

# FIFA 선수 분석

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## 1. 개요

### 주제

최근 손흥민 선수는 토트넘 소속으로 무서운 기세의 활약으로 한국선수중 최초로 푸스카스 상을 수상하는 등 만개하는 기량을 보여주며 어려운 시국 중 우리에게 몇없는 즐거움과 자부심을 안겨주고 있기 때문에, 본 조원들 뿐 아니라 많은 한국인들이 해외축구에 대한 관심과 애정을 높이고 있는것은 분명하다.

이렇게 높아지는 해외축구에 대한 국내 팬들의 열정을 빌어, 본 조는 해당 보고서에서 FIFA 선수 데이터와 UEFA 축구클럽 순위 데이터를 연계시켜 선수의 Feature 및 stat에 따라 선수가 어떤 등급의 팀에 소속될지 예측하는 분류모델을 제작하는것을 주 연구과제로 EDA, 다중회귀분석, 군집분석 등 추가적인 다양한 분석기법을 활용하며 FIFA 데이터에서 축구팬들이 정량적인 STAT를 통해 흥미를 가질 수 있는 다양한 insight를 제시해 내는 것을 연구 목표로 세우고자 한다.

### 데이터

#### 데이터셋

FIFA에서 제공하는 "**FIFA Dataset**" 과 UEFA의 "**20-21 Club Ranking**" 데이터를 사용하고자 한다.

#### 주요 변수 및 특징

- *position* :  
선수의 주 포지션을 표기하는 변수

- *stat* :  
경기내에서 측정되는 선수의 각종 수많은 stat을 관측한 변수들
- *club* :  
선수가 소속된 구단, UEFA에서 가져온 club순위 변수, 그리고 이를 통해 해당 보고서에서 만들어낼 클럽등급 파생변수
- *wage & value* :  
선수의 임금 및 가치 변수

## 연구법(분석기법)

- 1. EDA
- 2. 유사도분석
- 2. 군집분석
- 3. 다중회귀분석
- 4. 분류분석

## 세부 연구주제 & 가설

### (1) 유사도분석

손흥민, 케인 선수와 유사한 능력 및 포지션의 선수는 누구일까

### (2) 군집분석

FIFA 데이터상 선수들의 군집은 어떻게 나뉘질 것인가

### (3) 다중회귀분석

FIFA 데이터상 선수들의 overall에 가장 큰 영향을 주는 Feature은 무엇이며 나머지 선수개인특성의 변수들에 따라 overall이 어느정도로 설명될 것인가?

### (4) 분류분석

사전에 UEFA순위에 따라 임의로 나눠준 선수들이 현재 소속된 팀의 등급을 label로

분류를 실시하여, 손흥민 선수가 현재 소속된 팀의 grade에 적합한 선수인지 판별하는 분류기를 학습 및 테스트한다

## 2. 본문

### 2.1 데이터 수집 및 전처리

#### 2.1.1 패키지 준비

```
library(tidyverse)
library(magrittr)
library(DataExplorer)
library(maps)
library(plotly)
library(DT)
library(tidytext)
library(gridExtra)
library(factoextra)

library(readxl)
library(dplyr)
library(ggpubr)
library(caret)
library(corrplot)
library(car)
library(olsrr)
library(MASS)
library(ggplot2)
```

#### 2.1.2 데이터 불러오기

```
#데이터 불러오기 df
setwd('C:\\Users\\JSW\\Desktop\\학교\\1_학기\\3학년 2학기_2020_2\\강의
df <- read.csv("complete_data.csv", encoding = "UTF-8")[-1]

UEFA = read.csv('club_ranking.csv')
```

```
dim(df)
```

```
## [1] 18207    88
```

## FIFA 원데이터의 차원 갯수

```
glimpse(df)
```

```
## Rows: 18,207
## Columns: 88
## $ ID                <int> 158023, 20801, 190871, 193080, 1
## $ Name              <chr> "L. Messi", "Cristiano Ronaldo",
## $ Age               <int> 31, 33, 26, 27, 27, 27, 32, 31,
## $ Photo             <chr> "https://cdn.sofifa.org/players/
## $ Nationality       <chr> "Argentina", "Portugal", "Brazil
## $ Flag              <chr> "https://cdn.sofifa.org/flags/52
## $ Overall           <int> 94, 94, 92, 91, 91, 91, 91, 91,
## $ Potential         <int> 94, 94, 93, 93, 92, 91, 91, 91,
## $ Club              <chr> "FC Barcelona", "Juventus", "Par
## $ Club.Logo         <chr> "https://cdn.sofifa.org/teams/2/
## $ Value             <chr> "110.5M", "77M", "118.5M", "72M"
## $ Wage              <chr> "565K", "405K", "290K", "260K",
## $ Special           <int> 2202, 2228, 2143, 1471, 2281, 21
## $ Preferred.Foot    <chr> "Left", "Right", "Right", "Right
## $ International.Reputation <int> 5, 5, 5, 4, 4, 4, 4, 5, 4, 3, 4,
## $ Weak.Foot         <int> 4, 4, 5, 3, 5, 4, 4, 4, 3, 3, 4,
## $ Skill.Moves       <int> 4, 5, 5, 1, 4, 4, 4, 3, 3, 1, 4,
## $ Work.Rate         <chr> "Medium/ Medium", "High/ Low", "
## $ Body.Type         <chr> "Messi", "C. Ronaldo", "Neymar",
## $ Real.Face         <chr> "Yes", "Yes", "Yes", "Yes", "Yes
## $ Position          <chr> "RF", "ST", "LW", "GK", "RCM", "
## $ Jersey.Number    <int> 10, 7, 10, 1, 7, 10, 10, 9, 15,
## $ Joined            <chr> "01-Jul-04", "10-Jul-18", "03-Au
## $ Loaned.From       <chr> "", "", "", "", "", "", "", "",
## $ Contract.Valid.Until <chr> "2021", "2022", "2022", "2020",
## $ Height            <chr> "5'7", "6'2", "5'9", "6'4", "5'1
## $ Weight            <chr> "159lbs", "183lbs", "150lbs", "1
## $ LS               <chr> "88+2", "91+3", "84+3", "", "82+
```

## \$ ST	<chr> "88+2", "91+3", "84+3", "", "82+
## \$ RS	<chr> "88+2", "91+3", "84+3", "", "82+
## \$ LW	<chr> "92+2", "89+3", "89+3", "", "87+
## \$ LF	<chr> "93+2", "90+3", "89+3", "", "87+
## \$ CF	<chr> "93+2", "90+3", "89+3", "", "87+
## \$ RF	<chr> "93+2", "90+3", "89+3", "", "87+
## \$ RW	<chr> "92+2", "89+3", "89+3", "", "87+
## \$ LAM	<chr> "93+2", "88+3", "89+3", "", "88+
## \$ CAM	<chr> "93+2", "88+3", "89+3", "", "88+
## \$ RAM	<chr> "93+2", "88+3", "89+3", "", "88+
## \$ LM	<chr> "91+2", "88+3", "88+3", "", "88+
## \$ LCM	<chr> "84+2", "81+3", "81+3", "", "87+
## \$ CM	<chr> "84+2", "81+3", "81+3", "", "87+
## \$ RCM	<chr> "84+2", "81+3", "81+3", "", "87+
## \$ RM	<chr> "91+2", "88+3", "88+3", "", "88+
## \$ LWB	<chr> "64+2", "65+3", "65+3", "", "77+
## \$ LDM	<chr> "61+2", "61+3", "60+3", "", "77+
## \$ CDM	<chr> "61+2", "61+3", "60+3", "", "77+
## \$ RDM	<chr> "61+2", "61+3", "60+3", "", "77+
## \$ RWB	<chr> "64+2", "65+3", "65+3", "", "77+
## \$ LB	<chr> "59+2", "61+3", "60+3", "", "73+
## \$ LCB	<chr> "47+2", "53+3", "47+3", "", "66+
## \$ CB	<chr> "47+2", "53+3", "47+3", "", "66+
## \$ RCB	<chr> "47+2", "53+3", "47+3", "", "66+
## \$ RB	<chr> "59+2", "61+3", "60+3", "", "73+
## \$ Crossing	<int> 84, 84, 79, 17, 93, 81, 86, 77,
## \$ Finishing	<int> 95, 94, 87, 13, 82, 84, 72, 93,
## \$ HeadingAccuracy	<int> 70, 89, 62, 21, 55, 61, 55, 77,
## \$ ShortPassing	<int> 90, 81, 84, 50, 92, 89, 93, 82,
## \$ Volleys	<int> 86, 87, 84, 13, 82, 80, 76, 88,
## \$ Dribbling	<int> 97, 88, 96, 18, 86, 95, 90, 87,
## \$ Curve	<int> 93, 81, 88, 21, 85, 83, 85, 86,
## \$ FKAaccuracy	<int> 94, 76, 87, 19, 83, 79, 78, 84,
## \$ LongPassing	<int> 87, 77, 78, 51, 91, 83, 88, 64,
## \$ BallControl	<int> 96, 94, 95, 42, 91, 94, 93, 90,
## \$ Acceleration	<int> 91, 89, 94, 57, 78, 94, 80, 86,
## \$ SprintSpeed	<int> 86, 91, 90, 58, 76, 88, 72, 75,
## \$ Agility	<int> 91, 87, 96, 60, 79, 95, 93, 82,
## \$ Reactions	<int> 95, 96, 94, 90, 91, 90, 90, 92,
## \$ Balance	<int> 95, 70, 84, 43, 77, 94, 94, 83,
## \$ ShotPower	<int> 85, 95, 80, 31, 91, 82, 79, 86,
## \$ Jumping	<int> 68, 95, 61, 67, 63, 56, 68, 69,

```
## $ Stamina      <int> 72, 88, 81, 43, 90, 83, 89, 90,
## $ Strength     <int> 59, 79, 49, 64, 75, 66, 58, 83,
## $ LongShots    <int> 94, 93, 82, 12, 91, 80, 82, 85,
## $ Aggression   <int> 48, 63, 56, 38, 76, 54, 62, 87,
## $ Interceptions <int> 22, 29, 36, 30, 61, 41, 83, 41,
## $ Positioning  <int> 94, 95, 89, 12, 87, 87, 79, 92,
## $ Vision       <int> 94, 82, 87, 68, 94, 89, 92, 84,
## $ Penalties    <int> 75, 85, 81, 40, 79, 86, 82, 85,
## $ Composure    <int> 96, 95, 94, 68, 88, 91, 84, 85,
## $ Marking      <int> 33, 28, 27, 15, 68, 34, 60, 62,
## $ StandingTackle <int> 28, 31, 24, 21, 58, 27, 76, 45,
## $ SlidingTackle <int> 26, 23, 33, 13, 51, 22, 73, 38,
## $ GKDiving     <int> 6, 7, 9, 90, 15, 11, 13, 27, 11,
## $ GKHandling   <int> 11, 11, 9, 85, 13, 12, 9, 25, 8,
## $ GKKicking    <int> 15, 15, 15, 87, 5, 6, 7, 31, 9,
## $ GKPositioning <int> 14, 14, 15, 88, 10, 8, 14, 33, 7
## $ GKReflexes   <int> 8, 11, 11, 94, 13, 8, 9, 37, 11,
## $ Release.Clause <chr> "226.5M", "127.1M", "228.1M", "1
```

## FIFA 원데이터의 모습

```
glimpse(UEFA)
```

```
## Rows: 429
## Columns: 10
## $ rank <chr> "1?", "2?", "3?", "4?", "5?", "6?", "7?", "8?", "9?"
## $ club <chr> "FC Bayern Munchen", "FC Barcelona", "Juventus", "Re
## $ X    <chr> "GER", "ESP", "ITA", "ESP", "ESP", "ENG", "FRA", "ES
## $ X.1  <chr> "22", "23", "33", "33", "29", "18", "20", "19", "26"
## $ X.2  <chr> "29", "25", "23", "32", "28", "22", "19", "21", "20"
## $ X.3  <chr> "20", "30", "21", "19", "20", "25", "19", "13", "19"
## $ X.4  <chr> "36", "24", "22", "17", "22", "25", "31", "26", "22"
## $ X.5  <chr> "20", "19", "19", "16", "16", "20", "17", "18", "10"
## $ X.6  <dbl> 127.0, 121.0, 118.0, 117.0, 115.0, 110.0, 106.0, 97.
## $ X.7  <dbl> 13.799, 17.885, 14.030, 17.885, 17.885, 17.713, 10.8
```

## UEFA 클럽 순위 데이터의 모습

### 2.1.3 데이터 전처리

#### 1.club\_ranking data 전처리 및 임의의 클럽등급 파생변수 생성

이후 분류분석에 label로 활용하기 위한 선수 소속의 클럽등급 파생변수를 생성합니다.

```

UEFA = UEFA[,c(1,2)]

for(i in 1:429){
  UEFA$rank[i] = i
}

#해당 데이터에서 구간등급도 나눠 파생변수를 만들어준다
UEFA$rank = as.numeric(UEFA$rank)

club_grade = c(1:10)

UEFA$club_grade = c(rep('S',times=10),rep('A',times=10),rep('B',time

```

### 2.리그변수 할당

```

bundesliga <- c(
  "1. FC Nürnberg", "1. FSV Mainz 05", "Bayer 04 Leverkusen", "FC Ba
  "Borussia Dortmund", "Borussia Mönchengladbach", "Eintracht Frankf
  "FC Augsburg", "FC Schalke 04", "Fortuna Düsseldorf", "Hannover 96
  "Hertha BSC", "RB Leipzig", "SC Freiburg", "TSG 1899 Hoffenheim",
  "VfB Stuttgart", "VfL Wolfsburg", "SV Werder Bremen"
)

```



```

premierLeague <- c(
  "Arsenal", "Bournemouth", "Brighton & Hove Albion", "Burnley",
  "Cardiff City", "Chelsea", "Crystal Palace", "Everton", "Fulham",
  "Huddersfield Town", "Leicester City", "Liverpool", "Manchester Ci
  "Manchester United", "Newcastle United", "Southampton",
  "Tottenham Hotspur", "Watford", "West Ham United", "Wolverhampton
)

laliga <- c(
  "Athletic Club de Bilbao", "Atlético Madrid", "CD Leganés",
  "Deportivo Alavés", "FC Barcelona", "Getafe CF", "Girona FC",
  "Levante UD", "Rayo Vallecano", "RC Celta", "RCD Espanyol",
  "Real Betis", "Real Madrid", "Real Sociedad", "Real Valladolid CF"
  "SD Eibar", "SD Huesca", "Sevilla FC", "Valencia CF", "Villarreal
)

seriea <- c(
  "Atalanta", "Bologna", "Cagliari", "Chievo Verona", "Empoli", "Fiorent
  "Inter", "Juventus", "Lazio", "Milan", "Napoli", "Parma", "Roma", "Sampdo
  "Torino", "Udinese"
)

superlig <- c(
  "Akhisar Belediyespor", "Alanyaspor", "Antalyaspor", "Medipol Başakş
  "Bursaspor", "Çaykur Rizespor", "Fenerbahçe SK", "Galatasaray SK", "G
  "Kayserispor", "Atiker Konyaspor", "MKE Ankaragücü", "Sivasspor", "Tr
)

ligue1 <- c(
  "Amiens SC", "Angers SCO", "AS Monaco", "AS Saint-Étienne", "Dijon
  "FC Nantes", "FC Girondins de Bordeaux", "LOSC Lille", "Montpellier
  "OGC Nice", "Olympique Lyonnais", "Olympique de Marseille", "Paris
  "RC Strasbourg Alsace", "Stade Malherbe Caen", "Stade de Reims", "
)

eredivisie <- c(
  "ADO Den Haag", "Ajax", "AZ Alkmaar", "De Graafschap", "Excelsior", "
  "FC Utrecht", "Feyenoord", "Fortuna Sittard", "Heracles Almelo", "NA
  "PEC Zwolle", "PSV", "SC Heerenveen", "Vitesse", "VVV-Venlo", "Willem
)

```

```

liganos <- c(
  "Os Belenenses", "Boavista FC", "CD Feirense", "CD Tondela", "CD A
  "CD Nacional", "GD Chaves", "Clube Sport Marítimo", "Moreirense FC
  "Santa Clara", "SC Braga", "SL Benfica", "Sporting CP", "Vitória G
)

df %<>% mutate(
  League = case_when(
    Club %in% bundesliga ~ "Bundesliga",
    Club %in% premierLeague ~ "Premier League",
    Club %in% laliga ~ "La Liga",
    Club %in% seriea ~ "Serie A",
    Club %in% superlig ~ "Süper Lig",
    Club %in% ligue1 ~ "Ligue 1",
    Club %in% liganos ~ "Liga Nos",
    Club %in% eredivisie ~ "Eredivisie"
  ),
  Country = case_when(
    League == "Bundesliga" ~ "Germany",
    League == "Premier League" ~ "UK",
    League == "La Liga" ~ "Spain",
    League == "Serie A" ~ "Italy",
    League == "Süper Lig" ~ "Turkey",
    League == "Ligue 1" ~ "France",
    League == "Liga Nos" ~ "Portugal",
    League == "Eredivisie" ~ "Netherlands"
  )
) %>% filter(!is.na(League)) %>% mutate_if(is.factor, as.character)

rm(bundesliga, premierLeague, laliga, seriea, superlig, ligue1, ered

```

### 3.문자값 수정

```
head(df$Value)
```

```
## [1] "110.5M" "77M" "118.5M" "72M" "102M" "93M"
```

```
# Player Value
```

```
df$Values <- str_remove_all(df$Value, "€")
```

```
df$Values <- str_replace_all(df$Values, "K", "000")
```

```
df$Values <- str_remove_all(df$Values, "M")
```

```
df$Values <- as.numeric(df$Values)
```

```
# Player Wage
```

```
df$Wages <- str_remove_all(df$Wage, "€")
```

```
df$Wages <- str_replace_all(df$Wages, "K", "000")
```

```
df$Wages <- as.numeric(df$Wages)
```

```
df <- df %>% mutate(Values = if_else(df$Values < 1000 , Values * 10
```

#### 4.포지션 클래스 생성

```
unique(df$Position)
```

```
## [1] "RF" "ST" "LW" "GK" "RCM" "LF" "RS" "RCB" "LCM" "CB"
```

```
## [13] "CDM" "LS" "LCB" "RM" "LAM" "LM" "LB" "RDM" "RW" "CM"
```

```
## [25] "CF" "RWB" "LWB" ""
```

```
# 기존에 존재하던 position 변수로 파생 변수 Class 변수 생성
```

```
defence <- c("CB", "RB", "LB", "LWB", "RWB", "LCB", "RCB")
```

```
midfielder <- c("CM", "CDM", "CAM", "LM", "RM", "LAM", "RAM", "LCM", "R
```

```
df %>% mutate(Class = if_else(Position %in% "GK", "Goal Keeper",
```

```
if_else(Position %in% defence, "Defen
```

```
if_else(Position %in% midfiel
```

```
rm(defence, midfielder)
```

## 5.Height & Weight

```
df %<>%
  mutate(Height = round((as.numeric(str_sub(Height, start=1,end = 1)
    Weight = round(as.numeric(str_sub(Weight, start = 1, end =

levels(factor(df$Country))
```

```
## [1] "France"      "Germany"      "Italy"        "Netherlands" "Port
## [6] "Spain"       "Turkey"      "UK"
```

```
a = factor(df$Country)
levels(a)
```

```
## [1] "France"      "Germany"      "Italy"        "Netherlands" "Port
## [6] "Spain"       "Turkey"      "UK"
```

## 6.선호 주발의 수정

```
# 선호하는 주발의 교정

df %<>% filter(Preferred.Foot %in% c("Left", "Right"))
df$Preferred.Foot <- as.factor(as.character(df$Preferred.Foot))
```

## 7.변수명 재설정

```
#3.6 변수명 재설정
df %<>%
```

```

rename(
  "Heading.Accuracy"= HeadingAccuracy,
  "Short.Passing"= ShortPassing,
  "FK.Accuracy" = FKAccuracy,
  "Long.Passing"= LongPassing,
  "Ball.Control"= BallControl,
  "Sprint.Speed"= SprintSpeed,
  "Shot.Power"= ShotPower,
  "Long.Shots"= LongShots,
  "Standing.Tackle"= StandingTackle,
  "Sliding.Tackle"= SlidingTackle,
  "GK.Diving"= GKDiving,
  "GK.Handling"= GKHandling,
  "GK.Kicking"= GKKicking,
  "GK.Positioning"= GKPositioning,
  "GK.Reflexes"= GKReflexes
)

```

## 8. 불필요 변수 제거

#3.7 불필요 변수 제거

```
df %<>% dplyr::select(-ID, -Body.Type, -Real.Face, -Joined, -Loaned.
```

## 9. 데이터 병합

```
b = merge(df, UEFA, key="club") #FIFA와 UEFA 데이터 병합
```

```
#glimpse(b)
```

## 2.2 EDA & Visualization

### 2.2.1 각 리그 선수들의 분산과 평균

#5.1. 각 리그 선수들의 연령 분산과 평균

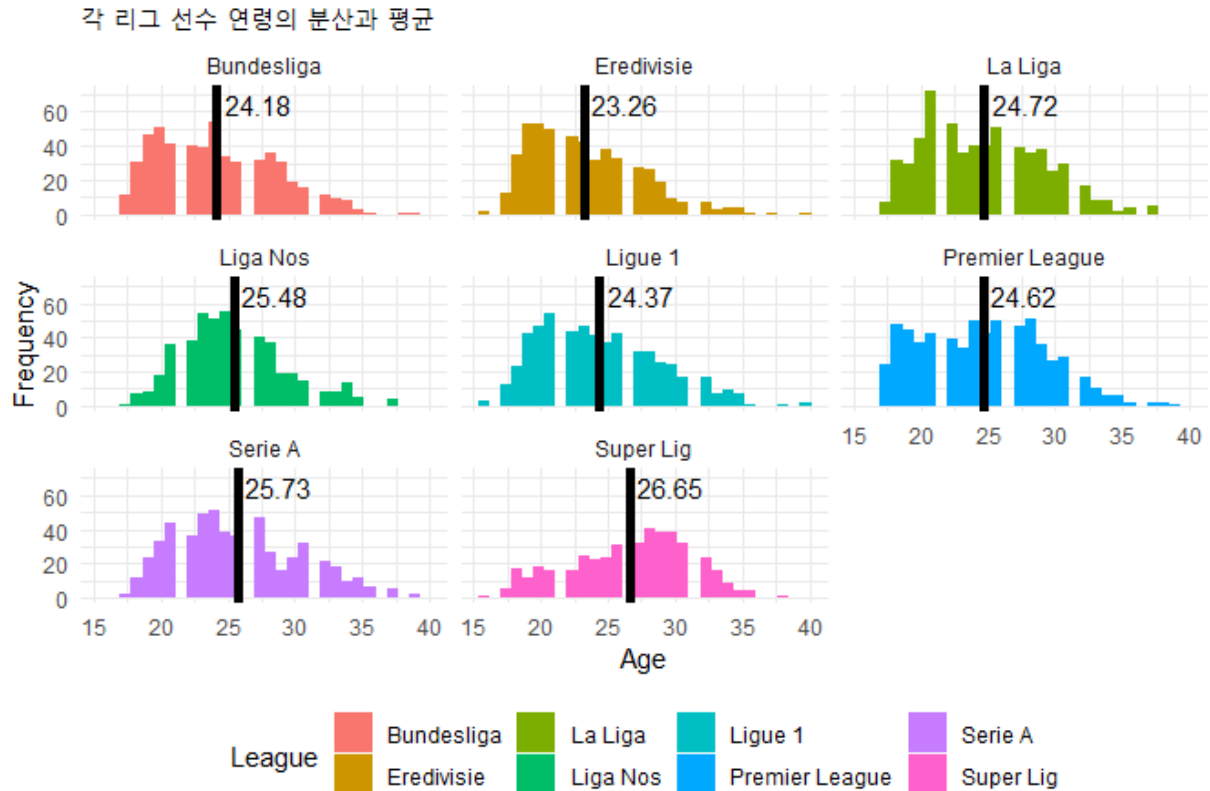
```
summ <- df %>%
  group_by(League) %>%
  summarise(age = mean(Age))
```

```
## `summarise()` ungrouping output (override with `.groups` argument)
```

```
options(repr.plot.width = 15, repr.plot.height = 6)

ggplot()+
  geom_histogram(df, mapping = aes(Age, fill = League))+
  geom_vline(summ, mapping = aes(xintercept = age), color = "black",
  geom_text(summ, mapping = aes(x = age+3, y = 65, label = round(age
  facet_wrap(League~.))+
  theme_minimal()+
  theme(legend.position = "bottom")+
  labs(y = "Frequency", title = "각 리그 선수 연령의 분산과 평균", cap
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`
```



