



American International University, Bangladesh

Introduction to Data Science

Project

Final Term

Topic: Web Scrapping

Section: C

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Project Overview

The project consists with two datasets. These datasets are scrapped from a website called “fifa index” here they publish the ratings of the football clubs all over the world every year. Here in our data basically contains the Attacking, Midfield, Defense, and the overall ratings. Also, the corresponding teams name and the league name where they belong to are also included. First dataset was collected from the 2022 edition of top 30 clubs and the 2nd datasets are collected from 2021 edition. The attributes are common in the datasets, and they are,

- Name of the top 30 clubs
- Corresponding League names
- Attack, defense, midfield, and overall rating lists for 30 clubs of two datasets.

After the scrapping the data then we use necessary cleaning process and data handling methods as needed. Lastly, we will give a descriptive statistics analysis and a graph for final analysis.

Project solution design

The solution for this project is, we scrapped the data from the desired website using selectorgadget tool and Rstudio. After importing the data to a csv file. Then we will clean and sorted data for our desired purpose through data pre-processing. This requires few steps of data pre-processing which are:

1. Data Cleaning
 - Smooth Noisy Data
 - Handling Missing Data
 - Data Wrangling or Munging
2. Data Reduction
3. Data Transformation
4. Data Integration
5. Data Discretization

These steps are required for processing the data to get a clean dataset for our desired work. Some data are dirty and noisy that is they are incomplete or is not consistence that is they have a different value set which might have resulted from typing error or any other mistake, data might be missing, or data might have extreme values that do not go with dataset ranges, data can have many more errors and outliers, negative values, small digits than its range, might have alphabetical or numerical errors. To overcome all these in our dataset, we will be preprocessing our data now.

Then by using quantitative analysis and simple graphics we will be able to give a descriptive Statistics report. At the very end, a plotting graph will be shown to visualize the data.

Data Scrapping

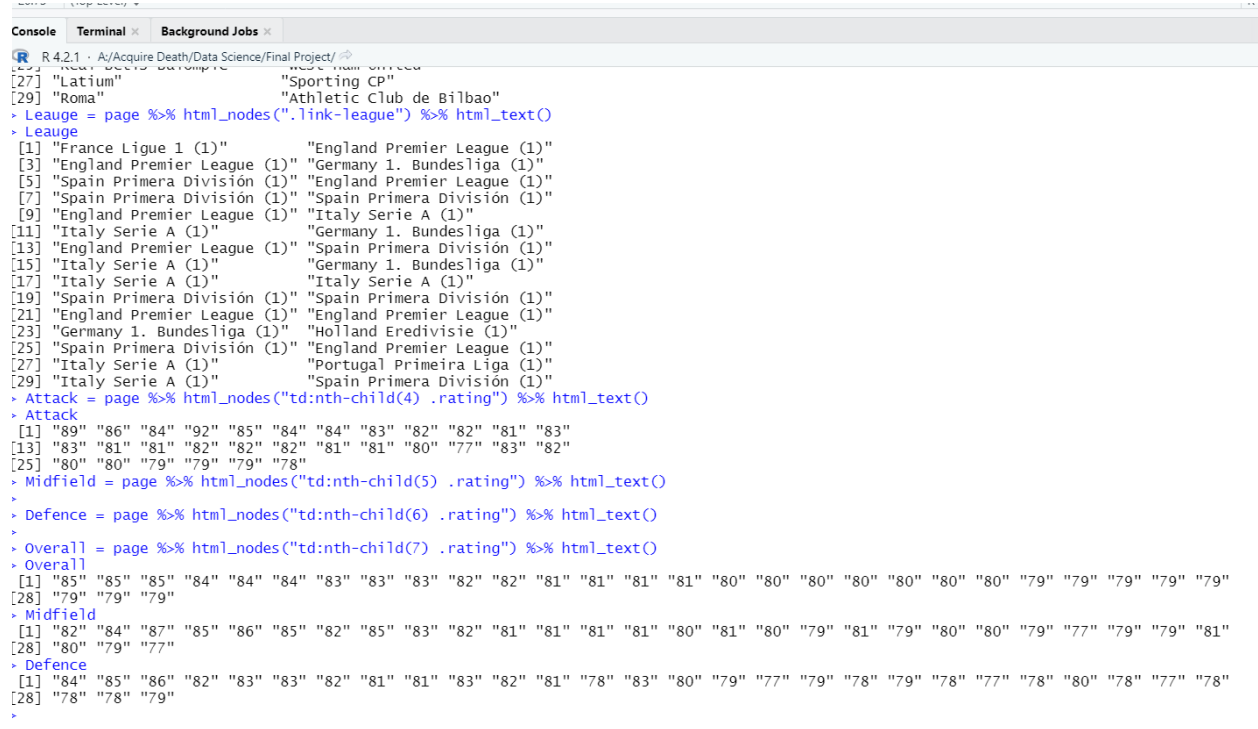
Data is scrapped from “<https://ifaindex.com/teams/>” this website. The scrapping process is done with SelectorGadget.

Code:

