

rbtl - Data wrangling with `tidyverse`

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Today

1. Part 1: Data types and vectors
 - Live Coding Exercise
2. Part 2: tidyr - long and wide formats
 - Live Coding Exercise
3. Part 3: dplyr - joining data
 - Live Coding Exercise
4. Homework Assignment 13
5. Programming Exercise

Learning Objectives

1. Learners can apply functions from the `tidyverse` (actually dplyr) R Package to join multiple data sets
2. Learners can apply functions from the `tidyverse` R Package to transform their data from a wide to a long format and vice versa

Part 1: Data types and vectors

Why care about data types?

via GIPHY

Example: survey data

```
1 survey_data_small  
  
# A tibble: 22 × 3  
  id   job      price_glass  
  <int> <chr>    <chr>  
1     1 Student    0  
2     2 Retired    0  
3     3 Other      0  
4     4 Employed   10  
5     5 Employed  See comment  
6     6 Student   05-Oct  
# ... with 16 more rows
```

Oh why won't you work?!

```
1 survey_data_small %>%
2   summarise(mean_price_glass = mean(price_glass))
```

```
# A tibble: 1 × 1
  mean_price_glass
  <dbl>
1 NA
```

Oh why won't you still work??!!

```
1 survey_data_small %>%
2   summarise(mean_price_glass = mean(price_glass, na.rm = TRUE))

# A tibble: 1 × 1
  mean_price_glass
  <dbl>
1       NA
```

Take a breath and look at your data

```
1 glimpse(survey_data_small)
```

Rows: 22

Columns: 3

```
$ id           <int> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, ...
$ job          <chr> "Student", "Retired", "Other", "Employed", ...
$ price_glass  <chr> "0", "0", "0", "10", "See comment", "05-Oct...
```

Very common data tidying step!

```
1 survey_data_small %>%
 2   mutate(price_glass_new = case_when(
 3     price_glass == "5 to 10" ~ "7.5",
 4     price_glass == "05-Oct" ~ "7.5",
 5     str_detect(price_glass, pattern = "20") == TRUE ~ "20",
 6     str_detect(price_glass, pattern = "See comment") == TRUE ~ NA_character_,
 7     TRUE ~ price_glass
 8   ))
```

Very common data tidying step!

```
# A tibble: 22 × 4
  id job      price_glass_new price_glass
  <int> <chr>    <chr>           <chr>
1     1 Student   0               0
2     2 Retired   0               0
3     3 Other     0               0
4     4 Employed  10              10
5     5 Employed  <NA>            See comment
6     6 Student   7.5             05-Oct
7     7 Student   0               0
8     8 Retired   0               0
9     9 Student   10              10
10    10 Employed 0               0
11    11 Employed 20              20 (2chf per person with 10 pe...
12    12 Student   10              10
```

Sumamrise? Argh!!!!

```
1 survey_data_small %>%
 2   mutate(price_glass_new = case_when(
 3     price_glass == "5 to 10" ~ "7.5",
 4     price_glass == "05-Oct" ~ "7.5",
 5     str_detect(price_glass, pattern = "20") == TRUE ~ "20",
 6     str_detect(price_glass, pattern = "See comment") == TRUE ~ NA_character_,
 7     TRUE ~ price_glass
 8   )) %>%
 9   summarise(mean_price_glass = mean(price_glass_new, na.rm = TRUE))
```

```
# A tibble: 1 × 1
  mean_price_glass
    <dbl>
1           NA
```

Always respect your data types!

```
1 survey_data_small %>%
 2   mutate(price_glass_new = case_when(
 3     price_glass == "5 to 10" ~ "7.5",
 4     price_glass == "05-Oct" ~ "7.5",
 5     str_detect(price_glass, pattern = "20") == TRUE ~ "20",
 6     str_detect(price_glass, pattern = "See comment") == TRUE ~ NA_character_,
 7     TRUE ~ price_glass
 8   )) %>%
 9   mutate(price_glass_new = as.numeric(price_glass_new)) %>%
10   summarise(mean_price_glass = mean(price_glass_new, na.rm = TRUE))
```

```
# A tibble: 1 × 1
  mean_price_glass
                <dbl>
1                 4.76
```

Live Coding Exercise

ae-13-data-wrangling-tidyr

1. Head over to the GitHub Organisation for the course.
2. Find the repo for week 13 that has your GitHub username.
3. Clone the repo with your username to the RStudio Cloud.
4. Open the file: **ae-13a-tidyr.qmd**
5. Use your Sticky Notes to let me know when you are ready.

Break One

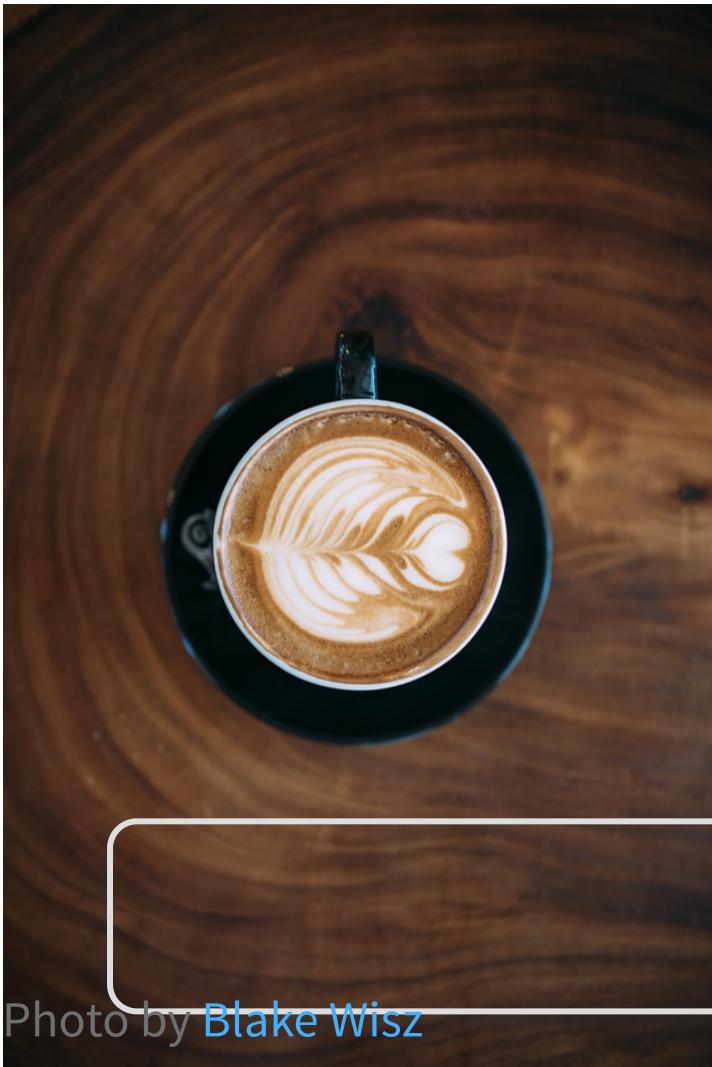


Photo by [Blake Wisz](#)

15:00

Part 2: tidyr - long and wide formats

TIDY DATA

is a standard way of mapping the meaning of a dataset to its structure.

