FPL points predictor Exploratory Data Analysis

This is a file for conductiong explorative data analysis of the premier league players' dataset

Importing the datasets

```
season17 <- read.csv("~/DSI-SRP1/season17.csv", encoding="UTF-8")
season18 <- read.csv("~/DSI-SRP1/season18.csv", encoding="UTF-8")
season19 <- read.csv("~/DSI-SRP1/season19.csv", encoding="UTF-8")</pre>
```

Loading the tidyverse package

Calculating correlation coefficients between goal scored and total FPL points

Correlation between goal scored and total fpl points for 2016/17: 0.700148307138786

```
message("Correlation between goal scored and total fpl points for 2017/18: ",
     with(season18, cor(goals_scored,total_points)))
```

Correlation between goal scored and total fpl points for 2017/18: 0.679586851190284

Correlation between goal scored and total fpl points for 2018/19: 0.703249889002981

Changing the datatype of the position_index to factor

```
season17$position_index <- as_factor(season17$position_index)
season18$position_index <- as_factor(season18$position_index)
season19$position_index <- as_factor(season19$position_index)</pre>
```

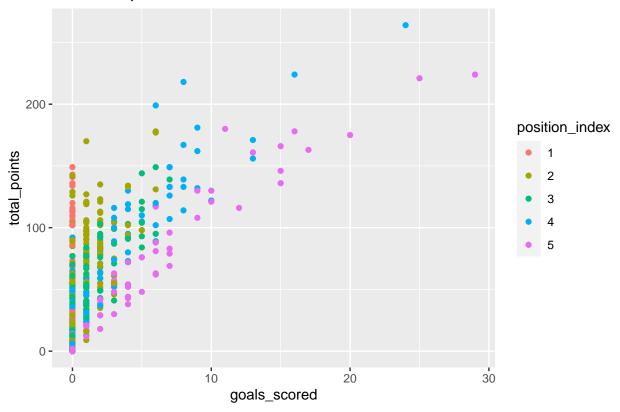
Creating a funtion to calculate a new metric called total fpl points to game played

```
fpl_to_gaming <- function(df) {
    df %>%
        mutate(fpl_to_game = (total_points/minutes.played)*90)
}
season17 <- fpl_to_gaming(season17)
season18 <- fpl_to_gaming(season18)
season19 <- fpl_to_gaming(season19)</pre>
```

Scatter plots graph to highlight how the goals scored vary with fpl points

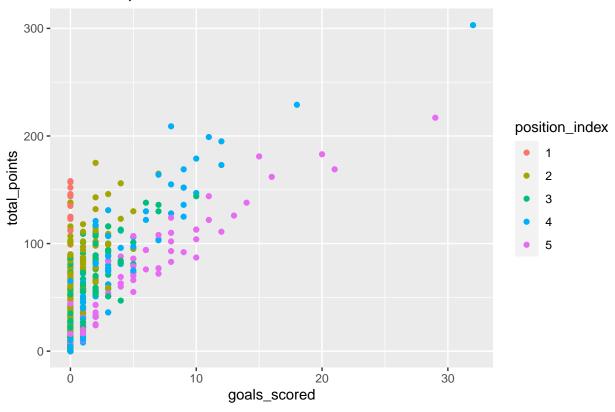
```
ggplot(season17, aes(goals_scored, total_points)) +
  geom_point(aes(color = position_index)) +
  labs(title = "Total FPL points vs Goal Scored in 2016/17")
```

Total FPL points vs Goal Scored in 2016/17



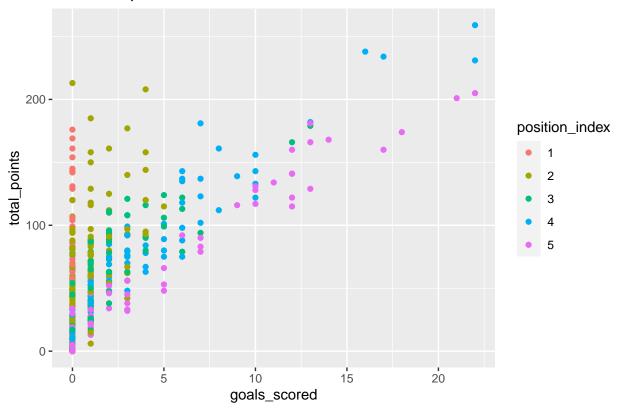
```
ggplot(season18, aes(goals_scored, total_points)) +
  geom_point(aes(color = position_index)) +
  labs(title = "Total FPL points vs Goal Scored in 2017/18")
```

Total FPL points vs Goal Scored in 2017/18



```
ggplot(season19, aes(goals_scored, total_points)) +
  geom_point(aes(color = position_index)) +
  labs(title = "Total FPL points vs Goal Scored in 2018/19")
```

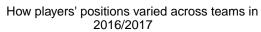
Total FPL points vs Goal Scored in 2018/19

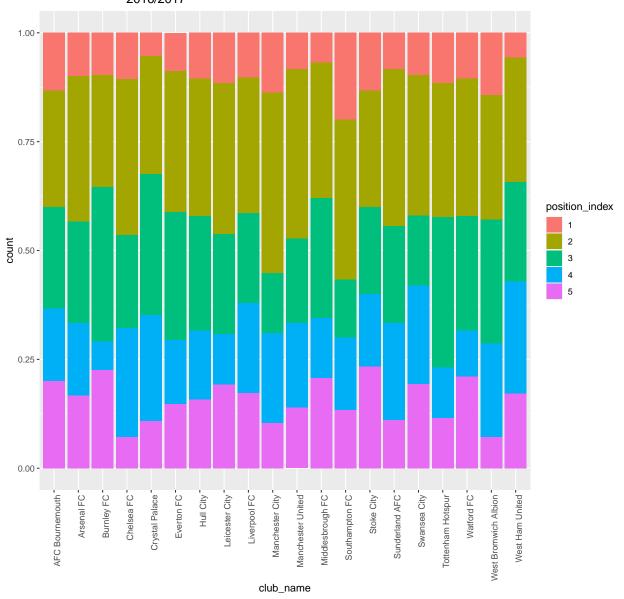


Bar graphs to show how the number of players vary per team with regards to position

