

# CS5801 Coursework Template Proforma

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## 0. Instructions

1. Remove the (italicised) guidance text but keep the section headings.
2. Add as many chunks of R code as required.
3. Add descriptions of your analysis plans and explanations of your code and findings. Please be detailed and where you have made choices explain the rationale for them.
4. Write your report using RMarkdown. For guidance see a helpful blog (<https://www.dataquest.io/blog/r-markdown-guide-cheatsheet/#tve-jump-17333da0719>) or use the R Markdown cheatsheet which can be accessed from within RStudio by selecting `HeLp > Cheatsheets > R Markdown Cheat Sheet`.
5. Your report should be clearly and professionally presented with appropriate use of cited external sources. (5 marks)
6. It should also be easy to understand, with well-documented code following the principles of literate programming. (5 marks)

```
# Add code here to load any required libraries with `library()`.
# We suggest you use `install.packages()` for any required packages externally to this document
# since installation only need be done once.
library(corrplot)
```

```
## corrplot 0.91 loaded
```

```
library(ggpubr)
```

```
## Loading required package: ggplot2
```

```
options(scipen = 999)
```

## 1. Organise and clean the data

### 1.1 Subset the data into the specific dataset allocated

A description of the data set provided, its contents and which subset you should select is documented in the assessment brief at ????.pdf  
Use R code to correctly select the subset of data allocated. (5 marks)

```
# Assign your student id into the variable SID, for example:
SID <- 2144810 # This is an example, replace 2101234 with your actual ID
SIDoffset <- (SID %% 25) + 1 # Your SID mod 25 + 1

load("CS5801_football_analysis.Rda")
```

```
## Warning in load("CS5801_football_analysis.Rda"): strings not representable in
## native encoding will be translated to UTF-8
```

```
# Now subset the football data set
# Pick every 25th observation starting from your offset
# Put into your data frame named mydf (you can rename it)
mydf <- football.analysis[seq(from=SIDoffset,to=nrow(football.analysis),by=25),]
write.csv(mydf, "data.csv")
```

### 1.2 Data quality analysis

Provide a description of a comprehensive plan to assess the quality of the data, and document your findings. Include all variables/columns (5 marks) from the data set and provide a full implementation (5 marks). NB even if no data quality issues are identified you should still check and report.

Here is my following plan to check the quality of the data for each column:

1. `sofifa_id` : First, it must be checked if the column is a string. Then it must be checked if each value is unique since it is an ID. It must be checked if missing values (NA) exist.
2. For columns `wage_eur`, `age`, `height_cm` and `weight_kg`, it must be checked that the columns are numerical. Then it must be checked that values are not below 0. It also must be checked for outliers and if they are reasonable or not. It must be checked if missing values (NA) exist.
3. For columns `potential`, `pace`, `shooting`, `passing`, `dribbling`, `defending`, `physic`, `power_strength` and `power_long_shots`, it must be checked that the column is numerical. Then it must be checked that the maximum value is not above 100 and that the minimum value is not below 0. It must be checked if missing values (NA) exist.
4. `club_name` : First, it must be checked that the column is a string. This column won't be used for analysis so not further studying will be done.
5. `preferred_foot` : First, it must be checked that the column is categorical (factor). This column must be constraint to two unique values: `Right` and `Left`. It exist the possibility that values like `left` or `Left` could be encountered and if they do, they need to be changed to the 2 unique values. It must be checked if missing values (NA) exist.

6. `high.wage.ind` : First, it must be checked that the column is categorical (factor). This column must be constraint to two unique values: `1` and `0`. It must be checked if missing values (NA) exist. It also must be checked that it has a value of `1` if weekly wage is above 8000 and `0` otherwise.

**Note:** a) NA refers to Not Available. b) There are many methods for calculating outliers. However, for this lab, 3 will be used. The first is by using a boxplot. Where points are visualized, outliers exist. The second method is by using the standard deviation (std). Values that are beyond 3 stds from the mean are considered outliers. The final method is using interquartile range (IQR). Values that are beyond quartile 3 plus 1.5 times the IQR and values that are below quartile 1 minus 1.5 times the IQR are considered outliers.

## Checking for quality issues:

- a. Checking data type of each column:

```
str(mydf)
```

```
## 'data.frame': 514 obs. of 17 variables:
## $ sofifa_id : int 155862 218667 189513 146536 212194 186345 208574 176635 210665 179846 ...
## $ potential : int 89 88 85 84 88 83 83 82 82 81 ...
## $ wage_eur : num 300000 230000 59000 43000 75000 71000 45000 100000 65000 87000 ...
## $ age : int 34 25 31 34 24 29 27 31 28 33 ...
## $ height_cm : int 184 173 182 172 185 173 184 180 188 189 ...
## $ weight_kg : int 82 64 74 60 83 71 82 71 87 90 ...
## $ club_name : chr "Real Madrid CF" "Manchester City" "Villarreal CF" "Sevilla FC" ...
## $ preferred_foot : chr "Right" "Left" "Right" "Right" ...
## $ pace : int 71 81 50 87 81 73 88 64 66 45 ...
## $ shooting : int 70 77 83 69 78 60 76 69 66 74 ...
## $ passing : int 76 83 90 79 82 81 79 85 73 77 ...
## $ dribbling : int 73 92 81 83 87 77 83 83 72 77 ...
## $ defending : int 88 51 71 80 51 79 68 24 81 79 ...
## $ physic : int 85 67 69 58 66 72 80 55 80 78 ...
## $ power_strength : int 85 54 70 39 71 64 77 56 79 85 ...
## $ power_long_shots: int 64 76 88 70 80 70 78 67 69 78 ...
## $ high.wage.ind : int 1 1 1 1 1 1 1 1 1 1 ...
```

- `power_long_shots`, `power_strength`, `physic`, `defending`, `dribbling`, `passing`, `shooting`, `pace` and `potential` have the right data type. They are integers (numerical).
- `high.wage.ind` is integer (numerical) and therefore it needs to be changed to categorical (factor).
- `preferred_foot` is a string (character) and needs to be changed to categorical (factor).
- `sofifa_id` is a integer (numerical) and needs to be changed to string (character).
- `wage_eur` is numerical therefore it is correct.
- `age`, `height_cm`, `weight_kg` are integer (numerical) and therefore they are correct.
- `club_name` is a string (character) and therefore it is correct.

- b. Checking for missing values (NA) for each column:

```
# df_col_names is a list of all the column names of our dataframe.
df_col_names <- c(colnames(mydf))
# iteration over each column to find missing values.
for (col_name in df_col_names){
  message = paste("Column", col_name, "has", sum(is.na(mydf$col_name)), "missing values (NA)")
  print(message)
}
```

```
## [1] "Column sofifa_id has 0 missing values (NA)"
## [1] "Column potential has 0 missing values (NA)"
## [1] "Column wage_eur has 0 missing values (NA)"
## [1] "Column age has 0 missing values (NA)"
## [1] "Column height_cm has 0 missing values (NA)"
## [1] "Column weight_kg has 0 missing values (NA)"
## [1] "Column club_name has 0 missing values (NA)"
## [1] "Column preferred_foot has 0 missing values (NA)"
## [1] "Column pace has 0 missing values (NA)"
## [1] "Column shooting has 0 missing values (NA)"
## [1] "Column passing has 0 missing values (NA)"
## [1] "Column dribbling has 0 missing values (NA)"
## [1] "Column defending has 0 missing values (NA)"
## [1] "Column physic has 0 missing values (NA)"
## [1] "Column power_strength has 0 missing values (NA)"
## [1] "Column power_long_shots has 0 missing values (NA)"
## [1] "Column high.wage.ind has 0 missing values (NA)"
```

No missing values were found in any column.

- c. Check maximum and minimum values for numerical columns:

```
summary(mydf)
```

```
##      sofifa_id      potential      wage_eur      age
## Min.   :143809   Min.   :54.00   Min.    :    4   Min.   :16.00
## 1st Qu.:211561   1st Qu.:67.00   1st Qu.: 1000   1st Qu.:22.00
## Median :232602   Median :72.00   Median : 3000   Median :25.00
## Mean   :227354   Mean   :71.67   Mean    :10821   Mean   :25.22
## 3rd Qu.:247061   3rd Qu.:76.00   3rd Qu.:10000   3rd Qu.:28.00
## Max.   :258923   Max.   :89.00   Max.   :300000   Max.   :77.00
##      height_cm      weight_kg      club_name      preferred_foot
## Min.   :160.0   Min.    : 58.00   Length:514   Length:514
## 1st Qu.:175.0   1st Qu.: 70.00   Class :character   Class :character
## Median :180.0   Median : 75.00   Mode  :character   Mode  :character
## Mean   :180.1   Mean    : 74.54
## 3rd Qu.:185.0   3rd Qu.: 79.00
## Max.   :220.0   Max.   :188.00
##      pace      shooting      passing      dribbling
## Min.   :-66.00   Min.   :21.00   Min.   :26.00   Min.   :-66.0
## 1st Qu.: 62.00   1st Qu.:42.00   1st Qu.:51.00   1st Qu.: 57.0
## Median : 69.00   Median :55.00   Median :58.00   Median : 64.0
## Mean   : 68.13   Mean   :53.04   Mean   :57.81   Mean   : 62.8
## 3rd Qu.: 76.00   3rd Qu.:65.00   3rd Qu.:65.00   3rd Qu.: 70.0
## Max.   : 94.00   Max.   :83.00   Max.   :90.00   Max.   : 92.0
##      defending      physic      power_strength      power_long_shots
## Min.   :18.00   Min.   :35.00   Min.   :29.00   Min.   :16.00
## 1st Qu.:36.00   1st Qu.:58.00   1st Qu.:57.00   1st Qu.:40.00
## Median :57.00   Median :66.00   Median :67.00   Median :54.00
## Mean   :51.53   Mean   :64.61   Mean   :65.31   Mean   :51.83
## 3rd Qu.:65.00   3rd Qu.:72.00   3rd Qu.:75.00   3rd Qu.:64.00
## Max.   :88.00   Max.   :85.00   Max.   :93.00   Max.   :88.00
##      high.wage.ind
## Min.   :0.0000
## 1st Qu.:0.0000
## Median :0.0000
## Mean   :0.2899
## 3rd Qu.:1.0000
## Max.   :1.0000
```

- Column `wage_eur` has minimum value of 4 and maximum value of 300000. All values are positive. 4 euros for weekly salary is very low. Low values in this column need further investigation. Also, I need to look for outliers.
- Column `age` has a minimum value of 16 and maximum value of 77. All values are positive. Now, I need to look for outliers.
- Column `height_cm` has a minimum value of 160 and a max value of 220. All values are positive. Now, I need to look for outliers.
- Column `weight_kg` has a minimum value of 56 and a maximum value of 188. All values are positive. Now, I need to look for outliers.
- Column `potential` has a minimum value of 54 and has a maximum value of 89 . Values are between 0 and 100 so no changes are needed.
- Column `pace` has a minimum value of -66 and has a maximum value of 94 . Values must be between 0 and 100. Therefore, changes are needed
- Column `shooting` has a minimum value of 21 and has a maximum value of 83 . Values are between 0 and 100 so no changes are needed.
- Column `passing` has a minimum value of 26 and has a maximum value of 90 . Values are between 0 and 100 so no changes are needed.
- Column `dribbling` has a minimum value of -66 and has a maximum value of 92 . Values must be between 0 and 100. Therefore, changes are needed
- Column `defending` has a minimum value of 18 and has a maximum value of 88 . Values are between 0 and 100 so no changes are needed.
- Column `physic` has a minimum value of 35 and has a maximum value of 85 . Values are between 0 and 100 so no changes are needed.
- Column `power_strength` has a minimum value of 29 and has a maximum value of 93 . Values are between 0 and 100 so no changes are needed.
- Column `power_long_shots` has a minimum value of 16 and has a maximum value of 88 . Values are between 0 and 100 so no changes are needed.

Checking for low values in the `wage_eur` column.

```
mydf[mydf["wage_eur"]<500,]
```

	sofifa_id<int>	potential<int>	wage_eur<dbl>	a...<int>	height_cm<int>	weight_kg<int>	club_name<chr>	preferred_foot<chr>	pace<int>
5111	222993	68	4.0001	30	177	68	Cerezo Osaka	Right	75

1 row | 1-10 of 18 columns

- There is one player which weekly wage in euros is 4. This player belongs to a team in Japan (Cerezo Osaka). This is consider an error since the minimum weekly wage for someone in Japan is around 340 euros. [1]
- Columns `pace` and `dribbling` need further investigation since their minimum value is -66.

```
mydf[mydf["pace"] < 0,]
```

	sofifa_id<int>	potential<int>	wage_eur<dbl>	a...<int>	height_cm<int>	weight_kg<int>	club_name<chr>
6011	237742	70	2000	25	179	80	Hokkaido Consadole Sapporo

1 row | 1-8 of 18 columns

Column `pace` has one row which value is -66 which is below 0.

```
mydf[mydf["dribbling"] < 0,]
```

	soffifa_id <int>	potential <int>	wage_eur <dbl>	... <int>	height_cm <int>	weight_kg <int>	club_name <chr>	preferred_foot <chr>	
	7561	248011	73	3000	22	180	76 Sepsis OSK Sf. Gheorghe	Right	
1 row   1-9 of 18 columns									

Column `dribbling` has one row which value is -66 which is below 0.

d. Looking for outliers in columns `wage_eur`, `age`, `height_cm` and `weight_kg`.

Outliers using IQR method:

```
c_col <- c("wage_eur", "age", "height_cm", "weight_kg")

# Outliers using IQR
for (col in c_col){
  # calculating iqr method
  iqr <- IQR(mydf[, col])
  # calculating q1 and q3
  quant <- quantile(mydf[, col], c(0.25, 0.75))
  # numbers below min limit will be considered outliers
  min_limit <- quant[1] - 1.5*iqr
  # numbers above max limit will be considered outliers
  max_limit <- quant[2] + 1.5*iqr
  print(mydf[mydf[col] > max_limit, ])
  print(mydf[mydf[col] < min_limit, ])
}
```

##	sofifa_id	potential	wage_eur	age	height_cm	weight_kg	
## 11	155862	89	300000	34	184	82	
## 36	218667	88	230000	25	173	64	
## 61	189513	85	59000	31	182	74	
## 86	146536	84	43000	34	172	60	
## 111	212194	88	75000	24	185	83	
## 136	186345	83	71000	29	173	71	
## 161	208574	83	45000	27	184	82	
## 186	176635	82	100000	31	180	71	
## 211	210665	82	65000	28	188	87	
## 236	179846	81	87000	33	189	90	
## 261	204884	84	115000	25	185	85	
## 286	225850	85	69000	24	189	80	
## 311	159145	80	53000	34	184	77	
## 336	189805	80	38000	29	188	86	
## 361	202750	80	84000	29	195	97	
## 386	219683	83	74000	25	181	81	
## 411	232656	85	37000	22	184	82	
## 436	178509	79	95000	33	193	91	
## 461	193983	79	43000	28	181	83	
## 486	206511	80	46000	26	184	74	
## 536	226456	83	41000	24	178	67	
## 561	234943	86	37000	23	183	74	
## 586	170733	78	38000	33	179	65	
## 611	192045	78	46000	29	177	76	
## 636	199915	80	56000	28	192	88	
## 661	207421	81	56000	25	172	67	
## 686	213160	81	51000	26	182	78	
## 711	223816	79	68000	26	177	79	
## 736	230442	78	33000	20	187	85	
## 786	183427	77	65000	30	174	78	
## 811	189681	77	36000	29	172	78	
## 886	208509	81	24000	25	177	67	
## 936	223627	77	42000	31	190	86	
## 961	230331	77	25000	36	177	77	
## 1011	164859	76	66000	31	176	68	
## 1036	184826	76	66000	31	175	69	
## 1086	201939	77	29000	27	189	86	
## 1111	208451	77	29000	26	183	77	
## 1161	219733	79	30000	26	197	82	
## 1186	225782	84	52000	22	177	70	
## 1211	230373	76	30000	24	172	69	
## 1236	235424	82	25000	24	176	68	
## 1286	162131	75	41000	35	193	92	
## 1361	191053	75	38000	32	177	76	
## 1386	194957	75	79000	28	185	71	
## 1411	200478	75	37000	28	183	84	
## 1436	205193	79	33000	25	186	84	
## 1536	222148	79	24000	25	181	75	
## 1586	229942	81	33000	22	190	86	
## 1661	236947	85	30000	22	191	80	
## 1811	199439	77	26000	26	193	87	
## 2086	247028	84	34000	22	186	76	
## 2111	163050	73	30000	34	175	70	
## 2361	224656	80	28000	23	184	82	
## 2386	229348	82	52000	22	183	70	
## 2711	208502	73	29000	27	178	73	
## 3036	241390	84	28000	20	184	74	
## 3411	227678	83	25000	22	183	77	
## 3486	234579	79	26000	23	177	78	
## 3786	209610	70	24000	32	166	64	
## 3811	212680	73	26000	25	174	61	
## 4636	234679	77	34000	23	190	82	
## 6186	253163	81	24000	21	188	79	
##		club_name	preferred_foot	pace	shooting	passing	dribbling
## 11		Real Madrid CF	Right	71	70	76	73
## 36		Manchester City	Left	81	77	83	92
## 61		Villarreal CF	Right	50	83	90	81
## 86		Sevilla FC	Right	87	69	79	83
## 111		Borussia Dortmund	Right	81	78	82	87
## 136		Atlético de Madrid	Right	73	60	81	77
## 161		Eintracht Frankfurt	Left	88	76	79	83
## 186		Arsenal	Left	64	69	85	83
## 211		RB Leipzig	Left	66	66	73	72
## 236		Juventus	Right	45	74	77	77
## 261		Manchester City	Left	80	54	77	77
## 286		Paris Saint-Germain	Left	73	39	67	67
## 311		Al Hilal	Right	67	82	67	76
## 336		Sevilla FC	Right	55	78	70	69
## 361		Wolverhampton Wanderers	Right	62	36	64	64
## 386		FC Bayern München	Right	72	77	78	78
## 411		AC Milan	Left	92	67	73	77
## 436		Chelsea	Left	39	79	70	71

## 461	Montpellier Hérault SC	Right	84	80	70	75
## 486	VfL Wolfsburg	Left	66	80	78	77
## 536	West Ham United	Right	68	76	78	80
## 561	Borussia Mönchengladbach	Right	67	69	76	81
## 586	Parma	Right	90	70	71	80
## 611	Tigres U.A.N.L.	Right	84	64	72	75
## 636	Brighton & Hove Albion	Right	51	49	62	63
## 661	Brighton & Hove Albion	Right	82	77	76	81
## 686	Spartak Moskva	Right	76	62	76	77
## 711	Cruz Azul	Right	87	78	68	81
## 736	São Paulo	Right	73	82	62	69
## 786	Everton	Left	66	68	74	76
## 811	Istanbul Basaksehir FK	Right	76	60	74	78
## 886	Stade de Reims	Right	75	57	70	72
## 936	Al Ahli	Right	69	79	69	72
## 961	Palmeiras	Right	67	72	75	73
## 1011	Everton	Right	86	70	69	78
## 1036	Leicester City	Right	53	73	77	75
## 1086	Atalanta	Right	76	42	64	70
## 1111	1. FSV Mainz 05	Right	75	77	68	79
## 1161	RC Strasbourg Alsace	Left	63	76	58	68
## 1186	Arsenal	Right	83	59	69	76
## 1211	São Paulo	Right	72	69	76	76
## 1236	TSG Hoffenheim	Right	60	49	67	72
## 1286	Napoli	Right	31	76	61	66
## 1361	Torino F.C.	Right	66	59	71	75
## 1386	Manchester United	Right	53	50	60	56
## 1411	Newcastle United	Right	67	68	74	76
## 1436	Hertha BSC	Left	69	39	58	62
## 1536	1. FC Köln	Right	68	73	72	79
## 1586	AS Monaco	Right	55	28	49	56
## 1661	Hertha BSC	Left	74	34	64	63
## 1811	FC Augsburg	Left	57	74	72	70
## 2086	Spartak Moskva	Right	75	68	67	74
## 2111	Sheffield United	Right	57	74	62	64
## 2361	Fulham	Right	86	61	68	74
## 2386	Fulham	Left	87	39	50	72
## 2711	Santos Laguna	Right	77	39	63	68
## 3036	Napoli	Right	67	69	71	76
## 3411	Aston Villa	Right	74	38	58	68
## 3486	Tigres U.A.N.L.	Right	88	69	60	72
## 3786	Club Deportivo Guadalajara	Right	72	53	68	66
## 3811	Rangers FC	Right	91	57	66	72
## 4636	Manchester City	Right	64	39	62	69
## 6186	FC Barcelona	Right	61	30	50	48
##	defending physic power_strength power_long_shots high.wage.ind					
## 11	88 85	85	64		1	
## 36	51 67	54	76		1	
## 61	71 69	70	88		1	
## 86	80 58	39	70		1	
## 111	51 66	71	80		1	
## 136	79 72	64	70		1	
## 161	68 80	77	78		1	
## 186	24 55	56	67		1	
## 211	81 80	79	69		1	
## 236	79 78	85	78		1	
## 261	77 77	79	65		1	
## 286	82 84	84	39		1	
## 311	36 74	83	75		1	
## 336	50 76	84	69		1	
## 361	80 83	93	23		1	
## 386	76 77	75	81		1	
## 411	74 77	78	69		1	
## 436	42 77	88	66		1	
## 461	33 81	84	74		1	
## 486	73 73	71	87		1	
## 536	70 67	63	80		1	
## 561	64 67	65	81		1	
## 586	35 57	52	64		1	
## 611	72 77	75	68		1	
## 636	79 78	83	67		1	
## 661	30 62	60	82		1	
## 686	76 72	66	62		1	
## 711	38 76	76	72		1	
## 736	31 75	79	80		1	
## 786	73 71	64	74		1	
## 811	74 76	72	62		1	
## 886	73 76	70	62		1	
## 936	39 80	88	76		1	
## 961	74 76	78	80		1	
## 1011	39 60	55	65		1	
## 1036	69 67	63	77		1	
## 1086	75 76	79	35		1	
## 1111	44 69	72	76		1	

```

## 1161      26      79      89      68      1
## 1186      69      68      66      62      1
## 1211      57      45      42      71      1
## 1236      75      74      68      54      1
## 1286      33      68      86      64      1
## 1361      74      78      72      72      1
## 1386      76      73      77      49      1
## 1411      70      75      73      68      1
## 1436      75      77      78      31      1
## 1536      54      61      55      68      1
## 1586      74      82      91      17      1
## 1661      77      74      78      25      1
## 1811      41      70      79      73      1
## 2086      72      79      76      72      1
## 2111      42      68      74      64      1
## 2361      70      73      75      68      1
## 2386      70      69      67      38      1
## 2711      68      71      66      40      1
## 3036      58      67      64      75      1
## 3411      71      72      72      32      1
## 3486      25      71      76      61      1
## 3786      71      70      63      64      1
## 3811      40      46      34      46      1
## 4636      70      68      73      52      1
## 6186      66      68      73      35      1
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)
##      sofifa_id potential wage_eur age height_cm weight_kg      club_name
## 4311      201965      69      4000 77      180      76      IFK Norrköping
## 6986      156478      65      2000 38      178      77      Urawa Red Diamonds
##      preferred_foot pace shooting passing dribbling defending physic
## 4311      Right      76      65      67      71      58      70
## 6986      Right      32      66      70      63      66      60
##      power_strength power_long_shots high.wage.ind
## 4311      66      73      0
## 6986      75      72      0
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)
##      sofifa_id potential wage_eur age height_cm weight_kg      club_name
## 9011      255592      69      1000 23      220      61      Yokohama FC
##      preferred_foot pace shooting passing dribbling defending physic
## 9011      Right      69      60      60      68      42      42
##      power_strength power_long_shots high.wage.ind
## 9011      36      64      0
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)
##      sofifa_id potential wage_eur age height_cm weight_kg
## 361      202750      80      84000 29      195      97
## 8511      211095      64      3000 27      191      188
## 9736      210820      66      550 25      189      93
##      club_name preferred_foot pace shooting passing dribbling
## 361      Wolverhampton Wanderers      Right      62      36      64      64
## 8511      AC Horsens      Right      60      60      48      57
## 9736      Varbergs BoIS FC      Right      57      51      48      50
##      defending physic power_strength power_long_shots high.wage.ind
## 361      80      83      93      23      1
## 8511      36      75      87      57      0
## 9736      59      71      82      49      0
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)

```

Outliers using Standard Deviation method:

```
# Outliers using sd
for (col in c_col){
  sd_col <- 3*sd(mydf[, col])
  mean_col <- mean(mydf[, col])
  min_limit <- mean_col - sd_col
  max_limit <- mean_col + sd_col
  print(mydf[mydf[col] > max_limit, ])
  print(mydf[mydf[col] < min_limit, ])
}
```

```
##      sofifa_id potential wage_eur age height_cm weight_kg
## 11      155862      89 300000 34      184      82
## 36      218667      88 230000 25      173      64
## 186     176635      82 100000 31      180      71
## 236     179846      81  87000 33      189      90
## 261     204884      84 115000 25      185      85
## 361     202750      80  84000 29      195      97
## 436     178509      79  95000 33      193      91
##      club_name preferred_foot pace shooting passing dribbling
## 11      Real Madrid CF      Right 71      70      76      73
## 36      Manchester City      Left 81      77      83      92
## 186      Arsenal      Left 64      69      85      83
## 236      Juventus      Right 45      74      77      77
## 261      Manchester City      Left 80      54      77      77
## 361 Wolverhampton Wanderers      Right 62      36      64      64
## 436      Chelsea      Left 39      79      70      71
##      defending physic power_strength power_long_shots high.wage.ind
## 11      88      85      85      64      1
## 36      51      67      54      76      1
## 186     24      55      56      67      1
## 236     79      78      85      78      1
## 261     77      77      79      65      1
## 361     80      83      93      23      1
## 436     42      77      88      66      1
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)
##      sofifa_id potential wage_eur age height_cm weight_kg      club_name
## 4311     201965      69  4000 77      180      76 IFK Norrköping
##      preferred_foot pace shooting passing dribbling defending physic
## 4311      Right 76      65      67      71      58      70
##      power_strength power_long_shots high.wage.ind
## 4311      66      73      0
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)
##      sofifa_id potential wage_eur age height_cm weight_kg      club_name
## 9011     255592      69 1000 23      220      61 Yokohama FC
##      preferred_foot pace shooting passing dribbling defending physic
## 9011      Right 69      60      60      68      42      42
##      power_strength power_long_shots high.wage.ind
## 9011      36      64      0
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)
##      sofifa_id potential wage_eur age height_cm weight_kg      club_name
## 8511     211095      64  3000 27      191      188 AC Horsens
##      preferred_foot pace shooting passing dribbling defending physic
## 8511      Right 60      60      48      57      36      75
##      power_strength power_long_shots high.wage.ind
## 8511      87      57      0
## [1] sofifa_id      potential      wage_eur      age
## [5] height_cm      weight_kg      club_name      preferred_foot
## [9] pace      shooting      passing      dribbling
## [13] defending      physic      power_strength      power_long_shots
## [17] high.wage.ind
## <0 rows> (or 0-length row.names)
```

- The first outlier found was a player which age is 77. This number I believe is incorrect because it is too old to be a football player at professional football. [2]
- The second outlier found was a player which height is 220cm. This number I believe is incorrect. [3]
- The third outlier found was a player which weight is 188kg. This number is to high to be correct. [4]

e. Checking for duplicates on the `sofifa_id`



```
# Check for duplicated values
num_duplic = sum(duplicated(mydf$sofifa_id))
message = paste("There are", num_duplic, "duplicated values")
print(message)
```

```
## [1] "There are 1 duplicated values"
```

```
mydf[duplicated(mydf$sofifa_id),]
```

	sofifa_id <int>	potential <int>	wage_eur <dbl>	... <int>	height_cm <int>	weight_kg <int>	club_name <chr>	preferred_foot <chr>	pace <int>	
	4011	230400	70	8000	31	185	84 Atlético Tucumán	Right	68	

1 row | 1-10 of 18 columns

There is 1 duplicated row on the `sofifa_id` column.

f. Checking for unique values on the `high.wage.ind` and `preferred_foot` columns.

```
table(mydf$"high.wage.ind")
```

```
##
##  0  1
## 365 149
```

There are 365 values for "0" and 149 values for "1"

```
table(mydf$"preferred_foot")
```

```
##
## Left right Right
## 121    1   392
```

Values must be constrained to 2 unique values: "Left", "Right", however, 1 "right" value was found and it must be changed. There are 121 values for "Left" and 392 values for "Right".

g. Checking for binary value to be correct. It should be "1" if weekly wage is above 8000 and "0" otherwise.

```
a <- mydf[mydf["wage_eur"] > 8000,]
sum(mydf[mydf["wage_eur"] > 8000,]["high.wage.ind"])
```

```
## [1] 149
```

```
sum(mydf[mydf["wage_eur"] <= 8000,]["high.wage.ind"])
```

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

All values above 8000 are "1" and "0" otherwise. Not mistakes were found.

## 1.3 Data cleaning

Explain any data quality issues found in 1.2 (5 marks), justify and document the responses made (if any) (5 marks).

Data quality issues found:

1. `preferred_foot` column has 1 value which is "right", however, it should be "Right".
2. `sofifa_id` column is int and it should be changed to string (char). `preferred_foot` column is a string (char) and should be changed to factor. `high.wage.ind` is a string (char) and should be changed to factor.
3. `pace` and `dribbling` columns both have 1 row which value is below 0. They need to be changed.
4. The ideal weight for a football player is 48kg for the first 152cm and then 2.7kg for every extra 2.5cm. Therefore, if we have a player that has an altitude of 191cm, it is impossible that his weight is 188kg. Therefore this outlier is considered a mistake. [4]
5. The tallest football player according to FIFA has a height of 206cm. However, one football player whose height is 220cm was found. Therefore, this outlier will be considered as a mistake. [3]
6. `sofifa_id` has 1 duplicated value. Id values can't have duplicate values since it is an identifier of a person.
7. It was found that 1 player has an age of 77. The oldest football player in professional football has an age of 54. Therefore, this outlier is consider a mistake. [2]
8. There is one football player in a football club in Japan which weekly wage is 4 euros. This does not make sense since the minimum weekly salary in Japan is around 340 euros. This was not identified as an outlier, however this is considered a mistake aswell. [1]

## Data Cleaning Implementation:

```
df_clean <- data.frame(mydf)
```

## 1. Fixing the "right" value to "Right"

```
df_clean[df_clean["preferred_foot"]=="right",]["preferred_foot"] <- "Right"
table(df_clean$preferred_foot)
```

```
##
## Left Right
## 121 393
```

## 2. Converting sofifa\_id, preferred\_foot and high.wage.ind to their factor data type.

```
df_clean$sofifa_id <- as.character(df_clean$sofifa_id)
df_clean$preferred_foot <- as.factor(df_clean$preferred_foot)
df_clean$high.wage.ind <- as.factor(df_clean$high.wage.ind)
str(df_clean)
```

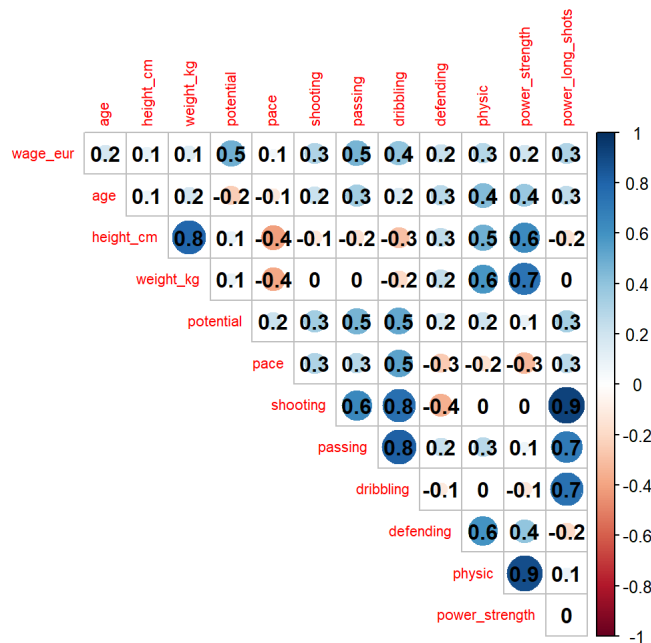
```
## 'data.frame': 514 obs. of 17 variables:
## $ sofifa_id : chr "155862" "218667" "189513" "146536" ...
## $ potential : int 89 88 85 84 88 83 83 82 82 81 ...
## $ wage_eur : num 300000 230000 59000 43000 75000 71000 45000 100000 65000 87000 ...
## $ age : int 34 25 31 34 24 29 27 31 28 33 ...
## $ height_cm : int 184 173 182 172 185 173 184 180 188 189 ...
## $ weight_kg : int 82 64 74 60 83 71 82 71 87 90 ...
## $ club_name : chr "Real Madrid CF" "Manchester City" "Villarreal CF" "Sevilla FC" ...
## $ preferred_foot : Factor w/ 2 levels "Left","Right": 2 1 2 2 2 1 1 1 2 ...
## $ pace : int 71 81 50 87 81 73 88 64 66 45 ...
## $ shooting : int 70 77 83 69 78 60 76 69 66 74 ...
## $ passing : int 76 83 90 79 82 81 79 85 73 77 ...
## $ dribbling : int 73 92 81 83 87 77 83 83 72 77 ...
## $ defending : int 88 51 71 80 51 79 68 24 81 79 ...
## $ physic : int 85 67 69 58 66 72 80 55 80 78 ...
## $ power_strength : int 85 54 70 39 71 64 77 56 79 85 ...
## $ power_long_shots: int 64 76 88 70 80 70 78 67 69 78 ...
## $ high.wage.ind : Factor w/ 2 levels "0","1": 2 2 2 2 2 2 2 2 2 2 ...
```

Now, all data types are correct.

We will try to use correlation to find the most suitable columns to be used in a regression model to fix the incorrect numbers.

```
num_col <- c("wage_eur",
            "age",
            "height_cm",
            "weight_kg",
            "potential",
            "pace",
            "shooting",
            "passing",
            "dribbling",
            "defending",
            "physic",
            "power_strength",
            "power_long_shots")

df_cor <- subset(df_clean, select=num_col)
df_cor <- df_cor[df_cor$pace > 0 & df_cor$dribbling > 0 & df_cor$weight_kg != 188 & df_cor$height_cm != 220, ]
cor_matrix <- cor(df_cor)
corrplot(cor_matrix, diag=F, type="upper", insig="p-value", number.digits=1, addCoef.col="black", tl.cex=0.7)
```



Since the `pace` value is not highly correlated with any value, the incorrect value will be replaced by the mean of the `pace` column. The column `dribbling` is highly correlated with the columns `passing` and `shooting`. Therefore, any of both values can be used to correct the negative values. The relation between `passing` and `dribbling` will be used to correct the negative values.

3. Correcting the value -66 in the `pace` column with the mean. The mean value was calculated without the -66 value. Also, correcting the value -66 in the `dribbling` column using the `passing` column since it is the most correlated to it.

```
mean_pace <- mean(df_clean[df_clean$"pace" != -66,]$"pace")
df_clean[df_clean$"pace" == -66,]$"pace" <- mean_pace
```

Checking if changes in the `pace` column were made.

```
df_clean[df_clean$"pace" < 0,]
```

0 rows | 1-10 of 17 columns

The column `pace` now does not have negative values.

Since `dribbling` is highly correlated with `passing`, I will build a regression model with `dribbling` as my dependent variable and `passing` as my independent variable to fix the -66 value.

```
model_dribbling <- lm(dribbling~passing, data=df_clean)
summary(model_dribbling)
```

```
##
## Call:
## lm(formula = dribbling ~ passing, data = df_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -131.300   -3.257    0.265    3.864   20.049
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  17.55846    2.05531   8.543 <0.000000000000002 ***
## passing      0.78265    0.03501  22.355 <0.000000000000002 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.139 on 512 degrees of freedom
## Multiple R-squared:  0.494, Adjusted R-squared:  0.493
## F-statistic: 499.8 on 1 and 512 DF, p-value: < 0.0000000000000022
```

61 is the number in the `passing` column where the error in the `dribbling` column is located.

```
passing_value <- as.data.frame(61)
colnames(passing_value) <- c("passing")
dribbling_value <- predict(model_dribbling, newdata=passing_value)
dribbling_value
```

```
##      1
## 65.29981
```

```
df_clean[df_clean$"dribbling" == -66,]$"dribbling" <- dribbling_value
```

```
df_clean[df_clean$"dribbling" < 0,]
```

0 rows | 1-10 of 17 columns

Dribbling values below 0 have been fixed.

