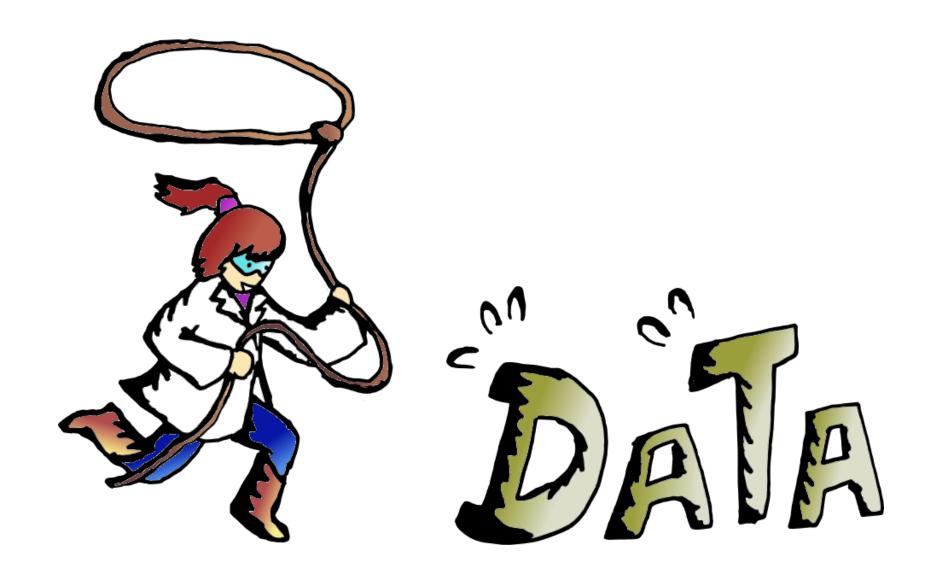


# **Data Wrangling**

Andrew Redd, PhD. R Bootcamp 2020



# Manipulations

- Data integrity
- Reshaping
- Filtering
- Merging
- Summarizing

### Packages that we will need

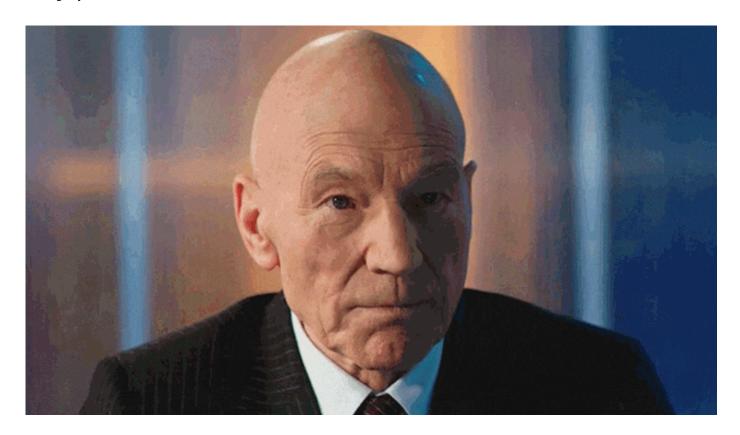
```
# Make tidyverse load quietly
options(tidyverse.quiet = TRUE)
library(tidyverse)  #< General use
library(tidyr)  #< Reshaping
library(wbstats)  #< World bank data.
library(countrycode) #< Country coding
library(assertthat)  #< Results checking
library(lubridate)  #< Date manipulations</pre>
```

### Loading data

```
for .RData files use load()
load("data/covid.data.wide.dfr.RData")
for .rds files use readRDS() and capture the results in a variable.
data <- readRDS("data/covid.data.wide.dfr.rds")</pre>
```

# Always check your data

Any problems with the data?



### 1st problem: Reshaping data

- · Wide Data
  - multiple observations for one unit are in columns
- Long Data
  - multiple observations for one unit are in rows.

### Wide <-> Long

Time for an aside on tidy selectors.

### Tidy selectors

The statement . %>% select(file, 1:10) %>% ... states to give back only the column named file then columns at indices 1 through 10.

Note file is not quoted. select and other tidy functions uses tidy selection, a form of lazy evaluation where the arguments are evaluated in the context of the data.