-HADLEY WICKHAM

In tidy data:

- each variable forms a column
- each observation forms a row
- each cell is a single measurement

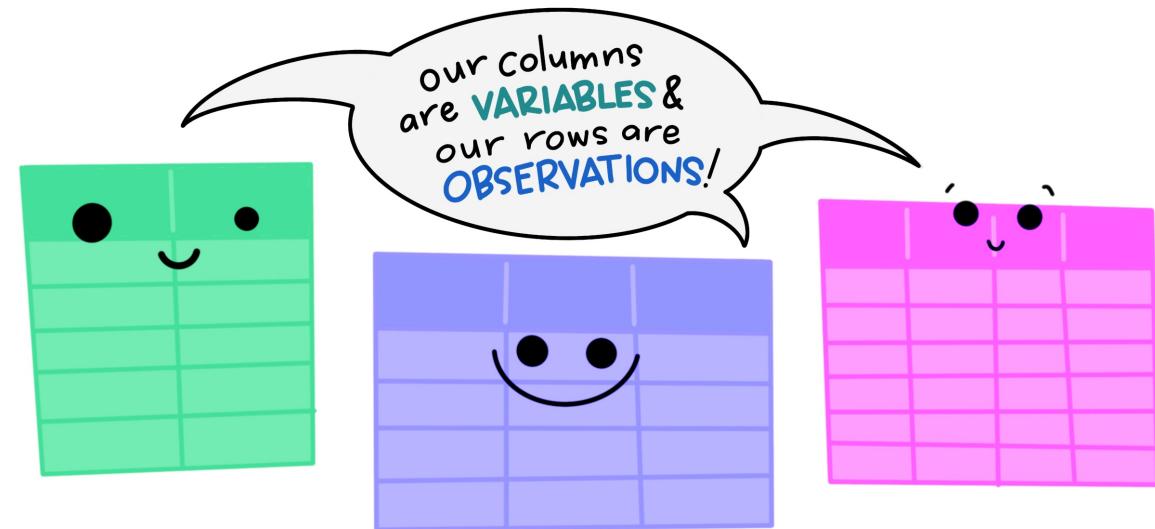
each column a variable

each row an observation

id	name	color
1	floof	gray
2	max	black
3	cat	orange
4	donut	gray
5	merlin	black
6	panda	calico

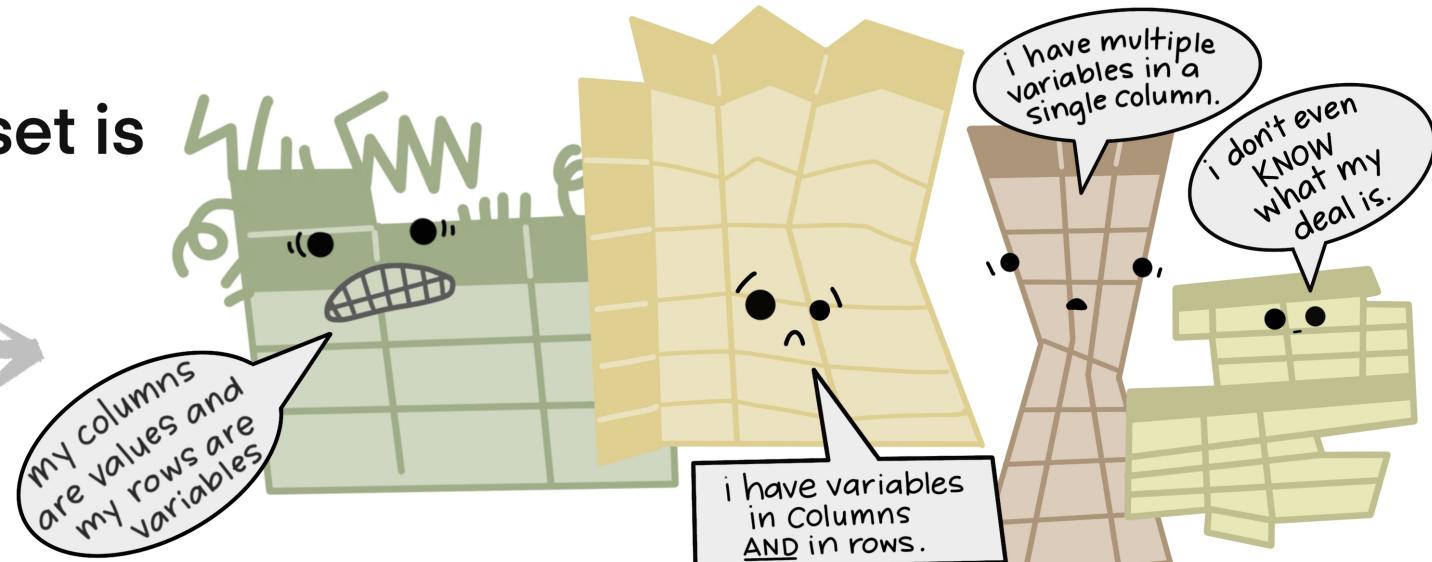
Illustrations from the [Openscapes](#) blog [Tidy Data for reproducibility, efficiency, and collaboration](#) by Wickham, H. (2014). Tidy Data. Journal of Statistical Software, 59(10). DOI: 10.18637/jss.v059.i10

The standard structure of
tidy data means that
“tidy datasets are all alike...”



“...but every messy dataset is
messy in its own way.”