R통계\_5조

## 2.2.2 두 선수들 간의 비교

#5.4 두 선수들 간의 비교

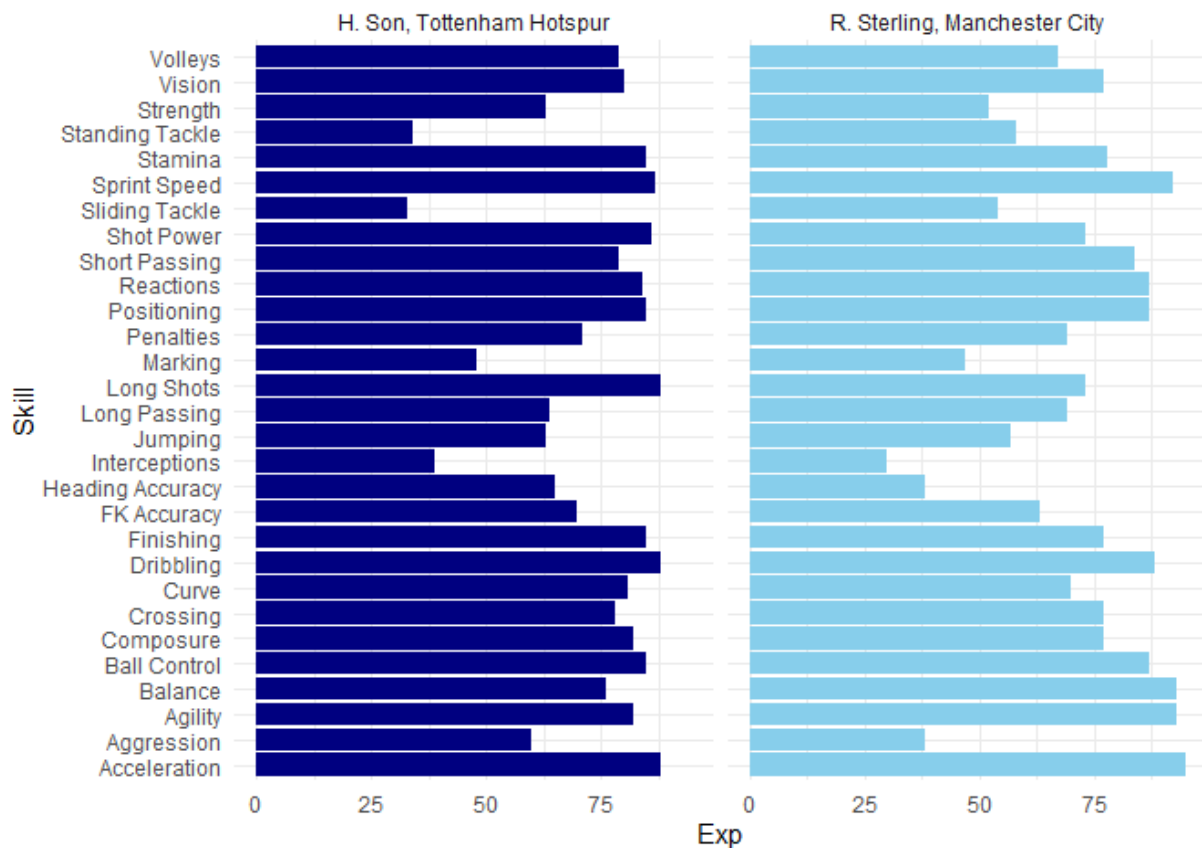
```
library(tidyr)
# 비교대상이 될 두 선수 선정
players <- df %>%
  filter(Name %in% c("H. Son", "R. Sterling")) %>%
  # 이름 및 소속 클럽명
  mutate(Name = paste0(Name, ", ", Club)) %>%
  # 선수들의 비교기준 스탯 선정
  dplyr::select(Name, Crossing:Sliding.Tackle) %>%
  # 문자열수정
  rename_all(funs(gsub("[[:punct:]]", " ", .))) %>%
  # 변수에서 관측치로의 변형
  gather(Skill, Exp, Crossing:`Sliding Tackle`, -Name)
head(players)
```

```
##
## 1 R. Sterling, Manchester City      Crossing 77
## 2   H. Son, Tottenham Hotspur      Crossing 78
## 3 R. Sterling, Manchester City      Finishing 77
## 4   H. Son, Tottenham Hotspur      Finishing 85
## 5 R. Sterling, Manchester City      Heading Accuracy 38
## 6   H. Son, Tottenham Hotspur      Heading Accuracy 65
```

#두 선수 비교 시각화 <<1>>

```
options(repr.plot.width = 15, repr.plot.height = 8)
```

```
ggplot(players, aes(Skill, Exp, fill = Name))+
  geom_col(show.legend = FALSE)+
  coord_flip()+
  facet_wrap(Name~.)+
  scale_fill_manual(values = c("navy", "skyblue"))+
  theme_minimal()
```

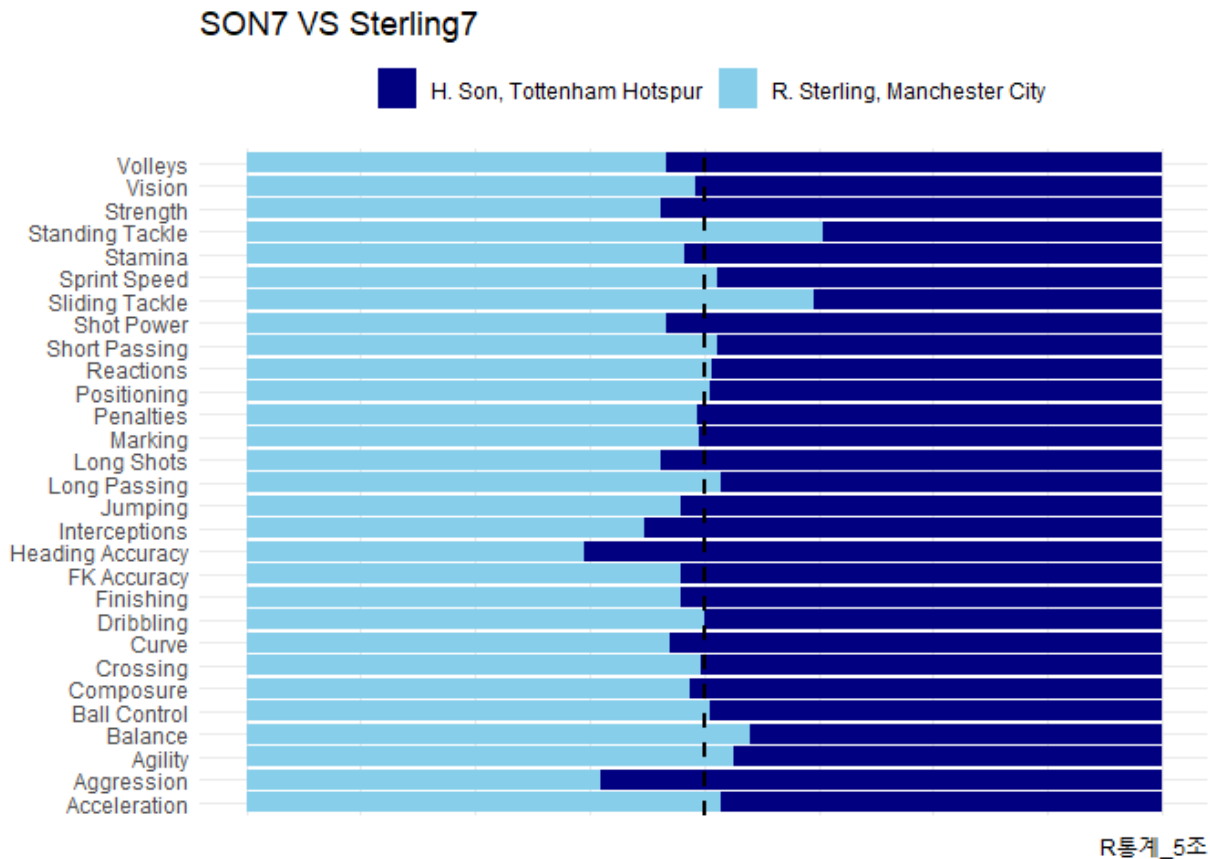


#두 선수 비교 시각화 <<2>>

```
options(repr.plot.width = 15, repr.plot.height = 8)
```



```
ggplot(players, aes(Skill, Exp, fill = Name))+
  geom_col(position = "fill")+
  coord_flip()+
  scale_fill_manual(values = c("navy", "skyblue"))+
  theme_minimal()+
  geom_hline(yintercept = 0.5, color = "black", size = 1, linetype =
  theme(legend.position = "top", axis.text.x=element_blank())+
  labs(title = "SON7 VS Sterling7",
        caption = "R통계_5조",
        fill = NULL,x = NULL, y = NULL)
```

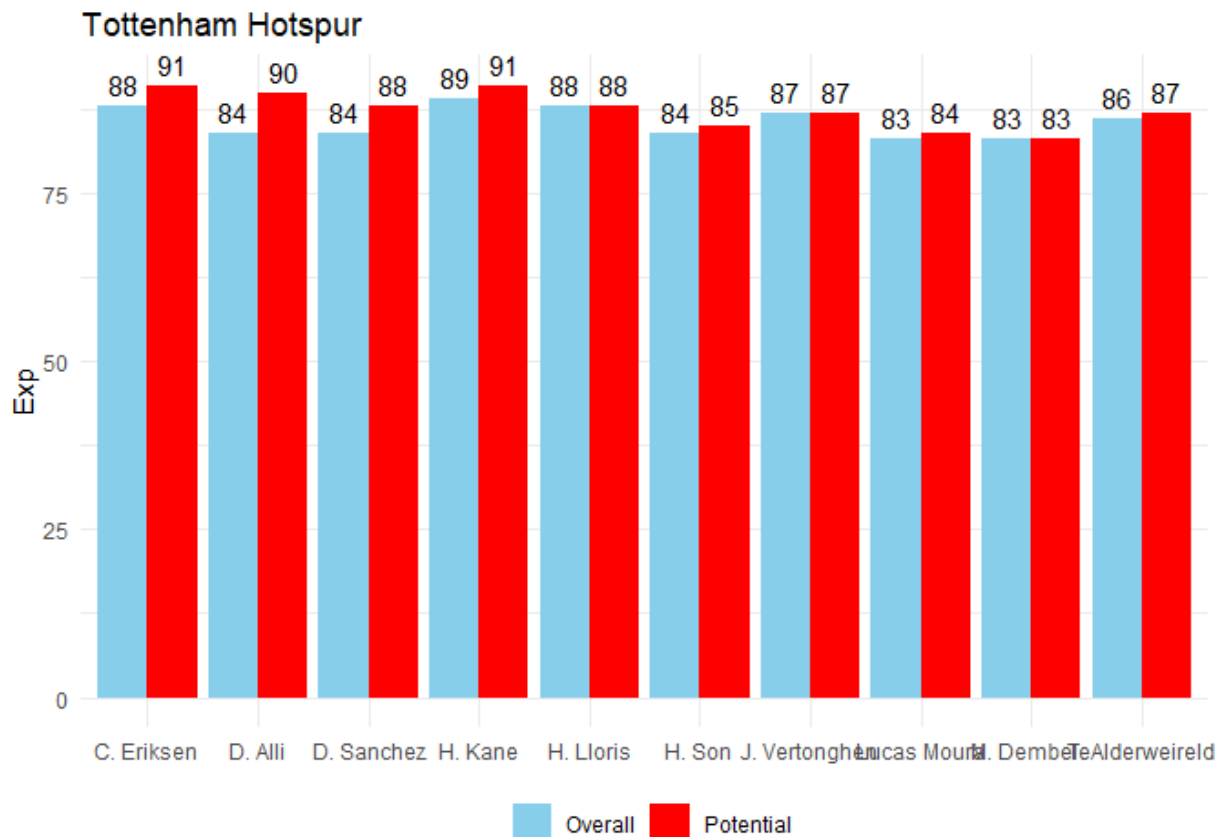


프리미어리그의 각 팀에서 등번호 7번을 달고 윙포워드로 활약하고있는 손흥민 선수와 스털링 선수의 *stat*을 비교해본 결과, 두 선수는 매우 유사한 *stat*분포를 가지고 있음을 확인할 수 있었다.

### 2.2.3 토트넘 선수들의 잠재력&통합스탯

```
options(repr.plot.width = 12, repr.plot.height = 8)

df %>%
  filter(Club == "Tottenham Hotspur") %>%
  dplyr::select(Name, Overall, Potential) %>%
  arrange(-Overall) %>%
  head(10) %>%
  gather(variable, Exp, -Name) %>%
  ggplot(aes(Name, Exp, fill = variable))+
  geom_col(position = "dodge")+
  geom_text(aes(label = Exp), position = position_dodge(width = 0.9),
  scale_fill_manual(values = c("skyblue", "red"))+
  theme_minimal()+
  theme(legend.position = "bottom")+
  labs(fill = NULL, x = NULL, title = "Tottenham Hotspur")
```



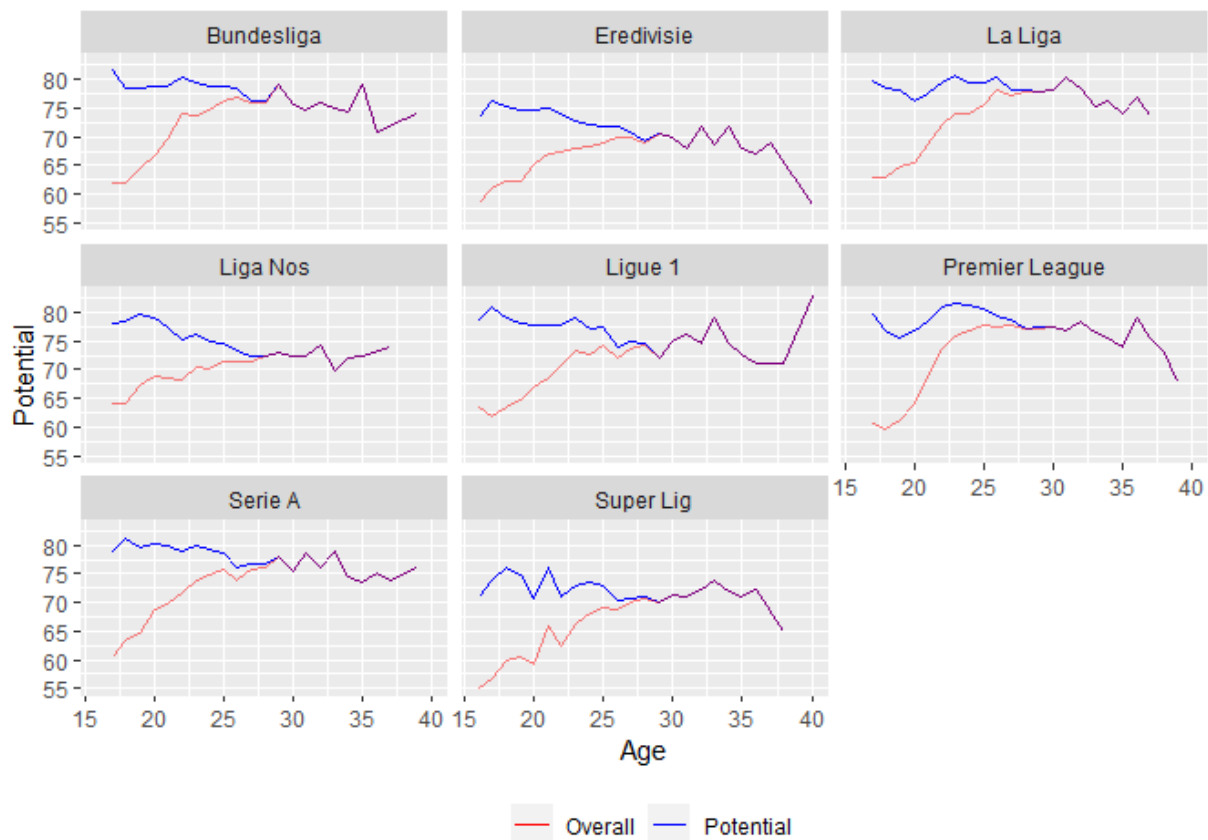
토티넘 내 선수들의 *Overall* 및 *Potential*을 시각화 해본 결과 *Harry Kane*이 가장 높은 *Overall*과 *Potential*을 가진 선수임을 확인할 수 있다

## 2.2.4 나이에 따른 소속 리그별 평균 잠재력 및 통합스탯

```
options(repr.plot.width = 12, repr.plot.height = 8)

df %>%
  group_by(League, Age) %>%
  summarise(Overall = mean(Overall),
            Potential = mean(Potential)) %>%
  ggplot()+
  geom_line(aes(Age, Potential, color = "Potential")) +
  geom_line(aes(Age, Overall, color = "Overall"), alpha = 0.5) +
  facet_wrap(League~.)+
  scale_color_manual(values = c("red", "blue"))+
  theme(legend.position = "bottom")+
  labs(color = NULL)
```

```
## `summarise()` regrouping output by 'League' (override with `.group`)
```



나이에 따른 소속 리그별 평균 잠재력 및 *Overall*의 분포를 확인해 본 결과 유럽의 4대 리그라 불리는 분데스리가,프리미어리그,프리메라리가,세리에가 예상처

럼 높은 *Overall* 및 *stat*분포를 보이고 있음을 확인할 수 있다.

한가지 더 주목해보자면, 모든 리그를 통틀어 약 25세쯤부터 선수들의 *Potential*과 *Overall*이 일치하여 동일선을 그리는 것을 확인할 수 있다.

## 2.2.5 나이에 따른 소속 리그별 평균 잠재력 및 통합스탯

#5.11 가장 강한 클럽

```
options(repr.plot.width = 12, repr.plot.height = 8)
```

```
powerful <- df %>%
  group_by(Club) %>%
  summarise(mean = mean(Overall)) %>%
  arrange(-mean) %>%
  head(20)
```

```
## `summarise()` ungrouping output (override with `.groups` argument
```

```
df %>%
  group_by(Club, Class) %>%
  summarise(mean = mean(Overall)) %>%
  ungroup() %>%
  filter(Club %in% powerful$Club) %>%
  ggplot(aes(reorder(Club, mean), mean, fill = Class))+
  geom_col(position = "fill")+
  geom_text(aes(label = round(mean,digits = 2)), position = position_
coord_flip()+
  theme_minimal()+
  theme(legend.position = "top")+
  labs(x = NULL, y = NULL, title = "Clubs' Overall by Position")
```

```
## `summarise()` regrouping output by 'Club' (override with `.groups`
```

## Clubs' Overall by Position



## 2.3 손흥민 &amp; 케인의 유사선수 분석

유클라디안 거리를 사용한 유사도 분석을 실시하여 알아본 두 선수와 유사한 선수들

```
options(repr.plot.width = 12, repr.plot.height = 8)

similarity <- function(df, player, selectLeague, fill_variable, fill
                        input = c("Eucledian", "Maximum", "Manhattan"
                                "Minkowski", "Pearson", "Spearman",
                                ){

  res <- NULL

  if(missing("df") | missing("input") | missing("player") | missing(
```

```

if(is.null(df) | is.null(input) | is.null(player) | is.null(fill_v
if(length(selectLeague) < 1) return(res)

if(input == "Euclidean"){
  distance_method <- "euclidean"
}else if(input == "Maximum"){
  distance_method <- "maximum"
}else if(input == "Manhattan"){
  distance_method <- "manhattan"
}else if(input == "Canberra"){
  distance_method <- "canberra"
}else if(input == "Minkowski"){
  distance_method <- "minkowski"
}else if(input == "Pearson"){
  distance_method <- "pearson"
}else if(input == "Spearman"){
  distance_method <- "spearman"
}else if(input == "Kendall"){
  distance_method <- "kendall"
}else{
  return(NULL)
}

plyr <- df %>% filter(Name == player)

smdf <- df %>% filter(League %in% selectLeague, !Name %in% plyr$Na
rwnam <- paste0(smdf$Name, " (", smdf$Club, ")")

smdf <- smdf %>% select_if(is.numeric) %>% dplyr::select(-Jersey.N

smdf <- apply(smdf, 2, scale)
rownames(smdf) <- rwnam

smdf <- get_dist(smdf, method = distance_method)
smdf <- fviz_dist(smdf)

smdf <- smdf$data %>%
  mutate(Var1 = str_sub(Var1, start = 1, end = str_length(Var1)-1)

```

```

        Var2 = str_sub(Var2, start = 1, end = str_length(Var2)-1)
    arrange(value) %>%
    filter(value > 0, Var1 == rwname[length(rwname)]) %>%
    head(30)

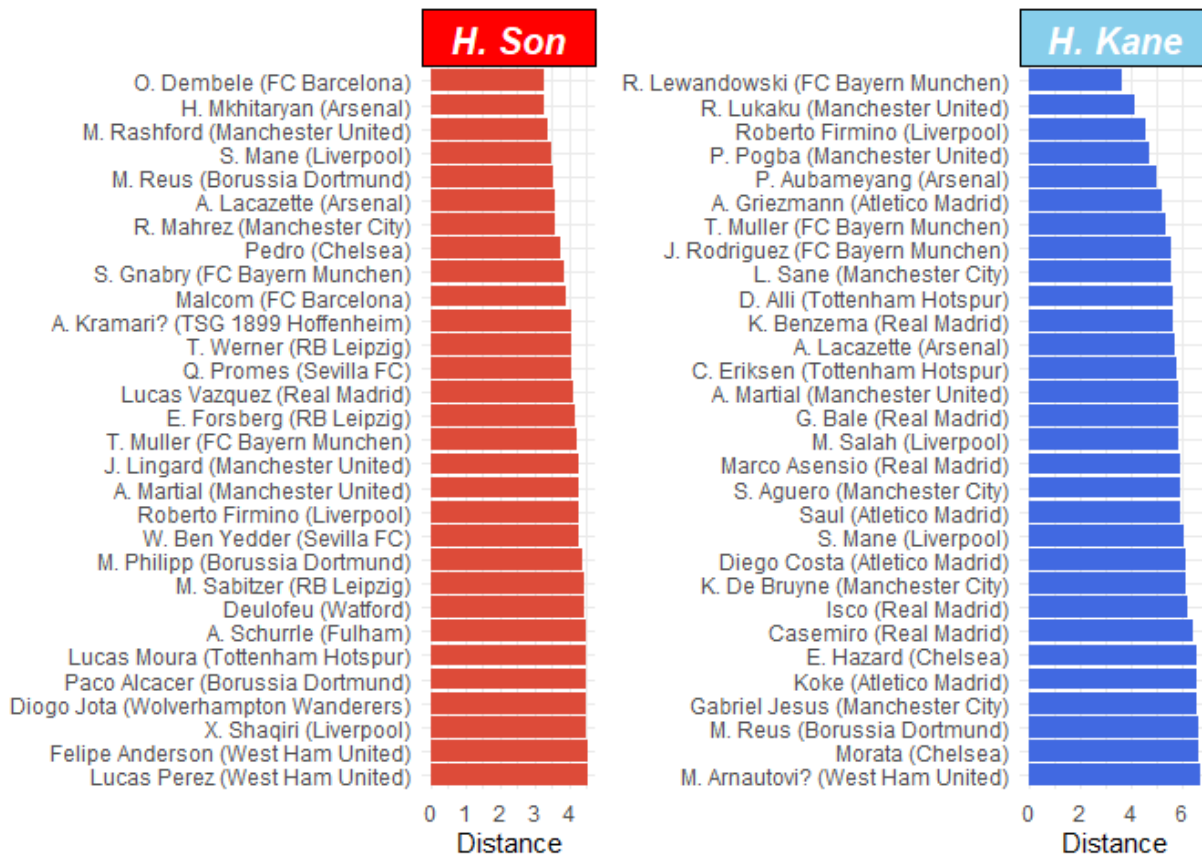
res <- ggplot(smdf, aes(fct_reorder(Var2, desc(value)), value))+
  geom_col(fill = fill_variable)+
  coord_flip()+
  labs(y = "Distance", x = NULL)+
  theme_minimal()+
  facet_wrap(~paste0(player))+
  theme(strip.background = element_rect(fill=fill_strip,color = "black"),
        strip.text.x = element_text(size = 15, colour = "white"),fa

return(res)

}

grid.arrange(
  ncol = 2,
  similarity(df, input = "Euclidean", selectLeague = c("Bundesli
    player = "H. Son", fill_variable = "#dd4b39", fill_
  similarity(df, input = "Euclidean", selectLeague = c("Bundesli
    player = "H. Kane", fill_variable = "royalblue", fi
  )

```



유클라디안 거리를 계산하여 손흥민선수와 케인 선수와 유사한 선수들을 확인해 보았다

손흥민 선수는 뎀벨레,르키타리안,래시포드,마네,라카제트 등 상대 수비진을 침투하는 플레이방식의 선수들과 유사하다는 결과를 얻을 수 있었다.

