

Data 607 - Project 1 - Data Cleaning

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Introduction

The goal of this project is to turn a sample of messy chess tournament data into a readable, ingestible .csv format. The .txt file containing the raw data can be found at the following link. The desired output of the data is to create a .csv file with the following variables: player name, player state, total number of points won, player's pre-tournament ELO rating, and average ELO rating of all the player's opponents.

Import Data from CSV

The first step is to import the data from the location above. The data is initially read in as a single string, and a sample of the raw data is printed out below using the `substr` function:

```
raw_data <- getURL('https://raw.githubusercontent.com/williamzjasmine/CUNY_SPS_DS/master/DATA_607/Project1/ChessTournamentData.txt')
print(substr(raw_data, 1, 1000))
```

```
## [1] "-----\n Pair | Player Name | Total | Round | Round | Round | Round | Round | Round | Round | "
```

Printing out the first 1,000 characters of the texts reveals a bit of a mess, but it is clear that the file is broken up by lines given the presence of the `\n` characters. The code chunk below splits the `raw_data` line by line using the `strsplit()` function to create a character vector called `file_lines`:

```
file_lines <- strsplit(raw_data, split="\n", fixed=TRUE)[[1]]
print(file_lines[1:10])
```

```
## [1] "-----"
## [2] " Pair | Player Name | Total | Round | Round | Round | Round | Round | Round | Round | "
## [3] " Num | USCF ID / Rtg (Pre->Post) | Pts | 1 | 2 | 3 | 4 | 5 | 6 | 7 | "
## [4] "-----"
## [5] " 1 | GARY HUA | 6.0 | W 39 | W 21 | W 18 | W 14 | W 7 | D 12 | D 4 | "
## [6] " ON | 15445895 / R: 1794 ->1817 | N:2 | W | B | W | B | W | B | W | "
## [7] "-----"
## [8] " 2 | DAKSHESH DARURI | 6.0 | W 63 | W 58 | L 4 | W 17 | W 16 | W 20 | W 7 | "
## [9] " MI | 14598900 / R: 1553 ->1663 | N:2 | B | W | B | W | B | W | B | "
## [10] "-----"
```

Looking at the first 10 lines outputted above, it is clear that the first four lines are header lines and contain no useful information. They are removed in the chunk below:

```
file_lines = file_lines[-1:-4]
file_lines[1:8]
```

```
## [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] "-----"
## [7] "      3 | ADITYA BAJAJ          |6.0 |L 8|W 61|W 25|W 21|W 11|W 13|W 12|"
## [8] "    MI | 14959604 / R: 1384   ->1640 |N:2 |W   |B   |W   |B   |W   |B   |W   |"
```

Printing the first 8 lines of the new `file_lines` vector above confirms that these header lines have now been removed.

Parse File

Now that the text information is in an iterable format based on line, we can go through the `file_lines` vector line-by-line to determine the information needed. Looking above it appears as though the lines have a particular pattern to them, and exist in groups of three. In this group of three lines:

1. The first line contains the pair number, name, number of points won, the results of each of the player's matches, and the pair number of each opponent they played. For instance, in line [1] above the pair number is 1, the player's name is GARY HUA, he scored 6.0 points in the tournament, and he played seven games (winning against player 39, 21, 18, 14, 7, 12, and 4).
2. The second line contains the players state and pre tournament ranking. For instance in line [2], we see that GARY HUA is from ON (Ontario) and had a pre tournament ELO ranking of 1,794.
3. The third line is a line break made up of - characters to separate out the information for the next player.

