# A Short Introduction to Working With Data in R EXTRAS

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#### Section 1

Reading a csv file with base R

## Load a csv file using read.csv()

#### ?read.csv

• If read.csv() encounters problems reading a file, it is more likely to trigger an error than read\_csv(), which gives a warning more often.

```
DF_path <- file.path("..", "data", "data_example.csv")
try( read.csv(DF_path) )</pre>
```

```
# Error in read.table(file = file, header = header, sep = sep, quote
# more columns than column names
```

#### Check the file contents

 Let's take a peek at the first few lines and see if we can identify the problem (this is more often necessary with read.csv()):

```
readLines(DF_path, n = 4)
```

- # [1] "Data from an experiment on the cold tolerance of the grass s
- # [2] "Modified from `data(CO2)`. See `?CO2`."
- # [3] "Type, Treatment, Plant Num, 95, 175, 250, 350, 500, 675, 1000"
- # [4] "Quebec, nonchilled, 1, 16, 30.4, 34.8, 37.2, 35.3, 39.2, 39.7"
  - The first 2 lines don't look like comma-separated values!
  - They look like extra information that is not part of the data table structure.

#### Load a csv file into R

We can tell R to skip the lines with no data:

```
DF <- read.csv(DF_path, skip = 2)
DF_readr <- readr::read_csv(DF_path, skip = 2)</pre>
```

• Just because there were no Errors from R, doesn't mean there's nothing wrong with the data!

## Loading data: readr vs base R functions

readr		base R	
read_csv()	comma separated values	read.csv()	
read_csv2()	<pre>';' as delimiter (allows ',' for decimals)</pre>	read.csv2()	<ul><li>',' for decimals,</li><li>';' as separator</li></ul>
<pre>read_tsv()</pre>	tab separated values	read.delim()	delimited files (tab is default)
<pre>read_delim()</pre>	(generic) files with any delimiter	<pre>read.table()</pre>	,
<pre>read_fwf()</pre>	fixed width files	<pre>read.fwf()</pre>	

readr descriptions based on #dsbox

## Comparison of read.csv() and read csv()

- In keeping with Tidyverse conventions, functions are names with words separated by "\_"
  - ▶ instead of "." or camelCase, as in many base R functions
- The column names are different.
  - ▶ read.csv() automatically applies make.names() to the column names to make 'syntactically valid' names to use in R.
  - convenient, but not always what we want.
  - there are other 'cleaning' functions available (e.g., clean\_names() in the janitor package)
- read\_csv() automatically replaced empty strings in the Treatment column with NAs.
- read\_csv() left the '675' column as numeric, but ignored the commas, resulting in larger numbers.
- read\_csv() produces a "tbl\_df" (tibble) object, not a simple data.frame

## Tibble examples

- Tibbles have an enhanced print() method
- and they will not do partial matching on variable names, triggering a warning instead for columns that do not exist.

```
print(DF_readr, n=2)
\# \# A + \text{tibble} \cdot 13 \times 10
   Type Treatment PlantNum '95' `175' `250' `350' `500'
    <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <</pre>
# 1 Quebec nonchilled
                              1 16 30.4 34.8 37.2 35.3
                              2 13.6 27.3 37.1 41.8 40.6
# 2 Quebec <NA>
# # i 11 more rows
# # i 2 more variables: `675` <dbl>, `1000` <dbl>
is.null(DF$Treat)
# [1] FALSE
is.null(DF readr$Treat)
 Warning: Unknown or uninitialised column: `Treat`.
```

[1] TRUE

## Writing data: readr vs base R functions

readr		base R
write_csv()	← comma separated values	write.csv()
<pre>write_csv2()</pre>	$\leftarrow$ allows ';' as delimiter	write.csv2()
	and ',' for decimals	',' for decimals,
	(depending on locale)	';' as separator
<pre>write_tsv()</pre>	$\leftarrow$ tab separated values	
<pre>write_delim()</pre>	$\leftarrow$ (generic) files with an	<pre>write.table()</pre>
	arbitrary delimiter	
<pre>write_excel_csv(),</pre>	$\leftarrow$ include a UTF-8 Byte	
<pre>write_excel_csv2()</pre>	order mark, which indicates	
	to Excel the csv is UTF-8	
	encoded	

#### Section 2

File Encoding, Windows, & Microsoft Excel<sup>TM</sup>

## Encoding non-English characters

 If you are running R in Windows, you may notice that some text values (character) look strange when read with read.csv():

"Québec" instead of "Québec"

