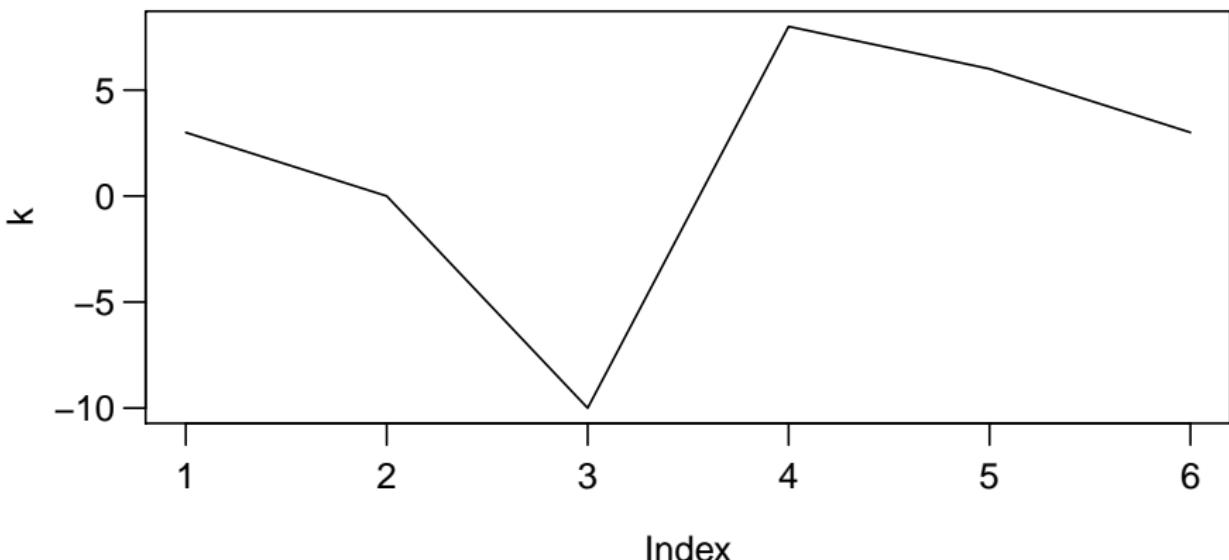
# Plot with lines I

```
k <- c(3,0,-10,8,6,3)      ; l <- c(-4,2,-6,-12,7,-3)
plot(k)  # standard (default) type: "p" for points
```



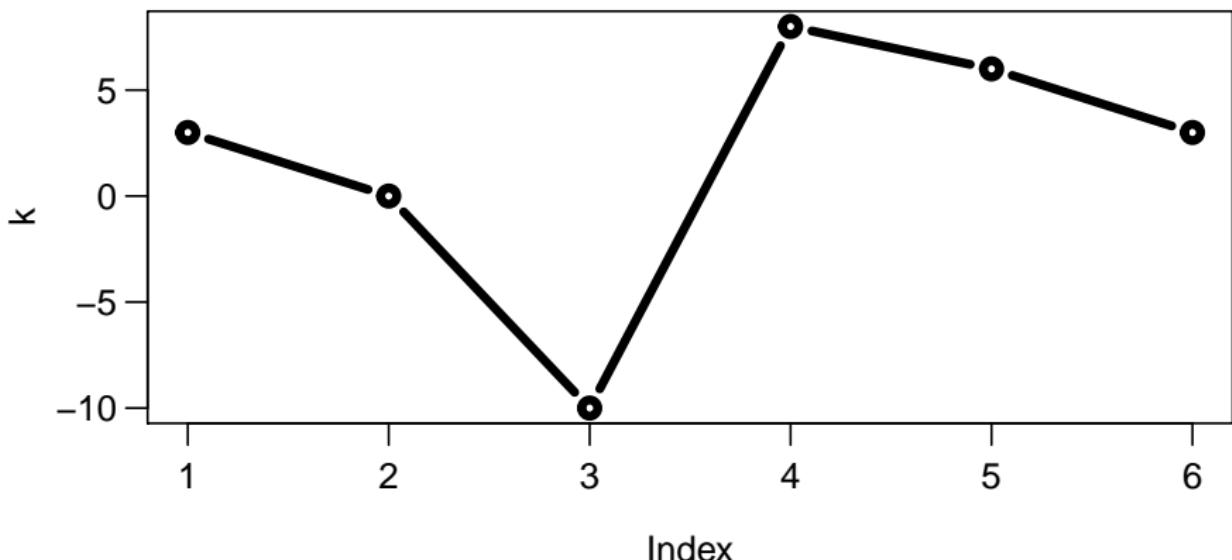
## Plot with lines II

```
plot(k, type="l")
# type="l" for lines
```



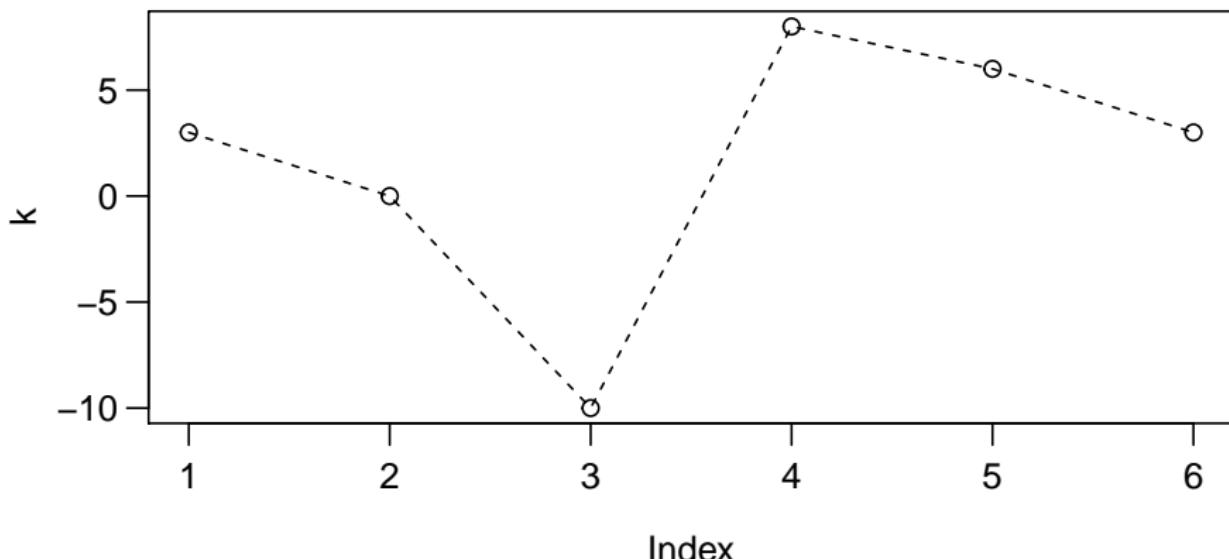
## Plot with lines III

```
plot(k, type="b", lwd=4)
# type="b" for both points and lines. # lwd: line width
```



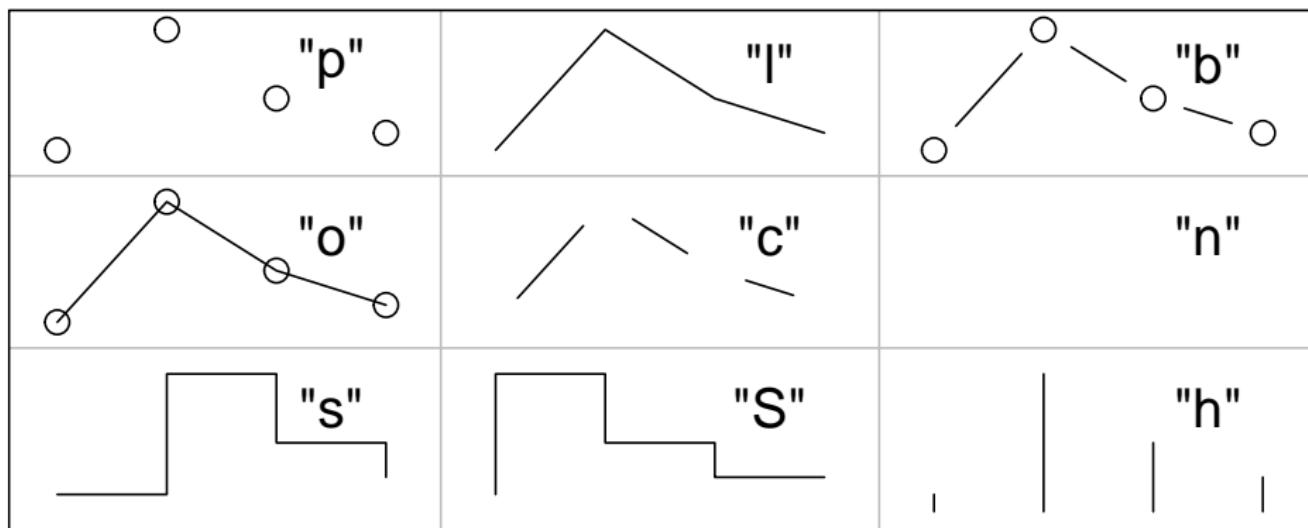
## Plot with lines IV

```
plot(k, type="o", lty=2)
# type="o" for both overplotted. # lty: line type
```



# Overview of types (more in bF Anhang)

**plot ( x, y, type = \_ )**



# Overview of line types (more in bF Anhang)

**plot ( x, y, lty = \_ )**

"blank" (no line) 0

"solid" (default) 1



"dashed" 2



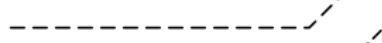
"dotted" 3



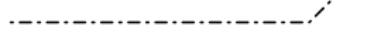
"dotdash" 4



"longdash" 5



"twodash" 6



# Outline

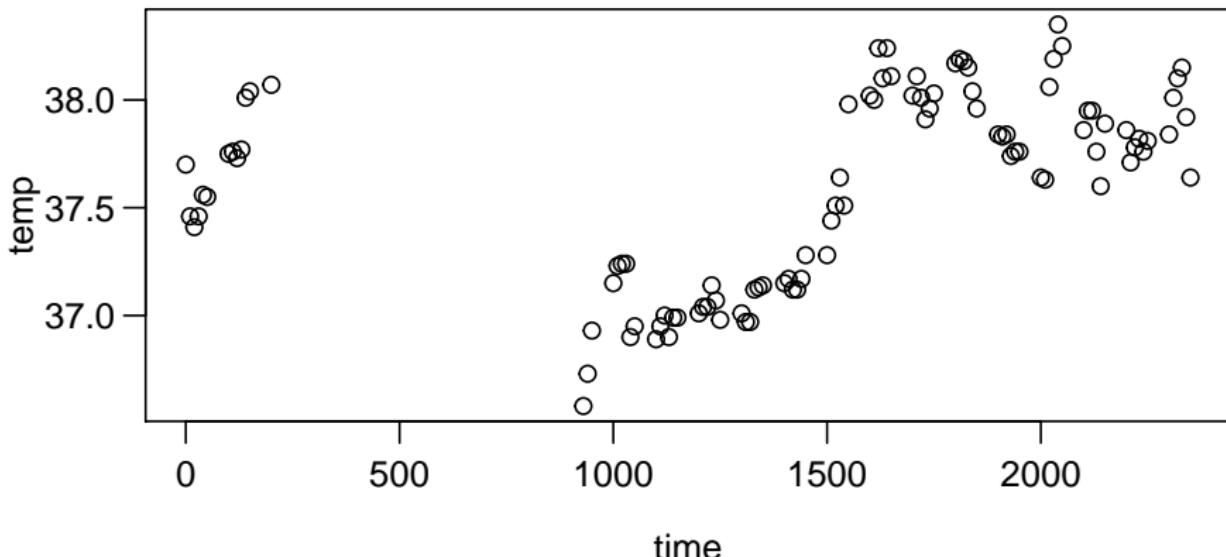
R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
    - Syntax, adjustments, labeling
    - Lines
    - Limits, adding to graphics
    - saving plots, multiple figures
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# One data.frame instead of two vectors

Default method if x is a data.frame:

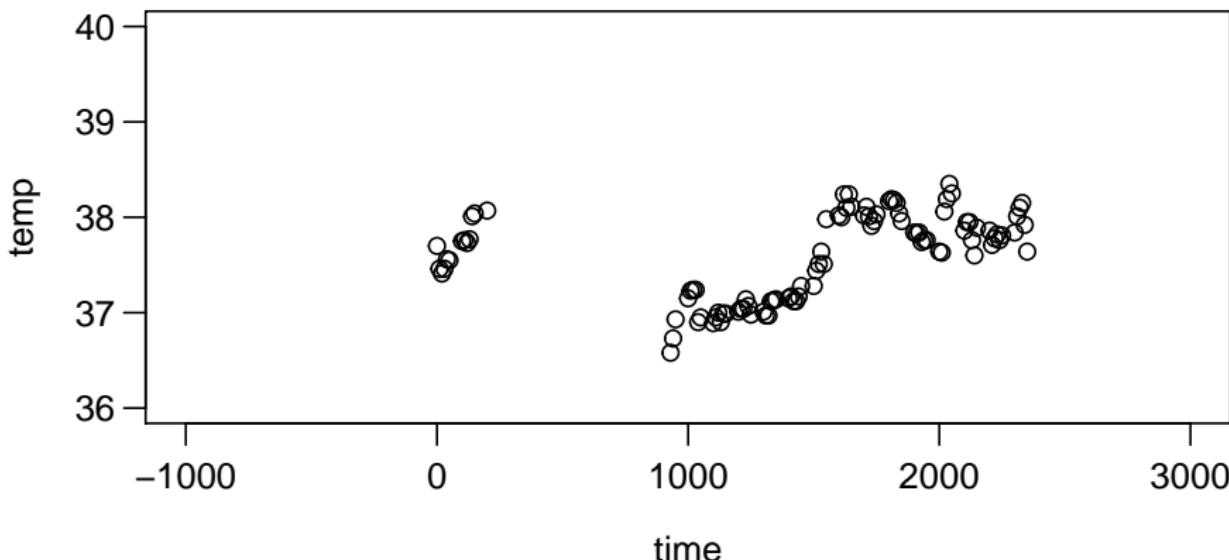
```
plot( beaver2[ ,2:3] )
```



# Plot limits I

xlim with a two-value vector with lower and upper limit.

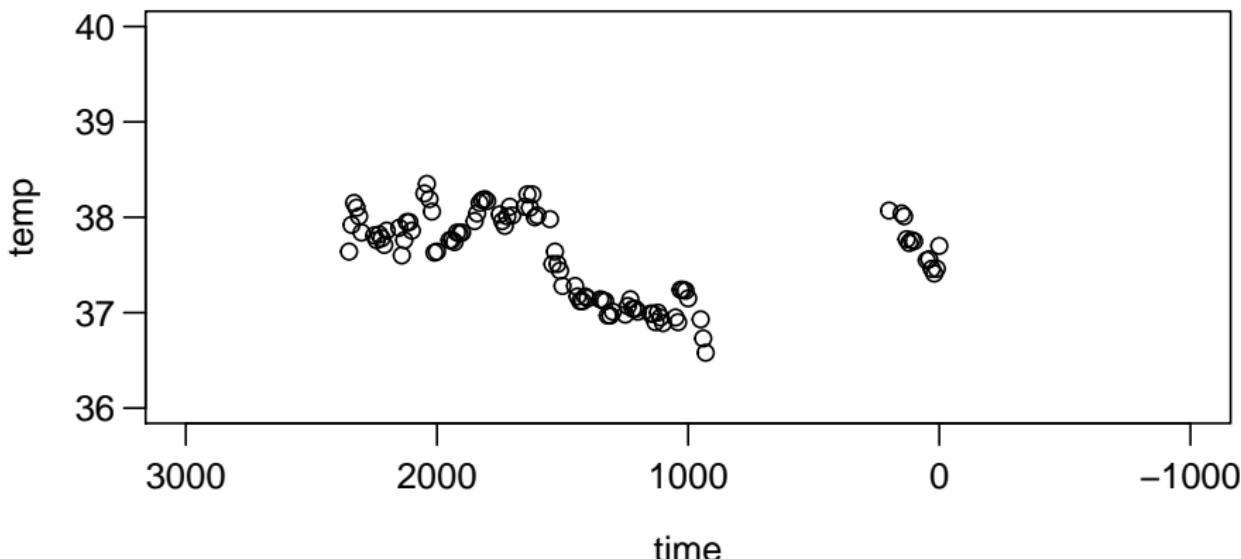
```
plot(beaver2[ ,2:3], xlim=c(-1000,3000), ylim=c(36,40))
```



## Plot limits II

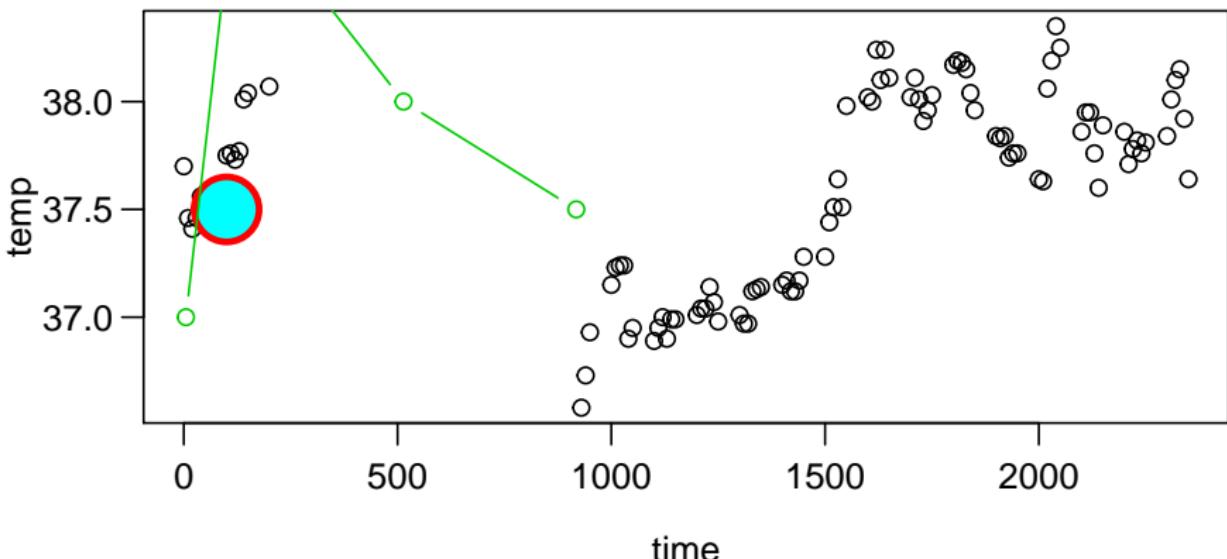
xaxis decreasing (can be very misleading!):

```
plot(beaver2[ ,2:3], xlim=c(3000,-1000), ylim=c(36,40))
```



## Adding datapoints

```
plot(beaver2[, 2:3])
points(100, 37.5, pch=21, bg=5, cex=4, col=2, lwd=3)
lines(x=c(5,120,514,918), y=c(37,39,38,37.5), col=3, type="b")
```



# Now it's your turn...

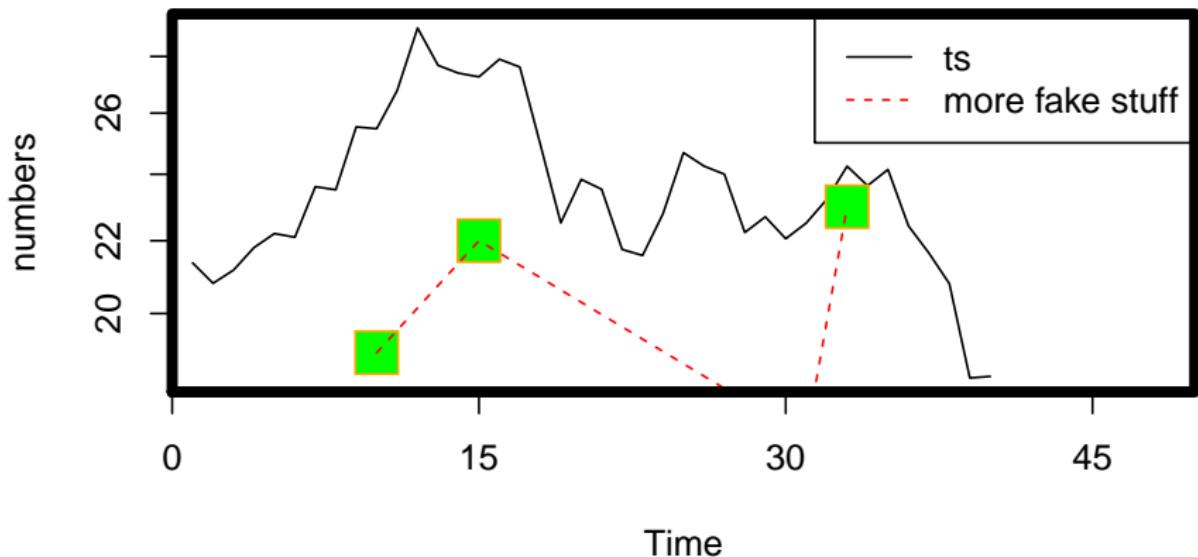
## Exercise 17: Plotting

- ① Plot a dataset of your choice (eg from `library(help = "datasets")`).
- ② Add some red dotted lines between big quadratic points, filled with blue color.
- ③ Label the axis and give your graph a big italic header with `cex.main`, `font.main`, see [?par](#).
- ④ Plot with a logarithmic scale on one axis.
- ⑤ Set the axis to a certain limit. With `yaxs`, suppress the default 4% range extension.
- ⑥ Using `xaxt`, plot without an x-axis. Look in the help of [par](#) how this can be done. Then add a custom `axis`.
- ⑦ Add a `legend` to the plot and thick-lined `box` around the final plot.

# Solution for exercise 17: Plotting I

```
# generate "random" data:  
set.seed(42) ; numbers <- cumsum(rnorm(40)) + 20  
x <- c(10, 15, 31, 33); y <- c(19, 22, 17, 23)  
# Basic plot:  
plot(numbers, type="l", xlab="Time",  
      main="Fake Time Series", cex.main=3, font.main=3,  
      log="y", xlim=c(0,50), xaxs="i", xaxt="n" )  
# Add points and lines  
points(x=x, y=y, pch=22, type="p", cex=3, col="orange",  
       bg="green")  
lines(x=c(10, 15, 31, 33), y=c(-6, -3, -8, -2)+25, col=2,  
      lty="dashed")  
# Custom axis, legend and box around plot:  
axis(side=1, at=c(0, 15, 30, 45) )  
legend("topright", legend=c("ts", "more fake stuff"),  
      lty=c(1,2), col=c(1,2) )  
box(lwd=5)
```

## Fake Time Series



# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. **Plotting**
    - Syntax, adjustments, labeling
    - Lines
    - Limits, adding to graphics
    - saving plots, multiple figures**
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Saving plots to a file

In the menu of the graphics window, you can save the current plot by hand.  
For reproducible sizes and if you're creating many plots, use

# Saving plots to a file

In the menu of the graphics window, you can save the current plot by hand.  
For reproducible sizes and if you're creating many plots, use

```
png("FileName.png", width=12, height=9, units="in",
     res=300, bg="transparent", pointsize=14)
plot(1:20)
title(main="Plot title")
points(5,9)
dev.off() # to close the external device
```

# Saving plots to a file

In the menu of the graphics window, you can save the current plot by hand.  
For reproducible sizes and if you're creating many plots, use

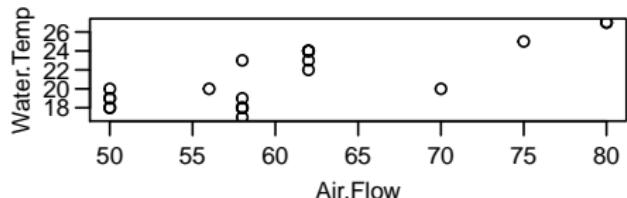
```
png("FileName.png", width=12, height=9, units="in",
     res=300, bg="transparent", pointsize=14)
plot(1:20)
title(main="Plot title")
points(5,9)
dev.off() # to close the external device
```

The functions `pdf`, `jpeg`, etc. are also available.

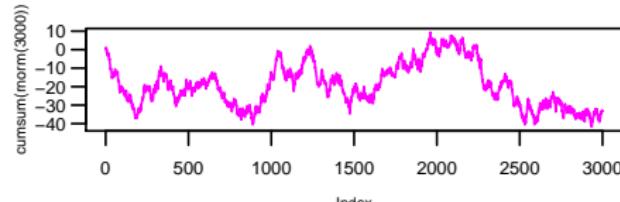
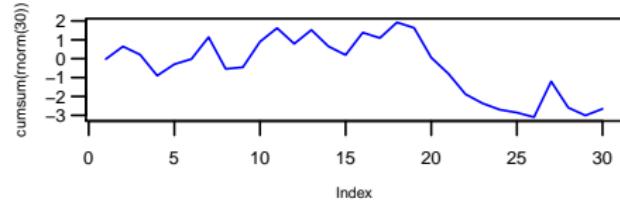
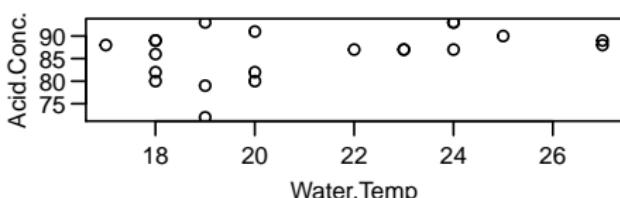
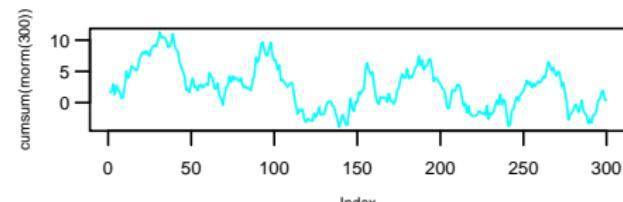
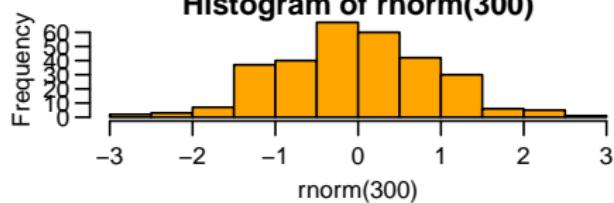
# multiple figure layout |

# 3 rows, 2 columns:

```
par(mfrow=c(3,2), mar=c(3,3,1,1), mgp=c(1.9, 0.7, 0), las=1)
plot(stackloss[,1:2]) # 6 plot commands
```

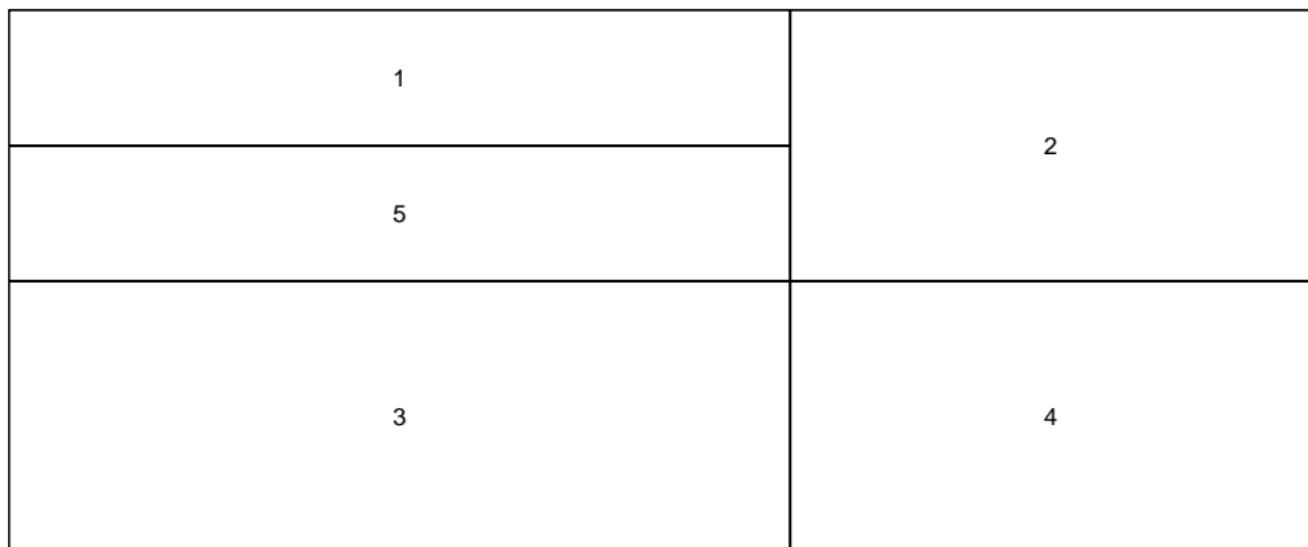


Histogram of  $rnorm(300)$



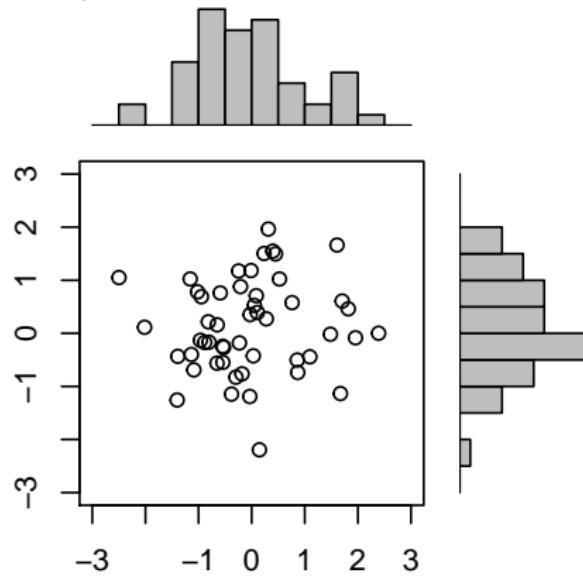
# multiple figure layout II

```
lay <- layout(matrix(c(1,1,2,2,
                      5,5,2,2,
                      3,3,4,4,
                      3,3,4,4), ncol=4, byrow=TRUE),
                 widths=c(6,6,4,4))
layout.show(lay)
```

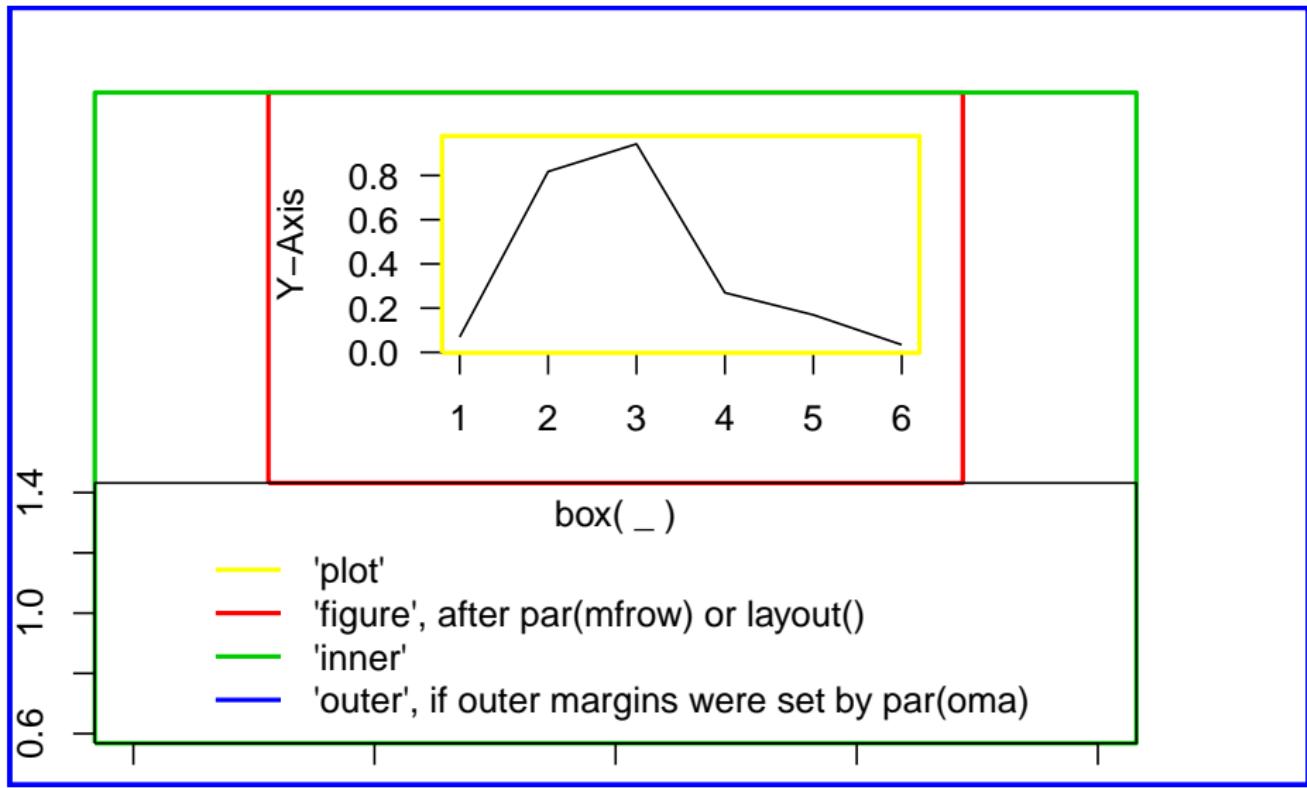


# multiple figure layout III

From the example in `?layout`:



# Box



# Enough talk, action...

## Exercise 18: Export (multipanel) plots

- ① Read `Zugspitze.txt` (*rightclick Raw, save as*) with monthly weather records
- ② With `par(mfrow=...)`, set up a figure with two panels (left and right)
- ③ Plot a histogram of the LUFTTEMPERATUR records in the left panel
- ④ Plot a line graph of the first 50 months in the right panel
- ⑤ Now export this graph as png with useful size and resolution
- ⑥ BONUS 1: also export as a vector graph (e.g. PDF, EPS, SVG, `svglite` package)
- ⑦ BONUS 2: Set smaller margins (`mar`) around each panel
- ⑧ BONUS 3: Set an outer margin (`oma`) with an outer `title` for the entire figure

## Solution for exercise 18: Export (multipanel) plots I

```
wzug <- read.table("data/Zugspitze.txt", header=TRUE)

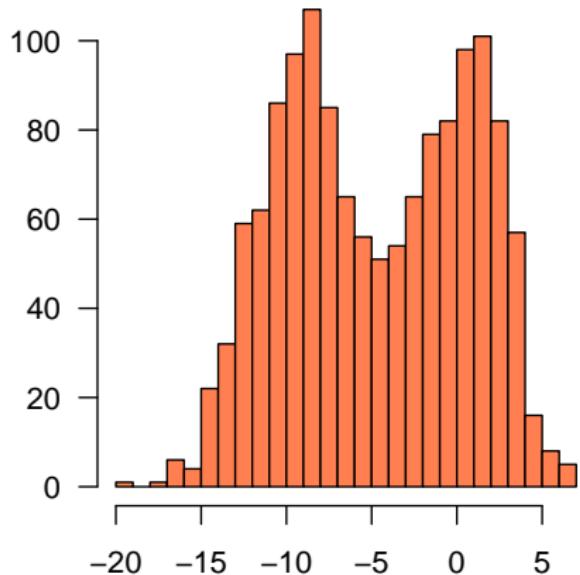
pdf("fig/zugspitze.pdf", height=4)
par(mfrow=c(1,2), mar=c(3,3,1,1), oma=c(0,0,2,0), las=1)
hist(wzug$LUFTTEMPERATUR, col="coral", breaks=25,
      main="Histogram monthly values")
plot(wzug$LUFTTEMPERATUR[1:50], type="l", lwd=2,
      col="darkorchid", main="First 50 months")
title(main="Temperature Zugspitze 1900:2015", outer=TRUE)
dev.off()

## pdf
## 2
```

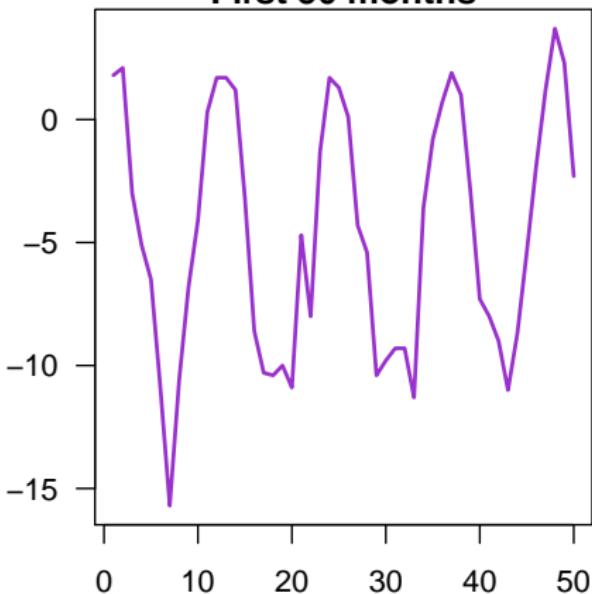
# Solution for exercise 18: Export (multipanel) plots II

Temperature Zugspitze 1900:2015

Histogram monthly values

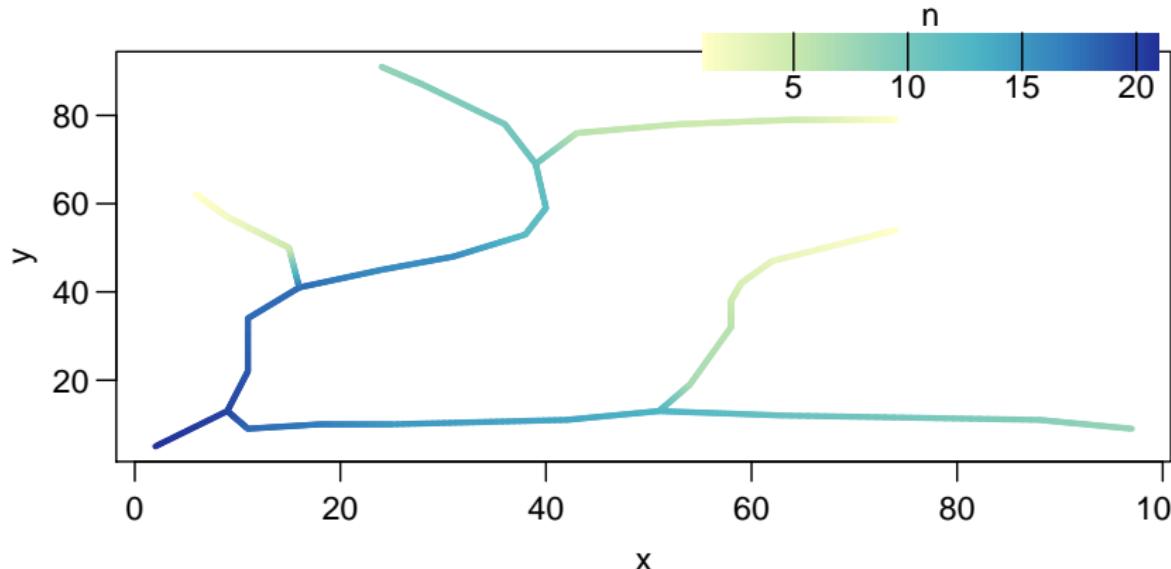


First 50 months



# Graphing paths: NA lines between entry groups

```
library("berryFunctions") # >= 1.9.2 (2015-10-22) for insertRows, colPoints  
tfile <- system.file("extdata/rivers.txt", package="berryFunctions")  
rivers <- read.table(tfile, header=TRUE, dec=",")  
colPoints(x,y,n, data=rivers, add=FALSE, lines=TRUE, lwd=3,  
          legargs=c(y1=0.8,y2=1, density=FALSE) )
```



# Base R vs ggplot2

This has all been graphing with the possibilities of base R.

