

DATA 607 Project 1: Chess Tournament Results

Fan Xu, Sin Ying Wong

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1. Project Description

In this project, you're given a text file with chess tournament results where the information has some structure. Your job is to create an R Markdown file that generates a .CSV file (that could for example be imported into a SQL database) with the following information for all of the players:

Player's Name, Player's State, Total Number of Points, Player's Pre-Rating, and Average Pre Chess Rating of Opponents For the first player, the information would be:

Gary Hua, ON, 6.0, 1794, 1605

1605 was calculated by using the pre-tournament opponents' ratings of 1436, 1563, 1600, 1610, 1649, 1663, 1716, and dividing by the total number of games played.

If you have questions about the meaning of the data or the results, please post them on the discussion forum. Data science, like chess, is a game of back and forth...

The chess rating system (invented by a Minnesota statistician named Arpad Elo) has been used in many other contexts, including assessing relative strength of employment candidates by human resource departments.

You may substitute another text file (or set of text files, or data scraped from web pages) of similar or greater complexity, and create your own assignment and solution. You may work in a small team. All of your code should be in an R markdown file (and published to rpubs.com); with your data accessible for the person running the script.

2. Project Raw Data

The raw data is a text file with unstructured data strings.

```
raw_data <- readLines('https://raw.githubusercontent.com/oggyluky11/DATA607-Project-1/master/tournamentinfo.txt')
```

```
## Warning in readLines("https://raw.githubusercontent.com/oggyluky11/DATA607-Project-1/master/tournamentinfo.txt"): incomplete final line found on
## 'https://raw.githubusercontent.com/oggyluky11/DATA607-Project-1/master/tournamentinfo.txt'
```

```
head(raw_data)
```

```
## [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 | "
```

3. Raw Data Manipulation

a. Restructure raw data into dataframe

In raw data, double lines of strings are used to store one 'row' of data. Data columns are separated by "|". Let's see what this raw data should look like in a real data frame. This step is not requested in this project but it is good to represent the raw data into the way it should look like as data preparation.

```
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(tidyr)
library(stringr)
#remove the last character "/"
raw_data <- str_trim(raw_data[str_detect(raw_data, '-----')==FALSE], 'right')
raw_data <- substr(raw_data, 1, nchar(raw_data)-1)

#Combine two lines of data into one row
raw_data2 <- character()
for (i in (1:length(raw_data))) {
  if(i%2==0) {
    line_odd <- str_trim(unlist(str_split(raw_data[i-1], '\\|')))
    line_even <- str_trim(unlist(str_split(raw_data[i], '\\|')))
    line_cat <- str_c(line_odd, line_even, sep = ' ', collapse = ',')
    raw_data2 <- append(raw_data2, line_cat)
  }
}
raw_data2 <- data.frame(raw_data2)
raw_data_frame <- raw_data2 %>% separate(raw_data2, as.character(c(1:10)), sep = ',', extra = 'drop')

#Set first row as header and rename columns
names(raw_data_frame) <- unlist(raw_data_frame[1,])
raw_data_frame <- raw_data_frame[-1,]
raw_data_frame
```

```
##      Pair Num      Player Name USCF ID / Rtg (Pre->Post)
## 2      1 ON      GARY HUA 15445895 / R: 1794 ->1817
## 3      2 MI      DAKSHESH DARURI 14598900 / R: 1553 ->1663
## 4      3 MI      ADITYA BAJAJ 14959604 / R: 1384 ->1640
## 5      4 MI      PATRICK H SCHILLING 12616049 / R: 1716 ->1744
## 6      5 MI      HANSHI ZUO 14601533 / R: 1655 ->1690
## 7      6 OH      HANSEN SONG 15055204 / R: 1686 ->1687
```

## 8	7 MI	GARY DEE SWATHELL 11146376 / R: 1649	->1673
## 9	8 MI	EZEKIEL HOUGHTON 15142253 / R: 1641P17->1657P24	
## 10	9 ON	STEFANO LEE 14954524 / R: 1411	->1564
## 11	10 MI	ANVIT RAO 14150362 / R: 1365	->1544
## 12	11 MI	CAMERON WILLIAM MC LEMAN 12581589 / R: 1712	->1696
## 13	12 MI	KENNETH J TACK 12681257 / R: 1663	->1670
## 14	13 MI	TORRANCE HENRY JR 15082995 / R: 1666	->1662
## 15	14 MI	BRADLEY SHAW 10131499 / R: 1610	->1618
## 16	15 MI	ZACHARY JAMES HOUGHTON 15619130 / R: 1220P13->1416P20	
## 17	16 MI	MIKE NIKITIN 10295068 / R: 1604	->1613
## 18	17 MI	RONALD GRZEGORCZYK 10297702 / R: 1629	->1610
## 19	18 MI	DAVID SUNDEEN 11342094 / R: 1600	->1600
## 20	19 MI	DIPANKAR ROY 14862333 / R: 1564	->1570
## 21	20 MI	JASON ZHENG 14529060 / R: 1595	->1569
## 22	21 ON	DINH DANG BUI 15495066 / R: 1563P22->1562	
## 23	22 MI	EUGENE L MCCLURE 12405534 / R: 1555	->1529
## 24	23 ON	ALAN BUI 15030142 / R: 1363	->1371
## 25	24 MI	MICHAEL R ALDRICH 13469010 / R: 1229	->1300
## 26	25 MI	LOREN SCHWIEBERT 12486656 / R: 1745	->1681
## 27	26 ON	MAX ZHU 15131520 / R: 1579	->1564
## 28	27 MI	GAURAV GIDWANI 14476567 / R: 1552	->1539
## 29	28 MI	SOFIA ADINA STANESCU-BELLU 14882954 / R: 1507	->1513
## 30	29 MI	CHIEDOZIE OKORIE 15323285 / R: 1602P6 ->1508P12	
## 31	30 ON	GEORGE AVERY JONES 12577178 / R: 1522	->1444
## 32	31 MI	RISHI SHETTY 15131618 / R: 1494	->1444
## 33	32 ON	JOSHUA PHILIP MATHEWS 14073750 / R: 1441	->1433
## 34	33 MI	JADE GE 14691842 / R: 1449	->1421
## 35	34 MI	MICHAEL JEFFERY THOMAS 15051807 / R: 1399	->1400
## 36	35 MI	JOSHUA DAVID LEE 14601397 / R: 1438	->1392
## 37	36 MI	SIDDHARTH JHA 14773163 / R: 1355	->1367
## 38	37 MI	AMIYATOSH PWNANANDAM 15489571 / R: 980P12->1077P17	
## 39	38 MI	BRIAN LIU 15108523 / R: 1423	->1439
## 40	39 MI	JOEL R HENDON 12923035 / R: 1436P23->1413	
## 41	40 MI	FOREST ZHANG 14892710 / R: 1348	->1346
## 42	41 MI	KYLE WILLIAM MURPHY 15761443 / R: 1403P5 ->1341P9	
## 43	42 MI	JARED GE 14462326 / R: 1332	->1256
## 44	43 MI	ROBERT GLEN VASEY 14101068 / R: 1283	->1244
## 45	44 MI	JUSTIN D SCHILLING 15323504 / R: 1199	->1199
## 46	45 MI	DEREK YAN 15372807 / R: 1242	->1191
## 47	46 MI	JACOB ALEXANDER LAVALLEY 15490981 / R: 377P3 ->1076P10	
## 48	47 MI	ERIC WRIGHT 12533115 / R: 1362	->1341
## 49	48 MI	DANIEL KHAIN 14369165 / R: 1382	->1335
## 50	49 MI	MICHAEL J MARTIN 12531685 / R: 1291P12->1259P17	
## 51	50 MI	SHIVAM JHA 14773178 / R: 1056	->1111
## 52	51 MI	TEJAS AYYAGARI 15205474 / R: 1011	->1097
## 53	52 MI	ETHAN GUO 14918803 / R: 935	->1092
## 54	53 MI	JOSE C YBARRA 12578849 / R: 1393	->1359
## 55	54 MI	LARRY HODGE 12836773 / R: 1270	->1200
## 56	55 MI	ALEX KONG 15412571 / R: 1186	->1163
## 57	56 MI	MARISA RICCI 14679887 / R: 1153	->1140
## 58	57 MI	MICHAEL LU 15113330 / R: 1092	->1079
## 59	58 MI	VIRAJ MOHILE 14700365 / R: 917	-> 941
## 60	59 MI	SEAN M MC CORMICK 12841036 / R: 853	-> 878
## 61	60 MI	JULIA SHEN 14579262 / R: 967	-> 984

