

Data Visualization in R with `ggplot2`

Data Transformation with `dplyr`

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Physalia Courses | November 9-13 2020

Photo by Richard Strozyński

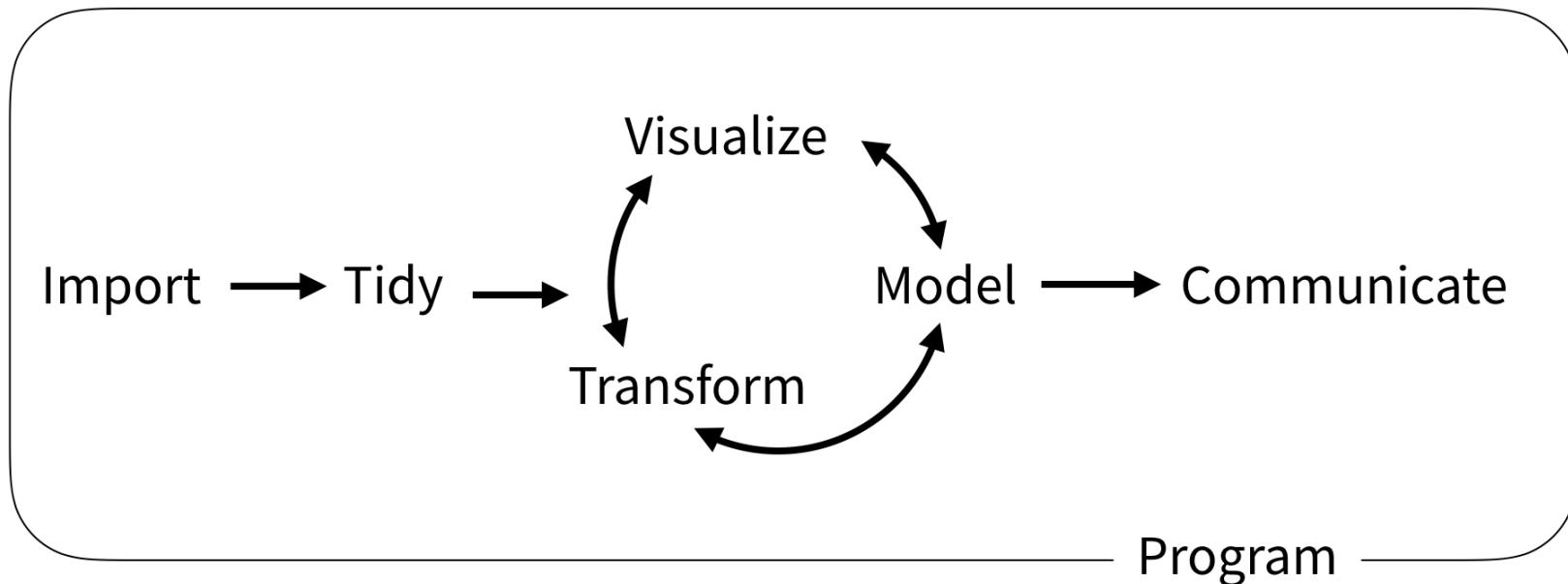
dplyr : go wrangling



Illustration by Allison Horst (github.com/allisonhorst/stats-illustrations)

The **tidyverse**

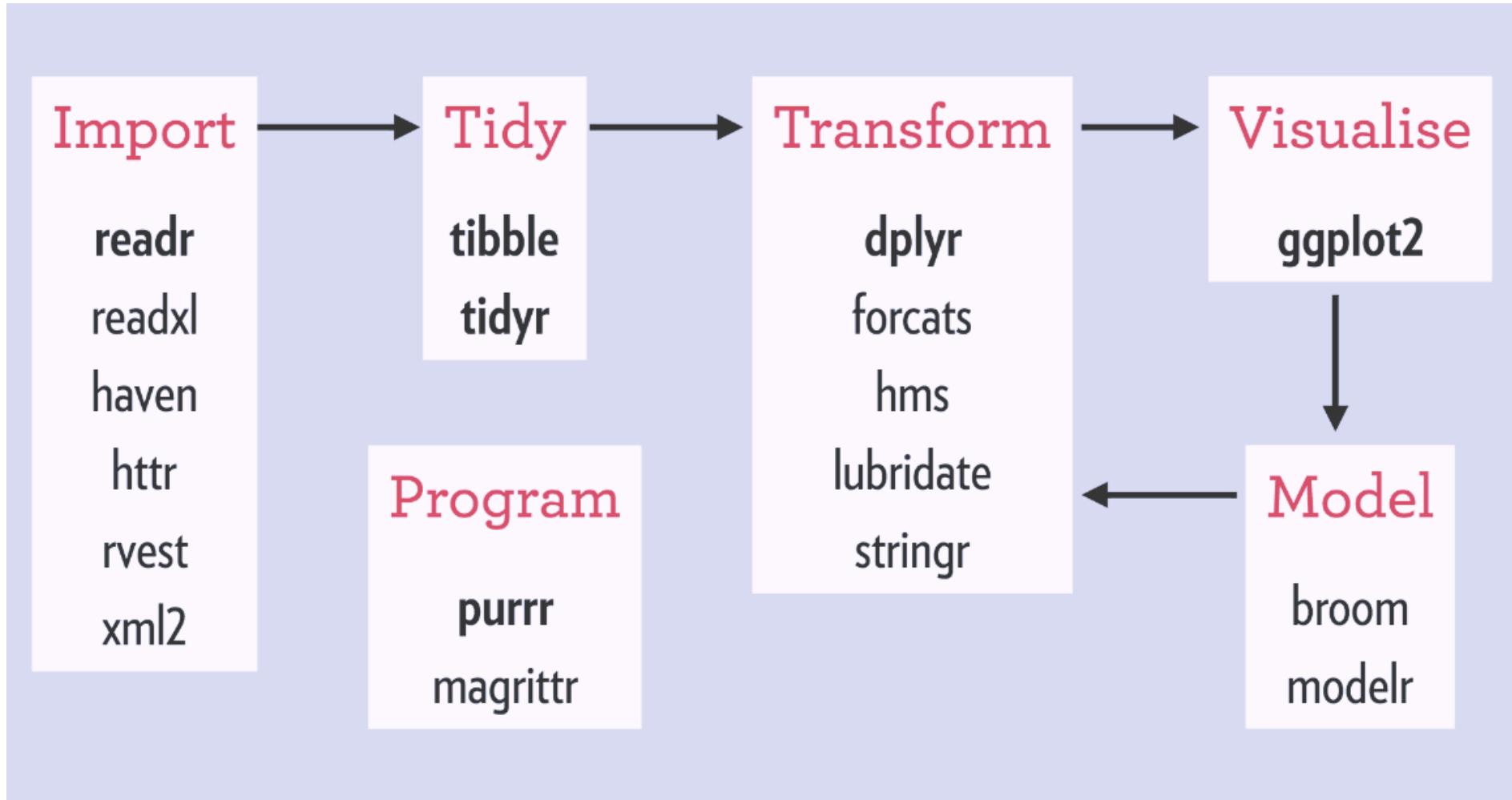
The **tidyverse** provides exemplary support for data wrangling and is the main reason for the recent popularity of R, especially in data-driven environments.



Source: rstudio-education.github.io/tidyverse-cookbook

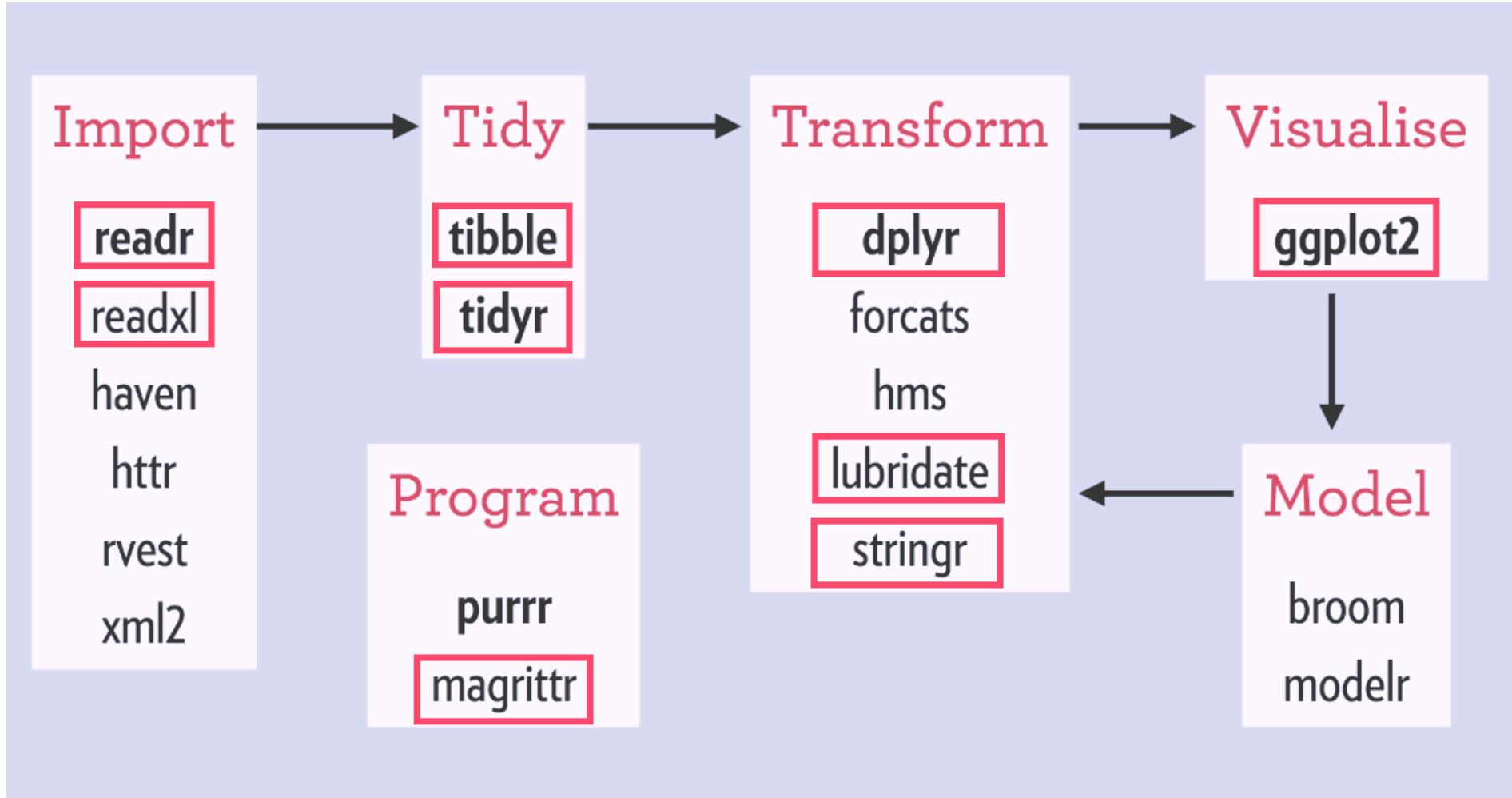
Modified from Hadley Wickham's "R for Data Science" (R4DS)

The Typical Data Science Project Flow



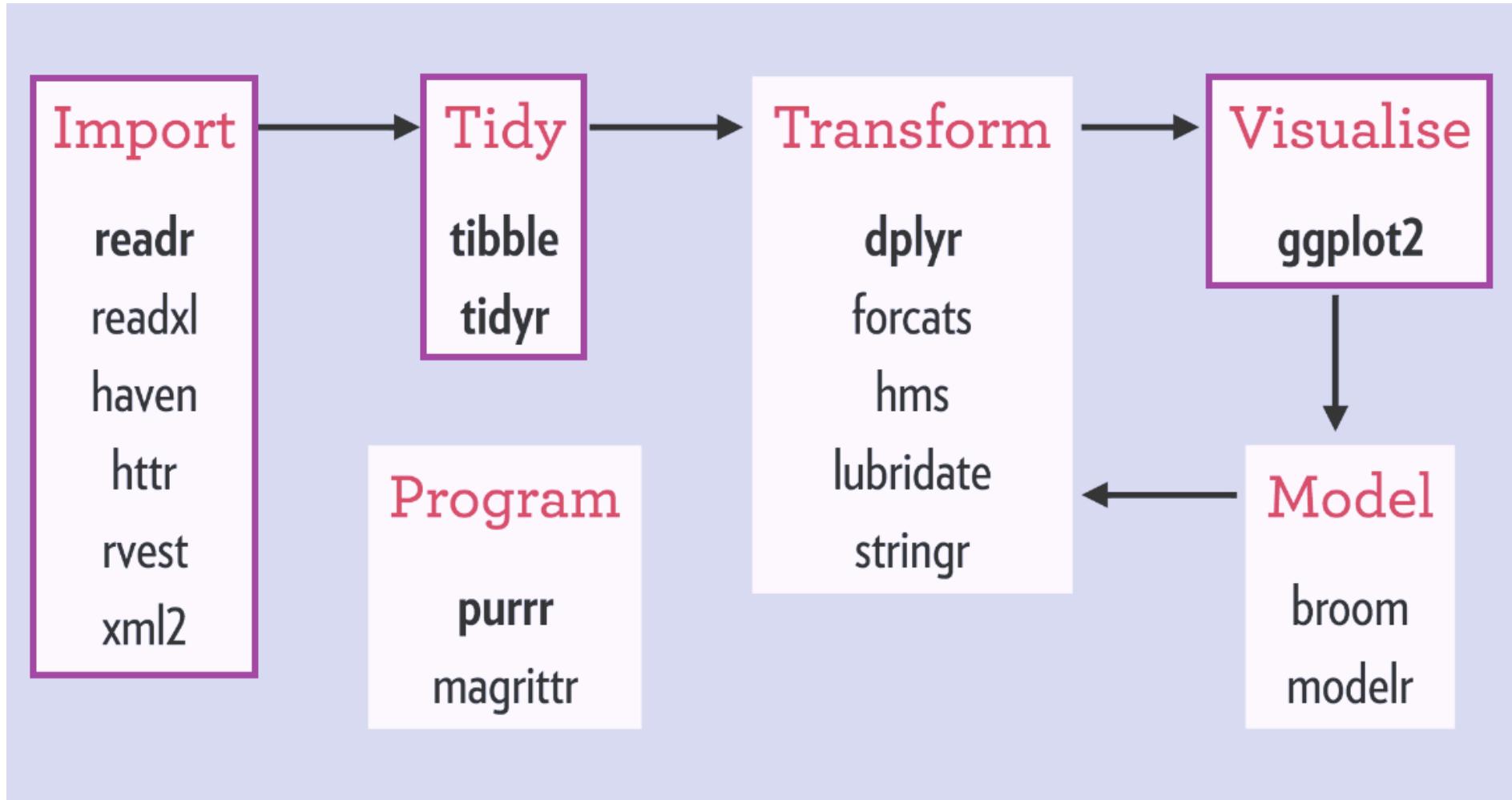
Source: www.rviews.rstudio.com/post/2017-06-09-What-is-the-tidyverse_files/tidyverse1.png

The Typical Data Science Project Flow



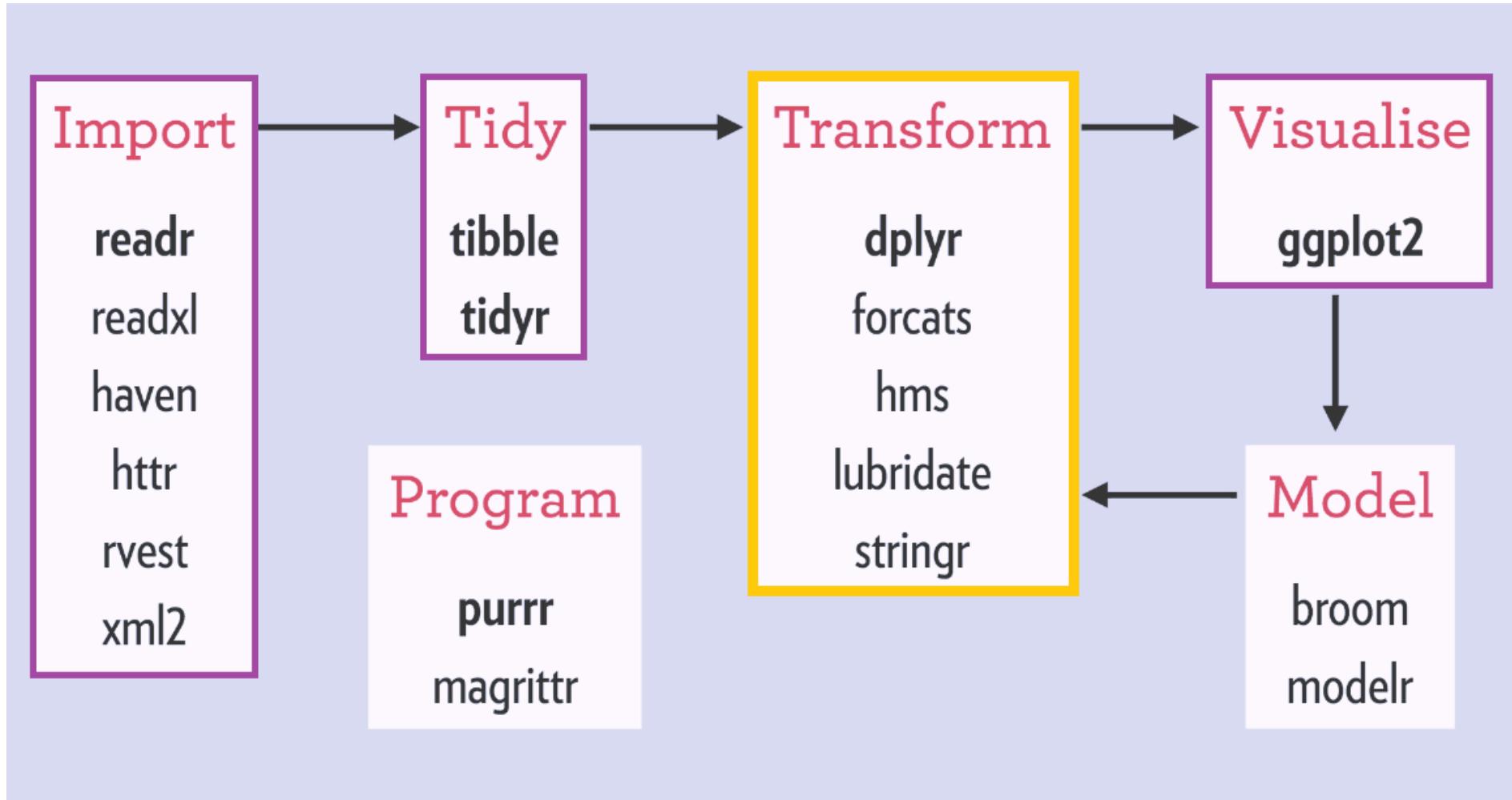
Source: www.rviews.rstudio.com/post/2017-06-09-What-is-the-tidyverse_files/tidyverse1.png

The Typical Data Science Project Flow



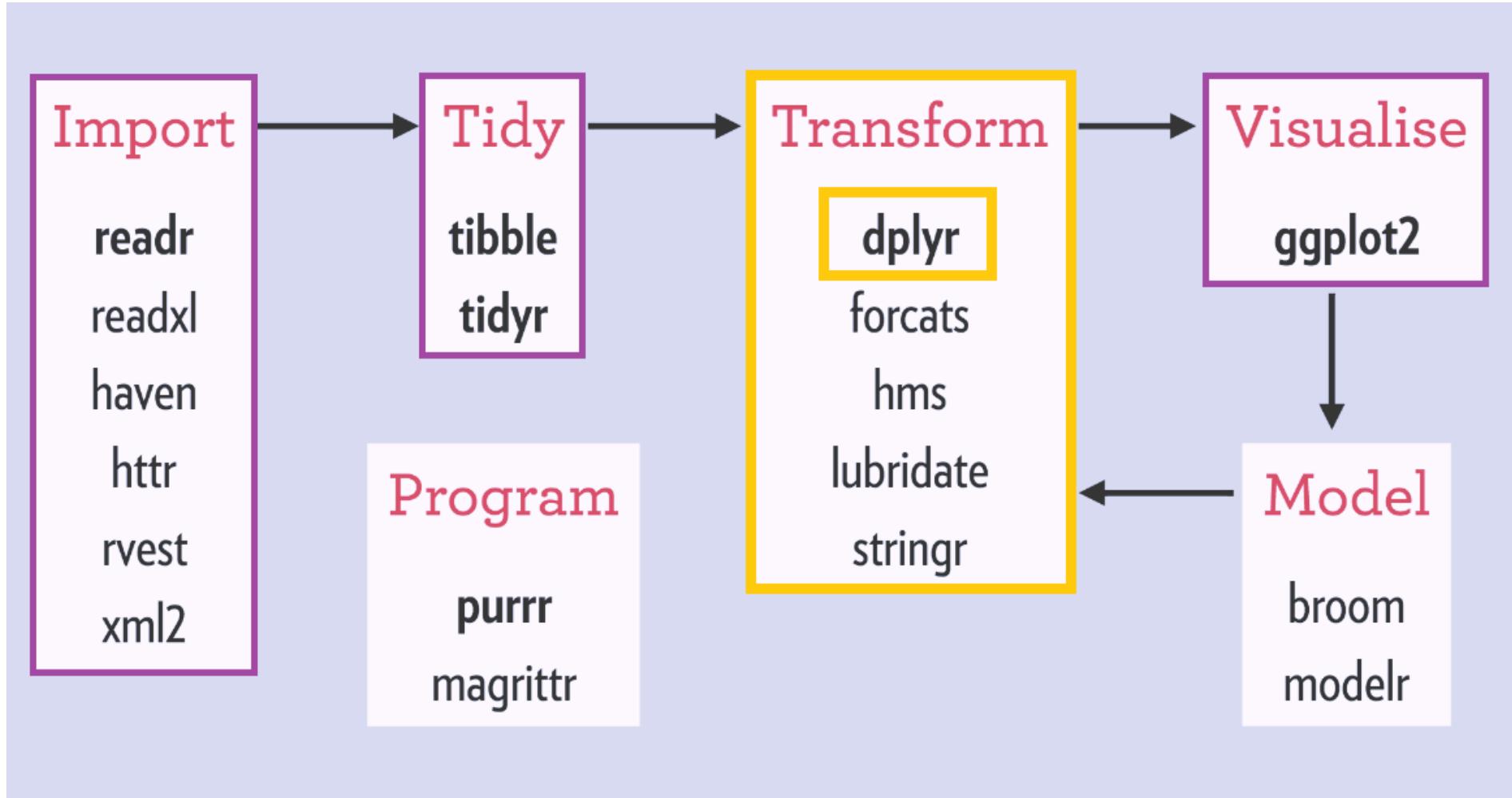
Source: www.rviews.rstudio.com/post/2017-06-09-What-is-the-tidyverse_files/tidyverse1.png

The Typical Data Science Project Flow



Source: www.rviews.rstudio.com/post/2017-06-09-What-is-the-tidyverse_files/tidyverse1.png

The Typical Data Science Project Flow



Source: www.rviews.rstudio.com/post/2017-06-09-What-is-the-tidyverse_files/tidyverse1.png

