

# Lecture 2: Atomic Data Types/Homogeneous vectors

Wim R.M. Cardoen, PhD

## Contents

<b>1</b>	<b>R Objects</b>	<b>2</b>
1.1	Examples . . . . .	2
<b>2</b>	<b>Atomic Data Types</b>	<b>4</b>
2.1	The core/atomic data types . . . . .	4
2.1.1	Examples . . . . .	4
2.2	Operations on atomic data types . . . . .	6
2.2.1	Examples . . . . .	6
<b>3</b>	<b>Atomic vectors</b>	<b>8</b>
3.1	Creation of atomic vectors . . . . .	8
3.1.1	Examples . . . . .	8
3.2	Operations on vectors: element-wise . . . . .	10
3.2.1	Examples . . . . .	10
3.3	Retrieving elements of vectors . . . . .	11
3.3.1	Examples . . . . .	11
3.4	Hash tables . . . . .	13
3.4.1	Examples . . . . .	14
3.5	NA (Not Available values) . . . . .	15
3.5.1	Examples . . . . .	15
3.6	Alia . . . . .	15
<b>4</b>	<b>Matrices &amp; Arrays</b>	<b>16</b>
4.1	Creation of matrices . . . . .	16
4.1.1	Examples . . . . .	16
4.2	Operations on matrices . . . . .	16
4.3	Retrieving elements of matrices . . . . .	16
4.4	Hash Tables . . . . .	16
<b>5</b>	<b>Other types</b>	<b>17</b>
	Other topics on Data structures . . . . .	18
	Conditionals & Loops . . . . .	18
	Environments . . . . .	18
	Functions . . . . .	18
	Capita selecta . . . . .	18

R can be summarized in three principles (John M. Chambers, 2016)

- Everything that exists in R is an **object**.
- Everything that happens in R is a **function call**.
- **Interfaces** to other languages are a part of R.

# 1 R Objects

- An object in R is (internally) represented as a pair: (**symbol**, **value**).
- A **symbol** is assigned a **value** by the use of an arrow pointing to the left (**<-**).
- There are **less favored** ways:
  - A simple equality sign (**=**).
  - Using the **assign()** function.

## 1.1 Examples

- Clean up the global environment i.e. remove all objects from the current R environment.  
**Recommended!**

```
rm(list=ls())  
ls()
```

```
## character(0)
```

- **preferred** way to assign variables

```
x <- 5.0  
x
```

```
## [1] 5
```

- alternative 1: mainly used to assign default function arguments

```
y = 5.0  
y
```

```
## [1] 5
```

```
mysamplevariance <- function(x, av=0){  
  
  n <- length(x)  
  if(n>1){  
    return(1.0/(n-1)*sum((x-av)^2))  
  }  
  else{  
    stop("ERROR:: Dividing by zero (n==1) || (n==0) ")  
  }  
}
```

```
x <- rnorm(10)  
mysamplevariance(x)
```

```
## [1] 1.896407
```

```
mysamplevariance(x,mean(x))
```

```
## [1] 1.827476
```

```
var(x)
```

```
## [1] 1.827476
```

- alternative 2: even less used

```
assign("z", 5.0)
```

```
z
```

```
## [1] 5
```

- functions are objects

```
f <- mean
```

```
f
```

```
## function (x, ...)
```

```
## UseMethod("mean")
```

```
## <bytecode: 0x55dd1bb20f70>
```

```
## <environment: namespace:base>
```

```
val <- f(1:10)
```

```
val
```

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

## 2 Atomic Data Types

Nothing exists except atoms and empty space; everything else is opinion. (Democritus)

### 2.1 The core/atomic data types

- R has the following 6 **atomic** data types:
  - logical (i.e. boolean)
  - integer
  - double
  - character (i.e. string)
  - complex
  - raw (i.e. byte)

The latter 2 types (i.e. complex and especially raw) are less common.

The **typeof()** function determines the **INTERNAL** storage/type of an R object.

