

Major Leagues - Caio Vigo Pereira

Loading the data

I choose to analyze the soccer datasets.

```
my_path<-"C:/Users/Caio Laptop/OneDrive - The University of Kansas/Documents/PhD/11. Courses/19. EECS 701/19. EECS 701 Datasets"
setwd(my_path)

spi_matches<-read.csv(paste(my_path,"/Datasets/spi_matches.csv", sep=""),header=T)
spi_global_rankings_intl<-read.csv(paste(my_path,"/Datasets/spi_global_rankings_intl.csv", sep=""),header=T)
spi_global_rankings<-read.csv(paste(my_path,"/Datasets/spi_global_rankings.csv", sep=""),header=T)
```

Information from the source

SPI Ratings This file contains links to the data behind our Club Soccer Predictions and Global Club Soccer Rankings. spi_matches.csv contains match-by-match SPI ratings and forecasts back to 2016. spi_global_rankings.csv contains current SPI ratings and rankings for men's club teams. spi_global_rankings_intl.csv contains current SPI ratings and rankings for men's international teams.

Analyzing my 3 datasets

```
names(spi_matches)
```

```
## [1] "date"          "league_id"      "league"         "team1"          "team2"
## [6] "spi1"          "spi2"          "prob1"          "prob2"          "probtie"
## [11] "proj_score1"   "proj_score2"    "importance1"     "importance2"    "score1"
## [16] "score2"        "xg1"           "xg2"           "nsxg1"          "nsxg2"
## [21] "adj_score1"    "adj_score2"
```

```
head(spi_matches)
```

```
##      date league_id league team1
## 1 2016-08-12    1843 French Ligue 1 Bastia
## 2 2016-08-12    1843 French Ligue 1 AS Monaco
## 3 2016-08-13    2411 Barclays Premier League Hull City
## 4 2016-08-13    2411 Barclays Premier League Burnley
## 5 2016-08-13    2411 Barclays Premier League Middlesbrough
## 6 2016-08-13    2411 Barclays Premier League Southampton
##      team2 spi1 spi2 prob1 prob2 probtie proj_score1
## 1 Paris Saint-Germain 51.16 85.68 0.0463 0.8380 0.1157 0.91
## 2 Guingamp 68.85 56.48 0.5714 0.1669 0.2617 1.82
## 3 Leicester City 53.57 66.81 0.3459 0.3621 0.2921 1.16
## 4 Swansea City 58.98 59.74 0.4482 0.2663 0.2854 1.37
## 5 Stoke City 56.32 60.35 0.4380 0.2692 0.2927 1.30
## 6 Watford 69.49 59.33 0.5759 0.1874 0.2367 1.91
##      proj_score2 importance1 importance2 score1 score2 xg1 xg2 nsxg1 nsxg2
## 1 2.36 32.4 67.7 0 1 0.97 0.63 0.43 0.45
## 2 0.86 53.7 22.9 2 2 2.45 0.77 1.75 0.42
## 3 1.24 38.1 22.2 2 1 0.85 2.77 0.17 1.25
## 4 1.05 36.5 29.1 0 1 1.24 1.84 1.71 1.56
## 5 1.01 33.9 32.5 1 1 1.40 0.55 1.13 1.06
```

```
## 6      1.05      34.1      30.7      1      1 1.05 0.22  1.52  0.41
##   adj_score1 adj_score2
## 1      0.00      1.05
## 2      2.10      2.10
## 3      2.10      1.05
## 4      0.00      1.05
## 5      1.05      1.05
## 6      1.05      1.05

dim(spi_matches)

## [1] 20879    22

typeof(spi_matches)

## [1] "list"

str(spi_matches)

## 'data.frame':   20879 obs. of  22 variables:
##  $ date       : Factor w/ 839 levels "2016-08-12","2016-08-13",...: 1 1 2 2 2 2 2 2 2 ...
##  $ league_id  : int   1843 1843 2411 2411 2411 2411 2411 2411 1843 2411 ...
##  $ league     : Factor w/ 37 levels "Argentina Primera Division",...: 13 13 4 4 4 4 4 4 13 4 ...
##  $ team1      : Factor w/ 698 levels "1. FC Heidenheim 1846",...: 78 50 319 120 406 581 213 180 103 3...
##  $ team2      : Factor w/ 698 levels "1. FC Heidenheim 1846",...: 473 295 369 613 604 682 634 685 593...
##  $ spi1       : num   51.2 68.8 53.6 59 56.3 ...