## 62	61 ON	JEZZEL FARKAS 15771592 / R: 955P11-> 979P18									
## 63	62 MI	ASHWIN BALAJI 15219542 / R: 1530 ->1535									
## 64	63 MI	THOMAS JOSEPH HOSMER 15057092 / R: 1175 ->1125									
## 65	64 MI	BEN LI 15006561 / R: 1163 ->1112									
##	Total Pts	Round 1	Round 2	Round 3	Round 4	Round 5	Round 6	Round 7			
## 2	6.0 N:2 W	39 W W	21 B W	18 W W	14 B W	7 W D	12 B D	4 W			
## 3	6.0 N:2 W	63 B W	58 W L	4 B W	17 W W	16 B W	20 W W	7 B			
## 4	6.0 N:2 L	8 W W	61 B W	25 W W	21 B W	11 W W	13 B W	12 W			
## 5	5.5 N:2 W	23 W D	28 B W	2 W W	26 B D	5 W W	19 B D	1 B			
## 6	5.5 N:2 W	45 B W	37 W D	12 B D	13 W D	4 B W	14 W W	17 B			
## 7	5.0 N:3 W	34 W D	29 B L	11 W W	35 B D	10 B W	27 W W	21 B			
## 8	5.0 N:3 W	57 W W	46 B W	13 W W	11 B L	1 B W	9 W L	2 W			
## 9	5.0 N:3 W	3 B W	32 W L	14 B L	9 W W	47 B W	28 W W	19 W			
## 10	5.0 N:2 W	25 W L	18 B W	59 W W	8 B W	26 W L	7 B W	20 B			
## 11	5.0 N:3 D	16 W L	19 W W	55 B W	31 B D	6 W W	25 B W	18 W			
## 12	4.5 N:3 D	38 B W	56 W W	6 B L	7 W L	3 B W	34 W W	26 B			
## 13	4.5 N:3 W	42 W W	33 B D	5 W W	38 B	H D	1 W L	3 B			
## 14	4.5 N:3 W	36 B W	27 W L	7 B D	5 B W	33 W L	3 W W	32 B			
## 15	4.5 N:3 W	54 W W	44 B W	8 W L	1 W D	27 B L	5 B W	31 W			
## 16	4.5 N:3 D	19 B L	16 B W	30 W L	22 W W	54 B W	33 B W	38 W			
## 17	4.0 N:3 D	10 B W	15 W	H W	39 B L	2 W W	36 B	U			
## 18	4.0 N:3 W	48 W W	41 B L	26 W L	2 B W	23 W W	22 B L	5 W			
## 19	4.0 N:3 W	47 B W	9 W L	1 B W	32 W L	19 B W	38 W L	10 B			
## 20	4.0 N:3 D	15 W W	10 B W	52 W D	28 B W	18 W L	4 W L	8 B			
## 21	4.0 N:4 L	40 W W	49 B W	23 W W	41 B W	28 W L	2 B L	9 W			
## 22	4.0 N:3 W	43 B L	1 W W	47 B L	3 W W	40 W W	39 B L	6 W			
## 23	4.0 N:4 W	64 W D	52 B L	28 W W	15 B	H L	17 W W	40 B			
## 24	4.0 L	4 B W	43 W L	20 B W	58 W L	17 B W	37 W W	46 B			
## 25	4.0 N:4 L	28 B L	47 W W	43 B L	25 B W	60 W W	44 W W	39 B			
## 26	3.5 N:4 L	9 B W	53 W L	3 B W	24 W D	34 B L	10 W W	47 B			
## 27	3.5 N:4 W	49 B W	40 W W	17 B L	4 W L	9 B D	32 W L	11 W			
## 28	3.5 N:4 W	51 W L	13 B W	46 W W	37 B D	14 W L	6 B	U			
## 29	3.5 N:3 W	24 W D	4 W W	22 B D	19 W L	20 B L	8 B D	36 W			
## 30	3.5 N:4 W	50 B D	6 W L	38 B L	34 W W	52 W W	48 B	U			
## 31	3.5 L	52 W D	64 B L	15 B W	55 W L	31 W W	61 B W	50 B			
## 32	3.5 L	58 B D	55 W W	64 B L	10 W W	30 B W	50 W L	14 B			
## 33	3.5 N:4 W	61 W L	8 B W	44 W L	18 B W	51 W D	26 B L	13 W			
## 34	3.5 W	60 B L	12 W W	50 B D	36 W L	13 B L	15 W W	51 B			
## 35	3.5 L	6 B W	60 W L	37 B W	29 B D	25 W L	11 B W	52 W			
## 36	3.5 L	46 W L	38 W W	56 B L	6 W W	57 B D	52 B W	48 W			
## 37	3.5 N:4 L	13 W W	57 B W	51 W D	33 B	H L	16 W D	28 B			
## 38	3.5	B L	5 B W	34 W L	27 W	H L	23 B W	61 W			
## 39	3.0 N:4 D	11 W W	35 B W	29 W L	12 W	H L	18 B L	15 B			
## 40	3.0 N:4 L	1 B W	54 W W	40 B L	16 W W	44 B L	21 W L	24 W			
## 41	3.0 W	20 B L	26 B L	39 W W	59 W L	21 B W	56 W L	22 W			
## 42	3.0 W	59 B L	17 W W	58 B L	20 W	X	U	U			
## 43	3.0 L	12 B L	50 W L	57 B D	60 B D	61 W W	64 W W	56 B			
## 44	3.0 L	21 W L	23 B L	24 W W	63 W W	59 B L	46 B W	55 W			
## 45	3.0	B L	14 W L	32 B W	53 B L	39 W L	24 B W	59 W			
## 46	3.0 L	5 W L	51 B D	60 W L	56 B W	63 W D	55 B W	58 W			
## 47	3.0 W	35 B L	7 W L	27 B L	50 W W	64 B W	43 W L	23 W			
## 48	2.5 L	18 W W	24 B L	21 W W	61 B L	8 W D	51 B L	25 W			
## 49	2.5 L	17 B W	63 W	H D	52 B	H L	29 W L	35 B			
## 50	2.5 L	26 W L	20 W D	63 B D	64 W W	58 B	H	U			

```
## 51      2.5  L  29 W W  42 B L  33 W W  46 B      H L  31 B L  30 W
## 52      2.5  L  27 B W  45 W L  36 B W  57 W L  32 B D  47 W L  33 W
## 53      2.5 N:4 W  30 B D  22 W L  19 B D  48 W L  29 B D  35 W L  34 B
## 54      2.0      H L  25 B      H L  44 W      U W  57 W      U
## 55      2.0  L  14 B L  39 B L  61 W      B L  15 W L  59 B W  64 W
## 56      2.0  L  62 W D  31 B L  10 W L  30 B      B D  45 W L  43 B
## 57      2.0      H L  11 B L  35 W W  45 W      H L  40 B L  42 W
## 58      2.0  L   7 B L  36 W W  42 W L  51 B L  35 W L  53 B      B
## 59      2.0  W  31 W L   2 B L  41 W L  23 B L  49 W      B L  45 B
## 60      2.0  L  41 W      B L   9 B L  40 B L  43 W W  54 W L  44 B
## 61      1.5  L  33 W L  34 B D  45 B D  42 W L  24 B      H      U
## 62      1.5  L  32 B L   3 W W  54 B L  47 W D  42 B L  30 W L  37 B
## 63      1.0  W  55 B      U      U      U      U      U      U
## 64      1.0  L   2 W L  48 B D  49 W L  43 B L  45 B      H      U
## 65      1.0  L  22 B D  30 W L  31 W D  49 B L  46 W L  42 B L  54 B
```

