

Project 1

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2022-09-20

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1 Solution

1.1 Read tournament data

Read input data. Since we will be extracting rows/columns we can convert to matrix format for easier downstream processing.

```
input_matrix <- read.csv("https://raw.githubusercontent.com/himalayahall/DATA607/main/Project1/tournament.csv")
input_matrix <- matrix(unlist(input_matrix))
```

1.2 Data preparation

We notice the first 3 rows are header rows and don't contain player info.

```
head(input_matrix, n = 10)
```

```
##      [,1]
## [1,] " Pair | Player Name          |Total|Round|Round|Round|Round|Round|Round|Round| "
## [2,] " Num  | USCF ID / Rtg (Pre->Post) | Pts | 1 | 2 | 3 | 4 | 5 | 6 | 7 | "
## [3,] "-----"
## [4,] "      1 | GARY HUA                |6.0  |W 39|W 21|W 18|W 14|W 7|D 12|D 4|"
```

```
## [5,] "    ON | 15445895 / R: 1794    ->1817    |N:2 |W    |B    |W    |B    |W    |B    |W    |"
## [6,] "-----"
## [7,] "    2 | DAKSHESH DARURI    |6.0 |W 63|W 58|L 4|W 17|W 16|W 20|W 7|"
## [8,] "    MI | 14598900 / R: 1553    ->1663    |N:2 |B    |W    |B    |W    |B    |W    |B    |"
## [9,] "-----"
## [10,] "    3 | ADITYA BAJAJ    |6.0 |L 8|W 61|W 25|W 21|W 11|W 13|W 12|"
```

Skip first 3 header rows.

```
input_matrix <- input_matrix[-1:-3]
head(input_matrix, n = 6)
```

```
## [1] "    1 | GARY HUA    |6.0 |W 39|W 21|W 18|W 14|W 7|D 12|D 4|"
## [2] "    ON | 15445895 / R: 1794    ->1817    |N:2 |W    |B    |W    |B    |W    |B    |W    |"
## [3] "-----"
## [4] "    2 | DAKSHESH DARURI    |6.0 |W 63|W 58|L 4|W 17|W 16|W 20|W 7|"
## [5] "    MI | 14598900 / R: 1553    ->1663    |N:2 |B    |W    |B    |W    |B    |W    |B    |"
## [6] "-----"
```

Data for a player is provided on 2 rows. First row gives the name of player and games played by them. The second row gives the State and initial rank of the player. For any given player, the 2 data rows appear consecutively, followed by a dashed separator line.

Using this observation, we split the input matrix into 2 components.

The 1st component will contain player name, total number of points, and games played by them. We can extract this data by starting at row 1 and scooping up every 3rd row from the input matrix (skipping over state/rank and separator line).

```
mPlayersAndGames <- input_matrix[seq(1, length(input_matrix), 3)]
head(mPlayersAndGames, n = 5)
```

```
## [1] "    1 | GARY HUA    |6.0 |W 39|W 21|W 18|W 14|W 7|D 12|D 4|"
## [2] "    2 | DAKSHESH DARURI    |6.0 |W 63|W 58|L 4|W 17|W 16|W 20|W 7|"
## [3] "    3 | ADITYA BAJAJ    |6.0 |L 8|W 61|W 25|W 21|W 11|W 13|W 12|"
## [4] "    4 | PATRICK H SCHILLING    |5.5 |W 23|D 28|W 2|W 26|D 5|W 19|D 1|"
## [5] "    5 | HANSHI ZUO    |5.5 |W 45|W 37|D 12|D 13|D 4|W 14|W 17|"
```

The 2nd component will contain player State and pre-rating. We extract this data by starting at row 2 and, as before, scooping up every 3rd row from the input matrix (skipping over the separator line and next player's name).

```
mStatesAndRanks <- input_matrix[seq(2, length(input_matrix), 3)]
head(mStatesAndRanks, n = 5)
```

```
## [1] "    ON | 15445895 / R: 1794    ->1817    |N:2 |W    |B    |W    |B    |W    |B    |W    |"
## [2] "    MI | 14598900 / R: 1553    ->1663    |N:2 |B    |W    |B    |W    |B    |W    |B    |"
## [3] "    MI | 14959604 / R: 1384    ->1640    |N:2 |W    |B    |W    |B    |W    |B    |W    |"
## [4] "    MI | 12616049 / R: 1716    ->1744    |N:2 |W    |B    |W    |B    |W    |B    |B    |"
## [5] "    MI | 14601533 / R: 1655    ->1690    |N:2 |B    |W    |B    |W    |B    |W    |B    |"
```

1.3 Generate static player data

At this point we have the necessary components to generate static data for each player - player id, name, state, total points, and pre-rating.

