Data 621 Business Analytics and Data Mining HW1 Xiaomeng(Vivian) Kong

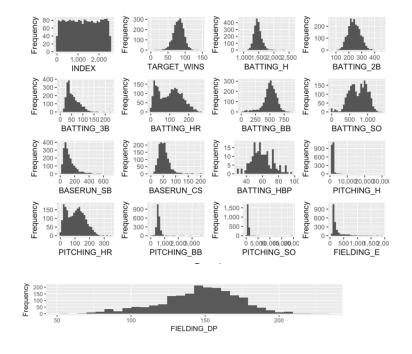
6/11/2018

1. Data Exploration

In our training data, I find 2276 rows and 16 columns. Each rows represent a professional baseball team's performance from the year 1871 to 2006 inclusive. We're interested in finding the trend of number of wins and predict them by using multiple linear regression model on the training data for the team. To find the model, we need to explore each variable first to get the better idea about our data.

• The shape of the distribution.

I used the histogram for each variables. From the histogram (Except the column of "Index"), I made a chart to categorize the shape of the distributions of all 15 variables



Shape of the distribution	Variables
Approximately normal	Target_Wins, Batting_H,Pitching_BB
Skewed Right	Batting_3B, Baserun_3B, Baserun_SB,Baserun_CS, Pitching_H, Fielding_E
Skewed left	Batting_BB, Fielding_DP

Bi-modal	Battting_2B, Batting_HR, Batting_SO, Pitching_HR
Abnormal shape	Batting_HBP, Pitching_SO

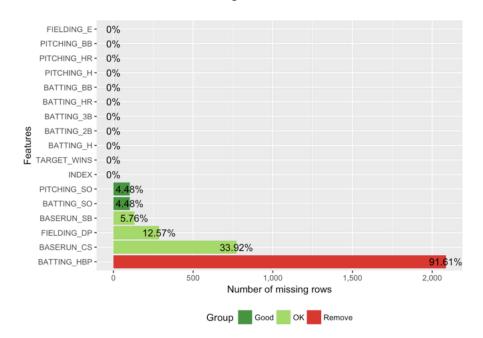
For the abnormal shape of distributions, we will suspect that something wrong with those variables. Before we move on to give the conclusion, I found the central tendency and standard deviation for these 17 variables.

• Central Tendency and Standard Deviation

	Mean	Median	Sd	NA
Target_ win	80.79	82	15.75	
Batting H	1469	1454	144.59	
2B	241.2	238	46.80	
3B	55.25	47	27.94	
HR	99.61	102	60.55	
BB	501.6	512	122.67	
So	735.6	750	248.53	102
SB	124.8	101	87.79	131
CS	52.8	49	22.96	772
HBP	59.36	58	12.97	2085
Pitching H	1779	1518	1406.8	
Pitching_HR	105.7	107	61.30	
P_BB	553	536.5	166.36	
P_so	817.7	813.5	553.09	102
Fielding_E	246.5	159	227.77	
Fielding DP	146.4	149	26.23	

• Missing values for the variables

There are 5 variables that have missing data set.



I replace the median for the missing values for these 4 variables except Batting_HBP. Batting_HBP missed 91.61% of the data values which should be removed. The reason why I chose the median for the other 4 variables because median is a better indicator for the central tendency, compare to the mean

Because the median is not affected by extreme values.

New Variables

I find out that the Batting_H=Batting_1B+ Batting_2B+ Batting_3B+Batting_HR, so in this case, I create the new variable Batting_1B= Batting_H- Batting_2B-Batting_3B-Batting_HR.

Another new variable is Total_Batting= 1*Batting_1B+2*Batting_2B+3*Batting_3B+4*Batting_HR.

