

# DATA621 HW 1

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**Overview** In this homework assignment, you will explore, analyze and model a data set containing approximately 2200 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. Your objective is to build a multiple linear regression model on the training data to predict the number of wins for the team.

## 1. DATA EXPLORATION

**Describe the size and the variables in the moneyball training data set. Consider that too much detail will cause a manager to lose interest while too little detail will make the manager consider that you aren't doing your job.**

Looking at the data provided, there are a total of 17 variables with 2276 records relevant to professional baseball teams. The 17 variables are each defined and evaluated based on impact on wins.

##	DEFINITION	THEORETICAL EFFECT
## INDEX	Indentification Variable (do not use)	None
## TARGET_WINS	Number of wins	
## TEAM_BATTING_H	Base Hits by batters (1B,2B,3B,HR)	Positive Impact on Wins
## TEAM_BATTING_2B	Doubles by batters (2B)	Positive Impact on Wins
## TEAM_BATTING_3B	Triples by batters (3B)	Positive Impact on Wins
## TEAM_BATTING_4B	Homeruns by batters (4B)	Positive Impact on Wins
## TEAM_BATTING_BB	Walks by batters	Positive Impact on Wins
## TEAM_BATTING_HBP	Batters hit by pitch (get a free base)	Positive Impact on Wins
## TEAM_BATTING_SO	Strikeouts by batters	Negative Impact on Wins
## TEAM_BASERUN_SB	Stolen bases	Positive Impact on Wins
## TEAM_BASERUN_CS	Caught stealing	Negative Impact on Wins
## TEAM_FIELDING_E	Errors	Negative Impact on Wins
## TEAM_FIELDING_DP	Double Plays	Positive Impact on Wins
## TEAM_PITCHING_BB	Walks allowed	Negative Impact on Wins
## TEAM_PITCHING_H	Hits allowed	Negative Impact on Wins
## TEAM_PITCHING_HR	Homeruns allowed	Negative Impact on Wins
## TEAM_PITCHING_SO	Strikeouts by pitchers	Positive Impact on Wins

We can see the minimum value, 1st and 3rd quantile, median value, average value (mean), and the maximum value for each variable.

##	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

```

The missing values are within the following variables and need to be addressed to make a predictive model:

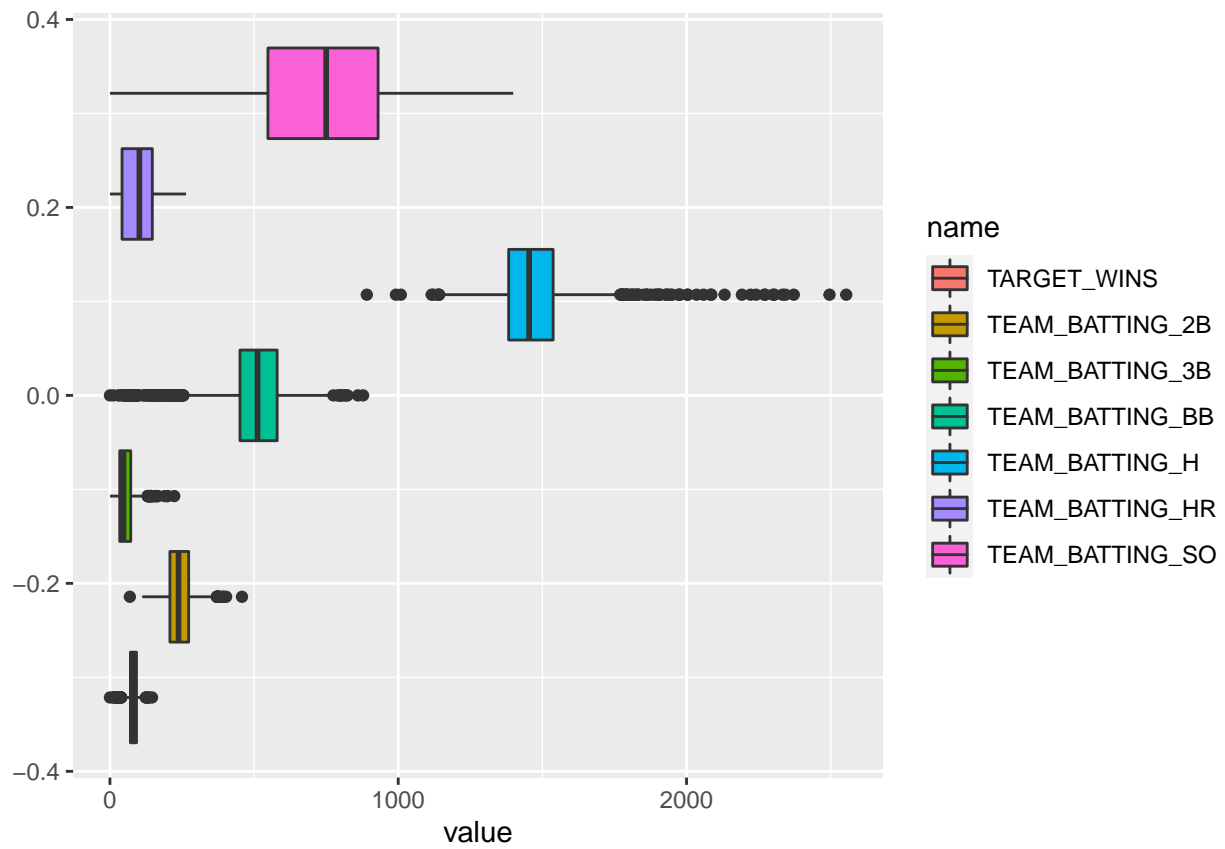
```