#### 2.1.1 Examples

- boolean/logical values: either **TRUE** or **FALSE**

```
x1 <- TRUE
x1
```

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

```
typeof(x1)
```

```
## [1] "logical"
```

- integer values ( $\in \mathbb{Z}$ ):

```
x2 <- 3L
x2
```

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

```
typeof(x2)
```

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

- double (precision) values:

```
x3 <- 3.14
x3
```

```
## [1] 3.14
```

```
typeof(x3)
```

```
## [1] "double"
```

- character values/strings

```
x4 <- "Hello world"
x4
```

```
## [1] "Hello world"
```

```
typeof(x4)
```

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

- complex values ( $\in \mathbb{C}$ ):

```
x5 <- 2.0 + 3i
x5
```

```
## [1] 2+3i
```

```
typeof(x5)
```

```
## [1] "complex"
```

## 2.2 Operations on atomic data types

- **logical** operators: `==`, `!=`, `&&`, `||`, `!`
- **numerical** operators: `+`, `-`, `*`, `/`, `^`, `**` (same as the caret), but also:
  - integer division: `%/%`
  - modulo operation: `%%`
  - **Note**: matrix multiplication will be performed using `%*%`
- **character/string** manipulation:
  - `nchar()`:
  - `paste()`:
  - `cat()`:
  - `sprintf()`:
  - `substr()`:
  - `strsplit()`:
  - **Note**: Specialized R libraries were developed to manipulate strings e.g. *stringr*
- explicit **cast**/conversion: <https://data-flair.training/blogs/r-string-manipulation/>
  - `as.{logical, integer, double, complex, character}()`
- explicit **test** of the type of a variable:
  - `is.{logical, integer, double, complex, character}()`

### 2.2.1 Examples

- Logical operators:

```
x <-3
y <-7
(x<=3) &&(y==7)
```

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

```
!(y<7)
```

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

- Mathematical operations

```
2**4
```

```
## [1] 16
```

```
7%%4
```

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

```
7/4
```

```
## [1] 1.75
```

```
7%/%4
```

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

- String operations

```
s <- "Hello"
nchar(s)

## [1] 5

news <- paste(s,"World")
news

## [1] "Hello World"

sprintf("My new string:%20s\n", news)

## [1] "My new string:          Hello World\n"

city <- "Witwatersrand"
substr(city,4,8)

## [1] "water"
```

- Conversion and testing of types

```
s <- "Hello World"
is.character(s)

## [1] TRUE
```

```
s1 <- "-500"
is.character(s1)

## [1] TRUE
```

```
s2 <- as.double(s1)
is.character(s2)

## [1] FALSE

is.double(s2)

## [1] TRUE
```

```
s3 <- as.complex(s2)
s3

## [1] -500+0i

sqrt(s3)

## [1] 0+22.36068i
```

## 3 Atomic vectors

- An **atomic** vector is a data structure containing elements of **only one atomic** data type. Therefore, an atomic vector is **homogeneous**.
- Atomic vectors are stored in a **linear** fashion.
- R does **NOT** have scalars:
  - An atomic vector of **length 1** plays the role of a scalar.
  - Vectors of **length 0** also exist (and they have some use!).
- A **list** is a vector not necessarily of the atomic type.  
A list is also known as a **recursive/generic** vector (*vide infra*).

### 3.1 Creation of atomic vectors

Atomic vectors can be created in a multiple ways:

- Use of the **vector()** function.
- Use of the **c()** function (**c** stands for concatenate).
- Use of the column operator **:**
- Use of the **seq()** and **rep()** functions.

The length of a vector can be retrieved using the **length()** function.

#### 3.1.1 Examples

- use of the **vector()** function:

```
x <- vector() # Empty vector (Default:'logical')
x
```

```
## logical(0)
```

```
length(x)
```

```
## [1] 0
```

```
typeof(x)
```

```
## [1] "logical"
```

```
x <- vector(mode="complex", length=4)
x
```

```
## [1] 0+0i 0+0i 0+0i 0+0i
```

```
length(x)
```

```
## [1] 4
```

```
x
```

```
## [1] 0+0i 0+0i 0+0i 0+0i
```

```
x[1] <- 4
```

```
x
```

```
## [1] 4+0i 0+0i 0+0i 0+0i
```



- use of the `c()` function:

```
x1 <- c(3, 2, 5.2, 7)
x1

## [1] 3.0 2.0 5.2 7.0

x2 <- c(8, 12, 13)
x2

## [1] 8 12 13

x3 <- c(x2, x1)
x3

## [1] 8.0 12.0 13.0 3.0 2.0 5.2 7.0

x4 <- c(FALSE, TRUE, FALSE)
x4

## [1] FALSE TRUE FALSE

x5 <- c("Hello", "Salt", "Lake", "City")
x5

## [1] "Hello" "Salt" "Lake" "City"
```

