Data Structures: Vectors

Enrico Toffalini

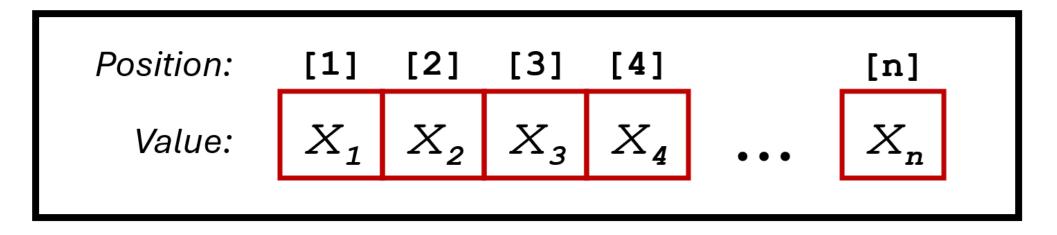
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What Are Data Structures

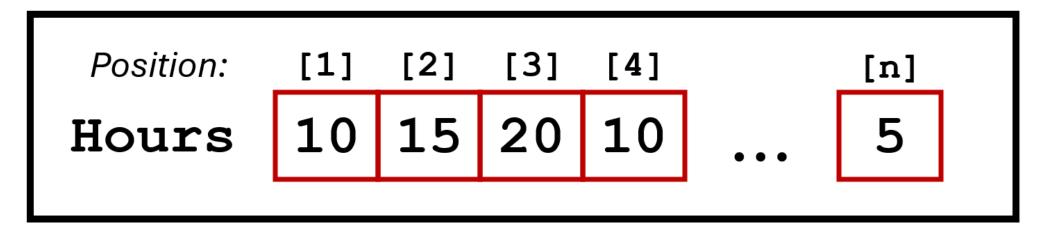
Data structures, like vectors, matrices, dataframes, **lists**, are fundamental tools that allow you to **organize** and store complex information, so that they can be easily **processes by functions** (e.g., 1m() function may fit a linear model on variables stored in a dataframe) Most operations you will perform in R (e.g., *processing*) data, fitting models, plotting outputs) are performed on these data structures

Vectors

Simple one-dimensional structures that store data of different types



Here is an actual **example** (of a *numerical* vector):



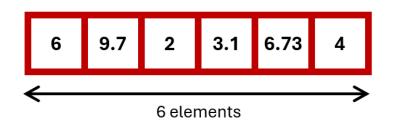
Vectors as 1-D Arrays

Vectors are just special cases of arrays

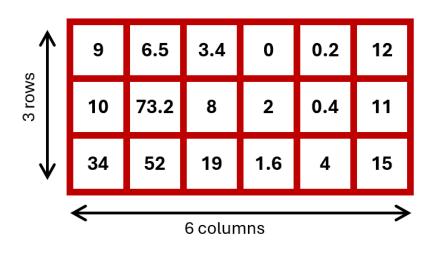
Scalar 0-Dimensional array



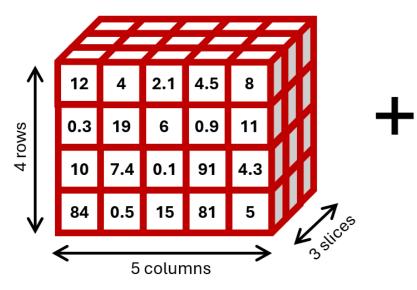
Vectors 1-Dimensional array



Matrix 2-Dimensional array



3-Dimensional *Array*



Create Vectors with c()

Vectors can easily be **created using the c()** base function, with a sequence of elements separated by *commas* ","

Vectors can be of different types. The following example shows a *character* vector (note the *quotes* " " around objects):

or numeric:

Hours = c(10, 15, 20, 10, 15, 5, 15, 5)

Vectors Must be Homogeneous

Vectors must contain elements of the **same type**. If you mix types, R will automatically **coerce** the elements to a single type, which may lead to undesired results.

Therefore, avoid mixing data types! Example:

Hours = c(10, 15, 20, 10, 15, "tbd", 15, 5)

Hours

[1] "10" "15" "20" "10" "15" "tbd" "15" "5"

everything was coerced to become a character!

If needed, use NA (Not Available):

Hours = c(10, 15, 20, 10, 15, NA, 15, 5) Hours # remains a numerical vector, NA does not affect type

[1] 10 15 20 10 15 NA 15 5

Vectors Must be Homogeneous

You may coerce a vector to be a particular type if needed

```
Hours = c(10, 15, 20, 10, 15, "tbd", 15, 5)
Hours
[1] "10" "15" "20" "10" "15" "tbd" "15" "5"
```

```
as.numeric(Hours)
```

```
Warning: NAs introduced by coercion
```

```
[1] 10 15 20 10 15 NA 15 5
```

But be careful! Elements that cannot be coerced to the target type, will be replace with NA

Hours = c("10", "15,", "20", " 10", "15 ", "tbd", "15.", "5_") as.numeric(Hours)

Warning: NAs introduced by coercion

```
[1] 10 NA 20 10 15 NA 15 NA
```

Indexing Vectors

Select/extract elements with **INDEXING** using square brackets []:

Hours = c(10, 15, 20, 10, 15, 5, 15, 5) Hours[4] # a single element

[1] 10

Hours [5:7] # a range of elements

[1] 15 5 15

Hours[c(1,3,6)] # specific elements

[1] 10 20 5

Know the **length** of a vector using the **length()** function, and use it:

length(Hours)

[1] 8

Hours[length(Hours)] # use it to extract the last element

[1] 5

Indexing Vectors

Negative indexing

You can use the *minus* sign - to select **all elements except some** from a vector. (This method is also applicable to dataframes)

