

EPRS 8530 Project 1 – Cover Page

1. Kayla Myers, Early Childhood Education and Mathematics
2. Relationship between Major League Baseball teams' 2014 winning percentage, average home game attendance, and median player salary
3. General interest
4. Descriptive statistics, Correlation, Frequency Tables, Graphs (Histograms & Scatterplots)
5. Microsoft Excel 2013
6. Sources: [mlb.com/standings](http://mlb.com/standings); [espn.go.com/mlb/attendance](http://espn.go.com/mlb/attendance); [usatoday.com](http://usatoday.com)

In this paper, I use measures of descriptive statistics and correlation to analyze the relationships between MLB teams' winning percentages, average home game attendance, and median player salary for the 2014 baseball season.

		<b>MLB Team</b>	<b>Win Percent</b>	<b>Avg Home Attendance</b>	<b>Median Player Salary</b>
Raw Data  $N = 30$		Los Angeles Angels	0.605	38,221	900,000
		Washington Nationals	0.593	31,844	2,850,000
		Baltimore Orioles	0.593	30,805	1,000,000
		Los Angeles Dodgers	0.58	46,695	4,300,000
		St Louis Cardinals	0.556	43,711	767,500
		Detroit Tigers	0.556	36,014	1,775,000
		Kansas City Royals	0.549	24,154	1,500,000
		San Francisco Giants	0.543	41,588	4,000,000
		Pittsburgh Pirates	0.543	30,155	1,200,000
		Oakland Athletics	0.543	25,045	775,000
		Seattle Mariners	0.537	25,485	750,000
		Cleveland Indians	0.525	18,428	975,000
		New York Yankees	0.519	42,520	3,684,426
		Toronto Blue Jays	0.512	29,327	3,000,000
		Milwaukee Brewers	0.506	34,535	850,000
		Atlanta Braves	0.488	29,065	1,142,857
		New York Mets	0.488	26,860	1,000,000
		San Diego Padres	0.475	27,103	1,980,000
		Miami Marlins	0.475	21,386	1,000,000
		Tampa Bay Rays	0.475	17,857	1,475,000
		Cincinnati Reds	0.469	30,576	1,200,000
		Chicago Cubs	0.451	32,742	1,622,857
		Philadelphia Phillies	0.451	29,924	1,587,500
		Chicago White Sox	0.451	20,896	1,500,000
		Boston Red Sox	0.438	36,494	3,100,000
		Minnesota Twins	0.432	27,785	1,000,000
		Houston Astros	0.432	21,627	507,600
		Texas Rangers	0.414	33,564	1,375,000
		Colorado Rockies	0.407	33,090	2,025,000
		Arizona Diamondbacks	0.395	25,601	1,937,500

# More Money, More Problems: An Analysis of MLB Teams' 2014 Winning Percentages, Average Home Game Attendance, and Median Player Salary

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## Description

As a baseball fan, I was interested in the relationship between money, fan support, and team performance. The politics surrounding ideas of salary caps and unfair advantages prompted this inquiry. Is a higher payroll an advantage? Does a higher payroll produce a higher game-day turn out from the fans? Do either of those affect the team's performance? This paper analyzes the relationships between all 30 Major League Baseball Teams' 2014 winning percentages, average attendance for home games in 2014, and median player salary amounts in 2014. Only regular season games were considered.

## Methods

I gathered data from 3 separate online sources. The first source, [mlb.com/standings](http://mlb.com/standings), provided the 2014 regular season standings. From this data, I collected each team's winning percentage ( $N = 30$ ). This number is calculated by dividing the team's total wins by total games in 2014. The second source, [espn.go.com/mlb/attendance](http://espn.go.com/mlb/attendance), provided the 2014 regular season average game attendance for each team. From this data, I collected each team's average home game attendance ( $N = 30$ ). The third source, [usatoday.com/sports/mlb/salaries/2014/team/all](http://usatoday.com/sports/mlb/salaries/2014/team/all), provided salary information for all MLB teams in 2014. From this data, I collected the median player salary amounts in 2014 for each team ( $N = 30$ ). With these three sets of data, I created frequency tables and histograms for attendance and salary, which tell a story by themselves. I used descriptive statistics to analyze each of the three data sets, including mean, standard deviation, median, mode, and range. Then, I used correlation (Pearson's  $r$  and  $r^2$ ) along with scatterplots to illustrate relationships between each of the three variables. Each of these graphs and tables are included in the results. From these correlation measures, I draw conclusions about relationships, which I include in the discussion.

