

# Intro to R

Manipulating Data in R

## Recap of Data Cleaning

- `is.na()`, `any(is.na())`, `count()`, and functions from `naniar` like `gg_miss_var()` can help determine if we have NA values
- `filter()` automatically removes NA values - can't confirm or deny if condition is met (need `| is.na()` to keep them)
- `drop_na()` can help you remove NA values from a variable or an entire data frame
- NA values can change your calculation results
- think about what NA values represent

## Recap of Data Cleaning

- `recode()` can help with simple recoding (not based on condition but simple swap)
- `case_when()` can recode **entire values** based on conditions
  - remember `case_when()` needs `TRUE ~ variable` to keep values that aren't specified by conditions, otherwise will be `NA`
- `stringr` package has great functions for looking for specific **parts of values** especially `filter()` and `str_detect()` combined
  - also has other useful string manipulation functions like `str_replace()` and more!
  - `separate()` can split columns into additional columns
  - `unite()` can combine columns

[Cheatsheet](#)

# Manipulating Data

In this module, we will show you how to:

1. Reshape data from wide to long
2. Reshape data from long to wide
3. Merge Data/Joins

# What is wide/long data?

<https://github.com/gadenbuie/tidyexplain/blob/main/images/tidyr-pivoting.gif>

wide

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

# What is wide/long data?

Data is stored *differently* in the tibble.

Wide: has many columns

```
# A tibble: 1 × 4
  State    June_vacc_rate May_vacc_rate April_vacc_rate
  <chr>    <chr>          <chr>         <chr>
1 Alabama 37.2%          36.0%         32.4%
```

Long: column names become data

```
# A tibble: 3 × 3
  State    name          value
  <chr>    <chr>         <chr>
1 Alabama June_vacc_rate 37.2%
2 Alabama May_vacc_rate  36.0%
3 Alabama April_vacc_rate 32.4%
```

## What is wide/long data?

Wide: multiple columns per individual, values spread across multiple columns

```
# A tibble: 2 × 4
```

	State	June_vacc_rate	May_vacc_rate	April_vacc_rate
	<chr>	<chr>	<chr>	<chr>
1	Alabama	37.2%	36.0%	32.4%
2	Alaska	47.5%	46.2%	41.7%

Long: multiple rows per observation, a single column contains the values

```
# A tibble: 6 × 3
```

	State	name	value
	<chr>	<chr>	<chr>
1	Alabama	June_vacc_rate	37.2%
2	Alabama	May_vacc_rate	36.0%
3	Alabama	April_vacc_rate	32.4%
4	Alaska	June_vacc_rate	47.5%
5	Alaska	May_vacc_rate	46.2%
6	Alaska	April_vacc_rate	41.7%

What is wide/long data?

Data is wide or long **with respect to** certain variables.

Wide

Long

	Day 1	Day 2	Day 3
Patient 1	A	B	C
Patient 2	D	E	F

	Day	Value
Patient 1	Day 1	A
Patient 1	Day 2	B
Patient 1	Day 3	C
Patient 2	Day 1	D
Patient 2	Day 2	E
Patient 2	Day 3	F

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# Why do we need to switch between wide/long data?

Wide: **Easier for humans to read**

```
# A tibble: 2 × 4
  State    June_vacc_rate May_vacc_rate April_vacc_rate
  <chr>    <chr>          <chr>          <chr>
1 Alabama 37.2%          36.0%          32.4%
2 Alaska 47.5%          46.2%          41.7%
```

Long: **Easier for R to make plots & do analysis**

```
# A tibble: 6 × 3
  State    name          value
  <chr>    <chr>          <chr>
1 Alabama June_vacc_rate 37.2%
2 Alabama May_vacc_rate  36.0%
3 Alabama April_vacc_rate 32.4%
4 Alaska  June_vacc_rate 47.5%
5 Alaska  May_vacc_rate  46.2%
6 Alaska  April_vacc_rate 41.7%
```

## Pivoting using **tidyr** package

`tidyr` allows you to “tidy” your data. We will be talking about:

- `pivot_longer` - make multiple columns into variables, (wide to long)
- `pivot_wider` - make a variable into multiple columns, (long to wide)
- `separate` - string into multiple columns (review)

The `reshape` command exists. Its arguments are considered more confusing, so we don't recommend it.

