

# Term Project Report

DA5020 Collect/Store/Retrieve Data



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MS in Bioinformatics (Spring 2017)

# Premier League 2016-17 Player Database

Motivation: I am a Bioinformatics student, so ideally I should select any biological database for the final term project. However, biological data is big responsibility to work with and I have worked on biological data during my other courses. So, my final choice was to work with sports data and I have decided to work on Premier League 2016-17 player and stats database.

## Introduction:

Premier League or English Premier League is a major soccer league contested between top 20 teams in England every year. It is the one of the most followed and major soccer league in the world. For the final term project, I have decided to collect data for the Premier League 2016-17 season, which is also current running season. I have tried to collect most of the possible and significant data for the season.

## Work Approach:

1)Collecting 2) Storing 3) Retrieval 4) Graphical Representation

## 1)Collecting:

For the collection of the data I will be scraping directly from the [www.premierleague.com](http://www.premierleague.com) and [www.nbcsports.com](http://www.nbcsports.com)

The screenshot shows the NBC Sports website interface. The top navigation bar includes links for NFL, NHL, NBA, MLB, SOCCER, NASCAR, GOLF, MOTORS, NCAA FB, and MORE. The main content area is titled 'WATCH NOW' and 'VIEW FULL SCHEDULE'. Below this, there are sections for 'LIVE' and 'UPCOMING EVENTS'. The 'LIVE' section shows a hockey game between Oilers vs. Sharks. The 'UPCOMING EVENTS' section shows a London Marathon and a soccer match between Burnley v. Man United. The 'CLUBS' section lists 20 Premier League teams, each with a link to Stats, Squad, and Fixtures.

Premier League	
Arsenal	Stats   Squad   Fixtures
Bournemouth	Stats   Squad   Fixtures
Burnley	Stats   Squad   Fixtures
Chelsea	Stats   Squad   Fixtures
Crystal Palace	Stats   Squad   Fixtures
Everton	Stats   Squad   Fixtures
Hull City	Stats   Squad   Fixtures
Leicester City	Stats   Squad   Fixtures
Liverpool	Stats   Squad   Fixtures
Manchester City	Stats   Squad   Fixtures
Manchester United	Stats   Squad   Fixtures
Middlesbrough	Stats   Squad   Fixtures
Southampton	Stats   Squad   Fixtures
Stoke City	Stats   Squad   Fixtures
Sunderland	Stats   Squad   Fixtures
Swansea City	Stats   Squad   Fixtures
Tottenham Hotspur	Stats   Squad   Fixtures

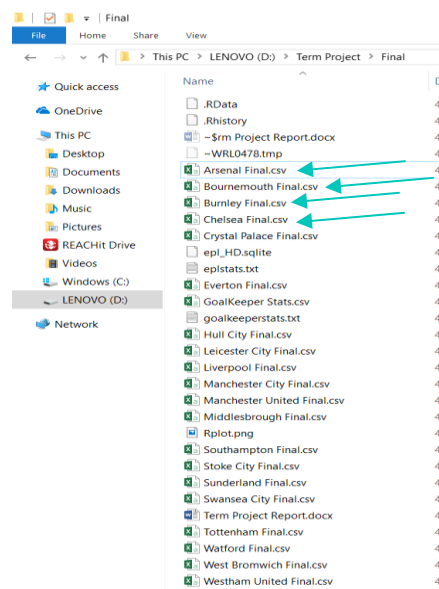
I am using [import.io](http://import.io) as a scraping tool and URL get query to collect data team wise using team id as URL query.

For example: <http://scores.nbcports.com/epl/teamstats.asp?team=21>, here team=21 is a team query for the team called Arsenal. I have extracted data using import.io tool using URL as input.

The screenshot shows the import.io interface with the URL `http://scores.nbcports.com/epl/teamstats.asp?team=21` entered. The table below lists player statistics for Arsenal (team 21).

	Pos	MP	Min	G	Ast	PS	SHTS	SOG	SOG%	YC	RC	
1												
2	Alexis Sánchez	F	31	2628	19	9	2	104	41	39.4	5	0
3	Nacho Monreal	D	29	2523	0	2	0	9	2	22.2	3	0
4	Petr Cech	GK	28	2468	0	0	0	0	0	-	2	0
5	Laurent Koscielny	D	28	2449	2	0	0	8	3	37.5	3	0
6	Mesut Özil	MF	26	2225	7	7	0	36	15	41.7	0	0
7	Héctor Bellerín	D	26	2064	0	2	0	17	4	23.5	2	0
8	Shkodran Mustafi	D	23	2005	2	2	0	18	3	16.7	9	0

Import.io allow us to store collected data into .csv files. So, I have stored 20 .csv file for each team separately.