There is also the package `ggplot2` which a lot of people love to use.

I get everything done with base R and don't see a need to learn ggplot grammar.

You can do whatever you want - it's both R, both have their advantages.

A very good place to start learning ggplot2 is:

<http://minimaxir.com/2017/08/ggplot2-web/>

To compare both paradigms, see:

<http://flowingdata.com/2016/03/22/>

[comparing-ggplot2-and-r-base-graphics/](http://flowingdata.com/2016/03/22/comparing-ggplot2-and-r-base-graphics/)

Be sure to follow the links in the beginning of the article...

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
    - Character strings: Intro
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Character strings I

```
paste("NiceThing", 3:5) # Creates vector of Mode character  
## [1] "NiceThing 3" "NiceThing 4" "NiceThing 5"  
  
paste(2, "hot", 4, "U", sep="_")  
## [1] "2_hot_4_U"  
  
letvec <- c("some different", "let", "ters")  
letvec  
  
## [1] "some different" "let"  
## [3] "ters"
```

## Character strings II

```
length(letvec)
## [1] 3

nchar(letvec)
## [1] 14  3  4

paste("nice", letvec, sep="_")
## [1] "nice_some different" "nice_let"
## [3] "nice_ters"
```

## Character strings III

```
paste(letvec, sep="")  
  
## [1] "some different" "let"  
## [3] "ters"  
  
paste(letvec, collapse="")  
  
## [1] "some differentletters"  
  
# remove spaces: substitute " " in letvec with ""  
sub(" ", "", letvec)  
  
## [1] "somedifferent" "let"  
## [3] "ters"
```

## Character strings IV

```
substr("Monster Party", start=2, stop=5)

## [1] "onst"

# If you need several string segments:
substring("Monster Party", first=c(1,3,5), last=c(2,5,9) )

## [1] "Mo"      "nst"     "ter P"

strsplit("Die Aerzte - Monster Party", split=" - ")[[1]]

## [1] "Die Aerzte"      "Monster Party"
```

## Backslash "\\" : Escape Character

- "\\n" for return or line break (newline), eg in graphictitles and -texts
- "\\t" for Tabstop, e.g. for `write.table("path/file.txt", yourdata, sep="\t")`
- "\\\\\" for Backslash as symbol in `title`, `text`, `legend` etc.
- "\\\" for " as character
- "\\U" for Unicode, e.g.:

```
plot(1)
# average symbol without using "expression":
mtext(paste("\\U00D8", 5.4, "\\U00B0 C"), col=2)
```

# Time to practice...

## Exercise 19: character string operations

In the German chocolate industry, we love

```
schoko <- c("Ritter Sport", "Lindor", "Frigor", "Milka", "Frey",  
"More Lindor")
```

- ① which element **is** or **contains** "Lindor"?

Find out with `match`, `grep`, `%in%` and `grepl`. What are the differences between those?

- ② Replace "or" with "nicor" using `sub` and `gsub`.
- ③ What's the problem with `"Ritter sport" %in% schoko`?
- ④ What does `"Ritter [s,S]port" %in% schoko` do?

# Solution for exercise 19.1: Character strings

```
schoko <- c("Lindor", "Ritter Sport", "Frigor", "Milka",
           "Frey", "More Lindor")
# Which element in schoko IS Lindor
match("Lindor", schoko)
match(schoko, "Lindor")
match(schoko, c("Lindor", "Frigor") )

# Which element in schoko CONTAINS Lindor
grep("Lindor", schoko)
grep("or", schoko)
grepl("Lindor", schoko) # logical value

# Is one of the elements of schoko == "Lindor" at all?
"Lindor" %in% schoko
any( grepl("Lindor", schoko) ) # any TRUE?
all( grepl("Lindor", schoko) ) # All TRUE
! grepl("Lindor", schoko)
```

## Solution for exercise 19.2-41: Character strings

```
# Replace Character strings (substitute)
sub(pattern="or", replacement="nicor", schoko)
# Replace all occurrences (global substitution)
gsub(pattern="or", replacement="nicor", schoko)

# Avoid capitalization issues with
tolower(schoko)
"Ritter [s,S]port" %in% schoko
# Does not work!!
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  - 7. Conditions & loops**
    - Conditions**
      - For Loops
      - lapply, Rcpp
      - Repeat + While Loops
      - Arrays
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

# Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

```
ifelse(condition, expression1, expression2)
```

# Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

```
ifelse(condition, expression1, expression2)
```

If condition == TRUE, then expression1 is evaluated, if condition == FALSE, then expression2 is evaluated.

# Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

```
ifelse(condition, expression1, expression2)
```

If condition == TRUE, then expression1 is evaluated, if condition == FALSE, then expression2 is evaluated.

```
7-3 > 2
class(7-3 > 2 )
if(7-3 > 2) 18
if(7-3 > 5) 18
if(7-3 > 5) 18 else 17
```

## Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

```
ifelse(condition, expression1, expression2)
```

If condition == TRUE, then expression1 is evaluated, if condition == FALSE, then expression2 is evaluated.

```
class(7-3 > 2 )
```

**if**(7-3 > 2) 18

**if**(7-3 > 5) 18

```
if(7-3 > 5) 18 else 17
```

## Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

**ifelse**(condition, expression1, expression2)

If condition == TRUE, then expression1 is evaluated, if condition == FALSE, then expression2 is evaluated.

$$7-3 > 2$$

`class(7-3 > 2 )` logical = truth value, boolean

**if**(7-3 > 2) 18

**if**(7-3 > 5) 18

```
if(7-3 > 5) 18 else 17
```

# Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

```
ifelse(condition, expression1, expression2)
```

If condition == TRUE, then expression1 is evaluated, if condition == FALSE, then expression2 is evaluated.

```
7-3 > 2
```

TRUE

```
class(7-3 > 2 )
```

logical = truth value, boolean

```
if(7-3 > 2) 18
```

Condition is TRUE, so 18 is returned

```
if(7-3 > 5) 18
```

```
if(7-3 > 5) 18 else 17
```

# Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

```
ifelse(condition, expression1, expression2)
```

If condition == TRUE, then expression1 is evaluated, if condition == FALSE, then expression2 is evaluated.

7-3 > 2	TRUE
class(7-3 > 2 )	logical = truth value, boolean
if(7-3 > 2) 18	Condition is TRUE, so 18 is returned
if(7-3 > 5) 18	Condition is FALSE, so nothing happens
if(7-3 > 5) 18 else 17	

# Conditional execution I

Syntax for a single logical value:

```
if(this_is_true) {do_something} else {do_other_thing}
```

Syntax for a vector with several T/F values:

```
ifelse(condition, expression1, expression2)
```

If condition == TRUE, then expression1 is evaluated, if condition == FALSE, then expression2 is evaluated.

7-3 > 2	TRUE
class(7-3 > 2 )	logical = truth value, boolean
if(7-3 > 2) 18	Condition is TRUE, so 18 is returned
if(7-3 > 5) 18	Condition is FALSE, so nothing happens
if(7-3 > 5) 18 else 17	Condition FALSE, so 17 is returned.

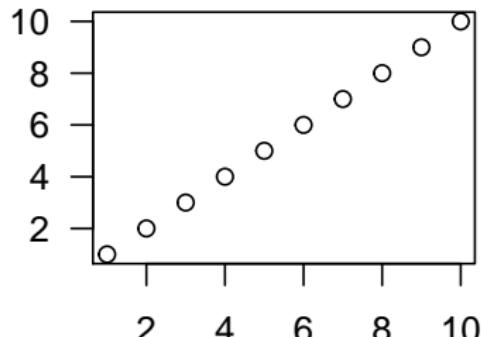
## Conditional execution II

```
# Several commands must be held together with curly braces:  
if(TRUE)  
{  
  plot(1:10, main="TRUE in code")  
  box("figure", col=4, lwd=3)  
} else  
# do something else: plot random numbers  
{  
  plot(rnorm(500), main="FALSE in code")  
}  
# these last brackets can (but should not) be left away  
# indenting code makes it readable for humans  
  
par(mfrow=c(1,2), cex=1, las=1)
```

# Conditional execution III

**TRUE in code**

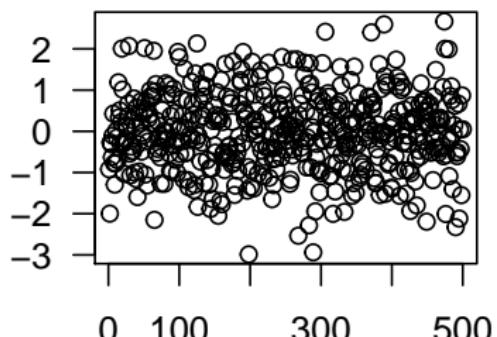
1:10



Index

**FALSE in code**

rnorm(500)



Index

Vectorising: `if(c) e1 else e2` vs `ifelse(c, e1, e2)`

```
v <- c(13, 14, 15, 16, 17)
v>14

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

## Vectorising: `if(c) e1 else e2` vs `ifelse(c, e1, e2)`

```
v <- c(13, 14, 15, 16, 17)
```

```
v>14
```

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

```
ifelse(v>14, v+10, NA) # can handle vector of input
```

```
## [1] NA NA 25 26 27
```

## Vectorising: `if(c) e1 else e2` vs `ifelse(c, e1, e2)`

```
v <- c(13, 14, 15, 16, 17)  
v>14
```

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

```
ifelse(v>14, v+10, NA) # can handle vector of input
```

```
## [1] NA NA 25 26 27
```

```
if(v>14) v+10 else NA
```

```
## Warning in if (v > 14) v + 10 else NA: the condition has  
length > 1 and only the first element will be used
```

```
## [1] NA
```

# Time to practice

## Exercise 20: if else - Conditional code execution

- ➊ `sqrt` returns NaN for the negative values in `v <- -3:5` and warns about it. With a conditional expression, pass 0 instead of negative values to `sqrt`.
- ➋ Construct a statement that checks whether the variable `input <- 4` is a number smaller than 3. Let it write a useful `message` to the console (for both cases). Now test it with `input <- 1.8` and `input <- -17`.
- ➌ Now restrict the correct value of `input` to *positive* numbers  $< 3$ , i.e. the number must be  $< 3$  AND  $\geq 0$ .
- ➍ BONUS 1: What happens if `input <- "2"` or if `input <- "b"`?
- ➎ BONUS 2: Create a character variable that, depending on the result of `rnorm(1)`, is initiated with `probable` or `unlikely`.
- ➏ BONUS 3: `replicate` this experiment 1000 times and examine the result with `table`.
- ➐ BONUS 4: How could you do this with `ifelse`?

## Solution for exercise 20.1: if else

```
v <- -3:5
sqrt(v)

## Warning in sqrt(v):  NaNs produced

## [1]      NaN      NaN      NaN 0.0000 1.0000 1.4142 1.7321 2.0000 2.2361
```

```
ifelse( v >= 0, sqrt(v), 0)

## Warning in sqrt(v):  NaNs produced

## [1] 0.0000 0.0000 0.0000 0.0000 1.0000 1.4142 1.7321 2.0000 2.2361
```

```
sqrt(ifelse( v >= 0, v, 0))

## [1] 0.0000 0.0000 0.0000 0.0000 1.0000 1.4142 1.7321 2.0000 2.2361
```

## Solution for exercise 20.2: if else

```
input <- 4
if( input >= 3 ) message("Input was wrong.
                           It should be <3") else
  message("Input OK")

## Input was wrong.
##                               It should be <3

# run it again after
input <- 1.8
input <- -17
```

## Solution for exercise 20.3: if else

```
# three different solutions:  
if( input >= 3 ) message("Input is > 3") else  
if( input < 0 ) message("Input is < 0") else  
    message("Input OK")  
  
if( input >= 3 | input < 0)  
    message("Input outside 0...3") else  
    message("Input OK")  
  
if( input > 0 & input <= 3 ) message("Input OK") else  
    message("Input (",input,") outside 0...3")
```

## Solution for exercise 20.3: if else BONUS

```
result <- if(rnorm(1)>2) "unlikely" else "probable"

result <- replicate(n=1000, expr=
    if(rnorm(1)>2) "unlikely" else "probable")
table(result)

result <- ifelse(rnorm(1000)>2, "unlikely", "probable")
table(result)
```

# Notes on logical values

- as you might have seen in `read.table(header=T)`, logical values (TRUE, FALSE) can be abbreviated.

## Notes on logical values

- as you might have seen in `read.table(header=T)`, logical values (TRUE, FALSE) can be abbreviated.
- If you want to play a mean prank on someone, write `T <- FALSE; F <- TRUE` in their `Rprofile` (see `?Startup`).

## Notes on logical values

- as you might have seen in `read.table(header=T)`, logical values (TRUE, FALSE) can be abbreviated.
- If you want to play a mean prank on someone, write `T <- FALSE; F <- TRUE` in their `Rprofile` (see `?Startup`).
- Logical (boolean) values F and T internally are often converted to integers 0 and 1, thus `sum(c(T,F,F,T,T,T,F,F,T))` is the number of TRUEs in a vector, `mean` yields the proportion of TRUEs.

# Notes on logical values

- as you might have seen in `read.table(header=T)`, logical values (TRUE, FALSE) can be abbreviated.
- If you want to play a mean prank on someone, write `T <- FALSE; F <- TRUE` in their `Rprofile` (see `?Startup`).
- Logical (boolean) values F and T internally are often converted to integers 0 and 1, thus `sum(c(T,F,F,T,T,T,F,F,T))` is the number of TRUEs in a vector, `mean` yields the proportion of TRUEs.
- `which(logicalVec)` gives the indices (positions) of TRUE values.

# Notes on logical values

- as you might have seen in `read.table(header=T)`, logical values (TRUE, FALSE) can be abbreviated.
- If you want to play a mean prank on someone, write `T <- FALSE; F <- TRUE` in their `Rprofile` (see `?Startup`).
- Logical (boolean) values F and T internally are often converted to integers 0 and 1, thus `sum(c(T,F,F,T,T,T,F,F,T))` is the number of TRUEs in a vector, `mean` yields the proportion of TRUEs.
- `which(logicalVec)` gives the indices (positions) of TRUE values.
- `Vec[logicalVec]` returns only the values of `vec` corresponding to TRUE in `logicalVec` (No need to wrap it into `which`).

# Notes on logical values

- as you might have seen in `read.table(header=T)`, logical values (TRUE, FALSE) can be abbreviated.
- If you want to play a mean prank on someone, write `T <- FALSE; F <- TRUE` in their `Rprofile` (see `?Startup`).
- Logical (boolean) values F and T internally are often converted to integers 0 and 1, thus `sum(c(T,F,F,T,T,T,F,F,T))` is the number of TRUEs in a vector, `mean` yields the proportion of TRUEs.
- `which(logicalVec)` gives the indices (positions) of TRUE values.
- `Vec[logicalVec]` returns only the values of `vec` corresponding to TRUE in `logicalVec` (No need to wrap it into `which`).
- Logical operators: `!`, `&`, `|`, `xor()` (not, and, or, exclusive or)

# Notes on conditional code execution

- `if(c){ex1}` is valid code, thus R doesn't expect `else` anymore.

# Notes on conditional code execution

- `if(c){ex1}` is valid code, thus R doesn't expect `else` anymore.
- If you execute code line by line (in a script, for example), `}` and `else` must be on the same line.

# Notes on conditional code execution

- `if(c){ex1}` is valid code, thus R doesn't expect `else` anymore.
- If you execute code line by line (in a script, for example), `}` and `else` must be on the same line.
- Many people consider it good practice to do this in functions as well, but for machine-readability, it is technically fine to write

```
if(cond)
{
  ex1a
  ex1b
}
else
  ex2
```

# Notes on conditional code execution

- `if(c){ex1}` is valid code, thus R doesn't expect `else` anymore.
- If you execute code line by line (in a script, for example), `}` and `else` must be on the same line.
- Many people consider it good practice to do this in functions as well, but for machine-readability, it is technically fine to write  
`if(cond)`  
  {  
    ex1a  
    ex1b  
  }  
`else`  
  ex2
- `if(logicalValue==TRUE) ...` is usually unnecessary, you can write

# Notes on conditional code execution

- `if(c){ex1}` is valid code, thus R doesn't expect `else` anymore.
- If you execute code line by line (in a script, for example), `}` and `else` must be on the same line.
- Many people consider it good practice to do this in functions as well, but for machine-readability, it is technically fine to write  
`if(cond)`  
  {  
    ex1a  
    ex1b  
  }  
`else`  
  ex2
- `if(logicalValue==TRUE) ...` is usually unnecessary, you can write `if(logicalValue) ...`, but sometimes,

# Notes on conditional code execution

- `if(c){ex1}` is valid code, thus R doesn't expect `else` anymore.
- If you execute code line by line (in a script, for example), `}` and `else` must be on the same line.
- Many people consider it good practice to do this in functions as well, but for machine-readability, it is technically fine to write

```
if(cond)
{
  ex1a
  ex1b
}
else
  ex2
```

- `if(logicalValue==TRUE) ...` is usually unnecessary, you can write `if(logicalValue) ...`, but sometimes, `if(isTRUE(logicalValue)) ...` helps to deal with NAs.

# Actual usage of if else statements

See the `hist` source code:

[github.com/wch/r-source](https://github.com/wch/r-source) -> src / library / graphics / R / hist.R

# Actual usage of if else statements

See the `hist` source code:

[github.com/wch/r-source](https://github.com/wch/r-source) -> src / library / graphics / R / hist.R

```
mad
```

# Actual usage of if else statements

See the `hist` source code:

[github.com/wch/r-source](https://github.com/wch/r-source) -> src / library / graphics / R / hist.R

```
mad
```

Multiple nested conditionals

`if(a) b else if(c) d else e`

can be avoided with `switch`.

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations

## 7. Conditions & loops

Conditions

For Loops

lapply, Rcpp

Repeat + While Loops

Arrays

## 8. Functions & packages

## 9. Distributions

## 10. Statistics

## 11. Time series handling and analysis

## 12. Spatial data and GIS functionality

## 13. Final tasks

Course plan

# For loops

Execute a block of code several times, with different input values.

Syntax: `for(aRunningVariable in aSequence){ doSomething }`

# For loops

Execute a block of code several times, with different input values.

Syntax: `for(aRunningVariable in aSequence){ doSomething }`

Often, `i` (for index) is used, thus `for(i in 1:n) doThis(i)`

# For loops

Execute a block of code several times, with different input values.

Syntax: `for(aRunningVariable in aSequence){ doSomething }`

Often, `i` (for index) is used, thus `for(i in 1:n) doThis(i)`

```
help("for") # needs quotation marks!
```

# For loops

Execute a block of code several times, with different input values.

Syntax: `for(aRunningVariable in aSequence){ doSomething }`

Often, `i` (for index) is used, thus `for(i in 1:n) doThis(i)`

```
help("for") # needs quotation marks!
```

```
print(1:2)
print(1:5)
print(1:9)
```

# For loops

Execute a block of code several times, with different input values.

Syntax: `for(aRunningVariable in aSequence){ doSomething }`

Often, `i` (for index) is used, thus `for(i in 1:n) doThis(i)`

```
help("for") # needs quotation marks!
```

```
print(1:2)
print(1:5)
print(1:9)
```

This is easier and less prone to human errors with:

# For loops

Execute a block of code several times, with different input values.

Syntax: `for(aRunningVariable in aSequence){ doSomething }`

Often, `i` (for index) is used, thus `for(i in 1:n) doThis(i)`

```
help("for") # needs quotation marks!
```

```
print(1:2)
print(1:5)
print(1:9)
```

This is easier and less prone to human errors with:

```
for(i in c(2,5,9) ) { print(1:i) }
```

```
## [1] 1 2
## [1] 1 2 3 4 5
## [1] 1 2 3 4 5 6 7 8 9
```

## For loops: fill a vector

```
v <- vector(mode="numeric", length=20)
v

## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
```

## For loops: fill a vector

```
v <- vector(mode="numeric", length=20)
v

## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

for(i in 3:17) { v[i] <- (i+2)^2 }
```

## For loops: fill a vector

```
v <- vector(mode="numeric", length=20)
v

## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

for(i in 3:17) { v[i] <- (i+2)^2 }

v # this code was executed once for each i

## [1] 0 0 25 36 49 64 81 100 121
## [10] 144 169 196 225 256 289 324 361 0
## [19] 0 0
```

## For loops: fill a vector

```
v <- vector(mode="numeric", length=20)
v

## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

for(i in 3:17) { v[i] <- (i+2)^2 }

v # this code was executed once for each i

## [1] 0 25 36 49 64 81 100 121
## [10] 144 169 196 225 256 289 324 361 0
## [19] 0 0
```

See [Rclick 10.4](#) for a realistic for loop example.

## For loops: fill a vector

```
v <- vector(mode="numeric", length=20)
v

## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

for(i in 3:17) { v[i] <- (i+2)^2 }

v # this code was executed once for each i

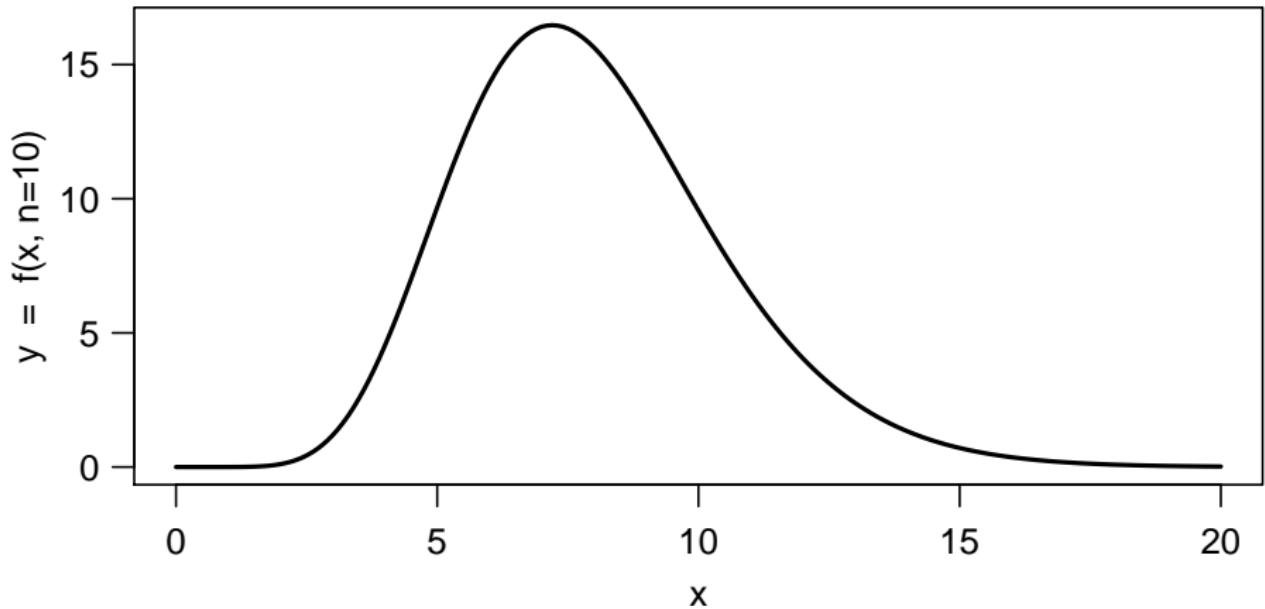
## [1] 0 25 36 49 64 81 100 121
## [10] 144 169 196 225 256 289 324 361 0
## [19] 0 0
```

See [Rclick 10.4](#) for a realistic for loop example.

In R, `for` loops are slow. Always try to vectorize (the best option, not always possible) or use `lapply` (saves you the initiation of the empty vector, easier to parallelize).

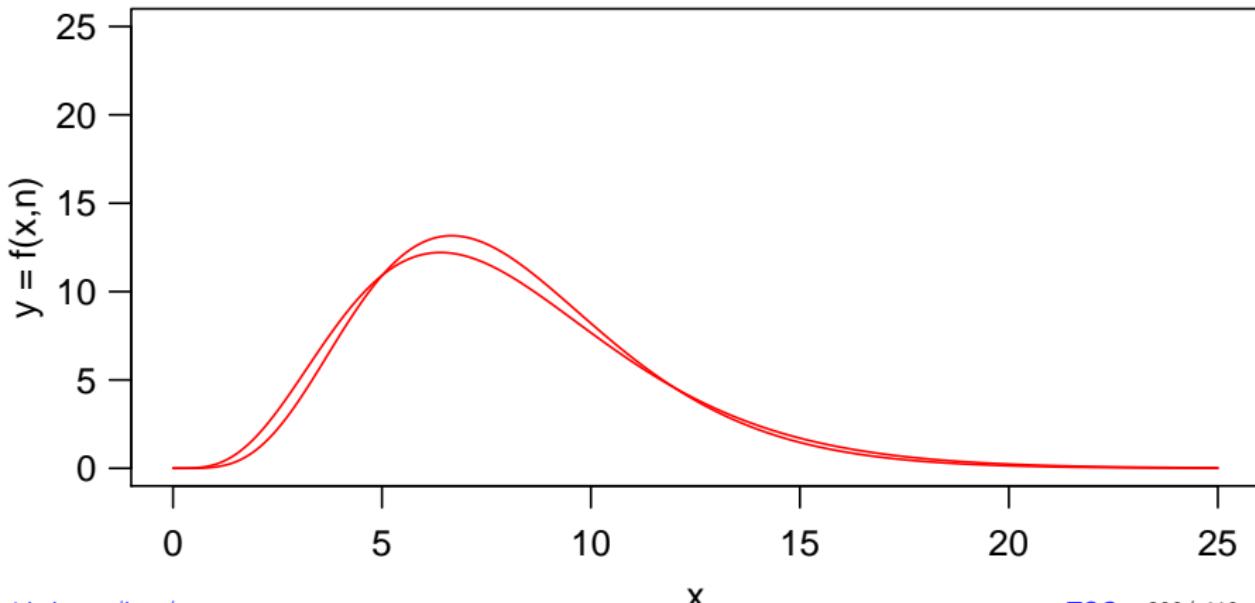
## For loops: execute code multiple times I

$$y = f(x, n) = \frac{12.5 \cdot n}{(n-1)!} \cdot \left(\frac{nx}{8}\right)^{(n-1)} \cdot e^{-\frac{nx}{8}}$$



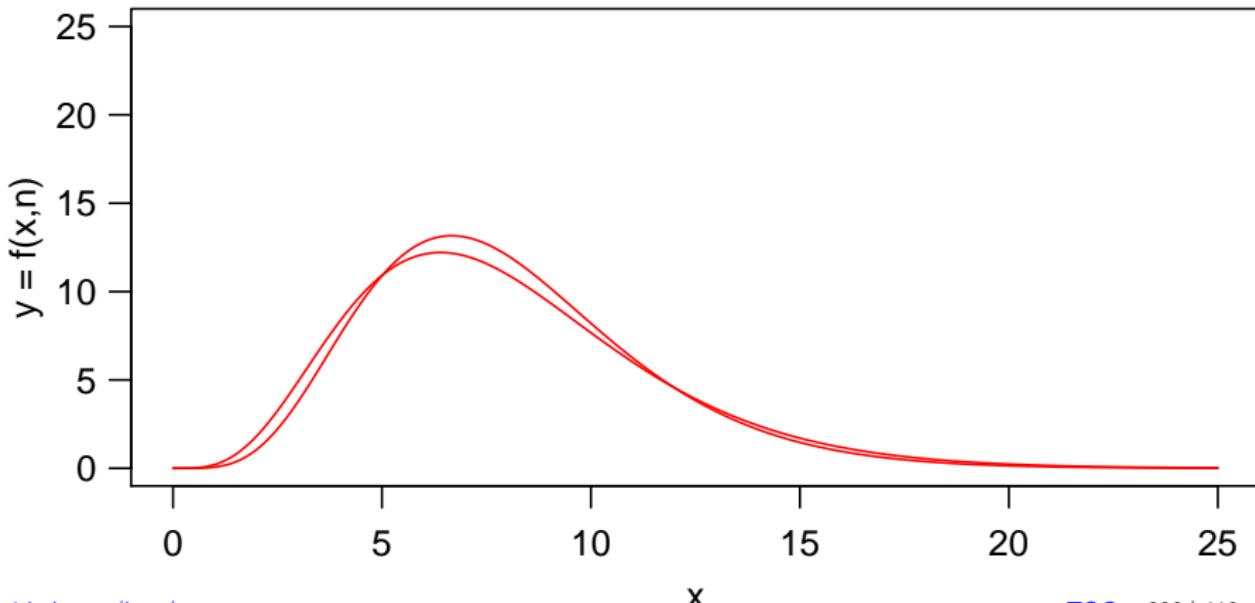
## For loops: execute code multiple times II

```
x <- seq(0,25,0.1)
plot(x,x, type="n", ylab="y = f(x,n)")
lines(x, 12.5*5/factorial(5-1)*(x/8*5)^(5-1)*exp(-x/8*5), col=2)
lines(x, 12.5*6/factorial(6-1)*(x/8*6)^(6-1)*exp(-x/8*6), col=2)
```



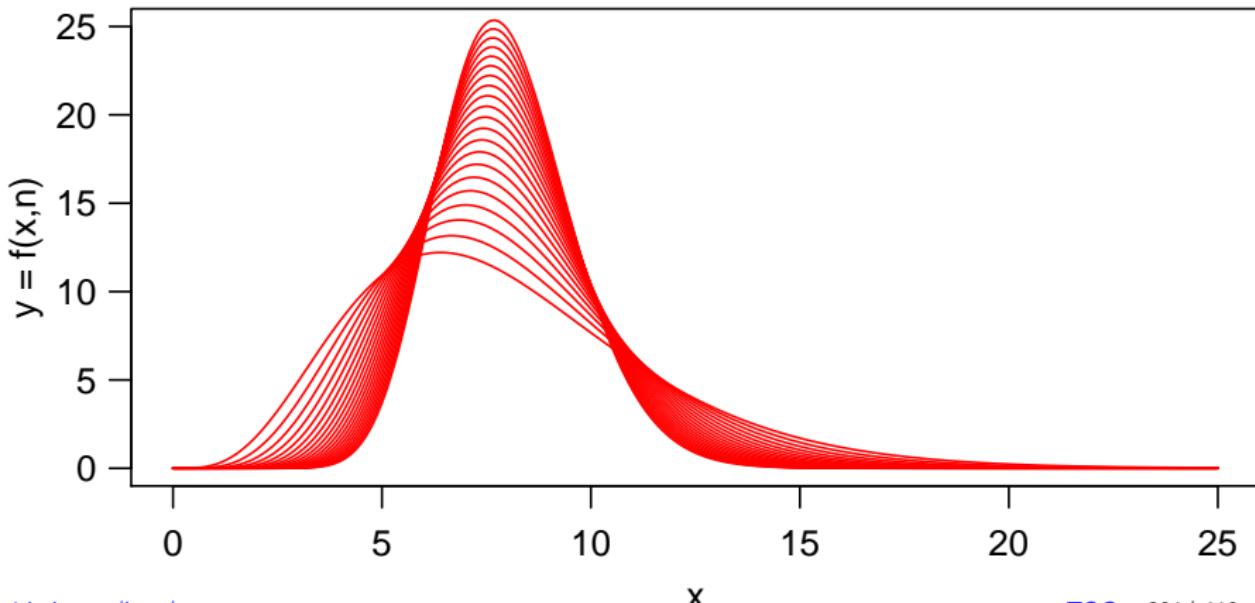
## For loops: execute code multiple times II

```
x <- seq(0,25,0.1)
plot(x,x, type="n", ylab="y = f(x,n)")
lines(x, 12.5*5/factorial(5-1)*(x/8*5)^(5-1)*exp(-x/8*5), col=2)
lines(x, 12.5*6/factorial(6-1)*(x/8*6)^(6-1)*exp(-x/8*6), col=2)
```



## For loops: execute code multiple times III

```
x <- seq(0,25,0.1)
plot(x,x, type="n", ylab="y = f(x,n)")
for (n in 5:25)
  lines(x, 12.5*n/factorial(n-1)*(x/8*n)^(n-1)*exp(-x/8*n), col=2)
```



# stocks data from finance.yahoo.com

```
# Download current datasets:  
if(!requireNamespace("quantmod")) install.packages("quantmod")  
if(!requireNamespace("pbapply")) install.packages("pbapply")  
dummy <- pbapply::pblapply(c("F", "VLKAF", "AMZN", "AAPL", "GOOG", "MSFT"),  
    function(x) zoo::write.zoo(x=quantmod::getSymbols(x, auto.assign=FALSE)[,6],  
        file=paste0("data/finance/", x, ".txt"), col.names=T))  
  
# read single files to R and merge into one file:  
stocks <- lapply(dir("data/finance", full=TRUE),  
    read.table, as.is=TRUE, header=TRUE)  
stocks <- Reduce(function(...) merge(..., all=T), stocks)  
  
# Get nicer column names:  
names <- sapply(strsplit(colnames(stocks), ".", fix=TRUE), "[", 1)  
colnames(stocks) <- c(Index="Date", F="FORD", VLKAF="VOLKSWAGEN",  
    AMZN="AMAZON", AAPL="APPLE", GOOG="GOOGLE", MSFT="MICROSOFT")[names]  
  
# Save to disc:  
write.table(stocks, file="data/stocks.txt", row.names=F, quote=F)
```

# For loops: multipanel graphics: the task

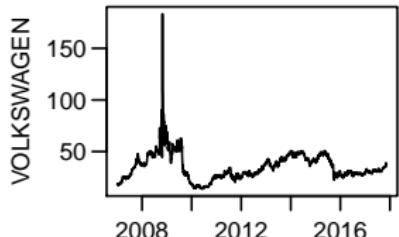
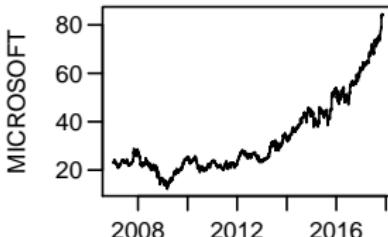
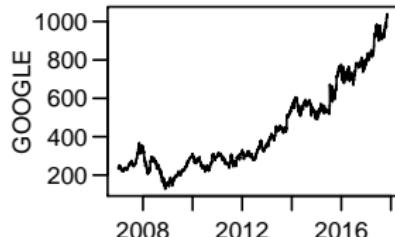
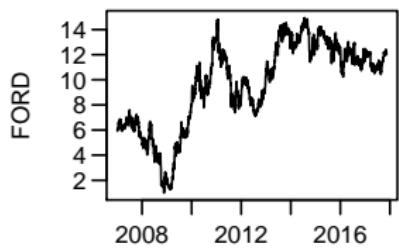
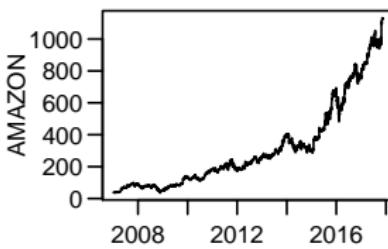
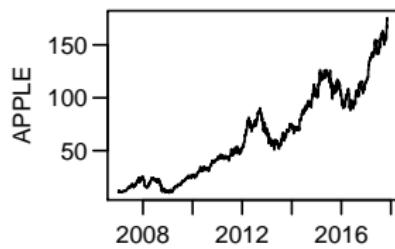
## Exercise 21: for loop

- ① Read `stocks.txt` (*rightclick Raw, save as*), so that there are no factors in the `data.frame`
- ② Change the first column type from `char` to `date` with `?as.Date`
- ③ What do you get with `plot(stocks[ ,1:2])`? Make it a line graph.
- ④ With `par(mfrow...,` set up a two by three panel plot
- ⑤ With a for loop, fill those with each stock time series
- ⑥ BONUS 1: Make good annotations, including a main title (`par oma, mtext` with the outer argument)
- ⑦ BONUS 2: Make the plot margins smaller (`par mar`), turn y axis labels upright (`las`) and move the axis labels closer to the plots (`mgp`).
- ⑧ BONUS 3: Understand and comment each line of the data preparation.

# For loops: multipanel graphics: the solution

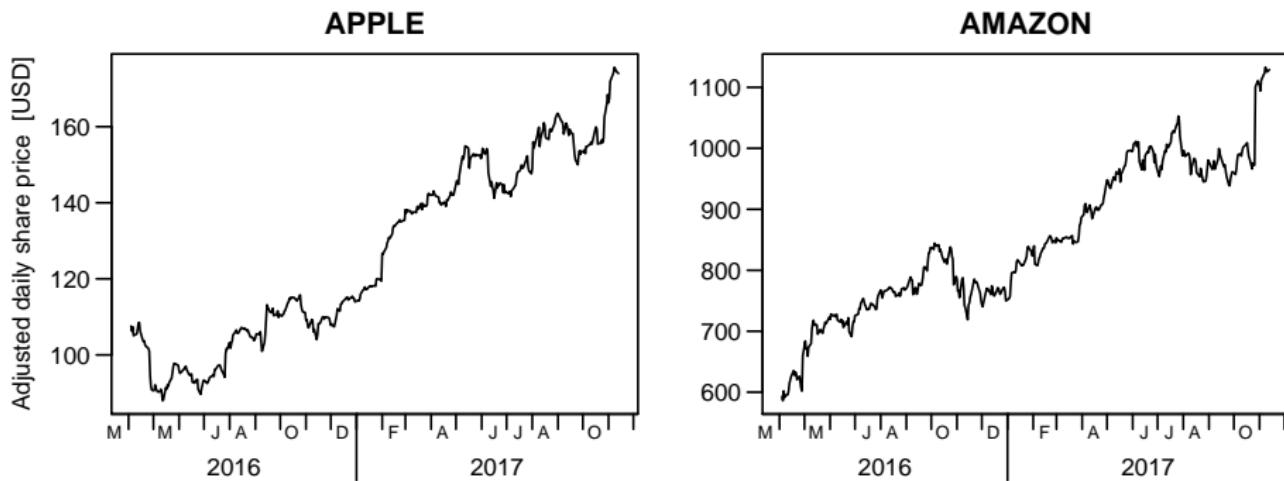
```
stocks <- read.table("data/stocks.txt", header=T, as.is=T)
stocks>Date <- as.Date(stocks>Date)
par(mfrow=c(2,3), mar=c(2,4,1,1), mgp=c(2.5,0.7,0), oma=c(0,0,2,0), las=1)
for(i in 2:7) plot(stocks[,c(1,i)], type="l")
mtext("stocks this decade", line=0, outer=TRUE)
```

stocks this decade



# For loops: multipanel graphics: beautifuller

```
par(mfrow=c(1,2), mar=c(4,4,2,0.1), mgp=c(2.5,0.7,0), cex=0.7, las=1)
for(i in 2:3)
{
  plot(stocks[ stocks$Date>as.Date("2016-04-01") , c(1,i) ],
    main=colnames(stocks)[i], xaxt="n", type="l", xlab=" ",
    ylab=if(i==2) "Adjusted daily share price [USD]" else "")
  berryFunctions::monthAxis(1)
}
```



# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  - 7. Conditions & loops**
    - Conditions
    - For Loops
    - lapply, Rcpp**
    - Repeat + While Loops
    - Arrays
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

## for -> lapply I: basics

```
files <- dir("../rawdata", pattern="*.csv", full=TRUE)
```

```
#
```

## for -> lapply I: basics

```
files <- dir("../rawdata", pattern="*.csv", full=TRUE)

# bad and slow way:
dfs <- list() # initiate empty list
for(i in 1:length(files))
    dfs[[i]] <- read.csv(files[i], as.is=TRUE)

#
```

## for -> lapply I: basics

```
files <- dir("../rawdata", pattern="*.csv", full=TRUE)

# bad and slow way:
dfs <- list() # initiate empty list
for(i in 1:length(files))
    dfs[[i]] <- read.csv(files[i], as.is=TRUE)

# much better way: apply function to each file
dfs <- lapply(X=files, FUN=read.csv, as.is=TRUE)

#
```

## for -> lapply I: basics

```
files <- dir("../rawdata", pattern="*.csv", full=TRUE)

# bad and slow way:
dfs <- list() # initiate empty list
for(i in 1:length(files))
  dfs[[i]] <- read.csv(files[i], as.is=TRUE)

# much better way: apply function to each file
dfs <- lapply(X=files, FUN=read.csv, as.is=TRUE)

# single data.frame if all files have n columns:
df <- do.call(rbind, dfs)

#
```

## for -> lapply I: basics

```
files <- dir("../rawdata", pattern="*.csv", full=TRUE)

# bad and slow way:
dfs <- list() # initiate empty list
for(i in 1:length(files))
    dfs[[i]] <- read.csv(files[i], as.is=TRUE)

# much better way: apply function to each file
dfs <- lapply(X=files, FUN=read.csv, as.is=TRUE)

# single data.frame if all files have n columns:
df <- do.call(rbind, dfs)

# PS: much faster in this example could be
library("data.table")
dfs <- lapply(X=files, FUN=fread, sep=",")
df <- rbindlist(dfs)
```

for -> lapply II: progress bar, names, indexing etc

```
dfs <- lapply(X=files, FUN=read.csv, as.is=TRUE)
```

```
#
```

## for -> lapply II: progress bar, names, indexing etc

```
dfs <- lapply(X=files, FUN=read.csv, as.is=TRUE)

# progress bar with remaining time
library("pbapply")
dfs <- pblapply(X=files, FUN=read.csv, as.is=TRUE)

#
```

## for -> lapply II: progress bar, names, indexing etc

```
dfs <- lapply(X=files, FUN=read.csv, as.is=TRUE)

# progress bar with remaining time
library("pbapply")
dfs <- pblapply(X=files, FUN=read.csv, as.is=TRUE)

# nice additional stuff:
names(dfs) <- files
str(dfs, max.level=1)
dfs[[2]] # second list element

#
```

## for -> lapply II: progress bar, names, indexing etc

```
dfs <- lapply(X=files, FUN=read.csv, as.is=TRUE)

# progress bar with remaining time
library("pbapply")
dfs <- pblapply(X=files, FUN=read.csv, as.is=TRUE)

# nice additional stuff:
names(dfs) <- files
str(dfs, max.level=1)
dfs[[2]] # second list element

# get third column / fifth row from each df:
sapply(dfs, "[", , j=3)
sapply(dfs, "[", 5, )
```

# lapply exercise

## Exercise 22: lapply

- ① Get the average of each element in the built-in dataset `Harman23.cor` with `lapply`.
- ② Now use `sapply` instead. What changes?
- ③ BONUS: Do this again with a `for` loop.
- ④ Please create `max_n <- function(...) max(rnorm(1e7))`. Now run this 10 times with an `sapply` loop.
- ⑤ Display a progress bar along with it.
- ⑥ BONUS: From each element of `Harman74.cor`, get the first 5 values.

