Untitled

```
library(tufte)
library(ggplot2)
library(kableExtra)
## Warning: package 'kableExtra' was built under R version 3.5.2
library(ggcorrplot)
library(Matrix)
library(gridExtra)
library(tidyverse)
## -- Attaching packages -----
                                                                        ---- tidyverse 1.2.1 --
## v tibble 1.4.2
                                0.2.5
                      v purrr
## v tidyr 0.8.2
                      v dplyr
                                0.7.8
## v readr
           1.2.1
                      v stringr 1.3.1
## v tibble 1.4.2
                      v forcats 0.3.0
## -- Conflicts -----
                                                 ----- tidyverse_conflicts() --
## x dplyr::combine() masks gridExtra::combine()
## x tidyr::expand() masks Matrix::expand()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                     masks stats::lag()
library(scales)
## Attaching package: 'scales'
## The following object is masked from 'package:purrr':
##
##
       discard
## The following object is masked from 'package:readr':
##
##
       col_factor
library(MASS)
##
## Attaching package: 'MASS'
## The following object is masked from 'package:dplyr':
##
##
       select
library(matrixcalc)
library(psych)
##
## Attaching package: 'psych'
## The following objects are masked from 'package:scales':
##
       alpha, rescale
## The following objects are masked from 'package:ggplot2':
```

```
##
##
       %+%, alpha
library(GGally)
##
## Attaching package: 'GGally'
## The following object is masked from 'package:dplyr':
##
##
       nasa
library(ggpubr)
## Loading required package: magrittr
##
## Attaching package: 'magrittr'
## The following object is masked from 'package:purrr':
##
##
       set_names
## The following object is masked from 'package:tidyr':
##
##
       extract
library(leaps)
# Load in data
mb_train <- read.csv("./data/moneyball-training-data.csv")[,-1] # use me
# Removes all rows with missing data
mb_complete <- mb_train[complete.cases(mb_train),]</pre>
Means <- sapply(mb_complete, mean)</pre>
Stan_Dev <- sapply(mb_complete, sd)</pre>
Data_Summary <- summary(mb_train)</pre>
Boxplots <- ggplot(stack(mb_complete), aes(x=ind, y=values)) +</pre>
    geom_boxplot() +
    theme(axis.text.x = element_text(angle = 90, hjust = 1))
Point_plots <- ggplot(data=mb_complete, aes(x=TARGET_WINS)) +</pre>
    geom_point(aes(y=TEAM_BATTING_H, color="Base Hits by Batters"), alpha=0.1) +
    geom_point(aes(y=TEAM_BATTING_2B, color="Doubles by Batters"), alpha=0.1) +
    geom_point(aes(y=TEAM_BATTING_3B, color="Triples by Batters"), alpha=0.1) +
    geom_point(aes(y=TEAM_BATTING_HR, color="Homeruns by Batters"), alpha=0.1) +
    geom_point(aes(y=TEAM_BATTING_BB, color="Walks by Batters"), alpha=0.1) +
    geom_point(aes(y=TEAM_BATTING_SO, color="Strikeouts by Batters"), alpha=0.1) +
    geom_point(aes(y=TEAM_BASERUN_SB, color="Stolen Bases"), alpha=0.1) +
    geom_point(aes(y=TEAM_BASERUN_CS, color="Caught Stealing"), alpha=0.1) +
    geom_point(aes(y=TEAM_BATTING_HBP, color="Batters Hit by Pitch"), alpha=0.1) +
    geom_point(aes(y=TEAM_PITCHING_H, color="Hits Allowed"), alpha=0.1) +
    geom point(aes(y=TEAM PITCHING HR, color="Homeruns Allowed"), alpha=0.1) +
    geom_point(aes(y=TEAM_PITCHING_SO, color="Strikeouts by Pitchers"), alpha=0.1) +
    geom_point(aes(y=TEAM_FIELDING_E, color="Errors"), alpha=0.05) +
    geom_point(aes(y=TEAM_FIELDING_DP, color="Double Plays"), alpha=0.1) +
```

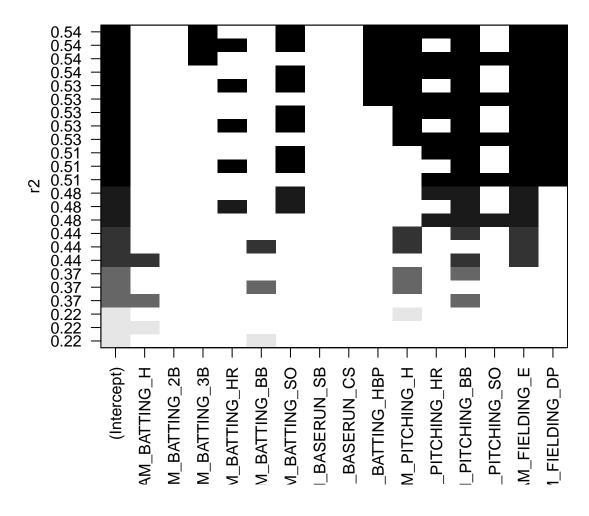
Linear Models

Unscaled

```
# Basic linear model with all variables
mb_lm <- lm(TARGET_WINS ~ ., mb_train)

LM_Summary <- summary(mb_lm)

# All Subsets Regression from leaps package
Leaps <- regsubsets(x=mb_complete[,2:16], y=mb_complete[,1], nbest=3)
# plot a table of models showing variables in each model.
# models are ordered by the selection statistic.
Leaps_plot <- plot(Leaps, scale="r2")</pre>
```



Scaled

```
# Scale all the predictor variables
mb_scaled <- as.data.frame(mb_complete[,2:16], center=Means, scale=Stan_Dev)
mb_scaled$TARGET_WINS <- mb_complete[,1]
# Linear model using all scaled predictors
mb_scaled_lm <- lm(TARGET_WINS ~ ., mb_scaled)

Scaled_LM_Summary <- summary(mb_scaled_lm)

# All Subsets Regression from leaps package on SCALED data
Scaled_Leaps <- regsubsets(x=mb_complete[,1:15], y=mb_scaled[,16], nbest=3)
# plot a table of models showing variables in each model.
# models are ordered by the selection statistic.
Scaled_Leaps_plot <- plot(Leaps, scale="r2")</pre>
```

