



Analysis of Top Chess Players

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Description of Dataset

Overview

Relative chess skill level is measured by Elo rating. The International Chess Federation, FIDE, acts as the governing body for international chess. FIDE assigns Elo ratings for three speeds of chess (in ascending order): Standard, Rapid, and Blitz. They also award various titles, based on both rating and performance.

This dataset has data of the chess rankings in September 2020.

<https://www.kaggle.com/hunterklotzburwell/top-chess-players>

Selected Subset

For our analysis, we have selected the topchesslist.csv table.

This dataset contains information on 746 active chess players whose Standard FIDE ratings were at least 2500 (as of September 2020).

The 2500 minimum was chosen to match the peak rating requirement for the Grandmaster (GM) title.

The dataset has the following columns :

- Name: Name of the Player
- Title: Whether the player is a GM or an IM
- Country: The player's country of origin
- Standard Rating: FIDE Rating of Standard Format (thinking time >60) of Chess
- Rapid Rating: FIDE Rating of Rapid Format (thinking time >10 and <60) of Chess
- Blitz Rating: FIDE Rating of Blitz Format (thinking time <10) of Chess
- Birth Year: Birth year of the player

Objective

By aggregating the data according to the age and country of the players, we can see how much a country dominates the sport and which age groups occupy the top positions.

We would also visualize our results to make them more intuitive. For this, we'll use pie charts, bar graphs, and other plots included in the ggplot2 and base package.

The following section contains our analysis, observations, and visualizations.

Descriptive Analysis

1. Dataset

```
players <- read.csv("topchesslist.csv")
print("DATASET")
print(players)
```

```
[1] "Top Chess Players DATASET"
```

| | Name | Title | Country | Standard | Rapid | Blitz | Birth.Year |
|----|--------------------------|-------|---------|----------|-------|-------|------------|
| 1 | Carlsen, Magnus | GM | NOR | 2863 | 2881 | 2886 | 1990 |
| 2 | Caruana, Fabiano | GM | USA | 2835 | 2773 | 2711 | 1992 |
| 3 | Ding, Liren | GM | CHN | 2791 | 2836 | 2788 | 1992 |
| 4 | Nepomniachtchi, Ian | GM | RUS | 2784 | 2778 | 2785 | 1990 |
| 5 | Vachier-Lagrave, Maxime | GM | FRA | 2778 | 2860 | 2822 | 1990 |
| 6 | Grischuk, Alexander | GM | RUS | 2777 | 2784 | 2765 | 1983 |
| 7 | Aronian, Levon | GM | ARM | 2773 | 2778 | 2739 | 1982 |
| 8 | So, Wesley | GM | USA | 2770 | 2741 | 2816 | 1993 |
| 9 | Radjabov, Teimour | GM | AZE | 2765 | 2758 | 2757 | 1987 |
| 10 | Mamedyarov, Shakhriyar | GM | AZE | 2764 | 2761 | 2716 | 1985 |
| 11 | Giri, Anish | GM | NED | 2764 | 2731 | 2752 | 1994 |
| 12 | Wang, Hao | GM | CHN | 2763 | 2750 | 2712 | 1989 |
| 13 | Rapport, Richard | GM | HUN | 2760 | 2727 | 2759 | 1996 |
| 14 | Dominguez Perez, Leinier | GM | USA | 2758 | 2786 | 2654 | 1983 |
| 15 | Anand, Viswanathan | GM | IND | 2753 | 2751 | 2785 | 1969 |
| 16 | Duda, Jan-Krzysztof | GM | POL | 2753 | 2774 | 2799 | 1998 |
| 17 | Karjakin, Sergey | GM | RUS | 2752 | 2709 | 2765 | 1990 |
| 18 | Nakamura, Hikaru | GM | USA | 2736 | 2829 | 2900 | 1987 |
| 19 | Topalov, Veselin | GM | BUL | 2735 | 2707 | 2667 | 1975 |
| 20 | Wei, Yi | GM | CHN | 2732 | 2752 | 2686 | 1999 |
| 21 | Harikrishna, Pentala | GM | IND | 2732 | 2705 | 2614 | 1986 |
| 22 | Firouzja, Alireza | GM | FID | 2728 | 2703 | 2770 | 2003 |
| 23 | Vidit, Santosh Gujrathi | GM | IND | 2726 | 2636 | 2735 | 1994 |
| 24 | Andreikin, Dmitry | GM | RUS | 2726 | 2740 | 2782 | 1990 |
| 25 | Svidler, Peter | GM | RUS | 2723 | 2742 | 2754 | 1976 |
| 26 | Vitiugov, Nikita | GM | RUS | 2722 | 2616 | 2673 | 1987 |
| 27 | Wojtaszek, Radoslaw | GM | POL | 2717 | 2703 | 2730 | 1987 |
| 28 | Artemiev, Vladislav | GM | RUS | 2716 | 2769 | 2783 | 1998 |
| 29 | Vallejo Pons, Francisco | GM | ESP | 2710 | 2656 | 2615 | 1982 |
| 30 | Yu, Yangyi | GM | CHN | 2709 | 2738 | 2808 | 1994 |
| 31 | Xiong, Jeffery | GM | USA | 2709 | 2730 | 2724 | 2000 |
| 32 | Le, Quang Liem | GM | VIE | 2709 | 2744 | 2690 | 1991 |

```
print("Type of dataset")
print(class(players))
print(mode(players))
print(typeof(players))

obs <- nrow(players)
var <- ncol(players)
print("no of observations")
print(obs)
print("no of variables")
print(var)
attach(players)
```

```
[1] "Type of Dataset : "
[1] "data.frame"
[1] "list"
[1] "list"
[1] "Number of observations : "
[1] 746
[1] "Number of variables : "
[1] 7
```

