

Data621 - HW1

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Contents

Data Overview	1
Objective	2
Data Exploration	2
Feature Boxplots and Histograms	9
Missing values	11
Fitting a Linear Model	68
Final Model using all Training Data	76
Evaluation Data	78

```
library(dplyr)
library(DataExplorer)
library(GGally)
library(ggplot2)
library(readr)
library(reshape2)
library(purrr)
library(tidyr)
library(corrplot)
library(MASS)
library(caret)
```

Data Overview

The data set contains approximately 2276 records. Each record represents a professional baseball team from the years 1871 to 2006 inclusive. Each record has the performance of the team for the given year, with all of the statistics adjusted to match the performance of a 162 game season. Below is a short description of the variables

- INDEX - Identification Variable
- TARGET_WINS - Number of wins
- TEAM_BATTING_H - Base Hits by batters (1B,2B,3B,HR)
- TEAM_BATTING_2B - Doubles by batters (2B)
- TEAM_BATTING_3B - Triples by batters (3B)
- TEAM_BATTING_HR - Homeruns by batters (4B)

- TEAM_BATTING_BB - Walks by batters
- TEAM_BATTING_HBP - Batters hit by pitch (get a free base)
- TEAM_BATTING_SO - Strikeouts by batters
- TEAM_BASERUN_SB - Stolen bases
- TEAM_BASERUN_CS - Caught stealing
- TEAM_FIELDING_E - Errors
- TEAM_FIELDING_DP - Double Plays
- TEAM_PITCHING_BB - Walks allowed
- TEAM_PITCHING_H - Hits allowed
- TEAM_PITCHING_HR - Homeruns allowed
- TEAM_PITCHING_SO - Strikeouts by pitchers

Objective

To build a multiple linear regression model on the training data to predict *TARGET_WINS*, which is the number of wins for the team.

Data Exploration

```
# read data
baseball_df <- read.csv('https://raw.githubusercontent.com/hillt5/DATA_621/master/HW1/moneyball-training')
baseball_eval <- read.csv('https://raw.githubusercontent.com/hillt5/DATA_621/master/HW1/moneyball-evaluation')

head(baseball_df)
```

```
##      INDEX TARGET_WINS TEAM_BATTING_H TEAM_BATTING_2B TEAM_BATTING_3B
## 1         1          39           1445           194           39
## 2         2          70           1339           219           22
## 3         3          86           1377           232           35
## 4         4          70           1387           209           38
## 5         5          82           1297           186           27
## 6         6          75           1279           200           36
##      TEAM_BATTING_HR TEAM_BATTING_BB TEAM_BATTING_SO TEAM_BASERUN_SB
## 1                 13             143             842             NA
## 2                 190             685            1075             37
## 3                 137             602             917             46
## 4                 96             451             922             43
## 5                 102             472             920             49
## 6                 92             443             973            107
##      TEAM_BASERUN_CS TEAM_BATTING_HBP TEAM_PITCHING_H TEAM_PITCHING_HR
## 1                 NA                NA             9364             84
## 2                 28                NA             1347            191
## 3                 27                NA             1377            137
## 4                 30                NA             1396             97
## 5                 39                NA             1297            102
## 6                 59                NA             1279             92
##      TEAM_PITCHING_BB TEAM_PITCHING_SO TEAM_FIELDING_E TEAM_FIELDING_DP
## 1                 927             5456             1011             NA
## 2                 689             1082              193            155
## 3                 602              917              175            153
## 4                 454              928              164            156
```

```
## 5          472          920          138          168
## 6          443          973          123          149
```

```
dim(baseball_df)
```

```
## [1] 2276  17
```

```
summary(baseball_df)
```

```
##      INDEX      TARGET_WINS      TEAM_BATTING_H TEAM_BATTING_2B
## Min.   : 1.0   Min.   : 0.00   Min.   : 891   Min.   : 69.0
## 1st Qu.: 630.8 1st Qu.: 71.00   1st Qu.:1383   1st Qu.:208.0
## Median :1270.5 Median : 82.00   Median :1454   Median :238.0
## Mean   :1268.5 Mean   : 80.79   Mean   :1469   Mean   :241.2
## 3rd Qu.:1915.5 3rd Qu.: 92.00   3rd Qu.:1537   3rd Qu.:273.0
## Max.   :2535.0 Max.   :146.00   Max.   :2554   Max.   :458.0
##
## TEAM_BATTING_3B TEAM_BATTING_HR TEAM_BATTING_BB TEAM_BATTING_SO
## Min.   : 0.00   Min.   : 0.00   Min.   : 0.0   Min.   : 0.0
## 1st Qu.: 34.00   1st Qu.: 42.00   1st Qu.:451.0   1st Qu.: 548.0
## Median : 47.00   Median :102.00   Median :512.0   Median : 750.0
## Mean   : 55.25   Mean   : 99.61   Mean   :501.6   Mean   : 735.6
## 3rd Qu.: 72.00   3rd Qu.:147.00   3rd Qu.:580.0   3rd Qu.: 930.0
## Max.   :223.00   Max.   :264.00   Max.   :878.0   Max.   :1399.0
##
##                                     NA's :102
## TEAM_BASERUN_SB TEAM_BASERUN_CS TEAM_BATTING_HBP TEAM_PITCHING_H
## Min.   : 0.0   Min.   : 0.0   Min.   :29.00   Min.   : 1137
## 1st Qu.: 66.0   1st Qu.: 38.0   1st Qu.:50.50   1st Qu.: 1419
## Median :101.0   Median : 49.0   Median :58.00   Median : 1518
## Mean   :124.8   Mean   : 52.8   Mean   :59.36   Mean   : 1779
## 3rd Qu.:156.0   3rd Qu.: 62.0   3rd Qu.:67.00   3rd Qu.: 1682
## Max.   :697.0   Max.   :201.0   Max.   :95.00   Max.   :30132
## NA's :131     NA's :772     NA's :2085
## TEAM_PITCHING_HR TEAM_PITCHING_BB TEAM_PITCHING_SO TEAM_FIELDING_E
## Min.   : 0.0   Min.   : 0.0   Min.   : 0.0   Min.   : 65.0
## 1st Qu.: 50.0   1st Qu.: 476.0   1st Qu.: 615.0   1st Qu.: 127.0
## Median :107.0   Median : 536.5   Median : 813.5   Median : 159.0
## Mean   :105.7   Mean   : 553.0   Mean   : 817.7   Mean   : 246.5
## 3rd Qu.:150.0   3rd Qu.: 611.0   3rd Qu.: 968.0   3rd Qu.: 249.2
## Max.   :343.0   Max.   :3645.0   Max.   :19278.0   Max.   :1898.0
##
##                                     NA's :102
## TEAM_FIELDING_DP
## Min.   : 52.0
## 1st Qu.:131.0
## Median :149.0
## Mean   :146.4
## 3rd Qu.:164.0
## Max.   :228.0
## NA's :286
```

```
print('Number of observations:')
```

```
## [1] "Number of observations:"
```

```
nrow(baseball_df)
```

```
## [1] 2276
```

```
print('Observations per year, 1871 - 2006:')
```

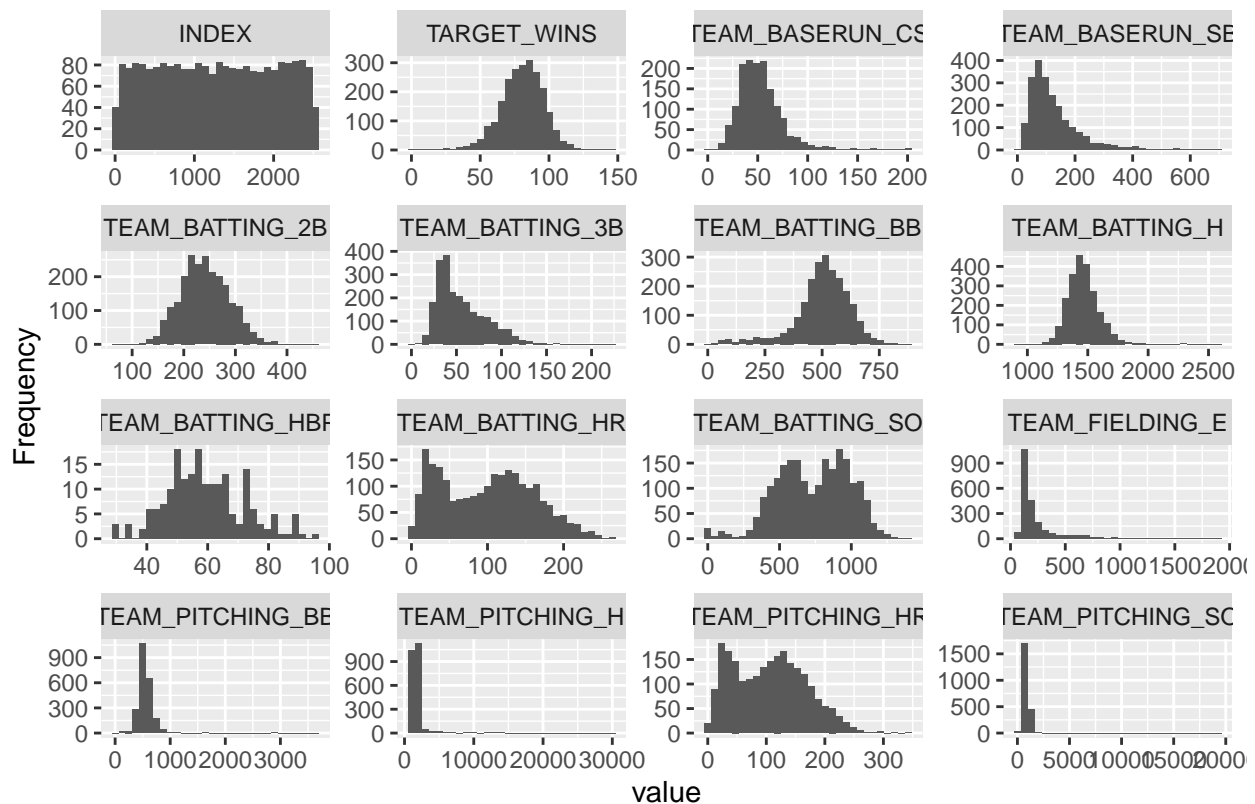
```
## [1] "Observations per year, 1871 - 2006:"
```

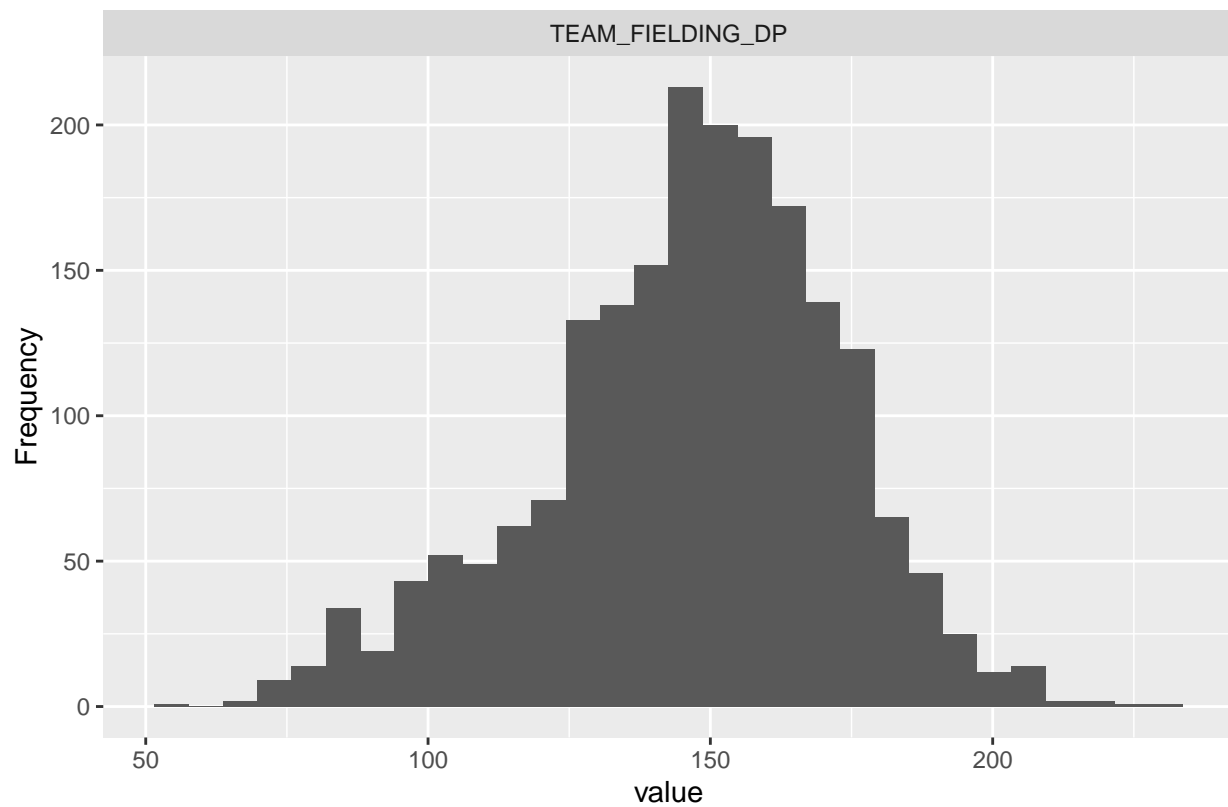
```
round(nrow(baseball_df)/(2006-1871),2)
```

```
## [1] 16.86
```

Some columns have maximum values that are clearly outliers, like `TEAM_PITCHING_H` AND `TEAM_PITCHING_HR`. The assignment mentions that some of the season records were adjusted to match the performance during a 162-game season. There are 2276 seasons in the training set. Observations span 128 years, with an average of 17 teams playing per year. Based on a quick Google search, there were initially 8 teams in the MLB, and 30 teams in 2006. The MLB has two leagues of the same size since 1901, with the league sizes increasing in the late 20th century.

```
# distribution  
plot_histogram(baseball_df)
```

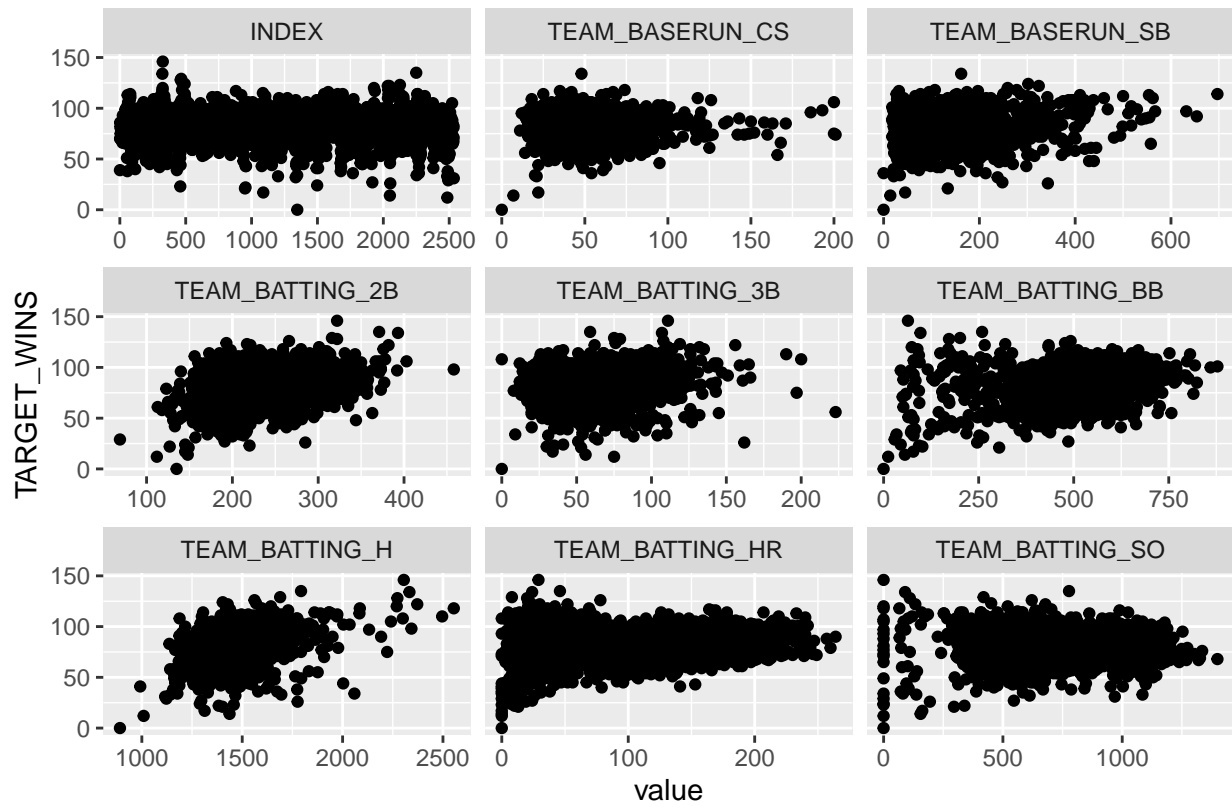




Page 2

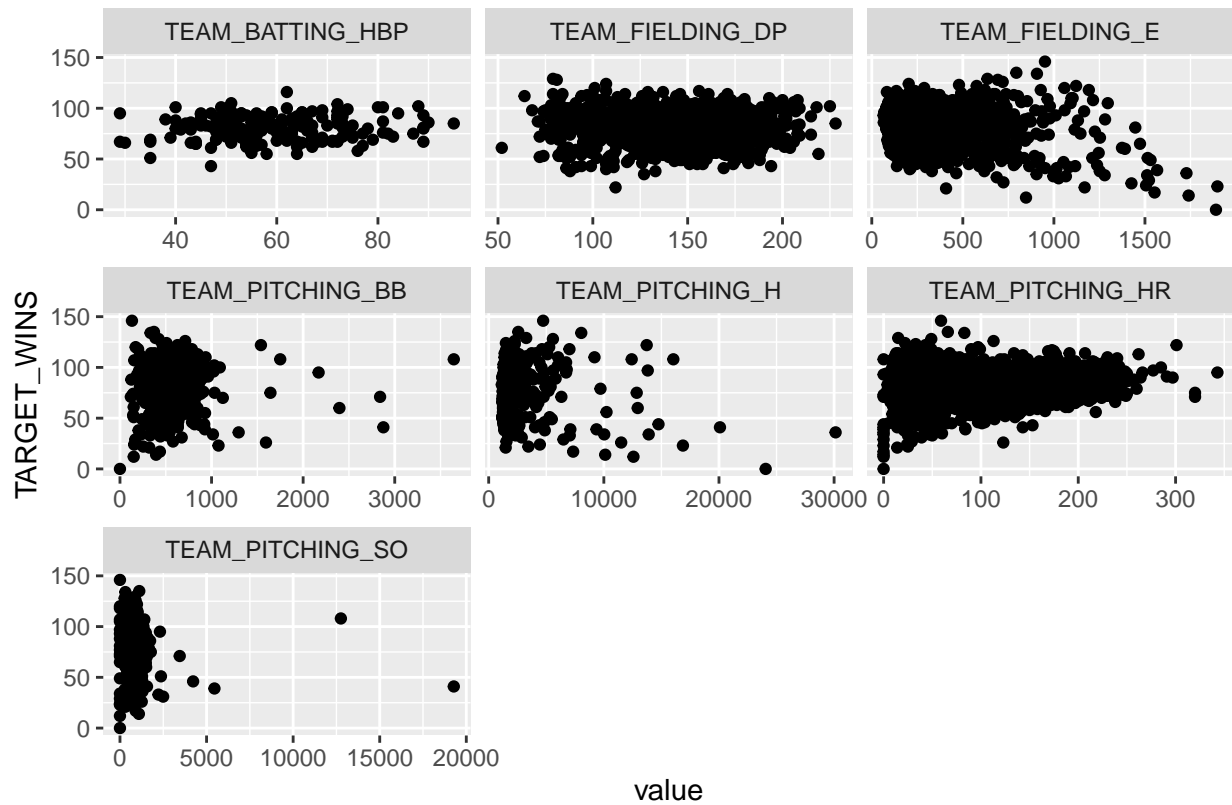
```
# against the response variable  
plot_scatterplot(baseball_df, by = "TARGET_WINS")
```

```
## Warning: Removed 1005 rows containing missing values (geom_point).
```



Page 1

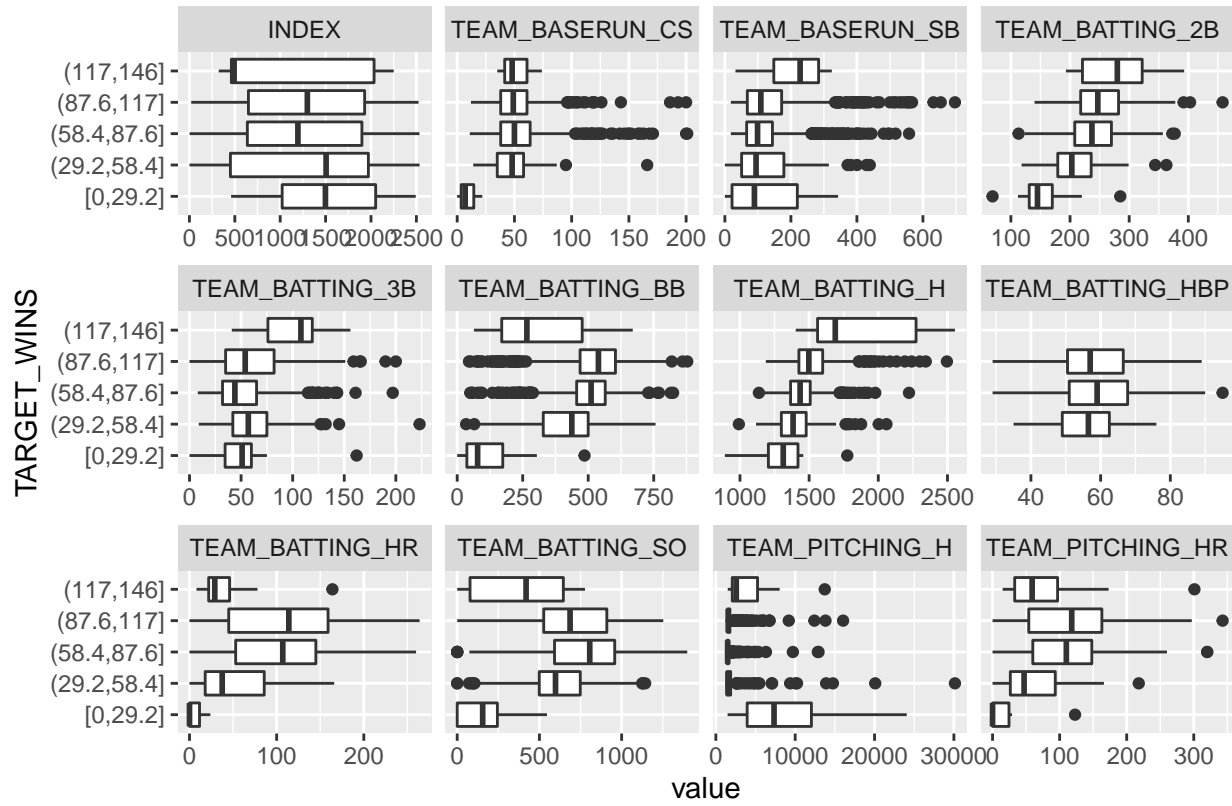
```
## Warning: Removed 2473 rows containing missing values (geom_point).
```



Page 2

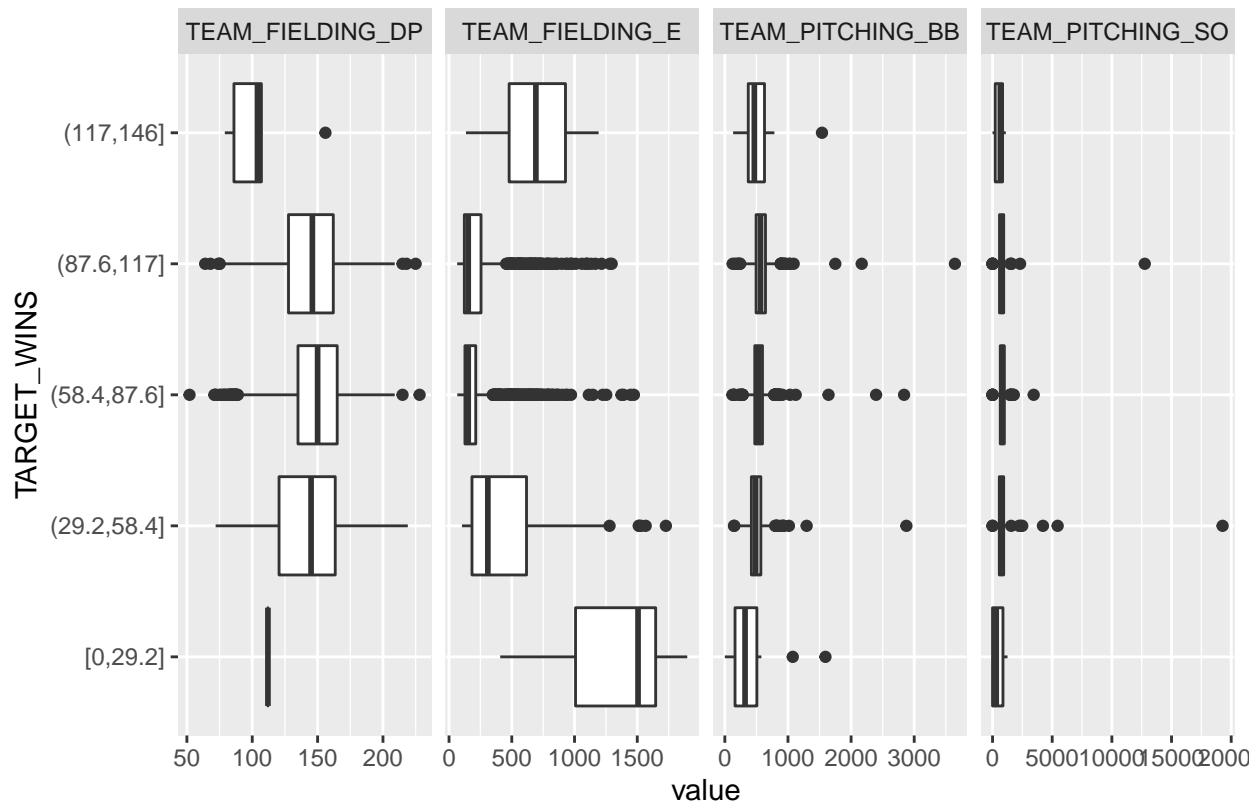
```
# boxplot for train dataset
plot_boxplot(baseball_df, by="TARGET_WINS")
```

```
## Warning: Removed 3090 rows containing non-finite values (stat_boxplot).
```



Page 1

Warning: Removed 388 rows containing non-finite values (stat_boxplot).