You might see old functions `gather` and `spread` when googling. These are older iterations of `pivot_longer` and `pivot_wider`, respectively.

**pivot\_longer...**

## Reshaping data from **wide** to **long**

`pivot_longer()` - puts column data into rows (`tidyr` package)

- First describe which columns we want to “pivot\_longer”

```
{long_data} <- {wide_data} %>% pivot_longer(cols = {columns to pivot})
```

# Reshaping data from wide to long

```
wide_data
```

```
# A tibble: 1 × 3
  June_vacc_rate May_vacc_rate April_vacc_rate
  <chr>          <chr>          <chr>
1 37.2%          36.0%          32.4%
```

```
long_data <- wide_data %>% pivot_longer(cols = everything())
long_data
```

```
# A tibble: 3 × 2
  name      value
  <chr>    <chr>
1 June_vacc_rate 37.2%
2 May_vacc_rate  36.0%
3 April_vacc_rate 32.4%
```

## Reshaping data from wide to long

`pivot_longer()` - puts column data into rows (tidyr package)

- First describe which columns we want to “pivot\_longer”
- `names_to` = gives a new name to the pivoted columns
- `values_to` = gives a new name to the values that used to be in those columns

```
{long_data} <- {wide_data} %>% pivot_longer(cols = {columns to pivot},  
                                             names_to = {New column name: contains old column names},  
                                             values_to = {New column name: contains cell values})
```

# Reshaping data from wide to long

wide\_data

```
# A tibble: 1 × 3
  June_vacc_rate May_vacc_rate April_vacc_rate
  <chr>          <chr>          <chr>
1 37.2%          36.0%          32.4%
```

```
long_data <- wide_data %>% pivot_longer(cols = everything(),
                                         names_to = "Month",
                                         values_to = "Rate")
```

long\_data

```
# A tibble: 3 × 2
  Month      Rate
  <chr>      <chr>
1 June_vacc_rate 37.2%
2 May_vacc_rate  36.0%
3 April_vacc_rate 32.4%
```

## Data used: Charm City Circulator

[http://jhudatascience.org/intro\\_to\\_r/data/Charm\\_City\\_Circulator\\_Ridership.csv](http://jhudatascience.org/intro_to_r/data/Charm_City_Circulator_Ridership.csv)

```
library(jhur)
circ <- read_circulator()
head(circ, 5)
```

```
# A tibble: 5 × 15
```

	day	date	orangeBoardings	orangeAlightings	orangeAverage	purpleBoardings
	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
1	Monday	01/1...	877	1027	952	
2	Tuesday	01/1...	777	815	796	
3	Wednesday	01/1...	1203	1220	1212.	
4	Thursday	01/1...	1194	1233	1214.	
5	Friday	01/1...	1645	1643	1644	

```
# ... with 9 more variables: purpleAlightings <dbl>, purpleAverage <dbl>,  
#   greenBoardings <dbl>, greenAlightings <dbl>, greenAverage <dbl>,  
#   bannerBoardings <dbl>, bannerAlightings <dbl>, bannerAverage <dbl>,  
#   daily <dbl>
```



## Reshaping data from wide to long

```
long <- circ %>%  
  pivot_longer(starts_with(c("orange", "purple", "green", "banner")))  
long
```

```
# A tibble: 13,752 × 5  
  day      date      daily name      value  
  <chr>   <chr>    <dbl> <chr>    <dbl>  
1 Monday 01/11/2010    952 orangeBoardings    877  
2 Monday 01/11/2010    952 orangeAlightings  1027  
3 Monday 01/11/2010    952 orangeAverage     952  
4 Monday 01/11/2010    952 purpleBoardings    NA  
5 Monday 01/11/2010    952 purpleAlightings    NA  
6 Monday 01/11/2010    952 purpleAverage       NA  
7 Monday 01/11/2010    952 greenBoardings     NA  
8 Monday 01/11/2010    952 greenAlightings     NA  
9 Monday 01/11/2010    952 greenAverage        NA  
10 Monday 01/11/2010    952 bannerBoardings    NA  
# ... with 13,742 more rows
```

