

# Roshan Shrestha Final Self Reflection

[Link to my Shiny app](#)

[Link to my Github](#)

## Q. Import, manage, and clean data.

In my shiny project there was many instants we I had to load csv files. I have loaded the csv dataset then cleaned the data set by removing row with NA and empty values and separated into different csv files as per my need in shiny project. I can also import different types of files by different ways.

*Importing csv file:*

```
fifa <- read.csv("Fifa.csv", header = T)
head(fifa)
```

```
##      short_name      long_name player_positions
## 1      L. Messi      Lionel Andr  s Messi Cuccittini      RW, ST, CF
## 2      R. Lewandowski      Robert Lewandowski      ST
## 3 Cristiano Ronaldo Cristiano Ronaldo dos Santos Aveiro      ST, LW
## 4      Neymar Jr      Neymar da Silva Santos J  nior      LW, CAM
## 5      K. De Bruyne      Kevin De Bruyne      CM, CAM
## 6      J. Oblak      Jan Oblak      GK
## overall potential value_eur wage_eur age      dob height_cm weight_kg
## 1      93      93 78000000 320000 34 6/24/1987      170      72
## 2      92      92 119500000 270000 32 8/21/1988      185      81
## 3      91      91 45000000 270000 36 02/05/1985      187      83
## 4      91      91 129000000 270000 29 02/05/1992      175      68
## 5      91      91 125500000 350000 30 6/28/1991      181      70
## 6      91      93 112000000 130000 28 01/07/1993      188      87
##      club_name Club_Rank      league_name club_position
## 1 Paris Saint-Germain      6      French Ligue 1      RW
## 2 FC Bayern M  nchen      1      German 1. Bundesliga      ST
## 3 Manchester United      9      English Premier League      ST
## 4 Paris Saint-Germain      6      French Ligue 1      LW
## 5 Manchester City      2      English Premier League      RCM
## 6 Atl  tico de Madrid      10      Spain Primera Division      GK
## club_jersey_number club_joined club_contract_valid_until Country_Rank
## 1      30 08/10/2021      2023      4
## 2      9 07/01/2014      2023      28
## 3      7 8/27/2021      2023      8
## 4      10 08/03/2017      2025      2
## 5      17 8/30/2015      2025      1
## 6      13 7/16/2014      2023      64
##      Rough Rough.1 nationality_name      Continent nation_position
## 1      Afghanistan      Asia      Argentina      South America      RW
## 2      Aland Islands      Europe      Poland      Europe      RS
## 3      Albania      Europe      Portugal      Europe      ST
## 4      Algeria      Africa      Brazil      South America
## 5      American Samoa      Oceania      Belgium      Europe      RCM
```

```
## 6      Andorra Europe      Slovenia      Europe
## nation_jersey_number preferred_foot weak_foot skill_moves
## 1          10      Left      4      4
## 2           9      Right     4      4
## 3           7      Right     4      5
## 4          NA      Right     5      5
## 5           7      Right     5      4
## 6          NA      Right     3      1
## international_reputation release_clause_eur pace shooting passing dribbling
## 1           5      144300000 85      92      91      95
## 2           5      197200000 78      92      79      86
## 3           5      833000000 87      94      80      88
## 4           5      238700000 91      83      86      94
## 5           4      232200000 76      86      93      88
## 6           5      238000000 NA      NA      NA      NA
## defending physic
## 1      34      65
## 2      44      82
## 3      34      75
## 4      37      63
## 5      64      78
## 6      NA      NA
```

Importing Excel file:

```
library(readxl)
fifaxl = read_excel("Fifa.xlsx")
```

```
## New names:
## * Rough -> Rough...20
## * Rough -> Rough...21
```

```
fifaxl
```

```
## # A tibble: 20 x 36
## short_name long_name player_positions overall potential value_eur wage_eur
## <chr> <chr> <chr> <dbl> <dbl> <dbl> <dbl>
## 1 L. Messi "Lionel An~ RW, ST, CF 93 93 78000000 320000
## 2 R. Lewando~ "Robert Le~ ST 92 92 119500000 270000
## 3 Cristiano ~ "Cristiano~ ST, LW 91 91 45000000 270000
## 4 Neymar Jr "Neymar da~ LW, CAM 91 0 129000000 270000
## 5 K. De Bruy~ "Kevin De ~ CM, CAM 91 91 125500000 350000
## 6 J. Oblak "Jan Oblak" GK 91 93 112000000 130000
## 7 K. Mbappe "Kylian Mb~ ST, LW 91 95 194000000 230000
## 8 M. Neuer "Manuel Pe~ GK 90 90 13500000 86000
## 9 M. ter Ste~ "Marc-Andr~ GK 90 NA 99000000 250000
## 10 H. Kane "Harry Kan~ ST 90 90 129500000 240000
## 11 N. Kante "N'Golo Ka~ CDM, CM 90 90 100000000 230000
## 12 K. Benzema "Karim Ben~ CF, ST 89 89 66000000 350000
## 13 T. Courtois "Thibaut C~ GK 89 91 85500000 250000
## 14 H. Son "i†\u0090i~ LM, CF, LW 89 89 104000000 220000
## 15 Casemiro "Carlos He~ CDM 89 89 88000000 310000
## 16 V. van Dijk "Virgil va~ CB 89 89 86000000 230000
## 17 S. Mane "Sadio Man~ LW 89 89 101000000 270000
## 18 M. Salah "Mohamed S~ RW 89 89 101000000 270000
## 19 Ederson "Ederson S~ GK 89 91 94000000 200000
```

```
## 20 J. Kimmich "Joshua Wa~ CDM, RB 89 90 108000000 160000
## # ... with 29 more variables: age <dbl>, dob <chr>, height_cm <dbl>,
## # weight_kg <dbl>, club_name <chr>, Club_Rank <chr>, league_name <chr>,
## # club_position <chr>, club_jersey_number <dbl>, club_joined <chr>,
## # club_contract_valid_until <dbl>, Country_Rank <dbl>, Rough...20 <chr>,
## # Rough...21 <chr>, nationality_name <chr>, Continent <chr>,
## # nation_position <chr>, nation_jersey_number <dbl>, preferred_foot <chr>,
## # weak_foot <dbl>, skill_moves <dbl>, international_reputation <dbl>,
## # release_clause_eur <dbl>, pace <dbl>, shooting <dbl>, passing <dbl>,
## # dribbling <dbl>, defending <dbl>, physic <dbl>
```

*Import files by here::here :*

```
library(here)
```

```
## here() starts at /home/shrestro/518 Stat/Project_R/Final
```

```
dt <- readr::read_csv(here::here('data', 'Fifa.csv'))
```

```
## Warning: Duplicated column names deduplicated: 'Rough' => 'Rough_1' [21]
```

```
##
## -- Column specification -----
## cols(
##   .default = col_double(),
##   short_name = col_character(),
##   long_name = col_character(),
##   player_positions = col_character(),
##   dob = col_character(),
##   club_name = col_character(),
##   Club_Rank = col_character(),
##   league_name = col_character(),
##   club_position = col_character(),
##   club_joined = col_character(),
##   Rough = col_character(),
##   Rough_1 = col_character(),
##   nationality_name = col_character(),
##   Continent = col_character(),
##   nation_position = col_character(),
##   preferred_foot = col_character()
## )
## i Use `spec()` for the full column specifications.
```

*Importing of image by using here::here.*

```
library(imager)
```

```
## Loading required package: magrittr
```

```
##
```

```
## Attaching package: 'imager'
```

```
## The following object is masked from 'package:magrittr':
```

```
##
```

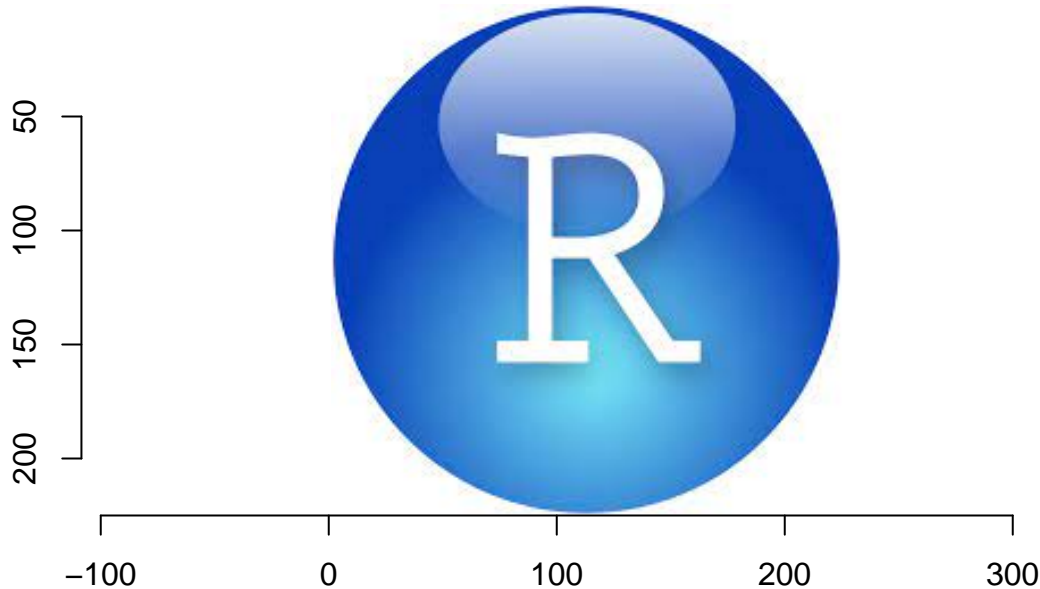
```
## add
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## convolve, spectrum
```

```
## The following object is masked from 'package:graphics':
##
##   frame
## The following object is masked from 'package:base':
##
##   save.image
img <- jpeg::readJPEG(here::here('images','r.jpeg'))
img1<-load.image('~/.518 Stat/Project_R/Final/images/r.jpeg')
plot(img1)
```