##  $ spi2       : num   85.7 56.5 66.8 59.7 60.4 ...
##  $ prob1      : num   0.0463 0.5714 0.3459 0.4482 0.438 ...
##  $ prob2      : num   0.838 0.167 0.362 0.266 0.269 ...
##  $ probtie    : num   0.116 0.262 0.292 0.285 0.293 ...
##  $ proj_score1: num   0.91 1.82 1.16 1.37 1.3 1.91 1.47 1.35 1.39 2.69 ...
##  $ proj_score2: num   2.36 0.86 1.24 1.05 1.01 1.05 1.38 1.14 1.14 0.48 ...
##  $ importance1: num   32.4 53.7 38.1 36.5 33.9 34.1 31.9 43.6 37.9 73 ...
##  $ importance2: num   67.7 22.9 22.2 29.1 32.5 30.7 48 34.6 44.2 27 ...
##  $ score1     : int    0 2 2 0 1 1 1 0 3 2 ...
##  $ score2     : int    1 2 1 1 1 1 1 1 2 1 ...
##  $ xg1        : num   0.97 2.45 0.85 1.24 1.4 1.05 0.73 1.11 1.03 2.14 ...
##  $ xg2        : num   0.63 0.77 2.77 1.84 0.55 0.22 1.11 0.68 1.84 1.25 ...
##  $ nsxg1      : num   0.43 1.75 0.17 1.71 1.13 1.52 0.88 0.84 1.1 1.81 ...
##  $ nsxg2      : num   0.45 0.42 1.25 1.56 1.06 0.41 1.81 1.6 2.26 0.92 ...
##  $ adj_score1 : num    0 2.1 2.1 0 1.05 1.05 1.05 0 3.12 2.1 ...
##  $ adj_score2 : num    1.05 2.1 1.05 1.05 1.05 1.05 1.05 1.05 2.1 1.05 ...

names(spi_global_rankings_intl)

## [1] "rank"  "name"  "confed" "off"   "def"   "spi"

head(spi_global_rankings_intl)

##   rank    name confed off def  spi
## 1     1  Brazil CONMEBOL 3.11 0.29 92.96
## 2     2   Spain      UEFA 3.46 0.48 92.54
## 3     3 Belgium      UEFA 3.06 0.54 89.10
## 4     4  France      UEFA 2.84 0.46 88.57
## 5     5 Germany      UEFA 2.96 0.56 87.93
## 6     6 Argentina CONMEBOL 2.57 0.49 85.53
```

```
dim(spi_global_rankings_intl)
```

```
## [1] 213 6
```

```
typeof(spi_global_rankings_intl)
```

```
## [1] "list"
```

```
str(spi_global_rankings_intl)
```

```
## 'data.frame': 213 obs. of 6 variables:
## $ rank : int 1 2 3 4 5 6 7 8 9 10 ...
## $ name : Factor w/ 213 levels "Afghanistan",...: 28 175 19 68 74 8 61 150 133 203 ...
## $ confed: Factor w/ 6 levels "AFC","CAF","CONCACAF",...: 4 6 6 6 6 4 6 6 6 4 ...
## $ off : num 3.11 3.46 3.06 2.84 2.96 2.57 2.32 2.38 2.55 2.3 ...
## $ def : num 0.29 0.48 0.54 0.46 0.56 0.49 0.51 0.56 0.68 0.54 ...
## $ spi : num 93 92.5 89.1 88.6 87.9 ...
```

```
names(spi_global_rankings)
```

```
## [1] "rank" "prev_rank" "name" "league" "off" "def"
## [7] "spi"
```

```
head(spi_global_rankings)
```

```
## rank prev_rank name league off def
## 1 1 1 Manchester City Barclays Premier League 2.92 0.20
## 2 2 3 Barcelona Spanish Primera Division 3.12 0.38
## 3 3 4 Real Madrid Spanish Primera Division 2.99 0.38
## 4 4 2 Bayern Munich German Bundesliga 2.94 0.40
## 5 5 6 Juventus Italy Serie A 2.66 0.29
## 6 6 7 Paris Saint-Germain French Ligue 1 3.09 0.49
## spi
## 1 93.78
## 2 92.41
## 3 91.75
## 4 90.93
## 5 90.72
## 6 90.70
```

```
dim(spi_global_rankings)
```

```
## [1] 628 7
```

```
typeof(spi_global_rankings)
```

```
## [1] "list"
```

```
str(spi_global_rankings)
```

```
## 'data.frame': 628 obs. of 7 variables:
## $ rank : int 1 2 3 4 5 6 7 8 9 10 ...
## $ prev_rank: int 1 3 4 2 6 7 5 8 9 10 ...
## $ name : Factor w/ 628 levels "1. FC Heidenheim 1846",...: 357 64 461 70 302 428 338 57 136 573
## $ league : Factor w/ 35 levels "Argentina Primera Division",...: 4 28 28 16 18 13 4 28 4 4 ...
## $ off : num 2.92 3.12 2.99 2.94 2.66 3.09 2.66 2.21 2.52 2.42 ...
## $ def : num 0.2 0.38 0.38 0.4 0.29 0.49 0.3 0.26 0.45 0.52 ...
## $ spi : num 93.8 92.4 91.8 90.9 90.7 ...
```