2. Analysis based on Standard Rating

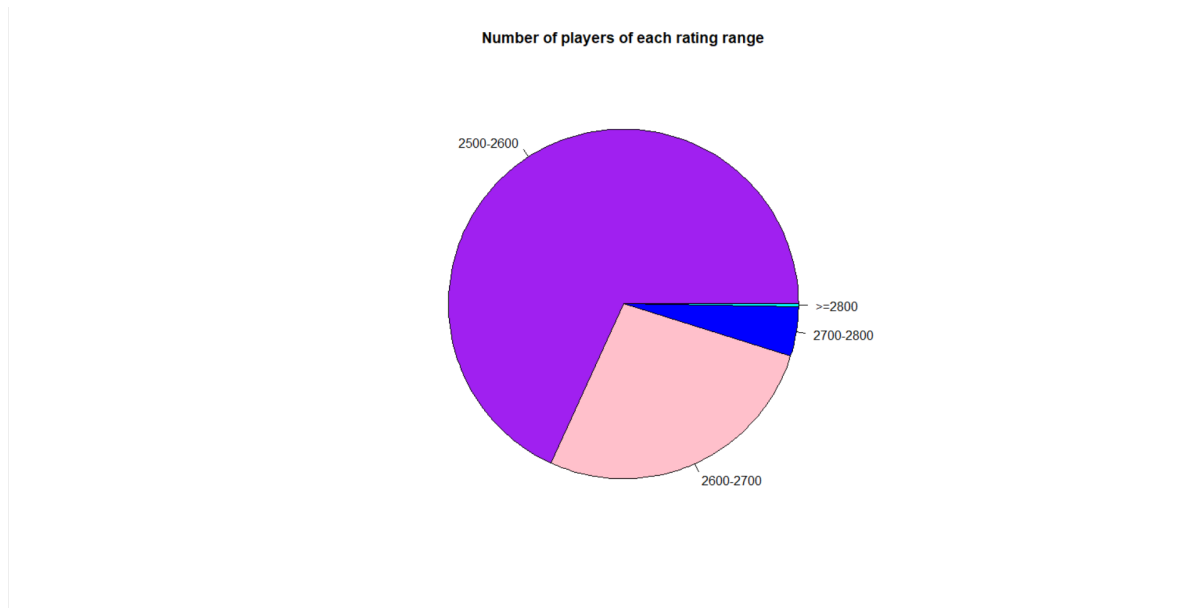
```
print("Info about Standard Rating")
print(summary(players$Standard))
```

```
[1] "Info about Standard Rating"
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
 2500   2525   2562   2577   2614   2863
```

```
print("Different range of Standard Ratings")
n <- c()
print("No. of players of rating 2500-2600")
n[1] <- length(Standard[Standard < 2600])
print(n[1])
print("No. of players of rating 2600-2700")
n[2] <- length(Standard[(Standard >= 2600) & (Standard < 2700)])
print(n[2])
print("No. of players of rating 2700-2800")
n[3] <- length(Standard[(Standard >= 2700) & (Standard < 2800)])
print(n[3])
print("No. of players of rating greater 2800")
n[4] <- length(Standard[Standard >= 2800])
print(n[4])
```

```
[1] "Different range of Standard Ratings"
[1] "No. of players of rating 2500-2600"
[1] 509
[1] "No. of players of rating 2600-2700"
[1] 201
[1] "No. of players of rating 2700-2800"
[1] 34
[1] "No. of players of rating greater 2800"
[1] 2
```

```
pie(n, c("2500-2600", "2600-2700", "2700-2800", ">=2800"), main = "Number of players of e
```



