

# Introduction to data wrangling

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05-11-2019

# What we will cover today

Workshop is aimed at advanced beginners and I will assume some familiarity with R and the tidyverse.

We will spend a few minutes refamiliarising ourselves with R.

You will need to have `tidyverse`, `tidylog` and `here` packages installed and loaded. You also need to load the `lubridate` package.

Open the script called `code`.

- ▶ `Dplyr` main verbs and their friends
- ▶ Reshaping data
- ▶ Joining two data sets

# Packages in R

The first time you use it you need to install the package.

```
install.packages("tidyverse")
```

Load the package

```
library(tidyverse)
```

# The pipe

## Simplifying R code with pipes (`%>%`)

- ▶ Easy way to pass data through functions without nesting
- ▶ First argument of each function is “piped” in to reduce redundancy
- ▶ `f(x)` is the same as `x %>% f()`
- ▶ `f(x, y)` is the same as `x %>% f(y)`
- ▶ Keyboard shortcut `ctrl+shift +m`

# The pipe

```
leave_house(get_dressed(get_out_of_bed(wake_up(me))))
```

VS

```
me %>%  
  wake_up() %>%  
  get_out_of_bed() %>%  
  get_dressed() %>%  
  leave_house()
```

# General R tips

## Getting help

- ▶ `?functionname` will search all loaded packages
- ▶ `??functionname` will search all installed packages
- ▶ Most packages have vignettes
- ▶ Most tidyverse packages have cheatsheet (under help in the menu)

## Debugging

- ▶ start a new R session (keyboard short `ctrl+shift+F10`)
- ▶ the `::` operator is your friend eg `dplyr::select()`

# Dplyr verbs

Grammar of data manipulation:

Main verbs

- ▶ `select()` picks variables (columns) based on their names
- ▶ `filter()` allows row selection based on given criteria
- ▶ `mutate()` creates new variables (columns) from existing ones
- ▶ `summarise()` reduces multiple values down to a single summary

helper verbs

- ▶ `group_by()` performs any of the above on a group-by-group basis
- ▶ `arrange()` changes the ordering of rows

## dplyr syntax

- ▶ All calls to dplyr verbs follow the same format:
1. The first argument is a dataframe
  2. The subsequent arguments describe what to do to that dataframe, using unquoted variable names.

Helper functions you can use with `select()`:

- ▶ `starts_with("e_")` matches names that begin with "e\_"
- ▶ `ends_with("_end")` matches names that ends with "\_end"
- ▶ `contains("_h12")` matches names that contain "\_h12"
- ▶ `matches()` allows you to do regex matching on names eg `matches("abc|abd")`



## Example select()

```
new_dat <- starwars %>%  
  select( -height, -mass)  
  
starwars %>%  
  select(name, -ends_with("color"))
```

## Which bit of code produces the output?

```
starwars %>% select(birth_year) (pink) or starwars %>%  
select(-birth_year) (green)
```

```
# A tibble: 87 x 1
```

```
  birth_year
```

```
    <dbl>
```

```
1      19
```

```
2     112
```

```
3      33
```

```
4     41.9
```

```
5      19
```

```
6      52
```

```
7      47
```

```
8      NA
```

```
9      24
```

```
10     57
```

```
# ... with 77 more rows
```

## filter()

- ▶ Allows pointed row selection based on given criteria
- ▶ First argument is the dataframe, subsequent arguments are logical expressions used to filter the dataframe

Useful functions when working with filter

- ▶ `is.na()` and `!` to negate
- ▶ `str_detect()` to search for a string
- ▶ Logical comparisons (`<`, `<=`, `>`, `>=`, `!=`)

## Example filter

```
starwars %>%  
  filter(mass>80 | hair_color=="white")
```

```
starwars %>%  
  filter(is.na(hair_color))
```

Will this code run without an error?