Loading some packages

```
library(stargazer)

##
## Please cite as:
## Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statistics Tables.
## R package version 5.2.2. https://CRAN.R-project.org/package=stargazer
library(Amelia) # for missmap() function

## Loading required package: Rcpp

## ##
## ## Amelia II: Multiple Imputation
## ## (Version 1.7.5, built: 2018-05-07)
## ## Copyright (C) 2005-2018 James Honaker, Gary King and Matthew Blackwell
## ## Refer to http://gking.harvard.edu/amelia/ for more information
## ##
```

Taking a look on the Descriptive Statistics

```
stargazer(spi_matches, type='text', align=TRUE, digits=2)

##
## =====
## Statistic      N      Mean   St. Dev.  Min  Pctl(25) Pctl(75)  Max
## -----
## league_id     20,879  2,141.44  738.39   1,818  1,849     2,160    5,641
## spi1          20,879   46.57    18.62    5.23   33.04     59.48    96.57
## spi2          20,879   46.53    18.61    4.97   33.04     59.41    96.78
## prob1         20,879    0.45     0.16    0.03    0.35      0.54     0.98
## prob2         20,879    0.29     0.14    0.00    0.20      0.36     0.88
## probtie       20,879    0.26     0.05    0.00    0.24      0.28     0.45
## proj_score1   20,879    1.52     0.43    0.25    1.24      1.72     4.03
## proj_score2   20,879    1.14     0.42    0.20    0.88      1.36     3.42
## importance1   10,515   30.79    25.35    0.00   10.90     44.60   100.00
## importance2   10,515   30.12    25.03    0.00   10.50     43.60   100.00
## score1        14,315    1.54     1.28    0.00    1.00      2.00     8.00
## score2        14,315    1.17     1.14    0.00    0.00      2.00     8.00
## xg1           8,664    1.47     0.83    0.00    0.85      1.94     7.04
## xg2           8,664    1.12     0.72    0.00    0.58      1.50     6.20
## nsxg1         8,664    1.40     0.65    0.00    0.95      1.74     6.58
## nsxg2         8,664    1.12     0.57    0.00    0.72      1.42     5.92
## adj_score1    8,664    1.55     1.26    0.00    1.05      2.10     7.97
## adj_score2    8,664    1.17     1.12    0.00    0.00      2.10     6.76
## -----

stargazer(spi_global_rankings, type='text', align=TRUE, digits=2)
```

```
##
## =====
## Statistic  N    Mean  St. Dev.  Min  Pctl(25) Pctl(75)  Max
## -----
```