Page 2

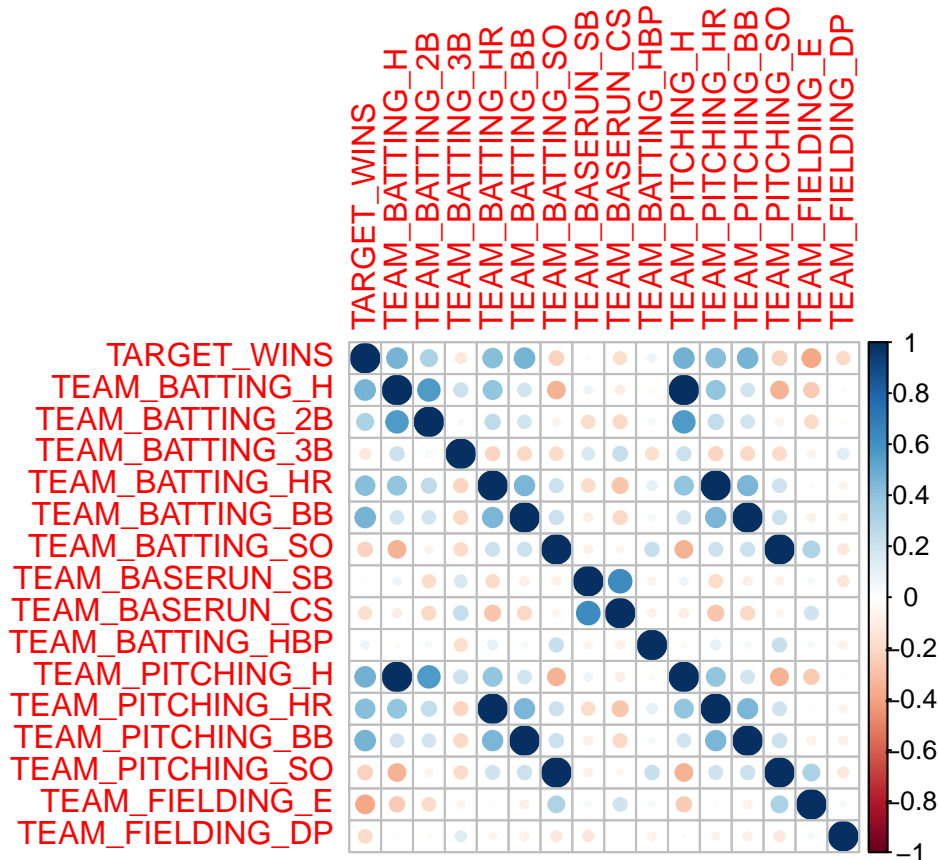
Feature Boxplots and Histograms

```
#baseball_df %>%
# keep(is.numeric) %>%
# gather() %>%
# ggplot(aes(value)) +
#   facet_wrap(~ key, scales = "free") +
#   geom_boxplot()
```

```
#baseball_df %>%
# keep(is.numeric) %>%
# gather() %>%
# ggplot(aes(value)) +
#   facet_wrap(~ key, scales = "free") +
#   geom_histogram()
```

Based on this quick look of boxplots and histograms, there are several variables with skewed distributions that would benefit from transformation. Additionally, there are a few variables with bimodal distributions.

```
corrplot(cor(baseball_df[,2:17], use = 'complete.obs'))
```



Looking at the correlation plot, there appear to be several strong correlations between explanatory variables and the target.

From an initial inspection, it appears the team should focus on getting players on base through hits or walks. Contrary to what I would expect, teams can still win if the pitchers allow homeruns, hits and walks to the other team. *Variables with Highest Positive Correlation with TARGET_WINS:*

* TEAM_BATTING_H = 0.47 * TEAM_BATTING_HR = 0.42 * TEAM_BATTING_BB = 0.47 * TEAM_PITCHING_H = 0.47 * TEAM_PITCHING_HR = 0.42 * TEAM_PITCHING_BB = 0.47

To win more games it makes sense the team will need to make fewer errors.

Variables with Strongly Negative Correlation with TARGET_WINS:

There were several batting variables which were related.

Positive Correlations between variables:

* TEAM_PITCHING_H and TEAM_BATTING_H = 0.99
 * TEAM_PITCHING_HR and TEAM_BATTING_HR = 0.99
 * TEAM_PITCHING_BB and TEAM_BATTING_BB = 0.99
 * TEAM_PITCHING_SO and TEAM_BATTING_SO = 0.99

The pitchers who have more strikeouts allow fewer hits, which makes sense. It's interesting that pitchers who have fewer strikeouts have fewer team batting hits. Potentially due to the game being over in fewer innings and lower score games. This would be an interesting point to look into more.

Negative Correlations between variables:

* TEAM_PITCHING_SO and TEAM_BATTING_H = -0.34
 * TEAM_PITCHING_SO and TEAM_PITCHING_H = -0.34

Missing values

```
round(100*colSums(is.na(baseball_df))/nrow(baseball_df),2)
```

```
##          INDEX      TARGET_WINS  TEAM_BATTING_H  TEAM_BATTING_2B
##          0.00          0.00          0.00          0.00
## TEAM_BATTING_3B TEAM_BATTING_HR  TEAM_BATTING_BB  TEAM_BATTING_SO
##          0.00          0.00          0.00          4.48
## TEAM_BASERUN_SB TEAM_BASERUN_CS TEAM_BATTING_HBP  TEAM_PITCHING_H
##          5.76          33.92          91.61          0.00
## TEAM_PITCHING_HR TEAM_PITCHING_BB TEAM_PITCHING_SO  TEAM_FIELDING_E
##          0.00          0.00          4.48          0.00
## TEAM_FIELDING_DP
##          12.57
```

In terms of missing values, there are two variables missing many observations. `TEAM_BATTING_HBP` is missing over 90% of its values, while `TEAM_BASERUN_CS` is missing just around 30%. Since so many observations are missing, imputing values could change the distributions considerably. To retain as many features as possible, I think it makes sense to explore these two variables first. The other affected missing values only have 5-10% missing values. None of these appear to be stand-ins for ‘zero’ values, so mean values can be used instead.

I’ll start by looking at `TEAM_BATTING_HBP`.

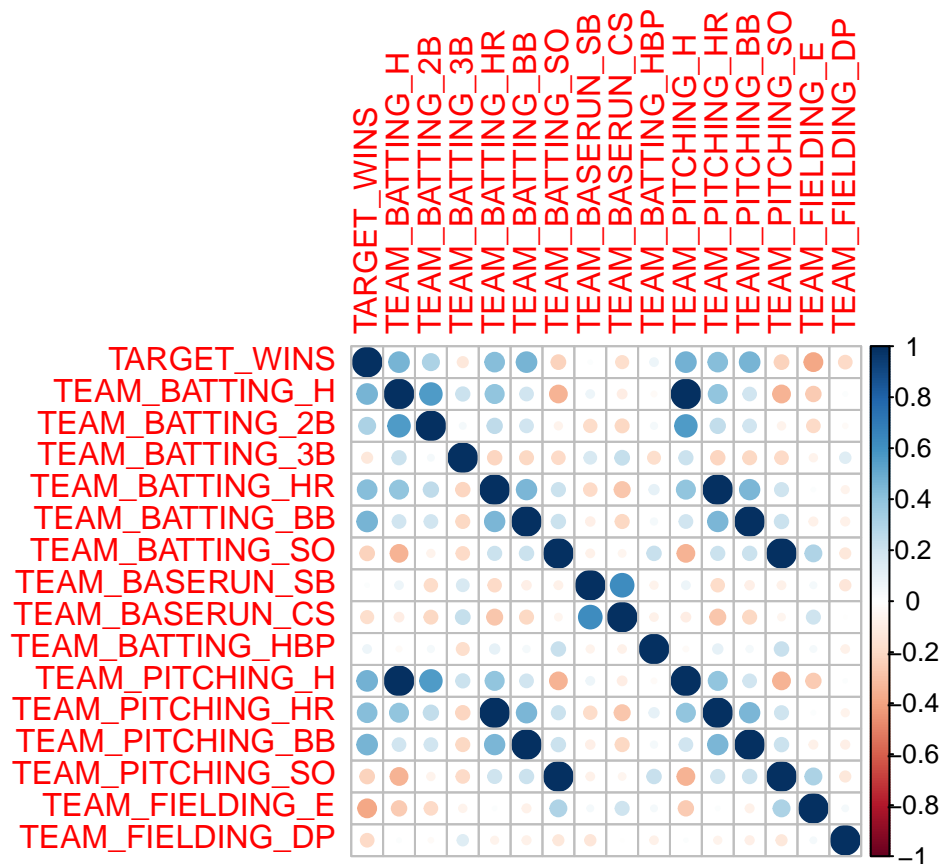
```
baseball_no_hbp <- baseball_df %>%
  filter(is.na(TEAM_BATTING_HBP)) %>% #missing values for hbp
dplyr::select(-TEAM_BATTING_HBP) ## select all rows except hbp

baseball_hbp <- baseball_df %>%
  filter(!is.na(TEAM_BATTING_HBP)) #not missing values for hbp
```

I separated training data into two smaller dataframes, one with complete values for HBP and one omitting this variable.

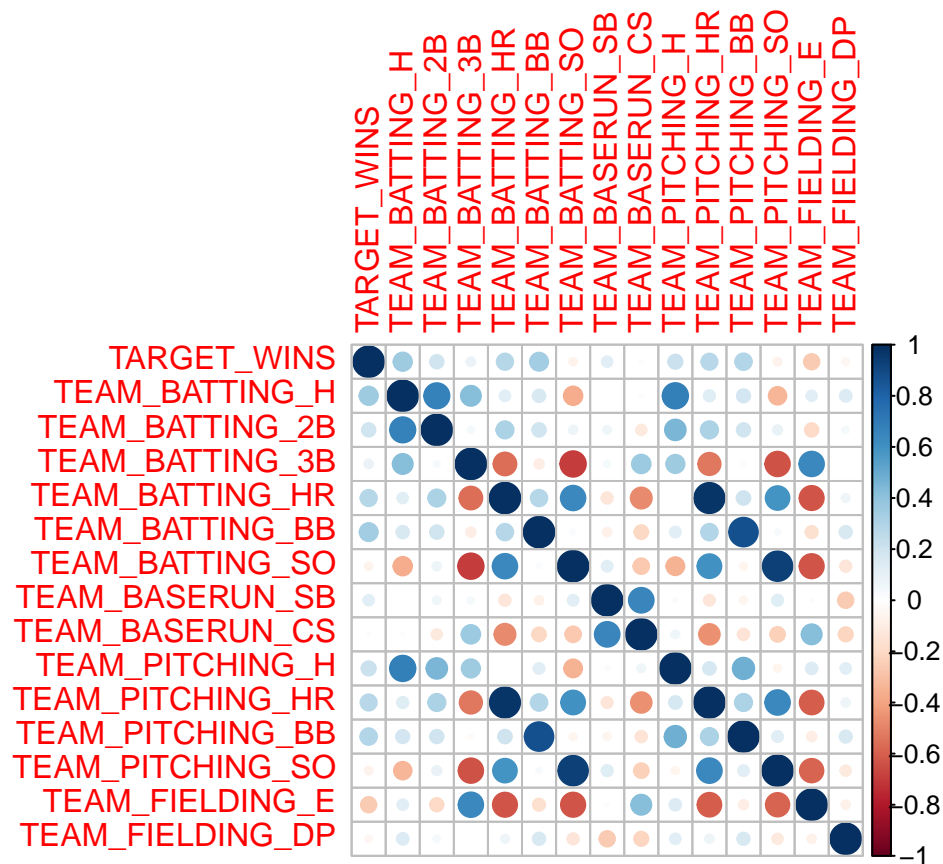
```
corrplot(cor(baseball_df[,2:17], use = 'complete.obs'))

corrplot(cor(baseball_hbp[,2:17], use = 'complete.obs'))
```

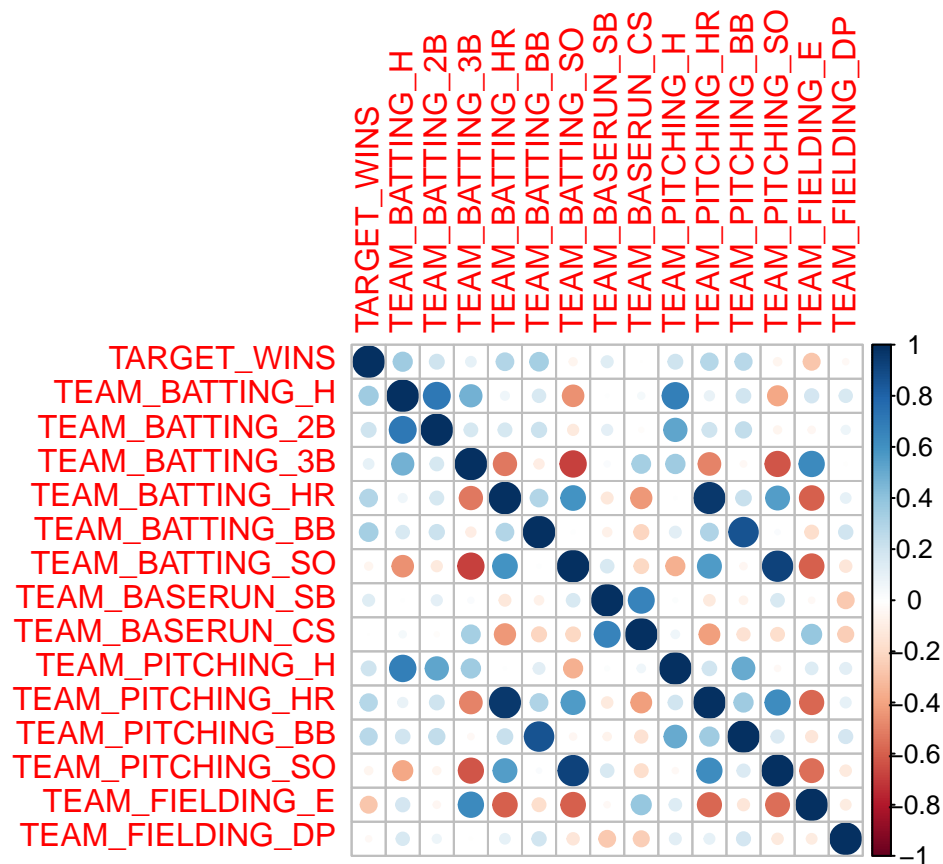


When HBP has values, it appears that there are no major changes in correlations.

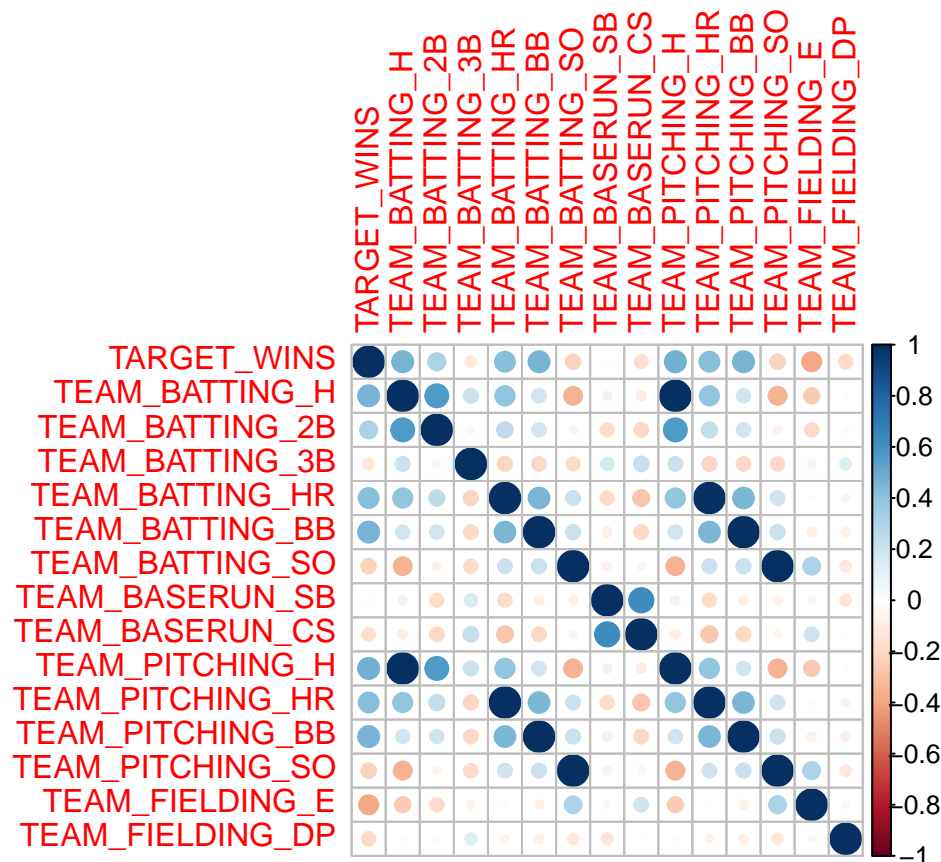
```
corrplot(cor(baseball_df[, -c(1,11)], use = 'complete.obs')) #all rows
```



```
corrplot(cor(baseball_no_hbp[,2:16], use = 'complete.obs')) #only rows missing values for hbp
```



```
corrplot(cor(baseball_hbp[, -c(1,11)], use = 'complete.obs')) #only rows with values for hbp
```



There are three new correlaton plots being considered: the first is all datapoints, then a plot with missing hbp values, and finally a plot for rows with hbp values same as the previous pair. There appear to be no major discrepancies between missing values and the overall set. However, comparing missing values to available values does illustrate there are some distinct changes correlation when the hbp was recorded. This may be because HBP only represents only a small proportion of entries and has more variation. However, there also appear to be stronger correlations when HBP is added, which may help predict wins better than omitting altogether.

```
hbp_lm <- lm(baseball_hbp, formula = TARGET_WINS ~.-INDEX-TEAM_BATTING_HBP-TEAM_BATTING_SO-TEAM_BATTING_HR-TEAM_BASERUN_CS-TEAM_PITCHING_BB-TEAM_BATTING_2B-TEAM_BATTING_3B, data = baseball_hbp)
```

```
summary(hbp_lm)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ . - INDEX - TEAM_BATTING_HBP - TEAM_BATTING_SO -
##     TEAM_BATTING_HR - TEAM_BASERUN_CS - TEAM_BATTING_H - TEAM_BASERUN_SB -
##     TEAM_PITCHING_BB - TEAM_BATTING_2B - TEAM_BATTING_3B, data = baseball_hbp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -20.8123  -5.9942  -0.0737   5.3098  22.2025
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    62.916227   19.219854   3.274 0.001269 **
```

```
## TEAM_BATTING_BB    0.055959    0.009466    5.912 1.61e-08 ***
## TEAM_PITCHING_H    0.026147    0.010184    2.567 0.011041 *
## TEAM_PITCHING_HR   0.091571    0.024033    3.810 0.000189 ***
## TEAM_PITCHING_SO  -0.028772    0.007191   -4.001 9.13e-05 ***
## TEAM_FIELDING_E    -0.173897    0.039905   -4.358 2.18e-05 ***
## TEAM_FIELDING_DP  -0.121570    0.035338   -3.440 0.000719 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.473 on 184 degrees of freedom
## Multiple R-squared:  0.5263, Adjusted R-squared:  0.5109
## F-statistic: 34.07 on 6 and 184 DF,  p-value: < 2.2e-16
```

```
baseball_hbp_dummy <- baseball_df %>%
  mutate(TEAM_HBP_YES_NO = case_when(!is.na(TEAM_BATTING_HBP) ~ 1, is.na(TEAM_BATTING_HBP) ~ 0)) %>%
  dplyr::select(-TEAM_BATTING_HBP)

summary(lm(baseball_hbp_dummy, formula = TARGET_WINS ~.-INDEX-TEAM_PITCHING_BB-TEAM_PITCHING_HR-TEAM_BA
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ . - INDEX - TEAM_PITCHING_BB - TEAM_PITCHING_HR -
##     TEAM_BATTING_H, data = baseball_hbp_dummy)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -30.3981  -6.6295   0.1545   6.4842  28.2220
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    58.585629    6.442796   9.093 < 2e-16 ***
## TEAM_BATTING_2B  -0.060017    0.009747  -6.158 9.50e-10 ***
## TEAM_BATTING_3B   0.166293    0.022021   7.552 7.51e-14 ***
## TEAM_BATTING_HR   0.100869    0.009172  10.998 < 2e-16 ***
## TEAM_BATTING_BB   0.038251    0.003366  11.363 < 2e-16 ***
## TEAM_BATTING_SO   0.040704    0.009102   4.472 8.35e-06 ***
## TEAM_BASERUN_SB   0.034100    0.008689   3.924 9.10e-05 ***
## TEAM_BASERUN_CS   0.052980    0.018176   2.915 0.00361 **
## TEAM_PITCHING_H    0.031740    0.004269   7.435 1.76e-13 ***
## TEAM_PITCHING_SO  -0.058995    0.007547  -7.817 1.02e-14 ***
## TEAM_FIELDING_E   -0.158154    0.009939 -15.912 < 2e-16 ***
## TEAM_FIELDING_DP  -0.112916    0.013095  -8.623 < 2e-16 ***
## TEAM_HBP_YES_NO  -2.456525    0.923761  -2.659 0.00792 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.532 on 1473 degrees of freedom
## (790 observations deleted due to missingness)
## Multiple R-squared:  0.4407, Adjusted R-squared:  0.4361
## F-statistic: 96.71 on 12 and 1473 DF,  p-value: < 2.2e-16
```

I compared two preliminary linear models that I arrived at through backward selection. Looking only at

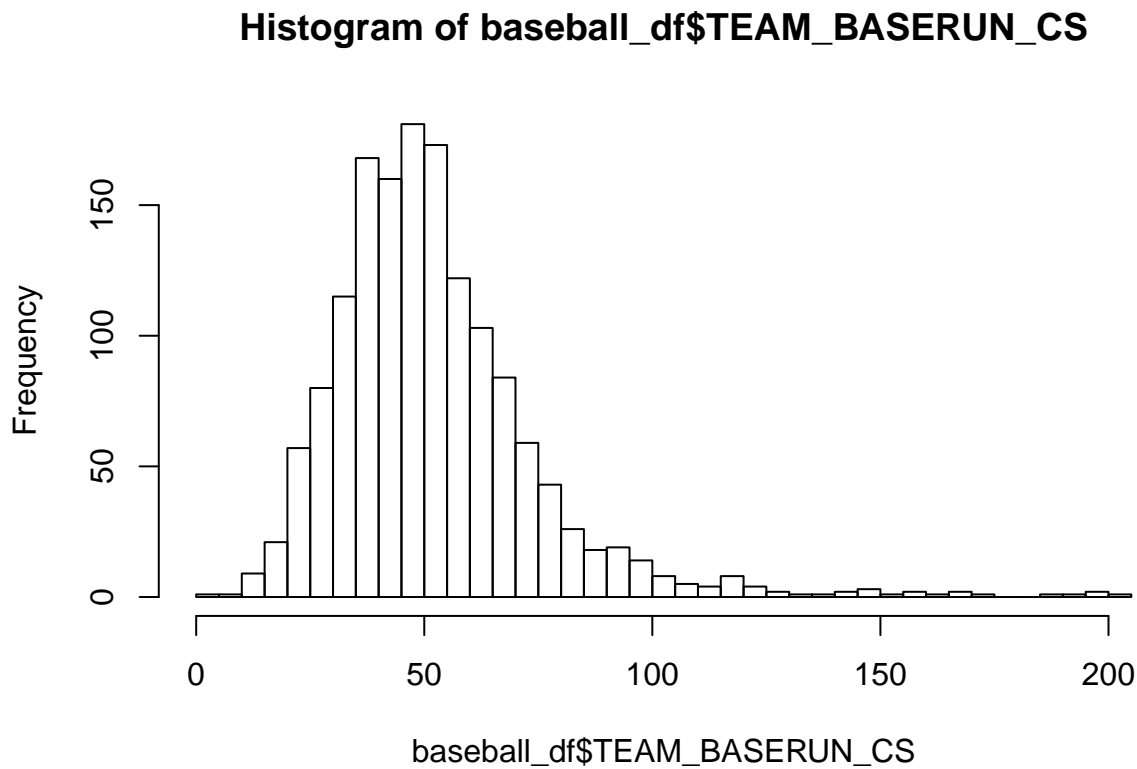
HBP-containing observations, there's a small increase in r-squared compared to a model that uses a dummy variable to consider whether values were available.

