Data wrangling

Importing and data cleaning

Jordan Creed
Moffitt Cancer Center

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Elements of "tidy" data

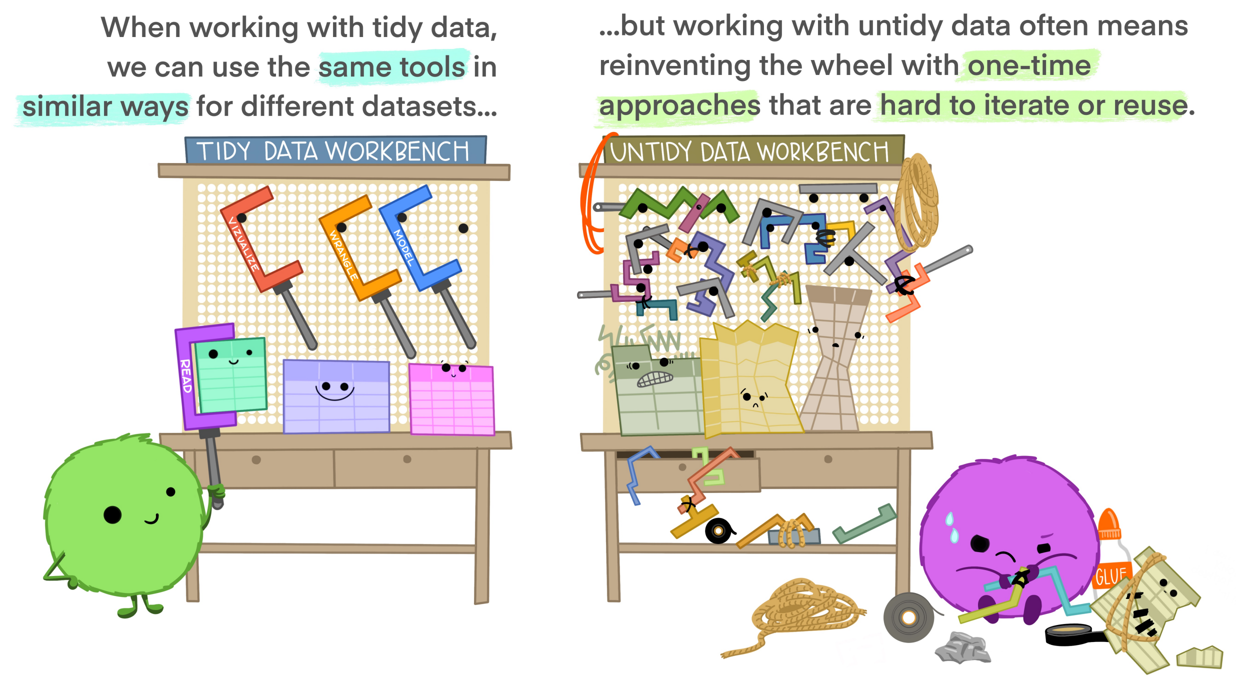
Data comes in all shapes and sizes, but not all data organization is made equal!

3 elements of tidy data:

1. Each variable has its own column
2. Each observation has its own row
3. Each value has it own cell

👉 For more information please see chapter 11 of R for Data Science

Elements of "tidy" data


When working with tidy data, we can use the same tools in similar ways for different datasets...
...but working with untidy data often means reinventing the wheel with one-time approaches that are hard to iterate or reuse.
TIDY DATA WORKBENCH
VIZUALIZE
WRANGLE
MODEL
READ
UNTIDY DATA WORKBENCH
GLUE

TCGA Data

The Cancer Genome Atlas (TCGA)

TCGA is the large cancer genomics program spanning 33 cancer types and multiple institutions.

- HNSC - Head and neck squamous cell carcinoma
- KIRC - Kidney renal clear cell carcinoma

