# Advanced R: Visualization and Programming

Computational Economics Practice Winter Term 2015/16 Stefan Feuerriegel

## Today's Lecture

## Objectives

- Visualizing data in R graphically as points, lines, contours or areas
- 2 Understanding the programming concepts of if-conditions and loops
- 3 Implementing simple functions in R
- 4 Measuring execution time

# Outline

## 1 Visualization

2 Control Flow

## 3 Timing

4 Wrap-Up

# Outline

## 1 Visualization

#### 2 Control Flow

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# Point Plot

- Creating simple point plots (also named scatter plots) via plot (...)
- Relies upon vectors denoting the x-axis and y-axis locations
- Various options can be added to change appearance

```
d <- read.csv("persons.csv", header=TRUE, sep=",",</pre>
               stringsAsFactors=FALSE)
plot(d$height, d$age)
```



# Adding Title, Labels and Annotations

- Title is added through additional parameter main
- Axis labels are set via xlab and ylab
- Annotations next to points with text(...)



Title

## Line Plot

Generate line plot using the additional option type="l"

```
x <- seq(0, 4, 0.01)
plot(x, x*x, type="l")
```



## Exercise: Plotting

x <- seq(-1, +1, 0.01)



## Question

How would you reproduce the above plot?

- > plot(x, kink(x), type="l", main="")
- > plot(x, kink(x), type="l", lab="")
- ▶ plot(x, abs(x), type="l", ylab="", xlab="")

Visit http://pingo.upb.de with code 1523

• Consider the function  $f(x,y) = x^3 + 3y - y^3 - 3x$ 

```
f <- function(x, y) x^3+3*y-y^3-3*x
```

Create axis ranges for plotting

```
x <- seq(-5, 5, 0.1)
y <- seq(-5, 5, 0.1)
```

► Function outer (x, y, f) evaluates f all combinations of x and y

z <- **outer**(x, y, f)

## Function $\texttt{persp}\left(\ldots\right)$ plots the plane through $x,\,y$ and z in 3D

persp(x, y, z)



Turn on ticks on axes via ticktype="detailed"

persp(x, y, z, ticktype="detailed")



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#### Parameters theta (left/right) and phi (up/down) control viewing angle

persp(x, y, z, theta=20, phi=0)

**persp**(x, y, z, theta=20, phi=35)





## **Contour Plots**

- ► A contour line is a curve along which the function has the same value
- ► image (...) plots a grid of pixels colored corresponding to z-value
- ► contour(..., add=TRUE) adds contour lines to an existing plot

```
image(x, y, z) # Plot colors
contour(x, y, z, add=TRUE) # Add contour lines
```



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## **Contour Plots**

```
f <- function(x, y) sqrt(x^2+y^2)
z <- outer(x, y, f)
image(x, y, z, asp=1) # set aspect ratio, i.e. same scale for x and y
contour(x, y, z, add=TRUE)</pre>
```

#### Question





Visit http://pingo.upb.de with code 1523

## **Plotting Regression Plane**

```
library(car) # for dataset Highway1
```

## Warning: no function found corresponding to methods exports from 'SparseM' for: 'coerce'

```
model <- lm(rate ~ len + slim, data=Highway1)</pre>
model
##
## Call:
## lm(formula = rate ~ len + slim, data = Highway1)
##
## Coefficients.
## (Intercept)
                     len
                                     slim
## 16.61050 -0.09151 -0.20906
x1r <- range(Highway1$len)</pre>
x1seq <- seq(x1r[1], x1r[2], length=30)</pre>
x2r <- range(Highway1$slim)</pre>
x2seg <- seg(x2r[1], x2r[2], length=30)</pre>
z <- outer(x1seq, x2seq,</pre>
           function(a,b) predict(model,
                                   newdata=data.frame(len=a,slim=b)))
```

## Plotting a Regression Plane



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## Managing Code Execution

- Control flow specifies order in which statements are executed
- Previous concepts can only execute R code in a linear fashion
- Control flow constructs can choose which execution path to follow

Functions: Combines sequence of statements into a self-contained task Conditional expressions: Different computations according to a specific condition