Plotting the bar charts

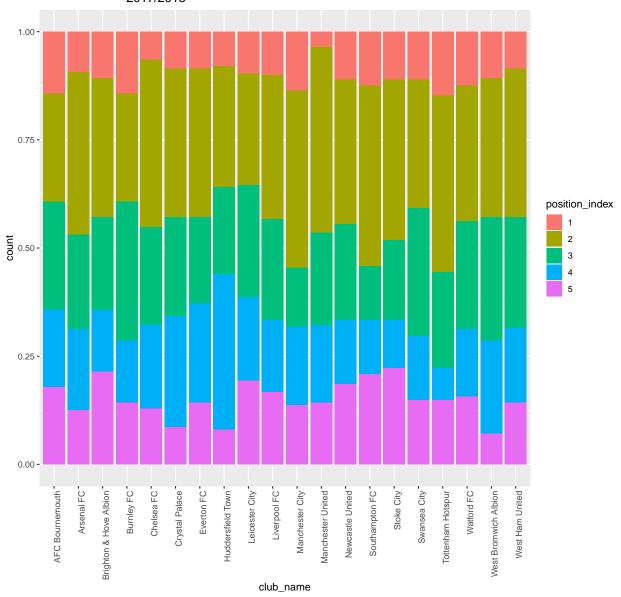
```
position_bar(season17)
```



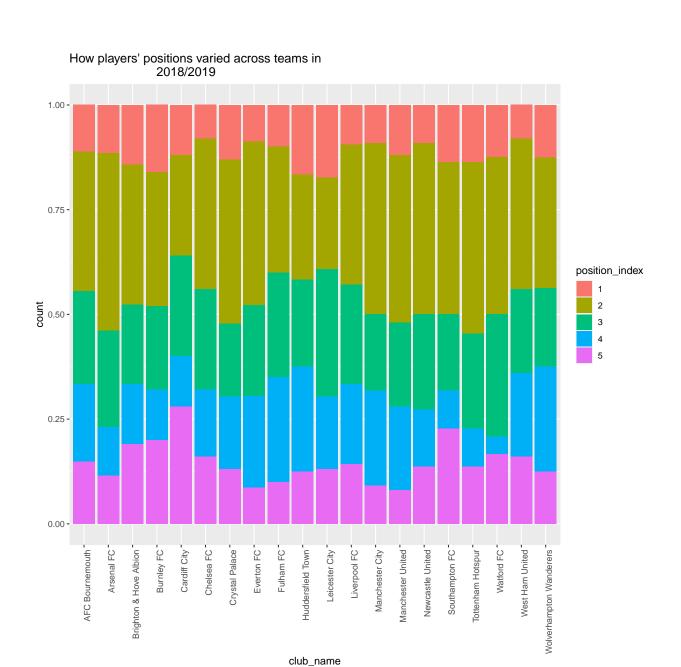


position_bar(season18)





position_bar(season19)

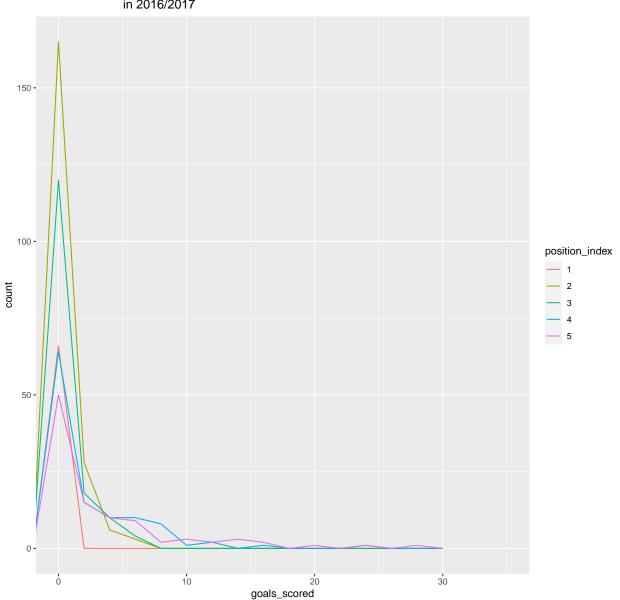


Creating frequency plot for goal scored and how it varies for the players' position

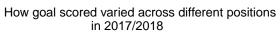
Plotting the graphs

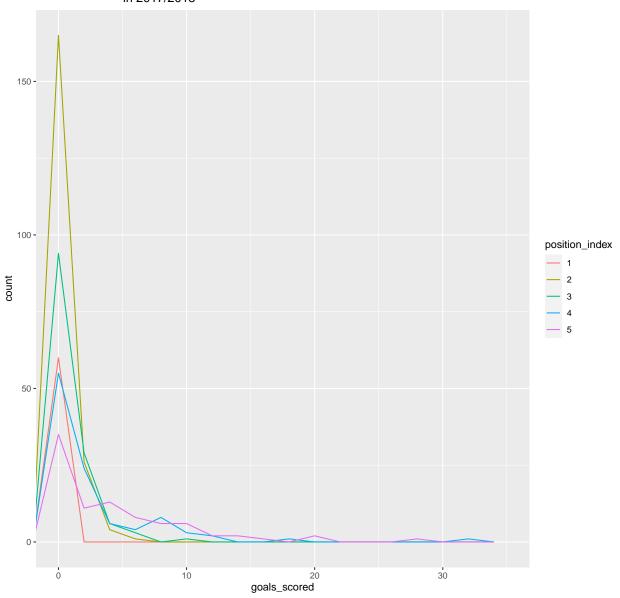
goal_freqplot(season17)

How goal scored varied across different positions in 2016/2017



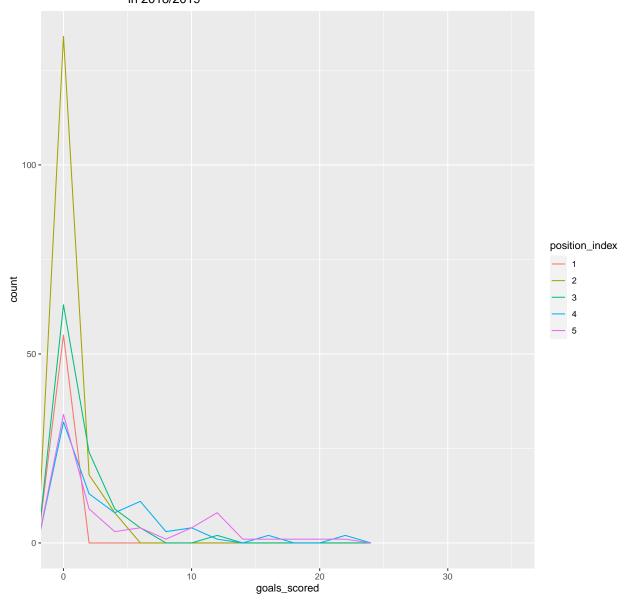
goal_freqplot(season18)





goal_freqplot(season19)

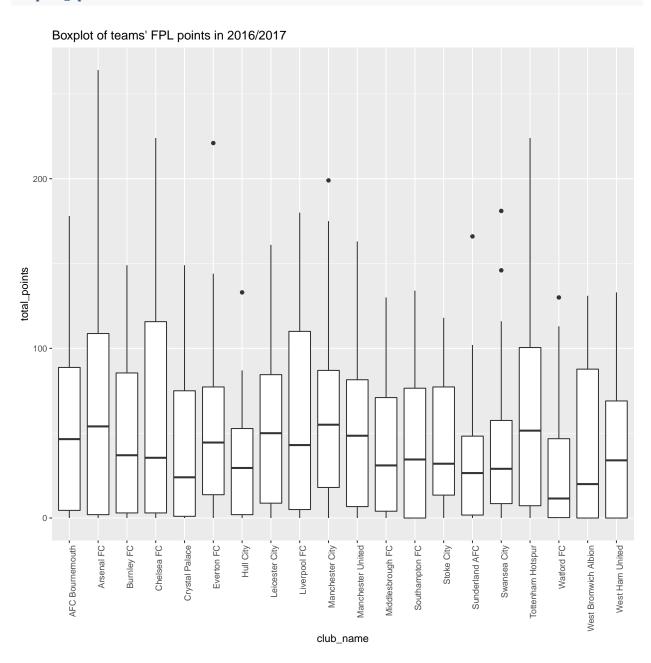
How goal scored varied across different positions in 2018/2019



