# Midterm: Data Science CS3072 Spring 2023

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# **Packages**

```
library(tidyverse)
```

#### Data

```
people <- read_csv("data/people.csv")
pitching <- read_csv("data/pitching.csv")
salaries <- read_csv("data/salaries.csv")
teams <- read_csv("data/teams.csv")</pre>
```

#### Exercise 1

1. Using group\_by(), find the top 3 players in pitching dataset who got the maximum total number of earned\_runs for the years 2014 through 2019.

```
pitching %>% filter (year_id >= 2014, year_id <= 2019) %>%
  group_by(player_id, year_id) %>%
  summarise(total_earned_runs = sum(earned_runs))%>%
  arrange(desc(total_earned_runs)) %>%
  head(3)
```

```
## 'summarise()' has grouped output by 'player_id'. You can override using the
## '.groups' argument.
## # A tibble: 3 x 3
## # Groups: player_id [3]
    player_id year_id total_earned_runs
##
                 <dbl>
     <chr>>
                                   <dbl>
## 1 giolilu01
                  2018
                                     118
                  2015
## 2 samarje01
                                     118
## 3 shielja02
                  2016
                                     118
```

## Exercise 2

2. Join the people data to the salaries data and mutate() a new variable with each player's approximate age. Call this dataset player\_income\_age. This dataset will have 26428 rows.

```
player_income_age <- inner_join(people,salaries,by="player_id") %>%
  mutate(age = year_id - birth_year)

## Warning in inner_join(people, salaries, by = "player_id"): Each row in 'x' is expected to match at m
## i Row 1 of 'x' matches multiple rows.

## i If multiple matches are expected, set 'multiple = "all"' to silence this
## warning.
```

#### Exercise 3

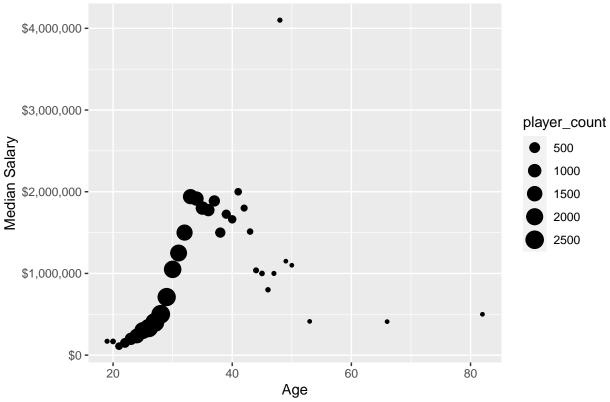
3. Based on the player\_income\_age dataset, create a new dataset called player\_stats\_by\_age with the median\_salary and count of players for all possible ages using group\_by() and summarize(). The new dataset will have 35 rows.

```
player_stats_by_age <- player_income_age %>%
  group_by(age) %>%
  summarise(median_salary = median(salary), player_count = n())
```

#### Exercise 4

4. Construct a plot of median salary versus age with points sized by how many pitchers are at that age. Describe what you observe.





## Exercise 5

- 5. Use the teams dataset to find each team's win percentage for the years 2011 through 2016. Save the result as a tibble named team\_stats. This tibble should have three columns (year\_id, team\_id, and win\_pct) and 180 rows. Steps:
- 6. Filter the data for the given years
- 7. Add a new variable (win pct)
- 8. Select only the three columns
- 9. Save the data in a new dataframe (team\_stats)

```
team_stats <- teams %>%
  filter(year_id >= 2011, year_id <= 2016) %>%
  mutate(win_pct = (wins / games)) %>%
  select(year_id, team_id, win_pct)
```

## Exercise 6

6. Use an appropriate join function to add salary information to players in the pitching data. You should only include observations that appear in both pitching and salaries. Then, using a group\_by() paired with a summarize(), create a new column giving the total amount of money (sum()) spent on pitching salaries by each team in each year. Save the result as a tibble named team\_spending. This dataset should have three columns (year\_id, team\_id, and pitching\_salaries) and 918 rows. Steps:

- 7. Join pitching and salaries by player, year, and team ().
- 8. Group by year and team
- 9. Create a new variable (pitching\_Salaries) using summarize
- 10. Save the data in new dataframe (team\_spending) Note: when you join two datasets based on more than one variable, the syntax is: inner\_join( dataframe1, dataframe2, by = c("var1", "var2", "var3"))

```
## 'summarise()' has grouped output by 'year_id'. You can override using the
## '.groups' argument.
```

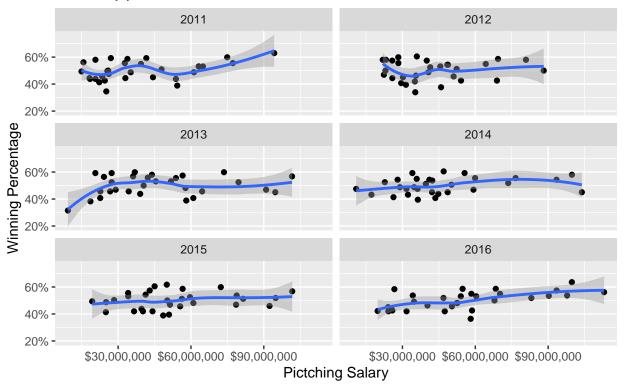
#### Exercise 7

- 7. Use an appropriate function to join team\_stats and team\_spending and use faceting to create subplots of win percentage versus pitching spending for each year. Steps:
- 8. Join team\_stats and team\_spending by year and team
- 9. Provide data visualization using geom point and geom smooth

## 'geom\_smooth()' using method = 'loess' and formula = 'y ~ x'

# Win Percentage by Pitching Salary

## faceted by year



# Exercise 8

- 8. Find the pitcher with the most strikeouts (strike\_outs in the pitching dataset) in each year, by league, from 1901 to the most recent year in the dataset. Save the result as a tibble named top\_so\_by\_year and glimpse() the dataset. Your final dataset should have 243 rows (there are some ties) and the columns year\_id, league\_id,first\_name, and last\_name, where the name information is from the people dataset. Steps:
- 9. Filter the pitching data from 1901 to the most recent year
- 10. Join the filtered data with people dataset by player
- 11. Group the joined data by year and league
- 12. Filter the data to have only the max strike outs
- 13. Save the data in ne dataframe (top\_so\_by\_year) year\_id, league\_id,first\_name, and last\_name

```
top_so_by_year <- pitching %>%
  filter(year_id >= 1901) %>%
  inner_join(y = people, by = "player_id") %>%
  group_by(year_id, league_id) %>%
  slice_max(strike_outs, n = 1) %>%
  select(year_id, league_id, first_name, last_name) %>%
  glimpse()
```

## Rows: 243 ## Columns: 4

## Groups: year\_id, league\_id [240]