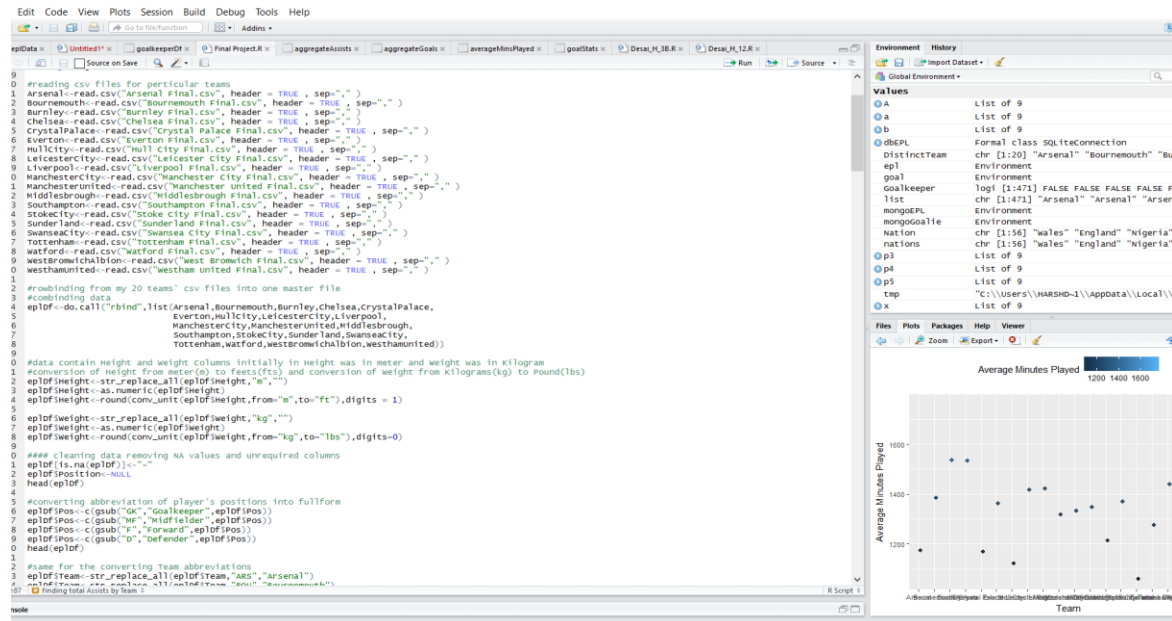


File Home Insert Page Layout Formulas Data Review View Tell me what you want to do												
A1 X Y Z Name												
	A	B	C	D	E	F	G	H	I	J	K	
1	Name	Pos	MP	Min	GS	Ast	PS	SHTS	SOTS	ST%	YC	
2	Aaron Ramsey	MF	16	772	0	2	0	29	6	20.7		
3	Ainsley Maitland-Niles	MF	1	1	0	0	0	0	0	0		
4	Alex Iwobi	F	24	1419	3	4	0	35	9	25.7		
5	Alex Oxlade-Chamberlain	MF	26	1352	2	5	0	26	7	26.9		
6	Alexis Sánchez	F	31	2628	19	9	2	104	41	39.4		
7	Carl Jenkinson	D	1	83	0	0	0	0	0	0		
8	Damián Martínez	GK	2	180	0	0	0	0	0	0		
9	Danny Welbeck	F	9	417	1	0	0	9	5	55.6		
10	David Ospina	GK	2	142	0	0	0	0	0	0		
11	Francis Coquelin	MF	24	1577	0	0	0	12	2	16.7		
12	Gabriel	D	16	1298	0	0	0	7	2	28.6		
13	Granit Xhaka	MF	25	1923	1	2	0	24	4	16.7		
14	Héctor Bellerín	D	26	2064	0	2	0	17	4	23.5		
15	Kieran Gibbs	D	6	295	0	1	0	3	2	66.7		
16	Laurent Koscielny	D	28	2449	2	0	0	8	3	37.5		
17	Lucas Pérez	F	11	265	1	0	0	9	4	44.4		
18	Mathieu Debuchy	D	1	16	0	0	0	0	0	0		
19	Mesut Özil	MF	26	2225	7	7	0	36	15	41.7		

Reading data into R using read.csv() function and combining all 20 files into master file using R functions. I might have combined data manually but I wanted to use R functions that's the main I choose to extract file per teams and not a single master file.

Packages used: stringr ,measurements

Loading .csv files, combining files and Cleaning data



Combined final data frame :

RStudio

File Edit Code View Plots Session Build Debug Tools Help

	Name	Position	Matches Played	Minutes Played	Goals	Assists	Penalties Scored	Shots	Shots on Target	Shots on Target %	Yellow Cards	Red Cards	Fouls	Fouls Suffered	Crosses	Offsides
1	Aaron Ramsey	Midfielder	16	772	0	2	0	29	6	20.7	3	0	11	15	24	
2	Ainsley Maitland-Niles	Midfielder	1	1	0	0	0	0	0	-	0	0	0	0	0	
3	Alex Iwobi	Forward	24	1419	3	4	0	35	9	25.7	1	0	6	12	33	
4	Alex Oxlade-Chamberlain	Midfielder	26	1352	2	5	0	26	7	26.9	1	0	23	16	74	
5	Alexis Sánchez	Forward	31	2628	19	9	2	104	41	39.4	5	0	40	52	136	
6	Carl Jenkinson	Defender	1	83	0	0	0	0	0	-	0	0	1	1	3	
7	Damián Martínez	Goalkeeper	2	180	0	0	0	0	0	0	0	0	1	0	0	
8	Danny Welbeck	Forward	9	417	1	0	0	9	5	55.6	0	0	5	5	2	
9	David Ospina	Goalkeeper	2	142	0	0	0	0	0	0	0	0	0	1	0	
10	Francis Coquelin	Midfielder	24	1577	0	0	0	12	2	16.7	5	0	33	34	9	
11	Gabriel	Defender	16	1298	0	0	0	7	2	28.6	4	0	14	8	14	
12	Granit Xhaka	Midfielder	25	1923	1	2	0	24	4	16.7	4	2	30	14	35	
13	Héctor Bellerín	Defender	26	2064	0	2	0	17	4	23.5	2	0	17	11	75	
14	Kieran Gibbs	Defender	6	295	0	1	0	3	2	66.7	3	0	7	1	19	
15	Laurent Koscielny	Defender	28	2449	2	0	0	8	3	37.5	3	0	14	24	4	
16	Lucas Pérez	Forward	11	265	1	0	0	9	4	44.4	0	0	3	2	11	
17	Mathieu Debuchy	Defender	1	16	0	0	0	0	0	-	0	0	0	0	0	
18	Mesut Özil	Midfielder	26	2225	7	7	0	36	15	41.7	0	0	8	20	194	
19	Mohamed Elneny	Midfielder	14	694	0	1	0	14	2	14.3	1	0	10	5	10	
20	Nacho Monreal	Defender	29	2523	0	2	0	9	2	22.2	3	0	26	33	78	
21	Olivier Giroud	Forward	23	913	9	3	0	30	14	46.7	1	0	16	10	3	
22	Petr Čech	Goalkeeper	28	2468	0	0	0	0	0	0	2	0	2	1	0	