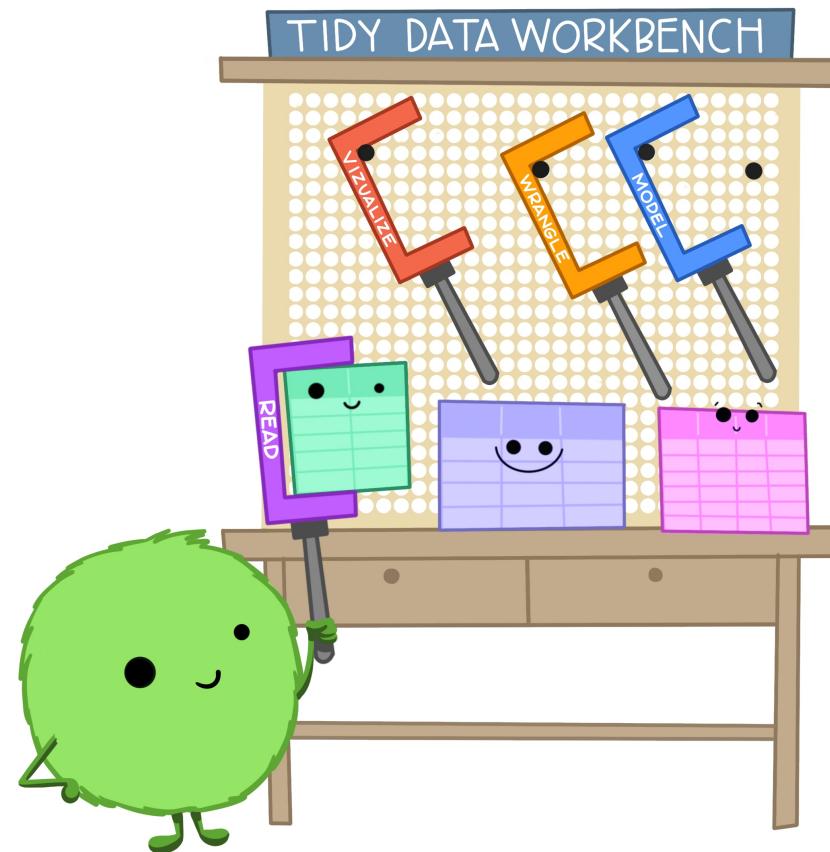
—HADLEY WICKHAM



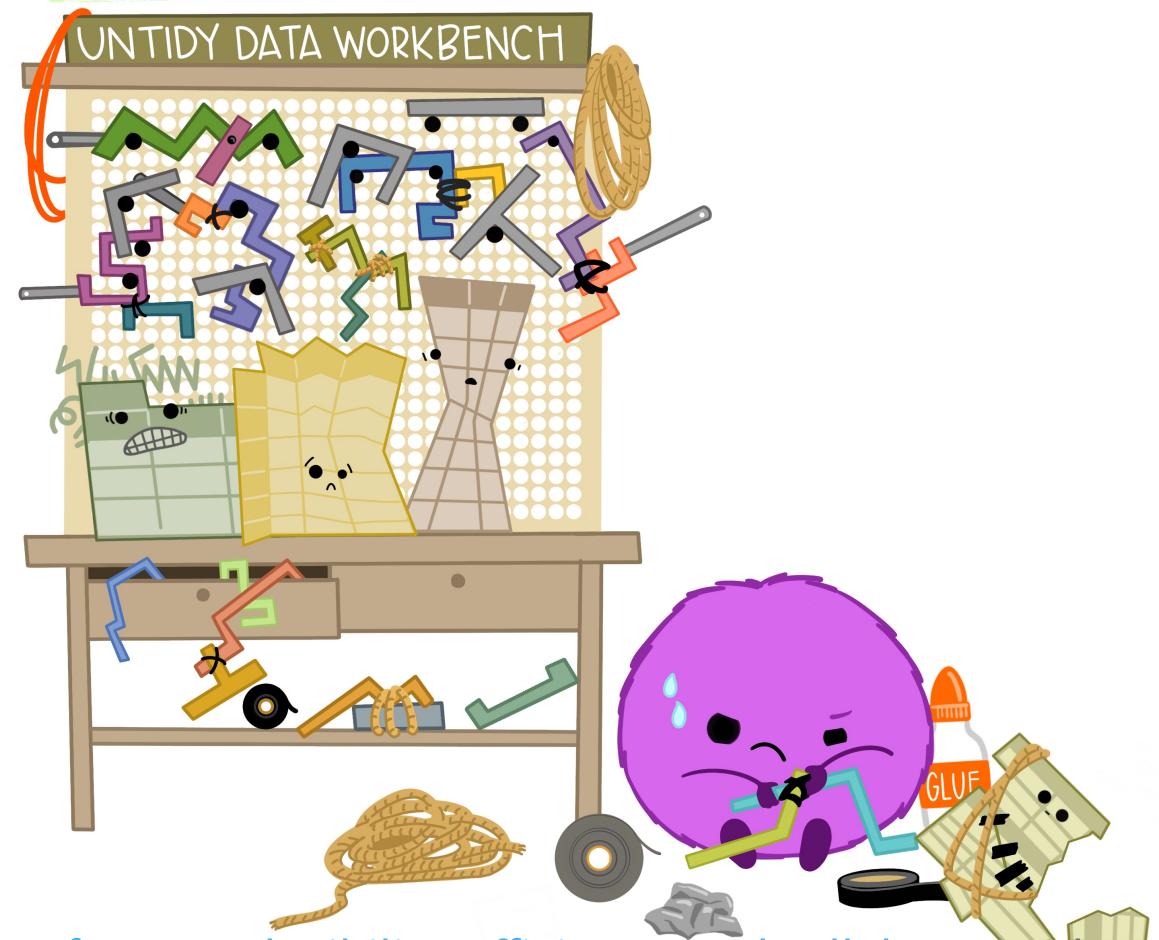
Illustrations from the [Openscapes](#) blog [Tidy Data for reproducibility, efficiency, and collaboration by](#)

-

When working with tidy data,
we can use the **same tools** in
similar ways for different datasets...

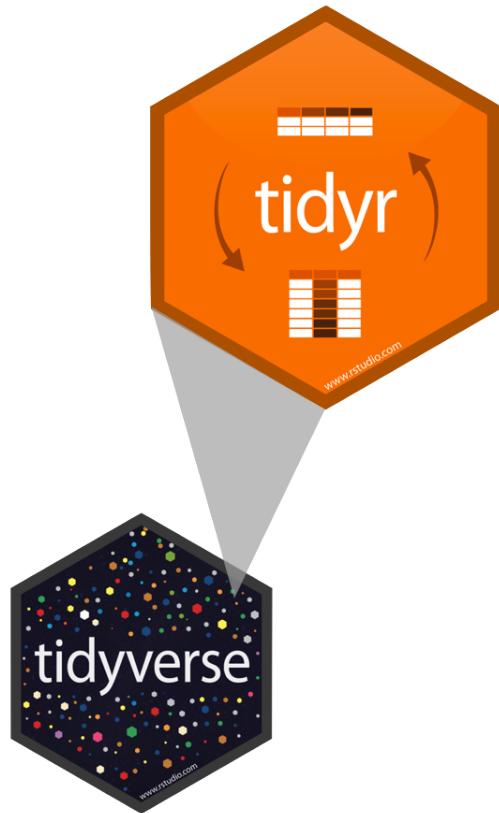


...but working with untidy data often means
reinventing the wheel with **one-time**
approaches that are **hard to iterate or reuse**.



Illustrations from the [Openscapes](#) blog [Tidy Data for reproducibility, efficiency, and collaboration](#) by

A grammar of data tidying



The goal of `tidyr` is to help you tidy your data via

- pivoting for going between wide and long data
- splitting and combining character columns
- nesting and unnesting columns
- clarifying how `NAs` should be treated

