# Session 2

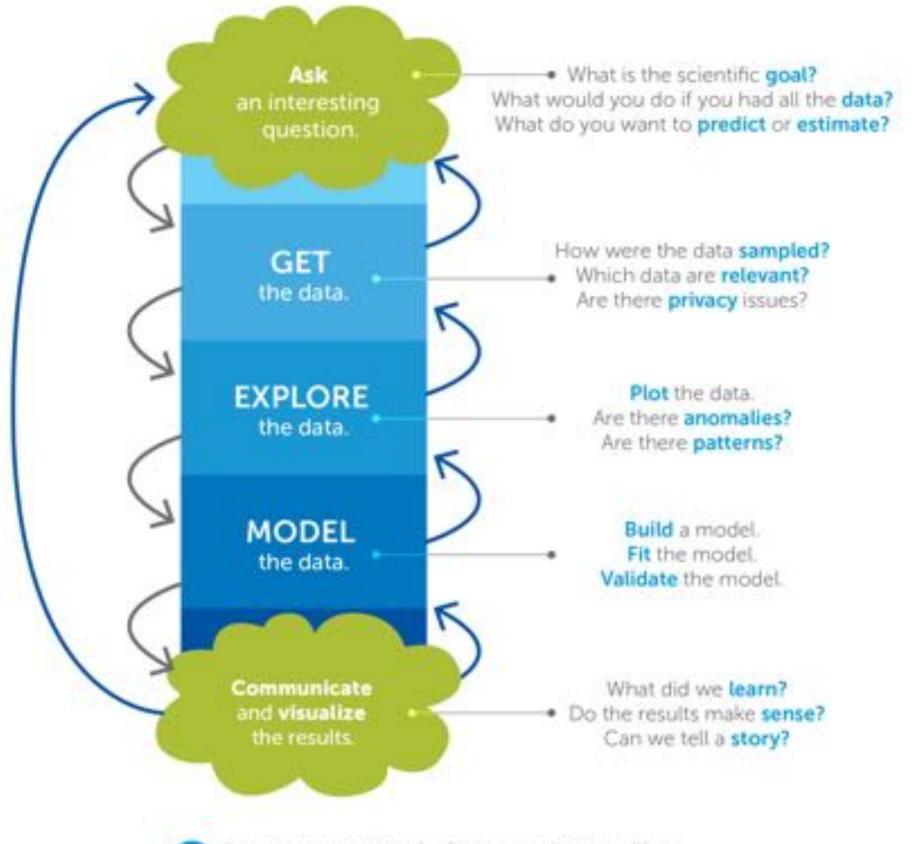
(Customer Analytics)

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#### The Data Science Process



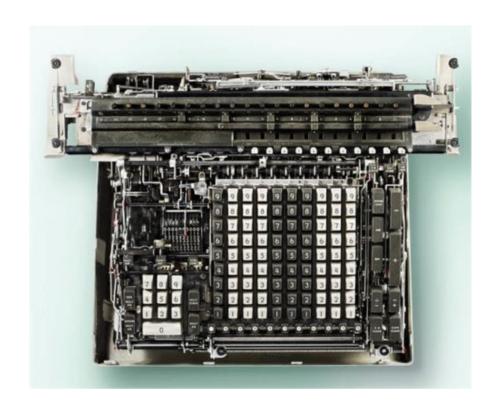
Derived from the work of Joe Blitzstein and Hanspeter Pfister, originally created for the Harvard data science course http://cs109.org/

#### R can be used:

- Data Import (access any type of data)
- Data Wrangling (tidy, prepare and sample)
- Exploratory Data Analysis
   (plot, anomalies & patterns)
- Analysis (build, fit, validate)
- Communicate

2







```
R-intro-session1.R* ×
    Z+3 #simple math
   x < -34.56 + 23.15 + log(456)
   print(x)
5
   X < -c(1,2,4)
   y < -c(5,6,7)
   print(x+y)
10
11
```