Collected Variables:

Basic Player Info: Name, Position, Team, Matches Played, Minutes Played, Nationality etc

Standard Stats: Goals, Assists, Shots, Yellow Cards, Red Cards, Crosses, Offside etc.

Goalkeeper Stats: Saves, Saves, Penalty Kicks Saved, Goals Allowed, Shots faced etc.

## 2) Storing:

Initially, I was planning to store collected data using Relational Database. After working on relational database schema, I realized that it will be difficult to store data into relational database due to complexity between player positions and their statistics. For example, Goalkeeper statistics attributes are totally different than other player position's attribute for soccer. So, I decided to go with non-relational (NoSQL) database system to store collected data. It is simpler than relational database system, other advantages of NoSQL database system are no unique constrain, massive parallel processing and tolerance for failure. Non-relational system I have used for project is MongoDB.

Packages used: mongolite

Steps for running MongoDB server

1. installing MongoDB from mongodb.com

2. Setting up MongoDB environment

mkdir \data\db

"C:\Program Files\MongoDB\Server\3.4\bin\mongod.exe" --dbpath d:\test\mongodb\data

3. C:\MongoDB\Server\3.2\bin\mongod.exe Code to start MongoDB server

3. Connect using R

After connecting MongoDB server with R studio using mongolite R package, next step is to insert collected and cleaned data into MongoDB system using `insert()` function provided my mongolite. Mongolite also provides option of exporting JSON structure to file using `export()` function.

Setting MongoDB connections, inserting files and exporting JSON structure to designated files.

The screenshot displays the RStudio interface with several components:

- Environment/Source View:** Shows a list of data frames including `epiData`, `goalkeeperDf`, and `Final Project`. The `goalkeeperDf` data frame is highlighted, showing columns for names, positions, and statistics.
- Console:** Contains R code for setting up MongoDB connections, inserting data, and exporting JSON structures to files. The code includes comments and function calls like `epi <- mongo("epiData")`, `goal <- mongo("goalkeeperDf")`, `insert(goalkeeperDf)`, and `export(goalkeeperDf, "goalkeeperstats.txt")`.
- Export Dialog:** A modal window titled "Export" is open, showing a list of data frames to be exported. The `goalkeeperDf` data frame is selected for export.

### 3.Retrieval:

Functions used : sort(),find(),distinct()

Packages used: dplyr,ggplot2,mongolite

find() function is used to find specific value from the stored data.

Firstly using find() function to search all the forward position players from the database.

```
> EPL$find({'Position':'Forward'})
```

	Name	Position	Matches Played	Minutes Played	Goals	Assists	Penalties Scored	Shots	Shots on Target
1	Alex Iwobi	Forward	24	1419	3	4	0	35	9
2	Alexis Sánchez	Forward	31	2628	19	9	2	104	41
3	Danny Welbeck	Forward	9	417	1	0	0	9	5
4	Lucas Pérez	Forward	11	265	1	0	0	9	4
5	Olivier Giroud	Forward	23	913	9	3	0	30	14
6	Theo Walcott	Forward	24	1830	10	2	0	58	27
7	Benik Afobe	Forward	28	1279	5	3	0	34	16
8	Callum Wilson	Forward	20	1363	6	0	3	32	12
9	Joshua King	Forward	32	2365	13	2	2	53	25
10	Andre Gray	Forward	27	2032	8	2	2	41	18
11	Ashley Barnes	Forward	23	1485	5	1	2	41	17
12	Daniel Agyei	Forward	2	8	0	0	0	0	0
13	Sam Vokes	Forward	33	1695	6	3	2	43	17
14	Diego Costa	Forward	30	2667	17	5	0	92	36
15	Michy Batshuayi	Forward	17	113	1	1	0	6	2
16	Pedro	Forward	29	1874	7	7	0	48	16
17	Christian Benteke	Forward	30	2627	12	2	2	89	33
18	Connor Wickham	Forward	8	410	2	0	0	10	5
19	Fraizer Campbell	Forward	8	113	1	0	0	2	1
20	Jeff Schlupp	Forward	10	547	0	0	0	3	0
21	Jonathan Benteke	Forward	1	6	0	0	0	1	1
22	Loïc Remy	Forward	4	112	0	0	0	5	2
23	Ademola Lookman	Forward	7	245	1	0	0	14	6
24	Arouna Koné	Forward	4	43	0	0	0	3	2
25	Dominic Calvert-Lewin	Forward	8	210	1	0	0	4	1
26	Enner Valencia	Forward	17	483	3	1	0	17	6
27	Kevin Mirallas	Forward	30	1724	4	7	0	53	21
28	Romelu Lukaku	Forward	32	2817	24	6	0	95	47
29	Yannick Bolasie	Forward	13	1028	1	4	0	26	6
30	Abel Hernández	Forward	22	1293	4	3	0	34	15
31	Adama Diomande	Forward	22	1169	2	0	0	23	6
32	Dieumerci Mbokani	Forward	11	750	0	1	0	15	4
33	Jarrod Bowen	Forward	5	61	0	0	0	0	0
34	Oumar Niasse	Forward	12	710	4	0	0	17	8
35	Will Keane	Forward	5	282	0	0	0	4	0
36	Ahmed Musa	Forward	20	741	2	0	0	17	6
37	Islam Slimani	Forward	19	1100	7	3	1	29	12
38	Jamie Vardy	Forward	29	2280	11	4	0	43	17
39	Leonardo Ulloa	Forward	14	342	1	0	0	11	5
40	Shinji Okazaki	Forward	25	1304	2	0	0	26	8