b. Get data elements

```
data_elements <- data.frame(str_extract(raw_data_frame$`Pair Num`, '\\d+'), #Player's ID
  str_trim(str_extract(raw_data_frame$`Player Name USCF ID / Rtg (Pre->Post)`, '[A-Z ]+\\b')), #Player's
  str_extract(raw_data_frame$`Pair Num`, '[A-Z]+'), #Player's State
  str_extract(raw_data_frame$`Total Pts`, '[0-9.]+'), #Player's Total Points
  str_replace(raw_data_frame$`Player Name USCF ID / Rtg (Pre->Post)`, '[A-z0-9 [:PUNCT:]]+\\ / R\\:\\: *([0-9.]*)', '\\1'),
  lapply(raw_data_frame[,4:10], function(x) str_extract(x, '\\d+'))) #IDs of components

#Rename columns
names(data_elements) <- c('ID', 'Player's Name', 'Player's State', 'Total Number of Points', 'Player's Pre-Rating')

#change data type of 'Player's Pre-Rating' to be numeric
data_elements$`Player's Pre-Rating` <- as.numeric(as.character(data_elements$`Player's Pre-Rating`))

#get ratings of all components for each player
data_elements[,6:12] <- data_elements$`Player's Pre-Rating`[match(unlist(data_elements[,6:12]), data_elements$`Player's Pre-Rating`)]

#compute average ratings of components for each player
data_elements$`Average Pre Chess Rating of Opponents` <- round(rowMeans(data_elements[,6:12], na.rm = TRUE), 1)
data_elements
```

```
##      ID      Player's Name Player's State Total Number 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	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's Pre-Rating Round 1 Round 2 Round 3 Round 4 Round 5 Round 6							
##	1	1794	1436	1563	1600	1610	1649	1663
##	2	1553	1175	917	1716	1629	1604	1595

## 3	1384	1641	955	1745	1563	1712	1666
## 4	1716	1363	1507	1553	1579	1655	1564
## 5	1655	1242	980	1663	1666	1716	1610
## 6	1686	1399	1602	1712	1438	1365	1552
## 7	1649	1092	377	1666	1712	1794	1411
## 8	1641	1384	1441	1610	1411	1362	1507
## 9	1411	1745	1600	853	1641	1579	1649
## 10	1365	1604	1564	1186	1494	1686	1745
## 11	1712	1423	1153	1686	1649	1384	1399
## 12	1663	1332	1449	1655	1423	NA	1794
## 13	1666	1355	1552	1649	1655	1449	1384
## 14	1610	1270	1199	1641	1794	1552	1655
## 15	1220	1564	1604	1522	1555	1270	1449
## 16	1604	1365	1220	NA	1436	1553	1355
## 17	1629	1382	1403	1579	1553	1363	1555
## 18	1600	1362	1411	1794	1441	1564	1423
## 19	1564	1220	1365	935	1507	1600	1716
## 20	1595	1348	1291	1363	1403	1507	1553
## 21	1563	1283	1794	1362	1384	1348	1436
## 22	1555	1163	935	1507	1220	NA	1629
## 23	1363	1716	1283	1595	917	1629	980
## 24	1229	1507	1362	1283	1745	967	1199
## 25	1745	1411	1393	1384	1229	1399	1365
## 26	1579	1291	1348	1629	1716	1411	1441
## 27	1552	1011	1666	377	980	1610	1686
## 28	1507	1229	1716	1555	1564	1595	1641
## 29	1602	1056	1686	1423	1399	935	1382
## 30	1522	935	1163	1220	1186	1494	955
## 31	1494	917	1186	1163	1365	1522	1056
## 32	1441	955	1641	1199	1600	1011	1579
## 33	1449	967	1663	1056	1355	1666	1220
## 34	1399	1686	967	980	1602	1745	1712
## 35	1438	377	1423	1153	1686	1092	935
## 36	1355	1666	1092	1011	1449	NA	1604
## 37	980	NA	1655	1399	1552	NA	1363
## 38	1423	1712	1438	1602	1663	NA	1600
## 39	1436	1794	1270	1348	1604	1199	1563
## 40	1348	1595	1579	1436	853	1563	1153
## 41	1403	853	1629	917	1595	NA	NA
## 42	1332	1663	1056	1092	967	955	1163
## 43	1283	1563	1363	1229	1175	853	377
## 44	1199	NA	1610	1441	1393	1436	1229
## 45	1242	1655	1011	967	1153	1175	1186
## 46	377	1438	1649	1552	1056	1163	1283
## 47	1362	1600	1229	1563	955	1641	1011
## 48	1382	1629	1175	NA	935	NA	1602
## 49	1291	1579	1595	1175	1163	917	NA
## 50	1056	1602	1332	1449	377	NA	1494
## 51	1011	1552	1242	1355	1092	1441	1362
## 52	935	1522	1555	1564	1382	1602	1438
## 53	1393	NA	1745	NA	1199	NA	1092
## 54	1270	1610	1436	955	NA	1220	853
## 55	1186	1530	1494	1365	1522	NA	1242
## 56	1153	NA	1712	1438	1242	NA	1348

## 57	1092	1649	1355	1332	1011	1438	1393
## 58	917	1494	1553	1403	1363	1291	NA
## 59	853	1403	NA	1411	1348	1283	1270
## 60	967	1449	1399	1242	1332	1229	NA
## 61	955	1441	1384	1270	1362	1332	1522
## 62	1530	1186	NA	NA	NA	NA	NA
## 63	1175	1553	1382	1291	1283	1242	NA
## 64	1163	1555	1522	1494	1291	377	1332

Round 7 Average Pre Chess Rating of Opponents

## 1	1716	1605
## 2	1649	1469
## 3	1663	1564
## 4	1794	1574
## 5	1629	1501
## 6	1563	1519
## 7	1553	1372
## 8	1564	1468
## 9	1595	1523
## 10	1600	1554
## 11	1579	1468
## 12	1384	1506
## 13	1441	1498
## 14	1494	1515
## 15	1423	1484
## 16	NA	1386
## 17	1655	1499
## 18	1365	1480
## 19	1641	1426
## 20	1411	1411
## 21	1686	1470
## 22	1348	1300
## 23	377	1214
## 24	1436	1357
## 25	1362	1363
## 26	1712	1507
## 27	NA	1222
## 28	1355	1522
## 29	NA	1314
## 30	1056	1144
## 31	1610	1260
## 32	1666	1379
## 33	1011	1277
## 34	935	1375
## 35	1382	1150
## 36	1507	1388
## 37	955	1385
## 38	1220	1539
## 39	1229	1430
## 40	1555	1391
## 41	NA	1248
## 42	1153	1150
## 43	1186	1107
## 44	853	1327
## 45	917	1152