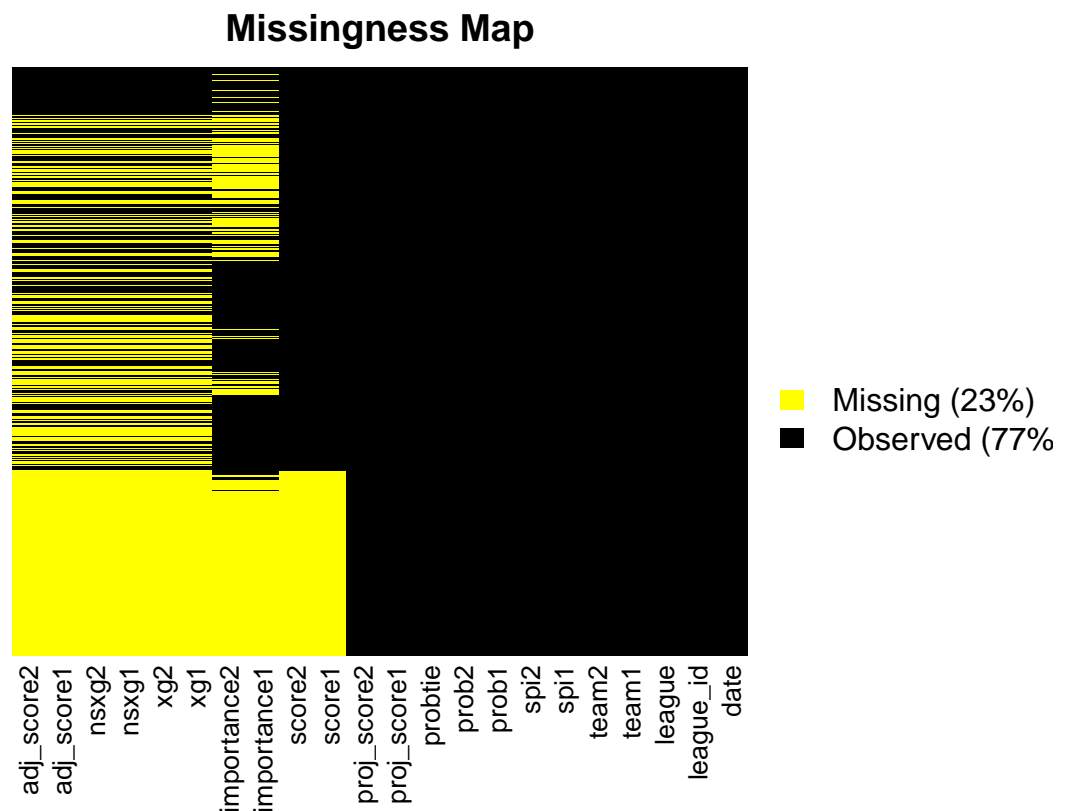
```
## rank      628 314.50  181.43   1    157.8    471.2    628
## prev_rank 628 314.50  181.43   1    157.8    471.2    628
## off       628  1.26   0.49   0.20  0.96     1.53    3.12
## def       628  1.39   0.44   0.20  1.09     1.68    2.84
## spi       628 42.99   18.08   4.97  29.98    55.66    93.78
## -----
```

```
stargazer(spi_global_rankings_intl, type='text', align=TRUE, digits=2)
```