For FIFA 22 Dataset

```
#install.packages("rvest")
#install.packages("dplyr")
library(rvest)
library(dplyr)
link = "https://www.fifaindex.com/teams/fifa22/"
page = read_html(link)

Club = page %>% html_nodes("td+ td .link-team") %>% html_text()
League = page %>% html_nodes(".link-league") %>% html_text()
Attack = page %>% html_nodes("td:nth-child(4) .rating") %>% html_text()
Midfield = page %>% html_nodes("td:nth-child(5) .rating") %>% html_text()
Defence = page %>% html_nodes("td:nth-child(6) .rating") %>% html_text()
Overall = page %>% html_nodes("td:nth-child(7) .rating") %>% html_text()
FIFA22 = data.frame(Club, League, Attack, Midfield, Defence, Overall, stringsAsFactors = FALSE)
write.csv(FIFA22, "fifa22.csv")
```



```
Console Terminal Background Jobs
R 4.2.1 - A:/Acquire Death/Data Science/Final Project/
[27] "Latium" "Sporting CP"
[29] "Roma" "Athletic Club de Bilbao"
> League = page %>% html_nodes(".link-league") %>% html_text()
> League
[1] "France Ligue 1 (1)" "England Premier League (1)"
[3] "England Premier League (1)" "Germany 1. Bundesliga (1)"
[5] "Spain Primera División (1)" "England Premier League (1)"
[7] "Spain Primera División (1)" "Spain Primera División (1)"
[9] "England Premier League (1)" "Italy Serie A (1)"
[11] "Italy Serie A (1)" "Germany 1. Bundesliga (1)"
[13] "England Premier League (1)" "Spain Primera División (1)"
[15] "Italy Serie A (1)" "Germany 1. Bundesliga (1)"
[17] "Italy Serie A (1)" "Italy Serie A (1)"
[19] "Spain Primera División (1)" "Spain Primera División (1)"
[21] "England Premier League (1)" "England Premier League (1)"
[23] "Germany 1. Bundesliga (1)" "Holland Eredivisie (1)"
[25] "Spain Primera División (1)" "England Premier League (1)"
[27] "Italy Serie A (1)" "Portugal Primeira Liga (1)"
[29] "Italy Serie A (1)" "Spain Primera División (1)"
> Attack = page %>% html_nodes("td:nth-child(4) .rating") %>% html_text()
> Attack
[1] "89" "86" "84" "92" "85" "84" "84" "83" "82" "81" "83"
[13] "83" "81" "81" "82" "82" "82" "81" "81" "80" "77" "83" "82"
[25] "80" "80" "79" "79" "79" "78"
> Midfield = page %>% html_nodes("td:nth-child(5) .rating") %>% html_text()
>
> Defence = page %>% html_nodes("td:nth-child(6) .rating") %>% html_text()
>
> Overall = page %>% html_nodes("td:nth-child(7) .rating") %>% html_text()
> Overall
[1] "85" "85" "85" "84" "84" "84" "83" "83" "83" "82" "82" "81" "81" "81" "81" "80" "80" "80" "80" "80" "80" "79" "79" "79" "79" "79"
[28] "79" "79" "79"
> Midfield
[1] "82" "84" "87" "85" "86" "85" "82" "85" "83" "82" "81" "81" "81" "81" "80" "81" "80" "79" "81" "79" "80" "80" "79" "77" "79" "79" "81"
[28] "80" "79" "77"
> Defence
[1] "84" "85" "86" "82" "83" "83" "82" "81" "81" "83" "82" "81" "78" "83" "80" "79" "77" "79" "78" "79" "78" "77" "78" "80" "78" "77" "78"
[28] "78" "78" "79"
>
```

For FIFA 21 Dataset

```
library(rvest)
```

```
library(dplyr)
```

```
link2 = "https://www.fifaindex.com/teams/fifa21_486/"
```

```
page2 = read_html(link2)
```

```
Club = page2 %>% html_nodes("td+ td .link-team") %>% html_text()
```

```
League = page2 %>% html_nodes(".link-league") %>% html_text()
```

```
Attack = page2 %>% html_nodes("td:nth-child(4) .rating") %>% html_text()
```

```
Midfield = page2 %>% html_nodes("td:nth-child(5) .rating") %>% html_text()
```

```
Defence = page2 %>% html_nodes("td:nth-child(6) .rating") %>% html_text()
```

```
Overall = page2 %>% html_nodes("td:nth-child(7) .rating") %>% html_text()
```

```
FIFA21 = data.frame(Club, League, Attack, Midfield, Defence, Overall, stringsAsFactors = FALSE)
```

```
write.csv(FIFA21, "fifa21.csv")
```

Datasets

A	B	C	D	E	F	G	H
	Club	League	Attack	Midfield	Defence	Overall	
1	Paris Saint-Germain	France Ligue 1 (1)	89	82	84	85	
2	Liverpool	England Premier League (1)	86	84	85	85	
3	Manchester City	England Premier League (1)	84	87	86	85	
4	Bayern M�nchen	Germany 1. Bundesliga (1)	92	85	82	84	
5	Real Madrid	Spain Primera Divisi�n (1)	85	86	83	84	
6	Chelsea	England Premier League (1)	84	85	83	84	
7	Atl�tico de Madrid	Spain Primera Divisi�n (1)	84	82	82	83	
8	FC Barcelona	Spain Primera Divisi�n (1)	83	85	81	83	
9	Manchester United	England Premier League (1)	82	83	81	83	
10	Inter	Italy Serie A (1)	82	82	83	82	
11	Juventus	Italy Serie A (1)	81	81	82	82	
12	Borussia Dortmund	Germany 1. Bundesliga (1)	83	81	81	81	
13	Tottenham Hotspur	England Premier League (1)	83	81	78	81	
14	Sevilla FC	Spain Primera Divisi�n (1)	81	81	83	81	
15	Milan	Italy Serie A (1)	81	80	80	81	
16	RB Leipzig	Germany 1. Bundesliga (1)	82	81	79	80	
17	Napoli	Italy Serie A (1)	82	80	77	80	
18	Bergamo Calcio	Italy Serie A (1)	82	79	79	80	
19	Real Sociedad	Spain Primera Divisi�n (1)	81	81	78	80	
20	Villarreal CF	Spain Primera Divisi�n (1)	81	79	79	80	
21	Leicester City	England Premier League (1)	80	80	78	80	
22	Arsenal	England Premier League (1)	77	80	77	80	
23	Bayer 04 Leverkusen	Germany 1. Bundesliga (1)	83	79	78	79	
24	Ajax	Holland Eredivisie (1)	82	77	80	79	
25	Real Betis Balomp�	Spain Primera Divisi�n (1)	80	79	78	79	
26	West Ham United	England Premier League (1)	80	79	77	79	
27	Latium	Italy Serie A (1)	79	81	78	79	
28	Sporting CP	Portugal Primeira Liga (1)	79	80	78	79	
29	Roma	Italy Serie A (1)	79	79	78	79	
30	Athletic Club de Bilbao	Spain Primera Divisi�n (1)	78	77	79	79	

FIG: Dataset 1 (fifa22)

A	B	C	D	E	F	G	H
	Club	League	Attack	Midfield	Defence	Overall	
1	Liverpool	Premier League	86	83	80	85	
2	Manchester City	Premier League	85	86	83	85	
3	Real Madrid	LaLiga Santander	84	86	84	85	
4	FC Bayern M�nchen	Bundesliga	92	85	82	84	
5	Paris Saint-Germain	Ligue 1 Uber Eats	87	82	82	84	
6	Juventus	Serie A TIM	87	81	83	83	
7	Atl�tico de Madrid	LaLiga Santander	85	82	83	83	
8	FC Barcelona	LaLiga Santander	84	82	81	83	
9	Tottenham Hotspur	Premier League	89	82	81	82	
10	Manchester United	Premier League	83	82	82	82	
11	Inter	Serie A TIM	83	81	83	82	
12	Chelsea	Premier League	82	84	82	82	
13	Borussia Dortmund	Bundesliga	87	81	81	81	
14	Sevilla FC	LaLiga Santander	80	81	82	81	
15	Arsenal	Premier League	83	79	79	80	
16	Napoli	Serie A TIM	82	80	80	80	
17	Lazio	Serie A TIM	82	80	79	80	
18	Villarreal CF	LaLiga Santander	82	79	79	80	
19	Leicester City	Premier League	80	81	79	80	
20	RB Leipzig	Bundesliga	78	81	80	80	
21	Milan	Serie A TIM	84	78	80	79	
22	Atalanta	Serie A TIM	82	79	78	79	
23	Real Sociedad	LaLiga Santander	81	81	77	79	
24	Everton	Premier League	81	79	80	79	
25	Athletic Club	LaLiga Santander	80	78	79	79	
26	Borussia M�nchengladbach	Bundesliga	78	79	80	79	
27	Bayer 04 Leverkusen	Bundesliga	78	78	78	79	
28	Wolverhampton Wanderers	Premier League	77	81	78	79	
29	SL Benfica	Liga NOS	77	79	78	79	
30	Ajax	Eredivisie	80	77	76	78	