## Reshaping data from wide to long

There are many ways to select the columns we want. Use `?tidyr_tidy_select` to look at more column selection options.

```
long <- circ %>%  
  pivot_longer( !c(day, date, daily))  
long
```

```
# A tibble: 13,752 × 5  
   day    date    daily name          value  
   <chr> <chr>    <dbl> <chr>          <dbl>  
1 Monday 01/11/2010  952 orangeBoardings    877  
2 Monday 01/11/2010  952 orangeAlightings  1027  
3 Monday 01/11/2010  952 orangeAverage     952  
4 Monday 01/11/2010  952 purpleBoardings    NA  
5 Monday 01/11/2010  952 purpleAlightings    NA  
6 Monday 01/11/2010  952 purpleAverage       NA  
7 Monday 01/11/2010  952 greenBoardings     NA  
8 Monday 01/11/2010  952 greenAlightings     NA  
9 Monday 01/11/2010  952 greenAverage        NA  
10 Monday 01/11/2010  952 bannerBoardings    NA  
# ... with 13,742 more rows
```

# Reshaping data from wide to long

```
long %>% count(name)
```

```
# A tibble: 12 × 2
```

	name <chr>	n <int>
1	bannerAlightings	1146
2	bannerAverage	1146
3	bannerBoardings	1146
4	greenAlightings	1146
5	greenAverage	1146
6	greenBoardings	1146
7	orangeAlightings	1146
8	orangeAverage	1146
9	orangeBoardings	1146
10	purpleAlightings	1146
11	purpleAverage	1146
12	purpleBoardings	1146

## Cleaning up long data

We will use `str_replace` from the `stringr` package to put `_` in the names

```
long <- long %>% mutate(  
  name = str_replace(name, "Board", "_Board"),  
  name = str_replace(name, "Alight", "_Alight"),  
  name = str_replace(name, "Average", "_Average")  
)  
long
```

# A tibble: 13,752 × 5

	day	date	daily	name	value
	<chr>	<chr>	<dbl>	<chr>	<dbl>
1	Monday	01/11/2010	952	orange_Boardings	877
2	Monday	01/11/2010	952	orange_Alightings	1027
3	Monday	01/11/2010	952	orange_Average	952
4	Monday	01/11/2010	952	purple_Boardings	NA
5	Monday	01/11/2010	952	purple_Alightings	NA
6	Monday	01/11/2010	952	purple_Average	NA
7	Monday	01/11/2010	952	green_Boardings	NA
8	Monday	01/11/2010	952	green_Alightings	NA
9	Monday	01/11/2010	952	green_Average	NA
10	Monday	01/11/2010	952	banner_Boardings	NA

# ... with 13,742 more rows

## Cleaning up long data with `separate()`

- first argument - which column should be split up?
- `"into ="` gives names to the new columns
- `"sep ="` to show where the separation should happen.

```
long <- long %>%  
  separate(name, into = c("line", "type"), sep = "_")  
long
```

```
# A tibble: 13,752 × 6  
  day    date      daily line    type      value  
  <chr> <chr>    <dbl> <chr> <chr>    <dbl>  
1 Monday 01/11/2010    952 orange Boardings    877  
2 Monday 01/11/2010    952 orange Alightings 1027  
3 Monday 01/11/2010    952 orange Average    952  
4 Monday 01/11/2010    952 purple Boardings    NA  
5 Monday 01/11/2010    952 purple Alightings    NA  
6 Monday 01/11/2010    952 purple Average    NA  
7 Monday 01/11/2010    952 green Boardings    NA  
8 Monday 01/11/2010    952 green Alightings    NA  
9 Monday 01/11/2010    952 green Average    NA  
10 Monday 01/11/2010    952 banner Boardings    NA  
# ... with 13,742 more rows
```

**pivot\_wider...**

## Reshaping data from **long** to **wide**

`pivot_wider()` - spreads row data into columns (`tidyr` package)

- `names_from` = the old column whose contents will be spread into multiple new column names.
- `values_from` = the old column whose contents will fill in the values of those new columns.

```
{wide_data} <- {long_data} %>%  
  pivot_wider(names_from = {Old column name: contains new column names},  
              values_from = {Old column name: contains new cell values})
```