# lapply exercise solution |

```
lapply(Harman23.cor, mean)
```

```
## $cov  
## [1] 0.578875  
##  
## $center  
## [1] 0  
##  
## $n.obs  
## [1] 305
```

```
sapply(Harman23.cor, mean) # simplified into a vector
```

```
##          cov      center      n.obs  
## 0.578875 0.000000 305.000000
```

## lapply exercise solution II

```
harmeans <- rep(NA,length(Harman23.cor))
for(i in 1:length(Harman23.cor) )
  harmeans[i] <- mean(Harman23.cor[[i]])
names(harmeans) <- names(Harman23.cor)
harmeans

##          cov      center      n.obs
## 0.578875  0.000000 305.000000

max_n <- function(...) max(rnorm(1e5))
pbapply::pbsapply(1:10, max_n)

## [1] 4.624598 4.204675 5.079964 4.255254 4.401126
## [6] 4.625683 4.250512 4.160242 4.281924 4.644114
```

```
lapply(Harman74.cor, "[", 1:5)
```

## for -> lapply IIIa: parallel on multiple CPU cores

```
library("parallel") # already comes with R
# -> don't try to install.packages("parallel")
nc <- detectCores()-1

#
```

## for -> lapply IIIa: parallel on multiple CPU cores

```
library("parallel") # already comes with R
# -> don't try to install.packages("parallel")
nc <- detectCores()-1

# multicore parallel execution
dfs <-      lapply(X=files, FUN=read.csv, as.is=TRUE)
dfs <-      mclapply(X=files, FUN=read.csv, as.is=TRUE,
                      mc.cores=nc) # easy on linux

#
```

## for -> lapply IIIa: parallel on multiple CPU cores

```
library("parallel") # already comes with R
# -> don't try to install.packages("parallel")
nc <- detectCores()-1

# multicore parallel execution
dfs <-      lapply(X=files, FUN=read.csv, as.is=TRUE)
dfs <-      mclapply(X=files, FUN=read.csv, as.is=TRUE,
                      mc.cores=nc) # easy on linux

# more code needed on windows:
cl <- makePSOCKcluster(nc)
dfs <- parLapply(cl, X=files, fun=read.csv, as.is=TRUE)
stopCluster(cl)

#
```

## for -> lapply IIIa: parallel on multiple CPU cores

```
library("parallel") # already comes with R
# -> don't try to install.packages("parallel")
nc <- detectCores()-1

# multicore parallel execution
dfs <-      lapply(X=files, FUN=read.csv, as.is=TRUE)
dfs <-      mclapply(X=files, FUN=read.csv, as.is=TRUE,
                      mc.cores=nc) # easy on linux

# more code needed on windows:
cl <- makePSOCKcluster(nc)
dfs <- parLapply(cl, X=files, fun=read.csv, as.is=TRUE)
stopCluster(cl)

# sometimes needed before parLapply call:
clusterExport(cl, c("files", "otherObjects"))
clusterEvalQ(cl, library("somePackage"))
```

## for -> lapply IIIb: parallel AND progress bar

```
library(parallel)
library(pbapply) # (>2016-10-30)
cl <- makeCluster( detectCores()-1 )
myfun <- function(...) mean(rnorm(1e6))
clusterExport(cl, "myfun")
dfs1 <- pblapply(X=1:500, FUN=myfun) # 36 sec on 1 CPU core
dfs2 <- pblapply(X=1:500, FUN=myfun, cl=cl) # 10 s on 7
stopCluster(cl)

# selects windows or linux method for you!
# Peter Solymos and Zygmunt Zawadzki are awesome!
```

## for -> lapply IV: code timing

```
begintime <- Sys.time(); begintime  
parLapply(cl, X, fun)  
Sys.time() - begintime ; rm(cl, begintime)
```

## for -> lapply IV: code timing

```
begintime <- Sys.time(); begintime
parLapply(cl, X, fun)
Sys.time() - begintime ; rm(cl, begintime)

#install.packages("microbenchmark")
library(microbenchmark)
forbad <- function(n)  # fibonacci
{
  fibvals <- c(1,1)
  for (i in 3:n) fibvals[i] <- fibvals[i-1]+fibvals[i-2]
  fibvals
}
forgood <- function(n)
{
  fibvals <- rep(1,n)
  for (i in 3:n) fibvals[i] <- fibvals[i-1]+fibvals[i-2]
  fibvals
}
mb <- microbenchmark(forbad(2000), forgood(2000)) ; mb

## Unit: microseconds
##          expr      min       lq     mean   median      uq     max neval
##  forbad(2000) 711.676 736.8235 986.0070 778.085 874.7620 5340.731    100
##  forgood(2000) 264.131 268.1965 402.9358 273.015 279.4905 4951.312    100
```

## for -> C++ code: Rcpp

Loops are fast in C++, so outsource that part of your code into C. The package [Rcpp](#) will compile it for you and make it available as a normally accessible R function.

for -> C++ code: Rcpp

Loops are fast in C++, so outsource that part of your code into C. The package [Rcpp](#) will compile it for you and make it available as a normally accessible R function.

Start learning at <http://adv-r.had.co.nz/Rcpp.html>

## for -> C++ code: Rcpp

Loops are fast in C++, so outsource that part of your code into C. The package [Rcpp](#) will compile it for you and make it available as a normally accessible R function.

Start learning at <http://adv-r.had.co.nz/Rcpp.html>

Speed gain is highly variable, of course, but you might be able to reduce your computing time from 2 hours to 20 seconds!

## for -> C++ code: Rcpp

Loops are fast in C++, so outsource that part of your code into C. The package `Rcpp` will compile it for you and make it available as a normally accessible R function.

Start learning at <http://adv-r.had.co.nz/Rcpp.html>

Speed gain is highly variable, of course, but you might be able to reduce your computing time from 2 hours to 20 seconds!

To find the slow parts of your code, you perform a process called "profiling":

<https://www.r-bloggers.com/profiling-r-code>

<https://www.rdocumentation.org/packages/utils/topics/Rprof?>

<https://adv-r.had.co.nz/Profiling.html>

## for -> C++ code: Rcpp

Loops are fast in C++, so outsource that part of your code into C. The package `Rcpp` will compile it for you and make it available as a normally accessible R function.

Start learning at <http://adv-r.had.co.nz/Rcpp.html>

Speed gain is highly variable, of course, but you might be able to reduce your computing time from 2 hours to 20 seconds!

To find the slow parts of your code, you perform a process called "profiling":

<https://www.r-bloggers.com/profiling-r-code>

<https://www.rdocumentation.org/packages/utils/topics/Rprof?>

<https://adv-r.had.co.nz/Profiling.html>

New and very promising:

<https://blog.rstudio.org/2016/05/23/profiling-with-rstudio-and-profvis>

<https://support.rstudio.com/hc/en-us/articles/218221837-Profiling-with-RStudio>

# For loop: Fibonacci

The Fibonacci sequence consists of the following integers:

1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, 144, ...

The first two values are 1, the following values are always the sum of the previous two elements.

## Exercise 23: Fibonacci

- ① Create the first 100 numbers of the fibonacci sequence with a for loop
- ② BONUS 1: "forget" to create an empty vector of the correct length first, and compare the computational time of the two expressions.  
Hint: Use `install.packages("microbenchmark")`;  
`library(microbenchmark); mb <- microbenchmark(...)`
- ③ BONUS 2: Can you also do this with sapply?
- ④ BONUS 3: Find a fast fibonacci creator function online and compare the timings of that as well
- ⑤ BONUS 4: With the package Rcpp, write the for loop in C++ and compare the speed with the other variants

# Fibonacci solution

```
len <- 20
fn <- rep(1, len)
for(i in 3:len) fn[i] <- fn[i-1] + fn[i-2]
fn

## [1] 1 1 2 3 5 8 13
## [8] 21 34 55 89 144 233 377
## [15] 610 987 1597 2584 4181 6765
```

# Fibonacci solution II

```
#install.packages("microbenchmark")
library(microbenchmark)
#install.packages("numbers")
suppressWarnings(library(numbers))
forbad <- function(n)
  {fibvals <- c(1,1)
  for (i in 3:len) fibvals[i] <- fibvals[i-1]+fibvals[i-2]
  fibvals
  }
forgood <- function(n)
  {fibvals <- rep(1,n)
  for (i in 3:len) fibvals[i] <- fibvals[i-1]+fibvals[i-2]
  fibvals
  }
numfib <- function(n) suppressWarnings(fibonacci(n, seq=T))
applfib <- function(n)
  {fibvals <- rep(1,n)
  sapply(3:n, function(i) {fibvals[i]<-fibvals[i-1]+fibvals[i-2]})
  }
mb <- microbenchmark(numfib(2000), forbad(2000), forgood(2000), applfib(2000))
mb

## Unit: microseconds
##          expr      min       lq     mean    median       uq      max neval
##  numfib(2000) 386.103  404.7765  684.79159  437.0010  534.1300 18523.326   100
##  forbad(2000)   8.132   10.8425  56.71100  13.8545  20.1790  4137.209   100
##  forgood(2000)   6.325   7.2290  57.59042   9.3365  13.4025  4547.105   100
##  applfib(2000) 4687.151 4921.0120 10810.46794 5080.4825 5273.6850 119569.480   100
```

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations

## 7. Conditions & loops

Conditions

For Loops

lapply, Rcpp

Repeat + While Loops

Arrays

## 8. Functions & packages

## 9. Distributions

## 10. Statistics

## 11. Time series handling and analysis

## 12. Spatial data and GIS functionality

## 13. Final tasks

Course plan

# Repeat loops

Syntax: `repeat {expressions}`

`next` # jump to next iteration

`break` # leave the loop

# Repeat loops

Syntax: `repeat {expressions}`

`next` # jump to next iteration  
`break` # leave the loop

```
i <- 0 # Create object i with the value 0
repeat # keep repeating the following block of code
{
  i <- i + 1          # Codeblock begins
  if (i < 3)  print(i) # overwrite object i with new number
  if (i == 3)  break   # conditional output to console
}                      # conditional leaving of loop
# Codeblock ends

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

What is now the value of i?

# Repeat loops

Syntax: `repeat {expressions}`

`next` # jump to next iteration

`break` # leave the loop

```
i <- 0 # Create object i with the value 0
repeat # keep repeating the following block of code
{
  i <- i + 1          # Codeblock begins
  if (i < 3)  print(i) # overwrite object i with new number
  if (i == 3)  break   # conditional output to console
}                      # conditional leaving of loop
# Codeblock ends

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

What is now the value of i? i is now 3, because in the last iteration, 1 was added to 2 and the result saved as i.

# While loops

Syntax: `while(condition){expressions}`

The following code is executed as long as (while) i is lesser then 3:

```
i <- 0
while(i < 3) i <- i + 1
i

## [1] 3
```

repeat loops are hardly ever used in R, and while loops fairly seldom as well.

# Practice while loops

## Exercise 24: while loop

- ① Why is the character string `folder <- "path/to/file///"` unsuitable for reading in a file?
- ② With a while loop, make it suitable. Hints: `substring`, `nchar`
- ③ BONUS: enable your solution to deal with `\\" at the end as well. Hint: %in%.`

## Solution to exercise 24: while loop

folder <- "path/to/file///" would link to empty-named subfolders at the end

```
folder <- "path/to/file///"  
while(substring(folder, nchar(folder))=="/")  
    folder <- substring(folder, 1, nchar(folder)-1)  
  
while( substring(folder, nchar(folder)) %in% c("/", "\\\\") )  
    folder <- substring(folder, 1, nchar(folder)-1)
```

# The grand duel (<http://www.janko.at/Raetsel/Mathematik/095.a.htm>)

Sir Adam and Sir Bart confront each other in a duel with pistols. By good old English custom, the two shoot at each other alternately, until one of them dies. Being gentlemen, they decide that the worse shooter may fire the first shot. Sir Adam has a hit ratio of 30% and Sir Bert one of 40%. What are the chances of survival for each of the two?

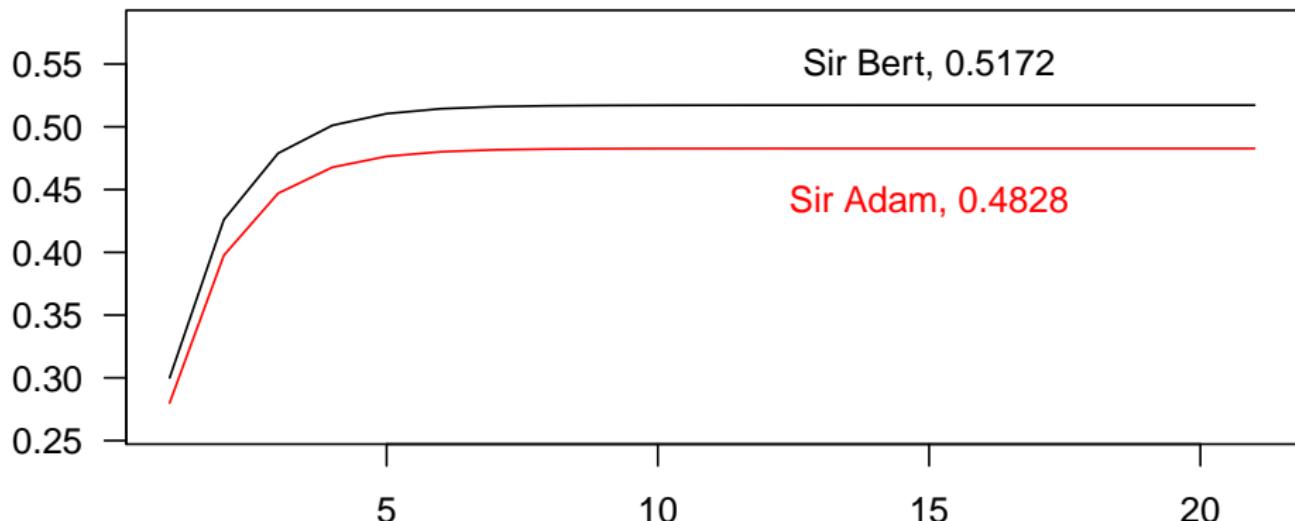
## Exercise 25: while loop duel

- ① BONUS 1: Derive the probabilities mathematically.
- ② Simulate the duel numerically. Hint: repeat the experiment `while` both are alive. In each step, let Sir Bert die with a 30% chance; if he survives, let Sir Adam die with a 40% chance.
- ③ BONUS 2: Write a function simulating the duel and returning the outcome.
- ④ Run the simulation 1000 (10k) times and assess Sir Bert's chance of survival.
- ⑤ BONUS 3: How many rounds are fought on average? (How is that quantity distributed?)

# The grand duel - resolved

```
plot(cumsum(sapply(0:20, function(i) 0.42^i*0.3)), type="l",
      ylim=c(0.26,0.58), main="cumulative death chance per round")
lines(cumsum(sapply(0:20, function(i) 0.42^i*0.7*0.4)), col=2)
```

**cumulative death chance per round**



# The grand duel - resolved

```
duel_simulation <- function()  
{  
  a <- b <- TRUE # TRUE = 1 = alive  
  n <- 0 # number of rounds  
  while(a&b)  
  {  
    if( runif(1)<0.3) b <- FALSE # = 0 = dead  
    if(b&runif(1)<0.4) a <- FALSE  
    n <- n+1  
  }  
  # return simulation result  
  c(a=a, b=b, n=n)  
}
```

# The grand duel - resolved

```
simu_res <- replicate(10000, duel_simulation())
# head(t(simu_res), 30)
apply(simu_res, 1, table) # / 10000 *100

## $a
##
##      0      1
## 4857 5143
##
## $b
##
##      0      1
## 5143 4857
##
## $n
##
##      1      2      3      4      5      6      7      8      9      10     13     19
## 5736 2437 1040  447   189    85    41    14     8      1      1      1
```

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations

## 7. Conditions & loops

Conditions

For Loops

lapply, Rcpp

Repeat + While Loops

Arrays

## 8. Functions & packages

## 9. Distributions

## 10. Statistics

## 11. Time series handling and analysis

## 12. Spatial data and GIS functionality

## 13. Final tasks

Course plan

# Read files in a loop, create an array

## Exercise 26: Create an array from files

- ① Download the `econ.zip` file (e.g. with `download.file`), then `unzip` it.
- ② Get a vector of filenames in the `econ` folder with `dir`. Remember to set the `full.names` argument.
- ③ Using `lapply`, read all the datasets into a list.
- ④ BONUS: With `basename` and `substr`, use the years in the filename as names of the list elements.
- ⑤ Convert each element of the list to a `?matrix`.
- ⑥ With `?berryFunctions::l2array`, convert the list to an array.
- ⑦ Examine it with `str`

## Array creation - solution

```
download.file(  
  "https://github.com/brry/course/raw/master/data/econ.zip",  
  destfile="econ.zip")  
unzip("econ.zip")
```

```
files <- dir("data/econ", full.names=TRUE)  
econ <- lapply(files, read.table)  
names(econ) <- substr(basename(files), 2,5)  
econ <- lapply(econ, as.matrix)  
econ <- berryFunctions::l2array(econ)  
str(econ)  
  
##  num [1:3, 1:3, 1:5] 25.2 2 2.5 321.4 32.1 ...  
##  - attr(*, "dimnames")=List of 3  
##    ..$ : chr [1:3] "DE" "CH" "GB"  
##    ..$ : chr [1:3] "Landwirtschaft" "Industrie" "Dienstleistung"  
##    ..$ : chr [1:5] "2012" "2013" "2014" "2015" ...
```

# Array subsetting and aggregation

```
some_array[x,y,z, ...]  
apply(some_array, MARGIN=c(1,3), FUN=mean)
```

MARGIN denotes the dimensions to keep (i.e. aggregate over the other dimensions)

## Exercise 27: subsets and aggregates from arrays

- ① Display all the values for Germany (by name). Extract the values in the second economic category (by number). Display a time series of the agricultural branch in Switzerland.
- ② Compute the median per category and country (aggregate over time). Compute the maximum of each year.
- ③ BONUS: Compute the max per year, excluding the industrial sector. If there are many items in the second dimension, how could you automatize the request for a particular entry (like "Industrie") without hard-coding the position? Hint: `str(some_array)` shows a useful attribute you'll need.

## Array subsetting - solution

```
econ["DE", ,]  
econ[, 2,]  
econ["CH", "Landwirtschaft", ]  
  
apply(econ, 1:2, median)  
apply(econ, 3, max)  
  
apply(econ[,-2,], 3, max)  
which(dimnames(econ)[[2]]=="Industrie")
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
    - Functions
    - Writing R packages
    - Debugging
    - Programming exercise
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Functions I

<http://r4ds.had.co.nz/functions.html>

# Functions I

<http://r4ds.had.co.nz/functions.html>

? "function"

# Functions I

<http://r4ds.had.co.nz/functions.html>

? "function"

Syntax:

```
Functionobjectname <- function(argument1, argument2, ...)  
                      {"DoSomething"}
```

# Functions I

<http://r4ds.had.co.nz/functions.html>

? "function"

Syntax:

```
Functionobjectname <- function(argument1, argument2, ...)  
                      {"DoSomething"}
```

```
myfunct <- function(grappig)  
           {plot(grappig, type="l"); return(grappig*7) }
```

# Functions I

<http://r4ds.had.co.nz/functions.html>

? "function"

Syntax:

```
Functionobjectname <- function(argument1, argument2, ...)  
                      {"DoSomething"}
```

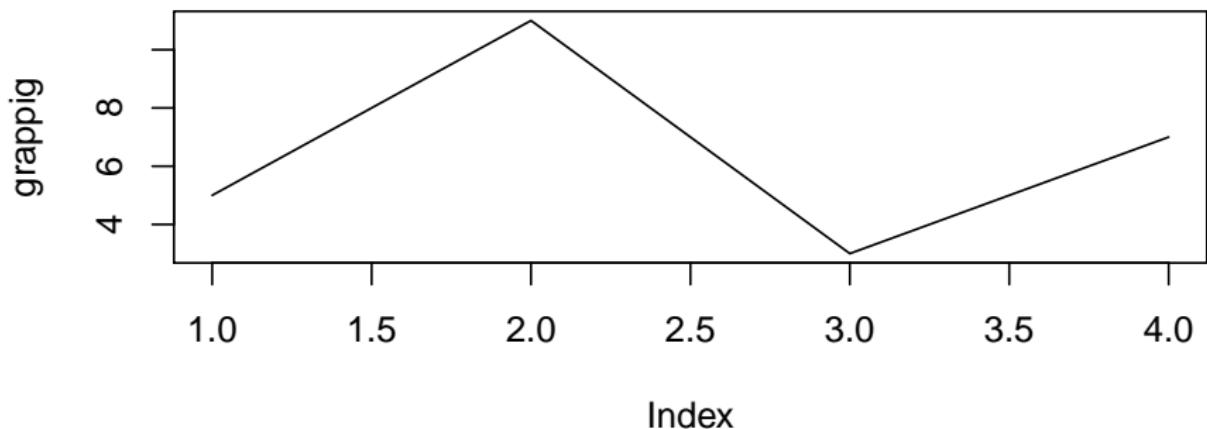
```
myfunct <- function(grappig)  
           {plot(grappig, type="l"); return(grappig*7) }
```

After **return()**ing, the execution of the function is terminated, so it should only be positioned at the end. It can also be left away, the last instruction ("expression") will then be returned.

## Functions II

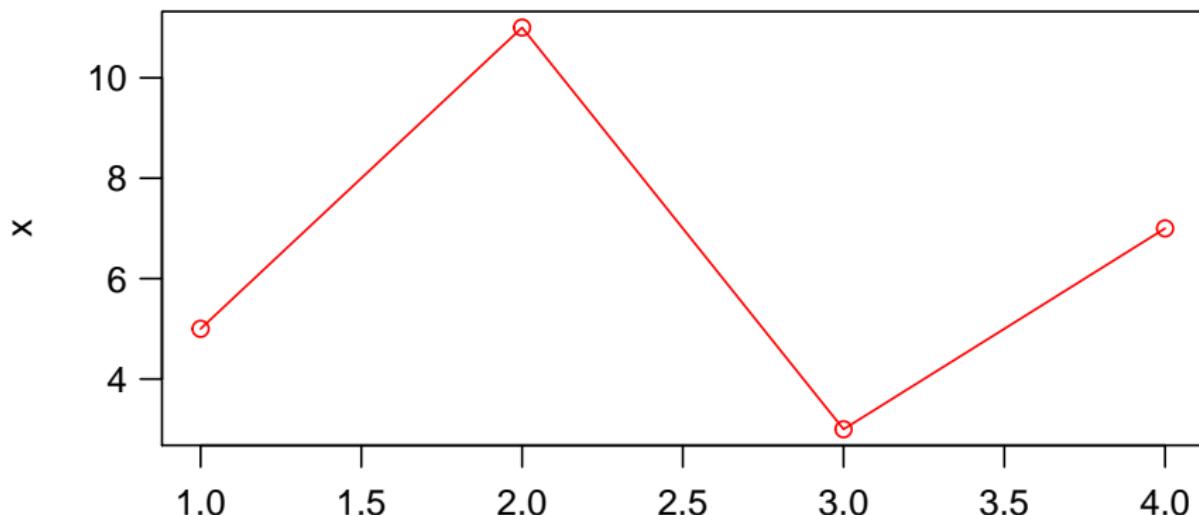
```
myfunct(    c(5,11,3,7)  )
```

```
## [1] 35 77 21 49
```



## Functions with more arguments

```
myfunct <- function(x, type="o", ...) plot(x, type=type, ...)  
# type="o" is now the default, thus used unless specified  
# The ellipsis (...) passes arguments to other functions  
myfunct( c(5,11,3,7) , col="red", las=1)
```



## Functions: example 1

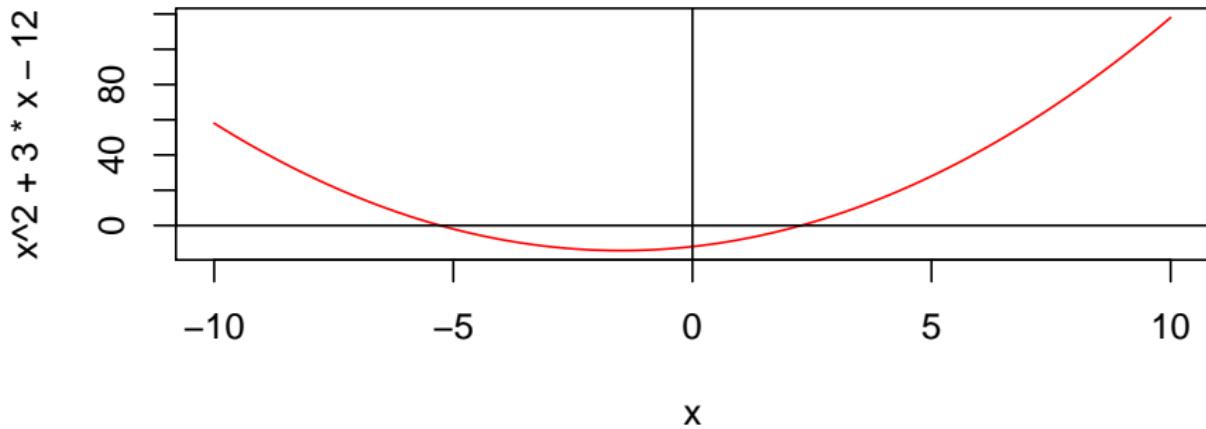
If you needed to find the zeros of quadratic functions very often, you could use

```
 pq <- function(p,q) #  $y = x^2 + px + q$ 
 {
   w <- sqrt( p^2 / 4 - q )
   c(-p/2-w, -p/2+w)
 } # End of function

pq(3, -12)
## [1] -5.2749  2.2749
```

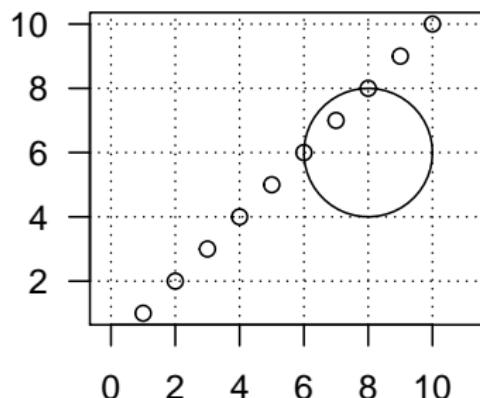
## Functions: example II

```
x <- seq(-10, 10, len=100)
plot(x, x^2 +3*x -12, type="l", col=2)
abline(h=0, v=0)
```



## Exercise: add circles with given radius

```
plot(1:10, asp=1) # aspect ratio y/x of graph range
grid(col=1) # the next part sould go into a function:
x <- 8 ; y <- 6 ; r <-2
p <- seq(0, 2*pi, len=50)
cx <- x+r*cos(p) ; cy <- y+r*sin(p)
polygon(cx, cy)
```



# Time to practice programming

## Exercise 28: Writing functions

Write a function that

- ① - draws a circle with a certain radius at user-specified locations of an existing plot (see last slide).
- ② - uses ellipsis to allow the user to customize the appearance
- ③ - checks all the arguments and gives useful warnings if the wrong type of input is provided
- ④ - has useful explanations for each argument (documentation)
- ⑤ - has readable indentation, spacing and comments explaining the code
- ⑥ Now let your neighbor use it without explaining how it is to be used (this should be inferred from the code and comments!)
- ⑦ Use your neighbor's function with a vector to draw several circles at once. (unintended use?) What happens?

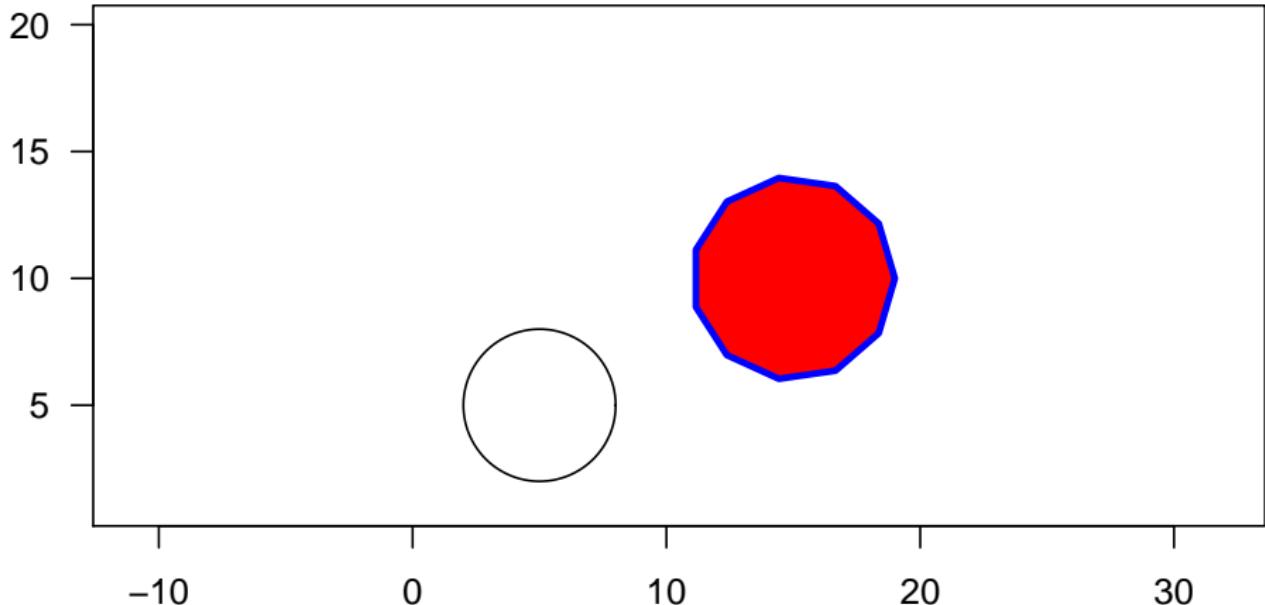
```

# Small helper function drawing circles into existing graphics
# Berry Boessenkool, berry-b@gmx.de, 2012
circle <- function(
  x, # x-coordinate of points, numeric value of length 1
  y, # ditto for y
  r, # radius of the circle, in the graphic's units
  locnum=100, # number of points on circle (more means smoother but slower)
  ...) # Further Arguments passed to polygon, like col, border, lwd
{
  # input checking - only one circle can be drawn:
  if(length(x) >1 | length(y) >1 | length(r) >1 | length(locnum) >1)
  {
    warning("Only the first element of the vectors is used.")
    x <- x[1]; y <- y[1]; r <- r[1]; locnum <- locnum[1]
  }
  # input checking - is every value numeric?
  if(!is.numeric(x)) stop("x must be numeric, not ", class(x))
  if(!is.numeric(y)) stop("y must be numeric, not ", class(y))
  if(!is.numeric(r)) stop("r must be numeric, not ", class(r))
  # prepare circle line coordinates:
  cx <- x+r*cos( seq(0,2*pi,len=locnum) )
  cy <- y+r*sin( seq(0,2*pi,len=locnum) )
  polygon(cx, cy, ...) # actually draw it
}
# Note: if circles look like ellipsis, use plot(... asp=1)

```

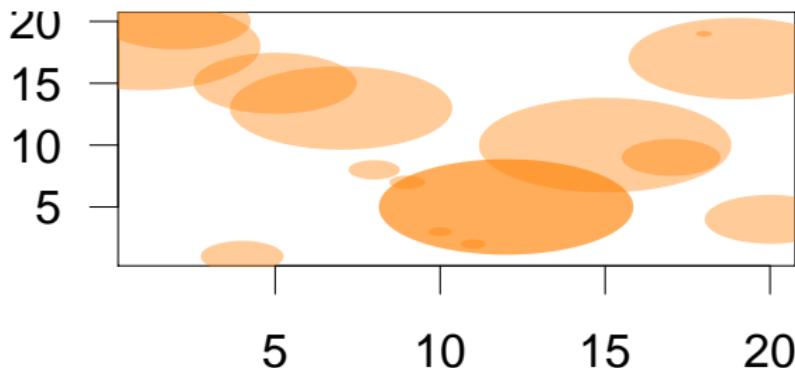
## Solution for exercise 28 II: functions

```
plot(1:20, type="n", asp=1, cex=2)
circle(5,5, r=3)
circle(15,10, r=4, locnum=12, col=2, border=4, lwd=3)
```



## Solution for exercise 28 III: functions

```
# can not be vectorized:  
x <- sample(1:20, 15) ; y <- sample(1:20, 15) ; r <- runif(20)*4  
circle(x,y,r, col=rgb(1,0.5,0,alpha=0.4), border=NA)  
  
## Warning in circle(x, y, r, col = rgb(1, 0.5, 0, alpha = 0.4), border = NA):  
Only the first element of the vectors is used.  
  
for(i in 1:15) circle(x[i],y[i],r[i], col=rgb(1,0.5,0,alpha=0.4), border=NA)
```



# Coding elegance I

Suppose you have two related vectors, perhaps as a result from `approx`:

a: 1 2 3 4 5 ...

b: 1 2 3 4 5 6 7 8 9 ...

# Coding elegance I

Suppose you have two related vectors, perhaps as a result from `approx`:

a: 1 2 3 4 5 ...

b: 1 2 3 4 5 6 7 8 9 ...

For any element  $i$  in a, you want to select the corresponding surrounding values of b.

# Coding elegance I

Suppose you have two related vectors, perhaps as a result from `approx`:

a: 1 2 3 4 5 ...

b: 1 2 3 4 5 6 7 8 9 ...

For any element  $i$  in a, you want to select the corresponding surrounding values of b. Example: if  $i=3$ , you want to select the elements 4,5 and 6.

# Coding elegance I

Suppose you have two related vectors, perhaps as a result from `approx`:

a: 1 2 3 4 5 ...

b: 1 2 3 4 5 6 7 8 9 ...

For any element  $i$  in a, you want to select the corresponding surrounding values of b. Example: if  $i=3$ , you want to select the elements 4,5 and 6. However, at the border, you cannot select with negative indices.

# Coding elegance I

Suppose you have two related vectors, perhaps as a result from `approx`:

a: 1 2 3 4 5 ...

b: 1 2 3 4 5 6 7 8 9 ...

For any element  $i$  in a, you want to select the corresponding surrounding values of b. Example: if  $i=3$ , you want to select the elements 4,5 and 6. However, at the border, you cannot select with negative indices.

## Exercise 29: Elegant Coding: surrounding indices

- ① Write a function using conditionals that performs the task described above.
- ② BONUS: If it's not a "one-liner", try to find a different approach.

## Coding elegance II

Here is how to write this with conditional statements:

## Coding elegance II

Here is how to write this with conditional statements:

```
index_1 <- function(i)
{
  if(i == 1)                      1 : 2
  else if(i == length(a)) ((i-1)*2) : (i*2-1)
  else                            ((i-1)*2) : (i*2)
}
```

## Coding elegance II

Here is how to write this with conditional statements:

```
index_1 <- function(i)
{
  if(i == 1)                      1 : 2
  else if(i == length(a)) ((i-1)*2) : (i*2-1)
  else                            ((i-1)*2) : (i*2)
}
```

Here's the elegant way:

```
index_2 <- function(i)
  pmax((i-1)*2, 1) : pmin(i*2, length(a)*2-1)
```

# Coding elegance III

Both functions yield the same result:

```
a <- 1:5  
index_1(1); index_2(1)
```

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

```
index_1(3); index_2(3)
```

```
## [1] 4 5 6  
## [1] 4 5 6
```

```
index_1(5); index_2(5)
```

```
## [1] 8 9  
## [1] 8 9
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. **Functions & packages**
    - Functions
    - Writing R packages
    - Debugging
    - Programming exercise
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Why you should write R packages

# Why you should write R packages

- Collect your own functions in one place

# Why you should write R packages

- Collect your own functions in one place
- Combine functions and documentation in the right way

# Why you should write R packages

- Collect your own functions in one place
- Combine functions and documentation in the right way
- Share code with others

# Why you should write R packages

- Collect your own functions in one place
- Combine functions and documentation in the right way
- Share code with others
- Make your research reproducible !

# Why you should write R packages

- Collect your own functions in one place
- Combine functions and documentation in the right way
- Share code with others
- Make your research reproducible !

Good introductions linked here: [github.com/brry/misc](https://github.com/brry/misc)

# Package development: devtools + github

We'll create a cat package together following the instructions in [packdev.R](#) (*rightclick Raw, save as*).

Version control with git:

- Once only: Create a github account (free), set up git on your computer, connect it to Rstudio. This is all very unfun and tedious, but at least there is a good guide that also includes a first test of the system: [r-bloggers.com/rstudio-and-github](http://r-bloggers.com/rstudio-and-github). See also [kbroman.org/github\\_tutorial](http://kbroman.org/github_tutorial) for a good intro.
- [github.com](https://github.com), log in, on "+" in top-right select "New repository", initialize with readme!
- On the green field "clone or download": copy link
- Rstudio - file - new project - version control - git: paste url and create
- `devtools::setup()` instead of `create`, choose option number to Overwrite 'yourPackage.Rproj'
- Work. Commit. Push. Repeat.