4. Since `weight_kg` is highly correlated with `height_cm`, I will build a regression model with `weight_kg` as my dependent variable and `height_cm` as my independent variable to fix the 188 value. 191cm is the height of the football player which weight is 188kg and that we consider a mistake.

```
# dataframe without the errors in the weight_kg column and the height_cm column
df_model_heigh_weight <- df_clean[df_clean$"weight_kg" != 188 & df_clean$"height_cm" != 220, ]
```

```
# model with weight_kg as dependent variable
model_weight_height <- lm(weight_kg~height_cm, data=df_model_heigh_weight)
summary(model_weight_height)
```

```
##
## Call:
## lm(formula = weight_kg ~ height_cm, data = df_model_heigh_weight)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -13.3609  -2.5446   0.0525   2.4765  12.0441
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)  -69.95846    4.89685  -14.29 <0.000000000000002 ***
## height_cm      0.80168     0.02718   29.49 <0.000000000000002 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.196 on 510 degrees of freedom
## Multiple R-squared:  0.6303, Adjusted R-squared:  0.6296
## F-statistic: 869.7 on 1 and 510 DF,  p-value: < 0.0000000000000022
```

```
height_value <- as.data.frame(191)
colnames(height_value) <- c("height_cm")
predicted_weight <- predict(model_weight_height, height_value)
predicted_weight
```

```
##      1
## 83.16263
```

```
df_clean[df_clean$"weight_kg" == 188, ]["weight_kg"] <- predicted_weight
```

The outlier on the `weight_kg` got fixed.

5. Since `height_cm` is highly correlated with `weight_kg`, I will build a regression model with `height_cm` as my dependent variable and `weight_kg` as my independent variable to fix the 220 value. 61kg is the weight of the football player which height is 220cm and that we consider a mistake.

```
model_height_weight <- lm(height_cm~weight_kg, data=df_model_heigh_weight)
summary(model_height_weight)
```

```
##
## Call:
## lm(formula = height_cm ~ weight_kg, data = df_model_heigh_weight)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -12.0890  -2.8007  -0.1655   2.7679  14.2012
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)  121.54567    1.99073   61.06 <0.000000000000002 ***
## weight_kg      0.78628     0.02666   29.49 <0.000000000000002 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.156 on 510 degrees of freedom
## Multiple R-squared:  0.6303, Adjusted R-squared:  0.6296
## F-statistic: 869.7 on 1 and 510 DF,  p-value: < 0.0000000000000022
```

```
weight_value <- as.data.frame(61)
colnames(weight_value) <- c("weight_kg")
predicted_height <- predict(model_height_weight, weight_value)
predicted_height
```

```
##           1
## 169.5085
```

```
df_clean[df_clean["height_cm"] == 220,]["height_cm"] <- predicted_height
```

The outlier on the height\_cm got fixed.

6. Changing the duplicated id to any string that is unique in the column

```
df_clean[duplicated(df_clean$sofifa_id),]["sofifa_id"] <- "101096"
df_clean[duplicated(df_clean$sofifa_id),]
```

0 rows | 1-10 of 17 columns

The duplicated value on the sofifa\_id was modified to 101096. Why? it is a random number I picked.

7. Changing the age value to the second max value in the column

```
df_clean[df_clean["age"]==77,]["age"] <- 38
max(df_clean$"age")
```

```
## [1] 38
```

We replaced the age of 77 by 38 because it is the max value after 77.

8. Changing the wage value to its minimum value after 4.

```
df_clean[df_clean["wage_eur"]<500,]["wage_eur"] <- 500
min(df_clean$"wage_eur")
```

```
## [1] 500
```

```
df <- data.frame(df_clean)
df
```

	sofifa_id <chr>	potential <int>	wage_eur <dbl>	a... <dbl>	height_cm <dbl>	weight_kg <dbl>	club_name <chr>	
11	155862	89	300000	34	184.0000	82.00000	Real Madrid CF	
36	218667	88	230000	25	173.0000	64.00000	Manchester City	
61	189513	85	59000	31	182.0000	74.00000	Villarreal CF	
86	146536	84	43000	34	172.0000	60.00000	Sevilla FC	
111	212194	88	75000	24	185.0000	83.00000	Borussia Dortmund	
136	186345	83	71000	29	173.0000	71.00000	Atlético de Madrid	
161	208574	83	45000	27	184.0000	82.00000	Eintracht Frankfurt	
186	176635	82	100000	31	180.0000	71.00000	Arsenal	
211	210665	82	65000	28	188.0000	87.00000	RB Leipzig	
236	179846	81	87000	33	189.0000	90.00000	Juventus	
1-10 of 514 rows   1-8 of 18 columns								Previous 1 2 3 4 5 6 ... 52 Next

List of modifications done:

1. The row which value on preferred\_foot was "right" was changed to "Right". This was done because this column only admits 2 values: "Right" and "Left".
2. sofifa\_id column was converted to string data type. preferred\_foot column was changed to factor data type. high.wage.ind column was converted to factor data type. Those changes were done because those columns are intended to be used in such way.
3. The column pace has 1 value below 0 which is -66. Column pace does not have any column which is highly correlated. Therefore, we calculated the mean of the column pace (without using the -66 value) and we change the -66 with the mean value. The column dribbling is highly correlated to passing (correlation of 0.8 without using the -66 value). Therefore, to fix the value, a linear regression model was built and used to predict the dribbling value.
4. weight\_kg is highly correlated to height\_cm (correlation of 0.8 without using the outliers). Therefore a linear regression model was used to fix the outlier of 188 on the weight\_kg column.
5. height\_cm is highly correlated to weight\_kg (correlation of 0.8 without using the outliers). Therefore a linear regression model was used to fix the outlier of 220 on the height\_cm column.
6. sofifa\_id column had a duplicated value. An id is a unique value that identifies a player. This id could be modified to any value as long as it is not repeated. We used the value of "101096". It is an arbitrary number but I decided to use this one as it is random and is not used in the column.
7. The second max value of a player is 38. Therefore, the player which age is 77 was changed to 38.
8. 500 is the minimum weekly wage after 4. This number was used to replace 4.

2. Exploratory Data Analysis (EDA)

## 2.1 EDA plan

*Outline a suitable plan to explore, describe and visualize your data. (5 marks)*

### Univariate EDA planning:

1. For columns `potential`, `wage_eur`, `age`, `height_cm`, `weight_kg`, `pace`, `shooting`, `passing`, `dribbling`, `defending`, `physic`, `power_strength`, `power_long_shots` columns, I will explore the maximum value, the minimum value, the range value, the standard deviation and the interquartile range (IQR). The mean value and median value will also be calculated. Furthermore, outliers will be identified by using boxplots. Histograms will be graphed to check if they follow a normal distribution or if they are skewed to any side. Furthermore, qqplot graphs and Shapiro tests will be done to check for normality.
2. For columns which datatype is factor ( `preferred_foot`, `high.wage.ind` ), barplots and tables will be done.

### Bivariate EDA planning:

Since our research questions are regarding columns `potential` and `high.wage.ind`, I will do the following:

1. Research question regarding column `potential` :
  - a. I will make the column `potential` as our dependent variable and we will create scatterplots against the following columns: `wage_eur`, `age`, `height_cm`, `weight_kg`, `pace`, `shooting`, `passing`, `dribbling`, `defending`, `physic`, `power_strength`, `power_long_shots`. Furthermore, a correlation plot will be created to see insight of the correlation that exist between `potential` and the other columns and also look for collinearity. Then, correlation tests will be done.
  - b. I will make `potential` as my dependent variable and perform ANOVA analysis with `preferred_foot` and `high.wage.ind` separately.
2. Research question regarding column `high.wage.ind` :
  - a. I will create boxplots with `high.wage.ind` on the x-axis and columns `wage_eur`, `age`, `height_cm`, `weight_kg`, `pace`, `shooting`, `passing`, `dribbling`, `defending`, `physic`, `power_strength`, `power_long_shots` and `potential` in the y-axis. The I will perform t-tests to see if there is significance difference or not between their means.
  - b. I will perform a chi-squared test to see if there is a relationship between `high.wage.ind` and `preferred_foot`

## 2.2 EDA and summary of results

*Undertake and summaries the findings of your data exploration, particularly with respect to the research questions. Use appropriate summary statistics (uni- and multi-variate) and visualizations. (10 marks)*

### Univariate EDA

```
# This function is used to calculate max, min, range, iqr, std, mean and medians of a single numerical column
numeric_eda <- function(col){

  max_value = max(df[col])
  min_value = min(df[col])
  range_value = max_value - min_value
  iqr_value = IQR(df[, col])
  std_value = sd(df[, col])
  mean_value = mean(df[, col])
  median_value = median(df[, col])

  max_message = paste("The max value of the column", col, "is", round(max_value,2))
  min_message = paste("The min value of the column", col, "is", round(min_value,2))
  range_message = paste("The range value of the column", col, "is", round(range_value,2))
  iqr_message = paste("The column", col, "has an interquartile range of", round(iqr_value,2))
  std_message = paste("The column", col, "has an standard deviation of", round(std_value,2))
  mean_message = paste("The mean value of the column", col, "is", round(mean_value,2))
  median_message = paste("The median value of the column", col, "is", round(median_value,2))

  print(max_message)
  print(min_message)
  print(range_message)
  print(iqr_message)
  print(std_message)
  print(mean_message)
  print(median_message)
  print("-----")
}
```

```
numeric_columns <- c("potential",  
                     "wage_eur",  
                     "age",  
                     "height_cm",  
                     "weight_kg",  
                     "pace",  
                     "shooting",  
                     "passing",  
                     "dribbling",  
                     "defending",  
                     "physic",  
                     "power_strength",  
                     "power_long_shots")  
  
for (col in numeric_columns){  
  numeric_eda(col)  
}
```

```

## [1] "The max value of the column potential is 89"
## [1] "The min value of the column potential is 54"
## [1] "The range value of the column potential is 35"
## [1] "The column potential has an interquartile range of 9"
## [1] "The column potential has an standard deviation of 6.39"
## [1] "The mean value of the column potential is 71.67"
## [1] "The median value of the column potential is 72"
## [1] "-----"
## [1] "The max value of the column wage_eur is 300000"
## [1] "The min value of the column wage_eur is 500"
## [1] "The range value of the column wage_eur is 299500"
## [1] "The column wage_eur has an interquartile range of 9000"
## [1] "The column wage_eur has an standard deviation of 22823.07"
## [1] "The mean value of the column wage_eur is 10821.6"
## [1] "The median value of the column wage_eur is 3000"
## [1] "-----"
## [1] "The max value of the column age is 38"
## [1] "The min value of the column age is 16"
## [1] "The range value of the column age is 22"
## [1] "The column age has an interquartile range of 6"
## [1] "The column age has an standard deviation of 4.56"
## [1] "The mean value of the column age is 25.15"
## [1] "The median value of the column age is 25"
## [1] "-----"
## [1] "The max value of the column height_cm is 200"
## [1] "The min value of the column height_cm is 160"
## [1] "The range value of the column height_cm is 40"
## [1] "The column height_cm has an interquartile range of 10"
## [1] "The column height_cm has an standard deviation of 6.85"
## [1] "The mean value of the column height_cm is 180"
## [1] "The median value of the column height_cm is 180"
## [1] "-----"
## [1] "The max value of the column weight_kg is 97"
## [1] "The min value of the column weight_kg is 58"
## [1] "The range value of the column weight_kg is 39"
## [1] "The column weight_kg has an interquartile range of 9"
## [1] "The column weight_kg has an standard deviation of 6.92"
## [1] "The mean value of the column weight_kg is 74.34"
## [1] "The median value of the column weight_kg is 75"
## [1] "-----"
## [1] "The max value of the column pace is 94"
## [1] "The min value of the column pace is 31"
## [1] "The range value of the column pace is 63"
## [1] "The column pace has an interquartile range of 14"
## [1] "The column pace has an standard deviation of 11.15"
## [1] "The mean value of the column pace is 68.39"
## [1] "The median value of the column pace is 69"
## [1] "-----"
## [1] "The max value of the column shooting is 83"
## [1] "The min value of the column shooting is 21"
## [1] "The range value of the column shooting is 62"
## [1] "The column shooting has an interquartile range of 23"
## [1] "The column shooting has an standard deviation of 14.2"
## [1] "The mean value of the column shooting is 53.04"
## [1] "The median value of the column shooting is 55"
## [1] "-----"
## [1] "The max value of the column passing is 90"
## [1] "The min value of the column passing is 26"
## [1] "The range value of the column passing is 64"
## [1] "The column passing has an interquartile range of 14"
## [1] "The column passing has an standard deviation of 10.26"
## [1] "The mean value of the column passing is 57.81"
## [1] "The median value of the column passing is 58"
## [1] "-----"
## [1] "The max value of the column dribbling is 92"
## [1] "The min value of the column dribbling is 32"
## [1] "The range value of the column dribbling is 60"
## [1] "The column dribbling has an interquartile range of 13"
## [1] "The column dribbling has an standard deviation of 9.91"
## [1] "The mean value of the column dribbling is 63.06"
## [1] "The median value of the column dribbling is 64"
## [1] "-----"
## [1] "The max value of the column defending is 88"
## [1] "The min value of the column defending is 18"
## [1] "The range value of the column defending is 70"
## [1] "The column defending has an interquartile range of 29"
## [1] "The column defending has an standard deviation of 16.85"
## [1] "The mean value of the column defending is 51.53"
## [1] "The median value of the column defending is 57"
## [1] "-----"
## [1] "The max value of the column physic is 85"
## [1] "The min value of the column physic is 35"
## [1] "The range value of the column physic is 50"

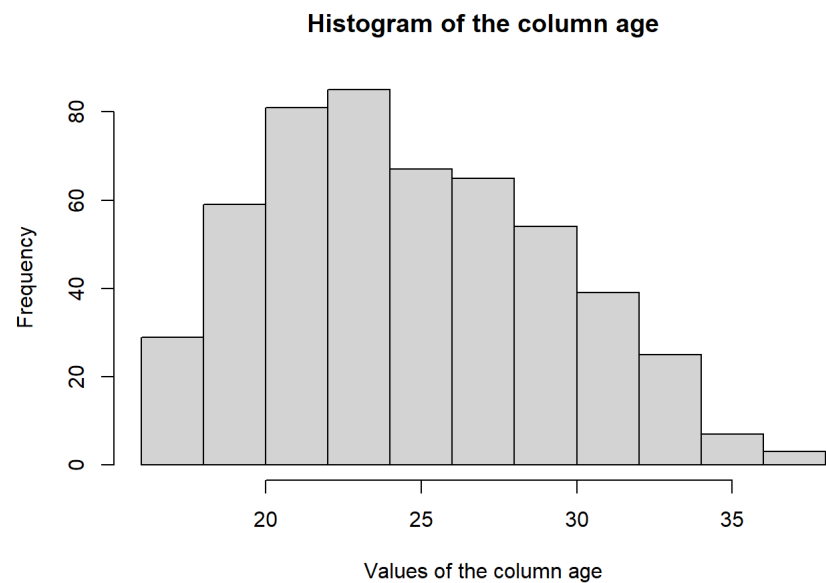
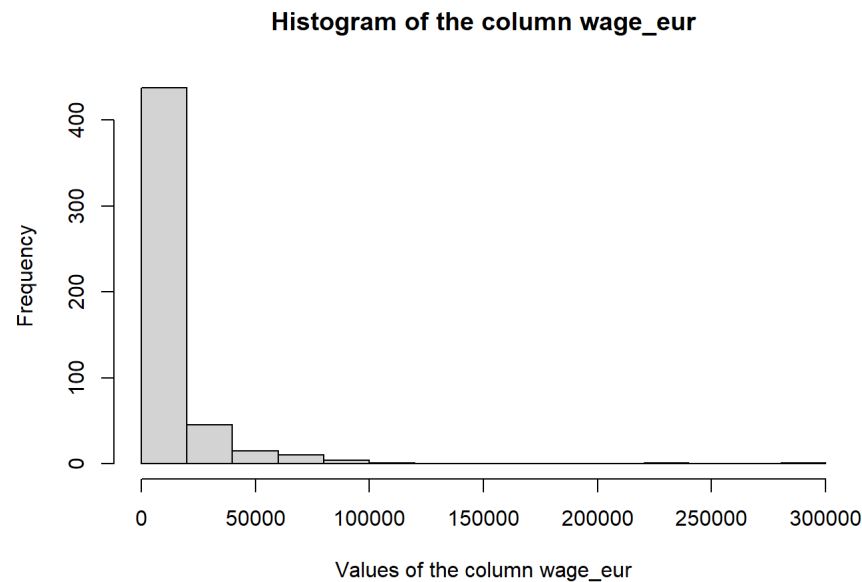
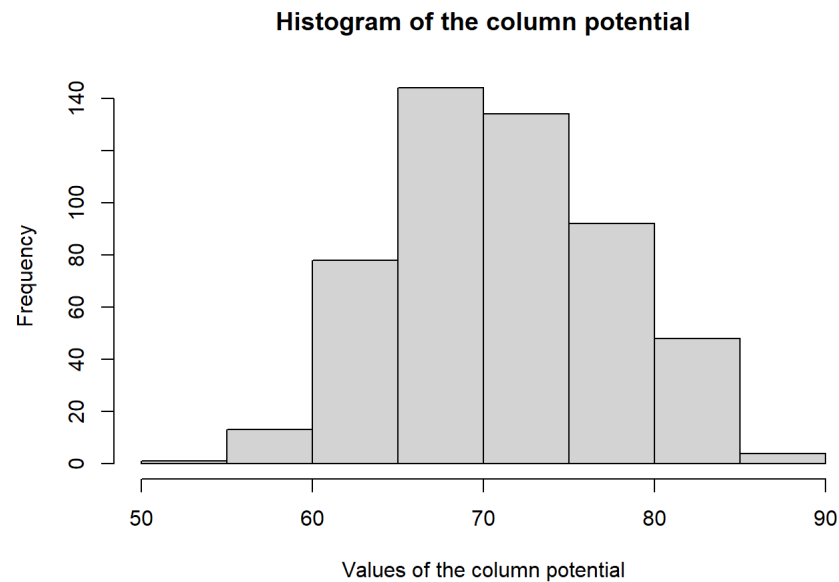
```

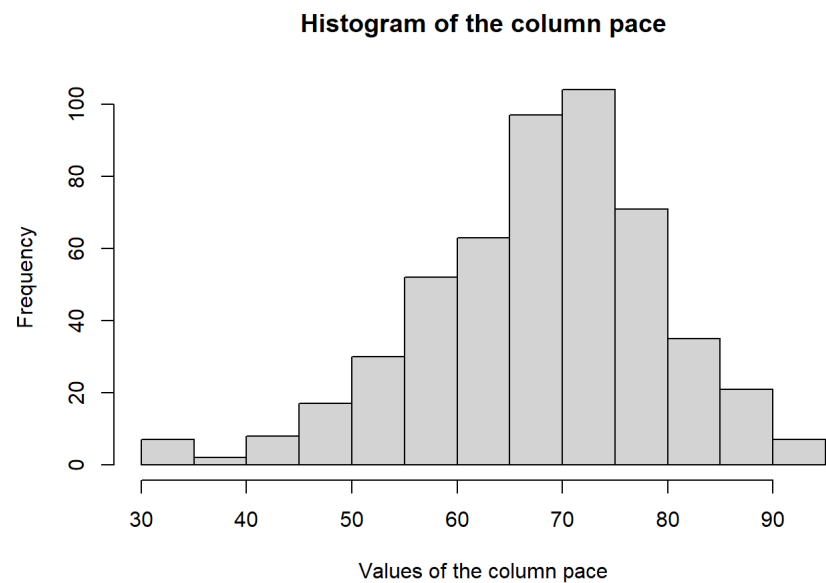
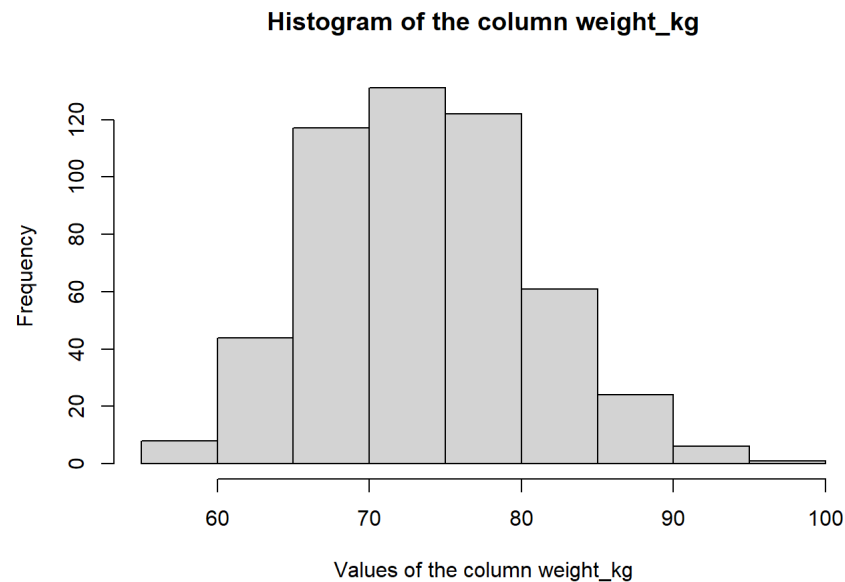
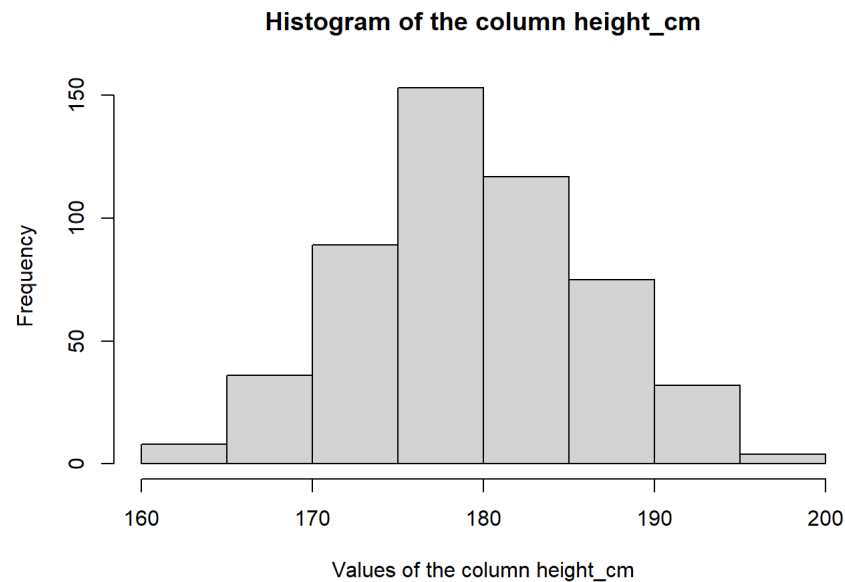


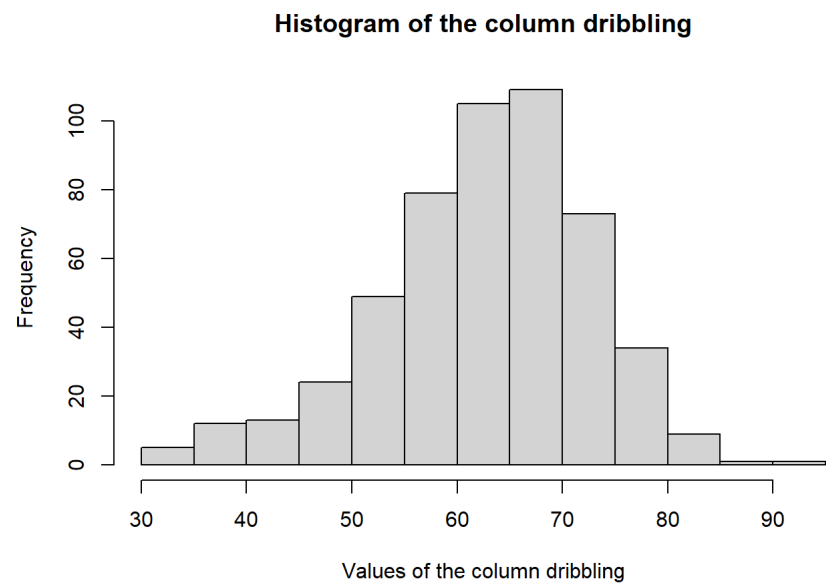
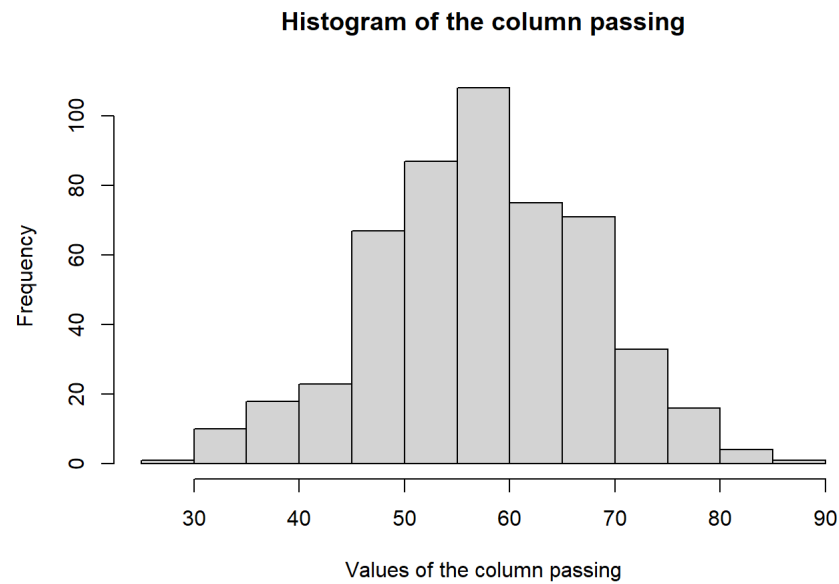
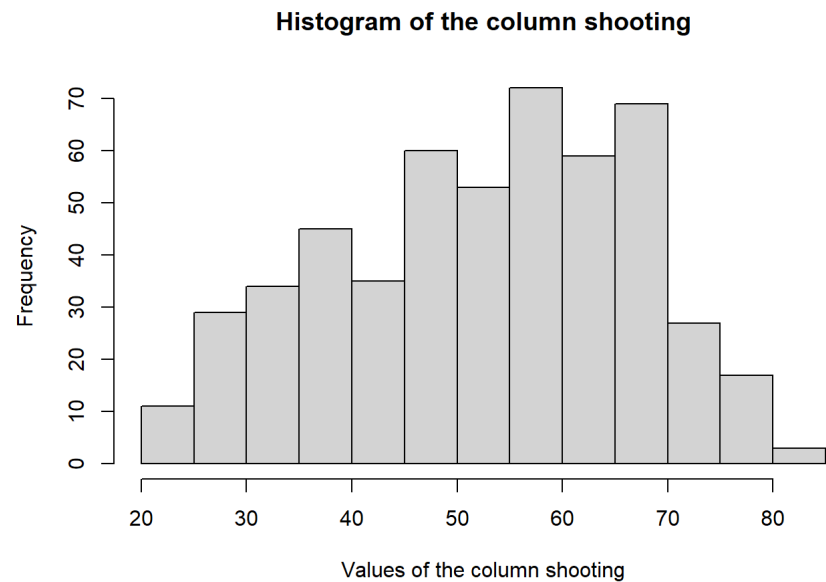
```
## [1] "The column physic has an interquartile range of 14"
## [1] "The column physic has an standard deviation of 10.03"
## [1] "The mean value of the column physic is 64.61"
## [1] "The median value of the column physic is 66"
## [1] "-----"
## [1] "The max value of the column power_strength is 93"
## [1] "The min value of the column power_strength is 29"
## [1] "The range value of the column power_strength is 64"
## [1] "The column power_strength has an interquartile range of 18"
## [1] "The column power_strength has an standard deviation of 13.24"
## [1] "The mean value of the column power_strength is 65.31"
## [1] "The median value of the column power_strength is 67"
## [1] "-----"
## [1] "The max value of the column power_long_shots is 88"
## [1] "The min value of the column power_long_shots is 16"
## [1] "The range value of the column power_long_shots is 72"
## [1] "The column power_long_shots has an interquartile range of 24"
## [1] "The column power_long_shots has an standard deviation of 15.65"
## [1] "The mean value of the column power_long_shots is 51.83"
## [1] "The median value of the column power_long_shots is 54"
## [1] "-----"
```

Now we will plot histograms for each column:

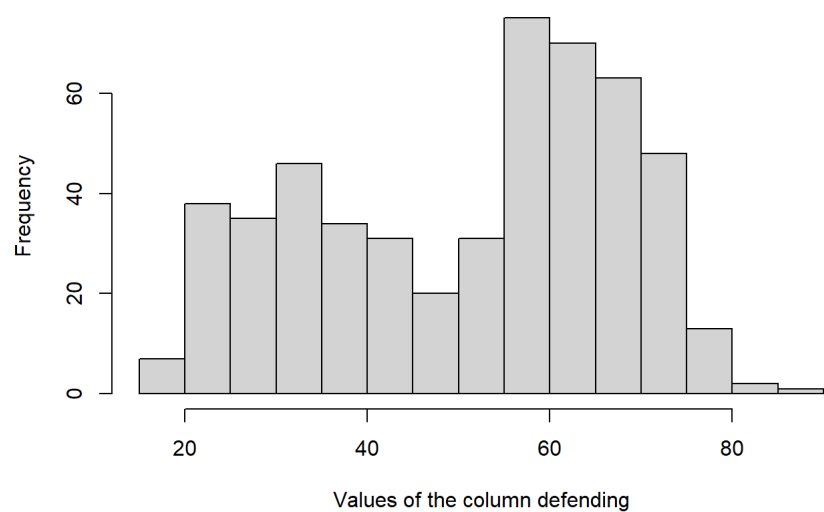
```
for (col in numeric_columns){
  title = paste("Histogram of the column", col)
  x_axis = paste("Values of the column", col)
  hist(df[,col], main=title, xlab=x_axis)
}
```



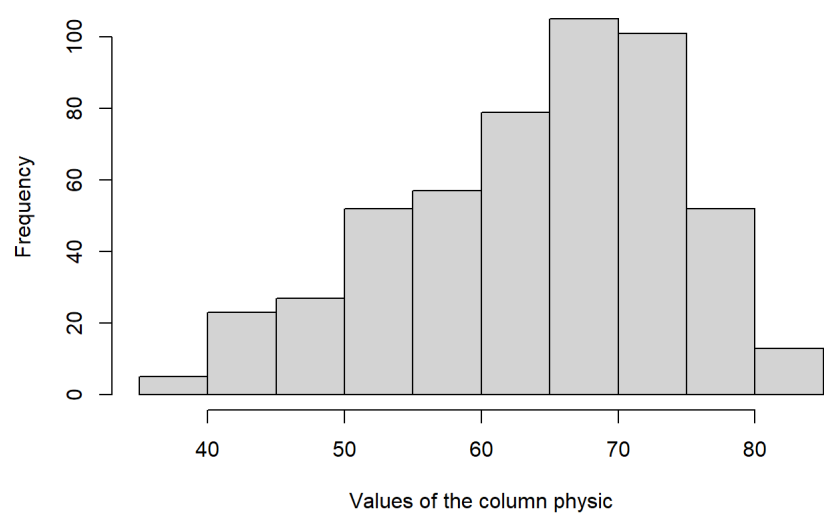




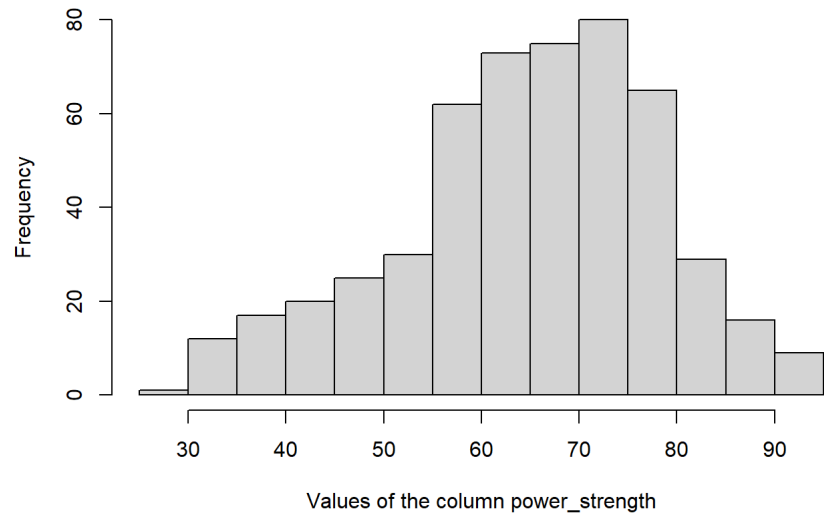
Histogram of the column defending

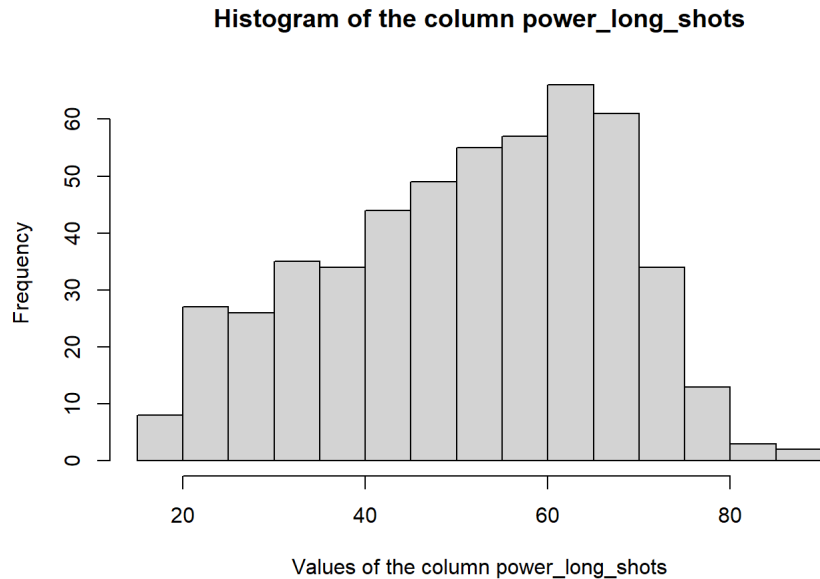


Histogram of the column physic



Histogram of the column power\_strength

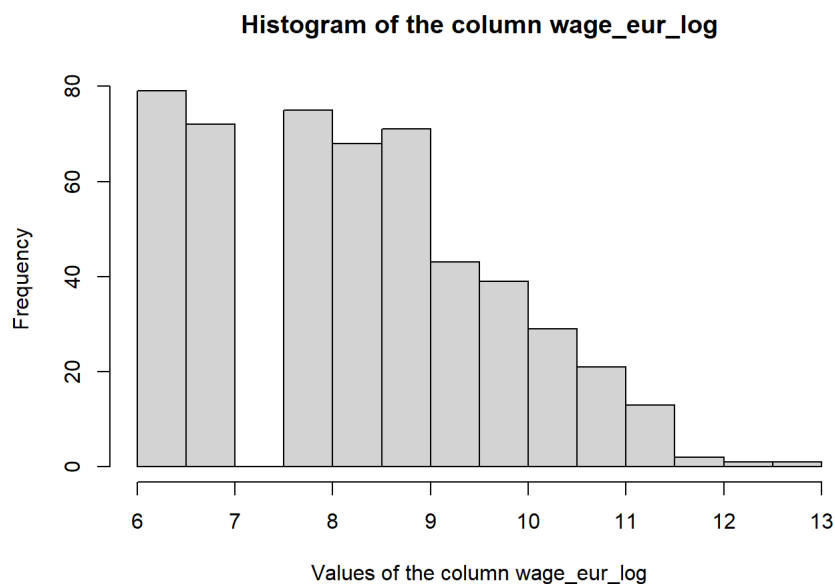




- potential column seems to follow a normal distribution.
- wage\_eur column seems to follow logarithmic distribution. [5]
- age column seems to be slightly skewed to the right.
- height\_cm column seems to follow a normal distribution.
- weight\_kg column seems to be slightly skewed to the right.
- pace column seems to be skewed to the left.
- shooting column seems to be skewed to the left.
- passing column seems to follow a normal distribution.
- dribbling column seems to be slightly skewed to the left.
- defending column seems to have 2 picks.
- physic column seems to be skewed to the left.
- power\_strength column seems to be slightly skewed to the left.
- power\_long\_shots column seems to be skewed to the left.

