# DA Lab 2

# Analysis of IPL-2019 Dataset

Name: Dheeraj Chaudhary

**Roll: 17BCS009** 

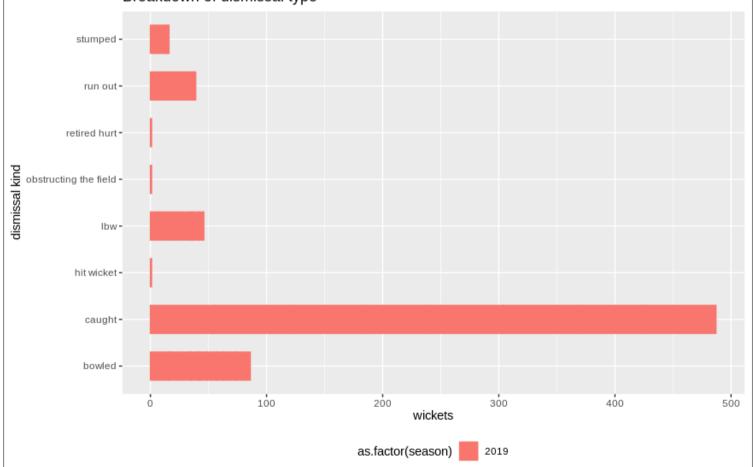
```
library(tabulizer)
library(dplyr)
library(ggplot2)
library(reshape2)
library(magrittr)
library(tidyr)
                          matches <- read.csv("/home/dheeraj/Desktop/Lecture/6th_sem_Academics/DataScience/Lab2/</pre>
matches.csv", stringsAsFactors = FALSE)
data <- read.csv("/home/dheeraj/Desktop/Lecture/6th_sem_Academics/DataScience/Lab2/</pre>
deliveries.csv", stringsAsFactors = FALSE)
matches <- matches[,-18]</pre>
data$wickets <- as.numeric(ifelse(data$player_dismissed =="" ,"",1))</pre>
####### Number of matches in the dataset
                                        (We can see 60 matches were played in IPL'2019)
summarize(matches, no of matches = n())
##### OUTPUT > no of matches 60
```

```
####### Which Team won by maximum runs?
                                               (We can see SRH won y 118 runs)
max_run <- matches[which.max(matches$win_by_runs),]</pre>
select(max_run, winner, win_by_runs)
###### Output >
                             winner win by runs
                      11 Sunrisers Hyderabad
                                                       118
######## Which Team won by maximum wickets? (We ca see SRH won by 9 wickets)
max run <- matches[which.max(matches$win by wickets),]</pre>
select(max run, winner, win by wickets)
###### Output > winner win_by_wickets
                      38 Sunrisers Hyderabad
                                                            9
matches%>%
  group_by(winner)%>%
  summarize(most_win = n())%>%
  ggplot(aes(x = winner,y = most_win,fill = winner))+
  geom_bar(stat = "identity")+
  coord_flip()+
  scale_y_continuous("Matches won")
       Sunrisers Hyderabad -
  Royal Challengers Bangalore -
          Rajasthan Royals -
                                                                           winner
                                                                               Chennai Super Kings
          Mumbai Indians -
                                                                               Delhi Capitals
                                                                               Kings XI Punjab
       Kolkata Knight Riders -
                                                                               Kolkata Knight Riders
                                                                               Mumbai Indians
          Kings XI Punjab -
                                                                               Rajasthan Royals
                                                                               Royal Challengers Bangalore
                                                                               Sunrisers Hyderabad
            Delhi Capitals -
       Chennai Super Kings -
                                          Matches won
```

```
teams <- data %>% select(batting team)%>%
  distinct()
teams <- rename(teams, team = batting team)</pre>
teams
###### Output >
                                              (following teams played in IPL 2019)
                              team
           1 Royal Challengers Bangalore
                     Chennai Super Kings
           3
                     Sunrisers Hyderabad
           4
                   Kolkata Knight Riders
           5
                           Delhi Capitals
                          Mumbai Indians
           6
           7
                          Kings XI Punjab
                         Rajasthan Royals
s team <- c("RCB","CSK","SRH","KKR","DC","MI","KXIP","RR")</pre>
s_team
##### OUTPUT > [1] "RCB" "CSK" "SRH" "KKR" "DC" "MI" "KXIP" "RR"
teams <- cbind(teams, s_team)</pre>
player_of_match <- matches%>% select(id,player_of_match,season) %>%
  distinct()
player_of_match <- rename(player_of_match, player=player_of_match)</pre>
matches$city <- as.character(matches$city)</pre>
matches$city[matches$city==""] <- "Dubai"</pre>
venue city <- matches %>%
  select(city)%>%
  distinct()
```