I am also able to write cvs from the data frame where we can easily download the the csv file:

```
write.csv(fifaxl,"~/518 Stat/Project_R/fifaxl.csv", row.names = TRUE)
```

### Clean the data set.

I am able to clean the data by removing the NA, empty values and zero values from the dataset. For my shiny app I have imported many files from various sources and clean the data as per the requirement of the project.

Removing the row with NA values, zero values and empty cell.

```
fifaxl = fifaxl[!(is.na(fifaxl$potential) | fifaxl$potential==" " | fifaxl$potential=="0" ),]
fifaxl
```

```
## # A tibble: 18 x 36
##   short_name long_name  player_positions overall potential value_eur wage_eur
##   <chr>      <chr>      <chr>          <dbl>    <dbl>    <dbl>    <dbl>
## 1 L. Messi   "Lionel An~ RW, ST, CF      93      93  78000000  320000
## 2 R. Lewando~ "Robert Le~ ST              92      92  119500000 270000
## 3 Cristiano ~ "Cristiano~ ST, LW          91      91   45000000 270000
## 4 K. De Bruy~ "Kevin De ~ CM, CAM      91      91  125500000 350000
## 5 J. Oblak   "Jan Oblak" GK              91      93  112000000 130000
## 6 K. Mbappe  "Kylia~ ST, LW        91      95  194000000 230000
## 7 M. Neuer   "Manuel Pe~ GK              90      90   13500000  86000
## 8 H. Kane    "Harry Kan~ ST              90      90  129500000 240000
```

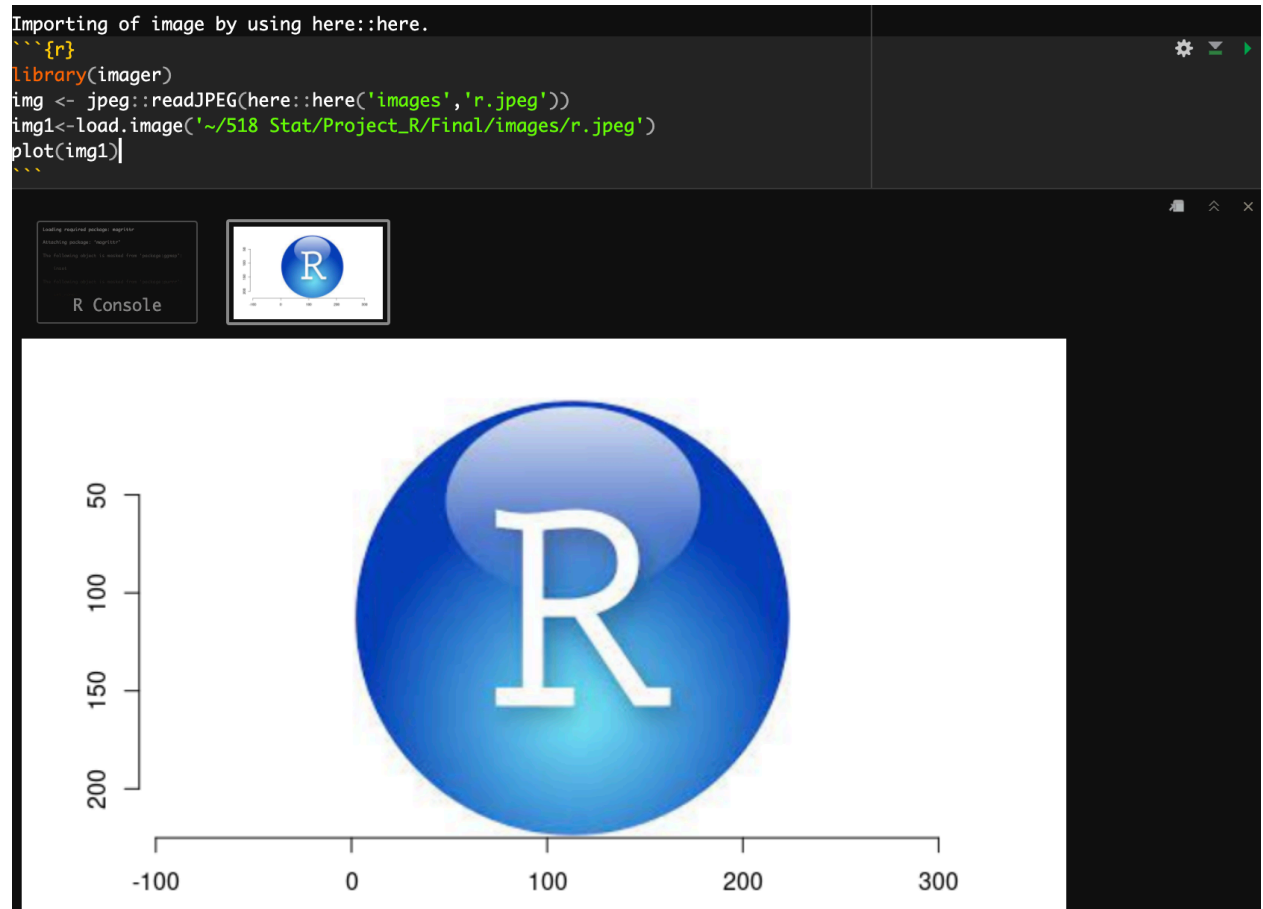


Figure 1: Image

```
## 9 N. Kante "N'Golo Ka~ CDM, CM 90 90 100000000 230000
## 10 K. Benzema "Karim Ben~ CF, ST 89 89 660000000 350000
## 11 T. Courtois "Thibaut C~ GK 89 91 855000000 250000
## 12 H. Son "i†\u0090i~ LM, CF, LW 89 89 104000000 220000
## 13 Casemiro "Carlos He~ CDM 89 89 880000000 310000
## 14 V. van Dijk "Virgil va~ CB 89 89 860000000 230000
## 15 S. Mane "Sadio Man~ LW 89 89 101000000 270000
## 16 M. Salah "Mohamed S~ RW 89 89 101000000 270000
## 17 Ederson "Ederson S~ GK 89 91 940000000 200000
## 18 J. Kimmich "Joshua Wa~ CDM, RB 89 90 108000000 160000
## # ... with 29 more variables: age <dbl>, dob <chr>, height_cm <dbl>,
## # weight_kg <dbl>, club_name <chr>, Club_Rank <chr>, league_name <chr>,
## # club_position <chr>, club_jersey_number <dbl>, club_joined <chr>,
## # club_contract_valid_until <dbl>, Country_Rank <dbl>, Rough...20 <chr>,
## # Rough...21 <chr>, nationality_name <chr>, Continent <chr>,
## # nation_position <chr>, nation_jersey_number <dbl>, preferred_foot <chr>,
## # weak_foot <dbl>, skill_moves <dbl>, international_reputation <dbl>,
## # release_clause_eur <dbl>, pace <dbl>, shooting <dbl>, passing <dbl>,
## # dribbling <dbl>, defending <dbl>, physic <dbl>
```

## Manage the data

I can isolation with in the data set by using dplyr function where I used select, filter, groupby, summarize and mutate.

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
## filter, lag

## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union
```

```
country <- fifaxl %>%
  select(nationality_name, wage_eur) %>%
  group_by(nationality_name) %>%
  summarize(Avg_Wages = mean(wage_eur, na.rm = TRUE))
```

```
head(country)
```

```
## # A tibble: 6 x 2
## nationality_name Avg_Wages
## <chr> <dbl>
## 1 Argentina 320000
## 2 Belgium 300000
## 3 Brazil 255000
## 4 Egypt 270000
## 5 England 240000
## 6 France 270000
```

In the above code, I am able to group the data set by nation name and calculate the average wages of each country with the help of dplyr and data pipelines.

I am able to write write loop function to calculate the mean from the given data set.

Writing loop to calculate mean of overall ranking of players.