FIG: Dataset 2 (fifa21)

Data pre-processing:

Data preprocessing is one of the most crucial parts for data analysis. Without data preprocessing data scientists can never produce an accurate data for their research. Hence, data preprocessing is an integral part data science or any field. There are various processes to clean and process data before they are utilized for data analysis. They are the following:

1. Data Cleaning

- Smooth Noisy Data:

Now smoothy noisy data mean if there is any negative value in the Dataset then it needed to be handled. But in our case the dataset contains no such data. So, we can simply skip this part of the process.

- **Handling Missing Data:**

Though we don't have missing value here in our datasets yet we are writing missing data handling code so that there is any missing data then it will be handled using mean function.

Code:

```
#Missing Data handle
```

```
mean(fifa22$Attack)
```

```
meanAs <- mean(fifa22$Attack, na.rm = TRUE)
```

```
print(meanAs)
```

```
fifa22[is.na(fifa22$Attack), "Attack"] <- meanAs
```

```
print(fifa22)
```

2. Data Integration

Here no need for data integration but we write code to classify the teams give them a tag if above 86 the S means super class, 82 then A class then 80 for B and below 80 is C class. Though it is unnecessary for our datasets.

Code:

```
#Data Integration
```

```
for(i in 1:nrow(fifa22)){
```

```
  if(fifa22$overall[i] <= 86 ){
```

```
    fifa22$Type[i] <- "S"
```

```
  }else if(fifa22$overall [i] <= 82){
```

```
    fifa22$Type [i] <- "A"
```

```
  }else if(fifa22$overall [i] <= 80){
```

```
    fifa22$Type [i] <- "B"
```

```
  }else { fifa22$Type [i] <- "C"}
```

```
}
```

3. Data Reduction: We do data reduction to reduce unnecessary data from the dataset that can be used to attain the same results. In our dataset,

fifa22, there were additional (1). We have reduced the additional 1s in brackets through data reduction.

Code: `fifa22$Leauge <- gsub("[(1)]", "", fifa22$Leauge)`
`view(fifa22)`

RStudio Source Editor							
fifa22 x							
Filter							
X	Club	Leauge	Attack	Midfield	Defence	Overall	
1	1 Paris Saint-Germain	France Ligue	89	82	84	85	
2	2 Liverpool	England Premier League	86	84	85	85	
3	3 Manchester City	England Premier League	84	87	86	85	
4	4 Bayern München	Germany . Bundesliga	92	85	82	84	
5	5 Real Madrid	Spain Primera División	85	86	83	84	
6	6 Chelsea	England Premier League	84	85	83	84	
7	7 Atlético de Madrid	Spain Primera División	84	82	82	83	
8	8 FC Barcelona	Spain Primera División	83	85	81	83	
9	9 Manchester United	England Premier League	82	83	81	83	
10	10 Inter	Italy Serie A	82	82	83	82	
11	11 Juventus	Italy Serie A	81	81	82	82	
12	12 Borussia Dortmund	Germany . Bundesliga	83	81	81	81	
13	13 Tottenham Hotspur	England Premier League	83	81	78	81	
14	14 Sevilla FC	Spain Primera División	81	81	83	81	
15	15 Milan	Italy Serie A	81	80	80	81	
16	16 RB Leipzig	Germany . Bundesliga	82	81	79	80	
17	17 Napoli	Italy Serie A	82	80	77	80	
18	18 Bergamo Calcio	Italy Serie A	82	79	79	80	
19	19 Real Sociedad	Spain Primera División	81	81	78	80	
20	20 Villarreal CF	Spain Primera División	81	79	79	80	
21	21 Leicester City	England Premier League	80	80	78	80	
22	22 Arsenal	England Premier League	77	80	77	80	
23	23 Bayer 04 Leverkusen	Germany . Bundesliga	83	79	78	79	
24	24 Ajax	Holland Eredivisie	82	77	80	79	
25	25 Real Betis Balompié	Spain Primera División	80	79	78	79	
26	26 West Ham United	England Premier League	80	79	77	79	
27	27 Latium	Italy Serie A	79	81	78	79	
28	28 Sporting CP	Portugal Primeira Liga	79	80	78	79	
29	29 Roma	Italy Serie A	79	79	78	79	
30	30 Athletic Club de Bilbao	Spain Primera División	78	77	79	79	

- 4. Data Discretization:** In our dataset, fifa22, we categorized the league row into one category. Earlier it was about all data individually written which looked unclear and messy, so we categorized it into whole rows.

Code:

```
class(fifa22$League)
fifa22$League[fifa22$League=="France Ligue "] <- "1"
fifa22$League[fifa22$League=="England Premier League "] <- "2"
fifa22$League[fifa22$League=="Germany . Bundesliga "] <- "3"
fifa22$League[fifa22$League=="Spain Primera División "] <- "4"
fifa22$League[fifa22$League=="Italy Serie A "] <- "5"
fifa22$League[fifa22$League=="Holland Eredivisie "] <- "6"
fifa22$League[fifa22$League=="Portugal Primeira Liga "] <- "7"
```

```
class(fifa21$League)
fifa21$League[fifa21$League=="Ligue 1 Uber Eats"] <- "1"
fifa21$League[fifa21$League=="Premier League"] <- "2"
fifa21$League[fifa21$League=="Bundesliga"] <- "3"
fifa21$League[fifa21$League=="LaLiga Santander"] <- "4"
fifa21$League[fifa21$League=="Serie A TIM"] <- "5"
fifa21$League[fifa21$League=="Eredivisie"] <- "6"
fifa21$League[fifa21$League=="Liga NOS"] <- "7"
```


