

Project_One

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```
options(warn=-1)
library(XML)
library(httr)
library(curl)
```

```
##
## Attaching package: 'curl'
```

```
## The following object is masked from 'package:httr':
##
##   handle_reset
```

```
library(RCurl)
```

```
## Loading required package: bitops
```

```
library(stringr)
library(tm)
```

```
## Loading required package: NLP
```

```
##
## Attaching package: 'NLP'
```

```
## The following object is masked from 'package:httr':
##
##   content
```

```
library(sm)
```

```
## Package 'sm', version 2.2-5.5: type help(sm) for summary information
```

DATA607 Fall 2018 - PROJECT ONE

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1. Read in the text file, collapse it as one large block of text, unlist and parse it into a vector using `str_split`, and go through a series of replace statements to search for strings and replace them.

```

txt<- character()
txt <- c(txt, readLines('tournamentinfo.txt'))
txt <- str_c(txt, collapse = "\n")
txt_parsed <- unlist(str_split(txt, "-----
-----"))
txt_parsed<-str_replace(txt_parsed,pattern="^\n","")
txt_parsed<-str_replace(txt_parsed,pattern="\n$", "")
txt_parsed<-str_replace(txt_parsed," \n Num ", "Num")
txt_parsed<-str_replace(txt_parsed," / R:", "|")
txt_parsed<-str_replace(txt_parsed,"USCF ID / Rtg ", "USCF ID|Rtg")
txt_parsed<-str_replace(txt_parsed,"->","|")
txt_parsed<-str_replace(txt_parsed,"Rtg\\(Pre|Post\\)", "Rating Prior|Rating Post")
txt_parsed<-str_replace(txt_parsed,"Post\\)", "")
txt_parsed<-str_replace(txt_parsed,"\n","")
txt_parsed<-txt_parsed[txt_parsed != ""]

```

2. The first line below shows the second element in the txt_parsed vector. The next statement locates the coordinates for where all of the “|” delimiters are located.

```
txt_parsed[2]
```

```
## [1] "      1 | GARY HUA          |6.0 |W 39|W 21|W 18|W 14|W 7|D 12|D 4
|   ON | 15445895| 1794   |1817   |N:2 |W   |B   |W   |B   |W   |B   |W   |"
```

```
str_locate_all(txt_parsed[2], "\\|")
```

```

## [[1]]
##      start end
## [1,]      7  7
## [2,]     41 41
## [3,]     47 47
## [4,]     53 53
## [5,]     59 59
## [6,]     65 65
## [7,]     71 71
## [8,]     77 77
## [9,]     83 83
## [10,]    89 89
## [11,]    96 96
## [12,]   106 106
## [13,]   115 115
## [14,]   125 125
## [15,]   131 131
## [16,]   137 137
## [17,]   143 143
## [18,]   149 149
## [19,]   155 155
## [20,]   161 161
## [21,]   167 167
## [22,]   173 173

```

3. With the coordinates of the “|”, the next step is to extract the data into individual vectors, `player_id`, `player_name`, `player_state`, `total_no_of_points`, and `player_pre_rating`. The vector, `player_opp`, contains the `player_ids` for a player's opponent. The vector `black_white` contains each instance of whether the player's chess pieces were Black or White.

```
player_id <- trimws(str_sub(txt_parsed[2:65],0,6))
player_name<- trimws(str_sub(txt_parsed[2:65],8,40))
player_state<- trimws(str_sub(txt_parsed[2:65],90,95))
total_no_of_points<-trimws(str_sub(txt_parsed[2:65],42,46))
player_pre_rating<-str_replace(trimws(str_sub(txt_parsed[2:65],107,114)),"P.*$","")
player_opp<-trimws(str_sub(txt_parsed[2:65],48,88))
black_white <- trimws(str_sub(txt_parsed[2:65],132,168))
```

```
player_id<-as.numeric(player_id)
player_id
```

```
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [24] 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46
## [47] 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64
```

