# Tidying Data tidyr

2/22/23

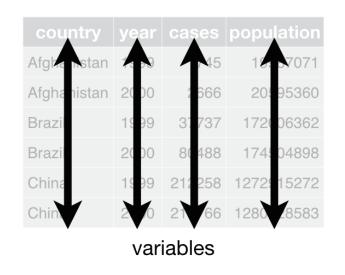
# tidyr

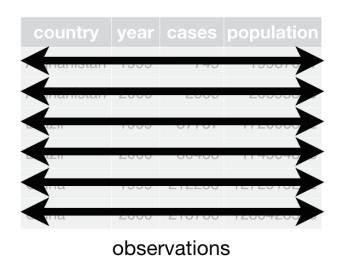
# Functions for tidying data. What is tidy data?

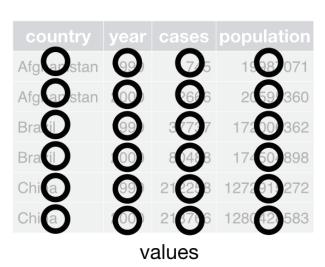


"Tidy datasets are all alike, but every messy dataset is messy in its own way." — Hadley Wickham

# **Tidy Data**







Each column is a single variable

Each row is a single observation

Each cell is a value

# pivot\_longer()

```
1 pivot_longer(<DATA>, <NAMES TO>, <VALUES TO>, <VARIABLES>)
```

# **Lord of the Rings**

```
lotr <- tribble(</pre>
                         ~film, ~race, ~female, ~male,
   "The Fellowship Of The Ring", "Elf", 1229L, 971L,
     "The Fellowship Of The Ring", "Hobbit", 14L, 3644L,
    "The Fellowship Of The Ring", "Man", OL, 1995L,
               "The Two Towers", "Elf", 331L, 513L,
               "The Two Towers", "Hobbit", 0L, 2463L,
               "The Two Towers", "Man", 401L, 3589L,
        "The Return Of The King", "Elf", 183L, 510L,
10
        "The Return Of The King", "Hobbit", 2L, 2673L,
        "The Return Of The King", "Man", 268L, 2459L
11
12 )
```

# **Lord of the Rings**

#### 1 lotr

```
# A tibble: 9 \times 4
  film
                                  female male
                           race
                           <chr> <int> <int>
 <chr>
1 The Fellowship Of The Ring Elf
                                   1229 971
2 The Fellowship Of The Ring Hobbit
                                      14 3644
 The Fellowship Of The Ring Man
                                         1995
                           Elf
                                     331 513
4 The Two Towers
5 The Two Towers
                           Hobbit
                                    0 2463
                                    401 3589
6 The Two Towers
                           Man
                           Elf
                                     183 510
7 The Return Of The King
8 The Return Of The King Hobbit
                                    2 2673
9 The Return Of The King
                                     268 2459
                           Man
```



# new data alert!



#### lotr

film female male 1 The Fellowship Of The Ring Elf 1229 971 2 The Fellowship Of The Ring Hobbit 14 3644 3 The Fellowship Of The Ring Man 1995 4 The Two Towers 331 513 5 The Two Towers Hobbit 0 2463 6 The Two Towers 401 3589 Man 7 The Return Of The King 183 510 8 The Return Of The King 2 2673 Hobbit 9 The Return Of The King 268 2459 Man

#### Where does it come from?

exercises

source:

github.com/jennybc/lotr-tidyr

#### How can I use it?

Run the code at the top of exercises

View(lotr)



this saves it in your global environment

# pivot\_longer()

```
1 lotr |>
2  pivot_longer(
3    names_to = "sex",
4    values_to = "words",
5    cols = female:male
6  )
```

# pivot\_longer()

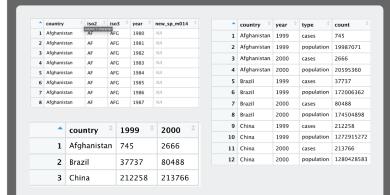
```
# A tibble: 18 \times 4
  film
                                          words
                                   sex
                             race
  <chr>
                             <chr> <chr> <int>
                                   female 1229
 1 The Fellowship Of The Ring Elf
 2 The Fellowship Of The Ring Elf
                                   male 971
 3 The Fellowship Of The Ring Hobbit female
                                             14
 4 The Fellowship Of The Ring Hobbit male
                                           3644
 5 The Fellowship Of The Ring Man
                                   female
 6 The Fellowship Of The Ring Man
                                   male
                                           1995
                             Elf female 331
 7 The Two Towers
                             Elf male
                                            513
 8 The Two Towers
                             Hobbit female
 9 The Two Towers
                            Hobbit male
10 The Two Towers
                                           2463
                                    £ _ __ _ 1 _
```



# new data alert!



#### table2, table4a, who



#### Where does it come from?

The tidyr R package

#### How can I use it?

library(tidyr)
 View(table2)
 View(table4a)
 View(who)



they're invisible!