Pivoting data

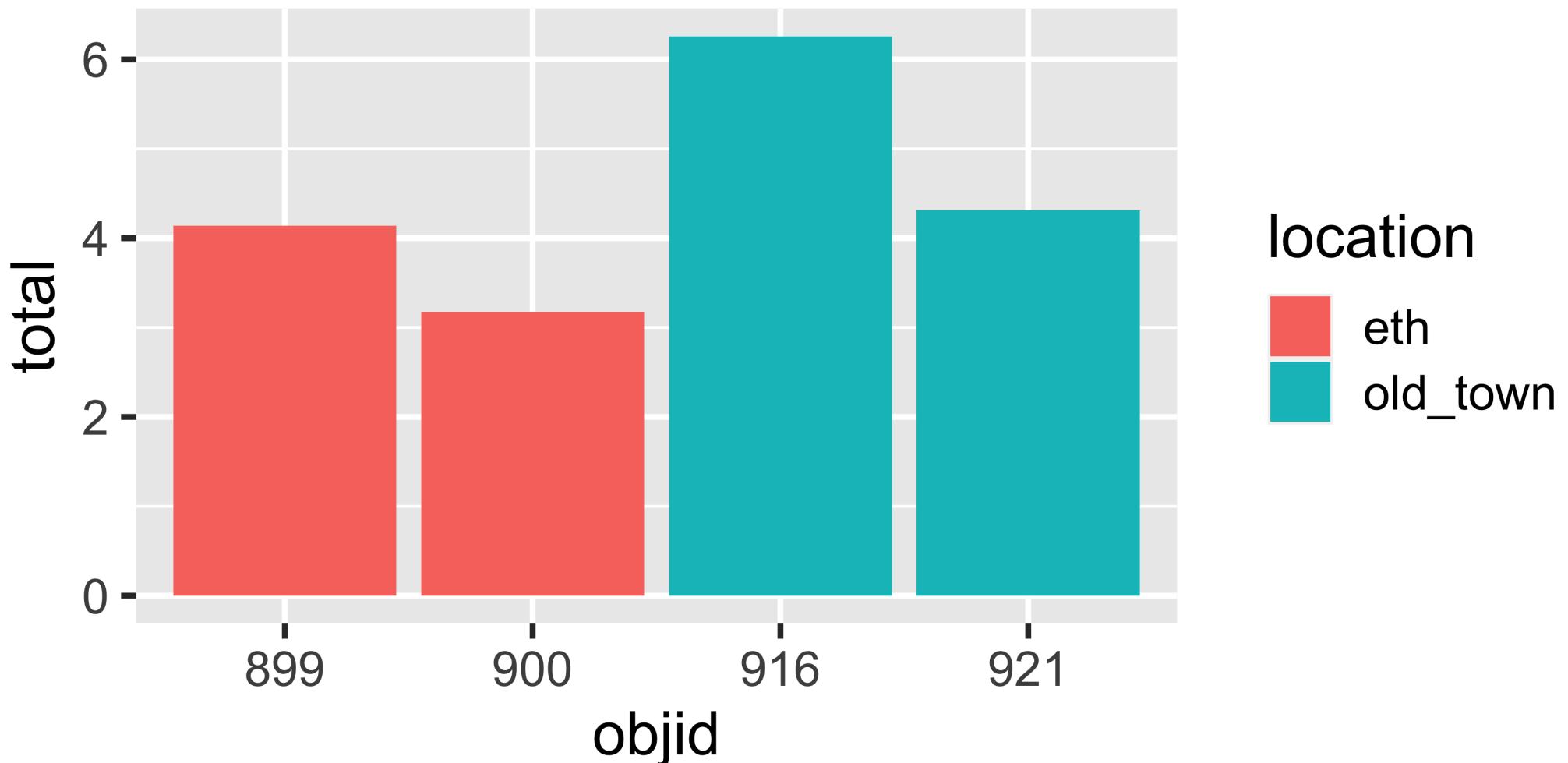
wide

id	x	y	z
1	a	c	e
2	b	d	f

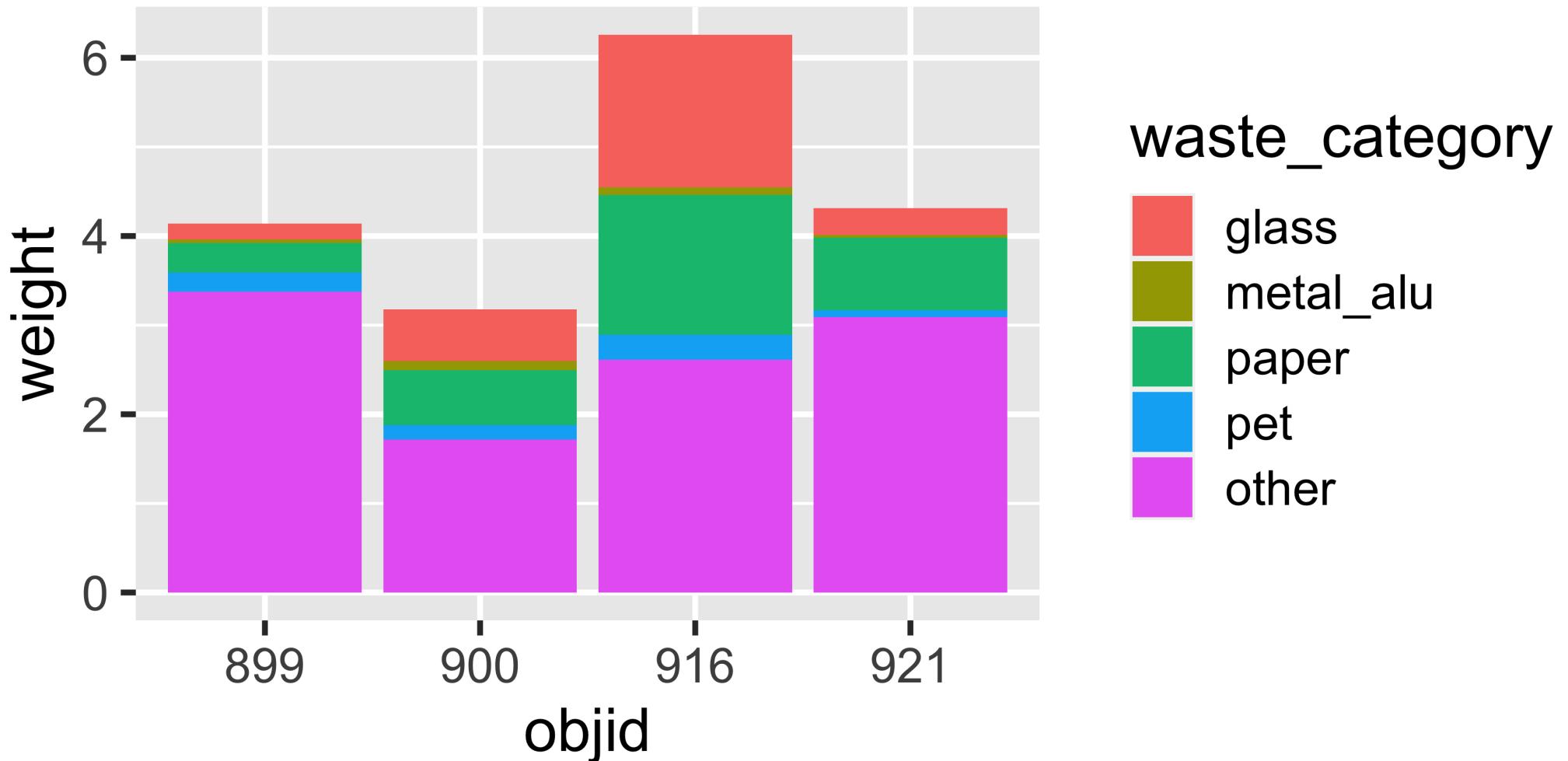
Waste characterisation data

objid	location	pet	metal_alu	glass	paper	recyclable	non_recyclable	total
900	eth	0.06	0.06	0.58	0.21	0.92	1.14	2.05
899	eth	0.14	0.01	0.18	0.28	0.61	3.04	3.64
921	old_town	0.00	0.00	0.00	0.41	0.41	1.57	1.99
916	old_town	0.17	0.04	0.80	0.55	1.56	0.62	2.19
900	eth	0.10	0.04	0.00	0.40	0.54	0.58	1.12
899	eth	0.08	0.03	0.00	0.05	0.16	0.34	0.50
921	old_town	0.08	0.03	0.30	0.40	0.81	1.52	2.33
916	old_town	0.11	0.04	0.92	1.01	2.08	1.99	4.07

How would you plot this?