```
player_name
```

```
## [1] "GARY HUA" "DAKSHESH DARURI"
## [3] "ADITYA BAJAJ" "PATRICK H SCHILLING"
## [5] "HANSHI ZUO" "HANSEN SONG"
## [7] "GARY DEE SWATHELL" "EZEKIEL HOUGHTON"
## [9] "STEFANO LEE" "ANVIT RAO"
## [11] "CAMERON WILLIAM MC LEMAN" "KENNETH J TACK"
## [13] "TORRANCE HENRY JR" "BRADLEY SHAW"
## [15] "ZACHARY JAMES HOUGHTON" "MIKE NIKITIN"
## [17] "RONALD GRZEGORCZYK" "DAVID SUNDEEN"
## [19] "DIPANKAR ROY" "JASON ZHENG"
## [21] "DINH DANG BUI" "EUGENE L MCCLURE"
## [23] "ALAN BUI" "MICHAEL R ALDRICH"
## [25] "LOREN SCHWIEBERT" "MAX ZHU"
## [27] "GAURAV GIDWANI" "SOFIA ADINA STANESCU-BELLU"
## [29] "CHIEDOZIE OKORIE" "GEORGE AVERY JONES"
## [31] "RISHI SHETTY" "JOSHUA PHILIP MATHEWS"
## [33] "JADE GE" "MICHAEL JEFFERY THOMAS"
## [35] "JOSHUA DAVID LEE" "SIDDHARTH JHA"
## [37] "AMIYATOSH PWNANANDAM" "BRIAN LIU"
## [39] "JOEL R HENDON" "FOREST ZHANG"
## [41] "KYLE WILLIAM MURPHY" "JARED GE"
## [43] "ROBERT GLEN VASEY" "JUSTIN D SCHILLING"
## [45] "DEREK YAN" "JACOB ALEXANDER LAVALLEY"
## [47] "ERIC WRIGHT" "DANIEL KHAIN"
## [49] "MICHAEL J MARTIN" "SHIVAM JHA"
## [51] "TEJAS AYYAGARI" "ETHAN GUO"
## [53] "JOSE C YBARRA" "LARRY HODGE"
## [55] "ALEX KONG" "MARISA RICCI"
## [57] "MICHAEL LU" "VIRAJ MOHILE"
## [59] "SEAN M MC CORMICK" "JULIA SHEN"
## [61] "JEZZEL FARKAS" "ASHWIN BALAJI"
## [63] "THOMAS JOSEPH HOSMER" "BEN LI"
```

```
player_state
```

```
## [1] "ON" "MI" "MI" "MI" "MI" "OH" "MI" "MI" "ON" "MI" "MI" "MI" "MI" "MI"
## [15] "MI" "MI" "MI" "MI" "MI" "MI" "ON" "MI" "ON" "MI" "MI" "ON" "MI" "MI"
## [29] "MI" "ON" "MI" "ON" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI"
## [43] "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI" "MI"
## [57] "MI" "MI" "MI" "MI" "ON" "MI" "MI" "MI"
```

```
total_no_of_points<-as.numeric(total_no_of_points)
total_no_of_points
```

```
## [1] 6.0 6.0 6.0 5.5 5.5 5.0 5.0 5.0 5.0 5.0 4.5 4.5 4.5 4.5 4.5 4.0 4.0
## [18] 4.0 4.0 4.0 4.0 4.0 4.0 4.0 3.5 3.5 3.5 3.5 3.5 3.5 3.5 3.5 3.5 3.5
## [35] 3.5 3.5 3.5 3.0 3.0 3.0 3.0 3.0 3.0 3.0 3.0 2.5 2.5 2.5 2.5 2.5
## [52] 2.5 2.0 2.0 2.0 2.0 2.0 2.0 2.0 1.5 1.5 1.0 1.0 1.0
```

```
player_pre_rating<-as.numeric(player_pre_rating)
player_pre_rating
```

```
## [1] 1794 1553 1384 1716 1655 1686 1649 1641 1411 1365 1712 1663 1666 1610
## [15] 1220 1604 1629 1600 1564 1595 1563 1555 1363 1229 1745 1579 1552 1507
## [29] 1602 1522 1494 1441 1449 1399 1438 1355 980 1423 1436 1348 1403 1332
## [43] 1283 1199 1242 377 1362 1382 1291 1056 1011 935 1393 1270 1186 1153
## [57] 1092 917 853 967 955 1530 1175 1163
```

4. The vectors `player_id`, `player_name`, `player_state`, `total_no_of_points`, and `player_pre_rating` were combined into a data frame.

