# Joining, mapping, and reshaping data



#### Overview

Joining data frames continued

Creating maps

Reshaping

If there is time: string manipulation

#### Announcement

Homework 9 late date is set for Sunday November 28<sup>th</sup>

• i.e., the Sunday at the end of the Thanksgiving break

Regrade requests for older homework are open again until Sunday the 28<sup>th</sup> at 11pm

They will not be opened again!

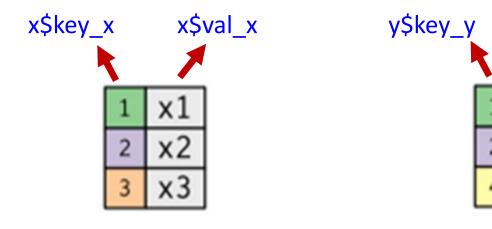
Keep thinking about your final project!

# Joining data frames

## Left and right tables

Suppose we have two data frames called x and y

- x have two variables called key\_x, and val\_x
- y has two variables called key\_y and val\_y



Data frame x

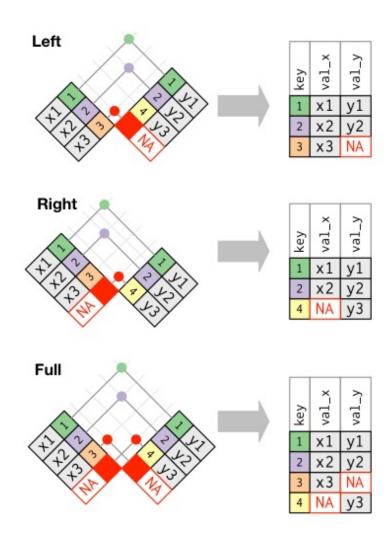
Data frame y

y\$val\_y

Joins have the general form:

$$join(x, y, by = c("key_x" = "key_y"))$$

#### Joining data frames



**Left joins** keep all rows in the <u>left</u> table.

$$left_join(x, y, by = c("key_x" = "key_y"))$$

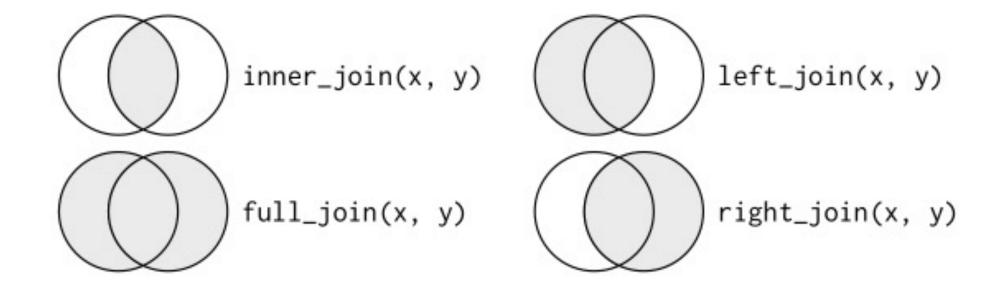
**Right joins** keep all rows in the <u>right</u> table.

right\_join(x, y, by = 
$$c("key_x" = "key_y"))$$

Full joins keep all rows in both tables.

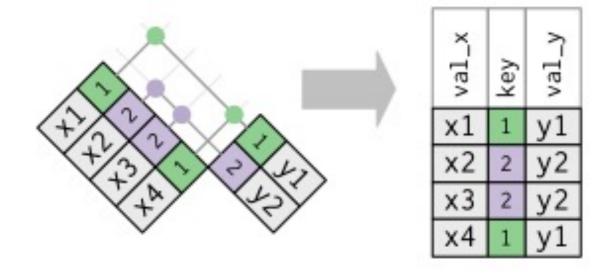
full\_join(x, y, by = 
$$c("key_x" = "key_y"))$$