Next, I'll look at TEAM_BASERUN_CS, which was missing about 30% of its values.

```
sum(baseball_df$TEAM_BASERUN_CS == 0, na.rm = TRUE)
```

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

```
hist(baseball_df$TEAM_BASERUN_CS, breaks = 30)
```



```
baseball_no_cs <- baseball_df %>%  
  filter(is.na(Team_BASERUN_CS)) %>% #missing values for hbp  
  dplyr::select(-Team_BASERUN_CS) ## select all rows except hbp
```

```
baseball_cs <- baseball_df %>%  
  filter(!is.na(Team_BASERUN_CS)) #not missing values for hbp
```

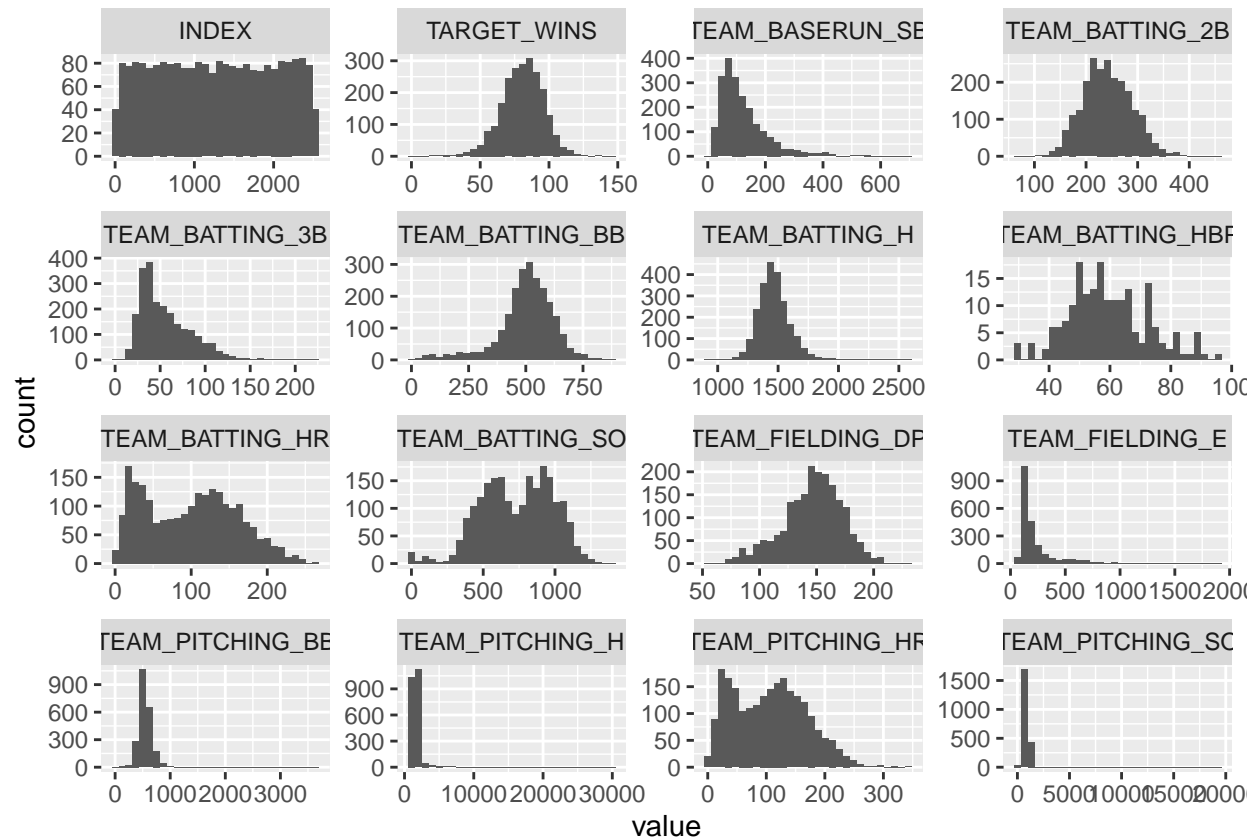
Same as HBP, it appears CS did not miscode values of 0 as NA. I'm going to separate the dataset in the same way as HBP to see if there are differences in its distribution and correlation plots.

```
baseball_df %>% ##original histograms  
  dplyr::select(-Team_BASERUN_CS) %>%  
  keep(is.numeric) %>%  
  gather() %>%  
  ggplot(aes(value)) +
```

```
facet_wrap(~ key, scales = "free") +
geom_histogram()
```

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

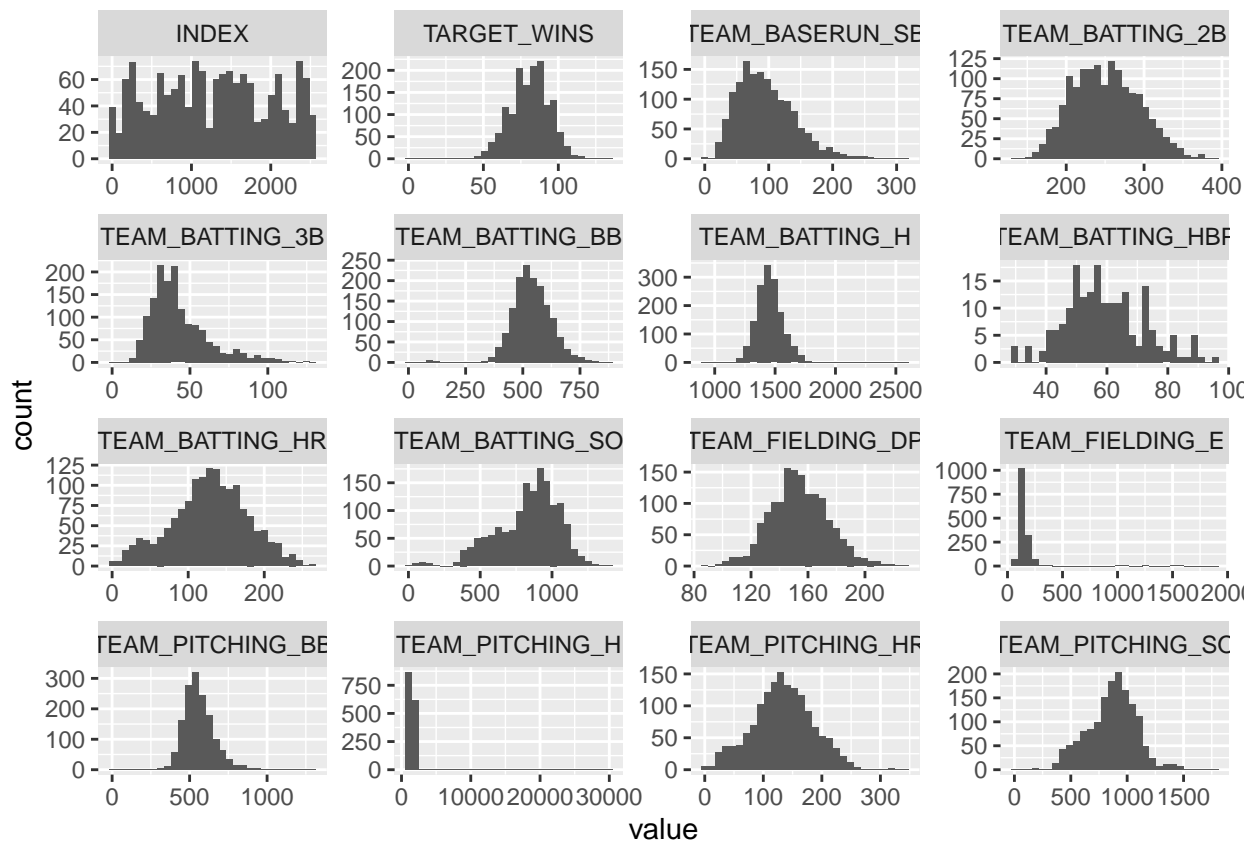
```
## Warning: Removed 2706 rows containing non-finite values (stat_bin).
```



```
baseball_cs %>% ##histograms with seasons having CS statistics
dplyr::select(-TEAM_BASERUN_CS) %>%
keep(is.numeric) %>%
gather() %>%
ggplot(aes(value)) +
  facet_wrap(~ key, scales = "free") +
  geom_histogram()
```

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

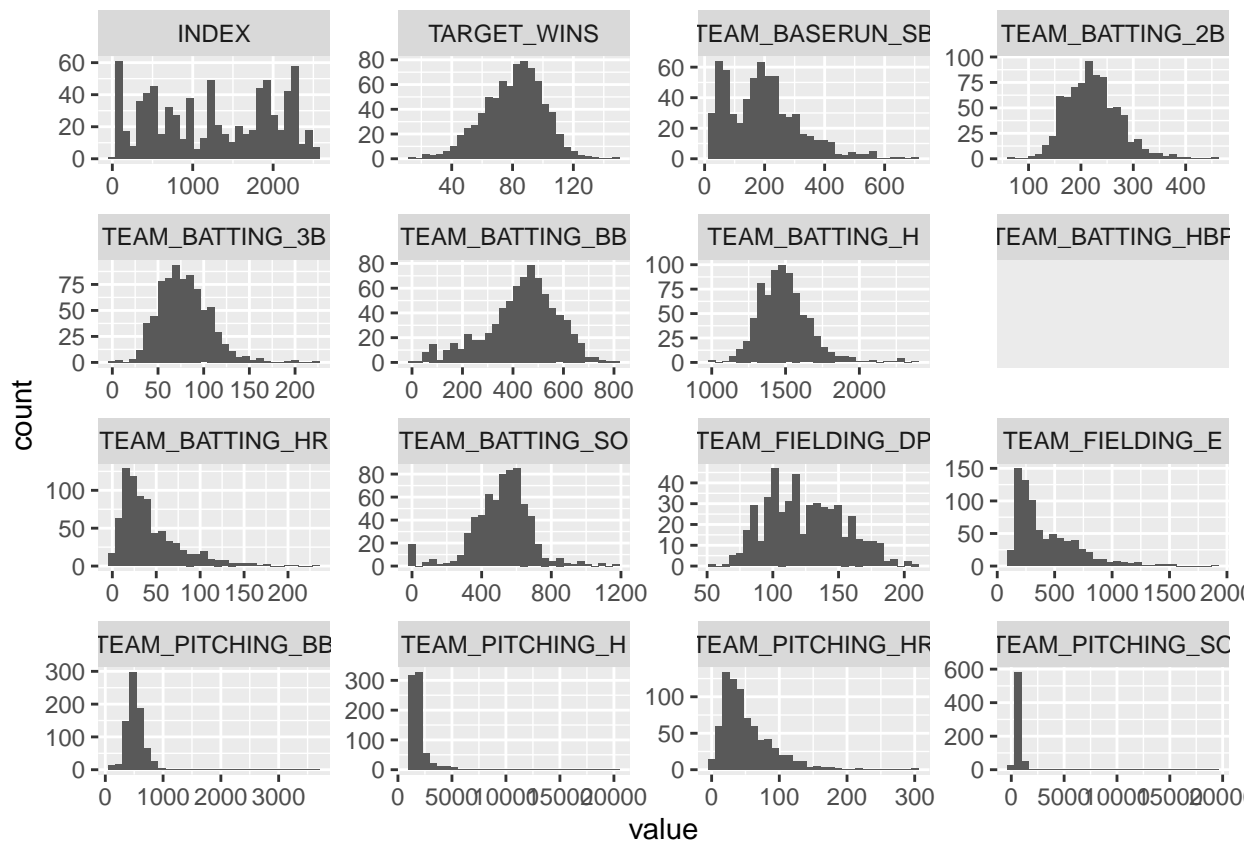
```
## Warning: Removed 1331 rows containing non-finite values (stat_bin).
```



```
baseball_no_cs %>% #histograms missing CS statistics
  keep(is.numeric) %>%
  gather() %>%
  ggplot(aes(value)) +
    facet_wrap(~ key, scales = "free") +
    geom_histogram()
```

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

```
## Warning: Removed 1375 rows containing non-finite values (stat_bin).
```

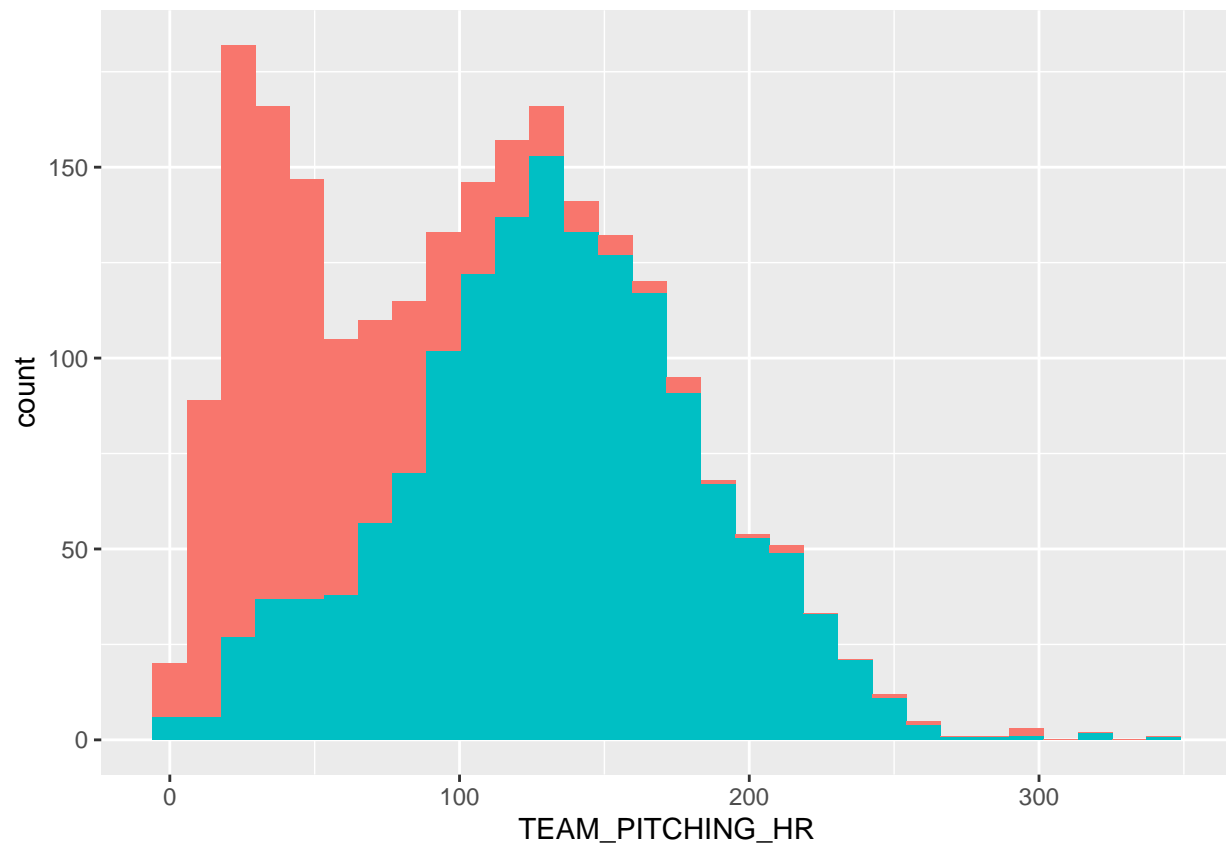


After subsetting for availability of CS statistics, an interesting pattern emerges: our three bimodal variables, TEAM_PITCHING_HR, TEAM_BATTING_SO, and TEAM_BATTING_HR, are no longer bimodal.

```
baseball_hbp_dummy <- baseball_hbp_dummy %>%
  mutate(TEAM_CS_YES_NO = case_when(!is.na(TEAM_BASERUN_CS) ~ '1', is.na(TEAM_BASERUN_CS) ~ '0'))

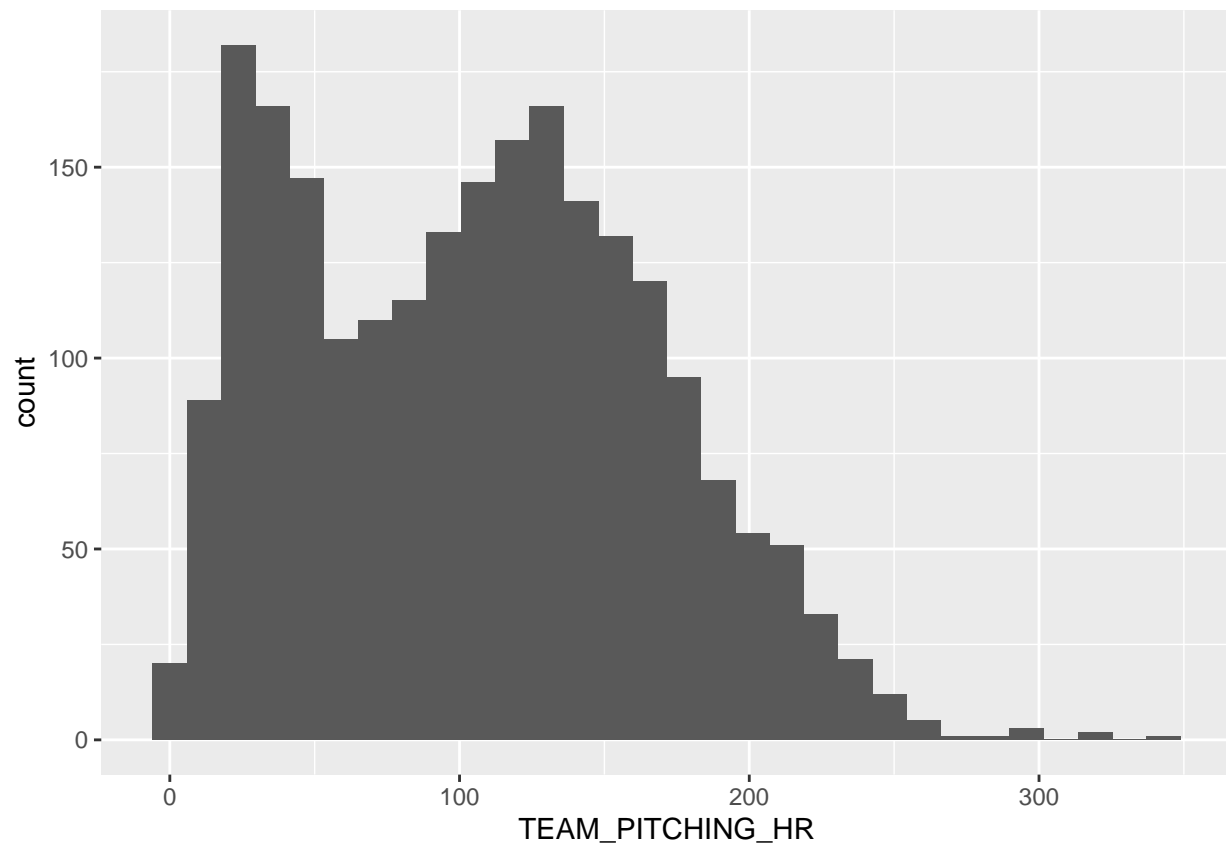
ggplot(baseball_hbp_dummy, aes(x = TEAM_PITCHING_HR, fill = TEAM_CS_YES_NO)) +
  geom_histogram() +
  theme(legend.position = 'none')
```

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



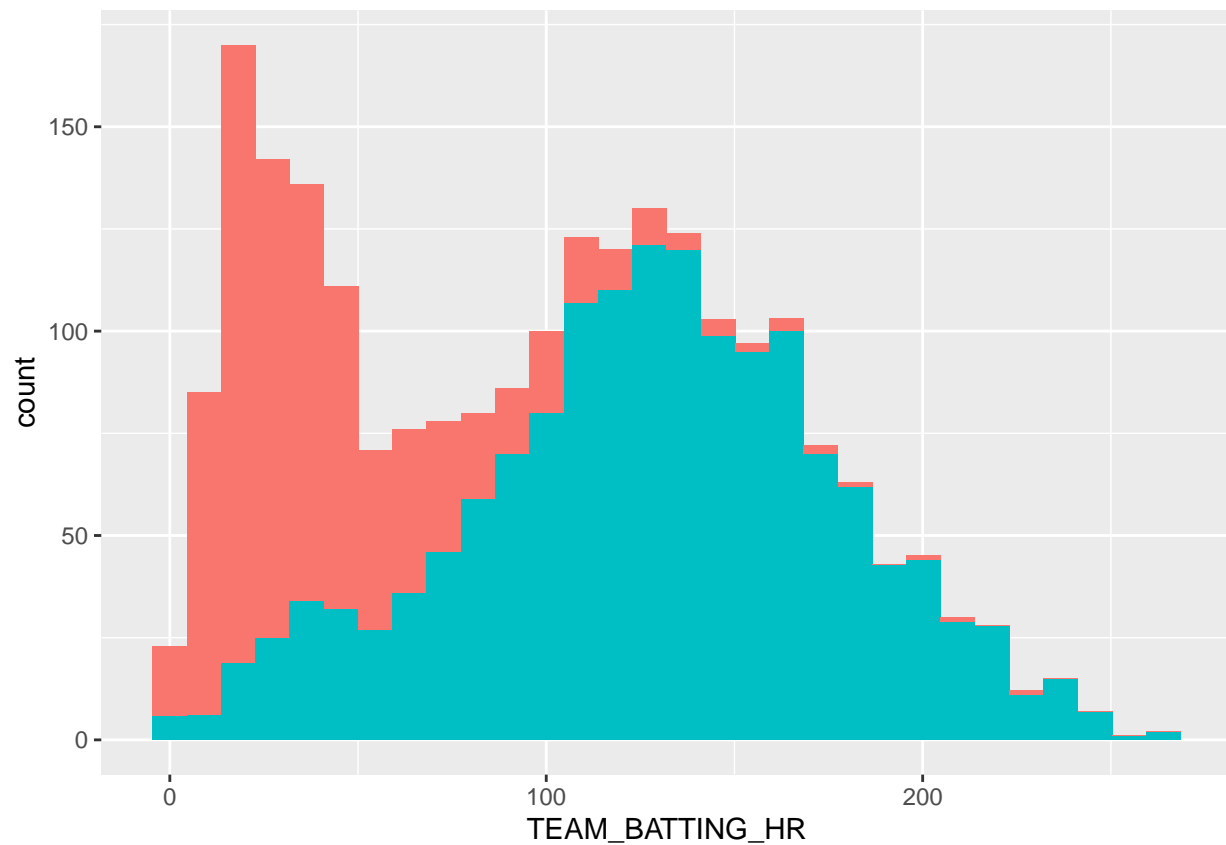
```
ggplot(baseball_df, aes(x = TEAM_PITCHING_HR)) +  
  geom_histogram()
```

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



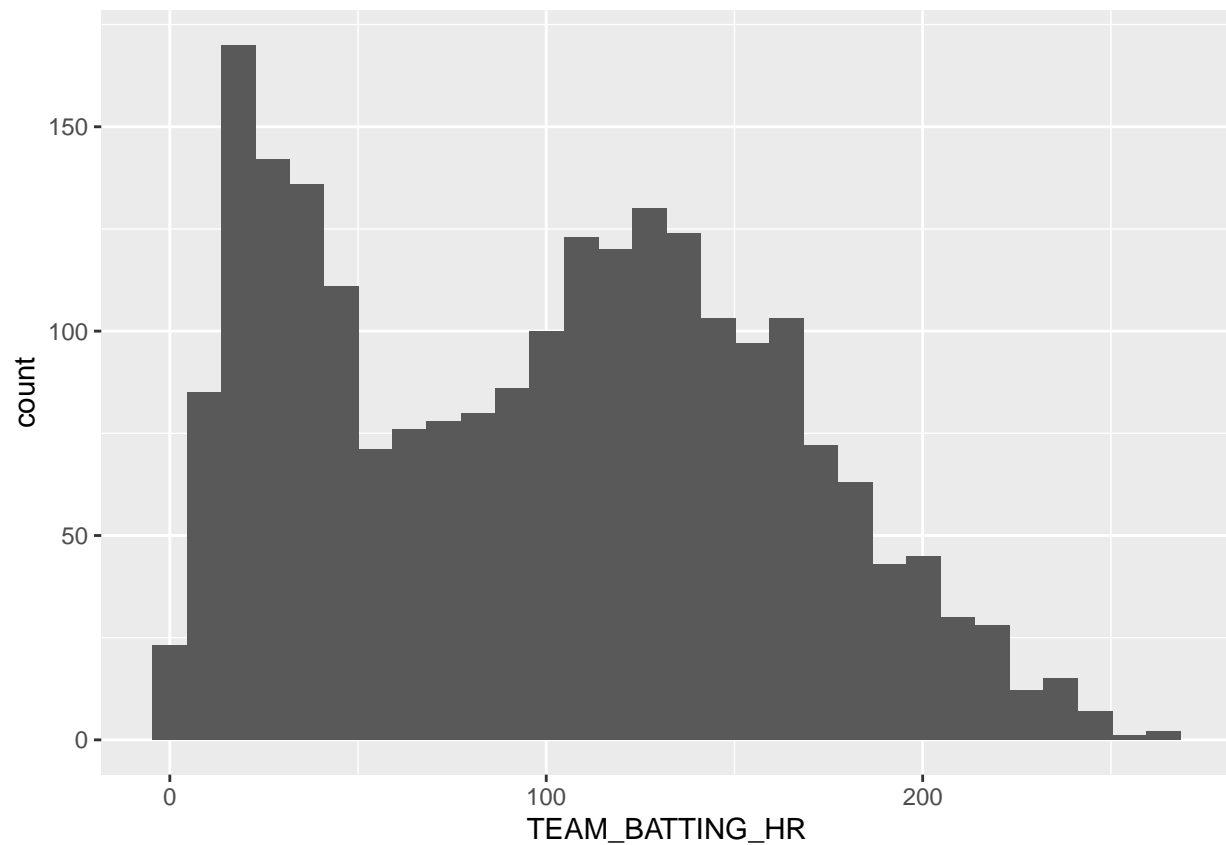
```
ggplot(baseball_hbp_dummy, aes(x = TEAM_BATTING_HR, fill = TEAM_CS_YES_NO)) +  
  geom_histogram() +  
  theme(legend.position = 'none')
```

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



```
ggplot(baseball_df, aes(x = TEAM_BATTING_HR)) +  
  geom_histogram()
```

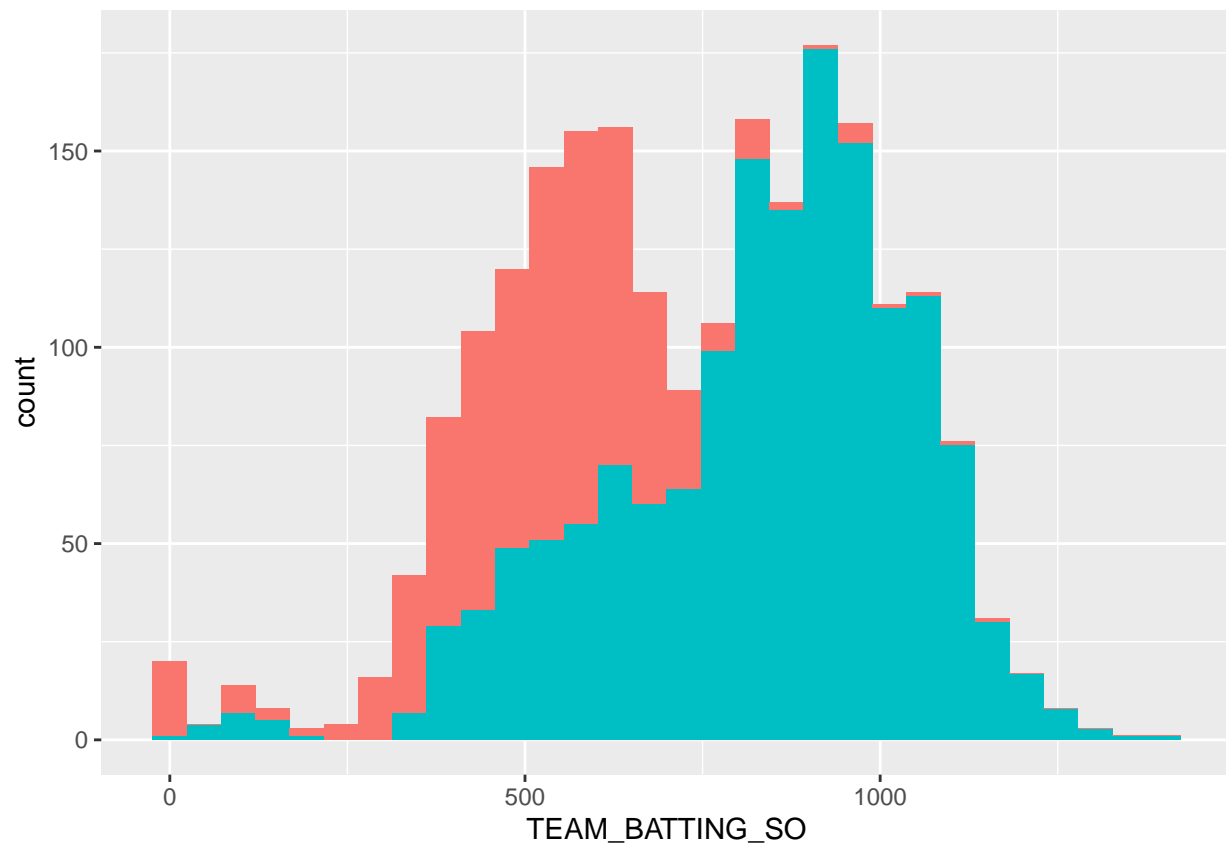
'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



```
ggplot(baseball_hbp_dummy, aes(x = TEAM_BATTING_SO, fill = TEAM_CS_YES_NO)) +  
  geom_histogram() +  
  theme(legend.position = 'none')
```