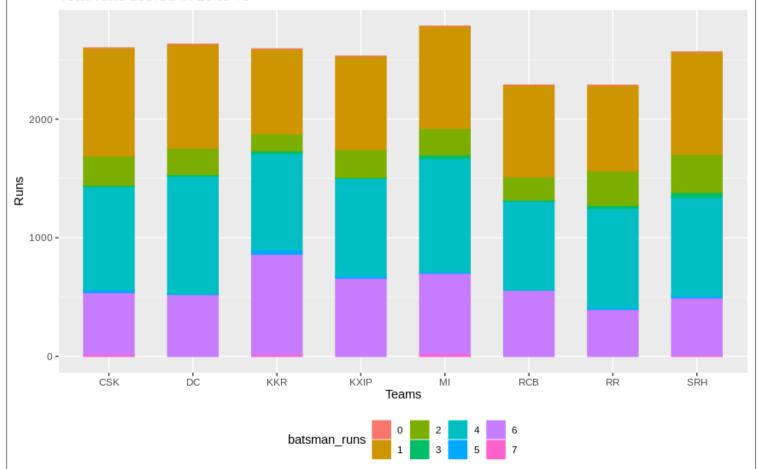
```
dismissal <- data%>%
 left join(matches, by=c("match id"="id"))%>%
 left join(teams,by=c("batting team"="team"))%>%
 filter(dismissal kind!="")%>%
 group by(season,dismissal kind,s team)%>%
 summarize(wickets =n())
ggplot(dismissal,aes(x=dismissal kind,y=wickets,colour=as.factor(season),
fill=as.factor(season)))+
 geom bar(position = "stack", show.legend = TRUE, width =.6,stat="identity")+
 theme(legend.position="bottom")+
 coord_flip()+
 theme(legend.direction = "horizontal") +
 scale_y_continuous(name="wickets")+
 scale_x_discrete(name="dismissal kind")+
 ggtitle("Breakdown of dismissal type ")
```

#### Breakdown of dismissal type



(We ca see in above plot that maximum dismissal was happened due to caught)

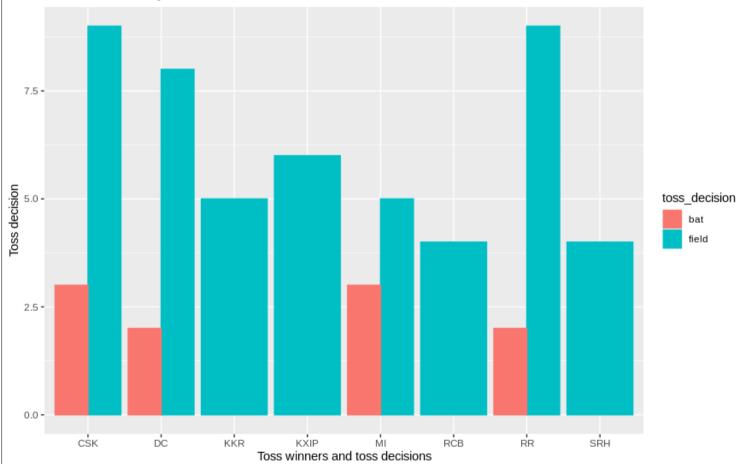




(We ca see in above plot that most of the runs were scored in 1st, 3rd and 6th ball)

Toss decisions by each Team

ggtitle("Toss decisions by each Team")