### Summary



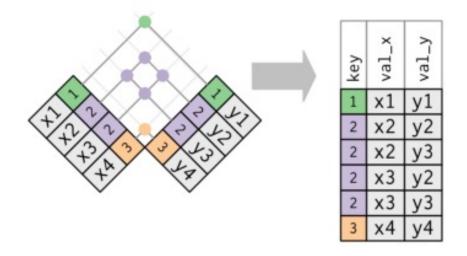
Duplicate keys are useful if there is a many-to-one relationship

• e.g., duplicates are useful in the left table when doing a left join



If both tables have duplicate keys you get all possible combinations of joined values (Cartesian product).

This is usually an error!



Always check the output size after you join a table because even if there is not a syntax error you might not get the table you are expecting!

You can check how many rows a data frame has using the nrow() function

To deal with duplicate keys in both tables, we can join the tables using multiple keys in order to make sure that each row is uniquely specified.

We can do this using the syntax:

```
join(x2, y2, by = c("key1_x" = "key1_y", "key2_x" = "key2_y"))
```

```
> x2 <- data.frame(key1_x = c(1, 2, 2),
          key2 x = c("a", "a", "b"),
          val x = c("y1", "y2", "y3"))
> y2 <- y2 <- data.frame(key1 y = c(1, 2, 2, 3, 3),
          key2 y = c("a", "a", "b", "a", "b"),
          val y = c("y1", "y2", "y3", "y4", "y5"))
> left join(x2, y2, c("key1 x" = "key1 y"))
> left join(x2, y2, c("key1 x" = "key1 y", "key2 x" = "key2 y"))
```

### Structured Query Language

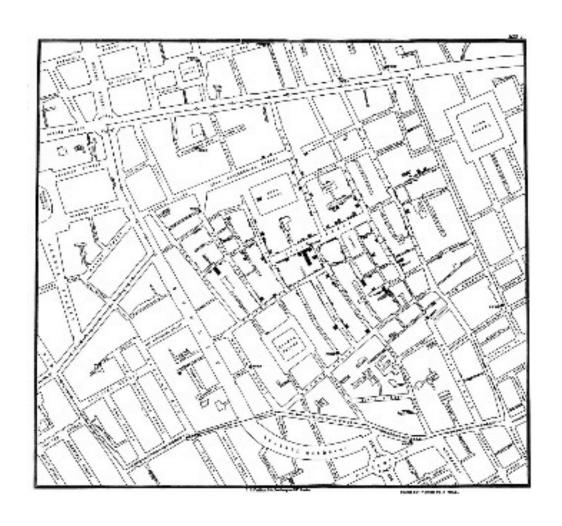
Having multiple tables that can be joined together is common in Relational Database Systems (RDBS).

A common language used by RDBS is Structured Query Language (SQL)

plyr	SQL		
inner_join(x, y, by = "z")	SELECT * FROM x INNER JOIN y USING (z)		
left_join(x, y, by = "z")	SELECT * FROM x LEFT OUTER JOIN y USING (z)		
right_join(x, y, by = "z")	SELECT * FROM x RIGHT OUTER JOIN y USING (z)		
full_join(x, y, by = "z")	SELECT * FROM x FULL OUTER JOIN y USING (z)		

Let's try it in R...

