Introduction to R

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This is from the first chapter of learn.r-journalism.com.

Syntax

You must follow pretty specific syntax rules for it to work. R won't guess for you.

- R is case-sensitive (unlike SQL) interreted language (unlike C)
- Can enter commands at prompt > or in batch
- Comments are preceded by #
 - This is important to use often in your code for documentation
 - You'll see it often in this course
- Statements are lines of code such as functions and object assignments
 - Enters in code (as in a new line) or semi-colons will separate statements

Working directories

Your working directory is the folder on your computer in which you are currently working. When you ask R to open a certain file, it will look in the working directory for this file, and when you tell R to save a data file or figure, it will save it in the working directory.

Before you start working, please set your working directory to where all your data and script files are or should be stored.

Tip: When you see code in a black box throughout this class, I want you to run that code in R unless noted otherwise. You can run it in the console, I'd prefer you do so in a script so you can see your history.

In the video, you'll see me swap back and forth between coding in the console and the script. This really depends on if I'm experimenting with code quickly or if it's something I know I should track. Often I will copy and paste code executed from the console to growing script as an afterthought. Also, **be sure to read the commented out code** because I try to give additional context, like tips on what to do if the command fails.

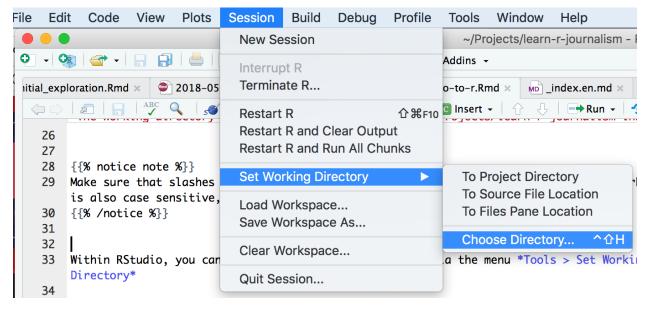
This is an example of setting a working directory to a manual directory on your computer.

```
# On a mac, it'd look like this
setwd("~/projects/learn-r-journalism")

# On a pc, it might look like this
setwd("C:/Documents/learn-r-journalism")
```

Note: Make sure that slashes are forward slashes and that you don't forget the quotation marks.

Within RStudio, you can also set the working directory via the menu Tools > Set Working Directory



The commands above, setwd() was an example of setting an absolute folder.

This works for you for now, but if you wanted to share your methodology and script in the future or if you wanted to save run the code on another computer it would likely **not** work because it would be looking through a folder structure that doesn't exist on any computer except the one where the original script was written. This is not ideal for reproducibility.

Working directories are a tough concept. We'll be going over best practices in Chapter 6 in more detail. Meanwhile, here's a link that might help explain some more.

Libraries

R can do many statistical and data analyses.

They are organized in so-called *packages* or *libraries*.

You can do a lot of statistical analysis in R without any additional libraries—this is called base R.

But other users have created libraries with functions that solve common problems. R package users download only the libraries that they need for an individual project

To get a list of all installed packages, go to the packages window or type library() in the console window. If the box in front of the package name is ticked in the packages window, the package is loaded and the functions within it are ready to be called.

There are many more packages available on the R website. If you want to install and use a package (for example, the packaged called "dplyr") you should:

- Install the package: click install packages in the packages window and type dplyr or type install.packages("dplyr") in the console window.
- Load the package: Check box in front of dplyr or type library("dplyr") in the console window.

Examples of R commands

Calculator

R can be used as a calculator.

Just type an equation in the console window after the >

Note: In those code sections, the code preceded by ## is the output of the code from the lines above.

```
10<sup>2</sup> + 26
## [1] 126
```

Workspace

You can give numbers a name.

By doing so, they become so-called variables which can be used later.

```
a <- 4
```

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

You can do calculations with a now.

```
a*5
```

```
## [1] 20
```

If you specify a again, it will forget what value you had before because you did not assign it to anything.