- · : sequence, i.e. all variables between given variables
- · ! compliment of what is provided.
- · - drop specified variable

### Tidy selector functions

- everything() everything not already specified.
- last\_col() last variable, or nth last variable
- String matches: starts\_with(), ends\_with(), contains()(exact match), and matches()(regular expression match)
- num\_range() example: X1, X2, X3, ... could be selected with num\_range('X')
- escaping the lazy evaluation
  - all\_of() and any\_of(), Example: select(my.data, any\_of(my.vars))
    would select any variables from my.data whose names were present in
    the variable my.vars.
  - where(), give a predicate to make the determination if variable should be kept or not. Ex. where(is.numeric) would select all numeric variables.



### Wide <-> Long

#### glimpse(covid.data.long.1)

### 2nd problem: fixing variables

#### Tasks:

- 1. Convert file names to better categories
- 2. Fix date to be an actual date
- Use
  - mutate (https://www.rdocumentation.org/packages/dplyr/versions/0.7.8/topics/mutate to add/alter variables,
  - gsub (https://www.rdocumentation.org/packages/base/versions/3.6.2/topics/grep) with regular expressions (https://cheatography.com/davechild/cheat-sheets/regular-expressions/) for string manipulation,
  - and lubridate (https://www.rdocumentation.org/packages/lubridate/versions/1.7.10) for dates.

### mutate() variants

- mutate() modify/add variables
- mutate\_at() modify a set of variables.
- mutate\_if() modify variables meeting a criteria
- transmute() create a new set of variables based on previous.

### Make COVID data Long

#### glimpse(covid.data.long.2)

```
## Rows: 359,115
## Columns: 8
## $ file
         <chr> "data/confirmed.csv", "data/confi~
## $ `Country/Region` <chr> "Afghanistan", "Afghanistan", "Af~
                  <dbl> 33.93911, 33.93911, 33.93911, 33.~
## $ Lat
## $ Long
                  <dbl> 67.70995, 67.70995, 67.70995, 67.~
               <date> 2020-01-22, 2020-01-23, 2020-01-~
## $ Date
## $ Count
                  <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ Metric
                  <chr> "confirmed", "confirmed", "confir~
```

### Now make it wider

```
covid.data.long.3 <-
    covid.data.long.2 %>%
    select(-file) %>% #< important to drop in order to get the right result
    tidyr::pivot_wider(names_from=Metric, values_from=Count)</pre>
```

#### glimpse(covid.data.long.3)

```
## Rows: 124,600
## Columns: 8
## $ `Country/Region` <chr> "Afghanistan", "Afghanistan", "Af-
## $ Lat
                   <dbl> 33.93911, 33.93911, 33.93911, 33.~
                  <dbl> 67.70995, 67.70995, 67.70995, 67.~
## $ Long
## $ Date
                  <date> 2020-01-22, 2020-01-23, 2020-01-~
                  <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ confirmed
                   <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ deaths
                   <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ recovered
```

### select() - Choosing variables

#### **Key Function**

Use select() to choose the variables desired.

### **Basic Usage**

```
select(data, ...)
```

Over the next few examples we will explore the forms ... can take

### select() - Variable Names

the easiest is with variable names:

```
covid.data.long.2 %>%
    select(`Country/Region`, `Province/State`
    , Date, Metric, Count) %>%
    head
```

Country/Region	Province/State	Date	Metric	Count
Afghanistan	NA	2020-01-22	confirmed	0
Afghanistan	NA	2020-01-23	confirmed	0
Afghanistan	NA	2020-01-24	confirmed	0
Afghanistan	NA	2020-01-25	confirmed	0
Afghanistan	NA	2020-01-26	confirmed	0
Afghanistan	NA	2020-01-27	confirmed	0

### select() - Dropping by Variable Names

You can select everything **but** a variable with the minus operator

covid.data.long.2 %>% select(-file) %>% head()

Province/State	Country/Region	Lat	Long	Date	Count	Metric
NA	Afghanistan	33.93911	67.70995	2020-01-22	0	confirmed
NA	Afghanistan	33.93911	67.70995	2020-01-23	0	confirmed
NA	Afghanistan	33.93911	67.70995	2020-01-24	0	confirmed
NA	Afghanistan	33.93911	67.70995	2020-01-25	0	confirmed
NA	Afghanistan	33.93911	67.70995	2020-01-26	0	confirmed
NA	Afghanistan	33.93911	67.70995	2020-01-27	0	confirmed

### select() - By the numbers

You can select by variable position as well.

```
covid.data.long.3 %>% select(2, 3, 6:9) %>% head()
## Error: Can't subset columns that don't exist.
## x Location 9 doesn't exist.
## i There are only 8 columns.
```