```
df = data.frame(player_id,player_name,player_state,total_no_of_points,player_pre_rating)
df
```

##	player_id	player_name	player_state	total_no_of_points
## 1	1	GARY HUA	ON	6.0
## 2	2	DAKSHESH DARURI	MI	6.0
## 3	3	ADITYA BAJAJ	MI	6.0
## 4	4	PATRICK H SCHILLING	MI	5.5
## 5	5	HANSHI ZUO	MI	5.5
## 6	6	HANSEN SONG	OH	5.0
## 7	7	GARY DEE SWATHELL	MI	5.0
## 8	8	EZEKIEL HOUGHTON	MI	5.0
## 9	9	STEFANO LEE	ON	5.0
## 10	10	ANVIT RAO	MI	5.0
## 11	11	CAMERON WILLIAM MC LEMAN	MI	4.5
## 12	12	KENNETH J TACK	MI	4.5
## 13	13	TORRANCE HENRY JR	MI	4.5
## 14	14	BRADLEY SHAW	MI	4.5
## 15	15	ZACHARY JAMES HOUGHTON	MI	4.5
## 16	16	MIKE NIKITIN	MI	4.0
## 17	17	RONALD GRZEGORCZYK	MI	4.0
## 18	18	DAVID SUNDEEN	MI	4.0
## 19	19	DIPANKAR ROY	MI	4.0
## 20	20	JASON ZHENG	MI	4.0
## 21	21	DINH DANG BUI	ON	4.0
## 22	22	EUGENE L MCCLURE	MI	4.0
## 23	23	ALAN BUI	ON	4.0
## 24	24	MICHAEL R ALDRICH	MI	4.0
## 25	25	LOREN SCHWIEBERT	MI	3.5
## 26	26	MAX ZHU	ON	3.5
## 27	27	GAURAV GIDWANI	MI	3.5
## 28	28	SOFIA ADINA STANESCU-BELLU	MI	3.5
## 29	29	CHIEDOZIE OKORIE	MI	3.5
## 30	30	GEORGE AVERY JONES	ON	3.5
## 31	31	RISHI SHETTY	MI	3.5
## 32	32	JOSHUA PHILIP MATHEWS	ON	3.5
## 33	33	JADE GE	MI	3.5
## 34	34	MICHAEL JEFFERY THOMAS	MI	3.5
## 35	35	JOSHUA DAVID LEE	MI	3.5
## 36	36	SIDDHARTH JHA	MI	3.5
## 37	37	AMIYATOSH PWNANANDAM	MI	3.5
## 38	38	BRIAN LIU	MI	3.0
## 39	39	JOEL R HENDON	MI	3.0
## 40	40	FOREST ZHANG	MI	3.0
## 41	41	KYLE WILLIAM MURPHY	MI	3.0
## 42	42	JARED GE	MI	3.0
## 43	43	ROBERT GLEN VASEY	MI	3.0
## 44	44	JUSTIN D SCHILLING	MI	3.0
## 45	45	DEREK YAN	MI	3.0
## 46	46	JACOB ALEXANDER LAVALLEY	MI	3.0
## 47	47	ERIC WRIGHT	MI	2.5
## 48	48	DANIEL KHAIN	MI	2.5
## 49	49	MICHAEL J MARTIN	MI	2.5
## 50	50	SHIVAM JHA	MI	2.5
## 51	51	TEJAS AYYAGARI	MI	2.5
## 52	52	ETHAN GUO	MI	2.5

## 53	53	JOSE C YBARRA	MI	2.0
## 54	54	LARRY HODGE	MI	2.0
## 55	55	ALEX KONG	MI	2.0
## 56	56	MARISA RICCI	MI	2.0
## 57	57	MICHAEL LU	MI	2.0
## 58	58	VIRAJ MOHILE	MI	2.0
## 59	59	SEAN M MC CORMICK	MI	2.0
## 60	60	JULIA SHEN	MI	1.5
## 61	61	JEZZEL FARKAS	ON	1.5
## 62	62	ASHWIN BALAJI	MI	1.0
## 63	63	THOMAS JOSEPH HOSMER	MI	1.0
## 64	64	BEN LI	MI	1.0

player_pre_rating

## 1	1794
## 2	1553
## 3	1384
## 4	1716
## 5	1655
## 6	1686
## 7	1649
## 8	1641
## 9	1411
## 10	1365
## 11	1712
## 12	1663
## 13	1666
## 14	1610
## 15	1220
## 16	1604
## 17	1629
## 18	1600
## 19	1564
## 20	1595
## 21	1563
## 22	1555
## 23	1363
## 24	1229
## 25	1745
## 26	1579
## 27	1552
## 28	1507
## 29	1602
## 30	1522
## 31	1494
## 32	1441
## 33	1449
## 34	1399
## 35	1438
## 36	1355
## 37	980
## 38	1423
## 39	1436
## 40	1348
## 41	1403