Correlations

After I fixed the missing values, I can find the correlations between variables. I also add the new variable BATTING 1B and Total batting in the chart.

```
TARGET_WINS BATTING_2B BATTING_3B BATTING_HR BATTING_BB
TARGET WINS 1.00000000 0.28910365 0.142608411 0.1761532 0.23255986
BATTING_2B 0.28910365 1.00000000 -0.107305824 0.4353973 0.25572610
BATTING 3B
            0.14260841 -0.10730582 1.000000000 -0.6355669 -0.28723584
BATTING_HR 0.17615320 0.43539729 -0.635566946 1.0000000 0.51373481
BATTING BB 0.23255986 0.25572610 -0.287235841 0.5137348 1.00000000
BATTING_SO -0.03058135 0.15173438 -0.655709613 0.6930076 0.37148892
BASERUN SB 0.12361087 -0.18340432 0.485740156 -0.4068891 -0.04268402
            0.01595982 -0.04584955 0.136181182 -0.2254587 -0.04581766
BASERUN CS
PITCHING H -0.10993705 0.02369219 0.194879411 -0.2501455 -0.44977762
PITCHING HR 0.18901373 0.45455082 -0.567836679 0.9693714 0.45955207
PITCHING BB 0.12417454 0.17805420 -0.002224148 0.1369276 0.48936126
PITCHING SO -0.07579967 0.06213042 -0.254238104 0.1774182 -0.02017989
FIELDING E -0.17648476 -0.23515099 0.509778447 -0.5873391 -0.65597081
FIELDING DP -0.03008630 0.25696798 -0.227771884 0.3916524 0.32963974
BATTING 1B 0.34579395 0.33580405 0.347822719 -0.0318712 -0.12886347
Total batting 0.39892151 0.75439415 -0.136638042 0.7493183 0.36408258
            BATTING SO BASERUN SB BASERUN CS PITCHING H PITCHING HR
TARGET WINS -0.03058135 0.12361087 0.01595982 -0.10993705 0.18901373
BATTING 2B 0.15173438 -0.18340432 -0.04584955 0.02369219 0.45455082
BATTING 3B -0.65570961 0.48574016 0.13618118 0.19487941 -0.56783668
BATTING HR 0.69300765 -0.40688907 -0.22545867 -0.25014548 0.96937140
BATTING BB 0.37148892 -0.04268402 -0.04581766 -0.44977762 0.45955207
            1.00000000 -0.21178758 -0.10250193 -0.37571553 0.63286033
BATTING SO
BASERUN SB -0.21178758 1.00000000 0.23324171 0.03957227 -0.38005624
BASERUN CS -0.10250193 0.23324171 1.00000000 -0.05259183 -0.22818525
PITCHING H -0.37571553 0.03957227 -0.05259183 1.00000000 -0.14161276
PITCHING HR 0.63286033 -0.38005624 -0.22818525 -0.14161276 1.00000000
PITCHING BB 0.03498809 0.12928969 -0.04722893 0.32067616 0.22193750
PITCHING SO 0.41618159 -0.06424741 -0.05653800 0.26693587 0.19691491
FIELDING E -0.58259305 0.32615276 -0.02917821 0.66775901 -0.49314447
FIELDING DP 0.11089804 -0.27023400 -0.10200214 -0.04464784 0.38959550
BATTING_1B -0.48464372 0.09474682 -0.01375594 0.33253091 0.04579447
Total batting 0.24141254 -0.21340675 -0.16245457 -0.01596413 0.77829405
```

```
PITCHING_BB PITCHING_SO FIELDING_E FIELDING_DP
TARGET WINS 0.124174536 -0.075799674 -0.17648476 -0.030086302
BATTING_2B 0.178054204 0.062130422 -0.23515099 0.256967975
BATTING_3B -0.002224148 -0.254238104 0.50977845 -0.227771884
BATTING HR 0.136927564 0.177418187 -0.58733910 0.391652434
BATTING_BB 0.489361263 -0.020179893 -0.65597081 0.329639737
BATTING_SO 0.034988093 0.416181592 -0.58259305 0.110898035
BASERUN_SB 0.129289686 -0.064247407 0.32615276 -0.270234003
BASERUN_CS -0.047228927 -0.056538002 -0.02917821 -0.102002137
PITCHING_H 0.320676162 0.266935871 0.66775901 -0.044647837
PITCHING HR 0.221937505 0.196914911 -0.49314447 0.389595503
PITCHING_BB 1.000000000 0.482172000 -0.02283756 0.192348657
PITCHING_SO 0.482172000 1.000000000 -0.02332278 0.009552324
FIELDING E -0.022837561 -0.023322782 1.00000000 -0.227394807
FIELDING_DP 0.192348657 0.009552324 -0.22739481 1.000000000
BATTING 1B 0.047792487 -0.279280625 0.31333793 0.110655544
Total batting 0.182240384 -0.012524535 -0.28654467 0.371830404
            BATTING_1B Total_batting
TARGET_WINS 0.34579395 0.39892151
BATTING_2B 0.33580405 0.75439415
BATTING 3B 0.34782272 -0.13663804
BATTING_HR -0.03187120 0.74931833
BATTING BB -0.12886347 0.36408258
BATTING_SO -0.48464372 0.24141254
BASERUN_SB 0.09474682 -0.21340675
BASERUN CS -0.01375594 -0.16245457
PITCHING H 0.33253091 -0.01596413
PITCHING_HR 0.04579447 0.77829405
PITCHING BB 0.04779249 0.18224038
PITCHING_SO -0.27928062 -0.01252454
FIELDING_E 0.31333793 -0.28654467
FIELDING DP 0.11065554 0.37183040
BATTING 1B 1.00000000
                          0.54607257
Total batting 0.54607257 1.00000000
```