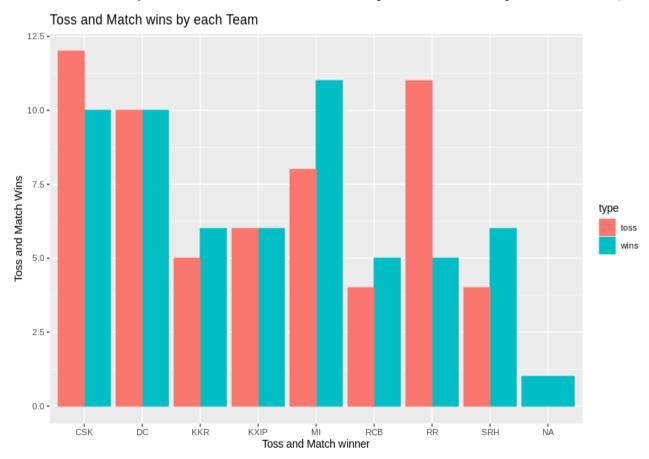


(We can see that CSK and RR choosen fielding after winning the toss and KKR, KXIP, SRH never batted first after winning the toss)

```
toss <- matches%>%
  left_join(teams,by=c("toss_winner"="team") )%>%
```

```
select(s_team,toss_winner)%>%
  group_by(s_team)%>%
  summarize(wins=n())
toss$type <- "toss"
wins <-matches%>%
  left join(teams,by=c("winner"="team") )%>%
  select(s team,winner)%>%
 group by(s team)%>%
  summarize(wins=n())
wins$type <- "wins"
toss_w <- rbind(toss,wins)</pre>
toss_w <- toss_w %>%
  group by(s team, type)%>%
  summarize(wins=sum(wins))
ggplot(toss w,aes(x=s team,y=wins,colour=type,fill=type))+
  geom bar(position = "dodge",stat = "identity")+
  theme(legend.position="right")+
  scale y continuous(name="Toss and Match Wins")+
  scale x discrete(name="Toss and Match winner")+
  ggtitle("Toss and Match wins by each Team")
```

#### (We can see in the below plot that DC and XXIP won every match when they won the toss)



```
venue_c <- data%>%
  left_join(matches,by=c("match_id"="id"))%>%
  select(match id,city,total runs,wickets)%>%
  group by(city)%>%
  summarize(runs=sum(total runs),wickets=sum(wickets,na.rm=TRUE))
city mat <- matches %>%
 group by(city)%>%
  summarize(matches=n())
venue_c <- venue_c %>%
  left_join(city_mat, by=c("city"="city"))%>%
 mutate(Avg_runs=runs/matches)%>%
  mutate(Avg wkt =wickets/matches)%>%
  arrange(city)
venue all <- venue c%>%
  left join(venue city, by=c("city"="city"))%>%
  arrange(Avg_runs)
venue_all$city <- factor(venue_all$city, levels = venue_all$city[order(venue_all$matches)])</pre>
ggplot(venue all,aes(x=city,y=matches,colour=city,fill=city))+
  geom_bar(position = "dodge",stat = "identity")+
  theme(legend.position="none")+ coord_flip()+
  scale_y_continuous(name="Total no of Matches in each city")+
  scale_x_discrete(name="Cities ")+
 ggtitle("Cities with most no of matches")
                   Cities with most no of matches
             Hyderabad ·
              Chennai ·
               Kolkata ·
```



(We can see in the above plot that most of the maximum of 9 matches were played in Chennain and Hyderabad)