Using find() function to fetch multiple columns from the database and to make data frame

```
> ep1Data<-EPL$find({'', fields = '{"Name":1,
+ "Position":1,
+ "Matches Played":1,
+ "Minutes Played":1,
+ "Goals":1,
+ "Assists":1,
+ "Shots":1,
+ "Fouls":1,
+ "Team":1,
+ "Nationality":1,
+ "Height(ft)":1,
+ "Weight(lbs)":1,
+ "_id":0 }'})
> ep1Data<-as.data.frame(ep1Data)
> head(ep1Data)
```

	Name	Position	Matches Played	Minutes Played	Goals	Assists	Shots	Fouls	Height(ft)	Weight(lbs)	Nationality
1	Aaron Ramsey	Midfielder	16	772	0	2	29	11	5.8	168	Wales
2	Ainsley Maitland-Niles	Midfielder	1	1	0	0	0	0	5.8	157	England
3	Alex Iwobi	Forward	24	1419	3	4	35	6	5.7	159	Nigeria
4	Alex Oxlade-Chamberlain	Midfielder	26	1352	2	5	26	23	5.9	154	England
5	Alexis Sánchez	Forward	31	2628	19	9	104	40	5.5	154	Chile
6	Carl Jenkinson	Defender	1	83	0	0	0	1	6.1	170	England



The other function used for retrieval was `distinct()`, which will find distinct values from particular columns.

Use of `distinct()` function to find distinct teams and players from different nations.

```
RStudio
File Edit Code View Plots Session Build Debug Tools Help
Go to file/function Addins
Source
Console D:/Term Project/Final/ >R>
> distinctTeam<-EPL$distinct("Team")
> distinctTeam
[1] "Arsenal" "Bournemouth" "Burnley" "Chelsea" "Crystal Palace"
[6] "Everton" "Hull City" "Leicester City" "Liverpool" "Manchester City"
[11] "Manchester United" "Middlesbrough" "Southampton" "Stoke City" "Sunderland"
[16] "Swansea City" "Tottenham Hotspur" "Watford" "West Bromwich Albion" "West Ham United"
> nations<-EPL$distinct("Nationality")
> nations
[1] "Wales" "England" "Nigeria" "Chile" "Argentina" "Colombia"
[7] "France" "Brazil" "Switzerland" "Spain" "Germany" "Egypt"
[13] "Czechoslovakia" "Australia" "South Africa" "Poland" "Norway" "Ivory Coast"
[19] "Scotland" "Ireland" "Iceland" "Belgium" "Yugoslavia" "Sierra Leone"
[25] "DR Congo" "South Korea" "Senegal" "Netherlands" "Ecuador" "Uruguay"
[31] "Italy" "Malaysia" "Austria" "Ghana" "Algeria" "Denmark"
[37] "Japan" "Serbia-Montenegro" "Ukraine" "Chile" "Jamaica" "Sweden"
[43] "USA" "Portugal" "Burundi" "Costa Rica" "Gabon" "California"
[49] "Kenya" "Romania" "Northern Ireland" "Brazil" "West Germany" "Zimbabwe"
[55] "Venezuela" "New Zealand"
```