Creating boxplots to show the distiribution of the teams' fpl points according to their players

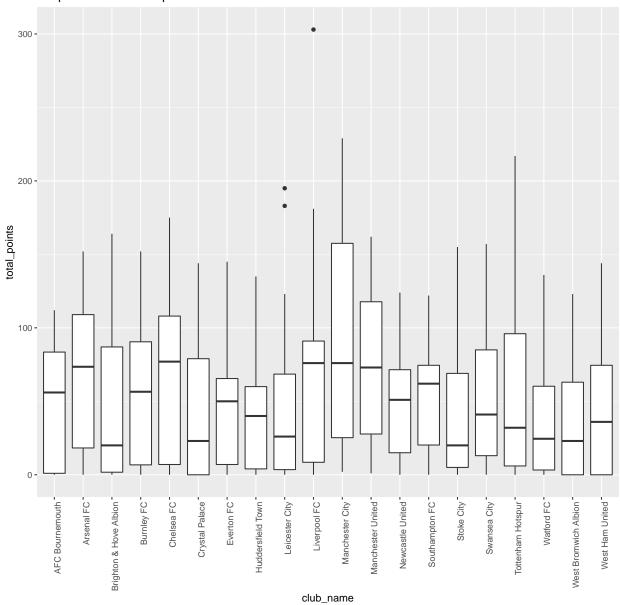
Plotting the graphs

boxplot_fpl(season17)



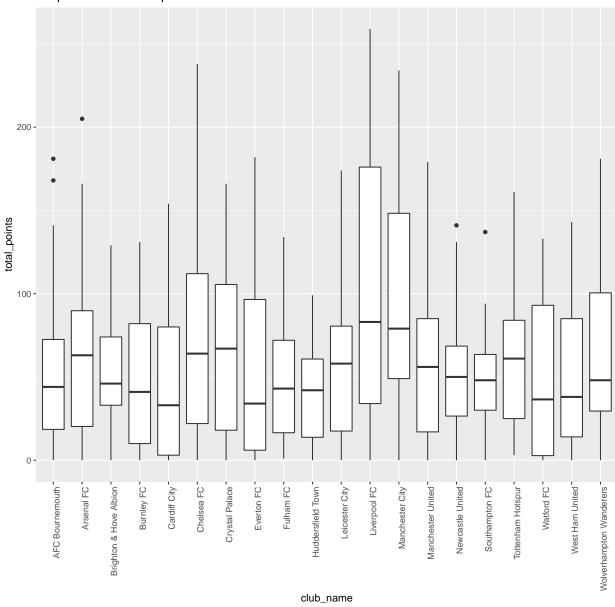
boxplot_fpl(season18)





boxplot_fpl(season19)

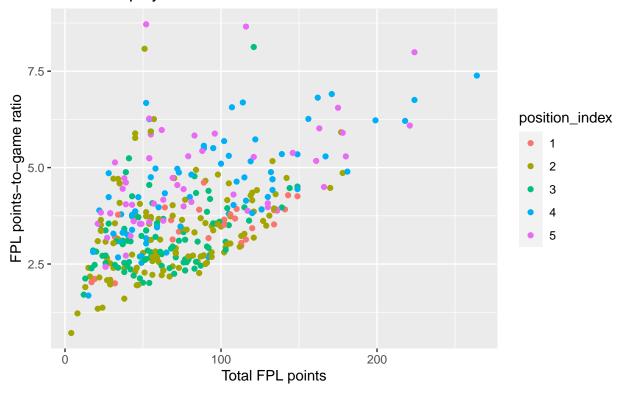




Scatter plots showing the FPL points to game ratio vs the total FPL points for each season

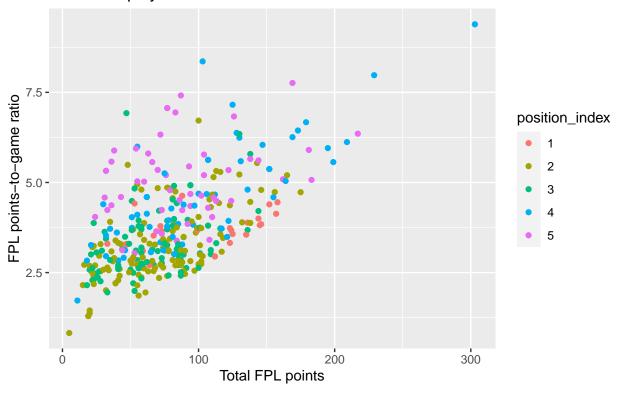
```
season17 %>%
  filter(minutes.played >= 500) %>%
  ggplot() +
  geom_point(aes(total_points, fpl_to_game, color = position_index), na.rm = TRUE) +
  labs(x = "Total FPL points", y = "FPL points-to-game ratio",
  title = title1)
```

FPL points to game ratio vs Total FPL points for players who have played more than 500 minutes in 2016/17

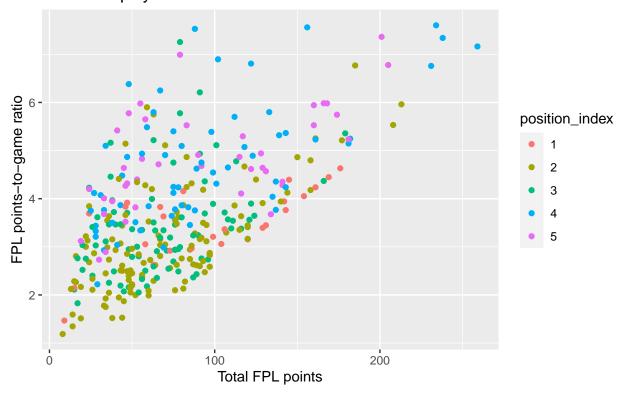


```
season18 %>%
  filter(minutes.played >= 500) %>%
  ggplot() +
  geom_point(aes(total_points, fpl_to_game, color = position_index), na.rm = TRUE) +
  labs(x = "Total FPL points", y = "FPL points-to-game ratio",
  title = title2)
```

FPL points to game ratio vs Total FPL points for players who have played more than 500 minutes in 2017/18