## INDEX TARGET_WINS TEAM_BATTING_H TEAM_BATTING_2B
## 0 0 0 0
## TEAM_BATTING_3B TEAM_BATTING_HR TEAM_BATTING_BB TEAM_BATTING_SO
## 0 0 0 102
## TEAM_BASERUN_SB TEAM_BASERUN_CS TEAM_BATTING_HBP TEAM_PITCHING_H
## 131 772 2085 0
## TEAM_PITCHING_HR TEAM_PITCHING_BB TEAM_PITCHING_SO TEAM_FIELDING_E
## 0 0 102 0
## TEAM_FIELDING_DP
## 286

```

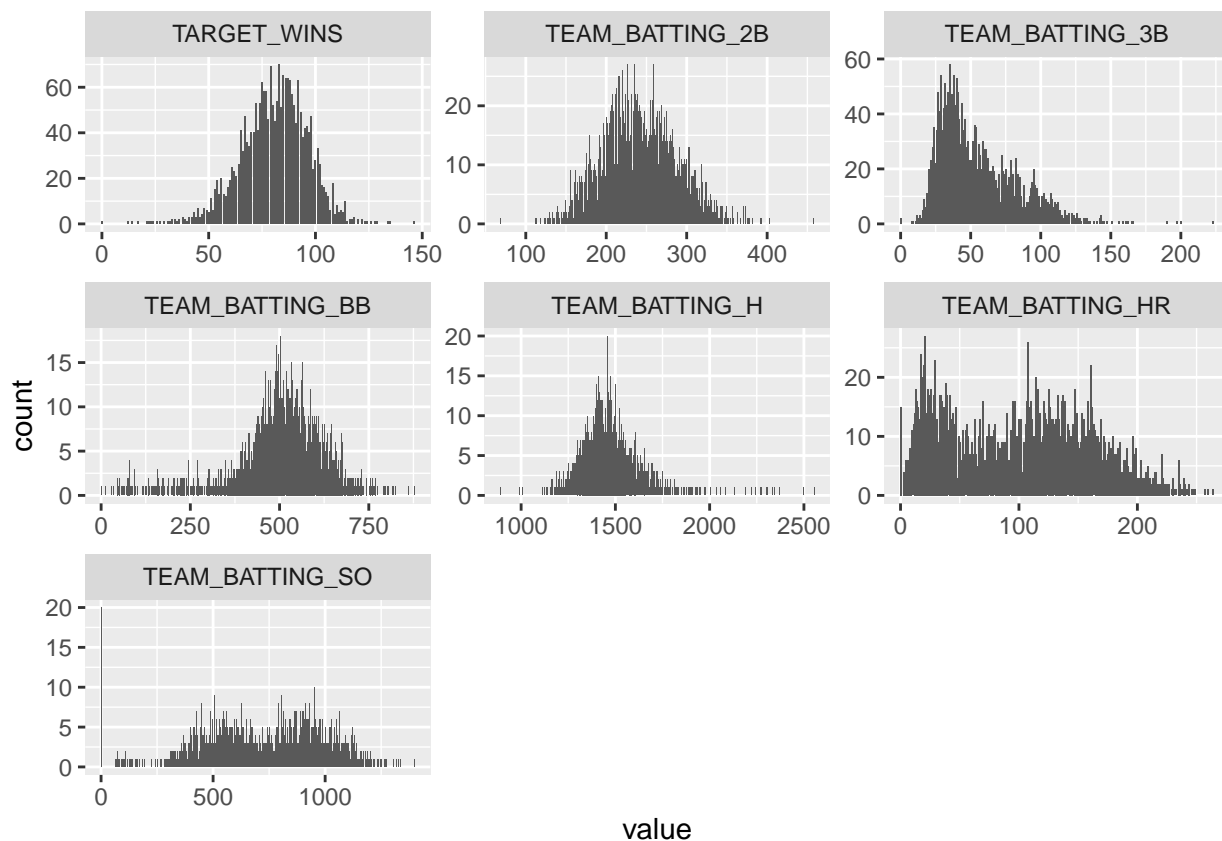
Here are boxplots of the variables in the data set. As we can see, the median, upper quartile, lower quartile, upper whisker, lower whisker, and outliers can be determined based on the plots.

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



Additionally, we can use a barplot to determine the count of each value for each variable.

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



Let's determine the correlation of our target variable with each remaining variable, where values range from -1 (negative linear correlation) and 1 (positive linear correlation).