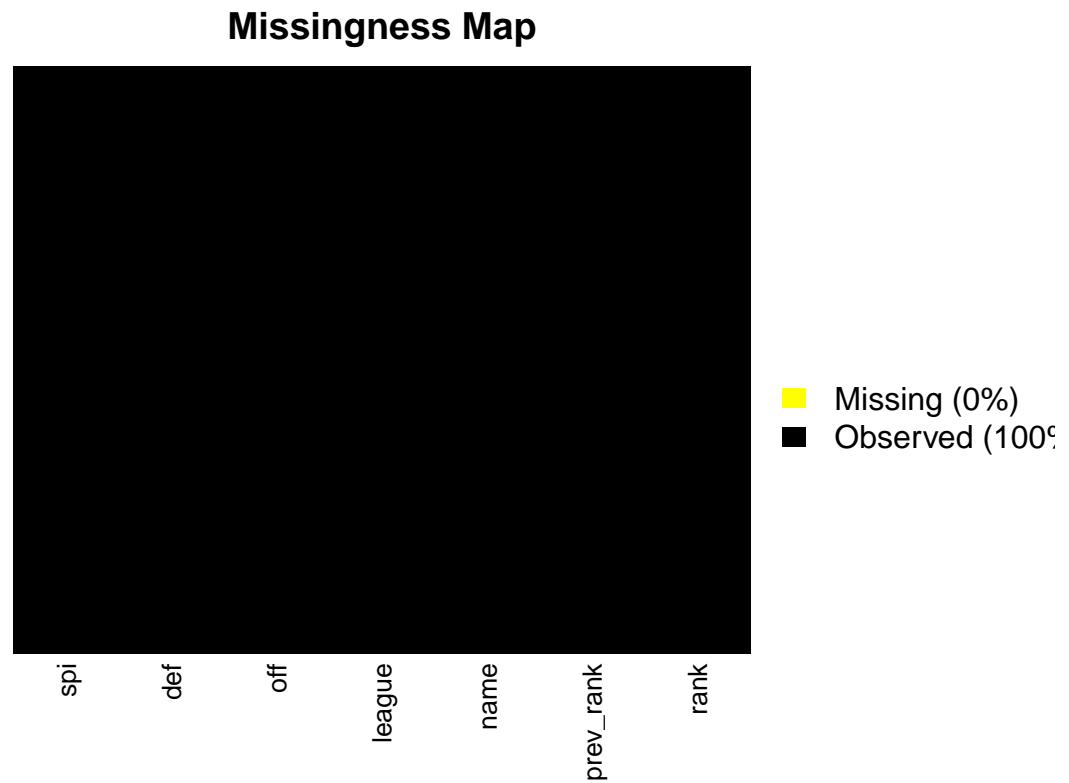
```
##
## =====
## Statistic  N    Mean  St. Dev. Min  Pctl(25) Pctl(75)  Max
## -----
## rank      213 107.00  61.63   1     54     160     213
## off       213  1.17   0.65   0.20  0.67     1.57    3.46
## def       213  1.64   1.10   0.29  0.93     1.89    6.08
## spi       213 39.91   24.45   0.26  19.84    59.64   92.96
## -----
```

Checking for any NA's in the dataframe.

```
missmap(spi_matches,col=c('yellow','black'),y.at=1,y.labels='',legend=TRUE)
```

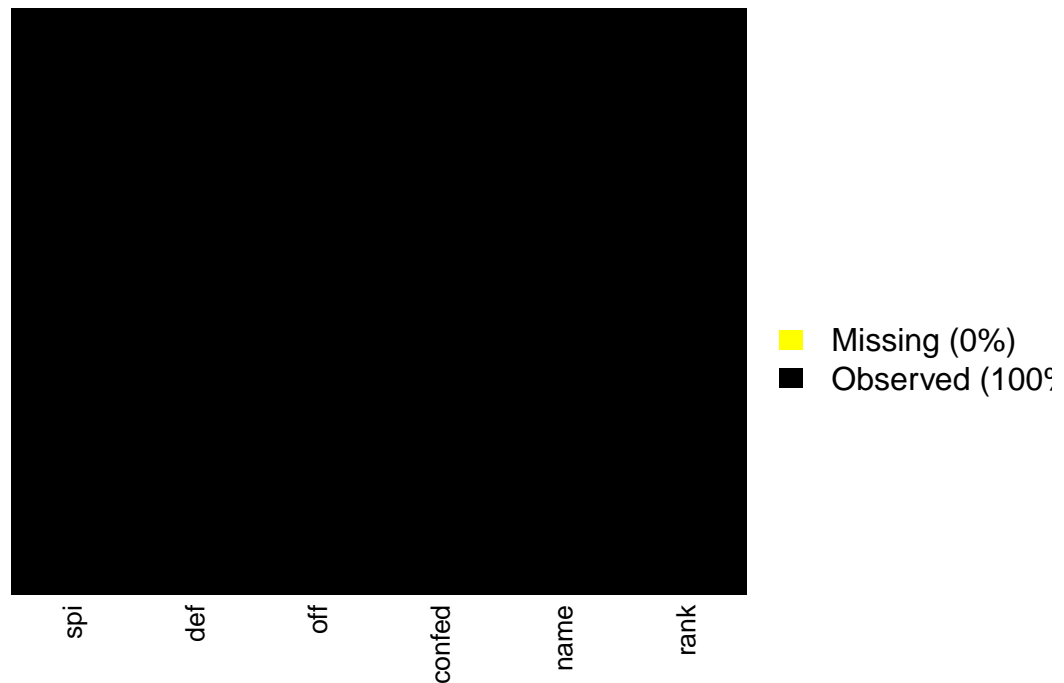


```
missmap(spi_global_rankings,col=c('yellow','black'),y.at=1,y.labels='',legend=TRUE)
```



```
missmap(spi_global_rankings_intl,col=c('yellow','black'),y.at=1,y.labels='',legend=TRUE)
```