```
## 42      1332
## 43      1283
## 44      1199
## 45      1242
## 46       377
## 47      1362
## 48      1382
## 49      1291
## 50      1056
## 51      1011
## 52       935
## 53      1393
## 54      1270
## 55      1186
## 56      1153
## 57      1092
## 58       917
## 59       853
## 60       967
## 61       955
## 62      1530
## 63      1175
## 64      1163
```

5. In the next block, I created a test to calculate the Average Pre Chess Rating of Opponents. This is done by extracting the player_ids of opponents by searching for only 1 to 2 digits in the string. Next these were added to a numeric vector. With the player_ids, I created a temporary data frame for a player's opponent by subsetting records for only the opponents. Finally, by taking the mean of these players, I have each the Average Pre Chess Rating of Opponents.

```
test<- unlist(str_extract_all(player_opp[1], "[:digit:]{1,2}"))
test <- as.numeric(test)
temp<- df[df$player_id %in% test,]
mean(temp$player_pre_rating)
```

```
## [1] 1605.286
```

5a. With a successful test, I created a for loop which calculates the Average Pre Chess Rating of Opponents for all of the players, and then adding that vector to the data frame.


```
opponent_avg <- vector(mode="numeric")

for (i in seq(1,length(player_opp)))
{

  test<- unlist(str_extract_all(player_opp[i],":digit:"){1,2}"))
  test <- as.numeric(test)
  temp<- df[df$player_id %in% test,]
  opponent_avg<-append(opponent_avg, mean(temp$player_pre_rating))
}

opponent_avg
```

```
## [1] 1605.286 1469.286 1563.571 1573.571 1500.857 1518.714 1372.143
## [8] 1468.429 1523.143 1554.143 1467.571 1506.167 1497.857 1515.000
## [15] 1483.857 1385.800 1498.571 1480.000 1426.286 1410.857 1470.429
## [22] 1300.333 1213.857 1357.000 1363.286 1506.857 1221.667 1522.143
## [29] 1313.500 1144.143 1259.857 1378.714 1276.857 1375.286 1149.714
## [36] 1388.167 1384.800 1539.167 1429.571 1390.571 1248.500 1149.857
## [43] 1106.571 1327.000 1152.000 1357.714 1392.000 1355.800 1285.800
## [50] 1296.000 1356.143 1494.571 1345.333 1206.167 1406.000 1414.400
## [57] 1363.000 1391.000 1319.000 1330.200 1327.286 1186.000 1350.200
## [64] 1263.000
```