## Results

Table 1  
*Average Home Game Attendance Frequency*  
( $N = 30$ ).

<i>Limits</i>	<i>F</i>
17,000-20,000	2
20,001-23,000	3
23,001-26,000	4
26,001-29,000	3
29,001-32,000	7
32,001-35,000	4
35,001-38,000	2
38,001-41,000	1
41,001-44,000	3
44,001-47,000	1

Table 2  
*Median Player Salary Frequency* ( $N = 30$ ).

<i>Limits</i>	<i>F</i>
500,000-880,000	5
880,001-1,260,000	9
1,260,001-1,640,000	6
1,640,001-2,020,000	3
2,020,001-2,400,000	1
2,400,001-2,780,000	0
2,780,001-3,160,000	3
3,160,001-3,540,000	0
3,540,001-3,920,000	1
3,920,001-4,300,000	2

Table 3  
 Descriptive Statistics (N = 30).

Category	Mean	SD	Median	Mode	Range
Winning Percentage	0.5	0.06	0.497	0.543	0.21
Average Attendance	30,436.57	7,391.06	30,039.50	n/a	28,838
Median Player Salary	1,692,674.67	1,019,935.47	1,425,000	1,000,000	3,792,400

Figure 1  
 Average Home Game Attendance Histogram (N = 30).

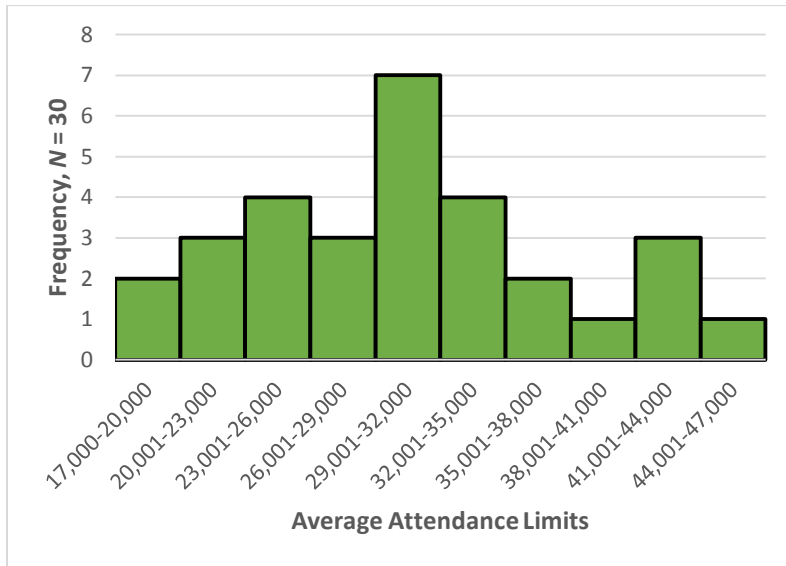


Figure 2  
 Median Player Salary Histogram (N = 30).