Between Target_wins and other variables, there are correlations but not strong. Most of them are showing weak positive/negative correlations. It means that when we set up the multi regression model later, we will not expect high R^2.

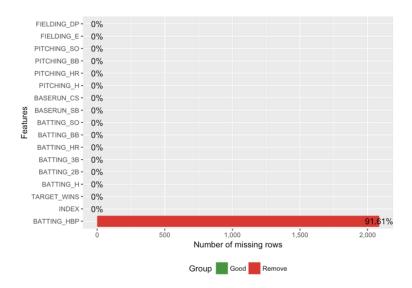
Between variables, I found that Batting_HR highly correlated to Pitching_HR. (Unfortunately, I don't really understand the Baseball. I just guess the following correlation. Because there is the same group of players batting and pitching in turns for the same game. I think that is the reason why they are very correlated.

2. Data Preparation

Fixing the missing values

As I did from part 1. Data exploration, I removed the "Batting_HBP" Since it lacks about 92% of data values so it is not necessary to fix it.

For Pitching_So, Batting_So, Baserun_SB, Baserun_CS, Fielding_DP, I used the median to replace the missing values since the median is a better indicator for the central measurement. The median is not affected by the extreme value.



• Combing variables.

I combined the Batting_H, Batting_2B, Batting_3B, Batting_HR and create the new variable: Batting_1B.

I also created another new variable:

Total_Batting=1*Batting_1B+2*Batting_2B+3*Batting_3B+4*Batting_HR

3. Building Models

1) Model 1

Backward Selections. For the first model, I will use all variables in the models, and then decide which one variable I will eliminate.

```
lm(formula = TARGET_WINS ~ ., data = moneyball)
Residuals:
   Min
             10 Median
                              30
                                     Max
-49.753 -8.626
                 0.120
                           8.395 58.561
Coefficients: (1 not defined because of singularities)
                Estimate Std. Error t value Pr(>|t|)
              23.6421579 5.3902272 4.386 1.21e-05 ***
(Intercept)
BATTING_2B
               0.0279578 0.0073363
                                      3.811 0.000142 ***
BATTING_3B
               0.1133940 0.0159335
                                      7.117 1.48e-12 ***
BATTING HR
               0.0527325 0.0274915
                                       1.918 0.055219
BATTING BB
               0.0104483 0.0058377
                                       1.790 0.073621 .
BATTING SO
              BASERUN SB
               0.0254236 0.0043565
                                      5.836 6.12e-09 ***
              -0.0110027 0.0157842 -0.697 0.485829
BASERUN CS
              -0.0008456 0.0003674 -2.302 0.021444 *
PITCHING_H
PITCHING HR
              0.0129626 0.0243894
                                      0.531 0.595135
PITCHING BB
              0.0007798 0.0041571
                                      0.188 0.851231
PITCHING_SO
              0.0028156 0.0009219
                                      3.054 0.002284 **
FIELDING E
              -0.0195325 0.0024609
                                      -7.937 3.23e-15 ***
FIELDING DP
              -0.1217801
                           0.0129421
                                      -9.410 < 2e-16 ***
                                               < 2e-16 ***
BATTING 1B
               0.0489152 0.0036949
                                      13.239
Total_batting
                      NA
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 13.07 on 2261 degrees of freedom
Multiple R-squared: 0.3154, Adjusted R-squared: 0.3111
F-statistic: 74.4 on 14 and 2261 DF, p-value: < 2.2e-16
                                                        Normal Q-Q
 9
 40
 20
 -20
 49
 09
              Fitted values
Im(TARGET_WINS ~ .)
                                                     Theoretical Quantiles
Im(TARGET_WINS ~ .)
  S,
  1.0
  9.5
                                                 0.1
                                                         0.3
                                                             0.4
                                                                 0.5
               Fitted values
Im(TARGET_WINS ~ .)
                                                       Leverage
Im(TARGET_WINS ~ .)
```

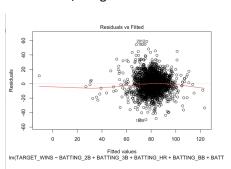
From this model, we get R^2 is 0.3154 which means that 31.54% of the total variance of Target_wins can be explained by the total variance of all variables. It is not quite a satisfying number. F statistics tell us if the model best fits the population from which the data are sampled. For our case, we get p-value is approximately 0 and it is statistically significant. It indicates that our model fits the population from the sample we took. However, in this case, we have "NA" in the Total_batting which means it may strongly correlated to some other independent variables.

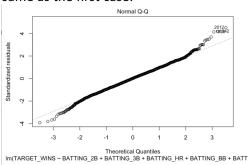
2) Model 2

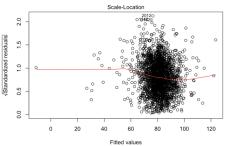
I got rid of "Total batting" and take a look at the linear model again.

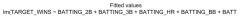
```
lm(formula = TARGET_WINS ~ BATTING_2B + BATTING_3B + BATTING_HR +
    BATTING_BB + BATTING_SO + BASERUN_SB + BASERUN_CS + PITCHING_H +
    PITCHING HR + PITCHING BB + PITCHING SO + FIELDING E + FIELDING DP +
    BATTING_1B, data = moneyball)
Residuals:
    Min
            10 Median
                            30
                                  Max
-49.753 -8.626
                0.120
                         8.395
                               58.561
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept) 23.6421579 5.3902272
                                 4.386 1.21e-05 ***
                                  3.811 0.000142 ***
BATTING 2B
            0.0279578 0.0073363
                                  7.117 1.48e-12 ***
BATTING 3B
            0.1133940 0.0159335
BATTING_HR
            0.0527325 0.0274915
                                  1.918 0.055219 .
BATTING BB
            0.0104483 0.0058377
                                  1.790 0.073621 .
                                 -3.312 0.000941 ***
BATTING SO -0.0084323 0.0025461
BASERUN_SB 0.0254236 0.0043565
                                  5.836 6.12e-09 ***
BASERUN_CS -0.0110027 0.0157842
                                 -0.697 0.485829
PITCHING_H -0.0008456 0.0003674 -2.302 0.021444 *
PITCHING_HR 0.0129626 0.0243894
                                  0.531 0.595135
PITCHING_BB 0.0007798 0.0041571
                                  0.188 0.851231
PITCHING SO 0.0028156 0.0009219
                                  3.054 0.002284 **
FIELDING_E -0.0195325 0.0024609 -7.937 3.23e-15 ***
FIELDING_DP -0.1217801 0.0129421 -9.410 < 2e-16 ***
BATTING_1B 0.0489152 0.0036949 13.239 < 2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 13.07 on 2261 degrees of freedom
Multiple R-squared: 0.3154, Adjusted R-squared: 0.3111
F-statistic: 74.4 on 14 and 2261 DF, p-value: < 2.2e-16
```