# Package exercise

## Exercise 30: create LSC package (solutions on next 2 slides)

- ① Create a package for linear storage cascade functions (lsc)
- ② BONUS: Use git from the beginning, commit often!
- ③ Check the package, correct description file and check again
- ④ Now add the functions in [lsc\\_functions.zip](#) to the R folder
- ⑤ Done? Help your peers.
- ⑥ Document the nse function (we'll do this one together)
- ⑦ Create datasets within the package (following instructions in data.R)
- ⑧ Document the rmse function
- ⑨ Check and install the package

## Solution for exercise 30: create LSC package

```
library(devtools)
create("lsc") # or devtools::setup() if using git
document() # calls roxygen2::roxygenise()
check() # will also update documentation
install() # install the package locally
load_all(".") # CTRL + SHIFT + L - faster than install!
```

# Solution for exercise 30: Documenting functions

```
#' @title      Nash-Sutcliffe efficiency
#' @description Nash-Sutcliffe efficiency as a measure of goodness of fit.
#'
#'           Removes incomplete observations with a warning.
#'
#' @return      Single numerical value
#'
#' @export
#'
#' @seealso     \code{\link{rmse}}
#'
#' @references based on eval.NSeff in RHydro Package
#'
#'           \url{https://r-forge.r-project.org/R/?group_id=411}
#'
#' @examples
#'
#' x <- rnorm(20) ; y <- 2*x + rnorm(20)
#' x[2:4] <- NA
#' nse(x,y)
#'
#'
#' @param obs Numerical vector with observed values
#' @param sim Simulated values (Num vector with same length as \code{obs})
#'
nse <- function(obs, sim)
{
  if(!(is.vector(obs) & is.vector(sim))) stop("Input is not a vector.")
  if(length(obs) != length(sim)) stop("Vectors are not of equal length.")
  # [...]
  1 - ( sum((obs - sim)^2) / sum((obs - mean(obs))^2) )
}
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. **Functions & packages**
    - Functions
    - Writing R packages
    - Debugging
    - Programming exercise
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Debugging

- Your code throws an error. You didn't call the mentioned function. Obviously, your code calls some function calling some function calling some function calling [you get the idea] which in the end creates an error. To trace back this path, you can use `traceback()`.

# Debugging

- Your code throws an error. You didn't call the mentioned function. Obviously, your code calls some function calling some function calling some function calling [you get the idea] which in the end creates an error. To trace back this path, you can use  `traceback()`.
- Now that you know where the error originates from, you set `options(error=recover)`. You run your code again, but this time R waits at the level creating an error. You examine the environment within the function, play around with the objects and internal function code, until the bug has been fixed. You have just debugged a function.

# Debugging

- Your code throws an error. You didn't call the mentioned function. Obviously, your code calls some function calling some function calling some function calling [you get the idea] which in the end creates an error. To trace back this path, you can use  `traceback()`.
- Now that you know where the error originates from, you set  `options(error=recover)`. You run your code again, but this time R waits at the level creating an error. You examine the environment within the function, play around with the objects and internal function code, until the bug has been fixed. You have just debugged a function.
- You want to step into the function you are developing at a specific point. You add  `browser()` at that point of the code. You want to go line by line in one specific function. You set  `debug(thatFunction)`.

# Debugging

- Your code throws an error. You didn't call the mentioned function. Obviously, your code calls some function calling some function calling some function calling [you get the idea] which in the end creates an error. To trace back this path, you can use  `traceback()`.
- Now that you know where the error originates from, you set `options(error=recover)`. You run your code again, but this time R waits at the level creating an error. You examine the environment within the function, play around with the objects and internal function code, until the bug has been fixed. You have just debugged a function.
- You want to step into the function you are developing at a specific point. You add  `browser()` at that point of the code. You want to go line by line in one specific function. You set  `debug(thatFunction)`.
- You want to learn about lexical scoping (Where does R find variables?).  
<http://trestletech.com/2013/04/package-wide-variables-cache-in-r-package/>  
<http://adv-r.had.co.nz/Environments.html>

# Debugging: useful functions

## Debugging: useful functions

```
source("projectFuns.R")    execute complete file
```

## Debugging: useful functions

<code>source("projectFuns.R")</code>	execute complete file
<code>traceback()</code>	find error source in sequence of function calls

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0
browser()	go into function environment: <code>n</code> , <code>s</code> , <code>f</code> , <code>c</code> , <code>Q</code>

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0
browser()	go into function environment: <code>n</code> , <code>s</code> , <code>f</code> , <code>c</code> , <code>Q</code>
options(error=recover)	open interactive session where error occurred

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0
browser()	go into function environment: <code>n</code> , <code>s</code> , <code>f</code> , <code>c</code> , <code>Q</code>
<code>options(error=recover)</code>	open interactive session where error occurred
debug(funct)	toggle linewise function execution

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0
browser()	go into function environment: <code>n</code> , <code>s</code> , <code>f</code> , <code>c</code> , <code>Q</code>
<code>options(error=recover)</code>	open interactive session where error occurred
debug(funct)	toggle linewise function execution
undebug(funct)	after calling and fixing funct

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0
browser()	go into function environment: n, s, f, c, Q
options(error=recover)	open interactive session where error occurred
debug(funct)	toggle linewise function execution
undebug(funct)	after calling and fixing funct

```
if(length(input)>1) stop("length must be 1, not ", length(input))
```

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0
browser()	go into function environment: n, s, f, c, Q
options(error=recover)	open interactive session where error occurred
debug(funct)	toggle linewise function execution
undebug(funct)	after calling and fixing funct

```
if(length(input)>1) stop("length must be 1, not ", length(input))
```

stop: Interrupts function execution and gives error

warning: continues but gives warning

message: to inform instead of worry the user

# Debugging: useful functions

source("projectFuns.R")	execute complete file
traceback()	find error source in sequence of function calls
options(warn=2)	warnings to error. default 0
browser()	go into function environment: n, s, f, c, Q
options(error=recover)	open interactive session where error occurred
debug(funct)	toggle linewise function execution
undebug(funct)	after calling and fixing funct

```
if(length(input)>1) stop("length must be 1, not ", length(input))
```

stop: Interrupts function execution and gives error

warning: continues but gives warning

message: to inform instead of worry the user

R. Peng (2002): Interactive Debugging Tools in R

D. Murdoch (2010): Debugging in R

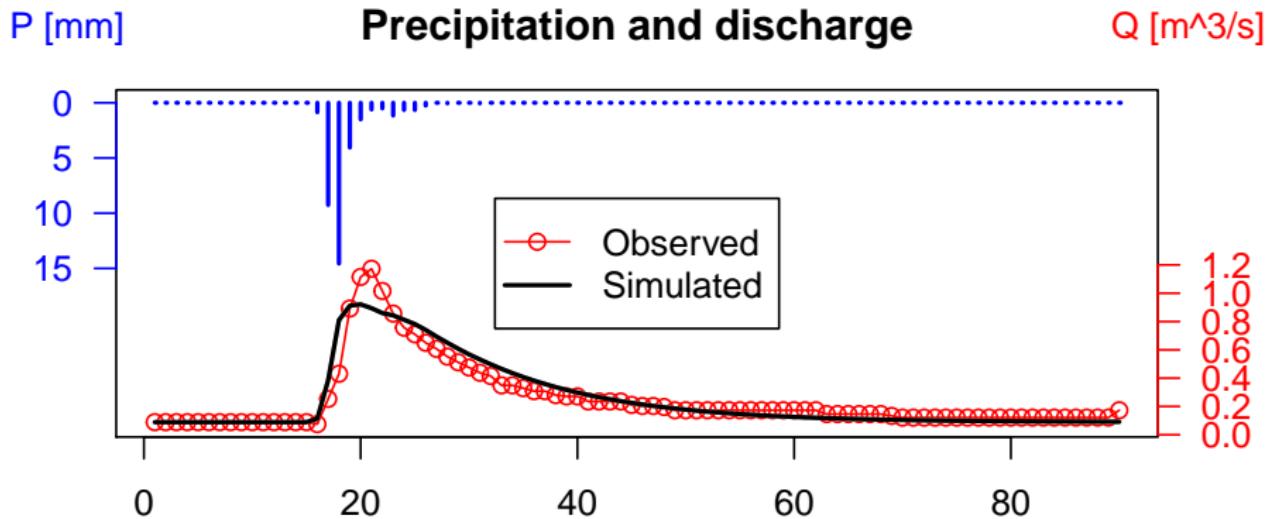
H. Wickham (2015): Advanced R: debugging

Example: Pete Werner Blog Post (2013)

Practice debugging, `lsc_functions.R` (*rightclick Raw, save as*)

## Exercise 31: Debugging

- ① Load your package and the datasets. Correct the functions until `lsc(calib$P, calib$Q, area=1.6)` returns the result below.
  - ② BONUS: commit each change to git.



## Solution for exercise 31: Debugging

- stupid error you can easily remove - traceback - find location of error - lsc#73 - just comment it out
- harder to find but still stupid - traceback - nse#11 - ditto
- Error in plot: need finite 'ylim' value -  
debug/browser/options(error=recover) - lsc#105 - NAs in Q -  
range(Q, na.rm=TRUE) - also in other applicable locations
- There were 50 or more warnings - come from rmse being called  
in optimization - add argument quietNA (or similar) to lsc that is  
passed to rmse in lsc#79

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. **Functions & packages**
    - Functions
    - Writing R packages
    - Debugging
    - Programming exercise**
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Chapter practice: sample size dependency

## Exercise 32: Programming part 1 - functions

- ① Write a function (#1) that computes the backtransformed arithmetic average of logtransformed values. It should `stop` with a useful error message if there are `any` negative values in the input vector.
- ② Write a second function (#2) taking a vector of values that returns the mean, sd, coefficient of variation and the output of function #1.
- ③ It should return these values with names as in `c(a=5, b=-3)`.
- ④ If the average is zero, it should inform the user that the CV cannot be computed. Remember to check only rounded values for equality.
- ⑤ Test both functions with `c(88, 31, 73, 21, 08, 52, 66, 48)`, with `c(4, 7, 5, 6, 4, 5, 3, NA, 6, 4, 5, 6, 8, 5)` and with `c(0, 0, 0)`.
- ⑥ Write a function (#3) taking `n`, `mean` and `sd` that
  - breaks if `n` is not a single value
  - draws `n` random numbers from the normal distribution with `mean` and `sd`
  - returns the output of function #2 of that sample.

# Programming practice solutions part 1 - functions 1

backtransformed arithmetic average of logtransformed values

```
logmean <- function(x)
{
  if(any(x<0)) stop("x has ", sum(x<0), " negative values.")
  lx <- log10(x)
  10^mean(lx)
}

logmean(c(88,31,73,21,08,52,66,48))

## [1] 39.14336

#logmean(c(4,7,5,6,4,5,3,NA,6,4,5,6,8,5))
logmean(c(0,0,0))

## [1] 0
```

# Programming practice solutions part 1 - functions 2

## Statistical properties of a vector

```
meanvar <- function(vals)
{
  m <- mean(vals) # average
  s <- sd(vals) # root of variation
  if(round(m)==0) stop("mean is 0, CV cannot be computed")
  c(mean=m, sd=s, CV=s/m, logmean=logmean(vals))
}
```

```
meanvar(c(88,31,73,21,08,52,66,48))
```

```
##           mean          sd          CV      logmean
## 48.3750000 27.2078639  0.5624365 39.1433623
```

```
#meanvar(c(4,7,5,6,4,5,3,NA,6,4,5,6,8,5))
#meanvar(c(0,0,0))
```

# Programming practice solutions part 1 - functions 3

## Dependency on sample size

```
dep_n <- function(n, mean, sd)
{
  if(length(n) != 1) stop("n must be a single value, not ",
                         length(n))
}
```

*# draw n numbers randomly from normal distribution:*

```
v <- rnorm(n, mean=mean, sd)
```

```
meanvar(v)
```

```
}
```

```
dep_n(20, 70, 5)
```

##	mean	sd	CV	logmean
##	70.95960010	6.56314211	0.09249125	70.65816749

```
#
```

# Chapter practice: sample size dependency

## Exercise 33: Programming part 2 - debugging

- ① Test function #3 with `n=20:30, mean=70, sd=5`
  - ② What happens if you use it with `n=NA, mean=70, sd=5?`
  - ③ After you call it with `n=20, mean=-70, sd=5`, run `traceback()`  
Explain the result.
  - ④ What is the result of `ls.str()`?
  - ⑤ Set your upper level function #3 to `debugging mode`. Now call it again. What happens? What is now the result of `ls.str()`?
- 
- With `enter`, you can now execute the function line by line and observe how objects (variables) in functions change over execution time, thus making it easier to find the source of an error.
  - Don't forget to `undebug` your function again.
  - If you only want to step into the function at a certain point, write `browser()` at that position, redefine the function and call it again.

## Programming practice solutions part 2 - debugging

```
dep_n(20, -70, 5)
```

```
## Error in logmean(vals): x has 20 negative values.
```

```
 traceback()
```

```
4: stop("x has ", sum(x < 0), " negative values.") at #3
3: logmean(vals) at #6
2: meanvar(v) at #5
1: dep_n(20, -70, 5)
```

In debugging mode, you step into the function when it is called (visible by `Browse[2]>` in the console). `ls.str()` now returns the structures of the objects defined within the function environment.

# Chapter practice: sample size dependency

## Exercise 34: Programming part 3 - loops for computation

For this task, use `n <- rep(1:50, each=100)`

- ① Create an empty `data.frame` with the column names according to the output of function #3.
- ② With a for loop, fill each row with the result of the n-corresponding call to function #3.
- ③ Obtain the computing time by either wrapping it in `system.time` or calling `Sys.time()` before and afterwards and computing the `difftime`.
- ④ Time the usage of `sapply` for the same task. How many times is it faster? What does that imply for a 5-hour for loop computation?
- ⑤ BONUS: Get uncertainties on the computing time estimation with `microbenchmark` (You might want to reduce the length of n for that).

## Programming practice solutions part 3 - loops computation

```
n <- rep(1:50, each=100)
results_for <- data.frame(mean=NA, sd=NA, CV=NA, logmean=NA)
system.time( for(i in 1:length(n))
{results_for[i, ] <- dep_n(n[i], 70,5)})

##      user    system elapsed
##      2.41     0.00    2.68

tail(results_for, 5)

##           mean        sd        CV   logmean
## 4996 70.07647 4.851932 0.06923768 69.90943
## 4997 70.30274 4.788358 0.06811055 70.14381
## 4998 69.69953 5.420097 0.07776375 69.49969
## 4999 69.35405 4.780247 0.06892527 69.19216
## 5000 68.47572 4.488871 0.06555420 68.32956
```

## Programming practice solutions part 3 - loops computation

```
system.time(res_lapply <- sapply(n, dep_n, mean=70, sd=5) )  
  
##      user    system elapsed  
##     0.26      0.00     0.47
```

In this case with (only) 5000 iterations, lapply is ca 7 times faster than the badly initiated for loop. 5 hrs of computation would reduce to 40 minutes. Notice you get similar to lapply speed results for properly initiated data.frames with the correct dimension.

For more speed hints, read [win-vector.com](http://win-vector.com) efficient accumulation in R.  
Remember: your code is much cleaner with lapply:

```
results <- matrix(NA, ncol=4, nrow=length(n2))  
for(i in 1:length(n2)) results[i, ] <- dep_n(n2[i], 70, 5)  
  
results <- sapply(n2, dep_n, mean=70, sd=5)
```

# Programming practice solutions part 3 - loops computation

```
n2 <- rep(1:50, each=10)

dep_n_for <- function()
{results <- data.frame(mean=NA, sd=NA, CV=NA, logmean=NA)
 for(i in 1:length(n2)) results[i, ] <- dep_n(n2[i], 70,5)}

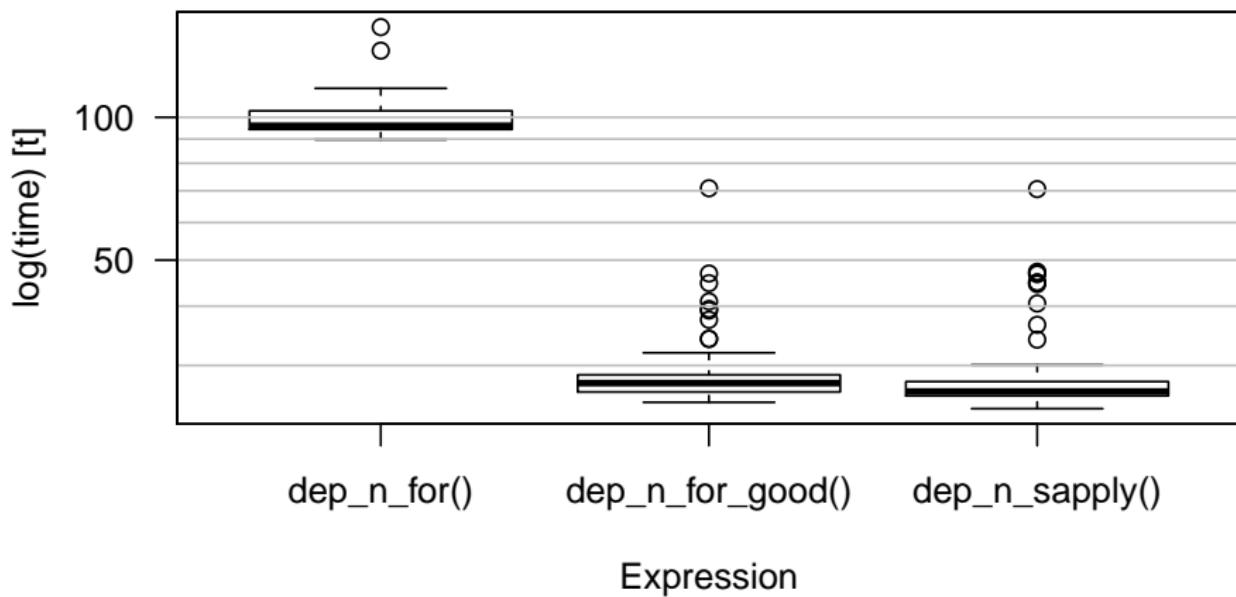
dep_n_for_good <- function()
{results <- matrix(NA, ncol=4, nrow=length(n2))
 for(i in 1:length(n2)) results[i, ] <- dep_n(n2[i], 70,5)}

dep_n_sapply <- function()
{results <- sapply(n2, dep_n, mean=70, sd=5)}

mb <- microbenchmark(dep_n_for(), dep_n_for_good(),
                      dep_n_sapply())
save(mb, file="data(mb.Rdata")
```

# Programming practice solutions part 3 - loops computation

```
load("data/mb.Rdata"); library(microbenchmark)
boxplot(mb, yaxt="n"); berryFunctions::logAxis(2)
```



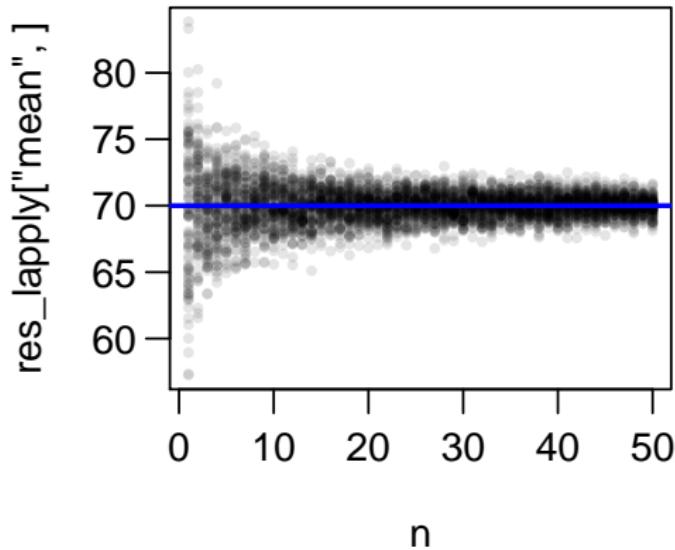
# Chapter practice: sample size dependency

## Exercise 35: Programming part 4 - loops for plotting

- ① Plot the mean estimate over sample size, using fully colored but semitransparent dots (see `rgb` or `berryFunctions::addAlpha`).
- ② Add a nicely visible horizontal line at the population mean. (we drew random samples from some defined distribution, which is the general population whose parameters are usually unknown in real life).
- ③ BONUS: add lines limiting the 90% confidence region.
- ④ Set up a multipanel plot to automatically hold a figure for each of the rows of the previous results (mean, sd, CV, logmean). To automatically dimension the panels, you could use `berryFunctions::panelDim(n)`
- ⑤ With a for loop, fill each of the panels and add nice annotations.

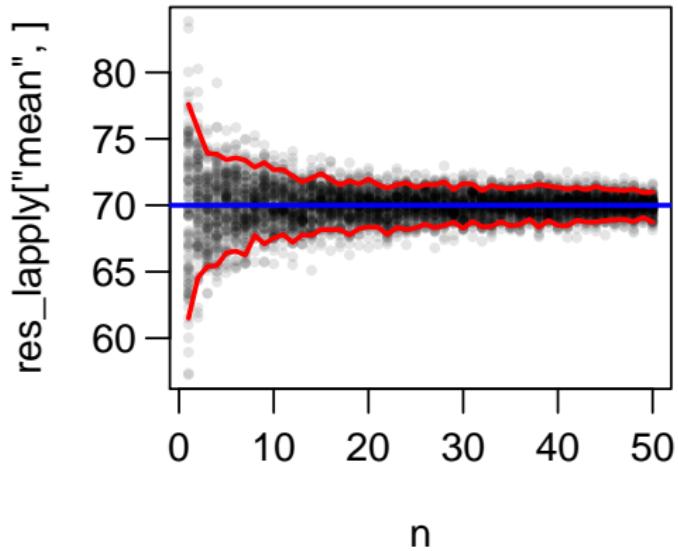
# Programming practice solutions part 4 - loops plotting

```
transp <- rgb(0,0,0, alpha=0.1)
plot(n, res_lapply["mean", ], pch=16, col=transp, cex=0.6)
abline(h=70, col=4, lwd=2)
```



# Programming practice solutions part 4 - loops plotting

```
conf <- tapply(res_lapply["mean", ], n, quantile, probs=c(0.05, 0.95))
conf <- sapply(conf, I)
conf_x <- as.numeric(colnames(conf))
for(i in 1:2) lines(conf_x, conf[i, ], col=2, lwd=2)
```



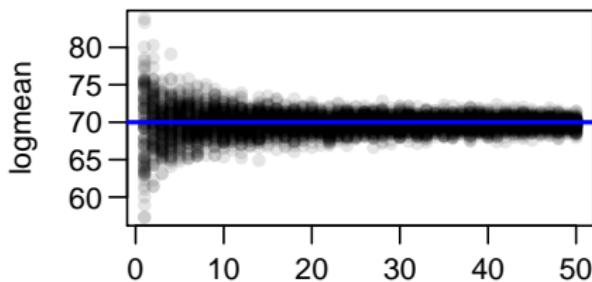
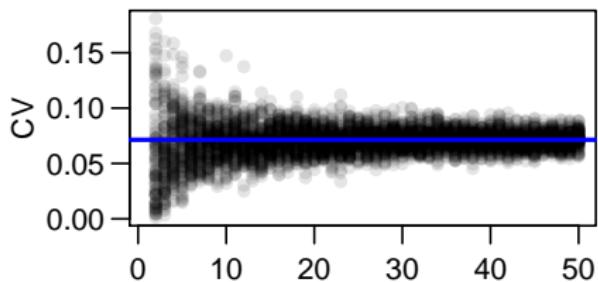
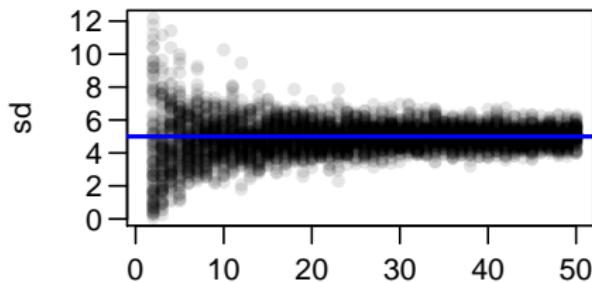
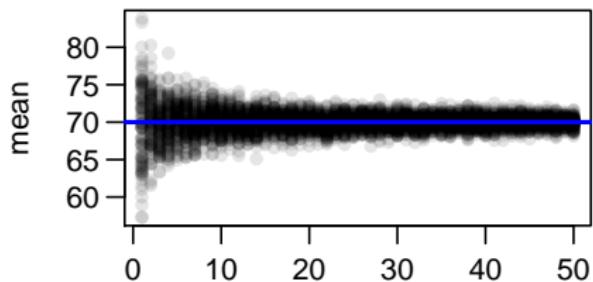
## Programming practice solutions part 4 - loops plotting

```
library(berryFunctions) # for panelDim

par(mfrow=panelDim(nrow(res_lapply)), mar=c(2,4,1,1),
    mgp=c(2.5,0.7,0), las=1 )

truevals <- c(mean=70, sd=5, CV=5/70, logmean=70)
for(nm in rownames(res_lapply))
{
  plot(n, res_lapply[nm, ], ylab=nm, pch=16, col=transp)
  abline(h=truevals[nm], col=4, lwd=2)
  if(nm=="sd") points(60,2, col=2)
}
```

# Programming practice solutions part 4 - loops plotting



# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. **Distributions**
    - Exploratory Data Analysis (EDA)
      - Normal distribution
      - Other distribution functions
      - Beta distribution
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# A quick example I

```
grades <- c(3.0, 2.3, 5, 1.3, 4, 1.7)
grades           # Show the object

## [1] 3.0 2.3 5.0 1.3 4.0 1.7

mode(grades)    # numeric

## [1] "numeric"

sum(grades)     # sum of all values

## [1] 17.3

diff(grades)    # difference from one value to the next

## [1] -0.7  2.7 -3.7  2.7 -2.3
```

```
mean(grades)    # Average  
## [1] 2.883333  
  
median(grades) # median*  
## [1] 2.65  
  
var(grades)     # variance  
## [1] 1.997667  
  
sd(grades)      # Standard deviation  
## [1] 1.413388
```

\* If you sort all the numbers ascendingly, the median is in the middle. Thus, it is not as sensitive to outliers as the mean. We'll get back to this later on.

# The dataset we'll be looking at



sources: [mirlab.org](http://mirlab.org), [desirableplants.com](http://desirableplants.com), [3.bp.blogspot.com](http://3.bp.blogspot.com)  
[github.com/brry/course](http://github.com/brry/course)

# iris data set