# Reshaping data from long to wide

long\_data

```
# A tibble: 3 × 2
  Month      Rate
<chr>      <chr>
1 June_vacc_rate 37.2%
2 May_vacc_rate  36.0%
3 April_vacc_rate 32.4%
```

```
wide_data <- long_data %>% pivot_wider(names_from = "Month",
                                       values_from = "Rate")
```

wide\_data

```
# A tibble: 1 × 3
  June_vacc_rate May_vacc_rate April_vacc_rate
<chr>          <chr>          <chr>
1 37.2%         36.0%         32.4%
```



# Reshaping Charm City Circulator

long

```
# A tibble: 13,752 × 6
```

	day <chr>	date <chr>	daily <dbl>	line <chr>	type <chr>	value <dbl>
1	Monday	01/11/2010	952	orange	Boardings	877
2	Monday	01/11/2010	952	orange	Alightings	1027
3	Monday	01/11/2010	952	orange	Average	952
4	Monday	01/11/2010	952	purple	Boardings	NA
5	Monday	01/11/2010	952	purple	Alightings	NA
6	Monday	01/11/2010	952	purple	Average	NA
7	Monday	01/11/2010	952	green	Boardings	NA
8	Monday	01/11/2010	952	green	Alightings	NA
9	Monday	01/11/2010	952	green	Average	NA
10	Monday	01/11/2010	952	banner	Boardings	NA

```
# ... with 13,742 more rows
```

# Reshaping Charm City Circulator

```
wide <- long %>% pivot_wider(names_from = "type",  
                             values_from = "value")
```

```
wide
```

```
# A tibble: 4,584 × 7
```

	day <chr>	date <chr>	daily <dbl>	line <chr>	Boardings <dbl>	Alightings <dbl>	Average <dbl>
1	Monday	01/11/2010	952	orange	877	1027	952
2	Monday	01/11/2010	952	purple	NA	NA	NA
3	Monday	01/11/2010	952	green	NA	NA	NA
4	Monday	01/11/2010	952	banner	NA	NA	NA
5	Tuesday	01/12/2010	796	orange	777	815	796
6	Tuesday	01/12/2010	796	purple	NA	NA	NA
7	Tuesday	01/12/2010	796	green	NA	NA	NA
8	Tuesday	01/12/2010	796	banner	NA	NA	NA
9	Wednesday	01/13/2010	1212.	orange	1203	1220	1212.
10	Wednesday	01/13/2010	1212.	purple	NA	NA	NA

```
# ... with 4,574 more rows
```

# Inconsistencies in non-pivoted columns?

Notice “daily” column has different values

long2

```
# A tibble: 13,752 × 6
  day      date      daily line  type      value
<chr> <chr>    <dbl> <chr> <chr>    <dbl>
1 Monday 01/11/2010 4000 orange Boardings    877
2 Monday 01/11/2010  10 orange Alightings 1027
3 Monday 01/11/2010 952 orange Average    952
4 Monday 01/11/2010 952 purple Boardings    NA
5 Monday 01/11/2010 952 purple Alightings    NA
6 Monday 01/11/2010 952 purple Average    NA
7 Monday 01/11/2010 952 green Boardings    NA
8 Monday 01/11/2010 952 green Alightings    NA
9 Monday 01/11/2010 952 green Average    NA
10 Monday 01/11/2010 952 banner Boardings    NA
# ... with 13,742 more rows
```

## Inconsistencies in non-pivoted columns?

R won't drop data while pivoting.

```
wide2 <- long2 %>% pivot_wider(names_from = "type",  
                               values_from = "value")
```

```
wide2
```

```
# A tibble: 4,586 × 7
```

	day <chr>	date <chr>	daily <dbl>	line <chr>	Boardings <dbl>	Alightings <dbl>	Average <dbl>
1	Monday	01/11/2010	4000	orange	877	NA	NA
2	Monday	01/11/2010	10	orange	NA	1027	NA
3	Monday	01/11/2010	952	orange	NA	NA	952
4	Monday	01/11/2010	952	purple	NA	NA	NA
5	Monday	01/11/2010	952	green	NA	NA	NA
6	Monday	01/11/2010	952	banner	NA	NA	NA
7	Tuesday	01/12/2010	796	orange	777	815	796
8	Tuesday	01/12/2010	796	purple	NA	NA	NA
9	Tuesday	01/12/2010	796	green	NA	NA	NA
10	Tuesday	01/12/2010	796	banner	NA	NA	NA

```
# ... with 4,576 more rows
```