Operator	Description	
+	Addition	
	Subtraction	
*	Multiplication	
1	Division	
^	Exponential	
%%	Modulus	
%/%	Integer division	



# f(argument1, argument2, ...)

```
# log to the base e of 2
log(2)
# antilog of 2
 exp(2)
# log to base 2 of 3
log(3,2)
# log to base 10 of 2
log10(2)
# square root of 2
sqrt(2)
# !5
factorial(4)
```

```
# largest interger smaller than 2
floor(2)
# smallest integer greater than 6
ceiling(6)
# round 3.14159 to three digits
round(3.14159, digits=3)
# cosine of 3
cos(3)
# sine of 3
 sin(3)
# tangent of 3
tan(3)
```





• Calculate the value of 1 euro after 1,3,5,8,10 years at a 5% compound rate

FV = value ^(years\*rate)

```
# Our first R code

years <- c(1,3,5,8,10)
result <- exp(0.05*years)
print(result)

# We can improve the code

rate <- 0.05
result <- exp(rate*years)
print(result)</pre>
```





dim(available.packages())

A Swiss Army Knife with its thousands of packages

- Structured, standardized unit of:
  - R code
  - documentation
  - data
  - external code

- · What packages do I have installed?
  - (.packages())



# Handling packages

- Install with install.packages(name)
  - install.packages("ggplot2")
- Uninstall with remove.packages(name)
  - remove.packages("ggplot2")
- Load with library(name)
  - library("ggplot2")
  - You can also use require() but it is a bad programming practice
- Package-level help:
  - · library(help=name)
- Unload with detach(package:name)
  - detach("package:ggplot2", unload=TRUE)

- See the packages tab in Rstudio
- Checkmark to load
- Some are already loaded!!
- Click on the name for help



# Exploring a dataset

- data(name)
  - data("diamonds")
    - Loads the dataset in memory
    - data() shows datasets available

- View(diamonds)
- head(diamonds)
- tail(diamonds)





# R-style Recommendations

#### Indentation

Indent lines with two spaces, not tabs. If code is inside parentheses, indent to the innermost parentheses

#### Spacing

Use only single spaces. Add spaces between binary operators and operands. Do not add spaces between a function name and the argument list. Add a single space between items in a list, after each comma

#### **Blocks**

Don't place an opening brace ("{") on its own line. Do place a closing brace ("}") on its own line. Indent inner blocks (by two spaces)

#### Semicolons

Omit semicolons at the end of lines when they are optional

#### Naming

Name objects with lowercase words, separated by periods. For function names, capitalize the name of each word that is joined together, with no periods. Try to make function names verbs.

#### Assignment

Use <-, not = for assignment statements



#### Values – value assignment Varibles - dynamically type variables

Unlike other languages, R variables do not need to be declared and typed

- Assigning a sequence of numbers to x forces x to be a numeric vector
- Given x, executing class(x) reports the class. This indicates which functions can be used on x
  - Everything is a vector

```
# Values assigment
a <- 5
b <- a
a <- 7
print(b)</pre>
```

```
# Dynamically type variables
x <- 5
class(x)
x <- "hello"
class(x)
x <- 5L
class(x)
x <- TRUE
class(x)
x[1]
class(x[1])</pre>
```



#### Variables – Mode vs. Class

In R, every object has a mode, which indicates how it is stored in memory: as a number, as a character string, as a list of pointers to other objects, as a function, and so forth:

<b>Object</b>	Example	Mode
Number	3.1415	numeric
Vector of numbers	c(2.7.182, 3.1415)	numeric
Character string	"Moe"	character
Vector of character strings	c("Moe", "Larry", "Curly")	character
Factor	factor(c("NY", "CA", "IL"))	numeric
List	list("Moe", "Larry", "Curly")	list
Data frame	data.frame(x=1:3, y=c("NY", "CA", "IL"))	list
Function	print	function



#### Variables – Mode vs. Class

Additionally every object also has a *class*, which defines its abstract type.