- There is nothing wrong with the file this indicates a mismatch between the encoding used to write the file, and what R used to read it.
- Even though '.csv' files are plain text, letters (especially non-english characters) can be encoded in different ways to represent them in the computer.
- "UTF-8" is a character encoding standard designed to handle many non-english characters.
  - ▶ The example data file was written in "UTF-8"
  - ▶ Most OSes and many programs use "UTF-8" encoding by default.
  - ▶ But Windows uses "latin1" by default, and so does R (< 4.2.0) when running in Windows.
  - ► Starting with v4.2.0, R uses "UTF-8" as the default encoding on Windows

## Read a csv file with a different encoding

 You can specify the encoding used in the file with the 'encoding' argument of read.csv()

```
DF <- read.csv(DF_path, skip = 2, encoding = "UTF-8")
```

 If reading a file that was created on a Windows computer and encoded in "latin1", on a different system (mac, Unix, linux, etc.) — or a recent version of R (>=4.2) on Windows — you can specify that, too:

```
read.csv(DF_path, skip = 2, encoding = "latin1")
```

## Encoding & Microsoft Excel<sup>TM</sup>

- Excel can save a .csv file using UTF-8 encoding, but in doing so, it adds "byte order mark" ("BOM") to the file.
  - ▶ This is a special character that Excel also uses to recognize that the file is encoded using UTF-8.
  - ▶ Thus a BOM can make the file "easier" to use with Excel, by allowing it to automatically recognize the UTF-8 encoding, but it can also cause problems for other programs (like R) that do not expect such a non-Unicode character.
- Without the BOM, Excel will assume the file is encoded in "latin1" if you double-click on the csv file to open it in Excel, even if it was actually encoded with UTF-8.
  - ▶ This can cause special characters to appear incorrectly.
  - ▶ You can still import a .csv file encoded in UTF-8 into Excel correctly, but it requires opening the file within Excel, or importing it using commands in the "Data" ribbon / menu

## Read a file with a BOM using read.csv() in R

 Reading a .csv file with a BOM using the usual method may cause the BOM to be included in the name of the first column (on Windows).

 The solution with read.csv() is to use the argument 'fileEncoding = "UTF-8-BOM"' (instead of the 'encoding' argument)

```
# Type Treatment PlantNum X95
# 4 Québec chilled 1 14.2
```

## Read a file with a BOM using read csv()

 The readr package uses "UTF-8" encoding by default, and automatically ignores a BOM, if present.

```
bom_readr <- readr::read_csv("../data/data_example_bom.csv")
bom_readr[4, 1:4] |> knitr::kable()
```

Туре	Treatment	PlantNum	95
Québec	chilled	1	14.2

- write\_csv() (in the readr package) automatically encodes output files using "UTF-8", for greater portability across systems.
  - except for older versions of base R (read.csv()) on Windows :(

Hopefully, these examples have demonstrated that the readr package makes it easy to work with "UTF-8" files by default, on any platform.

## Add a BOM to an output file

- It is possible to add a BOM to a csv file, but it must be done manually with base R:
  - code adapted from this StackOverflow answer

```
writeChar(
  iconv("\ufeff", to = "UTF-8"),
  "output.csv",
  eos = NULL
)
write.csv(Data, "output.csv", append = TRUE, ...)
```

 The readr package can do this directly with a special write\_excel\_csv function:

```
write_excel_csv(Data, "output.csv", ... )
```

!

R does not recommend doing this (see ?file), so use with caution.

## Using other encodings with readr

 You can control the encoding used by readr functions with the locale argument.

• See ?readr::read csv and ?readr::locale for details.

#### Section 3

## Downloading Data From the Internet

## Read data directly from a URL

You can use read.csv() & read\_csv() with URLs as file paths:

```
read_csv( paste0(
    "https://raw.githubusercontent.com/jawhiteley/",
    "R_training_jaw/main/R2_data_scripts/",
    "data/data example.csv"
    ),
    skip = 2
# Rows: 13 Columns: 10
# -- Column specification -----
# Delimiter: ","
# chr (3): Type, Treatment, 500
# dbl (6): PlantNum, 95, 175, 250, 350, 1000
# num (1): 675
#
# i Use `spec()` to retrieve the full column specification for this
```

csv web <-

## Packages: Canadian sources

Some packages provide convenient ways to download data from online sources.

