# Week 4: 'Tidying' data 2

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## Welcome!

Welcome to week 4!

**Record the meeting** 

## Discussion!

#### **Two questions**

- What is one particular "messy data" problem have you encountered (or do you anticipate encountering) in your own work?
- What is an example of multiple, separate datasets that you might wish to combine together into a single dataset in your work? If you cannot immediately think of one, consider information that might be recorded or created in two different files, but which could be useful to combine.

(10 minutes)

## Review of last week's class

#### Last week we discussed wrangling and tidying data:

- 1. Reading in Data
- 2. Tidying data
- 3. Our tidy data tools

#### Homework

#### Effective and ineffective ways of filtering the data

```
data %>%
   filter(content == 1, content == 2, content == 3, content == 4, content == 5)

data %>%
   filter(content == "1", content == "2", content == "3", content == "4", content == "5")

data %>%
   filter(content == "1" | content == "2" | content == "3" | content == "4" | content == "5")

data %>% filter(between(content, 1, 5))

data %>% filter(content %in% c("1", "2", "3", "4", "5"))

data %>% filter(str_detect(content, "[1-5]"))
```

## Reminder

Check the output of functions you write carefully; you may find yourself *constantly* looking at and viewing your data!

#### How do I check/view my data?

There are many ways:

- my\_data (just typing the name of your data will print info about it!)
- glance(my\_data)
- str(my\_data)
- View(my\_data) (do not include in .Rmd document)
- skimr::skim(my\_data) (must install "skimr" first)
- psych::describe() (must install "psych" first)

## This week's topics

#### **Overview**

- 1. Tidy data: Data reshaping
- 2. Working with multiple data frames: joins
- 3. Grouped data operations with dplyr

We are by no means done with the data tidying tools we discussed last week, so don't feel like you should already have those completely mastered yet.

#### **Outline**

What is reshaping?

Why would you reshape?

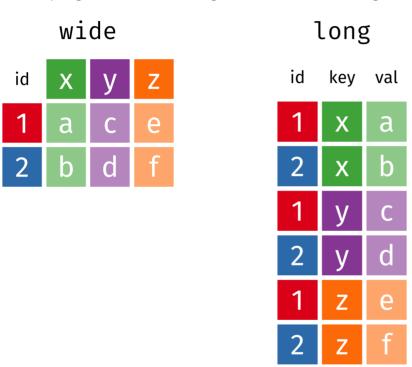
How do you reshape?

#### What is reshaping?

- Reshaping involves moving data between "long"er and "wide"er formats
  - Wide data has more columns and fewer rows
  - Long data has more rows and fewer columns
  - This is often a choice you can make with your data

#### What is reshaping?

• Reshaping involves moving data between "long"er and "wide"er formats



#### Why would you reshape?

One reason to reshape is to get data in a proper 'tidy' format.

The broader reason is that data can be in multiple shapes and you may need different ones for different purposes.

'Tidy' data doesn't always fully determine what you should do, particularly in the case of repeated measures data.

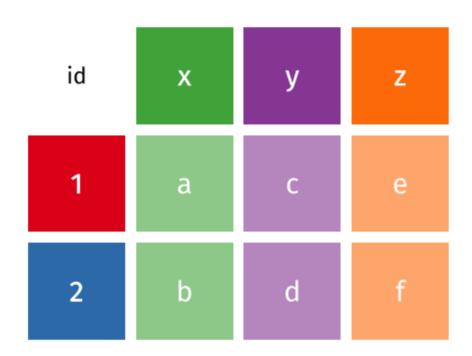
#### How do you reshape?

The tidyr package offers the gather() and spread() functions.

These are meant to be replaced by the "pivot" functions from dplyr, which are pivot\_longer() and pivot\_wider()

#### How do you reshape?

wide



#### What are joins?

Why are joins important?

What are the different types?

How do we know we're doing them right?