And this?



You need: A long format

objid	location	waste_category	weight
900	eth	pet	0.06
900	eth	metal_alu	0.06
900	eth	glass	0.58
900	eth	paper	0.21
900	eth	other	1.14
899	eth	pet	0.14
899	eth	metal_alu	0.01
899	eth	glass	0.18
899	eth	paper	0.28
899	eth	other	3.04
921	old_town	pet	0.00
921	old_town	metal_alu	0.00
921	old_town	glass	0.00

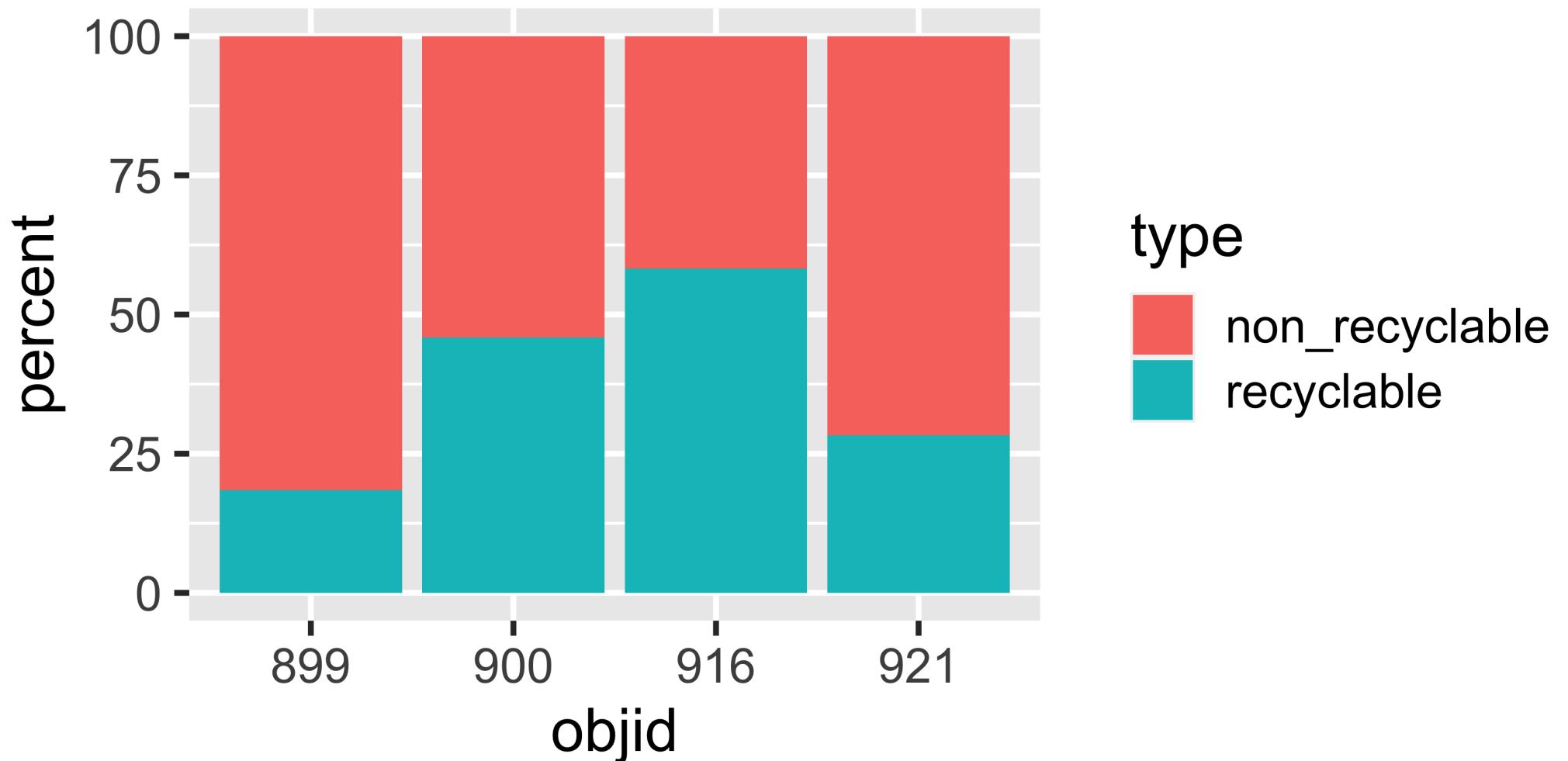
objid	location	waste_category	weight
921	old_town	paper	0.41
921	old_town	other	1.57
916	old_town	pet	0.17
916	old_town	metal_alu	0.04
916	old_town	glass	0.80
916	old_town	paper	0.55
916	old_town	other	0.62
900	eth	pet	0.10
900	eth	metal_alu	0.04
900	eth	glass	0.00
900	eth	paper	0.40
900	eth	other	0.58
899	eth	pet	0.08
899	eth	metal_alu	0.03
899	eth	glass	0.00
899	eth	paper	0.05

objid	location	waste_category	weight
899	eth	other	0.34
921	old_town	pet	0.08
921	old_town	metal_alu	0.03
921	old_town	glass	0.30
921	old_town	paper	0.40
921	old_town	other	1.52
916	old_town	pet	0.11
916	old_town	metal_alu	0.04
916	old_town	glass	0.92
916	old_town	paper	1.01
916	old_town	other	1.99

Reminder: The wide format

objid	location	pet	metal_alu	glass	paper	recyclable	non_recyclable	total
900	eth	0.06	0.06	0.58	0.21	0.92	1.14	2.05
899	eth	0.14	0.01	0.18	0.28	0.61	3.04	3.64
921	old_town	0.00	0.00	0.00	0.41	0.41	1.57	1.99
916	old_town	0.17	0.04	0.80	0.55	1.56	0.62	2.19
900	eth	0.10	0.04	0.00	0.40	0.54	0.58	1.12
899	eth	0.08	0.03	0.00	0.05	0.16	0.34	0.50
921	old_town	0.08	0.03	0.30	0.40	0.81	1.52	2.33
916	old_town	0.11	0.04	0.92	1.01	2.08	1.99	4.07

Or this?