- use of the column operator:

```
y1 <- 1:10
y1

## [1] 1 2 3 4 5 6 7 8 9 10

y2 <- 5:-5
y2

## [1] 5 4 3 2 1 0 -1 -2 -3 -4 -5

y3 <- 2.3:10
y3

## [1] 2.3 3.3 4.3 5.3 6.3 7.3 8.3 9.3

y4 <- 2.0*7:1
y4

## [1] 14 12 10 8 6 4 2

y5 <- 1:7-1
y5

## [1] 0 1 2 3 4 5 6
```

- `seq()` and `rep()` functions

```
z1 <- seq(from=1, to=15, by=3)
z1

## [1] 1 4 7 10 13
```

```
z2 <- seq(from=-2,to=5,length=4)
z2

## [1] -2.0000000  0.3333333  2.6666667  5.0000000
```

```
z3 <- rep(c(3,2,4), time=2)
z3
```

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

```
z4 <- rep(c(3,2,4), each=3)
z4
```

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

```
z5 <- rep(c(1,7), each=2, time=3)
z5
```

```
## [1] 1 1 7 7 1 1 7 7 1 1 7 7
```

```
length(z5)
```

```
## [1] 12
```

## 3.2 Operations on vectors: element-wise

- All operations on vectors in R happen **element by element** (cfr. *NumPy*).
- **Vector Recycling**:

If 2 vectors of **different** lengths are involved in an operation, the **shortest vector** will be repeated until all elements of the longest vector are matched.

A message will be sent to the stdout.

### 3.2.1 Examples

```
x <- -3:3
x
```

```
## [1] -3 -2 -1  0  1  2  3
```

```
y <- 1:7
y
```

```
## [1] 1 2 3 4 5 6 7
```

```
xy <- x*y
xy
```

```
## [1] -3 -4 -3  0  5 12 21
```

```
xpy <- x^y
xpy
```

```
## [1] -3  4 -1  0  1 64 2187
```

```

x <- 0:10
y <- 1:2
length(x)

## [1] 11
length(y)

## [1] 2
x

## [1] 0 1 2 3 4 5 6 7 8 9 10
y

## [1] 1 2
x+y

## Warning in x + y: longer object length is not a multiple of shorter object
## length
## [1] 1 3 3 5 5 7 7 9 9 11 11

```

### 3.3 Retrieving elements of vectors

- Indexing: starts at **1** (**not 0** like C/C++, Python, Java, ...) see also: [Edsger Dijkstra: Why numbering should start at zero](#)
- Use of vector with indices to extract values.
- Advanced features:
  - use of boolean values to extract values.
  - the membership operator: **%in%**.
  - the deselect/omit operator: **-**
  - **which()**: returns the indices for which the condition is true.
  - **any()/all()** functions.
    - \* **any()** : **TRUE** if at least 1 value is true
    - \* **all()** : **TRUE** if all values are true

#### 3.3.1 Examples

- Use of a simple index:

```

x <- seq(2,100,by=15)
x[4]

## [1] 47
x[1]

## [1] 2

```

- Select several indices at once using vectors:

```
x
## [1]  2 17 32 47 62 77 92
x[3:5]
## [1] 32 47 62
x[c(1,3,5,7)]
## [1]  2 32 62 92
x[seq(1,7,by=2)]
## [1]  2 32 62 92
```

- Extraction via booleans (i.e. retain only those values that are equal to **TRUE**):

```
x
## [1]  2 17 32 47 62 77 92
x>45
## [1] FALSE FALSE FALSE  TRUE  TRUE  TRUE  TRUE
x[x>45]
## [1] 47 62 77 92
```

- Use of the **%in%** operator:

```
x
## [1]  2 17 32 47 62 77 92
10 %in% x
## [1] FALSE
62 %in% x
## [1] TRUE
c(32,33,43) %in% x
## [1]  TRUE FALSE FALSE
!(c(32,33,43) %in% x)
## [1] FALSE  TRUE  TRUE
```

- Negate/filter out the elements with **negative** indices:

```
x
## [1]  2 17 32 47 62 77 92
```

```
x[-c(2,4,6)]

## [1]  2 32 62 92
z <- x[-1] - x[-length(x)]
z

## [1] 15 15 15 15 15 15
```

- The `which()` function returns **only those indices** of which the condition/expression is **true**.

```
# Sample 10 numbers from N(0,1)
vecnum <- rnorm(n=10)
vecnum

## [1] -1.41132225 -0.41176787  0.59329962  0.81029508  0.09852538 -0.89015581
## [7]  1.21601739 -0.49217006  0.05142668 -1.76690416
which(vecnum>1.0)

## [1] 7
```

- Use of the `any()/all()` functions.

```
y <- seq(0,100,by=10)
x

## [1]  2 17 32 47 62 77 92
y

## [1]  0 10 20 30 40 50 60 70 80 90 100
any(x<y)

## Warning in x < y: longer object length is not a multiple of shorter object
## length
## [1] TRUE
all(x[6:7]>y[2:3])

## [1] TRUE
```

### 3.4 Hash tables

A **hash table** is a data structure which implements an associative array or dictionary. It is an abstract data which maps data to keys.