반면 해리케인 선수는 레반도프스키,루카쿠,피르미누,포그바 등 득점능력 뿐 아니라 패스 등 전체적으로 높은 밸런스를 가진 선수들과 유사하다고 판단되어 완성형 공격수라는 별명에 어울리는 결과를 얻을 수 있었다.

## 2.4. K-means Clustering

전방 공격수들의 특징에 따른 군집을 나누는 모델

```
km <- df %>% filter(Class == "Forward") #포지션이 forward인 class의 선
```



```
x <- km

x$ID <- 1:nrow(x)

row.names(x) <- paste(x$ID, x$Name, sep = "-")
x$Name <- NULL
x$ID <- NULL

x <- x %>% select_if(is.numeric) %>% dplyr::select(-Jersey.Number, -

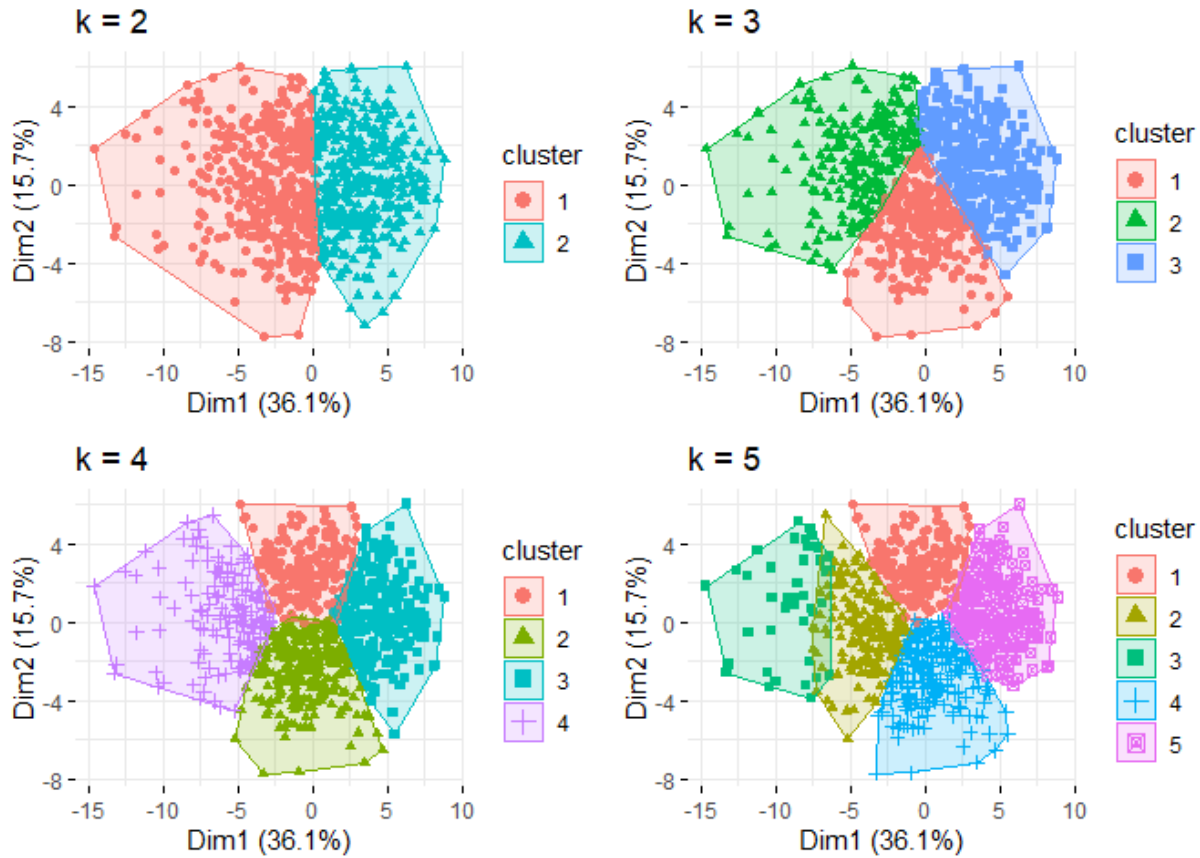
x <- apply(x, 2, scale)

x <- as.data.frame(x)

k2 <- kmeans(x, center = 2, nstart = 25) #k값이 2개인 경우
k3 <- kmeans(x, centers = 3, nstart = 25) #k값이 3개인 경우
k4 <- kmeans(x, centers = 4, nstart = 25) #k값이 4개인 경우
k5 <- kmeans(x, centers = 5, nstart = 25) #k값이 5개인 경우

p1 <- fviz_cluster(k2, geom = "point", data = x, ggtheme = theme_min
p2 <- fviz_cluster(k3, geom = "point", data = x, ggtheme = theme_min
p3 <- fviz_cluster(k4, geom = "point", data = x, ggtheme = theme_min
p4 <- fviz_cluster(k5, geom = "point", data = x, ggtheme = theme_min

options(repr.plot.width = 15, repr.plot.height = 8)
grid.arrange(p1,p2,p3,p4)
```



## 2.5 다중회귀분석

FIFA 선수의 인구통계학적 Feature와 경기중에 관측되는 각종 stat들을 설명변수로 선수의 overall score를 얼마나 설명할 수 있을것인지 다중회귀 모델을 설계하고 개선하는 과정을 통해 규명해보고자 한다.

변수

종속변수 :

**"Overall"**

설명변수 :

**"Age" "Value" "Wage" "Special" "International.Reputation"**  
**"Weak.Foot" "Skill.Moves" "Height" "Weight" "각종 경기내 STAT"**

## 가설

- Age는 Overall에 유의한 영향을 미칠 것이다.
- Value,Wage는 Overall에 유의한 영향을 미칠 것이다.
- Special은 Overall에 유의한 영향을 미칠 것이다.
- International Reputation은 Overall에 유의한 영향을 미칠 것이다.
- Weak.Foot은 Overall에 유의한 영향을 미칠 것이다.
- Skill.Moves는 Overall에 유의한 영향을 미칠 것이다.
- Height&Weight는 Overall에 유의한 영향을 미칠 것이다.
- 선수의 경기내 관측되는 각종 stat은 Overall에 유의한 영향을 미칠 것이다.

### 2.5.1 다중회귀분석을 위한 추가적인 전처리

```
#다중회귀분석을 위한 추가적인 전처리 필요
#불필요한 범주형 특징 제거
```

```
data_original = read.csv("complete_data.csv")
```

```
#data_original[c(1:5),]
```

```
cols.dont.want <- c("X","ID","Name", "Photo", "Nationality", "Flag",
data <- data_original[, ! names(data_original) %in% cols.dont.want,
#data[c(1:5),]
```

```

# 모든 관측값을 수치형으로 변환

data$Value <- gsub("€", "", data$Value)
data$Value <- as.numeric(gsub("M", "000", data$Value))
data$Release.Clause <- gsub("€", "", data$Release.Clause)
data$Release.Clause <- as.numeric(gsub("M", "000", data$Release.Clau
data$Wage <- gsub("€", "", data$Wage)
data$Wage <- as.numeric(gsub("K", "", data$Wage))
data$Height <- as.numeric(gsub("\'", ".", data$Height))
data$Weight <- as.numeric(gsub("lbs", "", data$Weight))

for (i in c(11:38)){
  data[,i]<- as.numeric(gsub("+", ".", data[,i],fixed=TRUE))
}

for(i in 1:ncol(data)){
  data[is.na(data[,i]), i] <- mean(data[,i], na.rm = TRUE)
}

```

#모델에서 가중치가 동일하게 받도록 데이터를 정규화시켜준다

```

# calculate the pre-process parameters from the dataset
preprocessParams <- preProcess(data, method=c("range"))

```

```
# summarize transform parameters
print(preprocessParams)
```

```
## Created from 18207 samples and 71 variables
##
## Pre-processing:
##   - ignored (0)
##   - re-scaling to [0, 1] (71)
```

```
# transform the dataset using the parameters
data_scaled <- predict(preprocessParams, data) + c(0.0000001)
# summarize the transformed dataset
for(i in 1:ncol(data))
  data_scaled[is.na(data_scaled[,i]), i] <- mean(data_scaled[,i], na
data_scaled[is.infinite(data_scaled[,i]), i] <- mean(data_scaled[,i]
#data_scaled
summary(data_scaled)
```

```
##      Age              Overall              Value
## Min.    :0.0000001  Min.    :0.0000001  Min.    :0.0000001
## 1st Qu.:0.1724139  1st Qu.:0.3333334  1st Qu.:0.0000736
## Median :0.3103449  Median :0.4166668  Median :0.0226927
## Mean    :0.3145589  Mean    :0.4216397  Mean    :0.0226927
## 3rd Qu.:0.4137932  3rd Qu.:0.5208334  3rd Qu.:0.0226927
## Max.    :1.0000001  Max.    :1.0000001  Max.    :1.0000001
##      Wage              Special              International.Reputation
## Min.    :0.0000001  Min.    :0.0000001  Min.    :0.0000001
## 1st Qu.:0.0017700  1st Qu.:0.4495357  1st Qu.:0.0000001
## Median :0.0053098  Median :0.5597524  Median :0.0000001
## Mean    :0.0172237  Mean    :0.5367245  Mean    :0.0283056
## 3rd Qu.:0.0159293  3rd Qu.:0.6538701  3rd Qu.:0.0000001
## Max.    :1.0000001  Max.    :1.0000001  Max.    :1.0000001
##      Weak.Foot      Skill.Moves              Height
## Min.    :0.0000001  Min.    :0.0000001  Min.    :0.0000001
## 1st Qu.:0.5000001  1st Qu.:0.2500001  1st Qu.:0.0055557
## Median :0.5000001  Median :0.2500001  Median :0.4444445
## Mean    :0.4868248  Mean    :0.3403272  Mean    :0.3874263
## 3rd Qu.:0.5000001  3rd Qu.:0.5000001  3rd Qu.:0.5555557
```

##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	Weight	LS	ST
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3308272	1st Qu.:0.3660567	1st Qu.:0.3660567
##	Median :0.4135339	Median :0.4462151	Median :0.4462151
##	Mean :0.4208958	Mean :0.4462151	Mean :0.4462151
##	3rd Qu.:0.4962407	3rd Qu.:0.5490850	3rd Qu.:0.5490850
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	RS	LW	LF
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3660567	1st Qu.:0.4328359	1st Qu.:0.4090910
##	Median :0.4462151	Median :0.5080554	Median :0.4806280
##	Mean :0.4462151	Mean :0.5080554	Mean :0.4806280
##	3rd Qu.:0.5490850	3rd Qu.:0.5970150	3rd Qu.:0.5757577
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	CF	RF	RW
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.4090910	1st Qu.:0.4090910	1st Qu.:0.4328359
##	Median :0.4806280	Median :0.4806280	Median :0.5080554
##	Mean :0.4806280	Mean :0.4806280	Mean :0.5080554
##	3rd Qu.:0.5757577	3rd Qu.:0.5757577	3rd Qu.:0.5970150
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	LAM	CAM	RAM
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.4090910	1st Qu.:0.4090910	1st Qu.:0.4090910
##	Median :0.4846241	Median :0.4846241	Median :0.4846241
##	Mean :0.4846241	Mean :0.4846241	Mean :0.4846241
##	3rd Qu.:0.5757577	3rd Qu.:0.5757577	3rd Qu.:0.5757577
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	LM	LCM	CM
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.4375001	1st Qu.:0.3958693	1st Qu.:0.3958693
##	Median :0.5115120	Median :0.4857648	Median :0.4857648
##	Mean :0.5115120	Mean :0.4857648	Mean :0.4857648
##	3rd Qu.:0.5937501	3rd Qu.:0.5679863	3rd Qu.:0.5679863
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	RCM	RM	LWB
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3958693	1st Qu.:0.4375001	1st Qu.:0.3992741
##	Median :0.4857648	Median :0.5115120	Median :0.4993708
##	Mean :0.4857648	Mean :0.5115120	Mean :0.4993708
##	3rd Qu.:0.5679863	3rd Qu.:0.5937501	3rd Qu.:0.5989112

##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	LDM	CDM	RDM
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3891710	1st Qu.:0.3891710	1st Qu.:0.3891710
##	Median :0.4877700	Median :0.4877700	Median :0.4877700
##	Mean :0.4877700	Mean :0.4877700	Mean :0.4877700
##	3rd Qu.:0.5922167	3rd Qu.:0.5922167	3rd Qu.:0.5922167
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	RWB	LB	LCB
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3992741	1st Qu.:0.3992741	1st Qu.:0.3542674
##	Median :0.4993708	Median :0.5046977	Median :0.4930223
##	Mean :0.4993708	Mean :0.5046977	Mean :0.4930223
##	3rd Qu.:0.5989112	3rd Qu.:0.6170600	3rd Qu.:0.6280194
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	CB	RCB	RB
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3542674	1st Qu.:0.3542674	1st Qu.:0.3992741
##	Median :0.4930223	Median :0.4930223	Median :0.5046977
##	Mean :0.4930223	Mean :0.4930223	Mean :0.5046977
##	3rd Qu.:0.6280194	3rd Qu.:0.6280194	3rd Qu.:0.6170600
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	Crossing	Finishing	HeadingAccuracy
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3750001	1st Qu.:0.3010754	1st Qu.:0.4555557
##	Median :0.5568183	Median :0.5053764	Median :0.5777779
##	Mean :0.5083431	Mean :0.4682895	Mean :0.5366461
##	3rd Qu.:0.6704546	3rd Qu.:0.6451614	3rd Qu.:0.6666668
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	ShortPassing	Volleys	Dribbling
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.5465117	1st Qu.:0.3023257	1st Qu.:0.4838711
##	Median :0.6395350	Median :0.4651164	Median :0.6129033
##	Mean :0.6010084	Mean :0.4524306	Mean :0.5523765
##	3rd Qu.:0.7093024	3rd Qu.:0.6162792	3rd Qu.:0.6881721
##	Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
##	Curve	FKAccuracy	LongPassing
##	Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
##	1st Qu.:0.3181819	1st Qu.:0.3076924	1st Qu.:0.4047620
##	Median :0.4772728	Median :0.4175825	Median :0.5595239
##	Mean :0.4678504	Mean :0.4380567	Mean :0.5203803
##	3rd Qu.:0.6363637	3rd Qu.:0.5824177	3rd Qu.:0.6547620

## Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
## BallControl	Acceleration	SprintSpeed
## Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
## 1st Qu.:0.5384616	1st Qu.:0.5294119	1st Qu.:0.5357144
## Median :0.6373627	Median :0.6470589	Median :0.6547620
## Mean :0.5864777	Mean :0.6189892	Mean :0.6277021
## 3rd Qu.:0.7032968	3rd Qu.:0.7411766	3rd Qu.:0.7500001
## Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
## Agility	Reactions	Balance
## Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
## 1st Qu.:0.5000001	1st Qu.:0.4666668	1st Qu.:0.5000001
## Median :0.6341464	Median :0.5466668	Median :0.6250001
## Mean :0.6037026	Mean :0.5444882	Mean :0.5995823
## 3rd Qu.:0.7317074	3rd Qu.:0.6266668	3rd Qu.:0.7250001
## Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
## ShotPower	Jumping	Stamina
## Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
## 1st Qu.:0.4623657	1st Qu.:0.5375001	1st Qu.:0.5238096
## Median :0.6129033	Median :0.6375001	Median :0.6428572
## Mean :0.5748393	Mean :0.6261180	Mean :0.6097614
## 3rd Qu.:0.7096775	3rd Qu.:0.7250001	3rd Qu.:0.7380953
## Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
## Strength	LongShots	Aggression
## Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
## 1st Qu.:0.5125001	1st Qu.:0.3296704	1st Qu.:0.3928572
## Median :0.6125001	Median :0.5274726	Median :0.5714287
## Mean :0.6038997	Mean :0.4847251	Mean :0.5341547
## 3rd Qu.:0.7125001	3rd Qu.:0.6483517	3rd Qu.:0.6904763
## Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
## Interceptions	Positioning	Vision
## Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
## 1st Qu.:0.2584271	1st Qu.:0.3978496	1st Qu.:0.4047620
## Median :0.5505619	Median :0.5698926	Median :0.5357144
## Mean :0.4909920	Mean :0.5156827	Mean :0.5166775
## 3rd Qu.:0.6853934	3rd Qu.:0.6666668	3rd Qu.:0.6428572
## Max. :1.0000001	Max. :1.0000001	Max. :1.0000001
## Penalties	Composure	Marking
## Min. :0.0000001	Min. :0.0000001	Min. :0.0000001
## 1st Qu.:0.3908047	1st Qu.:0.5161291	1st Qu.:0.2967034
## Median :0.5057472	Median :0.6021506	Median :0.5494506
## Mean :0.5005587	Mean :0.5983686	Mean :0.4866113
## 3rd Qu.:0.6321840	3rd Qu.:0.6881721	3rd Qu.:0.6703298



```
## Max.      :1.0000001 Max.      :1.0000001 Max.      :1.0000001
## StandingTackle SlidingTackle GKDiving
## Min.      :0.0000001 Min.      :0.0000001 Min.      :0.0000001
## 1st Qu.:0.2747254 1st Qu.:0.2386365 1st Qu.:0.0786518
## Median :0.5824177 Median :0.5568183 Median :0.1123597
## Mean    :0.5021741 Mean    :0.4847891 Mean    :0.1754633
## 3rd Qu.:0.7032968 3rd Qu.:0.6931819 3rd Qu.:0.1460675
## Max.      :1.0000001 Max.      :1.0000001 Max.      :1.0000001
## GKHandling      GKKicking      GKPositioning
## Min.      :0.0000001 Min.      :0.0000001 Min.      :0.0000001
## 1st Qu.:0.0769232 1st Qu.:0.0777779 1st Qu.:0.0786518
## Median :0.1098902 Median :0.1111112 Median :0.1123597
## Mean    :0.1691385 Mean    :0.1692452 Mean    :0.1729091
## 3rd Qu.:0.1428572 3rd Qu.:0.1444445 3rd Qu.:0.1460675
## Max.      :1.0000001 Max.      :1.0000001 Max.      :1.0000001
## GKReflexes      Release.Clause
## Min.      :0.0000001 Min.      :0.0000001
## 1st Qu.:0.0752689 1st Qu.:0.0000160
## Median :0.1075270 Median :0.0052538
## Mean    :0.1689344 Mean    :0.0052538
## 3rd Qu.:0.1397850 3rd Qu.:0.0052538
## Max.      :1.0000001 Max.      :1.0000001
```

## 2.5.2 Train set과 Test set 분할

# 모델 학습과 평가를 위해 *Train set*과 *Test set*을 7:3으로 분할한다.

```
data_shuffled <- data_scaled[sample(nrow(data_scaled)),]
r1 <- as.integer(nrow(data_shuffled)*0.7) #row number at which split
data_train <- data_shuffled[1:r1,]
data_test <- data_shuffled[(r1+1):nrow(data_shuffled),]

data_shuffled[c(1:2),]
```

```

##           Age  Overall           Value           Wage  Special
## 225  0.6551725 0.7708334 5.882363e-02 0.09911514 0.3517029
## 6525 0.4137932 0.4791668 1.088431e-05 0.01238948 0.6941177
##           International.Reputation Weak.Foot Skill.Moves      Height
## 225                0.5000001 0.5000001   0.0000001 0.6111112 0.6
## 6525                0.0000001 0.5000001   0.5000001 0.3888890 0.4
##           LS           ST           RS           LW           LF           CF
## 225  0.4462151 0.4462151 0.4462151 0.5080554 0.4806280 0.4806280
## 6525 0.5823628 0.5823628 0.5823628 0.6567165 0.6212122 0.6212122
##           RW           LAM           CAM           RAM           LM           LCM
## 225  0.5080554 0.4846241 0.4846241 0.4846241 0.5115120 0.4857648
## 6525 0.6567165 0.6212122 0.6212122 0.6212122 0.6562501 0.6196214
##           RCM           RM           LWB           LDM           CDM           RDM
## 225  0.4857648 0.5115120 0.4993708 0.4877700 0.4877700 0.4877700
## 6525 0.6196214 0.6562501 0.6170600 0.5583757 0.5583757 0.5583757
##           LB           LCB           CB           RCB           RB  Crossing
## 225  0.5046977 0.4930223 0.4930223 0.4930223 0.5046977 0.06818192
## 6525 0.5989112 0.4830919 0.4830919 0.4830919 0.5989112 0.70454555
##           HeadingAccuracy ShortPassing  Volleys  Dribbling      Curv
## 225                0.1000001   0.1395350 0.08139545 0.06451623 0.0568182
## 6525                0.4777779   0.6976745 0.54651173 0.74193558 0.7840910
##           LongPassing BallControl Acceleration SprintSpeed  Agility R
## 225  0.1547620 0.1978023 0.4470589 0.6071430 0.4390245 0
## 6525 0.7142858 0.6813188 0.8000001 0.7500001 0.6951221 0
##           Balance ShotPower  Jumping  Stamina  Strength LongShots
## 225  0.5500001 0.2365592 0.7250001 0.2619049 0.6875001 0.1318682
## 6525 0.7000001 0.7741936 0.4375001 0.6666668 0.5625001 0.7692309

##           Interceptions Positioning  Vision Penalties Composure  Mar
## 225  0.2134832 0.1182797 0.654762 0.2183909 0.6559141 0.186
## 6525 0.7303372 0.7634410 0.607143 0.6206898 0.7204302 0.373
##           StandingTackle SlidingTackle  GK Diving GKHandling GK Kicking
## 225  0.2197803 0.2386365 0.8876405 0.9010990 0.9000001
## 6525 0.5714287 0.6250001 0.1123597 0.0989012 0.1000001
##           GKReflexes Release.Clause
## 225  0.8709678 5.558818e-05
## 6525 0.1075270 2.539041e-06

