## Task-03

#### Import necessary libraries

1 ""

##

```
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':

##
## filter, lag

## The following objects are masked from 'package:base':

##
## intersect, setdiff, setequal, union

library(ggplot2)
library(knitr)

Loading data
```

#### Identify the 10 users who have written the most reviews

3.00

reviews <- read.csv("../data/reviews.csv", stringsAsFactors = FALSE)

Identify the 10 most active users by counting the number of reviews per user\_id from the reviews dataset, sorting them in descending order based on the number of reviews, and selecting the top 10 users top\_users\_by\_count. Next, filters the original reviews dataset to include only reviews from these top 10 users reviews\_from\_top\_users. After that, for each of the 10 users, the algorithm calculates the average star rating they have given, ignoring any missing values NA in the calculation average\_stars\_for\_top\_users.

```
top_users_by_count <- reviews %>%
  group_by(user_id) %>%
  summarise(review_count = n(), .groups = 'drop') %% # Group the rows in the reviews dataset by the us
  arrange(desc(review_count)) %>% # sorting the results
  slice_head(n = 10) # get 10 rows
reviews_from_top_users <- reviews %>%
  filter(user_id %in% top_users_by_count$user_id)
# calculate average stars for each of these top users.
# na.rm = TRUE ensures that NA values in the 'stars' column are ignored during mean calculation.
average_stars_for_top_users <- reviews_from_top_users %>%
  group_by(user_id) %>%
  summarise(average_stars = mean(stars, na.rm = TRUE), .groups = 'drop')
print(average_stars_for_top_users)
## # A tibble: 10 x 2
##
      user_id average_stars
##
      <chr>
                        <dbl>
```

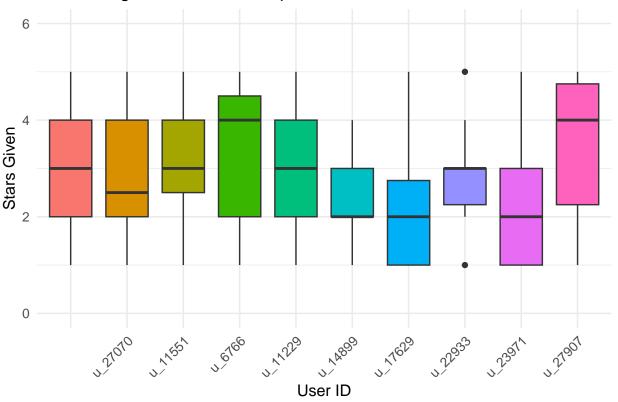
```
## 2 "u 11229"
                        3.07
## 3 "u_11551"
                        3.27
## 4 "u 14899"
                        2.57
## 5 "u_17629"
                        2.21
## 6 "u 22933"
                        2.93
## 7 "u 23971"
                        2.36
## 8 "u 27070"
                        2.83
## 9 "u 27907"
                        3.43
## 10 "u_6766"
                        3.27
```

Then, merges *left\_join* the review count information *top\_users\_by\_count* with the average star ratings *average\_stars\_for\_top\_users* for each user into a single summary table called *summary\_top\_users*, which is then sorted again based on the number of reviews

```
## Table: Top 10 Users: Review Count and Average Stars
## | User ID | Total Reviews | Average Stars |
## |:----:|:---
          ## |
                 5829
                                3.00
## | u_27070 |
                  18
                                2.83
                                3.27
## | u_11551 |
                 15
                 15
## | u_6766 |
                                3.27
## | u_11229 |
                  14
                          -
                                3.07
## | u_14899 |
                  14
                                2.57
                         ## | u 17629 |
                  14
                               2.21
## | u_22933 |
                               2.93
                  14
                          -
## | u_23971 |
                  14
                          2.36
                                         1
## | u_27907 |
                  14
                          3.43
```

### Visualize

# Star Rating Distribution for Top 10 Users



The boxplot shows how the top users vary in the way they give star ratings. Users such as u\_6766 and u\_27907 tend to give higher ratings overall, with medians above 3 and upper quartiles close to 5. In contrast, users like u\_17629 and u\_22933 give lower ratings more frequently, as indicated by their lower medians and compressed upper ranges. Some users show wide variability (e.g., u\_6766, u\_27907), while others have more consistent rating patterns (e.g., u\_14899). Outliers in a few users suggest occasional extreme ratings.