# More data Wrangling with joins and tidyR

Math 241, Week 4

```
# it's good practice to check that all the packages required are loaded and installed
libs <- c('tidyverse','knitr','viridis','mosaicData','babynames','mdsr','Lahman','nycflights13')
for(l in libs){
   if(!require(l,character.only = TRUE, quietly = TRUE)){
     message( sprintf('Did not have the required package << %s >> installed. Downloading now ... ',l))
   install.packages(l)
   }
   library(l, character.only = TRUE, quietly = TRUE)
}
```

### Goals of this in-class activity:

• Practice data wrangling and joins with tidyR

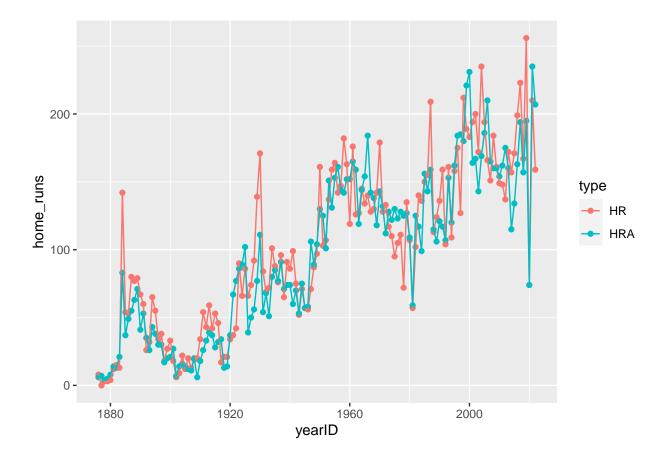
#### Notes:

• Be prepared to ask for help from me, Tory, and your classmates!

### Problem 1 (Medium):

Consider the number of home runs hit (HR) and home runs allowed (HRA) for the Chicago Cubs (CHN) baseball team. Reshape the Teams data from the Lahman package into "long" format and plot a time series conditioned on whether the HRs that involved the Cubs were hit by them or allowed by them.

```
Teams %>%
  filter(teamID == "CHN") %>%
  select(yearID, HR, HRA) %>%
  pivot_longer(-yearID, names_to = "type", values_to = "home_runs") %>%
  ggplot(aes(x = yearID, y = home_runs, color = type)) +
  geom_point() +
  geom_line()
```



# Problem 2 (Medium):

Use the nycflights13 package and the flights and planes tables to answer the following questions:

a. How many planes have a missing date of manufacture?

```
library(nycflights13)
planes2 <- select(planes, tailnum, year, manufacturer)
flights2 <- select(flights, tailnum)
nyc_flights <- left_join(planes2, flights2)
nyc_flights2 <- nyc_flights %>%
   filter(is.na(year)) %>%
   distinct(tailnum)
nrow(nyc_flights2)
```

### ## [1] 70

There are 70 airplanes with a missing date of manufacture.

b. What are the five most common manufacturers?

```
nyc_flights %>%
  select(manufacturer, tailnum, year) %>%
  unique() %>%
```

```
group_by(manufacturer) %>%
summarize(count = n()) %>%
arrange(desc(count))
```

```
## # A tibble: 35 x 2
##
      manufacturer
                                     count
##
      <chr>>
                                     <int>
##
   1 BOEING
                                      1630
    2 AIRBUS INDUSTRIE
                                       400
    3 BOMBARDIER INC
                                       368
##
##
  4 AIRBUS
                                       336
## 5 EMBRAER
                                       299
## 6 MCDONNELL DOUGLAS
                                       120
   7 MCDONNELL DOUGLAS AIRCRAFT CO
                                       103
## 8 MCDONNELL DOUGLAS CORPORATION
                                        14
## 9 CANADAIR
                                         9
## 10 CESSNA
                                         9
## # i 25 more rows
```

### Problem 3 (Medium):

Use the nycflights13 package and the flights and planes tables to answer the following questions:

a. What is the oldest plane (specified by the tailnum variable) that flew from New York City airports in 2013?

```
planes2 <- dplyr::select(planes, tailnum, year)
flights2 <- dplyr::select(flights, tailnum)
nyc_flights <- left_join(planes2, flights2) %>%
    arrange(year) %>%
    unique()
head(nyc_flights)
```

```
## # A tibble: 6 x 2
##
     tailnum year
##
     <chr>>
             <int>
## 1 N381AA
              1956
## 2 N201AA
              1959
## 3 N567AA
              1959
## 4 N378AA
              1963
## 5 N575AA
              1963
## 6 N14629
              1965
```

N381AA, manufactured in 1956, is the oldest plane that flew from NYC in 2013.

b. How many airplanes that flew from New York City are included in the planes table?

```
nyc_flights2 <- distinct(nyc_flights)
nrow(nyc_flights2)</pre>
```

```
## [1] 3322
```

There are 3322 unique airplanes.

## Problem 4 (Medium):

The knitr package allows the analyst to display nicely formatted tables and results when outputting to pdf files. Use the following code chunk as an example to create a similar display for the penguins dataset, in the palmerpenguins package, instead (you can model penguins' body\_mass\_g as a function of their flipper\_length\_mm and sex):

```
mod <- broom::tidy(lm(cesd ~ mcs + sex, data = HELPrct))
knitr::kable(
  mod,
  digits = c(0, 2, 2, 2, 4),
  caption = "Regression model from HELP clinical trial.",
  longtable = TRUE
)</pre>
```

Table 1: Regression model from HELP clinical trial.

| term        | estimate | std.error | statistic | p.value |
|-------------|----------|-----------|-----------|---------|
| (Intercept) | 55.79    | 1.31      | 42.62     | 0.0000  |
| mcs         | -0.65    | 0.03      | -19.48    | 0.0000  |
| sexmale     | -2.95    | 1.01      | -2.91     | 0.0038  |

```
library(palmerpenguins)
mod <- broom::tidy(lm(body_mass_g ~ flipper_length_mm + sex, data = penguins))
knitr::kable(mod, digits = c(0, 1, 1, 1, 4), longtable = TRUE)</pre>
```

| term              | estimate | std.error | statistic | p.value |
|-------------------|----------|-----------|-----------|---------|
| (Intercept)       | -5410.3  | 285.8     | -18.9     | 0       |
| flipper_length_mm | 47.0     | 1.4       | 32.6      | 0       |
| sexmale           | 347.9    | 40.3      | 8.6       | 0       |