Missingness Map



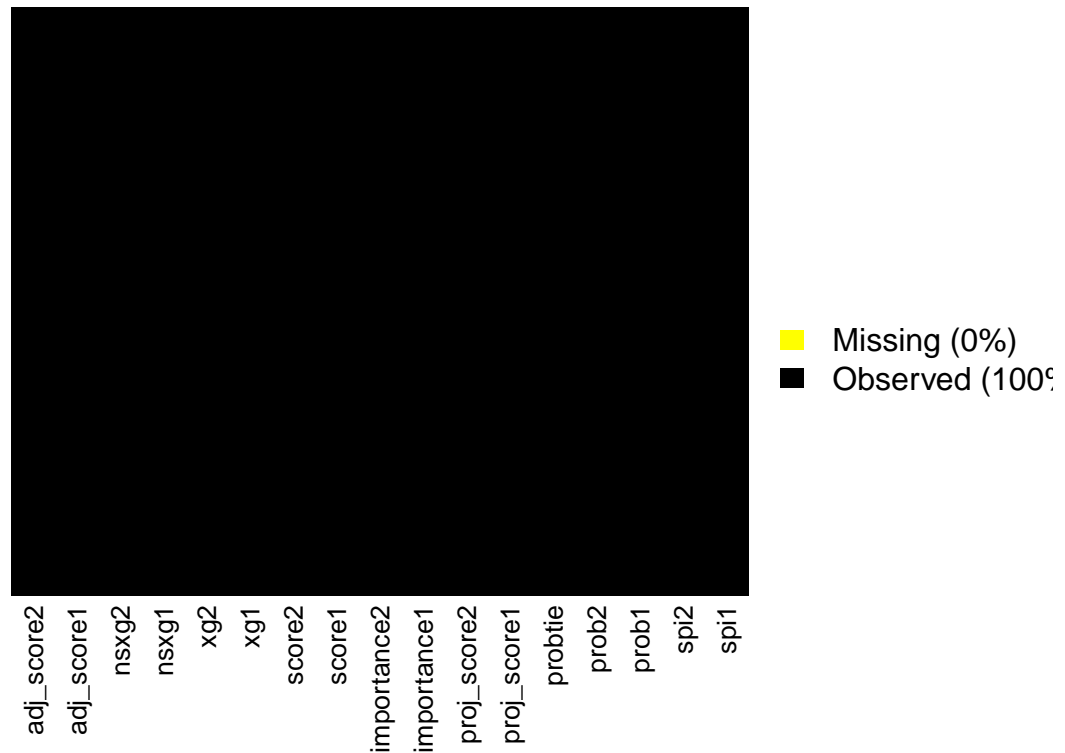
Replacing missing values with the mean

```
spi_matches<-spi_matches[,6:22]
for(i in 1:ncol(spi_matches)){
  spi_matches[is.na(spi_matches[,i]), i] <- mean(spi_matches[,i], na.rm = TRUE)
}
```

Checking if there is any missing value after the changes

```
missmap(spi_matches,col=c('yellow','black'),y.at=1,y.labels='',legend=TRUE)
```