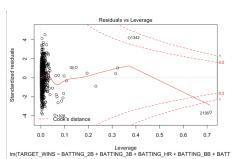
In this case, we get the R^2 which is the same as the first case.









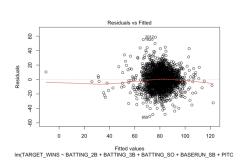


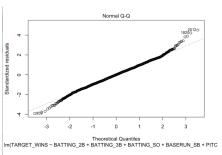
3) Model 3

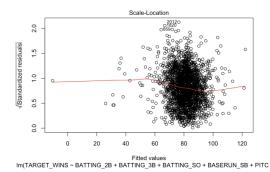
I got rid of the variables that are not statistically significant. In this model, I got rid of "Batting_HR, Batting_BB, Baserun_CS, Pitching_HR, and Pitching_BB.

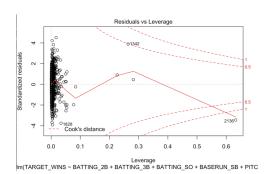
```
lm(formula = TARGET_WINS ~ BATTING_2B + BATTING_3B + BATTING_SO +
   BASERUN SB + PITCHING H + PITCHING SO + FIELDING E + FIELDING DP +
   BATTING 1B + Total batting, data = moneyball)
Residuals:
   Min
            10
                Median
                            30
-51.028 -8.645
                 0.079
                         8.538 58.554
Coefficients:
               Estimate Std. Error t value Pr(>|t|)
(Intercept)
             28.5417286 4.9219496
                                     5.799 7.61e-09 ***
BATTING_2B
             -0.0094319
                         0.0094685
                                    -0.996 0.319289
BATTING_3B
              0.0605133
                         0.0156578
                                     3.865 0.000114 ***
BATTING SO
             -0.0099177
                         0.0024244
                                    -4.091 4.45e-05 ***
                                     7.542 6.64e-14 ***
              0.0298779
                         0.0039613
BASERUN SB
PITCHING_H
              -0.0008022
                         0.0003216
                                   -2.494 0.012699 *
PITCHING_SO
              0.0029552
                         0.0006735
                                     4.388 1.20e-05 ***
                        0.0021496 -10.530 < 2e-16 ***
0.0126728 -8.879 < 2e-16 ***
FIELDING_E
              -0.0226345
             -0.1125269
FIELDING DP
BATTING 1B
              0.0284034
                        0.0050326
                                     5.644 1.87e-08 ***
8.290 < 2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 13.1 on 2265 degrees of freedom
Multiple R-squared: 0.3115, Adjusted R-squared: 0.3085
```

F-statistic: 102.5 on 10 and 2265 DF, p-value: < 2.2e-16









4) Model 4

Keep getting rid of the relatively statistically insignificant variable. Batting_2B and Piting_H is being got rid of.

```
lm(formula = TARGET_WINS ~ BATTING_3B + BATTING_SO + BASERUN_SB +
   PITCHING_SO + FIELDING_E + FIELDING_DP + BATTING_1B + Total_batting,
   data = moneyball)
```