#### What are joins?

- When you have multiple data frames that you want to combine
- Need to have overlap

#### Why are joins important?

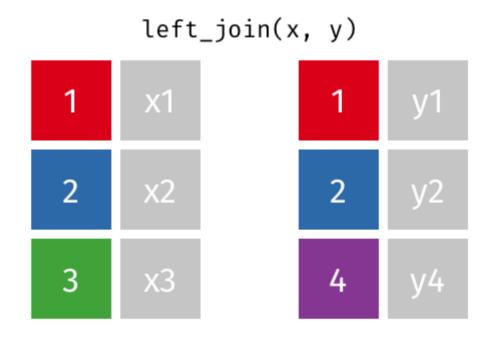
- They are the way we will use to integrate data from different sources for analysis
- Doing them incorrectly can substantially change your data set

#### Different types of joins

- Main types
  - left join
  - right join
  - full join
  - inner join
- Less common
  - semi join
  - anti join

dplyr::left\_join()

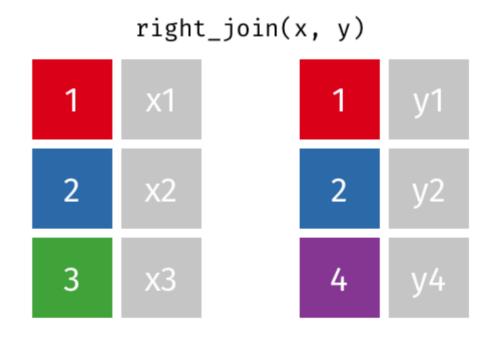
- Columns from both
- Matching rows from both



• non-matching rows from X only

dplyr::right\_join()

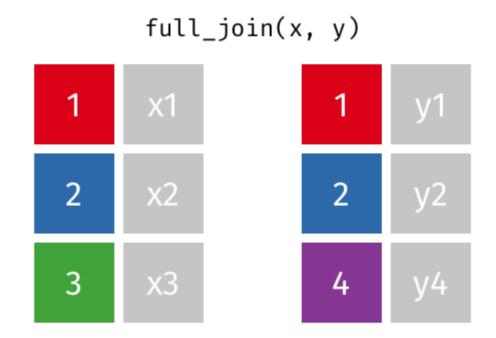
- Columns from both
- Matching rows from both



• non-matching rows from Y only

dplyr::full\_join()

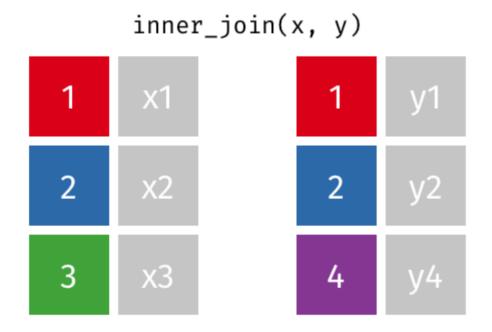
- Columns from both
- Matching rows from both



• non-matching rows from both

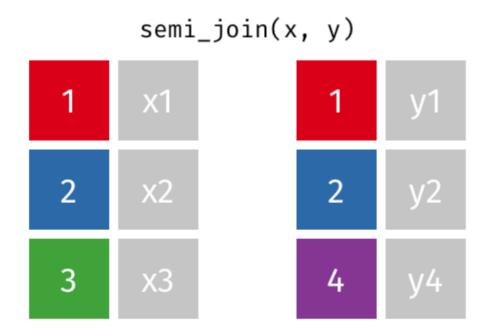
#### dplyr::inner\_join()

- Columns from both
- Matching rows from both
- non-matching rows from neither



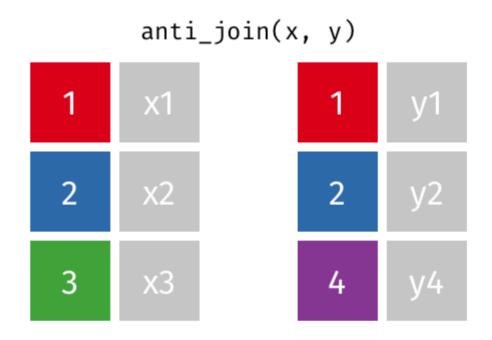
dplyr::semi\_join()

- Columns from X
- Matching rows from both
- non-matching rows from neither



dplyr::anti\_join()

- Columns from X
- No matching rows



• Only non-matching rows from X

How do we know we're doing them right?