```

```
data_train[c(1:2),]
```

##	Age	Overall	Value	Wage	Special	
## 225	0.6551725	0.7708334	5.882363e-02	0.09911514	0.3517029	
## 6525	0.4137932	0.4791668	1.088431e-05	0.01238948	0.6941177	
##	International.Reputation	Weak.Foot	Skill.Moves	Height		
## 225	0.5000001	0.5000001	0.0000001	0.6111112	0.6	
## 6525	0.0000001	0.5000001	0.5000001	0.3888890	0.4	
##	LS	ST	RS	LW	LF	CF
## 225	0.4462151	0.4462151	0.4462151	0.5080554	0.4806280	0.4806280
## 6525	0.5823628	0.5823628	0.5823628	0.6567165	0.6212122	0.6212122
##	RW	LAM	CAM	RAM	LM	LCM
## 225	0.5080554	0.4846241	0.4846241	0.4846241	0.5115120	0.4857648
## 6525	0.6567165	0.6212122	0.6212122	0.6212122	0.6562501	0.6196214
##	RCM	RM	LWB	LDM	CDM	RDM
## 225	0.4857648	0.5115120	0.4993708	0.4877700	0.4877700	0.4877700
## 6525	0.6196214	0.6562501	0.6170600	0.5583757	0.5583757	0.5583757
##	LB	LCB	CB	RCB	RB	Crossing
## 225	0.5046977	0.4930223	0.4930223	0.4930223	0.5046977	0.06818192
## 6525	0.5989112	0.4830919	0.4830919	0.4830919	0.5989112	0.70454555
##	HeadingAccuracy	ShortPassing	Volleys	Dribbling	Curv	
## 225	0.1000001	0.1395350	0.08139545	0.06451623	0.0568182	
## 6525	0.4777779	0.6976745	0.54651173	0.74193558	0.7840910	
##	LongPassing	BallControl	Acceleration	SprintSpeed	Agility	R
## 225	0.1547620	0.1978023	0.4470589	0.6071430	0.4390245	0
## 6525	0.7142858	0.6813188	0.8000001	0.7500001	0.6951221	0
##	Balance	ShotPower	Jumping	Stamina	Strength	LongShots
## 225	0.5500001	0.2365592	0.7250001	0.2619049	0.6875001	0.1318682
## 6525	0.7000001	0.7741936	0.4375001	0.6666668	0.5625001	0.7692309
##	Interceptions	Positioning	Vision	Penalties	Composure	Mar
## 225	0.2134832	0.1182797	0.654762	0.2183909	0.6559141	0.186
## 6525	0.7303372	0.7634410	0.607143	0.6206898	0.7204302	0.373
##	StandingTackle	SlidingTackle	GK Diving	GK Handling	GK Kicking	
## 225	0.2197803	0.2386365	0.8876405	0.9010990	0.9000001	
## 6525	0.5714287	0.6250001	0.1123597	0.0989012	0.1000001	
##	GK Reflexes	Release.Clause				
## 225	0.8709678	5.558818e-05				
## 6525	0.1075270	2.539041e-06				

```
data_test[c(1:2),]
```

```
##           Age    Overall    value    wage    special
## 14288 0.2413794 0.3125001 0.02269274 0.003539923 0.1962849
## 15642 0.2068967 0.2708334 0.02269274 0.014159392 0.1851394
##           International.Reputation Weak.Foot Skill.Moves    Height
## 14288                    1e-07 0.2500001            1e-07 0.7222223 0.
## 15642                    1e-07 0.2500001            1e-07 0.6666668 0.
##           LS      ST      RS      LW      LF      CF
## 14288 0.4462151 0.4462151 0.4462151 0.5080554 0.480628 0.480628 0
## 15642 0.4462151 0.4462151 0.4462151 0.5080554 0.480628 0.480628 0
##           RW      LAM      CAM      RAM      LM      LCM
## 14288 0.5080554 0.4846241 0.4846241 0.4846241 0.511512 0.4857648
## 15642 0.5080554 0.4846241 0.4846241 0.4846241 0.511512 0.4857648
##           RCM      RM      LWB      LDM      CDM      RDM
## 14288 0.4857648 0.511512 0.4993708 0.48777 0.48777 0.48777 0.4993
## 15642 0.4857648 0.511512 0.4993708 0.48777 0.48777 0.48777 0.4993
##           LCB      CB      RCB      RB    Crossing Finishin
## 14288 0.4930223 0.4930223 0.4930223 0.5046977 0.09090919 0.172043
## 15642 0.4930223 0.4930223 0.4930223 0.5046977 0.07954555 0.129032
##           HeadingAccuracy ShortPassing    Volleys Dribbling    Curve
## 14288      0.1111112      0.2441861 0.1744187 0.1397850 0.10227283
## 15642      0.1222223      0.2209303 0.1046513 0.1290324 0.07954555
##           LongPassing BallControl Acceleration SprintSpeed    Agility
## 14288      0.1309525      0.1538463      0.3294119      0.3571430 0.2073172
## 15642      0.2380953      0.1538463      0.4352942      0.3452382 0.2560977
##           Balance ShotPower    Jumping    Stamina    Strength    LongShot
## 14288 0.3250001 0.1935485 0.4750001 0.2023811 0.7625001 0.1208792
## 15642 0.5125001 0.2258066 0.5500001 0.3452382 0.4250001 0.0879121
##           Interceptions Positioning    Vision    Penalties Composure
## 14288      0.2471911      0.1827958 0.2500001 0.18390815 0.2688173 0.
## 15642      0.2471911      0.1182797 0.3095239 0.09195412 0.4516130 0.
##           StandingTackle SlidingTackle    GK Diving GK Handling GK Kicking
## 14288      0.1868133      0.1022728 0.6966293      0.6043957 0.5777779
## 15642      0.1318682      0.1818183 0.6741574      0.5934067 0.5888890
##           GK Reflexes Release.Clause
## 14288      0.6666668      0.005253827
## 15642      0.6559141      0.005253827
```

```
not_training_features <- c("Overall")
x_train <- as.matrix(data_train[, ! names(data_train) %in% not_train
y_train <- as.matrix(data_train[, "Overall", drop = F])
```

```
x_test <- as.matrix(data_test[, ! names(data_test) %in% not_training])
y_test <- as.matrix(data_test[, "Overall", drop = F])

print("Dimensions :")
```

```
## [1] "Dimensions :"
```

```
print("x_train : ")
```

```
## [1] "x_train : "
```

```
print(dim(x_train))
```

```
## [1] 12744    70
```

```
print("x_test : ")
```

```
## [1] "x_test : "
```

```
print(dim(x_test))
```

```
## [1] 5463    70
```

```
print("y_train : ")
```

```
## [1] "y_train : "
```

```
print(dim(y_train))
```

```
## [1] 12744      1
```

```
print("y_test : ")
```

```
## [1] "y_test : "
```

```
print(dim(y_test))
```

```
## [1] 5463      1
```

## 2.5.1 다중회귀선형모델 첫 Fitting

```
#다중선형회귀모델을 Fitting한다.
```

```
lin_model1<- Overall~Age+Value+Wage+Special+International.Reputation
```

```
fit1<-lm(lin_model1,data_train)
```

```
summary(fit1)
```

```
##
```

```
## Call:
```

```
## lm(formula = lin_model1, data = data_train)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
## -0.285555 -0.031844  0.000765  0.033002  0.198360
```

```
##
```

```
## Coefficients: (16 not defined because of singularities)
```

```
##              Estimate Std. Error t value Pr(>|t|)
```

## (Intercept)	-0.343128	0.031778	-10.798	< 2e-16	**
## Age	0.038337	0.003735	10.266	< 2e-16	**
## Value	-0.074122	0.013351	-5.552	2.88e-08	**
## Wage	0.409815	0.017787	23.040	< 2e-16	**
## Special	0.387188	0.093414	4.145	3.42e-05	**
## International.Reputation	0.047591	0.006422	7.411	1.33e-13	**
## Weak.Foot	0.004786	0.002935	1.631	0.103020	
## Skill.Moves	0.118945	0.004733	25.131	< 2e-16	**
## Height	0.015126	0.002143	7.058	1.78e-12	**
## Weight	0.029057	0.006317	4.600	4.27e-06	**
## LS	0.130226	0.059640	2.184	0.029016	*
## ST	NA	NA	NA	NA	
## RS	NA	NA	NA	NA	
## LW	0.090495	0.090023	1.005	0.314800	
## LF	-0.273815	0.088417	-3.097	0.001960	**
## CF	NA	NA	NA	NA	
## RF	NA	NA	NA	NA	
## RW	NA	NA	NA	NA	
## LAM	-0.222606	0.085120	-2.615	0.008928	**
## CAM	NA	NA	NA	NA	
## RAM	NA	NA	NA	NA	
## LM	0.166823	0.087318	1.911	0.056087	.
## LCM	-0.127102	0.065327	-1.946	0.051721	.
## CM	NA	NA	NA	NA	
## RCM	NA	NA	NA	NA	
## RM	NA	NA	NA	NA	
## LWB	0.091453	0.070928	1.289	0.197293	
## LDM	0.386897	0.071778	5.390	7.16e-08	**
## CDM	NA	NA	NA	NA	
## RDM	NA	NA	NA	NA	
## RWB	NA	NA	NA	NA	
## LB	0.015017	0.068851	0.218	0.827350	
## LCB	-0.299031	0.056771	-5.267	1.41e-07	**
## CB	NA	NA	NA	NA	
## RCB	NA	NA	NA	NA	
## RB	NA	NA	NA	NA	
## Crossing	-0.065420	0.014224	-4.599	4.28e-06	**
## Finishing	0.036184	0.014560	2.485	0.012963	*
## HeadingAccuracy	0.167247	0.012356	13.536	< 2e-16	**
## ShortPassing	0.132579	0.011695	11.336	< 2e-16	**
## Volleys	-0.041118	0.007703	-5.338	9.55e-08	**
## Dribbling	0.010128	0.015933	0.636	0.525033	

```
## Curve -0.007660 0.007642 -1.002 0.316183
## FKAccuracy -0.018693 0.007405 -2.525 0.011596 *
## LongPassing -0.072783 0.011394 -6.388 1.74e-10 **
## BallControl 0.222560 0.012677 17.556 < 2e-16 **
## Acceleration 0.029447 0.010060 2.927 0.003427 **
## SprintSpeed 0.022052 0.009340 2.361 0.018246 *
## Agility -0.013149 0.007978 -1.648 0.099355 .
## Reactions 0.370212 0.008455 43.789 < 2e-16 **
## Balance -0.049659 0.006787 -7.317 2.69e-13 **
## ShotPower 0.017012 0.011443 1.487 0.137118
## Jumping -0.001661 0.006352 -0.262 0.793656
## Stamina -0.033313 0.010397 -3.204 0.001358 **
## Strength 0.027074 0.008716 3.106 0.001899 **
## LongShots -0.027424 0.009943 -2.758 0.005822 **
## Aggression -0.018210 0.008102 -2.248 0.024614 *
## Interceptions -0.068016 0.012747 -5.336 9.66e-08 **
## Positioning -0.073190 0.012264 -5.968 2.47e-09 **
## Vision -0.020129 0.009077 -2.217 0.026609 *
## Penalties -0.021999 0.007414 -2.967 0.003009 **
## Composure 0.173944 0.006737 25.819 < 2e-16 **
## Marking 0.036942 0.010613 3.481 0.000502 **
## StandingTackle 0.015104 0.014950 1.010 0.312363
## SlidingTackle -0.045329 0.014159 -3.201 0.001371 **
## GKDiving 0.099761 0.013231 7.540 5.02e-14 **
## GKHandling 0.103507 0.013665 7.574 3.86e-14 **
## GK Kicking 0.036626 0.012725 2.878 0.004005 **
## GKPositioning 0.111748 0.012975 8.612 < 2e-16 **
## GKReflexes 0.119826 0.013670 8.766 < 2e-16 **
## Release.Clause 0.039520 0.022258 1.776 0.075830 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.05032 on 12689 degrees of freedom
## Multiple R-squared: 0.8767, Adjusted R-squared: 0.8761
## F-statistic: 1670 on 54 and 12689 DF, p-value: < 2.2e-16
```

```
print("Number of Parameters Learnt = ")
```

```
## [1] "Number of Parameters Learnt = "
```



```
print(length(fit1$coefficients))
```

```
## [1] 71
```

$p$ -value가 0.05보다 작으므로, 귀무가설이 기각되고 따라서 회귀성이 충족됨을 알 수 있고 이를 모델에 포함시키고자한다.

다중공선성이 존재함을 확인

상관 행렬을 사용하여 공선형 회귀 변수를 찾는다.

## 2.5.2 상관행렬을 이용한 공선형 회귀변수 확인

```
corr_x_train <- cor(x_train)
dim(corr_x_train)
```

```
## [1] 70 70
```

```
name_col_rem <- findCorrelation(corr_x_train, cutoff = 0.95, verbose
print("Regressors to Remove: ")
```

```
## [1] "Regressors to Remove: "
```

```
name_col_rem
```

```
## [1] "LCM"      "CM"       "LAM"      "CAM"
## [5] "RAM"      "LM"       "RM"       "LW"
## [9] "RW"       "LF"       "CF"       "LS"
## [13] "ST"       "RS"       "RWB"      "SlidingTack
```

```
## [17] "LDM"          "CDM"          "RDM"          "LB"
## [21] "RB"          "GKDividing"   "GKHandling"   "GKReflexes"
## [25] "GKKicking"    "CB"           "RCB"
```

### 2.5.3 다른 회귀자와 $> 0.95$ 의 상관 관계를 가지고 있기 때문에 제거해야하는 Regressor를 확인 및 이를 제거

```
cols.dont.want <- name_col_rem
data_train <- data_train[, ! names(data_train) %in% cols.dont.want,
data_test <- data_test[, ! names(data_test) %in% cols.dont.want, dro
dim(data_train)
```

```
## [1] 12744    44
```

```
dim(data_test)
```

```
## [1] 5463    44
```

```
not_training_features <- c("Overall")
x_train <- as.matrix(data_train[, ! names(data_train) %in% not_train
y_train <- as.matrix(data_train[, "Overall", drop = F])
x_test <- as.matrix(data_test[, ! names(data_test) %in% not_training
y_test <- as.matrix(data_test[, "Overall", drop = F])
#Regressor의 숫자가 44개로 줄어들었음을 확인할 수 있다.
```

### 2.5.4 선형모델 재적합

```
new_colnames <- colnames(x_train)
print("New Column Names")
```

```
## [1] "New Column Names"
```

```
new_colnames
```

```
## [1] "Age" "Value"
## [3] "Wage" "Special"
## [5] "International.Reputation" "Weak.Foot"
## [7] "Skill.Moves" "Height"
## [9] "Weight" "RF"
## [11] "RCM" "LWB"
## [13] "LCB" "Crossing"
## [15] "Finishing" "HeadingAccuracy"
## [17] "ShortPassing" "Volleys"
## [19] "Dribbling" "Curve"
## [21] "FKAccuracy" "LongPassing"
## [23] "BallControl" "Acceleration"
## [25] "SprintSpeed" "Agility"
## [27] "Reactions" "Balance"
## [29] "ShotPower" "Jumping"
## [31] "Stamina" "Strength"
## [33] "LongShots" "Aggression"
## [35] "Interceptions" "Positioning"
## [37] "Vision" "Penalties"
## [39] "Composure" "Marking"
## [41] "StandingTackle" "GKPositioning"
## [43] "Release.Clause"
```

```
model2_formula <- Overall~Age+Value+Wage+Special+International.Reput

lin_model2<-model2_formula
fit2<-lm(lin_model2,data_train)
summary(fit2)
```

```
##
## Call:
## lm(formula = lin_model2, data = data_train)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-0.301249	-0.032216	0.000604	0.033521	0.289856

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	-0.045271	0.016688	-2.713	0.00668	**
Age	0.040020	0.003785	10.572	< 2e-16	**
Value	-0.073563	0.013545	-5.431	5.70e-08	**
Wage	0.413992	0.018043	22.944	< 2e-16	**
Special	1.285366	0.056090	22.916	< 2e-16	**
International.Reputation	0.047869	0.006517	7.345	2.19e-13	**
Weak.Foot	0.005120	0.002978	1.719	0.08560	.
Skill.Moves	0.114960	0.004792	23.992	< 2e-16	**
Height	0.015637	0.002175	7.189	6.88e-13	**
Weight	0.032811	0.006405	5.122	3.06e-07	**
RCM	0.065054	0.026367	2.467	0.01363	*
LWB	0.069668	0.040485	1.721	0.08530	.
LCB	-0.101317	0.042927	-2.360	0.01828	*
Crossing	-0.078703	0.009846	-7.994	1.42e-15	**
Finishing	-0.020836	0.007955	-2.619	0.00883	**
HeadingAccuracy	0.090816	0.008674	10.469	< 2e-16	**
ShortPassing	0.069679	0.010278	6.779	1.26e-11	**
Volleys	-0.086557	0.006417	-13.489	< 2e-16	**
Dribbling	-0.093233	0.010196	-9.144	< 2e-16	**
Curve	-0.058275	0.006526	-8.930	< 2e-16	**
FKAccuracy	-0.071163	0.006155	-11.561	< 2e-16	**
LongPassing	-0.108008	0.008972	-12.038	< 2e-16	**
BallControl	0.144636	0.011487	12.591	< 2e-16	**
Acceleration	-0.013309	0.008642	-1.540	0.12359	
SprintSpeed	-0.024154	0.007947	-3.040	0.00237	**
Agility	-0.058664	0.006334	-9.262	< 2e-16	**
Reactions	0.341274	0.007708	44.277	< 2e-16	**
Balance	-0.097858	0.005759	-16.992	< 2e-16	**
ShotPower	-0.035120	0.007034	-4.993	6.04e-07	**
Jumping	-0.051278	0.005017	-10.221	< 2e-16	**
Stamina	-0.051299	0.007105	-7.220	5.48e-13	**
Strength	-0.009970	0.007325	-1.361	0.17351	

```
## LongShots          -0.108121    0.007694 -14.052 < 2e-16 **
## Aggression         -0.058856    0.006259  -9.404 < 2e-16 **
## Interceptions      -0.091095    0.007748 -11.757 < 2e-16 **
## Positioning        -0.169971    0.007985 -21.287 < 2e-16 **
## Vision            -0.103632    0.007586 -13.661 < 2e-16 **
## Penalties         -0.069204    0.006365 -10.872 < 2e-16 **
## Composure          0.177975    0.006825  26.078 < 2e-16 **
## Marking            -0.004702    0.007966  -0.590  0.55499
## StandingTackle     -0.091654    0.011294  -8.115 5.29e-16 **
## GKPositioning       0.155367    0.012802  12.136 < 2e-16 **
## Release.Clause      0.039519    0.022593   1.749 0.08030 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.05109 on 12701 degrees of freedom
## Multiple R-squared:  0.8727, Adjusted R-squared:  0.8723
## F-statistic: 2074 on 42 and 12701 DF, p-value: < 2.2e-16
```

```
print("Number of Parameters Learnt = ")
```

```
## [1] "Number of Parameters Learnt = "
```

```
print(length(fit2$coefficients))
```

```
## [1] 43
```

## 2.5.5 개별 가설 검정 :

이제  $p\text{-value} < 0.05$  인 매개 변수만 고려하고 다른 매개 변수는 중요하지 않기 때문에 다음 선형 모델을 선택한 회귀 변수에 맞춰준다.

```
#p-value < 0.05를 기준으로 모든 회귀 변수의 포함 / 제외를 위해 참 / 거짓
toselect.x <- summary(fit2)$coeff[-1,4] < 0.05
#p-value가 0.05보다 작은 regressors를 저장
relevant.x <- names(toselect.x)[toselect.x == TRUE]
```

```
#유의한 변수들만 남은 선형회귀
sig.formula <- as.formula(paste("Overall ~", paste(relevant.x, collapse=" + "), sep=""))
print("New Formula ")
```

```
## [1] "New Formula "
```

```
print(sig.formula)
```

```
## Overall ~ Age + Value + Wage + Special + International.Reputation
##      Skill.Moves + Height + Weight + RCM + LCB + Crossing + Finish
##      HeadingAccuracy + ShortPassing + Volleys + Dribbling + Curve
##      FKAaccuracy + LongPassing + BallControl + SprintSpeed + Agilit
##      Reactions + Balance + ShotPower + Jumping + Stamina + LongSho
##      Aggression + Interceptions + Positioning + Vision + Penalties
##      Composure + StandingTackle + GKPositioning
```

```
lin_model3 <- sig.formula
fit3<-lm(sig.formula,data_train)
summary(fit3)
```

```
##
## Call:
## lm(formula = sig.formula, data = data_train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.303756 -0.032115  0.000599  0.033651  0.274724
##
## Coefficients:
```

```

... -----
##
## (Intercept)      -0.061577    0.014808   -4.159 3.22e-05 **
## Age              0.038905    0.003737   10.410 < 2e-16 **
## Value            -0.070880    0.013410   -5.285 1.27e-07 **
## Wage             0.415470    0.018043   23.027 < 2e-16 **
## Special          1.232565    0.049161   25.072 < 2e-16 **
## International.Reputation 0.049353    0.006499    7.594 3.32e-14 **
## Skill.Moves      0.114459    0.004771   23.991 < 2e-16 **
## Height           0.015460    0.002171    7.122 1.12e-12 **
## Weight           0.026868    0.005898    4.556 5.27e-06 **
## RCM              0.111363    0.017127    6.502 8.21e-11 **
## LCB              -0.067558    0.021622   -3.125 0.00179 **
## Crossing         -0.061515    0.006100  -10.084 < 2e-16 **
## Finishing        -0.020393    0.007774   -2.623 0.00872 **
## HeadingAccuracy   0.086423    0.006461   13.376 < 2e-16 **
## ShortPassing     0.069459    0.010071    6.897 5.58e-12 **
## Volleys          -0.083690    0.006257  -13.376 < 2e-16 **
## Dribbling        -0.090559    0.010028   -9.031 < 2e-16 **
## Curve            -0.055116    0.006348   -8.683 < 2e-16 **
## FkAccuracy       -0.068086    0.005939  -11.464 < 2e-16 **
## LongPassing      -0.113029    0.008193  -13.796 < 2e-16 **
## BallControl      0.143611    0.011245   12.771 < 2e-16 **
## SprintSpeed      -0.021269    0.006603   -3.221 0.00128 **
## Agility          -0.055708    0.006233   -8.938 < 2e-16 **
## Reactions        0.345866    0.007568   45.699 < 2e-16 **
## Balance          -0.093824    0.005520  -16.997 < 2e-16 **
## ShotPower        -0.032583    0.006854   -4.754 2.01e-06 **
## Jumping          -0.050282    0.004508  -11.154 < 2e-16 **
## Stamina          -0.044733    0.005606   -7.980 1.59e-15 **
## LongShots        -0.108987    0.007439  -14.650 < 2e-16 **
## Aggression       -0.060359    0.004910  -12.292 < 2e-16 **
## Interceptions    -0.084842    0.007343  -11.555 < 2e-16 **
## Positioning      -0.172855    0.007654  -22.585 < 2e-16 **
## Vision           -0.105717    0.006964  -15.181 < 2e-16 **
## Penalties        -0.066690    0.006184  -10.785 < 2e-16 **
## Composure        0.177374    0.006819   26.012 < 2e-16 **
## StandingTackle   -0.083705    0.009648   -8.676 < 2e-16 **
## GKPositioning    0.165692    0.011767   14.081 < 2e-16 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.05111 on 12707 degrees of freedom

```

```
## Residual standard error: 37.511 on 12707 deg. of freedom
## Multiple R-squared:  0.8726, Adjusted R-squared:  0.8722
## F-statistic: 2417 on 36 and 12707 DF,  p-value: < 2.2e-16
```