```
my_sum = 0
len = length(fifaxl$overall)
for(i in 1:len){
  my_sum = my_sum + fifaxl$overall[i]
}
mean_overall = my_sum/len
mean_overall
```

```
## [1] 90
```

To avoid writing complicated and long code I am also able to write map function .

I am also able to use map function for finding the mean of the data

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --

## v ggplot2 3.3.4      v purrr 0.3.4
## v tibble 3.1.2       v stringr 1.4.0
## v tidyr 1.1.3        v forcats 0.5.1
## v readr 1.4.0

## -- Conflicts ----- tidyverse_conflicts() --
## x imager::add()      masks magrittr::add()
## x stringr::boundary() masks imager::boundary()
## x tidyr::extract()   masks magrittr::extract()
## x tidyr::fill()      masks imager::fill()
## x dplyr::filter()    masks stats::filter()
## x dplyr::lag()        masks stats::lag()
## x purrr::set_names() masks magrittr::set_names()

map1 <- fifaxl %>% select(overall,wage_eur) %>% map_dbl(mean)
map1

## overall wage_eur
## 90.0 243666.7
```

Here I have merge the data of two different csv to form one data frame.

```
one <- read.csv("one.csv")
two <- read.csv("two.csv")
# one
# two

total <- merge(one,two,by="short_name")
head(total)

##      short_name      long_name player_positions
## 1      Casemiro Carlos Henrique Venancio Casimiro CDM
## 2 Cristiano Ronaldo Cristiano Ronaldo dos Santos Aveiro ST, LW
## 3      Ederson      Ederson Santana de Moraes GK
## 4      H. Kane      Harry Kane ST
## 5      H. Son      i†\u0090i\u009d¥ë~¼ å å...`æ...æ LM, CF, LW
## 6      J. Kimmich      Joshua Walter Kimmich CDM, RB
## overall potential value_eur wage_eur age
```

```
## 1      89      89 88000000 310000 29
## 2      91      91 45000000 270000 36
## 3      89      91 94000000 200000 27
## 4      90      90 129500000 240000 27
## 5      89      89 104000000 220000 28
## 6      89      90 108000000 160000 26
```

According to the requirement of shiny app project I have used pivot wider and longer to manage the data as per the need of the project.

Below is the sample code where pivot wider is used to make more columns.

```
league <- read.csv("league.csv")
league
```

```
##           league Avg_Wages_league year
## 1           EPL          57433.79 2018
## 2 French Ligue 1          20421.40 2018
## 3      Bundesliga          30632.46 2018
## 4      Serie A          35257.07 2018
## 5      LaLiga          34658.25 2018
## 6           EPL          53981.82 2019
## 7 French Ligue 1          19522.43 2019
## 8      Bundesliga          25016.45 2019
## 9      Serie A          32028.21 2019
## 10      LaLiga          33810.32 2019
## 11          EPL          51273.03 2020
## 12 French Ligue 1          20263.94 2020
## 13      Bundesliga          27253.46 2020
## 14      Serie A          30241.71 2020
## 15      LaLiga          34279.67 2020
## 16          EPL          52107.80 2021
## 17 French Ligue 1          19515.58 2021
## 18      Bundesliga          24501.00 2021
## 19      Serie A          26623.64 2021
## 20      LaLiga          32397.91 2021
## 21          EPL          50847.70 2022
## 22 French Ligue 1          21462.74 2022
## 23      Bundesliga          24407.71 2022
## 24      Serie A          31004.53 2022
## 25      LaLiga          31128.83 2022
```

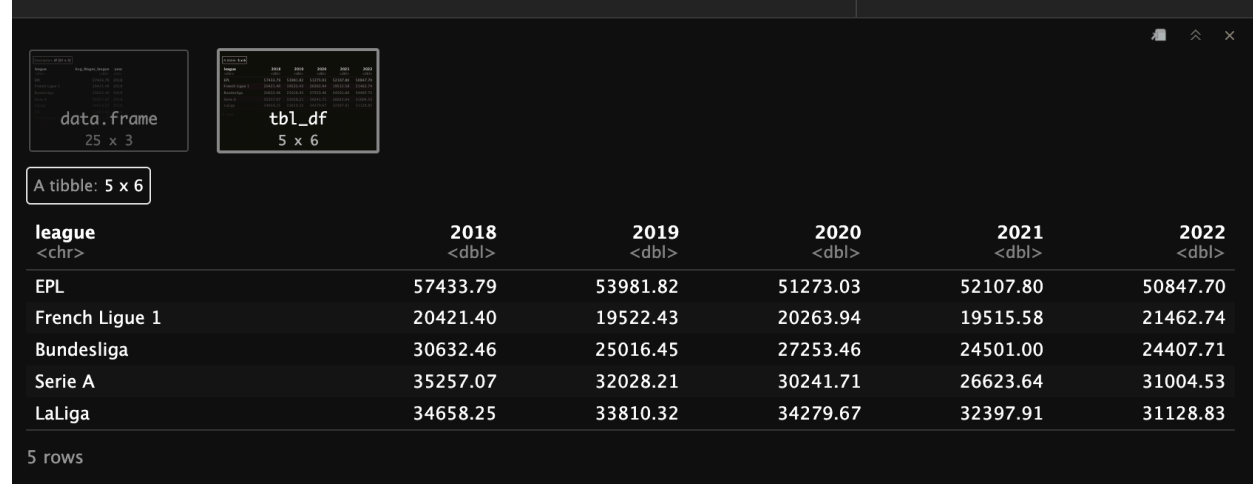
```
wide_league <- pivot_wider(league, names_from = year, values_from = Avg_Wages_league)
wide_league
```

```
## # A tibble: 5 x 6
##   league   `2018` `2019` `2020` `2021` `2022`
##   <chr>     <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 EPL      57434. 53982. 51273. 52108. 50848.
## 2 French Ligue 1 20421. 19522. 20264. 19516. 21463.
## 3 Bundesliga    30632. 25016. 27253. 24501. 24408.
## 4 Serie A      35257. 32028. 30242. 26624. 31005.
## 5 LaLiga      34658. 33810. 34280. 32398. 31129.
```



Below is the sample code where pivot wider is used to make more columns.

```
library(tidyverse)
league <- read.csv("league.csv")
wide_league <- pivot_wider(league, names_from = year, values_from = Avg_Wages_league)
```



The screenshot shows the RStudio interface. At the top, the R console contains the code to read a CSV file and pivot it wider. Below the console, two data frame visualizations are shown: a 'data.frame' with dimensions 25 x 3 and a 'tbl\_df' with dimensions 5 x 6. The main area displays a tibble with 5 rows and 6 columns, representing the pivoted data.

| league<br><chr> | 2018<br><dbl> | 2019<br><dbl> | 2020<br><dbl> | 2021<br><dbl> | 2022<br><dbl> |
|-----------------|---------------|---------------|---------------|---------------|---------------|
| EPL             | 57433.79      | 53981.82      | 51273.03      | 52107.80      | 50847.70      |
| French Ligue 1  | 20421.40      | 19522.43      | 20263.94      | 19515.58      | 21462.74      |
| Bundesliga      | 30632.46      | 25016.45      | 27253.46      | 24501.00      | 24407.71      |
| Serie A         | 35257.07      | 32028.21      | 30241.71      | 26623.64      | 31004.53      |
| LaLiga          | 34658.25      | 33810.32      | 34279.67      | 32397.91      | 31128.83      |

5 rows

Below I have use Stringr package to sepeate the values by coma.

```
library("stringr")
fifastr <- read.csv("fifastr.csv")
fifastr
```

```
##           short_name           player_tags
## 1      L. Messi      #Dribbler, #Distance Shooter
## 2    R. Lewandowski  #Aerial Threat, #Distance Shooter
## 3 Cristiano Ronaldo      #Aerial Threat, #Dribbler
## 4      Neymar Jr      #Speedster, #Dribbler
## 5      K. De Bruyne      #Dribbler, #Playmaker
## 6      K. Mbappé      #Speedster, #Dribbler
## 7      H. Kane #Distance Shooter, #Clinical Finisher
## 8      H. Son      #Dribbler, #Engine
## 9      Casemiro      #Engine, #Tackling
## 10     V. van Dijk      #Tackling, #Tactician
## 11     S. Mané      #Speedster, #Dribbler
## 12     M. Salah      #Speedster, #Dribbler
```

```
str_split_fixed(fifastr$player_tags, ",", 2)
```

```
##           [,1]           [,2]
## [1,] "#Dribbler"      " #Distance Shooter"
## [2,] "#Aerial Threat" " #Distance Shooter"
## [3,] "#Aerial Threat" " #Dribbler"
## [4,] "#Speedster"     " #Dribbler"
## [5,] "#Dribbler"     " #Playmaker"
## [6,] "#Speedster"     " #Dribbler"
## [7,] "#Distance Shooter" " #Clinical Finisher"
## [8,] "#Dribbler"     " #Engine"
## [9,] "#Engine"       " #Tackling"
## [10,] "#Tackling"    " #Tactician"
## [11,] "#Speedster"   " #Dribbler"
## [12,] "#Speedster"   " #Dribbler"
```

