

## Agenda

1. Review the five verbs of data analysis [`tidyverse`]
2. Introduce joins
3. Continue applying this language to study police stops

# Class 4: Grammar of data analysis - Part II

API-201

Quantitative Analysis and Empirical Methods I

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## Five verbs of data manipulation

- `filter`
- `select`
- `mutate`
- `summarize`
- `arrange`

## The group\_by operation

The five verbs of data manipulation are particularly powerful when we combine them with the group\_by operation.

### Select the two appropriate commands

You have a data frame of grant recipients. Each row contains the grantee's name, the funding area (e.g., education, health, climate, etc.), and funding amount.

You would like to compute the total amount of funding distributed to each area.

**Which two commands would you use?**

filter and summarize

group\_by and mutate

group\_by and summarize

select and summarize



## Joining datasets

<https://r4ds.hadley.nz/joins.html>

In many cases, the data we want to analyze is split across multiple data frames.

For example, police stops may be stored in one table but information about the involved officers may be stored in another.

**Joins** let us combine columns from multiple data frames

## Joining datasets [ *inner join* ]

<https://r4ds.hadley.nz/joins.html>

`inner_join(x, y)`

x1	1	
x2	2	
x3	3	

4	2	1
y3	y2	y1



key	x_val	y_val
1	x1	y1
2	x2	y2

## Joining datasets [ *left join* ]

<https://r4ds.hadley.nz/joins.html>

`left_join(x, y)`

x1	1
x2	2
x3	3

4	2	1
NA	y3	y1
NA	y3	y2



key	val_x	val_y
1	x1	y1
2	x2	y2
3	x3	NA

## Joining datasets [ *full join* ]

<https://r4ds.hadley.nz/joins.html>

`full_join(x, y)`

x1	1
x2	2
x3	3
NA	NA

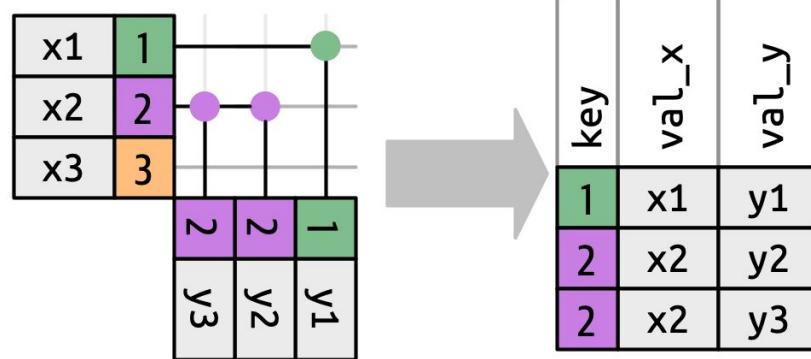
4	2	1
NA	y3	y1
NA	y3	y2
NA	NA	y3



key	val_x	val_y
1	x1	y1
2	x2	y2
3	x3	NA
4	NA	y3

## Joining datasets [ duplicates ]

<https://r4ds.hadley.nz/joins.html>



### Select the appropriate type of join

0

One data frame contains household income for every resident of Boston, with a unique person identifier.

Another data frame lists the set of individuals on government-sponsored health insurance across Massachusetts.

You would like to compute the proportion of Boston residents in each income bracket receiving the insurance.

**Which type of join would you do?**

Inner join (A)

Left join (B)

Full join (C)



## if\_else()

The command **if\_else** lets us select between different values depending on some *test*.

`if_else` takes three parameters:

- A *test* vector of TRUE/FALSE values
- A *yes* value to return when the corresponding entries in the TRUE/FALSE vector are TRUE
- A *no* value to return when the corresponding entries in the TRUE/FALSE vector are FALSE

## if\_else()

```
my_vec <- c(-1, 7, -3, 1, 5)  
if_else(my_vec > 0, my_vec*2, 0)
```

## Exercise

`residents` has two columns: `person_id` and `income_bracket`

`insurance` has two columns: `person_id` and `subsidy`

Your goal is to compute the average subsidy for individuals in each income bracket. This average should be computed over all individuals, not just those who received a subsidy. [ Discuss with your neighbor. ]

```
residents %>%
  _____(insurance, by = “_____”) %>%
  mutate(subsidy = if_else(is.na(subsidy), 0, _____)) %>%
  _____(income_bracket) %>%
  _____(avg_subsidy = _____(subsidy))
```

## Key ideas

The **five verbs** of data manipulation – plus the **group\_by** operation – allow us to carry out sophisticated descriptive statistical analyses.

**Joins** allows us to analyze complex patterns spread across multiple datasets.

## Putting it all together

Now we'll use what we've learned to continue investigating racial discrimination in police stops.

<https://bit.ly/API201-dplyr>