Our Example Data

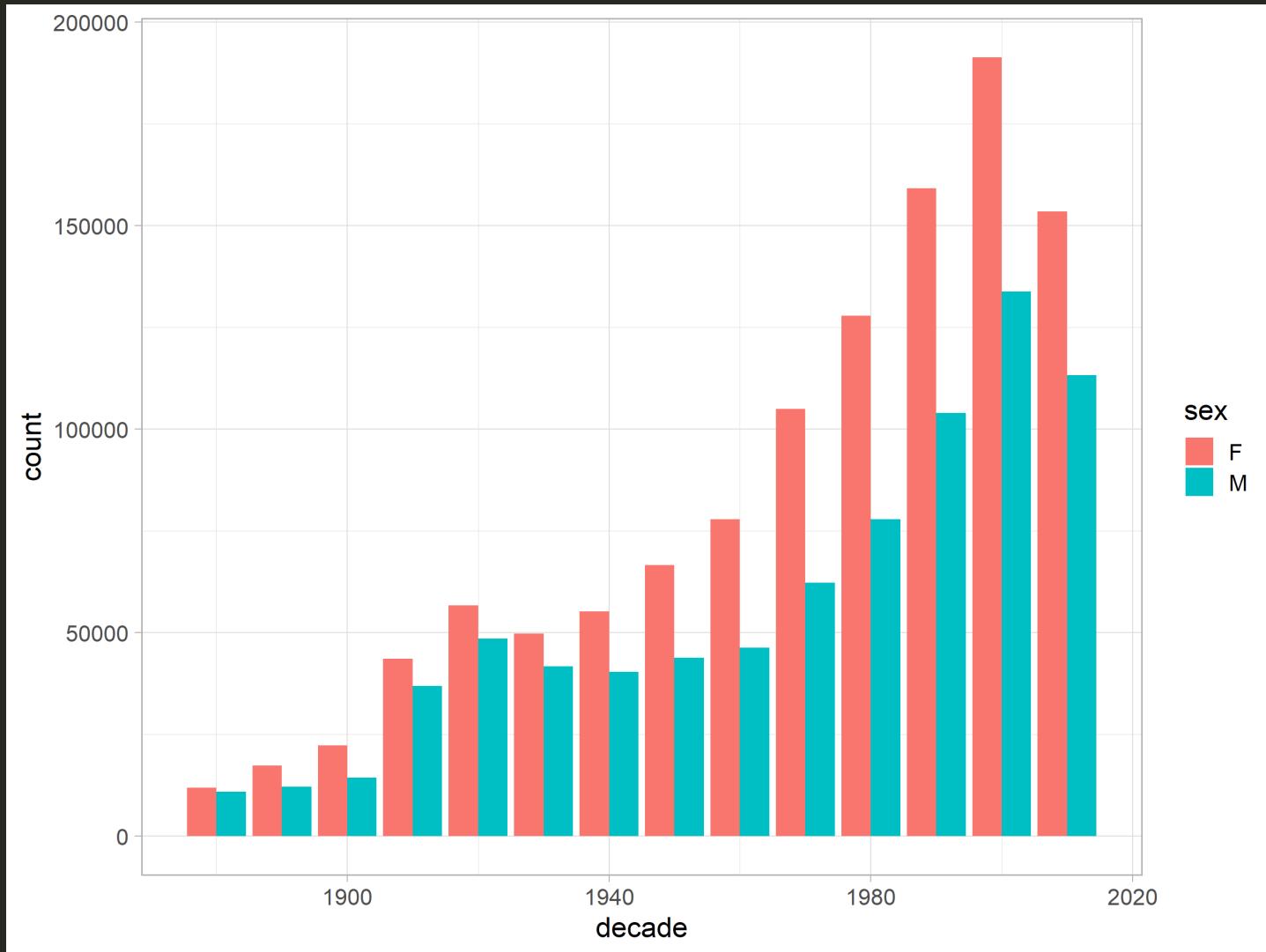
- Count and proportional data of baby names used for at least five children in the US per sex and year from 1880 to 2017
- 1,924,665 rows (observations) and 4 columns (variables)
- Data source: [US Social Security Administration \(SSA\)](#)
- Available as data set **babynames** in the package **babynames**:
`install.packages("babynames")` (includes 3 more data sets)
- See also [??babynames](#)

Our Example Data: babynames

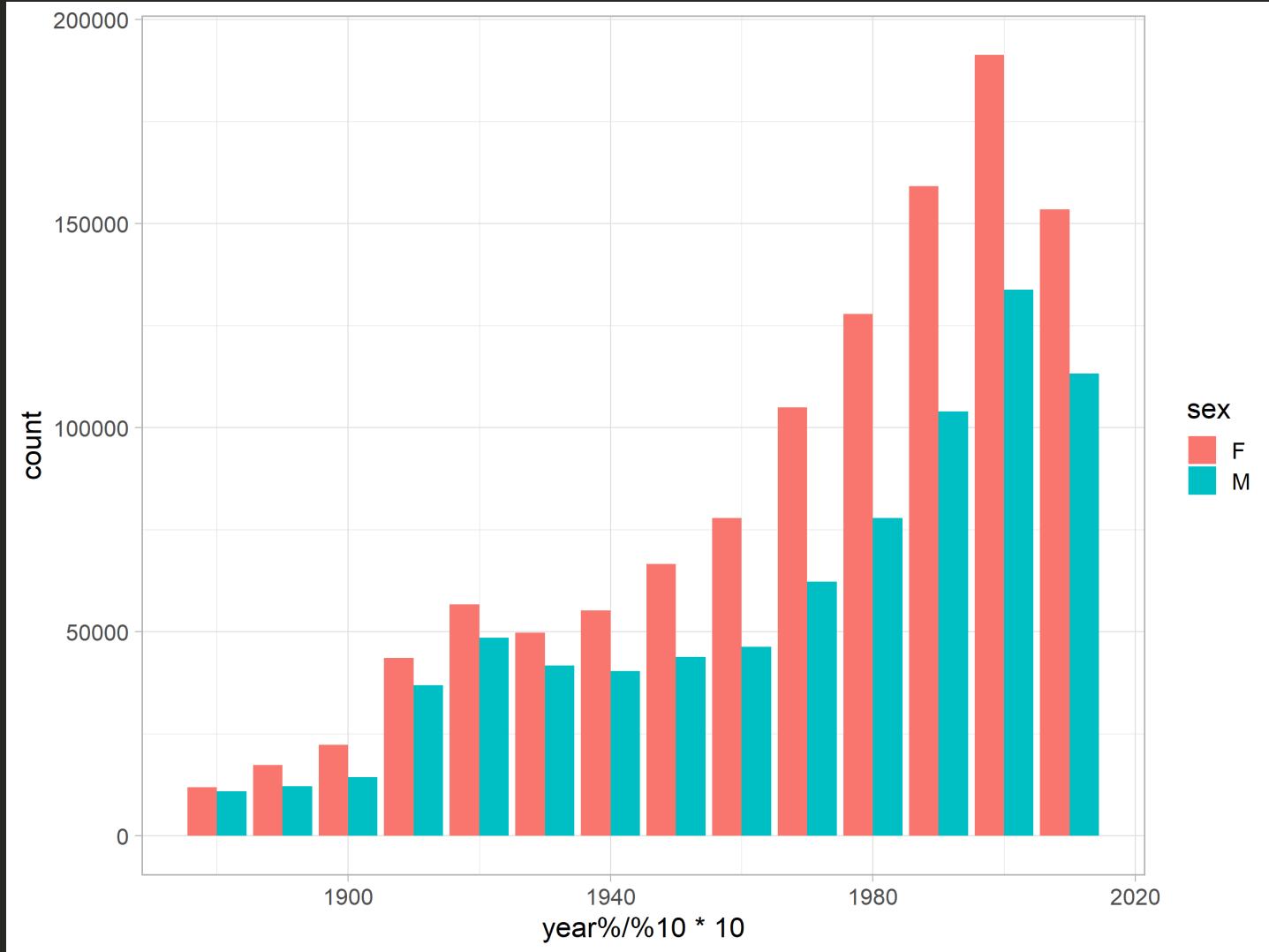
```
library(babynames)

glimpse(babynames)
## Rows: 1,924,665
## Columns: 5
## $ year <dbl> 1880, 1880, 1880, 1880, 1880, 1880, 1880, 1880, 1880, 18...
## $ sex <chr> "F", "F...
## $ name <chr> "Mary", "Anna", "Emma", "Elizabeth", "Minnie", "Margaret", "Id...
## $ n <int> 7065, 2604, 2003, 1939, 1746, 1578, 1472, 1414, 1320, 1288, 12...
## $ prop <dbl> 0.07238359, 0.02667896, 0.02052149, 0.01986579, 0.01788843, 0....
```

Your Turn!



Your Turn! (Less Mean)



Excuse: Integer Division

```
1 %% 10
## [1] 0
2 %% 10
## [1] 0
11 %% 10
## [1] 1
12 %% 10
## [1] 1
```

Excuse: Integer Division

```
1 %% 10
## [1] 0
2 %% 10
## [1] 0
11 %% 10
## [1] 1
12 %% 10
## [1] 1

2019 %% 10
## [1] 201
2020 %% 10
## [1] 202
```

Excourse: Integer Division

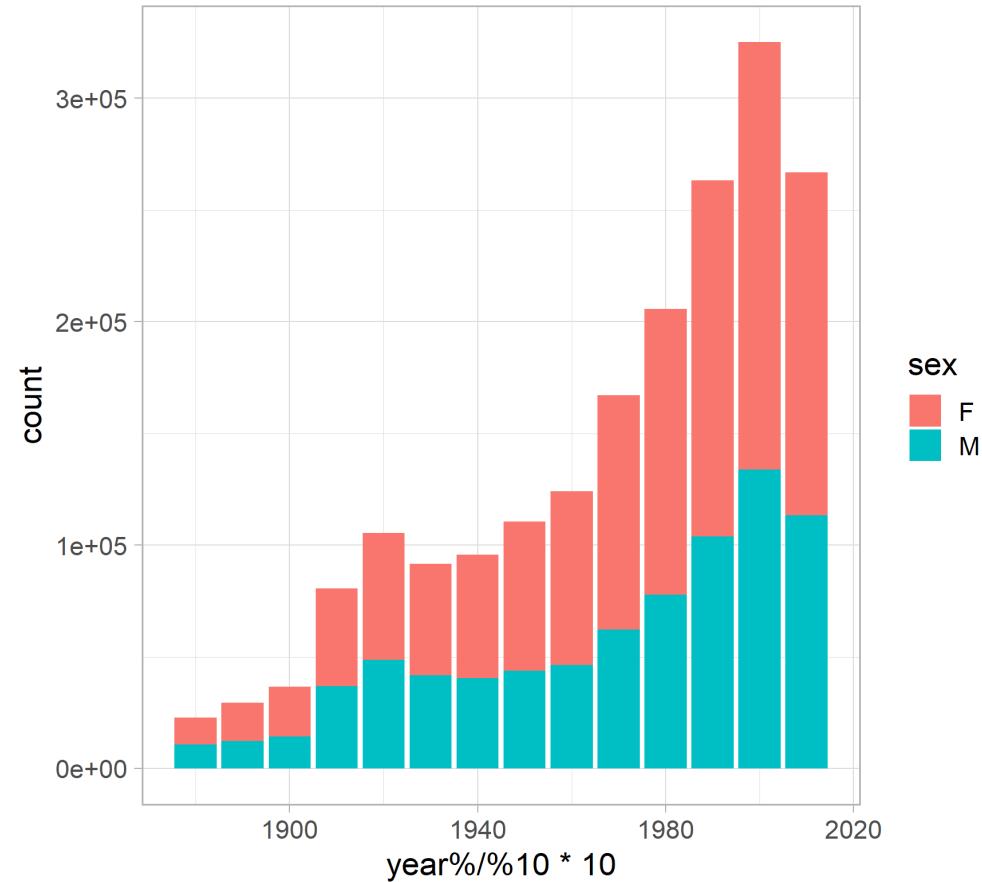
```
1 %% 10
## [1] 0
2 %% 10
## [1] 0
11 %% 10
## [1] 1
12 %% 10
## [1] 1

2019 %% 10
## [1] 201
2020 %% 10
## [1] 202

2019 %% 10 * 10
## [1] 2010
2020 %% 10 * 10
## [1] 2020
```

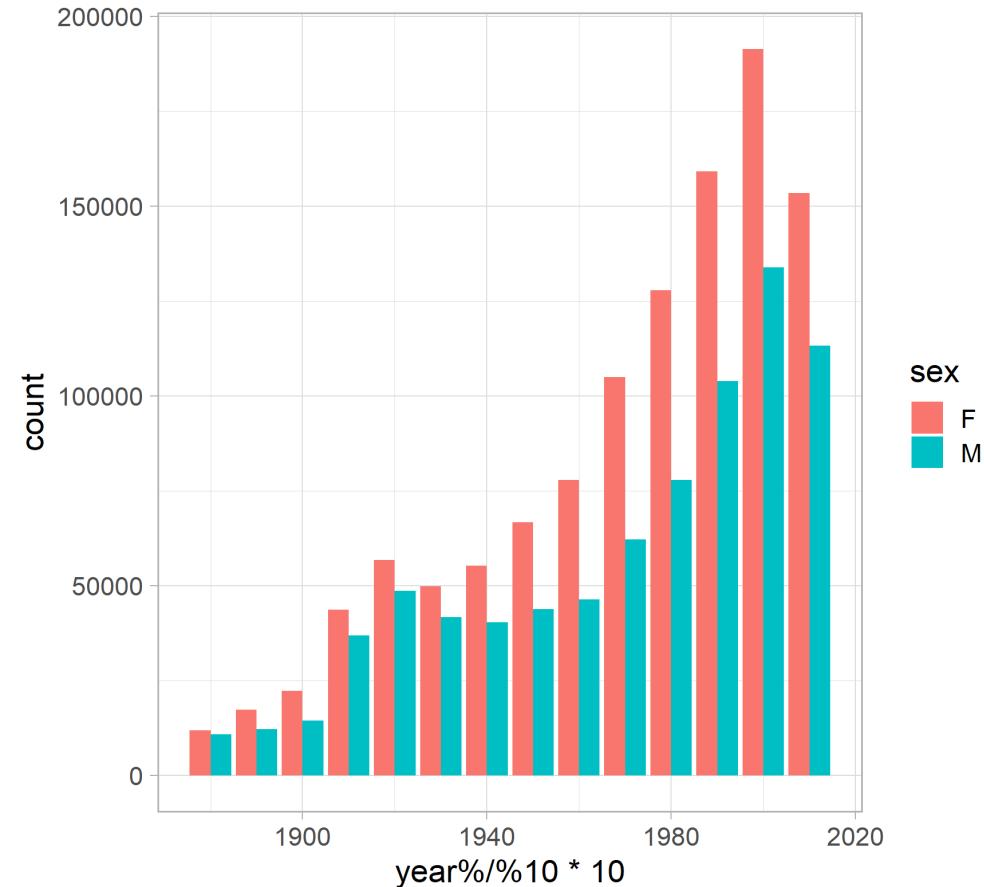
Your Turn: Dodged Bar Plots

```
ggplot(  
  babynames,  
  aes(  
    year %/% 10 * 10,  
    fill = sex  
  )) +  
  geom_bar(  
    stat = "count"  
  ) ## default: position = "stack"
```



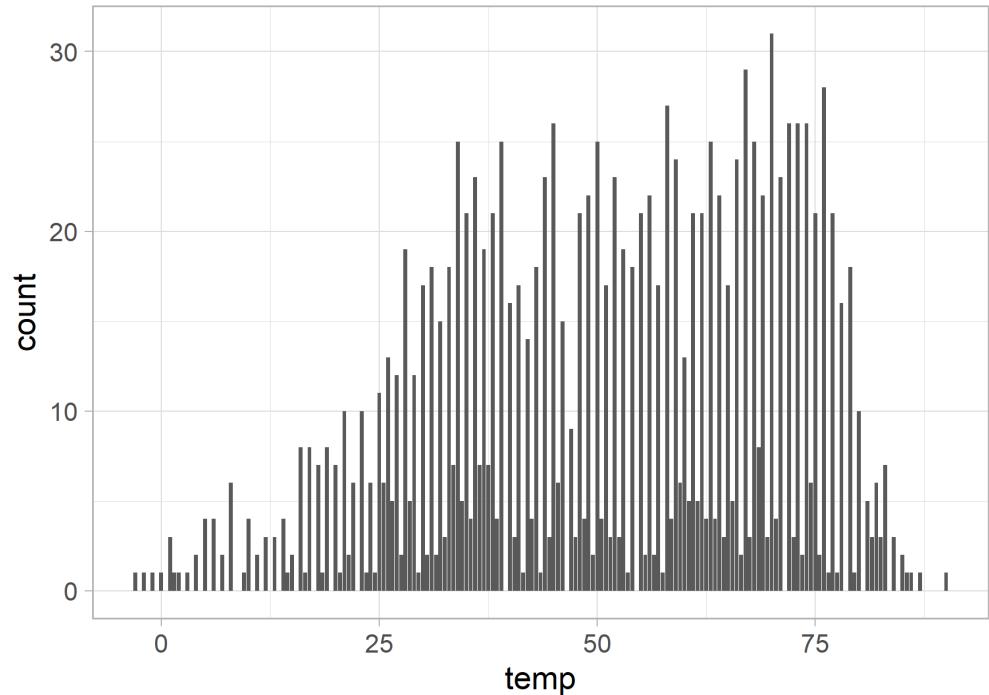
Your Turn: Dodged Bar Plots

```
ggplot(  
  babynames,  
  aes(  
    year %/% 10 * 10,  
    fill = sex  
  )) +  
  geom_bar(  
    stat = "count",  
    position = "dodge"  
  )
```

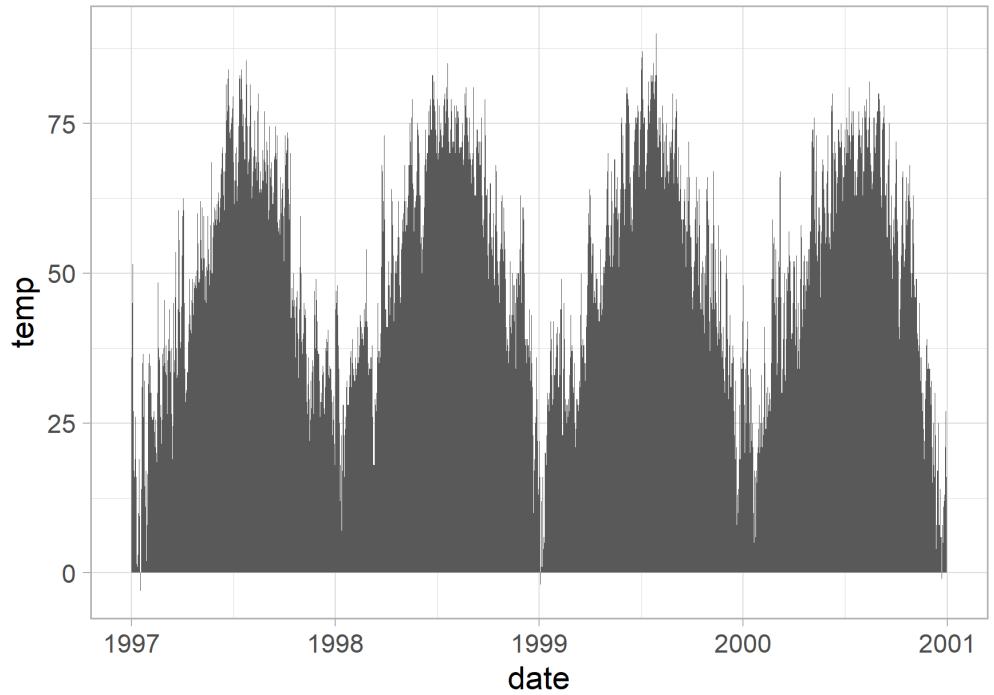


geom_bar()

```
ggplot(chic, aes(temp)) +  
  geom_bar()
```

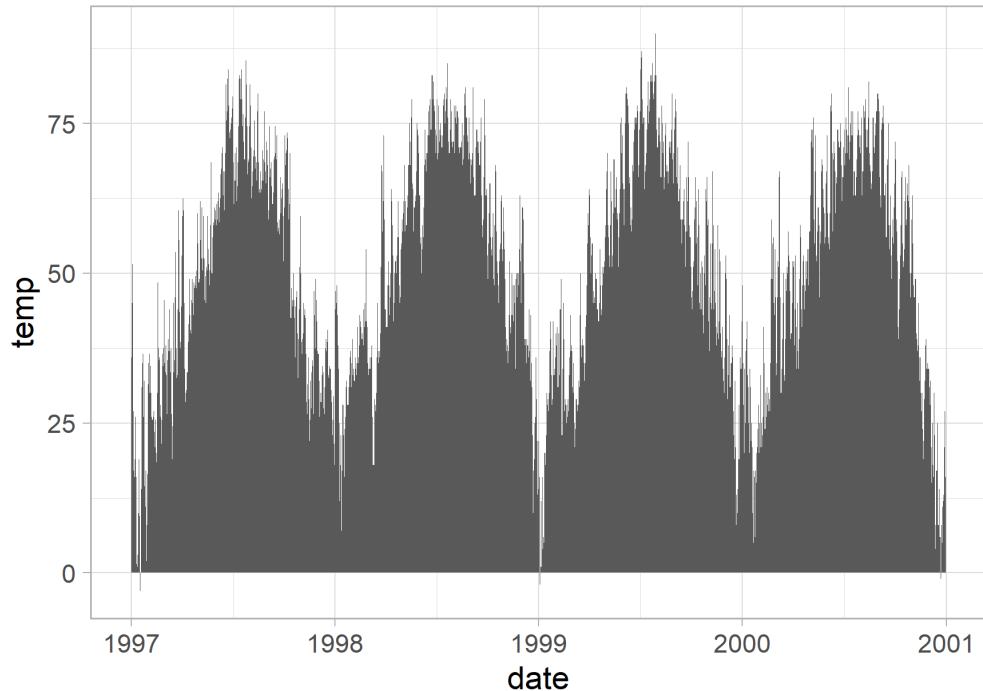


```
ggplot(chic, aes(date, temp)) +  
  geom_bar(stat = "identity")
```

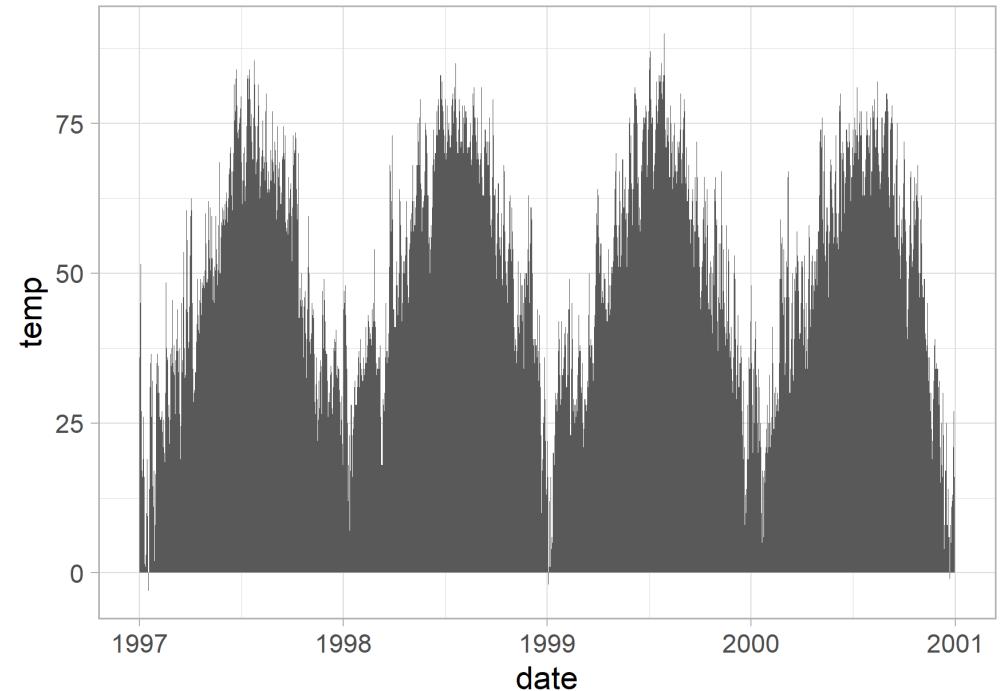


geom_bar() versus geom_col()

