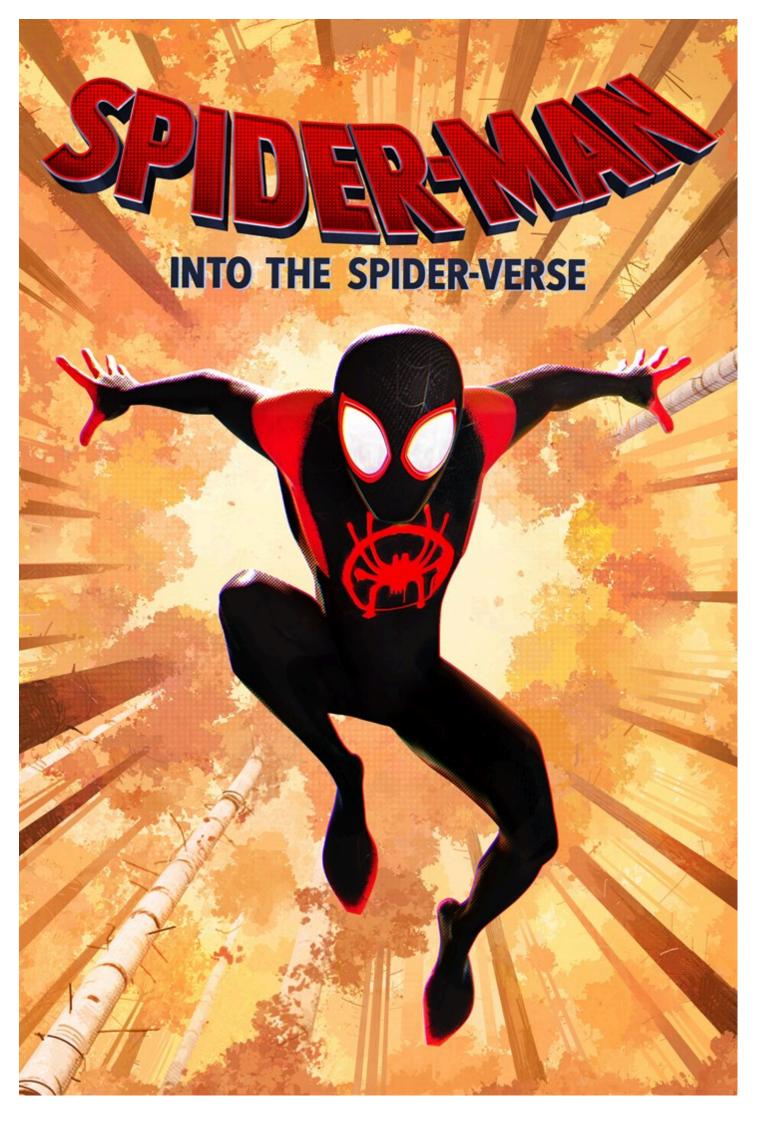
R08: Transforming and subetting data with tidyverse

Nicky Wakim 2024-11-04

Introduction to the tidyverse



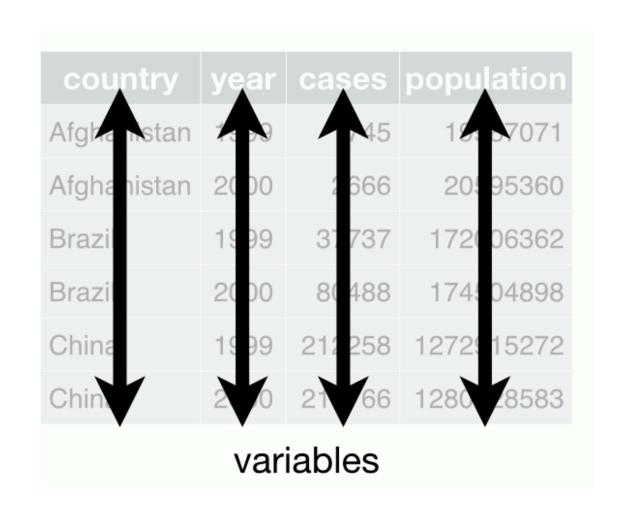
What is the tidyverse?