Below I have use Stringr package to sepeate the values by coma.

```
```{r}
```

```
library("stringr")
fifastr <- read.csv("fifastr.csv")
fifastr
str_split_fixed(fifastr$player_tags,",", 2)
```

```
```
```

```
data.frame
12 x 2
```

```
R Console
```

|       | [,1]                | [,2]                  |
|-------|---------------------|-----------------------|
| [1,]  | "#Dribbler"         | " #Distance Shooter"  |
| [2,]  | "#Aerial Threat"    | " #Distance Shooter"  |
| [3,]  | "#Aerial Threat"    | " #Dribbler"          |
| [4,]  | "#Speedster"        | " #Dribbler"          |
| [5,]  | "#Dribbler"         | " #Playmaker"         |
| [6,]  | "#Speedster"        | " #Dribbler"          |
| [7,]  | "#Distance Shooter" | " #Clinical Finisher" |
| [8,]  | "#Dribbler"         | " #Engine"            |
| [9,]  | "#Engine"           | " #Tackling"          |
| [10,] | "#Tackling"         | " #Tactician"         |
| [11,] | "#Speedster"        | " #Dribbler"          |
| [12,] | "#Speedster"        | " #Dribbler"          |

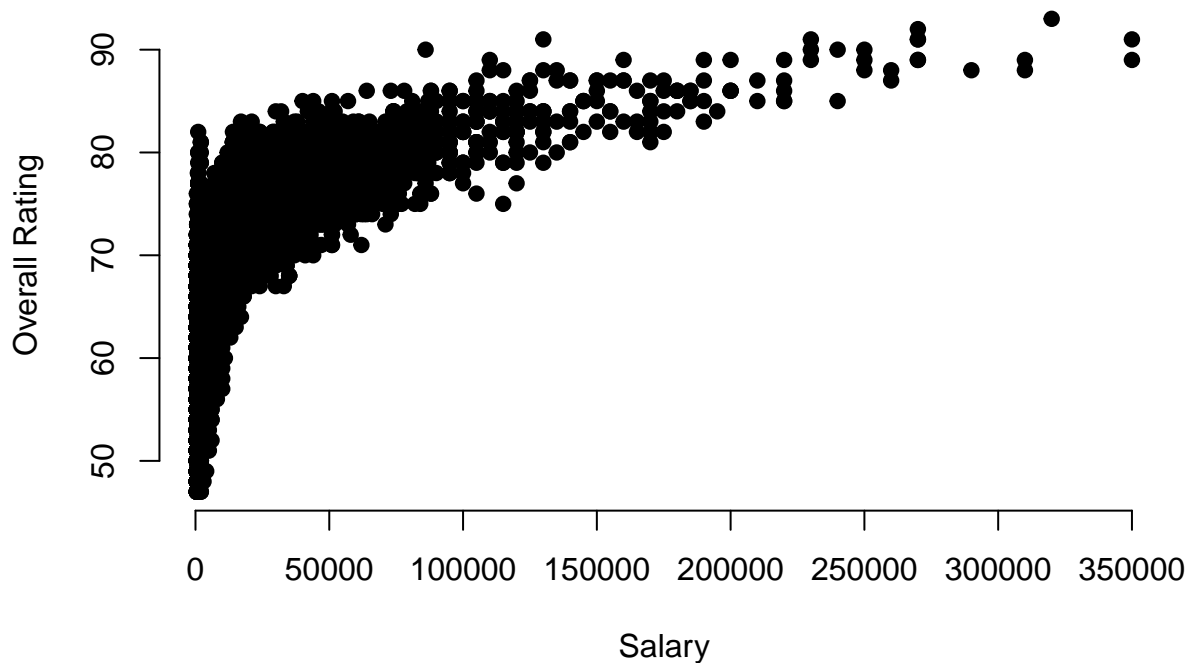
Figure 2: shinyr

Q. Create graphical displays and numerical summaries of data for exploratory analysis and presentations.

ggplot2 is a popular package used to make graphical diagram such as bar graph, line graph and many more graphs. By using ggplot2 I have made Bar graph, line graph, and scatter plot.

```
x <- fifa$wage_eur
y <- fifa$overall
plot(x, y, main = "Scatter Plot of overall rating vs wage",
     xlab = "Salary", ylab = "Overall Rating",
     pch = 19, frame = FALSE)
```

**Scatter Plot of overall rating vs wage**



```
bar<-read.csv("Bargraph.csv")
g <- ggplot(bar, aes( y = Avg_Wages_league, x = league,fill=league))
g+ geom_bar(stat='identity') +scale_x_discrete(guide = guide_axis(check.overlap = TRUE))
```

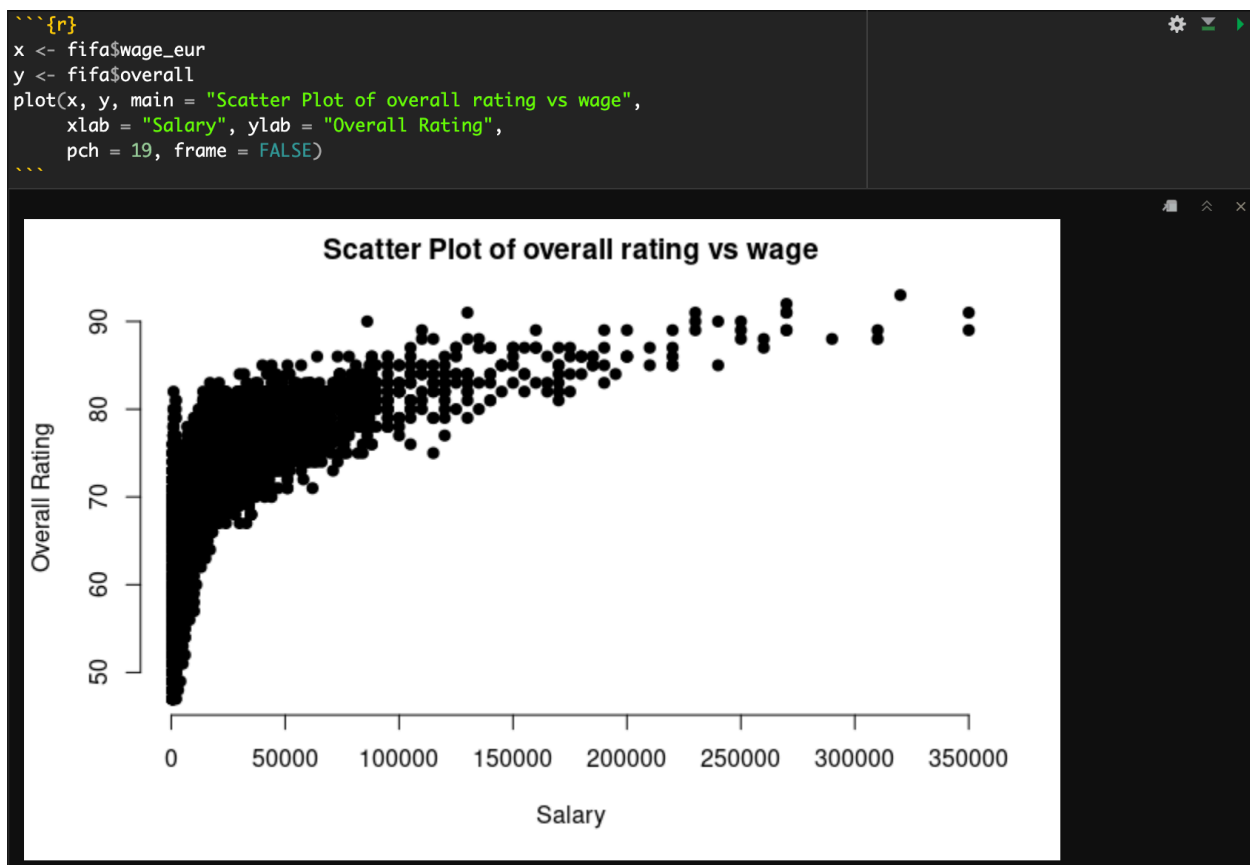
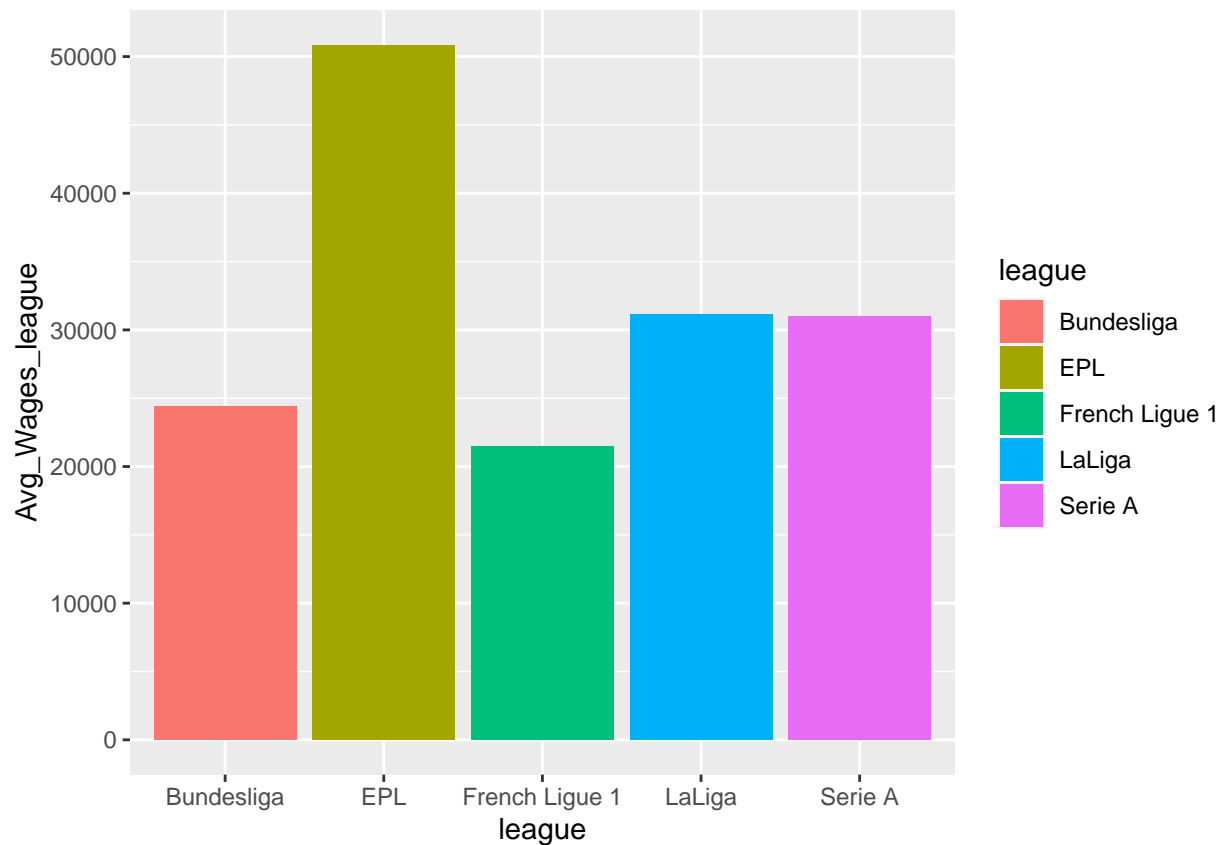


