Creative Data Mining

Lecture 02: Intro to RStudio and Clustering 29 February 2016

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What we'll cover today:

- HW1: Share what you have learned about urban data mining (20 min)
- Break (10 min)
- R tutorial (60 min)







Homework_1

1. Install R from http://cran.r-project.org/

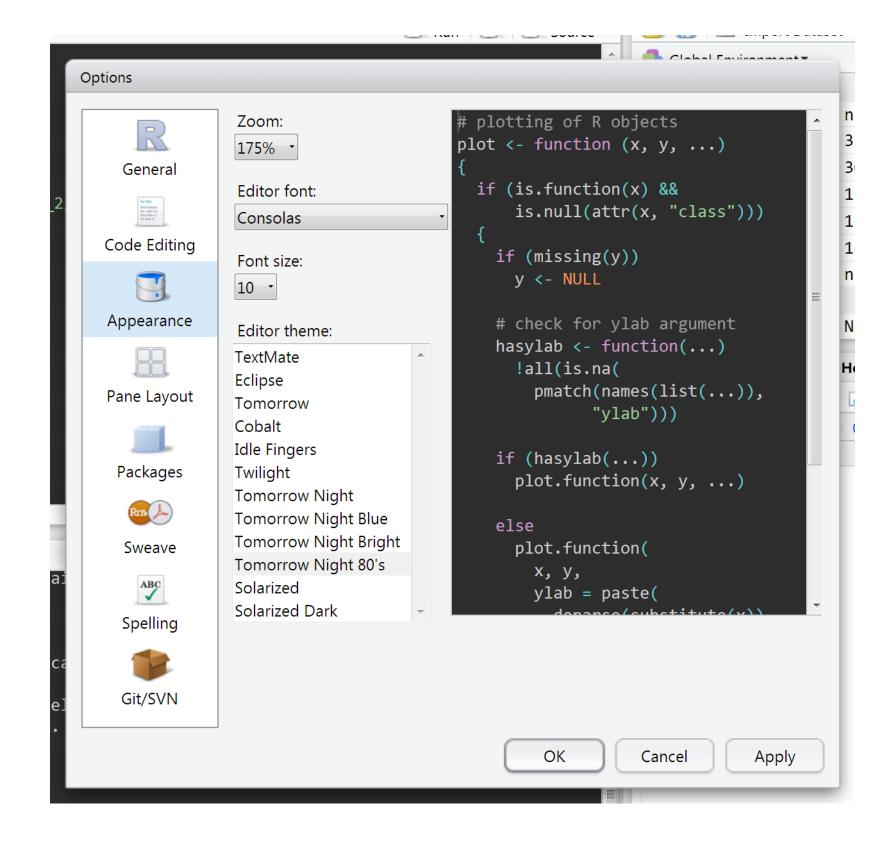


- 2. Install RStudio from http://www.rstudio.com/products/rstudio/download/
- 3. Research other examples of urban data mining and make 3 slides about the most interesting project/application/resarch group(s) that you find. Will be presented at the beginning of next lecture.