- There are several ways to create one:
  - Map names to an existing vector
  - Add names when creating the vector
- To remove the map, map the names to NULL

### 3.4.1 Examples

- Creation of 2 independent vectors

```
capitals <- c("Albany", "Providence", "Hartford", "Boston", "Montpelier", "Concord", "Augusta")
states <- c("NY", "RI", "CT", "MA", "VT", "NH", "ME")
capitals

## [1] "Albany"      "Providence" "Hartford"    "Boston"      "Montpelier"
## [6] "Concord"     "Augusta"
states

## [1] "NY" "RI" "CT" "MA" "VT" "NH" "ME"
capitals[3]

## [1] "Hartford"
```

- Create the hashtable/dictionary

```
# Method 1
names(capitals) <- states
capitals

##           NY           RI           CT           MA           VT           NH
## "Albany" "Providence" "Hartford" "Boston" "Montpelier" "Concord"
##           ME
## "Augusta"
capitals["MA"]

##           MA
## "Boston"
names(capitals)

## [1] "NY" "RI" "CT" "MA" "VT" "NH" "ME"

# Method 2
phonecode <- c("801"="SLC", "206"="Seattle", "307"="Wyoming")
phonecode

##           801           206           307
## "SLC" "Seattle" "Wyoming"
phonecode["801"]

##           801
## "SLC"
```

- Dissociate the 2 vectors

```
names(capitals) <- NULL
capitals
```

```
## [1] "Albany"      "Providence" "Hartford"   "Boston"     "Montpelier"
## [6] "Concord"     "Augusta"
```

### 3.5 NA (Not Available values)

- **NA**: stands for ‘Not Available’/Missing values
- has length of 1.
- **is.na()**: test all elements of a vector for NA values.
- some functions e.g. **mean()** return NA when an instance of NA is present.

#### 3.5.1 Examples

- Check of the NA availability

```
x <- c(NA, 1, 2, NA)
is.na(x)
```

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

- Functions on a vector containing NA

```
mean(x)
```

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

```
mean(x, na.rm=TRUE)
```

```
## [1] 1.5
```

### 3.6 Alia

**Still to be developed!**

- boolean: Vector operators vs. unique value
- && vs. &.
- || vs. |.
- xor()

## 4 Matrices & Arrays

Matrices and arrays are **homogeneous atomic vectors** with an **extra** attribute: `dimension`

By default, the elements are stored in a **column-major** fashion. (cfr. **Fortran**). However, we can store the elements in **row-major** order (cfr. **C**) as well.

### 4.1 Creation of matrices

Matrices can be created in different ways:

- use of the `matrix()` function
- use of `rbind()/cbind()`
- set the `attributes()` of a vector
- special functions like e.g. `diag()`

#### 4.1.1 Examples

### 4.2 Operations on matrices

- Being in essence a vector, operations like `*`, `/`, `+` happen element-wise.
- There are also specialized functions:
  - `rowMeans()`, `colMeans()`: row and column means
  - `%*%`: matrix multiplication
  - linear algebra functions, e.g.:
    - \* inversion
    - \* det
    - \* transpose
    - \* eigen
    - \* ...

### 4.3 Retrieving elements of matrices

### 4.4 Hash Tables



## 5 Other types

- Attributes
- Special types:
  - Factors
  - Date
  - Time
- NA, NaN, NULL

## Other topics on Data structures

- List
- Dataframe & Tibble
- IO (read.csv, read.file)
- Names
- Subsetting, `[[` vs. `[]`

## Conditionals & Loops

- if, else, else if switch and elseif
- for
- while
- repeat
- return()

## Environments

- search(), attach, detach
- library

## Functions

- lexical scoping
- simple functions
- args(), formals()
- default arg, ...
- lazy evaluation
- closure
- anonymous functions
- make your own operators
- loop functions: `{l,s,m}apply`, `split`

## Capita selecta

- profiling, debugging