R for Data Science

Using the tidyverse

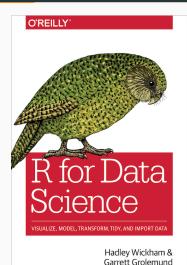
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Workshop material: bit.ly/comos-r

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References

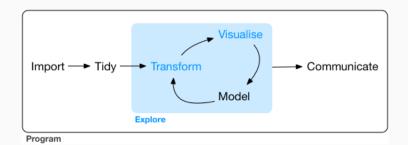


- Workshop material: bit.ly/comos-r
- RStudio Cheat sheets rstudio.com/resources/cheatsheets/
- Book R for Data Science (Garrett Grolemund, Hadley Wickham):
 - Online version: http://r4ds.had.co.nz/
 - Paperback: R for Data Science: Import, Tidy, Transform, Visualize, and Model Data, O'Reilly Media, 2017.

Workshop outline

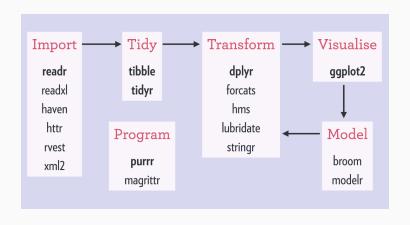
- Today:
 - Overview of Data Science workflow
 - tidyverse fundamentals
 - Data transformation with dplyr
 - Merging datasets with join operations
 - Data input strategies with readr
- Tomorrow:
 - String manipulation with stringr
 - Functional programming with purrr
 - Handling categorical variables with forcats
 - Data tidying with tidyr
- Friday:
 - Data visualization with ggplot2
 - Model estimation and selection with broom
 - Web scraping with rvest
 - ? More dataviz? (maps)

The Data Science Workflow



Credit R for Data Science

The Tidyverse Ecosystem



credit Joseph Rickert

Tidyverse philosophy

The tidyverse is organised around a few principles:

- Tidy data: each variable is a column, each observation is a row, and each type of observational unit is a table (see article by Hadley on JSS);
- Code readability: consistent, expressive syntax;
- Functional programming: functions applied to all elements of objects (no iterators); pass data along function pipeline %>%;
- Compatibility: classes and functions are mostly backward compatible with "Base R".

Tidyverse Philosophy

Key differences with "Base R"

- tibble replaces data.frame class for rectangular datasets:
 - better preview printing;
 - "lazy and surly": less type coercion (strings not converted to factors); doesn't change variable names;
 - doesn't use row.names
- style: function names are in snake_casee.g.: read_csv() instead of read.csv()
- data is first argument of functions.

Building the pipeline

Tidyverse uses the *pipe* %>% to concatenate operations on data.

%>% builds chains of function by passing ("piping") the **output** of one function (ie, data) as **input** of the next function.

 $RStudio \ shortcut: \ \boxed{\texttt{ctrl}} + \boxed{\^{\^{1}}} + \boxed{\texttt{M}} \ \text{or} \ \boxed{\texttt{\#}} + \boxed{\^{1}} + \boxed{\texttt{M}} \ \text{on} \ \text{macOS}.$

The pipe allows function composition: x %>% f() %>% g() \equiv g(f(x))

If a function h(x,y) has more than one argument (or data is not the first argument), . is an explicit placeholder: $y \%\% h(x, .) \equiv h(x, y)$.

Data manipulation with *dplyr*

In the tidyverse, *dplyr* provides a grammar of data manipulation, with consistent set of verbs that help you solve the most common data manipulation challenges:

- mutate() adds new variables that are functions of existing variables;
- select() picks variables based on their names;
- filter() picks cases based on their values;
- summarise() reduces multiple values down to a single summary;
- arrange() changes the ordering of the rows;

These all combine naturally with group_by() which allows to perform any operation by groups of values.

Data structure

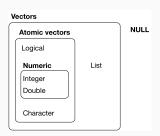
Vectors are the fundamental object class in R.

There are two types of vectors:

- Atomic vectors (homogeneous) of six types:
 - 1. double: most numbers
 - 2. integer: integer numbers
 - 3. logical: TRUE, FALSE, NA
 - 4. character: strings
 - 5. complex
 - 6. raw

Integer and double vectors are also known as *numeric* vectors.

 Lists (heterogeneous) which are sometimes called recursive vectors because lists can contain other lists.



Functional programming with purrr

With the *purrr* package, the function map() to maps (ie, applies) a function to every element of a vector (atomic or list).

(loose) mathematical definition:

- Given a set (in R, a vector) $X = \{x_1, x_2, \dots, x_n\}$
- And a function $f(x): X \mapsto ?$
- $Map(X, f) : \{f(x_1), f(x_2), \dots, f(x_n)\}$