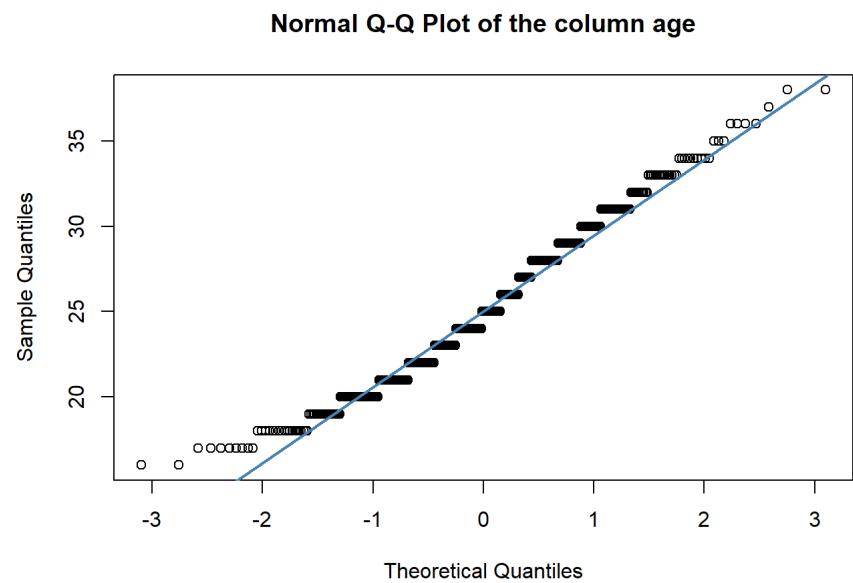
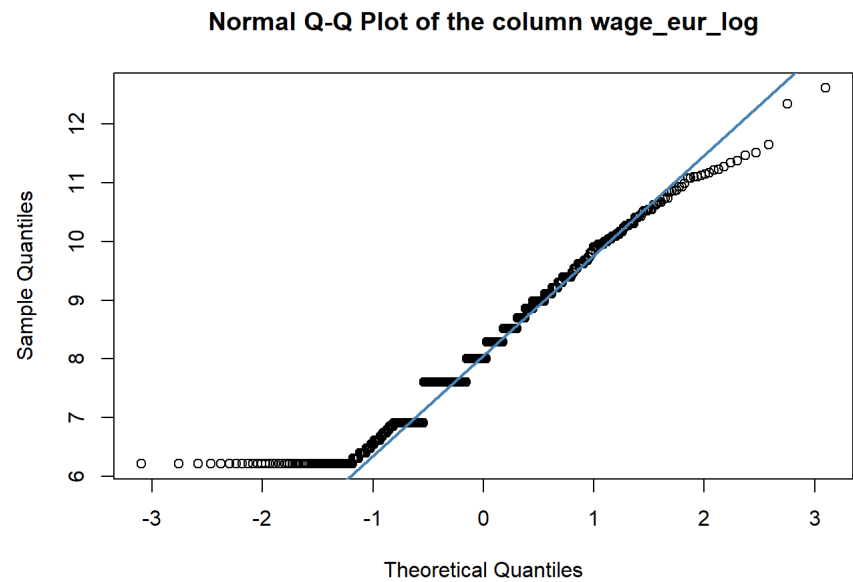
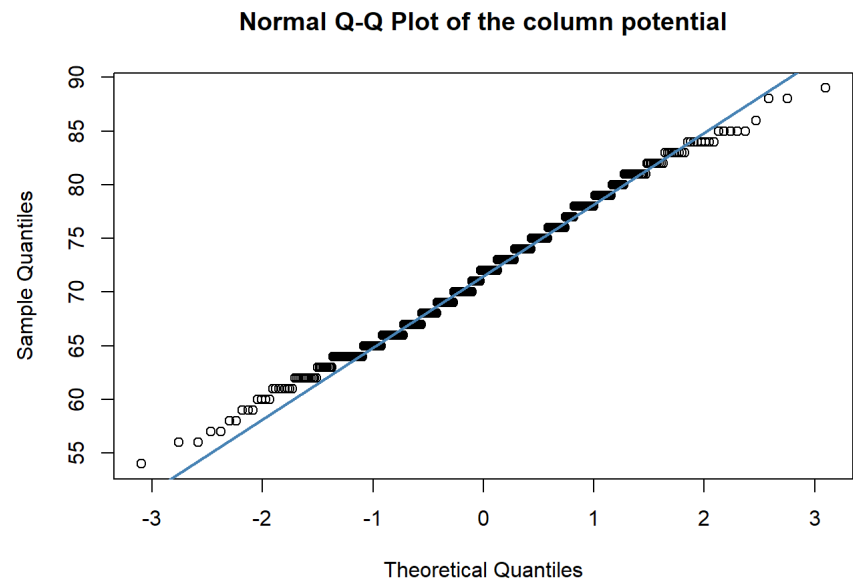
Since the column wage\_eur seems to follow a logarithmic distribution, I will use a log function to transform values so that it tries to follow now a normal distribution.

```
df$wage_eur_log <- log(df$wage_eur)
hist(df$wage_eur_log, main="Histogram of the column wage_eur_log", xlab="Values of the column wage_eur_log")
```

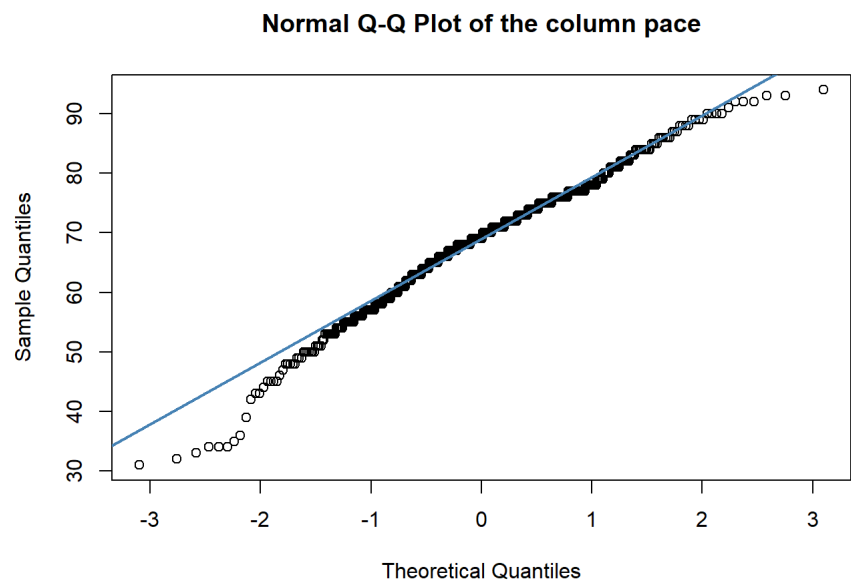
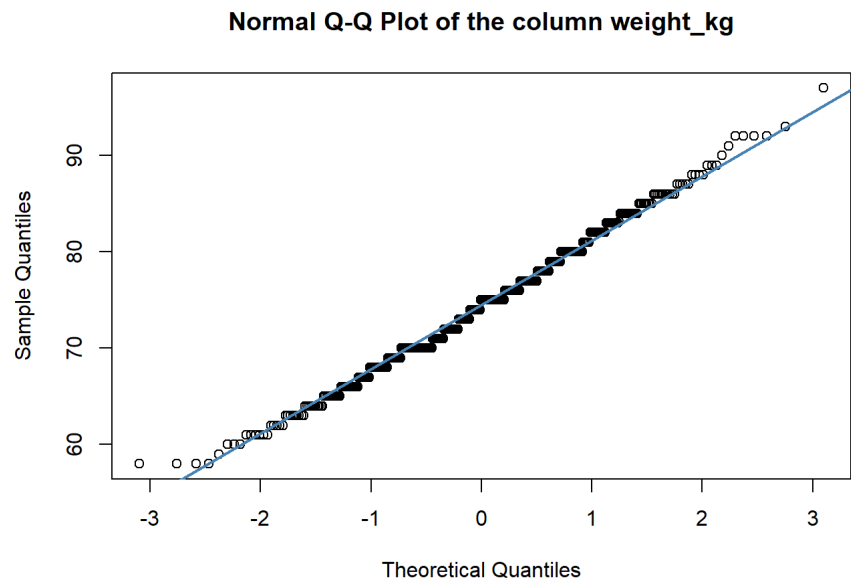
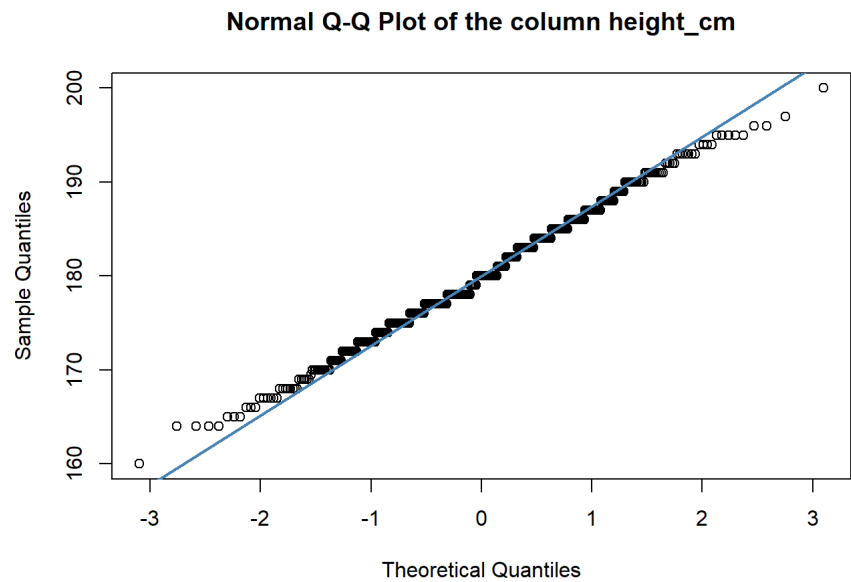


- wage\_eur\_log seems to be skewed to the right.

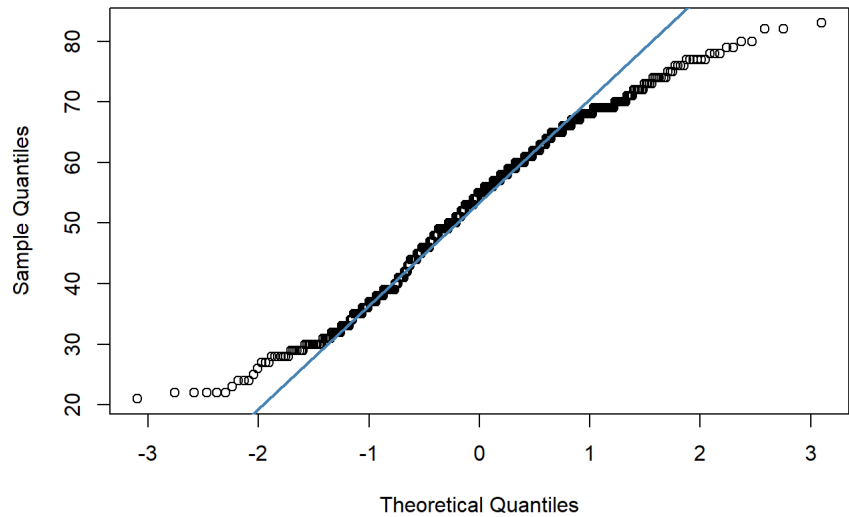
```
numeric_columns <- c("potential",  
                     "wage_eur_log",  
                     "age",  
                     "height_cm",  
                     "weight_kg",  
                     "pace",  
                     "shooting",  
                     "passing",  
                     "dribbling",  
                     "defending",  
                     "physic",  
                     "power_strength",  
                     "power_long_shots")  
  
for (col in numeric_columns){  
  title = paste("Normal Q-Q Plot of the column", col)  
  qqnorm(df[,col], main=title)  
  qqline(df[,col], col="steelblue", lwd=2)  
}
```



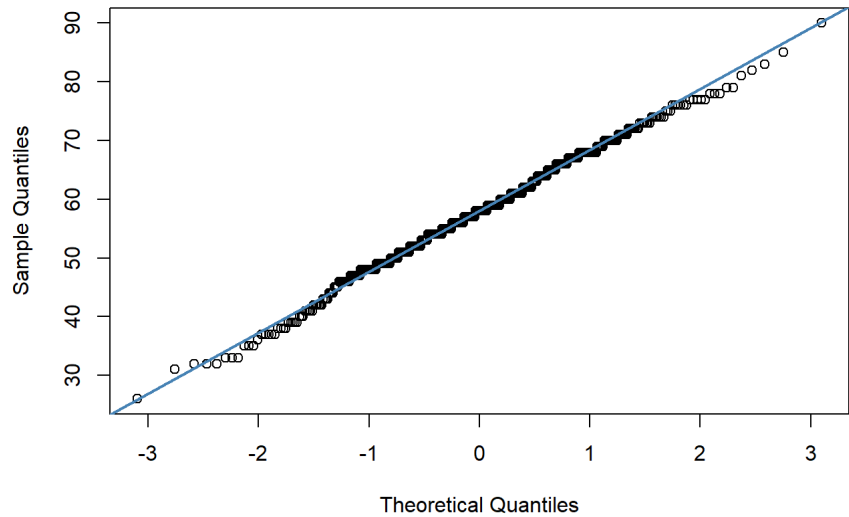




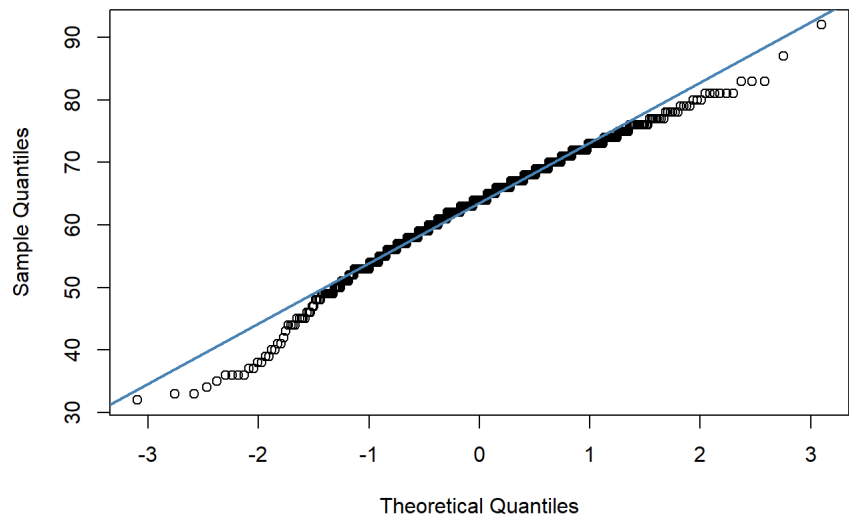
Normal Q-Q Plot of the column shooting



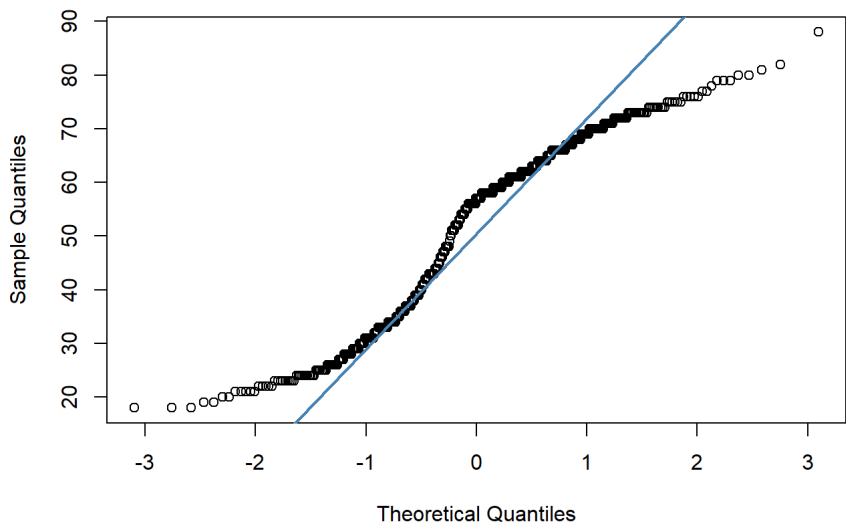
Normal Q-Q Plot of the column passing



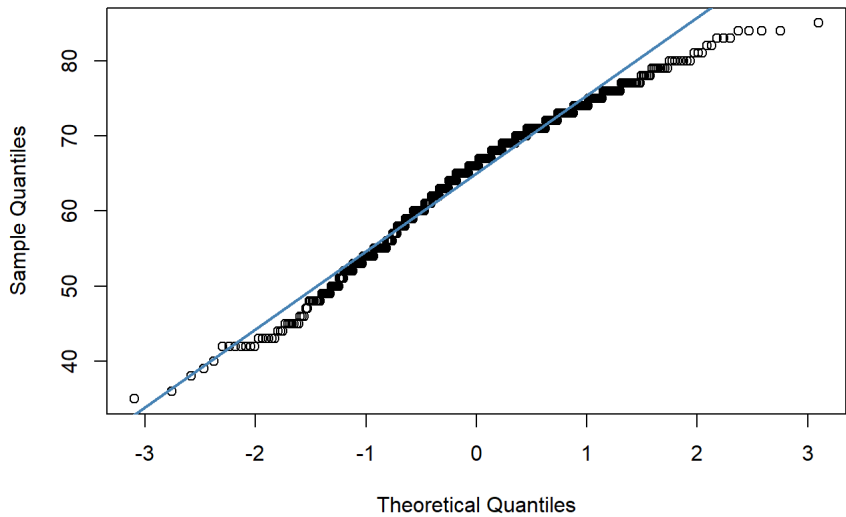
Normal Q-Q Plot of the column dribbling



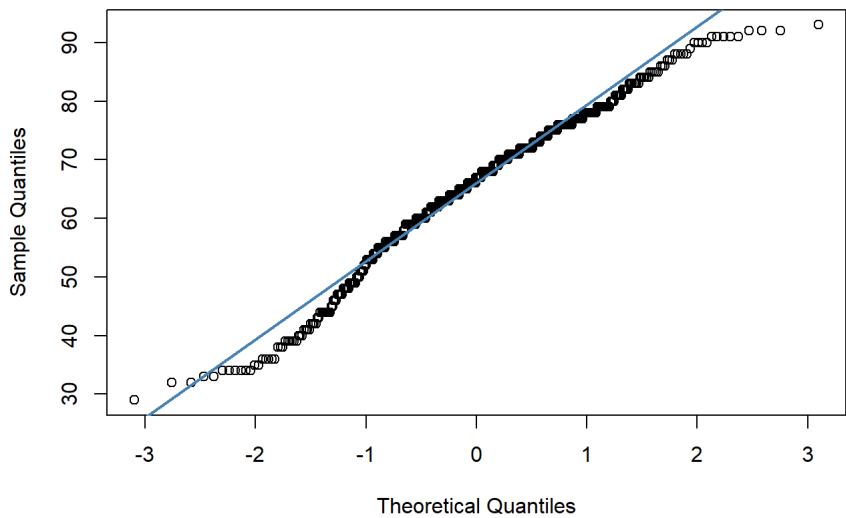
Normal Q-Q Plot of the column defending

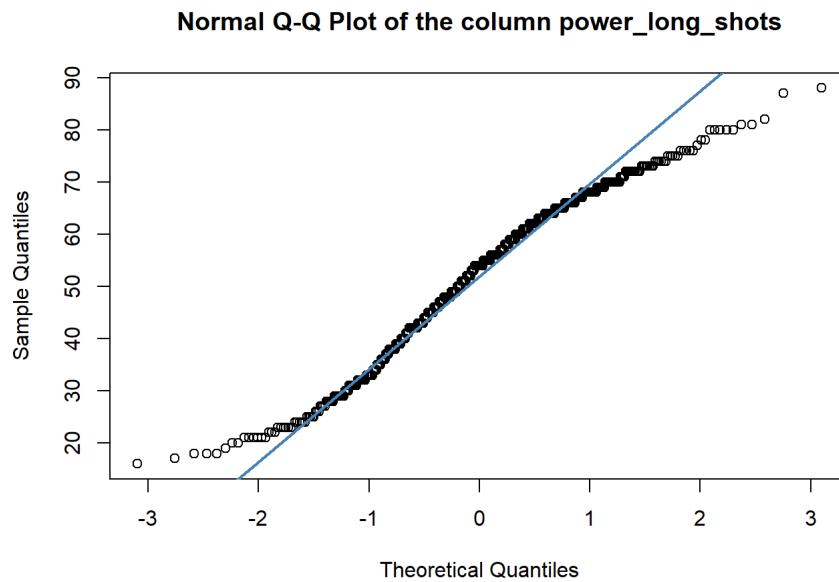


Normal Q-Q Plot of the column physic



Normal Q-Q Plot of the column power\_strength





```
for (col in numeric_columns){
  message = paste("Shapiro-Wilk normality test for the column", col)
  print(message)
  shapiro_test <- shapiro.test(df[,col])
  print(shapiro_test[2])
  if (shapiro_test[2] < 0.05){
    shap_message <- paste("This is a very small p-value, we reject the null hypothesis that", col, "in this sample is normal
ly distributed")
    print(shap_message)
  }
  else{
    print(paste("We failed to reject the null hypothesis. Column", col, "is normally distributed"))
  }
  cat("-----\n\n\n")
}
```

```

## [1] "Shapiro-Wilk normality test for the column potential"
## $p.value
## [1] 0.006241612
##
## [1] "This is a very small p-value, we reject the null hypothesis that potential in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column wage_eur_log"
## $p.value
## [1] 0.0000000002264849
##
## [1] "This is a very small p-value, we reject the null hypothesis that wage_eur_log in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column age"
## $p.value
## [1] 0.0000002278677
##
## [1] "This is a very small p-value, we reject the null hypothesis that age in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column height_cm"
## $p.value
## [1] 0.09240383
##
## [1] "We failed to reject the null hypothesis. Column height_cm is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column weight_kg"
## $p.value
## [1] 0.02663471
##
## [1] "This is a very small p-value, we reject the null hypothesis that weight_kg in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column pace"
## $p.value
## [1] 0.000004189181
##
## [1] "This is a very small p-value, we reject the null hypothesis that pace in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column shooting"
## $p.value
## [1] 0.000001774451
##
## [1] "This is a very small p-value, we reject the null hypothesis that shooting in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column passing"
## $p.value
## [1] 0.3160711
##
## [1] "We failed to reject the null hypothesis. Column passing is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column dribbling"
## $p.value
## [1] 0.000006314925
##
## [1] "This is a very small p-value, we reject the null hypothesis that dribbling in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column defending"
## $p.value
## [1] 0.00000000001639449
##
## [1] "This is a very small p-value, we reject the null hypothesis that defending in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column physic"
## $p.value

```

```
## [1] 0.00000003411265
##
## [1] "This is a very small p-value, we reject the null hypothesis that physic in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column power_strength"
## $p.value
## [1] 0.0000006843208
##
## [1] "This is a very small p-value, we reject the null hypothesis that power_strength in this sample is normally distributed"
## -----
##
##
## [1] "Shapiro-Wilk normality test for the column power_long_shots"
## $p.value
## [1] 0.0000006352206
##
## [1] "This is a very small p-value, we reject the null hypothesis that power_long_shots in this sample is normally distributed"
## -----
```

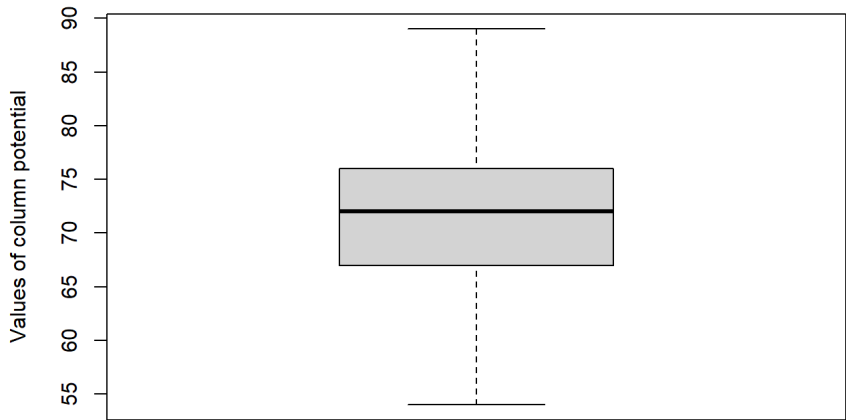
After graphing the qq-plots and performing shapiro tests, we can conclude the following:

- Columns `power_long_shots`, `power_strength`, `physic`, `physic`, `defending`, `dribbling`, `shooting`, `pace`, `weight_kg`, `age`, `wage_eur_log`, `potential` are not normally distributed.
- Columns `passing`, `height_cm` are normally distributed.

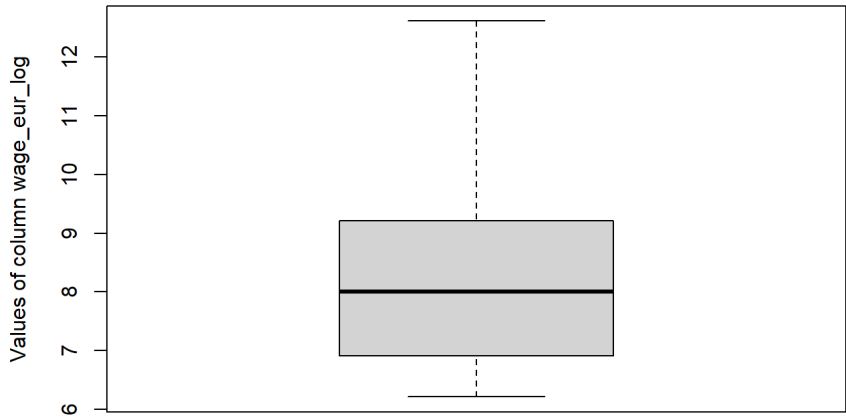
Boxplots:

```
for (col in numeric_columns){
  title = paste("Boxplot for column", col)
  y_axis = paste("Values of column", col)
  boxplot(df[, col], main=title, ylab=y_axis)
}
```

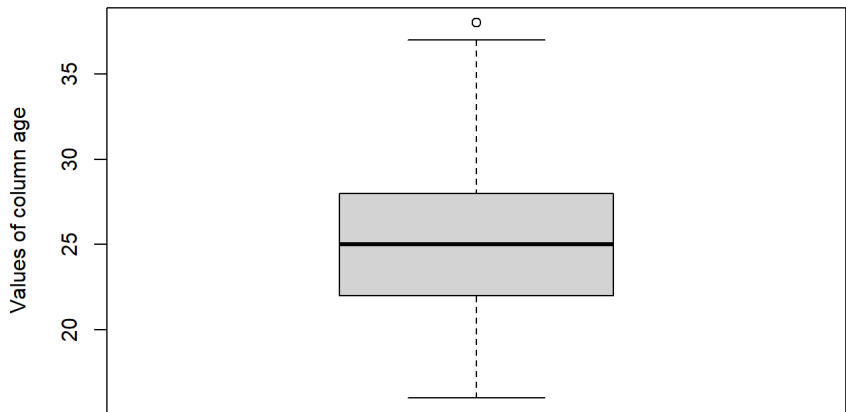
Boxplot for column potential

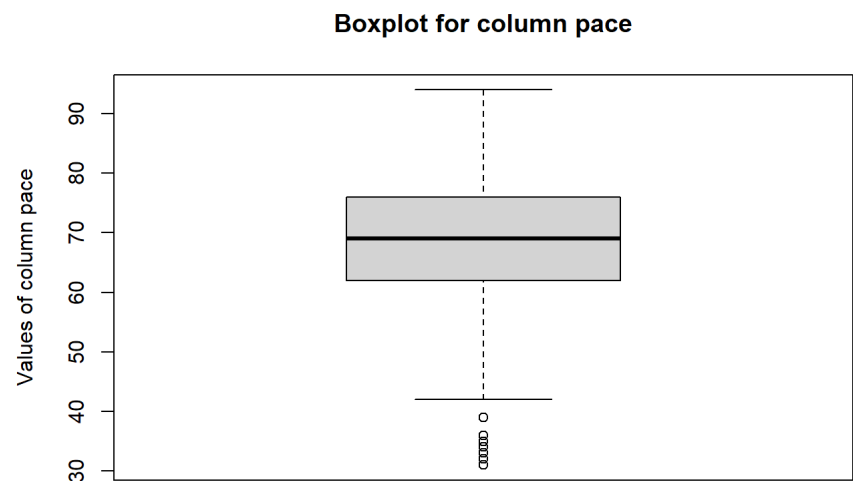
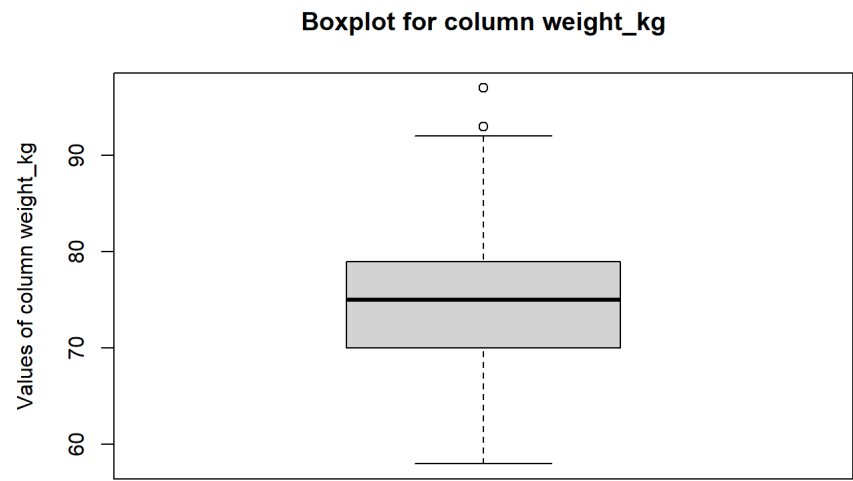
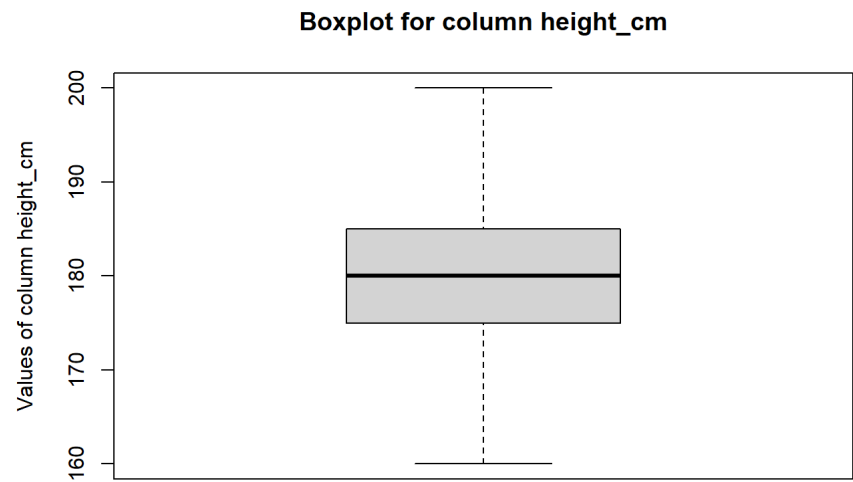


Boxplot for column wage\_eur\_log



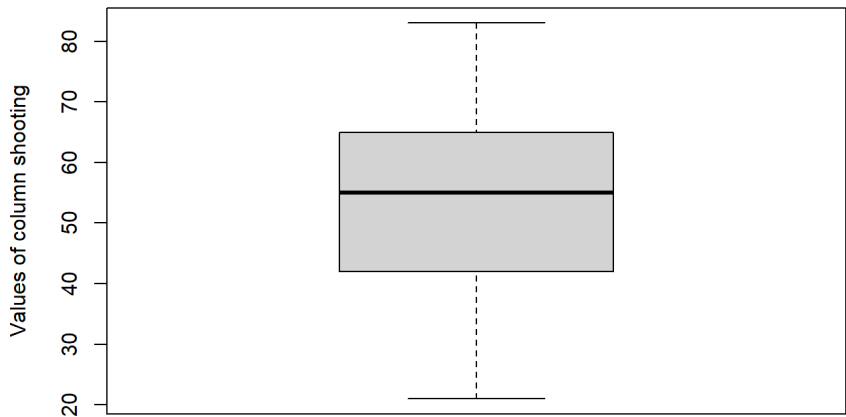
Boxplot for column age



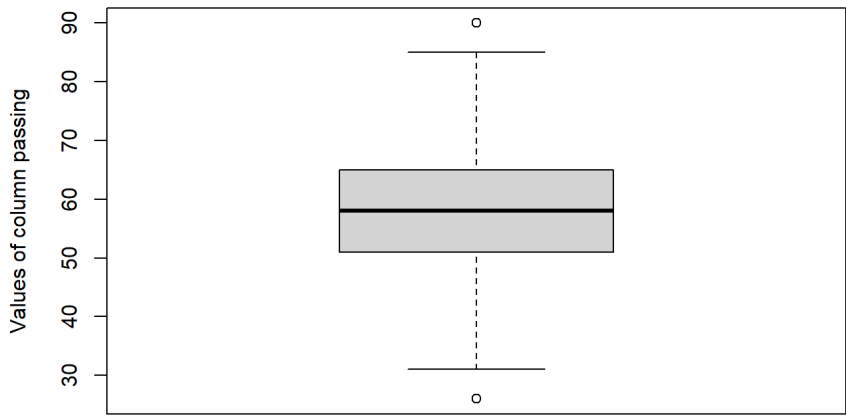




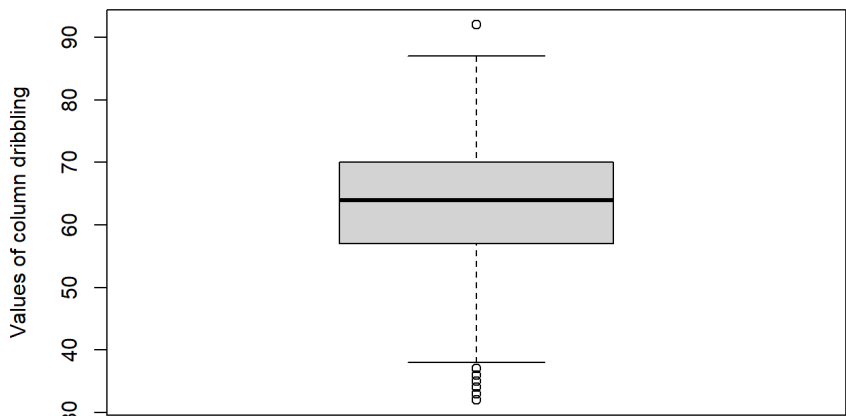
Boxplot for column shooting



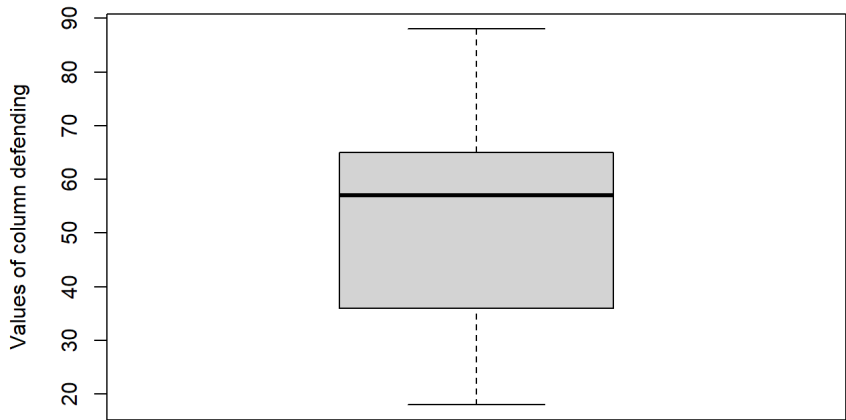
Boxplot for column passing



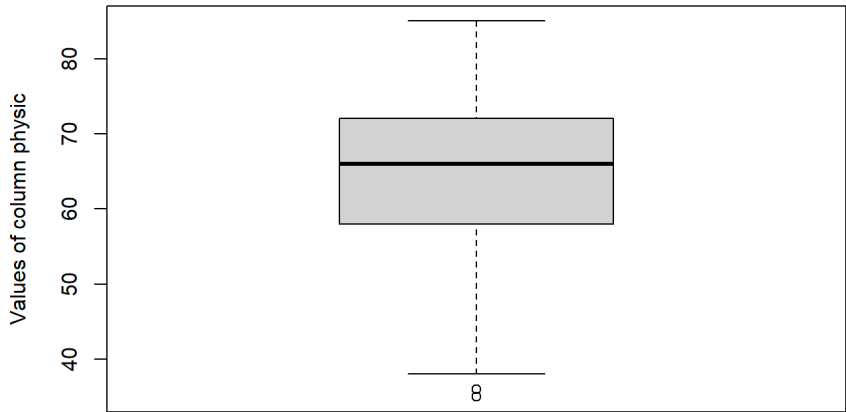
Boxplot for column dribbling



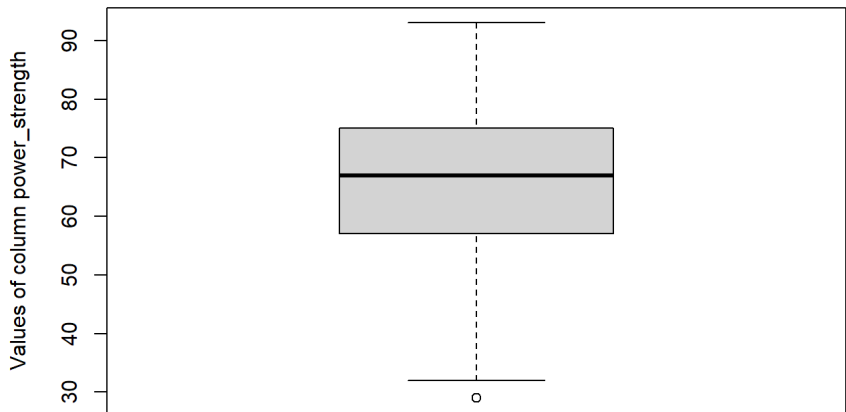
Boxplot for column defending

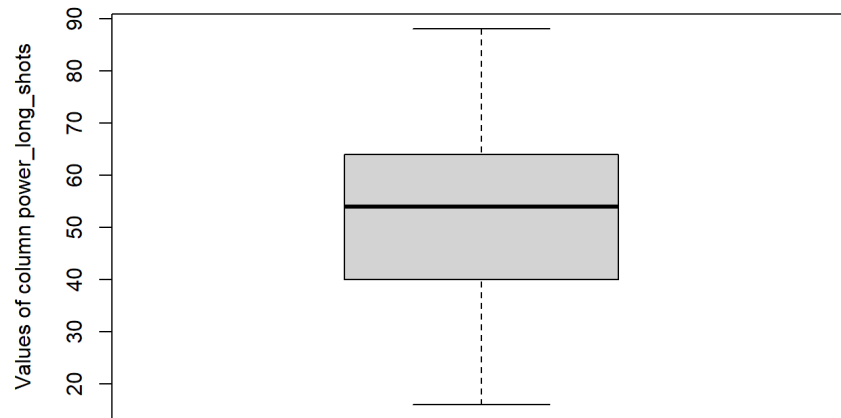


Boxplot for column physic



Boxplot for column power\_strength



**Boxplot for column power\_long\_shots**

- Even though the boxplots show that some columns do have outliers, those outliers do not represent any trouble when modeling as they seem to land on the range of normal values given by FIFA.