### select() - by variable range

Use single colon: with variable names to select variables named and eveything in between:

```
covid.data.long.3 %>%
    select(`Province/State`, `Country/Region`, Date
    , confirmed:recovered) %>% head()
```

Province/State	Country/Region	Date	confirmed	deaths	recovered
NA	Afghanistan	2020-01-22	0	0	0
NA	Afghanistan	2020-01-23	0	0	0
NA	Afghanistan	2020-01-24	0	0	0
NA	Afghanistan	2020-01-25	0	0	0
NA	Afghanistan	2020-01-26	0	0	0
NA	Afghanistan	2020-01-27	0	0	0

# select() - by helpers

selection helpers are also provided:

covid.data.long.3 %>% select(contains("/")) %>% head()

Province/State	Country/Region
NA	Afghanistan

### select() - The helpers

The available helpers are:

- starts\_with()
- ends\_with()
- contains() must match literally
- matches() Regular expression match
- num\_range() numerical ranged variables with a prefix
- all\_of() must match all of given variables named in a vector.
- any\_of() select any variables present in given vector, but no error if not present.
- everything() Matches all variables, useful when reordering variables.
- last\_col() The last column

### select() - Multiple

You may use multiple forms together.

```
covid.data.long.3 %>%
   select(2:3, where(is.Date), confirmed:recovered) %>%
   head()
```

Country/Region	Lat	Date	confirmed	deaths	recovered
Afghanistan	33.93911	2020-01-22	0	0	0
Afghanistan	33.93911	2020-01-23	0	0	0
Afghanistan	33.93911	2020-01-24	0	0	0
Afghanistan	33.93911	2020-01-25	0	0	0
Afghanistan	33.93911	2020-01-26	0	0	0
Afghanistan	33.93911	2020-01-27	0	0	0

### Subsetting data

### **Key Function**

Subset data with the filter() function.

The base R version is subset, but it is FAR less robust.

It takes the form of

```
filter(data, expr1, expr2, ...)
```

where data is the data set, and expr1, expr2, ... are the criteria expressions evaluated *in the context of the data*. Data must meet *all* ctriteria to remain.

# filter() Example

Subset data to only confirmed cases for Nigeria.

```
covid.data.long.3 %>%
   select(2:3, Date:recovered) %>%
   filter( `Country/Region` == "US") %>%
   head()
```

Country/Region	Lat	Date	confirmed	deaths	recovered
US	40	2020-01-22	1	0	0
US	40	2020-01-23	1	0	0
US	40	2020-01-24	2	0	0
US	40	2020-01-25	2	0	0
US	40	2020-01-26	5	0	0
US	40	2020-01-27	5	0	0

### filter() Or

to perform an or use the single |

Country/Region	Lat	Date	confirmed	deaths	recovered
Canada	53.9333	2020-01-22	0	0	NA
Canada	53.9333	2020-01-23	0	0	NA
Canada	53.9333	2020-01-24	0	0	NA
Canada	53.9333	2020-01-25	0	0	NA
Canada	53.9333	2020-01-26	0	0	NA
Canada	53.9333	2020-01-27	0	0	NA

### filter() and

Independent statements to filter are combined assuming an and. You can make an and explicit with a &.

an alternate form would be to use %in%

```
covid.data.long.3 %>%
    select(2:3, Date:recovered) %>%
    filter( `Country/Region` %in% c('US', 'Canada', 'Mexico')
        , is.na(`Province/State`)
        ) %>% head()

## Error: Problem with `filter()` input `..2`.
## i Input `..2` is `is.na(`Province/State`)`.
## x object 'Province/State' not found
```

### distinct() - normalizing

From the previous filter example note that report date is repeated week after week.

:::{.keyfunction} To get only distinct observations, use distinct().:::

### distinct() - normalizing

```
covid.data.long.2 %>%
    filter( `Country/Region` %in% c('US', 'Canada', 'Mexico')
        , is.na(`Province/State`)
        ) %>%
    select(2:3, Metric) %>%
    distinct()
```

Province/State	Country/Region	Metric
NA	Mexico	confirmed
NA	US	confirmed
NA	Mexico	deaths
NA	US	deaths
NA	Canada	recovered
NA	Mexico	recovered
NA	US	recovered

34/50

### **Sorting Data**

### **Key Function**

To sort data use arrange()

sort() is the base version but again, less robust.

Arrange allows you to give sorting criteria.