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ADJUST COLORS







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OVERVIEW

lecture_2.R



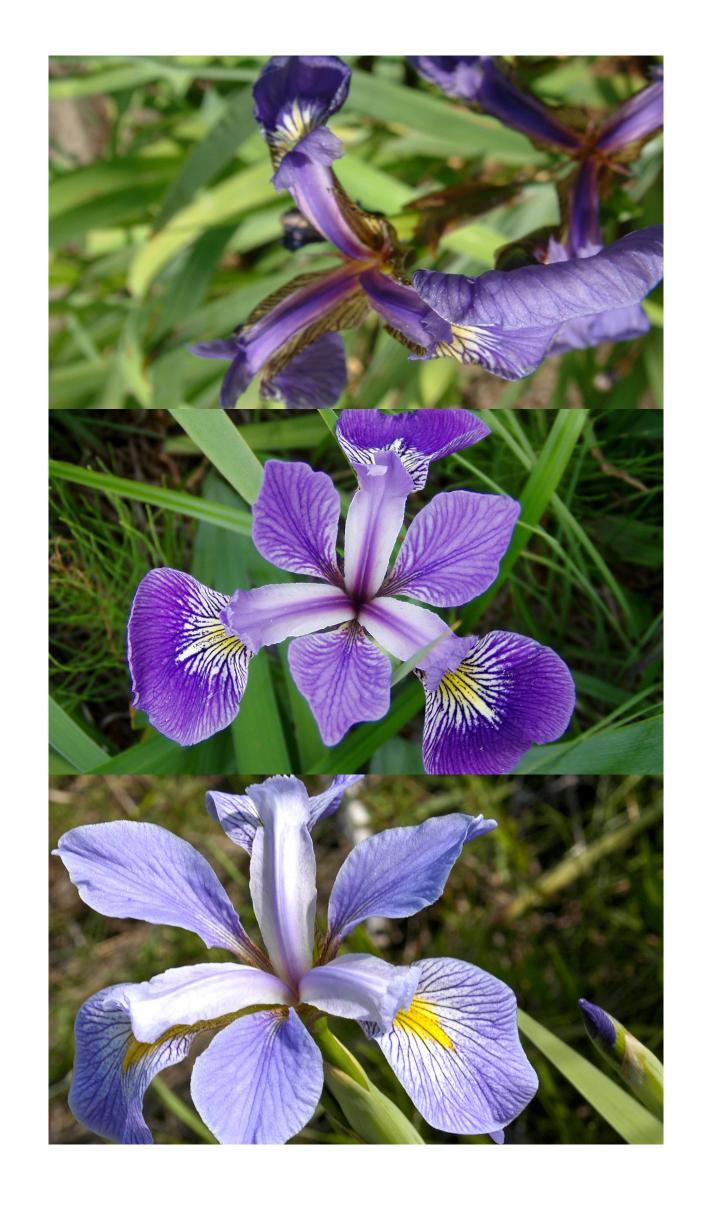


```
# CREATIVE DATA MINING - FS 2016
    # LECTURE 2 - R101
   # Matthias Standfest
    # Danielle Griego
    ## OVERVIEW
    # keyboard shortcuts: https://support.rstudio.com/hc/en-us/articles/200711853-Keyboard-Shortcuts
13
   # NUMBERS
   a <- 2 #store a number
    (b <- 5) #brackets for writing to console
    (a*b*pi) #simple multiplication with predefined variable
19
   # VECTORS
    (vector_a <- c(2,3,4,5)) #save a list of numbers
    (vector_b <- 5:8) #shortcut for number sequence</pre>
    (vector_a * vector_b) #multiplying lists
   # MATRICES
    (matrix_a <- matrix(1, nrow=4, ncol=4)) #create 4 x 4 matrix filled with 1
    (matrix_b <- matrix(1:16, nrow=4, ncol=4)) #fill matrix with number sequence
    (matrix_c <- matrix(1:16, nrow=4, ncol=4, byrow=TRUE)) #fill by rows not by columns
    matrix_b1<- t(matrix_b) #transposing also exchanges rows and columns
    (matrix_a *3) #multiplying matrix with scalar
    (matrix_b * matrix_c) #multiplying matrices elementwise
    test <- matrix_b %*% matrix_c #dot product of two matrices
   # PLOTTING
    plot(vector_a, type="o", col="blue") #simple plot of vector
    help(plot) #search plot elements, change plot type, col, etc.
   #install.library(fields)
   #library(fields) #import library need to import libraries for certain functions
    image.plot(matrix((data=matrix_a), ncol=4, nrow=4)) #plot matrix_a, compare plot with numeric values of matrix_a
    image.plot(matrix((data=matrix_b), ncol=4, nrow=4)) #plot matrix_b, compare plot with numeric values of matrix_b
    image.plot(matrix((data=matrix_b1), ncol=4, nrow=4)) #plot matrix_b1, compare plot with numeric values of matrix_b1
    image.plot(matrix((data=matrix_c), ncol=4, nrow=4)) #plot matrix_c, compare with matrix_b and matrix_b1
   # Note that you can transpose data in a matrix using byrow=TRUE when creating the matrix, or by t(matrix_b)
```