Now, we need to explore the categorical columns. First, we need to create tables.

```
table_foot <- table(df$preferred_foot)
print(table_foot)
```

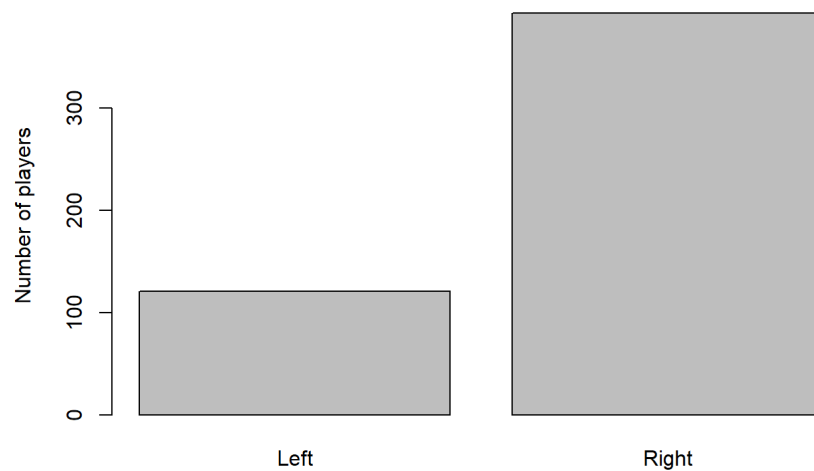
```
##
## Left Right
## 121 393
```

```
table_wage <- table(df$high.wage.ind)
print(table_wage)
```

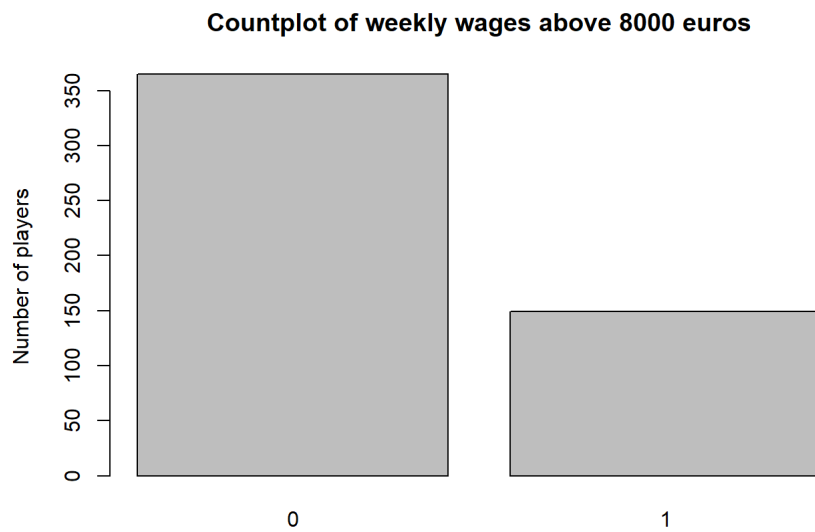
```
##
## 0 1
## 365 149
```

Now, we will create barplots

```
barplot(table_foot,
        names.arg=rownames(table_foot),
        ylab="Number of players",
        main="Countplot of preferred foot for players")
```

**Countplot of preferred foot for players**

```
barplot(table_wage,
        names.arg=rownames(table_wage),
        ylab="Number of players",
        main="Countplot of weekly wages above 8000 euros")
```



- It is clearly shown that both categorical columns are not balanced.

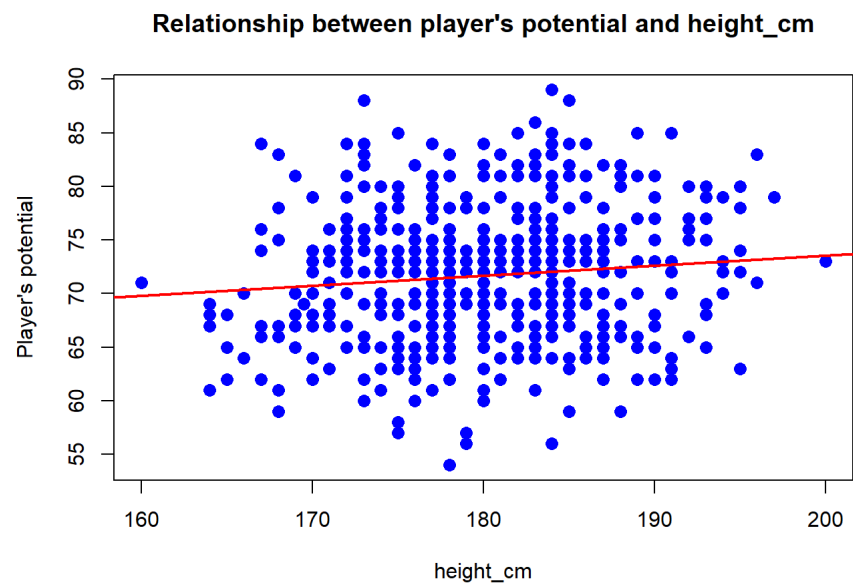
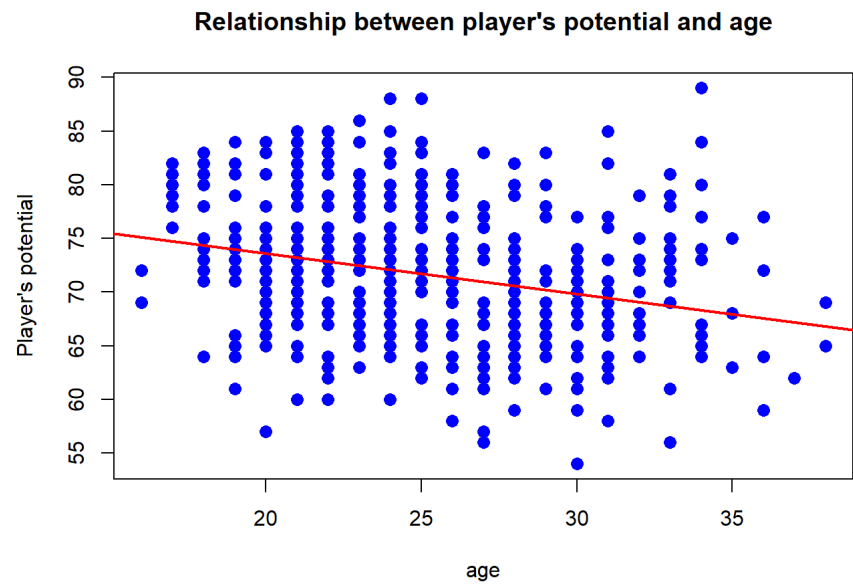
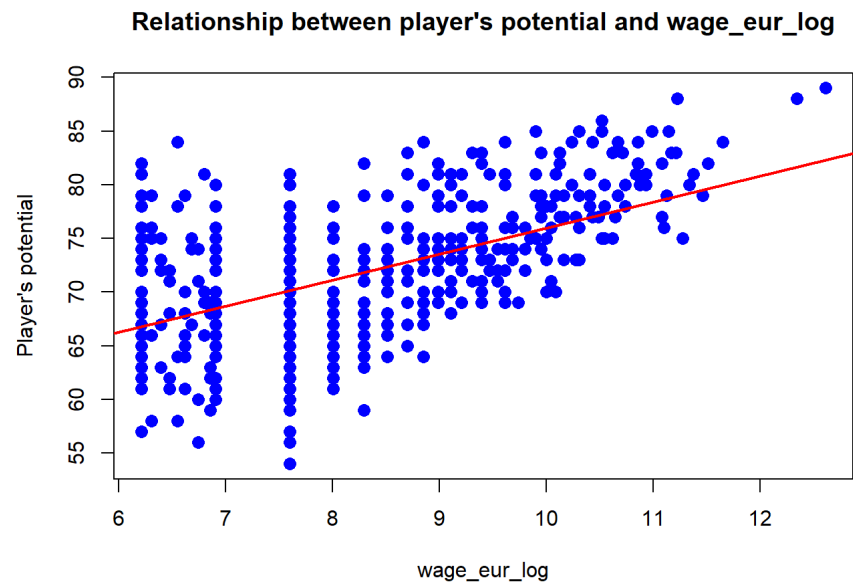
## Bivariate EDA

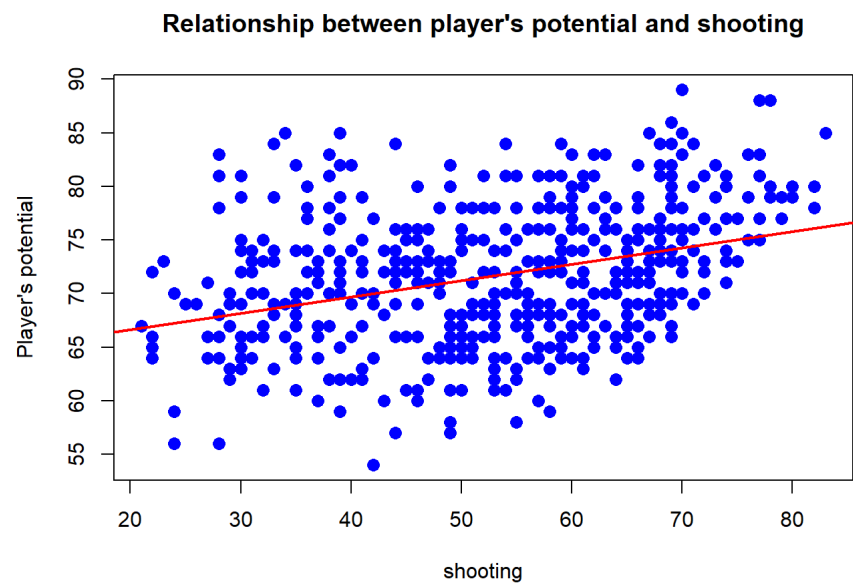
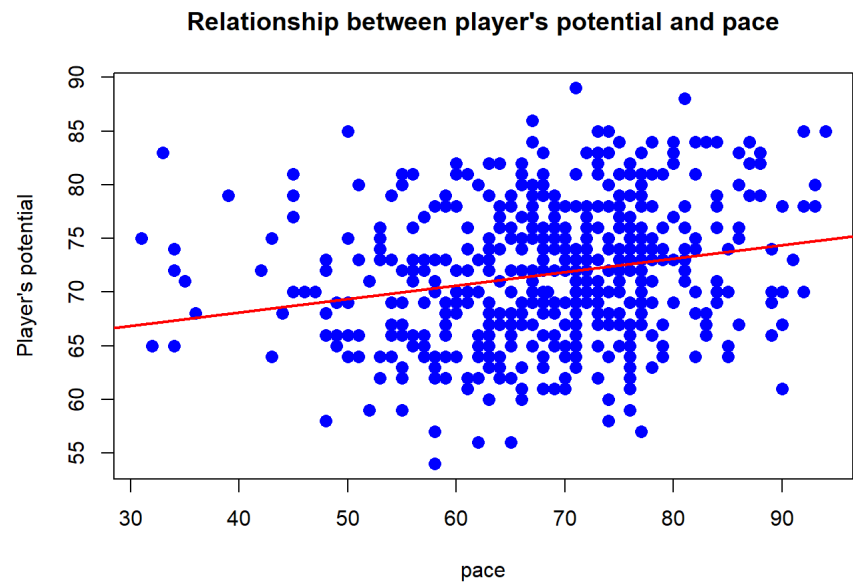
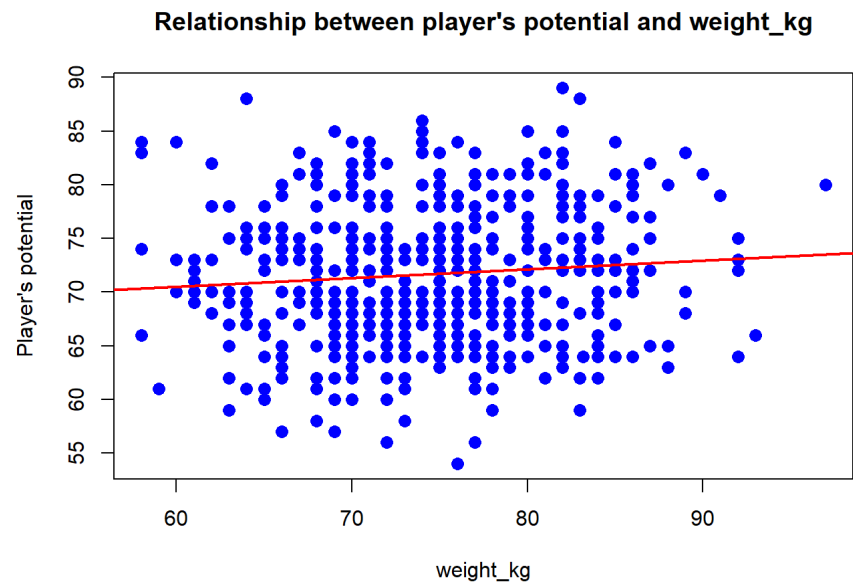
### Regarding column potential

```
numeric_columns <- c("wage_eur_log",
                     "age",
                     "height_cm",
                     "weight_kg",
                     "pace",
                     "shooting",
                     "passing",
                     "dribbling",
                     "defending",
                     "physic",
                     "power_strength",
                     "power_long_shots")

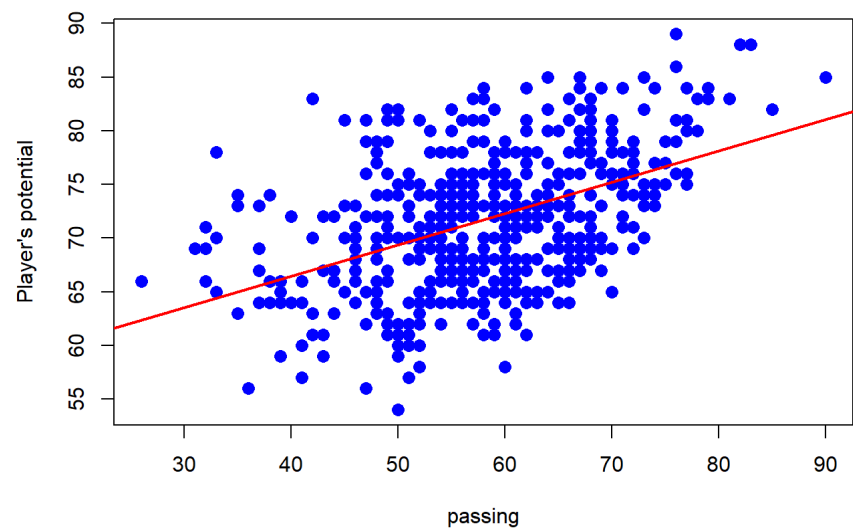
for (col in numeric_columns){
  title = paste("Relationship between player's potential and", col)
  x_axis = col
  y_axis = "Player's potential"
  plot(df[,col],
       df[, "potential"],
       pch=16,
       cex=1.3,
       col="blue",
       ylab=y_axis,
       xlab=x_axis,
       main=title)

  abline(lm(df[, "potential"]~df[,col]),
         lwd=2,
         col="red")
}
```

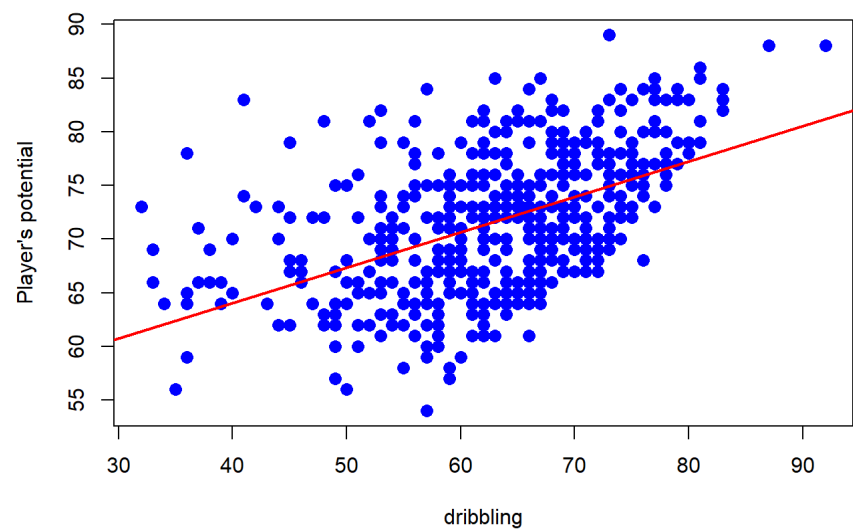




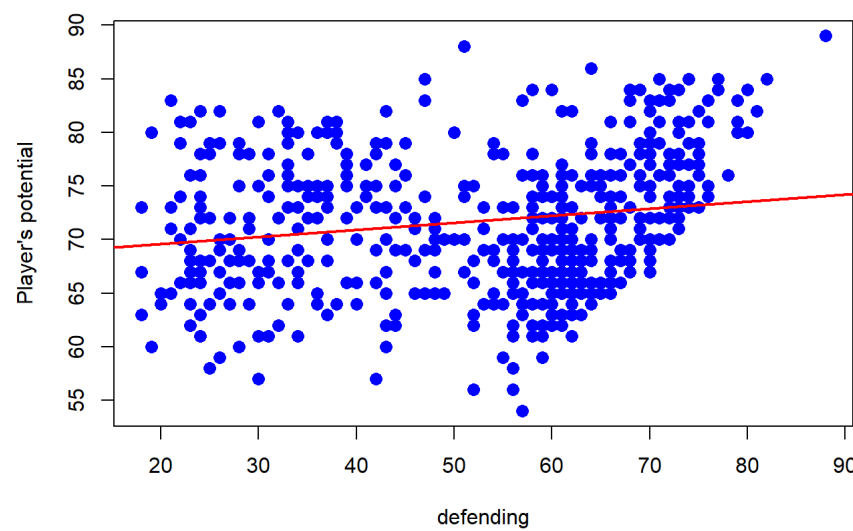
Relationship between player's potential and passing

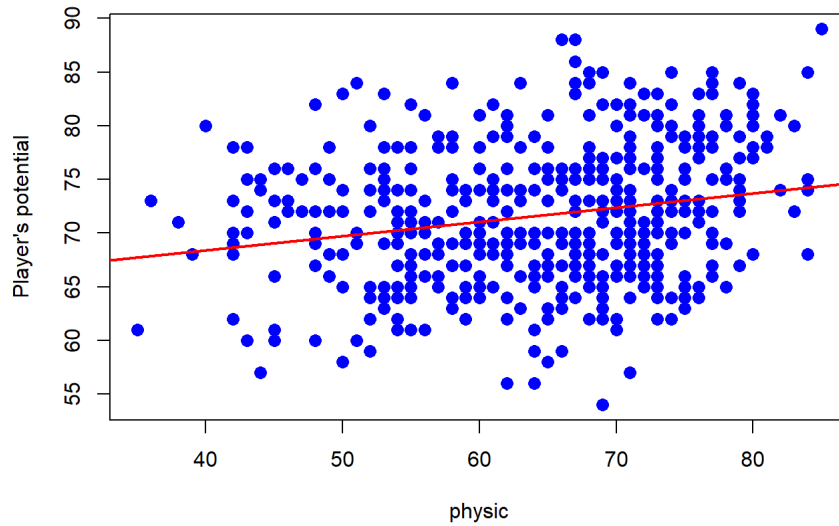
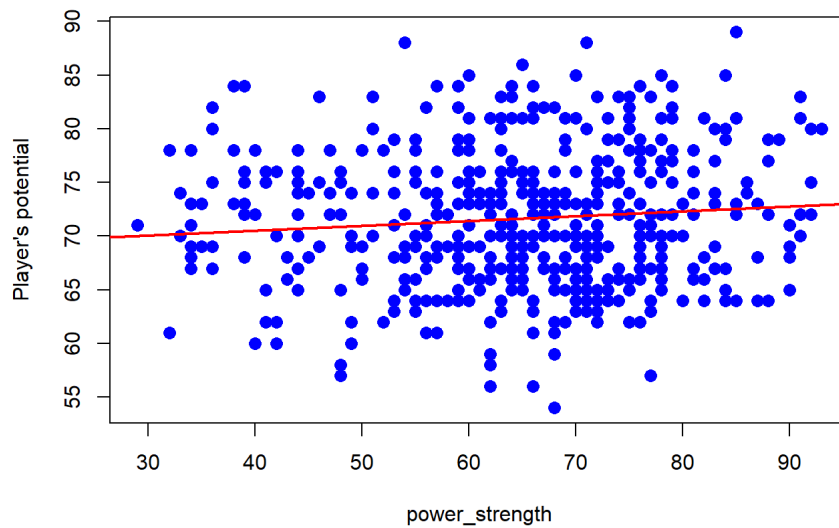
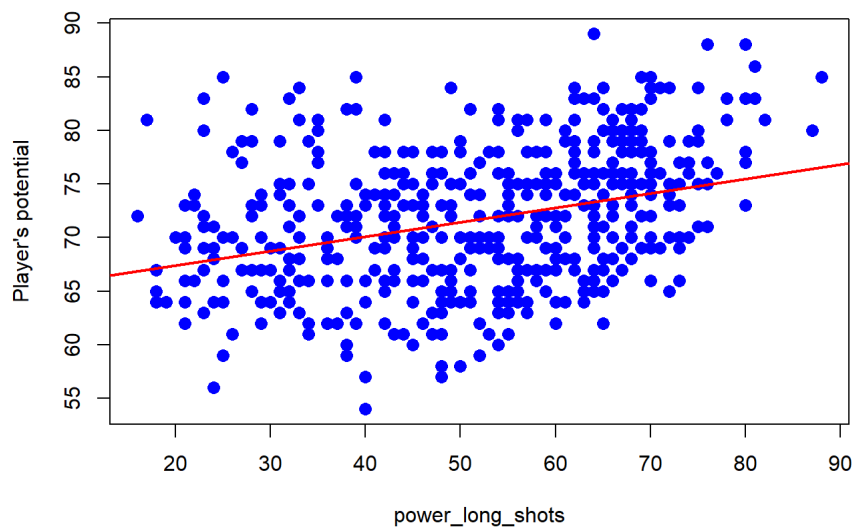


Relationship between player's potential and dribbling



Relationship between player's potential and defending



**Relationship between player's potential and physic****Relationship between player's potential and power\_strength****Relationship between player's potential and power\_long\_shots**

- potential seems to have a positive correlation with wage\_eur\_log .
- potential seems to have a quadratic relationship with age .
- potential seems to have no correlation with height\_cm .
- potential seems to have no correlation with weight\_cm .



- potential seems to have positive correlation with pace .
- potential seems to have positive correlation with shooting .
- potential seems to have positive correlation with passing .
- potential seems to have positive correlation with dribbling .
- potential seems to have a quadratic relationship with defending .
- potential seems to have no correlation with physic .
- potential seems to have no correlation with power\_strength .
- potential seems to have positive correlation with power\_long\_shots .

Making correlation tests:

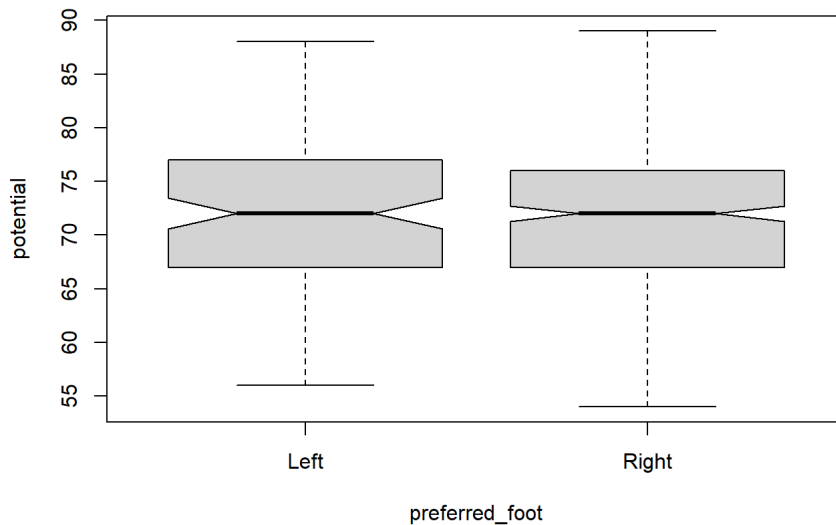
```
for (col in numeric_columns){
  cor_test <- cor.test(df[,col], df[, "potential"], method="spearman", exact=FALSE)
  p_value <- cor_test[3]
  corr_value <- cor_test[4]
  print(p_value)
  if (p_value < 0.05){
    message <- paste("Since, we have a small p-value, we reject the null hypothesis and we say there is a relationship between player's potential and", col, ". The correlation between the 2 variables is", corr_value)
    print(message)
  } else{
    message <- paste("p-value is not small. We failed to reject the null hypothesis. There is no relationship between player's potential and", col)
    print(message)
  }
  cat("-----\n\n")
}
```

[illegible]

Analyzing relationship between potential and preferred\_foot .

```
#Anova
boxplot(potential~preferred_foot,
        data=df,
        notch=TRUE,
        main="Boxplot of player's potential comparing left foot and right foot")
```

**Boxplot of player's potential comparing left foot and right foot**



```
summary(aov(df$potential~df$preferred_foot))
```

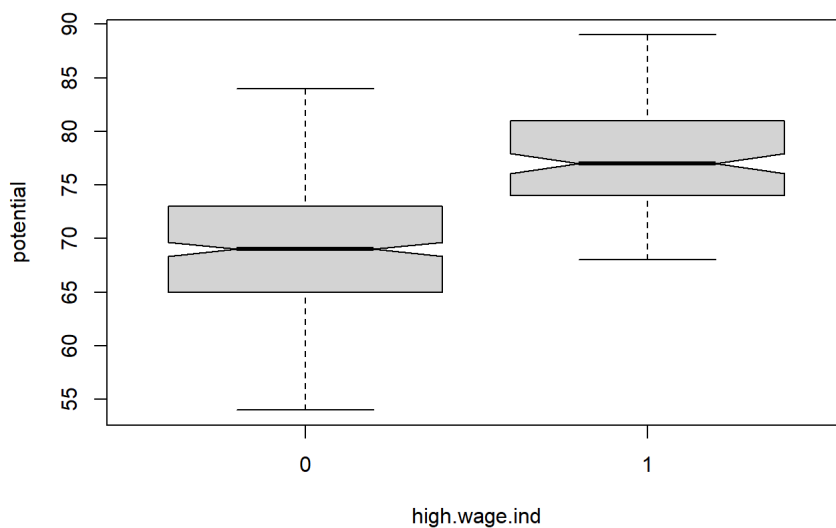
```
##               Df Sum Sq Mean Sq F value Pr(>F)
## df$preferred_foot  1    65    64.91    1.591  0.208
## Residuals       512  20882    40.78
```

- The boxplot shows that there is not statistical significance in the difference of medians. The ANOVA test shows that there is not statistical significance in the difference of means.

Analyzing relationship between potential and high.wage.ind .

```
#Anova
boxplot(potential~high.wage.ind,
        data=df,
        notch=TRUE,
        main="Boxplot of player's potential comparing left foot and right foot")
```

**Boxplot of player's potential comparing left foot and right foot**



```
summary(aov(df$potential~df$high.wage.ind))
```

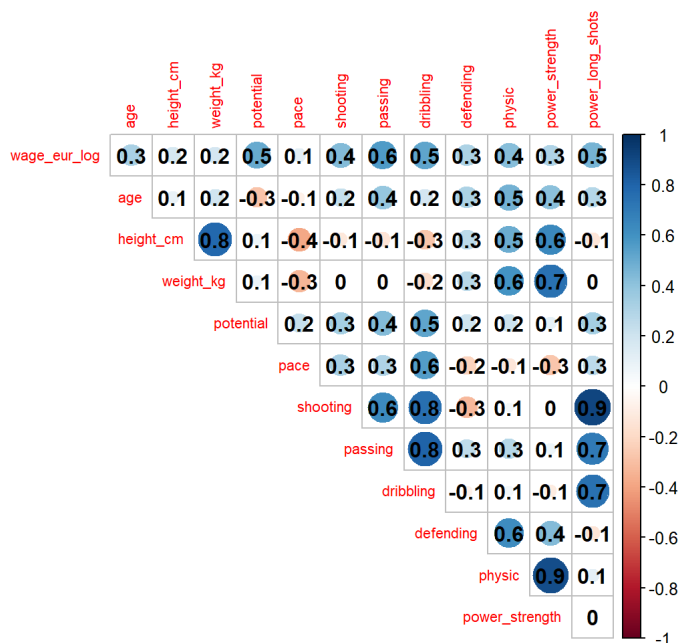
```
##               Df Sum Sq Mean Sq F value           Pr(>F)
## df$high.wage.ind  1   6384     6384   224.4 <0.000000000000002 ***
## Residuals      512  14563         28
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

The boxplot shows that there is statistical significance in the difference of medians. The ANOVA test shows that there is statistical significance in the difference of means.

Finding correlations to find issues of collinearity.