```
df$avg_pre_chess_rating_of_opponents<-c(opponent_avg)
```

```
df
```

##	player_id	player_name	player_state	total_no_of_points
## 1	1	GARY HUA	ON	6.0
## 2	2	DAKSHESH DARURI	MI	6.0
## 3	3	ADITYA BAJAJ	MI	6.0
## 4	4	PATRICK H SCHILLING	MI	5.5
## 5	5	HANSHI ZUO	MI	5.5
## 6	6	HANSEN SONG	OH	5.0
## 7	7	GARY DEE SWATHELL	MI	5.0
## 8	8	EZEKIEL HOUGHTON	MI	5.0
## 9	9	STEFANO LEE	ON	5.0
## 10	10	ANVIT RAO	MI	5.0
## 11	11	CAMERON WILLIAM MC LEMAN	MI	4.5
## 12	12	KENNETH J TACK	MI	4.5
## 13	13	TORRANCE HENRY JR	MI	4.5
## 14	14	BRADLEY SHAW	MI	4.5
## 15	15	ZACHARY JAMES HOUGHTON	MI	4.5
## 16	16	MIKE NIKITIN	MI	4.0
## 17	17	RONALD GRZEGORCZYK	MI	4.0
## 18	18	DAVID SUNDEEN	MI	4.0
## 19	19	DIPANKAR ROY	MI	4.0
## 20	20	JASON ZHENG	MI	4.0
## 21	21	DINH DANG BUI	ON	4.0
## 22	22	EUGENE L MCCLURE	MI	4.0
## 23	23	ALAN BUI	ON	4.0
## 24	24	MICHAEL R ALDRICH	MI	4.0
## 25	25	LOREN SCHWIEBERT	MI	3.5
## 26	26	MAX ZHU	ON	3.5
## 27	27	GAURAV GIDWANI	MI	3.5
## 28	28	SOFIA ADINA STANESCU-BELLU	MI	3.5
## 29	29	CHIEDOZIE OKORIE	MI	3.5
## 30	30	GEORGE AVERY JONES	ON	3.5
## 31	31	RISHI SHETTY	MI	3.5
## 32	32	JOSHUA PHILIP MATHEWS	ON	3.5
## 33	33	JADE GE	MI	3.5
## 34	34	MICHAEL JEFFERY THOMAS	MI	3.5
## 35	35	JOSHUA DAVID LEE	MI	3.5
## 36	36	SIDDHARTH JHA	MI	3.5
## 37	37	AMIYATOSH PWNANANDAM	MI	3.5
## 38	38	BRIAN LIU	MI	3.0
## 39	39	JOEL R HENDON	MI	3.0
## 40	40	FOREST ZHANG	MI	3.0
## 41	41	KYLE WILLIAM MURPHY	MI	3.0
## 42	42	JARED GE	MI	3.0
## 43	43	ROBERT GLEN VASEY	MI	3.0
## 44	44	JUSTIN D SCHILLING	MI	3.0
## 45	45	DEREK YAN	MI	3.0
## 46	46	JACOB ALEXANDER LAVALLEY	MI	3.0
## 47	47	ERIC WRIGHT	MI	2.5
## 48	48	DANIEL KHAIN	MI	2.5
## 49	49	MICHAEL J MARTIN	MI	2.5
## 50	50	SHIVAM JHA	MI	2.5
## 51	51	TEJAS AYYAGARI	MI	2.5
## 52	52	ETHAN GUO	MI	2.5

## 53	53	JOSE C YBARRA	MI	2.0
## 54	54	LARRY HODGE	MI	2.0
## 55	55	ALEX KONG	MI	2.0
## 56	56	MARISA RICCI	MI	2.0
## 57	57	MICHAEL LU	MI	2.0
## 58	58	VIRAJ MOHILE	MI	2.0
## 59	59	SEAN M MC CORMICK	MI	2.0
## 60	60	JULIA SHEN	MI	1.5
## 61	61	JEZZEL FARKAS	ON	1.5
## 62	62	ASHWIN BALAJI	MI	1.0
## 63	63	THOMAS JOSEPH HOSMER	MI	1.0
## 64	64	BEN LI	MI	1.0

##	player_pre_rating	avg_pre_chess_rating_of_opponents
----	-------------------	-----------------------------------

## 1	1794	1605.286
## 2	1553	1469.286
## 3	1384	1563.571
## 4	1716	1573.571
## 5	1655	1500.857
## 6	1686	1518.714
## 7	1649	1372.143
## 8	1641	1468.429
## 9	1411	1523.143
## 10	1365	1554.143
## 11	1712	1467.571
## 12	1663	1506.167
## 13	1666	1497.857
## 14	1610	1515.000
## 15	1220	1483.857
## 16	1604	1385.800
## 17	1629	1498.571
## 18	1600	1480.000
## 19	1564	1426.286
## 20	1595	1410.857
## 21	1563	1470.429
## 22	1555	1300.333
## 23	1363	1213.857
## 24	1229	1357.000
## 25	1745	1363.286
## 26	1579	1506.857
## 27	1552	1221.667
## 28	1507	1522.143
## 29	1602	1313.500
## 30	1522	1144.143
## 31	1494	1259.857
## 32	1441	1378.714
## 33	1449	1276.857
## 34	1399	1375.286
## 35	1438	1149.714
## 36	1355	1388.167
## 37	980	1384.800
## 38	1423	1539.167
## 39	1436	1429.571
## 40	1348	1390.571
## 41	1403	1248.500

## 42	1332	1149.857
## 43	1283	1106.571
## 44	1199	1327.000
## 45	1242	1152.000
## 46	377	1357.714
## 47	1362	1392.000
## 48	1382	1355.800
## 49	1291	1285.800
## 50	1056	1296.000
## 51	1011	1356.143
## 52	935	1494.571
## 53	1393	1345.333
## 54	1270	1206.167
## 55	1186	1406.000
## 56	1153	1414.400
## 57	1092	1363.000
## 58	917	1391.000
## 59	853	1319.000
## 60	967	1330.200
## 61	955	1327.286
## 62	1530	1186.000
## 63	1175	1350.200
## 64	1163	1263.000

6. I went a step further. I wanted to know how many Wins, Losses, Draws, U's, and H's that a player has. Additionally, I wanted to count the number of times a player has played with Black or White chess pieces. Using `extract all` on `player_opp` and `black_white` to search for "W", "L", "D", "U", and "H" and putting these values are put into vectors, I was able to count the length of each vector to calculate the number of Wins, Losses, Draws, Us, and Hs. I took the same approach to count the number of times a player played with either Black or White pieces.