- A single number could represent many different things: a distance, a point in time, a weight
- All those objects have a mode of "numeric" because they are stored as a number; but they could have different classes to indicate their interpretation.

```
# Modes vs Class
crashCourseDate <- as.Date("2016-10-19")
typeof(crashCourseDate)
class(crashCourseDate)</pre>
```



#### Variables – Testing and Changing types

Type	Testing	Coercing
Array	is.array	as.array
Character	is.character	as.character
Complex	is.complex	as.complex
Dataframe	is.data.frame	as.data.frame
Double	is.double	as.double
Factor	is.factor	as.factor
List	is.list	as.list
Logical	is.logical	as.logical
Matrix	is.matrix	as.matrix
Numeric	is.numeric	as.numeric
Raw	is.raw	as.raw
Time series (ts)	is.ts	as.ts
Vector	is.vector	as.vector



## Data Objects in R

- vector a sequence of numbers or characters, or higher- dimensional arrays like matrices
- list a collection of objects that may themselves be complicated
- factor a sequence assigning a category to each index
- data.frame a table-like structure (experimental results often collected in this form)
- environment hash table. A collection of key-value pairs



### Data Objects in R - Vectors

- Vectors can only <u>contain entries of the</u> <u>same type</u>: numeric or character; you can't mix them.
  - Note that characters should be surrounded by " ".

```
# Vectors
x <- c("a","b","c")
length(x)</pre>
```

• The most basic way to create a vector is with  $c(x_1, ..., x_n)$ , and it works for characters and numbers alike.



#### Data Objects in R - Vectors

Vector variables can and should be named to allow for future indexing

```
x <- c(s1=0.5,s2=0.8,s3=0.1)
x
sort(x)
names(x)

y <- c(0.5,0.8,0.1)
names(y) <- c("s1","s2","s3")
y</pre>
```



### Data Objects in R - Vectors

var, mean, summary, min and max are useful functions for working with vectors,

```
> X
[1] -1.6267904 -1.2337142 0.2257716 -0.6470700
> var(x)
[1] 0.6485375
> max(x)
[1] 0.2257716
> min(x)
[1] -1.62679
> summary(x)
  Min. 1st Qu. Median Mean 3rd Qu.
                                           Max.
-1.6270 -1.3320 -0.9404 -0.8205 -0.4289
                                         0.2258
```



#### Data Objects in R - Scalars are vectors

Scalars are vectors of length 1

```
> x < -34
> x[1]
[1] 34
> x[2]
[1] NA
> length(x)
[1] 1
```



# Vectors – Generating sequences

- c concatenate
- seq, :, and rep
- vector, numeric,
  character, etc

```
# Vector creation
a <- c(1,2,3)
a
b <- seq(0,10, by=2)
b
c <- rep(b, times=5)
c</pre>
```



# Vectors - Logical

Operators	Meaning	
<	Less than	
<=	Less than or equal to	
>	More than	
>=	More than or equal to	
==	Equal to	
!=	Not equal to	
a	Not a	
a b	a or b	
a&b	a and b	
isTRUE(a)	Test if a is true	



## Special values - NA

In R, the NA values are used to represent missing values. (NA stands for "not available.") You may encounter NA values in text loaded into R (to represent missing values) or in data loaded from databases (to replace NULL values).

```
> x <- c(1:3,NA)
> is.na(x)
[1] FALSE FALSE FALSE TRUE
```



# Special values – Inf & -Inf

If a computation results in a number that is too big, R will return Inf for a positive number and -Inf for a negative number (meaning positive and negative infinity, respectively):

```
> 1/0
[1] Inf
> 2^35670
[1] Inf
> -3^4567789
[1] -Inf
```