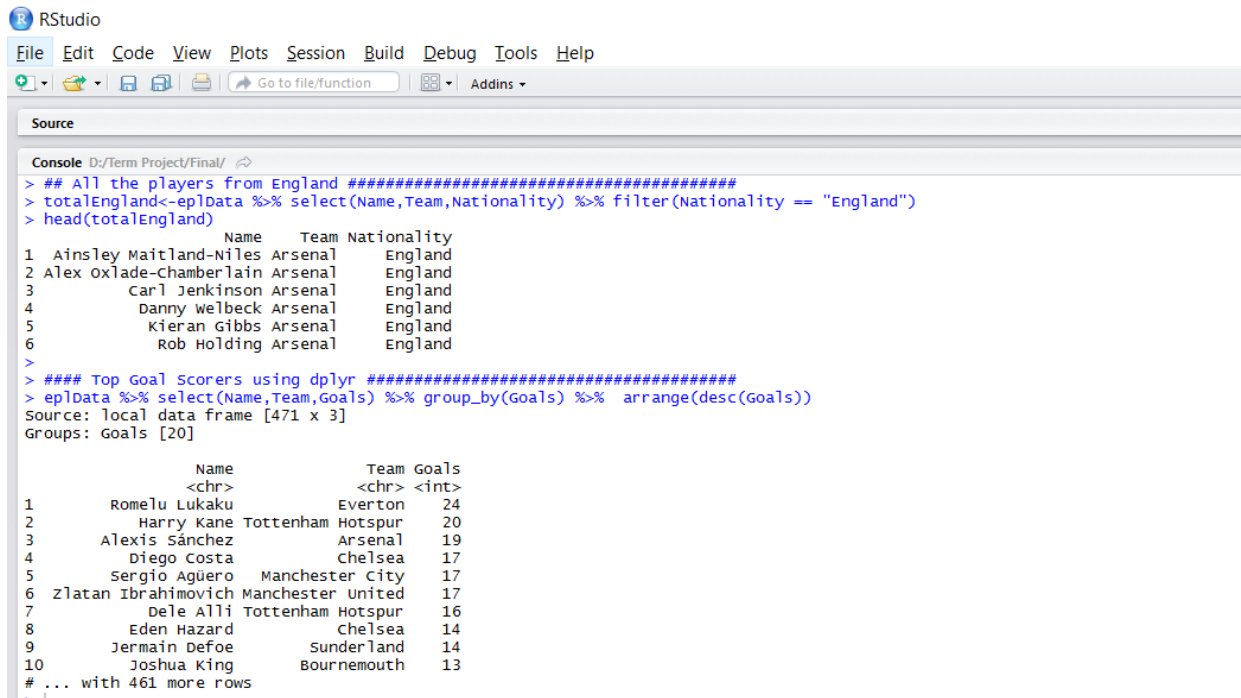
We can see that there are 20 distinct teams in a database and players are from 56 distinct countries.

Next function I have used for the retrieval is `sort()` function. I have used `sort()` function to find top 10 Goal scorers and top 10 Assists by players.

```
#####
> # finding top 10 goal scoring players using sort() function
> TopGoals<-EPL$find('Goals')
+ fields = '{"Name":1,
+ "Team":1,
+ "Goals":1,
+ "_id":0}',
+ sort='{"Goals":-1}'
> TopGoals<-TopGoals[1:10,]
> TopGoals
  Name Goals Team
1 Romelu Lukaku 24 Everton
2 Harry Kane 20 Tottenham Hotspur
3 Alexis Sánchez 19 Arsenal
4 Diego Costa 17 Chelsea
5 Sergio Agüero 17 Manchester City
6 Zlatan Ibrahimovich 17 Manchester United
7 Dele Alli 16 Tottenham Hotspur
8 Eden Hazard 14 Chelsea
9 Jermain Defoe 14 Sunderland
10 Joshua King 13 Bournemouth
#####
> # finding top 10 assists by players using sort() function and using plotting
> ## graph to present data graphically
> TopAssists<-EPL$find('Assists')
+ fields = '{"Name":1,
+ "Team":1,
+ "Assists":1,
+ "_id":0}',
+ sort='{"Assists":-1}'
> TopAssists<-TopAssists[1:10,]
> TopAssists$Assists<-as.numeric(TopAssists$Assists)
> TopAssists
  Name Assists Team
1 Kevin De Bruyne 13 Manchester City
2 Christian Eriksen 12 Tottenham Hotspur
3 Gylfi Sigurdsson 11 Swansea City
4 Alexis Sánchez 9 Arsenal
5 Wilfried Zaha 9 Crystal Palace
6 Ross Barkley 8 Everton
7 Matt Phillips 8 West Bromwich Albion
8 Mesut Özil 7 Arsenal
9 Cesc Fàbregas 7 Chelsea
10 Pedro 7 Chelsea
```

I have also used **dplyr package** to work on my retrieved data. I have used `filter`, `group_by`, `select`, `arrange` etc.