Missingness Map



```
#set a seed
set.seed(999)

library(caTools)
#Split the data , `split()` assigns a booleans to a new column based on the SplitRatio specified.

split <- sample.split(spi_matches,SplitRatio =0.75)

train <- subset(spi_matches,split==TRUE)
test <- subset(spi_matches,split==FALSE)
```

```
team_1_model <- lm(score1 ~ proj_score1 + importance1 + xg1 + nsxg1 + spi1 + probt1, data=train)
team_2_model <- lm(score2 ~ proj_score2 + importance2 + xg2 + nsxg2 + spi2 + probt2, data=train)
# summary(team_1_model)
# summary(team_2_model)
stargazer(team_1_model, type='text', align=TRUE, digits=2)
```

```
##
## =====
##               Dependent variable:
##               -----
##               score1
## -----
## proj_score1      0.35***
##                  (0.04)
##
```



```

## importance1          0.0003
##                    (0.0004)
##
## xg1                  0.93***
##                    (0.02)
##
## nsxg1               -0.40***
##                    (0.02)
##
## spi1                -0.0003
##                    (0.0005)
##
## prob1               0.37***
##                    (0.11)
##
## Constant            0.03
##                    (0.04)
##
## -----
## Observations        14,739
## R2                  0.24
## Adjusted R2         0.24
## Residual Std. Error  0.92 (df = 14732)
## F Statistic         773.89*** (df = 6; 14732)
## =====
## Note:                *p<0.1; **p<0.05; ***p<0.01
stargazer(team_2_model, type='text', align=TRUE, digits=2)

##
## =====
##                    Dependent variable:
##                    -----
##                    score2
## -----
## proj_score2         0.28***
##                    (0.04)
##
## importance2         0.0000
##                    (0.0004)
##
## xg2                 0.99***
##                    (0.02)
##
## nsxg2              -0.33***
##                    (0.02)
##
## spi2               -0.0001
##                    (0.0004)
##
## prob2              0.33**
##                    (0.13)
##
## Constant            0.02
##                    (0.03)

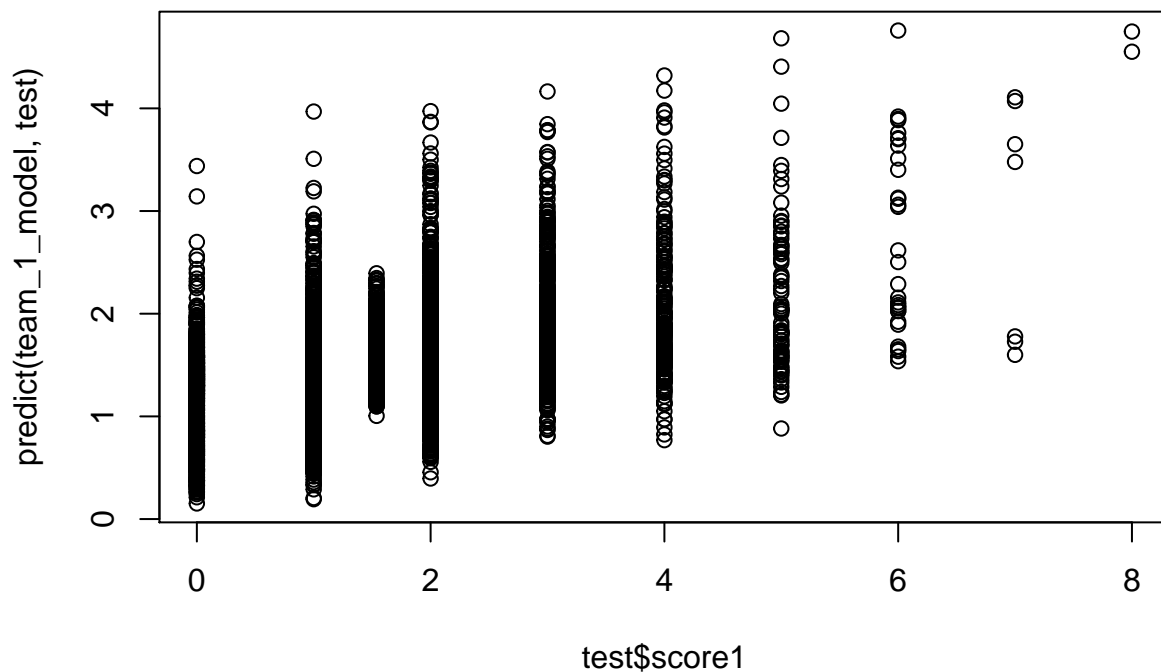
```

```
##
## -----
## Observations          14,739
## R2                    0.24
## Adjusted R2           0.24
## Residual Std. Error   0.82 (df = 14732)
## F Statistic           775.62*** (df = 6; 14732)
## =====
## Note:                  *p<0.1; **p<0.05; ***p<0.01
# test$predicted.medv <- predict(team_1_model,test)
# test<- na.omit(test$predicted.medv)
```