```
##                                [,1]
## INDEX                        -0.02105643
## TEAM_BATTING_H                0.38876752
## TEAM_BATTING_2B              0.28910365
## TEAM_BATTING_3B              0.14260841
## TEAM_BATTING_HR              0.17615320
## TEAM_BATTING_BB              0.23255986
## TEAM_BATTING_SO              NA
## TEAM_BASERUN_SB              NA
## TEAM_BASERUN_CS              NA
## TEAM_BATTING_HBP              NA
## TEAM_PITCHING_H              -0.10993705
## TEAM_PITCHING_HR             0.18901373
## TEAM_PITCHING_BB             0.12417454
## TEAM_PITCHING_SO             NA
## TEAM_FIELDING_E              -0.17648476
## TEAM_FIELDING_DP             NA
```

## 2. Data Preparation

‘Describe how you have transformed the data by changing the original variables or creating new variables. If you did transform the data or create new variables, discuss why you did this. Here are some possible transformations.

First, we need to address the missing values. From what we can recall, TEAM\_BATTING\_HBP have over 90% of missing values and should not be included in the model. Additionally, the INDEX variable has no relevance to the model and therefore will be removed as well. TEAM\_BASERUN\_CS is highly correlated with TEAM\_BASERUN\_SB and has a large amount of missing values. I will remove this variable from the model. In baseball, stolen bases can be derived from the batting and/or pitching rates. Therefore TEAM\_BASERUN\_SB can be removed from the model. The remaining variables (TEAM\_BATTING\_SO, TEAM\_PITCHING\_SO, TEAM\_FIELDING\_E, and TEAM\_FIELDING\_DP) will have their missing values replaced with the median values. This is, in my opinion, the best course of action because having a decimal value for each variable when they should be whole numbers does not make sense and will show in the model.

## 3. Build Models

Using the training data set, build at least three different multiple linear regression models, using different variables (or the same variables with different transformations). Since we have not yet covered automated variable selection methods, you should select the variables manually (unless you previously learned Forward or Stepwise selection, etc.). Since you manually selected a variable for inclusion into the model or exclusion into the model, indicate why this was done.

The first multiple linear regression model is based on only batting variables.