I have used dplyr package to find all the player from England and Top Goal Scorer for the season.



```

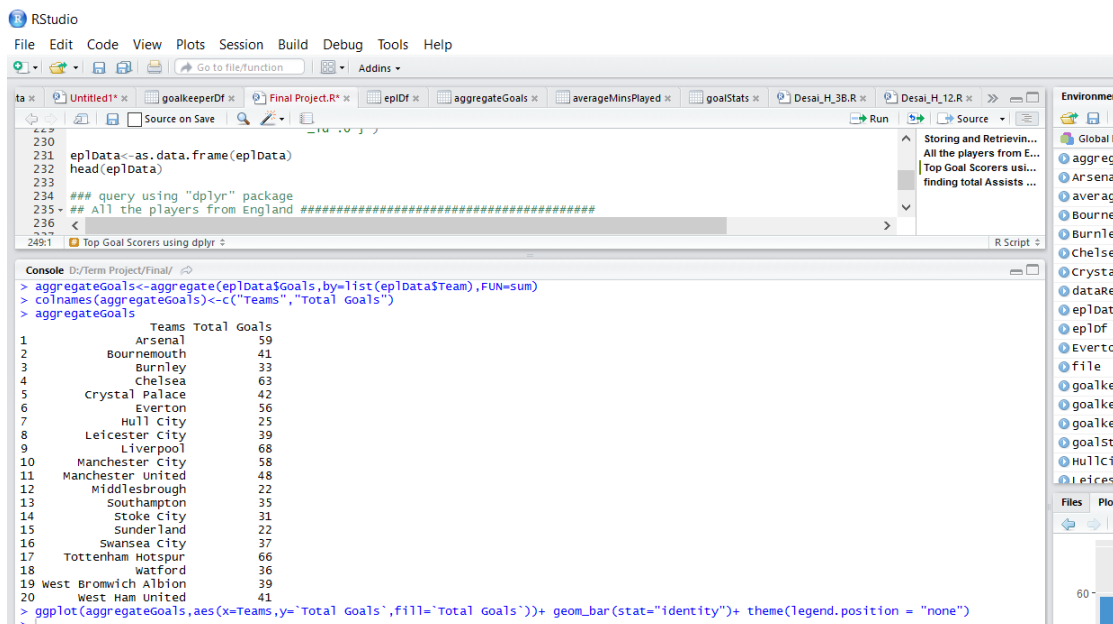
RStudio
File Edit Code View Plots Session Build Debug Tools Help
Source
Console D:/Term Project/Final/
> ## All the players from England #####
> totalEngland<-ep1Data %>% select(Name,Team,Nationality) %>% filter(Nationality == "England")
> head(totalEngland)
  Name      Team Nationality
1 Ainsley Maitland-Niles Arsenal      England
2 Alex Oxlade-Chamberlain Arsenal      England
3 Carl Jenkinson Arsenal      England
4 Danny Welbeck Arsenal      England
5 Kieran Gibbs Arsenal      England
6 Rob Holding Arsenal      England
>
> #### Top Goal Scorers using dplyr #####
> ep1Data %>% select(Name,Team,Goals) %>% group_by(Goals) %>% arrange(desc(Goals))
Source: local data frame [471 x 3]
Groups: Goals [20]
   Name      Team Goals
   <chr>    <chr> <int>
1  Romelu Lukaku Everton    24
2  Harry Kane Tottenham    20
3  Alexis Sánchez Arsenal    19
4  Diego Costa Chelsea    17
5  Sergio Agüero Manchester City    17
6  Zlatan Ibrahimović Manchester United    17
7  Dele Alli Tottenham    16
8  Eden Hazard Chelsea    14
9  Jermain Defoe Sunderland    14
10 Joshua King Bournemouth    13
# ... with 461 more rows

```

#### 4.Graphical Representation:

Last but not the list, I have used ggplot2 package to present my data and results graphically.

Use of aggregate function to calculate total number of goals by each team and present data graphically using ggplot.

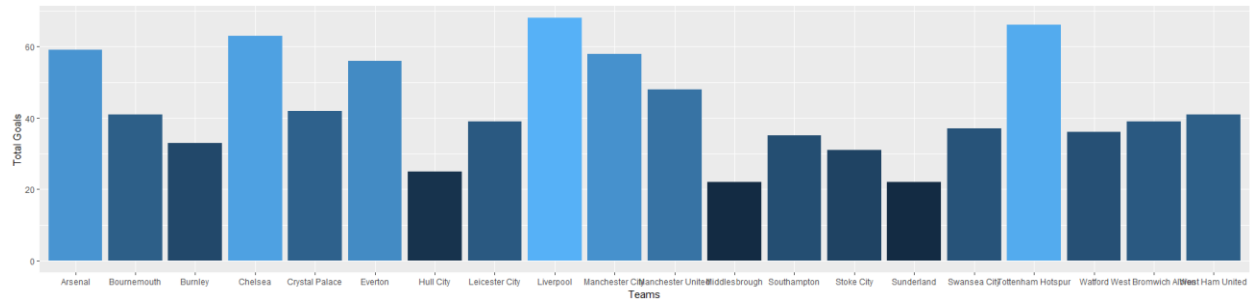


```

RStudio
File Edit Code View Plots Session Build Debug Tools Help
Source on Save
goalkeeperDf x Final Project.R x ep1Df x aggregateGoals x averageMinsPlayed x goalStats x Desai_H_3B.R x Desai_H_12.R x
230
231 ep1Data<-as.data.frame(ep1Data)
232 head(ep1Data)
233
234 ## query using "dplyr" package
235 > ## All the players from England #####
236 <
249:1 Top Goal Scorers using dplyr
Console D:/Term Project/Final/
> aggregateGoals<-aggregate(ep1Data$Goals,by=list(ep1Data$Team),FUN=sum)
> colnames(aggregateGoals)<-c("Teams","Total Goals")
> aggregateGoals
  Teams Total Goals
1 Arsenal         59
2 Bournemouth      41
3 Burnley          33
4 Chelsea         63
5 Crystal Palace   42
6 Everton          56
7 Hull City        25
8 Leicester City   39
9 Liverpool        68
10 Manchester City  58
11 Manchester United 48
12 Middlesbrough   22
13 Southampton     35
14 Stoke City      31
15 Sunderland      22
16 Swansea City    37
17 Tottenham Hotspur 66
18 Watford         36
19 West Bromwich Albion 39
20 West Ham United  41
> ggplot(aggregateGoals,aes(x=Teams,y='Total Goals',fill='Total Goals'))+ geom_bar(stat="identity")+ theme(legend.position = "none")
>