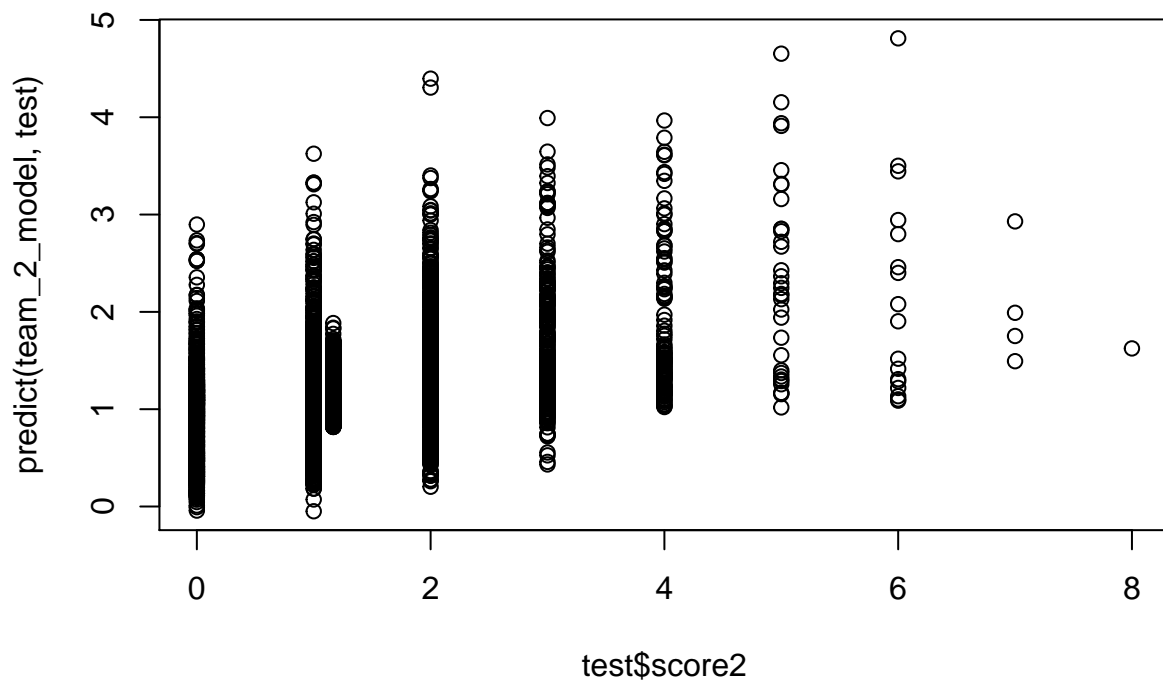
Predictions plots

```
predict_team_1_model <- predict(team_1_model,test)
predict_team_2_model <- na.omit(predict(team_2_model,test))

plot(test$score1,predict(team_1_model,test))
```



```
plot(test$score2,predict(team_2_model,test))
```



```
error <- test$score1-predict_team_1_model
rmse <- sqrt(mean(error)^2)
rmse
```

```
## [1] 0.008974758
```

Add a new chunk by clicking the *Insert Chunk* button on the toolbar or by pressing *Ctrl+Alt+I*.

When you save the notebook, an HTML file containing the code and output will be saved alongside it (click the *Preview* button or press *Ctrl+Shift+K* to preview the HTML file).

The preview shows you a rendered HTML copy of the contents of the editor. Consequently, unlike *Knit*, *Preview* does not run any R code chunks. Instead, the output of the chunk when it was last run in the editor is displayed.