Note, there are 64 players in data set.

```
player_id <- as.integer(str_extract(mPlayersAndGames, '\\d+'))

player_name <- str_trim(str_extract(mPlayersAndGames, "[A-Z][^\\|]+")) # assume names start with a lett

player_state <- str_extract(mStatesAndRanks, "[A-Z][A-Z]") # assume 2-letter abbreviation f

player_total_points <- as.numeric(str_extract(mPlayersAndGames, "[0-9]+\\.?[0-9]"))

player_pre_rating <- as.numeric(str_remove(str_extract(mStatesAndRanks, "R:[ ]+[0-9]{1,}"), "R:[ ]+"))

NROW(player_name)
```

```
## [1] 64
```

It's a good idea to spot check our static data by comparing with the original. Let's print some values from both original and static data to compare visually:

Full names of all players have been successfully extracted. For example, SOFIA ADINA STANESCU-BELLU
Player SOFIA ADINA STANESCU-BELLU has 3.5 total points and 1507 pre-rank in original and static data
Visual checks pass for other players as well so we are good to go!

```
idx <- c(1, 12, 22, 28, 58, 64)
sanity_check <- data.frame(
  player_name[idx],
  player_total_points[idx],
  player_pre_rating[idx]
)

colnames(sanity_check) <- c('name', 'total_points', 'init_rank')

sanity_check
```

```
##           name total_points init_rank
## 1      GARY HUA          6.0      1794
## 2  KENNETH J TACK          4.5      1663
## 3  EUGENE L MCCLURE          4.0      1555
## 4 SOFIA ADINA STANESCU-BELLU          3.5      1507
## 5      VIRAJ MOHILE          2.0        917
## 6        BEN LI          1.0      1163
```

```
mPlayersAndGames[idx]
```

```
## [1] "      1 | GARY HUA                |6.0 |W 39|W 21|W 18|W 14|W 7|D 12|D 4|"
## [2] "     12 | KENNETH J TACK           |4.5 |W 42|W 33|D 5|W 38|H   |D 1|L 3|"
## [3] "     22 | EUGENE L MCCLURE             |4.0 |W 64|D 52|L 28|W 15|H   |L 17|W 40|"
## [4] "     28 | SOFIA ADINA STANESCU-BELLU    |3.5 |W 24|D 4|W 22|D 19|L 20|L 8|D 36|"
## [5] "     58 | VIRAJ MOHILE                 |2.0 |W 31|L 2|L 41|L 23|L 49|B   |L 45|"
## [6] "     64 | BEN LI                       |1.0 |L 22|D 30|L 31|D 49|L 46|L 42|L 54|"
```

```
mStatesAndRanks[idx]
```

```
## [1] "   ON | 15445895 / R: 1794  ->1817  |N:2 |W   |B   |W   |B   |W   |B   |W   |"
## [2] "   MI | 12681257 / R: 1663  ->1670  |N:3 |W   |B   |W   |B   |   |   |W   |B   |"
## [3] "   MI | 12405534 / R: 1555  ->1529  |N:4 |W   |B   |W   |B   |   |   |W   |B   |"
## [4] "   MI | 14882954 / R: 1507  ->1513  |N:3 |W   |W   |B   |W   |B   |B   |W   |"
## [5] "   MI | 14700365 / R: 917    -> 941   |   |W   |B   |W   |B   |W   |   |B   |"
## [6] "   MI | 15006561 / R: 1163  ->1112  |   |B   |W   |W   |B   |W   |B   |B   |"
```

1.4 Missing opponent id

We observe that some games do not contain the id of the opposing player. For example, there is no id of the opposing player for games 6 and 7 played by JULIA SHEN. Similarly, only game 1 played by ASHWIN BALAJI has the opposing player id.

```
mPlayersAndGames[60]
```

```
## [1] "     60 | JULIA SHEN                |1.5 |L 33|L 34|D 45|D 42|L 24|H   |U   |"
```

```
mPlayersAndGames[62]
```

```
## [1] "     62 | ASHWIN BALAJI             |1.0 |W 55|U   |U   |U   |U   |U   |U   |"
```

Visual inspection of the full data shows missing player id for games with codes [H, U, B, X].