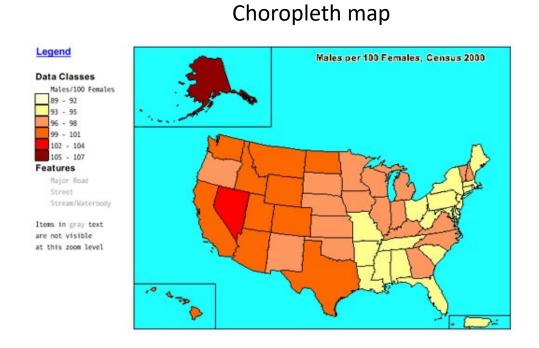
# Spatial mapping



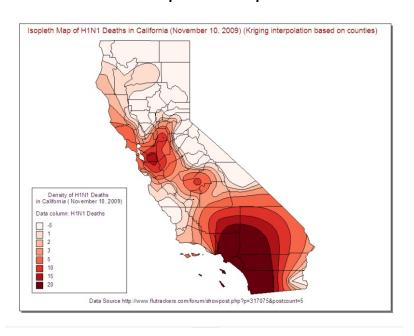
#### Maps

**Choropleth maps**: shades/colors in predefined areas based on properties of a variable

**Isopleth maps**: creates regions based on constant values



#### Isopleth map



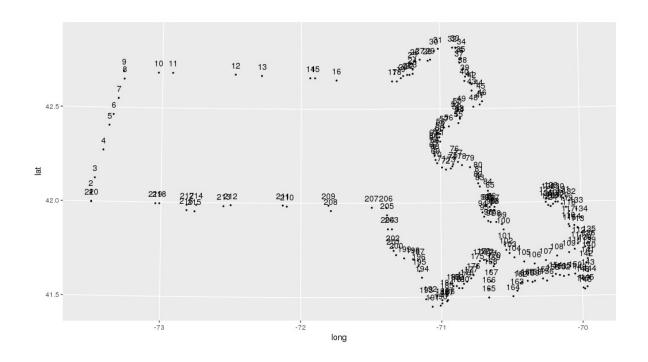
### Choropleth maps

- # has the coordinates for several maps
- > library('maps')
- # get a data frame with coordinates of states
- > states\_map <- map\_data("state")

_	long <sup>‡</sup>	lat <sup>‡</sup>	group <sup>‡</sup>	order <sup>‡</sup>	region 🗦	subregion <sup>‡</sup>
1	-87.46201	30.38968	1	1	alabama	NA
2	-87.48493	30.37249	1	2	alabama	NA
3	-87.52503	30.37249	1	3	alabama	NA
4	-87.53076	30.33239	1	4	alabama	NA
5	-87.57087	30.32665	1	5	alabama	NA

## Choropleth maps

geom\_polygon() works by connecting the dots:



Often need to arrange points first: <a href="mailto:arrange">arrange</a>(states\_map, group, order)

#### Choropleth maps

```
# has the coordinates for several maps
> library('maps')
# get a data frame with coordinates of states
> states_map <- map_data("state")
# filled white states with black borders
> ggplot(states map,
         aes(x = long, y = lat, group = group)) +
         geom polygon(fill = "white", color = "black")
```

Let's try it in R!



Reshaping data

#### Wide vs. Long data

Plotting data using ggplot requires that data is in the right format

• i.e., requires data transformations.

Often this involves converting data from a wide format to long format

#### Wide data

Person	Age	Height
Bob	32	72
Alice	24	65
Steve	64	70

#### Narrow data

Person	name	value
Bob	Age	32
Bob	Height	72
Alice	Age	24
Alice	Height	65
Steve	Age	64
Steve	Height	70

library(tidyr)

#### tidyr::pivot\_longer()

#### pivot\_longer(df, cols) converts data from wide to long

- Takes multiple columns and converts them into two columns: name and value
  - Column names become categorical variable levels of a new variable called name
  - The data in rows become entries in a variable called value

#### Long data

name

value

70

#### Wide data

Person	Age	Height
Bob	32	72
Alice	24	65
Steve	64	70



Height

Person

Steve

## tidyr::pivot\_wider()

pivot\_wider(df, names\_from, values\_from) converts data from narrow to wide

• Turns the levels of categorical data into columns in a data frame

#### **Narrow data**

person	name	value
Bob	Age	32
Bob	Height	72
Alice	Age	24
Alice	Height	65
Steve	Age	64
Steve	Height	70

#### Wide data

Person	Age	Height
Bob	32	72
Alice	24	65
Steve	64	70

Let's try it in R!