```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_2B +
##     TEAM_BATTING_3B + TEAM_BATTING_HR + TEAM_BATTING_BB + TEAM_BATTING_SO,
##     data = prep_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -64.644  -8.787   0.454   9.020  54.933
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -8.499099   5.023839  -1.692 0.090830 .
## TEAM_BATTING_H   0.044182   0.003702  11.933 < 2e-16 ***
## TEAM_BATTING_2B -0.015660   0.009321  -1.680 0.093104 .
## TEAM_BATTING_3B  0.099801   0.016369   6.097 1.27e-09 ***
## TEAM_BATTING_HR  0.031772   0.009378   3.388 0.000716 ***
## TEAM_BATTING_BB  0.028425   0.002805  10.135 < 2e-16 ***
## TEAM_BATTING_SO  0.007086   0.002184   3.244 0.001195 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 13.76 on 2269 degrees of freedom
## Multiple R-squared:  0.2391, Adjusted R-squared:  0.2371
## F-statistic: 118.8 on 6 and 2269 DF,  p-value: < 2.2e-16
```

Since TEAM\_BATTING\_2B has a p-value greater than 0.05, I will remove it from the model. This is a better model as all coefficients are positive, which means there is a positive correlation in relation to winning. Unfortunately, I was not expecting Batting Strike Outs being positively correlated to winnings, which does not make much sense.

```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_BATTING_H + TEAM_BATTING_3B +
##     TEAM_BATTING_HR + TEAM_BATTING_BB + TEAM_BATTING_SO, data = prep_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -65.329  -8.805   0.471   8.973  52.511
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -5.530298   4.704651  -1.175 0.239920
## TEAM_BATTING_H    0.040190   0.002840  14.149 < 2e-16 ***
## TEAM_BATTING_3B    0.104229   0.016162   6.449 1.37e-10 ***
## TEAM_BATTING_HR    0.031142   0.009374   3.322 0.000908 ***
## TEAM_BATTING_BB    0.027709   0.002773   9.992 < 2e-16 ***
## TEAM_BATTING_SO    0.006129   0.002109   2.906 0.003700 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 13.76 on 2270 degrees of freedom
## Multiple R-squared:  0.2382, Adjusted R-squared:  0.2365
## F-statistic: 141.9 on 5 and 2270 DF,  p-value: < 2.2e-16
```

The next model is made only off the pitching variables.

```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_PITCHING_H + TEAM_PITCHING_HR +
##     TEAM_PITCHING_BB + TEAM_PITCHING_SO, data = prep_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -62.284  -9.842   0.483   9.679  74.699
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   71.8253764   1.1577568  62.038 < 2e-16 ***
## TEAM_PITCHING_H  -0.0011826   0.0002476  -4.776 1.90e-06 ***
## TEAM_PITCHING_HR   0.0419494   0.0055003   7.627 3.52e-14 ***
## TEAM_PITCHING_BB   0.0197724   0.0022794   8.674 < 2e-16 ***
## TEAM_PITCHING_SO  -0.0052582   0.0006818  -7.712 1.84e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15.11 on 2271 degrees of freedom
## Multiple R-squared:  0.08207, Adjusted R-squared:  0.08046
## F-statistic: 50.76 on 4 and 2271 DF,  p-value: < 2.2e-16
```

From the model, it seems pitching has little to no correlation to winning the game as the coefficients are close to 0, whether negative or positive.

The final model is based on fielding only.

```
##
## Call:
## lm(formula = TARGET_WINS ~ TEAM_FIELDING_E + TEAM_FIELDING_DP,
##     data = prep_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -60.653  -9.992   0.632  10.038  74.737
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    91.060363    2.127771  42.796 < 2e-16 ***
## TEAM_FIELDING_E -0.013370    0.001462  -9.143 < 2e-16 ***
## TEAM_FIELDING_DP -0.047535    0.013574  -3.502 0.000471 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15.47 on 2273 degrees of freedom
## Multiple R-squared:  0.03635,    Adjusted R-squared:  0.0355
## F-statistic: 42.87 on 2 and 2273 DF,  p-value: < 2.2e-16
```

The fielding error variable has a negative correlation to winning the game, which makes sense. However, the fielding double play should have had a positive correlation, though it is close to 0.

## Select Model

**Decide on the criteria for selecting the best multiple linear regression model. Will you select a model with slightly worse performance if it makes more sense or is more parsimonious? Discuss why you selected your model.**

All three models have similar characteristics that would allow them to be implemented. The MSE and residual plots were fairly similar across the models. I have decided, however, to choose the batting multiple regression model because the F Stat and R-squared scores were significantly higher than the others. The F Stat explains the variability more than the other models and the R-squared explains better model fitting.

## Implement Prediction to Evaluation Data