# Analysis on Batsman of IPL 2019 by giving priorities to their performance measures

```
most runs <- read.csv("/home/dheeraj/Desktop/Lecture/6th sem Academics/DataScience/Lab2/</pre>
batting stats.csv")
head(most runs)
########### Ordering According to Priority ##############
a <- most runs[order(-most runs$RUNS),]</pre>
b <- a[order(-most runs$AVG),]</pre>
c <- b[order(most runs$INN),]</pre>
d <- c[order(-most_runs$SR),]</pre>
e <- d[order(-most_runs$X4S),]</pre>
 (I gave priority in the following order by highest runs scored, maximum avg of the player,
           minimum innings played, highest Strike rate, maximum number of fours)
########## Player who topped the list #################
select(head(e, n=1), PLAYER, RUNS)
###### Output > PLAYER
              David Warner
######### Player who scored maximum runs ###############
max_run <- e[which.max(e$RUNS),]</pre>
select(max run, PLAYER)
###### Output >
                     PLAYER
              David Warner
######### Player who've highest Strike Rate ############
max_sr <- e[which.max(e$SR),]</pre>
select(max_sr, PLAYER)
###### Output >
                    PLAYER
              Andre Russell
```

```
########## Player who hit highest 4rs ##################
max fours <- e[which.max(e$X4S),]</pre>
select(max fours, PLAYER)
###### Output > PLAYER
            Shikhar Dhawan
max avg <- e[which.max(e$AVG),]</pre>
select(max avg, PLAYER)
###### Output >
                 PLAYER
              MS Dhoni
########## Player who hit highest Sixes #################
max_sixes <- e[which.max(e$X6S),]</pre>
select(max sixes, PLAYER)
###### Output >
                   PLAYER
               Andre Russell
########## Player who played minimum match #############
min_match <- e[which.min(e$MATCHES),]</pre>
select(min_match, PLAYER)
###### Output >
                   PLAYER
            K Khaleel Ahmed
select(head(e, n=10), PLAYER)
###### Output >
                     PLAYER
                David Warner
            1
                Lokesh Rahul
            2
            3
             Shikhar Dhawan
           4 Jonny Bairstow
           5
                 Shreyas Iyer
           6 Ajinkya Rahane
           7 Quinton de Kock
           8
             Hardik Pandya
           9
                   MS Dhoni
                Shane Watson
```

```
######## Top ten player's with their data in my list #########
select(head(e, n=10), PLAYER,INN, RUNS,
                                            AVG,
                                                     SR ,X4S, X6S )
###### Output >
                        PLAYER INN RUNS
                                           AVG
                                                   SR X4S X6S
                 David Warner 12 692 69.20 143.87
Lokesh Rahul 14 593 53.91 135.39
           2
                                                       49
                                                           25
           3
               Shikhar Dhawan 16 521 34.73 135.68 64 11
              Jonny Bairstow 10 445 55.62 157.24 48 18
                 Shreyas Iver 16 463 30.87 119.95 41 14
           6 Ajinkya Rahane 13
                                   393 32.75 137.89 45
              Quinton de Kock 16 529 35.27 132.91 45 25
              Hardik Pandya 15 402 44.67 191.43
MS Dhoni 12 416 83.20 134.63
                                                          29
                                                      28
                                                      22
                                                          23
           10
                 Shane Watson 17 398 23.41 127.56 42 20
```

# Analysis on Bowlers of IPL 2019 by giving priorities to their performance measures

```
bowling stats <- read.csv("/home/dheeraj/Desktop/Lecture/6th sem Academics/DataScience/Lab2/
bowling stats.csv")
head(bowling_stats)
########### Ordering According to Priority ###############
a <- bowling_stats[order(-bowling_stats$WKTS),]</pre>
b <- a[order(bowling stats$BALLS),]</pre>
c <- b[order(bowling stats$MATCHES),]</pre>
e <- c[order(-bowling_stats$RUNS),]</pre>
  (Priority given in the following order by maximum wickets taken, minimum balls throw by
                    him , minimum match played, and runs given by him)
########## Player who topped the list #################
select(head(e, n=1), PLAYER)
###### Output >
                     PLAYER
                Imran Tahir
######## Player who taken maximum wicket ##############
max wkt <- e[which.max(e$WKTS),]</pre>
select(max_wkt, PLAYER)
###### Output >
                    PLAYER
                 Imran Tahir
```

```
######### Player who've thrown maximum balls ############
max run <- e[which.max(e$BALLS),]</pre>
select(max run, PLAYER)
###### Output >
                      PLAYER
                Deepak Chahar
########## Player who gave minimum runs #################
max run <- e[which.min(e$RUNS),]</pre>
select(max run, PLAYER)
###### Output >
                   PI AYFR
               Amit Mishra
########## Player who played minimum match #############
max run <- e[which.min(e$MATCHES),]</pre>
select(max run, PLAYER)
###### Output >
                     PLAYER
              K Khaleel Ahmed
select(head(e, n=10), PLAYER)
###### Output >
                     PLAYER
                   Imran Tahir
             1
                   Axar Patel
             2
             3 K Khaleel Ahmed
                Mohammed Shami
             4
             5
                  Amit Mishra
               Navdeep Saini
             6
             7
               Ishant Sharma
             8
                 Shreyas Gopal
             9
                   Sam Curran
             10
                   Kagiso Rabada
######## Top ten player's with their data in my list #########
select(head(e, n=10), PLAYER, MATCHES, BALLS, RUNS, WKTS )
###### Output >
                     PLAYER MATCHES BALLS RUNS WKTS
                Imran Tahir
                                    386 431
                               17
                                               26
         1
         2
                Axar Patel
                               14
                                    306 364
                                               10
            K Khaleel Ahmed
                               9
                                    209 287
                                               19
         4
             Mohammed Shami
                               14
                                    324
                                        469
                                               19
         5
               Amit Mishra
                                    240
                                        270
                               11
                                               11
             Navdeep Saini
         6
                               13
                                    288
                                         397
                                               11
         7
             Ishant Sharma
                               13
                                    276
                                         349
                                               13
                                        347
         8
              Shreyas Gopal
                              14
                                    288
                                               20
         9
                Sam Curran
                               9
                                    198 323
                                               10
                               12
         10
               Kagiso Rabada
                                    282
                                        368
                                               25
```