```
## 46      1363      1358
## 47      1745      1392
## 48      1438      1356
## 49        NA      1286
## 50      1522      1296
## 51      1449      1356
## 52      1399      1495
## 53        NA      1345
## 54      1163      1206
## 55      1283      1406
## 56      1332      1414
## 57        NA      1363
## 58      1242      1391
## 59      1199      1319
## 60        NA      1330
## 61       980      1327
## 62        NA      1186
## 63        NA      1350
## 64      1270      1263
```

3. Present final data

```
final_data <- subset(data_elements[,c(2:5,13)])
final_data
```

##	Player's Name	Player's State	Total Number of Points
## 1	GARY HUA	ON	6.0
## 2	DAKSHESH DARURI	MI	6.0
## 3	ADITYA BAJAJ	MI	6.0
## 4	PATRICK H SCHILLING	MI	5.5
## 5	HANSHI ZUO	MI	5.5
## 6	HANSEN SONG	OH	5.0
## 7	GARY DEE SWATHELL	MI	5.0
## 8	EZEKIEL HOUGHTON	MI	5.0
## 9	STEFANO LEE	ON	5.0
## 10	ANVIT RAO	MI	5.0
## 11	CAMERON WILLIAM MC LEMAN	MI	4.5
## 12	KENNETH J TACK	MI	4.5
## 13	TORRANCE HENRY JR	MI	4.5
## 14	BRADLEY SHAW	MI	4.5
## 15	ZACHARY JAMES HOUGHTON	MI	4.5
## 16	MIKE NIKITIN	MI	4.0
## 17	RONALD GRZEGORCZYK	MI	4.0
## 18	DAVID SUNDEEN	MI	4.0
## 19	DIPANKAR ROY	MI	4.0
## 20	JASON ZHENG	MI	4.0
## 21	DINH DANG BUI	ON	4.0
## 22	EUGENE L MCCLURE	MI	4.0
## 23	ALAN BUI	ON	4.0
## 24	MICHAEL R ALDRICH	MI	4.0
## 25	LOREN SCHWIEBERT	MI	3.5

## 26	MAX ZHU	ON	3.5
## 27	GAURAV GIDWANI	MI	3.5
## 28	SOFIA ADINA STANESCU	MI	3.5
## 29	CHIEDOZIE OKORIE	MI	3.5
## 30	GEORGE AVERY JONES	ON	3.5
## 31	RISHI SHETTY	MI	3.5
## 32	JOSHUA PHILIP MATHEWS	ON	3.5
## 33	JADE GE	MI	3.5
## 34	MICHAEL JEFFERY THOMAS	MI	3.5
## 35	JOSHUA DAVID LEE	MI	3.5
## 36	SIDDHARTH JHA	MI	3.5
## 37	AMIYATOSH PWNANANDAM	MI	3.5
## 38	BRIAN LIU	MI	3.0
## 39	JOEL R HENDON	MI	3.0
## 40	FOREST ZHANG	MI	3.0
## 41	KYLE WILLIAM MURPHY	MI	3.0
## 42	JARED GE	MI	3.0
## 43	ROBERT GLEN VASEY	MI	3.0
## 44	JUSTIN D SCHILLING	MI	3.0
## 45	DEREK YAN	MI	3.0
## 46	JACOB ALEXANDER LAVALLEY	MI	3.0
## 47	ERIC WRIGHT	MI	2.5
## 48	DANIEL KHAIN	MI	2.5
## 49	MICHAEL J MARTIN	MI	2.5
## 50	SHIVAM JHA	MI	2.5
## 51	TEJAS AYYAGARI	MI	2.5
## 52	ETHAN GUO	MI	2.5
## 53	JOSE C YBARRA	MI	2.0
## 54	LARRY HODGE	MI	2.0
## 55	ALEX KONG	MI	2.0
## 56	MARISA RICCI	MI	2.0
## 57	MICHAEL LU	MI	2.0
## 58	VIRAJ MOHILE	MI	2.0
## 59	SEAN M MC CORMICK	MI	2.0
## 60	JULIA SHEN	MI	1.5
## 61	JEZZEL FARKAS	ON	1.5
## 62	ASHWIN BALAJI	MI	1.0
## 63	THOMAS JOSEPH HOSMER	MI	1.0
## 64	BEN LI	MI	1.0
##	Player's Pre-Rating Average Pre Chess Rating of Opponents		
## 1	1794		1605
## 2	1553		1469
## 3	1384		1564
## 4	1716		1574
## 5	1655		1501
## 6	1686		1519
## 7	1649		1372
## 8	1641		1468
## 9	1411		1523
## 10	1365		1554
## 11	1712		1468
## 12	1663		1506
## 13	1666		1498
## 14	1610		1515

## 15	1220	1484
## 16	1604	1386
## 17	1629	1499
## 18	1600	1480
## 19	1564	1426
## 20	1595	1411
## 21	1563	1470
## 22	1555	1300
## 23	1363	1214
## 24	1229	1357
## 25	1745	1363
## 26	1579	1507
## 27	1552	1222
## 28	1507	1522
## 29	1602	1314
## 30	1522	1144
## 31	1494	1260
## 32	1441	1379
## 33	1449	1277
## 34	1399	1375
## 35	1438	1150
## 36	1355	1388
## 37	980	1385
## 38	1423	1539
## 39	1436	1430
## 40	1348	1391
## 41	1403	1248
## 42	1332	1150
## 43	1283	1107
## 44	1199	1327
## 45	1242	1152
## 46	377	1358
## 47	1362	1392
## 48	1382	1356
## 49	1291	1286
## 50	1056	1296
## 51	1011	1356
## 52	935	1495
## 53	1393	1345
## 54	1270	1206
## 55	1186	1406
## 56	1153	1414
## 57	1092	1363
## 58	917	1391
## 59	853	1319
## 60	967	1330
## 61	955	1327
## 62	1530	1186
## 63	1175	1350
## 64	1163	1263

4. Save Final Data as CSV

```
write.csv(final_data, 'Final_data.csv', row.names = FALSE)