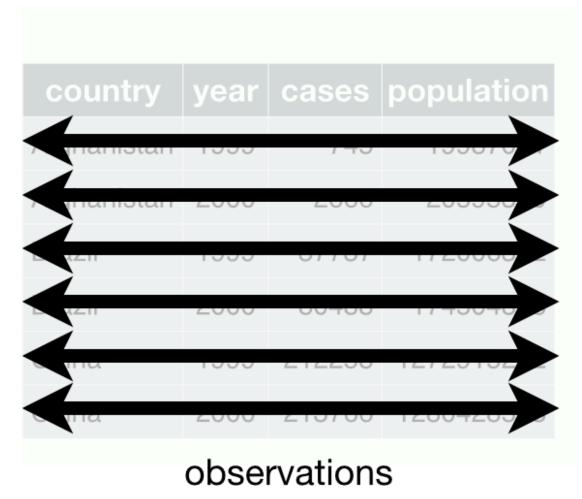
The **tidyverse** is a collection of R packages designed for data science. All packages share an underlying design philosophy, grammar, and data structures.

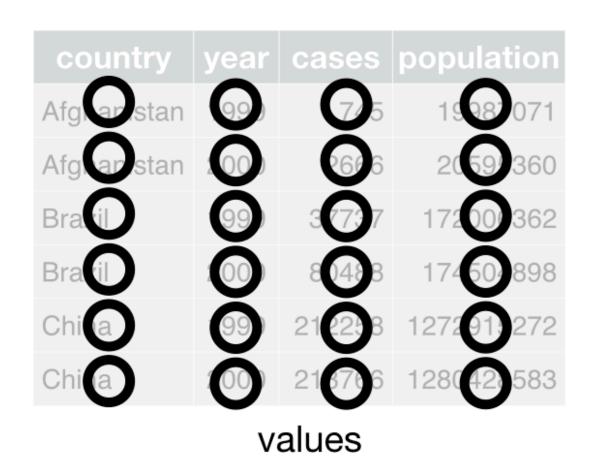
- ggplot2 data visualisation
- **dplyr** data manipulation
- tidyr tidy data
- readr read rectangular data
- purrr functional programming
- tibble modern data frames
- **stringr** string manipulation
- forcats factors
- and many more ...



Tidy data¹







- 1. Each variable must have its own column.
- 2. Each observation must have its own row.
- 3. Each value must have its own cell.

Pipe operator (magrittr)

• The pipe operator (%>%) allows us to step through sequential functions in the same way we follow if-then statements or steps from instructions

I want to find my keys, then start my car, then drive to work, then park my car.

Nested

Piped

```
1 find("keys") %>%
2  start_car() %>%
3  drive(to = "work") %>%
4  park()
```

Helpful functions for transforming and subsetting

Helpful functions for transforming and subsetting

Data transformation

- rename()
- mutate()
- pivot_longer() and pivot_wider()

Data subsetting

- filter()
- select()

New dataset: dds.discr

- In the US, individuals with developmental disabilities typically receive services and support from state governments
 - California allocates funds to developmentally disabled residents through the Department of Developmental Services (DDS)
- Dataset dds.discr
 - Sample of 1,000 people who received DDS funds (out of a total of ~ 250,000)
 - Data include age, sex, race/ethnicity, and annual DDS financial support per consumer

RO8 Slides

Let's look back at the dds.discr dataset

• We will load the data (This is a special case! dds.discr is a built-in R dataset)

```
1 data("dds.discr")
```

Now, let's take a glimpse at the dataset:

```
1 glimpse(dds.discr)
```

Helpful functions for transforming and subsetting

Data transformation

- rename()
- mutate()
- pivot_longer() and pivot_wider()

Data subsetting

- filter()
- select()

rename(): one of the first things I usually do

- I notice that two variables have values that don't necessarily match the variable name
 - Female and male are not genders (NIH page on sex and gender)
 - "White not Hispanic" combines race and ethnicity into one category (APA page on race and ethnicity)

I want to rename gender to sex (not sure if assigned at birth or current sex) and rename ethnicity to R_E (race and ethnicity)

rename(): one of the first things I usually do

- rename() can change the name of a column
- We use: data %>% rename(new_col_name = old_col_name)

Helpful functions for transforming and subsetting

Data transformation

rename()

- mutate()
- pivot_longer() and pivot_wider()

Data subsetting

- filter()
- select()

mutate(): constructing new variables from what you have

- We can create a new variable from other variables
 - Another way to say it: creates new columns that are functions of existing variables
- We often use it like:

```
1 data %>% mutate(new_variable = some_transformation_of_another_variable)
```

mutate(): create a new variable from two other variables

I want to make a variable that is the ratio of expenditures over age

```
1 dds.discr2 = dds.discr1 %>%
2  mutate(exp_to_age = expenditures/age)
3
4 glimpse(dds.discr2)
```