Figure 3: scatter



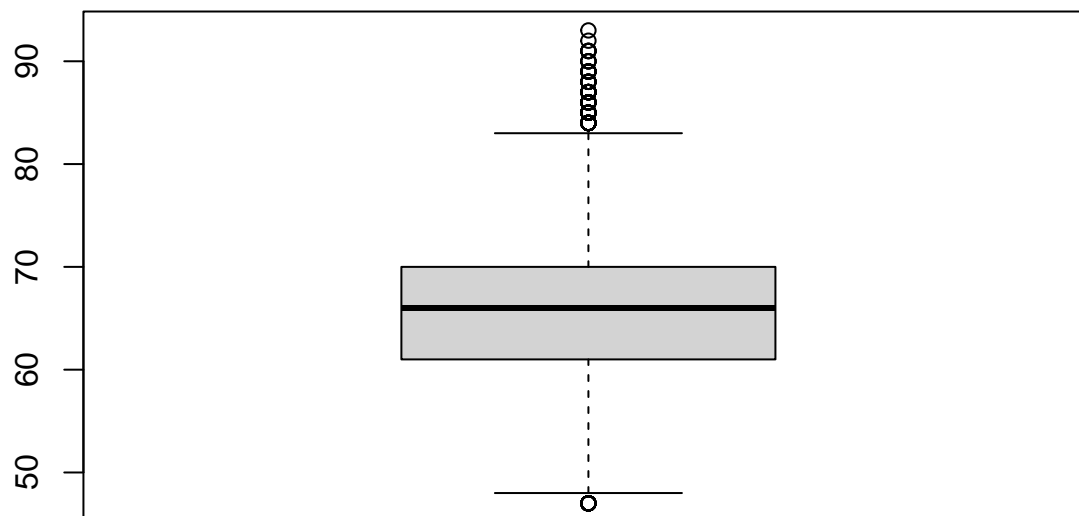
summary function is helpful in getting insight of the data where we get to know the median, maximum, minimum value.

```
summary(fifa$overall)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  47.00   61.00   66.00   65.77   70.00   93.00
```

We can also know median, maximum, minimum values from box plot.

```
boxplot(fifa$overall)
```



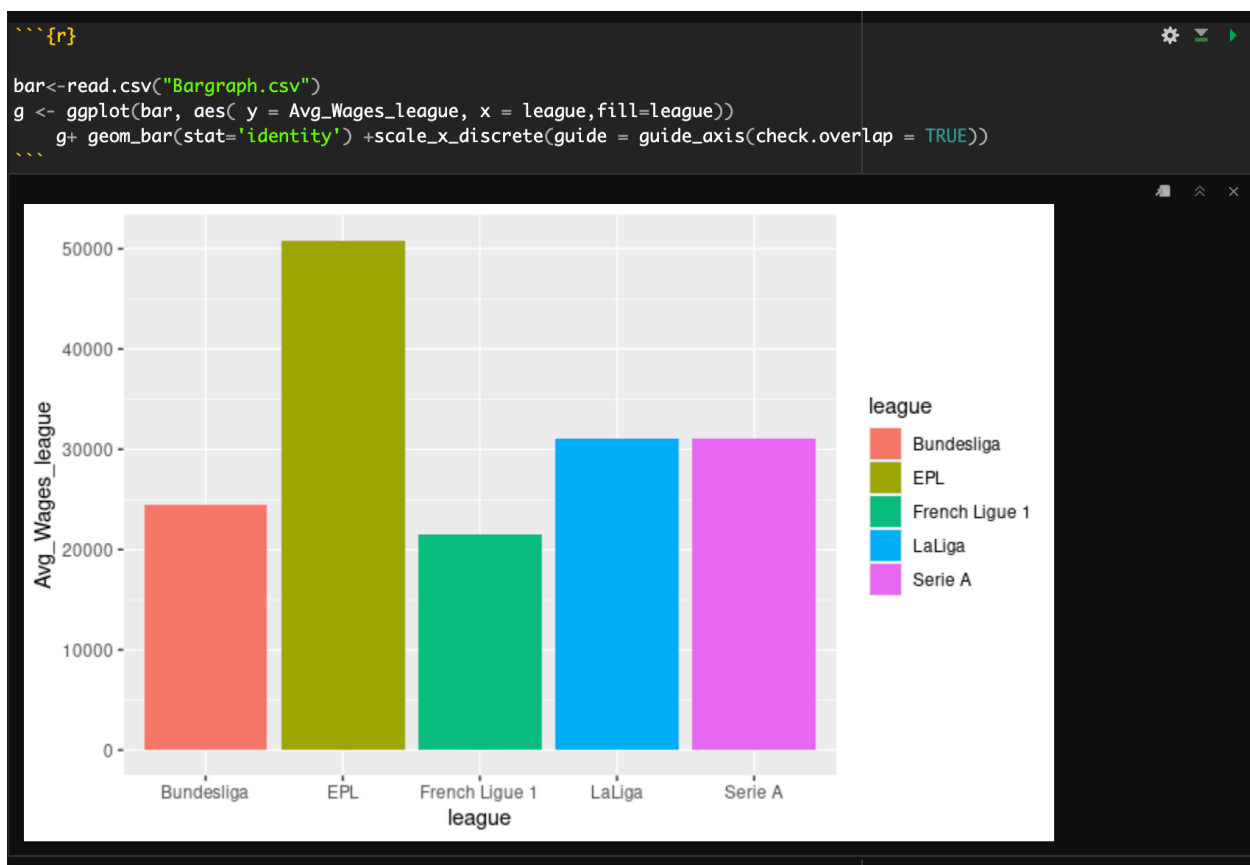


Figure 4: bargraph

we can also know median, maximum, minimum values from box plot.

```
{r}  
boxplot(fifa$overall)  
{r}
```