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

You can also assign a value to a using the old one

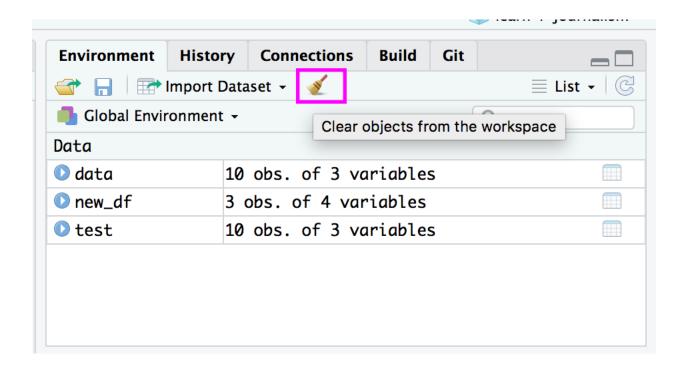
```
a <- a + 10
a
```

```
## [1] 14
```

To remove all variables from R's memory, type

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

or click the "clear all" broom button in the work space window.



Scalars and vectors

Like in many other programs, R organizes numbers in *scalars* (a single number 0-dimensional), *vectors* (a row of numbers, also called arrays - '-dimensional) and *matrices* (which we won't get into now).

The a you defined was scalar.

To define a vector with the numbers 3,4, and 5, you need the function c() which is short for concatenate (or paste together).

```
b=c(3,4,5)
b
```

[1] 3 4 5

Functions

If you would like to compute the mean of all the elements in the vector $\mathfrak b$ from the example above, you could type

```
(3+4+5)/3
```

[1] 4

But when the vector is very long, this is very boring and time-consuming work.

Functions do things to data. R is built on them. Some functions come with R, like median() or summary() and others come as part of packages that others have created.

When you use a function to compute an average, you'll type

```
mean(x=b)
```

[1] 4

Within the brackets you specify the arguments.

Arguments give extra information to the function. In this case, the argument x says of which set of numbers (vector) the mean should computed (namely of b).

Sometimes the name of the argument is not necessary:

```
mean(b)
```

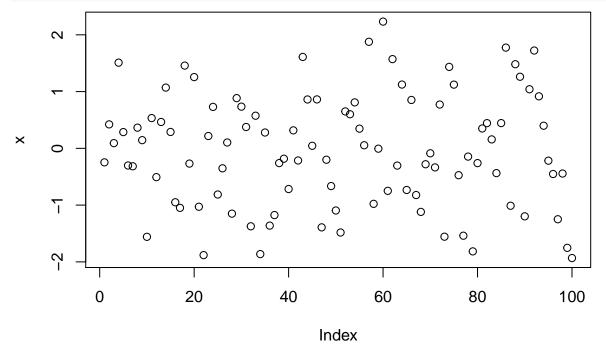
[1] 4

Also works.

Plots

R can make simple graphics right away.

```
# rnorm() is a base function that creates random samples from a random distribution
x <- rnorm(100)
# plot() is a base function that charts
plot(x)</pre>
```



- In the first line, 100 random numbers are assigned to the variable x, which becomes a vector by this operation.
- In the second line, all these values are plotted in the plot window.

Scripts

R is an interpreter that uses a command line based environment.

This means that you have to type commands, rather than use the mouse and menus.

This has the advantage that you do not always have to retype commands.

You can store your commands in files, the so-called scripts. These scripts have typically file names with the extension $.\mathbf{R}$ as in $script.\mathbf{R}$.

You can open an editor window to edit these files by clicking File > New or File > Open file...

You can run (send to the console window) part of the code by selecting lines and pressing **CTRL+ENTER** or **CMD+ENTER** or click the *Run* button at the top of the script editor window. If you do not select anything, R will run the line your cursor is on.

You can always run the whole script with the function source()

For example, to run the entire saved **script.R** if it's in the root directory of the working directory, type

```
source("script.R")
```

You can also click $Run\ all\ in$ the editor window or type $\mathbf{CTRL} + \mathbf{SHIFT} + \mathbf{S}$ or CMD + SHIFT + S

Not available data

When you work with real data, you will encounter missing values because instrumentation failed or human error.