By observing this pie chart, we can conclude that only few (2) players have breached the 2800 Standard rating while most of the GM Ratings lie in the 2500-2600 bracket.

3. Analysis based on Country

```
# Number of players by Country
agg_country <- aggregate(Standard ~ (Country), players, function(x) which.max(table(x)))
print(agg_country)
```

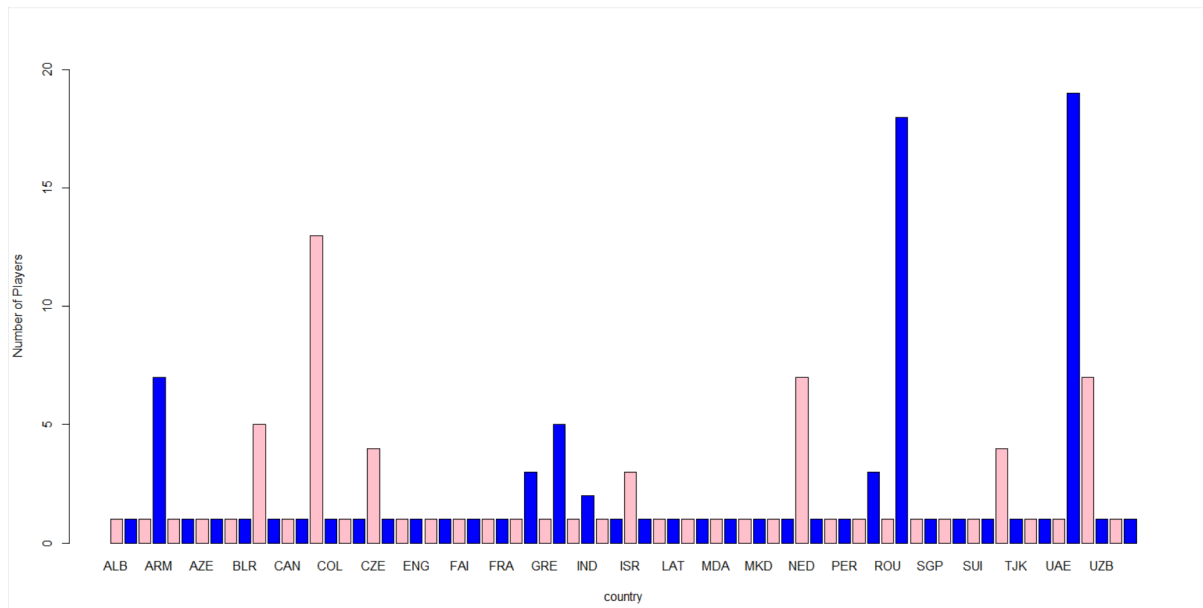
| | Country | Standard |
|----|---------|----------|
| 1 | ALB | 1 |
| 2 | ALG | 1 |
| 3 | ARG | 1 |
| 4 | ARM | 7 |
| 5 | AUS | 1 |
| 6 | AUT | 1 |
| 7 | AZE | 1 |
| 8 | BEL | 1 |
| 9 | BIH | 1 |
| 10 | BLR | 1 |
| 11 | BRA | 5 |
| 12 | BUL | 1 |
| 13 | CAN | 1 |
| 14 | CHI | 1 |
| 15 | CHN | 13 |
| 16 | COL | 1 |
| 17 | CRO | 1 |
| 18 | CUB | 1 |
| 19 | CZE | 4 |
| 20 | DEN | 1 |
| 21 | EGY | 1 |
| 22 | ENG | 1 |
| 23 | ESP | 1 |
| 24 | EST | 1 |
| 25 | FAI | 1 |
| 26 | FID | 1 |
| 27 | FIN | 1 |
| 28 | FRA | 1 |
| 29 | GEO | 1 |
| 30 | GER | 3 |
| 31 | GRE | 1 |
| 32 | HUN | 5 |
| 33 | INA | 1 |
| 34 | IND | 2 |
| 35 | IRI | 1 |
| 36 | ISL | 1 |
| 37 | ISR | 3 |
| 38 | ITA | 1 |