Yes (pink) **or** No (green)

```
starwars %>%  
  filter(height+5)
```

## mutate() and transmute()

- ▶ Mutate creates new variables (columns) from existing ones and keeps existing ones
- ▶ can be used to overwrite and old variable eg starwars %>%  
mutate(mass=mass+5)
- ▶ Note: columns created with mutate() are always added to end of dataset
- ▶ transmute() creates new variables and drops existing ones

## mutate() useful functions

- ▶ Arithmetic operators (+, -, \*, /, ^)
- ▶ Log functions (like log10())
- ▶ Offsets like lead() and lag()
- ▶ ifelse statements (if this, then this, else this)
- ▶ Or when more than 1 logical split then use case\_when
- ▶ Cumulative and rolling aggregates
- ▶ Ranking (like ntile())
- ▶ toupper, str\_replace, str\_to\_title for character variables

## Example mutate

```
starwars %>%  
  mutate(height_m=height/100, bmi=mass/(height_m^2), bmi=round(bmi, 1))
```



# How many variables will the resulting dataset have

1 variable (pink) **or** 88 variables (green)

```
starwars %>%  
  transmute(height_m=height/100)
```

## Example summarise

```
starwars %>%  
  summarise(height_mean = mean(height, na.rm=TRUE))
```

```
starwars %>%  
  group_by(homeworld) %>%  
  summarise(height_max = max(height, na.rm=TRUE))
```

How many rows will the resulting data set have?

1 row (pink) **or** 87 rows (green)

```
starwars %>%  
  summarise(mass_mean = mean(mass, na.rm=TRUE))
```

## Time for you to try!

- ▶ Find the rows where `height > 90` and `hair_color` is not brown
- ▶ Find the rows where `eye_color` is brown
- ▶ Select `name` and `mass`
- ▶ Find the rows where `hair_color` is **NOT** missing
- ▶ Find the rows where `hair_color` contains the word `white`
- ▶ Select columns with the word `color` in them
- ▶ Select all columns BUT `height` and `mass`
- ▶ Create a new variable called `half_mass` that is half of `mass`
- ▶ Create variable `height_cat` with 4 categories “short”, “medium”, “tall” if `height` `[0,70]`, `]70,90]`, `]90,inf[` and ‘Not recorded’ if `height` is missing.
- ▶ Create variable `small` that is 1 if `mass` is less than the mean of `mass` and 0 otherwise.

## Solution

```
starwars %>%  
  filter(height>90, hair_color!="brown")
```

```
starwars %>%  
  filter(eye_color=="brown")
```

```
starwars %>%  
  select(name,mass)
```

```
starwars %>%  
  filter(!is.na(hair_color))
```

```
starwars %>%  
  filter(str_detect(hair_color, "white"))
```

```
starwars %>%  
  select(contains("color"))
```

# Scoped verbs

- ▶ Terminology: we have been using “single table verbs”
- ▶ Now we can affect multiple variables simultaneously with the scoped verbs
- ▶ Three extensions
  - ▶ `_if` pick variables based on a predicate function like `is.numeric()` or a user defined function `function(x) do_this(x)`
  - ▶ `_at` pick variables using the same syntax as `select()`.
  - ▶ `_all` operates on all variables

## mutate friends

```
starwars %>%  
  mutate_if(is.numeric, ~round(.) )
```

```
starwars %>%  
  mutate_at(vars(contains('color')), toupper)
```

## select friends

```
colour_vars <- c("hair_color", "skin_color", "eye_color")

starwars %>%
  select_at(colour_vars)

starwars %>%
  select_at(vars(-colour_vars))

starwars %>%
  select_if(~n_distinct(.) < 10)

numeric_vars <- starwars %>%
  select_if(is.numeric) %>%
  names()
```



## summarise friends

```
starwars %>%  
  summarise_at(numeric_vars, ~mean(.x, na.rm=TRUE))
```

```
starwars %>%  
  group_by(hair_color) %>%  
  summarise_if(is.numeric,  
    list(min = ~min(., na.rm = FALSE), max = ~max(., na.rm =
```