# Let's check the csv output, looks good:)
read.csv('Final_data.csv')
```

##	Player.s.Name	Player.s.State	Total.Number.of.Points
## 1	GARY HUA	ON	6.0
## 2	DAKSHESH DARURI	MI	6.0
## 3	ADITYA BAJAJ	MI	6.0
## 4	PATRICK H SCHILLING	MI	5.5
## 5	HANSHI ZUO	MI	5.5
## 6	HANSEN SONG	OH	5.0
## 7	GARY DEE SWATHELL	MI	5.0
## 8	EZEKIEL HOUGHTON	MI	5.0
## 9	STEFANO LEE	ON	5.0
## 10	ANVIT RAO	MI	5.0
## 11	CAMERON WILLIAM MC LEMAN	MI	4.5
## 12	KENNETH J TACK	MI	4.5
## 13	TORRANCE HENRY JR	MI	4.5
## 14	BRADLEY SHAW	MI	4.5
## 15	ZACHARY JAMES HOUGHTON	MI	4.5
## 16	MIKE NIKITIN	MI	4.0
## 17	RONALD GRZEGORCZYK	MI	4.0
## 18	DAVID SUNDEEN	MI	4.0
## 19	DIPANKAR ROY	MI	4.0
## 20	JASON ZHENG	MI	4.0
## 21	DINH DANG BUI	ON	4.0
## 22	EUGENE L MCCLURE	MI	4.0
## 23	ALAN BUI	ON	4.0
## 24	MICHAEL R ALDRICH	MI	4.0
## 25	LOREN SCHWIEBERT	MI	3.5
## 26	MAX ZHU	ON	3.5
## 27	GAURAV GIDWANI	MI	3.5
## 28	SOFIA ADINA STANESCU	MI	3.5
## 29	CHIEDOZIE OKORIE	MI	3.5
## 30	GEORGE AVERY JONES	ON	3.5
## 31	RISHI SHETTY	MI	3.5
## 32	JOSHUA PHILIP MATHEWS	ON	3.5
## 33	JADE GE	MI	3.5
## 34	MICHAEL JEFFERY THOMAS	MI	3.5
## 35	JOSHUA DAVID LEE	MI	3.5
## 36	SIDDHARTH JHA	MI	3.5
## 37	AMIYATOSH PWNANANDAM	MI	3.5
## 38	BRIAN LIU	MI	3.0
## 39	JOEL R HENDON	MI	3.0
## 40	FOREST ZHANG	MI	3.0
## 41	KYLE WILLIAM MURPHY	MI	3.0
## 42	JARED GE	MI	3.0
## 43	ROBERT GLEN VASEY	MI	3.0
## 44	JUSTIN D SCHILLING	MI	3.0

## 45	DEREK YAN	MI	3.0
## 46	JACOB ALEXANDER LAVALLEY	MI	3.0
## 47	ERIC WRIGHT	MI	2.5
## 48	DANIEL KHAIN	MI	2.5
## 49	MICHAEL J MARTIN	MI	2.5
## 50	SHIVAM JHA	MI	2.5
## 51	TEJAS AYYAGARI	MI	2.5
## 52	ETHAN GUO	MI	2.5
## 53	JOSE C YBARRA	MI	2.0
## 54	LARRY HODGE	MI	2.0
## 55	ALEX KONG	MI	2.0
## 56	MARISA RICCI	MI	2.0
## 57	MICHAEL LU	MI	2.0
## 58	VIRAJ MOHILE	MI	2.0
## 59	SEAN M MC CORMICK	MI	2.0
## 60	JULIA SHEN	MI	1.5
## 61	JEZZEL FARKAS	ON	1.5
## 62	ASHWIN BALAJI	MI	1.0
## 63	THOMAS JOSEPH HOSMER	MI	1.0
## 64	BEN LI	MI	1.0
##	Player.s.Pre.Rating Average.Pre.Chess.Rating.of.Opponents		
## 1	1794		1605
## 2	1553		1469
## 3	1384		1564
## 4	1716		1574
## 5	1655		1501
## 6	1686		1519
## 7	1649		1372
## 8	1641		1468
## 9	1411		1523
## 10	1365		1554
## 11	1712		1468
## 12	1663		1506
## 13	1666		1498
## 14	1610		1515
## 15	1220		1484
## 16	1604		1386
## 17	1629		1499
## 18	1600		1480
## 19	1564		1426
## 20	1595		1411
## 21	1563		1470
## 22	1555		1300
## 23	1363		1214
## 24	1229		1357
## 25	1745		1363
## 26	1579		1507
## 27	1552		1222
## 28	1507		1522
## 29	1602		1314
## 30	1522		1144
## 31	1494		1260
## 32	1441		1379
## 33	1449		1277

## 34	1399	1375
## 35	1438	1150
## 36	1355	1388
## 37	980	1385
## 38	1423	1539
## 39	1436	1430
## 40	1348	1391
## 41	1403	1248
## 42	1332	1150
## 43	1283	1107
## 44	1199	1327
## 45	1242	1152
## 46	377	1358
## 47	1362	1392
## 48	1382	1356
## 49	1291	1286
## 50	1056	1296
## 51	1011	1356
## 52	935	1495
## 53	1393	1345
## 54	1270	1206
## 55	1186	1406
## 56	1153	1414
## 57	1092	1363
## 58	917	1391
## 59	853	1319
## 60	967	1330
## 61	955	1327
## 62	1530	1186
## 63	1175	1350
## 64	1163	1263

5. Extra Data Analysis

It is fun and challenge in a chess tournament where players are paried according to a certain rating algorithm which matches players with compatible components, which means if a player has high winning rate, he/she will be more likely to paired with a strong component in the next run. Let's see if this is the case in this data set.

```
library(reshape2)
```

```
##
## Attaching package: 'reshape2'

## The following object is masked from 'package:tidyr':
##
##      smiths
```

```
# extract a new subset of data which records results of each round, then compute the wining-rate of each
data <- data.frame(lapply(raw_data_frame[,c(2,4:10)], function(x) str_trim(str_extract(str_trim(x), '^A
```

```
# For a more accurate winning-rate, in this data set only players who completed all 7 rounds of game are
data <-na.omit(data)
names(data)<-c('Name', str_c('Round_', c(1:7)))
data
```

##	Name	Round_1	Round_2	Round_3	Round_4	Round_5
## 1	GARY HUA	W	W	W	W	W
## 2	DAKSHESH DARURI	W	W	L	W	W
## 3	ADITYA BAJAJ	L	W	W	W	W
## 4	PATRICK H SCHILLING	W	D	W	W	D
## 5	HANSHI ZUO	W	W	D	D	D
## 6	HANSEN SONG	W	D	L	W	D
## 7	GARY DEE SWATHELL	W	W	W	W	L
## 8	EZEKIEL HOUGHTON	W	W	L	L	W
## 9	STEFANO LEE	W	L	W	W	W
## 10	ANVIT RAO	D	L	W	W	D
## 11	CAMERON WILLIAM MC LEMAN	D	W	W	L	L
## 13	TORRANCE HENRY JR	W	W	L	D	W
## 14	BRADLEY SHAW	W	W	W	L	D
## 15	ZACHARY JAMES HOUGHTON	D	L	W	L	W
## 17	RONALD GRZEGORCZYK	W	W	L	L	W
## 18	DAVID SUNDEEN	W	W	L	W	L
## 19	DIPANKAR ROY	D	W	W	D	W
## 20	JASON ZHENG	L	W	W	W	W
## 21	DINH DANG BUI	W	L	W	L	W
## 23	ALAN BUI	L	W	L	W	L
## 24	MICHAEL R ALDRICH	L	L	W	L	W
## 25	LOREN SCHWIEBERT	L	W	L	W	D
## 26	MAX ZHU	W	W	W	L	L
## 28	SOFIA ADINA STANESCU	W	D	W	D	L
## 30	GEORGE AVERY JONES	L	D	L	W	L
## 31	RISHI SHETTY	L	D	W	L	W
## 32	JOSHUA PHILIP MATHEWS	W	L	W	L	W
## 33	JADE GE	W	L	W	D	L
## 34	MICHAEL JEFFERY THOMAS	L	W	L	W	D
## 35	JOSHUA DAVID LEE	L	L	W	L	W
## 39	JOEL R HENDON	L	W	W	L	W
## 40	FOREST ZHANG	W	L	L	W	L
## 42	JARED GE	L	L	L	D	D
## 43	ROBERT GLEN VASEY	L	L	L	W	W
## 45	DEREK YAN	L	L	D	L	W
## 46	JACOB ALEXANDER LAVALLEY	W	L	L	L	W
## 47	ERIC WRIGHT	L	W	L	W	L
## 51	TEJAS AYYAGARI	L	W	L	W	L
## 52	ETHAN GUO	W	D	L	D	L
## 61	JEZZEL FARKAS	L	L	W	L	D
## 64	BEN LI	L	D	L	D	L
##	Round_6	Round_7				
## 1	D	D				
## 2	W	W				
## 3	W	W				
## 4	W	D				
## 5	W	W				
## 6	W	W				