- An ecosystem of R packages to access and process Canadian data
  - cansim: download public data from Statistics Canada (CODR).
  - cancensus: "access, retrieve, and work with Canadian Census data and geography".

requires a valid CensusMapper API key

- cmhc: API wrapper for extracting CMHC data out of the CMHC Housing Market Information Portal.
- ▶ tongfen: prepare "data on disparate geographies for comparison by converting them to a common geography" ("slightly coarser least common denominator geography").
- ▶ See also: Analyzing Canadian Demographic and Housing Data

#### Section 4

# Reading Data in Other Formats

## Other tidyverse packages

- The Data Import Chapter of R for Data Science (2e) describes these tidyverse packages for other types of data:
  - haven reads SPSS, Stata, and SAS files.
  - ▶ DBI, along with a database specific backend (e.g. RMySQL, RSQLite, RPostgreSQL, etc.) allows you to run SQL queries on a database and return a data frame.

## Other options

- The foreign package can read data in a variety of formats, including: 'Minitab', 'S', 'SAS', 'SPSS', 'Stata', and others
  - ▶ May require access to external software to read their formats
- Parquet files: an efficient columnar format, popular with Big Data and cloud computing
  - ► Apache Arrow (i.e., the 'arrow' package)
- See the Import Section of R for Data Science (2nd edition) For more details on getting data into R from these and other sources.
- Other options are also described in the R Data Import/Export manual.

## Section 5

# Exporting to other formats

## Writing to Microsoft Excel<sup>TM</sup> files

Packages that can write to Excel files:

- xlsx: read, write, format Excel 2007 (.xlsx) and Excel 97/2000/XP/2003 (.xls) files.
  - Requires Java and the rJava package
- XLConnect: comprehensive and cross-platform R package for manipulating Microsoft Excel files (.xlsx & .xls) from within R.
  - ▶ Requires a Java Runtime Environment (JRE)
- openxlsx: simplified creation of Excel .xlsx files (not .xls).
  - ► No dependency on Java
- writexl: portable, light-weight data frame to xlsx exporter.
  - ▶ No Java or Excel required

I recommend *avoiding* exporting data to Excel files if possible. csv files are easier to read to & write from, and can be read by a wider variety of software (they are more portable).

Automated reports can be produced with R Markdown and output to a variety of more portable formats (pdf, HTML, etc.) instead.

#### Section 6

# Programming

#### Control flow: if conditions

#### ?Control

 Control what code in a script actually runs with conditional expressions and if statements

```
if (condition) {
  message("The condition is TRUE")
}
```

- **IF** the *expression* (condition) in the parentheses () evaluates to TRUE, **then** the code inside the braces {} will run.
  - Otherwise, it will not.

#### Control flow: if & NA

- The conditional expression must result in either TRUE or FALSE; anything else causes an error.
  - ▶ NA is a common source of problems here.
  - ▶ isTRUE() and isFALSE() are useful to ensure a TRUE/FALSE result.

```
if (NA) {
  message("The condition is NA")
}
# Error in if (NA) { : missing value where TRUE/FALSE needed
if (isTRUE(NA)) {
  message("The condition is NA")
}
if (is.na(NA)) {
  message("The condition is TRUE")
}
```

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# The condition is TRUE

## Control flow: if, else conditions

An else condition is optional:
 IF the expression (condition) in the if statement evaluates to FALSE,
 then the code inside the braces {} after else will run.

```
if (isTRUE(condition)) {
  message("The condition is TRUE")
} else {
  warning("The condition is FALSE")
if (isFALSE(0 == 1)) {
  message("The condition is TRUE")
} else {
  warning("The condition is FALSE")
```

# The condition is TRUE

## Control flow: multiple if, else conditions

You can have multiple "if-else" statements, as long as each if
after the first follows immediately after the else keyword.

```
if (isTRUE(NA)) {
   stop("NA is TRUE")
} else if (isFALSE(NA)) {
   warning("NA is FALSE")
} else {
   message("NA is neither TRUE nor FALSE")
}
```

# NA is neither TRUE nor FALSE

## Control flow: for loops

#### ?Control

• Tell R to run a block of code multiple times in a row with a loop

```
for (i in 1:3) {
  print( paste("i =", i) )
}
# [1] "i = 1"
# [1] "i = 2"
# [1] "i = 3"
```

#### for loops

- Put inside the parentheses (), in order:
  - ► The name of a variable to update each time through the loop (iteration)
  - ▶ The keyword 'in'
  - A vector of values
- The code inside the the braces {}
   is run with the specified variable
   assigned each value in the vector,
   in sequence.