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

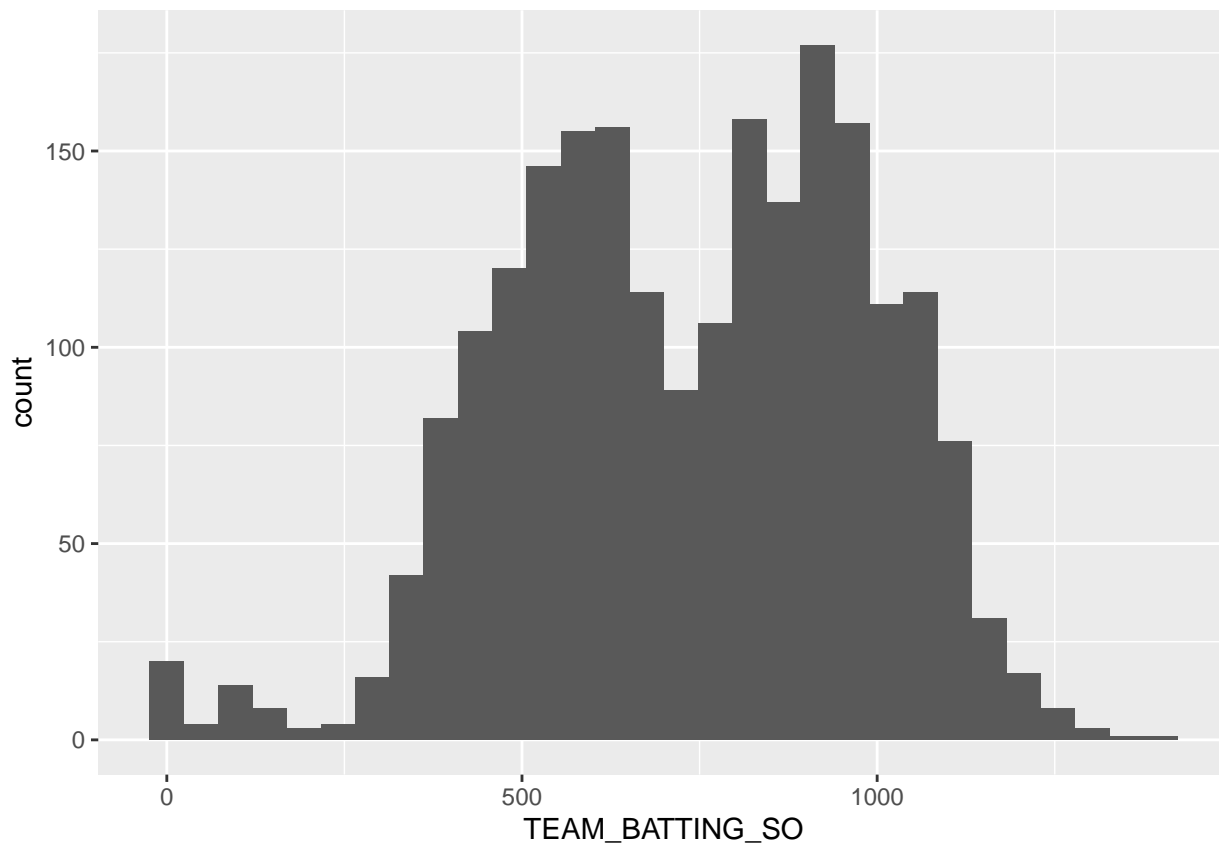
```
## Warning: Removed 102 rows containing non-finite values (stat_bin).
```

```
ggplot(baseball_df, aes(x = TEAM_BATTING_SO)) +  
  geom_histogram()
```

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

```
## Warning: Removed 102 rows containing non-finite values (stat_bin).
```



As these three histograms illustrate, the bimodal distributions are explained by missing CS values or not. Missing values explain both modes present in the overall histogram.

```
# Devin - start- baseball_df_fix not defined yet - did you want to move to after? I added the code here
baseball_df_fix <- baseball_df %>%
  mutate(TEAM_CS_YES_NO = case_when(!is.na(TEAM_BASERUN_CS) ~ 1, is.na(TEAM_BASERUN_CS) ~ 0)) %>%
  mutate(TEAM_HBP_YES_NO = case_when(!is.na(TEAM_BATTING_HBP) ~ 1, is.na(TEAM_BATTING_HBP) ~ 0)) %>%
  dplyr::select(-c(TEAM_BATTING_HBP, INDEX, TEAM_BASERUN_CS))
# Devin - end- baseball_df_fix not defined yet - did you want to move to after? I added the code here j

#Devin - still erroring out?
#image(is.na(baseball_df_fix),axes=FALSE,col=gray(1:0))
#axis(2, at=0:17/17, labels=colnames(baseball_df_fix))
#axis(1, at=0:2275/2275, labels=row.names(baseball_df_fix),las=2)
```

To better visualize the missing values, it looks like two of them overlap perfectly.

```
baseball_df_fix[rowSums(is.na(baseball_df_fix)) > 0,] %>%
  dplyr::select(TEAM_PITCHING_SO, TEAM_FIELDING_DP, TEAM_BATTING_SO, TEAM_BASERUN_SB)
```

##	TEAM_PITCHING_SO	TEAM_FIELDING_DP	TEAM_BATTING_SO	TEAM_BASERUN_SB
## 1	5456	NA	842	NA
## 53	272	88	99	NA
## 54	525	97	227	NA
## 55	883	97	327	NA
## 56	825	112	428	NA

## 57	822	104	426	NA
## 58	908	71	471	NA
## 59	1155	NA	699	NA
## 60	1405	NA	963	NA
## 61	1092	NA	755	NA
## 62	1030	NA	744	216
## 63	703	NA	525	499
## 64	765	NA	633	354
## 65	721	NA	570	419
## 66	764	NA	627	347
## 67	742	NA	632	339
## 68	461	NA	367	305
## 69	393	NA	320	296
## 70	361	NA	292	246
## 71	419	NA	339	298
## 72	395	NA	322	286
## 73	358	NA	329	187
## 74	306	NA	287	197
## 77	NA	104	NA	187
## 78	NA	96	NA	151
## 79	NA	94	NA	139
## 80	NA	109	NA	100
## 81	NA	140	NA	129
## 82	NA	95	NA	141
## 83	NA	107	NA	143
## 175	NA	87	NA	240
## 176	NA	103	NA	153
## 177	NA	103	NA	134
## 178	NA	116	NA	147
## 179	NA	119	NA	178
## 180	NA	111	NA	133
## 181	NA	134	NA	185
## 269	552	NA	450	196
## 272	252	NA	84	105
## 273	208	NA	72	61
## 274	1059	91	477	NA
## 275	943	NA	559	NA
## 276	1273	NA	833	NA
## 277	1168	NA	786	NA
## 278	923	NA	746	333
## 279	481	NA	401	654
## 280	669	NA	566	385
## 281	772	NA	643	373
## 282	3450	NA	724	481
## 283	797	NA	664	410
## 284	583	NA	529	250
## 285	502	NA	403	290
## 286	320	NA	253	410
## 287	378	NA	303	386
## 288	316	NA	252	554
## 289	398	NA	319	500
## 290	374	NA	344	272
## 291	459	NA	419	398
## 294	1552	NA	1006	NA

## 295	556	NA	103	394
## 296	310	NA	90	162
## 297	181	NA	66	107
## 298	0	NA	0	NA
## 299	0	NA	0	NA
## 304	NA	80	NA	140
## 305	NA	88	NA	104
## 306	NA	112	NA	136
## 307	NA	75	NA	176
## 308	NA	102	NA	231
## 309	NA	85	NA	205
## 310	NA	98	NA	201
## 311	NA	94	NA	197
## 391	460	NA	105	74
## 392	387	NA	129	54
## 393	0	NA	0	NA
## 394	1739	NA	1170	NA
## 395	1221	NA	746	298
## 396	686	NA	546	517
## 399	1354	129	652	NA
## 400	1276	109	646	NA
## 401	1029	94	527	NA
## 402	849	81	440	NA
## 403	954	NA	571	NA
## 404	975	NA	668	NA
## 405	796	NA	550	NA
## 406	556	NA	453	334
## 407	1144	NA	685	NA
## 408	909	NA	606	NA
## 409	836	83	423	NA
## 410	881	100	457	NA
## 411	833	131	432	NA
## 412	1019	140	516	NA
## 413	1044	NA	625	NA
## 414	1205	NA	833	NA
## 415	0	NA	0	NA
## 416	782	NA	140	101
## 417	707	NA	96	88
## 418	270	81	110	NA
## 419	837	118	305	NA
## 420	1145	100	424	NA
## 421	1237	107	603	NA
## 422	808	79	419	NA
## 423	833	104	432	NA
## 424	974	104	505	NA
## 425	1091	NA	660	NA
## 426	981	NA	678	NA
## 427	898	NA	621	NA
## 428	875	NA	670	278
## 429	718	NA	536	511
## 430	811	NA	676	344
## 431	777	NA	633	298
## 432	658	NA	548	286
## 433	594	NA	535	259

## 434	449	NA	366	401
## 435	464	NA	367	420
## 436	477	NA	389	324
## 437	460	NA	426	238
## 438	409	NA	374	270
## 442	NA	92	NA	304
## 443	NA	94	NA	240
## 444	NA	105	NA	283
## 445	NA	107	NA	302
## 446	NA	117	NA	250
## 447	NA	80	NA	223
## 448	NA	101	NA	198
## 539	574	NA	485	326
## 542	NA	68	NA	207
## 543	NA	86	NA	230
## 544	NA	108	NA	188
## 545	NA	87	NA	223
## 546	NA	108	NA	225
## 547	NA	105	NA	216
## 637	836	83	413	NA
## 638	712	NA	431	NA
## 639	892	NA	600	NA
## 640	879	NA	608	NA
## 641	872	NA	743	217
## 642	527	NA	439	632
## 643	811	NA	671	567
## 644	695	NA	596	538
## 645	568	NA	463	383
## 646	614	NA	519	289
## 647	553	NA	512	292
## 648	410	NA	324	301
## 649	391	NA	314	268
## 650	386	NA	310	406
## 651	329	NA	268	238
## 652	341	NA	320	176
## 653	345	NA	319	246
## 655	NA	98	NA	168
## 656	NA	86	NA	190
## 657	NA	129	NA	192
## 658	NA	104	NA	182
## 659	NA	125	NA	167
## 660	NA	76	NA	206
## 749	1011	NA	624	266
## 754	NA	88	NA	198
## 755	NA	118	NA	215
## 756	NA	146	NA	206
## 757	NA	116	NA	184
## 758	NA	114	NA	221
## 759	NA	131	NA	205
## 844	536	NA	430	224
## 845	839	NA	715	357
## 846	814	NA	673	427
## 847	741	NA	627	331
## 848	709	NA	573	439

## 849	842	NA	686	433
## 850	714	NA	582	187
## 851	631	NA	541	282
## 852	634	NA	583	245
## 853	367	NA	290	319
## 854	475	NA	378	276
## 855	561	NA	450	233
## 856	506	NA	400	221
## 857	526	NA	425	224
## 858	362	NA	333	101
## 859	310	NA	295	134
## 860	0	NA	0	NA
## 861	0	NA	0	NA
## 862	845	112	339	NA
## 863	890	94	313	NA
## 864	1029	100	381	NA
## 865	848	97	419	NA
## 866	1096	99	541	NA
## 881	1148	NA	744	NA
## 882	1600	NA	899	NA
## 886	NA	85	NA	137
## 887	NA	94	NA	224
## 888	NA	101	NA	175
## 889	NA	82	NA	290
## 890	NA	96	NA	286
## 976	1173	86	601	NA
## 977	1033	NA	625	NA
## 978	1016	NA	677	NA
## 979	739	NA	561	256
## 980	440	NA	337	349
## 981	606	NA	490	239
## 982	905	NA	162	45
## 996	443	64	186	NA
## 997	757	89	271	NA
## 998	0	NA	0	NA
## 999	0	NA	0	NA
## 1044	1436	100	532	NA
## 1045	1284	NA	848	NA
## 1046	626	NA	487	429
## 1047	708	NA	590	420
## 1048	653	NA	540	305
## 1049	909	NA	741	315
## 1050	875	NA	740	558
## 1082	2225	NA	1085	NA
## 1083	657	NA	77	136
## 1084	1013	NA	650	NA
## 1085	678	NA	469	NA
## 1086	731	NA	618	293
## 1087	533	NA	441	494
## 1088	560	NA	474	460
## 1089	569	NA	453	438
## 1090	608	NA	514	398
## 1091	560	NA	532	430
## 1092	475	NA	375	270

## 1093	450	NA	364	349
## 1094	486	NA	393	226
## 1095	412	NA	333	245
## 1096	384	NA	313	230
## 1097	392	NA	351	145
## 1101	NA	117	NA	325
## 1102	NA	108	NA	198
## 1103	NA	78	NA	187
## 1104	NA	91	NA	149
## 1191	592	108	241	NA
## 1192	1021	100	378	NA
## 1193	850	NA	509	NA
## 1194	918	NA	612	NA
## 1195	937	NA	648	NA
## 1196	792	NA	665	241
## 1197	505	NA	424	555
## 1198	870	NA	725	382
## 1199	718	NA	612	238
## 1200	693	NA	565	419
## 1201	661	NA	559	272
## 1202	577	NA	541	293
## 1203	515	NA	397	263
## 1204	566	NA	454	270
## 1205	493	NA	399	193
## 1206	653	NA	528	241
## 1207	704	NA	565	243
## 1208	494	NA	460	252
## 1209	426	NA	400	248
## 1210	547	NA	81	34
## 1211	0	NA	0	0
## 1249	NA	82	NA	181
## 1250	NA	84	NA	252
## 1251	NA	74	NA	239
## 1252	NA	107	NA	145
## 1253	NA	106	NA	206
## 1254	NA	95	NA	226
## 1255	NA	98	NA	292
## 1340	2309	NA	513	212
## 1341	1561	86	578	NA
## 1342	12758	NA	945	NA
## 1345	0	NA	0	NA
## 1346	363	NA	74	226
## 1347	468	NA	156	168
## 1348	205	NA	67	46
## 1349	0	NA	0	NA
## 1350	0	NA	0	NA
## 1351	556	NA	450	286
## 1393	737	NA	437	NA
## 1394	722	NA	477	NA
## 1395	833	NA	694	144
## 1396	687	NA	564	372
## 1397	292	52	101	NA
## 1401	NA	96	NA	217
## 1402	NA	74	NA	206

## 1403	NA	86	NA	225
## 1404	NA	82	NA	243
## 1405	NA	102	NA	309
## 1406	NA	106	NA	287
## 1407	NA	82	NA	263
## 1496	NA	104	NA	208
## 1497	NA	96	NA	184
## 1498	NA	75	NA	153
## 1499	NA	72	NA	123
## 1500	NA	97	NA	217
## 1501	NA	126	NA	224
## 1502	NA	107	NA	242
## 1584	1296	NA	72	0
## 1585	601	NA	475	315
## 1588	765	78	354	NA
## 1589	732	NA	443	NA
## 1590	974	NA	643	NA
## 1591	858	NA	593	NA
## 1592	1003	NA	836	341
## 1593	702	NA	576	529
## 1594	814	NA	663	374
## 1595	1072	NA	741	NA
## 1596	1042	NA	733	321
## 1597	601	NA	456	468
## 1598	753	NA	604	307
## 1599	574	NA	450	343
## 1600	616	NA	498	414
## 1601	576	NA	487	274
## 1602	577	NA	545	229
## 1603	529	NA	421	254
## 1604	392	NA	310	346
## 1605	461	NA	370	238
## 1606	450	NA	367	200
## 1607	451	NA	415	198
## 1608	387	NA	363	226
## 1612	NA	91	NA	144
## 1613	NA	88	NA	191
## 1614	NA	115	NA	170
## 1615	NA	79	NA	210
## 1616	NA	103	NA	196
## 1698	2367	NA	979	NA
## 1699	491	NA	397	251
## 1700	745	NA	639	174
## 1701	789	83	380	NA
## 1702	942	NA	570	NA
## 1703	926	NA	617	NA
## 1704	1144	NA	784	NA
## 1705	997	NA	843	307
## 1706	651	NA	498	289
## 1707	852	NA	705	347
## 1708	650	NA	546	248
## 1709	508	NA	480	235
## 1710	431	NA	343	264
## 1711	450	NA	367	315

## 1712	451	NA	359	273
## 1713	511	NA	413	210
## 1714	410	NA	375	117
## 1718	NA	116	NA	199
## 1719	NA	98	NA	188
## 1720	NA	119	NA	214
## 1721	NA	115	NA	172
## 1722	NA	79	NA	278
## 1723	NA	78	NA	196
## 1724	NA	107	NA	197
## 1810	770	NA	133	324
## 1811	637	NA	173	214
## 1812	0	NA	0	NA
## 1813	0	NA	0	NA
## 1814	1590	113	589	NA
## 1815	640	79	332	NA
## 1816	692	102	359	NA
## 1817	797	127	413	NA
## 1818	949	129	492	NA
## 1819	845	NA	511	NA
## 1820	981	NA	678	NA
## 1821	932	NA	633	NA
## 1822	364	NA	119	134
## 1823	0	NA	0	NA
## 1824	0	NA	0	NA
## 1825	1092	NA	155	14
## 1826	4224	NA	1095	NA
## 1827	890	NA	692	399
## 1828	1257	NA	194	343
## 1829	402	84	159	NA
## 1830	1072	78	397	NA
## 1895	845	NA	501	NA
## 1896	652	NA	451	NA
## 1897	760	NA	558	211
## 1898	565	NA	429	547
## 1899	697	NA	564	388
## 1900	638	NA	496	375
## 1901	732	NA	592	357
## 1902	594	NA	484	275
## 1903	540	NA	503	323
## 1904	420	NA	342	367
## 1905	326	NA	266	392
## 1906	446	NA	361	361
## 1907	414	NA	335	339
## 1908	500	NA	404	406
## 1909	434	NA	402	231
## 1910	420	NA	389	253
## 1914	NA	101	NA	308
## 1915	NA	98	NA	300
## 1916	NA	98	NA	308
## 1917	NA	90	NA	307
## 1918	NA	83	NA	190
## 1919	NA	105	NA	248
## 2012	1114	NA	777	NA

## 2013	927	NA	618	NA
## 2014	1157	NA	871	207
## 2015	0	NA	0	NA
## 2016	0	NA	0	NA
## 2017	927	83	458	NA
## 2018	656	NA	397	NA
## 2019	777	NA	513	NA
## 2020	590	NA	408	NA
## 2021	577	NA	495	392
## 2022	490	NA	408	697
## 2023	750	NA	625	562
## 2024	686	NA	572	403
## 2025	696	NA	584	366
## 2026	624	NA	524	337
## 2027	572	NA	530	226
## 2028	378	NA	308	307
## 2029	427	NA	345	254
## 2030	466	NA	374	231
## 2031	480	NA	388	213
## 2032	469	NA	434	112
## 2033	301	NA	281	225
## 2037	NA	131	NA	202
## 2038	NA	87	NA	209
## 2039	NA	87	NA	170
## 2040	NA	99	NA	119
## 2041	NA	111	NA	132
## 2042	NA	72	NA	158
## 2043	NA	96	NA	172
## 2136	19278	NA	952	NA
## 2137	1275	86	551	NA
## 2138	784	NA	615	372
## 2190	841	NA	685	517
## 2191	1313	NA	843	NA
## 2219	636	NA	110	359
## 2220	590	NA	91	58
## 2221	990	113	507	NA
## 2222	893	135	463	NA
## 2223	1136	137	582	NA
## 2224	698	NA	582	263
## 2225	636	NA	593	296
## 2226	374	NA	298	193
## 2227	565	NA	460	306
## 2228	634	NA	501	300
## 2229	524	NA	427	255
## 2230	438	NA	411	210
## 2231	387	NA	363	188
## 2232	569	NA	137	21
## 2233	0	NA	0	NA
## 2234	1167	NA	807	NA
## 2235	1061	NA	786	193
## 2236	598	NA	450	444
## 2237	729	NA	603	400
## 2238	779	NA	596	303
## 2239	0	NA	0	NA

```
## 2240          1060          96          543          NA
## 2241           660          99          334          NA
## 2242          1126         127          584          NA
## 2276          2492          NA          969          NA
```

```
summary( lm(baseball_df_fix, formula = TEAM_BATTING_SO ~.-TARGET_WINS))
```

```
##
## Call:
## lm(formula = TEAM_BATTING_SO ~ . - TARGET_WINS, data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -112.307   -3.515   -0.733    2.989   100.385
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   20.011705   6.396987   3.128  0.00179 **
## TEAM_BATTING_H    0.287453   0.016182  17.763 < 2e-16 ***
## TEAM_BATTING_2B    0.004292   0.009945   0.432  0.66608
## TEAM_BATTING_3B   -0.024591   0.020178  -1.219  0.22312
## TEAM_BATTING_HR    2.173703   0.071019  30.607 < 2e-16 ***
## TEAM_BATTING_BB    0.264486   0.044878   5.893 4.50e-09 ***
## TEAM_BASERUN_SB    0.017420   0.005876   2.964  0.00307 **
## TEAM_PITCHING_H   -0.281532   0.014499 -19.417 < 2e-16 ***
## TEAM_PITCHING_HR  -2.072216   0.068079 -30.438 < 2e-16 ***
## TEAM_PITCHING_BB  -0.253951   0.042640  -5.956 3.10e-09 ***
## TEAM_PITCHING_SO   0.945191   0.002419 390.665 < 2e-16 ***
## TEAM_FIELDING_E    0.007461   0.007813   0.955  0.33973
## TEAM_FIELDING_DP  -0.011230   0.013192  -0.851  0.39475
## TEAM_CS_YES_NO    -0.242665   0.854894  -0.284  0.77655
## TEAM_HBP_YES_NO    4.013018   1.027459   3.906 9.74e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.81 on 1820 degrees of freedom
## (441 observations deleted due to missingness)
## Multiple R-squared:  0.9975, Adjusted R-squared:  0.9975
## F-statistic: 5.264e+04 on 14 and 1820 DF,  p-value: < 2.2e-16
```

```
summary( lm(baseball_df_fix, formula = TEAM_PITCHING_SO ~.-TARGET_WINS))
```

```
##
## Call:
## lm(formula = TEAM_PITCHING_SO ~ . - TARGET_WINS, data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -109.203   -2.921    0.575    3.335   121.537
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.602953   6.745982   0.089  0.92879
```

```
## TEAM_BATTING_H -0.316956 0.016873 -18.785 < 2e-16 ***
## TEAM_BATTING_2B 0.005710 0.010459 0.546 0.58521
## TEAM_BATTING_3B 0.011791 0.021229 0.555 0.57866
## TEAM_BATTING_HR -2.299366 0.074470 -30.876 < 2e-16 ***
## TEAM_BATTING_BB -0.269092 0.047228 -5.698 1.41e-08 ***
## TEAM_BATTING_SO 1.045519 0.002676 390.665 < 2e-16 ***
## TEAM_BASERUN_SB -0.007573 0.006193 -1.223 0.22149
## TEAM_PITCHING_H 0.299708 0.015210 19.704 < 2e-16 ***
## TEAM_PITCHING_HR 2.218126 0.070945 31.265 < 2e-16 ***
## TEAM_PITCHING_BB 0.255925 0.044882 5.702 1.38e-08 ***
## TEAM_FIELDING_E -0.011783 0.008215 -1.434 0.15163
## TEAM_FIELDING_DP 0.005163 0.013877 0.372 0.70989
## TEAM_CS_YES_NO 1.134114 0.898748 1.262 0.20715
## TEAM_HBP_YES_NO -3.263506 1.082434 -3.015 0.00261 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.37 on 1820 degrees of freedom
## (441 observations deleted due to missingness)
## Multiple R-squared: 0.9974, Adjusted R-squared: 0.9974
## F-statistic: 4.991e+04 on 14 and 1820 DF, p-value: < 2.2e-16
```

It appears that TEAM_PITCHING_SO and TEAM_BATTING_SO are missing all of the same rows. By quickly running a linear model for either column shows that it's possible to approximate values from other season records.

```
baseball_df %>%
  dplyr::filter(TEAM_PITCHING_SO < 5)
```

```
##      INDEX TARGET_WINS TEAM_BATTING_H TEAM_BATTING_2B TEAM_BATTING_3B
## 1      325          120          2270          301          132
## 2      326          146          2305          322          111
## 3      435           65          1464          147           32
## 4      459           23          1458          220           35
## 5      952           77          1895          244           8
## 6      953           73          1685          206           31
## 7     1106           49          1794          281           58
## 8     1107          107          1725          194           67
## 9     1347           0           891          135           0
## 10    1498           24          1289          145           41
## 11    1502          105          1767          249           77
## 12    1503           71          1491          200           57
## 13    2037           97          1903          256           50
## 14    2038          118          2086          280          135
## 15    2048           81          1927          207          142
## 16    2049           88          1622          155           67
## 17    2253           34          1177          171           9
## 18    2254           93          1527          200           64
## 19    2486           12          1009          112           75
## 20    2493           29          1122           69           64
##      TEAM_BATTING_HR TEAM_BATTING_BB TEAM_BATTING_SO TEAM_BASERUN_SB
## 1              42          74          0          NA
## 2              29          64          0          NA
```