```
##           fit      lwr      upr
## 1    68.08942  66.64072  69.53811
## 2    69.29685  68.11673  70.47696
## 3    75.55903  74.59663  76.52142
## 4    83.36517  82.32353  84.40681
## 5    64.96974  62.94783  66.99165
## 6    66.08535  64.34267  67.82803
## 7    77.79507  76.53715  79.05300
## 8    69.13533  67.87089  70.39977
## 9    71.05055  69.69488  72.40622
```

## 10	72.90298	71.85225	73.95370
## 11	75.18580	74.02157	76.35003
## 12	82.34208	80.86427	83.81989
## 13	82.03545	80.20805	83.86284
## 14	79.02673	77.46249	80.59097
## 15	75.37674	74.01718	76.73629
## 16	76.59812	75.34915	77.84710
## 17	72.72117	71.69116	73.75118
## 18	82.39601	81.41707	83.37495
## 19	NA	NA	NA
## 20	91.54848	90.14901	92.94795
## 21	84.38659	83.04238	85.73080
## 22	86.54548	85.06329	88.02767
## 23	85.20318	84.08788	86.31847
## 24	75.15469	73.99257	76.31680
## 25	80.10412	78.94021	81.26803
## 26	82.71288	81.48628	83.93948
## 27	60.02618	56.75330	63.29906
## 28	74.42073	73.39384	75.44762
## 29	85.21959	83.62002	86.81916
## 30	75.46200	73.96388	76.96011
## 31	92.25479	90.84458	93.66499
## 32	86.95118	85.80719	88.09517
## 33	88.79667	87.49142	90.10191
## 34	91.83855	90.37263	93.30448
## 35	83.33418	82.42539	84.24297
## 36	83.68520	82.32668	85.04372
## 37	77.03349	76.23771	77.82927
## 38	90.00488	88.36854	91.64122
## 39	85.29554	84.08406	86.50701
## 40	88.03578	86.86580	89.20577
## 41	82.29417	81.06071	83.52763
## 42	87.17920	86.08031	88.27809
## 43	44.12510	39.93466	48.31554
## 44	98.88584	96.58767	101.18401
## 45	85.89691	84.92403	86.86980
## 46	93.22319	91.54780	94.89858
## 47	95.04292	93.61661	96.46922
## 48	72.19682	71.11182	73.28181
## 49	70.26227	69.07241	71.45213
## 50	75.14097	74.08225	76.19968
## 51	77.06988	76.06775	78.07200
## 52	83.53489	82.02354	85.04624
## 53	80.30853	79.21703	81.40002
## 54	72.74436	71.53790	73.95083
## 55	76.79893	75.92452	77.67334
## 56	75.60789	74.79366	76.42212
## 57	87.21737	85.82777	88.60697
## 58	68.52212	67.25246	69.79179
## 59	NA	NA	NA
## 60	NA	NA	NA
## 61	82.49638	81.45479	83.53797
## 62	84.02306	82.26443	85.78168
## 63	84.34261	83.42354	85.26169



## 64	83.64790	81.92827	85.36754
## 65	79.56597	78.12004	81.01189
## 66	86.02321	84.76045	87.28597
## 67	76.69226	75.65744	77.72708
## 68	81.53356	80.31551	82.75160
## 69	NA	NA	NA
## 70	86.21488	84.46725	87.96250
## 71	89.29295	87.64841	90.93749
## 72	74.60917	73.49096	75.72737
## 73	82.57260	81.30381	83.84140
## 74	84.95533	82.99559	86.91506
## 75	80.99296	79.41214	82.57378
## 76	86.30953	85.01612	87.60295
## 77	82.51679	81.55144	83.48213
## 78	79.10895	78.12165	80.09625
## 79	NA	NA	NA
## 80	NA	NA	NA
## 81	86.40439	85.06827	87.74051
## 82	90.19683	89.18432	91.20934
## 83	97.97194	96.44221	99.50166
## 84	82.70965	81.78967	83.62962
## 85	86.99759	85.92605	88.06914
## 86	77.31055	75.87592	78.74518
## 87	75.73002	