Calculate percentages

objid	location	waste_category	type	weight	percent
900	eth	pet	recyclable	0.06	2.02
900	eth	metal_alu	recyclable	0.06	1.95
900	eth	glass	recyclable	0.58	18.14
900	eth	paper	recyclable	0.21	6.74
900	eth	other	non_recyclable	1.14	35.78
899	eth	pet	recyclable	0.14	3.33
899	eth	metal_alu	recyclable	0.01	0.31
899	eth	glass	recyclable	0.18	4.30
899	eth	paper	recyclable	0.28	6.69
899	eth	other	non_recyclable	3.04	73.36
921	old_town	pet	recyclable	0.00	0.00
921	old_town	metal_alu	recyclable	0.00	0.00
921	old_town	glass	recyclable	0.00	0.00

objid	location	waste_category	type	weight	percent
921	old_town	paper	recyclable	0.41	9.60
921	old_town	other	non_recyclable	1.57	36.46
916	old_town	pet	recyclable	0.17	2.76
916	old_town	metal_alu	recyclable	0.04	0.69
916	old_town	glass	recyclable	0.80	12.73
916	old_town	paper	recyclable	0.55	8.82
916	old_town	other	non_recyclable	0.62	9.99
900	eth	pet	recyclable	0.10	3.09
900	eth	metal_alu	recyclable	0.04	1.35
900	eth	glass	recyclable	0.00	0.00
900	eth	paper	recyclable	0.40	12.60
900	eth	other	non_recyclable	0.58	18.33
899	eth	pet	recyclable	0.08	1.86
899	eth	metal_alu	recyclable	0.03	0.72
899	eth	glass	recyclable	0.00	0.00
899	eth	paper	recyclable	0.05	1.26

objid	location	waste_category	type	weight	percent
899	eth	other	non_recyclable	0.34	8.16
921	old_town	pet	recyclable	0.08	1.81
921	old_town	metal_alu	recyclable	0.03	0.70
921	old_town	glass	recyclable	0.30	6.89
921	old_town	paper	recyclable	0.40	9.32
921	old_town	other	non_recyclable	1.52	35.21
916	old_town	pet	recyclable	0.11	1.74
916	old_town	metal_alu	recyclable	0.04	0.70
916	old_town	glass	recyclable	0.92	14.63
916	old_town	paper	recyclable	1.01	16.20
916	old_town	other	non_recyclable	1.99	31.73

How to

```
1 waste_data_untidy
```

objid	location	pet	metal_alu	glass	paper	recyclable	non_recyclable	total
900	eth	0.06	0.06	0.58	0.21	0.92	1.14	2.05
899	eth	0.14	0.01	0.18	0.28	0.61	3.04	3.64
921	old_town	0.00	0.00	0.00	0.41	0.41	1.57	1.99
916	old_town	0.17	0.04	0.80	0.55	1.56	0.62	2.19
900	eth	0.10	0.04	0.00	0.40	0.54	0.58	1.12
899	eth	0.08	0.03	0.00	0.05	0.16	0.34	0.50
921	old_town	0.08	0.03	0.30	0.40	0.81	1.52	2.33
916	old_town	0.11	0.04	0.92	1.01	2.08	1.99	4.07

How to

```
1 waste_data_untidy %>%
2   select(objid:paper, non_recyclable)
```

objid	location	pet	metal_alu	glass	paper	non_recyclable
900	eth	0.06	0.06	0.58	0.21	1.14
899	eth	0.14	0.01	0.18	0.28	3.04
921	old_town	0.00	0.00	0.00	0.41	1.57
916	old_town	0.17	0.04	0.80	0.55	0.62
900	eth	0.10	0.04	0.00	0.40	0.58
899	eth	0.08	0.03	0.00	0.05	0.34
921	old_town	0.08	0.03	0.30	0.40	1.52
916	old_town	0.11	0.04	0.92	1.01	1.99

How to

```
1 waste_data_untidy %>%
2   select(objid:paper, non_recyclable) %>%
3   rename(other = non_recyclable)
```

objid	location	pet	metal_alu	glass	paper	other
900	eth	0.06	0.06	0.58	0.21	1.14
899	eth	0.14	0.01	0.18	0.28	3.04
921	old_town	0.00	0.00	0.00	0.41	1.57
916	old_town	0.17	0.04	0.80	0.55	0.62
900	eth	0.10	0.04	0.00	0.40	0.58
899	eth	0.08	0.03	0.00	0.05	0.34
921	old_town	0.08	0.03	0.30	0.40	1.52
916	old_town	0.11	0.04	0.92	1.01	1.99

How to

```
1 waste_category_levels <- c("glass", "metal_alu", "paper", "pet", "other"
2
3 waste_data_untidy %>%
4   select(objid:paper, non_recyclable) %>%
5   rename(other = non_recyclable) %>%
6   pivot_longer(cols = pet:other,
7                 names_to = "waste_category",
8                 values_to = "weight") %>%
9   mutate(waste_category = factor(waste_category,
10                     levels = waste_category_levels))
```

objid	location	waste_category	weight
900	eth	pet	0.06
900	eth	metal_alu	0.06
900	eth	glass	0.58
900	eth	paper	0.21
900	eth	other	1.14
899	eth	pet	0.14
899	eth	metal_alu	0.01

objid	location	waste_category	weight
899	eth	glass	0.18
899	eth	paper	0.28
899	eth	other	3.04
921	old_town	pet	0.00
921	old_town	metal_alu	0.00
921	old_town	glass	0.00
921	old_town	paper	0.41
921	old_town	other	1.57
916	old_town	pet	0.17
916	old_town	metal_alu	0.04
916	old_town	glass	0.80
916	old_town	paper	0.55
916	old_town	other	0.62
900	eth	pet	0.10
900	eth	metal_alu	0.04
900	eth	glass	0.00

objid	location	waste_category	weight
900	eth	paper	0.40
900	eth	other	0.58
899	eth	pet	0.08
899	eth	metal_alu	0.03
899	eth	glass	0.00
899	eth	paper	0.05
899	eth	other	0.34
921	old_town	pet	0.08
921	old_town	metal_alu	0.03
921	old_town	glass	0.30
921	old_town	paper	0.40
921	old_town	other	1.52
916	old_town	pet	0.11
916	old_town	metal_alu	0.04
916	old_town	glass	0.92
916	old_town	paper	1.01

How to

```
1 waste_category_levels <- c("glass", "metal_alu", "paper", "pet", "other"
2
3 waste_data_untidy %>%
4   select(objid:paper, non_recyclable) %>%
5   rename(other = non_recyclable) %>%
6   pivot_longer(cols = pet:other,
7                 names_to = "waste_category",
8                 values_to = "weight") %>%
9   mutate(waste_category = factor(waste_category,
10                      levels = waste_category_levels)) %>%
11  mutate(type = case_when(
12    waste_category == "other" ~ "non_recyclable",
13    TRUE ~ "recyclable")) %>%
14  relocate(type, .before = weight)
```

objid	location	waste_category	type	weight
900	eth	pet	recyclable	0.06
900	eth	metal_alu	recyclable	0.06
900	eth	glass	recyclable	0.58
900	eth	paper	recyclable	0.21

objid	location	waste_category	type	weight
900	eth	other	non_recyclable	1.14
899	eth	pet	recyclable	0.14
899	eth	metal_alu	recyclable	0.01
899	eth	glass	recyclable	0.18
899	eth	paper	recyclable	0.28
899	eth	other	non_recyclable	3.04
921	old_town	pet	recyclable	0.00
921	old_town	metal_alu	recyclable	0.00
921	old_town	glass	recyclable	0.00
921	old_town	paper	recyclable	0.41
921	old_town	other	non_recyclable	1.57
916	old_town	pet	recyclable	0.17
916	old_town	metal_alu	recyclable	0.04
916	old_town	glass	recyclable	0.80
916	old_town	paper	recyclable	0.55
916	old_town	other	non_recyclable	0.62

objid	location	waste_category	type	weight
900	eth	pet	recyclable	0.10
900	eth	metal_alu	recyclable	0.04
900	eth	glass	recyclable	0.00
900	eth	paper	recyclable	0.40
900	eth	other	non_recyclable	0.58
899	eth	pet	recyclable	0.08
899	eth	metal_alu	recyclable	0.03
899	eth	glass	recyclable	0.00
899	eth	paper	recyclable	0.05
899	eth	other	non_recyclable	0.34
921	old_town	pet	recyclable	0.08
921	old_town	metal_alu	recyclable	0.03
921	old_town	glass	recyclable	0.30
921	old_town	paper	recyclable	0.40
921	old_town	other	non_recyclable	1.52
916	old_town	pet	recyclable	0.11