# Analysis of 10 Players and their 10 match individual scores finding median, consistency, corelation between few pairs of Players

```
####### loading data of top 10 players and their individual 10 match score ######
ind ply <- read.csv("/home/dheeraj/Desktop/Lecture/6th sem Academics/DataScience/Lab2/</pre>
player_eachmatch.csv")
View(ind ply)
median(ind ply$DavidWarner)
###### Output > [1] 54
median(ind_ply$LokeshRahul)
###### Output > [1] 33.5
median(ind ply$ShikharDhawan)
###### Output > [1] 32.5
median(ind ply$JonnyBairstow)
####### Output > [1] 43
median(ind ply$ShreyasIyer)
###### Output > [1] 35.5
median(ind_ply$AjinkyaRahane)
###### Output > [1] 20
median(ind_ply$QuintondeKock )
###### Output > [1] 31
median(ind_ply$HardikPandya)
###### Output > [1] 26.5
median(ind_ply$MSDhoni)
###### Output > [1] 37
median(ind_ply$ShaneWatson )
###### Output > [1] 9.5
```

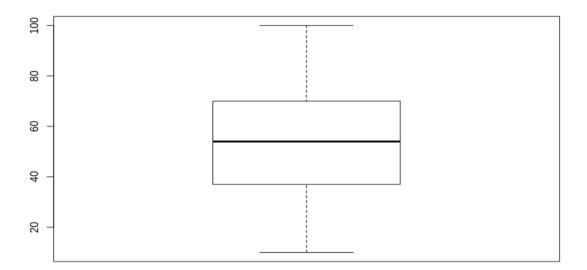
# Below is consitency of each players (player which have values near to 0 are more consistent and which're near to 1 are inconsistent)

```
####### Finding Coefficient of variance ##########
sd(ind ply$LokeshRahul)/mean(ind ply$DavidWarner)
###### Output > [1] 0.6536159
sd(ind ply$LokeshRahul)/mean(ind ply$LokeshRahul)
###### Output > [1] 0.9039726
sd(ind_ply$ShikharDhawan)/mean(ind_ply$ShikharDhawan)
###### Output > [1] 0.8334738
sd(ind ply$JonnyBairstow)/mean(ind ply$JonnyBairstow)
###### Output > [1] 0.7892764
sd(ind_ply$ShreyasIyer)/mean(ind_ply$ShreyasIyer)
###### Output > [1] 0.66783
sd(ind ply$AjinkyaRahane)/mean(ind ply$AjinkyaRahane)
###### Output > [1] 1.052911
sd(ind_ply$QuintondeKock )/mean(ind_ply$QuintondeKock )
###### Output > [1] 0.6318267
sd(ind_ply$HardikPandya)/mean(ind_ply$HardikPandya)
###### Output > [1] 0.4513528
sd(ind_ply$MSDhoni)/mean(ind_ply$MSDhoni)
###### Output > [1] 0.6289972
sd(ind_ply$ShaneWatson )/mean(ind_ply$ShaneWatson )
###### Output > [1] 1.002898
```

# Box plots for each 10 players We can see median from the plot at mid line is median

############ PLOT 1: Box plot for David Warner ##############

boxplot(ind\_ply\$DavidWarner)
num = as.numeric(ind\_ply\$DavidWarner)
outvalues = boxplot(num)\$out
which(ind\_ply\$DavidWarner %in% outvalues)