## 3	3	94	0	NA
## 4	0	93	0	NA
## 5	8	93	0	NA
## 6	0	58	0	NA
## 7	6	79	0	NA
## 8	4	79	0	NA
## 9	0	0	0	0
## 10	7	45	0	NA
## 11	20	95	0	NA
## 12	17	50	0	NA
## 13	18	71	0	NA
## 14	22	89	0	NA
## 15	8	78	0	NA
## 16	12	52	0	NA
## 17	0	119	0	NA
## 18	0	79	0	NA
## 19	0	12	0	NA
## 20	0	29	0	NA
##	TEAM_BASERUN_CS	TEAM_BATTING_HBP	TEAM_PITCHING_H	TEAM_PITCHING_HR
## 1	NA	NA	5253	97
## 2	NA	NA	4727	59
## 3	NA	NA	4312	9
## 4	NA	NA	16871	0
## 5	NA	NA	5203	22
## 6	NA	NA	4074	0
## 7	NA	NA	5484	18
## 8	NA	NA	3408	8
## 9	0	NA	24057	0
## 10	NA	NA	4443	24
## 11	NA	NA	4404	50
## 12	NA	NA	3552	41
## 13	NA	NA	5605	53
## 14	NA	NA	4629	49
## 15	NA	NA	5382	22
## 16	NA	NA	3864	29
## 17	NA	NA	10035	0
## 18	NA	NA	3638	0
## 19	NA	NA	12574	0
## 20	NA	NA	6492	0
##	TEAM_PITCHING_BB	TEAM_PITCHING_SO	TEAM_FIELDING_E	TEAM_FIELDING_DP
## 1	171	0	1058	NA
## 2	131	0	951	NA
## 3	277	0	1473	NA
## 4	1076	0	1898	NA
## 5	255	0	1225	NA
## 6	140	0	931	NA
## 7	241	0	1531	NA
## 8	156	0	853	NA
## 9	0	0	1890	NA
## 10	155	0	1506	NA
## 11	237	0	1092	NA
## 12	119	0	1253	NA
## 13	209	0	1166	NA
## 14	198	0	928	NA

## 15	218	0	1447	NA
## 16	124	0	1132	NA
## 17	1015	0	1279	NA
## 18	188	0	1010	NA
## 19	150	0	847	NA
## 20	168	0	1522	NA

```
baseball_df%>%
  dplyr::filter(TEAM_BATTING_SO < 5)
```

##	INDEX	TARGET_WINS	TEAM_BATTING_H	TEAM_BATTING_2B	TEAM_BATTING_3B
## 1	325	120	2270	301	132
## 2	326	146	2305	322	111
## 3	435	65	1464	147	32
## 4	459	23	1458	220	35
## 5	952	77	1895	244	8
## 6	953	73	1685	206	31
## 7	1106	49	1794	281	58
## 8	1107	107	1725	194	67
## 9	1347	0	891	135	0
## 10	1498	24	1289	145	41
## 11	1502	105	1767	249	77
## 12	1503	71	1491	200	57
## 13	2037	97	1903	256	50
## 14	2038	118	2086	280	135
## 15	2048	81	1927	207	142
## 16	2049	88	1622	155	67
## 17	2253	34	1177	171	9
## 18	2254	93	1527	200	64
## 19	2486	12	1009	112	75
## 20	2493	29	1122	69	64
##	TEAM_BATTING_HR	TEAM_BATTING_BB	TEAM_BATTING_SO	TEAM_BASERUN_SB	
## 1	42	74	0	NA	
## 2	29	64	0	NA	
## 3	3	94	0	NA	
## 4	0	93	0	NA	
## 5	8	93	0	NA	
## 6	0	58	0	NA	
## 7	6	79	0	NA	
## 8	4	79	0	NA	
## 9	0	0	0	0	
## 10	7	45	0	NA	
## 11	20	95	0	NA	
## 12	17	50	0	NA	
## 13	18	71	0	NA	
## 14	22	89	0	NA	
## 15	8	78	0	NA	
## 16	12	52	0	NA	
## 17	0	119	0	NA	
## 18	0	79	0	NA	
## 19	0	12	0	NA	
## 20	0	29	0	NA	
##	TEAM_BASERUN_CS	TEAM_BATTING_HBP	TEAM_PITCHING_H	TEAM_PITCHING_HR	
## 1	NA	NA	5253	97	

## 2	NA	NA	4727	59
## 3	NA	NA	4312	9
## 4	NA	NA	16871	0
## 5	NA	NA	5203	22
## 6	NA	NA	4074	0
## 7	NA	NA	5484	18
## 8	NA	NA	3408	8
## 9	0	NA	24057	0
## 10	NA	NA	4443	24
## 11	NA	NA	4404	50
## 12	NA	NA	3552	41
## 13	NA	NA	5605	53
## 14	NA	NA	4629	49
## 15	NA	NA	5382	22
## 16	NA	NA	3864	29
## 17	NA	NA	10035	0
## 18	NA	NA	3638	0
## 19	NA	NA	12574	0
## 20	NA	NA	6492	0
##	TEAM_PITCHING_BB	TEAM_PITCHING_SO	TEAM_FIELDING_E	TEAM_FIELDING_DP
## 1	171	0	1058	NA
## 2	131	0	951	NA
## 3	277	0	1473	NA
## 4	1076	0	1898	NA
## 5	255	0	1225	NA
## 6	140	0	931	NA
## 7	241	0	1531	NA
## 8	156	0	853	NA
## 9	0	0	1890	NA
## 10	155	0	1506	NA
## 11	237	0	1092	NA
## 12	119	0	1253	NA
## 13	209	0	1166	NA
## 14	198	0	928	NA
## 15	218	0	1447	NA
## 16	124	0	1132	NA
## 17	1015	0	1279	NA
## 18	188	0	1010	NA
## 19	150	0	847	NA
## 20	168	0	1522	NA

Lookin closer at these two variables, there are also some values that may be omitted if they are implausibly small. Going a whole season with zero strikeouts, pitching or batting, seems unlikely. It may make sense to recode these as NA and impute values onto them.

```
baseball_df %>%
  dplyr::select(TEAM_BASERUN_CS, TEAM_BASERUN_SB) %>%
  filter(is.na(TEAM_BASERUN_SB))
```

##	TEAM_BASERUN_CS	TEAM_BASERUN_SB
## 1	NA	NA
## 2	NA	NA
## 3	NA	NA

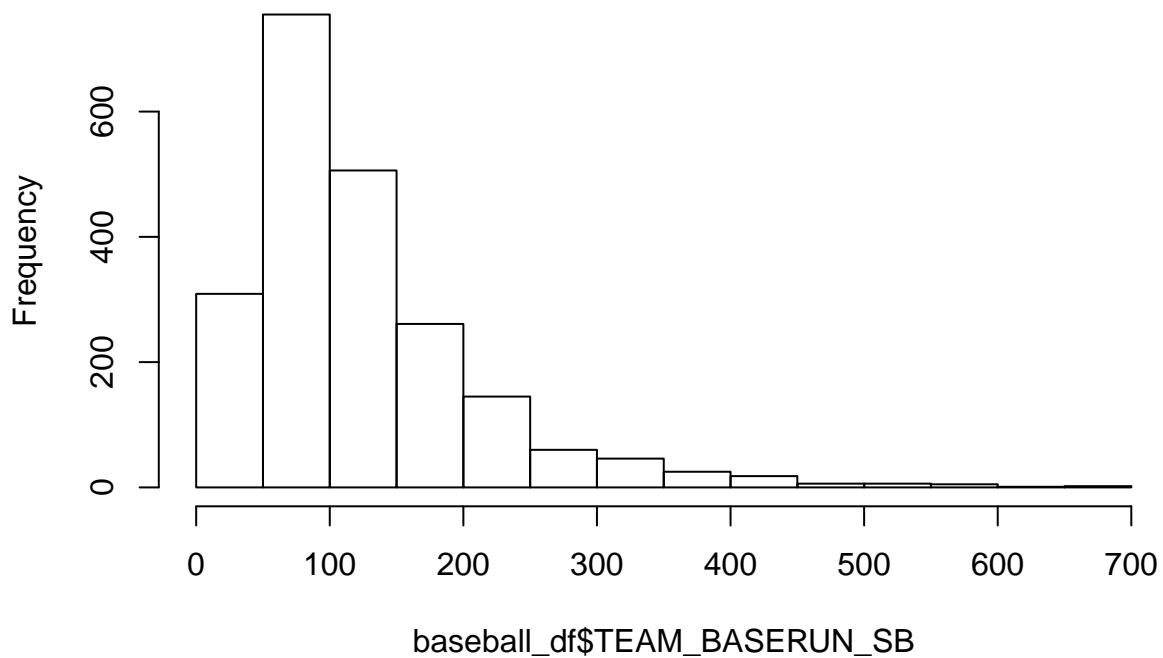
## 4	NA	NA
## 5	NA	NA
## 6	NA	NA
## 7	NA	NA
## 8	NA	NA
## 9	NA	NA
## 10	NA	NA
## 11	NA	NA
## 12	NA	NA
## 13	NA	NA
## 14	NA	NA
## 15	NA	NA
## 16	NA	NA
## 17	NA	NA
## 18	NA	NA
## 19	NA	NA
## 20	NA	NA
## 21	NA	NA
## 22	NA	NA
## 23	NA	NA
## 24	NA	NA
## 25	NA	NA
## 26	NA	NA
## 27	NA	NA
## 28	NA	NA
## 29	NA	NA
## 30	NA	NA
## 31	NA	NA
## 32	NA	NA
## 33	NA	NA
## 34	NA	NA
## 35	NA	NA
## 36	NA	NA
## 37	NA	NA
## 38	NA	NA
## 39	NA	NA
## 40	NA	NA
## 41	NA	NA
## 42	NA	NA
## 43	NA	NA
## 44	NA	NA
## 45	NA	NA
## 46	NA	NA
## 47	NA	NA
## 48	NA	NA
## 49	NA	NA
## 50	NA	NA
## 51	NA	NA
## 52	NA	NA
## 53	NA	NA
## 54	NA	NA
## 55	NA	NA
## 56	NA	NA
## 57	NA	NA

## 58	NA	NA
## 59	NA	NA
## 60	NA	NA
## 61	NA	NA
## 62	NA	NA
## 63	NA	NA
## 64	NA	NA
## 65	NA	NA
## 66	NA	NA
## 67	NA	NA
## 68	NA	NA
## 69	NA	NA
## 70	NA	NA
## 71	NA	NA
## 72	NA	NA
## 73	NA	NA
## 74	NA	NA
## 75	NA	NA
## 76	NA	NA
## 77	NA	NA
## 78	NA	NA
## 79	NA	NA
## 80	NA	NA
## 81	NA	NA
## 82	NA	NA
## 83	NA	NA
## 84	NA	NA
## 85	NA	NA
## 86	NA	NA
## 87	NA	NA
## 88	NA	NA
## 89	NA	NA
## 90	NA	NA
## 91	NA	NA
## 92	NA	NA
## 93	NA	NA
## 94	NA	NA
## 95	NA	NA
## 96	NA	NA
## 97	NA	NA
## 98	NA	NA
## 99	NA	NA
## 100	NA	NA
## 101	NA	NA
## 102	NA	NA
## 103	NA	NA
## 104	NA	NA
## 105	NA	NA
## 106	NA	NA
## 107	NA	NA
## 108	NA	NA
## 109	NA	NA
## 110	NA	NA
## 111	NA	NA

```
## 112      NA      NA
## 113      NA      NA
## 114      NA      NA
## 115      NA      NA
## 116      NA      NA
## 117      NA      NA
## 118      NA      NA
## 119      NA      NA
## 120      NA      NA
## 121      NA      NA
## 122      NA      NA
## 123      NA      NA
## 124      NA      NA
## 125      NA      NA
## 126      NA      NA
## 127      NA      NA
## 128      NA      NA
## 129      NA      NA
## 130      NA      NA
## 131      NA      NA
```

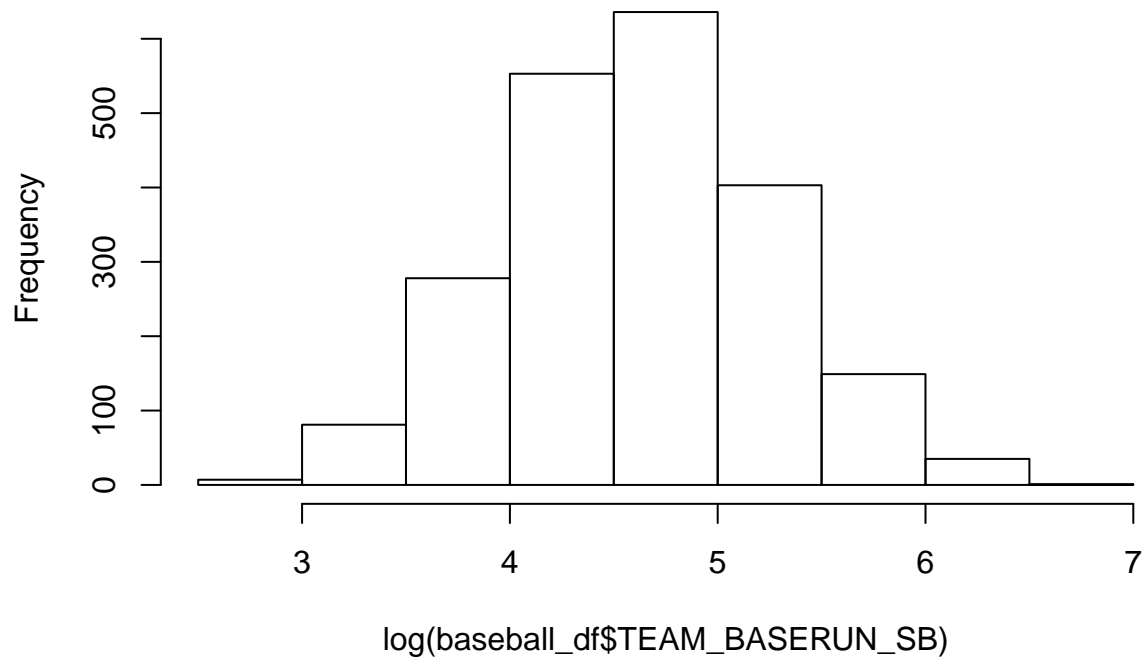
```
hist(baseball_df$TEAM_BASERUN_SB)
```

Histogram of baseball_df\$TEAM_BASERUN_SB



```
hist(log(baseball_df$TEAM_BASERUN_SB))
```

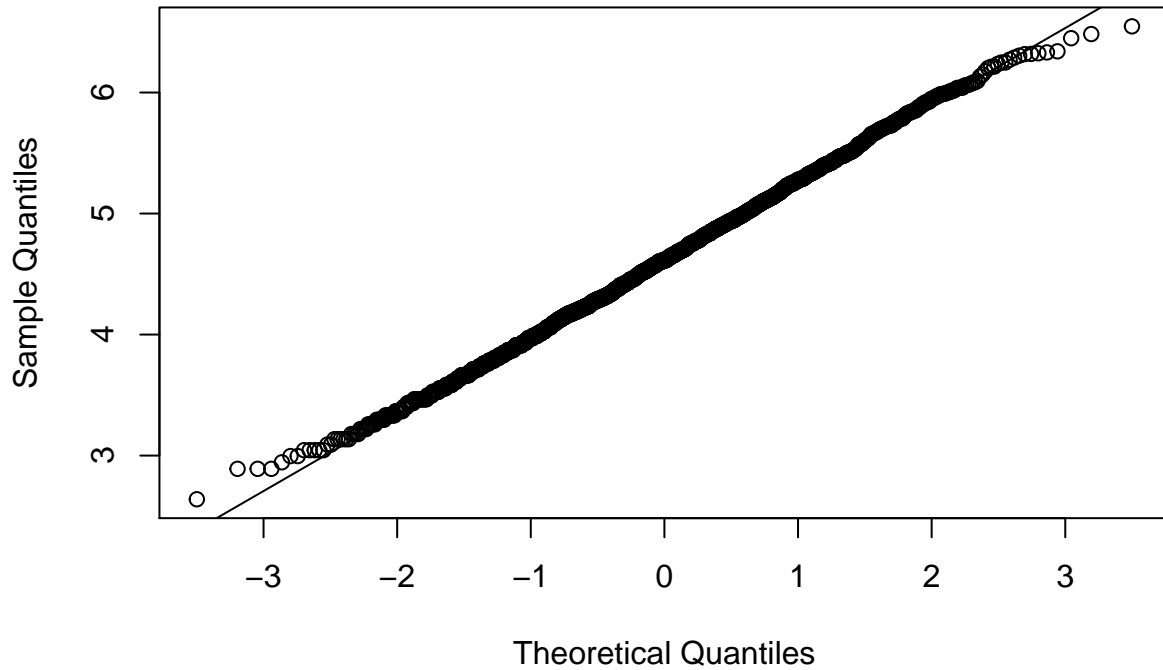
Histogram of $\log(\text{baseball_df\$TEAM_BASERUN_SB})$



```
baseball_log_sb <- baseball_df %>%  
  filter(!is.na(Team_Baserun_SB)) %>%  
  filter(Team_Baserun_SB != 0) %>%  
  mutate(LOG_Baserun_SB = log(Team_Baserun_SB))
```

```
qqnorm((baseball_log_sb$LOG_Baserun_SB))  
qqline((baseball_log_sb$LOG_Baserun_SB))
```

Normal Q-Q Plot



The column `TEAM_BASERUN_SB` is partly correlated with `TEAM_BASERUN_CS` in the training set. However, there are many missing values so single imputation may not be an option. In this case, `TEAM_BASERUN_SB` may qualify for multiple imputation after log transform to make it normally distributed.

```
baseball_df_fix %>%
  filter(is.na(Team_Fielding_DP))
```

##	TARGET_WINS	TEAM_BATTING_H	TEAM_BATTING_2B	TEAM_BATTING_3B	TEAM_BATTING_HR
## 1	39	1445	194	39	13
## 2	104	1670	345	142	56
## 3	107	1551	261	88	53
## 4	67	1323	208	77	32
## 5	78	1502	209	82	33
## 6	82	1680	248	126	71
## 7	85	1430	202	108	68
## 8	105	1583	248	68	53
## 9	93	1486	213	76	38
## 10	102	1484	212	94	63
## 11	108	1705	224	63	82
## 12	102	2035	334	115	126
## 13	88	1693	244	70	67
## 14	92	1751	216	92	45
## 15	114	1932	282	102	55
## 16	111	1665	207	60	58
## 17	101	1617	190	96	42
## 18	44	1533	221	79	25
## 19	105	2241	282	105	42
## 20	98	2343	344	113	35

## 21	47	1468	211	83	8
## 22	96	1369	203	128	49
## 23	61	1244	184	88	25
## 24	59	1169	153	63	10
## 25	92	1604	242	120	37
## 26	67	1263	192	83	22
## 27	84	1451	186	82	24
## 28	71	1325	162	76	10
## 29	85	1460	170	119	36
## 30	51	1479	176	122	33
## 31	75	1596	204	107	34
## 32	113	2084	343	190	42
## 33	108	1907	293	111	31
## 34	113	1944	260	126	29
## 35	112	1974	303	82	24
## 36	104	1722	167	84	13
## 37	94	1652	223	78	19
## 38	89	1469	231	40	26
## 39	108	2300	378	200	16
## 40	134	2333	393	107	24
## 41	118	2554	376	126	36
## 42	120	2270	301	132	42
## 43	146	2305	322	111	29
## 44	39	1620	201	44	0
## 45	51	1764	180	81	18
## 46	65	1464	147	32	3
## 47	86	1379	250	48	28
## 48	43	1258	190	77	21
## 49	102	1640	280	95	68
## 50	87	1767	307	99	13
## 51	93	1604	238	101	57
## 52	55	1418	216	72	33
## 53	93	1662	228	114	42
## 54	53	1426	169	132	25
## 55	104	1352	161	144	60
## 56	92	1423	307	63	13
## 57	51	1351	213	71	23
## 58	23	1458	220	35	0
## 59	56	1832	196	223	39
## 60	44	2003	272	44	0
## 61	98	1653	458	101	21
## 62	90	1701	234	72	205
## 63	126	1561	266	108	78
## 64	118	1598	259	114	69
## 65	95	1576	238	131	107
## 66	92	1441	176	114	92
## 67	82	1564	226	81	97
## 68	98	1480	191	106	72
## 69	78	1318	165	102	29
## 70	70	1908	325	106	80
## 71	90	1659	230	123	43
## 72	72	1664	232	119	47
## 73	92	1545	189	91	19
## 74	82	1563	189	90	30

## 75	89	1550	236	112	37
## 76	101	1589	202	122	56
## 77	101	1541	162	143	54
## 78	91	1513	156	111	38
## 79	76	1438	170	112	53
## 80	97	1542	215	122	44
## 81	97	1404	160	99	39
## 82	89	1523	230	112	61
## 83	95	1478	184	147	33
## 84	66	1369	175	106	47
## 85	89	1391	167	81	48
## 86	82	1512	204	82	37
## 87	69	1712	279	83	76
## 88	82	1738	293	131	45
## 89	93	1609	269	85	27
## 90	98	1543	221	108	20
## 91	90	1554	210	113	14
## 92	70	1358	170	94	45
## 93	69	1711	265	117	34
## 94	70	1464	201	112	42
## 95	96	1481	192	93	19
## 96	72	1316	182	72	24
## 97	48	1447	220	95	17
## 98	61	1321	157	72	15
## 99	54	1317	162	72	26
## 100	76	1508	213	103	26
## 101	101	1495	213	104	28
## 102	92	1804	281	124	41
## 103	85	1811	303	113	46
## 104	105	1773	242	83	36
## 105	101	1852	262	91	35
## 106	85	1699	237	109	20
## 107	88	1499	176	61	20
## 108	21	1402	149	53	13
## 109	77	1895	244	8	8
## 110	73	1685	206	31	0
## 111	106	1585	182	97	40
## 112	73	1321	226	46	18
## 113	66	1539	271	79	21
## 114	62	1376	224	99	38
## 115	115	1660	232	107	70
## 116	103	1834	278	165	72
## 117	84	1577	219	89	63
## 118	17	1313	145	34	0
## 119	49	1794	281	58	6
## 120	107	1725	194	67	4
## 121	44	1347	195	94	30
## 122	48	1389	208	90	42
## 123	60	1320	216	40	41
## 124	71	1639	276	42	75
## 125	53	1227	174	75	23
## 126	65	1485	192	90	21
## 127	33	1142	213	31	12
## 128	60	1518	162	68	17

## 129	62	1316	174	73	25
## 130	77	1397	175	94	20
## 131	90	1491	232	95	19
## 132	73	1549	242	99	30
## 133	110	1496	222	93	56
## 134	108	1464	231	94	54
## 135	72	1458	236	82	27
## 136	100	1514	193	110	32
## 137	82	1519	219	105	57
## 138	87	1864	282	161	52
## 139	88	1645	234	95	48
## 140	72	1598	215	108	35
## 141	75	1648	248	88	29
## 142	60	1468	174	74	19
## 143	87	1488	190	107	23
## 144	102	1506	228	104	26
## 145	77	1426	182	120	27
## 146	79	1541	217	105	24
## 147	91	1691	231	117	32
## 148	58	1412	220	80	17
## 149	32	1466	200	88	26
## 150	108	1608	191	80	18
## 151	64	1475	154	82	20
## 152	67	1287	142	65	19
## 153	65	1536	229	95	25
## 154	45	1503	216	110	52
## 155	43	1632	211	90	42
## 156	47	1480	176	99	46
## 157	65	1492	199	87	50
## 158	75	1490	161	76	34
## 159	80	1582	205	72	43
## 160	34	2059	209	54	7
## 161	0	891	135	0	0
## 162	95	1494	261	68	59
## 163	108	1188	338	0	0
## 164	24	1289	145	41	7
## 165	79	1978	211	103	5
## 166	102	2004	258	39	12
## 167	89	1901	205	125	15
## 168	105	1767	249	77	20
## 169	71	1491	200	57	17
## 170	92	1723	252	120	82
## 171	91	1490	187	98	10
## 172	114	1593	235	97	33
## 173	64	1256	130	86	22
## 174	54	1458	235	80	26
## 175	36	1674	216	54	0
## 176	76	1509	213	143	44
## 177	109	1607	246	83	33
## 178	92	1600	253	151	39
## 179	80	1590	244	110	43
## 180	76	1370	230	98	25
## 181	99	1473	223	108	38
## 182	66	1297	222	63	29

## 183	56	1351	216	56	20
## 184	101	1387	206	94	37
## 185	99	1671	281	117	62
## 186	86	1272	188	57	20
## 187	80	1592	274	66	56
## 188	96	1567	272	96	28
## 189	80	1471	213	60	25
## 190	92	1504	238	101	53
## 191	90	1950	309	113	100
## 192	90	2192	319	166	51
## 193	77	1722	292	105	61
## 194	68	1708	261	102	49
## 195	85	1556	259	88	36
## 196	100	1719	257	88	33
## 197	51	1494	261	85	17
## 198	84	1667	231	140	61
## 199	85	1516	212	143	64
## 200	51	1475	198	96	21
## 201	45	1166	158	75	3
## 202	82	1394	180	115	7
## 203	95	1385	220	114	19
## 204	72	1491	239	102	26
## 205	80	1294	181	59	17
## 206	27	1296	191	51	24
## 207	85	1364	151	114	40
## 208	102	1817	221	159	46
## 209	87	1656	233	109	32
## 210	83	1722	212	118	34
## 211	74	1566	173	134	31
## 212	79	1437	153	96	15
## 213	122	2372	382	156	52
## 214	110	2496	284	85	15
## 215	97	1903	256	50	18
## 216	118	2086	280	135	22
## 217	96	1655	312	98	35
## 218	122	1428	221	62	30
## 219	78	1208	168	44	9
## 220	110	1972	254	61	24
## 221	81	1927	207	142	8
## 222	88	1622	155	67	12
## 223	14	1437	148	56	0
## 224	46	1254	154	127	27
## 225	81	1399	168	82	40
## 226	26	1776	285	162	19
## 227	78	1519	235	116	41
## 228	123	1569	217	119	23
## 229	102	1574	238	93	29
## 230	90	1658	220	122	63
## 231	104	1421	161	94	68
## 232	107	1696	267	99	67
## 233	78	1546	257	110	31
## 234	87	1560	232	88	56
## 235	76	1423	186	91	42
## 236	83	1748	223	124	75

## 237	108	1775	242	118	53
## 238	82	1637	236	111	40
## 239	79	1710	197	108	49
## 240	103	1792	232	104	38
## 241	83	1536	205	93	37
## 242	65	1545	174	70	25
## 243	135	1793	371	59	46
## 244	54	1244	182	32	12
## 245	57	1329	243	61	40
## 246	34	1177	171	9	0
## 247	93	1527	200	64	0
## 248	107	1475	195	76	12
## 249	101	1493	229	91	17
## 250	114	1416	191	82	25
## 251	108	1591	240	99	23
## 252	114	1860	313	94	47
## 253	110	1427	179	56	43
## 254	108	1574	253	77	70
## 255	93	1558	212	87	57
## 256	101	1584	201	61	69
## 257	60	1282	149	57	49
## 258	70	1581	187	120	12
## 259	48	1662	192	109	47
## 260	50	1448	167	97	46
## 261	36	1579	184	83	38
## 262	42	1393	161	59	14
## 263	90	1624	185	94	50
## 264	41	992	263	20	0
## 265	70	1477	193	75	18
## 266	83	1414	187	133	29
## 267	72	1338	238	75	12
## 268	75	2222	295	197	35
## 269	97	2132	363	71	32
## 270	53	1420	176	101	23
## 271	62	1336	160	84	40
## 272	50	1582	226	104	29
## 273	55	1616	268	145	72
## 274	54	1663	262	128	70
## 275	75	1689	238	95	44
## 276	54	1517	189	85	38
## 277	58	1523	173	93	50
## 278	33	1695	187	104	12
## 279	12	1009	112	75	0
## 280	68	1347	174	38	6
## 281	38	1156	182	69	31
## 282	61	1380	198	84	62
## 283	58	1141	118	59	36
## 284	54	1444	197	74	33
## 285	29	1122	69	64	0
## 286	31	1116	157	62	15
##	TEAM_BATTING_BB	TEAM_BATTING_SO	TEAM_BASERUN_SB	TEAM_PITCHING_H	
## 1	143	842	NA	9364	
## 2	203	699	NA	2761	
## 3	302	963	NA	2264	