```
head(iris)
```

```
##   Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1          5.1         3.5         1.4         0.2  setosa
## 2          4.9         3.0         1.4         0.2  setosa
## 3          4.7         3.2         1.3         0.2  setosa
## 4          4.6         3.1         1.5         0.2  setosa
## 5          5.0         3.6         1.4         0.2  setosa
## 6          5.4         3.9         1.7         0.4  setosa
```

## Exercise 36: EDA - Exploratory Data Analysis

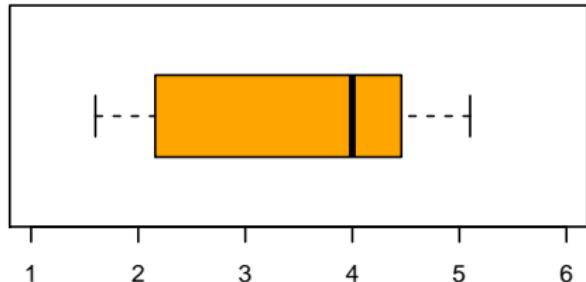
- ① Look at `iris` with `summary`.
- ② Assign `iris$Sepal.Length` to an object with a short name.
- ③ Calculate `min` and `max`, the `median` and the median absolute deviation (`mad`), the quartiles and octiles (via `quantile`), the fourthspread (`IQR`) and `diff(range)` of the `iris` values.
- ④ With `par(mfrow=c(2,1))`, set the graphic device to contain two graphics.
- ⑤ Draw a horizontal `boxplot` in the upper window and compare it with the statistics calculated above. Interpret them.
- ⑥ What happens when the argument `notch` is set to true?
- ⑦ Draw a `histogram` with the bars filled with orange color, and experiment with the number of breaks.
- ⑧ BONUS: What are the (dis-)advantages of boxplots and histograms?

# Mean vs. Median etc.

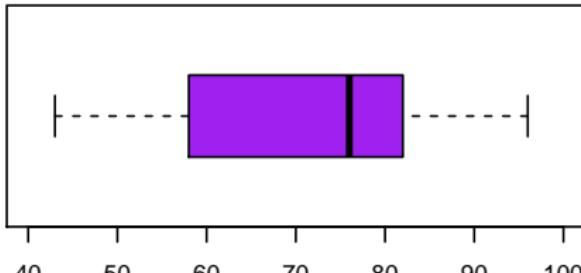
[mathwithbaddrawings.com/2016/07/13/why-not-to-trust-statistics](http://mathwithbaddrawings.com/2016/07/13/why-not-to-trust-statistics)

'faithful' – Old Faithful Geyser (Yellowstone)

Eruption duration [min]



Waiting Time between eruptions [min]

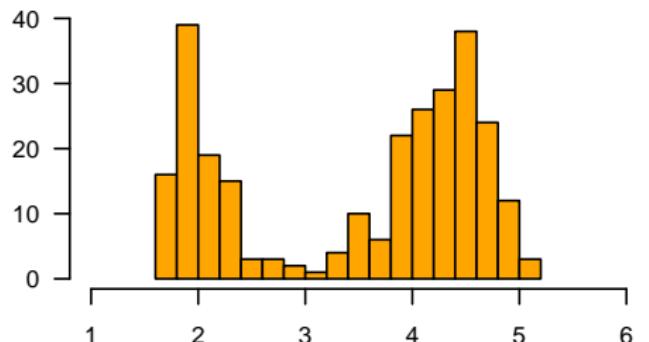
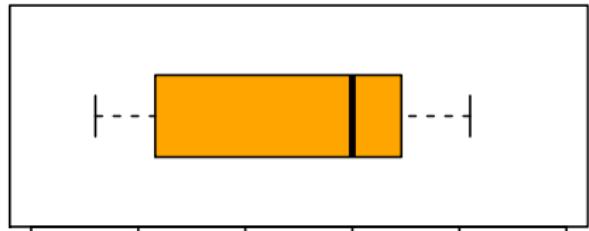


# Mean vs. Median etc.

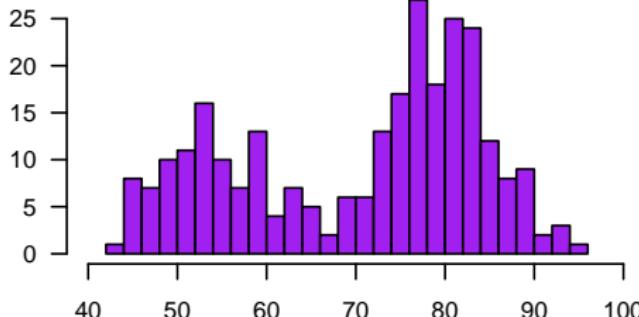
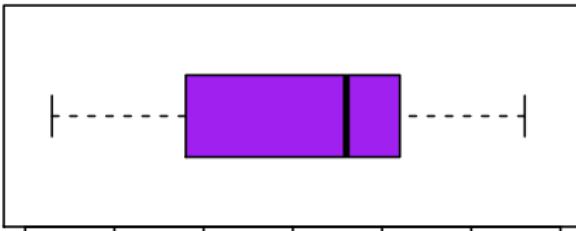
[mathwithbaddrawings.com/2016/07/13/why-not-to-trust-statistics](http://mathwithbaddrawings.com/2016/07/13/why-not-to-trust-statistics)

'faithful' – Old Faithful Geyser (Yellowstone)

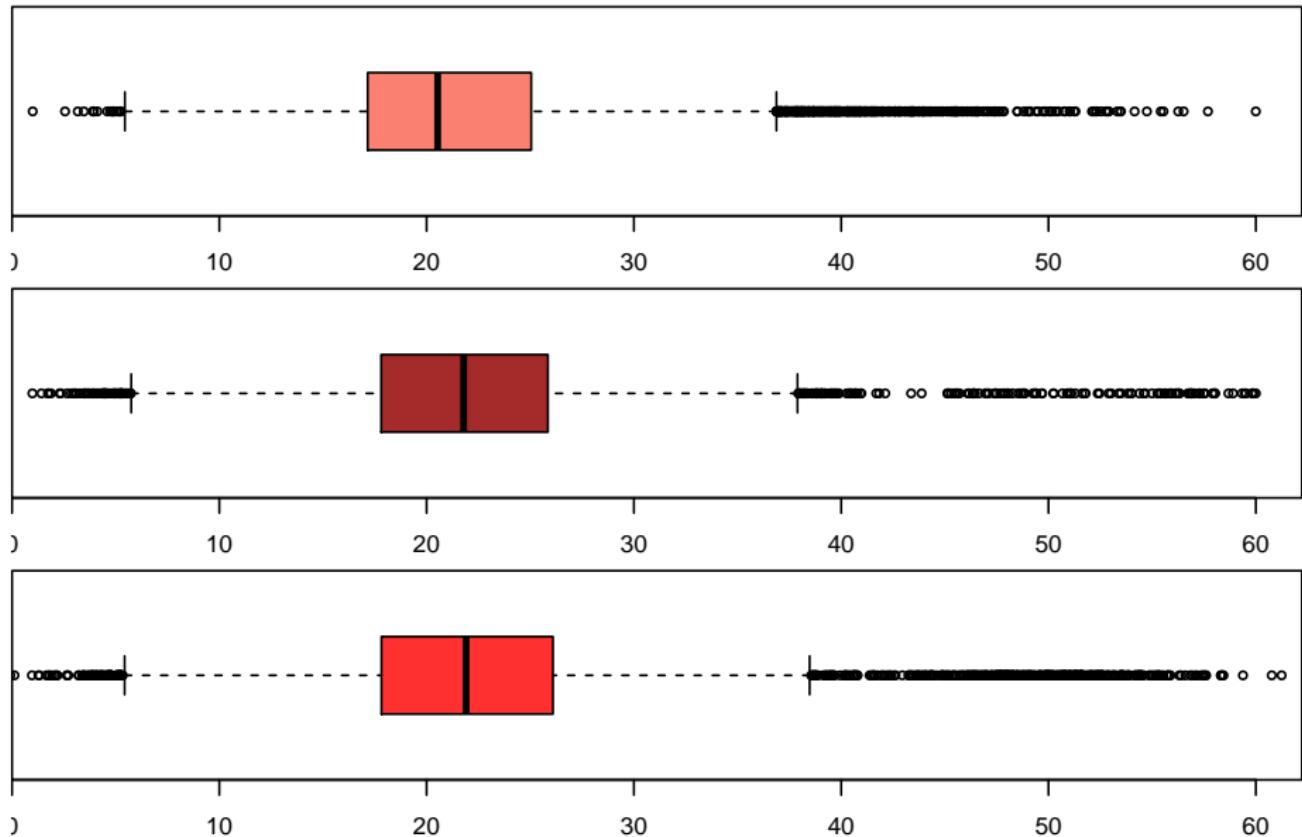
Eruption duration [min]



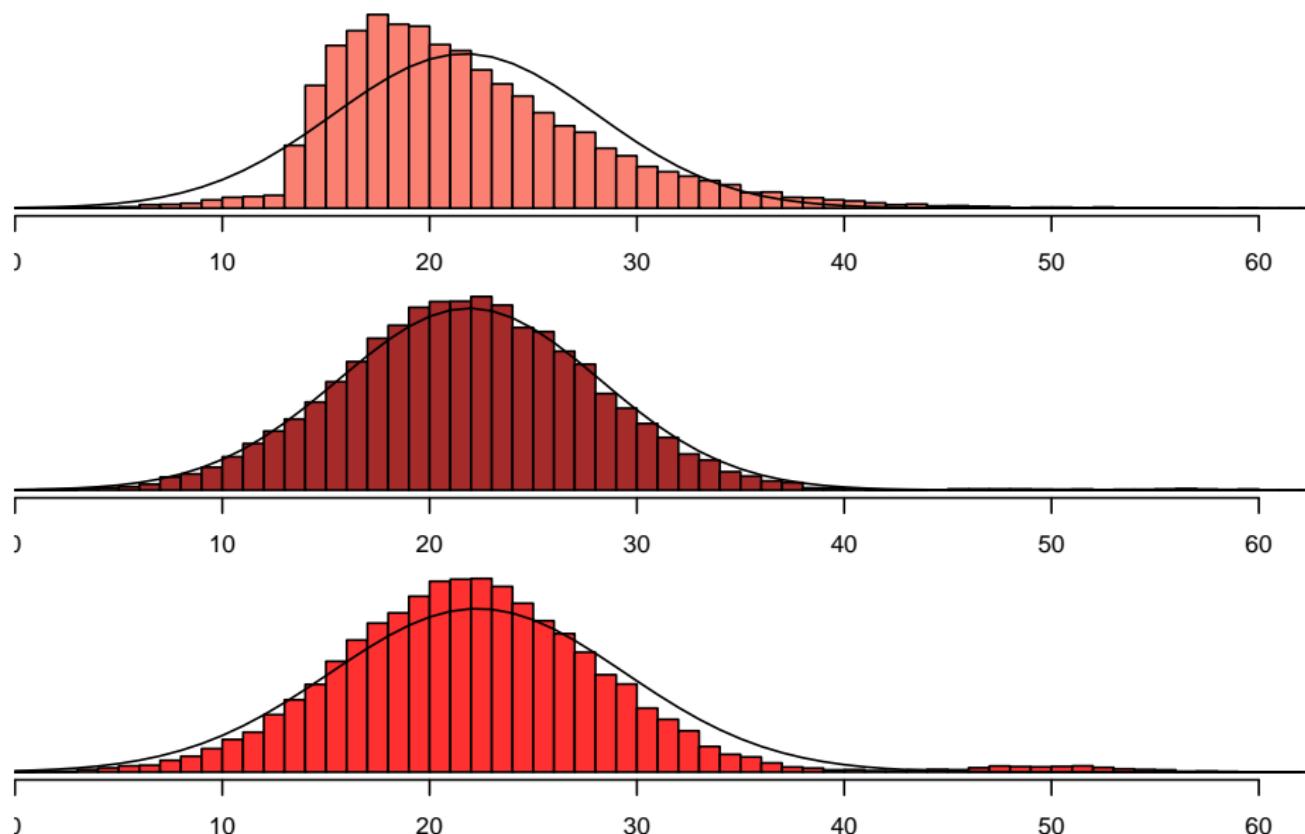
Waiting Time between eruptions [min]



## Similar boxplots...



... different distributions!



# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. **Distributions**
    - Exploratory Data Analysis (EDA)
    - Normal distribution
    - Other distribution functions
    - Beta distribution
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Random numbers

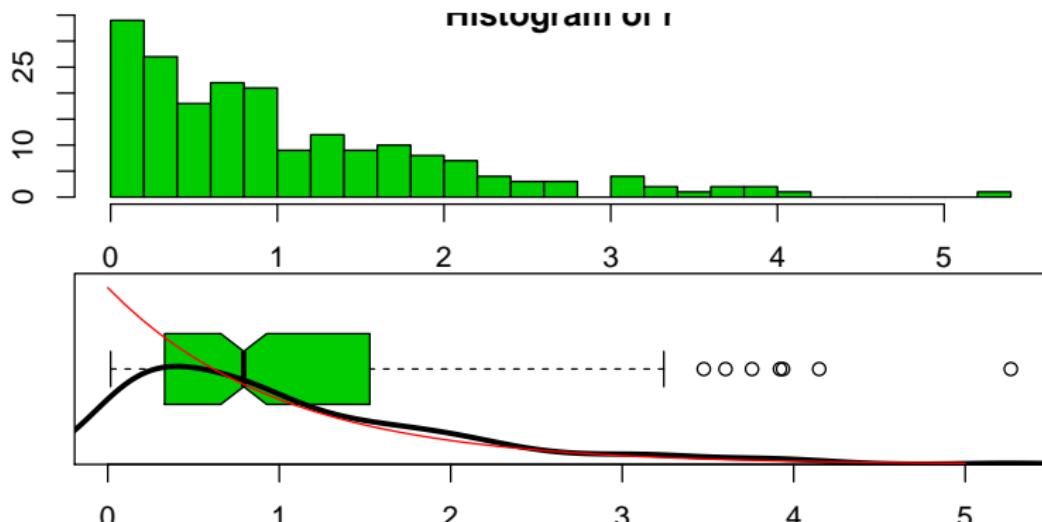
R has a very good random number generator. If you want to set a starting point for it, you can use `set.seed(n)`, with n being some number you like. After R has set the seed, it will always generate the same "random" numbers. This creates reproducible numbers.

## Exercise 37: Random numbers

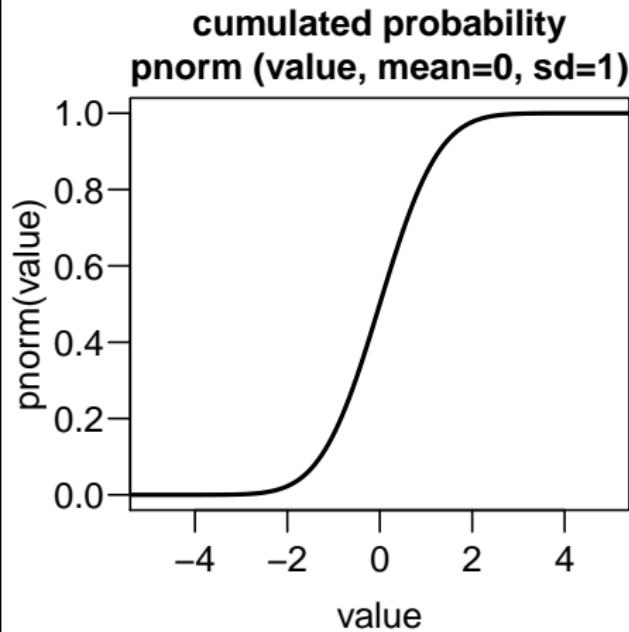
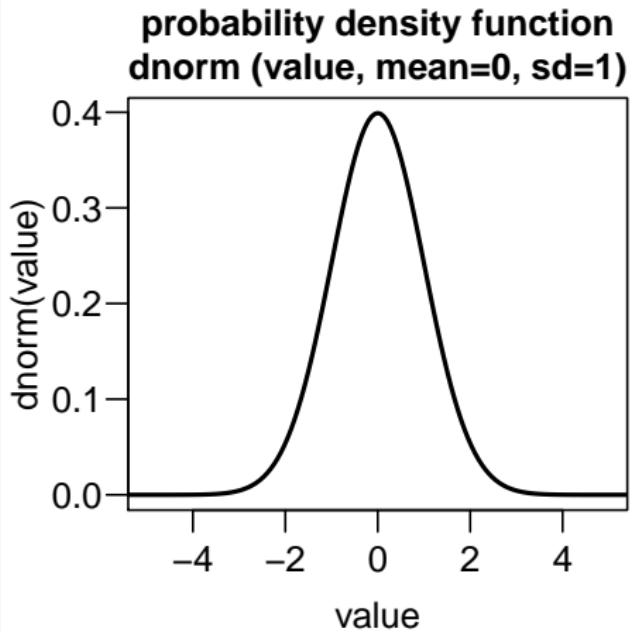
- ① With `sample`, put the numbers 5 to 15 in random order. Out of the possible numbers 1:3, draw 25 realizations.
- ② Via `?Distributions`, find out how to generate 200 random numbers from the exponential distribution and look at those with a boxplot and a histogram (BONUS: look at `ecdf` and `density`).

# Solution for exercise 37: random numbers

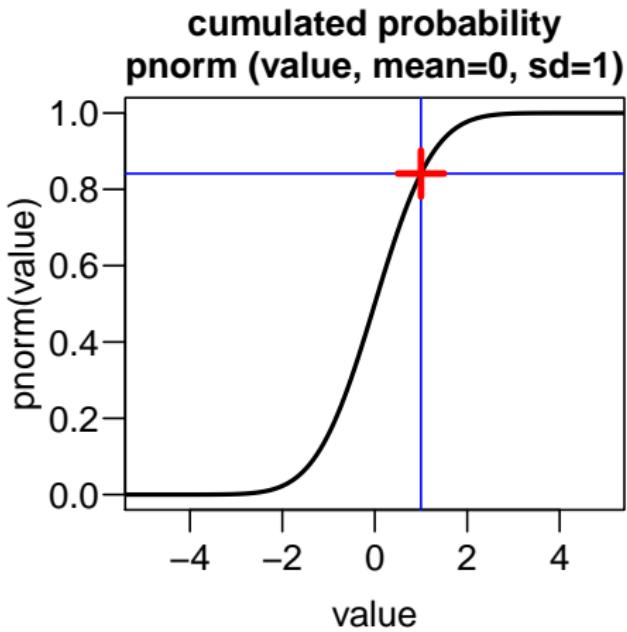
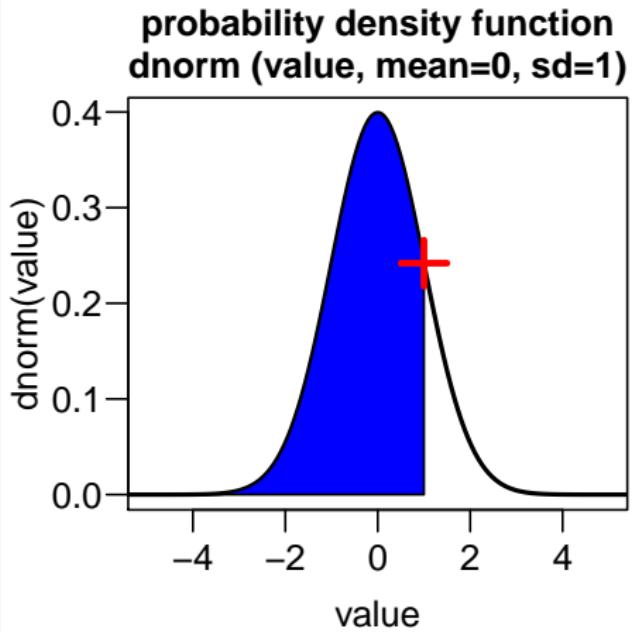
```
sample(5:15) ; sample(1:3, 25, replace=TRUE) ; r <- rexp(200)
par(mfrow=c(2,1), mar=c(2,2,0,0))
hist(r, breaks=20, col=3)
boxplot(r, horizontal=T, notch=T, col=3)
lines(density(r)$x, density(r)$y +par("usr")[3], lwd=3)
lines(0:50/10, dexp(0:50/10) +par("usr")[3], col=2)
```



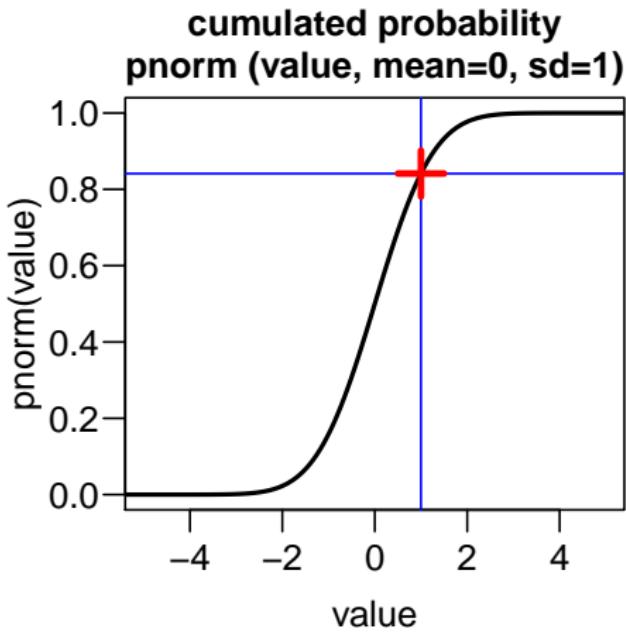
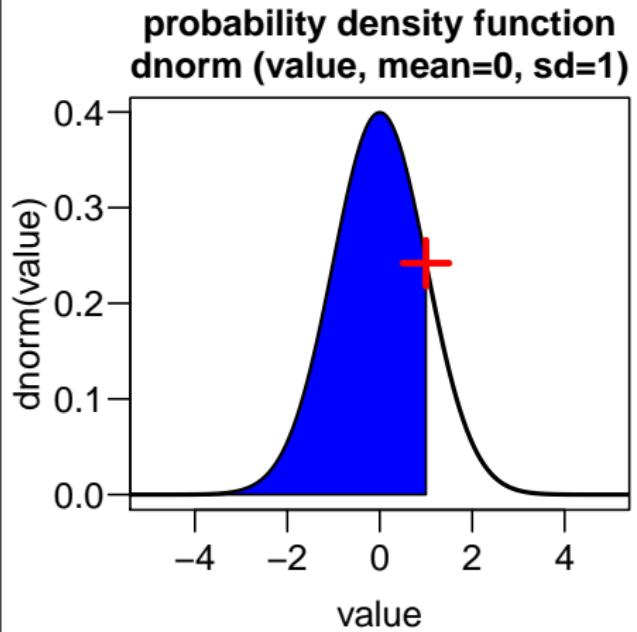
# Distributions: normal distribution I



## Distributions: normal distribution II



## Distributions: normal distribution II



`pnorm(q=1, mean=0, sd=1) = 84%` of the values of a standard normal distribution (SND) with  $\mu = 0$  and  $\sigma = 1$  are smaller than 1.

# Get familiar with the Gaussian distribution

## Exercise 38: normal distribution

In 1981, Fries and Crapo estimated future life expectancy to be 85 years with a standard deviation of 4.5 years.

Fries, J., Crapo, L.: Vitality and Aging: Implications of the Rectangular Curve. Freeman & Co Publishers, San Francisco (1981)

- ① Plot the distribution density curve
- ② Plot it again using `normPlot` in the package `berryFunctions`.
- ③ What proportion of people would die before age 75?  
Is that a realistic likelihood?
- ④ How many Germans (pop ~80 mio) would exceed 90 years?
- ⑤ In which symmetric age range would 95% of the people die?  
Give the rough estimate and the exact number.
- ⑥ How likely is it to die at age 85.3?
- ⑦ What are the biggest problems with this task?

# Solution for exercise 38: Normal distribution I

Examples taken from: Wolfgang Tschirk (2014): Statistik: Klassisch oder Bayes - Zwei Wege im Vergleich

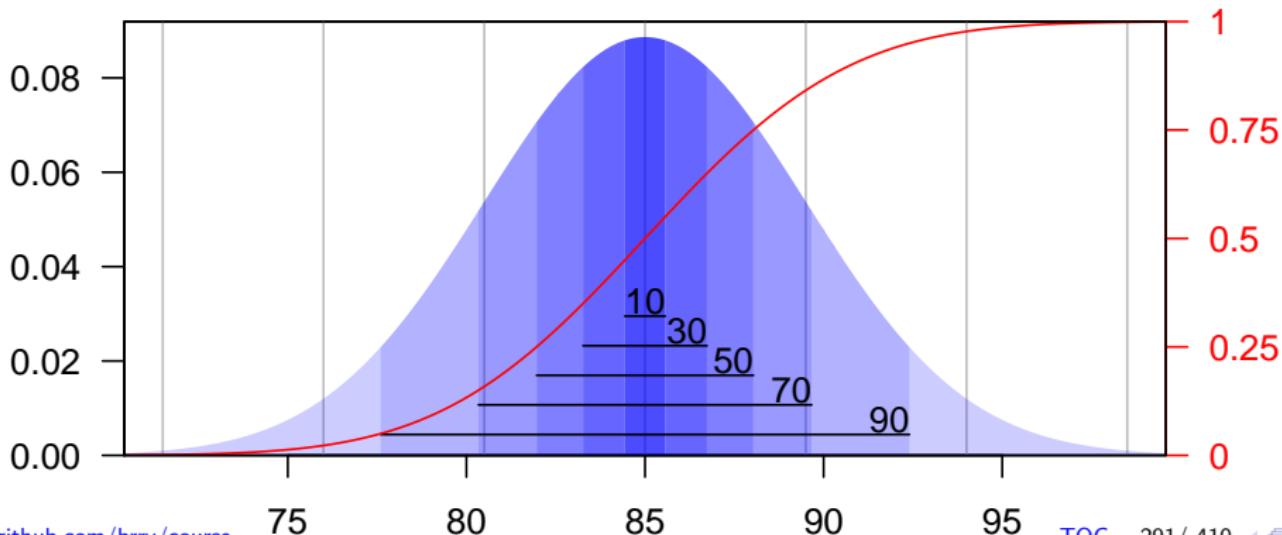
```
plot(70:100, dnorm(70:100, 85, 4.5), type="l")
berryFunctions::normPlot(85, 4.5)
pnorm(75, 85, 4.5)*100 # 1.3%
(1-pnorm(90, 85, 4.5))*80e6 # 10.7 mio
85-4.5*2; 85+4.5*2 # 76, 94
qnorm(c(0.025, 0.975), 85, 4.5) # 76.2 - 93.8
```

The probability to die at age 85.3 is 0 percent, as there are infinite possibilities in a continuous distribution. (You could also die at 85.331761 years). That's why we look at probability "density" in intervals.  
But: we don't know if life expectancy is normally distributed. We also don't know whether it is parametrized correctly.

## Solution for exercise 38: Normal distribution II

```
## [1] 1.313415  
## [1] 10660821  
## [1] 76.18016 93.81984
```

**Normal density with  
mean = 85 and sd = 4.5**



## Test for normality of a distribution

```
data <- rnorm(1000, mean=97, sd=8.9)
shapiro.test(data)
ks.test(data, "pnorm", mean(data), sd(data))
# if p > 0.05: accept Nullhypothesis
# (that data are normally distributed.)
```

More on [slide 327](#) et sequentes.

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. **Distributions**
    - Exploratory Data Analysis (EDA)
    - Normal distribution
    - Other distribution functions
      - Beta distribution
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Get familiar with the discrete Binomial distribution

## Exercise 39: binomial distribution

149'291 of 279'371 students in Austria are female. Statistik Austria: Studierende in Österreich im Wintersemester 2009/10. Statistik Austria, Wien (2010). This is a population for which we know the actual proportion, which happens rarely in inferential statistics. 100 students are randomly chosen.

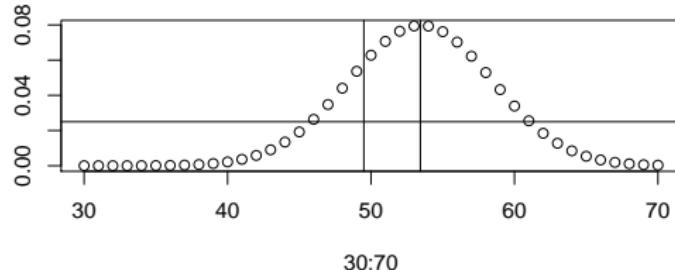
- ① Guess how likely it is that the majority (at least half) of those are female.
- ② Plot the probability for each reasonable possible outcome using `dbinom`. (This is the actual probability, not probability density, for each discrete number of females in a group of 100 randomly chosen students). Now guess again.
- ③ Actually calculate the probability.
- ④ At which proportion of females would you accuse the experimenter of selection bias (or cheating)?

## Solution for exercise 39: Binomial distribution

```
plot(30:70, dbinom(x=30:70, size=100, prob=149291/279371))
abline(v=c(49.5, 53.438), h=0.025)
1-pbinom(q=49.5, size=100, prob=149291/279371)

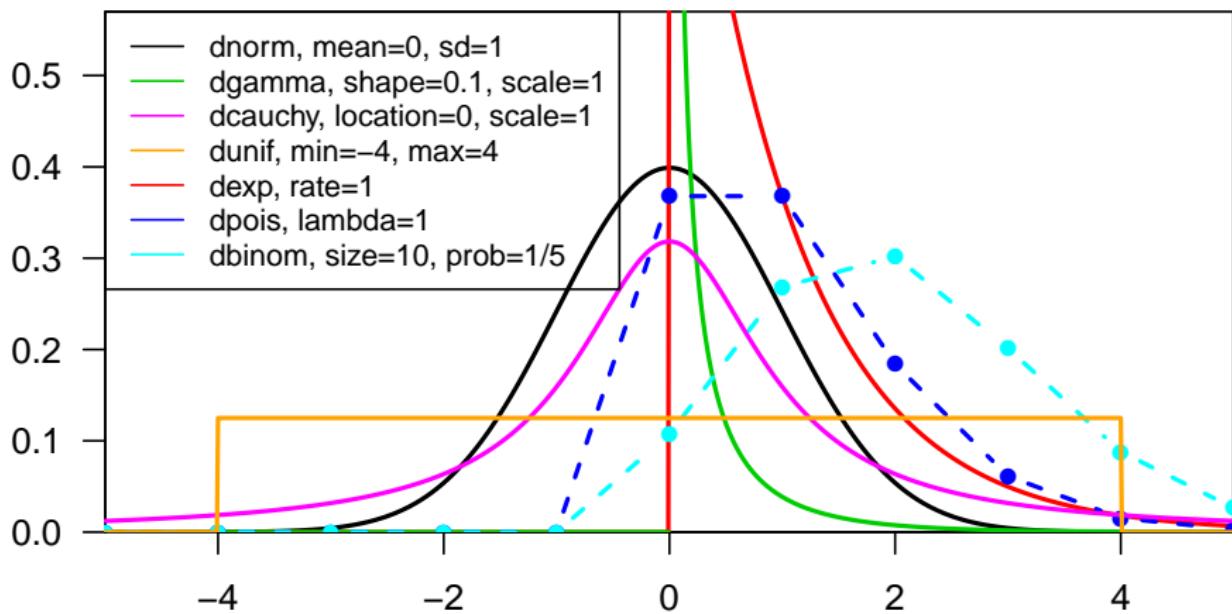
## [1] 0.7852772
```

## [1] 0.7852772  
im(x = 30:70, size = 100, prob = 149291/279371)

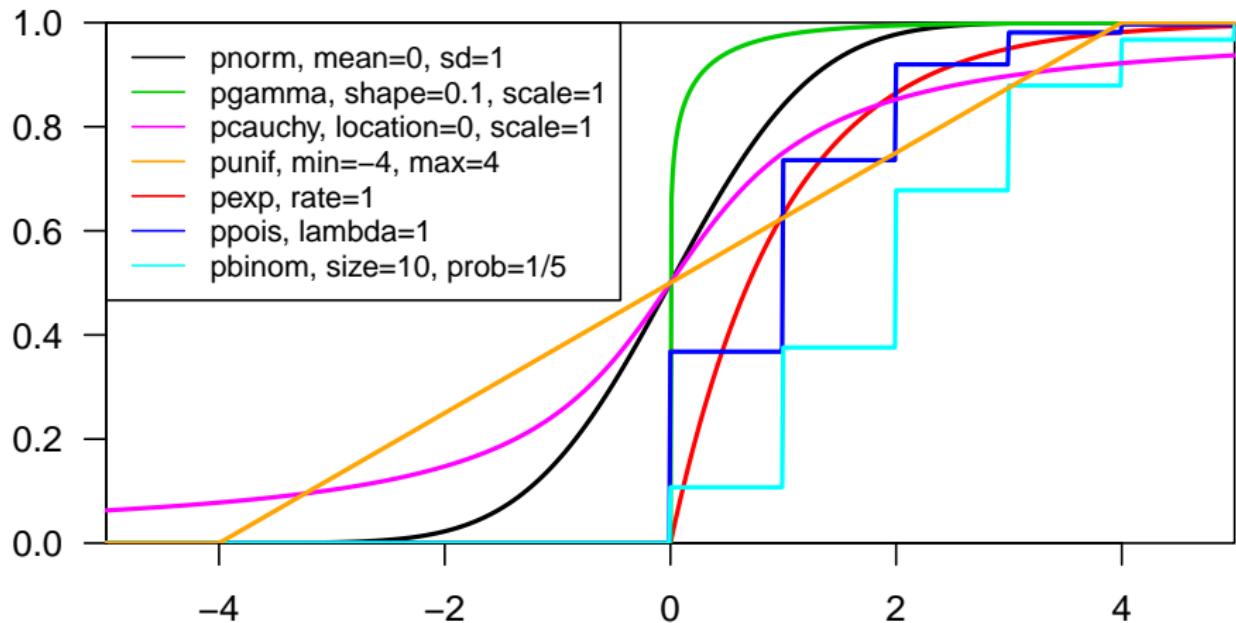


If the probability of the result happening by chance is under 5%, in this case  $\leq 45$  or  $\geq 61$  females (two-sided confidence interval).

# Other distributions: density



## Other distributions: cumulated



# Overview of distributions in R

prefix	meaning	usage
d	<b>density</b>	calculate probability (density) for value x
p	<b>cumulated probability</b>	area under density curve up until value x
q	<b>quantile</b>	reciprocal to p: calculate value x for a given cumulated probability
r	<b>random</b>	draw random numbers from distribution

# Overview of distributions in R

prefix	meaning	usage
d	<b>density</b>	calculate probability (density) for value x
p	<b>cumulated probability</b>	area under density curve up until value x
q	<b>quantile</b>	reciprocal to p: calculate value x for a given cumulated probability
r	<b>random</b>	draw random numbers from distribution

distribution	parameters	type
binom	size, prob	discrete
pois	lambda	
norm	mean, sd	
exp	rate	continuous
beta	shape1, shape2	

## Exercise 40: classical statistical analysis

- ① Calculate `mean`, `var`, `sd` and standard error of the mean ( $SEM = \frac{\sigma}{\sqrt{n}}$ ) of the `iris` sepal length data and interpret these values.
- ② Visualize the theoretical distribution of the dataset. Use a sequence of values and calculate `pnorm` and `dnorm` on it with mean and `sd` according to the previous task.
- ③ Compare this with the actual distribution of the sample. First draw a histogram with `freq=FALSE`, then add the theoretical distribution with `lines`.
- ④ If this distribution were from the general and complete "population" of iris sepal lengths, we could say: 20% of the sepal lengths are smaller than a certain value. Use `qnorm` to find out this value. What is the limit of this approach?
- ⑤ BONUS: What are the (dis-)advantages of mean and median?

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  - 9. Distributions**
    - Exploratory Data Analysis (EDA)
    - Normal distribution
    - Other distribution functions
    - Beta distribution**
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Beta distribution: motivating example

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India.

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic.

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction. You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration.

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction. You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction. You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?** To graph the uncertainty, you want to use the beta distribution.

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction. You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?** To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$$\alpha - 1 = \text{number of successes}$$

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

What are the parameters in this case?

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

What are the parameters in this case?

$\alpha =$

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

What are the parameters in this case?

$$\alpha = 2 + 1 =$$

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

What are the parameters in this case?

$$\alpha = 2 + 1 = 3$$

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

What are the parameters in this case?

$$\alpha = 2 + 1 = 3$$

$$\beta =$$

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

What are the parameters in this case?

$$\alpha = 2 + 1 = 3$$

$$\beta = 10 - 2 + 1 =$$

## Beta distribution: motivating example

Imagine you are the scientific advisor to the government of India. It has recently been found that some of the millions (yes, millions) of wells contain dangerous levels of arsenic. If more than 25% of the wells are affected, the government will pass a law constraining well construction.

You sample 10 wells and find that 2 surpass the safety limit of arsenic concentration. **What will you do?**

To graph the uncertainty, you want to use the beta distribution. You remember (from this class) that for a binomial trial, the two parameters can be defined as:

$\alpha - 1$  = number of successes

$\beta - 1$  = number of failures

What are the parameters in this case?

$$\alpha = 2 + 1 = 3$$

$$\beta = 10 - 2 + 1 = 9$$

# Beta distribution: some theory + density plot

## Beta distribution: some theory + density plot

- Continuous, values between 0 and 1

## Beta distribution: some theory + density plot

- Continuous, values between 0 and 1
- Suitable for proportions and probabilities

## Beta distribution: some theory + density plot

- Continuous, values between 0 and 1
- Suitable for proportions and probabilities
- Very versatile, different shapes

## Beta distribution: some theory + density plot

- Continuous, values between 0 and 1
- Suitable for proportions and probabilities
- Very versatile, different shapes
- Used extensively in Bayesian statistics, see e.g. [Liu and Kong \(2015\)](#).

## Beta distribution: some theory + density plot

- Continuous, values between 0 and 1
- Suitable for proportions and probabilities
- Very versatile, different shapes
- Used extensively in Bayesian statistics, see e.g. [Liu and Kong \(2015\)](#).
- Two parameters: `a=alpha=shape1` and `b=beta=shape2`.

# Beta distribution: some theory + density plot

- Continuous, values between 0 and 1
- Suitable for proportions and probabilities
- Very versatile, different shapes
- Used extensively in Bayesian statistics, see e.g. [Liu and Kong \(2015\)](#).
- Two parameters: `a=alpha=shape1` and `b=beta=shape2`.

## Exercise 41: Beta distribution density

- ① Plot the beta distribution density (`dbeta`) for `alpa=3` and `beta=9`.
- ② If this describes the true distribution of contaminated well proportions across subregions of India, how likely is it to have a sample with 25% or more of the wells contaminated?
- ③ BONUS: Plot the distribution with `polygon`.

Two hints:

# Beta distribution: some theory + density plot

- Continuous, values between 0 and 1
- Suitable for proportions and probabilities
- Very versatile, different shapes
- Used extensively in Bayesian statistics, see e.g. [Liu and Kong \(2015\)](#).
- Two parameters: `a=alpha=shape1` and `b=beta=shape2`.

## Exercise 41: Beta distribution density

- ① Plot the beta distribution density (`dbeta`) for `alpa=3` and `beta=9`.
- ② If this describes the true distribution of contaminated well proportions across subregions of India, how likely is it to have a sample with 25% or more of the wells contaminated?
- ③ BONUS: Plot the distribution with `polygon`.

Two hints:

Hint 1: Create a vector from 0 to 1 with, say, 200 values.

# Beta distribution: some theory + density plot

- Continuous, values between 0 and 1
- Suitable for proportions and probabilities
- Very versatile, different shapes
- Used extensively in Bayesian statistics, see e.g. [Liu and Kong \(2015\)](#).
- Two parameters: `a=alpha=shape1` and `b=beta=shape2`.

## Exercise 41: Beta distribution density

- ① Plot the beta distribution density (`dbeta`) for `alpa=3` and `beta=9`.
- ② If this describes the true distribution of contaminated well proportions across subregions of India, how likely is it to have a sample with 25% or more of the wells contaminated?
- ③ BONUS: Plot the distribution with `polygon`.

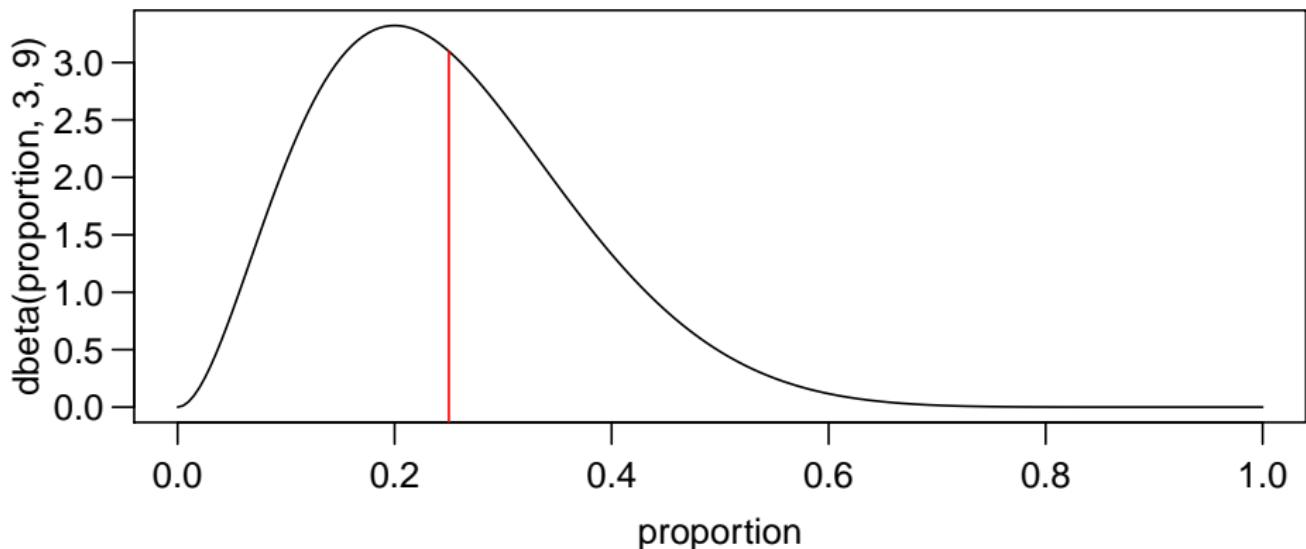
Two hints:

Hint 1: Create a vector from 0 to 1 with, say, 200 values.

Hint 2: Plot `dbeta(x)` over `x` using `type="l"`.

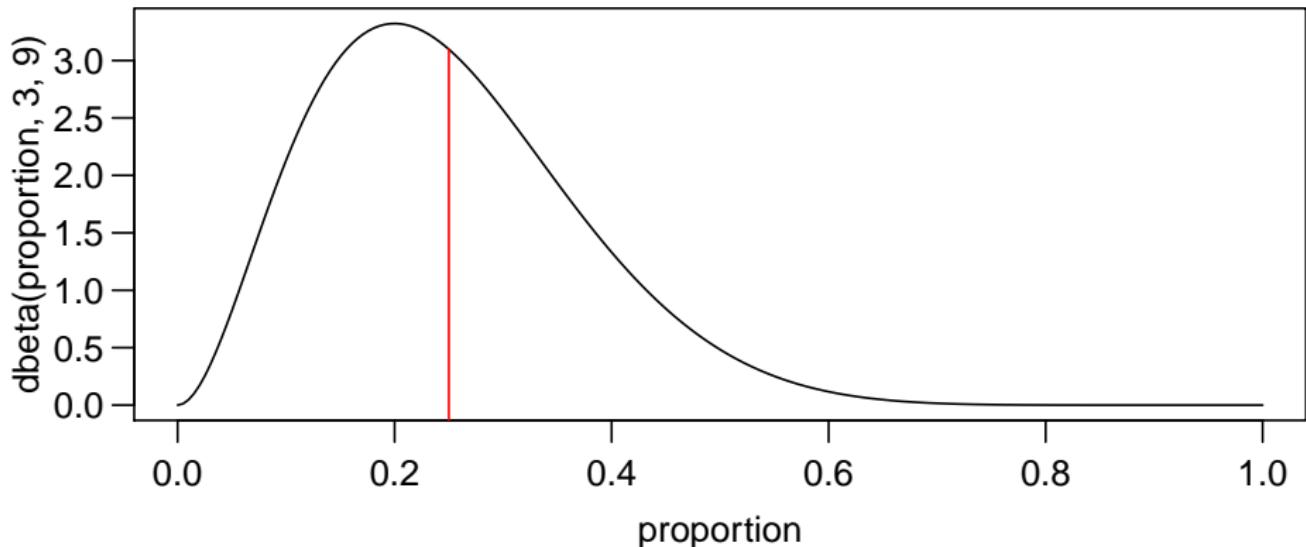
## Solution to exercise 41: Beta density

```
proportion <- seq(0,1, len=200)
plot(proportion, dbeta(proportion, 3,9), type="l")
segments(x0=0.25, y0=-1, y1=3.097, col=2)
```



## Solution to exercise 41: Beta density

```
proportion <- seq(0,1, len=200)
plot(proportion, dbeta(proportion, 3,9), type="l")
segments(x0=0.25, y0=-1, y1=3.097, col=2)
```



```
1-pbeta(0.25, 3,9)
```

```
## [1] 0.4552009
```

[github.com/brry/course](http://github.com/brry/course)

# Beta parameters

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the *left/right*?

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the right

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the right

Large  $\beta$ : HDI rather left, lower values

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the right

Large  $\beta$ : HDI rather left, lower values

Both large:

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the right

Large  $\beta$ : HDI rather left, lower values

Both large: (*more samples*)

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the right

Large  $\beta$ : HDI rather left, lower values

Both large: (*more samples*) narrow HDI

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the right

Large  $\beta$ : HDI rather left, lower values

Both large: (*more samples*) narrow HDI

```
install.packages("berryFunctions")
```

## Beta parameters

Large  $\alpha$ : HDI (Highest Density Interval) more to the right

Large  $\beta$ : HDI rather left, lower values

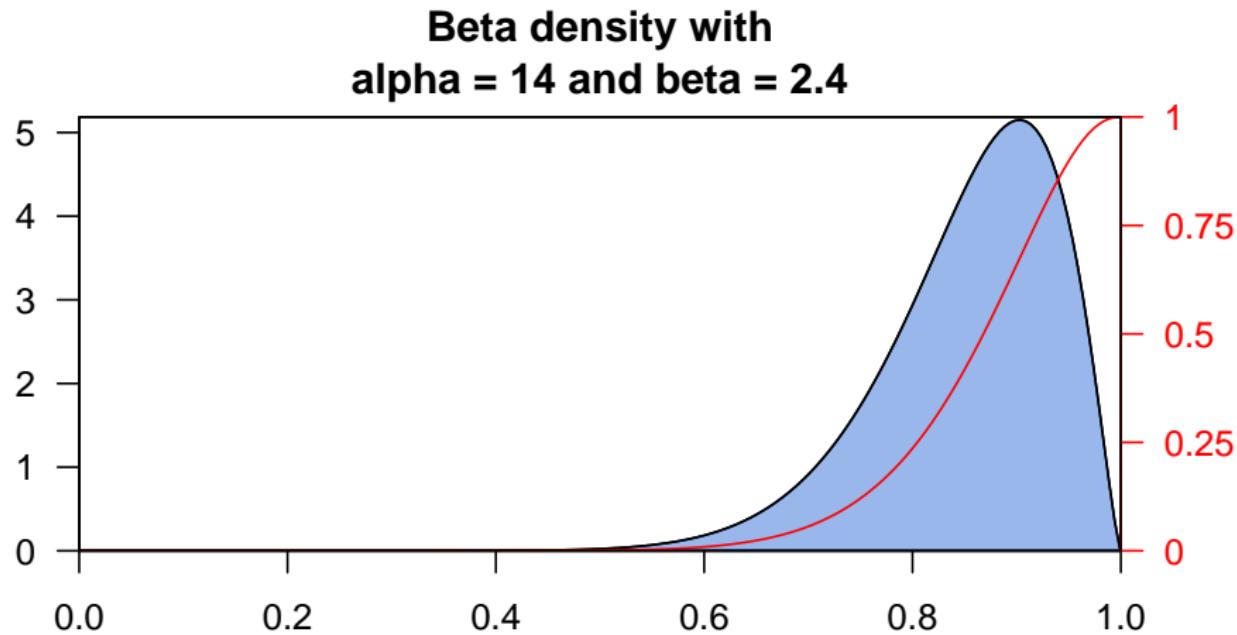
Both large: (*more samples*) narrow HDI

```
install.packages("berryFunctions")
```

```
library(berryFunctions) # >= version 1.9.6 (2015-12-18)
betaPlot(shape1=alpha, shape2=beta, ...)
betaPlotComp()
```

## Beta parameters

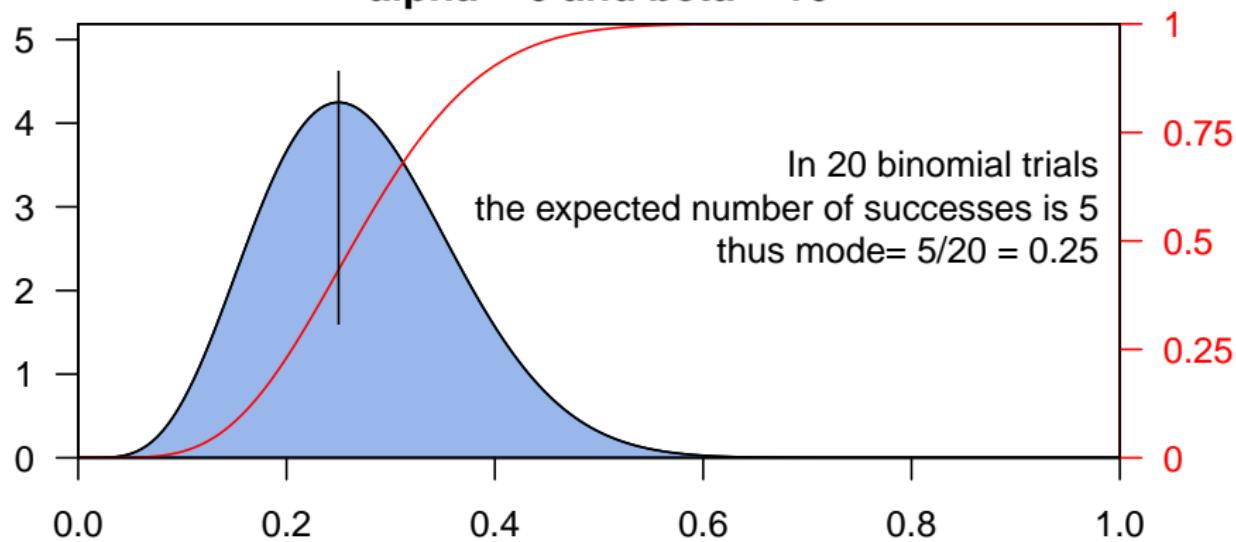
```
betaPlot(14, 2.4, ylim=lim0(5))
```



## Beta parameters

```
betaPlot(6, 16, ylim=lim0(5))
```

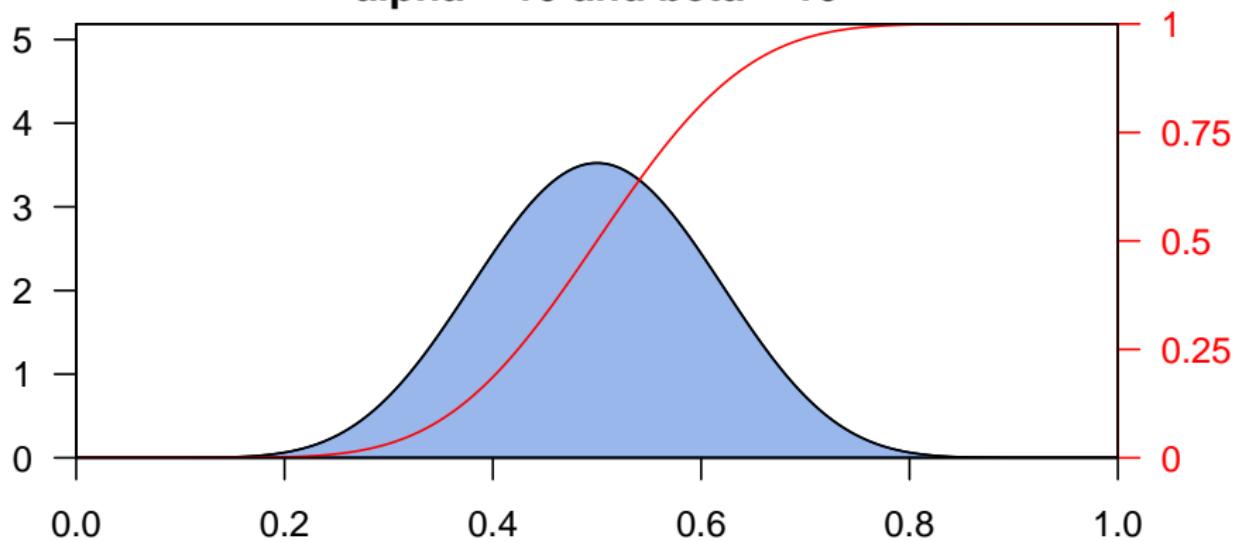
**Beta density with  
alpha = 6 and beta = 16**



## Beta parameters