### arrange() Example

```
covid.data.long.2 %>%
    arrange(Date, `Country/Region`, `Province/State`) %>%
    select(3:2, Date, Metric, Count) %>%
    head()
```

Country/Region	Province/State	Date	Metric	Count
Afghanistan	NA	2020-01-22	confirmed	0
Afghanistan	NA	2020-01-22	deaths	0
Afghanistan	NA	2020-01-22	recovered	0
Albania	NA	2020-01-22	confirmed	0
Albania	NA	2020-01-22	deaths	0
Albania	NA	2020-01-22	recovered	0

### **Combining data**

#### **Key Function**

Use the **join** family of functions to merge data together:

- inner\_join(a, b) keep only rows that match both a and b.
- · left\_join(a, b) keep all rows of a and add columns in b to the rows that match. Unmatched rows will contain missing values.
- right\_join(a, b) same as left but swap a and b.
- full\_join(a, b) keep all rows of both a and b.
- semi\_join(a, b) keep all rows of a that match b, but don't add columns from b.
- anti\_join(a, b) keep only those rows of a that don't match b.

#### Operations have these parameters:

- by variables to join on, defaults to common variables
- suffix sufixes to add to distinguish common variables that are not part of by

### **World Bank Data**

The wbstats package provides access to the world bank database.

```
library(wbstats)
wb_search('population', extra=TRUE)

(wb.pop.data <- wb_data(indicator ="SP.POP.TOTL", start_date = 2020, end_date = 2021))</pre>
```

Run these commands, investigate the output and then let's discuss.

## Example: Add country information to COVID data

Which join do we want to use?

```
covid.data.long.4 <-
    left_join(covid.data.long.3, wb.pop.data
    , by=c('Country/Region'='country')
)</pre>
```

Now we need to investigate...

# Summarization

#### **Summarization**

```
Key Function
summarise(data, ...)
```

Take the data and summarise it by performing the ... operations to it.

Total	N Missing	Number of countries	# of Reporting dates	max.cases	max.deaths
124600	12460	192	445	31151495	561783

## **Grouped Summarization**

```
Key Function
group_by(data, ...)

Take the data and group it by variables specified in ...,
all subsequent operations should be done by group.
```

## **Grouped Summarization**

```
covid.data.long.4 %>%
   group_by(`Country/Region`) %>%
   summarize( `Total` = n()
        , 'N Missing' = sum(is.na(SP.POP.TOTL))
        , 'Number of countries' = n_distinct(`Country/Region`)
        , "# of Reporting dates" = n_distinct(Date)
        , max.cases = max(confirmed, na.rm=TRUE)
        , max.deaths = max(deaths, na.rm=TRUE)
        )
```

Country/Region	Total	N Missing	Number of countries	# of Reporting dates	max.cases	max.deaths
Afghanistan	445	0	1	445	57144	2521
Albania	445	0	1	445	128155	2310
Algeria	445	0	1	445	118378	3126
Andorra	445	0	1	445	12497	120
Angola	445	0	1	445	23331	550 43/50

### **Exercise**

Find what didn't match?

2:00

### Solution

```
covid.data.long.4 %>%
    group_by(`Country/Region`) %>%
    summarize( `Total` = n()
        , 'N Missing' = sum(is.na(SP.POP.TOTL))
        , "# of Reporting dates" = n_distinct(Date)
        , max.cases = max(confirmed, na.rm=TRUE)
        , max.deaths = max(deaths, na.rm=TRUE)
        ) %>%
    filter(`N Missing` > 0)
```

Country/Region	Total	N Missing	# of Reporting dates	max.cases	max.deaths
Bahamas	445	445	445	9364	189
Brunei	445	445	445	219	3
Burma	445	445	445	142572	3206
Congo (Brazzaville)	445	445	445	10084	137
Congo (Kinshasa)	445	445	445	28542	745

## Question

What should we do with our data?

This data set on it's own is not very interesting.

Let's build something interesting.



### Get the desired population data

- SP.URB.TOTL.ZS Percentage of Population in Urban Areas (in % of Total Population)
- SP.POP.TOTL.MA.ZS Population, male (% of total)
- SP.POP.TOTL Population, total
- EN.POP.DNST Population density (people per sq km)
- IN.POV.HCR.EST.TOTL Poverty HCR Estimates (%) Total
- NY.GDP.PCAP.CD GDP per capita (current US\$)

### Look at the data

- 1. What format is it in?
- 2. Are there any problems?
- 3. Did we get get everything we expected?

# Join together

### Exercise/break

15:00