



# Data Processing for D

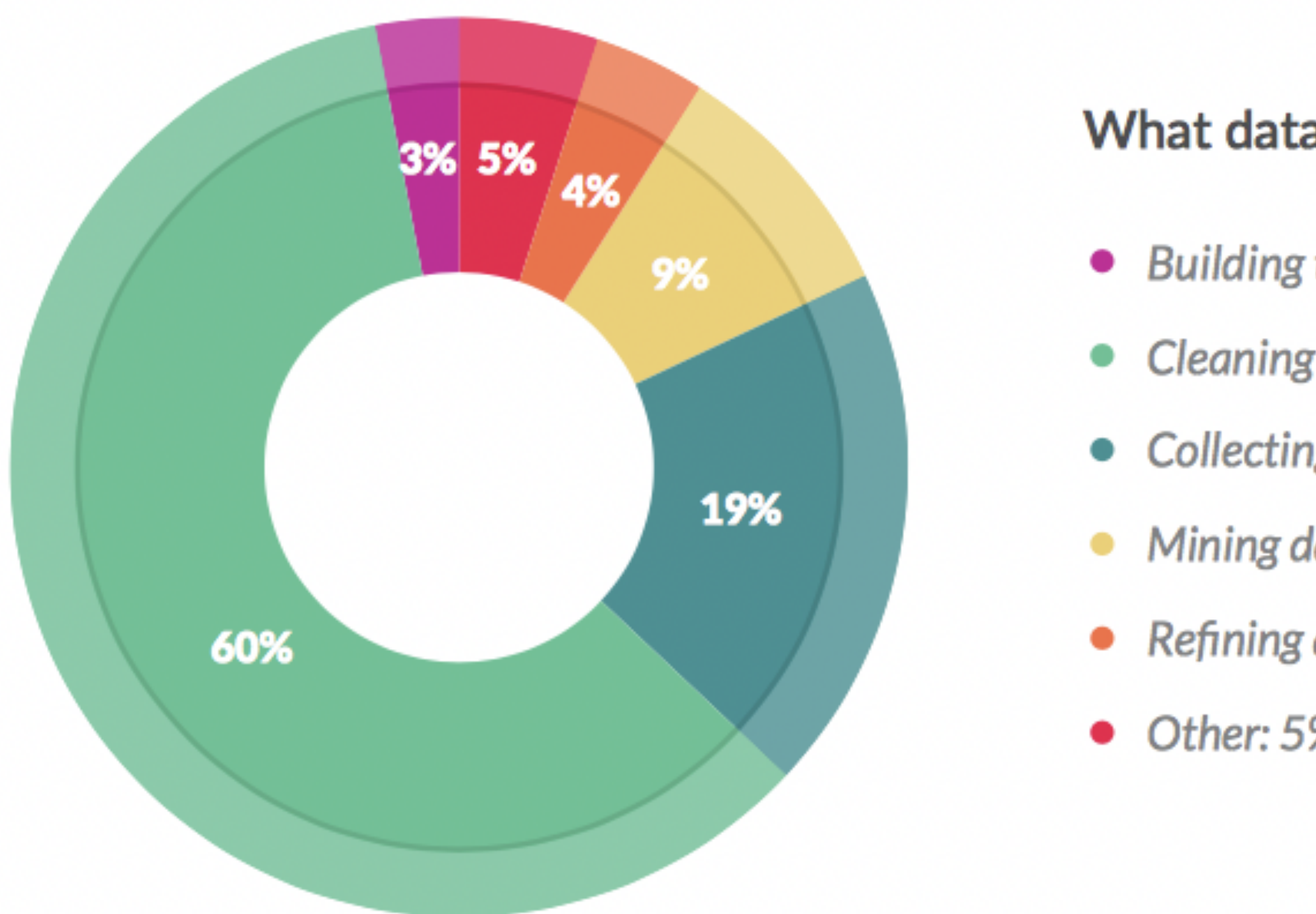
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