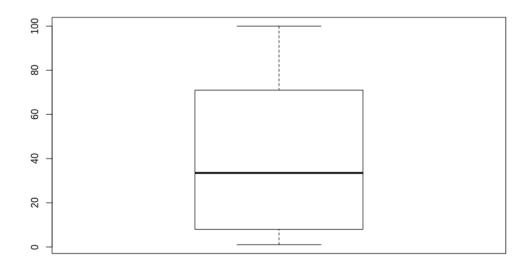


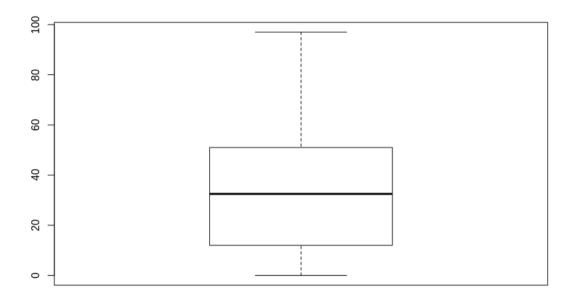
boxplot(ind\_ply\$LokeshRahul)

num = as.numeric(ind\_ply\$LokeshRahul)

outvalues = boxplot(num)\$out

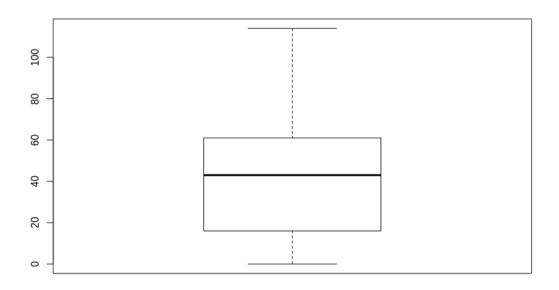
which(ind\_ply\$LokeshRahul %in% outvalues)

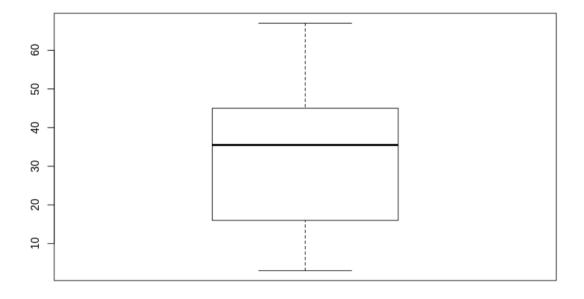




### ############ PLOT 4: Box plot for JonnyBairstow ##############

boxplot(ind\_ply\$JonnyBairstow)
num = as.numeric(ind\_ply\$JonnyBairstow)
outvalues = boxplot(num)\$out
which(ind\_ply\$JonnyBairstow %in% outvalues)





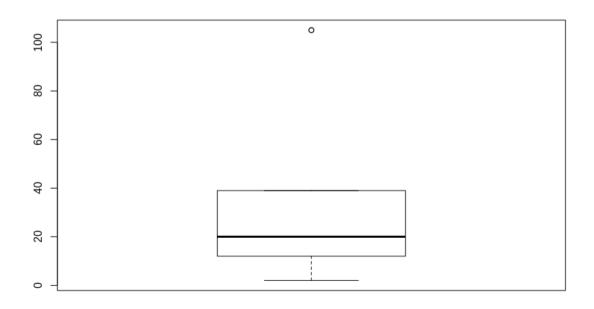
#### 

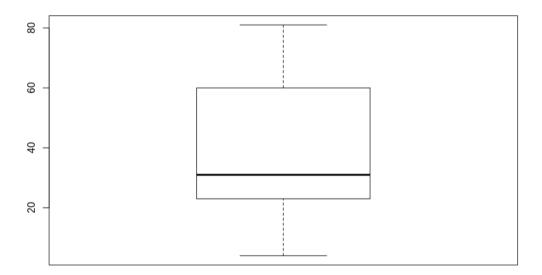
boxplot(ind\_ply\$AjinkyaRahane)

num = as.numeric(ind\_ply\$AjinkyaRahane)

outvalues = boxplot(num)\$out

which(ind\_ply\$AjinkyaRahane %in% outvalues) (We can see an Outliar here i.e, at score 2)