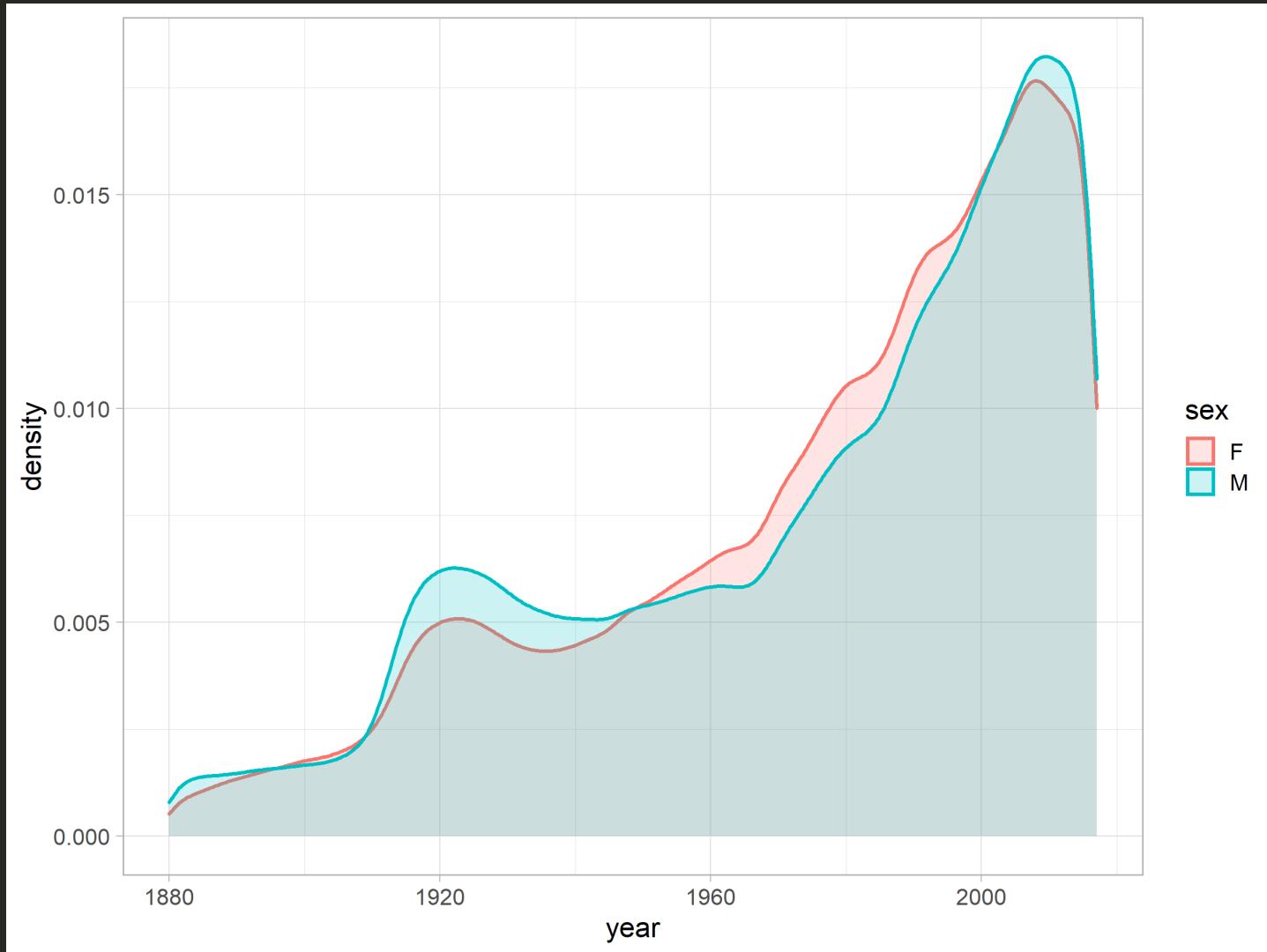
```
ggplot(chic, aes(date, temp)) +  
  geom_col()
```



```
ggplot(chic, aes(date, temp)) +  
  geom_bar(stat = "identity")
```

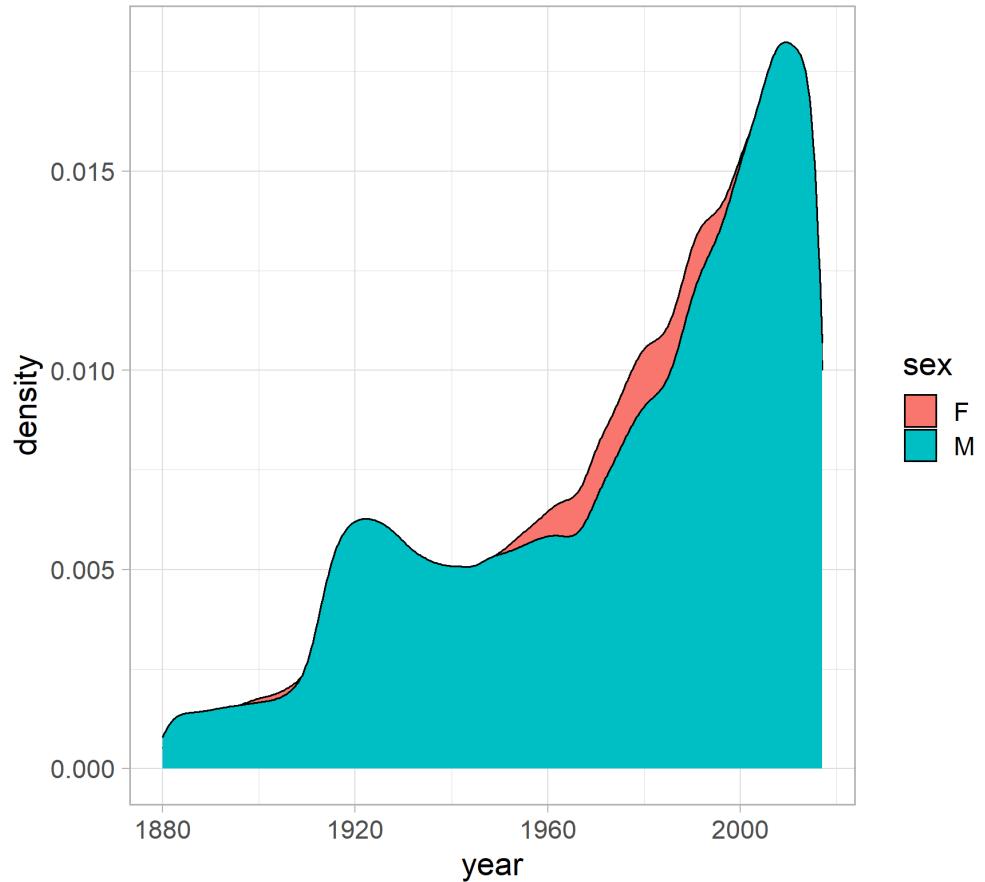


Your Turn!



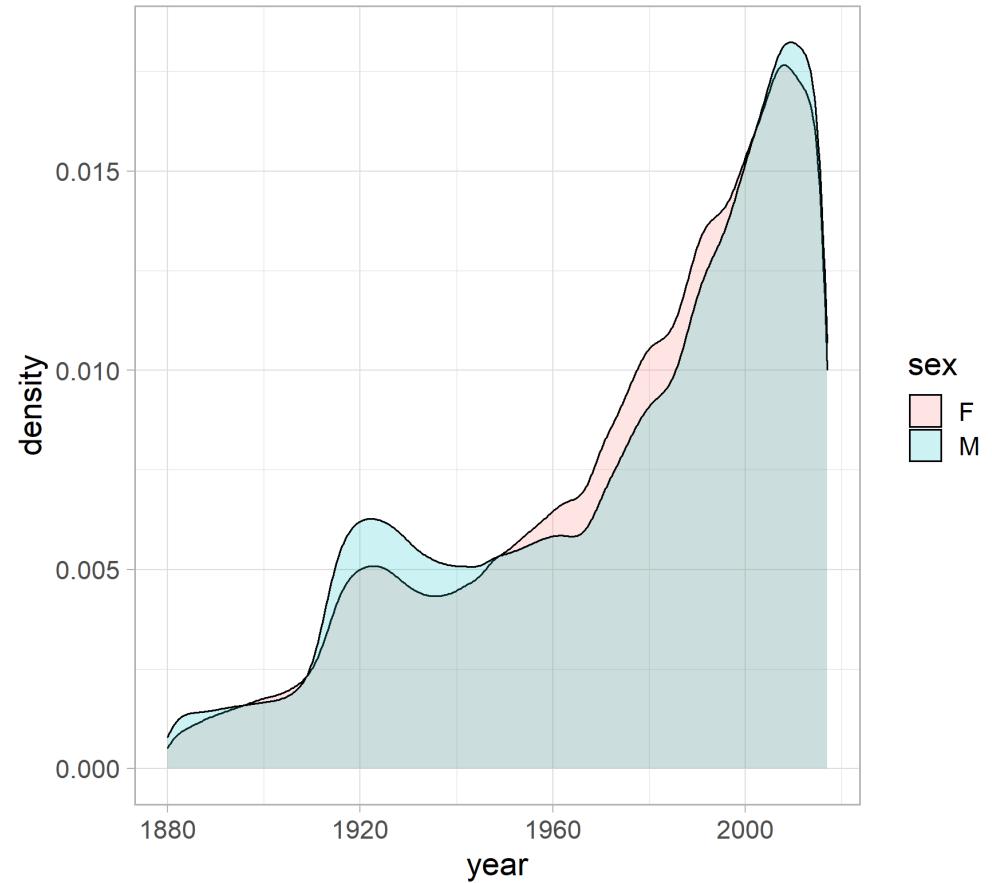
Your Turn: Density Plots

```
ggplot(  
  babynames,  
  aes(  
    year,  
    fill = sex  
  )) +  
  geom_density()
```



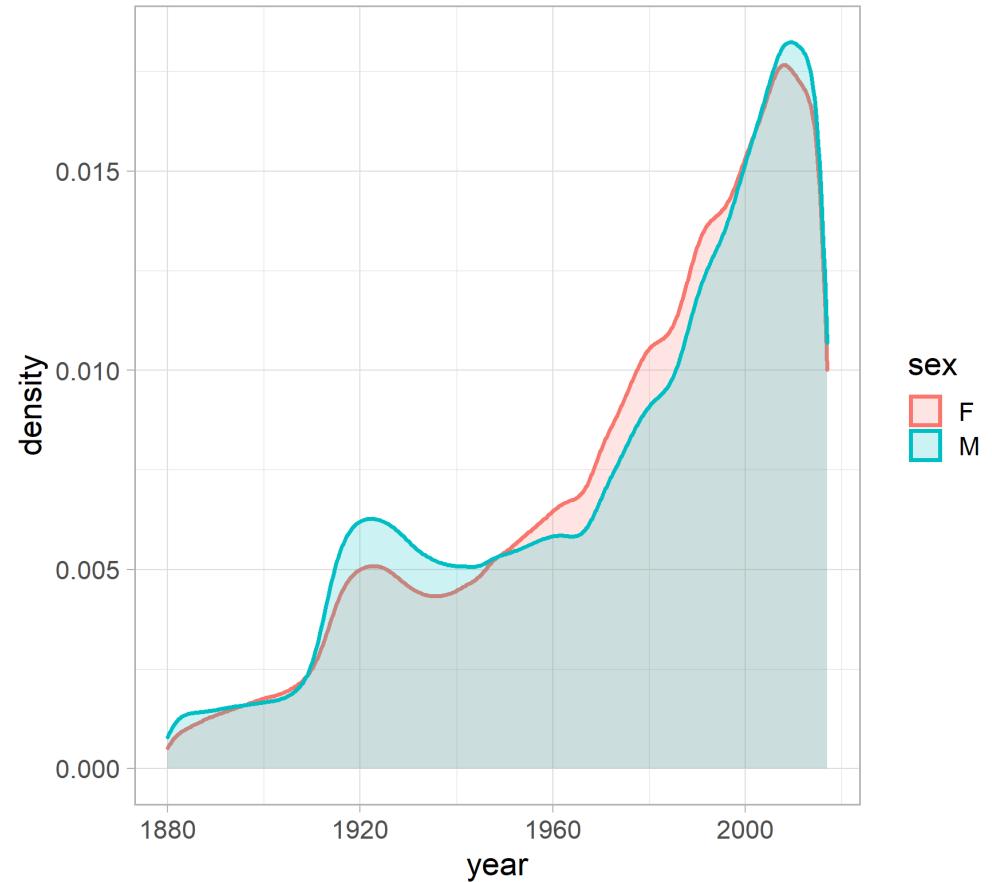
Your Turn: Density Plots

```
ggplot(  
  babynames,  
  aes(  
    year,  
    fill = sex  
  )) +  
  geom_density(  
    alpha = .2  
  )
```

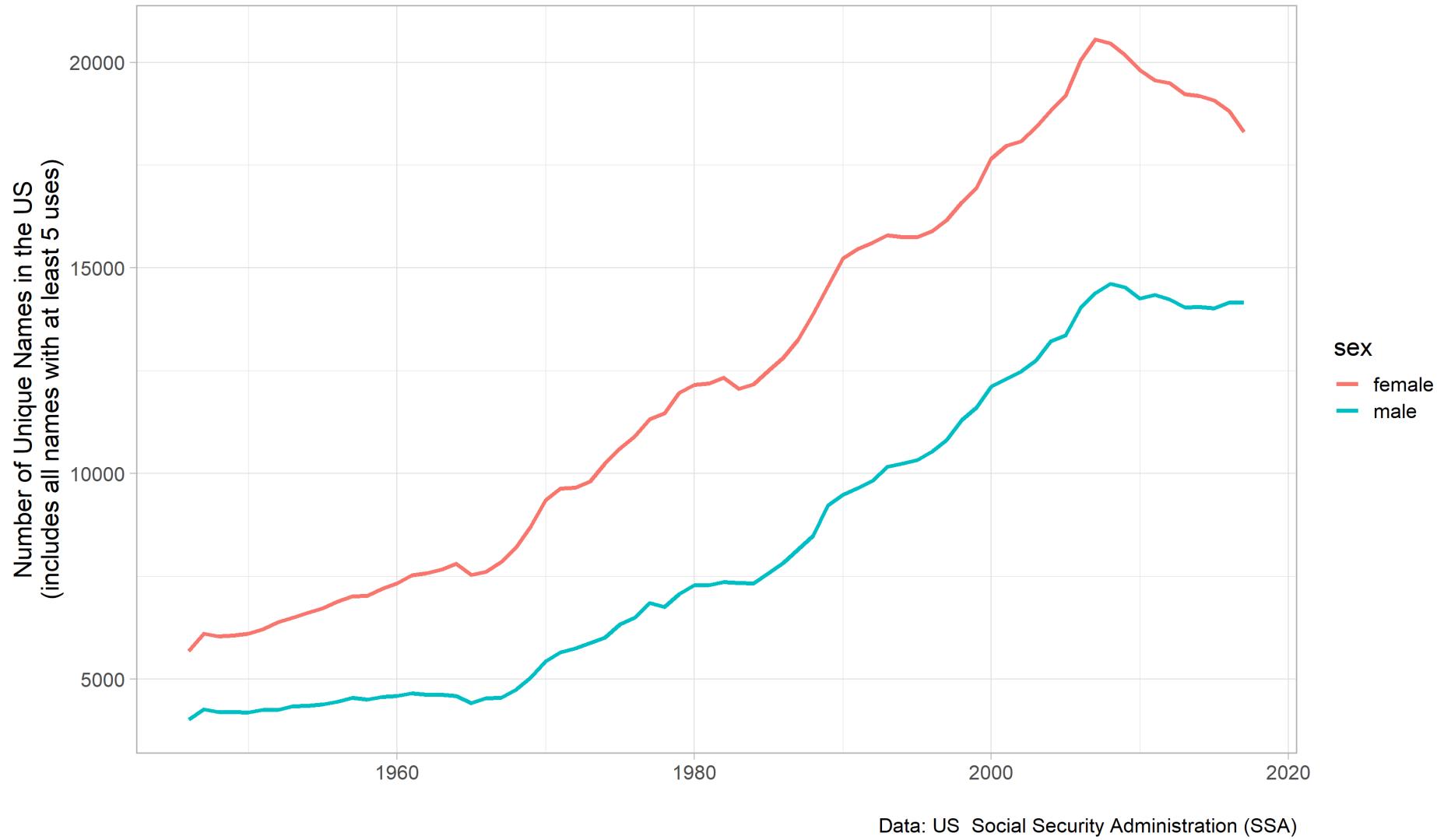


Your Turn: Density Plots

```
ggplot(  
  babynames,  
  aes(  
    year,  
    fill = sex,  
    color = sex  
  )) +  
  geom_density(  
    alpha = .2,  
    size = 1  
)
```

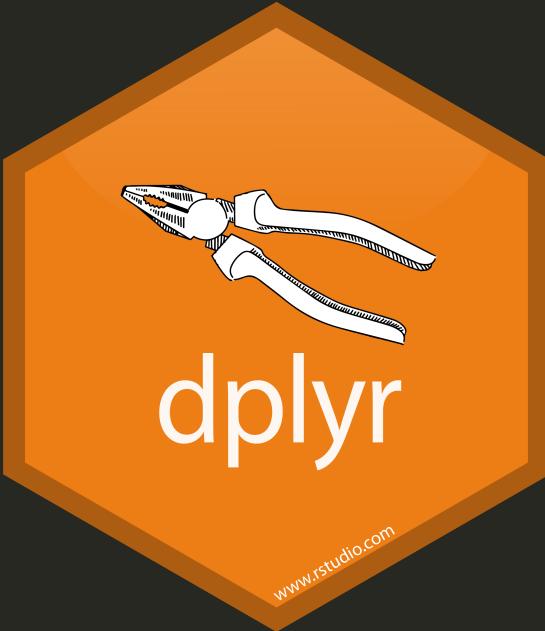


The Rise of Unique Baby Names in the Post-WWII Era



Your Turn!

- Instead of coding, discuss what would be a way to create this plot.
 - Which geom did I use here?
 - What are details you notice but we can't do until now?
 - What can you already achieve with your current knowledge?
If you like, give it a try afterwards!



The **dplyr** Package

Why should I care about `dplyr`?

- contains a set of convenient functions to perform common transformation and summary operations
- compared to base R:
 - syntax is more consistent
 - constrained options
 - always return a `data.frame` (actually, a `tibble`)
 - uses efficient data storage backends
 - can work with databases and data tables

The Main Verbs of dplyr

Function	Explanation
<code>filter()</code>	Pick rows with matching criteria
<code>select()</code>	Pick columns with matching criteria
<code>arrange()</code>	Reorder rows
<code>mutate()</code>	Create new variables
<code>summarize()</code> (or <code>summarise()</code>)	Sum up variables
<code>group_by()</code>	Create subsets

Consistent Syntax of `dplyr`

All functions take the same main arguments:

`verb(data, condition)`

The first argument is your data, subsequent arguments say what to do with the data frame, using the variable names.

filter()

Pick Rows with Matching Criteria



Pick Rows with Matching Criteria

`filter(data, condition)` allows you to select a subset of **rows**:

```
filter(babynames, year == 2000, n > 25000)
## # A tibble: 5 x 5
##   year sex   name      n    prop
##   <dbl> <chr> <chr> <int>  <dbl>
## 1 2000 F     Emily  25953 0.0130
## 2 2000 M     Jacob  34471 0.0165
## 3 2000 M     Michael 32035 0.0153
## 4 2000 M     Matthew 28572 0.0137
## 5 2000 M     Joshua  27538 0.0132
```

Pick Rows with Matching Criteria

`filter(data, condition)` allows you to select a subset of **rows**:

```
filter(babynames, year == 2000, n > 25000)
```

Equivalent code in base **R**:

```
babynames[babynames$year == 2000 & babynames$n > 25000, ]  
## # A tibble: 5 x 5  
##   year sex   name      n    prop  
##   <dbl> <chr> <chr> <int>  <dbl>  
## 1 2000 F   Emily  25953 0.0130  
## 2 2000 M   Jacob  34471 0.0165  
## 3 2000 M   Michael 32035 0.0153  
## 4 2000 M   Matthew 28572 0.0137  
## 5 2000 M   Joshua 27538 0.0132
```

Pick Rows with Matching Criteria

`filter(data, condition)` allows you to select a subset of `rows` → join filtering conditions with `&` and `|`:

```
filter(babynames, (name == "Cedric" | name == "Robert"))
## # A tibble: 426 x 5
##   year sex   name     n    prop
##   <dbl> <chr> <chr> <int>    <dbl>
## 1 1880 F    Robert  12 0.000123
## 2 1880 M    Robert 2415 0.0204
## 3 1881 F    Robert  9 0.0000910
## 4 1881 M    Robert 2140 0.0198
## 5 1882 F    Robert  12 0.000104
## 6 1882 M    Robert 2500 0.0205
## 7 1883 F    Robert 11 0.0000916
## 8 1883 M    Robert 2334 0.0208
## 9 1884 F    Robert  7 0.0000509
## 10 1884 M   Robert 2468 0.0201
## # ... with 416 more rows
```

Pick Rows with Matching Criteria

`filter(data, condition)` → join filtering conditions with `&` and `|`:

```
filter(babynames, (name == "Cedric" | name == "Robert") & year %in% 1900:1905)
## # A tibble: 15 x 5
##   year sex   name     n     prop
##   <dbl> <chr> <chr> <int>    <dbl>
## 1 1900 F    Robert  24 0.0000755
## 2 1900 M    Robert 3821 0.0236
## 3 1901 F    Robert  16 0.0000629
## 4 1901 M    Robert 2543 0.0220
## 5 1902 F    Robert  21 0.0000749
## 6 1902 M    Robert 3180 0.0240
## 7 1903 F    Robert  13 0.0000467
## 8 1903 M    Robert 3044 0.0235
## 9 1903 M    Cedric  7 0.0000541
## 10 1904 F   Robert  13 0.0000445
## 11 1904 M   Robert 3414 0.0246
## 12 1904 M   Cedric  8 0.0000578
## 13 1905 F   Robert  21 0.0000678
## 14 1905 M   Robert 3410 0.0238
## 15 1905 M   Cedric  5 0.0000349
```