## Time for you to try again!

- ▶ What variables are not characters?
- ▶ For all numeric variables, create new variables with the mean by homeworld. The new variables should have the suffix `mean` eg `height_mean`. (Bonus, move the new variables so they are after `name`. Try `?everything` )
- ▶ For all character variables, replace missing values by 'not recorded'

## Solution

```
“{r, echo=FALSE} starwars %>% select_if(~!is.character(.)) %>%  
names()
```

```
starwars %>% group_by(homeworld) %>% mutate_if(is.numeric,  
list(mean=~mean(.x, na.rm=TRUE))) %>% select(name,  
ends_with('mean'), everything())
```

```
starwars %>% group_by(homeworld) %>%  
summarise_if(is.numeric, mean, na.rm=TRUE)
```

```
starwars %>% mutate_if(is.character, list(~replace_na(., “not  
recorded”)))
```

```
““
```

# Tidy data

Data comes in all kinds of shapes and forms.

1. Each variable forms a column.
2. Each observation forms a row.
3. Each type of observational unit forms a table.

## Sitrep data

Downloaded data from NHSE website and saved in R data format.

```
sitrep <- readRDS(here::here('data', 'sitrep.rds')) # all data  
sitrep_60sec <- readRDS(here::here('data', 'sitrep_60sec.rds')) # 60 second data
```

Look at the data. Is it tidy?

## New tidyr package

The tidyr package was updated early September. Spread and gather have been replaced by pivot\_longer and pivot\_wider.

```
=====
```

```
sitrep_long <- sitrep %>%  
  pivot_longer(-c(NHS_111_area_name, year), names_to='day_r
```

Our data is long! But we can make it even tidier. Suggestions?

## Sorting out the date

```
sitrep_long <- sitrep_long %>%  
  mutate(day_month=str_replace(day_month, '_', '-'), date=)
```

## Time for you to try

Reshape the `sitrep_60sec` data frame and create a date variable.  
Call your new data frame `sitrep_60sec_long`.



## Solution

```
sitrep_60sec_long <- sitrep_60sec %>%  
  pivot_longer(-c(NHS_111_area_name, year), names_to='day_m', values_to='count')  
  mutate(day_month=str_replace(day_month, '_', '-'), date=year)
```

## Join the two data set

Joining data in R is very similar to sql.

# Mutating joins

A mutating join allows you to combine variables from two tables. It first matches observations by their keys, then copies across variables from one table to the other.

- ▶ `inner_join(x,y)` only keeps observations where in both
- ▶ `full_join(x,y)` keeps all observations even if not in both
- ▶ `left_join(x,y)` keeps all observations in x
- ▶ `right_join(x,y)` keeps all observations in y

Join our two data sets

```
sitrep_full <- full_join(sitrep_long, sitrep_60sec_long, by = "date")
```

## Play with the data set

- ▶ Drop all the extra variables
- ▶ Create a new variable called `calls_60_p` with % of calls answered within 60 sec.
- ▶ Which area had the most calls in total looking at the full time period?
- ▶ Calculate the % of calls answered within 60 sec over the full time period by area and sort by your new variable.

## Solution

```
sitrep_full <- sitrep_full %>%  
  select(-contains('year'), -contains('day_month'))  
  
sitrep_full %>%  
  mutate(calls_60_p=calls_60sec/calls_all*100)  
  
sitrep_full %>%  
  group_by(NHS_111_area_name) %>%  
  summarise(calls_60_p=sum(calls_60sec, na.rm=TRUE)/sum(calls_all))  
  arrange(calls_60_p)
```

## Filtering joins

Filtering joins match observations in the same way as mutating joins, but affect the observations, not the variables.

- ▶ `anti_join(x,y)` drops all observations in `x` that have a match in `y`
- ▶ `semi_join(x,y)` keeps all observations in `x` that have a match in `y`

## Messy data

There is another file with some more data but something has gone wrong.

Which regions are missing?

```
sitrep_clinical_input_corrupt_long <- readRDS(here::here('s
```



## Solution

```
anti_join(sitrep_clinical_input_corrupt_long, sitrep_full,
```