```
c_col <- c("wage_eur_log",
           "age",
           "height_cm",
           "weight_kg",
           "potential",
           "pace",
           "shooting",
           "passing",
           "dribbling",
           "defending",
           "physic",
           "power_strength",
           "power_long_shots")

df_cor <- subset(df, select=c_col)
cor_matrix <- cor(df_cor, method="spearman")
corrplot(cor_matrix, diag=F, type="upper", insig="p-value", number.digits=1, addCoef.col="black", tl.cex=0.7)
```



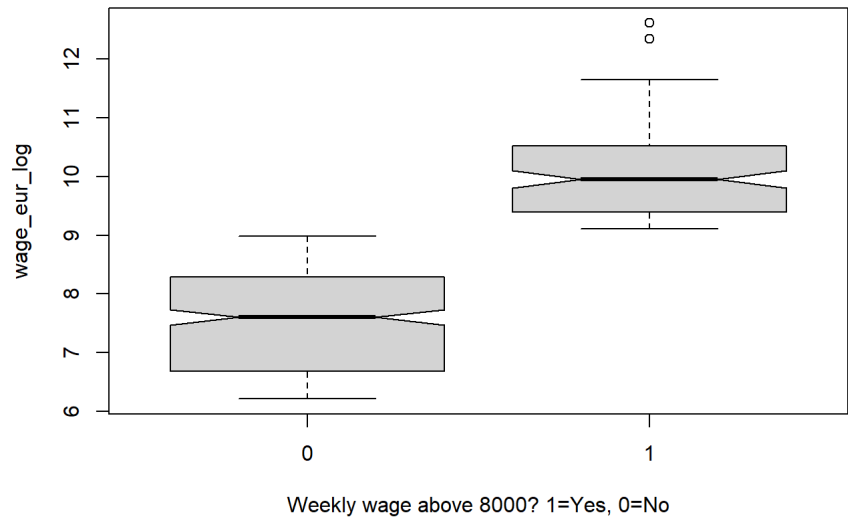
- Issues of collinearity could be found between physic and power\_strength, between dribbling and both shooting and passing, between power\_long\_shots and shooting and between height\_cm and weight\_kg.

Regarding column high.wage.ind

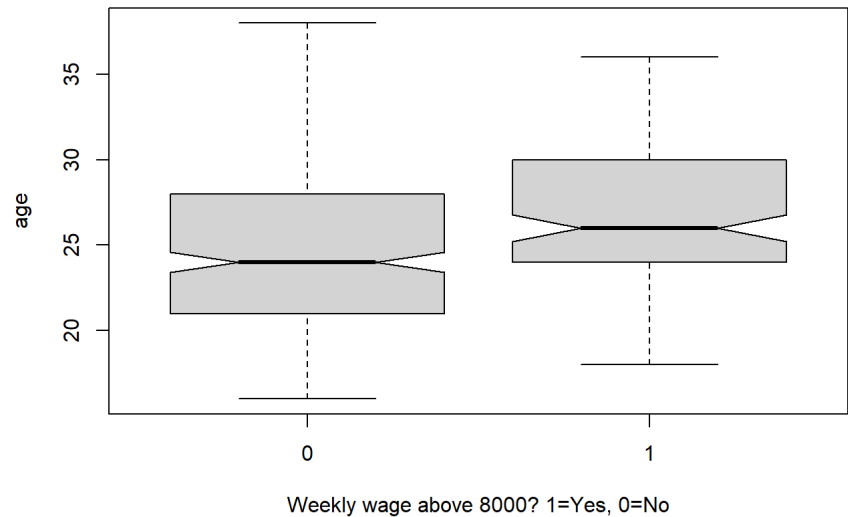
```
c_col <- c("wage_eur_log",
          "age",
          "height_cm",
          "weight_kg",
          "potential",
          "pace",
          "shooting",
          "passing",
          "dribbling",
          "defending",
          "physic",
          "power_strength",
          "power_long_shots")

for (col in c_col){
  title = paste("Boxplot comparing values on", col, "between players with weekly wages below 8000 and above 8000")
  x_axis = "Weekly wage above 8000? 1=Yes, 0=No"
  y_axis = col
  boxplot(df[,col]~df[, "high.wage.ind"],
          data=df,
          notch=TRUE,
          main=title,
          xlab=x_axis,
          ylab=y_axis)
}
```

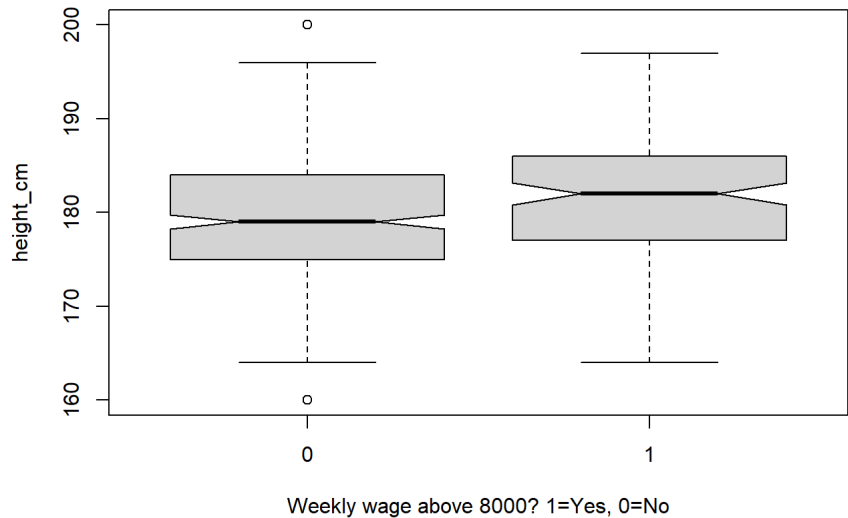
ring values on wage\_eur\_log between players with weekly wages below 8000 ar



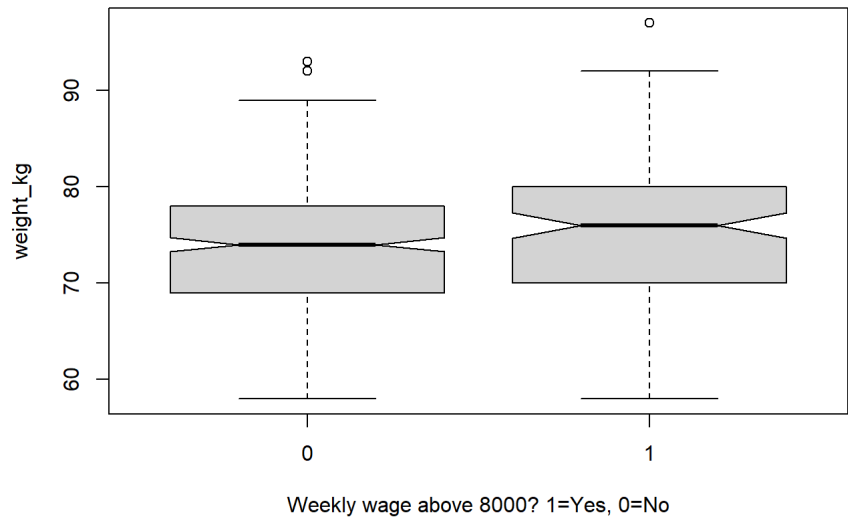
omparing values on age between players with weekly wages below 8000 ar



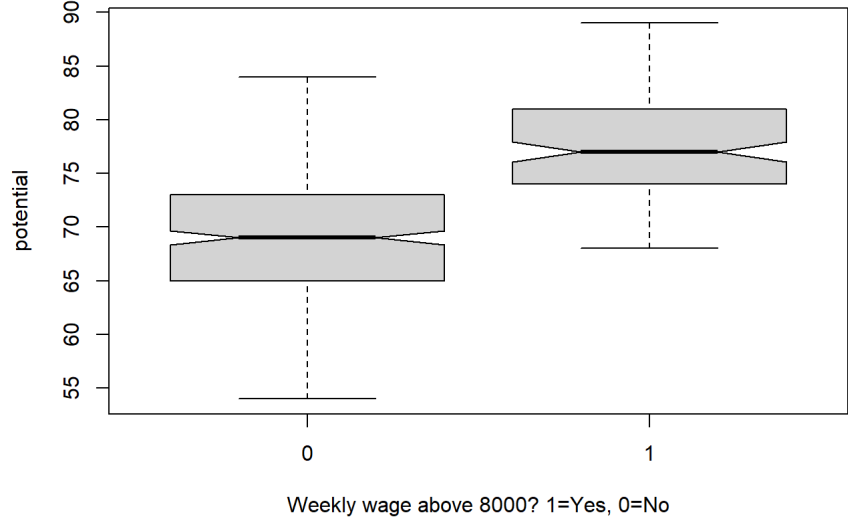
paring values on height\_cm between players with weekly wages below 8000 ar



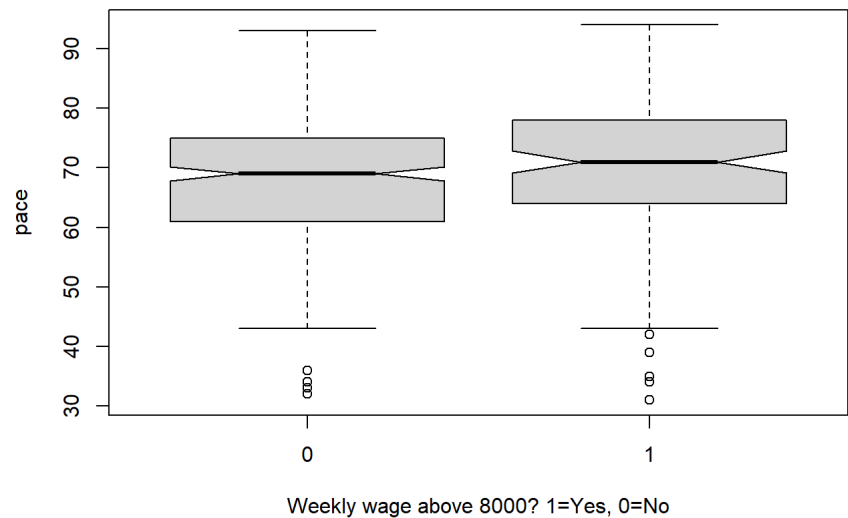
paring values on weight\_kg between players with weekly wages below 8000



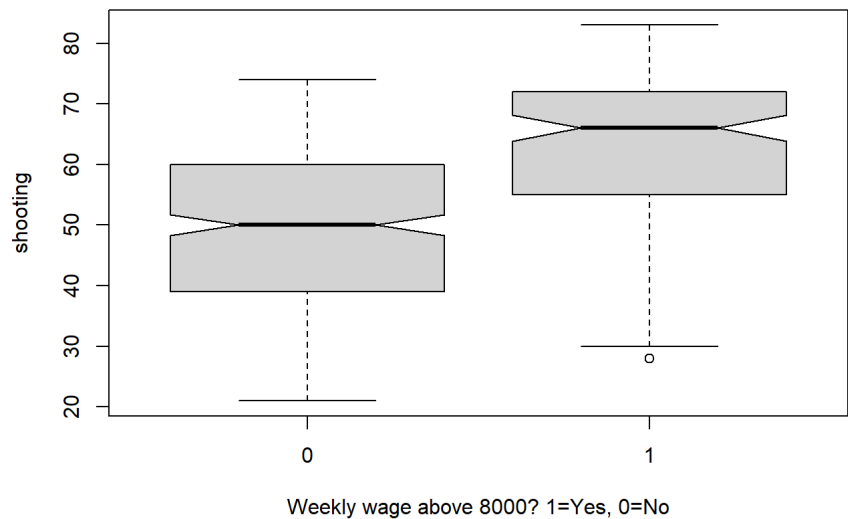
paring values on potential between players with weekly wages below 8000



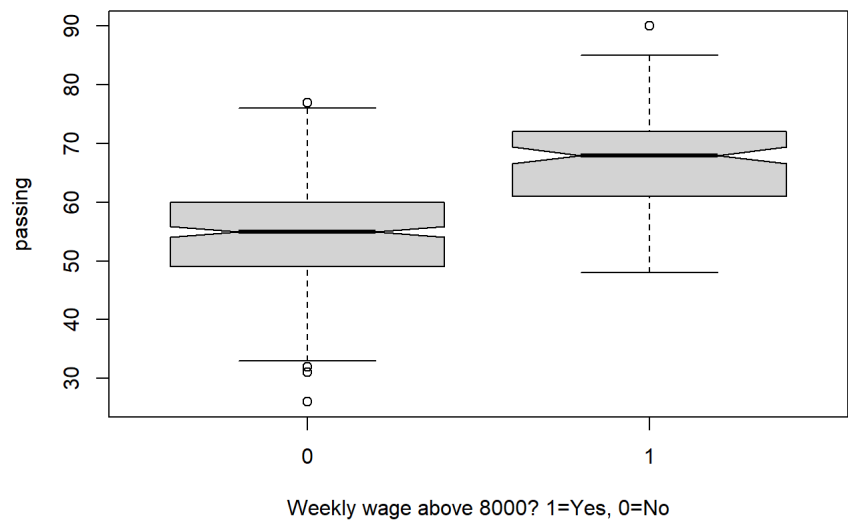
mparing values on pace between players with weekly wages below 8000 a



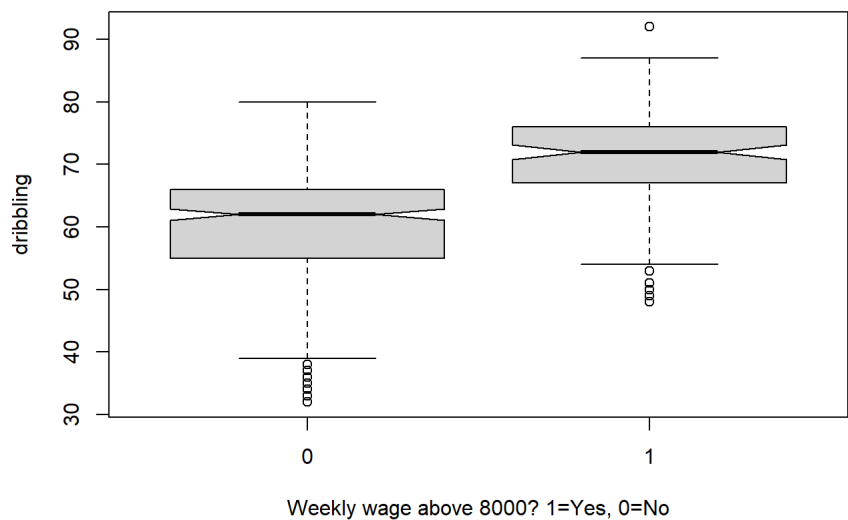
Comparing values on shooting between players with weekly wages below 8000



Comparing values on passing between players with weekly wages below 8000

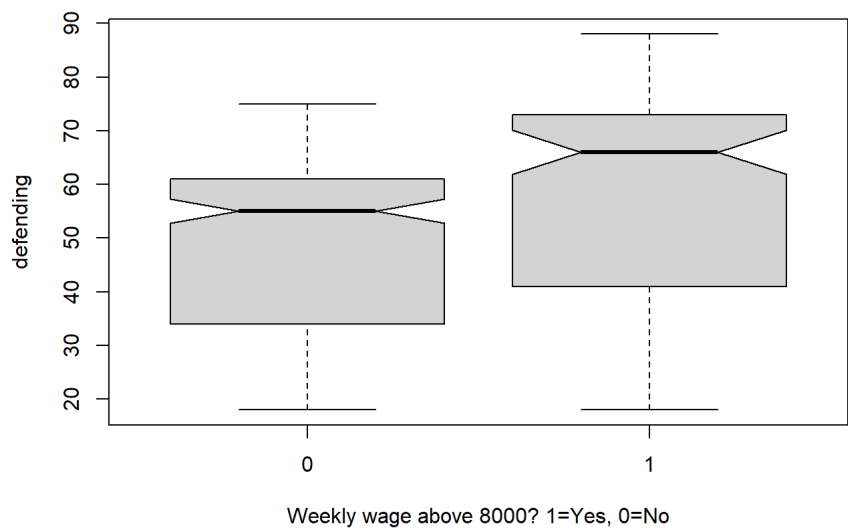


Comparing values on dribbling between players with weekly wages below 8000

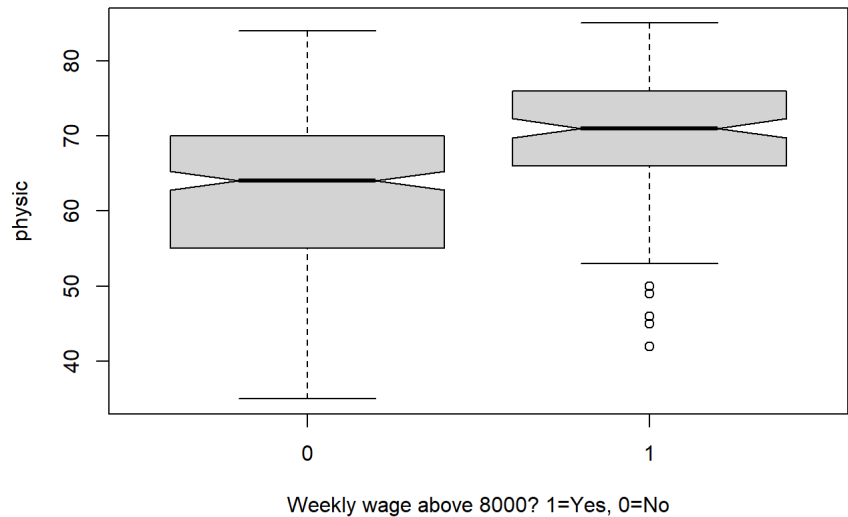




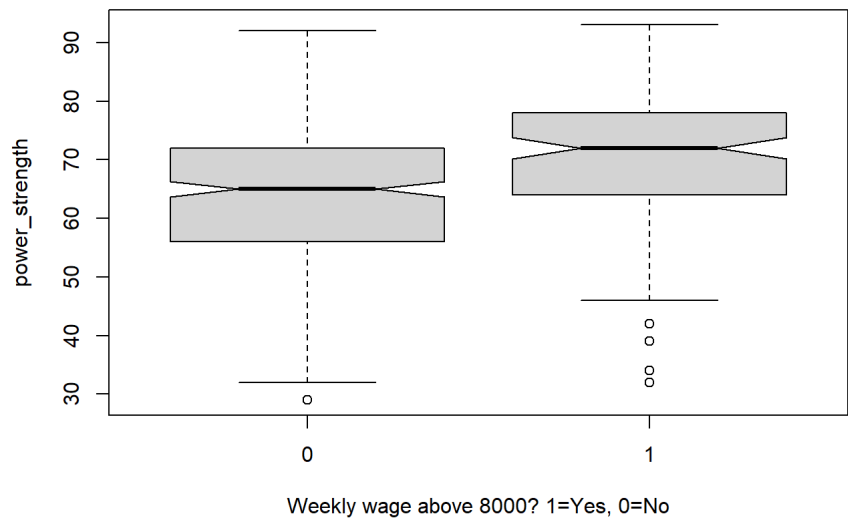
paring values on defending between players with weekly wages below 8000



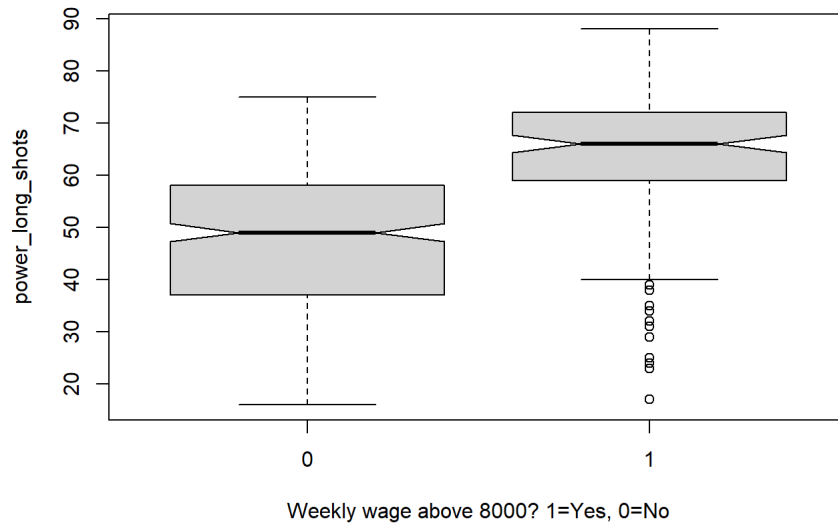
mparing values on physic between players with weekly wages below 8000 :



ring values on power\_strength between players with weekly wages below 8



### ng values on power\_long\_shots between players with weekly wages below



- Boxplots show that there is statistical significance in the difference of medians between `high.wage.ind` and the following columns: `wage_eur_log` (more than obvious since `high.wage.ind` is calculated from `wage_eur`), `age` (small), `height_cm` (barely), `potential`, `shooting`, `passing`, `dribbling`, `defending`, `physic`, `power_strength` and `power_long_shots`. Boxplots show that there is no statistical significance in the difference in medians between `high.wage.ind` and the following columns: `weight_kg` and `pace`.

Performing two sided t-test to find if there is significance difference in means between `high.wage.ind` and the other numerical columns.

```
for (col in c_col){
  message <- paste("Results for column", col)
  print(message)
  ttest_result <- t.test(df[,col]~df[, "high.wage.ind"], alternative="two.sided")
  print(ttest_result[5])
  print(ttest_result[3])
  if (ttest_result[3]<0.05){
    message_1 <- "P-value is very small, we reject null-hypothesis and we say there is statistical significance in the difference of means."
    print(message_1)
  }
  else {
    message_2 <- "P-value is big. We failed to reject the null-hypothesis. There is no evidence that there is statistical significance in the difference of means."
    print(message_2)
  }
  cat("-----\n\n")
}
```

[illegible]

[illegible]

```
eans."
```

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

Columns `wage_eur_log`, `age`, `height_cm`, `weight_cm`, `potential`, `shooting`, `passing`, `dribbling`, `defending`, `physic`, `power_strength` and `power_long_shots` have shown to have statistical significance in their difference of means between players that have weekly payments less than 8000 and above 8000. Column `pace` has shows to not have statistical significance in the difference of means.

Looking for a relationship between `preferred_foot` and `high.wage.ind`:

```
table_wage_foot <- table(df$high.wage.ind, df$preferred_foot)
table_wage_foot
```

```
##
##      Left Right
##    0   85   280
##    1   36   113
```

```
chisq.test(table_wage_foot)
```

```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data:  table_wage_foot
## X-squared = 0.0094454, df = 1, p-value = 0.9226
```

P-value is 0.9226, therefore failed to reject the null hypothesis and we say there is no relationship between `preferred_foot` and `high.wage.ind`.

## 2.3 Additional insights and issues

*Highlight potential further issues or insights uncovered in 2.2. This might include follow up to findings from your initial EDA. We accept that the boundary between 2.2 and 2.3 is somewhat arbitrary so use your judgement and maximize good structure and readability. (5 marks)*

No further issues were found during the EDA, all issues were found and fixed during the data cleaning.

## 3. Modelling

### 3.1 Build a model for player potential

*Given the research question (i.e., player potential) outline an analysis plan that incorporates/references any findings from the data cleaning (1.3) and EDA (2.2) (5 marks). Use R to build a suitable model (10 marks).*

*NB Submissions where suitable models do not have good fit due to the nature of the data will not be penalized.*

- From the EDA process, I have found that column `wage_eur` follows a logarithmic distribution. Therefore, a logarithmic transformation was applied to make the column more suitable for usage in the model. The transformation was stored in the `wage_eur_log` column. [6]
- From the EDA process, I have found that columns `power_long_shots`, `power_strength`, `physic`, `defending`, `dribbling`, `shooting`, `pace`, `weight_kg`, `wage_eur_log`, `age`, `potential` do not follow a normal distribution. Only columns `height_cm` and `passing` follow a normal distribution. Therefore, when trying to find correlation between the columns, the Spearman method was used.
- In order to avoid collinearity, features that are highly correlated ( $\geq 0.8$ ) between each other, will not be used. `height_cm` and `weight_kg` are highly correlated, for this case `height_cm` will be used because it follows a normal distribution. `dribbling` is highly correlated with `shooting` and `passing`, for this case we will use `dribbling` and remove `shooting` and `passing` in the model. In this way, we use one feature instead of two and make the model simpler, and also because `dribbling` has more correlation with `potential` than the other 2 features. `physic` and `power_strength` are also highly correlated, I will use `physic` and not use `power_strength` because `physic` has a higher correlation with `potential` than `power_strength` with `potential`. In summary, columns `weight_kg`, `shooting`, `passing`, `power_strength` will not be used in the model.
- Using ANOVA and boxplots with notches, we have found that there is not statistical significance in median and mean of player's potential between players who preferred right foot and left foot. Therefore, the column `preferred_foot` will not be used in the model. On the other hand, it was found that there was a statistical significance in median and mean of player's potential between players whose weekly wages are above 8000 euros and those below 8000 euros. However, the column `high.wage.ind` is taken from the `wage_eur` column which does not bring extra information, it only confirms that `wage_eur_log` is a good feature for our model. If `high.wage.ind` is used, an interaction problem with `wage_eur_log` will surged.
- `age` and `defending` may have a quadratic relationships on potential as seen as on the linear regressions performed on the EDA.
- First I will build a complex model and take note of the significant coefficients. Then I will build a very simple model and start adding significant coefficients found in the complex model.

**model0**

```
mr_model0 <- lm(potential ~ wage_eur_log*age*height_cm*pace*dribbling*defending*physic*power_long_shots +
                I(wage_eur_log^2)+
                I(age^2) +
                I(height_cm^2) +
                I(pace^2) +
                I(dribbling^2) +
                I(defending^2) +
                I(physic^2) +
                I(power_long_shots^2), data=df)
summary(mr_model0)
```

