

L02 - Tidy Data

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May 17, 2023

Tidy Data

“Happy families are all alike; every unhappy family is unhappy in its own way.” — Leo Tolstoy

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

- Key ideas:
 - *Cases* = *Rows*
 - *Variables* = *Columns*
- How should we define **case**?
- How do we identify **variables**?
- Advantages and Disadvantages

Vocabulary

Variable

- In data science, the word variable has a different meaning than in mathematics.
 - *In algebra, a variable is an unknown quantity.*
 - *In data, a variable is known; it represents a feature that has been measured or observed. “Variable” refers to a specific quantity or quality that can vary from one case to another.*
- Types of variables
 - *quantitative : a number*
 - *categorical (R calls these factors): tells which category or group a case falls into*
 - *all non-numerical values are categorical, but not all numerical values are quantitative*
 - e.g. zip code, IP address, dates

Cases

- Unit of observation or analysis
 - *this is extremely context specific*

What is Tidy Data

- Being neat is **not** what makes data tidy!

There are three interrelated rules which make a dataset tidy:

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

It is your job as the researcher to define the variables, observations, and values.

- The “tidiness” of the data set depends on the research question. It is not an inherent property to the data set itself.
- When data are in tidy form, it's often straightforward to transform the data into arrangements that are useful for answering interesting questions.

Example of Untidy data

	A	B	E	F	G	H	I	J
1	City of Minneapolis Statistics							
2	General Election November 5, 2013							
3	Ward	Precinct	Voters Registering by Absentee	Total Registrations	Voters at Polls	Absentee Voters	Total Ballots Cast	Total Turnout
4	City-Wide Total		708	6,634	75,145	4,954	80,099	33.38%
5								
6	1	1	3	28	492	27	519	27.23%
7	1	2	1	44	836	56	892	31.71%
8	1	3	0	40	905	19	924	38.87%
9	1	4	5	29	768	26	794	36.62%
10	1	5	0	31	683	31	714	37.46%
11	1	6	0	69	739	20	759	32.62%
12	1	7	0	47	291	8	299	15.79%
13	1	8	0	43	415	5	420	30.55%
14	1	9	0	42	596	25	621	25.42%
15	Ward 1 Subtotal		9	373	5,725	217	5,942	30.93%
16								
17	2	1	1	63	1,011	39	1,050	36.42%
18	2	2	5	44	679	37	716	50.39%
19	2	3	4	48	324	18	342	18.88%
20	2	4	0	53	117	3	120	7.34%
21	2	5	2	50	495	26	521	25.49%
22	2	6	1	36	433	19	452	39.10%
23	2	7	0	39	138	7	145	13.78%
24	2	8	1	50	1,206	36	1,242	47.90%
25	2	9	2	39	351	16	367	30.56%
26	2	10	0	87	196	5	201	6.91%
27	Ward 2 Subtotal		16	509	4,950	206	5,156	27.56%
28								
29	3	1	0	52	165	1	166	7.04%

Example of Tidy Data

ward	precinct	registered	voters	absentee	total.turnout
1	1	28	492	27	0.2723
1	4	29	768	26	0.3662
1	7	47	291	8	0.1579
2	1	63	1011	39	0.3642
2	4	53	117	3	0.0734
2	7	39	138	7	0.1378

... and so on for 117 rows altogether.

- Disadvantages
 - tidy data can be hard for human to quickly interpret
 - often not the ideal form for creating graphics
- Advantages
 - clear definitions
 - tidy data can easily be wrangled to a useful form for interpretation and visualization

Tidy Data Example

From <https://r4ds.had.co.nz/tidy-data.html>

You can represent the same underlying data in multiple ways. The example below shows the same data organised in four different ways. Each dataset shows the same values of four variables country, year, population, and cases, but each dataset organises the values in a different way.

Which ones of these is tidy?