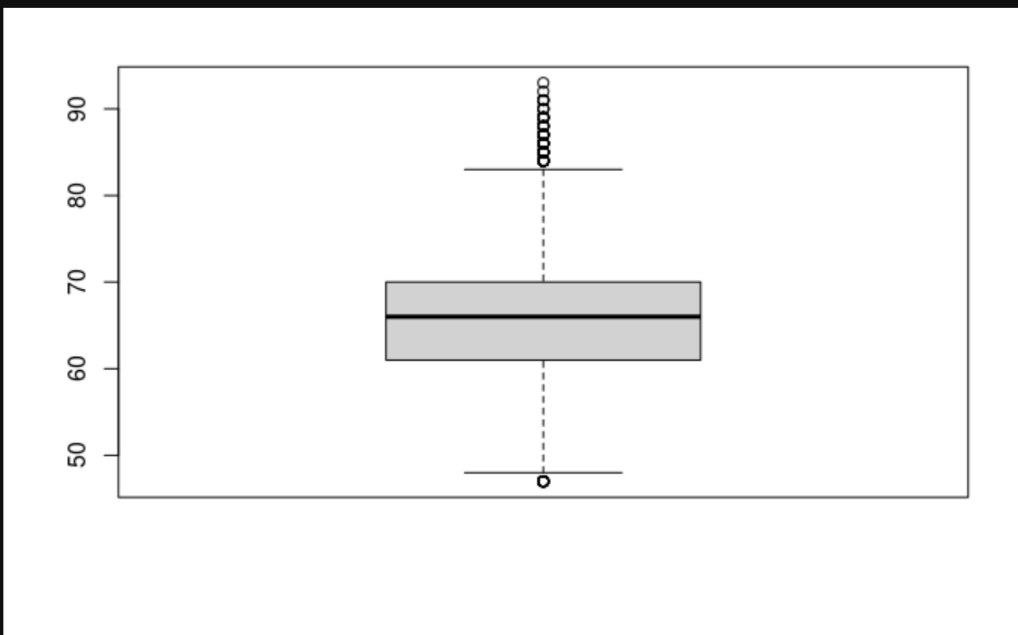


Figure 5: boxplot

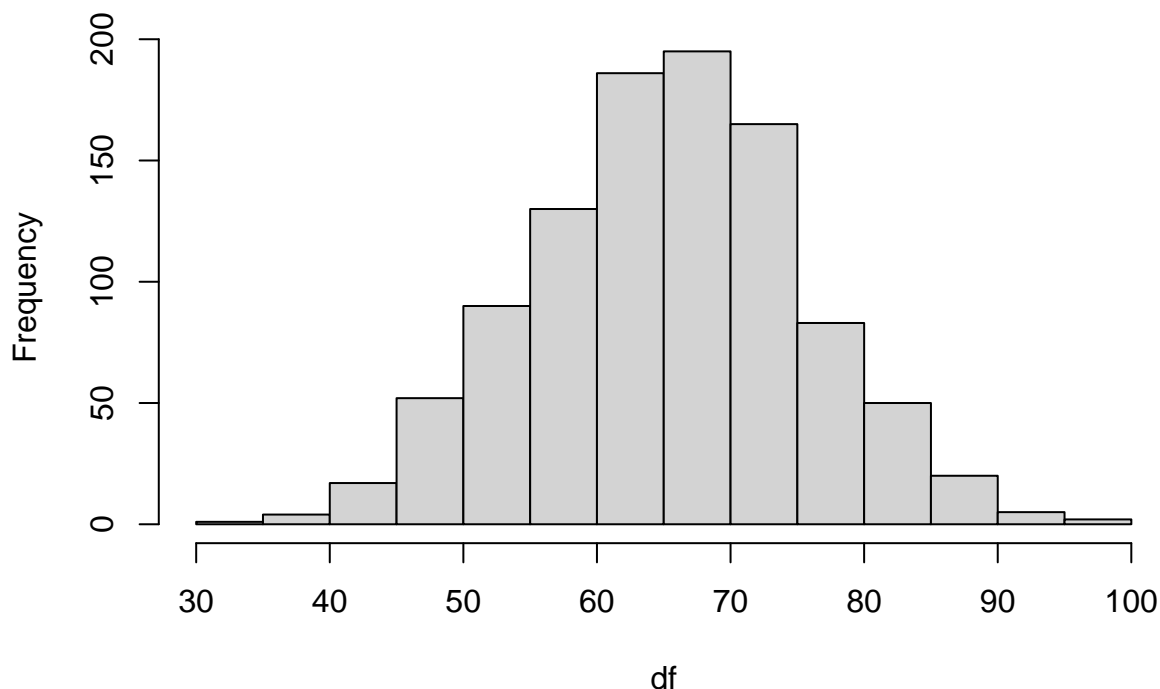
### Q. Analysis by simulation and bootstrapping

With simulation I can also create data similar to fifa overall dataset and perform hypothesis testing.

```
df <- rnorm(1000, mean = 65, sd = 10)
```

```
hist(df)
```

## Histogram of df



Here we have create random normal variable with mean 65 and stranded deviation 10 and shown the data in his-togram.

```
summary(df)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  33.97   58.60   65.42   65.29   72.11   96.28
```

Now lets create a sample from the df variable for our hypothesis testing. Where we will calculate the mean of the sample data and than compare it the df data frame mean. Our hypothesis would be H0: mean = 65.0928 and H1: mean 65.0928 .

```
set.seed(40)
```

```
sampldf = sample(df,100,replace = TRUE)
mean(sampldf)
```

```
## [1] 64.64108
```

```
sd(sampldf)
```

```
## [1] 9.867187
```

We can test the hypothesis by p-value with the mean and dataset of the sampldf to find if we accept the the hypothesis or fail to reject it at 95% coincidence interval.

```
t.test(sampldf,y = NULL, c("two.sided","less","greater"),mu = 65.0928, conf=0.95)
```

```
##
```

```
## One Sample t-test
```

```
##
```

```
## data:  sampldf
```

```
## t = -0.4578, df = 99, p-value = 0.6481
```

```
## alternative hypothesis: true mean is not equal to 65.0928
```



```
## 95 percent confidence interval:
## 62.68322 66.59894
## sample estimates:
## mean of x
## 64.64108
```

So p value is 0.1133 and the significance level is 0.05 where p- value is greater than confidence level. Hence we fail to reject the null hypothesis which means we accept the hypothesis H0.

```
summary(fifa$overall)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  47.00  61.00   66.00   65.77   70.00   93.00
```

Lets perform hypothesis testing on actual fifa 22 dataset

```
set.seed(520)
sample1 = sample(fifa$overall,50,replace = FALSE)
mean(sample(fifa$overall,50,replace = FALSE))
```

```
## [1] 66.64
```

```
sd(sample1)
```

```
## [1] 6.630203
```

```
mean(fifa$overall)
```

```
## [1] 65.77218
```

our hypothesis would be  $\text{mean}(H_0) = 65.77218$  and other hypothesis would be  $H_1$ : mean 65.77218.

```
t.test(sample1,y = NULL, c("two.sided","less","greater"),mu = 65.77218, conf=0.95)
```

```
##
## One Sample t-test
##
## data:  sample1
## t = 0.39228, df = 49, p-value = 0.6966
## alternative hypothesis: true mean is not equal to 65.77218
## 95 percent confidence interval:
## 64.25572 68.02428
## sample estimates:
## mean of x
## 66.14
```

So p value is 0.6966 and the significance level is 0.05 where p- value is greater than confidence level. Hence we fail to reject the null hypothesis which means we accept the hypothesis H0.

With the help of R I am also able to calculate the linear regression for the variable such as player wage and their overall rating.

```
fifa_reg <- lm(wage_eur~overall, data = fifax1)
summary(fifa_reg)
```

```
##
## Call:
## lm(formula = wage_eur ~ overall, data = fifax1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
```

```
## -157667 -20397 2872 34795 114795
##
## Coefficients:
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) -517872 1260703 -0.411 0.687
## overall 8462 14007 0.604 0.554
##
## Residual standard error: 71420 on 16 degrees of freedom
## Multiple R-squared: 0.0223, Adjusted R-squared: -0.03881
## F-statistic: 0.365 on 1 and 16 DF, p-value: 0.5542
```

*#bootstrapping:* My plan here is to take sample from overall player data than perform bootstrapping to those data to know the distribution of mean along with their frequency.

```
set.seed(520)
sample1 = sample(fifa$overall,50,replace = FALSE)
n <- length(sample1)
Boot <- 10000
```

Here I have taken 50 sample of overall player data from fifa dataset where replace is false, which means the data from the same row is not selected. I have set seed because set seed allows us to get same set of random values every time while executing the code and we do not get different means and different output.

```
Bootsample <- matrix(sample(sample1, size = n*Boot, replace=TRUE), nrow = n, ncol= Boot)
```

Here I have create a bootstrap which is 50 by 10000 matrix (where size = 50 \* 10000). Where 50 is number of row and 10000 is number of columns. I have taken sample from the sample1 data set which is again sample of 50 data from fifa overall player dataset. for the sample taken from sample1 in the Bootsampl I have set replace equals to true which means that repetition selection from the same rows are allowed while taking the sample to form the bootstrap.

```
dim(Bootsample)
```

```
## [1] 50 10000
```

We can see that we have created Bootsampl with 50 by 10000 matrix.

```
hist(Bootsample)
```