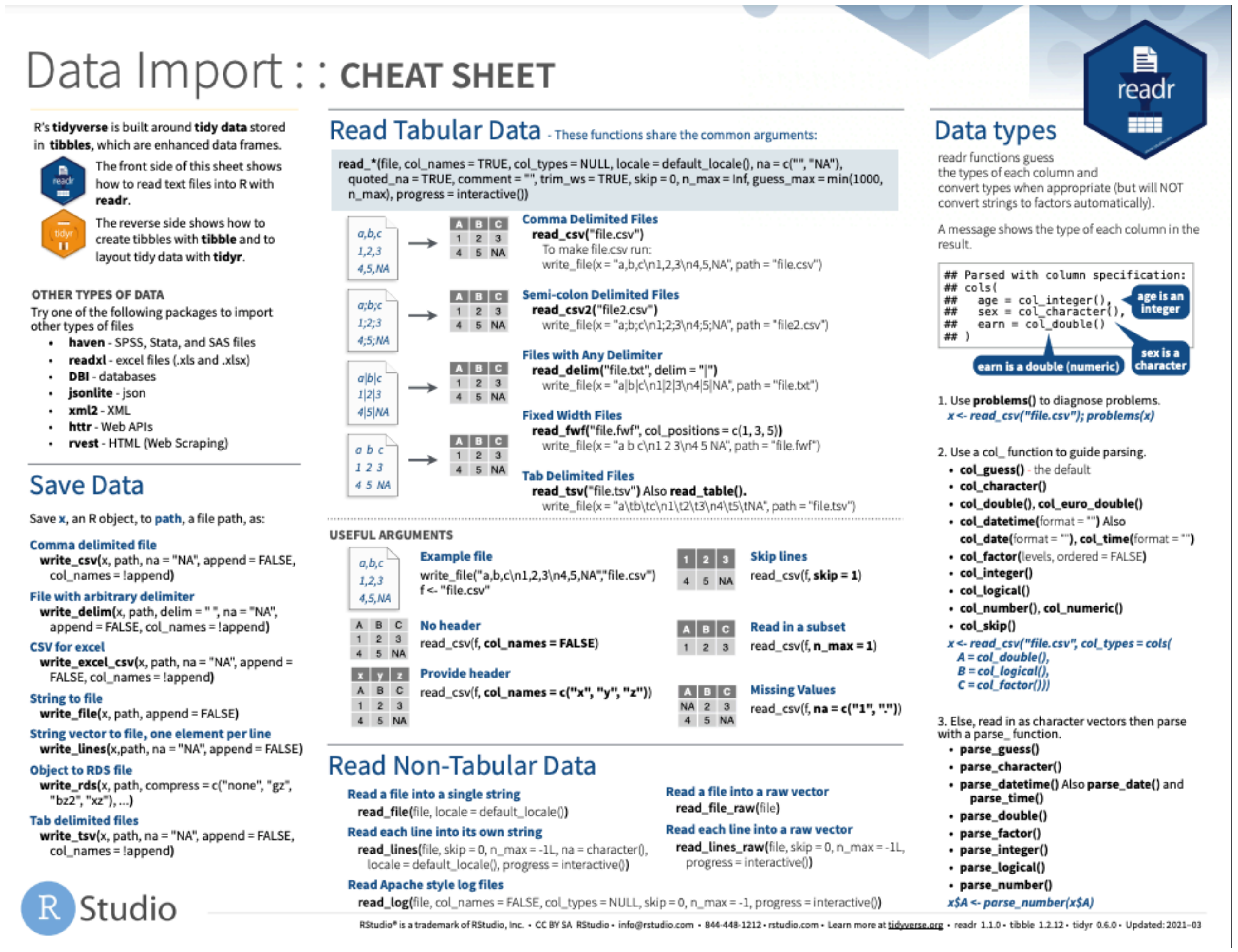
Data is saved as two `.csv`s : `data/tcga-clinical.csv` and `data/tcga-gene-exp.csv`

Importing files

Import method depends on file type!

- Most text files : `readr::read_delim()` or `readr::read_csv()`
- Microsoft excel files : `readxl::read_excel()`
- R data files : `load()`
- SAS data files: `haven::read_sas()`

👩‍💻 remember to copy code from the import gui to your scripts!



Data Import : : CHEAT SHEET

R's **tidyverse** is built around **tidy data** stored in **tibbles**, which are enhanced data frames.

The front side of this sheet shows how to read text files into R with **readr**.

The reverse side shows how to create tibbles with **tibble** and to layout tidy data with **tidyr**.

OTHER TYPES OF DATA

Try one of the following packages to import other types of files

- **haven** - SPSS, Stata, and SAS files
- **readxl** - excel files (.xls and .xlsx)
- **DBI** - databases
- **jsonlite** - json
- **xml2** - XML
- **httr** - Web APIs
- **rvest** - HTML (Web Scraping)

Save Data

Save **x**, an R object, to **path**, a file path, as:

Comma delimited file
write_csv(x, path, na = "NA", append = FALSE, col_names = !append)

File with arbitrary delimiter
write_delim(x, path, delim = " ", na = "NA", append = FALSE, col_names = !append)

CSV for excel
write_excel_csv(x, path, na = "NA", append = FALSE, col_names = !append)

String to file
write_file(x, path, append = FALSE)

String vector to file, one element per line
write_lines(x,path, na = "NA", append = FALSE)

Object to RDS file
write_rds(x, path, compress = c("none", "gz", "bz2", "xz"), ...)