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IRIS DATA SET (1936)

en.wikipedia.org/wiki/Iris_flower_data_set







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VIEW PREINSTALLED DATA FRAME

lecture_2_00.R

```
# CREATIVE DATA MINING - FS 2016
    # LECTURE 2 - R101
    # Matthias Standfest
   # Danielle Griego
    ## EXPLORING PREINSTALLED DATA
   # 1. set your working directory
   # or use Ctl+Shift+H and select
    setwd("~/Dropbox/00_Work/01_Teaching/Creative Data Mining/001_FS15-DataMining/lecture_2/RSCRIPTS_FS16/")
13 # 2. print the preinstalled data set
   print(iris)
   # 3. print first rows
   head(iris)
19 # 4. print structural information
   str(iris)
   colnames(iris)
  # 5. print summary
   summary(iris)
```





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SAVE AND READ DATA

lecture 2 01.R

```
# CREATIVE DATA MINING - FS 2016
    # LECTURE 2 - R101
    # Matthias Standfest
    # Danielle Griego
    ## READING AND WRITING CSV FILES TO A FOLDER
   # 1. set your working directory
   setwd("~/Dropbox/00_Work/01_Teaching/Creative Data Mining/001_FS15-DataMining/lecture_2/RSCRIPTS_FS16/")
12 # 2. use data
   myData <- iris
   # 3. write data in file, check that it's in your working directory
   write.table(myData, file = "myFlowerData.csv", sep = ";", col.names = NA, qmethod = "double")
   # 4. read data back into R
   myFile <- read.table("myFlowerData.csv", header = TRUE, sep = ";", row.names = 1)</pre>
   # 5. write data in R file (smaller and better structured), check that it's in your working directory
   saveRDS(myData, file="myFlowerDataUnreadableButBetter.RData")
   # 6. get data from R file
   myOtherFile <- readRDS(file="myFlowerDataUnreadableButBetter.RData")</pre>
```





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USE ONLY PARTS OF THE DATA

lecture_2_02.R

```
# CREATIVE DATA MINING - FS 2016
# LECTURE 2 - R101
# Matthias Standfest
# Danielle Griego

## ACCESSING SPECIFIC ROWS AND COLUMNS FROM A MATRIX

# 1. set your working directory
setwd("~/Dropbox/00_Work/01_Teaching/Creative Data Mining/001_FS15-DataMining/lecture_2/RSCRIPTS_FS16/")

# 2. use data partially
(iris[1:3,3) # rows 1 TO 3

(iris[1:3,3:5]) # rows 1 TO 3 and columns 3 TO 5

(iris[1:3,c(1,5)]) # rows 1 TO 3 and columns 1 AND 5

# 3. Question: create a matrix from the iris dataset without categorical data (column 5)

myData <- iris[,1:4]
```





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DATAFRAME AND MATRIX

lecture_2_03.R

```
2 # CREATIVE DATA MINING - FS 2016
    # LECTURE 2 - R101
    # Matthias Standfest
    # Danielle Griego
    ## MATRIX VERSUS DATA FRAMES?
9 # 1. set your working directory
    setwd("~/Dropbox/00_Work/01_Teaching/Creative Data Mining/001_FS15-DataMining/lecture_2/RSCRIPTS_FS16/")
12 # 2. WHAT IS A DATA FRAME?
    ## answer <- a data frame is a list of vectors with equal length, thus a matrix with mixed types
15 # 2.1. WHAT ARE TYPES?
# answer <- we have numerics (numbers), characters (text) and logicals (TRUE or FALSE)
    numeric_vector <- c(1,3.14,23,42)
    character_vector <- c("start", "pie", "the all seing eye", "the answer to everything")</pre>
    logical_vector <- c(TRUE, TRUE, FALSE, TRUE)</pre>
    # try decomposing the vector structure to understand how it works, for example remove the c, a generic function which combines its
    logical_vector <- (TRUE, TRUE, FALSE, TRUE)</pre>
    # combine vectors to a data.frame
    my_first_dataframe <- data.frame(numeric_vector, character_vector, logical_vector)</pre>
   # 4. WHAT IS A MATRIX?
28 # answer <- a matrix is a list of vectors with equal length of the SAME type
   myMatrix <- matrix(1:20, nrow = 5, ncol = 4)
    # label the matrix rows and columns
    dimnames(myMatrix) <- list(c("a","b","c","d","e"), character_vector)</pre>
32 # use those labels
   myMatrix[,c("start","the answer to everything")] #to identify columns
    myMatrix["a",] #to identify a single row (or more)
   # 5. labelling a data frame
    names(my_first_dataframe) <- c("values","nerdstuff","BOOL")</pre>
    my_first_dataframe[,"values"] # access like a matrix
    my_first_dataframe$values # shortcut to a vector
```