When a data is *not available*, you write NA instead of a number.

```
j < c(1,2,NA)
```

Computing statistics of incomplete data sets is strictly not possible.

maybe the largest value occurred during the weekend when you didn't measure. Therefore, R will say that it doesn't know what the largest value of j is

```
\max(j)
```

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

If you don't mind about the missing data and want to compute the statistics anyway, you can add the argument na.rm=TRUE (Should I remove the NAs? Yes)

```
max(j, na.rm=T)
```

```
## [1] 2
```

Warning: NAs will also affect any sort of math if you're not careful

```
sum(j)
```

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

```
# compared to
sum(j, na.rm=T)
```

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

Here are some links on how to handle NAs in your data

Data types

We've been working so far with numbers.

But sometimes data we work with can be specified as something else, like characters and strings or Boolean values like **TRUE** or **FALSE** or dates.

Characters

m <- "apples"

```
## [1] "apples"

To tell R that something is a character string, you should type the text between apostrophes, otherwise R will start looking for a defined variable with the same name.

n <- pears

## Error in eval(expr, envir, enclos): object 'pears' not found
You can't do math with characters.

m + 2</pre>
```

Dates

Dates and times are complicated.

R has to know that 3 o'clock comes after 2:59 and that February has 29 days in some years.

Error in m + 2: non-numeric argument to binary operator

The base way to tell R that something is a date-time combination is with the function strptime()

```
date1 <- strptime(c("20100225230000", "20100226000000", "20100226010000"), format="%Y%m%d%H%M%S")
date1
## [1] "2010-02-25 23:00:00 EST" "2010-02-26 00:00:00 EST"
## [3] "2010-02-26 01:00:00 EST"</pre>
```

A vector is created with c() and the numbers between the quotation marks are strings, because that's what the strptim() function requires.

That's followed by the argument **format** that defines how the character string should be read. In this instance, the year is denoted first (%Y), then the month (%M) and second (%S). You don't have to specify all of them, as long as the format corresponds to the character string.

In this course, we'll be using a less messy way to deal with dates using the package lubridate.

```
# If you don't currently have the lubridate package installed, uncomment the line below and run it
# install.packages("lubridate")

library(lubridate)

##
## Attaching package: 'lubridate'

## The following object is masked from 'package:base':
##
## date
```

```
date1 <- ymd_hms(c("20100225230000", "20100226000000", "20100226010000"))
```

The function ymd_hms() converted the year, month, date and hour, minutes, and seconds in the string. We'll go over this in more detail in Chapter 3.

Factors

Complicated.

- Data structure specifying categorical (nominal) or ordered categorical (ordinal) variables
- Tells R how to handle that variable in analyses
- Very important and misunderstood
- Any variable that is categorical or ordinal should usually be stored as a factor.

For example, Race might be input as "White", "Black", and "Hispanic"

When importing that data in from a spreadsheet, R will most often interpret it as a **Factor**.

Run these lines of code to create a new object, a dataframe called **sample_df**

⟨□ □ □ □ □ □ □ Filter					
_	id [‡]	name [‡]	age [‡]	race [‡]	
1	1001	Steve	26	White	
2	1002	Pam	65	Black	
3	1003	Jim	15	White	
4	1004	Dwight	7	Hispanic	

Let's take a look at the structure behind a dataframe we've created.

```
str(sample_df)
```

```
## 'data.frame': 4 obs. of 4 variables:
## $ id : Factor w/ 4 levels "1001","1002",..: 1 2 3 4
## $ name: chr "Steve" "Pam" "Jim" "Dwight"
## $ age : num 26 65 15 7
## $ race: Factor w/ 3 levels "Black","Hispanic",..: 3 1 3 2
```

R sees that the **Race** column is a factor variable with three levels.

```
levels(sample_df$race)
```

```
## [1] "Black" "Hispanic" "White"
```

This means that R groups statistics by these levels.

```
summary(sample_df$race)
```

```
## Black Hispanic White
## 1 1 2
```

Internally, R stores the integer values 1, 2, and 3, and maps the character strings in alphabetical order to these values. 1=Black, 2=Hispanic, and 3=White.

