Data wrangling with dplyr

Spring 2023

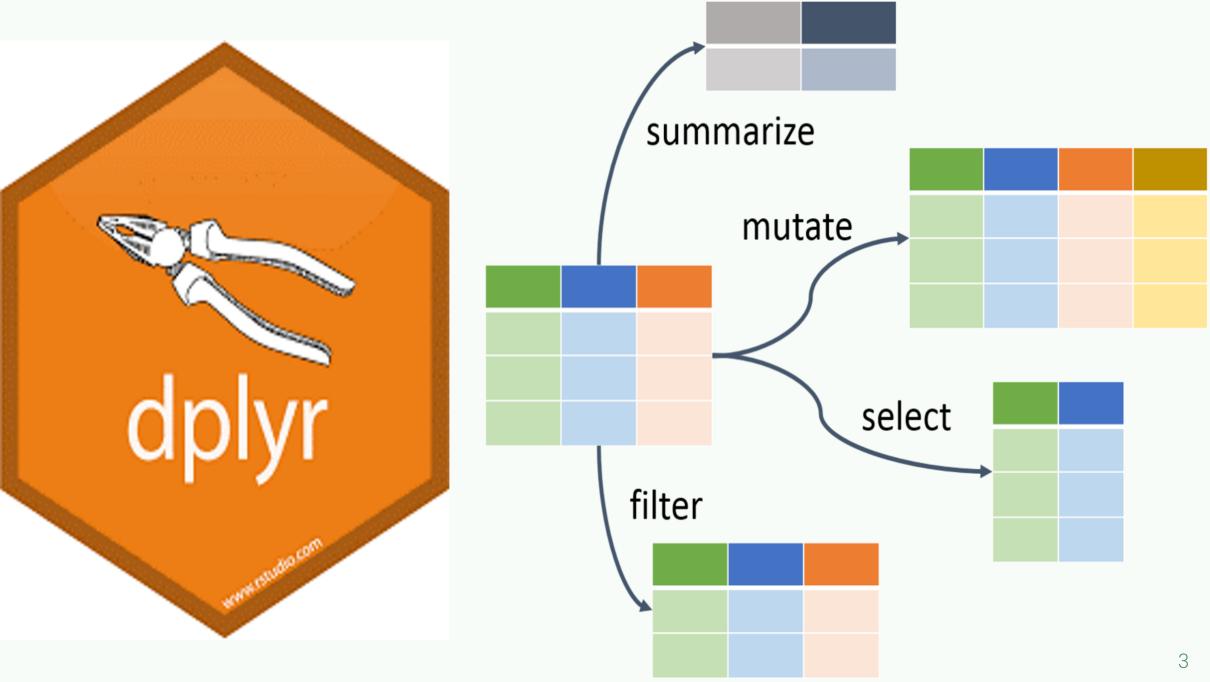
April 07 2023

Data Wrangling

the process of cleaning and unifying messy and complex data sets for easy access and analysis

- "data janitor work"
- importing data
- cleaning data
- changing shape of data

- fixing errors and poorly formatted data elements
- transforming columns and rows
- filtering, subsetting



The Five Verbs

Most of the operations on a data table can be achieved with

- select() : extract a subset of columns
- filter() : extract a subset of rows
- mutate() : create new columns
- arrange() : order the rows from smallest to largest (or largest to smallest)
- summarize() : compute a table of summary statistics

Some Operators

Operator	Definition
<	less than
<=	less than or equal to
>	greater than
>=	greater than or equal to
==	exactly equal to
[!=	not equal to
x & y	x AND y
x %in% y	test if x is in y

Find a subset of the columns using *select()*:

select(): take a subset of the columns (variables/features)

```
library(babynames)
babynames %>%
 select(year, name, n) %>%
 head()
# A tibble: 6 × 3
  vear name
  <dbl> <chr> <int>
  1880 Marv
              7065
             2604
  1880 Anna
  1880 Emma
             2003
  1880 Elizabeth 1939
  1880 Minnie
                 1746
  1880 Margaret
                 1578
```

- %>% passes result on left into first argument of function on right
- Chaining functions together lets you read Left-to-right, top-to-bottom

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```
babynames %>%  # dataframe first and then...
select(year, name, n) %>%  # select columns year, name, and n
head()  # display header of the data frame
```

- %>% passes result on left into first argument of function on right
- Chaining functions together lets you read Left-to-right, top-to-bottom

```
babynames %>%  # dataframe first and then...
select(year, name, n) %>%  # select columns year, name, and n
head()  # display header of the data frame
```

select() helpers