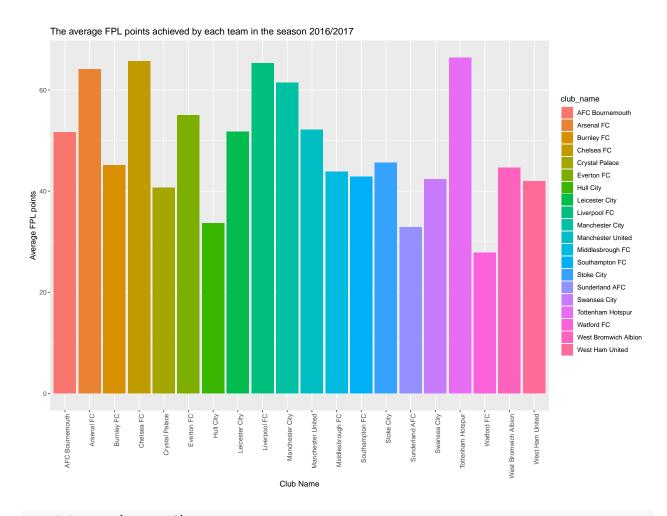
FPL points to game ratio vs Total FPL points for players who have played more than 500 minutes in 2018/19



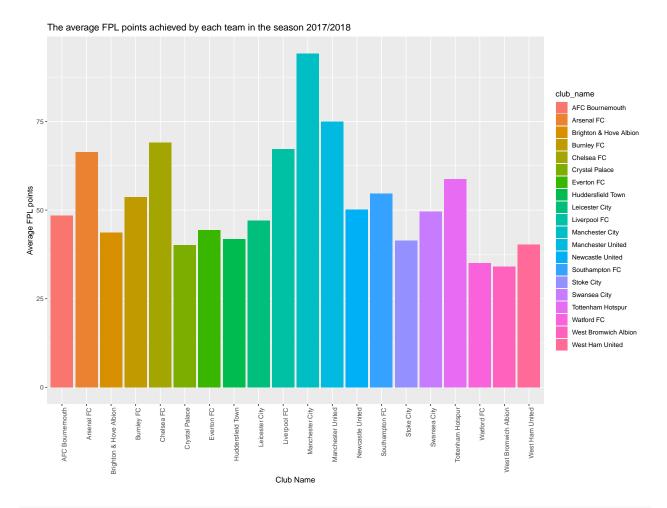
Calculating the average fpl points per team and building a bar chart to show the average fpl points per team

```
avg_fpl_point <- function(df) {
    df %>%
        group_by(club_name) %>%
        summarize(average = mean(total_points, na.rm = TRUE)) %>%
        ggplot() +
        geom_bar(aes(x = club_name, y = average, fill = club_name), stat = "identity") +
        theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
        labs(x = "Club Name", y = "Average FPL points",
        title = sprintf("The average FPL points achieved by each team in the season %s",
        unique(df$season)))
}
```

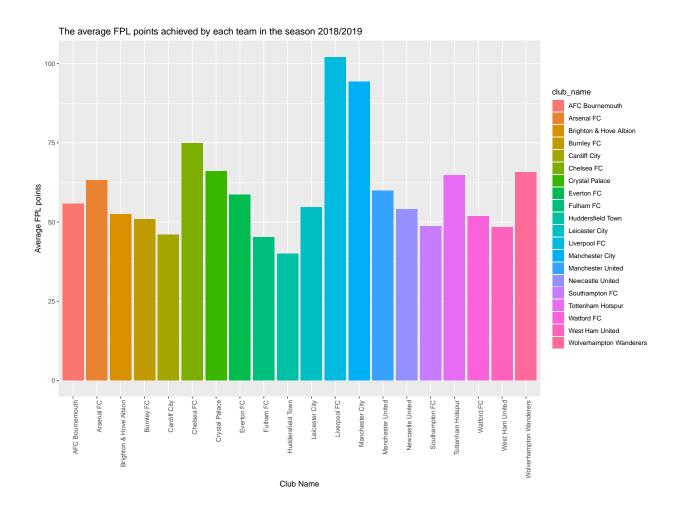
```
avg_fpl_point(season17)
```



avg_fpl_point(season18)



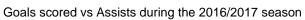
avg_fpl_point(season19)

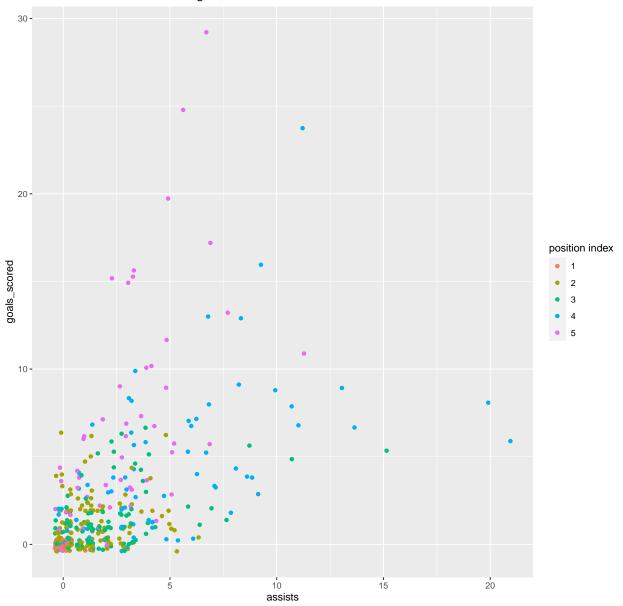


Plotting graphs of goals scored vs assists made for every season

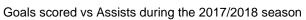
goals_vs_assist(season17)

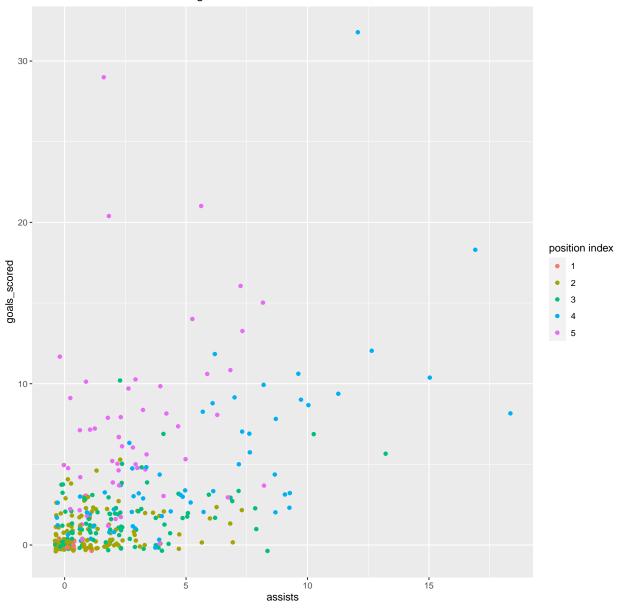
```
goals_vs_assist <- function(df) {
    df %>%
        filter(minutes.played >= 500) %>%
        ggplot(aes(assists, goals_scored)) +
        geom_point(aes(color = position_index), position = "jitter") +
        labs(title = sprintf("Goals scored vs Assists during the %s season",
        unique(df$season)), color = "position index")
}
```