fifa22 x							
Filter							
	X	Club	League	Attack	Midfield	Defence	Overall
1	1	Paris Saint-Germain	1	89	82	84	85
2	2	Liverpool	2	86	84	85	85
3	3	Manchester City	2	84	87	86	85
4	4	Bayern München	3	92	85	82	84
5	5	Real Madrid	4	85	86	83	84
6	6	Chelsea	2	84	85	83	84
7	7	Atlético de Madrid	4	84	82	82	83
8	8	FC Barcelona	4	83	85	81	83
9	9	Manchester United	2	82	83	81	83
10	10	Inter	5	82	82	83	82
11	11	Juventus	5	81	81	82	82
12	12	Borussia Dortmund	3	83	81	81	81
13	13	Tottenham Hotspur	2	83	81	78	81
14	14	Sevilla FC	4	81	81	83	81
15	15	Milan	5	81	80	80	81
16	16	RB Leipzig	3	82	81	79	80
17	17	Napoli	5	82	80	77	80
18	18	Bergamo Calcio	5	82	79	79	80
19	19	Real Sociedad	4	81	81	78	80
20	20	Villarreal CF	4	81	79	79	80
21	21	Leicester City	2	80	80	78	80
22	22	Arsenal	2	77	80	77	80
23	23	Bayer 04 Leverkusen	3	83	79	78	79
24	24	Ajax	6	82	77	80	79
25	25	Real Betis Balompié	4	80	79	78	79
26	26	West Ham United	2	80	79	77	79
27	27	Latium	5	79	81	78	79
28	28	Sporting CP	7	79	80	78	79
29	29	Roma	5	79	79	78	79
30	30	Athletic Club de Bilbao	4	78	77	79	79

And same goes for dataset2 (fifa21)

▲	X	Club	League	Attack	Midfield	Defence	Overall
1	1	Liverpool	2	86	83	80	85
2	2	Manchester City	2	85	86	83	85
3	3	Real Madrid	4	84	86	84	85
4	4	FC Bayern München	3	92	85	82	84
5	5	Paris Saint-Germain	1	87	82	82	84
6	6	Juventus	5	87	81	83	83
7	7	Atlético de Madrid	4	85	82	83	83
8	8	FC Barcelona	4	84	82	81	83
9	9	Tottenham Hotspur	2	89	82	81	82
10	10	Manchester United	2	83	82	82	82
11	11	Inter	5	83	81	83	82
12	12	Chelsea	2	82	84	82	82
13	13	Borussia Dortmund	3	87	81	81	81
14	14	Sevilla FC	4	80	81	82	81
15	15	Arsenal	2	83	79	79	80
16	16	Napoli	5	82	80	80	80
17	17	Lazio	5	82	80	79	80
18	18	Villarreal CF	4	82	79	79	80
19	19	Leicester City	2	80	81	79	80
20	20	RB Leipzig	3	78	81	80	80
21	21	Milan	5	84	78	80	79
22	22	Atalanta	5	82	79	78	79
23	23	Real Sociedad	4	81	81	77	79
24	24	Everton	2	81	79	80	79
25	25	Athletic Club	4	80	78	79	79
26	26	Borussia Mönchengladbach	3	78	79	80	79
27	27	Bayer 04 Leverkusen	3	78	78	78	79
28	28	Wolverhampton Wanderers	2	77	81	78	79
29	29	SL Benfica	7	77	79	78	79
30	30	Ajax	6	80	77	76	78

Descriptive Statistics:

Descriptive statistics applies the concepts, measures, and terms that are used to describe the basic features of the samples in a study. Here these procedures are followed for 2 of our datasets (fifa22, fifa21) as an approximation result.

First the quantitative analysis then the simple graphics. For the dataset (fifa22) all statistics are done separately with specific code on the other hand for fifa21 dataset the summaries () function used for direct analysis.

Dataset 1 (fifa22)

```
> mean(fifa22$Overall)
[1] 81.2
> median(fifa22$Overall)
[1] 80.5
> uv <- unique(fifa22$Overall)
> mode(fifa22$Overall)
[1] "numeric"
> max(fifa22$Overall)-min(fifa22$Overall)
[1] 6
> var(fifa22$Overall)
[1] 4.372414
> sd(fifa22$Overall)
[1] 2.091032
> quantile(fifa22$Overall)
 0%   25%   50%   75%  100%
79.00 79.25 80.50 83.00 85.00
> IQR(fifa22$Overall)
[1] 3.75
> |
```

Dataset 2 (fifa21)

```
> summary(fifa21)
      X      Club      League      Attack      Midfield      Defence
Min.   : 1.00  Length:30  Length:30  Min.   :77.00  Min.   :77.0  Min.   :76.0
1st Qu.: 8.25  Class :character  Class :character  1st Qu.:80.00  1st Qu.:79.0  1st Qu.:79.0
Median :15.50  Mode  :character  Mode  :character  Median :82.00  Median :81.0  Median :80.0
Mean   :15.50                                Mean   :82.63  Mean   :80.9  Mean   :80.3
3rd Qu.:22.75                                3rd Qu.:84.75  3rd Qu.:82.0  3rd Qu.:82.0
Max.   :30.00                                Max.   :92.00  Max.   :86.0  Max.   :84.0
Overall
Min.   :78.00
1st Qu.:79.00
Median :80.00
Mean   :81.03
3rd Qu.:82.75
Max.   :85.00
> var(fifa21$Overall)
[1] 4.585057
> sd (fifa21$Overall)
[1] 2.141275
> quantile(fifa21$Overall)
 0%   25%   50%   75%  100%
78.00 79.00 80.00 82.75 85.00
> IQR (fifa21$Overall)
[1] 3.75
> |
```

Mean: The mean value for fifa22 overall is 81.2 where in the dataset 2 the fifa21 mean values of attack, Mid and defense is respectively 82.63, 80.9 and 80.3.

Median: using R median () function to determine the overall median value of fifa22 is 80.5 where at fifa21 82, 81 and 80 are the median value for the attributes.

Mode: R does not have a built-in function to calculate the mode. So, using R with a user-defined function to find the modes of the values,
For fifa21 the mode is 79 and for fifa22 the value is also 79

```
> my_mode(x1)  > my_mode(x2)
[1] 79         [1] 79
```

Fifa21 dataset Code:

```
x1 <- c(fifa21$Overall)
my_mode <- function(x) {
  unique_x <- unique(x)
  tabulate_x <- tabulate(match(x, unique_x))
  unique_x[tabulate_x == max(tabulate_x)]
}
my_mode(x1)
```

Fifa22 dataset Code:

```
x2 <- c(fifa22$Overall)
my_mode <- function(x) {
  unique_x <- unique(x)
  tabulate_x <- tabulate(match(x, unique_x))
  unique_x[tabulate_x == max(tabulate_x)]
}
my_mode(x2)
```

Range: Use the R min() and max() functions to find the range of the values, here for fifa22 the range is 6 and for fifa21 dataset it is 85-78= 7.