```
* select range of columns
select(gapminder, income:population)
- select every column but
select(gapminder, -c(income,population))
starts_with() select columns
that start with...
select(gapminder, starts_with("p"))
```

```
ends_with() select columns
that end with...
select(gapminder, ends_with("y"))
contains() select columns
whose names contain...
select(gapminder, contains("e"))
```

Find a subset of the rows using *filter()*:

• filter(): take a subset of the rows (observations)

```
babynames %>%
  filter(name == "Bella") %>%
 head()
# A tibble: 6 × 5
  year sex
              name
                        n
                              prop
  <dbl> <chr> <chr> <int>
                             <dbl>
  1880 F
              Bella
                       13 0.000133
  1881 F
             Bella
                       24 0.000243
  1882 F
             Bella
                       16 0.000138
            Bella
  1883 F
                    17 0.000142
  1884 F
             Bella
                       31 0.000225
  1885 F
              Bella
                       25 0.000176
```

Use both filter() and select()

```
bella <- babynames %>%
  filter(name == "Bella") %>%
  select(year, name, sex, n)
```

```
head(bella, 10)
# A tibble: 10 × 4
    year name sex
                         n
   <dbl> <chr> <chr> <int>
   1880 Bella F
                        13
   1881 Bella F
                        24
    1882 Bella F
                        16
   1883 Bella F
                        17
    1884 Bella F
                        31
    1885 Bella F
                         25
    1886 Bella F
                        22
    1887 Bella F
                        26
    1888 Bella F
                        31
    1889 Bella F
10
                         37
```

```
michael_selected_years <- babynames %>%
  filter(name == "Michael", sex == "M",
          year %in% c(1990, 2000, 2010)) %>%
  select(year, name, sex, n)
```

arrange()

Order rows from smallest to largest

```
arrange(babynames, n)
# A tibble: 1,924,665 × 5
    year sex
               name
                                    prop
   <dbl> <chr> <chr>
                     <int>
                                   <dbl>
 1 1880 F
              Adelle
                             5 0.0000512
   1880 F
              Adina
                             5 0.0000512
   1880 F
              Adrienne
                             5 0.0000512
              Albertine
   1880 F
                             5 0.0000512
   1880 F
               Alys
                             5 0.0000512
   1880 F
               Ana
                             5 0.0000512
               Araminta
   1880 F
                             5 0.0000512
   1880 F
              Arthur
                             5 0.0000512
               Birtha
   1880 F
                             5 0.0000512
               Bulah
10
   1880 F
                             5 0.0000512
# ... with 1,924,655 more rows
```

desc()

Changes ordering from largest to smallest.

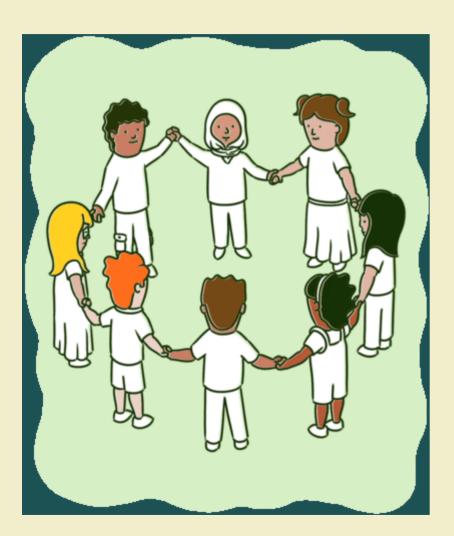
```
arrange(babynames, desc(n))
# A tibble: 1,924,665 × 5
   year sex
              name
                          n
                              prop
   <dbl> <chr> <int> <dbl>
   1947 F
              Linda
                     99686 0.0548
              Linda 96209 0.0552
   1948 F
   1947 M
              James
                      94756 0.0510
              Michael 92695 0.0424
   1957 M
   1947 M
              Robert 91642 0.0493
              Linda
   1949 F
                      91016 0.0518
              Michael 90620 0.0423
   1956 M
   1958 M
              Michael 90520 0.0420
                      88588 0.0497
   1948 M
              James
10
   1954 M
              Michael 88514 0.0428
# ... with 1,924,655 more rows
```

Most common names in 1990