| | | |
|----|-----|----|
| 39 | KAZ | 1 |
| 40 | LAT | 1 |
| 41 | LTU | 1 |
| 42 | MAS | 1 |
| 43 | MDA | 1 |
| 44 | MEX | 1 |
| 45 | MGL | 1 |
| 46 | MKD | 1 |
| 47 | MNE | 1 |
| 48 | MYA | 1 |
| 49 | NED | 7 |
| 50 | NOR | 1 |
| 51 | PAR | 1 |
| 52 | PER | 1 |
| 53 | PHI | 1 |
| 54 | POL | 3 |
| 55 | ROU | 1 |
| 56 | RUS | 18 |
| 57 | SCO | 1 |
| 58 | SGP | 1 |
| 59 | SLO | 1 |
| 60 | SRB | 1 |
| 61 | SUI | 1 |
| 62 | SVK | 1 |
| 63 | SWE | 4 |
| 64 | TJK | 1 |
| 65 | TKM | 1 |
| 66 | TUR | 1 |
| 67 | UAE | 1 |
| 68 | UKR | 19 |
| 69 | USA | 7 |
| 70 | UZB | 1 |
| 71 | VEN | 1 |
| 72 | VIE | 1 |

```

barplot(agg_country$Standard,
        width = 1, names.arg = agg_country$Country,
        horiz = FALSE, xlab = "country", ylab = "Number of Players", axes = TRUE,
        col = c("pink", "blue"), ylim = c(0, 20)
)

```

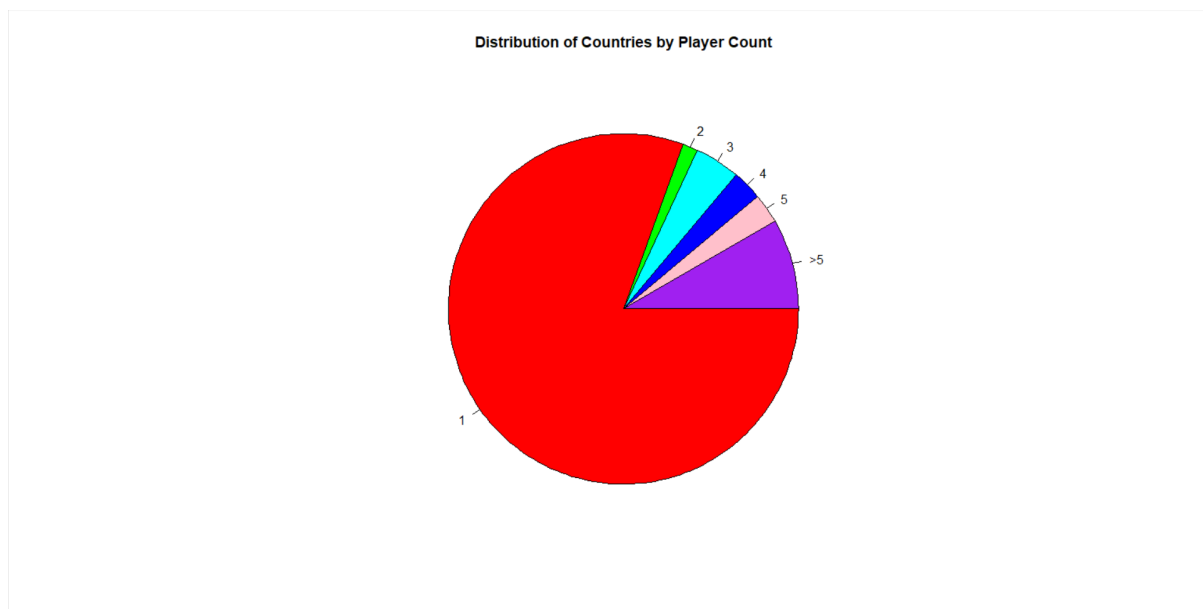


```
print("Different range of players by Country")
n <- c()
print("Countries with more than 5 players")
n[1] <- length(agg_country$Standard[agg_country$Standard > 5])
print(n[1])
print("Countries with 5 players")
n[2] <- length(agg_country$Standard[agg_country$Standard == 5])
print(n[2])
print("Countries with 4 players")
n[3] <- length(agg_country$Standard[agg_country$Standard == 4])
print(n[3])
print("Countries with 3 players")
n[4] <- length(agg_country$Standard[agg_country$Standard == 3])
print(n[4])
print("Countries with 2 players")
n[5] <- length(agg_country$Standard[agg_country$Standard == 2])
print(n[5])
print("Countries with only 1 player")
n[6] <- length(agg_country$Standard[agg_country$Standard == 1])
print(n[6])
```