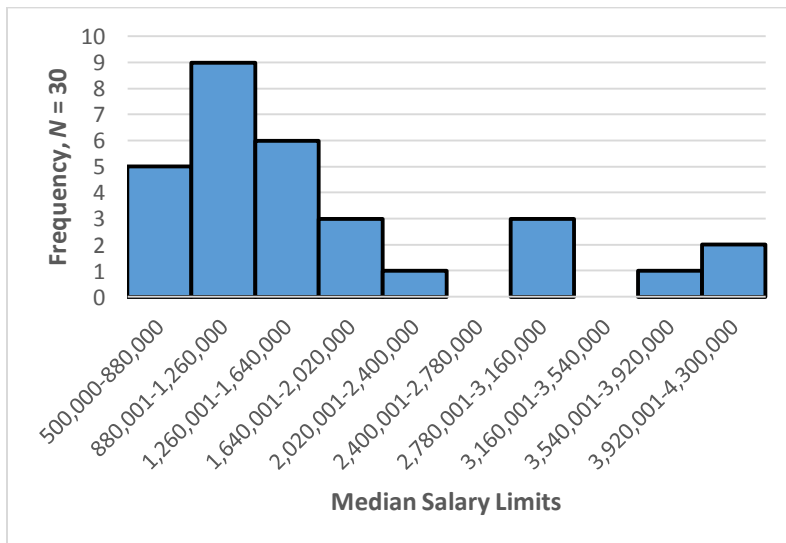


Table 4  
 Correlation Matrix (N = 30).

Category	Winning Percentage	Average Attendance	Median Player Salary
<b>Winning Percentage</b>			
Pearson Correlation	1		
Common Variance	1		
<b>Average Attendance</b>			
Pearson Correlation	0.352	1	
Common Variance	0.124	1	
<b>Median Player Salary</b>			
Pearson Correlation	0.126	0.561	1
Common Variance	0.016	0.315	1

Figure 3

Scatterplot: Winning Percentage and Average Home Game Attendance (N = 30).

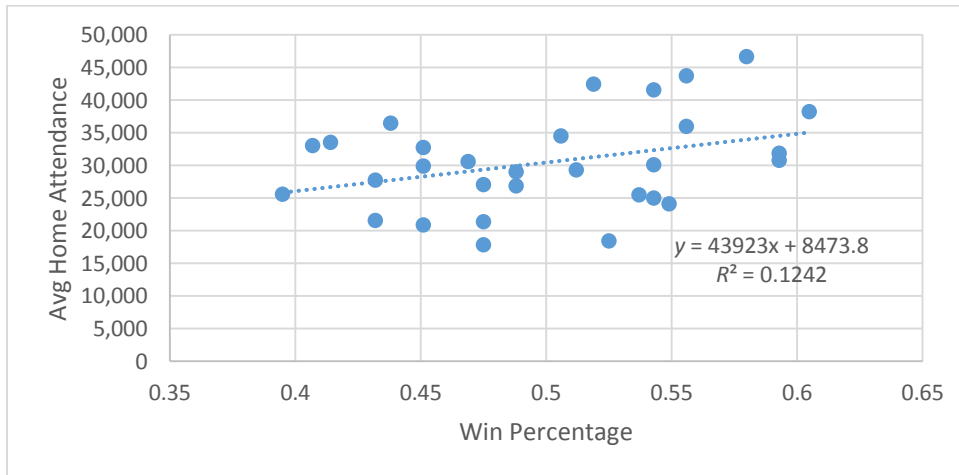


Figure 4

Scatterplot: Winning Percentage and Median Player Salary (N = 30).