Loops: Sequence of statements which may be executed more than once

## Functions

- Functions avoid repeating the same code more than once
- Leave the current evaluation context to execute pre-defined commands



## Functions

- Extend set of built-in functions with opportunity for customization
- ► Functions **can** consist of the following:
  - 1 Name to refer to (avoid existing function names in R)
  - 2 Function body is a sequence of statements
  - 3 Arguments define additional parameters passed to the function body
  - 4 Return value which can be used after executing the function
- Simple example

```
f <- function(x,y) {
   return(2*x + y^2)
}
f(-3, 5)
## [1] 19</pre>
```

## Functions

## General syntax

```
functionname <- function(argument1, argument2, ...) {
  function_body
  return(value)
}</pre>
```

Return value is the last evaluated expression

 $\rightarrow$  Alternative: set explicitly with return(...)

- Curly brackets can be omitted if the function contains only one statement (not recommended)
- Be cautious since the order of the arguments matters
- Values in functions are not printed in console

```
\rightarrow Remedy is print ( . . . )
```

## **Examples of Functions**

square <- function(x) x\*x # last value is return value
square(10)</pre>

## [1] 100

```
cubic <- function(x) {
    # Print value to screen from inside the function
    print(c("Value: ", x, " Cubic: ", x*x*x))
    # no return value
}
cubic(10)
## [1] "Value: " "10" " Cubic: " "1000"</pre>
```

## **Examples of Functions**

```
hello <- function() { # no arguments
    print("world")
}
hello()
## [1] "world"</pre>
```

```
my.mean <- function(x) {
   return (sum(x)/length(x))
}
my.mean(1:100)
### [1] 50.5</pre>
```

## Scope in Functions

- $\blacktriangleright$  Variables created inside a function only exists within it  $\rightarrow$  local
- They are thus inaccessible from outside of the function
- Scope denotes when the name binding of variable is valid

```
x <- "A"
g <- function(x) {
    x <- "B"
    return(x)
}
x <- "C"</pre>
```

What are the values?



Solution

```
## [1] "B"
## [1] "C"
```

## Scope in Functions

- $\blacktriangleright$  Variables created inside a function only exists within it  $\rightarrow$  local
- ► They are thus inaccessible from outside of the function
- Scope denotes when the name binding of variable is valid

```
x <- "A"
g <- function(x) {
    x <- "B"
    return(x)
}
x <- "C"</pre>
```

What are the values?

g(x) # Return value of function x
x # Value of x after function execution

## Solution

## [1] "B" ## [1] "C"

## **Unevaluated Expressions**

- Expressions can store symbolic mathematical statements for later modifications (e.g. symbolic derivatives)
- ► Let's define an example via expression ( . . . )

```
f <- expression(x^3+3*y-y^3-3*x)
f
## expression(x^3 + 3 * y - y^3 - 3 * x)</pre>
```

 If evaluation of certain parameters becomes necessary, one can use eval(...)

x <- 2
y <- 3
eval(f)
## [1] -16</pre>

# **If-Else Conditions**

Conditional execution requires a condition to be met



# **If-Else Conditions**

- Keyword if with optional else clause
- General syntax:
   if condition

```
if (condition) {
   statement1
}
```

```
If condition is true, then statement is executed
```

#### if-else condition

```
if (condition) {
   statement1
} else {
   statement2
}
```

If condition is true, then
statement1 is executed,
otherwise statement2

# **If-Else Conditions**

## ► Example



Condition must be of length 1 and evaluate as either TRUE or FALSE

```
if (c(TRUE, FALSE)) { # don't do this!
    print("something")
}
## Warning in if (c(TRUE, FALSE)) {: Bedingung hat Länge
> 1 und nur das erste Element wird benutzt
## [1] "something"
```

# **Else-If Clauses**

- Multiple conditions can be checked with else if clauses
- The last else clause applies when no other conditions are fulfilled
- The same behavior can also be achieved with nested if-clauses
   else-if clause
   Nested if-condition

```
if (grade == 1) {
    print("very good")
} else if (grade == 2) {
    print("good")
} else {
    print("not a good grade")
}
```



# **If-Else Function**

As an alternative, one can also reach the same control flow via the function ifelse(...)