By observing the bar graph, we can see that a select few countries (namely Ukraine(19), Russia(18), and China(13)) are dominating the top chess rankings.


```
[1] "Different range of players by Country"
[1] "Countries with more than 5 players"
[1] 6
[1] "Countries with 5 players"
[1] 2
[1] "Countries with 4 players"
[1] 2
[1] "Countries with 3 players"
[1] 3
[1] "Countries with 2 players"
[1] 1
[1] "Countries with only 1 player"
[1] 58
```

```
pie(n, c(">5", "5", "4", "3", "2", "1"), main = "Distribution of Countries by Player Count")
```



By observing this pie chart, we can see that most of the countries have only one player representing them with a rating above 2500 while only few countries have more than 5.

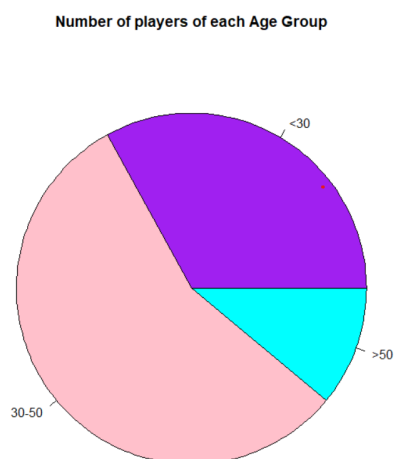
4. Analysis based on Age

```
players_age <- players
players_age$Birth.Year <- 2020 - players_age$Birth.Year
print("Different Age Range of Players")
n <- c()
print("Number of players of Age less than 30")
n[1] <- length(players_age$Birth.Year[players_age$Birth.Year < 30])
print(n[1])
```

```
print("Number of players of Age 30-50")
n[2] <- length(players_age$Birth.Year[(players_age$Birth.Year >= 30) & (players_age$Birth.Year <= 50)])
print(n[2])
print("Number of players of Age greater than 50")
n[3] <- length(players_age$Birth.Year[players_age$Birth.Year > 50])
print(n[3])
```

```
[1] "Different Age Range of Players"
[1] "Number of players of Age less than 30"
[1] 246
[1] "Number of players of Age 30-50"
[1] 418
[1] "Number of players of Age greater than 50"
[1] 82
```

```
pie(n, c("<30", "30-50", ">50"), main = "Number of players of each Age range", col = c("red", "blue", "green"))
```