```
## 7      W      L
## 8      W      W
## 9      L      W
## 10     W      W
## 11     W      W
## 13     L      W
## 14     L      W
## 15     W      W
## 17     W      L
## 18     W      L
## 19     L      L
## 20     L      L
## 21     W      L
## 23     W      W
## 24     W      W
## 25     L      W
## 26     D      L
## 28     L      D
## 30     W      W
## 31     W      L
## 32     D      L
## 33     L      W
## 34     L      W
## 35     D      W
## 39     L      L
## 40     W      L
## 42     W      W
## 43     L      W
## 45     D      W
## 46     W      L
## 47     D      L
## 51     D      L
## 52     D      L
## 61     L      L
## 64     L      L
```

```
# Reshape the data and compute winning-rate
```

```
data_melt <- melt(data, variable.name = 'Round', value.name = 'Result', id.vars = c('Name'))
data_cast <- dcast(data_melt, Name~Result, fun.aggregate = length)
```

```
## Using Result as value column: use value.var to override.
```

```
data_cast$Win_Rtg <- data_cast$W/7
```

```
# Join the final_data set in previous section
```

```
data_join <- merge(data_cast, final_data, by.x=c('Name'), by.y=c('Player's Name'))
data_join$Rtg_Diff <- data_join[,9]-data_join[,8]
data_join$Win_Seg <- ifelse(data_join$Win_Rtg>=0.5, 'Winning_Rate >= 50%', 'Winning_Rate < 50%')
data_join
```

