# Regression Variables Selector

This document shows a step-by-step procedure of how the variables were selected for modeling

### Loading the packages

```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.0 --
## v ggplot2 3.3.0 v purrr
                               0.3.4
## v tibble 3.0.1 v dplyr 0.8.5
## v tidyr 1.0.3 v stringr 1.4.0
## v readr 1.3.1 v forcats 0.5.0
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(lasso2)
## R Package to solve regression problems while imposing
    an L1 constraint on the parameters. Based on S-plus Release 2.1
## Copyright (C) 1998, 1999
## Justin Lokhorst <jlokhors@stats.adelaide.edu.au>
## Berwin A. Turlach <bturlach@stats.adelaide.edu.au>
## Bill Venables
                 <wvenable@stats.adelaide.edu.au>
## Copyright (C) 2002
## Martin Maechler <maechler@stat.math.ethz.ch>
```

## Loading the datasets

```
s17_1 <- read.csv("~/DSI-SRP1/season2017.csv", encoding = "UTF-8")
s17_2 <- read.csv("~/DSI-SRP1/FPL_2016_17_new.csv")
s18_1 <- read.csv("~/DSI-SRP1/season2018.csv", encoding = "UTF-8")
s18_2 <- read.csv("~/DSI-SRP1/FPL_2017_18_new.csv")
s19_1 <- read.csv("~/DSI-SRP1/season2019.csv", encoding = "UTF-8")
s19_2 <- read.csv("~/DSI-SRP1/FPL_2018_19_new.csv")</pre>
```

#### Variable selector

#### Part I

This is done using the seasonal datasets that is going to be used for modeling

```
s17_1n <- var_sel(s17_1)
s18_1n <- var_sel(s18_1)
s19_1n <- var_sel(s19_1)
s_combined <- rbind(s17_1n, s18_1n, s19_1n)
lasso.s <- l1ce(total_points~., data = s_combined)</pre>
summary(lasso.s)$coefficients
##
                             Value Std. Error
                                                 Z score
                                                             Pr(>|Z|)
## (Intercept)
                       16.46621655 1.438894135 11.443661 0.000000e+00
## goals_scored
                        0.00000000 0.313436323 0.000000 1.000000e+00
## assists
                        0.00000000 0.325963467 0.000000 1.000000e+00
                        0.00000000 0.003115412 0.000000 1.000000e+00
## minutes.played
## bonus
                        0.65961380 0.187782364 3.512650 4.436607e-04
                        0.08088562 0.010440822 7.747055 9.325873e-15
## bps
## goals_conceded
                        0.00000000 0.104186649 0.000000 1.000000e+00
## clean_sheets
                        0.67163353 0.346257177 1.939696 5.241669e-02
## ict_index
                        0.16194721 0.024962065 6.487733 8.713763e-11
                        0.00000000 0.445673259 0.000000 1.000000e+00
## position_index
## selected_by_percent 0.00000000 0.139350609 0.000000 1.000000e+00
lasso.s <- l1ce(total_points ~ ict_index + bps + clean_sheets + bonus,</pre>
                data = s_combined)
summary(lasso.s)$coefficients
                                                      Pr(>|Z|)
##
                      Value Std. Error
                                          Z score
```

```
## (Intercept) 25.47307218 0.814252771 31.283986 0.000000e+00
## ict_index 0.12862290 0.017572139 7.319707 2.484679e-13
## bps 0.07502564 0.007530691 9.962651 0.000000e+00
## clean_sheets 0.00000000 0.334994853 0.000000 1.000000e+00
## bonus 0.03378626 0.204580419 0.165149 8.688267e-01
```

#### Part II

This is done by combining all three weekly datasets

```
s_week <- rbind(s17_2, s18_2, s19_2)
s_week <- s_week %>%
  select(-player_name:-position, -X, -season, -GW)
sample.index <- sample(1:nrow(s_week), nrow(s_week)*0.85, replace = FALSE)</pre>
s_week1 <- s_week[-sample.index,]</pre>
lasso.week <- l1ce(total_points ~., data = s_week1)</pre>
summary(lasso.week)$coefficients
##
                       Value Std. Error
                                           Z score
                                                       Pr(>|Z|)
## (Intercept)
                 12.21488166 0.599686306 20.368785 0.000000e+00
## position_index 0.00000000 0.177974424 0.000000 1.000000e+00
## goals scored 0.00000000 0.142903576 0.000000 1.000000e+00
## assists
                  0.00000000 0.149556052 0.000000 1.000000e+00
## ict index
                  0.17388986 0.012033213 14.450825 0.000000e+00
## goals conceded 0.00000000 0.047386246 0.000000 1.000000e+00
## minutes 0.0000000 0.001446032 0.000000 1.000000e+00
             0.00000000 0.782731973 0.000000 1.000000e+00
## own_goals
                 0.08416582 0.004990104 16.866545 0.000000e+00
## bps
## bonus
                 0.68715478 0.084028529 8.177637 2.220446e-16
## clean_sheets 0.61700253 0.151895632 4.062016 4.865068e-05
lasso.week <- 11ce(total_points ~ ict_index + bps + bonus + clean_sheets,</pre>
                   data = s_week1)
summary(lasso.week)$coefficients
                     Value Std. Error
##
                                         Z score Pr(>|Z|)
## (Intercept) 18.76383751 0.30058377 62.4246538 0.0000000
               0.13731522 0.00889584 15.4358908 0.0000000
## ict index
## bps
                0.07635565 0.00361839 21.1021064 0.0000000
## bonus
                0.05955156 0.09753627 0.6105581 0.5414922
## clean_sheets 0.00000000 0.15001673 0.0000000 1.0000000
```