`select()`

Pick Columns with Matching Criteria



Pick Columns with Matching Criteria

`select(data, condition)` allows you to select a subset of **columns**:

```
select(babynames, name, year, n)
## # A tibble: 1,924,665 x 3
##   name      year     n
##   <chr>    <dbl> <int>
## 1 Mary      1880  7065
## 2 Anna      1880  2604
## 3 Emma      1880  2003
## 4 Elizabeth 1880  1939
## 5 Minnie    1880  1746
## 6 Margaret  1880  1578
## 7 Ida       1880  1472
## 8 Alice     1880  1414
## 9 Bertha    1880  1320
## 10 Sarah    1880  1288
## # ... with 1,924,655 more rows
```

Pick Columns with Matching Criteria

`select(data, condition)` allows you to select a subset of **columns**:

```
select(babynames, name, year, n)
```

Equivalent code in base **R**:

```
babynames[, c("name", "year", "n")]
## # A tibble: 1,924,665 x 3
##   name      year     n
##   <chr>    <dbl> <int>
## 1 Mary      1880    7065
## 2 Anna      1880    2604
## 3 Emma      1880    2003
## 4 Elizabeth 1880    1939
## 5 Minnie    1880    1746
## 6 Margaret  1880    1578
## 7 Ida       1880    1472
## 8 Alice     1880    1414
## 9 Bertha    1880    1320
## 10 Sarah    1880    1288
## # ... with 1,924,655 more rows
```

Pick Columns with Matching Criteria

`select(data, condition)` allows you to select a subset of **columns**:

```
select(babynames, name, year, n)
## # A tibble: 1,924,665 x 3
##       name      year     n
##   <chr>    <dbl> <int>
## 1 Mary     1880  7065
## 2 Anna     1880  2604
## 3 Emma     1880  2003
## 4 Elizabeth 1880 1939
## 5 Minnie   1880  1746
## 6 Margaret 1880  1578
## 7 Ida      1880  1472
## 8 Alice    1880  1414
## 9 Bertha   1880  1320
## 10 Sarah   1880  1288
## # ... with 1,924,655 more rows
```

```
select(babynames, -prop, -sex)
## # A tibble: 1,924,665 x 3
##       year     name     n
##   <dbl> <chr> <int>
## 1 1880 Mary   7065
## 2 1880 Anna   2604
## 3 1880 Emma   2003
## 4 1880 Elizabeth 1939
## 5 1880 Minnie  1746
## 6 1880 Margaret 1578
## 7 1880 Ida    1472
## 8 1880 Alice   1414
## 9 1880 Bertha  1320
## 10 1880 Sarah   1288
## # ... with 1,924,655 more rows
```

arrange()

Reorder Rows



Reorder Rows

`arrange(data, condition)` allows you to order rows by variables:

```
arrange(babynames, prop)
## # A tibble: 1,924,665 x 5
##   year sex   name        n     prop
##   <dbl> <chr> <chr>    <int>    <dbl>
## 1 2007 M   Aaban      5 0.00000226
## 2 2007 M   Aareon     5 0.00000226
## 3 2007 M   Aaris      5 0.00000226
## 4 2007 M   Abd        5 0.00000226
## 5 2007 M   Abdulazeez 5 0.00000226
## 6 2007 M   Abdulhadi 5 0.00000226
## 7 2007 M   Abdulhamid 5 0.00000226
## 8 2007 M   Abdulkadir 5 0.00000226
## 9 2007 M   Abdulraheem 5 0.00000226
## 10 2007 M  Abdulrahim 5 0.00000226
## # ... with 1,924,655 more rows
```

Reorder Rows

`arrange(data, condition)` allows you to order rows by variables:

```
arrange(babynames, prop)
```

Equivalent code in base R:

```
babynames[order(babynames$prop), ]  
## # A tibble: 1,924,665 x 5  
##   year sex   name        n      prop  
##   <dbl> <chr> <chr>     <int>    <dbl>  
## 1 2007 M   Aaban      5 0.00000226  
## 2 2007 M   Aareon      5 0.00000226  
## 3 2007 M   Aaris       5 0.00000226  
## 4 2007 M   Abd         5 0.00000226  
## 5 2007 M   Abdulazeez  5 0.00000226  
## 6 2007 M   Abdulhadi   5 0.00000226  
## 7 2007 M   Abdulhamid  5 0.00000226  
## 8 2007 M   Abdulkadir  5 0.00000226  
## 9 2007 M   Abdulraheem 5 0.00000226  
## 10 2007 M  Abdulrahim  5 0.00000226  
## # ... with 1,924,655 more rows
```

Reorder Rows

`arrange(data, condition)` allows you to order rows by variables:

```
arrange(babynames, year, sex, -prop)
## # A tibble: 1,924,665 x 5
##   year sex   name       n    prop
##   <dbl> <chr> <chr>     <int>   <dbl>
## 1 1880 F   Mary      7065 0.0724
## 2 1880 F   Anna     2604 0.0267
## 3 1880 F   Emma     2003 0.0205
## 4 1880 F   Elizabeth 1939 0.0199
## 5 1880 F   Minnie    1746 0.0179
## 6 1880 F   Margaret  1578 0.0162
## 7 1880 F   Ida       1472 0.0151
## 8 1880 F   Alice     1414 0.0145
## 9 1880 F   Bertha    1320 0.0135
## 10 1880 F  Sarah     1288 0.0132
## # ... with 1,924,655 more rows
```

mutate()

Create New Variables





Illustration by Allison Horst (github.com/allisonhorst/stats-illustrations)

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mutate(babynames, decade = year %/% 10 * 10)
## # A tibble: 1,924,665 x 6
##   year sex   name      n    prop decade
##   <dbl> <chr> <chr>  <int>   <dbl>   <dbl>
## 1 1880 F     Mary    7065 0.0724 1880
## 2 1880 F     Anna    2604 0.0267 1880
## 3 1880 F     Emma    2003 0.0205 1880
## 4 1880 F     Elizabeth 1939 0.0199 1880
## 5 1880 F     Minnie   1746 0.0179 1880
## 6 1880 F     Margaret 1578 0.0162 1880
## 7 1880 F     Ida      1472 0.0151 1880
## 8 1880 F     Alice    1414 0.0145 1880
## 9 1880 F     Bertha   1320 0.0135 1880
## 10 1880 F    Sarah    1288 0.0132 1880
## # ... with 1,924,655 more rows
```

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mutate(babynames, decade = year %/% 10 * 10)
```

Equivalent code in base R:

```
transform(babynames, decade = year %/% 10 * 10)
##      year sex        name    n      prop decade
## 1  1880   F       Mary 7065 0.07238359  1880
## 2  1880   F       Anna 2604 0.02667896  1880
## 3  1880   F       Emma 2003 0.02052149  1880
## 4  1880   F   Elizabeth 1939 0.01986579  1880
## 5  1880   F      Minnie 1746 0.01788843  1880
## 6  1880   F   Margaret 1578 0.01616720  1880
## 7  1880   F       Ida 1472 0.01508119  1880
## 8  1880   F       Alice 1414 0.01448696  1880
## 9  1880   F      Bertha 1320 0.01352390  1880
## 10 1880   F       Sarah 1288 0.01319605  1880
## 11 1880   F       Annie 1258 0.01288868  1880
## 12 1880   F       Clara 1226 0.01256083  1880
## 13 1880   F       Ella 1156 0.01184366  1880
```

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mutate(babynames, name = str_to_lower(name))
## # A tibble: 1,924,665 x 5
##   year sex   name       n    prop
##   <dbl> <chr> <chr>     <int>   <dbl>
## 1 1880 F   mary      7065 0.0724
## 2 1880 F   anna      2604 0.0267
## 3 1880 F   emma      2003 0.0205
## 4 1880 F   elizabeth 1939 0.0199
## 5 1880 F   minnie    1746 0.0179
## 6 1880 F   margaret  1578 0.0162
## 7 1880 F   ida        1472 0.0151
## 8 1880 F   alice      1414 0.0145
## 9 1880 F   bertha    1320 0.0135
## 10 1880 F  sarah     1288 0.0132
## # ... with 1,924,655 more rows
```

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mutate(babynames, note = "Please check!")
## # A tibble: 1,924,665 x 6
##   year sex   name       n    prop note
##   <dbl> <chr> <chr>     <int>  <dbl> <chr>
## 1 1880 F   Mary      7065 0.0724 Please check!
## 2 1880 F   Anna      2604 0.0267 Please check!
## 3 1880 F   Emma      2003 0.0205 Please check!
## 4 1880 F   Elizabeth 1939 0.0199 Please check!
## 5 1880 F   Minnie    1746 0.0179 Please check!
## 6 1880 F   Margaret  1578 0.0162 Please check!
## 7 1880 F   Ida       1472 0.0151 Please check!
## 8 1880 F   Alice     1414 0.0145 Please check!
## 9 1880 F   Bertha    1320 0.0135 Please check!
## 10 1880 F  Sarah     1288 0.0132 Please check!
## # ... with 1,924,655 more rows
```

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mutate(babynames, note = if_else(year < 1900, "Please check!", "Ok"))
## # A tibble: 1,924,665 x 6
##   year sex   name       n    prop note
##   <dbl> <chr> <chr>     <int>  <dbl> <chr>
## 1 1880 F   Mary      7065 0.0724 Please check!
## 2 1880 F   Anna      2604 0.0267 Please check!
## 3 1880 F   Emma      2003 0.0205 Please check!
## 4 1880 F   Elizabeth 1939 0.0199 Please check!
## 5 1880 F   Minnie    1746 0.0179 Please check!
## 6 1880 F   Margaret  1578 0.0162 Please check!
## 7 1880 F   Ida       1472 0.0151 Please check!
## 8 1880 F   Alice     1414 0.0145 Please check!
## 9 1880 F   Bertha    1320 0.0135 Please check!
## 10 1880 F  Sarah     1288 0.0132 Please check!
## # ... with 1,924,655 more rows
```

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mutate(babynames, id = row_number())
## # A tibble: 1,924,665 x 6
##   year sex   name       n    prop     id
##   <dbl> <chr> <chr> <int>  <dbl> <int>
## 1 1880 F   Mary     7065 0.0724     1
## 2 1880 F   Anna    2604 0.0267     2
## 3 1880 F   Emma    2003 0.0205     3
## 4 1880 F   Elizabeth 1939 0.0199     4
## 5 1880 F   Minnie   1746 0.0179     5
## 6 1880 F   Margaret 1578 0.0162     6
## 7 1880 F   Ida      1472 0.0151     7
## 8 1880 F   Alice    1414 0.0145     8
## 9 1880 F   Bertha   1320 0.0135     9
## 10 1880 F  Sarah    1288 0.0132    10
## # ... with 1,924,655 more rows
```

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mpg2 <- select(mpg, model, hwy, cty) ## only for visualization
mutate(mpg2, diff = hwy - cty, perc = diff / hwy)
## # A tibble: 234 x 5
##       model      hwy     cty   diff   perc
##       <chr>    <int>  <int>  <dbl>
## 1 a4            29     18     11  0.379
## 2 a4            29     21      8  0.276
## 3 a4            31     20     11  0.355
## 4 a4            30     21      9  0.3
## 5 a4            26     16     10  0.385
## 6 a4            26     18      8  0.308
## 7 a4            27     18      9  0.333
## 8 a4 quattro   26     18      8  0.308
## 9 a4 quattro   25     16      9  0.36
## 10 a4 quattro  28     20      8  0.286
## # ... with 224 more rows
```

Create New Variables

`mutate(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
mpg2 <- select(mpg, model, hwy, cty) ## only for visualization  
mutate(mpg2, diff = hwy - cty, perc = diff / hwy)
```

Equivalent code in base R:

```
mpg_diff <- transform(mpg2, diff = hwy - cty)  
transform(mpg_diff, perc = diff / hwy)  
##  
## 1           model hwy cty diff      perc  
## 2           a4    29   18   11 0.3793103  
## 3           a4    29   21    8 0.2758621  
## 4           a4    31   20   11 0.3548387  
## 5           a4    30   21    9 0.3000000  
## 6           a4    26   16   10 0.3846154  
## 7           a4    26   18    8 0.3076923  
## 8           a4    27   18    9 0.3333333  
## 9           a4    quattro 26   18    8 0.3076923  
## 10          a4    quattro 25   16    9 0.3600000  
## 10          a4    quattro 28   20    8 0.2857143  
## 11          a4    quattro 27   19    8 0.2962963
```

Create New Variables

`mutate_all(data, condition)` allows you to transform all columns in one step:

```
mutate_all(mpg2, mean, na.rm = TRUE)
## # A tibble: 234 x 3
##   model     hwy     cty
##   <dbl> <dbl> <dbl>
## 1 NA     23.4   16.9
## 2 NA     23.4   16.9
## 3 NA     23.4   16.9
## 4 NA     23.4   16.9
## 5 NA     23.4   16.9
## 6 NA     23.4   16.9
## 7 NA     23.4   16.9
## 8 NA     23.4   16.9
## 9 NA     23.4   16.9
## 10 NA    23.4   16.9
## # ... with 224 more rows
```

Create New Variables

`mutate_if(data, condition)` allows you to conditionally create several new columns in one step:

```
mutate_if(mpg2, is.numeric, as.character)
## # A tibble: 234 x 3
##   model      hwy     cty
##   <chr>     <chr>  <chr>
## 1 a4        29     18
## 2 a4        29     21
## 3 a4        31     20
## 4 a4        30     21
## 5 a4        26     16
## 6 a4        26     18
## 7 a4        27     18
## 8 a4 quattro 26     18
## 9 a4 quattro 25     16
## 10 a4 quattro 28    20
## # ... with 224 more rows
```

Create New Variables

`mutate_if(data, condition)` allows you to conditionally create several new columns in one step:

```
mutate_if(mpg2, is.numeric, list(avg = mean, log_2 = log2))
## # A tibble: 234 x 7
##   model      hwy     cty hwy_avg cty_avg hwy_log_2 cty_log_2
##   <chr>     <int>   <int>    <dbl>    <dbl>      <dbl>      <dbl>
## 1 a4          29      18     23.4    16.9       4.86      4.17
## 2 a4          29      21     23.4    16.9       4.86      4.39
## 3 a4          31      20     23.4    16.9       4.95      4.32
## 4 a4          30      21     23.4    16.9       4.91      4.39
## 5 a4          26      16     23.4    16.9       4.70      4
## 6 a4          26      18     23.4    16.9       4.70      4.17
## 7 a4          27      18     23.4    16.9       4.75      4.17
## 8 a4 quattro  26      18     23.4    16.9       4.70      4.17
## 9 a4 quattro  25      16     23.4    16.9       4.64      4
## 10 a4 quattro 28      20     23.4    16.9       4.81      4.32
## # ... with 224 more rows
```

Create New Variables

`mutate_if(data, condition)` allows you to conditionally create several new columns in one step:

```
mutate_if(mpg2, is.numeric, list(~mean(., na.rm = T), ~log2(.)))  
## # A tibble: 234 x 7  
##   model      hwy     cty hwy_mean cty_mean hwy_log2 cty_log2  
##   <chr>     <int>   <int>    <dbl>    <dbl>    <dbl>    <dbl>  
## 1 a4         29      18     23.4     16.9     4.86     4.17  
## 2 a4         29      21     23.4     16.9     4.86     4.39  
## 3 a4         31      20     23.4     16.9     4.95     4.32  
## 4 a4         30      21     23.4     16.9     4.91     4.39  
## 5 a4         26      16     23.4     16.9     4.70     4  
## 6 a4         26      18     23.4     16.9     4.70     4.17  
## 7 a4         27      18     23.4     16.9     4.75     4.17  
## 8 a4 quattro 26      18     23.4     16.9     4.70     4.17  
## 9 a4 quattro 25      16     23.4     16.9     4.64     4  
## 10 a4 quattro 28      20     23.4     16.9     4.81     4.32  
## # ... with 224 more rows
```

Create New Variables

`mutate_at(data, condition)` allows you to conditionally transform several columns in one step:

```
mutate_at(mpg2, c("hwy", "cty"), mean, na.rm = TRUE)
## # A tibble: 234 x 3
##   model      hwy     cty
##   <chr>    <dbl>   <dbl>
## 1 a4        23.4   16.9
## 2 a4        23.4   16.9
## 3 a4        23.4   16.9
## 4 a4        23.4   16.9
## 5 a4        23.4   16.9
## 6 a4        23.4   16.9
## 7 a4        23.4   16.9
## 8 a4 quattro 23.4   16.9
## 9 a4 quattro 23.4   16.9
## 10 a4 quattro 23.4   16.9
## # ... with 224 more rows
```

Create New Variables

`mutate_all(data, condition)` allows you to transform all columns in one step:

```
mutate_all(mpg2, list(~mean(., na.rm = T), ~nchar(.)))  
## # A tibble: 234 x 9  
##   model  hwy   cty model_mean hwy_mean cty_mean model_nchar hwy_nchar  
##   <chr> <int> <int>     <dbl>    <dbl>    <dbl>      <int>      <int>  
## 1 a4        29     18       NA     23.4    16.9        2         2  
## 2 a4        29     21       NA     23.4    16.9        2         2  
## 3 a4        31     20       NA     23.4    16.9        2         2  
## 4 a4        30     21       NA     23.4    16.9        2         2  
## 5 a4        26     16       NA     23.4    16.9        2         2  
## 6 a4        26     18       NA     23.4    16.9        2         2  
## 7 a4        27     18       NA     23.4    16.9        2         2  
## 8 a4 q~      26     18       NA     23.4    16.9       10         2  
## 9 a4 q~      25     16       NA     23.4    16.9       10         2  
## 10 a4 q~     28     20       NA     23.4    16.9       10         2  
## # ... with 224 more rows, and 1 more variable: cty_nchar <int>
```

summarize()

Sum Up Variables



Sum Up Variables

`summarize(data, condition)` allows you to calculate summary statistics for particular variables:

```
summarize(babynames, unique_children = sum(n, na.rm = TRUE))
## # A tibble: 1 x 1
##   unique_children
##             <int>
## 1            348120517
```

Sum Up Variables

`summarize(data, condition)` allows you to create new variables (columns) based or not based on other variables:

```
summarize(babynames, unique_children = sum(n, na.rm = TRUE))  
## # A tibble: 1 x 1  
##   unique_children  
##       <int>  
## 1     348120517
```

Equivalent code in base R:

```
sum(babynames$n, na.rm = TRUE)  
## [1] 348120517
```

Sum Up Variables

`summarize(data, condition)` allows you to calculate summary statistics for particular variables:

```
summarize(babynames, n = n())
## # A tibble: 1 x 1
##       n
##   <int>
## 1 1924665
```

group_by()

Create Subsets



Create Subsets

`group_by(data, condition)` allows you to break down your data into specified groups based on variables:

```
group_by(babynames, sex)
## # A tibble: 1,924,665 x 5
## # Groups:   sex [2]
##       year sex     name      n    prop
##       <dbl> <chr> <chr> <int>  <dbl>
## 1 1880 F     Mary     7065 0.0724
## 2 1880 F     Anna    2604 0.0267
## 3 1880 F     Emma    2003 0.0205
## 4 1880 F     Elizabeth 1939 0.0199
## 5 1880 F     Minnie   1746 0.0179
## 6 1880 F     Margaret 1578 0.0162
## 7 1880 F     Ida     1472 0.0151
## 8 1880 F     Alice    1414 0.0145
## 9 1880 F     Bertha   1320 0.0135
## 10 1880 F    Sarah    1288 0.0132
## # ... with 1,924,655 more rows
```

Create Subsets

`group_by(data, condition)` allows you to break down your data into specified groups based on variables:

```
group_by(babynames, sex, year)
## # A tibble: 1,924,665 x 5
## # Groups:   sex, year [276]
##       year sex   name      n    prop
##       <dbl> <chr> <chr> <int>  <dbl>
## 1 1880 F   Mary     7065 0.0724
## 2 1880 F   Anna    2604 0.0267
## 3 1880 F   Emma    2003 0.0205
## 4 1880 F   Elizabeth 1939 0.0199
## 5 1880 F   Minnie   1746 0.0179
## 6 1880 F   Margaret 1578 0.0162
## 7 1880 F   Ida     1472 0.0151
## 8 1880 F   Alice    1414 0.0145
## 9 1880 F   Bertha   1320 0.0135
## 10 1880 F  Sarah    1288 0.0132
## # ... with 1,924,655 more rows
```

Remove Subsets

... and `ungroup()` lets you remove any subsets:

```
grouped <- group_by(babynames, sex, year)
ungroup(grouped)
## # A tibble: 1,924,665 x 5
##       year   sex   name        n    prop
##   <dbl> <chr> <chr>    <int>    <dbl>
## 1 1880 F     Mary      7065 0.0724
## 2 1880 F     Anna     2604 0.0267
## 3 1880 F     Emma     2003 0.0205
## 4 1880 F     Elizabeth 1939 0.0199
## 5 1880 F     Minnie    1746 0.0179
## 6 1880 F     Margaret 1578 0.0162
## 7 1880 F     Ida      1472 0.0151
## 8 1880 F     Alice     1414 0.0145
## 9 1880 F     Bertha    1320 0.0135
## 10 1880 F    Sarah     1288 0.0132
## # ... with 1,924,655 more rows
```

A stylized illustration of a superhero character with a dark blue mask featuring white star-like markings. He has a red cape and is shown from the chest up, flying towards the viewer against a background of radiating grey and white lines resembling a sunburst or light rays.

THE SUPERPOWER OF `group_by()`

Verbs will be automatically applied "by group"!

group_by() SUPERPOWER!

group_by() in combination with other dplyr verbs allows you to calculate summary statistics for specified groups:

```
names_group <- group_by(babynames, sex)
names_group
## # A tibble: 1,924,665 x 5
## # Groups:   sex [2]
##       year sex   name      n    prop
##       <dbl> <chr> <chr>  <int>  <dbl>
## 1 1880 F   Mary     7065 0.0724
## 2 1880 F   Anna     2604 0.0267
## 3 1880 F   Emma     2003 0.0205
## 4 1880 F   Elizabeth 1939 0.0199
## 5 1880 F   Minnie    1746 0.0179
## 6 1880 F   Margaret 1578 0.0162
## 7 1880 F   Ida      1472 0.0151
## 8 1880 F   Alice     1414 0.0145
## 9 1880 F   Bertha    1320 0.0135
## 10 1880 F  Sarah     1288 0.0132
## # ... with 1,924,655 more rows
```

group_by() SUPERPOWER!

group_by() in combination with other dplyr verbs allows you to calculate summary statistics for specified groups:

```
names_group <- group_by(babynames, sex)
summarize(names_group, sum = sum(n))
## # A tibble: 2 x 2
##   sex      sum
##   <chr>    <int>
## 1 F        172371079
## 2 M        175749438
```

group_by() SUPERPOWER!

group_by() in combination with other dplyr verbs allows you to calculate summary statistics for specified groups:

```
names_cent <- filter(babynames, year %in% c(1890, 1900))
names_cent <- group_by(names_cent, sex, year)
names_cent
## # A tibble: 6,425 x 5
## # Groups:   sex, year [4]
##       year sex     name       n    prop
##       <dbl> <chr> <chr>     <int>   <dbl>
## 1 1890 F     Mary     12078 0.0599
## 2 1890 F     Anna     5233 0.0259
## 3 1890 F     Elizabeth 3112 0.0154
## 4 1890 F     Margaret 3100 0.0154
## 5 1890 F     Emma     2980 0.0148
## 6 1890 F     Florence 2744 0.0136
## 7 1890 F     Ethel    2718 0.0135
## 8 1890 F     Minnie   2650 0.0131
## 9 1890 F     Clara    2496 0.0124
## 10 1890 F    Bertha   2388 0.0118
## # ... with 6,415 more rows
```

group_by() SUPERPOWER!

group_by() in combination with other dplyr verbs allows you to calculate summary statistics for specified groups:

```
names_cent <- filter(babynames, year %in% c(1890, 1900))
names_cent <- group_by(names_cent, sex, year)
names_cent <- mutate(names_cent, sum = sum(n))
names_cent <- group_by(names_cent, sex, year, name)
names_cent

## # A tibble: 6,425 x 6
## # Groups:   sex, year, name [6,425]
##       year  sex    name        n    prop     sum
##       <dbl> <chr> <chr>    <int>  <dbl>   <int>
## 1 1890 F Mary 12078 0.0599 190376
## 2 1890 F Anna 5233 0.0259 190376
## 3 1890 F Elizabeth 3112 0.0154 190376
## 4 1890 F Margaret 3100 0.0154 190376
## 5 1890 F Emma 2980 0.0148 190376
## 6 1890 F Florence 2744 0.0136 190376
## 7 1890 F Ethel 2718 0.0135 190376
## 8 1890 F Minnie 2650 0.0131 190376
## 9 1890 F Clara 2496 0.0124 190376
## 10 1890 F Bertha 2388 0.0118 190376
## # ... with 6,415 more rows
```

group_by() SUPERPOWER!