## Summary

- `tidyr` package helps us convert between wide and long data
- `pivot_longer()` goes from wide -> long
  - Specify columns you want to pivot
  - Specify `names_to =` and `values_to =` for custom naming
- `pivot_wider()` goes from long -> wide
  - Specify `names_from =` and `values_from =`

# Lab Part 1

[Class Website](#)

[Lab](#)

## Joining in `dplyr`

- Merging/joining data sets together - usually on key variables, usually “id”
- `?join` - see different types of joining for `dplyr`
- `inner_join(x, y)` - only rows that match for x and y are kept
- `full_join(x, y)` - all rows of x and y are kept
- `left_join(x, y)` - all rows of x are kept even if not merged with y
- `right_join(x, y)` - all rows of y are kept even if not merged with x
- `anti_join(x, y)` - all rows from x not in y keeping just columns from x.

# Merging: Simple Data

data\_As

```
# A tibble: 2 × 3
  State      June_vacc_rate May_vacc_rate
<chr>      <chr>           <chr>
1 Alabama  37.2%             36.0%
2 Alaska   47.5%             46.2%
```

data\_cold

```
# A tibble: 2 × 2
  State      April_vacc_rate
<chr>      <chr>
1 Maine    32.4%
2 Alaska   41.7%
```



# Inner Join

<https://github.com/gadenbuie/tidyexplain/blob/main/images/inner-join.gif>

`inner_join(x, y)`



# Inner Join

```
ij = inner_join(data_As, data_cold)
```

```
Joining, by = "State"
```

```
ij
```

```
# A tibble: 1 × 4
```

	State	June_vacc_rate	May_vacc_rate	April_vacc_rate
	<chr>	<chr>	<chr>	<chr>
1	Alaska	47.5%	46.2%	41.7%

# Left Join

<https://raw.githubusercontent.com/gadenbuie/tidyexplain/main/images/left-join.gif>

`left_join(x, y)`



# Left Join

```
lj = left_join(data_As, data_cold)
```

```
Joining, by = "State"
```

```
lj
```

```
# A tibble: 2 × 4
```

	State	June_vacc_rate	May_vacc_rate	April_vacc_rate
	<chr>	<chr>	<chr>	<chr>
1	Alabama	37.2%	36.0%	<NA>
2	Alaska	47.5%	46.2%	41.7%

## Install **tidylog** package to log outputs

```
# install.packages("tidylog")
library(tidylog)
left_join(data_As, data_cold)
```

```
Joining, by = "State"
left_join: added one column (April_vacc_rate)
> rows only in x 1
> rows only in y (1)
> matched rows 1
> ===
> rows total 2
```

```
# A tibble: 2 × 4
  State      June_vacc_rate May_vacc_rate April_vacc_rate
  <chr>      <chr>          <chr>          <chr>
1 Alabama  37.2%           36.0%          <NA>
2 Alaska   47.5%           46.2%          41.7%
```

## Right Join

<https://raw.githubusercontent.com/gadenbuie/tidyexplain/main/images/right-join.gif>

`right_join(x, y)`



# Right Join

```
rj <- right_join(data_As, data_cold)
```

```
Joining, by = "State"  
right_join: added one column (April_vacc_rate)  
> rows only in x (1)  
> rows only in y 1  
> matched rows 1  
> ===  
> rows total 2
```

```
rj
```

```
# A tibble: 2 × 4  
  State   June_vacc_rate May_vacc_rate April_vacc_rate  
  <chr>   <chr>           <chr>         <chr>  
1 Alaska 47.5%           46.2%         41.7%  
2 Maine  <NA>            <NA>          32.4%
```