```
##           Name D L W   Win_Rtg Player's State
## 1      ADITYA BAJAJ 0 1 6 0.8571429      MI
## 2        ALAN BUI 0 3 4 0.5714286      ON
```

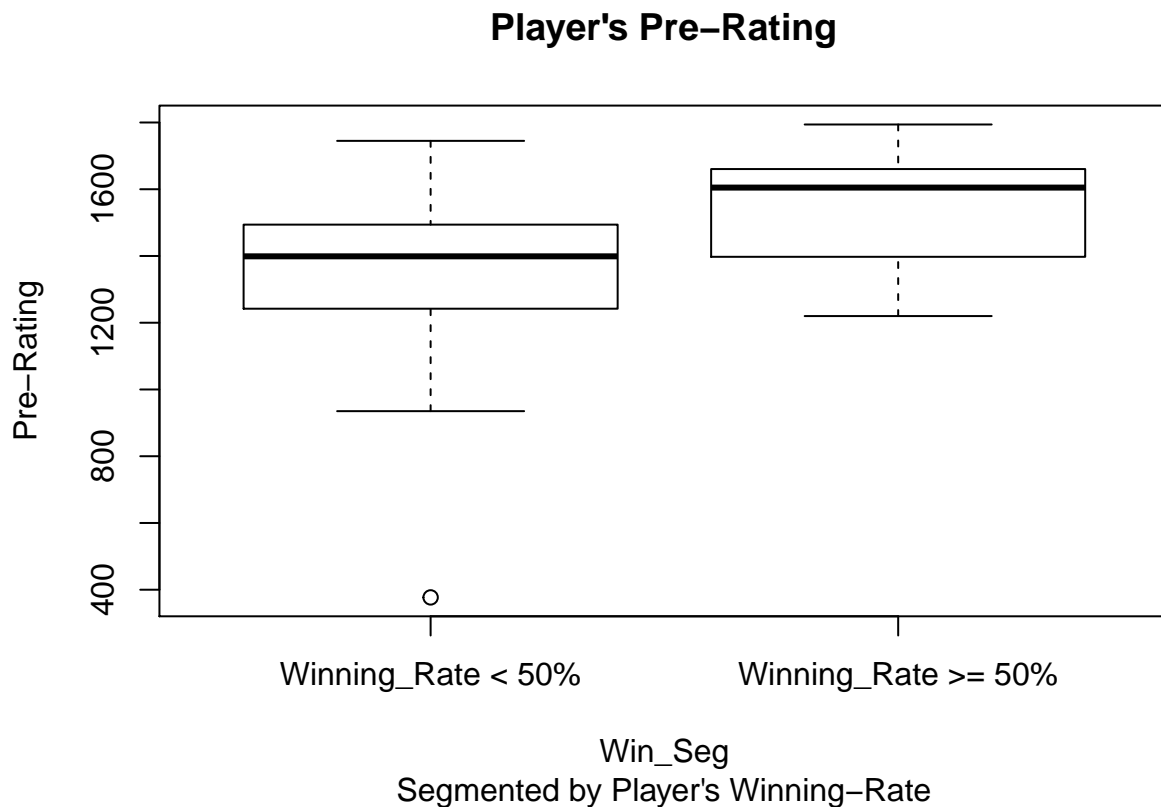

## 3	ANVIT RAO	2	1	4	0.5714286	MI
## 4	BEN LI	2	5	0	0.0000000	MI
## 5	BRADLEY SHAW	1	2	4	0.5714286	MI
## 6	CAMERON WILLIAM MC LEMAN	1	2	4	0.5714286	MI
## 7	DAKSHESH DARURI	0	1	6	0.8571429	MI
## 8	DAVID SUNDEEN	0	3	4	0.5714286	MI
## 9	DEREK YAN	2	3	2	0.2857143	MI
## 10	DINH DANG BUI	0	3	4	0.5714286	ON
## 11	DIPANKAR ROY	2	2	3	0.4285714	MI
## 12	ERIC WRIGHT	1	4	2	0.2857143	MI
## 13	ETHAN GUO	3	3	1	0.1428571	MI
## 14	EZEKIEL HOUGHTON	0	2	5	0.7142857	MI
## 15	FOREST ZHANG	0	4	3	0.4285714	MI
## 16	GARY DEE SWATHELL	0	2	5	0.7142857	MI
## 17	GARY HUA	2	0	5	0.7142857	ON
## 18	GEORGE AVERY JONES	1	3	3	0.4285714	ON
## 19	HANSEN SONG	2	1	4	0.5714286	OH
## 20	HANSHI ZUO	3	0	4	0.5714286	MI
## 21	JACOB ALEXANDER LAVALLEY	0	4	3	0.4285714	MI
## 22	JADE GE	1	3	3	0.4285714	MI
## 23	JARED GE	2	3	2	0.2857143	MI
## 24	JASON ZHENG	0	3	4	0.5714286	MI
## 25	JEZZEL FARKAS	1	5	1	0.1428571	ON
## 26	JOEL R HENDON	0	4	3	0.4285714	MI
## 27	JOSHUA DAVID LEE	1	3	3	0.4285714	MI
## 28	JOSHUA PHILIP MATHEWS	1	3	3	0.4285714	ON
## 29	LOREN SCHWIEBERT	1	3	3	0.4285714	MI
## 30	MAX ZHU	1	3	3	0.4285714	ON
## 31	MICHAEL JEFFERY THOMAS	1	3	3	0.4285714	MI
## 32	MICHAEL R ALDRICH	0	3	4	0.5714286	MI
## 33	PATRICK H SCHILLING	3	0	4	0.5714286	MI
## 34	RISHI SHETTY	1	3	3	0.4285714	MI
## 35	ROBERT GLEN VASEY	0	4	3	0.4285714	MI
## 36	RONALD GRZEGORCZYK	0	3	4	0.5714286	MI
## 37	SOFIA ADINA STANESCU	3	2	2	0.2857143	MI
## 38	STEFANO LEE	0	2	5	0.7142857	ON
## 39	TEJAS AYYAGARI	1	4	2	0.2857143	MI
## 40	TORRANCE HENRY JR	1	2	4	0.5714286	MI
## 41	ZACHARY JAMES HOUGHTON	1	2	4	0.5714286	MI
##	Total Number of Points Player's Pre-Rating					
## 1		6.0			1384	
## 2		4.0			1363	
## 3		5.0			1365	
## 4		1.0			1163	
## 5		4.5			1610	
## 6		4.5			1712	
## 7		6.0			1553	
## 8		4.0			1600	
## 9		3.0			1242	
## 10		4.0			1563	
## 11		4.0			1564	
## 12		2.5			1362	
## 13		2.5			935	
## 14		5.0			1641	

## 15	3.0	1348	
## 16	5.0	1649	
## 17	6.0	1794	
## 18	3.5	1522	
## 19	5.0	1686	
## 20	5.5	1655	
## 21	3.0	377	
## 22	3.5	1449	
## 23	3.0	1332	
## 24	4.0	1595	
## 25	1.5	955	
## 26	3.0	1436	
## 27	3.5	1438	
## 28	3.5	1441	
## 29	3.5	1745	
## 30	3.5	1579	
## 31	3.5	1399	
## 32	4.0	1229	
## 33	5.5	1716	
## 34	3.5	1494	
## 35	3.0	1283	
## 36	4.0	1629	
## 37	3.5	1507	
## 38	5.0	1411	
## 39	2.5	1011	
## 40	4.5	1666	
## 41	4.5	1220	