```
countofWins<- vector(mode="numeric")
countofLosses<-vector(mode="numeric")
countofDraws <-vector(mode="numeric")
countofU <-vector(mode="numeric")
countofH <-vector(mode="numeric")
countofWhite <-vector(mode="numeric")
countofBlack <- vector(mode="numeric")

for (i in seq(1,length(player_opp)))
{

  wins<- unlist(str_extract_all(player_opp[i],"[W]"))
  countofWins<-append(countofWins,length(wins))

  losses <- unlist(str_extract_all(player_opp[i],"[L]"))
  countofLosses<-append(countofLosses,length(losses))

  draws <- unlist(str_extract_all(player_opp[i],"[D]"))
  countofDraws<-append(countofDraws,length(draws))

  U <- unlist(str_extract_all(player_opp[i],"[U]"))
  countofU<-append(countofU,length(U))

  H<-unlist(str_extract_all(player_opp[i],"[H]"))
  countofH<-append(countofH,length(H))

  WWhite<-unlist(str_extract_all(black_white[i],"[W]"))
  countofWhite<-append(countofWhite,length(WWhite))

  Black<-unlist(str_extract_all(black_white[i],"[B]"))
  countofBlack<-append(countofBlack,length(Black))

}
```

7. These values were added to the data frame.