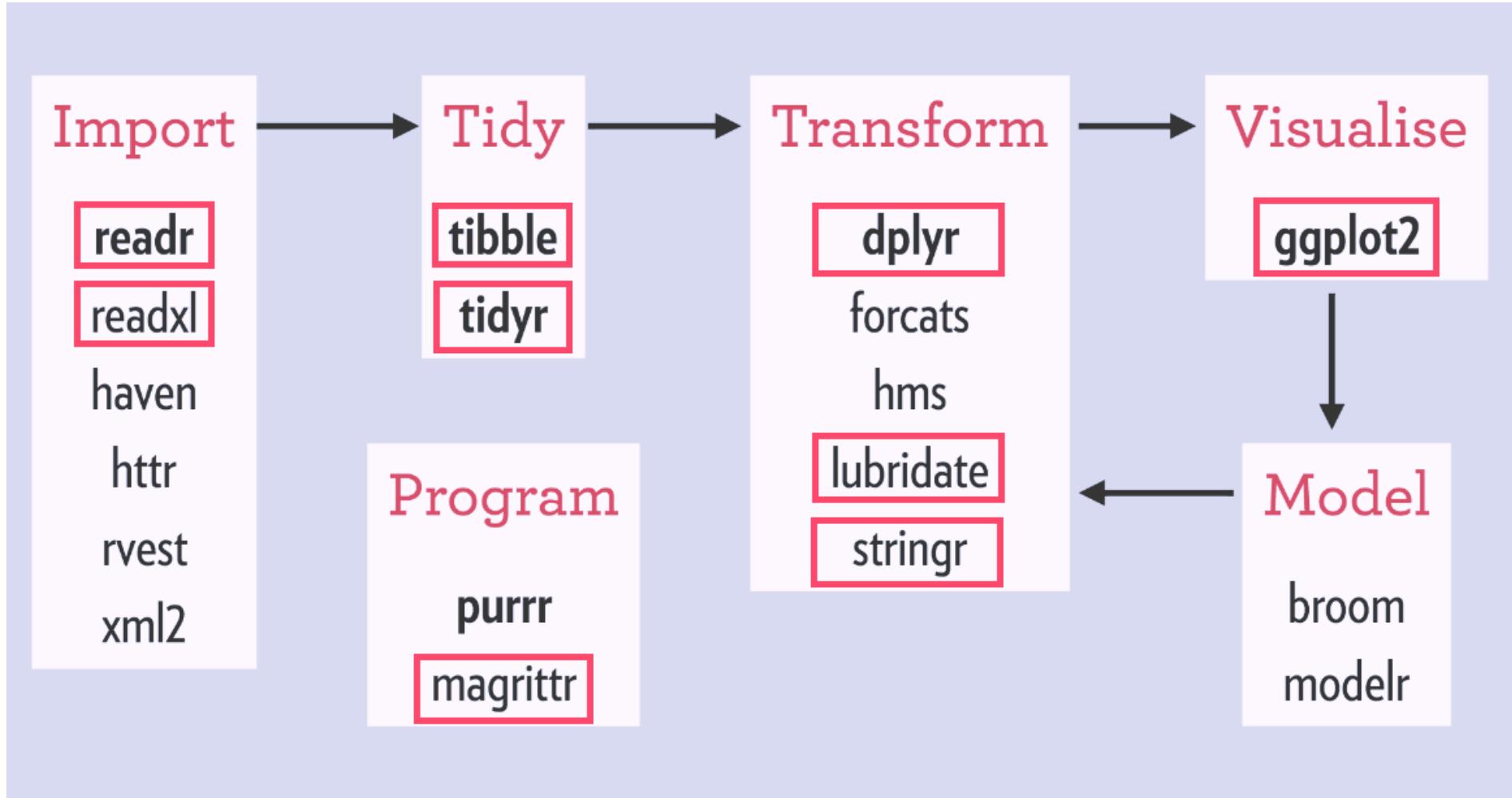
group_by() in combination with other dplyr verbs allows you to calculate summary statistics for specified groups:

```
names_cent <- filter(babynames, year %in% c(1890, 1900))
names_cent <- group_by(names_cent, sex, year)
names_cent <- mutate(names_cent, sum = sum(n))
names_cent <- group_by(names_cent, sex, year, name)
names_sum <- summarize(names_cent, prop = prop, prop_check = sum(n) / unique(sum))
arrange(names_sum, desc(sex), -prop_check)
## # A tibble: 6,425 x 5
## # Groups:   sex, year [4]
##       sex     year   name     prop prop_check
##       <chr>  <dbl> <chr>    <dbl>      <dbl>
## 1 M         1890 John     0.0710     0.0766
## 2 M         1890 William  0.0626     0.0675
## 3 M         1900 John     0.0606     0.0653
## 4 M         1900 William  0.0529     0.0570
## 5 M         1900 James    0.0447     0.0481
## 6 M         1890 James    0.0426     0.0459
## 7 M         1890 George   0.0372     0.0402
## 8 M         1890 Charles  0.0339     0.0366
## 9 M         1900 George   0.0333     0.0359
## 10 M        1890 Frank    0.0257     0.0277
```



The **magrittr** Package

The Typical Data Science Project Flow



Source: www.rviews.rstudio.com/post/2017-06-09-What-is-the-tidyverse_files/tidyverse6.png

Readability of Code

Have a look again at this chunk of code:

```
names_cent <- filter(babynames, year %in% c(1890, 1900))
names_cent <- group_by(names_cent, sex, year)
names_cent <- mutate(names_cent, sum = sum(n))
names_cent <- group_by(names_cent, sex, year, name)
names_sum <- summarize(names_cent, prop = prop, prop_check = sum(n) / unique(sum))
names_1990_sum <- ungroup(names_sum)
names_sum <- arrange(names_sum, desc(sex), -prop_check)
```

There is a lot redundancy and also no need to save each step as an object.

Readability of Code

One could rewrite the code like this:

```
names_cent <-  
  group_by(mutate(group_by(filter(babynames, year %in% c(1890, 1900)), sex, year), sum  
  
names_sum <-  
  ungroup(arrange(summarize(names_cent, prop = prop, prop_check = sum(n) / unique(sum  
  
names_sum  
## # A tibble: 6,425 x 5  
##   sex     year name      prop prop_check  
##   <chr>  <dbl> <chr>    <dbl>      <dbl>  
## 1 M       1890 John     0.0710     0.0766  
## 2 M       1890 William  0.0626     0.0675  
## 3 M       1900 John     0.0606     0.0653  
## 4 M       1900 William  0.0529     0.0570  
## 5 M       1900 James    0.0447     0.0481  
## 6 M       1890 James    0.0426     0.0459  
## 7 M       1890 George   0.0372     0.0402  
## 8 M       1890 Charles  0.0339     0.0366  
## 9 M       1900 George   0.0333     0.0359  
## 10 M      1890 Frank    0.0257     0.0277  
## # ... with 6,415 more rows
```

Readability of Code

One could rewrite the code like this (Part 1)

```
names_cent <-
  group_by(
    mutate(
      group_by(
        filter(
          babynames,
          year %in% c(1890, 1900)
        ),
        sex, year
      ),
      sum = sum(n)
    ),
    sex, year, name
  )
```

Readability of Code

One could rewrite the code like this (Part 2)

```
names_sum <-
  ungroup(
    arrange(
      summarize(
        names_cent,
        prop = prop,
        prop_check = sum(n) / unique(sum)
      ),
      desc(sex), -prop_check
    )
  )
```

```
physalia(  
  bvg(  
    breakfast(  
      shower(  
        wake_up(  
          Cédric,  
          7.0  
        ),  
        temp == 38  
      ),  
      c("cornflakes", "tea"),  
    ),  
    price = "EUR2.90",  
    delay = 10  
  ),  
  course = "Data Visualization in R with ggplot2"  
)
```

```
Cédric %>%
  wake_up(7.0) %>%
  shower(temp == 38) %>%
  breakfast(c("cereals", "tea")) %>%
  bvg(
    price = "EUR2.90",
    delay = 10
  ) %>%
  physalia(
    course = "Data Visualization in R with ggplot2"
  )
```

```
Cédric %>%
  wake_up(., 7.0) %>%
  shower(., temp == 38) %>%
  breakfast(., c("cereals", "tea")) %>%
  bvg(
   .,
    price = "EUR2.90",
    delay = 10
  ) %>%
  physalia(
   .,
    course = "Data Visualization in R with ggplot2"
  )
```

Readability of Code: %>% (The Pipe)

With the pipe one can rewrite the code like this:

```
names_cent <-  
babynames %>%  
filter(year %in% c(1890, 1900)) %>%  
group_by(sex, year) %>%  
mutate(sum = sum(n)) %>%  
group_by(sex, year, name)  
  
names_sum <-  
names_cent %>%  
summarize(  
  prop = prop,  
  prop_check = sum(n) / unique(sum)  
) %>%  
arrange(desc(sex), -prop_check) %>%  
ungroup()
```

Readability of Code: %>% (The Pipe)

... or like this by omitting the intermediate assignment:

```
names_sum <-  
babynames %>%  
filter(year %in% c(1890, 1900)) %>%  
group_by(sex, year) %>%  
mutate(sum = sum(n)) %>%  
group_by(sex, year, name) %>%  
summarize(  
  prop = prop,  
  prop_check = sum(n) / unique(sum)  
) %>%  
arrange(desc(sex), -prop_check) %>%  
ungroup()
```

Readability of Code: %>% (The Pipe)

(Just to show it worked.)

```
names_sum
## # A tibble: 6,425 x 5
##   sex     year name      prop prop_check
##   <chr>  <dbl> <chr>    <dbl>      <dbl>
## 1 M       1890 John     0.0710     0.0766
## 2 M       1890 William  0.0626     0.0675
## 3 M       1900 John     0.0606     0.0653
## 4 M       1900 William  0.0529     0.0570
## 5 M       1900 James    0.0447     0.0481
## 6 M       1890 James    0.0426     0.0459
## 7 M       1890 George   0.0372     0.0402
## 8 M       1890 Charles  0.0339     0.0366
## 9 M       1900 George   0.0333     0.0359
## 10 M      1890 Frank    0.0257     0.0277
## # ... with 6,415 more rows
```

Readability of Code: %>% (The Pipe)

This way, you can also easily inactivate verbs or single conditions:

```
names_sum <-  
babynames %>%  
filter(year %in% c(1890, 1900)) %>%  
group_by(sex, year) %>%  
mutate(sum = sum(n)) %>%  
group_by(sex, year, name) %>%  
summarize(  
  #prop = prop,  
  prop_check = sum(n) / unique(sum)  
) %>%  
arrange(desc(sex), -prop_check) %>%  
ungroup()
```

Readability of Code: %>% (The Pipe)

You can easily inactivate verbs or single conditions:

```
names_sum <-  
babynames %>%  
filter(year %in% c(1890, 1900)) %>%  
group_by(sex, year) %>%  
mutate(sum = sum(n)) %>%  
group_by(sex, year, name) %>%  
summarize(  
  #prop = prop,  
  prop_check = sum(n) / unique(sum)  
) %>%  
#arrange(desc(sex), -prop_check) %>%  
ungroup()
```

Readability of Code: %>% (The Pipe)

You can easily inactivate verbs or single conditions:

```
names_sum <-  
babynames %>%  
filter(year %in% c(1890, 1900)) %>%  
group_by(sex, year) %>%  
mutate(sum = sum(n)) %>%  
group_by(sex, year, name) %>%  
summarize(  
  #prop = prop,  
  prop_check = sum(n) / unique(sum)  
) %>% ## expecting another verb!  
#arrange(desc(sex), -prop_check) %>%  
#ungroup()
```

Readability of Code: %>% (The Pipe)

You can easily inactivate verbs or single conditions:

```
names_sum <-  
babynames %>%  
filter(year %in% c(1890, 1900)) %>%  
group_by(sex, year) %>%  
mutate(sum = sum(n)) %>%  
group_by(sex, year, name) %>%  
summarize(  
  #prop = prop,  
  prop_check = sum(n) / unique(sum)  
) #>% #<< ## comment as well  
#arrange(desc(sex), -prop_check) %>%  
#ungroup()
```

Readability of Code: %>% (The Pipe)

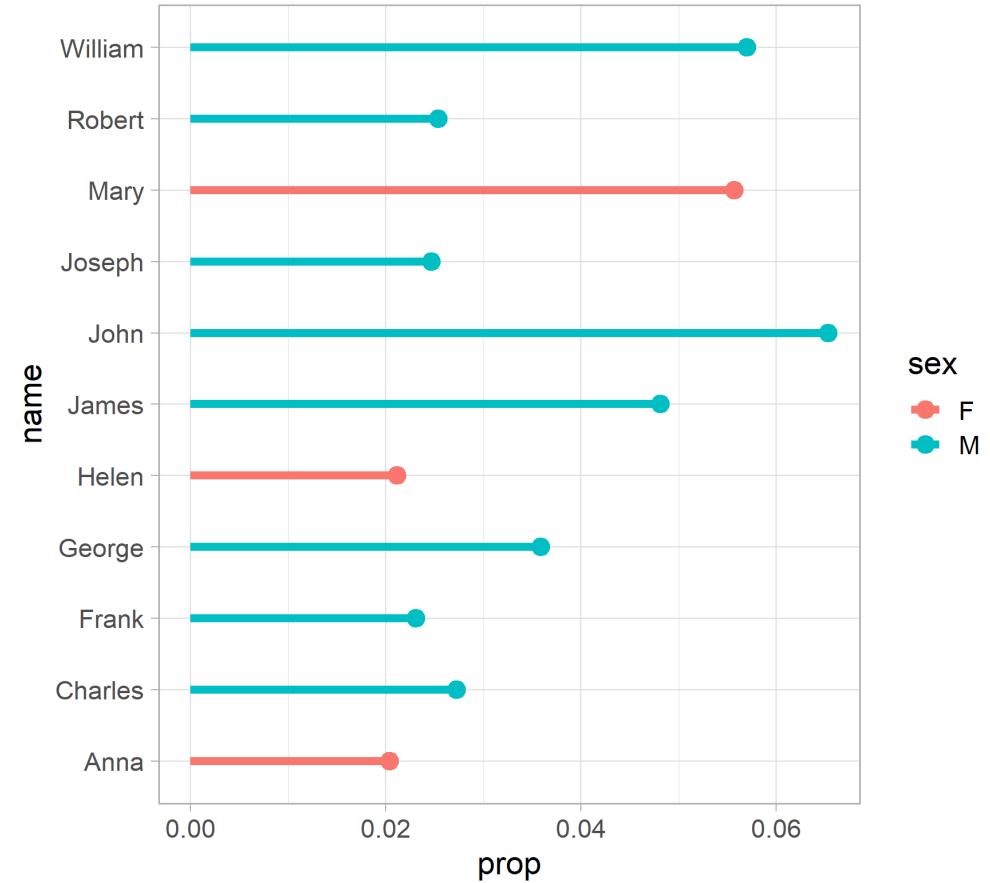
You can easily inactivate verbs or single conditions:

```
names_sum <-  
babynames %>%  
filter(year %in% c(1890, 1900)) %>%  
group_by(sex, year) %>%  
mutate(sum = sum(n)) %>%  
group_by(sex, year, name) %>%  
summarize(  
  #prop = prop,  
  prop_check = sum(n) / unique(sum)  
) %>%  
#arrange(desc(sex), -prop_check) %>%  
#ungroup() %>%  
{} ##<< ... or add this
```

Readability of Code: %>% (The Pipe)

You can also pipe into a `ggplot()` call (but don't forget to use `+` afterwards, not `%>%`):

```
babynames %>%
  filter(year == 1900) %>%
  group_by(sex) %>%
  mutate(sum = sum(n)) %>%
  group_by(sex, name) %>%
  summarize(
    prop = sum(n) / unique(sum)
  ) %>%
  filter(prop >= .02) %>%
  ggplot(aes(prop, name,
             color = sex)) +
  geom_segment(
    aes(xend = 0, yend = name),
    size = 2
  ) +
  geom_point(
    size = 4
  )
```



Your Turn!

- Inspect the `flights` data set from the `nycflights13` package.
- Count the number of flights in June and July 2013.
- Find out how many flights did catch up their departure delay.
- Create a table that contains the average delays per origin and month, ordered by maximum arrival delay. (Bonus: Delays with 1 digit only.)
- Explore the relationship between the average distance and average delay per season and destination.

Flights in June and July 2013

```
library(nycflights13)
filter(flights, month == 6 | month == 7)
## # A tibble: 57,668 x 19
##   year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time
##   <int> <int> <int>     <int>          <int>     <dbl>    <int>          <int>
## 1 2013     6     1       2            2359        3      341          350
## 2 2013     6     1      451           500       -9      624          640
## 3 2013     6     1      506           515       -9      715          800
## 4 2013     6     1      534           545      -11      800          829
## 5 2013     6     1      538           545       -7      925          922
## 6 2013     6     1      539           540       -1      832          840
## 7 2013     6     1      546           600      -14      850          910
## 8 2013     6     1      551           600       -9      828          850
## 9 2013     6     1      552           600       -8      647          655
## 10 2013    6     1      553           600       -7      700          711
## # ... with 57,658 more rows, and 11 more variables: arr_delay <dbl>,
## #   carrier <chr>, flight <int>, tailnum <chr>, origin <chr>, dest <chr>,
## #   air_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>, time_hour <dttm>
```

Flights in June and July 2013

```
filter(flights, month %in% c(6, 7))
## # A tibble: 57,668 x 19
##   year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time
##   <int> <int> <int>     <dbl>          <dbl>     <dbl>     <int>          <dbl>
## 1 2013    6     1       2        2359         3      341        350
## 2 2013    6     1      451        500        -9      624        640
## 3 2013    6     1      506        515        -9      715        800
## 4 2013    6     1      534        545       -11      800        829
## 5 2013    6     1      538        545        -7      925        922
## 6 2013    6     1      539        540        -1      832        840
## 7 2013    6     1      546        600       -14      850        910
## 8 2013    6     1      551        600        -9      828        850
## 9 2013    6     1      552        600        -8      647        655
## 10 2013   6     1      553        600        -7      700        711
## # ... with 57,658 more rows, and 11 more variables: arr_delay <dbl>,
## #   carrier <chr>, flight <int>, tailnum <chr>, origin <chr>, dest <chr>,
## #   air_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>, time_hour <dttm>
```

Catch-Up Delay

```
filter(flights, dep_delay > 0, arr_delay <= 0)
## # A tibble: 35,442 x 19
##   year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time
##   <int> <int> <int>     <int>           <int>     <dbl>     <int>           <int>
## 1 2013     1     1      601            600       1        844          850
## 2 2013     1     1      644            636       8        931          940
## 3 2013     1     1      646            645       1        910          916
## 4 2013     1     1      646            645       1       1023         1030
## 5 2013     1     1      701            700       1       1123         1154
## 6 2013     1     1      752            750       2       1025         1029
## 7 2013     1     1      803            800       3       1132         1144
## 8 2013     1     1      826            817       9       1145         1158
## 9 2013     1     1      846            845       1       1138         1205
## 10 2013    1     1      856            855       1       1140         1203
## # ... with 35,432 more rows, and 11 more variables: arr_delay <dbl>,
## #   carrier <chr>, flight <int>, tailnum <chr>, origin <chr>, dest <chr>,
## #   air_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>, time_hour <dttm>
```

Average Delays per Origin and Month

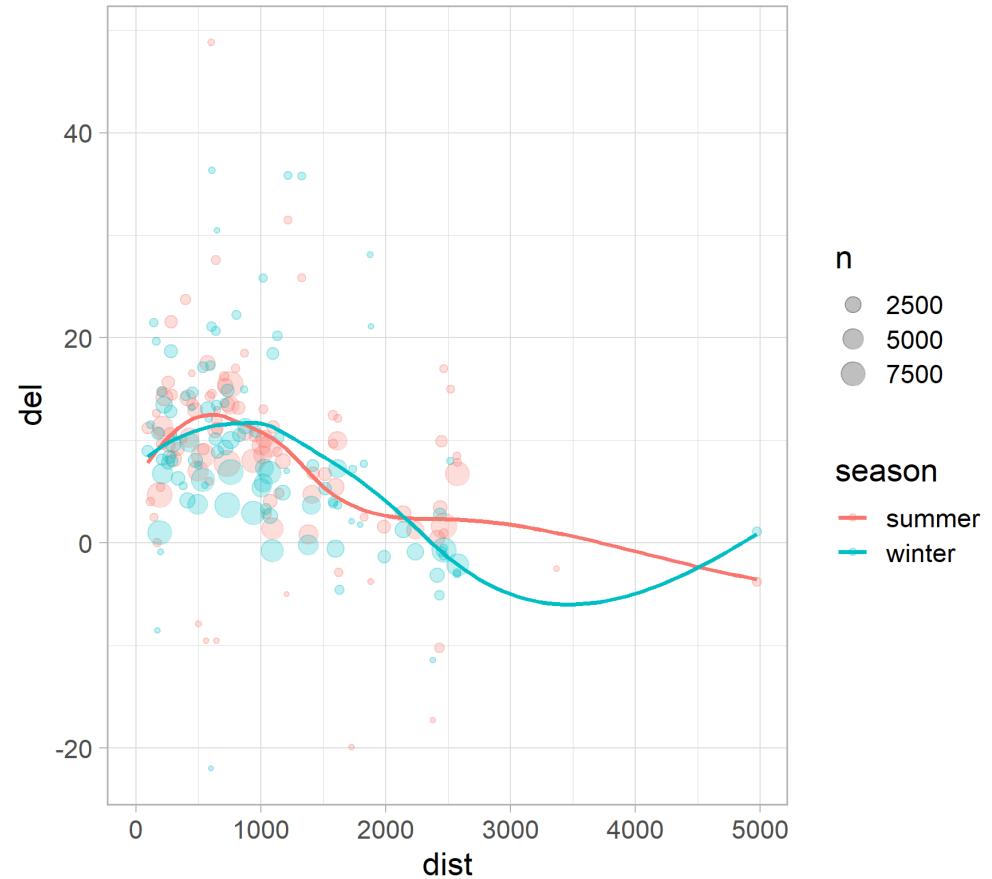
```
flights %>%
  group_by(month, origin) %>%
  summarize(
    dep_delay_avg = round(mean(dep_delay, na.rm = T), 1),
    arr_delay_avg = round(mean(arr_delay, na.rm = T), 1)
  ) %>%
  arrange(-arr_delay_avg)
## # A tibble: 36 x 4
## # Groups:   month [12]
##   month origin dep_delay_avg arr_delay_avg
##   <int> <chr>        <dbl>        <dbl>
## 1     7  JFK         23.8         20.2
## 2     12 EWR          21          19.6
## 3     6  JFK         20.5         17.6
## 4     6  EWR         22.5         16.9
## 5     7  EWR          22          15.5
## 6     6  LGA         19.3         14.8
## 7     7  LGA          19          14.2
## 8     4  EWR         17.4         14.1
## 9     1  EWR         14.9         12.8
## 10    12 JFK         14.8         12.7
## # ... with 26 more rows
```