Tab delimited files
write_tsv(x, path, na = "NA", append = FALSE, col_names = !append)

RStudio

Read Tabular Data - These functions share the common arguments:

```
read_*(file, col_names = TRUE, col_types = NULL, locale = default_locale(), na = c("", "NA"),
  quoted_na = TRUE, comment = "", trim_ws = TRUE, skip = 0, n_max = Inf, guess_max = min(1000,
  n_max), progress = interactive())
```

Comma Delimited Files
read_csv("file.csv")
To make file.csv run:
write_file(x = "a,b,c\n1,2,3\n4,5,NA", path = "file.csv")

Semi-colon Delimited Files
read_csv2("file2.csv")
write_file(x = "a;b;c\n1;2;3\n4;5;NA", path = "file2.csv")

Files with Any Delimiter
read_delim("file.txt", delim = "|")
write_file(x = "a|b|c\n1|2|3\n4|5|NA", path = "file.txt")

Fixed Width Files
read_fwf("file.fwf", col_positions = c(1, 3, 5))
write_file(x = "a b c\n1 2 3\n4 5 NA", path = "file.fwf")

Tab Delimited Files
read_tsv("file.tsv") Also **read_table()**.
write_file(x = "a\tb\tc\n1\t2\t3\n4\t5\tNA", path = "file.tsv")

USEFUL ARGUMENTS

Example file
write_file("a,b,c\n1,2,3\n4,5,NA","file.csv")
f <- "file.csv"

Skip lines
read_csv(f, **skip = 1**)

No header
read_csv(f, **col_names = FALSE**)

Read in a subset
read_csv(f, **n_max = 1**)

Provide header
read_csv(f, **col_names = c("x", "y", "z")**)

Missing Values
read_csv(f, **na = c("1", ".")**)

Read Non-Tabular Data

Read a file into a single string
read_file(file, locale = default_locale())

Read each line into its own string
read_lines(file, skip = 0, n_max = -1L, na = character(), locale = default_locale(), progress = interactive())

Read Apache style log files
read_log(file, col_names = FALSE, col_types = NULL, skip = 0, n_max = -1, progress = interactive())

Read a file into a raw vector
read_file_raw(file)

Read each line into a raw vector
read_lines_raw(file, skip = 0, n_max = -1L, progress = interactive())

Data types

readr functions guess the types of each column and convert types when appropriate (but will NOT convert strings to factors automatically).

A message shows the type of each column in the result.

```
## Parsed with column specification:
## cols(
##   age = col_integer(),
##   sex = col_character(),
##   earn = col_double()
## )
```

1. Use **problems()** to diagnose problems.
   x <- read_csv("file.csv"); problems(x)

2. Use a col_ function to guide parsing.
   - **col_guess()** - the default
   - **col_character()**
   - **col_double()**, **col_euro_double()**
   - **col_datetime**(format = "") Also **col_date**(format = ""), **col_time**(format = "")
   - **col_factor**(levels, ordered = FALSE)
   - **col_integer()**
   - **col_logical()**
   - **col_number()**, **col_numeric()**
   - **col_skip()**

   *x <- read_csv("file.csv", col_types = cols(
     A = col_double(),
     B = col_logical(),
     C = col_factor()))*

3. Else, read in as character vectors then parse with a parse_ function.
   - **parse_guess()**
   - **parse_character()**
   - **parse_datetime()** Also **parse_date()** and **parse_time()**
   - **parse_double()**
   - **parse_factor()**
   - **parse_integer()**
   - **parse_logical()**
   - **parse_number()**

   x$A <- parse_number(x$A)

Data cleaning with dplyr

`dplyr` verbs and syntax

`select()` subset columns

- `select(tibble_name, variable1, variable2)`

`mutate()` create new variables/columns

- `mutate(tibble_name, new_variable = variable1*2)`

common `select()` options

`:` selects a range of columns

`-` selects every column but those specified

`starts_with()` and `ends_with()` selects columns whose names start/end with the specified string