```



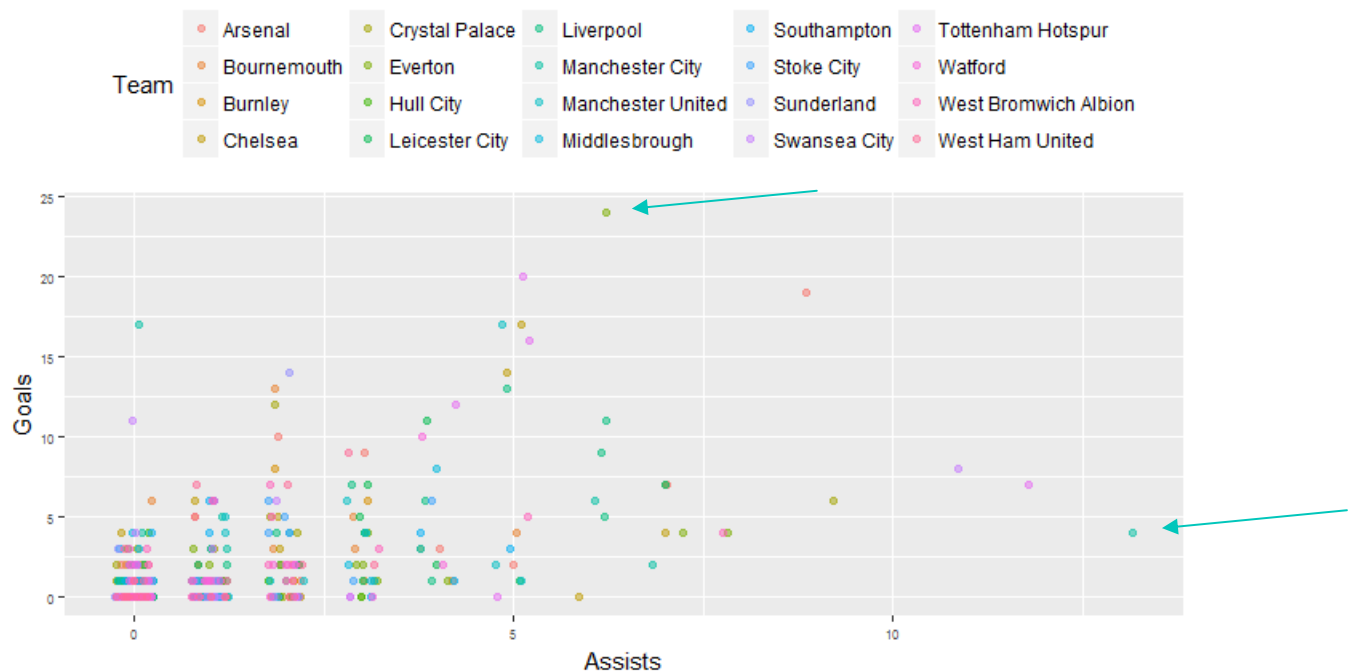


Then I plotted abundance graph for goals and assists for by the team.

```

RStudio
File Edit Code View Plots Session Build Debug Tools Help
Source
Console D:/Term Project/Final/
> a <- ggplot(epldata,
+           aes(x = Assists,
+             y = Goals)) +
+           theme(legend.position="top",
+             axis.text=element_text(size = 6))
> (b <- a + geom_point(aes(color = Team),
+           alpha = 0.5,
+           size = 1.5,
+           position = position_jitter(width = 0.25, height = 0)))
> |

```



Presented two points in graph shows highest goals and highest assists and color for the team matches with our data. Top goal scorer Romelu Lukaku is from Everton and Top Assists player Kevin De Bruyne is from Manchester City.