```
betaPlot(10, 10, ylim=lim0(5))
```

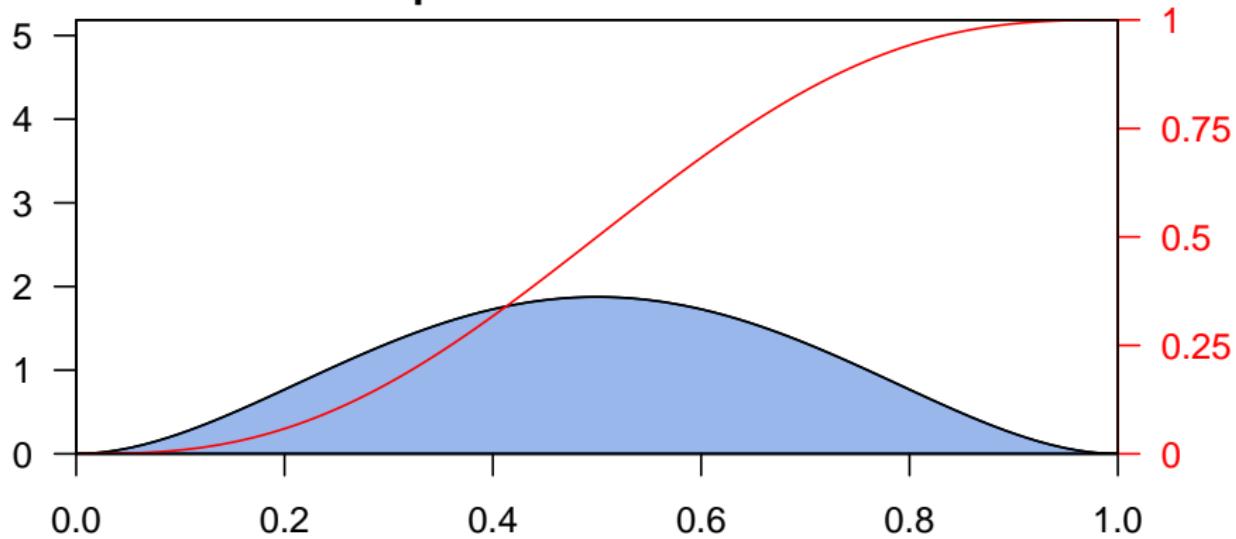
**Beta density with  
alpha = 10 and beta = 10**



## Beta parameters

```
betaPlot(3, 3, ylim=lim0(5))
```

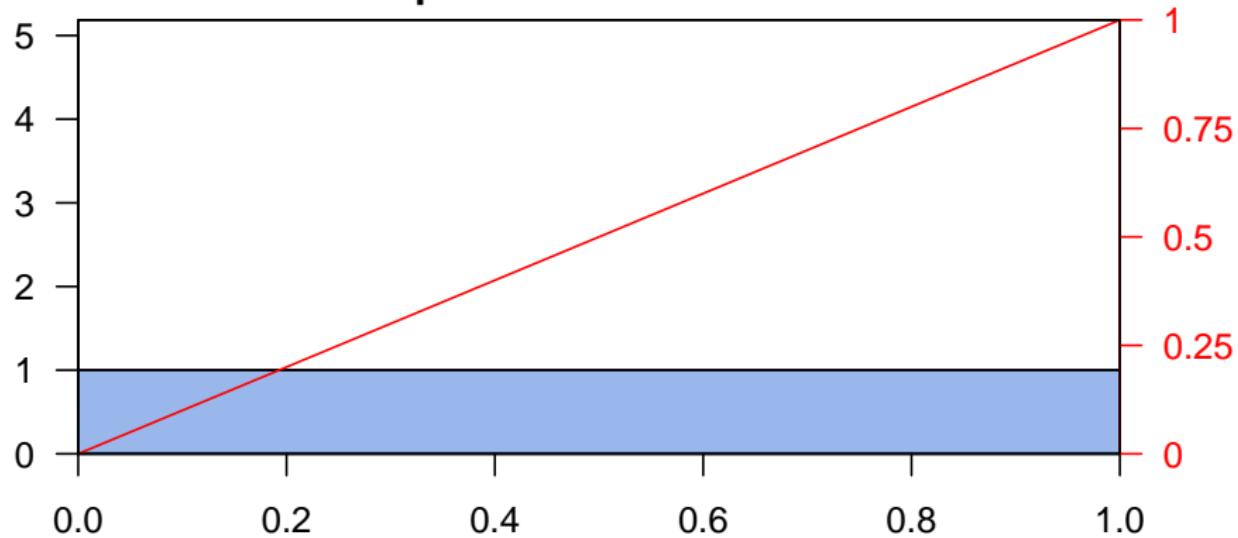
**Beta density with  
alpha = 3 and beta = 3**



## Beta parameters

```
betaPlot(1, 1, ylim=lim0(5))
```

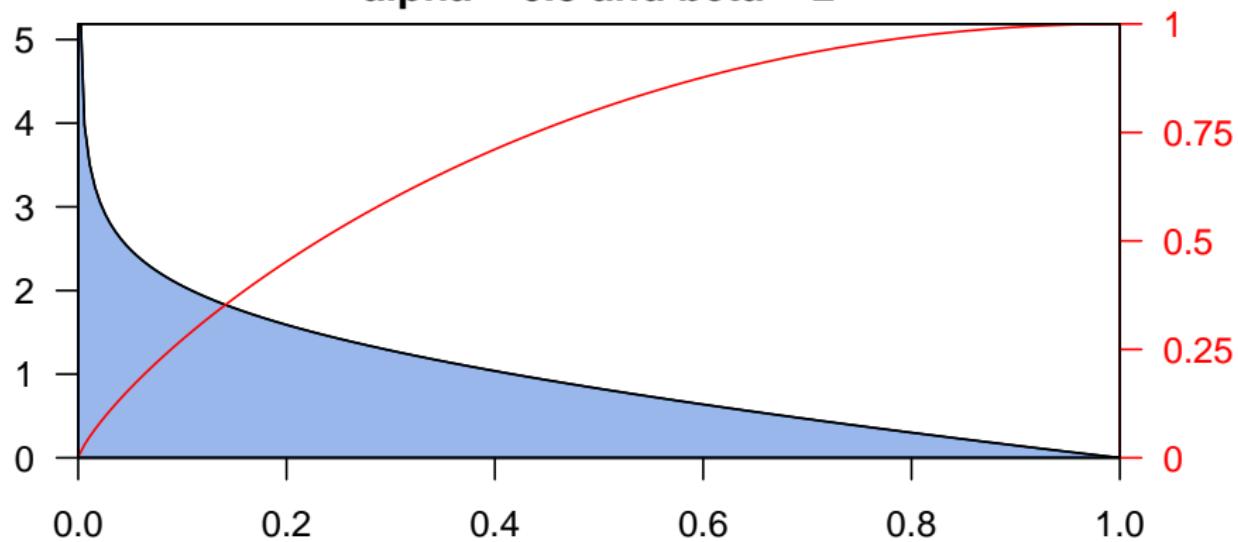
**Beta density with  
alpha = 1 and beta = 1**



## Beta parameters

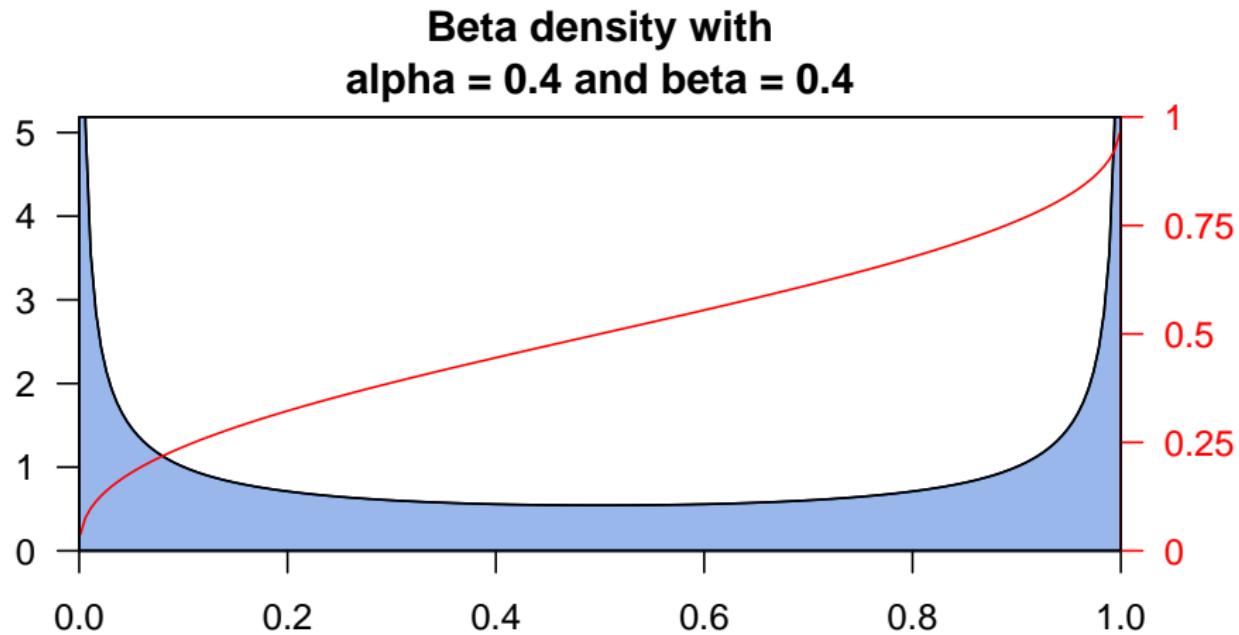
```
betaPlot(0.8, 2, ylim=lim0(5))
```

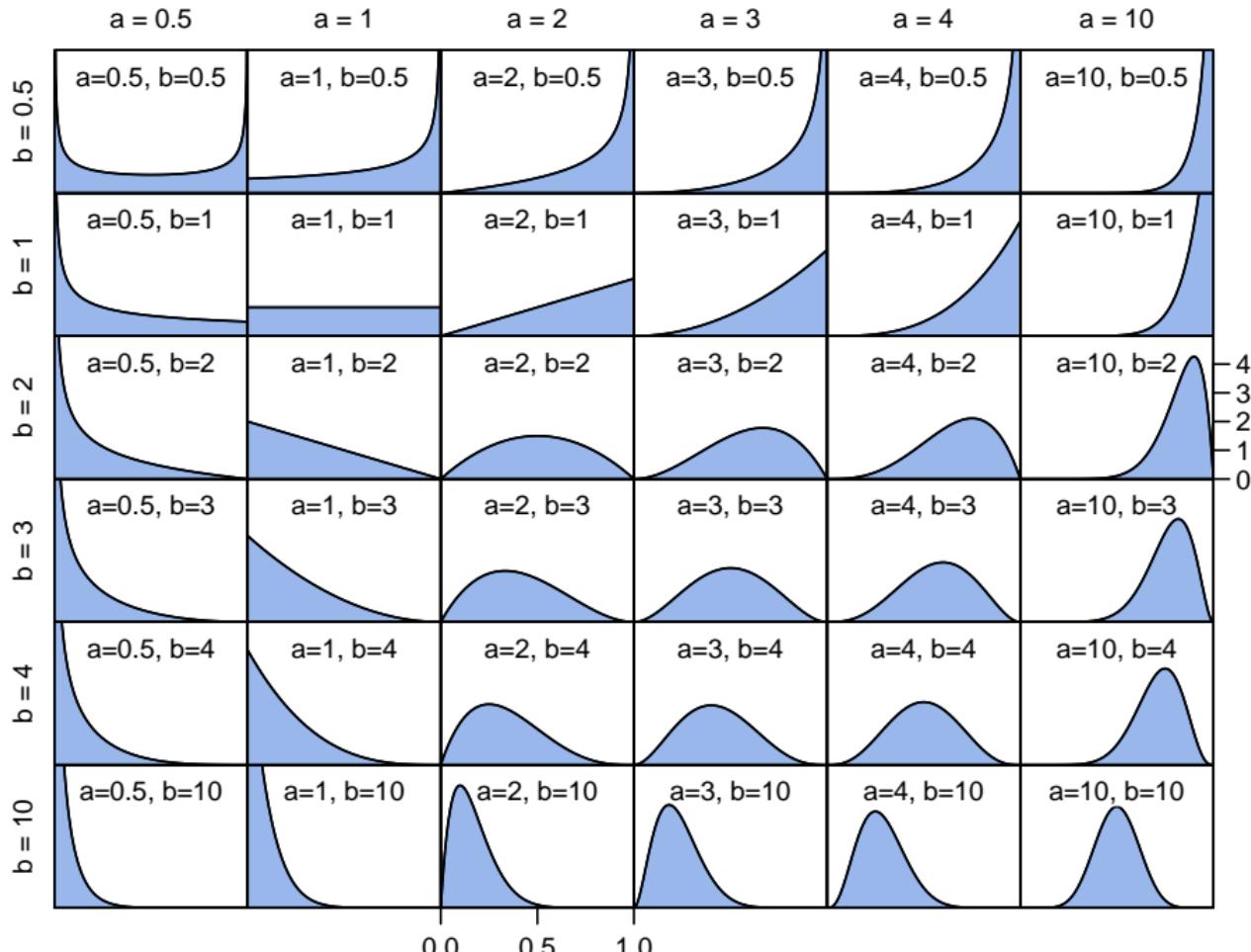
**Beta density with  
alpha = 0.8 and beta = 2**



## Beta parameters

```
betaPlot(0.4, 0.4, ylim=lim0(5))
```





# Beta distribution properties

# Beta distribution properties

$$\text{mean} = \mu = \frac{a}{a+b}$$

# Beta distribution properties

$$\text{mean} = \mu = \frac{a}{a+b}$$

$$\text{variance} = \sigma^2 = \frac{ab}{(a+b)^2 * (a+b+1)}$$

# Beta distribution properties

$$\text{mean} = \mu = \frac{a}{a+b} \qquad \text{variance} = \sigma^2 = \frac{ab}{(a+b)^2 * (a+b+1)}$$

With these formulas, if  $a$  and  $b \gg 1$ , you can estimate an approximation to the normal distribution.

To get  $a$  &  $b$  from mean & var, see

<https://stats.stackexchange.com/a/12239>

# Beta distribution properties

$$\text{mean} = \mu = \frac{a}{a+b} \qquad \text{variance} = \sigma^2 = \frac{ab}{(a+b)^2 * (a+b+1)}$$

With these formulas, if  $a$  and  $b \gg 1$ , you can estimate an approximation to the normal distribution.

To get  $a$  &  $b$  from mean & var, see

<https://stats.stackexchange.com/a/12239>

$$\text{mode} = \text{density peak} = m = \frac{a-1}{a+b-2} \quad \text{for } a, b > 1$$

# Beta distribution properties

$$\text{mean} = \mu = \frac{a}{a+b} \qquad \text{variance} = \sigma^2 = \frac{ab}{(a+b)^2 * (a+b+1)}$$

With these formulas, if  $a$  and  $b \gg 1$ , you can estimate an approximation to the normal distribution.

To get  $a$  &  $b$  from mean & var, see

<https://stats.stackexchange.com/a/12239>

$$\text{mode} = \text{density peak} = m = \frac{a-1}{a+b-2} \quad \text{for } a, b > 1$$

If you want to set the mode at a certain value, you can compute beta with

$$b = \frac{a-1}{m} - a + 2 \quad \text{and alpha with} \quad a = \frac{(2-b-1/m)*m}{m-1}$$

# Beta distribution specification

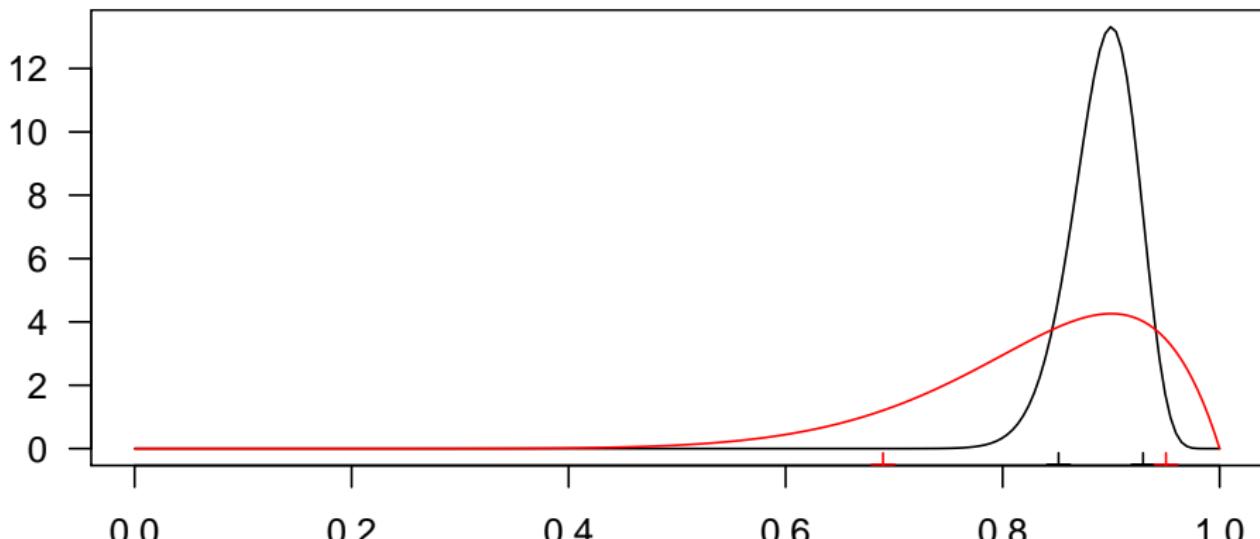
## Exercise 42: Beta distribution parameter determination

You want to specify a distribution with the peak at 90%, reached in a sample of 100.

- ① What are the parameters?
- ② What does the distribution look like? (Graph it)
- ③ BONUS: How does the HDI (highest density interval) width change for a sample size of 10?

## Beta distribution parametrization solution

```
v <- seq(0,1, len=200)
plot(v, dbeta(v, 91,11), type="l", las=1)
lines(v, dbeta(v, 10,2), col=2)
points(qbeta(c(0.1, 0.9), 91,11), rep(-0.5,2), pch=3)
points(qbeta(c(0.1, 0.9), 10,2), rep(-0.5,2), pch=3, col=2)
```



# Bayesian inference with beta distribution

Prior: weak knowledge that true proportion of cats among pets must be around 10 to 30%, but may also be more.

# Bayesian inference with beta distribution

Prior: weak knowledge that true proportion of cats among pets must be around 10 to 30%, but may also be more.

Data: observation in survey: 35 of 100 households with a pet have a cat.

# Bayesian inference with beta distribution

Prior: weak knowledge that true proportion of cats among pets must be around 10 to 30%, but may also be more.

Data: observation in survey: 35 of 100 households with a pet have a cat.

Posterior beta parameters:  $a_{\text{prior}} + \text{successes}$ ,  $b_{\text{prior}} + \text{failures}$

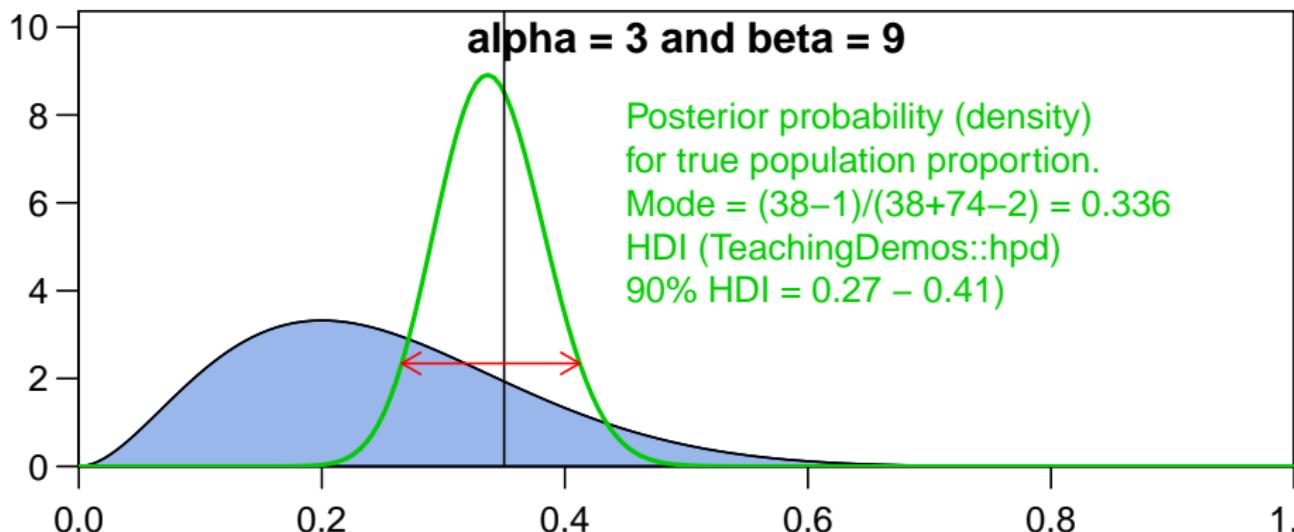
# Bayesian inference with beta distribution

Prior: weak knowledge that true proportion of cats among pets must be around 10 to 30%, but may also be more.

Data: observation in survey: 35 of 100 households with a pet have a cat.

Posterior beta parameters:  $a_{\text{prior}} + \text{successes}$ ,  $b_{\text{prior}} + \text{failures}$

```
betaPlot(3, 9, ylim=lim0(10), cum=F, mar=c(2,2,0,0), keeppar=T)
lines(x, dbeta(x, 3+35, 9+65), lwd=2, col=3)
```



## Shift + scale values

If values do not lie between 0 and 1, but between m and n, you can transform the raw values (y) to the 0:1 space (y') with

$$y' = \frac{(y-m)}{(n-m)}$$

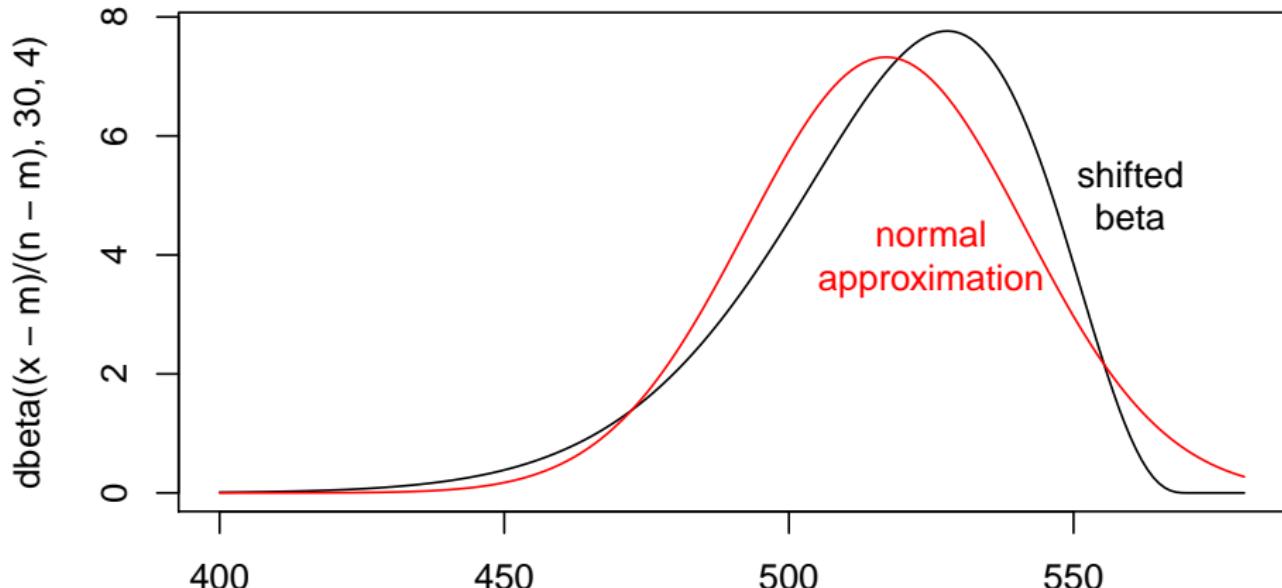
and beta values (y') to raw values (y) with

$$y = y' * (n - m) + m$$

Outlook: used for beta regression, see [Liu and Kong \(2015\)](#): zoib: An R Package for Bayesian Inference for Beta Regression and Zero/One Inflated Beta Regression. In: the R Journal Vol. 7/2, December 2015

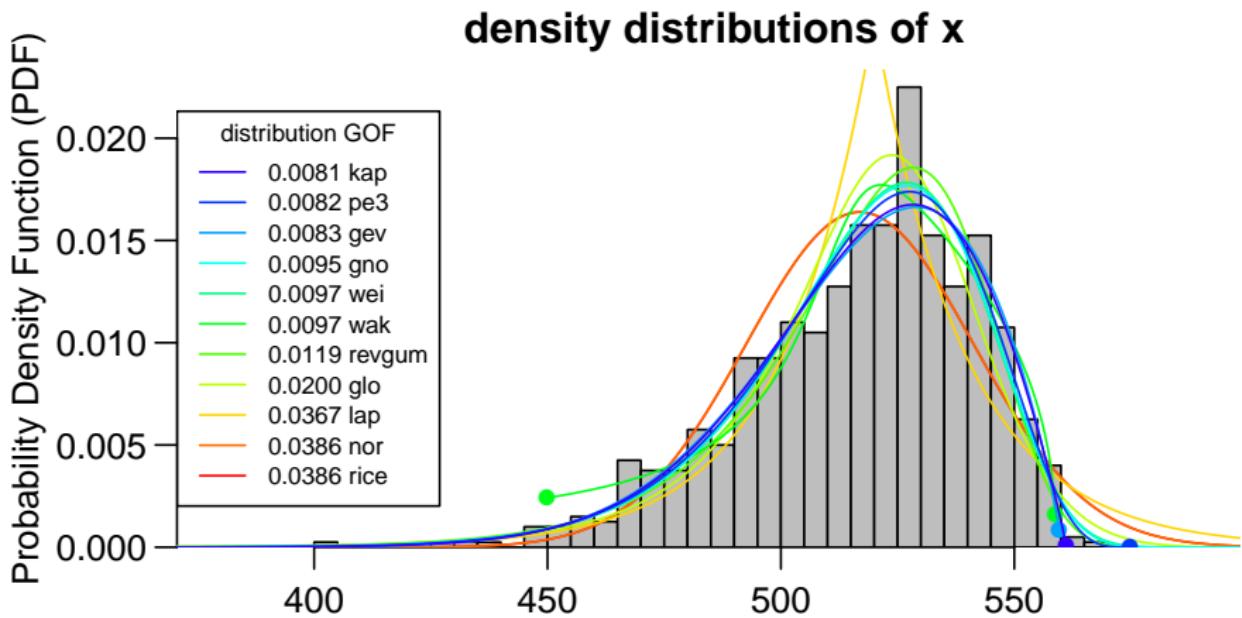
## Shift + scale values

```
x <- 400:580; m <- 120 ; n <- 570  
plot(x, dbeta((x-m)/(n-m), 30, 4), type="l")  
lines(x, dnorm((x-m)/(n-m), 30/34, sd=sqrt(120/(34^2*35))), col=2)
```



# Shift + scale values

```
m <- 120 ; n <- 570; set.seed(1)
x <- rbeta(800,30,4) * (n-m)+m
library(extremeStat) # github.com/brry/extremeStat
plotLfit(distLfit(x, quiet=T), breaks=30, nbest=11, legargs=c(x="left"))
```



# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  - 10. Statistics**
    - Simpsons Paradox**
    - Normality tests
    - Confidence Intervals
    - (Linear) regression
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Simpson's Paradox

- Kievit et al (2015): [Simpson's paradox in psychological science: a practical guide](#) (Frontiers in Psychology)
- [vudlab app](#)
- [Wikipedia entry](#)
- 2016 example: [development of crime rates in the USA](#)
- [minutephysics video \(2017\)](#)

## Simpson's Paradox: Kidney stone treatment success rate

	open surgery	small puncture
small stones		
large stones		
together	$\frac{273}{350} = 78\%$	$\frac{289}{350} = 83\%$

Wikipedia, Julious + Mullee (1994)

## Simpson's Paradox: Kidney stone treatment success rate

	open surgery	small puncture
small stones	$\frac{81}{87} = 93\%$	$\frac{234}{270} = 87\%$
large stones		
together	$\frac{273}{350} = 78\%$	$\frac{289}{350} = 83\%$

Wikipedia, Julious + Mullee (1994)

## Simpson's Paradox: Kidney stone treatment success rate

	open surgery	small puncture
small stones	$\frac{81}{87} = 93\%$	$\frac{234}{270} = 87\%$
large stones	$\frac{192}{263} = 73\%$	$\frac{55}{80} = 69\%$
together	$\frac{273}{350} = 78\%$	$\frac{289}{350} = 83\%$

Wikipedia, Julious + Mullee (1994)

## Simpson's Paradox: Kidney stone treatment success rate

	open surgery	small puncture
small stones	$\frac{81}{87} = 93\%$	$\frac{234}{270} = 87\%$
large stones	$\frac{192}{263} = 73\%$	$\frac{55}{80} = 69\%$
together	$\frac{273}{350} = 78\%$	$\frac{289}{350} = 83\%$

Wikipedia, Julius + Mullee (1994)

In less severe cases (small kidney stones), doctors favor **percutaneous nephrolithotomy (which involves only a small puncture)**. Doctors tend to give the severe cases (large stones) the **better treatment (open surgery)**. The success rate is more strongly influenced by the severity of the case than by the choice of treatment. When the **less effective treatment (small puncture)** is applied more frequently to less severe cases, it can appear to be a **more effective treatment**.

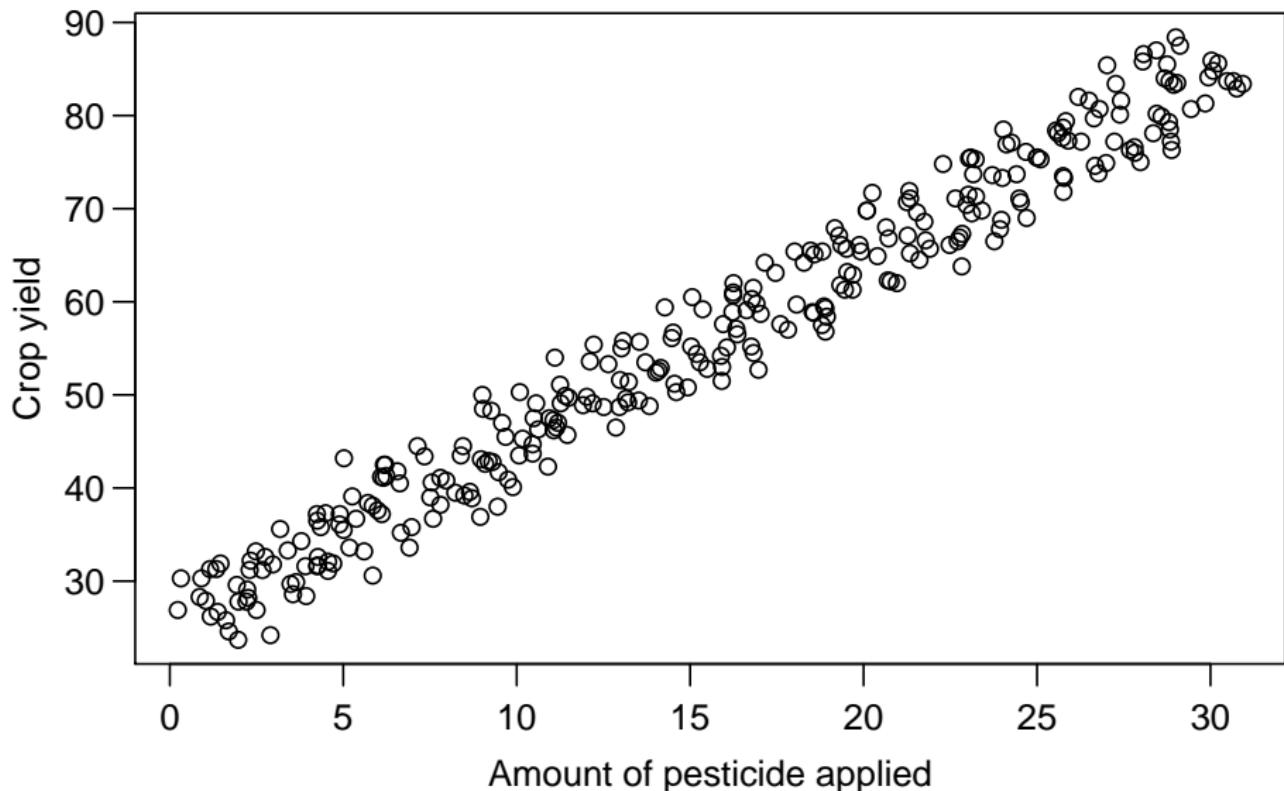
# Simpson's Exercise

## Exercise 43: Simpson

Visualize [farming.txt](#) (*rightclick Raw, save as*), a fictional dataset measuring crop yield on 30 study fields after pesticide application.

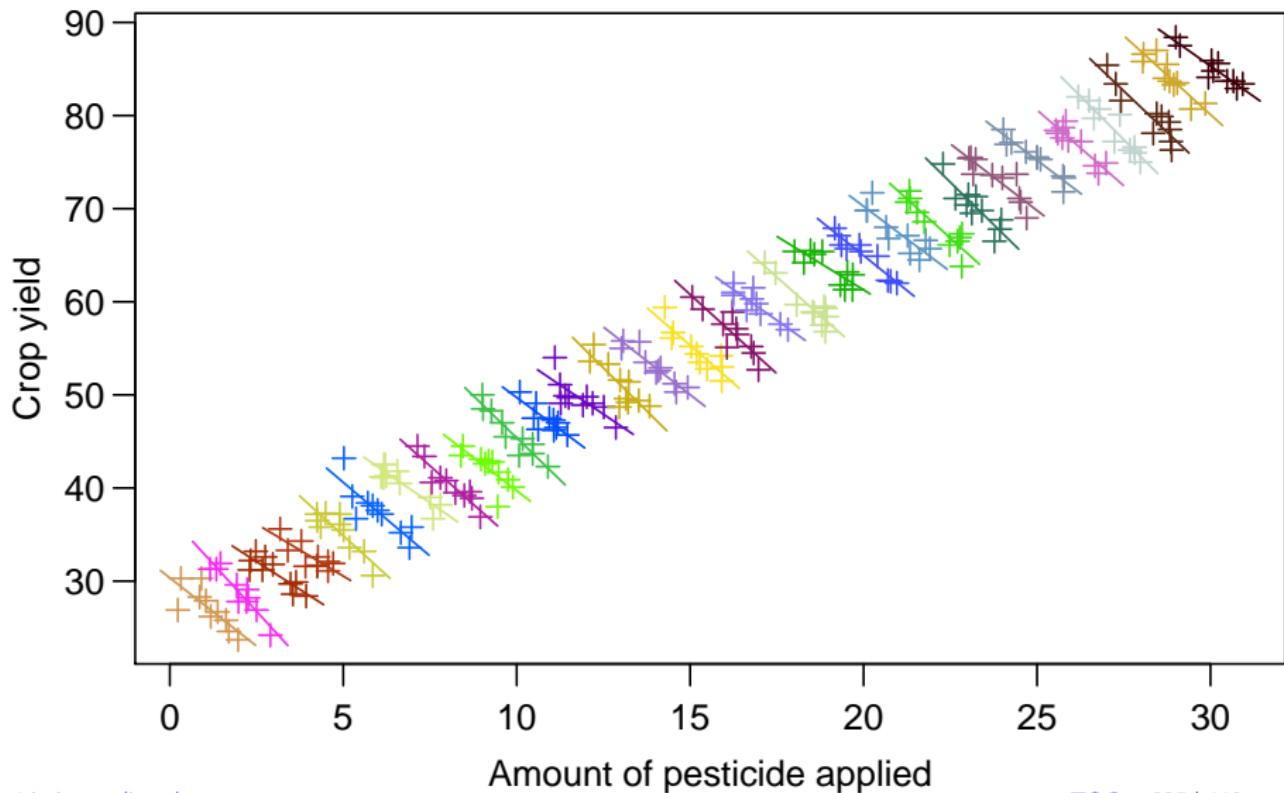
# Farming solution I

**Farmers applying pesticides can harvest more!**



## Farming solution II

On second thought: maybe they shouldn't.



# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  - 10. Statistics**
    - Simpsons Paradox
    - Normality tests**
    - Confidence Intervals
    - (Linear) regression
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Normality Test: Kolmogorov-Smirnov

## Normality Test: Kolmogorov-Smirnov

- If a population is normally distributed, it is described by only two parameters: the mean (position) and the sd (width, dispersion) of the bell shaped curve.

## Normality Test: Kolmogorov-Smirnov

- If a population is normally distributed, it is described by only two parameters: the mean (position) and the sd (width, dispersion) of the bell shaped curve.
- This is an important assumption for many classical statistical methods.

# Normality Test: Kolmogorov-Smirnov

- If a population is normally distributed, it is described by only two parameters: the mean (position) and the sd (width, dispersion) of the bell shaped curve.
- This is an important assumption for many classical statistical methods.
- Whether a dataset is normally distributed can be checked with a histogram (visually effective, but the class limits are subjective), with qqplots (I don't find them very intuitive), or with statistical tests.

# Normality Test: Kolmogorov-Smirnov

- If a population is normally distributed, it is described by only two parameters: the mean (position) and the sd (width, dispersion) of the bell shaped curve.
- This is an important assumption for many classical statistical methods.
- Whether a dataset is normally distributed can be checked with a histogram (visually effective, but the class limits are subjective), with qqplots (I don't find them very intuitive), or with statistical tests.