Based off this pattern outlined above, we can pick out all the relevant information to create the desired `.csv` file. This is done below:

```
matches_df <- data.frame(matrix(ncol = 2, nrow = 0))
names = c()
pair_nums = c()
states = c()
points = c()
ratings = c()

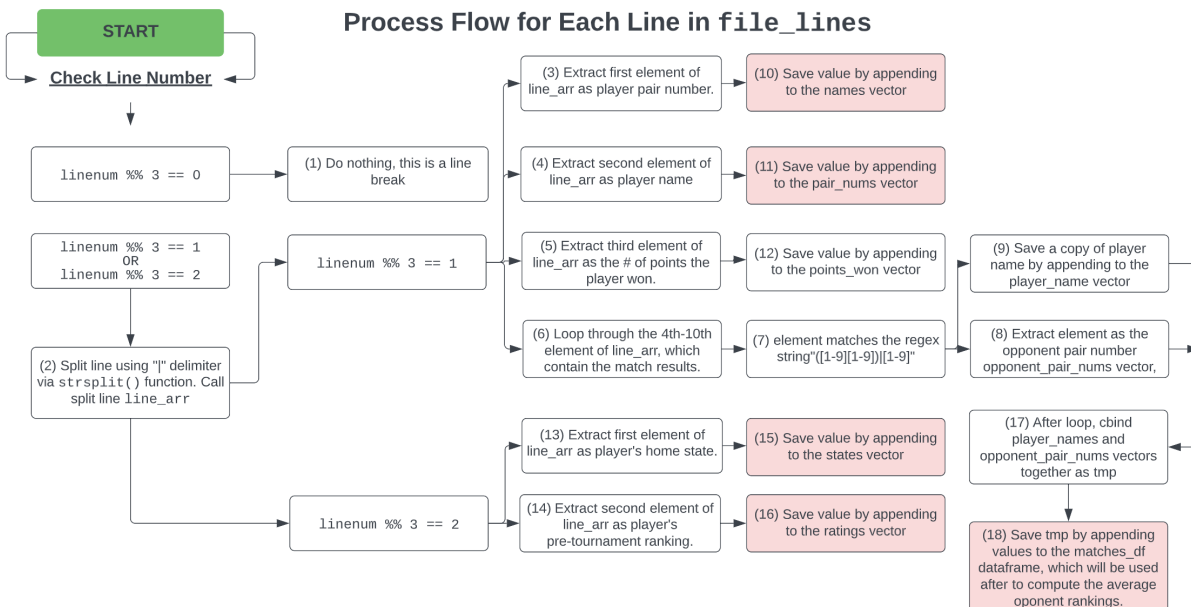
for (line_num in 1:length(file_lines)){
  if (line_num %% 3 == 0){
    next # (1)
  }
  else {
    line_arr <- strsplit(file_lines[line_num], "|", fixed = TRUE) # (2)
    #print(line_arr)
    if (line_num %% 3 == 1) {
      pair_num <- as.integer(
        gsub(line_arr[[1]][1], pattern = " ", replacement = "")) # (3)
      name <- str_extract(line_arr[[1]][2], '[A-Z].*[A-Z]') # (4)
      points_won = as.numeric(
        gsub(line_arr[[1]][3], pattern = " ", replacement = "")) # (5)
      opponent_pair_nums = c()
      for (col_num in 4:10){ # (6)
        opponent_pair_num = str_extract(line_arr[[1]][col_num],
                                          '([1-9][0-9])|([0-9])' ) # (7)
        if (is.na(opponent_pair_num) == FALSE) { # (7)
          opponent_pair_nums = append(as.integer(opponent_pair_num),
                                       opponent_pair_nums) # (8)
        }
      }
    }
  }
}
```

```

    }
    player_name = rep(name, length(opponent_pair_nums)) # (9)
    tmp = cbind.data.frame(player_name, opponent_pair_nums) # (17)
    matches_df = rbind(matches_df, tmp) # (18)
    pair_nums = append(pair_nums, pair_num) # (10)
    names = append(names, name) # (11)
    points = append(points, points_won) # (12)
  }
  else if (line_num %% 3 == 2) {
    state <- gsub(line_arr[[1]][1], pattern = " ", replacement = "") # (13)
    pre_rating <- as.integer(
      gsub(gsub(str_extract(line_arr[[1]][2], ' \\d{3,4}(( )|P)'),
        pattern=" ", replacement=''),
        pattern='P', replacement='')) # (14)
    states = append(states, state)
    ratings = append(ratings, pre_rating)
  }
  else {
    print('Something went wrong')
  }
}
}

```

The code above contains a lot of different `if` statements and `for` loops, making it a little difficult to understand the logical flow on the surface. As such, the following graphic provides a flowchart of what it performs:



The flowchart describes the logical flow of the code, but doesn't specify the exact regex matching or string manipulation that was used to extract the necessary values. In order to view this detail, the numbers in the flowchart cells correspond to the commented numbers in the above code specifying exactly where each action took place.