ifelse(condition, statement1, statement2)
# executes statement1 if condition is true,
# otherwise statement2

```
grade <- 2
ifelse(grade <= 4, "Passed", "Failed")
## [1] "Passed"</pre>
```

ifelse(...) can also work with vectors as if it was applied to each element separately

```
grades <- c(1, 2, 3, 4, 5)
ifelse(grades <= 4, "Passed", "Failed")
## [1] "Passed" "Passed" "Passed" "Failed"</pre>
```

This allows for the efficient comparison of vectors

# For Loop

for loops execute statements for a fixed number of repetitions



For Loop

General syntax

```
for (counter in looping_vector) {
    # code to be executed for each element in the sequence
}
```

- In every iteration of the loop, one value in the looping vector is assigned to the counter variable that can be used in the statements of the body of the loop.
- Examples

```
for (i in 4:7) {
    print(i)
}
### [1] 4
### [1] 5
### [1] 6
### [1] 7
```

```
a <- c()
for (i in 1:3) {
    a[i] <- sqrt(i)
}
a
## [1] 1.000000 1.414214 1.732051</pre>
```

# While Loop

- Loop where the number of iterations is controlled by a condition
- The condition is checked in every iteration
- ► When the condition is met, the loop body in curly brackets is executed
- General syntax

```
while (condition) {
    # code to be executed
}
```

## Examples

```
z <- 1
# same behavior as for loop
while (z <= 4) {
    print(z)
    z <- z + 1
}
## [1] 1
## [1] 2
## [1] 3
## [1] 3
ced R: CGNOFIENW</pre>
```

```
z <- 1
# iterates all odd numbers
while (z <= 5) {
    z <- z + 2
    print(z)
}
## [1] 3
## [1] 5
## [1] 7</pre>
```

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Advanced R: Timing

## Measuring Timings via Stopwatch

- Efficiency is a major issue with larger datasets and complex codes
- Timings can help in understanding scalability and bottlenecks
- Use a stopwatch approach measuring the duration between two proc.time() calls

```
start.time <- proc.time() # Start the clock
g <- rnorm(100000)
h <- rep(NA, 100000)
for (i in 1:100000) { # Loop over vector, always add +1
h[i] <- g[i] + 1
}
# Stop clock and measure duration
duration <- proc.time() - start.time</pre>
```

# Measuring Timings via Stopwatch

Results of duration have the following format

## user system elapsed ## 0.71 0.02 0.72

- Timings are generally grouped into 3 categories
  - User time measures the understanding of the R instructions
  - System time measures the underlying execution time
  - Elapsed is the difference since starting the stopwatch (= user + system)
- Alternative approach avoiding loop

```
start.time <- proc.time() # Start clock
g <- rnorm(100000)
h <- g + 1
proc.time() - start.time # Stop clock
## user system elapsed
## 0.08 0.00 0.08</pre>
```

#### Rule: vector operations are faster than loops

# Measuring Timings of Function Calls

Function system.time(...) can directly time function calls

```
slowloop <- function(v) {
    for (i in v) {
        tmp <- sqrt(i)
        }
}</pre>
```

system.time(slowloop(1:1000000))

# #	user	system	elapsed
##	2.06	0.05	2.13

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# Fancy Diagrams with ggplot2

library(ggplot2)



# Summary: Visualization and Timing

plot() Simple plot function text() Add text to an existing plot outer() Apply a function to two arrays persp() Plot a surface in 3D image() Plot a colored image Add contour lines to a plot contour() trans3d() Add point to an existing 3D plot points() Add points to a plot Stopwatch for measuring execution time proc.time() system.time(expr) Measures execution time of an expression

# Summary: Programming

<pre>function(){}</pre>	Self-defined function
expression()	Function with arguments not evaluated
eval()	Evaluate an expression
if,else	Conditional statement
for(){}	Loops over a fixed vector
while	Loops while a condition is fulfilled

# Outlook

## Additional Material

- Further exercises as homework
- Advanced materials beyond our scope
  - Advanced R (CRC Press, 2014, by Wickham) http://adv-r.had.co.nz/

#### **Future Exercises**

R will be used to implement optimization algorithms