```
print("Number of Parameters Learnt = ")
```

```
## [1] "Number of Parameters Learnt = "
```

```
print(length(fit3$coefficients))
```

```
## [1] 37
```

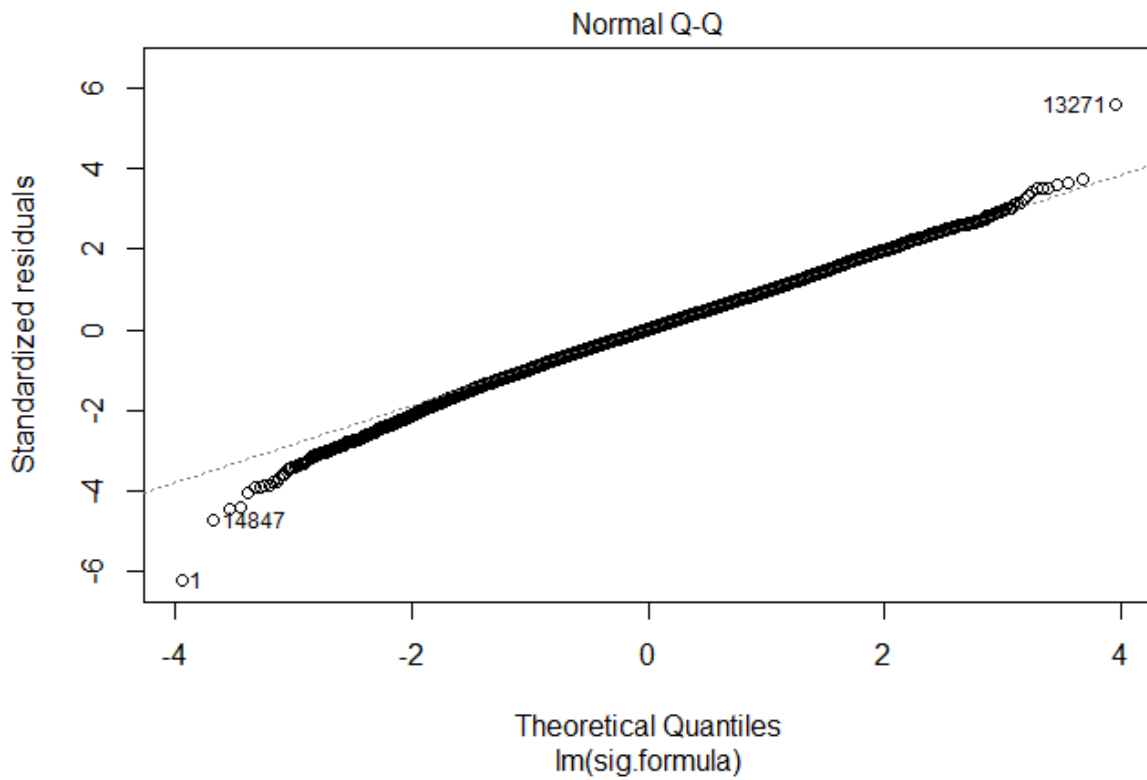
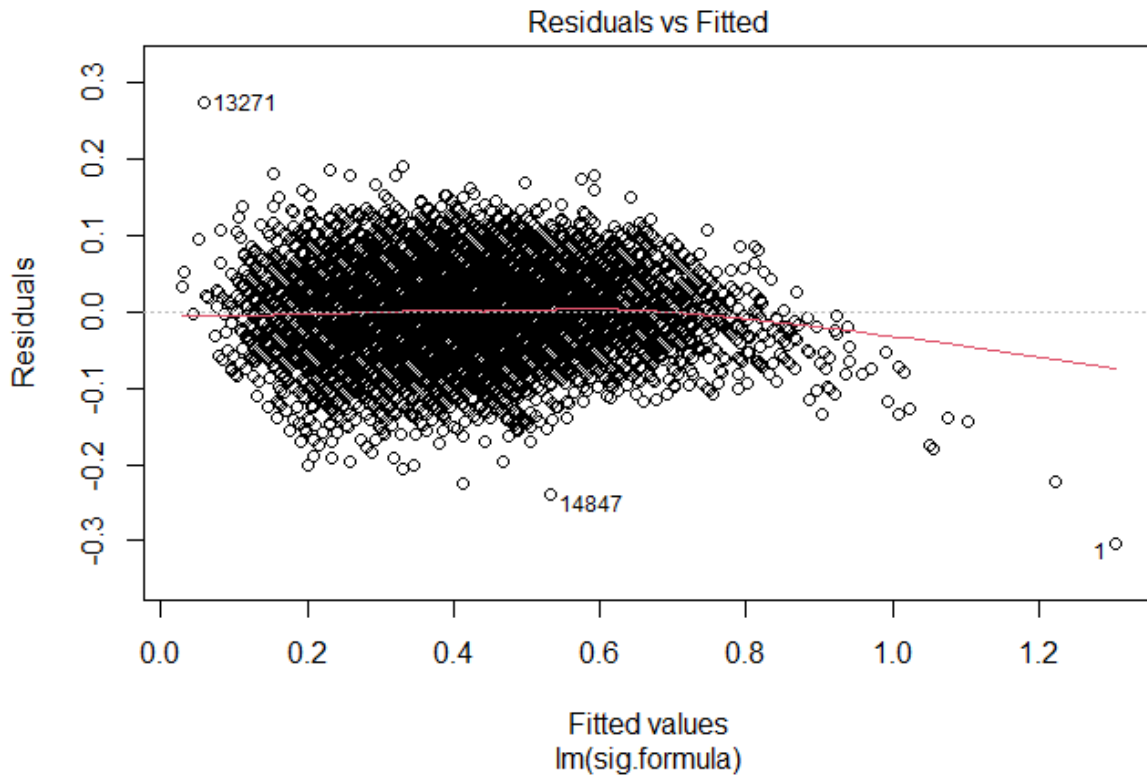
중간결과 : 학습 된 매개 변수 수를 초기 값 71에서 36으로 줄 였지만 *adjusted R<sup>2</sup>*는 거의 동일하다.

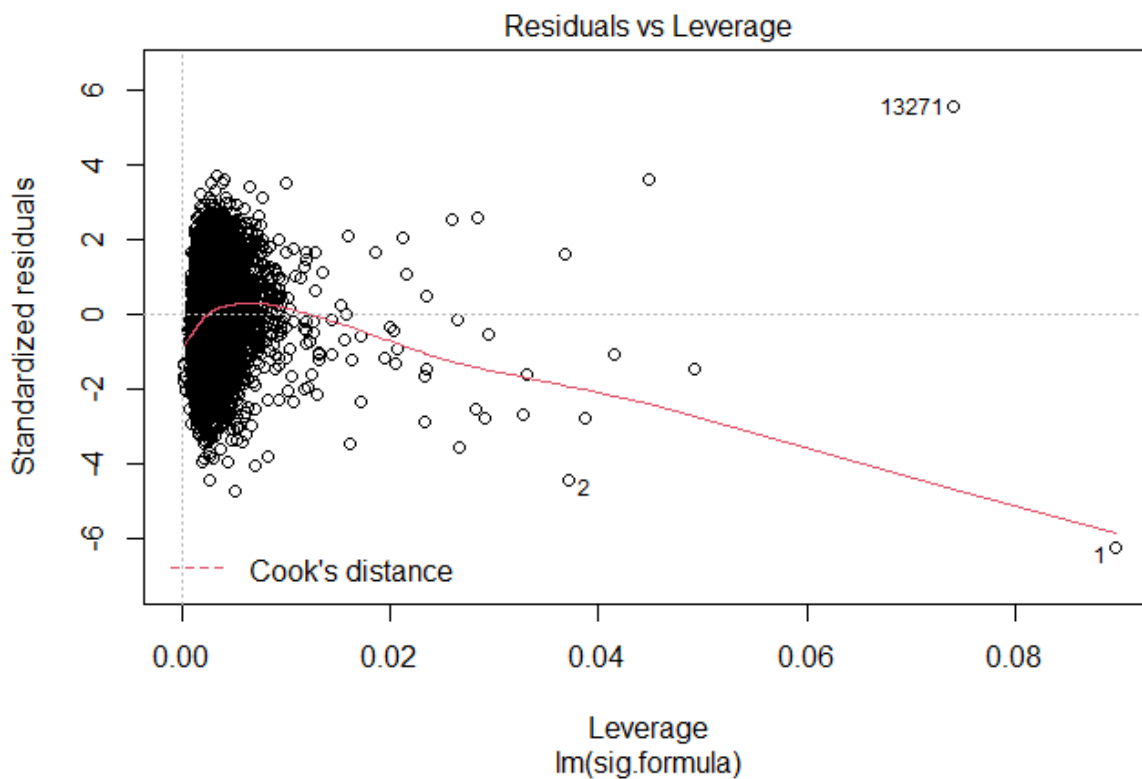
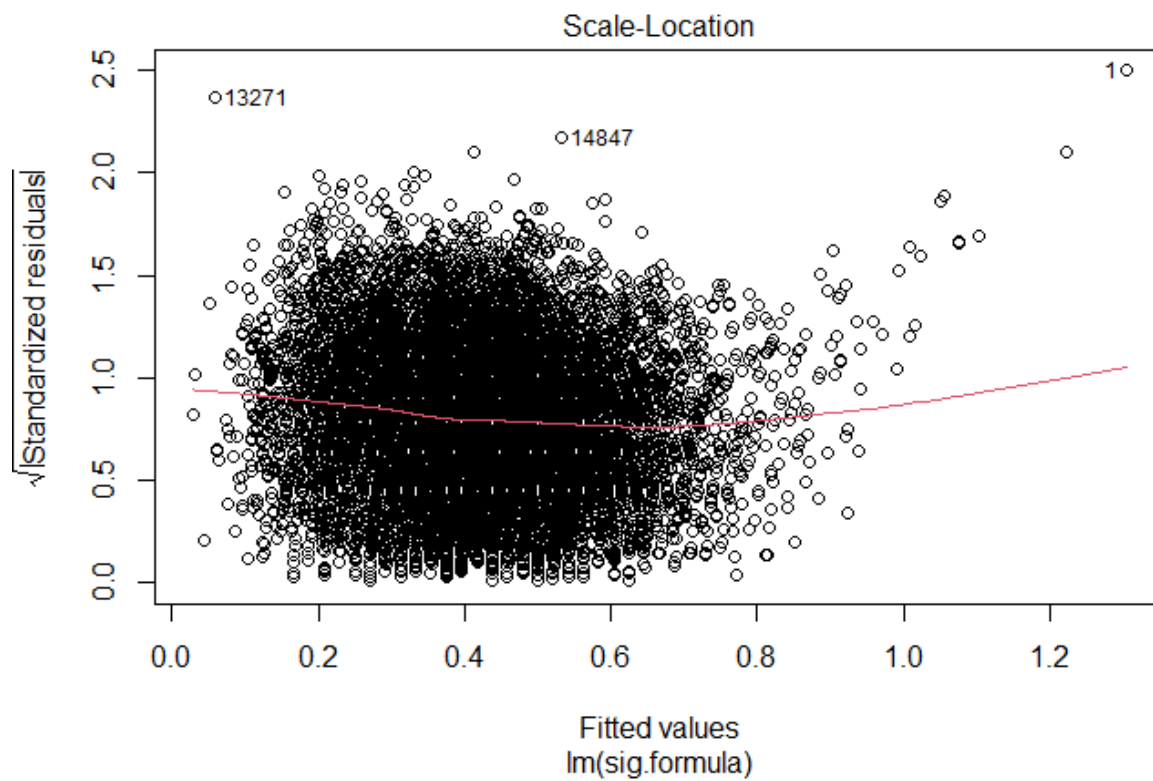
## 2.5.6 QQ Plot등 잔차확인

```
# 잔차확인 -
```

```
plot(fit3)
```



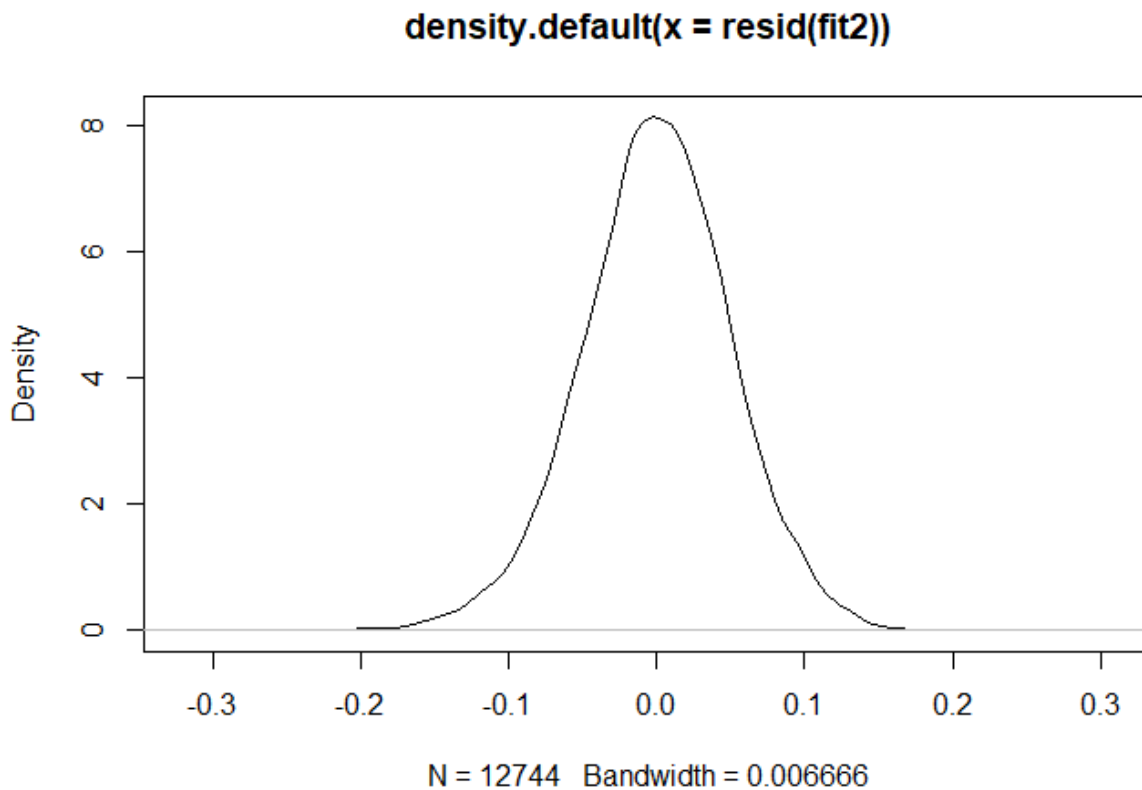




#

# Observations -

```
#  
# Residuals vs Fitted : Most Residuals are equally spread along th  
# Normal QQ Plot : Data is normally distributed except few outliers.  
# Scale Location Plot: Data is evenly spread except for outliers whe  
# Residuals vs Leverage : Cook's Distance Lines are not seen in plot  
  
# Density Plot of Residuals Looks like it has a Normal Distribution  
  
plot(density(resid(fit2)))
```



## 2.5.7 Shapiro-Wilk Normalcy Test

*#Applying Shapiro-Wilk's Normalcy Test :  $H_0$  : Sample is Normal,  $H_a$*

```
shapiro.test(resid(fit2)[1:5000])
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: resid(fit2)[1:5000]  
## W = 0.99811, p-value = 9.219e-06
```

```
shapiro.test(resid(fit2)[5001:10000])
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: resid(fit2)[5001:10000]  
## W = 0.99607, p-value = 2.691e-10
```

```
shapiro.test(resid(fit2)[10001:length(resid(fit2))])
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: resid(fit2)[10001:length(resid(fit2))]  
## W = 0.99482, p-value = 3.157e-08
```

```
x <- replicate(1000, { #각 분포에 1000개의 다른 test를 만들어준다
  c(shapiro.test(rnorm(1000)+c(0,0,1,0,0))$p.value, #$
    shapiro.test(rnorm(5000)+c(0,0,1,0,0))$p.value) #$
}) # rnorm은 정규분포로부터 random한 draw를 준다.
)
rownames(x) <- c("n1000", "n5000")
rowMeans(x<0.05)
```

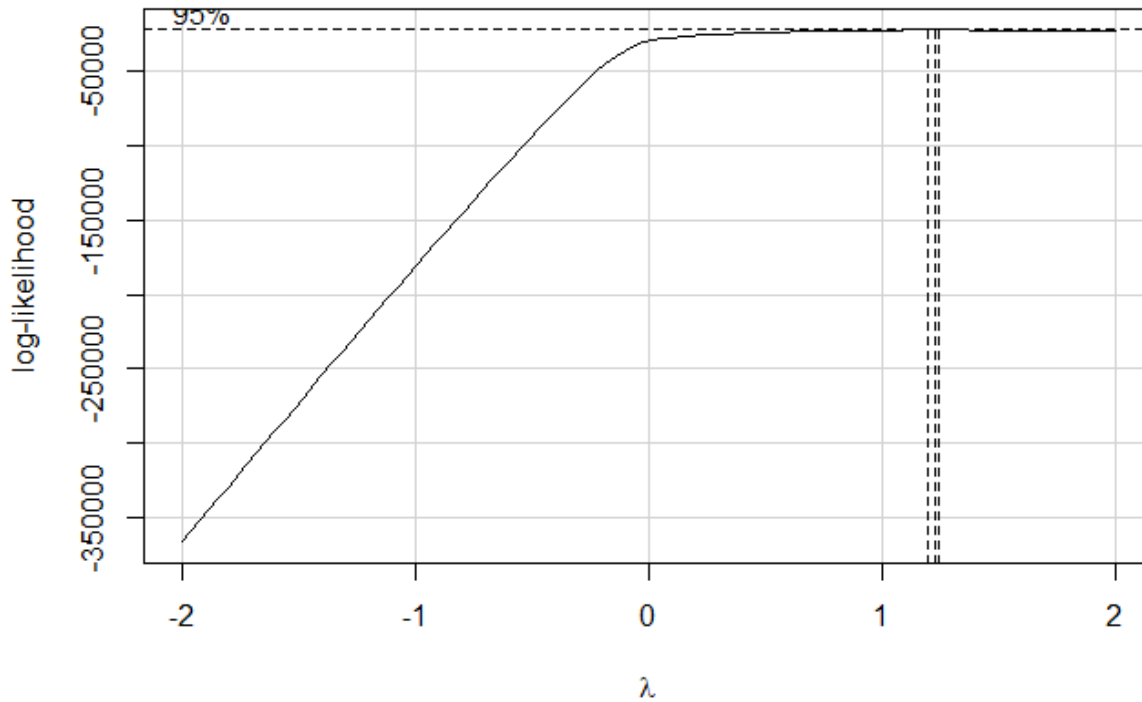
```
## n1000 n5000
## 0.132 0.459
```

잔차는 정규적으로 분포함을 확인할 수 있다

### ###2.5.8 Box-Cox Transformation

```
# BOX-COX TRANSFORMATION을 사용한다.
```

```
model4_formula <- as.formula(paste("data_train$Overall ~", paste("dat
lambda<-boxCox(model4_formula, objective.name="Log-Likelihood", plotit
```



```
ind<-which(lambda$y == max(lambda$y))
lambda.max<-lambda$x[ind]
data_train$Overall<-bcPower(data_train$Overall,lambda = lambda.max)
data_test$Overall<-bcPower(data_test$Overall,lambda = lambda.max)
not_training_features <- c("Overall")
x_train <- as.matrix(data_train[, ! names(data_train) %in% not_train
y_train <- as.matrix(data_train[, "Overall", drop = F])
x_test <- as.matrix(data_test[, ! names(data_test) %in% not_training
y_test <- as.matrix(data_test[, "Overall", drop = F])

print("Applied BoxCox. Fitting New Model")
```

```
## [1] "Applied BoxCox. Fitting New Model"
```

```
print(data_train$Overall[c(1,3)])
```

```
## [1] -0.2226667 -0.5184911
```

```
print(data_test$Overall[c(1,3)])
```

```
## [1] -0.6179369 -0.6179369
```

```
model4_formula <- as.formula(paste("Overall~",paste(relevant.x, coll
model4 <-lm(model4_formula
fit4<-lm(model4,data_train)
summary(fit4)
```

```
##
## Call:
## lm(formula = model4, data = data_train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.27131 -0.02601 -0.00020  0.02669  0.20890
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.909600   0.011796  -77.113  < 2e-16 **
## Age           0.023268   0.002977   7.815  5.90e-15 **
##
## Value        -0.041316   0.010683   -3.868  0.00011 **
## Wage          0.406294   0.014373   28.268  < 2e-16 **
## Special       0.965508   0.039162   24.654  < 2e-16 **
## International.Reputation 0.063416   0.005177   12.249  < 2e-16 **
## Skill.Moves   0.094076   0.003801   24.753  < 2e-16 **
## Height        0.011595   0.001729    6.705  2.09e-11 **
## Weight        0.019403   0.004698    4.130  3.65e-05 **
## RCM           0.088910   0.013643    6.517  7.46e-11 **
## LCB          -0.050605   0.017224   -2.938  0.00331 **
## Crossing      -0.049822   0.004860  -10.252  < 2e-16 **
## Finishing     -0.015799   0.006193   -2.551  0.01075 *
## HeadingAccuracy 0.068067   0.005147   13.225  < 2e-16 **
## ShortPassing  0.054140   0.008023    6.748  1.56e-11 **
## Volleys       -0.065520   0.004984  -13.146  < 2e-16 **
## Dribbling     -0.071669   0.007988   -8.972  < 2e-16 **
## Curve         -0.042596   0.005056   -8.424  < 2e-16 **
## FKAccuracy    -0.052627   0.004731  -11.124  < 2e-16 **
```

```

## LongPassing      -0.087681    0.006526 -13.435 < 2e-16 **
## BallControl      0.105026    0.008958  11.724 < 2e-16 **
## SprintSpeed     -0.016985    0.005260  -3.229 0.00124 **
## Agility         -0.045491    0.004965  -9.162 < 2e-16 **
## Reactions        0.280890    0.006029  46.590 < 2e-16 **
## Balance         -0.073722    0.004397 -16.765 < 2e-16 **
## ShotPower       -0.024565    0.005460  -4.499 6.87e-06 **
## Jumping         -0.039479    0.003591 -10.993 < 2e-16 **
## Stamina         -0.035939    0.004466  -8.048 9.15e-16 **
## LongShots       -0.086528    0.005926 -14.601 < 2e-16 **
## Aggression      -0.046575    0.003912 -11.907 < 2e-16 **
## Interceptions   -0.066467    0.005849 -11.364 < 2e-16 **
## Positioning     -0.135381    0.006097 -22.205 < 2e-16 **
## Vision          -0.080834    0.005547 -14.571 < 2e-16 **
## Penalties       -0.051351    0.004926 -10.424 < 2e-16 **
## Composure       0.141796    0.005432  26.104 < 2e-16 **
## StandingTackle  -0.067019    0.007685  -8.720 < 2e-16 **
## GKPositioning   0.129315    0.009374  13.795 < 2e-16 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.04071 on 12707 degrees of freedom
## Multiple R-squared:  0.8778, Adjusted R-squared:  0.8775
## F-statistic: 2537 on 36 and 12707 DF, p-value: < 2.2e-16

```