```
babynames %>%
 filter(year == 1990) %>%
 arrange(desc(prop))
# A tibble: 24,719 × 5
   year sex
             name
                            n
                                prop
  <dbl> <chr> <int> <dbl>
1 1990 M
          Michael
                         65282 0.0303
   1990 M
          Christopher 52332 0.0243
   1990 F
             Jessica
                        46475 0.0226
             Ashley 45558 0.0222
   1990 F
   1990 M
             Matthew
                        44800 0.0208
             Joshua
   1990 M
                         43216 0.0201
             Brittany
   1990 F
                         36538 0.0178
   1990 F
             Amanda
                         34408 0.0168
             Daniel
                         33815 0.0157
   1990 M
             David
   1990 M
                         33742 0.0157
# ... with 24,709 more rows
```

10:00

P GROUP ACTIVITY 1



- Let's go over to maize server/ local Rstudio and our class moodle
- Get the class activity 6.Rmd file
- Work on problems 1-3
- Ask me questions

summarize() or summarise()

If we want to compare summary statistics, we might use summarize().

```
babynames %>%
  filter(name == "Bella", sex == "F") %>%
  summarise(total = sum(n), max = max(n), mean = mean(n))
# A tibble: 1 × 3
  total max mean
  <int> <int> <dbl>
1 57411 5121 416.
```

summarize()

```
babynames %>%
  summarize(nname = n_distinct(name))
# A tibble: 1 × 1
  nname
  <int>
1 97310
```

Using group_by()

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

Using group_by() along with summarize()

```
babynames %>%
 group_by(year) %>%
 summarise(total = sum(n))
# A tibble: 138 × 2
   year total
   <dbl> <int>
   1880 201484
   1881 192696
   1882 221533
   1883 216946
   1884 243462
   1885 240854
   1886 255317
   1887 247394
   1888 299473
   1889 288946
# ... with 128 more rows
```

mutate()

• mutate() lets us create new variables based on manipulations of the old variables

```
babynames <- babynames %>%
 group_by(year) %>%
 mutate(percent = prop * 100)
head(babynames)
# A tibble: 6 × 6
# Groups:
         year [1]
  vear sex
             name
                              prop percent
  <dbl> <chr> <chr>
                   <int> <dbl>
                                     <dbl>
  1880 F
                                     7.24
             Marv
                      7065 0.0724
  1880 F
                       2604 0.0267
                                    2.67
             Anna
  1880 F
                                    2.05
            Emma
                       2003 0.0205
           Elizabeth 1939 0.0199
                                    1.99
  1880 F
            Minnie
  1880 F
                       1746 0.0179
                                     1.79
  1880 F
             Margaret
                      1578 0.0162
                                      1.62
```

top_n()

Most common name in each year

```
babynames %>%
 group_by(year) %>%
 top_n(1, prop)
# A tibble: 138 × 6
# Groups: year [138]
   year sex
               name
                         n
                             prop percent
   <dbl> <chr> <int> <dbl>
                                    <dbl>
   1880 M
               John
                      9655 0.0815
                                     8.15
    1881 M
               John
                      8769 0.0810
                                    8.10
   1882 M
               John
                      9557 0.0783
                                    7.83
   1883 M
               John
                      8894 0.0791
                                    7.91
    1884 M
               John
                      9388 0.0765
                                    7.65
    1885 M
               John
                      8756 0.0755
                                     7.55
    1886 M
               John
                      9026 0.0758
                                    7.58
    1887 M
               John
                                    7.42
                      8110 0.0742
    1888 M
               John
                      9247 0.0712
                                    7.12
   1889 M
               John
                                     7.18
10
                      8548 0.0718
# ... with 128 more rows
```

min_rank() : A go to ranking function (ties share the lowest rank)

```
min_rank(c(50, 100, 1000))
[1] 1 2 3
```

```
min_rank(desc(c(50, 100, 1000)))
[1] 3 2 1
```

Slicing and selecting data

```
The slice_ operators let you slice (subset) rows:
• slice_head(n=5) : view the first 5 rows
• slice_tail(n=5) : view the last 5 rows
• slice_sample(n=5) : view 5 random rows
• slice_min(column, n=5) : view the 5 smallest values of a
  column
 slice_max(column, n=5) : view the 5 largest values of a
  column
```

slice()

```
library(gapminder)
slice(gapminder, 1:5)
# A tibble: 5 × 6
 country continent year lifeExp
                                        pop gdpPercap
  <fct>
        <fct>
                      <int> <dbl>
                                      <int>
                                                <dbl>
 Afghanistan Asia
                       1952 28.8 8425333
                                                 779.
2 Afghanistan Asia
                       1957 30.3 9240934
                                                 821.
3 Afghanistan Asia
                       1962
                                                 853.
                               32.0 10267083
4 Afghanistan Asia
                       1967 34.0 11537966
                                                 836.
5 Afghanistan Asia
                       1972
                               36.1 13079460
                                                 740.
```

slice_max()