## Left Join: Switching arguments

```
lj2 <- left_join(data_cold, data_As)
```

```
Joining, by = "State"  
left_join: added 2 columns (June_vacc_rate, May_vacc_rate)  
> rows only in x 1  
> rows only in y (1)  
> matched rows 1  
> ===  
> rows total 2
```

```
lj2
```

```
# A tibble: 2 × 4  
  State April_vacc_rate June_vacc_rate May_vacc_rate  
  <chr>   <chr>           <chr>         <chr>  
1 Maine  32.4%           <NA>         <NA>  
2 Alaska 41.7%          47.5%        46.2%
```



# Full Join

<https://raw.githubusercontent.com/gadenbuie/tidyexplain/main/images/full-join.gif>

`full_join(x, y)`



# FullJoin

```
fj <- full_join(data_As, data_cold)
```

```
Joining, by = "State"  
full_join: added one column (April_vacc_rate)  
> rows only in x 1  
> rows only in y 1  
> matched rows 1  
> ===  
> rows total 3
```

```
fj
```

```
# A tibble: 3 × 4  
  State    June_vacc_rate May_vacc_rate April_vacc_rate  
  <chr>    <chr>           <chr>         <chr>  
1 Alabama 37.2%           36.0%         <NA>  
2 Alaska 47.5%           46.2%         41.7%  
3 Maine  <NA>            <NA>         32.4%
```

## Watch out for “includes duplicates”

```
data_As
```

```
# A tibble: 2 × 2
  State    state_bird
  <chr>    <chr>
1 Alabama wild turkey
2 Alaska  willow ptarmigan
```

```
data_cold
```

```
# A tibble: 3 × 3
  State    vacc_rate month
  <chr>    <chr>    <chr>
1 Maine   32.4%    April
2 Alaska  41.7%    April
3 Alaska  46.2%    May
```

## Watch out for “includes duplicates”

```
lj <- left_join(data_As, data_cold)
```

```
Joining, by = "State"  
left_join: added 2 columns (vacc_rate, month)  
> rows only in x 1  
> rows only in y (1)  
> matched rows 2 (includes duplicates)  
> ===  
> rows total 3
```

## Watch out for “includes duplicates”

Data including the joining column (“State”) has been duplicated.

```
lj
```

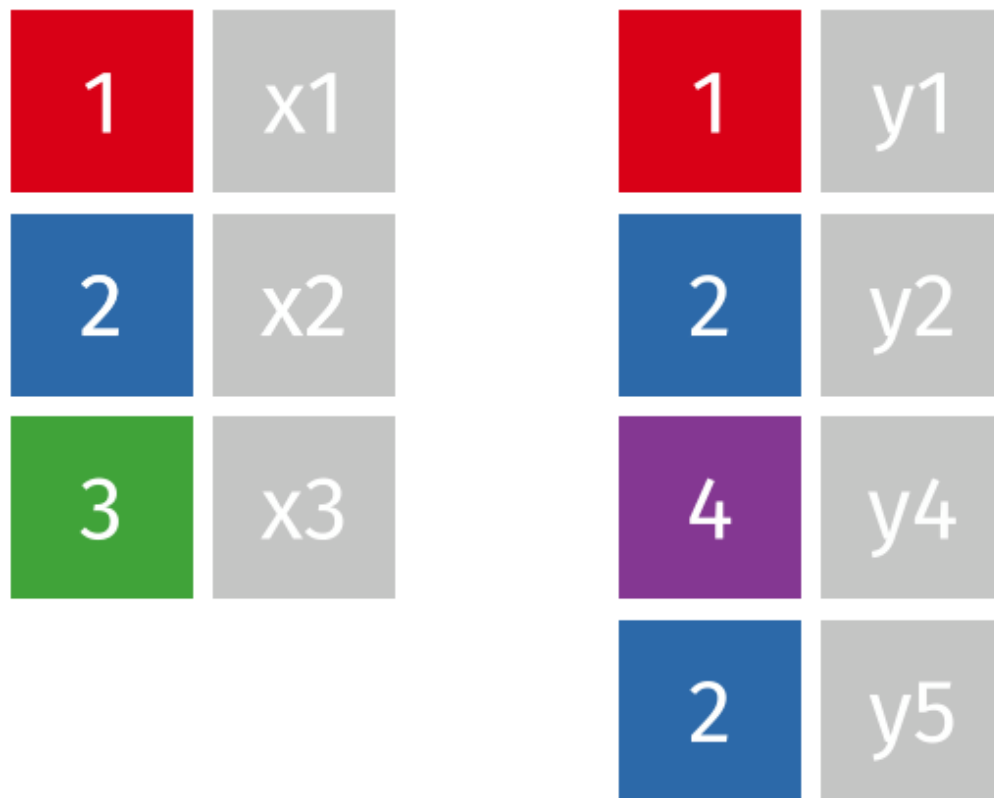
```
# A tibble: 3 × 4
  State    state_bird      vacc_rate month
  <chr>    <chr>          <chr>    <chr>
1 Alabama wild turkey    <NA>     <NA>
2 Alaska  willow ptarmigan 41.7%    April
3 Alaska  willow ptarmigan 46.2%    May
```