`contains()` selects columns whose names contain the specified string

`dplyr` verbs and syntax

`filter()` subset rows

- `filter(tibble_name, variable1 == "specific value")`

`summarise()` aggregates rows

- `summarise(tibble_name, avg_value = mean(variable1))`

`arrange()` orders rows

- `arrange(tibble_name, variable1)`

Operations can be chained together with the pipe operator (`%>%`)

logical tests for `filter()`

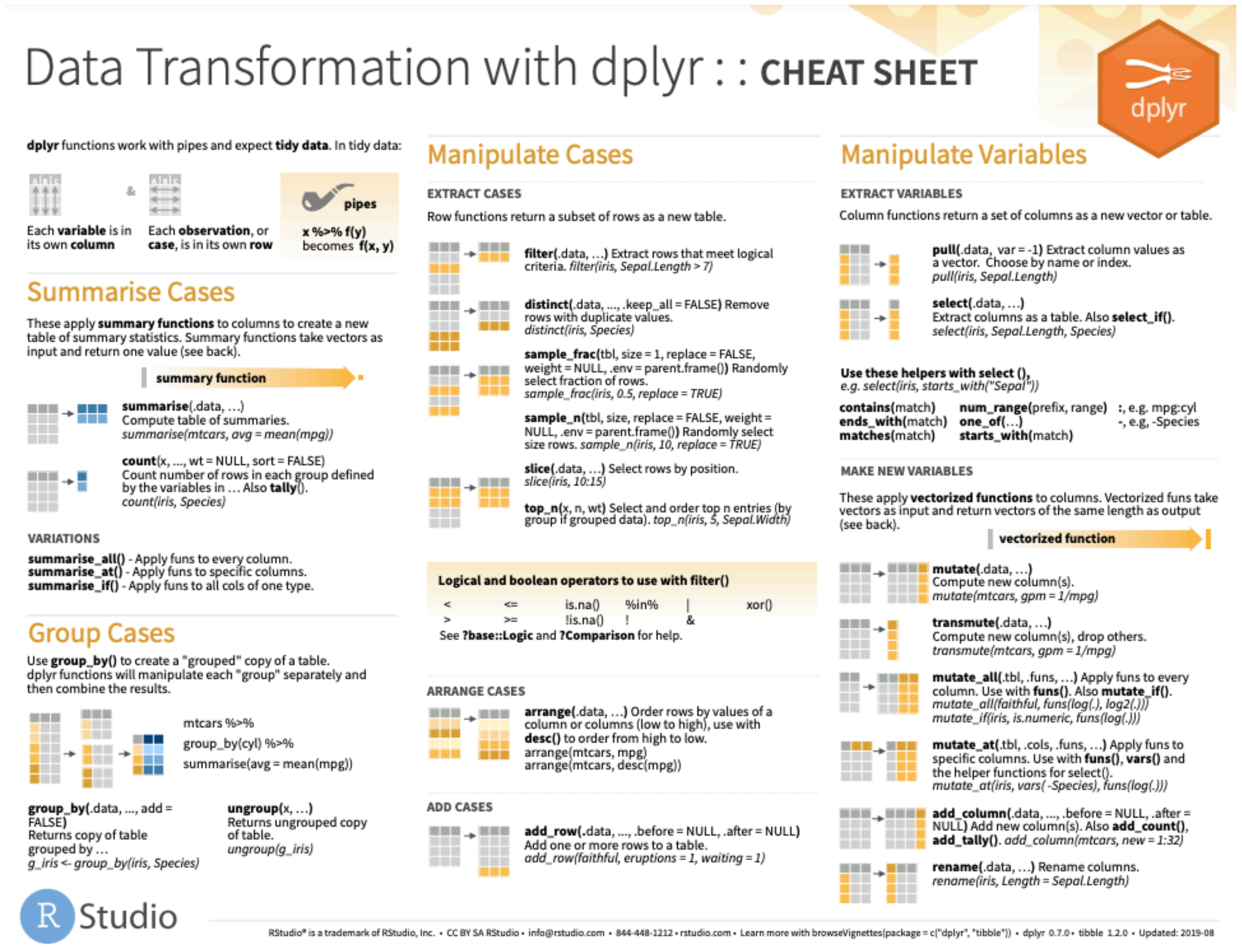
- `==` : equal
- `!=` : not equal
- `<` & `<=` : less than & less than or equal to
- `>` & `>=` : greater than & greater than or equal to
- `|` : or
- `&` : and
- `!` : not
- `%in%` : in the set

`dplyr` helpers

`n()` the number of rows

`n_distinct()` the number of unique values for a variable

`group_by()` collects observations by a common value



Data Transformation with dplyr : : CHEAT SHEET

dplyr functions work with pipes and expect **tidy data**. In tidy data:

Each **variable** is in its own **column** & Each **observation**, or **case**, is in its own **row**

pipes

x %>% f(y) becomes **f(x, y)**

Summarise Cases

These apply **summary functions** to columns to create a new table of summary statistics. Summary functions take vectors as input and return one value (see back).

summary function

summarise(.data, …)
Compute table of summaries.
summarise(mtcars, avg = mean(mpg))

count(x, ..., wt = NULL, sort = FALSE)
Count number of rows in each group defined by the variables in … Also **tally**().
count(iris, Species)

VARIATIONS

summarise_all() - Apply funs to every column.
summarise_at() - Apply funs to specific columns.
summarise_if() - Apply funs to all cols of one type.