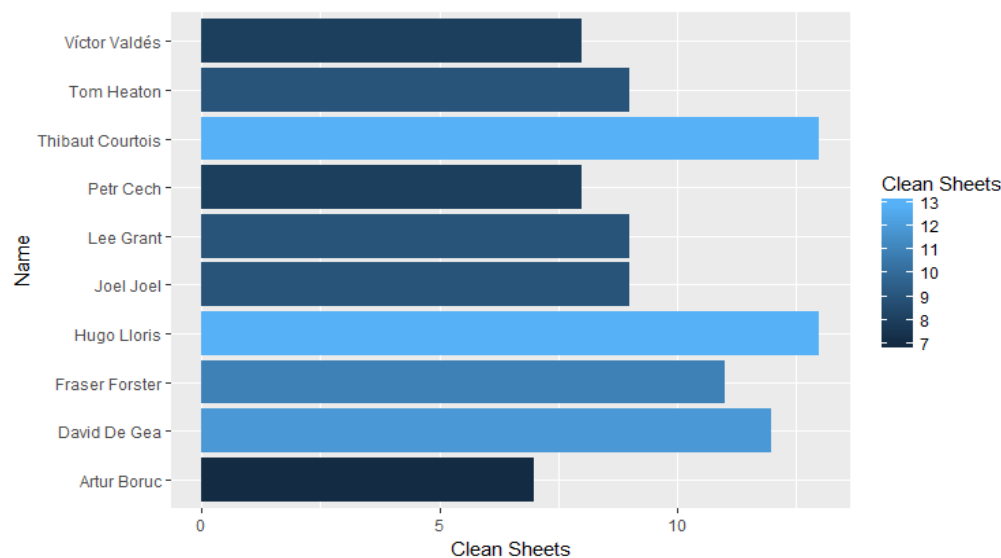
Other graph would be to find Top 10 Cleansheets by Goalkeeper and to plot it.

```
RStudio
File Edit Code View Plots Session Build Debug Tools Help

ta * Untitled1 * goalKeeperDf Final Project.R * eplDf * aggregateGoals * averageMinsPlayed * goalStats * Desai_H_3B.R * Desa
292:1 finding total Assists by Team +

Console D:/Term Project/Final/
> ## to find highest Cleansheet by Goalkeepers and Plot a graph
> TopCleansheets<-GOALIE$find('{}',
+                               fields = '{"Name":1,
+                               "Team":1,
+                               "Clean Sheets":1,
+                               "_id":0}',
+                               sort='{"Clean Sheets":-1}')
> TopCleansheets<-TopCleansheets[1:10,]
> TopCleansheets$`Clean Sheets`<-as.numeric(TopCleansheets$`clean Sheets`)
> head(TopCleansheets)
  Name Team Clean Sheets
1 Thibaut Courtois Chelsea 13
2 Hugo Lloris Tottenham Hotspur 13
3 David De Gea Manchester United 12
4 Fraser Forster Southampton 11
5 Tom Heaton Burnley 9
6 Joel Joel Everton 9

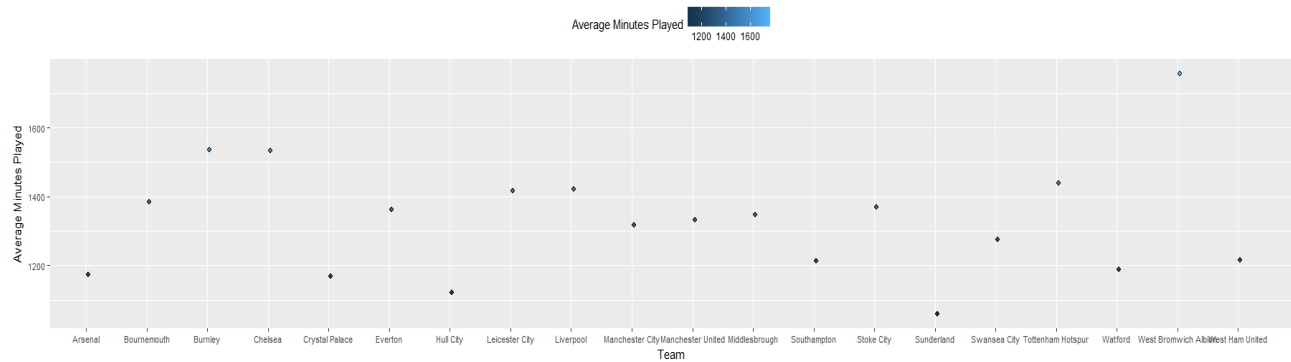
> ggplot() +
+   layer(
+     data = TopCleansheets, mapping = aes(x = Name, y = `Clean Sheets`, fill = `clean Sheets`),
+     geom = "bar", stat = "identity", position = "identity"
+   ) +
+   scale_y_continuous() +
+   coord_flip()
> |
```



In the end I calculated average minutes played by every team and plotted a graph for it.

```
Console D:/Term Project/Final/
> ## To find average minutes played by team and plot a graph to represent it graphically
> eplData$`Minutes Played`<-as.numeric(eplData$`Minutes Played`)
> averageMinsPlayed<-aggregate(eplData$`Minutes Played`,by=(list(eplData$Team)),FUN=mean)
> colnames(averageMinsPlayed)<-c("Team","Average Minutes Played")
> averageMinsPlayed
  Team Average Minutes Played
1 Arsenal 1171.769
2 Bournemouth 1183.217
3 Burnley 1136.524
4 Chelsea 1132.000
5 Crystal Palace 1167.259
6 Everton 1162.261
7 Hull City 1121.885
8 Leicester City 1116.000
9 Liverpool 1120.435
10 Manchester City 1116.768
11 Manchester United 1132.043
12 Middlesbrough 1147.348
13 Southampton 1213.583
14 Stoke City 1169.435
15 Sunderland 1059.214
16 Swansea City 1174.958
17 Tottenham Hotspur 1139.955
18 West Bromwich Albion 1187.577
19 West Ham United 1176.056
20 West Ham United 1215.520

> x<-ggplot(averageMinsPlayed,aes(x=Team,y=`Average Minutes Played`,fill=`Average Minutes Played`))
> x<-geom_dotplot(binaxis = "y",binwidth = 15)+ theme(legend.position="top")
> x<-x+axis.text.element_text(size = 8)
> |
```



### Future Work:

- To collect more stats for the same seasons and previous seasons
- To use better and dynamic scraping technique
- To collect data for different soccer leagues like La Liga, Champions League etc.
- After collecting other Premier League Seasons data, produce comparative statistics.
- Better graphical representation and will try to create soccer statistics application.

### References:

- Collecting, Storing and Retrieving Data by Yatish Jain and Martin Schedlbauer
- Data Manipulation with R by Jaynal Abedin
- Seven Database in Seven Weeks by Eric Redmond and Jim R. Wilson
- [www.premierleague.com](http://www.premierleague.com)
- [www.nbcsports.com](http://www.nbcsports.com)
- <http://ggplot2.org/>

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