Use pivot\_longer() to reorganize table4a into three columns: country, year, and cases.

```
1 table4a |>
2  pivot_longer(
3    names_to = "year",
4    values_to = "cases",
5    cols = -country
6 )
```

```
# A tibble: 6 \times 3
 country year
                 cases
 <chr> <chr> <chr> <dbl>
1 Afghanistan 1999
                    745
2 Afghanistan 2000 2666
3 Brazil
            1999 37737
4 Brazil
            2000 80488
5 China
            1999 212258
6 China
            2000
                 213766
```

# pivot\_wider()

1 pivot\_wider(<DATA>, <NAMES FROM>, <VALUES FROM>)

wide						
id	х	У	z			
1	a	С	е			
2	b	d	f			

wide

# pivot\_wider()

```
1 lotr |>
2  pivot_longer(
3    names_to = "sex",
4    values_to = "words",
5    cols = female:male
6  ) |>
7  pivot_wider(
8    names_from = race,
9    values_from = words
10  )
```

Use pivot\_wider() to reorganize table2 into four columns: country, year, cases, and population.

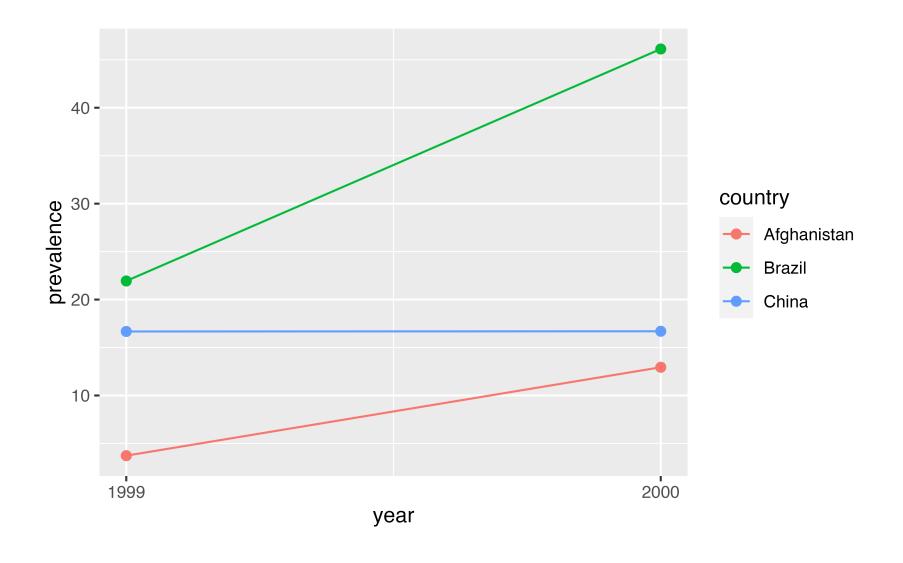
Create a new variable called prevalence that divides cases by population multiplied by 100000.

Pass the data frame to a ggplot. Make a scatter plot with year on the x axis and prevalence on the y axis. Set the color aesthetic (aes()) to country. Use size = 2 for the points. Add a line geom.

1 table2

```
country year cases population prevalence
 <chr> <dbl> <dbl>
                         <dbl>
                                 <dbl>
1 Afghanistan 1999 745 19987071
                                 3.73
2 Afghanistan 2000 2666 20595360
                                 12.9
3 Brazil 1999 37737 172006362
                                 21.9
4 Brazil 2000 80488 174504898
                                 46.1
5 China
           1999 212258 1272915272
                                 16.7
6 China
           2000 213766 1280428583
                                  16.7
```

```
1 table2 |>
2  pivot_wider(
3    names_from = type,
4   values_from = count
5  ) |>
6  mutate(prevalence = (cases / population) * 100000) |>
7  ggplot(aes(x = year, y = prevalence, color = country)) +
8  geom_point(size = 2) +
9  geom_line() +
10  scale_x_continuous(breaks = c(1999L, 2000L))
```



Pivot the 5th through 60th columns of who so that the names of the columns go into a new variable called codes and the values go into a new variable called n. Then select just the country, year, codes and n variables.

1 who

```
1 who |>
2    pivot_longer(
3         names_to = "codes",
4         values_to = "n",
5         cols = 5:60
6    ) |>
7         select(country, year, codes, n)
```

```
# A tibble: 405,440 \times 4
  country
               year codes
                                     n
  <chr> <dbl> <chr>
                                 <dbl>
 1 Afghanistan 1980 new sp m014
                                    NA
               1980 new sp_m1524
 2 Afghanistan
                                    NA
 3 Afghanistan
               1980 new sp m2534
                                    NA
 4 Afghanistan
               1980 new sp m3544
                                    NA
 5 Afghanistan
               1980 new sp m4554
                                    NA
 6 Afghanistan
               1980 new sp m5564
                                    NA
 7 Afghanistan
               1980 new sp m65
                                    NA
 8 Afghanistan
               1980 new sp f014
                                    NA
 9 Afghanistan
               1980 new sp f1524
                                    NA
10 Afghanistan 1980 new sp_f2534
                                    NA
   --<u>-</u>-----
```