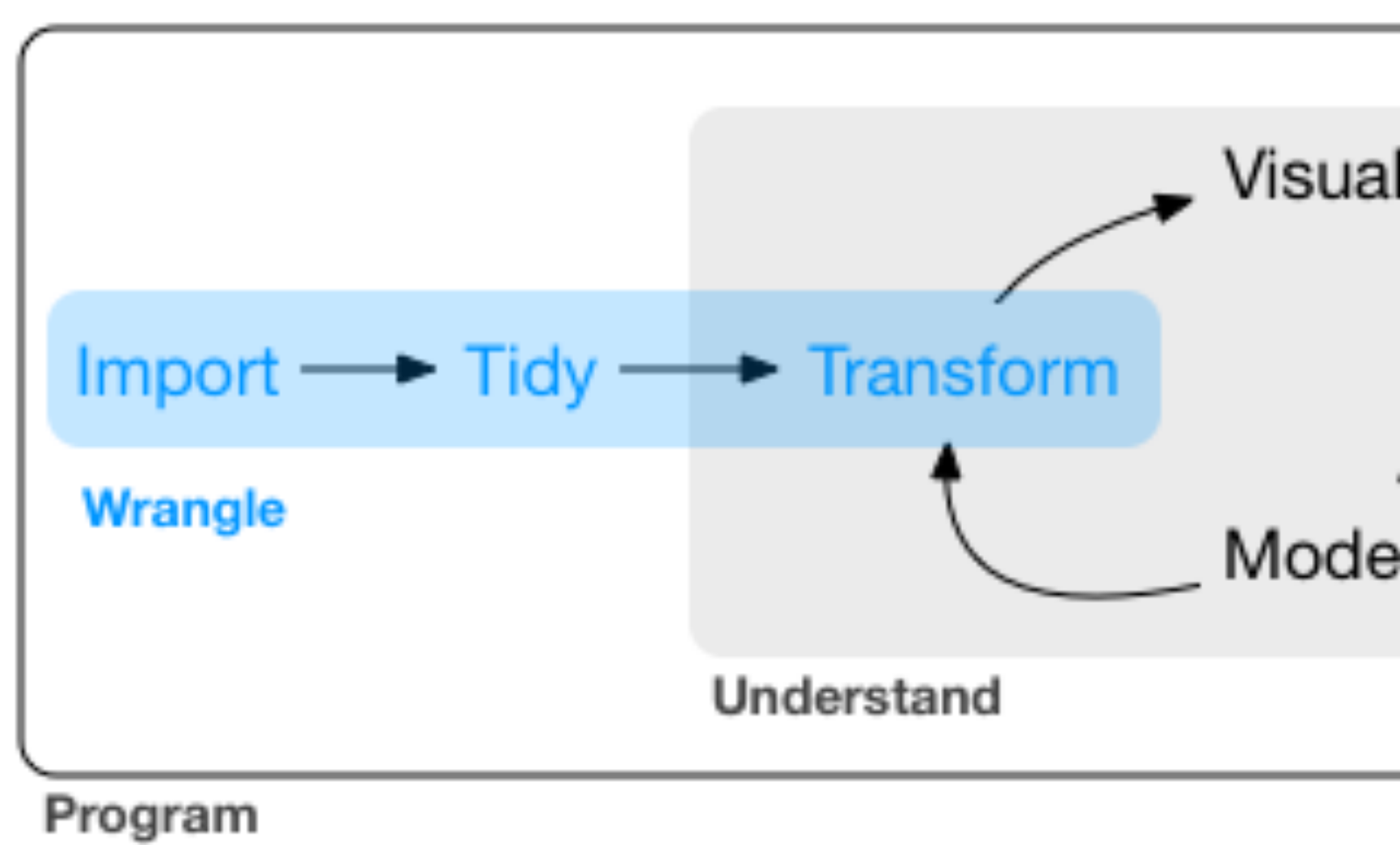
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# Wrangling o