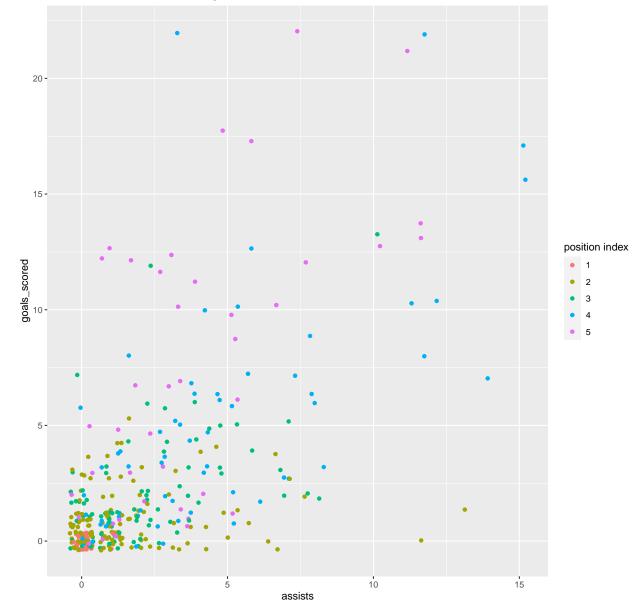
goals_vs_assist(season18)





goals_vs_assist(season19)





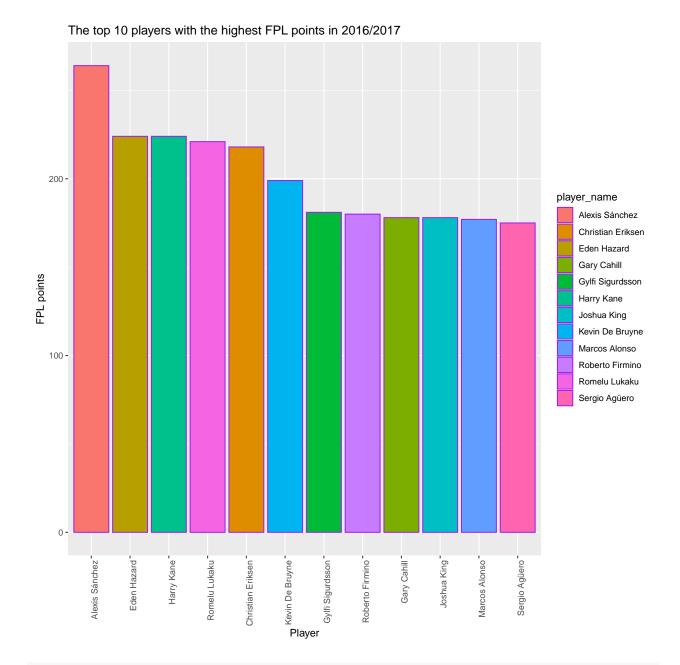
Ranking the players with the top 10 fpl points per season

```
fpl_point_rank <- function(df) {
   df %>%
    mutate(ranking = dense_rank(desc(total_points))) %>%
    filter(ranking <= 10) %>%
        arrange(desc(total_points)) %>%
        select(ranking, player_name, total_points) %>%
        print() %>%
        ggplot(aes(reorder(player_name, -total_points), total_points, fill = player_name)) +
        geom_bar(stat = "identity", color = "purple") +
        theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
```

```
labs(title = sprintf("The top 10 players with the highest FPL points in %s",
  unique(df$season)), x = "Player", y = "FPL points")
}
```

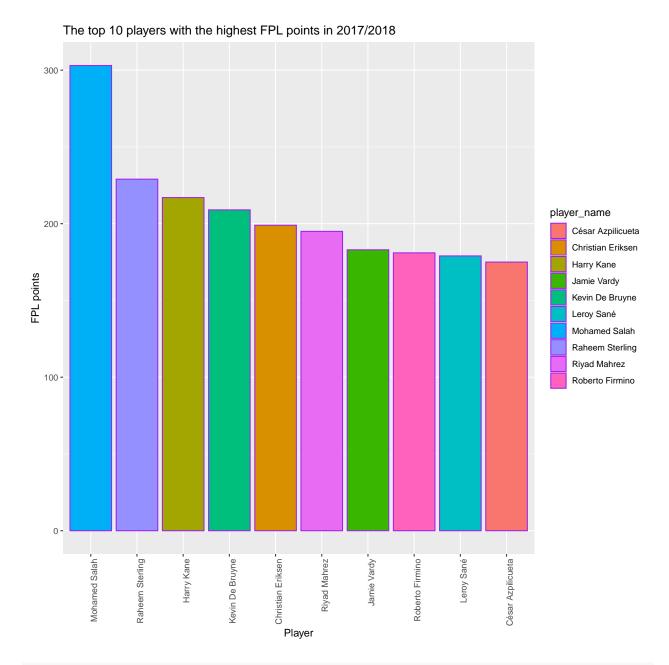
fpl_point_rank(season17)

##		ranking	player_name	total_points
##	1	1	Alexis Sánchez	264
##	2	2	Eden Hazard	224
##	3	2	Harry Kane	224
##	4	3	Romelu Lukaku	221
##	5	4	Christian Eriksen	218
##	6	5	Kevin De Bruyne	199
##	7	6	Gylfi Sigurdsson	181
##	8	7	Roberto Firmino	180
##	9	8	Gary Cahill	178
##	10	8	Joshua King	178
##	11	9	Marcos Alonso	177
##	12	10	Sergio Agüero	175



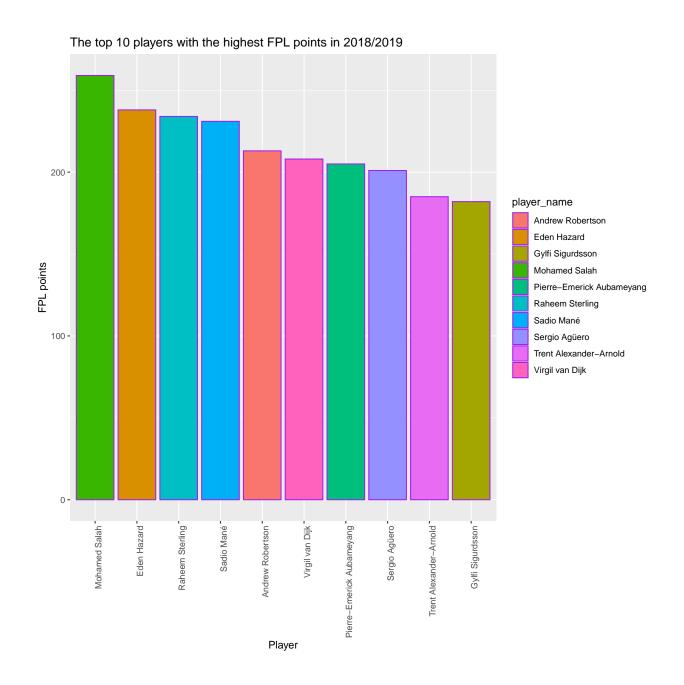
fpl_point_rank(season18)