Recoding a numeric variable into categorical

Can we recreate age cohort using the age varible?

```
1 summary(dds.discr2)
      id
                                                       expenditures
               age.cohort
                                              SAB
                                age
       :10210 0-5 : 82
                                  : 0.0
                                          Female:503
                                                      Min.
                                                              : 222
Min.
                           Min.
1st Qu.:31809 6-12:175
                           1st Qu.:12.0
                                          Male :497
                                                      1st Qu.: 2899
Median :55384 13-17:212
                           Median :18.0
                                                       Median: 7026
      :54663 18-21:199
                                :22.8
                                                              :18066
Mean
                           Mean
                                                       Mean
3rd Qu.:76135 22-50:226
                           3rd Qu.:26.0
                                                       3rd Qu.: 37713
      :99898 51+ :106
                           Max. :95.0
                                                              :75098
                                                       Max.
Max.
               R E
                          exp to age
                               : 27.57
White not Hispanic:401
                        Min.
Hispanic
                 :376
                        1st Qu.:273.88
Asian
                        Median :461.75
                 :129
Black
                 : 59
                        Mean
                                   Inf
Multi Race
                 : 26
                        3rd Qu.:938.12
                 : 4
American Indian
                        Max.
                                   Inf
(Other)
```

RO8 Slides

Recoding a numeric variable into categorical (2/2)

- We can integrate other functions into mutate()
- For example, case_when() is a helpful function for mapping values to a category

Tidyverse:

```
1 dds.discr3 <- dds.discr2 %>%
2  mutate(
3    age.cohort2 = case_when(
4         age <= 5 ~ "0-5",
5         age <= 12 ~ "6-12",
6         age <= 17 ~ "13-17",
7         age <= 21 ~ "18-21",
8         age <= 50 ~ "22-50",
9         age >= 51 ~ "51+"
10    )
11  )
```

RO8 Slides

Have you noticed that I change the number on dds.discr?

- I change the number so that R saves a new dataset
- And I do not overwrite the previous dataset
- Can be annoying, but VERY helpful when you have to go back and change code
- When you run things in real time and troubleshoot, it will be helpful to have different versions of the same dataframe

Helpful functions for transforming and subsetting

Data transformation

- rename()
- mutate()
- pivot_longer() and pivot_wider()

Data subsetting

- filter()
- select()

filter(): keep rows that match a condition

• What if I want to subset the data frame? (keep certain rows of observations)

I want to look at the data for people who between 50 and 60 years old

```
1 dds.discr4 = dds.discr3 %>%
2  filter(age >= 50 & age <= 60)
3
4 glimpse(dds.discr4)</pre>
```

```
Rows: 23
Columns: 8
$ id
             <int> 15970, 19412, 29506, 31658, 36123, 39287, 39672, 43455, 4...
             $ age.cohort
             <int> 51, 60, 56, 60, 59, 59, 54, 57, 52, 57, 55, 52, 59, 54, 5...
$ age
$ SAB
             <fct> Female, Female, Female, Male, Female, Female, Mal...
$ expenditures <int> 54267, 57702, 48215, 46873, 42739, 44734, 52833, 48363, 5...
$ R E
             <fct> White not Hispanic, White not Hispanic, White not Hispani...
$ exp to age
             <dbl> 1064.0588, 961.7000, 860.9821, 781.2167, 724.3898, 758.20...
             <chr> "51+", "51+", "51+", "51+", "51+", "51+", "51+", "51+", "...
$ age.cohort2
```

Helpful functions for transforming and subsetting

Data transformation

- rename()
- mutate()
- pivot_longer() and pivot_wider()

Data subsetting

• filter()

• select()

select(): keep or drop columns using their names and types

What if I want to remove or keep certain variables?

I want to only have age and expenditure in my data frame

```
1 dds.discr5 = dds.discr2 %>%
2    select(age, expenditures)
3
4 glimpse(dds.discr5)
```

Resources

dplyr resources

• More dpylr functions to reference!

Additional details and examples are available in the vignettes:

- column-wise operations vignette
- row-wise operations vignette

and the dplyr 1.0.0 release blog posts:

- working across columns
- working within rows

R programming class at OHSU!

You can check out Dr. Jessica Minnier's R class page if you want more notes, videos, etc.

The larger tidy ecosystem

Just to name a few...