```
##
## Call:
## lm(formula = potential ~ wage_eur_log * age * height_cm * pace *
##   dribbling * defending * physic * power_long_shots + I(wage_eur_log^2) +
##   I(age^2) + I(height_cm^2) + I(pace^2) + I(dribbling^2) +
##   I(defending^2) + I(physic^2) + I(power_long_shots^2), data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.8602 -0.9462 -0.0291  0.9572  6.1651
##
## Coefficients: (1 not defined because of singularities)
##                                     Estimate
## (Intercept)                      427405.034363127954
## wage_eur_log                     -40301.401010620160
## age                               32839.128403441748
## height_cm                         -2067.053106715392
## pace                             2233.809102541663
## dribbling                       -14884.233093490198
## defending                          -7296.221614994129
## physic                           -8818.663226317689
## power_long_shots                 -25832.658239883862
## I(wage_eur_log^2)                  0.690066755844
## I(age^2)                           0.074923577237
## I(height_cm^2)                     0.000072019611
## I(pace^2)                          0.002342111916
## I(dribbling^2)                     0.009574794368
## I(defending^2)                     0.005052403570
## I(physic^2)                        0.002876070656
## I(power_long_shots^2)              0.000385620411
## wage_eur_log:age                   -4643.973125183560
## wage_eur_log:height_cm             164.469514853294
## age:height_cm                     -190.875866432536
## wage_eur_log:pace                  -251.577958288820
## age:pace                           -773.111244189751
## height_cm:pace                     -18.408316336233
## wage_eur_log:dribbling             1837.433287764628
## age:dribbling                      -192.306002859222
## height_cm:dribbling                78.751112293509
## pace:dribbling                     68.925695086549
## wage_eur_log:defending             905.845026878650
## age:defending                      -640.004713376964
## height_cm:defending                38.536618968758
## pace:defending                     -69.832757523856
## dribbling:defending                287.583337501923
## wage_eur_log:physic                816.443769894335
## age:physic                         -399.005091641252
## height_cm:physic                   42.122136683417
## pace:physic                        -7.376064034482
## dribbling:physic                   252.834595842318
## defending:physic                     131.167522657696
## wage_eur_log:power_long_shots      2725.573886186652
## age:power_long_shots                196.587233110556
## height_cm:power_long_shots          141.765502009691
## pace:power_long_shots               217.153729202686
## dribbling:power_long_shots          535.555278311568
## defending:power_long_shots            418.840459733903
## physic:power_long_shots             435.949900754030
## wage_eur_log:age:height_cm          27.670895722173
## wage_eur_log:age:pace                95.748419523381
## wage_eur_log:height_cm:pace         2.473345916236
## age:height_cm:pace                  4.489834544539
## wage_eur_log:age:dribbling           24.138134960707
## wage_eur_log:height_cm:dribbling     -9.391899196477
## age:height_cm:dribbling              1.147638897488
## wage_eur_log:pace:dribbling          -11.560601876349
## age:pace:dribbling                   7.982890090867
## height_cm:pace:dribbling             -0.298437568790
## wage_eur_log:age:defending           82.886588920742
## wage_eur_log:height_cm:defending     -4.430593494107
## age:height_cm:defending              3.533507513222
## wage_eur_log:pace:defending          3.719355543005
## age:pace:defending                  15.559338855313
## height_cm:pace:defending             0.437914808316
## wage_eur_log:dribbling:defending     -37.557164060069
## age:dribbling:defending              3.029865001191
## height_cm:dribbling:defending        -1.573908641575
## pace:dribbling:defending             -1.064875329101
## wage_eur_log:age:physic              60.370931720731
## wage_eur_log:height_cm:physic        -3.373219686841
## age:height_cm:physic                 2.436804292044
## wage_eur_log:pace:physic             1.455268190336
## age:pace:physic                     10.465294651960
```

## height_cm:pace:physic	0.165325966696
## wage_eur_log:dribbling:physic	-30.424028890896
## age:dribbling:physic	2.104039576387
## height_cm:dribbling:physic	-1.307304495445
## pace:dribbling:physic	-1.371776214836
## wage_eur_log:defending:physic	-14.814773499251
## age:defending:physic	8.737357573090
## height_cm:defending:physic	-0.664239090650
## pace:defending:physic	0.928964529540
## dribbling:defending:physic	-4.490455155564
## wage_eur_log:age:power_long_shots	-1.969179879260
## wage_eur_log:height_cm:power_long_shots	-14.636278360019
## age:height_cm:power_long_shots	-1.044826669618
## wage_eur_log:pace:power_long_shots	-24.259833937985
## age:pace:power_long_shots	2.746395431753
## height_cm:pace:power_long_shots	-1.167692774238
## wage_eur_log:dribbling:power_long_shots	-61.529809179846
## age:dribbling:power_long_shots	-8.259402359958
## height_cm:dribbling:power_long_shots	-2.956533479249
## pace:dribbling:power_long_shots	-5.039526955630
## wage_eur_log:defending:power_long_shots	-46.434719022420
## age:defending:power_long_shots	-1.480816794065
## height_cm:defending:power_long_shots	-2.357297670236
## pace:defending:power_long_shots	-2.925917162302
## dribbling:defending:power_long_shots	-9.541379712589
## wage_eur_log:physic:power_long_shots	-44.683526963194
## age:physic:power_long_shots	-5.065746000559
## height_cm:physic:power_long_shots	-2.361370506880
## pace:physic:power_long_shots	-3.674735397104
## dribbling:physic:power_long_shots	-8.691655623448
## defending:physic:power_long_shots	-6.755120442155
## wage_eur_log:age:height_cm:pace	-0.569396854116
## wage_eur_log:age:height_cm:dribbling	-0.156593029568
## wage_eur_log:age:pace:dribbling	-0.869442651129
## wage_eur_log:height_cm:pace:dribbling	0.048709290732
## age:height_cm:pace:dribbling	-0.046764250088
## wage_eur_log:age:height_cm:defending	-0.471419600369
## wage_eur_log:age:pace:defending	-1.788464253776
## wage_eur_log:height_cm:pace:defending	-0.032253231126
## age:height_cm:pace:defending	-0.087109872400
## wage_eur_log:age:dribbling:defending	-0.341385760460
## wage_eur_log:height_cm:dribbling:defending	0.200097748995
## age:height_cm:dribbling:defending	-0.016003839951
## wage_eur_log:pace:dribbling:defending	0.235256399100
## age:pace:dribbling:defending	-0.156033764796
## height_cm:pace:dribbling:defending	0.005187886891
## wage_eur_log:age:height_cm:physic	-0.374664684422
## wage_eur_log:age:pace:physic	-1.318082428827
## wage_eur_log:height_cm:pace:physic	-0.028363420057
## age:height_cm:pace:physic	-0.062480467285
## wage_eur_log:age:dribbling:physic	-0.276144740434
## wage_eur_log:height_cm:dribbling:physic	0.152142993267
## age:height_cm:dribbling:physic	-0.014409410323
## wage_eur_log:pace:dribbling:physic	0.208501141767
## age:pace:dribbling:physic	-0.109541114481
## height_cm:pace:dribbling:physic	0.005776045847
## wage_eur_log:age:defending:physic	-1.174821714638
## wage_eur_log:height_cm:defending:physic	0.069235342143
## age:height_cm:defending:physic	-0.049636991435
## wage_eur_log:pace:defending:physic	-0.055335121516
## age:pace:defending:physic	-0.226013261879
## height_cm:pace:defending:physic	-0.006442295215
## wage_eur_log:dribbling:defending:physic	0.571161208370
## age:dribbling:defending:physic	-0.044774097852
## height_cm:dribbling:defending:physic	0.024023080452
## pace:dribbling:defending:physic	0.016342244766
## wage_eur_log:age:height_cm:power_long_shots	-0.005630508917
## wage_eur_log:age:pace:power_long_shots	-0.491737407209
## wage_eur_log:height_cm:pace:power_long_shots	0.125387024943
## age:height_cm:pace:power_long_shots	-0.016634143324
## wage_eur_log:age:dribbling:power_long_shots	0.795897661233
## wage_eur_log:height_cm:dribbling:power_long_shots	0.334970401997
## age:height_cm:dribbling:power_long_shots	0.045851891741
## wage_eur_log:pace:dribbling:power_long_shots	0.618424114457
## age:pace:dribbling:power_long_shots	0.022400579435
## height_cm:pace:dribbling:power_long_shots	0.027447459339
## wage_eur_log:age:defending:power_long_shots	-0.139067144832
## wage_eur_log:height_cm:defending:power_long_shots	0.256830052830
## age:height_cm:defending:power_long_shots	0.010465545204
## wage_eur_log:pace:defending:power_long_shots	0.381311744017
## age:pace:defending:power_long_shots	-0.090849062487
## height_cm:pace:defending:power_long_shots	0.016568453595
## wage_eur_log:dribbling:defending:power_long_shots	1.117930607660
## age:dribbling:defending:power_long_shots	0.144128592577

## height_cm:dribbling:defending:power_long_shots	0.053523941376
## pace:dribbling:defending:power_long_shots	0.082846642157
## wage_eur_log:age:physic:power_long_shots	0.222215897170
## wage_eur_log:height_cm:physic:power_long_shots	0.236779689334
## age:height_cm:physic:power_long_shots	0.025948674716
## wage_eur_log:pace:physic:power_long_shots	0.391325119039
## age:pace:physic:power_long_shots	-0.020182800019
## height_cm:pace:physic:power_long_shots	0.019333396696
## wage_eur_log:dribbling:physic:power_long_shots	0.979063749558
## age:dribbling:physic:power_long_shots	0.146741928532
## height_cm:dribbling:physic:power_long_shots	0.047399052070
## pace:dribbling:physic:power_long_shots	0.081595498733
## wage_eur_log:defending:physic:power_long_shots	0.721719016533
## age:defending:physic:power_long_shots	0.044247612340
## height_cm:defending:physic:power_long_shots	0.037412923685
## pace:defending:physic:power_long_shots	0.044247957282
## dribbling:defending:physic:power_long_shots	0.148354404333
## wage_eur_log:age:height_cm:pace:dribbling	0.005331123392
## wage_eur_log:age:height_cm:pace:defending	0.010235174316
## wage_eur_log:age:height_cm:dribbling:defending	0.002010092467
## wage_eur_log:age:pace:dribbling:defending	0.015912542654
## wage_eur_log:height_cm:pace:dribbling:defending	-0.001132841309
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## wage_eur_log:age:height_cm:dribbling:physic	0.002056054895
## wage_eur_log:age:pace:dribbling:physic	0.011767038188
## wage_eur_log:height_cm:pace:dribbling:physic	-0.000855335841
## age:height_cm:pace:dribbling:physic	0.000667888974
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## wage_eur_log:age:pace:defending:physic	0.026235829030
## wage_eur_log:height_cm:pace:defending:physic	0.000544426591
## age:height_cm:pace:defending:physic	0.001287598198
## wage_eur_log:age:dribbling:defending:physic	0.005221425669
## wage_eur_log:height_cm:dribbling:defending:physic	-0.002977468311
## age:height_cm:dribbling:defending:physic	0.000258948566
## wage_eur_log:pace:dribbling:defending:physic	-0.003470312447
## age:pace:dribbling:defending:physic	0.002366722350
## height_cm:pace:dribbling:defending:physic	-0.000071178432
## wage_eur_log:age:height_cm:pace:power_long_shots	0.003080992979
## wage_eur_log:age:height_cm:dribbling:power_long_shots	-0.004248840353
## wage_eur_log:age:pace:dribbling:power_long_shots	-0.002072876097
## wage_eur_log:height_cm:pace:dribbling:power_long_shots	-0.003296584231
## age:height_cm:pace:dribbling:power_long_shots	-0.000110382213
## wage_eur_log:age:height_cm:defending:power_long_shots	0.000673611437
## wage_eur_log:age:pace:defending:power_long_shots	0.012066518281
## wage_eur_log:height_cm:pace:defending:power_long_shots	-0.002079772333
## age:height_cm:pace:defending:power_long_shots	0.000479697000
## wage_eur_log:age:dribbling:defending:power_long_shots	-0.013940982606
## wage_eur_log:height_cm:dribbling:defending:power_long_shots	-0.006194988670
## age:height_cm:dribbling:defending:power_long_shots	-0.000841499173
## wage_eur_log:pace:dribbling:defending:power_long_shots	-0.010838602315
## age:pace:dribbling:defending:power_long_shots	-0.000088096069
## height_cm:pace:dribbling:defending:power_long_shots	-0.000464535729
## wage_eur_log:age:height_cm:physic:power_long_shots	-0.000817337459
## wage_eur_log:age:pace:physic:power_long_shots	0.005521936543
## wage_eur_log:height_cm:pace:physic:power_long_shots	-0.001975891046
## age:height_cm:pace:physic:power_long_shots	0.000154357856
## wage_eur_log:age:dribbling:physic:power_long_shots	-0.014101702417
## wage_eur_log:height_cm:dribbling:physic:power_long_shots	-0.005266498366
## age:height_cm:dribbling:physic:power_long_shots	-0.000789704318
## wage_eur_log:pace:dribbling:physic:power_long_shots	-0.009765182913
## age:pace:dribbling:physic:power_long_shots	-0.000561818968
## height_cm:pace:dribbling:physic:power_long_shots	-0.000436201436
## wage_eur_log:age:defending:physic:power_long_shots	0.000341945661
## wage_eur_log:height_cm:defending:physic:power_long_shots	-0.003929900706
## age:height_cm:defending:physic:power_long_shots	-0.000254878088
## wage_eur_log:pace:defending:physic:power_long_shots	-0.005451187035
## age:pace:defending:physic:power_long_shots	0.001259932116
## height_cm:pace:defending:physic:power_long_shots	-0.000241988059
## wage_eur_log:dribbling:defending:physic:power_long_shots	-0.017042227869
## age:dribbling:defending:physic:power_long_shots	-0.002307978473
## height_cm:dribbling:defending:physic:power_long_shots	-0.000822126844
## pace:dribbling:defending:physic:power_long_shots	-0.001234261089
## wage_eur_log:age:height_cm:pace:dribbling:defending	-0.000092453578
## wage_eur_log:age:height_cm:pace:dribbling:physic	-0.000075669089
## wage_eur_log:age:height_cm:pace:defending:physic	-0.000152823531
## wage_eur_log:age:height_cm:dribbling:defending:physic	-0.000033387244
## wage_eur_log:age:pace:dribbling:defending:physic	-0.000241294735
## wage_eur_log:height_cm:pace:dribbling:defending:physic	0.000015624284
## age:height_cm:pace:dribbling:defending:physic	-0.000013579181
## wage_eur_log:age:height_cm:pace:dribbling:power_long_shots	0.000007008036
## wage_eur_log:age:height_cm:pace:defending:power_long_shots	-0.000066910323
## wage_eur_log:age:height_cm:dribbling:defending:power_long_shots	0.000079573103
## wage_eur_log:age:pace:dribbling:defending:power_long_shots	0.000013337758



## wage_eur_log:height_cm:pace:dribbling:defending:power_long_shots	0.000059515606
## age:height_cm:pace:dribbling:defending:power_long_shots	0.000000917258
## wage_eur_log:age:height_cm:pace:physic:power_long_shots	-0.000038378643
## wage_eur_log:age:height_cm:dribbling:physic:power_long_shots	0.000072867433
## wage_eur_log:age:pace:dribbling:physic:power_long_shots	0.000053245126
## wage_eur_log:height_cm:pace:dribbling:physic:power_long_shots	0.000051135106
## age:height_cm:pace:dribbling:physic:power_long_shots	0.000002545317
## wage_eur_log:age:height_cm:defending:physic:power_long_shots	-0.000003131975
## wage_eur_log:age:pace:defending:physic:power_long_shots	-0.000176859814
## age:height_cm:height_cm:pace:defending:physic:power_long_shots	0.000028753066
## age:height_cm:pace:defending:physic:power_long_shots	-0.000006988649
## wage_eur_log:age:dribbling:defending:physic:power_long_shots	0.000218769832
## wage_eur_log:height_cm:dribbling:defending:physic:power_long_shots	0.000093345658
## age:height_cm:dribbling:defending:physic:power_long_shots	0.000013031219
## wage_eur_log:pace:dribbling:defending:physic:power_long_shots	0.000158595081
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## height_cm:pace:dribbling:defending:physic:power_long_shots	0.000006775439
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic	0.000001442517
## wage_eur_log:age:height_cm:pace:dribbling:defending:power_long_shots	-0.000000079616
## wage_eur_log:age:height_cm:pace:dribbling:physic:power_long_shots	-0.000000185991
## wage_eur_log:age:height_cm:pace:defending:physic:power_long_shots	0.000001021421
## wage_eur_log:age:height_cm:dribbling:defending:physic:power_long_shots	-0.000001201982
## wage_eur_log:age:pace:dribbling:defending:physic:power_long_shots	-0.000000106841
## wage_eur_log:height_cm:pace:dribbling:defending:physic:power_long_shots	-0.000000854063
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## wage_eur_log:age:height_cm:pace:dribbling:defending:physic:power_long_shots	NA
##	Std. Error
## (Intercept)	810426.497181685059
## wage_eur_log	122813.547944598744
## age	38735.313375743157
## height_cm	4496.184263362489
## pace	11686.555341824876
## dribbling	13945.644640084638
## defending	13459.418218453275
## physic	15295.182671274888
## power_long_shots	13951.856270178532
## I(wage_eur_log^2)	0.145235171449
## I(age^2)	0.010733368283
## I(height_cm^2)	0.004381462650
## I(pace^2)	0.002072281810
## I(dribbling^2)	0.005148935274
## I(defending^2)	0.001082375557
## I(physic^2)	0.003275170894
## I(power_long_shots^2)	0.001315203615
## wage_eur_log:age	5539.777740979263
## wage_eur_log:height_cm	682.920075444882
## age:height_cm	217.621862305997
## wage_eur_log:pace	1950.456345370333
## age:pace	571.705965557261
## height_cm:pace	66.479856435376
## wage_eur_log:dribbling	2005.842981212068
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## height_cm:dribbling	77.521294908730
## pace:dribbling	192.299598234302
## wage_eur_log:defending	1758.858024059992
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## height_cm:defending	72.684572926952
## pace:defending	177.971443259344
## dribbling:defending	248.339805038806
## wage_eur_log:physic	2075.629593573948
## age:physic	660.762005142207
## height_cm:physic	83.288026796875
## pace:physic	218.891982513219
## dribbling:physic	260.335989978091
## defending:physic	194.378551307029
## wage_eur_log:power_long_shots	1963.687746257762
## age:power_long_shots	579.858098894247
## height_cm:power_long_shots	77.195051475967
## pace:power_long_shots	179.189188027685
## dribbling:power_long_shots	215.202741688037
## defending:power_long_shots	258.629949478439
## physic:power_long_shots	264.629238461172
## wage_eur_log:age:height_cm	31.075209731985
## wage_eur_log:age:pace	87.312472940094
## wage_eur_log:height_cm:pace	10.989598525918
## age:height_cm:pace	3.272989564296
## wage_eur_log:age:dribbling	87.061533591231
## wage_eur_log:height_cm:dribbling	11.148392141102
## age:height_cm:dribbling	3.544543921475
## wage_eur_log:pace:dribbling	30.498146007141
## age:pace:dribbling	8.899333727820
## height_cm:pace:dribbling	1.090049579773
## wage_eur_log:age:defending	74.514440418277
## wage_eur_log:height_cm:defending	9.546249739779

## age:height_cm:defending	3.235288286923
## wage_eur_log:pace:defending	25.685120250357
## age:pace:defending	8.069949577474
## height_cm:pace:defending	0.985849145268
## wage_eur_log:dribbling:defending	31.606458718278
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## height_cm:dribbling:defending	1.358181876024
## pace:dribbling:defending	3.312045073890
## wage_eur_log:age:physic	88.941948241338
## wage_eur_log:height_cm:physic	11.393973665137
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## wage_eur_log:pace:physic	31.937537108410
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## height_cm:pace:physic	1.215575174577
## wage_eur_log:dribbling:physic	34.337476044516
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## height_cm:dribbling:physic	1.421707002409
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## wage_eur_log:defending:physic	23.491570063860
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## height_cm:defending:physic	1.006931384934
## pace:defending:physic	2.312676080519
## dribbling:defending:physic	3.517828638066
## wage_eur_log:age:power_long_shots	81.555731086879
## wage_eur_log:height_cm:power_long_shots	10.932890051673
## age:height_cm:power_long_shots	3.267745375642
## wage_eur_log:pace:power_long_shots	28.509417496257
## age:pace:power_long_shots	7.995464778055
## height_cm:pace:power_long_shots	1.027004618366
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## height_cm:dribbling:power_long_shots	1.193165064917
## pace:dribbling:power_long_shots	2.628342240032
## wage_eur_log:defending:power_long_shots	28.767919329372
## age:defending:power_long_shots	9.015658198205
## height_cm:defending:power_long_shots	1.396369412633
## pace:defending:power_long_shots	3.237479608423
## dribbling:defending:power_long_shots	4.241270283782
## wage_eur_log:physic:power_long_shots	33.050638297695
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## height_cm:physic:power_long_shots	1.437081213398
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## dribbling:physic:power_long_shots	4.156027851116
## defending:physic:power_long_shots	3.928929803536
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## wage_eur_log:age:height_cm:dribbling	0.486929002184
## wage_eur_log:age:pace:dribbling	1.317773778297
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## wage_eur_log:age:height_cm:defending	0.409872451170
## wage_eur_log:age:pace:defending	1.095140818957
## wage_eur_log:height_cm:pace:defending	0.142503519257
## age:height_cm:pace:defending	0.045255138759
## wage_eur_log:age:dribbling:defending	1.283984735203
## wage_eur_log:height_cm:dribbling:defending	0.172999108263
## age:height_cm:dribbling:defending	0.057392329299
## wage_eur_log:pace:dribbling:defending	0.451132636250
## age:pace:dribbling:defending	0.140902904494
## height_cm:pace:dribbling:defending	0.018419560791
## wage_eur_log:age:height_cm:physic	0.492628455912
## wage_eur_log:age:pace:physic	1.378967302031
## wage_eur_log:height_cm:pace:physic	0.177914057260
## age:height_cm:pace:physic	0.054451714054
## wage_eur_log:age:dribbling:physic	1.417352241947
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## age:height_cm:dribbling:physic	0.059479309606
## wage_eur_log:pace:dribbling:physic	0.506096944054
## age:pace:dribbling:physic	0.151715147610
## height_cm:pace:dribbling:physic	0.019900877826
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## wage_eur_log:height_cm:defending:physic	0.122910820051
## age:height_cm:defending:physic	0.041550873731
## wage_eur_log:pace:defending:physic	0.313478170565
## age:pace:defending:physic	0.098909842899
## height_cm:pace:defending:physic	0.012176743681
## wage_eur_log:dribbling:defending:physic	0.426914844337
## age:dribbling:defending:physic	0.136550122352
## height_cm:dribbling:defending:physic	0.018632010090
## pace:dribbling:defending:physic	0.042625231119
## wage_eur_log:age:height_cm:power_long_shots	0.459533551961
## wage_eur_log:age:pace:power_long_shots	1.220544535044
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## age:height_cm:dribbling:power_long_shots	0.047297828848
## wage_eur_log:pace:dribbling:power_long_shots	0.400264819299
## age:pace:dribbling:power_long_shots	0.107995787230
## height_cm:pace:dribbling:power_long_shots	0.015075769968
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## wage_eur_log:height_cm:defending:power_long_shots	0.155127189501
## age:height_cm:defending:power_long_shots	0.049010084355
## wage_eur_log:pace:defending:power_long_shots	0.361058556808
## age:pace:defending:power_long_shots	0.112970471569
## height_cm:pace:defending:power_long_shots	0.017937315452
## wage_eur_log:dribbling:defending:power_long_shots	0.456462945977
## age:dribbling:defending:power_long_shots	0.143116343817
## height_cm:dribbling:defending:power_long_shots	0.023085497056
## pace:dribbling:defending:power_long_shots	0.052889070797
## wage_eur_log:age:physic:power_long_shots	1.325598445062
## wage_eur_log:height_cm:physic:power_long_shots	0.181468336192
## age:height_cm:physic:power_long_shots	0.057183438902
## wage_eur_log:pace:physic:power_long_shots	0.475070201201
## age:pace:physic:power_long_shots	0.146573183938
## height_cm:pace:physic:power_long_shots	0.019755459279
## wage_eur_log:dribbling:physic:power_long_shots	0.498608026165
## age:dribbling:physic:power_long_shots	0.155369335433
## height_cm:dribbling:physic:power_long_shots	0.022586321254
## pace:dribbling:physic:power_long_shots	0.053923054024
## wage_eur_log:defending:physic:power_long_shots	0.392803175064
## age:defending:physic:power_long_shots	0.128093699710
## height_cm:defending:physic:power_long_shots	0.020533824469
## pace:defending:physic:power_long_shots	0.046583510415
## dribbling:defending:physic:power_long_shots	0.062808876106
## wage_eur_log:age:height_cm:pace:dribbling	0.007447327905
## wage_eur_log:age:height_cm:pace:defending	0.006150260399
## wage_eur_log:age:height_cm:dribbling:defending	0.007074562404
## wage_eur_log:age:pace:dribbling:defending	0.018335336424
## wage_eur_log:height_cm:pace:dribbling:defending	0.002505433324
## age:height_cm:pace:dribbling:defending	0.000787836699
## wage_eur_log:age:height_cm:pace:physic	0.007738205979
## wage_eur_log:age:height_cm:dribbling:physic	0.007817902767
## wage_eur_log:age:pace:dribbling:physic	0.021052544142
## wage_eur_log:height_cm:pace:dribbling:physic	0.002810466964
## age:height_cm:pace:dribbling:physic	0.000847866067
## wage_eur_log:age:height_cm:defending:physic	0.005048748438
## wage_eur_log:age:pace:defending:physic	0.013064928607
## wage_eur_log:height_cm:pace:defending:physic	0.001686374482
## age:height_cm:pace:defending:physic	0.000536316735
## wage_eur_log:age:dribbling:defending:physic	0.016315719295
## wage_eur_log:height_cm:dribbling:defending:physic	0.002272555235
## age:height_cm:dribbling:defending:physic	0.000726451856
## wage_eur_log:pace:dribbling:defending:physic	0.005598512333
## age:pace:dribbling:defending:physic	0.001677126692
## height_cm:pace:dribbling:defending:physic	0.000228631408
## wage_eur_log:age:height_cm:pace:power_long_shots	0.007014161366
## wage_eur_log:age:height_cm:dribbling:power_long_shots	0.006311165002
## wage_eur_log:age:pace:dribbling:power_long_shots	0.016133233641
## wage_eur_log:height_cm:pace:dribbling:power_long_shots	0.002279719233
## age:height_cm:pace:dribbling:power_long_shots	0.000632392570
## wage_eur_log:age:height_cm:defending:power_long_shots	0.005292539974
## wage_eur_log:age:pace:defending:power_long_shots	0.012044834405
## wage_eur_log:height_cm:pace:defending:power_long_shots	0.002010985946
## age:height_cm:pace:defending:power_long_shots	0.000640247845
## wage_eur_log:age:dribbling:defending:power_long_shots	0.014330313535
## wage_eur_log:height_cm:dribbling:defending:power_long_shots	0.002479154330
## age:height_cm:dribbling:defending:power_long_shots	0.000780389956
## wage_eur_log:pace:dribbling:defending:power_long_shots	0.005622557787
## age:pace:dribbling:defending:power_long_shots	0.001741546727
## height_cm:pace:dribbling:defending:power_long_shots	0.000293900624
## wage_eur_log:age:height_cm:physic:power_long_shots	0.007358465640
## wage_eur_log:age:pace:physic:power_long_shots	0.019744110286
## wage_eur_log:height_cm:pace:physic:power_long_shots	0.002659916889
## age:height_cm:pace:physic:power_long_shots	0.000827061062
## wage_eur_log:age:dribbling:physic:power_long_shots	0.018924046383
## wage_eur_log:height_cm:dribbling:physic:power_long_shots	0.002732418869
## age:height_cm:dribbling:physic:power_long_shots	0.000851102862
## wage_eur_log:pace:dribbling:physic:power_long_shots	0.006897191156
## age:pace:dribbling:physic:power_long_shots	0.002076854978
## height_cm:pace:dribbling:physic:power_long_shots	0.000298498302
## wage_eur_log:age:defending:physic:power_long_shots	0.011794522819
## wage_eur_log:height_cm:defending:physic:power_long_shots	0.002029321862
## age:height_cm:defending:physic:power_long_shots	0.000663453535
## wage_eur_log:pace:defending:physic:power_long_shots	0.004279757673
## age:pace:defending:physic:power_long_shots	0.001475819221
## height_cm:pace:defending:physic:power_long_shots	0.000246737048
## wage_eur_log:dribbling:defending:physic:power_long_shots	0.006127762410
## age:dribbling:defending:physic:power_long_shots	0.001927996631

## height_cm:dribbling:defending:physic:power_long_shots	0.000330953228
## pace:dribbling:defending:physic:power_long_shots	0.000725021602
## wage_eur_log:age:height_cm:pace:dribbling:defending	0.000102592562
## wage_eur_log:age:height_cm:pace:dribbling:physic	0.000117571672
## wage_eur_log:age:height_cm:pace:defending:physic	0.000071979248
## wage_eur_log:age:height_cm:dribbling:defending:physic	0.000087373376
## wage_eur_log:age:pace:dribbling:defending:physic	0.000215912396
## wage_eur_log:height_cm:pace:dribbling:defending:physic	0.000030326307
## age:height_cm:pace:dribbling:defending:physic	0.00009099105
## wage_eur_log:age:height_cm:pace:dribbling:power_long_shots	0.000093121501
## wage_eur_log:age:height_cm:pace:defending:power_long_shots	0.000069363086
## wage_eur_log:age:height_cm:dribbling:defending:power_long_shots	0.000078111739
## wage_eur_log:age:pace:dribbling:defending:power_long_shots	0.000165875754
## wage_eur_log:height_cm:pace:dribbling:defending:power_long_shots	0.000031375403
## age:height_cm:pace:dribbling:defending:power_long_shots	0.00009835699
## wage_eur_log:age:height_cm:pace:physic:power_long_shots	0.00011784899
## wage_eur_log:age:height_cm:dribbling:physic:power_long_shots	0.000104541874
## wage_eur_log:age:pace:dribbling:physic:power_long_shots	0.000270231998
## wage_eur_log:height_cm:pace:dribbling:physic:power_long_shots	0.000038512093
## age:height_cm:pace:dribbling:physic:power_long_shots	0.000011671824
## wage_eur_log:age:height_cm:defending:physic:power_long_shots	0.000059966062
## wage_eur_log:age:pace:defending:physic:power_long_shots	0.000112317648
## wage_eur_log:height_cm:pace:defending:physic:power_long_shots	0.000022535367
## age:height_cm:pace:defending:physic:power_long_shots	0.00007904211
## wage_eur_log:age:dribbling:defending:physic:power_long_shots	0.000157743008
## wage_eur_log:height_cm:dribbling:defending:physic:power_long_shots	0.000031855198
## age:height_cm:dribbling:defending:physic:power_long_shots	0.000009953589
## wage_eur_log:pace:dribbling:defending:physic:power_long_shots	0.000062502034
## age:pace:dribbling:defending:physic:power_long_shots	0.000019937464
## height_cm:pace:dribbling:defending:physic:power_long_shots	0.000003852540
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic	0.000001183418
## wage_eur_log:age:height_cm:pace:dribbling:defending:power_long_shots	0.00000956830
## wage_eur_log:age:height_cm:pace:dribbling:physic:power_long_shots	0.000001526263
## wage_eur_log:age:height_cm:pace:defending:physic:power_long_shots	0.000000610931
## wage_eur_log:age:height_cm:dribbling:defending:physic:power_long_shots	0.000000773746
## wage_eur_log:age:pace:dribbling:defending:physic:power_long_shots	0.000000498939
## wage_eur_log:height_cm:pace:dribbling:defending:physic:power_long_shots	0.000000327218
## age:height_cm:pace:dribbling:defending:physic:power_long_shots	0.000000104755
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic:power_long_shots	NA
##	t value
## (Intercept)	0.527
## wage_eur_log	-0.328
## age	0.848
## height_cm	-0.460
## pace	0.191
## dribbling	-1.067
## defending	-0.542
## physic	-0.577
## power_long_shots	-1.852
## I(wage_eur_log^2)	4.751
## I(age^2)	6.980
## I(height_cm^2)	0.016
## I(pace^2)	1.130
## I(dribbling^2)	1.860
## I(defending^2)	4.668
## I(physic^2)	0.878
## I(power_long_shots^2)	0.293
## wage_eur_log:age	-0.838
## wage_eur_log:height_cm	0.241
## age:height_cm	-0.877
## wage_eur_log:pace	-0.129
## age:pace	-1.352
## height_cm:pace	-0.277
## wage_eur_log:dribbling	0.916
## age:dribbling	-0.304
## height_cm:dribbling	1.016
## pace:dribbling	0.358
## wage_eur_log:defending	0.515
## age:defending	-1.083
## height_cm:defending	0.530
## pace:defending	-0.392
## dribbling:defending	1.158
## wage_eur_log:physic	0.393
## age:physic	-0.604
## height_cm:physic	0.506
## pace:physic	-0.034
## dribbling:physic	0.971
## defending:physic	0.675
## wage_eur_log:power_long_shots	1.388
## age:power_long_shots	0.339
## height_cm:power_long_shots	1.836
## pace:power_long_shots	1.212
## dribbling:power_long_shots	2.489
## defending:power_long_shots	1.619

## physic:power_long_shots	1.647
## wage_eur_log:age:height_cm	0.890
## wage_eur_log:age:pace	1.097
## wage_eur_log:height_cm:pace	0.225
## age:height_cm:pace	1.372
## wage_eur_log:age:dribbling	0.277
## wage_eur_log:height_cm:dribbling	-0.842
## age:height_cm:dribbling	0.324
## wage_eur_log:pace:dribbling	-0.379
## age:pace:dribbling	0.897
## height_cm:pace:dribbling	-0.274
## wage_eur_log:age:defending	1.112
## wage_eur_log:height_cm:defending	-0.464
## age:height_cm:defending	1.092
## wage_eur_log:pace:defending	0.145
## age:pace:defending	1.928
## height_cm:pace:defending	0.444
## wage_eur_log:dribbling:defending	-1.188
## age:dribbling:defending	0.290
## height_cm:dribbling:defending	-1.159
## pace:dribbling:defending	-0.322
## wage_eur_log:age:physic	0.679
## wage_eur_log:height_cm:physic	-0.296
## age:height_cm:physic	0.669
## wage_eur_log:pace:physic	0.046
## age:pace:physic	1.079
## height_cm:pace:physic	0.136
## wage_eur_log:dribbling:physic	-0.886
## age:dribbling:physic	0.195
## height_cm:dribbling:physic	-0.920
## pace:dribbling:physic	-0.382
## wage_eur_log:defending:physic	-0.631
## age:defending:physic	1.108
## height_cm:defending:physic	-0.660
## pace:defending:physic	0.402
## dribbling:defending:physic	-1.276
## wage_eur_log:age:power_long_shots	-0.024
## wage_eur_log:height_cm:power_long_shots	-1.339
## age:height_cm:power_long_shots	-0.320
## wage_eur_log:pace:power_long_shots	-0.851
## age:pace:power_long_shots	0.343
## height_cm:pace:power_long_shots	-1.137
## wage_eur_log:dribbling:power_long_shots	-2.150
## age:dribbling:power_long_shots	-0.983
## height_cm:dribbling:power_long_shots	-2.478
## pace:dribbling:power_long_shots	-1.917
## wage_eur_log:defending:power_long_shots	-1.614
## age:defending:power_long_shots	-0.164
## height_cm:defending:power_long_shots	-1.688
## pace:defending:power_long_shots	-0.904
## dribbling:defending:power_long_shots	-2.250
## wage_eur_log:physic:power_long_shots	-1.352
## age:physic:power_long_shots	-0.487
## height_cm:physic:power_long_shots	-1.643
## pace:physic:power_long_shots	-1.032
## dribbling:physic:power_long_shots	-2.091
## defending:physic:power_long_shots	-1.719
## wage_eur_log:age:height_cm:pace	-1.150
## wage_eur_log:age:height_cm:dribbling	-0.322
## wage_eur_log:age:pace:dribbling	-0.660
## wage_eur_log:height_cm:pace:dribbling	0.284
## age:height_cm:pace:dribbling	-0.922
## wage_eur_log:age:height_cm:defending	-1.150
## wage_eur_log:age:pace:defending	-1.633
## wage_eur_log:height_cm:pace:defending	-0.226
## age:height_cm:pace:defending	-1.925
## wage_eur_log:age:dribbling:defending	-0.266
## wage_eur_log:height_cm:dribbling:defending	1.157
## age:height_cm:dribbling:defending	-0.279
## wage_eur_log:pace:dribbling:defending	0.521
## age:pace:dribbling:defending	-1.107
## height_cm:pace:dribbling:defending	0.282
## wage_eur_log:age:height_cm:physic	-0.761
## wage_eur_log:age:pace:physic	-0.956
## wage_eur_log:height_cm:pace:physic	-0.159
## age:height_cm:pace:physic	-1.147
## wage_eur_log:age:dribbling:physic	-0.195
## wage_eur_log:height_cm:dribbling:physic	0.808
## age:height_cm:dribbling:physic	-0.242
## wage_eur_log:pace:dribbling:physic	0.412
## age:pace:dribbling:physic	-0.722
## height_cm:pace:dribbling:physic	0.290
## wage_eur_log:age:defending:physic	-1.241
## wage_eur_log:height_cm:defending:physic	0.563

## age:height_cm:defending:physic	-1.195
## wage_eur_log:pace:defending:physic	-0.177
## age:pace:defending:physic	-2.285
## height_cm:pace:defending:physic	-0.529
## wage_eur_log:dribbling:defending:physic	1.338
## age:dribbling:defending:physic	-0.328
## height_cm:dribbling:defending:physic	1.289
## pace:dribbling:defending:physic	0.383
## wage_eur_log:age:height_cm:power_long_shots	-0.012
## wage_eur_log:age:pace:power_long_shots	-0.403
## wage_eur_log:height_cm:pace:power_long_shots	0.773
## age:height_cm:pace:power_long_shots	-0.357
## wage_eur_log:age:dribbling:power_long_shots	0.711
## wage_eur_log:height_cm:dribbling:power_long_shots	2.101
## age:height_cm:dribbling:power_long_shots	0.969
## wage_eur_log:pace:dribbling:power_long_shots	1.545
## age:pace:dribbling:power_long_shots	0.207
## height_cm:pace:dribbling:power_long_shots	1.821
## wage_eur_log:age:defending:power_long_shots	-0.143
## wage_eur_log:height_cm:defending:power_long_shots	1.656
## age:height_cm:defending:power_long_shots	0.214
## wage_eur_log:pace:defending:power_long_shots	1.056
## age:pace:defending:power_long_shots	-0.804
## height_cm:pace:defending:power_long_shots	0.924
## wage_eur_log:dribbling:defending:power_long_shots	2.449
## age:dribbling:defending:power_long_shots	1.007
## height_cm:dribbling:defending:power_long_shots	2.319
## pace:dribbling:defending:power_long_shots	1.566
## wage_eur_log:age:physic:power_long_shots	0.168
## wage_eur_log:height_cm:physic:power_long_shots	1.305
## age:height_cm:physic:power_long_shots	0.454
## wage_eur_log:pace:physic:power_long_shots	0.824
## age:pace:physic:power_long_shots	-0.138
## height_cm:pace:physic:power_long_shots	0.979
## wage_eur_log:dribbling:physic:power_long_shots	1.964
## age:dribbling:physic:power_long_shots	0.944
## height_cm:dribbling:physic:power_long_shots	2.099
## pace:dribbling:physic:power_long_shots	1.513
## wage_eur_log:defending:physic:power_long_shots	1.837
## age:defending:physic:power_long_shots	0.345
## height_cm:defending:physic:power_long_shots	1.822
## pace:defending:physic:power_long_shots	0.950
## dribbling:defending:physic:power_long_shots	2.362
## wage_eur_log:age:height_cm:pace:dribbling	0.716
## wage_eur_log:age:height_cm:pace:defending	1.664
## wage_eur_log:age:height_cm:dribbling:defending	0.284
## wage_eur_log:age:pace:dribbling:defending	0.868
## wage_eur_log:height_cm:pace:dribbling:defending	-0.452
## age:height_cm:pace:dribbling:defending	1.107
## wage_eur_log:age:height_cm:pace:physic	1.041
## wage_eur_log:age:height_cm:dribbling:physic	0.263
## wage_eur_log:age:pace:dribbling:physic	0.559
## wage_eur_log:height_cm:pace:dribbling:physic	-0.304
## age:height_cm:pace:dribbling:physic	0.788
## wage_eur_log:age:height_cm:defending:physic	1.358
## wage_eur_log:age:pace:defending:physic	2.008
## wage_eur_log:height_cm:pace:defending:physic	0.323
## age:height_cm:pace:defending:physic	2.401
## wage_eur_log:age:dribbling:defending:physic	0.320
## wage_eur_log:height_cm:dribbling:defending:physic	-1.310
## age:height_cm:dribbling:defending:physic	0.356
## wage_eur_log:pace:dribbling:defending:physic	-0.620
## age:pace:dribbling:defending:physic	1.411
## height_cm:pace:dribbling:defending:physic	-0.311
## wage_eur_log:age:height_cm:pace:power_long_shots	0.439
## wage_eur_log:age:height_cm:dribbling:power_long_shots	-0.673
## wage_eur_log:age:pace:dribbling:power_long_shots	-0.128
## wage_eur_log:height_cm:pace:dribbling:power_long_shots	-1.446
## age:height_cm:pace:dribbling:power_long_shots	-0.175
## wage_eur_log:age:height_cm:defending:power_long_shots	0.127
## wage_eur_log:age:pace:defending:power_long_shots	1.002
## wage_eur_log:height_cm:pace:defending:power_long_shots	-1.034
## age:height_cm:pace:defending:power_long_shots	0.749
## wage_eur_log:age:dribbling:defending:power_long_shots	-0.973
## wage_eur_log:height_cm:dribbling:defending:power_long_shots	-2.499
## age:height_cm:dribbling:defending:power_long_shots	-1.078
## wage_eur_log:pace:dribbling:defending:power_long_shots	-1.928
## age:pace:dribbling:defending:power_long_shots	-0.051
## height_cm:pace:dribbling:defending:power_long_shots	-1.581
## wage_eur_log:age:height_cm:physic:power_long_shots	-0.111
## wage_eur_log:age:pace:physic:power_long_shots	0.280
## wage_eur_log:height_cm:pace:physic:power_long_shots	-0.743
## age:height_cm:pace:physic:power_long_shots	0.187
## wage_eur_log:age:dribbling:physic:power_long_shots	-0.745

## wage_eur_log:height_cm:dribbling:physic:power_long_shots	-1.927
## age:height_cm:dribbling:physic:power_long_shots	-0.928
## wage_eur_log:pace:dribbling:physic:power_long_shots	-1.416
## age:pace:dribbling:physic:power_long_shots	-0.271
## height_cm:pace:dribbling:physic:power_long_shots	-1.461
## wage_eur_log:age:defending:physic:power_long_shots	0.029
## wage_eur_log:height_cm:defending:physic:power_long_shots	-1.937
## age:height_cm:defending:physic:power_long_shots	-0.384
## wage_eur_log:pace:defending:physic:power_long_shots	-1.274
## age:pace:defending:physic:power_long_shots	0.854
## height_cm:pace:defending:physic:power_long_shots	-0.981
## wage_eur_log:dribbling:defending:physic:power_long_shots	-2.781
## age:dribbling:defending:physic:power_long_shots	-1.197
## height_cm:dribbling:defending:physic:power_long_shots	-2.484
## pace:dribbling:defending:physic:power_long_shots	-1.702
## wage_eur_log:age:height_cm:pace:dribbling:defending	-0.901
## wage_eur_log:age:height_cm:pace:dribbling:physic	-0.644
## wage_eur_log:age:height_cm:pace:defending:physic	-2.123
## wage_eur_log:age:height_cm:dribbling:defending:physic	-0.382
## wage_eur_log:age:pace:dribbling:defending:physic	-1.118
## wage_eur_log:height_cm:pace:dribbling:defending:physic	0.515
## age:height_cm:pace:dribbling:defending:physic	-1.492
## wage_eur_log:age:height_cm:pace:dribbling:power_long_shots	0.075
## wage_eur_log:age:height_cm:pace:defending:power_long_shots	-0.965
## wage_eur_log:age:height_cm:dribbling:defending:power_long_shots	1.019
## wage_eur_log:age:pace:dribbling:defending:power_long_shots	0.080
## wage_eur_log:height_cm:pace:dribbling:defending:power_long_shots	1.897
## age:height_cm:pace:dribbling:defending:power_long_shots	0.093
## wage_eur_log:age:height_cm:pace:physic:power_long_shots	-0.343
## wage_eur_log:age:height_cm:dribbling:physic:power_long_shots	0.697
## wage_eur_log:age:pace:dribbling:physic:power_long_shots	0.197
## wage_eur_log:height_cm:pace:dribbling:physic:power_long_shots	1.328
## age:height_cm:pace:dribbling:physic:power_long_shots	0.218
## wage_eur_log:age:height_cm:defending:physic:power_long_shots	-0.052
## wage_eur_log:age:pace:defending:physic:power_long_shots	-1.575
## wage_eur_log:height_cm:pace:defending:physic:power_long_shots	1.276
## age:height_cm:pace:defending:physic:power_long_shots	-0.884
## wage_eur_log:age:dribbling:defending:physic:power_long_shots	1.387
## wage_eur_log:height_cm:dribbling:defending:physic:power_long_shots	2.930
## age:height_cm:dribbling:defending:physic:power_long_shots	1.309
## wage_eur_log:pace:dribbling:defending:physic:power_long_shots	2.537
## age:pace:dribbling:defending:physic:power_long_shots	0.044
## height_cm:pace:dribbling:defending:physic:power_long_shots	1.759
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic	1.219
## wage_eur_log:age:height_cm:pace:dribbling:defending:power_long_shots	-0.083
## wage_eur_log:age:height_cm:pace:dribbling:physic:power_long_shots	-0.122
## wage_eur_log:age:height_cm:pace:defending:physic:power_long_shots	1.672
## wage_eur_log:age:height_cm:dribbling:defending:physic:power_long_shots	-1.553
## wage_eur_log:age:pace:dribbling:defending:physic:power_long_shots	-0.214
## wage_eur_log:height_cm:pace:dribbling:defending:physic:power_long_shots	-2.610
## age:height_cm:pace:dribbling:defending:physic:power_long_shots	-0.051
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic:power_long_shots	NA
##	Pr(> t )
## (Intercept)	0.59839
## wage_eur_log	0.74307
## age	0.39737
## height_cm	0.64610
## pace	0.84857
## dribbling	0.28686
## defending	0.58824
## physic	0.56475
## power_long_shots	0.06526
## I(wage_eur_log^2)	0.000034032091
## I(age^2)	0.000000000262
## I(height_cm^2)	0.98690
## I(pace^2)	0.25947
## I(dribbling^2)	0.06412
## I(defending^2)	0.000049578581
## I(physic^2)	0.38071
## I(power_long_shots^2)	0.76961
## wage_eur_log:age	0.40266
## wage_eur_log:height_cm	0.80988
## age:height_cm	0.38127
## wage_eur_log:pace	0.89747
## age:pace	0.17750
## height_cm:pace	0.78208
## wage_eur_log:dribbling	0.36052
## age:dribbling	0.76150
## height_cm:dribbling	0.31067
## pace:dribbling	0.72032
## wage_eur_log:defending	0.60699
## age:defending	0.27986
## height_cm:defending	0.59645
## pace:defending	0.69511

## dribbling:defending	0.24796
## wage_eur_log:physic	0.69440
## age:physic	0.54648
## height_cm:physic	0.61348
## pace:physic	0.97315
## dribbling:physic	0.33239
## defending:physic	0.50042
## wage_eur_log:power_long_shots	0.16637
## age:power_long_shots	0.73487
## height_cm:power_long_shots	0.06747
## pace:power_long_shots	0.22670
## dribbling:power_long_shots	0.01347
## defending:power_long_shots	0.10660
## physic:power_long_shots	0.10073
## wage_eur_log:age:height_cm	0.37408
## wage_eur_log:age:pace	0.27386
## wage_eur_log:height_cm:pace	0.82211
## age:height_cm:pace	0.17136
## wage_eur_log:age:dribbling	0.78181
## wage_eur_log:height_cm:dribbling	0.40034
## age:height_cm:dribbling	0.74638
## wage_eur_log:pace:dribbling	0.70496
## age:pace:dribbling	0.37057
## height_cm:pace:dribbling	0.78448
## wage_eur_log:age:defending	0.26705
## wage_eur_log:height_cm:defending	0.64296
## age:height_cm:defending	0.27580
## wage_eur_log:pace:defending	0.88498
## age:pace:defending	0.05498
## height_cm:pace:defending	0.65728
## wage_eur_log:dribbling:defending	0.23585
## age:dribbling:defending	0.77188
## height_cm:dribbling:defending	0.24762
## pace:dribbling:defending	0.74809
## wage_eur_log:age:physic	0.49791
## wage_eur_log:height_cm:physic	0.76743
## age:height_cm:physic	0.50416
## wage_eur_log:pace:physic	0.96369
## age:pace:physic	0.28146
## height_cm:pace:physic	0.89193
## wage_eur_log:dribbling:physic	0.37645
## age:dribbling:physic	0.84592
## height_cm:dribbling:physic	0.35870
## pace:dribbling:physic	0.70270
## wage_eur_log:defending:physic	0.52885
## age:defending:physic	0.26892
## height_cm:defending:physic	0.51007
## pace:defending:physic	0.68826
## dribbling:defending:physic	0.20296
## wage_eur_log:age:power_long_shots	0.98076
## wage_eur_log:height_cm:power_long_shots	0.18187
## age:height_cm:power_long_shots	0.74943
## wage_eur_log:pace:power_long_shots	0.39561
## age:pace:power_long_shots	0.73151
## height_cm:pace:power_long_shots	0.25663
## wage_eur_log:dribbling:power_long_shots	0.03249
## age:dribbling:power_long_shots	0.32659
## height_cm:dribbling:power_long_shots	0.01387
## pace:dribbling:power_long_shots	0.05632
## wage_eur_log:defending:power_long_shots	0.10776
## age:defending:power_long_shots	0.86967
## height_cm:defending:power_long_shots	0.09262
## pace:defending:power_long_shots	0.36699
## dribbling:defending:power_long_shots	0.02534
## wage_eur_log:physic:power_long_shots	0.17760
## age:physic:power_long_shots	0.62644
## height_cm:physic:power_long_shots	0.10160
## pace:physic:power_long_shots	0.30299
## dribbling:physic:power_long_shots	0.03750
## defending:physic:power_long_shots	0.08679
## wage_eur_log:age:height_cm:pace	0.25118
## wage_eur_log:age:height_cm:dribbling	0.74803
## wage_eur_log:age:pace:dribbling	0.51000
## wage_eur_log:height_cm:pace:dribbling	0.77652
## age:height_cm:pace:dribbling	0.35723
## wage_eur_log:age:height_cm:defending	0.25117
## wage_eur_log:age:pace:defending	0.10370
## wage_eur_log:height_cm:pace:defending	0.82113
## age:height_cm:pace:defending	0.05538
## wage_eur_log:age:dribbling:defending	0.79055
## wage_eur_log:height_cm:dribbling:defending	0.24852
## age:height_cm:dribbling:defending	0.78059
## wage_eur_log:pace:dribbling:defending	0.60249
## age:pace:dribbling:defending	0.26919



## height_cm:pace:dribbling:defending	0.77844
## wage_eur_log:age:height_cm:physic	0.44764
## wage_eur_log:age:pace:physic	0.34007
## wage_eur_log:height_cm:pace:physic	0.87346
## age:height_cm:pace:physic	0.25229
## wage_eur_log:age:dribbling:physic	0.84568
## wage_eur_log:height_cm:dribbling:physic	0.42001
## age:height_cm:dribbling:physic	0.80878
## wage_eur_log:pace:dribbling:physic	0.68071
## age:pace:dribbling:physic	0.47096
## height_cm:pace:dribbling:physic	0.77187
## wage_eur_log:age:defending:physic	0.21585
## wage_eur_log:height_cm:defending:physic	0.57374
## age:height_cm:defending:physic	0.23337
## wage_eur_log:pace:defending:physic	0.86003
## age:pace:defending:physic	0.02315
## height_cm:pace:defending:physic	0.59723
## wage_eur_log:dribbling:defending:physic	0.18215
## age:dribbling:defending:physic	0.74326
## height_cm:dribbling:defending:physic	0.19847
## pace:dribbling:defending:physic	0.70175
## wage_eur_log:age:height_cm:power_long_shots	0.99023
## wage_eur_log:age:pace:power_long_shots	0.68738
## wage_eur_log:height_cm:pace:power_long_shots	0.44002
## age:height_cm:pace:power_long_shots	0.72148
## wage_eur_log:age:dribbling:power_long_shots	0.47774
## wage_eur_log:height_cm:dribbling:power_long_shots	0.03666
## age:height_cm:dribbling:power_long_shots	0.33326
## wage_eur_log:pace:dribbling:power_long_shots	0.12360
## age:pace:dribbling:power_long_shots	0.83585
## height_cm:pace:dribbling:power_long_shots	0.06985
## wage_eur_log:age:defending:power_long_shots	0.88628
## wage_eur_log:height_cm:defending:power_long_shots	0.09905
## age:height_cm:defending:power_long_shots	0.83108
## wage_eur_log:pace:defending:power_long_shots	0.29194
## age:pace:defending:power_long_shots	0.42205
## height_cm:pace:defending:power_long_shots	0.35654
## wage_eur_log:dribbling:defending:power_long_shots	0.01501
## age:dribbling:defending:power_long_shots	0.31487
## height_cm:dribbling:defending:power_long_shots	0.02123
## pace:dribbling:defending:power_long_shots	0.11851
## wage_eur_log:age:physic:power_long_shots	0.86701
## wage_eur_log:height_cm:physic:power_long_shots	0.19316
## age:height_cm:physic:power_long_shots	0.65038
## wage_eur_log:pace:physic:power_long_shots	0.41088
## age:pace:physic:power_long_shots	0.89059
## height_cm:pace:physic:power_long_shots	0.32870
## wage_eur_log:dribbling:physic:power_long_shots	0.05068
## age:dribbling:physic:power_long_shots	0.34584
## height_cm:dribbling:physic:power_long_shots	0.03685
## pace:dribbling:physic:power_long_shots	0.13149
## wage_eur_log:defending:physic:power_long_shots	0.06734
## age:defending:physic:power_long_shots	0.73006
## height_cm:defending:physic:power_long_shots	0.06964
## pace:defending:physic:power_long_shots	0.34310
## dribbling:defending:physic:power_long_shots	0.01894
## wage_eur_log:age:height_cm:pace:dribbling	0.47475
## wage_eur_log:age:height_cm:pace:defending	0.09732
## wage_eur_log:age:height_cm:dribbling:defending	0.77655
## wage_eur_log:age:pace:dribbling:defending	0.38630
## wage_eur_log:height_cm:pace:dribbling:defending	0.65155
## age:height_cm:pace:dribbling:defending	0.26919
## wage_eur_log:age:height_cm:pace:physic	0.29897
## wage_eur_log:age:height_cm:dribbling:physic	0.79277
## wage_eur_log:age:pace:dribbling:physic	0.57670
## wage_eur_log:height_cm:pace:dribbling:physic	0.76112
## age:height_cm:pace:dribbling:physic	0.43160
## wage_eur_log:age:height_cm:defending:physic	0.17559
## wage_eur_log:age:pace:defending:physic	0.04570
## wage_eur_log:height_cm:pace:defending:physic	0.74709
## age:height_cm:pace:defending:physic	0.01709
## wage_eur_log:age:dribbling:defending:physic	0.74922
## wage_eur_log:height_cm:dribbling:defending:physic	0.19133
## age:height_cm:dribbling:defending:physic	0.72180
## wage_eur_log:pace:dribbling:defending:physic	0.53591
## age:pace:dribbling:defending:physic	0.15943
## height_cm:pace:dribbling:defending:physic	0.75581
## wage_eur_log:age:height_cm:pace:power_long_shots	0.66086
## wage_eur_log:age:height_cm:dribbling:power_long_shots	0.50142
## wage_eur_log:age:pace:dribbling:power_long_shots	0.89787
## wage_eur_log:height_cm:pace:dribbling:power_long_shots	0.14941
## age:height_cm:pace:dribbling:power_long_shots	0.86158
## wage_eur_log:age:height_cm:defending:power_long_shots	0.89882
## wage_eur_log:age:pace:defending:power_long_shots	0.31740