The red cells indicate the times in which desired values are saved to various vectors to be used in the following code. Five of these vectors (`names`, `pair_nums`, `states`, `points`, and `ratings`) are combined below into what will be our final dataframe (`final_df`) using the `cbind` function. Appropriate column names are also given

to this dataframe.

```
final_df = cbind.data.frame(names, pair_nums, states, points, ratings)
colnames(final_df) <- c('player_name', 'pair_number',
                        'state', 'points_won', 'pre_rating')
colnames(matches_df) <- c('player_name', 'pair_number')

head(final_df)
```

```
##           player_name pair_number state points_won pre_rating
## 1           GARY HUA           1    ON           6.0       1794
## 2    DAKSHESH DARURI           2    MI           6.0       1553
## 3      ADITYA BAJAJ           3    MI           6.0       1384
## 4 PATRICK H SCHILLING           4    MI           5.5       1716
## 5           HANSHI ZUO           5    MI           5.5       1655
## 6           HANSEN SONG           6    OH           5.0       1686
```

The output above shows that our data is now very close to being in the desired format: the only column now missing is the average rating of each player's opponent.

Calc Average Ratings

To calculate the average rating of each opponent, we can use the `matches_df` dataframe created in the cell chunk containing the primary parsing code. This dataframe contains a record of all the games that were played, namely a list of each player name and all of their associated opponent pair numbers. This can also be clarified in the graphic above by following the logical flow after cell (6). The chunk below calculates the average rating of each opponent using a `groupby()`:

```
avg_opponent_score_df <-
  matches_df %>%
    left_join(final_df, by = "pair_number", suffix=c("", "_y")) %>%
    group_by(player_name) %>%
    summarise(avg_opponent_score=mean(pre_rating))
```

The results are stored in a dataframe called `avg_opponent_score_df`.

Putting it All Together

The final step to having all the required data is to join our `final_df` to the newly created `avg_opponent_score_df` and pulling the average opponent score. This is done below:

```
final_df <-
  final_df %>%
    left_join(avg_opponent_score_df, by='player_name')

final_df <- final_df[-2]
final_df
```

```
##           player_name state points_won pre_rating avg_opponent_score
## 1           GARY HUA    ON           6.0       1794          1605.286
## 2    DAKSHESH DARURI    MI           6.0       1553          1469.286
## 3      ADITYA BAJAJ    MI           6.0       1384          1563.571
## 4 PATRICK H SCHILLING    MI           5.5       1716          1573.571
## 5           HANSHI ZUO    MI           5.5       1655          1500.857
## 6           HANSEN SONG    OH           5.0       1686          1518.714
```