```
print("Number of Coeff learnt =")
```

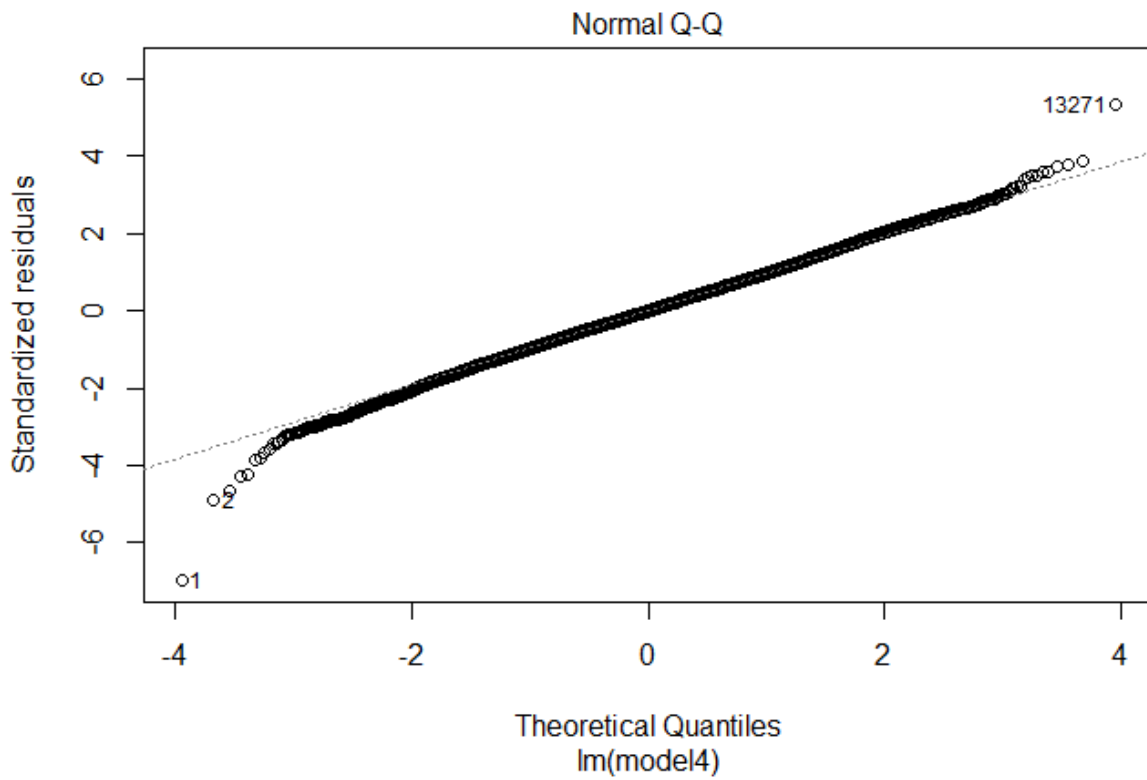
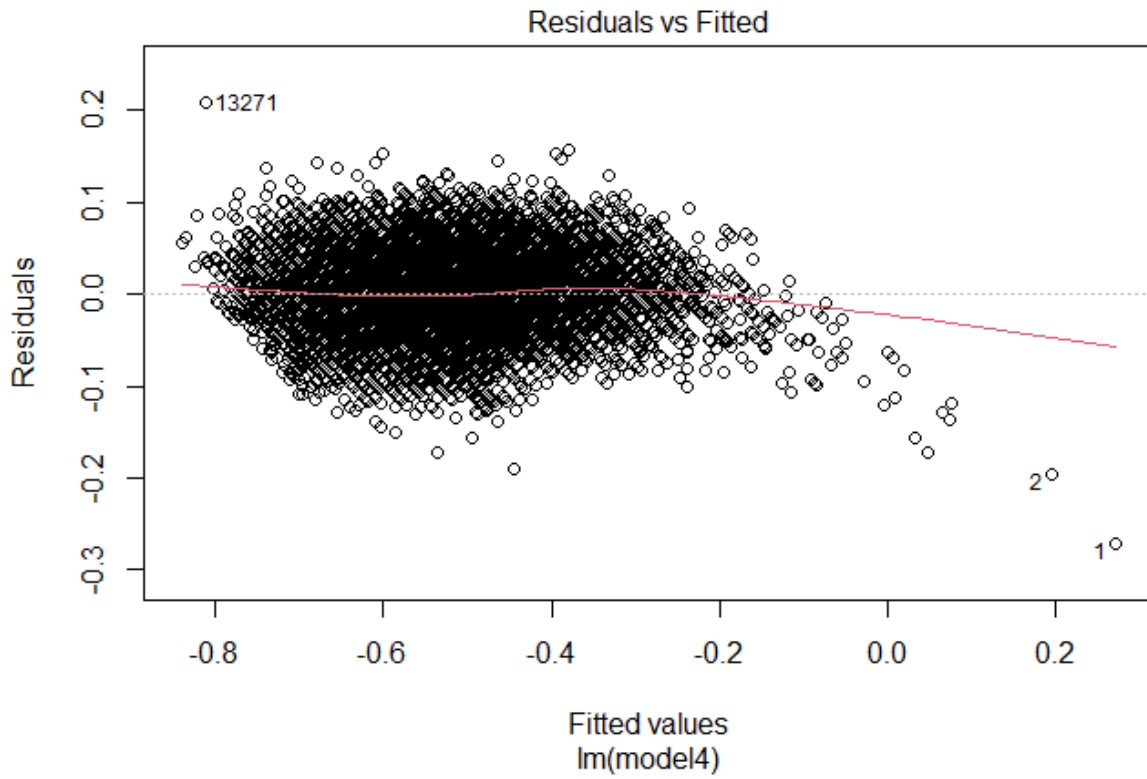
```
## [1] "Number of Coeff learnt ="
```

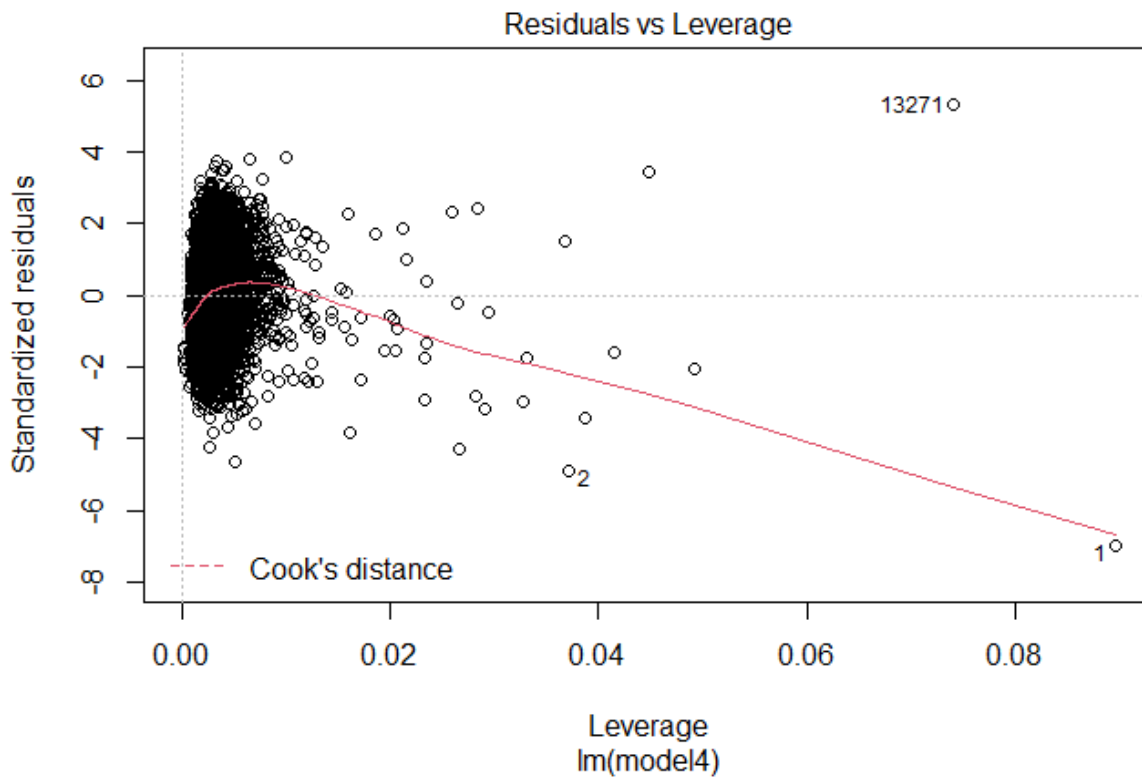
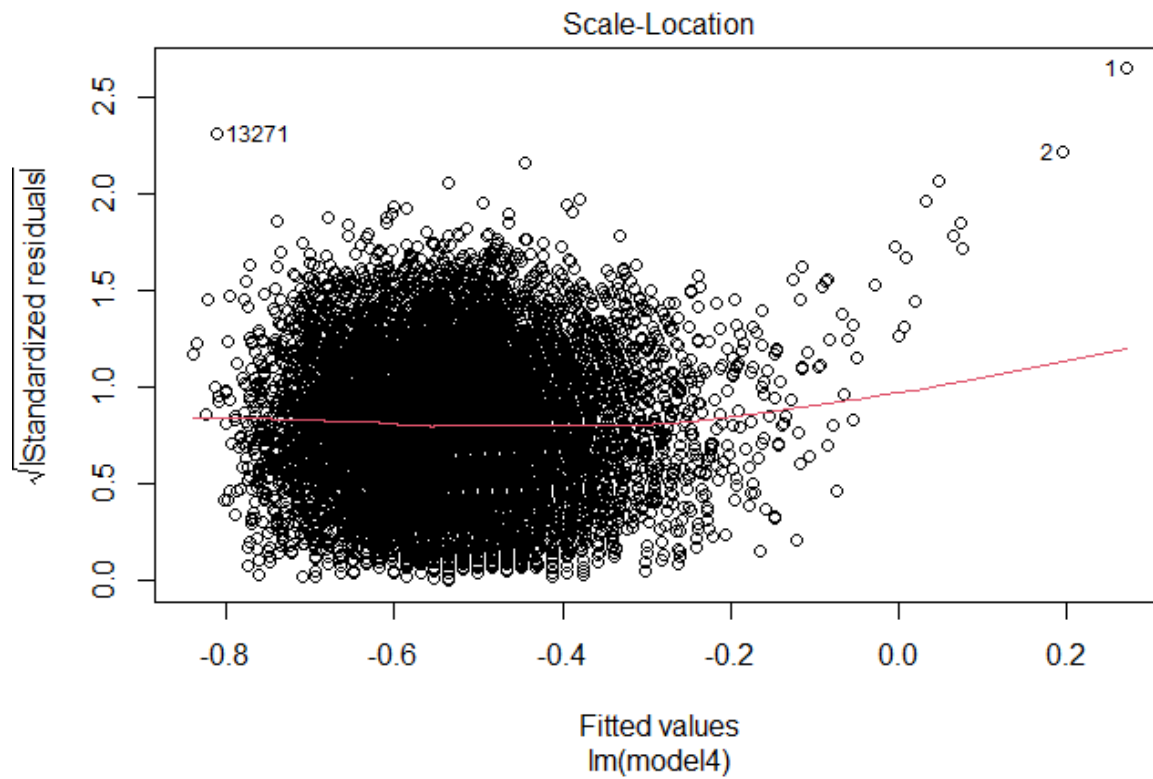
```
print(length(fit4$coefficients))
```

```
## [1] 37
```

```
plot(fit4)
```







중간결과 : 모델 개선을 통해  $R^2$  값과  $Adjusted R^2$  값이 유의미하게 상승했음을 확인할 수 있다.

&lt;br

## 2.5.9 모델에서 이상치 확인 및 제거

```
##### Handling Outliers :
```

```
p=dim(data_train)[2]
n=dim(data_train)[1]
print("Original Dim of Training Set = ")
```

```
## [1] "Original Dim of Training Set = "
```

```
print(dim(data_train))
```

```
## [1] 12744 44
```

```
data_train_outrem <- data_train
```

```
 #(1)Using Leverage방법론
```

```
print("Detecting Outliers using Leverage")
```

```
## [1] "Detecting Outliers using Leverage"
```

```
lev<- lm.influence(fit1)$hat
lev_rem <- lev>3*(p+1)/n
print("Number of Outliers Found = ")
```

```
## [1] "Number of Outliers Found = "
```

```
print(sum(lev_rem == TRUE))
```

```
## [1] 208
```

```
#(2)Cook's Distance방법론  
print("Detecting Outliers using Cook's Distance")
```

```
## [1] "Detecting Outliers using Cook's Distance"
```

```
cd<-cooks.distance(fit1)  
cd_rem <- cd>1  
print("Number of Outliers Found = ")
```

```
## [1] "Number of Outliers Found = "
```

```
print(sum(cd_rem == TRUE))
```

```
## [1] 0
```

```
#(3)DFBETAS방법론  
print("Detecting Outliers using DFBETAS")
```

```
## [1] "Detecting Outliers using DFBETAS"
```

```
dfb<-dfbeta(fit1)  
dim(dfb)
```

```
## [1] 12744    55
```

```
n_dfb = dim(dfb)[1]
p_dfb = dim(dfb)[2]
#Intializing dfb_rem
dfb_rem <- cd_rem
for (i in c(1:n_dfb))
{
  dfb_rem[i]=FALSE
  for (j in c(1:p_dfb))
    if (dfb[i,j]>(2/sqrt(n)))
      dfb_rem[i]<-TRUE
}
print("Number of Outliers Found = ")
```

```
## [1] "Number of Outliers Found = "
```

```
print(sum(dfb_rem == TRUE))
```

```
## [1] 0
```

```
#(4)DFFITS방법론
print("Detecting Outliers using DFFITS")
```

```
## [1] "Detecting Outliers using DFFITS"
```

```
dfft<-dffits(fit1)
dfft_rem <- dfft>(2*sqrt((p+1)/n))
print("Number of Outliers Found = ")
```

```
## [1] "Number of Outliers Found = "
```

```
print(sum(dfft_rem == TRUE))
```

```
## [1] 507
```

#(5) COVRATIO 방법론

```
print("Detecting Outliers using COVRATIO")
```

```
## [1] "Detecting Outliers using COVRATIO"
```

```
covr<-covratio(fit1)
covr_rem <- ( covr>(1 + (3*((p+1)/n))) | covr<(1 - (3*((p+1)/n)) ) )
print("Number of Outliers Found = ")
```

```
## [1] "Number of Outliers Found = "
```

```
print(sum(covr_rem == TRUE))
```

```
## [1] 950
```

#위의 모든 방법으로 검출한 이상치들을 제거해준다

```
data_train_outrem <- data_train
out_rem <- lev_rem | cd_rem | dfb_rem | dfft_rem | covr_rem
data_train_outrem <- data_train_outrem[out_rem==FALSE,]
dim(data_train_outrem)
```

```
## [1] 11462    44
```

```
lin_model5<-sig.formula
fit5<-lm(lin_model5,data_train_outrem)
summary(fit5)
```

```
##
## Call:
## lm(formula = lin_model5, data = data_train_outrem)

##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.146630 -0.023886  0.000155  0.024617  0.101435
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.852792    0.012527  -68.074 < 2e-16 **
## Age             0.025292    0.002737   9.240 < 2e-16 **
## Value          -0.080360    0.014871  -5.404 6.66e-08 **
## Wage           0.664776    0.019077  34.847 < 2e-16 **
## Special        1.106625    0.043196  25.619 < 2e-16 **
## International.Reputation 0.033516    0.005308   6.314 2.83e-10 **
## Skill.Moves     0.086996    0.003410  25.513 < 2e-16 **
## Height          0.011040    0.001531   7.211 5.93e-13 **
## Weight          0.017310    0.004262   4.062 4.90e-05 **
## RCM             0.122705    0.013778   8.906 < 2e-16 **
## LCB            -0.095683    0.017746  -5.392 7.11e-08 **
## Crossing       -0.056562    0.004562 -12.398 < 2e-16 **
## Finishing      -0.029861    0.005789  -5.158 2.54e-07 **
## HeadingAccuracy 0.067144    0.004702  14.280 < 2e-16 **
## ShortPassing    0.048573    0.007715   6.296 3.17e-10 **
## Volleys        -0.073092    0.004660 -15.684 < 2e-16 **
## Dribbling      -0.090467    0.007580 -11.935 < 2e-16 **
## Curve          -0.050955    0.004701 -10.840 < 2e-16 **
## FKAaccuracy    -0.056201    0.004485 -12.530 < 2e-16 **
## LongPassing    -0.102665    0.006335 -16.207 < 2e-16 **
## BallControl     0.100719    0.008547  11.785 < 2e-16 **
## SprintSpeed    -0.027524    0.005234  -5.259 1.48e-07 **
## Agility        -0.047373    0.004759  -9.954 < 2e-16 **
## Reactions       0.258430    0.005725  45.139 < 2e-16 **
## Balance        -0.082239    0.004185 -19.651 < 2e-16 **
## ShotPower      -0.029838    0.005149  -5.795 7.03e-09 **
## Jumping        -0.043826    0.003391 -12.923 < 2e-16 **
## Stamina        -0.049211    0.004373 -11.254 < 2e-16 **
## LongShots      -0.097242    0.005610 -17.333 < 2e-16 **
## Aggression     -0.053215    0.003574 -14.891 < 2e-16 **
## Interceptions  -0.069126    0.005353 -12.913 < 2e-16 **
```

```
## Positioning      -0.150757    0.005770 -26.128 < 2e-16 **
## Vision          -0.101010    0.005565 -18.150 < 2e-16 **
## Penalties       -0.060900    0.004681 -13.011 < 2e-16 **
## Composure       0.139787    0.005018  27.856 < 2e-16 **

## StandingTackle  -0.071835    0.007126 -10.080 < 2e-16 **
## GKPositioning   0.079294    0.009336   8.494 < 2e-16 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.03443 on 11425 degrees of freedom
## Multiple R-squared:  0.8979, Adjusted R-squared:  0.8976
## F-statistic: 2792 on 36 and 11425 DF, p-value: < 2.2e-16
```

중간점검 : 이상치 제거 후 *adjusted R<sup>2</sup>* 값이 상승한것을 확인할 수 있다

## 2.5.10 Test set을 통한 y값 예측

```
##차원확인
dim(data_test)
```

```
## [1] 5463 44
```

```
dim(data_train)
```

```
## [1] 12744 44
```

```
dim(x_train)
```

```
## [1] 12744 43
```



```
dim(x_test)
```

```
## [1] 5463 43
```

```
final_prediction <- predict(fit5, newdata=as.data.frame(x_test)) #최종 예측값

#length(final_prediction)
print("See first 10 Prediction ")
```

```
## [1] "See first 10 Prediction "
```

```
print(final_prediction[c(1:10)])
```

```
##      14288      15642      14068      10288      12638      6241
## -0.6251796 -0.6278009 -0.5711662 -0.5215758 -0.5707499 -0.5208227
##      3498      17069      2396
## -0.4974333 -0.7295942 -0.4456081
```

```
print("See first 10 Actual ")
```

```
## [1] "See first 10 Actual "
```

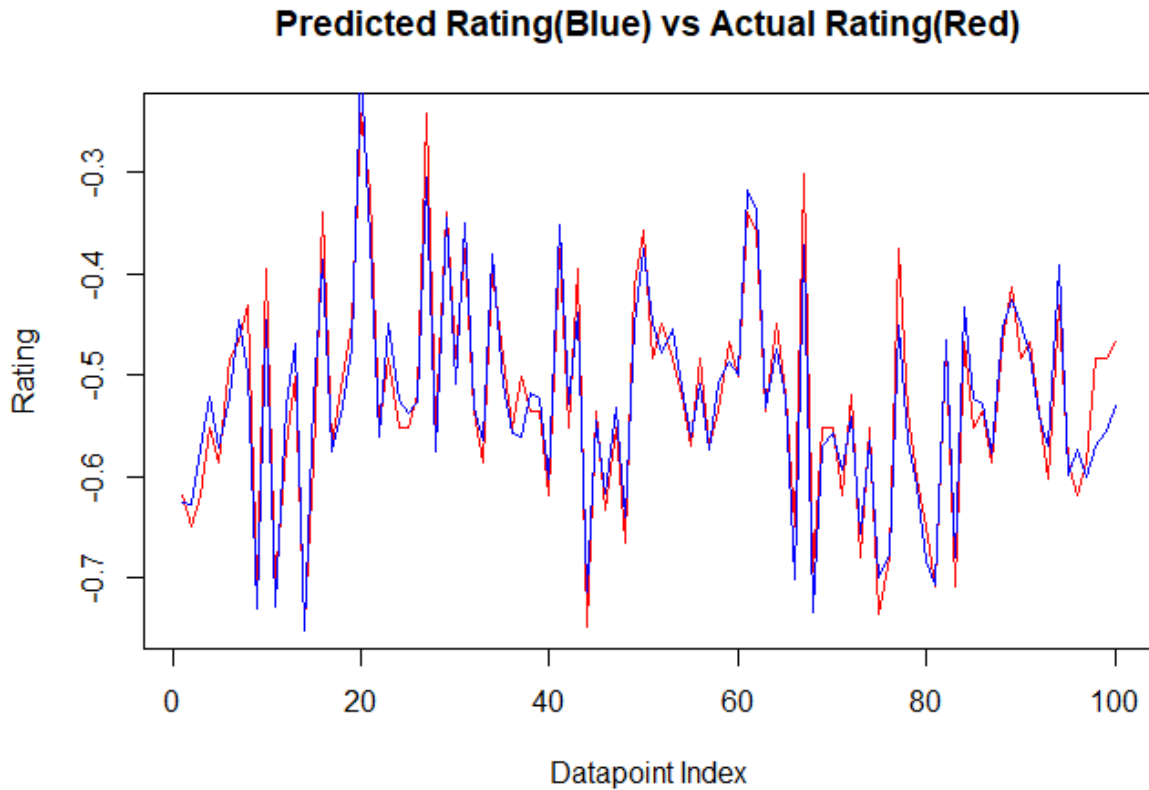
```
print(y_test[c(1:10)])
```

```
## [1] -0.6179369 -0.6492267 -0.6179369 -0.5524868 -0.5856600 -0.48
## [7] -0.4660858 -0.4302803 -0.7083470 -0.3938285
```

```
#Plotting Predicted and Actual Curve for Test Set (Only first 100 Va
```

```
ind = c(1:100)

plot(ind,y_test[c(1:100)],type='l',col="red",xlab="Datapoint Index",
lines(ind,final_prediction[c(1:100)],col="blue")
```



### 2.5.11 변수선택법 : Step-Forward AIC method

```
#Finding RMSE
rmse_val <- sqrt(norm(y_test-final_prediction)/n)
print("RMSE=")
```

```
## [1] "RMSE="
```

```
print(rmse_val)
```

```
## [1] 0.1169155
```

```
# - step-forward AIC method
```

```
lin_model6<-sig.formula
data_new<-as.data.frame(data_train)
step_model<-lm(lin_model6,data_new)
ols_step_forward_aic(step_model,details = TRUE)
```

```
## Forward Selection Method
## -----
##
## Candidate Terms:
##
## 1 . Age
## 2 . Value
## 3 . Wage
## 4 . Special
## 5 . International.Reputation
## 6 . Skill.Moves
## 7 . Height
## 8 . Weight
## 9 . RCM
## 10 . LCB
## 11 . Crossing
## 12 . Finishing
## 13 . HeadingAccuracy
## 14 . ShortPassing
## 15 . Volleys
## 16 . Dribbling
## 17 . Curve
## 18 . FKAccuracy
## 19 . LongPassing
## 20 . BallControl
## 21 . SprintSpeed
## 22 . Agility
## 23 . Reactions
```

```
## 24 . Balance
## 25 . ShotPower
## 26 . Jumping
## 27 . Stamina
## 28 . LongShots
## 29 . Aggression
## 30 . Interceptions
## 31 . Positioning
## 32 . Vision
## 33 . Penalties
## 34 . Composure
## 35 . StandingTackle
## 36 . GKPositioning
##
## Step 0: AIC = -18666.29
## Overall ~ 1
##
```

```
## -----
## Variable                DF      AIC      Sum Sq      RSS
## -----
## Reactions                1    -34899.647    124.188     48.22
## Composure                1    -28038.310     89.788     82.62
## RCM                      1    -27788.920     88.156     84.26
## Wage                     1    -24297.022     61.595    110.82
## Special                  1    -24251.446     61.198    111.21
## International.Reputation 1    -22869.438     48.459    123.95
## Vision                   1    -22295.368     42.747    129.66
## ShortPassing             1    -22190.358     41.674    130.74
## LongPassing              1    -21918.262     38.853    133.56
## BallControl              1    -21490.024     34.288    138.12
## Age                     1    -21344.312     32.700    139.71
## ShotPower                1    -21284.354     32.041    140.37
## LCB                     1    -21160.865     30.674    141.74
## Curve                   1    -21016.235     29.056    143.36
## LongShots                1    -20996.918     28.839    143.57
## Skill.Moves              1    -20990.183     28.763    143.65
## FkAccuracy               1    -20777.112     26.341    146.07
## Volleys                  1    -20685.418     25.286    147.13
## Crossing                 1    -20681.029     25.236    147.18
## Aggression               1    -20676.249     25.180    147.23
## Dribbling                1    -20455.950     22.613    149.80
## Stamina                  1    -20306.135     20.842    151.57
```

```

## Positioning      1      -20278.232      20.509      151.90
## Penalties       1      -20140.601      18.860      153.55
## HeadingAccuracy  1      -20056.473      17.843      154.57
## Finishing        1      -20055.088      17.826      154.59
## Interceptions    1      -19937.451      16.392      156.02
## Agility          1      -19516.121      11.148      161.26
## Value            1      -19499.746      10.941      161.47
## Jumping          1      -19491.274      10.833      161.58
## StandingTackle   1      -19434.382      10.110      162.30
## SprintSpeed      1      -19193.517       7.013      165.40
## Weight           1      -18963.516       4.001      168.41
## Balance          1      -18784.888       1.624      170.79
## Height           1      -18689.677       0.343      172.07
## GKPositioning    1      -18664.337       0.001      172.41
## -----
##
##
## - Reactions
##
##
## Step 1 : AIC = -34899.65
## Overall ~ Reactions
##
## -----
## Variable          DF          AIC          Sum Sq          RSS
## -----
## Wage              1      -36992.285          7.310          40.919
## Composure         1      -36773.392          6.601          41.628
## RCM               1      -36575.118          5.948          42.280
## International.Reputation  1      -36230.456          4.789          43.439
## Special           1      -35545.687          2.391          45.837
## ShortPassing      1      -35375.764          1.776          46.453
## Skill.Moves       1      -35374.164          1.770          46.459
## LongPassing       1      -35362.397          1.727          46.501
## LCB              1      -35309.168          1.532          46.696
## ShotPower         1      -35272.574          1.398          46.830
## BallControl       1      -35245.913          1.300          46.928
## Weight            1      -35213.479          1.181          47.048
## Vision            1      -35210.894          1.171          47.058
## Value             1      -35204.163          1.146          47.082
## Curve             1      -35138.523          0.903          47.326
## LongShots         1      -35086.843          0.711          47.518

```

## HeadingAccuracy	1	-35072.374	0.657	47.572
## Age	1	-35069.204	0.645	47.584
## Crossing	1	-35069.356	0.645	47.583
## FKAaccuracy	1	-35066.820	0.636	47.593
## Volleys	1	-35063.467	0.623	47.605
## Dribbling	1	-35056.394	0.597	47.632
## Height	1	-35050.186	0.574	47.655
## Aggression	1	-35025.207	0.480	47.748
## Finishing	1	-35019.524	0.459	47.770
## Stamina	1	-35003.220	0.398	47.831
## Penalties	1	-34993.331	0.361	47.868
## SprintSpeed	1	-34987.204	0.338	47.891
## Jumping	1	-34975.070	0.292	47.936
## StandingTackle	1	-34950.479	0.200	48.029
## GKPositioning	1	-34947.091	0.187	48.042
## Interceptions	1	-34942.323	0.169	48.060
## Balance	1	-34937.001	0.149	48.080
## Positioning	1	-34931.602	0.128	48.100
## Agility	1	-34927.644	0.113	48.115
## -----				
##				
## - Wage				
##				
##				
## Step 2 : AIC = -36992.29				
## Overall ~ Reactions + Wage				
##				
## -----				
## Variable	DF	AIC	Sum Sq	RSS
## -----				
## Composure	1	-38689.025	5.106	35.812
## RCM	1	-38400.620	4.287	36.632
## Special	1	-37573.351	1.830	39.089
## LCB	1	-37423.171	1.367	39.552
## LongPassing	1	-37404.383	1.308	39.611
## ShortPassing	1	-37401.437	1.299	39.620
## Age	1	-37363.428	1.181	39.738
## Skill.Moves	1	-37355.832	1.157	39.762
## Weight	1	-37317.973	1.039	39.880
## ShotPower	1	-37317.231	1.036	39.882
## BallControl	1	-37274.730	0.903	40.016
## Vision	1	-37218.233	0.725	40.193

## Curve	1	-37174.600	0.588	40.331
## International.Reputation	1	-37173.897	0.585	40.333
## HeadingAccuracy	1	-37159.785	0.541	40.378
## Aggression	1	-37154.373	0.523	40.395
## LongShots	1	-37148.741	0.506	40.413
## Crossing	1	-37132.837	0.455	40.464
## FKAccuracy	1	-37126.694	0.436	40.483
## Height	1	-37123.016	0.424	40.495
## Stamina	1	-37117.410	0.406	40.513
## Dribbling	1	-37102.690	0.359	40.559
## Volleys	1	-37100.470	0.352	40.567
## Jumping	1	-37078.002	0.281	40.638
## Finishing	1	-37069.890	0.255	40.664
## SprintSpeed	1	-37056.670	0.213	40.706
## StandingTackle	1	-37052.971	0.201	40.718
## Interceptions	1	-37052.244	0.198	40.720
## Value	1	-37051.653	0.197	40.722
## Penalties	1	-37050.045	0.191	40.727
## GKPositioning	1	-37046.903	0.181	40.737
## Balance	1	-37044.419	0.173	40.745
## Agility	1	-37013.215	0.074	40.845
## Positioning	1	-37008.214	0.058	40.861
## -----				
##				
## - Composure				
##				
##				
## Step 3 : AIC = -38689.03				
## Overall ~ Reactions + Wage + Composure				
##				
## -----				
## Variable	DF	AIC	Sum Sq	RSS
## -----				
## GKPositioning	1	-39708.277	2.758	33.054
## RCM	1	-39523.033	2.274	33.538
## Weight	1	-39326.608	1.753	34.059
## LCB	1	-39155.495	1.293	34.520
## Balance	1	-39143.087	1.259	34.553
## Positioning	1	-38991.836	0.846	34.966
## Height	1	-38989.721	0.841	34.972
## Age	1	-38911.423	0.625	35.187
## Penalties	1	-38838.790	0.424	35.388