74.64869	76.81136
## 88	81.12983	80.42372	81.83595
## 89	81.66588	80.32692	83.00485
## 90	89.21333	88.02150	90.40516
## 91	77.62391	76.57258	78.67525
## 92	94.50043	92.43066	96.57020
## 93	72.47554	71.09452	73.85657
## 94	NA	NA	NA
## 95	NA	NA	NA
## 96	NA	NA	NA
## 97	86.76518	84.86046	88.66990
## 98	100.27256	98.43955	102.10557
## 99	91.12815	89.63324	92.62306
## 100	92.39052	90.89663	93.88442
## 101	84.64627	83.82141	85.47112
## 102	74.42346	73.04834	75.79859
## 103	83.97725	83.11662	84.83788
## 104	81.59866	80.40856	82.78876
## 105	82.70115	81.24141	84.16089
## 106	76.78864	75.25292	78.32437
## 107	65.38717	63.41417	67.36017
## 108	81.52514	80.18784	82.86244
## 109	85.11409	84.09137	86.13680
## 110	68.92951	67.57369	70.28534
## 111	82.21495	81.26315	83.16674
## 112	80.45662	79.77893	81.13431
## 113	88.25270	87.34528	89.16012
## 114	85.94376	84.98162	86.90589
## 115	79.47902	78.57030	80.38773
## 116	80.74175	79.85312	81.63038
## 117	89.34286	88.24729	90.43843

## 118	80.24073	79.49758	80.98387
## 119	77.93898	76.79564	79.08232
## 120	72.97312	71.80593	74.14031
## 121	85.86022	84.43120	87.28923
## 122	NA	NA	NA
## 123	NA	NA	NA
## 124	NA	NA	NA
## 125	70.15680	68.72547	71.58812
## 126	82.94453	81.66910	84.21996
## 127	89.56987	88.39985	90.73988
## 128	73.79284	72.71109	74.87458
## 129	88.40188	87.33274	89.47102
## 130	93.92047	92.65171	95.18923
## 131	87.97776	86.77158	89.18394
## 132	79.16626	77.63084	80.70168
## 133	74.32412	73.20772	75.44052
## 134	83.61427	82.30577	84.92278
## 135	83.73923	82.49900	84.97946
## 136	69.78458	68.27238	71.29677
## 137	76.39372	75.50268	77.28476
## 138	75.92216	75.00281	76.84151
## 139	78.96605	77.86885	80.06325
## 140	79.37017	78.40316	80.33719
## 141	65.02154	63.42657	66.61652
## 142	NA	NA	NA
## 143	93.40640	92.01065	94.80215
## 144	81.02677	80.23983	81.81371
## 145	75.95016	74.64104	77.25929
## 146	76.21359	75.30355	77.12362
## 147	80.91402	80.07156	81.75647
## 148	82.01046	80.99287	83.02806
## 149	83.62807	82.86302	84.39312
## 150	80.21488	79.40566	81.02411
## 151	82.62385	81.31041	83.93729
## 152	79.98538	78.67536	81.29539
## 153	60.31141	56.98412	63.63871
## 154	71.54299	70.31605	72.76993
## 155	76.43235	75.32176	77.54295
## 156	71.87775	70.71345	73.04205
## 157	85.73202	84.45313	87.01091
## 158	72.94072	71.66300	74.21844
## 159	90.69042	88.87673	92.50411
## 160	NA	NA	NA
## 161	105.10975	102.83732	107.38219
## 162	104.70676	102.67172	106.74180
## 163	91.14299	89.81007	92.47591
## 164	105.37722	103.25832	107.49613
## 165	98.49251	96.60207	100.38296
## 166	91.05878	89.30814	92.80943
## 167	85.80615	84.76832	86.84398
## 168	80.12362	78.68624	81.56101
## 169	72.28575	71.08617	73.48533
## 170	80.56660	79.55834	81.57486
## 171	NA	NA	NA

## 172	83.70763	82.57633	84.83892
## 173	81.46669	80.41953	82.51385
## 174	90.07446	88.60931	91.53961
## 175	83.79660	