# separate()/unite()

```
1 separate(<DATA>, <VARIABLE>, into = c("<VARIABLE1>", "<VARIABLE2
2 unite(<DATA>, <VARIABLES>)
```

# Use the cases data below. Separate the sex\_age column into sex and age columns.

```
1 cases <- tribble(
2    ~id,    ~sex_age,
3    "1",    "male_56",
4    "2",    "female_77",
5    "3",    "female_49"
6 )
7 separate(_____, ____, into = c("____", "___"))</pre>
```

```
# A tibble: 3 × 3
  id sex age
  <chr>      <chr>      <chr> 1 1 male 56
2 2 female 77
3 3 female 49
```

```
1 cases <- tribble(</pre>
   ~id, ~sex_age,
  "1", "male_56",
4 "2", "female_77",
5 "3", "female 49"
6
   separate(
   cases,
   sex age,
into = c("sex", "age"),
11 convert = TRUE
12 )
```

#### Your Turn 5: Challenge!

There are two CSV files in this folder containing SEER data in breast cancer incidence in white and black women. For both sets of data:

Import the data

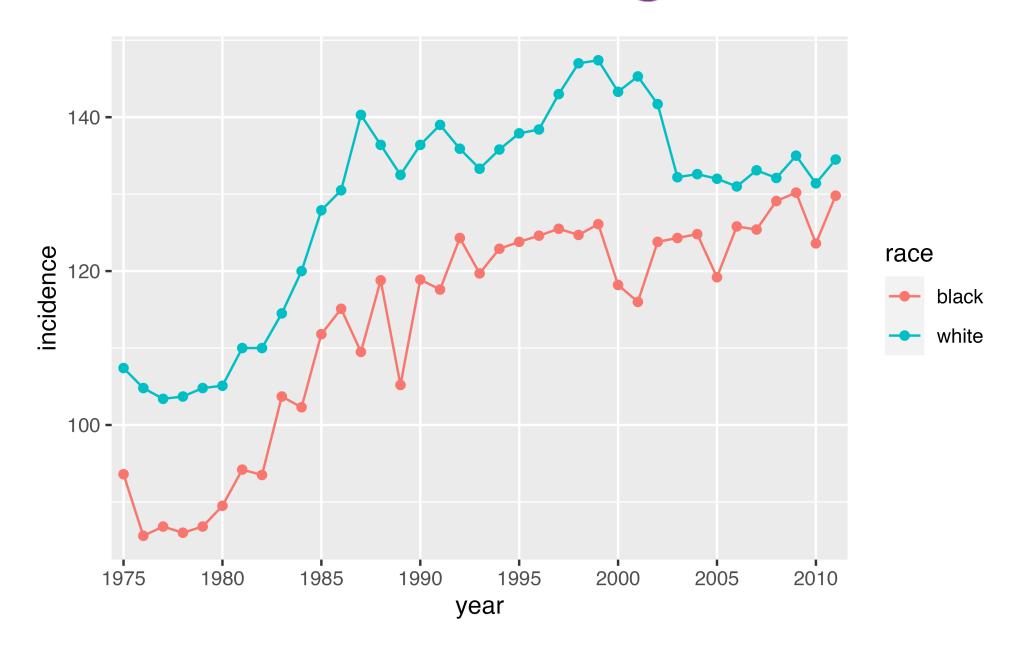
Pivot the columns into 2 new columns called year and incidence

Add a new variable called race. Remember that each data set corresponds to a single race.

Bind the data sets together using bind\_rows() from the dplyr package. Either save it as a new object or pipe the result directly into the ggplot2 code.

Plot the data using the code below. Fill in the blanks to have year on the x-axis, incidence on the y-axis, and race as the color aesthetic.

# Your Turn 5: No solution **Turn**



# Other neat tidyr tools Uncounting frequency tables

```
1 lotr |>
2  pivot_longer(
3    names_to = "sex",
4    values_to = "count",
5    cols = c(female, male)
6  ) |>
7    uncount(count)
```

## Other neat tidyr tools

```
# A tibble: 21,245 \times 3
   film
                               race
                                     sex
   <chr>
                               <chr> <chr>
                                     female
 1 The Fellowship Of The Ring Elf
                                     female
 2 The Fellowship Of The Ring Elf
 3 The Fellowship Of The Ring Elf
                                     female
 4 The Fellowship Of The Ring Elf
                                     female
                                     female
 5 The Fellowship Of The Ring Elf
 6 The Fellowship Of The Ring Elf
                                     female
                                     female
 7 The Fellowship Of The Ring Elf
 8 The Fellowship Of The Ring Elf
                                     female
 9 The Fellowship Of The Ring Elf
                                     female
10 The Fellowship Of The Ring Elf
                                     female
```

## Other neat tidyr tools

Work with data frames

```
crossing() and expand()
nest() and unnest()
```

## Other neat tidyr tools

Work with missing data

```
complete()
drop_na() and replace_na()
```

#### Resources

R for Data Science: A comprehensive but friendly introduction to the tidyverse. Free online.

RStudio Primers: Free interactive courses in the Tidyverse