Note that “Alaska willow ptarmigan” appears twice.

## Watch out for “includes duplicates”

<https://github.com/gadenbuie/tidyexplain/blob/main/images/left-join-extra.gif>

`left_join(x, y)`



## Stop tidylog

```
unloadNamespace("tidylog")
```

# Duplicated

- The `uplicated` function can give you indications if there are duplicates in a **vector**:

```
uplicated(1:5)
```

```
[1] FALSE FALSE FALSE FALSE FALSE
```

```
uplicated(c(1:5, 1))
```

```
[1] FALSE FALSE FALSE FALSE FALSE TRUE
```

```
lj %>% mutate(dup_State = duplicated(State))
```

```
# A tibble: 3 × 5
```

	State	state_bird	vacc_rate	month	dup_State
	<chr>	<chr>	<chr>	<chr>	<lgl>
1	Alabama	wild turkey	<NA>	<NA>	FALSE
2	Alaska	willow ptarmigan	41.7%	April	FALSE
3	Alaska	willow ptarmigan	46.2%	May	TRUE



## Using the **by** argument

By default joins use the intersection of column names. If **by** is specified, it uses that.

```
full_join(data_As, data_cold, by = "State")
```

```
# A tibble: 4 × 4
  State    state_bird      vacc_rate month
  <chr>    <chr>          <chr>    <chr>
1 Alabama wild turkey    <NA>     <NA>
2 Alaska  willow ptarmigan 41.7%    April
3 Alaska  willow ptarmigan 46.2%    May
4 Maine   <NA>             32.4%    April
```

## Using the **by** argument

You can join based on multiple columns by using something like `by = c(col1, col2)`.

If the datasets have two different names for the same data, use:

```
full_join(x, y, by = c("a" = "b"))
```

## Using “setdiff”

We might want to determine what indexes ARE in the first dataset that AREN'T in the second:

```
data_As
```

```
# A tibble: 2 × 2
  State    state_bird
<chr>    <chr>
1 Alabama wild turkey
2 Alaska  willow ptarmigan
```

```
data_cold
```

```
# A tibble: 3 × 3
  State    vacc_rate month
<chr>    <chr>    <chr>
1 Maine  32.4%    April
2 Alaska 41.7%    April
3 Alaska 46.2%    May
```

## Using “`setdiff`”

Use `setdiff` to determine what indexes ARE in the first dataset that AREN'T in the second:

```
A_states <- data_As %>% pull(State)
cold_states <- data_cold %>% pull(State)
```

```
setdiff(A_states, cold_states)
```

```
[1] "Alabama"
```

```
setdiff(cold_states, A_states)
```

```
[1] "Maine"
```

## Summary

- Merging/joining data sets together - assumes all column names that overlap
  - use the `by = c("a" = "b")` if they differ
- `inner_join(x, y)` - only rows that match for x and y are kept
- `full_join(x, y)` - all rows of x and y are kept
- `left_join(x, y)` - all rows of x are kept even if not merged with y
- `right_join(x, y)` - all rows of y are kept even if not merged with x
- Use the `tidylog` package for a detailed summary
- `setdiff(x, y)` shows what in x is missing from y

## Lab Part 2

[Class Website](#)

[Lab](#)



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