Average Delays per Origin and Month

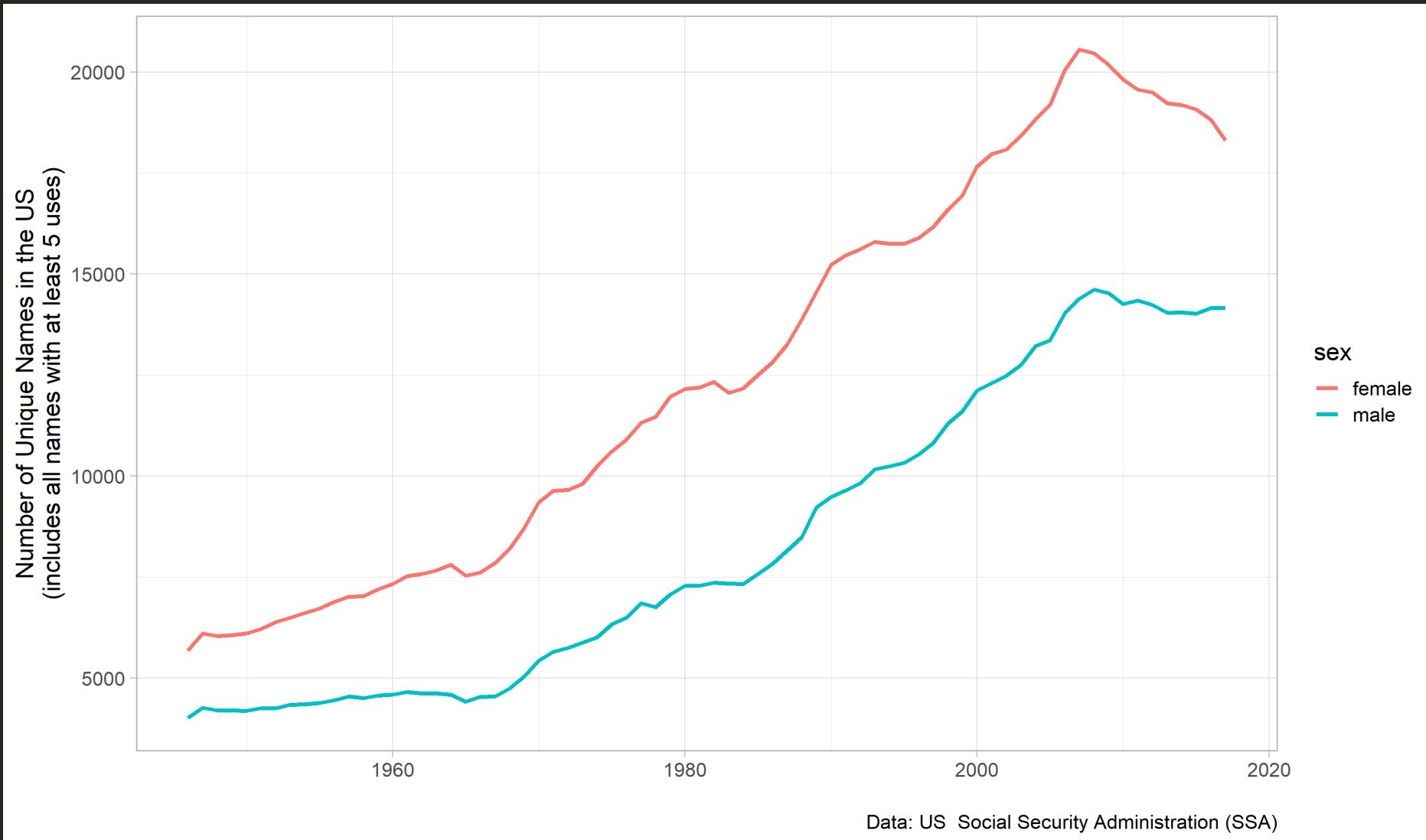
```
flights %>%
  select(month, origin, ends_with("delay")) %>%
  group_by(month, origin) %>%
  summarize_all(list(~round(mean(., na.rm = T), 1))) %>%
  arrange(-arr_delay)
## # A tibble: 36 x 4
## # Groups:   month [12]
##   month origin dep_delay arr_delay
##   <int> <chr>     <dbl>     <dbl>
## 1     1  JFK      23.8      20.2
## 2     2  EWR      21        19.6
## 3     3  JFK      20.5      17.6
## 4     4  EWR      22.5      16.9
## 5     5  EWR      22        15.5
## 6     6  LGA      19.3      14.8
## 7     7  LGA      19        14.2
## 8     8  EWR      17.4      14.1
## 9     9  EWR      14.9      12.8
## 10    10  JFK      14.8      12.7
## # ... with 26 more rows
```

Distance versus Delay

```
flights %>%
  mutate(
    season = if_else(month %in% 4:9,
                      "summer",
                      "winter")
  ) %>%
  group_by(season, dest) %>%
  summarize(
    del = mean(arr_delay, na.rm = T),
    dist = mean(distance, na.rm = T),
    n = n()
  ) %>%
  ggplot(
    aes(
      dist,
      del,
      color = season
    )) +
  geom_point(
    aes(size = n),
    alpha = .25
  ) +
```



Your Turn: Create this Line Plot!

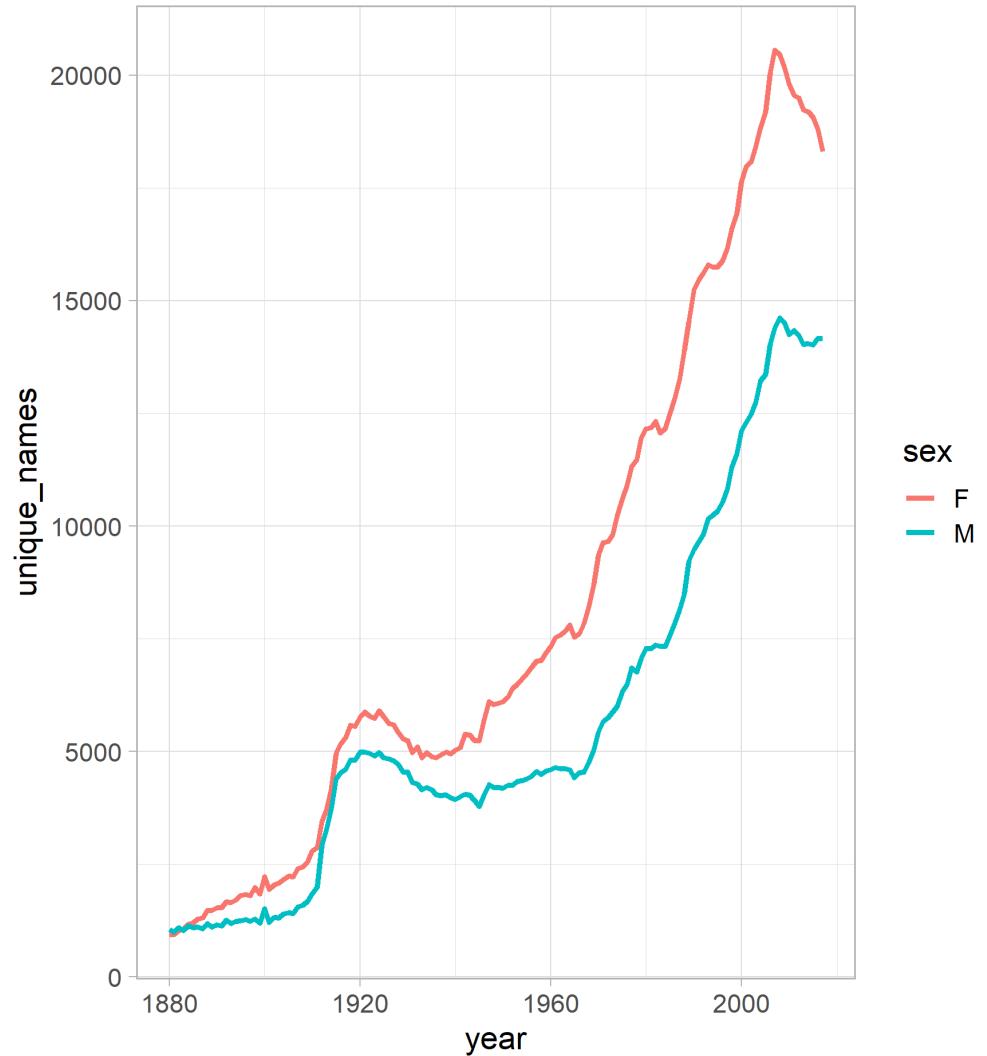


Your Turn: Post-WWII Line Plot

```
babynames %>%
  group_by(year, sex) %>%
  summarize(unique_names = n_distinct(name))
## # A tibble: 276 x 3
## # Groups:   year [138]
##       year  sex unique_names
##       <dbl> <chr>     <int>
## 1 1880 F         942
## 2 1880 M        1058
## 3 1881 F         938
## 4 1881 M         997
## 5 1882 F        1028
## 6 1882 M        1099
## 7 1883 F        1054
## 8 1883 M        1030
## 9 1884 F        1172
## 10 1884 M       1125
## # ... with 266 more rows
```

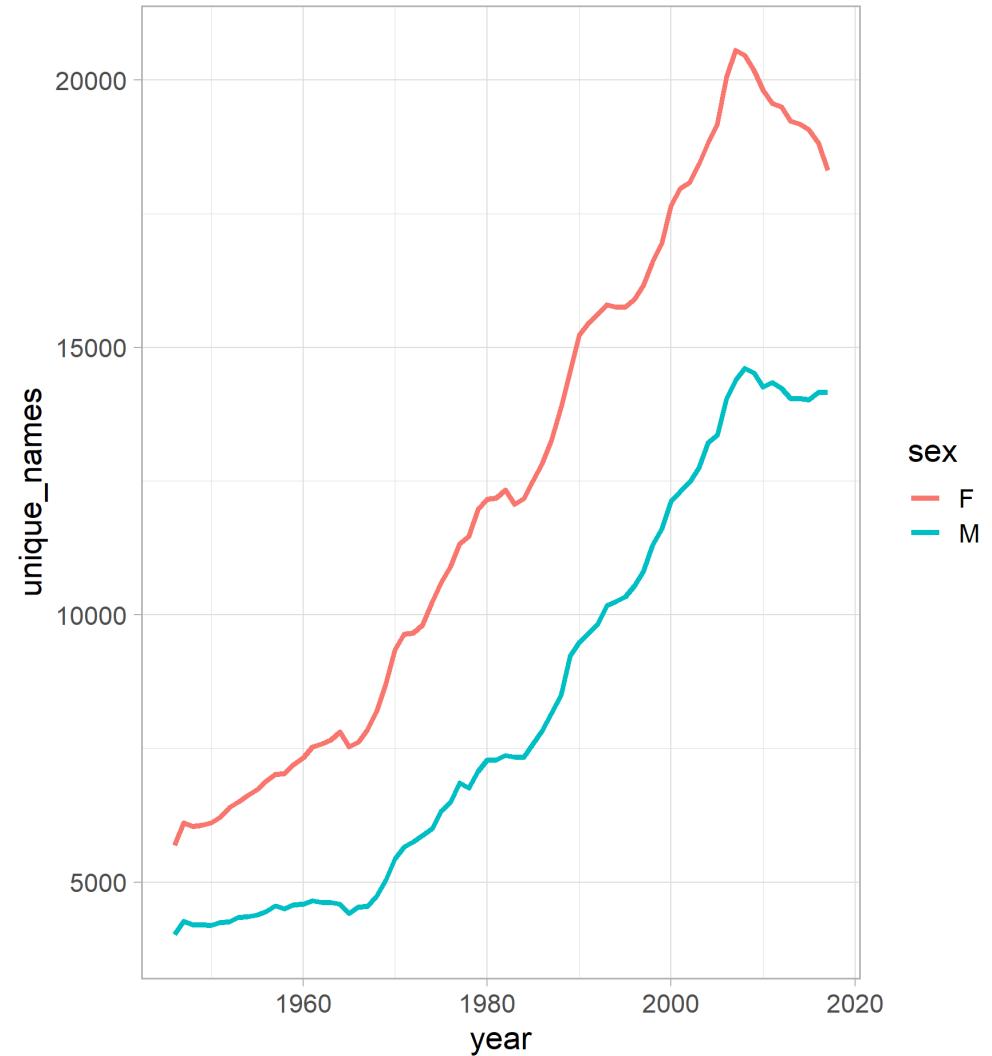
Your Turn: Post-WWII Line Plot

```
babynames %>%
  group_by(year, sex) %>%
  summarize(
    unique_names = n_distinct(name)
  ) %>%
  ggplot(
    aes(
      year,
      unique_names,
      color = sex
    )) +
  geom_line(size = 1.3)
```



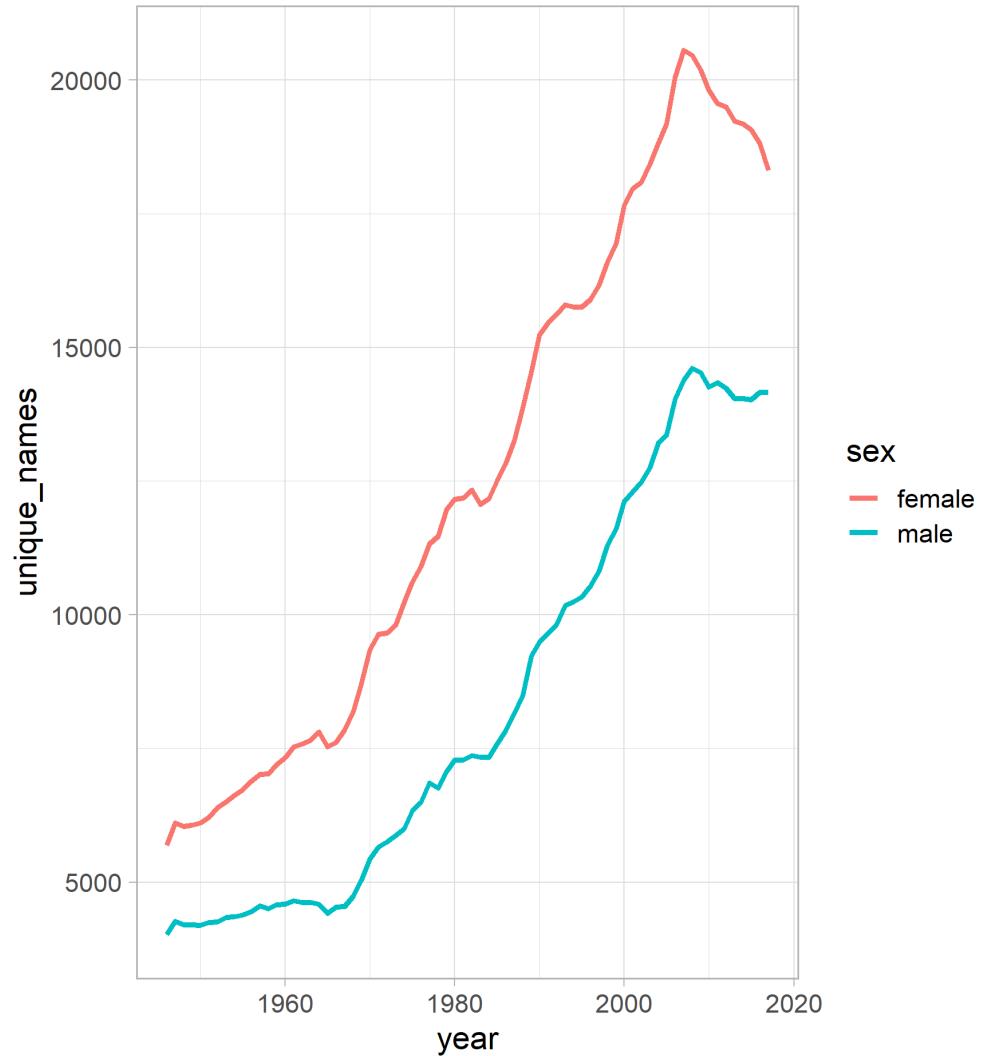
Your Turn: Post-WWII Line Plot

```
babynames %>%
  filter(year > 1945) %>%
  group_by(year, sex) %>%
  summarize(
    unique_names = n_distinct(name)
  ) %>%
  ggplot(
    aes(
      year,
      unique_names,
      color = sex
    )) +
  geom_line(size = 1.3)
```



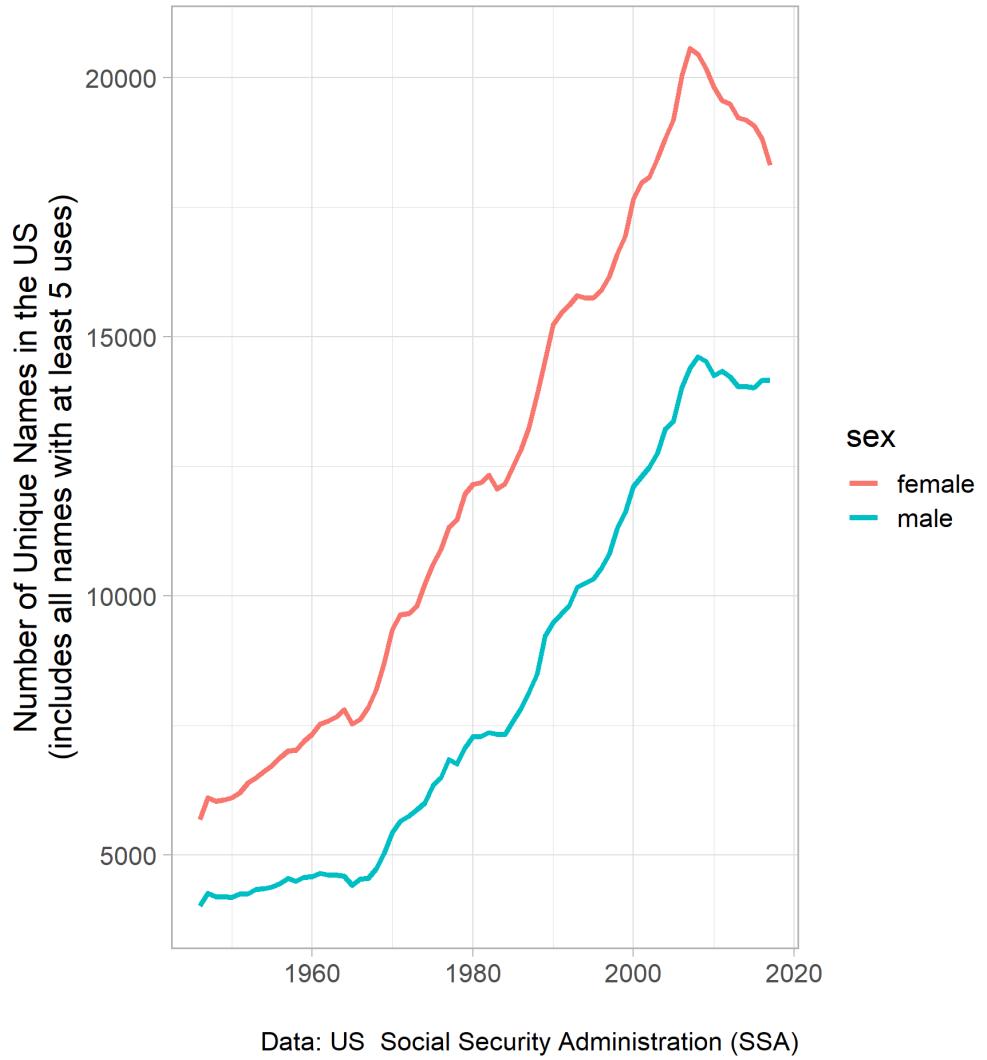
Your Turn: Post-WWII Line Plot

```
babynames %>%
  filter(year > 1945) %>%
  mutate(
    sex = if_else(sex == "M",
                  "male",
                  "female")
  ) %>%
  group_by(year, sex) %>%
  summarize(
    unique_names = n_distinct(name)
  ) %>%
  ggplot(
    aes(
      year,
      unique_names,
      color = sex
    )) +
  geom_line(size = 1.3)
```



Your Turn: Post-WWII Line Plot

```
babynames %>%
  filter(year > 1945) %>%
  mutate(
    sex = if_else(sex == "M",
                  "male",
                  "female")
  ) %>%
  group_by(year, sex) %>%
  summarize(
    unique_names = n_distinct(name)
  ) %>%
  ggplot(
    aes(
      year,
      unique_names,
      color = sex
    )) +
  geom_line(size = 1.3) +
  labs(
    x = NULL,
    y = "Number of Unique Names in the US",
    caption = "\nData: US Social Security Administration (SSA)")
```



Other Helpful `dplyr` and `tidyverse` Functions

Function	Explanation
<code>slice()</code>	Extract rows by ordinal position
<code>distinct()</code> and <code>n_distinct()</code>	Find and count unique values
<code>sample_n()</code> and <code>sample_frac()</code>	Select rows randomly
<code>top_n()</code> and <code>top_frac()</code>	Pick top or bottom values by variable
<code>count()</code> and <code>top_frac()</code>	Count number of observations
<code>complete()</code>	Create all possible combinations of two variables
<code>*_join()</code>	Merge two tables
<code>pivot_longer()</code> and <code>pivot_wider()</code>	Turn wide data into long and vice versa
<code>fct_*</code> ()	Change factor levels dynamically
<code>str_*</code> ()	Manipulate character strings
<code>glue()</code>	Combine strings

Extract Rows by Ordinal Position

`slice()` allows you to filter rows based on their row value:

```
slice(mpg, 100:120)
## # A tibble: 21 x 11
##   manufacturer model  displ  year   cyl trans  drv   cty   hwy fl class
##   <chr>        <chr>  <dbl> <int> <int> <chr>  <chr> <int> <int> <chr> <chr>
## 1 honda         civic    1.6  1999     4 manual~ f      28    33 r  subcom~
## 2 honda         civic    1.6  1999     4 auto(l~ f      24    32 r  subcom~
## 3 honda         civic    1.6  1999     4 manual~ f      25    32 r  subcom~
## 4 honda         civic    1.6  1999     4 manual~ f      23    29 p  subcom~
## 5 honda         civic    1.6  1999     4 auto(l~ f      24    32 r  subcom~
## 6 honda         civic    1.8  2008     4 manual~ f      26    34 r  subcom~
## 7 honda         civic    1.8  2008     4 auto(l~ f      25    36 r  subcom~
## 8 honda         civic    1.8  2008     4 auto(l~ f      24    36 c  subcom~
## 9 honda         civic     2   2008     4 manual~ f      21    29 p  subcom~
## 10 hyundai      sonata   2.4  1999     4 auto(l~ f      18    26 r  midsize
## # ... with 11 more rows
```