Variance: now for fifa22 the variance is 4.372414 and for fifa21 the variance is 4.585057.

Standard Deviation: The standard deviation is simply the square root of the variance. Standard deviation measures how far a 'typical' observation is from the average of the data. Use the R `sd()` function to find the sample standard deviation of `fifa22` is 2.091032 and for `fifa21` it is 2.141575.

Quartiles: Quartiles are values that separate the data into four equal parts.

```
> quantile(fifa21$Overall)    > quantile(fifa22$Overall)
 0%   25%   50%   75%  100%   0%   25%   50%   75%  100%
78.00 79.00 80.00 82.75 85.00 79.00 79.25 80.50 83.00 85.00
```

Interquartile Range: Interquartile range is the difference between the first and third quartiles (Q1 and Q3).

`Fifa22 IQR()` = 3.75 and `fifa21 IQR()` = 3.75

Data Distributor: A histogram is a widely used graph to show the distribution of quantitative (numerical) data. It shows the frequency of rating values in the data, usually in intervals of values. Frequency is the number of times that rating value appeared in the data. We try to display all the attributes data using histogram and summaries the frequency rate.

For `fifa22` dataset

Code:

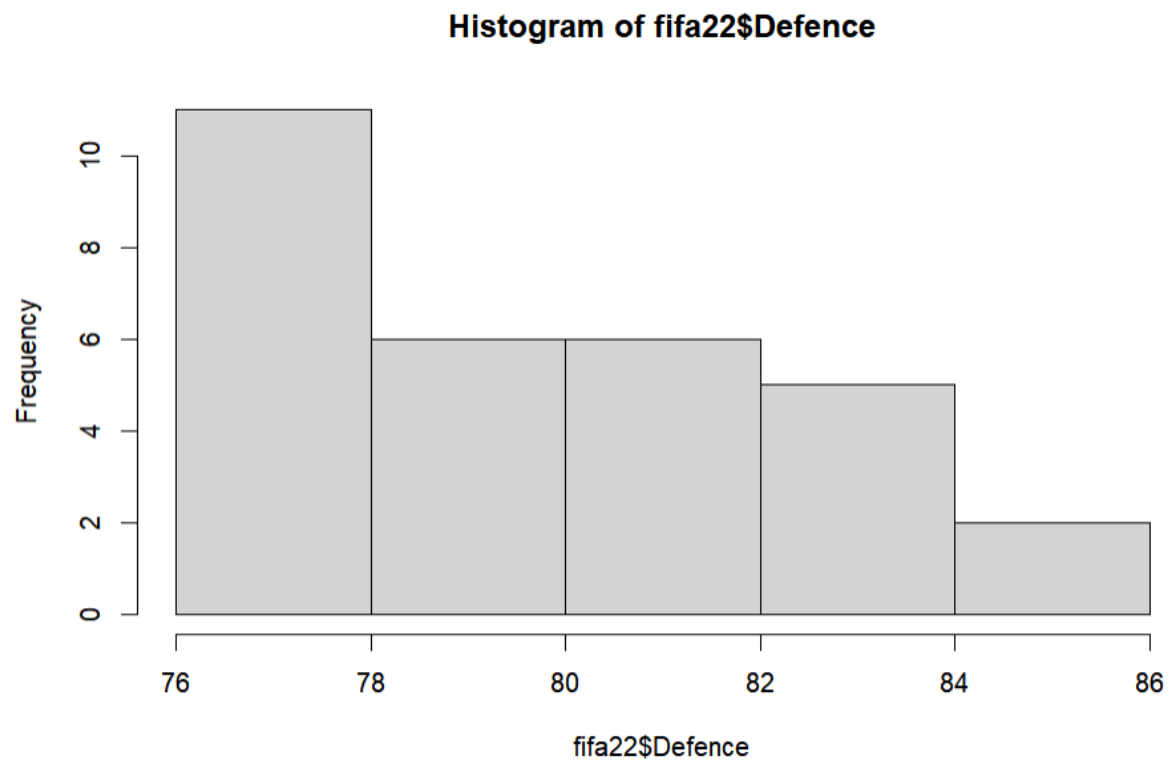
#Histogram

`hist(fifa22$Overall)`

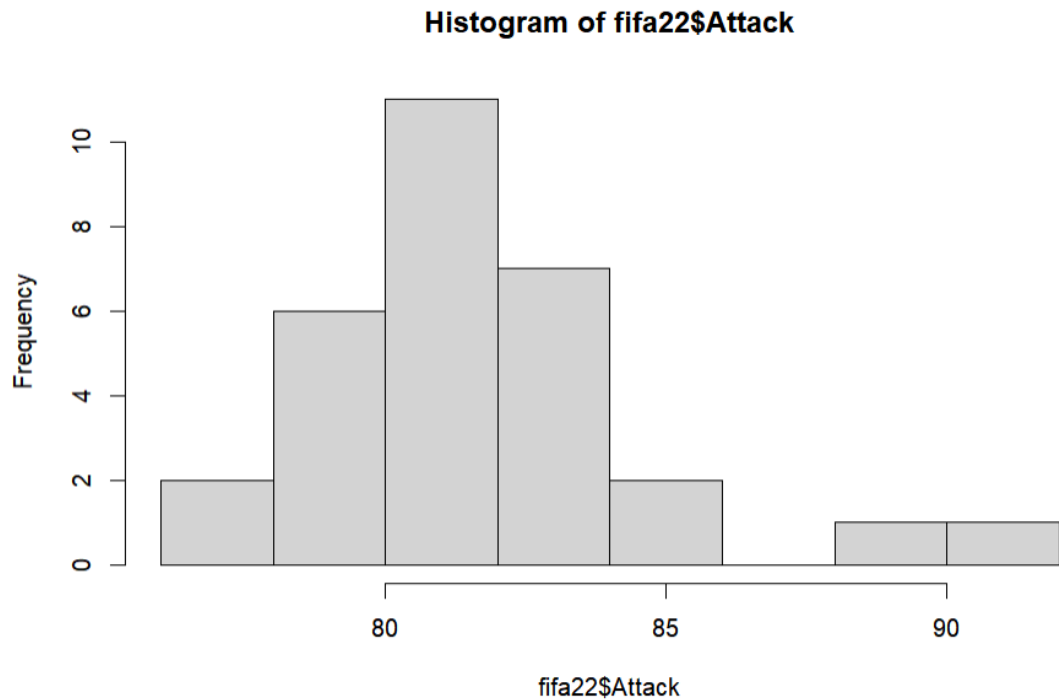
`hist(fifa22$Attack)`

`hist(fifa22$Defence)`

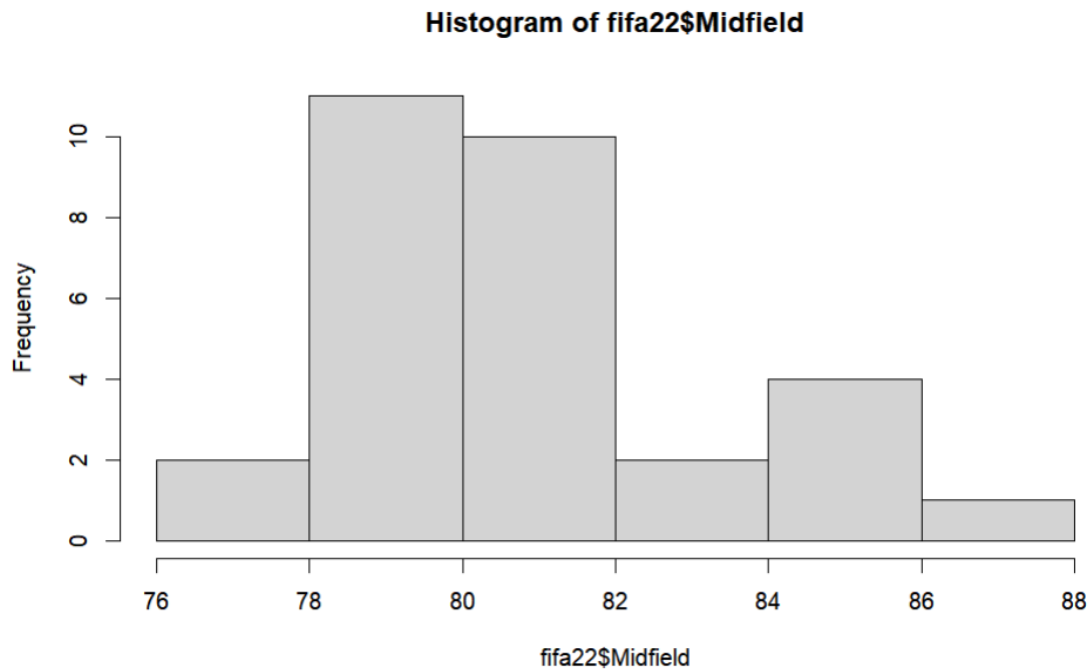
`hist(fifa22$Midfield)`



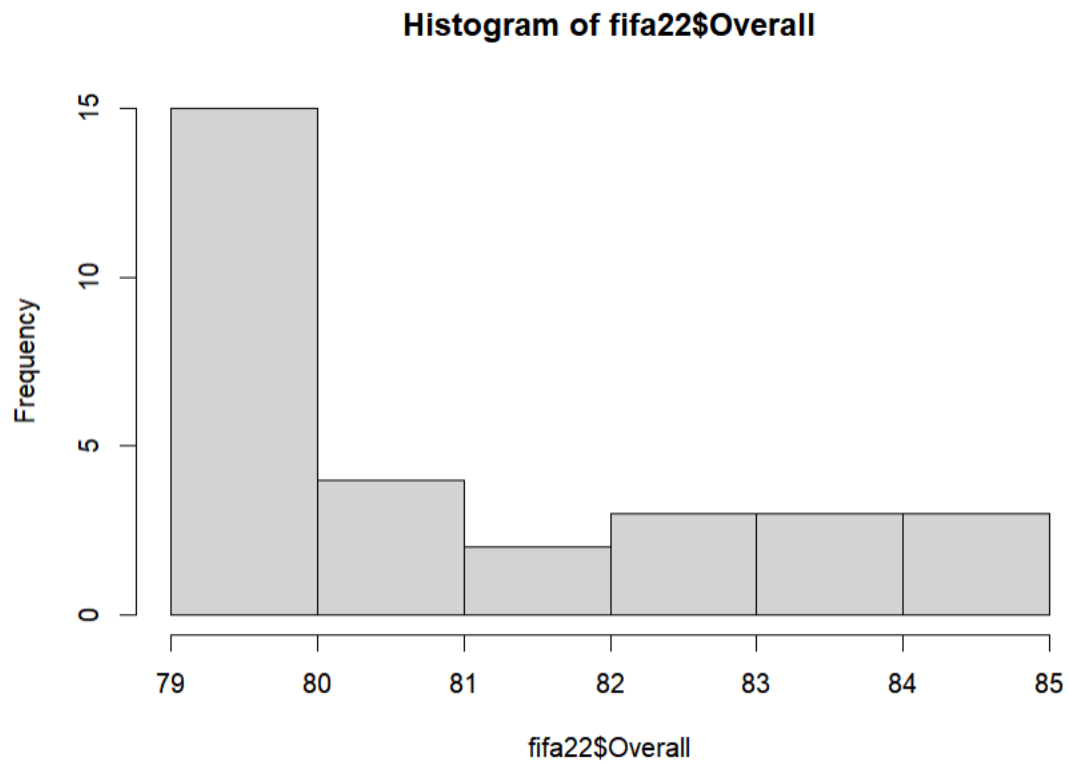
fifa22\$defence the most teams have defense rating around 76-78 which is very low in case of defense. Defenders are not very highly rated as the attacker.



fifa22\$Attack graph we can notice that most of the team's attack rating is around 80-82. Very few club possess high class striker.



Where The midfielder apart from one team most of them are having 78 – 82 rating status.



. So the overall view is that most teams are around 79 to 80. Only top teams have the best overall ratings no doubts.