82.94920	84.64400
## 176	78.86884	77.59812	80.13956
## 177	79.94570	78.26951	81.62188
## 178	77.33317	76.58710	78.07925
## 179	76.69705	75.89659	77.49751
## 180	81.20086	80.30876	82.09295
## 181	76.35058	75.39064	77.31053
## 182	84.56267	83.41446	85.71089
## 183	82.60796	81.64134	83.57459
## 184	85.10092	83.94347	86.25838
## 185	99.04875	96.48779	101.60970
## 186	87.00154	85.75493	88.24815
## 187	91.58599	90.00473	93.16725
## 188	69.98346	68.53005	71.43686
## 189	66.12218	64.82235	67.42200
## 190	106.76010	104.13804	109.38216
## 191	NA	NA	NA
## 192	NA	NA	NA
## 193	73.20437	72.01471	74.39403
## 194	77.08226	75.90128	78.26325
## 195	80.55491	79.02408	82.08574
## 196	69.54963	68.04272	71.05654
## 197	76.21057	75.31578	77.10536
## 198	82.65042	81.21464	84.08620
## 199	80.72763	79.61255	81.84271
## 200	87.19532	86.24386	88.14677
## 201	80.83480	79.56419	82.10541
## 202	81.48662	80.56821	82.40503
## 203	77.26073	75.91779	78.60366
## 204	82.24614	81.25423	83.23804
## 205	76.62681	75.69447	77.55915
## 206	80.67658	79.71049	81.64268
## 207	81.49961	80.21715	82.78206
## 208	78.30078	77.24688	79.35468
## 209	81.48989	80.63939	82.34038
## 210	77.69007	76.57125	78.80889
## 211	102.09699	99.64678	104.54720
## 212	91.88123	90.13394	93.62852
## 213	83.68423	81.98636	85.38210
## 214	70.41563	69.27764	71.55362
## 215	75.31877	74.17984	76.45771
## 216	86.65858	85.87931	87.43785
## 217	84.62488	83.49822	85.75155
## 218	85.36282	84.43411	86.29152
## 219	75.00080	74.09362	75.90798
## 220	78.13146	77.23682	79.02610
## 221	80.82165	79.63426	82.00905
## 222	74.90177	73.52816	76.27538
## 223	85.41220	84.36683	86.45757
## 224	78.90019	77.63416	80.16623
## 225	93.08378	89.28187	96.88568

##	226	75.76272	74.88305	76.64240
##	227	78.29652	77.37794	79.21509
##	228	83.89904	82.59321	85.20487
##	229	82.10592	81.20322	83.00862
##	230	81.15634	79.74136	82.57132
##	231	NA	NA	NA
##	232	90.48406	89.23190	91.73622
##	233	83.84944	82.52407	85.17481
##	234	84.33990	82.93444	85.74536
##	235	79.76023	79.00411	80.51635
##	236	73.88387	72.99846	74.76928
##	237	81.54950	80.14457	82.95443
##	238	76.93411	75.84156	78.02666
##	239	93.31488	91.10435	95.52541
##	240	72.67397	71.51547	73.83247
##	241	88.81799	87.81270	89.82329
##	242	86.80489	85.73392	87.87586
##	243	82.85278	81.54910	84.15646
##	244	81.93886	81.09569	82.78202
##	245	64.98955	63.29840	66.68071
##	246	83.54372	82.09603	84.99140
##	247	76.67608	75.69917	77.65300
##	248	82.74580	81.60182	83.88979
##	249	72.99194	71.93834	74.04554
##	250	83.86636	82.48596	85.24676
##	251	84.27467	82.98045	85.56889
##	252	63.70279	61.55413	65.85145
##	253	92.86270	91.39385	94.33155
##	254	47.72203	44.22261	51.22144
##	255	69.00722	67.85843	70.15602
##	256	78.83549	77.43877	80.23220
##	257	75.81659	74.80303	76.83016
##	258	78.97870	78.27716	79.68024
##	259	78.84236	77.49893	80.18578