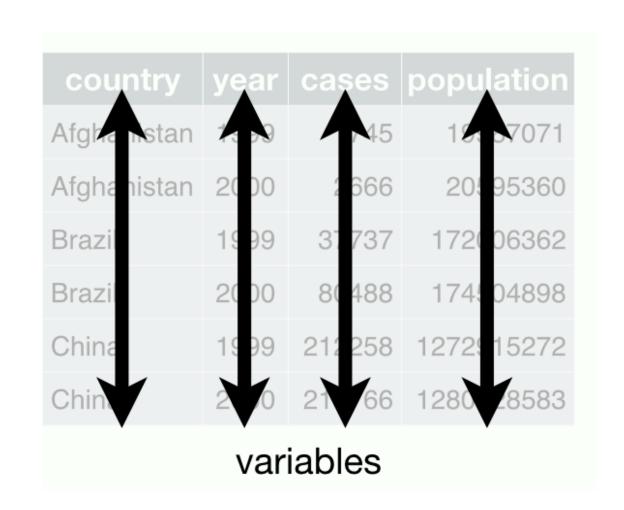
- janitor
- kableExtra
- patchwork
- gghighlight
- tidybayes

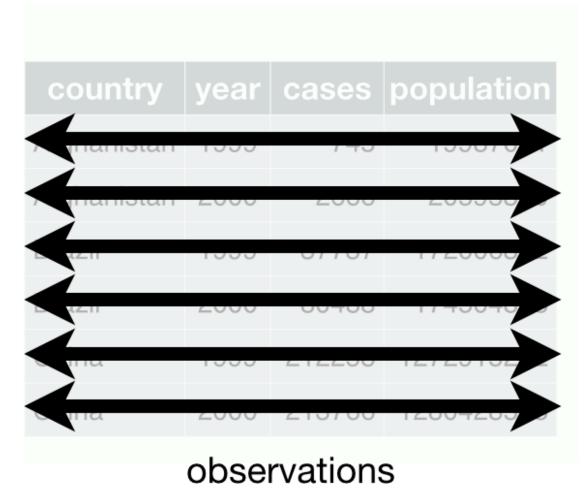
Credit to Mine Çetinkaya-Rundel

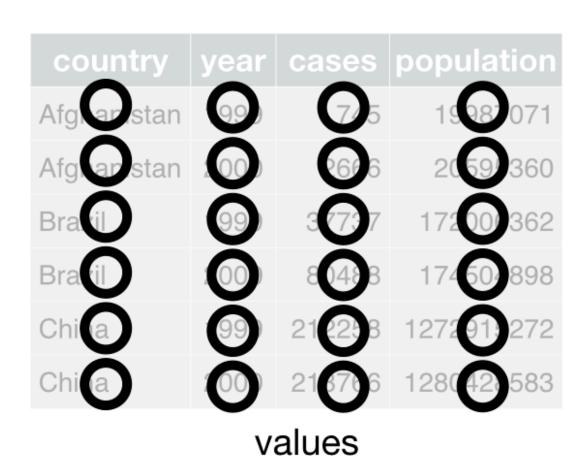
- These notes were built from Mine's notes
 - Most pages and code were left as she made them
 - I changed a few things to match our class
- Please see her Github repository for the original notes

If time

Tidy data¹







- 1. Each variable must have its own column.
- 2. Each observation must have its own row.
- 3. Each value must have its own cell.

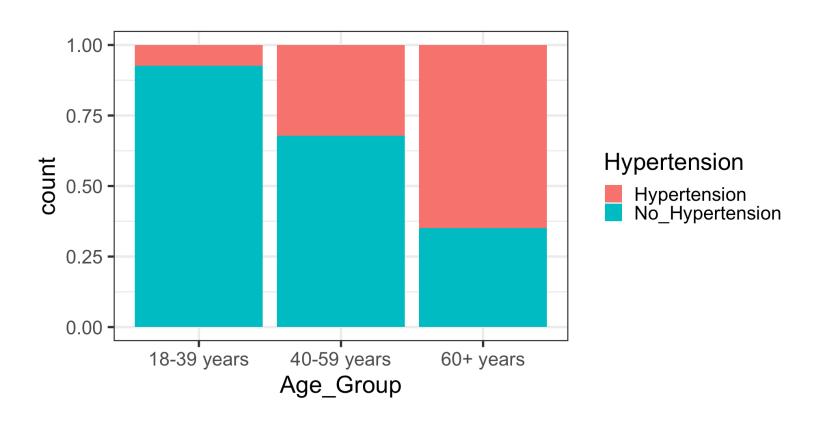
How do we make our data tidy??