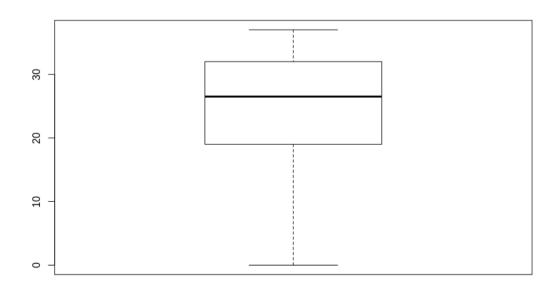
#### ############ PLOT 8: Box plot for HardikPandya ##############

boxplot(ind\_ply\$HardikPandya)

num = as.numeric(ind\_ply\$HardikPandya)

outvalues = boxplot(num)\$out

which(ind\_ply\$HardikPandya %in% outvalues)

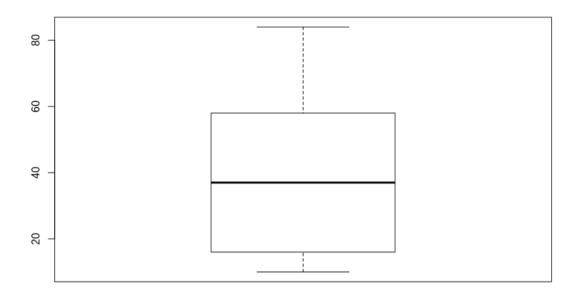


```
boxplot(ind_ply$MSDhoni)
```

num = as.numeric(ind\_ply\$MSDhoni)

outvalues = boxplot(num)\$out

which(ind\_ply\$MSDhoni %in% outvalues)



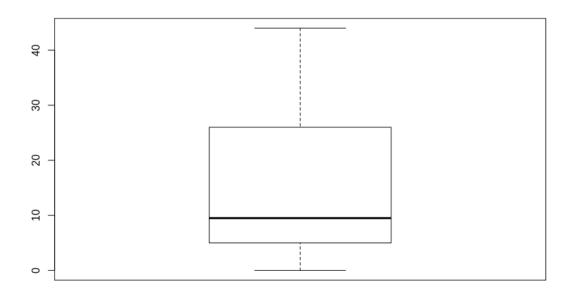
#### ############ PLOT 10: Box plot for ShaneWatson ################

boxplot(ind\_ply\$ShaneWatson)

num = as.numeric(ind\_ply\$ShaneWatson)

outvalues = boxplot(num)\$out

which(ind\_ply\$ShaneWatson %in% outvalues)



## Correlation between two few pairs of players

```
####### corelation between Davidwarner and Lokesh rahul ##########
cor.test(ind ply$DavidWarner, ind ply$LokeshRahul, method = "spearman")
###### Output >
                         Spearman's rank correlation rho
data: ind_ply$DavidWarner and ind_ply$LokeshRahul
S = 150.96, p-value = 0.8152
alternative hypothesis: true rho is not equal to 0
sample estimates:
       rho
0.08510678
####### corelation between ShikharDhawan and JonnyBairstow ###########
cor.test(ind ply$ShikharDhawan, ind ply$JonnyBairstow, method = "spearman")
###### Output >
                       Spearman's rank correlation rho
data: ind_ply$ShikharDhawan and ind_ply$JonnyBairstow
S = 188, p-value = 0.7072
alternative hypothesis: true rho is not equal to 0
sample estimates:
       rho
-0.1393939
####### corelation between ShreyasIyer and AjinkyaRahane ##########
cor.test(ind_ply$ShreyasIyer, ind_ply$AjinkyaRahane, method = "spearman")
###### Output >
                       Spearman's rank correlation rho
data: ind_ply$ShreyasIyer and ind_ply$AjinkyaRahane
S = 248.52, p-value = 0.1355
alternative hypothesis: true rho is not equal to 0
sample estimates:
       rho
-0.5062114
####### corelation between QuintondeKock and HardikPandya ###########
cor.test(ind ply$QuintondeKock, ind ply$HardikPandya, method = "spearman")
###### Output >
                         Spearman's rank correlation rho
data: ind_ply$QuintondeKock and ind_ply$HardikPandya
S = 120.87, p-value = 0.455
alternative hypothesis: true rho is not equal to 0
sample estimates:
      гhо
0.2674784
####### corelation between MSDhoni warner and ShaneWatson ##########
cor.test(ind_ply$MSDhoni, ind_ply$ShaneWatson, method = "spearman")
###### Output >
                        Spearman's rank correlation rho
data: ind_ply$MSDhoni and ind_ply$ShaneWatson
S = 189.22, p-value = 0.6857
alternative hypothesis: true rho is not equal to 0
sample estimates:
       rho
-0.1467897
```