```

## wage_eur_log:height_cm:pace:defending:power_long_shots 0.30204
## age:height_cm:pace:defending:power_long_shots 0.45442
## wage_eur_log:age:dribbling:defending:power_long_shots 0.33157
## wage_eur_log:height_cm:dribbling:defending:power_long_shots 0.01310
## age:height_cm:dribbling:defending:power_long_shots 0.28193
## wage_eur_log:pace:dribbling:defending:power_long_shots 0.05502
## age:pace:dribbling:defending:power_long_shots 0.95970
## height_cm:pace:dribbling:defending:power_long_shots 0.11523
## wage_eur_log:age:height_cm:physic:power_long_shots 0.91165
## wage_eur_log:age:pace:physic:power_long_shots 0.77996
## wage_eur_log:height_cm:pace:physic:power_long_shots 0.45827
## age:height_cm:pace:physic:power_long_shots 0.85210
## wage_eur_log:age:dribbling:physic:power_long_shots 0.45686
## wage_eur_log:height_cm:dribbling:physic:power_long_shots 0.05506
## age:height_cm:dribbling:physic:power_long_shots 0.35437
## wage_eur_log:pace:dribbling:physic:power_long_shots 0.15807
## age:pace:dribbling:physic:power_long_shots 0.78699
## height_cm:pace:dribbling:physic:power_long_shots 0.14518
## wage_eur_log:age:defending:physic:power_long_shots 0.97689
## wage_eur_log:height_cm:defending:physic:power_long_shots 0.05392
## age:height_cm:defending:physic:power_long_shots 0.70118
## wage_eur_log:pace:defending:physic:power_long_shots 0.20394
## age:pace:defending:physic:power_long_shots 0.39408
## height_cm:pace:defending:physic:power_long_shots 0.32766
## wage_eur_log:dribbling:defending:physic:power_long_shots 0.00583
## age:dribbling:defending:physic:power_long_shots 0.23240
## height_cm:dribbling:defending:physic:power_long_shots 0.01364
## pace:dribbling:defending:physic:power_long_shots 0.08992
## wage_eur_log:age:height_cm:pace:dribbling:defending 0.36836
## wage_eur_log:age:height_cm:pace:dribbling:physic 0.52042
## wage_eur_log:age:height_cm:pace:defending:physic 0.03472
## wage_eur_log:age:height_cm:dribbling:defending:physic 0.70269
## wage_eur_log:age:pace:dribbling:defending:physic 0.26482
## wage_eur_log:height_cm:pace:dribbling:defending:physic 0.60686
## age:height_cm:pace:dribbling:defending:physic 0.13686
## wage_eur_log:age:height_cm:pace:dribbling:power_long_shots 0.94007
## wage_eur_log:age:height_cm:pace:defending:power_long_shots 0.33565
## wage_eur_log:age:height_cm:dribbling:defending:power_long_shots 0.30932
## wage_eur_log:age:pace:dribbling:defending:power_long_shots 0.93598
## wage_eur_log:height_cm:pace:dribbling:defending:power_long_shots 0.05899
## age:height_cm:pace:dribbling:defending:power_long_shots 0.92577
## wage_eur_log:age:height_cm:pace:physic:power_long_shots 0.73164
## wage_eur_log:age:height_cm:dribbling:physic:power_long_shots 0.48644
## wage_eur_log:age:pace:dribbling:physic:power_long_shots 0.84396
## wage_eur_log:height_cm:pace:dribbling:physic:power_long_shots 0.18546
## age:height_cm:pace:dribbling:physic:power_long_shots 0.82755
## wage_eur_log:age:height_cm:defending:physic:power_long_shots 0.95839
## wage_eur_log:age:pace:defending:physic:power_long_shots 0.11660
## wage_eur_log:height_cm:pace:defending:physic:power_long_shots 0.20317
## age:height_cm:pace:defending:physic:power_long_shots 0.37745
## wage_eur_log:age:dribbling:defending:physic:power_long_shots 0.16671
## wage_eur_log:height_cm:dribbling:defending:physic:power_long_shots 0.00370
## age:height_cm:dribbling:defending:physic:power_long_shots 0.19166
## wage_eur_log:pace:dribbling:defending:physic:power_long_shots 0.01177
## age:pace:dribbling:defending:physic:power_long_shots 0.96462
## height_cm:pace:dribbling:defending:physic:power_long_shots 0.07985
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic 0.22401
## wage_eur_log:age:height_cm:pace:dribbling:defending:power_long_shots 0.93375
## wage_eur_log:age:height_cm:pace:dribbling:physic:power_long_shots 0.90311
## wage_eur_log:age:height_cm:pace:defending:physic:power_long_shots 0.09579
## wage_eur_log:age:height_cm:dribbling:defending:physic:power_long_shots 0.12157
## wage_eur_log:age:pace:dribbling:defending:physic:power_long_shots 0.83062
## wage_eur_log:height_cm:pace:dribbling:defending:physic:power_long_shots 0.00960
## age:height_cm:pace:dribbling:defending:physic:power_long_shots 0.95920
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic:power_long_shots NA
##
## (Intercept)
## wage_eur_log
## age
## height_cm
## pace
## dribbling
## defending
## physic
## power_long_shots .
## I(wage_eur_log^2) ***
## I(age^2) ***
## I(height_cm^2)
## I(pace^2)
## I(dribbling^2) .
## I(defending^2) ***
## I(physic^2)
## I(power_long_shots^2)
## wage_eur_log:age

```

```

## wage_eur_log:height_cm
## age:height_cm
## wage_eur_log:pace
## age:pace
## height_cm:pace
## wage_eur_log:dribbling
## age:dribbling
## height_cm:dribbling
## pace:dribbling
## wage_eur_log:defending
## age:defending
## height_cm:defending
## pace:defending
## dribbling:defending
## wage_eur_log:physic
## age:physic
## height_cm:physic
## pace:physic
## dribbling:physic
## defending:physic
## wage_eur_log:power_long_shots
## age:power_long_shots
## height_cm:power_long_shots .
## pace:power_long_shots
## dribbling:power_long_shots *
## defending:power_long_shots
## physic:power_long_shots
## wage_eur_log:age:height_cm
## wage_eur_log:age:pace
## wage_eur_log:height_cm:pace
## age:height_cm:pace
## wage_eur_log:age:dribbling
## wage_eur_log:height_cm:dribbling
## age:height_cm:dribbling
## wage_eur_log:pace:dribbling
## age:pace:dribbling
## height_cm:pace:dribbling
## wage_eur_log:age:defending
## wage_eur_log:height_cm:defending
## age:height_cm:defending
## wage_eur_log:pace:defending
## age:pace:defending .
## height_cm:pace:defending
## wage_eur_log:dribbling:defending
## age:dribbling:defending
## height_cm:dribbling:defending
## pace:dribbling:defending
## wage_eur_log:age:physic
## wage_eur_log:height_cm:physic
## age:height_cm:physic
## wage_eur_log:pace:physic
## age:pace:physic
## height_cm:pace:physic
## wage_eur_log:dribbling:physic
## age:dribbling:physic
## height_cm:dribbling:physic
## pace:dribbling:physic
## wage_eur_log:defending:physic
## age:defending:physic
## height_cm:defending:physic
## pace:defending:physic
## dribbling:defending:physic
## wage_eur_log:age:power_long_shots
## wage_eur_log:height_cm:power_long_shots
## age:height_cm:power_long_shots
## wage_eur_log:pace:power_long_shots
## age:pace:power_long_shots
## height_cm:pace:power_long_shots
## wage_eur_log:dribbling:power_long_shots *
## age:dribbling:power_long_shots
## height_cm:dribbling:power_long_shots *
## pace:dribbling:power_long_shots .
## wage_eur_log:defending:power_long_shots
## age:defending:power_long_shots
## height_cm:defending:power_long_shots .
## pace:defending:power_long_shots
## dribbling:defending:power_long_shots *
## wage_eur_log:physic:power_long_shots
## age:physic:power_long_shots
## height_cm:physic:power_long_shots
## pace:physic:power_long_shots
## dribbling:physic:power_long_shots *
## defending:physic:power_long_shots .
## wage_eur_log:age:height_cm:pace

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## wage_eur_log:age:height_cm:dribbling
## wage_eur_log:age:pace:dribbling
## wage_eur_log:height_cm:pace:dribbling
## age:height_cm:pace:dribbling
## wage_eur_log:age:height_cm:defending
## wage_eur_log:age:pace:defending
## wage_eur_log:height_cm:pace:defending
## age:height_cm:pace:defending
## wage_eur_log:age:dribbling:defending
## wage_eur_log:height_cm:dribbling:defending
## age:height_cm:dribbling:defending
## wage_eur_log:pace:dribbling:defending
## age:pace:dribbling:defending
## height_cm:pace:dribbling:defending
## wage_eur_log:age:height_cm:physic
## wage_eur_log:age:pace:physic
## wage_eur_log:height_cm:pace:physic
## age:height_cm:pace:physic
## wage_eur_log:age:dribbling:physic
## wage_eur_log:height_cm:dribbling:physic
## age:height_cm:dribbling:physic
## wage_eur_log:pace:dribbling:physic
## age:pace:dribbling:physic
## height_cm:pace:dribbling:physic
## wage_eur_log:age:defending:physic
## wage_eur_log:height_cm:defending:physic
## age:height_cm:defending:physic
## wage_eur_log:pace:defending:physic
## age:pace:defending:physic
## height_cm:pace:defending:physic
## wage_eur_log:dribbling:defending:physic
## age:dribbling:defending:physic
## height_cm:dribbling:defending:physic
## pace:dribbling:defending:physic
## wage_eur_log:age:height_cm:power_long_shots
## wage_eur_log:age:pace:power_long_shots
## wage_eur_log:height_cm:pace:power_long_shots
## age:height_cm:pace:power_long_shots
## wage_eur_log:age:dribbling:power_long_shots
## wage_eur_log:height_cm:dribbling:power_long_shots
## age:height_cm:dribbling:power_long_shots
## wage_eur_log:pace:dribbling:power_long_shots
## age:pace:dribbling:power_long_shots
## height_cm:pace:dribbling:power_long_shots
## wage_eur_log:age:defending:power_long_shots
## wage_eur_log:height_cm:defending:power_long_shots
## age:height_cm:defending:power_long_shots
## wage_eur_log:pace:defending:power_long_shots
## age:pace:defending:power_long_shots
## height_cm:pace:defending:power_long_shots
## wage_eur_log:dribbling:defending:power_long_shots
## age:dribbling:defending:power_long_shots
## height_cm:dribbling:defending:power_long_shots
## pace:dribbling:defending:power_long_shots
## wage_eur_log:age:physic:power_long_shots
## wage_eur_log:height_cm:physic:power_long_shots
## age:height_cm:physic:power_long_shots
## wage_eur_log:pace:physic:power_long_shots
## age:pace:physic:power_long_shots
## height_cm:pace:physic:power_long_shots
## wage_eur_log:dribbling:physic:power_long_shots
## age:dribbling:physic:power_long_shots
## height_cm:dribbling:physic:power_long_shots
## pace:dribbling:physic:power_long_shots
## wage_eur_log:defending:physic:power_long_shots
## age:defending:physic:power_long_shots
## height_cm:defending:physic:power_long_shots
## pace:defending:physic:power_long_shots
## dribbling:defending:physic:power_long_shots
## wage_eur_log:age:height_cm:pace:dribbling
## wage_eur_log:age:height_cm:pace:defending
## wage_eur_log:age:height_cm:dribbling:defending
## wage_eur_log:age:pace:dribbling:defending
## wage_eur_log:height_cm:pace:dribbling:defending
## age:height_cm:pace:dribbling:defending
## wage_eur_log:age:height_cm:pace:physic
## wage_eur_log:age:height_cm:dribbling:physic
## wage_eur_log:age:pace:dribbling:physic
## wage_eur_log:height_cm:pace:dribbling:physic
## age:height_cm:pace:dribbling:physic
## wage_eur_log:age:height_cm:defending:physic
## wage_eur_log:age:pace:defending:physic
## wage_eur_log:height_cm:pace:defending:physic
## age:height_cm:pace:defending:physic

```

```

## wage_eur_log:age:dribbling:defending:physic
## wage_eur_log:height_cm:dribbling:defending:physic
## age:height_cm:dribbling:defending:physic
## wage_eur_log:pace:dribbling:defending:physic
## age:pace:dribbling:defending:physic
## height_cm:pace:dribbling:defending:physic
## wage_eur_log:age:height_cm:pace:power_long_shots
## wage_eur_log:age:height_cm:dribbling:power_long_shots
## wage_eur_log:age:pace:dribbling:power_long_shots
## wage_eur_log:height_cm:pace:dribbling:power_long_shots
## age:height_cm:pace:dribbling:power_long_shots
## wage_eur_log:age:height_cm:defending:power_long_shots
## wage_eur_log:age:pace:defending:power_long_shots
## wage_eur_log:height_cm:pace:defending:power_long_shots
## age:height_cm:pace:defending:power_long_shots
## wage_eur_log:age:dribbling:defending:power_long_shots
## wage_eur_log:height_cm:dribbling:defending:power_long_shots *
## age:height_cm:dribbling:defending:power_long_shots
## wage_eur_log:pace:dribbling:defending:power_long_shots .
## age:pace:dribbling:defending:power_long_shots
## height_cm:pace:dribbling:defending:power_long_shots
## wage_eur_log:age:height_cm:physic:power_long_shots
## wage_eur_log:age:pace:physic:power_long_shots
## wage_eur_log:height_cm:pace:physic:power_long_shots
## age:height_cm:pace:physic:power_long_shots
## wage_eur_log:age:dribbling:physic:power_long_shots
## wage_eur_log:height_cm:dribbling:physic:power_long_shots .
## age:height_cm:dribbling:physic:power_long_shots
## wage_eur_log:pace:dribbling:physic:power_long_shots
## age:pace:dribbling:physic:power_long_shots
## height_cm:pace:dribbling:physic:power_long_shots
## wage_eur_log:age:defending:physic:power_long_shots
## wage_eur_log:height_cm:defending:physic:power_long_shots .
## age:height_cm:defending:physic:power_long_shots
## wage_eur_log:pace:defending:physic:power_long_shots
## age:pace:defending:physic:power_long_shots
## height_cm:pace:defending:physic:power_long_shots
## wage_eur_log:dribbling:defending:physic:power_long_shots **
## age:dribbling:defending:physic:power_long_shots
## height_cm:dribbling:defending:physic:power_long_shots *
## pace:dribbling:defending:physic:power_long_shots .
## wage_eur_log:age:height_cm:pace:dribbling:defending
## wage_eur_log:age:height_cm:pace:dribbling:physic
## wage_eur_log:age:height_cm:pace:defending:physic *
## wage_eur_log:age:height_cm:dribbling:defending:physic
## wage_eur_log:age:pace:dribbling:defending:physic
## wage_eur_log:height_cm:pace:dribbling:defending:physic
## age:height_cm:pace:dribbling:defending:physic
## wage_eur_log:age:height_cm:pace:dribbling:power_long_shots
## wage_eur_log:age:height_cm:pace:defending:power_long_shots
## wage_eur_log:age:height_cm:dribbling:defending:power_long_shots
## wage_eur_log:age:pace:dribbling:defending:power_long_shots
## wage_eur_log:height_cm:pace:dribbling:defending:power_long_shots .
## age:height_cm:pace:dribbling:defending:power_long_shots
## wage_eur_log:age:height_cm:pace:physic:power_long_shots
## wage_eur_log:age:height_cm:dribbling:physic:power_long_shots
## wage_eur_log:age:pace:dribbling:physic:power_long_shots
## wage_eur_log:height_cm:pace:dribbling:physic:power_long_shots
## age:height_cm:pace:dribbling:physic:power_long_shots
## wage_eur_log:age:height_cm:defending:physic:power_long_shots
## wage_eur_log:age:pace:defending:physic:power_long_shots
## wage_eur_log:height_cm:pace:defending:physic:power_long_shots
## age:height_cm:pace:defending:physic:power_long_shots
## wage_eur_log:age:dribbling:defending:physic:power_long_shots
## wage_eur_log:height_cm:dribbling:defending:physic:power_long_shots **
## age:height_cm:dribbling:defending:physic:power_long_shots
## wage_eur_log:pace:dribbling:defending:physic:power_long_shots *
## age:pace:dribbling:defending:physic:power_long_shots
## height_cm:pace:dribbling:defending:physic:power_long_shots .
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic
## wage_eur_log:age:height_cm:pace:dribbling:defending:power_long_shots
## wage_eur_log:age:height_cm:pace:dribbling:physic:power_long_shots
## wage_eur_log:age:height_cm:pace:defending:physic:power_long_shots .
## wage_eur_log:age:height_cm:dribbling:defending:physic:power_long_shots
## wage_eur_log:age:pace:dribbling:defending:physic:power_long_shots
## wage_eur_log:height_cm:pace:dribbling:defending:physic:power_long_shots **
## age:height_cm:pace:dribbling:defending:physic:power_long_shots
## wage_eur_log:age:height_cm:pace:dribbling:defending:physic:power_long_shots
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.196 on 251 degrees of freedom
## Multiple R-squared:  0.9422, Adjusted R-squared:  0.8818
## F-statistic: 15.61 on 262 and 251 DF,  p-value: < 0.0000000000000022

```

Coefficients  $I(\text{wage\_eur\_log}^2)$ ,  $I(\text{age}^2)$ ,  $I(\text{defending}^2)$ ,  $\text{dribbling}:\text{power\_long\_shots}$  are significant here. There are other significant coefficients as well but they will not be used because it makes the model too complex (e.g.  $\text{wage\_eur\_log}:\text{dribbling}:\text{power\_long\_shots}$ )

### model1

Starting with a very simple model:

```
mr_model1 <- lm(potential ~ wage_eur_log +
                age +
                height_cm +
                pace +
                dribbling +
                defending +
                physic +
                power_long_shots, data=df)

summary(mr_model1)
```

```
##
## Call:
## lm(formula = potential ~ wage_eur_log + age + height_cm + pace +
##     dribbling + defending + physic + power_long_shots, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.5916  -2.4986  -0.3558   2.2130  12.2437
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)  37.51410     6.01978   6.232 0.000000000973 ***
## wage_eur_log   1.53614     0.15548   9.880 < 0.000000000000002 ***
## age          -0.85509     0.04340  -19.704 < 0.000000000000002 ***
## height_cm      0.07956     0.03186   2.497    0.0128 *
## pace          -0.03197     0.01965  -1.627    0.1045
## dribbling      0.31234     0.03173   9.844 < 0.000000000000002 ***
## defending        0.07029     0.01333   5.274 0.00000197843 ***
## physic         0.10910     0.02611   4.179 0.000034561395 ***
## power_long_shots 0.01020     0.01707   0.598    0.5503
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.677 on 505 degrees of freedom
## Multiple R-squared:  0.6741, Adjusted R-squared:  0.6689
## F-statistic: 130.6 on 8 and 505 DF,  p-value: < 0.0000000000000022
```

### model2

`power_long_shots` has the least significant coefficient. It will be removed.

```
mr_model2 <- update(mr_model1, ~. - power_long_shots)

summary(mr_model2)
```

```
##
## Call:
## lm(formula = potential ~ wage_eur_log + age + height_cm + pace +
##     dribbling + defending + physic, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.6786  -2.4719  -0.3568   2.2162  12.4024
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)  37.44423     6.01482   6.225 0.00000000101 ***
## wage_eur_log   1.54437     0.15477   9.979 < 0.000000000000002 ***
## age          -0.85094     0.04281  -19.876 < 0.000000000000002 ***
## height_cm      0.07883     0.03182   2.477    0.0136 *
## pace          -0.03466     0.01912  -1.812    0.0705 .
## dribbling      0.32428     0.02465  13.158 < 0.000000000000002 ***
## defending        0.06771     0.01260   5.375 0.0000011729 ***
## physic         0.11099     0.02590   4.285 0.00002186890 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.674 on 506 degrees of freedom
## Multiple R-squared:  0.6739, Adjusted R-squared:  0.6694
## F-statistic: 149.4 on 7 and 506 DF,  p-value: < 0.0000000000000022
```

### model3

`pace` has the least significant coefficient. It will be removed.

```
mr_model3 <- update(mr_model2,~.-pace)
summary(mr_model3)
```

```
##
## Call:
## lm(formula = potential ~ wage_eur_log + age + height_cm + dribbling +
##     defending + physic, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.4664  -2.5034  -0.2909   2.2721  12.3385
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)  33.00158    5.50507   5.995 0.0000000387 ***
## wage_eur_log   1.56838    0.15455  10.148 < 0.000000000000002 ***
## age          -0.83232    0.04165 -19.982 < 0.000000000000002 ***
## height_cm      0.09518    0.03058   3.112   0.00196 **
## dribbling      0.30461    0.02218  13.736 < 0.000000000000002 ***
## defending        0.07092    0.01250   5.674   0.0000002347 ***
## physic         0.10384    0.02566   4.047   0.0005990173 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.683 on 507 degrees of freedom
## Multiple R-squared:  0.6718, Adjusted R-squared:  0.6679
## F-statistic: 172.9 on 6 and 507 DF,  p-value: < 0.0000000000000022
```

#### model4

I(wage\_eur\_log^2) will be added since it has been shown in the complex model that it possesses a significant coefficient.

```
mr_model4 <- update(mr_model3,~.+I(wage_eur_log^2))
summary(mr_model4)
```

```
##
## Call:
## lm(formula = potential ~ wage_eur_log + age + height_cm + dribbling +
##     defending + physic + I(wage_eur_log^2), data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.9439  -2.4533  -0.1034   2.2268  11.9656
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)  72.71138    7.52941   9.657 < 0.000000000000002 ***
## wage_eur_log  -7.11112    1.19123  -5.970   0.00000004480176 ***
## age          -0.81786    0.03969 -20.608 < 0.000000000000002 ***
## height_cm      0.07950    0.02918   2.724   0.00667 **
## dribbling      0.27281    0.02154  12.664 < 0.000000000000002 ***
## defending        0.05894    0.01201   4.909   0.000001235064317 ***
## physic         0.11235    0.02444   4.596   0.000005433483767 ***
## I(wage_eur_log^2) 0.51938    0.07074   7.342   0.000000000000844 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.504 on 506 degrees of freedom
## Multiple R-squared:  0.7034, Adjusted R-squared:  0.6993
## F-statistic: 171.4 on 7 and 506 DF,  p-value: < 0.0000000000000022
```

#### model5

I(age^2) will be added since it has been shown in the complex model that it possesses a significant coefficient.

```
mr_model5 <- update(mr_model4,~.+I(age^2))
summary(mr_model5)
```

```
##
## Call:
## lm(formula = potential ~ wage_eur_log + age + height_cm + dribbling +
##     defending + physic + I(wage_eur_log^2) + I(age^2), data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.5296  -2.0687  -0.1365   1.9882  11.5235
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)    112.32393     7.21462   15.569 < 0.0000000000000002 ***
## wage_eur_log     -4.66319     1.05065    -4.438  0.0000111381845827 ***
## age             -4.96452     0.32246   -15.396 < 0.0000000000000002 ***
## height_cm        0.06487     0.02534     2.560    0.0108 *
## dribbling        0.30173     0.01882    16.030 < 0.0000000000000002 ***
## defending          0.06045     0.01042     5.804  0.0000000114695501 ***
## physic           0.17157     0.02169     7.908  0.0000000000000165 ***
## I(wage_eur_log^2) 0.37698     0.06235     6.046  0.0000000028856705 ***
## I(age^2)         0.07864     0.00608    12.934 < 0.0000000000000002 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.04 on 505 degrees of freedom
## Multiple R-squared:  0.7772, Adjusted R-squared:  0.7736
## F-statistic: 220.2 on 8 and 505 DF,  p-value: < 0.0000000000000022
```

### model6

I(defending^2) will be added since it has been shown in the complex model that it posses a significant coefficient.

```
mr_model6 <- update(mr_model5,~.+I(defending^2))
summary(mr_model6)
```

```
##
## Call:
## lm(formula = potential ~ wage_eur_log + age + height_cm + dribbling +
##     defending + physic + I(wage_eur_log^2) + I(age^2) + I(defending^2),
##     data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.0805  -2.0434  -0.1617   2.0537  10.3276
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)    114.1618122     7.0543648   16.183 < 0.0000000000000002 ***
## wage_eur_log     -3.6916435     1.0437626    -3.537  0.000442 ***
## age             -4.9081759     0.3150715   -15.578 < 0.0000000000000002 ***
## height_cm        0.0636173     0.0247475     2.571    0.010436 *
## dribbling        0.3124827     0.0185030    16.888 < 0.0000000000000002 ***
## defending         -0.2199382     0.0563329    -3.904    0.000107 ***
## physic           0.1647190     0.0212273     7.760  0.0000000000000476 ***
## I(wage_eur_log^2) 0.3073120     0.0624208     4.923  0.0000011554437612 ***
## I(age^2)         0.0776776     0.0059405    13.076 < 0.0000000000000002 ***
## I(defending^2)    0.0029532     0.0005836     5.061  0.0000005867605736 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.969 on 504 degrees of freedom
## Multiple R-squared:  0.7879, Adjusted R-squared:  0.7842
## F-statistic: 208.1 on 9 and 504 DF,  p-value: < 0.0000000000000022
```

### model7

dribbling:power\_long\_shots will be added since it has been shown in the complex model that it posses a significant coefficient.