##		ranking	player_name	total_points
##	1	1	Mohamed Salah	303
##	2	2	Raheem Sterling	229
##	3	3	Harry Kane	217
##	4	4	Kevin De Bruyne	209
##	5	5	Christian Eriksen	199
##	6	6	Riyad Mahrez	195
##	7	7	Jamie Vardy	183
##	8	8	Roberto Firmino	181
##	9	9	Leroy Sané	179
##	10	10	César Azpilicueta	175



fpl_point_rank(season19)

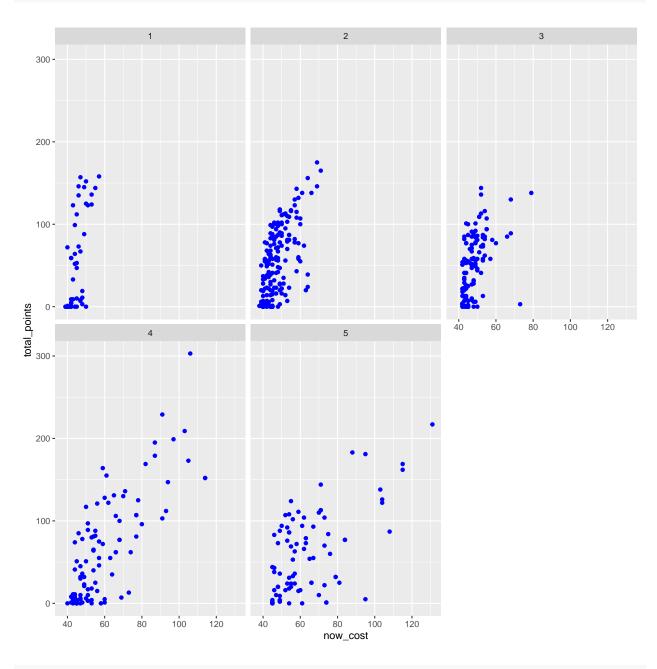
##		ranking	player_name	total_points
##	1	1	Mohamed Salah	259
##	2	2	Eden Hazard	238
##	3	3	Raheem Sterling	234
##	4	4	Sadio Mané	231
##	5	5	Andrew Robertson	213
##	6	6	Virgil van Dijk	208
##	7	7	Pierre-Emerick Aubameyang	205
##	8	8	Sergio Agüero	201
##	9	9	Trent Alexander-Arnold	185
##	10	10	Gylfi Sigurdsson	182



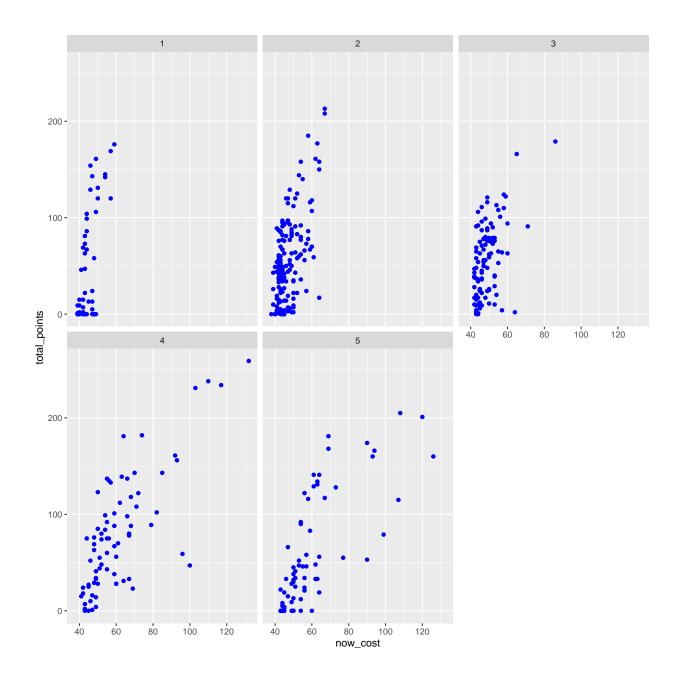
Plotting the total FPL points vs players' cost for every season

```
library(modelr)
options(na.action = na.warn)
value <- function(df) {
    ggplot(df, aes(x = now_cost)) +
    geom_point(aes(y = total_points), color = "blue") +
    facet_wrap(~position_index)
}</pre>
```

value(season18)



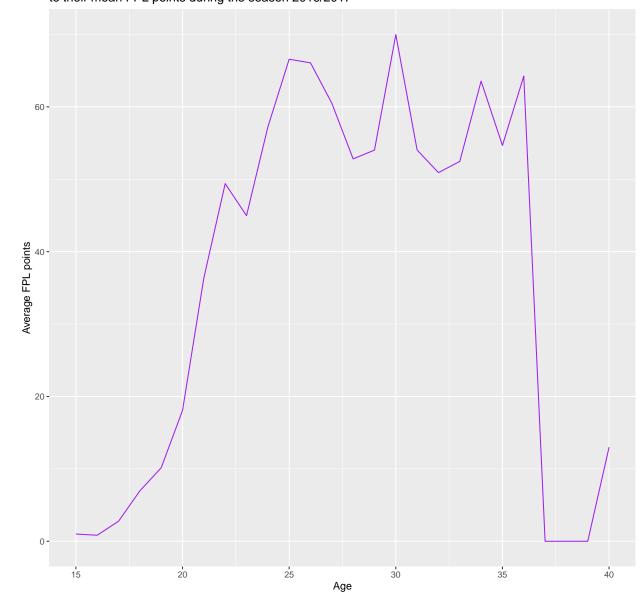
value(season19)



Visualizing how age has an impact on FPL points

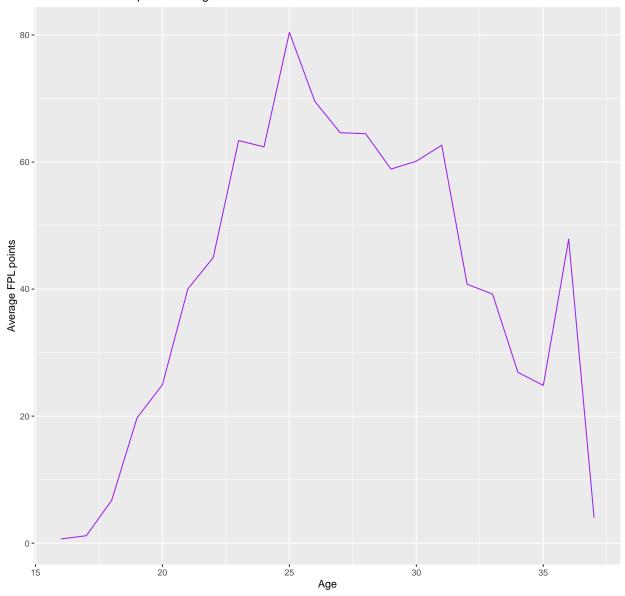
```
age_freq <- function(df) {
    df %>%
        group_by(age) %>%
        summarize(mean = mean(total_points, na.rm = TRUE)) %>%
        ggplot(aes(x = age, y = mean)) + geom_freqpoly(stat = "identity", color = "purple") +
labs(title = sprintf(title, unique(df$season)), x = "Age",
        y = "Average FPL points")
}
```