<https://visit.figure-eight.com/rs/416-ZBE-142/images/Crow>

# Wrangling data



*R for Data Science*, Wickham and

## Learning goals

- Types of data
- Structuring data for data science
- Data wrangling and data cleaning
- Summarizing data

# TYPES OF D

Data comes in many forms

- **Structured** data is highly organized and is easy to query, transform, and analyze
- **Semi-structured** data has some organization but requires some processing
- **Unstructured** data is unorganized and requires significant tidying

## Unstructured

- Text, video, images, etc.
- Vast majority of data in abundant on the internet
- Requires significant processing useable for data analysis

## Semi-structure

- Structured text, JSON, XML
- Follows a structure (e.g. CSV)  
requires transformation
- Structured elements often have a specific purpose besides data analysis
- Required amount of processing



## Structured data

- Tables in a database or spreadsheets
- High level of organization
- May follow a schema (blueprint)
- Easy to query, transform

## Tabular data

- Most common kind of structured data
- Follows a “table” format
  - ◆ Rows and columns
  - ◆ Values in cells
- Tables in a RDBMS
- Data frames in R, Python

# STRUCTURIN

## Data model

- A **data model** is a concept to organize elements of data
- A **data model** is analogous to a data type in computer programming

## Common data models

- Relational data
- Key-value pairs
- Graphics and networks
- Arrays and matrices
- Tree structures

## Common data models

- Relational data —
- Key-value pairs —
- Graphics and networks
- Arrays and matrices
- Tree structures —

## Goals of structure

- Make the data easier to work with
- Ideal structure may differ from the structure required for the desired computation
  - ◆ Exploratory analysis — “tidy” tabular
  - ◆ Machine learning — arrays and matrices
- May need to transform data from one structure to another or different data models

## “Tidy” data

- Each variable forms a column
- Each observation forms a row
- Each value is a cell
  - ◆ Stricter: *Each type of observation*

Hadley Wickham. “Tidy Data.” *Journal of Statistical Software*



## Useful definitions

- A dataset is a collection of data
- An **observational unit** is a unit of analysis on which values are measured
- A **variable** is a quantity, quality or attribute that is measured
- An **observation** is a set of values for all variables made under similar conditions

# Tidy data

| country     | year | cases  | population |
|-------------|------|--------|------------|
| Afghanistan | 1999 | 745    | 19337071   |
| Afghanistan | 2000 | 2666   | 20595360   |
| Brazil      | 1999 | 37737  | 172006362  |
| Brazil      | 2000 | 80488  | 174504898  |
| China       | 1999 | 212258 | 1272915272 |
| China       | 2000 | 213766 | 128028583  |

variables

| country     | year | cases  | p |
|-------------|------|--------|---|
| Afghanistan | 1999 | 745    |   |
| Afghanistan | 2000 | 2666   |   |
| Brazil      | 1999 | 37737  |   |
| Brazil      | 2000 | 80488  |   |
| China       | 1999 | 212258 |   |
| China       | 2000 | 213766 |   |

observations

## Why tidy data

- Easy to query, transform
- Consistent format allows a variety of tools (e.g., `ggplot2`)
- Relationship to RDBMS
  - ◆ Concept of “tidy” data mirrors but framed in language of statistics

## “Messy” data

- “Messy” data frequently
- Storage and/or computation
  - ◆ Messy form may be more common
  - ◆ Matrices/arrays preferable for storage
- Ease of data entry
  - ◆ Data entry by hand
  - ◆ Recording instrument
- “Tidy” form not easily entered

## Common symptoms of

- Column headers are values
- Multiple variables are stored
- Variables are stored in both
- A single observation is stor

Is it tidy?

|    |    |                |       |       |
|----|----|----------------|-------|-------|
| ## | #  | tibble: 12 x 4 |       |       |
| ## |    | country        | year  | type  |
| ## |    | <chr>          | <int> | <chr> |
| ## | 1  | fghanistan     | 1999  | cases |
| ## | 2  | fghanistan     | 1999  | popul |
| ## | 3  | fghanistan     | 2000  | cases |
| ## | 4  | fghanistan     | 2000  | popul |
| ## | 5  | Brazil         | 1999  | cases |
| ## | 6  | Brazil         | 1999  | popul |
| ## | 7  | Brazil         | 2000  | cases |
| ## | 8  | Brazil         | 2000  | popul |
| ## | 9  | China          | 1999  | cases |
| ## | 10 | China          | 1999  | popul |
| ## | 11 | China          | 2000  | cases |
| ## | 12 | China          | 2000  | popul |

Is it tidy? —

```
## #      tibble: 12 x 4
##      country      year type
##      <chr>        <int> <chr>
##  1  fghanistan    1999 cases
##  2  fghanistan    1999 popul
##  3  fghanistan    2000 cases
##  4  fghanistan    2000 popul
##  5  Brazil        1999 cases
##  6  Brazil        1999 popul
##  7  Brazil        2000 cases
##  8  Brazil        2000 popul
##  9  China         1999 cases
## 10  China         1999 popul
## 11  China         2000 cases
## 12  China         2000 popul
```

“cases” and “population” should

Is it tidy?

```
## #      tibble: 6 x 3
##   country      year  ra
## * <chr>        <int> <c
## 1  fghanistan  1999  74
## 2  fghanistan  2000  26
## 3  Brazil      1999  37
## 4  Brazil      2000  80
## 5  China       1999  21
## 6  China       2000  21
```



Is it tidy? —

```
## #      tibble: 6 x 3
##   country      year rate
## * <chr>      <int> <dbl>
## 1 fghanistan  1999  74.0
## 2 fghanistan  2000  26.0
## 3 Brazil      1999  37.0
## 4 Brazil      2000  80.0
## 5 China       1999  21.0
## 6 China       2000  21.0
```