Option 1

library(tidyverse)

```
## -- Attaching core tidyverse packages -- tidyverse 2.0.0 --
## ✓ dplyr      1.1.4    ✓ readr      2.1.5
## ✓ forcats    1.0.0    ✓ stringr   1.5.1
## ✓ ggplot2    3.5.1    ✓ tibble     3.2.1
## ✓ lubridate  1.9.3    ✓ tidyr      1.3.1
## ✓ purrr      1.0.2
## -- Conflicts -- tidyverse_conflicts() --
## ✖ dplyr::filter() masks stats::filter()
## ✖ dplyr::lag()     masks stats::lag()
## ! Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

table1

```
## # A tibble: 6 × 4
##   country year cases population
##   <chr>   <dbl> <dbl>   <dbl>
## 1 Afghanistan 1999    745  19987071
## 2 Afghanistan 2000   2666  20595360
## 3 Brazil      1999  37737  172006362
## 4 Brazil      2000  80488  174504898
## 5 China       1999 212258 1272915272
## 6 China       2000 213766 1280428583
```

Option 2

table2

```
## # A tibble: 12 × 4
##   country year type      count
##   <chr>   <dbl> <chr>   <dbl>
## 1 Afghanistan 1999 cases      745
## 2 Afghanistan 1999 population 19987071
## 3 Afghanistan 2000 cases      2666
## 4 Afghanistan 2000 population 20595360
## 5 Brazil      1999 cases      37737
## 6 Brazil      1999 population 172006362
## 7 Brazil      2000 cases      80488
## 8 Brazil      2000 population 174504898
## 9 China       1999 cases      212258
## 10 China      1999 population 1272915272
## 11 China      2000 cases      213766
## 12 China      2000 population 1280428583
```

Option 3

table3

```
## # A tibble: 6 × 3
##   country year rate
##   <chr>   <dbl> <chr>
## 1 Afghanistan 1999 745/19987071
## 2 Afghanistan 2000 2666/20595360
## 3 Brazil      1999 37737/172006362
## 4 Brazil      2000 80488/174504898
## 5 China       1999 212258/1272915272
## 6 China       2000 213766/1280428583
```

Option 4

table4a

```
## # A tibble: 3 × 3
##   country `1999` `2000`
##   <chr>   <dbl> <dbl>
## 1 Afghanistan    745    2666
## 2 Brazil      37737  80488
## 3 China     212258 213766
```

table4b

```
## # A tibble: 3 × 3
##   country `1999` `2000`
##   <chr>   <dbl> <dbl>
## 1 Afghanistan 19987071 20595360
## 2 Brazil    172006362 174504898
## 3 China    1272915272 1280428583
```

Example Continued

Table 1!

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	1280428583

variables

observations

values

Figure 12.1: Following three rules makes a dataset tidy: variables are in columns, observations are in rows, and values are in cells.

- Note that all tables contain the same information, just represented differently. Thus, we can transform Tables 2, 3, 4a/4b into Table 1, and vice versa.

Table 2 to Table 1

country	year	key	value
Afghanistan	1999	cases	745
Afghanistan	1999	population	19987071
Afghanistan	2000	cases	2666
Afghanistan	2000	population	20595360
Brazil	1999	cases	37737
Brazil	1999	population	172006362
Brazil	2000	cases	80488
Brazil	2000	population	174504898
China	1999	cases	212258
China	1999	population	1272915272
China	2000	cases	213766
China	2000	population	1280428583

table2

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	1280428583

Table 3 to Table 1

country	year	rate
Afghanistan	1999	745 / 19987071
Afghanistan	2000	2666 / 20595360
Brazil	1999	37737 / 172006362
Brazil	2000	80488 / 174504898
China	1999	212258 / 1272915272
China	2000	213766 / 1280428583

table3

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	1280428583

Table 4 to Table 1

Make each table tidy individually, then combine the two tables.

(you don't need to be able to interpret this code right now, just look at the end tables along the way.)