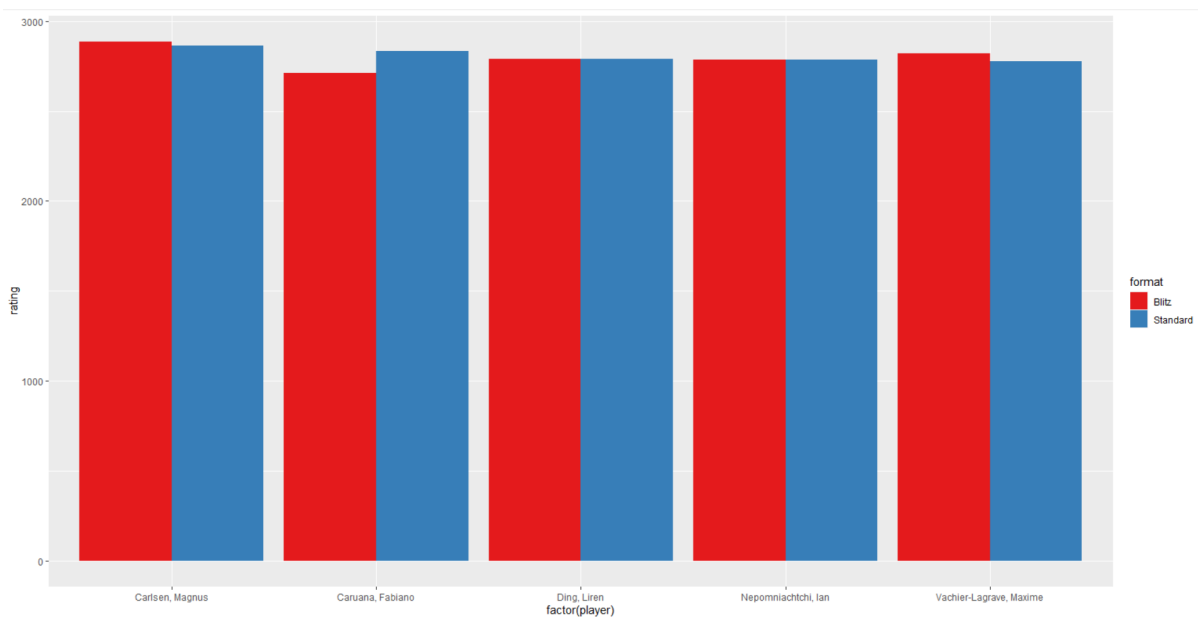
By observing this pie chart, we can notice that most of the Chess Grandmasters are aged between 30 and 50 while there are quite a few younger GM as well who are aged under 30. Old people i.e. aged greater than 50 are less likely to be Chess GMs, quite contrary to the public perception of the sport.

5. Analysis of Standard Vs. Blitz Rating

```
# Comparison Between Standard and Blitz Ratings
players_ratings <- subset(players, select = c("Name", "Standard", "Blitz"))
print("Ratings of the top 5 players")
print(head(players_ratings, 5))
```

```
[1] "Ratings of the top 5 players"
      Name Standard Blitz
1  Carlsen, Magnus    2863  2886
2  Caruana, Fabiano    2835  2711
3    Ding, Liren      2791  2788
4 Nepomniachtchi, Ian    2784  2785
5 Vachier-Lagrave, Maxime 2778  2822
```

```
rating_gen <- data.frame(
  format <- c("Standard", "Standard", "Standard", "Standard", "Standard", "Blitz", "Blitz"),
  player <- c(
    "Carlsen, Magnus", "Caruana, Fabiano", "Ding, Liren", "Nepomniachtchi, Ian", "Vachier-Lagrave, Maxime",
    "Carlsen, Magnus", "Caruana, Fabiano", "Ding, Liren", "Nepomniachtchi, Ian", "Vachier-Lagrave, Maxime"
  ),
  rating <-
    c(
      players_ratings$Standard[1], players_ratings$Standard[2],
      players_ratings$Standard[3], players_ratings$Standard[4],
      players_ratings$Standard[5], players_ratings$Blitz[1],
      players_ratings$Blitz[2], players_ratings$Blitz[3],
      players_ratings$Blitz[4], players_ratings$Blitz[5]
    )
)
print(ggplot(rating_gen, aes(factor(player), rating, fill = format)) +
  geom_bar(stat = "identity", position = "dodge") +
  scale_fill_brewer(palette = "Set1"))
```



As we can see in the graph, for the top players, there isn't that great of a difference between their Standard Rating and their Blitz Rating. This implies that time constraint doesn't have that much of an effect on the performance of the top chess players.

Source Code

The source code for this project can be accessed using the following link :

https://github.com/Chaitanya-Raj/Data_Science_Project