## Special values - Nana

If the result of a statement doesn't make sense Not a Number will be the result

```
> log(-34)
[1] NaN
Warning message:
In log(-34) : Se han producido NaNs
> 0/0
[1] NaN
```



# Special values - Null

NULL represents the lack of existence (vs. Na or NaN)

```
> x<-c(8,2,3,NULL)
> x
[1] 8 2 3
> x[2]<- NULL
Error in x[2] <- NULL : replacement has length zero
> x
[1] 8 2 3
> x[2]<-NaN
> x
[1] 8 NaN 3
```



# Vectors – Subsetting

- An element of a vector v is assigned an index by its position in the sequence, starting with 1
  - The basic function for subsetting is []
  - x[1] is the first element,
     x[length(x)] is the last.

```
> x <- c("a","b","c","d","e","f")
> x[2]
[1] "b"
> x[length(x)]
[1] "f"
```



# Vectors – Adding an element

```
> x < -c(1,2,3)
> X
[1] 1 2 3
> x[3] <- 45
> X
[1] 1 2 45
> x[7] <- 48
> X
[1] 1 2 45 NA NA NA 48
> append(x, c(1,23))
[1] 1 2 45 NA NA NA 48 1 23
> y <- c(45,23,32)
> x \leftarrow c(x,y)
> X
 [1] 1 2 45 NA NA NA 48 45 23 32
> x <- append(x, 34, after=4)
> X
 [1]
        2 45 NA 34 NA NA 48 45 23 32
```



# Exercice - Vector Manipulation

- Create a variable and store the number 25
- Calculate the square root of 25
- Create a vector with the first 100 numbers
- Create a vector to hold the data of a student (grade, course and class, all of them numeric)
  - access the element with the name "grade"
  - Change the data to character data
- Create a vector with the first 50 numbers
  - Create another vector withe the numbers 51-100
  - Add them together
  - Multiply them
  - Join both of them in a larger vector



#### Useful functions

#### Keeping track of your work

```
# Save the commands used during the session
savehistory(file="mylog.Rhistory")
# Load the commands used in a previous session
loadhistory(file="mylog.Rhistory")
# Display the last 25 commands
history()
# You can read mylog.Rhistory with any word processor. Notice that the file has to have the extension *.Rhistory
```



#### Useful functions

```
objects()  # Lists the objects in the workspace

ls()  # Same as objects()

remove()  # Remove objects from the workspace

rm(list=ls())  #clearing memory space

detach(package:ABC)  # Detached packages when no longer need them

search()  # Shows the loaded packages

library()  # Shows the installed packages

dir()  # show files in the working directory
```



#### An ordered collection of heterogenous variables

#### Lists

```
> student <-list(name="Josep", surname="Curto Díaz",
                 male=TRUE, age=39,grades=list(5,6,7,8))
> student[1]
$name
[1] "Josep"
> student$age
[1] 39
> student
$name
[1] "Josep"
$surname
[1] "Curto Díaz"
$male
[1] TRUE
$age
[1] 39
$grades
$grades[[1]]
[1] 5
$grades[[2]]
[1] 6
$grades[[3]]
[1] 7
$grades[[4]]
[1] 8
```



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#### Matrices

A multi-dimension object that keeps the underlying class

```
> m <- matrix(data=1:12, nrow=4, ncol=3, dimnames = list(c("r1","r2","r3","r4"),c("c1","c2"
,"c3")), byrow=TRUE)
> m
    c1 c2 c3
r1    1    2    3
r2    4    5    6
r3    7    8    9
r4    10    11    12
```



### Matrices - indexing

```
> m[1,3]
[1] 3
> m[1,]
c1 c2 c3
1 2 3
> m[,3]
r1 r2 r3 r4
3 6 9 12
> m["r1", "c2"]
[1] 2
> m[c(1:2),c(1:3)]
  c1 c2 c3
```