```
table4a.temp <-  
  table4a %>%  
  pivot_longer(!country, names_to = "year", values_to = "cases")  
  
table4b.temp <-  
  table4b %>%  
  pivot_longer(!country, names_to = "year", values_to = "population")  
  
left_join(table4a.temp, table4b.temp)
```

```
## Joining with `by = join_by(country, year)`
```

```
## # A tibble: 6 x 4  
##   country    year cases population  
##   <chr>      <chr> <dbl>      <dbl>  
## 1 Afghanistan 1999     745    19987071  
## 2 Afghanistan 2000    2666    20595360  
## 3 Brazil      1999    37737   172006362  
## 4 Brazil      2000    80488   174504898  
## 5 China       1999   212258  1272915272  
## 6 China       2000   213766  1280428583
```

Galton Data

In the 1880s, Francis Galton started to make a mathematical theory of evolution.

Here's part of a page from his lab notebook. Discuss the following in groups:

- What might he investigate with these data (e.g., **Research Question**)?
- Are these data **tidy** according to our definition?
- What are the **cases**?
- What are the **variables**?
- How many **rows** of data should the result have?
- How many **columns** of data should the result have? What is the data type of each column?
- What are some additional variables (not yet shown) that might be of interest? How would you recommend showing that information in the data table?

FAMILY HEIGHTS. from REF
(add 60 inches to every entry in the Table)

	Father	Mother	Sons in order of height	Daughters in order of height.
1	18.5	7.0	13.2	9.2, 9.0, 9.0
2	15.5	6.5	13.5, 12.5	5.5, 5.5
3	15.0	about 4.0	11.0	8.0
4	15.0	4.0	10.5, 8.5	7.0, 4.5, 3.0
5	15.0	1.5	12.0, 9.0, 8.0	6.5, 2.5, 2.5

A page from Francis Galton's notebook.

Activity 01: Tidy Data

Work to put these tables in tidy form

- Work with your partner
- As a team, you will put two different data sets into “tidy” form.
- **See Canvas for details**
 - View-only source data is provided
 - use any software you like
 - must submit a CSV to Canvas
 - do not use spaces in your file names
- **Tip: Sketch things out together on paper before you do anything in the computer**

Table 1: Galton's Height measurements data

	Father	Mother	Sons in order of height	Daughters in order of height
1	18.5	7.0	13.2	9.2, 9.0, 9.0
2	15.5	6.5	13.5, 12.5	5.5, 5.5
3	15.0	about 4.0	11.0	8.0
4	15.0	4.0	10.5, 8.5	7.0, 4.5, 3.0
5	15.0	1.5	12.0, 9.0, 8.0	6.5, 2.5, 2.5

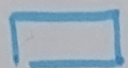
A page from Francis Galton's notebook.

Table 2: Presidents

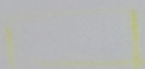
US Presidents

Name	Start Date	End date	VP
Adams, John	March 4, 1797	March 4, 1801	Thomas Jefferson
James Madison	March 4, 1809	March 4, 1817	George Clinton (03/04/1809 - 04/20/1812)
Martin VanBuren	4 th March, 1837	4 th March 1841	Elbridge Gerry (03/04/1813 - 11/23/1814)
William Henry Harrison	09/04/1841	04/04/1841	Richard Mentor Johnson
John Tyler	April 4 th , 1841	March 4 th , 1845	John Tyler Vacant throughout entire presidency

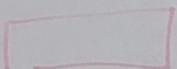
Leg:



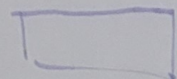
= Federalist



= Democratic-Republican



= Democrat



= Whig

Code Books

What is a code book?

- A **codebook** describes the contents, structure, and layout of a data collection.
- A well-documented codebook contains information intended to be complete and self-explanatory for each variable in a data file
- <https://www.icpsr.umich.edu/web/ICPSR/cms/1983>
- Federal Elections Commission
 - <https://www.fec.gov/data/browse-data/?tab=bulk-data>

References

- <https://dtkaplan.github.io/DataComputingEbook/chap-tidy-data.html#chap:tidy-data>
- <https://r4ds.had.co.nz/tidy-data.html>
- <https://www.icpsr.umich.edu/web/ICPSR/cms/1983>