Why is this important to know?

Journalists are less concerned by factors and will often find themselves converting factors to strings and characters. But when you reach the point that you are wanting to create models and linear regressions then you'll be happy that it's an option.

Tip: Most odd quirks when it comes to R can be traced back to the fact that R was created by and for statisticians. R has grown a lot since then and the community has helped make it evolve to handle data the way we are more used to. But some habits die hard and are ingrained.

Integers & Numbers

These are what they sound like, whole or decimal numbers that can be used in calculations

R doesn't handle units properly if they are combined with numbers.

For example, "4in" will be treated as a string and not as 4.

Converting between the different types

Here's a warning.

[1] 1 2 3 4

• You can convert factors into strings.

```
sample_df$race
## [1] White
                Black
                          White
                                    Hispanic
## Levels: Black Hispanic White
as.character(sample_df$race)
                                          "Hispanic"
## [1] "White"
                   "Black"
                               "White"
  • You can convert strings into factors
sample_df$name
## [1] "Steve" "Pam"
                          "Jim"
                                    "Dwight"
factor(sample_df$name)
## [1] Steve Pam
                      Jim
                             Dwight
## Levels: Dwight Jim Pam Steve
  • You cannot convert factors into numbers.
sample_df$id
## [1] 1001 1002 1003 1004
## Levels: 1001 1002 1003 1004
as.numeric(sample_df$id)
```

Because R stores Factors as Integer values.

You must convert factors into characters first before converting it to numbers.

So you can nest it.

```
sample_df$id

## [1] 1001 1002 1003 1004

## Levels: 1001 1002 1003 1004

as.numeric(as.character(sample_df$id))

## [1] 1001 1002 1003 1004
```

Tip: It's okay if you don't fully understand this next section. It's pretty advanced. I give you permission to skip down to **Your turn**:.

Revisiting Functions

You can also save code you've written that simplifies your process into a function.

```
percent_change <- function(first_number, second_number) {
  pc <- (second_number-first_number)/first_number*100
  return(pc)
}

percent_change(100,150)</pre>
```

```
## [1] 50
```

This is what's happening in the code above:

- percent change is the name of the function, and assigned to it is the function function()
- Two variables are necessary to be passed to this function, first_number and second_number
- A new object pc is created using some math calculating percent change from the two variables passed
- the function return() assigns the result of the math to percent_change from the first line

Build enough functions and you can save them as your own package.

The important thing about functions is that they're programmed by humans.

I constructed this function because that's how I know that I'll only include two inputs and that each one will be numeric and that they'll be in order of first then second.

If you're working in R and a function you're using is giving an error, it most likely means there's something wrong with one or more of the variables you're passing to the function.

Here's what happens when I pass a string to my percent_change() function.

```
percent_change("what", "happens")
```

Error in second_number - first_number: non-numeric argument to binary operator

This is the point:

Sometimes really great R programmers will anticipate errors and catch bad inputs and try to output helpful suggestions instead of a generic error.

This particular error isn't very explicit. It needs to be interpreted but **you** know that the function needs numbers to work correctly because you created it.

New users might not know that intuitively.

And that's how you're going to feel when functions you run don't work.

You'll have to Google the error or peek into code to see if you can see what it expects and how it might break down thanks to what you've passed it.

And it won't be entirely your fault.

When we're coding and sharing it for others we can't anticipate all the ways in which others might want to use it in the future.

Shoot the function writer a message or if you wrote the package, welcome feedback from others.

This is what makes participating in the R community so great. We just want to do better.

Your turn

Challenge yourself with these exercises so you'll retain the knowledge of this section.

Instructions on how to run the exercise app are in the intro page to this section.