# **Tidy Data**

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# Introduction

### Goals for this talk

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- Displaying and organizing tidy data

#### Where did this come from?

Most of what follows is based off of Hadley Wickham's paper on tidy data.

If you're looking for a practical introduction (in R), Hadley Wickham has one of those too.



I also borrow from other resources (listed at the end), as well as my own experience working with tidy and untidy datasets.

# Idea (the theory)

# What is tidy data?

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- 2. Each observation forms a row.
- 3. Each type of observational unit forms a table.

### The Language of Tidy Data

- Dataset: a collection of values (e.g. iris data)
- Variable: all values that measure the same underlying attribute (e.g. height, width)
- Values: a specific measurement or attribute for a variable (e.g. \$100)
- Observation: all values measured on the same unit (like a person, or a day, or a game) across variables

It's usually easy to figure out things like *observations* and *variables* for a given dataset, but defining them in the abstract can be difficult.

# **Execution** (the practice)

# Five common types of untidy data

Here are the five most common types of untidy data you're likely to experience "in the wild":

- Column headers are values, not variable names.
- Multiple variables are stored in one column.
- Variables are stored in both rows and columns.
- Multiple types of observational units are stored in the same table.
- A single observational unit is stored in multiple tables.

We'll go through examples of three of the five.



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### 1. Column headers are values, not variable names

The first dataset we'll look at comes from the WHO and displays the number of TB cases for three countries in two years.

| country     | 1999    | 2000    |
|-------------|---------|---------|
| Afghanistan | 745     | 2,666   |
| Brazil      | 37,737  | 80,488  |
| China       | 212,258 | 213,766 |

This data is too *wide*, as 1999 and 2000 are **values** for a **variable** we could call year.

Although difficult to analyze, this format is helpful for presentation and data entry.

# Tidying # 1

Need to gather (like UNPIVOT in SQL) columns into key-value (year-cases) pairs:

| country     | year | tb_cases |
|-------------|------|----------|
| Afghanistan | 1999 | 745      |
| Brazil      | 1999 | 37,737   |
| China       | 1999 | 212,258  |
| Afghanistan | 2000 | 2,666    |
| Brazil      | 2000 | 80,488   |
| China       | 2000 | 213,766  |

# 2. Multiple variables are stored in one column

The next table has two columns, but it should have four. How would you work with this data without tidying it first?

| Hair - Eye - Sex     | n  |
|----------------------|----|
| Black - Brown - Male | 32 |
| Brown - Brown - Male | 53 |
| Red - Brown - Male   | 10 |
| Blond - Brown - Male | 3  |
| Black - Blue - Male  | 11 |
| Brown - Blue - Male  | 50 |
|                      |    |

The Hair - Eye - Sex variable actually has values for three separate variables stored within it.

# Tidying #2

We need to separate one column (Hair - Eye - Sex) into multiple columns (hair, eye, sex)

| hair  | eye   | sex  | n  |  |
|-------|-------|------|----|--|
| Black | Brown | Male | 32 |  |
| Brown | Brown | Male | 53 |  |
| Red   | Brown | Male | 10 |  |
| Blond | Brown | Male | 3  |  |
| Black | Blue  | Male | 11 |  |
| Brown | Blue  | Male | 50 |  |

### 3. Variables are stored in both rows and columns

This is the most complicated form of untidy data, and typically requires a bit more massaging.

| id      | year | month | element | d1 | d2   | d3   | d4 | d5   |
|---------|------|-------|---------|----|------|------|----|------|
| MX17004 | 2010 | 2     | tmax    | NA | 27.3 | 24.1 | NA | NA   |
| MX17004 | 2010 | 2     | tmin    | NA | 14.4 | 14.4 | NA | NA   |
| MX17004 | 2010 | 3     | tmax    | NA | NA   | NA   | NA | 32.1 |
| MX17004 | 2010 | 3     | tmin    | NA | NA   | NA   | NA | 14.2 |
|         |      |       |         |    |      |      |    |      |

Think carefully about what the **observation** is for this data.

# Tidying #3

Requires gathering, spreading (like PIVOT in SQL), and unite-ing.

| id      | date       | tmax | tmin |
|---------|------------|------|------|
| MX17004 | 2010-02-01 | NA   | NA   |
| MX17004 | 2010-02-02 | 27.3 | 14.4 |
| MX17004 | 2010-02-03 | 24.1 | 14.4 |
| MX17004 | 2010-02-04 | NA   | NA   |
| MX17004 | 2010-02-05 | NA   | NA   |
| MX17004 | 2010-03-01 | NA   | NA   |
|         |            |      |      |

# **Displaying and Organizing Data**

How can we make it easier to scan raw values in a data table?

 Determine the roles of variables in your analysis (fixed by design of experiment vs. measured during course of experiment)

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  - Order from L-R by degree of fixed-ness. The "most fixed" variables are the key describers of an observation, and are useful when we're trying to scan values.
- Put related variables next to each other
- Order rows based on the first variable and then break ties with the second and subsequent (fixed) variables after that.

# Organizing data in spreadsheets

Broman & Woo (2018) wrote a short paper with 12 tips for organizing data in spreadsheets for sharing, analysis, reproducibility, and collaboration. After reading the tidy data paper, I would recommend reading it.

- Be consistent
  - Codes, NA, names, ID, layout, files, dates, phrases
- Write dates like YYYY-MM-DD
- Do not leave any cells empty
- Put just one thing in a cell
- Organize the data as a single rectangle (with subjects as rows, variables as columns, and with a single header row)
- Create a data dictionary

- Do not include calculations in the raw data files
- Do not use font color or highlighting as data
- Choose good names for things
- Make backups
- Use data validation to avoid data entry errors
- Save the data in plain text files

# **Conclusion**

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  - Code is for humans, data is for computers (relevant tweet from Vince Buffalo)
- Be consciously aware of your values, variables, and observations
- Normalization can be your friend
- Be assertive and understanding

#### Other resources

There's a bevy of resources I consulted for this presentation. I've arranged these in descending order of importance.

- The tidy data paper
- Data Organization in Spreadsheets
- Informal version of tidy data paper
- Practical introduction to tidy data
- Tidy data presentation
- Tidy data analysis (an extension of the tidy data paradigm)
- Tidy Data in Python
- Database Normalization
- Codd's 3rd Normal Form

Here's the link to my GitHub repository.

**Questions?** 

Thanks for listening!