Group Cases

Use **group_by()** to create a "grouped" copy of a table. dplyr functions will manipulate each "group" separately and then combine the results.

mtcars %>%
group_by(cyl) %>%
summarise(avg = mean(mpg))

group_by(.data, ..., add = FALSE**)**
Returns copy of table grouped by …
g_iris <- group_by(iris, Species)

ungroup(x, …**)**
Returns ungrouped copy of table.
ungroup(g_iris)

Manipulate Cases

EXTRACT CASES

Row functions return a subset of rows as a new table.

filter(.data, …) Extract rows that meet logical criteria. *filter(iris, Sepal.Length > 7)*

distinct(.data, ..., .keep_all = FALSE) Remove rows with duplicate values.
distinct(iris, Species)

sample_frac(tbl, size = 1, replace = FALSE, weight = NULL, .env = parent.frame()) Randomly select fraction of rows.
sample_frac(iris, 0.5, replace = TRUE)

sample_n(tbl, size, replace = FALSE, weight = NULL, .env = parent.frame()) Randomly select size rows. *sample_n(iris, 10, replace = TRUE)*

slice(.data, …) Select rows by position.
slice(iris, 10:15)

top_n(x, n, wt) Select and order top n entries (by group if grouped data). *top_n(iris, 5, Sepal.Width)*

Logical and boolean operators to use with filter()

<	<=	is.na()	%in%	|	xor()
>	>=	!is.na()	!	&	

See **?base::Logic** and **?Comparison** for help.

ARRANGE CASES

arrange(.data, …) Order rows by values of a column or columns (low to high), use with **desc()** to order from high to low.
arrange(mtcars, mpg)
arrange(mtcars, desc(mpg))

ADD CASES

add_row(.data, ..., .before = NULL, .after = NULL**)**
Add one or more rows to a table.
add_row(faithful, eruptions = 1, waiting = 1)

Manipulate Variables

EXTRACT VARIABLES

Column functions return a set of columns as a new vector or table.

pull(.data, var = -1) Extract column values as a vector. Choose by name or index.
pull(iris, Sepal.Length)

select(.data, …)
Extract columns as a table. Also **select_if()**.
select(iris, Sepal.Length, Species)

Use these helpers with select (),
e.g. select(iris, starts_with("Sepal"))

contains(match)	**num_range**(prefix, range)	:, e.g. mpg:cyl
ends_with(match)	**one_of**(…)	-, e.g, -Species
matches(match)	**starts_with**(match)	

MAKE NEW VARIABLES

These apply **vectorized functions** to columns. Vectorized funs take vectors as input and return vectors of the same length as output (see back).

vectorized function

mutate(.data, …)
Compute new column(s).
mutate(mtcars, gpm = 1/mpg)

transmute(.data, …)
Compute new column(s), drop others.
transmute(mtcars, gpm = 1/mpg)

mutate_all(.tbl, .funs, …**)** Apply funs to every column. Use with **funs()**. Also **mutate_if()**.
mutate_all(faithful, funs(log(.), log2(.)))
mutate_if(iris, is.numeric, funs(log(.)))

mutate_at(.tbl, .cols, .funs, …**)** Apply funs to specific columns. Use with **funs()**, **vars()** and the helper functions for select().
mutate_at(iris, vars(-Species), funs(log(.)))

add_column(.data, ..., .before = NULL, .after = NULL) Add new column(s). Also **add_count()**, **add_tally()**. *add_column(mtcars, new = 1:32)*

rename(.data, …) Rename columns.
rename(iris, Length = Sepal.Length)

RStudio

RStudio® is a trademark of RStudio, Inc. • CC BY SA RStudio • info@rstudio.com • 844-448-1212 • rstudio.com • Learn more with browseVignettes(package = c("dplyr", "tibble")) • dplyr 0.7.0 • tibble 1.2.0 • Updated: 2019-08