A frequency polygon graph of the ages of Premier League players with respect to their mean FPL points during the season 2016/2017



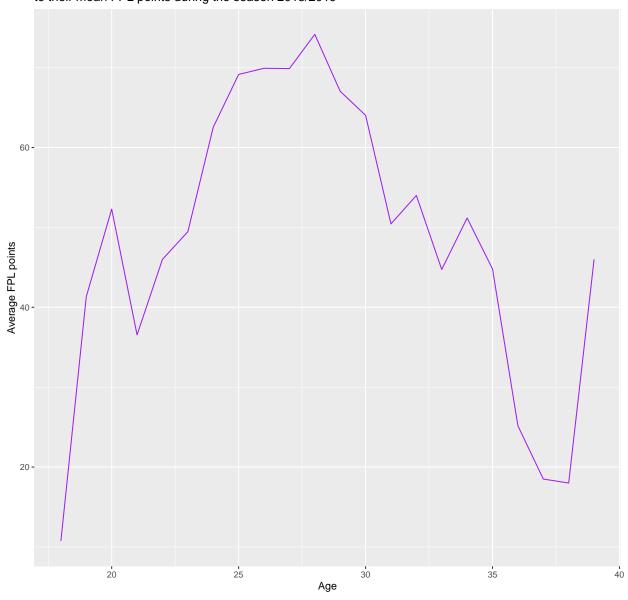
age_freq(season18)

A frequency polygon graph of the ages of Premier League players with respect to their mean FPL points during the season 2017/2018 $\,$



age_freq(season19)

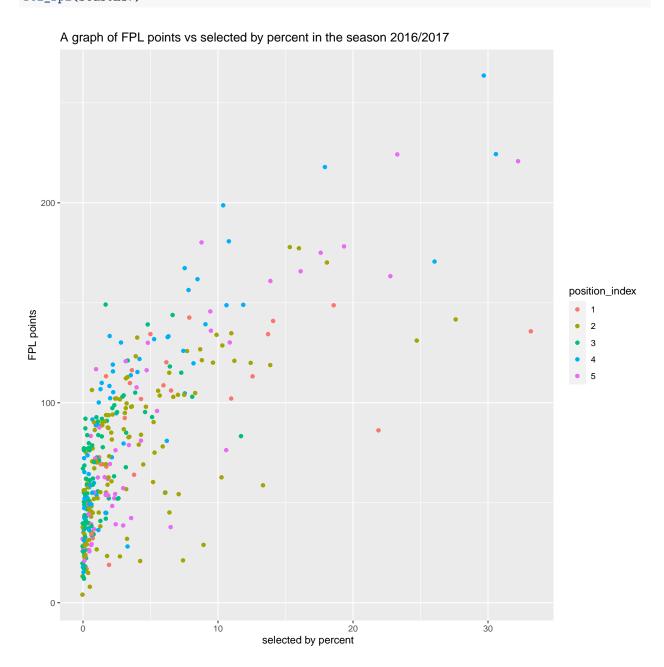
A frequency polygon graph of the ages of Premier League players with respect to their mean FPL points during the season 2018/2019



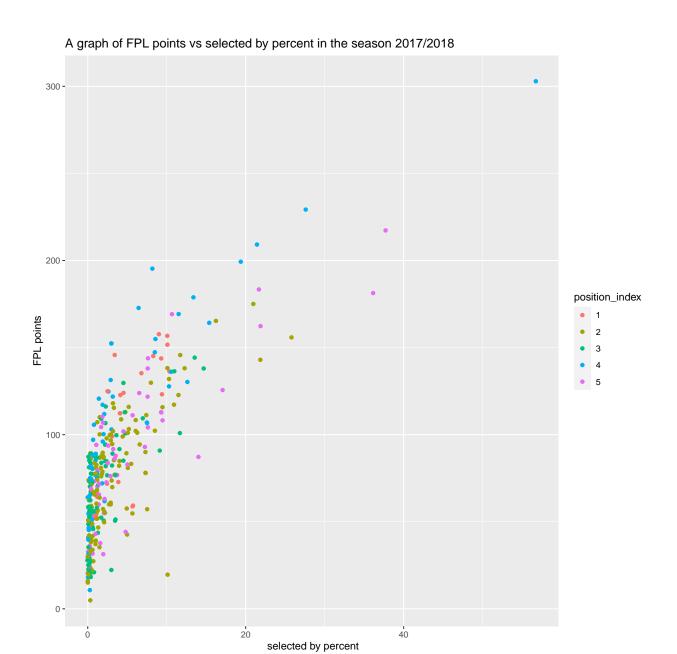
How selected by percent varies with the total FPL points per season

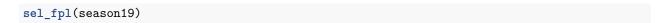
```
sel_fpl <- function(df) {
    df %>%
        filter(minutes.played >= 500) %>%
        ggplot(aes(selected_by_percent, total_points)) +
        geom_point(aes(color = position_index), position = "jitter") +
        labs(
        title = sprintf("A graph of FPL points vs selected by percent in the season %s",
        unique(df$season)), x = " selected by percent", y = "FPL points")
}
```

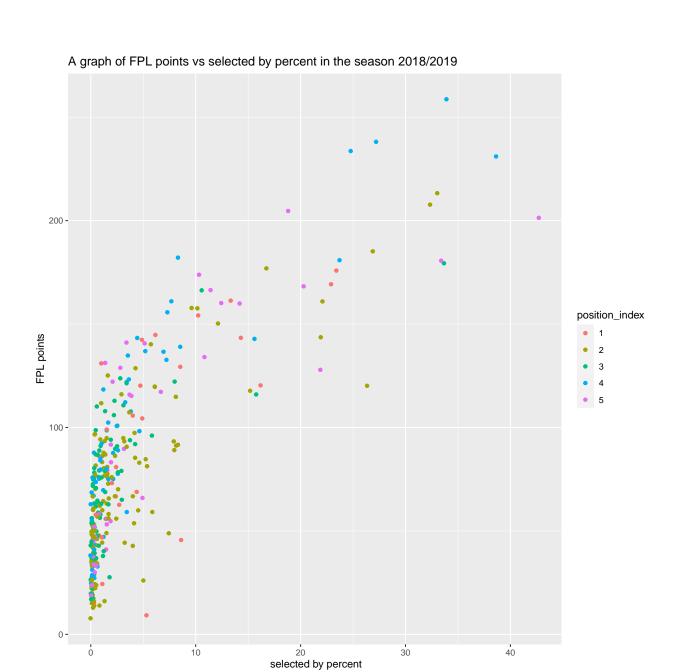
sel_fpl(season17)



sel_fpl(season18)