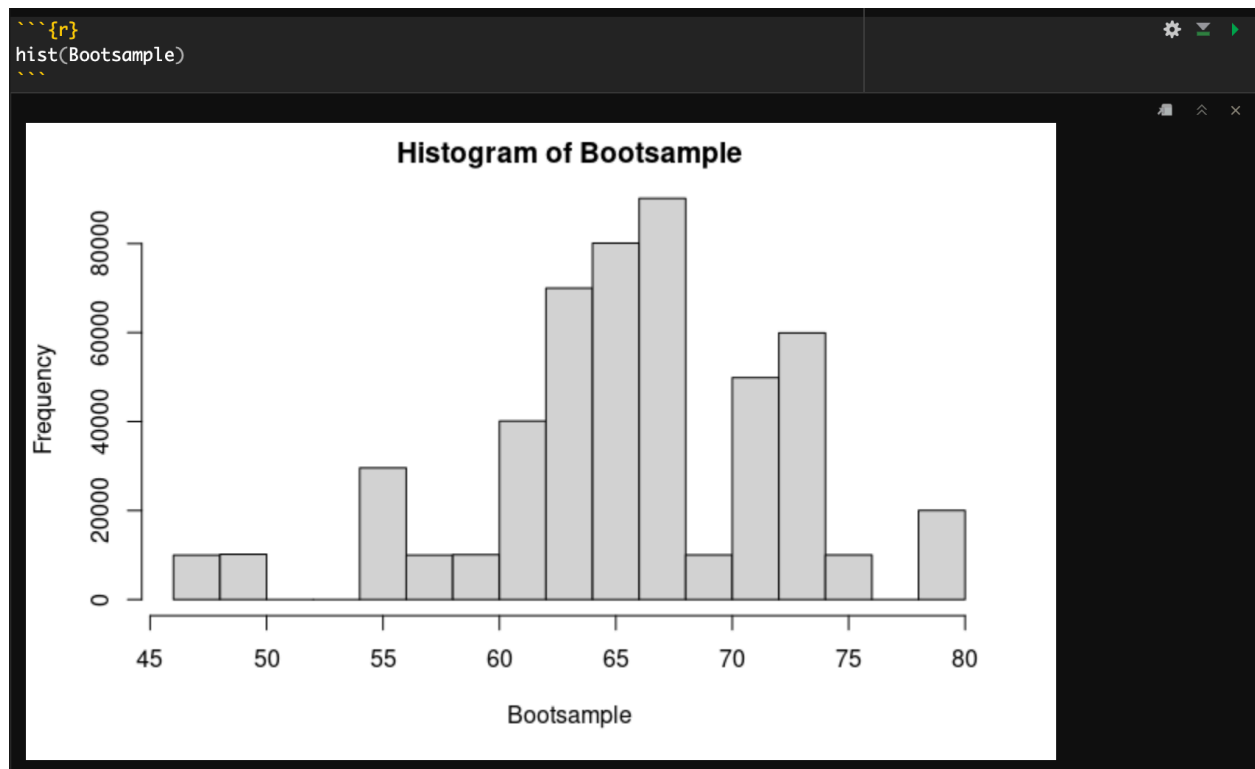
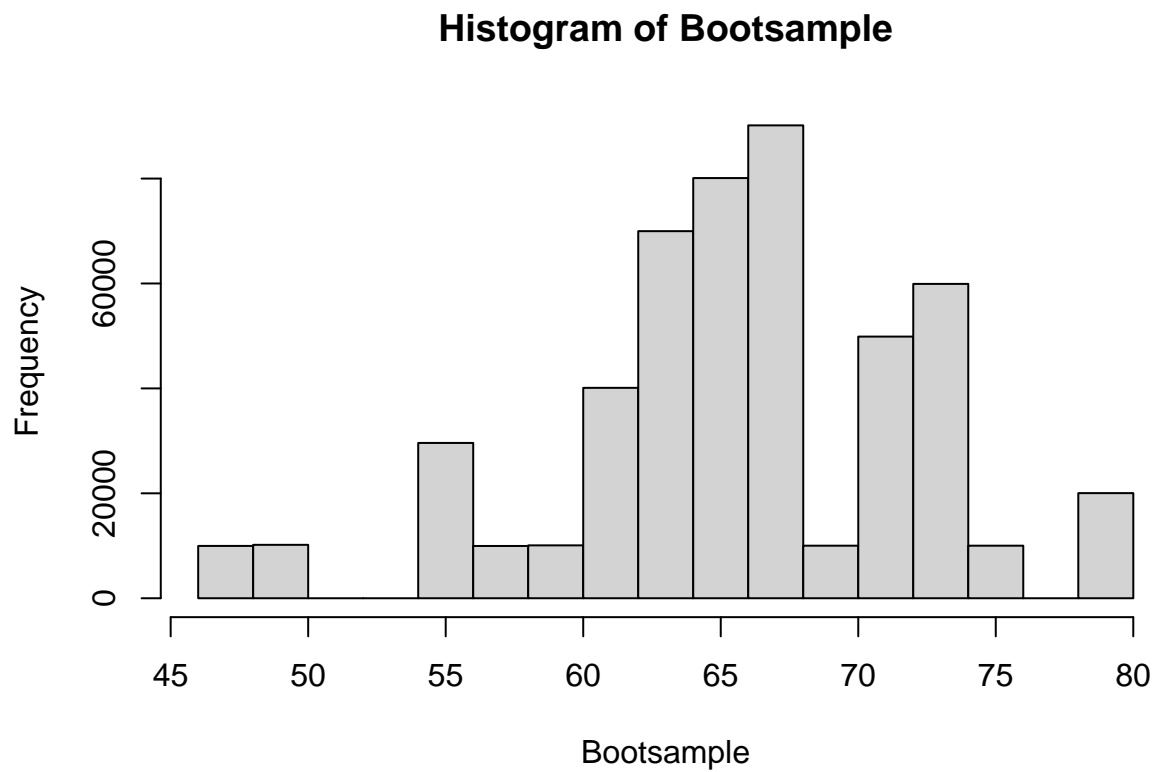


Figure 6: hist

I have created a histogram of the the Bootsample to get more insight about the data.

```
sort(sample1)
```

```
## [1] 47 50 55 55 56 57 60 61 62 62 62 63 63 64 64 64 64 65 65 65 66 66 66 66  
## [26] 66 67 67 67 67 68 68 68 68 68 70 71 71 71 71 72 73 73 73 73 74 74 76 79 80
```

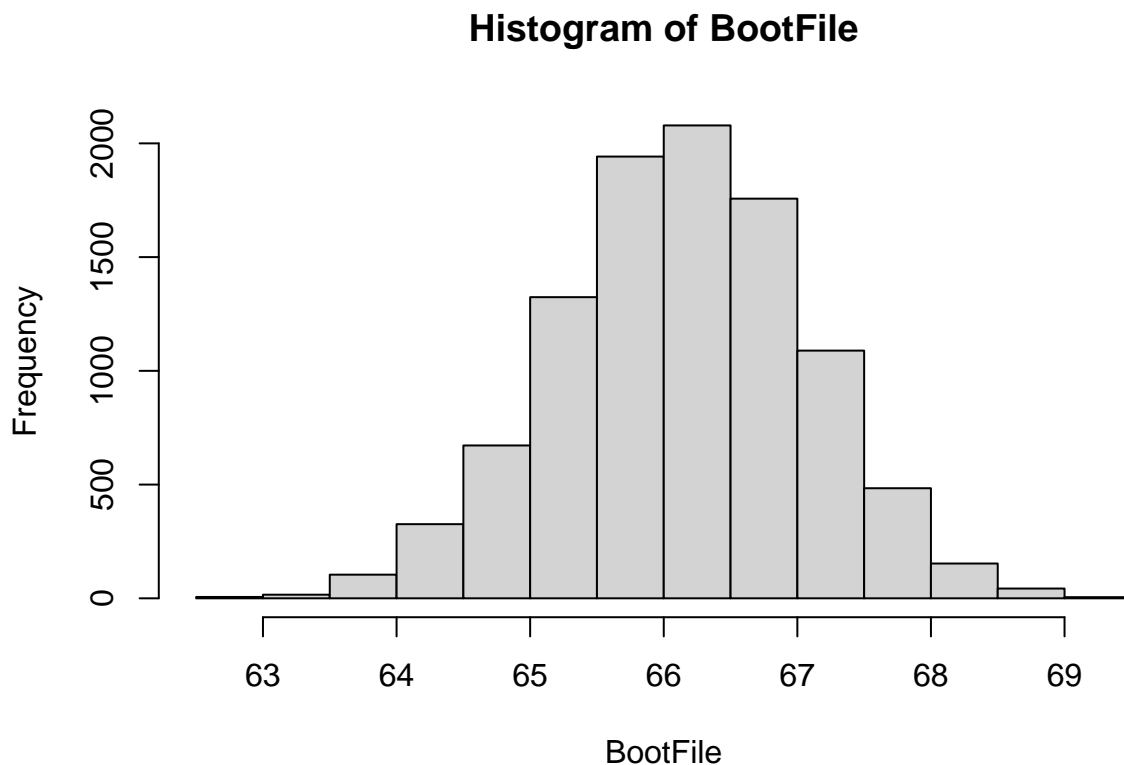
There is gap between 50 and 55 because in our data set sample1 we do not have the values between 50 and 55 and similar is the case between 76 and 79.

```
BootFile <- rep(0,Boot)  
for (i in 1:Boot){  
  BootFile[i] <- mean(Bootsample[1:50,i])  
}
```

I have created a vector with the name BootFile with 0 values. Again I have written code to calculate the mean from every 10000 columns and assign those values to BootFile.