```
df$Wins <- countofWins
df$Losses <- countofLosses
df$Draws <- countofDraws
df$U <- countofU
df$H <- countofH
df$White<- countofWhite
df$Black<-countofBlack
df
```

##	player_id	player_name	player_state	total_no_of_points
## 1	1	GARY HUA	ON	6.0
## 2	2	DAKSHESH DARURI	MI	6.0
## 3	3	ADITYA BAJAJ	MI	6.0
## 4	4	PATRICK H SCHILLING	MI	5.5
## 5	5	HANSHI ZUO	MI	5.5
## 6	6	HANSEN SONG	OH	5.0
## 7	7	GARY DEE SWATHELL	MI	5.0
## 8	8	EZEKIEL HOUGHTON	MI	5.0
## 9	9	STEFANO LEE	ON	5.0
## 10	10	ANVIT RAO	MI	5.0
## 11	11	CAMERON WILLIAM MC LEMAN	MI	4.5
## 12	12	KENNETH J TACK	MI	4.5
## 13	13	TORRANCE HENRY JR	MI	4.5
## 14	14	BRADLEY SHAW	MI	4.5
## 15	15	ZACHARY JAMES HOUGHTON	MI	4.5
## 16	16	MIKE NIKITIN	MI	4.0
## 17	17	RONALD GRZEGORCZYK	MI	4.0
## 18	18	DAVID SUNDEEN	MI	4.0
## 19	19	DIPANKAR ROY	MI	4.0
## 20	20	JASON ZHENG	MI	4.0
## 21	21	DINH DANG BUI	ON	4.0
## 22	22	EUGENE L MCCLURE	MI	4.0
## 23	23	ALAN BUI	ON	4.0
## 24	24	MICHAEL R ALDRICH	MI	4.0
## 25	25	LOREN SCHWIEBERT	MI	3.5
## 26	26	MAX ZHU	ON	3.5
## 27	27	GAURAV GIDWANI	MI	3.5
## 28	28	SOFIA ADINA STANESCU-BELLU	MI	3.5
## 29	29	CHIEDOZIE OKORIE	MI	3.5
## 30	30	GEORGE AVERY JONES	ON	3.5
## 31	31	RISHI SHETTY	MI	3.5
## 32	32	JOSHUA PHILIP MATHEWS	ON	3.5
## 33	33	JADE GE	MI	3.5
## 34	34	MICHAEL JEFFERY THOMAS	MI	3.5
## 35	35	JOSHUA DAVID LEE	MI	3.5
## 36	36	SIDDHARTH JHA	MI	3.5
## 37	37	AMIYATOSH PWNANANDAM	MI	3.5
## 38	38	BRIAN LIU	MI	3.0
## 39	39	JOEL R HENDON	MI	3.0
## 40	40	FOREST ZHANG	MI	3.0
## 41	41	KYLE WILLIAM MURPHY	MI	3.0
## 42	42	JARED GE	MI	3.0
## 43	43	ROBERT GLEN VASEY	MI	3.0
## 44	44	JUSTIN D SCHILLING	MI	3.0
## 45	45	DEREK YAN	MI	3.0
## 46	46	JACOB ALEXANDER LAVALLEY	MI	3.0
## 47	47	ERIC WRIGHT	MI	2.5
## 48	48	DANIEL KHAIN	MI	2.5
## 49	49	MICHAEL J MARTIN	MI	2.5
## 50	50	SHIVAM JHA	MI	2.5
## 51	51	TEJAS AYYAGARI	MI	2.5
## 52	52	ETHAN GUO	MI	2.5

## 53	53	JOSE C YBARRA	MI	2.0		
## 54	54	LARRY HODGE	MI	2.0		
## 55	55	ALEX KONG	MI	2.0		
## 56	56	MARISA RICCI	MI	2.0		
## 57	57	MICHAEL LU	MI	2.0		
## 58	58	VIRAJ MOHILE	MI	2.0		
## 59	59	SEAN M MC CORMICK	MI	2.0		
## 60	60	JULIA SHEN	MI	1.5		
## 61	61	JEZZEL FARKAS	ON	1.5		
## 62	62	ASHWIN BALAJI	MI	1.0		
## 63	63	THOMAS JOSEPH HOSMER	MI	1.0		
## 64	64	BEN LI	MI	1.0		
##	player_pre_rating	avg_pre_chess_rating_of_opponents	Wins	Losses	Draws	U
## 1	1794	1605.286	5	0	2	0
## 2	1553	1469.286	6	1	0	0
## 3	1384	1563.571	6	1	0	0
## 4	1716	1573.571	4	0	3	0
## 5	1655	1500.857	4	0	3	0
## 6	1686	1518.714	4	1	2	0
## 7	1649	1372.143	5	2	0	0
## 8	1641	1468.429	5	2	0	0
## 9	1411	1523.143	5	2	0	0
## 10	1365	1554.143	4	1	2	0
## 11	1712	1467.571	4	2	1	0
## 12	1663	1506.167	3	1	2	0
## 13	1666	1497.857	4	2	1	0
## 14	1610	1515.000	4	2	1	0
## 15	1220	1483.857	4	2	1	0
## 16	1604	1385.800	3	1	1	1
## 17	1629	1498.571	4	3	0	0
## 18	1600	1480.000	4	3	0	0
## 19	1564	1426.286	3	2	2	0
## 20	1595	1410.857	4	3	0	0
## 21	1563	1470.429	4	3	0	0
## 22	1555	1300.333	3	2	1	0
## 23	1363	1213.857	4	3	0	0
## 24	1229	1357.000	4	3	0	0
## 25	1745	1363.286	3	3	1	0
## 26	1579	1506.857	3	3	1	0
## 27	1552	1221.667	3	2	1	1
## 28	1507	1522.143	2	2	3	0
## 29	1602	1313.500	3	2	1	1
## 30	1522	1144.143	3	3	1	0
## 31	1494	1259.857	3	3	1	0
## 32	1441	1378.714	3	3	1	0
## 33	1449	1276.857	3	3	1	0
## 34	1399	1375.286	3	3	1	0
## 35	1438	1149.714	3	3	1	0
## 36	1355	1388.167	2	2	2	0
## 37	980	1384.800	2	3	0	0
## 38	1423	1539.167	2	3	1	0
## 39	1436	1429.571	3	4	0	0
## 40	1348	1390.571	3	4	0	0
## 41	1403	1248.500	2	2	0	2

## 42	1332	1149.857	2	3	2 0
## 43	1283	1106.571	3	4	0 0
## 44	1199	1327.000	2	4	0 0
## 45	1242	1152.000	2	3	2 0
## 46	377	1357.714	3	4	0 0
## 47	1362	1392.000	2	4	1 0
## 48	1382	1355.800	1	3	1 0
## 49	1291	1285.800	1	2	2 1
## 50	1056	1296.000	2	4	0 0
## 51	1011	1356.143	2	4	1 0
## 52	935	1494.571	1	3	3 0
## 53	1393	1345.333	1	2	0 2
## 54	1270	1206.167	1	5	0 0
## 55	1186	1406.000	0	4	2 0
## 56	1153	1414.400	1	4	0 0
## 57	1092	1363.000	1	5	0 0
## 58	917	1391.000	1	5	0 0
## 59	853	1319.000	1	5	0 0
## 60	967	1330.200	0	3	2 1
## 61	955	1327.286	1	5	1 0
## 62	1530	1186.000	1	0	0 6
## 63	1175	1350.200	0	4	1 1
## 64	1163	1263.000	0	5	2 0

##	H	White	Black
----	---	-------	-------

## 1	0	4	3
## 2	0	3	4
## 3	0	4	3
## 4	0	3	4
## 5	0	3	4
## 6	0	3	4
## 7	0	4	3
## 8	0	4	3
## 9	0	3	4
## 10	0	4	3
## 11	0	3	4
## 12	1	3	3
## 13	0	3	4
## 14	0	4	3
## 15	0	3	4
## 16	1	2	3
## 17	0	4	3
## 18	0	3	4
## 19	0	4	3
## 20	0	4	3
## 21	0	4	3
## 22	1	3	3
## 23	0	3	4
## 24	0	3	4
## 25	0	3	4
## 26	0	4	3
## 27	0	3	3
## 28	0	4	3
## 29	0	3	3
## 30	0	3	4