## 4	275	755	NA	1914
## 5	346	744	216	2080
## 6	455	525	499	2249
## 7	341	633	354	1729
## 8	596	570	419	2003
## 9	646	627	347	1810
## 10	625	632	339	1742
## 11	705	367	305	2141
## 12	657	320	296	2498
## 13	618	292	246	2094
## 14	512	339	298	2165
## 15	519	322	286	2371
## 16	440	329	187	1810
## 17	459	287	197	1723
## 18	664	450	196	1881
## 19	81	84	105	6723
## 20	116	72	61	6778
## 21	273	559	NA	2477
## 22	322	833	NA	2092
## 23	415	786	NA	1849
## 24	469	746	333	1446
## 25	563	401	654	1925
## 26	352	566	385	1493
## 27	502	643	373	1741
## 28	596	724	481	6313
## 29	661	664	410	1752
## 30	550	529	250	1630
## 31	669	403	290	1989
## 32	653	253	410	2638
## 33	442	303	386	2376
## 34	485	252	554	2441
## 35	545	319	500	2460
## 36	564	344	272	1872
## 37	458	419	398	1808
## 38	222	1006	NA	2266
## 39	324	103	394	12420
## 40	97	90	162	8041
## 41	170	66	107	7013
## 42	74	0	NA	5253
## 43	64	0	NA	4727
## 44	83	105	74	7093
## 45	159	129	54	5292
## 46	94	0	NA	4312
## 47	190	1170	NA	2050
## 48	537	746	298	2059
## 49	819	546	517	2060
## 50	246	571	NA	2951
## 51	314	668	NA	2341
## 52	259	550	NA	2051
## 53	616	453	334	2040
## 54	224	685	NA	2382
## 55	294	606	NA	2028
## 56	232	625	NA	2377
## 57	246	833	NA	1954

## 58	93	0	NA	16871
## 59	145	140	101	10234
## 60	125	96	88	14749
## 61	213	660	NA	2733
## 62	382	678	NA	2460
## 63	492	621	NA	2258
## 64	601	670	278	2088
## 65	545	536	511	2110
## 66	348	676	344	1729
## 67	636	633	298	1919
## 68	631	548	286	1776
## 69	474	535	259	1462
## 70	609	366	401	2342
## 71	518	367	420	2100
## 72	528	389	324	2042
## 73	514	426	238	1669
## 74	444	374	270	1711
## 75	582	485	326	1833
## 76	230	431	NA	2627
## 77	229	600	NA	2290
## 78	221	608	NA	2188
## 79	439	743	217	1688
## 80	458	439	632	1850
## 81	417	671	567	1697
## 82	527	596	538	1775
## 83	531	463	383	1814
## 84	490	519	289	1619
## 85	543	512	292	1502
## 86	673	324	301	1914
## 87	633	314	268	2133
## 88	516	310	406	2166
## 89	466	268	238	1975
## 90	485	320	176	1645
## 91	524	319	246	1678
## 92	693	624	266	2200
## 93	634	430	224	2132
## 94	595	715	357	1719
## 95	659	673	427	1790
## 96	626	627	331	1556
## 97	464	573	439	1789
## 98	387	686	433	1621
## 99	610	582	187	1616
## 100	605	541	282	1758
## 101	600	583	245	1625
## 102	673	290	319	2283
## 103	591	378	276	2274
## 104	588	450	233	2209
## 105	552	400	221	2344
## 106	538	425	224	2101
## 107	593	333	101	1630
## 108	304	295	134	1475
## 109	93	0	NA	5203
## 110	58	0	NA	4074
## 111	227	744	NA	2445

## 112	212	899	NA	2352
## 113	274	625	NA	2544
## 114	324	677	NA	2064
## 115	493	561	256	2186
## 116	460	337	349	2396
## 117	380	490	239	1950
## 118	78	162	45	7335
## 119	79	0	NA	5484
## 120	79	0	NA	3408
## 121	189	848	NA	2039
## 122	386	487	429	1786
## 123	283	590	420	1584
## 124	456	540	305	1981
## 125	353	741	315	1506
## 126	508	740	558	1756
## 127	252	1085	NA	2342
## 128	281	77	136	12943
## 129	279	650	NA	2050
## 130	344	469	NA	2021
## 131	512	618	293	1763
## 132	551	441	494	1873
## 133	650	474	460	1769
## 134	649	453	438	1839
## 135	549	514	398	1724
## 136	662	532	430	1593
## 137	599	375	270	1922
## 138	576	364	349	2305
## 139	491	393	226	2034
## 140	425	333	245	1976
## 141	431	313	230	2023
## 142	366	351	145	1640
## 143	234	509	NA	2485
## 144	219	612	NA	2259
## 145	220	648	NA	2063
## 146	488	665	241	1836
## 147	519	424	555	2014
## 148	386	725	382	1694
## 149	376	612	238	1721
## 150	503	565	419	1973
## 151	524	559	272	1744
## 152	461	541	293	1372
## 153	629	397	263	1991
## 154	436	454	270	1873
## 155	428	399	193	2018
## 156	459	528	241	1830
## 157	461	565	243	1859
## 158	402	460	252	1599
## 159	465	400	248	1686
## 160	34	81	34	13898
## 161	0	0	0	24057
## 162	482	513	212	6723
## 163	270	945	NA	16038
## 164	45	0	NA	4443
## 165	162	74	226	9710

## 166	165	156	168	6012
## 167	128	67	46	5811
## 168	95	0	NA	4404
## 169	50	0	NA	3552
## 170	601	450	286	2131
## 171	240	437	NA	2514
## 172	307	477	NA	2412
## 173	396	694	144	1507
## 174	535	564	372	1776
## 175	72	72	0	30132
## 176	720	475	315	1910
## 177	321	443	NA	2656
## 178	232	643	NA	2422
## 179	323	593	NA	2300
## 180	454	836	341	1644
## 181	369	576	529	1794
## 182	583	663	374	1592
## 183	302	741	NA	1954
## 184	401	733	321	1971
## 185	507	456	468	2201
## 186	334	604	307	1585
## 187	501	450	343	2031
## 188	646	498	414	1938
## 189	570	487	274	1739
## 190	559	545	229	1592
## 191	588	421	254	2449
## 192	628	310	346	2774
## 193	546	370	238	2146
## 194	490	367	200	2096
## 195	513	415	198	1692
## 196	470	363	226	1832
## 197	249	979	NA	3612
## 198	533	397	251	2061
## 199	521	639	174	1767
## 200	268	570	NA	2438
## 201	215	617	NA	1749
## 202	276	784	NA	2034
## 203	565	843	307	1638
## 204	417	498	289	1948
## 205	235	705	347	1564
## 206	486	546	248	1544
## 207	461	480	235	1444
## 208	674	343	264	2282
## 209	461	367	315	2032
## 210	486	359	273	2163
## 211	444	413	210	1937
## 212	368	375	117	1573
## 213	266	133	324	13724
## 214	254	173	214	9190
## 215	71	0	NA	5605
## 216	89	0	NA	4629
## 217	246	511	NA	2736
## 218	434	678	NA	2066
## 219	390	633	NA	1779

## 220	190	119	134	6028
## 221	78	0	NA	5382
## 222	52	0	NA	3864
## 223	56	155	14	10121
## 224	204	1095	NA	4837
## 225	573	692	399	1799
## 226	246	194	343	11508
## 227	214	501	NA	2563
## 228	320	451	NA	2269
## 229	323	558	211	2143
## 230	475	429	547	2184
## 231	334	564	388	1757
## 232	692	496	375	2181
## 233	433	592	357	1912
## 234	538	484	275	1915
## 235	547	503	323	1527
## 236	619	342	367	2145
## 237	584	266	392	2178
## 238	561	361	361	2024
## 239	543	335	339	2115
## 240	500	404	406	2216
## 241	462	402	231	1659
## 242	418	389	253	1669
## 243	259	777	NA	2570
## 244	321	618	NA	1866
## 245	312	871	207	1765
## 246	119	0	NA	10035
## 247	79	0	NA	3638
## 248	207	397	NA	2438
## 249	263	513	NA	2260
## 250	338	408	NA	2048
## 251	466	495	392	1854
## 252	530	408	697	2232
## 253	492	625	562	1712
## 254	592	572	403	1889
## 255	565	584	366	1856
## 256	744	524	337	1887
## 257	656	530	226	1385
## 258	643	308	307	1940
## 259	475	345	254	2055
## 260	414	374	231	1804
## 261	438	388	213	1953
## 262	414	434	112	1504
## 263	502	281	225	1742
## 264	142	952	NA	20088
## 265	583	615	372	1884
## 266	596	685	517	1735
## 267	245	843	NA	2084
## 268	284	110	359	12856
## 269	45	91	58	13815
## 270	562	582	263	1704
## 271	568	593	296	1433
## 272	658	298	193	1987
## 273	757	460	306	1983

## 274	656	501	300	2105
## 275	459	427	255	2073
## 276	394	411	210	1617
## 277	373	363	188	1623
## 278	79	137	21	7041
## 279	12	0	NA	12574
## 280	171	807	NA	1948
## 281	358	786	193	1561
## 282	357	450	444	1832
## 283	297	603	400	1379
## 284	609	596	303	1887
## 285	29	0	NA	6492
## 286	262	969	NA	2870
##	TEAM_PITCHING_HR	TEAM_PITCHING_BB	TEAM_PITCHING_SO	TEAM_FIELDING_E
## 1	84	927	5456	1011
## 2	93	336	1155	631
## 3	77	441	1405	546
## 4	46	398	1092	644
## 5	46	479	1030	644
## 6	95	609	703	699
## 7	82	412	765	597
## 8	67	754	721	523
## 9	46	787	764	436
## 10	74	734	742	420
## 11	103	885	461	443
## 12	155	806	393	509
## 13	83	764	361	450
## 14	56	633	419	455
## 15	68	637	395	334
## 16	63	478	358	325
## 17	45	489	306	314
## 18	31	815	552	603
## 19	126	243	252	1296
## 20	101	336	208	1059
## 21	14	461	943	1053
## 22	75	492	1273	705
## 23	37	617	1168	623
## 24	12	580	923	648
## 25	44	676	481	659
## 26	26	416	669	545
## 27	29	602	772	643
## 28	48	2840	3450	519
## 29	43	793	797	604
## 30	36	606	583	644
## 31	42	834	502	479
## 32	53	826	320	371
## 33	39	551	378	359
## 34	36	609	316	372
## 35	30	679	398	345
## 36	14	613	374	354
## 37	21	501	459	337
## 38	40	343	1552	941
## 39	86	1750	556	1215
## 40	83	334	310	907

## 41	99	467	181	1192
## 42	97	171	0	1058
## 43	59	131	0	951
## 44	0	363	460	1567
## 45	54	477	387	1515
## 46	9	277	0	1473
## 47	42	282	1739	941
## 48	34	879	1221	645
## 49	85	1029	686	576
## 50	22	411	954	743
## 51	83	458	975	676
## 52	48	375	796	671
## 53	52	756	556	652
## 54	42	374	1144	894
## 55	90	441	909	650
## 56	22	387	1044	650
## 57	33	356	1205	741
## 58	0	1076	0	1898
## 59	218	810	782	1246
## 60	0	920	707	1237
## 61	35	352	1091	898
## 62	297	553	981	861
## 63	113	712	898	719
## 64	90	785	875	621
## 65	143	730	718	632
## 66	110	418	811	500
## 67	119	781	777	568
## 68	86	757	658	476
## 69	32	526	594	470
## 70	98	747	449	555
## 71	54	656	464	464
## 72	58	648	477	484
## 73	21	555	460	445
## 74	33	486	409	468
## 75	44	688	574	582
## 76	93	380	712	595
## 77	80	340	892	612
## 78	55	320	879	603
## 79	62	515	872	681
## 80	53	550	527	581
## 81	47	504	811	551
## 82	71	614	695	513
## 83	41	652	568	468
## 84	56	579	614	484
## 85	52	586	553	434
## 86	47	852	410	406
## 87	95	789	391	523
## 88	56	643	386	470
## 89	33	572	329	335
## 90	21	517	341	346
## 91	15	566	345	366
## 92	73	1123	1011	630
## 93	42	790	536	663
## 94	49	698	839	573

## 95	23	797	814	475
## 96	28	740	741	448
## 97	21	574	709	712
## 98	18	475	842	589
## 99	32	749	714	497
## 100	30	705	631	565
## 101	30	652	634	441
## 102	52	852	367	492
## 103	58	742	475	432
## 104	45	733	561	434
## 105	44	699	506	354
## 106	25	665	526	318
## 107	22	645	362	322
## 108	14	320	310	408
## 109	22	255	0	1225
## 110	0	140	0	931
## 111	62	350	1148	821
## 112	32	377	1600	817
## 113	35	453	1033	734
## 114	57	486	1016	668
## 115	92	649	739	491
## 116	94	601	440	502
## 117	78	470	606	573
## 118	0	436	905	1553
## 119	18	241	0	1531
## 120	8	156	0	853
## 121	45	286	1284	780
## 122	54	496	626	616
## 123	49	340	708	539
## 124	91	551	653	507
## 125	28	433	909	622
## 126	25	601	875	722
## 127	25	517	2225	1066
## 128	145	2396	657	1390
## 129	39	435	1013	791
## 130	29	498	678	631
## 131	22	605	731	721
## 132	36	666	533	679
## 133	66	769	560	498
## 134	68	815	569	401
## 135	32	649	608	511
## 136	34	696	560	419
## 137	72	758	475	487
## 138	64	712	450	482
## 139	59	607	486	402
## 140	43	526	412	367
## 141	36	529	384	447
## 142	21	409	392	373
## 143	38	391	850	787
## 144	39	329	918	639
## 145	39	318	937	665
## 146	29	581	792	706
## 147	38	618	505	684
## 148	20	463	870	731

## 149	31	441	718	686
## 150	22	617	693	466
## 151	24	620	661	542
## 152	20	491	577	500
## 153	32	815	515	425
## 154	65	543	566	533
## 155	52	529	493	590
## 156	57	568	653	587
## 157	62	574	704	492
## 158	36	431	494	410
## 159	46	496	426	407
## 160	47	230	547	1512
## 161	0	0	0	1890
## 162	266	2169	2309	522
## 163	0	3645	12758	716
## 164	24	155	0	1506
## 165	25	795	363	1114
## 166	36	495	468	978
## 167	46	391	205	1281
## 168	50	237	0	1092
## 169	41	119	0	1253
## 170	101	743	556	555
## 171	17	405	737	660
## 172	50	465	722	668
## 173	26	475	833	654
## 174	32	652	687	765
## 175	0	1296	1296	1728
## 176	56	911	601	648
## 177	55	531	732	965
## 178	59	351	974	693
## 179	62	467	858	697
## 180	30	545	1003	764
## 181	46	449	702	579
## 182	36	716	814	556
## 183	29	437	1072	775
## 184	53	570	1042	558
## 185	82	668	601	620
## 186	25	416	753	528
## 187	71	639	574	594
## 188	35	799	616	490
## 189	30	674	576	524
## 190	56	592	577	416
## 191	126	738	529	399
## 192	65	795	392	428
## 193	76	680	461	390
## 194	60	601	450	363
## 195	39	558	451	412
## 196	35	501	387	404
## 197	41	602	2367	1211
## 198	75	659	491	631
## 199	75	607	745	453
## 200	35	443	942	836
## 201	5	323	926	785
## 202	10	403	1144	616

## 203	22	668	997	576
## 204	34	545	651	555
## 205	21	284	852	503
## 206	29	579	650	723
## 207	42	488	508	508
## 208	58	846	431	435
## 209	39	566	450	481
## 210	43	610	451	398
## 211	38	549	511	428
## 212	16	403	410	372
## 213	301	1539	770	1122
## 214	55	935	637	1097
## 215	53	209	0	1166
## 216	49	198	0	928
## 217	58	407	845	693
## 218	43	628	981	576
## 219	13	574	932	676
## 220	73	581	364	1103
## 221	22	218	0	1447
## 222	29	124	0	1132
## 223	0	394	1092	1740
## 224	104	787	4224	918
## 225	51	737	890	535
## 226	123	1594	1257	1426
## 227	69	361	845	790
## 228	33	463	652	479
## 229	39	440	760	489
## 230	83	626	565	566
## 231	84	413	697	532
## 232	86	890	638	561
## 233	38	535	732	544
## 234	69	660	594	471
## 235	45	587	540	606
## 236	92	760	420	530
## 237	65	717	326	544
## 238	49	694	446	542
## 239	61	671	414	451
## 240	47	618	500	491
## 241	40	499	434	483
## 242	27	451	420	468
## 243	66	371	1114	794
## 244	18	482	927	597
## 245	53	414	1157	600
## 246	0	1015	0	1279
## 247	0	188	0	1010
## 248	20	342	656	641
## 249	26	398	777	740
## 250	36	489	590	551
## 251	27	543	577	576
## 252	56	636	490	577
## 253	52	590	750	516
## 254	84	710	686	526
## 255	68	673	696	562
## 256	82	886	624	555

## 257	53	708	572	489
## 258	15	789	378	488
## 259	58	587	427	470
## 260	57	516	466	430
## 261	47	542	480	464
## 262	15	447	469	419
## 263	54	539	301	428
## 264	0	2876	19278	952
## 265	23	744	784	499
## 266	36	731	841	514
## 267	19	382	1313	731
## 268	203	1643	636	1146
## 269	207	292	590	933
## 270	28	674	698	707
## 271	43	609	636	587
## 272	36	826	374	624
## 273	88	929	565	612
## 274	89	830	634	562
## 275	54	563	524	453
## 276	41	420	438	471
## 277	53	398	387	430
## 278	50	328	569	997
## 279	0	150	0	847
## 280	9	247	1167	796
## 281	42	483	1061	618
## 282	82	474	598	623
## 283	44	359	729	573
## 284	43	796	779	678
## 285	0	168	0	1522
## 286	39	674	2492	1026
##	TEAM_FIELDING_DP	TEAM_CS_YES_NO	TEAM_HBP_YES_NO	
## 1	NA	0	0	
## 2	NA	0	0	
## 3	NA	0	0	
## 4	NA	0	0	
## 5	NA	0	0	
## 6	NA	0	0	
## 7	NA	0	0	
## 8	NA	0	0	
## 9	NA	0	0	
## 10	NA	0	0	
## 11	NA	0	0	
## 12	NA	0	0	
## 13	NA	0	0	
## 14	NA	0	0	
## 15	NA	0	0	
## 16	NA	0	0	
## 17	NA	0	0	
## 18	NA	0	0	
## 19	NA	1	0	
## 20	NA	1	0	
## 21	NA	0	0	
## 22	NA	0	0	
## 23	NA	0	0	

## 24	NA	0	0
## 25	NA	0	0
## 26	NA	0	0
## 27	NA	0	0
## 28	NA	0	0
## 29	NA	0	0
## 30	NA	0	0
## 31	NA	0	0
## 32	NA	0	0
## 33	NA	0	0
## 34	NA	0	0
## 35	NA	0	0
## 36	NA	0	0
## 37	NA	0	0
## 38	NA	0	0
## 39	NA	0	0
## 40	NA	1	0
## 41	NA	1	0
## 42	NA	0	0
## 43	NA	0	0
## 44	NA	1	0
## 45	NA	1	0
## 46	NA	0	0
## 47	NA	0	0
## 48	NA	0	0
## 49	NA	0	0
## 50	NA	0	0
## 51	NA	0	0
## 52	NA	0	0
## 53	NA	0	0
## 54	NA	0	0
## 55	NA	0	0
## 56	NA	0	0
## 57	NA	0	0
## 58	NA	0	0
## 59	NA	0	0
## 60	NA	1	0
## 61	NA	0	0
## 62	NA	0	0
## 63	NA	0	0
## 64	NA	0	0
## 65	NA	0	0
## 66	NA	0	0
## 67	NA	0	0
## 68	NA	0	0
## 69	NA	0	0
## 70	NA	0	0
## 71	NA	0	0
## 72	NA	0	0
## 73	NA	0	0
## 74	NA	0	0
## 75	NA	0	0
## 76	NA	0	0
## 77	NA	0	0

## 78	NA	0	0
## 79	NA	0	0
## 80	NA	0	0
## 81	NA	0	0
## 82	NA	0	0
## 83	NA	0	0
## 84	NA	0	0
## 85	NA	0	0
## 86	NA	0	0
## 87	NA	0	0
## 88	NA	0	0
## 89	NA	0	0
## 90	NA	0	0
## 91	NA	0	0
## 92	NA	0	0
## 93	NA	0	0
## 94	NA	0	0
## 95	NA	0	0
## 96	NA	0	0
## 97	NA	0	0
## 98	NA	0	0
## 99	NA	0	0
## 100	NA	0	0
## 101	NA	0	0
## 102	NA	0	0
## 103	NA	0	0
## 104	NA	0	0
## 105	NA	0	0
## 106	NA	0	0
## 107	NA	0	0
## 108	NA	0	0
## 109	NA	0	0
## 110	NA	0	0
## 111	NA	0	0
## 112	NA	0	0
## 113	NA	0	0
## 114	NA	0	0
## 115	NA	0	0
## 116	NA	0	0
## 117	NA	0	0
## 118	NA	1	0
## 119	NA	0	0
## 120	NA	0	0
## 121	NA	0	0
## 122	NA	0	0
## 123	NA	0	0
## 124	NA	0	0
## 125	NA	0	0
## 126	NA	0	0
## 127	NA	0	0
## 128	NA	0	0
## 129	NA	0	0
## 130	NA	0	0
## 131	NA	0	0

## 132	NA	0	0
## 133	NA	0	0
## 134	NA	0	0
## 135	NA	0	0
## 136	NA	0	0
## 137	NA	0	0
## 138	NA	0	0
## 139	NA	0	0
## 140	NA	0	0
## 141	NA	0	0
## 142	NA	0	0
## 143	NA	0	0
## 144	NA	0	0
## 145	NA	0	0
## 146	NA	0	0
## 147	NA	0	0
## 148	NA	0	0
## 149	NA	0	0
## 150	NA	0	0
## 151	NA	0	0
## 152	NA	0	0
## 153	NA	0	0
## 154	NA	0	0
## 155	NA	0	0
## 156	NA	0	0
## 157	NA	0	0
## 158	NA	0	0
## 159	NA	0	0
## 160	NA	1	0
## 161	NA	1	0
## 162	NA	0	0
## 163	NA	0	0
## 164	NA	0	0
## 165	NA	0	0
## 166	NA	1	0
## 167	NA	1	0
## 168	NA	0	0
## 169	NA	0	0
## 170	NA	0	0
## 171	NA	0	0
## 172	NA	0	0
## 173	NA	0	0
## 174	NA	0	0
## 175	NA	1	0
## 176	NA	0	0
## 177	NA	0	0
## 178	NA	0	0
## 179	NA	0	0
## 180	NA	0	0
## 181	NA	0	0
## 182	NA	0	0
## 183	NA	0	0
## 184	NA	0	0
## 185	NA	0	0

## 186	NA	0	0
## 187	NA	0	0
## 188	NA	0	0
## 189	NA	0	0
## 190	NA	0	0
## 191	NA	0	0
## 192	NA	0	0
## 193	NA	0	0
## 194	NA	0	0
## 195	NA	0	0
## 196	NA	0	0
## 197	NA	0	0
## 198	NA	0	0
## 199	NA	0	0
## 200	NA	0	0
## 201	NA	0	0
## 202	NA	0	0
## 203	NA	0	0
## 204	NA	0	0
## 205	NA	0	0
## 206	NA	0	0
## 207	NA	0	0
## 208	NA	0	0
## 209	NA	0	0
## 210	NA	0	0
## 211	NA	0	0
## 212	NA	0	0
## 213	NA	0	0
## 214	NA	1	0
## 215	NA	0	0
## 216	NA	0	0
## 217	NA	0	0
## 218	NA	0	0
## 219	NA	0	0
## 220	NA	1	0
## 221	NA	0	0
## 222	NA	0	0
## 223	NA	1	0
## 224	NA	0	0
## 225	NA	0	0
## 226	NA	0	0
## 227	NA	0	0
## 228	NA	0	0
## 229	NA	0	0
## 230	NA	0	0
## 231	NA	0	0
## 232	NA	0	0
## 233	NA	0	0
## 234	NA	0	0
## 235	NA	0	0
## 236	NA	0	0
## 237	NA	0	0
## 238	NA	0	0
## 239	NA	0	0

## 240	NA	0	0
## 241	NA	0	0
## 242	NA	0	0
## 243	NA	0	0
## 244	NA	0	0
## 245	NA	0	0
## 246	NA	0	0
## 247	NA	0	0
## 248	NA	0	0
## 249	NA	0	0
## 250	NA	0	0
## 251	NA	0	0
## 252	NA	0	0
## 253	NA	0	0
## 254	NA	0	0
## 255	NA	0	0
## 256	NA	0	0
## 257	NA	0	0
## 258	NA	0	0
## 259	NA	0	0
## 260	NA	0	0
## 261	NA	0	0
## 262	NA	0	0
## 263	NA	0	0
## 264	NA	0	0
## 265	NA	0	0
## 266	NA	0	0
## 267	NA	0	0
## 268	NA	0	0
## 269	NA	1	0
## 270	NA	0	0
## 271	NA	0	0
## 272	NA	0	0
## 273	NA	0	0
## 274	NA	0	0
## 275	NA	0	0
## 276	NA	0	0
## 277	NA	0	0
## 278	NA	1	0
## 279	NA	0	0
## 280	NA	0	0
## 281	NA	0	0
## 282	NA	0	0
## 283	NA	0	0
## 284	NA	0	0
## 285	NA	0	0
## 286	NA	0	0

