# R Basics

36-600

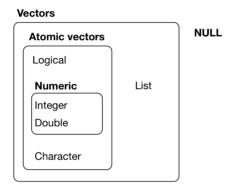
Week 1 Thursday – Fall 2021

#### **Motivation**

A *vector* in R is a homogeneous collection of numbers, strings, or TRUE/FALSE values (i.e., logicals).

A collection of (column) vectors, all of the same length, can be bound together into a data table (called a *data frame*) that might provide the input to, e.g., a regression analysis. So in our discussion of statistical learning, it makes sense to start with a discussion of R vectors and how you might use (and manipulate!) them.

#### Vectors in R



Atomic vectors are *homogeneous*, i.e., all elements of the vectors are of the same type. The important types of atomic vectors (or just vectors) for our purposes are:

- double: double-precision floating-point numbers (8 bytes per element);
- integer: integer numbers (4 bytes per element);
- logical: TRUE and FALSE; and
- character: individual strings (at 1 byte per individual character within each string).

Note: the integer and double types are collectively (and at times confusingly) dubbed numeric. (Confusingly because one can cast to a numeric type, which is equivalent to casting to double.)

My advice: unless memory storage is an issue, just forget that R supports integer vectors, and assume that all the numerical variables you will deal with will be contained in double-precision numerical vectors.

#### **Initializing Vectors**

Let's show the various ways in which one can initialize a vector of five integers:

Function Call	
x <- c(0,0,0,0,0)	c = "collection" or "container"
x <- rep(0,5)	rep = "repeat"
<pre>x &lt;- vector("numeric",5)</pre>	
x <- numeric(5)	
x <- seq(1,5,by=1)	seq = "sequence"
x <- 1:5	steps by 1

We can use all six of these functions to initialize numeric vectors, and the first four to initialize those of mode logical or character, as with, e.g.,

```
vector("logical",5)
```

## [1] FALSE FALSE FALSE FALSE

```
character(5)
```

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

Note that in my own life, I almost always use rep() (e.g., rep("",100) or rep(TRUE,200), etc.).

#### **Initializing Vectors**

A few more points to make here:

• You can combine initialization functions, which can be helpful:

```
x <- c(rep(0,5),11:14,numeric(3))
x
## [1] 0 0 0 0 0 11 12 13 14 0 0 0</pre>
```

• You can concatenate vectors too:

```
x <- 1:3
y <- 78:83
(z <- append(x,y))
## [1] 1 2 3 78 79 80 81 82 83</pre>
```

(Why the parentheses? It's an R trick: you can assign to a new variable *and* print its contents in one line of code.)

• Note that the assignment operator <-. An equals sign, =, works too, but purists use <- and thus we'll start purely. (You will see me slip up throughout the semester and type = because I learned R on my own long ago and wasn't indoctrinated in the use of <-.)

#### Handy Vector Functions

To determine the type of a vector:

```
x <- c(1,0,3,2)
typeof(x)
```

## [1] "double"

To determine the number of elements in a vector:

```
length(x)
```

## [1] 4

To display the  $n^{th}$  element of a vector, where  $n \in [1, length(x)]$ :

```
x[1]
```

## [1] 1

To explicitly cast from one type to another:

```
as.character(x)
```

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

#### Handy Vector Functions

To display multiple vector elements, pass in a defined vector:

```
x[c(1,4)]
## [1] 1 2
x[1:2]
## [1] 1 0
```

To remove a vector element (from output...not from the vector itself!), utilize the minus sign:

```
x[-2]
```

## [1] 1 3 2

To remove multiple vector elements, pass in a minus sign and a defined vector:

## [1] 3 2 7/14

### Handy Vector Functions

To sort a vector in ascending order, and to retrieve the sorted vector indices:

```
Χ
## [1] 1 0 3 2
sort(x)
## [1] 0 1 2 3
order(x)
## [1] 2 1 4 3
To display the unique values of a vector:
unique(x)
## [1] 1 0 3 2
table(x)
## x
## 0 1 2 3
## 1 1 1 1
```

### **Logical Subsetting**

Relational Operators in R	
Operator	Description
<	Less than
>	Greater than
<=	Less than or equal to
>=	Greater than or equal to
==	Equal to
!=	Not equal to

If you apply a relational operator to a vector, the output will be a logical vector:

```
set.seed(101)
x <- rnorm(10)
x>0
```

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

### **Logical Subsetting**

**;MUY IMPORTANTE!** If you apply a logical vector of length n to a vector of length n, then *only the elements of the second vector associated with the value TRUE will be displayed!* For instance:

```
x
## [1] -0.3260365 0.5524619 -0.6749438 0.2143595 0.3107692 1.1739663 0.6187899 -0.
x[x>0]
```

### **Logical Subsetting**

The output from relational operators can be combined using the logical and operator (&) or the logical or operator (|):

```
y <- x>0 & x<0.5

x[y]

## [1] 0.2143595 0.3107692

y <- x<0 | x>0.5

x[y]
```

## [1] -0.3260365 0.5524619 -0.6749438 1.1739663 0.6187899 -0.1127343 0.9170283 -0.2

# Logical Subsetting: sum()

To determine how many values in your vector satisfy a condition, combine one or more relational operators with the sum() function:

```
sum(x>-0.5 \& x<0)
```

## [1] 3

## Logical Subsetting: which()

To determine which elements of the original vector satisfy a condition, combine one or more relational operators with the which() function:

```
which(x>-0.5 & x<0)
## [1] 1 8 10</pre>
```

Another means by which to subset a vector is to apply the output of the which() function. Note how adding a minus sign changes the output!

```
w <- which(x<0)
x[w]

## [1] -0.3260365 -0.6749438 -0.1127343 -0.2232594

x[-w]</pre>
```

## [1] 0.5524619 0.2143595 0.3107692 1.1739663 0.6187899 0.9170283

#### Missing Data: NA

NA means "Not Available" and is the preferred way in R to denote missing data.

To determine whether vector elements are NA, we can use the is.na() function, which returns a logical vector.

```
x <- c(1,NA,3)
is.na(x)
```

## [1] FALSE TRUE FALSE