Residuals:

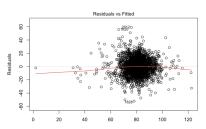
Min 1Q Median 3Q -52.140 -8.690 0.037 8.443 59.016

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept)
            27.996811 4.798551 5.834 6.17e-09 ***
                        0.015282 4.578 4.94e-06 ***
BATTING 3B
             0.069968
BATTING_SO
             -0.007831
                        0.002273 -3.446 0.000580 ***
BASERUN SB
                                  7.851 6.33e-15 ***
              0.030903
                        0.003936
PITCHING_SO
              0.002079
                        0.000591
                                  3.517 0.000444 ***
FIELDING_E
             -0.025823
                        0.001712 -15.081 < 2e-16 ***
FIELDING_DP
             -0.111019
                        0.012628 -8.792 < 2e-16 ***
BATTING 1B
             0.028995
                        0.004849
                                  5.979 2.60e-09 ***
                       0.001734 9.830 < 2e-16 ***
Total_batting 0.017044
```

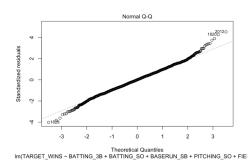
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 13.11 on 2267 degrees of freedom Multiple R-squared: 0.3093, Adjusted R-squared: 0.3068 F-statistic: 126.9 on 8 and 2267 DF, p-value: < 2.2e-16



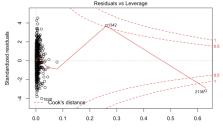
 $\label{eq:fitted values} Fitted \ values \\ Im(TARGET_WINS \sim BATTING_3B + BATTING_SO + BASERUN_SB + PITCHING_SO + FIELD + FIE$

2.0 5. 1.0 9.0





 $\label{eq:fitted} Fitted \ values $$ Im(TARGET_WINS \sim BATTING_3B + BATTING_SO + BASERUN_SB + PITCHING_SO + FIELD | FIRST | F$



 $\label{lower} Leverage $$ Im(TARGET_WINS \sim BATTING_3B + BATTING_SO + BASERUN_SB + PITCHING_SO + FIELD + FIEL$

4. Select Models

Assume that the model with the highest adjusted R^2 is the "best" model especially for this case. Since all the independent variables are not strongly correlated to the Target_wins. Our Adjusted R^2 are 31.11%, 31.11%, 30.85%, 30.68%. Based on our assumption. The first model with all independent variables and the second model without "Total_batting" are the best models.

All the models, F_test gives us the result that all independent variables have correlation with the Target_wins in a strong statistically significant level since the P-value is approximately 0.

In the meanwhile, for all the models, the residual plots seem to be reasonable but not the satisfying results. The plots show that our models can predict Target_wins but the models do show the significant amount of residuals when we compare the expected value and real values. It means that the model still need a lot of improvement. It requires more variables adjustment. (Unfortunately, I have no idea about the baseball, so a lot of technical terms I felt confused with, even with a lot of "Wikipedia" learning.) It also requires that more data collections because we have a lot of missing values and values probably outdated as well.

For selecting models, I am going to use the highest adjusted R^2 although I felt the worse model that makes sense. The "Worse" model may be not that "worse", it just needs more data collections and more manipulation for the variables.

Prediction

I chose the first model to make the prediction. In order to make the prediction, I fixed my "Evaluation" data set to fit my model, I added Batting_Base 1 and Total_batting, removed the Batting_HBP, and replaced the median to the missing values. The following pic is just the first a few rows of the result between the predicted value and actual value.

pred	V2
64.02285	39
65.73235	70
75.27890	86
85.74341	70
66.37855	82
69.83817	75

Appendix

Github: https://github.com/xkong100/data-621/blob/master/Hw1/XKong_Assignment1_data621.Rmd