Extract Rows by Ordinal Position

`slice()` allows you to filter rows based on their row value - works also per group:

```
mpg %>% group_by(manufacturer) %>% slice(1)
## # A tibble: 15 x 11
## # Groups:   manufacturer [15]
##   manufacturer model   displ  year   cyl trans drv   cty   hwy fl   class
##   <chr>        <chr>   <dbl> <int> <int> <chr> <chr> <int> <int> <chr> <chr>
## 1 audi         a4      1.8   1999     4 auto(~ f       18    29 p   compa~
## 2 chevrolet    c1500 s~  5.3   2008     8 auto(~ r       14    20 r   suv
## 3 dodge        caravan~ 2.4   1999     4 auto(~ f       18    24 r   miniv~
## 4 ford         expedit~ 4.6   1999     8 auto(~ r       11    17 r   suv
## 5 honda        civic   1.6   1999     4 manua~ f       28    33 r   subco~
## 6 hyundai      sonata  2.4   1999     4 auto(~ f       18    26 r   midsi~
## 7 jeep         grand c~  3     2008     6 auto(~ 4      17    22 d   suv
## 8 land rover   range r~  4     1999     8 auto(~ 4      11    15 p   suv
## 9 lincoln      navigat~ 5.4   1999     8 auto(~ r       11    17 r   suv
## 10 mercury     mountai~  4     1999     6 auto(~ 4      14    17 r   suv
## 11 nissan      altima   2.4   1999     4 manua~ f       21    29 r   compa~
## 12 pontiac     grand p~  3.1   1999     6 auto(~ f       18    26 r   midsi~
## 13 subaru      foreste~  2.5   1999     4 manua~ 4      18    25 r   suv
## 14 toyota      4runner~  2.7   1999     4 manua~ 4      15    20 r   suv
## 15 volkswagen  gti     2     1999     4 manua~ f       21    29 r   compa~
```

Find Unique Values

`distinct()` allows you to retain only unique values (as `unique()` does):

```
distinct(mpg, manufacturer)
## # A tibble: 15 x 1
##   manufacturer
##   <chr>
## 1 audi
## 2 chevrolet
## 3 dodge
## 4 ford
## 5 honda
## 6 hyundai
## 7 jeep
## 8 land rover
## 9 lincoln
## 10 mercury
## 11 nissan
## 12 pontiac
## 13 subaru
## 14 toyota
## 15 volkswagen
```

Find Unique Values

`distinct()` allows you to retain only unique values (as `unique()` does):

```
distinct(mpg, hwy)
## # A tibble: 27 x 1
##       hwy
##   <int>
## 1     29
## 2     31
## 3     30
## 4     26
## 5     27
## 6     25
## 7     28
## 8     24
## 9     23
## 10    20
## # ... with 17 more rows
```

Count Unique Values

`n_distinct()` allows you to calculate the number of unique values (as `length(unique())` does):

```
n_distinct(mpg$model)
## [1] 38
```

Select Rows Randomly

`sample_n()` allows you to create random subsets based on a number of rows:

```
sample_n(mpg, 5)
## # A tibble: 5 x 11
##   manufacturer model      displ  year   cyl trans  drv   cty   hwy fl class
##   <chr>        <chr>     <dbl> <int> <int> <chr> <chr> <int> <int> <chr> <chr>
## 1 subaru       forester~    2.5  2008     4 auto(l~ 4          20    26 r   suv 
## 2 chevrolet    malibu      2.4  2008     4 auto(l~ f          22    30 r   mids~
## 3 ford         explorer~    4     1999     6 manual~ 4          15    19 r   suv 
## 4 volkswagen   jetta       2.5  2008     5 manual~ f          21    29 r   comp~
## 5 mercury      mountain~    4     1999     6 auto(l~ 4          14    17 r   suv 
```

Select Rows Randomly

`sample_n()` allows you to create random subsets based on a number of rows - also works with groups:

```
mpg %>% group_by(year, manufacturer) %>% sample_n(1)
## # A tibble: 30 x 11
## # Groups:   year, manufacturer [30]
##   manufacturer model      displ  year   cyl trans  drv   cty   hwy fl class
##   <chr>        <chr>     <dbl> <int> <int> <chr> <chr> <int> <int> <chr> <chr>
## 1 audi         a4 quat~    1.8  1999     4 manua~ 4          18    26 p   compa~
## 2 chevrolet    malibu     3.1  1999     6 auto(~ f 18    26 r   midsi~
## 3 dodge        caravan~   3.8  1999     6 auto(~ f 15    21 r   miniv~
## 4 ford         mustang    3.8  1999     6 auto(~ r 18    25 r   subco~
## 5 honda        civic      1.6  1999     4 manua~ f 25    32 r   subco~
## 6 hyundai      sonata    2.4  1999     4 auto(~ f 18    26 r   midsi~
## 7 jeep         grand c~  4     1999     6 auto(~ 4 15    20 r   suv
## 8 land rover   range r~  4.6  1999     8 auto(~ 4 11    15 p   suv
## 9 lincoln      navigat~  5.4  1999     8 auto(~ r 11    16 p   suv
## 10 mercury     mountai~  5     1999     8 auto(~ 4 13    17 r   suv
## # ... with 20 more rows
```

Select Rows Randomly

`sample_frac()` allows you to create random subsets based on a fraction:

```
sample_frac(mpg, .05)
## # A tibble: 12 x 11
##   manufacturer model      displ  year   cyl trans  drv   cty   hwy fl class
##   <chr>        <chr>     <dbl> <int> <int> <chr> <chr> <int> <int> <chr> <chr>
## 1 volkswagen  jetta     1.9   1999     4 manua~ f       33    44 d   comp~
## 2 dodge       caravan ~ 3.8   1999     6 auto(~ f       15    22 r   mini~
## 3 toyota      camry     2.2   1999     4 auto(~ f       21    27 r   mids~
## 4 audi        a4        3.1   2008     6 auto(~ f       18    27 p   comp~
## 5 ford        f150 pic~ 4.6   1999     8 auto(~ 4       13    16 r   pick~
## 6 nissan      maxima    3.5   2008     6 auto(~ f       19    25 p   mids~
## 7 dodge       dakota p~ 4.7   2008     8 auto(~ 4       9     12 e   pick~
## 8 chevrolet   malibu    2.4   2008     4 auto(~ f       22    30 r   mids~
## 9 dodge       ram 1500~ 4.7   2008     8 manua~ 4       12    16 r   pick~
## 10 dodge      caravan ~ 3.3   2008     6 auto(~ f      17    24 r   mini~
## 11 volkswagen gti       2     2008     4 auto(~ f      22    29 p   comp~
## 12 volkswagen gti       2.8   1999     6 manua~ f      17    24 r   comp~
```

Pick Top or Bottom Values

`top_n()` allows you to filter the **top** or bottom values, ranked by a variable:

```
top_n(mpg, 5, hwy)
## # A tibble: 6 x 11
##   manufacturer model   displ  year   cyl trans   drv   cty   hwy fl class
##   <chr>        <chr>   <dbl> <int> <int> <chr>   <chr> <int> <int> <chr> <chr>
## 1 honda        civic    1.8   2008     4 auto(l~ f       25    36 r   subcom~
## 2 honda        civic    1.8   2008     4 auto(l~ f       24    36 c   subcom~
## 3 toyota       corolla  1.8   2008     4 manual~ f      28    37 r   compact
## 4 volkswagen  jetta    1.9   1999     4 manual~ f      33    44 d   compact
## 5 volkswagen  new be~  1.9   1999     4 manual~ f      35    44 d   subcom~
## 6 volkswagen  new be~  1.9   1999     4 auto(l~ f       29    41 d   subcom~
```

Pick Top or Bottom Values

`top_n()` allows you to filter the **top** or **bottom** values, ranked by a variable:

```
top_n(mpg, 5, -hwy)
## # A tibble: 5 x 11
##   manufacturer model      displ  year cyl trans drv cty hwy fl class
##   <chr>        <chr>     <dbl> <int> <int> <chr> <chr> <int> <int> <chr> <chr>
## 1 dodge        dakota pi~    4.7  2008     8 auto(~ 4          9    12 e   pick~
## 2 dodge        durango 4~   4.7  2008     8 auto(~ 4          9    12 e   suv 
## 3 dodge        ram 1500 ~  4.7  2008     8 auto(~ 4          9    12 e   pick~
## 4 dodge        ram 1500 ~  4.7  2008     8 manua~ 4          9    12 e   pick~
## 5 jeep         grand che~  4.7  2008     8 auto(~ 4          9    12 e   suv 
```

Pick Top or Bottom Values

`top_n()` allows you to filter the top or bottom values, ranked by a variable:

```
mpg %>% group_by(manufacturer) %>% top_n(1, hwy)
## # A tibble: 25 x 11
## # Groups:   manufacturer [15]
##   manufacturer model  displ  year   cyl trans   drv   cty   hwy fl class
##   <chr>        <chr>  <dbl> <int> <int> <chr>   <chr> <int> <int> <chr> <chr>
## 1 audi         a4      2     2008     4 manual~ f       20    31 p   compa~
## 2 chevrolet    malibu  2.4   2008     4 auto(l~ f      22    30 r   midsi~
## 3 dodge        carava~ 2.4   1999     4 auto(l~ f      18    24 r   miniv~
## 4 dodge        carava~ 3     1999     6 auto(l~ f      17    24 r   miniv~
## 5 dodge        carava~ 3.3   2008     6 auto(l~ f      17    24 r   miniv~
## 6 dodge        carava~ 3.3   2008     6 auto(l~ f      17    24 r   miniv~
## 7 ford         mustang 3.8   1999     6 manual~ r     18    26 r   subco~
## 8 ford         mustang 4     2008     6 manual~ r     17    26 r   subco~
## 9 honda        civic   1.8   2008     4 auto(l~ f     25    36 r   subco~
## 10 honda       civic   1.8   2008     4 auto(l~ f    24    36 c   subco~
## # ... with 15 more rows
```

Pick Top or Bottom Values

`top_frac()` allows you to filter the top or bottom values, ranked by a variable:

```
top_frac(mpg, .01, displ)
## # A tibble: 2 x 11
##   manufacturer model      displ  year   cyl trans  drv   cty   hwy fl class
##   <chr>        <chr>     <dbl> <int> <int> <chr> <chr> <int> <int> <chr> <chr>
## 1 chevrolet    corvette     7    2008     8 manual~ r         15     24 p   2sea~
## 2 chevrolet    k1500 ta~   6.5   1999     8 auto(l~ 4         14     17 d   suv
```

Count Observations

`count()` calculates the sample size of your data as summary statistic:

```
count(mpg)
## # A tibble: 1 x 1
##       n
##   <int>
## 1 234
```

Count Observations

`count()` calculates the sample size of your data as summary statistic:

```
count(mpg)
## # A tibble: 1 x 1
##       n
##   <int>
## 1    234
```

```
summarize(mpg, n = n())
## # A tibble: 1 x 1
##       n
##   <int>
## 1    234
```

Count Observations

`count()` calculates the sample size of your data as summary statistic - also per group:

```
mpg %>% group_by(manufacturer) %>% count()
## # A tibble: 15 x 2
## # Groups:   manufacturer [15]
##       manufacturer     n
##       <chr>           <int>
## 1 audi                 18
## 2 chevrolet            19
## 3 dodge                37
## 4 ford                 25
## 5 honda                9
## 6 hyundai              14
## 7 jeep                 8
## 8 land rover            4
## 9 lincoln               3
## 10 mercury              4
## 11 nissan               13
## 12 pontiac              5
## 13 subaru               14
## 14 toyota                34
## 15 volkswagen            27
```

Count Observations

`count()` calculates the sample size of your data as summary statistic - also per group:

```
mpg %>% group_by(manufacturer) %>% count(sort = TRUE)
## # A tibble: 15 x 2
## # Groups:   manufacturer [15]
##       manufacturer     n
##       <chr>           <int>
## 1 dodge              37
## 2 toyota             34
## 3 volkswagen         27
## 4 ford               25
## 5 chevrolet          19
## 6 audi               18
## 7 hyundai            14
## 8 subaru             14
## 9 nissan              13
## 10 honda              9
## 11 jeep               8
## 12 pontiac            5
## 13 land rover          4
## 14 mercury             4
## 15 lincoln            3
```

Count Observations

`tally()` does basically the same as `count()`:

```
tally(mpg)
## # A tibble: 1 x 1
##       n
##   <int>
## 1    234
```

Count Observations

`add_count()` adds the sample size of your data as additional column:

```
add_count(mpg2)
## # A tibble: 234 x 4
##   model      hwy     cty     n
##   <chr>    <int> <int> <int>
## 1 a4          29     18    234
## 2 a4          29     21    234
## 3 a4          31     20    234
## 4 a4          30     21    234
## 5 a4          26     16    234
## 6 a4          26     18    234
## 7 a4          27     18    234
## 8 a4 quattro  26     18    234
## 9 a4 quattro  25     16    234
## 10 a4 quattro 28     20    234
## # ... with 224 more rows
```

Count Observations

`add_count()` adds the sample size of your data as additional column:

```
add_count(mpg2)
## # A tibble: 234 x 4
##   model      hwy     cty     n
##   <chr>    <int>  <int>  <int>
## 1 a4          29     18    234
## 2 a4          29     21    234
## 3 a4          31     20    234
## 4 a4          30     21    234
## 5 a4          26     16    234
## 6 a4          26     18    234
## 7 a4          27     18    234
## 8 a4 quattro  26     18    234
## 9 a4 quattro  25     16    234
## 10 a4 quattro 28     20    234
## # ... with 224 more rows
```

```
summarize(mpg2, n = n())
## # A tibble: 1 x 1
##       n
##   <int>
## 1 234
```

Create All Combinations of Two Columns

`complete()` turns missing values into explicit missing values:

```
mpg3 <- mpg %>% group_by(manufacturer, class) %>% count() %>% ungroup()
complete(mpg3, manufacturer, nesting(class))
## # A tibble: 105 x 3
##   manufacturer class       n
##   <chr>        <chr>     <int>
## 1 audi         2seater    NA
## 2 audi         compact     15
## 3 audi         midsize     3
## 4 audi         minivan    NA
## 5 audi         pickup      NA
## 6 audi         subcompact  NA
## 7 audi         suv         NA
## 8 chevrolet    2seater     5
## 9 chevrolet    compact     NA
## 10 chevrolet   midsize    5
## # ... with 95 more rows
```

Create All Combinations of Two Columns

`complete()` turns missing values into explicit missing values:

```
mpg3 <- mpg %>% group_by(manufacturer, class) %>% count() %>% ungroup()
complete(mpg3, manufacturer, nesting(class), fill = list(n = 0))
## # A tibble: 105 x 3
##       manufacturer   class     n
##       <chr>        <chr> <dbl>
## 1 audi         2seater     0
## 2 audi         compact     15
## 3 audi         midsize     3
## 4 audi         minivan     0
## 5 audi         pickup      0
## 6 audi         subcompact   0
## 7 audi         suv         0
## 8 chevrolet    2seater     5
## 9 chevrolet    compact      0
## 10 chevrolet   midsize     5
## # ... with 95 more rows
```

The `*_join()` Family

Several `*_join()` functions let you merge two data tables:

```
band_members
## # A tibble: 3 x 2
##   name   band
##   <chr> <chr>
## 1 Mick  Stones
## 2 John  Beatles
## 3 Paul  Beatles
```

```
band_instruments
## # A tibble: 3 x 2
##   name   plays
##   <chr> <chr>
## 1 John  guitar
## 2 Paul  bass
## 3 Keith guitar
```

The `*_join()` Family

Several `*_join()` functions let you merge two data tables:

```
left_join(band_members, band_instruments)
## # A tibble: 3 x 3
##   name   band   plays
##   <chr> <chr>   <chr>
## 1 Mick  Stones  <NA>
## 2 John  Beatles guitar
## 3 Paul  Beatles bass
```

The `*_join()` Family

Several `*_join()` functions let you merge two data tables:

```
right_join(band_members, band_instruments)
## # A tibble: 3 x 3
##   name   band   plays
##   <chr> <chr>   <chr>
## 1 John  Beatles guitar
## 2 Paul  Beatles bass
## 3 Keith <NA>    guitar
```

The `join()` Family

Several `*_join()` functions let you merge two data tables:

```
inner_join(band_members, band_instruments)
## # A tibble: 2 x 3
##   name   band   plays
##   <chr> <chr>   <chr>
## 1 John  Beatles guitar
## 2 Paul  Beatles bass
```

The `join()` Family

Several `*_join()` functions let you merge two data tables:

```
full_join(band_members, band_instruments)
## # A tibble: 4 x 3
##   name   band   plays
##   <chr> <chr>   <chr>
## 1 Mick  Stones  <NA>
## 2 John  Beatles guitar
## 3 Paul  Beatles bass
## 4 Keith <NA>    guitar
```

The `join()` Family

Several `*_join()` functions let you merge two data tables:

```
full_join(band_members, band_instruments, by = "name")
## # A tibble: 4 x 3
##   name   band   plays
##   <chr> <chr>   <chr>
## 1 Mick  Stones  <NA>
## 2 John  Beatles guitar
## 3 Paul  Beatles bass
## 4 Keith <NA>    guitar
```

The `*_join()` Family

Several `*_join()` functions let you merge two data tables:

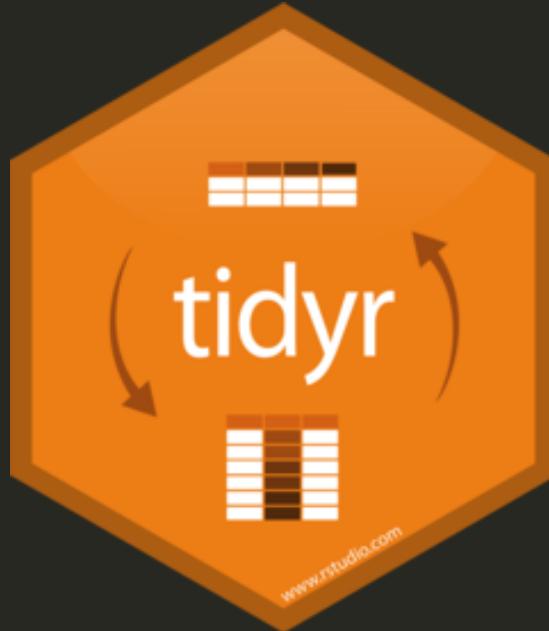
```
band_members
## # A tibble: 3 x 2
##   name   band
##   <chr> <chr>
## 1 Mick  Stones
## 2 John  Beatles
## 3 Paul  Beatles
```

```
band_instruments2
## # A tibble: 3 x 2
##   artist plays
##   <chr> <chr>
## 1 John   guitar
## 2 Paul   bass
## 3 Keith  guitar
```

The `join()` Family

Several `*_join()` functions let you merge two data tables:

```
full_join(band_members, band_instruments2, by = c("name" = "artist"))
## # A tibble: 4 x 3
##   name   band   plays
##   <chr> <chr>   <chr>
## 1 Mick  Stones <NA>
## 2 John  Beatles guitar
## 3 Paul  Beatles bass
## 4 Keith <NA>    guitar
```



The **tidyr** Package

`pivot_wider()` and `pivot_longer()` (tidyverse)

wide

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

long

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

`pivot_wider()` and `pivot_longer()` (tidyverse)

`pivot_wider(names_from = , values_from =)` let's you turn a long data frame into a wide one:

```
(mpg_wide <-
  mpg %>%
  group_by(manufacturer, model, trans, cyl, year) %>%
  summarize(cty = mean(cty, na.rm = T)) %>%
  pivot_wider(
    id_cols = c(manufacturer, model, trans, cyl),
    names_from = year,
    values_from = cty,
    names_prefix = "cty_"
  ))
## # A tibble: 157 x 6
## # Groups:   manufacturer, model, trans, cyl [157]
##   manufacturer model     trans      cyl cty_2008 cty_1999
##   <chr>        <chr>     <chr>     <int>    <dbl>    <dbl>
## 1 audi         a4       auto(av)     4      21      NA
## 2 audi         a4       auto(av)     6      18      NA
## 3 audi         a4       auto(l5)     4      NA      18
## 4 audi         a4       auto(l5)     6      NA      16
## 5 audi         a4       manual(m5)  4      NA      21
## 6 audi         a4       manual(m5)  6      NA      18
```

`pivot_wider()` and `pivot_longer()` (tidyverse)

`pivot_longer(names_to = , values_to =)` let's you turn a wide data frame into a long one:

```
mpg_wide %>%
  pivot_longer(
    cols = starts_with("cty_"),
    names_to = "year",
    values_to = "cty",
    names_prefix = "cty_"
)
## # A tibble: 314 x 6
## # Groups:   manufacturer, model, trans, cyl [157]
##   manufacturer model trans      cyl year   cty
##   <chr>        <chr> <chr>     <int> <chr> <dbl>
## 1 audi         a4    auto(av)    4  2008    21
## 2 audi         a4    auto(av)    4  1999    NA
## 3 audi         a4    auto(av)    6  2008    18
## 4 audi         a4    auto(av)    6  1999    NA
## 5 audi         a4    auto(l5)    4  2008    NA
## 6 audi         a4    auto(l5)    4  1999    18
## 7 audi         a4    auto(l5)    6  2008    NA
## 8 audi         a4    auto(l5)    6  1999    16
## 9 audi         a4    manual(m5)  4  2008    NA
```

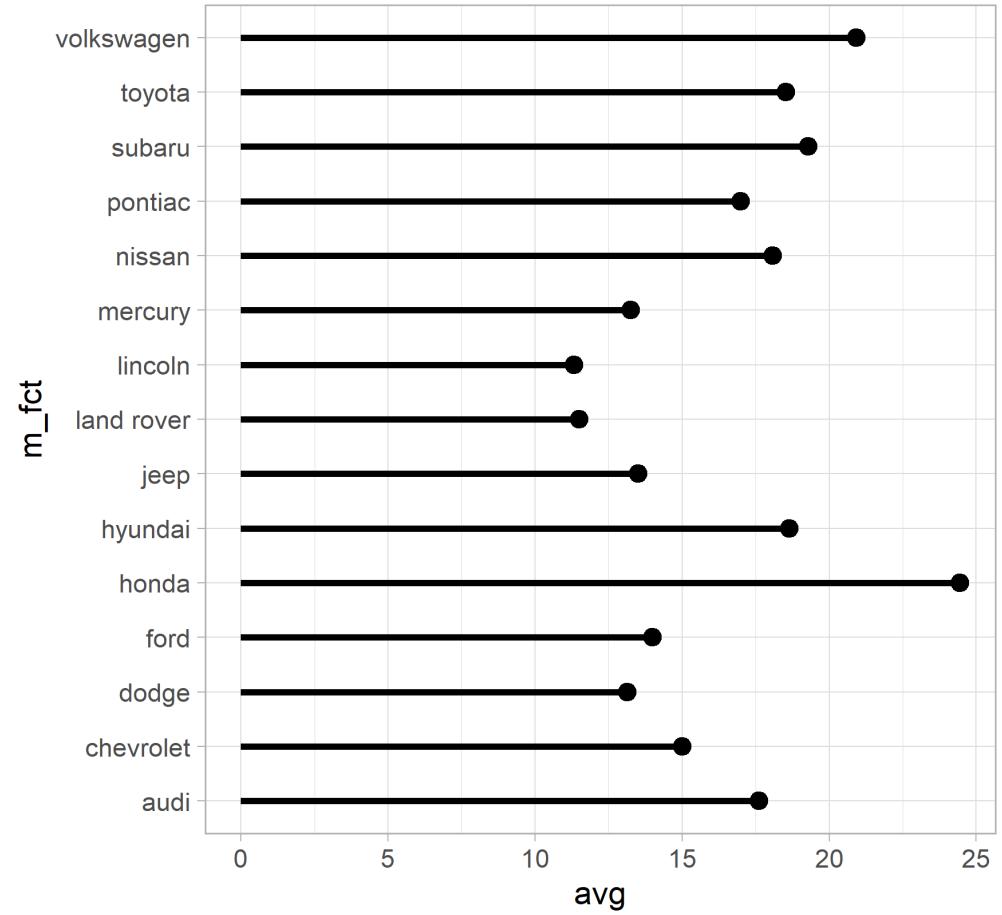


The **forcats** Package

Functions from the `forcats` Package

The `forcats` package provides helpers for reordering factor levels:

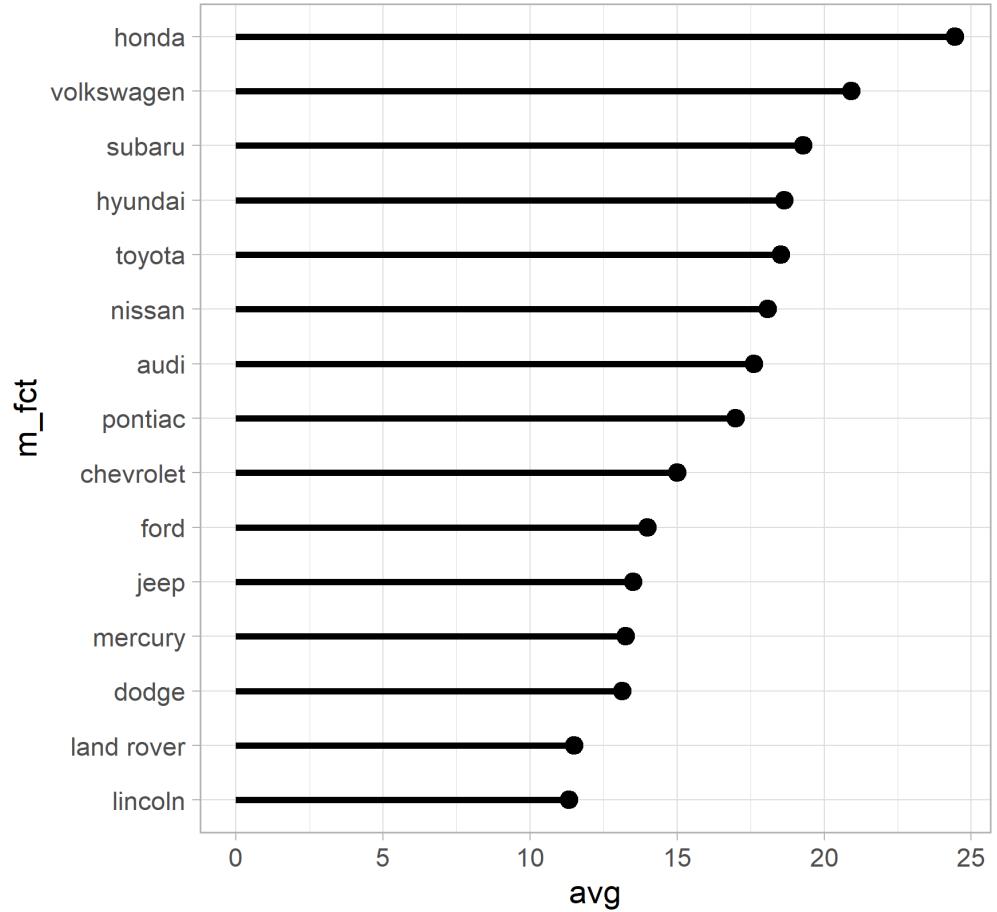
```
mpg_avg <-  
  mpg %>%  
  mutate(  
    m_fct = factor(manufacturer)  
  ) %>%  
  filter(!is.na(cty)) %>%  
  group_by(m_fct) %>%  
  summarize(avg = mean(cty)) %>%  
  ungroup()  
  
mpg_avg %>%  
  ggplot(aes(avg, m_fct)) +  
    geom_segment(  
      aes(xend = 0, yend = m_fct),  
      size = 1.5  
    ) +  
    geom_point(size = 4)
```



Functions from the `forcats` Package

The `forcats` package provides helpers for reordering factor levels:

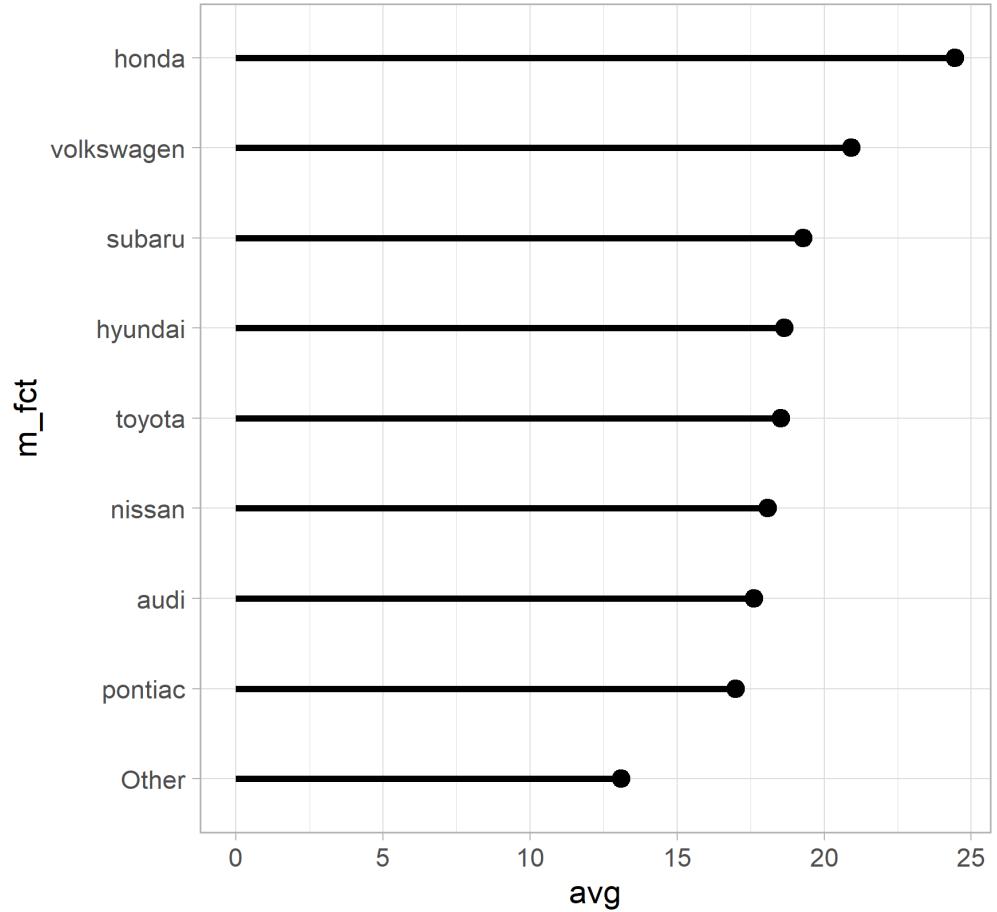
```
mpg_avg %>%
  mutate(
    m_fct = fct_reorder(m_fct, avg)
  ) %>%
  ggplot(aes(avg, m_fct)) +
  geom_segment(
    aes(xend = 0, yend = m_fct),
    size = 1.5
  ) +
  geom_point(size = 4)
```



Functions from the `forcats` Package

The `forcats` package provides helpers for reordering factor levels:

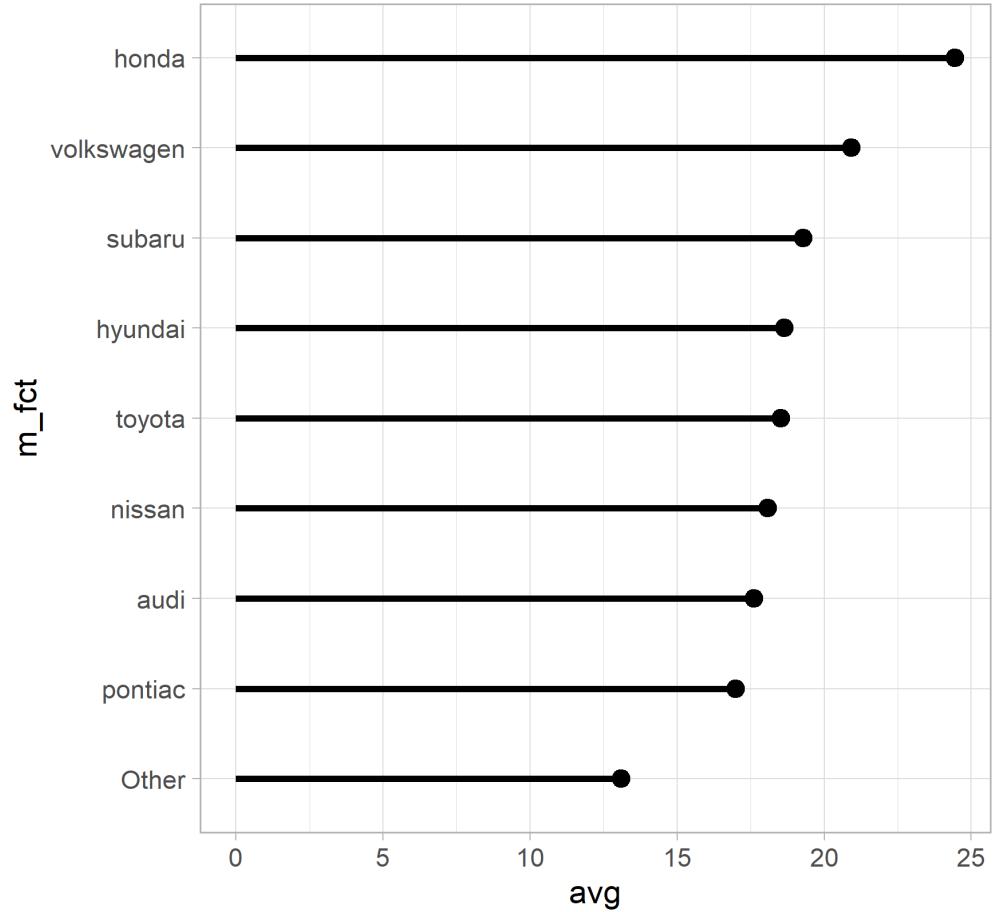
```
mpg_avg %>%
  mutate(m_fct = fct_lump(
    m_fct,
    n = 8,
    w = avg
  )) %>%
  group_by(m_fct) %>%
  summarize(avg = mean(avg)) %>%
  ungroup() %>%
  mutate(m_fct = fct_reorder(m_fct, avg))
ggplot(aes(avg, m_fct)) +
  geom_segment(
    aes(xend = 0, yend = m_fct),
    size = 1.5
  ) +
  geom_point(size = 4)
```



Functions from the `forcats` Package

The `forcats` package provides helpers for reordering factor levels:

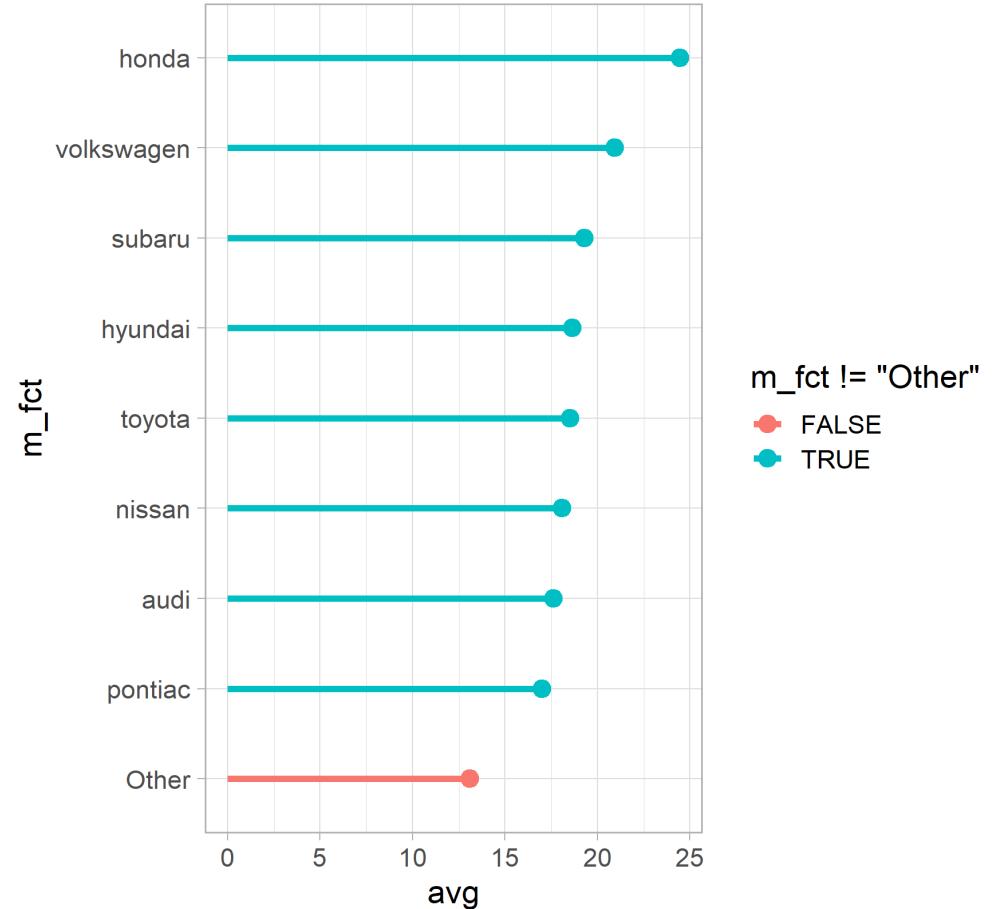
```
mpg_avg %>%
  mutate(m_fct = fct_lump(
    m_fct,
    n = 8,
    w = avg
  )) %>%
  group_by(m_fct) %>%
  summarize(avg = mean(avg)) %>%
  ungroup() %>%
  mutate(m_fct = fct_reorder(m_fct, avg))
ggplot(aes(avg, m_fct)) +
  geom_segment(
    aes(xend = 0, yend = m_fct),
    size = 1.5
  ) +
  geom_point(size = 4)
```



Functions from the `forcats` Package

The `forcats` package provides helpers for reordering factor levels:

```
mpg_avg %>%
  mutate(m_fct = fct_lump(
    m_fct,
    n = 8,
    w = avg
  )) %>%
  group_by(m_fct) %>%
  summarize(avg = mean(avg)) %>%
  ungroup() %>%
  mutate(m_fct = fct_reorder(m_fct, avg))
ggplot(aes(avg, m_fct,
           color = m_fct != "Other")) +
  geom_segment(
    aes(xend = 0, yend = m_fct),
    size = 1.5
  ) +
  geom_point(size = 4)
```





The **stringr** Package

Functions from the `stringr` Package

Functions from the `stringr` package operate on character strings:

```
(manufacturers <- mpg %>% distinct(manufacturer) %>% pull(manufacturer))  
## [1] "audi"       "chevrolet"   "dodge"       "ford"        "honda"  
## [6] "hyundai"    "jeep"        "land rover"  "lincoln"     "mercury"  
## [11] "nissan"     "pontiac"     "subaru"      "toyota"      "volkswagen"
```

Functions from the `stringr` Package

Functions from the `stringr` package operate on character strings:

```
manufacturers <- mpg %>% distinct(manufacturer) %>% pull(manufacturer)

str_sub(manufacturers, 1, 3)
## [1] "aud" "che" "dod" "for" "hon" "hyu" "jee" "lan" "lin" "mer" "nis" "pon"
## [13] "sub" "toy" "vol"
```

Functions from the `stringr` Package

Functions from the `stringr` package operate on character strings, often using REGular EXPRESSIONS:

```
manufacturers <- mpg %>% distinct(manufacturer) %>% pull(manufacturer)

str_sub(manufacturers, 1, 3)
## [1] "aud" "che" "dod" "for" "hon" "hyu" "jee" "lan" "lin" "mer" "nis" "pon"
## [13] "sub" "toy" "vol"

str_subset(manufacturers, "y")
## [1] "hyundai" "mercury" "toyota"
str_subset(manufacturers, "[ab]")
## [1] "audi"      "honda"     "hyundai"    "land rover" "nissan"
## [6] "pontiac"   "subaru"    "toyota"     "volkswagen"
```

Functions from the `stringr` Package

Functions from the `stringr` package operate on character strings, often using REGular EXPRESSIONS:

```
manufacturers <- mpg %>% distinct(manufacturer) %>% pull(manufacturer)

str_sub(manufacturers, 1, 3)
## [1] "aud" "che" "dod" "for" "hon" "hyu" "jee" "lan" "lin" "mer" "nis" "pon"
## [13] "sub" "toy" "vol"

str_subset(manufacturers, "y")
## [1] "hyundai" "mercury" "toyota"
str_subset(manufacturers, "[ab]")
## [1] "audi"      "honda"     "hyundai"    "land rover" "nissan"
## [6] "pontiac"   "subaru"    "toyota"     "volkswagen"

str_detect(manufacturers, "[ab]")
## [1] TRUE FALSE FALSE FALSE  TRUE  TRUE FALSE  TRUE FALSE FALSE  TRUE  TRUE
## [13] TRUE  TRUE  TRUE
```

Functions from the `stringr` Package

Functions from the `stringr` package operate on character strings, often using REGular EXPRESSIONS:

```
manufacturers <- mpg %>% distinct(manufacturer) %>% pull(manufacturer)

str_sub(manufacturers, 1, 3)
## [1] "aud" "che" "dod" "for" "hon" "hyu" "jee" "lan" "lin" "mer" "nis" "pon"
## [13] "sub" "toy" "vol"

str_subset(manufacturers, "y")
## [1] "hyundai" "mercury" "toyota"
str_subset(manufacturers, "[ab]")
## [1] "audi"      "honda"     "hyundai"    "land rover" "nissan"
## [6] "pontiac"   "subaru"    "toyota"     "volkswagen"

str_detect(manufacturers, "[ab]")
## [1] TRUE FALSE FALSE FALSE TRUE  TRUE FALSE TRUE FALSE FALSE TRUE  TRUE
## [13] TRUE  TRUE  TRUE

str_count(manufacturers, "[aeiou]")
## [1] 3 3 2 1 2 3 2 3 2 2 2 3 3 3 3
```

Functions from the `stringr` Package

Functions from the `stringr` package operate on character strings, often using **REGular EXPressions**:

```
manufacturers <- mpg %>% distinct(manufacturer) %>% pull(manufacturer)

str_sub(manufacturers, 1, 3)
## [1] "aud" "che" "dod" "for" "hon" "hyu" "jee" "lan" "lin" "mer" "nis" "pon"
## [13] "sub" "toy" "vol"

str_subset(manufacturers, "y")
## [1] "hyundai" "mercury" "toyota"
str_subset(manufacturers, "[aeiou]$")
## [1] "audi"     "dodge"    "honda"   "hyundai"  "subaru"   "toyota"

str_detect(manufacturers, "[ab]")
## [1] TRUE FALSE FALSE FALSE  TRUE  TRUE FALSE  TRUE FALSE FALSE  TRUE  TRUE
## [13] TRUE  TRUE  TRUE

str_count(manufacturers, "[aeiou]")
## [1] 3 3 2 1 2 3 2 3 2 2 2 3 3 3 3

str_replace(manufacturers, "[aeiou]", "X")
## [1] "Xudi"      "chXvrolet" "dXdge"     "fXrd"      "hXnda"
## [6] "hyXndai"   "jXep"      "lXnd rover" "lXncolin" "mXrcury"
```



The **glue** Package

glue

`glue()` allows you to *glue strings together*.

```
mpg %>%
  select(manufacturer, model, year, trans) %>%
  mutate(
    text = glue::glue("The model {manufacturer} {model} with {trans} as transmission")
  )
## # A tibble: 234 x 5
##   manufacturer model     year trans      text
##   <chr>        <chr>   <int> <chr>      <glue>
## 1 audi         a4      1999 auto(l5) The model audi a4 with auto(l5) as transmission
## 2 audi         a4      1999 manual(~ The model audi a4 with manual(m5) as transmission
## 3 audi         a4      2008 manual(~ The model audi a4 with manual(m6) as transmission
## 4 audi         a4      2008 auto(av) The model audi a4 with auto(av) as transmission
## 5 audi         a4      1999 auto(l5) The model audi a4 with auto(l5) as transmission
## 6 audi         a4      1999 manual(~ The model audi a4 with manual(m5) as transmission
## 7 audi         a4      2008 auto(av) The model audi a4 with auto(av) as transmission
## 8 audi         a4 quat~ 1999 manual(~ The model audi a4 quattro with manual(m5) as transmission
## 9 audi         a4 quat~ 1999 auto(l5) The model audi a4 quattro with auto(l5) as transmission
## 10 audi        a4 quat~ 2008 manual(~ The model audi a4 quattro with manual(m6) as transmission
## # ... with 224 more rows
```

glue in combination with stringr

`glue()` allows you to *glue strings together*.

```
mpg %>%
  select(manufacturer, model, year, trans) %>%
  mutate(text =
    glue::glue("The {str_to_title(manufacturer)} {str_to_title(model)} with {trans} as")
  )
## # A tibble: 234 x 5
##   manufacturer model     year trans      text
##   <chr>        <chr>   <int> <chr>      <glue>
## 1 audi         a4      1999 auto(l5) The Audi A4 with auto(l5) as transissio~
## 2 audi         a4      1999 manual(~ The Audi A4 with manual(m5) as transiss~
## 3 audi         a4      2008 manual(~ The Audi A4 with manual(m6) as transiss~
## 4 audi         a4      2008 auto(av) The Audi A4 with auto(av) as transissio~
## 5 audi         a4      1999 auto(l5) The Audi A4 with auto(l5) as transissio~
## 6 audi         a4      1999 manual(~ The Audi A4 with manual(m5) as transiss~
## 7 audi         a4      2008 auto(av) The Audi A4 with auto(av) as transissio~
## 8 audi         a4 quat~ 1999 manual(~ The Audi A4 Quattro with manual(m5) as ~
## 9 audi         a4 quat~ 1999 auto(l5) The Audi A4 Quattro with auto(l5) as tr~
## 10 audi        a4 quat~ 2008 manual(~ The Audi A4 Quattro with manual(m6) as ~
## ... with 224 more rows
```

Data Transformation with dplyr::

A guide to 37^{*} different behaviours applied to one tibble (tbl)



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