For fifa21 dataset

Code:

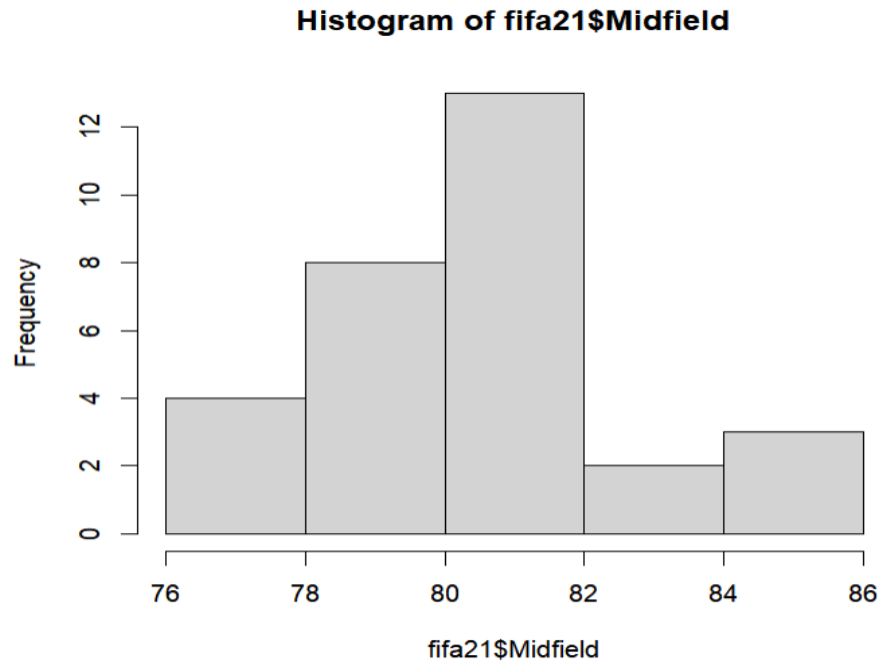
```
#Histogram
```

```
hist(fifa21$Overall)
```

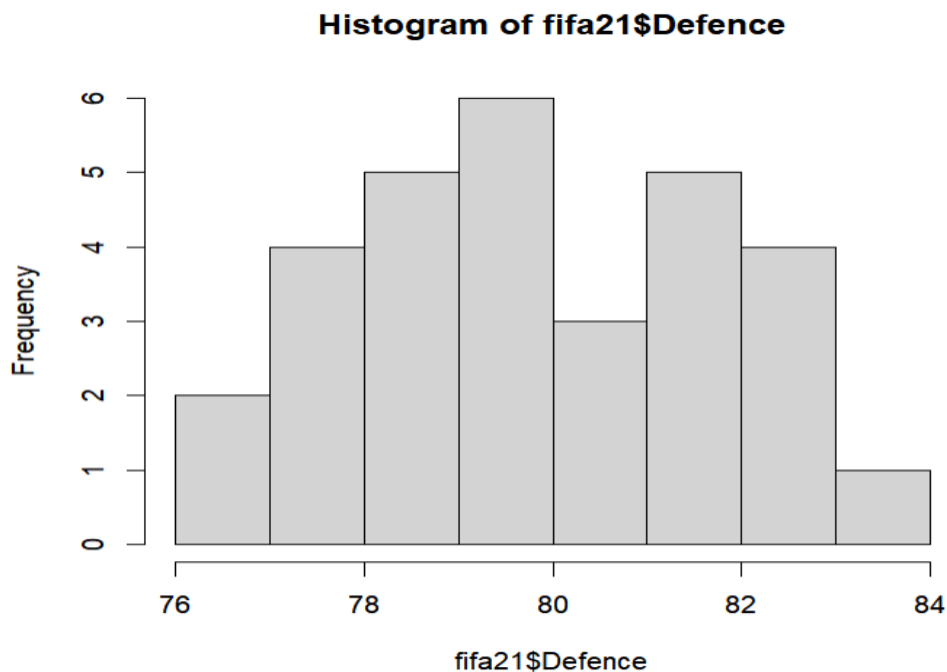
```
hist(fifa21$Attack)
```

```
hist(fifa21$Defence)
```

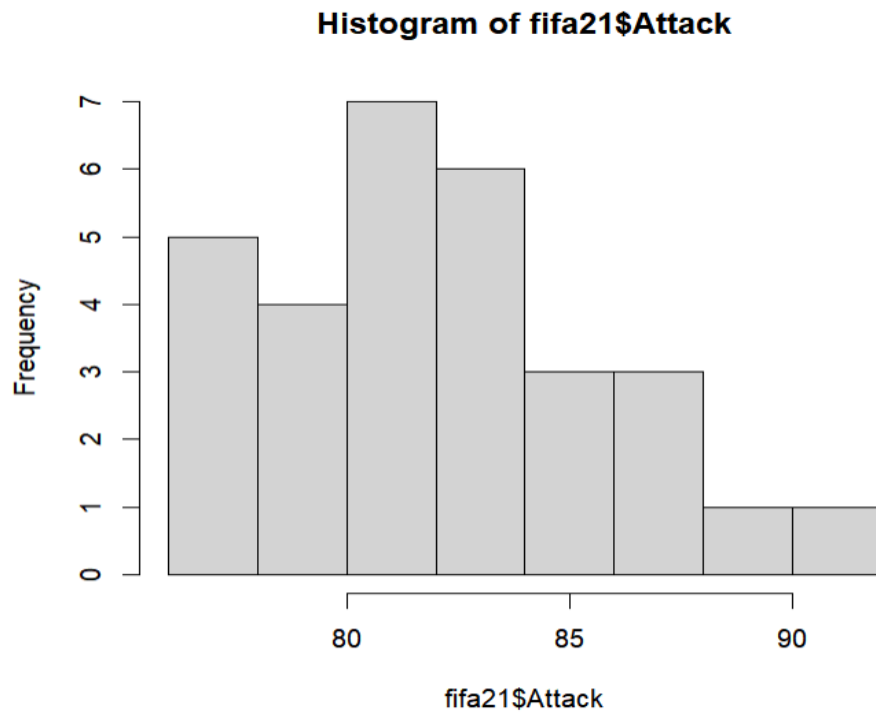
```
hist(fifa21$Midfield)
```

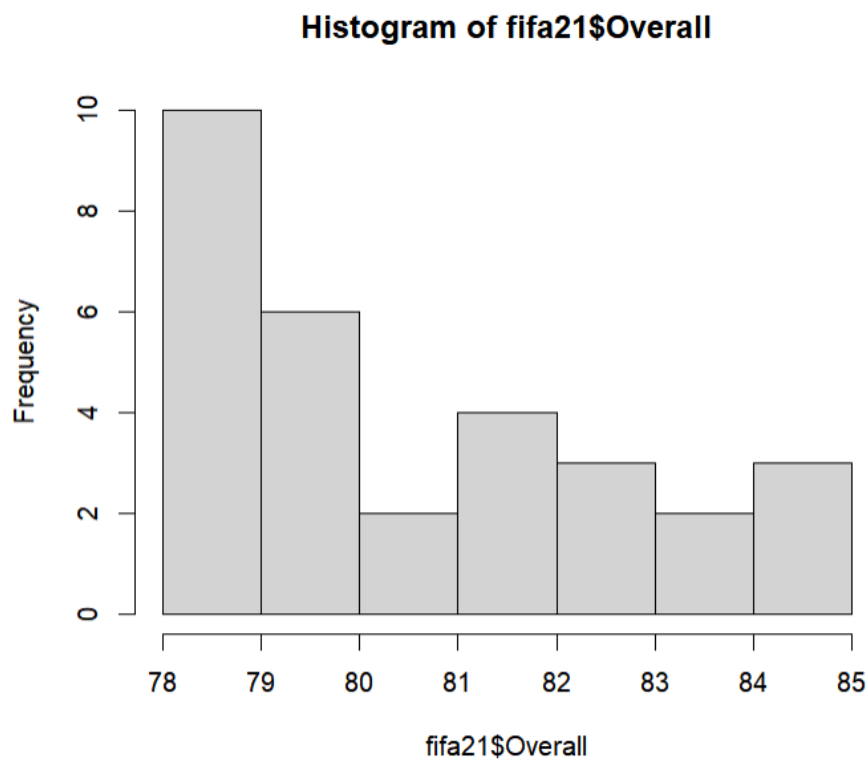
The midfielder apart from one team most of them are having 80 – 82 rating status. In 21 clubs are mainly focused on strengthening midfield for their tactics purpose.