```
mr_model7 <- update(mr_model6,~. + dribbling:power_long_shots)
summary(mr_model7)
```

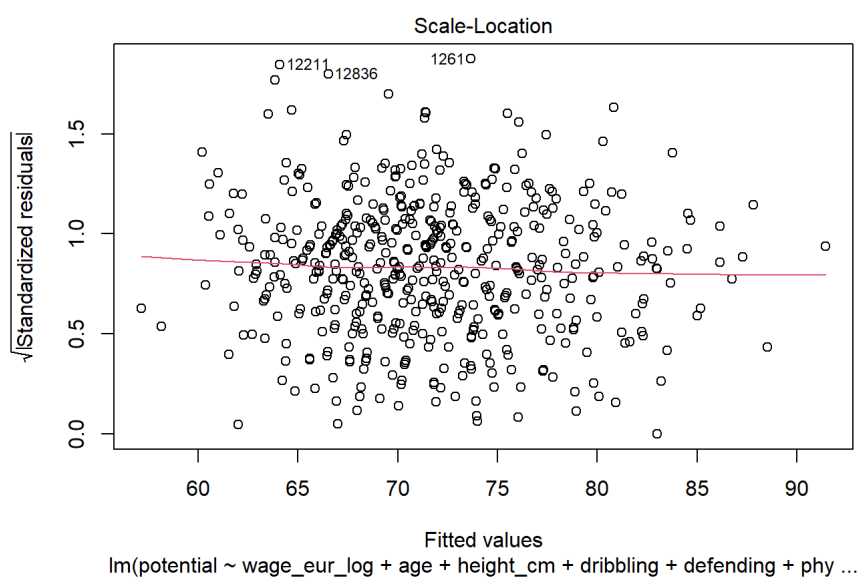
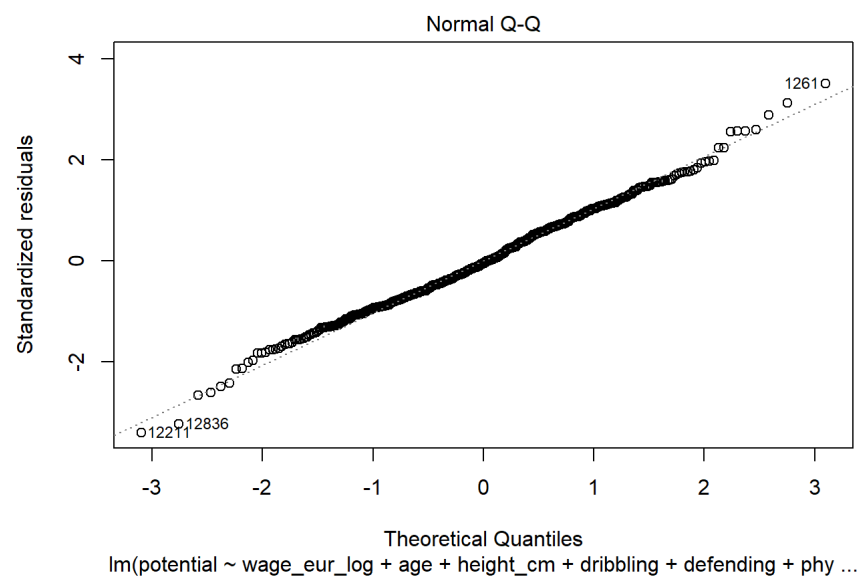
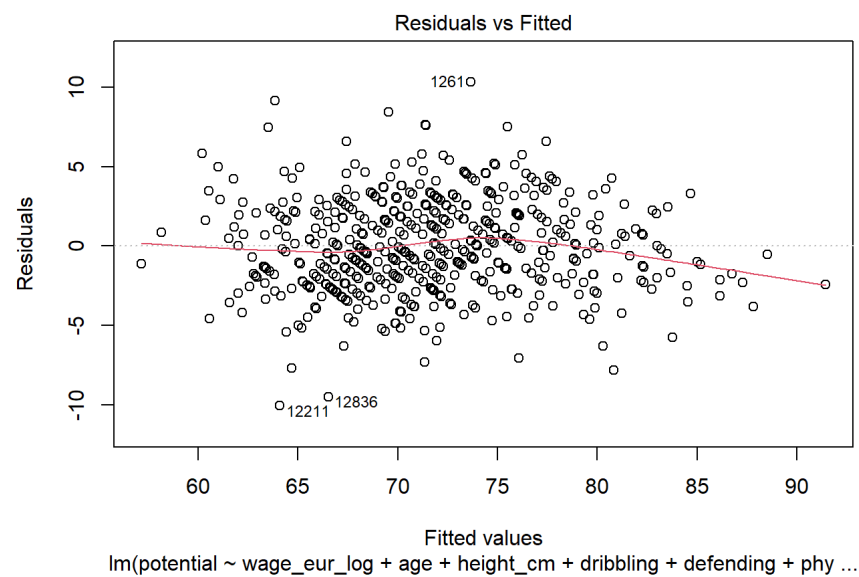


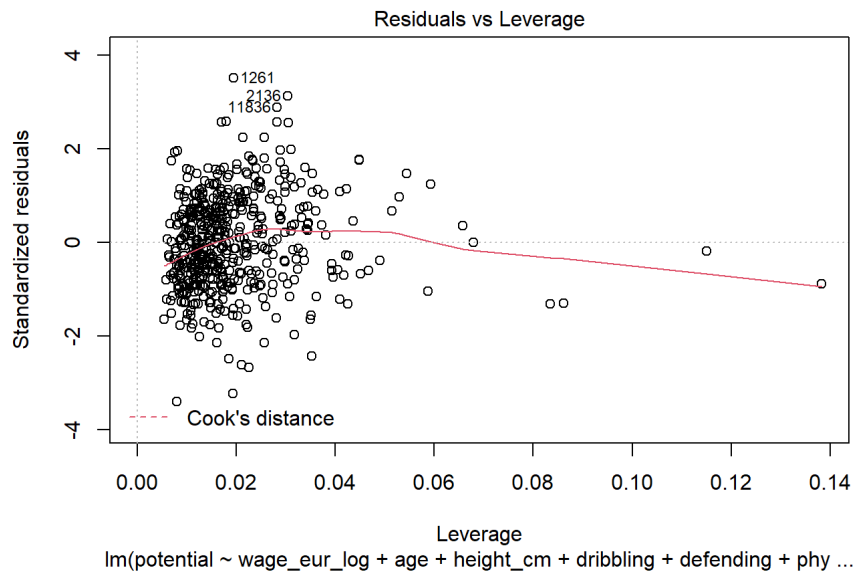
```
##
## Call:
## lm(formula = potential ~ wage_eur_log + age + height_cm + dribbling +
##     defending + physic + I(wage_eur_log^2) + I(age^2) + I(defending^2) +
##     dribbling:power_long_shots, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.9620 -2.0128 -0.1722  1.9967 10.1432
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    114.9109253    7.0808954   16.228 < 0.0000000000000002
## wage_eur_log     -3.6360710    1.0444681   -3.481    0.000542
## age             -4.8902573    0.3153306  -15.508 < 0.0000000000000002
## height_cm        0.0630427    0.0247434    2.548    0.011135
## dribbling        0.2868151    0.0287175    9.987 < 0.0000000000000002
## defending        -0.2210063    0.0563199   -3.924    0.000099154674066
## physic          0.1623905    0.0213129    7.619    0.0000000000000128
## I(wage_eur_log^2)  0.3021367    0.0625552    4.830    0.000001815737424
## I(age^2)         0.0771825    0.0059534   12.964 < 0.0000000000000002
## I(defending^2)    0.0030008    0.0005848    5.132    0.000000411107175
## dribbling:power_long_shots  0.0002408    0.0002061    1.168    0.243190
##
## (Intercept)          ***
## wage_eur_log          ***
## age                   ***
## height_cm             *
## dribbling             ***
## defending              ***
## physic               ***
## I(wage_eur_log^2)     ***
## I(age^2)              ***
## I(defending^2)        ***
## dribbling:power_long_shots
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.968 on 503 degrees of freedom
## Multiple R-squared:  0.7885, Adjusted R-squared:  0.7843
## F-statistic: 187.5 on 10 and 503 DF, p-value: < 0.0000000000000022
```

Model 7 shows that `dribbling:power_long_shots` is not significant so we will keep **model6** as our final model.

Analyzing model6:

```
plot(mr_model6)
```





In the graph "Residuals vs Fitted", residuals are randomly distributed which means that there are not signs of heteroscedasticity. The Normal Q-Q plot shows that the standardized residuals do follow a normal distribution which means our model is good.

## 3.2 Critique model using relevant diagnostics

Offer an interpretation of the model characteristics, goodness of fit and graphical diagnostics (5 marks) for the model built in 3.1. Explain any potential weaknesses (5 marks).

- First, I built a very complex model (model0) using the variables `wage_eur_log`, `age`, `height_cm`, `pace`, `dribbling`, `defending`, `physic`, `power_long_shots`, their quadratic values and their interactions between them. They were used since they have been shown to be relevant during the EDA. The output of the model was an R-squared of 0.9422 but a very low F-statistic of 15.61. However, I found that coefficients for  $I(\text{wage\_eur\_log}^2)$ ,  $I(\text{age}^2)$ ,  $I(\text{defending}^2)$ , `dribbling:power_long_shots` are significant. Others were too but we ignore them since adding them would make the model very complex.
- Second, I built a very simple model (model1) using the variables `wage_eur_log`, `age`, `height_cm`, `pace`, `dribbling`, `defending`, `physic`, `power_long_shots`. We found that `pace` and `power_long_shots` were the least significant. This model gave us a lower R-squared of 0.6741 but a significant higher F-statistic of 130.6 (improvement of 114.99).
- Third, I built a model (model2) by removing the `power_long_shots` column from model1 since it was no significant in model1. The R-squared slightly decreased to 0.6739 but the F-statistic increased to 149.4. Column `pace` remain to be not significant.
- Fourth, I built a model (model3) by removing the `pace` column from model2 since it was no significant in model2. The R-squared slightly increased from model2 to 0.6718 but still lower than model1 (0.6741). The F-statistic increased to 172.9.
- Fifth, I built a model (model4) by adding  $I(\text{wage\_eur\_log}^2)$  to model3. The R-squared increased to 0.7034 (the highest so far) and the F-statistic slightly decreased to 171.4.
- Sixth, I built a model (model5) by adding  $I(\text{age}^2)$  to model4. The R-squared increased to 0.7772 (the highest so far) and the F-statistic had a big increase to 220.2 (highest so far).
- Seventh, I built a model (model6) by adding  $I(\text{defending}^2)$  to model5. The R-squared slightly increased to 0.7879 (highest so far). However, the F-statistic decreased to 208.1.
- Finally, I built a model (model7) by adding `dribbling:power_long_shots` to model6. However, this column resulted to be no significant. Not only that, it led to a significant decrease in both R-squared and F-statistic.
- Therefore our final model will be model6 which had the highest R-squared and the second highest F-statistic from all my models. Furthermore, the graph "Residuals vs Fitted" shows that residuals are randomly distributed which means that there are not signs of heteroscedasticity and the Normal Q-Q plot shows that the standardized residuals do follow a normal distribution which means our model is good.
- Our final model is interpreted as follows:

$$\text{potential} = 114.1618122 - 3.6916435 \times \log(\text{wage\_eur}) - 4.9081759 \times \text{age} + 0.0636173 \times \text{height\_cm} + 0.3124827 \times \text{dribbling} - 0.219938$$

- The multi-regression coefficients gives you the size of effect that the feature is having on potential and the sign gives you whether that effect is positive or negative (e.g., for every unit increase in age there is 4.9 decrease in potential).

## 3.3 Suggest improvements to your model

Based on the findings in 3.2 articulates possible alternative approaches to address them (5 marks).

Since Multiple Regression is an algorithm that uses distance between points for the loss function, variables which range are higher (e.g., `wage_eur_log` range is significantly smaller than the range in `height_cm`) will have a bigger impact in the model without being necessary more important than the other variables. A good way to fix this is to standardize numerical columns. In that way, your columns will follow a normal distribution and have a mean of 0 and a standard deviation of 1. In other words, they will have similar scale and will help the algorithm remove that bias.

## 4. Extension work

### 4.1 Model the likelihood of a player having a weekly wage above 8000 Euro (using the high.wage.ind variable provided).

Given this second research question (i.e., involving the binary target attribute) provide a plan of analysis based on relevant EDA for this attribute (10 marks). The model is described, explained and critiqued (10 marks). NB Submissions where suitable models do not have good fit due to the nature of the data will not be penalized.

Plan:

- Column `wage_eur_log` will not be used in the model because `high.wage.ind` is calculated from `wage_eur`. Therefore, when building the model, `wage_eur_log` will perfectly separate 1 and 0s and will lead the model to not converge. [7]
- From the EDA process, I have found that columns `power_long_shots`, `power_strength`, `physic`, `defending`, `dribbling`, `shooting`, `pace`, `weight_kg`, `wage_eur_log`, `age`, `potential` do not follow a normal distribution. Only columns `height_cm`, and `passing` follow a normal distribution.
- In order to avoid collinearity, features that are highly correlated ( $\geq 0.8$ ) between each other, will not be used. `height_cm` and `weight_kg` are highly correlated, for this case `height_cm` will be used because it follows a normal distribution. `dribbling` is highly correlated with `shooting` and `passing`, for this case we will use `dribbling` and remove `shooting` and `passing` in the model. In this way, we use one feature instead of two and make the model simpler, and also because `dribbling` has more correlation with `potential` than the other 2 features. `physic` and `power_strength` are also highly correlated, I will use `physic` and not use `power_strength` because `physic` has a higher correlation with `potential` than `power_strength` with `potential`. In summary, columns `weight_kg`, `shooting`, `passing`, `power_strength` will not be used in the model.
- `pace` has been shown to not have statistical significance in the difference in means between `high.wage_ind = 0` and `high.wage_ind = 1`. I will not use this column in the model.
- `weight_kg` and `pace` have been shown to not have statistical significance in the difference in medians between `high.wage_ind = 0` and `high.wage_ind = 1`. We don't need to worry since we already said `weight_kg` will not be used to avoid collinearity and we already say we need to be careful with the `pace` column.
- First I will build a complex model and take note of the significant coefficients. Then I will build a very simple model and start adding significant coefficients found in the complex model.

model0

```
lg_model0 <- glm(high.wage.ind ~ age*height_cm*dribbling*defending*physic*power_long_shots +
                  I(age^2) +
                  I(height_cm^2) +
                  I(dribbling^2) +
                  I(defending^2) +
                  I(physic^2) +
                  I(power_long_shots^2) +
                  I(potential^2), data=df, family=binomial)
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
summary(lg_model0)
```

```
##
## Call:
## glm(formula = high.wage.ind ~ age * height_cm * dribbling * defending *
##   physic * power_long_shots + I(age^2) + I(height_cm^2) + I(dribbling^2) +
##   I(defending^2) + I(physic^2) + I(power_long_shots^2) + I(potential^2),
##   family = binomial, data = df)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5438  -0.1276  -0.0095   0.0266   2.5214
##
## Coefficients:
##                                     Estimate
## (Intercept)                    -104802.945687643
## age                             5065.505866037
## height_cm                       550.708445183
## dribbling                       1652.854974104
## defending                         935.947507931
## physic                          1878.641544470
## power_long_shots                1576.187413386
## I(age^2)                        -0.063979068
## I(height_cm^2)                   0.002322525
## I(dribbling^2)                   0.008882192
## I(defending^2)                   0.003725136
## I(physic^2)                      0.002383801
## I(power_long_shots^2)            -0.000452889
## I(potential^2)                   0.004064609
## age:height_cm                   -26.867367385
## age:dribbling                   -78.921868822
## height_cm:dribbling              -8.707971840
## age:defending                   -57.902904586
## height_cm:defending              -4.852886988
## dribbling:defending             -15.785583857
## age:physic                      -84.474018943
## height_cm:physic                -9.858653918
## dribbling:physic                -29.025406883
## defending:physic                  -18.762725039
## age:power_long_shots            -79.464142280
## height_cm:power_long_shots       -8.275082835
## dribbling:power_long_shots       -25.329073554
## defending:power_long_shots        -12.347570355
## physic:power_long_shots          -28.733485006
## age:height_cm:dribbling          0.418933033
## age:height_cm:defending           0.303951247
## age:dribbling:defending           0.912710711
## height_cm:dribbling:defending     0.081821250
## age:height_cm:physic             0.446539130
## age:dribbling:physic              1.305326838
## height_cm:dribbling:physic        0.152339634
## age:defending:physic              0.996685213
## height_cm:defending:physic        0.097326785
## dribbling:defending:physic        0.299600971
## age:height_cm:power_long_shots    0.421839868
## age:dribbling:power_long_shots    1.250206036
## height_cm:dribbling:power_long_shots 0.133300766
## age:defending:power_long_shots    0.877933090
## height_cm:defending:power_long_shots 0.063931680
## dribbling:defending:power_long_shots 0.225766343
## age:physic:power_long_shots       1.329560143
## height_cm:physic:power_long_shots 0.150437571
## dribbling:physic:power_long_shots 0.450780614
## defending:physic:power_long_shots   0.263605106
## age:height_cm:dribbling:defending -0.004775410
## age:height_cm:dribbling:physic    -0.006901944
## age:height_cm:defending:physic    -0.005219596
## age:dribbling:defending:physic    -0.015481253
## height_cm:dribbling:defending:physic -0.001552512
## age:height_cm:dribbling:power_long_shots -0.006643097
## age:height_cm:defending:power_long_shots -0.004628939
## age:dribbling:defending:power_long_shots -0.014294147
## height_cm:dribbling:defending:power_long_shots -0.001175531
## age:height_cm:physic:power_long_shots -0.007023534
## age:dribbling:physic:power_long_shots -0.020721362
## height_cm:dribbling:physic:power_long_shots -0.002364090
## age:defending:physic:power_long_shots -0.015188846
## height_cm:defending:physic:power_long_shots -0.001364800
## dribbling:defending:physic:power_long_shots -0.004440696
## age:height_cm:dribbling:defending:physic 0.000080878
## age:height_cm:dribbling:defending:power_long_shots 0.000075310
## age:height_cm:dribbling:physic:power_long_shots 0.000109595
## age:height_cm:defending:physic:power_long_shots 0.000079716
## age:dribbling:defending:physic:power_long_shots 0.000242649
## height_cm:dribbling:defending:physic:power_long_shots 0.000023066
```

```

## age:height_cm:dribbling:defending:physic:power_long_shots      -0.000001274
##                               Std. Error
## (Intercept)                                135418.612237753
## age                                           5276.956841708
## height_cm                                   745.845280639
## dribbling                                   1976.668166054
## defending                                   2096.610294464
## physic                                    2183.865143696
## power_long_shots                             2306.738032733
## I(age^2)                                     0.019368303
## I(height_cm^2)                               0.007423794
## I(dribbling^2)                              0.009734737
## I(defending^2)                              0.001456330
## I(physic^2)                                 0.004622177
## I(power_long_shots^2)                       0.001969161
## I(potential^2)                             0.000771771
## age:height_cm                               29.185765624
## age:dribbling                              77.201417306
## height_cm:dribbling                       10.895064995
## age:defending                              81.724976387
## height_cm:defending                       11.540210209
## dribbling:defending                       30.887492019
## age:physic                                84.226520790
## height_cm:physic                          11.975421007
## dribbling:physic                          31.896950820
## defending:physic                           32.814751850
## age:power_long_shots                      89.866121064
## height_cm:power_long_shots                12.693723134
## dribbling:power_long_shots                33.473916643
## defending:power_long_shots                 36.050875450
## physic:power_long_shots                   36.856371310
## age:height_cm:dribbling                   0.427309032
## age:height_cm:defending                   0.451460316
## age:dribbling:defending                   1.206044413
## height_cm:dribbling:defending             0.170373429
## age:height_cm:physic                     0.463373162
## age:dribbling:physic                     1.232081610
## height_cm:dribbling:physic                0.175024904
## age:defending:physic                     1.266252832
## height_cm:defending:physic               0.180022446
## dribbling:defending:physic               0.483524559
## age:height_cm:power_long_shots           0.496177307
## age:dribbling:power_long_shots           1.305359609
## height_cm:dribbling:power_long_shots     0.184379318
## age:defending:power_long_shots           1.404063364
## height_cm:defending:power_long_shots     0.198278784
## dribbling:defending:power_long_shots     0.527728112
## age:physic:power_long_shots              1.419044162
## height_cm:physic:power_long_shots        0.202059382
## dribbling:physic:power_long_shots        0.535302914
## defending:physic:power_long_shots          0.558935651
## age:height_cm:dribbling:defending        0.006674756
## age:height_cm:dribbling:physic           0.006782747
## age:height_cm:defending:physic           0.006967519
## age:dribbling:defending:physic           0.018688359
## height_cm:dribbling:defending:physic     0.002657522
## age:height_cm:dribbling:power_long_shots 0.007213850
## age:height_cm:defending:power_long_shots 0.007743780
## age:dribbling:defending:power_long_shots 0.020565429
## height_cm:dribbling:defending:power_long_shots 0.002908870
## age:height_cm:physic:power_long_shots    0.007798884
## age:dribbling:physic:power_long_shots    0.020621541
## height_cm:dribbling:physic:power_long_shots 0.002937000
## age:defending:physic:power_long_shots    0.021530624
## height_cm:defending:physic:power_long_shots 0.003065135
## dribbling:defending:physic:power_long_shots 0.008186648
## age:height_cm:dribbling:defending:physic 0.000103005
## age:height_cm:dribbling:defending:power_long_shots 0.000113644
## age:height_cm:dribbling:physic:power_long_shots 0.000113423
## age:height_cm:defending:physic:power_long_shots 0.000118335
## age:dribbling:defending:physic:power_long_shots 0.000315523
## height_cm:dribbling:defending:physic:power_long_shots 0.000044983
## age:height_cm:dribbling:defending:physic:power_long_shots 0.000001737
##                               z value      Pr(>|z|)
## (Intercept)                   -0.774      0.438979
## age                           0.960      0.337091
## height_cm                     0.738      0.460291
## dribbling                     0.836      0.403052
## defending                      0.446      0.655301
## physic                        0.860      0.389658
## power_long_shots              0.683      0.494419
## I(age^2)                      -3.303      0.000956
## I(height_cm^2)                 0.313      0.754396
## I(dribbling^2)                 0.912      0.361546

```

## I(defending^2)	2.558	0.010531
## I(physic^2)	0.516	0.606042
## I(power_long_shots^2)	-0.230	0.818099
## I(potential^2)	5.267	0.00000139
## age:height_cm	-0.921	0.357278
## age:dribbling	-1.022	0.306646
## height_cm:dribbling	-0.799	0.424141
## age:defending	-0.709	0.478629
## height_cm:defending	-0.421	0.674106
## dribbling:defending	-0.511	0.609304
## age:physic	-1.003	0.315891
## height_cm:physic	-0.823	0.410371
## dribbling:physic	-0.910	0.362836
## defending:physic	-0.572	0.567473
## age:power_long_shots	-0.884	0.376561
## height_cm:power_long_shots	-0.652	0.514463
## dribbling:power_long_shots	-0.757	0.449241
## defending:power_long_shots	-0.343	0.731972
## physic:power_long_shots	-0.780	0.435622
## age:height_cm:dribbling	0.980	0.326890
## age:height_cm:defending	0.673	0.500780
## age:dribbling:defending	0.757	0.449181
## height_cm:dribbling:defending	0.480	0.631052
## age:height_cm:physic	0.964	0.335211
## age:dribbling:physic	1.059	0.289396
## height_cm:dribbling:physic	0.870	0.384088
## age:defending:physic	0.787	0.431215
## height_cm:defending:physic	0.541	0.588758
## dribbling:defending:physic	0.620	0.535509
## age:height_cm:power_long_shots	0.850	0.395225
## age:dribbling:power_long_shots	0.958	0.338190
## height_cm:dribbling:power_long_shots	0.723	0.469698
## age:defending:power_long_shots	0.625	0.531787
## height_cm:defending:power_long_shots	0.322	0.747124
## dribbling:defending:power_long_shots	0.428	0.668791
## age:physic:power_long_shots	0.937	0.348789
## height_cm:physic:power_long_shots	0.745	0.456561
## dribbling:physic:power_long_shots	0.842	0.399730
## defending:physic:power_long_shots	0.472	0.637198
## age:height_cm:dribbling:defending	-0.715	0.474335
## age:height_cm:dribbling:physic	-1.018	0.308881
## age:height_cm:defending:physic	-0.749	0.453777
## age:dribbling:defending:physic	-0.828	0.407450
## height_cm:dribbling:defending:physic	-0.584	0.559089
## age:height_cm:dribbling:power_long_shots	-0.921	0.357113
## age:height_cm:defending:power_long_shots	-0.598	0.549999
## age:dribbling:defending:power_long_shots	-0.695	0.487020
## height_cm:dribbling:defending:power_long_shots	-0.404	0.686125
## age:height_cm:physic:power_long_shots	-0.901	0.367811
## age:dribbling:physic:power_long_shots	-1.005	0.314974
## height_cm:dribbling:physic:power_long_shots	-0.805	0.420858
## age:defending:physic:power_long_shots	-0.705	0.480528
## height_cm:defending:physic:power_long_shots	-0.445	0.656128
## dribbling:defending:physic:power_long_shots	-0.542	0.587521
## age:height_cm:dribbling:defending:physic	0.785	0.432347
## age:height_cm:dribbling:defending:power_long_shots	0.663	0.507535
## age:height_cm:dribbling:physic:power_long_shots	0.966	0.333921
## age:height_cm:defending:physic:power_long_shots	0.674	0.500535
## age:dribbling:defending:physic:power_long_shots	0.769	0.441871
## height_cm:dribbling:defending:physic:power_long_shots	0.513	0.608105
## age:height_cm:dribbling:defending:physic:power_long_shots	-0.733	0.463524
##		
## (Intercept)		
## age		
## height_cm		
## dribbling		
## defending		
## physic		
## power_long_shots		
## I(age^2)	***	
## I(height_cm^2)		
## I(dribbling^2)		
## I(defending^2)	*	
## I(physic^2)		
## I(power_long_shots^2)		
## I(potential^2)	***	
## age:height_cm		
## age:dribbling		
## height_cm:dribbling		
## age:defending		
## height_cm:defending		
## dribbling:defending		
## age:physic		
## height_cm:physic		

```
## dribbling:physic
## defending:physic
## age:power_long_shots
## height_cm:power_long_shots
## dribbling:power_long_shots
## defending:power_long_shots
## physic:power_long_shots
## age:height_cm:dribbling
## age:height_cm:defending
## age:dribbling:defending
## height_cm:dribbling:defending
## age:height_cm:physic
## age:dribbling:physic
## height_cm:dribbling:physic
## age:defending:physic
## height_cm:defending:physic
## dribbling:defending:physic
## age:height_cm:power_long_shots
## age:dribbling:power_long_shots
## height_cm:dribbling:power_long_shots
## age:defending:power_long_shots
## height_cm:defending:power_long_shots
## dribbling:defending:power_long_shots
## age:physic:power_long_shots
## height_cm:physic:power_long_shots
## dribbling:physic:power_long_shots
## defending:physic:power_long_shots
## age:height_cm:dribbling:defending
## age:height_cm:dribbling:physic
## age:height_cm:defending:physic
## age:dribbling:defending:physic
## height_cm:dribbling:defending:physic
## age:height_cm:dribbling:power_long_shots
## age:height_cm:defending:power_long_shots
## age:dribbling:defending:power_long_shots
## height_cm:dribbling:defending:power_long_shots
## age:height_cm:physic:power_long_shots
## age:dribbling:physic:power_long_shots
## height_cm:dribbling:physic:power_long_shots
## age:defending:physic:power_long_shots
## height_cm:defending:physic:power_long_shots
## dribbling:defending:physic:power_long_shots
## age:height_cm:dribbling:defending:physic
## age:height_cm:dribbling:defending:power_long_shots
## age:height_cm:dribbling:physic:power_long_shots
## age:height_cm:defending:physic:power_long_shots
## age:dribbling:defending:physic:power_long_shots
## height_cm:dribbling:defending:physic:power_long_shots
## age:height_cm:dribbling:defending:physic:power_long_shots
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 618.90  on 513  degrees of freedom
## Residual deviance: 174.95  on 443  degrees of freedom
## AIC: 316.95
##
## Number of Fisher Scoring iterations: 11
```

Columns  $I(\text{age}^2)$ ,  $I(\text{defending}^2)$ ,  $I(\text{potential}^2)$  were found to be significant.

#### model1

Simple model is built.

```
lg_model1 <- glm(high.wage.ind ~ age +
                  height_cm +
                  dribbling +
                  defending +
                  physic +
                  power_long_shots +
                  potential, data=df, family=binomial)
summary(lg_model1)
```



```
##
## Call:
## glm(formula = high.wage.ind ~ age + height_cm + dribbling + defending +
##      physic + power_long_shots + potential, family = binomial,
##      data = df)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.5562  -0.3949  -0.1215   0.1460   2.5659
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -57.201931   7.230179  -7.912 0.0000000000000254 ***
## age           0.272083   0.054216   5.018 0.00000052078334155 ***
## height_cm     0.070703   0.029736   2.378   0.017422 *
## dribbling     0.112784   0.031888   3.537   0.000405 ***
## defending       0.029214   0.011970   2.441   0.014667 *
## physic        -0.003968   0.023945  -0.166   0.868375
## power_long_shots 0.026991   0.015248   1.770   0.076711 .
## potential     0.359693   0.048786   7.373 0.0000000000016692 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 618.9  on 513  degrees of freedom
## Residual deviance: 278.6  on 506  degrees of freedom
## AIC: 294.6
##
## Number of Fisher Scoring iterations: 7
```

### model2

Now I will add the significant coefficients found on model0 to model2 which are  $I(\text{age}^2)$ ,  $I(\text{defending}^2)$ , and  $I(\text{potential}^2)$ .

```
lg_model2 <- glm(high.wage.ind ~ age +
                  height_cm +
                  dribbling +
                  defending +
                  physic +
                  power_long_shots +
                  potential +
                  I(age^2) +
                  I(defending^2) +
                  I(potential^2), data=df, family=binomial)
summary(lg_model2)
```

```
##
## Call:
## glm(formula = high.wage.ind ~ age + height_cm + dribbling + defending +
##      physic + power_long_shots + potential + I(age^2) + I(defending^2) +
##      I(potential^2), family = binomial, data = df)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.71845  -0.30933  -0.06546   0.14135   2.45539
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -144.2591414   41.5384562  -3.473 0.000515 ***
## age           2.4575823   0.6037411   4.071 0.0000469 ***
## height_cm     0.0489837   0.0319654   1.532 0.125425
## dribbling     0.0755562   0.0385326   1.961 0.049898 *
## defending       -0.2104275   0.0854109  -2.464 0.013751 *
## physic        -0.0378334   0.0274971  -1.376 0.168852
## power_long_shots 0.0385787   0.0169750   2.273 0.023046 *
## potential     2.2036432   1.1110701   1.983 0.047328 *
## I(age^2)       -0.0399280   0.0108332  -3.686 0.000228 ***
## I(defending^2)  0.0024210   0.0008875   2.728 0.006373 **
## I(potential^2) -0.0118716   0.0074457  -1.594 0.110839
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 618.90  on 513  degrees of freedom
## Residual deviance: 240.12  on 503  degrees of freedom
## AIC: 262.12
##
## Number of Fisher Scoring iterations: 7
```

### model3

physic column is not a significant coefficient so it will be removed from model2.

```
lg_model3 <- update(lg_model2, ~. -physic)

summary(lg_model3)
```

```
##
## Call:
## glm(formula = high.wage.ind ~ age + height_cm + dribbling + defending +
##      power_long_shots + potential + I(age^2) + I(defending^2) +
##      I(potential^2), family = binomial, data = df)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.80516  -0.32330  -0.06313   0.14219   2.45454
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -142.9301065    40.8844611  -3.496  0.000472 ***
## age           2.2328591     0.5733234   3.895  0.0000984 ***
## height_cm     0.0329031     0.0296654   1.109  0.267370
## dribbling     0.0860689     0.0379864   2.266  0.023465 *
## defending      -0.2188485     0.0845144  -2.589  0.009612 **
## power_long_shots 0.0352084     0.0166060   2.120  0.033987 *
## potential     2.2935743     1.0912843   2.102  0.035578 *
## I(age^2)      -0.0363405     0.0103747  -3.503  0.000460 ***
## I(defending^2)  0.0024432     0.0008817   2.771  0.005586 **
## I(potential^2) -0.0126722     0.0073044  -1.735  0.082763 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 618.90  on 513  degrees of freedom
## Residual deviance: 242.06  on 504  degrees of freedom
## AIC: 262.06
##
## Number of Fisher Scoring iterations: 7
```

#### model4

height\_cm column is not a significant coefficient so it will be removed from model3.

```
lg_model4 <- update(lg_model3, ~. -height_cm)

summary(lg_model4)
```

```
##
## Call:
## glm(formula = high.wage.ind ~ age + dribbling + defending + power_long_shots +
##      potential + I(age^2) + I(defending^2) + I(potential^2), family = binomial,
##      data = df)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.8620  -0.3202  -0.0668   0.1361   2.5007
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -138.8253817    40.4575875  -3.431  0.000601 ***
## age           2.3847270     0.5613062   4.249  0.0000215 ***
## dribbling     0.0656418     0.0329937   1.990  0.046643 *
## defending      -0.2202835     0.0837244  -2.631  0.008512 **
## power_long_shots 0.0379794     0.0163750   2.319  0.020376 *
## potential     2.2971500     1.0862434   2.115  0.034450 *
## I(age^2)      -0.0389509     0.0101871  -3.824  0.000132 ***
## I(defending^2)  0.0024510     0.0008724   2.810  0.004961 **
## I(potential^2) -0.0125484     0.0072689  -1.726  0.084290 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 618.9  on 513  degrees of freedom
## Residual deviance: 243.3  on 505  degrees of freedom
## AIC: 261.3
##
## Number of Fisher Scoring iterations: 7
```

#### model5

I(potential^2) has been shown to have low significance. Therefore it will be removed from model 4.

```
lg_model5 <- update(lg_model4, ~. - I(potential^2))

summary(lg_model5)
```

```
##
## Call:
## glm(formula = high.wage.ind ~ age + dribbling + defending + power_long_shots +
##     potential + I(age^2) + I(defending^2), family = binomial,
##     data = df)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.86148  -0.32449  -0.08563   0.11175   2.45009
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -73.0617277  10.3315670  -7.072 0.0000000000015305 ***
## age           2.6121957   0.5614537   4.653 0.0000032784273046 ***
## dribbling     0.0615119   0.0324929   1.893   0.05835 .
## defending     -0.2239394   0.0830827  -2.695   0.00703 **
## power_long_shots 0.0393219   0.0164313   2.393   0.01671 *
## potential     0.4380596   0.0567646   7.717 0.000000000000119 ***
## I(age^2)      -0.0429493   0.0101425  -4.235 0.0000228961024058 ***
## I(defending^2)  0.0024820   0.0008672   2.862   0.00421 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 618.90  on 513  degrees of freedom
## Residual deviance: 246.36  on 506  degrees of freedom
## AIC: 262.36
##
## Number of Fisher Scoring iterations: 7
```

#### Description, explanation and critique of the model:

- First, I built a complex model with the features that were found important for `high.wage.ind` during the EDA with their quadratic relationship and interactions between them. `I(age^2)`, `I(defending^2)`, `I(potential^2)` were found to be significant. The AIC score of model0 is 316.95
- Second, I built a very simple model (model1). The column `physic` was no significant and column `power_long_shots` had very small significance. The AIC score decrease to 294.6 (lowest so far).
- Third, I added the significant coefficients found in model0 which were `I(age^2)`, `I(defending^2)`, and `I(potential^2)` to model1 to build model2. The AIC score decrease to 262.12 (lowest so far).
- Fourth, for model3, I removed column `physic` from model2 because it was not significant. The AIC score slightly decreased to 262.06 (lowest so far).
- Fifth, for model4, I removed column `height_cm` from model3 because it was not significant. The AIC score slightly decreased to 261.3 (lowest so far).
- Finally, for model5, I removed `I(potential^2)` from model4 because it had very low significance. However, the AIC score slightly increased to 262.36.
- Since model4 is the model which has the smallest AIC score, then this will be our model I will use.
- The model can be interpreted as follows:

$$\log\left(\frac{p}{1-p}\right) = -138.8253817 + 2.3847270 \times \text{age} + 0.0656418 \times \text{dribbling} - 0.2202835 \times \text{defending} + 0.0379794 \times \text{power\_long\_shots} + 2.$$

- The logistic regression coefficients give the change in the log odds of the outcome for a one unit increase in the explanatory variable (e.g., for every unit change for age the log odds of making having a weekly wage above 8000 increases by 2.38).
- To improve the model, same concept for multi-regression applies. Since Logistic Regression is an algorithm that uses distance between points for the loss function, variables which range are higher (e.g., `wage_eur_log` range is significantly smaller than the range in `height_cm`) will have a bigger impact in the model without being necessary more important than the other variables. A good way to fix this is to standardize numerical columns. In that way, your columns will follow a normal distribution and have a mean of 0 and a standard deviation of 1. In other words, they will have similar scale and will help the algorithm remove that bias.
- Another issue found in the model is that the variable `high.wage_ind` is imbalanced which means that the model will have bias towards learning more about values 1 than from values 0. A way to fix this is to apply a "rare events correction to the intercept" [8]

## References

Add any references here. NB You can either do this manually or automatically with a `.bib` file (which then must be submitted along with your `.Rmd` file). See the RMarkdown documentation (<https://bookdown.org/yihui/rmarkdown-cookbook/bibliography.html>) for guidance.

[1] J. Press, "The Japan News," 13 August 2021. [Online]. Available: <https://the-japan-news.com/news/article/0007672187> (<https://the-japan-news.com/news/article/0007672187>). [Accessed 05 January 2022].

[2] FIFA, "FIFA," 25 February 2021. [Online]. Available: <https://www.fifa.com/news/kazuyoshi-miura-i-don-t-think-i-ll-ever-leave-football> (<https://www.fifa.com/news/kazuyoshi-miura-i-don-t-think-i-ll-ever-leave-football>). [Accessed 06 January 2022].

[3] "Transfer Markt," [Online]. Available: <https://www.transfermarkt.com/kyle-hudlin/profil/spieler/829766> (<https://www.transfermarkt.com/kyle-hudlin/profil/spieler/829766>). [Accessed 06 January 2022].

[4] "Colgadosporelftbol.com," [Online]. Available: <https://colgadosporelftbol.com/el-peso-ideal-de-un-futbolista-profesional-y-su-importancia/#> (<https://colgadosporelftbol.com/el-peso-ideal-de-un-futbolista-profesional-y-su-importancia/#>);~:text=El%20margen%20estimado%20es%20de,porm%20cada%202.5%20cent%C3%ADmetros%20adicionales. [Accessed 06 January 2022].

[5] S. Glen, "Statistics How to," [Online]. Available: <https://www.statisticshowto.com/logarithmic-distribution/> (<https://www.statisticshowto.com/logarithmic-distribution/>). [Accessed 06 January 2022].

[6] K. S. Htoon, "Medium," 29 February 2020. [Online]. Available: <https://medium.com/@kyawsawhtoon/log-transformation-purpose-and-interpretation-9444b4b049c9> (<https://medium.com/@kyawsawhtoon/log-transformation-purpose-and-interpretation-9444b4b049c9>). [Accessed 06 January 2022].

[7] "Statology," 01 October 2021. [Online]. Available: <https://www.statology.org/glm-fit-algorithm-did-not-converge/> (<https://www.statology.org/glm-fit-algorithm-did-not-converge/>). [Accessed 06 January 2022].

[8] G. King y L. Zeng, «Logistic Regression in Rare,» Harvard University, Cambridge, MA, , 2001.