- From a contingency table, we need to create the dataframe using the counts
- In Lesson 4, we saw this contingency table:

Table: Contingency table showing hypertension status and age group, in thousands.

Age Group	Hypertension	No Hypertension
18-39 yrs	8836	112206
40-59 yrs	42109	88663
60+ yrs	39917	21589

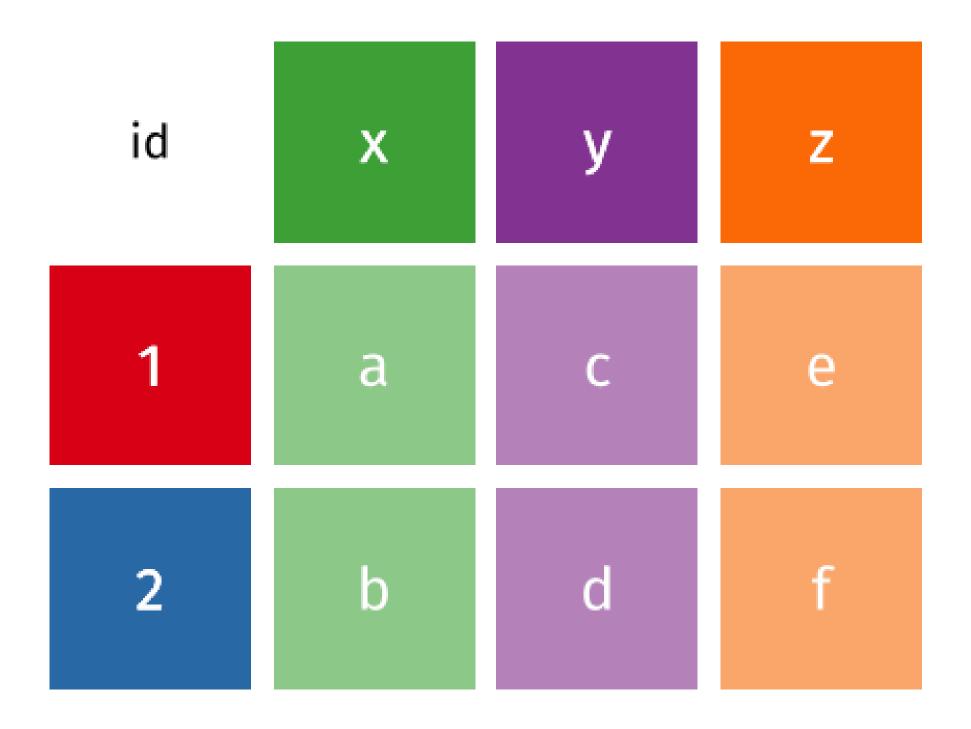
- And then I magically had it in a new format so I could make this plot:
- ► Code





pivot_*() functions

wide



l used pivot_longer() to create tidy data (1/2)

- Note that you won't be required to use pivot_longer()
 - I will give you data in a tidy form
- Here's the original data frame:

```
1 hyp_cont <- data.frame(
2 Age_Group = c("18-39 years", "40-59 years", "60+ years"),
3 Hypertension = c(8836, 42109, 39917),
4 No_Hypertension = c(112206, 88663, 21589))</pre>
```

- Note that I use use data frame () to make a data frame
- Then I can name each column that we saw in the contingency table
- Note that information about hypertension vs no hypertension is split between columns
 - And that we only have 3 rows of data to show all 313320 observations

l used pivot_longer() to create tidy data (2/2)

We need to tell pivot_longer():

- Which column must be repeated (pivoted) (all other columns are not repeating)
- The name of the new column that will contain the old variable names
- Where the values in each cell under the old variables will go

One more step to make it tidy

- Aka we need one more step to make it so every row is an observation
 - In this case, we want each row to represent data from one person

```
1 hyp_data = hyp_data1 %>% uncount(Counts)
 2 head(hyp data, 10)
# A tibble: 10 \times 2
  Age Group Hypertension
  <chr>
              <chr>
 1 18-39 years Hypertension
 2 18-39 years Hypertension
 3 18-39 years Hypertension
 4 18-39 years Hypertension
 5 18-39 years Hypertension
 6 18-39 years Hypertension
 7 18-39 years Hypertension
 8 18-39 years Hypertension
 9 18-39 years Hypertension
10 18-39 years Hypertension
```