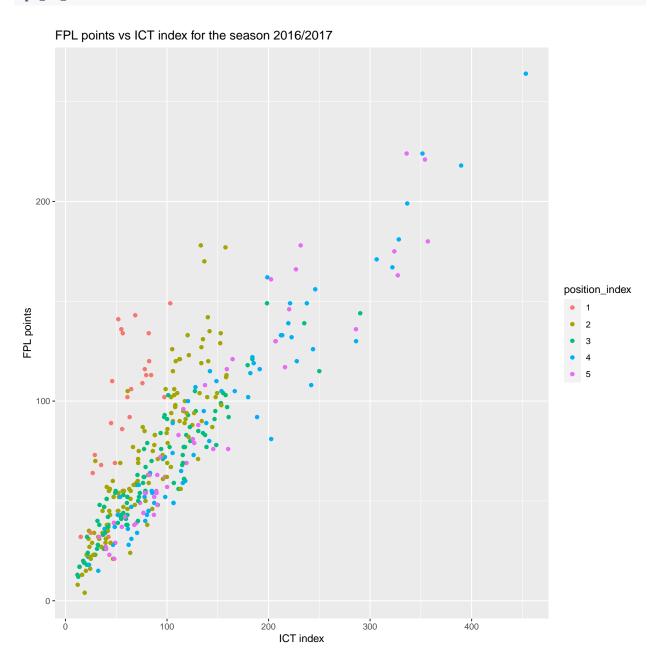




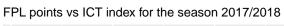
Creating a function that plots FPL points vs ICT index

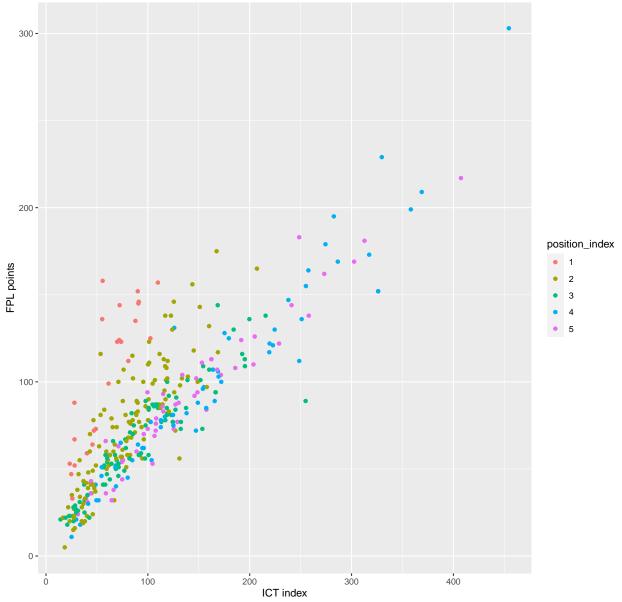
Plotting a graph of FPL points vs ICT index

fpl_vs_ict(season17)

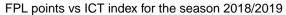


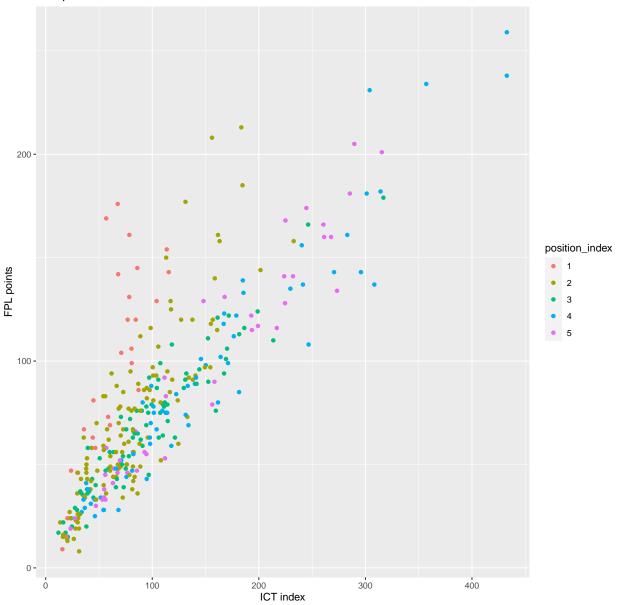
fpl_vs_ict(season18)





fpl_vs_ict(season19)





Creating a function for printing out the first ten rows alongside selected columns

Applying the function to the datasets

quickview(season17)

Table 1: The first 10 rows of the season 2016/2017 dataset with specific columns

player_name	club_name	total_points	minutes.played	fpl_to_game
Joe Hart	Manchester City	0	0	NaN
Claudio Bravo	Manchester City	73	1968	3.338415
Willy Caballero	Manchester City	64	1452	3.966942
Angus Gunn	Manchester City	0	0	NaN
Nicolás Otamendi	Manchester City	100	2592	3.472222
Vincent Kompany	Manchester City	57	820	6.256098
John Stones	Manchester City	59	2014	2.636544
Eliaquim Mangala	Manchester City	0	0	NaN
Jason Denayer	Manchester City	56	1878	2.683706
Aleksandar Kolarov	Manchester City	95	2535	3.372781

quickview(season18)

Table 2: The first 10 rows of the season 2017/2018 dataset with specific columns

$club_name$	$total_points$	minutes.played	fpl_to_game
Chelsea FC	136	3150	3.885714
Chelsea FC	11	270	3.666667
Chelsea FC	115	2336	4.430651
Chelsea FC	79	2067	3.439768
Chelsea FC	74	2082	3.198847
Chelsea FC	87	2887	2.712158
Chelsea FC	0	0	NaN
Chelsea FC	0	0	NaN
Chelsea FC	1	10	9.000000
Chelsea FC	165	2855	5.201401
	Chelsea FC	Chelsea FC 136 Chelsea FC 11 Chelsea FC 115 Chelsea FC 79 Chelsea FC 74 Chelsea FC 87 Chelsea FC 0 Chelsea FC 0 Chelsea FC 1	Chelsea FC 136 3150 Chelsea FC 11 270 Chelsea FC 115 2336 Chelsea FC 79 2067 Chelsea FC 74 2082 Chelsea FC 87 2887 Chelsea FC 0 0 Chelsea FC 0 0 Chelsea FC 1 10

quickview(season19)

Table 3: The first 10 rows of the season 2018/2019 dataset with specific columns

player_name	club_name	$total_points$	minutes.played	fpl_to_game
Ederson Santana de Moraes	Manchester City	169	3420	4.447368
Claudio Bravo	Manchester City	0	0	NaN
John Stones	Manchester City	83	1761	4.241908
Aymeric Laporte	Manchester City	177	3056	5.212696
Nicolás Otamendi	Manchester City	70	1233	5.109489
Vincent Kompany	Manchester City	58	1220	4.278689

player_name	club_name	total_points	minutes.played	fpl_to_game
Philippe Sandler	Manchester City	0	0	NaN
Benjamin Mendy	Manchester City	59	900	5.900000
Oleksandr Zinchenko	Manchester City	44	1151	3.440487
Kyle Walker	Manchester City	150	2776	4.863112

Writing the files to csv files

```
s17 <- file('season2017.csv', encoding = "UTF-8")
write.csv(season17, file = s17)
s18 <- file('season2018.csv', encoding = 'UTF-8')
write.csv(season18, file = s18)
s19 <- file('season2019.csv', encoding = "UTF-8")
write.csv(season19, file = s19)</pre>
```