“rate” column encodes two variables

Is it tidy?

```
## #      tibble: 3 x 3
##   country      `1999` `
## * <chr>          <int>
## 1  fghanistan    745
## 2  Brazil        37737
## 3  China         212258 2

## #      tibble: 3 x 3
##   country      `199
## * <chr>          <in
## 1  fghanistan    199870
## 2  Brazil        1720063
## 3  China         12729152
```

Is it tidy? —

```
## #      tibble: 3 x 3
##   country      `1999` `
## * <chr>         <int>
## 1  fghanistan    745
## 2  Brazil        37737
## 3  China         212258 2
```

```
## #      tibble: 3 x 3
##   country      `199
## * <chr>         <in
## 1  fghanistan    199870
## 2  Brazil        1720063
## 3  China         12729152
```

observations in multiple tables; co

Is it tidy?

```
## #      tibble: 6 x 4
##   country      year  ca
##   <chr>        <int> <
## 1  fghanistan  1999
## 2  fghanistan  2000
## 3  Brazil      1999  3
## 4  Brazil      2000  8
## 5  China       1999  21
## 6  China       2000  21
```

Is it tidy? —

```
## #      tibble: 6 x 4
##   country      year  ca
##   <chr>        <int> <dbl>
## 1  fghanistan  1999    2
## 2  fghanistan  2000    2
## 3  Brazil      1999    3
## 4  Brazil      2000    8
## 5  China       1999   21
## 6  China       2000   21
```

## Tidying data

- Pre-requisite step to analyzing data
- Makes additional data collection easier
- Reshape the dataset into a tidy format
  - ◆ “Wider” — more columns
  - ◆ “Longer” — more rows
- Process improperly coded data

# Going “wide”

- Single observations (country-year) scatter plot
- Values of “key” column should be variable

| 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

# Going “long”


- Single variable (“cases”) spread across
- Column names are values (1999 and

| country     | year | cases  |             |
|-------------|------|--------|-------------|
| Afghanistan | 1999 | 745    | Afghanistan |
| Afghanistan | 2000 | 2666   | Brazil      |
| Brazil      | 1999 | 37737  | China       |
| Brazil      | 2000 | 80488  |             |
| China       | 1999 | 212258 |             |
| China       | 2000 | 213766 |             |



# Process improperly-co

- Single column (“rate”) encodes two vari
- Strings used to represent quantitative (n



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

table3

TIDYR

## Summary: “tidy”

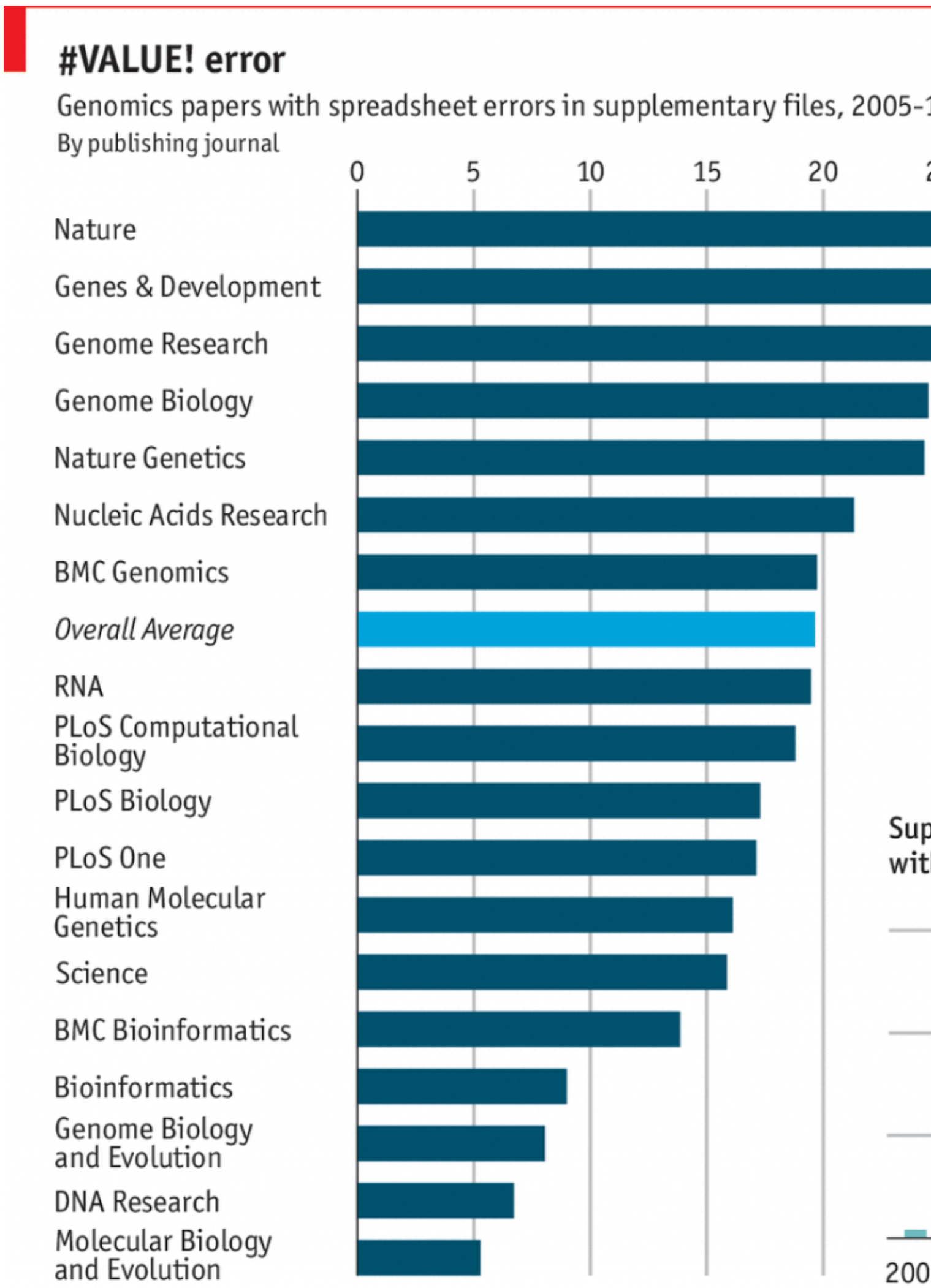
- Easy to query, transform
- Frames database normal in language of statistical