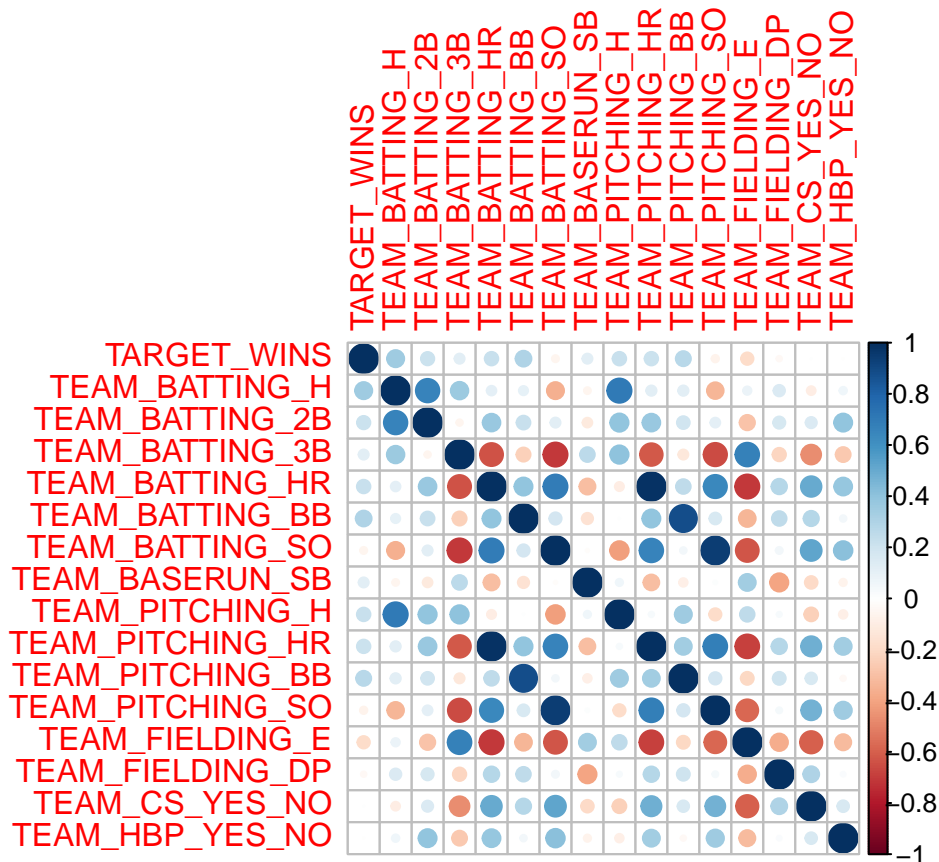
```
baseball_df_na_dp <- baseball_df_fix %>%
  filter(!is.na(Team_Fielding_DP))

summary(lm(baseball_df_na_dp, formula = Team_Fielding_DP ~ . - Target_Wins))
```

##

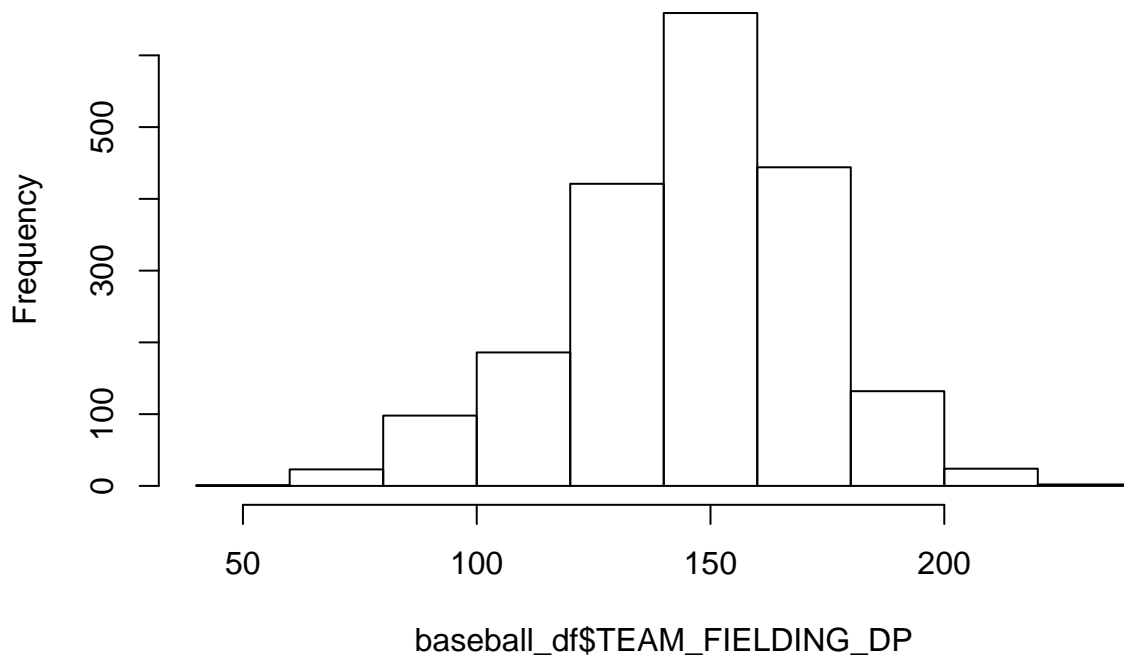
```
## Call:
## lm(formula = TEAM_FIELDING_DP ~ . - TARGET_WINS, data = baseball_df_na_dp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -54.601 -12.873  -0.814  12.674  61.123
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    127.88475    10.99343   11.633 < 2e-16 ***
## TEAM_BATTING_H     0.07151     0.03110    2.300  0.02158 *
## TEAM_BATTING_2B   -0.02827     0.01766   -1.601  0.10955
## TEAM_BATTING_3B   -0.10305     0.03578   -2.880  0.00402 **
## TEAM_BATTING_HR   -0.12579     0.15525   -0.810  0.41791
## TEAM_BATTING_BB   -0.01834     0.08048   -0.228  0.81978
## TEAM_BATTING_SO   -0.03544     0.04163   -0.851  0.39475
## TEAM_BASERUN_SB   -0.10282     0.01018  -10.097 < 2e-16 ***
## TEAM_PITCHING_H   -0.03362     0.02829   -1.188  0.23484
## TEAM_PITCHING_HR    0.12631     0.14854    0.850  0.39526
## TEAM_PITCHING_BB    0.04148     0.07648    0.542  0.58761
## TEAM_PITCHING_SO    0.01473     0.03959    0.372  0.70989
## TEAM_FIELDING_E   -0.08529     0.01374   -6.208 6.63e-10 ***
## TEAM_CS_YES_NO     9.51649     1.50229    6.335 2.99e-10 ***
## TEAM_HBP_YES_NO   -1.95290     1.83235   -1.066  0.28666
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 19.21 on 1820 degrees of freedom
## (155 observations deleted due to missingness)
## Multiple R-squared:  0.2889, Adjusted R-squared:  0.2834
## F-statistic: 52.81 on 14 and 1820 DF, p-value: < 2.2e-16
```

```
corrplot(cor(baseball_df_na_dp, use = 'complete.obs'))
```



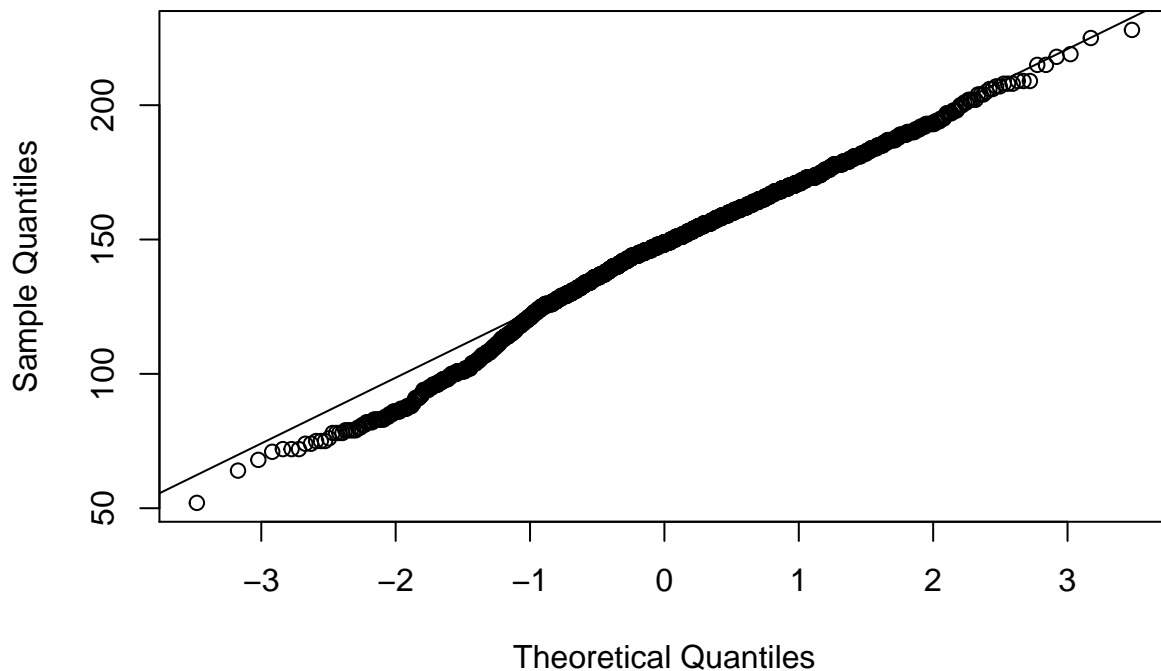
```
hist(baseball_df$TEAM_FIELDING_DP)
```

Histogram of baseball_df\$TEAM_FIELDING_DP



```
qqnorm(baseball_df$TEAM_FIELDING_DP)
qqline(baseball_df$TEAM_FIELDING_DP)
```

Normal Q-Q Plot



Fitting a Linear Model

My first change to the data was to eliminate the index and, replace HBP and BASERUN_CS with dummy variables.

```
baseball_df_fix <- baseball_df %>%
  mutate(TEAM_CS_YES_NO = case_when(!is.na(TEAM_BASERUN_CS) ~ 1, is.na(TEAM_BASERUN_CS) ~ 0)) %>%
  mutate(TEAM_HBP_YES_NO = case_when(!is.na(TEAM_BATTING_HBP) ~ 1, is.na(TEAM_BATTING_HBP) ~ 0)) %>%
  dplyr::select(-c(TEAM_BATTING_HBP, INDEX, TEAM_BASERUN_CS))

baseball_lm <- lm(baseball_df_fix, formula = TARGET_WINS ~.)

summary(baseball_lm)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ ., data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -32.424  -6.972   0.192   6.983  28.645
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)      57.987326   5.994875   9.673 < 2e-16 ***
## TEAM_BATTING_H   -0.027606   0.016383  -1.685 0.092156 .
## TEAM_BATTING_2B  -0.043602   0.009296  -4.691 2.93e-06 ***
## TEAM_BATTING_3B   0.186256   0.018867   9.872 < 2e-16 ***
## TEAM_BATTING_HR   0.155277   0.081692   1.901 0.057493 .
## TEAM_BATTING_BB   0.102630   0.042342   2.424 0.015456 *
## TEAM_BATTING_SO   0.030619   0.021908   1.398 0.162398
## TEAM_BASERUN_SB   0.068643   0.005505  12.469 < 2e-16 ***
## TEAM_PITCHING_H   0.053979   0.014889   3.625 0.000296 ***
## TEAM_PITCHING_HR -0.052190   0.078164  -0.668 0.504413
## TEAM_PITCHING_BB -0.064794   0.040239  -1.610 0.107522
## TEAM_PITCHING_SO -0.047628   0.020830  -2.286 0.022341 *
## TEAM_FIELDING_E  -0.127819   0.007304 -17.499 < 2e-16 ***
## TEAM_FIELDING_DP -0.104483   0.012332  -8.472 < 2e-16 ***
## TEAM_CS_YES_NO   -3.839845   0.799028  -4.806 1.67e-06 ***
## TEAM_HBP_YES_NO  -2.647365   0.964312  -2.745 0.006104 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.1 on 1819 degrees of freedom
## (441 observations deleted due to missingness)
## Multiple R-squared:  0.4152, Adjusted R-squared:  0.4103
## F-statistic: 86.09 on 15 and 1819 DF, p-value: < 2.2e-16
```

The initial linear model explains 41% of variation. Next, I'll add some log transformations of skewed columns: TEAM_PITCHING_BB, TEAM_PITCHING_SO, TEAM_BASERUN_SB, and TEAM_FIELDING_E.

```
baseball_log_lm <- lm(baseball_df_fix, formula = TARGET_WINS ~. + log(TEAM_FIELDING_E) + log(TEAM_PITCHING_BB) +
summary(baseball_log_lm)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ . + log(TEAM_FIELDING_E) + log(TEAM_PITCHING_BB) +
##     log(TEAM_PITCHING_SO) + log(TEAM_BASERUN_SB), data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -33.639  -6.850   0.083   6.851  29.725
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    209.367276   78.152118   2.679 0.007452 **
## TEAM_BATTING_H    -0.029153   0.016874  -1.728 0.084208 .
## TEAM_BATTING_2B   -0.038719   0.009338  -4.146 3.54e-05 ***
## TEAM_BATTING_3B    0.195250   0.019028  10.261 < 2e-16 ***
## TEAM_BATTING_HR    0.142934   0.082824   1.726 0.084560 .
## TEAM_BATTING_BB    0.140220   0.044797   3.130 0.001775 **
## TEAM_BATTING_SO    0.013917   0.022007   0.632 0.527220
## TEAM_BASERUN_SB    0.084057   0.015811   5.316 1.19e-07 ***
## TEAM_PITCHING_H    0.056607   0.015418   3.671 0.000248 ***
## TEAM_PITCHING_HR   -0.048334   0.079109  -0.611 0.541295
## TEAM_PITCHING_BB   -0.035418   0.041146  -0.861 0.389466
```

```
## TEAM_PITCHING_SO      -0.053456    0.021981   -2.432 0.015117 *
## TEAM_FIELDING_E       -0.069984    0.021909   -3.194 0.001426 **
## TEAM_FIELDING_DP      -0.104523    0.012260   -8.525 < 2e-16 ***
## TEAM_CS_YES_NO        -3.568860    0.803825   -4.440 9.54e-06 ***
## TEAM_HBP_YES_NO       -3.061485    1.007413   -3.039 0.002408 **
## log(TEAM_FIELDING_E)  -11.718073    4.037533   -2.902 0.003749 **
## log(TEAM_PITCHING_BB) -36.670662   12.641719   -2.901 0.003767 **
## log(TEAM_PITCHING_SO)  17.416259    6.179754    2.818 0.004881 **
## log(TEAM_BASERUN_SB)  -2.068123    1.483764   -1.394 0.163538
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.03 on 1815 degrees of freedom
## (441 observations deleted due to missingness)
## Multiple R-squared:  0.4248, Adjusted R-squared:  0.4188
## F-statistic: 70.55 on 19 and 1815 DF, p-value: < 2.2e-16
```

This model explains more variation, but the F-statistic decreased relative to the original model. Next, I'm going to add a few features I'm curious about. TEAM_BATTING_H considers all base hits, including 2B, 3B, and HR. I will create a new variable only looking at singles called TEAM_BATTING_1B. Related to this, I will also incorporate an approximation of an important baseball statistic, slugging. Because some base hits convert to runs at different rates, slugging weighs, singles, doubles, triples and home runs with increasing weight. Usually, slugging also has a denominator of at-bats, which is unavailable in this dataset. Instead, I'll approximate this by dividing by the number of hits. The weights I'm assigning are proportional to the number of bases, so 1 for single, 2 for double... 4 for HR.

```
baseball_df_fix <- baseball_df_fix %>%
  mutate(TEAM_BATTING_1B = TEAM_BATTING_H - TEAM_BATTING_2B - TEAM_BATTING_3B - TEAM_BATTING_HR) %>%
  mutate(TEAM_BATTING_SLG = (TEAM_BATTING_H + TEAM_BATTING_2B + 2 * TEAM_BATTING_3B + 3 * TEAM_BATTING_HR) / TEAM_BATTING_H)

baseball_vars_lm <- lm(baseball_df_fix, formula = TARGET_WINS ~ . + log(TEAM_FIELDING_E) + log(TEAM_PITCHING_BB) +
  log(TEAM_PITCHING_SO) + log(TEAM_BASERUN_SB), data = baseball_df_fix)

summary(baseball_vars_lm)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ . + log(TEAM_FIELDING_E) + log(TEAM_PITCHING_BB) +
##     log(TEAM_PITCHING_SO) + log(TEAM_BASERUN_SB), data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -33.610  -6.830   0.047   6.788  29.845
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    178.96333    99.72432   1.795 0.072887 .
## TEAM_BATTING_H     -0.02227     0.02194  -1.015 0.310078
## TEAM_BATTING_2B    -0.05084     0.02640  -1.926 0.054310 .
## TEAM_BATTING_3B     0.17007     0.05471   3.109 0.001909 **
## TEAM_BATTING_HR     0.10812     0.10905   0.991 0.321574
## TEAM_BATTING_BB     0.13896     0.04488   3.096 0.001989 **
## TEAM_BATTING_SO     0.01206     0.02233   0.540 0.589305
```

```
## TEAM_BASERUN_SB      0.08407    0.01581    5.316 1.19e-07 ***
## TEAM_PITCHING_H      0.05546    0.01560    3.556 0.000386 ***
## TEAM_PITCHING_HR     -0.05053    0.07925   -0.638 0.523812
## TEAM_PITCHING_BB     -0.03328    0.04138   -0.804 0.421329
## TEAM_PITCHING_SO     -0.05302    0.02200   -2.410 0.016069 *
## TEAM_FIELDING_E      -0.07007    0.02191   -3.197 0.001411 **
## TEAM_FIELDING_DP     -0.10475    0.01227   -8.536 < 2e-16 ***
## TEAM_CS_YES_NO      -3.62119    0.81103   -4.465 8.50e-06 ***
## TEAM_HBP_YES_NO     -3.00343    1.01454   -2.960 0.003112 **
## TEAM_BATTING_1B      NA          NA          NA          NA
## TEAM_BATTING_SLG     18.12471   36.91502    0.491 0.623497
## log(TEAM_FIELDING_E) -11.66550    4.03980   -2.888 0.003927 **
## log(TEAM_PITCHING_BB) -37.13340   12.67944   -2.929 0.003447 **
## log(TEAM_PITCHING_SO) 18.47861    6.54882    2.822 0.004829 **
## log(TEAM_BASERUN_SB) -2.03114    1.48598   -1.367 0.171838
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.03 on 1814 degrees of freedom
## (441 observations deleted due to missingness)
## Multiple R-squared:  0.4249, Adjusted R-squared:  0.4185
## F-statistic: 67.01 on 20 and 1814 DF, p-value: < 2.2e-16
```

Neither of these features offered additional significance. Finally, I'll use back-selection to eliminate non-contributing variables.

```
baseball_back_lm <- lm(baseball_df_fix, formula = TARGET_WINS ~.-TEAM_BATTING_1B+log(TEAM_FIELDING_E) +
summary(baseball_back_lm)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ . - TEAM_BATTING_1B + log(TEAM_FIELDING_E) +
##     log(TEAM_PITCHING_BB) + log(TEAM_PITCHING_SO) + log(TEAM_BASERUN_SB) -
##     TEAM_BATTING_SLG - TEAM_PITCHING_H - TEAM_BATTING_BB - TEAM_BATTING_SO -
##     TEAM_PITCHING_HR - TEAM_PITCHING_BB - TEAM_FIELDING_E, data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -31.5371  -6.9106   0.1119   7.0369  28.4639
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -26.850089   38.695974  -0.694 0.487850
## TEAM_BATTING_H    0.029999    0.004279   7.010 3.34e-12 ***
## TEAM_BATTING_2B   -0.040948    0.009282  -4.412 1.09e-05 ***
## TEAM_BATTING_3B    0.199025    0.019098  10.421 < 2e-16 ***
## TEAM_BATTING_HR    0.094197    0.008831  10.666 < 2e-16 ***
## TEAM_BASERUN_SB   0.073484    0.015580   4.717 2.58e-06 ***
## TEAM_PITCHING_SO  -0.039449    0.006805  -5.797 7.94e-09 ***
## TEAM_FIELDING_DP  -0.103899    0.012282  -8.459 < 2e-16 ***
## TEAM_CS_YES_NO   -3.536451    0.786013  -4.499 7.25e-06 ***
## TEAM_HBP_YES_NO  -3.574125    0.935744  -3.820 0.000138 ***
```

```
## log(TEAM_FIELDING_E) -23.337095 1.331275 -17.530 < 2e-16 ***
## log(TEAM_PITCHING_BB) 17.535210 1.507901 11.629 < 2e-16 ***
## log(TEAM_PITCHING_SO) 16.615041 5.730249 2.900 0.003782 **
## log(TEAM_BASERUN_SB) -1.478015 1.483772 -0.996 0.319324
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.1 on 1821 degrees of freedom
## (441 observations deleted due to missingness)
## Multiple R-squared: 0.4148, Adjusted R-squared: 0.4106
## F-statistic: 99.28 on 13 and 1821 DF, p-value: < 2.2e-16
```

Before moving to the final model, I want to try creating a simple model with fewer predictors to see how it performs compared to our other models. To start I chose a few variables that were highly positively and negatively correlated with TARGET_WINS.

```
* TEAM_BATTING_H * TEAM_BATTING_HR * TEAM_BATTING_BB * TEAM_PITCHING_H *
TEAM_PITCHING_HR * TEAM_PITCHING_BB * TEAM_FIELDING_E * TEAM_FIELDING_DP
* TEAM_BATTING_SO * TEAM_CS_YES_NO
```

From there I removed multiple predictors at once. To do this we need to construct a null hypothesis test which states that removing the variables doesn't make a better model. We construct a F-test and compare both versions of the model. If the p-value is under 0.05 we reject the null hypothesis, which indicates our new model isn't different than the first model. If the p-value is greater than 0.05, the model isn't better with those variables, so I will remove them. The simpler the model the better.

To determine which variables I removed, I chose the variable that was not proving to be significant in the linear regression (where the p-value was greater than 0.05). While this doesn't mean the variable itself isn't significant, it means the variable alongside the other combination of variables in the model is not significant.

Steps: * Remove TEAM_PITCHING_BB *

- TEAM_PITCHING_BB & TEAM_BATTING_BB
- TEAM_PITCHING_SO & TEAM_BATTING_SO
- TEAM_PITCHING_HR & TEAM_BATTING_HR
- TEAM_PITCHING_H & TEAM_BATTING_H

Steps: * Remove TEAM_PITCHING_BB & TEAM_PITCHING_SO

```
m1 <- lm(TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR + TEAM_BATTING_BB + TEAM_BATTING_SO + TEAM_PITCHING_H + TEAM_PITCHING_HR + TEAM_PITCHING_BB + TEAM_PITCHING_SO + TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO, data = baseball_df_fix)
summary(m1)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR +
##     TEAM_BATTING_BB + TEAM_BATTING_SO + TEAM_PITCHING_H + TEAM_PITCHING_HR +
##     TEAM_PITCHING_BB + TEAM_PITCHING_SO + TEAM_FIELDING_E + TEAM_FIELDING_DP +
##     TEAM_CS_YES_NO, data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -45.525  -7.848   0.161   7.880  40.171
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
```



```
## (Intercept)      54.830197    5.962944    9.195 < 2e-16 ***
## TEAM_BATTING_H    0.014811    0.004169    3.553 0.000391 ***
## TEAM_BATTING_HR   0.254967    0.057271    4.452 9.01e-06 ***
## TEAM_BATTING_BB   0.052542    0.019298    2.723 0.006534 **
## TEAM_BATTING_SO  -0.027645    0.009317   -2.967 0.003043 **
## TEAM_PITCHING_H   0.018217    0.002332    7.813 9.22e-15 ***
## TEAM_PITCHING_HR -0.194561    0.054263   -3.586 0.000345 ***
## TEAM_PITCHING_BB -0.018421    0.017909   -1.029 0.303814
## TEAM_PITCHING_SO  0.011678    0.008444    1.383 0.166820
## TEAM_FIELDING_E  -0.061318    0.005580  -10.988 < 2e-16 ***
## TEAM_FIELDING_DP -0.149379    0.013160  -11.351 < 2e-16 ***
## TEAM_CS_YES_NO   -3.046049    0.874662   -3.483 0.000508 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.38 on 1876 degrees of freedom
## (388 observations deleted due to missingness)
## Multiple R-squared:  0.3036, Adjusted R-squared:  0.2995
## F-statistic: 74.35 on 11 and 1876 DF, p-value: < 2.2e-16
```

```
#remove TEAM_PITCHING_BB & TEAM_PITCHING_SO
```

```
m2<- lm(TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR +TEAM_BATTING_BB + TEAM_BATTING_SO + TEAM_PITCHING_H + TEAM_PITCHING_HR + TEAM_PITCHING_BB + TEAM_PITCHING_SO + TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO, data = baseball_df_fix)
```

```
summary(m2)
```