This is the means of every 50 rows in the bootstrap.

```
#mean of 10000 Bootstrapping  
hist(BootFile)
```



Here is the histogram of the means from the bootstrapping.

```
sum(BootFile > 64 & BootFile < 68)
```

```
## [1] 9662
```

```
sum(BootFile > 65 & BootFile < 67)
```

```
## [1] 7054
```

```
sum(BootFile > 62.5 & BootFile < 69.5)
```

```
## [1] 9999
```

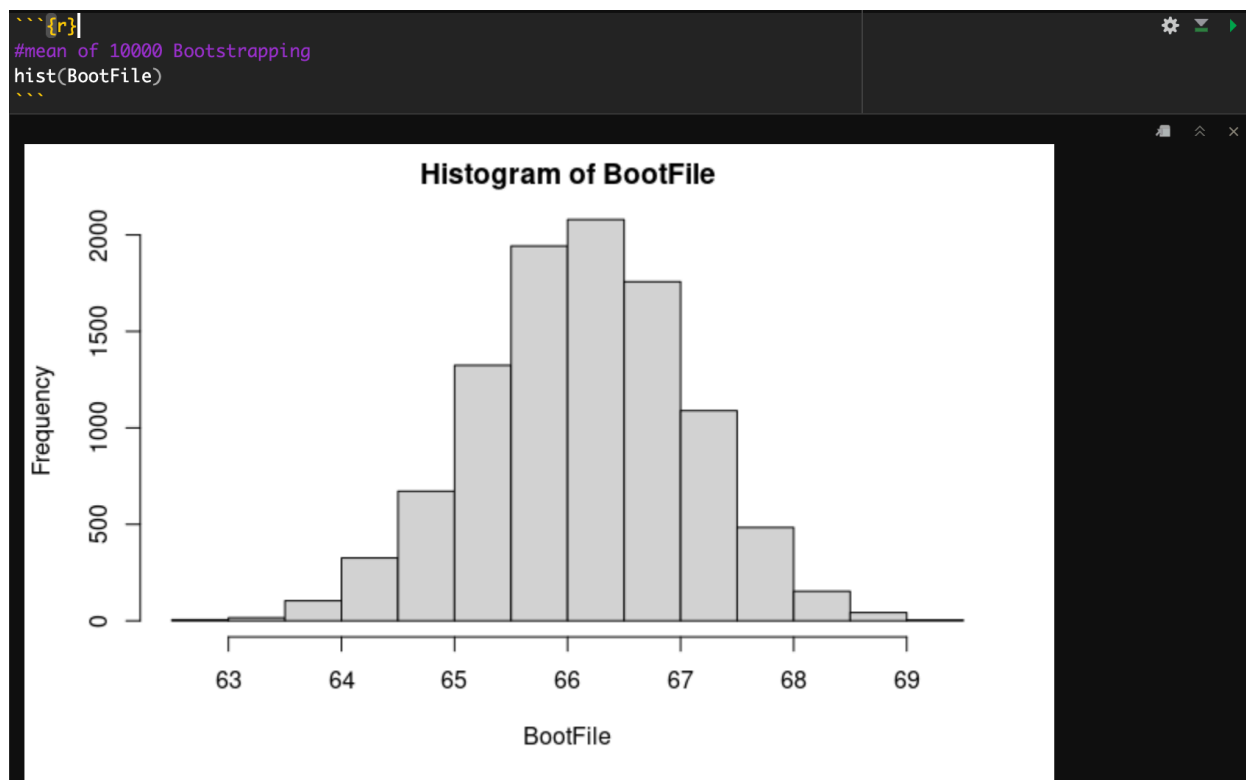


Figure 7: histboot

We can see that from the sample1 data when we calculate the the mean from the sample1 dataset when replace is true we can get mean between 62.5 and 69.5 but there is the 96.9% probability of getting mean between 64 and 68. We can also say that the data are normally distributed.

**Q. Use source documentation and other resources to troubleshoot and extend R programs.**

While learning, doing activity and project I had face my problem with the code and to solve it I refereed to r documentation, stackoverflow, and google to solve my issue. While solving my issue I got more insight in R and its syntax. Facing problem and reading documentation has help to develop my r knowledge and improve my understanding in R. Further I have also learn about many new packages like gganimate, esquisse and many more packages. Below is the snapshot of esquisse package used to make bar graph with drag and drop like tableau.

**Q. Write clear, efficient, and well-documented R programs.**

From the class activities and from R documentations online I have learned to write a well documented R code. I have also learn to write clean and effective code by using map, pivot, join, stringr, dplyr, tidyverse. The class activities has been effective and helped me grasp a knowledge of R. Additionally, working on my final project has strengthen my knowledge on working with R and Rmd files.

**Question 4:** Based on the progress you have made (i.e., see your response in (3)), what final grade would you give yourself for this course? Try to stick to the major grade levels (“A”, “B”, “C”, or “D or below”). Please reach out to me if you have concerns or were unable to finish your final project.

Since I have met all the objective, done the project, class activities and implemented it in this rmd file as well as in shiny app. I have given my best to learn R programming and performed many hundred line of code and learned from fixing the error while writing the code. Hence I will rate myself an “A” for the progress made through this semester.

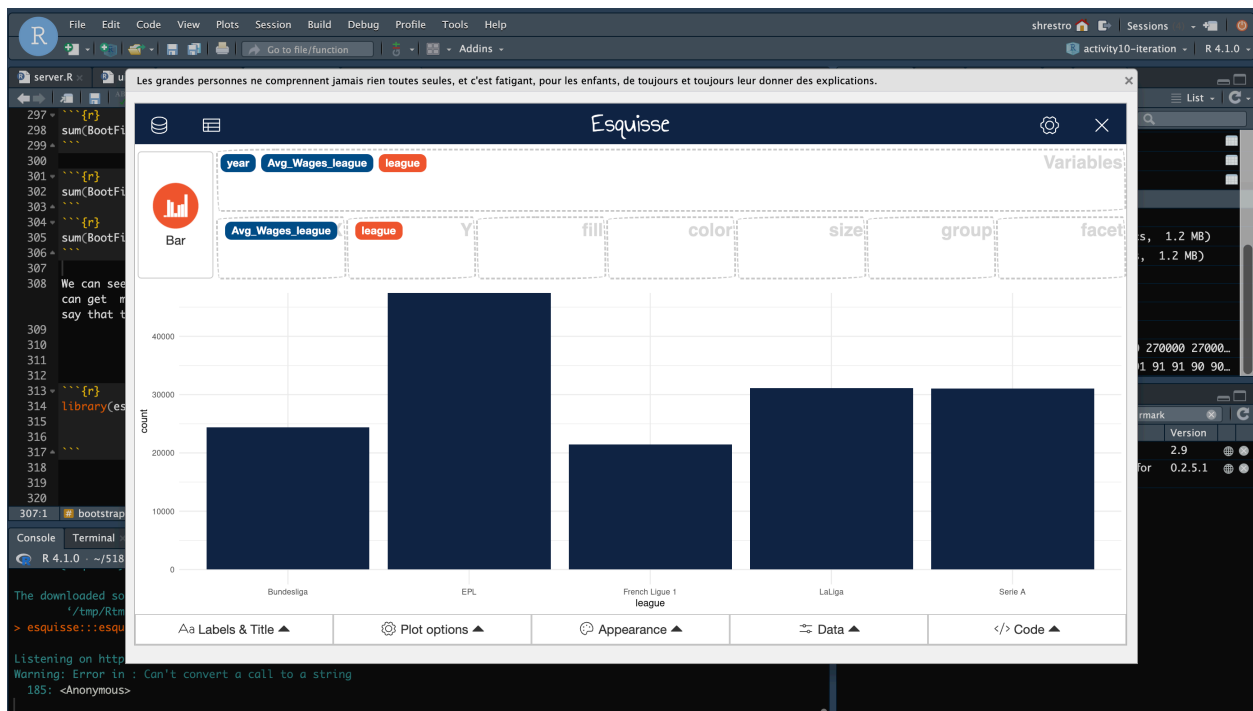


Figure 8: esquisse

**Question 5:** Do you have any other thoughts or reflections about the course that you'd like to share?

R is very powerful tool in doing statistical analysis and that is why I wanted to learn R programming. Hence my further plan would be to continue learn R and implement thing learned from this course in performing data analysis. Additionally, I would like to make interactive site similar like our final project and creating my own portfolio in R. Thank for showing me path to be better at R program and would also like to Thank you for your contribution.