DATA CLEA

## Wrangling c

- Post-structuring, additional cleaning is often necessary on real data
- Consistency
  - ◆ Dates and strings often need to be standardized
  - ◆ Label levels of categorical variables
- Missing data
  - ◆ Check for patterns of missing data
  - ◆ To impute or not to impute

# Never trust a spreadsheet



# Strings

- Trim/pad white space
- Normalization and punctuation
  - ◆ Singular vs plural, verb forms, etc
- Capitalization/case-folding
  - ◆ Proper vs common nouns
- Special characters and encoding

## Dates and times

- Consistent input formats
  - ◆ MM/DD/YY vs DD-MM-YYYY
- Convert to appropriate format
- Consider time zones
- Be careful of your assumptions
  - ◆ Leap years vs leap seconds, oh yes
  - ◆ *Use a good library!!!*



## Missing data

- Why is the data missing
- What to do about it

## Types of missing

- Unit non-response
  - ◆ Entire rows of data are missing
  - ◆ Usually not directly observed in
  - ◆ Very dangerous — sampling bi
- Item non-response
  - ◆ Missing values/cells in a column
  - ◆ Can be directly inspected in th

## Patterns of missingness

- Missing Completely at Random
  - ◆ Missing data are non-systematic and
- Missing at Random (MAR)
  - ◆ Missing data are independent of their missingness are related to features of
- Missing Not at Random (MNAR)
  - ◆ Missing data are dependent on their

## Patterns of missing data

- Missing Completely at Random
  - ◆ Data missing, randomly
- Missing at Random (MAR)
  - ◆ Data from earlier years more likely to
- Missing Not at Random (MNAR)
  - ◆ Data values near zero more likely to

## Methods of imputation

- Do nothing
  - ◆ Easiest
  - ◆ Adequate for some visualization
  - ◆ Not always possible or appropriate
- Mean/median/mode imputation
  - ◆ Easy
  - ◆ Distorts data — underestimates variance
  - ◆ Appropriate for MCAR and MAR

## Methods of imputation

- Zero/constant imputation
  - ◆ Easy
  - ◆ Introduces bias to the data
  - ◆ Can be appropriate for certain situations
- Algorithmic/model-based
  - ◆ Difficult
  - ◆ Can be more accurate and less biased
  - ◆ Many methods to choose from

## Missing data: final

- Look for patterns of missing data
  - ◆ Understand why data is missing
- How does the missingness impact the analysis?
  - ◆ Does it introduce bias?
- Do you need to impute the missing data?
  - ◆ How does it impact the analysis if you do?
- Always report what you did

# DATA TRANSFORM SUMMARIZA



## Key tasks

- Select columns of interest
- Filter/subset data based on
- Order/rank rows based on
- Transform data and create
- Group and aggregate sum

DPLYR