```

## Dribbling 1 -38831.983 0.405 35.407
## International.Reputation 1 -38821.041 0.375 35.438
## Volleys 1 -38802.933 0.324 35.488
## Agility 1 -38795.712 0.304 35.508
## Finishing 1 -38794.953 0.302 35.510
## LongShots 1 -38763.828 0.215 35.597
## FKAccuracy 1 -38751.722 0.181 35.631
## Curve 1 -38746.829 0.168 35.645
## BallControl 1 -38744.046 0.160 35.652
## Crossing 1 -38743.436 0.158 35.654
## Value 1 -38721.774 0.098 35.715
## Stamina 1 -38717.136 0.085 35.728
## Jumping 1 -38715.414 0.080 35.733
## SprintSpeed 1 -38707.763 0.058 35.754
## HeadingAccuracy 1 -38703.808 0.047 35.765
## Vision 1 -38701.948 0.042 35.770
## ShotPower 1 -38699.228 0.034 35.778

## ShortPassing 1 -38693.792 0.019 35.793
## StandingTackle 1 -38692.300 0.015 35.798
## Aggression 1 -38690.714 0.010 35.802
## Interceptions 1 -38690.643 0.010 35.802
## Special 1 -38687.754 0.002 35.810
## Skill.Moves 1 -38687.093 0.000 35.812
## LongPassing 1 -38687.041 0.000 35.812
## -----
##
## - GKPositioning
##
##
## Step 4 : AIC = -39708.28
## Overall ~ Reactions + Wage + Composure + GKPositioning
##
## -----
## Variable DF AIC Sum Sq RSS
## -----
## Special 1 -41097.419 3.418 29.636
## ShortPassing 1 -40751.146 2.602 30.452
## BallControl 1 -40729.875 2.551 30.503
## HeadingAccuracy 1 -40476.788 1.939 31.115
## RCM 1 -40267.812 1.425 31.630
## Stamina 1 -40181.607 1.210 31.844
## LCB 1 -40162.087 1.161 31.893

```



## Skill.Moves	1	-40159.758	1.156	31.899
## LongPassing	1	-40129.446	1.080	31.975
## ShotPower	1	-40106.478	1.022	32.033
## Aggression	1	-40027.100	0.822	32.233
## Dribbling	1	-39972.977	0.685	32.370
## Weight	1	-39968.544	0.673	32.381
## Crossing	1	-39948.819	0.623	32.431
## StandingTackle	1	-39943.207	0.609	32.446
## SprintSpeed	1	-39933.201	0.583	32.471
## Interceptions	1	-39898.083	0.494	32.561
## Curve	1	-39831.695	0.324	32.731
## Jumping	1	-39816.895	0.286	32.769
## LongShots	1	-39810.524	0.269	32.785
## Height	1	-39805.696	0.257	32.798
## International.Reputation	1	-39789.512	0.215	32.839
## Balance	1	-39783.676	0.200	32.854
## FKAccuracy	1	-39766.438	0.156	32.899
## Age	1	-39763.058	0.147	32.908
## Finishing	1	-39746.312	0.104	32.951
## Volleys	1	-39743.990	0.098	32.957
## Value	1	-39741.367	0.091	32.964
## Penalties	1	-39740.965	0.090	32.965
## Vision	1	-39724.669	0.048	33.007
## Agility	1	-39724.298	0.047	33.008
## Positioning	1	-39716.307	0.026	33.028
## -----				
##				
## - Special				
##				
##				
## Step 5 : AIC = -41097.42				
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special				
##				
## -----				
## Variable	DF	AIC	Sum Sq	RSS
## -----				
## HeadingAccuracy	1	-42242.820	2.552	27.084
## Balance	1	-41972.437	1.971	27.665
## Weight	1	-41877.542	1.764	27.872
## Height	1	-41553.724	1.047	28.589
## Vision	1	-41506.067	0.940	28.696
## Positioning	1	-41473.583	0.866	28.770

## FKAccuracy	1	-41425.112	0.757	28.879
## LCB	1	-41423.230	0.753	28.884
## Agility	1	-41421.402	0.748	28.888
## Curve	1	-41381.229	0.657	28.979
## LongShots	1	-41311.919	0.499	29.137
## Crossing	1	-41279.107	0.424	29.212
## ShortPassing	1	-41250.278	0.358	29.278
## Aggression	1	-41240.329	0.335	29.301
## Volleys	1	-41223.400	0.296	29.340
## BallControl	1	-41217.143	0.282	29.354
## StandingTackle	1	-41199.930	0.242	29.394
## Dribbling	1	-41183.807	0.205	29.431
## International.Reputation	1	-41181.413	0.199	29.437
## Finishing	1	-41176.604	0.188	29.448
## Skill.Moves	1	-41162.639	0.156	29.480
## Interceptions	1	-41162.264	0.155	29.481
## Penalties	1	-41155.952	0.140	29.496
## Jumping	1	-41141.920	0.108	29.528
## Stamina	1	-41131.792	0.084	29.552
## Age	1	-41130.456	0.081	29.555
## Value	1	-41119.294	0.055	29.581
## RCM	1	-41103.623	0.019	29.617
## ShotPower	1	-41098.092	0.006	29.630
## LongPassing	1	-41095.709	0.001	29.635
## SprintSpeed	1	-41095.623	0.000	29.636
## -----				
##				
## - HeadingAccuracy				
##				
##				
## Step 6 : AIC = -42242.82				
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special				
##				
## -----				
## Variable	DF	AIC	Sum Sq	RSS
## -----				
## Skill.Moves	1	-42648.268	0.852	26.232
## BallControl	1	-42556.721	0.663	26.421
## ShortPassing	1	-42505.132	0.556	26.529
## Balance	1	-42486.135	0.516	26.568
## Weight	1	-42467.123	0.477	26.608
## Positioning	1	-42447.563	0.436	26.649

## Height	1	-42421.384	0.381	26.703
## LongShots	1	-42353.408	0.238	26.846
## RCM	1	-42337.694	0.205	26.879
## FKAaccuracy	1	-42332.475	0.194	26.890
## Volleys	1	-42326.100	0.181	26.904
## Vision	1	-42324.858	0.178	26.906
## Penalties	1	-42316.509	0.160	26.924
## International.Reputation	1	-42312.465	0.152	26.933
## LCB	1	-42297.565	0.120	26.964
## SprintSpeed	1	-42293.686	0.112	26.972
## Curve	1	-42276.496	0.076	27.009
## Finishing	1	-42275.314	0.073	27.011
## LongPassing	1	-42259.542	0.040	27.045
## Value	1	-42257.853	0.036	27.048
## Jumping	1	-42254.592	0.029	27.055
## Agility	1	-42252.967	0.026	27.059
## Dribbling	1	-42249.574	0.019	27.066
## StandingTackle	1	-42249.765	0.019	27.065
## Stamina	1	-42247.017	0.013	27.071
## ShotPower	1	-42242.049	0.003	27.082
## Crossing	1	-42241.951	0.002	27.082
## Aggression	1	-42241.581	0.002	27.083
## Interceptions	1	-42241.654	0.002	27.083
## Age	1	-42240.924	0.000	27.084
## -----				
##				
## - Skill.Moves				
##				
##				
## Step 7 : AIC = -42648.27				
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special				
##				
## -----				
## Variable	DF	AIC	Sum Sq	RSS
## -----				
## Positioning	1	-43187.212	1.090	25.142
## LCB	1	-42983.772	0.686	25.547
## Balance	1	-42980.094	0.678	25.554
## Volleys	1	-42952.651	0.623	25.609
## LongShots	1	-42946.221	0.610	25.622
## Weight	1	-42943.519	0.605	25.627
## ShortPassing	1	-42907.647	0.533	25.700

## Finishing	1	-42873.199	0.463	25.769
## Penalties	1	-42865.968	0.448	25.784
## Height	1	-42858.501	0.433	25.799
## Vision	1	-42833.844	0.383	25.849
## FKAccuracy	1	-42815.534	0.346	25.886
## StandingTackle	1	-42808.155	0.331	25.901
## Curve	1	-42794.386	0.303	25.929
## BallControl	1	-42770.679	0.255	25.977
## Interceptions	1	-42761.589	0.236	25.996
## International.Reputation	1	-42722.025	0.155	26.077
## RCM	1	-42719.113	0.150	26.083
## Agility	1	-42716.288	0.144	26.089
## LongPassing	1	-42704.718	0.120	26.112
## Dribbling	1	-42699.297	0.109	26.123
## Aggression	1	-42685.062	0.080	26.153
## ShotPower	1	-42677.899	0.065	26.167
## Stamina	1	-42677.328	0.064	26.168
## SprintSpeed	1	-42660.773	0.030	26.202
## Value	1	-42660.299	0.029	26.203
## Crossing	1	-42657.844	0.024	26.208
## Age	1	-42652.966	0.014	26.218
## Jumping	1	-42646.979	0.001	26.231
## -----				
##				
## - Positioning				
##				
##				
## Step 8 : AIC = -43187.21				
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special				
##				
## -----				
## Variable	DF	AIC	Sum Sq	RSS
## -----				
## BallControl	1	-43564.880	0.738	24.404
## Balance	1	-43458.456	0.533	24.609
## Weight	1	-43433.634	0.485	24.657
## ShortPassing	1	-43387.113	0.395	24.747
## Height	1	-43357.530	0.338	24.804
## FKAccuracy	1	-43307.368	0.240	24.902
## RCM	1	-43280.822	0.188	24.954
## International.Reputation	1	-43265.775	0.158	24.984
## SprintSpeed	1	-43244.236	0.116	25.026

## LongShots	1	-43236.256	0.101	25.042
## Curve	1	-43232.910	0.094	25.048
## Volleys	1	-43229.098	0.086	25.056
## Vision	1	-43228.192	0.085	25.057
## Interceptions	1	-43221.012	0.071	25.072
## Penalties	1	-43213.932	0.057	25.085
## Dribbling	1	-43211.135	0.051	25.091
## StandingTackle	1	-43203.768	0.037	25.105
## Agility	1	-43202.121	0.033	25.109
## Jumping	1	-43201.263	0.032	25.110
## Value	1	-43196.311	0.022	25.120
## LCB	1	-43196.298	0.022	25.120
## Stamina	1	-43195.643	0.021	25.121
## Aggression	1	-43195.604	0.020	25.122
## Age	1	-43191.109	0.012	25.130
## ShotPower	1	-43188.730	0.007	25.135
## Crossing	1	-43187.202	0.004	25.138
## Finishing	1	-43187.328	0.004	25.138
## LongPassing	1	-43185.240	0.000	25.142
## -----				
##				
## - BallControl				
##				
##				
## Step 9 : AIC = -43564.88				
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special				
##				
## -----				
## Variable	DF	AIC	Sum Sq	RSS
## -----				
## Weight	1	-43863.173	0.568	23.836
## Balance	1	-43854.826	0.553	23.851
## Height	1	-43738.231	0.333	24.071
## FKAaccuracy	1	-43678.809	0.221	24.183
## International.Reputation	1	-43650.868	0.168	24.236
## LongShots	1	-43649.045	0.164	24.240
## Vision	1	-43647.263	0.161	24.243
## LCB	1	-43632.568	0.133	24.271
## SprintSpeed	1	-43631.496	0.131	24.273
## Volleys	1	-43628.799	0.126	24.278
## Curve	1	-43624.582	0.118	24.286
## ShortPassing	1	-43612.476	0.095	24.309

```

## Penalties          1    -43610.338    0.091    24.313
## Stamina            1    -43602.816    0.076    24.328
## RCM                1    -43593.564    0.059    24.345
## Agility            1    -43589.352    0.051    24.353
## Dribbling          1    -43587.804    0.048    24.356
## Age                1    -43586.928    0.046    24.358
## LongPassing        1    -43576.473    0.026    24.378
## Value              1    -43572.032    0.018    24.387
## Crossing           1    -43570.044    0.014    24.390
## Interceptions      1    -43567.740    0.009    24.395
## Finishing           1    -43565.649    0.005    24.399
## Aggression          1    -43563.569    0.001    24.403
## StandingTackle     1    -43563.256    0.001    24.403
## ShotPower          1    -43562.897    0.000    24.404
## Jumping            1    -43563.120    0.000    24.404
## -----
##
## - Weight
##
##
## Step 10 : AIC = -43863.17
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable          DF      AIC      Sum Sq      RSS
## -----
## FKAccuracy         1    -43992.277    0.244    23.592
## LongShots           1    -43989.432    0.239    23.597
## SprintSpeed         1    -43977.546    0.217    23.619
## Balance             1    -43960.777    0.186    23.650
## Vision              1    -43946.193    0.158    23.677
## Volleys             1    -43945.195    0.157    23.679
## Height              1    -43944.876    0.156    23.680
## International.Reputation 1    -43944.135    0.155    23.681
## Curve              1    -43925.225    0.120    23.716
## ShortPassing        1    -43915.029    0.101    23.735
## Penalties           1    -43911.265    0.094    23.742
## Stamina             1    -43897.999    0.069    23.767
## LCB                 1    -43897.386    0.068    23.768
## RCM                 1    -43881.313    0.038    23.798
## LongPassing         1    -43873.967    0.024    23.812
## Dribbling           1    -43871.642    0.020    23.816

```

```

## Age 1 -43870.445 0.017 23.818
## ShotPower 1 -43869.475 0.016 23.820
## Value 1 -43869.029 0.015 23.821
## Interceptions 1 -43868.449 0.014 23.822
## Finishing 1 -43867.064 0.011 23.825
## Aggression 1 -43865.358 0.008 23.828
## Crossing 1 -43862.933 0.003 23.832
## Jumping 1 -43862.874 0.003 23.833
## StandingTackle 1 -43862.143 0.002 23.834
## Agility 1 -43861.208 0.000 23.836
## -----
##
## - FKAccuracy
##
##
## Step 11 : AIC = -43992.28
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable DF AIC Sum Sq RSS
## -----
## Balance 1 -44120.985 0.241 23.351
## International.Reputation 1 -44087.989 0.180 23.412
## Height 1 -44074.339 0.155 23.437
## LongShots 1 -44056.587 0.122 23.469
## SprintSpeed 1 -44049.315 0.109 23.483
## Vision 1 -44048.464 0.107 23.484
## ShortPassing 1 -44047.468 0.106 23.486
## Volleys 1 -44028.948 0.071 23.520
## Interceptions 1 -44020.767 0.056 23.535
## RCM 1 -44017.538 0.050 23.541
## Age 1 -44012.557 0.041 23.551
## Aggression 1 -44008.754 0.034 23.558
## Dribbling 1 -44007.080 0.031 23.561
## StandingTackle 1 -44006.312 0.030 23.562
## Penalties 1 -44000.793 0.019 23.572
## LCB 1 -43999.629 0.017 23.575
## Curve 1 -43999.133 0.016 23.575
## Stamina 1 -43998.481 0.015 23.577
## Value 1 -43997.772 0.014 23.578
## LongPassing 1 -43997.828 0.014 23.578
## Agility 1 -43993.846 0.007 23.585

```

```

## Jumping          1      -43991.699      0.003      23.589
## Crossing         1      -43990.785      0.001      23.591
## Finishing        1      -43990.289      0.000      23.592
## ShotPower        1      -43990.334      0.000      23.592
## -----
##
## - Balance
##
##
## Step 12 : AIC = -44120.99
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable          DF          AIC          Sum Sq          RSS
## -----
## SprintSpeed        1      -44226.069          0.195      23.156
## International.Reputation  1      -44220.406          0.185      23.166
##
## Vision             1      -44198.625          0.145      23.206
## LongShots          1      -44195.240          0.139      23.212
## Interceptions      1      -44191.420          0.132      23.219
## Height             1      -44171.317          0.096      23.255
## StandingTackle     1      -44170.405          0.094      23.257
## Aggression         1      -44159.096          0.073      23.278
## ShortPassing       1      -44157.463          0.070      23.281
## Volleys            1      -44152.927          0.062      23.289
## LongPassing        1      -44145.896          0.049      23.302
## Age                1      -44139.291          0.037      23.314
## Dribbling          1      -44130.267          0.021      23.330
## Curve              1      -44129.425          0.019      23.332
## Value              1      -44126.086          0.013      23.338
## Stamina            1      -44126.245          0.013      23.338
## Penalties          1      -44125.452          0.012      23.339
## Agility            1      -44124.337          0.010      23.341
## RCM                1      -44123.154          0.008      23.343
## Crossing           1      -44120.889          0.003      23.348
## Jumping            1      -44120.617          0.003      23.348
## ShotPower          1      -44120.178          0.002      23.349
## LCB                1      -44119.748          0.001      23.350
## Finishing          1      -44119.193          0.000      23.351
## -----
##
## - SprintSpeed

```



```

##
##
## Step 13 : AIC = -44226.07
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable                DF          AIC          Sum Sq          RSS
## -----
## International.Reputation    1    -44343.403         0.216        22.940
## ShortPassing                1    -44304.236         0.145        23.010
## LongShots                   1    -44285.621         0.112        23.044
## Vision                      1    -44273.431         0.090        23.066
## Height                     1    -44272.849         0.088        23.067
## RCM                        1    -44271.872         0.087        23.069
## Age                       1    -44270.512         0.084        23.071
## Volleys                    1    -44259.001         0.063        23.092
## Dribbling                   1    -44252.993         0.052        23.103
##
## Interceptions              1    -44250.206         0.047        23.108
## Aggression                 1    -44241.277         0.031        23.124
## LCB                       1    -44240.204         0.029        23.126
## StandingTackle             1    -44238.081         0.025        23.130
## Curve                      1    -44232.850         0.016        23.140
## Value                     1    -44231.496         0.013        23.142
## Penalties                  1    -44230.710         0.012        23.144
## LongPassing                1    -44228.166         0.007        23.148
## Agility                    1    -44227.923         0.007        23.149
## Crossing                   1    -44227.305         0.006        23.150
## Stamina                    1    -44225.322         0.002        23.153
## ShotPower                  1    -44224.583         0.001        23.155
## Finishing                  1    -44224.244         0.000        23.155
## Jumping                    1    -44224.283         0.000        23.155
## -----
##
## - International.Reputation
##
##
## Step 14 : AIC = -44343.4
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable                DF          AIC          Sum Sq          RSS          R-Sq
## -----

```

```

## ShortPassing      1    -44423.280      0.147      22.793      0.868
## LongShots          1    -44401.268      0.108      22.832      0.868
## Vision             1    -44394.669      0.096      22.844      0.868
## RCM                1    -44393.853      0.094      22.846      0.867
## Height            1    -44389.909      0.087      22.853      0.867
## Volleys           1    -44386.990      0.082      22.858      0.867
## Age               1    -44374.893      0.060      22.880      0.867
## Dribbling         1    -44371.705      0.054      22.885      0.867
## Interceptions     1    -44365.865      0.044      22.896      0.867
## LCB               1    -44360.340      0.034      22.906      0.867
## Value             1    -44359.020      0.032      22.908      0.867
## Aggression        1    -44357.275      0.029      22.911      0.867
## Curve            1    -44353.772      0.022      22.918      0.867
## Penalties         1    -44353.071      0.021      22.919      0.867
## StandingTackle    1    -44353.088      0.021      22.919      0.867
## Stamina           1    -44346.863      0.010      22.930      0.867
## Crossing          1    -44346.063      0.008      22.931      0.867

## LongPassing       1    -44345.658      0.008      22.932      0.867
## Agility           1    -44344.860      0.006      22.934      0.867
## ShotPower         1    -44342.139      0.001      22.939      0.867
## Finishing         1    -44341.596      0.000      22.940      0.867
## Jumping           1    -44341.471      0.000      22.940      0.867
## -----
##
## - ShortPassing
##
##
## Step 15 : AIC = -44423.28
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable          DF          AIC          Sum Sq          RSS          R-Sq
## -----
## Vision            1    -44498.700      0.138      22.655      0.869
## LongPassing       1    -44489.924      0.122      22.671      0.869
## LongShots         1    -44470.011      0.087      22.706      0.868
## Age              1    -44467.258      0.082      22.711      0.868
## Height           1    -44467.469      0.082      22.710      0.868
## Volleys          1    -44456.406      0.063      22.730      0.868
## Dribbling        1    -44451.519      0.054      22.739      0.868
## Interceptions     1    -44445.452      0.043      22.750      0.868
## RCM              1    -44442.461      0.038      22.755      0.868

```

```

## LCB 1 -44441.367 0.036 22.757 0.868
## Value 1 -44439.552 0.033 22.760 0.868
## StandingTackle 1 -44434.587 0.024 22.769 0.868
## Aggression 1 -44431.039 0.017 22.776 0.868
## Stamina 1 -44430.385 0.016 22.777 0.868
## Curve 1 -44429.520 0.015 22.778 0.868
## Penalties 1 -44428.311 0.013 22.780 0.868
## Crossing 1 -44426.626 0.010 22.783 0.868
## Jumping 1 -44424.475 0.006 22.787 0.868
## Agility 1 -44422.703 0.003 22.790 0.868
## Finishing 1 -44421.372 0.000 22.793 0.868
## ShotPower 1 -44421.407 0.000 22.793 0.868
## -----
##
## - Vision
##
##

## Step 16 : AIC = -44498.7
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable DF AIC Sum Sq RSS R-Sq
## -----
## Interceptions 1 -44549.554 0.094 22.561 0.869
## LongPassing 1 -44544.983 0.086 22.569 0.869
## RCM 1 -44544.539 0.085 22.570 0.869
## Height 1 -44543.426 0.083 22.572 0.869
## Age 1 -44536.666 0.071 22.584 0.869
## StandingTackle 1 -44534.900 0.068 22.587 0.869
## LongShots 1 -44531.774 0.062 22.593 0.869
## Volleys 1 -44523.481 0.048 22.607 0.869
## Dribbling 1 -44519.345 0.040 22.615 0.869
## Aggression 1 -44516.614 0.035 22.620 0.869
## Value 1 -44513.715 0.030 22.625 0.869
## Crossing 1 -44505.657 0.016 22.639 0.869
## Curve 1 -44504.214 0.013 22.642 0.869
## Stamina 1 -44501.853 0.009 22.646 0.869
## LCB 1 -44501.097 0.008 22.647 0.869
## Penalties 1 -44500.825 0.007 22.648 0.869
## Finishing 1 -44500.192 0.006 22.649 0.869
## Jumping 1 -44497.850 0.002 22.653 0.869
## Agility 1 -44497.219 0.001 22.654 0.869

```

```

## ShotPower          1    -44496.710      0.000      22.655      0.869
## -----
##
## - Interceptions
##
##
## Step 17 : AIC = -44549.55
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable          DF          AIC          Sum Sq          RSS          R-Sq
## -----
## LCB                1    -44659.729      0.198      22.363      0.870
## LongShots          1    -44655.166      0.190      22.371      0.870
## Volleys            1    -44648.409      0.178      22.383      0.870
## RCM                1    -44602.619      0.097      22.464      0.870
## Height             1    -44593.306      0.081      22.480      0.870
##
## LongPassing        1    -44587.799      0.071      22.490      0.870
## Age                1    -44586.171      0.068      22.493      0.870
## Dribbling          1    -44585.867      0.068      22.493      0.870
## Penalties          1    -44579.384      0.056      22.505      0.869
## Curve              1    -44569.676      0.039      22.522      0.869
## Value              1    -44564.522      0.030      22.531      0.869
## ShotPower          1    -44560.236      0.022      22.539      0.869
## Aggression         1    -44558.853      0.020      22.541      0.869
## Crossing           1    -44555.259      0.014      22.548      0.869
## Finishing          1    -44554.194      0.012      22.549      0.869
## Stamina            1    -44553.742      0.011      22.550      0.869
## Agility            1    -44552.294      0.008      22.553      0.869
## StandingTackle     1    -44549.088      0.003      22.558      0.869
## Jumping            1    -44547.578      0.000      22.561      0.869
## -----
##
## - LCB
##
##
## Step 18 : AIC = -44659.73
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable          DF          AIC          Sum Sq          RSS          R-Sq
## -----

```