```
gapminder %>% slice_max(gdpPercap, n=6)
# A tibble: 6 × 6
  country continent year lifeExp
                                      pop gdpPercap
  <fct>
          <fct>
                    <int>
                            <dbl>
                                    <int>
                                              <dbl>
        Asia
1 Kuwait
                     1957
                             58.0
                                   212846
                                            113523.
2 Kuwait
        Asia
                     1972
                             67.7 841934
                                            109348.
3 Kuwait
        Asia
                     1952
                             55.6
                                   160000
                                            108382.
4 Kuwait
        Asia
                     1962
                             60.5
                                   358266
                                             95458.
5 Kuwait
         Asia
                     1967
                             64.6 575003
                                             80895.
6 Kuwait
         Asia
                     1977
                                             59265.
                             69.3 1140357
```

summarize() vs. mutate()

```
summarize() : summarize collapses all variable values down to one number
(by group)
```

```
gapminder %>%
  group_by(continent) %>%
 summarize(avg_life_expectancy = mean(lifeExp))
# A tibble: 5 × 2
  continent avg_life_expectancy
  <fct>
                          <dbl>
1 Africa
                           48.9
2 Americas
                           64.7
3 Asia
                           60.1
4 Europe
                           71.9
5 Oceania
                           74.3
```

summarize() vs. mutate()

mutate() : transforms all variable values but preserves the variable length (by group)

```
gapminder %>%
 group_by(continent) %>%
 mutate(meanPop = mean(pop)/1000000)
# A tibble: 1.704 × 7
           continent [5]
# Groups:
              continent year lifeExp pop gdpPercap meanPop
  country
  <fct>
              <fct>
                       <int>
                               <dbl>
                                        <int>
                                                 <dbl>
                                                        <dbl>
1 Afghanistan Asia
                       1952
                                28.8
                                      8425333
                                                  779.
                                                        77.0
2 Afghanistan Asia
                        1957 30.3
                                      9240934
                                                  821. 77.0
3 Afghanistan Asia
                        1962
                               32.0 10267083
                                                  853. 77.0
4 Afghanistan Asia
                        1967
                                34.0 11537966
                                                  836.
                                                        77.0
 5 Afghanistan Asia
                        1972
                                36.1 13079460
                                                  740.
                                                         77.0
6 Afghanistan Asia
                        1977
                                38.4 14880372
                                                  786.
                                                        77.0
7 Afghanistan Asia
                        1982
                                39.9 12881816
                                                  978.
                                                        77.0
8 Afghanistan Asia
                                40.8 13867957
                        1987
                                                  852.
                                                        77.0
9 Afghanistan Asia
                        1992
                                41.7 16317921
                                                  649. 77.0
10 Afghanistan Asia
                        1997
                               41.8 22227415
                                                  635. 77.0
# ... with 1,694 more rows
```

group_by(var1, var2)

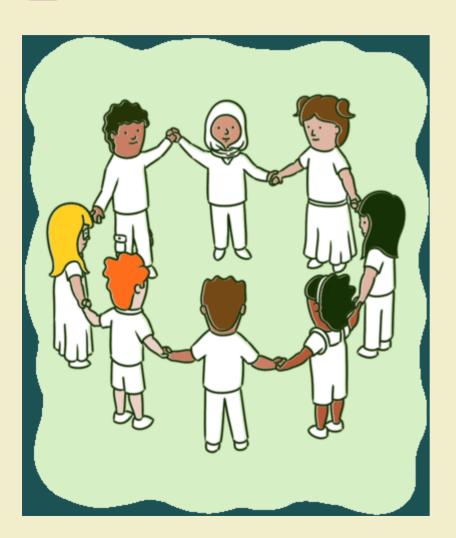
```
gapminder %>%
  group_by(continent, year) %>%
  summarise(avg_life_expectancy = mean(lifeExp)) %>%
  slice_max(avg_life_expectancy, n = 1)
# A tibble: 5 × 3
# Groups: continent [5]
  continent year avg_life_expectancy
  <fct>
           <int>
                                <dbl>
1 Africa
            2007
                                 54.8
2 Americas
            2007
                                 73.6
3 Asia
            2007
                                70.7
4 Europe
            2007
                                 77.6
5 Oceania
            2007
                                 80.7
```

ungroup()

Any further mutations called on it would not use the grouping for aggregate statistics.

10:00

B GROUP ACTIVITY 2



- Work on problems 4-6
- Ask me questions