# Matrices – adding elements

```
> m <- rbind(m, r5=c(13,14,15))
> m
   c1 c2 c3
r1 1 2 3
r2 4 5 6
                                    > m <- cbind(m, c4=c(34,23,12,11,67))
r3 7 8 9
                                    > m
r4 10 11 12
                                       c1 c2 c3 c4
r5 13 14 15
                                    r1 1 2 3 34
                                    r2 4 5 6 23
                                    r3 7 8 9 12
                                    r4 10 11 12 11
                                    r5 13 14 15 67
```



### Matrices – editing values



### Matrices – combining

```
> m2 <- matrix(data=13:32, nrow=5, ncol=4,
            dimnames=list(c("r1", "r2", "r3", "r4", "r5"), c("c1", "c2", "c3", "c4")),byrow=TRUE)
  c1 c2 c3 c4
r1 13 14 15 16
r2 17 18 19 20
r3 21 22 23 24
r4 25 26 27 28
r5 29 30 31 32
> cbind(m,m2)
    col1 col2 col3 col4 c1 c2 c3 c4
row1
       4 5 6 23 17 18 19 20
row2
     7 8 9 12 21 22 23 24
row3
      10 11 12 11 25 26 27 28
row4
           14
               15 67 29 30 31 32
row5
> rbind(m,m2)
    col1 col2 col3 col4
row1
row2
                9 12
row3
row4
row5
           14
           14
                31
                     32
```



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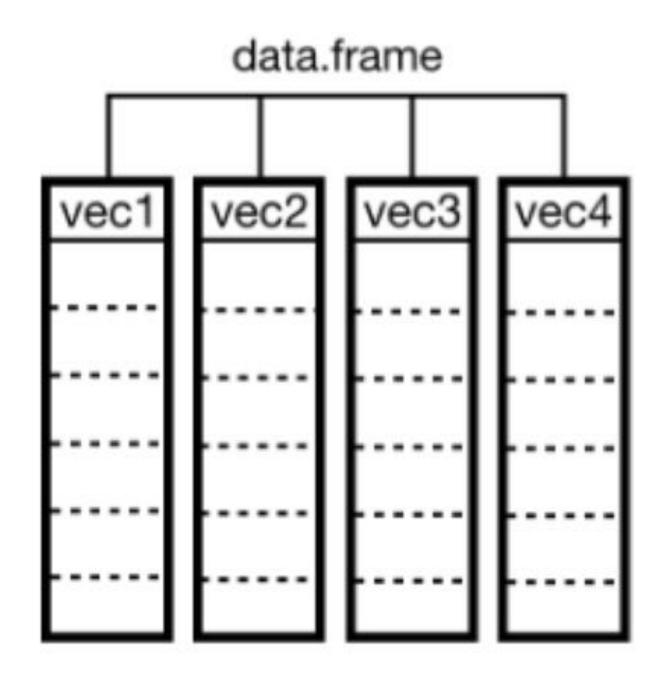
## Matrices – operations

```
> x <- matrix(data=1:12, nrow=4, ncol=3, byrow=TR
> X
     [,1] [,2] [,3]
[1,]
[2,]
[3,]
[4,]
> y[4] <- 4
> dim(y) <- c(4,1)
> y
     [,1]
[1,]
       45
[2,]
      23
[3,]
       32
[4,]
> t(x)
     [,1] [,2] [,3] [,4]
[1,]
[2,]
                     11
[3,]
                  9 12
> diag(4)
     [,1] [,2] [,3] [,4]
[1,]
[2,]
[3,]
[4,]
```



#### Dataframe – collection of vectors

Tabular (rectangular) data structure, which means that it has rows and columns. It is not implemented by a matrix, however. Rather, a data frame is a list:





#### Exercise - Dataframe manipulation

- Load the test dataset
  - load("dataTB.rdatq")
- What is the median of the final price?
- What is the relationship between final price and user price?
- What is the event with the highest final price?



# Session Wrap-up







# 