##	Average Pre Chess Rating of Opponents	Rtg_Diff	Win_Seg
## 1	1564	180	Winning_Rate >= 50%
## 2	1214	-149	Winning_Rate >= 50%
## 3	1554	189	Winning_Rate >= 50%
## 4	1263	100	Winning_Rate < 50%
## 5	1515	-95	Winning_Rate >= 50%
## 6	1468	-244	Winning_Rate >= 50%
## 7	1469	-84	Winning_Rate >= 50%
## 8	1480	-120	Winning_Rate >= 50%
## 9	1152	-90	Winning_Rate < 50%
## 10	1470	-93	Winning_Rate >= 50%
## 11	1426	-138	Winning_Rate < 50%
## 12	1392	30	Winning_Rate < 50%
## 13	1495	560	Winning_Rate < 50%
## 14	1468	-173	Winning_Rate >= 50%
## 15	1391	43	Winning_Rate < 50%
## 16	1372	-277	Winning_Rate >= 50%
## 17	1605	-189	Winning_Rate >= 50%
## 18	1144	-378	Winning_Rate < 50%
## 19	1519	-167	Winning_Rate >= 50%
## 20	1501	-154	Winning_Rate >= 50%
## 21	1358	981	Winning_Rate < 50%
## 22	1277	-172	Winning_Rate < 50%
## 23	1150	-182	Winning_Rate < 50%
## 24	1411	-184	Winning_Rate >= 50%
## 25	1327	372	Winning_Rate < 50%
## 26	1430	-6	Winning_Rate < 50%

## 27	1150	-288	Winning_Rate < 50%
## 28	1379	-62	Winning_Rate < 50%
## 29	1363	-382	Winning_Rate < 50%
## 30	1507	-72	Winning_Rate < 50%
## 31	1375	-24	Winning_Rate < 50%
## 32	1357	128	Winning_Rate >= 50%
## 33	1574	-142	Winning_Rate >= 50%
## 34	1260	-234	Winning_Rate < 50%
## 35	1107	-176	Winning_Rate < 50%
## 36	1499	-130	Winning_Rate >= 50%
## 37	1522	15	Winning_Rate < 50%
## 38	1523	112	Winning_Rate >= 50%
## 39	1356	345	Winning_Rate < 50%
## 40	1498	-168	Winning_Rate >= 50%
## 41	1484	264	Winning_Rate >= 50%

A player who has high wining rate always has high rating. This can be shown in the boxplot below.

```
boxplot(data_join$"Player's Pre-Rating"~Win_Seg, data=data_join, main="Player's Pre-Rating",
        sub="Segmented by Player's Winning-Rate", ylab = 'Pre-Rating')
```



The second boxplot below shows that players with low winning rates have wider range of components in terms of the rating of components, while players with high winning rates are more likely to be paired components with equal or close ratings. combine with the observation in the first boxplot above, players with high winning rates always have higher ratings, we can see that the more games the player win, the higher his/her rating is, and the more likely he/she will pair with components of similar high level.

```
boxplot(Rtg_Diff~Win_Seg, data=data_join,
        main="Difference between Average Rating of Components and Player's Rating",
        sub="Segmented by Player's Winning-Rate", ylab = 'Rating-Difference')
```