```

## StandingTackle      1    -44727.718      0.122      22.241      0.871
## LongShots            1    -44715.289      0.101      22.263      0.871
## Age                  1    -44710.178      0.092      22.272      0.871
## Volleys              1    -44706.872      0.086      22.277      0.871
## Height               1    -44703.113      0.079      22.284      0.871
## LongPassing          1    -44699.800      0.074      22.290      0.871
## Aggression           1    -44686.679      0.051      22.313      0.871
## Dribbling            1    -44682.728      0.044      22.320      0.871
## Value                1    -44673.894      0.028      22.335      0.870
## Curve                1    -44668.892      0.020      22.344      0.870
## Crossing             1    -44666.752      0.016      22.348      0.870
## Stamina              1    -44666.028      0.015      22.349      0.870
## Penalties            1    -44663.278      0.010      22.354      0.870
## RCM                  1    -44659.980      0.004      22.359      0.870
## Finishing            1    -44660.022      0.004      22.359      0.870
## ShotPower            1    -44659.748      0.004      22.360      0.870
## Agility              1    -44657.791      0.000      22.363      0.870

## Jumping              1    -44657.744      0.000      22.363      0.870
## -----
##
## - StandingTackle
##
##
## Step 19 : AIC = -44727.72
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      A
## -----
## LongShots      1    -44802.620      0.134      22.107      0.872
## Volleys         1    -44788.704      0.110      22.131      0.872
## Height          1    -44771.923      0.080      22.160      0.871
## Age             1    -44769.656      0.077      22.164      0.871
## LongPassing     1    -44764.862      0.068      22.173      0.871
## Aggression      1    -44761.079      0.062      22.179      0.871
## Dribbling       1    -44747.469      0.038      22.203      0.871
## Value           1    -44741.909      0.028      22.213      0.871
## Curve           1    -44739.158      0.023      22.218      0.871
## Penalties       1    -44733.598      0.014      22.227      0.871
## RCM             1    -44732.909      0.013      22.228      0.871
## Stamina         1    -44733.344      0.013      22.228      0.871
## Crossing        1    -44732.009      0.011      22.230      0.871

```

```

## ShotPower      1    -44731.688    0.010    22.231    0.871
## Jumping        1    -44727.828    0.004    22.237    0.871
## Agility        1    -44726.073    0.001    22.240    0.871
## Finishing      1    -44725.843    0.000    22.241    0.871
## -----
##
## - LongShots
##
##
## Step 20 : AIC = -44802.62
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      A
## -----
## Volleys       1    -44849.504    0.085    22.023    0.872
## Aggression    1    -44849.583    0.085    22.022    0.872
##
## Height        1    -44847.520    0.081    22.026    0.872
## LongPassing   1    -44844.227    0.076    22.032    0.872
## Age           1    -44838.735    0.066    22.041    0.872
## Dribbling     1    -44826.018    0.044    22.063    0.872
## Curve         1    -44820.013    0.034    22.074    0.872
## Crossing      1    -44818.841    0.032    22.076    0.872
## Value         1    -44817.981    0.030    22.077    0.872
## Jumping       1    -44809.411    0.015    22.092    0.872
## Finishing     1    -44808.687    0.014    22.093    0.872
## Penalties     1    -44808.693    0.014    22.093    0.872
## Agility       1    -44805.624    0.009    22.098    0.872
## Stamina       1    -44803.550    0.005    22.102    0.872
## RCM           1    -44800.751    0.000    22.107    0.872
## ShotPower     1    -44800.865    0.000    22.107    0.872
## -----
##
##
## - Aggression
##
##
## Step 21 : AIC = -44849.58
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      A
## -----

```

```

## Volleys      1    -44903.459      0.096      21.926      0.873
## LongPassing  1    -44898.006      0.087      21.935      0.873
## Height       1    -44896.735      0.085      21.938      0.873
## Age          1    -44893.195      0.079      21.944      0.873
## Dribbling    1    -44880.403      0.057      21.966      0.873
## Crossing     1    -44878.449      0.053      21.969      0.873
## Curve        1    -44877.291      0.051      21.971      0.873
## Value        1    -44864.372      0.029      21.993      0.872
## Penalties    1    -44859.925      0.021      22.001      0.872
## Jumping      1    -44856.526      0.015      22.007      0.872
## RCM          1    -44855.325      0.013      22.009      0.872
## Agility      1    -44854.864      0.013      22.010      0.872
## Finishing     1    -44852.443      0.008      22.014      0.872
## Stamina      1    -44850.804      0.006      22.017      0.872
## ShotPower    1    -44848.160      0.001      22.021      0.872
## -----
##
## - Volleys
##
##
## Step 22 : AIC = -44903.46
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      A
## -----
## LongPassing    1    -44965.191      0.109      21.817      0.873
## Height         1    -44953.136      0.089      21.837      0.873
## Age            1    -44946.334      0.077      21.849      0.873
## Crossing       1    -44942.551      0.071      21.855      0.873
## Dribbling      1    -44939.851      0.066      21.860      0.873
## Curve          1    -44927.841      0.045      21.881      0.873
## Value          1    -44918.323      0.029      21.897      0.873
## Jumping        1    -44918.142      0.029      21.897      0.873
## Agility        1    -44915.773      0.025      21.901      0.873
## Finishing      1    -44913.747      0.021      21.905      0.873
## Penalties      1    -44909.691      0.014      21.912      0.873
## RCM            1    -44907.218      0.010      21.916      0.873
## ShotPower      1    -44902.972      0.003      21.923      0.873
## Stamina        1    -44901.707      0.000      21.926      0.873
## -----
##

```

```
## - LongPassing
##
##
## Step 23 : AIC = -44965.19
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Height         1    -45016.470      0.091     21.726     0.874
## Age            1    -45008.211      0.077     21.740     0.874
## Dribbling      1    -45003.193      0.068     21.748     0.874
## Crossing       1    -45000.727      0.064     21.752     0.874
## Curve          1    -44991.875      0.049     21.768     0.874
## Jumping        1    -44990.250      0.046     21.770     0.874
## Agility        1    -44981.590      0.031     21.785     0.874
## Value          1    -44979.672      0.028     21.788     0.874
##
## Penalties      1    -44977.782      0.025     21.792     0.874
## Finishing      1    -44971.650      0.014     21.802     0.874
## ShotPower      1    -44964.500      0.002     21.814     0.873
## RCM            1    -44963.860      0.001     21.816     0.873
## Stamina        1    -44963.360      0.000     21.816     0.873
## -----
##
## - Height
##
##
## Step 24 : AIC = -45016.47
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Age            1    -45063.550      0.084     21.642     0.874
## Dribbling      1    -45053.150      0.066     21.660     0.874
## Crossing       1    -45046.943      0.055     21.670     0.874
## Curve          1    -45042.620      0.048     21.678     0.874
## Jumping        1    -45040.766      0.045     21.681     0.874
## Agility        1    -45032.215      0.030     21.695     0.874
## Value          1    -45030.253      0.027     21.699     0.874
## Penalties      1    -45028.806      0.024     21.701     0.874
## Finishing      1    -45022.091      0.013     21.713     0.874
```



```

## ShotPower      1      -45015.377      0.002      21.724      0.874
## RCM             1      -45015.123      0.001      21.725      0.874
## Stamina        1      -45014.778      0.001      21.725      0.874
## -----
##
## - Age
##
##
## Step 25 : AIC = -45063.55
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Crossing       1      -45105.014      0.074      21.568      0.875
## Dribbling      1      -45091.798      0.051      21.591      0.875
## Jumping        1      -45089.544      0.047      21.595      0.875
##
## Curve          1      -45088.553      0.046      21.596      0.875
## Agility        1      -45081.106      0.033      21.609      0.875
## Penalties      1      -45077.031      0.026      21.616      0.875
## Value          1      -45075.296      0.023      21.619      0.875
## Finishing      1      -45070.795      0.016      21.626      0.875
## RCM            1      -45063.550      0.003      21.639      0.874
## ShotPower      1      -45063.345      0.003      21.639      0.874
## Stamina        1      -45061.749      0.000      21.642      0.874
## -----
##
##
## - Crossing
##
##
## Step 26 : AIC = -45105.01
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Jumping        1      -45151.238      0.081      21.487      0.875
## Agility        1      -45127.826      0.042      21.526      0.875
## Penalties      1      -45127.630      0.042      21.527      0.875
## Dribbling      1      -45122.389      0.033      21.536      0.875
## Curve          1      -45119.174      0.027      21.541      0.875
## Value          1      -45117.744      0.025      21.544      0.875

```

```

## RCM          1      -45108.354      0.009      21.559      0.875
## Finishing    1      -45106.815      0.006      21.562      0.875
## Stamina      1      -45104.731      0.003      21.566      0.875
## ShotPower    1      -45103.293      0.000      21.568      0.875
## -----
##
## - Jumping
##
##
## Step 27 : AIC = -45151.24
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF          AIC          Sum Sq          RSS          R-Sq          Adj
## -----
## Penalties      1      -45188.120          0.065          21.422          0.876
## Dribbling       1      -45177.841          0.048          21.439          0.876
##
## Curve          1      -45174.184          0.042          21.445          0.876
## Agility         1      -45174.465          0.042          21.444          0.876
## Value           1      -45163.195          0.024          21.463          0.876
## Stamina         1      -45152.341          0.005          21.482          0.875
## RCM             1      -45151.141          0.003          21.484          0.875
## Finishing       1      -45150.333          0.002          21.485          0.875
## ShotPower       1      -45149.359          0.000          21.487          0.875
## -----
##
## - Penalties
##
##
## Step 28 : AIC = -45188.12
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF          AIC          Sum Sq          RSS          R-Sq          Adj
## -----
## Agility         1      -45226.361          0.068          21.354          0.876
## Dribbling       1      -45219.629          0.056          21.365          0.876
## Curve           1      -45216.593          0.051          21.370          0.876
## Value           1      -45200.597          0.024          21.397          0.876
## Stamina         1      -45194.672          0.014          21.407          0.876
## Finishing       1      -45188.502          0.004          21.418          0.876
## RCM             1      -45188.100          0.003          21.418          0.876

```

```

## ShotPower      1      -45186.320      0.000      21.421      0.876
## -----
##
## - Agility
##
##
## Step 29 : AIC = -45226.36
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Curve          1      -45259.807      0.059      21.295      0.876
## Dribbling       1      -45259.154      0.058      21.296      0.876
## Value          1      -45239.165      0.025      21.329      0.876
## Stamina        1      -45237.219      0.022      21.332      0.876
## RCM            1      -45227.931      0.006      21.348      0.876
##
## ShotPower      1      -45226.790      0.004      21.350      0.876
## Finishing       1      -45225.071      0.001      21.353      0.876
## -----
##
## - Curve
##
##
## Step 30 : AIC = -45259.81
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Dribbling       1      -45293.651      0.060      21.235      0.877
## Stamina          1      -45280.120      0.037      21.257      0.877
## Value           1      -45273.354      0.026      21.269      0.877
## RCM             1      -45263.301      0.009      21.285      0.877
## ShotPower       1      -45261.076      0.005      21.289      0.877
## Finishing       1      -45257.811      0.000      21.295      0.876
## -----
##
## - Dribbling
##
##
## Step 31 : AIC = -45293.65

```

```
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Stamina       1    -45324.234      0.054     21.181     0.877
## Value         1    -45307.394      0.026     21.209     0.877
## RCM           1    -45306.916      0.025     21.209     0.877
## ShotPower     1    -45298.039      0.011     21.224     0.877
## Finishing     1    -45291.652      0.000     21.235     0.877
## -----
##
## - Stamina
##
##
## Step 32 : AIC = -45324.23
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## RCM           1    -45354.920      0.054     21.126     0.877
## Value         1    -45336.779      0.024     21.156     0.877
## ShotPower     1    -45335.396      0.022     21.159     0.877
## Finishing     1    -45322.815      0.001     21.180     0.877
## -----
##
## - RCM
##
##
## Step 33 : AIC = -45354.92
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## ShotPower     1    -45370.333      0.029     21.098     0.878
## Value         1    -45367.481      0.024     21.102     0.878
## Finishing     1    -45357.465      0.008     21.119     0.878
## -----
##
## - ShotPower
```

```
##
##
## Step 34 : AIC = -45370.33
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Value          1    -45383.563      0.025     21.072     0.878
## Finishing       1    -45375.096      0.011     21.086     0.878
## -----
##
## - Value
##
##
## Step 35 : AIC = -45383.56
## Overall ~ Reactions + Wage + Composure + GKPositioning + Special
##
## -----
## Variable      DF      AIC      Sum Sq      RSS      R-Sq      Adj
## -----
## Finishing       1    -45388.089      0.011     21.062     0.878
## -----
##
## - Finishing
##
##
## Variables Entered:
##
## - Reactions
## - Wage
## - Composure
## - GKPositioning
## - Special
## - HeadingAccuracy
## - Skill.Moves
## - Positioning
## - BallControl
## - Weight
## - FKAccuracy
## - Balance
## - SprintSpeed
```

```
## - International.Reputation
## - ShortPassing
## - Vision
## - Interceptions
## - LCB
## - StandingTackle
## - LongShots
## - Aggression
## - Volleys
## - LongPassing
## - Height
## - Age
## - Crossing
## - Jumping
## - Penalties
## - Agility
## - Curve
```

```
## - Dribbling
## - Stamina
## - RCM
## - ShotPower
## - Value
## - Finishing
```

```
##
##
```

```
## Final Model Output
```

```
## -----
```

```
##
```

```
## Model Summary
```

```
## -----
```

## R	0.937	RMSE	0.041
## R-Squared	0.878	Coef. Var	-7.745
## Adj. R-Squared	0.877	MSE	0.002
## Pred R-Squared	0.877	MAE	0.032

```
## -----
```

```
## RMSE: Root Mean Square Error
```

```
## MSE: Mean Square Error
```

```
## MAE: Mean Absolute Error
```

```
##
```

```
## ANOVA
```

```
## -----
```

```
## Sum of
```

##	Squares	DF	Mean Square	F
##	-----			
## Regression	151.355	36	4.204	2536.57
## Residual	21.062	12707	0.002	
## Total	172.416	12743		
##	-----			
##				
##	Parameter Estimates			
##	-----			
##	model	Beta	Std. Error	Std. Beta
##	-----			
##	(Intercept)	-0.910	0.012	
##	Reactions	0.281	0.006	0.287
##	Wage	0.406	0.014	0.139
##	Composure	0.142	0.005	0.149
##	GKPositioning	0.129	0.009	0.212
##	Special	0.966	0.039	1.391
##	HeadingAccuracy	0.068	0.005	0.112
##	Skill.Moves	0.094	0.004	0.152
##	Positioning	-0.135	0.006	-0.243
##	BallControl	0.105	0.009	0.165
##	Weight	0.019	0.005	0.020
##	FKAccuracy	-0.053	0.005	-0.087
##	Balance	-0.074	0.004	-0.112
##	SprintSpeed	-0.017	0.005	-0.025
##	International.Reputation	0.063	0.005	0.054
##	ShortPassing	0.054	0.008	0.079
##	Vision	-0.081	0.006	-0.117
##	Interceptions	-0.066	0.006	-0.133
##	LCB	-0.051	0.017	-0.077
##	StandingTackle	-0.067	0.008	-0.137
##	LongShots	-0.087	0.006	-0.157
##	Aggression	-0.047	0.004	-0.082
##	Volleys	-0.066	0.005	-0.115
##	LongPassing	-0.088	0.007	-0.137
##	Height	0.012	0.002	0.025
##	Age	0.023	0.003	0.032
##	Crossing	-0.050	0.005	-0.089
##	Jumping	-0.039	0.004	-0.050
##	Penalties	-0.051	0.005	-0.079
##	Agility	-0.045	0.005	-0.070
##	Curve	-0.043	0.005	-0.076

##	Dribbling	-0.072	0.008	-0.124
##	Stamina	-0.036	0.004	-0.058
##	RCM	0.089	0.014	0.109
##	ShotPower	-0.025	0.005	-0.039
##	Value	-0.041	0.011	-0.015
##	Finishing	-0.016	0.006	-0.028
##	-----			

Selection Summary				
## Variable	AIC	Sum Sq	RSS	R
## Reactions	-34899.647	124.188	48.229	0.
## Wage	-36992.285	131.498	40.919	0.
## Composure	-38689.025	136.604	35.812	0.
## GKPositioning	-39708.277	139.362	33.054	0.
## Special	-41097.419	142.780	29.636	0.
## HeadingAccuracy	-42242.820	145.332	27.084	0.
## Skill.Moves	-42648.268	146.184	26.232	0.
## Positioning	-43187.212	147.274	25.142	0.
## BallControl	-43564.880	148.012	24.404	0.
## Weight	-43863.173	148.581	23.836	0.
## FKAccuracy	-43992.277	148.825	23.592	0.
## Balance	-44120.985	149.065	23.351	0.
## SprintSpeed	-44226.069	149.261	23.156	0.
## International.Reputation	-44343.403	149.477	22.940	0.
## ShortPassing	-44423.280	149.624	22.793	0.
## Vision	-44498.700	149.762	22.655	0.
## Interceptions	-44549.554	149.855	22.561	0.
## LCB	-44659.729	150.053	22.363	0.
## StandingTackle	-44727.718	150.176	22.241	0.
## LongShots	-44802.620	150.309	22.107	0.
## Aggression	-44849.583	150.394	22.022	0.
## Volleys	-44903.459	150.490	21.926	0.
## LongPassing	-44965.191	150.600	21.817	0.
## Height	-45016.470	150.691	21.726	0.
## Age	-45063.550	150.774	21.642	0.
## Crossing	-45105.014	150.848	21.568	0.
## Jumping	-45151.238	150.930	21.487	0.



## Penalties	-45188.120	150.995	21.422	0.
## Agility	-45226.361	151.063	21.354	0.
## Curve	-45259.807	151.122	21.295	0.
## Dribbling	-45293.651	151.182	21.235	0.
## Stamina	-45324.234	151.236	21.181	0.
## RCM	-45354.920	151.290	21.126	0.
## ShotPower	-45370.333	151.319	21.098	0.
## Value	-45383.563	151.344	21.072	0.
## Finishing	-45388.089	151.355	21.062	0.
## -----				

## 2.5.12 Model Output

### Model Summary

- R 0.939
- RMSE 0.042
- R-Squared 0.881
- Coef. Var -7.842
- Adj. R-Squared 0.880
- MSE 0.002
- Pred R-Squared 0.880
- MAE 0.033

### 다중 회귀분석 최종모델 결과 해석

다중회귀분석으로 설계하고 학습 및 개선시킨 모델로 사전에 마련해둔 Test data를 예측해본 결과

Adjusted R-squared 값이 개선 후 모델에서 최종적으로 사용된 44개의 변수가 Overall을 88% 설명할 수 있음을 확인해, 설명력이 높은 모델을 만들어냈음을

알 수 있었다.

,br>

## 2.6 분류분석

UEFA의 club순위를 참고하여 임의의 파생변수로 생성한 범주형 구간 변수인 Club\_grade를 label로 활용하여 선수들이 가진 stat과 feature에 따라 어떤 등급의 팀에 속해야 적절한지 예측하는 분류모델을 만들어 학습시키고, 이를 기반으로 30%의 test 데이터를 예측해보고자 한다.

### 너무긴학습시간문제

rpart에 너무 많은 변수와 관측값을 가진 데이터를 학습시켜 불순도 연산에 의해 너무 긴 시간이 걸리는것으로 짐작된다, 첫 학습이 진행되지 않아 스케줄 상 더이상의 연구 진행이 어려울것이라 판단하여 해당 파트의 코드는 주석처리하고 추후 연구에서 개선해보고자 한다

```
#Decision Tree

#
# library(rpart)
#
# # 데이터 샘플링
#
# library(doBy)
# train = sampleBy(~club_grade,frac=0.7,data=b) #70%의 랜덤(train)데
#
#
#
```

```

#
#
#
# # 70%에 포함된 train데이터의 행 번호를 추출하기 위한 사용자 함수 생성
#
# f1 <- function(x) {
#
#   as.numeric(strsplit(x, '\\.')[[1]][2])                                #
#
# }
#
# rn <- as.vector(sapply(rownames(train), f1))                            # 70%에 포
#
# test <- b[-rn,]                                                         #
#
#
#
#
#
#
#
#
#
#
# train[10,]
# #####2. 모델 생성
#
# m <- rpart(club_grade~.,data = train)
#
# # 모델 학습에 필요한 데이터 셋(70%의 train 데이터)
#
#
#                                     # 각 컬럼마다의 불순도 연산이 수행되므로 데이터가
#
#
#
#
#
#
#
#
#
#
#
#

```

```
# ###3. 모델평가
#
# # 3-1) 새로운 데이터셋(Test)에 대한 예측력 확인
#
# val_var <- predict()
#
# sum() / nrow() * 100
#
#
#
#
#
# val_var <- predict(m, newdata = train, type = 'class')
#
# sum(val_var == ) / nrow(train) * 100
#
#
#
#
#
# ##4. 모델을 통한 예측
#
```

### 3. 결론

## 의의

해당 보고서에선 손흥민 선수에 의해 점화된 해외축구에 대한 전국적인 관심에 힘입어, 데이터가 생소한 보통의 축구팬들에게도 다소 접근하기 쉬운 기초통계량 기반의 EDA분석과 시각화로 시작하여 유사도분석, 분류분석, 군집분석, 다중회귀분석까지 다소 심화된 개념의 분석 및 모델링과 결과들을 통해 저번 중간보

고서의 연구에 이어 데이터를 통해 스포츠를 더욱 깊고 폭넓게 즐기고 유용하게 분석해 볼 수 있다는 시사점을 다시 한번 강조할 수 있었다.

## 결과정리

### 2.2 EDA 분석결과

- 각 리그 선수들의 평균 연령은 25세로 형성돼있다.
- 프리미어리그 내 등번호 7번의 잉어인 손흥민과 라힘 스털링 선수를 비교해 본 결과 두 선수는 스탯의 유사한 분포를 가지고 있다.
- 토트넘에서 가장 높은 potential과 Overall을 가진 선수는 Harry Kane 선수이다.
- 나이에 따른 소속 리그별 평균 잠재력 및 통합스탯을 분석해 본 결과 약 25세부터 선수들의 overall과 potential이 일치하여 동일선을 그리기 시작한다.

### 2.3 선수 유사도 분석결과

\*\*\*

유클라디안 거리계산법을 활용하여 손흥민, Harry Kane 선수와 유사한 선수들을 알아본 결과

- 손흥민 선수는 뎀벨레, 므키타리안, 래시포드, 마네, 라카제트 등 상대 수비진을 침투하는 플레이방식의 선수들과 유사하다는 결과를 얻을 수 있었다.

\* 반면 해리케인 선수는 레반도프스키, 루카쿠, 피르미누, 포그바 등 득점능력 뿐

아니라 패스 등 전체적으로 높은 밸런스를 가진 선수들과 유사하다고 판단되어 완성형 공격수라는 별명에 어울리는 결과를 얻을 수 있었다.

## 2.5. 다중회귀 분석 결과

FIFA선수의 인구통계학적 Feature와 경기중에 관측되는 각종 선수stat이 Overall을 얼마나 설명할 수 있을 지 다중회귀분석으로 규명해보았다.

### 최종 모델 결과

- \* R 0.939
- \* RMSE 0.042
- \* R-Squared 0.881
- \* Coef. Var -7.842
- \* Adj. R-Squared 0.880
- \* MSE 0.002
- \* Pred R-Squared 0.880
- \* MAE 0.033

**Adjusted R-squared** 값이 개선 후 모델에서 최종적으로 사용된 **44개의 변수**가 **Overall**을 **88%** 설명할 수 있음을 확인해, 설명력이 높은 모델을 만들어냈음을 알 수 있었다.

### 한계점

다중회귀분석의 종속변수로 사용한 Overall은 본래 그 의미가 선수의 경기내 stat을 전체적으로 반영한 점수이므로 당연히 높은 설명력을 가질 수 밖에 없었다.

또한 너무 큰 dataset을 사용하여 분류모델을 학습시키려 할 경우 너무 많은 불순도연산으로 매우 긴 연산기간을 야기하는 것을 고려하지 못하여 원활히 연구를 진행할 수 없었다. 해당 부분의 문제는 추후 머신러닝 및 캡스톤 디자인 과목에서 보완법을 알아보고 추가 연구를 진행해보고자 한다.