How to

```
1 waste_category_levels <- c("glass", "metal_alu", "paper", "pet", "other")
2
3 waste_data_untidy %>%
4   select(objid:paper, non_recyclable) %>%
5   rename(other = non_recyclable) %>%
6   pivot_longer(cols = pet:other,
7                 names_to = "waste_category",
8                 values_to = "weight") %>%
9   mutate(waste_category = factor(waste_category,
10                     levels = waste_category_levels)) %>%
11  mutate(type = case_when(
12    waste_category == "other" ~ "non_recyclable",
13    TRUE ~ "recyclable")) %>%
14  relocate(type, .before = weight) %>%
15  group_by(objid) %>%
16  mutate(percent = weight / sum(weight) * 100)
```

objid	location	waste_category	type	weight	percent
900	eth	pet	recyclable	0.06	2.02
900	eth	metal_alu	recyclable	0.06	1.95
900	eth	glass	recyclable	0.58	18.14

objid	location	waste_category	type	weight	percent
900	eth	paper	recyclable	0.21	6.74
900	eth	other	non_recyclable	1.14	35.78
899	eth	pet	recyclable	0.14	3.33
899	eth	metal_alu	recyclable	0.01	0.31
899	eth	glass	recyclable	0.18	4.30
899	eth	paper	recyclable	0.28	6.69
899	eth	other	non_recyclable	3.04	73.36
921	old_town	pet	recyclable	0.00	0.00
921	old_town	metal_alu	recyclable	0.00	0.00
921	old_town	glass	recyclable	0.00	0.00
921	old_town	paper	recyclable	0.41	9.60
921	old_town	other	non_recyclable	1.57	36.46
916	old_town	pet	recyclable	0.17	2.76
916	old_town	metal_alu	recyclable	0.04	0.69
916	old_town	glass	recyclable	0.80	12.73
916	old_town	paper	recyclable	0.55	8.82

objid	location	waste_category	type	weight	percent
916	old_town	other	non_recyclable	0.62	9.99
900	eth	pet	recyclable	0.10	3.09
900	eth	metal_alu	recyclable	0.04	1.35
900	eth	glass	recyclable	0.00	0.00
900	eth	paper	recyclable	0.40	12.60
900	eth	other	non_recyclable	0.58	18.33
899	eth	pet	recyclable	0.08	1.86
899	eth	metal_alu	recyclable	0.03	0.72
899	eth	glass	recyclable	0.00	0.00
899	eth	paper	recyclable	0.05	1.26
899	eth	other	non_recyclable	0.34	8.16
921	old_town	pet	recyclable	0.08	1.81
921	old_town	metal_alu	recyclable	0.03	0.70
921	old_town	glass	recyclable	0.30	6.89
921	old_town	paper	recyclable	0.40	9.32
921	old_town	other	non_recyclable	1.52	35.21

objid	location	waste_category	type	weight	percent
916	old_town	pet	recyclable	0.11	1.74
916	old_town	metal_alu	recyclable	0.04	0.70
916	old_town	glass	recyclable	0.92	14.63
916	old_town	paper	recyclable	1.01	16.20
916	old_town	other	non_recyclable	1.99	31.73

Live Coding Exercise

ae-13-data-wrangling-tidyr

1. Back to [ae-13a-tidyr.qmd](#)

Break Two

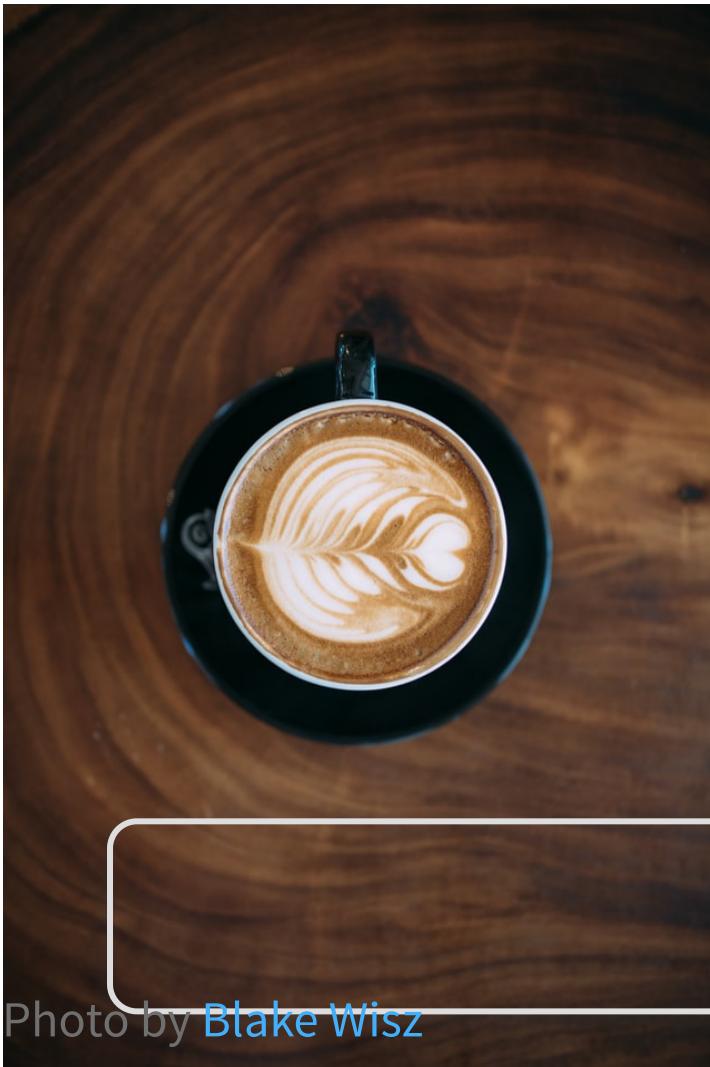


Photo by [Blake Wisz](#)

10 : 00

Part 3: dplyr - joining data

We...