To make downstream processing more robust let's replace missing opposing player ids with NA. After the transformation we observe that missing values have been replaced by NA.

```
mPlayersAndGames <- str_replace_all(mPlayersAndGames, "\\|([HUBX])([ \\t\\f\\n])+", "\\|\\1\\2 \\N\\A")
```

```
mPlayersAndGames[60]
```

```
## [1] "     60 | JULIA SHEN                |1.5 |L 33|L 34|D 45|D 42|L 24|H   NA|U   NA|"
```

```
mPlayersAndGames[62]
```

```
## [1] "     62 | ASHWIN BALAJI             |1.0 |W 55|U   NA|U   NA|U   NA|U   NA|U   NA|"
```

1.5 Data preparation for computing average opponent ranking

Now we extract all opposing player ids into a flattened list. Validate that there are exactly 64 * 7 ids (64 players and 7 games per player) since we made sure that missing ids were replaced by 0.

```
p_opponent_ids <- as.integer(str_remove(unlist(str_extract_all(mPlayersAndGames, "[A-Z][ ]+([0-9]+|NA)"),
```

```
## Warning: NAs introduced by coercion
```

```
length(p_opponent_ids) == 64 * 7
```

```
## [1] TRUE
```

Scores for exactly 7 games were reported for each player. So we can split opposing player ids into partitions of 7 each. Index into the resulting opponents list is the player id!

For example, ADITYA BAJAJ has player id 3, so ids of ADITYA's opponents will be found at index 3.

```
p_opponents <- split(p_opponent_ids,
  cut(seq_along(p_opponent_ids),
    length(mPlayersAndGames),
    labels = FALSE))
```

```
test_id <- 3 # ADITYA's id
```

```
mPlayersAndGames[test_id] # ADITYA's games
```

```
## [1] "      3 | ADITYA BAJAJ                |6.0 |L  8|W 61|W 25|W 21|W 11|W 13|W 12|"
```

```
p_opponents[test_id] # ADITYA's opponents
```

```
## $'3'
```

```
## [1]  8 61 25 21 11 13 12
```

1.6 Calculate average rank of opponents for all players

Now we are ready to calculate the average rank of opponents for each player.

- Count the number of opponents, ignoring NA opponent id
- Sum the ranks of all opponents
- Compute average opponent rank

```
opp_count <- map(p_opponents,
  function(x) sum(!is.na(x))) # count all opponent ids that are not NA
```

```
opp_prerating_sum <- map(p_opponents, # sum pre-rating of all opponents
  function(x) sum(player_pre_rating[x],
    na.rm = TRUE))
```

```
opp_avg_prerating <- map2(opp_prerating_sum, # compute player average score
  opp_count,
  ~ round(.x / .y, 0))
```

```
tail(unlist(opp_count, # opponent count for last 5 players
  use.names = FALSE),
  5)
```

```
## [1] 5 7 1 5 7
```

```
tail(unlist(opp_prerating_sum, # sum of opponent pre-ratings
           use.names = FALSE),
     5)
```

```
## [1] 6651 9291 1186 6751 8841
```

```
tail(unlist(opp_avg_prerating, # average opponent score
           use.names = FALSE),
     5)
```

```
## [1] 1330 1327 1186 1350 1263
```

1.7 Create result data frame

Now let's warp all computed attributes into the result data frame.

```
player_opp_avg_prerating <- unlist(opp_avg_prerating) # unlist

result_df <- data.frame(player_id, # create data.frame
                        player_name,
                        player_state,
                        player_total_points,
                        player_pre_rating,
                        player_opp_avg_prerating)

colnames(result_df) <- c('id', # change column names
                        'name',
                        'state',
                        'tot_points',
                        'pre_rating',
                        'avg_opponent_pre_rating')
```

Lets take a look at the final results. As sanity check, notice there is data for 64 players.

```
NROW(result_df)
```

```
## [1] 64
```

```
head(result_df, n = 5)
```

##	id	name	state	tot_points	pre_rating	avg_opponent_pre_rating
## 1	1	GARY HUA	ON	6.0	1794	1605
## 2	2	DAKSHESH DARURI	MI	6.0	1553	1469
## 3	3	ADITYA BAJAJ	MI	6.0	1384	1564
## 4	4	PATRICK H SCHILLING	MI	5.5	1716	1574
## 5	5	HANSHI ZUO	MI	5.5	1655	1501

1.8 Generate output CSV

Now we can generate the output CSV file in current working directory. No need to generate row names, player ids already start at 1 and increase in increments of 1.

```
write.csv(result_df, "player_analysis.csv", row.names = FALSE)
```