```
## 31 0      3      4
## 32 0      4      3
## 33 0      3      4
## 34 0      3      4
## 35 0      4      3
## 36 1      3      3
## 37 1      3      2
## 38 1      3      3
## 39 0      4      3
## 40 0      4      3
## 41 0      2      2
## 42 0      3      4
## 43 0      4      3
## 44 0      3      3
## 45 0      4      3
## 46 0      4      3
## 47 0      4      3
## 48 2      2      3
## 49 1      3      2
## 50 1      3      3
## 51 0      4      3
## 52 0      3      4
## 53 2      2      1
## 54 0      3      3
## 55 0      3      3
## 56 2      3      2
## 57 0      3      3
## 58 0      3      3
## 59 0      3      3
## 60 1      2      3
## 61 0      3      4
## 62 0      0      1
## 63 1      2      3
## 64 0      3      4
```

```
summary(df)
```

```

##      player_id      player_name player_state total_no_of_points
## Min.   : 1.00    ADITYA BAJAJ      : 1    MI:55      Min.   :1.000
## 1st Qu.:16.75    ALAN BUI        : 1    OH: 1      1st Qu.:2.500
## Median :32.50    ALEX KONG        : 1    ON: 8      Median :3.500
## Mean   :32.50    AMIYATOSH PWNANANDAM: 1      Mean   :3.438
## 3rd Qu.:48.25    ANVIT RAO        : 1      3rd Qu.:4.000
## Max.   :64.00    ASHWIN BALAJI    : 1      Max.   :6.000
##      (Other)      :58
## player_pre_rating avg_pre_chess_rating_of_opponents      Wins
## Min.   : 377      Min.   :1107      Min.   :0.000
## 1st Qu.:1227      1st Qu.:1310      1st Qu.:2.000
## Median :1407      Median :1382      Median :3.000
## Mean   :1378      Mean   :1379      Mean   :2.734
## 3rd Qu.:1583      3rd Qu.:1481      3rd Qu.:4.000
## Max.   :1794      Max.   :1605      Max.   :6.000
##
##      Losses      Draws      U      H
## Min.   :0.000    Min.   :0.0000    Min.   :0.00    Min.   :0.00
## 1st Qu.:2.000    1st Qu.:0.0000    1st Qu.:0.00    1st Qu.:0.00
## Median :3.000    Median :1.0000    Median :0.00    Median :0.00
## Mean   :2.734    Mean   :0.9062    Mean   :0.25    Mean   :0.25
## 3rd Qu.:4.000    3rd Qu.:1.2500    3rd Qu.:0.00    3rd Qu.:0.00
## Max.   :5.000    Max.   :3.0000    Max.   :6.00    Max.   :2.00
##
##      White      Black
## Min.   :0.000    Min.   :1.000
## 1st Qu.:3.000    1st Qu.:3.000
## Median :3.000    Median :3.000
## Mean   :3.188    Mean   :3.188
## 3rd Qu.:4.000    3rd Qu.:4.000
## Max.   :4.000    Max.   :4.000
##

```

8. Finally, write the results to a csv file for loading into a database.

```
write.csv(df, file = "Chess Tournament Results.csv", row.names = FALSE)
```