• Short answer: Carefully inspecting your data before and after

How do we know which one to choose?

- Short answer: Knowing what the different types do and knowing your data
- It all depends on what you want.

#### Matching variables:

- Need to make sure matching variables are correct
- Join functions by default will match all names, but you can specify

#### Need to align variable names:

- Can do this by renaming variables
- Can also specify corresponding pairs in the join functions

#### Other issues to consider:

- Multiple matches
- Matching NAs
- Duplicate variable labels

#### How do we look at summary information about our variables?

What do we mean by grouping?

How do we group?

What can you do with grouped data?

How do we look at summary information about our data?

summarize() from dplyr allows us to look at various sorts of summary info.

#### What do we mean by grouping?

Grouping is when we create groups in our data based on a categorical variable (or something that can be transformed into one).

#### How do we group?

The dplyr function group\_by() allows us to create grouped data frames.

```
library(dplyr)
storms %>%
  group_by(name)
## # A tibble: 5 × 13
## # Groups:
              name [1]
           year month
                        day hour
                                  lat long status
                                                     category wind pressure
    name
    <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dr>
                                                          <ord>
##
                                                                   <int>
                                                                            <int>
           1975
                         27
                                0 27.5 -79 tropical de... -1
                                                                             1013
## 1 Amy
                                                                      25
## 2 Amy
           1975
                      27
                               6 28.5 -79 tropical de... -1
                                                                      25
                                                                             1013
           1975
                    6 27 12 29.5 -79 tropical de... -1
                                                                     25
                                                                            1013
## 3 Amy
## 4 Amy
           1975
                         27
                            18 30.5 -79 tropical de... -1
                                                                      25
                                                                             1013
## 5 Amy
                         28
                                0 31.5 -78.8 tropical de... -1
                                                                      25
           1975
                                                                             1012
## # ... with 2 more variables: ts diameter <dbl>, hu diameter <dbl>
```

#### What can we do with grouped data?

We can combine **summarize()** with grouped data frames and get summary information by group.

We can also do different sorts of data cleaning operations on grouped data.

#### Finding the mean wind speed for storms

```
library(dplyr)
storms %>%
  group_by(name) %>%
  summarize(mean wind speed = mean(wind))
## # A tibble: 198 × 2
               mean wind speed
      name
      <chr>
##
                         <dbl>
    1 AL011993
                          27.5
## 2 AL012000
                          25
   3 AL021992
                          29
## 4 AL021994
                          24.2
                          28.8
   5 AL021999
## 6 AL022000
                          29.2
## 7 AL022001
                          25
## 8 AL022003
                          30
   9 AL022006
                          38
## 10 AL031987
                          21.2
## # ... with 188 more rows
```

#### Finding the mean wind speed for storms and arranging

```
library(dplyr)
storms %>%
  group_by(name) %>%
  summarize(mean wind speed = mean(wind)) %>%
  arrange(desc(mean wind speed))
## # A tibble: 198 × 2
##
      name
              mean wind speed
     <chr>
                        <dbl>
   1 Wilma
                         91.9
                         89.2
## 2 Luis
                         88.1
   3 Hugo
## 4 David
                         86.8
## 5 Gonzalo
                         86.1
## 6 Ike
                         83.2
                         81.8
## 7 Igor
                         81.8
## 8 Joaquin
## 9 Rita
                         80.1
## 10 Gilbert
                         79.6
## # ... with 188 more rows
```

# Wrapping up

#### On Slack:

- What is one thing you took away from today?
- What is something you want to learn more about?