...have multiple data frames

...want to bring them together

```
1 professions <- read_csv(here::here("data/scientists/professions.csv"))
2 dates <- read_csv(here::here("data/scientists/dates.csv"))
3 works <- read_csv(here::here("scientists/works.csv"))
```

Data: Women in science

Information on 10 women in science who changed the world

name

Ada Lovelace

Marie Curie

Janaki Ammal

Chien-Shiung Wu

Katherine Johnson

Rosalind Franklin

Vera Rubin

Gladys West

Flossie Wong-Staal

Jennifer Doudna

Inputs

professions

dates

works

name	profession
Ada Lovelace	Mathematician
Marie Curie	Physicist and Chemist
Janaki Ammal	Botanist
Chien-Shiung Wu	Physicist
Katherine Johnson	Mathematician
Rosalind Franklin	Chemist
Vera Rubin	Astronomer
Gladys West	Mathematician
Flossie Wong-Staal	Virologist and Molecular Biologist
Jennifer Doudna	Biochemist

Desired output

name	profession	birth_year	death_year	known_for
Ada Lovelace	Mathematician	NA	NA	first computer algorithm
Marie Curie	Physicist and Chemist	NA	NA	theory of radioactivity, discovery of elements polonium and radium, first woman to win a Nobel Prize
Janaki Ammal	Botanist	1897	1984	hybrid species, biodiversity protection
Chien-Shiung Wu	Physicist	1912	1997	confirm and refine theory of radioactive beta decay, Wu experiment overturning theory of parity
Katherine Johnson	Mathematician	1918	2020	calculations of orbital mechanics critical to sending the first Americans into space
Rosalind Franklin	Chemist	1920	1958	NA

name	profession	birth_year	death_year	known_for
Vera Rubin	Astronomer	1928	2016	existence of dark matter
Gladys West	Mathematician	1930	NA	mathematical modeling of the shape of the Earth which served as the foundation of GPS technology
Flossie Wong-Staal	Virologist and Molecular Biologist	1947	NA	first scientist to clone HIV and create a map of its genes which led to a test for the virus
Jennifer Doudna	Biochemist	1964	NA	one of the primary developers of CRISPR, a ground-breaking technology for editing genomes

Inputs, reminder

```
1 names(professions)
```

```
[1] "name"      "profession"
```

```
1 names(dates)
```

```
[1] "name"      "birth_year"  
"death_year"
```

```
1 names(works)
```

```
[1] "name"      "known_for"
```

```
1 nrow(professions)
```

```
[1] 10
```

```
1 nrow(dates)
```

```
[1] 8
```

```
1 nrow(works)
```

```
[1] 9
```

Joining data frames

Joining data frames

```
1 something_join(x, y)
```

- `left_join()`: all rows from x
- `right_join()`: all rows from y
- `full_join()`: all rows from both x and y
- ...

Setup

For the next few slides...

```
1 x <- tibble(  
2   id = c(1, 2, 3),  
3   value_x = c("x1", "x2", "x3")  
4 )
```

```
1 x
```

```
# A tibble: 3 × 2  
  id    value_x  
  <dbl> <chr>  
1     1  x1  
2     2  x2  
3     3  x3
```

```
1 y <- tibble(  
2   id = c(1, 2, 4),  
3   value_y = c("y1", "y2", "y4")  
4 )
```

```
1 y
```

```
# A tibble: 3 × 2  
  id    value_y  
  <dbl> <chr>  
1     1  y1  
2     2  y2  
3     4  y4
```

left_join()

```
left_join(x, y)
```

1	x1
2	x2
3	x3

1	y1
2	y2
4	y4

```
1 left_join(x, y)
```

```
# A tibble: 3 × 3
  id value_x value_y
  <dbl> <chr>   <chr>
1     1 x1      y1
2     2 x2      y2
3     3 x3      <NA>
```

left_join()

```
1 professions %>%
2   left_join(dates)

# A tibble: 10 × 4
  name      profession    birth_year death_year
  <chr>     <chr>          <dbl>        <dbl>
1 Ada Lovelace Mathematician      NA          NA
2 Marie Curie  Physicist and Chemist  NA          NA
3 Janaki Ammal Botanist           1897         1984
4 Chien-Shiung Wu Physicist        1912         1997
5 Katherine Johnson Mathematician 1918         2020
6 Rosalind Franklin Chemist       1920         1958
# ... with 4 more rows
```

right_join()

right_join(x, y)

1	x1
2	x2
3	x3

1	y1
2	y2
4	y4

```
1 right_join(x, y)
```

```
# A tibble: 3 × 3
  id value_x value_y
  <dbl> <chr>   <chr>
1     1 x1      y1
2     2 x2      y2
3     4 <NA>    y4
```

right_join()

```
1 professions %>%
  2   right_join(dates)

# A tibble: 8 × 4
  name      profession birth_year death_year
  <chr>     <chr>        <dbl>       <dbl>
1 Janaki Ammal Botanist      1897       1984
2 Chien-Shiung Wu Physicist    1912       1997
3 Katherine Johnson Mathematician 1918       2020
4 Rosalind Franklin Chemist     1920       1958
5 Vera Rubin Astronomer       1928       2016
6 Gladys West Mathematician    1930        NA
# ... with 2 more rows
```

full_join()

full_join(x, y)

1	x1
2	x2
3	x3

1	y1
2	y2
4	y4

```
1 full_join(x, y)
```

```
# A tibble: 4 × 3
  id value_x value_y
  <dbl> <chr>   <chr>
1     1 x1      y1
2     2 x2      y2
3     3 x3      <NA>
4     4 <NA>    y4
```

full_join()

```
1 dates %>%
2   full_join(works)

# A tibble: 10 × 4
  name      birth_year death_year known_for
  <chr>        <dbl>     <dbl> <chr>
1 Janaki Ammal      1897      1984 hybrid species, biodiv...
2 Chien-Shiung Wu    1912      1997 confim and refine theo...
3 Katherine Johnson  1918      2020 calculations of orbita...
4 Rosalind Franklin  1920      1958 <NA>
5 Vera Rubin         1928      2016 existence of dark matt...
6 Gladys West        1930          NA mathematical modeling ...
# ... with 4 more rows
```

Putting it altogether

```
1 professions %>%
2   left_join(dates) %>%
3   left_join(works)

# A tibble: 10 × 5
  name      profession birth_year death_year known_for
  <chr>     <chr>        <dbl>       <dbl> <chr>
1 Ada Lovelace Mathematician      NA          NA  first co...
2 Marie Curie  Physicist       an...      NA          NA  theory o...
3 Janaki Ammal Botanist         1897        1984 hybrid s...
4 Chien-Shiung Wu Physicist       1912        1997 confirm a...
5 Katherine Johnson Mathematician 1918        2020 calculat...
6 Rosalind Franklin Chemist       1920        1958 <NA>
# ... with 4 more rows
```

Live Coding Exercise

ae-13-data-wrangling-tidyr

1. Back to [ae-13a-tidyr.qmd](#)

Homework Assignment

Submission

- All details in assignment week 13
- Due: Wednesday, 26th May at 23:59 (2 points)

Evaluation

- 5 mins
- anonymous
- after each lecture

kutt.it/rbtl-eval

Programming

ae-13-data-wrangling-tidyr

1. Open the file: **ae-13b-dplyr.qmd**
2. Work through the exercises
3. Finalise as part of your homework

Thanks! 🌻

A large proportion of slides in this presentation are either taken from or adapted from [Data Science in a Box](#)]

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