Where fifa21\$defence the most teams have defense rating around 80 and the defense is very balanced according orderly. Though the defenders are not very high rated as the attacker.



fifa21\$Attack graph we can notice that most of the team's attack rating is around 82-85. Which is higher than the previous dataset.



So, the overall view is that most teams are around 79 to 80. Only top teams have the best overall ratings no doubt.

Data Visualization

Creating a graph plot using ggplot2.

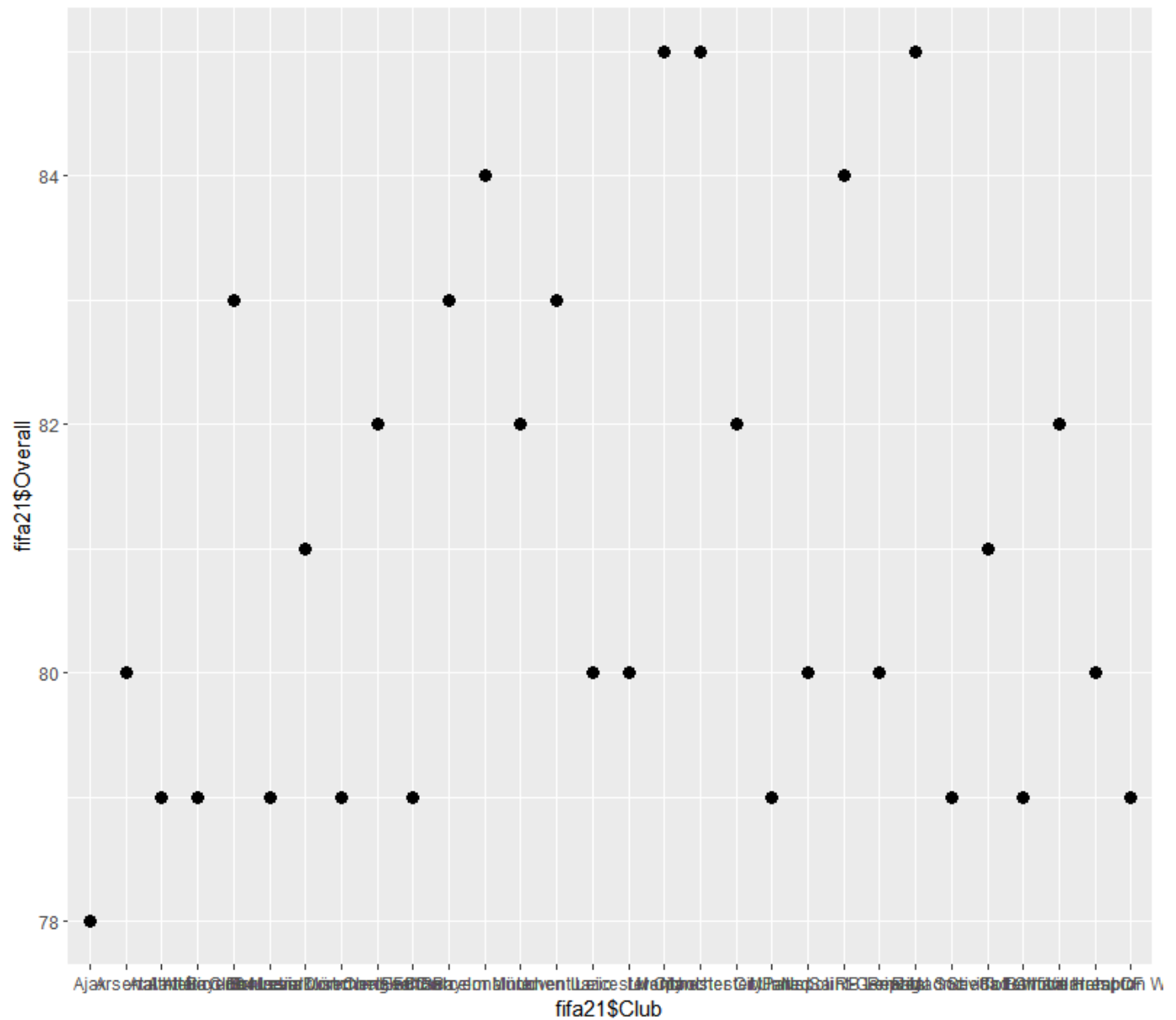
Dataset 1 (FIFA22)

Code:

```
#ggplot visualize
data22=data.frame(fifa22$Overall)
str(data22)
#install.packages("ggplot2")
library(ggplot2)
ggplot(data22, aes(fifa22$Club, fifa22$Overall))+
  geom_point(size = 3)
```



```
library(ggplot2)
ggplot(data21, aes(fifa21$Club, fifa21$Overall))+
  geom_point(size = 3)
```



Here in this scatter diagram, all teams are almost balanced. Though most of the clubs' ratings got reduced from the previous diagram which is around 81-79. And also, clubs between 82-84 have increased in great number. It means the clubs

have improved themselves in the competition. Also, a huge number of talented players may have joined the clubs. The 3 top teams with 85 rating like previous.

Discussion & Conclusion

After all the necessary process finally, our data set complete and ready to be used for further analysis. From the descriptive statistics and the graph plot we can say that in 2021 the club were balanced in every position aspect because all the positions ratings are well balanced but in the 2022 clubs are focusing more on defense rather than attacking. There is also a reason that a handful of good players are retiring so the defense player gets more ratings. This difficult thing is easily done by using Rstudio with R language thanks to the unlimited library makes our work easy. I also use Selectorgadget for scrapping data table from the website.