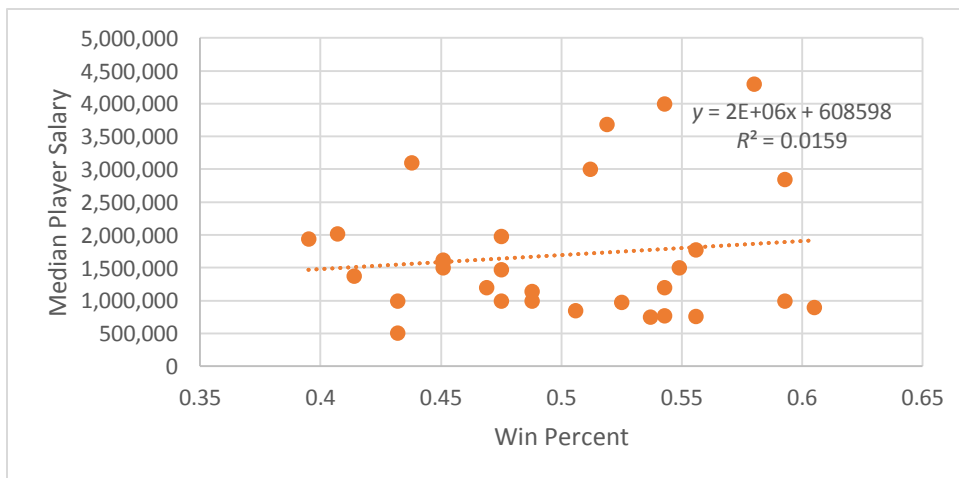
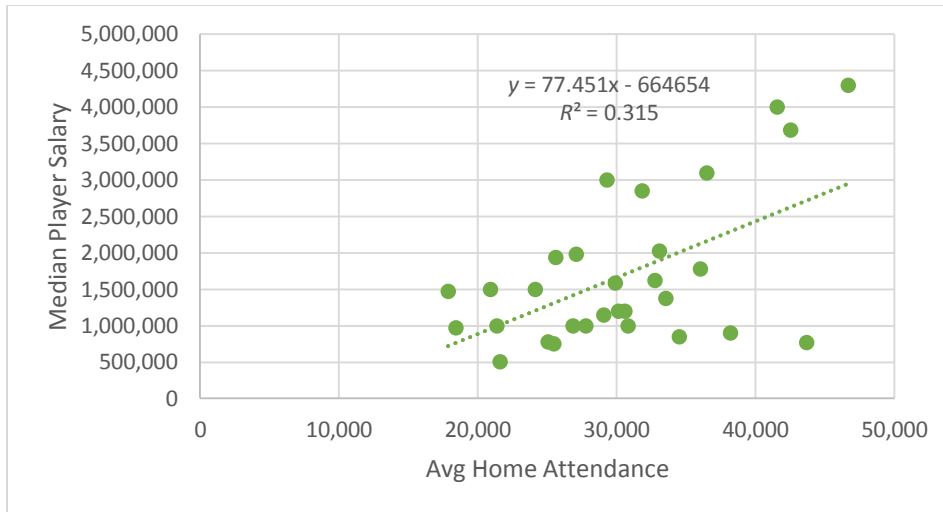


Figure 5

Scatterplot: Average Home Game Attendance and Median Player Salary (N = 30).



## Discussion

The descriptive statistics I found for attendance and salary were surprising. I had no idea just how vast the range was for these statistics. Different ball parks have different seating capacities, but the range of attendance is greater than the range of available seating. The graph for attendance is relatively normal. In the future, I would be interested to look at the percentage for each ball park. I chose to use median instead of mean player salary, expecting that the median would be a truer representation of the data, but even the median produced a positively skewed graph. The range of salaries is more than alarming, it's nauseating.

The correlational statistics were just as surprising. Each of the correlations was positive, but none were as strong as expected. The strongest correlation was between attendance and salary ( $r = 0.561$ ). This does not imply that a higher payroll causes higher ticket sales, nor does it imply that higher ticket sales cause higher payroll, but it does mean there is a relationship. There was a slight correlation between attendance and win percentage ( $r = 0.352$ ). Again, I expected this correlation to be stronger, since a team would expect to win more games if the attendance is higher. This may imply that a certain population of super-fans attend games for their favorite team regardless of win percentage, while others are more likely to go to games when the team is doing well.

The weakest correlation was between salary and win percentage ( $r = 0.126$ )! This does not imply causality (or a lack thereof), but it does imply that there is very little relationship between the team's performance and how much players are paid. This, in my humble opinion, is a big step in the direction of discrediting arguments for implementing salary caps for Major League Baseball Organizations. As these results indicate, a team having more money on payroll does not necessarily mean the team will win more games. This interesting result has implication for future research, possibly considering more variables of player salary or winning records over longer periods of time.

Go Braves!