```
• data <- rnorm(1000, mean=97, sd=8.9)
  shapiro.test(data)
  ks.test(data, "pnorm", mean(data), sd(data))
  # if p > 0.05: accept Nullhypothesis that data are normally distributed.

  if(!requireNamespace("nortest")) install.packages("nortest")
  library(nortest)          # with Lilliefors-correction!
  help(package="nortest")
  lillie.test(data)          # Lilliefors (Kolmogorov-Smirnov) test for normality
  ad.test(data)              # Anderson-Darling test for normality
  cvm.test(data)             # Cramer-von Mises test for normality
  pearson.test(data)         # Pearson chi-square test for normality
  sf.test(data)              # Shapiro-Francia test for normality
```

## Normality Test: Kolmogorov-Smirnov

KS Hypothesis Test: Reject the null hypothesis ( $H_0$ ) that the values are drawn from a normal distribution if  $D < \frac{k_\alpha}{\sqrt{n}}$ .

## Normality Test: Kolmogorov-Smirnov

KS Hypothesis Test: Reject the null hypothesis ( $H_0$ ) that the values are drawn from a normal distribution if  $D < \frac{k_\alpha}{\sqrt{n}}$ .

$D = \max(|F_{(n)}(x) - F(x)|)$  Difference between *empirical* and *statistical (theoretical, population)* Cumulated Density Function (CDF)

## Normality Test: Kolmogorov-Smirnov

KS Hypothesis Test: Reject the null hypothesis ( $H_0$ ) that the values are drawn from a normal distribution if  $D < \frac{k_\alpha}{\sqrt{n}}$ .

$D = \max(|F_{(n)}(x) - F(x)|)$  Difference between *empirical* and *statistical (theoretical, population)* Cumulated Density Function (CDF)

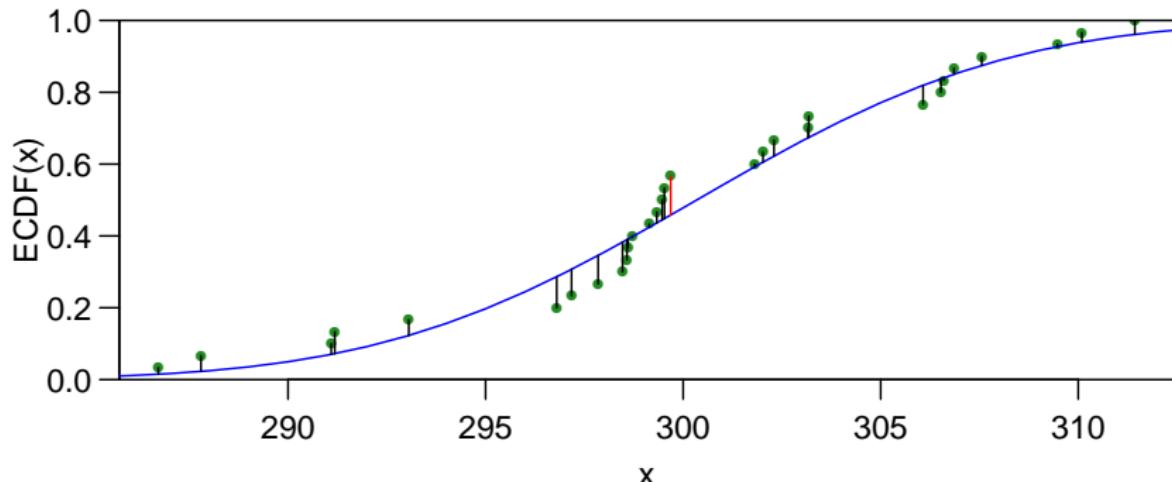
$$k_\alpha = \frac{\sqrt{-0.5 * \ln(\frac{\alpha}{2})}}{\sqrt{n}} \quad (\text{for } n > 35) \quad \alpha \text{ often } 0.05 \text{ (95% confidence)}$$

# Normality Test: Kolmogorov-Smirnov

KS Hypothesis Test: Reject the null hypothesis ( $H_0$ ) that the values are drawn from a normal distribution if  $D < \frac{k_\alpha}{\sqrt{n}}$ .

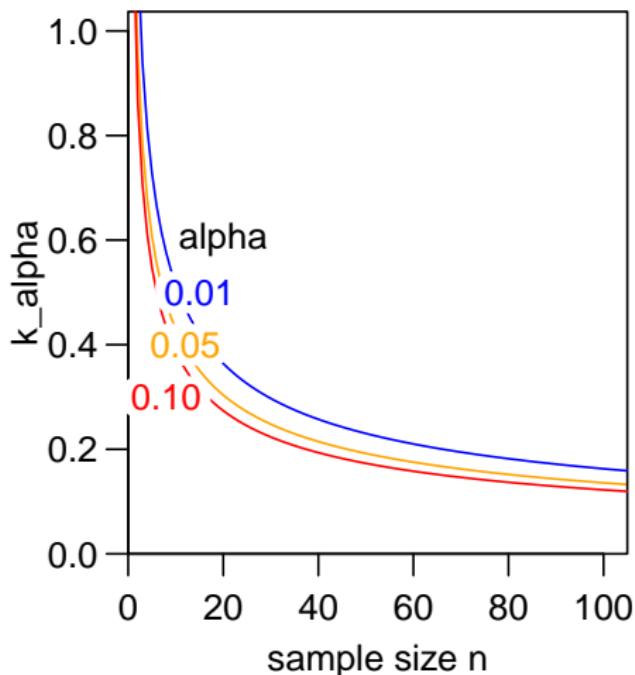
$D = \max(|F_{(n)}(x) - F(x)|)$  Difference between *empirical* and *statistical (theoretical, population)* Cumulated Density Function (CDF)

$$k_\alpha = \frac{\sqrt{-0.5 \cdot \ln(\frac{\alpha}{2})}}{\sqrt{n}} \quad (\text{for } n > 35) \quad \alpha \text{ often } 0.05 \text{ (95% confidence)}$$

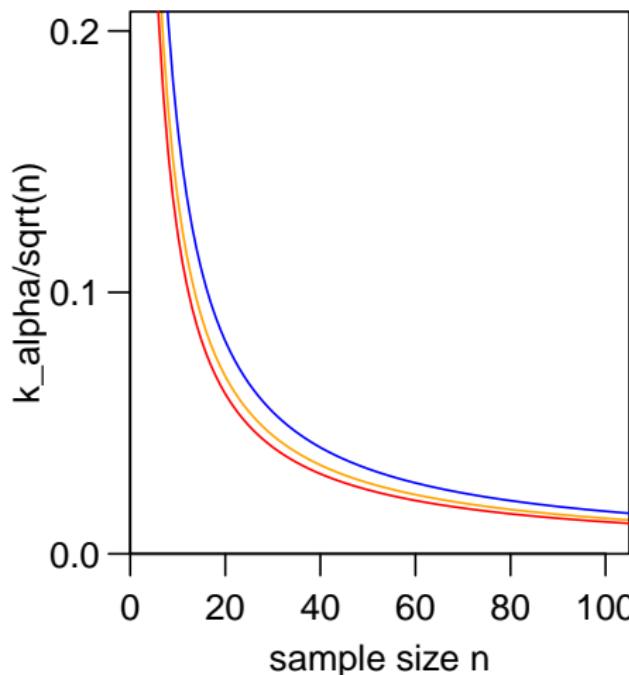


# Normality Test: Kolmogorov-Smirnov

KS-test critical value



comparison with D



# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  - 10. Statistics**
    - Simpsons Paradox
    - Normality tests
    - Confidence Intervals**
    - (Linear) regression
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Background confidence intervals

# Background confidence intervals

- About 70% of the values in a Gaussian distribution lie between  $\text{mean}-1\text{sd}$  and  $\text{mean}+1\text{sd}$ , as is shown by `pnorm(1) - pnorm(-1)`: 0.6826895.

# Background confidence intervals

- About 70% of the values in a Gaussian distribution lie between  $\text{mean}-1\text{sd}$  and  $\text{mean}+1\text{sd}$ , as is shown by `pnorm(1) - pnorm(-1)`: 0.6826895.
- 95% of the values are between  $\text{mean} \pm 1.96 \text{ sd}$ , as is shown by `qnorm(0.975)`: 1.959964. (As the bell curve is symmetrical, we want to know below what value 97.5% of the values are in the cumulated density).

# Background confidence intervals

- About 70% of the values in a Gaussian distribution lie between mean-1sd and mean+1sd, as is shown by `pnorm(1) - pnorm(-1)`: 0.6826895.
- 95% of the values are between  $\text{mean} \pm 1.96 \text{ sd}$ , as is shown by `qnorm(0.975)`: 1.959964. (As the bell curve is symmetrical, we want to know below what value 97.5% of the values are in the cumulated density).
- This is the famous parameter in calculating the limits of confidence intervals (CI).

$$CI = \mu \pm p * \frac{sd}{\sqrt{n}}$$

# Background confidence intervals

- About 70% of the values in a Gaussian distribution lie between mean-1sd and mean+1sd, as is shown by `pnorm(1) - pnorm(-1)`: 0.6826895.
- 95% of the values are between  $\text{mean} \pm 1.96 \text{ sd}$ , as is shown by `qnorm(0.975)`: 1.959964. (As the bell curve is symmetrical, we want to know below what value 97.5% of the values are in the cumulated density).
- This is the famous parameter in calculating the limits of confidence intervals (CI).

$$CI = \mu \pm p * \frac{sd}{\sqrt{n}}$$

$$p = qt(1-\text{alpha}/2, \text{ df=n-1})$$

## Parameter confidence intervals

[en.wikipedia.org/wiki/Student's\\_t-distribution#Table\\_of\\_selected\\_values](https://en.wikipedia.org/wiki/Student's_t-distribution#Table_of_selected_values)

For a two-sided confidence interval with  $qt(1-\alpha/2, df=n-1)$ :

	a=0.01	a=0.05	a=0.10
n=10	3.2498	2.2622	1.8331
n=30	2.7564	2.0452	1.6991
n=50	2.6800	2.0096	1.6766
n=100	2.6264	1.9842	1.6604
n=1000	2.5808	1.9623	1.6464

## Parameter confidence intervals

[en.wikipedia.org/wiki/Student's\\_t-distribution#Table\\_of\\_selected\\_values](https://en.wikipedia.org/wiki/Student's_t-distribution#Table_of_selected_values)

For a two-sided confidence interval with  $qt(1-\alpha/2, df=n-1)$ :

	a=0.01	a=0.05	a=0.10
n=10	3.2498	2.2622	1.8331
n=30	2.7564	2.0452	1.6991
n=50	2.6800	2.0096	1.6766
n=100	2.6264	1.9842	1.6604
n=1000	2.5808	1.9623	1.6464

code for generating this table:

```
alpha <- c(0.01, 0.05, 0.10)
n <- c(10, 30, 50, 100, 1000)
ttab <- t(sapply(n, function(x) qt(1-alpha/2 ,df=x-1)))
colnames(ttab) <- paste0("a=",alpha)
rownames(ttab) <- paste0("n=",n)
ttab
```

# CI example

## Exercise 44: Confidence intervals

- ① Comment the code below.
- ② plot the points along a number line and add the mean value with `points` and the confidence interval with `arrows` with `angle=90`, `code=3`. BONUS: plot the points as filled dots in a semitransparent color, see `rgb`.

```
values <- rnorm(80, 53, 4)
m <- mean(values)      ; m #
v <- var(values)       ; v #
n <- length(values)   ; n #
p <- qnorm(0.975)      ; p #
p <- qt(0.975, df=n-1) ; p #
cl <- m - p * sqrt(v/n) ; cl #
cu <- m + p * sqrt(v/n) ; cu #
```

# CI in a faster way

Scripting this manually needs several LOC (lines of code):

```
m <- mean(values)
v <- var(values)
n <- length(values)
p <- qt(0.975, df=n-1)

m + c(-1,1)* p * sqrt(v/n)

## [1] 52.34151 54.00713
```

# CI in a faster way

Scripting this manually needs several LOC (lines of code):

```
m <- mean(values)
v <- var(values)
n <- length(values)
p <- qt(0.975, df=n-1)

m + c(-1,1)* p * sqrt(v/n)

## [1] 52.34151 54.00713
```

You can use the side effect of `t.test`:

```
t.test(values, conf.level=.95)$conf.int

## [1] 52.34151 54.00713
## attr("conf.level")
## [1] 0.95
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  - 10. Statistics**
    - Simpsons Paradox
    - Normality tests
    - Confidence Intervals
    - (Linear) regression**
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Linear Regression

This section (linear models) was originally written by Matthias Seibert (GFZ).

It is excluded from this main document and can be obtained as [pdf](#).

# Linear Regression

[Storks Deliver Babies \( \$p = 0.008\$ \)](#)

Fit multiple functions: `berryFunctions::mReg`. See the `mReg` sections in the [Vignette](#) and [Rclick](#) page

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  - 11. Time series handling and analysis**
    - Date and time formatting
    - Time series
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

# Char to date / time and back

## Exercise 45: char time

- ① Convert the character string "2017-11-29 13:41:16" to a time object with `?strptime` (string parsed to time). Check the `structure`.
- ② BONUS: reflect the local or GMT time zone.
- ③ `?paste` the following charstrings together "14.2.2016" and "17:18", separated with a minus symbol.
- ④ Convert the result with `strptime`. Convert it to a number (what does it mean?) and with `as.POSIXct`.
- ⑤ BONUS: format the current `Sys.Date()` as "Mi, 29.Nov.2017 (= ein Mittwoch im November)". On what weekday were you born?
- ⑥ BONUS: Print the names in English, even if the locale is German. See Rclick chapter 13 on `?Sys.setlocale`.

# Char - Time Solutions

```
now <- strptime("2017-11-29 13:41:16", format="%Y-%m-%d %H:%M:%S")
now ; class(now); str(now)

## [1] "2017-11-29 13:41:16 CET"
## [1] "POSIXlt" "POSIXt"
## POSIXlt[1:1], format: "2017-11-29 13:41:16"

strptime("2017-11-29 13:41:16", format="%F %T", tz="UTC")

## [1] "2017-11-29 13:41:16 UTC"
```

# Char - Time Solutions

```
now <- paste("14.2.2016", "17:19", sep="-")
now <- strptime(now, "%d.%m.%Y-%H:%M")
as.numeric(now) # number of seconds since 1970-01-01

## [1] 1455466740

str(as.POSIXct(now))

## POSIXct[1:1], format: "2016-02-14 17:19:00"

format(Sys.Date(), "%a, %d.%b.%Y (= some %A in %B)")

## [1] "Tue, 05.Dec.2017 (= some Tuesday in December)"

format(as.Date("1997-04-13"), "%A") # Born on Sunday

## [1] "Sunday"
```

# Time sequences

## Exercise 46: equally spaced time sequences

- ① Create a weekly sequence with `seq()` starting

```
t1 <- as.Date("2007/10/18") # and ending at  
t2 <- as.Date("2007/11/25")
```

- ② BONUS: How about monthly sequences? (by = number of days not helpful)

- ③ From the time range "tr", create a sequence in steps of 15 minutes:

```
tr <- strptime(c("05.08.2010, 12:15",  
                 "09.02.2013, 14:45"),  
                 format="%d.%m.%Y, %H:%M")
```

- ④ Create a `data.frame` with that time sequence and values from a random walk (see `rnorm` and `cumsum`). Both columns should be named aptly.

- ⑤ BONUS: Correct the by argument in `seq(as.Date("2017-08-30"), as.Date("2017-06-30"), by=7)`.

# Time sequence solutions

```
seq(t1, t2, by=7)

## [1] "2007-10-18" "2007-10-25" "2007-11-01"
## [4] "2007-11-08" "2007-11-15" "2007-11-22"

seq(t1,t2+150, by="month") # uses seq.Date

## [1] "2007-10-18" "2007-11-18" "2007-12-18"
## [4] "2008-01-18" "2008-02-18" "2008-03-18"
## [7] "2008-04-18"

tseq <- seq(tr[1], tr[2], by="15 min")
fakeTS <- data.frame(time=tseq,
                      values=cumsum(rnorm(length(tseq))))
)
# str(fakeTS)
```

# Time Series aggregation

## Exercise 47: TS aggregation

- ① Plot the random walk time series from the previous task. BONUS: label the axis nicely with `berryFunctions::monthAxis()`
- ② Add a column uniquely identifying the month of each data point.
- ③ Compute monthly averages with `?tapply`
- ④ BONUS: Browse through Rclick chapter 13

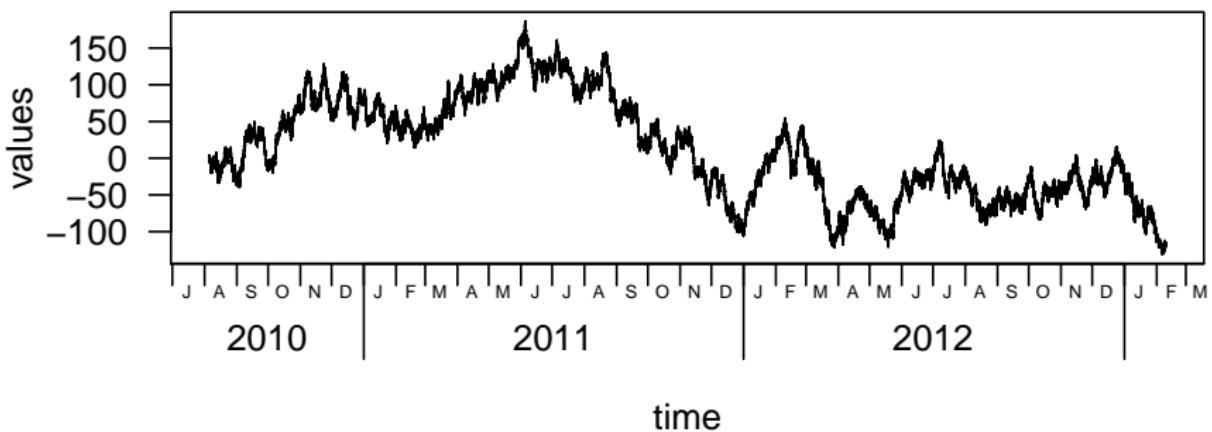
# TS aggregation solutions

```
plot(fakeTS, type="l", xaxt="n", las=1)
berryFunctions::monthAxis(mcex=0.5)
#
fakeTS$month <- format(fakeTS$time, "%Y-%m-15")
agg_month <- tapply(X=fakeTS$values, INDEX=fakeTS$month,
                     FUN=mean)
# compute mean of values, split into month groups

head(agg_month) # View(data.frame(agg_month))

## 2010-08-15 2010-09-15 2010-10-15 2010-11-15
## -8.496484 15.601002 35.889134 87.180137
## 2010-12-15 2011-01-15
## 76.125288 58.158087
```

# TS aggregation solutions



# Task

# Task

Your professor asks you to visually examine temperature and precipitation trends at certain meteorological stations in Germany. She hands you the files [Potsdam.txt](#) and [Zugspitze.txt](#) (*rightclick Raw, save as*). How do you approach the project?

# Task

Your professor asks you to visually examine temperature and precipitation trends at certain meteorological stations in Germany. She hands you the files `Potsdam.txt` and `Zugspitze.txt` (*rightclick Raw, save as*). How do you approach the project?

Data selection, download, processing and saving:

```
library(rdwd) # github.com/brry/rdwd
links <- selectDWD(c("Potsdam", "Zugspitze"), current=TRUE,
                     res="monthly", var="kl", per="hist", outvec=TRUE)
clim <- dataDWD(links)
write.table(clim[[1]], file="data/Potsdam.txt", row.names=F, quote=F)
write.table(clim[[2]], file="data/Zugspitze.txt", row.names=F, quote=F)
```

# Task

Your professor asks you to visually examine temperature and precipitation trends at certain meteorological stations in Germany. She hands you the files [Potsdam.txt](#) and [Zugspitze.txt](#) (*rightclick Raw, save as*). How do you approach the project?

Data selection, download, processing and saving:

```
library(rdwd) # github.com/brry/rdwd
links <- selectDWD(c("Potsdam", "Zugspitze"), current=TRUE,
                     res="monthly", var="kl", per="hist", outvec=TRUE)
clim <- dataDWD(links)
write.table(clim[[1]], file="data/Potsdam.txt", row.names=F, quote=F)
write.table(clim[[2]], file="data/Zugspitze.txt", row.names=F, quote=F)

potsdam$date <- as.Date(as.character(potsdam$MESS_DATUM_BEGINN), "%Y%m%d")
```

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  - 11. Time series handling and analysis**
    - Date and time formatting
    - Time series**
  12. Spatial data and GIS functionality
  13. Final tasks
- Course plan

## Authorship note

This section (time series) was originally written by Matthias Seibert (GFZ). It is currently excluded from this main document but can be obtained as [pdf](#).

```
library(xts)
```

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
  - Geostatistics: Kriging
  - R as GIS
  - NETcdf files
13. Final tasks
- Course plan

# Kriging: Overview

Geostatistics: Kriging - spatial interpolation between points, using semivariance

- ① Packages
- ② read shapefile
- ③ Variogram
- ④ Kriging
- ⑤ Plotting

This material has also been featured by Packt publishing:

[datahub.packtpub.com/analytics/kriging-interpolation-geostatistics](http://datahub.packtpub.com/analytics/kriging-interpolation-geostatistics)

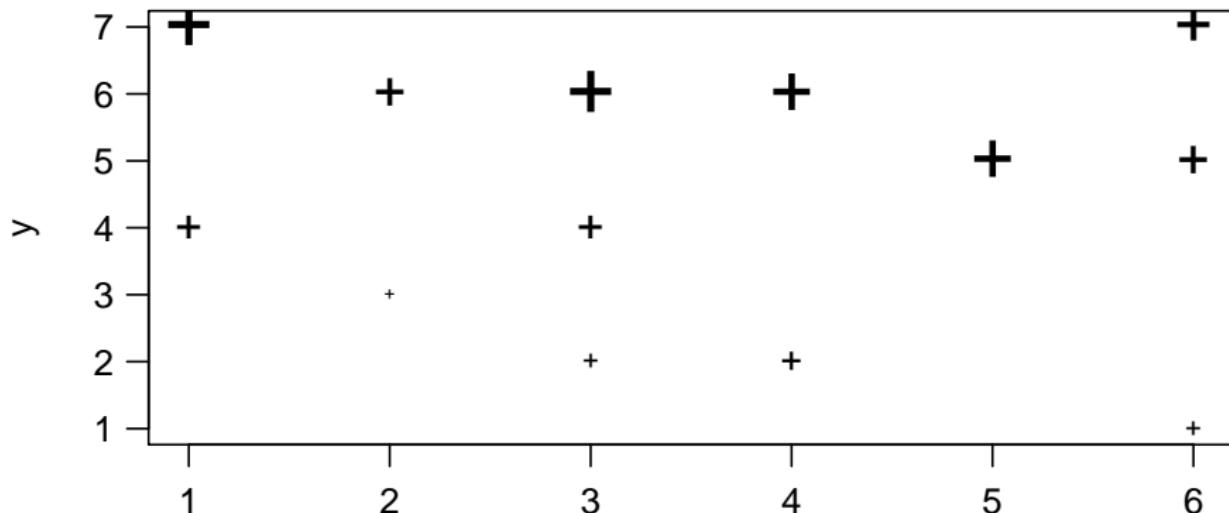
## Kriging: packages

```
install.packages("rgeos")
install.packages("sf")
install.packages("geoR")
```

```
library(sf)      # for st_read (read shapefiles),
                  # st_centroid, st_area, st_union
library(geoR)    # as.geodata, variog, variofit,
                  # krige.control, krige.conv, legend.krige
```

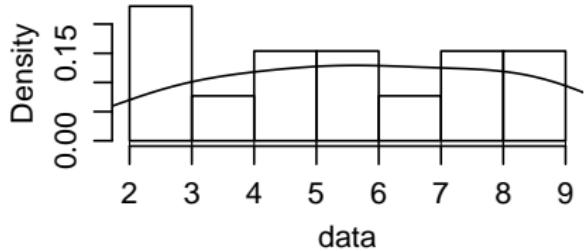
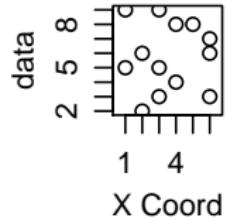
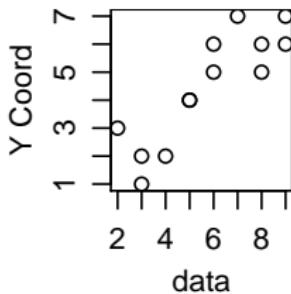
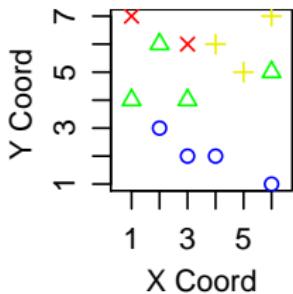
## Kriging: read shapefile / few points for demonstration

```
x <- c(1,1,2,2,3,3,3,4,4,5,6,6,6)  
y <- c(4,7,3,6,2,4,6,2,6,5,1,5,7)  
z <- c(5,9,2,6,3,5,9,4,8,8,3,6,7)  
plot(x,y, pch="+", cex=z/4)
```



# Kriging: read shapefile II

```
GEODATA <- as.geodata(cbind(x,y,z))  
plot(GEODATA)
```



# Kriging: Variogram I

```
EMP_VARIOGRAM <- variog(GEODATA)

## variog: computing omnidirectional variogram

FIT_VARIOGRAM <- variofit(EMP_VARIOGRAM)

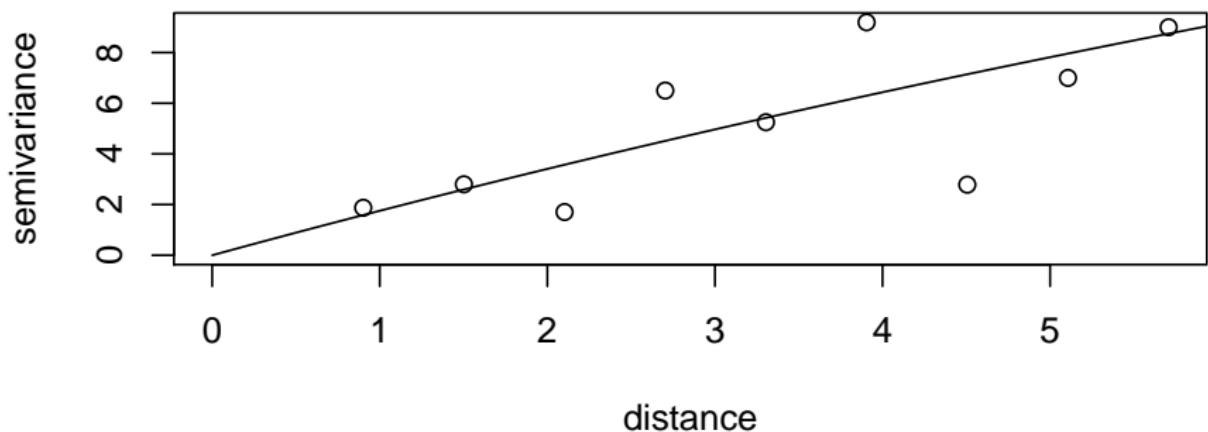
## variofit: covariance model used is matern
## variofit: weights used: npairs
## variofit: minimisation function used: optim

## Warning in variofit(EMP_VARIOGRAM): initial values not provided - running the
## default search

## variofit: searching for best initial value ... selected values:
##           sigmasq   phi   tausq   kappa
## initial.value "9.19"  "3.65"  "0"    "0.5"
## status        "est"   "est"   "est"   "fix"
## loss value: 401.578968904954
```

## Kriging: Variogram II

```
plot(EMP_VARIOGRAM)  
lines(FIT_VARIOGRAM)
```



# Kriging: Kriging

```
res <- 0.1
grid <- expand.grid(seq(min(x),max(x),res),
                     seq(min(y),max(y),res))
krico <- krig.control(type.krige="OK",
                       obj.model=FIT_VARIOGRAM)
krobj <- krig.conv(GEODATA,
                    locations=grid, krig=krico)

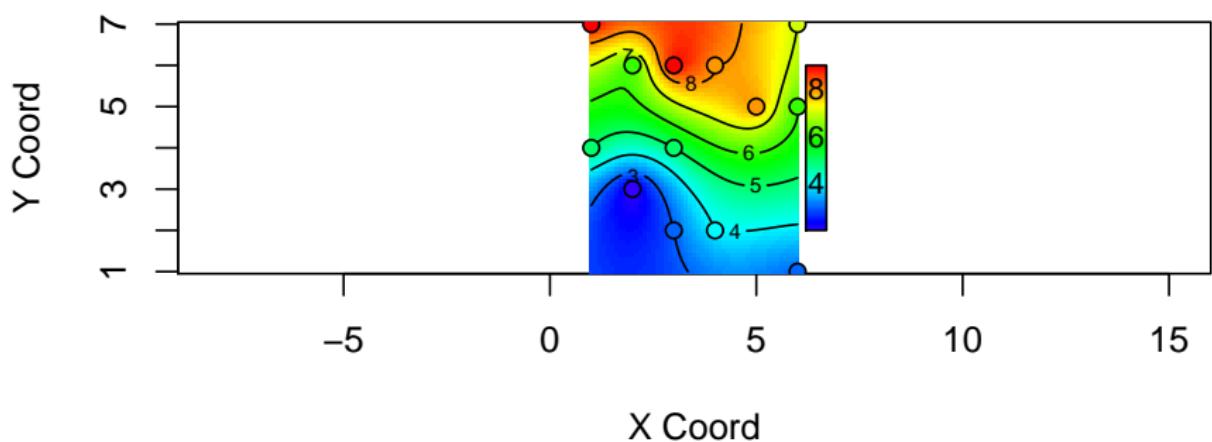
## krig.conv: model with constant mean
## krig.conv: Kriging performed using global neighbourhood

# KRigingObjekt
```

# Kriging: Plotting I

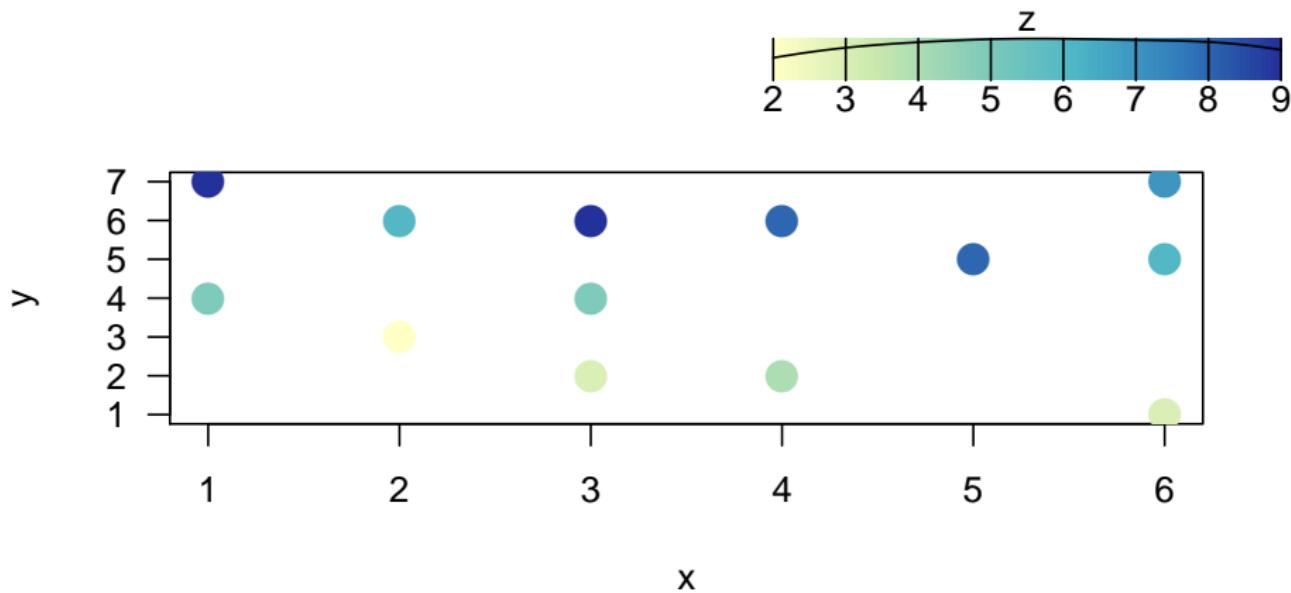
```
image(krobj, col=rainbow2(100))
legend.krige(col=rainbow2(100),
             x.leg=c(6.2,6.7), y.leg=c(2,6),
             vert=T, off=-0.5,
             values=krobj$predict)
contour(krobj, add=T)
colPoints(x,y,z, col=rainbow2(100), legend=F)
points(x,y)
```

# Kriging: Plotting II



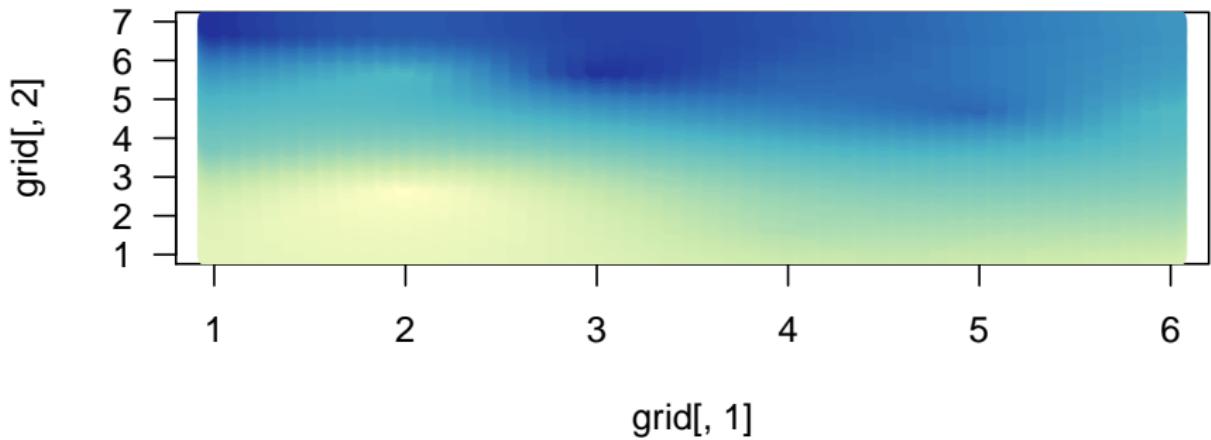
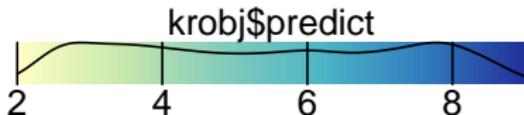
# Kriging: Plotting III

```
library("berryFunctions") # scatterpoints by color  
colPoints(x,y,z, add=F, cex=2, legargs=list(y1=0.8,y2=1))
```



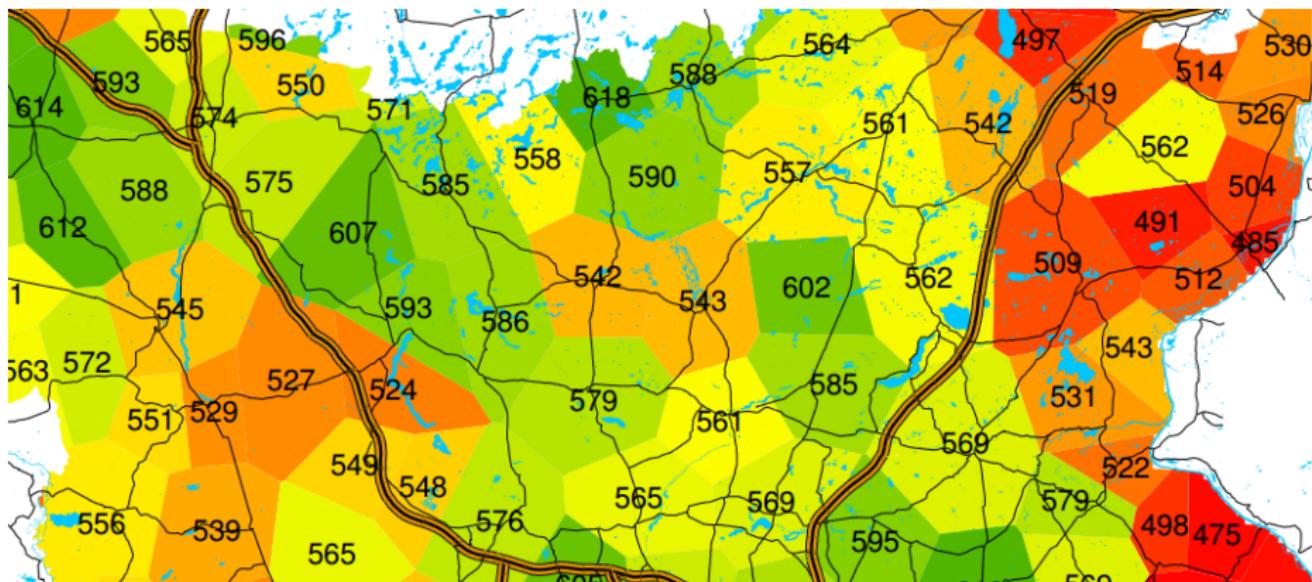
## Kriging: Plotting IV

```
colPoints(grid[,1], grid[,2], krobj$predict, add=F,  
         cex=2, col2=NA, legargs=list(y1=0.8,y2=1))
```



## Time for a real dataset

Precipitation from ca 250 gauges in Brandenburg as Thiessen Polygons with steep gradients at edges:



# Time for a real dataset

## Exercise 48: Kriging

- ① Load and plot the shapefile in `PrecBrandenburg.zip` with `sf::st_read`.
- ② With `colPoints` in the package `berryFunctions`, add the precipitation values at the centroids of the polygons.
- ③ Calculate the variogram and fit a semivariance curve.
- ④ Perform kriging on a grid with a useful resolution (keep in mind that computing time rises exponentially with grid size).
- ⑤ Plot the interpolated values with `image` or an equivalent (Rclick 4.15) and add contour lines.
- ⑥ What went wrong? (if you used the defaults, the result will be dissatisfying.) How can you fix it?