#### User-defined function

- Define your own functions with the function function!
  - function code goes between braces: {}
  - specify what value is returned with the return() function

```
\frac{1}{6} = \frac{1}{6} \cdot \frac{1}
                            same \leftarrow (v1 == v2) | (is.na(v1) & is.na(v2))
                            same[is.na(same)] <- FALSE</pre>
                            return(same)
                                                                                                                                                                                                                                                                                                                                                                                                # return the result
}
   # test it:
 c(1, NA, 3, 4, NaN) = (1, NA, 1, NA, NaN)
 # [1]
                                                                       TRUE TRUE FALSE FALSE TRUE
c(1, NA, 3, 4, NaN) == c(1, NA, 1, NA, NaN)
                         Г17
                                                                                      TRUE
                                                                                                                                                                                         NA FALSE
                                                                                                                                                                                                                                                                                                                                              NA
                                                                                                                                                                                                                                                                                                                                                                                                                           NA
```

• This code defines a function that compares two vectors, accounting for missing values (NA) (code adapted from Cookbook for R)

## An infix operator

```
c(1, NA, 3, 4 , NaN) %==% c(1, NA, 1, NA, NaN)
```

- # [1] TRUE TRUE FALSE FALSE TRUE
  - This function is also a special type called an "infix operator", which
    goes between two objects (it's arguments) like an operator, instead of
    a 'typical' function call
    - it has exactly 2 arguments (lhs, rhs)
    - ▶ the name begins and ends with a percent symbol (%)

## Debugging

These are some useful functions for troubleshooting or *debugging* code that is not doing what you want:

- ?traceback: prints a list ("stack") showing the order of expressions that triggered the last error
  - ▶ useful in cases where functions call other functions iteratively, and you want to know which function in the "stack" caused the error.
- ?debug: flag a function for debugging
  - ▶ The next time the function is called, it will open an interactive 'browser' session that lets you run the code in the function 1 line at a time, and even explore how objects change between lines.
- ?browser: open the same interactive session as debug(), but at a specific location in the code (anywhere)

For more information, check out these (and other) sources: Advanced R, R Programming for Data Science, Debugging with the RStudio IDE

#### Section 7

# Some Advanced dplyr Examples

## Find values systematically

Find columns that are character, but we expect to be numeric:

```
cols charn <- DF %>%
  select(starts_with("X") & where(is.character)) %>%
  names()
```

- Loop through the identified column names (using for) and print values that would be NA if converted to numeric
  - suppressWarnings() suppresses the warnings we expect from as.numeric() in this case.

```
DF %>% select(1:3, all_of(col)) %>%
    filter(get(col) %>% as.numeric() %>% is.na() %>%
               suppressWarnings() ) %>%
    print()
     Type Treatment PlantNum
                                            X500
 1 Québec
            chilled
                           1 32.5 (umol/m<sup>2</sup> sec)
     Type Treatment PlantNum X675
 1 Québec
           chilled
                           1 35.4
 2 Québec
                           2 37,5
                           3 39.6
# 3 Québec
```

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for (col in cols charn) {

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## summarize() multiple columns with across()

- summarize() uses mutate semantics, but across() applies a function to multiple columns, specified using select semantics
  - i.e., across() lets you use *select semantics* in a context where you would normally use *mutate semantics*
- A function can be specified by name

```
DF %>%
 summarize( across(where(is.numeric), max) )
#
   PlantNum X95 X175 X250 X350 X1000
# 1
        3 16.2 32.4
                    NA 42.1
                             NA
DF %>% group_by(Treatment) %>%
 summarize( across(starts with("X"), max) )
# # A tibble: 3 x 8
#
   Treatment
             X95
                   X175
                        X250
                             X350 X500
                                          X675
                                               X1000
   <chr> <dbl>
              16.2 32.4 NA 42.1 42.9
                                       43.9
                                                NA
 2 "chilled" 14.2 24.1 30.3 34.6 32.5 (um~ 35,4 38.7
   "nonchilled"
              16
                   30.4 34.8 37.2 35.3
                                          39.2
                                                39.7
```

#### summarize() across() with ad hoc function

 For more complex operations, you may have to define a custom function or define an ad hoc function to include additional arguments

```
# PlantNum X95 X175 X250 X350 X1000
# 1 3 16.2 32.4 40.3 42.1 45.5
```

#### summarize() across() with "lambda notation"

- You can also define ad hoc functions using a special "lambda" notation
  - refer to the value in the column with '.x'

```
DF %>%
   summarize( across(everything(), ~ sum(is.na(.x))) )
# Type Treatment PlantNum X95 X175 X250 X350 X500 X675
```

```
# 1 0 0 0 0 0 1 0 0 # X1000
```

# 1 1

## Tip

sum(is.na()) is a great way to count the number of missing values in a column.