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ADDITIONAL STUFF

lecture 2 04.R

```
# CREATIVE DATA MINING - FS 2016
# LECTURE 2 - R101
# Matthias Standfest
# Danielle Griego
## YOU CAN ALSO ORGANIZE DATA IN LISTS
# 1. set your working directory
setwd("~/Dropbox/00_Work/01_Teaching/Creative Data Mining/001_FS15-DataMining/lecture_2/RSCRIPTS_FS16/")
# 2. LISTS
# additional type <- a list is just an ordered collection of all objects under one name
myList <- list(letters = c("a","b","c"),</pre>
               numbers = 1:50,
               datatable = matrix(1:2, nrow=2, ncol=2),
               answer = 42,
               innerlist = list(sequence2=1:25, sequence3=26:35)) # you can use linebreaks to make your code more readable
myList[2] #access element by index
myList$numbers #access element by name
myList[["numbers"]] #access element via double bracket convention
myList$innerlist$sequence3 #access nested element
myList$innerlist #access nested element
# 3. additional functions
length(myList) # number of elements
class(myList$numbers) #class of element
class(myList$letters) # now try this for another variable
cbind(1,2,3) #instead of c, binding in matrix form
rbind(1,2,3) #matrix bind row wise
ls() # list all current objects, notice they are also in the upper righthand object pane
```





Schedule



Creative Data Mining Intuitively Analysing Design Ideas

The goal of this course is to introduce various data mining techniques for design and urban planning applications. Students will learn how to select relevant data sources and collect their own data using a "sensor backpack". Various methods will be applied to a common project to evaluate the predominant influencing factors of the urban environment on our perceptual experiences. A select neighborhood in the city will be used as a case study. Final results will be presented in the last class.

The course will start with an initial overview to data mining and the relevant mathematics as well as an introduction to the programming tool (RStudio). Then students will learn how to use and interpret results from a machine-learning tool to cluster self-made design sketches, which automatically generate qualitative collages. Finally, students will collect data using a "sensor backpack" with environmental sensors such as noise, temperature, illuminance, and air particulates. Students will also generate the data for perceptual quality in this neighborhood through time-stamped and geo-referenced surveys and biofeedback wristbands. Students will be given a work-flow to collect, process, analyze and interpret this data which may be used in their final projects.

Where

HIT H 12

Supervision

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Swiss Federal Institute of Technology Zurich





22.02.2016 Course Introduction

Introduce data-mining techniques and case study

Introduction to the Environment 29.02.2016 Introduction to R Studio and clustering

07.03.2016 From analog to digital analysis

Use hand-drawn sketched to auto-generated collages

14.03.2016 Seminar week (No lecture)

Analysis and interpretation I 21.03.2016

Evaluate auto-generated collages

28.03.2016 Holiday (No lecture)

04.04.2016 Time-series data analysis and Urban Planning

Introduction to time-series analysis

Data collection with sensor backpack 11.04.2016

Collect data and introduce workflows

Holiday (No lecture) 18.04.2016

Analysis and interpretation II

Evaluate sensor backpack data

02.05.2016 Q&A Feedback Workshop

Finalise semester projects

09.05.2016 Final iA critique

Combined critique with the other iA courses

(14:00 - 16:00)

Requirement Former knowledge of any digital tool or coding language is most welcome but NOT required. You only need to provide a reasonable amount of motivation and of course a notebook.

* Total 60 h = 2 ECTS

Exercises 40% (documentations) Final Presentation 40% (Final project) Attendance 20%

The most recent outline will be found on www.ia.arch.ethz.ch

CREATIVE DATA MINING Intuitively Analysing Design Ideas

Homework:

- 1. Review the R-tutorials lecture 2_00 through lecture 2_04
- 2. Complete HW_2
- 3. Optional: Review how to export images from maps.stamen.com according to the tutorial on the next pages. We will prepare the street maps for the next course exercise, but this is one method to create an image database for clustering analysis in the first block.