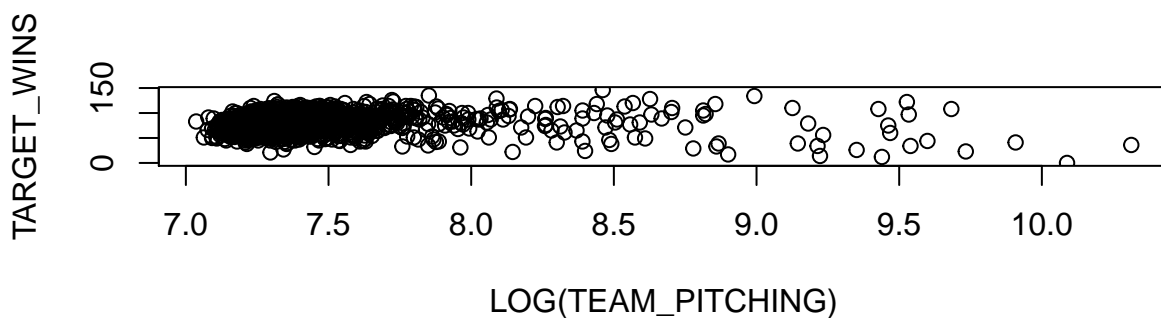
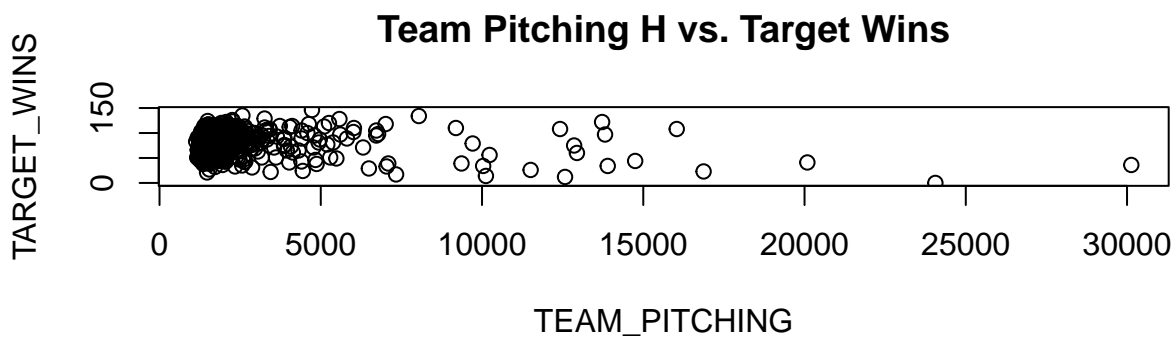
```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR +
##     TEAM_BATTING_BB + TEAM_BATTING_SO + TEAM_PITCHING_H + TEAM_PITCHING_HR +
##     TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO, data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -45.397  -7.831   0.217   7.924  40.201
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    54.329471    5.937886    9.150 < 2e-16 ***
## TEAM_BATTING_H     0.014275    0.003800    3.756 0.000178 ***
## TEAM_BATTING_HR    0.254520    0.032747    7.772 1.26e-14 ***
## TEAM_BATTING_BB    0.032968    0.003447    9.564 < 2e-16 ***
## TEAM_BATTING_SO   -0.015132    0.002258   -6.701 2.72e-11 ***
## TEAM_PITCHING_H     0.018472    0.001841   10.035 < 2e-16 ***
## TEAM_PITCHING_HR  -0.193336    0.031119   -6.213 6.40e-10 ***
## TEAM_FIELDING_E   -0.059177    0.005370  -11.020 < 2e-16 ***
## TEAM_FIELDING_DP  -0.147447    0.013090  -11.264 < 2e-16 ***
## TEAM_CS_YES_NO    -2.979270    0.865032   -3.444 0.000586 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.38 on 1878 degrees of freedom
## (388 observations deleted due to missingness)
## Multiple R-squared:  0.3028, Adjusted R-squared:  0.2995
## F-statistic: 90.65 on 9 and 1878 DF, p-value: < 2.2e-16
```

```
anova(m1, m2)
```

```
## Analysis of Variance Table
##
## Model 1: TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR + TEAM_BATTING_BB +
##     TEAM_BATTING_SO + TEAM_PITCHING_H + TEAM_PITCHING_HR + TEAM_PITCHING_BB +
##     TEAM_PITCHING_SO + TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO
## Model 2: TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR + TEAM_BATTING_BB +
##     TEAM_BATTING_SO + TEAM_PITCHING_H + TEAM_PITCHING_HR + TEAM_FIELDING_E +
##     TEAM_FIELDING_DP + TEAM_CS_YES_NO
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1   1876 243143
## 2   1878 243407 -2    -263.79 1.0176 0.3616
```

- Took the log of TEAM_PITCHING_H it's relationship to TARGET_WINS more linear

```
par(mfrow=c(2,1))
plot(baseball_df_fix$TEAM_PITCHING_H,baseball_df_fix$TARGET_WINS,xlab = 'TEAM_PITCHING',ylab = 'TARGET_WINS')
plot(log(baseball_df_fix$TEAM_PITCHING_H),baseball_df_fix$TARGET_WINS,xlab = 'LOG(TEAM_PITCHING)',ylab = 'TARGET_WINS')
```



```
#log TEAM_PITCHING_H
m3 <- lm(TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR + TEAM_BATTING_BB + TEAM_BATTING_SO + log(TEAM_PITCHING_H))
summary(m3)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR +
```

```
## TEAM_BATTING_BB + TEAM_BATTING_SO + log(TEAM_PITCHING_H) +
## TEAM_PITCHING_HR + TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO,
## data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -44.877  -7.714   0.053   8.006  37.267
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -3.068e+02  3.376e+01  -9.087  < 2e-16 ***
## TEAM_BATTING_H    -1.708e-03  4.855e-03  -0.352  0.72499
## TEAM_BATTING_HR     3.880e-01  4.108e-02   9.447  < 2e-16 ***
## TEAM_BATTING_BB     3.305e-02  3.440e-03   9.608  < 2e-16 ***
## TEAM_BATTING_SO    -1.409e-02  2.243e-03  -6.281  4.18e-10 ***
## log(TEAM_PITCHING_H)  5.639e+01  5.432e+00  10.382  < 2e-16 ***
## TEAM_PITCHING_HR    -3.254e-01  3.980e-02  -8.176  5.36e-16 ***
## TEAM_FIELDING_E     -6.592e-02  5.775e-03 -11.415  < 2e-16 ***
## TEAM_FIELDING_DP    -1.492e-01  1.307e-02 -11.414  < 2e-16 ***
## TEAM_CS_YES_NO     -2.390e+00  8.585e-01  -2.784  0.00543 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.36 on 1878 degrees of freedom
## (388 observations deleted due to missingness)
## Multiple R-squared:  0.3053, Adjusted R-squared:  0.302
## F-statistic: 91.72 on 9 and 1878 DF, p-value: < 2.2e-16
```

- Remove TEAM_BATTING_H

```
#Remove TEAM_BATTING_H
m4 <- lm(TARGET_WINS ~ TEAM_BATTING_HR + TEAM_BATTING_BB + TEAM_BATTING_SO + log(TEAM_PITCHING_H) + TEAM_PITCHING_HR + TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO, data = baseball_df_fix)
summary(m4)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_BATTING_HR + TEAM_BATTING_BB +
##      TEAM_BATTING_SO + log(TEAM_PITCHING_H) + TEAM_PITCHING_HR +
##      TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO, data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -44.874  -7.721   0.062   7.971  37.408
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -2.985e+02  2.407e+01 -12.400  < 2e-16 ***
## TEAM_BATTING_HR     3.786e-01  3.101e-02  12.208  < 2e-16 ***
## TEAM_BATTING_BB     3.294e-02  3.425e-03   9.618  < 2e-16 ***
## TEAM_BATTING_SO    -1.378e-02  2.069e-03  -6.663  3.51e-11 ***
## log(TEAM_PITCHING_H)  5.487e+01  3.280e+00  16.728  < 2e-16 ***
## TEAM_PITCHING_HR    -3.168e-01  3.130e-02 -10.122  < 2e-16 ***
## TEAM_FIELDING_E     -6.466e-02  4.526e-03 -14.287  < 2e-16 ***
```

```
## TEAM_FIELDING_DP      -1.491e-01  1.307e-02 -11.413  < 2e-16 ***
## TEAM_CS_YES_NO       -2.379e+00  8.578e-01  -2.773   0.0056 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.36 on 1879 degrees of freedom
## (388 observations deleted due to missingness)
## Multiple R-squared:  0.3053, Adjusted R-squared:  0.3023
## F-statistic: 103.2 on 8 and 1879 DF,  p-value: < 2.2e-16
```

```
anova(m3, m4)
```

```
## Analysis of Variance Table
##
## Model 1: TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_HR + TEAM_BATTING_BB +
##   TEAM_BATTING_SO + log(TEAM_PITCHING_H) + TEAM_PITCHING_HR +
##   TEAM_FIELDING_E + TEAM_FIELDING_DP + TEAM_CS_YES_NO
## Model 2: TARGET_WINS ~ TEAM_BATTING_HR + TEAM_BATTING_BB + TEAM_BATTING_SO +
##   log(TEAM_PITCHING_H) + TEAM_PITCHING_HR + TEAM_FIELDING_E +
##   TEAM_FIELDING_DP + TEAM_CS_YES_NO
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1    1878 242538
## 2    1879 242554 -1    -15.989 0.1238  0.725
```

This leaves a model with an R-squared value of ~30, which means the model accounts for 30% of the variance in the data.

Final Model using all Training Data

For my final model I considered, I originally modeled all of the dummy variables but they ended up not contributing anything to the model. This final model eliminates several features altogether, transforms three, and considers four different interaction effects.

```
baseball_interactions <- lm(baseball_df_fix, formula = TARGET_WINS ~ (TEAM_BATTING_H * TEAM_BATTING_2B +
summary(baseball_interactions)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ (TEAM_BATTING_H * TEAM_BATTING_2B +
##   TEAM_BATTING_H * TEAM_BATTING_3B + TEAM_BATTING_H * TEAM_BATTING_HR),
##   data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -68.807  -8.816   0.569   9.585  58.270
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -7.961e+00  1.291e+01  -0.616  0.537654
## TEAM_BATTING_H    4.868e-02  9.039e-03   5.386  7.97e-08 ***
## TEAM_BATTING_2B  -2.802e-02  5.708e-02  -0.491  0.623581
```

```
## TEAM_BATTING_3B          4.678e-01  1.016e-01  4.605 4.34e-06 ***
## TEAM_BATTING_HR          2.462e-01  6.928e-02  3.553 0.000388 ***
## TEAM_BATTING_H:TEAM_BATTING_2B 2.323e-05  3.814e-05  0.609 0.542590
## TEAM_BATTING_H:TEAM_BATTING_3B -2.231e-04  6.497e-05 -3.434 0.000606 ***
## TEAM_BATTING_H:TEAM_BATTING_HR -1.108e-04  4.622e-05 -2.397 0.016596 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 14.05 on 2268 degrees of freedom
## Multiple R-squared:  0.2074, Adjusted R-squared:  0.2049
## F-statistic: 84.77 on 7 and 2268 DF,  p-value: < 2.2e-16
```

```
baseball_lm2 <- lm(baseball_df_fix, formula = TARGET_WINS ~.-TEAM_BATTING_1B+log(TEAM_FIELDING_E) + log
summary(baseball_lm2)
```

```
##
## Call:
## lm(formula = TARGET_WINS ~ . - TEAM_BATTING_1B + log(TEAM_FIELDING_E) +
##      log(TEAM_PITCHING_BB) + log(TEAM_PITCHING_SO) - TEAM_BATTING_SLG -
##      TEAM_PITCHING_H - TEAM_BATTING_BB - TEAM_BATTING_SO - TEAM_PITCHING_HR -
##      TEAM_PITCHING_BB - TEAM_FIELDING_E + log(TEAM_FIELDING_E) +
##      log(TEAM_PITCHING_SO) + TEAM_BATTING_3B:TEAM_BATTING_HR +
##      TEAM_BATTING_2B:TEAM_BATTING_HR + TEAM_BATTING_H:TEAM_BATTING_HR +
##      TEAM_BATTING_H:TEAM_BATTING_3B - TEAM_BATTING_3B - TEAM_BATTING_SO -
##      TEAM_BATTING_2B - TEAM_BATTING_BB - TEAM_BATTING_HR - TEAM_BATTING_H -
##      TEAM_BATTING_HR - TEAM_PITCHING_HR, data = baseball_df_fix)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -31.0990  -7.0521   0.1861   6.9307  27.4218
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -1.369e+01  3.690e+01  -0.371  0.71058
## TEAM_BASERUN_SB    5.819e-02  5.596e-03  10.400 < 2e-16 ***
## TEAM_PITCHING_SO   -4.162e-02  6.988e-03  -5.955 3.11e-09 ***
## TEAM_FIELDING_DP   -9.923e-02  1.223e-02  -8.116 8.79e-16 ***
## TEAM_CS_YES_NO    -3.889e+00  7.742e-01  -5.023 5.59e-07 ***
## TEAM_HBP_YES_NO   -3.183e+00  1.000e+00  -3.182  0.00149 **
## log(TEAM_FIELDING_E) -2.358e+01  1.313e+00 -17.952 < 2e-16 ***
## log(TEAM_PITCHING_BB)  1.788e+01  1.497e+00  11.946 < 2e-16 ***
## log(TEAM_PITCHING_SO)  1.829e+01  5.822e+00   3.142  0.00171 **
## TEAM_BATTING_3B:TEAM_BATTING_HR -7.815e-04  2.553e-04  -3.061  0.00224 **
## TEAM_BATTING_2B:TEAM_BATTING_HR -3.637e-04  7.112e-05  -5.114 3.48e-07 ***
## TEAM_BATTING_H:TEAM_BATTING_HR  1.526e-04  1.597e-05   9.556 < 2e-16 ***
## TEAM_BATTING_H:TEAM_BATTING_3B  1.923e-04  1.760e-05  10.930 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.07 on 1822 degrees of freedom
## (441 observations deleted due to missingness)
## Multiple R-squared:  0.4177, Adjusted R-squared:  0.4139
```

```
## F-statistic: 108.9 on 12 and 1822 DF, p-value: < 2.2e-16
```

The R-squared statistic indicates that this model predicts less than half of the variation in wins with the included features. For a next step, I hope to use cross-validation techniques to split the training data further and allow me to compare RMSE of various models.

```
#Tom - still working on this  
#set.seed(123)
```

```
#baseball_cv <-
```

```
#cv_model <- train(TARGET_WINS ~ ., baseball_df_fix, method = 'lm', trControl = trainControl(method = 'cv'))
```

Evaluation Data

I also loaded the evaluation data and predicted the wins using my final model. Since the actual wins are withheld, I compared the distribution of predictions to the actual wins in the training set. The means were similar but the training data included much more variation between teams. It's also worth mentioning as well that using the predict function creates missing values as the evaluation data is missing. In fact, for TEAM_BATTING_HBP, over 90% of rows are missing entries.

```
round(100*colSums(is.na(baseball_eval))/nrow(baseball_eval),2)
```

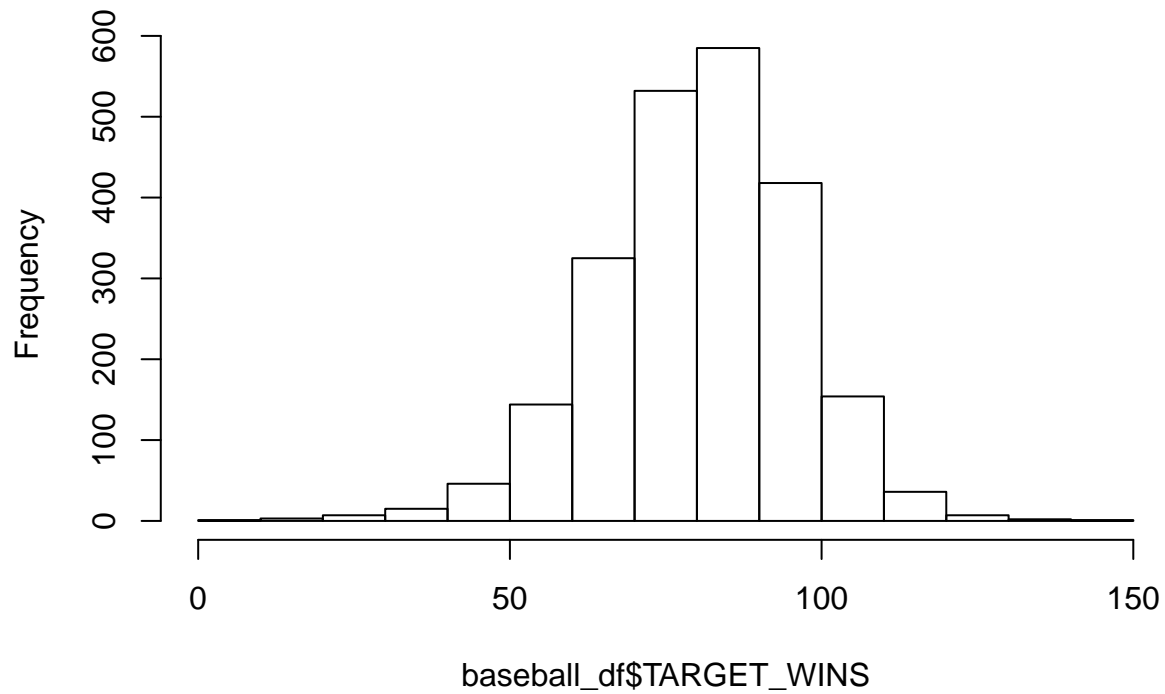
##	INDEX	TEAM_BATTING_H	TEAM_BATTING_2B	TEAM_BATTING_3B
##	0.00	0.00	0.00	0.00
##	TEAM_BATTING_HR	TEAM_BATTING_BB	TEAM_BATTING_SO	TEAM_BASERUN_SB
##	0.00	0.00	6.95	5.02
##	TEAM_BASERUN_CS	TEAM_BATTING_HBP	TEAM_PITCHING_H	TEAM_PITCHING_HR
##	33.59	92.66	0.00	0.00
##	TEAM_PITCHING_BB	TEAM_PITCHING_SO	TEAM_FIELDING_E	TEAM_FIELDING_DP
##	0.00	6.95	0.00	11.97

The prediction data also has missing values, which are approximately the same as the training data.

```
baseball_vars <- baseball_eval %>%  
  dplyr::select(TEAM_PITCHING_H, TEAM_PITCHING_HR, TEAM_FIELDING_DP, TEAM_BATTING_3B, TEAM_FIELDING_E, TEAM_PITCHING_BB, TEAM_PITCHING_SO, TEAM_FIELDING_E, TEAM_FIELDING_DP)  
  
eval_predict <- predict(baseball_interactions, newdata = baseball_eval)
```

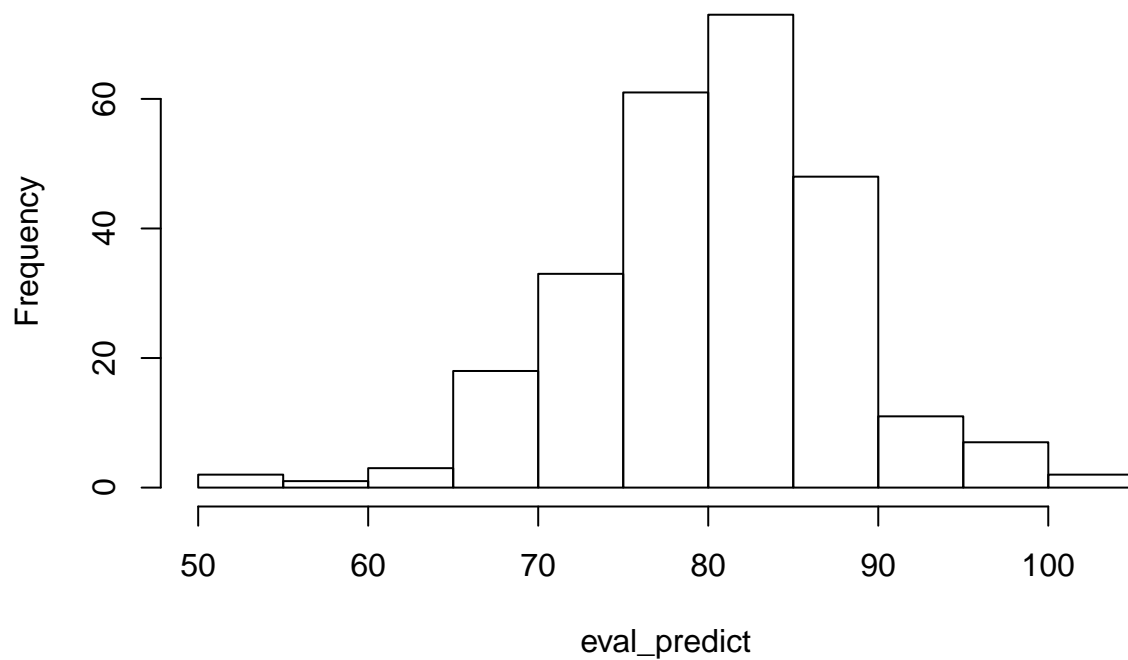
```
hist(baseball_df$TARGET_WINS)
```

Histogram of baseball_df\$TARGET_WINS



```
hist(eval_predict)
```

Histogram of eval_predict



```
summary(eval_predict)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    54.33  75.52   80.85   80.47  85.46  102.01
```

```
sd(eval_predict)
```

```
## [1] 7.711588
```

```
summary(baseball_df$TARGET_WINS)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.00  71.00   82.00   80.79  92.00  146.00
```

```
sd(baseball_df$TARGET_WINS)
```

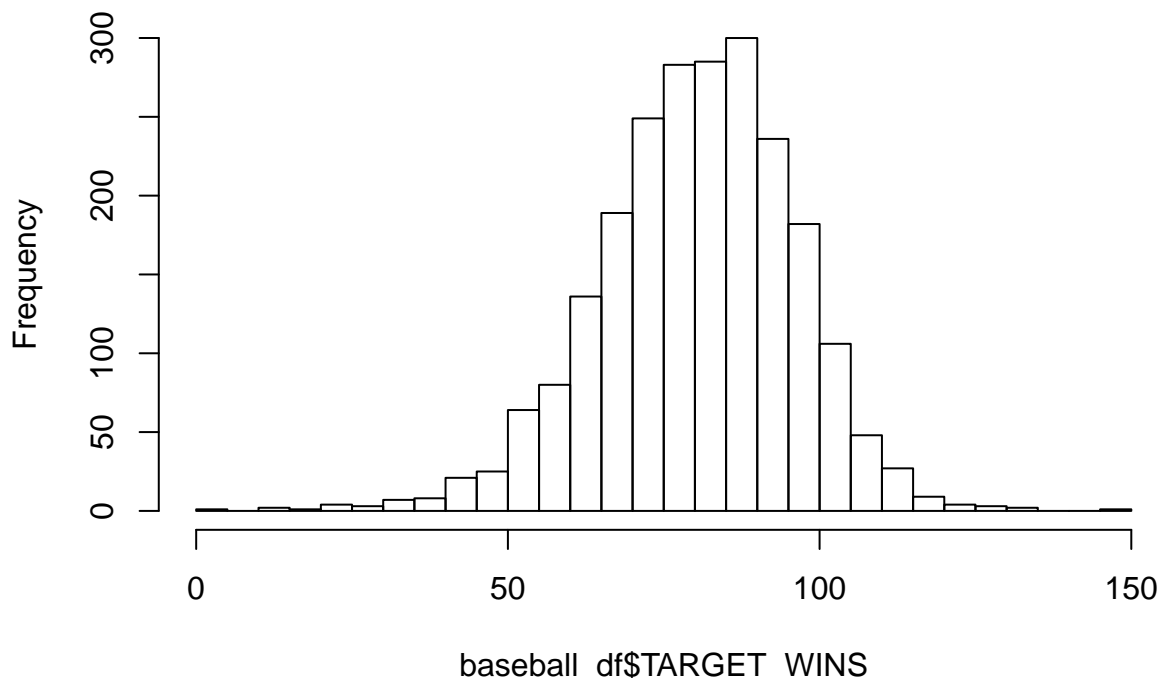
```
## [1] 15.75215
```

```
baseball_eval <- baseball_eval %>% ##added new features to eval data so predict could run
  mutate(TEAM_CS_YES_NO = case_when(!is.na(TEAM_BASERUN_CS) ~ 1, is.na(TEAM_BASERUN_CS) ~ 0)) %>%
  mutate(TEAM_HBP_YES_NO = case_when(!is.na(TEAM_BATTING_HBP) ~ 1, is.na(TEAM_BATTING_HBP) ~ 0)) %>%
  mutate(TEAM_BATTING_1B = TEAM_BATTING_H - TEAM_BATTING_2B - TEAM_BATTING_3B - TEAM_BATTING_HR) %>%
  mutate(TEAM_BATTING_SLG = (TEAM_BATTING_H + TEAM_BATTING_2B + 2 * TEAM_BATTING_3B + 3 * TEAM_BATTING_HR) / TEAM_BATTING_H)
```

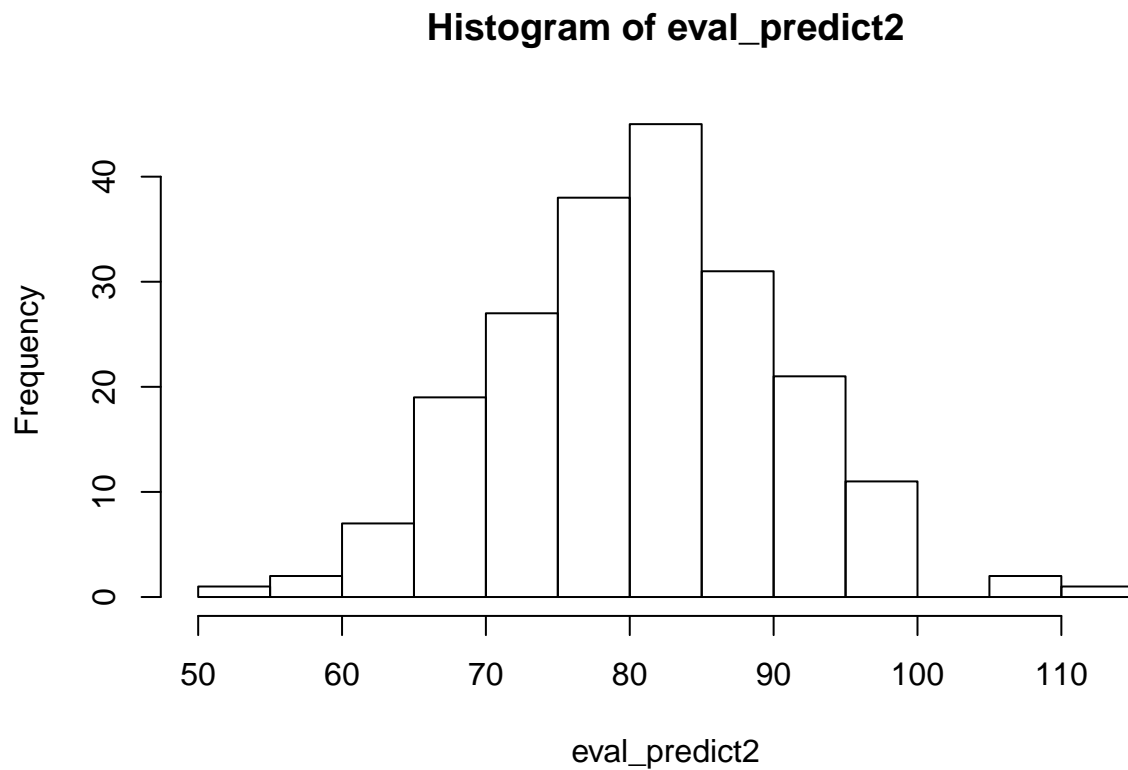
```
eval_predict2 <- predict(baseball_lm2, newdata = baseball_eval)
```

```
hist(baseball_df$TARGET_WINS, breaks = 40)
```

Histogram of baseball_df\$TARGET_WINS




```
hist(eval_predict2)
```



```
summary(eval_predict2)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.     NA's  
##   54.80   74.15   80.69   80.76   87.47  111.48      54
```

```
sd(eval_predict2, na.rm = T)
```

```
## [1] 9.677568
```

```
n_test <- nrow(baseball_eval)  
n_train <- nrow(baseball_df)
```

```
summary(baseball_df$TARGET_WINS)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
##    0.00   71.00   82.00   80.79   92.00  146.00
```

```
sd(baseball_df$TARGET_WINS)
```

```
## [1] 15.75215
```