# Solution for exercise 48.1-2: Kriging Data

```
# Shapefile:  
p <- sf::st_read("data/PrecBrandenburg/niederschlag.shp",  
                  quiet=TRUE)  
  
# Plot prep  
pcol <- colorRampPalette(c("red", "yellow", "blue"))(50)  
clss <- berryFunctions::classify(p$P1, breaks=50)$index  
  
# Plot  
par(mar = c(1, 1, 1.2, 1)) # for sf plots / colPointsLegend  
sf:::plot.sf(p, col=pcol[clss], max.plot=1) # P1: Precipitation  
  
# kriging coordinates  
cent <- sf::st_centroid(p)  
berryFunctions::colPoints(cent$x, cent$y, p$P1, add=T, cex=0.7,  
                         legargs=list(y1=0.8,y2=1), col=pcol)  
points(cent$x, cent$y, cex=0.7)
```

# Solution for exercise 48.3: Variogram

```
library(geoR)
# Semivariance:
geoprec <- as.geodata(cbind(cent$x,cent$y,p$P1))
vario <- variog(geoprec, max.dist=130000)

## variog: computing omnidirectional variogram

fit <- variofit(vario)

## variofit: covariance model used is matern
## variofit: weights used: npairs
## variofit: minimisation function used: optim

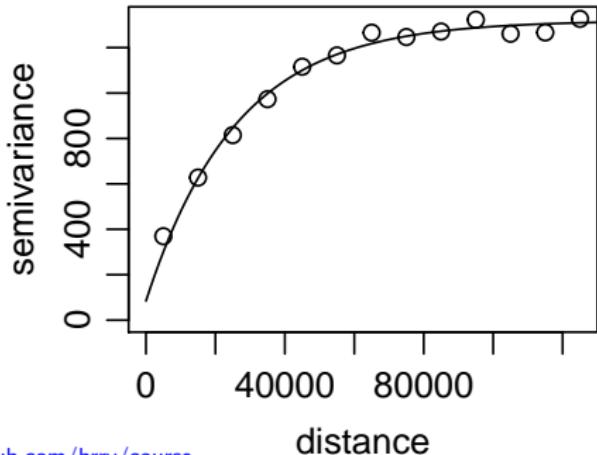
## Warning in variofit(vario): initial values not provided - running the default
## search

## variofit: searching for best initial value ... selected values:
##           sigmasq   phi      tausq kappa
## initial.value "1326.72" "19999.93" "0"    "0.5"
## status        "est"     "est"     "est"   "fix"
## loss value: 107266266.76371
```

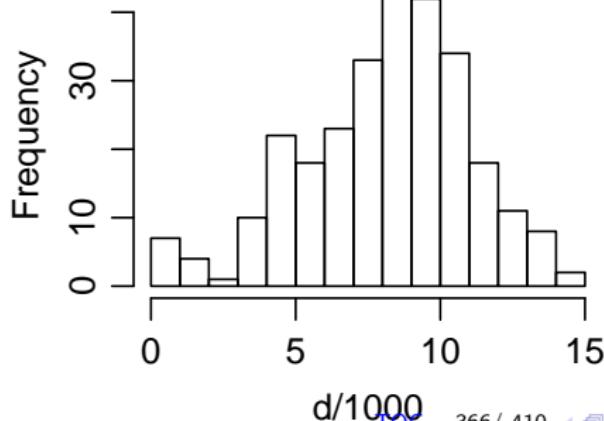
# Solution for exercise 48.3: Variogram

```
plot(vario) ; lines(fit)
# distance to closest other point:
d <- sapply(1:nrow(cent), function(i) min(berryFunctions::distance(
    cent$x[i], cent$y[i], cent$x[-i], cent$y[-i])))
hist(d/1000, breaks=20, main="distance to closest gauge [km]")
mean(d) # 8 km - grid resolution in next slide

## [1] 8165.633
```



distance to closest gauge [l]



# Solution for exercise 48.4-5: Kriging

# Kriging:

```
res <- 1000 # 1 km, since stations are 8 km apart on average
grid <- expand.grid(seq(min(cent$x),max(cent$x),res),
                     seq(min(cent$y),max(cent$y),res))
krico <- krige.control(type.krige="OK", obj.model=fit)
krobj <- krige.conv(geoprec, locations=grid, krige=krico)
```

## krige.conv: model with constant mean

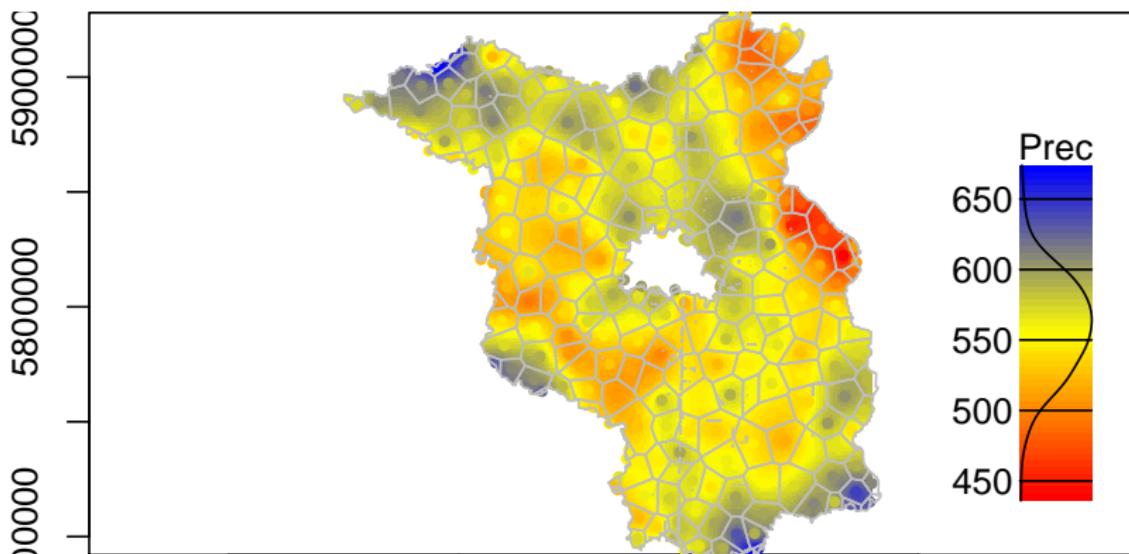
## krige.conv: Kriging performed using global neighbourhood

# Set values outside of Brandenburg to NA:

```
grid_sf <- sf::st_as_sf(grid, coords=1:2, crs=sf::st_crs(p))
isinp <- sapply(sf::st_within(grid_sf, p), length) > 0
krobj2 <- krobj
krobj2$predict[!isinp] <- NA
```

# Solution for exercise 48.5: Kriging Visualization

```
geoR:::image.kriging(krobj2, col=pcol)
colPoints(cent$x, cent$y, p$P1, col=pcol, zlab="Prec", cex=0.7,
          legargs=list(y1=0.1,y2=0.8, x1=0.78, x2=0.87, horizontal=F))
plot(p, add=T, col=NA, border=8) #; points(cent$x, cent$y, cex=0.8)
```



# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
- 12. Spatial data and GIS functionality**
  - Geostatistics: Kriging
  - R as GIS
  - NETcdf files
13. Final tasks
- Course plan

interactive shapefile editing - <http://r-spatial.org>

interactive shapefile editing - <http://r-spatial.org>

This is R. There is no if. Only how.

# interactive shapefile editing - <http://r-spatial.org>

This is R. There is no if. Only how.

```
# Data and Packages:  
download.file(  
  "https://github.com/brry/course/raw/master/data/PrecBrandenburg.zip",  
  destfile="prec.zip")  
unzip("prec.zip")  
  
library(sf)  
library(mapview)  
library(mapedit)  
  
p <- sf:::st_read("data/PrecBrandenburg/niederschlag.shp")  
map <- mapview::mapview(p)  
map  
  
p2 <- mapedit::editMap(map) # add polygons, lines, points, rectangles  
mapview::mapview(p2$finished)  
names(p)[2] = "x1" # x column problem, see Issue 2017-12-05  
p3 <- mapedit::editFeatures(p)  
?franconia # from mapview package  
ed <- mapedit::editFeatures(franconia[1:5,])
```

# shapefile operations

`sf` replaces the old `sp` package since 2017 (`sp` will be deprecated).

Vignettes:

1. intro: simple features, geometry, reading/writing, projections, units
2. reading/writing, conversion
3. transformation, coordinate reference systems, geometrical operations
4. subsetting, aggregation, joining by attribute / geometry
5. plotting, projections
6. miscellaneous

## point data (coordinates) - leaflet

```
ddd <- read.table(header=TRUE, sep=",", as.is=TRUE, text="  
lat, lon, city, LOA # LOA: Level of Awesomeness  
48.858271, 2.294245, Paris , 5  
51.503162, -0.131082, London, 3  
52.514687, 13.350012, Berlin, 6  
52.407160, 12.972761, Golm , 9  
")  
ddd$display <- berryFunctions::popleaf(ddd)
```

## point data (coordinates) - leaflet

```
ddd <- read.table(header=TRUE, sep=",", as.is=TRUE, text="  
lat, lon, city, LOA # LOA: Level of Awesomeness  
48.858271, 2.294245, Paris , 5  
51.503162, -0.131082, London, 3  
52.514687, 13.350012, Berlin, 6  
52.407160, 12.972761, Golm , 9  
")  
ddd$display <- berryFunctions::popleaf(ddd)
```

Interactive plots:

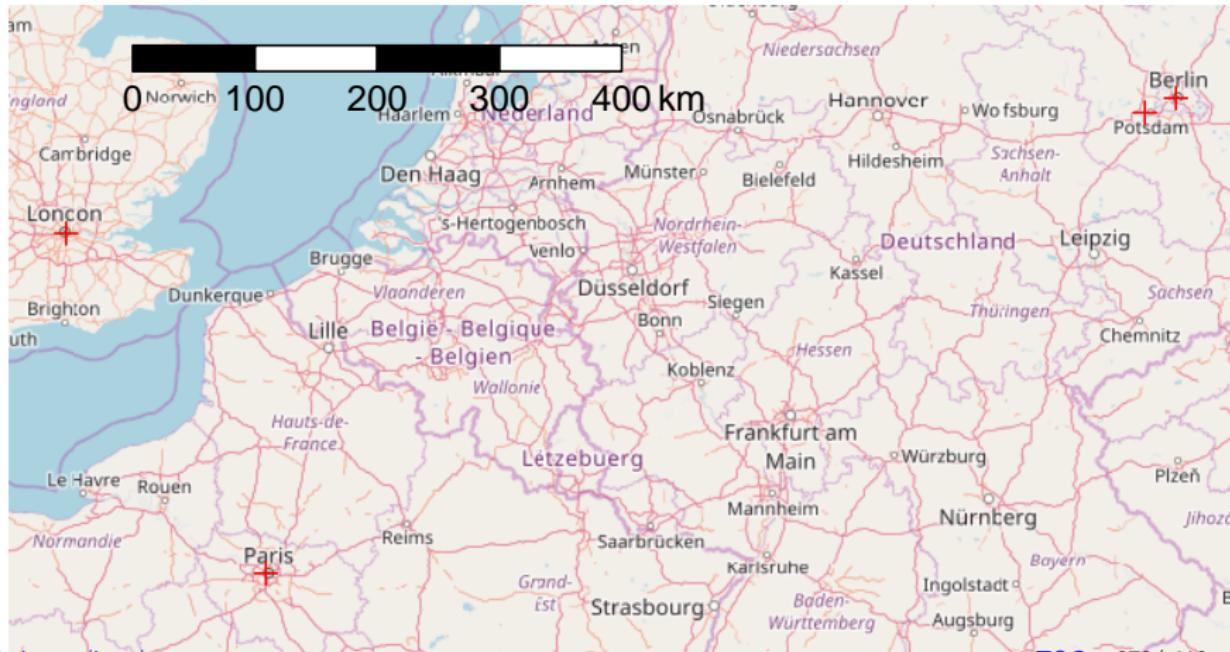
```
library(leaflet)  
leaflet(ddd) %>% addTiles() %>%  
  addCircleMarkers(~lon, ~lat, popup=~display)
```

Different background maps

# point data (coordinates) - OSMscale

Static maps:

```
library(OSMscale)  
pointsMap(lat, lon, data=ddd, quiet=TRUE)
```



# R geo, further material

Use the `sf`, `mapview`, `osmar` and `osmplotr` packages!

Combine city names with geolocations:

[blog.revolutionanalytics.com/2015/11/marriott.html](http://blog.revolutionanalytics.com/2015/11/marriott.html)

For this section (spatial data / GIS), a lot was originally written by Matthias Seibert (GFZ). With the advance of the `sf` package (late 2016), that material is mostly outdated. Until new material is created here, his slides can be obtained as [pdf](#).

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
- 12. Spatial data and GIS functionality**
  - Geostatistics: Kriging
  - R as GIS
  - NETcdf files**
13. Final tasks
- Course plan

read nc files: ncdf4::nc\_open

Install mhmVis with the instructions on github:

[github.com/brry/mhmVis](https://github.com/brry/mhmVis)

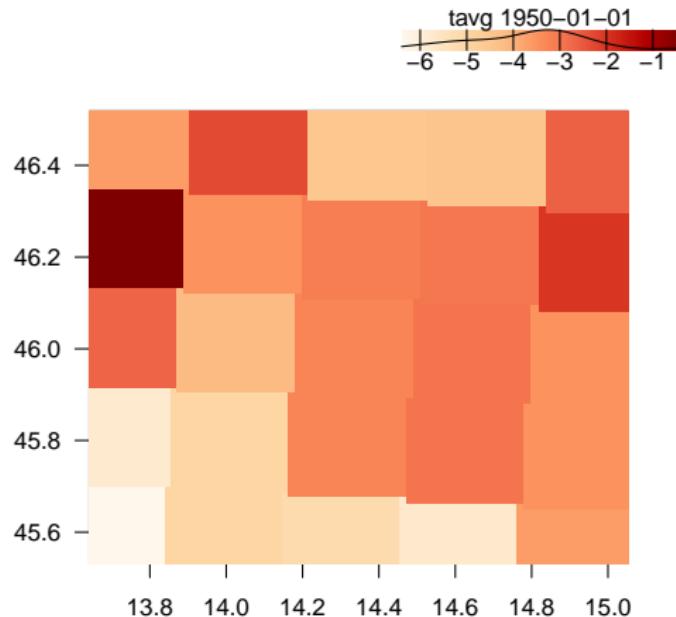
Get the example file: [tavg.zip](#)

```
unzip("data/tavg.zip", exdir="data")
library(mhmVis)
nc <- read_nc("data/tavg.nc")

## nc file has been read. Now extracting lat, lon, tavg.
This may take a minute.
## I'm done! (took 0.54 secs)
```

## visualize nc files: `mhmVis::vis_nc`

```
vis_nc(nc, cex=11.5)
```



# visualize nc files

`mhmVis::vis_nc_all` and `mhmVis::vis_nc_film`

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality

## 13. Final tasks

Practice tasks

Exam task

Course plan

## Exercise 49: Final exercises to solve at home I

- ① What is the difference between `c(5,3,-8)` and `vec <- c(5,3,-8)`?
- ② What methods do you know to generate a vector?
- ③ What are object, element, command, argument, assignment, Workspace, Working Directory?
- ④ What data and object types do you know?
- ⑤ How do you call the help about the `lines` command?
- ⑥ How do you read a text file? What are the main arguments of the function?
- ⑦ How do you examine the imported table?
- ⑧ How do you obtain / change subsets of the table (column, row, element)?

## Exercise 50: Final exercises to solve at home II

- ① How often is Peter Pan mentioned in the E-book? [`peterpan.txt`](#) (*rightclick Raw, save as*). Exclude the meta data about the book!
- ② Interpret the results from `pairs(iris)` and `cor(iris[, 1:4])`. Why does [`cor\(iris\)`](#) not work?
- ③ What are the most important functions and arguments for creating graphics?
- ④ Write a for loop including a while statement to distribute points randomly, but with a certain minimal distance to each other, along a line (segment sampling design). A solution is in my [`RclickHandbuch`](#) Chapter 10.4.
- ⑤ What functions and packages are useful for dealing with time series?

# Outline

R course info

1. One-Session-Intro
  2. Getting started: Background, GUI and first steps
  3. Objects
  4. "Real" data
  5. Plotting
  6. Character Operations
  7. Conditions & loops
  8. Functions & packages
  9. Distributions
  10. Statistics
  11. Time series handling and analysis
  12. Spatial data and GIS functionality
  - 13. Final tasks**
    - Practice tasks
    - Exam task**
- Course plan

## Exercise 51: Final exam

- ① Download a long time series (e.g. weather data from DWD, NOAA), see [slide 141](#).
- ② Read the data into R and control the data types. Change the date (or time) to appropriate types with `as.Date`, `as.POSIXct` or `strptime`.
- ③ Write a subsection of the data (eg your birth year) to a tab-separated file so it's easy to copypaste this to Excel/Calc.
- ④ Plot the development of some aggregate (e.g. monthly 90% Quantile of windspeed, annual precipitation sum or temperature mean depending on [hour of the day](#)).  
Hints: `tapply`, `strftime`, `berryFunctions::monthLab`
- ⑤ Examine [correlations](#) between variables.
- ⑥ Write a small report on your data analysis project with max 3 pages, including (beautiful) graphs. You can mail the code in a separate file.  
**This is a requirement for the participation certificate with credit points.**
- ⑦ BONUS 1: Write the report via Rnw, Rhtml, Rmd or Rpres (see [slide 140](#)).
- ⑧ BONUS 2: Make some interactive stuff (e.g. user can choose decade of analysis), e.g. via [HTMLwidgets](#) or [Shiny \(see more\)](#). This is a lot of initial effort, but you'll enable an awesome interactive graphical analyses!

# Outline

## R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

## Course plan

Session 1/4: Why R, read file, select and plot data

Session 2/4: Objects, file reading, Rstudio

Session 3/4: if-else conditionals, for loops

Session 4/4: functions and apply loops

Session x: Recap all

Feedback, additional material

# Session 1: Aim (Chapter 1)

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

afterwards, you should be able to:

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

afterwards, you should be able to:

- start a new script in Rstudio and execute lines of code

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

afterwards, you should be able to:

- start a new script in Rstudio and execute lines of code
- read a data file into R

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

afterwards, you should be able to:

- start a new script in Rstudio and execute lines of code
- read a data file into R
- select a column and rows that fullfill a certain criterion

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

afterwards, you should be able to:

- start a new script in Rstudio and execute lines of code
- read a data file into R
- select a column and rows that fullfill a certain criterion
- generate basic graphics with annotation and added elements

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

afterwards, you should be able to:

- start a new script in Rstudio and execute lines of code
- read a data file into R
- select a column and rows that fullfill a certain criterion
- generate basic graphics with annotation and added elements
- open and read the documentation for R functions

# Session 1: Aim (Chapter 1)

get a general impression on how R works, without going too much into technical detail

afterwards, you should be able to:

- start a new script in Rstudio and execute lines of code
- read a data file into R
- select a column and rows that fullfill a certain criterion
- generate basic graphics with annotation and added elements
- open and read the documentation for R functions

[Jump to Rstudio](#)

# Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with >, waiting with +

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with >, waiting with +
- Script for code writing (you) + exchange

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with >, waiting with +
- Script for code writing (you) + exchange
- Objects are created with the assignment operator <-

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1
- Create vectors with `c(42,3,-4)`, `1:n`

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1
- Create vectors with `c(42,3,-4)`, `1:n`
- `data.frames` hold tables with one data type per column

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1
- Create vectors with `c(42,3,-4)`, `1:n`
- `data.frames` hold tables with one data type per column
- `read.table`, with `header=TRUE` to read files into R

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1
- Create vectors with `c(42,3,-4)`, `1:n`
- `data.frames` hold tables with one data type per column
- `read.table`, with `header=TRUE` to read files into R
- Index objects with `values[8:5]`, `tab[4,2:6]`

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1
- Create vectors with `c(42,3,-4)`, `1:n`
- `data.frames` hold tables with one data type per column
- `read.table`, with `header=TRUE` to read files into R
- Index objects with `values[8:5]`, `tab[4,2:6]`
- df columns with `tab[ ,3]`, `tab[ , "precip"]`, `tab$wind`

## Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1
- Create vectors with `c(42,3,-4)`, `1:n`
- `data.frames` hold tables with one data type per column
- `read.table`, with `header=TRUE` to read files into R
- Index objects with `values[8:5]`, `tab[4,2:6]`
- df columns with `tab[ ,3]`, `tab[ , "precip"]`, `tab$wind`
- `hist(vals); plot(x,y, las=1, pch=16, type="b")`

# Session 1: Recap (They tried to make me go to recap and I said, 'Yes, yes, yes.')

- R is awesome. You're here, thus you will be R-some, too
- Console for code execution (R), ready with `>`, waiting with `+`
- Script for code writing (you) + exchange
- Objects are created with the assignment operator `<-`
- Get help with `help(median)`, `?matrix`, or pressing F1
- Create vectors with `c(42,3,-4)`, `1:n`
- `data.frames` hold tables with one data type per column
- `read.table`, with `header=TRUE` to read files into R
- Index objects with `values[8:5]`, `tab[4,2:6]`
- df columns with `tab[,3]`, `tab[, "precip"]`, `tab$wind`
- `hist(vals); plot(x,y, las=1, pch=16, type="b")`

The homework is at

<https://github.com/brry/course/blob/master/uni.md#homework>

# Outline

## R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

## Course plan

Session 1/4: Why R, read file, select and plot data

**Session 2/4: Objects, file reading, Rstudio**

Session 3/4: if-else conditionals, for loops

Session 4/4: functions and apply loops

Session x: Recap all

Feedback, additional material

## Session 2: Aim (Chapters 2 - 4.3)

## Session 2: Aim (Chapters 2 - 4.3)

- intensify and fortify the knowledge obtained in session 1

## Session 2: Aim (Chapters 2 - 4.3)

- intensify and fortify the knowledge obtained in session 1
- learn about Rstudio and objects

## Session 2: Aim (Chapters 2 - 4.3)

- intensify and fortify the knowledge obtained in session 1
- learn about Rstudio and objects

afterwards, you should be able to:

## Session 2: Aim (Chapters 2 - 4.3)

- intensify and fortify the knowledge obtained in session 1
- learn about Rstudio and objects

afterwards, you should be able to:

- find the appropriate place and way to get help if you're stuck

## Session 2: Aim (Chapters 2 - 4.3)

- intensify and fortify the knowledge obtained in session 1
- learn about Rstudio and objects

afterwards, you should be able to:

- find the appropriate place and way to get help if you're stuck
- distinguish common data and object types

## Session 2: Aim (Chapters 2 - 4.3)

- intensify and fortify the knowledge obtained in session 1
- learn about Rstudio and objects

afterwards, you should be able to:

- find the appropriate place and way to get help if you're stuck
- distinguish common data and object types
- use Rstudio more effectively

# Rstudio Tips and Tricks

## ① Rstudio settings slide

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and  
[rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and  
[rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ **ALT** + mouse for multiline cursor

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ **ALT** + mouse for multiline cursor
- ⑤ **CTRL** + **SHIFT** + **1** panel full view

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ **ALT** + mouse for multiline cursor
- ⑤ **CTRL** + **SHIFT** + **1** panel full view
- ⑥ **ALT** + **UP** / **DOWN** move line of code

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ ALT + mouse for multiline cursor
- ⑤ CTRL + SHIFT + 1 panel full view
- ⑥ ALT + UP / DOWN move line of code
- ⑦ CTRL + SHIFT + P rerun previous code region

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ ALT + mouse for multiline cursor
- ⑤ CTRL + SHIFT + 1 panel full view
- ⑥ ALT + UP / DOWN move line of code
- ⑦ CTRL + SHIFT + P rerun previous code region
- ⑧ CTRL + SHIFT + S / ENTER source document

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ **ALT** + mouse for multiline cursor
- ⑤ **CTRL** + **SHIFT** + **1** panel full view
- ⑥ **ALT** + **UP** / **DOWN** move line of code
- ⑦ **CTRL** + **SHIFT** + **P** rerun previous code region
- ⑧ **CTRL** + **SHIFT** + **S** / **ENTER** source document
- ⑨ **CTRL** + **UP** (in console) command history

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ **ALT** + mouse for multiline cursor
- ⑤ **CTRL** + **SHIFT** + **1** panel full view
- ⑥ **ALT** + **UP / DOWN** move line of code
- ⑦ **CTRL** + **SHIFT** + **P** rerun previous code region
- ⑧ **CTRL** + **SHIFT** + **S / ENTER** source document
- ⑨ **CTRL** + **UP** (in console) command history
- ⑩ tearable panes

# Rstudio Tips and Tricks

- ① Rstudio settings slide
- ② RStudio IDE Easy Tricks You Might've Missed and [rviews.rstudio.com/categories/tips-and-tricks](http://rviews.rstudio.com/categories/tips-and-tricks)
- ③ Work with Rstudio projects!
- ④ **ALT** + mouse for multiline cursor
- ⑤ **CTRL** + **SHIFT** + **1** panel full view
- ⑥ **ALT** + **UP / DOWN** move line of code
- ⑦ **CTRL** + **SHIFT** + **P** rerun previous code region
- ⑧ **CTRL** + **SHIFT** + **S / ENTER** source document
- ⑨ **CTRL** + **UP** (in console) command history
- ⑩ tearable panes
- ⑪ **CTRL** + **SHIFT** + **O** Document outline

## Session 2: Take home message

- Reading data: `setwd`, `dir`, `read.table`, `str`

## Session 2: Take home message

- Reading data: `setwd`, `dir`, `read.table`, `str`
- Function documentation with `help` (or F1), `StackOverflow` for questions, Reference Cards ([slide 47](#)) for an overview and books / tutorials ([slide 48](#)) for self-learning

## Session 2: Take home message

- Reading data: `setwd`, `dir`, `read.table`, `str`
- Function documentation with `help` (or F1), `StackOverflow` for questions, Reference Cards ([slide 47](#)) for an overview and books / tutorials ([slide 48](#)) for self-learning
- Rstudio is awesome!

## Session 2: Take home message

- Reading data: `setwd`, `dir`, `read.table`, `str`
- Function documentation with `help` (or F1), `StackOverflow` for questions, Reference Cards ([slide 47](#)) for an overview and books / tutorials ([slide 48](#)) for self-learning
- Rstudio is awesome!
- Objects (vector, data.frame, matrix, list, function) can have several data types (numeric, logical, character), see [slide 108](#)

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

## Course plan

Session 1/4: Why R, read file, select and plot data

Session 2/4: Objects, file reading, Rstudio

**Session 3/4: if-else conditionals, for loops**

Session 4/4: functions and apply loops

Session x: Recap all

Feedback, additional material

# to R or not to R

# to R or not to R

*Lucius Annaeus Seneca (4BC - 65AD):*

# to R or not to R

*Lucius Annaeus Seneca (4BC - 65AD):*

To err is human

# to R or not to R

*Lucius Annaeus Seneca (4BC - 65AD):*

To err is human

to  is better

## Session 3: Aim (Chapter 7.1 + 7.2)

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)
- repetitional coding (for-loops)

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)
- repetitional coding (for-loops)

afterwards, you should be able to:

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)
- repetitional coding (for-loops)

afterwards, you should be able to:

- make reasonable guesses about logical values by looking at code

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)
- repetitional coding (for-loops)

afterwards, you should be able to:

- make reasonable guesses about logical values by looking at code
- write conditionally executed code using if/else

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)
- repetitional coding (for-loops)

afterwards, you should be able to:

- make reasonable guesses about logical values by looking at code
- write conditionally executed code using if/else
- decide whether to use `if(cond) A else B` or `ifelse(cond, A, B)`

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)
- repetitional coding (for-loops)

afterwards, you should be able to:

- make reasonable guesses about logical values by looking at code
- write conditionally executed code using if/else
- decide whether to use `if(cond) A else B` or `ifelse(cond, A, B)`
- explain the concept of a loop

## Session 3: Aim (Chapter 7.1 + 7.2)

- conditional code execution (if-else)
- repetitional coding (for-loops)

afterwards, you should be able to:

- make reasonable guesses about logical values by looking at code
- write conditionally executed code using if/else
- decide whether to use `if(cond) A else B` or `ifelse(cond, A, B)`
- explain the concept of a loop
- write simple for-loops

## Session 3: Recap

- `if(TRUE) message("Yo!") else message("Dude...")`

## Session 3: Recap

- `if(TRUE) message("Yo!") else message("Dude...")`
- `ifelse(c(T,T,F,T), 7, else 7.3)` for vectors

## Session 3: Recap

- `if(TRUE) message("Yo!") else message("Dude...")`
- `ifelse(c(T,T,F,T), 7, else 7.3)` for vectors
- for loops repeat a block of code with a different `i` each time

## Session 3: Recap

- `if(TRUE) message("Yo!") else message("Dude...")`
- `ifelse(c(T,T,F,T), 7, else 7.3)` for vectors
- for loops repeat a block of code with a different `i` each time
- `for(i in 1:20) plot(df[,i], main=colnames(df)[i])`

## Session 3: Recap

- `if(TRUE) message("Yo!") else message("Dude...")`
- `ifelse(c(T,T,F,T), 7, else 7.3)` for vectors
- for loops repeat a block of code with a different `i` each time
- `for(i in 1:20) plot(df[,i], main=colnames(df)[i])`
- `for(cn in colnames(df)) plot(df[,cn], main=cn)`

## Session 3: Recap

- `if(TRUE) message("Yo!") else message("Dude...")`
- `ifelse(c(T,T,F,T), 7, else 7.3)` for vectors
- for loops repeat a block of code with a different `i` each time
- `for(i in 1:20) plot(df[,i], main=colnames(df)[i])`
- `for(cn in colnames(df)) plot(df[,cn], main=cn)`
- for loops should not be called too often, as they are slow

## Session 3: Recap

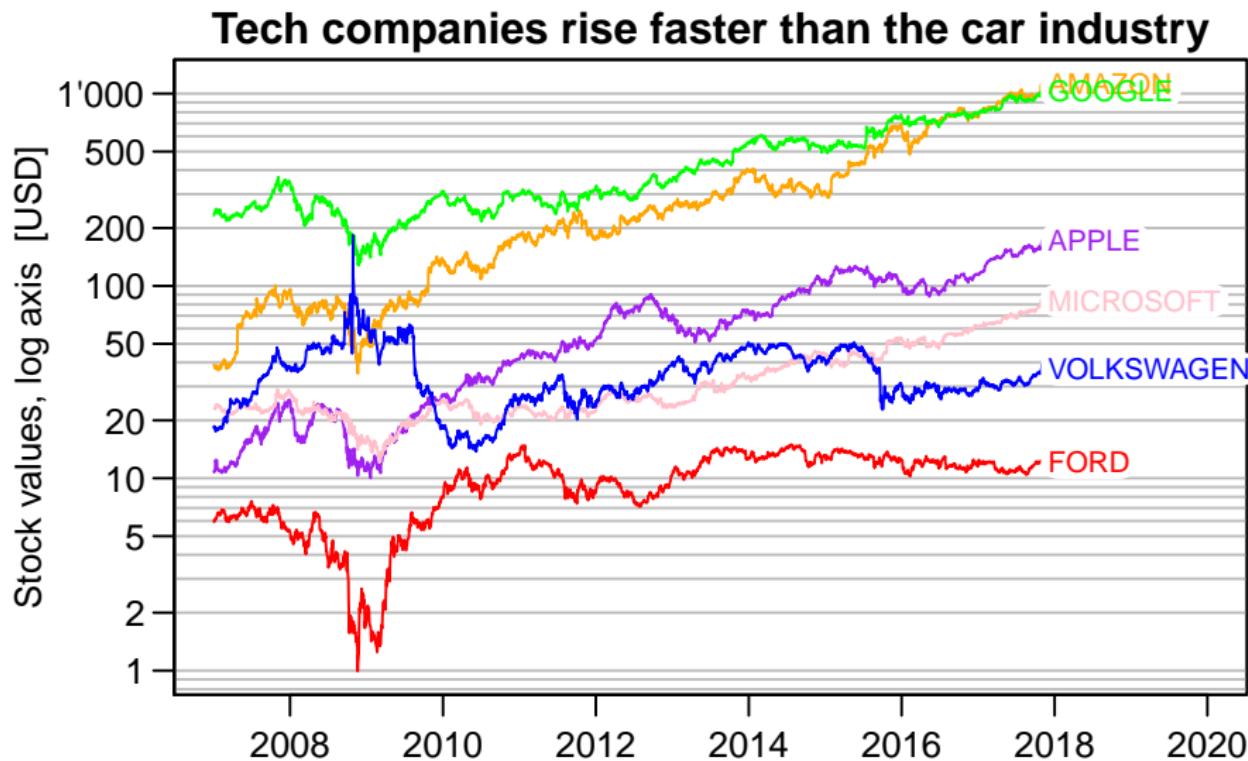
- `if(TRUE) message("Yo!") else message("Dude...")`
- `ifelse(c(T,T,F,T), 7, else 7.3)` for vectors
- for loops repeat a block of code with a different `i` each time
- `for(i in 1:20) plot(df[,i], main=colnames(df)[i])`
- `for(cn in colnames(df)) plot(df[,cn], main=cn)`
- for loops should not be called too often, as they are slow
- instead: vectorize, use `lapply`, write C++ code

# Homework for loop solution

```
stocks <- read.table("data/stocks.txt", header=T, as.is=T)
stocks>Date <- as.Date(stocks>Date)
companies <- colnames(stocks)[-1]
cols <- c("purple", "orange", "red", "green", "pink", "blue")
names(cols) <- companies      # cols["AMAZON"]

plot(stocks>Date, stocks$APPLE, type="n", log="y", yaxt="n",
      xlim=as.Date(c("2007-01-03", "2020-01-01")),
      ylim=range(stocks[,-1]),
      ylab="Stock values, log axis [USD]", xlab="Date", las=1,
      main="Tech companies rise faster than the car industry")
berryFunctions::logAxis(2)
for(comp in companies)
  lines(stocks>Date, stocks[,comp], col=cols[comp])
#legend("topleft", companies, fill=cols)
berryFunctions::textField(as.Date("2017-12-01"), cex=0.8,
                         y=as.numeric(tail(stocks[,-1], 1)),
                         companies, adj=0, col=cols, xpd=T)
```

# Homework for loop solution



# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

## Course plan

Session 1/4: Why R, read file, select and plot data

Session 2/4: Objects, file reading, Rstudio

Session 3/4: if-else conditionals, for loops

**Session 4/4: functions and apply loops**

Session x: Recap all

Feedback, additional material

# Unicode



Collaborative editing can quickly become a textual rap battle fought with increasingly convoluted invocations of U+202a to U+202e

([xkcd.com/1137](http://xkcd.com/1137))

# Today's plan

- ➊ Recap if else + for loop

# Today's plan

- ➊ Recap if else + for loop
- ➋ Writing functions

## Recap if else |

- ① Create a new R file and write `x <- 3.14` into it.
- ② Now assign the absolute value of x to `y` by using conditional code execution (control structures).
- ③ `source` the complete document.
- ④ Test the result of `y` by setting `x` to a negative number in the script and running the document again.
- ⑤ BONUS: Make it also work for vectors like `x <- (-5):8` .
- ⑥ BONUS 2: Which R function gives the absolute value? How is it implemented?

What's in [github.com/wch/r-source](https://github.com/wch/r-source) -> `src/main/names.c` L278 and  
-> `src/main/arithmetic.c` L1315?

## Recap if else |

- ① Create a new R file and write `x <- 3.14` into it.
- ② Now assign the absolute value of `x` to `y` by using conditional code execution (control structures).
- ③ `source` the complete document.
- ④ Test the result of `y` by setting `x` to a negative number in the script and running the document again.
- ⑤ BONUS: Make it also work for vectors like `x <- (-5):8`.
- ⑥ BONUS 2: Which R function gives the absolute value? How is it implemented?

What's in [github.com/wch/r-source](https://github.com/wch/r-source) -> `src/main/names.c` L278 and  
-> `src/main/arithmetic.c` L1315?

```
x <- -5:8
y <- ifelse(x<0, -x, x)
y

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

## Recap `if else` II

```
( requireNamespace("some_package", quietly=TRUE) )  
## [1] FALSE
```

invisibly returns TRUE if it succeeds loading a package. The outer brackets are for printing in the slides.

- ① Use it to conditionally install the package `quantmod` (i.e. if it is not already available).

## Recap `if else` II

```
( requireNamespace("some_package", quietly=TRUE) )  
## [1] FALSE
```

invisibly returns TRUE if it succeeds loading a package. The outer brackets are for printing in the slides.

- ① Use it to conditionally install the package `quantmod` (i.e. if it is not already available).

```
if(!requireNamespace("quantmod")) install.packages("quantmod")  
## Loading required namespace: quantmod
```

# Recap for loop

## Recap **for** loop

With `rexp(n, rate)`, random numbers can be generated from the exponential distribution.

With `paste0("someText_", 1)`, character strings and numbers can be concatenated.

With `write.table(x, file)`, data can be written to a file.

Using a **for** loop, write three datasets to separate .txt files with 5, 20 and 100 exponentially distributed random numbers.

BONUS: format the file nicely with the `write.table()` options.

## Recap **for** loop

With `rexp(n, rate)`, random numbers can be generated from the exponential distribution.

With `paste0("someText_", 1)`, character strings and numbers can be concatenated.

With `write.table(x, file)`, data can be written to a file.

Using a **for** loop, write three datasets to separate .txt files with 5, 20 and 100 exponentially distributed random numbers.

BONUS: format the file nicely with the `write.table()` options.

```
for(n in c(5,20,100))
  write.table(rexp(n), file= paste0("random_", n, ".txt"))
```

# Now let's start learning about functions

[Jump to Functions](#)

# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

## Course plan

Session 1/4: Why R, read file, select and plot data

Session 2/4: Objects, file reading, Rstudio

Session 3/4: if-else conditionals, for loops

Session 4/4: functions and apply loops

Session x: Recap all

Feedback, additional material

# Floyd Recap

Download the recap tasks at [FloydRecap.pdf](#)

# Floyd Recap Solutions

1.

```
setwd("S:/Dropbox/R/Project")
reed <- read.table("file.name", header=T, sep="\t", skip=2)
```

2. Spaces in column names, comments in the middle of data fields (use NA instead), tab-stop separated columns are not the safest option (semicolon ";" is better). `str(reed)` output:

```
data.frame, n observations in 4 variables (rows, columns)
$Site           : factor with n levels ...
$moisture       : factor with 3 levels "wet", "moist", ...: 1 2 2...
$plant.height   : int 154 120 173 ...
$stem_diameter: num 1.80.95 2.6 ...
```

3. see [slide 108](#). `c(pi, "pi")` gives a vector with charstrings, number gets converted in order of coercion. You cannot compute with a characterstring, but you could run `as.numeric(c(pi, "pi")[1])*1.8`.

# Floyd Recap Solutions

4. join "reed" with second dataset "chem":

```
merge(reed, chem, by.x="Site", by.y="columnname", all=TRUE)
```

5.

```
plot(reed$chem, reed$growth)
```

6.

```
par(mfrow=c(2,3), mar=c(2.9,3,1,0.5), mgp=c(2.1,0.7,0), las=1)
```

```
plot(a,b)
```

```
plot(C,D, col="pink")
```

...

7-9.

```
par(mfrow=c(4,4), mar=rep(0,4), oma=c(3,5,2,1), las=1)
```

```
for(i in 1:16)
```

```
{plot(rnorm(20), type="l", yaxt="n", xaxt="n")}
```

```
if( i %in% c(1,5,9,13) ) axis(side=2)
```

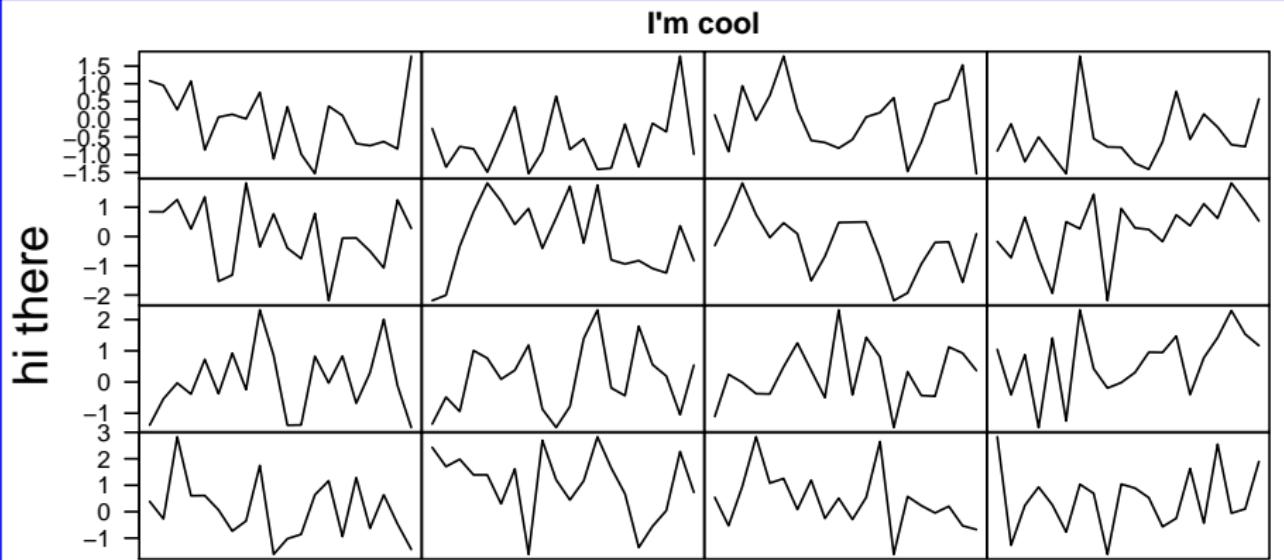
```
}
```

```
title(ylab="hi there", outer=TRUE, cex.lab=2)
```

```
title(main="I'm cool", outer=TRUE)
```

```
box("outer", col="blue", lwd=5)
```

# Floyd Recap Solutions



# Outline

R course info

1. One-Session-Intro
2. Getting started: Background, GUI and first steps
3. Objects
4. "Real" data
5. Plotting
6. Character Operations
7. Conditions & loops
8. Functions & packages
9. Distributions
10. Statistics
11. Time series handling and analysis
12. Spatial data and GIS functionality
13. Final tasks

## Course plan

Session 1/4: Why R, read file, select and plot data

Session 2/4: Objects, file reading, Rstudio

Session 3/4: if-else conditionals, for loops

Session 4/4: functions and apply loops

Session x: Recap all

**Feedback, additional material**

# Feedback

Please fill out the feedback form at

[bit.ly/feedbackR](https://bit.ly/feedbackR)

(it only takes a few minutes and helps to improve the course)

Thanks!