## 7	GARY DEE SWATHELL	MI	5.0	1649	1372.143
## 8	EZEKIEL HOUGHTON	MI	5.0	1641	1468.429
## 9	STEFANO LEE	ON	5.0	1411	1523.143
## 10	ANVIT RAO	MI	5.0	1365	1554.143
## 11	CAMERON WILLIAM MC LEMAN	MI	4.5	1712	1467.571
## 12	KENNETH J TACK	MI	4.5	1663	1506.167
## 13	TORRANCE HENRY JR	MI	4.5	1666	1497.857
## 14	BRADLEY SHAW	MI	4.5	1610	1515.000
## 15	ZACHARY JAMES HOUGHTON	MI	4.5	1220	1483.857
## 16	MIKE NIKITIN	MI	4.0	1604	1385.800
## 17	RONALD GRZEGORCZYK	MI	4.0	1629	1498.571
## 18	DAVID SUNDEEN	MI	4.0	1600	1480.000
## 19	DIPANKAR ROY	MI	4.0	1564	1426.286
## 20	JASON ZHENG	MI	4.0	1595	1410.857
## 21	DINH DANG BUI	ON	4.0	1563	1470.429
## 22	EUGENE L MCCLURE	MI	4.0	1555	1300.333
## 23	ALAN BUI	ON	4.0	1363	1213.857
## 24	MICHAEL R ALDRICH	MI	4.0	1229	1357.000
## 25	LOREN SCHWIEBERT	MI	3.5	1745	1363.286
## 26	MAX ZHU	ON	3.5	1579	1506.857
## 27	GAURAV GIDWANI	MI	3.5	1552	1221.667
## 28	SOFIA ADINA STANESCU-BELLU	MI	3.5	1507	1522.143
## 29	CHIEDOZIE OKORIE	MI	3.5	1602	1313.500
## 30	GEORGE AVERY JONES	ON	3.5	1522	1144.143
## 31	RISHI SHETTY	MI	3.5	1494	1259.857
## 32	JOSHUA PHILIP MATHEWS	ON	3.5	1441	1378.714
## 33	JADE GE	MI	3.5	1449	1276.857
## 34	MICHAEL JEFFERY THOMAS	MI	3.5	1399	1375.286
## 35	JOSHUA DAVID LEE	MI	3.5	1438	1149.714
## 36	SIDDHARTH JHA	MI	3.5	1355	1388.167
## 37	AMIYATOSH PWNANANDAM	MI	3.5	980	1384.800
## 38	BRIAN LIU	MI	3.0	1423	1539.167
## 39	JOEL R HENDON	MI	3.0	1436	1429.571
## 40	FOREST ZHANG	MI	3.0	1348	1390.571
## 41	KYLE WILLIAM MURPHY	MI	3.0	1403	1248.500
## 42	JARED GE	MI	3.0	1332	1149.857
## 43	ROBERT GLEN VASEY	MI	3.0	1283	1106.571
## 44	JUSTIN D SCHILLING	MI	3.0	1199	1327.000
## 45	DEREK YAN	MI	3.0	1242	1152.000
## 46	JACOB ALEXANDER LAVALLEY	MI	3.0	377	1357.714
## 47	ERIC WRIGHT	MI	2.5	1362	1392.000
## 48	DANIEL KHAIN	MI	2.5	1382	1355.800
## 49	MICHAEL J MARTIN	MI	2.5	1291	1285.800
## 50	SHIVAM JHA	MI	2.5	1056	1296.000
## 51	TEJAS AYYAGARI	MI	2.5	1011	1356.143
## 52	ETHAN GUO	MI	2.5	935	1494.571
## 53	JOSE C YBARRA	MI	2.0	1393	1345.333
## 54	LARRY HODGE	MI	2.0	1270	1206.167
## 55	ALEX KONG	MI	2.0	1186	1406.000
## 56	MARISA RICCI	MI	2.0	1153	1414.400
## 57	MICHAEL LU	MI	2.0	1092	1363.000
## 58	VIRAJ MOHILE	MI	2.0	917	1391.000
## 59	SEAN M MC CORMICK	MI	2.0	853	1319.000
## 60	JULIA SHEN	MI	1.5	967	1330.200

## 61	JEZZEL FARKAS	ON	1.5	955	1327.286
## 62	ASHWIN BALAJI	MI	1.0	1530	1186.000
## 63	THOMAS JOSEPH HOSMER	MI	1.0	1175	1350.200
## 64	BEN LI	MI	1.0	1163	1263.000

And that's it! The text file has now been transformed into an R dataframe containing the desired columns. This R dataframe is converted into a `.csv` in the cell below, and can be used for further analysis.

```
write.csv(final_df, "tournamentinfo.csv", row.names = FALSE)
```