Hours = c(10, 15, 20, 10, 15, 5, 15, 5) Hours[-4] # ALL BUT a single element [1] 10 15 20 15 5 15 5 Hours[-c(5:7)] # ALL BUT a range of elements [1] 10 15 20 10 5 Hours[-c(1,3,6)] # ALL BUT specific elements [1] 15 10 15 15 5 Hours[-length(Hours)] # ALL BUT the last element [1] 10 15 20 10 15 5 15

Logical Indexing

Often, you'll need to extract values from a vector based on specific *logical* conditions. Here's an example:

Hours = c(10, 15, 20, 10, 15, 5, 15, 5)

Hours [Hours >= 15] # extract only values greater than or equal to 15

[1] 15 20 15 15

This is called *logical indexing* because you are selecting elements based on a logical vector (i.e., a sequence of TRUE, FALSE):

Hours >= 15 # the logical vector actually inside the square brackets

[1] FALSE TRUE TRUE FALSE TRUE FALSE TRUE FALSE

Also, you can use a vector to extract values **from another vector**:

Teachers[Hours >= 15]

[1] "Kiesner" "Granziol" "Calignano" "Bastianelli"

Indexing and Assignment

With indexing, you can not only select, but also **assign or modify** elements in a vector:

Hours = $c($	0, 15, 20, 10, 15, 5, 15, 5)
	<pre>0 # assign a new value Hours[3]+50 # modify an existing element</pre>

[1] 0 15 70 10 15 5 15 5

You can even assign values **outside the current range** of the vector. But what happens?

Hours $[20] = 5$ Hours										
	[1]	0 15 70 10 15	5 15	5 NA N	ia na na	A NA NA	NA NA	NA NA	NA 5	

Operating on Vectors

you can simultaneously apply an operation to a whole vector, like

```
Hours = c(10, 15, 20, 10, 15, 5, 15, 5)
Hours / 5
```

[1] 2 3 4 2 3 1 3 1

Of course, this is useful when you want to save the result as a new vector:

ECTS = Hours / 5

Similarly, you can apply functions to all elements of a vector:

sqrt(Hours) # computes square root of each element

[1] 3.162278 3.872983 4.472136 3.162278 3.872983 2.236068 3.872983 2.236068

log(Hours) # computes the natural logarithm of each element

[1] 2.302585 2.708050 2.995732 2.302585 2.708050 1.609438 2.708050 1.609438

Summary Statistics on Vectors

A whole vector may serve to compute summary statistics, for example using functions such as mean(), sd(), median(), quantile(), max(), min():

mean(Hours) # returns the average value (mean) of the vect
[1] 11.875
<pre>sd(Hours) # returns the Standard Deviation of the vector</pre>
[1] 5.303301
<pre>median(Hours) # returns the median value of the vector</pre>
[1] 12.5

Summary Statistics on Vectors

A whole vector may serve to compute summary statistics, for example using functions such as mean(), sd(), median(), quantile(), max(), min():

	quantile(H	ours, prob	s=c(.25,	.50, .	.75))	# returns	desired
	50% 75% 2.50 15.00						
	max(Hours)	# returns	s largest	value			
[1] 20							
	min(Hours)	# returns	s smalles	t value	2		
[1] 5							

Summary Statistics - Managing Missing (NA) Values

All of the previous summary statistics will **fail** if there is even a single NA value:

Hours = c(10, 15, 20, 10, 15, NA, 15, 5)

mean(Hours) # a single NA value implies that the average is impossible

[1] NA

quantile(Hours, probs=c(.25, .75)) # quantile() will even return an Err

Error in quantile.default(Hours, probs = c(0.25, 0.75)): missing values and NaN's not allowed if 'na.rm' is FALSE

You can easily manage missing values by adding the **na.rm=TRUE** argument:

mean(Hours, na.rm=TRUE) # NA values are ignored

[1] 12.85714

```
quantile(Hours, probs=c(.25, .75), na.rm=TRUE) # NA values are ignored
25% 75%
10 15
```

Replacing NA With the Average Value

Replacing a missing value with the average across valid values is risky, as it may alter many other summary statistics, but it is a good example for understanding different concepts seen so far:

```
Hours = c(10, 15, 20, 10, 15, NA, 15, 5)
```

compute the average value ignoring NAs, and put it wherever # there is a NA value in the vector Hours[is.na(Hours)] = mean(Hours, na.rm=TRUE)

now let's inspect the updated content of the vector
Hours

[1] 10.00000 15.00000 20.00000 10.00000 15.00000 12.85714 15.00000 5.00000

by the way... na.rm=TRUE is no longer needed now, as NA is no longer mean(Hours)

```
[1] 12.85714
```

Frequency Counts

Another useful summary statistic is the **frequency count**, which shows how often each unique value appears in a vector. You can use the **table()** function to calculate frequencies easily:

```
type = c("METHODOLOGY", "METHODOLOGY", "PROGRAMMING", "SOFT SKILLS", "S
                  "METHODOLOGY", "SOFT SKILLS", "METHODOLOGY", "PROGRAMMING")
         table(type)
type
METHODOLOGY PROGRAMMING SOFT SKILLS
                       2
                                    3
          4
Be careful: R is case sensitive!
         type = c("METHODOLOGY", "methodology", "PROGRAMMING", "SOFT SKILLS", "$
                  "METHODOLOGY", "SOFT SKILLS", "METHODOLOGY", "Programming")
         table(type)
type
methodology METHODOLOGY Programming PROGRAMMING SOFT SKILLS
                       3
                                                             3
                                    1
                                                1
```